

**Modeling, Control Analysis, and Multi-Physics Co-Simulation Supporting  
High-Performance Hybrid-Electric Vehicles**

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By

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# Modeling, Control Analysis, and Multi-Physics Co-Simulation Supporting High-Performance Hybrid-Electric Vehicles

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## **DEDICATION**

*Dedicated to my father and mother.*

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## LIST OF ABBREVIATIONS

CAA	Clean Air Act
CCCV	Constant Current, Constant Voltage
CFD	Computational Fluid Dynamics
CG	Center of Gravity
CSE	Co-simulation Engine
CVL	Charging Voltage Limit
DP	Dynamic Programming
DUT	Device under Testing
DVL	Discharge Voltage Limit
ECMS	Equivalent Consumption Minimization Strategy
EES	Electric Energy Storage
EM	Electric Machine
EV	Electric Vehicle
FBD	Free Body Diagram
FE	Fuel Economy
FEM	Finite Element Method
FMI	Functional Mock-up Interface

FMU	Functional Mock-up Unit
HEV	Hybrid Electric Vehicle
ICE	Internal Combustion Engine
MPG	Miles per gallon
MPGGE	Miles per gallon of gasoline equivalent
OCV	Open Circuit Voltage
SOC	State of Charge
SOH	State of Health
UDDS	Urban Dynamometer Driving Schedule
USEPA	United States Environmental Protection Agency
XML	eXtensible Mark-up Language

## LIST OF SYMBOLS

$G$	Gravitational acceleration
$A$	Incline angle
$M$	Vehicle mass
$H$	Height of vehicle CG above the ground
$a, b$	Distance of front and rear axles, respectively, from the normal projection point of vehicle CG onto the common axle plane
$V_x$	Longitudinal vehicle velocity
$V_w$	Headwind speed
$N$	Number of wheels on each axle
$F_{xf}, F_{xr}$	Longitudinal forces on each wheel at the front and rear ground contact points, respectively
$F_{zf}, F_{zr}$	Normal load forces on each wheel at the front and rear ground contact points, respectively
$A$	Effective frontal cross-sectional area
$C_d$	Aerodynamic drag coefficient
$P$	Mass density of air
$F_d$	Aerodynamic drag force
$T$	Traction Torque
$N$	Effective gear ratio
$R$	Tire radius
$v_f$	Final velocity

$T_{EM}$	EM Torque
$\omega_{EM}$	EM Speed
$P_{EM}$	EM power
$I_{Battery}$	Battery Current
$V_{Battery}$	Battery Voltage
$Ah_{consumed}$	Amp-hour energy consumed
$E_{Battery}$	Battery energy use
$E_{Electrical}$	Electrical energy stored in battery
$E_{Fuel}$	Fuel energy used
$E_{Total}$	Total energy used (sum of fuel and battery energy)
$R_{Series}$	Series impedance
$R_{Internal}$	Battery internal resistance
$x_k$	State variable
$u_k$	Control variable
$F_k$	Function defining state variables
$J_{0,\pi}$	Total cost
$g_0$	Cost associated with the initial step
$g_N$	Cost associated with the final step
$\phi_k$	Cost associated with application of constraints
$h_k$	Incremental cost of application of controls
$\dot{m}_{fuel}$	Mass fuel rate
$SOC_{ini}$	Initial state of charge
<i>OCVtablecharge</i>	Look-up table for the OCV vs. SOC during charge



<i>OCVtableDischarge</i>	Look-up table for the OCV vs. SOC during discharge
$Q_{ini}$	Initial transferred charge
$Q_{total}$	Total transferable charge
$C_0$	Capacity at $T_{ref}$ for $Q_{total} = 0$ and $t = 0$
<i>useHeatPort</i>	Boolean variable for using the heat port
$T_{ref}$	Reference temperature
<i>alphaRs</i>	Linear temperature coefficient for $R_s$
$R_{sref}$	Ohmic resistance at reference temperature
$K_{CQ_{abs}}$	Linear $Q_{abs}$ dependency of capacity
<i>alphaC</i>	Linear temperature coefficient for capacity
$Q_{Heat}$	Cell heat generated
$Q_{cooling}$	Heat removed from the cell
$C_{Heat}$	Cell heat capacity

## SUMMARY

This thesis presents a series of model-based studies and associated considerations supporting the development of a high-performance HEV. Due to increasingly strict governmental regulations and consumer demand, automakers have taken steps to reduce fuel consumption and greenhouse emissions. HEV's can provide a balance between fuel economy and vehicle performance by exploiting engine load-point shifting, regenerative braking, pure electric operation, and hybrid traction modes. The existence of a multitude of HEV architectures with different emissions and performance characteristics necessitates the development of simulation platforms which can assist in specifying and selecting critical components.

Recent advancements in the automotive industry, especially the introduction of hybrid technology, have resulted in lower emissions and improved fuel economy; however, hybrid technology can also be utilized in order to enhance the performance characteristics of traditional internal combustion high-performance vehicles. The complexity of the hybrid systems and high power demand of high-performance vehicles requires a detail analysis of critical system components, such as the energy storage systems, to ensure safe and optimal operation. The collaboration between Georgia Tech researchers and Ferrari S.p.A. is illustrative of the need for the further development of innovative and model-based tools to enhance the design and performance of high-performance hybrid electric vehicles.

This thesis also features a series hybrid electric vehicle model developed using Simulink modeling software as part of a tutorial which may be provided to students in order to teach the basic principles underlying the operation, control, and design of hybrid electric vehicles.

The final chapter of this thesis features a modeling approach developed in order to analyze the battery pack in high-performance hybrid-electric vehicles using a multi-physics co-simulation approach. This modeling capability can be extended to other multi-physics systems in order to develop high fidelity models while significantly decreasing computational costs.

## **Chapter 1: Introduction**

Vehicle electrification is being driven both by growing consumer awareness as well as increasing efficiency standards. However both technical and market challenges remain. The development of hybrid electric vehicles requires access to a high skilled workforce as well as the necessary tools in order to reduce the development costs and expedite the evaluation of each design alternative. Virtual simulations can alleviate the product development cost and enable faster introduction of new vehicles. However, many new graduates lack the technical expertise to utilize the new simulation tools available to product development engineers. The second chapter of this thesis describes the development of a tutorial made available to engineering students in order to introduce the sequential process of vehicle development. The third chapter provides a brief description of the application of Bellman's principle of optimality to evaluate the fuel economy of hybrid electric vehicles. Chapter four provides a multi-physics co-simulation to support development of energy storage systems.

### **1.1 Hybrid Electric Vehicle Technology**

By the 1960's, public awareness that improvements in quality of life are offset by increased emissions and detrimental effects on the atmosphere led to the passage of the Clean Air Act (CAA) and the establishment of the Environmental Protection Agency (USEPA). Since then, both government and industry have taken major steps to reduce emissions and improve the efficiency of our energy systems [1].

Since the early days of large scale vehicle manufacturing, which coincided with the development of new technology that provided easy access to gasoline, the majority of vehicles manufactured used an internal combustion power plant. The high energy density of gasoline

makes it a good source to power vehicles, providing an ideal balance between range and power. However, the byproducts of combustion have been proven to have numerous negative consequences.

Advancements in manufacturing and fuel technology has dramatically increased the energy efficiency of internal combustion engines (ICE); however, even the most efficient engines manufactured today are limited by the Carnot efficiency of the engine. In addition, energy wasted due to friction of moving parts, cooling of engine block and incomplete combustion in cold temperatures reduce the overall efficiency of the vehicle.

One of the major advancements by the transportation industry has been the development of hybrid technology. With the addition of a secondary power plant to the vehicle, the ICE can be used more efficiently. During periods of high power demand, the engine management system can send power toward the secondary power source, which can significantly increase the efficiency of the system. During periods of low power demand, the ICE can operate at higher and more efficient operating points and the extra power is directed to energy storage devices such batteries or capacitors [1].

Hybrid electric vehicles on the market today feature superior fuel efficiency compared to their traditional competitors. While the hybrid technology can be used to dramatically increase the fuel economy and decrease the emissions, Ferrari S.p.A. has shown that the same technology can be used to dramatically increase vehicle performance. Unlike ICEs, electric machines (EM) provide maximum torque at low speeds, which can significantly increase vehicle performance [2-4].

With the addition of a secondary power plant, engineers need to carefully consider the layout of the vehicle powertrain. Each individual layout of the powertrain is referred to as the vehicle architecture. Traditionally HEVs have been classified into two categories: series and parallel. Figure 1.1 illustrates the difference between different hybrid architectures.

The series HEV is the simplest architecture, featuring an ICE powering a generator which charges a power storage device. The current from the generator, as well as the battery if needed, is then used to power the traction motor that drives the wheels. In this configuration, the engine does not provide torque directly to the wheels, rather it only produces electricity [1].

The advantage of the series architecture is in its simplicity as the engine can be designed to run at its most efficient operating point as the secondary on-board generator, and otherwise the vehicle is just an electric vehicle which simplifies the vehicle control strategy. Also a smaller engine can be used, as the role of the engine is to provide a steady source of power rather than provide a wide range of torques and speeds. The downside to this architecture, however, is in the number of energy conversions, each adding an associated efficiency factor. It should also be noted that as the traction motor is the only source of power to the wheels, it must be large enough to power the vehicle in a range of driving scenarios [1].

A parallel HEV architecture takes a different approach to traction by providing parallel energy paths. The ICE and any number of motors are connected directly to the wheels, and this availability of multiple sources of power enables the selection of any combination or all for different driving situations [1,5]. For example, the engine can be used for steady state highway speeds where its operating characteristics will provide effective combustion, while the motor can be used during stop-and-go urban driving where its regenerative braking capabilities will allow for much of the energy spent during acceleration to be recouped while braking [5]. At times

when the driver demands high acceleration, all sources can be used to propel the vehicle, and if the battery state of charge gets too low, the motor will only provide regenerative braking during stopping events or even act against the engine, thus acting as a generator and forcing the engine into a more efficient operating point.

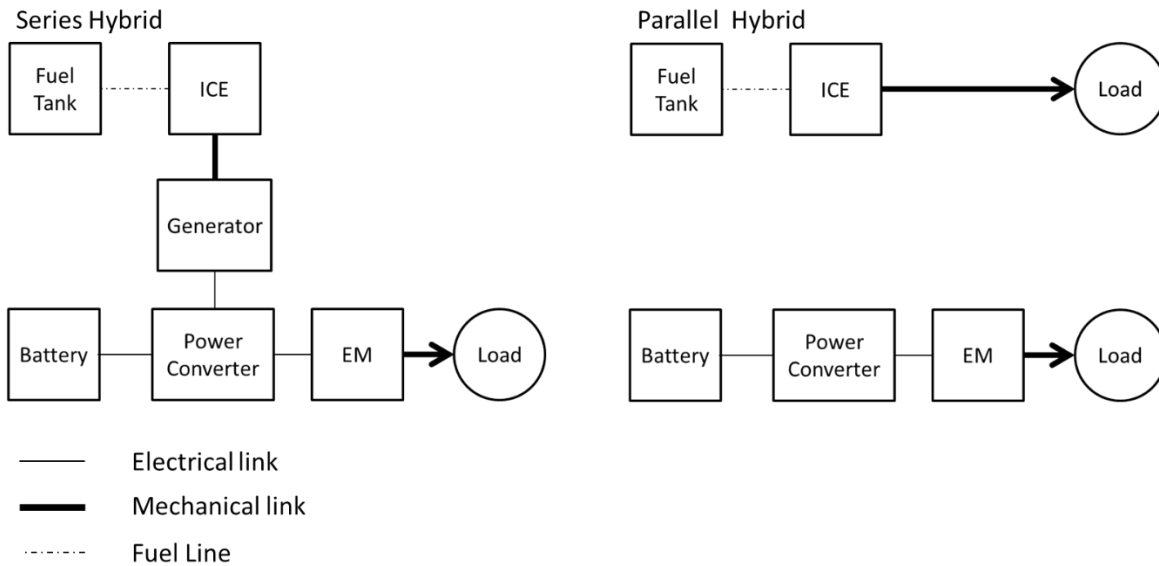


Figure 1.1: Classification of hybrid electric vehicles

Beyond these control options, another benefit of the parallel architecture is in component sizing [1,5]. As all power sources are available to provide power to the wheels, they can all be used for high torque demands, and therefore components can be downsized. The use of gears, shafts, and belts allow for this architecture to be laid out in numerous different configurations.

Due to the added complexity of HEVs and the growing demand for skilled talent, the automotive industry has invested heavily in collaborative research with universities. Examples of such work include the collaborative research between Georgia Tech and Ferrari S.p.A. One of the main challenges for students recruited into the field of automotive research is the knowledge

gap between theoretical models for individual components and the overall system operation of vehicles. Providing students with an overall systems preview of vehicle powertrain has proven to significantly increase the performance of student teams. Chapter 2 of this thesis is dedicated to work done on developing a tutorial designed to familiarize students with system level operation of a vehicle. The tutorial features a sequential refinement approach to model development in order to introduce students to complex model components.

## **1.2 Evaluation of HEV Performance using Global Optimization Algorithms**

The global optimization of a HEV architecture can be realized for a known time horizon and operating conditions, such as in the case of optimizing for a drive cycle. DP is used to find the optimal operation regime for a specific drive cycle, and it serves as a benchmark for other non-optimal control strategies [4,6-8]. Using DP multi-objective optimization is possible, as an example, for optimizing fuel economy (FE) and CO<sub>2</sub> emissions simultaneously [7]. However, as DP requires the future knowledge of the drive cycle conditions and large computational resources, vehicle online implementation has not materialized [4].

DP is applied using a backward-looking approach where the simulation requires knowledge of the complete drive cycle and simulation starts at the final simulation time and progresses using a backward step size [2,4]. At each time step the incremental cost of applying control is evaluated and penalty terms are applied to increase the cost of applying control. The optimal control strategy of the power cycle is determined by minimizing the overall cost [4,9]. Chapter 3 of this thesis provides details about DP and approach used to implement in HEV.

Providing benchmarking on performance and emissions results for vehicle architectures in the product development phase provides engineers with an unbiased metric to evaluate and



recognize the differences between different HEV architectures. Ferrari S.p.A. used the benchmarking results to identify final vehicle architectures.

### **1.3 Model-based Design and Ferrari Partnership**

In recent years, the development of automotive technology has changed from an exclusively electrical or mechanical engineering discipline to include software engineering. This shift has dramatically changed product development as well as methods, tools and engineering skills required. Due to the complexity of design in the development of hybrid electric vehicles, the need to use model-based design practices to evaluate the performance of each design alternative has increased dramatically.

Ferrari S.p.A. and Georgia Tech have formed a research partnership with a focus on the development of model-based analysis of HEVs. In the first stage of the partnership, focus was given to benchmarking the performance and fuel economy of each powertrain architecture. Fuel economy and CO<sub>2</sub> emissions were evaluated for each design alternative, using DP. Chapter 3 of this thesis features an overview of DP and work done in the first phase of the partnership.

After the completion of the first stage, focus has shifted toward providing model-based analysis of battery performance. One of the unique challenges when designing high-performance HEVs is the design of the battery pack and providing an accurate model of battery performance and aging under high load conditions. Due to high power demand, providing sufficient cooling is crucial to ensure safe operation and preventing damage to battery cells. The ultimate goal of this phase of the research partnership is to develop the capability to analyze the performance of different battery pack layouts and incorporate this capability in a complete vehicle model. Chapter 4 of this thesis is dedicated to work done on developing the battery model.

## Chapter 2: Vehicle Model

The vehicle model presented in in this chapter was developed as part of a tutorial which may be provided to students in order to:

- 1) teach the basic principles underlying the operation, control, and design of hybrid electric vehicles,
- 2) educate students to model, formulate, and simulate the operation of hybrid vehicles, and
- 3) provide an overview of the modeling capabilities of Simulink software.

The vehicle model developed next uses a sequential refinement methodology, to create a complex system model using simple components and basic control strategies. The tutorial starts by developing a simple, but complete, vehicle model and then makes further component refinements in order to provide a more realistic model.

The vehicle model is developed using Matlab and Simulink 2016a software. Simulink is a graphical programming environment within which models are created by creating links between pre-defined function blocks and user-defined subcomponents. The predefined function blocks can be categorized under two categories: utility and functional. Utility blocks can be used in a wide array of applications such as applying simple and complex mathematical operations or perform logical operations. Functional blocks however, are designed for specific applications such as calculating the vehicle longitudinal dynamics, evaluating speed, torque and energy loss of a compound planetary gear, or estimating the slip and braking force in a tire. Examples of utility and functional blocks are presented in Figures 2.1 and 2.2.

The vehicle powertrain design is based on a series hybrid architecture. A diesel powered ICE is mechanically connected to a generator (EM-A) with a 1-to-1 gear ratio that converts engine power to electricity which can be stored in the battery or sent directly to EM-B. EM-B then sends power to the wheels. The architecture is designed to operate within five different power modes. Figure 2.3 provides a visual representation of the power modes and Table 2.1 describes the operating conditions as implement by the controller unit.

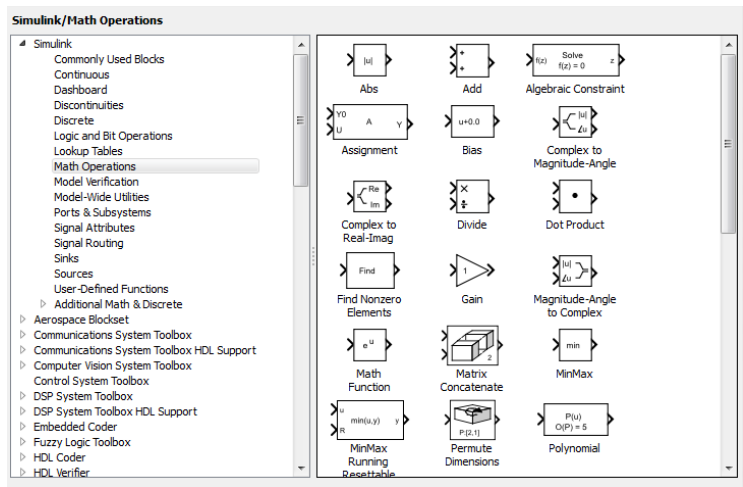


Figure 2.1: Simulink utility blocks

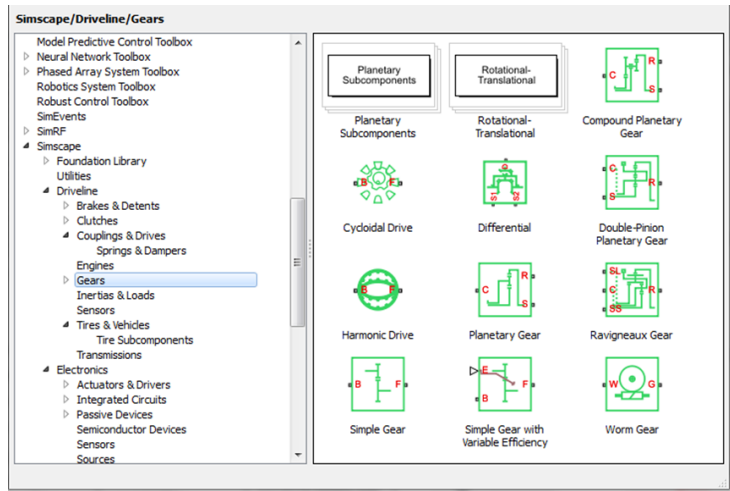


Figure 2.2: Simulink functional blocks

Table 2.1: Driving modes

Power Mode	Battery state of charge	Traffic conditions
Pure EV	High	Low power demand, Stop-and-go traffic
Pure ICE	Low	Constant high speed travel
Combined	Mid rang or high	High power demand, High performance setting
ICE – Charging	Low	Constant speed travel
Charging	As needed	Braking, coast down, hill descent

Based on the battery state of charge (SOC) and power demand, the vehicle controller selects the appropriate power mode. If the battery SOC is above a pre-determined value and the power demand can be supplied by the EM-B alone, the vehicle may operate in pure EV mode. Reducing the emissions to zero while allowing the EM-B to use energy stored in the battery. Under conditions that battery SOC is low, the controller will select the pure ICE driving mode to prevent damage to the battery. If the power demand is low, the controller may go into ICE-Charging mode and request additional power output from the ICE and store the extra energy in the battery. Under conditions when the SOC is within a pre-determined range, the controller may select the combined power model and reduce the fuel consumption by optimizing the power split between the ICE and EM-B. During periods of deceleration the controller may select the charging mode and allow EM-B to regenerate power and store the kinetic energy of the vehicle in the battery pack.

At the highest level, the model consists of three subsystems: Driver, Plant, and Controller. The Driver subsystem receives the speedometer and tachometer data and transmits torque request signals to the controller. The Plant subsystem includes system component models including ICE,

EM-A and EM-B. The controller subsystem uses a state-flow controller which receives torque request data from the Driver subsystem and based on pre-determined state conditions sends appropriate control signals to the system components such as the ICE, EMs, and transmission.

The vehicle model can be used to simulate a pre-determined drive cycle and calculate the vehicle performance parameters such as fuel consumption, speed, acceleration, and battery state of charge. The Simulink control model can also be uploaded to a vehicle control unit and used for testing purposes in Hardware-in-the-loop applications.

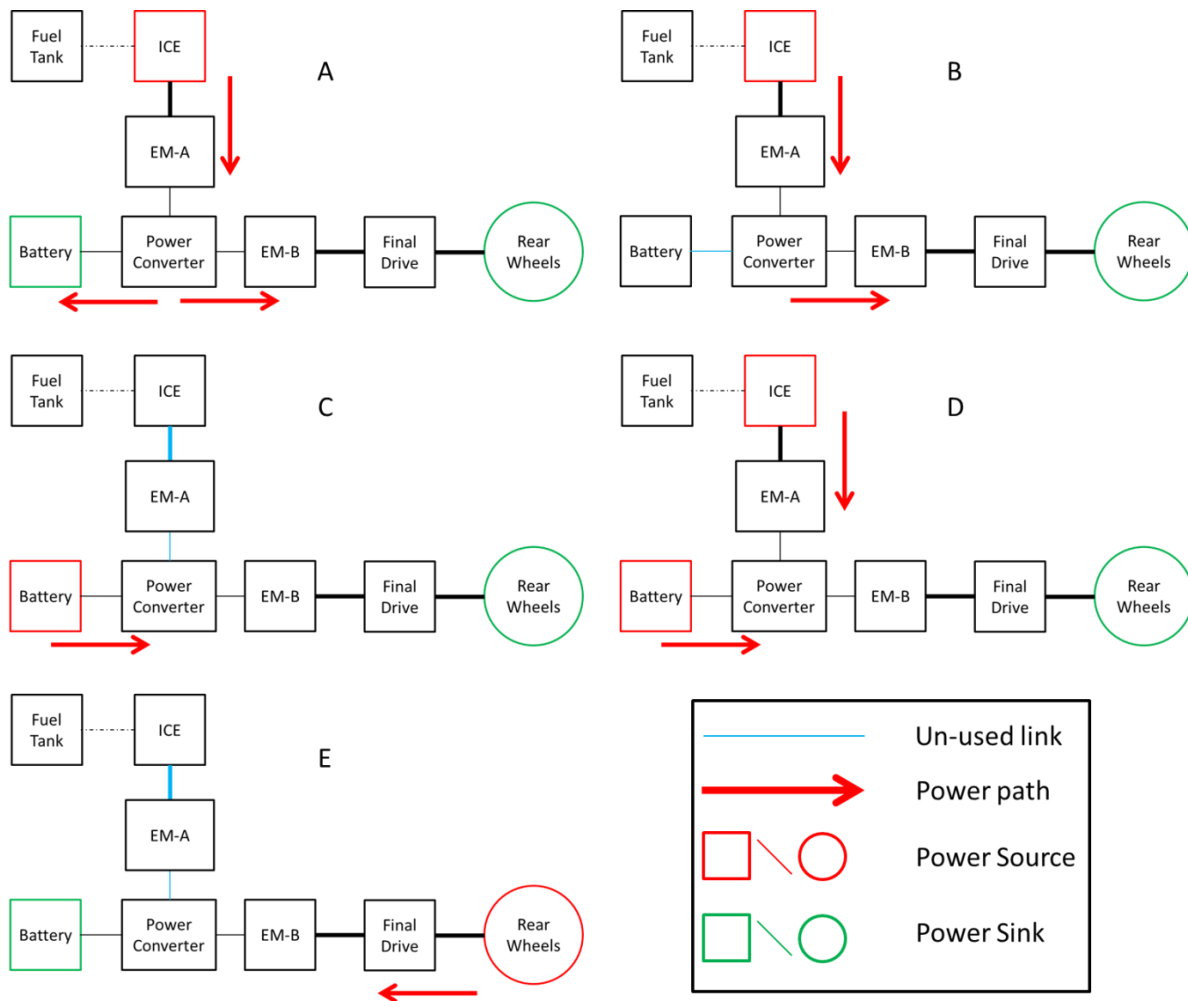


Figure 2.3: Hybrid vehicle power paths. A:ICE-Charging, B:Pure ICE, C:Pure EV, D:Combined, E:Charging

## 2.1 Simple Vehicle Model (EV-mode only)

The first step in the development of vehicle model is a simple vehicle subsystem with ideal components. The vehicle is propelled only by an ideal EM and a single gear pair is used to transmit power to the wheels. The main components of the simple vehicle subsystem are developed using the following functional Simulink blocks: Vehicle body, Tire (Magic formula), and Simple Gear blocks.

A vehicle body Simulink block is used to model a vehicle with two axles in longitudinal motion. The vehicle wheels are assumed to be identical in size. The vehicle is assumed to be in normal equilibrium and does pitch or have any vertical movement. The vehicle axles are parallel and form a plane. The longitudinal  $x$  direction lies in this plane and is perpendicular to the axels. The model also includes traveling on an incline slope  $\alpha$ , the normal  $z$  direction is not parallel to gravity but is always perpendicular to the axel longitudinal plane.

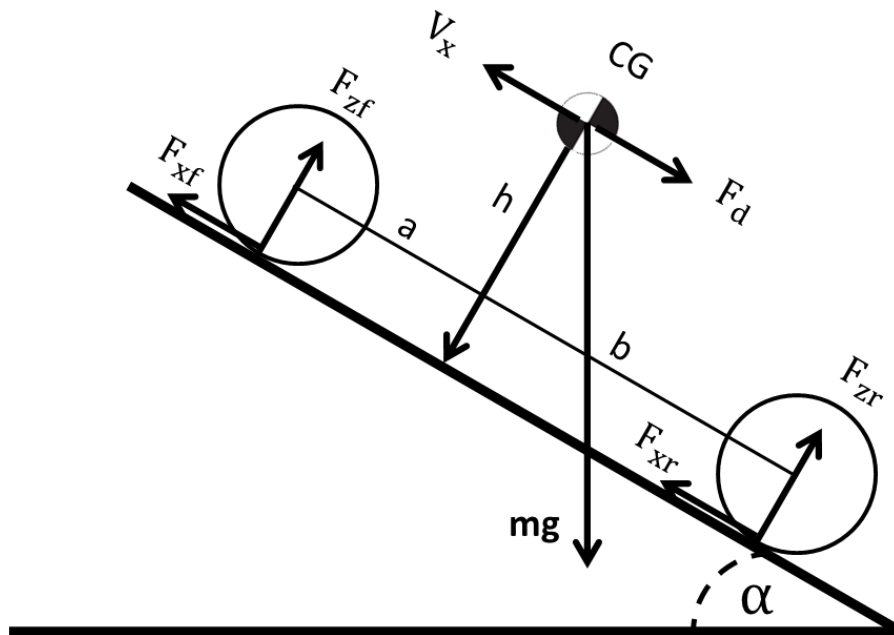


Figure 2.4 Vehicle free body diagram (FBD)

The vehicle motion is determined by the net effect of all forces and torques acting on it. The longitudinal tire forces push the vehicle forward or backward. The weight,  $mg$ , of the vehicle acts through its center of gravity (CG). Depending on the incline angle, the weight pulls the vehicle to the ground and pulls it either backward or forward. Whether the vehicle travels forward or backward, aerodynamic drag force acts in the opposite direction of travel. It is assumed that the drag force acts through the CG. Figure 2.4 and Table 2.2 define the vehicle motion model variables.

Table 2.2: Vehicle model variables

Symbol	Description and Unit
G	Gravitational acceleration = $9.81 \text{ [m/s}^2\text{]}$
A	Incline angle, [deg]
M	Vehicle mass, [kg]
H	Height of vehicle CG above the ground, [m]
a, b	Distance of front and rear axles, respectively, from the normal projection point of vehicle CG onto the common axle plane, [m]
$V_x$	Longitudinal vehicle velocity, [ $\text{m/s}$ ]
$V_w$	Headwind speed, [ $\text{m/s}$ ]
n	Number of wheels on each axle
$F_{xf}, F_{xr}$	Longitudinal forces on each wheel at the front and rear ground contact points, respectively, [N]
$F_{zf}, F_{zr}$	Normal load forces on each wheel at the front and rear ground contact points, respectively [N]
A	Effective frontal cross-sectional area [ $\text{m}^2$ ]
$C_d$	Aerodynamic drag coefficient
$\rho$	Mass density of air = $1.18, \text{ [kg/m}^3\text{]}$
$F_d$	Aerodynamic drag force, [N]

The vehicle dynamics are described as follows:

$$m\dot{V}_x = F_x - F_d - mgsin\alpha \quad (1)$$

$$F_x = n(F_{xr} + F_{xf}) \quad (2)$$

$$F_d = \frac{1}{2}C_d\rho A(V_x - V_w)^2 \cdot \text{sgn}(V_x - V_w) \quad (3)$$

In order to determine the normal force on each front and rear wheel, the normal force acting on the front and rear wheels are expressed by:

$$F_{zf} = \frac{-h(F_d + mgsin\alpha + m\dot{V}_x) + b \cdot mgcos\alpha}{n(a + b)} \quad (4)$$

$$F_{zr} = \frac{h(F_d + mgsin\alpha + m\dot{V}_x) + a \cdot mgcos\alpha}{n(a + b)} \quad (5)$$

while the wheel normal forces satisfy  $F_{zf} + F_{zr} = mgcos\alpha/n$ .

The Tire (Magic Formula) block is used to model a tire with longitudinal behavior given by the Magic formula<sup>1</sup>. The block can calculate the tire slip, however do to the nature of drive cycles used for analysis, it is assumed that tire slip is negligible. The effects of tire inertia, stiffness, and damping are ignored. The vehicle longitudinal velocity is calculate by  $V_x = r_w\omega$ . Where  $r_w$ , is the tire effective radius in meters and  $\omega$  is the wheel angular velocity in  $rad/s$ .

A Simple Gear block represents a gearbox that constrains two connected driveline axles to co-rotate with a fixed ratio. The output shaft rotates in the same direction as the input shaft. It is assumed that energy loss in the gearbox due to meshing and viscous losses is negligible for the purposes of this analysis. In practice, based on the type of gears used, the quality of the gear meshing, and lubrication, the efficiency of the gear may be as low as 0.9. Gear inertia is

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<sup>1</sup> Series of tire design models developed by Hans B. Pacejka. The term ‘magic formula’ refers to lack of physical basis for the models. Magic formula models are widely used in the industry in a variety of tire constructions and operating conditions.



assumed to be negligible compared to the engine and shaft inertia; therefore it is not included in the model. This assumption is only valid if the gearbox dynamics are not relevant to the study.

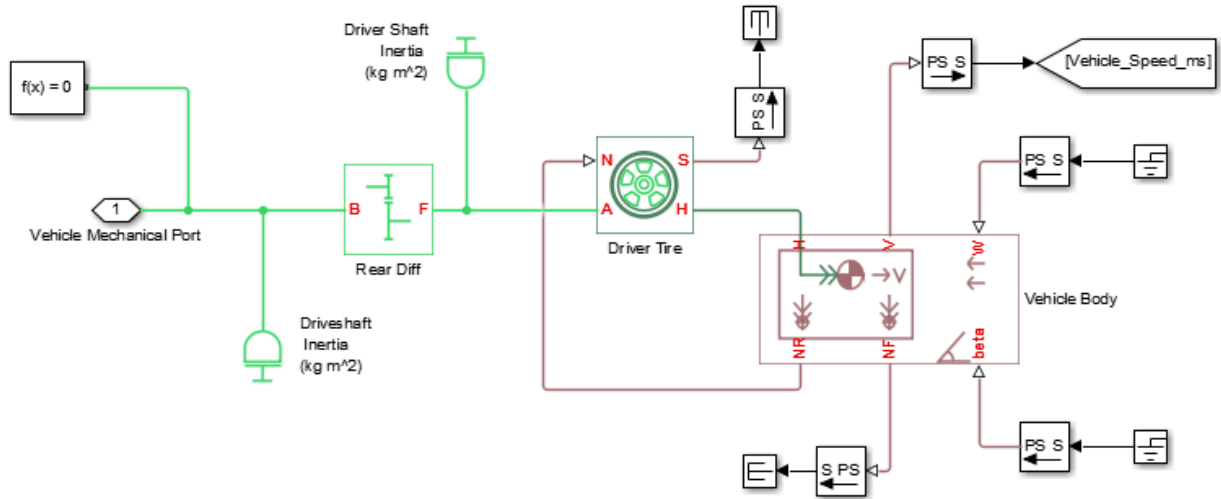


Figure 2.5: Simple vehicle model, Simulink diagram including functional blocks (vehicle body, Tire magic formula and simple gear)

Figure 2.5 illustrates the main components of the simple vehicle model developed using functional Simulink blocks and Table 2.3 includes the model parameters.

Table 2.3: Vehicle model parameters

System parameter	Value
Vehicle Mass	1200 kg
Tire Radius	0.3 m
Driveshaft inertia	$0.0015 \frac{\text{kg}}{\text{m}^3}$
Driver shaft inertia	$0.003 \frac{\text{kg}}{\text{m}^3}$
Gear Ratio	2

In order to test the validity of the model at this initial step, a constant torque of 200 N.m is supplied from EM-B for duration of 10 seconds. The velocity of the vehicle after 10 seconds can be calculated as follows:

$$\frac{dv_x}{dt} = \frac{1}{m} \left( \frac{TN}{r} \right) \quad (6)$$

$$\int_{v_i}^{v_f} dv_x = \int_{t_i}^{t_f} \left( \frac{TN}{mr} \right) dt \quad (7)$$

$$v_f = \frac{(200 \text{ Nm})(2)}{(1200 \text{ kg})(0.3 \text{ m})} = 11.1 \text{ m/s} = 24.9 \text{ mph} \quad (8)$$

Figure 2.6 illustrates the vehicle speed profile as described in Equations 6-8. The simple vehicle model does not capture effects such as tire slip, therefore the application of a constant torque results in a linear increase in speed with constant acceleration.

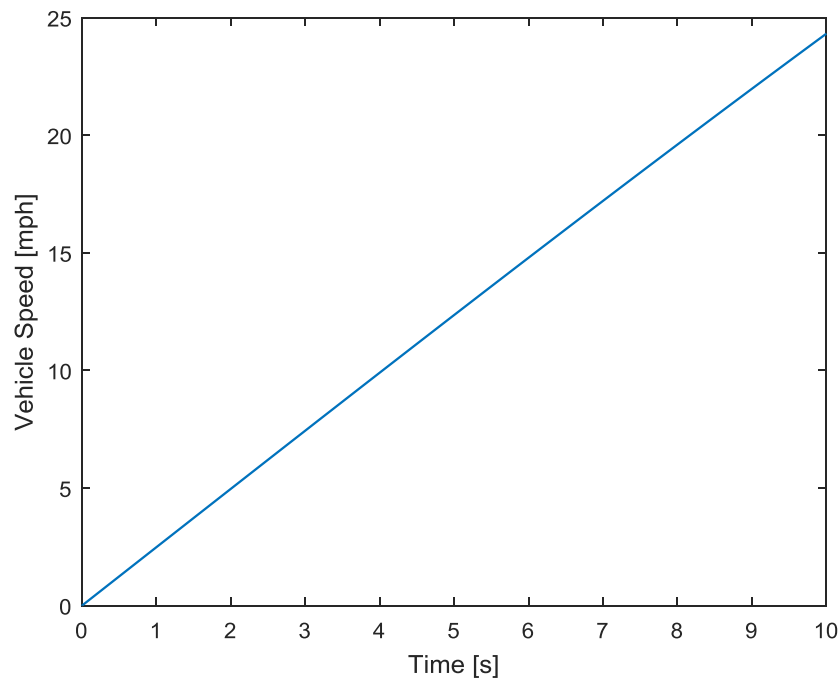


Figure 2.6: Vehicle speed profile

The Driver subsystem functions as a proportional feedback loop and sends a normalized torque signal to the controller. The proportional controller will send a maximum torque signal if the speed error is larger than 2 mph. Figure 2.7 illustrates the design of the proportional feedback loop which allows the vehicle to follow a drive cycle.

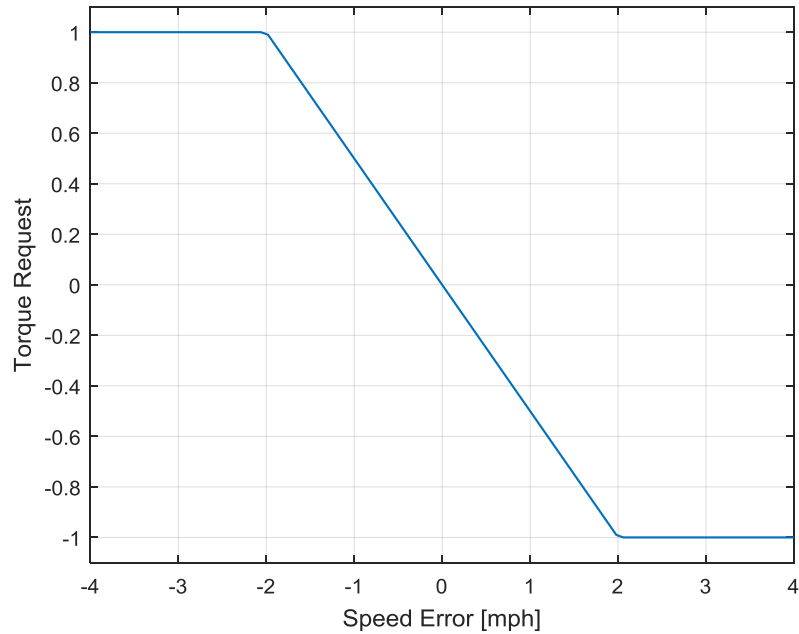


Figure 2.7: Normalized torque request

Using the first 500 seconds of the EPA Urban Dynamometer Driving Schedule (UDDS), it can be shown that the torque request signal from the proportional feedback controller will allow the vehicle to follow the drive cycle closely. Figure 2.8 illustrates the vehicle speed profile as well as the torque request as it follows the UDDS.

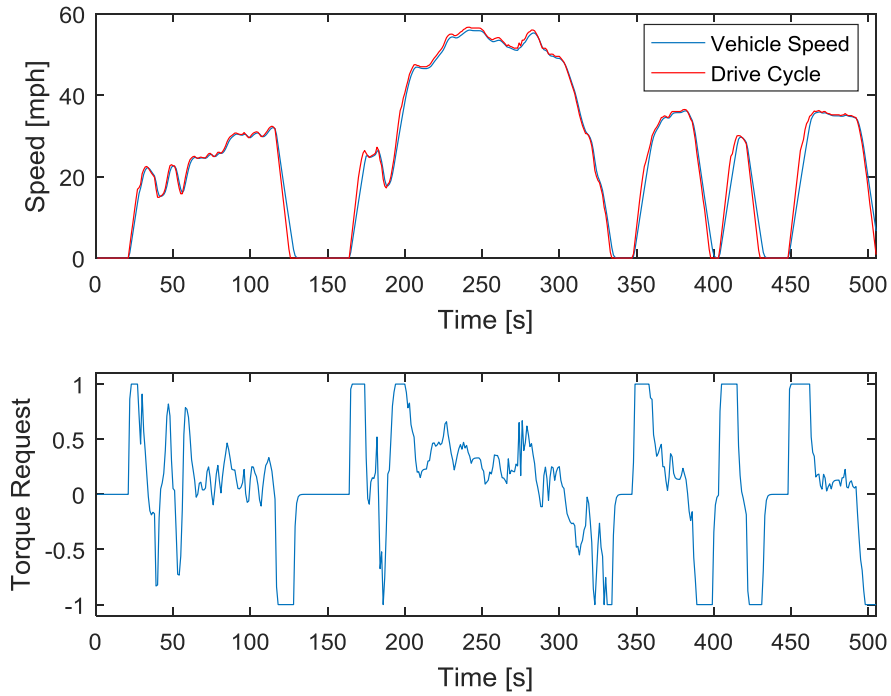


Figure 2.8: Vehicle speed profile, UDDS drive cycle

## 2.2 Simple battery and EM model

Following the development of the simple vehicle model, battery and EM models are developed using utility blocks as shown in Figure 2.9. The simple EM model features a constant maximum torque, no rpm limits, and is modulated by the controller. The battery model features a constant voltage power source with unlimited current and capacity.

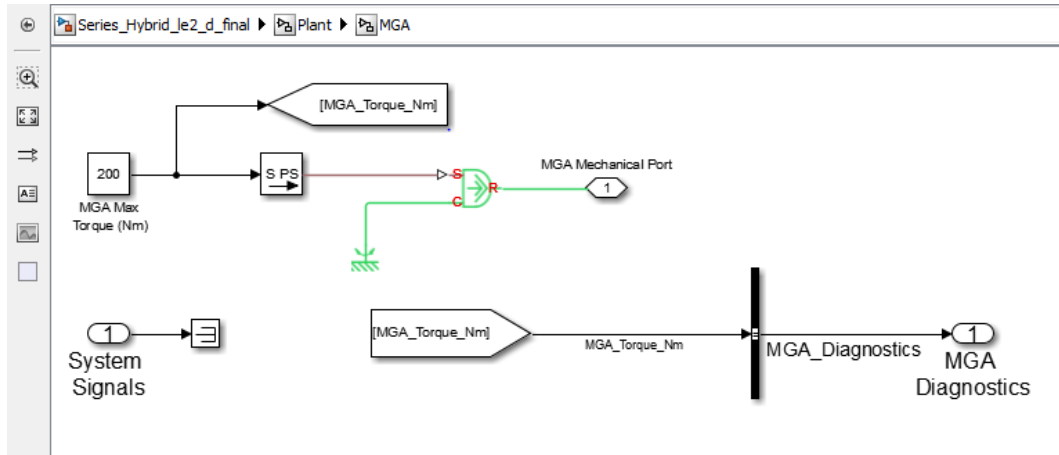


Figure 2.9: Simple EM model, utility blocks used in EM subcomponent

The power output of the EM is calculated by evaluating the speed and torque of the EM. The torque is directly proportional to the torque request from the driver subsystem as follows:

$$T_{EM} = T_{req} \text{Max}(T_{EM}) \quad (9)$$

$$\omega_{EM} = \frac{V_x}{r} N \quad (10)$$

$$P_{EM} = T_{EM} \omega_{EM} \quad (11)$$

The power generated by the EM, is directly proportional to the battery current draw and voltage as described by:

$$I_{\text{Battery}} = -\frac{P_{EM}}{V_{\text{Battery}}} \quad (12)$$

and the battery state of charge (SOC) is calculated using:

$$\text{SOC} = \frac{1}{3600} \int_{t_0}^t I_{\text{Battery}} dt / \text{Capacity} + \text{SOC}_{\text{Initial}} \quad (13)$$

In this calculation, the battery capacity is in the units of Amp-hour. Equations 12 and 13 were implemented using Simulink utility blocks as shown in Figure 2.10.

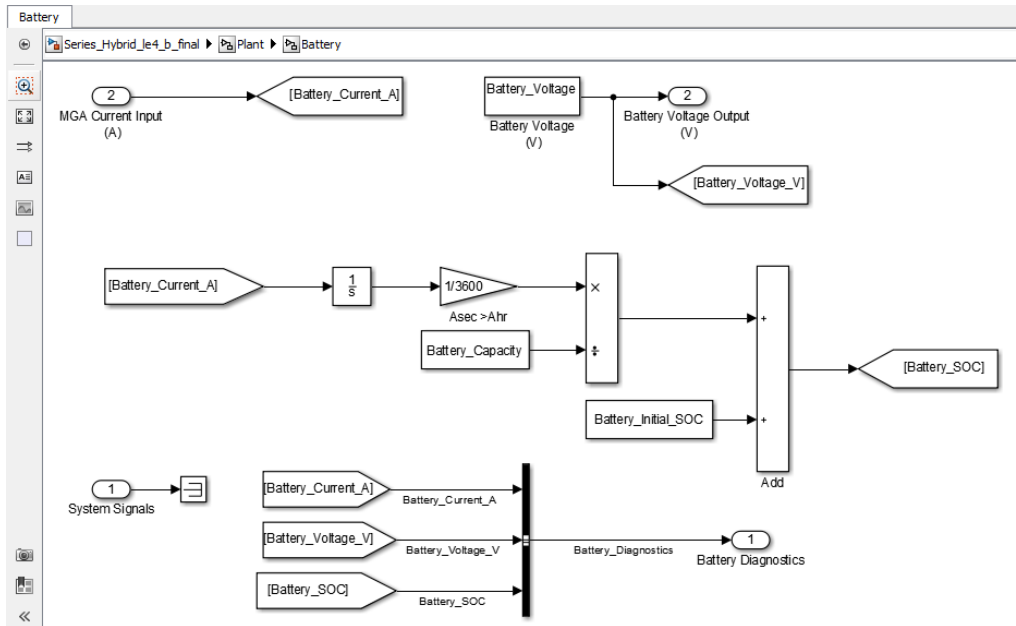


Figure 2.10: Simple battery model, Simulink diagram

As shown in Figure 2.11, during periods of acceleration, negative battery current results in lowering the SOC, whereas during deceleration, positive battery current results in an increase in the SOC. Most traditional HEVs are designed to maintain the battery SOC within an operable range. Continuous deep discharge and charge can damage the battery by lowering the battery life. Allowing the battery to achieve a high or low SOC can also prevent the vehicle controller from charging the battery when the SOC is high, or discharging the battery when the SOC is low.

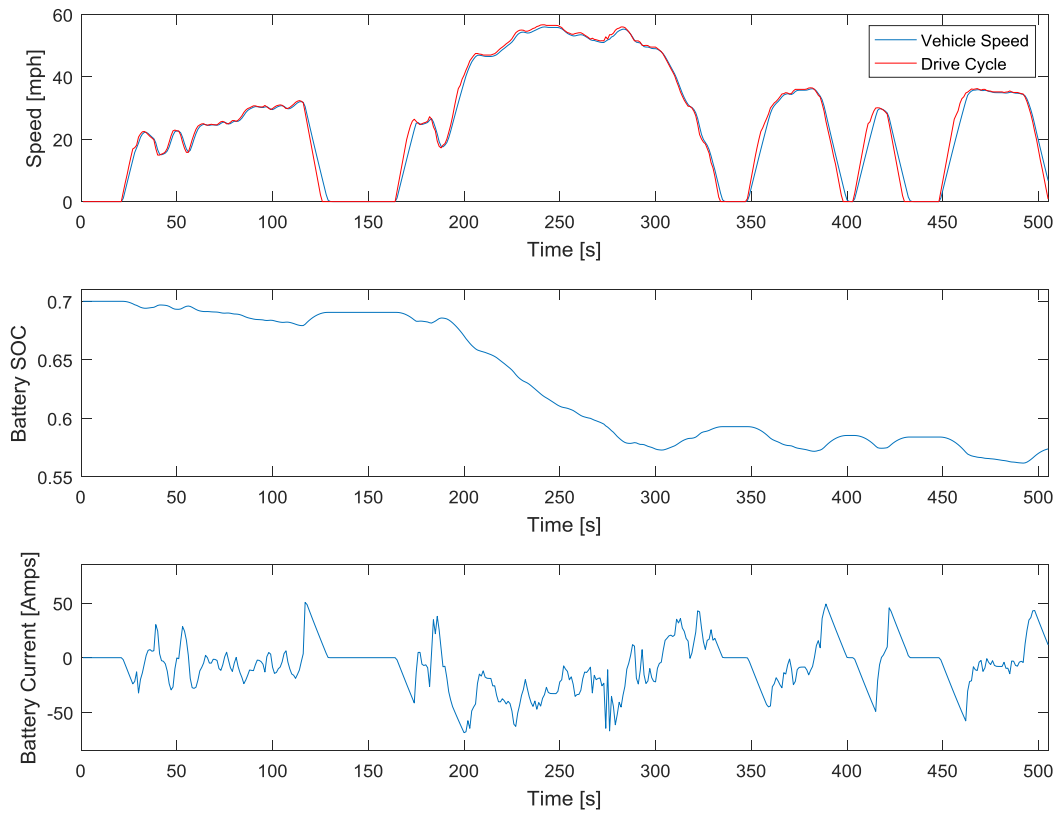


Figure 2.11: Vehicle performance with application of simple battery and EM model

### 2.3 Simple Charging Logic

AS shown in Figure 2.11, the vehicle SOC decreased which can lead to permanent damage to the battery and decrease the range. A secondary EM, EM-A which is identical to EM-B, is added to the model in order to charge the battery. EM-A is powered by an on-board generation unit in the form of a diesel engine.

A rule based controller unit is added to the controller subsystem to turn EM-A on and off based on the SOC. Applying this control strategy, referred to as charge sustaining, prevents the vehicle from depleting the battery. The effects of applying the charge sustaining strategy are

shown in Figure 2.12. Controller prevents the SOC to drop below 60%, allowing EM-A to recharge the battery. Similarly the controller EM-A is turned off as the SOC reaches 70% to prevent overcharging.

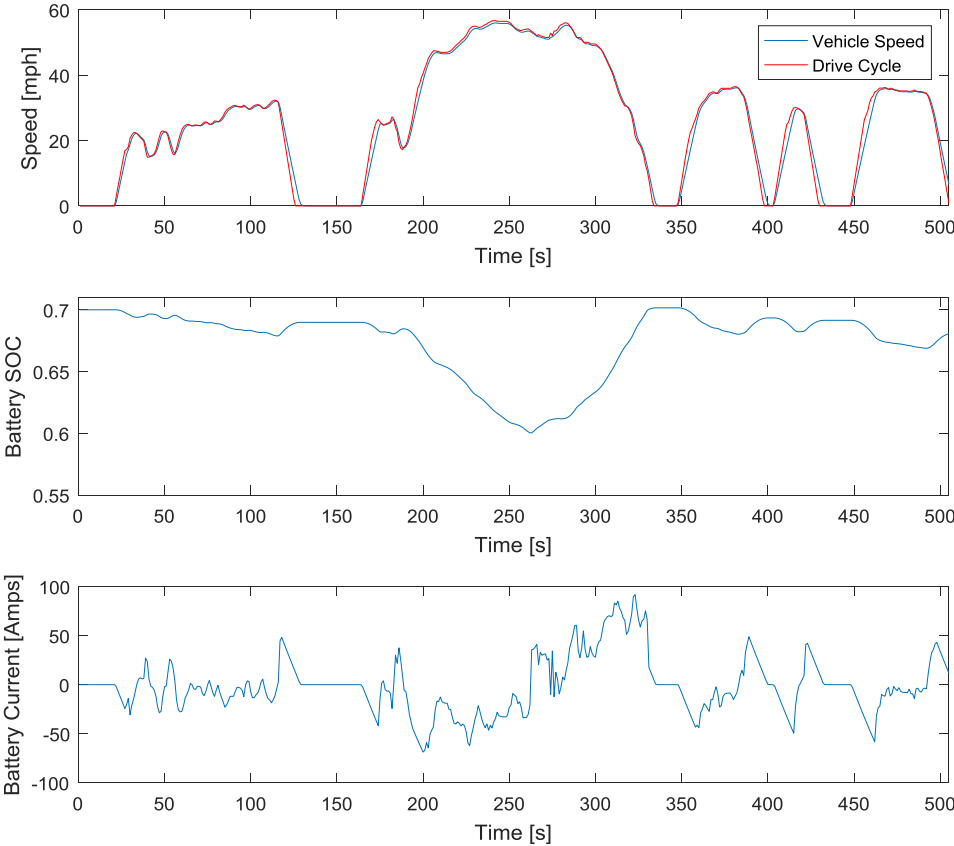


Figure 2.12: Vehicle performance with application of d simple charging logic

The Driver subsystem sends a normalized torque demand to EM-B, however the controller unit only sends an on and off signal to EM-A. In order to prevent damage to the EM, a feedback controller is implemented which sends a normalized torque request to EM-A. In order to power EM-A, a diesel engine is added to the Plant subsystem.



### 2.4 Improved Engine Model and Fuel Efficiency

In order to increase the engine model fidelity, an experimental torque curve for the engine is added to the engine model using look-up tables. The look-up table uses linear interpolation and extrapolation methods to evaluate the engine torque output based on the engine speed and throttle. An engine throttle request is provided from the controller unit. Figure 2.13 illustrates the experimental torque data included in the model.

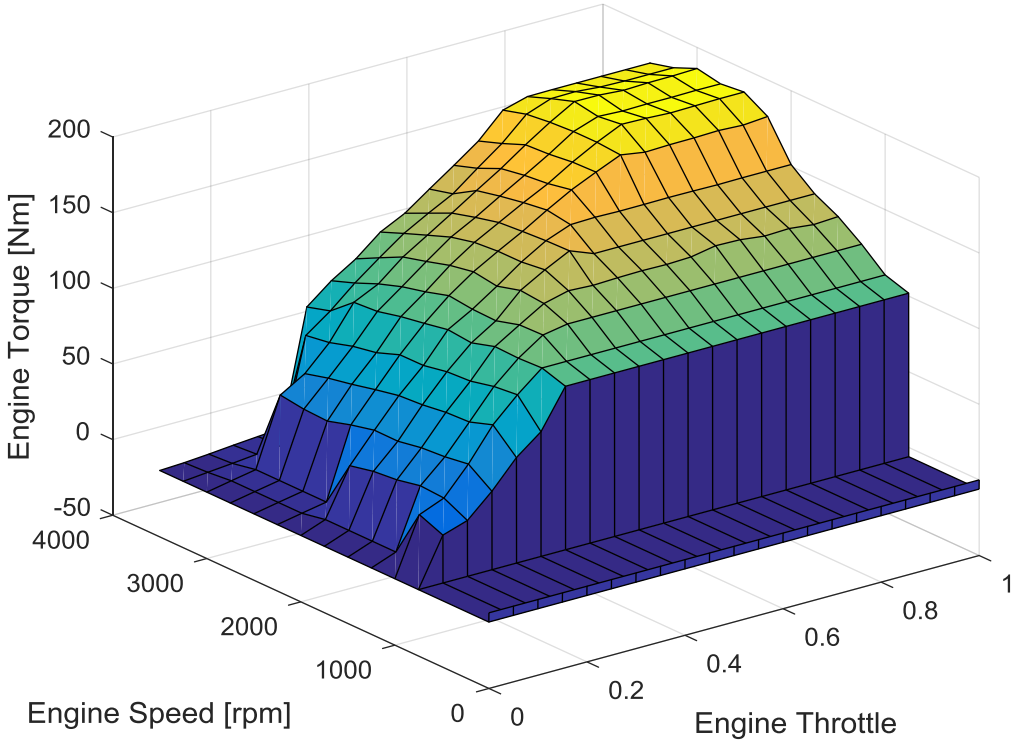


Figure 2.13: Engine Torque map

Similar to experimental engine torque data, fuel consumption data is added to the engine model using a look-up table. The look-up table uses linear interpolation and extrapolation methods to evaluate the engine fuel consumption based on the engine speed and throttle. Figure

2.14 presents the experimental fuel consumption data and Figure 2.15 illustrates the application of look-up blocks to incorporate the experimental data.

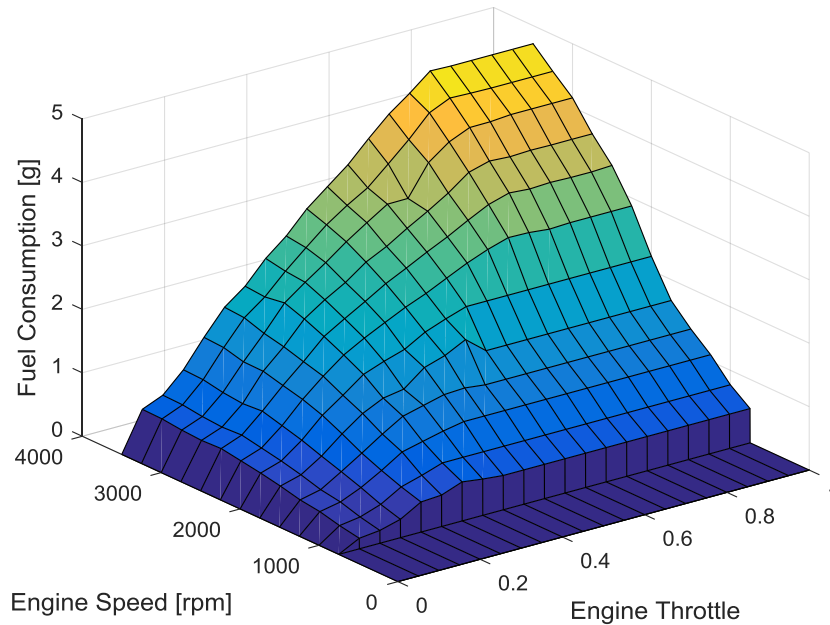


Figure 2.14: Engine fuel consumption map

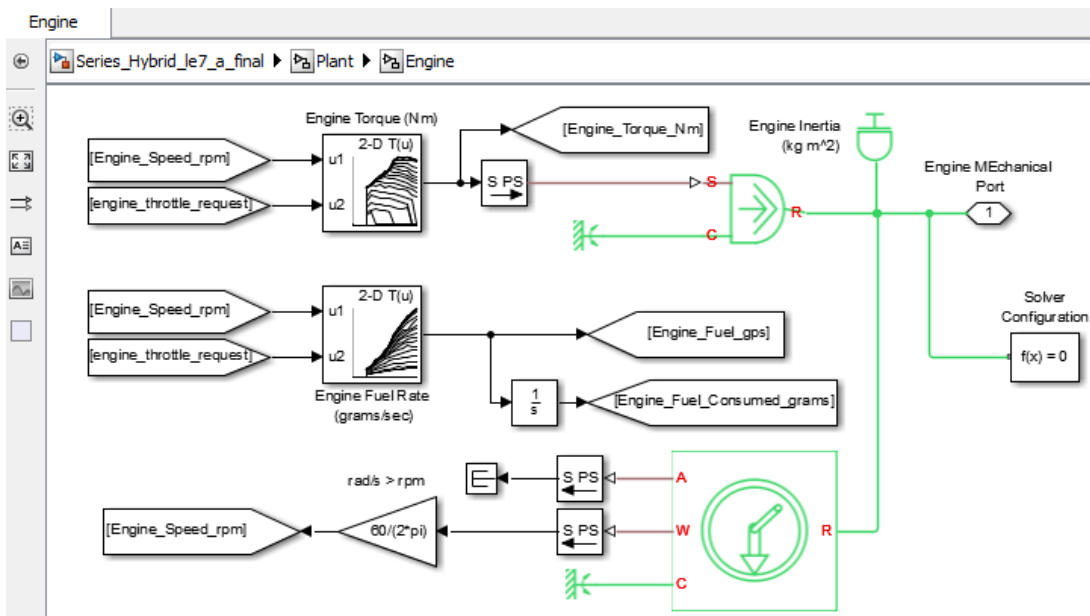


Figure 2.15: Improved engine model

The main advantage of using a series architecture in an HEV is the lack of direct mechanical connection between the engine and wheels which allows the engine to operate at a desired speed, independent from the wheels. The vehicle controller can operate the engine at the most efficient operating points, and reduce fuel consumption.

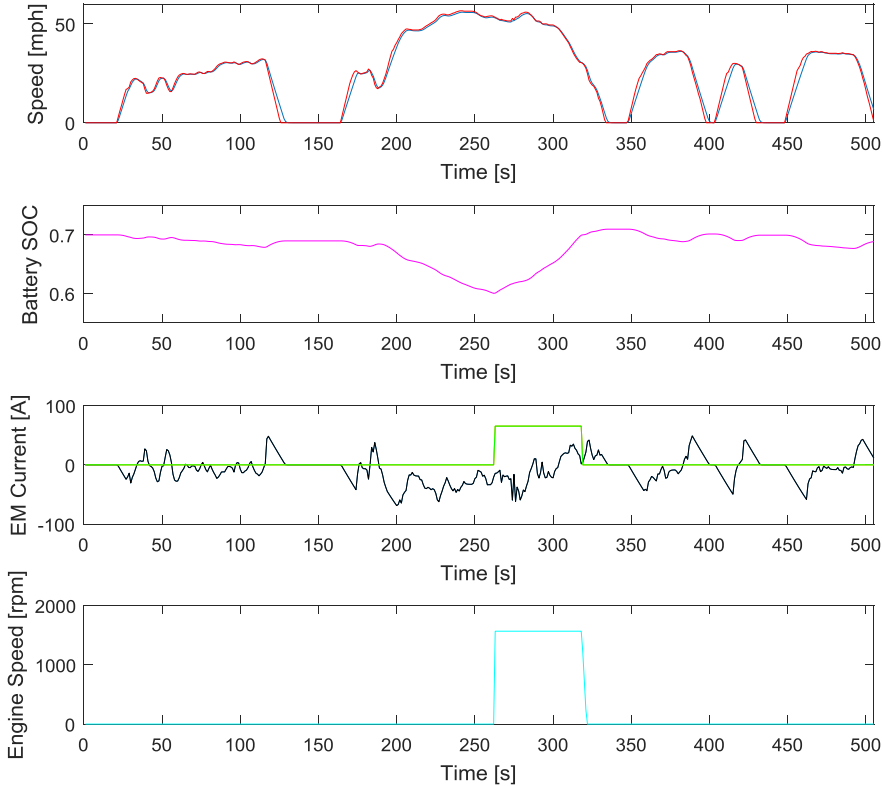


Figure 2.16: Vehicle performance, Engine charging enabled

Based on the engine fuel map and Brake specific fuel consumption (BSFC) map, the engine efficiency is maximized at a speed of 1800 rpm and engine throttle<sup>2</sup> of 0.5. The controller will operate the engine, exclusively at this operating point unless the battery SOC drops below a pre-defined threshold, at which point the engine throttle request increases to provide more torque

<sup>2</sup> Engine throttle is equal to the normalized torque request from the controller unit.

and to increase the SOC. Figure 2.16 presents the vehicle performance after the application of experimental engine data and Figure 2.17 illustrates the BSFC map.

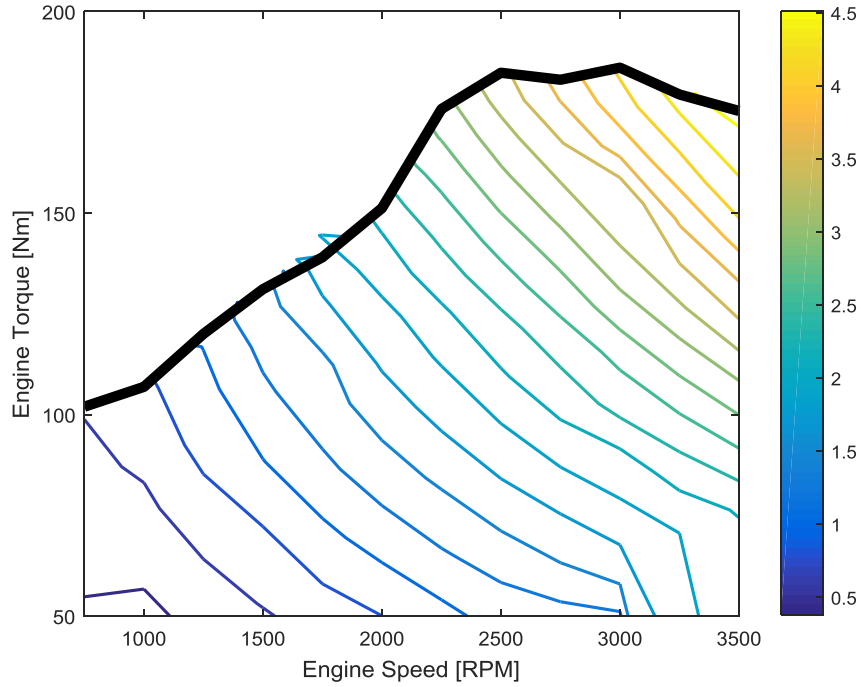


Figure 2.17: BSFC map

Fuel economy is traditionally reported by dividing the distance traveled by the volume of fuel consumed. However, in the case of HEVs, if the final battery SOC is lower than the initial SOC, fuel required to replenish the battery to the initial SOC should be considered in fuel economy calculations. Similarly if the final SOC is higher than the initial SOC, extra fuel consumed to charge the battery above the initial SOC should be subtracted.

First the electrical energy to change the battery SOC should be calculated as follows:

$$Ah_{\text{consumed}}[\text{Amp. hour}] = (\text{SOC}_i - \text{SOC}_f) \times \text{Capacity} \quad (14)$$

$$E_{\text{Battery}}[\text{W. hour}] = Ah_{\text{consumed}} \times \text{OCV}[\text{V}] \quad (15)$$

and the efficiency of the system in converting fuel energy into electrical energy stored in the battery pack should also be considered by:

$$E_{\text{Electrical}}[\text{BTU}] = \frac{3.412^3 \times E_{\text{Battery}}}{\eta_{\text{conversion}}} \quad (16)$$

and the gas equivalent fuel economy with SOC correction can be calculated as described by:

$$E_{\text{Fuel}}[\text{BTU}] = \text{Fuel}_{\text{consumed}} \times 133393.1^4 \quad (17)$$

$$E_{\text{Total}}[\text{BTU}] = E_{\text{Fuel}}[\text{BTU}] + E_{\text{Electrical}}[\text{BTU}] \quad (18)$$

$$\text{Fuel}_{\text{consumed-RFG}} = \frac{E_{\text{Total}}[\text{BTU}]}{114871.74^5} \quad (19)$$

$$\text{MPGGE} = \frac{\text{Distance Traveled}}{\text{Fuel}_{\text{consumed-RFG}}}. \quad (20)$$

The fuel economy reported for the vehicle model (MPGGE) is 103 mpg. The calculations are unrealistic as the inefficiencies of regenerative braking, the battery, and the EMs are not included in the model.

## 2.5 Improved Battery Model

Thus far the battery model featured: 1) a constant voltage equal to the open circuit voltage (OCV), 2) no energy conversion losses, 3) infinite current draw. In order to increase the model fidelity, a linear battery model as shown in Figure 2.18, is included where the battery voltage is a function of current.

---

<sup>3</sup> The 3.412 converts W.hour to BTU

<sup>4</sup> The 1.33393.1  $[BTU/gal]$  is the heating value of diesel fuel

<sup>5</sup> The 114871.74  $[BTU/gal]$  is the heating value of RFG (Reformulated gasoline)

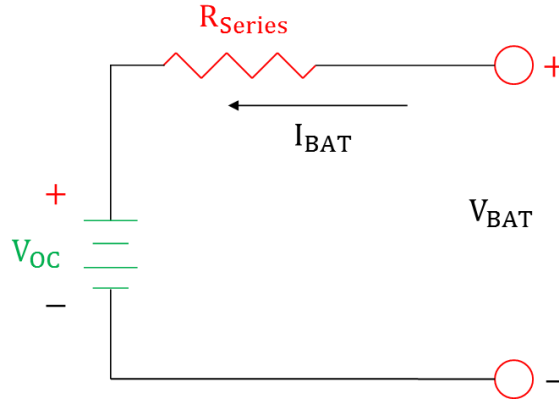


Figure 2.18: Improved battery model

$$V_{\text{BAT}} = V_{\text{OC}} + I_{\text{BAT}} \times R_{\text{Series}} \quad (21)$$

$$R_{\text{Series}} = R_{\text{Internal}} + R_o \quad (22)$$

Series resistance of the battery is the sum of battery internal resistance and the resistance of other electrical components such as the wiring. The battery OCV is a function of temperature and SOC. Experimental data, as shown in Figure 2.19, is used to include the temperature effects on the battery performance.

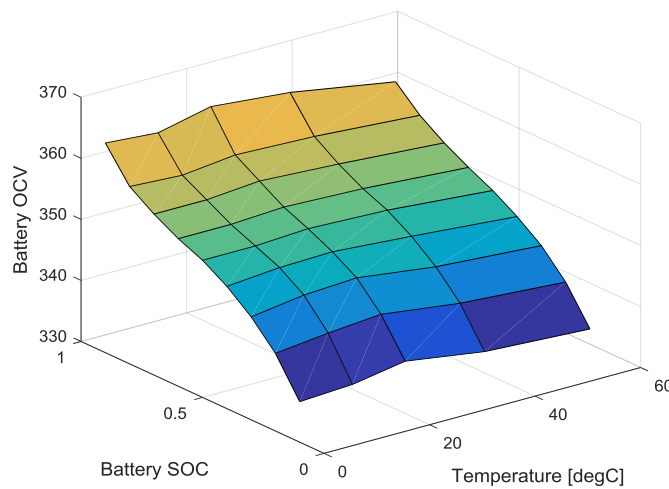


Figure 2.19: Battery Open Circuit Voltage experimental data

Experiments have also shown that the battery internal resistance is different during charge and discharge. Internal resistance is also a function of SOC and temperature. Figures 2.20 and 2.21 illustrate the internal resistance experimental data included in the model.

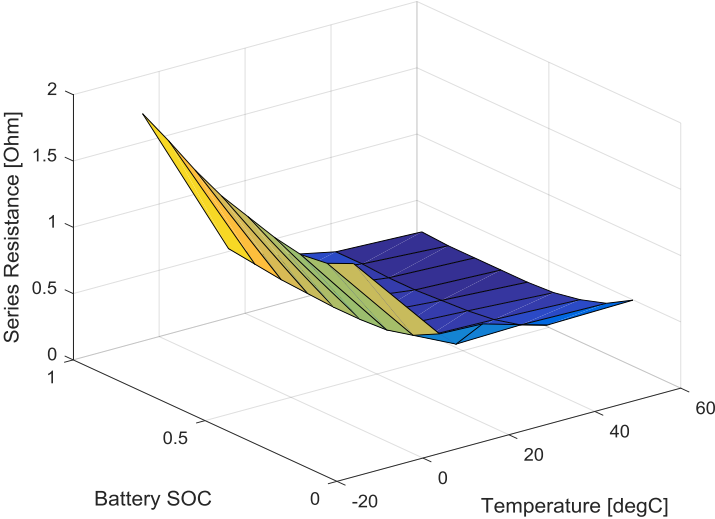


Figure 2.20: Total series resistance-Discharge

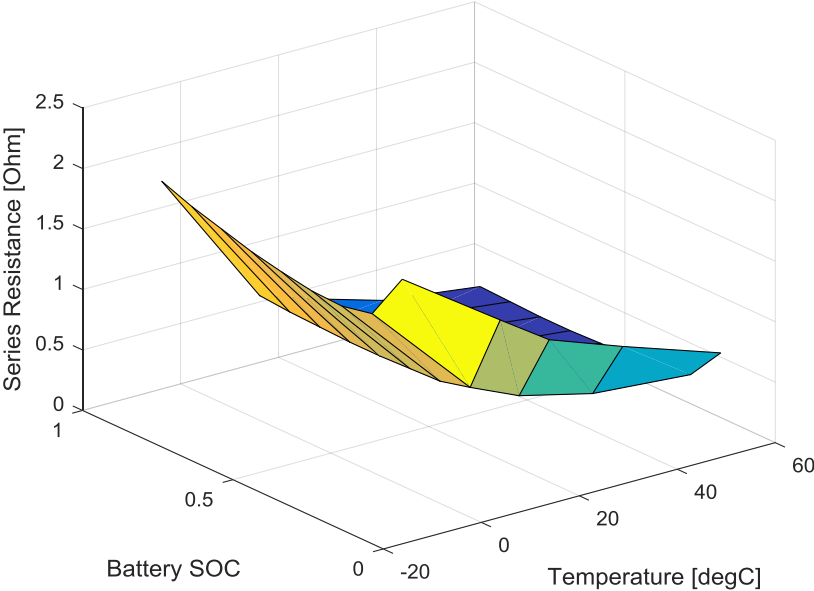


Figure 2.21: Total series resistance-Charge

The experimental data is added using a look-up table. Look-up table uses linear interpolation and extrapolation methods to evaluate the output based on the battery SOC and temperature. The improved battery model lowers the fuel efficiency of the vehicle model (MPGGE with SOC correction) to 90 mpg.

### 2.6 Improved EM Model

Thus far the EM model featured: 1) ideal conversion efficiency and, 2) constant maximum torque output. In order to improve the fidelity of the EM model, experimental torque-speed data is used which relates the maximum EM torque to the shaft speed as shown in Figure 2.22.

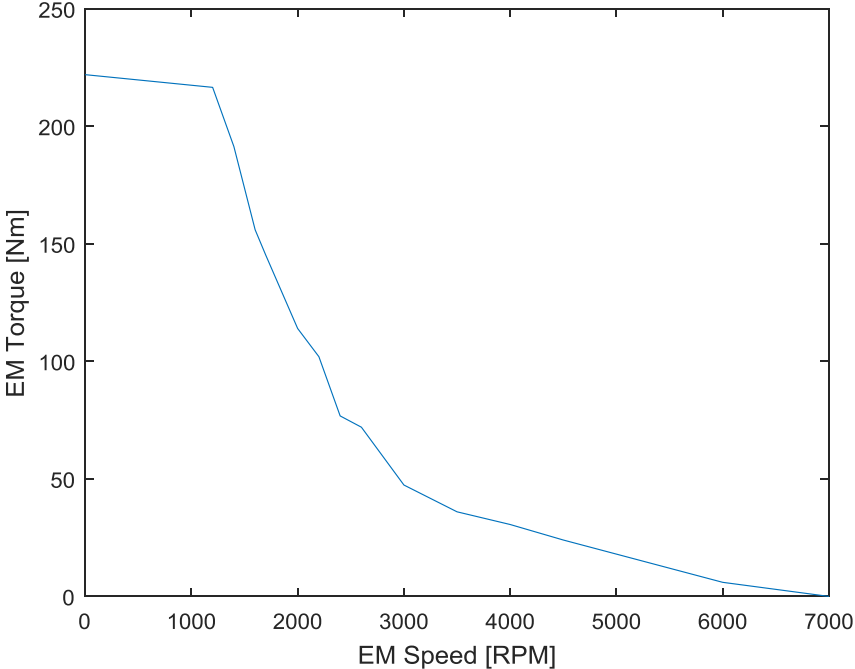


Figure 2.22: Electric machine torque curve



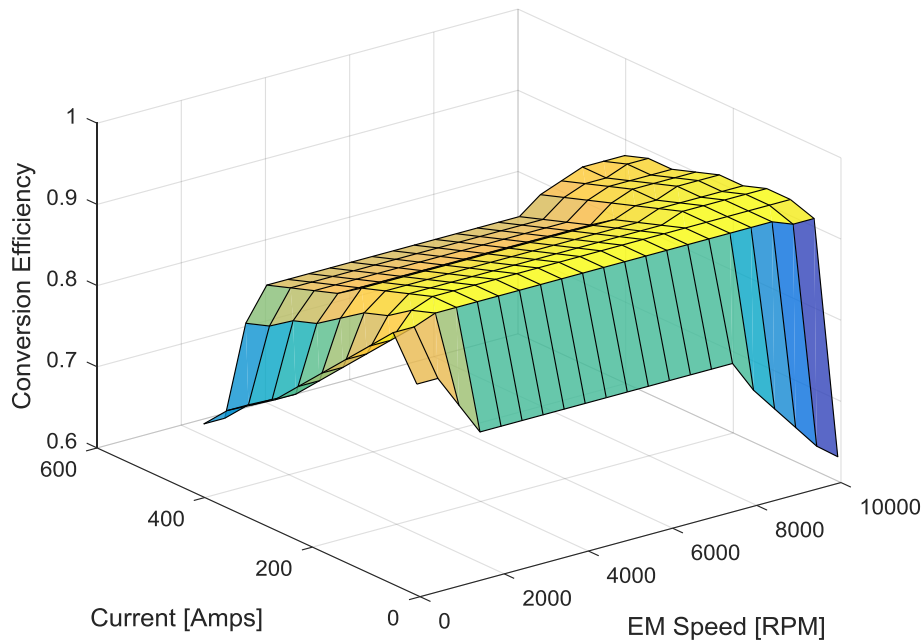


Figure 2.23: EM conversion efficiency map

Electromechanical energy conversion can go two ways: 1) electrical to mechanical and, 2) mechanical to electrical. Experimental results have shown that the conversion efficiency of the EM is dependent on shaft speed and current. Figure 2.23 illustrates the EM conversion efficiency data included in the model.

The addition of experimental data to the motor and generator model, increases the model fidelity. The fuel economy of the model decreases to 65.4 mpg.

## 2.7 Conclusion

Using the sequential refinement modeling method described in this chapter, students and engineers will be able to develop suitably complex vehicle models and develop the vehicle control strategy to operate the vehicle. In addition, the model can also be used in Hardware-in-

the-loop tests where the model can be uploaded into the control unit and used for testing the vehicle operation in the next steps of vehicle development.

The hybrid vehicle model developed in this chapter is a series design which features a simple control strategy. However, the development of more complex control strategies, for example parallel and power-split architectures, can be accomplished using the instructions provided in this thesis and the supplementary tutorial in the Appendix.

In addition to the improvements to the vehicle model in Sections 2.4-2.6, further additions and improvements to the model can improve the model fidelity and ensure better operation in Hardware-in-the-loop scenarios. Simple additional improvements to the model can include the addition of current limits, engine emergency defueling at high engine speeds and limiting the auxiliary motor speed during engine startup.

### **Chapter 3: Evaluation of Vehicle Performance using Dynamic Programming**

The introduction of HEVs into the fleet of performance vehicles manufactured by Ferrari S.p.A. requires a new design approach which ensures that new vehicles meet improved fuel economy and emissions standards while maximizing the utility and performance characteristics.

In order to evaluate each powertrain under consideration, a simulation-based design approach can ensure an unbiased comparison. DP can provide an optimally controlled performance metric for each architecture under consideration over pre-determined drive cycles. Therefore it evaluates each architecture using a consistent control strategy and thus avoids the uncertainty associated with different supervisory control approaches [4,9]. The DP algorithm has been developed and used in many applications in order to determine the optimal control strategy.

DP provides a globally optimal control path for a predetermined drive cycle using Bellman's Principle of Optimality. Since DP requires the future knowledge of the operation conditions and heavy computational loads, vehicle online implementation is not possible [9]. Nevertheless, attempts have been made to extract heuristic rules from dynamic programming results. Vehicle control units use alternative non-optimal control regimes such as Equivalent Consumption Minimization Strategy (ECMS) [4].

Bellman's Principle of Optimality states that the optimality of a future control action will not be affected by the optimality of any past control input [10]. DP uses this principle to progress backwards in time through a pre-determined drive cycle with identified states and control variables, and provides an optimal control path within constraints of the control space.

Chapter 3 provides a brief description of Dynamic Programming, followed by the method used to implement DP in HEVs and an example is provided.

### 3.1 Dynamic Programming

The DP algorithm is used on a class of discrete-time models in the following form,

$$x_{k+1} = F_k(x_k, u_k), \quad k = [0, N - 1] \quad (23)$$

where  $k$  denotes the index of discretized time,  $x_k$  the state variable,  $u_k$  the control variable, and  $F_k$  the function defining the state variable. In addition, for application of DP, the state and control variables are discretized.

The total cost of employing the control strategy  $\pi = \{u_0, u_1, \dots, u_{N-1}\}$  with the initial state  $x_0$  is defined by,

$$J_{0,\pi}(x_0) = g_0(x_0) + g_N(x_N) + \phi_N(x_N) + \sum_{k=0}^{N-1} [h_k(x_k, u_k) + \phi_k(x_k, u_k)] \quad (24)$$

where  $J_{0,\pi}(x_0)$  represents the total cost,  $g_0(x_0)$  and  $g_N(x_N)$  include the cost of initial and final steps respectively,  $\phi_k(x_k, u_k)$  the penalty function enforcing the constraints on the state and control variables, and  $h_k(x_k, u_k)$  the incremental cost of applying the control at time  $k$ . The optimal control path is one that minimizes the total cost represented in Equation 24.

### 3.2 Application of DP to HEV Supervisory Control

The DP algorithm is implemented in MATLAB 2016a as a backward-looking simulation in which the vehicle follows a pre-determined drive cycle and the steady state kinematic and torque relationships are used to compute component operation states. The main advantage of implementing a backward-looking simulation is the faster computation time, which comes at the cost of overlooking energy due to transient effects.

The DP control problem of the parallel HEV is characterized as,

$$\mathbf{x} = (\text{SOC}, v, F_{\text{req}}) \quad (25)$$

$$\mathbf{u} = (T_e, \omega_e, T_{\text{EM}}, \omega_{\text{EM}}, i_n) \quad (26)$$

$$h_k = \dot{m}_{\text{fuel}}(\mathbf{x}, \mathbf{u}) \quad (27)$$

where  $\dot{m}_{\text{fuel}}$  is the mass rate of fuel consumed,  $v$  is the vehicle speed, and  $F_{\text{req}}$  refers to force required.

The DP algorithm applied to HEVs, seeks to minimize the forward fuel consumption at any point of discretized state-time space. This minimizing operation can be summarized in Equation 28,

$$J_k(\text{SOC}_k^i) = \min[J_{k+1}(\text{SOC}_{k+1}^i) + \dot{m}_{\text{fuel}} + \phi_k] \quad (28)$$

The system design constraints, based on the operational limit of each component, are summarized as,

$$\text{Parallel HEV Component capability constraints: } \begin{cases} T_{e,\min} \leq T_e \leq T_{e,\max} \\ \omega_{e,\min} \leq \omega_e \leq \omega_{e,\max} \\ T_{\text{EM},\min} \leq T_{\text{EM}} \leq T_{\text{EM},\max} \\ \omega_{\text{EM},\min} \leq \omega_{\text{EM}} \leq \omega_{\text{EM},\max} \\ i_n \in [i_1, i_2, i_3, i_4, i_5] \end{cases} \quad (29)$$

where  $T_e$  and  $\omega_e$  refer to engine torque and speed respectively, and  $T_{\text{EM}}$  and  $\omega_{\text{EM}}$  refer to the EM torque and speed respectively, and  $i_n$  refers to the transmission gear ratio. Choices of control outside of the range specified in equation 29, results in application of a large penalty term.

In order to ensure the vehicle operates within a charge sustaining regime and the battery charge is within an allowable range, the following constraints are also applied,

$$\text{SOC}_{\min} \leq \text{SOC} \leq \text{SOC}_{\max} \quad (30)$$

$$\text{SOC}_0 = \text{SOC}_N = \text{SOC}_{\text{ref}} \quad (31)$$

using the penalty terms described earlier in Equation 24. The penalty terms are several orders of magnitude larger than the incremental cost of using fuel; otherwise they are zero.

In general, the computation time of a DP algorithm increases exponentially with the number of independent state and control variables since all permissible values of state and control variables are visited at each time step. Due to the nature of a backward-looking simulation and the prior knowledge of the vehicle speed at each time step, the drive cycle prescribes  $v$  and  $F_{req}$ , leaving SOC as the only independent state variable. In addition, by applying the steady state constraints, Equations 25-27, the independent control variables are reduced to  $u = (T_e, \omega_e)$ . Following this method, the control variables will be reduced to the set meeting the speed and torque requirements at the wheels. Due to the transmission coupling constraint,  $\omega_e$  choices are also limited to at most the number of gear ratios in the transmission due the kinematic constraint from Equation 27. Lastly,  $\dot{m}_{fuel}$  is assumed to be only a function of engine operation points characterized by  $T_e$  and  $\omega_e$ . During application of DP to an HEV, the discretized engine operation point  $(T_e, \omega_e)$  candidates are searched exhaustively to find the minimization operation point for Equation 25. An example Matlab code, for the application of the DP for a series architecture as shown in Chapter 2 is included in the appendix. For the purposes of the DP analysis a 9 speed transmission is added to the architecture to illustrate a more comprehensive application of DP.

Calculating the force required to power the vehicle using Equation 32,

$$F_{req} = \delta Mma + f_r mg + \frac{1}{2} C_d \rho_a A_{frontal} v^2 \quad (32)$$

and estimating the incremental cost of using fuel by referring the experimental fuel data, the total cost of applying control is calculated by adding the fixed costs as stated above. The DP

algorithm then determines the optimal control path by minimizing the total cost for the duration of the drive cycle.

Figure 3.1 represents the changes in the battery SOC for the duration of the drive cycle. In the design of hybrid electric vehicles, charge sustaining control strategies are utilized to ensure battery is available to power the vehicle at all times and store energy as needed. Another advantage of using charge sustaining control strategy is extended battery life. In order to ensure a charge sustaining operation, a high penalty term is applied when the SOC drops below or rises above the desirable SOC level.

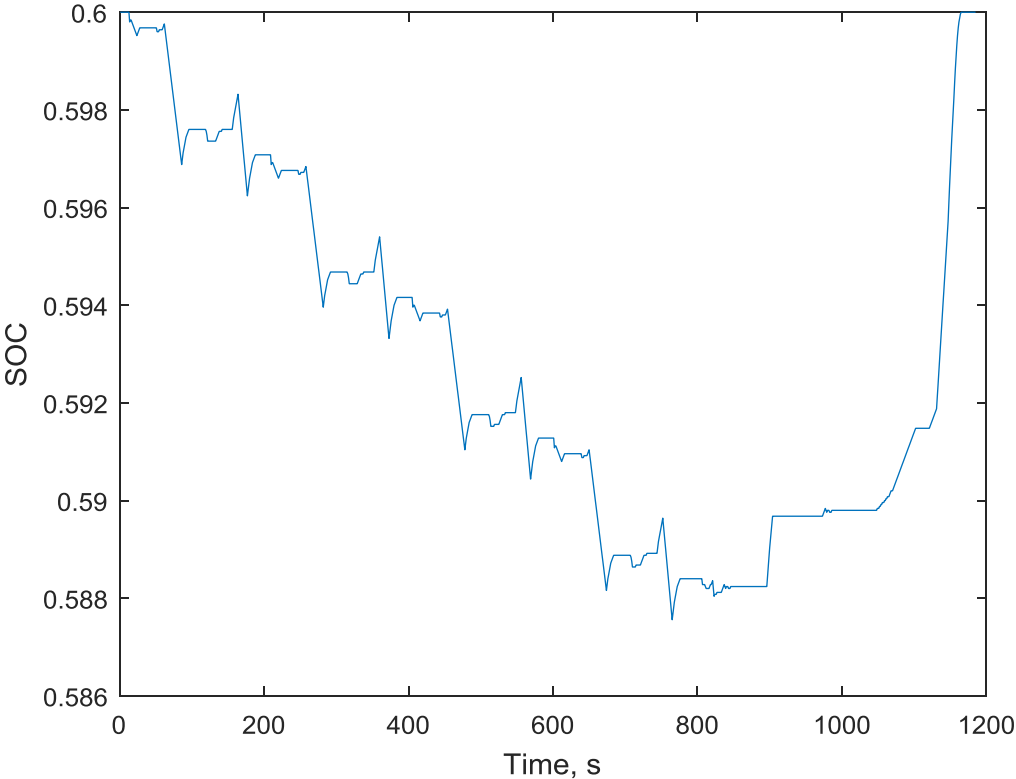


Figure 3.1: Battery SOC

Selecting the correct gear can also improve the efficiency of the drive as the engine can operate at a more desirable speed. The efficiency of the power transmission through the gearbox and other architecture components should also be considered when applying DP. Experimental data provided by Ferrari S.p.A. were used in order to estimate the energy loss through the powertrain.

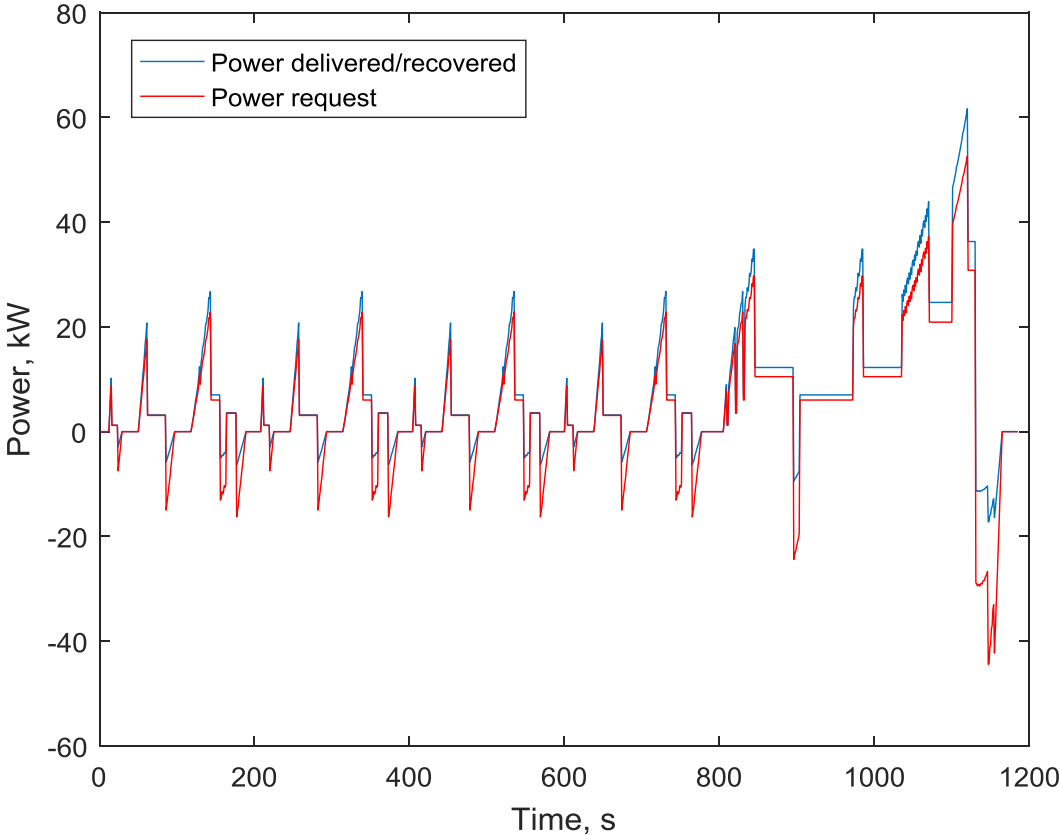


Figure 3.2: Power required vs. Power supplied/recovered

Figure 3.2 illustrates the effects of powertrain inefficiencies considered in the DP algorithm. During periods of deceleration due to the inefficiencies of the EM and other powertrain components some of the power available for recovery in the energy storage unit is wasted.



## **Chapter 4: Multi-physics co-simulation of battery thermal model**

The analysis and proper design of the energy storage system is a critical concern in the product development phase of HEVs. Rechargeable Li-ion batteries (LIBs) feature many advantages compared to alternative storage devices. LIBs provide high power and energy density over a broad operating temperature range, as well as low self-discharge, stable chemistry, and long battery life [11-12]. In addition, the growth in the market demand and production of high power density LIBs has led to a sharp drop in price of such batteries. However, challenges remain in the development of HEVs using LIBs due to stringent safety and performance requirements. Some of the limitations of the current LIBs include local underutilization, capacity drop, localized thermal degradation, and stress-induced material damage.

Significant efforts have been put forward to develop simulation based tools to analyze battery performance and predict operational states of LIBs. The porous electrode theory, which predicts the lithium diffusion dynamics and charge transfer in an electrode system has been investigated to predict the electrical dynamics of a cell [12-13]. This approach can be used to investigate the microscopic behavior of the cell. Others have proposed and investigated an equivalent circuit model to estimate the electrical response and heat generation [12, 14-15]. Researchers have also focused on the structural response of the LIBs and effects on performance degradation. The volume change of electrode material has been investigated to estimate the electro-thermal induced stress and strain [12] and the potential effects on capacity have been thoroughly investigated. Several lumped parameter thermal models have been proposed and validated through experiments [16-17]. Great effort has been placed on developing coupled models between electro-chemical and heat transfer domains in order to investigate the coupled

effects of discharge/charge current, voltage, and temperature on the SOC and battery state of health (SOH) [18-20].

Recent research has focused on development of coupled electro-thermal and heat transfer models in order to investigate the cell's temperature distribution under extreme conditions, such as over-charging/discharging, internal short circuit, and extrusion [16]. Coupled heat transfer and computational fluid dynamics models with a lumped value of heat generation were used to visualize the temperature gradient for both time-dependent and steady state simulations [16,21-22]. Multiple co-simulation platforms were used to simulate coupled thermo-electrical and heat transfer models. The effects of manufacturing variations and battery pack organization on cell thermal and power behavior were also investigated [21]. Moreover, researchers have been able to characterize the effects of localized high temperature regions on cell performance and SOH [22]. However, research on development of fully coupled electro-thermal, heat transfer, and computational fluid dynamics models to investigate temperature distribution at the battery pack level, which identify hot spots, is still in early stages. To the knowledge of the author, the coupled electro-thermal cell models have not been developed further to include the structural layout of battery pack with high capacity cooling units.

This chapter proposes a co-simulation approach to simulate modular and fully-coupled multi-physics models of a LIB battery with high fidelity thermal domain modeling using commercial off the shelf (COTS) software. The proposed multi-physics model couples a discrete electric model and continuous thermal model domains using the co-simulation capability of the SIMULIA simulation platform for rapid and scalable model development. Furthermore, the computational cost has been reduced by the development of a discrete electrical domain, as well as limiting the data exchange between co-simulation model components to discrete

communication points. The main purpose of this multi-physics co-simulation is to predict the temperature distribution and identify temperature hot spots. The critical system parameters, which govern the electrical and thermal dynamics, were provided by the cell manufacturer. Experimental test data was used to calibrate the system response and ensure high model fidelity.

#### **4.1 Model Development**

One of the unique challenges facing Ferrari engineers and others in development of HEVs is the design and verification of the battery pack. Due to the high-performance characteristics of the vehicle, the battery pack should be designed to operate under high-frequency-high-load conditions, including high current charge and discharge which can lead to areas of high temperature within the battery pack. As such, the battery pack model includes a thermal domain in order to investigate:

- 1) The spatial distribution of temperature to identify hot spots,
- 2) The effects of high temperature on performance characteristics such as battery life,
- 3) The cooling power required to maintain battery temperature.

The modeling approach includes three simulation domains: electro-thermal, thermal-structural and fluid. The discrete electrical model refers to the mathematical representation of the battery dynamics, using ordinary differential equations, relating the transfer of chemical energy to electrical energy and vice versa, during discharge and charge, respectively. The model also relates the battery charge/discharge current to heat generated and uses ordinary differential equations to estimate the heat loss to the environment and cell core temperature. The continuous thermal-structural model, developed using COTS software, refers to the heat transfer from each individual cell through the battery structure, which includes a cooling plate, to the flow of

coolant in the cooling system and the surroundings air through convection. The continuous fluid domain refers to the heat transfer to the cooling fluid within the cooling plate. Figure 4.1 illustrates the interactions between the three model domains.

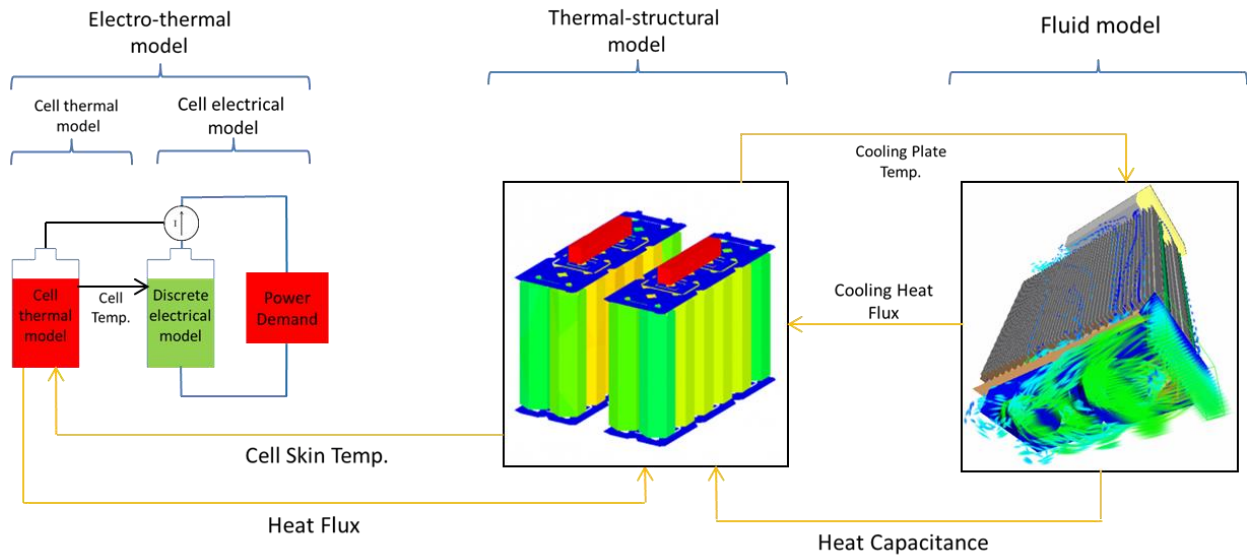


Figure 4.1: Battery model domains, orange arrows represent the flow of communication signals between co-simulation components. Note: Images representing the thermal-structural and fluid domain models only represent the hypothetical layout of the battery module and the coolant flow

The three simulation domains were coupled based on the conservation of energy. The heat loss from the electro-thermal model is imposed as heat flux input at the internal boundary of the cell thermal-structural model. In turn the temperature at the internal surface of the cell thermal-structural model is enforced as the temperature of the electro-thermal model. The thermal-structural model is coupled with the fluid model at an interface surface region. At this interface, the temperature of thermal-structure surface is imposed to the fluid model as a boundary condition. In turn, the fluid model heat capacitance and heat flux are enforced as the thermal-structural model boundary conditions and input heat flux.

Co-simulation requires the identification of master and slave components. The thermal-structural model is identified as the master component, while the fluid and electro-thermal domain function as slave components. The master initiates co-simulation by taking the first time step. The results and co-simulation time are communicated to the slave components and each slave takes appropriate steps to reach the co-simulation time, after which the results are communicated to the master and other slave components. The co-simulation progresses as the master takes the next step.

In order to establish communication between the three model components using a co-simulation approach, the electro-thermal model is translated into a tool-independent library, Functional Mock-up Unit (FMU), which can be used to execute the model in any Functional Mock-up Interface (FMI) compatible environment and establish communication between the electro-thermal model and other FMUs, or model components developed in the host, FMI compatible, modeling environment. The electro-thermal domain includes an individual cell (multiple cells can be arranged to form a battery module and/or pack), control unit, and power cycle in the form of a look-up table. The thermal-structural heat transfer model is developed using Finite Element Modeling (FEM) software, Abaqus 2016, and includes all the battery pack components. Similarly, the fluid domain model is developed using Abaqus 2016 and includes the model of cooling fluid that removes heat from the cooling plate. Abaqus 2016, which is part of the SIMULIA simulation platform, can establish communication with the electro-thermal domain FMU following the FMI standard. Development of high fidelity models for each domain is discussed in Section 4.1. In Section 4.2, the cell model was calibrated using experimental test results. Sections 4.3-4.7 focus on the next steps of the research partnership and potential applications of the battery pack model.

### **4.1.1 Electro-thermal Domain**

In order to develop a modular model of the battery cell, the Modelica language is used. Modelica is an open-source, object-oriented, and equation-based modeling language which can be used to easily model complex physical systems containing electrical, mechanical, hydraulic, thermal, electronic, control and process oriented subcomponents [23,24]. A wide array of open-source Modelica Libraries have been developed, providing rapid and scalable model development capability.

Dymola is a commercial modeling and simulation environment based on the Modelica modeling language. Dymola has a built-in capability to simulate object-oriented models as well as export models as Functional Mock-up Units. FMUs are executable function files, generated following the Functional Mock-up Interface, which is a tool-independent standard to support model exchange and co-simulation of dynamic models. FMI allows for the coupling of several independent simulation tools to develop modular coupled models. Data exchange is restricted to discrete communication points, as specified by the host simulation tool, and each subsystem is solved independently between the communication points. More information about FMI and its functionality has been published elsewhere [25-26].

In order to streamline the development of a cell model, the open-source Electric Energy Storage (EES) library developed jointly at the Austrian Institute of Technology and Vienna University of Technology was used. The library contains sub-components of different complexity, such as individual cells and stacks interacting with loads, battery management components and charging devices.

The EES library components are designed as universal components allowing the end users to easily simulate specific scenarios by varying their parameterization. The use of the Modelica language allows seamless modifications of the equations used in each subcomponent. For the purposes of this analysis, the built-in LinearDynamicImpedance cell model was modified in order to properly define the cell model, based on experimental data. The EES library is structured as shown in Figure 4.2.

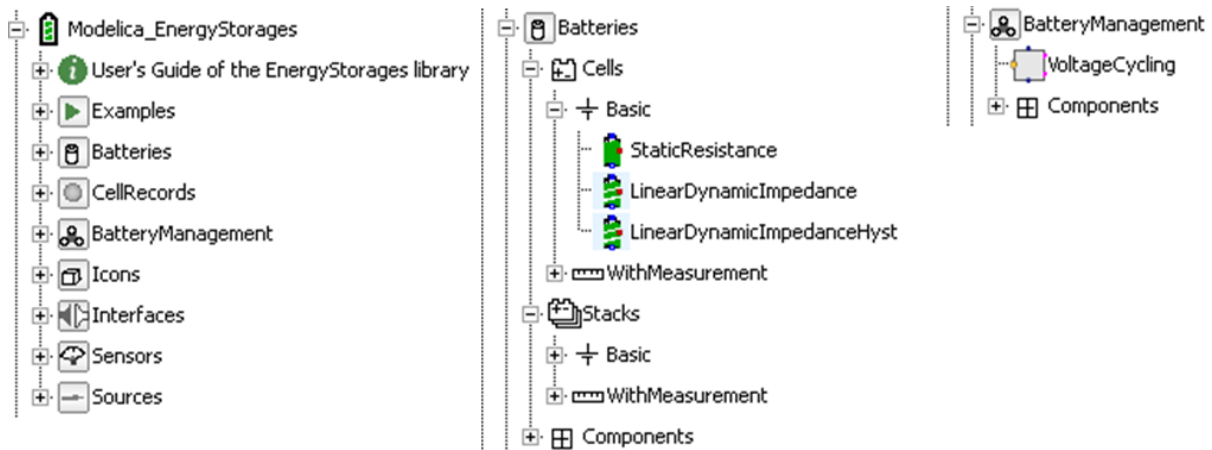


Figure 4.2: Electric Energy Storage library, Batteries and Battery Management subsystems

The Batteries package includes models for individual cells as well as stacks of  $n_s$  cells connected in series and  $n_p$  cells in parallel. A single cell component is selected for the purpose of this analysis. The single cell model features a positive (pin\_p), a negative (pin\_n) and an optional temperature connector (heatPort) which is activated herein as cell parameters were defined as a function of temperature [27]. The common parameters used in the cell model are given in Table 4.1.

Table 4.1: Common cell parameters [27]. Starred values\* are calibrated

Name	Unit	Description	Value*
$SOC_{ini}$		Initial state of charge	$0 \leq SOC \leq 1$
$OCV_{tablecharge}$	V	Look-up table for the OCV vs. SOC during charge	Figure 4.3
$OCV_{tabledischarge}$	V	Look-up table for the OCV vs. SOC during discharge	Figure 4.3
$Q_{ini}$	C	Initial transferred charge	0
$C_0$	C	Capacity at $T_{ref}$ for $Q_{ini} = 0$ and $t = 0$	2880
$useHeatPort$		Boolean variable for using the heat port	On/off
$T_{ref}$	K	Reference temperature	293
$alphaR_s$	$K^{-1}$	Linear temperature coefficient for $R_s$	0.013*
$R_{sref}$	$\Omega$	Ohmic resistance at reference temperature	0.014
$alphaC$	$K^{-1}$	Linear temperature coefficient for capacity	-1.6
$C_{Heat}$	J/K	Cell heat capacity	96*

The model output quantities are given in Table 4.2 and the calculations are presented next. The removed charge is given by,

$$Q = \int_{t_{start}}^{t_{stop}} I(t) dt \quad (33)$$

and the total transferred charge is,

$$Q_{abs} = Q_{ini} + \int_{t_{start}}^{t_{stop}} |I(t)| dt \quad (34)$$



The OCV of a battery cell changes with the SOC. Experimental results have shown that the OCV function is different during charge and discharge of the cell. As shown in Figure 4.3, OCV is provided using two linearly interpolated look-up tables for charge and discharge. The open circuit voltage is limited between the charging voltage limit CVL (SOC=1) and the discharge voltage limit DVL (SOC=0) [27].

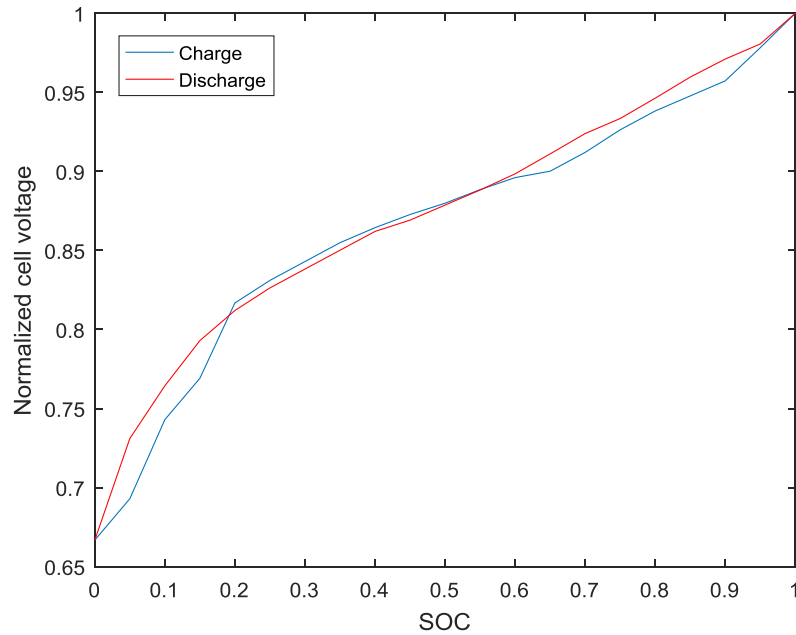


Figure 4.3: Normalized Cell Voltage

The model can consider the effects of aging by considering both calendaric aging and cycling. Calendaric aging is estimated from time, which for the purposes of the current analysis, is minimal and therefore removed from the model. Cycling is directly proportional to absolute transferred charge  $Q_{abs}$ . Aging of the cell mainly influences the performance by decreasing capacity and increasing the internal impedance [27].

The cell capacity calculated by,

$$C = (C_0 + K_{CQ_{abs}} \times Q_{abs}) [1 + \alpha_C \times (T_{heatPort} - T_{ref})] \quad (35)$$

is temperature dependent and decreases with increased transferred charge due to cycling.

The single cell model features a single, temperature dependent, ohmic impedance, modeled as,

$$R_s = R_{ref} [1 + \alpha_{Rs} \times (T_{heatPort} - T_{ref})] \quad (36)$$

which does not consider the increase in impedance due to aging.

The SOC is given at each time step as,

$$SOC = SOC_{ini} - \frac{Q}{C} \quad (37)$$

and the equivalent number of cycles

$$cycles = cycles_{ini} + \int_{t_{start}}^{t_{stop}} \frac{|I(t)|}{2C} dt \quad (38)$$

with

$$cycles_{ini} = \frac{Q_{ini}}{C_0 + C} \quad (39)$$

relates the total transferred charge to the cell capacity C. Therefore, one cycle is equivalent to the charge transfer of one full charge and one full discharge of the capacity C [27].

The Cell model includes *num* serially connected RC elements, as shown in Figure 4.4. The cell model was manually edited to include two sets of RC elements representing the cell dynamics in charge and discharge. Based on the direction of cell current, a switch is used to connect and disconnect the corresponding RC component.

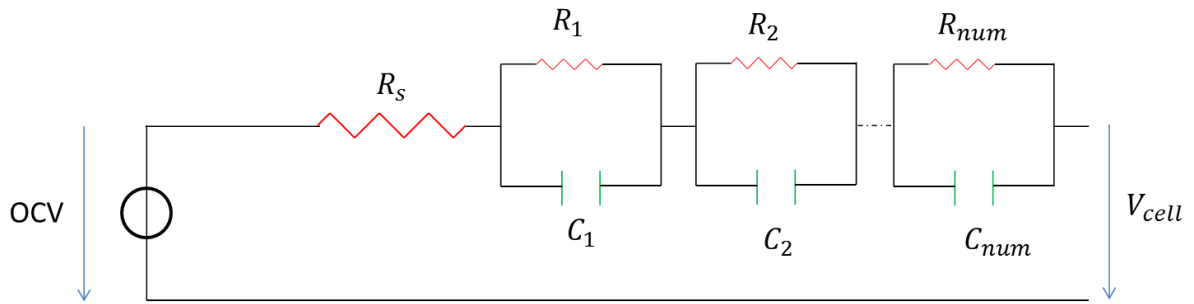


Figure 4.4: Cell model with one ohmic impedance and  $num$  serially connected RC elements

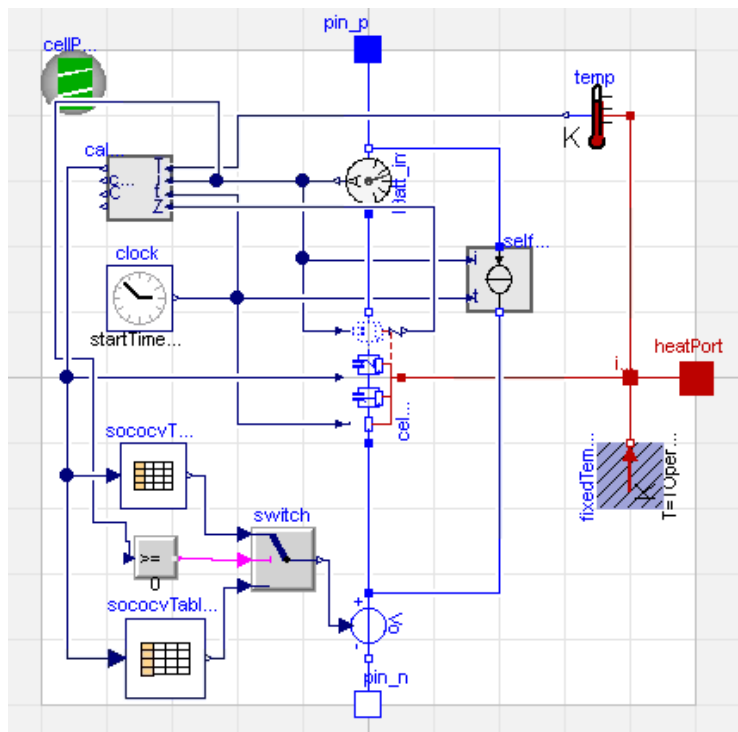


Figure 4.5: Cell model diagram

A list of calculated output variables of the modified *LinearDynamicImpedance* cell model used in the analysis is available in Table 4.2.

Table 4.2: Calculated output variables from cell model [27]

Name	Unit	Description
$SOC$		State of charge
$OCV$	V	Open circuit voltage
$Q_{abs}$	C	Total transferred charge
$cycles$		Number of equivalent cycles
$t$	S	Calendaric cell time
$C$	C	Cell capacity
$V$	V	Cell voltage

The EES library contains a battery management block (*Cycling*) which is used to implement a rule-based control of the battery cell. It has three Boolean outputs which control the operation of the charging unit, power cycle look-up table and the cell. The battery management block can be used to cycle the cell over a predefined voltage range  $[V_{min}, V_{max}]$ , while limiting the maximum current draw during discharge and charging current. The EES also includes a constant current, constant voltage (CCCV) charging device which charges the cell with constant current until the constant voltage level is reached [27]. The battery module under study contains twelve individual cells arranged in two series sets of six parallel cells, as shown in Figure 4.6.

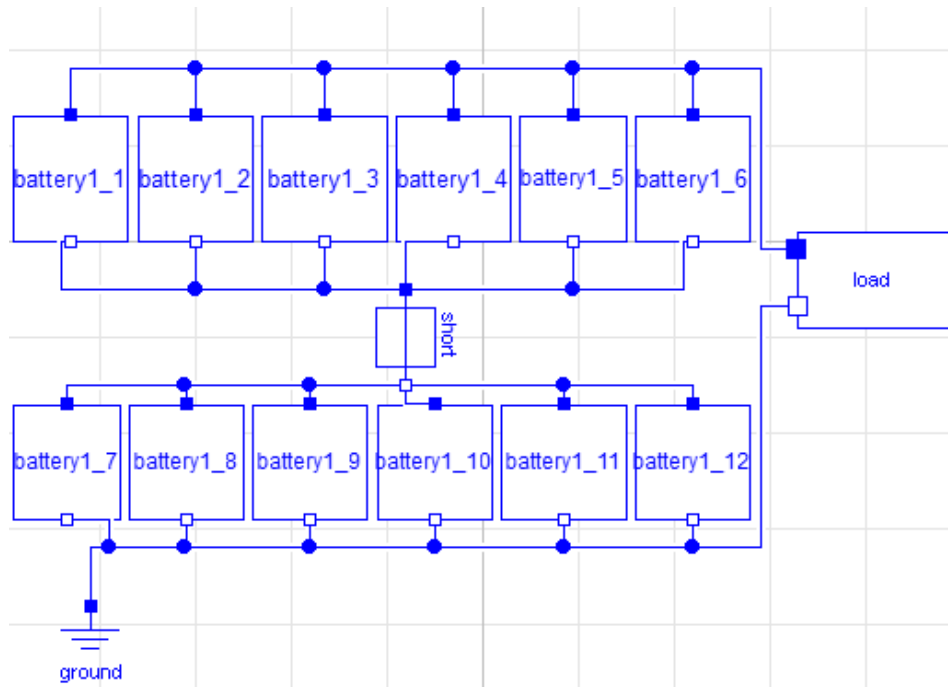


Figure 4.6: Diagram of module with 12 cells

The subcomponent labeled *load* includes the charging unit, power cycle look-up table, a switch and the cycling rule-based control unit. The switch works by disconnecting the power cycle when the cell/module drops below  $V_{min}$  and connecting the charger. The switch features a short delay to prevent numerical instability. Similarly, as the voltage reaches the constant voltage level, the switch disconnects the charger and connects the power cycle to continue the power cycle [27]. Figure 4.7 illustrates the construction of the *load* subcomponent.

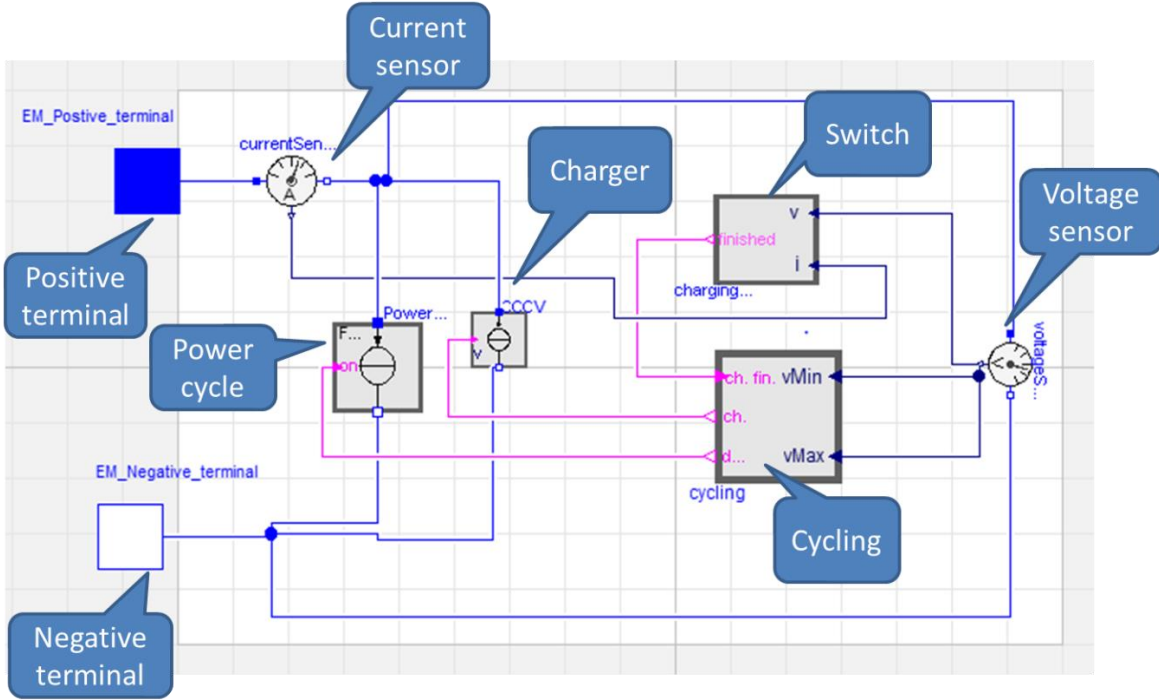


Figure 4.7: Charging unit and model controller

In order to study the effects of heating on battery performance and investigate the cooling power needed to ensure the safe operation of the battery pack, the heat generated during cycling is calculated by

$$Q_{Heat} = R_s I^2 \quad (40)$$

and the temperature of the cell is calculated by the heatCapacitor model by solving,

$$T = \int_{t_{start}}^{t_{stop}} \frac{Q_{Heat} - Q_{cooling}}{C_{Heat}} dt \quad (41)$$

where  $C_{Heat}$  is the heat capacity of the battery cell and  $Q_{cooling}$  is the heat dissipated through the cooling plate. An additional thermal resistance is included in the model to represent the heat

resistance of the skin of the cell. Figure 4.8 illustrates the application of Equations 40 and 41 to the electric domain model.

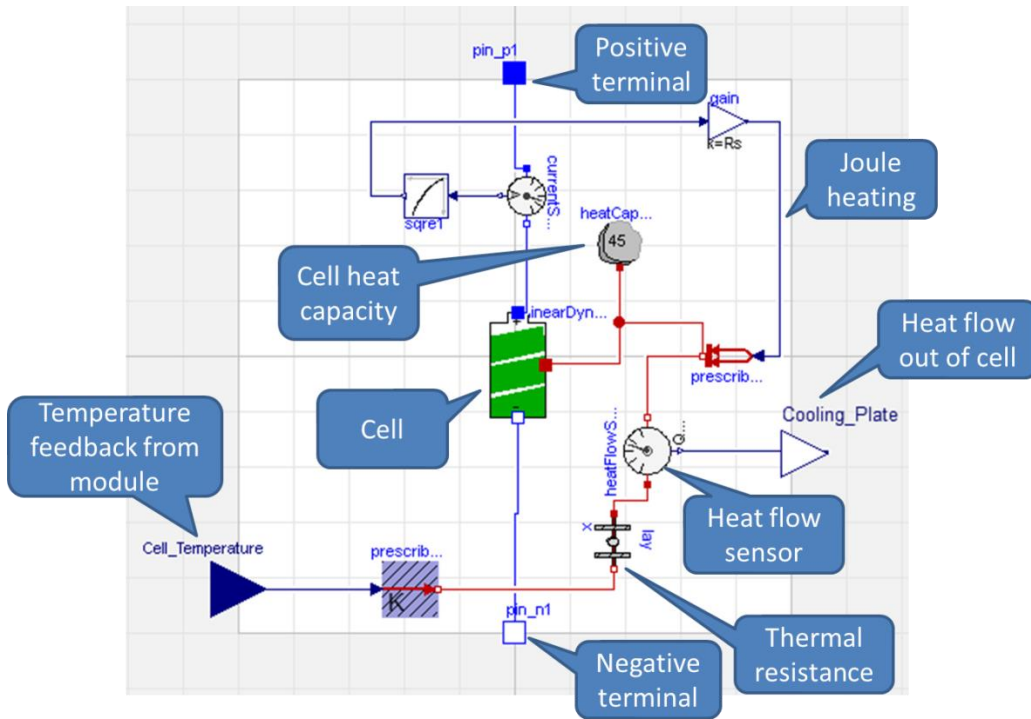


Figure 4.8: Single cell model interaction with the co-simulation engine

Due to the complex geometry of the battery module shown in Figure 4.9, the heat transfer model for the module and cooling plate is developed using FEM as discussed in Sections 4.1.2 and 4.1.3. The co-simulation approach used to communicate between the three domains is introduced in Section 4.1.4.

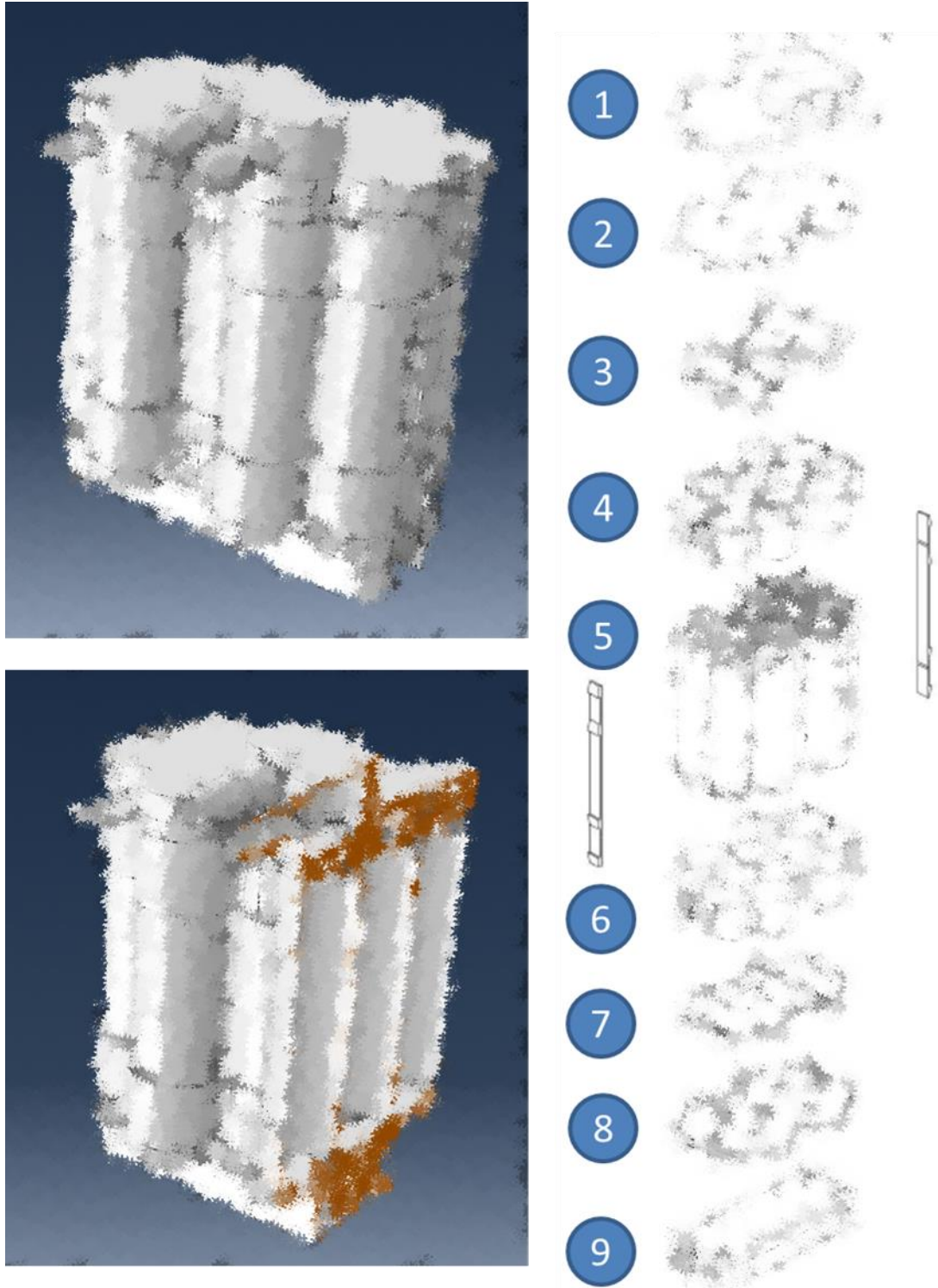


Figure 4.9: CAD drawing and assembly of the battery module. Module sections: 1: Cooling plate, 2&8: Cover, 3&7: Connecting plate, 4&6: Housing, 5: Cylindrical cells and clips, 9: Battery Management System (BMS) \*\* *Image obscured for proprietary reasons.*



### 4.1.2 Thermal-structural Domain

In order to create a high fidelity model of the battery module, a transient heat transfer model is also developed. The thermal properties of each component is assigned as shown in Table 4.3. Each part of the module was discretized on part using standard linear tetrahedron elements (DC3D4). Figure 4.10 provides an overview of the thermal-structural heat transfer model.

Table 4.3: Material properties of the thermal-structural model components

Sections	Conductivity [ $\frac{W}{m.K}$ ]	Density [ $\frac{kg}{m^3}$ ]	Specific Heat [ $\frac{J}{kg.K}$ ]
Cooling Plate	167	2700	902
Cover, Housing, Cell, Clips, BMS cover	0.2	2000	1800
Connecting plate	400	8960	385
BMS circuit	149	2329	700

Under the assembly module a dependent instance was created containing all components and a surface of type geometry, named *INTERFACE*, was created using the internal surfaces of the cooling plate as shown in Figure 4.11 (marked in red). The surface is used to establish interaction with the fluid model. Temperature at the surface is defined as in output from the structural-thermal to the fluid model (applied as boundary condition in the fluid model). The surface was used to define a heat flux interaction with the fluid domain model developed in Section 4.1.3

In order to define heat transfer within the module, a conduction (contact) interaction property is defined as a function of clearance between surfaces, with conductance of 20000  $[W/m.K]$  with zero clearance and 0 conductance with 2mm clearance. Convection heat transfer is also defined as an interaction property of type film condition, with a film coefficient of  $0.01[W/m.K]$ . Convection is only defined on the external surface of the module.

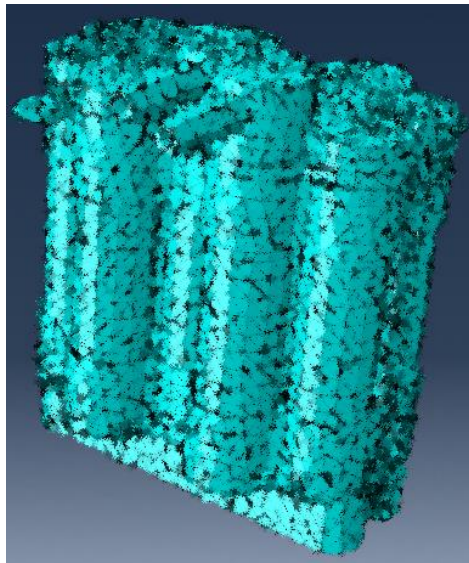


Figure 4.10: Thermal-Structural model *\*\* Image obscured for proprietary reasons.*

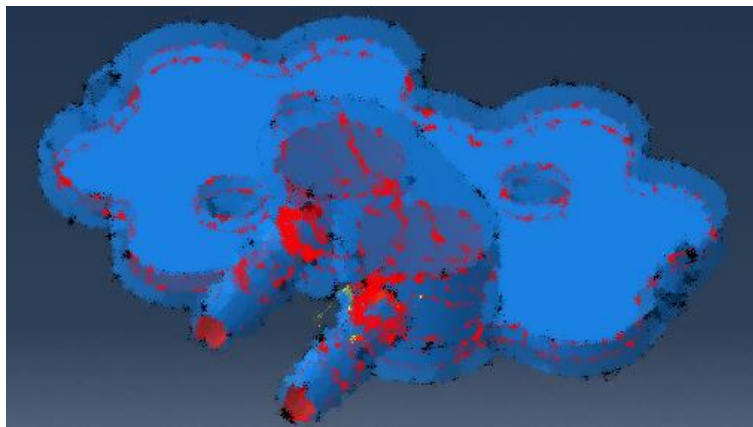


Figure 4.11: Fluid model interaction. Internal surface (marked in red) is identified as *INTERACTION* for interaction with fluid model *\*\* Image obscured for proprietary reasons.*

Similar to interaction with the fluid model, an interface surface was created, named *INTERFACE2*, in order to establish communication with the electro-thermal model. Figure 4.12 displays a cut section of the thermal-structural model with the cell internal surfaces marked in red. Heat flux from the electro-thermal model,  $Q_{cooling}$ , is identified as an input to the thermal-structural model from the electro-thermal model. In turn, the surface temperature at *INTERFACE2* is reported back to the electro-thermal model.

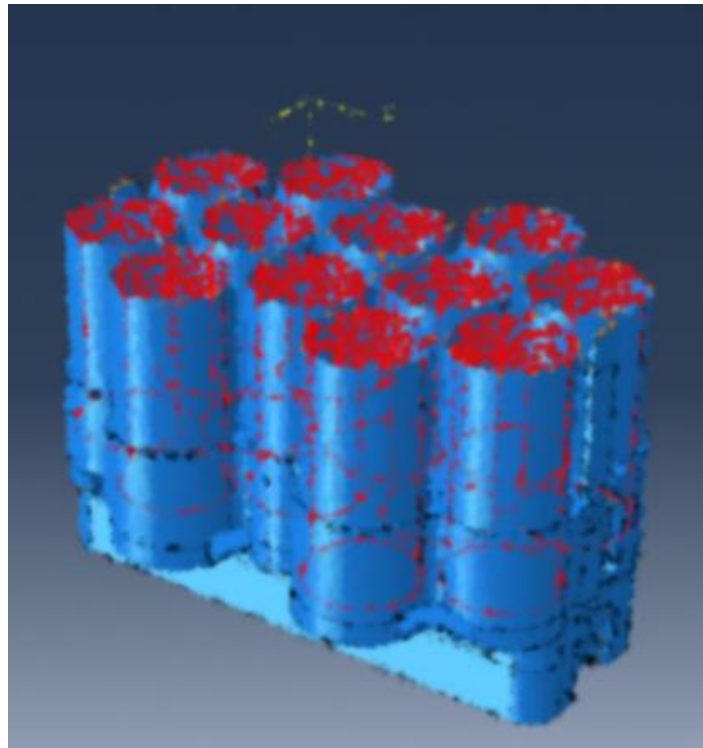


Figure 4.12: Electro-thermal model interaction. Internal surface of cells (marked in red) were identified as *INTERACTION2* for interaction with electro-thermal model *\*\* Image obscured for proprietary reasons.*

In order to ensure correct initialization of the analysis, a constant initial temperature is defined for the entire model at the initial time step. A transient step, named *Co-simulation*, of length  $t_{simulation}$  is created with the maximum number of increments set at  $t_{simulation} \times 50$  to

ensure simulation is completed. A load of type *Body heat flux* is created and applied to all 12 cells representing input heat flux,  $Q_{cooling}$  from the electro-thermal model. Finally an amplitude of type *Actuator*, named *BODYFLUX*, was defined and the amplitude of the load was updated.

In order to define the co-simulation interaction property between electro-thermal, thermal-structural, and fluid domain models, an input file, named *Standard.inp*, was generated using the thermal-structural heat transfer model created. Abaqus CAE interface does not allow the definition of the co-simulation interaction, thus the input file was edited as shown in Section 4.1.4.

### 4.1.3 Fluid Domain

In order to create a high fidelity model of the cooling fluid, a transient CFD model is developed. The thermal properties of the coolant, which is a mixture of 50% water and 50% Ethylene glycol, is assigned as shown in Table 4.4. The fluid model, which was generated to fill the void in the cooling plate part, was discretized on part using CFD linear tetrahedron elements (FC3D4), as shown in Figure 4.13.

Table 4.4: Coolant thermal properties

Fluid	Conductivity [ $\frac{W}{m.K}$ ]	Density [ $\frac{kg}{m^3}$ ]	Viscosity [ $\frac{N.s}{m^2}$ ]
Coolant	0.43	1056	0.0085

Under the assembly module a dependent instance was created and three surfaces of type mesh, named *INLET*, *OUTLET*, and *INTERFACE* were created as shown in Figure 4.14. The *INTERFACE* surface was used to define a heat flux interaction with the thermal-structural domain model as described in Section 4.1.2. The *INLET* surface was used to define the fluid inlet

boundary conditions. Similarly, the outlet boundary conditions were applied to the *OUTLET* surface, as described in Table 4.5.

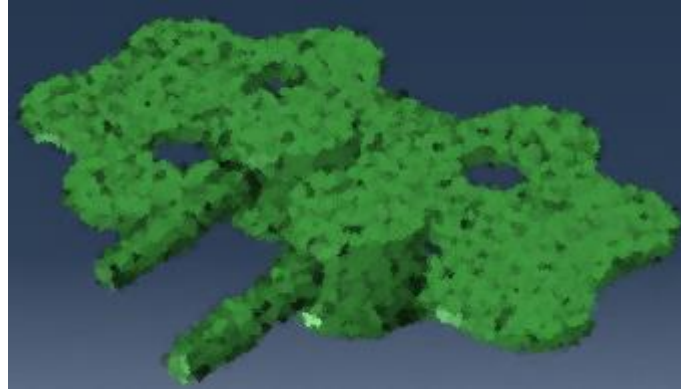


Figure 4.13: Fluid domain model (discretized) *\*\* Image obscured for proprietary reasons.*

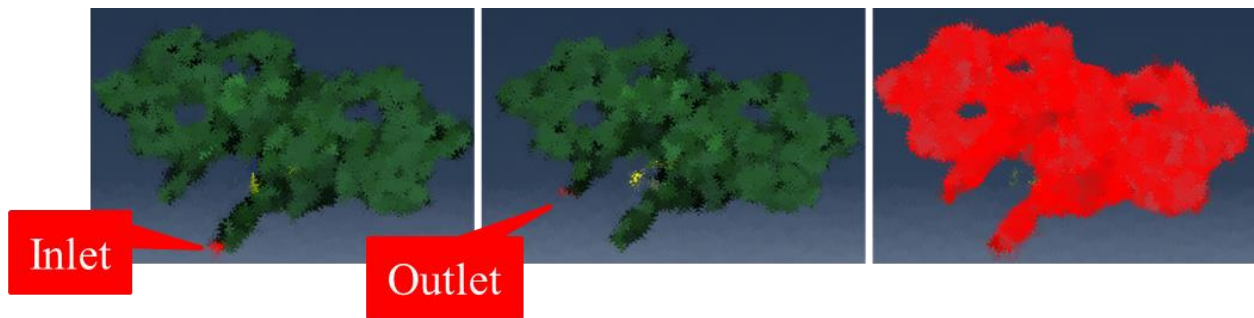


Figure 4.14: Fluid domain interactions (Marked in red). Left: Inlet, Middle: Outlet, Right: Interface *\*\* Image obscured for proprietary reasons.*

Table 4.5: Flow properties of the coolant

Boundary Region	Inlet	Outlet
Thermal properties	15 C <sup>o</sup>	18 C <sup>o</sup>
Flow properties	10 $\frac{Lit}{min}$	0 relative pressure

In order to ensure correct initialization of the analysis, a constant initial temperature is defined for the entire model at the initial time step. A transient step, named *Co-simulation*, of length  $t_{simulation}$  (equal to duration of the thermal-structural heat transfer model) was created to ensure compatible simulation time between the co-simulation components.

Similar to the thermal-structural heat transfer model, in order to define the Co-simulation interaction property between electro-thermal, thermal-structural, and fluid domain models, an input file, named *Fluid.inp*, was generated using the CFD model created. The Abaqus CAE interface does not allow the definition of the co-simulation interaction, thus the input file was edited as shown in Section 4.1.4.

#### **4.1.4 Co-simulation**

Co-simulation between Abaqus/Standard (standard heat transfer model) and Abaqus/CFD (fluid domain model) is governed by an additional process, the SIMULIA Co-Simulation Engine (CSE) director. Typically, the CSE director is automatically invoked and the co-simulation parameters are stored in the co-simulation configuration file. The co-simulation as specified by the CSE director is illustrated in Figure 4.15. An overview of the Abaqus co-simulation procedure is described in Section 17.2.1 of the Abaqus User's Guide [28].

In order to execute co-simulation between models generated using Abaqus products and co-simulation format FMU files, the Abaqus command line should be used to invoke the CSE director and the co-simulation configuration file should be manually generated. Proper execution of co-simulation requires:

- 1) Proper definition of the interaction between the three models as outlined in the CSE User's Guide [29] and API reference Guide [30].

- 2) Compatible and consistent definition of co-simulation time and step size.
- 3) Definition of co-simulation parameters in the co-simulation configuration file as specified by the CSE director.
- 4) Definition of the co-simulation master and slave.

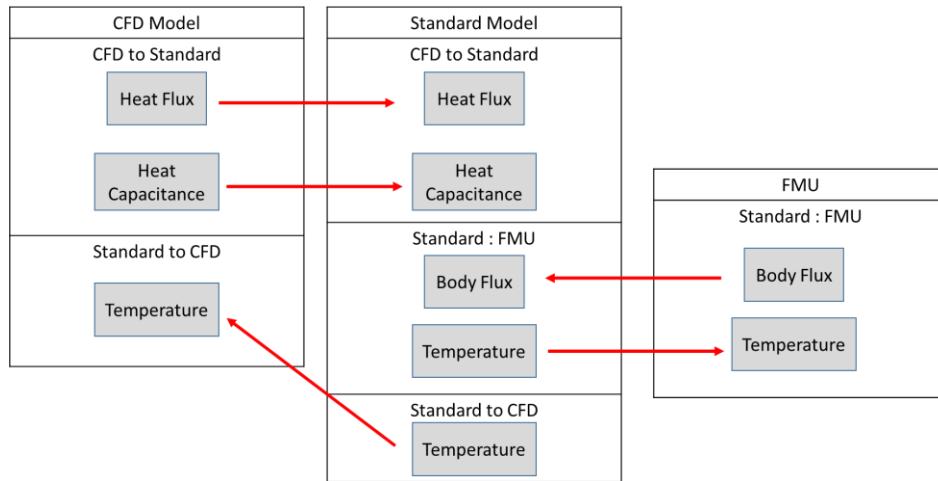


Figure 4.15: Co-simulation process

The CSE establishes communication between co-simulation models by assigning priority to individual models. The model designated as the master leads the co-simulation by taking the next time step,  $t_{next}$ , and results are communicated to slave models. The slave models will then individually take appropriate step sizes to advance to  $t_{next}$  and report the results the CSE director, which communicates the information to the master. The master will then take the next time step,  $t_{next+1}$ , and the co-simulation will continue, so on and such forth.

In preparing the Dymola model of the battery module for co-simulation, an FMU of type co-simulation and version 1.0 was generated. More information about generating an FMU using Dymola is available in the Dymola User's Guide [31]. The FMU file generated using Dymola is a compressed file which contains a Dynamic-link library (DLL) executable file as well as a

model description file in XML (eXtensible Mark-up Language) format. The CSE director automatically assigns FMU model instances as slave models.

As stated previously in Section 4.1.2, definition of the interaction between the three co-simulation models required manually editing the thermal-structural model input file, *Standard.inp*. Lines, listed in Figure 4.16, were added to the input file within the definition of the transient time step in order to establish proper communication between the standard heat transfer model and fluid model.

```
** STEP: Transient Step Name
...
...
...

** Interaction: Name
*Co-simulation, name=Name, controls=Name_Ctrls, program=MULTIPHYSICS
*Co-simulation Region, import, type=SURFACE
interface, CFL
interface, LUMPEDHEATCAPACITANCE
*Co-simulation Region, export, type=SURFACE
interface, NT
*Co-simulation Controls, name=Name_Ctrls, coupling scheme=gauss-seidel, time incrementation=subcycle, time
marks=yes,
step size=min, scheme modifier=lead
...
...
...
*End Step
```

Figure 4.16: Definition of co-simulation parameters (Abaqus Standard model). \*Sections of the code referring to definition of boundary conditions and other step parameters are omitted to ensure generality and replaced with ...

For the purpose of this analysis, communication between three models was defined using surface regions. For a complete description of fields available for exchange between co-simulation models interface regions, please refer to Section 17.2.1 of the Abaqus User's Guide [28]. As shown in Figure 4.15, the co-simulation controls, using the keyword *scheme modifier*,



was used to assign the thermal-structural heat transfer model as the master. The keyword *incrementation*, was used to allow the master to independently take an appropriate step size.

Similar to the thermal-structural heat transfer model, definition of the interaction between the three co-simulation models required manually editing the thermal-structural model input file, *Fluid.inp*. Lines, listed in Figure 4.17, were added to the input file within the definition of the time step in order to establish proper communication between the standard heat transfer model and fluid model.

```
** STEP: Fluid Step Name
...
...
...

** Interaction: Name
*Co-simulation, name=Name, controls=Name_Ctrls, program=MULTIPHYSICS
*Co-simulation Region, import, type=SURFACE
interface, TEMP
*Co-simulation Region, export, type=SURFACE
interface, HFL
interface, LUMPEDHEATCAPACITANCE
*Co-simulation Controls, name=Name_Ctrls, coupling scheme=gauss-seidel, time incrementation=lockstep, time
marks=yes,
step size=min, scheme modifier=lag
...
...
...
*End Step
```

Figure 4.17: Definition of co-simulation parameters (Abaqus Fluid model). \*Sections of the code referring to definition of boundary conditions and other step parameters are omitted to ensure generality and replaced with ...

The co-simulation controls, using the keyword *scheme modifier*, was used to assign the fluid model as a slave. The keyword *incrementation*, was used to force the slave to take appropriate step sizes and report the results at time  $t_{next}$ .

The co-simulation configuration file defines the simulation properties of the system and the numerical methods employed to simulate the system in its defined environment. This includes:

- **components:** subsystem simulator programs used in the co-simulation,
- **component instances:** subsystem simulations performed using identified simulators,
- **connectors:** available input and output simulation results from each subsystem,
- **connection sets:** pairing of subsystem simulation results,
- **connection categories:** time in the co-simulation when these pairings are relevant or active, and
- **model of computation:** numerical method used in the co-simulation.

In order for the configuration file to correctly identify the co-simulation parameters and be accepted by the CSE director, the following requirements must be satisfied:

- 1) XML well-formedness. The CSE director or any other XML authoring tool can confirm the well-formedness.
- 2) Abiding by the CSE schema definition. The rules and constraints are described formally in the CSE API kit.
- 3) Internal consistency of references.
- 4) Topological and algorithmic consistency. The definition of co-simulation and model parameters within the configuration file should be consistent with the problem definition in each subcomponent.
- 5) External consistency of references. The author of the configuration file should ensure consistency of names registered with the CSE director and each subcomponent.

The first four requirements can be verified by using the `datacheck` option. External consistency of references can only be verified at run time.

The configuration file contains six sections as described in the following paragraphs. The complete text of the configuration file created for the purpose of this analysis is include in the Appendix. For more details about the co-simulation configuration file refer to the CSE User's Guide and the CSE configuration schema documentation.

### 1) Document header

The first two lines, illustrated in Figure 4.18, in the file identify the document as an XML compliant and define the XML root element as `<CoupledMultiphysicsSimulation>`, which uniquely defines the document as relevant to the CSE. The next line segment will add the header. Various information can be included within the header however the only required information is the schema version.

```
<?xml version="1.0" encoding="UTF-8"?>
<CoupledMultiphysicsSimulation>
...
<header>
  <SchemaVersion>1.1</SchemaVersion>
</header>
```

Figure 4.18: Co-simulation configuration file document header

### 2) Components

The components section of the configuration file defines each model included in the co-simulation (purple highlighting). As illustrated in Figure 4.19, using the keyword `<bottomUpImplementation>`, the XML file introduces co-simulation components that are developed within the SIMULIA package, and the phrase `<topDownImplementation>` identifies

external components such as FMUs (green highlighting). In addition, a brief description of the master model, including the unit definitions (yellow highlighting) and co-simulation variables, will identify the master subcomponent to the CSE director. For example, the co-simulation parameters associated with the thermal-structural model, *BODYFLUX* and *TEMP*, are identified as well as their respective units. The CSE director can only recognize base units, therefore units such as kcal/kg and degC are further defined in regards to their respective base unit  $\frac{m^2}{s^2}$  and K. For further details about the different implementation techniques refer to API User's kit [30].

```

<components>
  <component name="Abaqus/Standard">
    bottomUpImplementation
    <codeName>Abaqus/Standard</codeName>
    <modelDescription>
      <UnitDefinitions>
        <Unit name="kcal/kg">
          <BaseUnit factor="4186.8" m="2" s="-2"></BaseUnit>
        </Unit>
        <Unit name="degC">
          <BaseUnit K="1" factor="1.0" offset="273.15"></BaseUnit>
        </Unit>
      </UnitDefinitions>
      <ModelVariables>
        <ScalarVariable name="BODYFLUX">
          <Real unit="kcal/kg"></Real>
        </ScalarVariable>
        <ScalarVariable name="TEMP">
          <Real unit="degC"></Real>
        </ScalarVariable>
      </ModelVariables>
    </modelDescription>
  </bottomUpImplementation>
</component>
  <component name="CFD">
    bottomUpImplementation
    <codeName>Abaqus/Cfd</codeName>
  </bottomUpImplementation>
</component>
  <component name="BatteryCoSim_System">
    topDownImplementation
    <identifier>BatteryCoSim_System</identifier>
  </topDownImplementation>
</component>
</components>

```

Figure 4.19: Configuration file sections: Components

### 3) Component Instances

This section identifies the subsystem models participating in the co-simulation, highlighted by yellow, and associates these models with the simulation executables, highlighted purple, defined in the components section. Each co-simulation sub-component (Abaqus/Standard, CFD, and BatteryCoSim\_System), can be used to identify *num* independent component instances which allows rapid and scalable model development. The XML text in Figure 4.20, *Fluid*, *Standard*, and *BatteryCoSim\_System* instances are created using the *CFD*, *Abaqus/Standard*, and *BatteryCosim\_System* subcomponents respectively.

```
...
<componentInstances>
  <componentInstance name="Fluid">
    <component>CFD</component>
    <timeIncrementation>
      <lockstep>>false</lockstep>
    </timeIncrementation>
    <initialConditions>
      sendBeforeReceive false sendBeforeReceive
    </initialConditions>
  </componentInstance>
  <componentInstance name="Standard">
    <component>Abaqus/Standard</component>
    <timeIncrementation>
      <lockstep>>false</lockstep>
    </timeIncrementation>
  </componentInstance>
  <componentInstance name="BatteryCoSim_System">
    <component>BatteryCoSim_System</component>
  </componentInstance>
</componentInstances>
```

Figure 4.20: Configuration file sections: Component Instances

For the purpose of the co-simulation performed for this study, the two Abaqus simulation products each correspond to a single model (there is a one-to-one relation between the simulation executable and the simulation model). In a more general use, one-to-many

relationships can be defined; for example, an Abaqus/CFD to Abaqus/Standard pairing of two separate models.

Using the initial conditions option and setting the *sendBeforeReceive* option to false, the Fluid model will not communicate its initial conditions until it receives the initial conditions from the master model (Standard). The default mode for the *sendBeforeReceive* option is true and therefore it is not included the Standard component instant definition.

#### 4) Connectors

Connectors define the variables that enter or leave a component instance. The connector element provides the association between the component instance and the variable name and declares whether the variable is input or output. Text included in Figure 4.21 describes six connectors, highlighted in purple, and defines the model variables, highlighted in green, which establish communication between the co-simulation components. The Connector named *CFD\_to\_STD\_INPUT* defines the input variables to the Standard model instance while the connector named *CFD\_to\_STD\_OUTPUT* assigns the same variable as outputs from the Fluid model instance. Table 4.6 includes the complete list of connectors as defined in the co-simulation configuration file.

The name attribute for the connector element is used for internal reference in the configuration file. For Abaqus simulation models and FMUs, there is no need to match the variables listed in input files with corresponding variables in the configuration file.

```
...
<connectors>
  <connector name="CFD_to_STD_INPUT">
    <componentInstance>Standard</componentInstance>
    <variables>
      <input>
```

```

    <variable>heat_flux</variable>
    <variable>heat_capacitance</variable>
  </input>
</variables>
</connector>
<connector name="CFD_to_STD_OUTPUT">
  <componentInstance>Fluid</componentInstance>
  <variables>
    <output>
      <variable>heat_flux</variable>
      <variable>heat_capacitance</variable>
    </output>
  </variables>
</connector>
<connector name="STD_to_CFD_INPUT">
  <componentInstance>Fluid</componentInstance>
  <variables>
    <input>
      <variable>temperature</variable>
    </input>
  </variables>
</connector>
<connector name="STD_to_CFD_OUTPUT">
  <componentInstance>Standard</componentInstance>
  <variables>
    <output>
      <variable>temperature</variable>
    </output>
  </variables>
</connector>
<connector name="STD_from_FMI">
  <componentInstance>Standard</componentInstance>
  <variables>
    <output>
      <variable>Cell_Temperature</variable>
    </output>
    <input>
      <variable>Cooling_Plate</variable>
    </input>
  </variables>
</connector>
<connector name="FMI_from_STD">
  <componentInstance>BatteryCoSim_System</componentInstance>
  <variables>
    <input>
      <variable>cell_temperature</variable>
    </input>
    <output>
      <variable>cooling_plate</variable>
    </output>
  </variables>
</connector>
</connectors>

```

Figure 4.21: Configuration file sections: Connectors

Table 4.6: Definition of Connectors in co-simulation configuration file

Connector Name	Model Instance	Connector Role	Variable(s)
CFD_to_STD_INPUT	Standard	Input	heat_flux heat_capacitance
CFD_to_STD_OUTPUT	Fluid	Output	heat_flux heat_capacitance
STD_to_CFD_INPUT	Fluid	Input	temperature
STD_to_CFD_OUTPUT	Standard	Output	temperature
STD_from_FMI	Standard	Output	Cell_Temperature
		Input	Cooling_Plate
FMI_from_STD	BatteryCoSim_System	Input	cell_temperature
		Output	cooling_plate

#### 5) Connection Sets

Connection sets establish the associations between Connectors. As illustrated in Figure 4.22, pairing between all input and output variables is checked by the CSE director. In cases where connector elements have multiple variables defined, the connection elements must pair two connector elements that exactly complement each other in the sense that one connector's input variable count match the other's output variable count, and vice versa.

In the case of co-simulation performed for this analysis, three independent connection sets are defined. The first two connection sets, of type field, establish communication between the fluid and standard model. The first set defines the communication from the CFD to the Standard model, and the second set defines communication from the Standard to the CFD model. Type



field connection ports will communicate both the variable magnitude as well as field units. The third connection set establishes communication of signals between the Standard model and the FMU.

```

...
<connectionSets>
  <connectionSet name="CFD_to_STD" type="FIELD">
    <connection>
      <connector>CFD_to_STD_INPUT</connector>
      <connector>CFD_to_STD_OUTPUT</connector>
    </connection>
  </connectionSet>
  <connectionSet name="STD_to_CFD" type="FIELD">
    <connection>
      <connector>STD_to_CFD_INPUT</connector>
      <connector>STD_to_CFD_OUTPUT</connector>
    </connection>
  </connectionSet>
  <connectionSet name="STD_and_FMI" type="SIGNAL">
    <connection>
      <connector>STD_from_FMI</connector>
      <connector>FMI_from_STD</connector>
    </connection>
  </connectionSet>
</connectionSets>

```

Figure 4.22: Configuration file sections: Connection sets

## 6) Execution

The execution element describes all the details of the numerical method used to perform the co-simulation. As shown in Figure 4.23, execution is generalized to enable a hierarchical arrangement of an arbitrary number of component instances. The term *atomicActor*, highlighted in green, refers to each individual simulation participant in co-simulation; whereas, a *compositeactor*, highlighted in yellow, refers to a group of simulation participants. Model instance(s) defined within a single *compositeactor* will communicate using the assigned solver, highlighted in purple, and negotiation method, highlighted in gray.

In the case of the co-simulation performed for this analysis, the CSE uses the Gauss-Seidel algorithm and the CSE will negotiate the variable exchange time between participants by selecting the maximum of the time increments preferred by the three codes. Two connection categories, highlighted in blue, are defined, *InitialConditions* and *CouplingStep*. Each category defines the connection sets participating in the respective connection category. For more information about the algorithm types and exchange time negotiation options, please refer to the CSE configuration schema.

```

...
<execution>
  <compositeActors>
    <compositeActor name="twoCodeContinuousTime">
      <actors>
        <atomicActor>Standard</atomicActor>
        <atomicActor>Fluid</atomicActor>
        <atomicActor>BatteryCoSim_System</atomicActor>
      </actors>
      <modelOfComputation>
        <continuousTime>
          <algorithm>GAUSS-SEIDEL</algorithm>
          <negotiationMethod>MAX</negotiationMethod>
        </continuousTime>
      </modelOfComputation>
    </compositeActor>
  </compositeActors>
  <connectionGroups>
    <connectionCategory name="InitialConditions">
      <connectionSet>CFD_to_STD</connectionSet>
      <connectionSet>STD_to_CFD</connectionSet>
      <connectionSet>STD_and_FMI</connectionSet>
    </connectionCategory>
    <connectionCategory name="CouplingStep">
      <connectionSet>CFD_to_STD</connectionSet>
      <connectionSet>STD_to_CFD</connectionSet>
      <connectionSet>STD_and_FMI</connectionSet>
    </connectionCategory>
  </connectionGroups>
  <scenario>
    <duration>300.</duration>
  </scenario>
</execution>

```

Figure 4.23: Configuration file sections: Execution

Finally, in order to perform the co-simulation, all input files (Abaqus input files and the FMU) as well as the co-simulation configuration file are stored in the current directory where the the Abaqus Command is issued. The commands, listed in Figure 4.24, will initiate the co-simulation on machine earth and communication with the CSE director is established through port 65533. For a complete list of available ports for co-simulation refer to the Abaqus User's Guide [28].

```
call abaqus cse -j Configure_File -config Configure_File.xml -listener 65533
call abaqus fmu -fmu Electrical_System.fmu -instance Electrical_System -csedirector earth:65533
call abaqus -j Standard -csedirector earth:65533
call abaqus -j Fluid -csedirector earth:65533
```

Figure 4.24: Co-simulation execution commands

After the first command prompt is used, the CSE director waits for the other co-simulation components, which can be invoked in an arbitrary order. During co-simulation, the CSE director generates individual log files for each co-simulation file, which can be used for debugging. A separate .msg file is also generated which provides a brief description of the co-simulation steps and possible errors encountered during co-simulation.

## 4.2 Calibration

The calibration of the battery model will be completed using experimental data in three phases: 1) Cell model calibration, 2) Battery module calibration, and 3) Battery pack calibration. The calibration of the cell model focuses on the calibration of the electro-thermal domain model. Cell parameters such internal resistance, OCV, and capacity are provided by the manufacturer. However, parameters such as cell thermal capacity, cell skin thermal conductance, temperature dependency of parameters such as internal resistance and capacity, and RC components are calibrated using experimental test results. Calibrating the battery module and battery pack

models require further testing of the battery module and battery pack, which will be completed in the next phase of the research partnership.

In order to calibrate the cell model, a single battery cell was tested under high current charge and discharge conditions and experimental results were provided. For the purposes of the discharge test, the ambient temperature was at  $23 \pm 2^{\circ}\text{C}$ . The cell was initially charged fully. Afterwards, the sample was paused for 1 hour. The test cell was placed under a thermal camera and the device under testing (DUT) was discharged with a current of 40 A until the voltage dropped below the minimum cell voltage, or until the thermal camera detected a temperature of over  $80^{\circ}\text{C}$ . For the purpose of the calibration, the distribution of temperature was constant over the surface.

During discharge the cell is discharged with a current of 40A and after 210 seconds the cell reaches  $80^{\circ}\text{C}$ , at which point the discharge is stopped. In order to calibrate the model, system parameters were adjusted until good agreement between the cell temperature experimental results and the model output temperature was reached. Figure 4.25 illustrates the cell skin temperature as recorded by the thermal camera and the cell model temperature.

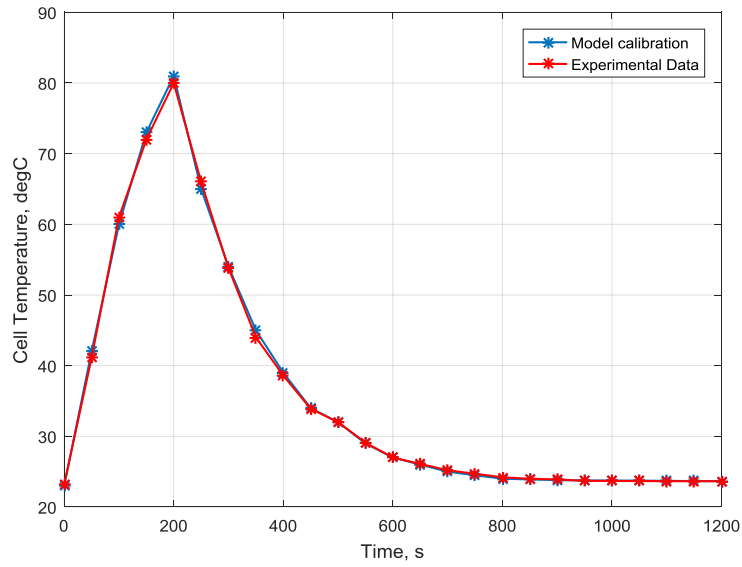


Figure 4.25: Cell calibration, discharge of fully charged cell

In order to test the effectiveness of the model calibration, the calibrated model was simulated at a discharge rate of 40 A. Figure 4.26 illustrates good agreement between the model predicted and experimental normalized cell voltage. During the cell discharge (first 210 seconds), the model predicted voltage drops at a slower pace compared to the experimental results. The cell electro-thermal domain model dynamic in fast discharge is governed by the cell capacity and internal resistance, both of which are defined as functions of temperature and calendric aging. The time constant of the cell thermal domain model (governed by the thermal capacity and conductance) is much larger compared to the electro-thermal domain. Therefore the slow change in temperature, and in turn cell internal resistance, results in the discrepancy between the model prediction and experimental results. After the discharge is completed, both experimental results and model-predicted results illustrate a fast voltage recovery (from 210 to 400 seconds). However, the model prediction fails to capture the slow continued voltage recovery beyond 400 seconds. Redefining the cell internal resistance and capacity as functions of

current, temperature, and calendric aging may reduce the discrepancy between model prediction and experimental results.

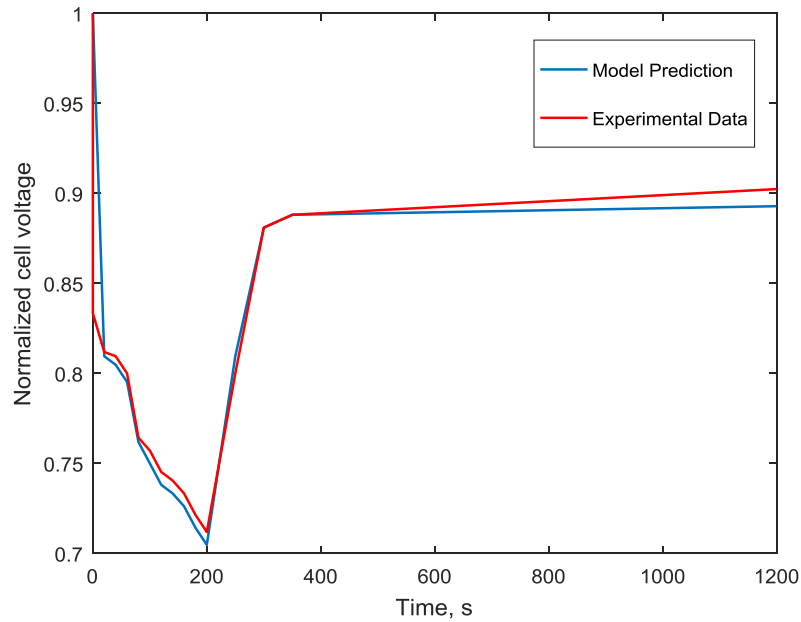


Figure 4.26: Verification of model calibration, normalized cell voltage - Discharge

Similar to the discharge test, during the charge test, the ambient temperature was at  $23\pm 2^{\circ}\text{C}$  and the cell was discharged with a current of 2.6 A until the cell voltage dropped below the minimum cell voltage. After a 1 hour pause period the cell was charged with a maximum current of 30A for one hour or until the thermal camera detected a temperature of  $80^{\circ}\text{C}$ .

In order to calibrate the model, system parameters were adjusted until good agreement between the cell temperature experimental results and the model output temperature was reached. Figure 4.27 illustrates the cell skin temperature as recorded by the thermal camera and the cell model temperature.

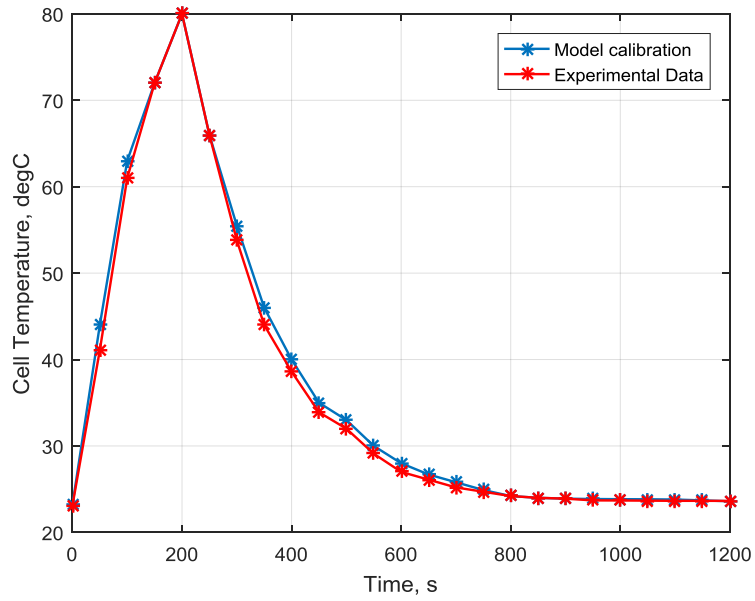


Figure 4.27: Cell calibration, Charging

Figures 4.27 and 4.28 illustrate a comparison between the model predicted and experimental current and normalized cell voltage respectively.

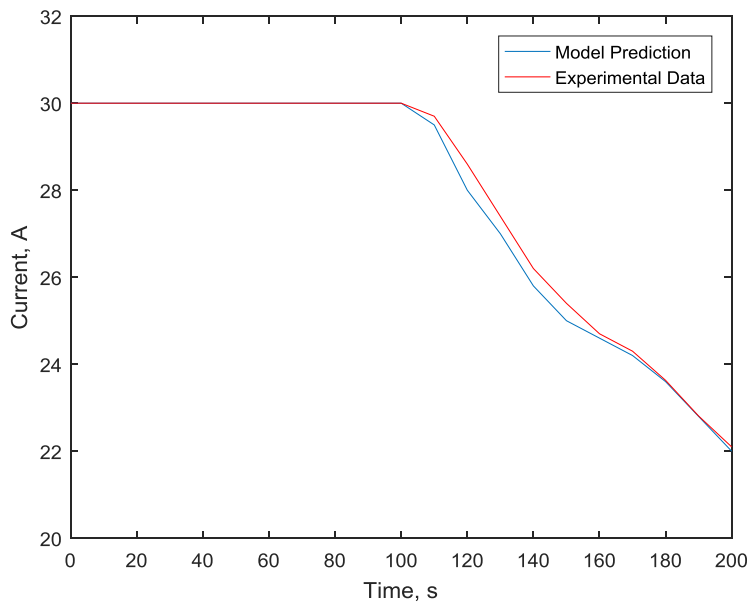


Figure 4.28: Cell current during charging

During the first 100 seconds of cell charging, both experimental and model-predicted discharge current is at the maximum allowed limit of 30 A. After the first 100 seconds, the charge current decreases rapidly until at 200 seconds the cell temperature reaches 80°C. Figures 4.28 and 4.29 illustrate good agreement between the experimental results and model prediction.

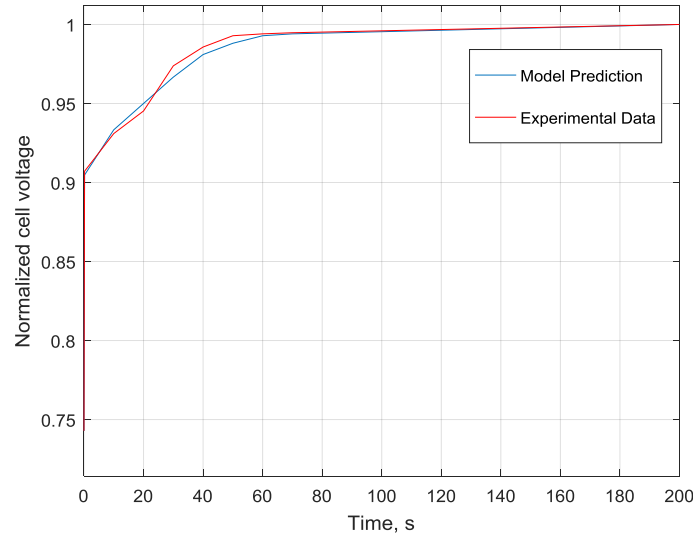


Figure 4.29: Cell normalized voltage during charging

After the first 100 seconds the cell model predicts a more rapid decline in charge current than experimental results. Similar to the discharge test, this discrepancy may be corrected by redefining the cell internal resistance and capacity as a function of charge current.

### 4.3 Battery module co-simulation

A battery module, following the model development and co-simulation execution procedure outlined in section 4.1, and using the calibrated cell parameters as described in 4.2, was created. Results presented in this section display the capability to identify hot spots and predict the system dynamics; however, the module has not yet been calibrated and thus predicted results are only of qualitative value. System parameters such as fluid film coefficient



and thermal contact conductance coefficient were extracted from reference text [32]. Calibration of the battery module model using experimental data will be completed in the next phase of the research project.

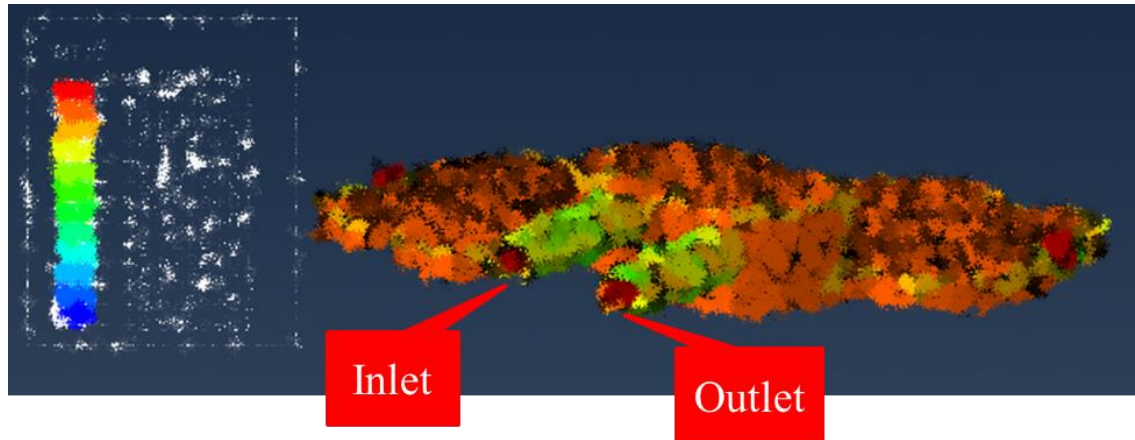


Figure 4.30: Cooling plate temperature distribution. Temperature contour displays areas of high temperature in red and low temperature in blue \*\* *Image obscured for proprietary reasons.*

Figure 4.30 illustrates the temperature distribution on the cooling plate. The initial module temperature was set at  $30\text{ }C^{\circ}$ . The cooling fluid temperature at the inlet was fixed at  $15\text{ }C^{\circ}$  and the fluid temperature at the outlet is presented in Figure 4.31.

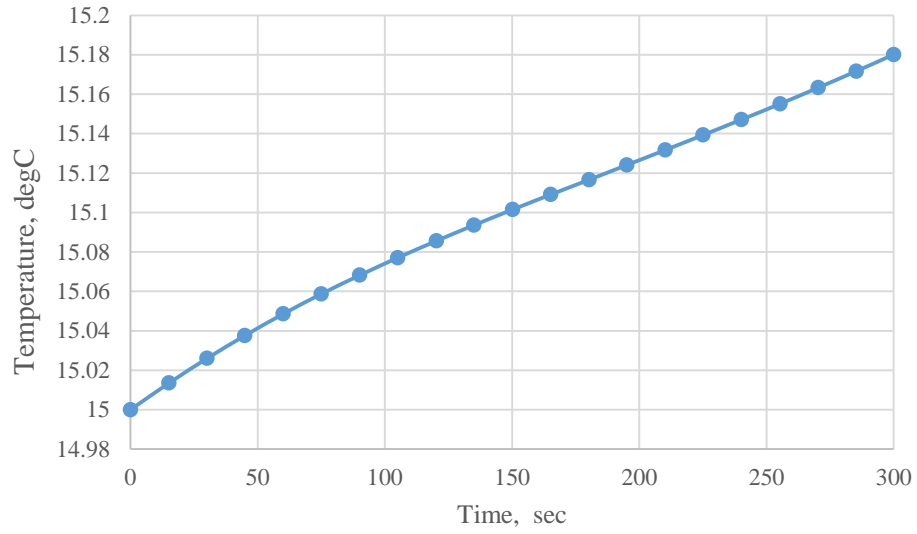


Figure 4.31: Coolant outflow temperature.

Similar to the cooling plate, hot spots can be identified on the cells as shown in Figure 4.32. The minimum cell temperature occurs at the contact points with the cooling plate at the top and bottom of each cell, while the heat loss through convection reduces the cell skin temperature on all sides. Maximum temperature regions, marked in red, occur at the top and bottom of the cell due to reduced heat transfer with the cooling plate and minimal heat loss through convection.

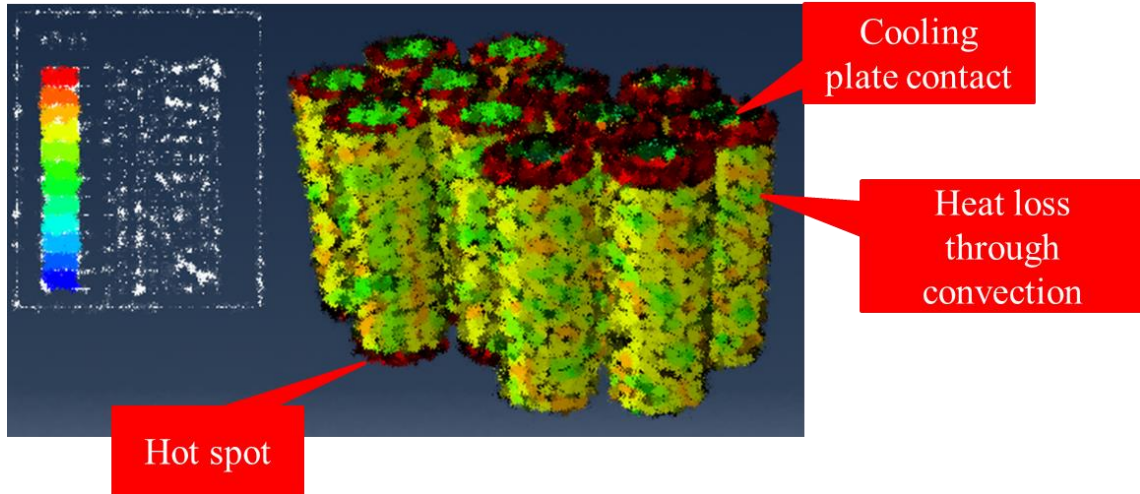


Figure 4.32: Cell temperature distribution. Temperature contour displays areas of high temperature in red and low temperature in blue. *\*\* Image obscured for proprietary reasons.*

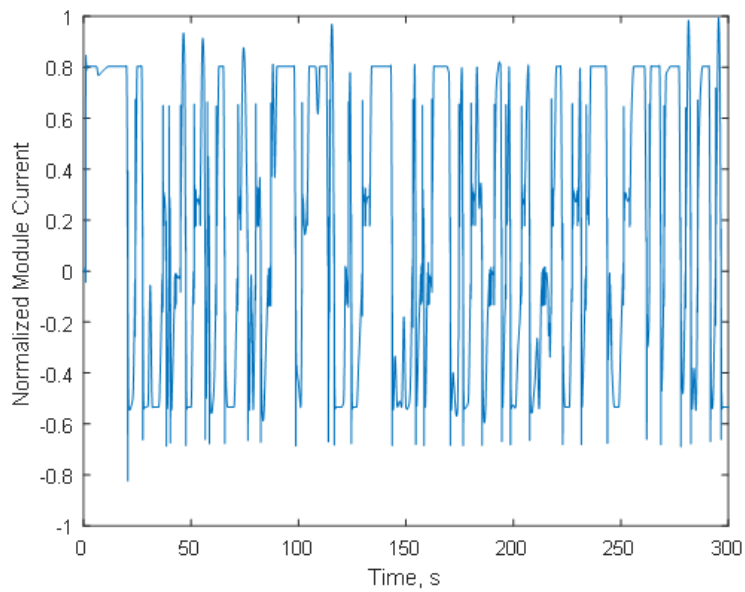


Figure 4.33: Normalized module current

In addition to the temperature distribution in the thermal-structural model, the electro-thermal model output is also calculated. Figure 4.33 displays the normalized module current

operating under a high-frequency-high-load power cycle. Similarly, Figure 4.34 illustrates the module voltage during the co-simulation.

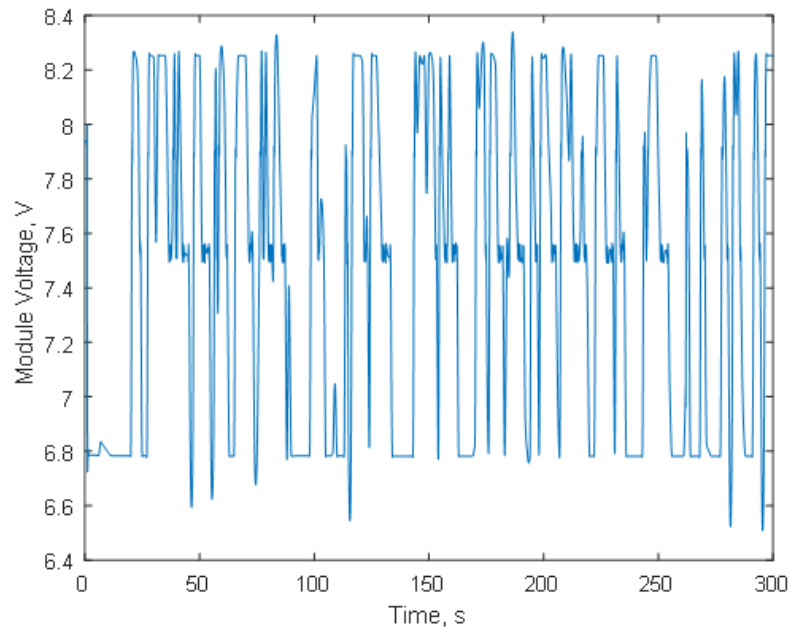


Figure 4.34: Module voltage under high-frequency-high-load conditions

#### 4.4 Performance under high loading conditions

Performance of the battery pack under high loading conditions is of great concern in the development of high-performance HEVs. The current draw from the battery pack can be limited; however, extended high current draw can result in development of local high temperature regions which can significantly reduce the battery performance and battery life. The purpose of this study is to estimate the effects of high current draw on battery life and ensure the cooling system can efficiently remove heat and eliminate high temperature regions.

For the purpose of this study the model will undergo two types of loading conditions. In order to study the long term effects on battery life, the model will be executed with loading

cycles that represent the long term use of the battery pack. In order to study the temperature profile in the battery pack, the loading cycle will feature long periods with maximum current draw. In order to complete this analysis, which will be completed in the next phase of research, a model of the battery pack to include hundreds of battery modules will be developed. This model will also include control logic that will monitor the temperature of the battery pack at predetermined locations and limit the current draw to prevent thermal or electrical damage to the battery pack.

The development of the battery pack model will also require further testing of an individual battery cell to better define the fast and slow transient behavior of the cell. The transient behavior is represented by several RC elements included in the cell model.

#### **4.5 Battery life estimation**

In order to study the battery life and characterize the battery aging, three battery parameters are studied. Battery current draw/ charging current, temperature and SOC significantly influence the battery life. A 2.5 year study documented in [32] has shown that cell temperature, depth-of-discharge, and current rates are the primary factors influencing battery state of health and its useful life.

The cell model included in the EES library has built-in functionality to estimate the battery life. The capacity, as well as the internal impedance of the cell, can change due to aging. The effects of aging on the battery state of health are identified by the battery state of health (SOH).

$$SOH = SOH_c \times SOH_z \quad (42)$$

For each cell,  $SOH_C$  and  $SOH_Z$  are estimated by calibrating aging factors,  $x_c$  and  $x_z$ , respectively [27]. Characterizing the battery life and the effects of high current draw and charging will be investigated in the next steps of the study.

$$SOH_C = \frac{1}{C_0(1 - x_c)} \times C - \frac{x_c}{1 - x_c} \quad (43)$$

$$SOH_Z = \frac{1}{Z_0(1 - x_z)} \times Z - \frac{x_z}{1 - x_z} \quad (44)$$

#### 4.6 High and low ambient temperature testing

Vehicle manufactures are required to ensure the vehicle can perform in both high and low ambient temperature conditions. A significant area of interest is performance in high ambient temperatures as the high ambient temperature can reduce the ability of the cooling system to remove heat from the battery pack, and additional cooling may be required in order to prevent long term damage to the battery pack.

Each battery cell model is temperature dependent and therefore the effects of ambient temperature will be investigated in later stages of the analysis. In order to ensure the safe operation of the battery pack, it is highly encouraged that each battery pack design is tested before final integration in the vehicle.

#### 4.7 Vehicle integration

The final stage of this project will include the integration of the battery pack model into the vehicle model. The battery pack model developed in this chapter is an example of a high fidelity multi-physics model which can take supersede models of lower fidelity. The development of high fidelity models within a single modeling environment, such as Matlab or

Simulink, is often unfeasible, or at best can lead to a significant increase in the computational costs as many functional components of the vehicle feature multi-physics systems. It should also be noted that the development of finite element model of system components in junction with equation-based models and rule-based control is often impossible within a single modeling environment. Therefore the use of co-simulation in the development of subcomponent models and integration with the vehicle model can lead to a significant increase in model fidelity and reduced computational cost. It is highly encouraged that the battery pack model developed as part of this analysis be integrated with the vehicle model after model verification and development of the complete vehicle model.

## **Chapter 5: Concluding Remarks and Future Work**

This thesis presented simulation tools, control analysis, and a multi-physics co-simulation approach supporting high-performance HEVs. It focused on the development of vehicle simulation and control of HEVs using a sequential approach, as well as a multi-physics modeling approach to develop high fidelity battery models. This chapter will provide concluding remarks for the thesis and suggest future work.

### **5.1 Adaptation of Vehicle Model to Simulate and Control Hybrid Electric Vehicles**

A sequential modeling approach was introduced in Chapter 2 in order to develop a vehicle model for a series HEV. The same modeling approach can be adapted to develop vehicle supervisory control strategies for parallel and power-split HEVs.

The supervisory control strategy applied to the vehicle model in Chapter 2 was a rule-based strategy which provides a good compromise between vehicle performance and efficiency for series hybrid architectures. Parallel and power-split hybrid architectures, however, require the application of more complex supervisory control strategies such as Equivalent Cost Minimization Strategy (ECMS), which assigns an equivalent fuel for battery power. Application of ECMS can be further enhanced by utilizing DP results to choose and update the equivalence factor.

In Chapter 3, DP and its application to a series hybrid architecture was introduced. DP is a Backward-Looking simulation which can be used as an unbiased tool to evaluate the controller performance. Applying DP to hybrid architectures and comparing the vehicle emissions and performance against the vehicle model can provide further insight in the development of the vehicle supervisory controller.



In both Chapters 2 and 3 the controller performance was evaluated based on fuel economy and CO<sub>2</sub> emissions. While the fuel economy and CO<sub>2</sub> emissions are of great importance in the design of vehicles, manufacturers require each new vehicle design to meet a long list of design goals, such as emission, drivability, and performance. One area for future work is to develop a robust supervisory control that can simultaneously minimize fuel economy in normal city driving conditions and maximize vehicle performance. Another possible improvement is to investigate the use of traffic light and congestion data, to interactively influence driver behavior and improve the fuel economy.

## **5.2 Battery Modeling and Multi-Physics Co-simulation**

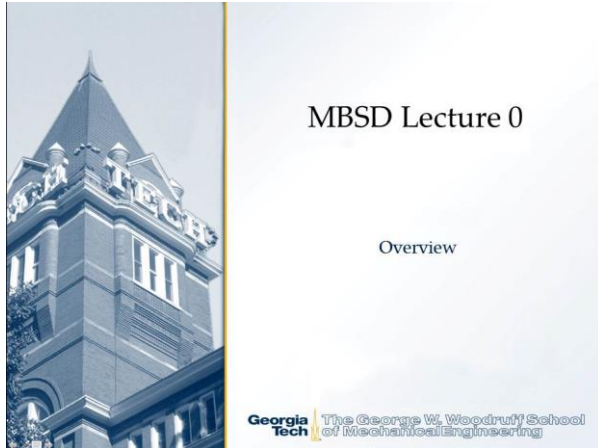
In Chapter 4 a multi-physics co-simulation approach was used to develop a high fidelity battery model in order to investigate the temperature distribution in the spatial framework of the battery pack. The co-simulation approach provides a fast computational time by incorporating a discrete electro-thermal domain model developed using Modelica, an open-source, object-oriented, and equation-based modeling language. The spatial context of the heat transfer problem was captured by the Finite Element Model (FEM) of battery components including a transient thermal-structural heat transfer model and a fluid heat transfer model of the cooling system.

Communication between the three model components was established following the Functional Mock-up Interface (FMI), a tool-independent standard to support model exchange and co-simulation of dynamic models. FMI allowed for the coupling of the discrete electro-thermal domain model and the FEM heat transfer models to develop a modular coupled model. Data exchange is established at discrete communication points, as specified by the host simulation tool, and each subsystem is solved independently between the communication points.

The battery model developed will be used in future studies to investigate the effects of high current charge/discharge in high performance applications to ensure safe operation of the battery pack. The model will also be used to design the cooling system and study the effects of high current on the battery life. Another area for future work is to incorporate the battery model into a vehicle model. Simulink software used to develop the vehicle model in Chapter 2 is an FMI compliant software which allows for the integration of the battery pack model. The high fidelity multi-physics models can replace the battery model developed using experimental results. In addition to improved fidelity, the model can be used to identify the areas of high temperature, design the cooling system to allow for proper cooling, investigate the effects of current draw on life, and in turn observe the effects of battery inefficacy on vehicle performance.

# Appendix:

## 1) MBSD Tutorial

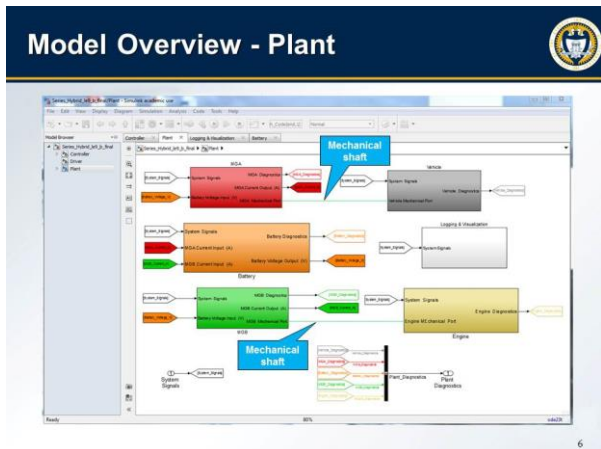
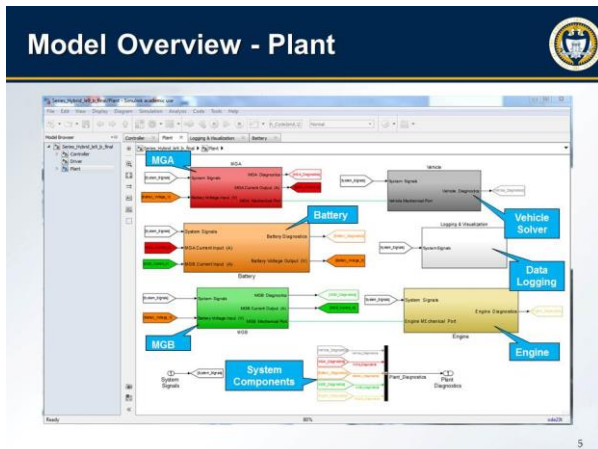
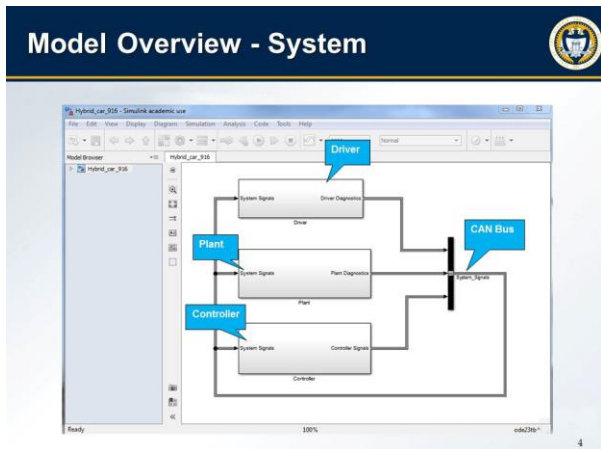


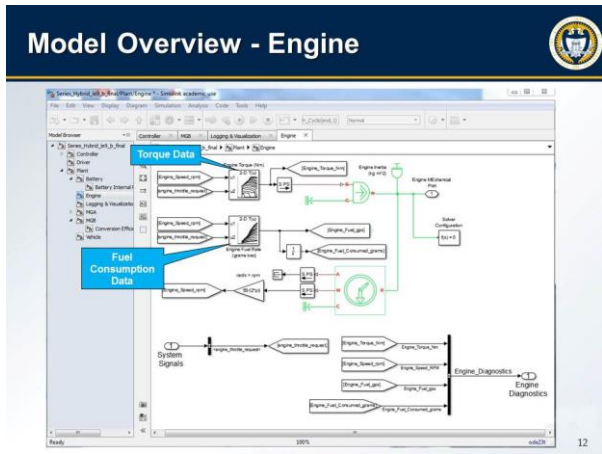
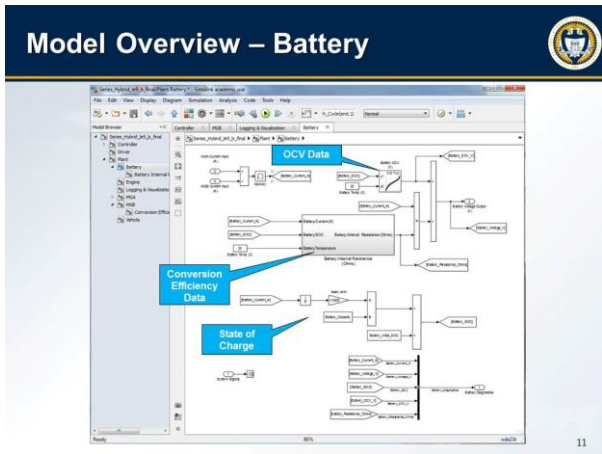
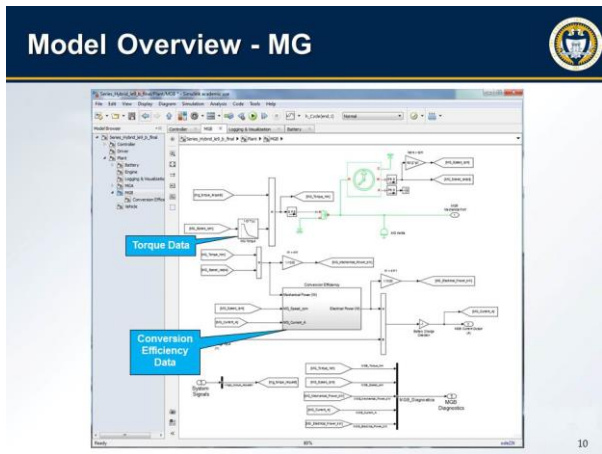
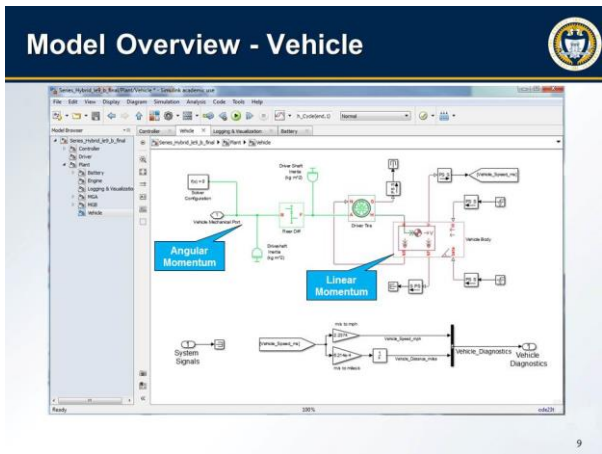
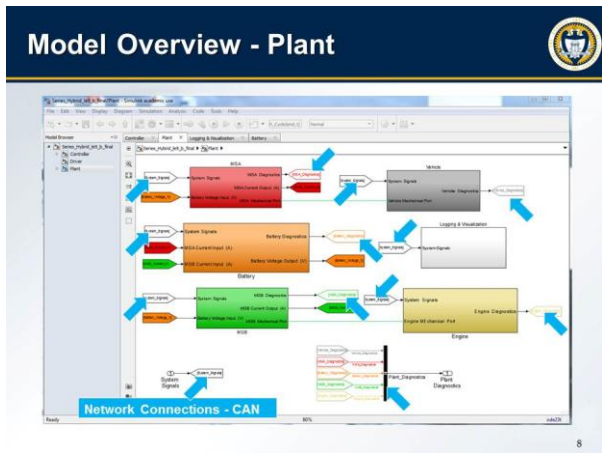
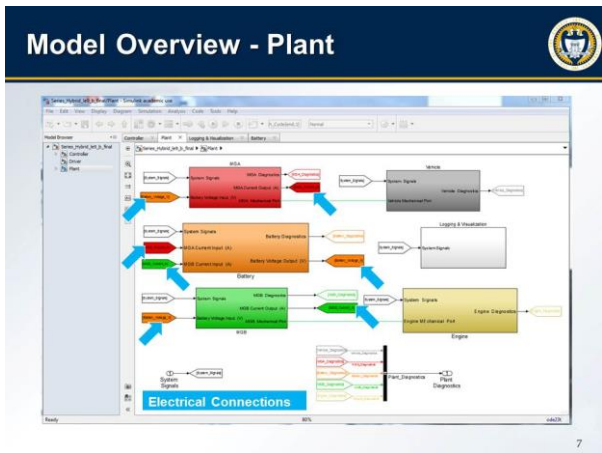
### Course Goals

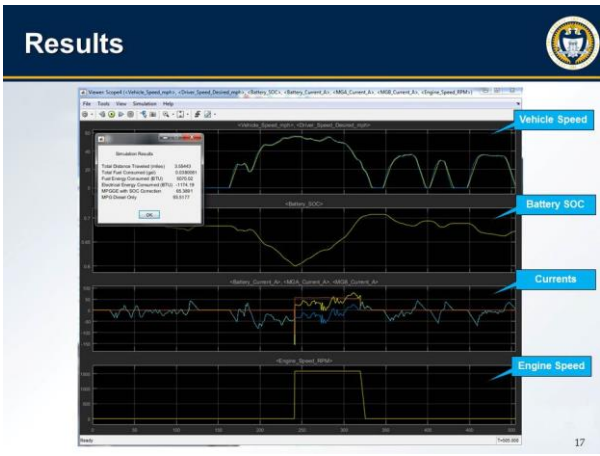
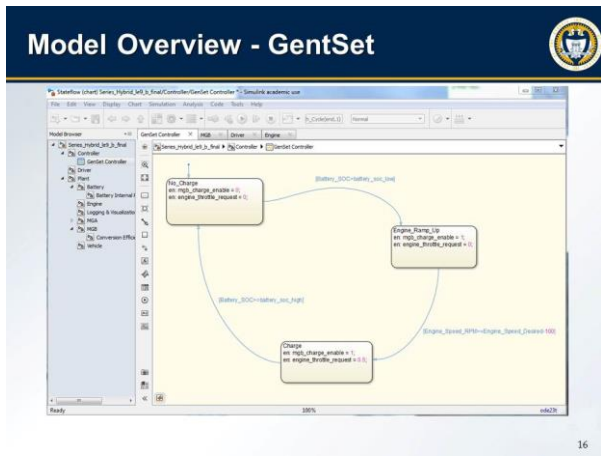
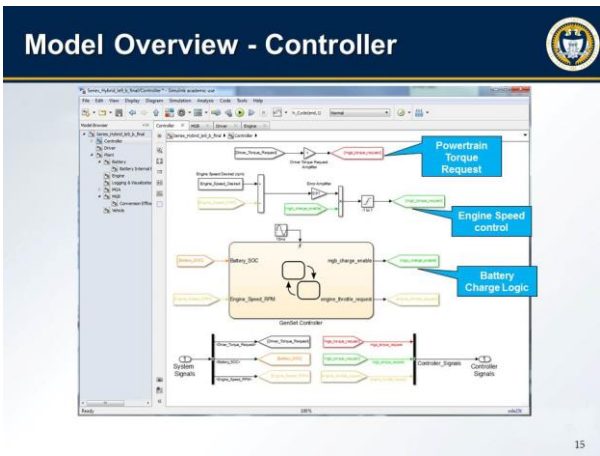
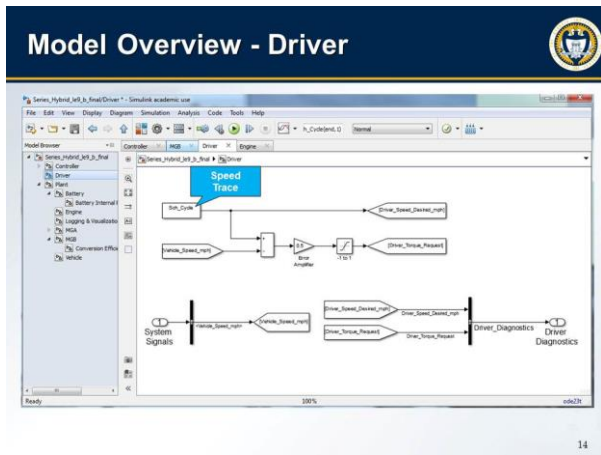
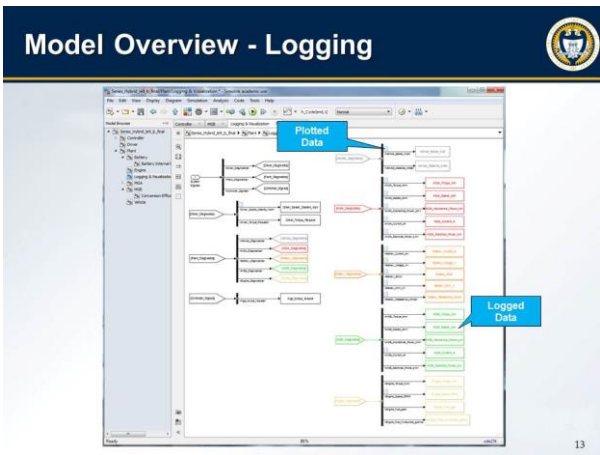
- Present and emphasize MBSD approach
  - Simple models
  - Complex system
  - Basic control
  - Iterative refinement
- Utilize simulation tools
  - Matlab
  - Simulink
  - Stateflow

### Course Outline

- Simple Series Electric Vehicle
  - Model hierarchy
  - Ideal Component Models
  - Controls
- Improved Component Models
  - Inefficiencies
  - Experimental Data
- Improved Controls







## MBSD Lecture 1

Getting Started with the High Level System

Georgia Tech The George W. Woodruff School of Mechanical Engineering

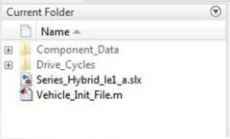
## Lecture Goals

- Getting Started with Simulink
- Developing model hierarchy (subsystems)
- Establishing CAN bus

19

## Getting Started

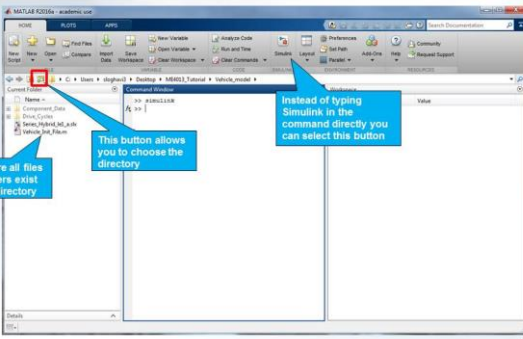
- Start Matlab
- Make sure you are in the correct directory



- Type **simulink** in the command window

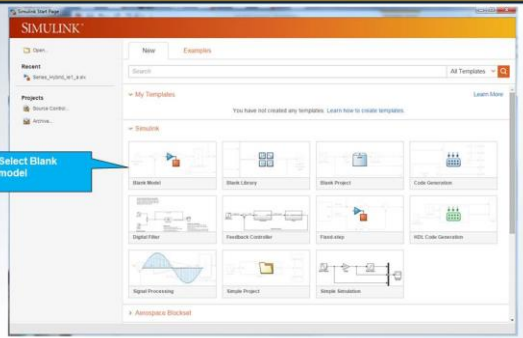
20

## Getting Started



21

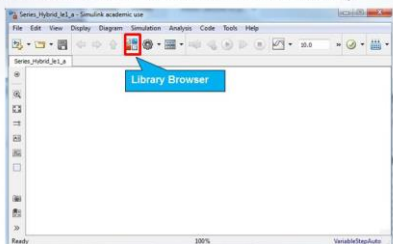
## Getting Started



22

## Getting Started

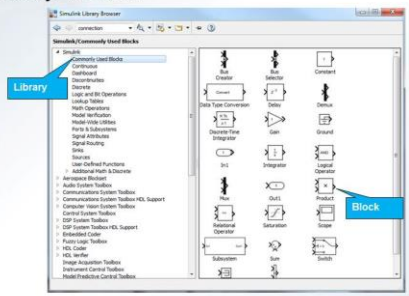
- Simulink is a graphical programming environment which enables you to link function blocks to create simulations
- In the Simulink model window, select library browser



23

## Getting Started

- Get yourself familiar with the libraries and blocks in the library browser



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## Getting Started



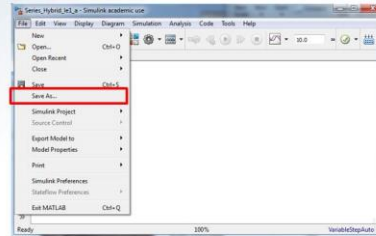
- We will primarily use the following libraries
  - Simulink \ Commonly Used Blocks
  - Simulink \ Signal Routing
  - Simulink \ Lookup Tables
  - Simulink \ Sinks
  - Simulink \ Sources
  - Simscape \ Driveline
  - Stateflow

25

## Getting Started



- Save the Simulink model as :
  - Series\_Hybrid\_1e1\_a
- Do not use spaces or variable names in file names



26

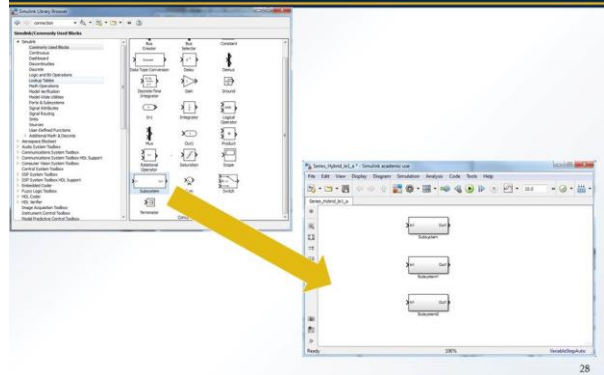
## High Level Model



- We will first create the high level plant and controller model
- From
  - Simulink \ Commonly Used Blocks
- Drag three **Subsystem** blocks into the model
  - Driver block
  - Controller block
  - Vehicle block

27

## High Level Model



28

## High Level Model



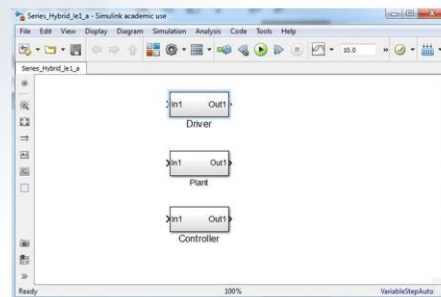
- Rename the upper subsystem
  - Driver
- The middle subsystem
  - Plant
- The lower subsystem
  - Controller

29

## High Level Model



- You can change the font size by right clicking on each subsystem/Format/Font Style...



30

## High Level Model



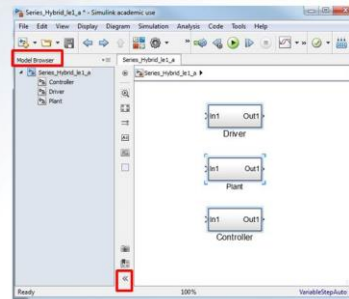
- At this high level, signals will be sent out by the Driver, Plant, and Controller
  - Driver Torque Request
  - Battery State of Charge
  - Engine Throttle Request
- These signals will be put onto a common bus so that all three subsystems can receive all signals
  - Only desired signals will be selected

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## High Level Model



- Open Model Browser by selecting the arrow
  - Shows an outline form of the model



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## Driver Subsystem



- The Driver receives information
  - Speedometer
  - Tachometer
  - Idiot lights (engine hot, .....
- The Driver sends information
  - Torque request (gas/brake pedal position)
  - Gear selection

33

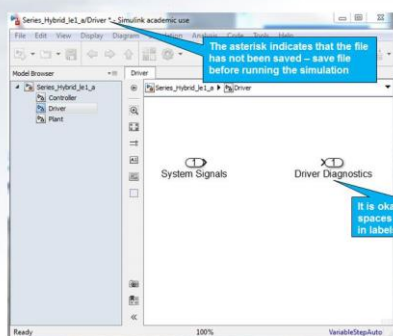
## Driver Subsystem



- Open the Driver subsystem by:
  - clicking on Driver in Model Browser or
  - double click on the Driver block
- Delete the wire that connects the In1 block to out1 block
- Rename
  - In1 to System Signals
  - Out1 to Driver Diagnostics

34

## Driver Subsystem



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## Driver Subsystem - CAN

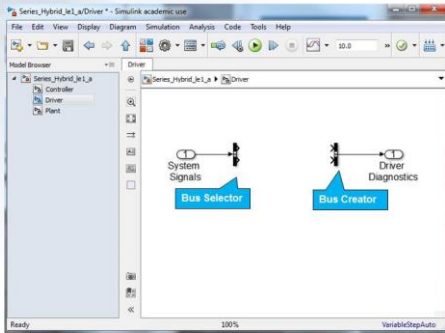


- From
  - Simulink \ Commonly Used Blocks
- Drag in a Bus Selector and a Bus Creator
  - The Bus Creator puts signals on the bus
  - The Bus Selector extracts them from the bus
- Connect the System Signals input to the Bus Selector
- Connect the Driver Signals input to the Bus Creator

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## Driver Subsystem - CAN



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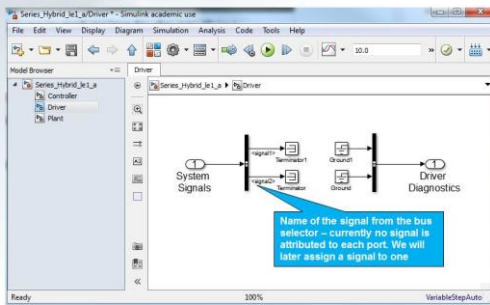
## Driver Subsystem - CAN



- For now, the Driver is not going to do anything
- From
  - Simulink \ Commonly Used Blocks
- Drag in two **Terminators** and two **Grounds**
- Connect the two outputs of the Bus Selector to the **Terminators**
- Connect the two inputs of the Bus Selector to the **Grounds**

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## Driver Subsystem - CAN



39

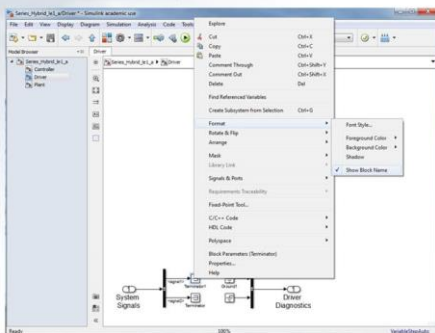
## Driver Subsystem - CAN



- The symbols for the ground and terminator are pretty obvious – there is no need for them to be labeled
- On one of the **Terminators**
  - Right Click
  - Select Format
  - Deselect Show Block Name
- We will do this for many of the blocks
- Hide the name of all terminator and ground blocks

40

## Driver Subsystem - CAN



41

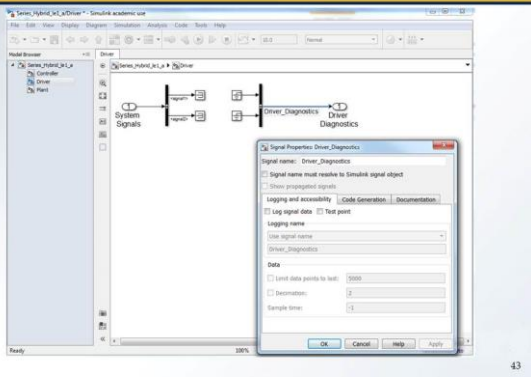
## Driver Subsystem - CAN



- While no signals are being sent by the Driver, the Driver Diagnostics signal should be identified on the bus
- On the wire connecting the **Bus Creator** to the Driver Diagnostics output
  - Right Click
  - Select Properties
  - For the Signal name enter **Driver\_Diagnostics**
  - Click OK

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## Driver Subsystem - CAN



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## Driver Subsystem - CAN



- Move the blocks around, enlarge if necessary, and make things look nice

44

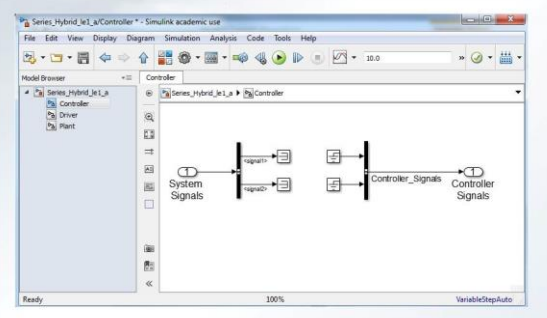
## Controller Subsystem



- For now, the controller is not going to do anything
- Repeat everything done to the Driver subsystem for the Controller subsystem
  - Rename the Out1 to Controller Signals
  - Rename the CAN signal to Controller\_Signals

45

## Controller Subsystem



46

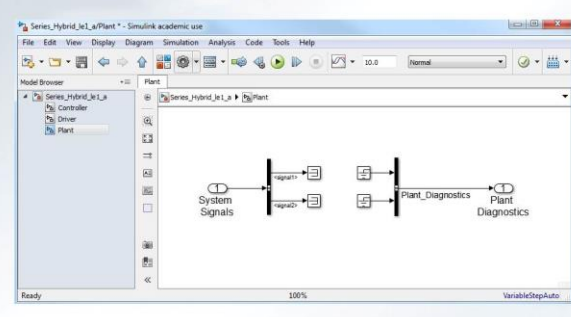
## Plant Subsystem



- For now, the Plant is not going to do anything
- Repeat everything done to the Controller subsystem for the Plant subsystem
  - Rename the Out1 to Plant Diagnostics
  - Rename the CAN signal to Plant\_Diagnostics

47

## Plant Subsystem



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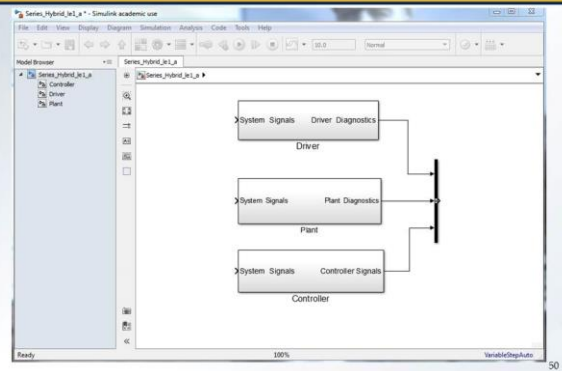
## High Level



- Return to the top level of the model
- Increase the size of the subsystems to make them legible
- Drag in a **Bus Creator**
  - Double click of the Bus Creator
  - Change the number of inputs to 3
  - Click OK
- Connect each subsystem to the **Bus Creator**
- If you Double click on the **Bus Creator** after connecting the subsystem you can view the input signals

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## High Level



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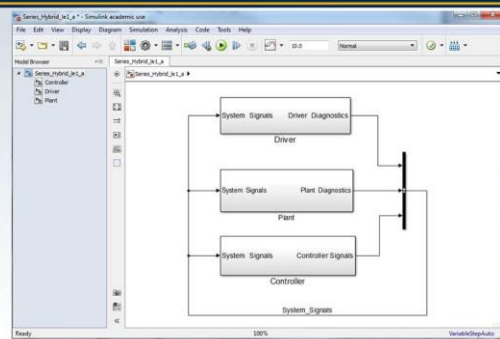
## High Level



- Connect the output of the Bus Creator to each of the subsystem inputs
- Rename the output signal of the Bus Creator to System\_Signals

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## High Level



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## High Level



- Congratulations! You have
  - Created a high level hierarchical system
  - Established communication between the three key subsystems
  - Gained experience with Simulink

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## MBSD Lecture 2

Simple Vehicle Model

Georgia Tech The George W. Woodruff School of Mechanical Engineering

## Lecture Goals

- Create the Vehicle subsystem
- Create the MGA subsystem with a preposterous motor model
- Observe and assess the results

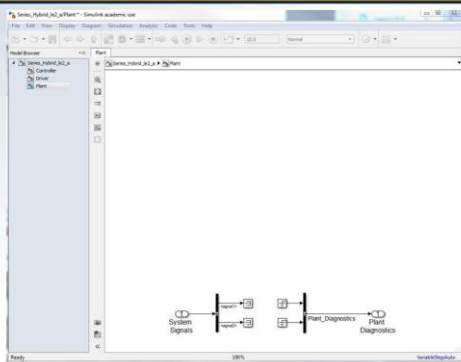
55

## Simple Vehicle

- Save your previous model (Series\_Hybrid\_le1\_a.slx) as Series\_Hybrid\_le2\_a
- In the Model Browser window, click on the Plant subsystem
- Move the buses to the bottom of the window

56

## Simple Vehicle



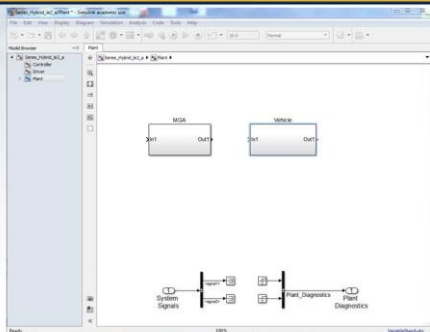
57

## Simple Vehicle

- Drag two **Subsystems** into the Plant window
- Rename them
  - Vehicle
  - MGA
- Right click on the Vehicle subsystem
  - Select Rotate & Flip
  - Flip Block Name
- Repeat for the MGA subsystem

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## Simple Vehicle



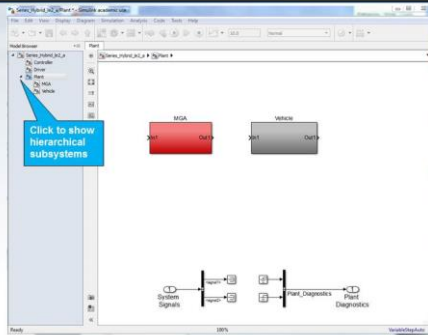
59

## Simple Vehicle

- Let's also color code our subsystems to make things easy to identify
  - Right click on the subsystem
  - Format/Background Color
- Make the vehicle gray and MGA red

60

## Simple Vehicle



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## Vehicle Subsystem

- The Vehicle subsystem will mechanically receive torque via a shaft from MGA
- We will skip mechanical brakes and use exclusively regenerative braking as applied by MGA
- The Vehicle subsystem will contain
  - Posi-trac rear axle
  - Rear tires
  - Vehicle solver
  - Inertias for the driveshaft and axle half shafts

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## Vehicle Subsystem

- Delete the **In1** and **Out1** blocks as well as the connecting wire
- Copy the **In1**, **Out1**, **Bus**, **Terminator**, and **Ground** blocks from the Plant subsystem
- Rename Plant to Vehicle
- From
  - Simscape / Utilities
- Drag in a **Connection Port** and rename it Vehicle Mechanical Port

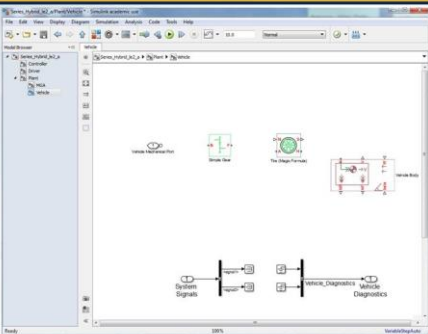
63

## Vehicle Subsystem

- From
  - Simscape / Driveline / Tires & Vehicles
- Drag in the **Vehicle Body** and one **Tire(Magic Formula)** blocks
- From
  - Simscape / Driveline / Gears
- Drag in a **Simple Gear** block

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## Vehicle Subsystem



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## Vehicle Subsystem

- Double click on the **Simple Gear**
- We can specify the gear ratio and relative rotational directions
  - Change the output shaft rotation to the same direction as input shaft
  - Leave the ratio as 2

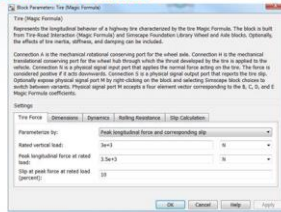


66

## Vehicle Subsystem



- Double click on the **Tire** block
  - This block converts the applied torque to force in the x-direction
  - It also determines whether or not the tire slips
  - Keep the default values

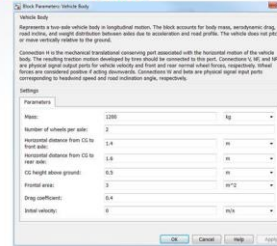


67

## Vehicle Subsystem



- Double click on the **Vehicle Body** block
  - This block solves linear momentum (with drag and pitch) from the tire force
  - Keep the default values



68

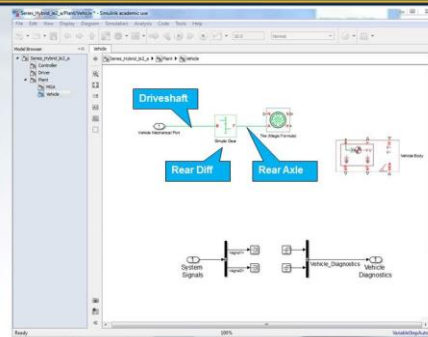
## Vehicle Subsystem



- Connect the Vehicle Mechanical Port to the Base side of the **Simple Gear**
- Connect the Follower side of the **Simple Gear** to the A port on the **Tire**
- We now have a model of the driveshaft, rear differential, and axle shaft

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## Vehicle Subsystem



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## Vehicle Subsystem



- Note the difference in between the Driveline connection ports and the Simulink connection ports
- They will not connect to each other
- Driveline blocks are used to solve the Conservation of Angular Momentum principle during each time step

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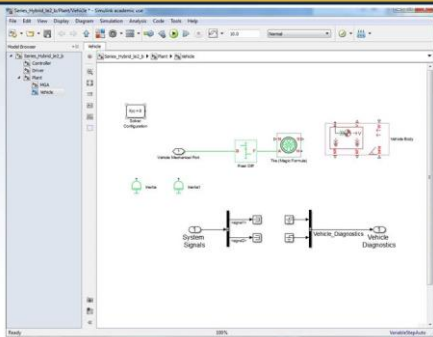
## Vehicle Subsystem



- Every shaft must have a mass moment of inertia
- Also, a special solver must be used
- From
  - Simscape / Foundation Library / Mechanical / Rotational Elements
- Drag in two **Inertia** blocks
- From
  - Simscape / Utilities
- Drag in a **Solver Configuration** block

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## Vehicle Subsystem



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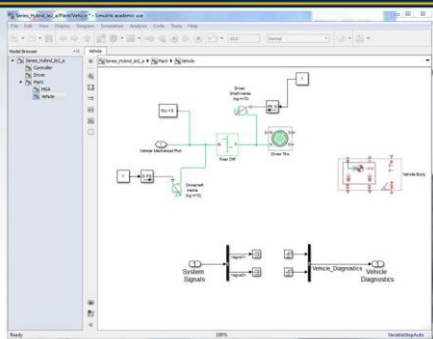
## Vehicle Subsystem



- Connect the **Inertias** to the driveshaft and rear axle and Rename them:
  - Driveshaft Inertia ( $\text{kg m}^2$ )
  - Driver Halfshaft Inertia ( $\text{kg m}^2$ )
- Connect the **Solver Configuration** to the driveshaft
- Hide its name
- Rename the Tire to Driver Tire
- Rename the Simple Gear to Rear Diff

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## Vehicle Subsystem



75

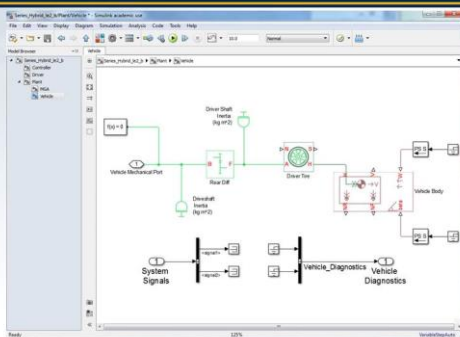
## Vehicle Subsystem



- Rename and save you your model as Series\_Hybrid\_le2\_b
- Now we can connect the H port of the **Drive Tire** block to the H port of the **Vehicle Body** block
- From
  - Simscape / Utilities
- Drag in two **Simulink-PS Converter** blocks
- These blocks transfer Simulink signals to physical signals
- Drag in two **Ground** blocks
- Hide their names
- Connect the **Ground** blocks to the **Simulink-PS Converters**
- Connect the **Simulink-PS Converters** to the beta and W ports of the **Vehicle Body** block

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## Vehicle Subsystem



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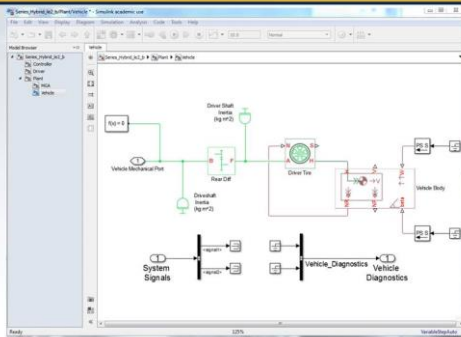
## Vehicle Subsystem



- The NR port of the **Vehicle Body** block is a physical signal reporting the normal force applied on the rear wheels
- Connect the NR port of the **Vehicle Body** block to the N port of the **Driver Tire** block
- Rearrange the block and wire to make model look good
- S port of the Drive Tire block is a physical signal reporting tire slip

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## Vehicle Subsystem



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## Vehicle Subsystem



- S port of the **Drive Tire** block is a physical signal reporting tire slip
- NF port of the **Vehicle Body** block reports the normal load on the front tires
- We are not interested in tire slip or the normal force on the front tire (this is a rear wheel vehicle)
- From
  - **Simscape / Utilities**
- Grab two **PS-Simulink Converter** blocks (opposite of **Simulink-PS Converter**)
- Grab two **Terminator** blocks
- Hide name of **PS-Simulink Converter** and **Terminator** both blocks
- Connect the **PS-Simulink Converter** blocks to the **Terminators**
- Connect the S port of the Drive Tire block and NF port of the Vehicle Body block to the **PS-Simulink Converter** blocks
- Rearrange the block to make the model look good and compact

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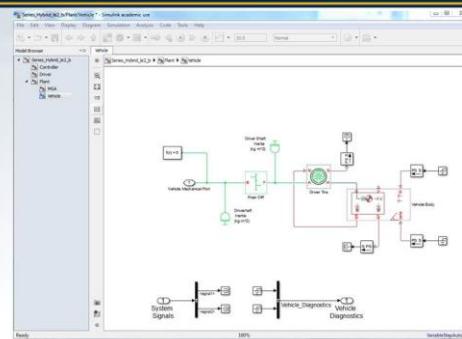
## Vehicle Subsystem



- It is preferred for the wires not to cross as much as possible. You should place your blocks in position to avoid the wire crossing as much as possible

81

## Vehicle Subsystem



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## Vehicle Subsystem



- We need to send the vehicle velocity signal (from the V port of the **Vehicle Body** block) to other parts of the model
- Grab a **PS-Simulink Converter** block and connect the V port of the **Vehicle Body** block to the **PS-Simulink Converter**
- From
  - **Simulink / Signal Routing**
- Grab a **Goto** block
- Connect the **PS-Simulink Converter** to the **Goto** block

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## Vehicle Subsystem



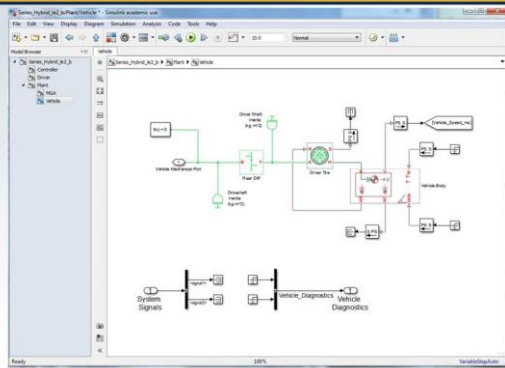
- Double click on the **Goto** block and change the tag to **Vehicle\_Speed\_ms**
- Make sure the visibility is set to local
- Click OK
- Drag the sides of the **Goto** block to make the name visible
- Hide the name of **Goto** and **PS-Simulink Converter** blocks



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## Vehicle Subsystem

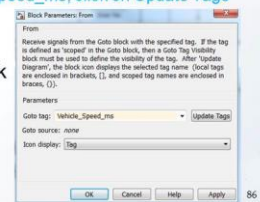


85

## Vehicle Subsystem

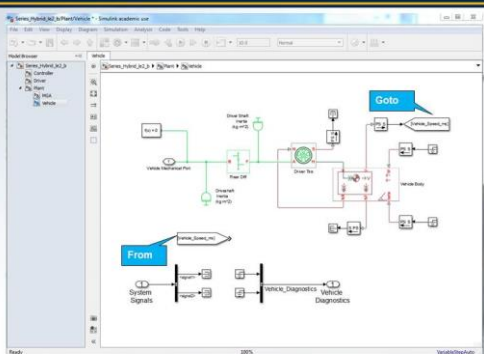


- From
  - Simulink / Signal Routing
- Grab a **From** block
- Double click on the **From** block
- Open the Goto tag bar and select **Vehicle\_Speed\_ms**
  - Note: if you don't see **Vehicle\_Speed\_ms**, click on **Update Tags** and try again
- Click OK
- Hide the name of the **From** block



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## Vehicle Subsystem



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## Vehicle Subsystem



- Congratulations – you have a model of a vehicle that will (hopefully) move when subjected to a torque
- From a vehicle diagnostics perspective, we are interested in two things
  - Speed (mph)
  - Distance Traveled (miles)
- Move the **Vehicle\_Speed\_ms From** block down near the **Vehicle Diagnostics Bus**
- From
  - Simulink / Commonly Used Blocks
- Drag in two **Gain** blocks and one **Integrator** block
- Hide the name of the **Integrator** block

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## Vehicle Subsystem



- Congratulations – you have a model of a vehicle that will (hopefully) move when subjected to a torque
- From a vehicle diagnostics perspective, we are interested in two things
  - Speed (mph)
  - Distance Traveled (miles)
- Move the **Vehicle\_Speed\_ms From** block down near the **Vehicle Diagnostics Bus**
- From
  - Simulink / Commonly Used Blocks
- Drag in two **Gain** blocks and one **Integrator** block
- Hide the name of the **Integrator** block

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## Vehicle Subsystem



- Set the first **Gain** to 2.2374 (m/s to mph)
- Set the second **Gain** to 6.214 e-4 (m/s to miles/s)
- Name them accordingly
- Connect the output of the **From** block to the input of the **Gain** blocks
- Connect the second **Gain** block to the **Integrator** block
- Delete the **Grounds** going into the **Bus**
- Connect the two signals
- Name them: (Double click on the wires)
  - **Vehicle\_Speed\_mph**
  - **Vehicle\_Distance\_miles**

90

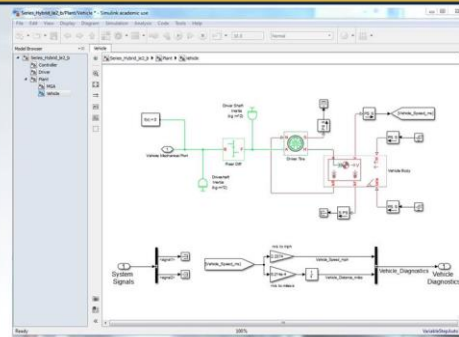
## Vehicle Subsystem



- The Vehicle subsystem is done
  - Note that we did not use any of the system signals
  - When we get around to mechanical brakes we will need a brake signal
- Take some time to move things around and make it look nice

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## Vehicle Subsystem



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## MGA Subsystem



- Our motor model is going to be preposterously simple
  - Constant torque output
  - No RPM limits
  - No input power required
- Rename and save you model as:
  - Series\_Hybrid\_le2\_c
- Use the Model Browser to go to the MGA subsystem

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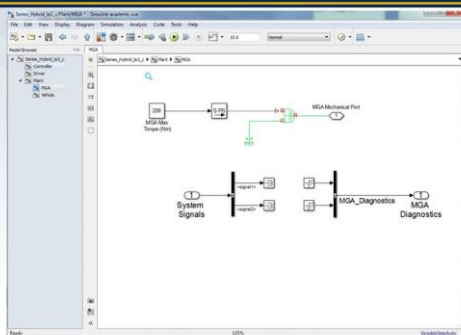
## MGA Subsystem



- Delete the **In1**, **Out1**, and the connecting wire
- Copy the Bus from the Plant subsystem
  - Rename accordingly
- Drag in a Constant block and rename it MGA Max Torque (Nm)
- Set the constant value at 200 Nm
- From
  - Simscape / Foundation Library / Mechanical / Mechanical Sources
- Drag in an **Ideal Torque Source** block and hide its name
- Drag in a **Connection Port** and rename it MGA Mechanical Port
- Drag in a **Simulink-PS Converter** and hide its name
- From
  - Simscape / Foundation Library / Mechanical / Rotational Elements
- Grab a **Mechanical Rotational Reference** block and hide its name
- Wire everything together as shown in the next page

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## MGA Subsystem

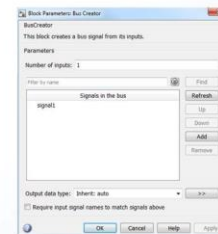


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## MGA Subsystem



- The only MGA diagnostic is the torque
- Delete the **Terminators**
- Double click on the **Bus Creator** and change the Number of inputs to 1
- Click OK



96

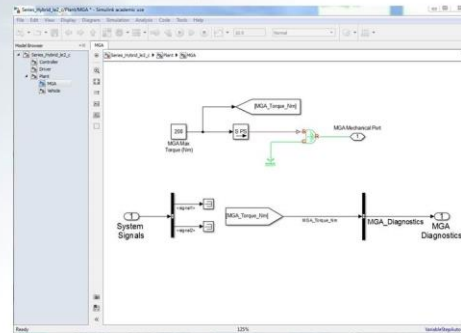
## MGA Subsystem



- Drag in a pair of **Goto** and **From** blocks
- Name them accordingly
- Use them to route the MGA torque signal to the **Bus Creator** block
- Name the signal going into the Bus
- Compare with next slides

97

## MGA Subsystem



98

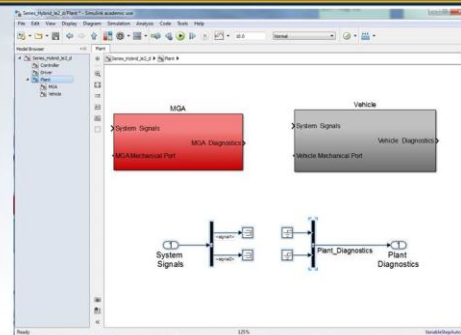
## MGA Subsystem



- Congratulations – you've made a motor model
- It is now time to connect MGA to the Vehicle and see what happens
- Save your model as [Series\\_Hybrid\\_1e2\\_d](#)
- Use the Model Browser to go to the Plant level of the model
- Make the MGA and Vehicle subsystems larger

99

## Plant Level

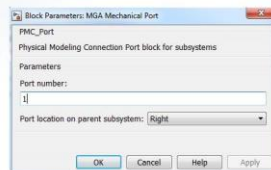


100

## Plant Level



- The MGA Mechanical Port is on the wrong side of the block
- Go into the MGA subsystem and double click on the Port
- Change the location from left to right
- Click OK



101

## Plant Level



- Connect the two Mechanical Ports
- Route the Vehicle and MGA Diagnostics to the Plant Signal CAN Bus
  - Name the signal inputs accordingly
- Change the foreground of the Vehicle routing blocks to gray
- Change the foreground of the MGA routing blocks to red

102

### Plant Level

103

### Plant Level

- All that is left is to route the System Signals to the Vehicle and MGA
- Delete the **Terminators** and the **Bus Selector**
- Use the **Goto** and **From** blocks to route the System Signals to MGA and the Vehicle

104

### Plant Level

105

### Running the Model

- It is now time to run the model
- Go to the Vehicle subsystem
- From
  - **Simulink / Sinks**
- Drag in a **Scope**
- Connect it to the **Vehicle\_Speed\_mph** signal
- Click on the Run button to start the simulation

106

### Running the Model

107

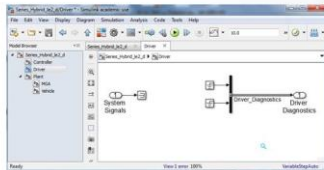
### Running the Model

- Oh! There is an error

108

## Running the Model

- Since we do not yet need any signals from the System Diagnostics bus, replace the **Bus Selector** with a **Terminator** in the
  - Driver
  - Controller
  - Plant / Vehicle
  - Plant / MGA
- subsystems



109

## Running the Model

- Rerun the model
- If the scope is not open, double click on the block
- Observe the results
  - 0 to 25 mph in 10 seconds
- You can check the results by hand

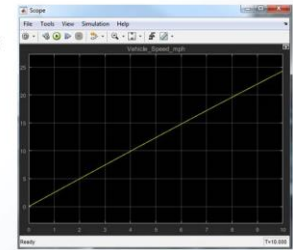
$$m \frac{dv_x}{dt} = \sum F_x$$

$$\frac{dv_x}{dt} = \frac{1}{m} \left( \frac{TN}{r} \right)$$

$$\int_{v_i}^{v_f} dv_x = \int_{t_i}^{t_f} \left( \frac{TN}{mr} \right) dt$$

$$v_f = \frac{(200 \text{ Nm})(2)}{(1200 \text{ kg})(0.3 \text{ m})} = 11.1 \text{ m/s} = 24.9 \text{ mph}$$

Gear ratio  
Tire radius



110

## Simple Vehicle

- Review
  - Vehicle subsystem
    - Receives torque from MGA
    - Solves vehicle velocity
    - Sends/Receives CAN messages
  - MGA subsystem
    - Provides torque to Vehicle
    - Sends/Receives CAN messages

111

## MBSD Lecture 3

Initialization, Solver, Logging & Visualization, and Driver

Georgia Tech The George W. Woodruff School of Mechanical Engineering

## Lecture Goals

- Create an initialization file for data management
- Select the correct ODE solver
- Create a Data Logging and Visualization subsystem
- Create a Driver subsystem
- Have the vehicle follow a drive cycle
- Experiment with feedback gain

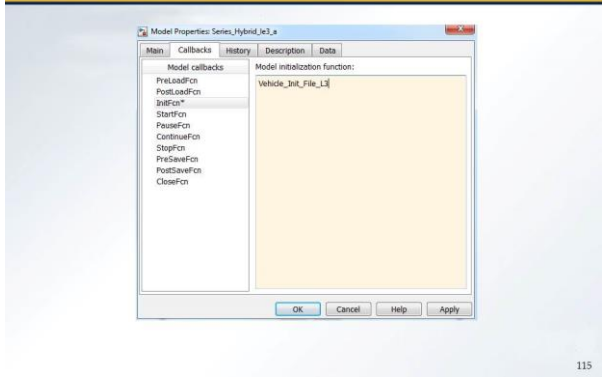
113

## Initialization File

- Save your previous model as
  - Series\_Hybrid\_1e3\_a
- Go to the System Level of the model
- Right click on any vacant white space on the model
  - Select Model Properties
  - Select the Callbacks tab
  - Click on the InitFcn
  - Type Vehicle\_Init\_File\_L3 in the second column

114

# Initialization File



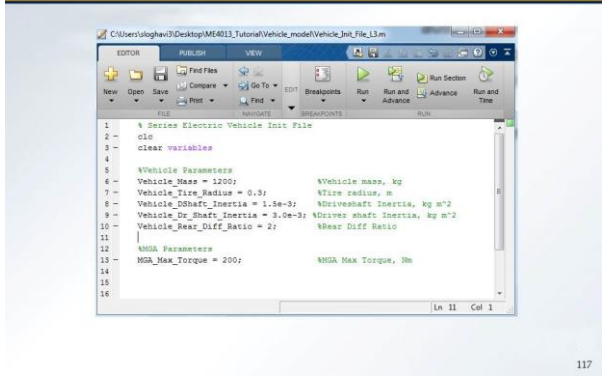
115

# Initialization File

- This is going to be a Matlab M-file which will automatically execute each time the model is run
- All of our component parameters will go in this file
- Start by going to the Matlab window
- Open file
  - [Vehicle\\_Init\\_File\\_L3.m](#)

116

# Initialization File



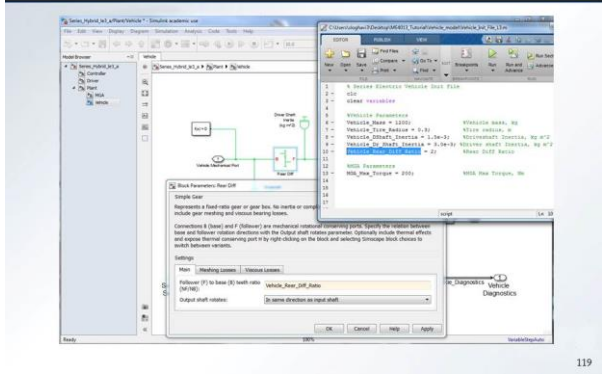
117

# Initialization File

- We will replace the numeric values in the blocks with these variable names
  - This consolidates all key parameters into one location
  - Help prevent making mistakes
  - Tip: copy the variable names from the M-file and paste into the block to prevent spelling errors
- Update all relevant blocks

118

# Initialization File



119

# ODE Solver

- While the Solver Configuration calls the hooks for solving the driveline angular momentum, an overall ODE solver must be specified
- Our model contains a set of stiff ODEs and as recommended by The MathWorks, best solver suited for this type of ODEs is ODE23t

120

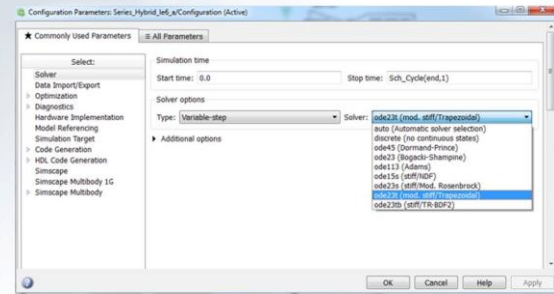
## ODE Solver



- As with the Init file, right click on the model background
- Select
  - Model Configuration Parameters
  - Solver: ode23t (mod. stiff/Trapezoidal)
  - Click OK
- Run the model and note how much faster it runs!

121

## ODE Solver



122

## Logging & Visualization



- Let's now create a subsystem to represent the on-board data logger
- This will also be a good place for having all the visualization
- First, go to the Vehicle subsystem and delete the Scope block
  - Having scopes in the model is bad for auto-code generation

123

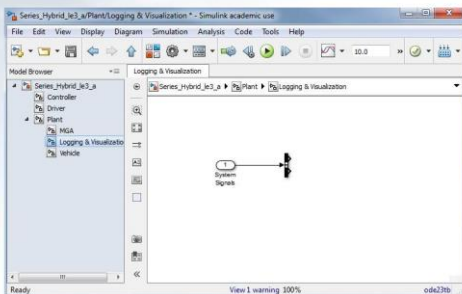
## Logging & Visualization



- Drag a new **Subsystem** into the Plant level of the model
- Rename it Logging & Visualization
- Rename the **In1** block to System Signals
- Delete the **Out1** block
  - The logger does not send any diagnostics
- Drag a **Bus Selector** into the subsystem
- Go back into the Plant level of the model
- Copy a System Signals **From** block and connect it to System Signals of Logging & Visualization subsystem

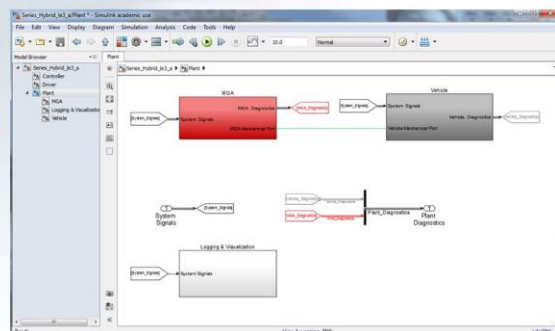
124

## Logging & Visualization



125

## Logging & Visualization



126

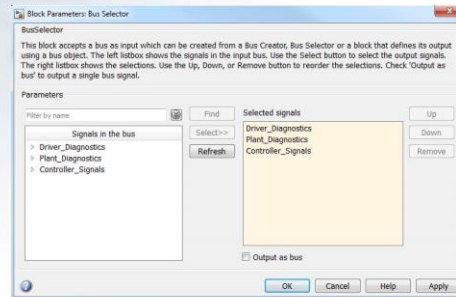
## Logging & Visualization



- In the Logging & Visualization subsystem
- Double click on the **Bus Selector** block
- In the Selected signals column
  - Click on signal1 and then Remove
  - Click on signal1 and then Remove
- In the Signals in the bus column
  - Click on Driver\_Diagnostics and then Select
  - Click on Plant\_Diagnostics and then Select
  - Click on Controller\_Diagnostics and then select
- Click OK

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## Logging & Visualization



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## Logging & Visualization



- Drag in three **Goto** blocks and name them Driver\_Diagnostics, Plant\_Diagnostics, and Controller\_Signals
- Wire them to the **Bus Selector**
- Drag in three **From** blocks
- **Terminate** the Driver\_Diagnostics and the Controller\_Signals **From** blocks
- Drag in a **Bus Selector** and wire it to the Plant\_Diagnostics **From** block

129

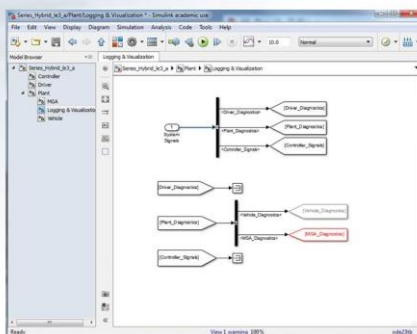
## Logging & Visualization



- Double click on the **Bus Selector**
- Remove signal1 and signal2
- Select the Vehicle\_Diagnostics and the MGA\_Diagnostics signals
- Drag in two more **Goto** blocks
  - Connect them to the **Bus Selector**
  - Name them accordingly
  - Color them accordingly

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## Logging & Visualization



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## Logging & Visualization



- Drag in two more **From** and **Bus Selector** blocks
- Select
  - Vehicle\_Diagnostics – first **From** block
  - MGA\_Diagnostics – second **From** block
- Connect the **From** blocks to the **Bus Selector** blocks
- Use the **Bus Selector** blocks to select all signals

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### Logging & Visualization

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### Logging & Visualization

- From
  - Simulink / Sinks
- Drag in a **To Workspace** block
  - This will write the desired signal values to an array for review when the simulation is done
- Double click on the **To Workspace** block
  - Make the following changes
    - Variable name - Vehicle\_Speed\_mph
    - Sample time - Logging\_Sample\_Time\_s
    - Save format – Array
- Update the input file

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### Logging & Visualization

135

### Logging & Visualization

- Connect the **To Workspace** block to the **Bus Selector**
- Repeat for the remaining signals coming from the Plant
- Color everything accordingly

136

### Logging & Visualization

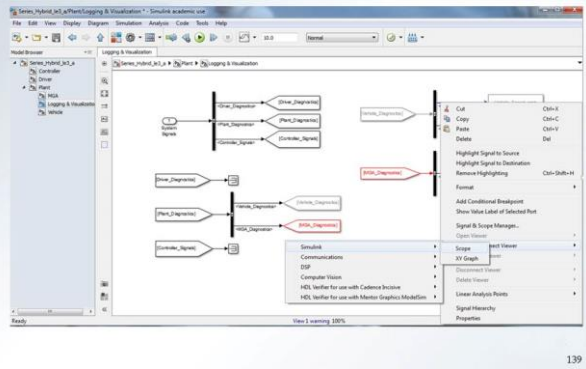
137

### Logging & Visualization

- Rather than using Scope blocks, we will use the Signal and Scope Manager to view the signals
- Right click on the Vehicle\_Speed\_mph signal and select
  - Create & Connect Viewer / Simulink / Scope

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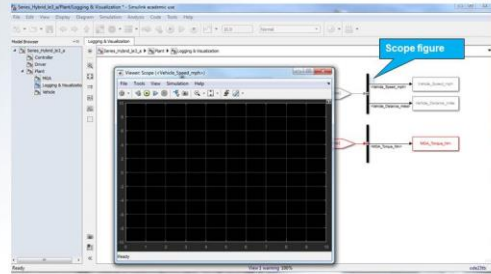
## Logging & Visualization



139

## Logging & Visualization

- Note the scope figure on the Vehicle\_Speed\_mph signal – that denotes that the signal is being scoped



140

## Driver Subsystem

- The Driver subsystem will be our first proportional feedback loop
  - The vehicle will be at a certain speed
  - The driver will want to go a different speed
  - The driver will send a torque request (-1 to 1) to the controller
  - The vehicle will change speed
  - Repeat for the duration of the drive cycle

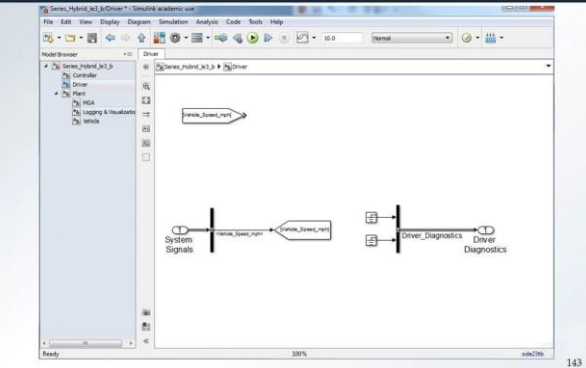
141

## Driver Subsystem

- Save the model as Series\_Hybrid\_le3\_b
- Use the Model Browser to go into the established Driver subsystem
- Drag in a **Bus Selector**, a **Goto**, and a **From** block
- Delete the **Terminator** and connect the **Bus Selector** block
- Select the Vehicle\_Speed\_mph signal
- Connect this signal to the **Goto** block

142

## Driver Subsystem



143

## Driver Subsystem

- The drive cycle will be read in from the workspace
  - It will contain time (s) and desired speed (mph) at that time
  - We will use the standard FU505 drive cycle
    - 505 seconds long
    - Collected data
  - The data is stored as a .mat file
  - The .mat file will be loaded with the init file

144

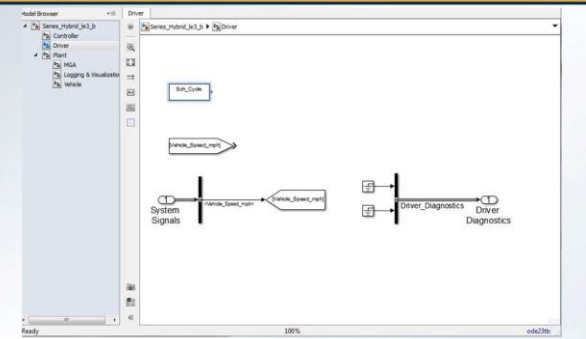
## Driver Subsystem



- From
  - Simulink / Sources
- Drag in a **From Workspace** block
  - Double click on it
  - Rename the Data to: **Sch\_Cycle**
  - Click OK

145

## Driver Subsystem



146

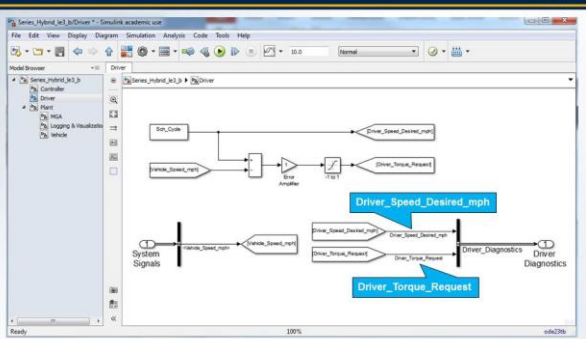
## Driver Subsystem



- From
  - Simulink / Commonly Used Blocks
- Drag in a **Sum**, **Gain**, and **Saturation** block
  - Make the Sum block rectangular with a +
- Drag in another two **Gain** and **From** blocks
- Arrange, wire, and name as shown on the next slide

147

## Driver Subsystem



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## Driver Subsystem



- We must now determine the value of the error amplifier
- Classical controls theory would have us generate a transfer function and look for poles
- This system will soon become so complex that classic approaches do not work
- Instead use some experience
- If you are driving a car and want to go
  - 1 mph faster
  - 10 mph faster
  - 20 mph faster
- You step on the gas (torque request)
  - Very little
  - Half way
  - To the floor

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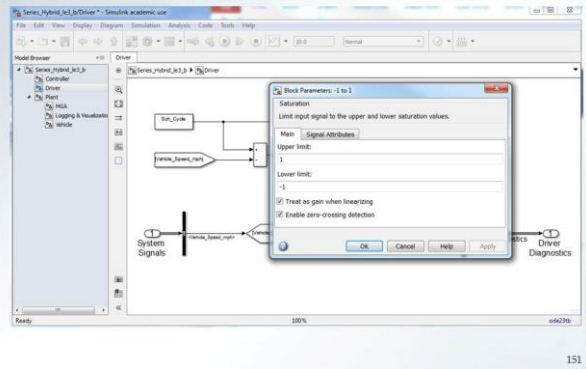
## Driver Subsystem



- The torque request will vary from
  - +1
    - full positive torque requested
    - 20 mph+ too slow
  - 1
    - Full negative torque requested
    - 20 mph+ too fast
- A gain of 0.05 will send a  $\pm 1$  at 20 mph error
- A Saturation will limit the request to  $\pm 1$

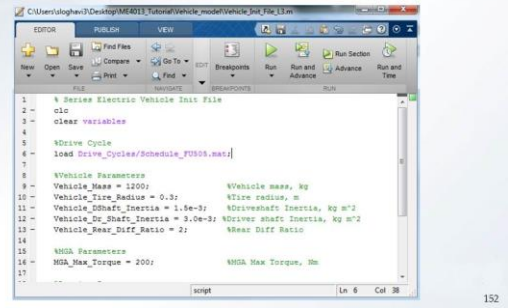
150

## Driver Subsystem



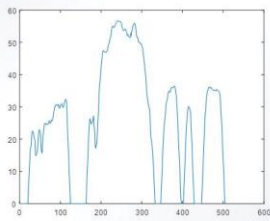
## Driver Subsystem

- The final step in the Driver subsystem is to add the desired drive cycle to the initialization file



## Driver Subsystem

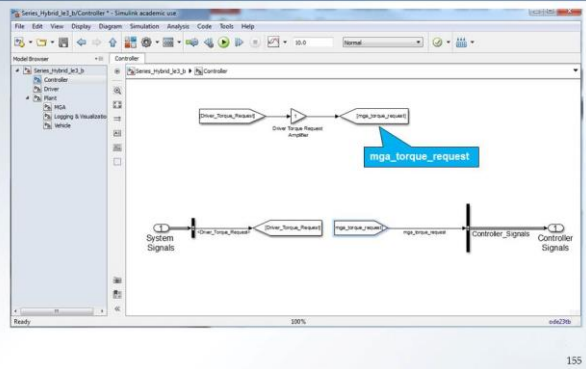
- Run the init file manually to load the drive cycle
- At the Matlab command line type
  - `plot(Sch_Cycle(:,1),Sch_Cycle(:,2))`



## Controller Subsystem

- The Driver torque request will next be sent through the CAN bus to the Controller subsystem
- Go to the Controller subsystem, delete the Terminators and Grounds, and drag in a Gain, a Bus Selector, two Goto, and two From blocks
- On the Bus Selector, select the Driver\_Torque\_Request signal
- Wire and label as shown on the next slide

## Controller Subsystem



## Controller Subsystem

- The driver torque request passes through the controller where it is sent to MGA
  - Just like a real vehicle
- The Gain block can be modified
  - Sporty : increase request
  - Economy : decrease request
- Naming convention
  - All Control signals will be lowercase
  - End with either "request" or "enable"

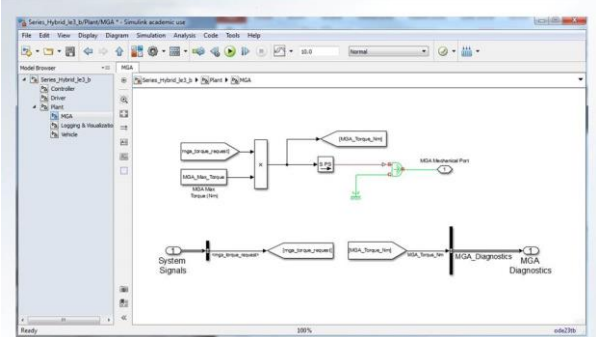
## MGA Subsystem



- The last step is to pass the torque request to MGA
- Drag in a **Bus Selector**, a **Goto**, a **From**, and a **Product** block
- Delete the **Terminator**
- Select the `mga_torque_request` signal
- Wire and label as shown on the next slide

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## MGA Subsystem



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## Run the simulation



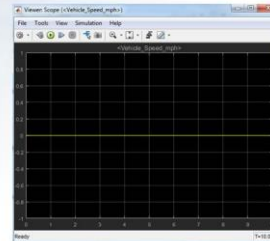
- Go to the model level and Run the model

159

## Results



- Run the model and observe the scope



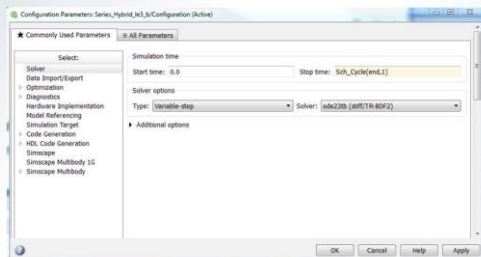
- It only ran for 10 seconds

160

## Results



- Open the Configuration Parameters
- Change the Stop Time to  
– `Sch_Cycle(end,1)`

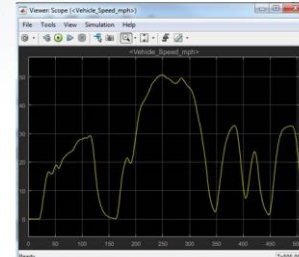


161

## Results



- Run the simulation again
- Awesome! The vehicle is doing something!
- Note: in the Scope, go to Tools / Axes Scaling
- Select: Automatically Scale Axis Limits



- But How well?

162

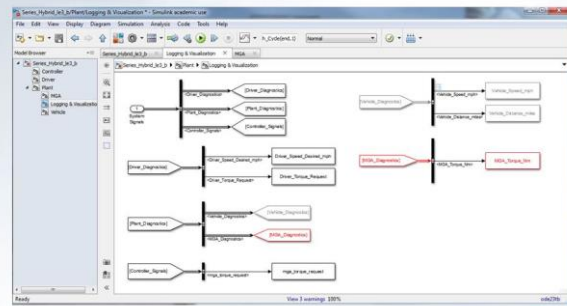
## Logging & Visualization



- Go to the Logging & Visualization subsystem
- Delete the **Terminator** on the Driver\_Diagnostics **From** block
- Use a **Bus Selector** to extract all Driver signals
- Drag in two **To Workspace** blocks to log the Driver Data
- Repeat for the Controller Signals

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## Logging & Visualization



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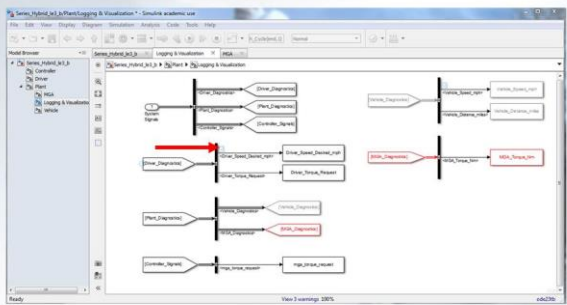
## Logging & Visualization



- Right click on the Driver\_Speed\_Desired\_mph signal
- Select
  - **Connect to Existing Viewer**
  - **Scope**
- Rerun the simulation and observe the scope

165

## Logging & Visualization

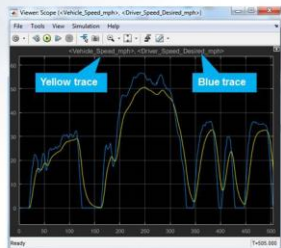


166

## Results



- Our vehicle does a poor job of following the drive cycle



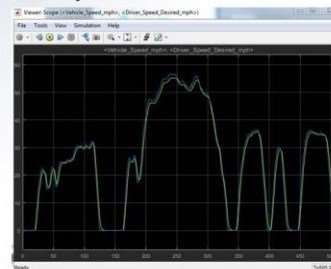
- Increase the MGA max torque in the init file until the traces are similar

167

## Results



- Even with 1000 Nm, the vehicle still can not follow the trace exactly



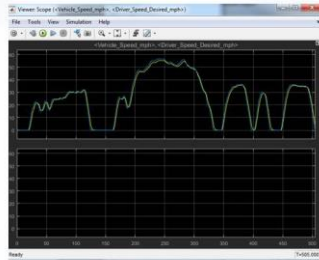
- Put the torque back to 200 Nm

168

## Results



- Let's see what the driver is doing
- On the scope
- Select
  - View / Layout
- Choose a  $2 \times 1$  layout
- Now we can add signals to both Axis 1 and Axis 2



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## Results



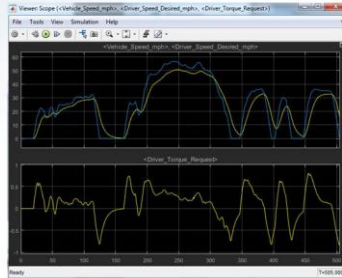
- In the Logging & Visualization subsystem
  - Right click on the Driver\_Torque\_Request signal
  - Select
    - Connect to Viewer / Scope / Axes 2
- This will put the Driver torque request directly below the drive cycle
- Rerun the simulation

170

## Results



- Wow – the driver torque request never goes above about 0.7



- Why?

171

## Results



- Note that the vehicle is “chasing” a velocity and only occasionally off by more than 10 mph
- Change the error amplifier in the Driver subsystem to make the driver more aggressive
  - Decreases the proportional error
  - Allows the system to respond faster
  - Can cause overshoot and oscillations

172

## Results



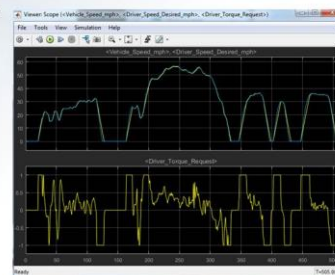
- An error amplifier of 1 should work quite well

173

## Results



- As we expected, error amplifier of 1 worked quite well for this drive cycle



- What about for a different drive cycle?

174

## Results



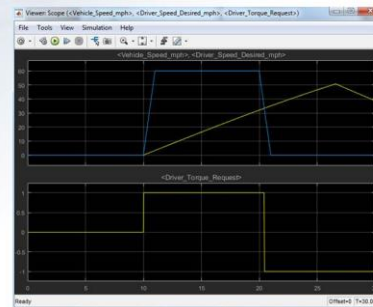
- In the init file
  - Change the drive cycle to Schedule\_Boston\_Cab.mat
  - Increase the MGA torque to 1000 Nm
- Rerun the simulation

175

## Results



- Definitely some overshoot



176

## Results



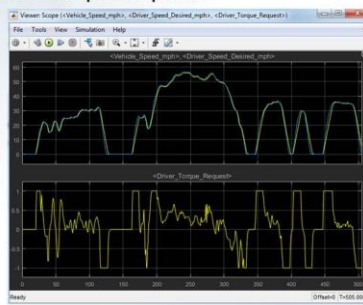
- In the init file
  - Change the drive cycle back to the FU505
  - Change the MGA torque back to 200 Nm
- In the Driver subsystem
  - Change the error amplifier to 0.5
- Rerun the simulation

177

## Results



- This provides a good balance of trace correlation and driver torque request



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## Initialization, Solver, and Driver



- Review
  - Created an initialization file to manage data
  - Changed to appropriate ODE solver
  - Created a logging and graphics subsystem
  - Developed driver proportional feedback loop
    - Velocity mismatch generates an error signal
    - Error signal converted to a driver torque request
    - Driver torque request goes to controller
    - Controller requests MGA torque
  - System response very sensitive to feedback gain value

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## MBSD Lecture 4

Improved MGA and Simple Battery

Georgia Tech The George W. Woodruff School of Mechanical Engineering



## Outline

- Make MGA an ideal electromechanical device
- Create a preposterous battery to power MGA
- Track the battery state of charge (SOC)

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## MGA

- Save your model as [Series\\_Hybrid\\_1e4\\_a.slx](#)
- Go into the MGA subsystem
- MGA will still
  - Have a constant maximum torque
  - Have no rpm limits
  - Be modulated by the Controller
- By providing torque at a given RPM, power is transferred to the Vehicle

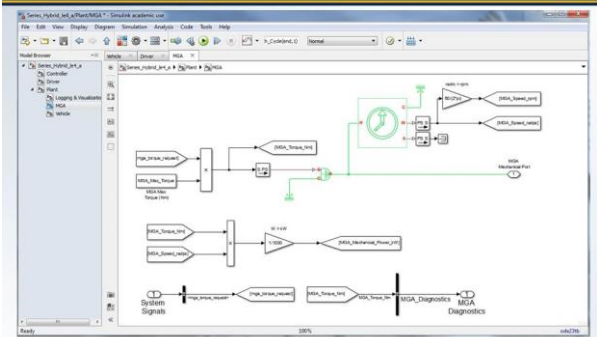
182

## MGA

- From
  - [Simscape / Foundation Library / Mechanical / Mechanical Sensors](#)
- Drag in an **Ideal Rotational Motion Sensor** block
- From
  - [Simscape / Foundation Library / Mechanical / Rotational Elements](#)
- Drag in a **Mechanical Rotational Reference** block
- Also drag in three **Gotos**, two **Froms**, two **Gains**, two **PS-Simulink Converters**, One **Terminator** and a **Product** block
- Arrange and wire as shown on the next slide

183

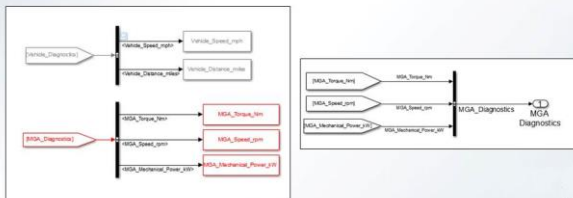
## MGA



184

## MGA

- Add the MGA\_Speed\_rpm and MGA\_Mechanical\_Power\_kW to the MGA diagnostics bus
- In the Logging & Visualization subsystem, extract and log these signals



185

## MGA & Battery

- To make the mechanical power, the motor will need to receive electrical power from the battery
- This power will be the battery voltage times the current drawn
- Let's now add an ideal battery to the Plant
- Drag in a **Subsystem** and rename it Battery
- Make the background orange

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## Battery



- The Battery will receive the system signals and broadcast its diagnostics
  - Delete the wire connecting the **In1** and **Out1** blocks and rename them System Signals and Battery Diagnostics
  - Terminate the System Signals
  - Drag in a **Bus Creator**, connect it to the Battery Diagnostics, and ground the inputs

187

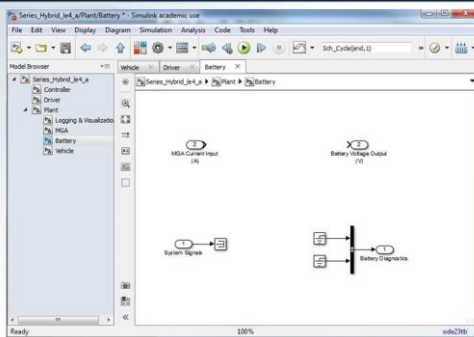
## Battery



- Electrically, the battery will
  - Tell MGA its voltage
  - Receive a current demand from MGA
- From
  - Simulink / Commonly Used Blocks
- Drag in an **In1** and an **Out1** block
- Rename them
  - MGA Current Input (A)
  - Battery Voltage Output (V)

188

## Battery



189

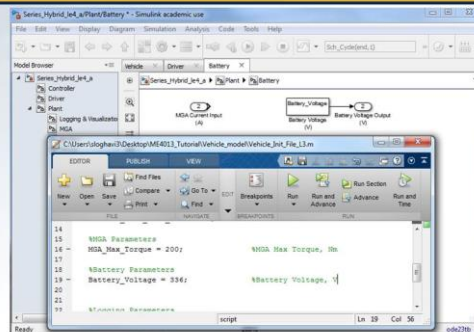
## Battery



- For now, the battery voltage will be constant and defined in the init file
- Drag in a **Constant** block
  - Rename it Battery Voltage (V)
  - Give it a variable name of Battery\_Voltage
  - Connect it to the Battery Voltage Output
- Update the init file for the battery voltage to be 336V

190

## Battery

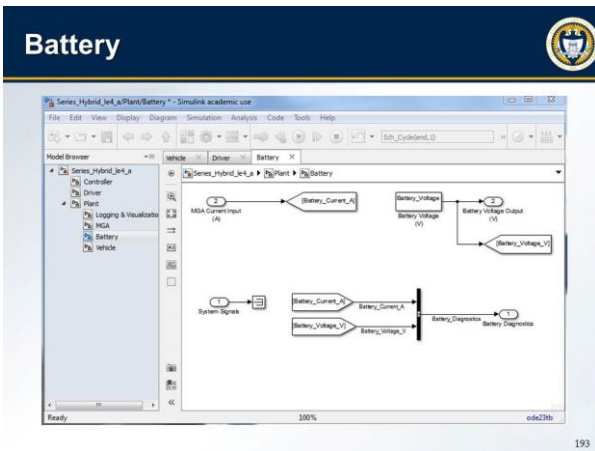


## Battery

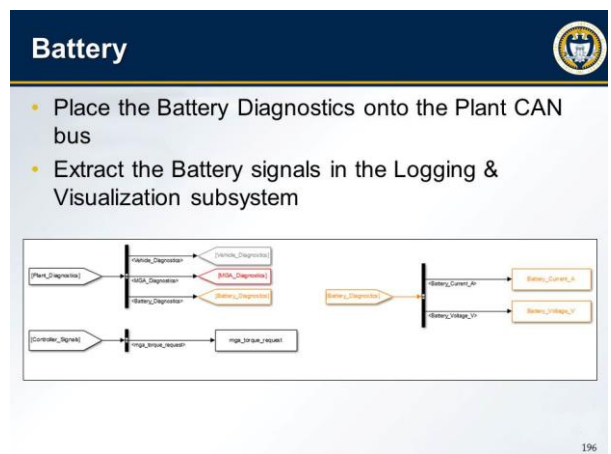
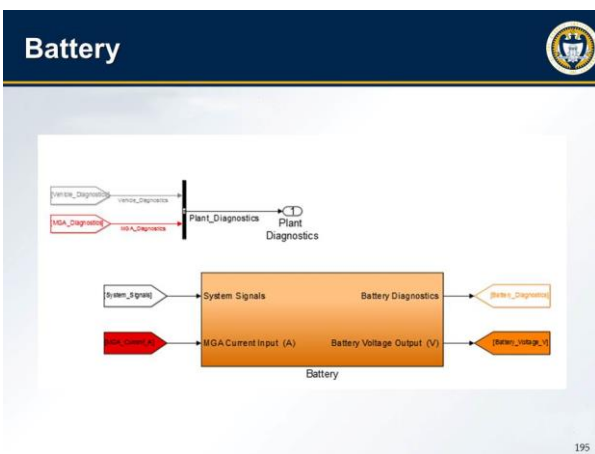


- Our magic battery can provide any amount of current required by the motor
- Put the battery current and the battery voltage on the Battery Diagnostics bus

192



- ## Battery
- Outstanding, we now have a constant voltage power source with unlimited amperage and capacity
  - Return to the Plant level of the model
    - Make the Battery subsystem larger
    - Drag in three From and two Goto blocks
    - Wire and label as shown on the next slide
  - Convention
    - Physical signals get colored background
    - CAN signals get colored foreground
- 194



- ## MGA
- Save the model as [Series\\_Hybrid\\_1e4\\_b.slx](#)
  - In the MGA subsystem, drag in an **In1** and **Out1** block
  - Rename them
    - Battery Voltage Input (V)
    - MGA Current Output (A)
  - For now, the electrical power from the battery will equal the mechanical power from the motor
- 197

- ## MGA
- Thus the current required will be the mechanical power divided by the battery voltage
  - From
    - Simulink / Math Operations
  - Drag in a **Divide** block
  - When MGA is drawing current, it will come out of the battery
  - Drag in a **Gain** block and give it a value of -1
- 198

### MGA

199

### MGA

- Place the MGA Current signal onto the diagnostics bus
  - Label it **MGA\_Current\_A**
- Extract the signal in the Logging & Visualization subsystem

200

### MGA / Logging & Visualization

201

### MGA

- Return to the Plant level of the model
- Make the MGA subsystem larger
- Use a **Goto** and a **From** block to connect MGA to the Battery
  - Color accordingly

202

### MGA

203

### Logging & Visualization

- Let's now observe the MGA current as it follows the FU505 drive cycle
  - Add a third axis to the Scope
  - Place the **MGA\_Current\_A** signal on the third axis
  - Run the simulation

204

## Results



205

## Battery SOC

- Let's add battery charge to our model
- Battery Capacity is in units of Amp hours
  - The amount of charge transferred by a constant current of one amp for one hour
  - Thus we can integrate the battery current to find out how much charge has been removed
- This value can be non-dimensionalized by dividing by the maximum capacity of the battery to get the State of Charge (SOC)
  - 1 fully charged
  - 0 fully discharged

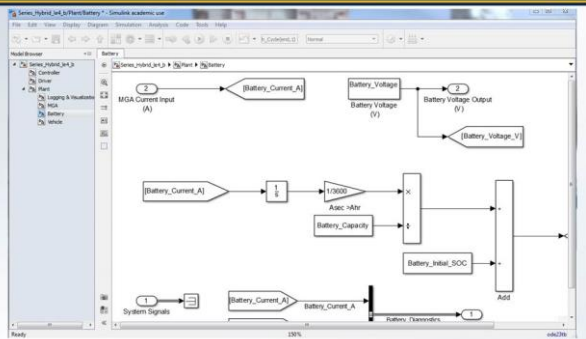
206

## Battery SOC

- In the Battery subsystem, drag in
  - An Integrator block
  - A Gain block
    - 1/3600
  - Two Constant blocks
    - Battery\_Capacity 8.5 Ahrs
    - Battery\_Initial\_SOC 0.7
  - A Divide block
  - And a Sum block
- Arrange as shown on the next slide

207

## Battery SOC



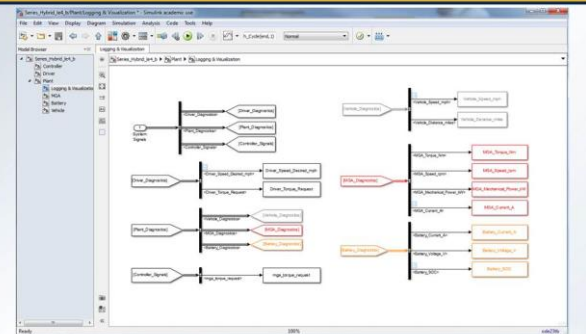
208

## Battery SOC

- Add the Battery\_SOC signal to the Battery Diagnostics bus
- Extract the signal in the Logging & Visualization subsystem
- Right click on the Driver\_Torque\_Request signal
  - Select Disconnect Viewer
  - Select Scope
  - Select Axes 2
- Add the SOC signal to Axes 2

209

## Logging & Visualization



210

## Initialization



- The last step is to add the two new battery parameters to the init file

```
C:\Users\loghavi\Desktop\ME4013_Tutorial\Vehicle_model\Vehicle_init_File_L3.m
EDITOR  FILE  VIEW  SIMULATION  RUN
New  Open  Save  Compare  Go To  Breakpoints  Run  Run and Advance  Run and Test
14  MGA Parameters
15  MGA_Max_Torque = 200;          MGA_Max_Torque, Nm
17
18  %Battery Parameters
19  Battery_Voltage = 336;        %Battery Voltage, V
20  Battery_Capacity = 8.5;       %Battery Max Capacity, Ah
21  Battery_Initial_SOC = 0.7;    %Battery initial SOC
22  |
23  %Logging Parameters
24  Logging_Sample_Time_s = 1;    %Logging time, s
script  Ln 22  Col 1
```

- Run the Simulation

211

## Results



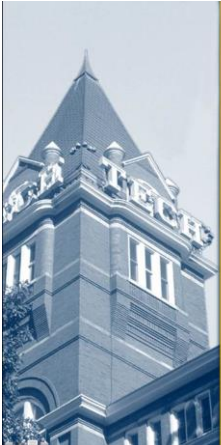
212

## Review



- Developed an electromechanical model of MGA
  - 100% efficient
  - Constant torque capable
  - No rpm limits
- Developed an electrical battery model
  - 100% efficient
  - Constant voltage capable
  - No current limits
  - SOC ready

213



MBSD Lecture 5

Simple Charging logic

Georgia Tech The George W. Woodruff School of Mechanical Engineering

## Outline



- Simple Battery charging
- Simple GenSet charge logic

215

## Battery SOC



- While having our vehicle follow a drive cycle, the Battery SOC decreased
- Eventually, the Battery SOC will get too low
  - Permanent damage
  - Stranded motorist
- This will require an on-board power generating unit
  - Diesel engine and MGB

216

## Battery SOC / GenSet



- The net effect of the GenSet is to provide power to the battery
  - MGB current at Battery voltage
- Let's make a ridiculously simple MGB
  - Constant current output
- Rename the model `Series_Hybrid_le5_a.slx`
- In the plant level of the model, drag in a new **Subsystem** and rename it MGB
  - Make the background green

217

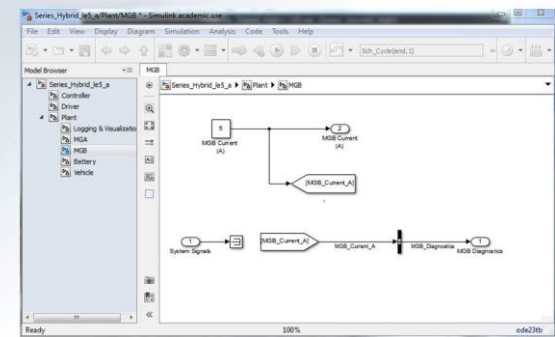
## MGB



- In the MGB subsystem
  - Delete the connecting wire
  - Create the System Signals and MGB Diagnostics bus
  - Drag in a **Constant** block
    - Rename it MGB Current (A)
    - Give it a value of 5
  - Drag in an **Out1** block
    - Rename it MGB Current Output (A)
  - Put the current on the Diagnostics Bus

218

## MGB



219

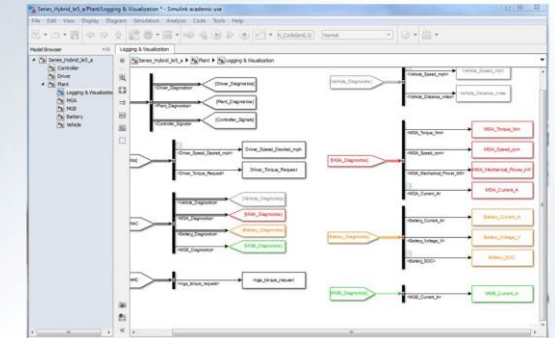
## MGB



- Return to the Plant level and put the MGB Diagnostics on the Plant Diagnostics Bus
- Connect a System Signals **From** block
- Go to the Logging & Visualization subsystem and extract the MGB diagnostics
- Color accordingly!

220

## MGB



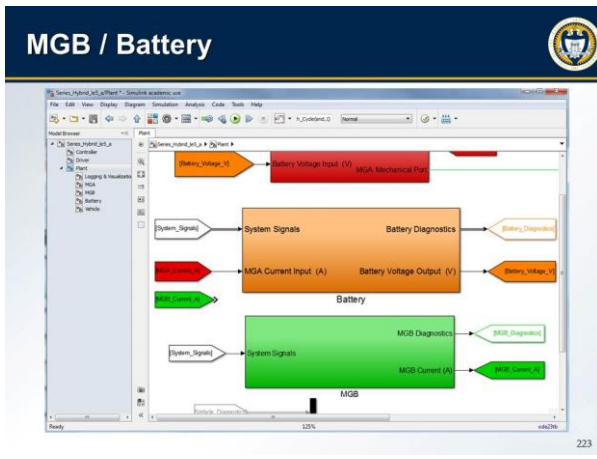
221

## MGB / Battery

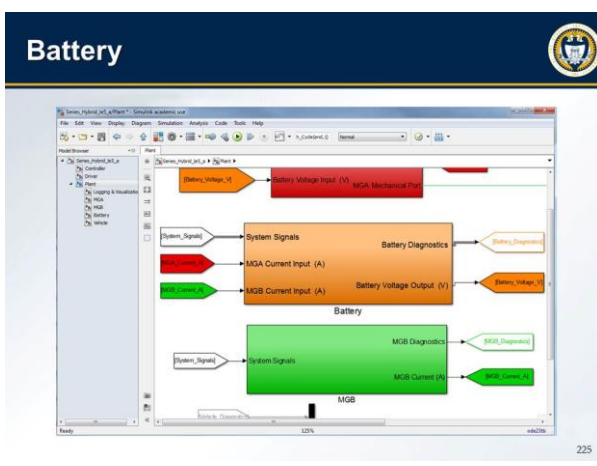


- Return to the Plant level of the model and use a **Goto** and a **From** block to make the electrical connection to the Battery

222



- ### Battery
- Drag in an **In1** block and rename it MGB Current Input (A)
  - Drag in a **Sum** block and add the two Current Inputs together
  - Connect the output to the Battery\_Current **Goto** block
  - Excellent – both MGs can remove/provide current from/to the Battery
  - Return to the Plant level and connect the MGB\_Current\_A **From** block
- 224



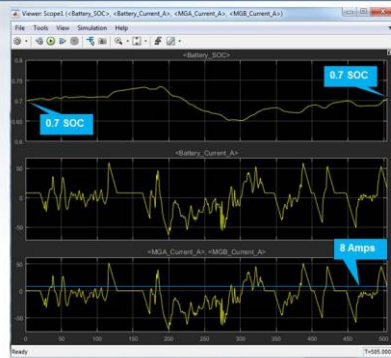
- ### Battery
- In the Logging & Visualization subsystem
    - Right click on the Battery\_SOC signal
    - Create a new scope
    - Give it three axis
      - Axis 1 – Battery SOC
      - Axis 2 – Battery Current
      - Axis 3 – MGA and MGB Current
  - Run the simulation
- 226



- ### Results
- The "trickle charge" of a constant 5 amps kept the SOC up but the Battery was still *charge depleting*
  - Vary the value of the MGB current until the Battery is relatively *charge sustaining*
- 228



## Results



229

## GenSet Controls

- Our series electric vehicle will use a genset which will be activated intermittently
  - SOC too low : start charging battery
  - SOC too high : stop charging battery
- Let's make the idealized genset do exactly that via MGB and a state machine
- Go to the Controller level of the model

230

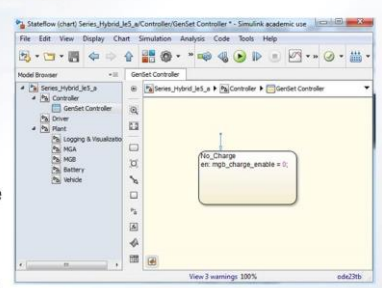
## GenSet Controls

- From
  - Simulink / Stateflow
- Drag in a **Chart** block
  - Rename it GenSet Controller
  - Double click on it
- The first logical state will be the “no charge” state
  - The Battery SOC is above the lower limit

231

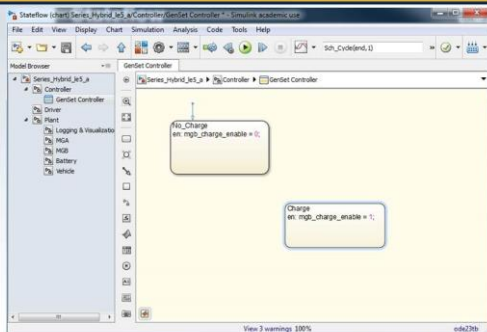
## Stateflow

- Left click on the State icon
- Left click on the window
- Name the state
  - No\_Charge
- On entry, set the MGB charge enable to 0
- Create a charge state



232

## Stateflow



233

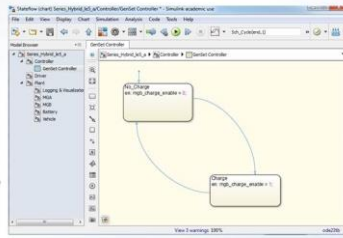
## Stateflow

- To go from one state to another, a *transition* must occur
- These transitions are controlled by parameters called *guards*
- If, at the instance the guard is queried, the parameters are true then the transition occurs
- Let's now create the transitions followed by the guards

234

## Stateflow

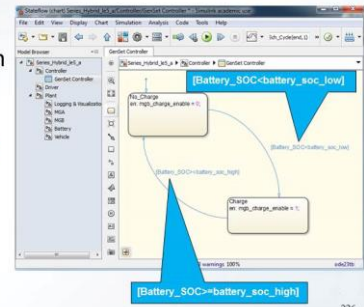
- Place the mouse over the No\_Charge state
- Left Click
- Hold and drag to the Charge state
- Connect the Charge back to the No\_Charge



235

## Stateflow

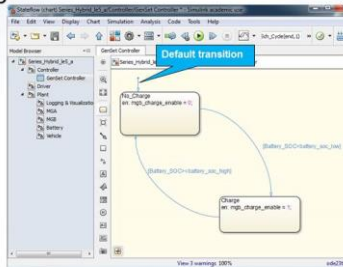
- Left click on the right transition
- Replace the question mark with the code at right
  - Enclosed in square brackets
- Repeat for the second transition



236

## Stateflow

- Stateflow needs to know which state to "wake up" in
- The default transition was created when the first state was placed



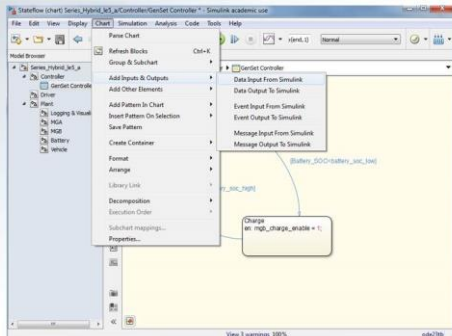
237

## Stateflow

- We need to add the Battery\_SOC input from Simulink into Stateflow
  - Select
    - Chart / Add Inputs & Outputs / Data Input From Simulink
    - Change the Name to Battery\_SOC
    - Click OK
  - If you select
    - Tools / Model Explorer
- You can see that Battery\_SOC has been added
- Close Model Explorer

238

## Stateflow



239

## Stateflow

- Similarly we need to add the value of mgb\_charge\_enable to Simulink
  - Select
    - Chart / Add Inputs & Outputs / Data Output to Simulink
    - Change the Name to mgb\_charge\_enable
    - Click OK
- We also need to add the guards as parameters
  - Select
    - Chart / Add Other Elements / Parameter...
    - Change the Name to battery\_soc\_low
    - Click OK
  - Repeat for battery\_soc\_high

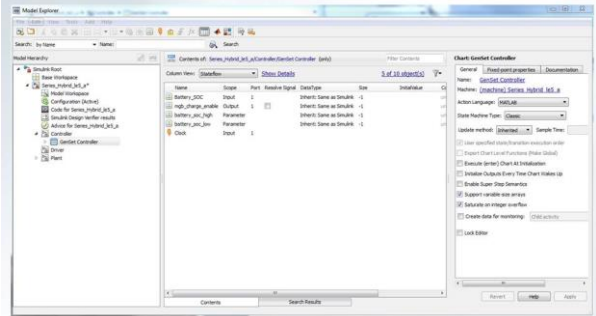
240

## Stateflow

- The final step is to set the frequency at which the guards are queried
  - Select
    - Chart / Add Inputs & Outputs / Event Input From Simulink
  - Change the Name to Clock
  - Click OK
- After this step, Model Explorer should be as shown in the next slide

241

## Model Explorer



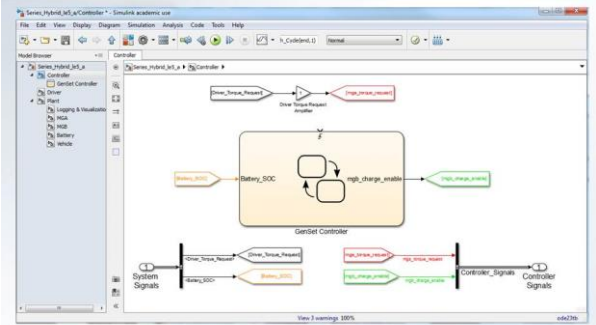
242

## GenSet Controller

- Return to the Controller level of the model
- Make the GenSet Controller subsystem larger
- Extract the Battery\_SOC from the System Signals bus and route it to the GenSet Controller
- Route the mga\_charge\_enable signal to the Control Signals bus and add it

243

## GenSet Controller



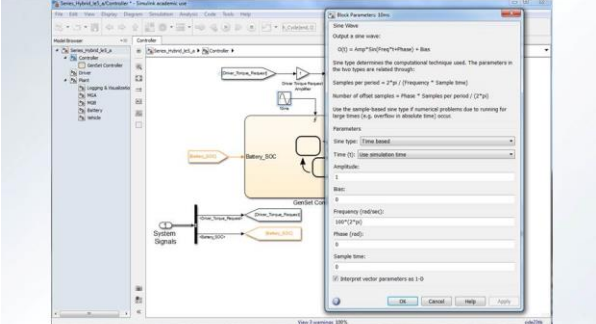
244

## GenSet Controller

- From
  - Simulink / Sources
- Drag in a Sine Wave block
  - Rename it 10ms
- Connect it to the Clock Port
- Double Click on it
  - Change the Frequency to  $100 \cdot (2 \cdot \pi)$
  - Click OK

245

## GenSet Controller



246

## GenSet Controller



- Excellent, we have built a simple two state controller which, based on the Battery SOC, will determine whether or not it should be charged
- This check will be performed every 10 ms
- In the init file, we need to add guards
  - Low SOC : 0.6
  - High SOC : 0.7
- Add to the init file

247

## GenSet Controller



```

17
18
19 - Battery Parameters
20 - Battery_Voltage = 336;
21 - Battery_Capacity = 5.5;
22 - Battery_Initial_SOC = 0.7;
23
24 - MController Parameters
25 - battery_soc_low = 0.6;
26 - battery_soc_high = 0.7;
27
    
```

248

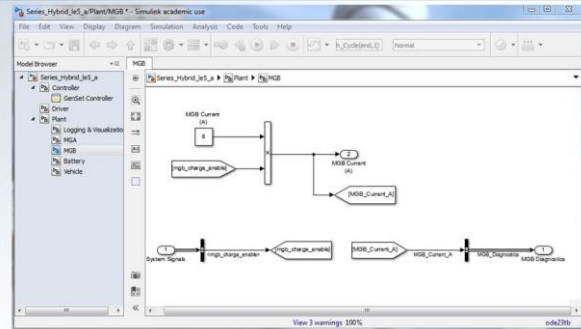
## MGB



- Return to the MGB subsystem
- Extract the mgb\_charge\_enable signal from the System Signals
- Drag in a **Product** block and multiply the mgb\_charge\_enable signal with the MGB Current block
- Run the simulation

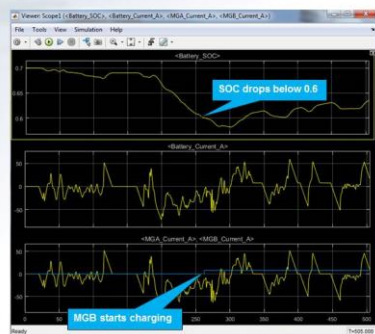
249

## MGB



250

## Results



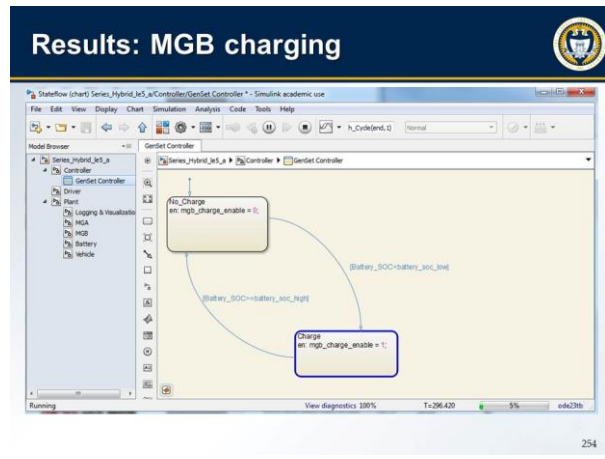
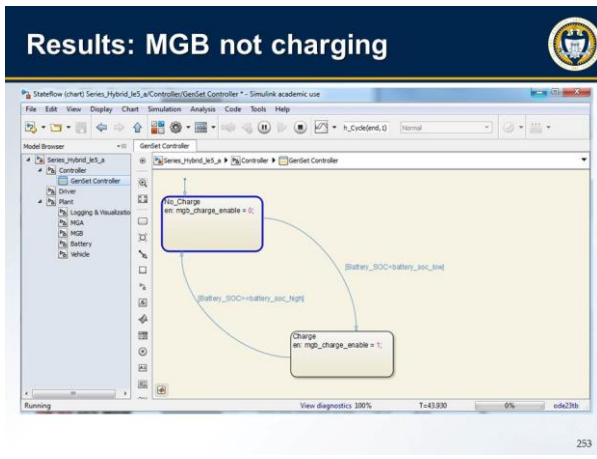
251

## Results



- In the init file, change the drive cycle to
  - [Schedule\\_FU505\\_Ten\\_Times.mat](#)
- Increase the MGB current to 50 amps
- Run the simulation
  - MGB now cycles on and off
  - Watch the Stateflow chart
    - The current state is highlighted during the simulation

252



- ### GenSet and Stateflow
- Review
    - Created a simple GenSet which provides current to recharge the battery
    - Tweaked the trickle charge current for charge sustaining
    - Created a state controller to turn the charge on and off

### MBSD Lecture 6

Improved GenSet and Controls

Georgia Tech The George W. Woodruff School of Mechanical Engineering

- ### Outline
- Improved MGB model
  - Simple diesel engine model
  - Charging logic

- ### Improved GenSet
- Previously, the GenSet simply provided a constant current
  - Improvements
    - Diesel Engine
      - Constant torque
      - Speed control
    - MGB
      - Recycle MGA model
  - Rename your model
    - Series\_Hybrid\_1e6\_a.slx

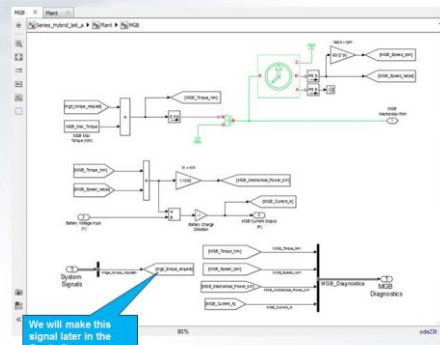
## Improved MGB



- Delete the MGB model you just got done making
- Copy the MGA model
  - Make it green
  - Rename EVERYTHING from MGA to MGB
- Add the MGB max torque to the init file
  - 200 Nm
- Update the Logging & Visualization for MGB
  - Note: We will create the signal for mgb\_torque\_request later in the controller subsystem

259

## Improved MGB



260

## Logging & Visualization



261

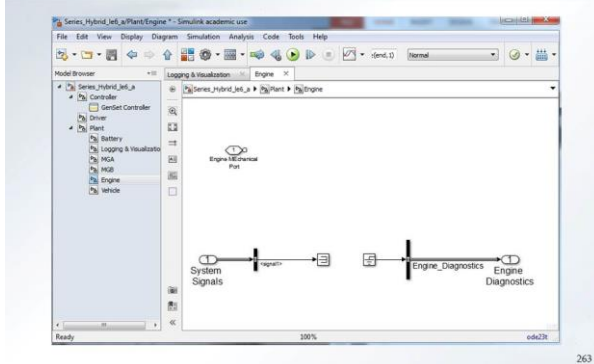
## Engine



- Return to the Plant level, drag in a **Subsystem**, and rename it Engine
  - Make it yellow
- In the Engine subsystem, create the System Signals and Engine Diagnostics bus
- Drag in a **Connection Port** for the mechanical connection to MGB

262

## Engine



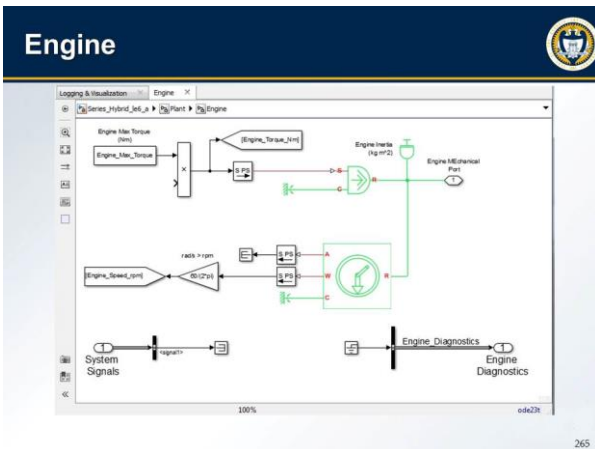
263

## Engine



- Drag in
  - One Constant
  - One Simulink-PS Converter
  - One Ideal Torque Source
  - One Inertia
  - One Ideal Rotational Motion Sensor
  - One Gain
  - One Terminator
  - One Product
  - Two Goto
  - Two PS-Simulink Converters
  - Two Mechanical Rotational References
- Arrange and label as shown on the next slide

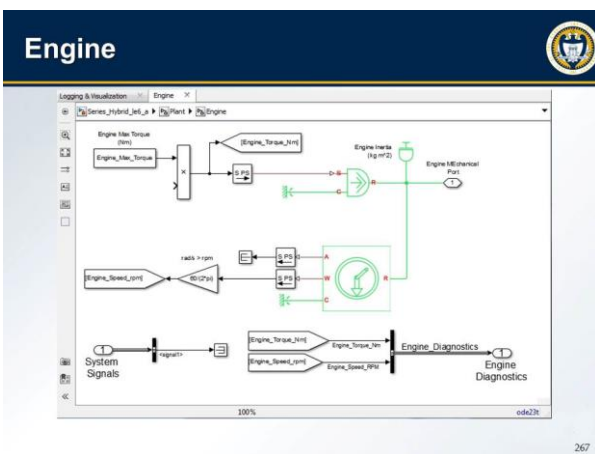
264



265

- ### Engine
- Our engine torque will be modulated with a throttle request from the Controller
    - Range of 0 to 1
    - Analogous to a driver stepping on the accelerator pedal
  - Place the engine torque and speed on the Diagnostic bus

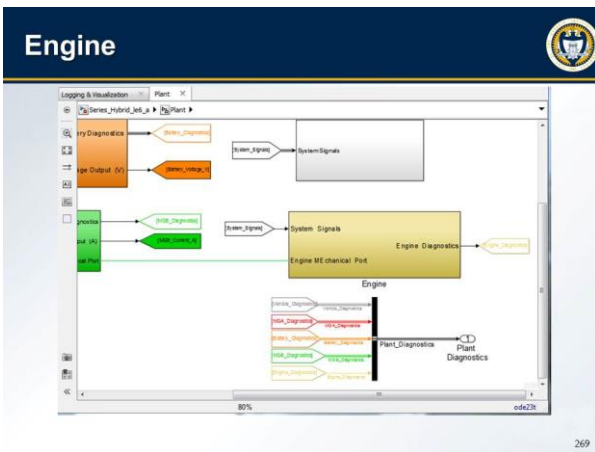
266



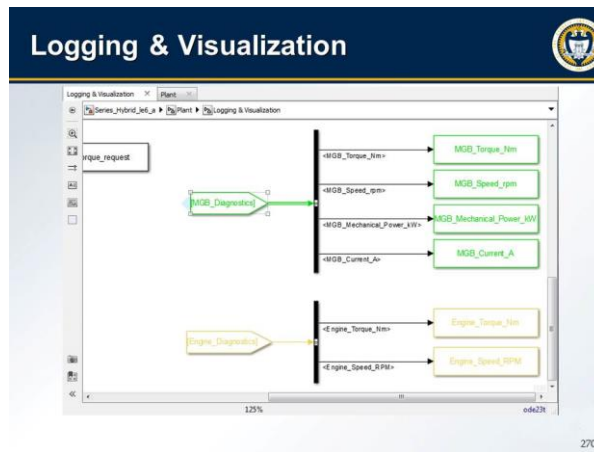
267

- ### Engine
- Return to the Plant level of the model
  - Connect a System Signals **From** block
  - Place the Engine Diagnostics on the CAN Bus
  - Connect the Engine mechanical port to the MGB mechanical port
  - Go to the Logging & Visualization and extract the Engine diagnostics

268



269



270

## Engine

- Now place the Engine parameters into the init file
  - Engine\_Max\_Torque = 200 Nm
  - Engine\_Inertia = 1.5e-3 kg m<sup>2</sup>

```

19 %Battery Parameters
20 %Battery Voltage = 136;
21 %Battery Capacity = 8.8;
22 %Battery_Initial_SOC = 0.7;
23
24 %Engine Parameters
25 %Engine_Max_Torque = 200;
26 %Engine_Inertia = 1.5e-3;
27
28 %Controller Parameters
29 %battery_soc_low = 0.6;
    
```

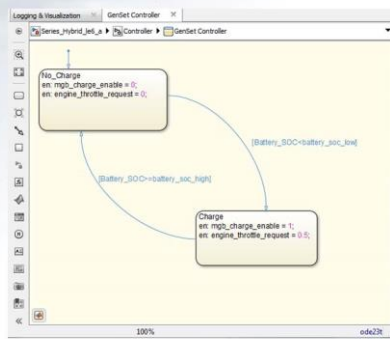
271

## Engine / Controller

- To keep things simple
  - No Charge State
    - MGB charge enable = 0
    - Engine throttle request = 0
  - Charge State
    - MGB charge enable = 1
    - Engine throttle request = 0.5
- Add the new engine throttle data to the Stateflow Controller
- Place the engine throttle signal on the bus

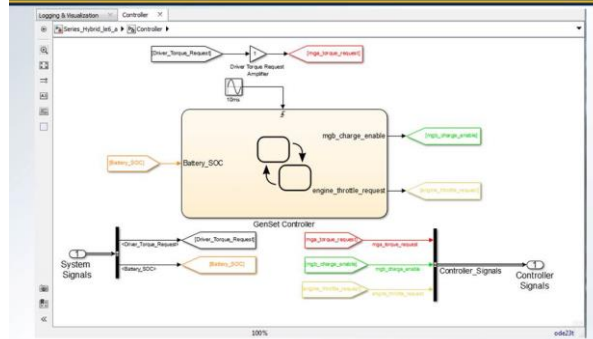
272

## Controller



273

## Controller



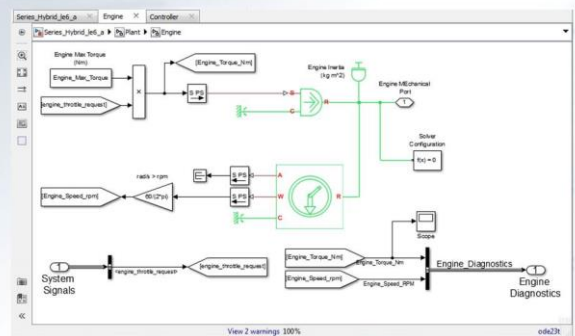
274

## Engine / GenSet

- Return to the Engine subsystem and extract the throttle request signal
- Excellent, the first pass Engine and associated improved GenSet is complete
- Drag in a **Solver Configuration** block and connect the block as shown in the next slide

275

## Engine



276



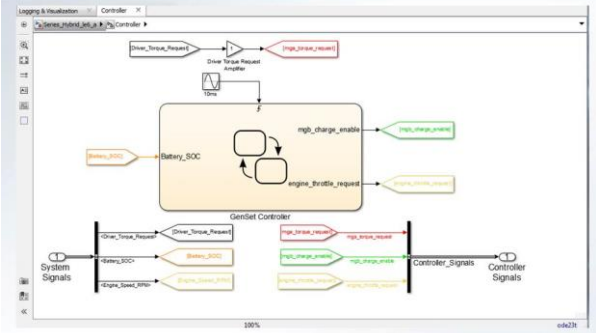
## Controller



- The controller will need to send a torque request to MGB to hold the engine at a constant desired speed
  - Engine too fast – increase torque
  - Engine too slow – decrease torque
- Go to the Controller level of the model
- Extract the engine speed signal

277

## Controller



278

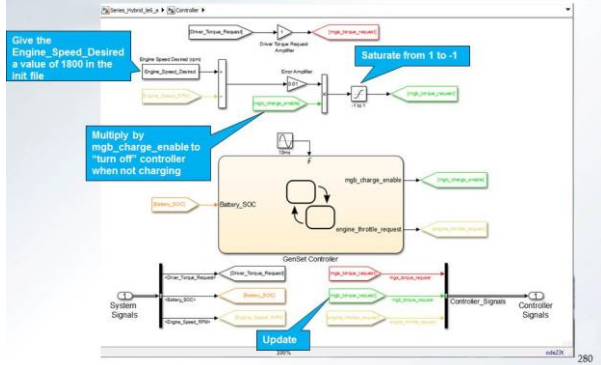
## Controller



- As done with the Driver subsystem
  - The actual engine speed will be subtracted from the desired speed
  - This will create an error signal
  - The error signal will be amplified
  - The amplified signal will be saturated
  - The resulting torque request will be sent to MGB
- Implement this feedback controller
- Update the Control Signals bus

279

## Controller



280

## MGB and Visualization



- Go to the MGB subsystem and extract the mgb\_torque\_request signal
- Go to the Logging & Visualization subsystem
  - Right click on the Vehicle\_Speed\_mph signal
  - Open the Scope
  - Add the MGB\_Current\_A signal to the third axis
  - Create a fourth axis and put the Engine\_Speed\_RPM signal on it

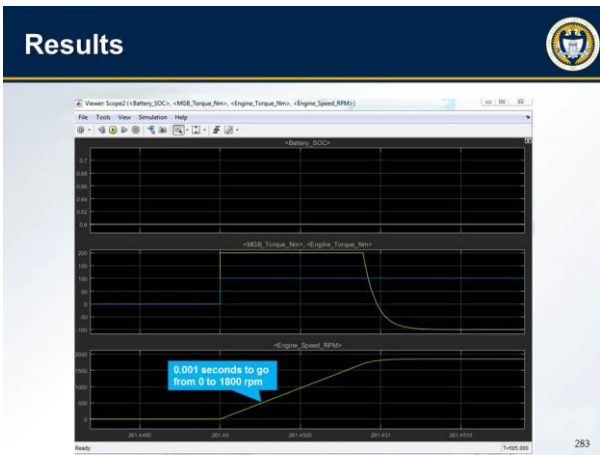
281

## Visualization



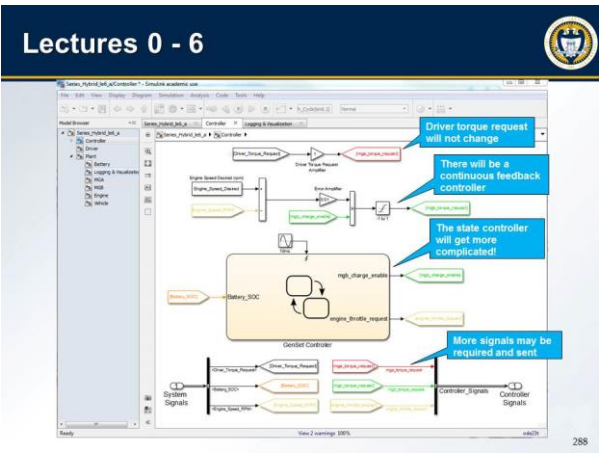
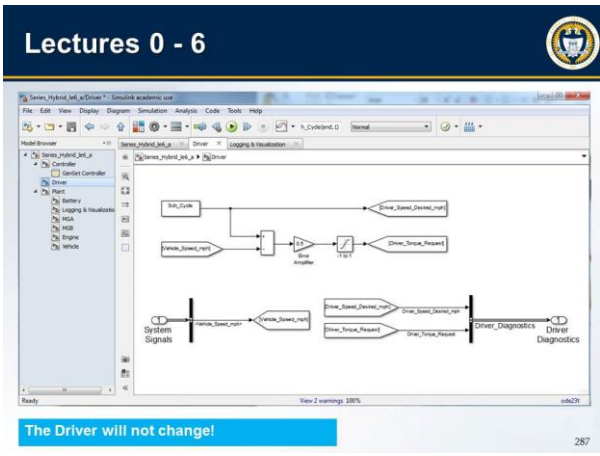
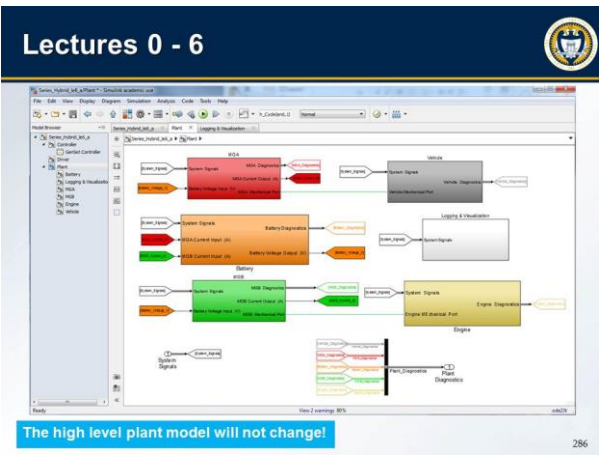
- Create a new scope
  - Axis 1 : Battery SOC
  - Axis 2 : Engine Torque and MGB Torque
  - Axis 3 : Engine Speed
- Run the FU505 cycle to save time


282



- ### Improved GenSet & Controls
- Review
    - Made MGB a duplicate of MGA
    - Created a very simple diesel engine
    - Updated the controller to intermittently obtain current from the GenSet

- ### Lectures 0 - 6
- We built a model of a series electric vehicle
    - Plant
    - Driver
    - Controller
  - All of the plant components are crude
    - System response could be intuited
    - Complex system was established
  - Component refinement is next
    - Controller will need to be updated





# MBSD Lecture 7

## Improved Engine Model and MPGGE SOC Fuel Efficiency

Georgia Tech The George W. Woodruff School of Mechanical Engineering

## Outline

- Improve Engine model
  - Experimental data for torque curves
  - Experimental data for fuel consumption
  - Controller refinements
- Fuel Efficiency
  - Calculate MPG, MPGGE, and MPGGE with SOC correction

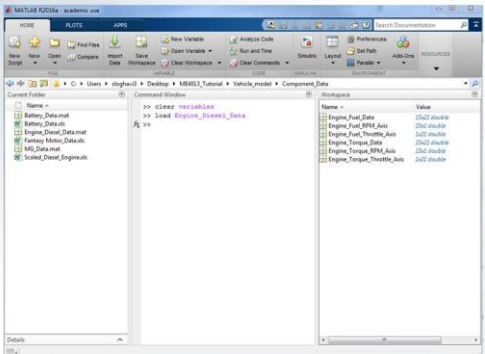
## Experimental Data

- Using the engine dynamometer at Rose-Hulman, a small diesel engine was tested at various engine speeds and throttle positions
- Data for torque (Nm) and fuel consumption (grams/sec) were collected
- Let's add this to our model
- Rename your model
  - `Series_Hybrid_le7_a.slx`

## Experimental Data

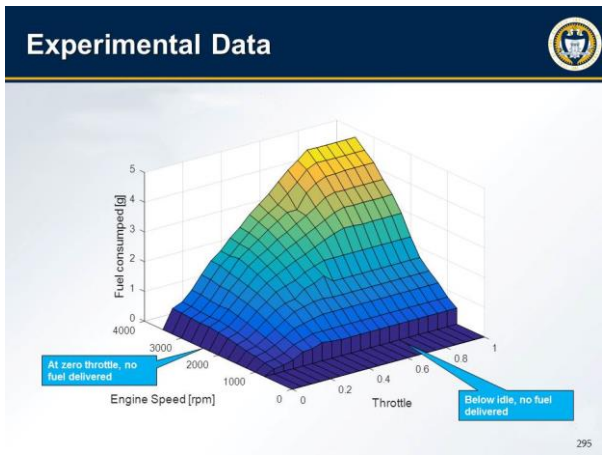
- In the Matlab Window, go to
  - Current Directory window
  - Double click on the Component\_Data folder
- At the command line, type
  - `clear variables`
  - `load Engine_Diesel_Data`
- The loaded variables will appear in the Workspace window

## Experimental Data

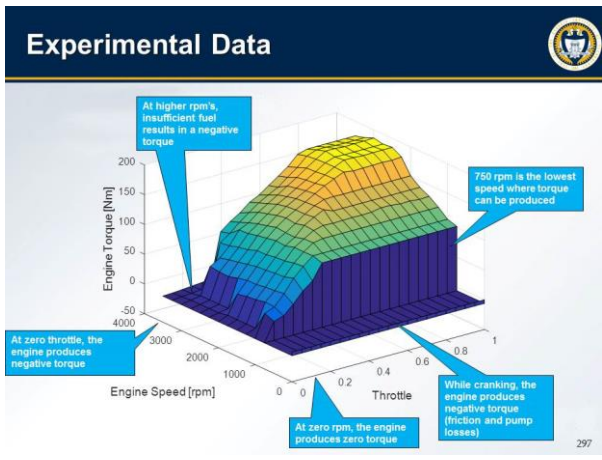


## Experimental Data

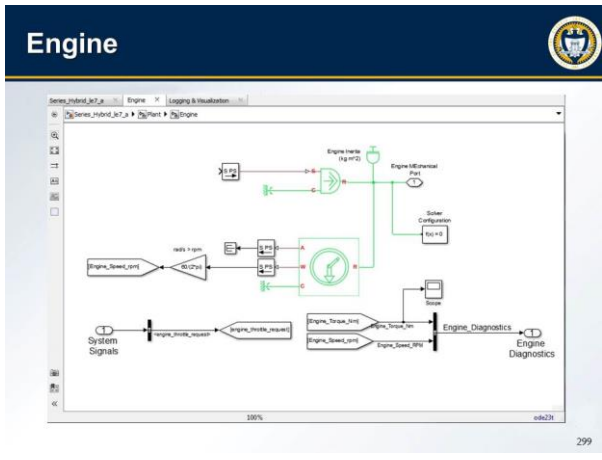
- Note the size of the matrices
  - `Engine_Fuel_Data` : 15x21
  - `Engine_Fuel_RPM_Axis`: 15x1
  - `Engine_Fuel_Throttle_Axis`: 1x21
- Thus
  - The data rows are for given rpms
  - The data columns are for given throttle positions
- At the command line type
  - `surf(Engine_Fuel_Throttle_Axis, Engine_Fuel_RPM_Axis, Engine_Fuel_Data)`



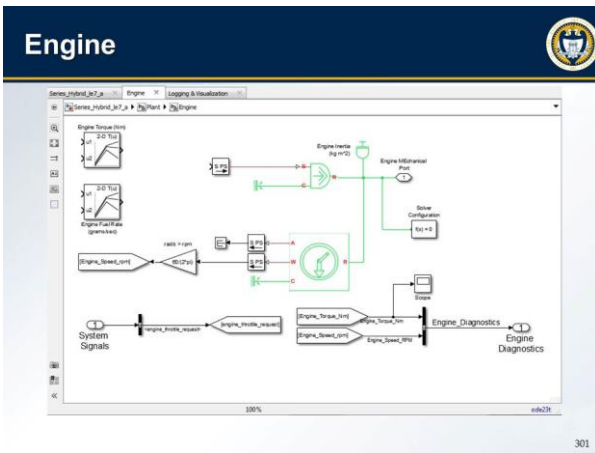
- ### Experimental Data
- This engine does not have an idle circuit
    - 750 rpm is the lowest it can run
    - A throttle signal of zero will send no fuel
    - It will stall out
  - This is great!
    - Our controller completely rules the throttle signal
  - Modify the previous surf command to view the torque data
- 296



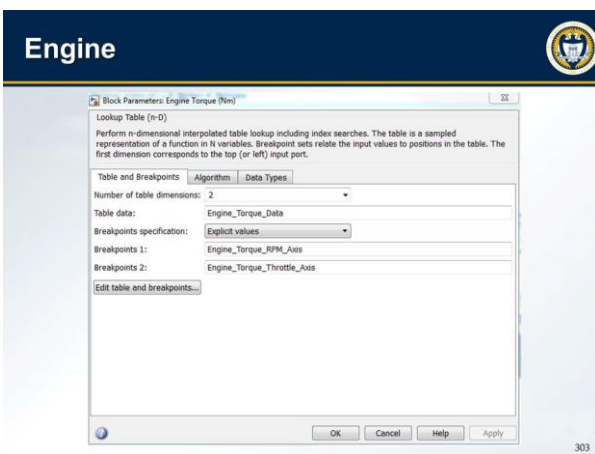
- ### Engine
- We will need to read in this data with a 2-d lookup table
  - Go to the Engine subsystem and delete the
    - Engine\_Max\_Torque Constant
    - engine\_throttle\_request From
    - Product block
    - Engine\_Torque\_Nm Goto
- 298



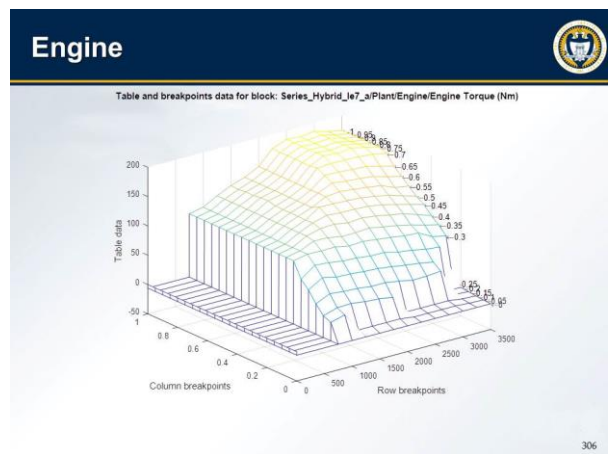
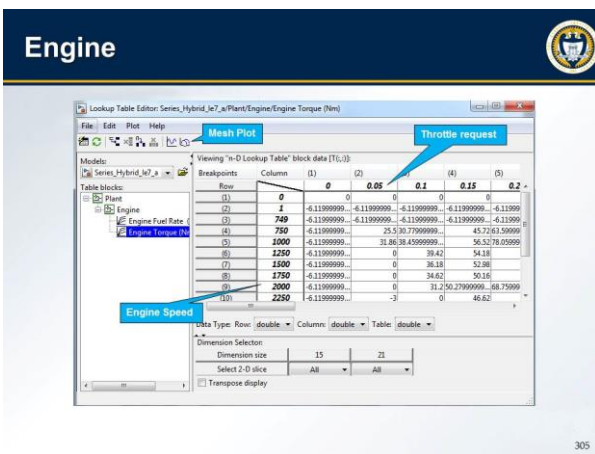
- ### Engine
- From
    - Simulink / Lookup Tables
  - drag in two
    - 2-D Lookup Table blocks
  - Rename them
    - Engine Torque (Nm)
    - Engine Fuel Rate (grams/sec)
- 300



- ### Engine
- Double click on the Engine Torque table
  - In the dialogue box, change
    - Table data: Engine\_Torque\_Data
    - Breakpoints 1: Engine\_Torque\_RPM\_Axis
    - Breakpoints 2: Engine\_Torque\_Throttle\_Axis
  - Click OK
  - Note the shape of the table
    - It looks like the data!
  - Repeat for the Fuel Rate table
- 
- 302



- ### Engine
- Double click on the Engine Torque table
  - Click the Edit table and breakpoints... button
    - Note that all the data shows up
  - Click on the Mesh Plot icon
  - Verify that the shape is correct
    - Easy to make a mistake, especially with square matrix data!
  - Close the mesh plot and the Editor
  - Click Cancel on the dialogue box
- 304

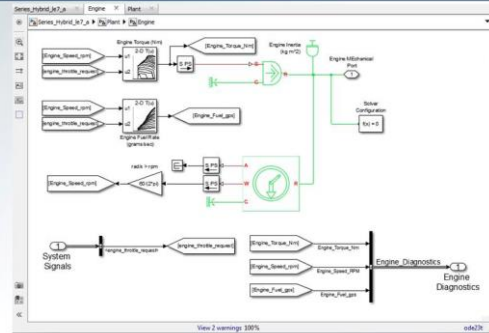


## Engine & Logging

- In the Engine subsystem, route
  - An engine rpm signal to each table
    - Top input
  - A throttle request signal to each table
    - Bottom input
  - The torque output to the **Simulink-PS Converter**
  - The fuel rate output to the Engine Diagnostics bus
  - Put a **Goto** block on the Engine Torque signal: **Engine\_Torque\_Nm**
- In the Logging & Visualization subsystem
  - Extract the engine fuel rate signal

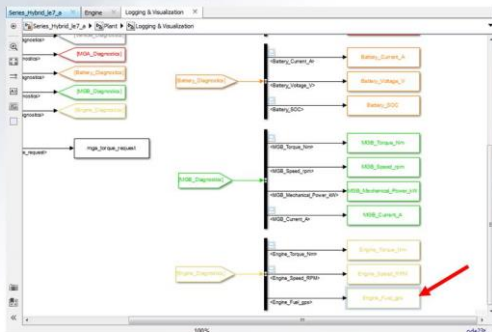
307

## Engine & Logging



308

## Engine & Logging



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## Engine & Initialization

- In the init file, add the line
  - `load Component_Data/Engine_Diesel_Data.mat`
- Update the engine inertia to 0.12
  - The crankshaft and flywheel

```

EDITOR  Path:  VIEW
C:\Users\logan\Desktop\ME4013_Tutorial\Vehicle_model\Vehicle_Initial_Files\
- Battery Parameters
- Battery_Voltage = 336;          %Battery Voltage, V
- Battery_Capacity = 6.5;       %Battery Max Capacity, Ahr
- Battery_initial_SOC = 0.7;    %Battery initial SOC

%Engine Parameters
- load Component_Data/Engine_Diesel_Data.mat
- Engine_Inertia = 0.12;        %Engine Inertia, kg m^2
- Engine_Speed_Desired = 1800; %Desired Engine Speed, rpm

%Controller Parameters
script [Ln 25 Col 43]
    
```

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## Directory

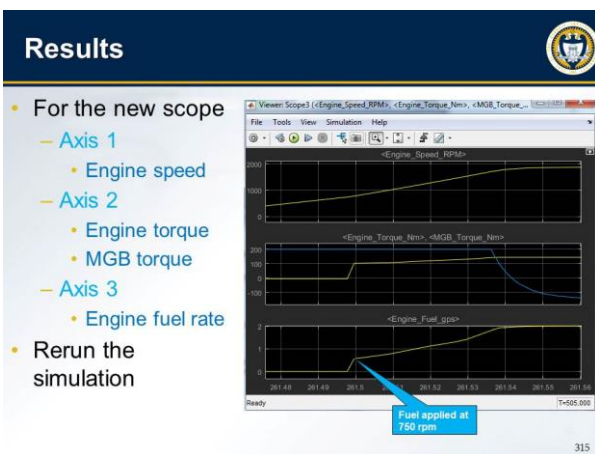
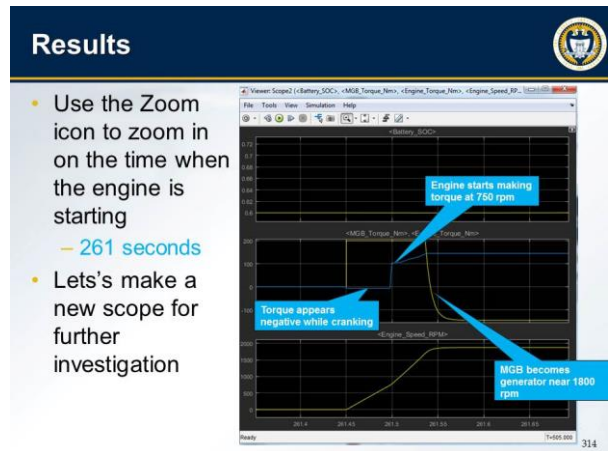
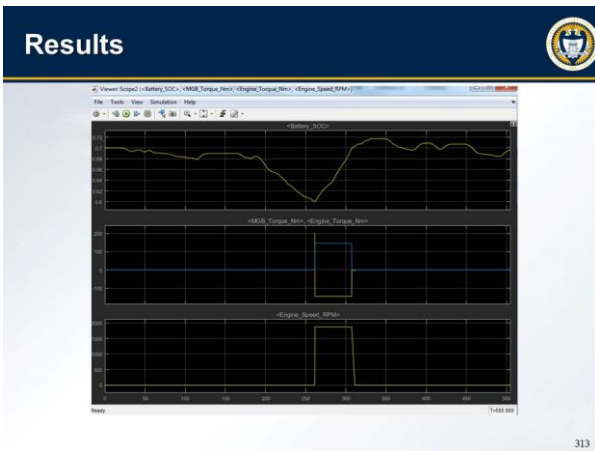
- Recall that we used the Matlab window to browse into the Component\_Data directory
- Matlab MUST be in the same working directory as the Simulink model
- In the Matlab window
  - Click on the Go up one level icon
- Run the Simulink model

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## Results

- The results look very similar to those previous
  - The model still appears to be responding correctly
  - At 1800 rpm and 0.5 throttle, the engine torque should be about 150 Nm
  - Verify

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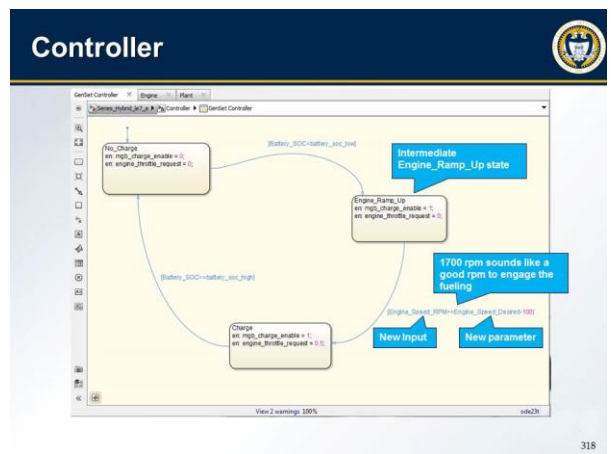


### Results & Controller

- We don't want the engine to start making power until it is closer to the operating point
- An additional logic state will be needed
  - Set mgb\_charge\_enable to 1 to start engine
  - Set engine\_throttle\_request to 0 until "close" to operating rpm
- Then go into the Charge state

### Controller

- Go to the Controller subsystem
  - Open the Stateflow GenSet Controller
  - Add this new state
    - Engine\_Speed\_RPM is a Simulink input
    - Engine\_Speed\_Desired is a parameter
  - What is a good value for "close" to 1800?



## Controller

Model Explorer showing the GenSet Controller block highlighted in red. The block is located under the Controller hierarchy.

## Controller

Block diagram of the GenSet Controller. It shows inputs for Battery\_SOC, Engine\_Speed\_RPM, and mgb\_charge\_enable. The controller outputs mgb\_charge\_enable, engine\_fuel\_request, and engine\_fuel\_request. A 'New Input' label points to the Engine\_Speed\_RPM input.

## Results

- X-axis zoom in on the engine ramp up
- Y-axis zoom in on the engine speed
- The fuel rate is about 90 rpm late
  - Why?

Three zoomed-in plots showing engine speed, engine torque, and engine fuel rate during a ramp-up. Annotations indicate 'Fuel should start flowing' and 'Fuel starts about 90 rpm late'.

## Results

- The ridiculous MGB starts the engine far too quickly
  - The transition guards are checked every 10 ms
  - In 10ms the engine speed goes from 1500 to almost 1800 rpm
  - The transition point was almost missed!
- Engine ramp-up rate will be addressed after we improve the MGB model

## Results

- X-axis zoom in on the engine coast down
- Everything here looks fine
  - Fuel shuts off
  - Engine coasts down in about 4 seconds

Three zoomed-in plots showing engine speed, engine torque, and engine fuel rate during a coast-down. The fuel rate drops to zero as the engine speed decreases.

## MPG

- Let's now calculate the vehicle fuel efficiency
  - MPG
  - SOC correction
  - GE (gas equivalent)
- We will do this with a post-processing file
  - Simply for experience
  - Could be done in the Logging & Visualization subsystem



## MPG



- Calculating MPG is easy!
  - Distance traveled / gallons consumed
- We already have a signal for distance traveled in miles
- We need a fuel consumed signal
  - Add an **Integrator** into the Engine subsystem to calculate the amount of fuel consumed in grams
  - Add the signal to the Engine Diagnostics bus
  - Extract it in the Logging & Visualization
- All of the remaining calculations are done in post file

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## Engine & Logging



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## SOC



- If we removed energy from the battery during the drive cycle, we must calculate the additional amount of fuel required to replace it
- If we added energy to the battery, we must calculate the amount of fuel required to produce that energy and subtract the fuel from our total

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## SOC



- Step 1 – Amp Hours Consumed
  - $Ah\_Consumed = (Initial\_SOC - Final\_SOC) * Battery\_Capacity$ 
    - Battery capacity is in Ah (8.5 Ah for our model)
- Step 2 – Convert to Energy
  - $Electrical\_Energy = Ah\_Consumed * Battery\_OCV$ 
    - OCV (open circuit voltage) is determined experimentally from V-I plots (366 V in our model) and changes with the SOC. Energy is in Watt-hr.

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## SOC



- Step 3 – Conversion Efficiency
  - $Electrical\_Energy\_BTU = (3.412 * Electrical\_Energy) / Efficiency$ 
    - Conversion efficiency is also determined experimentally by ANL (25% approximation generous)
    - The 3.412 converts Watt-hr to BTU
- Excellent – we now know the energy in BTU needed to account for the charge added to or removed from the battery

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## MPGGE



- As a standard for comparison, we employ the MPGGE with SOC correction
- This gives the equivalent amount of RFG required to produce the same energy used in the drive cycle
- We already know the BTUs of energy needed to correct the SOC

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### Outline

- Improve Battery model
  - First order math model
    - Variable battery voltage
  - Algebraic loops
  - Variable open circuit voltage
  - Variable charge/discharge resistance
  - Numeric error

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### Simple Battery

- Thus far, the battery model
  - Has had a constant voltage of 336V
  - Has had a constant OCV of 366V
  - Been able to provide infinite current
  - Has had no energy conversion losses
- Let's now improve the battery model by including the internal resistance
 
$$V_{BAT} = V_{OC} + I_{BAT} \times R_{Series}$$
  - The battery voltage is now a function of the current

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### Battery Terminal Voltage

- Current is defined as positive into the battery
- Positive current charges the battery and increases the battery SOC

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### Future Improvements

- For now, the open circuit voltage  $V_{OC}$  and the battery series resistance  $R_{Series}$  are a constant
- As our understanding of the model increases, we can make the battery model less ideal by:
  - Making  $V_{OC}$  a function of SOC and temperature
  - Having a different charge and discharge series resistances, both of which are a function of SOC and temperature
- These will all involve experimental (or manufacturer) data and 2-d lookup tables

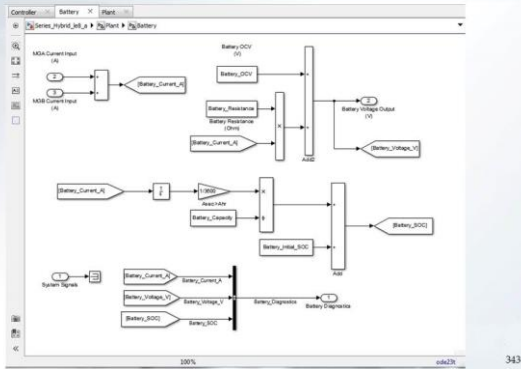
341

### First Order Model

- The battery OCV and series resistance will be constants read from the init file
  - Battery\_OCV = 366
  - Battery\_Resistance = 0.4
- Add to the init file
- Remove the Battery\_Voltage
- Remove the OCV from the post file
- Rename your model `Series_Hybrid_1e8_a.slx` and implement the battery voltage equation

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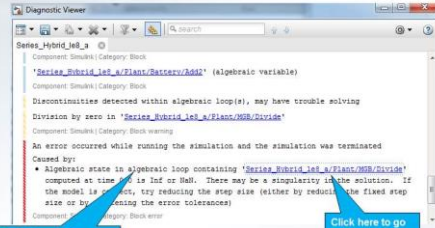
## First Order Model



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## Results

- Run the Simulation – Dooh!

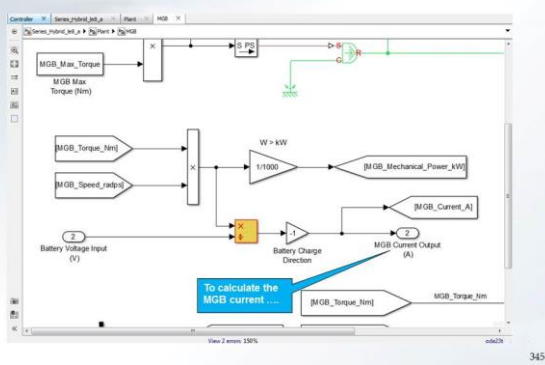


Algebraic Loop: This means a variable is a function of itself.

Click here to go to the problem

344

## Algebraic Loop



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## Algebraic Loop

- Aha! The loop involves the battery voltage which is a function of the battery current which is a function of the battery voltage ...
- To break the algebraic loop a delay will be used to have either
  - The battery voltage signal lag the battery current calculation or
  - The battery current signal lag the battery voltage calculation
- Close the Simulations Diagnostics box

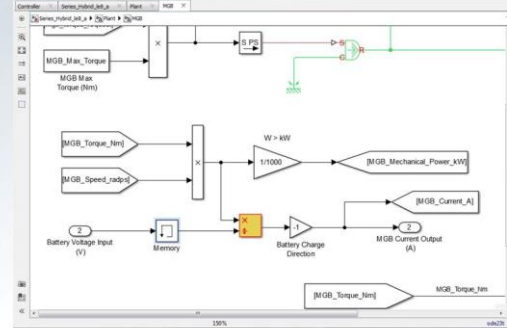
346

## Algebraic Loop

- Go to the MGB subsystem where the algebraic loop was first identified
- From
  - Simulink / Discrete
- Drag in a **Memory** block
- Insert it just after the Voltage Input block
- Thinking ahead, MGA is also used to calculate the battery current
- Put a **Memory** block there too

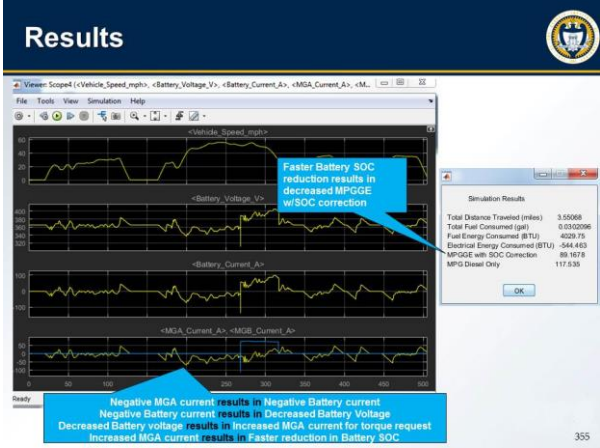
347

## MGA/MGB Voltage Delay

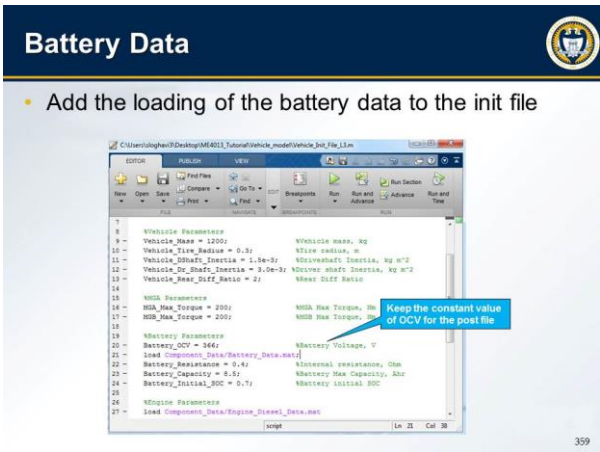
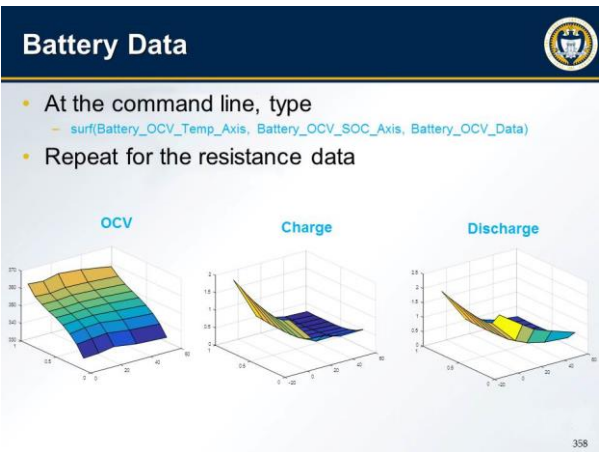
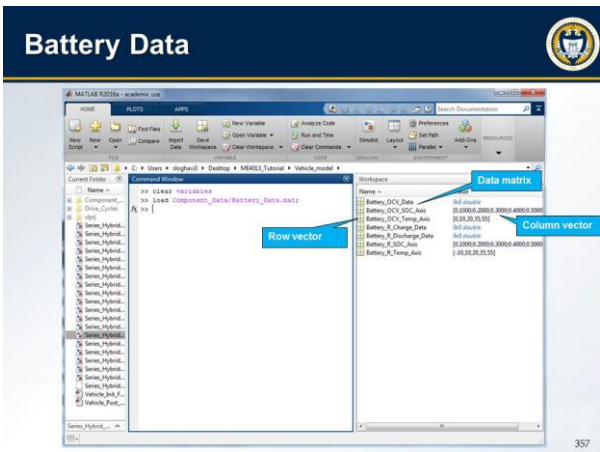


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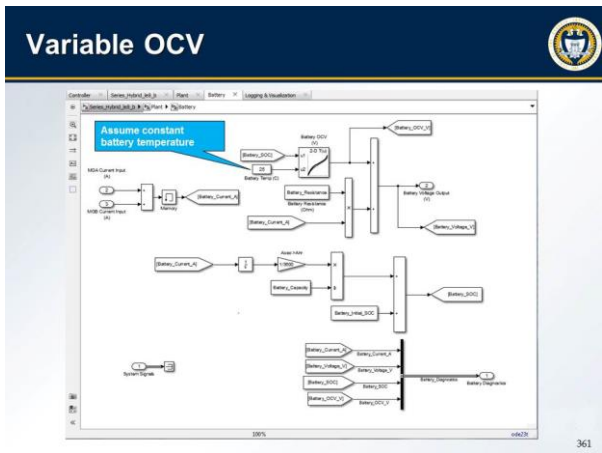




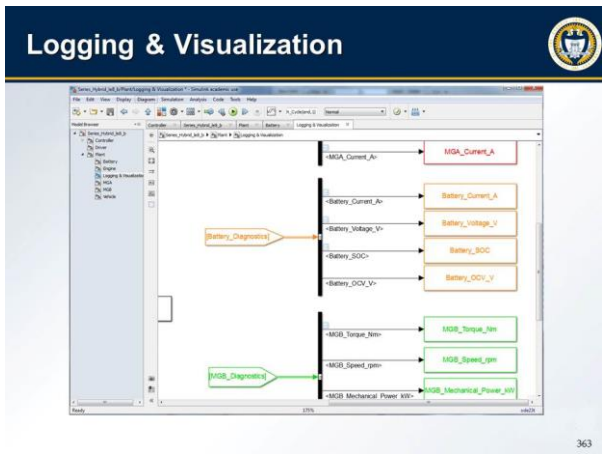
- ## Results
- This simple improvement has substantially decreased our SOC corrected MPGGE
  - Let's improve the model further and see how the fuel economy is affected
  - First, we will add a variable OCV as a function of SOC and temperature
  - At the Matlab command line type
    - clear variables
    - load Component\_Data/Battery\_Data.mat;
    - Click on the Workspace tab



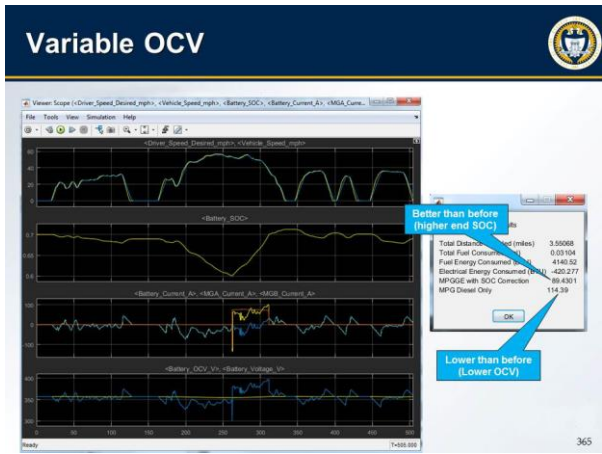
- ## Variable OCV
- Save your model as
    - Series\_Hybrid\_1e8\_b.slx
  - Go to the Battery subsystem
  - Delete the Battery OCV Constant block
  - Drag in a
    - 2-D Look up Table block
    - Constant block
  - Add the OCV signal to the Battery Diagnostics Bus and extract it in the Logger



### Variable OCV



- ### Variable OCV
- We've got a lot of scopes – delete them all
  - Create a new one
    - Axis 1 – Vehicle Speed (Desired & Actual)
    - Axis 2 – Battery SOC
    - Axis 3 – Battery, MGA, & MGB current
    - Axis 4 – Battery OCV and Voltage
  - Run the simulation



- ### Variable Internal Resistance
- The internal resistance of the battery is also a function of the temperature and SOC
  - It also varies for charging and discharging
  - Drag in
    - A Subsystem block
  - And rename it
    - Battery Internal Resistance (Ohms)

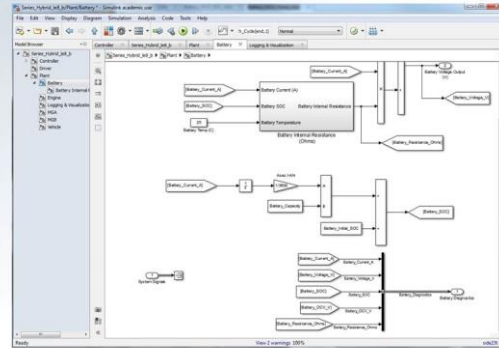
## Variable Internal Resistance



- It will have three inputs
  - Battery Current
  - Battery SOC
  - Battery Temperature
- Add one output
  - Battery Internal Resistance
- Put the Battery Resistance on the Diagnostics Bus and extract it in the Logger

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## Variable Internal Resistance



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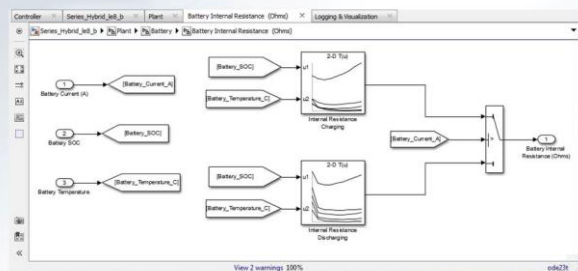
## Variable Internal Resistance



- We will use the current to tell if we are charging or discharging
- From
  - Simulink / Commonly Used Blocks
- Drag a **Switch** block into the Battery Resistance subsystem
- Also drag in
  - two 2-D Lookup Table blocks
  - three Goto and five From blocks

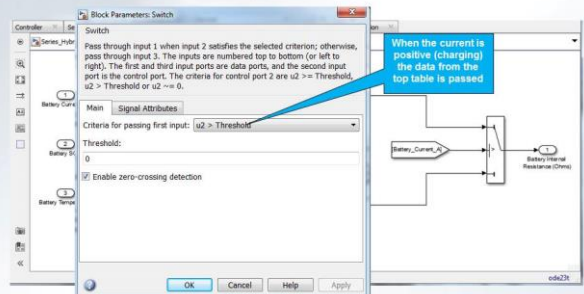
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## Variable Internal Resistance



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## Variable Internal Resistance

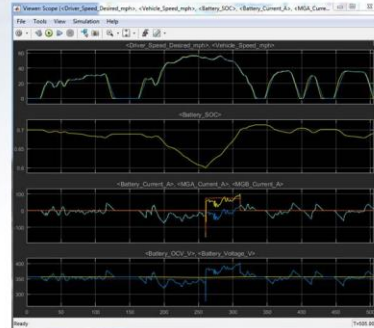


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## Variable Internal Resistance

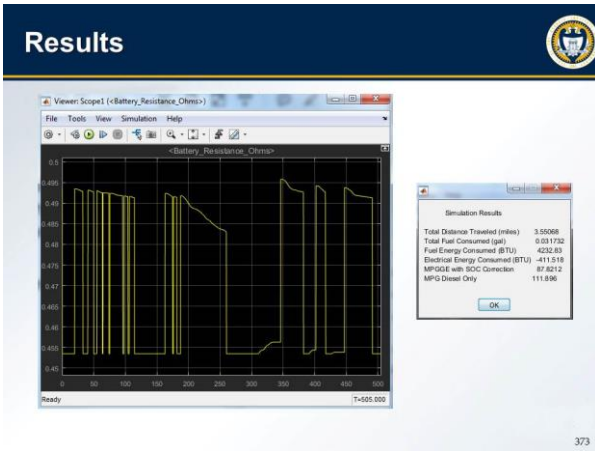


- Put the resistance on a new scope and run the simulation



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- ## Results
- Higher internal resistance lowered the fuel economy
  - Run the FU505\_Ten\_Times.mat drive cycle



- ## Review
- Improved the battery model
    - Used first order model battery model to incorporate internal resistance, making battery voltage a function of battery current
    - Discovered and corrected an algebraic loop with a memory block
    - Improved the model with open circuit voltage data based on SOC and battery temperature
    - Improved the model with variable charge and discharge internal resistances based on SOC and battery temperature

# MBSD Lecture 9

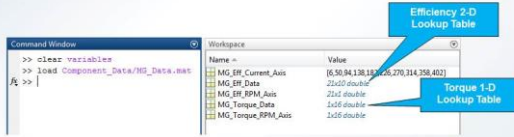
## Improved MG Model

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- ## Outline
- Improved Motor/Generator model
    - Variable torque
    - Constant conversion efficiency
    - Variable conversion efficiency

## MG Data

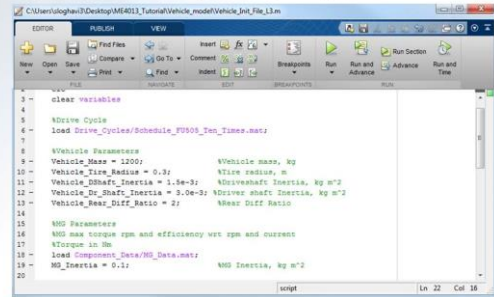
- At the Matlab command line type
  - clear variables
  - load Component\_Data/MG\_Data.mat



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## Initialization

- Add the MG data loading to the init file
- Also add an MG inertia of 0.1 kg m<sup>2</sup>

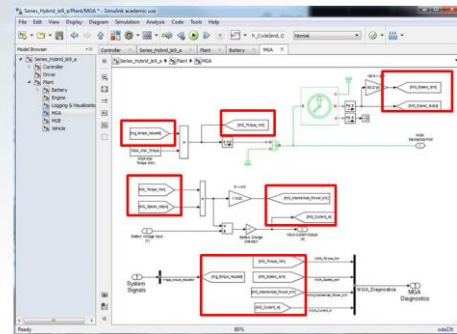


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## Variable Torque

- We will start with the MGA subsystem
  - Once we have it refined, we will copy it to MGB
- Rename your model
  - Series\_Hybrid\_1e9\_a.slx
- Go to the MGA subsystem
  - Change the name of all Goto and From blocks to MG

## Variable Torque



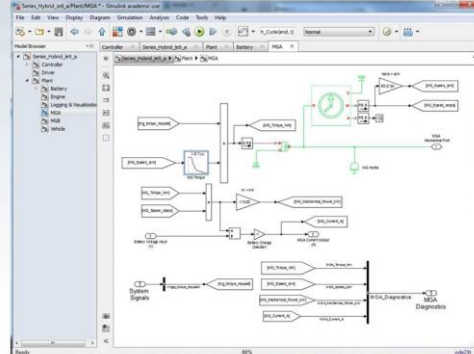
381

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## Variable Torque

- Drag in
  - An Inertia block
  - A 1-D Lookup Table block
- Delete the MGA\_Max\_Torque block
- Arrange as shown on the next slide
- Add the
  - MG\_Torque\_RPM\_Axis
  - MG\_Torque\_Data
- to the Lookup Table

## Variable Torque



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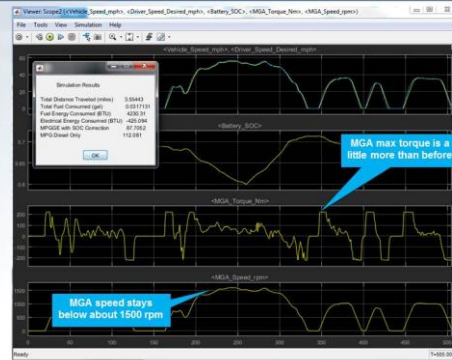
## Results



- Let's create a new scope
  - Axis 1 – Vehicle speed (desired and actual)
  - Axis 2 – Battery SOC
  - Axis 3 – MGA Torque
  - Axis 4 – MGA Speed rpm
- Run the simulation for the FU505 drive cycle

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## Results



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## Efficiency



- Electromechanical energy conversion can go two ways
  - Electrical to mechanical
    - Accelerating the vehicle
  - Mechanical to Electrical
    - Regenerative braking
- This is a little more challenging than the battery efficiency
- We will start by assuming a constant conversion efficiency of 0.85

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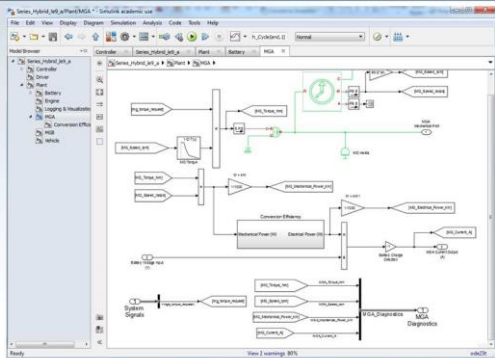
## Efficiency



- Drag a **Subsystem** block into MGA and
- Rename
  - Subsystem to Conversion Efficiency
  - In1 to Mechanical Power (W)
  - Out1 to Electrical Power (W)
- Add a **Gain** and a **Goto** for the Electrical Power

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## Efficiency



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## Efficiency



- Go to the Conversion Efficiency subsystem
- Drag in
  - A **Constant** block
  - A **Switch** block
  - Two **Product** blocks
  - A **Goto** block
  - Two **From** blocks

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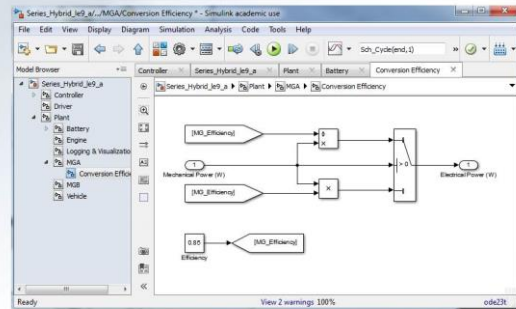
## Efficiency



- If the mechanical power is positive
  - Top of switch
  - The motor is accelerating the vehicle
  - A higher electrical power is required
  - Therefore divide MP by the efficiency
- If the mechanical power is negative
  - Bottom of switch
  - The generator is making power
  - A lower electrical power is made
  - Therefore multiply MP by the efficiency
- Arrange as shown on the next slide

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## Efficiency



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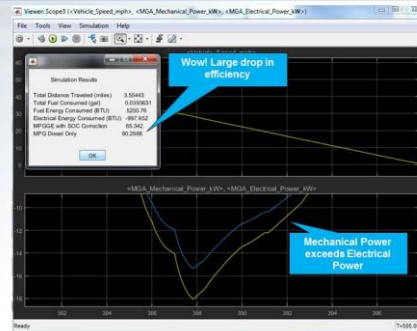
## Efficiency



- Add the Electrical Power to the Diagnostics bus
- Extract it in the Logging & Visualization subsystem
- Create a scope
  - Axis 1 – Vehicle Speed (actual)
  - Axis 2 – Mechanical and Electrical Power

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## Results



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## Variable Efficiency



- Conversion efficiency is a function of both MG speed and current
- Let's add that in
- Rename your model
  - Series\_Hybrid\_le9\_b.slx
- In the Conversion Efficiency subsystem
  - Delete the Efficiency block and Goto/From blocks

395

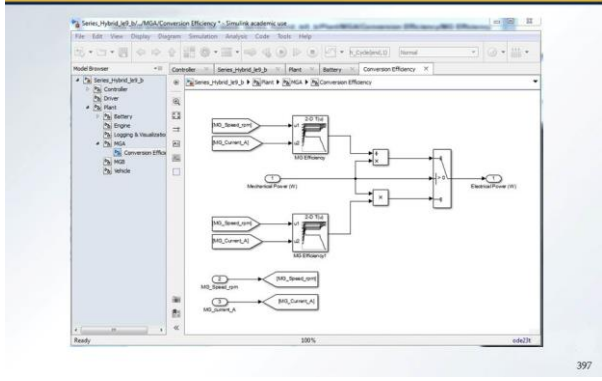
## Variable Efficiency



- Drag in
  - Two 2-D Lookup Table blocks
  - Two In1 blocks
  - Two Goto blocks
  - Four From blocks
- Arrange and label as shown on the next slide

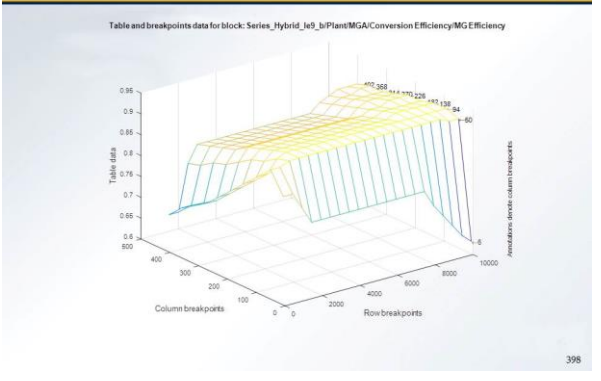
396

# Variable Efficiency



397

# Variable Efficiency



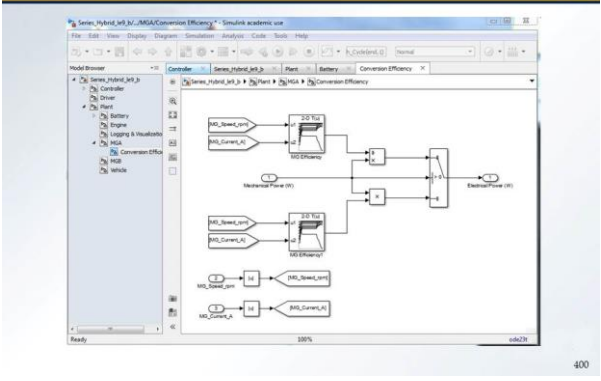
398

# Variable Efficiency

- Hmm, our efficiency data is for positive values of current and MG speed
  - Can either of these be negative?
- From
  - Simulink / Math Operations
- Drag in two Absolute Value blocks
- Return to the MGA subsystem
  - Connect the necessary From blocks
- Run the simulation

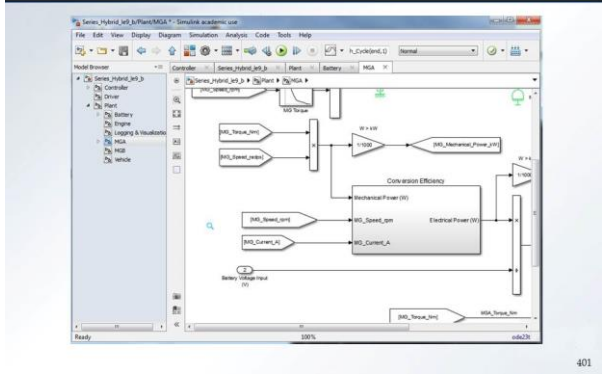
399

# Variable Efficiency



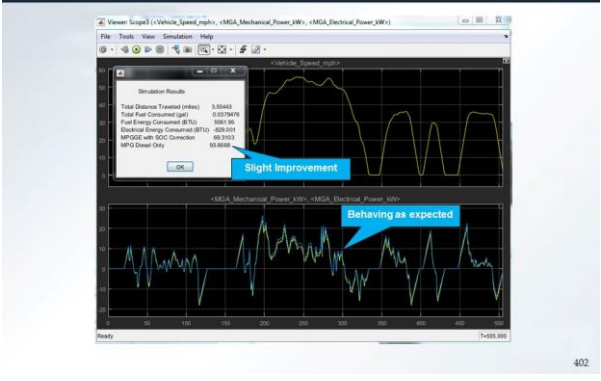
400

# Variable Efficiency



401

# Variable Efficiency



402

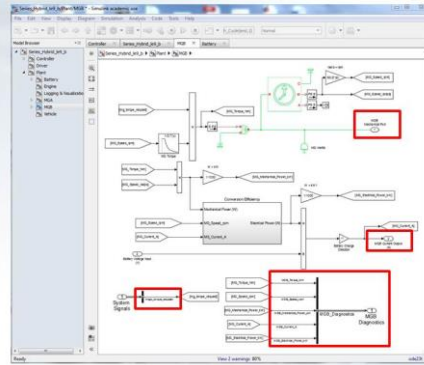
## Variable Efficiency



- Outstanding, we now have a pretty good model of a motor/generator
- Go to the Plant level of the model
  - Delete MGB
  - Copy MGA
  - Update the names accordingly
- Extract the MGB electrical power in the Logging & Visualization subsystem

403

## MGB



404

## Results



- One last new scope
  - Axis 1 – Vehicle speed (actual and desired)
  - Axis 2 – SOC
  - Axis 3 – Battery, MGA, & MGB current
  - Axis 4 – Engine rpm
- Run the Simulation

405

## Results



406

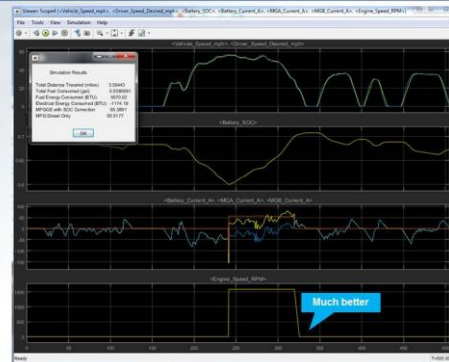
## Results



- Aha – our desired engine speed is 1800 rpm
  - Engine torque at 0.5 throttle 145 Nm
  - MGB max torque 140 Nm
- MGB can't control the engine!
- Decrease the desired engine speed to 1500 rpm

407

## Results



408

## Review



- Improved MG Model
  - Variable maximum torque wrt MG speed
  - Constant efficiency to explore electromechanical conversion losses
  - Variable efficient wrt MG speed and current

409

## Future Direction



- We blew up our battery
  - Need to add current limits
- We blew up our engine
  - Need to add an emergency defuel at redline engine speeds
- We are destroying the engine
  - Need to limit MGB speed for starting engine
- We haven't done that much control work
  - Other ways to control the system?

410

## 2) Dynamic Programming Matlab code

```
% Dynamic Programming HEV script
%Written by Saeid Loghavi, August, 2015
clc
clear all
close all
tic
global max_engine_RPM
global radius;
global i_final;
global gear_ratios;
global EM_gear_ratio;

global Engine_Fuel_Data;
global Engine_Fuel_RPM_Axis;
global Engine_Fuel_Throttle_Axis;
global Engine_Torque_Data;
global Engine_Torque_RPM_Axis;
global Engine_Torque_Throttle_Axis;

% fuel consumption rate (next 3 lines taken from Rose-Hulman tutorial)
load Engine_Diesel_Data.mat; % This is NOT BSFC, but instead grams/sec
as a function of throttle request and engine RPM
engine_max_torque = 200; % Engine Max Torque, Nm
engine_inertia = 0.12; % Engine Inertia, kg m^2
max_engine_RPM = max(Engine_Fuel_RPM_Axis);

% Vehicle data based on a 2011 VW Jetta
mass = 1389; % mass in kg (2011 Prius)
g = 9.81; % accel of gravity in m/s^2
Area = 2.642; % frontal area in m^2 (guessed)
rho_air = 1.25; % density of air in kg/m^3
Cd = 0.32; % drag coefficient (guessed)
delta_M = 1.2; % mass factor
fr = 0.013; % rolling resistance coefficient
radius = 0.25; % displaced tire radius, m
eta = 0.97; % transmission efficiency
% http://www.bebo.com/new-cars/2011-volkswagen-jetta-sportwagen/specs/
i_final = 3.65; % Final drive ratio
gear_ratios = [3.78 2.12 1.36 1.03 0.84]; % Gear ratios
[nnn,NUM_GEARs] = size(gear_ratios);

% Additional Hybrid data (made up, and from http://www.eaa-
phev.org/wiki/Toyota\_Prius\_Battery\_Specs)
battery_capacity = 1.5; % energy capacity in kW-h

SOC_low_range = 0.4; % low desirable operating range for battery
SOC_high_range = 0.8; % high desirable operating range for battery
SOC_initial=0.8;

SOC_final_low = 0.7; % low SOC at final step (otherwise penalized)
SOC_final_high = 0.8; % high SOC at final step (otherwise penalized)
```



```

SOC_penalty = 100000.0; % proportional penalty parameter for operating
outside of SOC desirable (or final) range
NOT_ALLOW_PENALTY = 100000;

EM_max_torque = 600; % Prius motor maximum torque is 400 Nm from Ehsani,
2nd ed for just one motor
EM_gear_ratio = 1.0; % Gear ratio between EM and the final drive

regen_efficiency = 0.60; % The efficiency of regenerative braking, taken into
account only braking at front can occur
EM_efficiency=0.6;

%
% Drive cycle
%

%load Schedule_FU505_Ten_Times.mat
load Schedule_FU505.mat;
%load Schedule_Boston_Cab.mat;

[NUM_STEPS,n] = size(Sch_Cycle);
t = Sch_Cycle(:,1);
t_final = Sch_Cycle(NUM_STEPS,1);

% Convert vehicle speed from mph to m/s and compute acceleration
v = Sch_Cycle(:,2)*0.44704;
a = diff(v)./diff(t);

% Remove last data points for velocity and time since not present in
acceleration
v(end) = [];
t(end) = [];

[NUM_STEPS,n] = size(v);

% Calculate tractive force and power at each point in the drive cycle
force = delta_M*mass*a + fr*mass*g + 0.5*Cd*rho_air*Area*v.^2;
power_required = force.*v;

%
% Establish the state and control grid points
%

state_GRID_size = 400;
SOC_grid_size=state_GRID_size;
control_GRID_size = 400;
% engine_throttle_grid_size=length(Engine_Torque_Throttle_Axis);
ICE_thr_grid_size=control_GRID_size;
thr_vec = linspace(0,1,control_GRID_size);

SOC_vector=linspace(SOC_low_range,SOC_high_range,state_GRID_size);

% Evaluate the recursive cost starting at the last step in the drive cycle

```

```

% and working backwards. The only state variable is SOC, so the cost
% function J_cost is size NUM_STEPS X SOC_grid_size.
% We also need to store the optimal control decisions at each time step.
% This is given as U_store and is size NUM_STEPS X SOC_grid_size X 3
% since we have three control variables (gear ratio, throttle request,
% EM torque)

J_cost = zeros(NUM_STEPS,SOC_grid_size);
U_store = zeros(NUM_STEPS,SOC_grid_size,3);

%
% First calculate J_cost(NUM_STEPS,:) - i.e., for all SOC in the grid at
% the last point on the drive cycle. Penalize any SOC not between
% SOC_final_low and SOC_final_high.
%
for L = 1:SOC_grid_size;
    if SOC_vector(1,L)>SOC_final_low;
        J_cost(NUM_STEPS,L)=0;
    end
    if SOC_vector(1,L)<=SOC_final_low;
        J_cost(NUM_STEPS,L)=NOT_ALLOW_PENALTY;
    end
end;
%
% Next do the cost at all other points on the drive cycle using the
% recursive statement
%
[ig1_grid, thr_grid] = meshgrid (gear_ratios, thr_vec);
for STEP=(NUM_STEPS-1):-1:1
    STEP
    ICE_RPM_choices=v(STEP)./radius.*i_final.*ig1_grid./2./pi.*60;%RPM
    choices for the time step (5 options based on gear)

    ICE_torque_choices=interp2(Engine_Torque_Throttle_Axis,Engine_Torque_RPM_Axis
    ,Engine_Torque_Data,thr_grid,ICE_RPM_choices);
    ICE_torque_choices(isnan(ICE_torque_choices))=0;

    ICE_fuel_consumption_choices=interp2(Engine_Fuel_Throttle_Axis,Engine_Fuel_RP
    M_Axis,Engine_Fuel_Data,thr_grid,ICE_RPM_choices);
    Torque_tran=force(STEP)*radius/i_final;
    P_Wheel=power_required(STEP);
    Power_ICE=ICE_torque_choices.*ICE_RPM_choices*2*pi/60;

    for L = 1:SOC_grid_size;
        EM_Analysis; % Move up, and only compute a delta_SOC
        min_cost = inf; % This keeps track of the minimum cost of using the
        three controls

        phi_cost_choices=Not_Possible*NOT_ALLOW_PENALTY;
        if STEP==1 & L~=SOC_grid_size;
            phi_cost_choices=phi_cost_choices+NOT_ALLOW_PENALTY;
        end

        J_cost_next=zeros(control_GRID_size,NUM_GEARs);
        for s=1:NUM_GEARs

```

```

J_cost_next_geari=interp1(SOC_vector,J_cost(STEP+1,:),SOC_next_possible(:,s))
;
    J_cost_next(:,s)=J_cost_next_geari;
    check=isnan(J_cost_next);
    for ch=1:5;
        for ch2=1:100;
            if check(ch2,ch)==1
                J_cost_next(ch2,ch)=NOT_ALLOW_PENALTY;
            end
        end
    end
end
end
end

Cost_possible=Possible_options.*ICE_fuel_consumption_choices+phi_cost_choices
+J_cost_next;
    % For each SOC on the grid, seek the controls gear ratio, engine
    throttle, and battery current at this time
    % step which minimize the cost from this point in the drive cycle
forward
    [M,min_cost_indicies_row]=min(Cost_possible);
    SOC_location=zeros(1,5);
    for count=1:5;
SOC_location(1,count)=SOC_next_possible(min_cost_indicies_row(count),count);
        end
        [min_cost,min_cost_indicies_column]=min(M);
        minimum_row_index = min_cost_indicies_row(min_cost_indicies_column);
        optimal_next_state(STEP,L) = SOC_next_possible(minimum_row_index,
min_cost_indicies_column);
        SOC_next_possible_indicies=interp1(SOC_vector, 1:state_GRID_size ,
SOC_next_possible,'nearest', 'extrap');

optimal_next_state_index(STEP,L)=SOC_next_possible_indicies(minimum_row_index
,min_cost_indicies_column);
    for S=1:NUM_GEARs;
        if S==1;
            if M(S)==min_cost;
                Store_Gear=S;
                Store_thr=min_cost_indicies_row(S);
            end
        end
        if S==2;
            if M(S)==min_cost & M(S-1)~=min_cost;
                Store_Gear=S;
                Store_thr=min_cost_indicies_row(S);
            end
        end
        if S==3;
            if M(S)==min_cost & M(S-1)~=min_cost & M(S-2)~=min_cost;
                Store_Gear=S;
                Store_thr=min_cost_indicies_row(S);
            end
        end
        if S==4;

```

```

        if M(S)==min_cost & M(S-1)~=min_cost & M(S-2)~=min_cost &
M(S-3)~=min_cost;
            Store_Gear=S;
            Store_thr=min_cost_indicies_row(S);
        end
    end
    if S==5;
        if M(S)==min_cost & M(S-1)~=min_cost & M(S-2)~=min_cost &
M(S-3)~=min_cost & M(S-4)~=min_cost;
            Store_Gear=S;
            Store_thr=min_cost_indicies_row(S);
        end
    end
    end
    U_store_1(STEP,L)=Store_Gear;
    U_store_2(STEP,L)=Store_thr;
    U_store_3(STEP,L)=SOC_location(Store_Gear);
    J_cost(STEP,L) = min_cost;

end;
%     U_store(STEP,L,3)=SOC_vector(L);

end;
toc
%%
close all
[global_min global_min_indices]=min(J_cost(1,:));
SOC_indices = global_min_indices;
for count3=1:NUM_STEPS-1
    O_SOC(count3)=SOC_vector(SOC_indices);
    O_Throttle(count3)=U_store_2(count3,SOC_indices);
    O_gear(count3)=U_store_1(count3,SOC_indices);
    if O_Throttle(count3)==1
        O_gear(count3)=0;
    end
    SOC_indices = optimal_next_state_index(count3,SOC_indices);
end
O_SOC(end)=SOC_vector(SOC_indices);

```

### 3) Co-simulation components

#### Configuration file:

```
<?xml version="1.0" encoding="UTF-8"?>
<CoupledMultiphysicsSimulation xmlns:cse="http://www.simulia.com/CSESchema">
  <header>
    <SchemaVersion>1.1</SchemaVersion>
  </header>
  <components>
    <component name="Abaqus/Standard">
      <bottomUpImplementation>
        <codeName>Abaqus/Standard</codeName>
        <modelDescription>
          <UnitDefinitions>
            <Unit name="kcal/kg">
              <BaseUnit factor="4186.8" m="2" s="-2"></BaseUnit>
            </Unit>
            <Unit name="degC">
              <BaseUnit K="1" factor="1.0" offset="273.15"></BaseUnit>
            </Unit>
          </UnitDefinitions>
          <ModelVariables>
            <ScalarVariable name="BODYFLUX">
              <Real unit="kcal/kg"></Real>
            </ScalarVariable>
            <ScalarVariable name="TEMP">
              <Real unit="degC"></Real>
            </ScalarVariable>
          </ModelVariables>
        </modelDescription>
      </bottomUpImplementation>
    </component>
    <component name="CFD">
      <bottomUpImplementation>
        <codeName>Abaqus/Cfd</codeName>
      </bottomUpImplementation>
    </component>
  </components>
</CoupledMultiphysicsSimulation>
```

```

<component name="BatteryCoSim_System">
  <topDownImplementation>
    <identifier>BatteryCoSim_System</identifier>
  </topDownImplementation>
</component>
</components>
<componentInstances>
  <componentInstance name="Fluidz">
    <component>CFD</component>
    <timeIncrementation>
      <lockstep>>false</lockstep>
    </timeIncrementation>
    <initialConditions>
      <sendBeforeReceive>>false</sendBeforeReceive>
    </initialConditions>
  </componentInstance>
  <componentInstance name="Module1">
    <component>Abaqus/Standard</component>
    <timeIncrementation>
      <lockstep>>false</lockstep>
    </timeIncrementation>
  </componentInstance>
  <componentInstance name="BatteryCoSim_System">
    <component>BatteryCoSim_System</component>
  </componentInstance>
</componentInstances>
<connectors>
  <connector name="CFD_to_STD_INPUT">
    <componentInstance>Module1</componentInstance>
    <variables>
      <input>
        <variable>heat_flux</variable>
        <variable>heat_capacitance</variable>
      </input>
    </variables>
  </connector>
  <connector name="CFD_to_STD_OUTPUT">
    <componentInstance>Fluidz</componentInstance>
  </connector>

```

```

<variables>
  <output>
    <variable>heat_flux</variable>
    <variable>heat_capacitance</variable>
  </output>
</variables>
</connector>
<connector name="STD_to_CFD_INPUT">
  <componentInstance>Fluidz</componentInstance>
  <variables>
    <input>
      <variable>temperature</variable>
    </input>
  </variables>
</connector>
<connector name="STD_to_CFD_OUTPUT">
  <componentInstance>Module1</componentInstance>
  <variables>
    <output>
      <variable>temperature</variable>
    </output>
  </variables>
</connector>
<connector name="STD_from_FMI">
  <componentInstance>Module1</componentInstance>
  <variables>
    <output>
      <variable>TEMP</variable>
    </output>
    <input>
      <variable>BODYFLUX</variable>
    </input>
  </variables>
</connector>
<connector name="FMI_from_STD">
  <componentInstance>BatteryCoSim_System</componentInstance>
  <variables>
    <input>

```

```

        <variable>temp</variable>
    </input>
    <output>
        <variable>bodyflux</variable>
    </output>
</variables>
</connector>
</connectors>
<connectionSets>
    <connectionSet name="CFD_to_STD" type="FIELD">
        <connection>
            <connector>CFD_to_STD_INPUT</connector>
            <connector>CFD_to_STD_OUTPUT</connector>
        </connection>
    </connectionSet>
    <connectionSet name="STD_to_CFD" type="FIELD">
        <connection>
            <connector>STD_to_CFD_INPUT</connector>
            <connector>STD_to_CFD_OUTPUT</connector>
        </connection>
    </connectionSet>
    <connectionSet name="STD_and_FMI" type="SIGNAL">
        <connection>
            <connector>STD_from_FMI</connector>
            <connector>FMI_from_STD</connector>
        </connection>
    </connectionSet>
</connectionSets>
<execution>
    <compositeActors>
        <compositeActor name="twoCodeContinuousTime">
            <actors>
                <atomicActor>Module1</atomicActor>
                <atomicActor>Fluidz</atomicActor>
                <atomicActor>BatteryCoSim_System</atomicActor>
            </actors>
            <modelOfComputation>
                <continuousTime>

```



```

        <algorithm>GAUSS-SEIDEL</algorithm>
        <negotiationMethod>MIN</negotiationMethod>
    </continuousTime>
</modelOfComputation>
</compositeActor>
</compositeActors>
<connectionGroups>
    <connectionCategory name="InitialConditions">
        <connectionSet>CFD_to_STD</connectionSet>
        <connectionSet>STD_to_CFD</connectionSet>
        <connectionSet>STD_and_FMI</connectionSet>
    </connectionCategory>
    <connectionCategory name="CouplingStep">
        <connectionSet>CFD_to_STD</connectionSet>
        <connectionSet>STD_to_CFD</connectionSet>
        <connectionSet>STD_and_FMI</connectionSet>
    </connectionCategory>
</connectionGroups>
<scenario>
    <duration>300.</duration>
</scenario>
</execution>
</CoupledMultiphysicsSimulation>

```

## Standard Model:

```

*Heading
** Job name: Module1 Model name: Model-1
** Generated by: Abaqus/CAE 2016
*Preprint, echo=NO, model=NO, history=NO, contact=NO
**
** PARTS
**
*Part, name=Module
*Node
    1, 7.94627619, 10.1000004, 64.5999985
...

```

```

25426, -0.0205669049, -10.0534954, -1.76602697
*Element, type=DC3D4
    1, 14320, 14464, 14449, 14450
...
72829, 14137, 24455, 24456, 24487
*Surface, type=ELEMENT, name=s_Surf-26
_s_Surf-26_S3, S3
_s_Surf-26_S1, S1
_s_Surf-26_S4, S4
_s_Surf-26_S2, S2
*End Assembly
*Amplitude, name=BODYFLUX, definition=ACTUATOR
**
** MATERIALS
**
*Material, name=Aluminum
*Conductivity
167.,
*Density
2700.,
*Specific Heat
902.,
*Material, name=Copper
*Conductivity
400.,
*Density
8960.,
*Specific Heat
385.,
*Material, name=Plastic
*Conductivity
0.2,
*Density
2000.,
*Specific Heat
1800.,
*Material, name=Silicon
*Conductivity

```

```

149.,
*Density
2329.,
*Specific Heat
700.,
**
** INTERACTION PROPERTIES
**
*Surface Interaction, name=Conduction
1.,
*Gap Conductance
20000.,0.
    0.,2.
**
** PREDEFINED FIELDS
**
** Name: Predefined Field-1   Type: Temperature
*Initial Conditions, type=TEMPERATURE
Set-5, 30.
**
** INTERACTIONS
**
** Interaction: Int-1
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-1, m_Surf-1
** Interaction: Int-2
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
m_Surf-3, s_Surf-3
** Interaction: Int-3
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
m_Surf-5, s_Surf-5
** Interaction: Int-4
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
m_Surf-9, s_Surf-9
** Interaction: Int-5
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
m_Surf-11, s_Surf-11
** Interaction: Int-6

```

```

*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-15, m_Surf-13
** Interaction: Int-7
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-16, m_Surf-16
** Interaction: Int-8
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-18, m_Surf-18
** Interaction: Int-9
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
m_Surf-20, s_Surf-20
** Interaction: Int-10
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-22, m_Surf-22
** Interaction: Int-11
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-24, m_Surf-24
** Interaction: Int-12
*Contact Pair, interaction=Conduction, type=SURFACE TO SURFACE
s_Surf-26, m_Surf-26
** -----
**
** STEP: Step-1
**
*Step, name=Step-1, nlgeom=NO, inc=1000
*Heat Transfer, end=PERIOD, deltmx=4.
0.1, 300., 0.003, 250.,
**
** LOADS
**
** Name: Load-1   Type: Body heat flux
*Dflux, amplitude=BODYFLUX
Set-6, BF, 500000.
** Interaction: Int-1
*Co-simulation, name=Int-1, controls=Int-1_Ctrls, program=MULTIPHYSICS
*Co-simulation Region, import, type=SURFACE
interface, CFL
interface, LUMPEDHEATCAPACITANCE

```

```

*Co-simulation Region, export, type=SURFACE
interface, NT
*Co-simulation Controls, name=Int-1_Ctrls, coupling scheme=gauss-seidel, time
incrementation=subcycle, time marks=yes,
step size=min, scheme modifier=lead
**
** OUTPUT REQUESTS
**
*Restart, write, frequency=0
**
** FIELD OUTPUT: F-Output-1
**
*Output, field, variable=PRESELECT
*Output, history, frequency=0
*End Step

```

## Fluid Model:

```

*Heading
** Job name: Fluidz Model name: Model-1
** Generated by: Abaqus/CAE 2016
*Preprint, echo=NO, model=NO, history=NO, contact=NO
**
** PARTS
**
*Part, name=Shell2-1
*Node
    1, 16.4462757, 40.3134003, -11.5000019
...
...
    1155, 15.4851465, 31.3644524, -14.9610233
*Element, type=FC3D4
    1, 578, 995, 576, 35
3412, 1002, 998, 508, 507
...
*Surface, type=ELEMENT, name=INTERFACE
_INTERFACE_S3, S3

```

```

_INTERFACE_S4, S4
_INTERFACE_S2, S2
_INTERFACE_S1, S1
*Elset, elset=_OUTLET_S3, internal, instance=Shell2-1-1
    21, 214, 479, 480, 2183
*Elset, elset=_OUTLET_S1, internal, instance=Shell2-1-1
    139, 1432
*Elset, elset=_OUTLET_S4, internal, instance=Shell2-1-1
    477, 879, 2252
*Surface, type=ELEMENT, name=OUTLET
_OUTLET_S3, S3
_OUTLET_S1, S1
_OUTLET_S4, S4
*Elset, elset=_PDF_WholeModel, internal, instance=Shell2-1-1, generate
    1, 3412, 1
*End Assembly
**
** MATERIALS
**
*Material, name=COOLANT
*Conductivity
    0.43,
*Density
    1056.,
*Specific Heat, type=CONSTANTPRESSURE
    50.,
*Viscosity
    0.008495,
**
** PREDEFINED FIELDS
**
** Name: Predefined Field-1    Type: Fluid thermal energy
*Initial Conditions, type=TEMPERATURE, Element Average
    _PDF_WholeModel, 15.
** Name: Predefined Field-2    Type: Fluid velocity
*Initial Conditions, type=VELOCITY, Element Average
    _PDF_WholeModel, 1, 0.
    _PDF_WholeModel, 2, 0.

```

```

_PDF_WholeModel, 3, 0.
** -----
**
** STEP: Step-1
**
*Step, name=Step-1
*CFD, incompressible navier stokes, energy equation=TEMPERATURE
0.01, 300., 0.025, 0.45, 1
1e-10, 0.5, , 0.5, 0.5
*Momentum Equation Solver
250, 2, 1e-05
*Pressure Equation Solver
250, 2, 1e-05
ICC, 1, 1, CG
*Transport Equation Solver
250, 2, 1e-05
**
** BOUNDARY CONDITIONS
**
** Name: BC-1 Type: Fluid wall condition
*Fluid Boundary, type=PHYSICAL, wall, surface=INTERFACE
** Name: BC-2 Type: Fluid inlet/outlet
*Fluid Boundary, type=PHYSICAL, velocity inlet, surface=INLET
VELX, 0.
VELY, -0.07536
VELZ, 0.
TEMP, 15.
** Name: BC-3 Type: Fluid inlet/outlet
*Fluid Boundary, type=PHYSICAL, pressure outlet, surface=OUTLET
P, 0.
passiveoutflow, 0.0
TEMP, 18.
** Interaction: Int-1
*Co-simulation, name=Int-1, controls=Int-1_Ctrls, program=MULTIPHYSICS
*Co-simulation Region, import, type=SURFACE
interface, TEMP
*Co-simulation Region, export, type=SURFACE
interface, HFL

```

```
interface, LUMPEDHEATCAPACITANCE

*Co-simulation Controls, name=Int-1_Ctrls, coupling scheme=gauss-seidel, time
incrementation=lockstep, time marks=yes,

step size=min, scheme modifier=lag

**

** OUTPUT REQUESTS

**

*Restart, write, frequency=0

**

** FIELD OUTPUT: F-Output-1

**

*Output, field, variable=PRESELECT

*Output, history, frequency=0

*End Step
```



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