



U.S. DEPARTMENT OF ENERGY

SMARTMOBILITY

Systems and Modeling for Accelerated Research in Transportation

Energy Efficient Connected and Automated Vehicles

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ARGONNE NATIONAL LABORATORY

2019 DOE Vehicle Technologies Annual Merit Review - June 2019



Project Overview

Timeline

- Project start: Oct. 2016
- Project end: Sep. 2019
- Percent complete: 75%

Budget

- Total Funding (3 years): \$2.3M
- FY18 Funding: \$660,000
- FY19 Funding: \$850,000

Partners

- Argonne: lead
- LLNL, LBNL: test data

Barriers

- Eco-driving research rarely integrates **advanced powertrain technologies**
- Combining dynamics and powertrain control results in **complex control problems**
- **Real-world implementation** often challenging
- Many **exogenous factors** (e.g. traffic), affect energy saving potential of eco-driving
- **Lack of practical tools** for “powertrain-aware” eco-driving algorithm development

Project Objectives and Relevance

CAV Eco-Driving Control: can we save energy through control of speed *and* powertrain? How much?

- For multiple powertrains: Conv., HEV, EV
- Automation levels 2 to 5
- For entire missions: real-world routes with mix of cruise-control, car-following, intersection approach and departure
- “Real-world implementable:” realistic I/Os, high-frequency feedback loop, adapted to powertrain response time

Enable energy efficient future mobility systems

CAV Simulation (RoadRunner): how to simulate CAVs, baseline, current and future technologies?

- Advanced powertrain technologies
- Multiple vehicles
- Road (speed limits, traffic lights, etc.)
- Driver (human or automated) reacting to environment
- Communication (V2X) & sensors

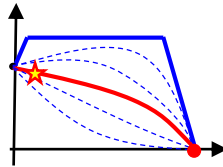
Tool for eco-driving research

Approach

CAV Eco-Driving: Control of Powertrain AND Longitudinal Speed

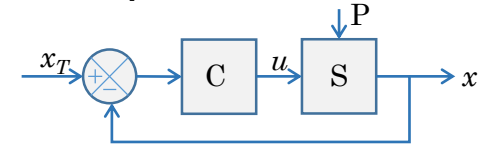
A. Optimization & Trajectory planning

- Future horizon \gg most energy-efficient state & control trajectory (speed, torque, SOC, etc.)
- Optimal control, Quadratic programming, etc.



B. Real-Time Control

- Current state \gg what command to follow optimal state & ctrl. trajectory?
- MPC, Feedback loops, transients, dynamics



Driving Scenarios

Powertrains

Real-World Data

Chassis Dyno
Track
On-road



**ROAD
RUNNER**

Energy
Impacts

CAV Modeling and Validation

Platooning, ACC, Human Driving

Milestones

Energy impact of eco-driving strategies under different scenarios

✓ **Complete**

Eco-driving: Demonstrate a "real-world implementable" controller working online in RoadRunner environment

✓ **On track**

Quantify energy impact of advanced optimal eco-driving over a range of representative scenarios

✓ **On track**

RoadRunner: Complete the development and validation of human and automated driver models.

✓ **On track**

Validate model of a CAV

✓ **Complete**

2018
Q3

2018
Q4

2019
Q1

2019
Q2

2019
Q3

2019
Q4



Accomplishments: Model Validation in RoadRunner

Validated Model of a 3-Truck Platoon

With EEMS033

Field test of CACC trucks



Test data of 3-truck CACC operation at freeway speed for modeling & validation. (**EEMS033**)

Data analysis

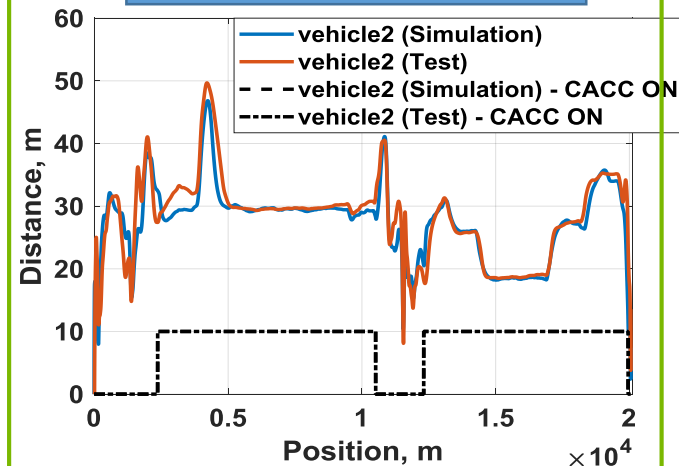
Importing test data into Autonomie, Analysis on shifting pattern, initial position,...

Model validation



Compare test data with simulation results

Inter-vehicle gap



Fuel Consumption :
($\Delta < .5\%$)

IDM/ACC/CACC development



Vehicle model /controller calibration

Conv. Class8 Linehaul with 12 speed transmission in Autonomie

Validating Prius Prime PHEV ACC Model

With EEMS041



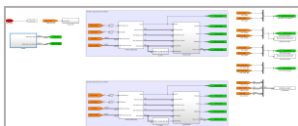
Ego vehicle

Chassis Dynamometer

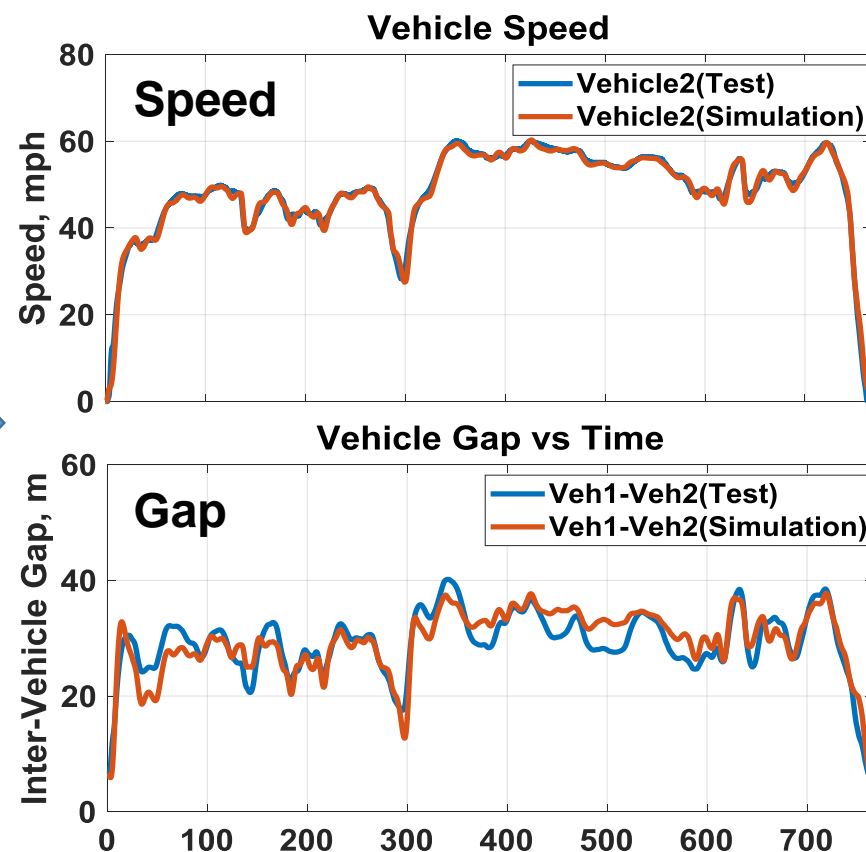


*Toyota Prius Prime PHEV w/ sensor override to emulate distance to preceding vehicle, stock Adaptive Cruise Control (ACC) – **EEMS041***

RoadRunner



Previously validated Prius Prime PHEV Autonomie model + new ACC model

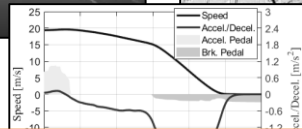
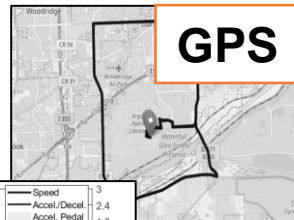


Developed Human Driver Model

With EEMS045

“Feature-Rich” On-Road Data Collection

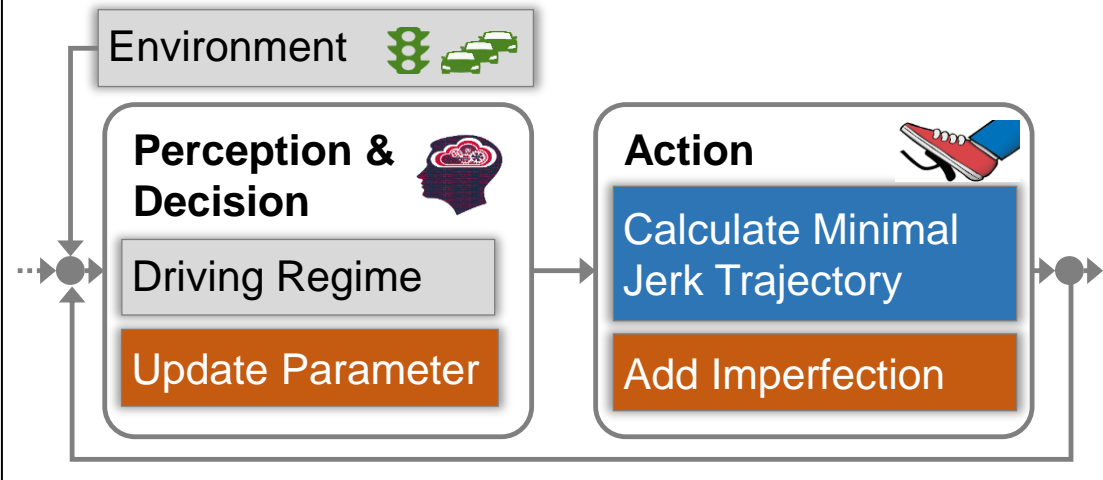
Instrumented vehicles



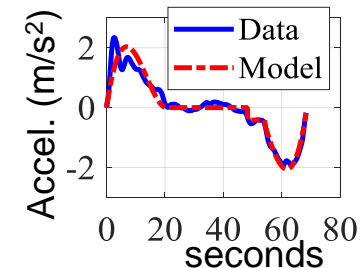
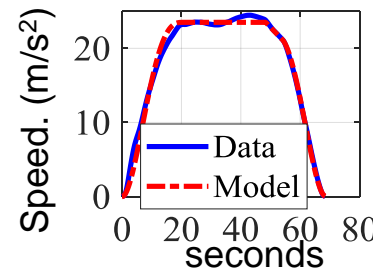
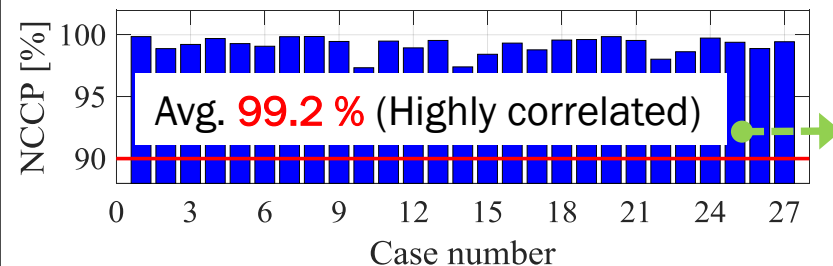
CAN & Radar

Human Driver Model Development

- Data-driven (■) and Analytical Approach (■)

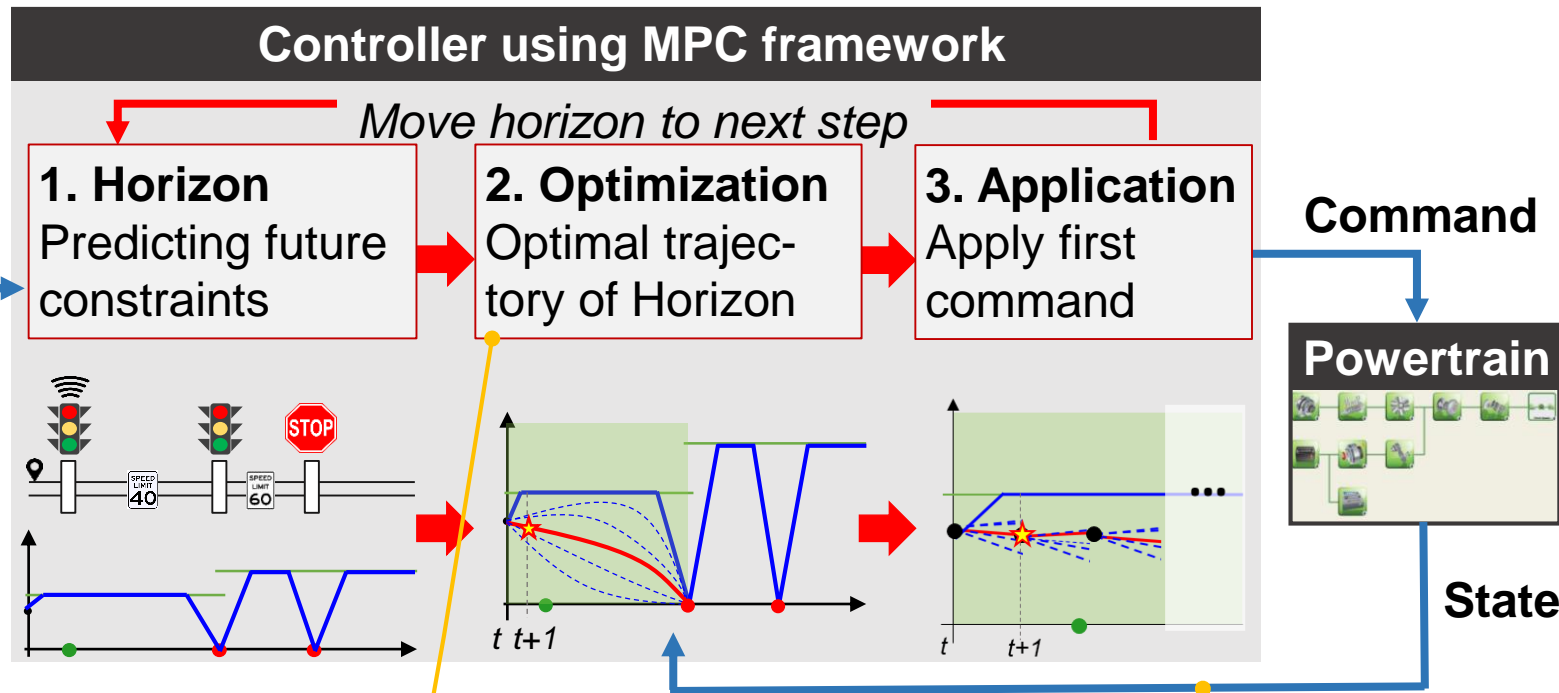


Model Validation (Compare experimental data with model)



Accomplishments: Real-World Implementable Eco-Driving Control for CAVs

Model-Predictive Control (MPC) Framework for Real-Time Implementation of Optimization Algorithms



Optimization algorithms (solvers)

- Quadratic Programming (QP)
- PMP for dynamics only
- PMP for dyn. + powertrain

Feedback loop allows real-time implementation:

- Optimization model is simplified
- Knowledge of environment is imperfect
- Future speed of preceding vehicle unknown

MPC Controller Implemented in RoadRunner with Various Optimization Options

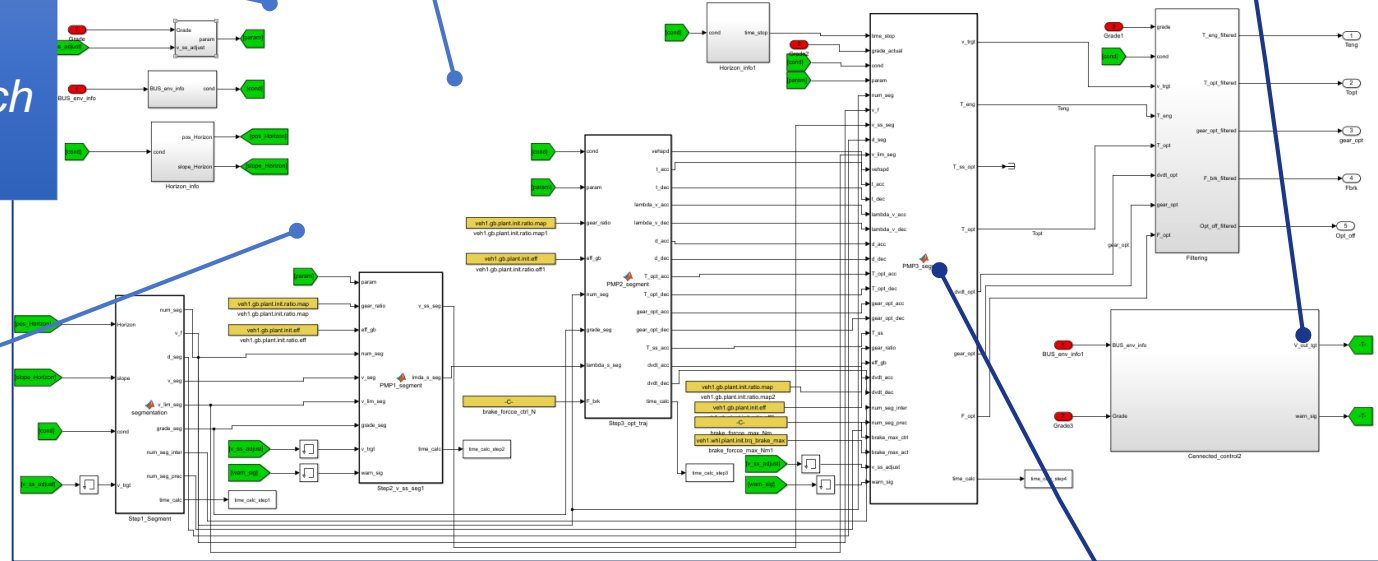
Interfaces with complex dynamic systems; deals with **transients, constraints, delays**, etc.

Works for entire mission:

- *cruising*
- *car-following*
- *intersection approach and departure*

Real- world Implementable

Eco-approach with V2I:
Can use traffic signal phase and timing if available for further energy savings



Two optimization options:

- 1. **Speed/Acceleration optimization only** (EcoDrv Spd/Accel)
- 2. **Powertrain and speed co-optimization** (EcoDrv PT+Spd)

1. Speed Optimization with Analytical Solution (No Powertrain)

EcoDrv Spd/Accel

Eco-Driving Optimal Control Problem

$$J = \int_{t_0}^{t_f} a^2 d\tau \quad + \quad \begin{array}{l} \dot{s} = v \\ \dot{v} = u \end{array}$$

Minimize Acceleration **Vehicle Dynamics**

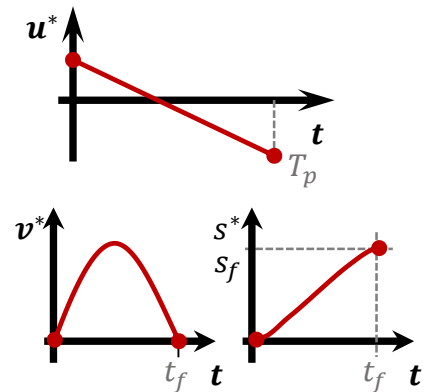
Command $u = a$ (acceleration)
States s : distance, v : speed

Simple analytical solution

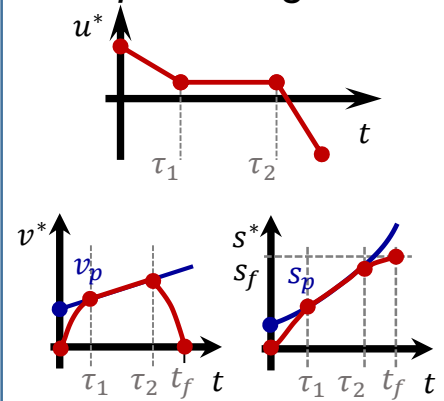
Pontryagin Minimum Principle (PMP)

- Simple model allows analytical solution
- Powertrain not taken into account, so it can be applied to any vehicle
- Not sensitive to grade

Without preceding vehicle



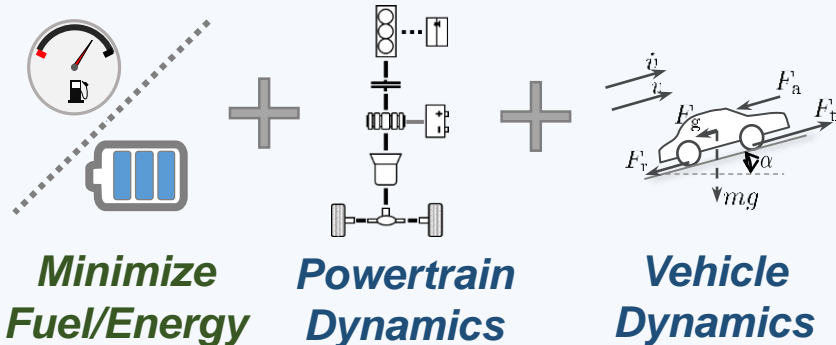
With preceding vehicle



2. Optimization for Combined Powertrain and Speed Control

EcoDrv PT+Spd

Eco-Driving Optimal Control Problem



↓ **Pontryagin Minimum Principle (PMP)**

Minimize Hamiltonian

$$H = \underline{P} + \lambda_v \dot{v} + \lambda_s \dot{s}$$

Power

P_f Fuel power (Conv)

P_b Electric power (EV)

$P_f + \lambda_E P_b$ Equivalent power (HEV)

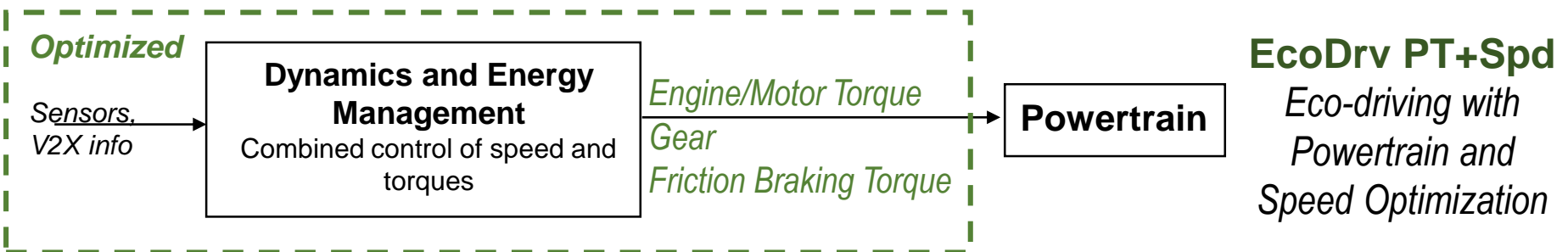
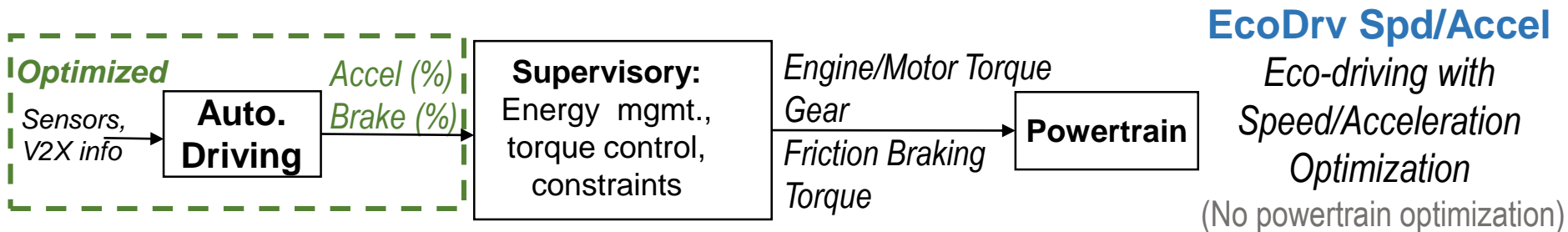
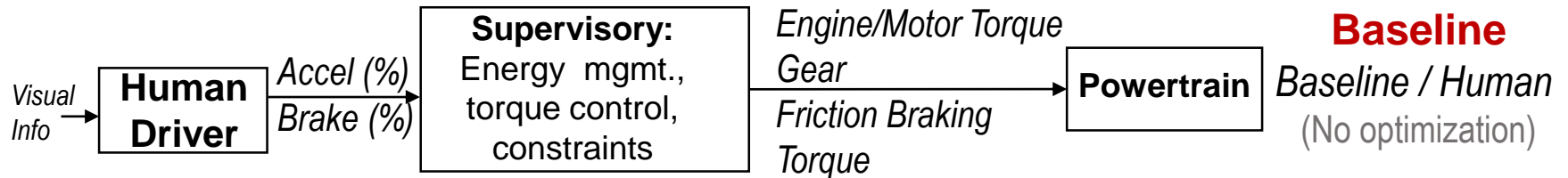
veh. dynamics

- Powertrain-specific to achieve maximum energy savings
- Takes into account grade
- Requires some numerical solvers

Completed optimization algorithm for:

- 3 powertrains (Conv., EV, HEV)
- Periodic control (“Pulse and Glide”)
- Stop approach and departure
- Car-following

Summary of Control Strategies Used in Case Study



Accomplishments: Energy Impact of CAV Eco-Driving Control (Preliminary Results)

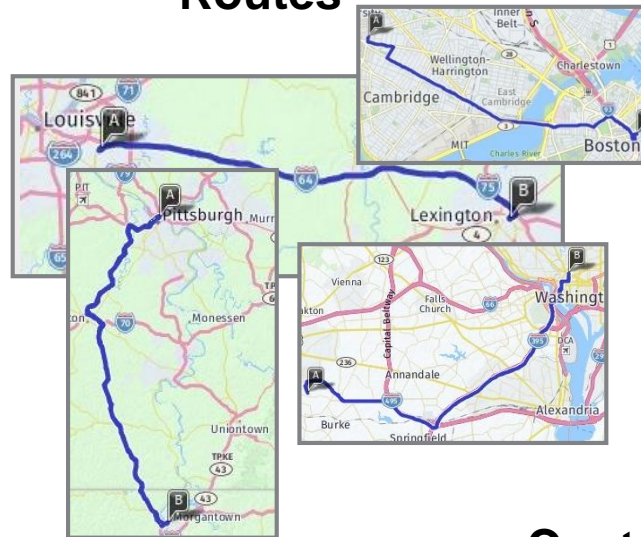
Simulation Setup: A Large Number of Scenarios for a Representative Energy Impact Evaluation

Vehicles

Midsized, Current Tech.



Routes



*Real-world routes
Data from HERE maps
14 Mixed
6 Suburban
8 Urban
16 Highway*

Scenario



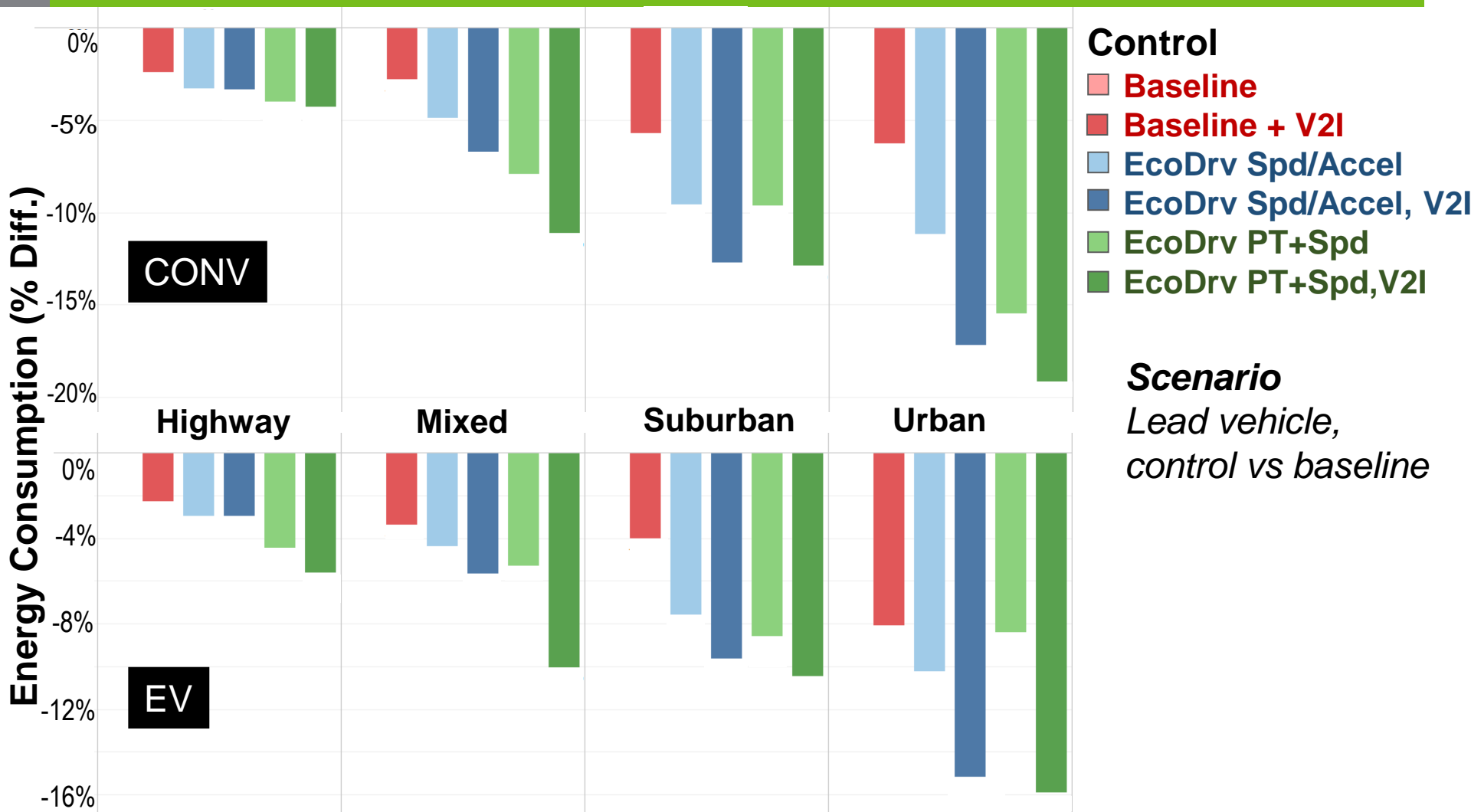
2 vehicles

Traffic signal phase and timing info (V2I): 0% or 100%

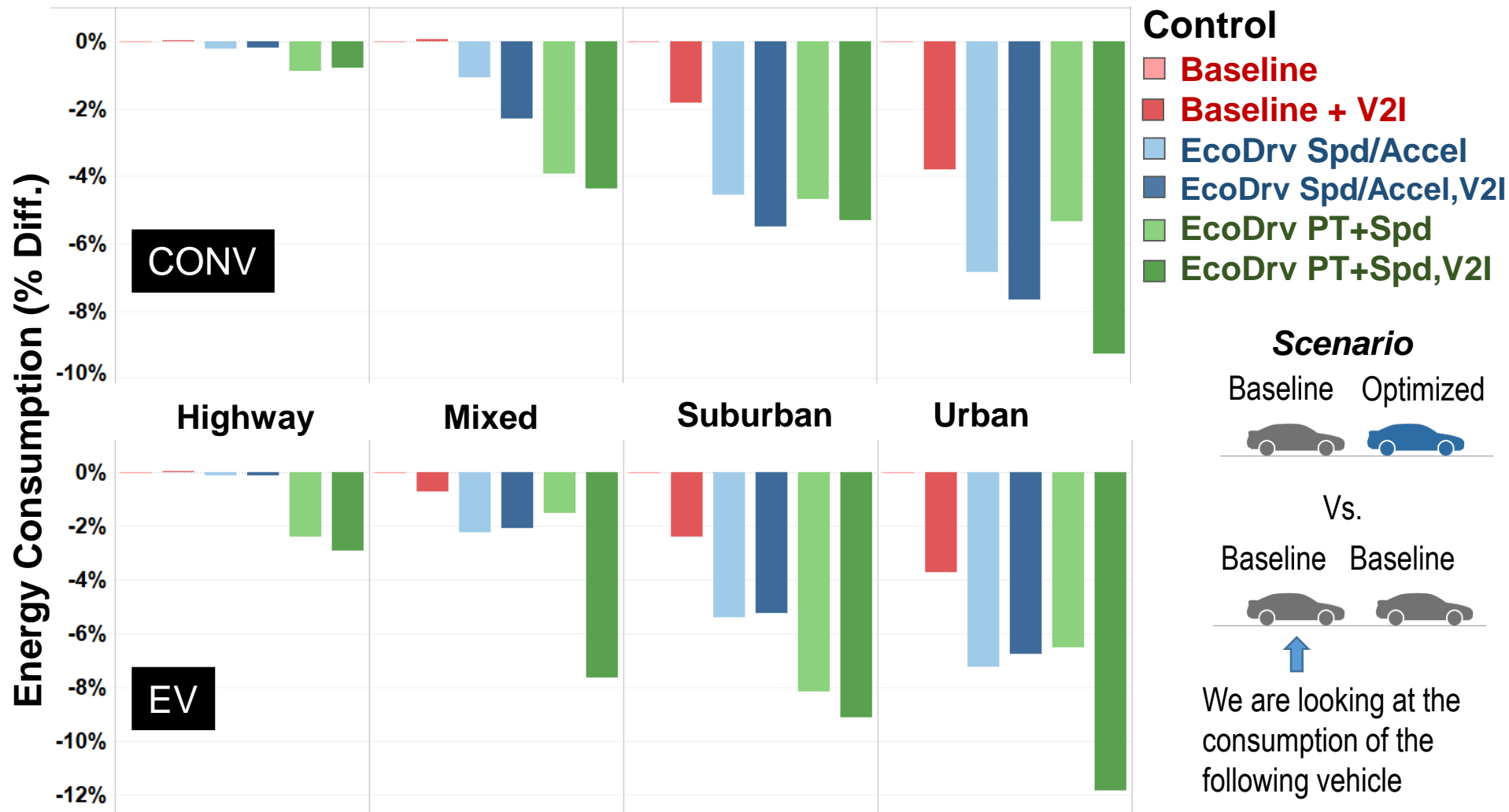
Control

Control	Description
Baseline	Baseline, no optimization
EcoDrv Spd/Accel	Eco-driving with Speed/Acceleration Optimization
EcoDrv PT+Spd	Eco-driving with Powertrain and Speed Optimization

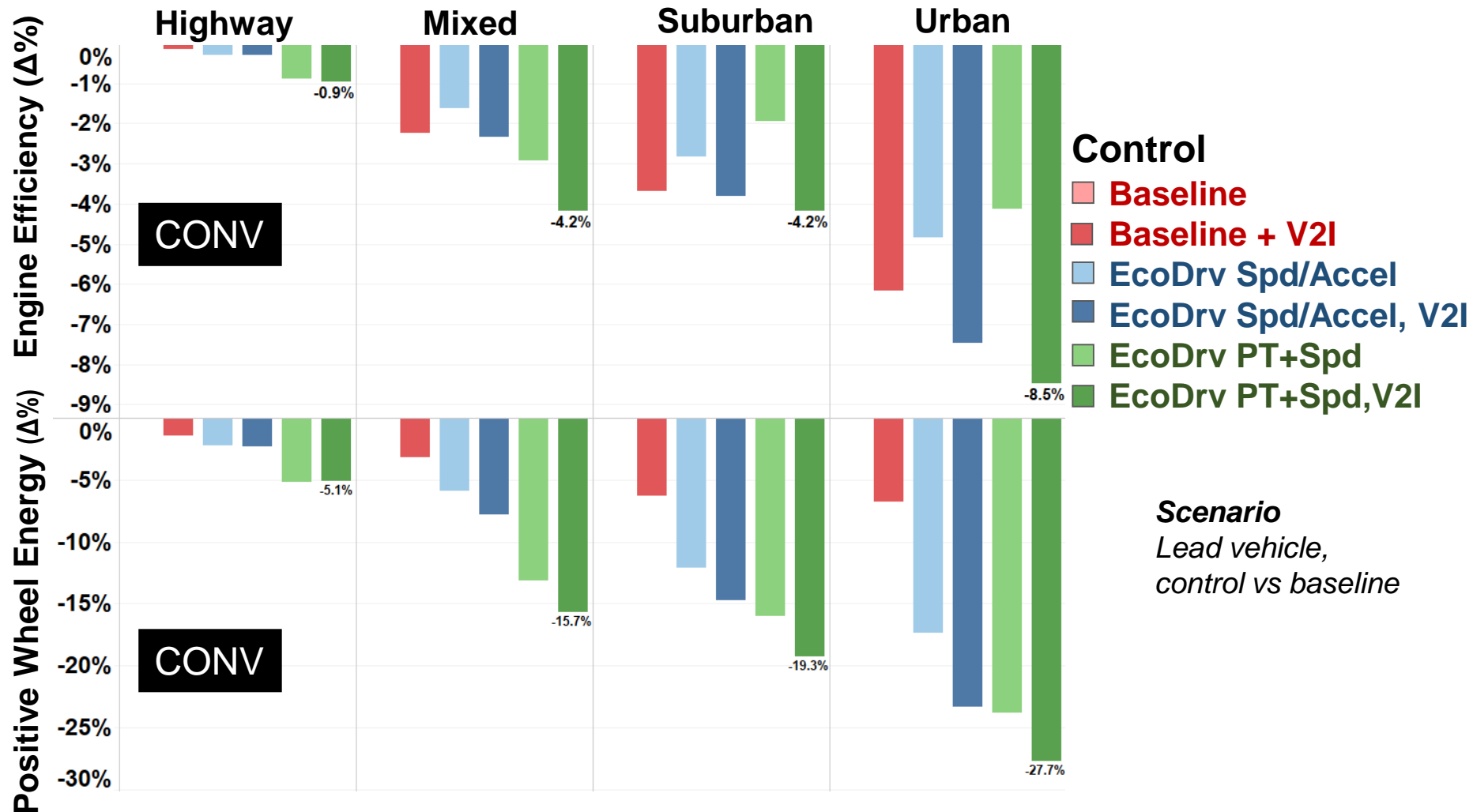
Lead Vehicle: Eco-Driving Saves Energy; Greater Savings with Powertrain Awareness and V2I



Non-Equipped Vehicles also Save Energy Thanks to Optimized Lead Vehicle



Eco-Driving Reduces Engine Efficiency and Positive Tractive Energy







Closing Remarks

Response to Previous Year Reviewers' Comments

Reviewer Comments	Response
<i>Reviewer prefers this approach over stating large fuel economy improvements for very specific situations</i>	We significantly expanded the number of scenarios to make our results even more representative
<i>Why both an offline (optimization) and online (model predictive control [MPC]) framework had been used [?]</i>	Optimization finds optimal trajectories over a horizon. MPC provides a framework to use these optimization algorithms in real-time with a feedback loop.
<i>Increased collaboration with OEMs is needed to ensure optimal control assumptions are possible in a production setting, and to correlate results with currently developed CAV systems.</i>	<ul style="list-style-type: none">• Models of existing CAV validated when data becomes available.• Team had positive feedback from OEM R&D.• We are building partnerships to demonstrate our control algorithms in actual vehicles (FY20 and beyond, provided funding)

Partnerships and Collaborations

	<p>Truck platooning data [EEMS033] CACC testing data (future work) [EEMS059] Mortar task (RoadRunner + micro-simulation) [EEMS076]</p>
	<p>Aerodynamic drag reduction coefficients from 3D modeling and wind tunnel</p>
	<p>Demonstrating RoadRunner for TARDEC use cases</p>
 Advanced Mobility Research	<p>“Feature-rich” on-road test data from [EEMS045] “Math-to-lab-to-road” initiative [EEMS041]</p>
<p>Other SMART</p>	<p>RoadRunner Core development [EEMS058] Autonomie vehicles for SMART studies [EEMS058]</p>
<p>OEMs</p>	<p>Presentations, discussions, sharing of prototypes</p>

Remaining Challenges and Barriers

- *Human and CAV model development in RoadRunner:*
 - **Quality data is in short supply**, limiting how representative models are
 - Human driver model improvements so far focused on **stop approach**
- *Eco-driving controller with optimization of speed and powertrain:*
 - **Conventional** and **EV** powertrain for now
 - Some **manual tuning** required (e.g. setting target speed)
 - **Powertrain dynamics** & Drivability/NVH need to be improved (shifting events, accelerations)
- *CAV energy impact case study:*
 - **2 Autonomie** vehicles for now (Conv., EV with current technology)
 - **Traffic** situations not modeled at this point
 - **Travel time** not exactly the same

Proposed Future Research*

- *Continue human and CAV model development in RoadRunner:*
 - Improve acceleration and cruising modes for the **human driver model**
 - Complete CAV validation work (**Prius ACC, CACC**)
- *Finalize eco-driving controller with optimization of speed and powertrain:*
 - Complete controller for **HEV**
 - Improve handling of **transients**
- *Rerun larger and more representative CAV energy impact case study:*
 - Add vehicles with **future technologies** (e.g. better efficiency and higher accessory load for future vehicles) developed for SMART
 - Add scenarios with **traffic [EEMS076]**
 - Analyze **energy savings/travel time trade off** for each controller

**Any proposed future work is subject to change based on funding levels*

Summary: On-Track to Complete Original Year 3 Goals

Developed models and algorithms

- ✓ **Validated RoadRunner models** of human driving, truck platooning, ACC
- ✓ **Developed real-world implementable CAV eco-driving controllers:**
 - *Optimization of powertrain AND speed*
 - *Multiple powertrains: conventional, EV, HEV (in progress)*

Evaluated energy impact

- **5% to 17%** energy savings for CAV in lead position w/ V2I
- Optimizing powertrain AND speed saves extra **1-5%**
- V2I brings **3% to 8%** (pts) extra savings
- Non-CAVs also benefit (**~10%** savings) [following a CAV]

Future research FY20 & Beyond (Provided funding)

- Expanded impact case study
- Collaborative eco-driving
- In-vehicle/on-road testing and demo
- Powertrain design optimization for CAVs

EEMS016

Energy Efficient Connected and Automated Vehicles

Thank you!



**Dominik
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Jeong**



**Daliang
Shen**

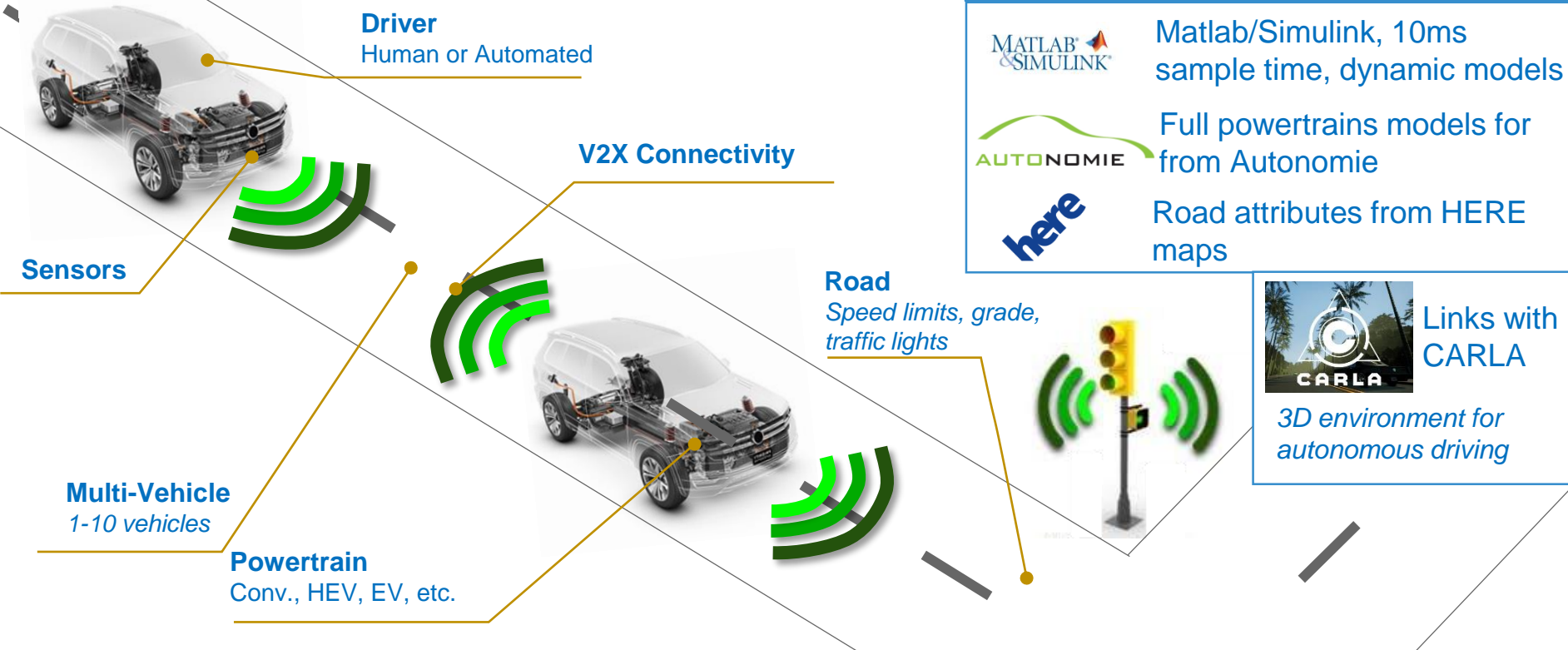


**Jihun
Han**

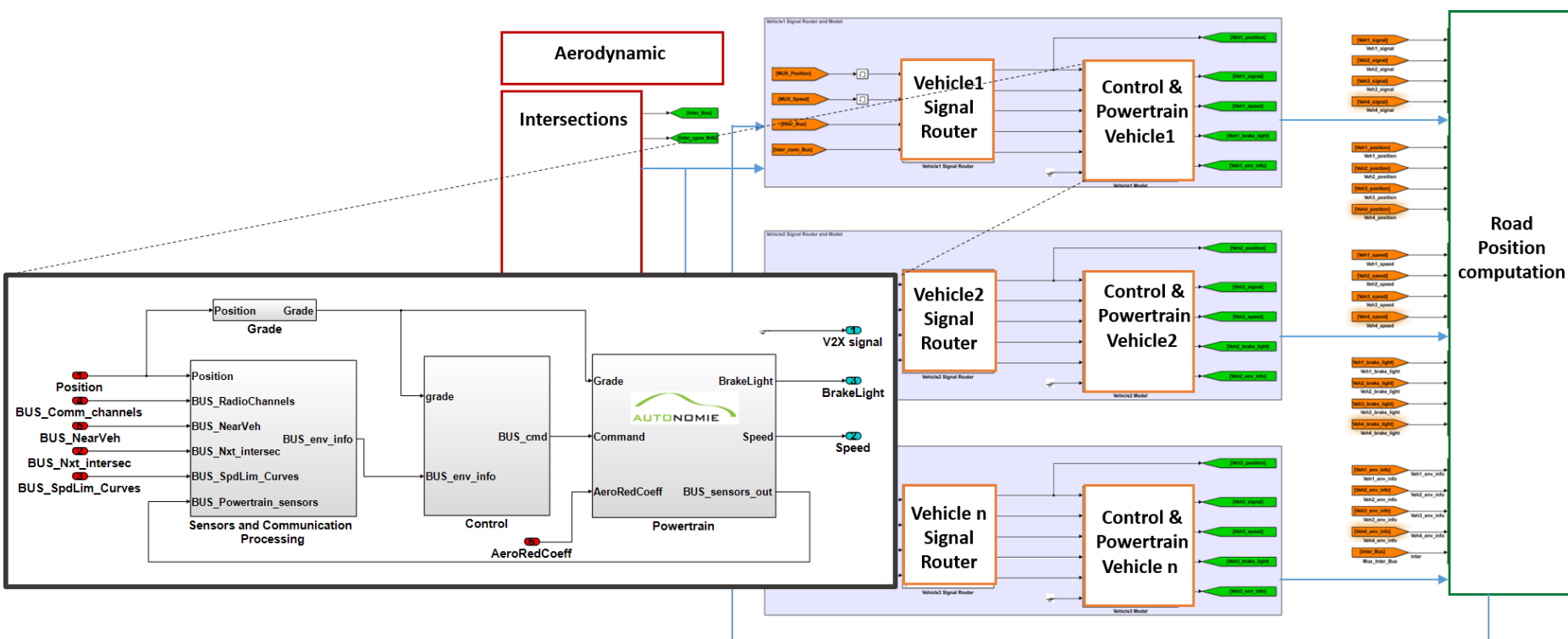
Technical Back-up Slides

RoadRunner: a Framework to Simulate Powertrain and Driving Dynamics for CAVs

ROAD RUNNER



RoadRunner Model Architecture



Validated Human Driver Model

- Two Measures

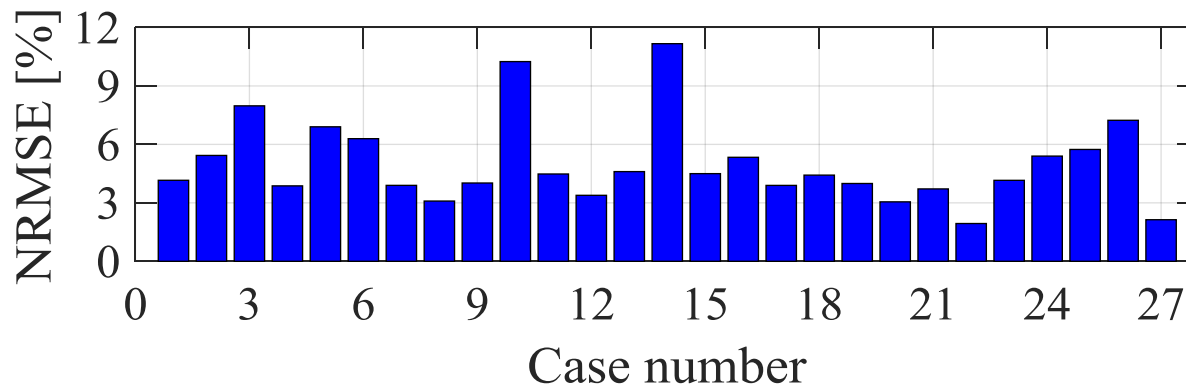
- Normalized Cross Correlation Power (NCCP)

- $$\text{NCCP} = \frac{\max[R_{xy}(\tau)]}{\max[R_{xx}(\tau), R_{yy}(\tau)]}, \text{ where } R_{xy} = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x(t) \circ y(t - \tau) dt$$

- Normalized Root-Mean-Squared-Error (NRMSE)

- $$\text{NRMSE} = \frac{\sqrt{\sum_{i=0}^T (x_i - y_i)^2 / T}}{(y_{\max} - y_{\min})}$$

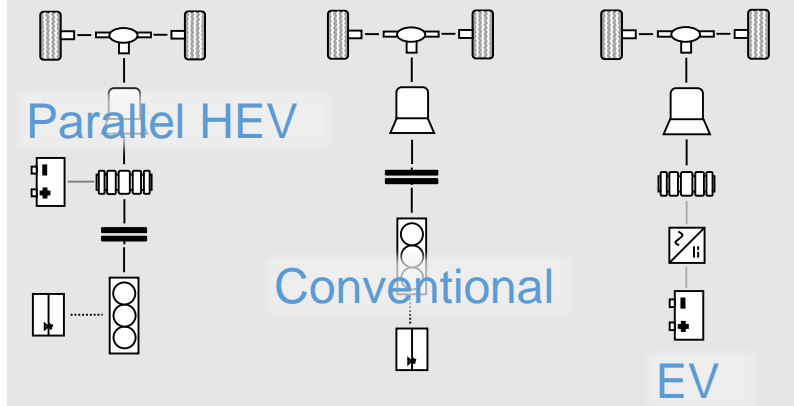
- Validation Results for 27 cases



Avg. **5.02 %**
(small error)

Optimization for Combined Powertrain and Speed Control in Various CAV Scenarios

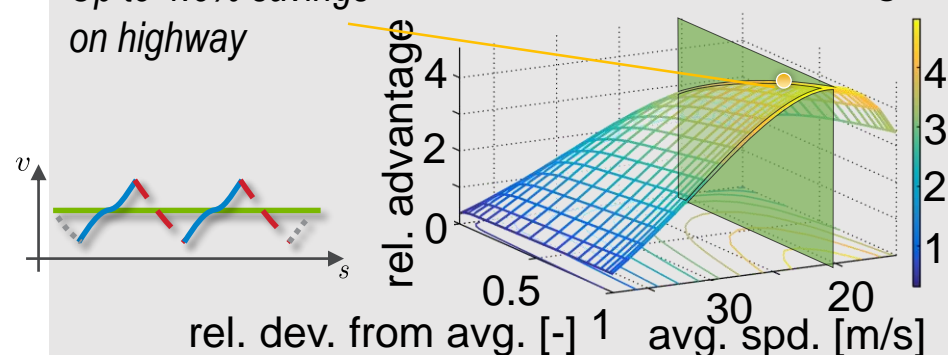
3 Powertrain Options



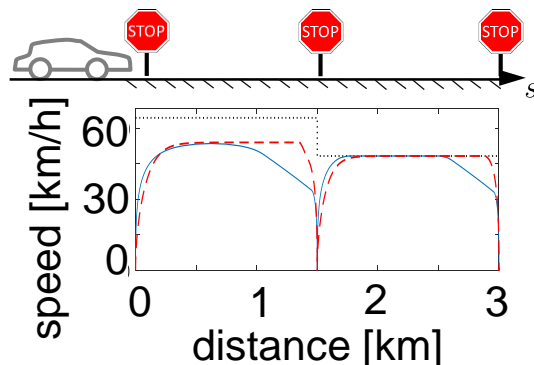
Periodic Cruise-Control

Up to 4.8% savings
on highway

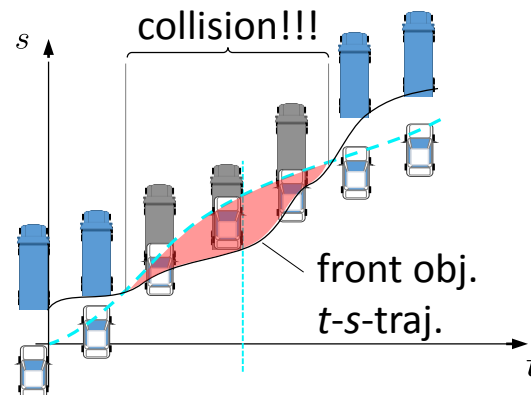
relative fuel-eco advantage



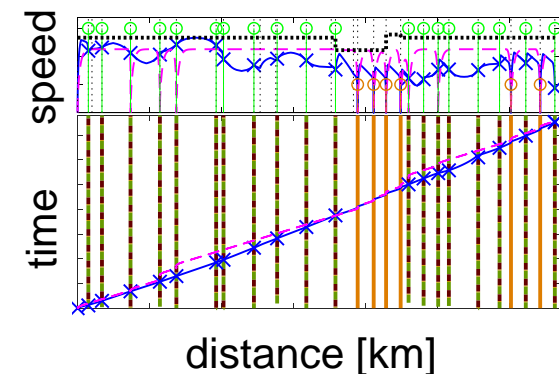
Stop Approach & Departure



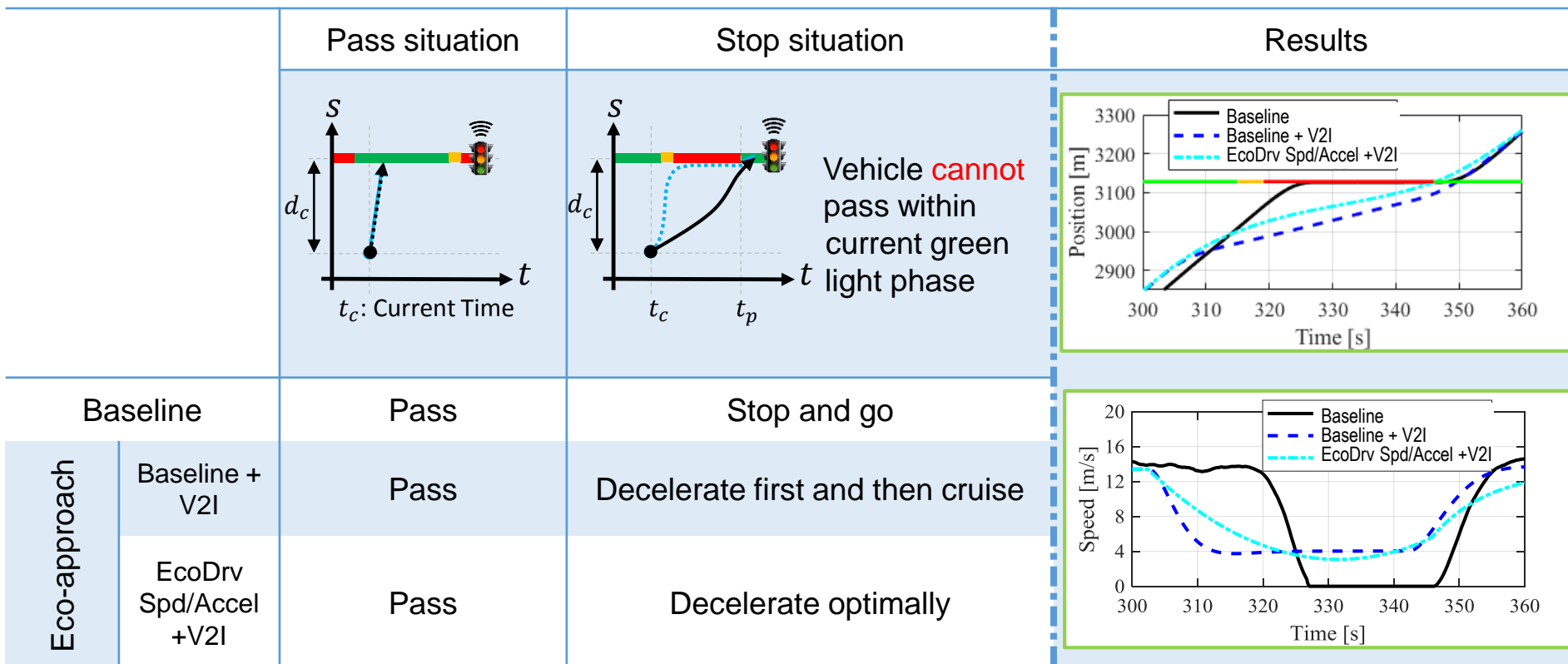
Car-Following



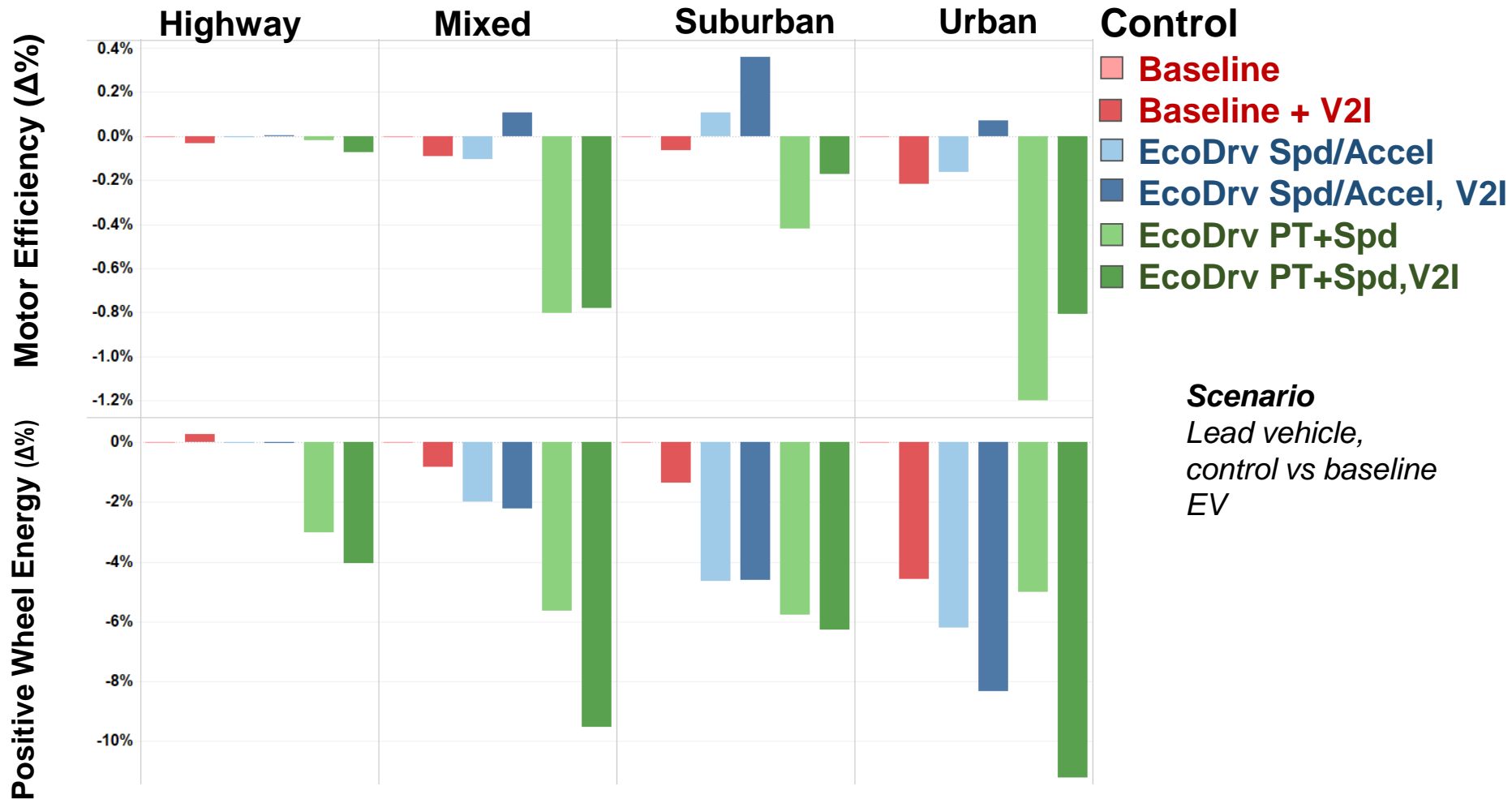
Signal Phase & Time



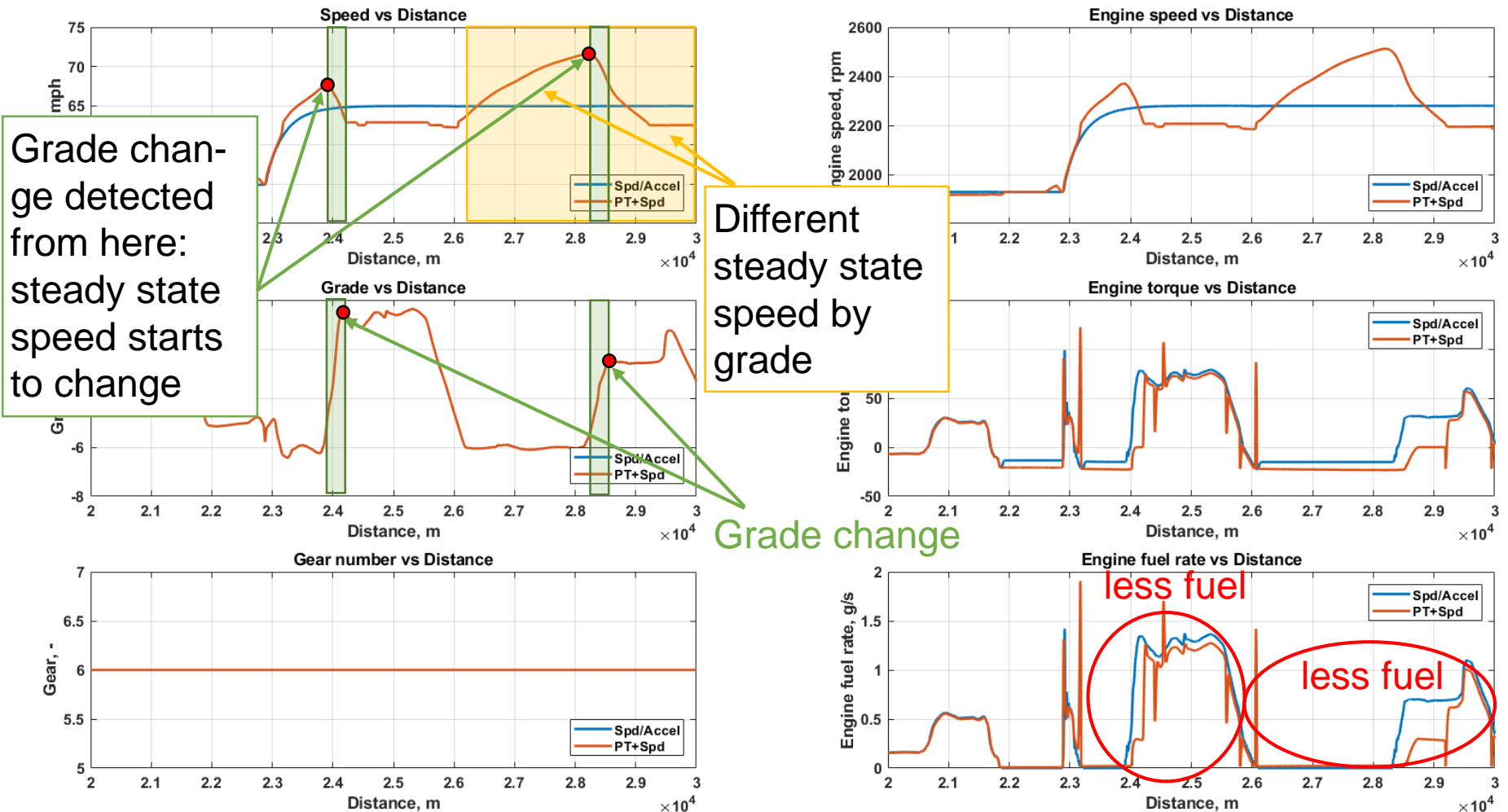
Traffic Signal Eco-Approach with V2I Approach



Eco-Driving Reduces Motor Efficiency and Positive Tractive Energy



How Does Eco-Driving (PT+Spd) Save Fuel? *Case of cruising with grade information*



How Does Eco-Driving (PT+Spd) Save Fuel?

Urban Driving with V2I

