











# NREL/Industry Range-Extended Electric Vehicle for Package Delivery

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NREL Vehicle Technology Evaluations Team: Ken Kelly, Adam Duran, Mike Lammert, and Eric Miller

SAE 2017 Range-Extenders for Electric Vehicles Symposium November 14, 2017, Dearborn, Michigan

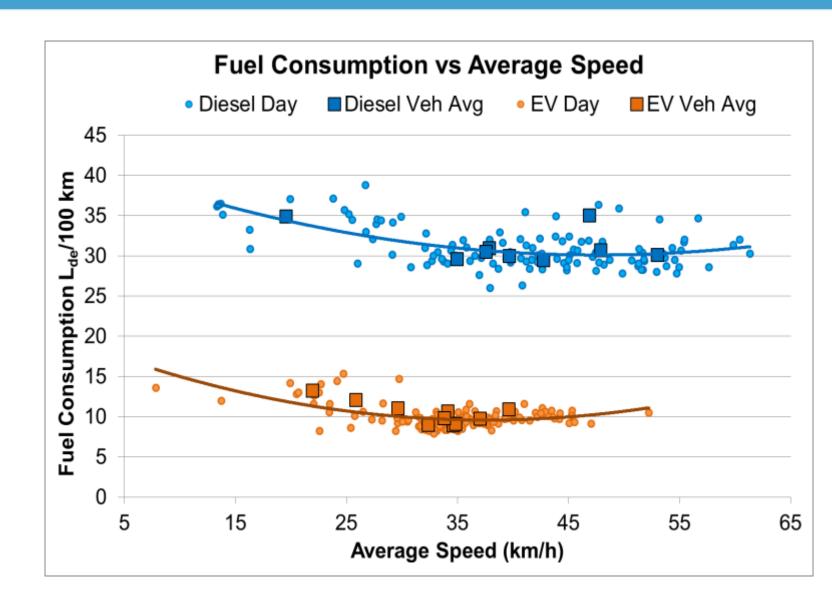
NREL/PR-5400-70558

## Approach and Overarching Questions

- Range extenders can be viable technology option for reducing fuel consumption from medium-duty (MD) and heavy-duty (HD) engines
- MD and HD engines have wide variations in use and duty cycle
- How best to identify vocations/duty cycles most suitable for range-extender applications?
- How to optimize powertrain and design requirements?

## Benefits of MD/HD Electrification

- Up to 3–4x FE benefit (energy equivalent basis)
- Lower maintenance costs
- Vocation-specific benefits
  - Reduced tailpipe emissions
  - Quiet operation
  - Off-board power
  - Potential for grid services
- U.S. manufacturing and innovation opportunities



## MD/HD Electrification: Remaining Challenges

- Low-volume production limits economies of scale
- Energy storage requirements, high cost and range limitations
- Fleet education, unbiased information, and long-term support
- Electric accessories significant impact on energy consumption
- Centralized fleets may experience significant utility demand charges



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# Overview

I. Data collection and analysis approach

II. Parcel delivery range-extender results

## NREL Field Data, Evaluation, and Analysis Tools

Data from field evaluations help populate Fleet DNA database

DOE fleet tools (DRIVE, FASTSim, AFLEET, etc.) used to analyze and investigate impacts – data used to validate and improve tools

Published information and data used by fleets, industry, DOE, and other research programs and agencies

Collect Lab and Field Data

Capture, Store, and Analyze

Laboratory Testing **Explore & Optimize** 

Communicate & Inform

Identify Barriers, New R&D Opportunities, Validate Efforts











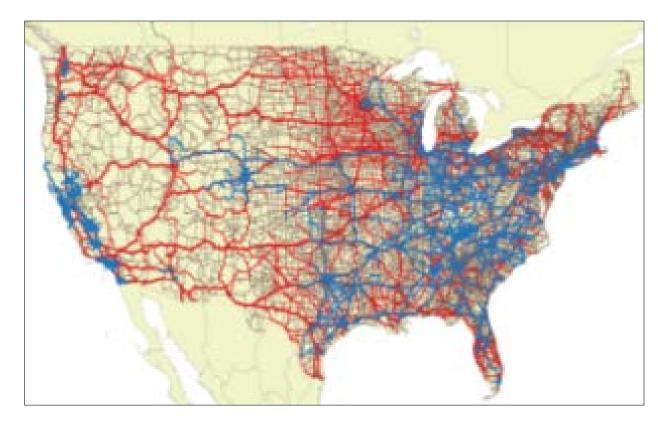




Partnership with Fleets and Technology Providers = Relevant Results & Optimized Solutions for Real-World Applications

## Fleet DNA: Clearinghouse of Fleet Vehicle Operating Data

- Repository of broad range of operational data for commercial vehicles across vocations/weight classes
- Features 11.5 million miles of 1-Hz engine CAN, GPS, and component data from 1,700 vocational vehicles operated by fleet partners
- Data processed to produce more than 350 unique duty-cycle metrics characterizing vehicle operating behavior

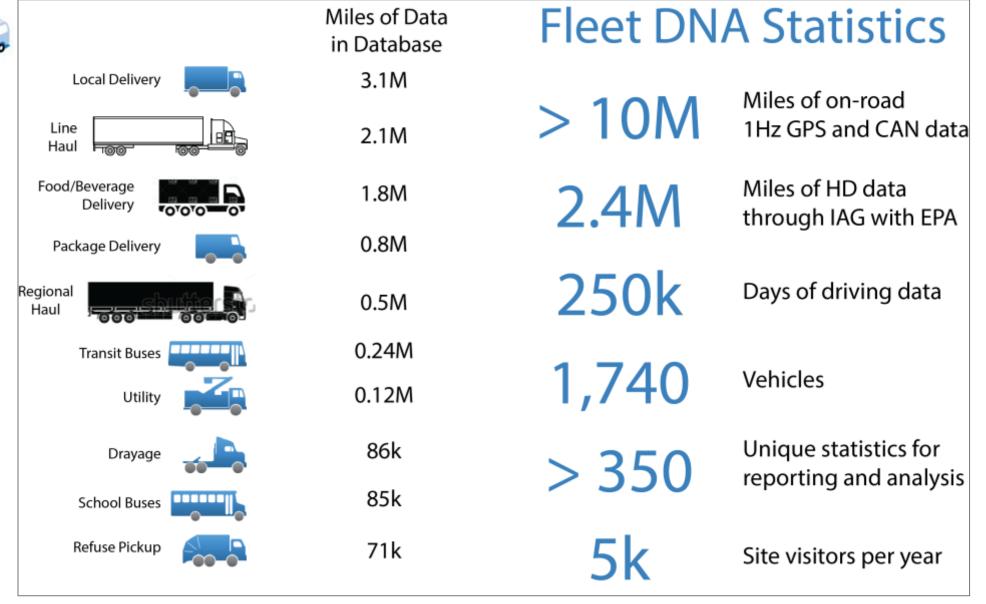


Freight volumes (red) along major U.S. roadways and Fleet DNA data coverage (blue) along those routes

https://www.nrel.gov/fleetdna

## Fleet DNA: Clearinghouse of Fleet Vehicle Operating Data

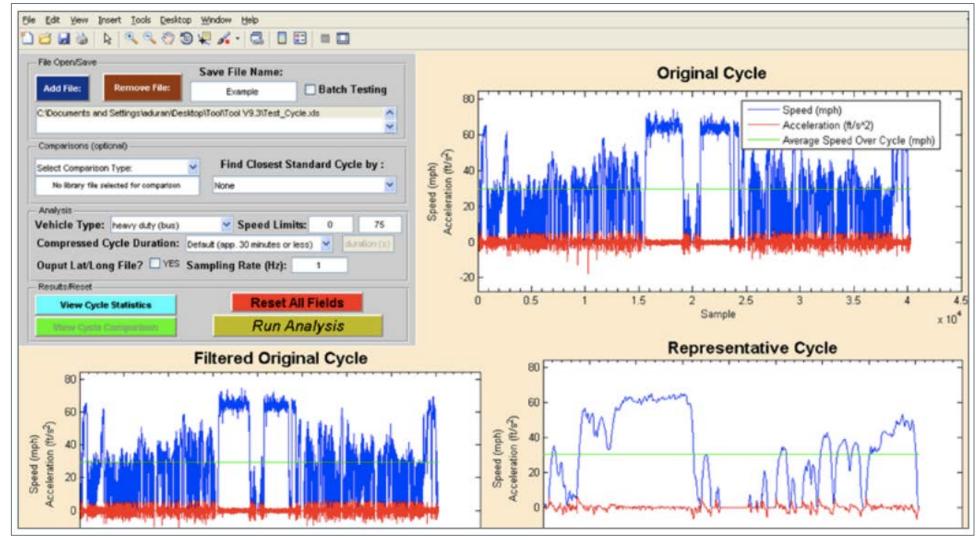




## DRIVE: Representative Drive Cycles from Vehicle Data



- Uses GPS/CAN data to produce statistically representative drive cycles
- 168 unique drive cycle metrics to generate custom representative drive cycles



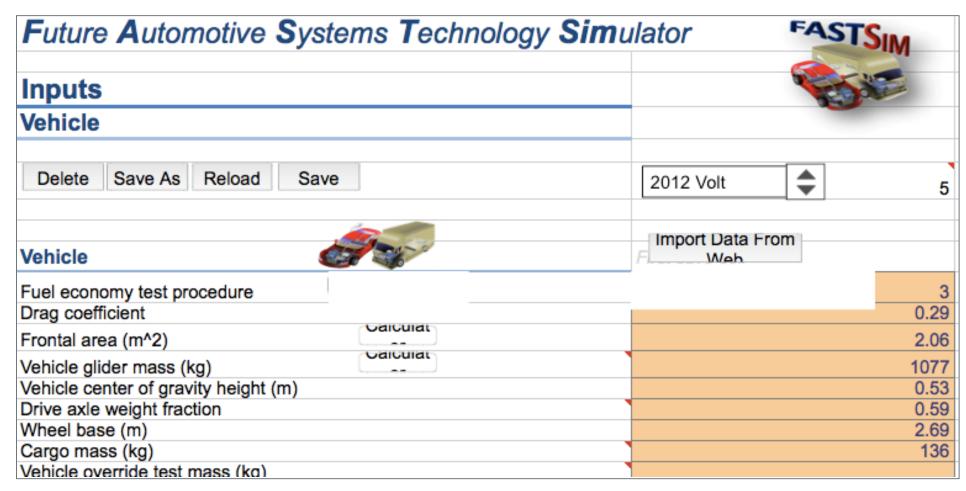
https://www.nrel.gov/transportation/drive.html

## FASTSim: Future Automotive Systems Technology Simulator

Used for comparing powertrain technologies and estimating the impact of technology improvements on vehicle efficiency, performance, cost, and battery life

#### Incorporates:

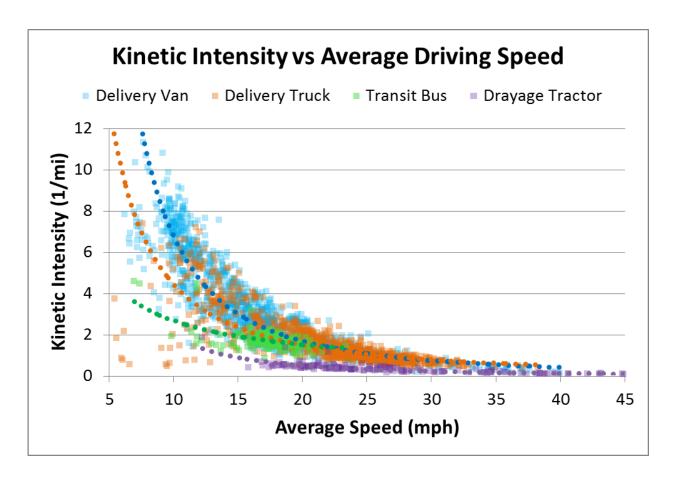
- Speed vs. time simulation
- Powertrain components
- Regenerative braking
- Energy management
- Battery life estimates
- Cost estimates
- Distribution of driving distances



https://www.nrel.gov/transportation/fastsim.html

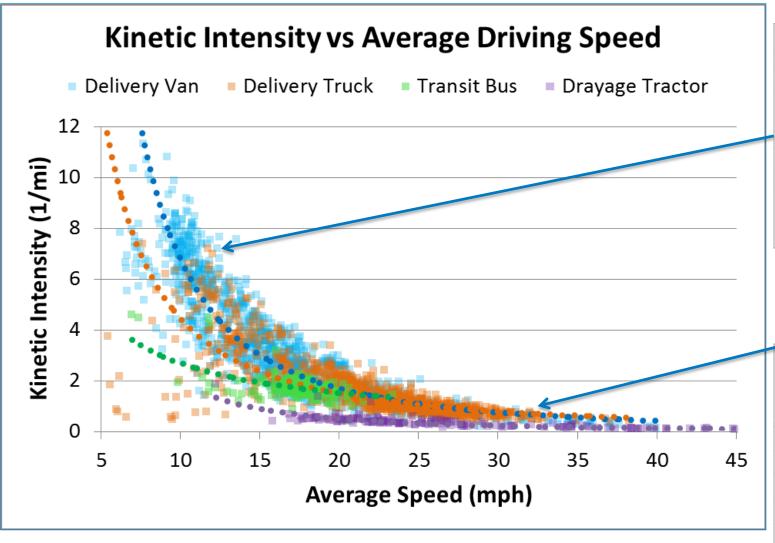
# An Aside on Kinetic Intensity

- Kinetic intensity (KI) characterizes relative driving aggressiveness
  - Ratio of characteristic acceleration and aerodynamic speed, where
  - Characteristic acceleration: Reflects inertial work to accelerate and/or raise the vehicle per unit mass per unit distance over the cycle
  - Aerodynamic speed: Ratio of overall average cubic speed to average speed

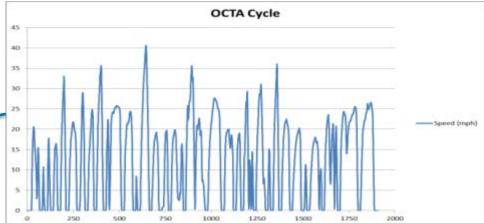


See SAE 2007-01-0302 for derivation

# An Aside on Kinetic Intensity



#### **High KI Low Speed Cycle**



#### **Low KI High Speed Cycle**



## Step 1

- Leverage Fleet DNA's parcel delivery data for representative drive/duty cycle characteristics
- Identify suitable standard drive cycles to model

#### Step 2

 Perform trip-level clustering analysis to identify distinct operating modes and driving patterns

#### Step 3

Create custom representative drive cycles from real-world data

#### Step 4

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Provider	Class	City	Miles	Number Days	Number of vehicles
FedEx	5	Los Angeles	12,657.79	279	21
UPS	6	Baltimore	7,591.46	275	15
UPS	4	Minneapolis	11,895.30	241	13
UPS	4	Minneapolis	19,381.11	364	11
UPS	6	Houston	7,179.90	114	6
UPS	6	Los Angeles	16,138.91	324	15
		Total	74,844.47	1,597	81

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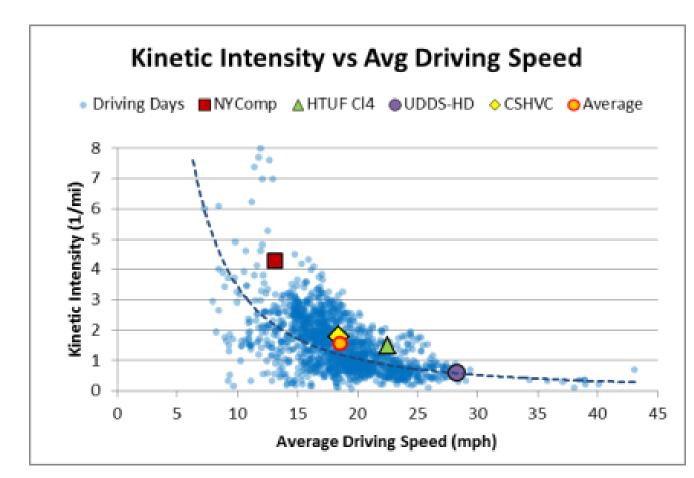
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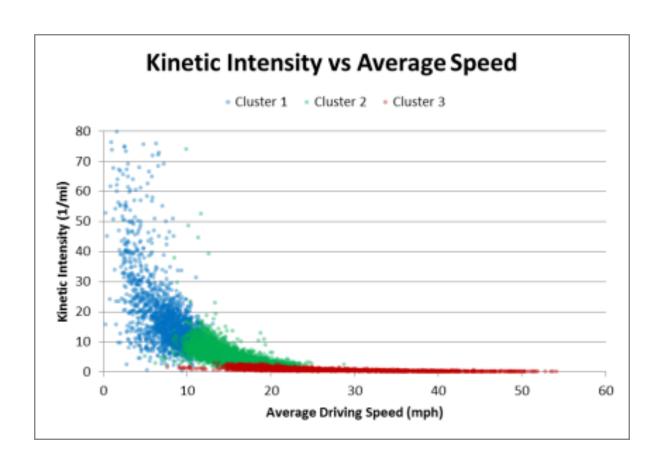
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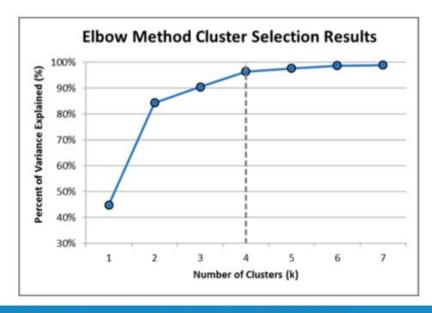
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#### **Clustering Analysis**

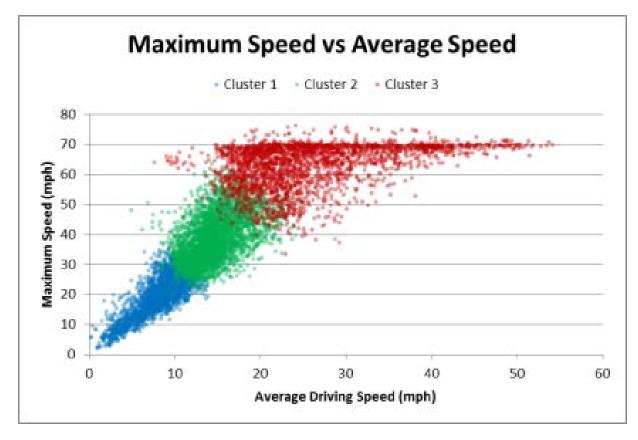
- 1. Trip-level metrics selected to define kinematic driving behavior
- 2. Clustering metrics scaled using the z-score scaling method
- 3. Identify number of clusters using both mean shift and elbow method
- 4. Use k-medoid clustering algorithm

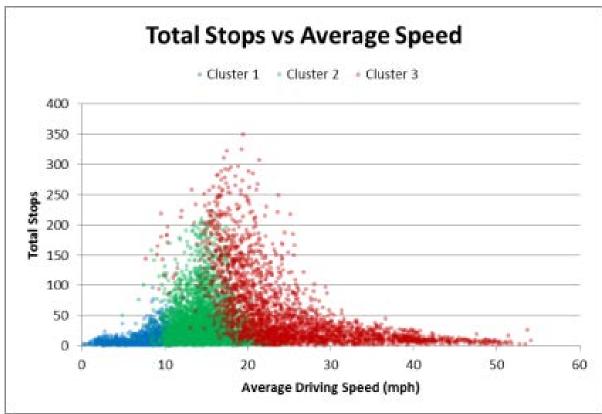


#### **Clustering Metrics**

- Aerodynamic speed
- Average driving speed
- Characteristic acceleration
- Kinetic intensity
- Maximum speed
- Stops per mile
- Total average speed
- Total distance
- Total stops

$$F(x) = minimize \sum_{i=1}^{n} \sum_{j=1}^{n} d(i,j)z_{ij}$$





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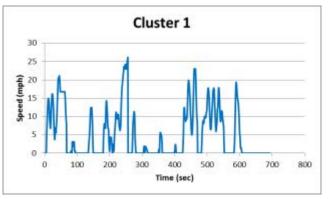
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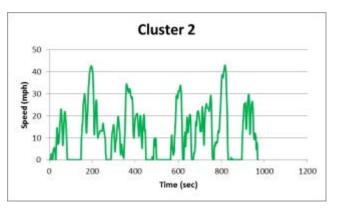
## Step 4

 Develop validated FASTSim vehicle model for system component sizing and system performance modeling over various drive cycles



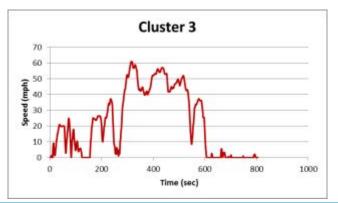
# Cluster 1 Slow moving

Slow moving, short trips with frequent stops



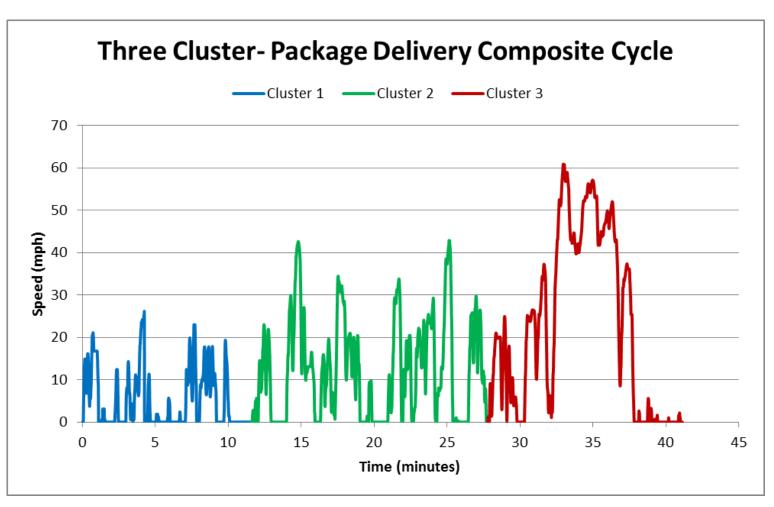
#### **Cluster 2**

Medium speed driving activity

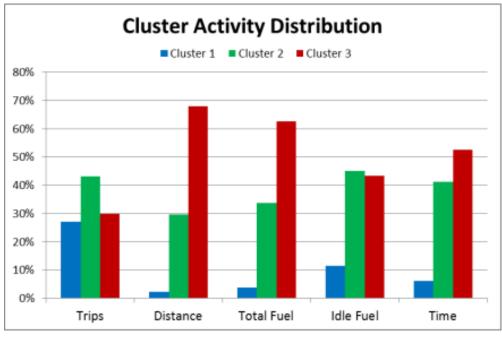


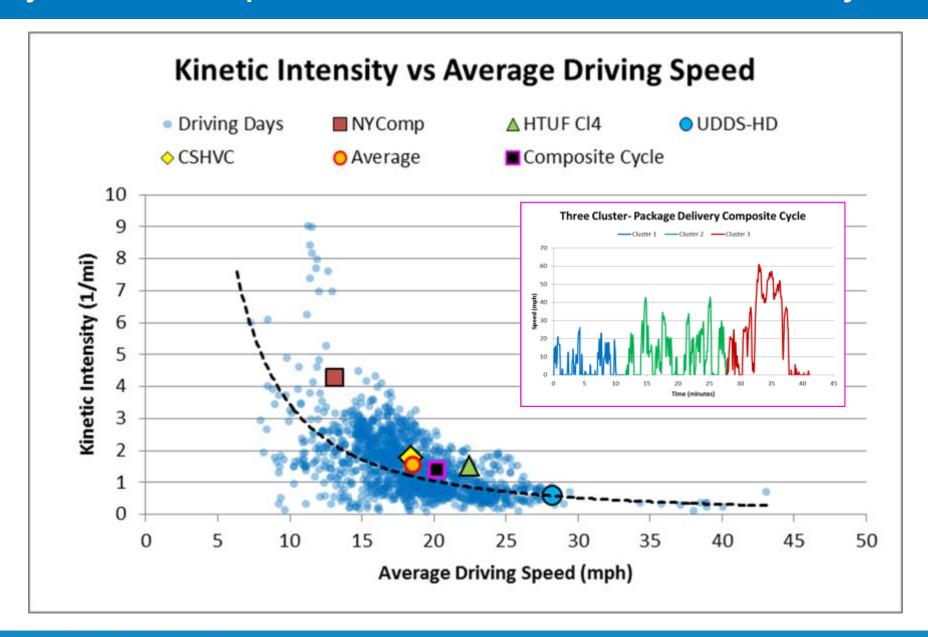
#### **Cluster 3**

Higher speed driving activity with fewer stops



#### **Weighting Factors**





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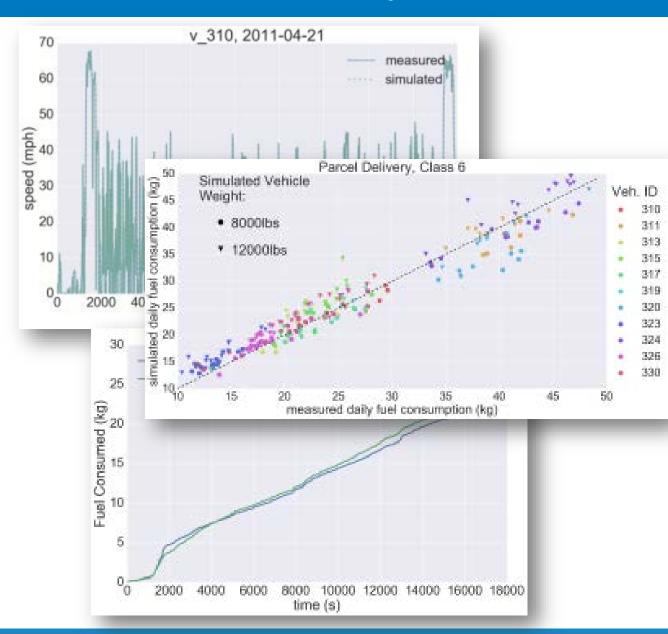
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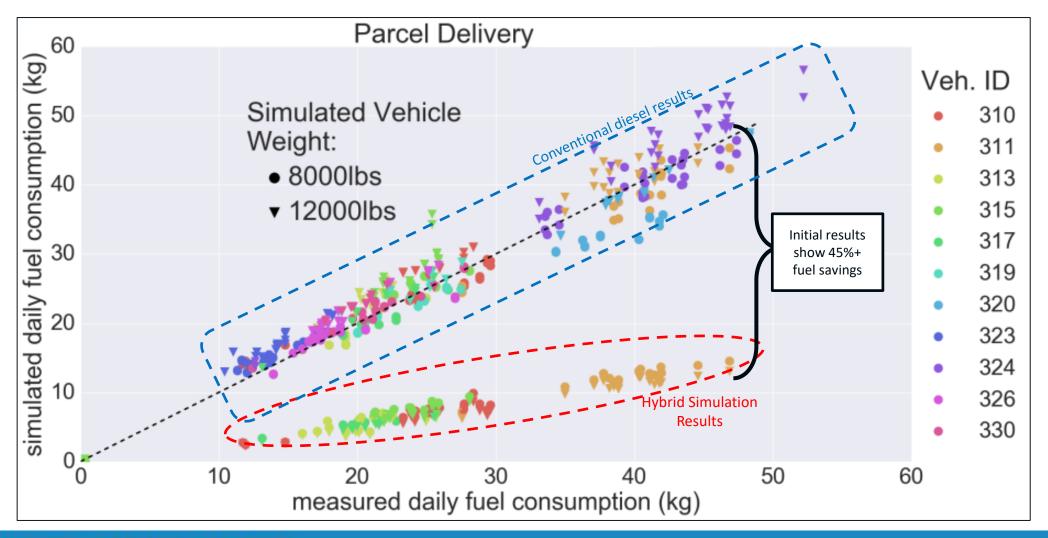
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## FASTSim Vehicle Modeling – Initial Results

- Conventional and initial hybrid vehicle model run over 231 days at two different vehicle weights
- Conventional diesel model results correlate to observed field data



# Cummins FOA Project

## Electric Truck with Range-Extending Engine (ETREE)

#### **Team**

 Cummins (lead), PACCAR, NREL, Argonne National Laboratory, Ohio State University

#### **Objectives**

- Develop, demonstrate, validate Class 6 vehicle with performance/range equivalent to conventional vehicle
- Use electricity to reduce fuel consumption by at least
   50% over wide range of real-world drive cycles
- Evaluate performance in fleet service

#### **Vehicle Features**

- Series hybrid architecture
- 112 kWh battery
- 155 kW ISB 4.5 diesel engine
- 160 kW motor, 130 kW generator

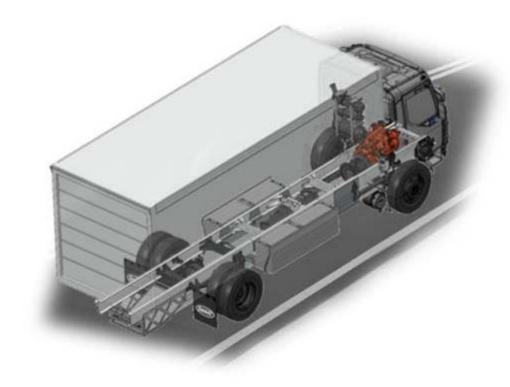


**Kenworth K270** 

Figure credit: Cummins 2017 Annual Merit Review presentation https://energy.gov/sites/prod/files/2017/06/f34/gi189\_kresse\_2017\_o.pdf

## Need: Expanded Drive Cycle Development

- Developing hybrid system controls technology focused on battery state-ofcharge trajectory management and vehicle integration
  - Electrified accessories and thermal management systems
- EV design optimization requires a "workday" drive cycle due to limited energy density and state-of-charge impact on continuous power
- Also need grade and key-on information



**K270/PB220 ETREE** 

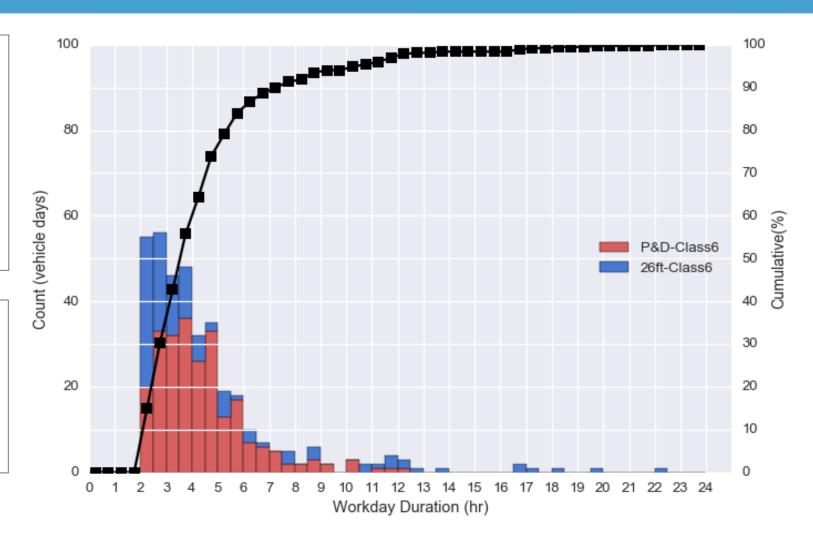
Figure credit: Cummins 2017 Annual Merit Review presentation https://energy.gov/sites/prod/files/2017/06/f34/gi189 kresse 2017 o.pdf

## Work Day Drive Cycle Development: Class 6 P&D

Days of operation data **Vehicles** Miles of real-world data 2.5M

analyzed

- Median operating day duration ~ 4 hours
- 95% days operate 8 hours or less



## Work Day Drive Cycle Development: Class 6 P&D

78,000 Days of operation data

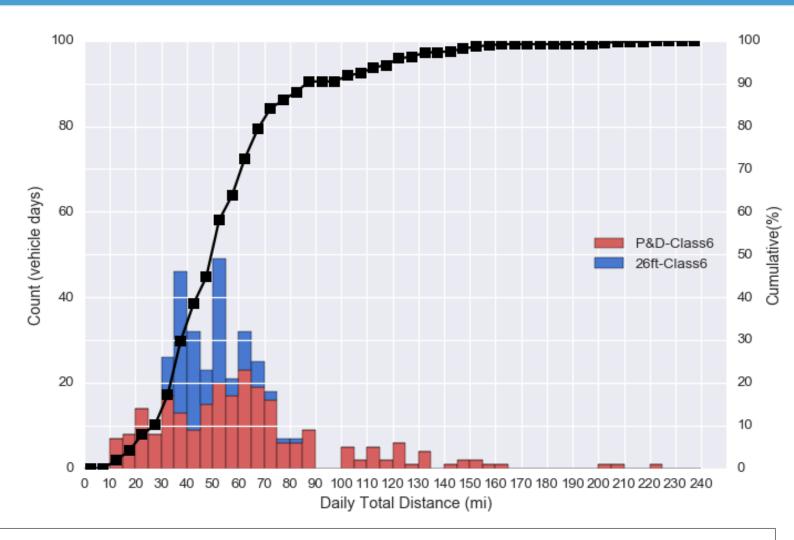
260 Vehicles

2.5M Miles of real-world data analyzed

Median daily VMT ~ 55 miles

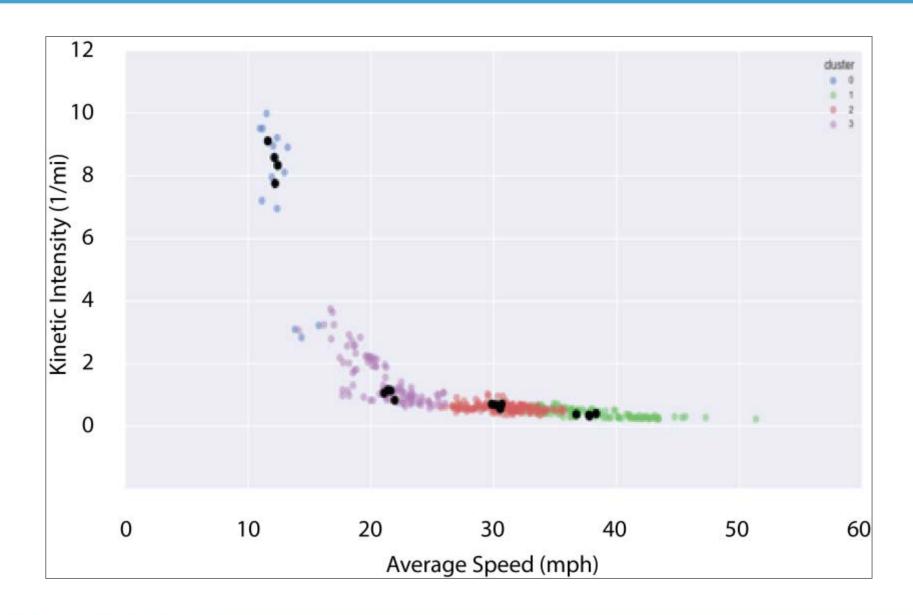
• 90<sup>th</sup> percentile: 80 miles

• 95<sup>th</sup> percentile: 100 miles



Results consistent with other data (VIUS, CalHEAT, etc.)

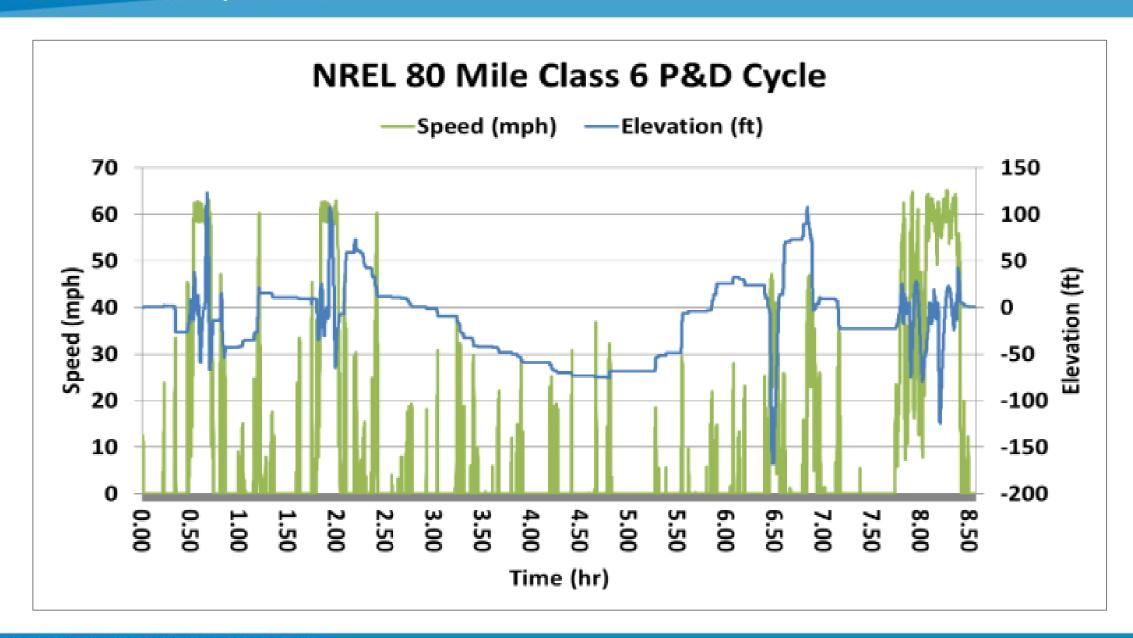
# 80- and 100-Mile Cycles



## Road Grade/Vehicle Speed Constraints



## Final Drive Cycle with Elevation



## **Next Steps**

System development, vehicle integration, and vehicle demonstration underway

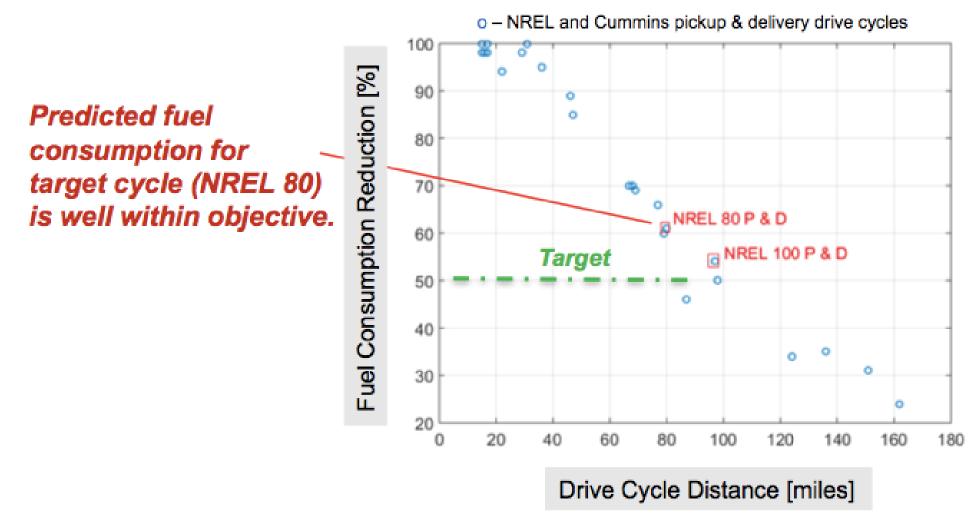


Figure credit: Cummins 2017 Annual Merit Review presentation https://energy.gov/sites/prod/files/2017/06/f34/gi189 kresse 2017 o.pdf

#### Conclusions

- Range extenders can be viable technology option for reducing fuel consumption from MD and HD engines by ~50% or more
  - Need to maximize e-miles to minimize payback period
- MD and HD engines have wide variations in use and duty cycle; real-world data and sophisticated analysis can:
  - Identify vocations/duty cycles best suited for range-extender applications
  - Help guide powertrain optimization and design requirements
- We have only just begun to make use of all the data (and computational power) available — talk to us!



# Thank you!

#### For more information:

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www.nrel.gov/transportation

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