ENHANCING MOVES FOR CONNECTED AND AUTOMATED VEHICLE ANALYSIS

Matthew Barth
Yeager Families Professor, CE-CERT Director
barth@cert.ucr.edu

David Oswald
CE-CERT Graduate Student Researcher



Transportation is undergoing FOUR major Revolutions

Shared Mobility:

- carsharing, ride hailing companies (e.g., Uber, Lyft), and advanced transit
- Drivers: Internet connectivity, convenience, and transportation costs



Electrification:

- electric drivetrains are becoming more common
- Drivers: advances in motors, controls, and batteries



Connectivity:

- Vehicles are increasingly "connected"
- Drivers: cellular communications, dedicated short range communications

Automation:

- Vehicle automation is emerging in many forms
- Automation comes with many social implications



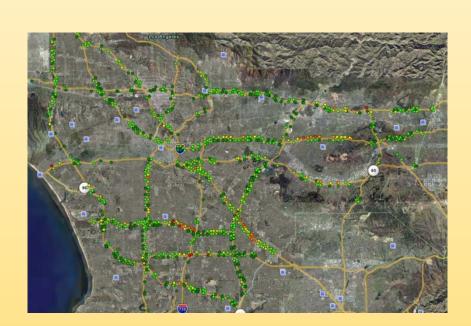


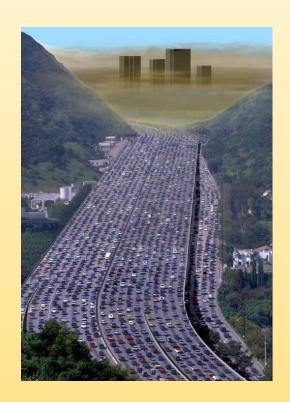
General Components of a Transportation-based Emissions/Energy Inventory:

- emissions/energy factors
- vehicle activity
- fleet composition









FUTURE TRANSPORTATION: MODELING ENERGY & EMISSIONS

Shared Mobility:

- Vehicle Activity: Travel Demand Models need to change
- Vehicle Operation should remain the same

Electrification:

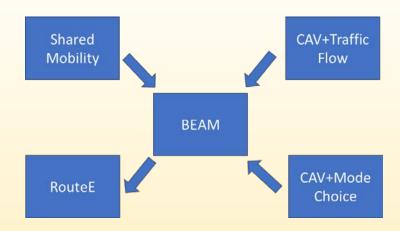
- Vehicle category is simply changed within MOVES
- Indirect Emissions: need to know energy sources that produce electricity

Connectivity:

- Vehicle types will likely be newer, energy efficient, and low emissions
- Vehicle Operation/Activity will likely be smoother

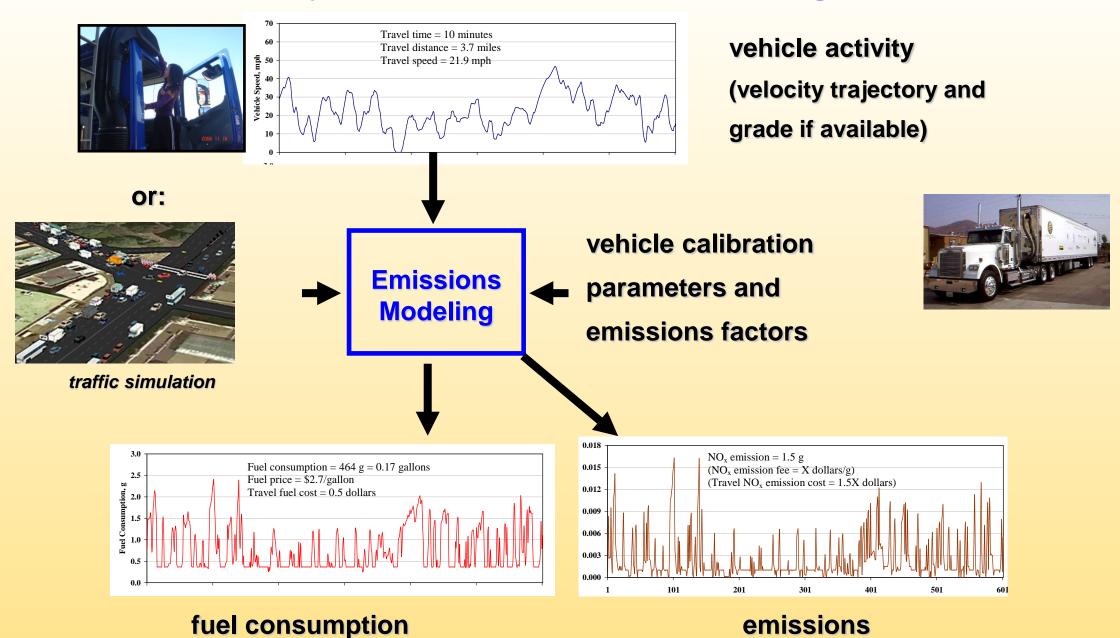
Automation:

- Vehicle types will likely be newer, energy efficient, and low emissions
- Vehicle Operation/Activity will likely be smoother

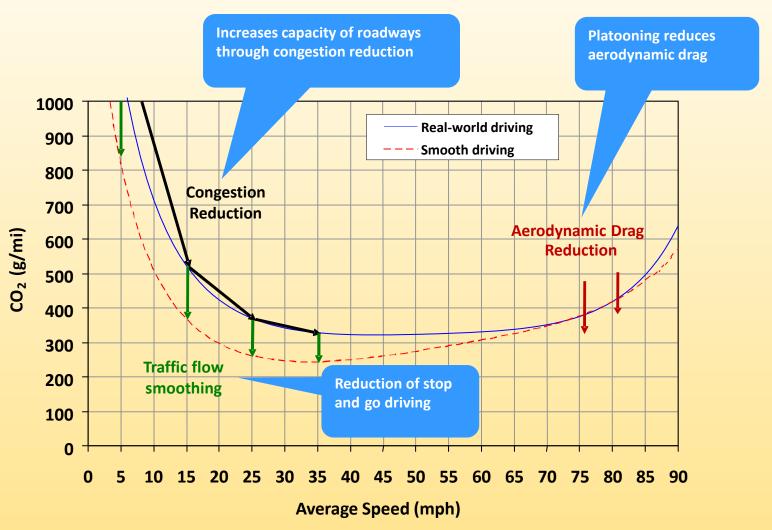




Project-Level Emission Modeling

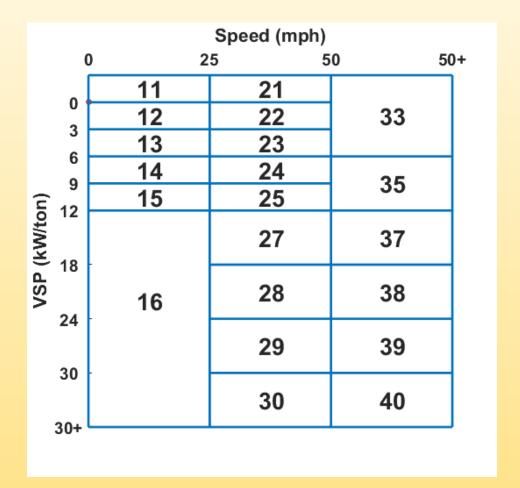


Three regimes on how Connected & Automated Vehicles can reduce on-road energy and emissions



Motor Vehicle Emission Simulator (MOVES)

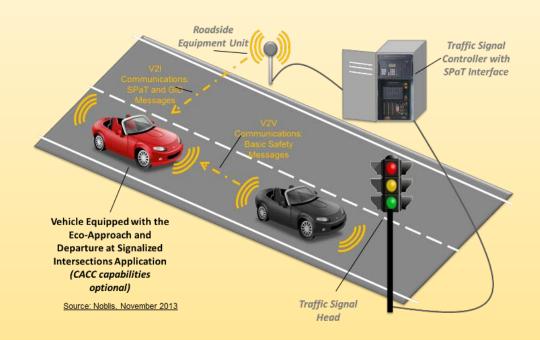
- Data-Driven Emission Model
- Uses a Binning Approach for Vehicle Operation Mode (OpMode) and Emission Factors



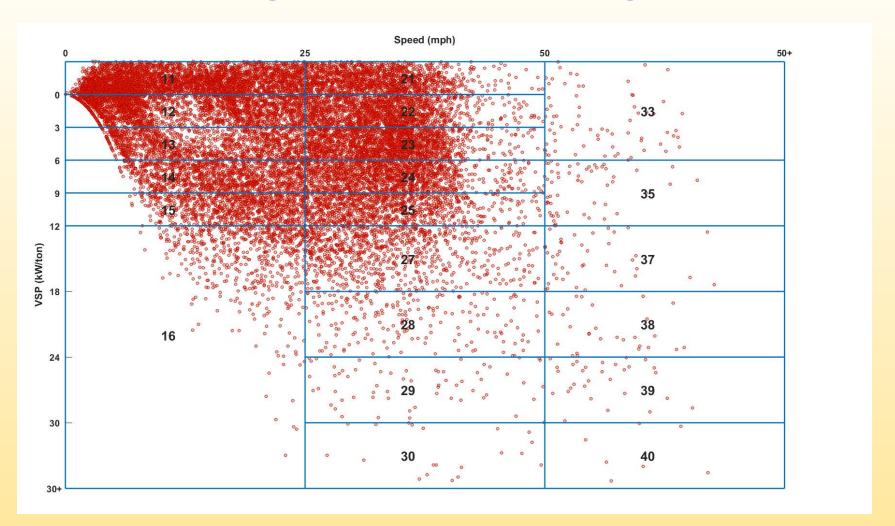
MOVES Sensitivity Analysis

- In our real-world connected and automated vehicle experiments, we noticed that MOVES was under predicting the energy and emissions benefits
- We initiated a study to compare real-world fuel consumption, MOVES, and CMEM



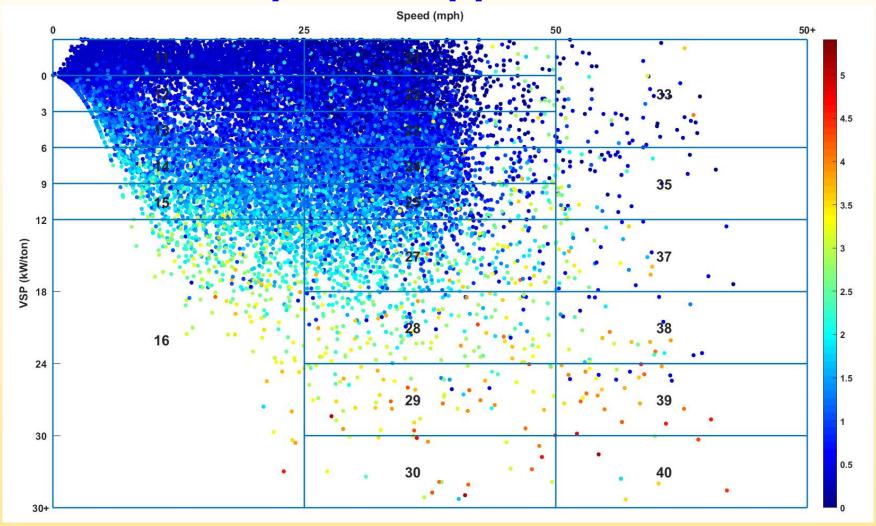


Calibrating MOVES: Training Data Set



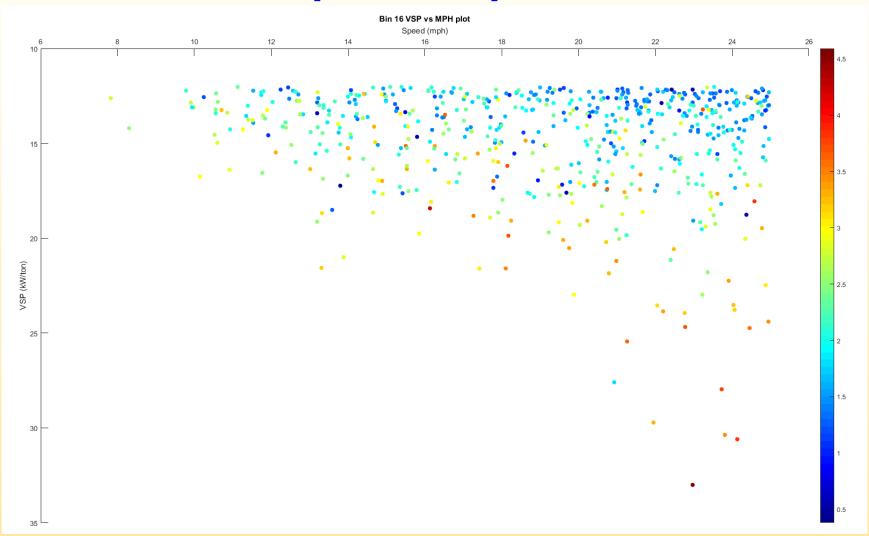
- Second-by-second data were collected from numerous driving trips and experiments
- Data are used to calibrate MOVES fuel consumption factors and activity in the various OpModes

Fuel Consumption mapped to MOVES bins



- Second-by-second fuel consumption (grams/s)
- Max value: 5.41 grams/s

Fuel Consumption map of MOVES Bin 16



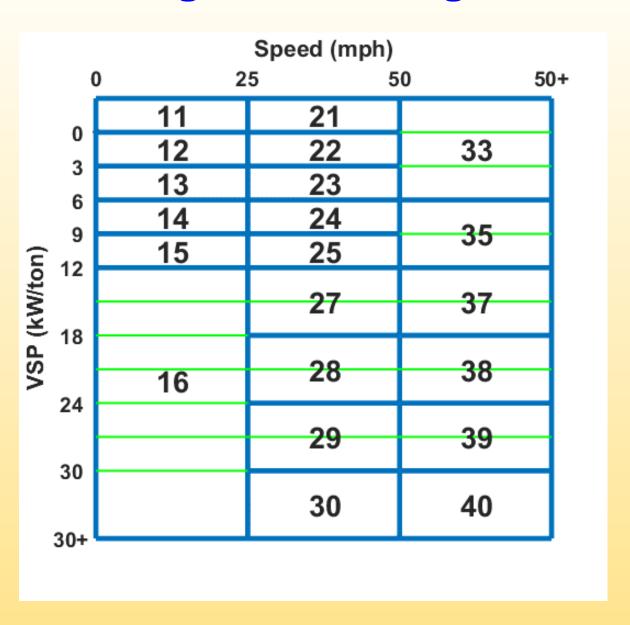
- Wide range of values with lower VSP values dominated by lower fuel values
- Range: [0.38, 4.59]

MOVES Bin Statistics

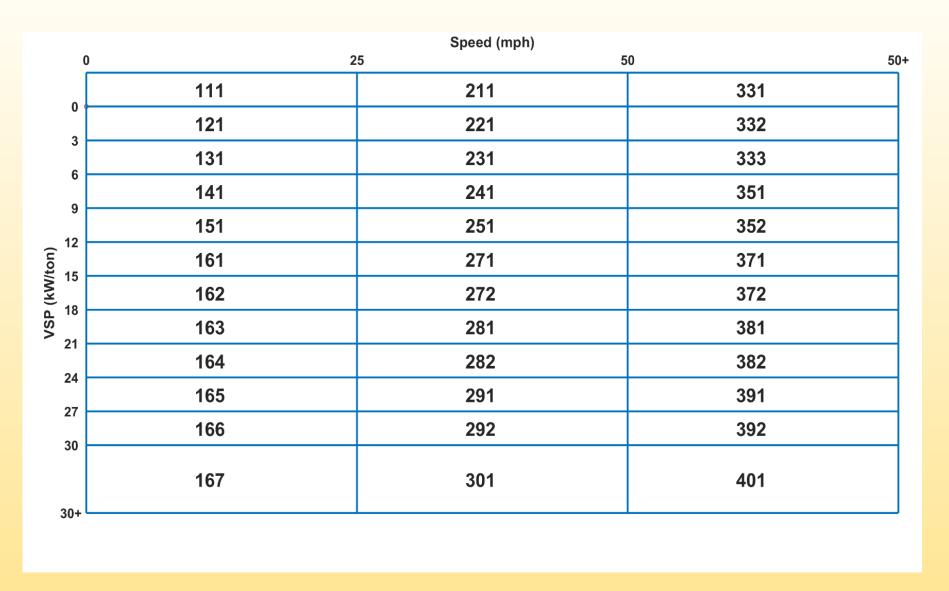
| 11 | # of Values: 6057 Range: [0.199, 3.01] | Mean = 0.324 S.D. = 0.177 | 21 | # of Values: 2923 Range: [0.23, 3.6266] | Mean = 0.195 S.D. = 0.323 | 33 | | |
|----|---|------------------------------|----|--|------------------------------|----|---|-------------------------------|
| 12 | # of Values: 2335 Range: [0.21,2.967] | Mean = 0.624 S.D. = 0.357 | 22 | # of Values: 1631 Range: [0.238,2.73] | Mean = 0.465 S.D. = 0.403 | | | Mean = 0.511 S.D. = 0.869 |
| 13 | # of Values: 1288 Range: [0.24, 2.856] | Mean = 0.953 S.D. = 0.429 | 23 | # of Values: 1931 Range: [0.24, 3.566] | Mean = 0.714 S.D. = 0.412 | | | |
| 14 | # of Values: 1229 Range: [0.27, 3.5] | Mean = 1.22 S.D. = 0.437 | 24 | # of Values: 1470 Range: [0.26, 3.744] | Mean = 0.939 S.D. = 0.417 | | | Mean = 1.2 S.D. = 1.139 |
| 15 | # of Values: 914 Range: [0.34, 3.586] | Mean = 1.501 S.D. = 0.451 | 25 | # of Values: 1002 Range: [0.247, 3.59] | Mean = 1.18 S.D. = 0.507 | 35 | | |
| | # of Values: 600 Range: [0.38, 4.59] | Mean = 2.07 S.D. = 0.699 | 27 | # of Values: 842 Range: [0.31, 4.11] | Mean = 1.64 S.D. = 0.651 | 37 | # of Values: 42 Range: [0.368, 3.75] | Mean = 1.908 S.D. = 1.1454 |
| 16 | | | 28 | # of Values: 188 Range: [0.43, 4.37] | Mean = 2.39 S.D. = 0.831 | 38 | # of Values: 28 Range: [0.4, 4.1159] | Mean = 2.14 S.D. = 1.282 |
| | | | 29 | # of Values: 62 Range: [0.36, 5.1] | Mean = 3.17 S.D. = 0.892 | 39 | # of Values: 30 Range: [0.4566, 4.745] | Mean = 3.23 S.D. = 1.435 |
| | | | 30 | # of Values: 30 Range: [2.44, 5.22] | Mean = 3.745 S.D. = 0.856 | 40 | # of Values: 30 Range: [0.67, 5.41] | Mean = 3.83 S.D. = 1.54 |

Improving Resolution Using Sub-Binning

• Green lines show new bin borders

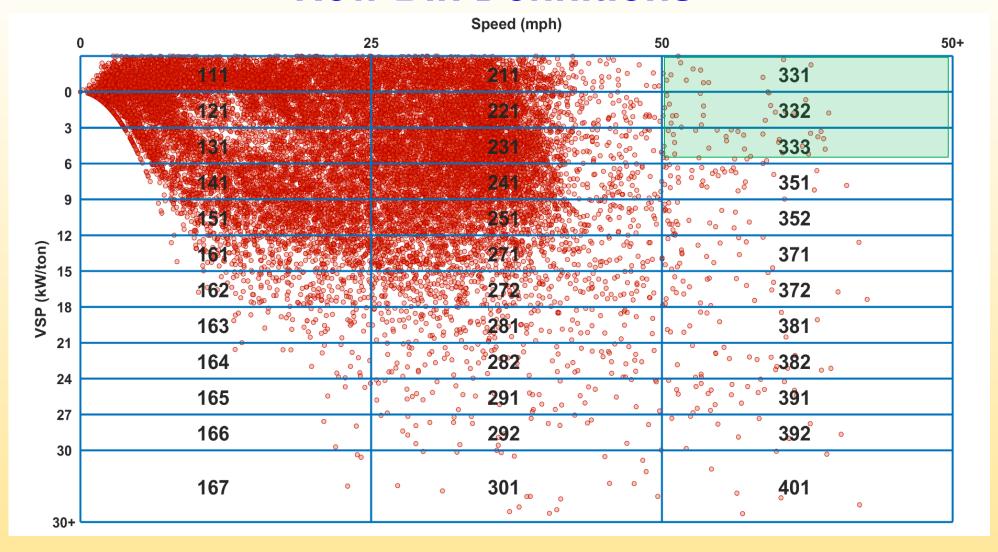


New Bin Definitions



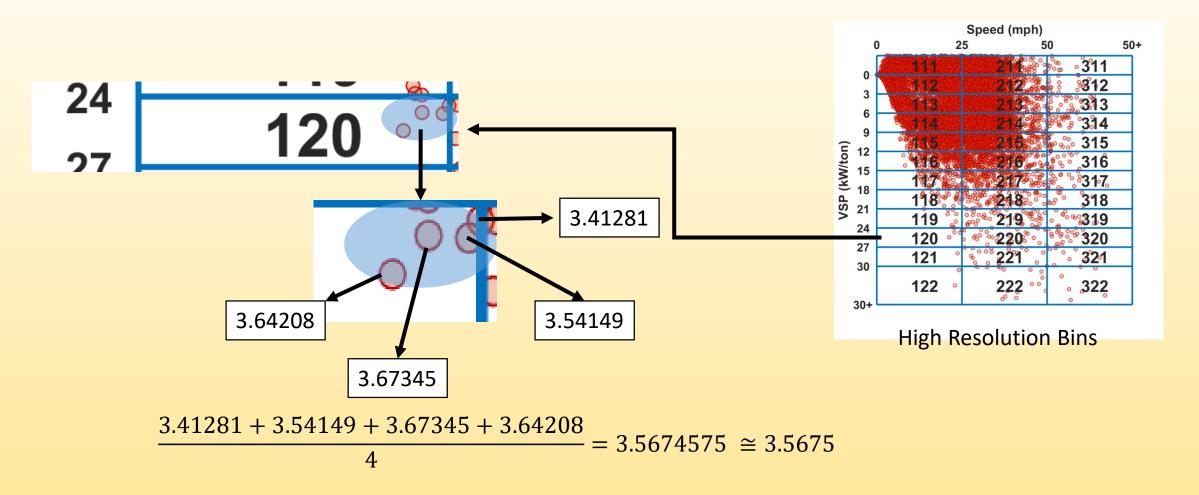
New Bin Numbering System

New Bin Definitions



- Training data plotted in new bins
- Green boxes highlight what was changed

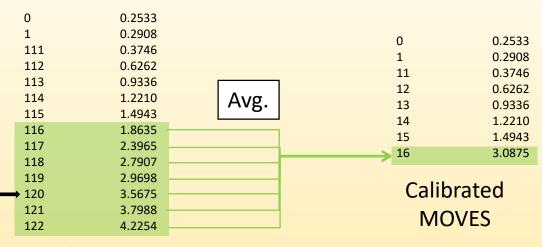
Emissions & Fuel Factors



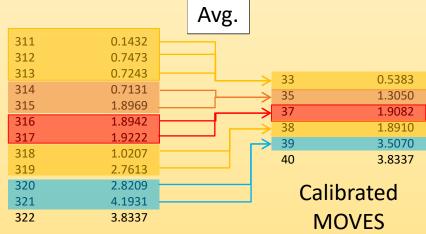
Emission factor for bin 120 is 3.5675

Do the same process for each bin of High Res. MOVES

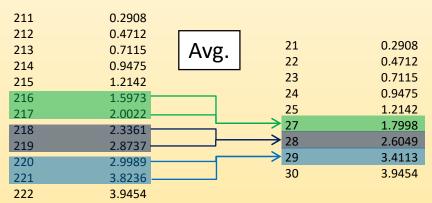
Calibration of Emissions & Fuel Factors



High Res.



High Res.



High Res.

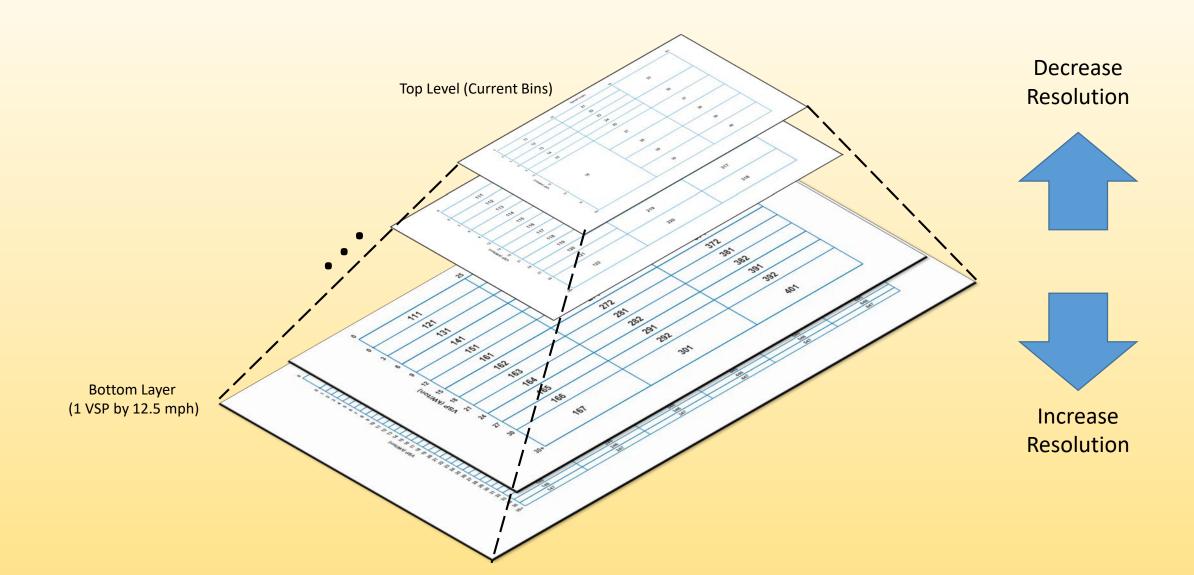
Calibrated MOVES

Comparison Results

| Fuel Consumption Avg. g/mile | | | | | | | |
|------------------------------|----------|--------|------------------|--------------------|--|--|--|
| Method | Measured | MOVES | Calibrated MOVES | High Res. MOVES | | | |
| gram/mile | 147.9 | 170.5 | 154.69 | 150.41 | | | |
| % Diff | | +15.3% | +4.59% | +1.69% | | | |

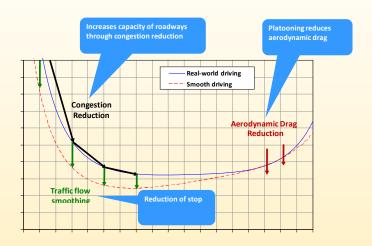
- Data from typical driving
- MOVES: uncalibrated results
- Calibrated MOVES: MOVES calibrated based on vehicle fuel consumption data
- High Resolution MOVES: uses sub-bins

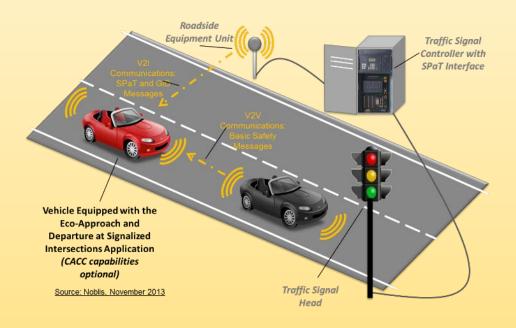
Extrapolation: Developing a Bin-Pyramid

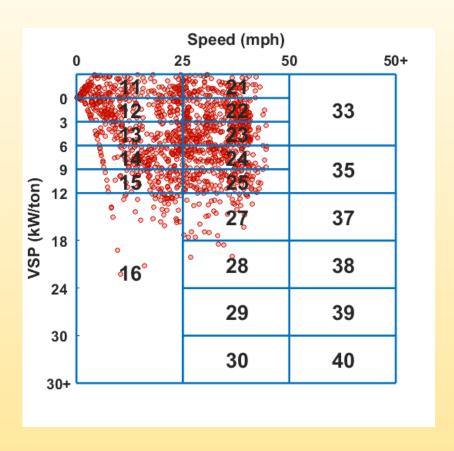


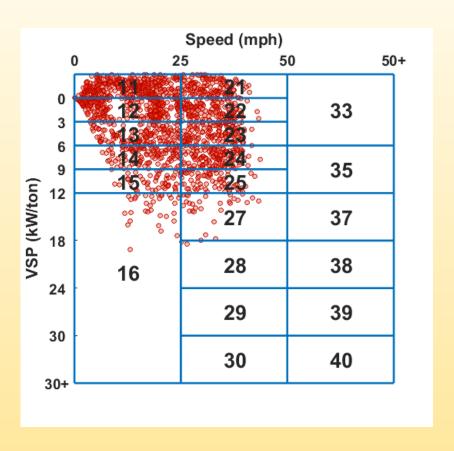
Trajectory Smoothing

- There are many applications that attempt to "smooth" trajectories without loss of travel time: Eco-Pedal, traffic-light anticipation, etc.
- Example: the Eco-Approach and Departure (EAD) connected vehicle application
- Two cars drove at the same time on the same street, one using EAD techniques and the other driving normally

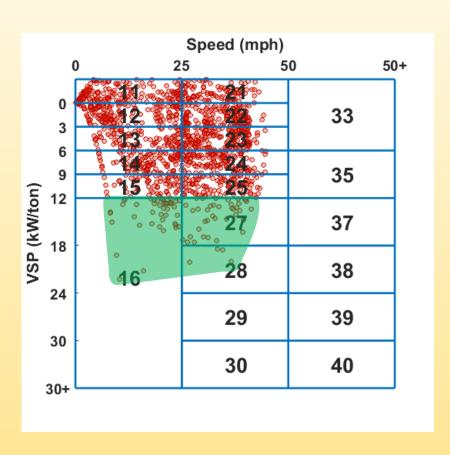


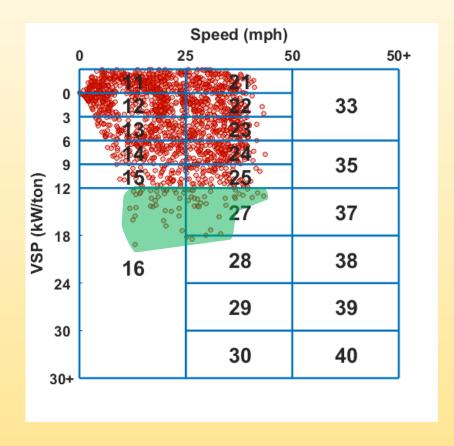


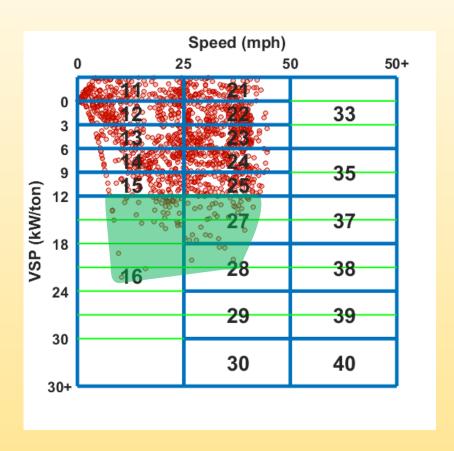


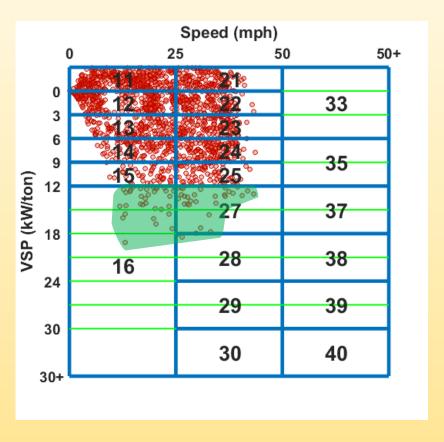


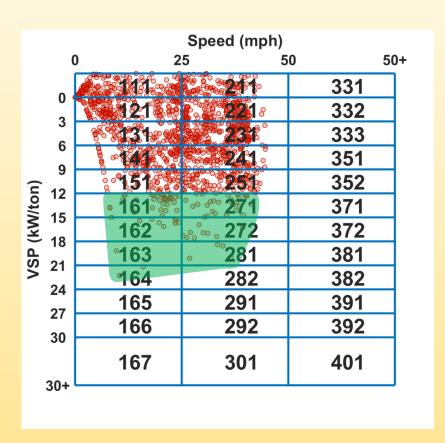
| Fuel Consumption Avg. g/mile | | | | | | |
|------------------------------|----------------|-------|------------------|--------------------|--|--|
| Method | Measured MOVES | | Calibrated MOVES | High Res. MOVES | | |
| No EAD | 137.63 | 158.3 | 144.9 | 140.63 | | |
| EAD | 128.51 | 154.6 | 141.2 | 136.3 | | |
| Improvement | 6.63% | 2.33% | 2.55% | 3.08% | | |

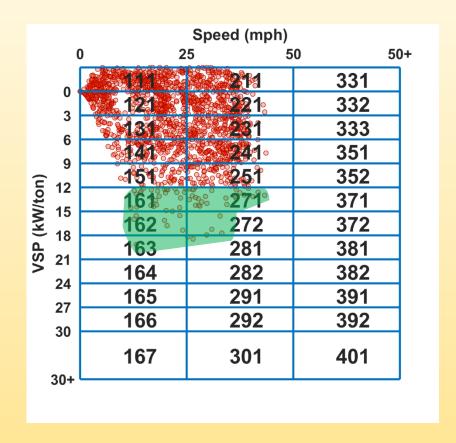












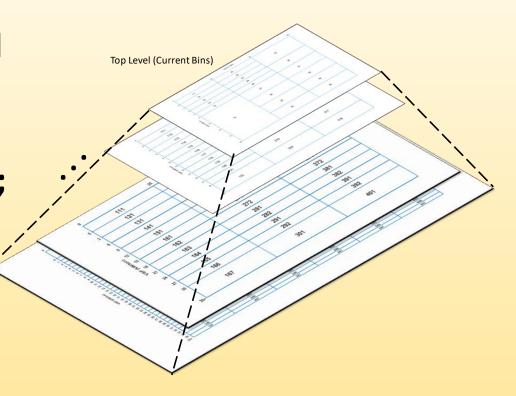
Conclusions and Recommendations

 Traffic Smoothing effects tend to get washed out in MOVES due to bin size

 Recommendation: MOVES can be preserved and enhanced with a sub-binning approach

 MOVES could be used at different "resolutions" using a Bin-Pyramid approach; original MOVES model is preserved

 New Guidance Documents can be written that suggests what resolution should be used → automated resolution determination



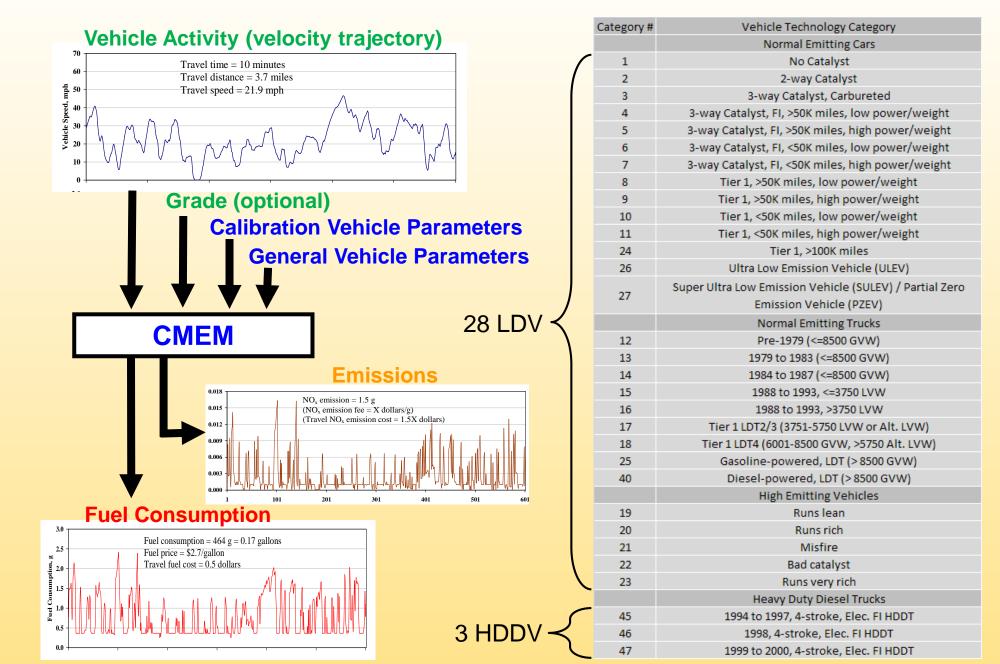
Alternative Vehicle Emission Modeling Approaches

Array of new modeling techniques developed since the late 1990's:

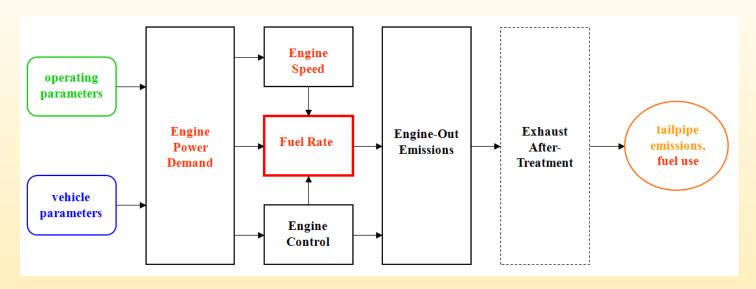
- Fuel-Based Emission Inventories
 - normalizes vehicle emissions to fuel consumption, not VMT
 - requires estimates of fuel use, e.g., from fuel tax
 - generates reasonable emission inventories for large databases
- Modal and instantaneous vehicle emission models:
 - concerned with estimating emissions as a function of vehicle operating mode, (e.g., idle, acceleration, cruise, deceleration)
 - predicts emissions second-by-second
- Statistical Models:
 - Many models exist...

COMPREHENSIVE MODAL EMISSIONS MODEL (CMEM)

- Microscale emission model
 - Developed at UCR CE-CERT
 - Initially developed in the 1990's, lightly maintained
- Sponsorship
 - National Cooperative Highway Research Program (NCHRP)
 - U.S. Environmental Protection Agency (EPA)
- Objective
 - Model vehicle emissions at the project level (sec-by-sec)
 - Accurately reflect the impact on emissions from various operating conditions/parameters
 - vehicle speed, acceleration, and road grade
 - starting conditions, temperature (history effects)
 - secondary engine load, etc.



CMEM Emission Model Structure



- Fuel is a function of Engine Power Demand and Engine Speed
- Fuel rate is related to emissions through analysis based on measured data
- Model Inputs
 - Operating parameters vehicle speed, road grade, accessory power, etc.
 - Vehicle parameters weight, gear ratios, calibrated emission parameters, etc.
- Model Outputs
 - Second-by-second emission data and fuel use

EAD VS. NON-EAD WITH CMEM

| Fuel | Cor | nsum | nptior |
|------|-----|------|--------|
| Α | vg. | g/m | ile |

| Method | Measured | MOVES | Calibrated MOVES | High Res. MOVES | СМЕМ |
|-------------|----------|-------|------------------|--------------------|--------|
| No EAD | 137.63 | 158.3 | 144.9 | 140.63 | 138.97 |
| EAD | 128.51 | 154.6 | 141.2 | 136.3 | 132.5 |
| Improvement | 6.63% | 2.33% | 2.55% | 3.08% | 4.65% |

CONCLUSIONS AND RECOMMENDATIONS

- MOVES Modeling Approach tends to under-estimate traffic smoothing effects due to connected and automated applications
- Sub-Binning Approach can improve resolution
- All Modeling Approaches tends to miss effects of aerodynamic drag reduction effects
- Consider adopting a complementary physical modal or instantaneous emissions model for connected and automated vehicle scenarios, as well as others that have a strong history effect in their emissions generation (SCR, after treatment, etc.)
- Dust off PERE emissions generator model that was previously used for MOVES?

