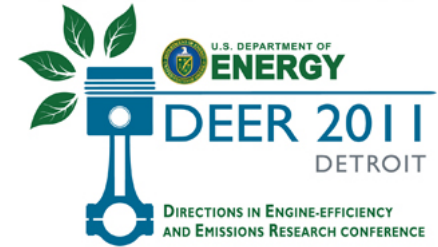




# DAIMLER

**DETROIT DIESEL**



**ATKINSON LLC**

## Demonstrating Fuel Consumption and Emissions Reductions with Next Generation Model-Based Diesel Engine Control

Chris Atkinson  
Atkinson LLC

Marc Allain & Kevin Sisken  
Detroit Diesel, DTNA

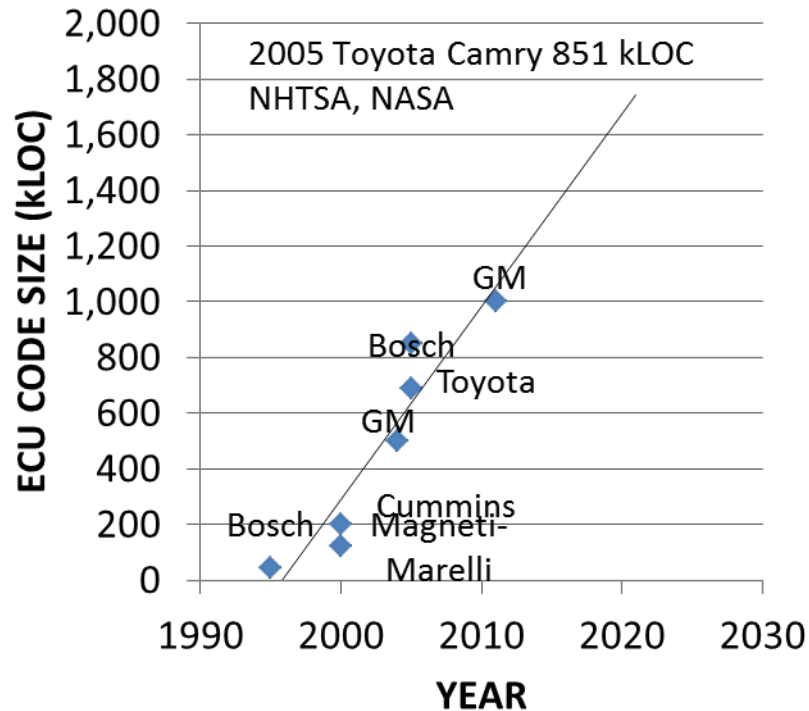
October 2011

- The state of the art in engine control today
- Requirements for engine control in the future
- Our approach to model-based engine control
- The implementation of model-based control and its results
- Accomplishments to date and conclusions

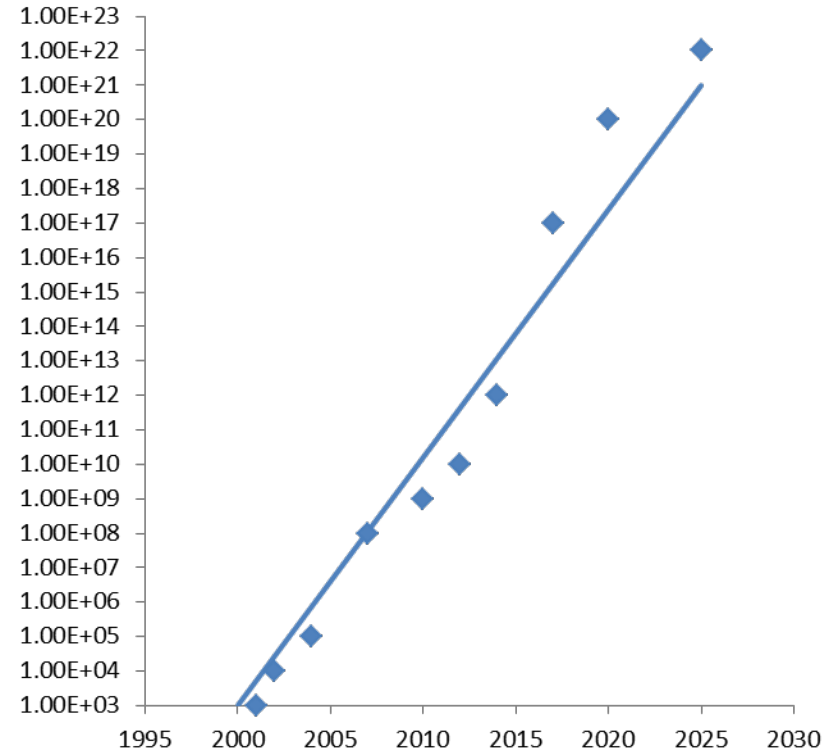
# Control System Complexity – today and in the future

Independent Control Parameters or Orthogonal Variables	Cumulative Number of Control Variables	Date Implemented (actual or projected)
Injection Timing	1	1990s
Injection Pressure Control	2	2002
EGR	3	2002
VGT	4	2007
Aftertreatment Control - DPF	5	2007
Aftertreatment Control - SCR	6	2010
In-cylinder Combustion Feedback	7	2012
Multiple Injection Strategies	8-10	2012
Multiple Combustion Regimes (LTC)	10-12	2014
Waste Heat Recovery	12-14	2017
Hybridization/ Auxiliary Electrification/ Energy Recovery	14-16	2017
Fuel Tolerance/ Advanced Biofuel Capable	16-18	2020
Fully Independent Valve Actuation	18-20	2025
Individual Cylinder Control	20-22	2025
Cycle-by-cycle Control	22-25	2025

# Engine Control Software – Complexity Increase



Software Lines of Code (LOC)



Full Factorial Calibration Space  
(for 10 level variation in each parameter)

- To date, HDD engine control has been focused on and based around emissions reduction on an integrated, cycle-based basis.
- Emphasis moving from emissions reduction to real-time fuel consumption or energy usage minimization.
- We are now at about one-quarter the number of independent control parameters that we will see implemented before 2025.
  - adding roughly one independent control parameter every 1-2 years.
- Each additional independent control parameter – to first order – increases the calibration space by a factor of 10x.

## “The Curse of Dimensionality”

- Currently at about 1,000,000 lines of code in engine controllers.
- Engine control today is a calibration-intensive set of hundreds of algorithms & thousands (or tens of thousands of calibration parameters).
- The trajectory of conventional engine control is an unsustainable increase in cost and effort required to control and calibrate engines.

- Engine control needs to be transformed.
- To date engine control has been dedicated to emissions reduction and compliance.
- But it is transitioning to fuel consumption or CO<sub>2</sub> reduction and energy minimization with tremendous complexity to come.
- Engine control must become more integrated with overall vehicle control.
- Current control and calibration targets will transition to
  - fuel consumption or energy use minimization
  - with power/ energy blending
  - and exhaust conditions amenable to near-zero tail-pipe out emissions levels for emissions compliance.

# An Alternative to Conventional Engine Control

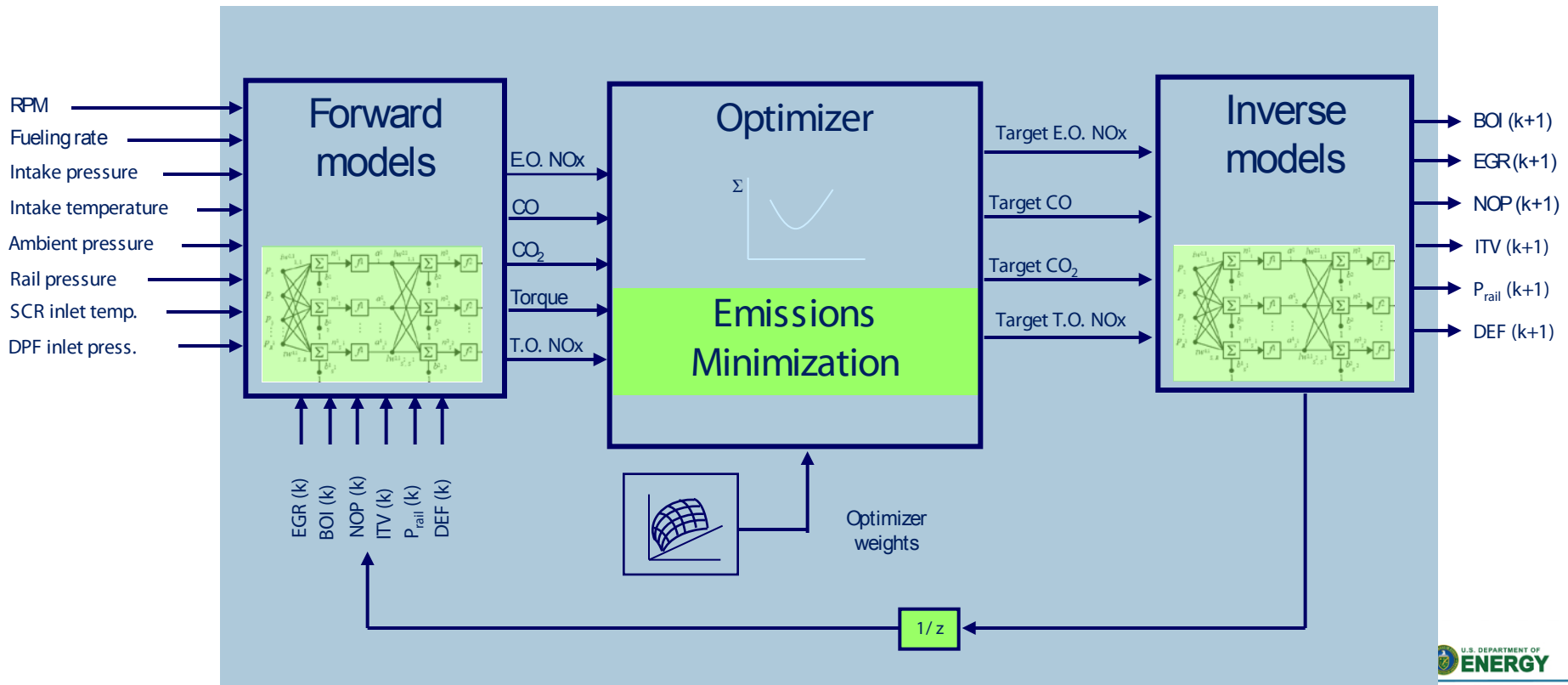
- Model-based engine control
  - Removes the requirement for the exhaustive development of algorithms and strategies.
  - Reduces the calibration requirement significantly.
  - Front-loads the engine testing effort.
  - Shifts the majority of the engineering effort to computational environment and out of the high cost engine test cell.
- Why data-driven models specifically?
  - Are able to determine the nonlinearities between engine cycle demand inputs, engine operating parameters, and outputs (emissions, fuel consumption and performance).
  - Able to make associations automatically and capable of learning.
  - Reduce data and testing requirements to a minimum.
  - Utilize immediate operating history of engine for fully dynamic, transient prediction.

# Predictive Model-Based Engine Controller

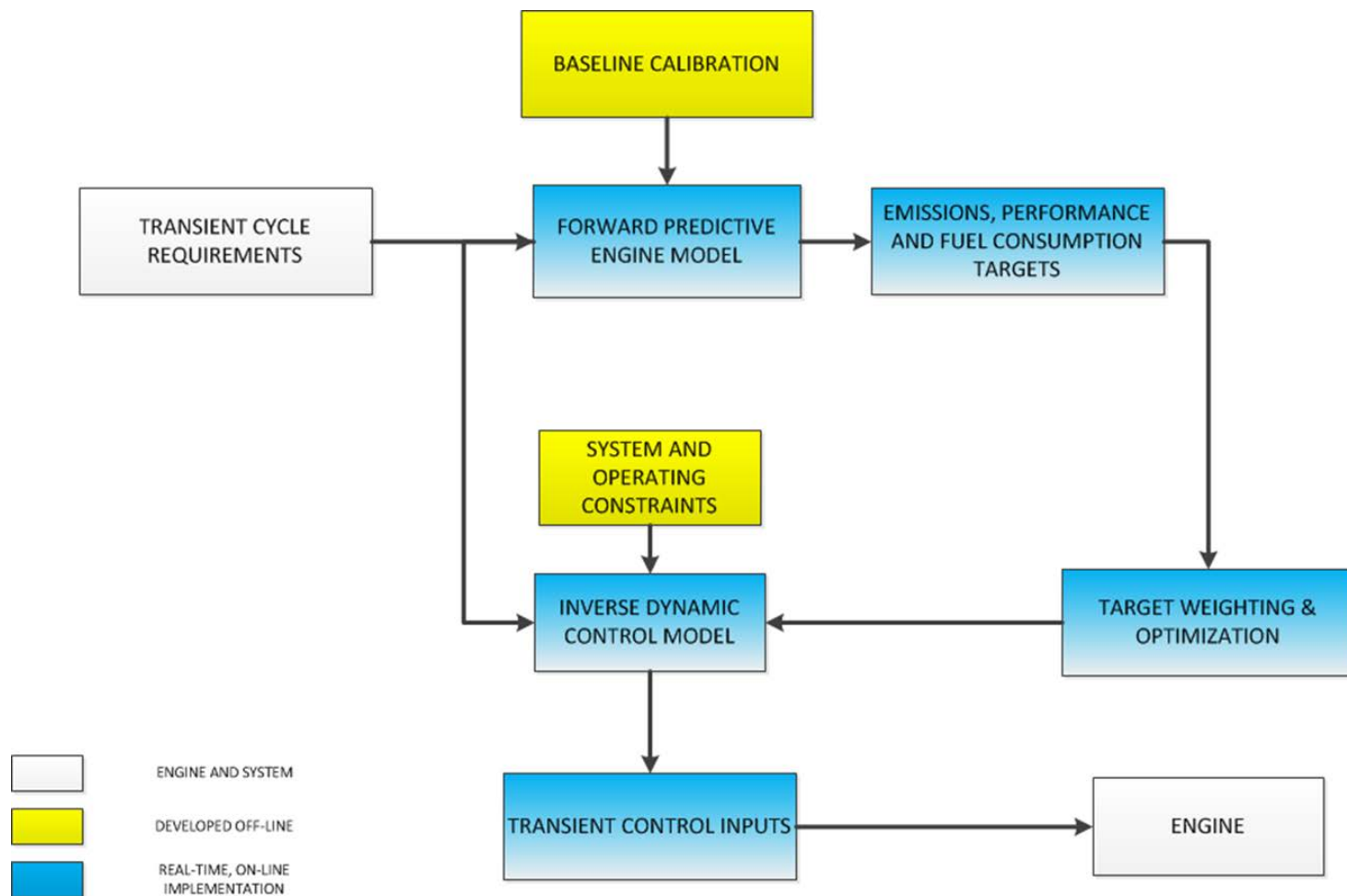
Calculates torque &  
emissions at every  
time step

Calculates “best”  
NO<sub>x</sub>/ CO/ CO<sub>2</sub> combination  
based on optimization weights

Calculates actuator  
outputs based on  
“best” emissions





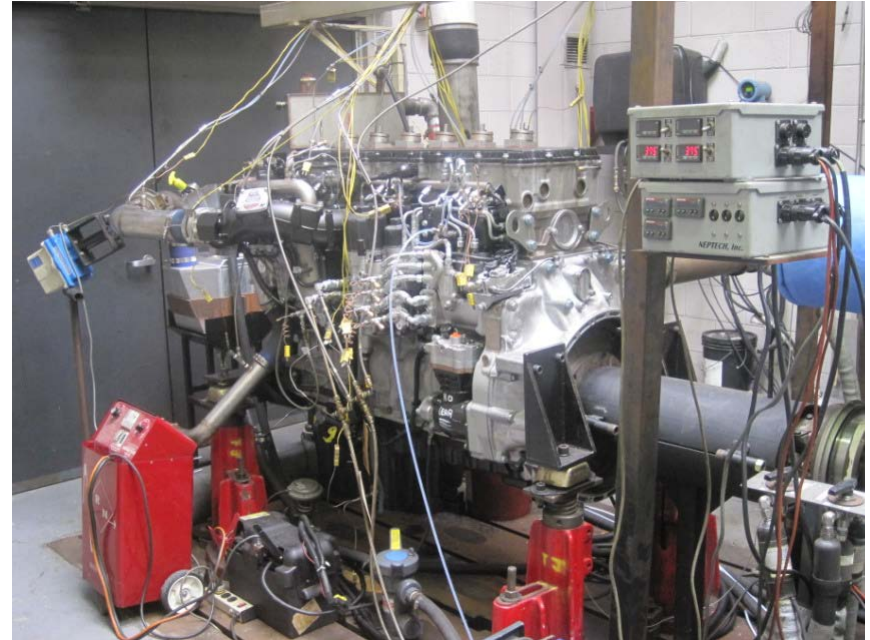


MODEL-BASED  
CONTROL  
SYSTEM

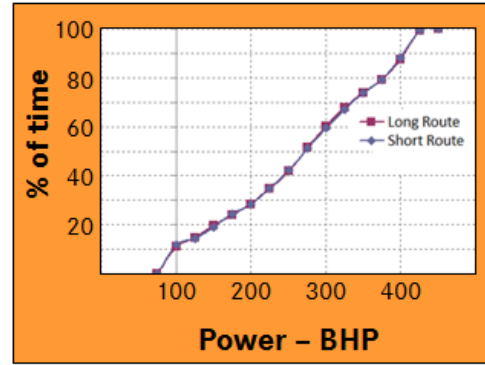
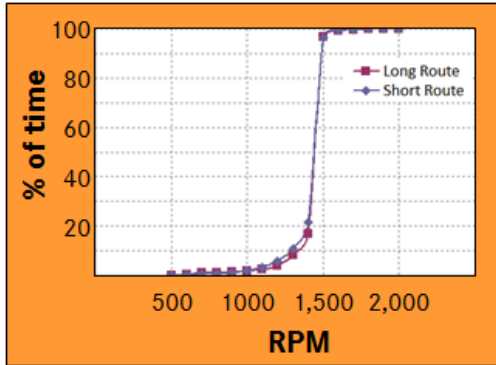
# Application of the Model-Based Engine Controller

## 12.8 liter Detroit Diesel DD13 Engine

- 5 independent control parameters (in addition to speed and fueling)
  - Injection timing
  - Injection pressure
  - EGR
  - Wastegate actuation
  - Rail pressure
- Target values include
  - Instantaneous NO<sub>x</sub>, CO and CO<sub>2</sub>
  - Real-time TQ
- Required ~ 10 hours of high fidelity dynamometer data to develop



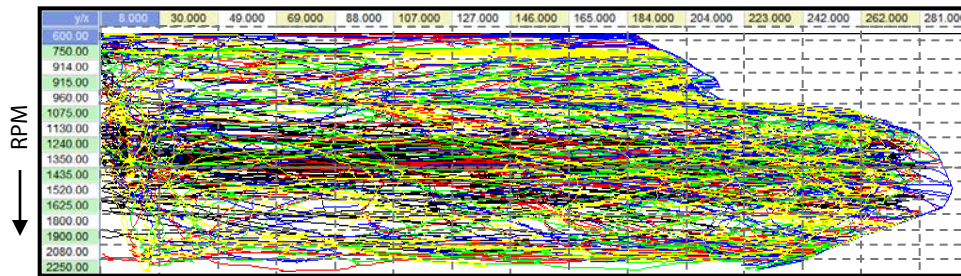
# Controller Development – Data Collection



## Step 1

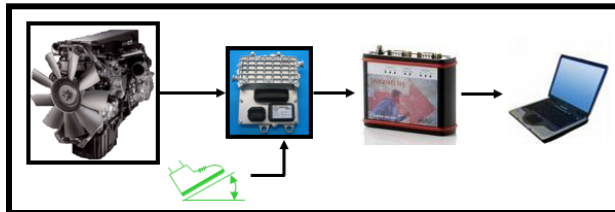
- Generate 20-40 minute dynamometer cycles representative of SuperTruck RPM/ load profiles

Fueling Rate (mg/ stroke) →



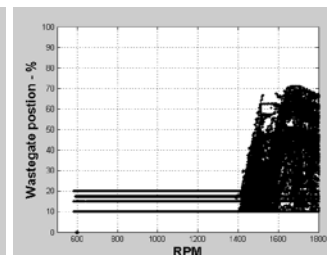
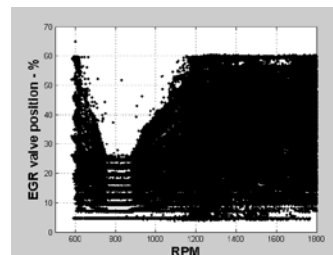
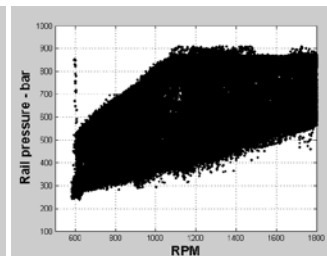
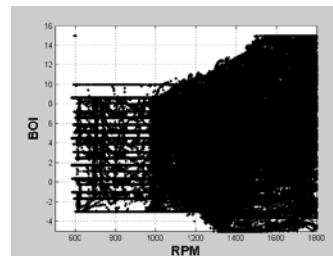
## Step 2

- Generate additional cycles that cover a wide range of transient excursions



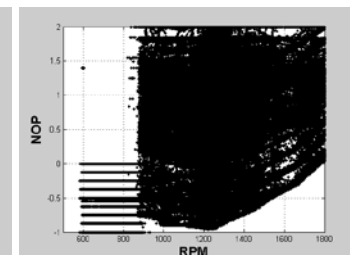
## Step 3

- Enable production ECM bypass



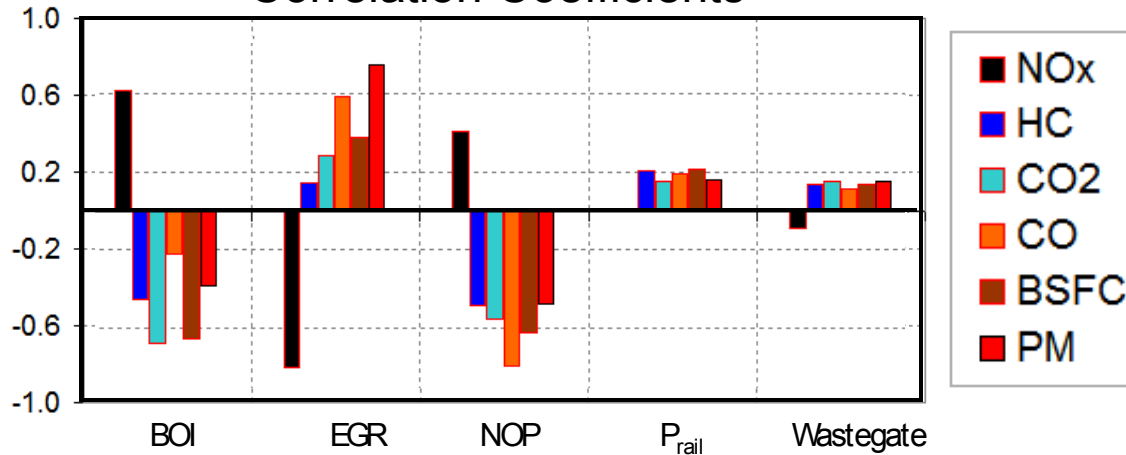
## Step 4

- Exercise engine actuators over a wide range of settings



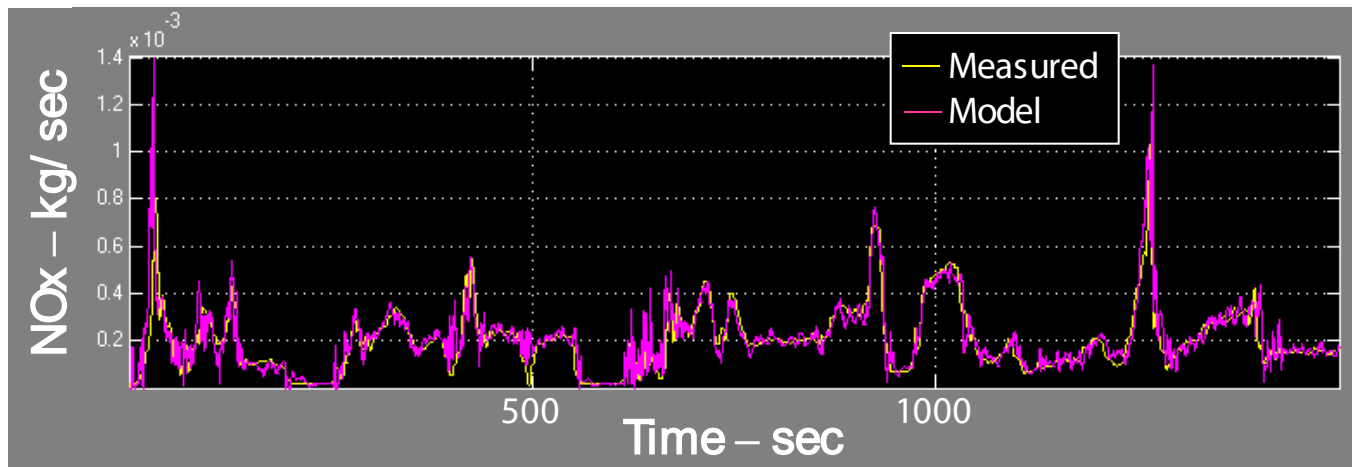
# Controller Development – Neural Network Models

## Correlation Coefficients



### Step 5

- Establish correlation between individual performance parameters and engine control variables
- Define predictive model inputs



### Step 6

- Train models
- Verify model's correlation to measured data

# Model-Based Engine Controller Implementation

- Forward Predictive Models.
- Inverse Control Models.
- Real-Time Optimizer with emissions and fuel efficiency cost function to 'steer' real-time emissions and fuel consumption levels.

## Parameter Description

$k$  – current time period,  $k-1$  – previous time period, etc.

Engine operating trajectory of speed and fueling

$u(k)$  – actual engine control inputs at current time step

$u(k-1)$  – actual engine control inputs at previous time period (history)

$y(k+1)$  – actual, unmeasured, engine outputs (emissions, fuel consumption, performance) at future time step

$Y(k+1)$  – predicted engine outputs (emissions, fuel consumption, performance) at future time step

$U_i(k+1)$  – predicted control inputs, subject to variable emissions, fuel consumption and performance targets (denoted  $i$ )

$B_i$  – modeled forward weights and biases (fixed)

$D_i$  – modeled inverse weights and biases (fixed)

$C_i$  – output emissions, fuel consumption and emissions targets (variable)

Predicted Outputs (calculated using Forward Predictive Models)

$$Y(k+1) = B_1 \cdot u(k) + B_2 \cdot u(k-1) + B_3 \cdot u(k-2)$$

Predicted Controller Parameters (calculated using Inverse Models for single step look ahead)

$$U_i(k+1) = C_i \cdot [D_1 \cdot Y(k+1) + D_2 \cdot Y(k) + D_3 \cdot Y(k-1)]$$

Controller Parameter Option Selection

$$\hat{U}_i(k+1) = \text{optimum}\{U_i(k+1)\}$$

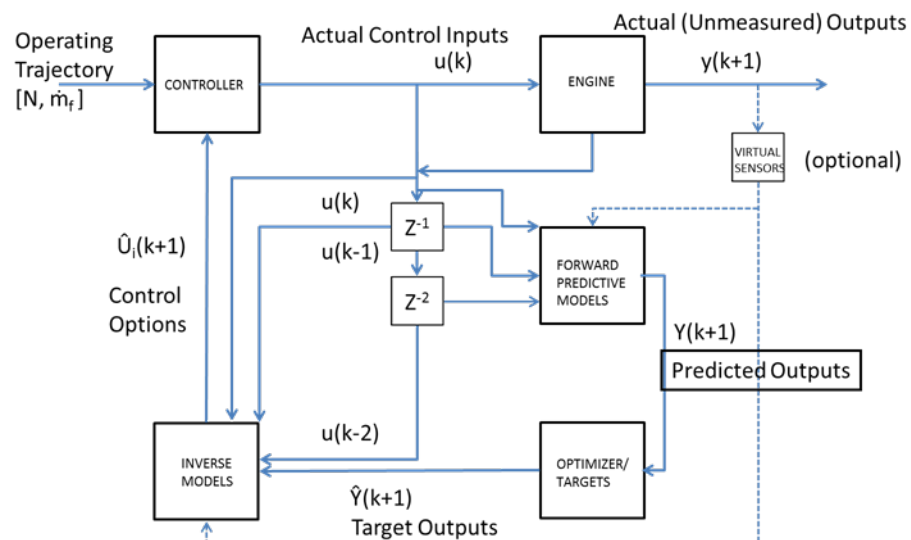
Subject to the constraints:

$$U(k) \in \{u_{\min}, u_{\max}\}$$

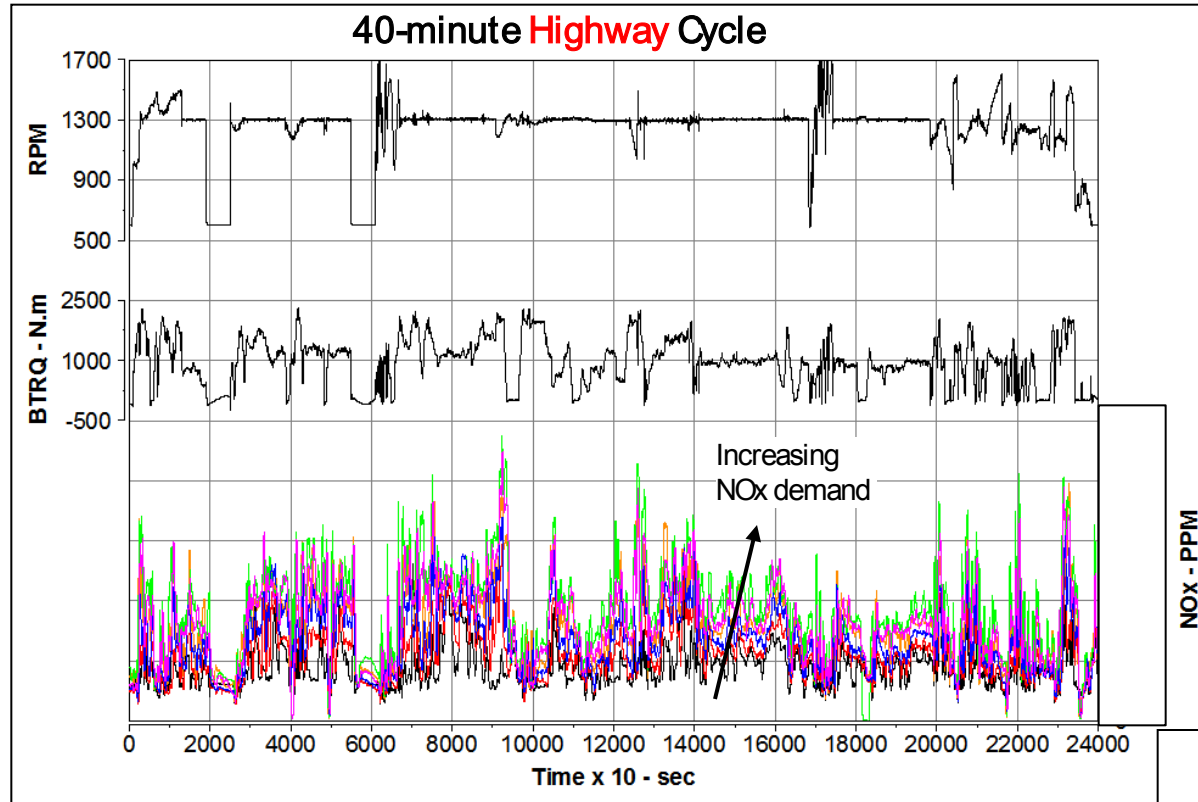
$$|U(k+1) - U(k)| < \Delta U_{\max \text{ slew}}$$

$$|Y_i(k+1) - Y_i(k)| < \Delta Y_{i \text{ max slew}} \text{ (primarily torque)}$$

$$Y_i(k+1) < Y_{i \text{ max}} \text{ (emissions constraints)}$$

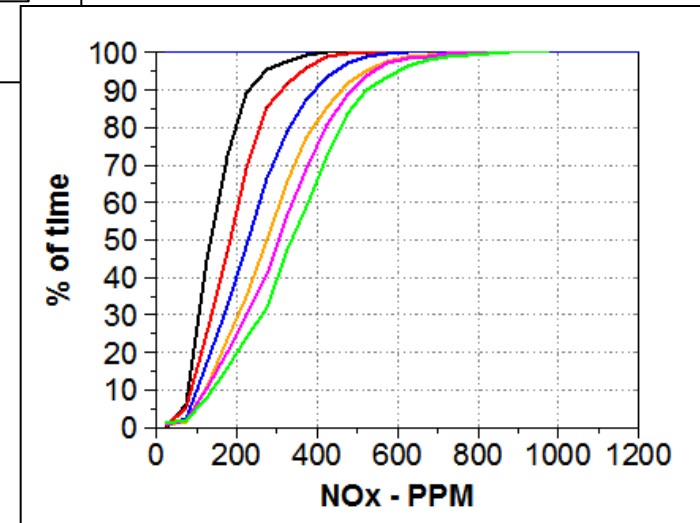


# Evaluation of Model-Based Controller Performance



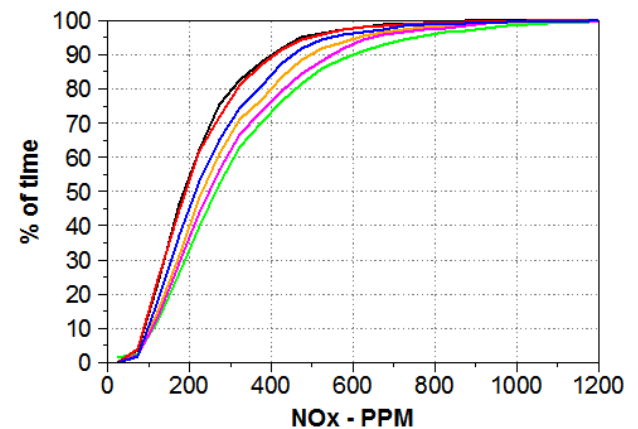
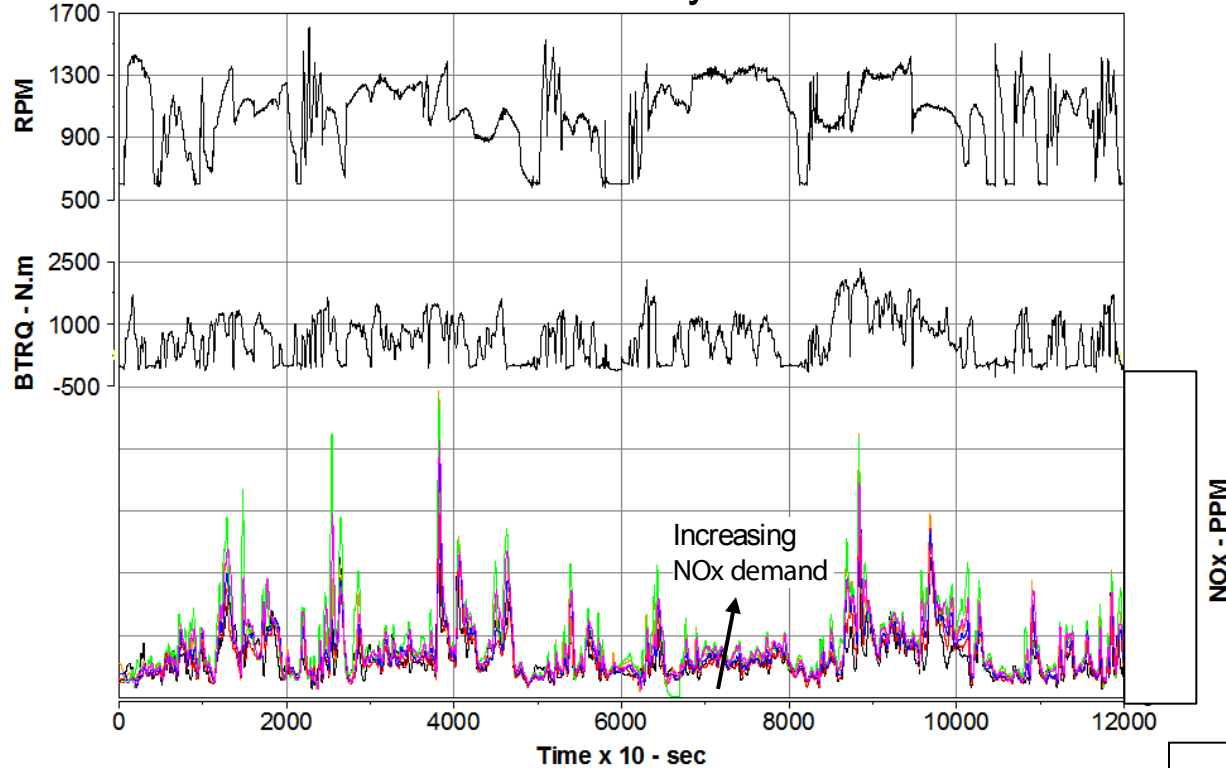
- A single input (NO<sub>x</sub> gain) is needed to drive the controller to higher/ lower NO<sub>x</sub> levels
- Controller response is predictable and repeatable
- NO<sub>x</sub> levels are scaled across the spectrum
- In general fuel efficiency increases with increasing NO<sub>x</sub> levels

- 6 discrete cycles with 6 different levels of NO<sub>x</sub> emissions output requested
- Controller is able to 'steer' emissions levels in real-time



# DD13 Transient Cycle Results

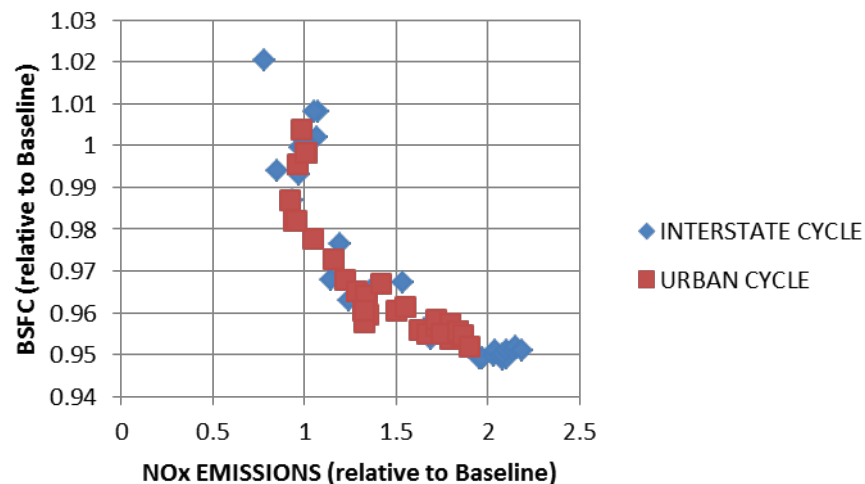
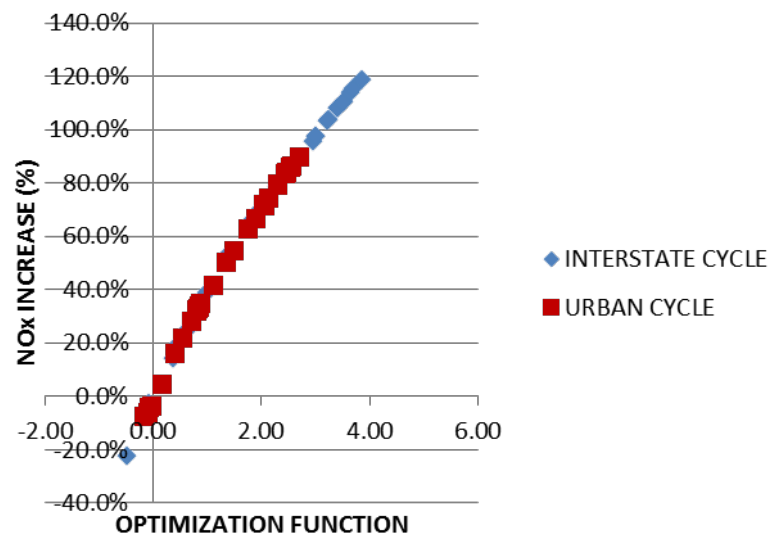
20-minute **Urban** Cycle





## Actual Integrated Cycle Results

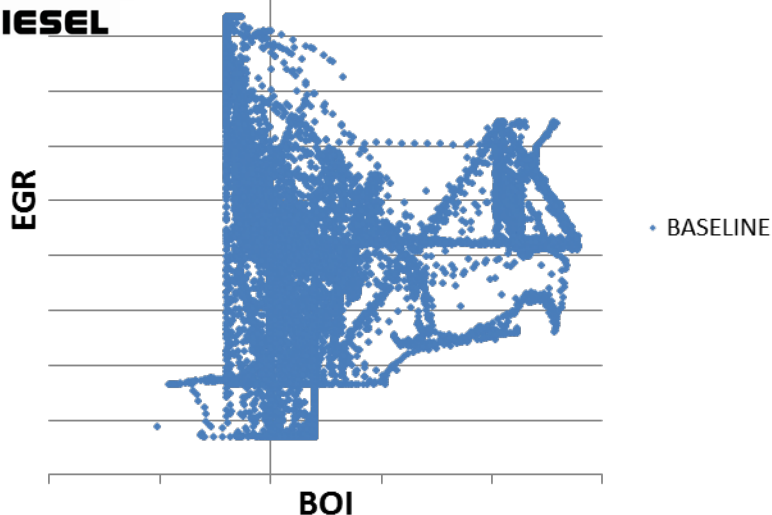
- BS NO<sub>x</sub> varies as demanded by the Optimization Function
- Optimization Function weights can be constant across a cycle (as here) or varied on a point-by-point basis
- BSFC varies with BS NO<sub>x</sub>
  - 2% reduction at the same NO<sub>x</sub> level
  - 4% reduction at 30% higher NO<sub>x</sub>
- Model-based controller demonstrates lower emissions with better fuel economy





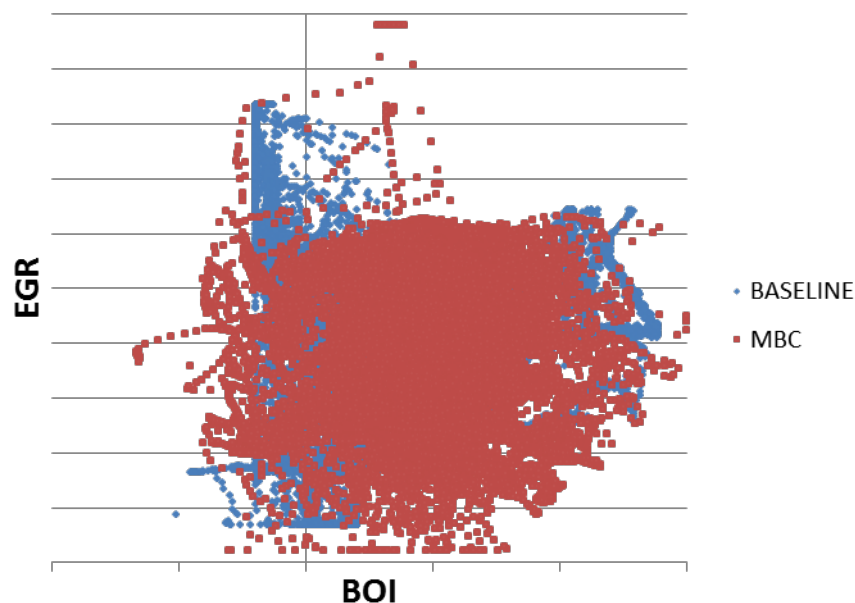


# Model-Based Control reduces Algorithm and Calibration requirements



## Conventional Engine Controller

- Algorithm intensive
- Calibration intensive



## Model-based Controller

- Requires no a priori algorithm development – algorithms replaced by fully predictive models
- Calibration replaced by real-time optimization

- Model-based control has been demonstrated and validated on 3 different engine displacements to date.
- Able to accommodate a range of engine technologies.
- Applicable to a wide range of engine operation and driving cycles.
- Two in-vehicle proof-of-concept tests successfully completed.
- Lower emissions and lower fuel consumption has been demonstrated in a much reduced time frame (and hence at much lower cost).
- Scalable to accommodate future control parameter requirements.

- With model-based control, the calibration task is transformed into one of setting real-time emissions and performance targets.
- Majority of the experimental test cell work is performed upfront in data collection, and not after the fact in calibration.
- Validation and verification in the engine test cell are still required.
- Shifts the emphasis from the high cost physical test environment, while reducing effort required to manageable levels,
- Compatible with virtual sensing, OBD and model-based calibration efforts.
- Model-based engine control allows interaction with vehicle control to allow look-ahead capability and the continuous optimization of fuel consumption (SuperTruck Program).

*Control becomes predictive rather than reactive, with substantial emissions, fuel efficiency and cost benefits.*





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- Falk Beier
- Ken Ball