

Model Based Calibration Optimization Using Machine Learning Algorithms

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Electrified Powertrain Engineering (EPE)
Ford Motor Company

Part I : Introduction

Part II : Infrastructure and Process Development

Part III : Example Study

Part I : Introduction

Manual powertrain calibration optimization

Requires hardware prototypes

Substantial cost

Engineering (design validation) time

Design Validation (DV) efficiency improvement

Use high fidelity CAE powertrain models (Simulink) for initial calibration

Optimized combination of design variables

Reduce prototype vehicles and engineering time

Explore designs which would otherwise not be possible

Electrified vehicles

- Complex hardware architecture and software controls
- 1000s of design (calibration) variables

Finding the optimum design

- Millions of design combinations
- Higher computation time for high fidelity Simulink SIL models
- Substantial amount of computing resources
- Cluster/ parallel computing for simulation
- Machine learning (Genetic Algorithm) for optimization
- 100s (even 1000s) of designs to run in parallel

Parallalization, Licenses and Scalability

Parallel computing

Running 100s/1000s of designs in parallel

Not cost effective with standalone licenses

Parallel/distributed computing licenses for MATLAB & Simulink

Parallel Computing Toolbox (PCT)

MATLAB Distributed Computing Server (MDCS)

Scalable parallel/distributed computing

Final solution test in workstations (before deployed to HPC)

HPC utilization for computation

With varying number of workers

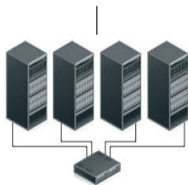
Scalable from sequential to 100s CPUs in parallel

Part II : Infrastructure and Process Development

Distributed Computing Capability

Distributed Computing

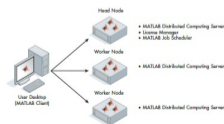
HPC
Ford Internal



Online Clusters
Pay per Use



Local Cluster (Unused CPUs)
MATLAB DCS/Other SW



Cost Effective Parallel/Distributed Computing Licenses

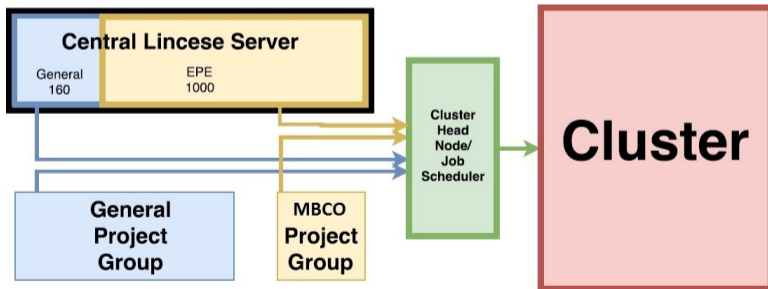
Model Based Calibration Optimization (MBCO) team

Purchased 1000 MDCS licenses

Added to existing PCT/MDCS license pool

Dedicated "Project Group"

For license management



Machine Learning Algorithms for Optimization

Need of machine learning algorithms

All design possibilities can't be evaluated

 With in the simulation budget and available time

Change calibration variables

Optimize for multi-objectives

Subjected to multi-constraints

Alternatives evaluating (in-progress)

No "all-in-one" solution algorithm

MATLAB Global Optimization Toolbox

modeFrontier

Several other commercial optimization SW packages

Currently modeFrontier is used as the optimization tool

Compatibility with non MathWorks SW tools

Off the shelf optimization algorithms

"Off the shelf" optimization SW packages - Issues

Parallel jobs are launched using internal schedulers

Individual job submitted to cluster

Uses standalone license

Do not use PCT and MDCS for MATLAB & Simulink

Not suitable for large scale parallel computing

Final Requirements and Solution

Combining modeFrontier and MATLAB

MATLAB - Capable of parallelization to any scale

modeFrontier - License issue

Simulations to run in HPC

Must use distributed computing licenses

PCT and MDCS

Jobs submitted to HPC from a remote MATLAB Client

New in-house tool and a process developed

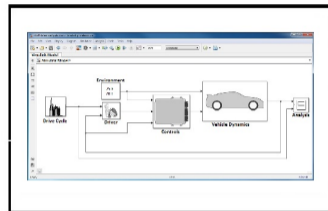
MATLAB GUI - ModefrontierSCVSP

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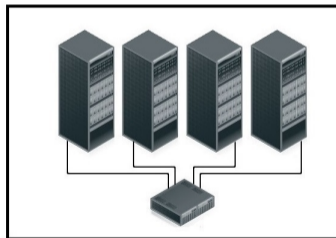
Combines modeFrontier and MATLAB & Simulink

Meets above requirements

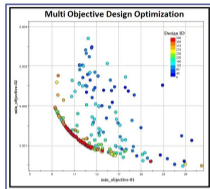
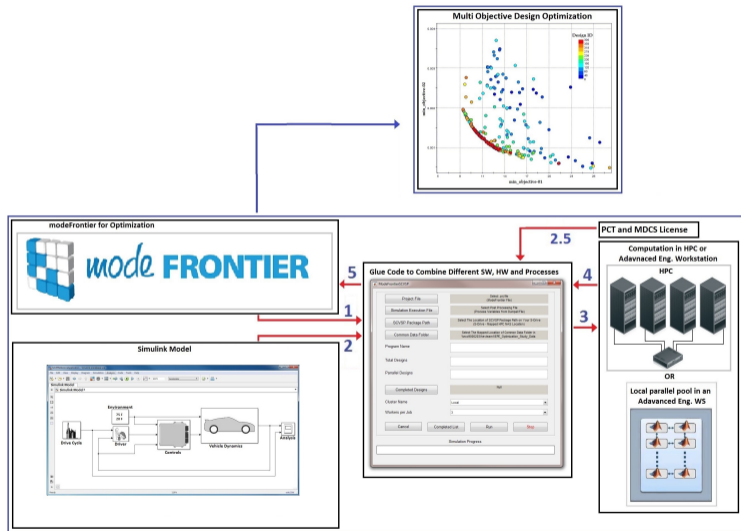
Challenge - 4 Unconnected Dots



**License Usage
PCT and MDCS**



Connecting Dots via a MATLAB GUI - ModefrontierSCVSP



1. Genetic algorithm
2. Model + cal file
- 2.5 License mgt.
3. Job submission
4. Results collection
5. Results reporting

MATLAB GUI - Glue Code

- Combines HW, SW and Licenses seamlessly
- Across both Windows and Linux platforms
- Uses PCT/MDCS
- Manages the large scale Simulink simulations

Modular (runs simulations in)

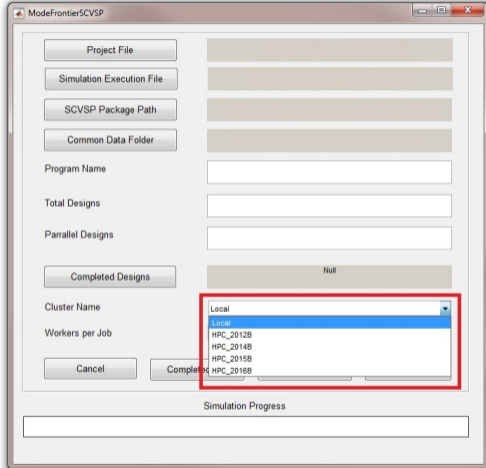
- In workstation with PCT (limited by physical cores)
- HPC with MDCS (limited by MDCS licenses)

User friendly GUI

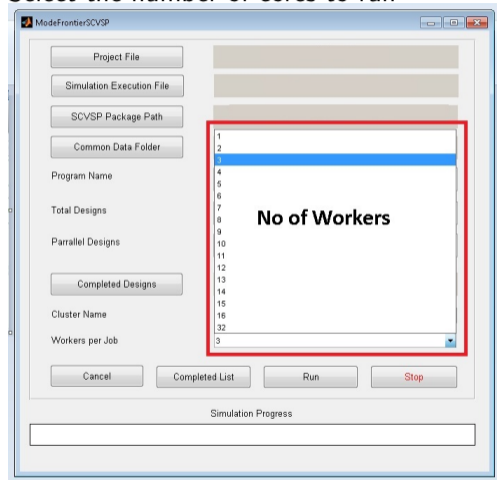
- No coding/programming required
- Easy to train, learn and apply

ModeFrontierSCVSP - Standardized Modular Solution

Select to run in local workstation or HPC



Select the number of cores to run



Possibilities With the New Tool

Scalable parallel/distributed computing

Sequential computing(1 CPU) to parallel in HPC (1000 CPUs)

Click of a button

12000 Simulink simulations (4000-6000 hours)

Completed in less than 15 hours (approx.)

Using 512 simulations in parallel

modeFrontier combined with MATLAB & Simulink

Machine learning algorithms

Design of Experiments (DoE)

Scheduling (Genetic Algorithms)

Multi-objectives (+ multi-constraints) optimization

Selected 12000 out of 1000000+ design combinations

Deployed and used by the MBCO team

Part III : Example Study

Optimizing Vehicle Attributes with Battery Power Table

Battery power table

Optimum power - different conditions
 $f(\text{Driver Pwr, Bat. SOC})$

Linked attributes

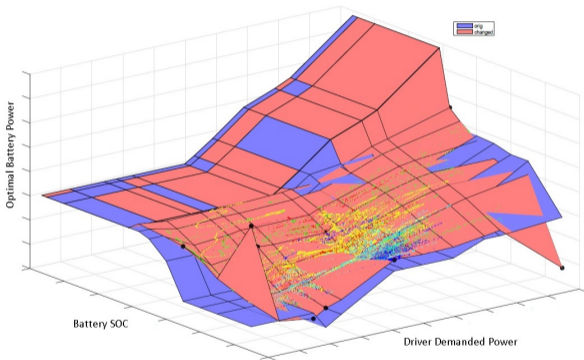
Fuel Economy/Emissions
Performance/Regeneration/NVH etc.

modeFrontier - machine learning

Optimize battery power table*
Using *Genetic Algorithms (MOGA-II)*
Evolutionary (generation) algorithm

Simulation in HPC

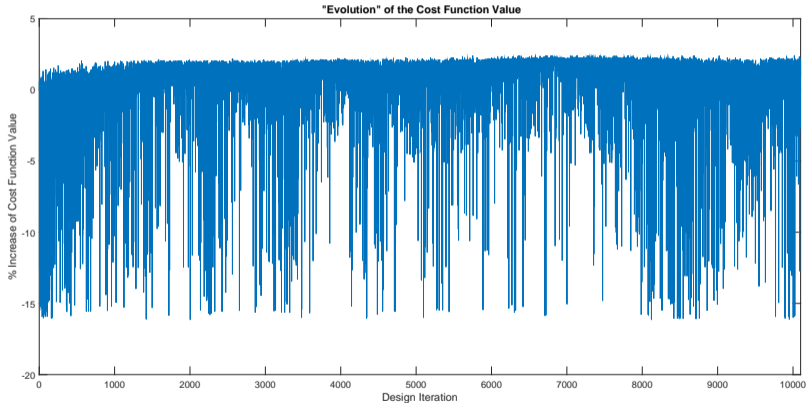
512 in parallel
Population size



*Indirect method used - Non Uniform Rational Basis Spline (NURBS) based
*Developed by Ford Model Based Systems Engineering (MBSE) team (@ RIC)

Optimizing Vehicle Attributes with Battery Power Table*

Cost function - maximize vehicle attribute(s) mentioned above
Single objective optimization**



*Exact details not shared due to confidentiality

*Not validated with prototypes yet

**Unfeasible/error designs excluded



Acknowledgments



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Thank You!!!

