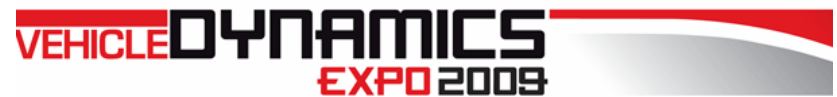


A system engineering approach for the design optimization of an active suspension

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 Task Manager Engineering Services
 Project Engineer Engineering Services



16, 17, 18 JUNE 2009 STUTTGART



1

The intelligent vehicle paradigm

2

The optimal design approach

3

Vehicle Model

4

Actuator Model

5

Conclusions & Outlooks

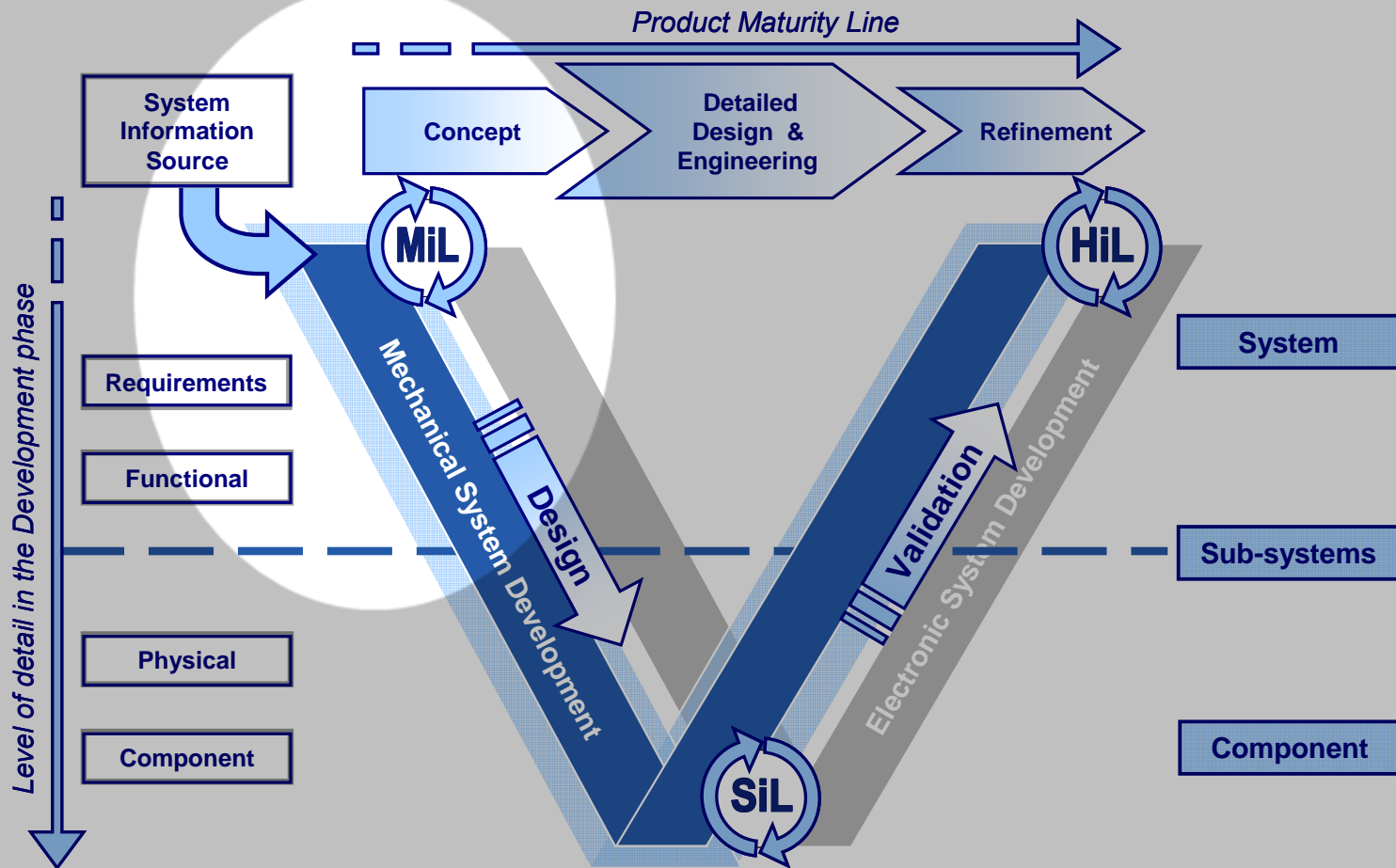
1

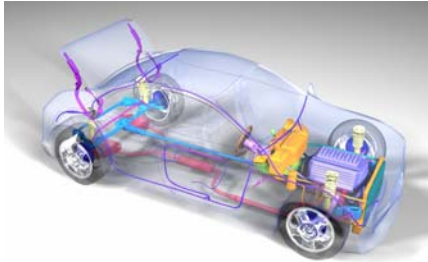
The intelligent vehicle paradigm

- Engineering challenges for intelligent vehicle systems
- LMS Imagine.Lab AMESim®
- noesis OPTIMUS®

Engineering challenges for intelligent vehicle systems

1 ●●●●● 2 ●●●●● 3 ●●●●● 4 ●●●●● 5 ●●●●●





Transmission

Performance and losses,
Comfort, NVH



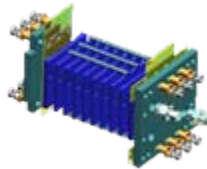
Vehicle Systems Dynamics

Braking, Steering, Suspension,
Vehicle dynamics



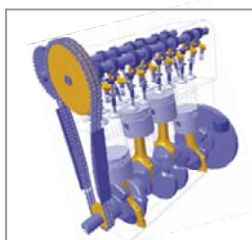
Energy Storage

Fuel Cell, Battery



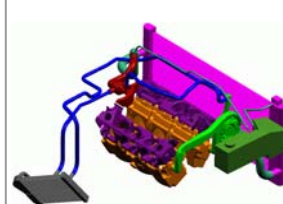
ICE Related Hydraulics

Fuel Injection, VVT, VVA, Engine
compression brake



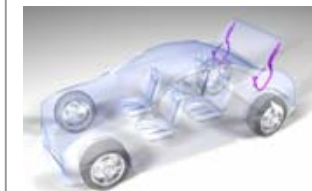
Internal Combustion Engine

Engine control, Air Path
Management, Combustion,
Hybrid Vehicle



Thermal Management

Lubrication, Cooling System,
Air conditioning



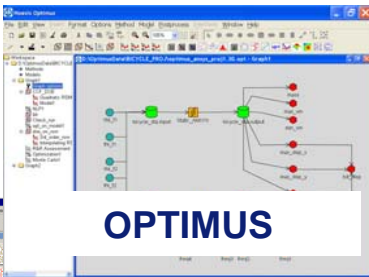
Electrical Systems

Electromechanical components,
Electrical networks

Simulation process capturing



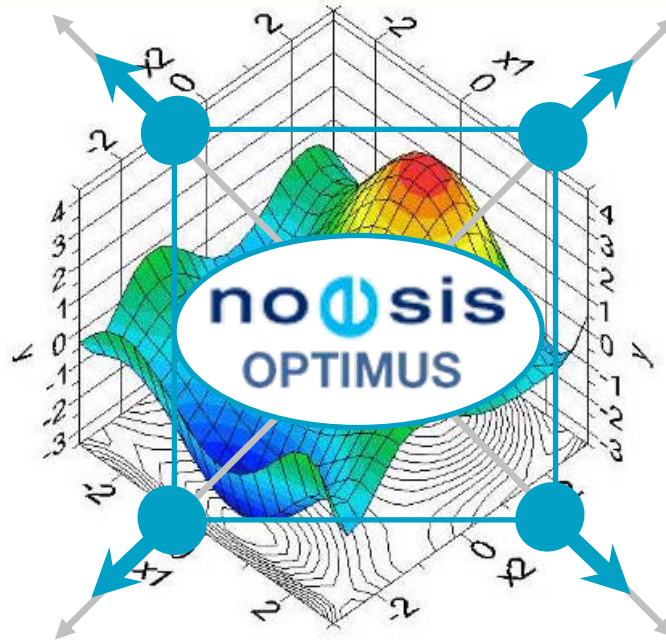
Simulation execution



OPTIMUS

Imagine.Lab

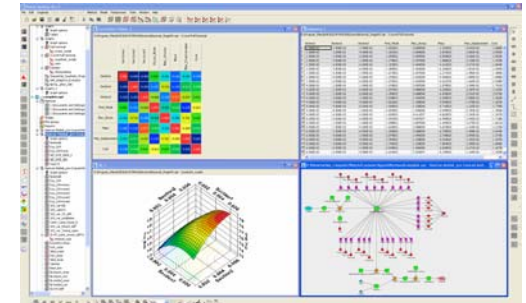
Virtual.Lab



Design Optimization

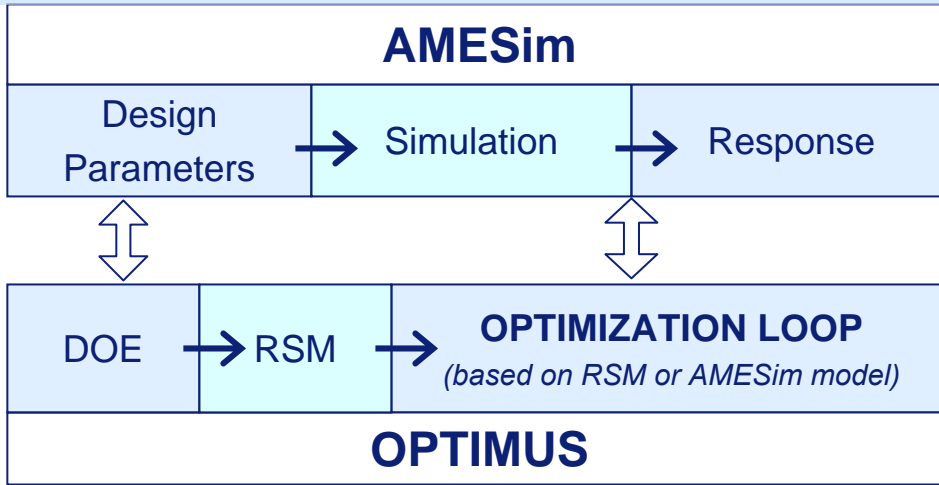
- Design-Of-Experiments
- Multi Objectives Optimization
- Local & Global Algorithms
- Robustness and Reliability analysis

Post-Processing

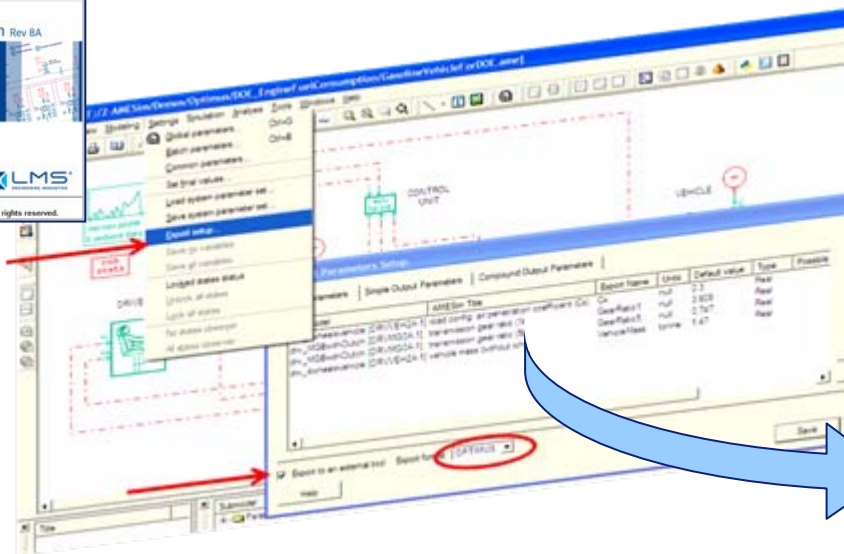


Imagine.Lab + OPTIMUS Functionality overview

1 ● ● ● ● ● ○ 2 ● 3 ● ● ● ● ● ● ● ● ● ● 4 ● ● ● ● ● ● ● ● ● ● 5 ●



- Parameterize and automate any analysis sequence
- Experiment and fit models through the generated data
- Optimize the models or the analysis sequence
- Ensure the reliability and robustness of the design
- Process supported by an intuitive, user-friendly Graphical User Interface



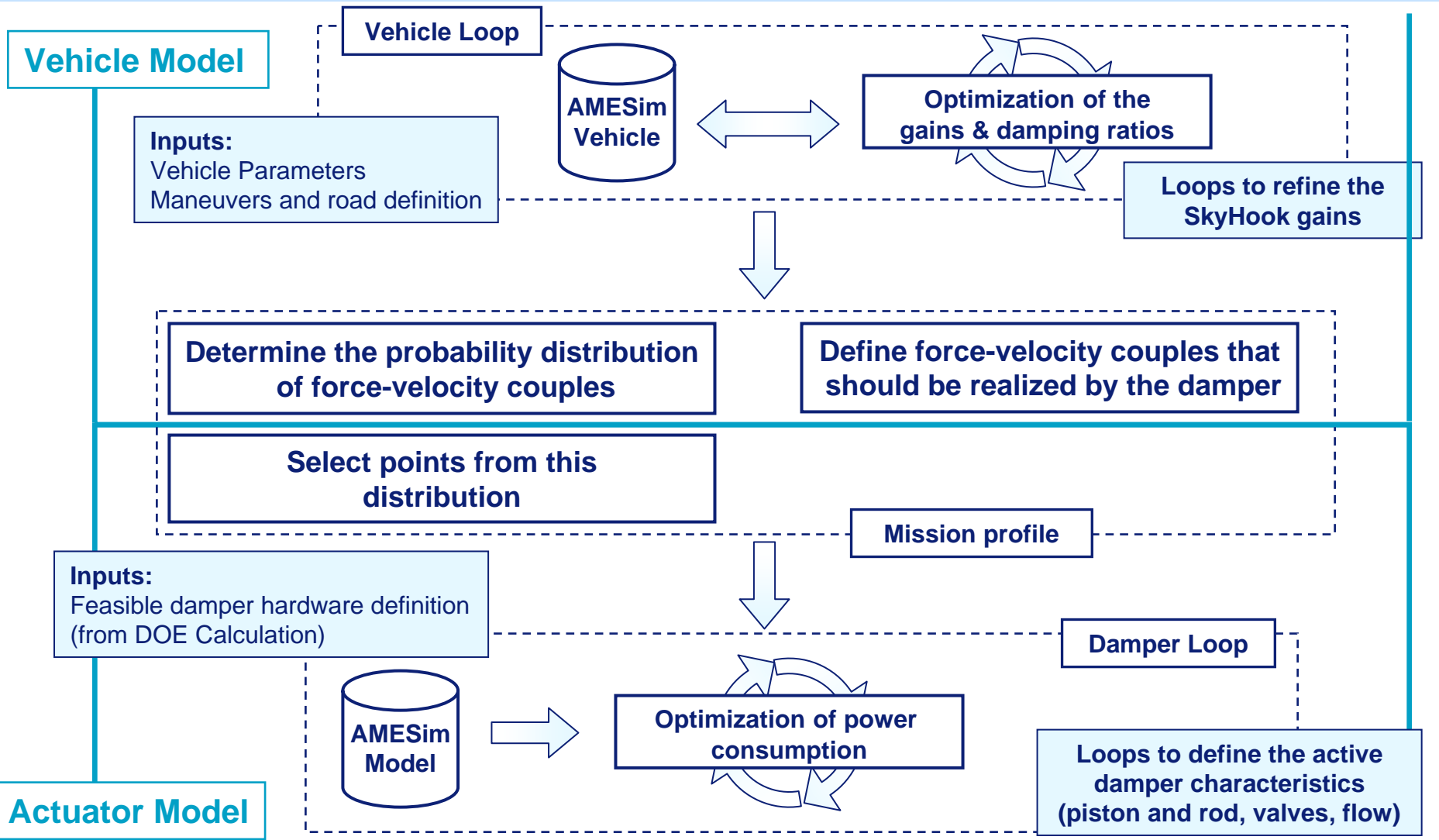
2

The optimal design approach

- Structured design scheme

Structured design scheme

1 ● ● ● 2 ○ 3 ● ● ● ● ● 4 ● ● ● ● ● 5 ●

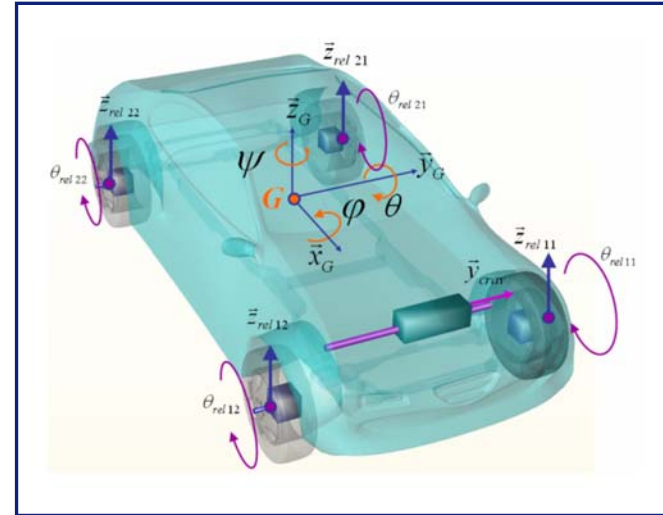
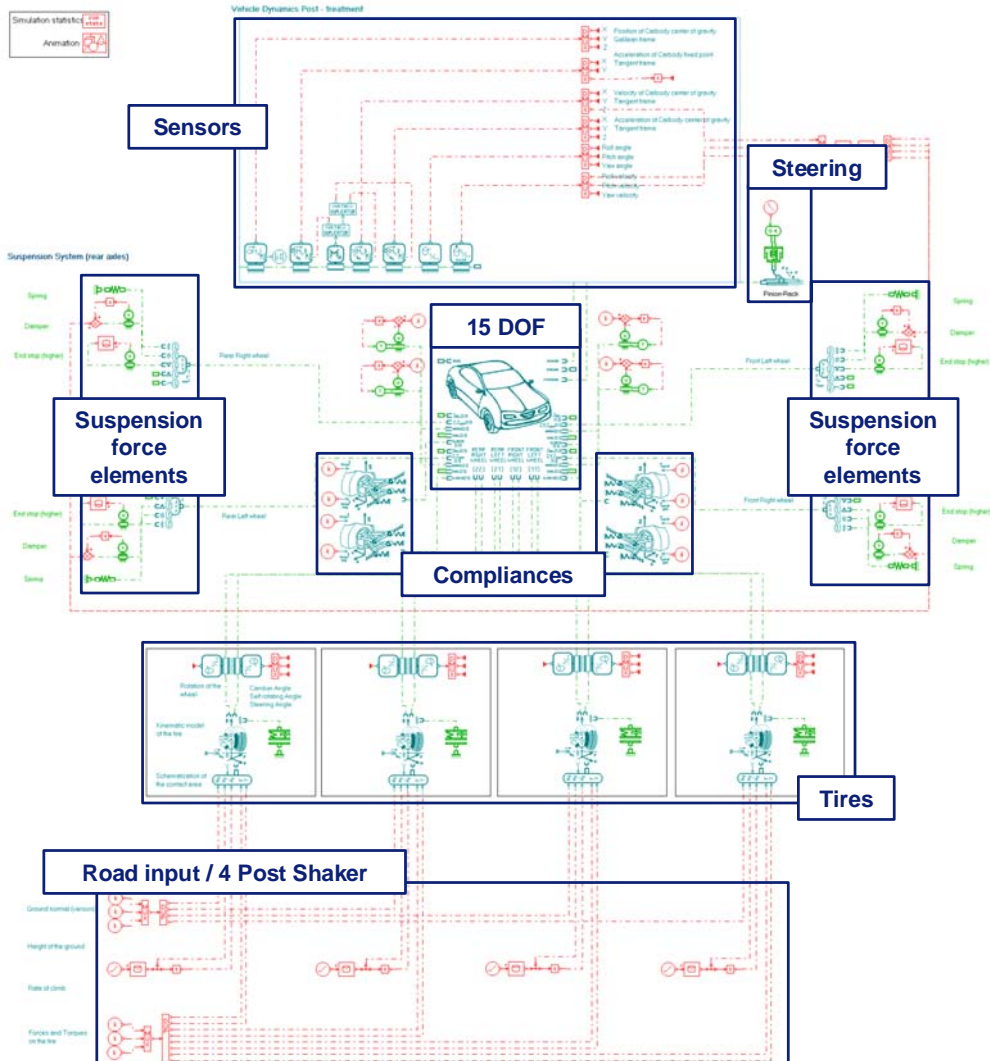


3

Vehicle Model

- Virtual model overview
- Measurement correlation – Model validity range
- The controlled vehicle – Active model set up
- Controller gains optimization

Virtual model overview



Car Body : 6 DOF	$X_G, Y_G, Z_G, \varphi, \vartheta, \psi$	} 15 DOF
Steering Rack Body : 1 DOF	y_n	
4 Spindle Body : 4 x 1 DOF	$4 \times z$	
4 Wheel Body : 4 x 1 DOF	$4 \times \omega$	

Virtual model overview



Wheel's attitude

$$\gg Trans_{Front} = \mathbf{d}_{rel\ KIN}(z, y_n, z_{opp}) = \begin{bmatrix} x(z, y_n, z_{opp}) \\ y(z, y_n, z_{opp}) \\ z \end{bmatrix}$$

$$\gg Rot_{Front} = \mathbf{r}_{rel\ KIN}(z, y_n, z_{opp}) = \begin{cases} \delta(z, y_n, z_{opp}) & \text{Toe in} \\ \varepsilon(z, y_n, z_{opp}) & \text{Camber} \\ \eta(z, y_n, z_{opp}) & \text{Self Rotating} \end{cases}$$

$$\mathbf{d}_{rel\ TOT} = \mathbf{d}_{rel\ ELAS} + \mathbf{d}_{rel\ KIN}$$

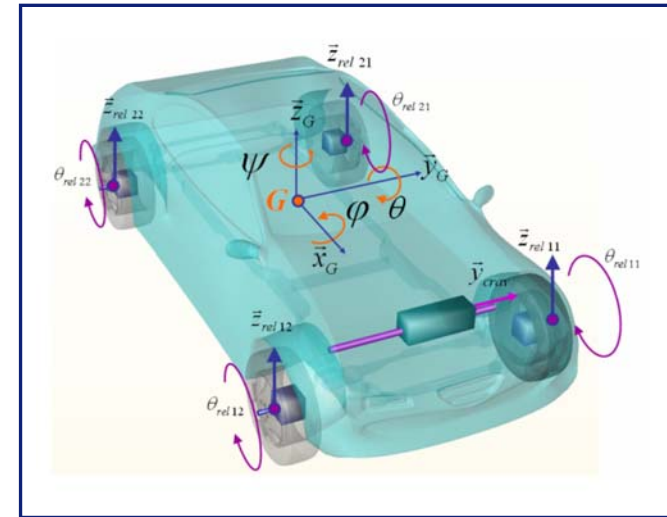
$$\mathbf{r}_{rel\ TOT} = \mathbf{r}_{rel\ ELAS} + \mathbf{r}_{rel\ KIN}$$

Compliances

$$\gg \begin{bmatrix} \Delta x_{elas} \\ \Delta y_{elas} \\ \Delta \varepsilon_{elas} \\ \Delta \eta_{elas} \\ \Delta \delta_{elas} \end{bmatrix} = \begin{bmatrix} M_{Flexibility} \\ M_{CurrentWheel} \end{bmatrix} \cdot \begin{bmatrix} F_x \\ F_y \\ M_\varepsilon \\ M_\eta \\ M_\delta \end{bmatrix} + \begin{bmatrix} M_{Coupling} \\ M_{CurrentWheel} \end{bmatrix} \cdot \begin{bmatrix} F_{xopp} \\ F_{yopp} \\ M_{\varepsilon opp} \\ M_{\eta opp} \\ M_{\delta opp} \end{bmatrix}$$

$$\begin{bmatrix} M_{Coupling} \\ M_{CurrentWheel} \end{bmatrix} = \begin{bmatrix} q_{Coupling} \\ q_{CONSTANT} \\ q_{CurrentWheel} \end{bmatrix}$$

$$\begin{bmatrix} M_{Flexibility} \\ M_{CurrentWheel} \end{bmatrix} = \begin{bmatrix} q_{Flexibility} \\ q_{CONSTANT} \\ q_{CurrentWheel} \end{bmatrix} + \begin{bmatrix} m_{Flexibility} \\ m_{REL\ to\ dz} \\ m_{CurrentWheel} \end{bmatrix} \cdot z$$



Car Body : 6 DOF	$X_G, Y_G, Z_G, \phi, \vartheta, \psi$	} 15 DOF
Steering Rack Body : 1 DOF	y_n	
4 Spindle Body : 4x1 DOF	$4 \times z$	
4 Wheel Body : 4x1 DOF	$4 \times \omega$	

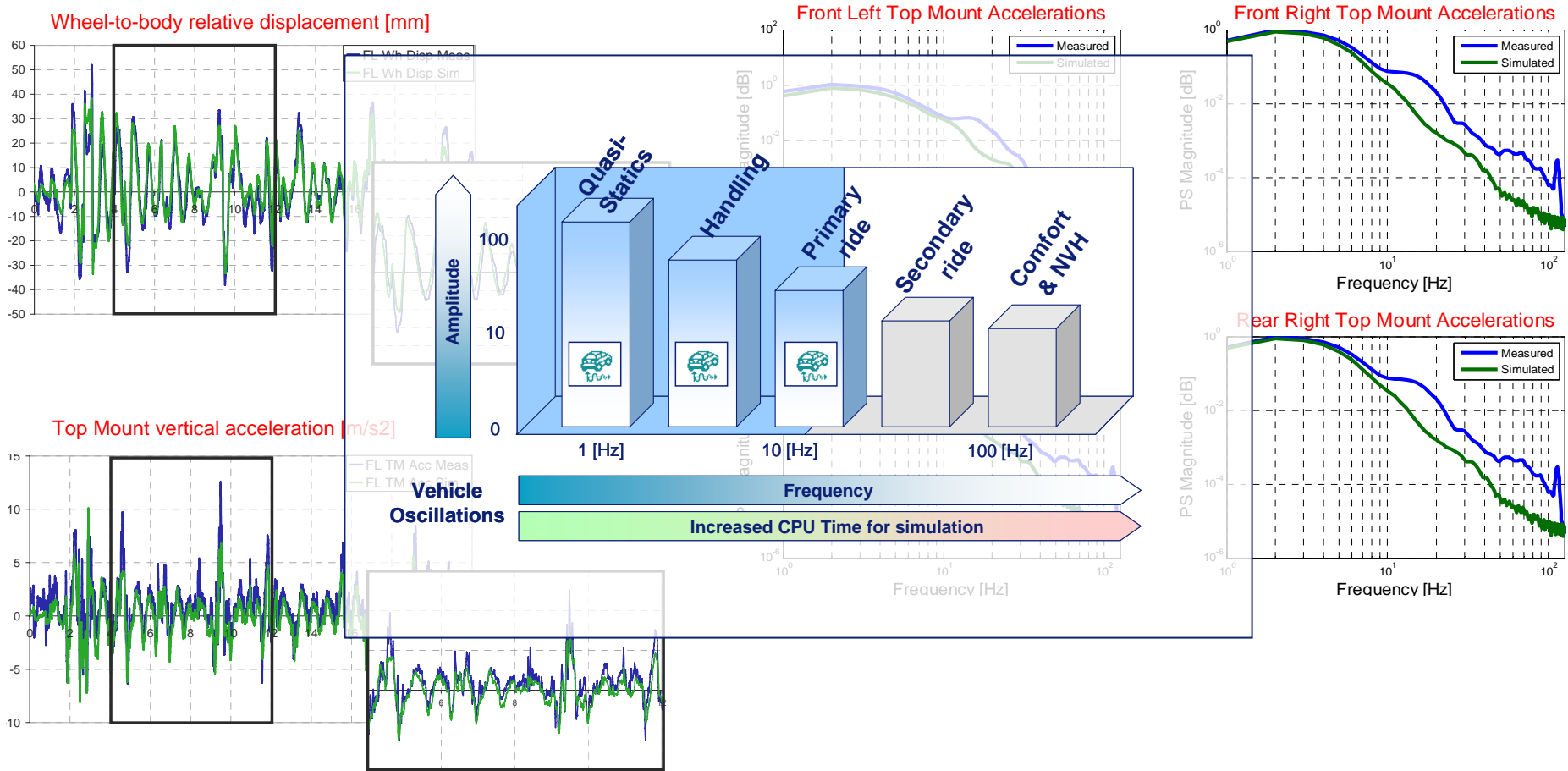
Force elements tables

$$\gg \Delta_{Spring} = f(z, y_n, z_{opp}) \longrightarrow \begin{cases} F_z = \frac{\partial f(z, y_n, z_{opp})}{\partial z} \cdot F_{Spring} \\ F_{y_n} = \frac{\partial f(z, y_n, z_{opp})}{\partial y_n} \cdot F_{Spring} \\ F_{z_{opp}} = \frac{\partial f(z, y_n, z_{opp})}{\partial z_{opp}} \cdot F_{Spring} \end{cases}$$

Measurement correlation – Model validity range

1 ● ● ● ● 2 ● 3 ● ○ ● ● ● ● 4 ● ● ● ● ● 5 ●

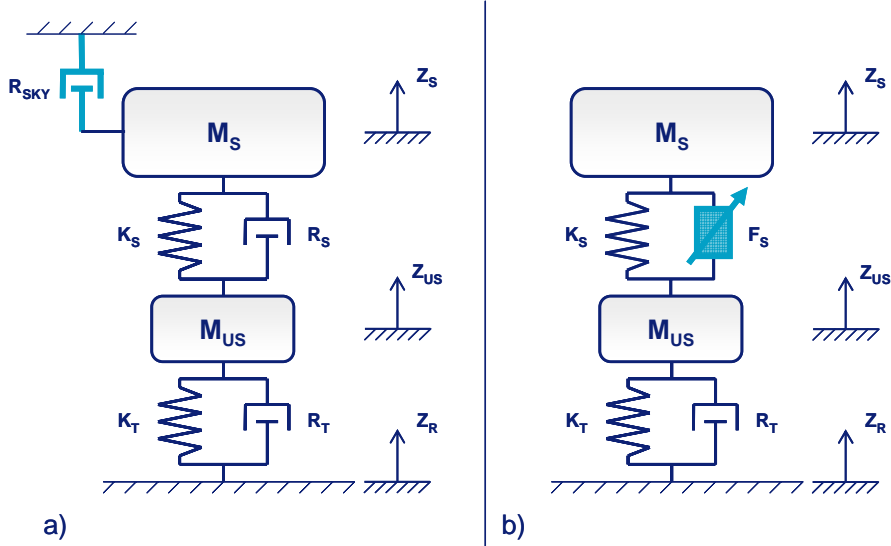
Measurements and virtual test with the vehicle running on *Road_1* at 80km/h



The controlled vehicle - Active model set up

- 1 ● 2 ● 3 ● 4 ● 5 ●

The sky-hook approach



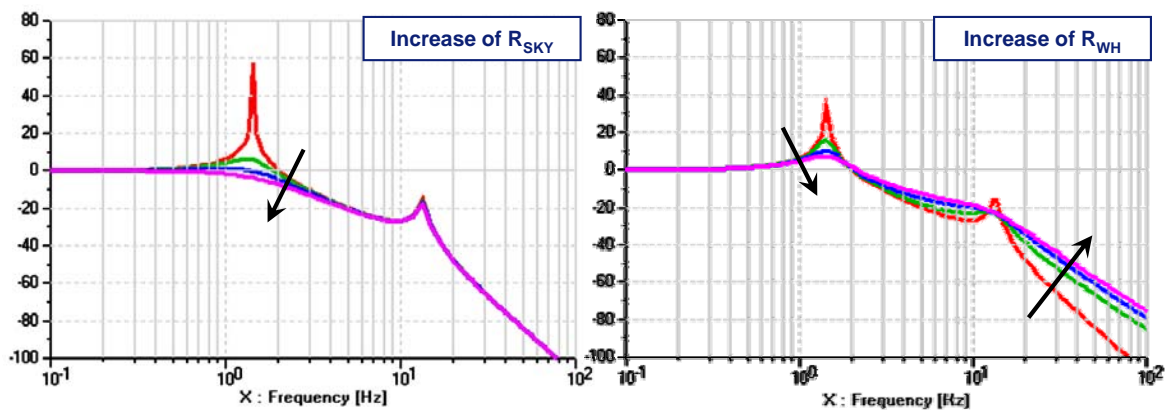
$$F_{SKY} = -\frac{dZ_S}{dt} \cdot R_{SKY}$$

$$F_S = F_{SKY} + F_{WH} = -\frac{dZ_S}{dt} \cdot R_{SKY} - (\dot{Z}_S - \dot{Z}_{US}) \cdot R_{WH}$$

Generalized on the full vehicle body modes

$$\begin{Bmatrix} F_Z \\ M_X \\ M_Y \end{Bmatrix} = \begin{bmatrix} -R_{SKY_z} & 0 & 0 \\ 0 & -R_{SKY_rx} & 0 \\ 0 & 0 & -R_{SKY_ry} \end{bmatrix} \cdot \begin{Bmatrix} \dot{Z} \\ \dot{\phi} \\ \dot{\theta} \end{Bmatrix}$$

Magnitude of HEAVE Displacement / HEAVE Road Input [dB]



Controller gains optimization - 1



Full vehicle NON LINEAR model

Complete Ride Maneuver:

Soft Spring Set RIDE Road_1
 Normal Spring Set RIDE Road_2
 RIDE Road_3

Tuning of the Wheel-Hop dampings in function of the performances requirements



Full vehicle LINEARIZED model

Static settling & linearization:

Soft Spring Set
 Normal Spring Set

Tuning of the Sky-hook gains with respect to the modal damping ratios

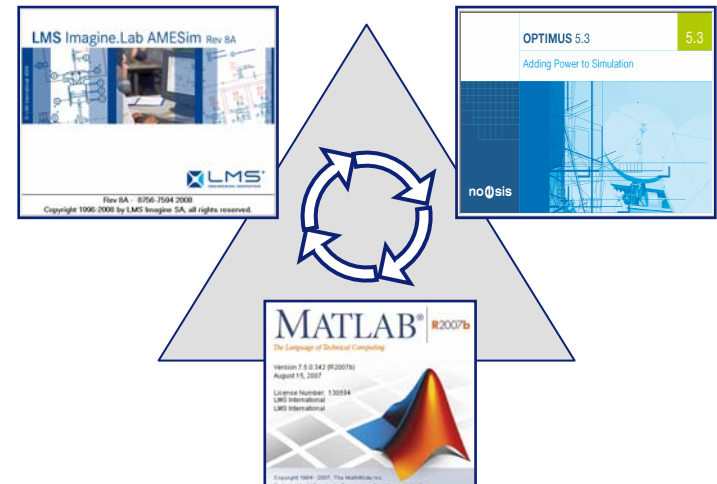
① Minimize

$$\begin{cases} abs(Max\ and\ Min\ \theta) \\ abs(Max\ and\ Min\ \varphi) \\ abs(Max\ and\ Min\ \dot{Z}_{CG}) \\ abs(Max\ and\ Min\ \ddot{Z}_{CG}) \end{cases}$$

② Modal Z, θ , φ damping $\xi = \frac{1}{\sqrt{2}}$

The optimizations ran with the *Differential Evolution Algorithm* (Global Optimization Scheme) in a nested loop

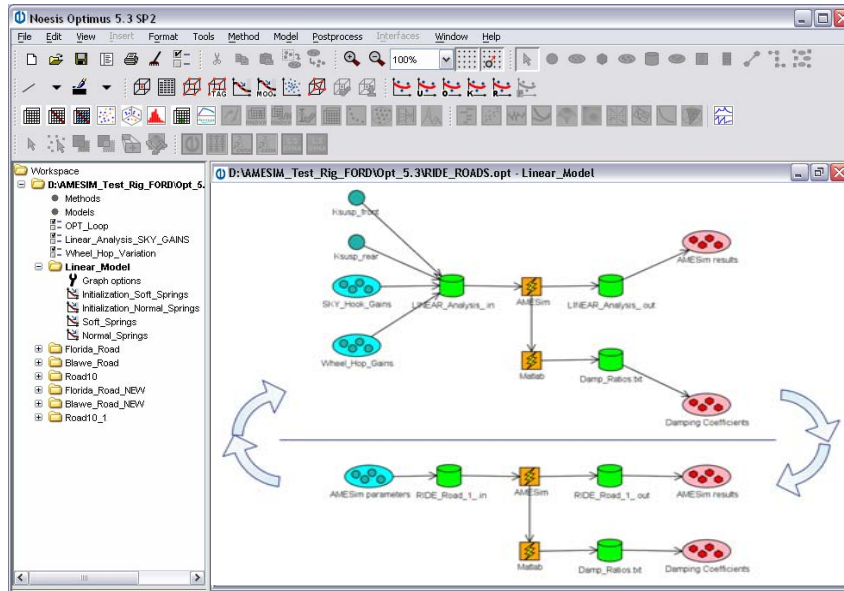
Interfacing Optimus – AMESim – Matlab



Controller gains optimization - 2

1 ● ● ● ● 2 ● 3 ● ● ● ● ● ○ 4 ● ● ● ● ● ● 5 ●

The set up of the active vehicle has been done with two different sets of springs (“**Normal**” and “**Soft**”), which leads to a small influence on the optimization results of this selection, because of the compensation offered by the active elements.



Soft Springs Set up			
Ride Roads			
	1	2	3
R WH_Front	1155	1541	1348
R WH_Rear	1899	542	1221
R SKY_z	16025	18046	27290
R SKY_rx	1402	1402	1251
R SKY_ry	788	681	849
MAX_HeaveSpeed	0.372	0.136	0.041
MIN_HeaveSpeed	0.389	0.157	0.039
MAX_Roll	1.322	0.932	0.273
MIN_Roll	1.088	1.068	0.288
MAX_Pitch	0.556	0.537	0.080
MIN_Pitch	0.842	0.614	0.126
DampCoeff 1.1	0.499	0.751	0.964
DampCoeff 1.2	0.499	0.751	0.964
DampCoeff 2.1	0.733	0.446	0.631
DampCoeff 2.2	0.733	0.446	0.631
DampCoeff 3.1	0.692	0.746	0.673
DampCoeff 3.2	0.692	0.746	0.673

From the three ride maneuvers the **Mission Profiles** have been extracted.

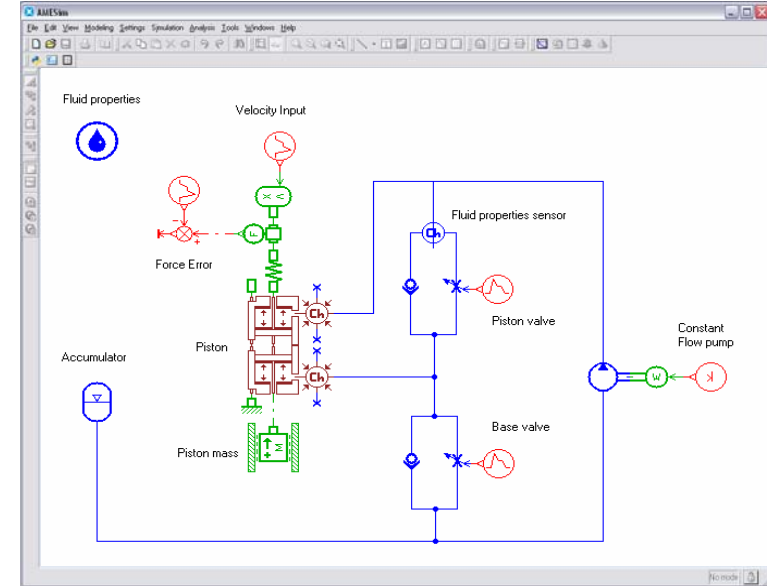
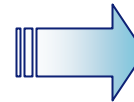
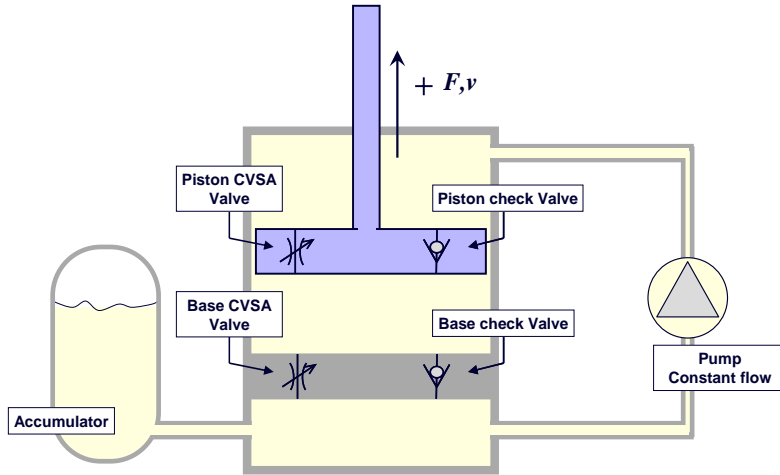
4

Actuator Model

- Actuator representation and functional model
- DOE for performance achievement
- Mission Profiles & Iso-Power curves
- Optimization for energy consumption minimization
- The optima achievement

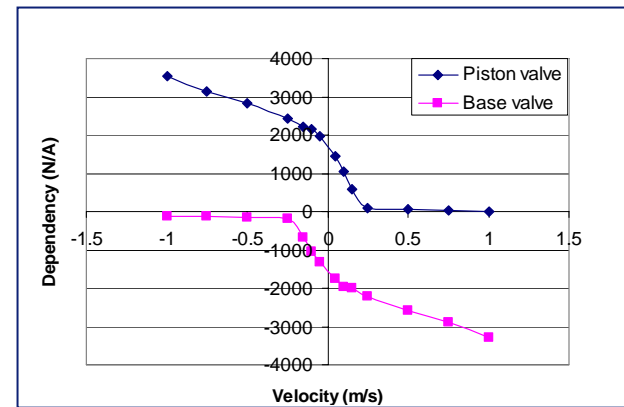
Actuator representation and functional model

1 ● ● ● ● 2 ● 3 ● ● ● ● ● 4 ○ ● ● ● ● ● 5 ●



$$F = \begin{cases} \frac{\rho \cdot S_1 \cdot (Q - S_2 \cdot \dot{x})^2}{2 \cdot C_q^2 \cdot S_{pv}^2} - p_3 \cdot (S_2 - S_1) & \text{if } \dot{x} > 0 \\ - \left[\frac{\rho \cdot (Q + (S_2 - S_1) \cdot \dot{x})^2}{2 \cdot C_q^2 \cdot S_{bv}^2} + p_3 \right] \cdot (S_2 - S_1) & \text{if } \dot{x} < 0 \end{cases}$$

- ρ : fluid density
- Q : pump flow
- p_3 : accumulator pressure
- S_1 : surface on the rod side
- S_2 : surface on the piston side
- S_{pv} : orifice of the piston valve
- S_{bv} : orifice of the base valve
- C_q : flow coefficient
- \dot{x} : relative velocity of the piston w.r.t. the cylinder



DOE for performance achievement

1 ●●● 2 ● 3 ●●●●● 4 ●○●●● 5 ●

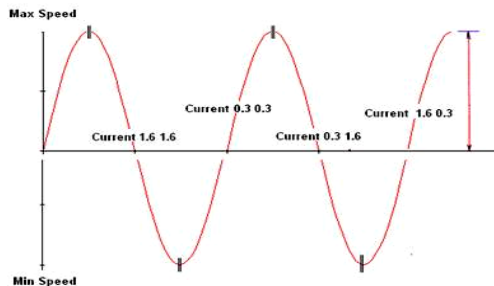
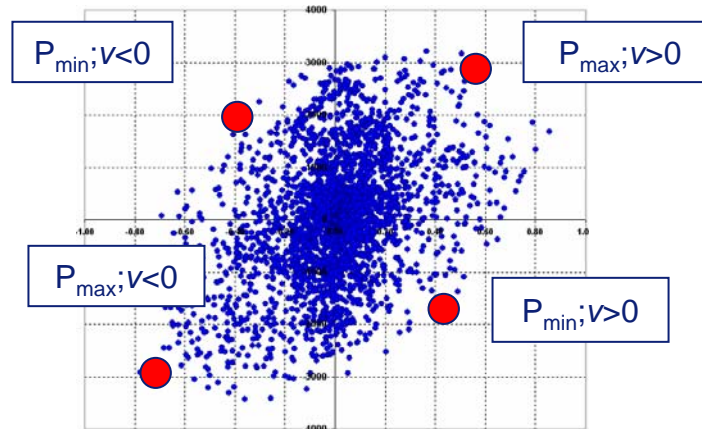
Define the feasible combinations to achieve the performance requirements defined by the mission profiles in terms of deliverable power achievement

This DOE simulates all combinations of damper, valves and flow (discretized in 3 flows: 1, 5 and 10 L/min.) for a total of 1296

A **combination** is composed of:

- Rod (4)
- Piston (3)
- Piston valve (6)
- Base valve (6)
- Flow (3)

Selection of 4 critical points (P, v)

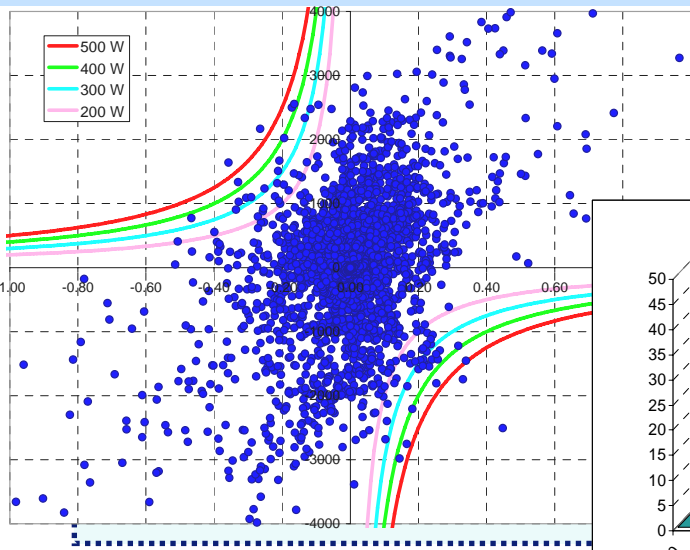


Index	Rod diameter [mm]	Cylinder diameter [mm]
1	12,4	25,4
2	15,8	
3	12,4	27
4	15,8	30
5	18	
6	20	
7	18	32
8	20	
9	22	35
10	20	
11	25	45
12	25	

Those *feasible combinations* will be considered in the optimization loop

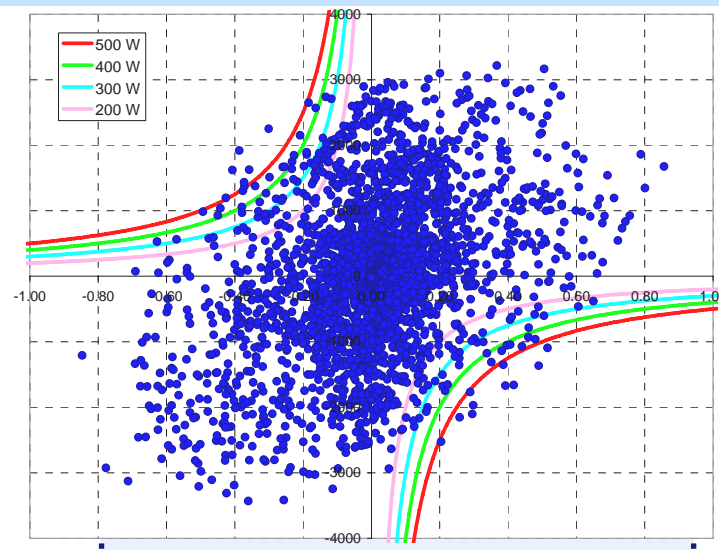
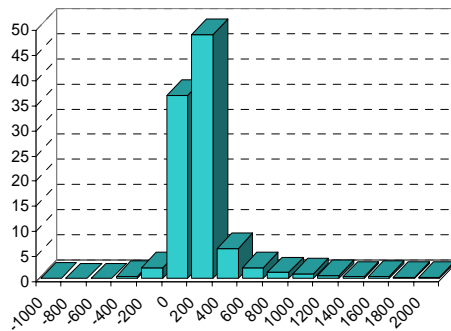
Mission Profiles & Iso-Power curves

1 ● ● ● 2 ● 3 ● ● ● ● ● 4 ● ● ● ○ ● ● 5 ●

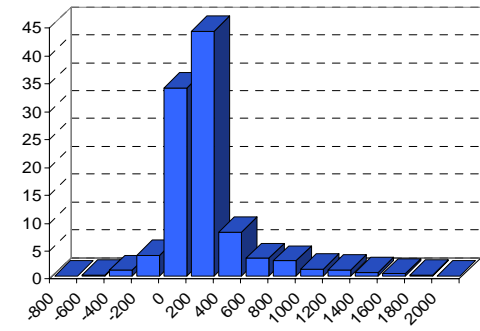


RIDE Road 1 - (5%)

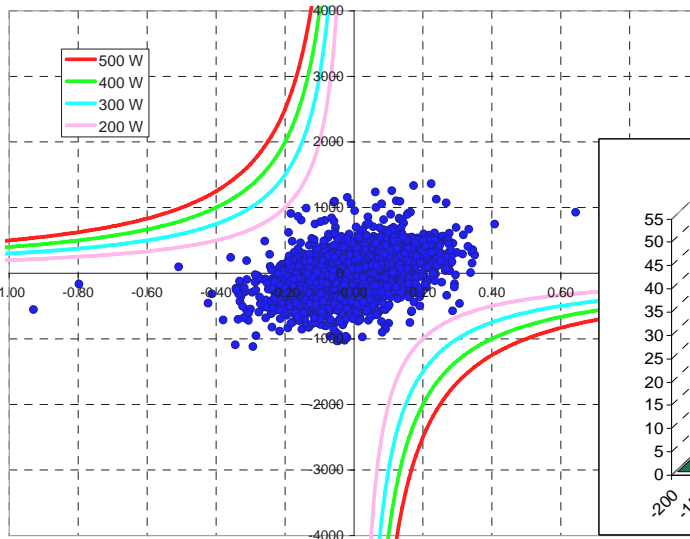
Occurrences of Power Values (in%)



Occurrences of Power Values (in%)

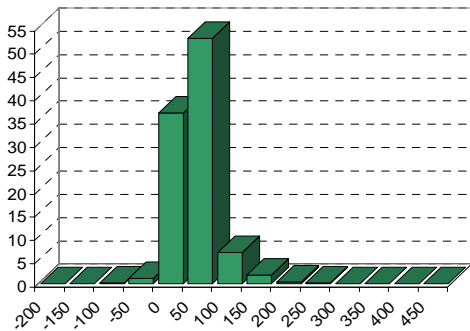


RIDE Road 2 - (5%)



RIDE Road 3 - (90%)

Occurrences of Power Values (in%)



Optimization for energy consumption minimization

1 ● ● ● 2 ● 3 ● ● ● ● ● 4 ● ● ● ○ ● 5 ●

The objective function takes into account the mean error and the mean power consumption.

- **Cost function:**

- Power Consumption & Mean Error with weighting factor

$$F(p) = w \cdot \int |error_{Force\ Observed}| + w \cdot \int |torque_{Pump} \cdot speed_{Pump}|$$

- **Optimization parameters:**

- Rod and piston diameters
- Valves characteristics
- Pump flow
- Input currents (at discretized time steps)

- **Criterion:**

- Selection of random points in the required performance subset

5% from RIDE Road 1

5% from RIDE Road 2

90% from RIDE Road 2

Optimization run for selected points using Genetic Algorithm

The optima achievement



Front Damper

- ✓ Damper: 35-20 (piston diameter 35mm and rod diameter 20mm)
- ✓ Valves: High Pressure with minimal characteristic
- ✓ Flow rate: 1L/min

Cylinder Ø [mm]	Rod Ø [mm]
30	20
32	20
	22
35	20

These combinations all use high pressure (HP) valves (min or max)

	Mean error (N)	Mean consumption (W)
Optimum (weight 5)	488	23
Optimum (weight 10)	314	53

Rear Damper

- ✓ Damper 45-25
- ✓ Piston valve: High Pressure maximum
- ✓ Base valve: Normal minimum
- ✓ Flow rate: 1L/min

Cylinder Ø [mm]	Rod Ø [mm]	Piston valve	Base valve
32	22	<i>N min</i>	<i>HP min</i>
		<i>N max</i>	
35	25	<i>HP min</i>	<i>N min</i>
		<i>HP max</i>	
45	25	<i>HP min</i>	
		<i>HP max</i>	

	Mean error (N)	Mean consumption (W)
Optimum (weight 10)	368 N	57 W

5

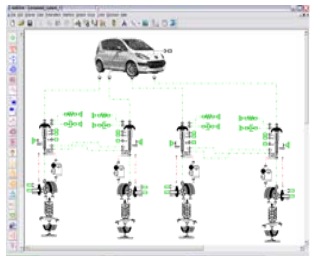
Conclusions & Outlooks

- The Intelligent Product development workflow

The Intelligent Product development workflow

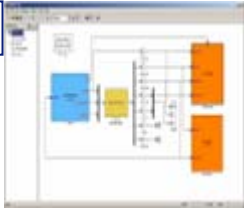
1 ●●● 2 ● 3 ●●●●● 4 ●●●●● 5 ○

“Physical” models
connected to design data

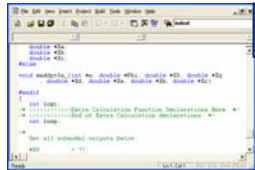
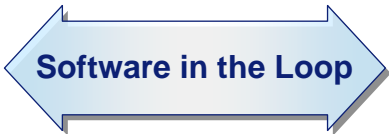


Control development

Control Logic Modeling



Real Time Requirement



Control Code Libraries



Controller
Testing & Calibration
In *Virtual*
Product Environment



Emulated ECU
“Rapid Prototyping”

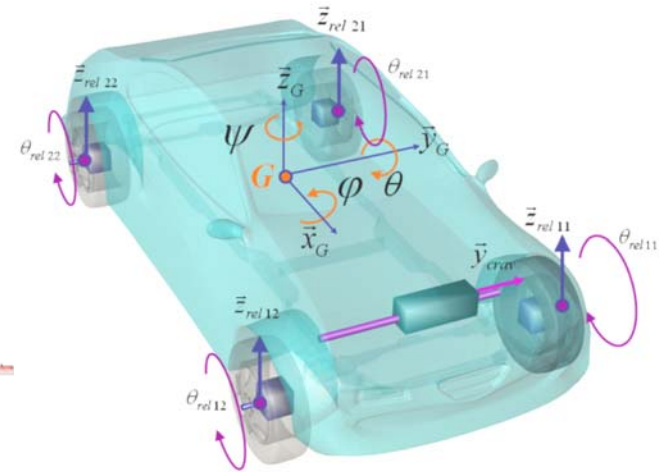
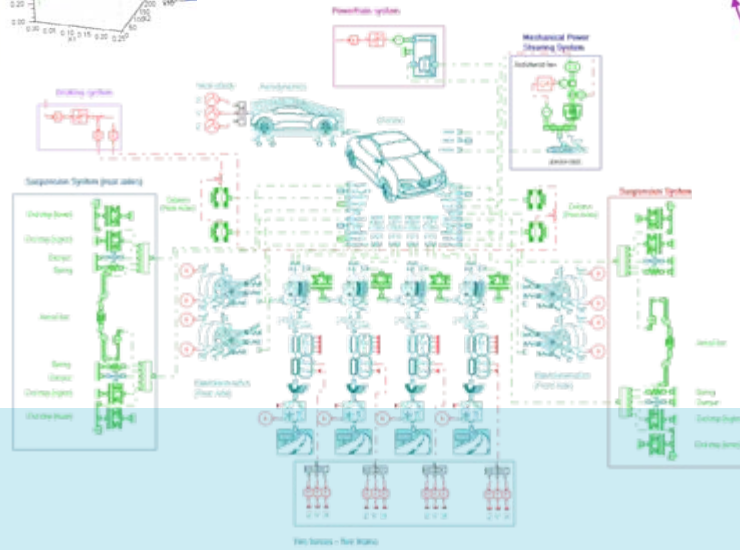
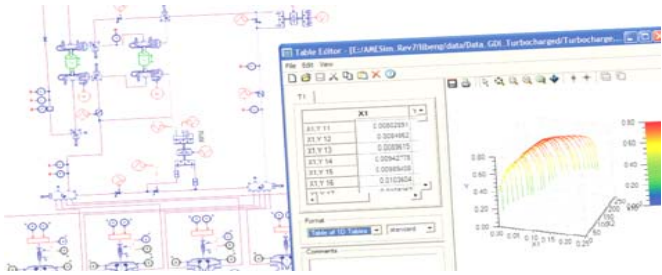


Controller
Testing & Calibration
in *Physical*
Vehicle Environment



Vehicle Integration





Thank you!
Time for Q & A...

