



Machine Learning im CAE

Saxsim 2023

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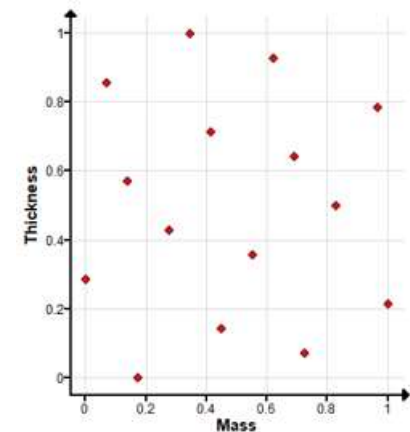
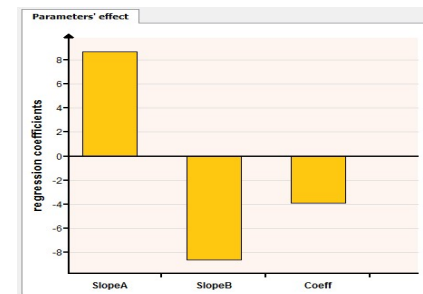


ODYSSEE – Overview

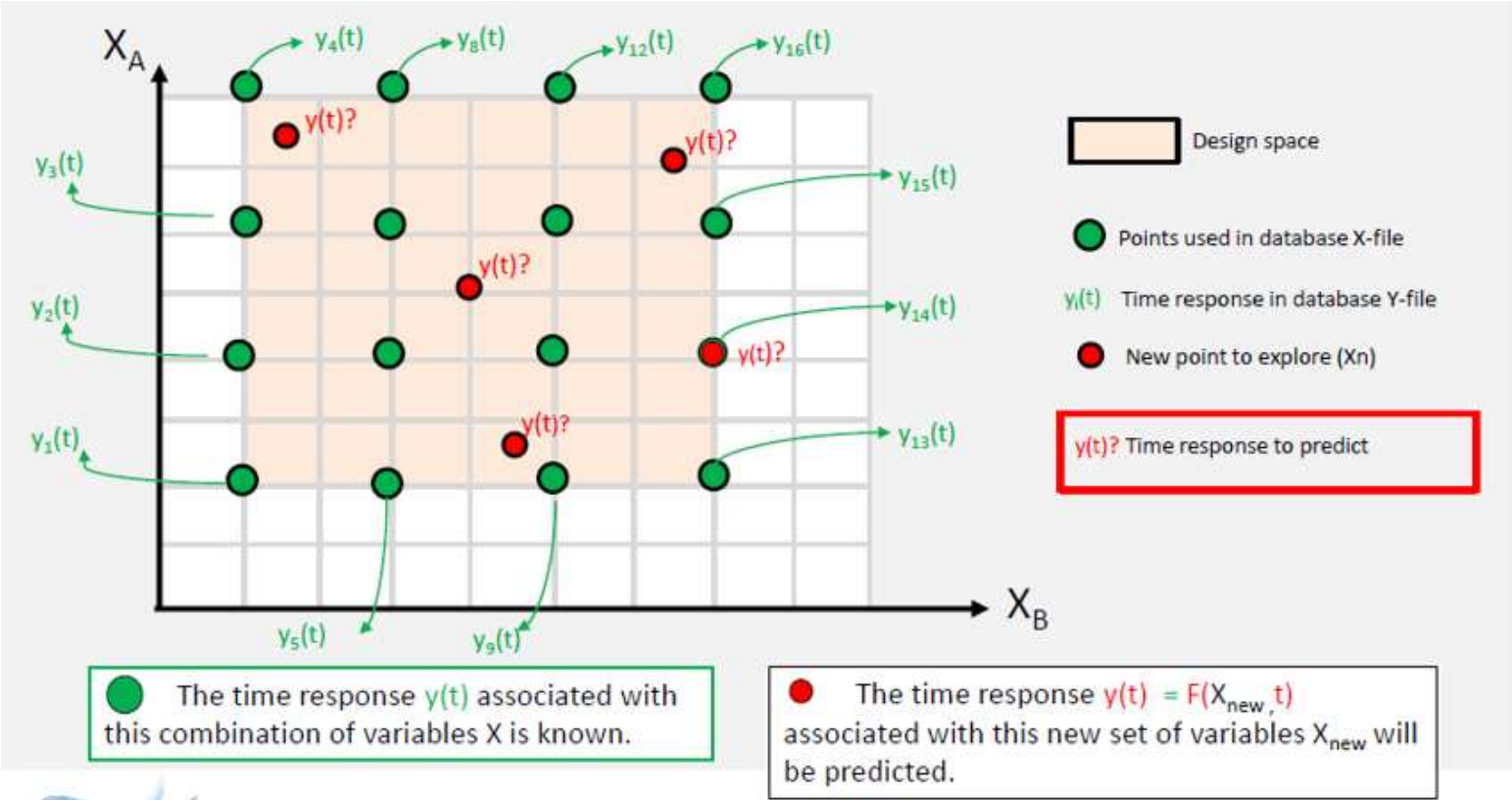
File Export Tools Preferences Help



- From available analysis, test results or measurements, predict responses for further data points
- ODYSSEE delivers the response in seconds
- Predict results values, curves and animations
- Improve parameters of manufacturing processes
- Input: csv files
- Can use images instead of parameters
- Easy-to-use GUI
- Scripting possible
- Create and improve DOE
- Plot tool for correlation, PCA, heat map, etc.
- Optimization, e.g. adapt analysis model to test result curve



Task Formulation

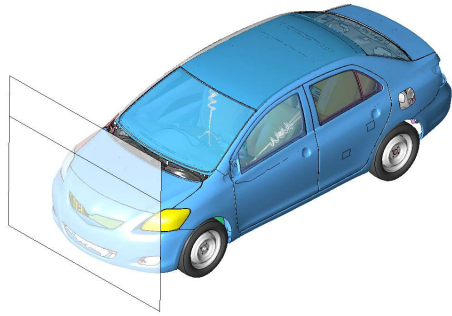


Input: Learning base
 $F(x) = Y$
 (Existing analysis or test data)

Output: $F(x_{new}) = ?$

Y can be a curve $Y(t)$

Example – ‘Yaris Front crash’



Source Model :
<https://www.ccsa.gmu.edu/models/2010-toyota-yaris/>

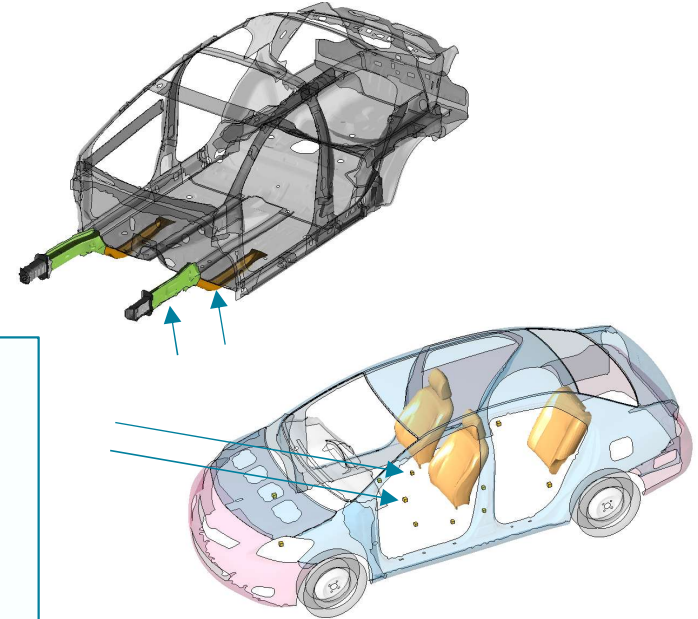
DOE of 17 runs

3 Model parameters:

Rail inner thickness => X1
Floor support thickness => X2
Velocity => X3

Output channels:

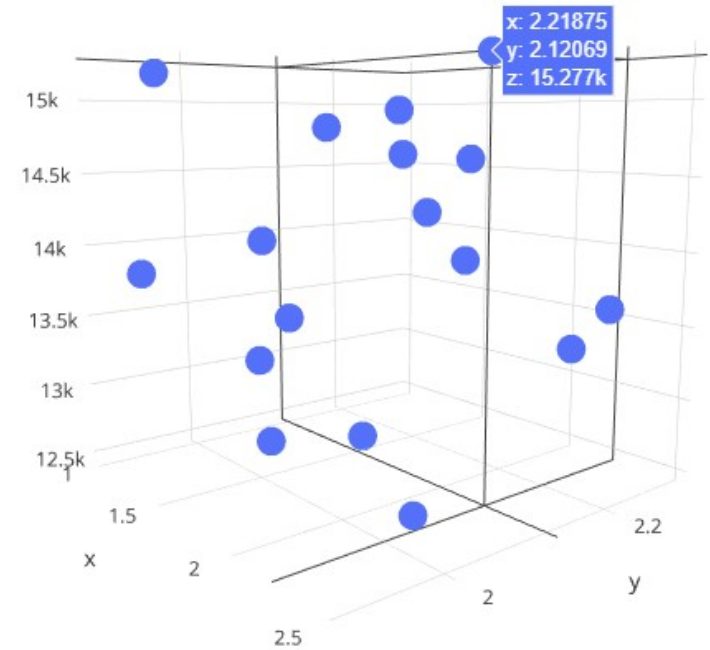
Driver seat acceleration => Y1
Rigid wall Force => Y2
Gravity center acceleration => Y3



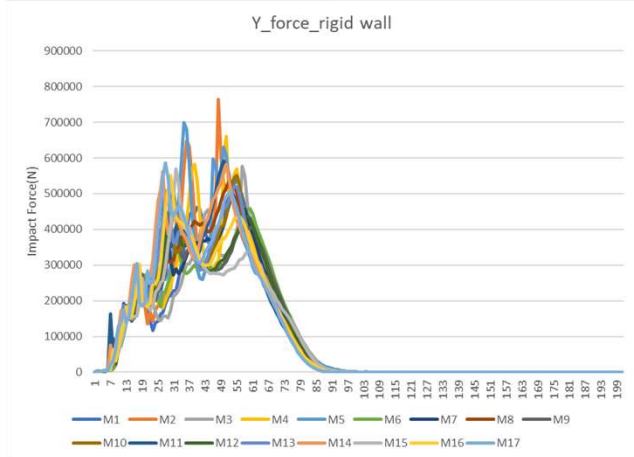
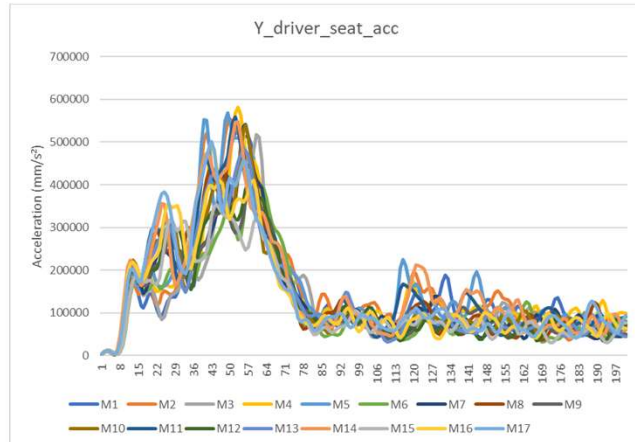
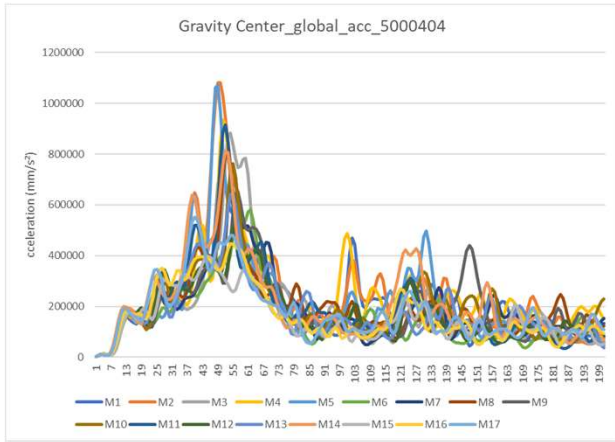
| Finite Elements | Lunar |
|-------------------|-----------|
| 3h per simulation | 2 seconds |

Example – ‘Yaris Front crash’

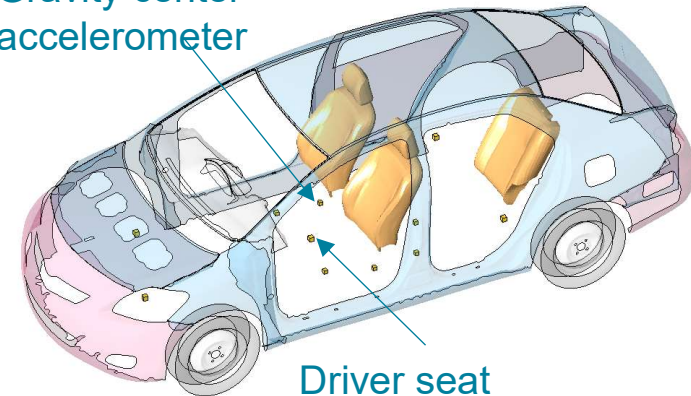
| | Rail inner thickness (mm) | Floor support thickness (mm) | Velocity (mm/s) |
|-----|---------------------------|------------------------------|-----------------|
| M1 | 1,00 | 1,94 | 13772,80 |
| M2 | 1,09 | 2,16 | 14814,20 |
| M3 | 1,19 | 2,05 | 13078,50 |
| M4 | 1,28 | 2,27 | 14119,90 |
| M5 | 1,38 | 1,91 | 15161,30 |
| M6 | 1,47 | 2,13 | 12500,00 |
| M7 | 1,56 | 2,02 | 13541,40 |
| M8 | 1,66 | 2,24 | 14582,80 |
| M9 | 1,75 | 1,97 | 12847,10 |
| M10 | 1,84 | 2,19 | 13888,50 |
| M11 | 1,94 | 2,08 | 14929,90 |
| M12 | 2,03 | 2,30 | 13194,20 |
| M13 | 2,13 | 1,90 | 14235,60 |
| M14 | 2,22 | 2,12 | 15277,00 |
| M15 | 2,31 | 2,01 | 12615,70 |
| M16 | 2,41 | 2,23 | 13657,10 |
| M17 | 2,50 | 1,96 | 14698,50 |



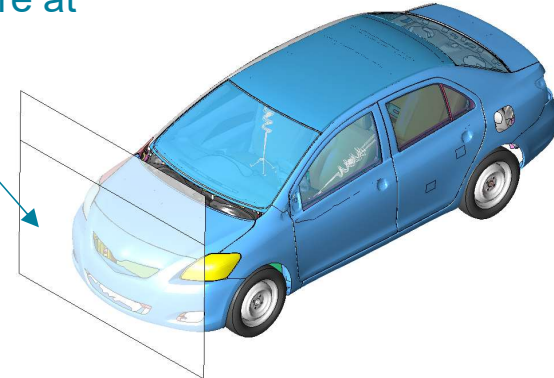
Example – ‘Yaris Front crash’



Gravity center accelerometer

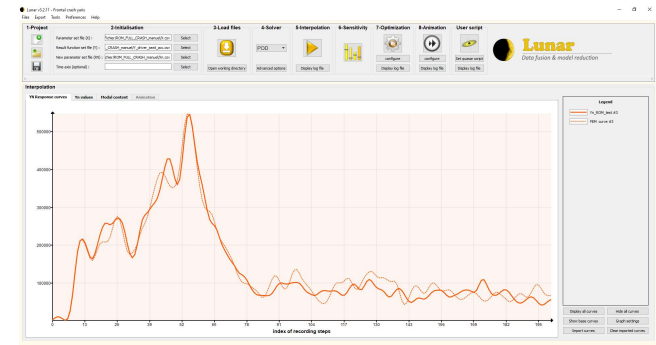
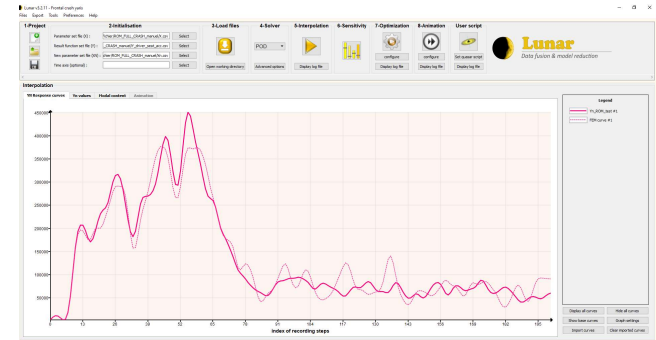
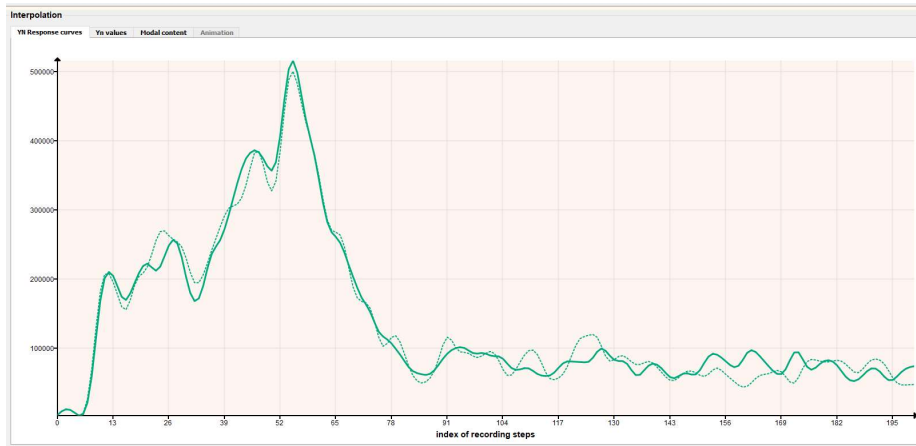


Force measure at the rigid wall



Example – ‘Yaris Front crash’

Driver seat acceleration POD KRG quadratic h1

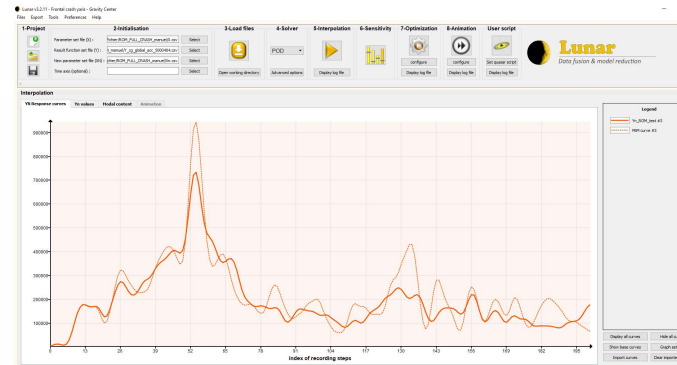
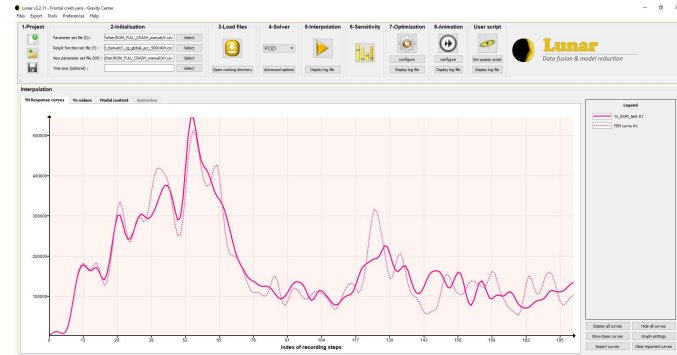
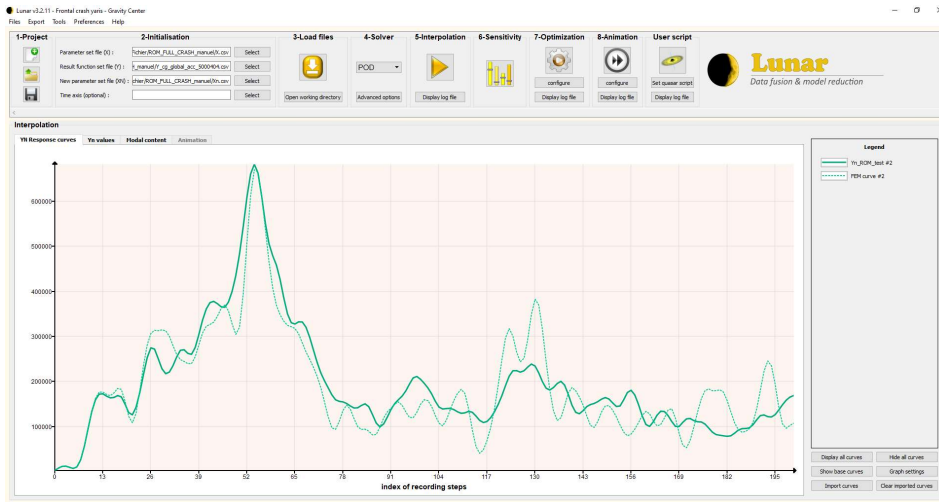


— Lunar: 1 s - - - FEM : 3h

Example – ‘Yaris Front crash’

Gravity centre acceleration

POD ARBF

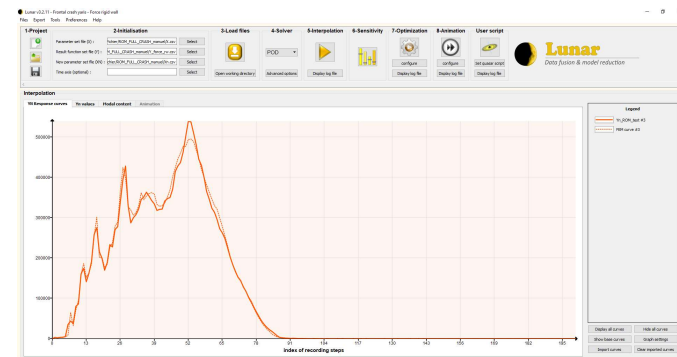
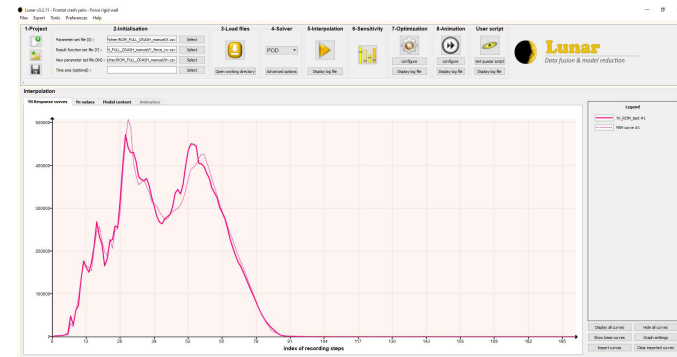
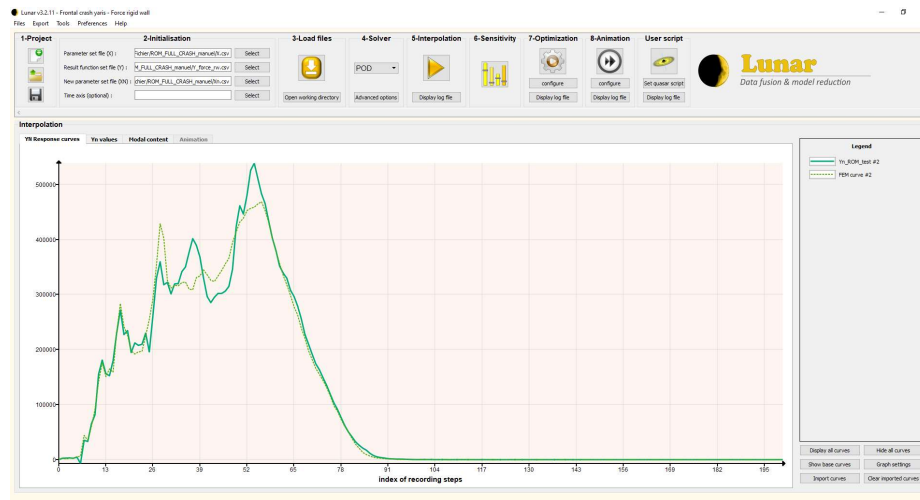


— Lunar: 1 s - - - FEM : 3h

Example – ‘Yaris Front crash’

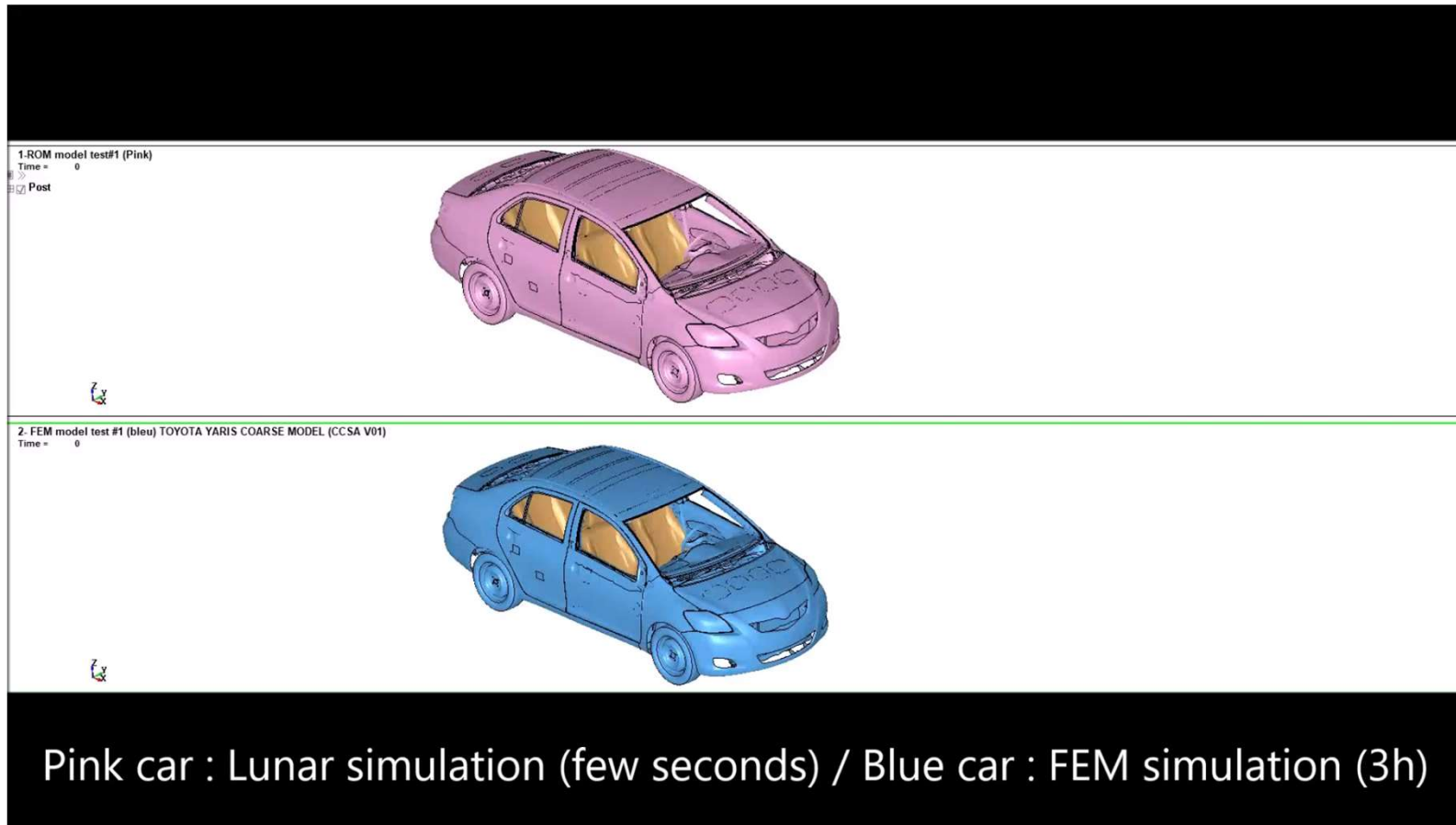
Rigid Wall force

POD ARBF



— Lunar: 1 s
- - - FEM : 3h

Example – ‘Yaris Front crash’



Source Model: <https://www.ccsa.gmu.edu/models/2010-toyota-yaris/>

Romax-Odyssee Example

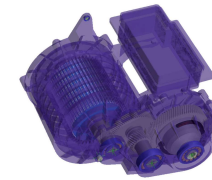
Prediction of contact stress on tooth gear of input Gear set of electric vehicle powertrain GBTE

Parameters (surface modifications) :

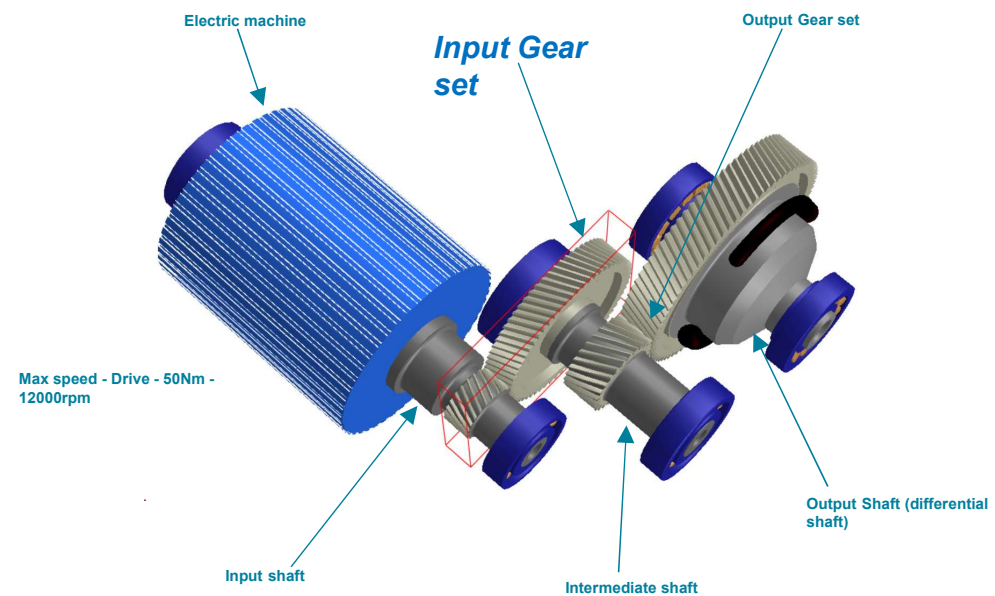
- Lead slope modification (μm)
- Involute slope modification (μm)

Outputs:

- Contact Stress max (Mpa)
- Khbeta (design parameter)
- Repartition of contact stress over tooth surface



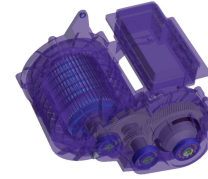
[Electrical vehicle powertrain](#)



Requires 2000 runs and about 10h-15h of computation time !

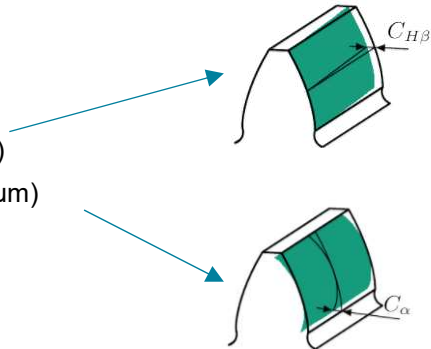
Romax-Odysee Example

Prediction of contact stress on tooth gear of input Gear set of electric vehicle powertrain GBTE



- INPUTS:

- Lead slope modification (μm)
- Involute slope modification (μm)



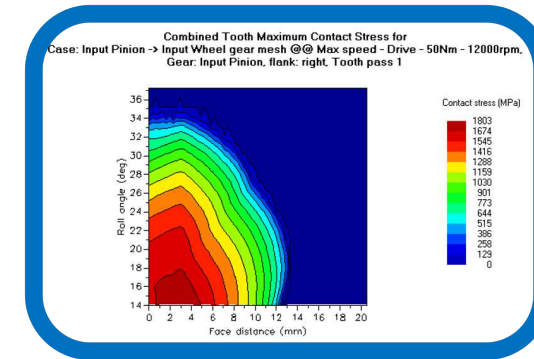
| 1 st Harmonic GBTE | Contact Stress GBTE | Khbeta GBTE (I) |
|-------------------------------|---------------------|-----------------|
| 2.15 | 1606.19 | 3.80 |
| 3.01 | 1729.55 | 4.12 |
| 2.61 | 1980.78 | 5.61 |
| 2.03 | 2124.66 | 3.56 |
| 1.42 | 1712.18 | 4.20 |
| 2.41 | 1788.34 | 3.43 |
| 0.98 | 1127.00 | 1.65 |
| 6.64 | 2732.24 | 3.89 |
| 1.46 | 2370.46 | 4.56 |
| 3.53 | 2457.38 | 4.44 |
| 3.96 | 1934.18 | 5.01 |

- OUTPUTS:

- Contact Stress max (Mpa) (**scalar**)
- 1st Harmonic (**scalar**)
- Khbeta (**scalar**)
- Repartition of contact stress over tooth surface (Matrix)
 - Is a **matrix 24*64**

- Validation :

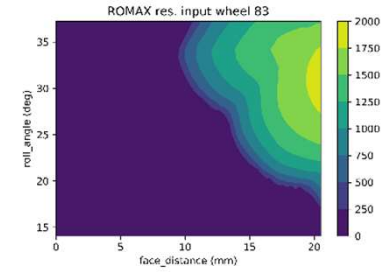
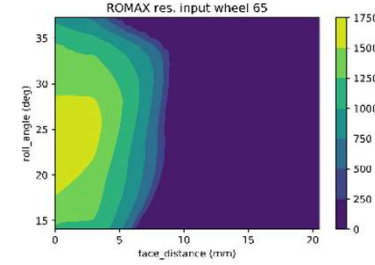
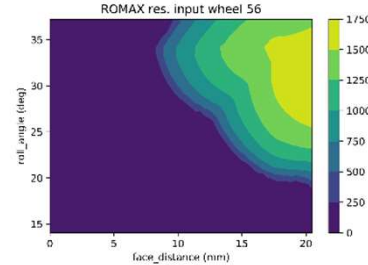
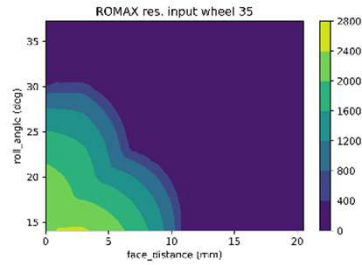
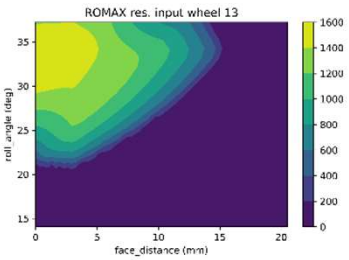
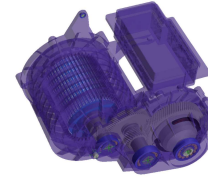
- 5 cases out of the 200 of the design space



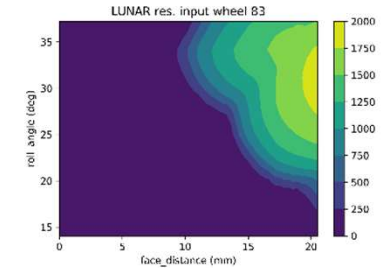
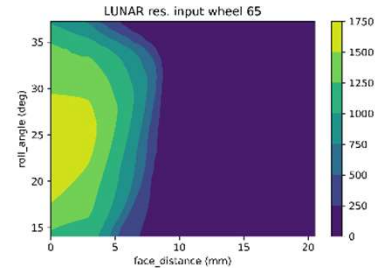
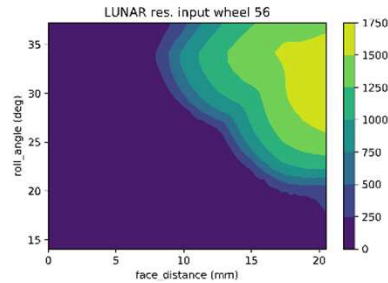
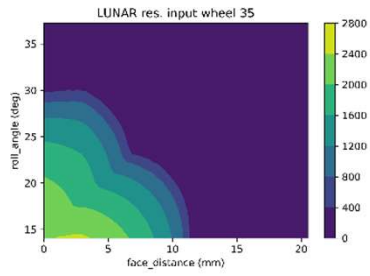
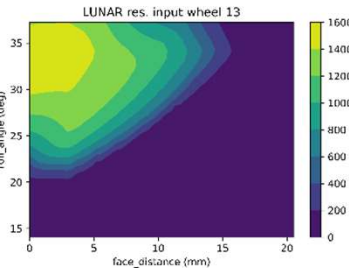
200 runs and about 1h of computation!

Romax-Odyssee Example

Validation: Comparison of stress distribution results between Romax and Odyssee results



Romax results



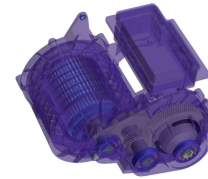
Odyssee results



VERY GOOD RESULTS!  **HEXAGON**

Romax-Odyssee Example

Comparison between Romax and Odyssee



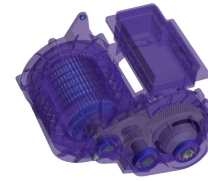
| 1st Harmonic | | | |
|--------------------|--------------|------------------|--------|
| | ROMAX result | LUNAR prediction | Error |
| cas 13 | 1.8905 | 1.8909 | 0.02% |
| cas 35 | 4 | 3.9956 | -0.11% |
| cas 56 | 2.0804 | 2.1348 | 2.61% |
| cas 65 | 2.2009 | 2.1655 | -1.61% |
| cas 83 | 3.014 | 3.011 | -0.10% |
| Max contact stress | | | |
| | ROMAX result | LUNAR prediction | Error |
| cas 13 | 1587.71 | 1585.84 | -0.12% |
| cas 35 | 2458.11 | 2458.42 | 0.01% |
| cas 56 | 1723.15 | 1724.84 | 0.10% |
| cas 65 | 1676.21 | 1675.7 | -0.03% |
| cas 83 | 1837.91 | 1838.7 | 0.04% |
| Khheta | | | |
| | ROMAX result | LUNAR prediction | Error |
| cas 13 | 3.4819 | 3.477 | -0.14% |
| cas 35 | 4.682 | 4.672 | -0.21% |
| cas 56 | 4.2957 | 4.3091 | 0.31% |
| cas 65 | 3.6167 | 3.6436 | 0.74% |
| cas 83 | 4.6959 | 4.6994 | 0.07% |



VERY GOOD RESULTS !

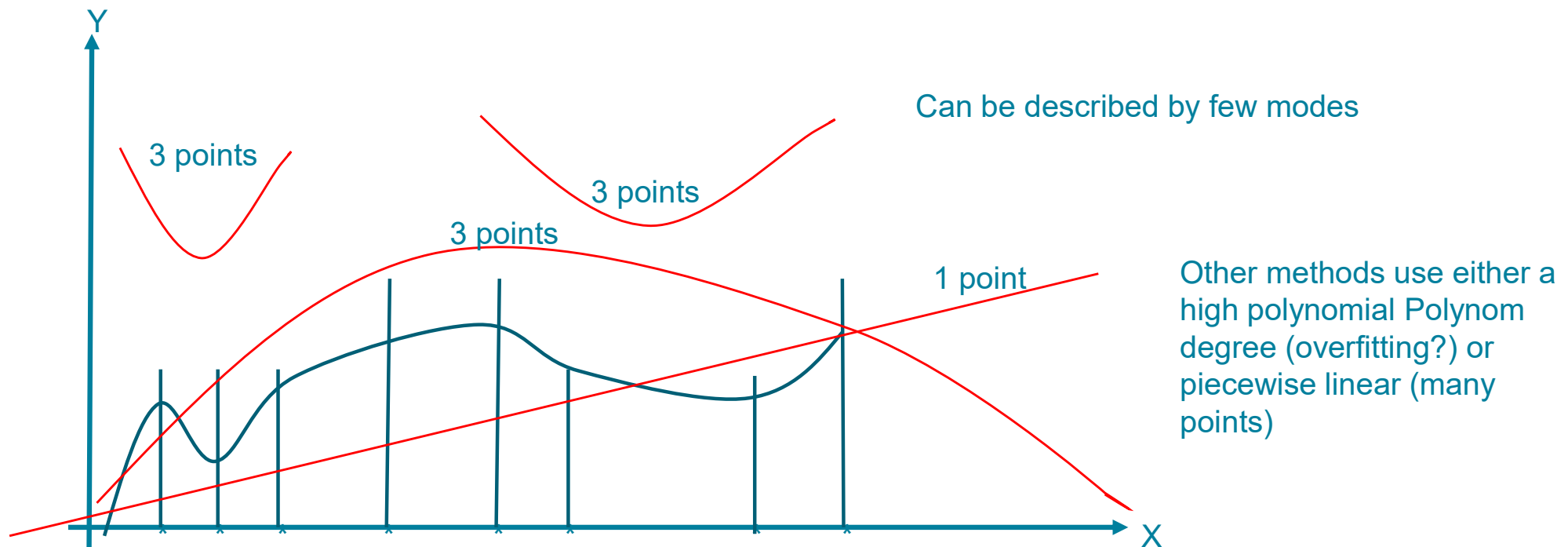
Romax-Odysee Example

Conclusion:



- Lunar gives very good prediction of ROMAX results with a maximum error of 2% in this case with less runs and less calculation time (10h vs 1h).
- With the help of Lunar it's now possible to reduce the design time and calculation without losing the precision of the calculus
- The process is very simple and can be in ROMAX to build a databases of gearbox configuration. Because with the time available for the results of one configuration it's possible to explore 10-15 configurations.

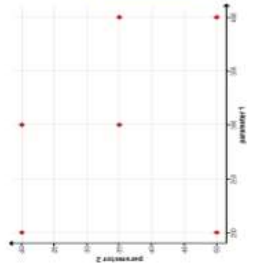
Why does the Modal Method Require Less Data Points?



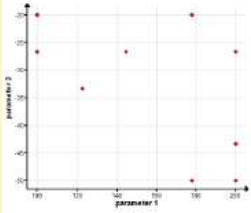
Odyssee uses reduced order models and typically represents the model by its Eigenmodes.

DOE examples with the different methods

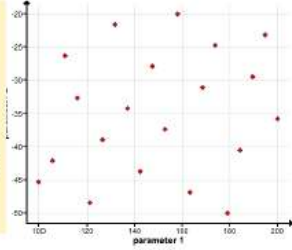
A-optimal



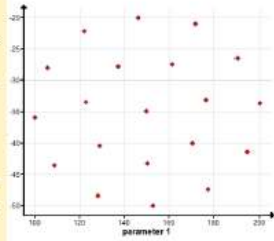
OLH Space filling



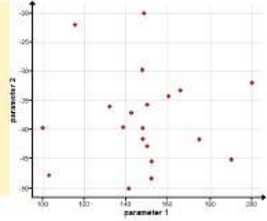
Optimal Latin Hypercube



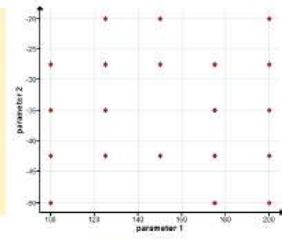
Voronoi (Spherique)



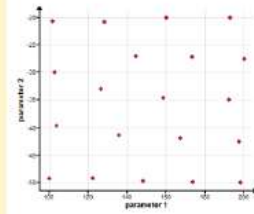
NORMAL



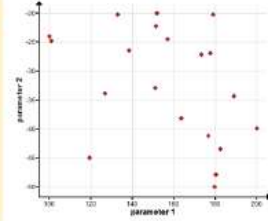
MIN / MAX



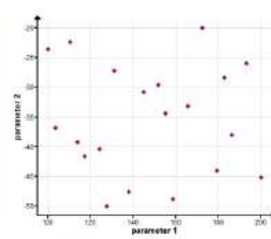
Voronoi (Uniform)



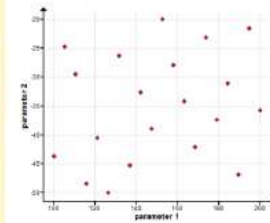
MONTECARLO



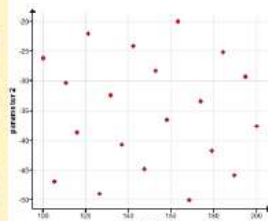
HALTON



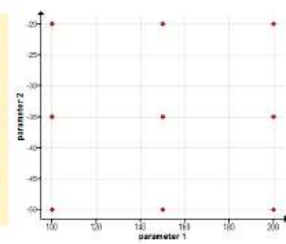
Improved Latin Hypercube



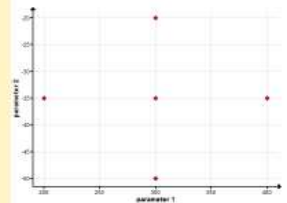
HAMMERSLEY



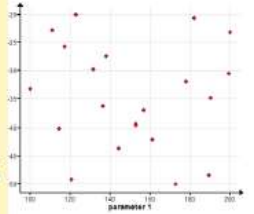
Full Factorial



Box behnken



Latin hypercube



DOE tool: Minimum number of sampling points

The number of sample depends on the budget available to run simulations. We recommend you to use a polynomial formula (order $d=2$) to estimate the number of sampling point. But if the number of variables n is high, we suggest a full factorial formula.

n = variables
 d^* = order

$$(d+1)^n$$

Full factorial formula

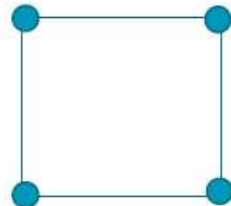
>>

$$C_d^{n+d}$$

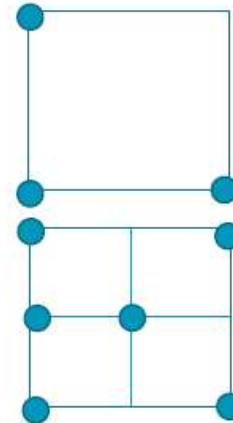
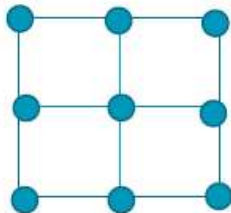
Polynomial formula

$$= (n+d)! / (d! n!)$$

$d=1$ (linear)
 $n=2$



$d=2$ (parabolic)
 $n=2$



$$(2+1)! / (1! 2!) = 3$$

$$(2+2)! / (2! 2!) = 6$$

Image Based

Use of Past Experience to Predict New Outcomes

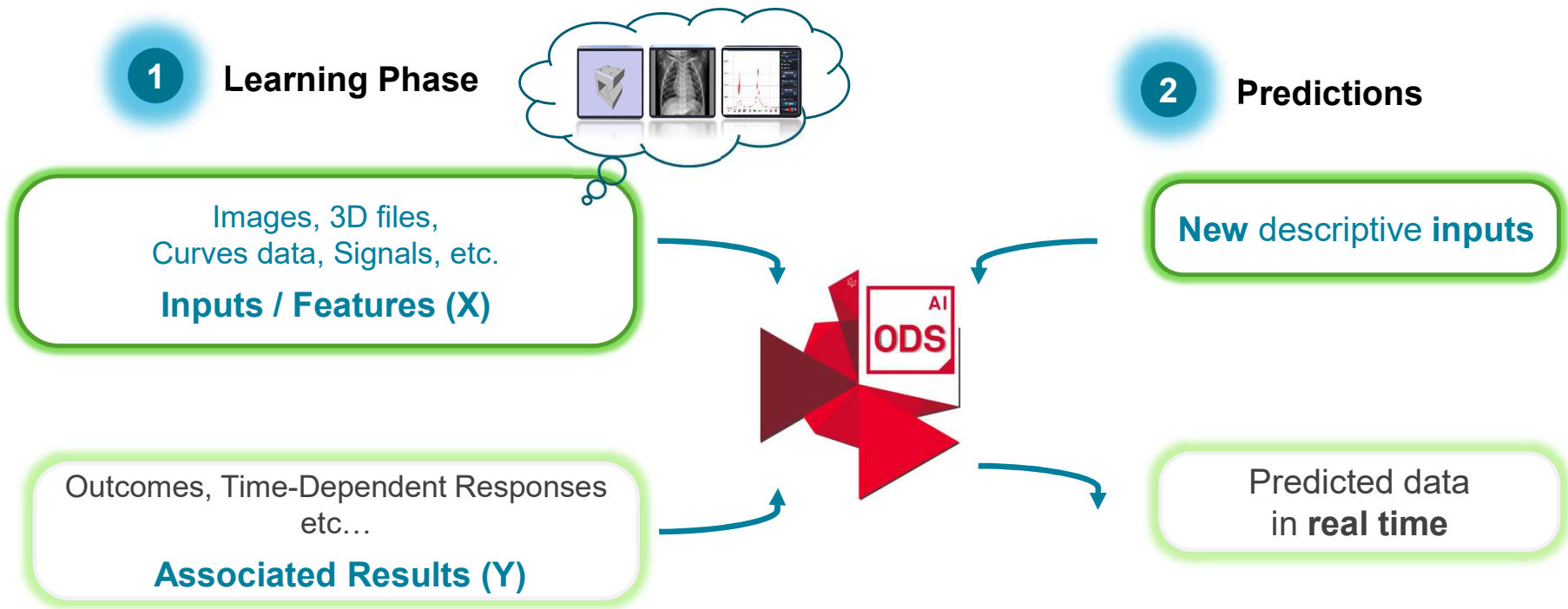
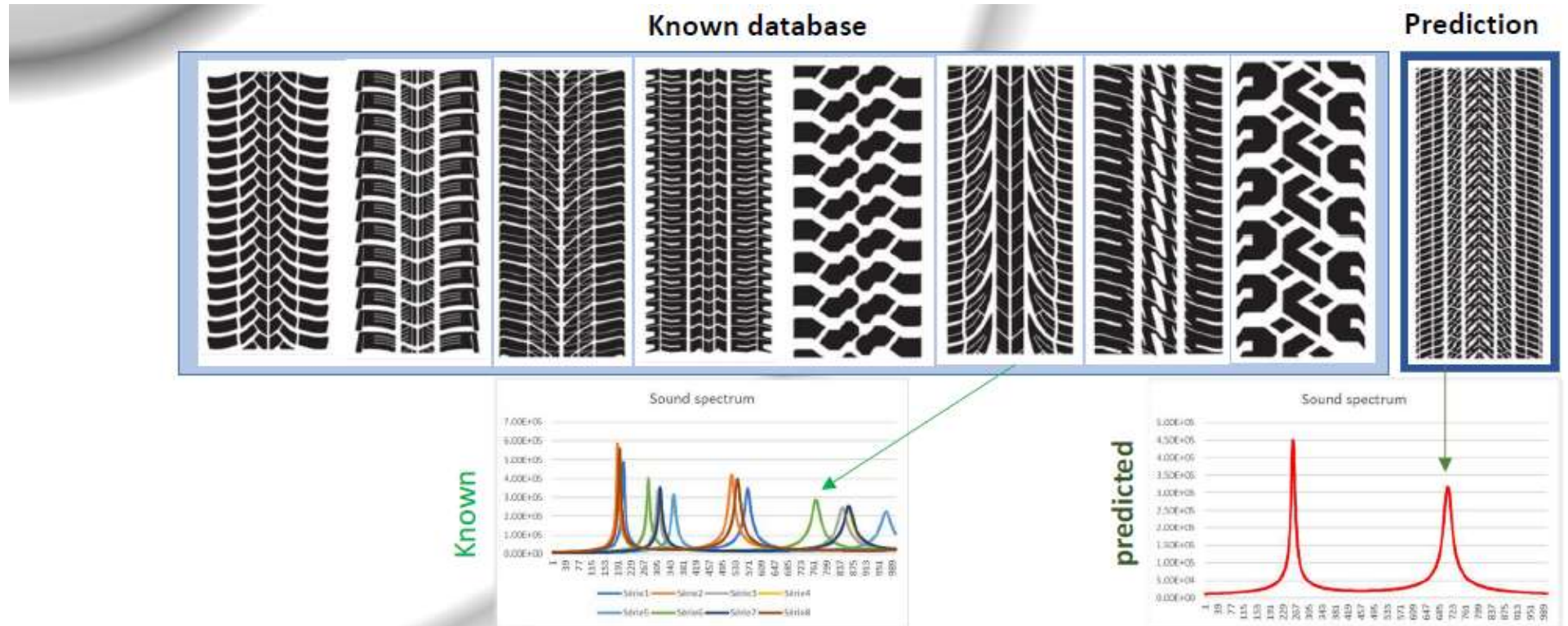


Image Based Learning: Predict Sound Pressure for Tire Profile



Real-Time Product Quality Inspection

A-Eye is not only used for CAE, but also many other applications



Example of valid parts:



Example of defect parts:

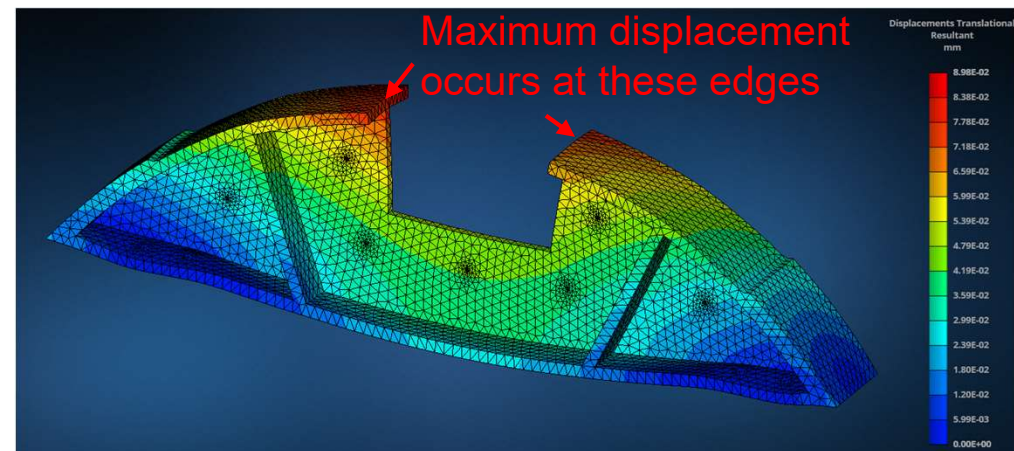
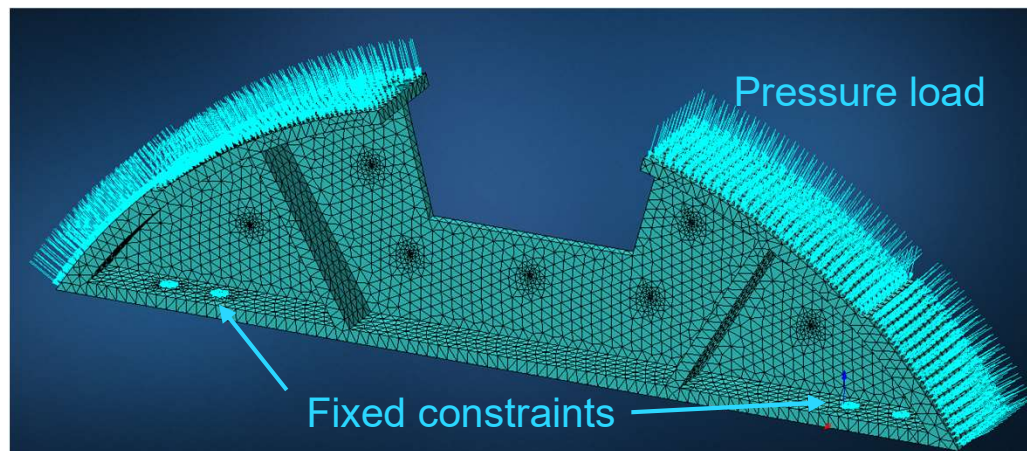


| Learning | Base Viewer |
|--|---------------------------|
| Names Use a file to fill names | RESULTS Results |
| ↓ | ↓ |
| - | Defect |

The state result is displayed

Bracket with Holes – Predict Maximum Displacement from Numbers or Images

- The diameter of the holes is variable
- Predict the maximum displacement for different hole diameters



Bracket with Holes – Predict Maximum Displacement from Numbers or Images

Learning base: Images and corresponding max. displacement values



image1.png



image2.png



image3.png



image4.png



image5.png



image7.png



image8.png



image9.png



image10.png

0.0898

0.0901

0.0903

0.0906

0.0916

0.0952

0.0966

0.101

0.104

Prediction: Predict max. displacement value for new image



image6.png

0.093

Thank you for listening!

Contact:

Cornelia Thieme

Manager Presales DACH

Design and Engineering

Manufacturing Intelligence Division

Hexagon

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E: cornelia.thieme@hexagon.com