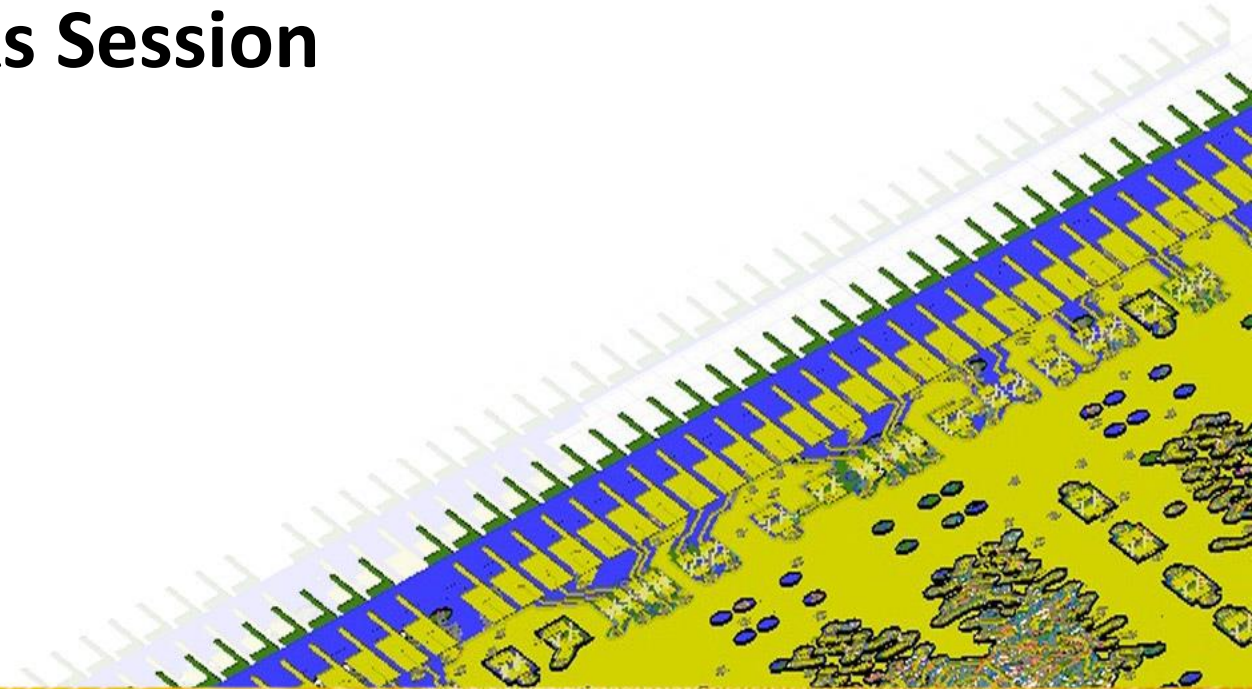


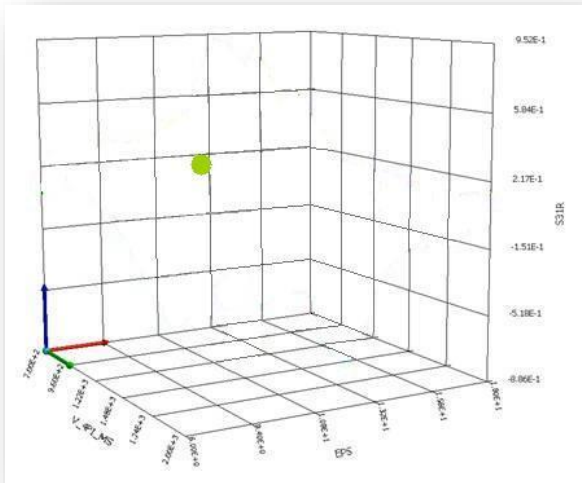


# DesignXplorer Tip & Tricks Session

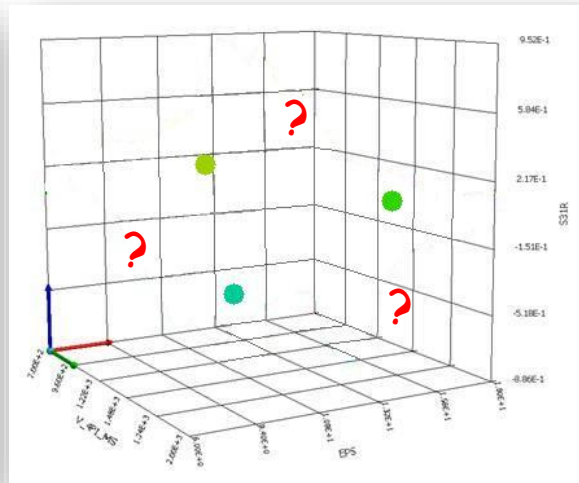


# What is DesignXplorer?

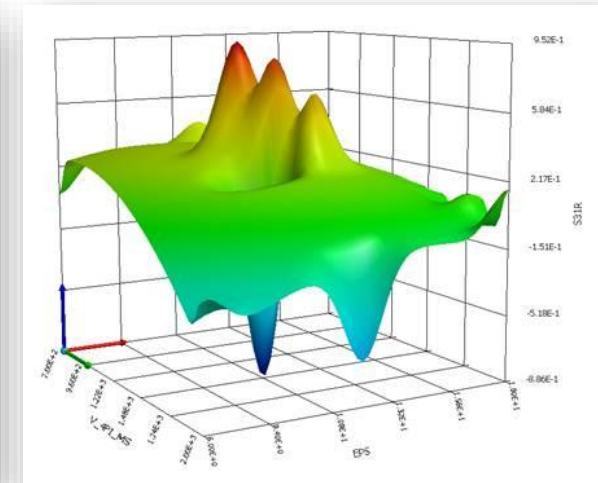
- DesignXplorer, also called DX, is a powerful approach to explore, understand and optimize your engineering challenges
  - Determine the key parameters influencing the design
  - Explore and understand the performance at other design or operating conditions
  - Find the conditions which give the best performance
  - Explore the robustness of the design



*Single Point*



*What If?*



*Response  
Surface*

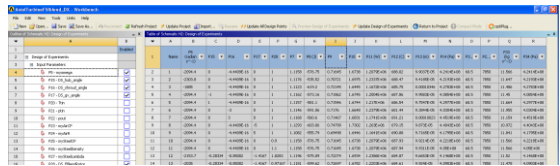
# DesignXplorer Features

## What if Study

Manual Search

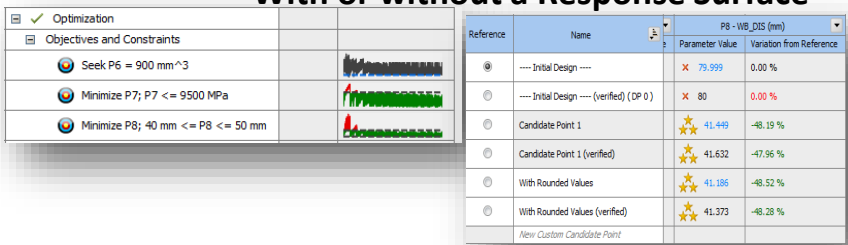
## Design of Experiments

Run a smart set of Design Points



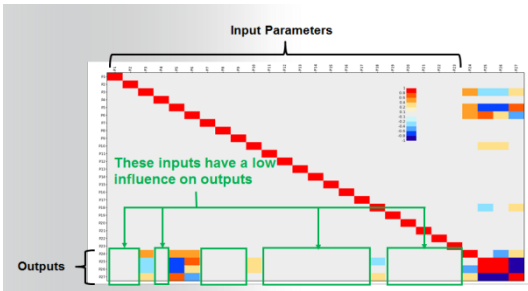
## Optimization

With or without a Response Surface



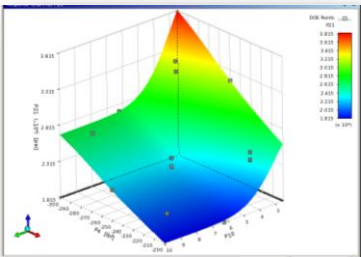
## Sensitivity / Correlation Analysis

Find the relevant parameters



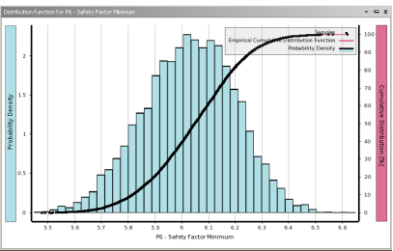
## Response Surface

Build a Mathematical model



## Robust Design

Six Sigma Analysis



*Optimized and Robust Design*

# Free ACT Extensions on the Customer Portal (For DesignXplorer)



DOE from  
Correlation v1



Response surface  
reader



**MATLAB Optimizers for  
DX 17.0**  
Version: 170.3



Direct Optimization  
from RSO v1



Full Factorial DOE v1

Many more ...

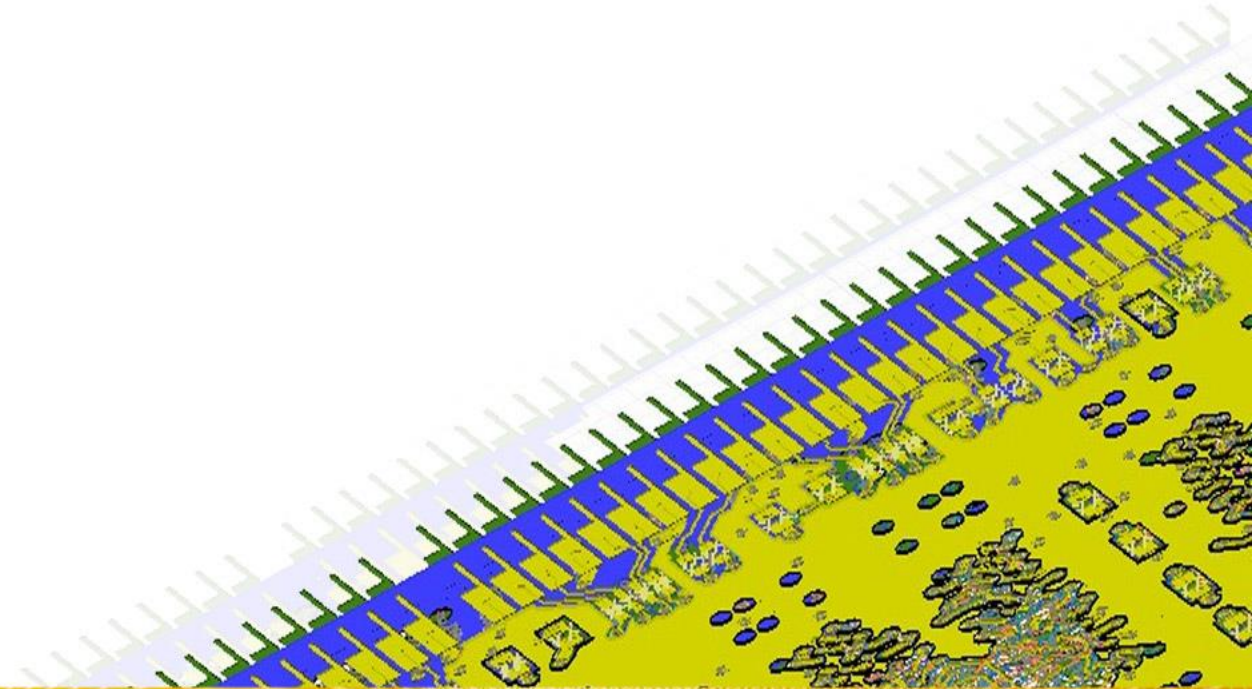
[https://support.ansys.com/AnsysCustomerPortal/en\\_us/Downloads/ANSYS+ACT+Application+Store](https://support.ansys.com/AnsysCustomerPortal/en_us/Downloads/ANSYS+ACT+Application+Store)

**ANSYS**<sup>®</sup>



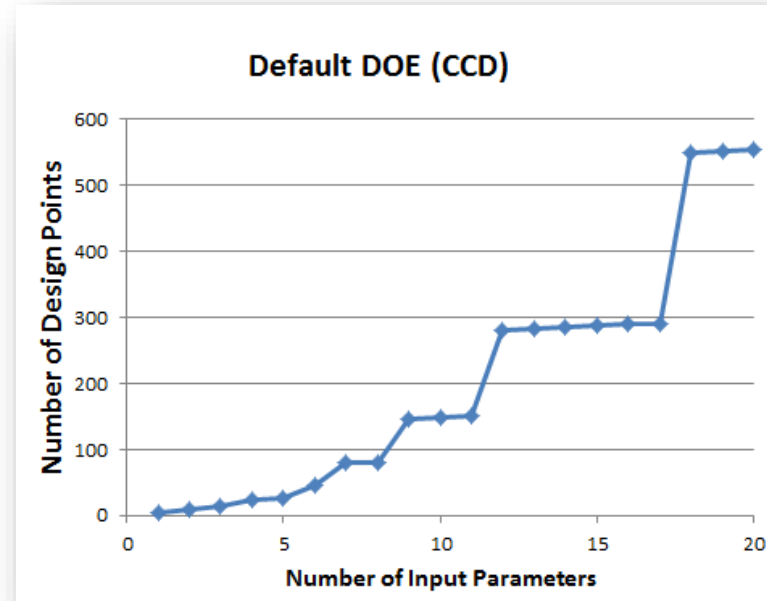


## Parameters Correlation



# Parameters Correlation

- In a DOE study, the amount of design points increases quickly as the number of input parameters increases, which can reduce the efficiency of the analysis process.
- It is recommended to exclude unimportant input parameters from the DOE sampling in order to reduce unnecessary sampling points.



# Parameters Correlation

- The parameter correlation tool allows us to identify important parameters
- Sampling in based on Latin Hypercube Sampling
- Correlation Type :Spearman or Pearson
- Filtering method

Outline of Schematic B2: Parameters Correlation

	A	B
1		Enabled
2	✓ Parameters Correlation	
3	Input Parameters	
4	⚠ Mechanical APDL (A1)	
5	P1 - WIDTH	✓
6	P2 - HEIGHT	✓
7	P3 - LENGTH	✓
8	P4 - FORCE	✓
9	P5 - YOUNG	✓
10	Output Parameters	

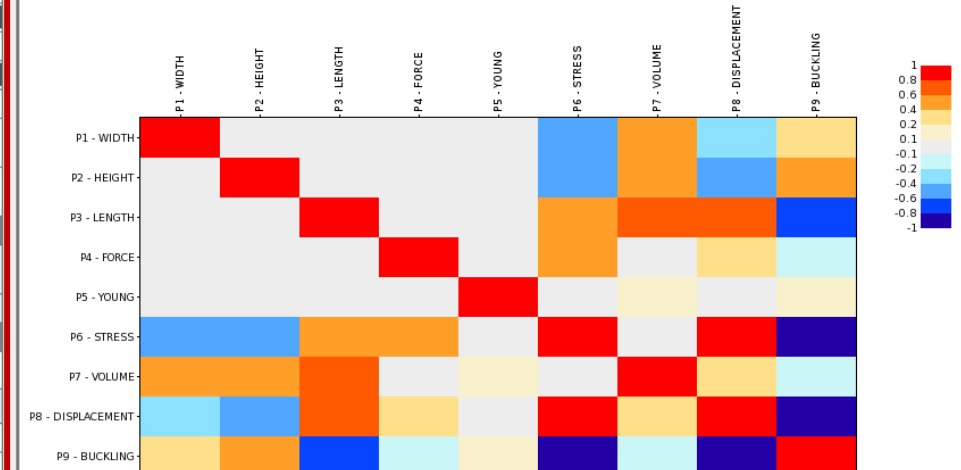
Properties of Outline A2: Parameters Correlation

	A	B
1	Property	Value
2	Design Points	
3	Preserve Design Points After DX Run	<input type="checkbox"/>
4	Failed Design Points Management	
5	Number of Retries	0
6	Parameters Correlation	
7	Reuse the samples already generated	<input checked="" type="checkbox"/>
8	Correlation Type	Spearman
9	Number Of Samples	30
10	Auto Stop Type	Enable Auto Stop
11	Mean Value Accuracy	0.01
12	Standard Deviation Accuracy	0.02
13	Convergence Check Frequency	10
14	Correlation Status	
15	Size of Generated Sample Set	30
16	Converged	No
17	Filtering Method	
18	Relevance Threshold	0.5
19	Correlation Filtering	<input checked="" type="checkbox"/>
20	R2 Contribution Filtering	<input checked="" type="checkbox"/>
21	Maximum Number of Major Inputs	5

Table of Schematic B2: Parameters Correlation

	A	B	C	D	E
1	Filtering Method				
2	Relevance Threshold	0.5			
3	Configuration	Filtering on Correlation Value and R2 Contribution, with a maximum of 5 major input parameters			
4	Filtering Output Parameters	P6 - STRESS, P7 - VOLUME, P8 - DISPLACEMENT, P9 - BUCKLING			
5	Major Input Parameters	Best Relationship With Filtering Output Parameter			
6	Input Parameter	Relevance	Output Parameter	R2 Contribution	Correlation Value
8	P2 - HEIGHT	1	P6 - STRESS	0.1864	-0.61042
9	P3 - LENGTH	1	P7 - VOLUME	0.6055	0.76841
10	P4 - FORCE	0.97229	P6 - STRESS	0.17133	0.59066
11	P1 - WIDTH	0.80185	P6 - STRESS	0.096101	0.52696
12	Minor Input Parameters	Best Relationship With Filtering Output Parameter			
13	Input Parameter	Relevance	Output Parameter	R2 Contribution	Correlation Value
15	P5 - YOUNG	0.45028	P9 - BUCKLING	0.023264	0.16786

Chart: No data



# DOE from Correlation ACT



Project\_Parameters\_Correlation - Workbench

File View Tools Units Extensions Jobs Help

Direct Optimization from RSO DOE from Correlation Import Parameters and DOE Parameter Sweep

Project B2:Parameters Correlation ACT Start Page

Import... Reconnect Refresh Project Update Project Update All Design Points ACT Start Page

Toolbox

- Analysis Systems
- Component Systems
- ACP (Post)
- ACP (Pre)
- BladeGen
- CFX
- Engineering Data
- Explicit Dynamics (LS-DYNA Export)
- External Data
- External Model
- Feedback Iterator
- Finite ElementModeler
- Fluent
- Fluent (with Fluent Meshing)
- Geometry
- Icepak
- Mechanical APDL
- Mechanical Model
- Mesh
- Microsoft OfficeExcel
- Performance Map

Project Schematic

A

- 1 Mechanical APDL
- 2 Analysis
- 3 Parameters

Mechanical APDL

Parameter Set

B

- 1 Parameters Correlation
- 2 Parameters Correlation

Parameters Correlation

DOE from Correlation - Transfer Settings

Correlation System ID: CMT

Reuse Design Points: Yes

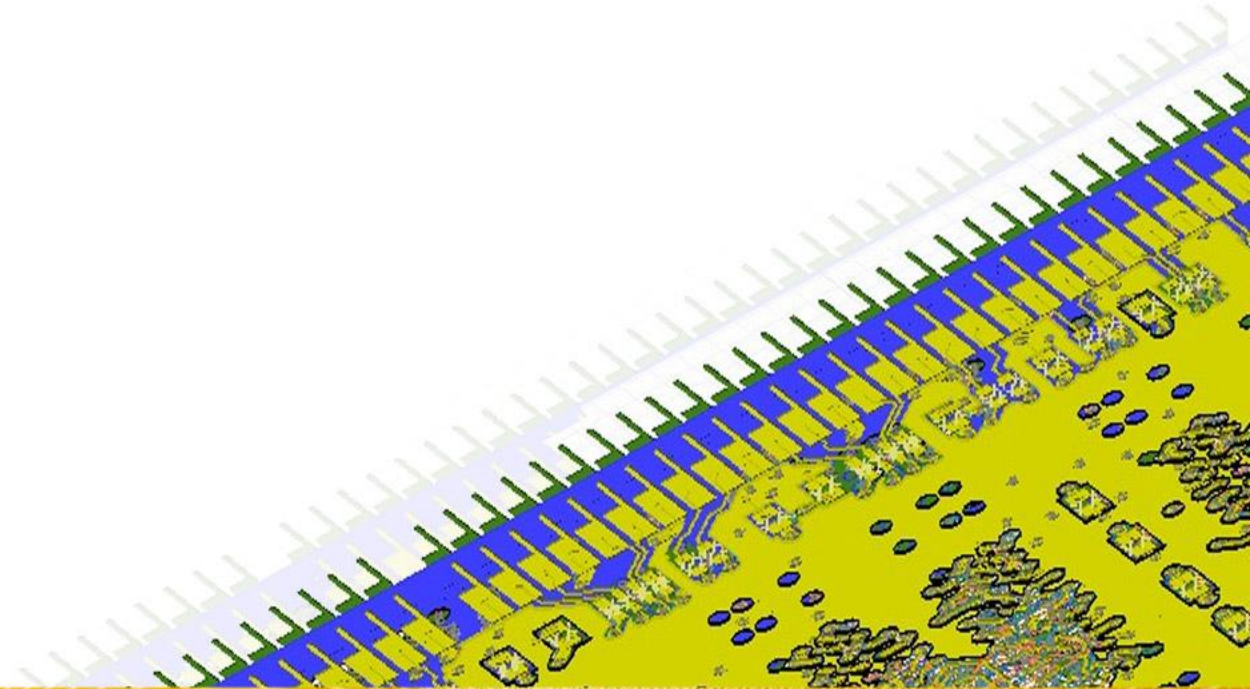
Warning: reusing DPs may lead to inaccurate results. Please see the extension documentation.

OK Cancel





# Design of Experiments



# DOE Purpose and Schemes

The purpose of a Design Of Experiments is to gather a representative set of data to compute a Response Surface, and then run an Optimization (for a Response Surface Optimization)

Basically, a set of Design Points will be calculated

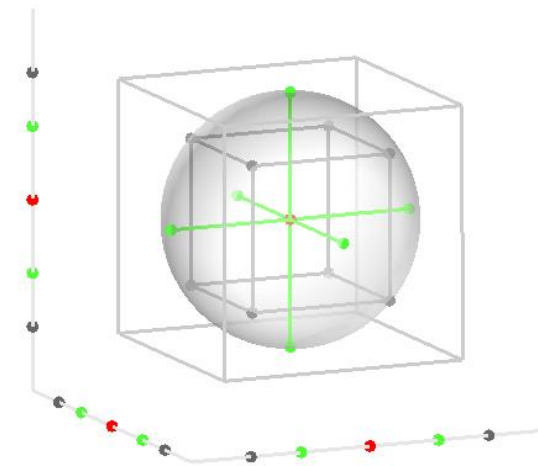
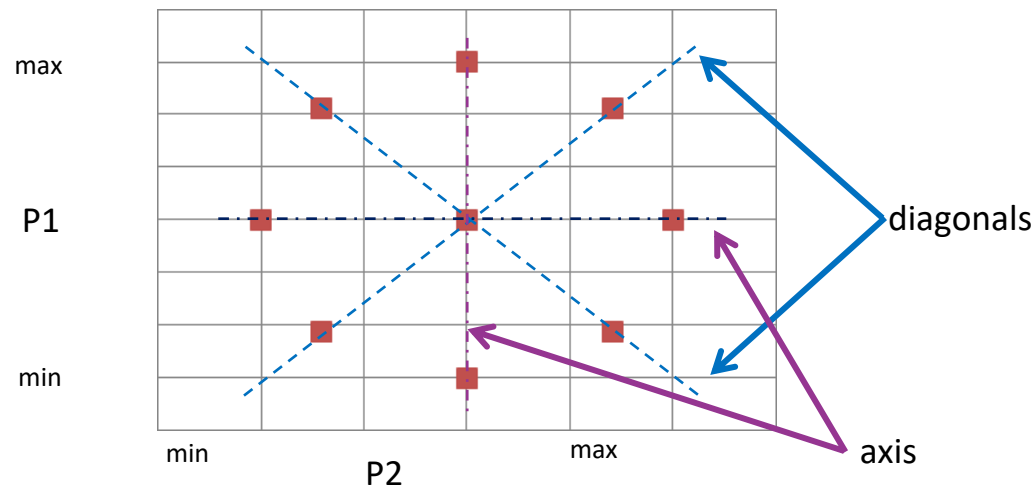
The Response Surface accuracy will greatly depend on the DOE scheme, and especially the number of Design Points that were calculated

DesignXplorer proposes several DOE schemes. Design Points are automatically chosen to explore the parametric space efficiently:

- Central Composite Design (CCD) [default]
- Box Behnken Design
- Optimal Space Filling Design
- Custom + Sampling
- Sparse Grid Initialization
- Latin Hypercube Sampling Design

# Central Composite Design (CCD)

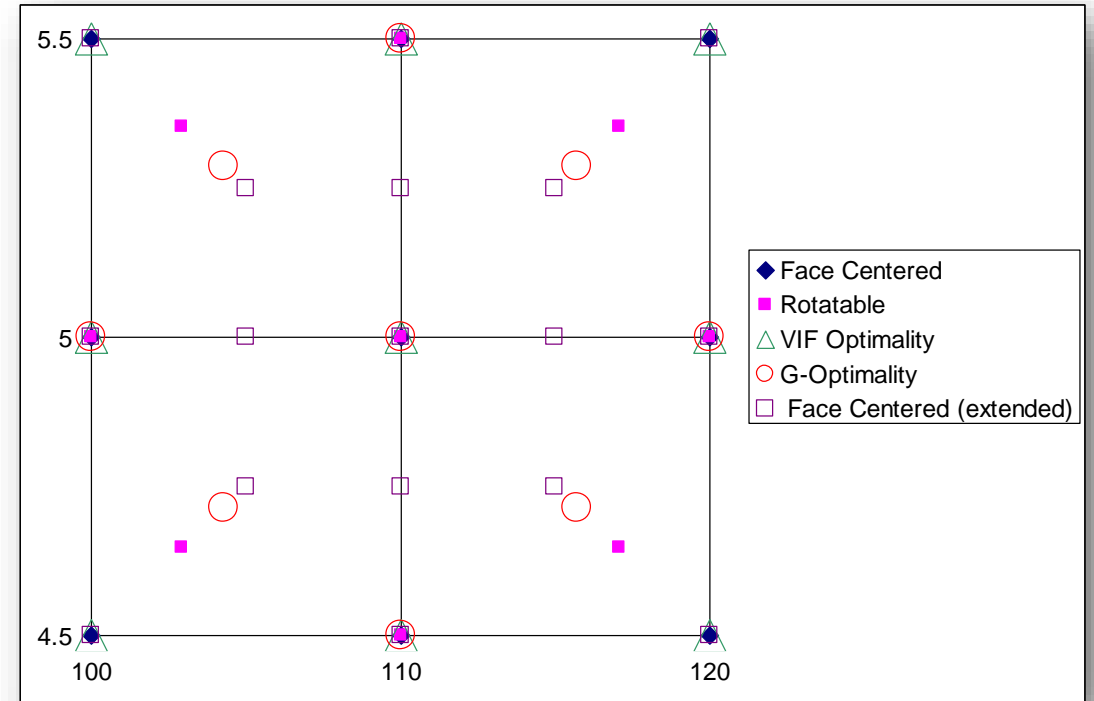
- Central Composite designs are five-level fractional factorial designs that are suitable for calibrating the quadratic response model
- A CCD consists of:
  - 1 center point
  - $2 \times N$  axis points located at the  $-a$  and  $+a$  positions on each axis of the selected input parameter
  - $2^{(N-f)}$  factorial points located at the  $-1$  and  $+1$  positions along the diagonals of the input parameter space [factorial number  $f$  discussed on next slide]



# Central Composite Design (CCD)

- There are 5 types of CCDs available each with their own benefits and drawbacks

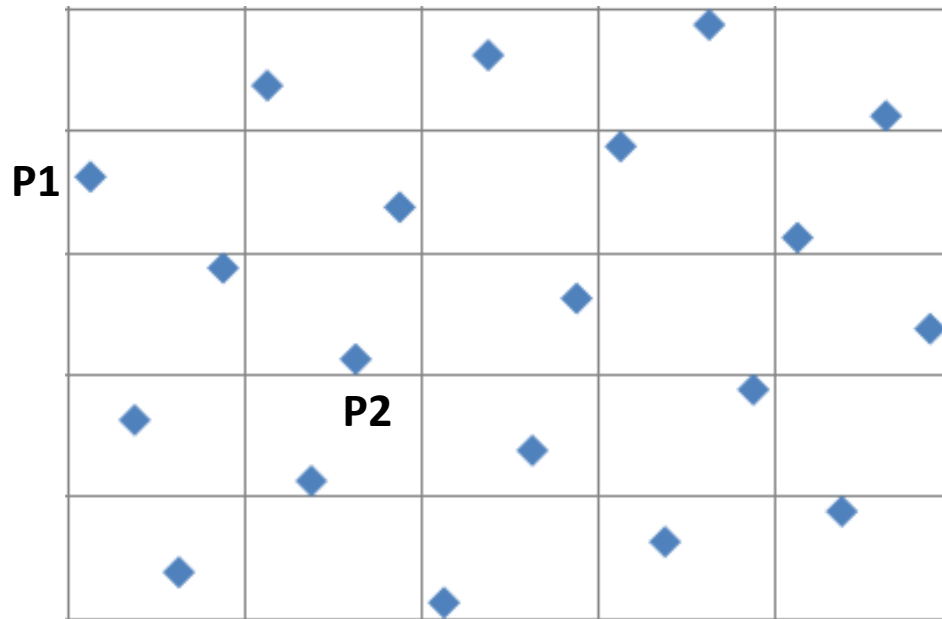
1. **Auto Defined (default):** Automatically switches between the G-Optimal (if the number of input variables is 5) or VIF-optimal otherwise
2. **Face Centered:** 3 levels, not rotatable. Benefit is that it gets sampling points at all extremes.
3. **Rotatable:** 5 levels and rotatable. Drawback: does not get sampling points at all extremes. Advantage: variance prediction is the same for any two locations that are the same distance from the design center
4. **VIF (Variance Inflation Factor) Optimality:** Maximizes orthogonality
5. **G-Optimality:** Minimizes leverage



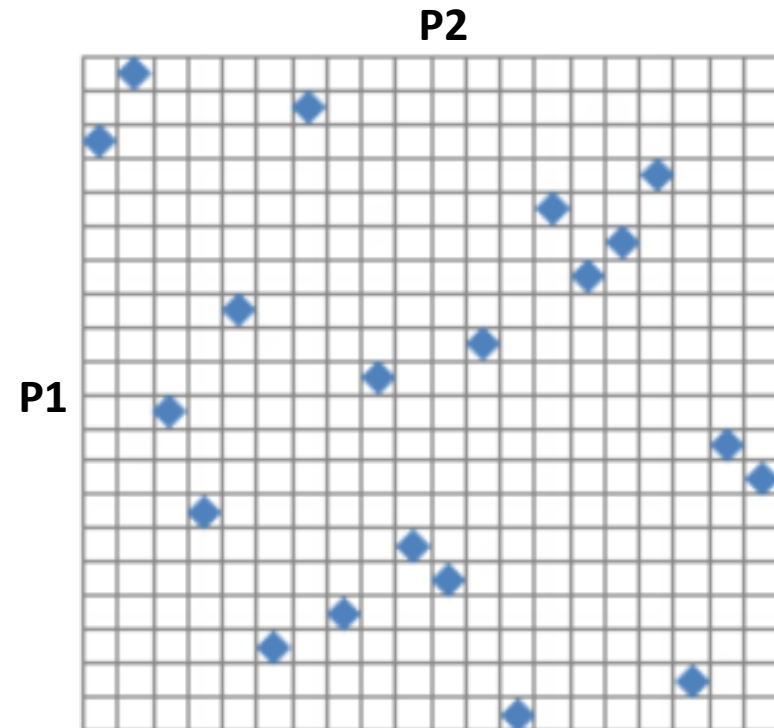
- For 2 parameters, the standard CCD schemes are based on 9 points
- The “extended” face centered is made of 17 points (note: there is also a 17 points Rotatable scheme)

## Other DOE's

## Optimal Space Filling Design



# Latin Hypercube Sampling

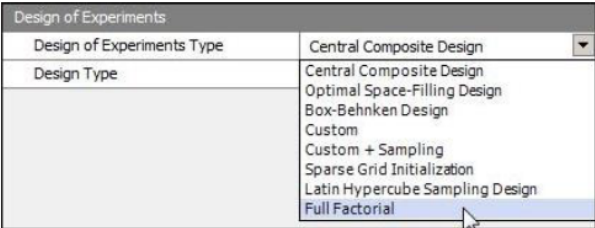
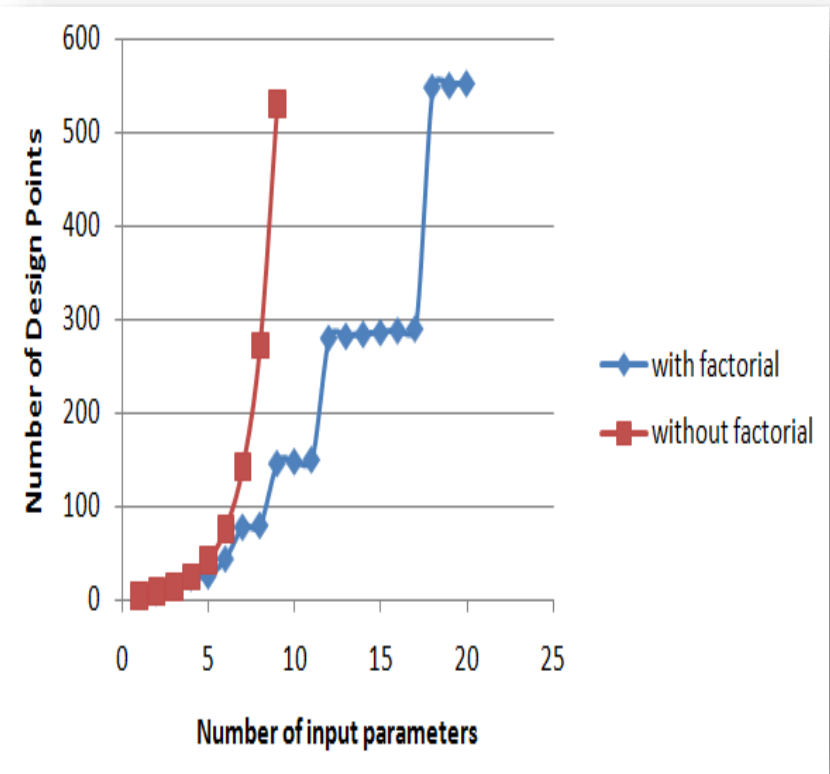




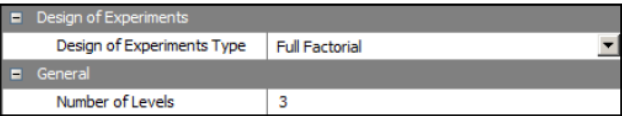
# Full Factorial DOE ACT



Number of Parameters	f	with factorial	without factorial
1	0	5	5
2	0	9	9
3	0	15	15
4	0	25	25
5	1	27	43
6	1	45	77
7	1	79	143
8	2	81	273
9	2	147	531
10	3	149	1,045
11	4	151	2,071
12	4	281	4,121
13	5	283	8,219
14	6	285	16,413
15	7	287	32,799
16	8	289	65,569
17	9	291	131,107
18	9	549	262,181
19	10	551	524,327
20	11	553	1,048,617

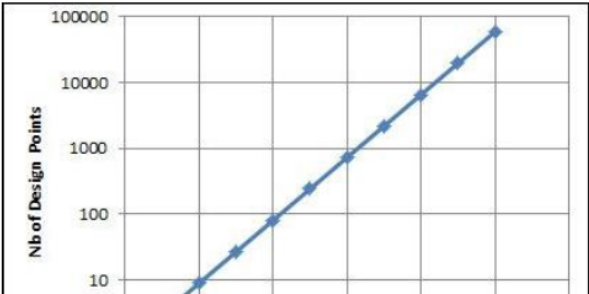


Under **General**, the **Number of Levels** property is shown. The default is 3 to create lower, mid, and upper parameter bounds.



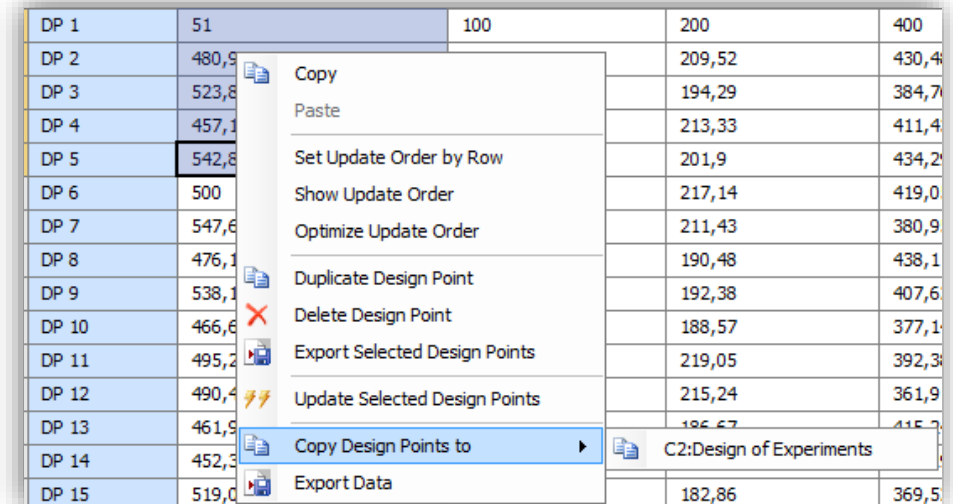
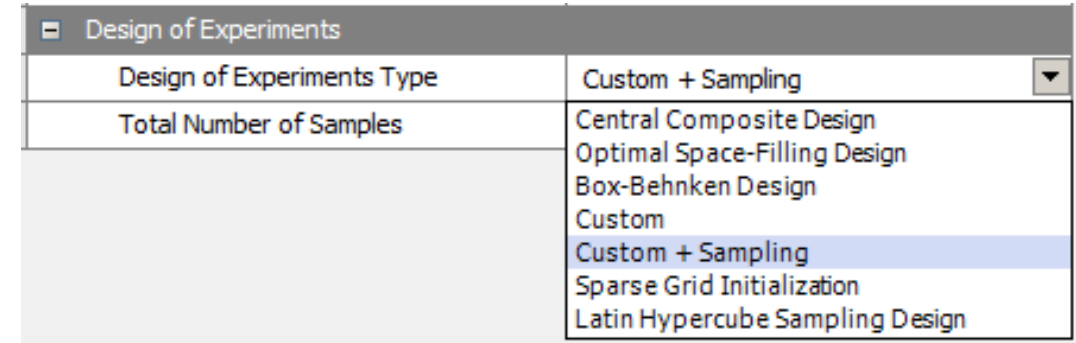
4. Specify the number of levels that you want to use.

CAUTION: The number of required design points increases very quickly when the **Full Factorial** DOE type is used. For a three-level full factorial design, the following graph shows the log scale on the Y axis.



# Custom Design of Experiments

- To import a DOE generated with 3rd party tool
- To import external DOE or experimental data with output parameters values
- To reuse design points (CSV import, or copy from the Parameter Set)



# CSV Input File Requirements

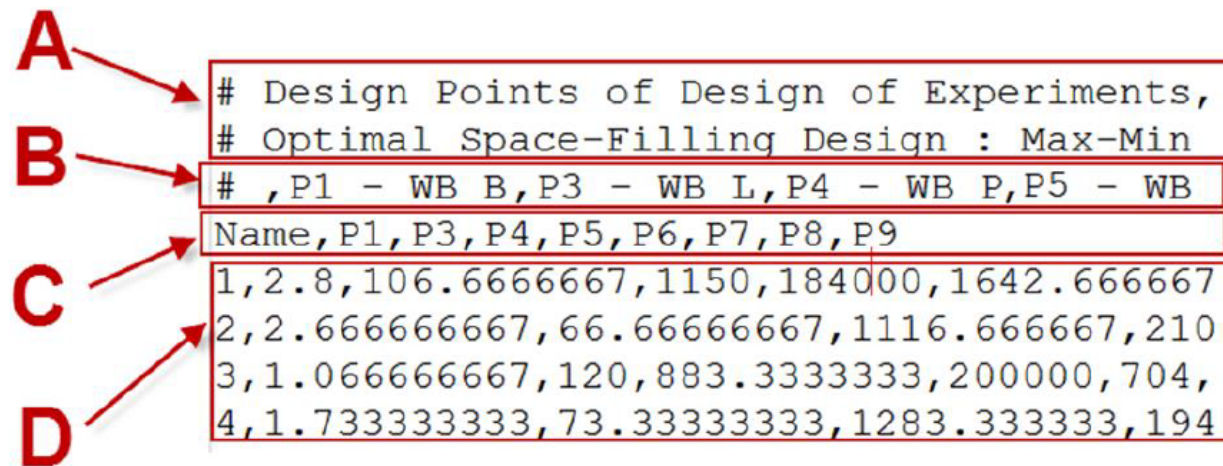
The CSV input file must be in the extended format supported by Workbench. The sample CSV file below is truncated for the purposes of illustration.

**A:** The lines beginning with “#” are comment lines. Except for the line defining the parameter names (see B), these comment lines are ignored.

**B:** The parameter names must be listed one line above the line defining the IDs. Parameter full names are extracted from this line. Based on the comma separator, the parameter names should be aligned with the corresponding parameter IDs in the line below. Parameter names cannot include commas.

**C:** The parameter IDs are extracted from this line. The line must be below the line defining the parameter full names.

**D:** The DOE values must be listed below the line defining the parameter IDs.



```
# Design Points of Design of Experiments,  
# Optimal Space-Filling Design : Max-Min  
# ,P1 - WB B,P3 - WB L,P4 - WB P,P5 - WB  
Name,P1,P3,P4,P5,P6,P7,P8,P9  
1,2.8,106.6666667,1150,184000,1642.666667  
2,2.66666667,66.6666667,1116.666667,210  
3,1.06666667,120,883.3333333,200000,704,  
4,1.733333333,73.33333333,1283.333333,194
```

# DOE with failed DPs: how to proceed?

DOE table with a failed DP:

	A	B	C
1	Name ▾	P1 ▾	P2 ▾
2	1	3	0.33333
3	2 DP 1	0	
4	3	6	0.16667
5	4	1.5	0.66667
6	5	4.5	0.22222

1. Set the DOE type to Custom:

Design of Experiments	
Design of Experiments Type	Custom

2. Right click on the failed DP > Delete

Name ▾	P1 ▾	P2 ▾
1	3	0.33333
2 DP 1	0	
3	6	0.1666
4	1.5	0.6666
5	4.5	0.2222
New Design Point		

Show Update Order

Set Update Order by Row

Optimize Update Order

Import Design Points...

Insert as Design Point

Delete

Set Output Values as Editable

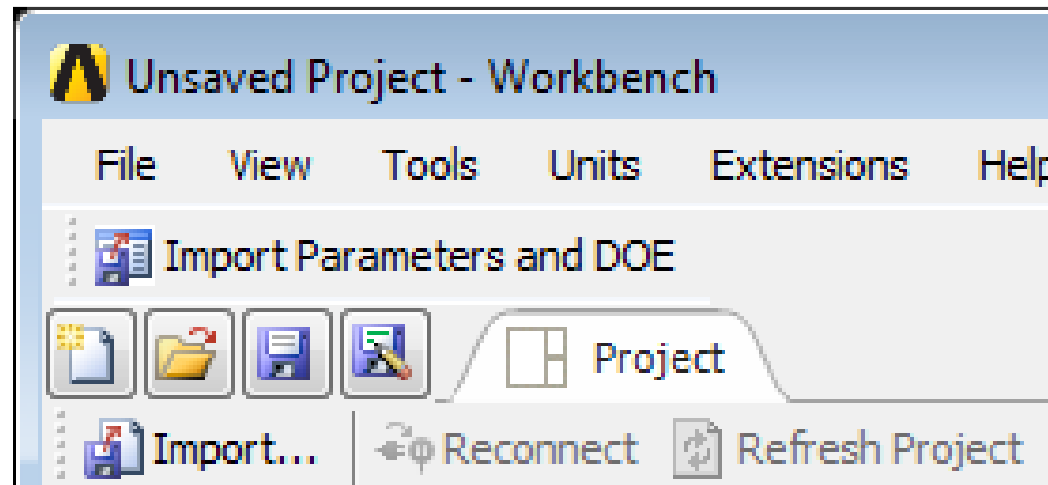
Failed DPs can also be deleted using python script

3. Update DOE (no solve is required)

# Import Parameters and DOE ACT



The Import Parameters and DOE extension enables you to automatically create parameters and import their DOE values into an empty Workbench project. For example, if a DOE was solved in Workbench (or any other tool) and saved as a CSV file, you can load it into Workbench to generate a response surface and eventually run an optimization using a standalone DesignXplorer system.





# Easily Switch Between Solved Design Points

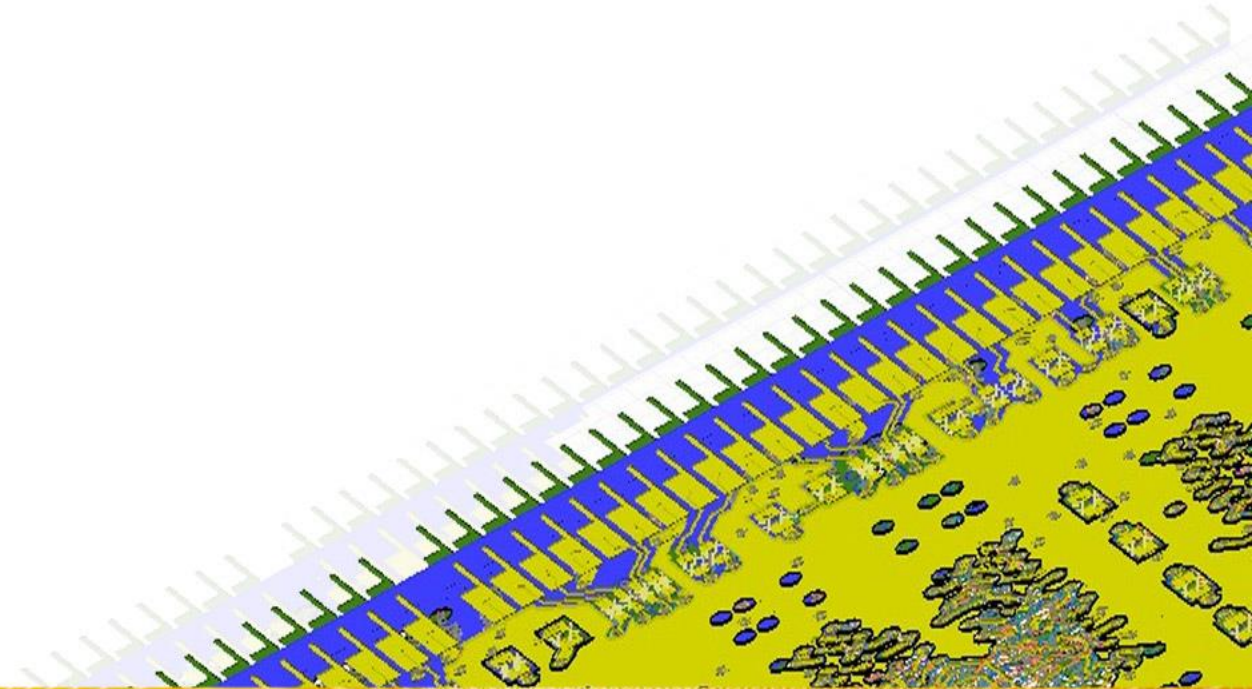
- Take advantage of “Retained Data”
  - Quickly switch between retained design points
    - “Set as Current” option
  - Export retained design points



Table of Design Points							
	A	B	C	D	E	F	G
1	Name ▼	P1 - Pressure Magnitude ▼	P2 - Equivalent Stress Minimum ▼	P3 - Equivalent Stress Maximum ▼	<input type="checkbox"/> Retain	Retained Data	Note ▼
2	Units	Pa ▼	Pa	Pa			
3	DP 0 (Current)	2000	19.512	1.0622E+05	<input checked="" type="checkbox"/>	✓	Harmonic testing
4	DP 1	5000	48.779	2.6556E+05	<input checked="" type="checkbox"/>	✓	
5	DP 2	3000	29.267	1.5934E+05	<input checked="" type="checkbox"/>	⚡	Harmonic testing
6	DP 3	4000	⚡ 58.535	⚡ 3.1867E+05	<input checked="" type="checkbox"/>	⚡ ✓	Harmonic testing
7	DP 4	6000	39.023	2.1245E+05	<input type="checkbox"/>		
*					<input type="checkbox"/>		



## Response Surface



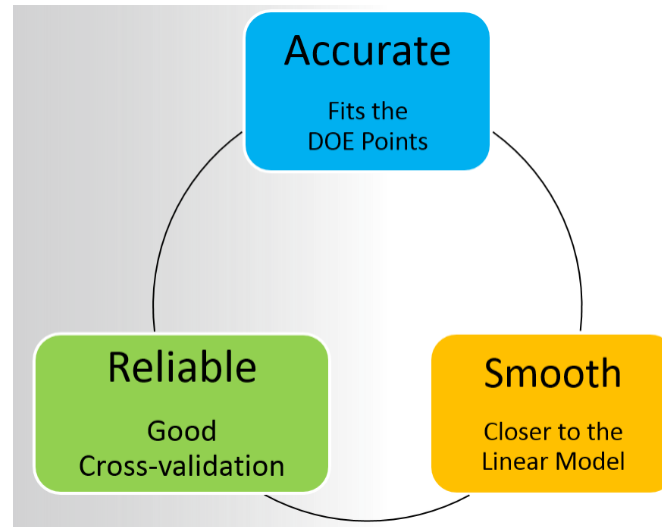
# Response Surface

## Definition

- **Response Surface = Surrogate Model = Meta Model = Approximation Model**
- **Response Surfaces are functions of different nature where the output parameters are described in terms of the input parameters**
- **Response Surfaces provide the approximated values of the output parameters, everywhere in the analyzed design space, without the need to perform a complete solution**
- **The response surface methods described here are suitable for problems using up to ~10-15 input parameters**

# Genetic Aggregation

- It runs an iterative genetic algorithm to find the best Response Surface type and settings for each output parameter. It selects the best ones and combines them to build an “aggregation” of several Response Surfaces. It results to the best RS quality and different settings for each output parameter
- Even if the algorithm runs on several CPUs, this method can still be long when dealing with high number of design points and/or parameters
- The goal is to meet the 3 main criteria to obtain the best Response Surface:



# Genetic Aggregation

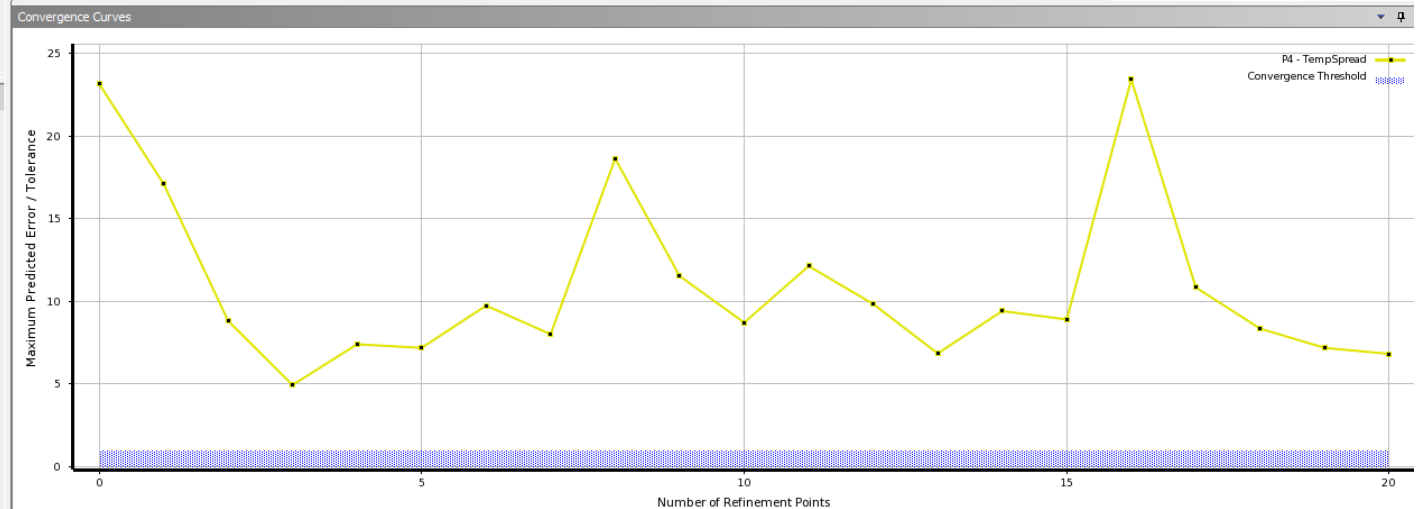
Automatic Refinement **new from R17.1**

Auto refinement adds design points until response surface accuracy meets user requirements

Outline of Schematic B3: Response Surface		
	A	B
1		Enabled
2	Response Surface	
3	Input Parameters	
4	Fluid Flow (FLUENT) (A1)	
5	P1 - inletcold_velocity	<input checked="" type="checkbox"/>
6	P2 - inlethot_velocity	<input checked="" type="checkbox"/>
7	Output Parameters	
8	Fluid Flow (FLUENT) (A1)	
9	P3 - PressureDrop	
10	P4 - TempSpread	
11	Min-Max Search	<input checked="" type="checkbox"/>
12	Refinement	
13	Tolerances	
14	Refinement Points	
15	Quality	

Properties of Outline A2: Response Surface		
	A	B
1	Property	Value
2	Design Points	
3	Preserve Design Points After DX Run	<input type="checkbox"/>
4	Failed Design Points Management	
5	Number of Retries	0
6	Meta Model	
7	Response Surface Type	Genetic Aggregation
8	Refinement	
9	Maximum Number of Refinement Points	20
10	Number of Refinement Points	20
11	Convergence State	Not Converged: Maximum Number of Points reached
12	Verification Points	
13	Generate Verification Points	<input type="checkbox"/>

Table of Schematic B3: Response Surface: Tolerances						
	A	B	C	D	E	F
1	Name	Calculated Minimum	Calculated Maximum	Maximum Predicted Error	Refinement	Tolerance
2	P3 - PressureDrop (Pa)	12.235	144.26	0.90251	<input type="checkbox"/>	
3	P4 - TempSpread (K)	0.33964	0.98977	0.06829	<input checked="" type="checkbox"/>	0.01

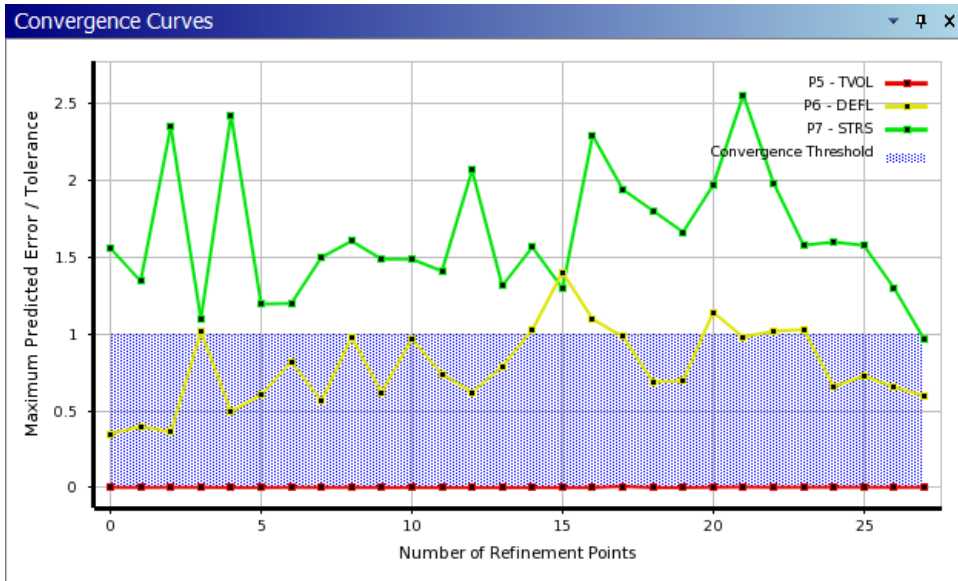




# Genetic Aggregation

Automatic Refinement Convergence **new from R17.1**

Table of Schematic B3: Response Surface								
	A	B	C	D	E	F	G	H
1	Name	P1 - TK16	P2 - TK27	P3 - TK38	P4 - TK49	P5 - TVOL	P6 - DEFL	P7 - STRS
22	20	0.19896	0.15	0.27	0.22162	4.0297	0.63451	34352
23	21	0.19329	0.17568	0.25328	0.21978	4.0915	0.47772	27603
24	22	0.20472	0.19524	0.27	0.15	3.881	0.47407	43324
25	23	0.2693	0.21473	0.27	0.20692	4.4619	0.28046	28690
26	24	0.2586	0.19297	0.26845	0.15	3.9604	0.42261	43447
27	25	0.15	0.15	0.20441	0.218	3.6404	0.79337	33818
28	26	0.20816	0.27	0.27	0.15	4.3389	0.3179	40722
29	27	0.26664	0.15	0.27	0.22273	4.1617	0.50478	29969
*	New Refiner Point							



- Refinement points are added automatically
  - Algorithm to place new refinement points efficiently uses NLPQL optimization to find the location where predicted error is greatest
- X-axis shows number of refinement points
- Y-axis shows ratio between the max predicted error and the tolerance for each output
- Convergence occurs when all the outputs are within the convergence threshold

# Goodness of Fit Table

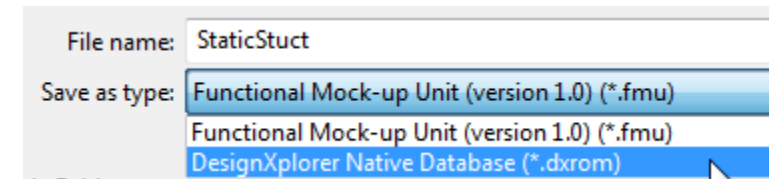
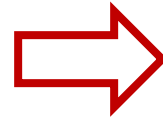
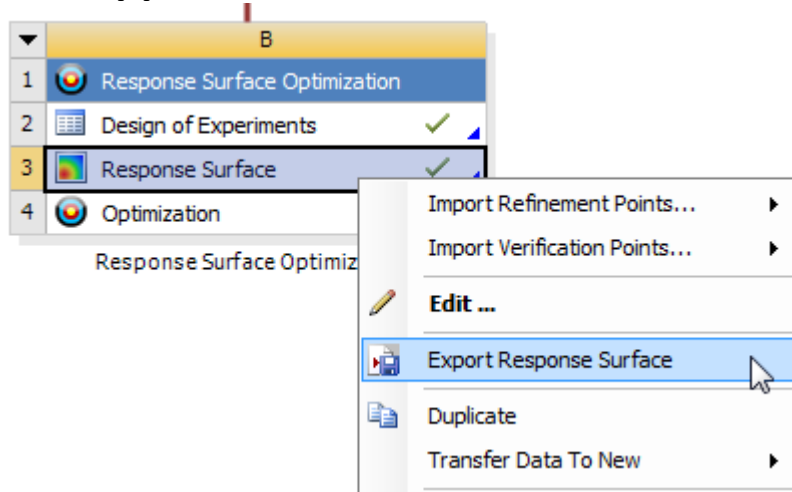
Table of Schematic C3: Response Surface				
	A	B	C	D
1		P6 - Geometry Mass	P14 - Safety Factor HOOP Minimum	P15 - Safety Factor PIPE Minimum
2	☐ Coefficient of Determination (Best Value = 1)			
3	Learning Points	★★★ 1	★★★ 0.99997	★★★ 0.99998
4	Cross-Validation on Learning Points	★★★ 1	★★★ 0.9994	★★★ 0.99949
5	☐ Root Mean Square Error (Best Value = 0)			
6	Learning Points	1.8876E-06	0.0044675	0.00051469
7	Verification Points	4.4587E-06	0.01796	0.00025673
8	Cross-Validation on Learning Points	5.0531E-06	0.020459	0.0023788
9	☐ Relative Maximum Absolute Error (Best Value = 0%)			
10	Learning Points	★★★ 0	★★ 1.3133	★★ 1.2211
11	Verification Points	★★★ 0	★ 2.1574	★★★ 0.24427
12	Cross-Validation on Learning Points	★★★ 0.011966	— 5.3961	— 6.4131
13	☐ Relative Average Absolute Error (Best Value = 0%)			
14	Learning Points	★★★ 0	★★★ 0.41866	★★★ 0.36291
15	Verification Points	★★★ 0	★ 2.1574	★★★ 0.24427
16	Cross-Validation on Learning Points	★★★ 0.0027904	★ 2.0526	★★ 1.7183

Type of Point:	GOF Measures:
Learning Points	Quality of the interpolation
Verification Points	Quality of the prediction
Cross-Validation Learning Points	Stability/reliability of the response surface

# Response Surface

## Export

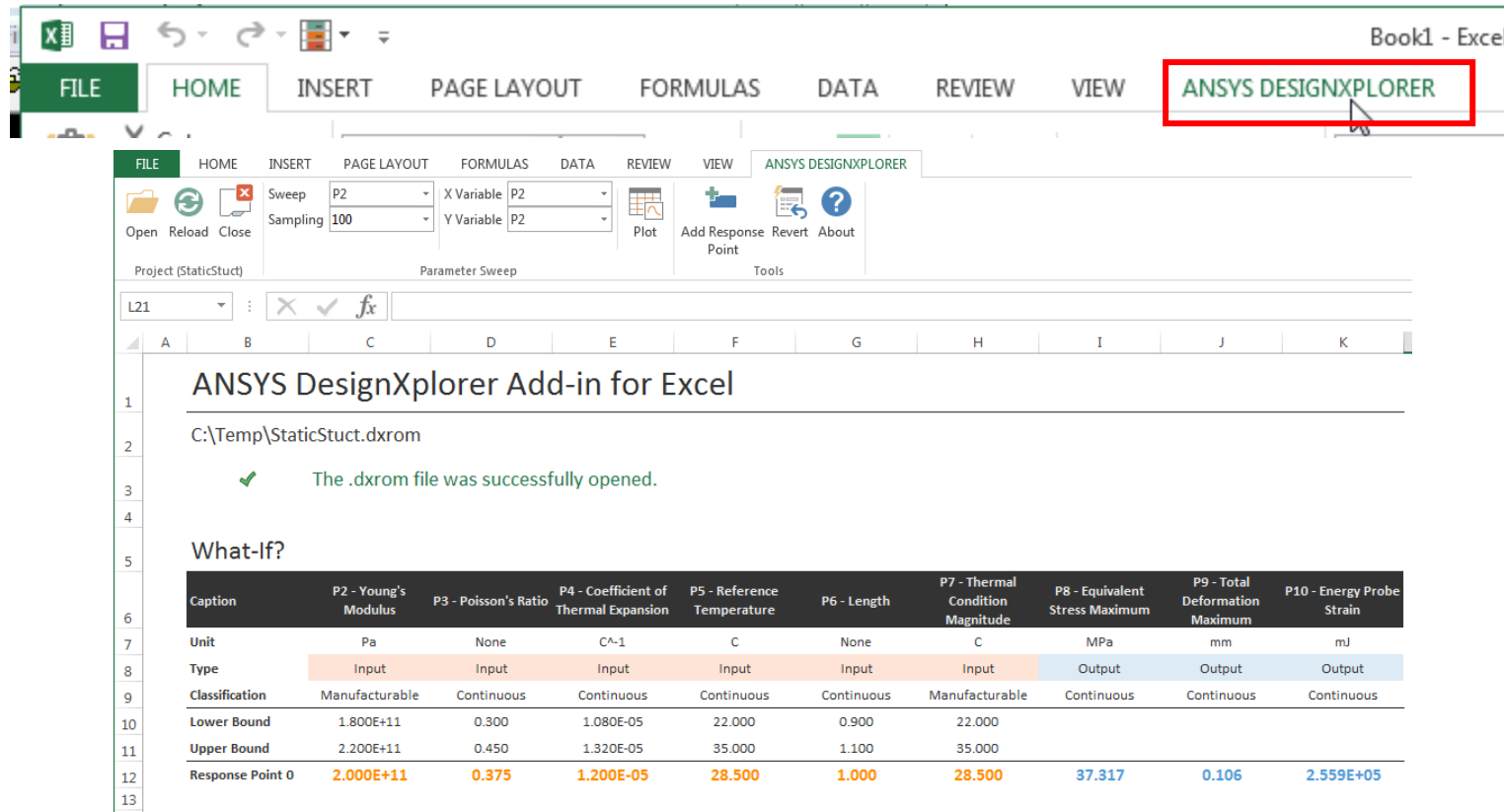
- All Response Surfaces can be exported, except Neural Network and Sparse Grid
- DesignXplorer can export a DX-ROM response surface in the following file formats:
  - Functional Mock-up Unit: \*.fmu
    - Broadest applicability; can be used by Simplorer, MATLAB, or any software with Functional Mock-up Interface (FMI) support. Recommended format for export to external software
  - DesignXplorer Native Database: \*.dxrom
    - Can be imported into Excel or Workbench using custom DX-ROM tools provided by ANSYS on the ACT Application Store



# Response Surface

Import in Excel using the 'Response Surface Reader' ACT extension

- Response Surfaces that were exported as \*.dxrom files can be imported in Excel
- \*.fmu file, we can rename it to \*.zip, unpack it and grab the dxrom file under the folder



ANSYS DesignXplorer Add-in for Excel

C:\Temp\StaticStuct.dxrom

✓ The .dxrom file was successfully opened.

What-If?

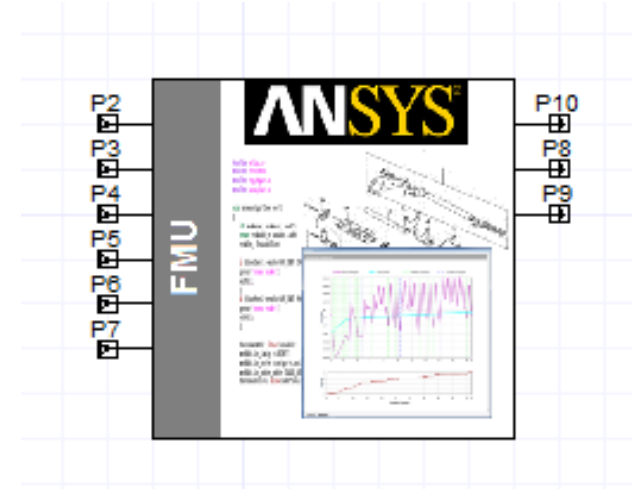
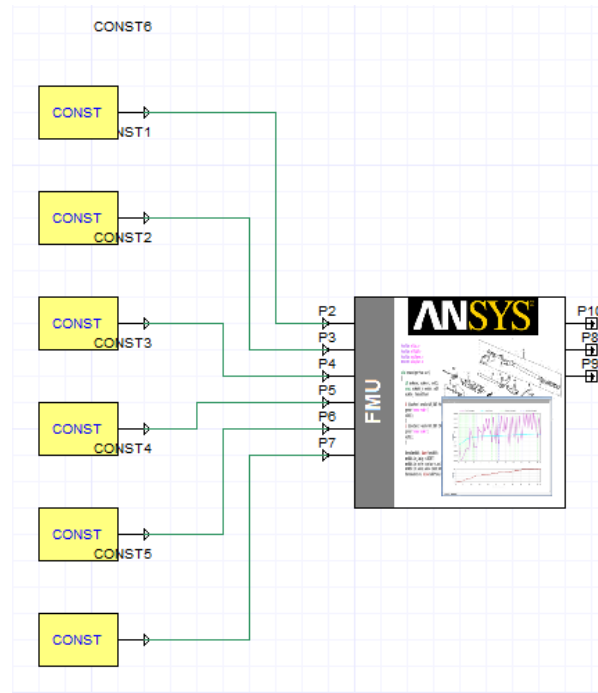
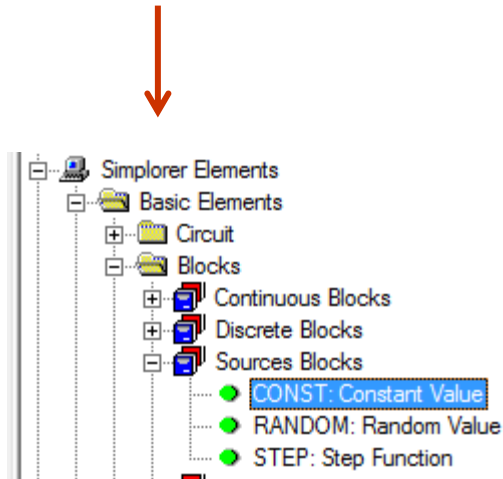
Caption	P2 - Young's Modulus	P3 - Poisson's Ratio	P4 - Coefficient of Thermal Expansion	P5 - Reference Temperature	P6 - Length	P7 - Thermal Condition Magnitude	P8 - Equivalent Stress Maximum	P9 - Total Deformation Maximum	P10 - Energy Probe Strain
Unit	Pa	None	C^-1	C	None	C	MPa	mm	mJ
Type	Input	Input	Input	Input	Input	Input	Output	Output	Output
Classification	Manufacturable	Continuous	Continuous	Continuous	Continuous	Manufacturable	Continuous	Continuous	Continuous
Lower Bound	1.800E+11	0.300	1.080E-05	22.000	0.900	22.000			
Upper Bound	2.200E+11	0.450	1.320E-05	35.000	1.100	35.000			
Response Point 0	2.000E+11	0.375	1.200E-05	28.500	1.000	28.500	37.317	0.106	2.559E+05

Importing the Response Surface in Excel does not necessarily need an ANSYS license

# Response Surface

## Import in Simplorer

- **Response Surfaces that were exported as \*.fmu files can be imported in Simplorer**
  - In Simplorer, go to Simplorer Circuit > Sub-circuit > Add FMU Component
  - Select the \*.fmu file > Open > OK
  - Click to place the MU system into the circuit. Then press Esc
  - Then link the input/output pins to the other component
  - To test this component, you can add some constant inputs and check the output values





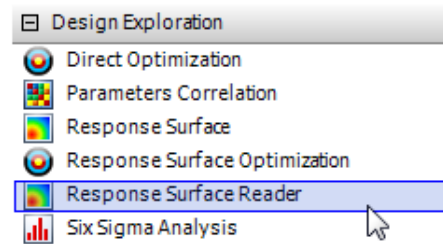
# RS Reader ACT



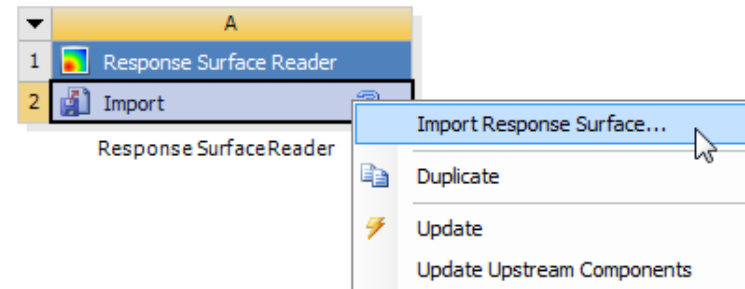
## Import a Response Surface in Workbench

Once the extension is loaded, you can create a Response Surface Reader system.

- Double click on the RS Reader system from the Toolbox, under Design Exploration



- Right click on Import > Import Response Surface...



- Select the \*.dxrom file > Open  
The file is loaded. Input and output parameters are automatically created.

# Quadratic RS: how to extract the equations?

## Generate the Advanced Goodness of Fit Report

1. Edit the Response Surface component, select Goodness of Fit
2. Right click on the output parameter you want to assess > Generate Advanced Goodness of Fit Report

The screenshot shows the ANSYS Workbench interface. On the left, the 'Response Surface' component is expanded, showing 'Input Parameters' (P1 - WB\_Thickness, P2 - WB\_Radius) and 'Output Parameters' (P3 - WB\_Mass, P4 - WB\_Deformation, P5 - WB\_Stress, P6 - WB\_Sinus). The 'Metrics' section is highlighted, showing 'Goodness Of Fit' selected. On the right, the 'Goodness Of Fit' report table is displayed, showing various metrics for each output parameter. A right-click context menu is open over the 'P4 - WB\_Deformation' column, with the 'Generate Advanced Goodness of Fit Report' option highlighted.

	Name	P3 - WB_Mass	P4 - WB_Deformation	P5 - WB_Stress	P6 - WB_Sinus
1					
2	Goodness Of Fit				
3	Coefficient of Determination (Best Value = 1)	★★★ 1	★★★ 1	★★★ 1	★★★ 1
4	Adjusted Coeff of Determination (Best Value = 1)	★★★ 1	★★★ 1	★★★ 1	★★★ 1
5	Maximum Relative Residual (Best Value = 0%)	★★★ 0	★★★ 0	★★★ 0	★★★ 0
6	Root Mean Square Error (Best Value = 0)	4.0238E-09	5.7629E-16	5.3169E-12	2.0751E-15
7	Relative Root Mean Square Error (Best Value = 0%)	★★★ 0	★★★ 0	★★★ 0	★★★ 0
8	Relative Maximum Absolute Error (Best Value = 0%)	★★★ 0	★★★ 0	★★★ 0	★★★ 0
9	Relative Average Absolute Error (Best Value = 0%)	★★★ 0	★★★		

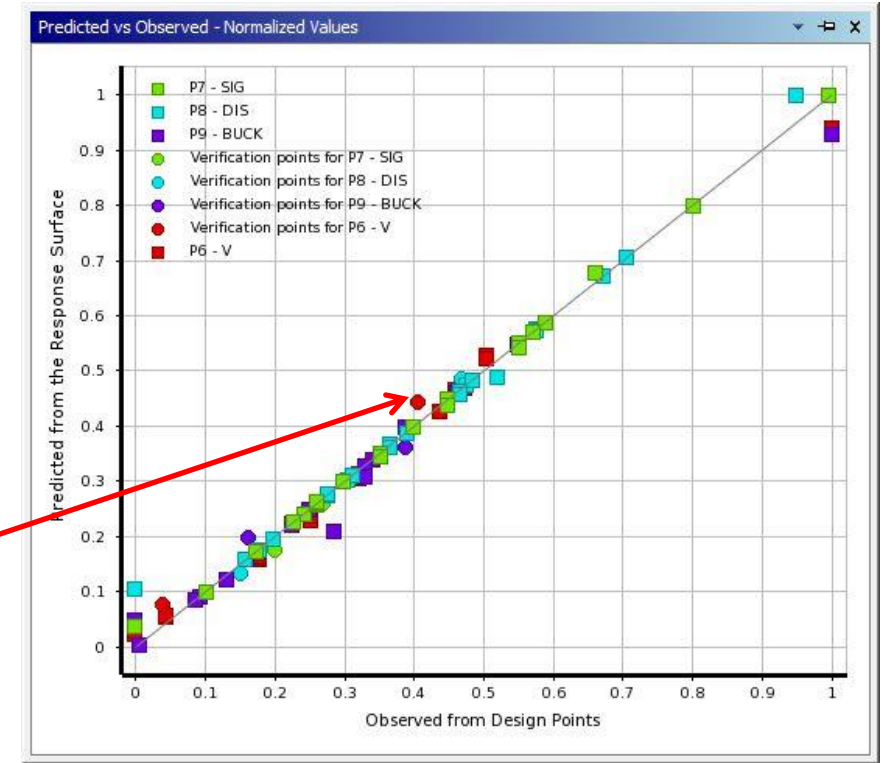
Context Menu Options:

- Copy
- Generate Advanced Goodness of Fit Report
- Export Data
- Expand All
- Collapse All

# Always check the response surface quality

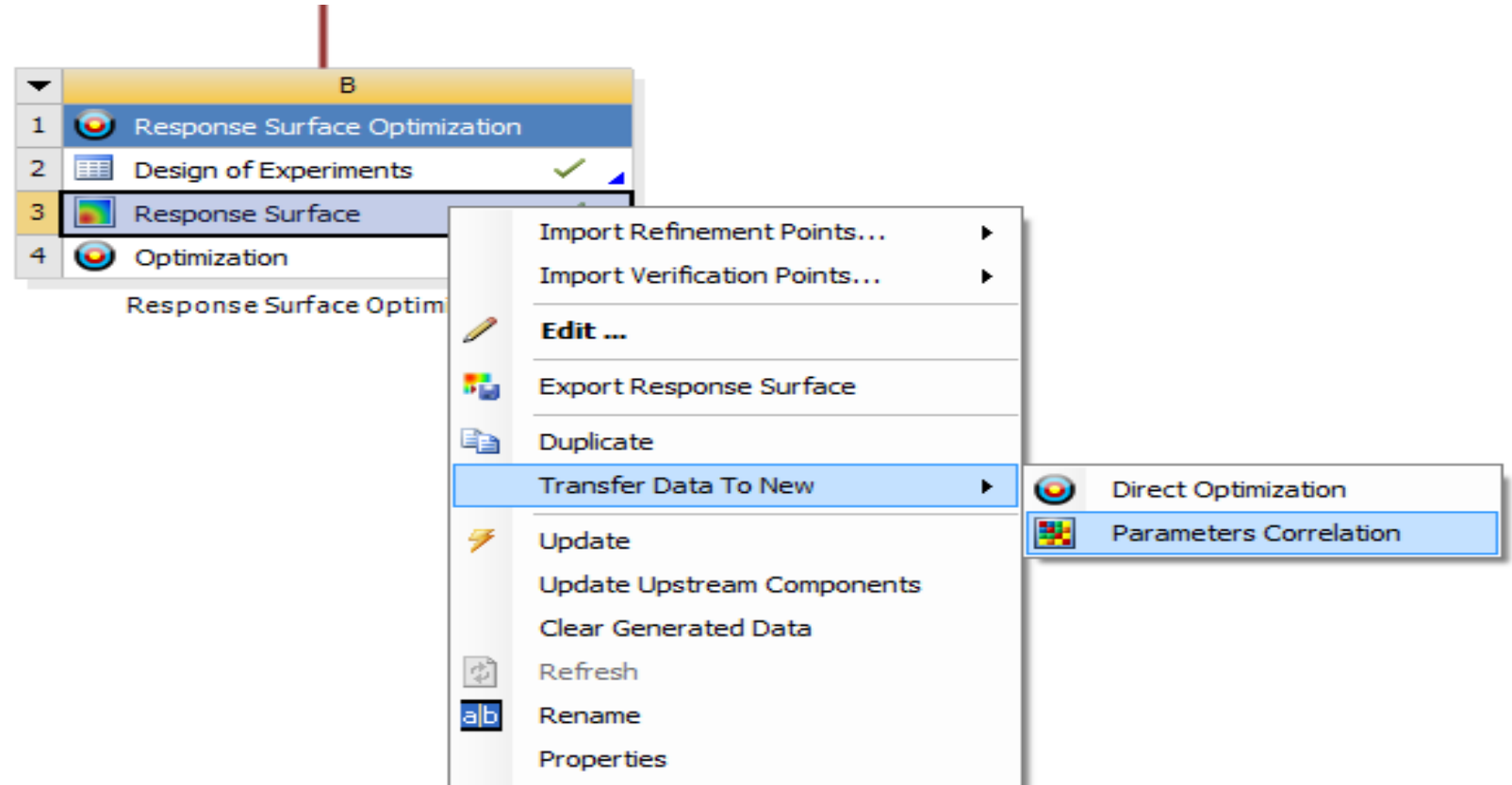
- To estimate the approximation accuracy on DOE points
- To check if output variations are explained by the variation of inputs (with verification points)
- Using Goodness of Fit metrics and associated chart
- Using verification points:
  - Automatically generated
  - Random and/or candidates
  - Manually defined
  - Not needed when GARS (cross validation) is used

Manual refinement: select a point to be appended to the DOE as a refinement point



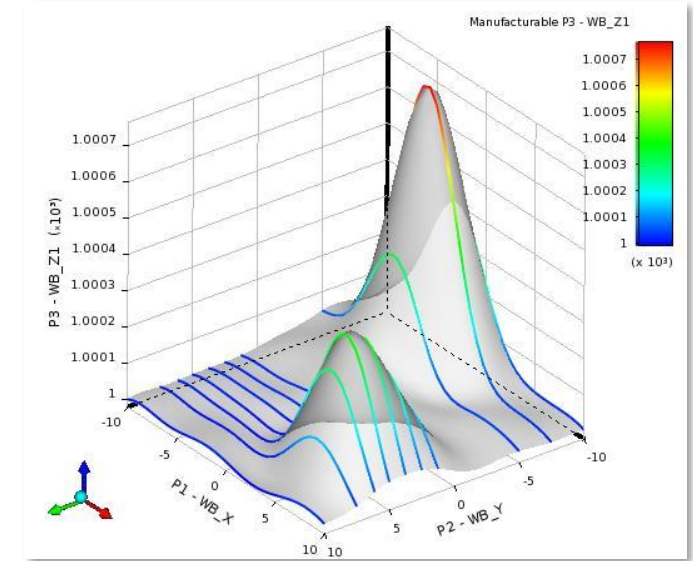
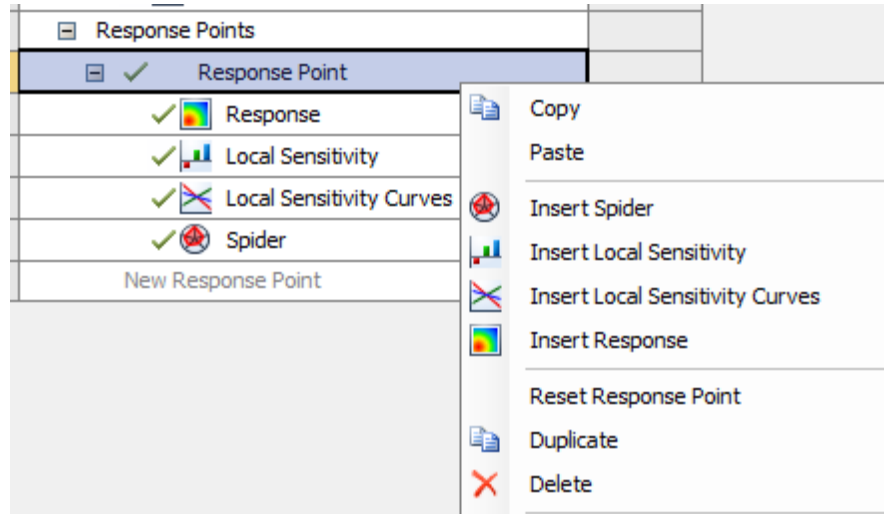
# Response Surface data to Parameters Correlation

The Parameter Correlation runs are done based on the Response Surface data and no real solve. Right click Response surface and export that data to Parameters Correlation



# Min-Max Search

- Prefer Manufacturable values to Discrete when possible  
=> more efficient and accurate, less limitations in DX features
- Check results of the Min-Max Search to estimate feasibility of the design specification
- Create multiple Response Points and charts to store interesting results and compare them

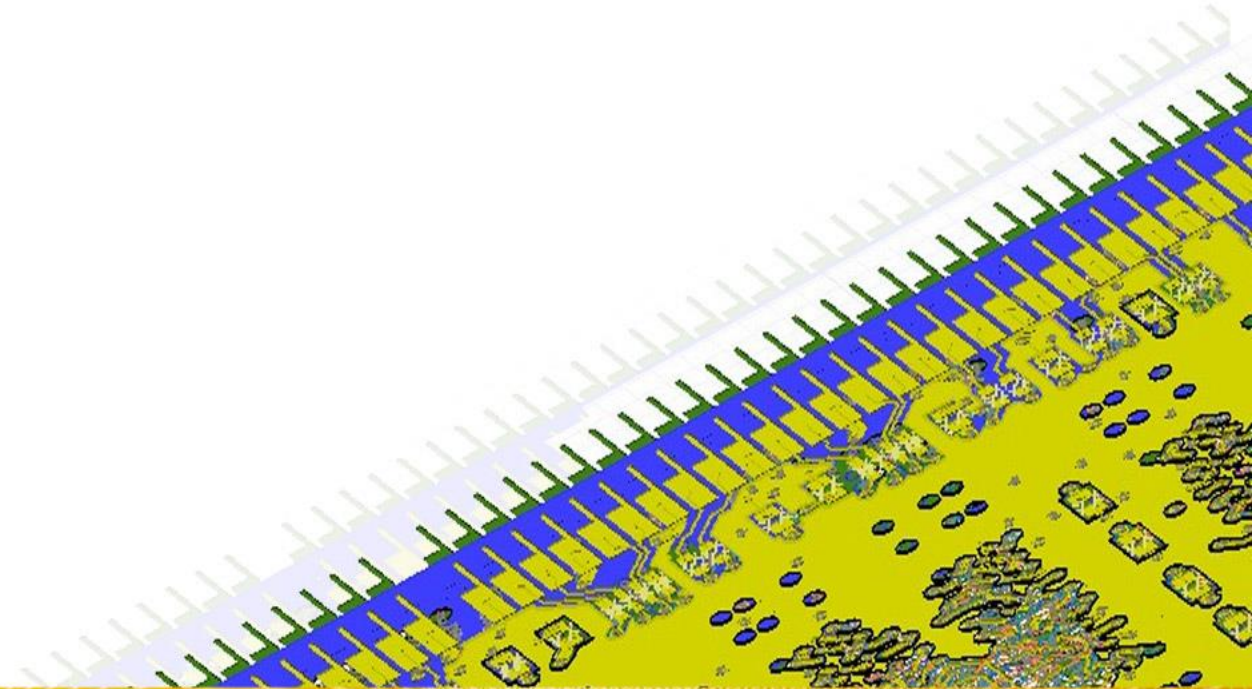


✓  Min-Max Search



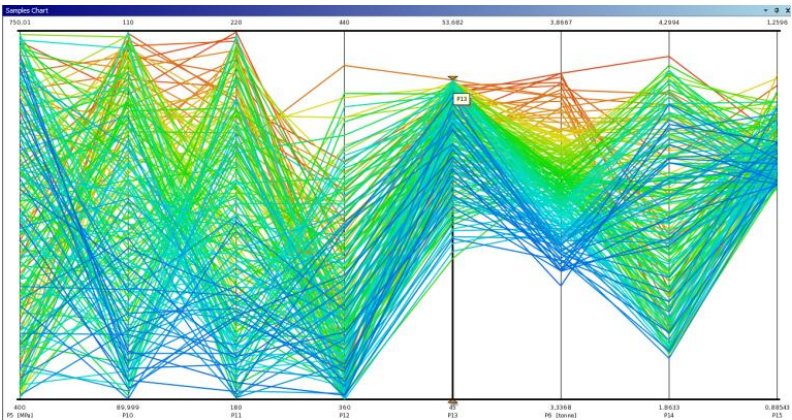


# Optimization



# Optimization

- Screening and manual exploration with parallel chart can provide you good insights
- For a single objective, pick a screening candidate as the starting point for gradient based optimization (NLPQL, MISQP).
- Verify candidates automatically, or manually, to measure approximation error. If possible, also launch a direct optimization using the best candidate as a starting point.



Candidate Point 1	Candidate Point 2	Candidate
2.0169	1.9282	2.084
5.1893	5.4542	5.184
92.593	95.501	93.90
926.99	905.48	903.3
1.8567E+05	1.8908E+05	1.9508E
★★★ 969.28	★★★ 1004.4	★★★ 1015
★★★ 9477.4		
★★★ 56.20		

Copy

Explore Response Surface at Point

Insert as Design Point

Insert as Refinement Point

Insert as Verification Point

Insert as Custom Candidate Point

Verify by Design Point Update

Candidate Point 1	Candidate Point 1 (verified)
2.0169	
5.1893	
92.593	
926.99	
1.8567E+05	
★★★ 969.28	★★★ 969.1
★★★ 9477.4	★★★ 9482
★★★ 56.204	★★★ 56.25

# Parameter Relationships in Optimization

- Parameter relationships give you greater flexibility in defining optimization limits on input parameters than the standard single-parameter constraints such as Greater Than, Less Than, and Inside Bounds. When you define parameter relationships, you can specify expression-based relationships between multiple input parameters, with the values remaining physically bounded and reflecting the constraints on the optimization problem.

Outline of Schematic C4: Optimization			
	A	B	C
1		Enabled	Monitoring
2	Optimization		
3	Objectives and Constraints		
4	Minimize P8		
5	Domain		
6	Microsoft Office Excel (A1)		
7	P1 - WB_B	<input checked="" type="checkbox"/>	
8	P2 - WB_D	<input checked="" type="checkbox"/>	
9	P3 - WB_L	<input checked="" type="checkbox"/>	
10	P4 - WB_P	<input checked="" type="checkbox"/>	
11	P5 - WB_E	<input checked="" type="checkbox"/>	
12	Parameter Relationships		
13	$2 * P1 \geq P2$	<input checked="" type="checkbox"/>	
14	$P1 + P2 \leq 0.3 [\text{in}]$	<input checked="" type="checkbox"/>	

Table of Schematic C4: Optimization				
	A	B	C	D
1	Input Parameters			
2	Name	Lower Bound	Upper Bound	Starting Value
3	P1 - WB_B (mm)	1.8	2.2	2
4	P2 - WB_D (mm)	4.5	5.5	5
5	P3 - WB_L (mm)	90	110	100
6	P4 - WB_P (N)	900	1100	1000
7	P5 - WB_E (MPa)	1.8E+05	2.2E+05	2E+05
8	Parameter Relationships			
9	Name	Left Expression	Operator	Right Expression
10	$2 * P1 \geq P2$	$2 * P1$	$\geq$	P2
11		4 [mm]		5 [mm]
12	$P1 + P2 \leq 0.3 [\text{in}]$	$P1 + P2$	$\leq$	0.3 [in]
13		7 [mm]		7.62 [mm]
*	New Parameter Relationship	New Expression	$\leq$	New Expression

# Direct Optimization from RSO ACT



mixingTank\_CCD - Workbench

File Edit View Tools Units Extensions Jobs Help

Import Parameters and DOE DOE from Correlation Parameter Sweep **Direct Optimization from RSO**

Project B4:Optimization C2:Optimization

Update Clear Generated Data Refresh

Outline of Schematic C2: Optimization

	A	B	C
1		Enabled	Monitoring
2	Optimization		
3	Objectives and Constraints		
4	Minimize P3		
5	Domain		
6	Fluid Flow (FLUENT) (A1)		
7	P1 - inletcold_velocity	<input checked="" type="checkbox"/>	
8	P2 - inlethot_velocity	<input checked="" type="checkbox"/>	
9	Parameter Relationships		
10	Results		

Table of Schematic C2: Optimization

	A	B	C	D
1	Input Parameters			
2	Name	Lower Bound	Upper Bound	Starting Value

Direct Optimization Settings

RSO System ID: RSO

Candidate of Interest: Candidate Point 1

New Parameter Bounds Definition: Focused on RSO Candidate

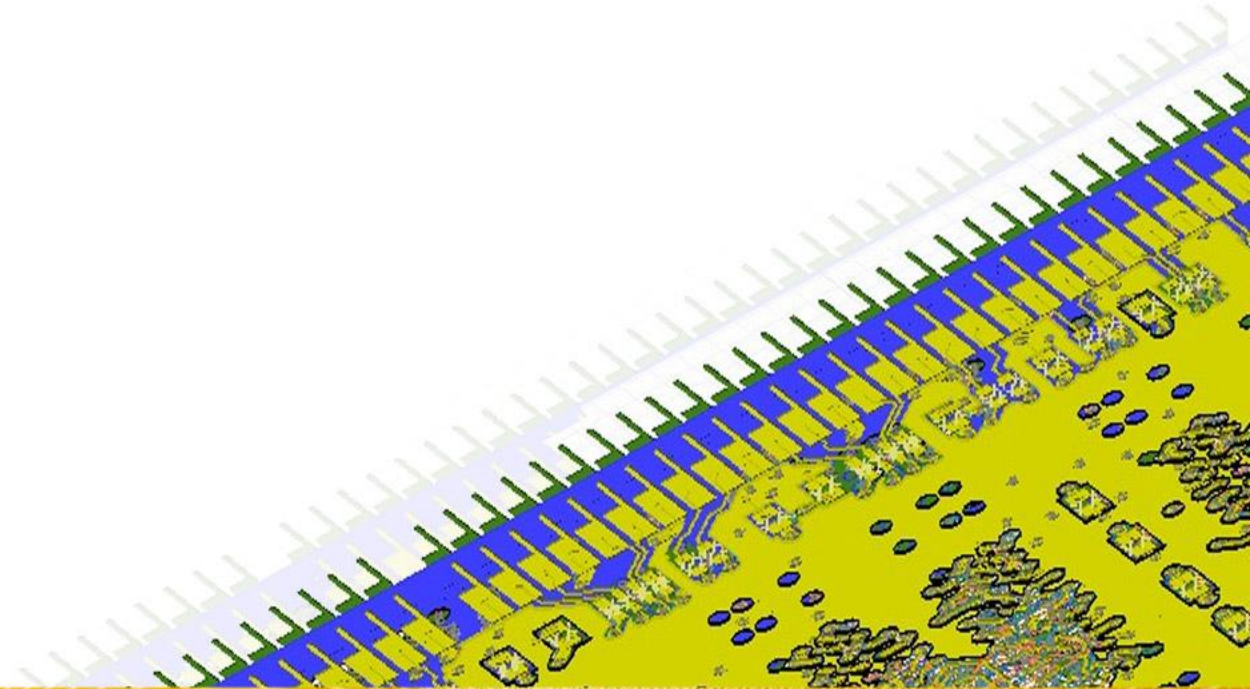
New Parameter Ranges (% of RSO range): 20

OK Cancel





## Miscellaneous



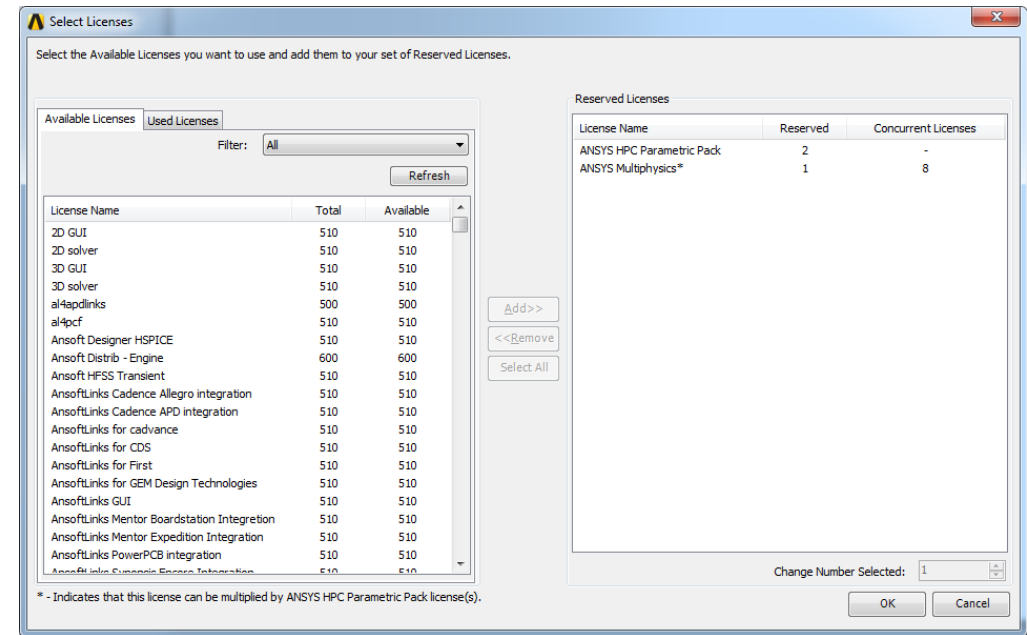
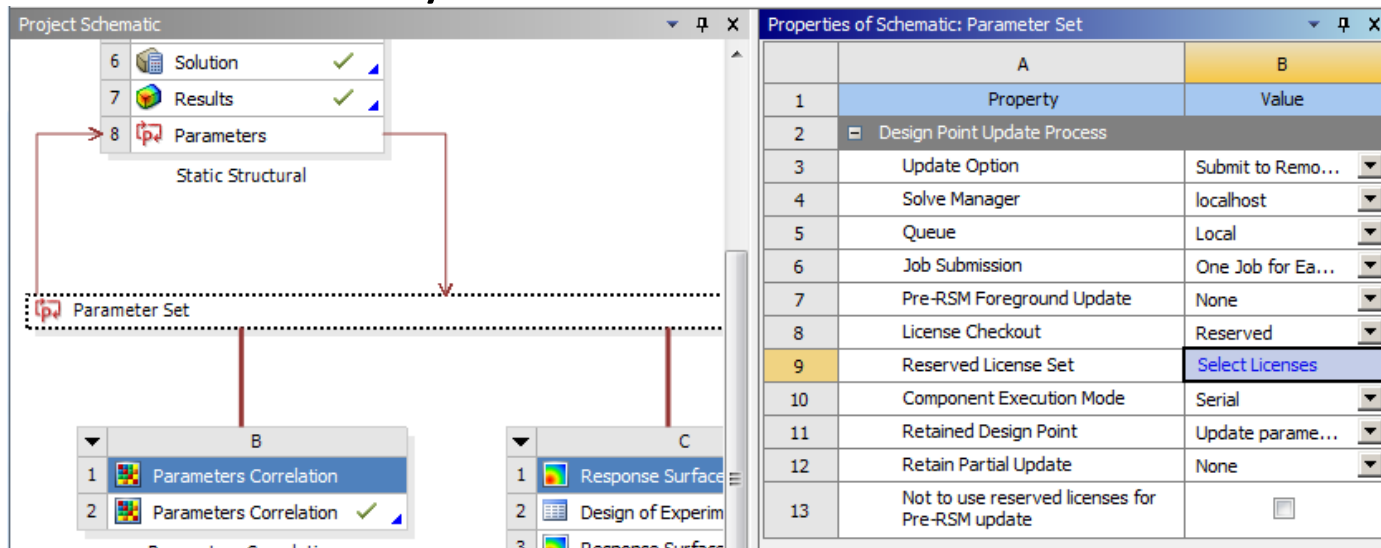


# Working with Design Points

- R16.0: if geometry is complicated, check the geometry by updating several design points (just some extremes or the entire DOE) using “Partial Update = Geometry”
- Update failure? Before opening the editor to investigate, copy the project (save as or export DP) to avoid invalidating parametric results.
- RSM: use the mode “One Job for Each Design Point”, so that several design points can be updated simultaneously
  - Look into HPC Parametric packs for licensing
- The file `user_files/DesignPoint.log` contains parameter values of all calculated DPs. It is a CSV file that you can import in a Custom DOE.

# Licences

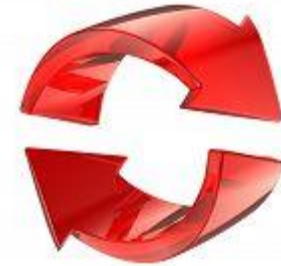
- Reserve licenses to avoid losing them between Design Points updates (common cause of failure)
- Use the HPC Parametric Pack to cost effectively update multiple design points simultaneously



# Failing design Points?

- Update of Design Points from DX
- Use the “Retry” option, to reduce DP failures.

Failed Design Points Management		
Number of Retries	3	
Retry Delay	120	s ▼



- Available in each DX component
- Some minor failures are just due to license server hiccups and a short pause is all that is needed before you can try again...

# DX Scripting

## DX Scripting

- Workbench platform actions are journaled
  - By default, available under %TEMP%\WorkbenchJournals
- All DX operations and data are journaled and accessible via commands and queries.
- Power of IronPython (.Net)

Design of Experiments	
Design of Experiments Type	Latin Hypercube Sampling Design
Samples Type	User-Defined Samples
Random Generator Seed	0
Number of Samples	30

```
dOEModel1 = designofExperiment1.GetModel()  
dOEModel1.MethodName = "Latin Hypercube Sampling Design"  
dOEModel1.SampType = "SFD_USER"  
dOEModel1.NumSamp = 30
```



## Questions & Answers

