

Workshop - Machine Learning - Structural Optimization of a 10 Bar Truss with MSC Nastran SOL 400

AN MSC NASTRAN MACHINE LEARNING WEB APP TUTORIAL

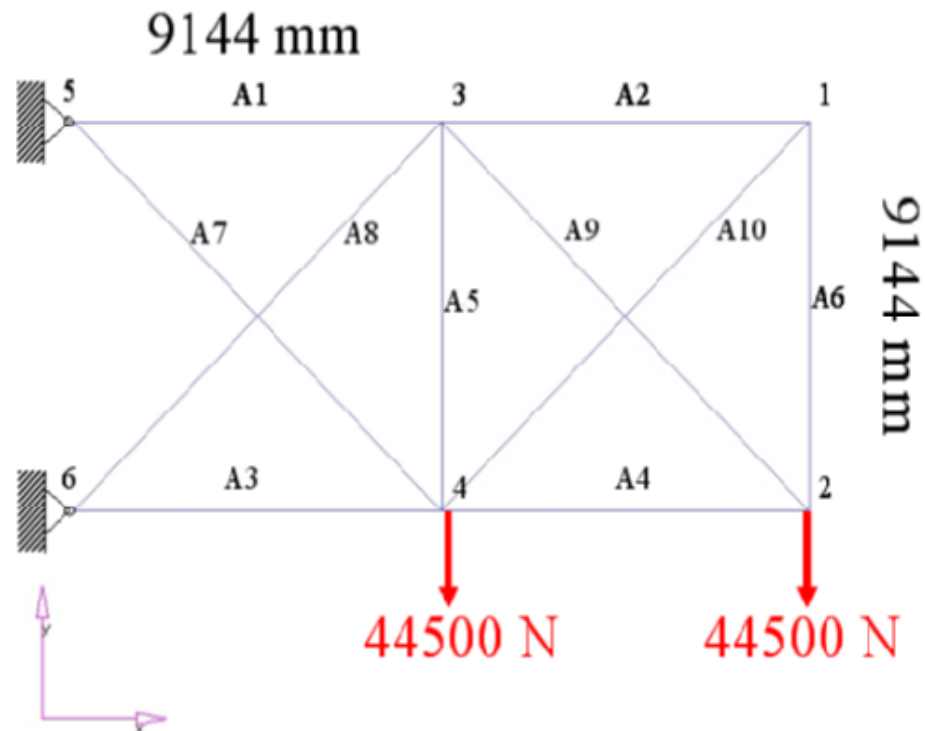
Goal: Use Machine Learning for Nonlinear Response Optimization

Before Optimization

- Weight: .7653

After Optimization

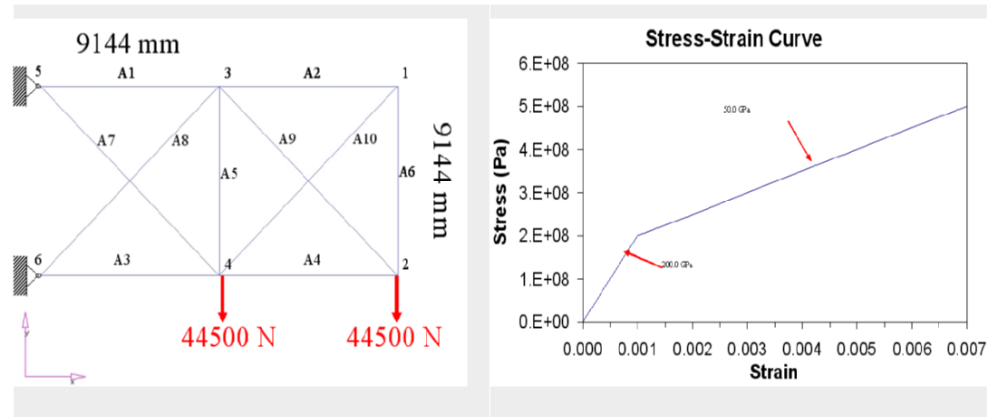
- Weight: Less than .6



Details of the Structural Model

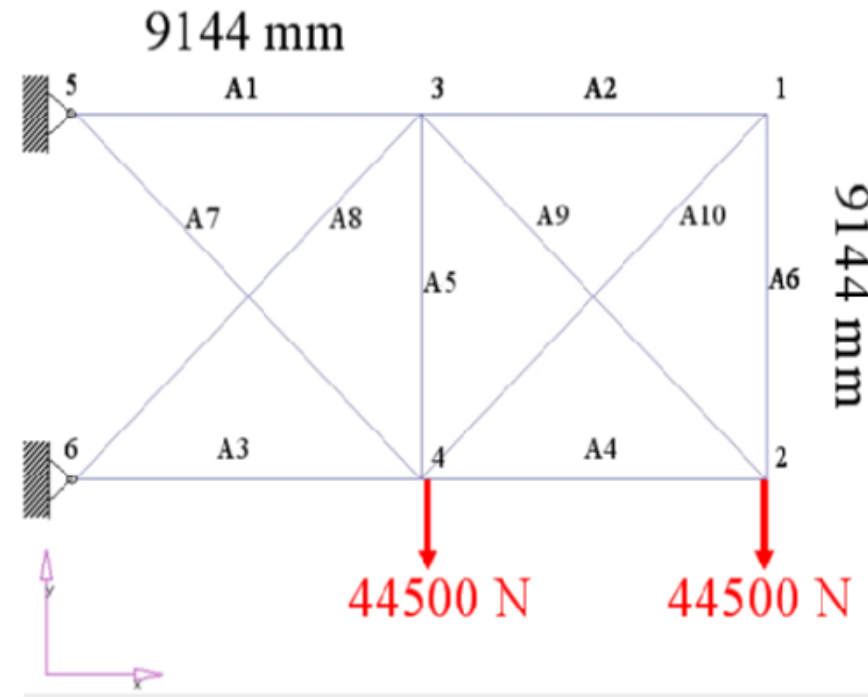
Examples

Example 1 10 Bar Truss (test library problem: deslo.dat)



(cross sectional areas)	
Find:	
Minimize:	Weight
	$ \sigma_j \leq 220 \text{ MPa} \quad (j = 1, \dots, 10)$
	$ \delta_{all} \leq 100.0 \text{ mm} \quad (\text{both x and y directions of all nodes})$
	$78.5 \text{ mm}^2 \leq X_i \leq 2826.0 \text{ mm}^2 \quad (i = 1, \dots, 10)$

MSC Nastran Design Sensitivity and Optimization User's Guide
Chapter 8 – Special Topics - Optimization of Nonlinear
Structural Responses

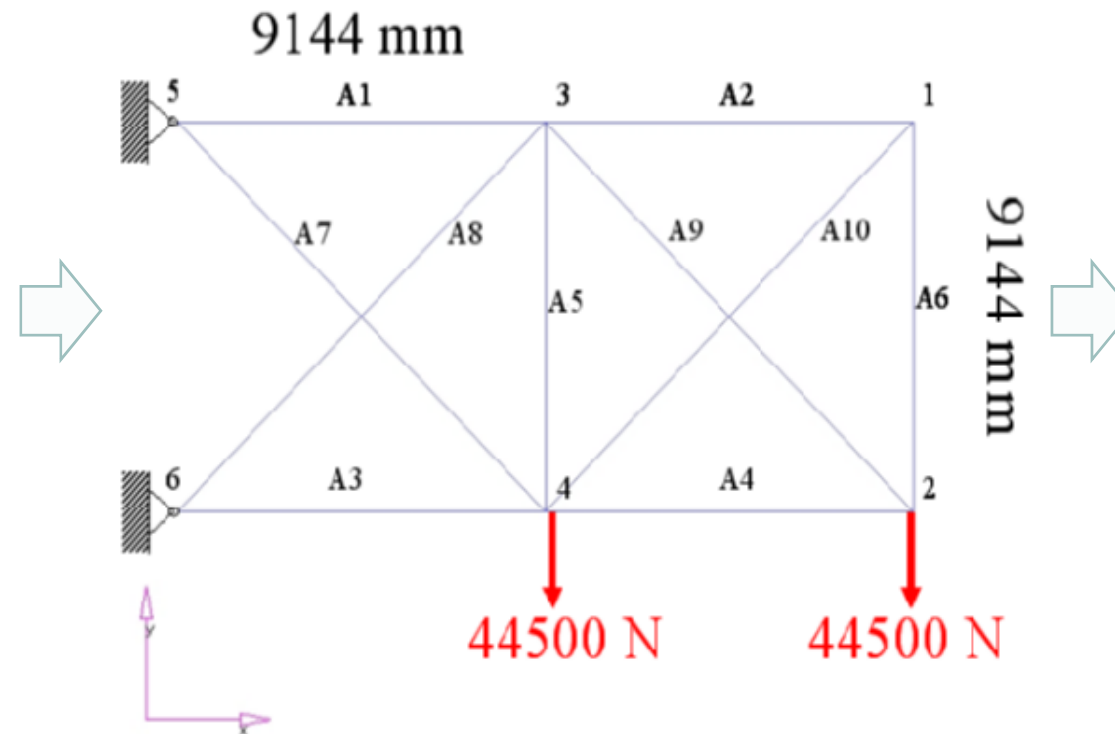


Optimization Problem Statement

Design Variables

x1: Area of PROD 1
x2: Area of PROD 2
x3: Area of PROD 3
x4: Area of PROD 4
x5: Area of PROD 5
x6: Area of PROD 6
x7: Area of PROD 7
x8: Area of PROD 8
x9: Area of PROD 9
x10: Area of PROD 10

$$78.5 < x_i < 2826.0$$



Design Objective

Minimize r1: Weight

Design Constraints

r2: Stress, axial, of PROD 1

r3: Stress, axial, of PROD 2

...

r11: Stress, axial, of PROD 10

...

r12: Displacement, x component, of grid 1

r13: Displacement, y component, of grid 1

...

$$-220 < r2, r3, \dots r11 < 220$$

$$-100 < r12, r13, \dots r19 < 100$$

Contact me

- Nastran SOL 200 training
- Nastran SOL 200 questions
- Structural or mechanical optimization questions
- Access to the SOL 200 Web App

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Tutorial

Tutorial Overview

1. Start with a .bdf and .h5 file
2. Use the SOL 200 Web App to:
 - Configure a Machine Learning task
 - Design Variables
 - Design Objective
 - Design Constraints
 - Perform optimization
3. Plot the Optimization Results

Special Topics Covered

Nonlinear Response Optimization - The existing optimization capability in MSC Nastran SOL 200 is mostly applicable to the linear solution sequences, e.g. SOL 101, 103, 105, etc. Optimization of nonlinear responses, such as those from SOL 400 or 700, pose many optimization challenges. This example demonstrates the use of machine learning to optimize nonlinear responses.

Training Data - The machine learning process utilizes regression models of the objective and constraints to determine likely optimal locations. Before the machine learning process begins, the regression models are first constructed by using training data. The training data consists of the objective and constrained responses at various points, or samples, throughout the design space. This example uses a Latin Hypercube design of 50 samples to generate the necessary initial training data.

SOL 200 Web App Capabilities

The Post-processor Web App and HDF5 Explorer are free to MSC Nastran users.

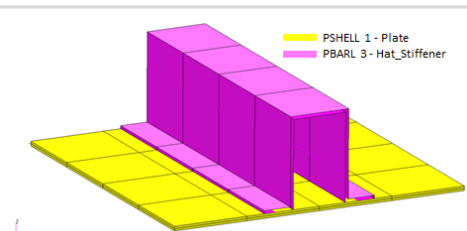
Compatibility

- Google Chrome, Mozilla Firefox or Microsoft Edge
- Windows and Red Hat Linux
- Installable on a company laptop, workstation or server. All data remains within your company.

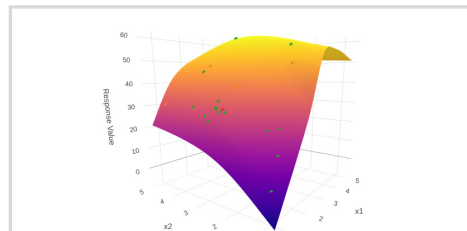
Web Apps

Benefits

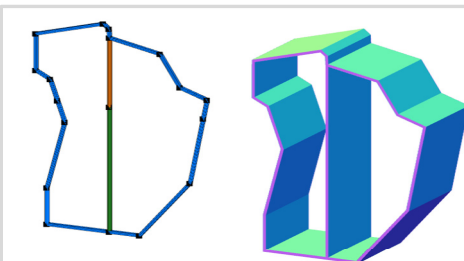
- REAL TIME error detection. 200+ error validations.
- REAL TIME creation of bulk data entries.
- Web browser accessible
- Free Post-processor web apps
- +80 tutorials



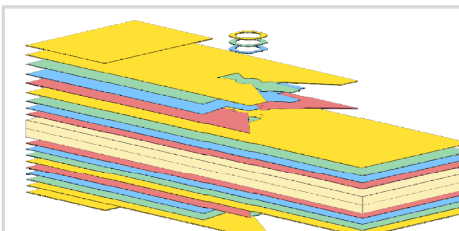
Web Apps for MSC Nastran SOL 200
Pre/post for MSC Nastran SOL 200.
Support for size, topology, topometry, topography, multi-model optimization.



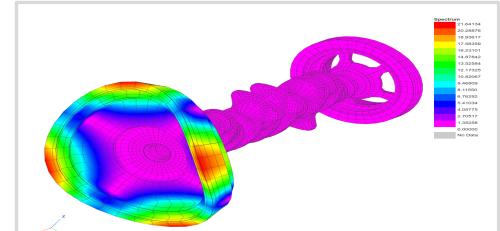
Machine Learning Web App
Bayesian Optimization for nonlinear response optimization (SOL 400)



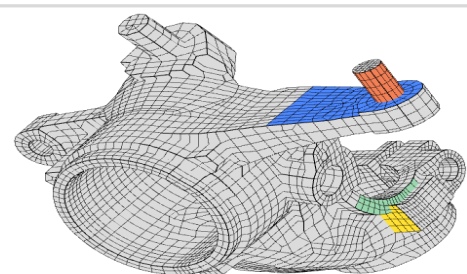
PBMSECT Web App
Generate PBMSECT and PBRSECT entries graphically



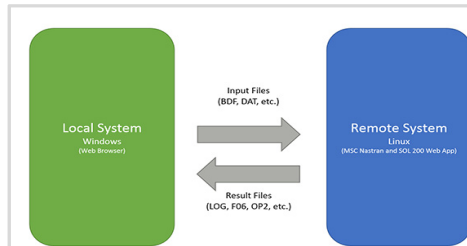
Ply Shape Optimization Web App
Optimize composite ply drop-off locations, and generate new PCOMPG entries



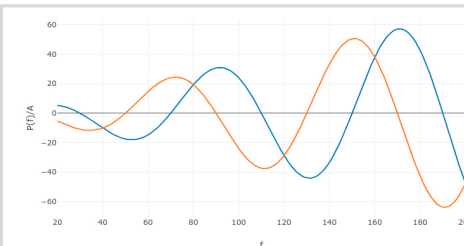
Post-processor Web App
View MSC Nastran results in a web browser on Windows and Linux



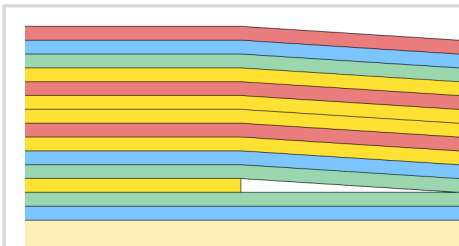
Shape Optimization Web App
Use a web application to configure and perform shape optimization.



Remote Execution Web App
Run MSC Nastran jobs on remote Linux or Windows systems available on the local network



Dynamic Loads Web App
Generate RLOAD1, RLOAD2 and DLOAD entries graphically



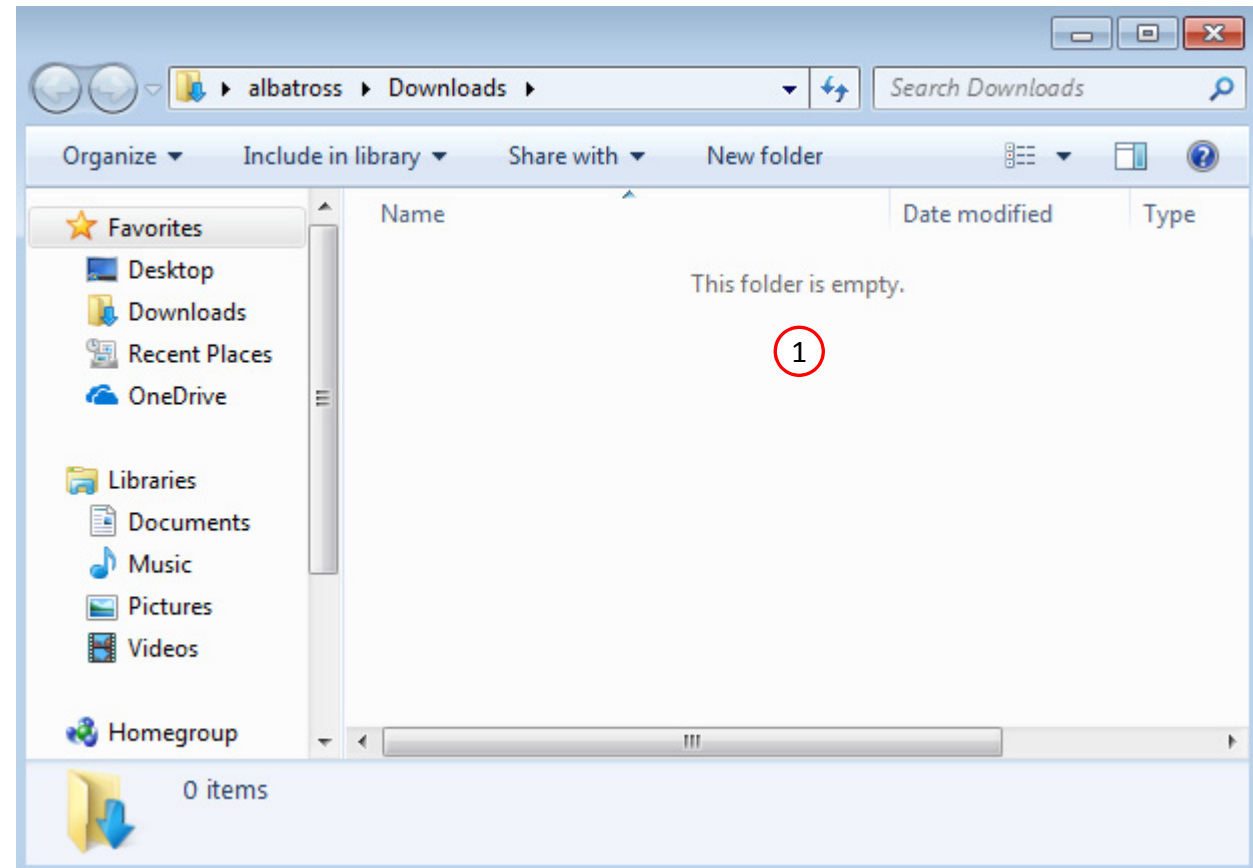
Stacking Sequence Web App
Optimize the stacking sequence of composite laminate plies



HDF5 Explorer Web App
Create graphs (XY plots) using data from the H5 file

Before Starting

1. Ensure the Downloads directory is empty in order to prevent confusion with other files



Go to the User's Guide

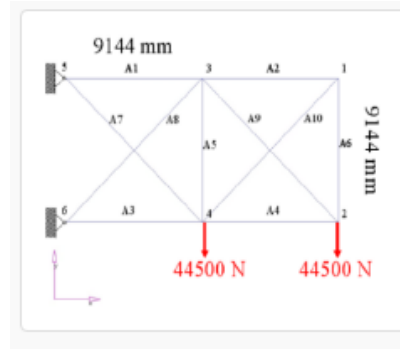
1. Click on the indicated link

- The necessary BDF files for this tutorial are available in the Tutorials section of the User's Guide.



Obtain Starting Files

1. Find the indicated example
2. Click Link
3. The starting file has been downloaded

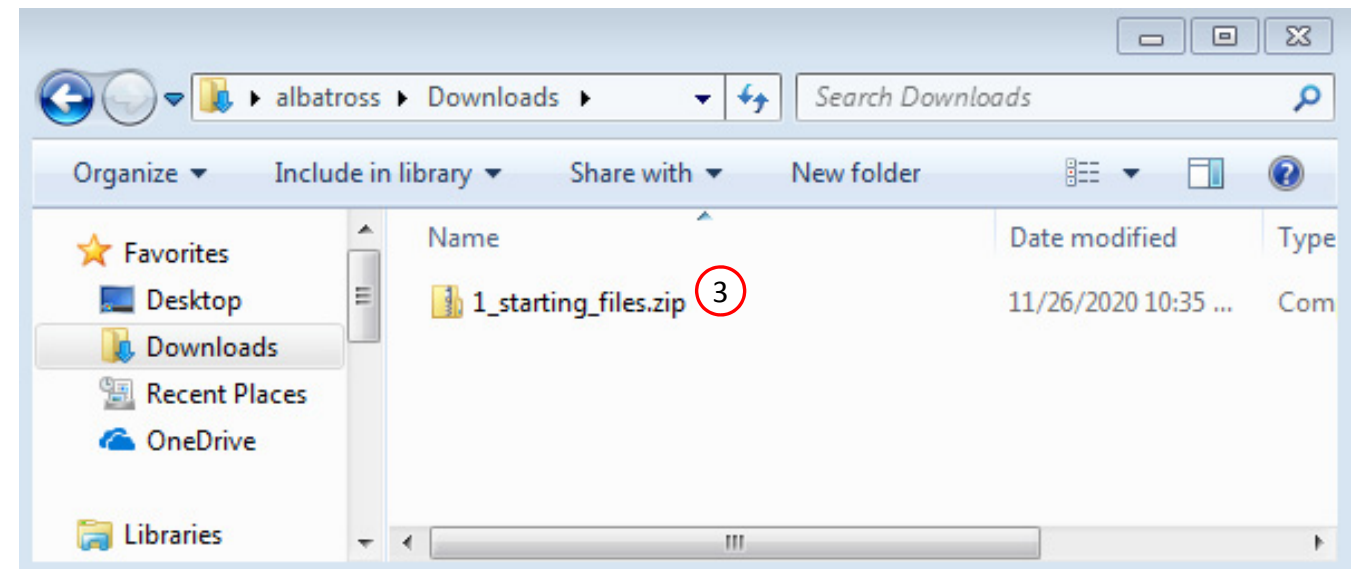


1

Machine Learning, Structural Optimization of a 10 Bar Truss

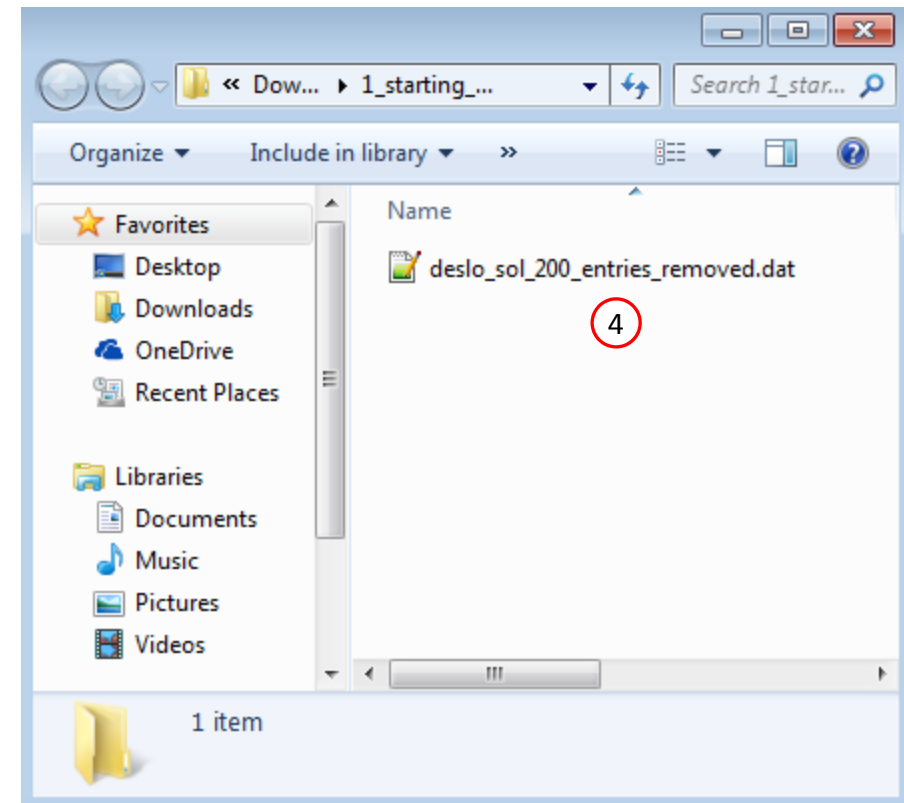
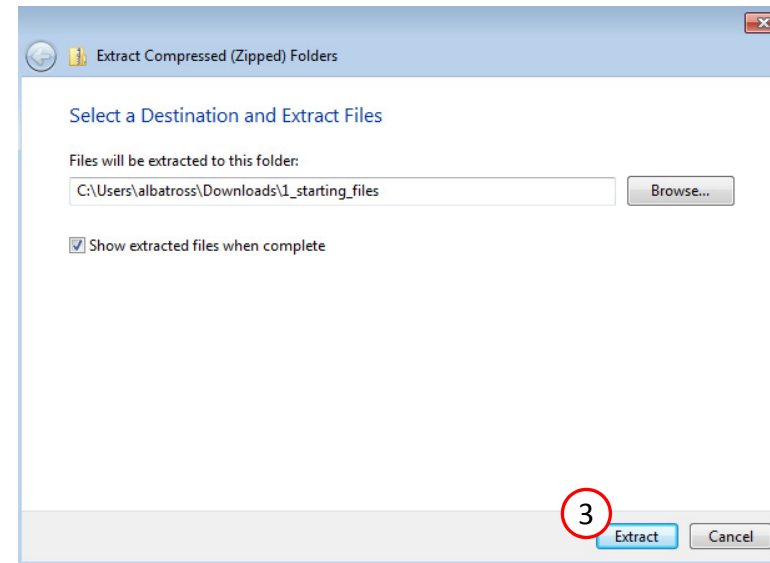
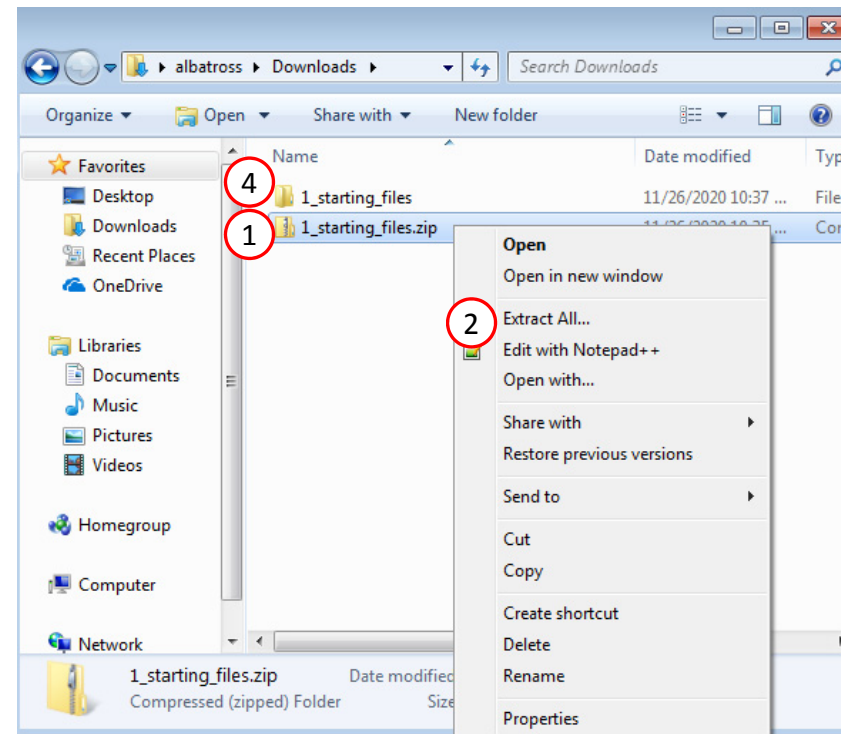
Machine learning methods are used to optimize a truss structure. MSC Nastran is used to evaluate the FE model. The design variables are the cross-sectional areas of the rod elements. The objective is to minimize the weight of the structure while constraining the axial stresses and displacements. Most machine learning examples are lower dimension problems with 1 to 5 parameters. This tutorial demonstrates a higher dimension scenario with 10 parameters.

Starting Files: [Link](#) 2
Solution BDF Files: [Link](#)



Obtain Starting Files

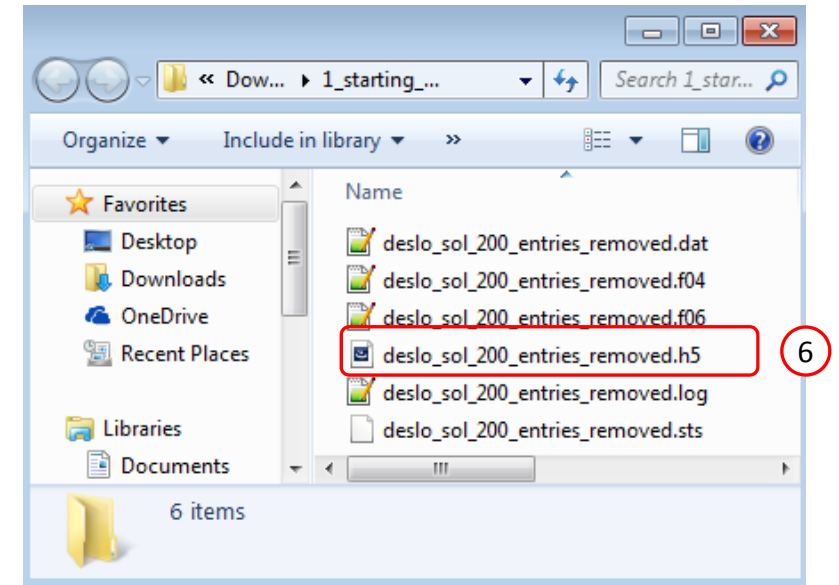
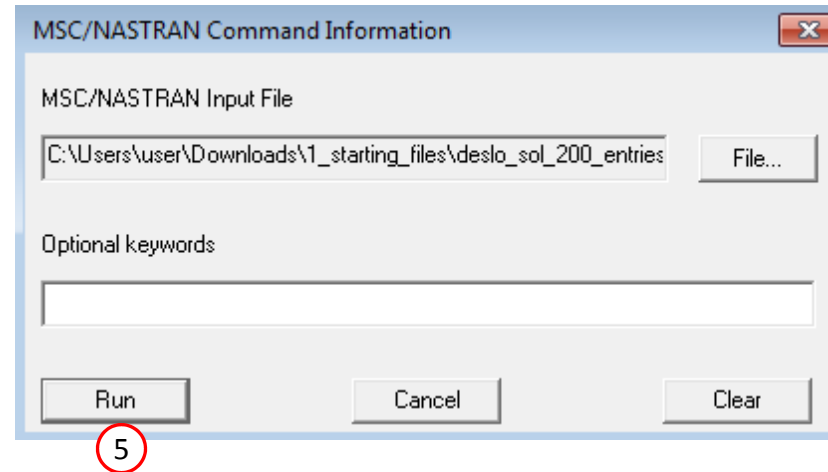
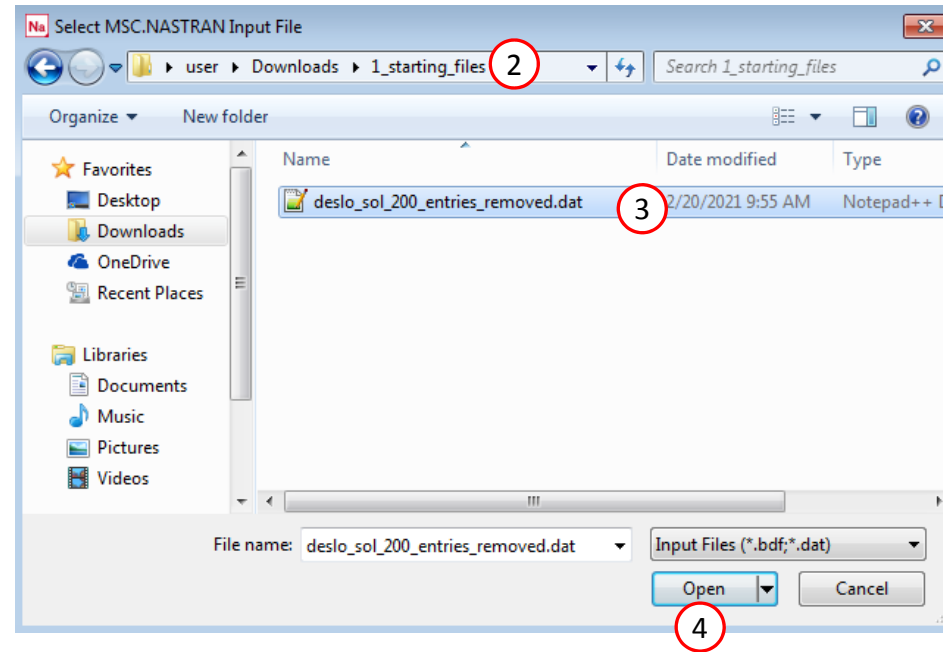
1. Right click on the zip file
2. Select Extract All...
3. Click Extract
4. The starting files are now available in a folder



Create the Starting H5 File

A starting H5 file must be created. This H5 file will be used to configure the responses later on.

1. Double click the MSC Nastran desktop shortcut
2. Navigate to the directory named 1_starting_files
3. Select the indicated file
4. Click Open
5. Click Run
6. The starting H5 file is created



Use the same MSC Nastran version throughout this exercise

The following applies if you have multiple versions of MSC Nastran installed.

To ensure compatibility, use the same MSC Nastran version throughout this exercise. For example, scenario 1 is OK but scenario 2 is NOT OK.

- Scenario 1 - OK
 - MSC Nastran 2021 is used to create the starting H5 file.
 - MSC Nastran 2021 is used for each run during Machine Learning or Parameter study.
- Scenario 2 – NOT OK
 - MSC Nastran 2018.2 is used to create the starting H5 file.
 - MSC Nastran 2021 is used for each run during Machine Learning or Parameter study.

Using the same MSC Nastran version is critical for consistent response extraction from the H5 file. A response configured for Nastran version X may not match in Nastran version Y, which leads to unsuccessful response extraction from the H5 files. The goal is to make sure all H5 files generated are from the same MSC Nastran version.

Open the Correct Page

1. Click on the indicated link

- MSC Nastran can perform many optimization types. The SOL 200 Web App includes dedicated web apps for the following:
 - Optimization for SOL 200 (Size, Topology, Topometry, Topography, Local Optimization, Sensitivity Analysis and Global Optimization)
 - Multi Model Optimization
 - Machine Learning
- The web app also features the HDF5 Explorer, a web application to extract results from the H5 file type.



Select BDF Files

1. Click Select files
2. Select the indicated file
3. Click Open
4. Click Upload files

- When starting the procedure, all the necessary BDF, or DAT, files must be collected and uploaded together. Relevant INCLUDE files must also be collected and uploaded.



Select BDF Files

1

1. Select files

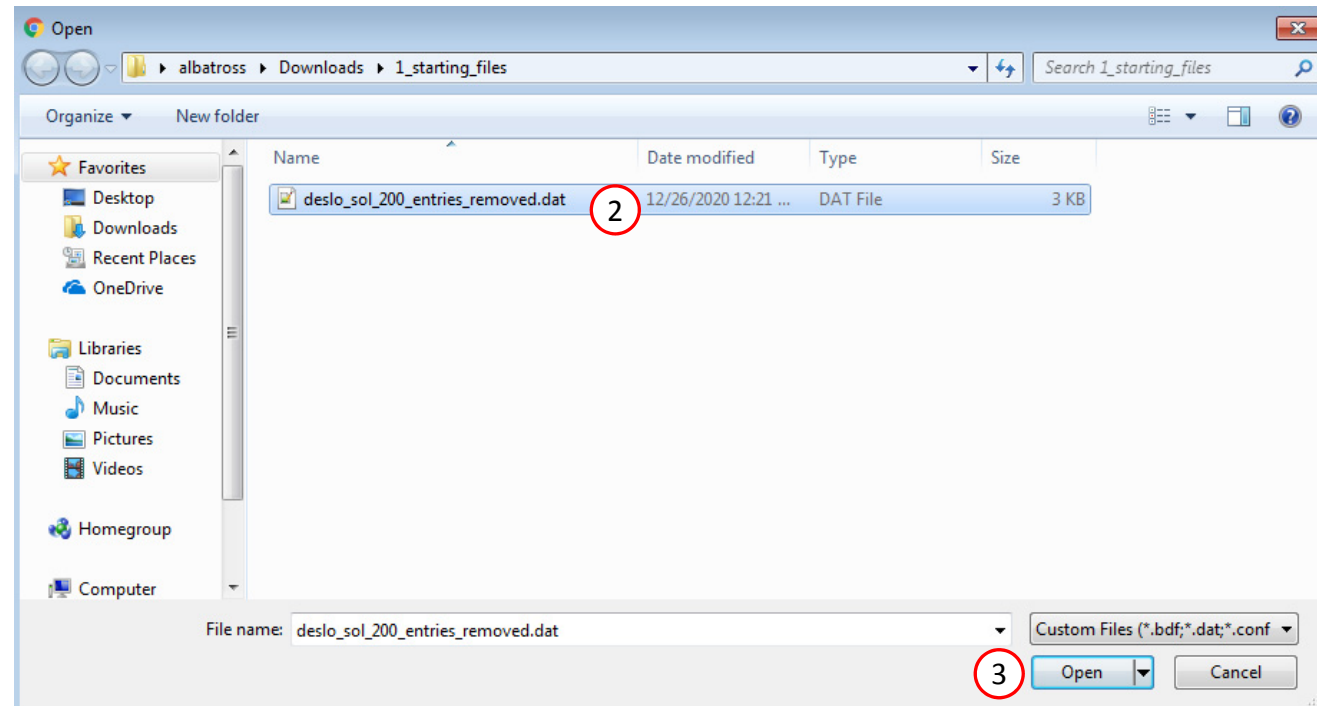
deslo_sol_200_entries_removed.dat

Inspecting: 100%

4

2. Upload files

Uploading: 100 %



Parameters

1. Set the following fields as parameters

- x1: Thickness, field 4, of PROD 1
- x2: Thickness, field 4, of PROD 2
- x3: Thickness, field 4, of PROD 2
- x4: Thickness, field 4, of PROD 2
- x5: Thickness, field 4, of PROD 2
- x6: Thickness, field 4, of PROD 2
- x7: Thickness, field 4, of PROD 2
- x8: Thickness, field 4, of PROD 2
- x9: Thickness, field 4, of PROD 2
- x10: Thickness, field 4, of PROD 2

2. Use the following bounds for all parameters

- Low: 78.5
- High: 2826.0

- Bulk data entries will always be displayed in the small field format.
- Only fields that have real or integer data entries may be selected as parameters. If the field is blank or contains only characters, the field may not be selected.

Select Parameters

```
$ _1 _2 _3 _4 _5 _6 _7 _8 _9 _10 _  
  
FORCE 300 2 4.45E4 0.0 -1.0 0.0  
FORCE 300 4 4.45E4 0.0 -1.0 0.0  
MAT1 1 2.E5 0.0 0.3 7.86E-09  
MATS1 1 1 PLASTIC 1 1 200.  
PROD 1 1 %x1% 0.e0 0.e0 0.e0  
PROD 2 1 %x2% 0.e0 0.e0 0.e0  
PROD 3 1 %x3% 0.e0 0.e0 0.e0  
PROD 4 1 %x4% 0.e0 0.e0 0.e0  
PROD 5 1 %x5% 0.e0 0.e0 0.e0  
PROD 6 1 %x6% 0.e0 0.e0 0.e0  
PROD 7 1 %x7% 0.e0 0.e0 0.e0  
PROD 8 1 %x8% 0.e0 0.e0 0.e0  
PROD 9 1 %x9% 0.e0 0.e0 0.e0  
PROD 10 1 %x10% 0.e0 0.e0 0.e0  
SPC 1 5 123 0.0  
SPC 1 6 123 0.0  
TABLES1 1  
0.0 0.0 0.001 200. .0126221500. ENDT
```



Configure Parameters

Delete	Parameter	Status	Low	High	Comments
<input type="checkbox"/>	x1	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x2	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x3	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x4	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x5	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x6	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x7	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x8	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x9	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR
<input type="checkbox"/>	x10	<input checked="" type="checkbox"/>	78.5	2826.0	Field 4 of PR

Samples

Configure 50 MSC Nastran runs at different values for the parameters

1. Click Samples
2. Set Design as Latin Hypercube, Reproducible
3. Set Number of Samples as 50
4. The table contains the 50 samples
5. The number of table rows to display or the page to view can be controlled with the indicated controls

- Before machine learning can be performed, a regression model must be constructed. This section configures the points that will be used to generate the training data (monitored responses) that is used to train the regression model.
- For example, the objective response from each sample is used to create a regression model for the objective.

SOL 200 Web App - Machine Learning

Parameters **Samples** Responses Download Results Connection Settings Home

Configure Samples

Design
Latin Hypercube, Reproducible

+ Info

Number of Samples
50

Samples to Run

+ Options

	Parameters			
Sample Number	x1	x2	x3	x4
1	1389.3	1938.5	2003.1	1139.7
2	1540.7	1572.9	1157.4	1411.1
3	1488.	1432.8	2627.1	2104.3
4	914.61	980.5	1535.6	735.08
5	1597.5	2034.	862.25	2191.9

« 1 2 3 4 5 6 7 ... 10 »

5 10 20 30 40 50

Responses

1. Click Responses
2. Click Select files
3. Select the indicated file
4. Click Open
5. Click Upload files

- On this page, the H5 file is uploaded to the web app.

1

Upload .h5 File

2

1. Select files

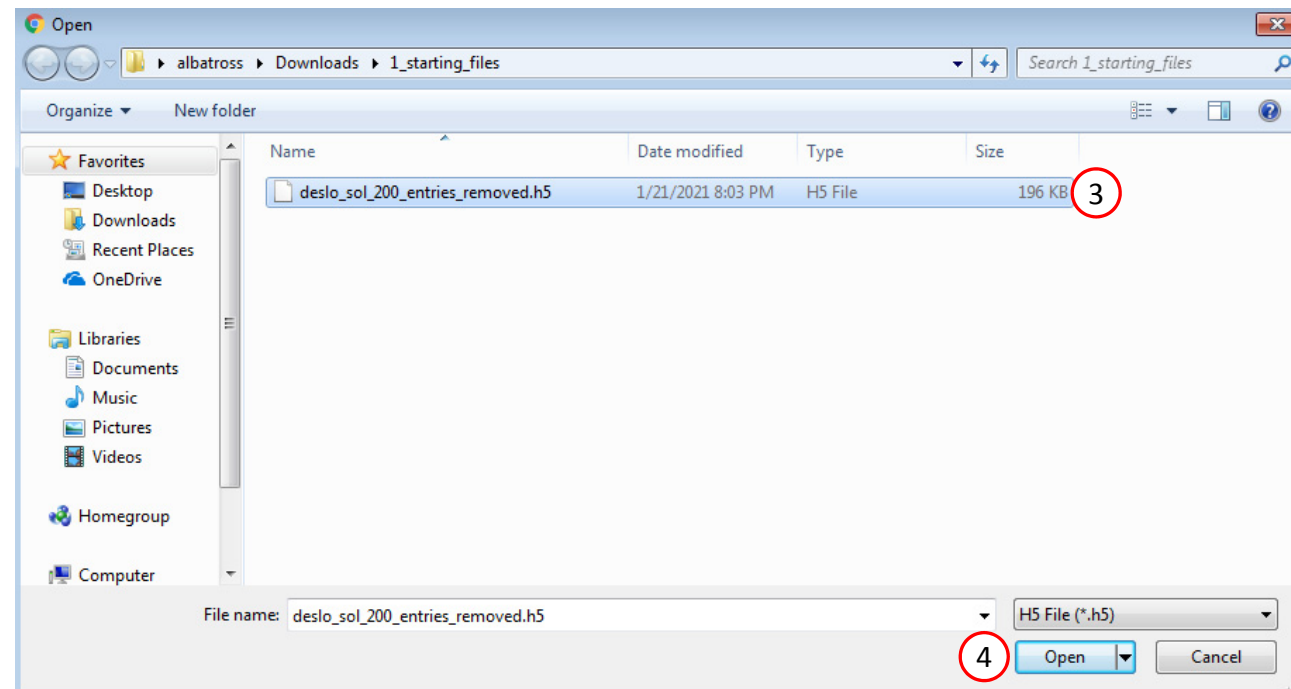
deslo_sol_200_entries_removed.h5

5

2. Upload files

Uploading

Loading



Adjust the Column Width

1. Optional - Use at your liking the buttons at the top right hand corner to adjust the width of the left and right columns

SOL 200 Web App - Machine Learning Parameters Samples Responses Download Results Connection Settings Home

Select Responses to Monitor

Session ID: 3710 HDF5

Select Dataset

- NODAL/ACCELERATION
- NODAL/DISPLACEMENT
- NODAL/GRID_WEIGHT
- NODAL/VELOCITY_CPLX
- SUMMARY/EIGENVALUE

Acquired Dataset

NODAL/GRID_WEIGHT - 0

SAMPLE	DOMAIN_ID	SUBCASE	STEP	ANA
--------	-----------	---------	------	-----

Reset Filters

View Responses to Monitor

Monitored Responses

Hide/Show Columns Reset Filters Download CSV

Delete	Label	Status	Objective	Lower Bound	Upper Bound	Monitor th of the Fil cycle (SC
	r1			Lower	Upper	

SOL 200 Web App - Machine Learning Parameters Samples Responses Download Results Connection Settings Home

Select Responses to Monitor

Session ID: 3710 HDF5

Select Dataset

- NODAL/ACCELERATION_CPLX
- NODAL/DISPLACEMENT_CPLX
- NODAL/GRID_WEIGHT
- NODAL/VELOCITY_CPLX
- SUMMARY/EIGENVALUE

Specify Entities

0

(ID)
Examples: 0, etc.

Acquired Dataset

NODAL/GRID_WEIGHT - 0

SAMPLE	DOMAIN_ID	SUBCASE	STEP	ANALYSIS	TIME_FREQ_EIGR
--------	-----------	---------	------	----------	----------------

Reset Filters

ame of H5 File**

Subcase number

Step number

Analysis type

Time, frequency or real part of eigen value

Im
eig
api

View Responses to Monitor

Monitored Responses

Hide/Show Columns Reset Filters Download CSV

Delete	Label	Status	Objective	L B
	r1			Lc

5 10 20 30 50 100

Select Responses

1. Select the following dataset:
NODAL/GRID_WEIGHT
2. Select the indicated cell
3. The newly created Response to Monitor is listed as r1
4. Set this response to Objective: MIN

- Any cell that includes a single decimal point can be set as a response to monitor.
- For this example, cells in the MX column can be selected. Cells in the MO and S column cannot be selected because the cells contain bracket characters ([and]).

Select Responses to Monitor

Session ID: 7231

HDF5

Select Dataset

ELEMENTAL/STRESS/ROD
ELEMENTAL/STRESS/ROD_NL
NODALDISPLACEMENT
NODAL/GRID_WEIGHT

1

Specify Entities

0

(ID)

Examples: 0, etc.

☒ Auto Execute

Acquire Dataset

Acquisition complete and successful

Acquired Dataset

NODAL/GRID_WEIGHT - 0

ID	MO	S	MX
0	[0.26306906...	[1,0,0,0,1,0,...	0.26306906...

2

View Responses to Monitor

Monitored Responses

Hide/Show Columns Reset Filters Download CSV

Delete	Label	Status	Objective	Lower Bound	Upper Bound	Monitor the response of the FINAL design cycle (SOL 200 only)	Mor r
3	r1		4 MIN	Lower	Upper		

Select Responses

1. Select the following dataset:
ELEMENTAL/STRESS/ROD
2. Select the indicated cells
3. The newly created Response to Monitor is listed as r2-r11
4. Use the following values to specify the constraints
 - Lower Bound: -220.
 - Upper Bound: 220.

Select Responses to Monitor

Session ID: 7231

HDF5

Select Dataset

ELEMENTAL/STRESS/ROD

ELEMENTAL/STRESS/ROD_NL

NODAL/DISPLACEMENT

NODAL/GRID_WEIGHT

Specify Entities

1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Element identification number (EID)
Examples: 1, 2, 3, etc.

☒ Auto Execute

Acquire Dataset

Acquisition complete and successful

Acquired Dataset

ELEMENTAL/STRESS/ROD - 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

EID	A	MSA	T
Element identification number	Axial stress	Axial Safety Margin*	Total stress
1	280.427640...	5e-324	0
2	56.3053243...	5e-324	0
3	-283.986703...	5e-324	0
4	-84.0714028...	5e-324	0
5	53.5135181...	5e-324	0
6	56.3599789...	5e-324	0
7	201.026683...	5e-324	0
8	-196.942249...	5e-324	0
9	120.698688...	5e-324	0
10	-79.7015753...	5e-324	0

View Responses to Monitor

Monitored Responses

Hide/Show Columns Reset Filters Download CSV

Delete	Label	Status	Objective	Lower Bound	Upper Bound	Monitor the response of the FINAL design cycle (SOL 200 only)
<input checked="" type="checkbox"/>	r1	<input checked="" type="checkbox"/>	MIN	Lower	Upper	
<input checked="" type="checkbox"/>	r2	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r3	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r4	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r5	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r6	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r7	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r8	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r9	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r10	<input checked="" type="checkbox"/>		-220.	220.	
<input checked="" type="checkbox"/>	r11	<input checked="" type="checkbox"/>		-220.	220.	

Select Responses

1. Select the following dataset:
NODAL/DISPLACEMENT
2. Select the indicated cells
3. Select the indicated cells
4. Adjust the scroll bar to view the new responses
5. The newly created Response to Monitor is listed as r12-r19
6. Use the following values to specify the constraints
 - Lower Bound: -100.
 - Upper Bound: 100.

Select Responses to Monitor

Session ID: 151

Select Dataset

ELEMENTAL/STRESS/ROD
ELEMENTAL/STRESS/ROD_N
NODAL/DISPLACEMENT
NODAL/GRID_WEIGHT

1

Specify Entities

1, 2, 3, 4, 5, 6

Grid identifier (ID)

Examples: 1, 2, 3, etc.

☒ Auto Execute

Acquire Dataset

Acquisition complete and successful

Acquired Dataset

NODAL/DISPLACEMENT - 1, 2, 3, 4, 5, 6

ID	X	Y	Z
Grid identifier	X component	Y component	Z component
1			
2			
3			
4			
5			
6			

2

3

5

4

6

Settings

1. Click Settings
2. Set Procedure to Machine Learning
3. Set n_iter to 50
4. Set acquisition_function_objective to Probability of Improvement

1



Settings

Procedure

Machine Learning

2

Advanced Settings

Setting	Description	Configure
Bayesian Optimization		
n_iter	This is the number of machine learning iterations. The total number of MSC Nastran runs is the sum of number of samples and n_iter. (Default = 20)	50
Acquisition Function		
acquisition_function_objective	Acquisition function to use for the objective (Default = Expected Improvement)	Probability of Improvement

3

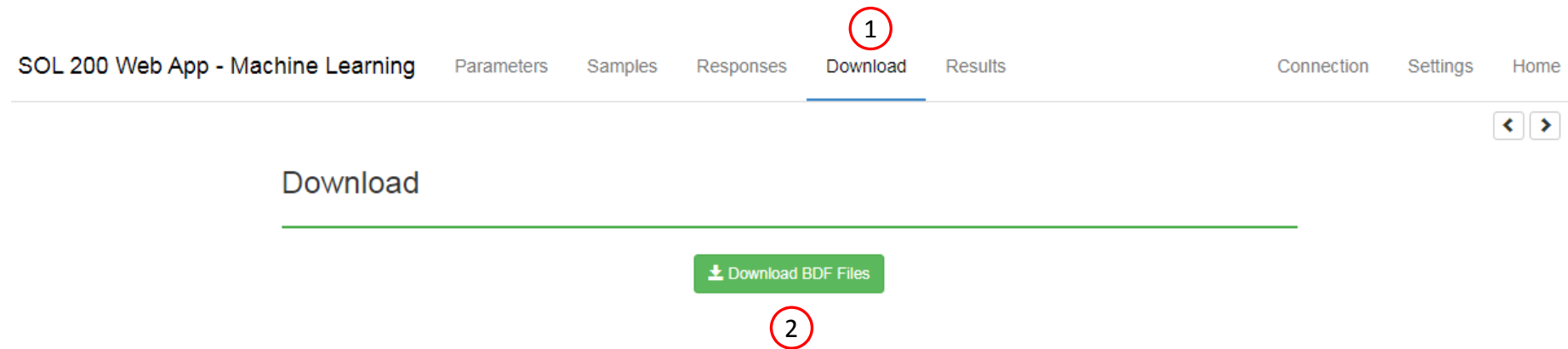
4

Settings Output

```
===== SETTINGS OUTPUT =====  
procedure,n_iter,optimize_restarts,nsamples,acquisition_function_objective,sta  
machineLearning,50,10,200,pi,incomplete  
=====
```


Download

1. Click Download
2. Click Download BDF Files

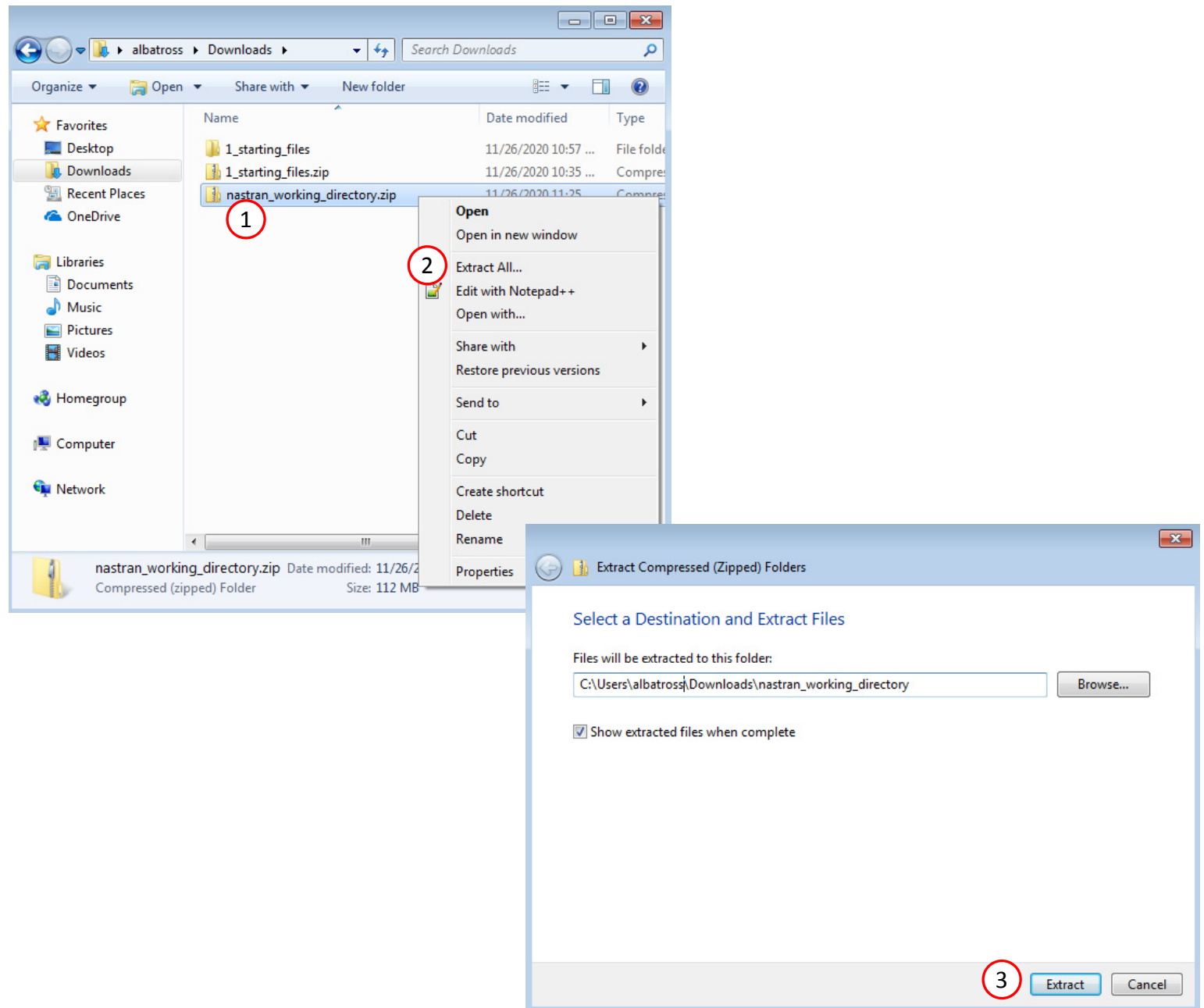


Start MSC Nastran

A new .zip file has been downloaded

1. Right click on the file
2. Click Extract All
3. Click Extract on the following window

- Always extract the contents of the ZIP file to a new, empty folder.



Start Desktop App

1. Inside of the new folder, double click on Start Desktop App
2. Click Open, Run or Allow Access on any subsequent windows
3. The Desktop App will now start

- One can run the Nastran job on a remote machine as follows:
 - 1) Copy the BDF files and the INCLUDE files to a remote machine.
 - 2) Run the MSC Nastran job on the remote machine.
 - 3) After completion, copy the BDF, F06, LOG, H5 files to the local machine.
 - 4) Click "Start Desktop App" to display the results.

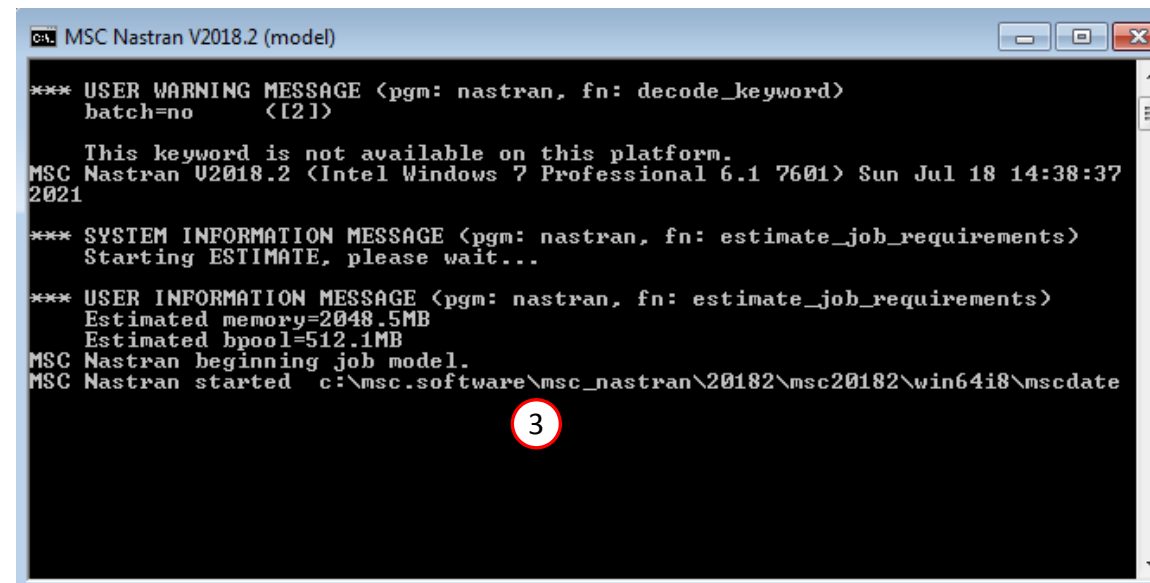
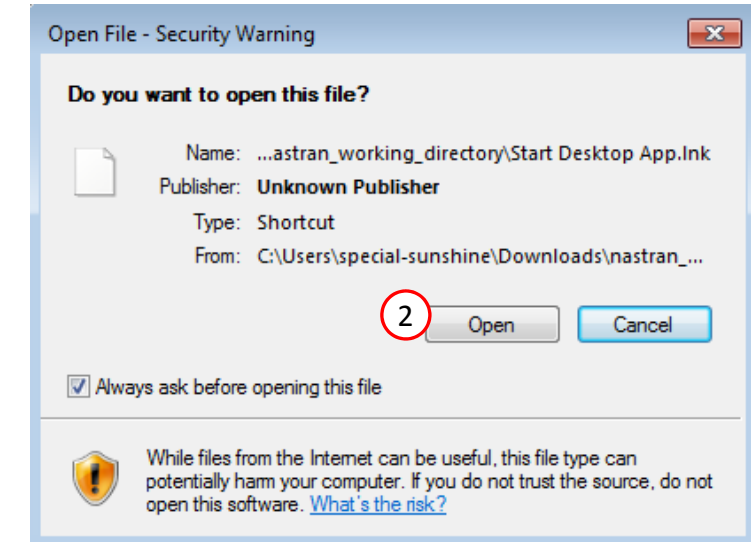
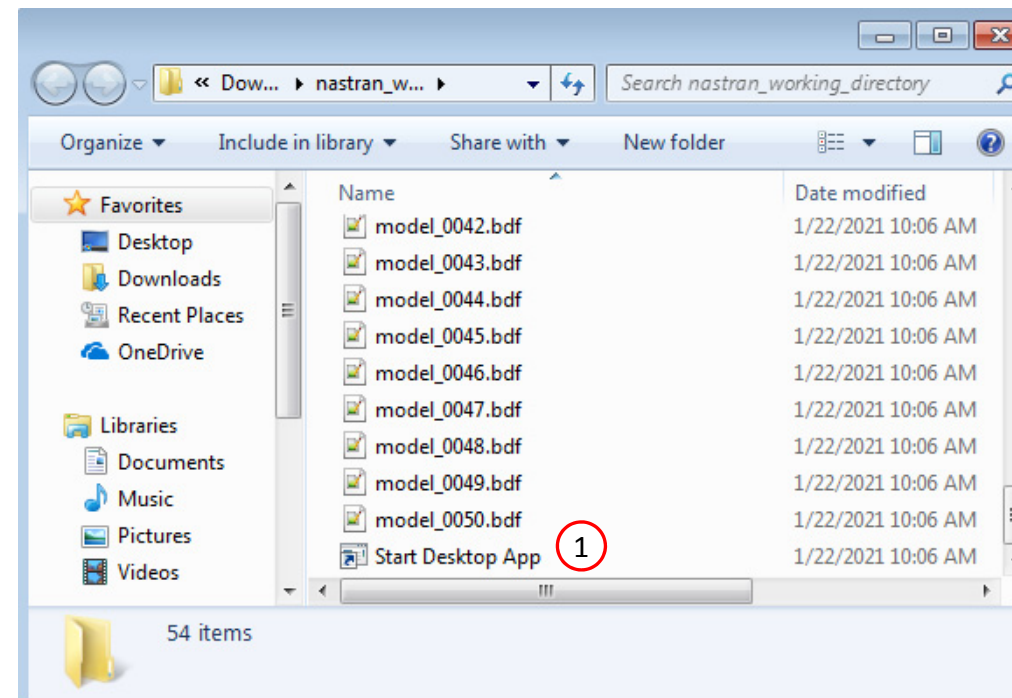
Using Linux?

Follow these instructions:

- 1) Open Terminal
- 2) Navigate to the nastran_working_directory
`cd ./nastran_working_directory`
- 3) Use this command to start the process
`./Start_MSC_Nastran.sh`

In some instances, execute permission must be granted to the directory. Use this command. This command assumes you are one folder level up.

```
sudo chmod -R u+x ./nastran_working_directory
```



Status

- While MSC Nastran is running, a status page will show the current state of MSC Nastran

SOL 200 Web App - Status

 Python

 MSC Nastran


Status

Name	Status of Job	Design Cycle	RUN TERMINATED DUE TO
model.bdf	Running	None	

Machine Learning Results

- During execution of the Desktop App, you will get a progress update regarding the best feasible design

Session ID:
60629

 In progress

Upload .csv File

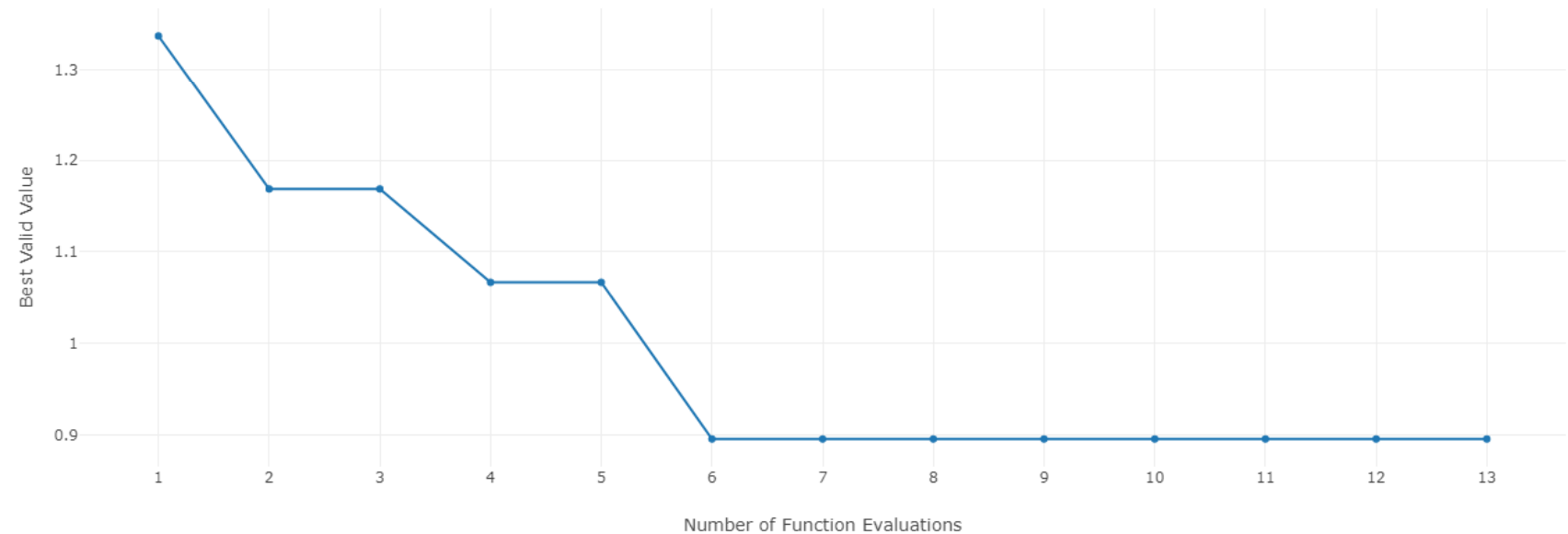
CSV Import

 Select files

a_tmp_best_valid_value.csv

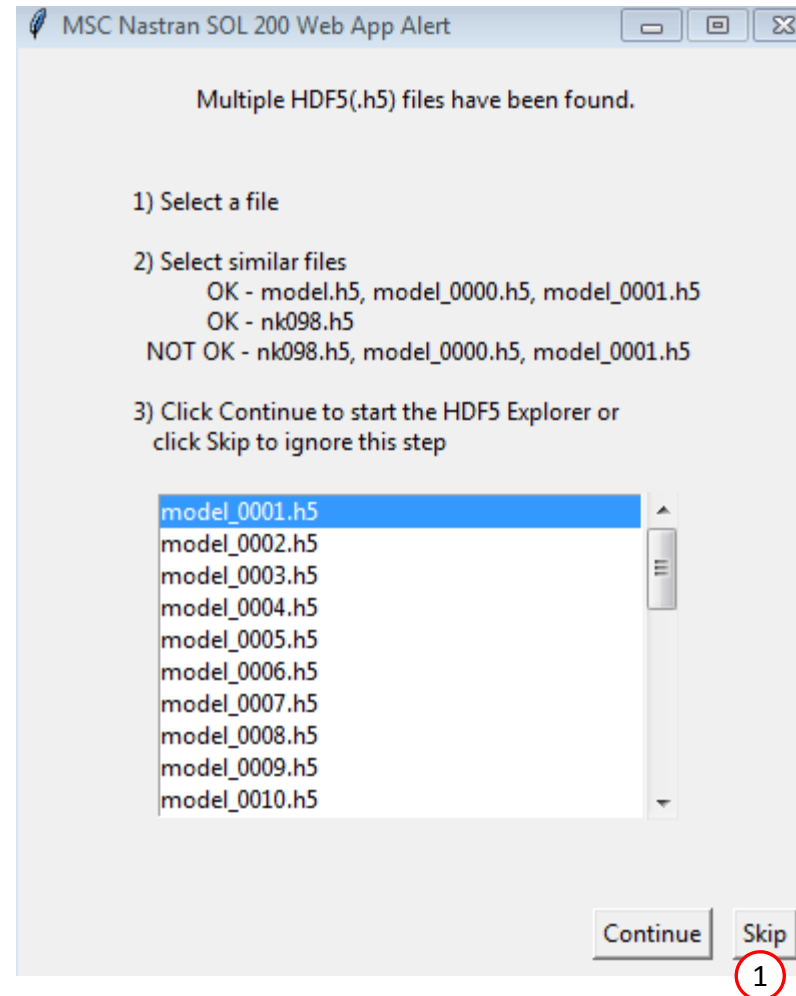
 Import

Best Valid Value



Machine Learning Results

1. After the process is complete, you will be asked to start the HDF5 Explorer. You can click Skip to skip opening the HDF5 Explorer.



Machine Learning Results

- The entire process consists of 2 phases.
 - Phase A – Initial Training Data Acquisition
 - This phase involves evaluating the FE model at different sampling points and recovering the monitored responses for the objective and constraints. The recovered monitored responses are referred to as training data.
 - This training data is used to train the regression model at the start of phase B, the machine learning phase.
 - Phase B – Machine Learning
 - This phase involves the machine learning process. The regression models for the objective and constraints are used to determine the next sample point to evaluate.
 - After each sample evaluated, the regression models are updated with the latest training data.
- This example was initially configured for a 50 sample Latin Hypercube design. After the initial training data was required, machine learning was executed for 50 runs. A total of 100 runs were performed.

Session
ID: 83016

Completed
successfully

Upload .csv File

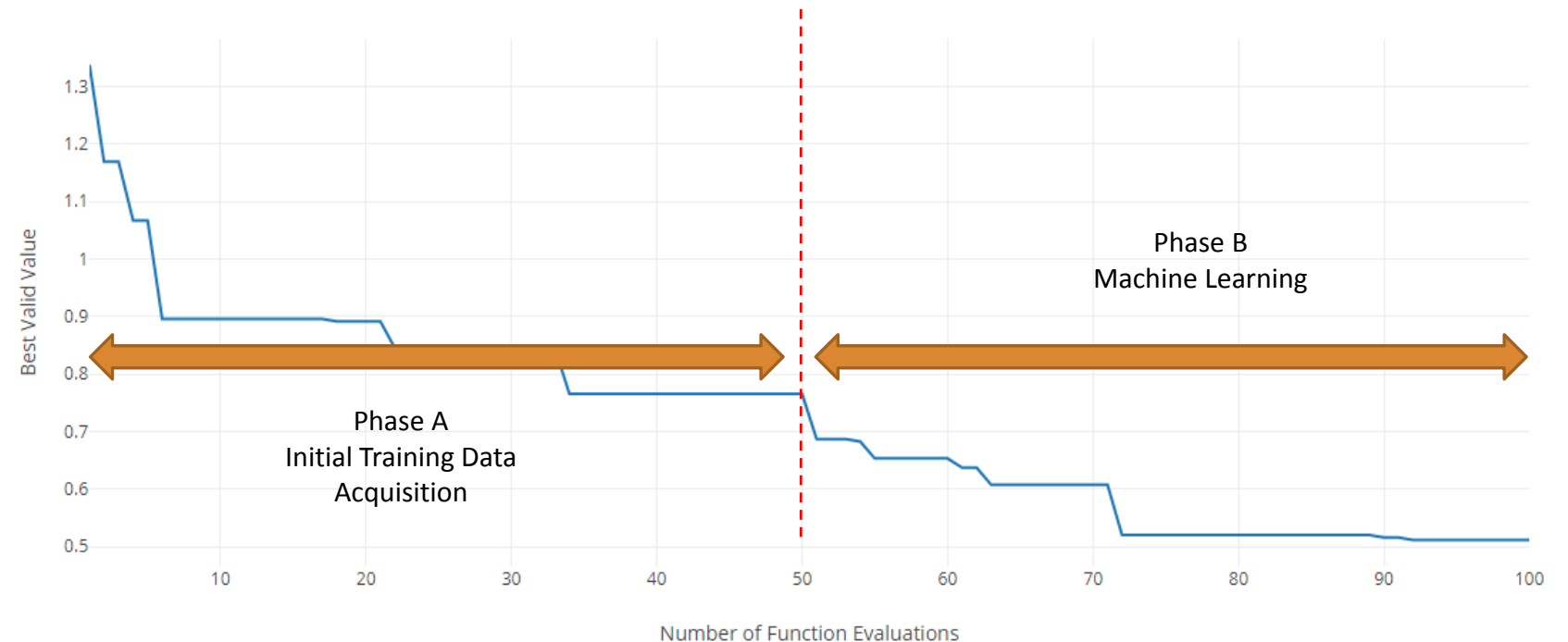
CSV Import

Select files

a_tmp_best_valid_value.csv

Import

Best Valid Value



Session
ID: 83016Completed
successfully

Upload .csv File

CSV Import

Select files

a_tmp_best_valid_value.csv

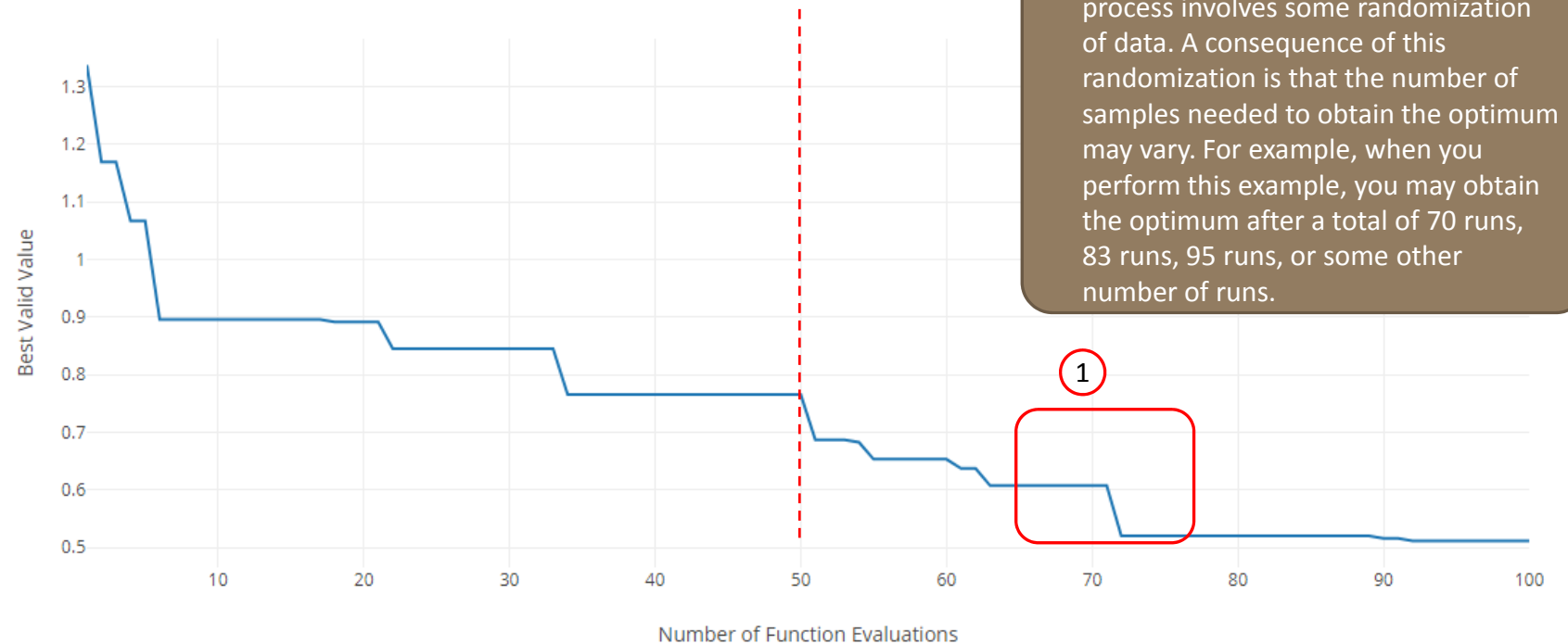
Import

Machine Learning Results

1. Each drop in the Best Valid Value (BVV) indicates a better feasible design has been obtained. For example, the machine learning process evaluates samples 69, 70 and 71 but the BVV is flat, indicating a better design has NOT been found. After sample 72 is evaluated, a better feasible design has been found, indicated by a drop in the BVV.

- Refer to the appendix for more explanation of the BVV.

Best Valid Value



- Important! The machine learning process involves some randomization of data. A consequence of this randomization is that the number of samples needed to obtain the optimum may vary. For example, when you perform this example, you may obtain the optimum after a total of 70 runs, 83 runs, 95 runs, or some other number of runs.

Machine Learning Results

1. A bar chart displays the objective value after each sample. A green colored bar indicates the constraints are satisfied for that sample. A gray colored bar indicates the constraints are NOT satisfied for that sample.
2. The Status message indicates sample 92 is the best design. Your solution might be different.
3. Click 10 to display 10 rows in the table.

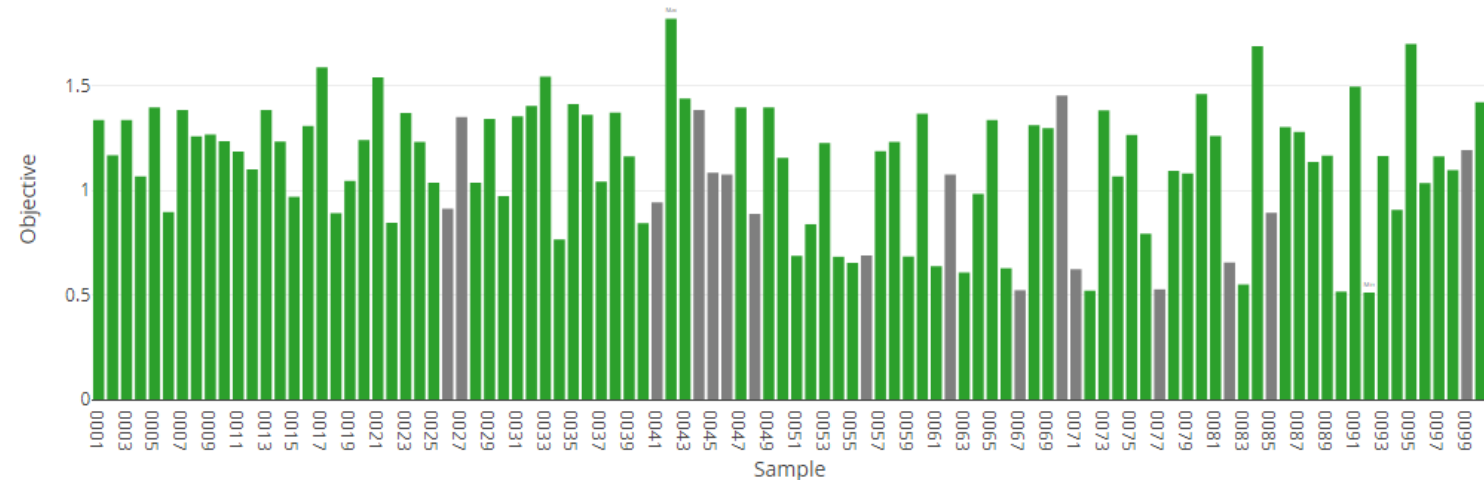
- Important! The machine learning process involves some randomization of data. A consequence of this randomization is that the number of samples needed to obtain the optimum may vary. For example, when you perform this example, you may obtain the optimum after a total of 70 runs, 83 runs, 95 runs, or some other number of runs.

Status

1

THE LATEST OPTIMAL SOLUTION IS SAMPLE # 92 (MIN), SAMPLE # 42 (MAX)
 OBJECTIVE = 5.1139E-1 (MIN), 1.8211E+0 (MAX)
 MAXIMUM CONSTRAINT VALUE = -8.1962E-2, -7.9325E-1 (FEASIBLE DESIGNS)

Objective for Each Sample



Data for Each Sample

Item	Sample 0001	Sample 0002	Sample 0003	Sample 0004	Sample 0005	Sample 0006	Sample 0007	Sample 0008	Sample 0009	Sa
Extrema (Max/Min)										
Objective	1.3369E+0	1.1688E+0	1.3359E+0	1.0667E+0	1.3968E+0	8.9552E-1	1.3836E+0	1.2591E+0	1.2673E+0	1.2
Normalized Constraint	-7.3836E-1	-6.2325E-1	-7.2014E-1	-5.5561E-1	-5.9428E-1	-4.1216E-1	-7.3948E-1	-7.2133E-1	-2.4485E-1	-1.8
X1	1.3893E+3	1.5407E+3	1.4880E+3	9.1461E+2	1.5975E+3	1.1944E+3	1.6273E+3	1.4271E+3	4.1338E+2	7.3
X2	1.9385E+3	1.5729E+3	1.4328E+3	9.8050E+2	2.0340E+3	1.1491E+3	1.9546E+3	6.8583E+2	2.2972E+3	5.2
X3	2.0031E+3	1.1574E+3	2.6271E+3	1.5356E+3	8.6225E+2	6.3967E+2	2.1268E+3	2.1268E+3	1.6599E+3	4.4

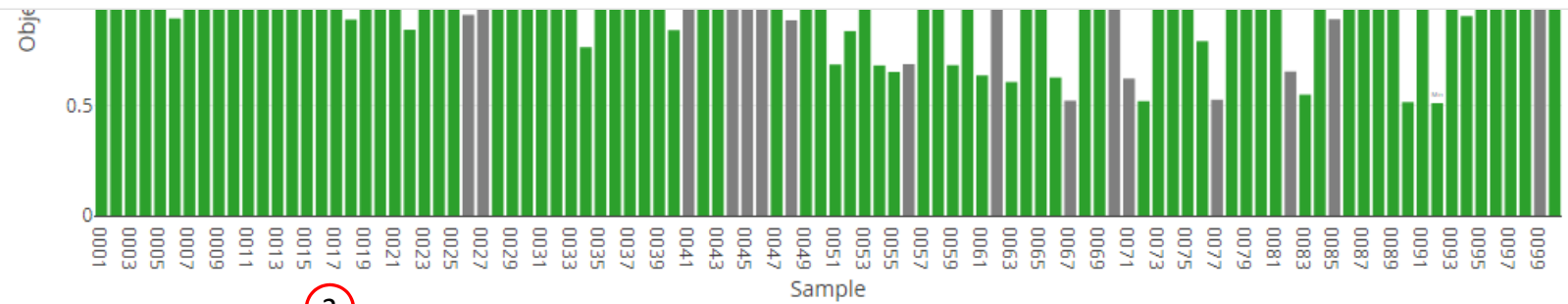
« 1 2 3 »

5 10 20 30 40 50

Machine Learning Results

1. Use the horizontal bar to locate sample 92 in the table.
2. The column displays sample 92's objective, max normalized constraint and parameter values

- Important! The machine learning process involves some randomization of data. A consequence of this randomization is that the number of samples needed to obtain the optimum may vary. For example, when you perform this example, you may obtain the optimum after a total of 70 runs, 83 runs, 95 runs, or some other number of runs.



Data for Each Sample

Item	0091	Sample 0092	Sample 0093	Sample 0094	Sample 0095	Sample 0096	Sample 0097	Sample 0098	Sample 0099	Sample 0100
Extrema (Max/Min)		Min								
Objective	0	5.1139E-1	1.1649E+0	9.0674E-1	1.7001E+0	1.0347E+0	1.1627E+0	1.0962E+0	1.1925E+0	1.4213E+0
Normalized Constraint	-1	-8.1962E-2	-4.9696E-1	-2.3263E-2	-7.5743E-1	-4.4317E-1	-4.4757E-1	-3.0209E-1	3.2400E-1	-7.2265E-1
X1	3	4.3229E+2	1.6624E+3	2.6706E+2	2.3264E+3	2.3350E+3	9.7353E+2	3.1907E+2	5.0395E+2	2.7617E+3
X2	3	1.0888E+3	2.0396E+3	1.4003E+3	4.4479E+2	1.3272E+3	1.2498E+3	1.5463E+3	2.4697E+3	9.0749E+2
X3	3	4.5861E+2	9.2647E+2	1.3339E+3	1.5161E+3	1.8653E+3	2.3104E+3	2.4021E+3	1.4302E+2	2.6936E+3
X4	2	1.9385E+2	1.5154E+3	1.0051E+2	2.8084E+3	2.9713E+2	1.8185E+3	2.2287E+3	4.6185E+2	2.0907E+3
X5	2	1.6987E+3	1.2410E+3	4.3725E+2	2.7434E+3	2.6571E+3	4.6141E+2	7.8762E+2	2.4295E+3	2.2135E+3
X6	3	5.2754E+2	2.7906E+3	8.7306E+2	1.5325E+3	7.0263E+2	2.3977E+3	1.4591E+3	1.0277E+3	2.6854E+3
X7	3	6.6888E+2	1.1000E+3	8.6979E+2	1.9908E+3	7.3925E+2	4.4721E+2	2.2023E+3	1.2633E+3	2.2149E+3



End of Tutorial

Appendix

Appendix Contents

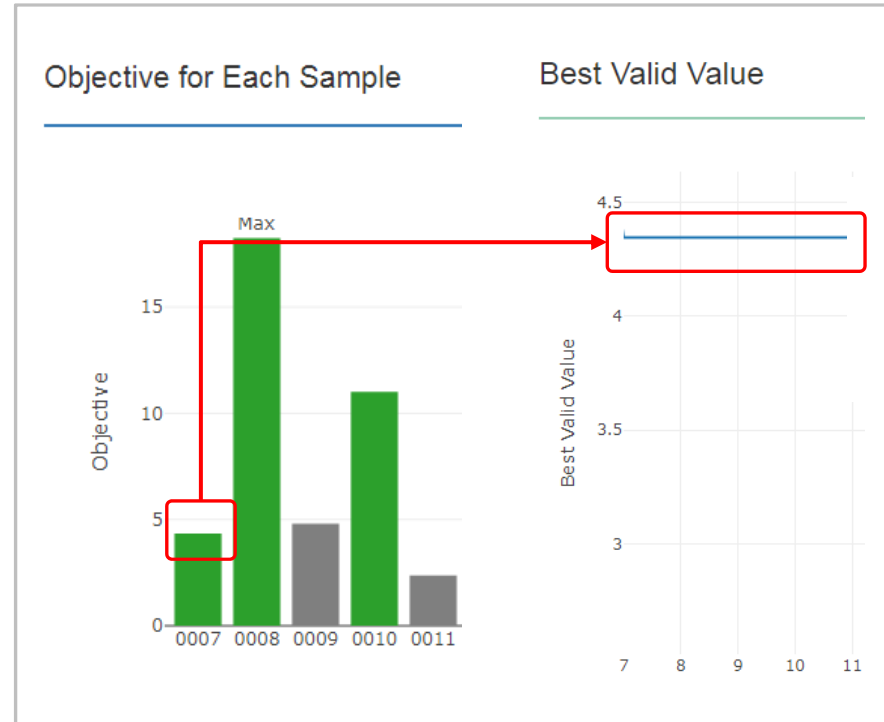
What is the Best Valid Value?

How to import and edit previous files

What is the Best Valid Value (BVV)?

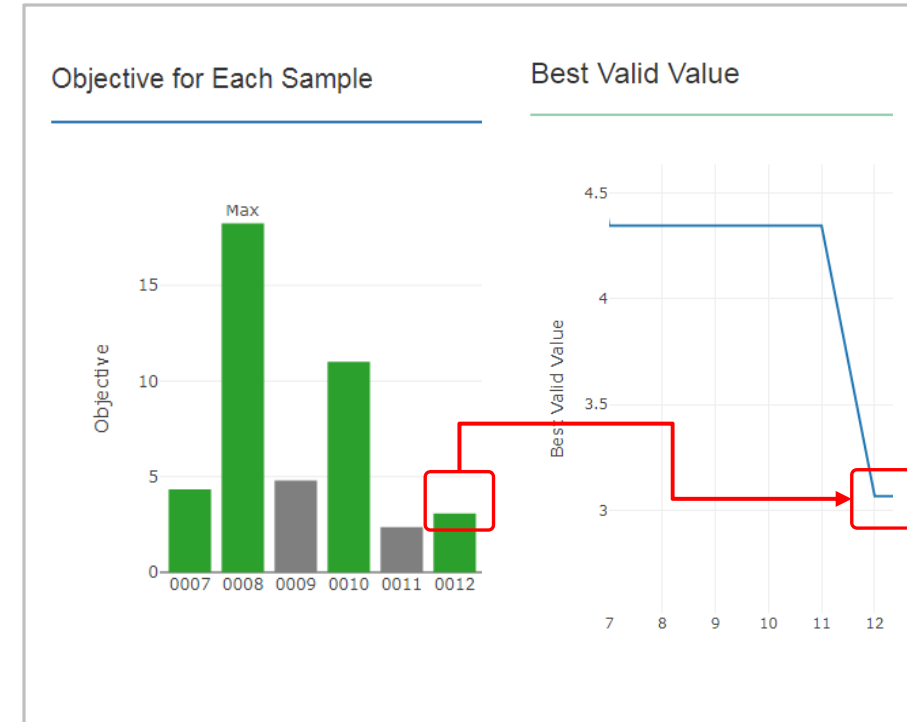
The Best Valid Value (BVV) is the best objective value of all previously evaluated feasible samples.

1. Consider samples 7-11. Out of samples 7-11, sample 7 is the best feasible design because its objective value is the lowest and the sample is a feasible design.
 - The BVV plot shows a flat line for samples 8, 9, 10 and 11 because a better feasible design has NOT been found.
2. Now consider samples 7-12. Samples 12 is now the best feasible design, the objective value of sample 12 is now less than sample 7.
 - The BVV plot is updated to reflect the new best design.



Samples 7-11

①



Samples 7-12

②

How to import and edit previous files

How to import and edit previous files

The parameters, samples and responses are contained in the following files

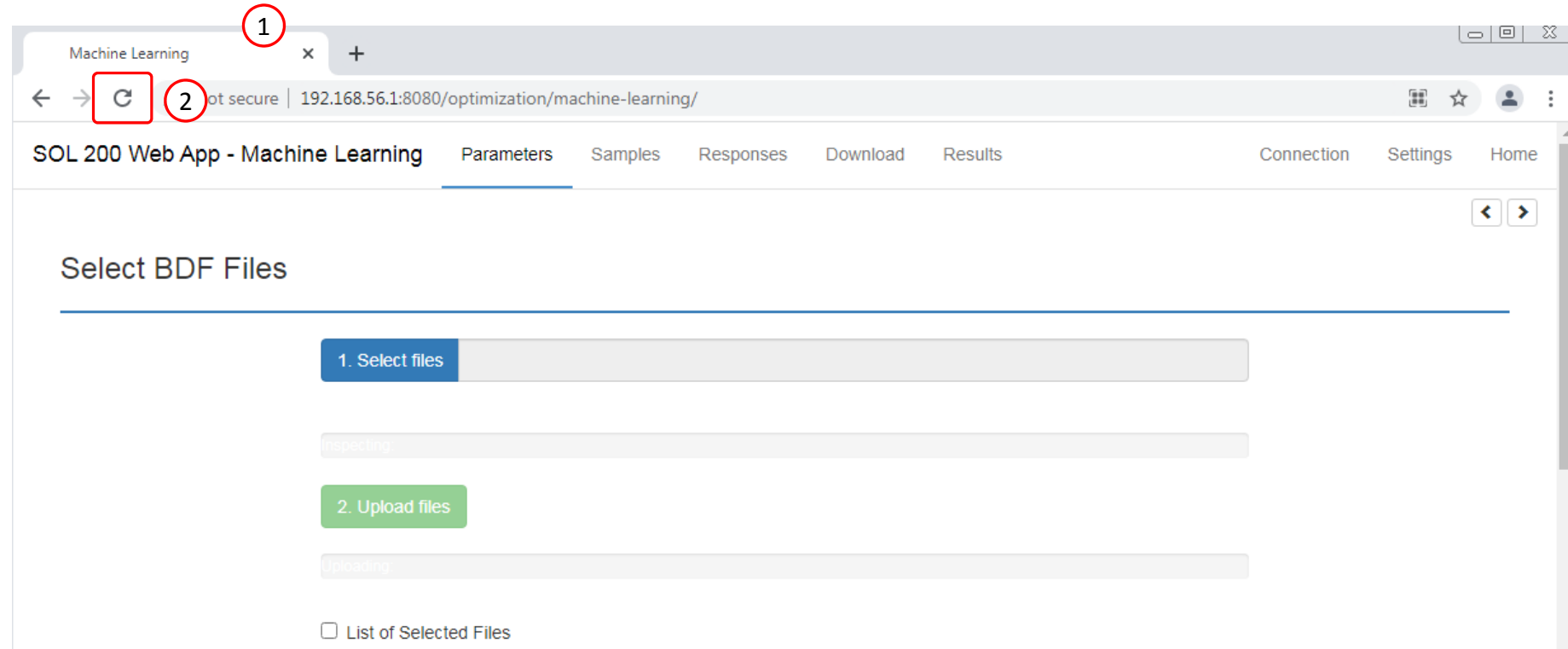
- app.config
- BDF files

These files may be imported back to the Machine Learning web app, and any parameters, samples and responses can be reconfigured

Import

1. Return to the window or tab that has the Machine Learning web app opened
2. Refresh the web page to start a new session

- Refreshing the page is only required when the *Select files* button is disabled.



Import

1. Click Select Files
2. Navigate to the folder named nastran_working_directory
3. Select all the BDF files AND the app.config file.
4. Click Open
5. Click Upload files

- All imports require the app.config file to be selected.

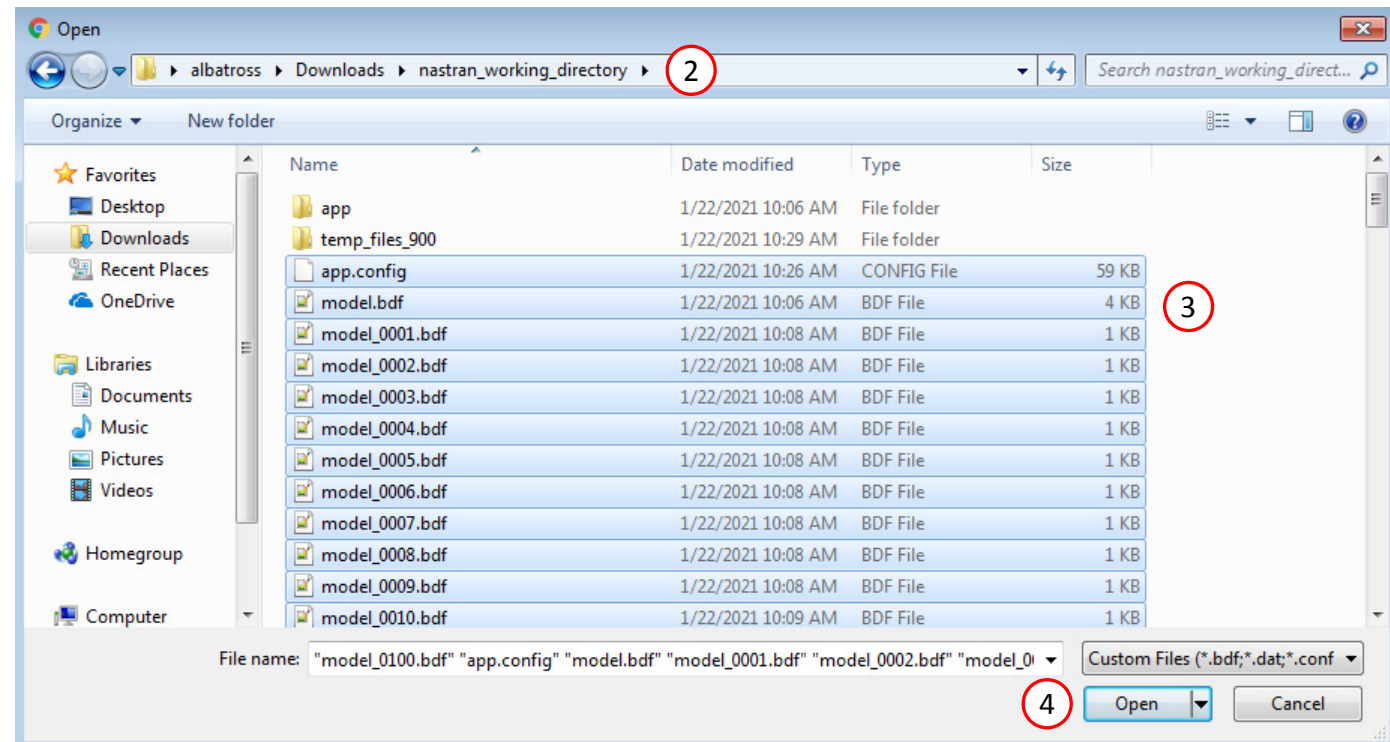
Select BDF Files

1. Select files 2 files selected

Inspecting: 100%

5. 2. Upload files

Uploading:



Import

For the Response section, the H5 file will need to be re-uploaded.

1. Click Responses
2. Select the H5 file
3. Click Upload
4. Data from the H5 is loaded and ready to use

SOL 200 Web App - Machine Learning Parameters Samples **Responses** Download Results Connection Settings Home

1

Upload .h5 File

2 1. Select files deslo_sol_200_entries_removed.h5

3 2. Upload files

Uploading

View Responses to Monitor

Monitored Responses Hide/Show Columns Reset Filters Download CSV

							Monitor the

SOL 200 Web App - Machine Learning Parameters Samples **Responses** Download Results Connection Settings Home

Select Responses to Monitor

Session ID: 2340 HDF5

Select Dataset

- ELEMENTAL/STRESS
- ELEMENTAL/STRESS
- NODAL/DISPLACEMENT
- NODAL/GRID_WEIGHT

Specify Entities

1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Examples: 1, 2, 3, etc.

☒ Auto Execute

Acquired Dataset

ELEMENTAL/STRESS/ROD - 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

EID	A	MSA
Element identification number	Axial stress	Axial Safety Margin*
1	280.427640...	5e-324

4

View Responses to Monitor

Monitored Responses Hide/Show Columns Reset Filters Download CSV

Delete	Label	Status	Objective	Lower Bound	Upper Bound
	r1				
	r2				
	r3				
	r4				
	r5				
	r1		MIN	Lower	Upper
	r2			-220.	220.
	r3			-220.	220.

Import

After import, any Parameter, Samples or Responses can be modified.

Select BDF Files

1. Select files

2 files selected

Inspecting: 100%

2. Upload files





















Uploading: 100 %

Select Parameters

\$ _1 _ || _2 _ || _3 _ || _4 _ || _5 _ || _6 _ || _7 _ || _8 _ || _9 _ || _10 _ |

```
FORCE 300 2 4.45E4 0.0 -1.0 0.0
FORCE 300 4 4.45E4 0.0 -1.0 0.0
MAT1 1 2.E5 0.0 0.3 7.86E-09
MATS1 1 1 PLASTIC 1 1 200.
PROD 1 1 %x1% 0.e0 0.e0 0.e0
PROD 2 1 %x2% 0.e0 0.e0 0.e0
PROD 3 1 %x3% 0.e0 0.e0 0.e0
PROD 4 1 %x4% 0.e0 0.e0 0.e0
PROD 5 1 %x5% 0.e0 0.e0 0.e0
PROD 6 1 %x6% 0.e0 0.e0 0.e0
PROD 7 1 %x7% 0.e0 0.e0 0.e0
PROD 8 1 %x8% 0.e0 0.e0 0.e0
PROD 9 1 %x9% 0.e0 0.e0 0.e0
PROD 10 1 %x10% 0.e0 0.e0 0.e0
SPC 1 5 123 0.0
SPC 1 6 123 0.0
TABLES1 1
0.0 0.0 0.001 200. .0126221500. ENDT
```

Configure Parameters

Delete	Parameter	Status	Low	High	Comments
	x1		78.5	2826.0	Field 4 of PI
	x2		78.5	2826.0	Field 4 of PI
	x3		78.5	2826.0	Field 4 of PI
	x4		78.5	2826.0	Field 4 of PI
	x5		78.5	2826.0	Field 4 of PI
	x6		78.5	2826.0	Field 4 of PI
	x7		78.5	2826.0	Field 4 of PI
	x8		78.5	2826.0	Field 4 of PI
	x9		78.5	2826.0	Field 4 of PI
	x10		78.5	2826.0	Field 4 of PI