

Simplifying Progress

SIMCA® 16 What's new



SIMCA® 16 Overview

- Umetrics[®] suite SIMCA[®] is focused on delivering a full data analysis experience, from data organization to data based decision making supported by multivariate models adopted for both single and multiblock analysis.
- SIMCA[®] 16 delivers improvements in three focus areas for the current, experienced, user as well as the beginner
- Usability improvements
 - By providing new users with a guided introduction to SIMCA and existing users with smoother plot interactivity, quick raw data
 visualization, and relevant workflows we want to give all users a rewarding data analytic experience
- Model use and understanding
 - Giving users new and extended tools for interrogating and experimenting with generated models, we help turn model knowledge into real world interpretations
- New technology
 - Aiming at driving innovation within the multivariate analytics field we provide our users access to new and advanced analytical technology



SIMCA® 16 – Some highlights

- Usability improvements
 - A "Start screen" when no project is open
 - A guided "getting started" tour for beginners
 - Flexible merging of datasets
 - Context based ribbon structure
 - Workset wizard with integrated quick info pane
 - Data explorer side pane with new and updated information panes
 - Plot properties are in a context based side pane
 - An updated report generator
 - Improved legend handling for BCC multiplot and others

- Model use and understanding
 - Score space exploration
 - Multivariate solver
- New technology
 - MOCA Multiblock Orthogonal Component Analysis
- Python additions and improvements
 - Install external packages inside SIMCA
 - Python plugin creation for File import formats and Data preprocessing (filters)



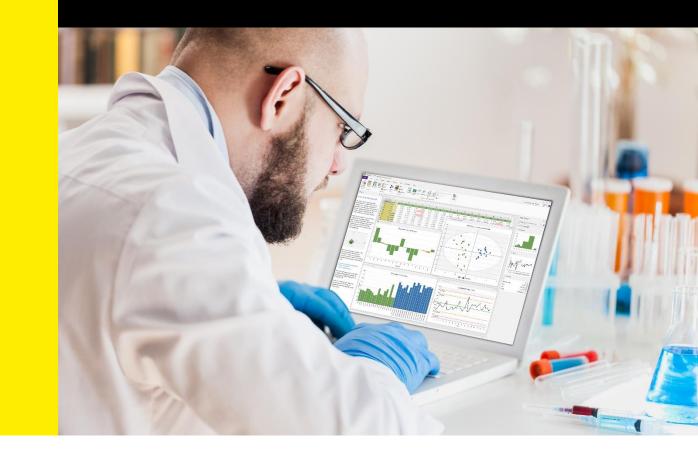
Learning what's new in SIMCA 16

- In the following slides an overview is given to the changes and additions made in SIMCA 16
- For more descriptions on how to use the features, please watch the videos that can be reached from the start page of SIMCA 16
- You can also search for Sartorius Data Analytics on YouTube



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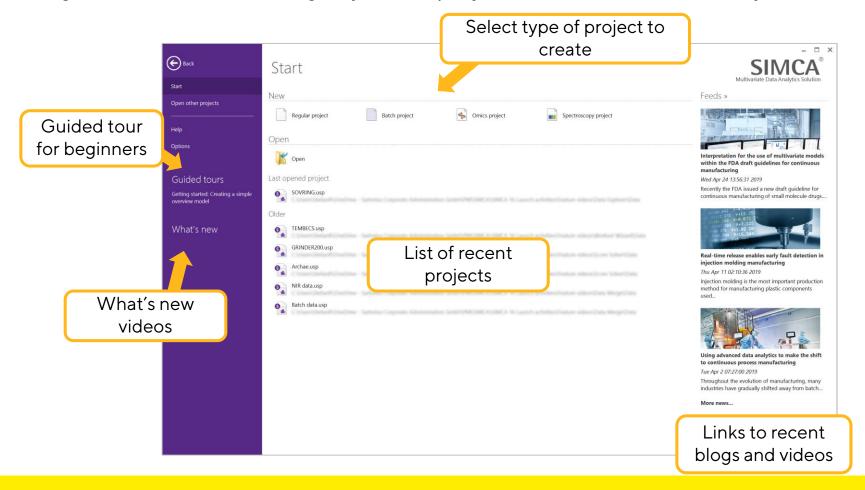
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Usability improvements

SIMCA[®] start screen and feeds

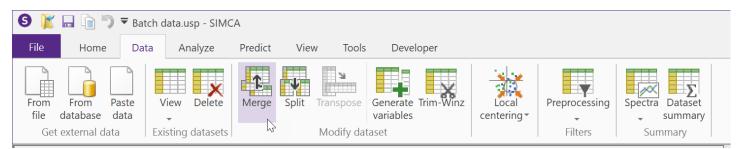
• When starting SIMCA without having any active projects, the start screen let's you chose what to do





Flexible merging of datasets

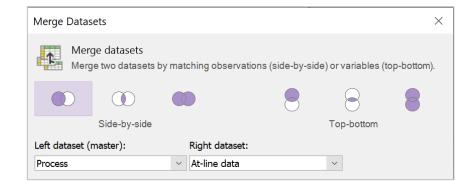
- Many users struggle to align and merge data before they get the data into SIMCA
 - In earlier versions you could only merge datasets based on primary ID inside SIMCA
- In SIMCA 16 we introduce a flexible merging of dataset
 - Merge data Side-by-Side or Top-Bottom
 - Merge by identifiers and variables
- Merge is found on the Data tab





Flexible merging of datasets

- Types of merge
 - Master/slave, intersection or full merge
- Options
 - Exact, nearest and intrapolate
 - Fill with last good value or empty cells
- Typical application for Side-by-Side merging
 - At-line data sampled at long and irregular intervals with process data sampled at high frequency
- Typical application for Top-Bottom merging
 - Spectral data with different resolution or not identical wavelengths





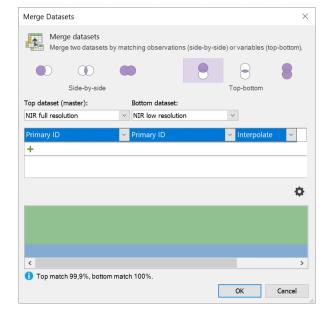
Flexible merging of datasets

- Side-by-Side example
 - Process data and At-line data
 - Left merge (master/slave)
 - Match by Batch ID and timestamps
 - Fill with last good value

Merge Datasets		×
Merge datasets Merge two datasets	by matching observations (side-l	by-side) or variables (top-bottom).
Side-by-side		Top-bottom
Left dataset (master):	Right dataset:	×
\$BatchID	\$BatchID	Exact
Timestamp	 Sample time 	 Forward X
+ Non matching rows: Use	last good Insert missing	Å

		^

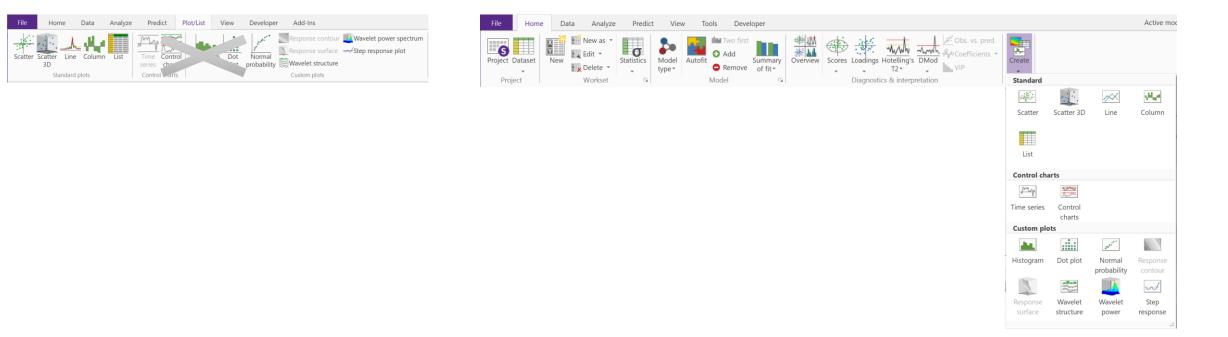
- Top-Bottom
 - Spectral data of different resolutions
 - Top merge (master/slave)
 - Match by Primary ID (Wavelength)
 - Intrapolate





Cleaned ribbon structure

- Plot/List tab removed
 - All generic plots are available in the Create gallery on the Home tab





Context based ribbon structure

Batch tab removed for batch projects

Analyze Predict Plot/List View Developer Add-Ins Data Batch **1** Variable DModX Hotelling's Obs vs. List Scores Variable DModX Hotelling time/maturity List Observed vs. Unaligned Create batch Create hierarchica T2 BCC time/maturity BCC BCC BCC -PS BCC - PS BCC PS BCC PS BCC batch models smoothed Y vs. aligned Time/Maturit Variable summar Prediction control chart Dataset

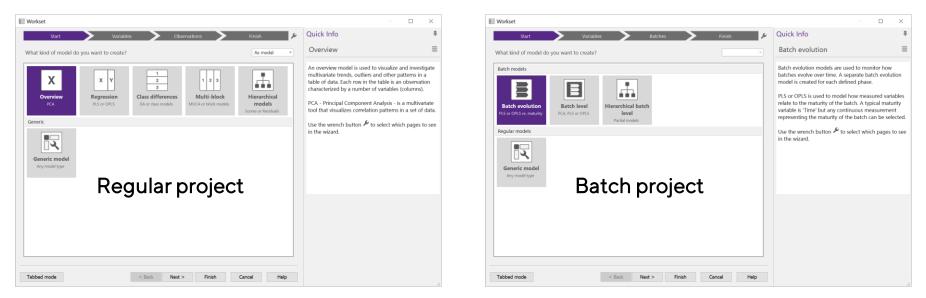
- The Home tab provides relevant plots for the selected model type (BEM or BLM)
 - The Batch prediction plots are now in the Predict tab

	File Hom	e Data Analyze	Predict	View Tools Developer	
BEM	Project	New Workset	Statistics	Autofit O Add Summary e Model Model G	Variable Scores Hotelling's DModX BCC list Batch control charts Predicted Time/Maturity Plot/list
	File Hom	,	Predict	View Tools Developer	
BLM	Project Dataset	New as *	Statistics Mo	Autofit • Add • Summary	Image: Construction of the second
	Project Dataset	Workset	v typ		T2 VIP VIP

• Batch level dataset creation is now handled by the Workset wizard.

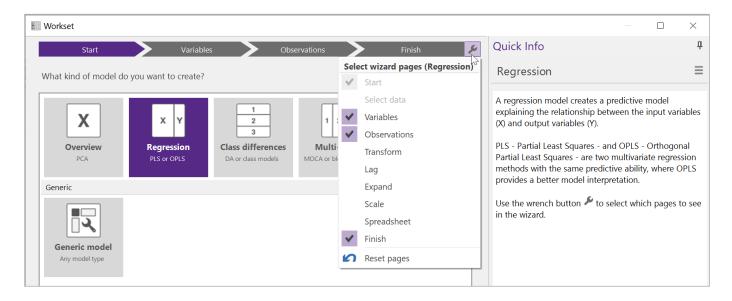


- The workset dialog has been converted to a workset wizard
- The wizard starts with an Start page where you define your objective
 - Different objectives for regular and batch projects
 - The selected objective controls what options will appear later in the wizard





- A progress bar indicates where you are in the workset definition
 - Clicking Next will lead you through the workset creation
 - You can also click in the progress bar to go directly to a selected step
- The steps in the progress bar can be modified to include other steps
 - The modifications will be remembered by objective



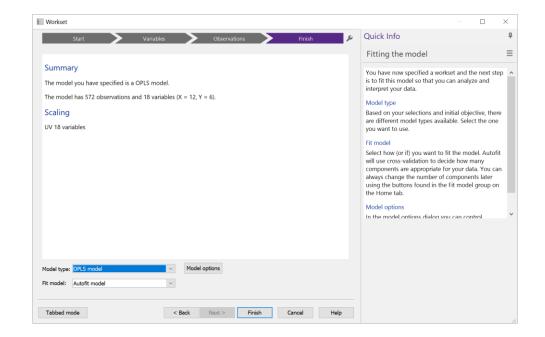


- The Updated Quick Info is now integrated on all steps in the workset wizard
- On the Spreadsheet step (needs to be added) you have access to Trim-Winz
 - Trimming done on the workset is only affecting the individual model and can be brought to SIMCA-online

E Workset	- 🗆 X	Workset	– 🗆 X
Start 💙 Variables 💙 Observations 💙 Spreadsheet 💙 Finish	🗾 🔎 Quick Info 🛛 🗜	Start 💙 Variables 💙 Observations 💙 Spreadsheet 💙 Finish 🔑	Quick Info 🕴
Variables: included: 18 (X=12, Y=6), excluded: 1, selected: 1	Variable Ton_in	Workset Workset as raw data	Variable ™_in
Variable ID Info) Deter/Time w jumps	🔺 Histogram	1 2 3 4 5 6 7 ^	∡ Trim-Winz
X Ton in		1 Primary ID Date/Time w jumps Select Ton_in KR30_IN KR40_IN PARM	Find variable values outside 💌
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X KR40_IN Excl	120 -	3 2 1992-03-05 11:01:00 O 0 0 4,65 1,2	0 and 1717,76
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× H5_2 × PKR_30	40-	6 5 1992-03-05 11:04:00 O 0 0 1,462	and
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Y PAR		11 10 1992-03-05 11:09:00 O 0 0 0,212	✓ Histogram
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		22 21 1992-03-05 11:20:00 O 0 0 0,	✓ Trend
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Find Variable ID			0 50 100 150 200 250 300 350 400 450 500 550
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Tabbed mode < Back Next > Finish Cancel H	elp	Tabbed mode < Back Next > Finish Cancel Help	Quick Info Trimming Info



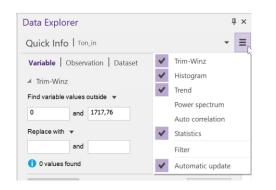
- The last step in the wizard let's you chose model type and if you want to fit the model automatically
 - The selection is remembered
- Model options (e.g. CV-groups and significance) can be accessed on the last page of the workset wizard

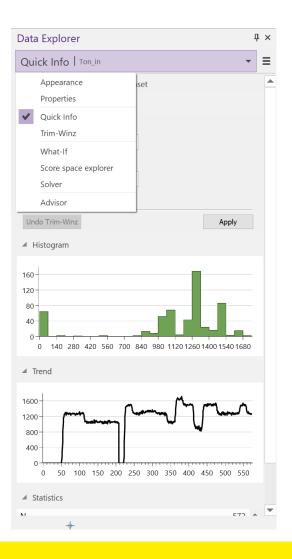




Data Explorer side pane

- The Data Explorer pane holds a collection of context based informative panes
 - Some have been modified and extended and some completely new side panes
- The Data Explorer pane will show information relevant to the active window in the SIMCA work area
- An options menu for customization is available for each individual pane

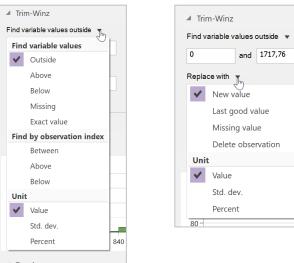


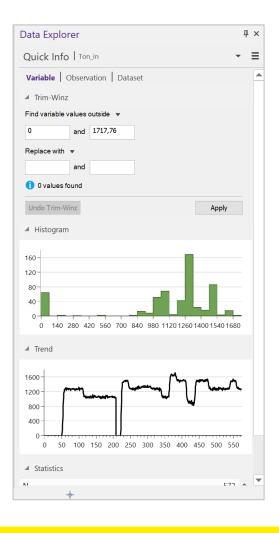




Data Explorer side pane - Quick Info with Trim-Winz

- An extended Quick Info is activated when clicking in lists, e.g. Datasets
- Trim-Winz has it's own section but is also included in the Quick Info for datasets
 - Trim-Winz has been extended with more search and replace functions including find "Missing" and "Exact"

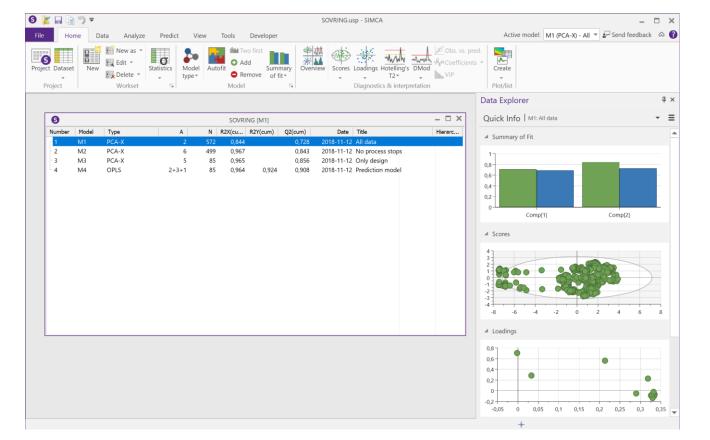






Data Explorer side pane – Quick Info on models

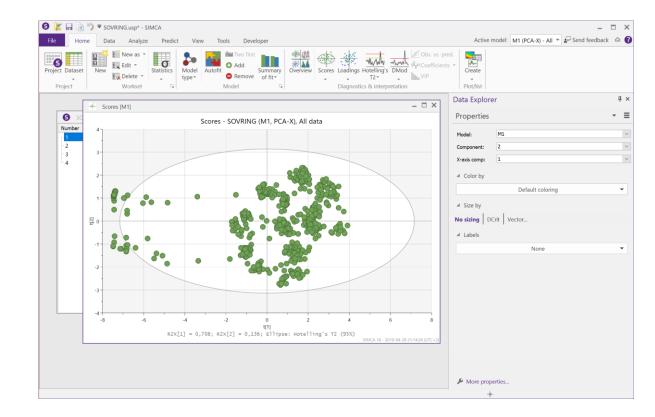
- When selecting a model in the project window, a selection of model plots appear
 - You can add or remove plots using the options menu next to the Quick Info header
 - Double-clicking one of these Quick info plots pops it out as a full size plot





Data Explorer side pane - Plot properties

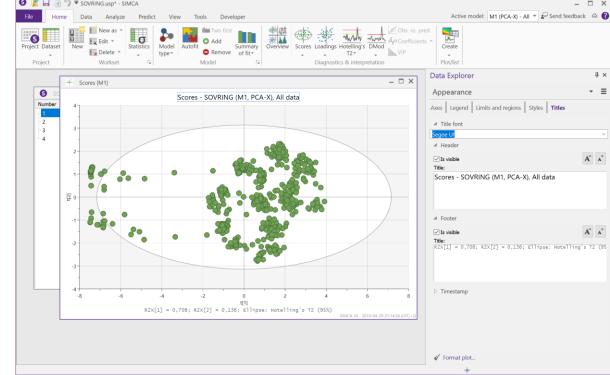
- The Properties pane opens up when a plot becomes active
 - Control content, colors and labels





Data Explorer side pane – Plot appearance

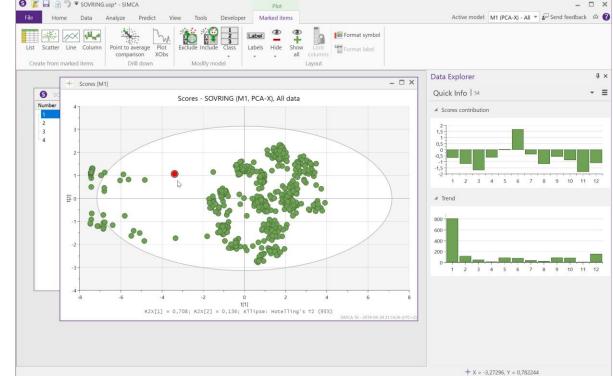
 When clicking on a title or an axis, the new Appearance pane becomes active letting you modify the appearance of the active plot





Data Explorer side pane - Observation contribution

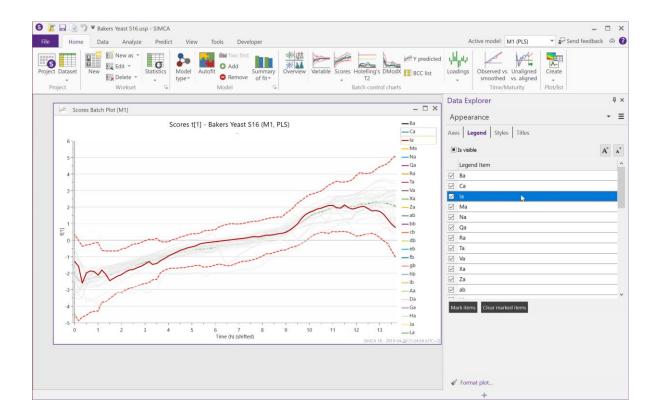
 Selecting an individual point or a group of points in an observation plot will open up the Quick Info with a contribution plot





Data Explorer side pane - Legend interaction

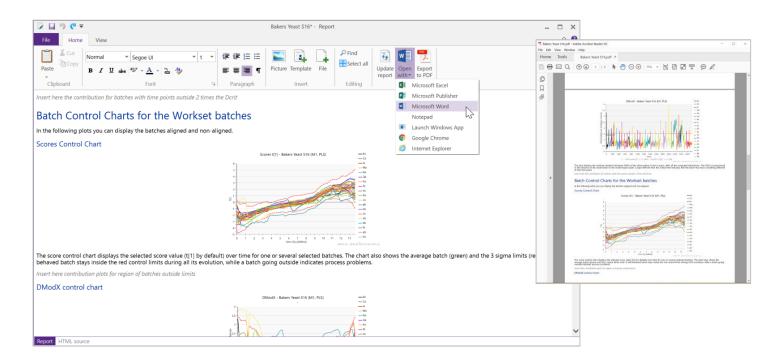
- Clicking on the legend opens the new, scrollable, legend pane
 - The legend pane lets you highlight objects and select which objects to display





An updated report generator - Formats

- The report generator has been updated and now allows for saving and exchanging content in a variety of formats
 - The report always starts as HTML
 - The report is by default saved in the SIMCA project file and can also be exported to an external file





An updated report generator - Templates

- Each model type has its own default report template
 - The included templates contain descriptive texts and some standard plots
 - The default templates can be modified to include other plots and descriptions to suit your needs
 - The templates can be shared with other SIMCA users

Insert Tem	nplate						\times
Default repor	Introduction for Batch	PCA-Y Template	PCA-X Template	PCA-XY Template	PLS Template	PLS-DA Template	
Batch Template	OPLS Template	OPLS-DA Template	Conclusion	MOCA Template	PCA-X BLM Template		
Template:	Default report						
					ОК	Cance	



Multiplots for Summary of Fit and Batch Control Charts

- Multiplots have been implemented for some plots
 - Summary of fit for class or phase models
 - Batch control charts (BCC:s) for multiple phases



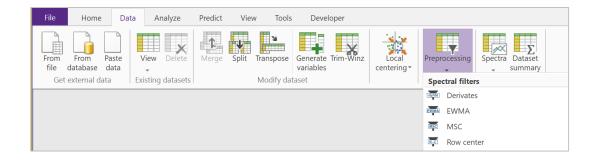


SIMCA® 16 – Other additions and modifications

- OPLS has been extended to include all O2PLS calculations
 - O2PLS has therefore been removed
 - Select Regression objective in Workset wizard and on the Finish step, select OPLS
 - Use model options to include the additional O2PLS components

	Model options
	Model Options
Model type: OPLS model V Model options	Model Distance to model Coefficients Residuals / R2 Predictions CV groups M
Fit model: Autofit model	Use cross-validation when fitting
	⊡ Calculate PCA on residuals (O2PLS)
Tabbed mode Back Next > Finish Cancel	

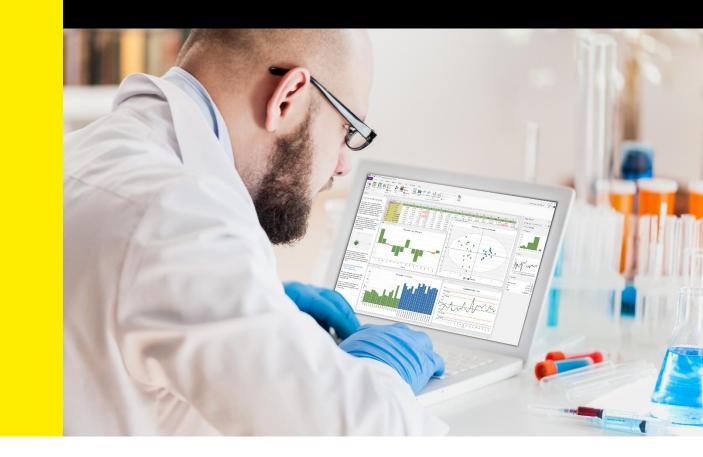
- Spectral filters have changed name to Preprocessing (on the Data tab)
 - The new Python preprocessing plugin has more capabilities than only spectral filtering





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Model use and understanding

Model interrogation innovations - Score space explorer

Works on regular projects

- Score space explorer
 - Several customers have asked for a tool to easier investigate and convert score points to original variables

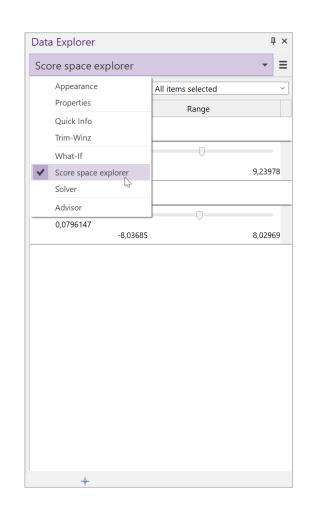
- Score space explorer scenario
 - You have a PC model of historical batches of raw materials and see an area in the score plot that is empty and you want to know what type of material is missing





Score space explorer

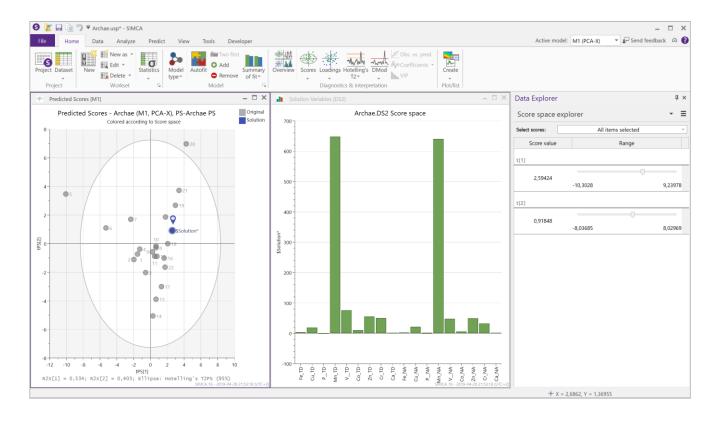
- Finding the variable profile for a specific location in a score scatter plot
 - The multivariate correlation is retained
 - Found in the Data explorer pane
- The interface consist of two plots and a slider section in the side pane
 - Predicted score scatter plot
 - An observation variable column chart for the selected point
 - A set of score vector sliders (in the side pane)





Score space explorer

- The sliders and the score scatter plot indicate the selected point
- The variable plot provides the profile corresponding to the selected score point

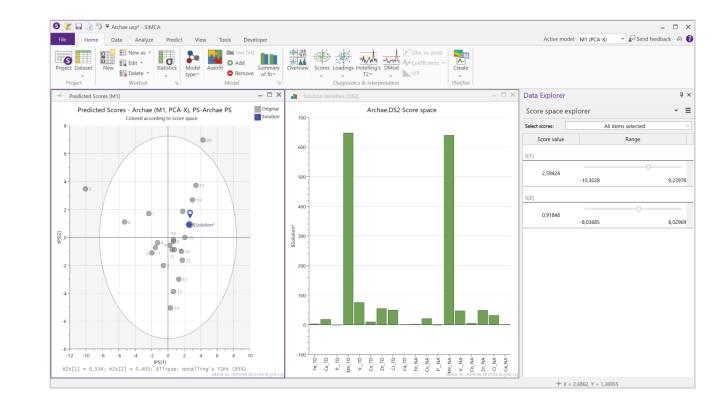




Score space explorer

- You can interact with the Score space explorer in two ways
 - Use the sliders to move the selected point in the score plot or
 - Click in the scatter plot

The variable plot instantly displays the corresponding variable settings





Model interrogation innovations – Multivariate solver

- Works on regular project OPLS models
- Multivariate solver
 - Similar to an optimizer but with fixed target values (i.e. not optimize or minimize)
 - Thanks to OPLS, the solver gives instant answer without iterations

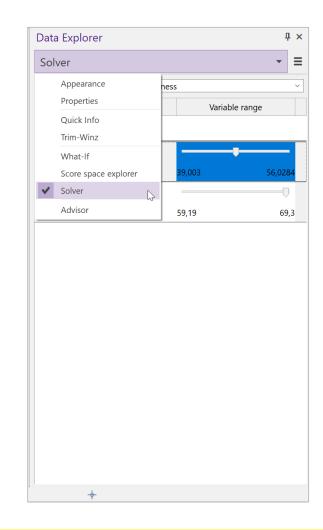
Solver scenario

- You have a predictive OPLS model and some of the inputs have to change (e.g. a new batch of raw material) and you need to know how to compensate to reach the same output
- Some X-variables are fixed and target Y is known



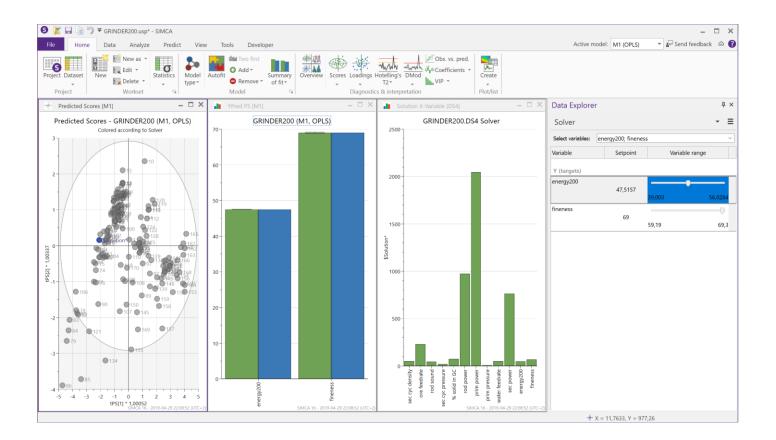


- The solver is accessed through the Data Explorer pane
- The solver works for single or multiple Y:s
 - It is recommended that the Y-variables are uncorrelated or only weakly correlated
- The solver works in two steps
 - The score values, corresponding to the desired target Y, are calculated
 - The variable settings for the score point are calculated
 - Only solutions where the multivariate pattern is retained are considered when no X-constraints are used
- When using constraints on X-variables the correlation pattern is broken
 - When the suggested solution no longer belongs to the model (PModX+) a warning will be given



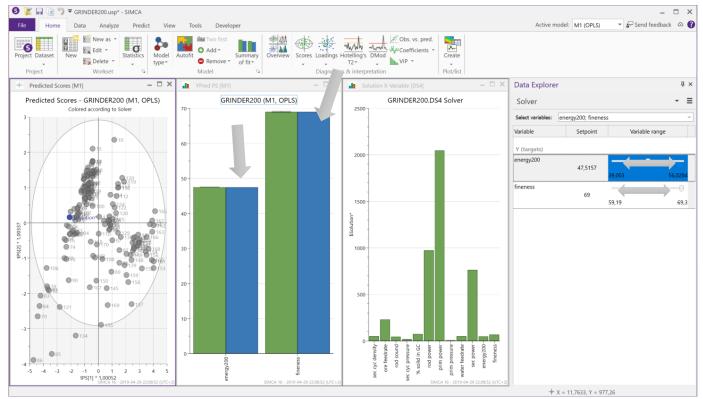


- The interface consist of three plots and a slider section in the side pane
 - Predicted score scatter plot
 - A column plot showing desired and predicted Y-values
 - An observation variable column chart for the suggested solution



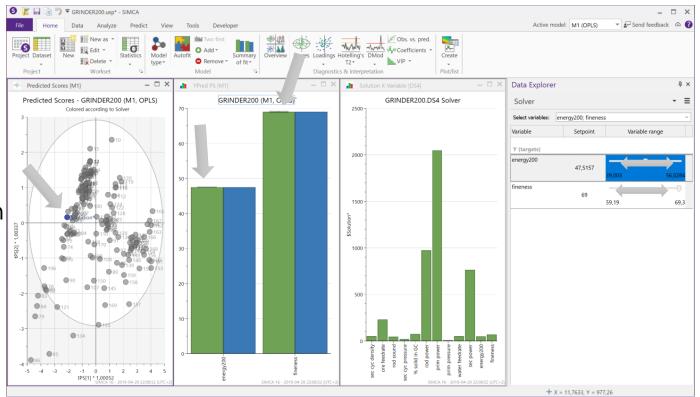


- The sliders are used to set the target values for Y
 - You can also type in the Setpoint field to the left of the sliders
- The desired Y target values are displayed as blue bars in the middle plot





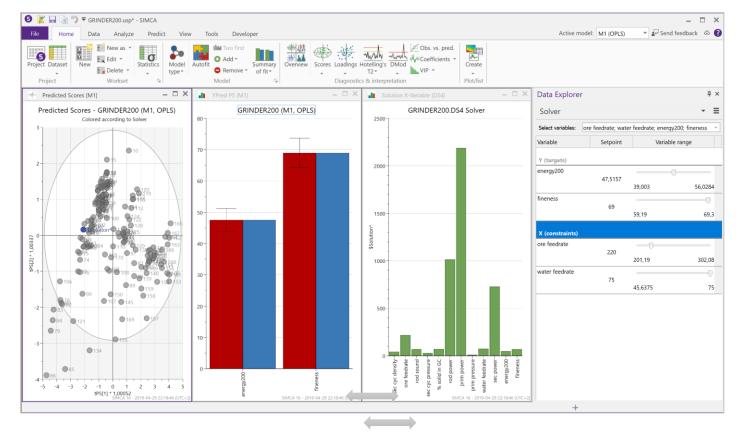
- The score scatter plot indicate the suggested solution
 - And the green bars in the middle plot show the predicted Y-values
- The variable plot to the right shows the variable profile for the suggested solution





Multivariate solver

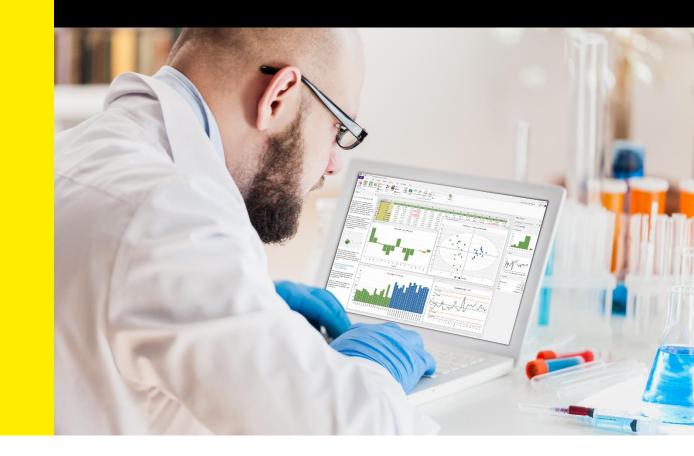
- When using constraints on X-variables the correlation pattern is broken
 - Some level of X-constraint is OK but not too much
 - Select X-variables in Select variables menu and set the constraints using the sliders
 - When suggested solution no longer belongs to the model (PModX+) the prediction plot turns red





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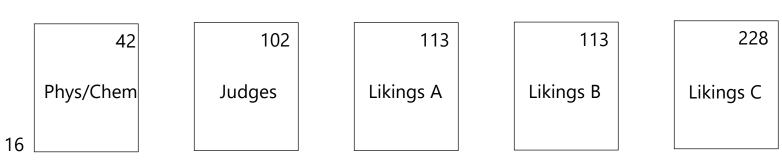
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New technology

Multiblock Orthogonal Component Analysis - MOCA

- A new and novel method for analyzing multiblock data is introduced
 - All blocks have the same observations but the variables differ
- Some data examples are:
 - Systems biology where samples are analyzed using e.g. metabolomics, proteomics, lipidomics
 - Manufacturing data where process signals are complemented with spectroscopy
 - Sensory data where expert panel data are compared to chemical analysis and consumer preference

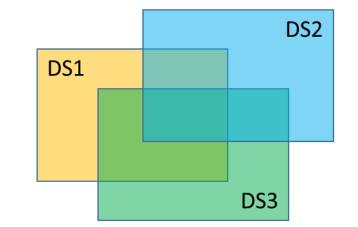


Sensory data example - 5 blocks of data



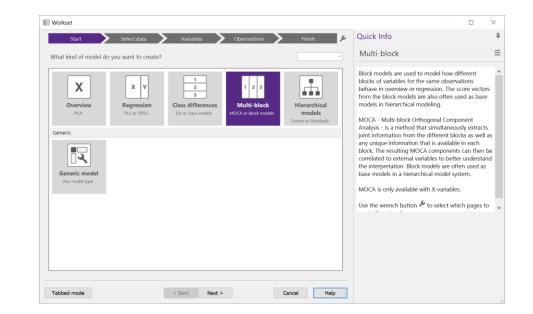
Multiblock Orthogonal Component Analysis - MOCA

- MOCA provides an overview of the information in all blocks simultaneously, similar to what PCA provides for one block and O2PLS provides for two blocks
- The method answers the questions
 - How much of the information in the blocks is common/joint among the blocks?
 - How much of the variation is unique for each specific block?
- MOCA is based on the OnPLS-algorithm which is using OPLS and PCA in an innovative way to find the joint and unique block information in a single analysis





- A process dataset with three blocks of data in three separate datasets
 - 92 observations
 - xin : 7 variables (input related variables)
 - xmd : 18 variables (intermediate process variables)
 - y : 8 variables (output related variables)
 - In addition there is one external data signal : Ext1
- Start the Workset wizard and select the Multiblock objective





- Select all datasets from the select data list
 - The blocks do not have to be in separate datasets but it is recommended

Workset					
Start	Select data Variables	Observations	Finish 🔑	Quick Info	џ
Specify which datasets	to include and in which order to include the	m.		Process one	Ξ
Dataset	Size (Var./Obs.)	Details	f	Select the datasets that hold the data you	want to
i∕xin	7 x 92			analyze.	
√xmd	18 x 92			You can control the order in which variable	
√у	8 x 92			observations are displayed in the workset a dialogs by changing the dataset order usin	
✓Ext	1 x 92			in the header of the dataset list. For instan	ce, if you
Primary ID match:	100 %			have spectral data with Y-variables in a sep dataset, you can move the Y-variables data easy access. The datasets are automatically aligned acc primary IDs.	aset first for
Tabbed mode	< Back Next	Cancel	Help		



• In the variable step, set blocks according to the datasets

					Variable x8md	
ariables: included: 34 (X=34, Y	*					
/ariable ID x1in	Info	Dataset xin		X	 Histogram 	
x2in		xin		Y		
x3in		xin			25	
x4in		xin		Exclude	20-	
x5in		xin				
x6in		xin		Blocks:	15	
x7in		xin		Set block 🔻	10	
x8md		xmd			5	
x9md		xmd	1			
xamd		xmd	Set all		-3 -2,5 -2 -1,5 -1 -0,5 0 0,5	1 1,5
xbmd		xmd	Set all		-3 -2,3 -2 -1,5 -1 -0,5 0 0,5	C,1 1
xcmd		xmd	Set block from	dataset		
xdmd		xmd		1	Trend	
xemd		xmd	Set block from	variable ID 🕨 🕨		
xfmd		xmd	Reset selected			
xgnx		xmd	Reset selected			
xhnx		xmd	Reset all			
xinx		xmd				~
xjnx		xmd				- m
xknx		xmd			-1,5	<u>γ</u>
xlnx		xmd			25	
xmen		xmd			-2,5 -2,5 -2,5 -2,5 -2,5 -2,5 -2,5 -2,5	70 75 00 05 0
xnen		xmd			5 10 15 20 25 50 55 40 45 50 55 80 05	101000039
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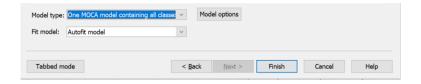
- Variables not part of any block can still be kept as X-variables (Optional step)
 - E.g. sample treatments, run order, groups, external disturbances,...
 - Reset potential external variables so that they are defined as X but not assigned to any block

X xbmd X xcmd X xdmd X xfmd X xfmd X xfmd X xgnx X xhnx X xinx X xinx X xjnx X xjnx X xknx		lected: 1 Info	Dataset xmd xmd xmd xmd xmd xmd xmd xmd xmd		∧ X Y Exclude Blocks: Set block ▼	Variable Ext1	
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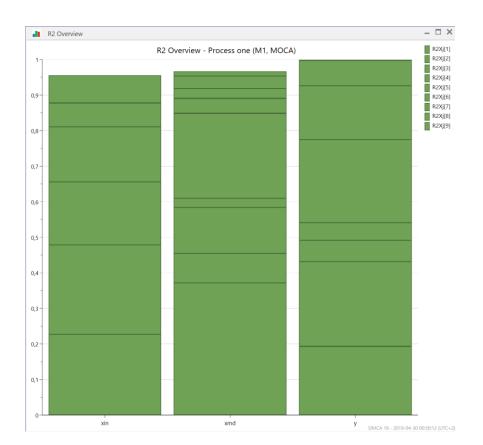


MOCA Analysis - Fitting the model

 When reaching the Finish step in the Workset wizard, select MOCA analysis and Autofit model (only option for MOCA)



 The MOCA model R2 overview has one column for each block and summarizes the explained variation for all blocks





MOCA Analysis - Model window and R2 Overview

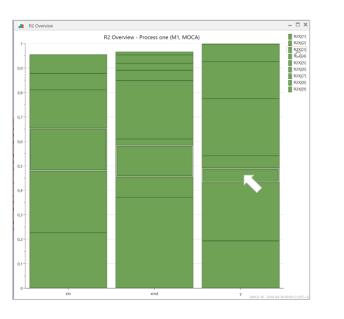
- The model window and the R2 overview show the same information
 - The explained variation by component and cumulative
- In this example we only have joint components
 - Joint components are green in the cumulative R2 overview (unique are blue)
 - The first four are globally joint (components including all blocks)
 - The next three (components 5-7) are locally joint between xmd and y
 - The last two (components 8 and 9) are locally joint between xin and xmd

9	Model M1						-
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	2	0,251		0,0831		0,238	0,431
	3	0,177		0,129			0,491
	5	0,155		0,0257		0,0495 0,234	0,541
	5			0,239		0,234	0,926
	-			0,0417		0,0715	0,920
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MOCA Analysis - R2 overview drill down

- Double clicking on one of the sections in the R2 overview brings up the explained variation of that component per variable
 - Displayed as a multiplot, one plot per block







MOCA Analysis - Scores correlation matrix

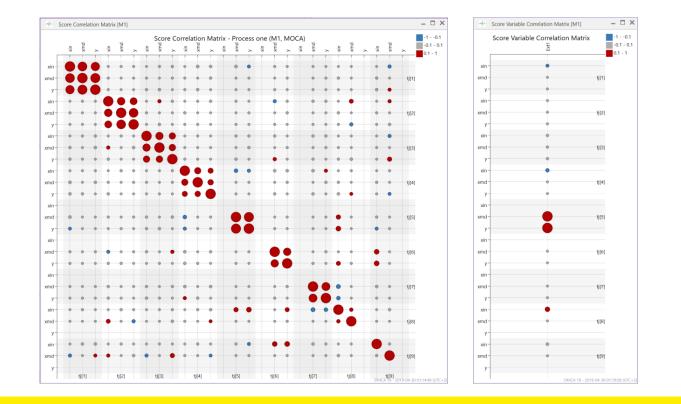
- The score correlation matrix is used to check the quality of the model
- With three blocks and 9 components the score correlation matrix is built up by 9x3=27 rows and columns divided into 3x3 squares (one for each component)
- The component squares on the diagonal are important. These are the only squares where we want to see large points.
 - The size of the points indicate the correlation strength

- This looks like a good model
- Score Correlation Matrix IM Score Correlation Matrix - Process one (M1, MOCA cin xmd xin xin xin xmd xin xir
- Big points only in the 3x3 squares on the diagonal



MOCA Analysis - Score variable correlation matrix (optional)

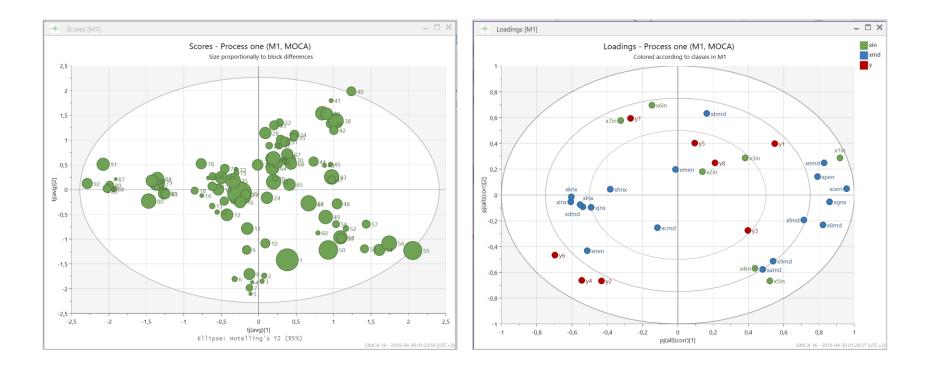
- If external vectors (not part of any block) were included as X-variables we can investigate if they
 are correlated to any of the MOCA score vectors
 - The external vector in this example seems to be correlated to the 5:th joint component





MOCA Analysis - Scores and loading scatter plots

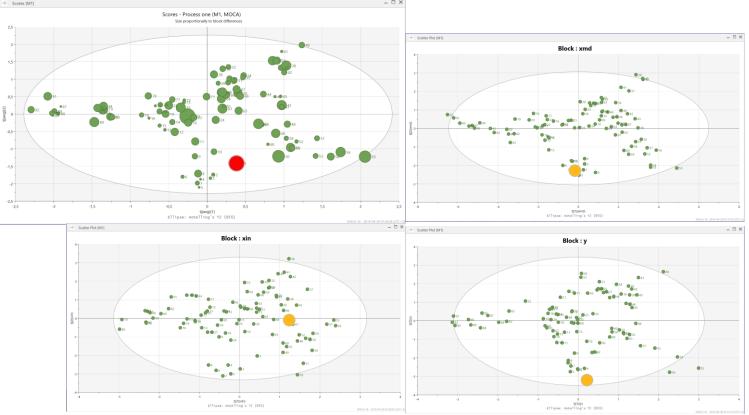
- The score and loadings plots from the MOCA model are interpreted in the same way as normal PCA scatter plots
 - The point location in the MOCA score scatter plot is the average score value for all blocks, and the size reflects how different the location of the point is in the different blocks





MOCA Analysis - Score scatter plots

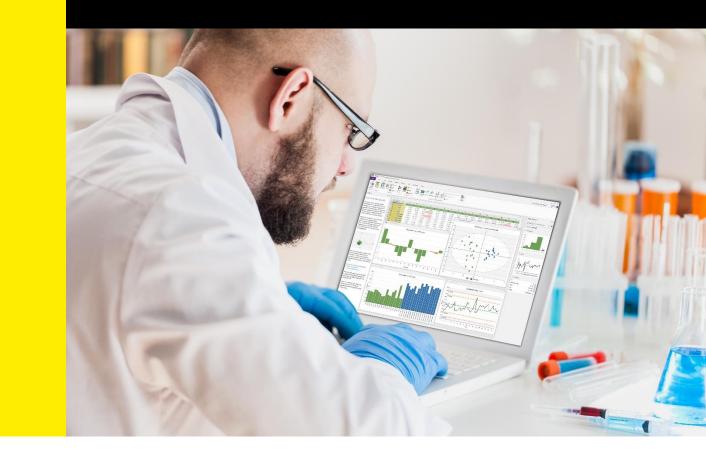
- Comparing the score plot from the MOCA model with the individual block scores
 - The location is almost the same in **xmd** and **y** for point "1" but very different in **xin**. This deviation in should be investigated further





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Simplifying Progress



Python additions and improvements

Python implementation news

- Customer feedback
 - Why do I have to install Python on the side to be able to access other packages (e.g. NumPy)?
 - How can I debug my scripts?
 - I have a spectral preprocessing method that works well in Python. When can that method be implemented in SIMCA?
- With SIMCA 16 we address all of these requests
- Python (3.7) is now fully installed with SIMCA
 - No need to install Python on the side

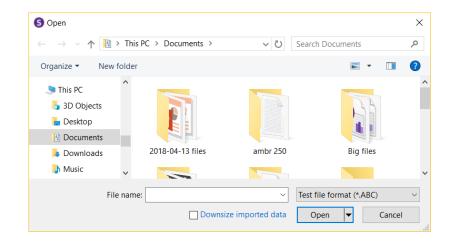


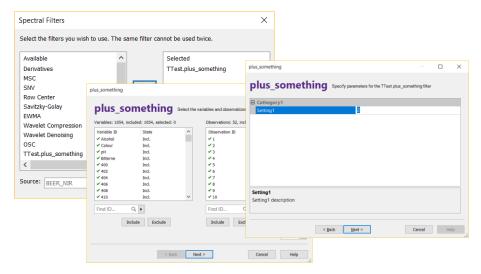
- Users can add additional Python packages directly in the console using PIP
 - Using virtual environments the user can also work with different combination of packages
- Python scripts can be debugged with VS Code



SIMCA Python plugins

- SIMCA supports two types of plugins
 - Users or third party companies can create custom plugins for both file reading and data preprocessing
 - Plugins can be shared with other users
- File reader plugins
 - Adds a new file type to the normal SIMCA file open dialog
- Data preprocessing plugin
 - Adds a new preprocessing choice to the list of implemented preprocessing techniques
 - Works as all the other filters (i.e. also for predictions)
- Plugins make SIMCA very flexible since users can create their own file import and data preprocessing
 - No need to wait for SSDA to implement functionality

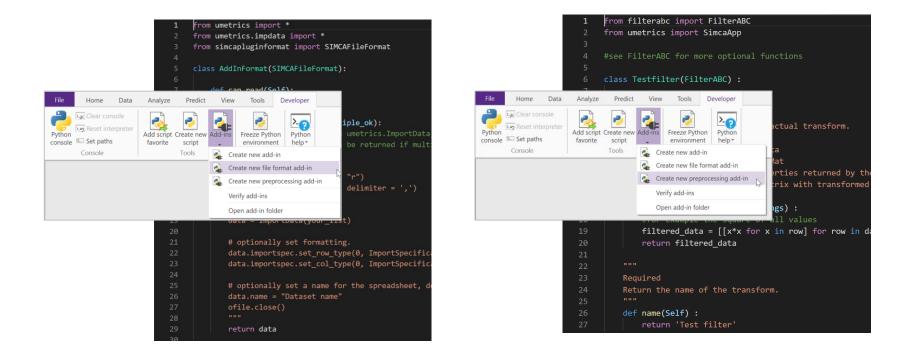






SIMCA Python file and spectral filter plugins

 Initiate the plugin creation in SIMCA and a boilerplate script is prepared for the user to add functional code to





Thank you for your interest in SIMCA® 16

Don't forget to check out instructional videos Search for Sartorius Data Analytics on YouTube

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