



## Development of pellets for oral lysozyme delivery by using a quality by design approach

The objectives of this research paper are to identify how the critical process parameters influence the critical quality attributes of a lysozyme-containing multiparticulate dosage form, and to determine the critical points of API activity preservation. Design of experiments (DoE) methodology is implemented in order to determine the effects of the adjustable critical process parameters.

The factors (independent variables) examined are:  $X_1$  = impeller speed (rpm),  $X_2$  = liquid addition (ml/min),  $X_3$  = extrusion speed (rpm),  $X_4$  = spheronizer speed (rpm) and  $X_5$  = spheronization time (min). All the factors are continuous. The responses (dependent variables) examined are:  $Y_1$  = activity (%),  $Y_2$  = hardness (N) and  $Y_3$  = roundness. The applied DoE method is  $2^5$  full factorial design.

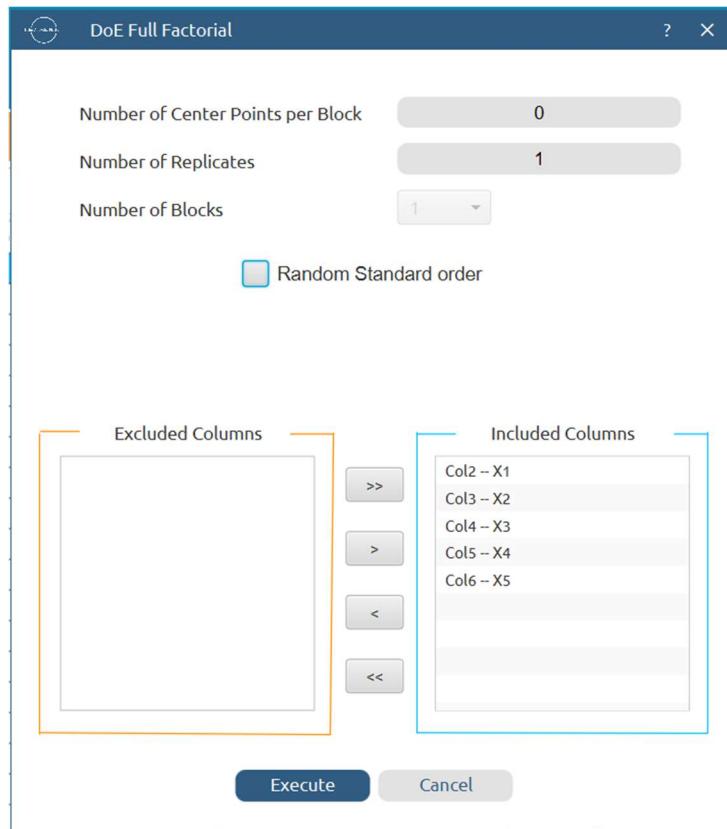
*Isalos version used: 2.0.6*

Scientific article: <https://www.sciencedirect.com/science/article/abs/pii/S0263876215004918>

### Step 1: Full Factorial Design

In the first tab named “Action” define the factors in the column headers and fill each column with the low and high levels of the corresponding factors. This tab can be renamed “Full Factorial”. Afterwards, apply the full factorial method: DOE → Factorial → Full Factorial

	Col1	Col2 (I)	Col3 (I)	Col4 (I)	Col5 (I)	Col6 (I)
User Header	User Row ID	X1	X2	X3	X4	X5
1		500	5	70	1000	15
2		1500	10	120	2000	30



Execute

Cancel

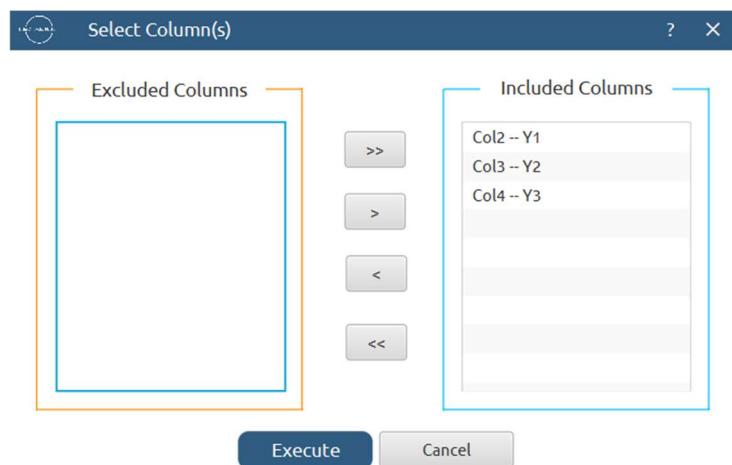
## Results (right spreadsheet):

	Col1	Col2 (I)	Col3 (S)	Col4 (S)	Col5 (S)	Col6 (D)	Col7 (D)	Col8 (D)	Col9 (D)	Col10 (D)
User Header	User Row ID	Standard Order	Block Number	Replicate Number	Point Type	X1	X2	X3	X4	X5
1		1	Block: 1	Replicate: 1	Design Point	500.0	5.0	70.0	1000.0	15.0
2		2	Block: 1	Replicate: 1	Design Point	1500.0	5.0	70.0	1000.0	15.0
3		3	Block: 1	Replicate: 1	Design Point	500.0	10.0	70.0	1000.0	15.0
4		4	Block: 1	Replicate: 1	Design Point	1500.0	10.0	70.0	1000.0	15.0
5		5	Block: 1	Replicate: 1	Design Point	500.0	5.0	120.0	1000.0	15.0
6		6	Block: 1	Replicate: 1	Design Point	1500.0	5.0	120.0	1000.0	15.0
7		7	Block: 1	Replicate: 1	Design Point	500.0	10.0	120.0	1000.0	15.0
8		8	Block: 1	Replicate: 1	Design Point	1500.0	10.0	120.0	1000.0	15.0
9		9	Block: 1	Replicate: 1	Design Point	500.0	5.0	70.0	2000.0	15.0
10		10	Block: 1	Replicate: 1	Design Point	1500.0	5.0	70.0	2000.0	15.0
11		11	Block: 1	Replicate: 1	Design Point	500.0	10.0	70.0	2000.0	15.0
12		12	Block: 1	Replicate: 1	Design Point	1500.0	10.0	70.0	2000.0	15.0
13		13	Block: 1	Replicate: 1	Design Point	500.0	5.0	120.0	2000.0	15.0
14		14	Block: 1	Replicate: 1	Design Point	1500.0	5.0	120.0	2000.0	15.0
15		15	Block: 1	Replicate: 1	Design Point	500.0	10.0	120.0	2000.0	15.0
16		16	Block: 1	Replicate: 1	Design Point	1500.0	10.0	120.0	2000.0	15.0
17		17	Block: 1	Replicate: 1	Design Point	500.0	5.0	70.0	1000.0	30.0
18		18	Block: 1	Replicate: 1	Design Point	1500.0	5.0	70.0	1000.0	30.0
19		19	Block: 1	Replicate: 1	Design Point	500.0	10.0	70.0	1000.0	30.0
20		20	Block: 1	Replicate: 1	Design Point	1500.0	10.0	70.0	1000.0	30.0
21		21	Block: 1	Replicate: 1	Design Point	500.0	5.0	120.0	1000.0	30.0

## Step 2: Definition of response variables

Create a new tab named “Responses” and define the responses in the column headers. Fill each column with the values of the corresponding responses that were observed and make sure the values follow the order of the experiments as given by the full factorial method. Then, select all columns to be transferred to the right spreadsheet: [Data Transformation → Data Manipulation → Select Column\(s\)](#)

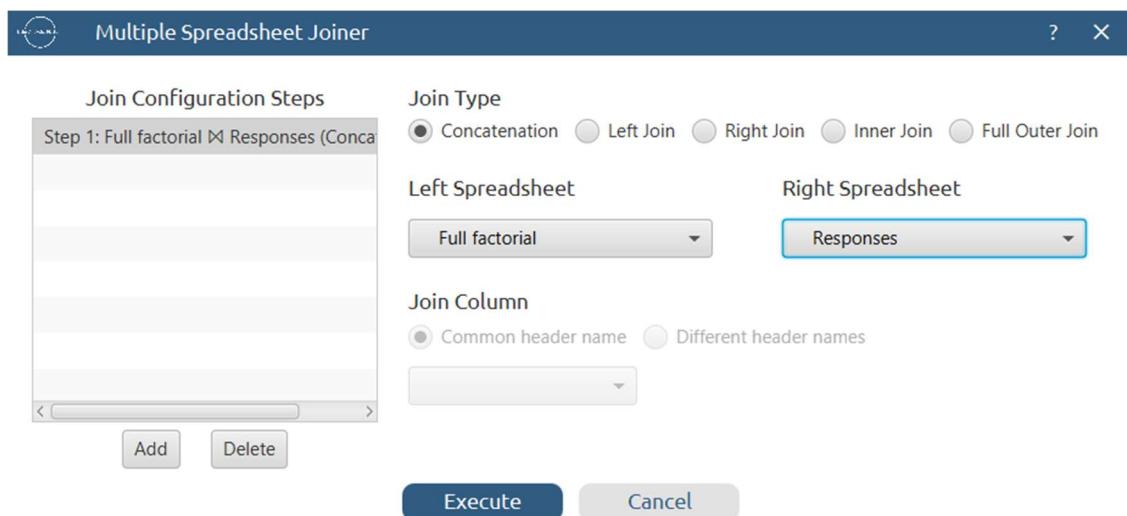
	Col1	Col2 (D)	Col3 (D)	Col4 (D)
User Header	User Row ID	Y1	Y2	Y3
1		68.39378	9.18	1.27
2		69.43005	4.99	1.37
3		96.89119	2.42	1.3
4		70.46632	5.29	1.4
5		71.50259	5.94	1.2
6		79.79275	4.45	1.39
7		84.45596	5.71	1.34
8		67.87565	8.87	1.38
9		70.46632	6.39	1.16
10		75.12953	3.87	1.27
11		71.50259	3.01	1.22
12		83.93782	5.36	1.23
13		77.72021	3.86	1.17
14		81.34715	3.48	1.28
15		65.80311	3.14	1.34
16		47.15026	6.57	1.27
17		72.53886	6.39	1.25
18		62.6943	4.79	1.4
19		86.5285	2.55	1.2
20		56.47668	7.71	1.39
21		67.87565	6.82	1.22



## Step 3: Data isolation

Create a new tab named “Data” and import the results from the “Full Factorial” and “Responses” spreadsheets by right clicking on the left spreadsheet. Then, select only the factors and responses columns to be transferred to the right spreadsheet: Data Transformation → Data Manipulation → Select Column(s)

	Col1	Col2	Col3	Col4	Col5	Col6
User Header	User Row ID					
1						
2						
3						
4						
5						
6						
7						
8						
9						
10						



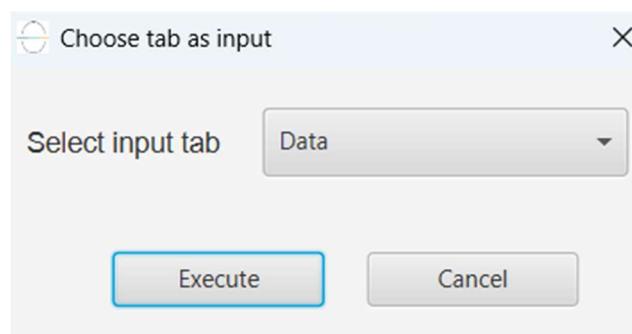
Results:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)	Col8 (D)	Col9 (D)
User Header	User Row ID	X1	X2	X3	X4	X5	Y1	Y2	Y3
1		500.0	5.0	70.0	1000.0	15.0	68.39378	9.18	1.27
2		1500.0	5.0	70.0	1000.0	15.0	69.43005	4.99	1.37
3		500.0	10.0	70.0	1000.0	15.0	96.89119	2.42	1.3
4		1500.0	10.0	70.0	1000.0	15.0	70.46632	5.29	1.4
5		500.0	5.0	120.0	1000.0	15.0	71.50259	5.94	1.2
6		1500.0	5.0	120.0	1000.0	15.0	79.79275	4.45	1.39
7		500.0	10.0	120.0	1000.0	15.0	84.45596	5.71	1.34
8		1500.0	10.0	120.0	1000.0	15.0	67.87565	8.87	1.38
9		500.0	5.0	70.0	2000.0	15.0	70.46632	6.39	1.16
10		1500.0	5.0	70.0	2000.0	15.0	75.12953	3.87	1.27
11		500.0	10.0	70.0	2000.0	15.0	71.50259	3.01	1.22
12		1500.0	10.0	70.0	2000.0	15.0	83.93782	5.36	1.23
13		500.0	5.0	120.0	2000.0	15.0	77.72021	3.86	1.17
14		1500.0	5.0	120.0	2000.0	15.0	81.34715	3.48	1.28
15		500.0	10.0	120.0	2000.0	15.0	65.80311	3.14	1.34
16		1500.0	10.0	120.0	2000.0	15.0	47.15026	6.57	1.27
17		500.0	5.0	70.0	1000.0	30.0	72.53886	6.39	1.25
18		1500.0	5.0	70.0	1000.0	30.0	62.6943	4.79	1.4
19		500.0	10.0	70.0	1000.0	30.0	86.5285	2.55	1.2
20		1500.0	10.0	70.0	1000.0	30.0	56.47668	7.71	1.39
21		500.0	5.0	120.0	1000.0	30.0	67.87565	6.82	1.22
22		1500.0	5.0	120.0	1000.0	30.0	72.53886	3.48	1.42
23		500.0	10.0	120.0	1000.0	30.0	78.75648	2.42	1.29
24		1500.0	10.0	120.0	1000.0	30.0	74.09326	4.9	1.4
25		500.0	5.0	70.0	2000.0	30.0	75.64767	6.48	1.14
26		1500.0	5.0	70.0	2000.0	30.0	77.20207	3.27	1.29
27		500.0	10.0	70.0	2000.0	30.0	94.81865	3.16	1.27
28		1500.0	10.0	70.0	2000.0	30.0	50.7772	6	1.29
29		500.0	5.0	120.0	2000.0	30.0	80.82902	4.24	1.31
30		1500.0	5.0	120.0	2000.0	30.0	84.97409	5.41	1.39
31		500.0	10.0	120.0	2000.0	30.0	62.6943	3.78	1.32
32		1500.0	10.0	120.0	2000.0	30.0	66.83938	7.18	1.35

## Step 4: Normalization

Create a new tab named “Normalized data” and import the results from the “Data” spreadsheet. Afterwards, normalize the factor columns to take values in the range [-1, 1]: [Data Transformation → Normalizers → Min-Max](#)

	Col1	Col2	Col3	Col4	Col5	Col6
User Header	User Row ID					
1						
2						
3						
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10						

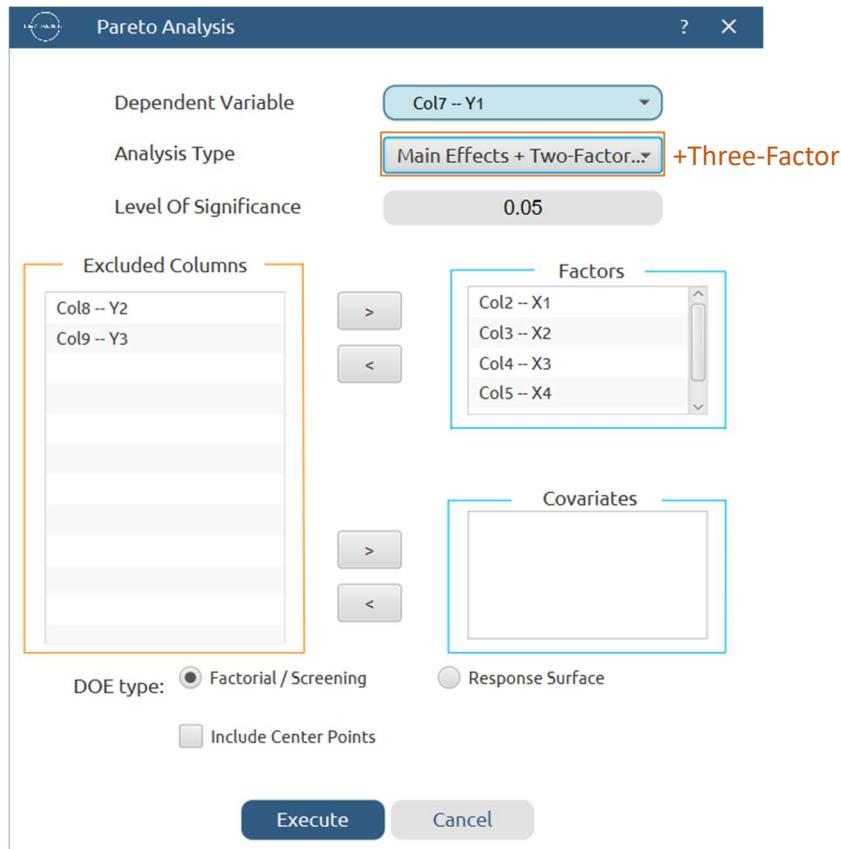


Results:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)	Col8 (D)	Col9 (D)
User Header	User Row ID	X1	X2	X3	X4	X5	Y1	Y2	Y3
1		-1.0	-1.0	-1.0	-1.0	-1.0	68.39378	9.18	1.27
2		1.0	-1.0	-1.0	-1.0	-1.0	69.43005	4.99	1.37
3		-1.0	1.0	-1.0	-1.0	-1.0	96.89119	2.42	1.3
4		1.0	1.0	-1.0	-1.0	-1.0	70.46632	5.29	1.4
5		-1.0	-1.0	1.0	-1.0	-1.0	71.50259	5.94	1.2
6		1.0	-1.0	1.0	-1.0	-1.0	79.79275	4.45	1.39
7		-1.0	1.0	1.0	-1.0	-1.0	84.45596	5.71	1.34
8		1.0	1.0	1.0	-1.0	-1.0	67.87565	8.87	1.38
9		-1.0	-1.0	-1.0	1.0	-1.0	70.46632	6.39	1.16
10		1.0	-1.0	-1.0	1.0	-1.0	75.12953	3.87	1.27
11		-1.0	1.0	-1.0	1.0	-1.0	71.50259	3.01	1.22
12		1.0	1.0	-1.0	1.0	-1.0	83.93782	5.36	1.23
13		-1.0	-1.0	1.0	1.0	-1.0	77.72021	3.86	1.17
14		1.0	-1.0	1.0	1.0	-1.0	81.34715	3.48	1.28
15		-1.0	1.0	1.0	1.0	-1.0	65.80311	3.14	1.34
16		1.0	1.0	1.0	1.0	-1.0	47.15026	6.57	1.27
17		-1.0	-1.0	-1.0	-1.0	1.0	72.53886	6.39	1.25
18		1.0	-1.0	-1.0	-1.0	1.0	62.6943	4.79	1.4
19		-1.0	1.0	-1.0	-1.0	1.0	86.5285	2.55	1.2
20		1.0	1.0	-1.0	-1.0	1.0	56.47668	7.71	1.39
21		-1.0	-1.0	1.0	-1.0	1.0	67.87565	6.82	1.22
22		1.0	-1.0	1.0	-1.0	1.0	72.53886	3.48	1.42
23		-1.0	1.0	1.0	-1.0	1.0	78.75648	2.42	1.29
24		1.0	1.0	1.0	-1.0	1.0	74.09326	4.9	1.4
25		-1.0	-1.0	-1.0	1.0	1.0	75.64767	6.48	1.14
26		1.0	-1.0	-1.0	1.0	1.0	77.20207	3.27	1.29
27		-1.0	1.0	-1.0	1.0	1.0	94.81865	3.16	1.27
28		1.0	1.0	-1.0	1.0	1.0	50.7772	6.0	1.29
29		-1.0	-1.0	1.0	1.0	1.0	80.82902	4.24	1.31
30		1.0	-1.0	1.0	1.0	1.0	84.97409	5.41	1.39
31		-1.0	1.0	1.0	1.0	1.0	62.6943	3.78	1.32
32		1.0	1.0	1.0	1.0	1.0	66.83938	7.18	1.35

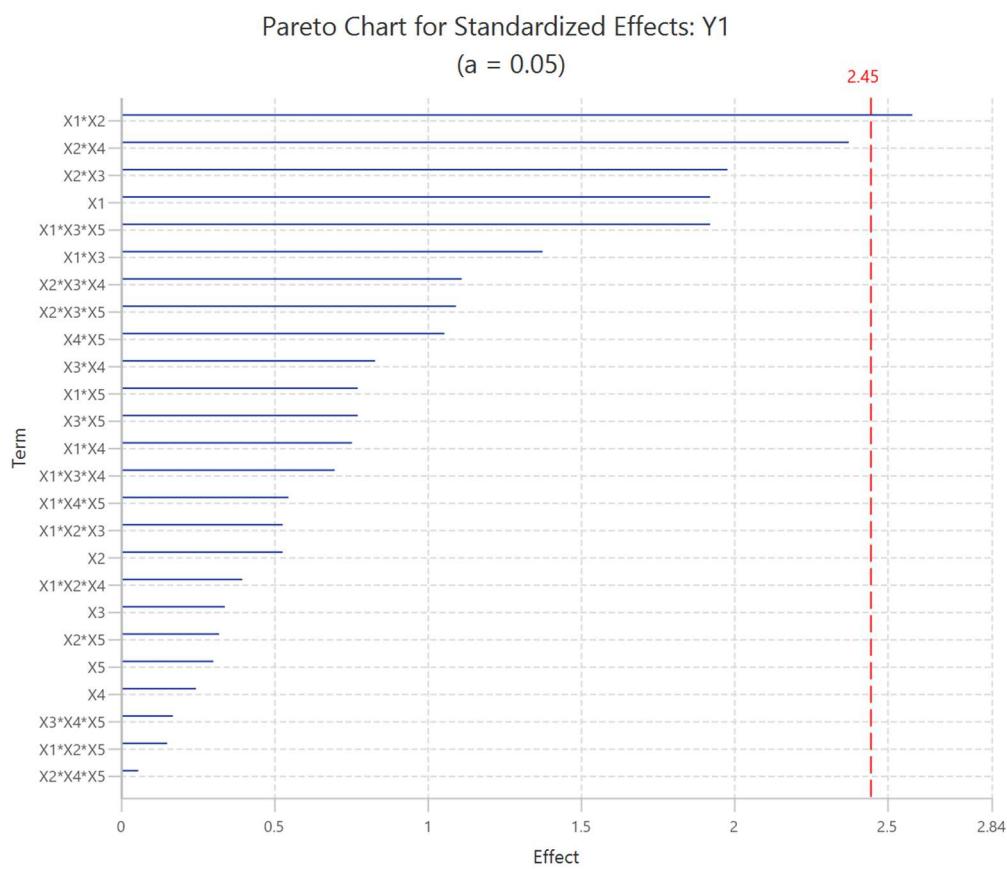
## Step 5: Pareto analysis

Create a new tab named “Pareto Analysis – Y1” and import the results either from the spreadsheet “Data” or “Normalized data”. Then, conduct pareto analysis for the first response variable, Y<sub>1</sub>: DOE → Post DoE Analysis → Pareto Analysis



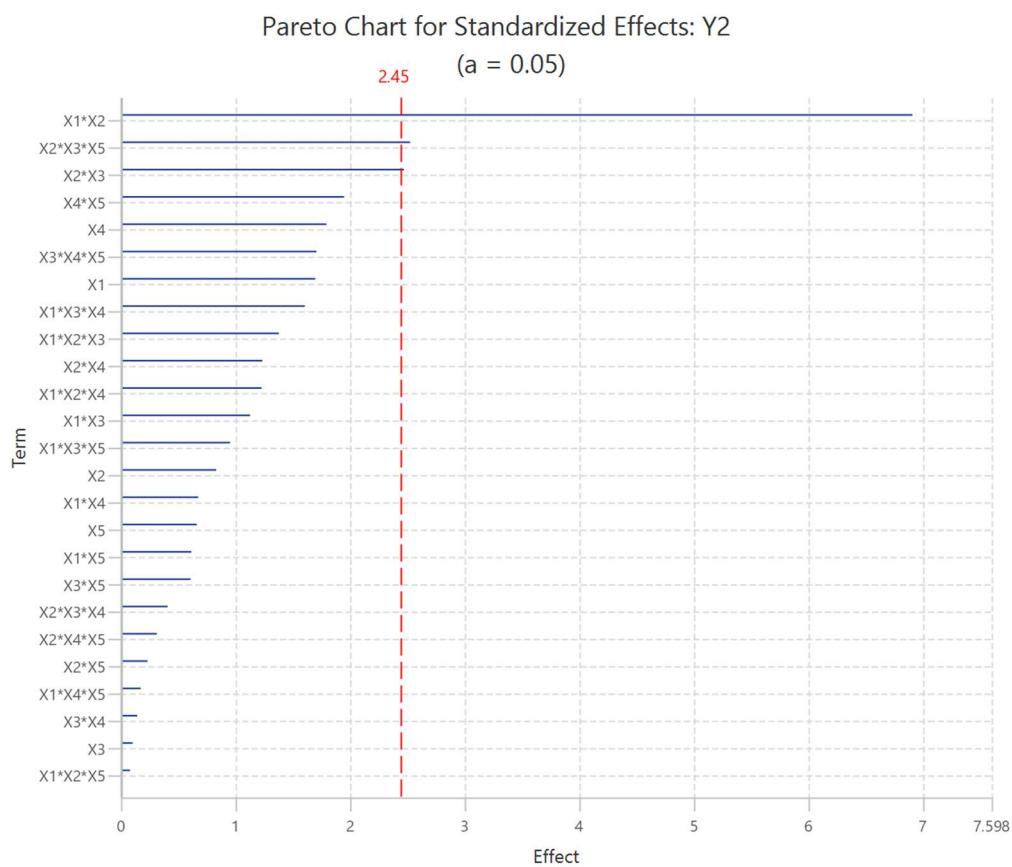
Results:

	Col1	Col2 (S)	Col3 (S)
User Header	User Row ID	Pareto Analysis of :	Standardized Effects
1		Variable	Effect
2		X1*X2	2.5818438
3		X2*X4	2.3745425
4		X2*X3	1.9787854
5		X1	1.9222492
6		X1*X3*X5	1.9222484
7		X1*X3	1.3757270
8		X2*X3*X4	1.1118888
9		X2*X3*X5	1.0930434
10		X4*X5	1.0553522
11		X3*X4	0.8292052
12		X1*X5	0.7726690
13		X3*X5	0.7726683
14		X1*X4	0.7538229
15		X1*X3*X4	0.6972863
16		X1*X4*X5	0.5465213
17		X1*X2*X3	0.5276762
18		X2	0.5276762
19		X1*X2*X4	0.3957577
20		X3	0.3392196
21		X2*X5	0.3203750
22		X5	0.3015292
23		X4	0.2449926
24		X3*X4*X5	0.1696103
25		X1*X2*X5	0.1507645
26		X2*X4*X5	0.0565368
27		Significance Value	2.4469119



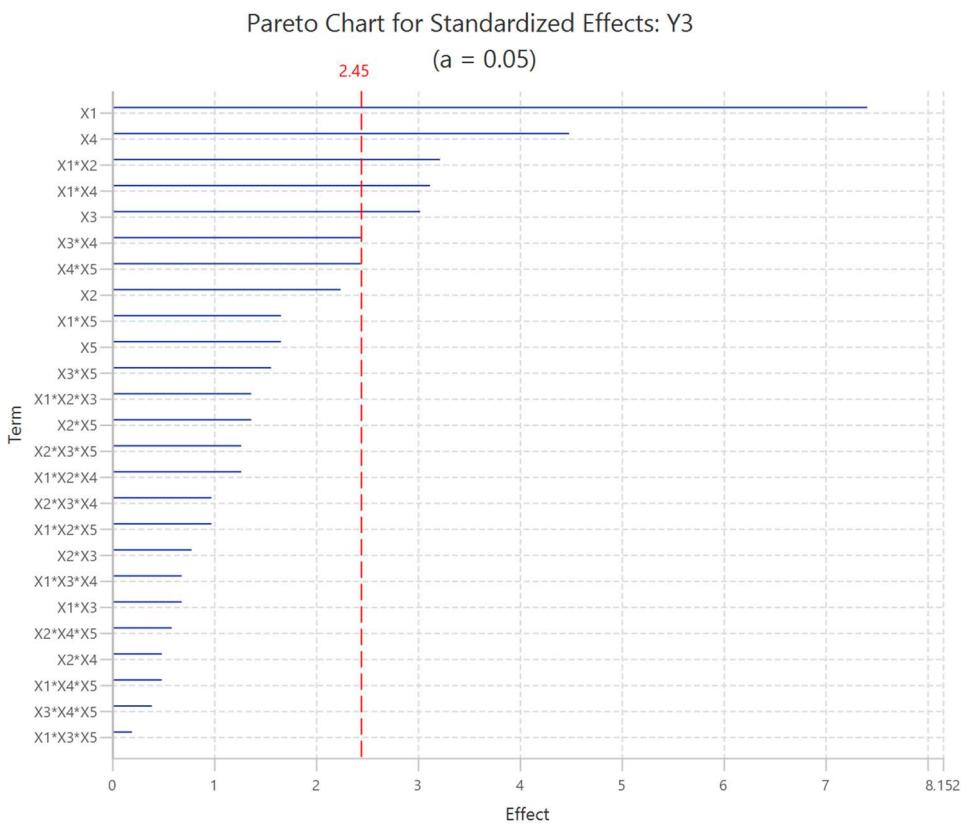
Repeat this step for the rest of the response variables. Results, Y<sub>2</sub>:

	Col1	Col2 (S)	Col3 (S)
User Header	User Row ID	Pareto Analysis of :	Standardized Effects
1		Variable	Effect
2		X1*X2	6.9072096
3		X2*X3*X5	2.5234339
4		X2*X3	2.4698507
5		X4*X5	1.9474145
6		X4	1.7933628
7		X3*X4*X5	1.7062901
8		X1	1.6962432
9		X1*X3*X4	1.6058216
10		X1*X2*X3	1.3780930
11		X2*X4	1.2340881
12		X1*X2*X4	1.2273902
13		X1*X3	1.1269217
14		X1*X3*X5	0.9527763
15		X2	0.8322141
16		X1*X4	0.6748134
17		X5	0.6614176
18		X1*X5	0.6145323
19		X3*X5	0.6078344
20		X2*X3*X4	0.4068974
21		X2*X4*X5	0.3131268
22		X2*X5	0.2327520
23		X1*X4*X5	0.1724709
24		X3*X4	0.1423304
25		X3	0.1021430
26		X1*X2*X5	0.0787003
27		Significance Value	2.4469119



Results, Y<sub>3</sub>:

User Header	Col1	Col2 (S)	Col3 (S)
User Row ID	Pareto Analysis of :	Standardized Effects	
1	Variable	Effect	
2	X1	7.4109612	
3	X4	4.4855818	
4	X1*X2	3.2179174	
5	X1*X4	3.1204047	
6	X3	3.0228921	
7	X3*X4	2.4378162	
8	X4*X5	2.4378162	
9	X2	2.2427909	
10	X1*X5	1.6577150	
11	X5	1.6577150	
12	X3*X5	1.5602024	
13	X1*X2*X3	1.3651771	
14	X2*X5	1.3651771	
15	X2*X3*X5	1.2676644	
16	X1*X2*X4	1.2676644	
17	X2*X3*X4	0.9751265	
18	X1*X2*X5	0.9751265	
19	X2*X3	0.7801012	
20	X1*X3*X4	0.6825885	
21	X1*X3	0.6825885	
22	X2*X4*X5	0.5850759	
23	X2*X4	0.4875632	
24	X1*X4*X5	0.4875632	
25	X3*X4*X5	0.3900506	
26	X1*X3*X5	0.1950253	
27	Significance Value	2.4469119	

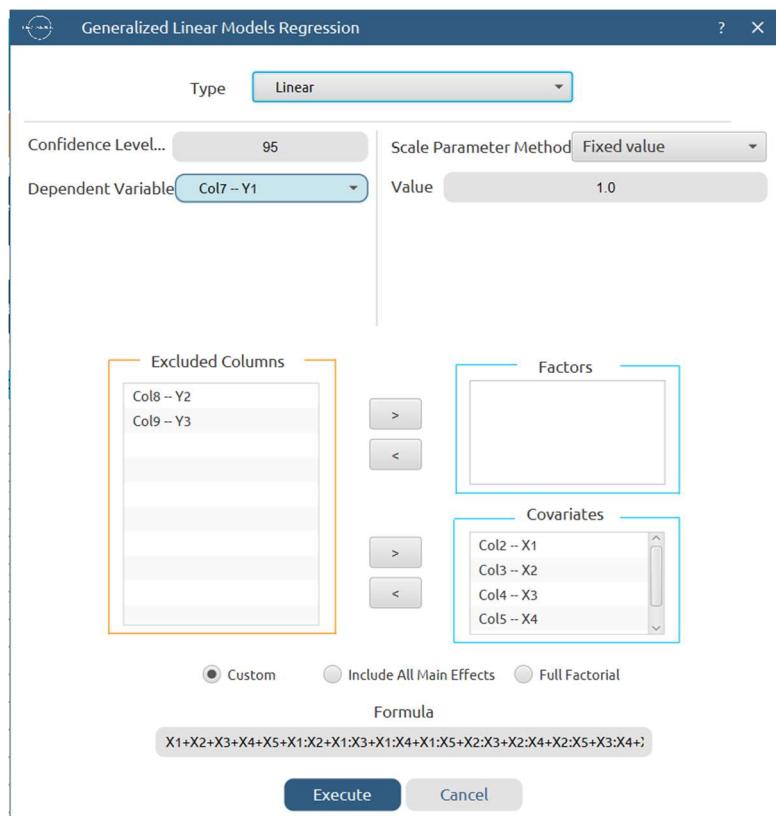


## Step 6: Regression

The goal here is to produce a regression equation that includes main effects, two-factor and three-factor interactions for Y<sub>1</sub>:

$$Y = b_0 + b_1X_1 + b_2X_2 + b_3X_3 + b_4X_4 + b_5X_5 + b_{12}X_1X_2 + b_{13}X_1X_3 + b_{14}X_1X_4 + b_{15}X_1X_5 + b_{23}X_2X_3 + b_{24}X_2X_4 + b_{25}X_2X_5 + b_{34}X_3X_4 + b_{35}X_3X_5 + b_{45}X_4X_5 + b_{123}X_1X_2X_3 + b_{124}X_1X_2X_4 + b_{125}X_1X_2X_5 + b_{134}X_1X_3X_4 + b_{135}X_1X_3X_5 + b_{145}X_1X_4X_5 + b_{234}X_2X_3X_4 + b_{235}X_2X_3X_5 + b_{245}X_2X_4X_5 + b_{345}X_3X_4X_5$$

Create a new tab named “Regression – Y1” and import the results from the spreadsheet “Normalized data”. Afterwards, fit a generalized linear model to the data: [Analytics → Regression → Statistical fitting → Generalized Linear Models](#)



## Results:

User Header	Col1	Col2 (D)	Col3 (D)		
User Row ID	Y1	Prediction		Goodness of Fit	
1	68.39378	67.8108769			
2	69.43005	70.4015537			
3	96.89119	93.7176162			
4	70.46632	73.2512931			
5	71.50259	75.4533675			
6	79.79275	75.4533719			
7	84.45596	84.2616594			
8	67.87565	68.4585512			
9	70.46632	65.8031087			
10	75.12953	79.4041406			
11	71.50259	79.9222781			
12	83.93782	75.9067325			
13	77.72021	79.0155469			
14	81.34715	80.4404137			
15	65.80311	60.7512962			
16	47.15026	51.8134731			
17	72.53886	72.9922288			
18	62.6943	61.8523306			
19	86.5285	89.8316081			
20	56.47668	53.5621725			
21	67.87565	64.0544069			
22	72.53886	76.7487037			
23	78.75648	78.8212462			
24	74.09326	73.6398931			
25	75.64767	80.4404156			
26	77.20207	72.797925			
27	94.81865	86.2694275			
28	50.7772	58.9378219			
29	80.82902	79.4041488			
30	84.97409	86.0103606			
31	62.6943	67.8756481			
32	66.83938	62.0466325			

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
intercept	73.3484453	0.1767767	73.0019694	73.6949213	172159.82175	1	0.0
X1	-3.3031097	0.1767767	-3.6495856	-2.9566337	349.1370754	1	0.0
X2	-0.9067359	0.1767767	-1.2532119	-0.5602600	26.3094419	1	3E-7
X3	-0.5829003	0.1767767	-0.9293763	-0.2364244	10.8727288	1	0.0009759
X4	-0.4209847	0.1767767	-0.7674606	-0.0745087	5.6712994	1	0.0172447
X5	-0.5181347	0.1767767	-0.8646106	-0.1716587	8.5908337	1	0.0033786
X1*X5	-1.3277209	0.1767767	-1.6741969	-0.9812450	56.4109724	1	0E-7
X1*X4	1.2953366	0.1767767	0.9486606	1.6418125	53.6926979	1	0E-7
X2*X5	-0.5505184	0.1767767	-0.8969944	-0.2040425	9.6982576	1	0.0018444
X1*X3	2.3639897	0.1767767	2.0175137	2.7104656	178.8303118	1	0.0
X2*X4	-4.0803109	0.1767767	-4.4267869	-3.7338350	532.7659951	1	0.0
X3*X5	1.3277197	0.1767767	0.9812437	1.6741956	56.4108662	1	0E-7
X1*X2	-4.4365284	0.1767767	-4.7830044	-4.0900525	629.8491065	1	0.0
X2*X3	-3.4002591	0.1767767	-3.7467350	-3.0537831	369.9763741	1	0.0
X3*X4	-1.4246703	0.1767767	-1.7713463	-1.0783944	64.9681730	1	0E-7
X4*X5	1.8134716	0.1767767	1.4669956	2.1599475	105.2377315	1	0.0
X2*X3*X4	-1.9106216	0.1767767	-2.2570975	-1.5641456	116.8151922	1	0.0
X2*X4*X5	0.0971503	0.1767767	-0.2493256	0.4436263	0.3020219	1	0.5826177
X2*X3*X5	1.8782384	0.1767767	1.5317625	2.2247144	112.8889481	1	0.0
X1*X2*X4	0.6800528	0.1767767	0.3335769	1.0265288	14.7990985	1	0.0001196
X1*X2*X3	0.9067359	0.1767767	0.5602600	1.2532119	26.3094419	1	3E-7
X1*X3*X4	-1.1981866	0.1767767	-1.5446625	-0.8517106	45.9408332	1	0E-7
X1*X4*X5	-0.9391184	0.1767767	-1.2855944	-0.5926425	28.2221901	1	1E-7
X1*X2*X5	-0.2590672	0.1767767	-0.6055431	0.0874088	2.1477058	1	0.1427831
X1*X3*X5	3.3031084	0.1767767	2.9566325	3.6495844	349.1368112	1	0.0
X3*X4*X5	0.2914509	0.1767767	-0.0550250	0.6379269	2.7181968	1	0.0992100

Repeat this step for the rest of the response variables. Results, Y<sub>2</sub>:

User Header	Col1	Col2 (D)	Col3 (D)		
User Row ID	Y2	Prediction		Goodness of Fit	
1		9.18	8.5668750		
2		4.99	4.9562500		
3		2.42	2.4625000		
4		5.29	5.8943750		
5		5.94	6.6325000		
6		4.45	4.404375		
7		5.71	5.5881250		
8		8.87	8.3450000		
9		6.39	7.0837500		
10		3.87	3.8231250		
11		3.01	2.8868750		
12		5.36	4.8362500		
13		3.86	3.0868750		
14		3.48	3.6062500		
15		3.14	3.3425000		
16		6.57	7.0143750		
17		6.39	7.0987500		
18		4.79	4.7281250		
19		2.55	2.4118750		
20		7.71	7.2012500		
21		6.82	6.0318750		
22		3.48	3.6212500		
23		2.42	2.6375000		
24		4.9	5.3293750		
25		6.48	5.6906250		
26		3.27	3.4125000		
27		3.16	3.3787500		
28		6.0	6.4281250		
29		4.24	5.1087500		
30		5.41	5.1881250		
31		3.78	3.4818750		
32		7.18	6.8312500		

Goodness of Fit							
Value							
Deviance	6.6871937						
Scaled Deviance	6.6871937						
Pearson Chi-Square	6.6871937						
Scaled Pearson Chi-Square	6.6871937						
Log Likelihood	-32.7496299						
Akaike's Information Criterion (AIC)	117.4992599						
Finite Sample Corrected AIC (AICC)	398.2992599						
Bayesian Information Criterion (BIC)	155.6083933						
Consistent AIC (CAIC)	181.6083933						

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
intercept	5.0346875	0.1767767	4.6882115	5.3811635	811.1385031	1	0.0
X1	0.3165625	0.1767767	-0.0299135	0.6630385	3.2067781	1	0.0733338
X2	-0.1553125	0.1767767	-0.5017885	0.1911635	0.7719031	1	0.3796289
X3	-0.0190625	0.1767767	-0.3655385	0.3274135	0.0116281	1	0.9141275
X4	-0.3346875	0.1767767	-0.6811635	0.0117885	3.5845031	1	0.0583209
X5	-0.1234375	0.1767767	-0.4699135	0.2230385	0.4875781	1	0.4850096
X1*X5	0.1146875	0.1767767	-0.2317885	0.4611635	0.4209031	1	0.5164867
X1*X4	0.1259375	0.1767767	-0.2205385	0.4724135	0.5075281	1	0.4762109
X2*X5	-0.0434375	0.1767767	-0.3899135	0.3030385	0.0603781	1	0.8058993
X1*X3	0.2103125	0.1767767	-0.1361635	0.5567885	1.4154031	1	0.2341615
X2*X4	0.2303125	0.1767767	-0.1161635	0.5767885	1.6974031	1	0.1926279
X3*X5	-0.1134375	0.1767767	-0.4599135	0.2330385	0.4117781	1	0.5210684
X1*X2	1.2890625	0.1767767	0.9425865	1.6355385	53.1738281	1	0E-7
X2*X3	0.4609375	0.1767767	0.1144615	0.8074135	6.7988281	1	0.0091218
X3*X4	0.0265625	0.1767767	-0.3199135	0.3730385	0.0225781	1	0.8805593
X4*X5	0.3634375	0.1767767	0.0169615	0.7099135	4.2267781	1	0.0397909
X2*X3*X4	-0.0759375	0.1767767	-0.4224135	0.2705385	0.1845281	1	0.6675104
X2*X4*X5	0.0584375	0.1767767	-0.2880385	0.4049135	0.1092781	1	0.7409675
X2*X3*X5	-0.4709375	0.1767767	-0.8174135	-0.1244615	7.0970281	1	0.0077212
X1*X2*X4	-0.2290625	0.1767767	-0.5755385	0.1174135	1.6790281	1	0.1950536
X1*X2*X3	-0.2571875	0.1767767	-0.6036635	0.0892885	2.1166531	1	0.1457046
X1*X3*X4	0.2996875	0.1767767	-0.0467885	0.6461635	2.8740031	1	0.0900207
X1*X4*X5	-0.0321875	0.1767767	-0.3786635	0.3142885	0.0331531	1	0.8555199
X1*X2*X5	0.0146875	0.1767767	-0.3317885	0.3611635	0.0069031	1	0.9337839
X1*X3*X5	-0.1778125	0.1767767	-0.5242885	0.1686635	1.0117531	1	0.3144832
X3*X4*X5	0.3184375	0.1767767	-0.0280385	0.6649135	3.2448781	1	0.0716470

Results, Y<sub>3</sub>:

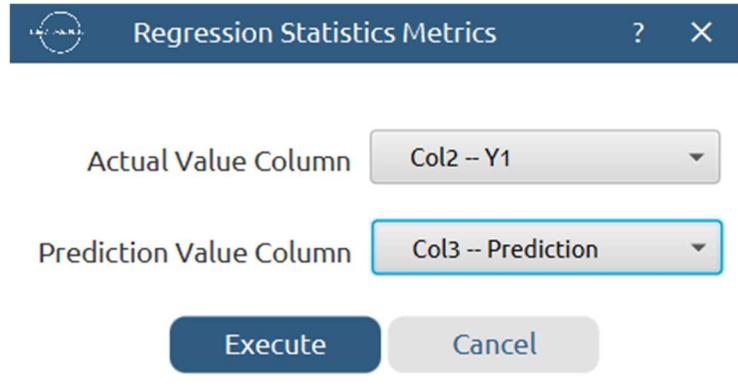
User Header	Col1	Col2 (D)	Col3 (D)		
User Row ID	Y3	Prediction		Goodness of Fit	
1		1.27	1.2681250		
2		1.37	1.3931250		
3		1.3	1.2968750		
4		1.4	1.3818750		
5		1.2	1.2018750		
6		1.39	1.366875		
7		1.34	1.3431250		
8		1.38	1.398125		
9		1.16	1.1506250		
10		1.27	1.258125		
11		1.22	1.2343750		
12		1.23	1.2368750		
13		1.17	1.1793750		
14		1.28	1.2918750		
15		1.34	1.3256250		
16		1.27	1.2631250		
17		1.25	1.2343750		
18		1.4	1.3943750		
19		1.2	1.2206250		
20		1.39	1.3906250		
21		1.22	1.235625		
22		1.42	1.4245625		
23		1.29	1.2693750		
24		1.4	1.3993750		
25		1.14	1.1668750		
26		1.29	1.284375		
27		1.27	1.238125		
28		1.29	1.3006250		
29		1.31	1.2831250		
30		1.39	1.3956250		
31		1.32	1.351875		
32		1.35	1.33993750		

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
Intercept	1.2975000	0.1767767	0.9510240	1.6439760	53.8722000	1	0E-7
X1	0.0475000	0.1767767	-0.2989760	0.3939760	0.0722000	1	0.7881601
X2	0.0143750	0.1767767	-0.3321010	0.3608510	0.0066125	1	0.9351896
X3	0.0193750	0.1767767	-0.3271010	0.3658510	0.0120125	1	0.9127254
X4	-0.0287500	0.1767767	-0.3752260	0.3177260	0.0264500	1	0.8708062
X5	0.0106250	0.1767767	-0.3358510	0.3571010	0.0036125	1	0.9520727
X1*X5	0.0106250	0.1767767	-0.3358510	0.3571010	0.0036125	1	0.9520727
X1*X4	-0.0200000	0.1767767	-0.3664760	0.3264760	0.0128000	1	0.9099219
X2*X5	-0.0087500	0.1767767	-0.3552260	0.3377260	0.0024500	1	0.9605228
X1*X3	-0.0043750	0.1767767	-0.3508510	0.3421010	0.0006125	1	0.9802554
X2*X4	0.0031250	0.1767767	-0.3433510	0.3496010	0.0003125	1	0.9858960
X3*X5	0.0100000	0.1767767	-0.3364760	0.3564760	0.0032000	1	0.9548889
X1*X2	-0.0206250	0.1767767	-0.3671010	0.3258510	0.0136125	1	0.9071195
X2*X3	0.0050000	0.1767767	-0.3414760	0.3514760	0.0008000	1	0.9774354
X3*X4	0.0156250	0.1767767	-0.3308510	0.3621010	0.0078125	1	0.9295680
X4*X5	0.0156250	0.1767767	-0.3308510	0.3621010	0.0078125	1	0.9295680
X2*X3*X4	-0.0062500	0.1767767	-0.3527260	0.3402260	0.0012500	1	0.9717964
X2*X4*X5	0.0037500	0.1767767	-0.3427260	0.3502260	0.0004500	1	0.9830756
X2*X3*X5	-0.0081250	0.1767767	-0.3546010	0.3383510	0.0021125	1	0.9633406
X1*X2*X4	-0.0081250	0.1767767	-0.3546010	0.3383510	0.0021125	1	0.9633406
X1*X2*X3	-0.0087500	0.1767767	-0.3552260	0.3377260	0.0024500	1	0.9605228
X1*X3*X4	-0.0043750	0.1767767	-0.3508510	0.3421010	0.0006125	1	0.9802554
X1*X4*X5	-0.0031250	0.1767767	-0.3496010	0.3433510	0.0003125	1	0.9858960
X1*X2*X5	0.0062500	0.1767767	-0.3402260	0.3527260	0.0012500	1	0.9717964
X1*X3*X5	-0.0012500	0.1767767	-0.3477260	0.3452260	0.0005000	1	0.9943582
X3*X4*X5	0.0025000	0.1767767	-0.3439760	0.3489760	0.0002000	1	0.9887166

## Step 7: Regression Metrics

Create a tab named “Metrics – Y1” and import the results from the spreadsheet “Regression – Y1”. Then, produce the regression metrics for the Y<sub>1</sub> regression equation: Statistics → Model Metrics → Regression Metrics



Results:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)
User Header	User Row ID	Mean Squared Error	Root Mean Squared Error	Mean Absolute Error	R Squared
1		17.7164979	4.2090970	3.3840673	0.8477895

Repeat this step for the rest of the response variables. Results, Y<sub>2</sub>:

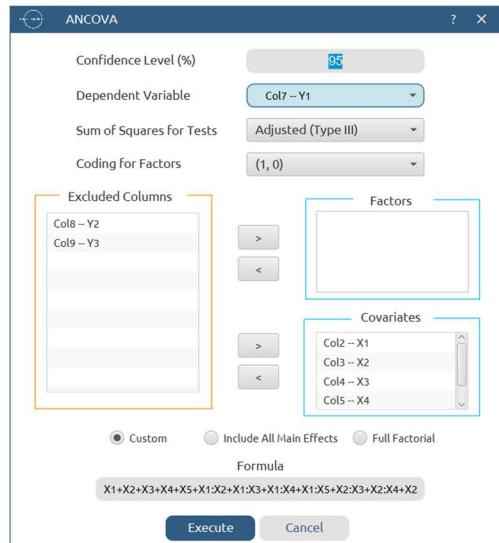
	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)
User Header	User Row ID	Mean Squared Error	Root Mean Squared Error	Mean Absolute Error	R Squared
1		0.2089748	0.4571376	0.3725781	0.9343377

Results, Y<sub>3</sub>:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)
User Header	User Row ID	Mean Squared Error	Root Mean Squared Error	Mean Absolute Error	R Squared
1		0.0002465	0.0156998	0.0128906	0.9590472

## Step 8: Analysis of Covariance

Create a new tab named “ANCOVA – Y1” and import the results from the spreadsheet “Normalized data”. Afterwards perform analysis of covariance for Y<sub>1</sub>: Statistics → Analysis of (Co)Variance → ANCOVA



Results:

	Col1	Col2 (S)	Col3 (I)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)
User Header	User Row ID	Source	DF	Adj SS	Adj MS	F-Value	P-Value
1		X1	1	349.1370754	349.1370754	3.6950419	0.1029446
2		X2	1	26.3094419	26.3094419	0.2784422	0.6166594
3		X3	1	10.8727288	10.8727288	0.1150700	0.7460022
4		X4	1	5.6712994	5.6712994	0.0600214	0.8146261
5		X5	1	8.5908337	8.5908337	0.0909198	0.7731884
6		X1*X2	1	629.8491065	629.8491065	6.6659172	0.0416661
7		X1*X3	1	178.8303118	178.8303118	1.8926248	0.2180538
8		X1*X4	1	53.6926979	53.6926979	0.5682489	0.4794826
9		X1*X5	1	56.4109724	56.4109724	0.5970174	0.4690523
10		X2*X3	1	369.9763741	369.9763741	3.9155916	0.0951840
11		X2*X4	1	532.7659951	532.7659951	5.6384521	0.0551775
12		X2*X5	1	9.6982576	9.6982576	0.1026401	0.7595478
13		X3*X4	1	64.9681730	64.9681730	0.6875813	0.4387244
14		X3*X5	1	56.4108662	56.4108662	0.5970163	0.4690527
15		X4*X5	1	105.2377315	105.2377315	1.1137684	0.3318946
16		X1*X2*X3	1	26.3094419	26.3094419	0.2784422	0.6166594
17		X1*X2*X4	1	14.7990985	14.7990985	0.1566241	0.7059727
18		X1*X2*X5	1	2.1477058	2.1477058	0.0227299	0.8851024
19		X1*X3*X4	1	45.9408332	45.9408332	0.4862082	0.5117232
20		X1*X3*X5	1	349.1368112	349.1368112	3.6950391	0.1029447
21		X1*X4*X5	1	28.2221901	28.2221901	0.2986855	0.6044205
22		X2*X3*X4	1	116.8151922	116.8151922	1.2362967	0.3087421
23		X2*X3*X5	1	112.8889481	112.8889481	1.1947439	0.3163060
24		X2*X4*X5	1	0.3020219	0.3020219	0.0031964	0.9567499
25		X3*X4*X5	1	2.7181968	2.7181968	0.0287676	0.8708907
26		Error	6	566.9279329	94.4879888		
27		Total	31	3724.6302380			

Repeat this step for the rest of the response variables. Results, Y<sub>2</sub>:

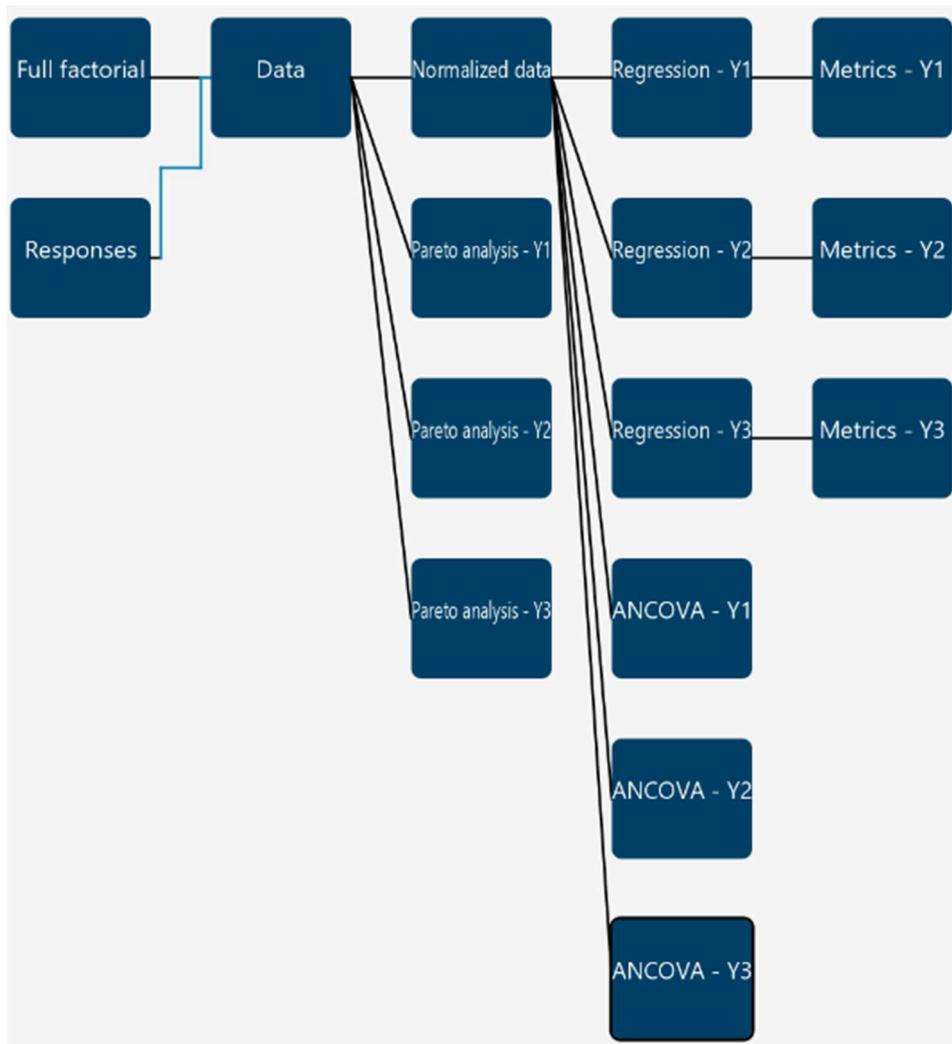
	Col1	Col2 (S)	Col3 (I)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)
User Header	User Row ID	Source	DF	Adj SS	Adj MS	F-Value	P-Value
1		X1	1	3.2067781	3.2067781	2.8772411	0.1407678
2		X2	1	0.7719031	0.7719031	0.6925803	0.4371510
3		X3	1	0.0116281	0.0116281	0.0104332	0.9219713
4		X4	1	3.5845031	3.5845031	3.2161501	0.1230785
5		X5	1	0.4875781	0.4875781	0.4374733	0.5329063
6		X1*X2	1	53.1738281	53.1738281	47.7095446	0.0004552
7		X1*X3	1	1.4154031	1.4154031	1.2699525	0.3028173
8		X1*X4	1	0.5075281	0.5075281	0.4553732	0.5249297
9		X1*X5	1	0.4209031	0.4209031	0.3776500	0.5614282
10		X2*X3	1	6.7988281	6.7988281	6.1001625	0.0484675
11		X2*X4	1	1.6974031	1.6974031	1.5229735	0.2633120
12		X2*X5	1	0.0603781	0.0603781	0.0541735	0.8236905
13		X3*X4	1	0.0225781	0.0225781	0.0202579	0.8914782
14		X3*X5	1	0.4117781	0.4117781	0.3694627	0.5655783
15		X4*X5	1	4.2267781	4.2267781	3.7924232	0.0994147
16		X1*X2*X3	1	2.1166531	2.1166531	1.8991402	0.2173611
17		X1*X2*X4	1	1.6790281	1.6790281	1.5064867	0.2656438
18		X1*X2*X5	1	0.0069031	0.0069031	0.0061937	0.9398301
19		X1*X3*X4	1	2.8740031	2.8740031	2.5786629	0.1594361
20		X1*X3*X5	1	1.0117531	1.0117531	0.9077827	0.3774926
21		X1*X4*X5	1	0.0331531	0.0331531	0.0297462	0.8687380
22		X2*X3*X4	1	0.1845281	0.1845281	0.1655655	0.6982007
23		X2*X3*X5	1	7.0970281	7.0970281	6.3677187	0.0450775
24		X2*X4*X5	1	0.1092781	0.1092781	0.0980484	0.7647831
25		X3*X4*X5	1	3.2448781	3.2448781	2.9114259	0.1388287
26		Error	6	6.6871937	1.1145323		
27		Total	31	101.8421969			

Results, Y<sub>3</sub>:

	Col1	Col2 (S)	Col3 (I)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)
User Header	User Row ID	Source	DF	Adj SS	Adj MS	F-Value	P-Value
1		X1	1	0.0722000	0.0722000	54.9223455	0.0003102
2		X2	1	0.0066125	0.0066125	5.0301109	0.0660930
3		X3	1	0.0120125	0.0120125	9.1378764	0.0233102
4		X4	1	0.0264500	0.0264500	20.1204437	0.0041675
5		X5	1	0.0036125	0.0036125	2.7480190	0.1484497
6		X1*X2	1	0.0136125	0.0136125	10.3549921	0.0181848
7		X1*X3	1	0.0006125	0.0006125	0.4659271	0.5203357
8		X1*X4	1	0.0128000	0.0128000	9.7369255	0.0205743
9		X1*X5	1	0.0036125	0.0036125	2.7480190	0.1484497
10		X2*X3	1	0.0008000	0.0008000	0.6085578	0.4649829
11		X2*X4	1	0.0003125	0.0003125	0.2377179	0.6431610
12		X2*X5	1	0.0024500	0.0024500	1.8637084	0.2211672
13		X3*X4	1	0.0078125	0.0078125	5.9429477	0.0506217
14		X3*X5	1	0.0032000	0.0032000	2.4342314	0.1697311
15		X4*X5	1	0.0078125	0.0078125	5.9429477	0.0506217
16		X1*X2*X3	1	0.0024500	0.0024500	1.8637084	0.2211672
17		X1*X2*X4	1	0.0021125	0.0021125	1.6069731	0.2518901
18		X1*X2*X5	1	0.0012500	0.0012500	0.9508716	0.3671567
19		X1*X3*X4	1	0.0006125	0.0006125	0.4659271	0.5203357
20		X1*X3*X5	1	0.0000500	0.0000500	0.0380349	0.8518095
21		X1*X4*X5	1	0.0003125	0.0003125	0.2377179	0.6431610
22		X2*X3*X4	1	0.0012500	0.0012500	0.9508716	0.3671567
23		X2*X3*X5	1	0.0021125	0.0021125	1.6069731	0.2518901
24		X2*X4*X5	1	0.0004500	0.0004500	0.3423138	0.5798183
25		X3*X4*X5	1	0.0002000	0.0002000	0.1521395	0.7099696
26		Error	6	0.0078875	0.0013146		
27		Total	31	0.1926000			

## Final Isalos Workflow

The final workflow is presented below:



## References

- (1) Sovány, T.; Csordás, K.; Kelemen, A.; Regdon, G.; Pintye-Hódi, K. Development of Pellets for Oral Lysozyme Delivery by Using a Quality by Design Approach. *Chemical Engineering Research and Design* **2016**, 106, 92–100. <https://doi.org/10.1016/j.cherd.2015.11.022>.