



Development and evaluation of metronidazole loaded microsponge based gel for superficial surgical wound infections

The purpose of this research paper is to optimize, characterize and develop a microsponge drug delivery system for topical delivery. Design of experiments (DoE) methodology is implemented for the statistical optimization process.

The factors (independent variables) examined are: X_1 = amount of dichloromethane (mL), X_2 = amount of xanthan gum (% w/v) and X_3 = amount of sodium chloride (% w/v). All the factors are continuous. The responses (dependent variables) examined are: Y_1 = drug entrapment efficiency (%), Y_2 = particle size (μm) and Y_3 = cumulative permeability drug release at 8 h (%). The applied DoE method is Box Behnken design.

Isalos version used: 2.0.6

Scientific article: <https://www.sciencedirect.com/science/article/abs/pii/S1773224715300162>

Step 1: Box Behnken 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 “Box Behnken”. Afterwards, apply the Box Behnken method: *DOE → Response Surface → Box Behnken*

	Col1	Col2 (I)	Col3 (D)	Col4 (D)
User Header	User Row ID	X1	X2	X3
1		10	0.1	0.05
2		50	1	0.1

DoE Box Behnken

Number of Center Points per Block: 5

Number of Replicates: 1

Number of Blocks: 1

Random Standard order

Excluded Columns

Included Columns

Col2 - X1
Col3 - X2
Col4 - X3

>>
>
<
<<

Execute Cancel

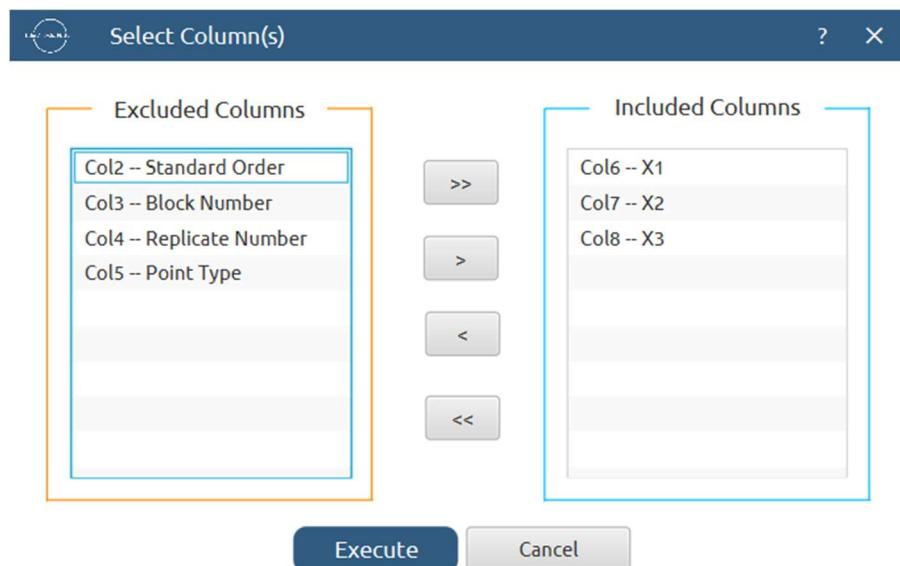
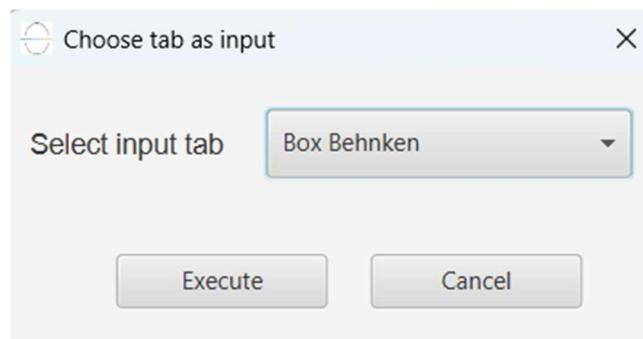
Results (right spreadsheet):

	Col1	Col2 (I)	Col3 (S)	Col4 (S)	Col5 (S)	Col6 (D)	Col7 (D)	Col8 (D)
User Header	User Row ID	Standard Order	Block Number	Replicate Number	Point Type	X1	X2	X3
1		1	Block: 1	Replicate: 1	Design Point	10.0	0.1	0.0750000
2		2	Block: 1	Replicate: 1	Design Point	50.0	0.1	0.0750000
3		3	Block: 1	Replicate: 1	Design Point	10.0	1.0	0.0750000
4		4	Block: 1	Replicate: 1	Design Point	50.0	1.0	0.0750000
5		5	Block: 1	Replicate: 1	Design Point	10.0	0.55	0.05
6		6	Block: 1	Replicate: 1	Design Point	50.0	0.55	0.05
7		7	Block: 1	Replicate: 1	Design Point	10.0	0.55	0.1
8		8	Block: 1	Replicate: 1	Design Point	50.0	0.55	0.1
9		9	Block: 1	Replicate: 1	Design Point	30.0	0.1	0.05
10		10	Block: 1	Replicate: 1	Design Point	30.0	1.0	0.05
11		11	Block: 1	Replicate: 1	Design Point	30.0	0.1	0.1
12		12	Block: 1	Replicate: 1	Design Point	30.0	1.0	0.1
13		13	Block: 1	----	Center Point	30.0	0.55	0.0750000
14		14	Block: 1	----	Center Point	30.0	0.55	0.0750000
15		15	Block: 1	----	Center Point	30.0	0.55	0.0750000
16		16	Block: 1	----	Center Point	30.0	0.55	0.0750000
17		17	Block: 1	----	Center Point	30.0	0.55	0.0750000

Step 2: Factor isolation

Create a new tab named “Factors” and import the results from the “Box Behnken” spreadsheet by right clicking on the left spreadsheet. Then, select only the factor 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							



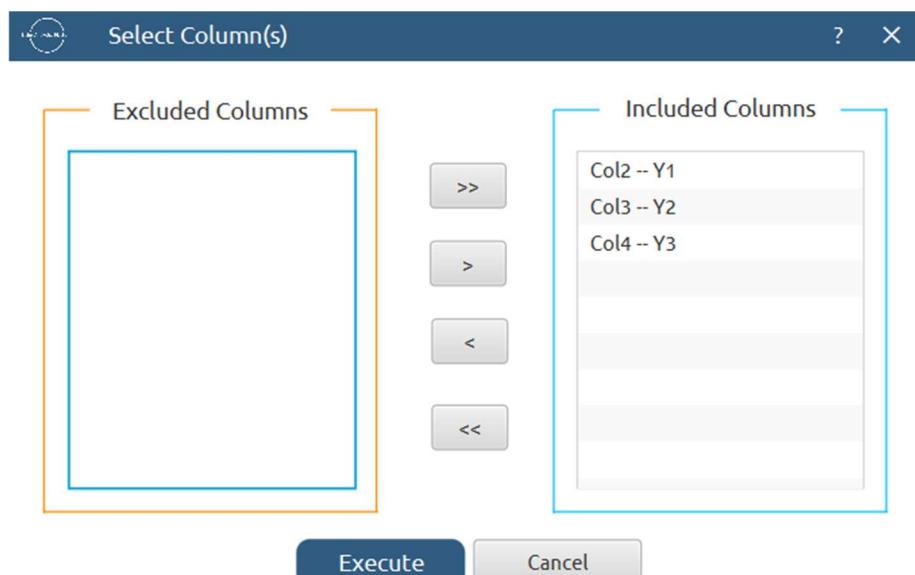
Results:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)
User Header	User Row ID	X1	X2	X3
1		10.0	0.1	0.0750000
2		50.0	0.1	0.0750000
3		10.0	1.0	0.0750000
4		50.0	1.0	0.0750000
5		10.0	0.55	0.05
6		50.0	0.55	0.05
7		10.0	0.55	0.1
8		50.0	0.55	0.1
9		30.0	0.1	0.05
10		30.0	1.0	0.05
11		30.0	0.1	0.1
12		30.0	1.0	0.1
13		30.0	0.55	0.0750000
14		30.0	0.55	0.0750000
15		30.0	0.55	0.0750000
16		30.0	0.55	0.0750000
17		30.0	0.55	0.0750000

Step 3: 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 Box Behnken design. 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		59.6	0.0000267	6.7268120
2		17.71	0.0000605	4.7822589
3		46.85	0.0000332	4.4855323
4		19.59	0.0000461	3.3630343
5		61.62	0.0035391	4.2107007
6		14.36	0.0000350	4.0074930
7		53.81	0.0001144	4.2107007
8		19.09	0.0000583	4.0681691
9		67.54	0.0002838	7.4020267
10		23.99	0.0000834	4.6043458
11		19.02	0.0001990	5.2028838
12		19.98	0.0000700	4.4888751
13		25.93	0.0005007	4.7518417
14		25.93	0.0000950	4.7455242
15		25.93	0.0000702	4.7822589
16		25.93	0.0005007	4.7518417
17		25.93	0.0000950	4.7455242



Step 4: Normalization

Create a new tab named “Normalized data” and import the results from the “Factors” and “Responses” spreadsheets. 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						
4						
5						
6						
7						
8						
9						
10						

Multiple Spreadsheet Joiner

Join Configuration Steps

Step 1: Factors \bowtie Responses (Concatenation)

Join Type: Concatenation Left Join Right Join Inner Join Full Outer Join

Left Spreadsheet: Factors

Right Spreadsheet: Responses

Join Column: Common header name Different header names

Add Cancel

Min-Max normalizer

Excluded Columns: Col5 -- Y1, Col6 -- Y2, Col7 -- Y3

Included Columns: Col2 -- X1, Col3 -- X2, Col4 -- X3

Min: -1.0 Cancel

Max: 1.0

Results:

	Col1	Col2 (D)	Col3 (D)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)
User Header	User Row ID	X1	X2	X3	Y1	Y2	Y3
1		-1.0	-1.0	0E-7	59.6	0.0000267	6.7268120
2		1.0	-1.0	0E-7	17.71	0.0000605	4.7822589
3		-1.0	1.0	0E-7	46.85	0.0000332	4.4855323
4		1.0	1.0	0E-7	19.59	0.0000461	3.3630343
5		-1.0	0E-7	-1.0	61.62	0.0035391	4.2107007
6		1.0	0E-7	-1.0	14.36	0.0000350	4.0074930
7		-1.0	0E-7	1.0	53.81	0.0001144	4.2107007
8		1.0	0E-7	1.0	19.09	0.0000583	4.0681691
9		0.0	-1.0	-1.0	67.54	0.0002838	7.4020267
10		0.0	1.0	-1.0	23.99	0.0000834	4.6043458
11		0.0	-1.0	1.0	19.02	0.0001990	5.2028838
12		0.0	1.0	1.0	19.98	0.0000700	4.4888751
13		0.0	0E-7	0E-7	25.93	0.0005007	4.7518417
14		0.0	0E-7	0E-7	25.93	0.0000950	4.7455242
15		0.0	0E-7	0E-7	25.93	0.0000702	4.7822589
16		0.0	0E-7	0E-7	25.93	0.0005007	4.7518417
17		0.0	0E-7	0E-7	25.93	0.0000950	4.7455242

Step 5: Regression

The goal here is to produce a regression equation that includes main effects, two-factor interactions and quadratic effects for Y₁:

$$Y = b_0 + b_1X_1 + b_2X_2 + b_3X_3 + b_{12}X_1X_2 + b_{13}X_1X_3 + b_{23}X_2X_3 + b_{11}X_1^2 + b_{22}X_2^2 + b_{33}X_3^2$$

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](#)

Generalized Linear Models Regression

Type: Linear

Confidence Level...: 95

Scale Parameter Method: Fixed value

Dependent Variable: Col5 - Y1

Value: 1.0

Excluded Columns:

- Col6 - Y2
- Col7 - Y3

Factors:

- >
- <

Covariates:

- Col2 - X1
- Col3 - X2
- Col4 - X3

Formula: X1+X2+X3+X1:X1+X2:X2+X3:X3+X1:X2+X2:X3+X1:X3

Custom Include All Main Effects Full Factorial

Execute Cancel

Results:

Y1	Prediction		
59.6	65.1687500		
17.71	20.0712500		
46.85	44.48875		
19.59	14.0212500		
61.62	66.1975		
14.36	22.1450000		
53.81	46.025		
19.09	14.5125000		
67.54	57.3937500		
23.99	21.7737500		
19.02	21.2362500		
19.98	30.12625		
25.93	25.9300000		
25.93	25.9300000		
25.93	25.9300000		
25.93	25.9300000		

Goodness of Fit	Value
Deviance	452.0087250
Scaled Deviance	452.0087250
Pearson Chi-Square	452.0087250
Scaled Pearson Chi-Square	452.0087250
Log Likelihood	-241.6263176
Akaike's Information Criterion (AIC)	503.2526351
Finite Sample Corrected AIC (AICC)	539.9193018
Bayesian Information Criterion (BIC)	511.5847686
Consistent AIC (CAIC)	521.5847686

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
intercept	25.9300000	0.4472136	25.0534775	26.8065225	3361.8245000	1	0.0
X1	-18.8912500	0.3535534	-19.5842019	-18.1982981	2855.0346125	1	0.0
X2	-6.6825	0.3535534	-7.3754519	-5.9895481	357.2464500	1	0.0
X3	-6.9512500	0.3535534	-7.6442019	-6.2582981	386.5590125	1	0.0
X1*X3	3.1350000	0.5	2.1550180	4.1149820	39.3129000	1	0E-7
X1*X2	3.6575000	0.5	2.6775180	4.6374820	53.5092250	1	0E-7
X2*X3	11.1275000	0.5	10.1475180	12.1074820	495.2850250	1	0.0
X1*X1	7.2975	0.4873397	6.3423317	8.2526683	224.2252895	1	0.0
X2*X2	2.7100000	0.4873397	1.7548317	3.6651683	30.9225263	1	0E-7
X3*X3	3.9925	0.4873397	3.0373317	4.9476683	67.1160263	1	0E-7

Repeat this step for the rest of the response variables. Results, Y₂:

Y2	Prediction
0.0000267	0.0005178
0.0000605	-0.0003501
0.0000332	0.0004439
0.0000461	-0.0004450
0.0035391	0.0026753
0.0000350	0.0000730
0.0001144	0.0000765
0.0000583	0.0009220
0.0002838	0.0006565
0.0000834	0.0005365
0.0001990	-0.0002541
0.0000700	-0.0003027
0.0005007	0.0002523
0.0000950	0.0002523
0.0000702	0.0002523
0.0005007	0.0002523
0.0000950	0.0002523

Goodness of Fit	
	Value
Deviance	0.0000032
Scaled Deviance	0.0000032
Pearson Chi-Square	0.0000032
Scaled Pearson Chi-Square	0.0000032
Log Likelihood	-15.6219567
Akaike's Information Criterion (AIC)	51.2439133
Finite Sample Corrected AIC (AIACC)	87.9105800
Bayesian Information Criterion (BIC)	59.5760468
Consistent AIC (CAIC)	69.5760468

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
intercept	0.0002523	0.4472136	-0.8762702	0.8767749	3E-7	1	0.9995498
X1	-0.0004392	0.3535534	-0.6933911	0.6925127	0.0000015	1	0.9990089
X2	-0.0000422	0.3535534	-0.6929941	0.6929097	0E-7	1	0.9999048
X3	-0.0004374	0.3535534	-0.6933894	0.6925145	0.0000015	1	0.9990128
X1*X3	0.0008620	0.5	-0.9791200	0.9808440	0.0000030	1	0.9986245
X1*X2	-0.0000052	0.5	-0.9799872	0.9799767	0E-7	1	0.9999916
X2*X3	0.0000179	0.5	-0.9799641	0.9799999	0E-7	1	0.9999715
X1*X1	0.0002835	0.4873397	-0.9548848	0.9554518	3E-7	1	0.9995359
X2*X2	-0.0004942	0.4873397	-0.9556625	0.9546741	0.0000010	1	0.9991909
X3*X3	0.0004009	0.4873397	-0.9547674	0.9555692	7E-7	1	0.9993436

Results, Y₃:

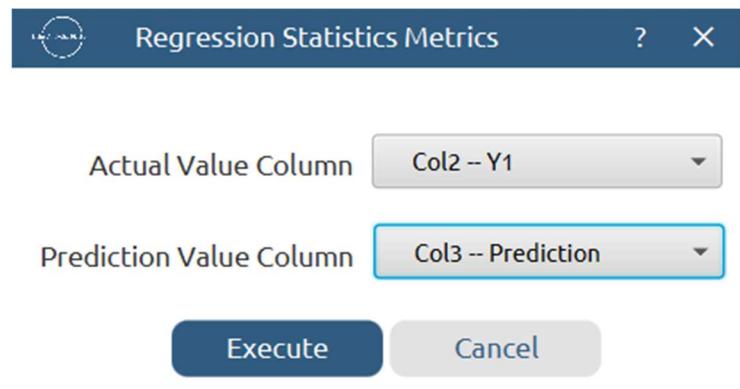
Y3	Prediction
6.7268120	6.3680462
4.7822589	5.1038210
4.4855323	4.1639701
3.3630343	3.7218002
4.2107007	4.8477759
4.0074930	3.9642402
4.2107007	4.2539534
4.0681691	3.4310939
7.4020267	7.1237174
4.6043458	4.2888327
5.2028838	5.5183969
4.4888751	4.7671845
4.7518417	4.7553982
4.7455242	4.7553982
4.7822589	4.7553982
4.7518417	4.7553982
4.7455242	4.7553982

Goodness of Fit	
	Value
Deviance	1.6346525
Scaled Deviance	1.6346525
Pearson Chi-Square	1.6346525
Scaled Pearson Chi-Square	1.6346525
Log Likelihood	-16.4392813
Akaike's Information Criterion (AIC)	52.8785626
Finite Sample Corrected AIC (AIACC)	89.5452293
Bayesian Information Criterion (BIC)	61.2106960
Consistent AIC (CAIC)	71.2106960

Parameter Estimates							
Variable	Coefficient	Std. Error	Lower CI	Upper CI	Test Statistic	df	p-value
intercept	4.7553982	0.4472136	3.8788756	5.6319207	113.0690582	1	0.0
X1	-0.4265988	0.3535534	-1.1195507	0.2663531	1.4558922	1	0.2275848
X2	-0.8965242	0.3535534	-1.5894762	-0.2035723	6.4300457	1	0.0112206
X3	-0.2817422	0.3535534	-0.9746941	0.4112097	0.6350292	1	0.4255165
X1*X3	0.0151690	0.5	-0.9648130	0.9951510	0.0009204	1	0.9757974
X1*X2	0.2055138	0.5	-0.7744682	1.1854958	0.1689437	1	0.6810523
X2*X3	0.5209181	0.5	-0.4590639	1.5009001	1.0854226	1	0.2974877
X1*X1	-0.6081279	0.4873397	-1.5632962	0.3470404	1.5571349	1	0.2120852
X2*X2	0.6921391	0.4873397	-0.2630292	1.6473074	2.0170803	1	0.1555380
X3*X3	-0.0230044	0.4873397	-0.9781727	0.9321639	0.0022282	1	0.9623506

Step 6: Regression Metrics

Create a tab named “Metrics – Y1” and import the results from the spreadsheet “Regression – Y1”. Then, produce the regression metrics for the Y₁ 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		26.5887485	5.1564279	3.8417647	0.9094393

Repeat this step for the rest of the response variables. Results, Y₂:

	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		2E-7	0.0004345	0.0003677	0.7140028

Results, Y₃:

	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		0.0961560	0.3100904	0.2330987	0.8896206

Step 7: 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 Y₁: Statistics → Analysis of (Co)Variance → ANCOVA

ANCOVA

Confidence Level (%) 95

Dependent Variable Col5 -- Y1

Sum of Squares for Tests Adjusted (Type III)

Coding for Factors (-1, 0, +1)

Excluded Columns

Col6 -- Y2
Col7 -- Y3

> <

Factors

Covariates

Col2 -- X1
Col3 -- X2
Col4 -- X3

> <

Custom Include All Main Effects Full Factorial

Formula

X1+X2+X3+X1:X1+X2:X2+X3:X3+X1:X2+X2:X3+X1:X3

Execute Cancel

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	2855.0346125	2855.0346125	44.2142843	0.0002907
2		X2	1	357.2464500	357.2464500	5.5324710	0.0509283
3		X3	1	386.5590125	386.5590125	5.9864178	0.0443202
4		X1*X2	1	53.5092250	53.5092250	0.8286667	0.3929242
5		X2*X3	1	495.2850250	495.2850250	7.6701952	0.0277136
6		X1*X3	1	39.3129000	39.3129000	0.6088163	0.4607988
7		X1*X1	1	224.2252895	224.2252895	3.4724485	0.1046798
8		X2*X2	1	30.9225263	30.9225263	0.4788794	0.5112243
9		X3*X3	1	67.1160263	67.1160263	1.0393874	0.3418995
10		Error	7	452.0087250	64.5726750		
11		Total	16	4991.2242471			

Repeat this step for the rest of the response variables. Results, Y₂:

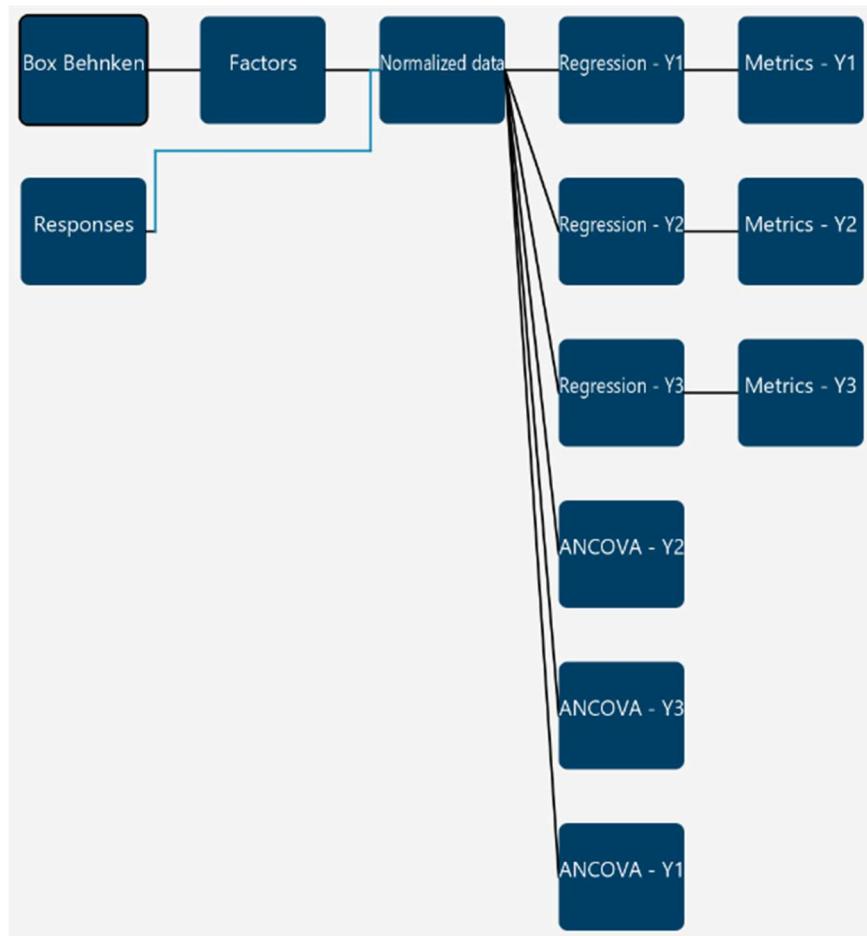
	Col1	Col2 (S)	Col3 (I)	Col4 (D)	Col5 (D)	Col6 (D)	Col7 (D)
User Header	User Row ID	Source	DF	Adj SS	Adj MS	F-Value	P-Value
1		X1	1	0.0000015	0.0000015	3.3661503	0.1091892
2		X2	1	0E-7	0E-7	0.0310394	0.8651411
3		X3	1	0.0000015	0.0000015	3.3394990	0.1103604
4		X1*X2	1	0E-7	0E-7	0.0002402	0.9880666
5		X2*X3	1	0E-7	0E-7	0.0027872	0.9593709
6		X1*X3	1	0.0000030	0.0000030	6.4830913	0.0383160
7		X1*X1	1	3E-7	3E-7	0.7381098	0.4187129
8		X2*X2	1	0.0000010	0.0000010	2.2429905	0.1778899
9		X3*X3	1	7E-7	7E-7	1.4761475	0.2637648
10		Error	7	0.0000032	5E-7		
11		Total	16	0.0000112			

Results, Y₃:

	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	1.4558922	1.4558922	6.2345029	0.0411787
2		X2	1	6.4300457	6.4300457	27.5351006	0.0011896
3		X3	1	0.6350292	0.6350292	2.7193577	0.1431263
4		X1*X2	1	0.1689437	0.1689437	0.7234601	0.4231479
5		X2*X3	1	1.0854226	1.0854226	4.6480571	0.0680149
6		X1*X3	1	0.0009204	0.0009204	0.0039414	0.9516964
7		X1*X1	1	1.5571349	1.5571349	6.6680499	0.0363526
8		X2*X2	1	2.0170803	2.0170803	8.6376537	0.0217445
9		X3*X3	1	0.0022282	0.0022282	0.0095418	0.9249228
10		Error	7	1.6346525	0.2335218		
11		Total	16	14.8094032			

Final Isalos Workflow

The final workflow is presented below:



References

- (1) Mahesh Kumar, P.; Ghosh, A. Development and Evaluation of Metronidazole Loaded Microsponge Based Gel for Superficial Surgical Wound Infections. *Journal of Drug Delivery Science and Technology* **2015**, 30, 15–29. <https://doi.org/10.1016/j.jddst.2015.09.006>.