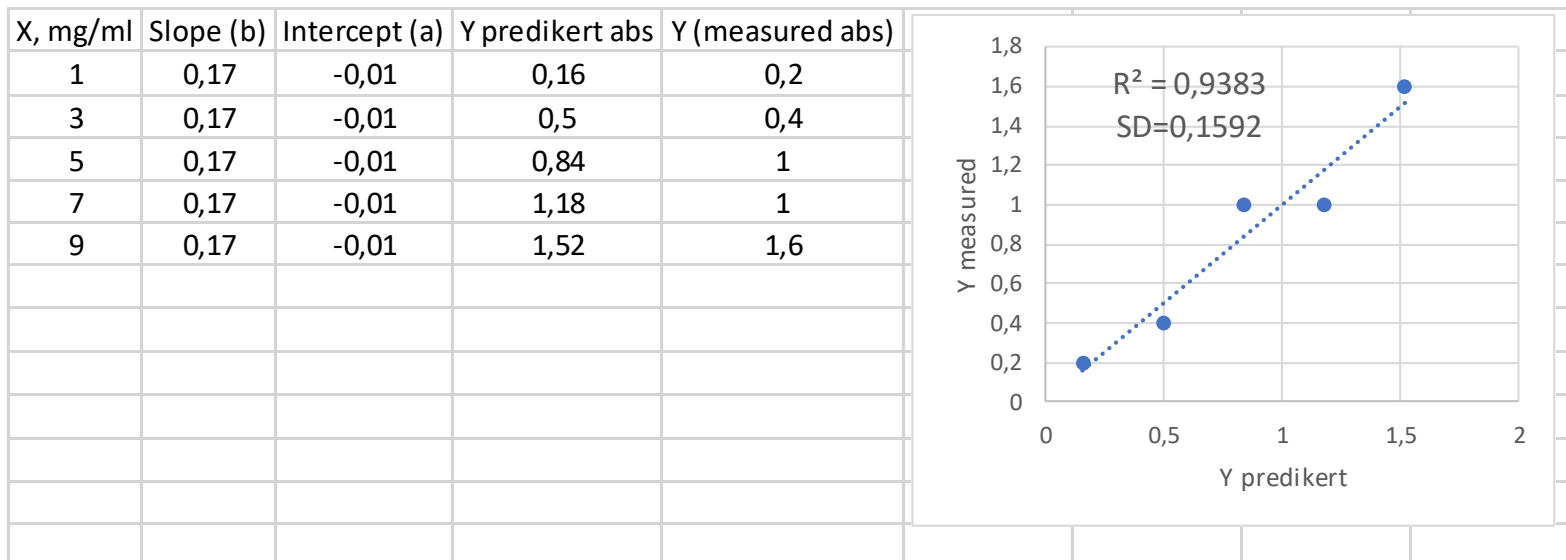
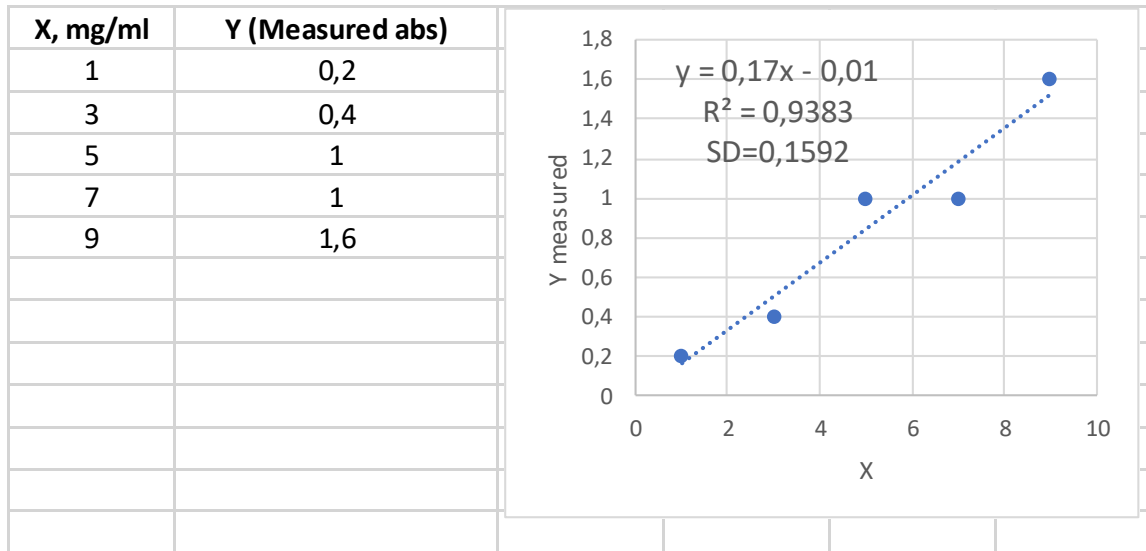


UiO

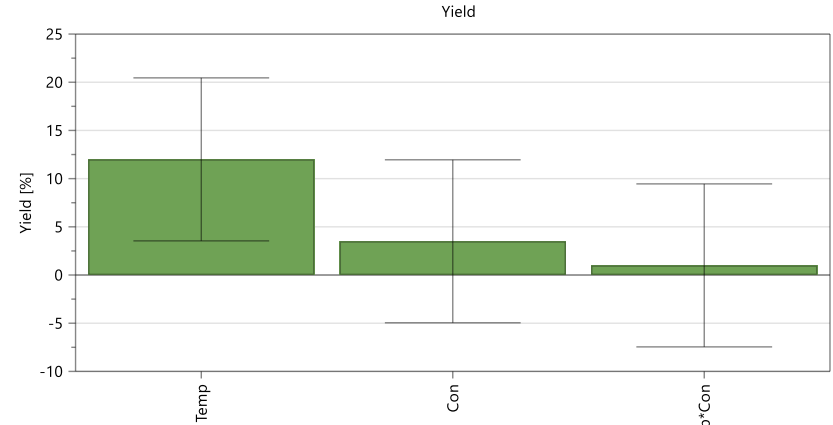
Two 'similar' ways of presenting X-Y correlation



Regression analysis (MLR) of screening DOE

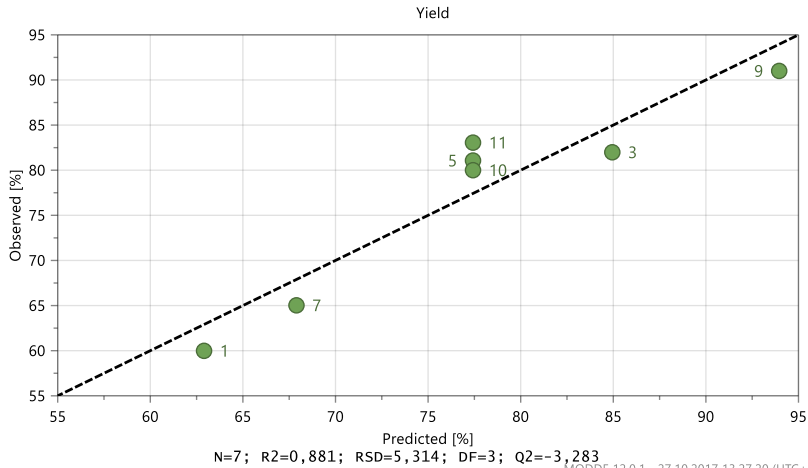
Experiment	Run	Temperature (°C)	Concentration (mg/ml)	Yield (%)
Screening	1	120	10	60
Screening	2	130	10	82
Screening	3	120	30	65
Screening	4	130	30	91
Screening, center	5	125	20	80
Screening, center	6	125	20	81
Screening, center	7	125	20	83

Coefficients (scaled and centered) - Training syntese 2 var (MLR)



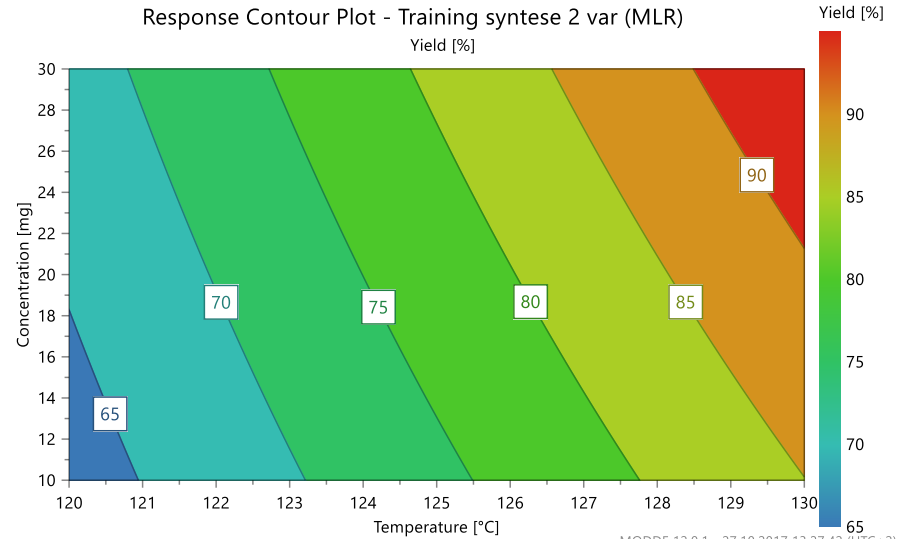
N=7; R2=0,881; RSD=5,314; DF=3; Q2=-3,283; confidence=0,95
 MODDE 12.0.1 - 27.10.2017 13.26.50 (UTC+2)

Observed vs. Predicted - Training syntese 2 var (MLR)



N=7; R2=0,881; RSD=5,314; DF=3; Q2=-3,283
 MODDE 12.0.1 - 27.10.2017 13.27.20 (UTC+2)

Response Contour Plot - Training syntese 2 var (MLR)



MODDE 12.0.1 - 27.10.2017 13.27.42 (UTC+2)

UiO

Regression analysis (MLR) of screening DOE, ANOVA table

Regression model sum of squares, SS_{model}:

The sum of the squares of the deviations of the predicted values from the mean value of a response variable

Residual (error) sum of squares, SS_{error}:

The error sum of squares is the variation in the response Y that is not explained by the variables X. It is a measure of the discrepancy between the data and an the estimated model

Total sum of squares, SS_{total}:

$$SS_{total} = SS_{model} + SS_{error} = 713.7$$

The mean square estimates:

$$MS_{model} = SS_{model}/df = 209.7$$

$$MS_{error} = SS_{error}/df = 28.2$$

$$MS_{total} = SS_{total} = 119.0$$

Significance of model (i.e. is the model describing much of Y variation or not):

$$F = MS_{model}/MS_{error} = 7.425$$

High F = significant model, i.e. MS_{model} (and SS_{model}) is large compared to MS_{error} (and SS_{error})

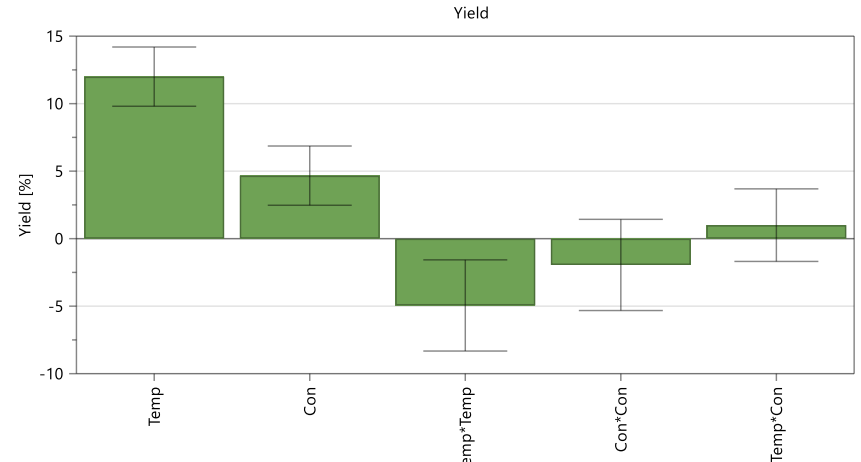
Probability: If F is higher than table reference F for a selected probability level the model is 'significant'

	SS	DF	MS	F-ratio	p-value	B-coefficients	STDerr
Summary							
Model	629.000	3	209.667	7.425	0.0669		
Error	84.714	3	28.238				
Adjusted Total	713.714	6	118.952				
Variable							
Intercept	334.850	1	334.850	11.858	0.0411	-229.571	66.667
Temp	576.000	1	576.000	20.398	0.0203	2.400	0.531
Conc	49.000	1	49.000	1.735	0.2793	0.350	0.266
Temp*Conc	4.000	1	4.000	0.142	0.7317	0.667	1.771

Regression analysis (MLR) of optimization DOE

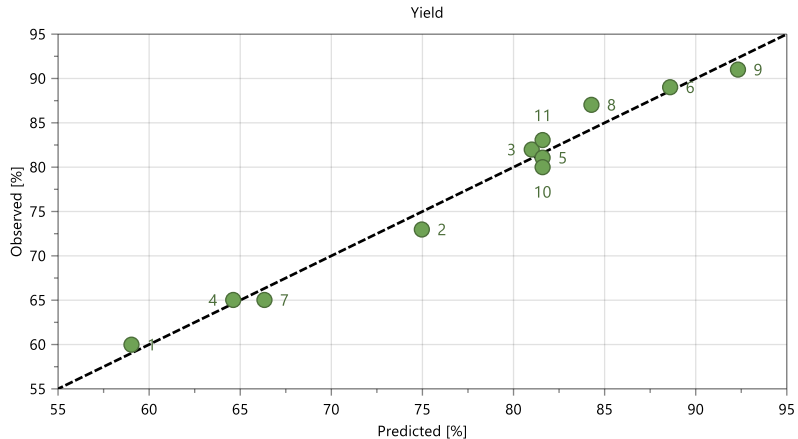
Experiment	Run	Temperature (°C)	Concentration (mg/ml)	Yield (%)
Screening	1	120	10	60
Screening	2	130	10	82
Screening	3	120	30	65
Screening	4	130	30	91
Screening, center	5	125	20	80
Screening, center	6	125	20	81
Screening, center	7	125	20	83
Optimization	8	125	10	73
Optimization	9	125	30	87
Optimization	10	120	20	65
Optimization	11	130	20	89

Coefficients (scaled and centered) - Training syntese 2 var (MLR)



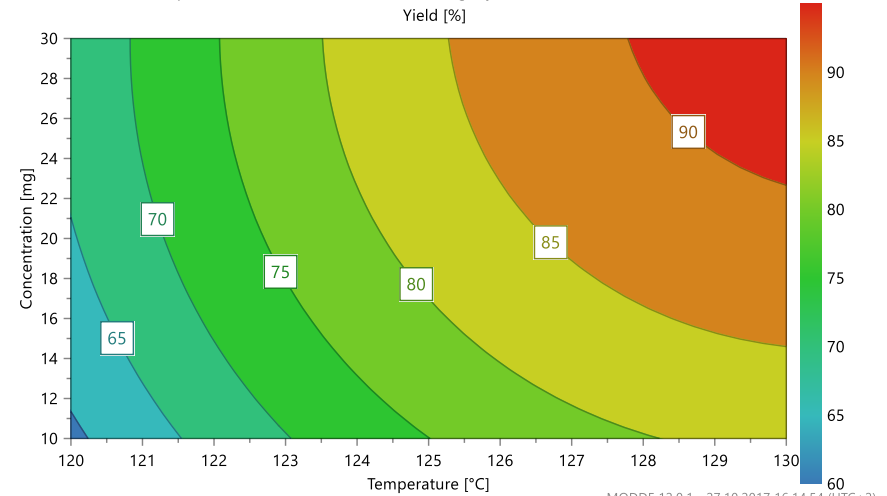
N=11; R2=0,980; RSD=2,091; DF=5; Q2=0,831; Confidence=0,95
MODDE 12.0.1 - 27.10.2017 16.14.05 (UTC+2)

Observed vs. Predicted - Training syntese 2 var (MLR)



N=11; R2=0,980; RSD=2,091; DF=5; Q2=0,831
MODDE 12.0.1 - 27.10.2017 16.14.30 (UTC+2)

Response Contour Plot - Training syntese 2 var (MLR)



MODDE 12.0.1 - 27.10.2017 16.14.54 (UTC+2)

Regression analysis (MLR) of optimization DOE, ANOVA table

	SS	DF	MS	F-ratio	p-value	B-coefficients	STDerr
Summary							
Model	1.090e+03	5	217.955	49.853	0.0003		
Error	21.860	5	4.372				
Adjusted Total	1.112e+03	10	111.164				
Variable							
Intercept	493.565	1	493.565	112.894	0.0001	-227.754	21.435
Temp	863.999	1	863.999	197.624	0.0000	2.400	0.171
Conc	130.666	1	130.666	29.888	0.0028	0.467	8.536e-02
Temp*Conc	4.000	1	4.000	0.915	0.3827	0.600	0.627
Temp**2	62.007	1	62.007	14.183	0.0131	-2.968	0.788
Conc**2	9.607	1	9.607	2.197	0.1983	-1.168	0.788

DOE randomization, block, fold-over

- **Randomization:** Randomize runs to avoid interference from a non-designed variables as build-up effects or other covariate effects
 - Nested design means that of different reasons, as practical ones, the DOE is not completely randomised
- **Blocking:** Divide the design in different blocks, often due to practical reasons (Different randomization within blocks may be selected, for example different randomization at day 1 and 2)
 - A block variable can be applied as a dummy variable checked for significance
- **Fold-over:** Extend the design and increase resolution by mirroring (multiply with -1) resolution III design
 - To understand the new confounding structure one can construct a reference design of similar size as the new design and study the confounding given by the stat software

One-way ANOVA (not mandatory, from excel)

	Replikat 1	Replikat 2	Replikat 3	Average	One way anova						
Analyse metode 1	80	79	80	79,66666667	Variansanalyse: en-faktor						
Analyse metode 2	82	78	80	80							
Analyse metode 3	86	85	83	84,66666667	SAMMENDRAG						
Average	82,66666667	80,66666667	81		<i>Grupper</i>	<i>Antall</i>	<i>Sum</i>	<i>Gjennomsnitt</i>	<i>Varians</i>		
					Analyse metode 1	3	239	79,66666667	0,333333333		
					Analyse metode 2	3	240	80	4		
					Analyse metode 3	3	254	84,66666667	2,333333333		
					Variansanalyse						
					<i>Variasjonskilde</i>	<i>SK</i>	<i>fg</i>	<i>GK</i>	<i>F</i>	<i>P-verdi</i>	<i>F-krit</i>
					Mellom grupper	46,88888889	2	23,44444444	10,55	0,010852902	5,14325285
					Innenfor grupper	13,33333333	6	2,22222222			
					Totalt	60,22222222	8				

The variation between the analytical methods is significant greater than the variation among replicates (P=0.0108)

Two-way ANOVA (not mandatory, from excel)

	Batch 1	Batch 2	Batch 3	Average	Two way ANOVA									
Analyse metode 1	80	79	80	79,66666667	Variansanalyse: To-faktor uten tilbakelegging									
Analyse metode 2	82	78	80	80										
Analyse metode 3	86	85	83	84,66666667										
Average	82,66666667	80,66666667	81											
					SAMMENDRAG				<i>Antall</i>	<i>Sum</i>	<i>Gjennomsnitt</i>	<i>Varians</i>		
					Analyse metode 1	3	239	79,66666667	0,333333333					
					Analyse metode 2	3	240	80	4					
					Analyse metode 3	3	254	84,66666667	2,333333333					
					Batch 1	3	248	82,66666667	9,333333333					
					Batch 2	3	242	80,66666667	14,33333333					
					Batch 3	3	243	81	3					
					Variansanalyse									
					<i>Variasjonskilde</i>				<i>SK</i>	<i>fg</i>	<i>GK</i>	<i>F</i>	<i>P-verdi</i>	<i>F-krit</i>
					Rader (analyse metode)	46,88888889	2	23,44444444	14,55172414	0,014600694	6,94427191			
					Kolonner (Batcher)	6,88888889	2	3,44444444	2,137931034	0,233611111	6,94427191			
					Feil	6,44444444	4	1,611111111						
					Totalt	60,22222222	8							

The difference between the analytical methods is significant ($P=0.0146$) while the difference between batches is not ($P=0.234$) when compared to the error (given by GK, i.e. mean square)