

## DOE Wizard – Multilevel Factorial Designs

### Summary

The *Multilevel Factorial* designs are used to study the effects of  $q$  quantitative factors. The user begins by specifying a range over which each factor is to be varied and the number of different levels at which it will be studied. The program creates a datasheet containing all combinations of the different levels of the variables.

One important use of this type of design is to create a set of candidate runs from which an optimal subset can be selected using the *Select runs* feature of the DOE Wizard, which selects a subset of runs so as to achieve a D-optimal design.

### Example

As an example, suppose the experimenter wished to study 3 factors over the region shown below:

Factor	Low	High	Units
temperature	160	180	degrees
flow rate	50	80	liters/min
concentration	20	40	%

Unfortunately, rather than designing a well-balanced experiment from the beginning, he decided to try a few runs that he thought might work well. The runs performed are shown below, together with the measured response:

<i>run</i>	<i>temperature</i>	<i>flow rate</i>	<i>concentration</i>	<i>yield</i>
1	160	50	20	18.7
2	180	75	40	32.9
3	165	80	25	22.1
4	175	80	35	29.0
5	170	55	25	22.8
6	180	75	35	29.9

After 6 runs and little progress, he decided to stop and design a good experiment. He did not, however, want to throw away the runs that had already been performed, each of which was costly and time-consuming.

This document will demonstrate how the DOE Wizard can be used to help select additional experiments to augment the runs that have already been performed.

**Sample StatFolio:** doewiz multilevel.sgp

## Design Creation

To begin the design creation process, start with an empty StatFolio. Select *DOE – Experimental Design Wizard* to load the DOE Wizard’s main window. Then push each button in sequence to create the design.

### Step #1 – Define Responses

The first step of the design creation process displays a dialog box used to specify the response variables. For the current example, there is a single response variable:

Design file: <untitled>  
 Comment: A fix-up design created using a multilevel factorial  
 Number of responses: 1

Response	Name	Units	Analyze	Goal	Target	Impact (1-5)	Sensitivity	Minimum	Maximum
1	yield		Mean	Maximize	0.5	3.0	Medium		
2	Var_2		Mean	Maximize	0.5	3.0	Medium		
3	Var_3		Mean	Maximize	0.5	3.0	Medium		
4	Var_4		Mean	Maximize	0.5	3.0	Medium		
5	Var_5		Mean	Maximize	0.5	3.0	Medium		
6	Var_6		Mean	Maximize	0.5	3.0	Medium		
7	Var_7		Mean	Maximize	0.5	3.0	Medium		
8	Var_8		Mean	Maximize	0.5	3.0	Medium		
9	Var_9		Mean	Maximize	0.5	3.0	Medium		
10	Var_10		Mean	Maximize	0.5	3.0	Medium		
11	Var_11		Mean	Maximize	0.5	3.0	Medium		
12	Var_12		Mean	Maximize	0.5	3.0	Medium		
13	Var_13		Mean	Maximize	0.5	3.0	Medium		
14	Var_14		Mean	Maximize	0.5	3.0	Medium		
15	Var_15		Mean	Maximize	0.5	3.0	Medium		
16	Var_16		Mean	Maximize	0.5	3.0	Medium		

OK Cancel Help

- **Name:** The name for the variable is *yield*.
- **Units:** The units are not relevant in this example.
- **Analyze:** The parameter of interest is the *mean* yield.
- **Goal:** The goal of the experiment is to maximize the yield.
- **Impact:** The relative importance of each response (not relevant if only one response).
- **Sensitivity:** The importance of being close to the best desired value.
- **Minimum and Maximum:** Range of desirable values for the response.

*Step #2 – Define Experimental Factors*

The second step displays a dialog box on which to specify the factors that will be varied. In the current example, there are 3 factors:

Design file: <untitled>  
 Comment: A fix-up design created using a multilevel factorial

Number of controllable process factors: 3    Number of controllable mixture components: 0    Number of noise factors: 0

Factor	Name	Units	Type	Role	Low	High	Levels
A	temperature	degrees	Continuous	Controllable	160	180	1,2,3,4
B	flow rate	liters/min	Continuous	Controllable	50	80	1,2,3,4
C	concentration	%	Continuous	Controllable	20	40	1,2,3,4
D	Factor_D		Continuous		-1.0	1.0	1,2,3,4
E	Factor_E		Continuous		-1.0	1.0	1,2,3,4
F	Factor_F		Continuous		-1.0	1.0	1,2,3,4
G	Factor_G		Continuous		-1.0	1.0	1,2,3,4
H	Factor_H		Continuous		-1.0	1.0	1,2,3,4
I	Factor_I		Continuous		-1.0	1.0	1,2,3,4
J	Factor_J		Continuous		-1.0	1.0	1,2,3,4
K	Factor_K		Continuous		-1.0	1.0	1,2,3,4
L	Factor_L		Continuous		-1.0	1.0	1,2,3,4
M	Factor_M		Continuous		-1.0	1.0	1,2,3,4

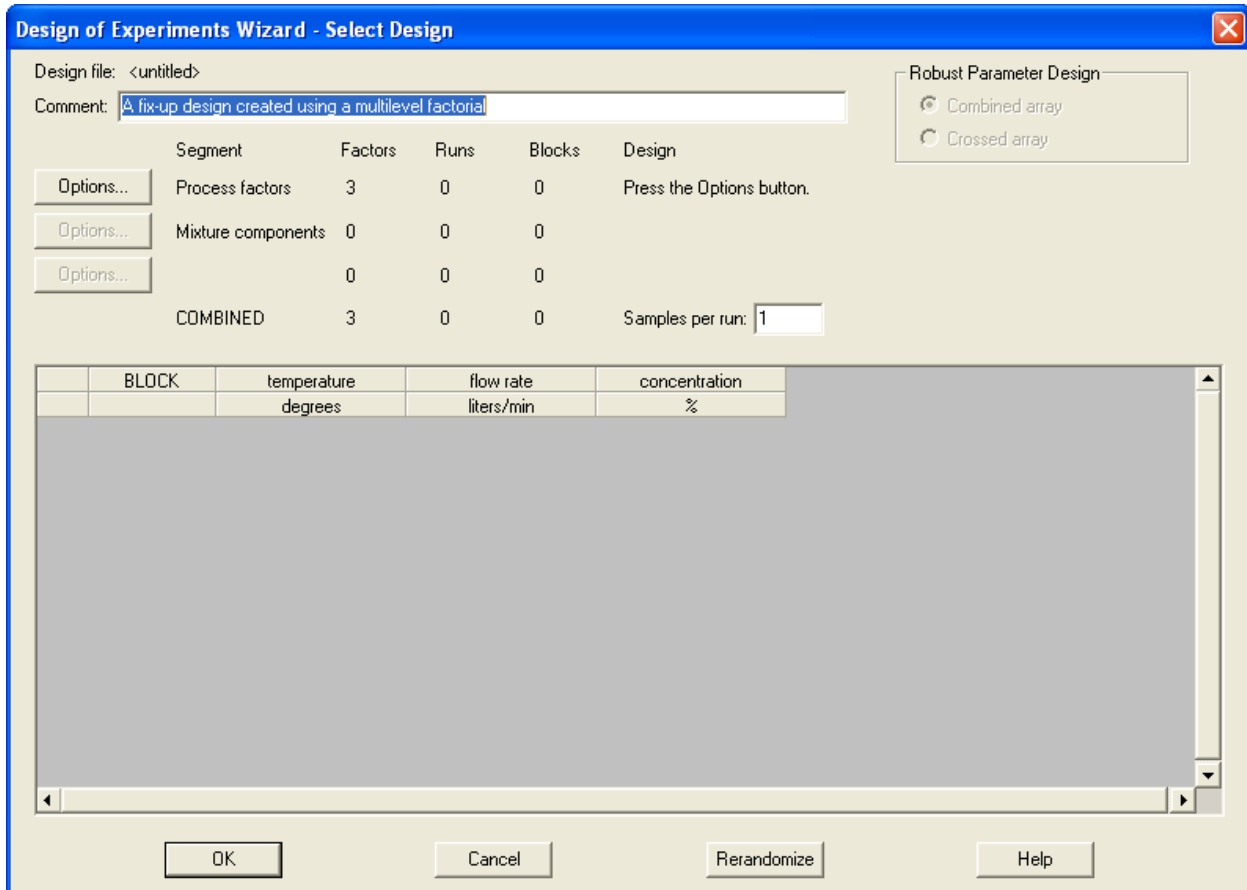
Total for controllable mixture components: 100.0    Factors A-M    Factors N-Z

OK    Back    Cancel    Help

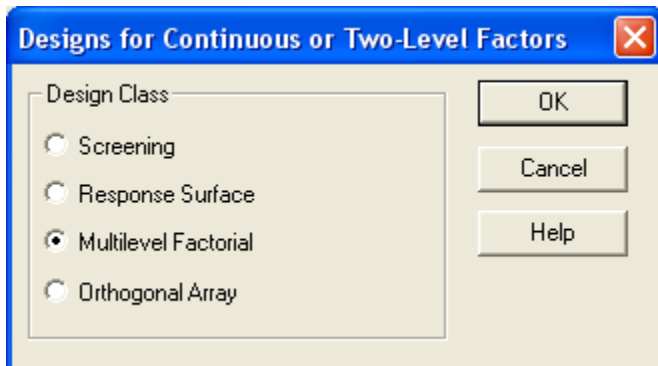
- **Name** – Each factor must be assigned a unique name.
- **Units** – Units are optional.
- **Type** – Set the type of each factor to *Continuous*, since they can be set at any value within a continuous interval.
- The three factors are all *Controllable*.
- **Low** - the lower level  $L_j$  for the factor.
- **High** - the upper level  $U_j$  for the factor.

*Step #3 – Select Design*

The third step begins by displaying the dialog box shown below:



Since all of the factors are controllable process factors, only one *Options* button is enabled. Pressing that button displays a second dialog box:



Four general classes of designs are offered:

1. *Screening* - designs intended to select the most important factors affecting a response. Most of the designs involve only 2 levels of each factor. The factors may be quantitative or categorical.
2. *Response Surface* - designs intended to select the optimal settings of a set of experimental factors. The designs involve at least 3 levels of the experimental factors, which must be quantitative.

3. *Multilevel Factorial* - designs involving different numbers of levels for each experimental factor. The factors must be quantitative.
4. *Orthogonal Array* – a general class of designs developed by Genichi Taguchi. The factors may be quantitative or categorical.

If *Multilevel Factorial* is selected, a third dialog box will be displayed on which to indicate the number of levels at which each factor should be run:

Factor	Levels
temperature	5
flow rate	5
concentration	5

Runs: 125    Error d.f.: 115

Replicate Design  
Number: 0

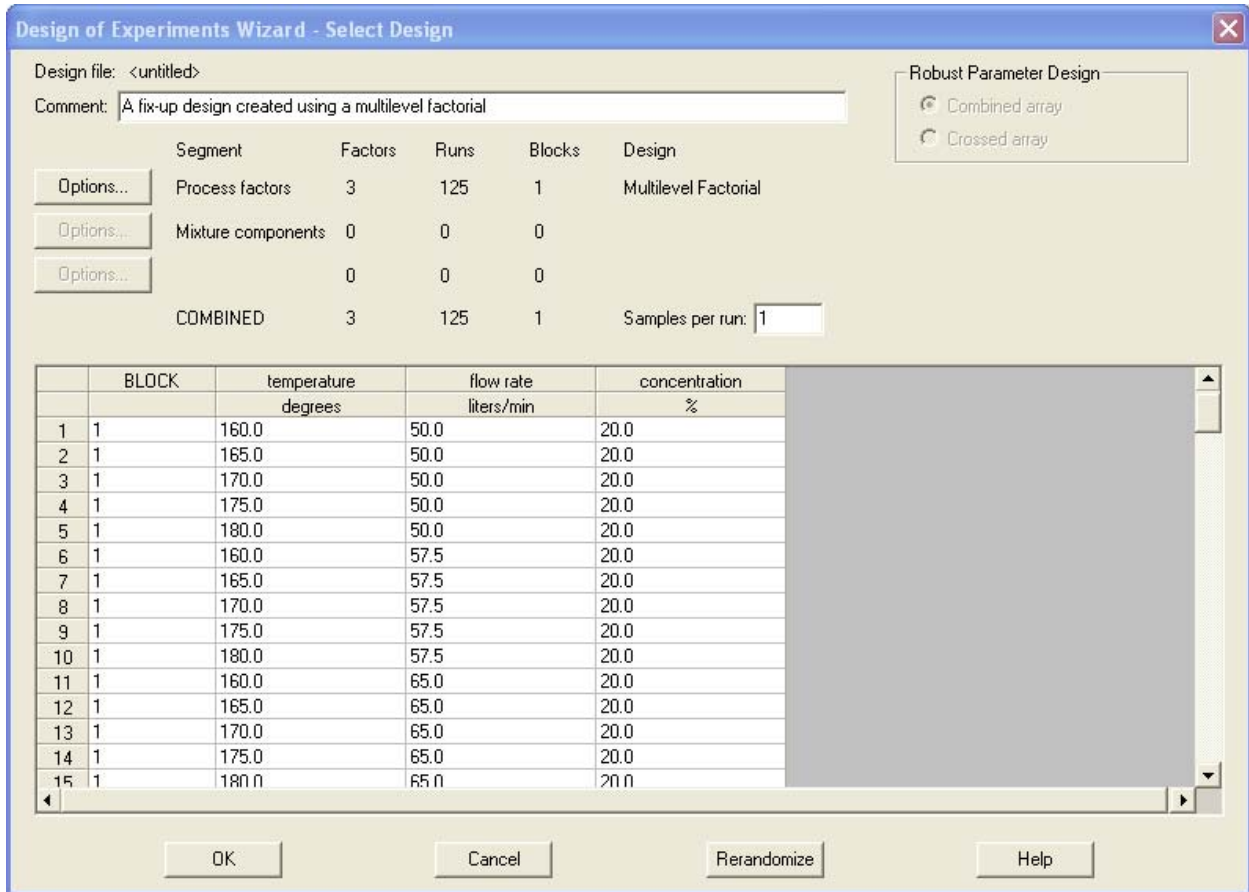
Randomize

Buttons: OK, Cancel, Back, Help

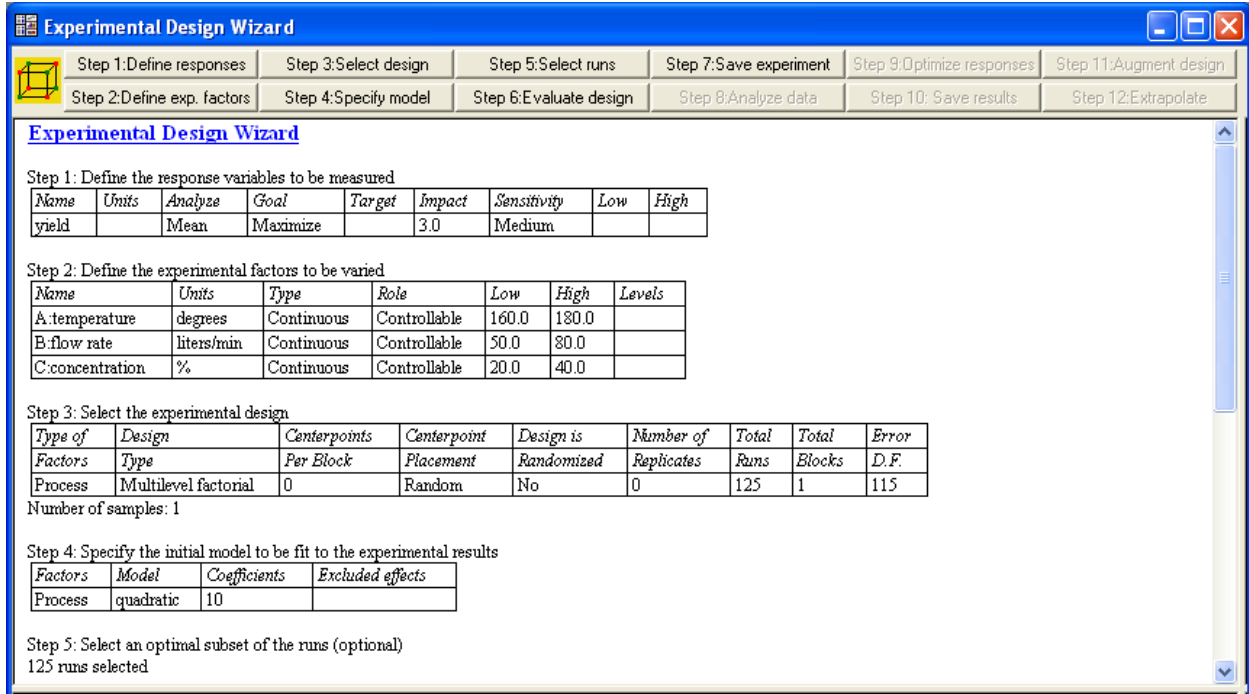
- **Levels** – the number of levels at which each factor should be run.
- **Replicate design** - if a number other than 0 is entered, the entire design will be repeated the indicated number of times.
- **Randomize** - check this box to randomly order the runs in the experiment. Randomization is generally a good idea, since it can reduce the effect of lurking variables such as trends over time. However, when replicating the examples in this documentation, **do not** randomize the designs.

For the current example, select 5 for each factor, resulting in a total of  $5 \times 5 \times 5 = 125$  runs at each combination of the factors.

The tentatively selected design is displayed in the *Select Design* dialog box:

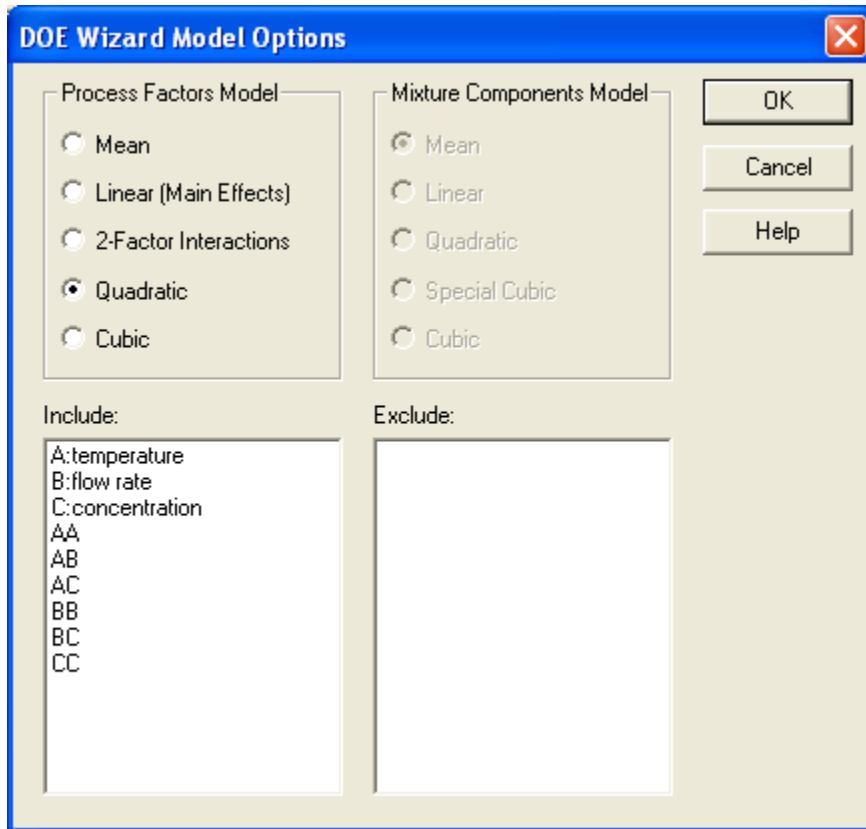


If the design is acceptable, press OK to save it to the STATGRAPHICS DataBook and return to the DOE Wizard’s main window, which should now contain a summary of the design:



*Step #4: Specify Model*

The next step in the design selection process is to specify the model that will be fit to the response data. Pressing the fourth button on the DOE Wizard's toolbar displays a dialog box to make that choice:



The default *quadratic* model includes main effects for each of the 3 experimental factors, 3 two-factor interactions (shown as two-letter combinations with different factors), and 3 quadratic terms (shown as two-letter combinations with identical factors). Selected terms could be excluded by double-clicking on them with the left mouse button.

*Step #5: Select runs*

The next step is to select a subset of the 125 runs to add to the 6 runs already performed. Before pressing the button for step 5, the original 6 runs need to be added to the datasheet. Click on the DataBook and scroll to the 126<sup>th</sup> row. Then add the information for the original 6 runs, including the block number and measured response values:

	BLOCK	temperature	flow rate	concentration	yield
		degrees	liters/min	%	
120	1	180.0	72.5	40.0	
121	1	160.0	80.0	40.0	
122	1	165.0	80.0	40.0	
123	1	170.0	80.0	40.0	
124	1	175.0	80.0	40.0	
125	1	180.0	80.0	40.0	
126	1	160	50	20	18.7
127	1	180	75	40	32.9
128	1	165	80	25	22.1
129	1	175	80	35	29.0
130	1	170	55	25	22.8
131	1	180	75	35	29.9
132					
133					

After entering the original 6 runs, return to the DOE Wizard and press the button labeled *Step 5*. The following dialog box will be displayed:

**Design of Experiments Wizard - Select Runs**

	BLOCK	temperature	flow rate	concentration	yield
		degrees	liters/min	%	
1	1	160.0	50.0	20.0	
2	1	165.0	50.0	20.0	
3	1	170.0	50.0	20.0	
4	1	175.0	50.0	20.0	
5	1	180.0	50.0	20.0	
6	1	160.0	57.5	20.0	
7	1	165.0	57.5	20.0	
8	1	170.0	57.5	20.0	
9	1	175.0	57.5	20.0	
10	1	180.0	57.5	20.0	
11	1	160.0	65.0	20.0	
12	1	165.0	65.0	20.0	
13	1	170.0	65.0	20.0	
14	1	175.0	65.0	20.0	
15	1	180.0	65.0	20.0	
16	1	160.0	72.5	20.0	
17	1	165.0	72.5	20.0	
18	1	170.0	72.5	20.0	
19	1	175.0	72.5	20.0	
20	1	180.0	72.5	20.0	
21	1	160.0	80.0	20.0	
22	1	165.0	80.0	20.0	

Number of runs desired:

Model coefficients: 10

Apply exchange algorithm at end

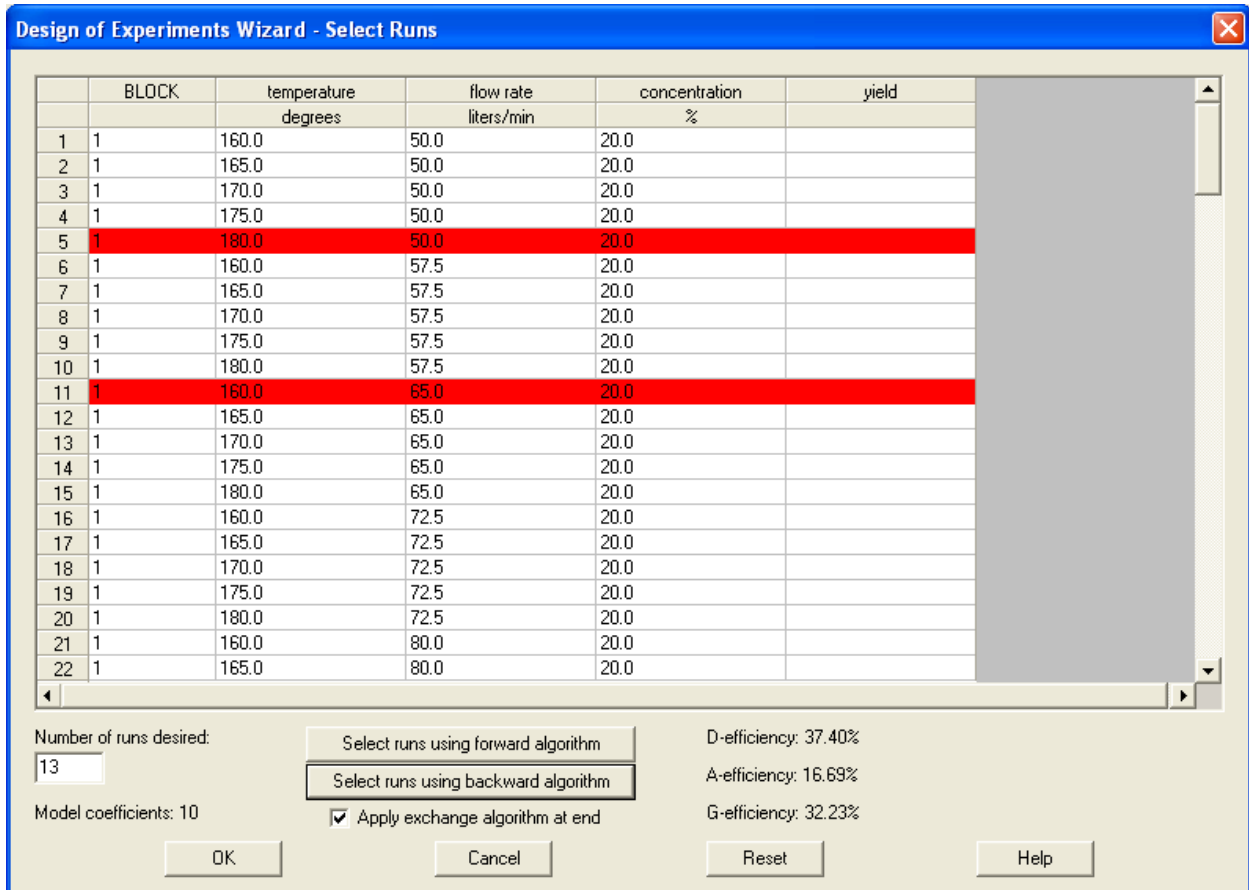
D-efficiency: 
 A-efficiency: 
 G-efficiency:



In the bottom left is a field where the desired number of runs should be specified. It is usually a good idea to select at least 3 more runs than there are coefficients in the selected model.

To select the runs, press either of the 2 buttons on the dialog box. Since the number of ways of choosing subsets of the candidate runs is too large to check all possibilities, STATGRAPHICS (like other programs) uses a selection algorithm to choose a subset. The *Forward* method begins with the runs that have already been performed (if any) and adds runs one at a time, adding at each step the run that adds the most to the D-efficiency of the experiment. The *Backward* method begins with all of the candidate runs and removes runs one at a time, removing at each step the run that adds the least to the D-efficiency of the experiment. In either case, once the desired number of runs has been selected, an *exchange* algorithm can be performed. This algorithm tests all pairs of runs consisting of one that has been selected and one that has not, making any exchanges that would increase the efficiency of the experiment. Exchanges continue until no further improvements can be made by switching one run that's been selected with one run that has not been selected.

When the algorithm is complete, the selected rows will be highlighted in red:



The efficiencies of the selected design will also be displayed. You can try another algorithm or press *OK* to accept the selection, at which point the rows of the datasheet will be reduced to the selected runs:

	BLOCK	temperature	flow rate	concentration	yield
		degrees	liters/min	%	
1	1	180	50	20	
2	1	160	65	20	
3	1	180	80	20	
4	1	160	50	30	
5	1	170	50	40	
6	1	180	50	40	
7	1	160	80	40	
8	1	160	50	20	18.7
9	1	180	75	40	32.9
10	1	165	80	25	22.1
11	1	175	80	35	29.0
12	1	170	55	25	22.8
13	1	180	75	35	29.9
14					

The main DOE Wizard window will reflect the design:

**Experimental Design Wizard**

Step 1: Define responses | Step 3: Select design | **Step 5: Select runs** | Step 7: Save experiment | Step 9: Optimize responses | Step 11: Augment design

Step 2: Define exp. factors | Step 4: Specify model | Step 6: Evaluate design | Step 8: Analyze data | Step 10: Save results | Step 12: Extrapolate

**Experimental Design Wizard**

Step 1: Define the response variables to be measured

Name	Units	Analyze	Goal	Target	Impact	Sensitivity	Low	High
yield		Mean	Maximize		3.0	Medium		

Step 2: Define the experimental factors to be varied

Name	Units	Type	Role	Low	High	Levels
A: temperature	degrees	Continuous	Controllable	160.0	180.0	
B: flow rate	liters/min	Continuous	Controllable	50.0	80.0	
C: concentration	%	Continuous	Controllable	20.0	40.0	

Step 3: Select the experimental design

Type of	Design	Centerpoints	Centerpoint	Design is	Number of	Total	Total	Error
Factors	Type	Per Block	Placement	Randomized	Replicates	Runs	Blocks	D.F.
Process	Multilevel factorial	0	Random	No	0	125	1	115

Number of samples: 1

Step 4: Specify the initial model to be fit to the experimental results

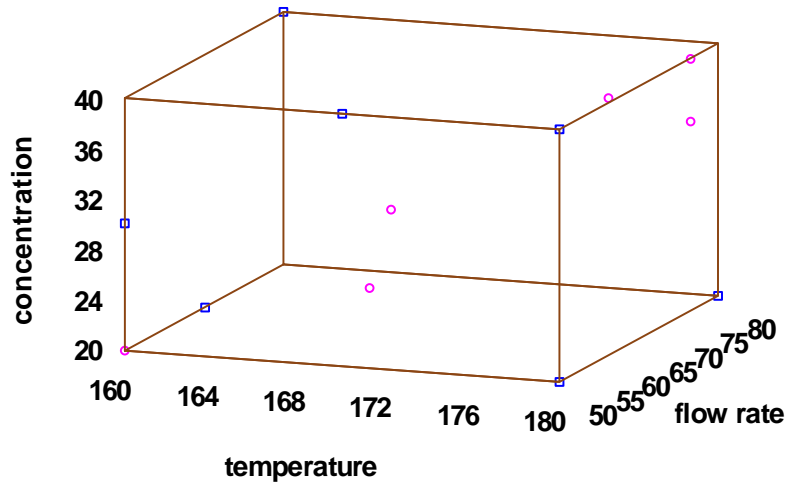
Factors	Model	Coefficients	Excluded effects
Process	quadratic	10	

Step 5: Select an optimal subset of the runs (optional)  
13 runs selected

If the selection is acceptable, press *Step 7: Save experiment* to save the reduced number of runs.

You can also use the *Design Plot* to display the final design:

A fix-up design created using a multilevel factorial



The original runs are shown as red asterisks, while the added runs are shown as blue squares.