

Categoric Mixture Components Proportion Going to Zero

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- **Categoric Mixture Components
Proportion Going to Zero**
- **What's the problem?**
- Shelf life example
- Conclusion

Categoric Factor with Proportion Zero

Four Component and One Categoric

Let's start with a four component example where (D) is proportion of one, or the other, of two categoric components:

- A, B, C and D vary from 0 to 1
- D is the proportion of categoric factor e, which has two levels (e_1 and e_2)

Build a combined design for a quadratic mixture crossed with a main effect model for the categoric factor (e):

$$\begin{aligned}\hat{y} = & \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD \\ & + \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce + \beta_{45} De + \beta_{125} ABe + \beta_{135} ACe + \beta_{145} ADe \\ & + \beta_{235} BCe + \beta_{245} BDe + \beta_{345} CDe\end{aligned}$$

Categoric Factor with Proportion Zero

Four Component and One Categoric

Component 1 A:A mg	Component 2 B:B mg	Component 3 C:C mg	Component 4 D:D mg	Factor 5 E:E
100.000	1000.000	0.000	0.000	e2
320.392	728.066	51.542	0.000	e2
406.6	51.542	0.000	0.000	e2
456	52.695	0.000	0.000	e1
223	120.000	0.000	0.000	e2
456	52.695	0.000	0.000	e1
000	112.000	0.000	0.000	e2
485.374	500.000	114.626	0.000	e2
292.094	687.906	120.000	0.000	e1
493.681	606.319	0.000	0.000	e1
600.000	500.000	0.000	0.000	e2

Problem: When the proportion of a categoric factor goes to zero, then the different levels are all the same – **absent!**

Categoric Factor with Proportion Zero

Four Component and One Categoric

Problem: When the proportion of a categoric factor goes to zero, then the different levels are all the same – absent!

Four component example:

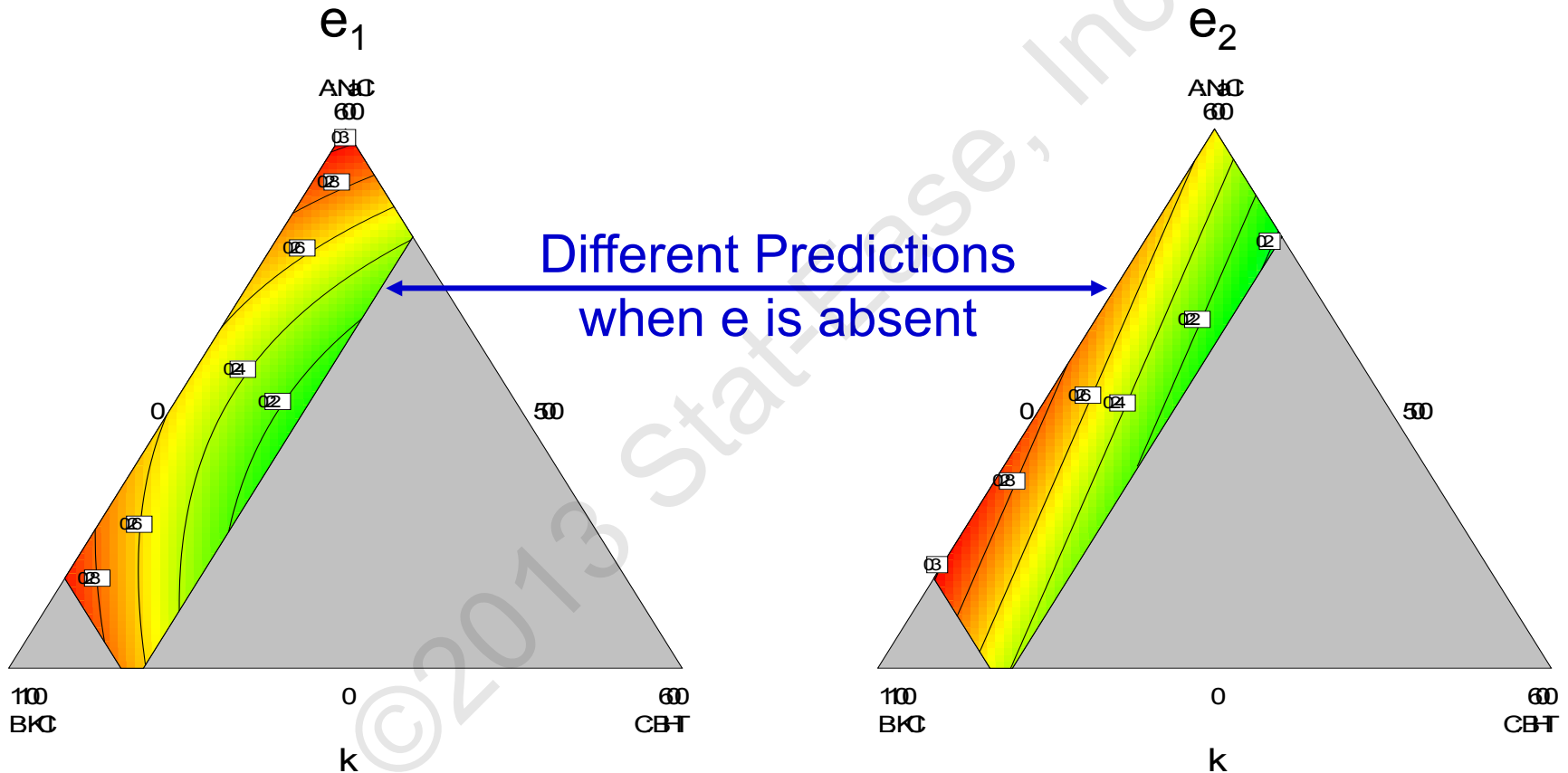
$$\begin{aligned}\hat{y} = & \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD \\ & + \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce + \beta_{45} De + \beta_{125} ABe + \beta_{135} ACe + \beta_{145} ADe \\ & + \beta_{235} BCe + \beta_{245} BDe + \beta_{345} CDe\end{aligned}$$

Predictions can vary for each level of the categoric component even when the categoric component is absent ($D=0$)!

(see next slide)

Categoric Factor Absent, i.e. $D = 0$

Different Predictions for each Levels



Categoric Factor with Proportion Zero

Four Component and One Categoric

Solution: Start with the mixture model and add:

- The mixture terms with “D” crossed with “e”.
- The mixture terms without “D” crossed with “De”.

Never let “e” appear without “D”.

$$\begin{aligned}\hat{y} = & \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD \\ & + \beta_{45} De + \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe + \beta_{1245} ABDe + \beta_{1345} ACDe \\ & + \beta_{2345} BCDe\end{aligned}$$

Now the predictions are the same for each level of the categoric component when the categoric component is absent (D=0)!

(see next slide)

Mixture model

- + the mixture terms with “D” crossed with “e”
- + the mixture terms without “D” crossed with “De”:

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD$$
$$(+\beta_4 D + \beta_{14} AD + \beta_{24} BD + \beta_{34} CD)(e)$$
$$(+\beta_1 A + \beta_2 B + \beta_3 C + \beta_{12} AB + \beta_{13} AC + \beta_{23} BC)(De)$$

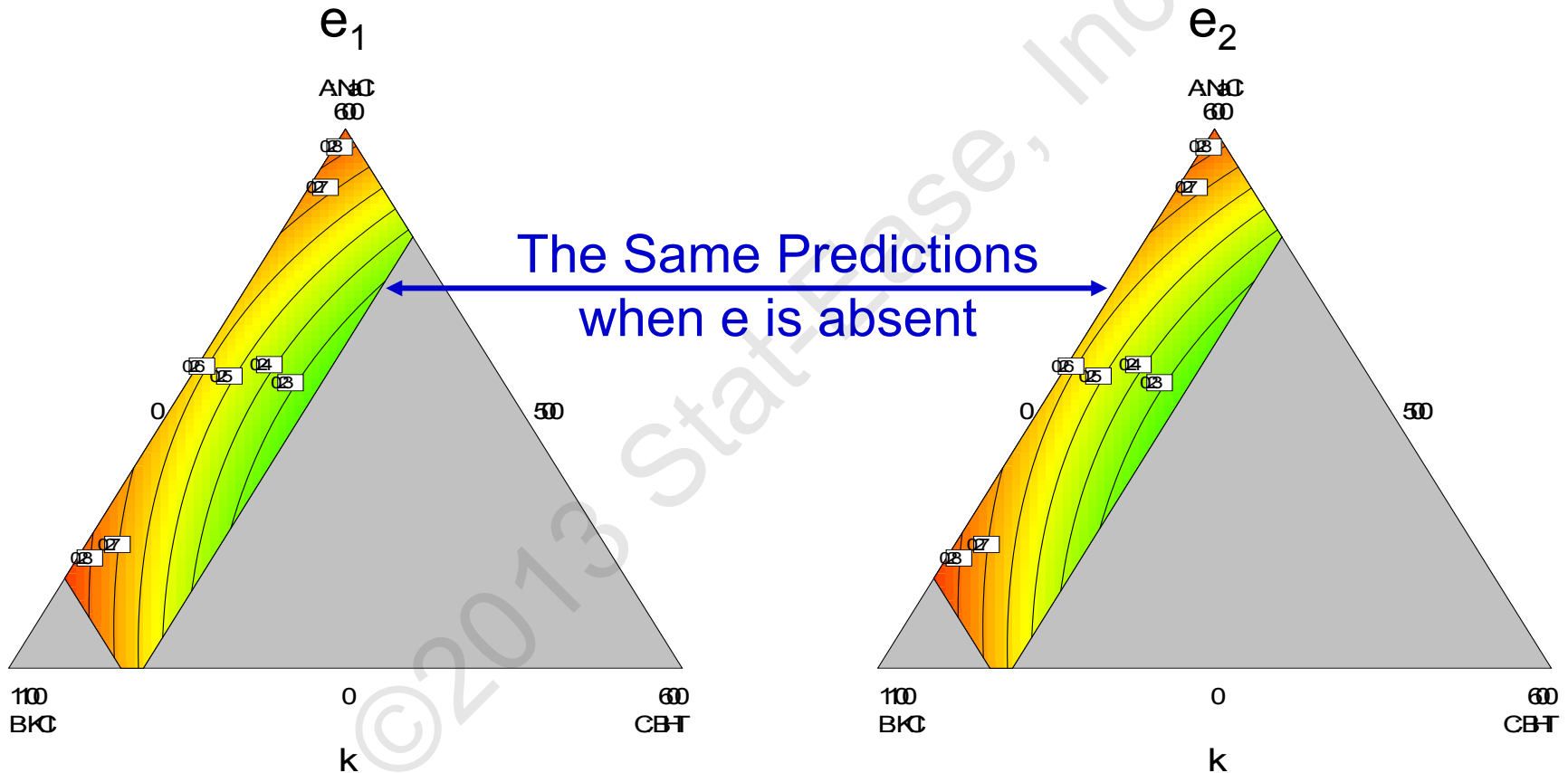
$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD$$
$$+ \beta_{45} De + \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe$$
$$+ \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe + \beta_{1245} ABDe + \beta_{1345} ACDe + \beta_{2345} BCDe$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD$$
$$+ \beta_{45} De + \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe + \beta_{1245} ABDe + \beta_{1345} ACDe + \beta_{2345} BCDe$$

“e” never appears without “D”

Categorical Factor Absent, i.e. $D = 0$

Same Predictions for each Levels



Categoric Factor with Proportion Zero

Mixture Quadratic with One Categoric

x_q is the proportion of a component and z_1 is the levels (categoric level) of that component :

$$\eta(\mathbf{x}, \mathbf{z}) = \underbrace{\sum_{i=1}^q \gamma_i^0 x_i + \sum_{i < j}^q \gamma_{ij}^0 x_i x_j}_{\text{average blending properties}} + \underbrace{\gamma_1^0 x_q z_1 + \left[\sum_{i=1}^{q-1} \gamma_i^1 x_i + \sum_{i=1 < j}^{q-1} \gamma_{ij}^1 x_i x_j \right]}_{\text{correction to average blending for categoric amount and levels}} (x_q z_1)$$

treat $x_q z_1$ as one term (amount)(levels) of a component

q = number of mixture components

m = number of mixture components with categoric levels
(in this case m is fixed at one: $m = 1$)

Categoric Factor with Proportion Zero

Mixture Quadratic with One Categoric

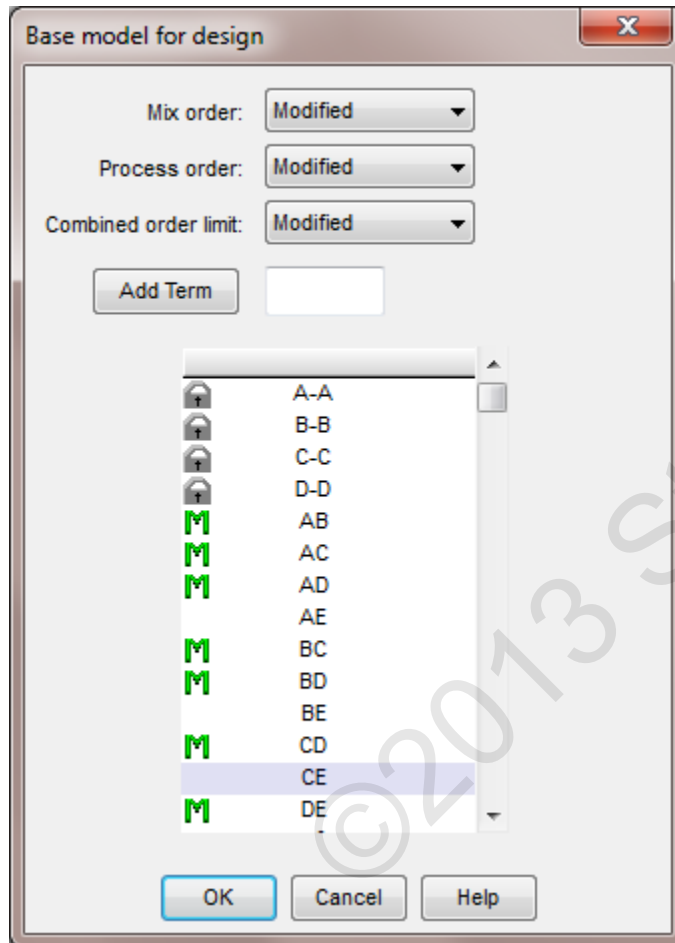
$$\eta(\mathbf{x}, \mathbf{z}) = \underbrace{\sum_{i=1}^q \gamma_i^0 x_i + \sum_{i < j}^q \gamma_{ij}^0 x_i x_j}_{\text{average blending properties}} + \underbrace{\gamma_1^0 \mathbf{x}_q \mathbf{z}_1 + \left[\sum_{i=1}^{q-1} \gamma_i^1 x_i + \sum_{i=1 < j}^{q-1} \gamma_{ij}^1 x_i x_j \right]}_{\text{correction to average blending for categoric amount and levels}} (\mathbf{x}_q \mathbf{z}_1)$$

for $q=4$, $m=1$

$$\begin{aligned} \hat{y} = & \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 \\ & + \beta_{12} x_1 x_2 + \beta_{13} x_1 x_3 + \beta_{14} x_1 x_4 + \beta_{23} x_2 x_3 + \beta_{24} x_2 x_4 + \beta_{34} x_3 x_4 \\ & + \beta_{41} x_4 z_1 + \beta_{141} x_1 x_4 z_1 + \beta_{241} x_2 x_4 z_1 + \beta_{341} x_3 x_4 z_1 \\ & + \beta_{1241} x_1 x_2 x_4 z_1 + \beta_{1341} x_1 x_3 x_4 z_1 + \beta_{2341} x_2 x_3 x_4 z_1 \end{aligned}$$

Categorical Factor with Proportion Zero

Four Component and One Categorical



model terms ($q=4$, $m=1$):

A, B, C, D,

AB, AC, AD, BC, BD, CD,

DE,

ADE, BDE, CDE,

ABDE, ACDE, BCDE

Categoric Factor with Proportion Zero

Four Component and One Categoric

The coupled model we specify while building the design is called the “Design Model”:

A, B, C, D, AB, AC, AD, BC, BD, CD,
DE, ADE, BDE, CDE, ABDE, ACDE, BCDE.

Analyzing a response, start with the “Design Model” and reduce it.

It is important to note the design model is not hierarchical, e.g. for ADE there is no AE. Therefore we say “NO” to Design-Expert’s query “Would you like the hierarchy corrected automatically?”

We must check hierarchy ourselves. E.g. if ABDE is selected be sure to include AB, AD, BD, DE, ADE and BDE, but not AE or BE.

When D = 0 then the equations are different

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD$$

$$+ \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce + \beta_{45} De + \beta_{125} ABe + \beta_{135} ACE + \beta_{145} ADe$$

$$+ \beta_{235} BCe + \beta_{245} BDe + \beta_{345} CDe$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_{12} AB + \beta_{13} AC + \beta_{23} BC + \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce$$

$$+ \beta_{125} ABe + \beta_{135} ACE + \beta_{235} BCe \quad \left. \vphantom{\hat{y}} \right\} \text{D is zero}$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD + \beta_{45} De$$

$$+ \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe + \beta_{1245} ABDe + \beta_{1345} ACDe + \beta_{2345} BCDe$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_{12} AB + \beta_{13} AC + \beta_{23} BC \quad \left. \vphantom{\hat{y}} \right\} \text{D is zero}$$

When $D > 0$

then the equations are equivalent

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD \\ + \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce + \beta_{45} De + \beta_{125} ABe + \beta_{135} ACE + \beta_{145} ADe \\ + \beta_{235} BCe + \beta_{245} BDe + \beta_{345} CDe$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_{12} AB + \beta_{13} AC + \beta_{23} BC + \beta_{45} e \\ + \beta_{15} Ae + \beta_{25} Be + \beta_{35} Ce + \beta_{125} ABe + \beta_{135} ACE + \beta_{235} BCe \quad \left. \vphantom{\hat{y}} \right\} \text{D is a fixed value } > 0$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 D + \beta_{12} AB + \beta_{13} AC + \beta_{14} AD + \beta_{23} BC + \beta_{24} BD + \beta_{34} CD + \beta_{45} De \\ + \beta_{145} ADe + \beta_{245} BDe + \beta_{345} CDe + \beta_{1245} ABDe + \beta_{1345} ACDe + \beta_{2345} BCDe$$

$$\hat{y} = \beta_1 A + \beta_2 B + \beta_3 C + \beta_{12} AB + \beta_{13} AC + \beta_{23} BC + \beta_{45} e \\ + \beta_{145} Ae + \beta_{245} Be + \beta_{345} Ce + \beta_{1245} ABe + \beta_{1345} ACE + \beta_{2345} BCe \quad \left. \vphantom{\hat{y}} \right\} \text{D is a fixed value } > 0$$

Categoric Factor with Proportion Zero

Four Component and One Categoric

Since:

- The different levels of a component are represented as a categoric factor.
- Actual proportions of the categoric components are key.
- Strict hierarchy is not maintained.

The analysis should be done using “**Real**” coding.

- **Categoric Mixture Components
Proportion Going to Zero**
- What's the problem?
- **Shelf life example**
- Conclusion

Shelf Life Characterization

Categoric Factor Going to Zero

Optimize a preservative blend to maximize shelf life of a new food:

- One of two additional preservatives (either sulfur dioxide or calcium propionate) will be used.

Component	Range
A: NaCl	0 – 600 mg
B: KCl	0 – 500 mg
C: BHT	0 – 120 mg
D: Preservative	0 – 120 mg
Total =	1100 mg

- MLC: the sum of the antioxidant BHT (butylated hydroxytoluene) and Preservative must be ≤ 130 mg.

Shelf Life Characterization

Degradation Rate

Degradation is a function of time:

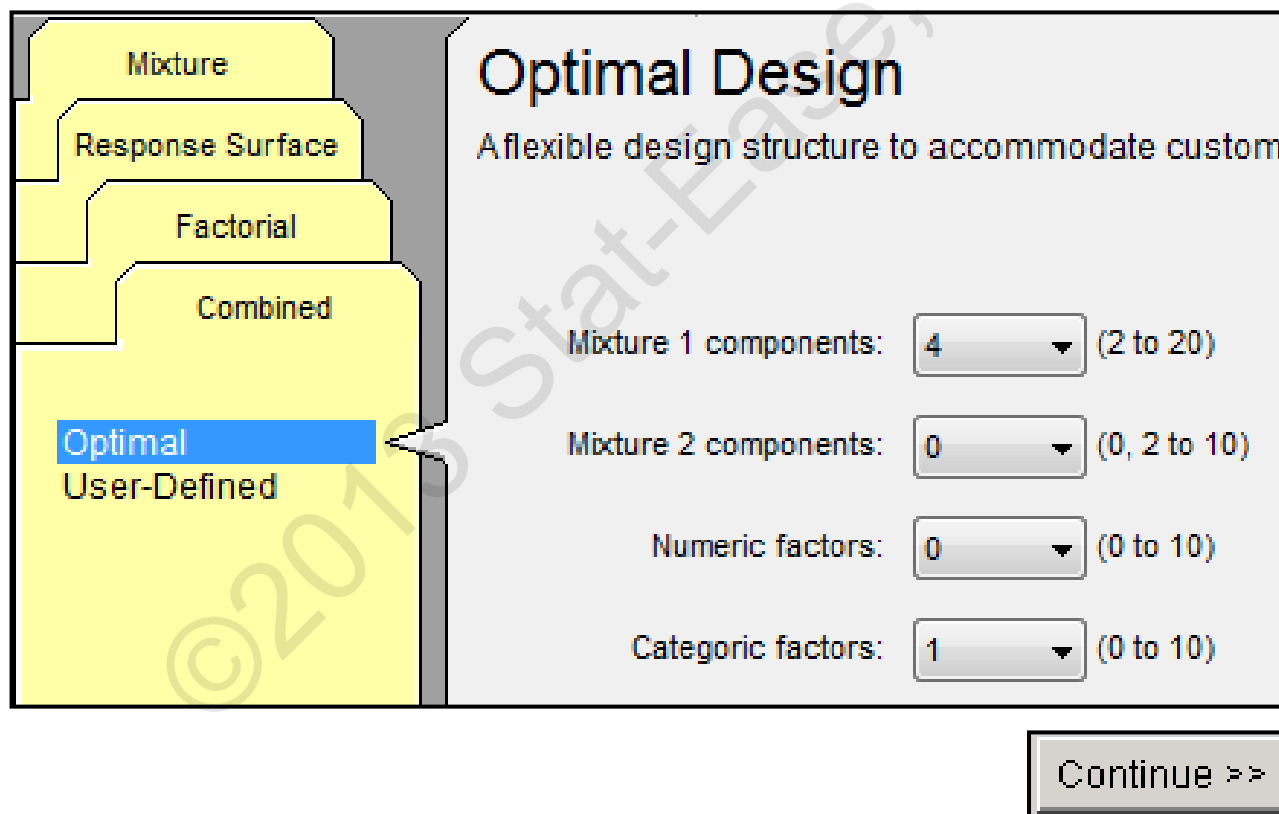
- It is assumed that degradation rate is zero order because of a low concentration of degradation products.
- If degradation is zero order, then a plot of the amount of degradation product versus time is linear with a slope of k .
- The major degradation was measured over time using accelerated life testing for each of the 27 runs and the rate (k) determined for each mixture combination.

Lower rates (k values) are better!

Categorical Factor Going to Zero

Shelf Life Characterization: Build (page 1 of 6)

1. Build an “**Optimal**” **Combined** design with four components and one categorical factor:



The screenshot shows the 'Optimal Design' configuration screen in Stat-Ease. On the left, a vertical menu lists design types: Mixture, Response Surface, Factorial, Combined, Optimal (highlighted in blue), and User-Defined. A callout box points from the 'Optimal' option to the main configuration area. The main area is titled 'Optimal Design' and has the subtitle 'A flexible design structure to accommodate custom'. It contains four configuration rows, each with a label, a dropdown menu, and a range in parentheses:

- Mixture 1 components: 4 (2 to 20)
- Mixture 2 components: 0 (0, 2 to 10)
- Numeric factors: 0 (0 to 10)
- Categorical factors: 1 (0 to 10)

At the bottom right, there is a 'Continue >>' button.

Categorical Factor Going to Zero

Shelf Life Characterization: Build (page 2 of 6)

2. Enter mixture components:

Mix Components: 4 Total:

Horizontal Vertical Units:

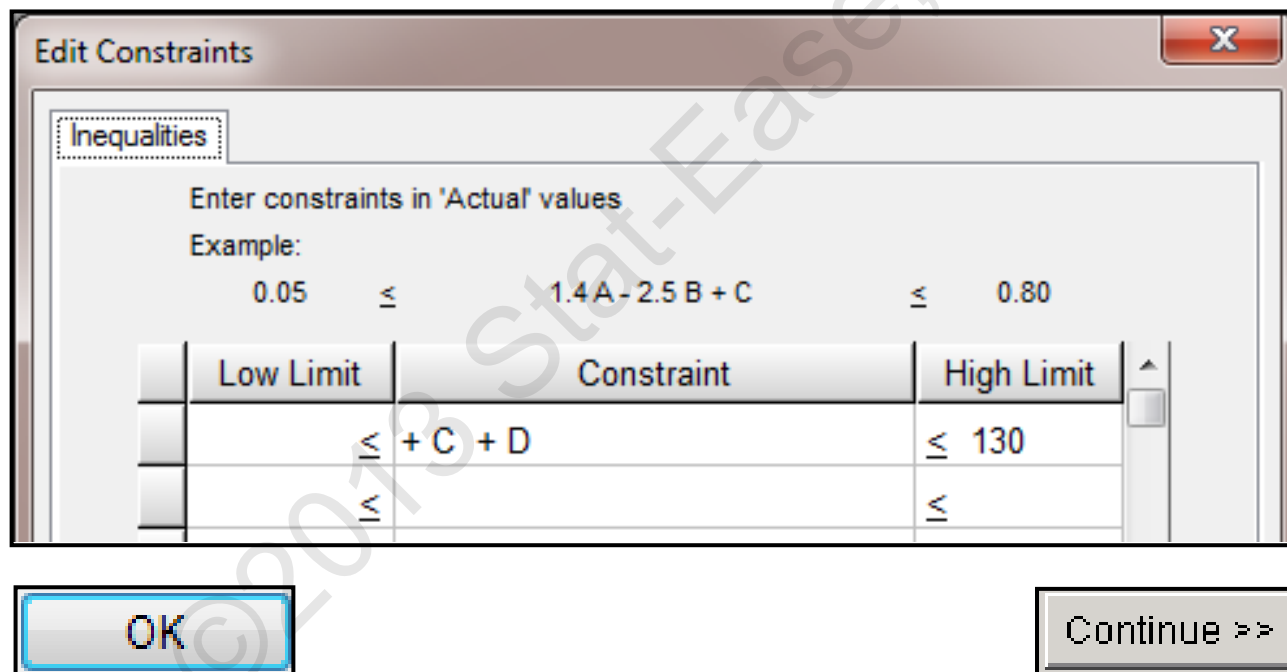
	Name	Low	High
A [Mixture]	NaCl	0	600
B [Mixture]	KCl	500	1000
C [Mixture]	BHT	0	120
D [Mixture]	Preservative	0	120

enter the MLC constraint on the next slide.

Categorical Factor Going to Zero

Shelf Life Characterization: Build (page 3 of 6)

3. Click on the “Edit constraints...” button and enter the constraint for the maximum of C (BHT) + D(preservative) ≤ 130 mg :



Edit Constraints

Inequalities

Enter constraints in 'Actual' values

Example:

0.05 ≤ 1.4A - 2.5B + C ≤ 0.80

Low Limit	Constraint	High Limit
	≤ + C + D	≤ 130
	≤	≤

OK

Continue >>

Categorical Factor Going to Zero

Shelf Life Characterization: Build *(page 4 of 6)*

4. Enter categorical factors:

Numeric Factors: 0

Categorical factors: 1

Horizontal

Vertical

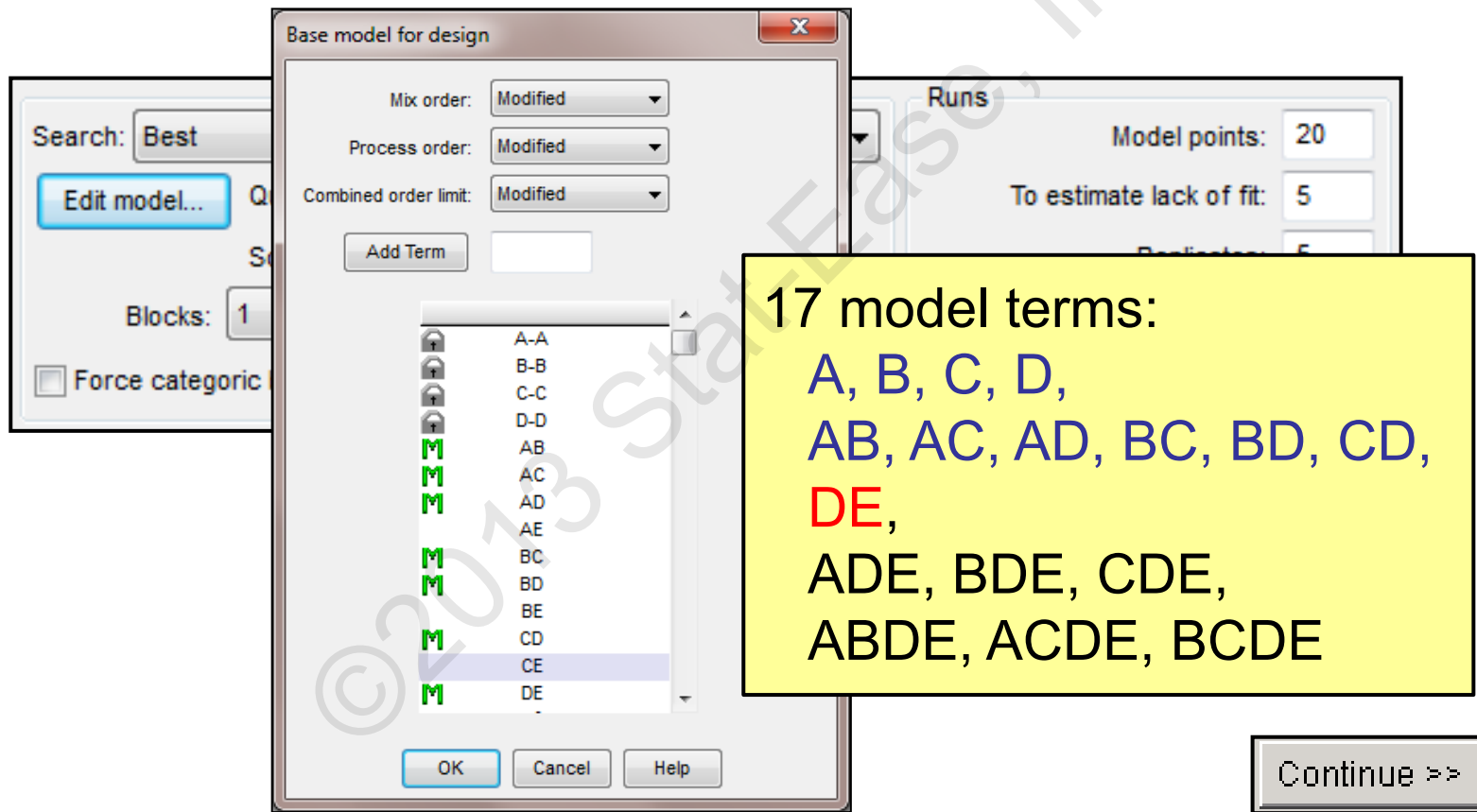
	E [Categorical]
Name	Species
Units	preservative
Type	Nominal
Levels	2
L[1]	S dioxide
L[2]	Ca propionate

Continue >>

Categorical Factor Going to Zero

Shelf Life Characterization: Build (page 5 of 6)

5. Click **“Edit model...”** and design for a “going to zero” model:



Base model for design

Mix order: Modified
Process order: Modified
Combined order limit: Modified

Add Term

Search: Best
Edit model...
Blocks: 1
Force categoric

Runs
Model points: 20
To estimate lack of fit: 5

17 model terms:
A, B, C, D,
AB, AC, AD, BC, BD, CD,
DE,
ADE, BDE, CDE,
ABDE, ACDE, BCDE

Continue >>

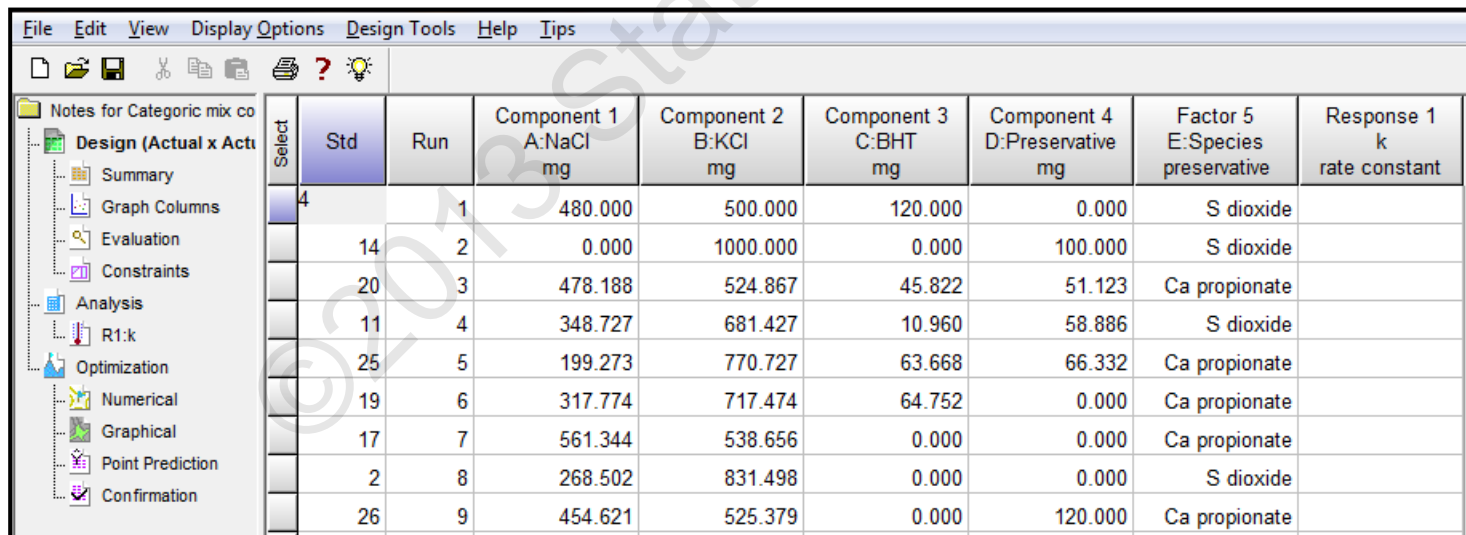
Categoric Factor Going to Zero

Shelf Life Characterization: Build (page 6 of 6)

6. There is one response:

Name	Units
k	rate constant

Continue >>

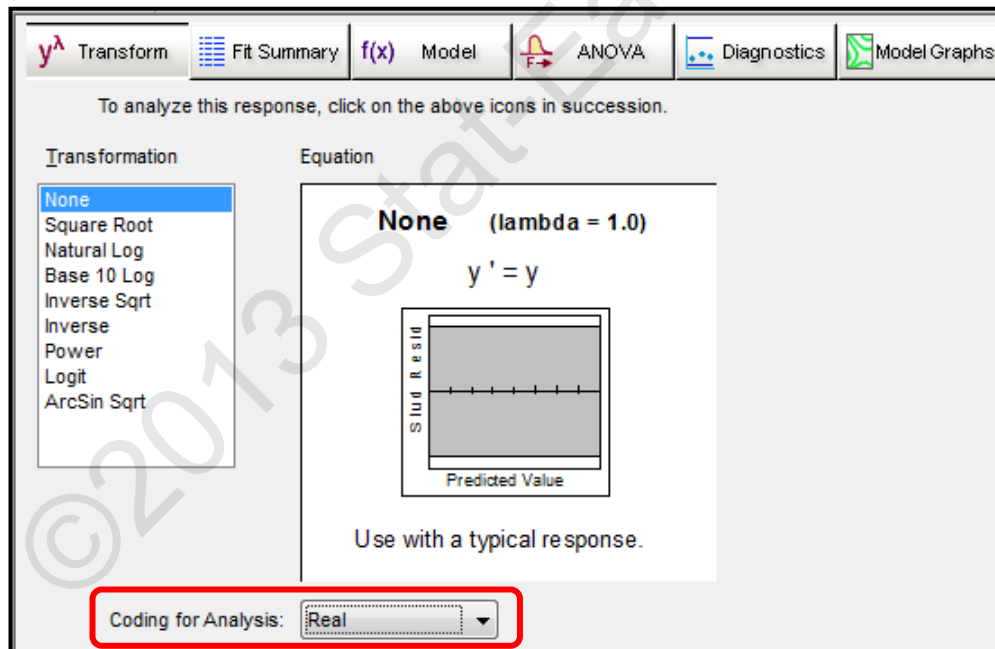


Select	Std	Run	Component 1 A:NaCl mg	Component 2 B:KCl mg	Component 3 C:BHT mg	Component 4 D:Preservative mg	Factor 5 E:Species preservative	Response 1 k rate constant
4		1	480.000	500.000	120.000	0.000	S dioxide	
	14	2	0.000	1000.000	0.000	100.000	S dioxide	
	20	3	478.188	524.867	45.822	51.123	Ca propionate	
	11	4	348.727	681.427	10.960	58.886	S dioxide	
	25	5	199.273	770.727	63.668	66.332	Ca propionate	
	19	6	317.774	717.474	64.752	0.000	Ca propionate	
	17	7	561.344	538.656	0.000	0.000	Ca propionate	
	2	8	268.502	831.498	0.000	0.000	S dioxide	
	26	9	454.621	525.379	0.000	120.000	Ca propionate	

Categorical Factor Going to Zero

Shelf Life Characterization: Analysis (page 1 of 2)

1. Right click on the response column header and “Simulate Response” using “**degradation rate.sim**”.
2. Remember: the analysis should be done use “**Real**” coding:



To analyze this response, click on the above icons in succession.

Transformation

- None
- Square Root
- Natural Log
- Base 10 Log
- Inverse Sqrt
- Inverse
- Power
- Logit
- ArcSin Sqrt

Equation

None (lambda = 1.0)

$$y' = y$$

Standard Residual

Predicted Value

Use with a typical response.

Coding for Analysis: Real

Categorical Factor Going to Zero

Shelf Life Characterization: Analysis (page 2 of 2)

3. Fit an appropriate model to the response.

This requires skipping the fit summary and using the “**Design Model**”.



(Say “**No**” to hierarchy – see notes below.)

4. Recommend a formula to minimize degradation rate (k):

NaCl _____ mg

Preservative _____ mg of

KCl _____ mg

“Sulfur dioxide” or “Calcium propionate”

BHT _____ mg

↑ _____ (circle one) _____ ↑

- **Categoric Mixture Components
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Multiple Categorical Factors

Going to Zero (without categorical interaction)

x_{q-m+k} is the proportion of a component and z_k is the levels (categorical level) of that component:

$$\eta(x, z) = \sum_{i=1}^q \gamma_i^0 x_i + \sum_{i < j}^q \gamma_{ij}^0 x_i x_j + \sum_{k=1}^m \left[\gamma_k^0 x_{q-m+k} z_k + \left[\sum_{i=1}^{q-m} \gamma_i^1 x_i + \sum_{i < j}^{q-m} \gamma_{ij}^1 x_i x_j \right] (x_{q-m+k} z_k) \right]$$

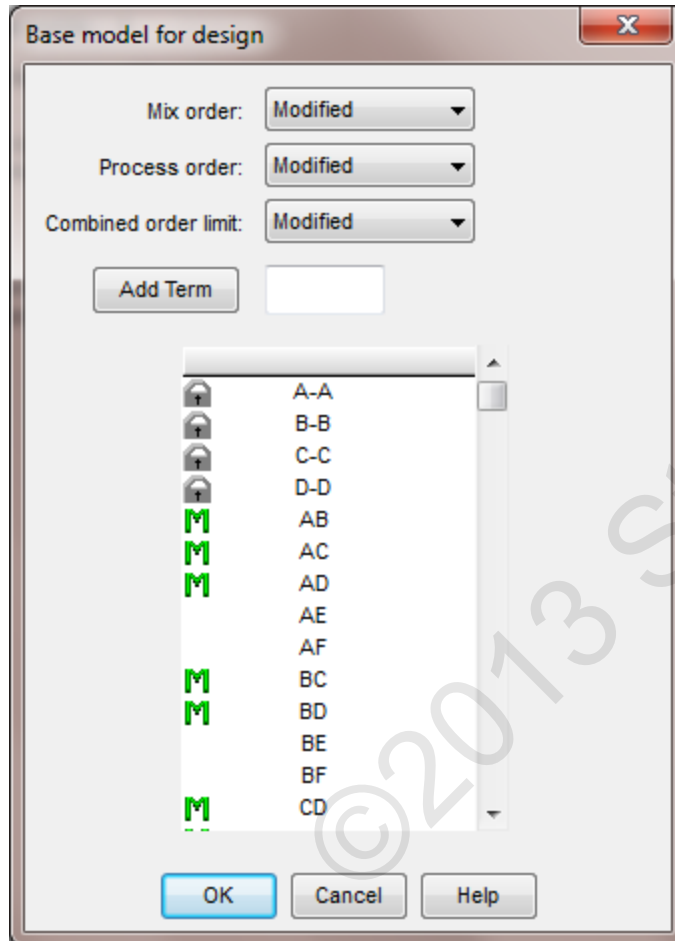
q – number of components

m – number of components with different levels

Model terms for q=4 and m=2:

A, B, C, D, AB, AC, AD, BC, BD, CD,
CE, ACE, BCE, ABCE, DF, ADF, BDF, ABDF

Two Components with Categorical Factors Going to Zero



model terms ($q=4$, $m=2$):

A, B, C, D,

AB, AC, AD, BC, BD, CD,

CE, DF,

ACE, BCE, ABCE,

ADF, BDF, ABDF

Multiple Categorical Factors

Going to Zero (with categoric interaction)

x_{q-m+k} is the proportion of a component and z_k is the levels (categoric level) of that component:

$$\eta(x, z) = \sum_{i=1}^q \gamma_i^0 x_i + \sum_{i < j} \sum_{j}^q \gamma_{ij}^0 x_i x_j + \sum_{k=1}^m \left[\gamma_k^0 x_{q-m+k} z_k + \left[\sum_{i \neq k}^q \gamma_i^1 x_i + \sum_{i \neq k < j \neq k}^q \gamma_{ij}^1 x_i x_j \right] (x_{q-m+k} z_k) \right] + \sum_k^{m-1} (x_{q-m+k} z_k) (x_{q-m+k+1} z_{k+1})$$

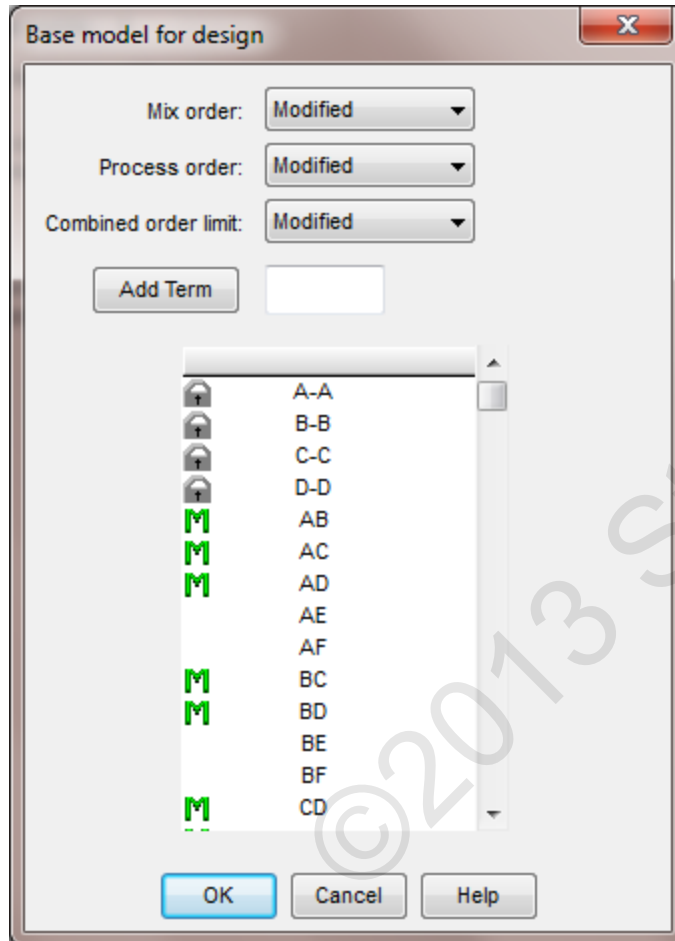
q – number of components

m – number of components with different levels

Model terms for $q=4$ and $m=2$:

A, B, C, D, AB, AC, AD, BC, BD, CD,
CE, ACE, BCE, DCE, ABCE, ADCE, BDCE,
DF, ADF, BDF, CDF, ABDF, ACDF, BCDF, **CEDF**

Two Components with Categorical Factors Going to Zero



Quadratic model terms
($q=4$, $m=2$):

A, B, C, D,
AB, AC, AD, BC, BD, CD,
CE, ACE, BCE, CDE,
ABCE, ACDE, BCDE,
DF, ADF, BDF, CDF,
ABDF, ACDF, BCDF,
CDEF