



# UTILIZING EXTRAPOLATION CONTROL FEATURE IN JMP® PRO 16 TO OPTIMIZE REAL-WORLD MINING PROCESS

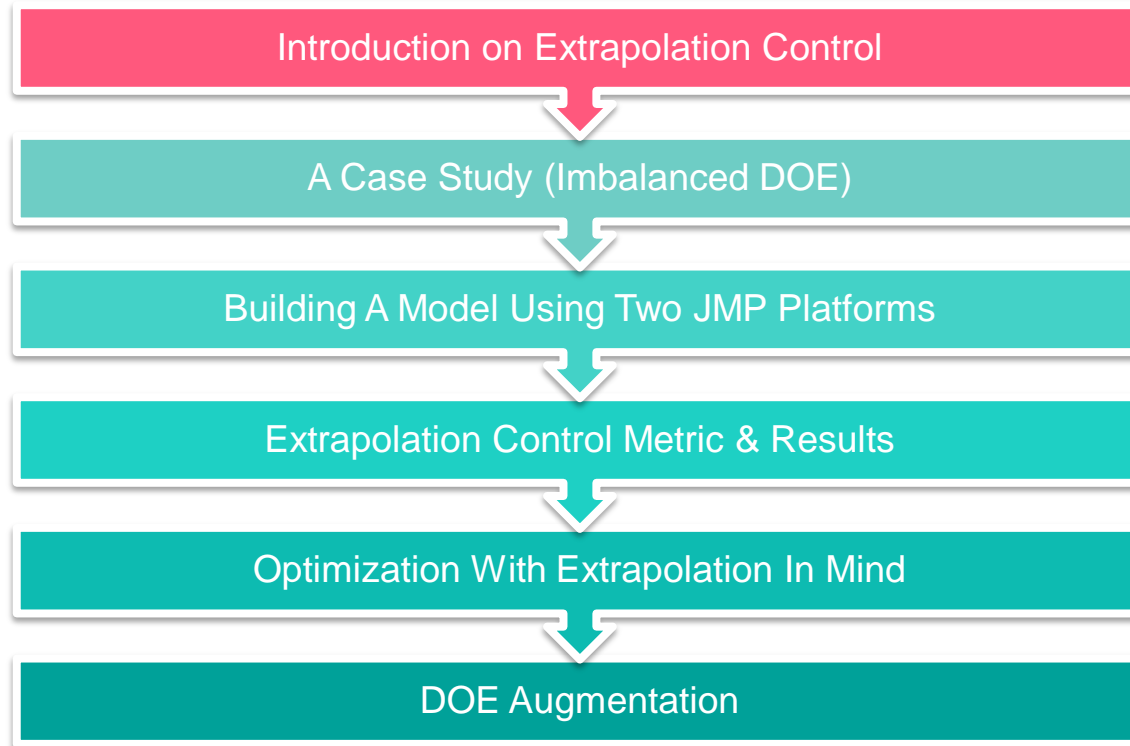
XINJIE TONG

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PRESENT AT DISCOVERY SUMMIT AMERICAS 2021

# OUTLINE

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# JMP 16 Pro - EXTRAPOLATION CONTROL CRITERIA\*

## ■ Linear Model

- Use **leverage** (hat) to detect extrapolation.
- Available for Fit Least Squares.
- $Leverage(x) = x^T (X^T X)^{-1} x$  where  $x$  is the prediction point and  $X$  is the design matrix of training data
- For a linear model, leverage/hat is equivalent to a scaled prediction variance (for mean).  
 $standard\ error = RMSE * \sqrt{h_x}$
- Extrapolation happens when
  - ✓  $h > h_{max}$
  - ✓  $h > 3h_{avg} = 3p/n$
- Hat is a function of model terms and design matrix.

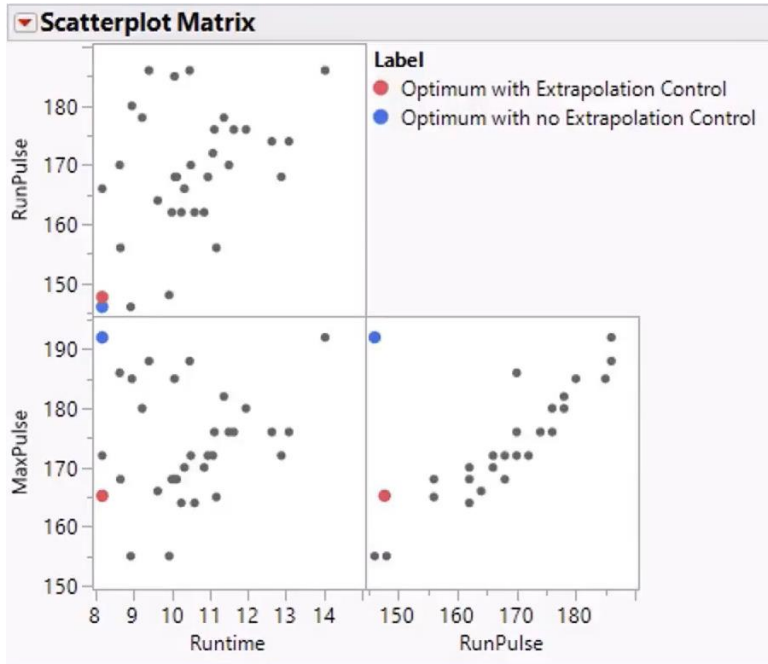
## ■ Nonlinear Model

- Use **regularized Hotelling's T square** to detect extrapolation (unsupervised). Not a function of model terms.
- Extrapolation happens when  $> \text{mean} + 3 * \text{standard deviation of the sample regularized T Squares}$
- Available for Generalized Regression, Neural, Support Vector Machines, Partial Least Squares, Naïve Bayes

Hotelling's  $T^2$  defined as:

$$T^2 = (x - \bar{x})^T \hat{\Sigma}^{-1} (x - \bar{x})$$

## EXAMPLE OF EXTRAPOLATION CONTROL FOR LINEAR MODEL



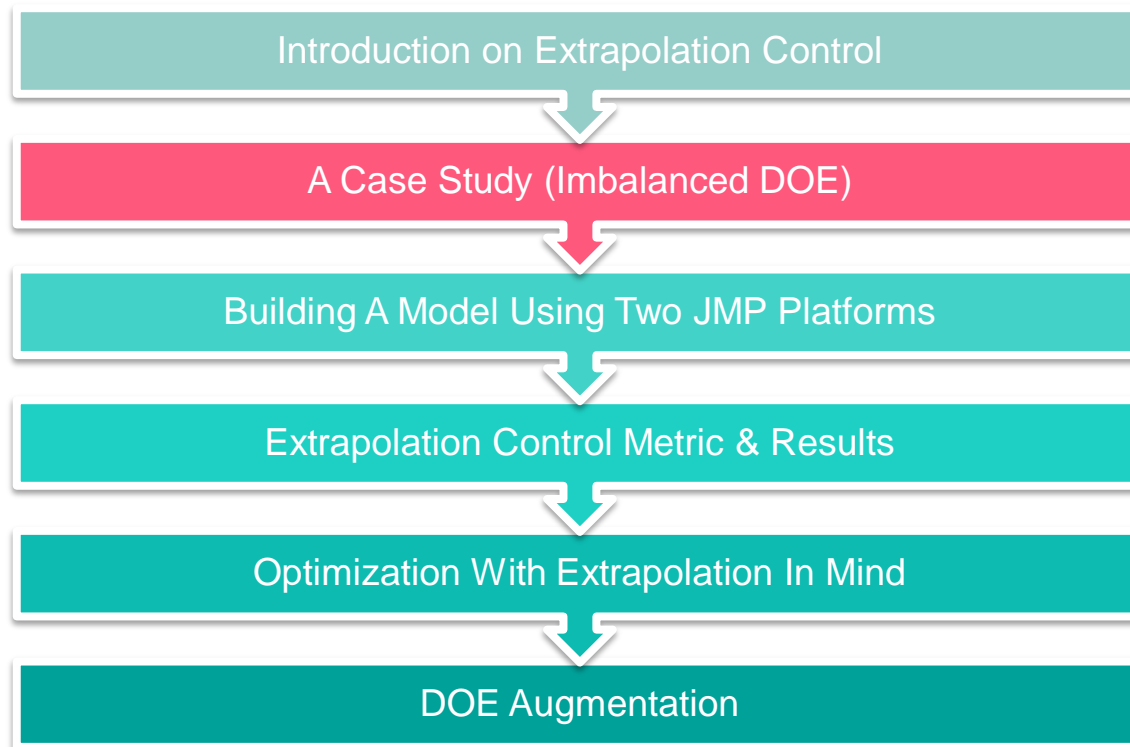
An example from SAS video\*.

- Model = least squares
- Extrapolation metric = leverage
- Threshold =  $3 \times$  average leverage

Extrapolation control detects the conditions that do not follow the relationship among input variables.

# OUTLINE

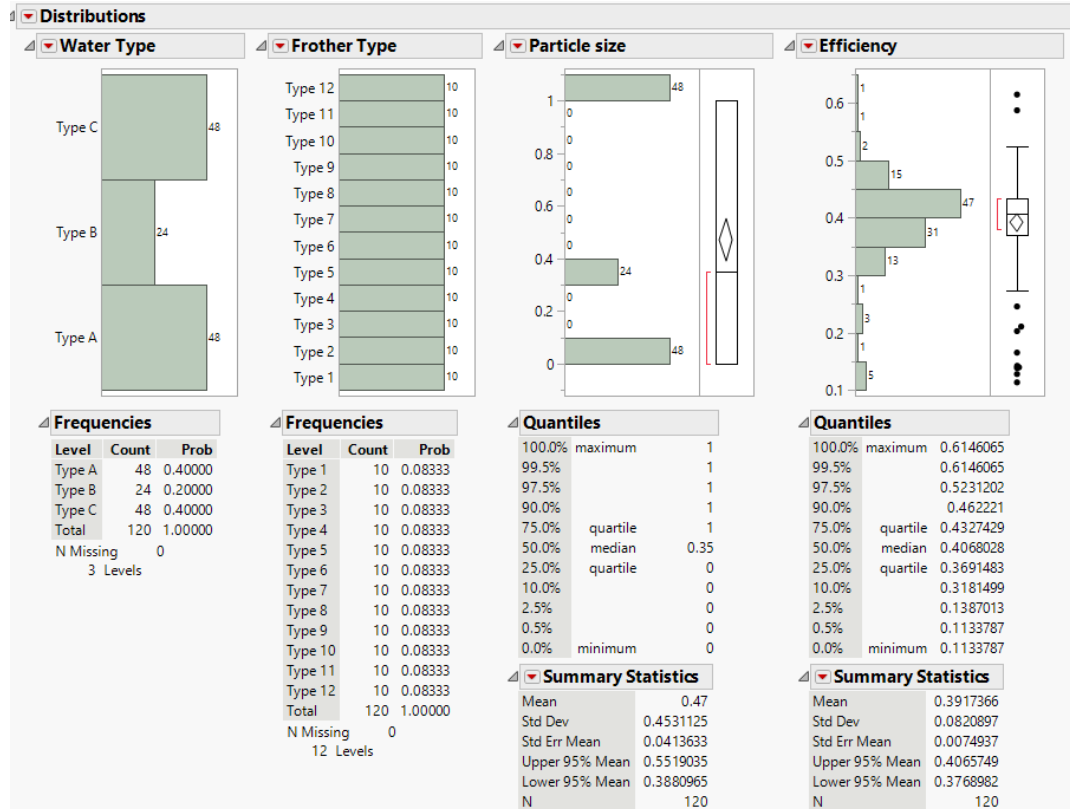
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# THE CASE STUDY

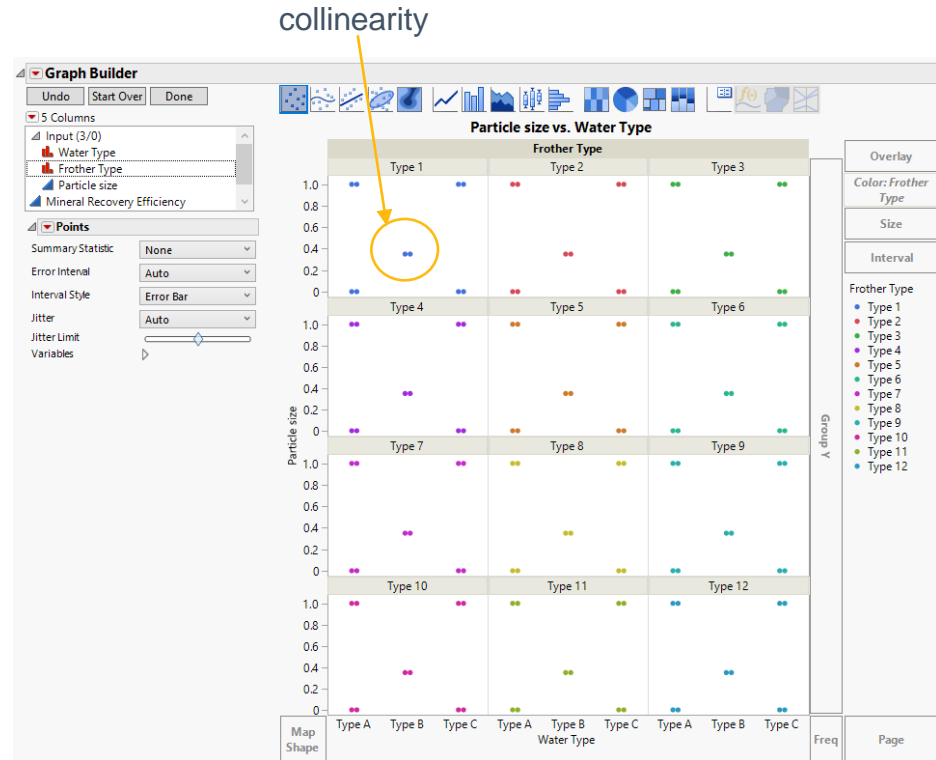
- Objective: optimize mining conditions for enhanced efficiency
- Input factors:
  1. Water type (categorical) – 3 levels
  2. Frother type (categorical) – 12 levels
  3. Particle size (continuous) – 3 levels
- Output factor:
  1. Efficiency (a value between 0 and 1)

All values of the input variables and the response variable were coded.



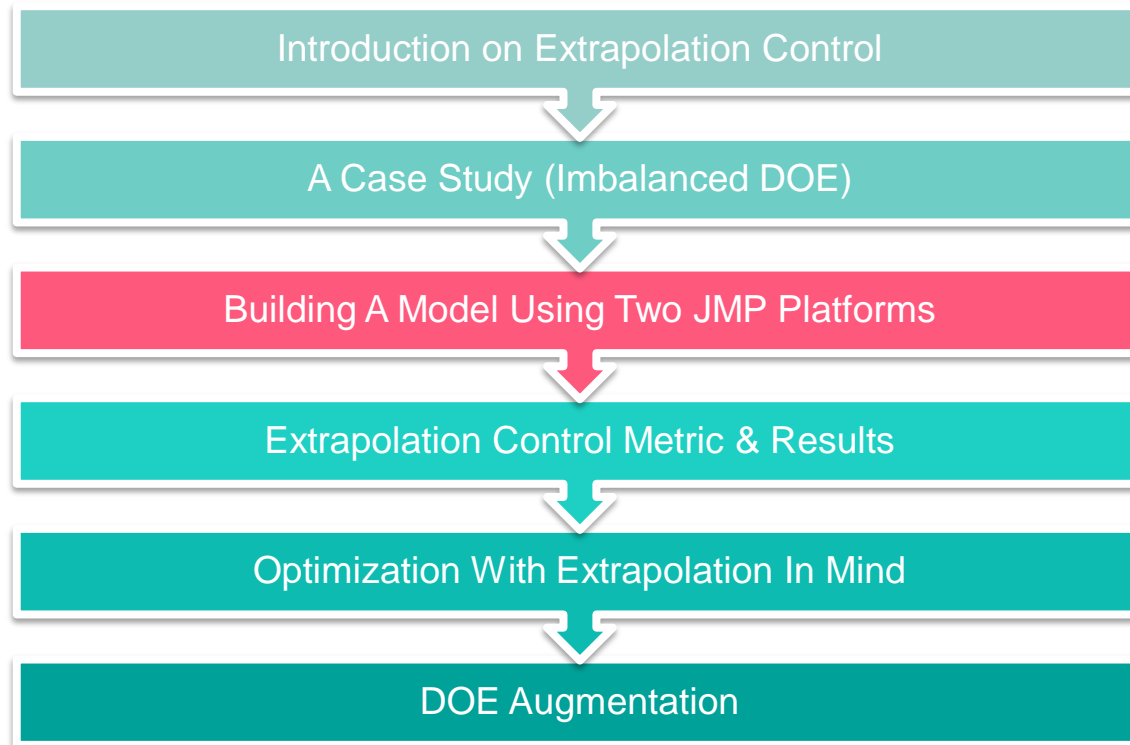
# THE DATA IMBALANCE

- Collinearity exists for Water Type B.
- Extrapolation control can help prevent the risks associated with imbalanced data.



# OUTLINE

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# EVALUATE DESIGN

▾ Evaluate Design

▸ Factors

▾ Model

Main Effects Interactions ▼ RSM Cross Powers ▼ Remove Term

Intercept  
Water Type  
Frother Type  
Particle size  
Water Type\*Frother Type  
Frother Type\*Particle size

Cannot estimate Particle Size & Water Type interaction

# FIT MODEL USING TWO PLATFORMS IN JMP

## Generalized Regression (nonlinear)

Fit Model 2 - JMP Pro

**Model Specification**

Select Columns: **Efficiency**

Pick Role Variables: Y: **Efficiency** (optional)

Personality: **Generalized Regression**

Distribution: **Beta**

Buttons: Help, Run, Recall, Remove, Keep dialog open

Construct Model Effects: Add, Cross, Nest, Macros, Degree: 2, Attributes, Transform, No Intercept

**Actual by Predicted Plot**

## Fit Least Square (linear)

Report: Fit Model - JMP Pro

**Model Specification**

Select Columns: 13 Columns (Input (3/0), Water Type, Frother Type, Particle size, Replication, Treatment, Efficiency)

Pick Role Variables: Y: **Efficiency** (optional)

Personality: **Standard Least Squares**

Emphasis: **Effect Leverage**

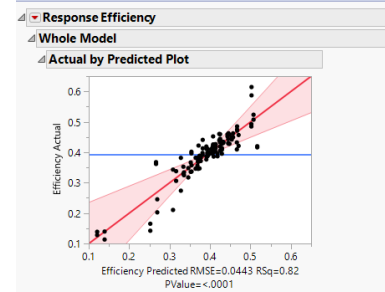
Buttons: Help, Run, Recall, Remove, Keep dialog open

Construct Model Effects: Add, Cross, Nest, Macros, Degree: 2, Attributes, Transform, No Intercept

**Generalized Regression for Efficiency**

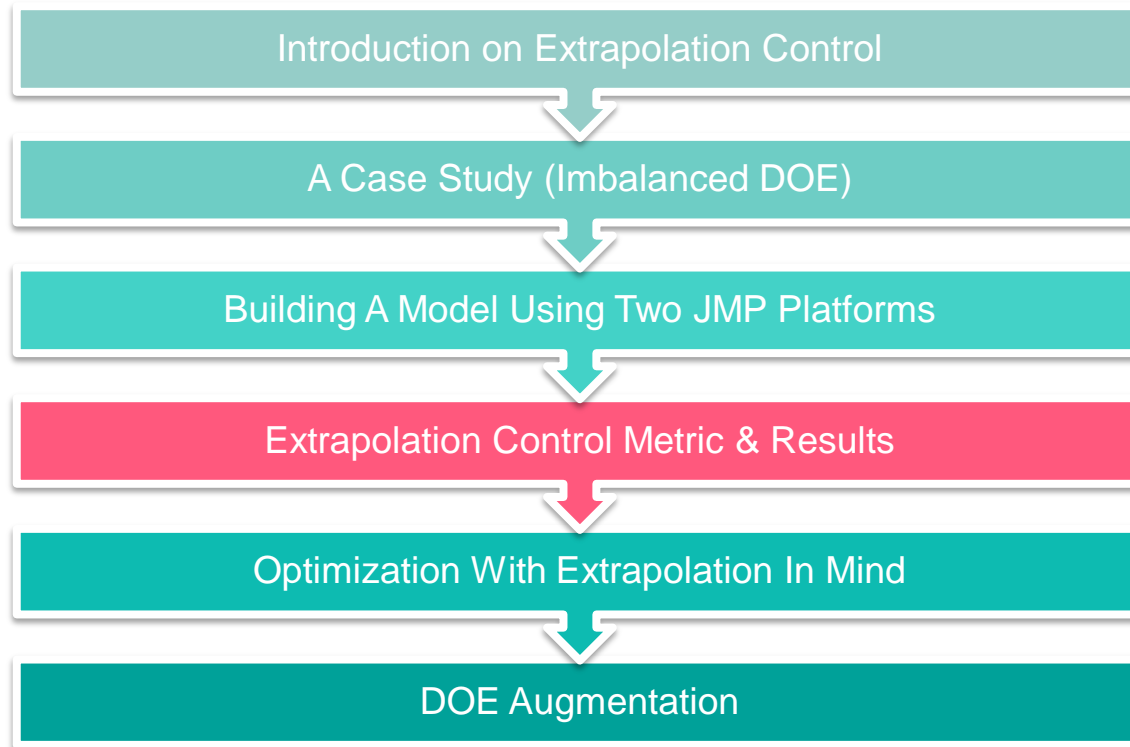
**Model Comparison**

Show	Response Distribution	Estimation Method	Validation Method	Nonzero Parameters	AICc	BIC	Generalized RSquare
<input type="checkbox"/>	Beta	Maximum Likelihood	None	49	-289.2205	-222.6334	0.8344784
<input type="checkbox"/>	Beta	Forward Selection	AICc	47	-292.137	-223.7915	0.8224517
<input type="checkbox"/>	Beta	Backward Elimination	AICc	47	-292.137	-223.7915	0.8224517
<input type="checkbox"/>	Beta	Pruned Forward Selection	AICc	47	-292.137	-223.7915	0.8224517
<input type="checkbox"/>	Beta	Elastic Net	AICc	28	-322.5243	-262.3207	0.7251609
<input type="checkbox"/>	Beta	Lasso	AICc	28	-323.1779	-262.9743	0.7266538
<input type="checkbox"/>	Normal	Standard Least Squares	None	49	-300.7216	-234.1345	0.8237122
<input checked="" type="checkbox"/>	Normal	Lasso	AICc	28	-334.4061	-274.2025	0.7082116

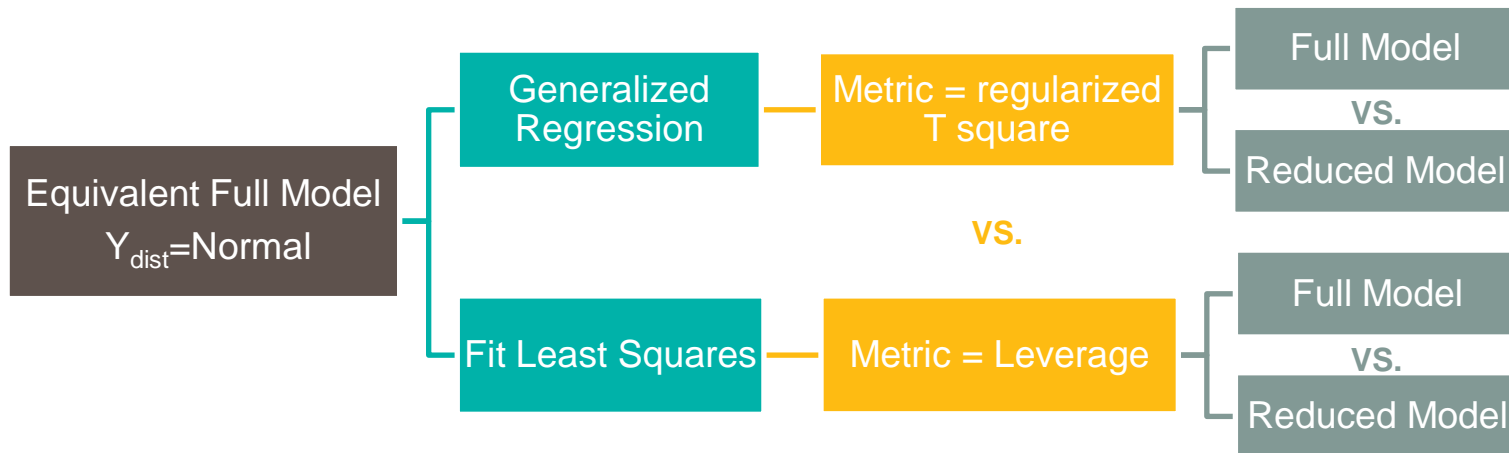


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# COMPARISON OF EXTRAPOLATION MATRICS



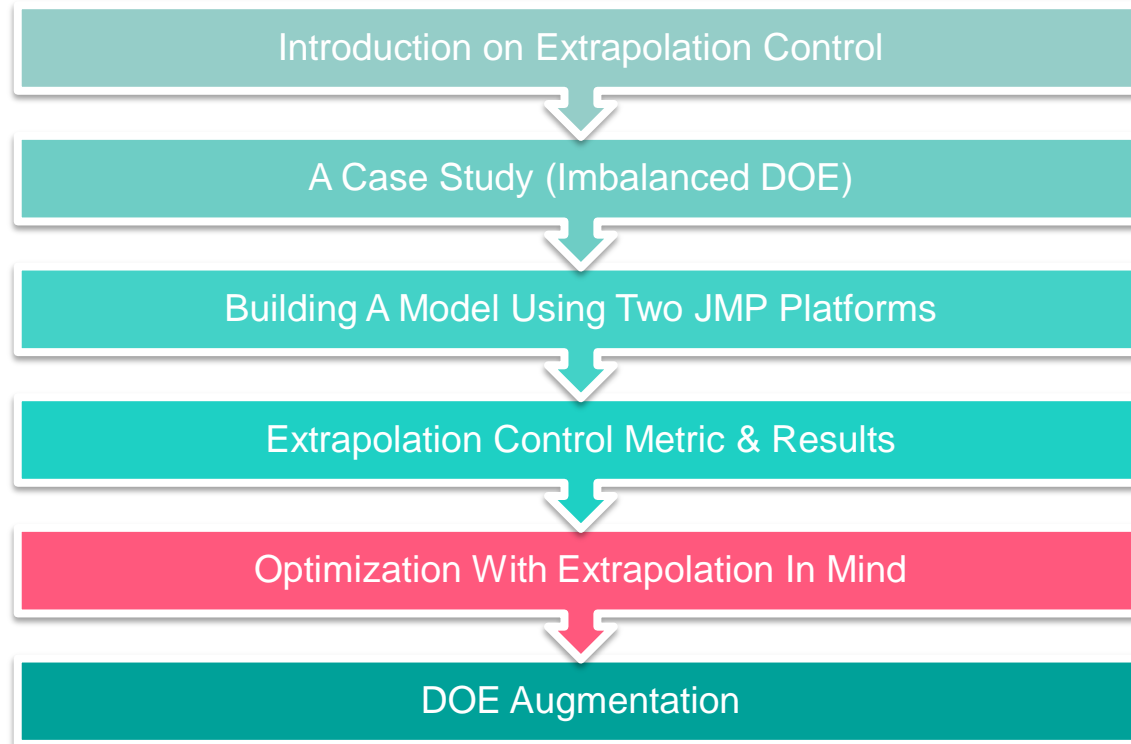
## COMPARISON

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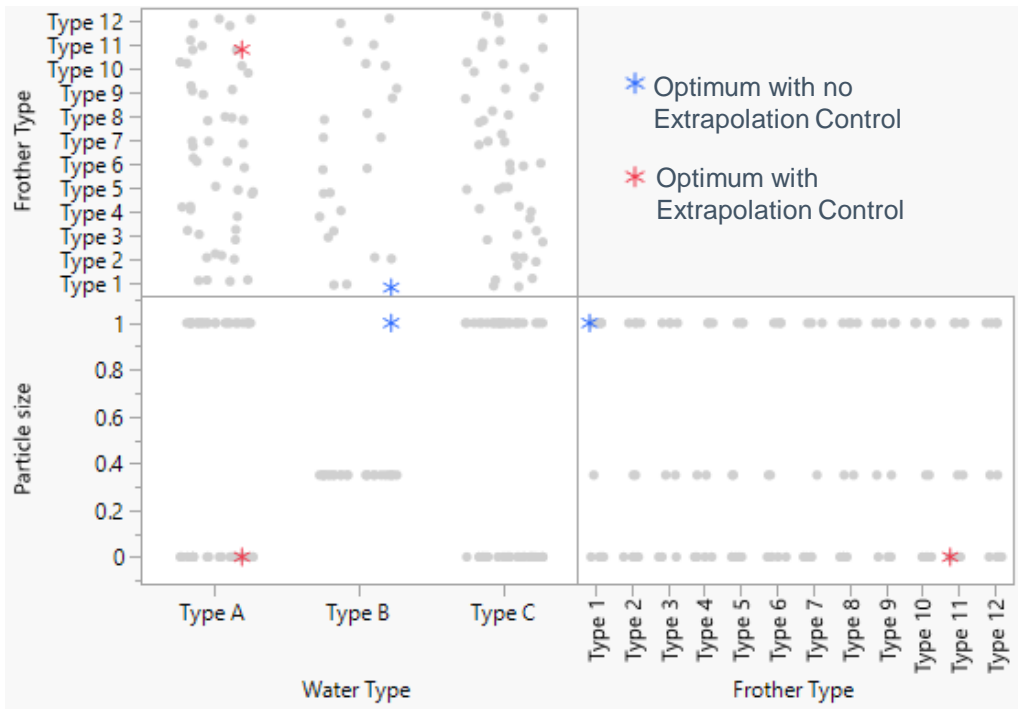
	Extrapolation Control Metric			
	Leverage		Regularized T Square	
Model	Full	Reduced	Full	Reduced
Threshold	0.50	0.13	16.3	16.3
Dependency of model form & design matrix	Yes		No	
Strictness	Strict		Flexible	

# OUTLINE

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# OPTIMIZATION

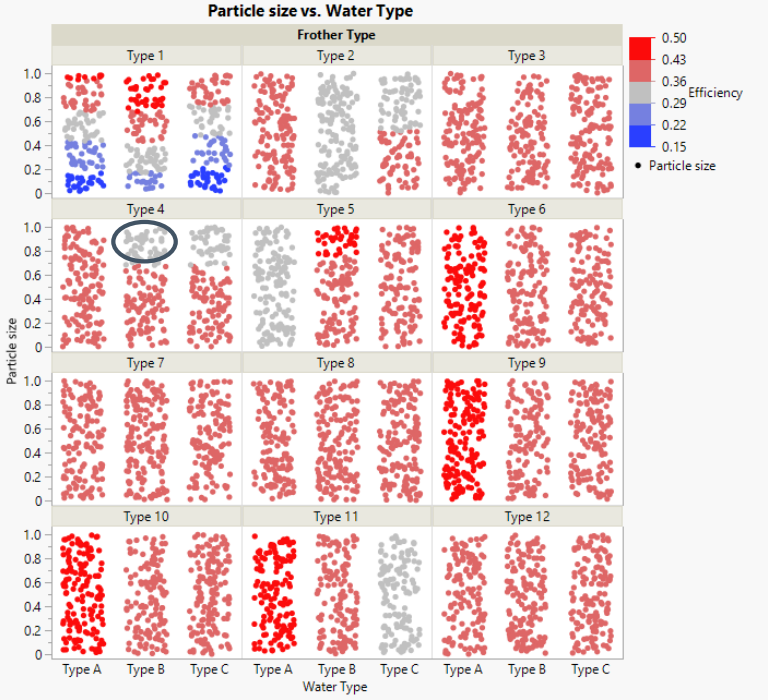


- Model = least squares
- Extrapolation metric = leverage
- Threshold =  $3 \times \text{average leverage}$

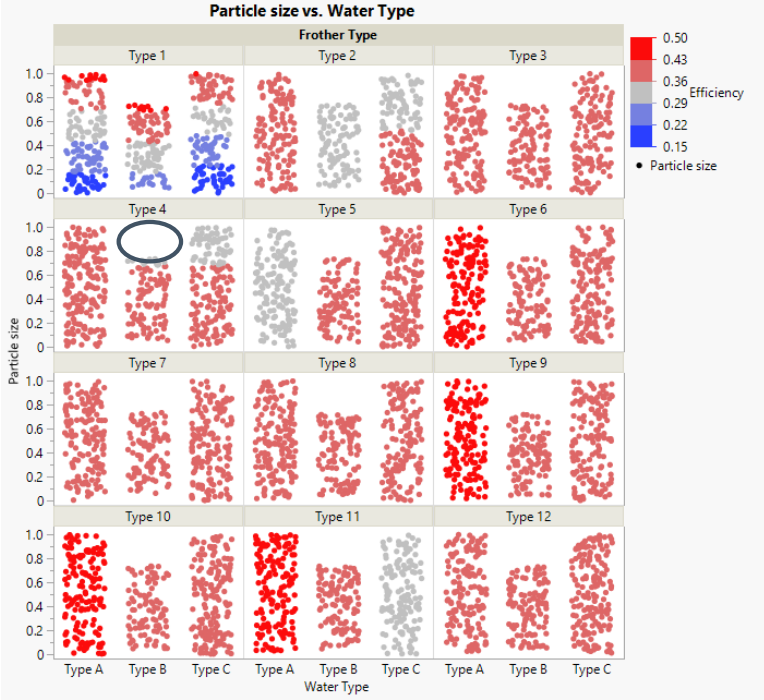
# APPLICATION OF THE MODEL

(GenReg;  $Y_{dist}$ =Normal, Lasso with AICc)

With No Extrapolation Control



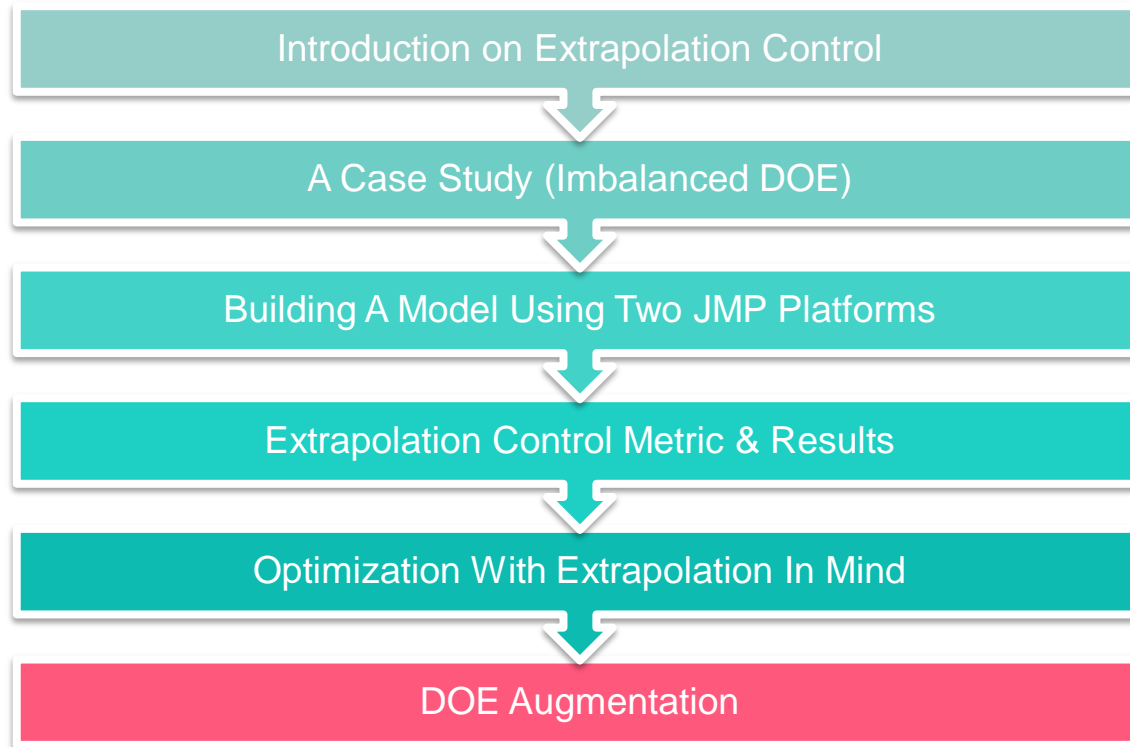
With Extrapolation Control





# OUTLINE

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# DOE AUGMENTATION

**Augment Design**

**Factors**

Name	Role	Changes	Values
Water Type	Categorical	Easy	Type A   Type B   Type C
Frother Type	Categorical	Easy	Type Type Type Type Type Type Type Type Type Type Type
Particle size	Continuous	Easy	0   1
Block	Blocking	Easy	1   2

Group new runs into separate block

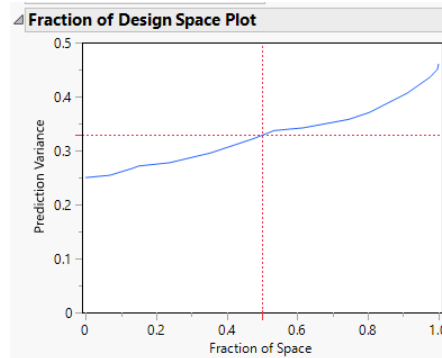
**Define Factor Constraints**

**Model**

Main Effects | Interactions | RSM | Cross | Powers | Remove Term

Name	Estimability
Intercept	Necessary
Water Type	Necessary
Frother Type	Necessary
Particle size	Necessary
Block	Necessary
Water Type*Frother Type	Necessary
Water Type*Particle size	Necessary
Frother Type*Particle size	Necessary

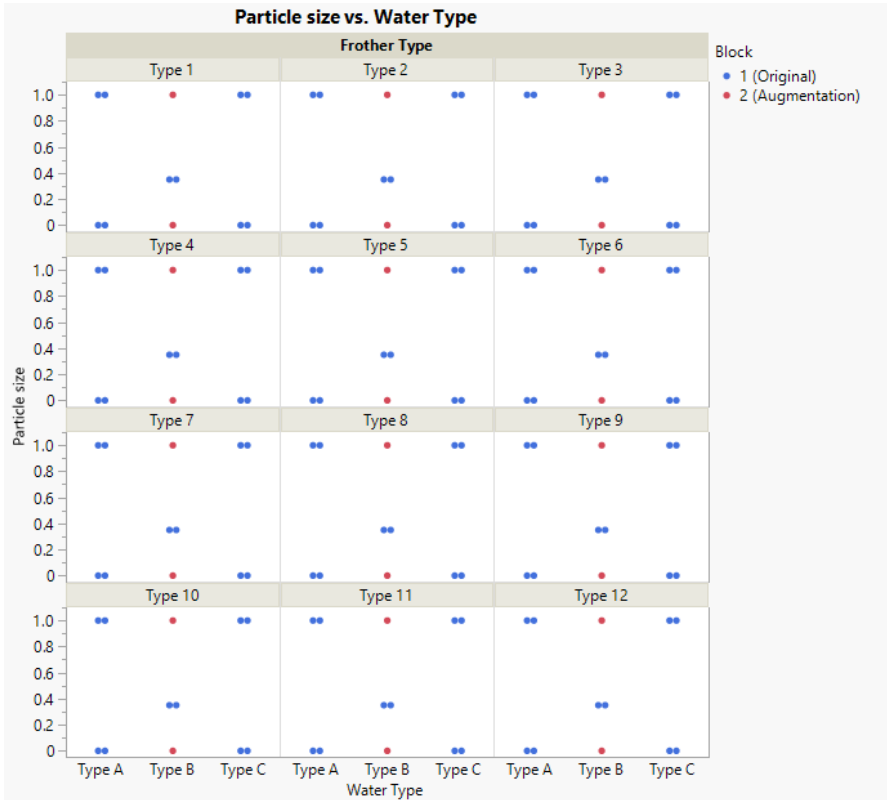
120 → 144 runs



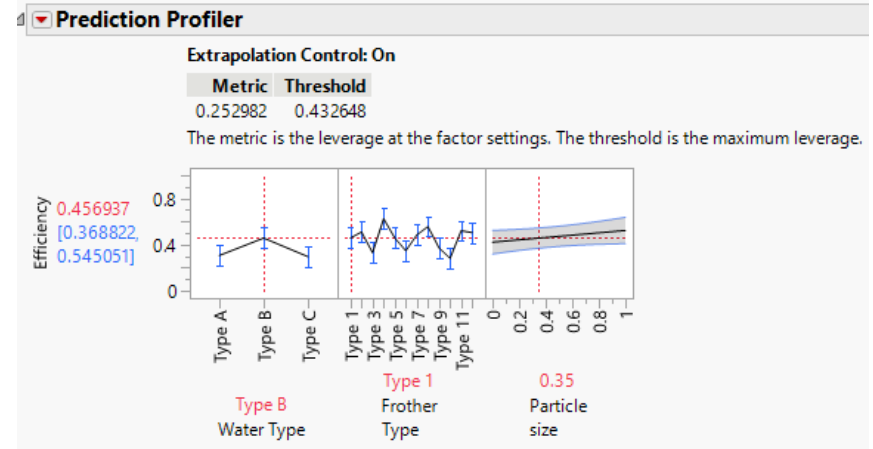
**Design Diagnostics**

D Optimal Design	
D Efficiency	93.00431
G Efficiency	76.82095
A Efficiency	87.2535
Average Variance of Prediction	0.326821
Design Creation Time (seconds)	21

# THE AUGMENTED DOE – NO EXTRAPOLATION ISSUE



Leverage  
Threshold = 0.43  
DOE size = 144



The augmented DOE does not have extrapolation issue.



## CONCLUSION

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- Sources of extrapolation:
  - Correlation among input variables
  - Inestimable model terms
- Extrapolation control is the “sanity check” for the use of predictive model.
- The extrapolation control feature in JMP 16 Pro made it simple to practice, and provide easy understanding on the effectiveness of augmenting a dataset.

## REFERENCE

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- Lancaster, L.; Ash, J.; Gotwalt, C. Controlling Extrapolation in the Prediction Profiler in JMP Pro 16. SAS. <https://community.jmp.com/t5/Discovery-Summit-Europe-2021/Controlling-Extrapolation-in-the-Prediction-Profiler-in-JMP-Pro/ta-p/349218>

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  - Contact: Xinjie Tong, [xtong1@dow.com](mailto:xtong1@dow.com), Statistician at Dow Inc.