

### MIXTURE DESIGN OF EXPERIMENTS USING CUSTOM DOE PLATFORM

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#### WHY USE DOE? QUICKER ANSWERS, LOWER COSTS, SOLVE BIGGER PROBLEMS, MAKE MORE MONEY!

- More rapidly answer "what if?" questions
- Do sensitivity and trade-space analysis
- Optimize across multiple responses
- By running efficient subsets of all possible combinations, one can for the same resources and constraints – solve bigger problems
- By running sequences of designs one can be as *cost effective as possible* & run *no more trials than are needed* to get a useful answer

### SAME HOLDS TRUE FOR MIXTURE DOE





#### AGENDA

- Do trade-space analysis using models fit to a mixture DOE
- What makes mixture factors (components) and formulation DOE different?
- Several Examples
  - Simple three-component designs using Custom DOE platform "Make Designs Fit the Problem – NOT Make Problems Fit the Designs!"
  - Five-component mixture DOE with 3 constraints and response data (revisited)
    - Visualizing process in Fit Model platform
    - Use transformation to prevent physically impossible predictions
  - PDF Ten-factor = 6 mixture, 2 continuous, 1 categorical and 1 block
    - "Real-world" several type of factors
    - Additional constraints including holding some of mixture constant
  - PDF Seven-component mixture DOE with 5 and 7 constraints
    - · Use constraints to define "mixtures within mixture"
    - Can I find a 3-component blend that's nearly as good as a 7-component blend?
  - Technically Speaking Optimizing Performance of a Multi-Layer Packaging Film
    - · Layer thickness expressed as proportions that sum to one = the mixture constraint
    - · Can I trade-off thickness and layer resin concentration to target 2 performance metrics & minimize a third?
  - Computational Chemistry Space-Filling Mixture Design
    - US Army explosive formulation of "Bread."
    - Presented last week at JMP Discovery Summit 2017, St. Louis, MO, USA



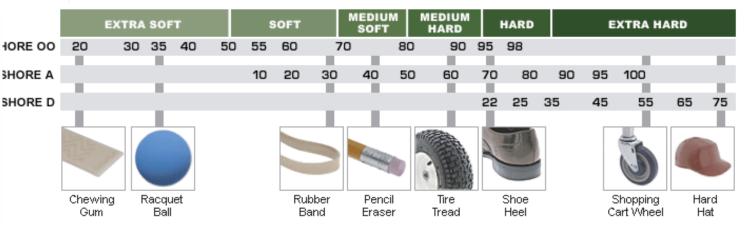




#### NEED TO PREDICT HARDNESS AND COST OF PLASTIC

#### WANT TO MAKE INFORMED BUSINESS DECISIONS TRADING OFF PRODUCT PERFORMANCE AND COST

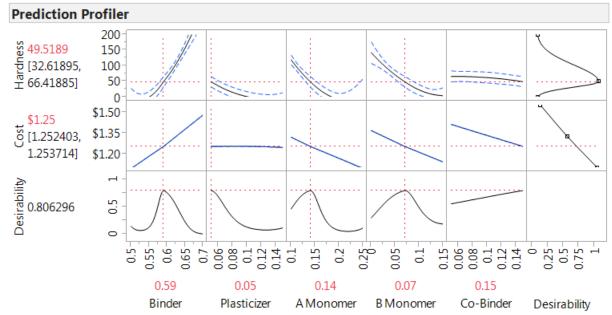




What formulations yield a Shore A hardness of 50?

What do these formulations cost?

Can I trade-off hardness and cost?



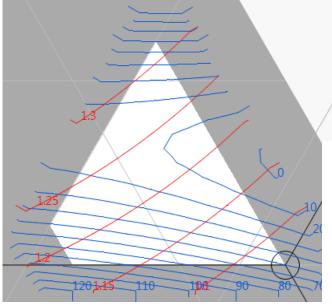
**Sas** 

THE POWER TO KNOW

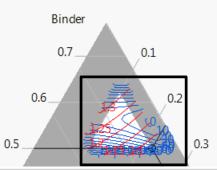


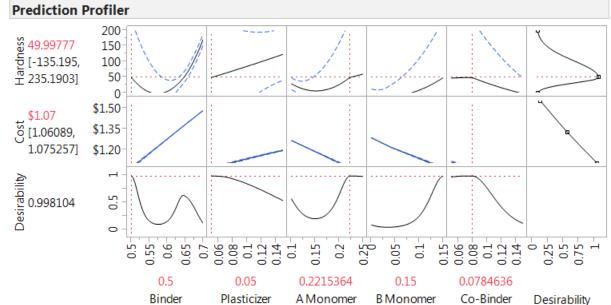
#### MODEL OPTIMIZATION SUGGESTS LOWER COST IS POSSIBLE

DOES THAT MAKE SENSE? DOES DATA SUPPORT IT? RUN CHECKPOINTS THERE.



Mixture Profiler			
T L R Factor	Current	X Lo Limit	Hi Limit
O O Binder	C	.5 0.5	0.7
🔿 💿 🔿 Plasticizer	0.0	0.05	0.15
<ul> <li>O          <ul> <li>A Monomer</li> <li>A Monomer</li> </ul> </li> </ul>	0.22153	54 0.1	0.25
○ ○ ○ B Monomer	0.1	15 0	0.15
○ ○ ○ Co-Binder	0.07846	36 0.05	0.15
Response	Contour Curre	ent Y Lo Limi	t Hi Limit
- Hardness	100 49.99	7766	
- Cost	1.325 1.068	0737	







0.2

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#### REAL-WORLD DESIGN ISSUES ADDRESSED BY CUSTOM DOE PLATFORM

*"Make Designs Fit the Problem – NOT Make Problems Fit the Designs!"* 

- Work with these different kinds of control variables/factors:
  - » Continuous/quantitative? (Finely adjustable like temperature, speed, force)
  - » Categorical/qualitative? (Comes in types, like material = rubber, polycarbonate, steel with mixed # of levels; 3 chemical agents, 4 decontaminants, 8 coupon materials...)
  - » Mixture/formulation? (Blend different amounts of *ingredients* and the process performance is dependent on the *proportions* more than on the amounts)
  - » Blocking? (e.g. "lots" of the same raw materials, multiple "same" machines, samples get processed in "groups" – like "eight in a tray," run tests over multiple days – i.e. variables for which there *shouldn't* be a causal effect
- Work with **combinations of these four kinds** of variables?
- Certain combinations cannot be run? (too costly, unsafe, breaks the process, subject matter experts say to avoid as "impractical.") Use constraints.
- Certain factors are hard-to-change (temperature takes a day to stabilize)
- Would like to add onto existing trials? (really expensive/time consuming to run)







#### MIXTURE VARIABLES

#### SIMPLE MIXTURE – MAKING SALAD DRESSING

- Relative *proportions* of factors or components is more important than actual quantity
- Three liquid components -Oil, Water, and Vinegar
- 8 oz. in Cruet vs. 4 gal. in Jug 5 oz. "O" 320 oz. 5/8 1 oz. "W" 64 oz. 1/8 2 oz. "V" 128 oz. 1/4
- To study these mixture components in a DOE use ranges that are proportions:
  - O: 0.500 to 0.750 (1/2 to 3/4)
  - W: 0.000 to 0.250 (0 to <sup>1</sup>/<sub>4</sub>)
  - V: 0.125 to 0.375 (1/8 to 3/8)
- Sum of proportions
   constrained to equal 1.

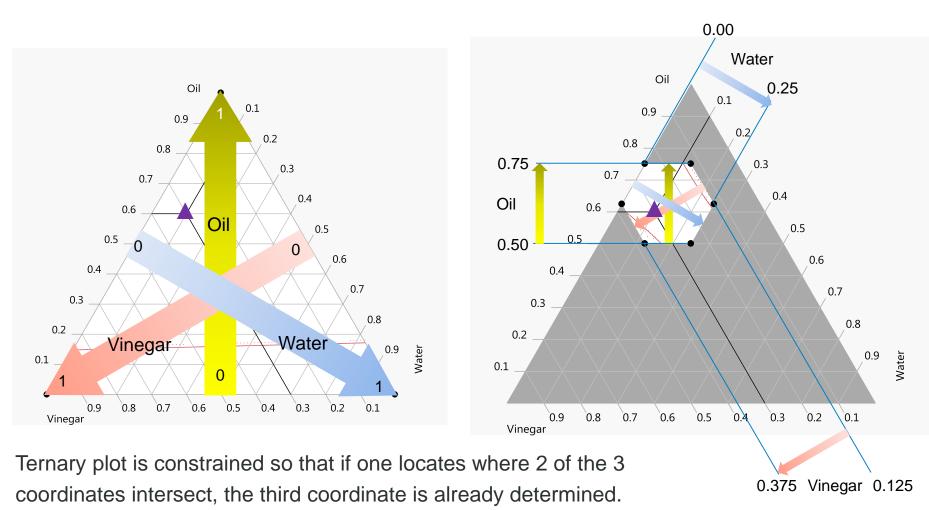
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1 = O + W + V so therefore... W = 1 - (O + V), O = 1 - (V + W), & V = 1 - (O + W)



# READINGINCREASE IN PROPORTION IS FROM BASE TO VERTEXREADINGLEFT:FULL RANGE: 0 TO 1TERNARY PLOTSRIGHT:EQUAL WIDTH RANGES: ± 0.125



If Oil = 0.6 and Vinegar = 0.3, then Water = 1 - (0.6 + 0.3) = 0.1 (See  $\blacktriangle$ )

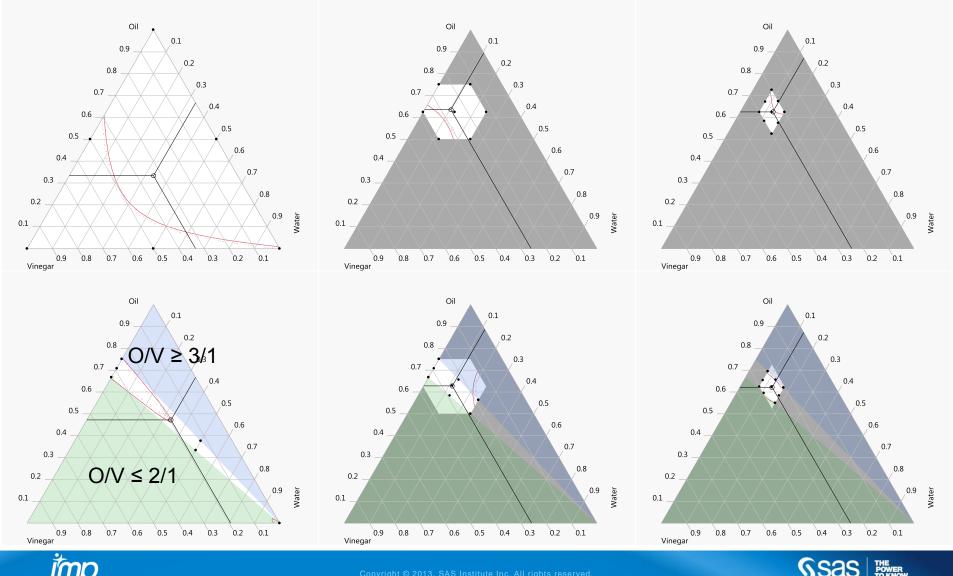


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#### SIX DESIGNS: Left: **TOP: NO CONSTRAINTS** Middle: **BOTTOM: 2 CONSTRAINTS** Right:

Full Range: 0 to 1

Equal width proportion:  $\pm 0.125$  about nominal Equal %change: ± 10% of nominal



#### INEQUALITY CONSTRAINT ALGEBRA

- 1. Express constraints as proportions
- 2. Clear fractions (note keeping unit multiplier)
- 3. Bring all factors to left side of inequality sign
- 4. Fill in boxes with coefficients and select  $\leq$  or  $\geq$

Oil/Vinegar ≤ 3/1	and	Oil/Vinegar ≥ $2/1$
1*Oil ≤ 3*Vinegar	and	1*Oil ≥ 2*Vinegar
1*Oil - 3*Vinegar ≤ 0	and	1*Oil - 2*Vinegar ≥ 0

DOE - Custom Design - JMP Pro										
Custom Design										
<sup>⊿</sup> Define Factor Constraints										
Add Constraint										
1 Oil + -3 Vinegar + 0 Wa	ter <b>≤</b> ▼ 0									
1 Oil + -2 Vinegar + 0 Wa	ter ≥ ▼ 0									

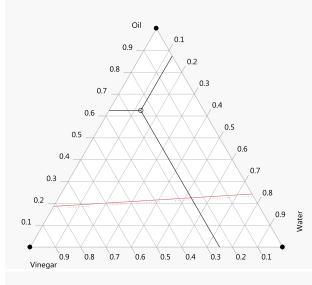
NOTE: Factors not in constraint get multiplied by zero

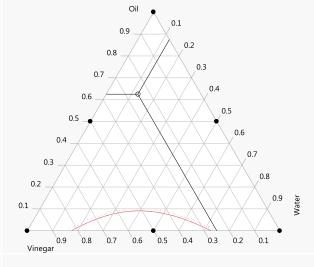


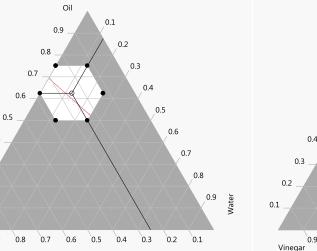


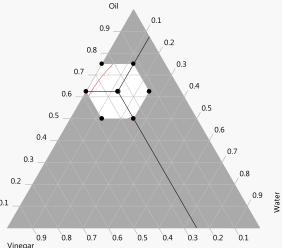
#### SIX DESIGNS: Left: TOP: 0 TO 1 RANGE BOTTOM: EQUAL WIDTH ± 0.125 ABOUT NOMINAL Right:

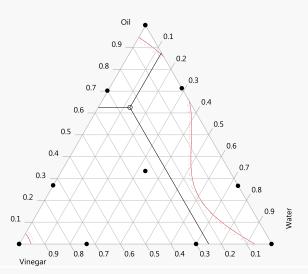
Main Effects Model – 1<sup>st</sup> order Interaction = Quadratic model! – 2<sup>nd</sup> Order Scheffé Special Cubic model – 3<sup>rd</sup> Order

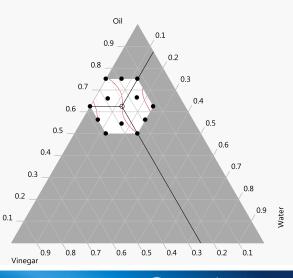














0.4

0.9

0.3

0.2

Vinegar

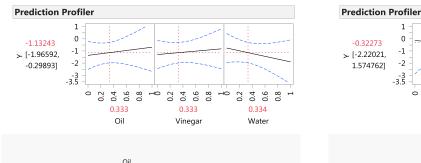
0.1

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#### MODEL COMPLEXIT

Main Effects Model – 1<sup>st</sup> order Left: Interaction = Quadratic model! – 2<sup>nd</sup> Order Middle: Right: Scheffé Special Cubic model – 3<sup>rd</sup> Order

**Prediction Profiler** 



0.1

0.2

0.3

0.5

0.3 0.2 0.1

0.4

0.6

07

0.8

0.9

Nater

0.9

0.8

0.7

0.6

0.8 0.7 0.6 0.5

0.5

0.4

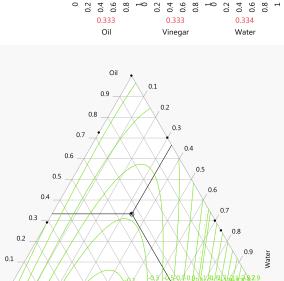
0.9

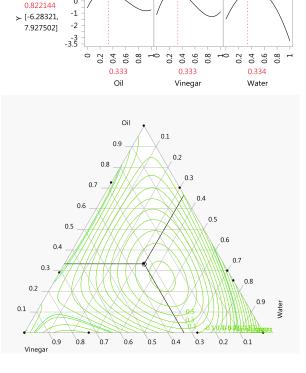
0.3

0.2

Vinegar

0.1





1<sup>st</sup> order for screening – finding the critical few

-0.32273

1.574762]

-1

-2

-3 -3.5

- 2<sup>nd</sup> order for prediction and optimization
- 3<sup>rd</sup> order when 2<sup>nd</sup> order proves inadequate for prediction (lack-of-fit) NOTE: For low numbers of components one might consider making a design to support a 3<sup>rd</sup> order model but analyze first with 2<sup>nd</sup> order model

0.7 0.6 0.5 0.4 0.3 0.2 0.1

0.9 0.8

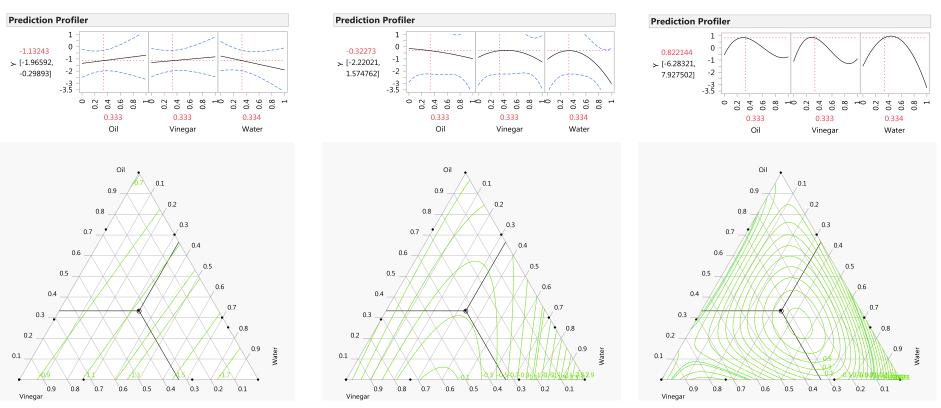
Vinegar





#### MODEL COMPLEXITY

Left:Linear Blending\* - 1st orderMiddle:Nonlinear Blending\* - 2nd OrderRight:Very Nonlinear Blending - 3rd Order



- Linear (additive) blending need only pure component response values
- Synergistic blending\* improvement in response exceeds additive prediction
- Antagonistic blending\* improvement in response is less than additive prediction

\*From Ron Snee's JMP Explorers Event on DOE Strategies for Accelerating Formulation Development



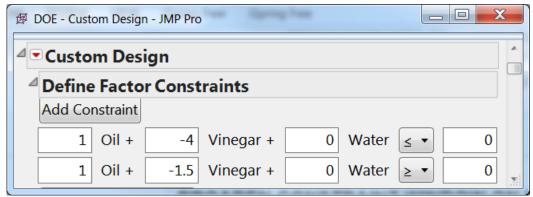


#### MAKE THIS DESIGN

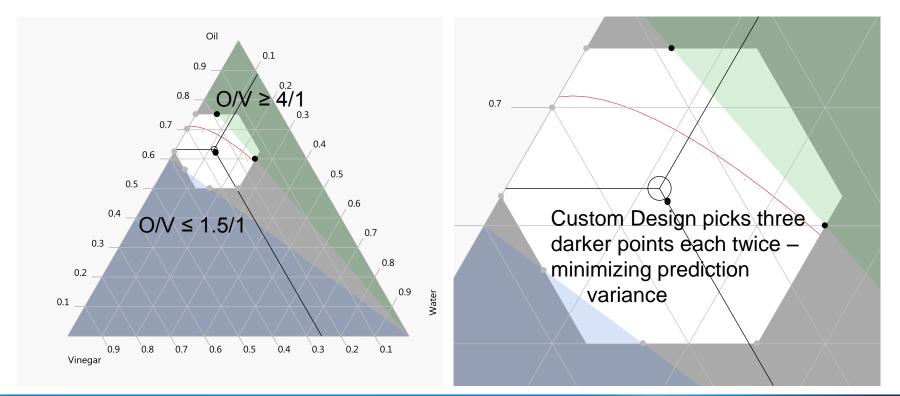
# BROADEN CONSTRAINT WINDOW ON RATIO OF OIL/VINEGAR FROM 2 $\leq$ 0/V $\leq$ 3 TO 1.5 $\leq$ 0/V $\leq$ 4

- O: 0.500 to 0.750 (<sup>1</sup>/<sub>2</sub> to <sup>3</sup>/<sub>4</sub>)
- V: 0.125 to 0.375 (<sup>1</sup>/<sub>8</sub> to <sup>3</sup>/<sub>8</sub>)
- W: 0.000 to 0.250 (0 to <sup>1</sup>/<sub>4</sub>)

Use a 2<sup>nd</sup> order model



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#### REVISIT PLASTIC FORMULATION

• 5 components - names & ranges

Binder	0.50	-	0.70
Plasticizer	0.05	-	0.15
A Monomer	0.10	-	0.25
B Monomer	0.00	-	0.15
Co-Binder	0.05	-	0.15

- 3 additional constraints
  - $0.18 \le A Mon + B Mon \le 0.26$
  - A\_Mon + B\_Mon + Plas  $\leq 0.35$
- model is 2<sup>nd</sup> order = nonlinear blending

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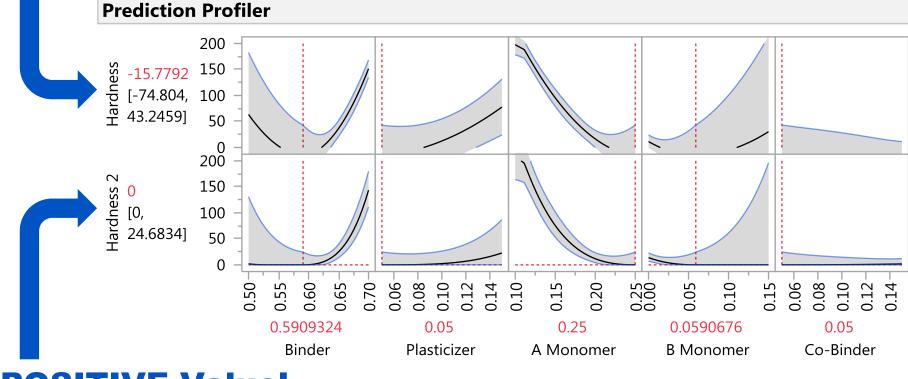
Define Factor Constraints										
◎ None										
Specify Linear Constraints										
O Use Disallowed Combinations Filter										
O Use Disallowed Combinations Script										
Linear Constraints										
Add										
0Binder +1Plasticizer +1A Monomer +1B Monomer +0Co-Binder $\leq$ 0.35										
0 Binder + 0 Plasticizer + 1 A Monomer + 1 B Monomer + 0 Co-Binder $\leq$ 0.26										
0Binder +0Plasticizer +1A Monomer +1B Monomer +0Co-Binder $\geq$ 0.18										
Remove Last Constraint										
Check Constraints										



#### POTENTIALLY EMBARRASSING PREDICTIONS

FITTING HARDNESS OF PLASTIC WITHOUT (TOP) AND WITH A SQRT TRANSFORMATION (BOTTOM)

## **NEGATIVE Value? NEGATIVE Low Limit?**



POSITIVE Value! ZERO Low Limit!

On Transformed Scale (Bottom), Predictions Make Physical Sense

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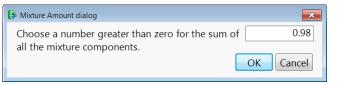


#### 10-FACTOR = 6-MIX, 2-CON, 1-CAT, 1-BLK COMPLEX DOE

 See step-by-step PDF for details of complex design construction

<b>#</b> [	DOE -	- Custom De	sign - JM	IP Pro													_ • <b>×</b>	
<u>F</u> ile	<u>E</u> c	dit <u>T</u> ables	<u>R</u> ows	<u>C</u> ols <u>I</u>	<u>D</u> OE	<u>A</u> nalyze	<u>G</u> raph	Six Sigma Too	ls T <u>o</u> ols	Add-I <u>n</u> s	View	<u>W</u> indow	<u>H</u> elp					
4	Cu	ustom D	esign	1													A	h
Þ	Re	sponses	5															
⊿	Fac	ctors																
Add Factor  Remove Add N Factors 1												E						
	N	Jame				Role		Char	iges	Values								
	<mark>∕⊿</mark> ₿	Base				Mixtu	ire	Easy		0.4				0.55				
				Mixtu	ire	Easy		0.2				0.4						
				Mixtu	ire	Easy		0.01				0.03						
	A-polymer			Mixtu	ire	Easy		0			0.3							
	<sup>A</sup> B-Polymer				Mixtu	ire	Easy		0			0.3						
	C-Polymer				Mixtu	ire	Easy		0			0.3						
	Cure Time			Conti	nuous	Easy		15			45							
	Temperature				Conti	nuous	Easy	140				160						
	۲ <mark>۸</mark>	Aixer				Cated	orical	Easy		Brand-	A			Brand	-B			
	<b>-</b> C	Day				Block	ing	Easy		1				2				
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7



Portion of mixture held

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constant at 2%



#### 7-COMPONENT AND 7-CONSTRAINTS MIXTURE DOE

🖉 DOE - Custom Design - JMP Pro

- Eile Edit Tables Rows Cols DOE Analyze Graph Six Sigma Tools Tools Add-Ins View Window <u>H</u>elp Custom Design Responses Factors Add Factor 
  Remove Add N Factors 1 Name Role Changes Values **A**1 Mixture Easy 0 0.36 **A**A2 0 0.36 Mixture Easy **4**A3 0.36 0 Mixture Easy <sup>4</sup>B1 0.24 0 Mixture Easv <sup>4</sup>B2 0.24 Mixture 0 Easy  $\mathbf{A}_{C1}$ 0 1 Mixture Easy <sup>4</sup>C2 Mixture 0 1 Easy Define Factor Constraints None Specify Linear Constraints Use Disallowed Combinations Filter Use Disallowed Combinations Script Linear Constraints Add 1 A1 + 1 A2 + 1 A3 + 0 B1 + 0 B2 + 0 C1 + 0 C2 0.36 ≤ ▼ 1 A2 + 1 A1 + 1 A3 + 0 B1 + 0 B2 + 0 C1 + 0 C2 0.2 ≥ ▼ 0 A1 + 0 A2 + 0 A3 + 1 B1 + 1 B2 + 0 C1 + 0 C2 0.24 ≤ 0 A1 + 0 A2 + 0 A3 + 1 B1 + 1 B2 + 0 C1 + 0 0.1 C2 ≥ 0 A2 + 0 0 A3 + 1 B1 + -3 B2 + 0 C1 + 0 A1 + C2 0 ≤ ▼ -1 A2 + -1 A3 + 0.7 B1 + 0.7 B2 + -1 A1 + 0 C1 + 0 C2 0 ≤ ▼ 1 A2 + 1 A3 + -1.3 B1 + -1.3 B2 + 1 A1 + 0 C1 + 0 C2 0 ≤ ▼ Remove Last Constraint Check Constraints 🏠 🔳 🔻
- See step-by-step PDF for details of mixture design construction

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#### THREE-LAYER FILM STRUCTURE, FACTORS AND RANGES

Factor choice and ranges come from you and/or your subject matter experts!

Total *Thickness* of Three-Layer Film is 24 to 48 microns

*Layer A* is 25% to 55% of Total Thickness

*Layer B* is 30% to 70% of Total Thickness

Layer C is 5% to 15% of Total Thickness

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Layer A = R1 Layer B = R2 + R1 Layer C = R2 + R3

Layer A is 100% Resin 1

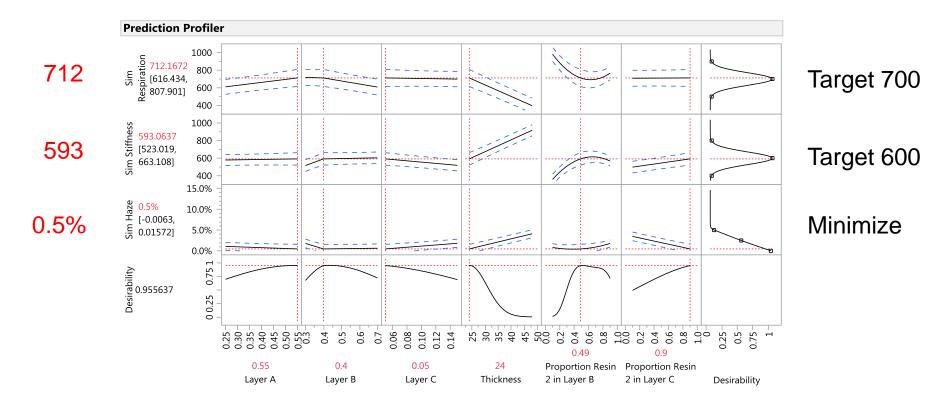
Layer B is 10% to 90% *Resin 2* and 90% to 10% Resin 1

Layer C is 10% to 90% *Resin 2* and 90% to 10% Resin 3



#### WWW.JMP.COM/TECHNICALLY-SPEAKING

#### **GO TO** "BOOSTING PERFORMANCE WITH CUSTOM DESIGNED EXPERIMENTS"

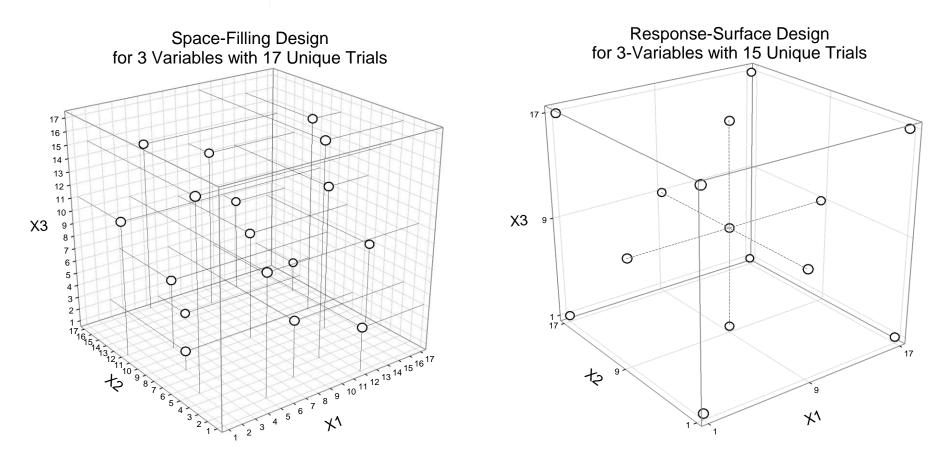


Ask JMP to find the best trade-off in performance among multiple responses for multiple factors





# FROM EFFICIENTHOW ARE SPACE-FILLING DESIGNS DIFFERENT FROMM&S TUTORIALTRADITIONAL RESPONSE-SURFACE DESIGNS?



Rather than emphasizing high leverage trials ("corners") for a simple polynomial model, space-filling designs "spread" their trials more uniformly through the space to better capture the local complexities of the simulation model.





#### **US ARMY EXAMPLE** SPACE-FILLING MIXTURE DESIGNS

In order to create a space filling design with mixture variables, we have two options-, we can use the mixture design platform with the space filling design option, or we can use the space filling design platform with the appropriate constraints to enforce the sum of the individual ingredient proportions equal to one. As Figure 2 visually demonstrates with an unconstrained three component mixture design, we have nearly equivalent space filling efficiency with either option.

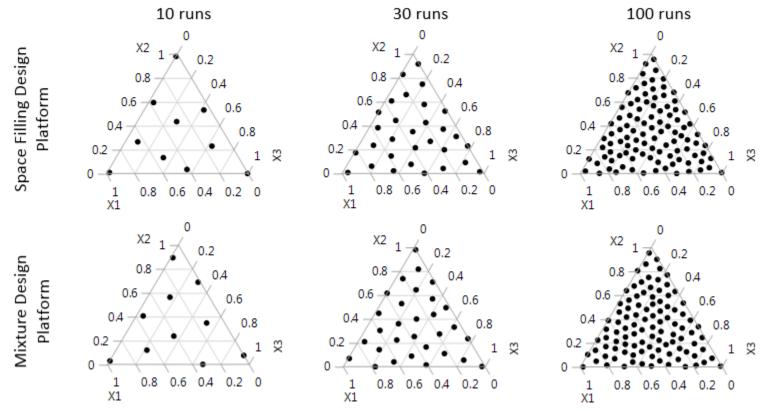


Figure 2: Ternary Plots of 10, 30 and 100-run Space Filling Mixture Designs using the Space Filling Design Platform and the Mixture Design Platform

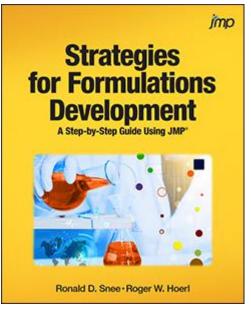




#### RESOURCES

#### LINKS TO WEBCASTS, DOWNLOAD PDFS, AND BOOK

- https://www.jmp.com/en\_us/events/ondemand/mastering-jmp/mixture-designs.html
- <u>https://www.jmp.com/en\_us/events/ondemand/technically-speaking/boosting-performance-with-</u> custom-designed-experiments.html
- https://www.jmp.com/en\_us/events/ondemand/mastering-jmp/transforming-data.html
- <u>https://community.jmp.com/t5/Discovery-Summit-2017/An-Uncertainty-Quantification-Case-Study-Using-Space-Filling/ta-p/44055</u>
- <u>https://community.jmp.com/t5/US-Federal-Government-JMP-Users/Step-by-Step-JMP-DOE-Examples/ta-p/22176</u>



https://www.sas.com/store/prodBK\_68410\_en.html?storeCode=SAS\_US









# Thanks. Questions or comments?

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