

Application of Consumer Ideal Point Mapping to a 3-Factor Experimental Design and its Graphical Representation

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Preference Mapping

- Group of multivariate statistical techniques
 - Develop a deeper understanding of consumer liking for products
 - Category appraisal
 - Internal, external
 - Why preference mapping?
 - Import sensory attributes: Drivers of liking
 - Assist product developers with optimizing sensory properties
 - What is the ideal product like?
-

Context of Preference Mapping

Consumer panel
assesses the products for
liking



Set of
competitive
products



Trained panelists
describe the products
in sensory terms



Hedonic
Scores

	C1	C2	...	Cn-1	Cn
P1					
P2					
⋮					
Pk					

Statistical modeling

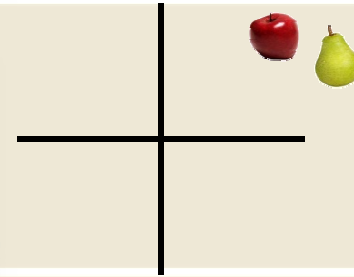
K=number of products
T=number of sensory attributes
N=number of consumers

Sensory
Profiles

	S1	S2	...	St-1	St
P1					
P2					
⋮					
Pk					

Concepts

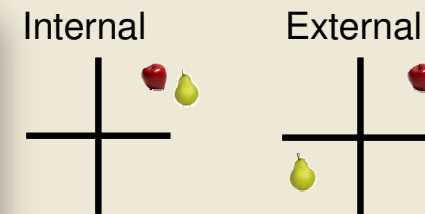
Mapping methods yield a graphical representation of consumer preference and/or sensory differences for a set of products



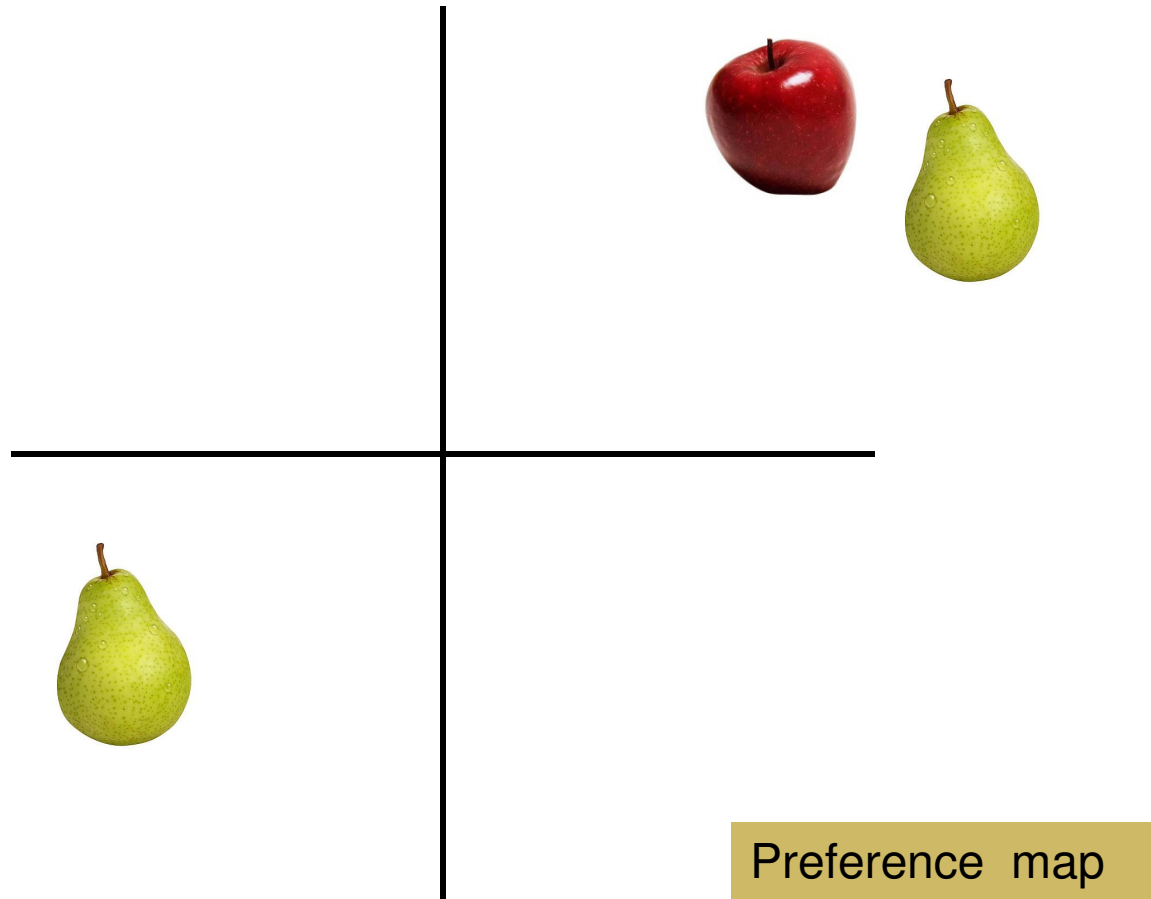
Consumers evaluate 6 or more products

- Some competitor products
- Some potential prototypes

External versus Internal Preference Mapping



Mapping perceptions or preferences?



Internal vs. External

Internal preference analysis

- Stimulus location based on liking (hedonic data drives orientation of the map)
- Sensory attributes can be fitted into preference space afterwards
- First dimension explains maximum variability in hedonic directions

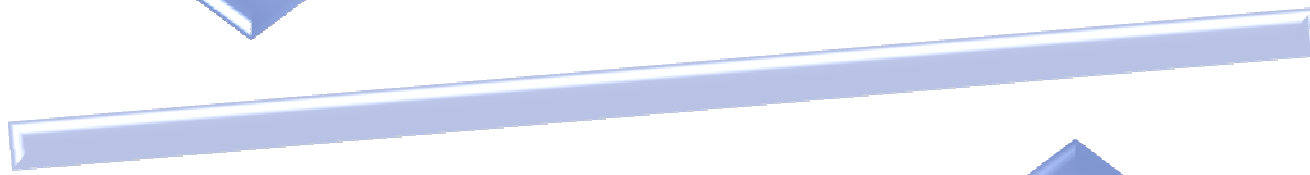
External preference analysis

- Stimulus locations based on similarity in sensory properties (sensory data drive orientation of the map)
- Preference data can be fitted into fixed space afterwards
- First Dimension explains maximum variance in sensory attribute descriptions

Objectives



Optimize the formulation of strawberry yogurt on viscosity, sweetness and strawberry sensory dimensions

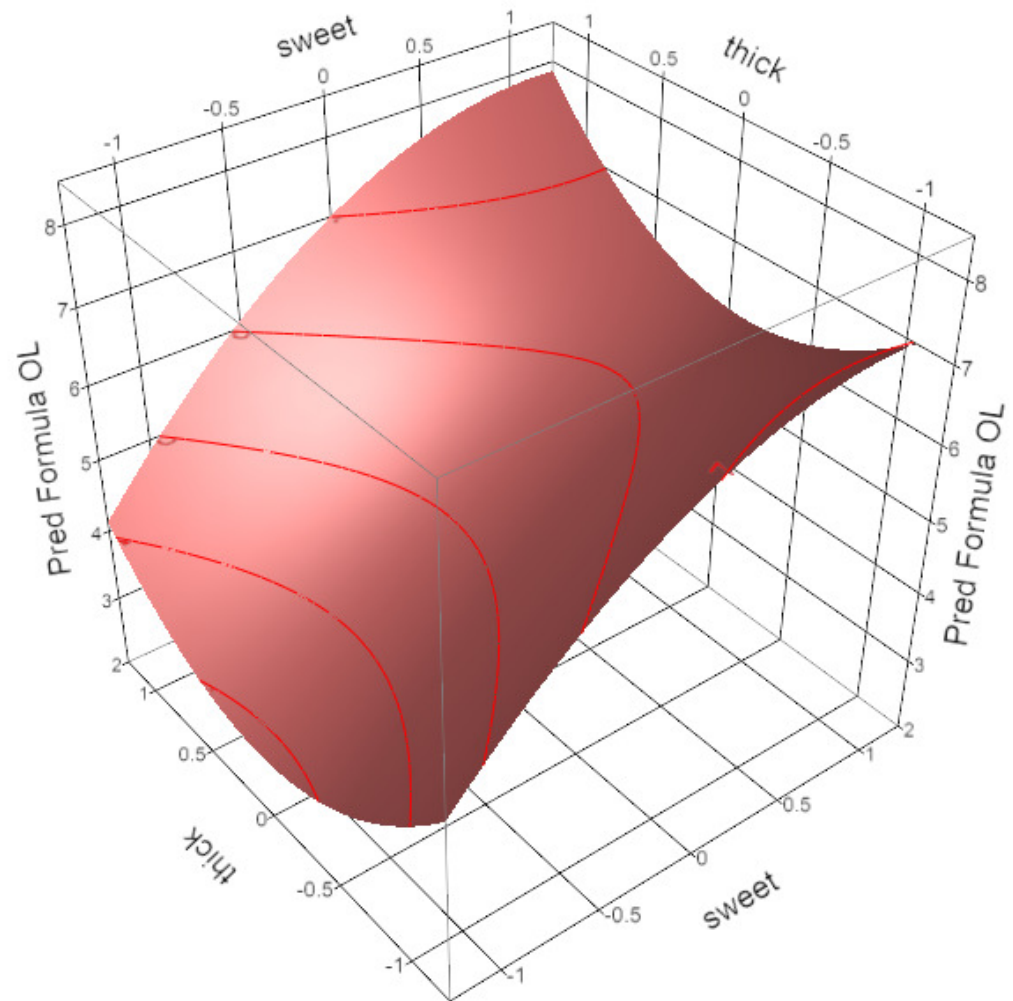


Comparison of DOE Ideal Point Mapping to more conventional methods



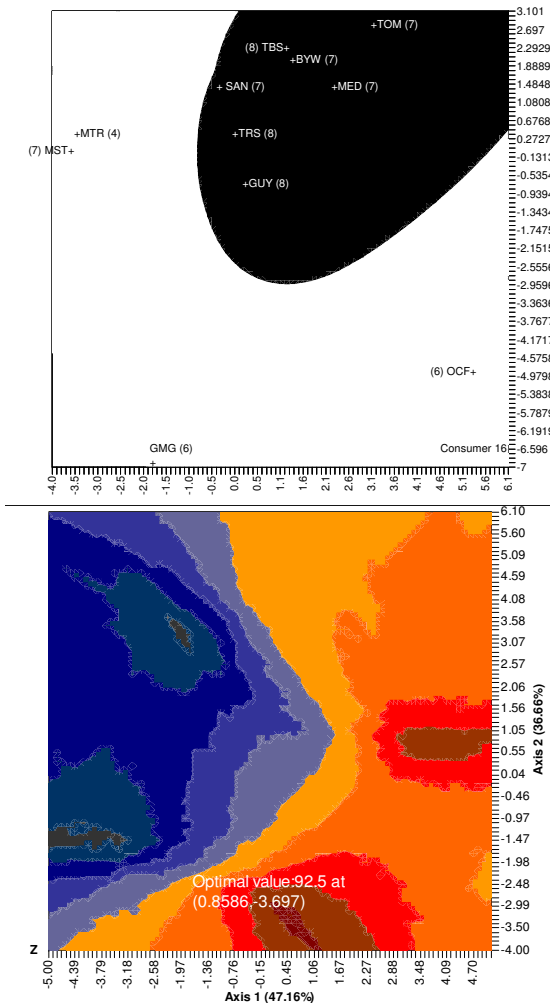
Modeling liking (DOE)

- Liking data fitted to DOE using RSM models (quadratic)
- Hedonic data averaged across consumers
 - Loss of information from averaging
 - Overfitting and number of treatments
 - 2 factors, 6df
 - 3 factors, 11df



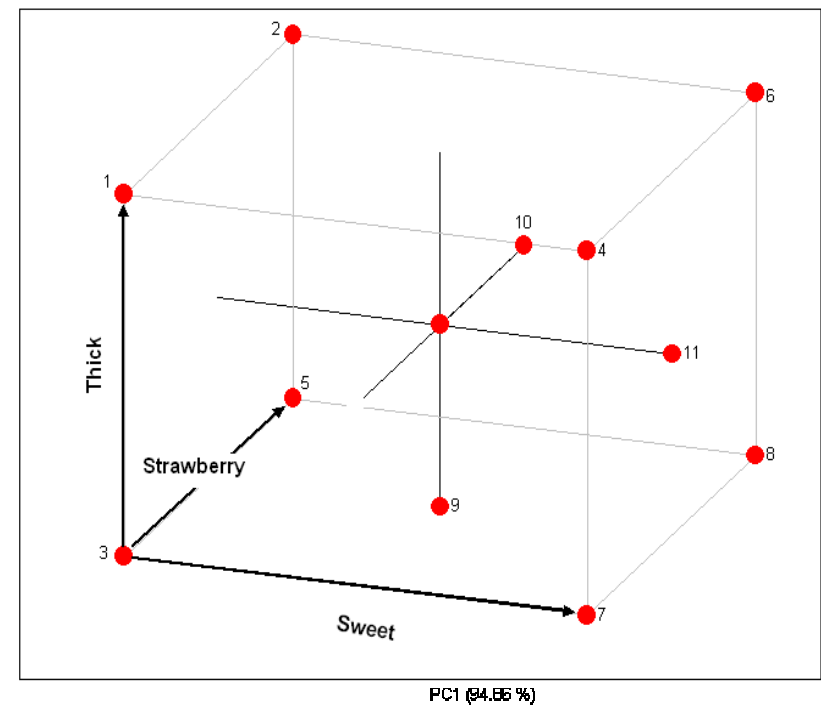
Euclidean Distance Ideal Point Mapping

- EDIPM, an enhancement to internal preference mapping (MDPREF vector models) to identify ideal points (Meullenet et al., 2007)
 - Individual ideals identified
 - Density of individual ideals
 - Group Ideals
 - Projection of sensory attributes to determine ideal profiles



EDIPM

- Starting point: a multidimensional representation of products in a space
- Derived from
 - consumer liking (OL) data (internal framework)
 - sensory profiles (external framework)
 - DOE



Ideal Point Mapping

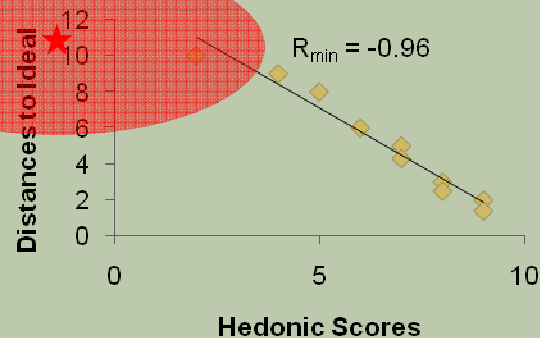
Point of maximum density in the sensory space taken as group ideal

If correlation not different from the minimum correlation this second location is added to the acceptable region of the sensory space for a consumer

Group Ideal

Individual consumer Ideal

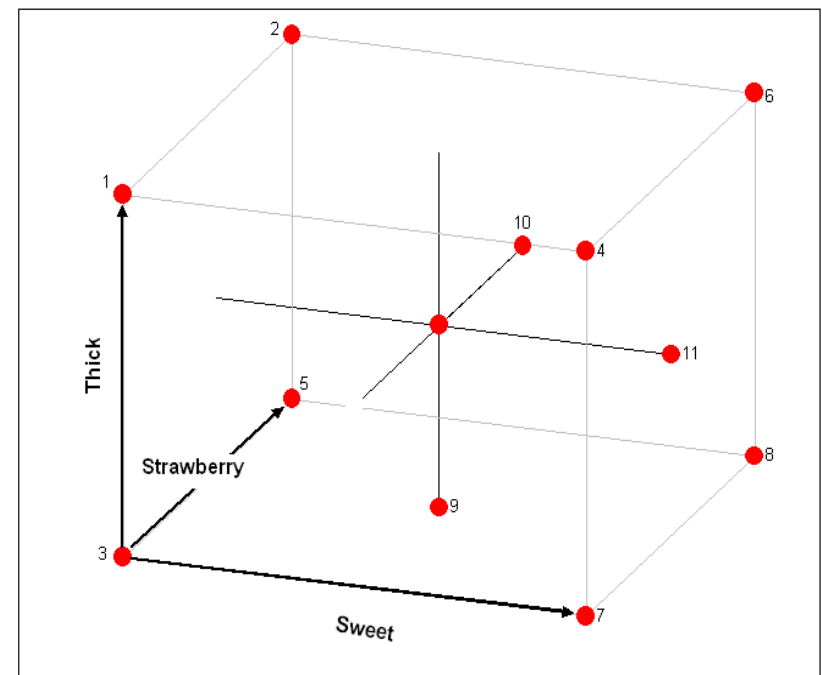
Significance of the correlation can be assessed (defined by $\alpha_1=0.05$)



Methods

Experimental Design

- 3 variables and 3 levels
 - Thickness, Sweetness, and Strawberry Flavor
 - High (1), medium (0), and low (-1)
- Design-Expert® 7.1, Stat-Ease, Inc.
 - D-optimal, 3 factor design with 11 treatments



Method of Production



Milk + SMP +
gelatin

- heated at 85°C
- 5 minutes



Cooled in ice
bath

- 40°C

Inoculation

- plain yogurt



Incubation in oven

- 5.5 hours
- 37°C to 40°C



Addition of:

- Strawberry flavor
- Sugar Syrup
- Strawberries
- Food coloring

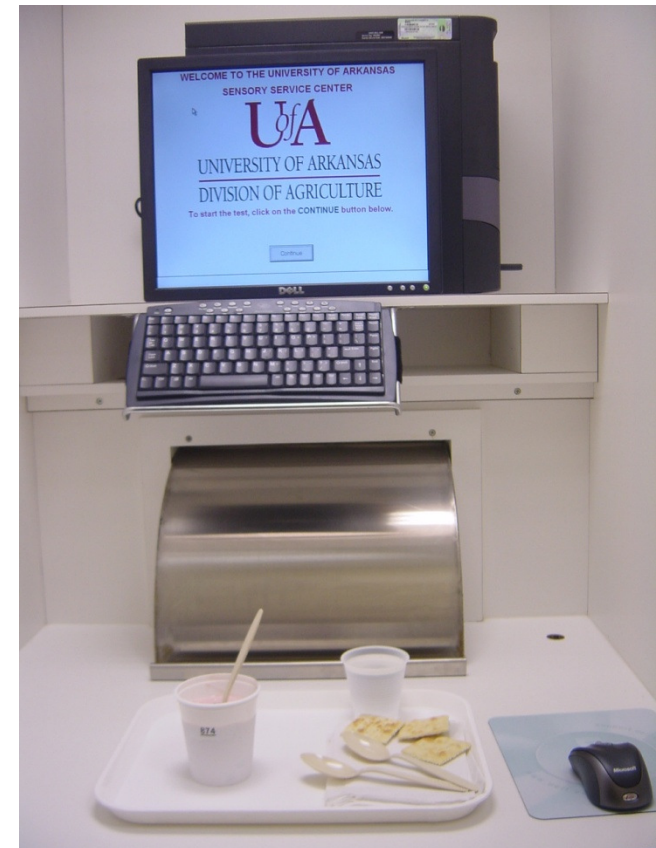
8oz plastic cups

Descriptive Analysis Methods

- Spectrum Method®
trained panelists
 - Texture and Flavor Evaluation
 - 0 to 15 intensity scale
 - Universal Scale for flavor
 - Texture References
 - Based on previous work of panel
 - Current commercial products
 - Visual Texture:
 - spoon impression, clumpy, thickness, and smooth
 - Oral Texture:
 - thickness, stickiness, chalky, and dairy film
 - Basic Tastes:
sweet, salt, sour, bitter
 - Aqueous solutions as references for 0 to 15 scale
 - Aromatics:
 - overall strawberry impression, musty/overripe, caramelized/cooked, green/unripe, vanillin, cultured dairy, butter fat, milky, and other.
-

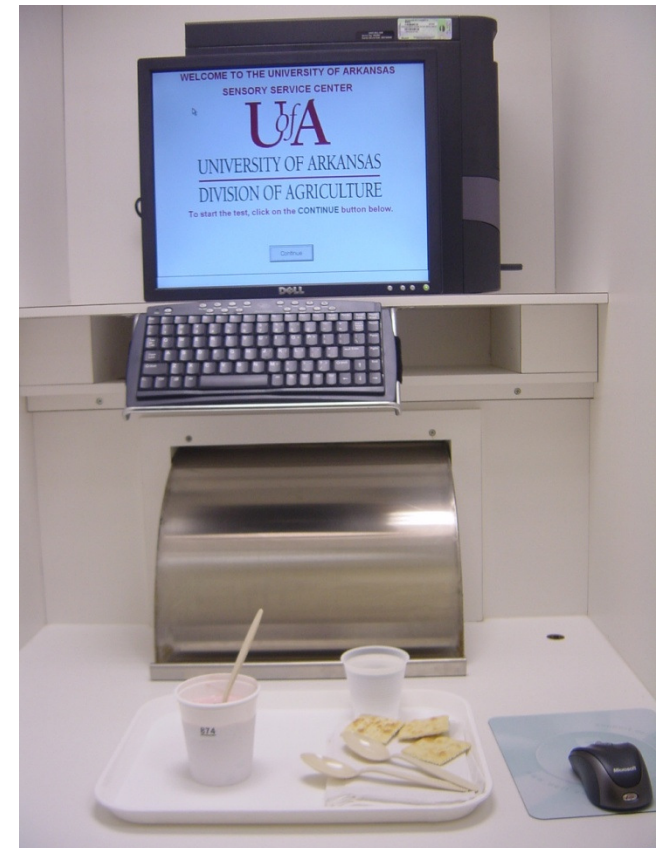
Consumer Testing

- Email recruiting from UofA Sensory Laboratory database (N=2500)
- 120 self-reported strawberry yogurt consumers (70% female, 30% male)
- Testing over 2 days: 11 samples
 - Balanced randomization across both days
 - 6 samples on day one, 5 samples on day 2



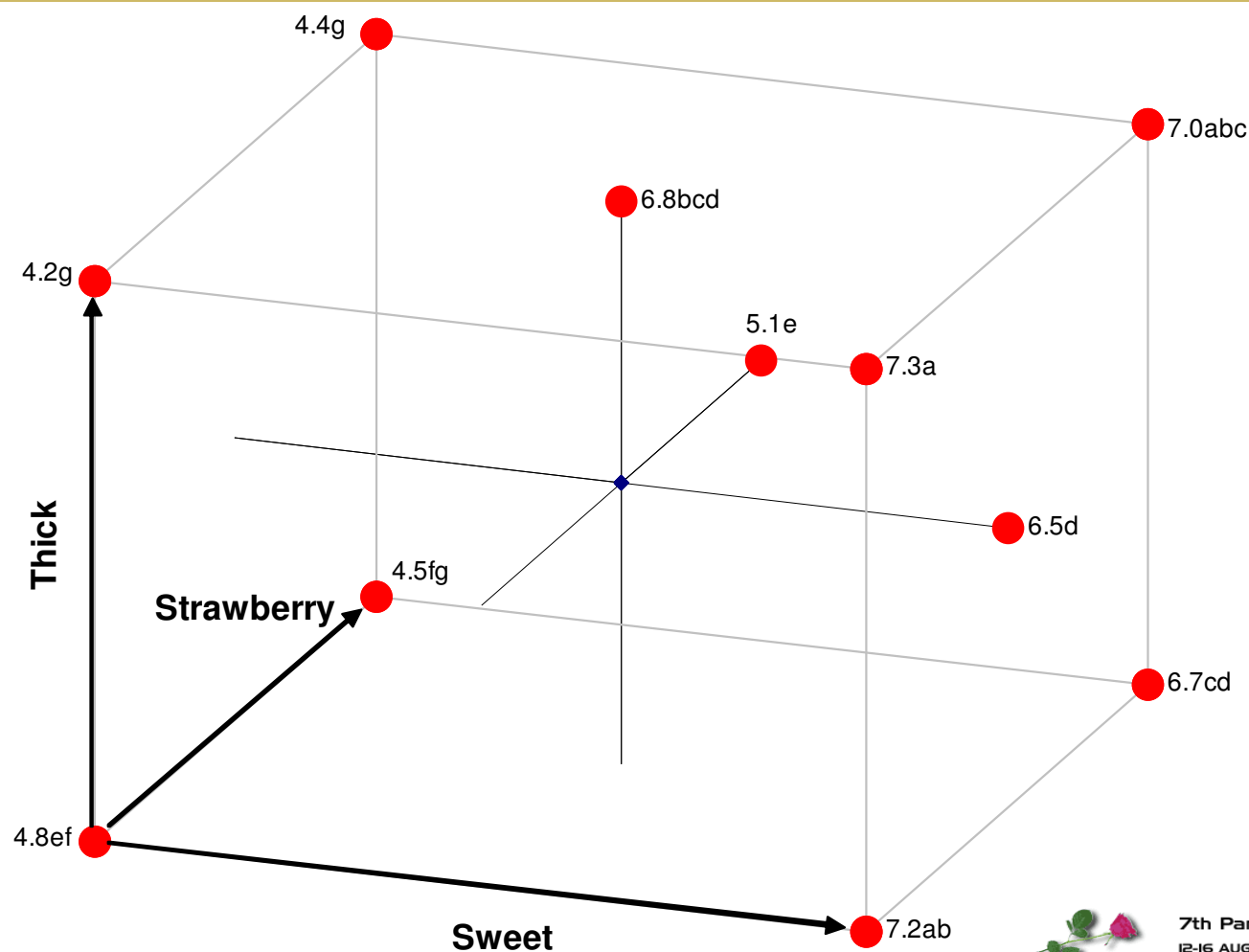
Consumer Testing

- 9-pt hedonic scale:
 - 1= dislike extremely, 5=neither like nor dislike, 9= like extremely
 - Overall impression, Appearance, Flavor, Texture
- Just-About-Right scales:
 - 1=not nearly sweet enough, 3= just about right, 5= much too sweet
 - Overall flavor, Sourness, Strawberry flavor, Thickness, Creaminess, Smoothness, Sweetness, Amount of fruit



Results

Overall Liking Means



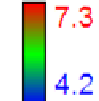
Response Surface Methodology

Fitting a quadratic model on mean liking data results in multiple optima
Consumers liked either thick or thin

Design-Expert® Software

Overall

● Design Points



X1 = A: Sweet

X2 = B: Thick

Actual Factor

C: Strawberry = -1.00

Factors Tool

Gauges Sheet

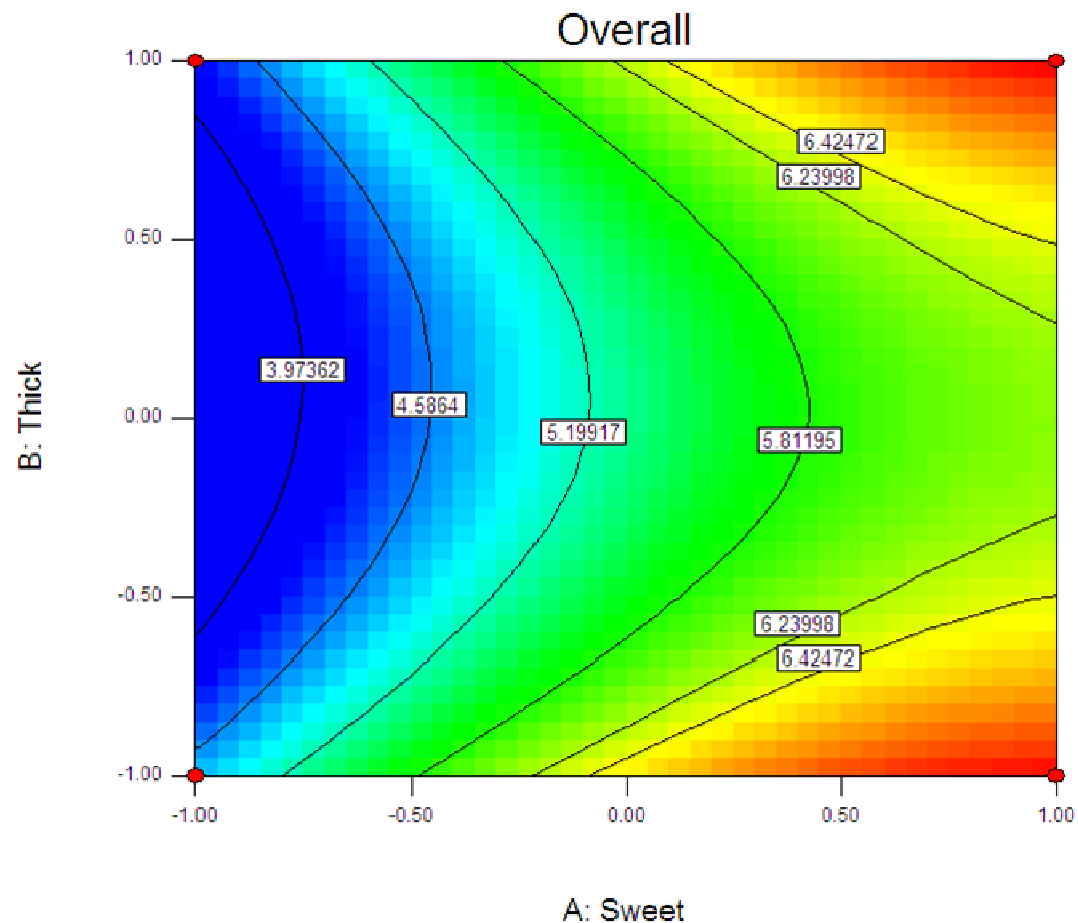
Default

X1 A: Sweet

X2 B: Thick

C: Strawberry -1.00

Term AB



RSM

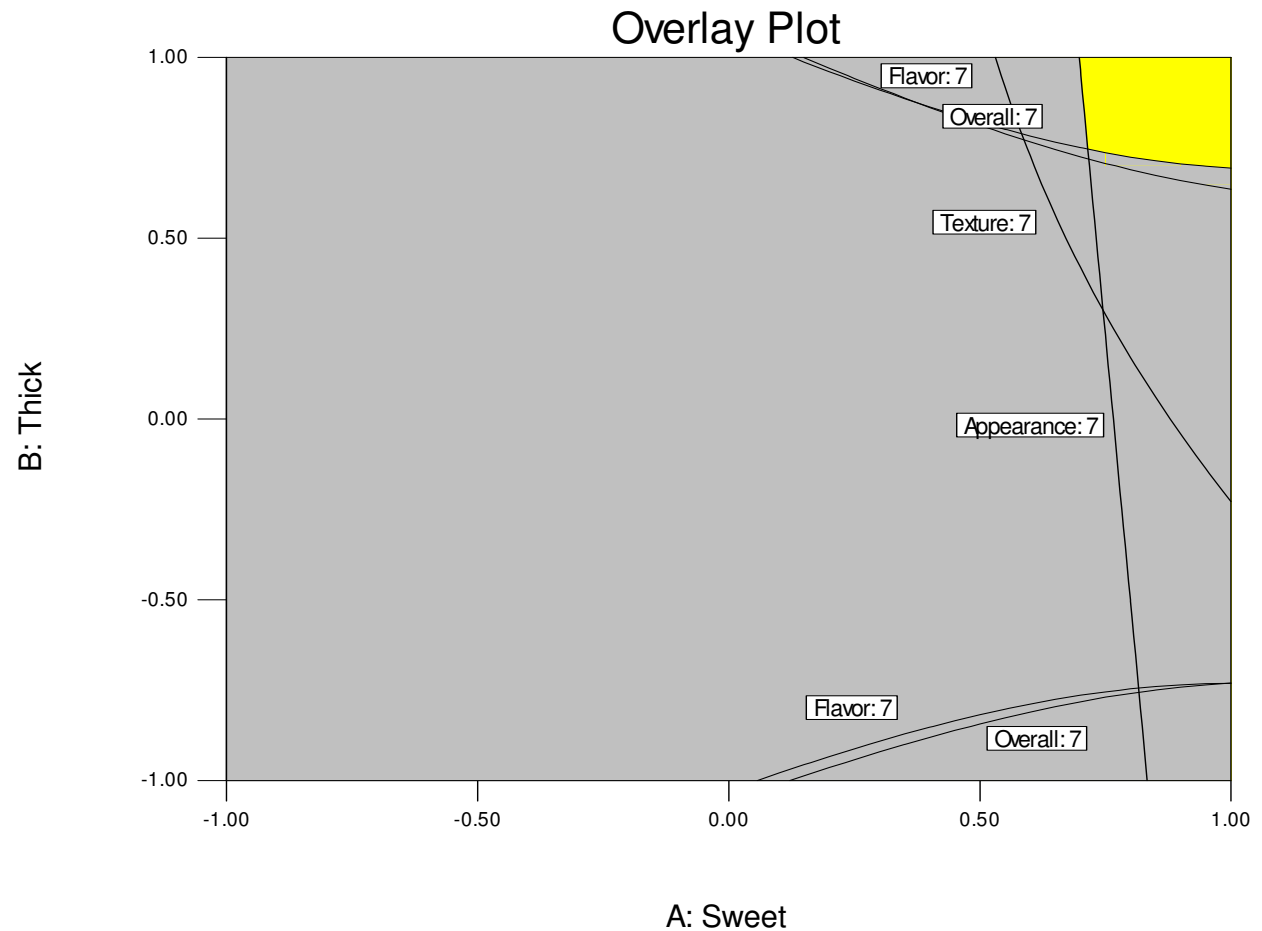
RSM

Considering multiple hedonic responses

- Overall Impression
- Appearance
- Flavor
- Texture

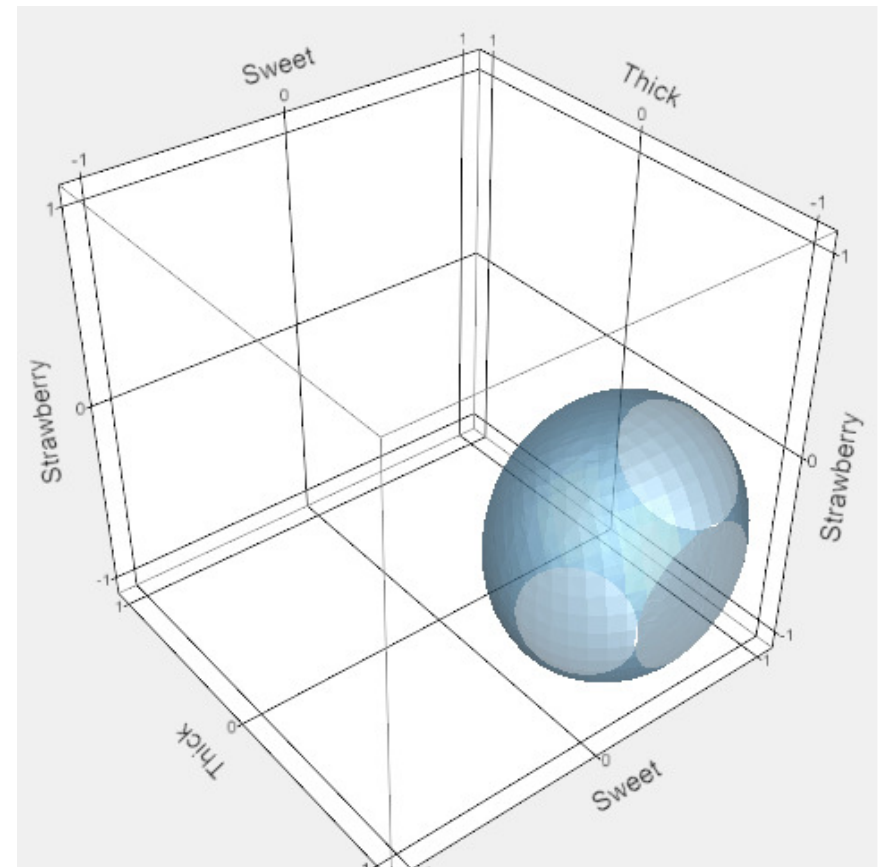
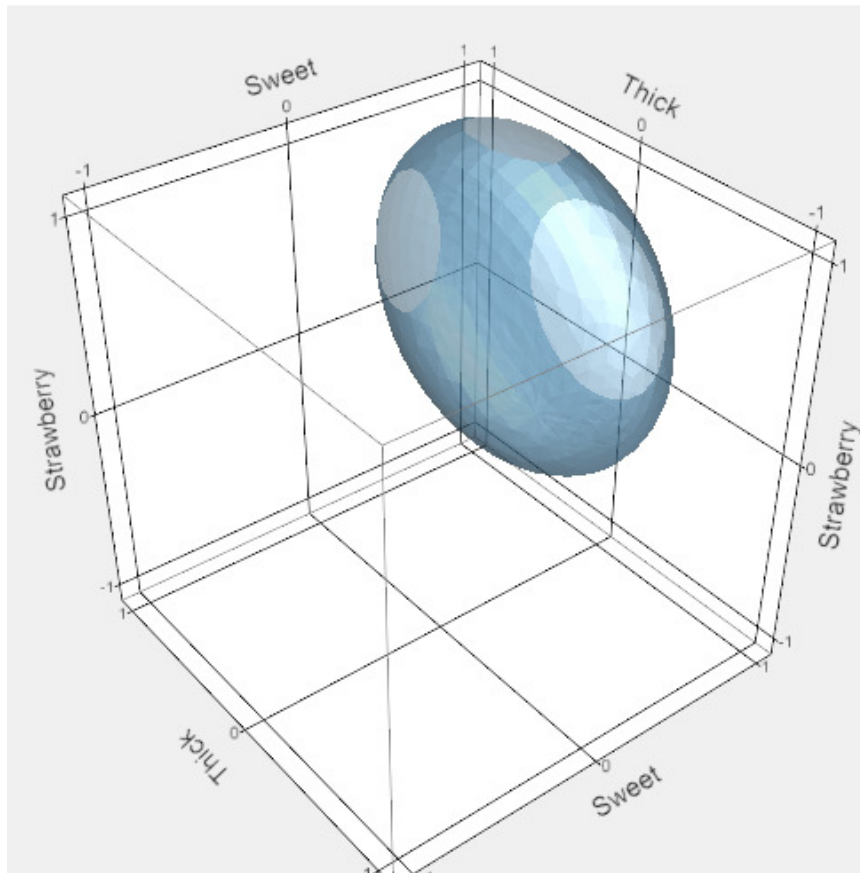
Acceptable formulation

- Thick > 0.75
- Sweet > 0.80

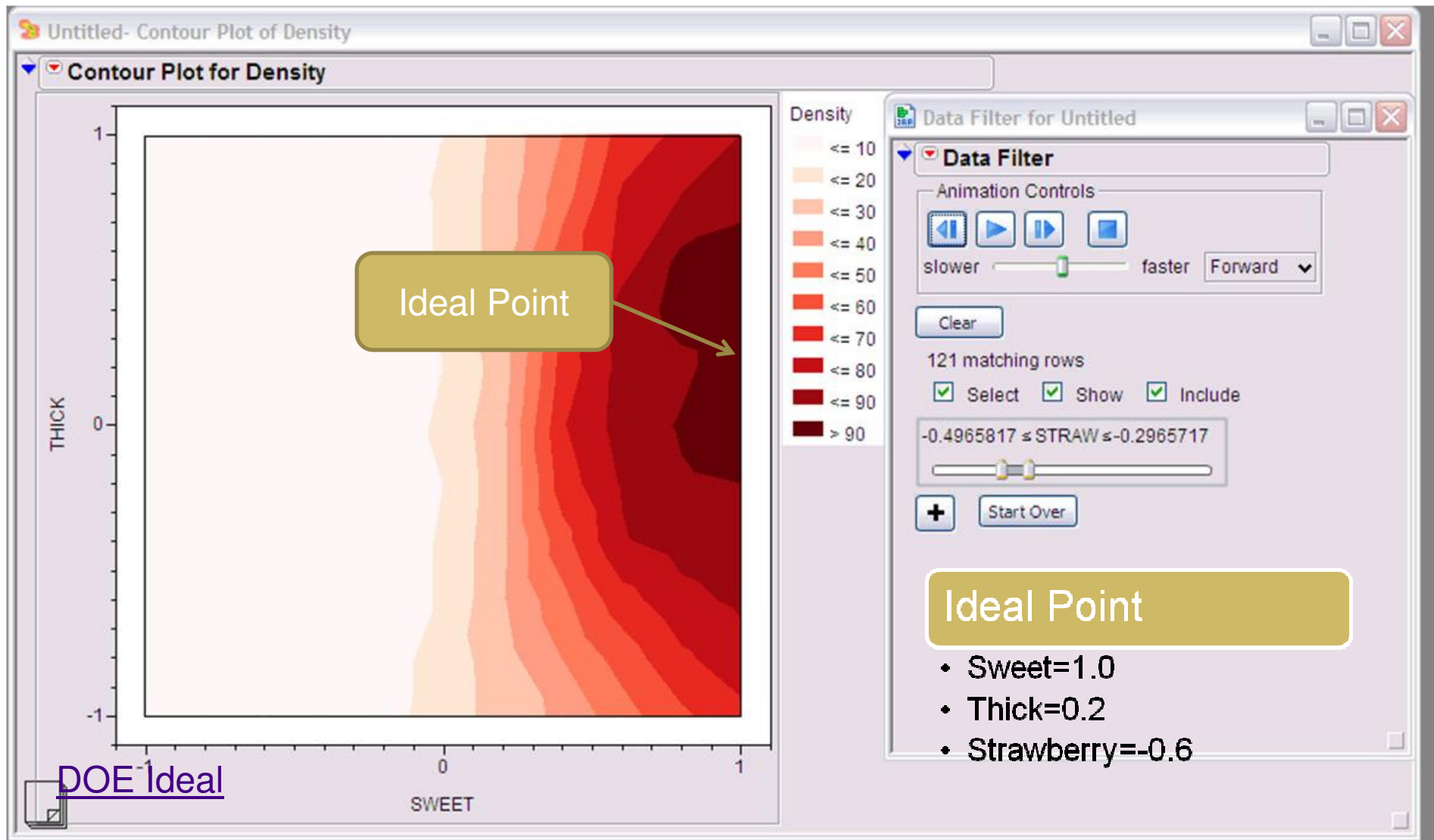


Individual Consumers

Ideal DOE: Ellipses represent acceptable areas for two different consumers



DOE Ideal Point Mapping



Optimal Formulations

DOE EDIPM is different from preference mapping solutions especially for thickness, the second most important factor

Method	Thick	Sweet	Strawberry
LSA	0.45	0.58	1.09
JAR	0.52	1.04	-1.1
EXT	1.06	0.81	0.17
EDIPM	0.88	1.1	-1.08
DOE EDIPM	0.20	1.00	-0.60

1=high

0=medium

-1=low



7th Pangborn Sensory Science Symposium
12-16 AUGUST 2007, HYATT REGENCY, MINNEAPOLIS, USA

Consumer Fit

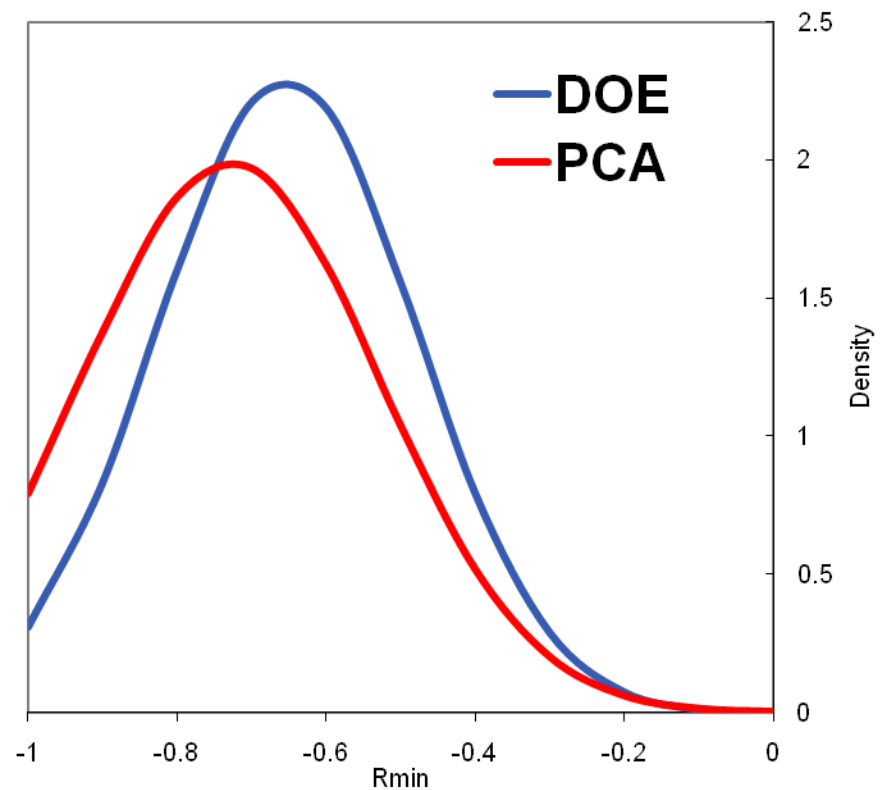
Is the hedonic data (individual consumers) well fitted in the DOE?
How does it compare to Internal Preference mapping?

Distribution of R_{\min} values for consumers in:

- Internal map space
- DOE space

Consumers slightly better fitted in internal map space

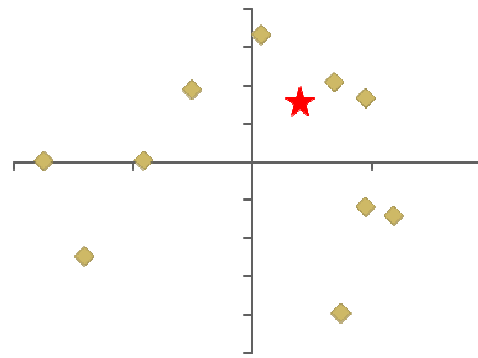
Why use DOE Ideal Point Mapping?



Internal mapping with DOE

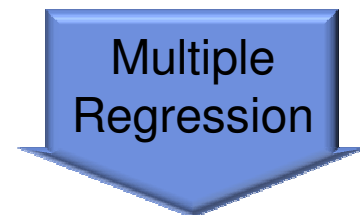
Many modeling steps are necessary when internal preference mapping is used with a DOE...multiple errors

Hedonic Scores					
	C1	C2	...	Cn-1	Cn
P1					
P2					
⋮					
⋮					
⋮					
Pk					



Information loss

	PC1	PC2
P1		
P2		
⋮		
⋮		
⋮		
Pk		
Ideal	PC1 _{Ideal}	PC2 _{Ideal}

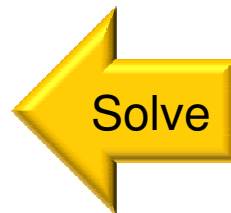


PC1=f(sweet, thick, strawberry)
PC2=f(sweet, thick, strawberry)
⋮
S1=f(sweet, thick, strawberry)
S2=f(sweet, thick, strawberry)
⋮
Sp=f(sweet, thick, strawberry)

Information loss



PC1 _{Ideal}
PC2 _{Ideal}
⋮
S1 _{Ideal}
S2 _{Ideal}
⋮
Sp _{Ideal}



Ideal
• Sweet
• Thick
• Strawberry

DOE Ideal Points

DOE Ideal Point Modeling is a more direct way to identify ideal
Retains consumer individuality

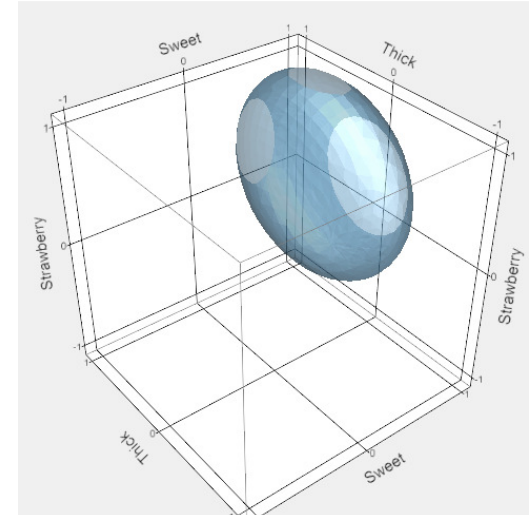
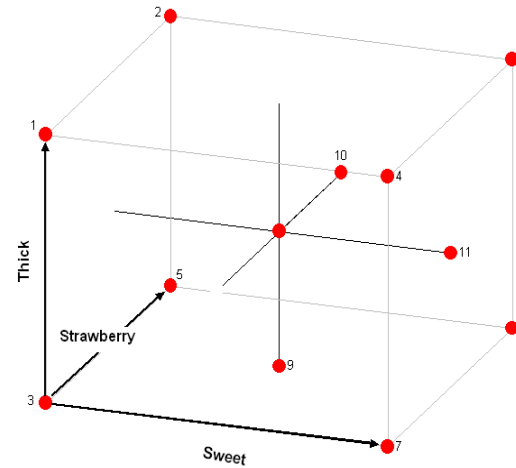
Hedonic
Scores

	C1	C2	...	Cn-1	Cn
P1					
P2					
⋮					
Pk					

Averaging

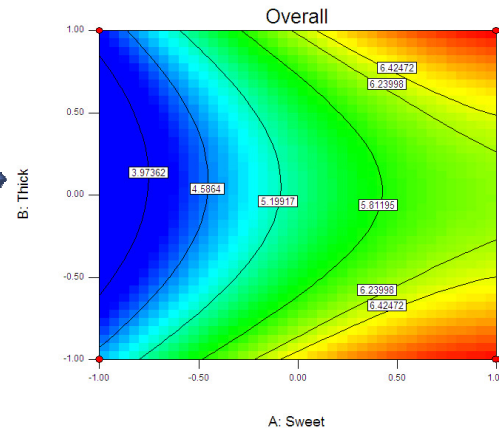
	Liking avg
P1	
P2	
⋮	
Pk	

Ideal
Points



Max
Density

RSM



Max
Predicted

Ideal

- Sweet
- Thick
- Strawberry

Conclusions

Ideal solutions for Ideal Point DOE and RSM are different

- Preserving consumer individuality seems more sensible!

Ideal Point DOE also yielded different answers than internal or external preference mapping

- Internal or external preference mapping applied to a DOE results in cumbersome modeling
- Prediction errors for various modeling steps are cumulated

IPM not regression based

- no overfitting
 - not limited to 3 factors
-



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DIVISION OF AGRICULTURE

<http://www.uark.edu/ua/sensory>

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