



World Class. Face to Face.

### Correlation Between Sensory Analysis and Volatile Composition of Beer using Multivariate Analysis: Effect of the Beer Matrix on the Sensory Perception and Volatile Fraction Behavior

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### **Beer Volatiles**



# **Flavor Perception**

• Volatiles have to be released from the beer

• Not uniformly released

- Release dependent on:
  - » Concentration
  - » Interactions with non-volatile ingredients

# **Previous Studies**

• Interactions among volatile aroma compounds and the non-volatile matrix influence flavor perception

• Relationships between chemical and sensory data would help understand how interactions affect flavor perception

# **OBJECTIVES**

• Study the effect of non volatile levels on volatile fraction behavior and sensory perception

• Hypothesis: Interaction between volatile and non-volatile fractions will impact partitioning and sensory perception of the beer

# **EXPERIMENTAL APPROACH**



# **Trained Panel**



• 8 hour training sessions

• 15 cm-line scales

• Formal evaluations

• Aroma and Flavor perception



• Beer Flavor Solution

• Non-Volatiles

• Isomerized Hop Acid

#### **Beer Flavor Solution**

- Isoamyl acetate (Banana)
- Ethyl hexanoate (Apple)
- Benzaldehyde (Almond)
- Myrcene (Dry-Hop)

- **Non-Volatiles** 
  - Glucose
  - Fructose
- Maltodextrin
- Protein Extract





• 3 levels of CHO (Low, Medium, High)

• 3 levels of Pro (Low, Medium, High)

• 9 combinations of CHO/Pro

# **Sensory Results**

	<b>Protein Level</b>			Carbohydrate Level		
Attribute	L	Μ	Η	L	Μ	Η
Apple aroma	3.7 <sup>a</sup>	4.8 <sup>b</sup>	5.0 <sup>b</sup>	4.2 <sup>a</sup>	4.7 <sup>a</sup>	4.5 <sup>a</sup>
Banana aroma	2.4 <sup>a</sup>	3.4 <sup>b</sup>	3.4 <sup>b</sup>	3.1 <sup>a</sup>	3.1 <sup>a</sup>	3.0 <sup>a</sup>
Almond aroma	2.5 <sup>a</sup>	2.9 <sup>a</sup>	2.8 <sup>a</sup>	2.8 <sup>a</sup>	2.6 <sup>a</sup>	2.9 <sup>a</sup>
Dry-Hop aroma	3.1 <sup>a</sup>	2.9 <sup>a</sup>	2.9 <sup>a</sup>	3.4 <sup>a</sup>	3.2ª	2.5 <sup>b</sup>
Apple flavor	3.3 <sup>a</sup>	<b>4.</b> 1 <sup>a</sup>	<b>3.</b> 8 <sup>a</sup>	3.4 <sup>a</sup>	4.0 <sup>a</sup>	3.9 <sup>a</sup>
Banana flavor	2.1 <sup>a</sup>	2.6 <sup>a</sup>	2.6 <sup>a</sup>	2.2ª	2.4 <sup>a</sup>	2.6 <sup>a</sup>
Almond flavor	3.3 <sup>a</sup>	3.2 <sup>a</sup>	2.9 <sup>a</sup>	2.9 <sup>a</sup>	3.1 <sup>a</sup>	3.4ª
Dry-Hop flavor	3.1 <sup>a</sup>	2.9 <sup>a</sup>	2.7 <sup>a</sup>	3.4 <sup>a</sup>	2.8 <sup>ab</sup>	2.6 <sup>b</sup>

# **Sensory Aroma Results**

• Unexpected increase in aroma intensity with increasing levels of protein

• Proteins bind volatile flavor components



### **Aroma Results**

- Presence of protein-protein interactions
- Retention varies depending on volatile compound
- Reciprocal aroma suppression

# **Reciprocal Aroma Suppression**

• An odorant decreases the perceived intensity of others

• Suppression effect was reduced due to binding

• Processing of mixtures by the brain is not fully understood

### **Aroma Results**

• Dry-hop reduction with increase in carbohydrate concentration

• Increase in solution hydrophobicity

• Myrcene hydrophobicity

# **Flavor Results**

• No trend in the results

• No effect except for dry-hop flavor

• Components in saliva could affect partitioning

• Individual panelist differences

### **Instrumental Analysis**



### SPDESBSESPME

#### Concentration (mg/L) of Compounds at Different Protein Concentrations using SPME



#### **Concentration (mg/L) of Compounds at Different CHO Concentrations using SPME**



# Proteins

- Proteins bind flavor compounds strongly
- Hydrophobic bonding
- High affinity of carbonyls to proteins
- Expected to find lower concentrations of compounds in headspace

# Carbohydrates

- Main effect due to modification of solutions viscosity
- Moderate effect due to molecular interaction
- Both retention and salting out effects have been observed

## **PCA Aroma Data**



### PCA SBSE Data







# **Partial Least Squares Regression**

• Used for predicting sensory data from instrumental data

• Creates linear models to relate Y to X

• Creates models to predict Y from X



PLS models to predict aroma sensory data for (A) banana, (B) almond, (C) dryhop and (D) apple from instrumental data obtained by SBSE-GCFID analysis

#### **R<sup>2</sup> and Q<sup>2</sup> values of the models to predict aroma** sensory descriptive data

	SBSE		SPME		SPDE	
Attribute	<b>R</b> <sup>2</sup>	Q <sup>2</sup>	<b>R</b> <sup>2</sup>	$\mathbf{Q}^2$	<b>R</b> <sup>2</sup>	<b>Q</b> <sup>2</sup>
Banana	0.583	0.162	0.209	0.042	0.595	0.195
Almond	0.351	-0.076	0.180	-1.036	0.558	-0.044
Dry-Hop	0.755	-0.279	0.074	-0.540	0.963	-0.218
Banana	0.775	0.205	0.141	0.075	0.850	0.267

### **Observations**











# CONCLUSIONS

- Non-volatile fraction of model beer solutions influenced the volatile fraction behavior and sensory perception of the beverage
- Proteins played a bigger role in sensory perception than carbohydrates
- Results differed between the instrumental and the sensory results

# CONCLUSIONS

• The PLS results indicated a weak correlation between the sensory and the instrumental data

• The results question the validity of directly relating instrumental data to sensory evaluation

# CONCLUSIONS

- PCA showed clear differentiation among model beer samples mainly driven by the protein concentration
- Chemical volatile analysis can provide valuable information about volatile composition but it is not able to provide a complete flavor profile for beer

### THANK YOU!!!!!

