

New Approach to Discovery in Food Flavor



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The goal

- To chemically characterize the flavor of a food i.e. the chemical stimulants in a food delivered to the sensory systems that are responsible for flavor perception.
 - Standardization
 - Trouble shooting flavor issues – stability, quality
 - Increasing flavor intensity
 - Understanding pathways/chemistry of flavor development
 - Packaging issues

The journey



- Modern flavor chemistry is 50+ years old
- Rather young in comparison to some other fields of study

Evolution of knowledge (academic)

- 1960 - 70s – focused on the identification of all volatile compounds in foods
 - Hypothesis – if we can identify all of the aroma stimuli, we can reproduce the flavor of a food through reformulation
 - Flawed
 - Some compounds not available or not approved
 - Need quantitative data as well as qualitative
 - Wasn't that "simple" (e.g. interactions/delivery)

Concurrently – Studies on the Maillard reaction



- Identification of volatiles
- Characterization of reaction systems (process flavors)
- Identification of pathways for volatile formation
 - Could not reproduce a “process” flavor through reformulation based on volatiles

1970 → 80s Quantitative data



- Obtained some quantitative data
 - Still did not permit reconstruction of the food flavor

1970s – 1980s Statistical Relationships

- ❑ Compared gas chromatographic profiles of different food products
 - Pepsi Vs Coke
 - Wine from the north Vs south of Italy
- ❑ Could differentiate products but Did not have enough data to understand what was giving flavor

1970 → 1990s (present) Flavor interactions

- Strong initial focus on proteins but also work done on lipids and carbohydrates
 - Attempting to explain why different foods taste different when adding the same flavor
- Obtained a qualitative appreciation for the issue but not a quantitative
 - Little help!

1980 → present – Identification of “key” aroma compounds



- Initially – Werner Grosch and then Peter Schieberle’s groups (Germany) – global effort
- Developed methods to select key compounds
 - Initially considered defining – after much work, been able to convince them this is a screening method (at best)

Select possible “Key” aroma compounds

(Semmelroch et. al., 1995).

	Activity Value	Mechanism
(E)-p-Damascenone	2.7×10^5	Carotene degrad.
2-Furfurylthiol	1.7×10^5	
3-Mercapto-3-methylbutylformate	3.7×10^4	
5-Ethyl-4-hydroxy-2-methyl-3(2H)-furanone	1.5×10^4	Maillard reaction
Guaicol	1.7×10^3	Phenol degrad.
4-Vinylguaicol	1.0×10^3	Phenol degrad.
Methional	1.2×10^3	Maillard reaction
2 3-Diethyl-5-methylpyrazine	95	

Add the selected compounds to an unflavored base (deodorized base).

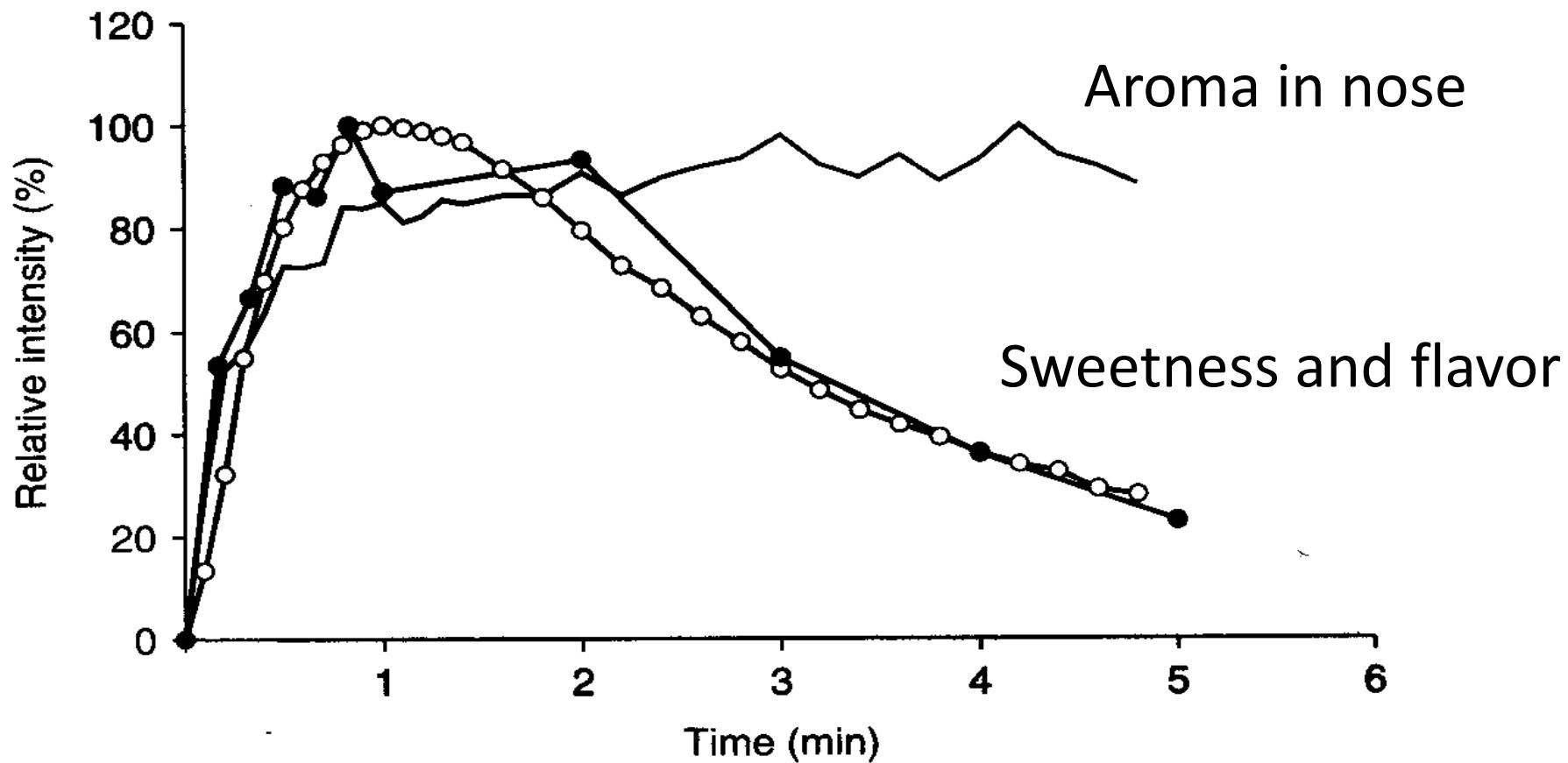
- Subject the sample to sensory analysis to validate selection and optimize quantities
 - FAILED!
- How to? Select 30 compounds and put them into model system to optimize the mixture – HOW?
- 30 variables at perhaps 3 concentrations

Mid 1990s – Aroma release

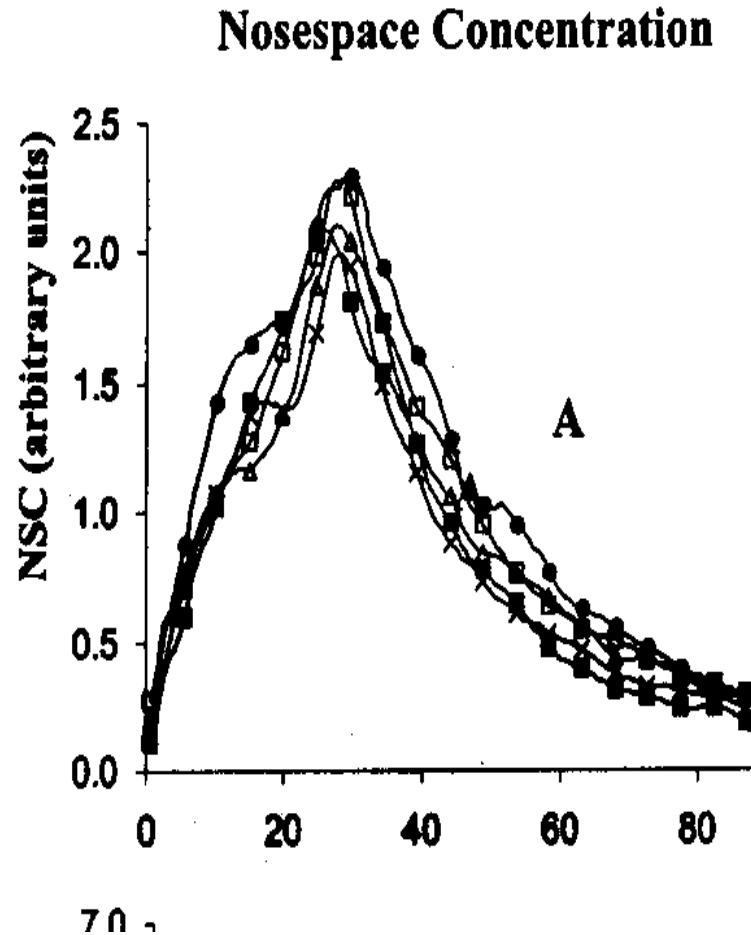
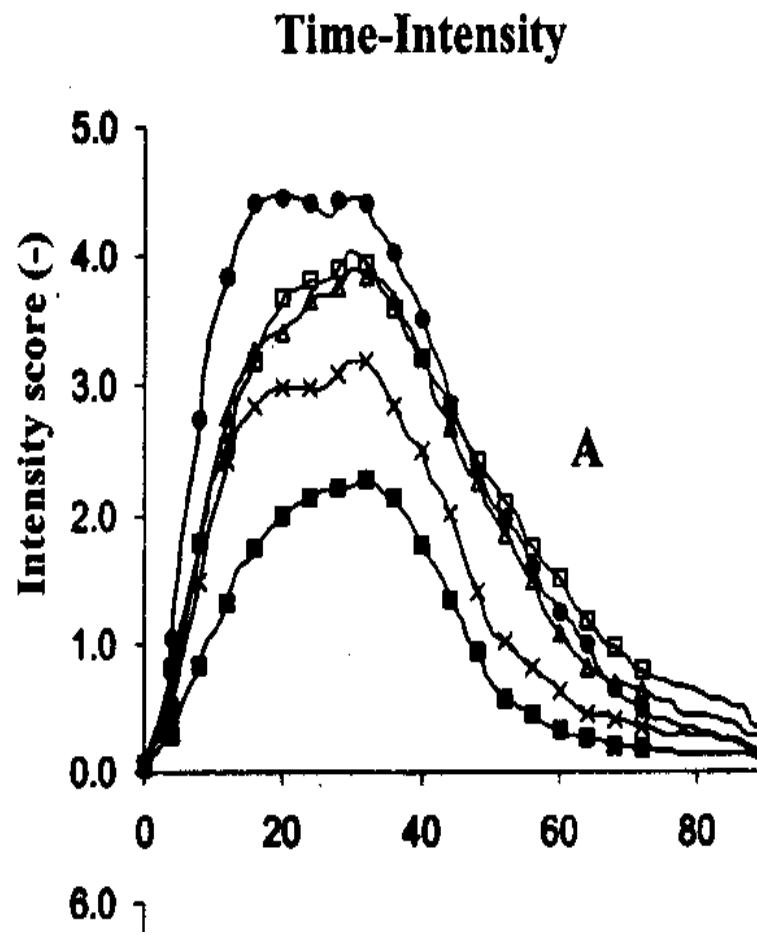


- ❑ Andrew Taylor et al. - in-vivo; many others developed artificial mouths
- ❑ API MS of volatiles in one's breath on eating a food; later included taste
- ❑ Learnings
 - Studies on how aroma release is linked to food composition and structure
 - A start on how stimuli relates to perception

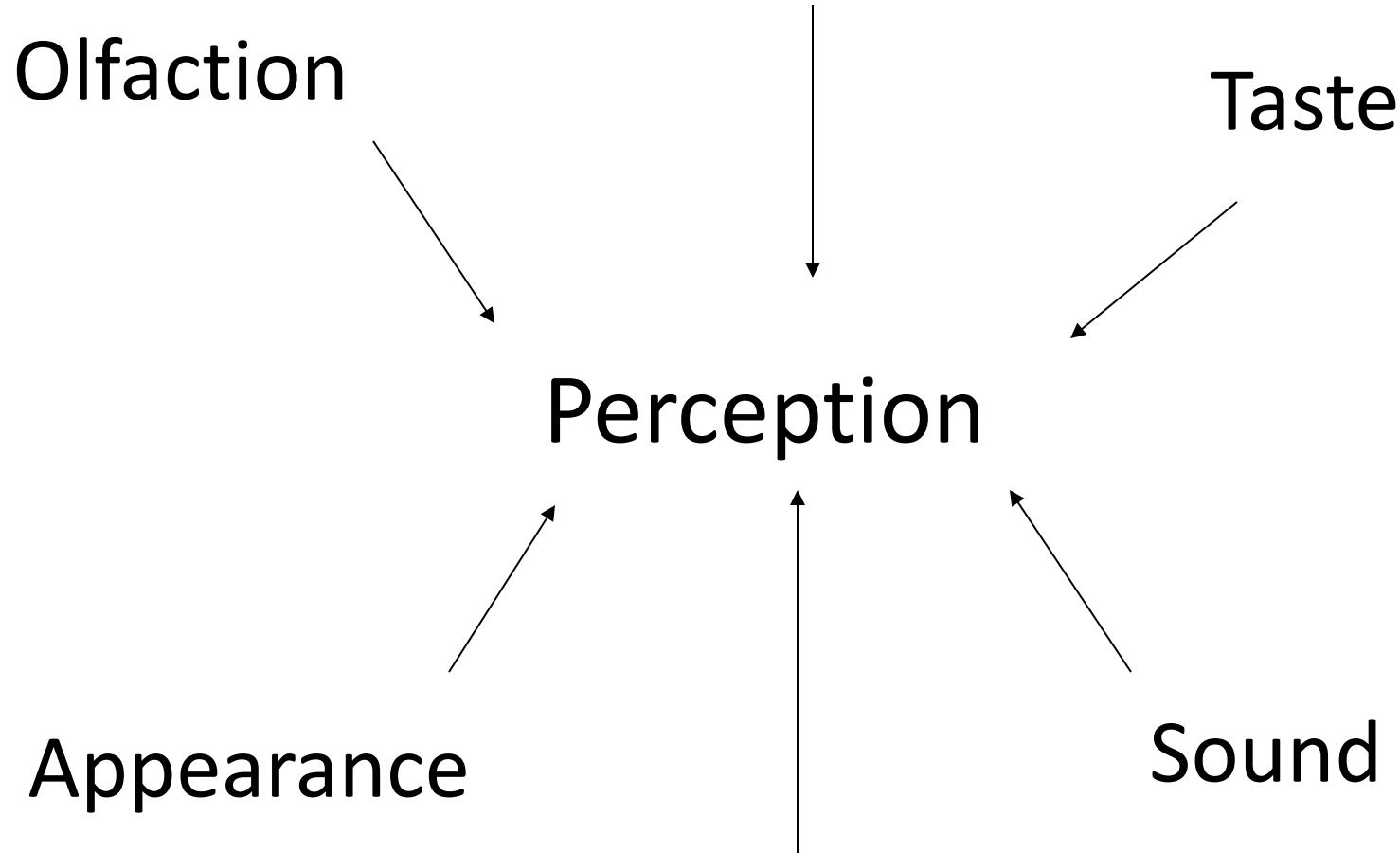
Sweetness, aroma and perception in chewing gums



Effect of Whey Protein Gel Strength on Aroma Release (no sweetener)



Mouthfeel aspects



Chemesthesia

Challenge remains



- How do we take chemical data and link it to perception i.e. Chemically define the stimuli that define perception of a given flavor, in a given food matrix?
- Need a method that takes interactions and release of an extremely complex set of stimulants into consideration.

Chemometrics (Flavoromics)

- ✓ Comprehensive & *data-driven*

Non-targeted

Flavor – inputs from
all measurable
chemical stimuli

**ALL instrumental data collected
are valuable *a priori***

(not restricted to earlier “thinking”)

Unbiased view of the food system

Flavoromics

- ✓ Comprehensive & data-driven
- ✓ Multi-disciplinary
- ✓ Chemometrics

Mathematical & statistical tools used to make rationale analysis of chemical measurements

Why needed?

Large data sets with multiple variables
Data visualization & interpretation

DATA ≠ INFORMATION

Applications

FLAVOROMICS

**Prediction of
sensory properties
(flavor)**

Linking chemical stimuli
to flavor perception

Discovery

Single mode interactions (olfaction – mixtures > 3 cpds take on a new character)

Cross modal interactions

Contribute to perception without having flavor itself

Some challenges (Compromises)

- Broad range of compounds (physicochemical & concentration)- multiple platforms for sufficient compound coverage (sensitive and comprehensive)
- High throughput – Severe limitation
- Success of flavor prediction depends strongly on samples used – sample needs
- Data handling - amounts & complexity of data generated

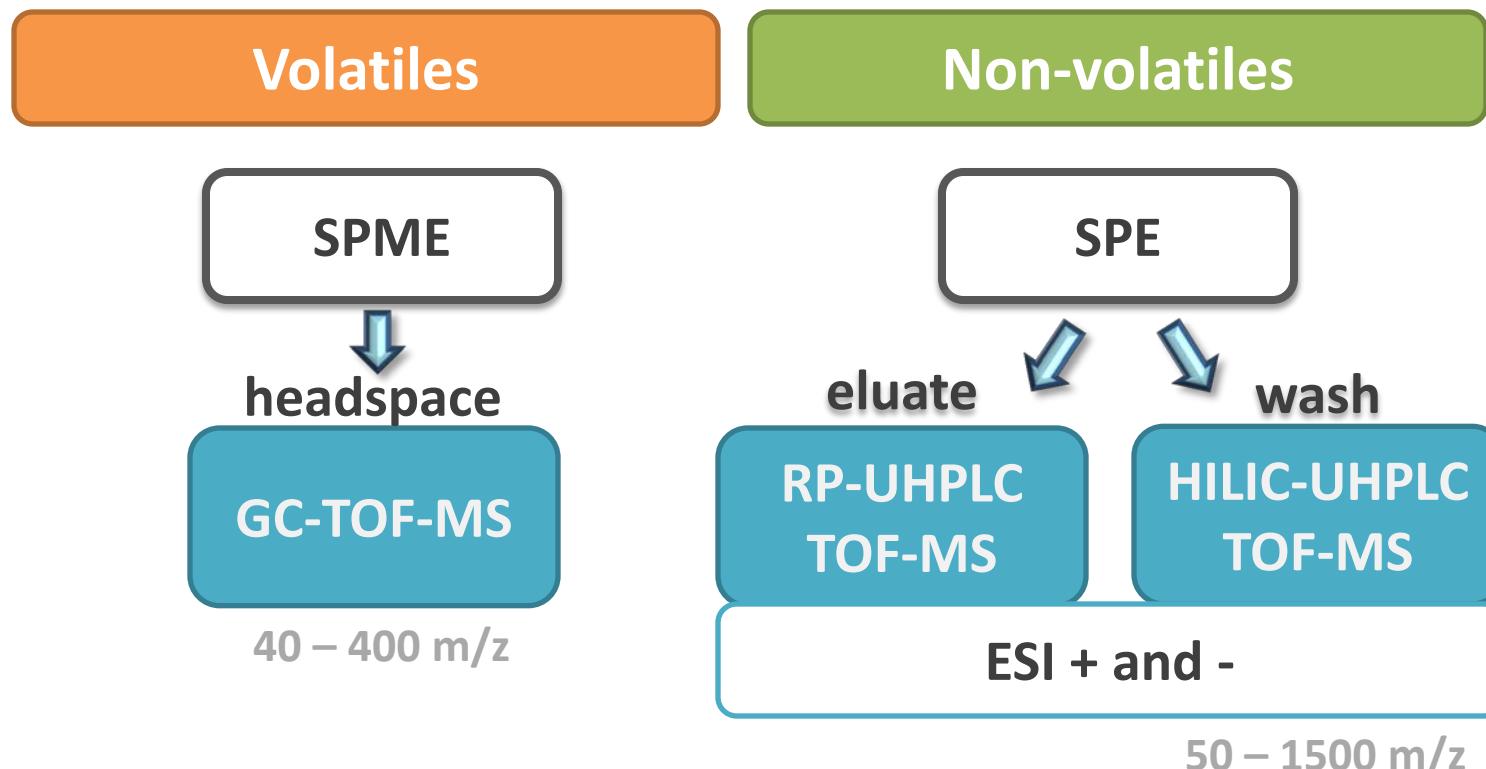
Flavoromics & mandarin fruits

- Characterization of sensory profile of new hybrids
 - trained panel
- Identification of “flavor markers”
 - earlier selection of fruits with “potential”
 - knowledge of inheritance mechanisms of related genes



Non-targeted analysis of chemical stimuli

MS-based & complementary platforms



PRE-PROCESSING

Collected data (GC & UHPLC-MS)

Variable extraction (RT – m/z)

Alignment, noise subtraction

Filtering, normalization

Variable reduction

Averaging, centering, scaling

variables

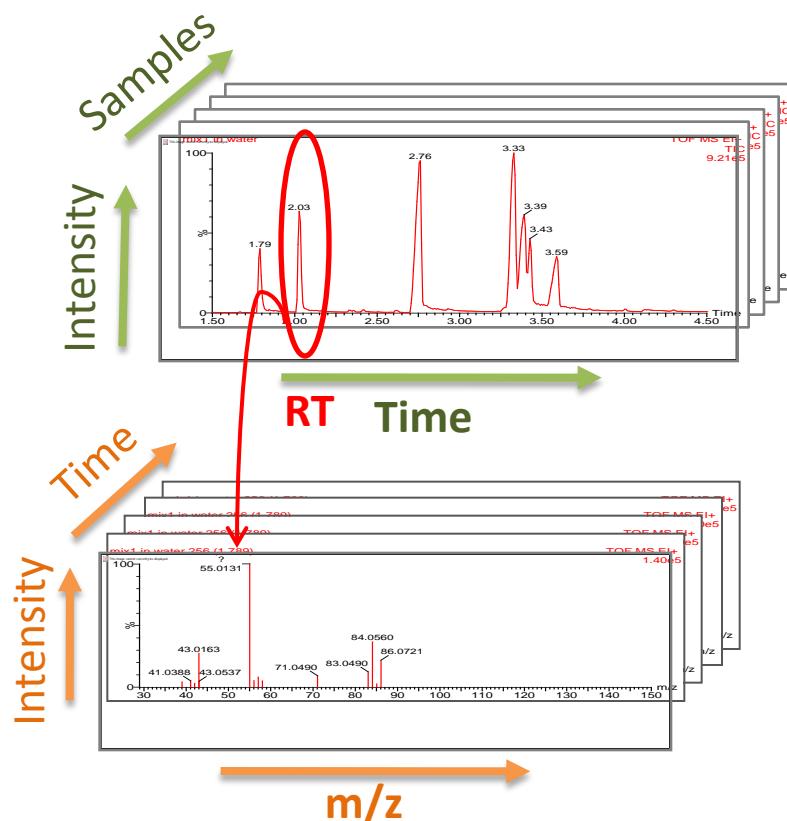
GC: 134 (9420)

RP neg: 439 (1116)

RP pos: 309 (1143)

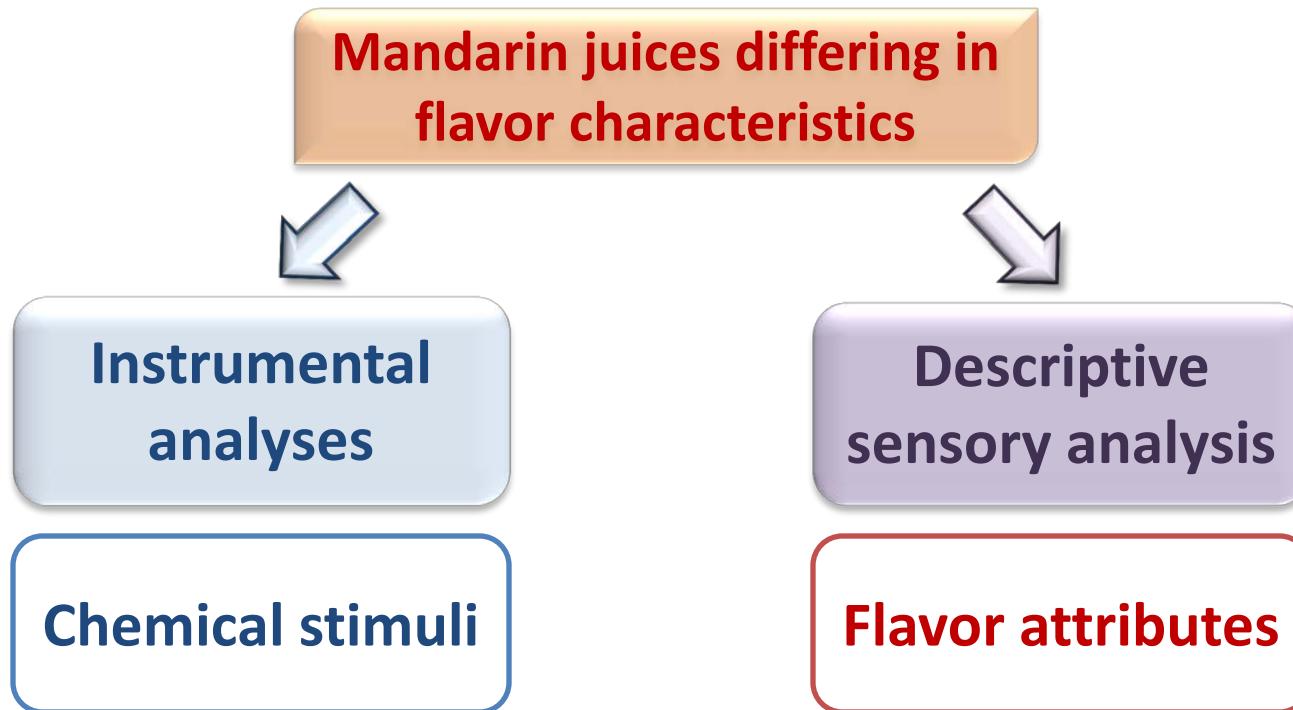
Amide neg: 300 (1002)

Amide pos: 255 (1249)



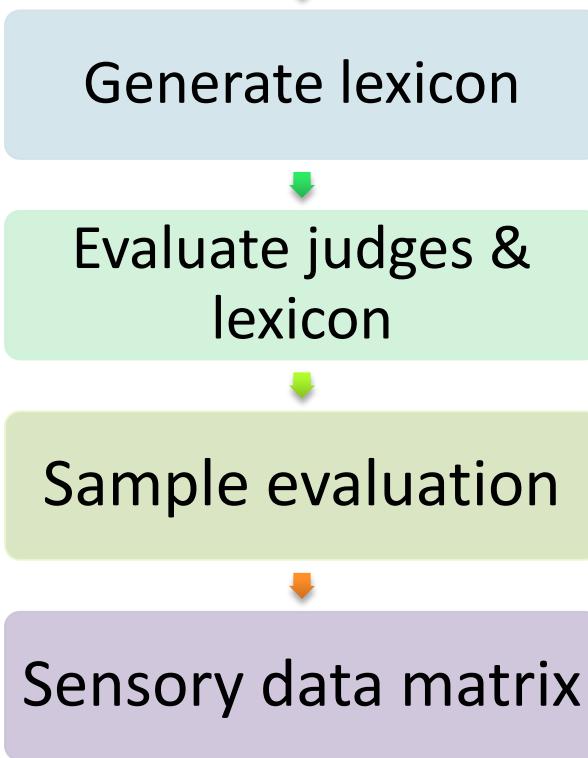
	RT- m/z			
	2.0210 - 146.0591	6.9022 - 425.1216	7.0587 - 373.1280
Juice A	Normalized relative intensities			
....				
Juice D				

Project workflow



Descriptive sensory analysis

Recruit panel



13-15 subjects

USDA/ARS, Citrus and Subtropical Products Laboratory (FL)

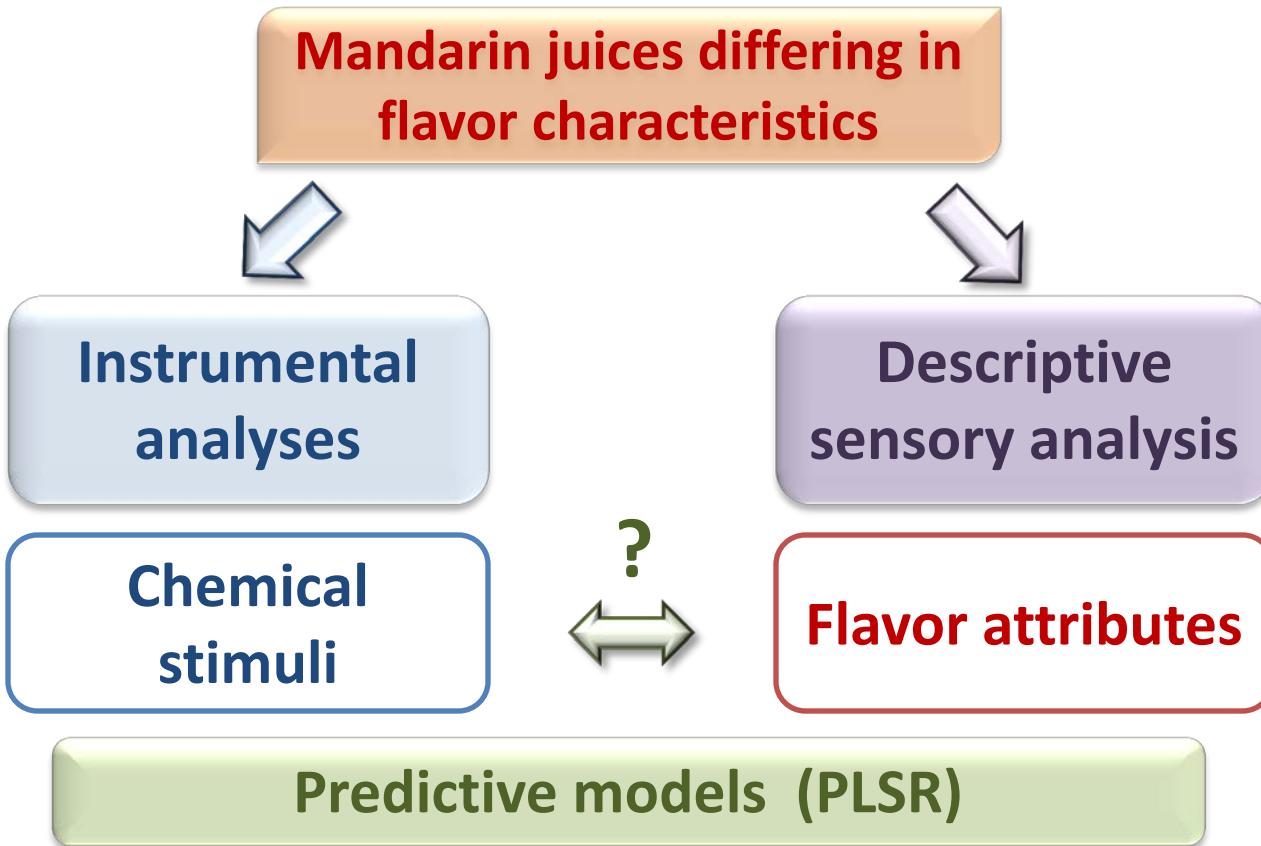
Training (~ 20 h each year)

19 descriptors (aroma & taste)

Sensory test (over 2 years)

	Green	Orange	Floral	...
Sample 1	Panel average scores			
....	Panel average scores			

Project workflow



Partial Least Squares regression – 1

Modeling method that finds a linear multivariate model to link two data matrices

INSTRUMENTAL

	RT- m/z	RT- m/z	
Juice 1			
...			
Juice i			

X
Predictors

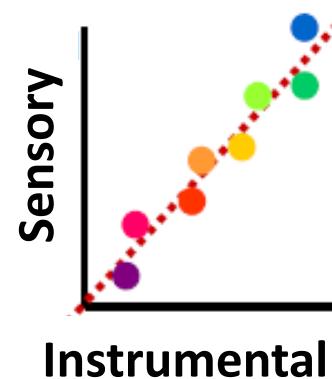
Find directions in X
which are predictive
of directions in Y

Maximizes correlation

SENSORY

	citrus	floral	bitter
Juice 1				
...				
Juice i				

Y
Responses



CORRELATION ≠ CAUSATION

PLSR – model selection 2

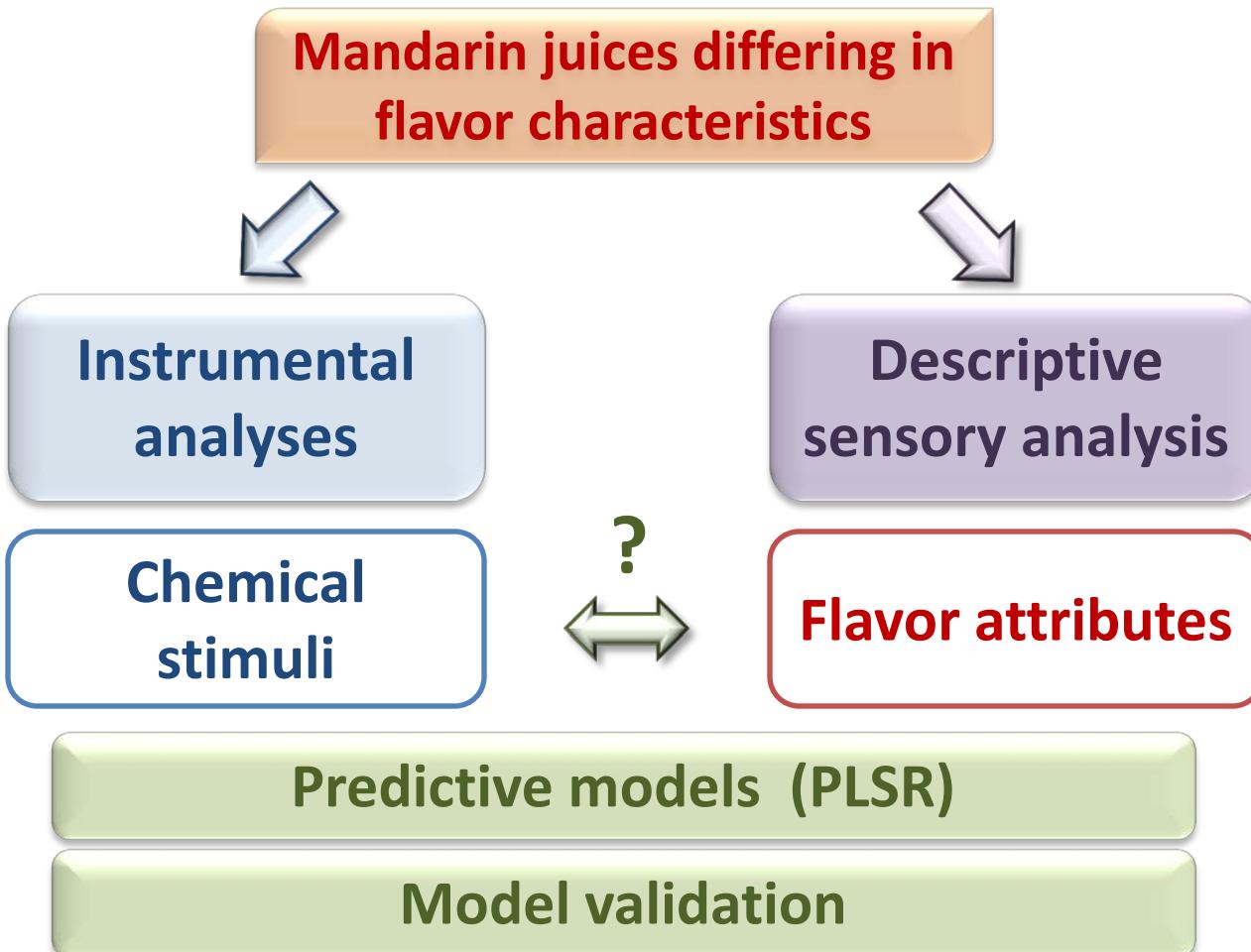
- Improved explanatory & predictive performances of the models when merging all instrumental data into one single data set (as opposed to individual ones)
- Best model with combined data & variable selection
 - 576 instrumental variables

Combined data & variable selection

GC			RP +			RP -			Amide +			Amide -		
	...	Var _j												
Juice 1			Juice 1			Juice 1			Juice 1			Juice 1		
...				
Juice i			Juice i			Juice i			Juice i			Juice i		



Project workflow

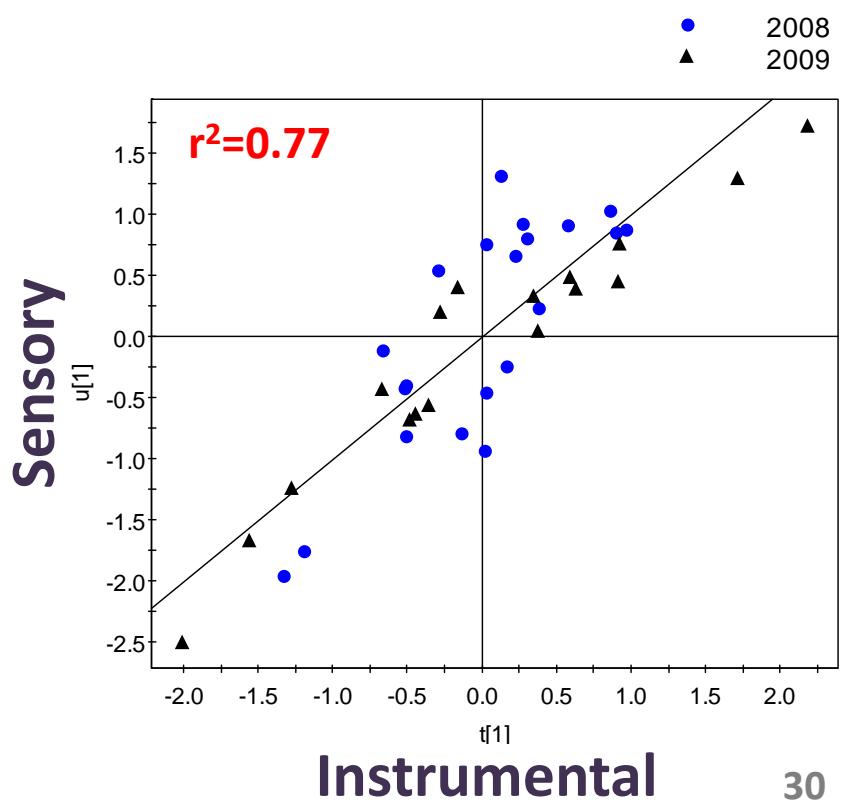


PLSR – model validation 1

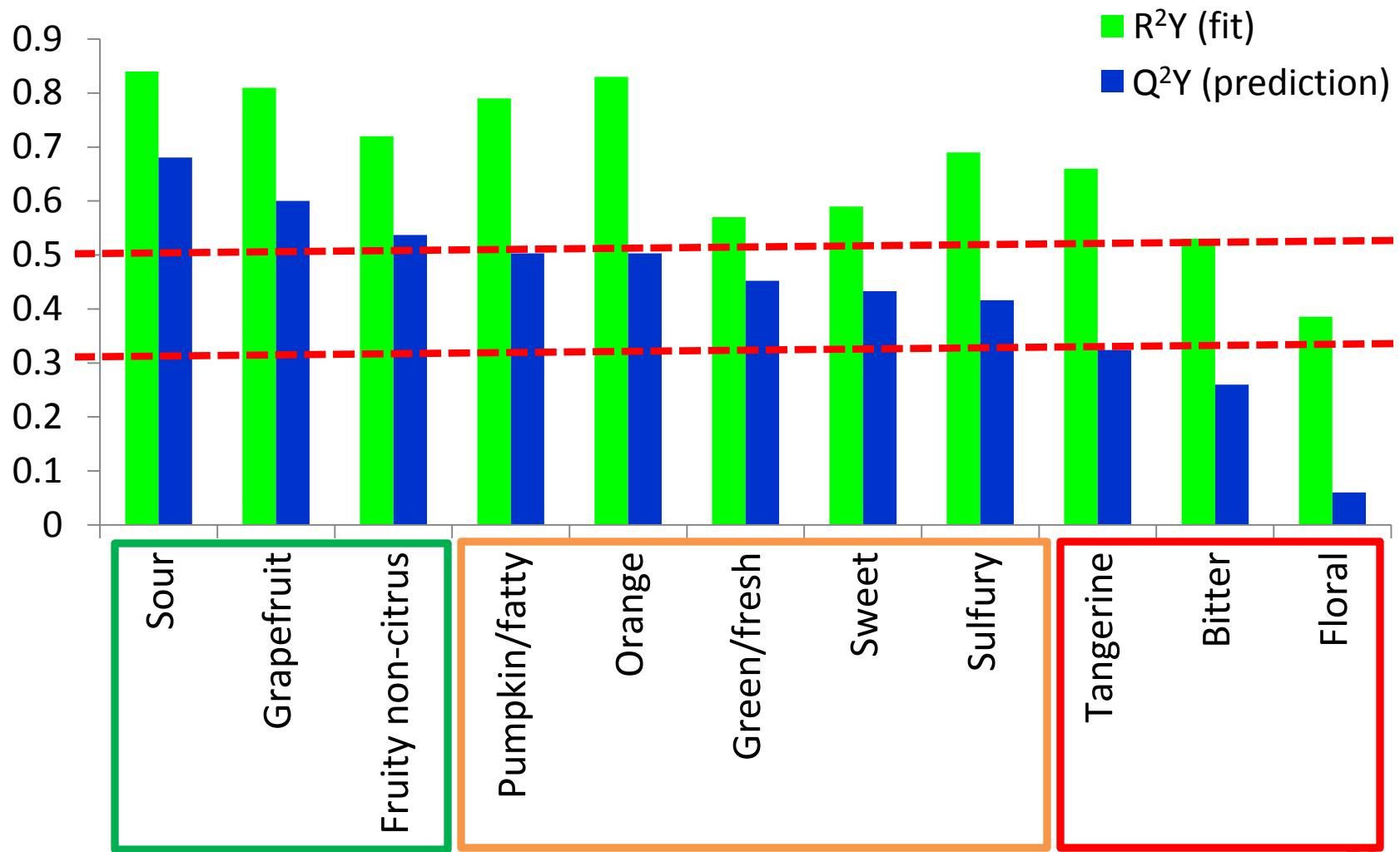
- External validation

- **calibration set** for model development (38 juices)
 - **prediction set** for model testing (8 juices)

Strong relationship between instrumental & sensory

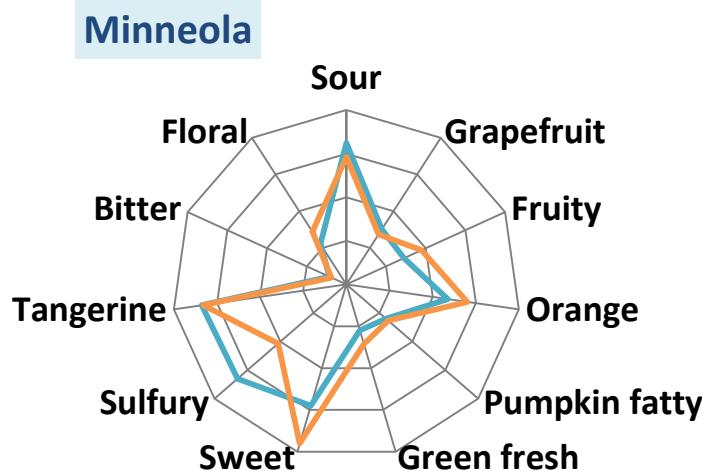


PLSR – model validation 2

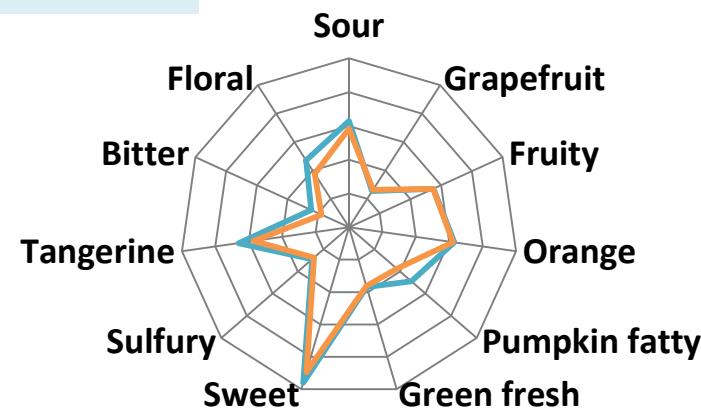


PLSR – model validation 3

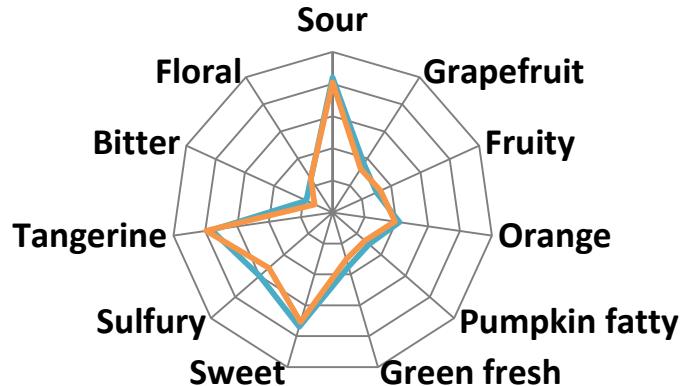
Sensory profile by trained panel



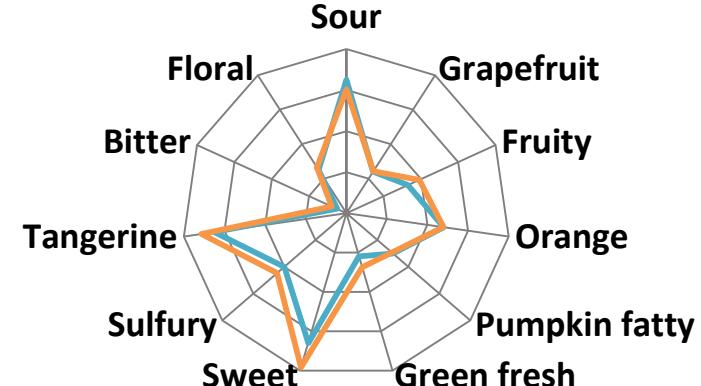
Predicted by PLS model



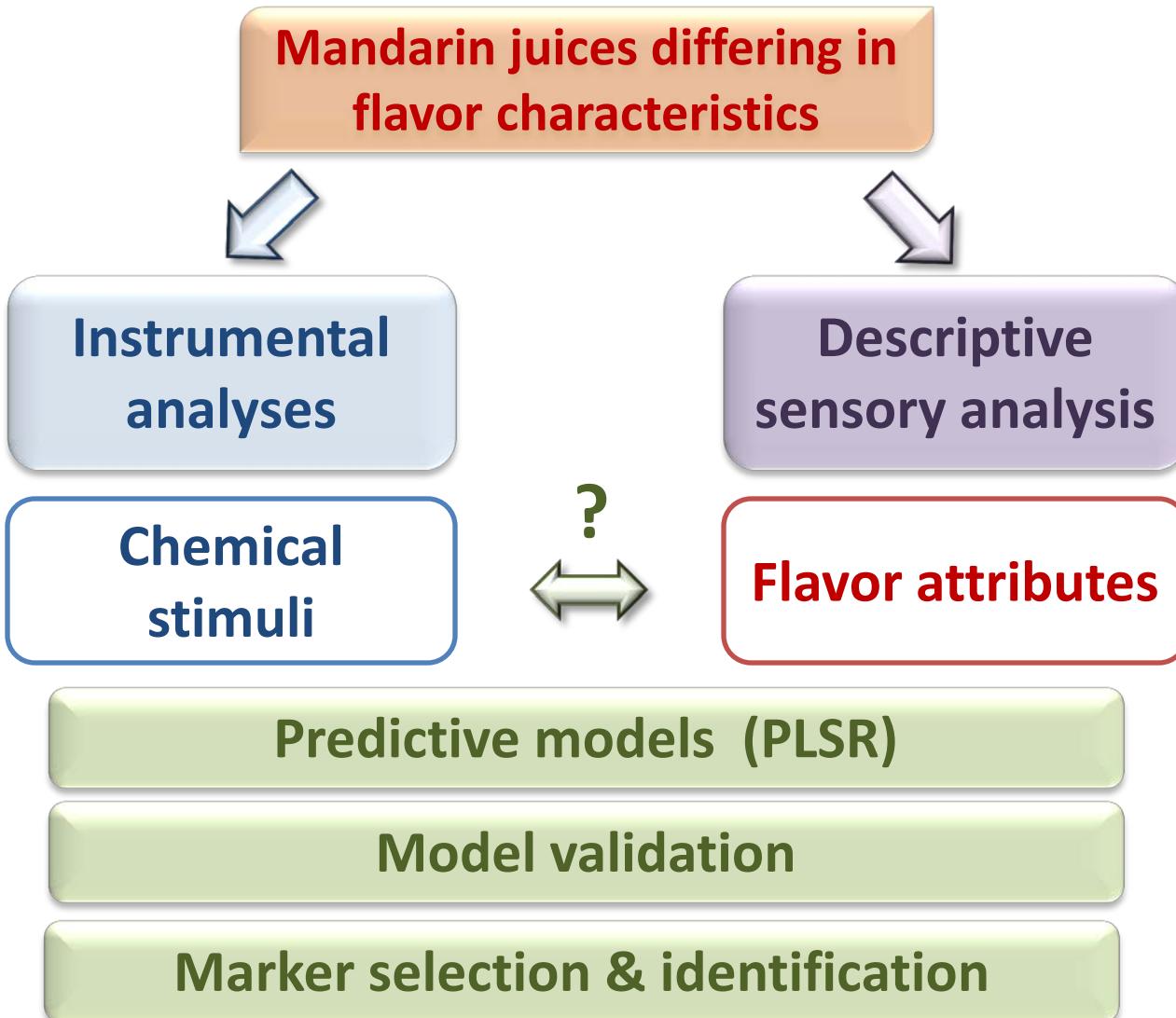
8-9 x Murcott (d)



Fallglo



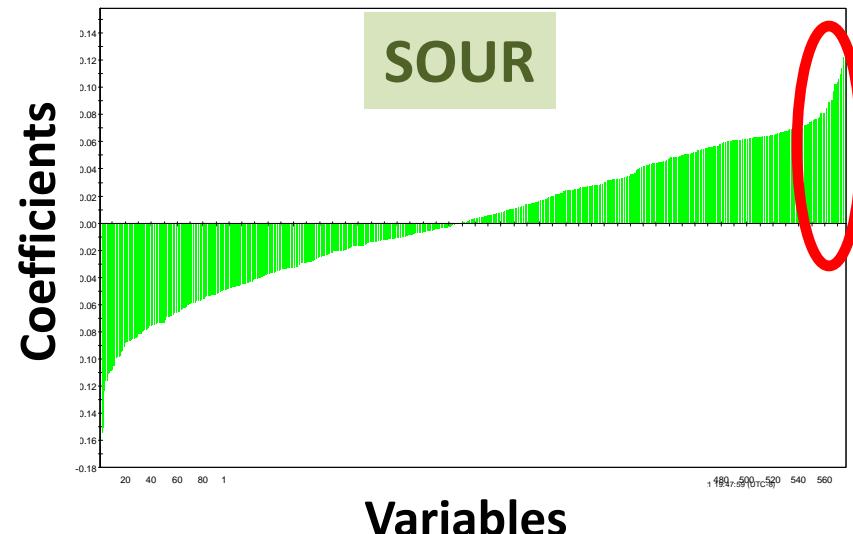
Project workflow



“Flavor marker” selection

Flavor marker = variable (compounds) most influencing the prediction of some sensory descriptors

- Regression coefficients
 - amplitude & direction of relationship of variables with selected response
 - variables w/ high & positive regression coefficient



Strategy for marker chemical identification

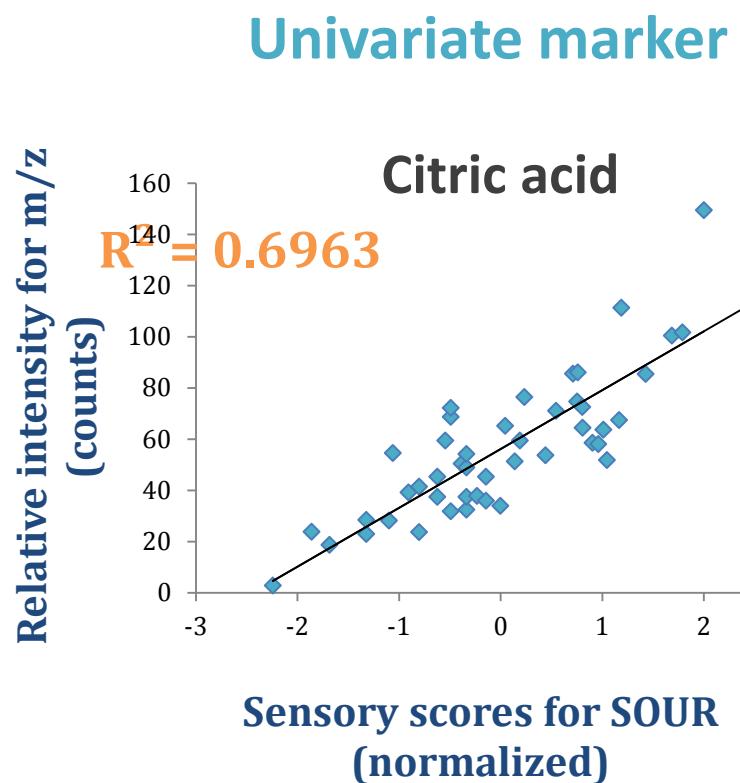
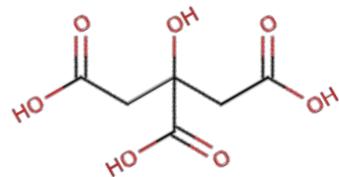
RT – m/z ???

- Mass spectra library & retention time indices
- Metabolite databases (accurate mass)
- MS/MS analyses
- Injection of authentic standard (if available)

Markers identification - 1

Sour

An_6.2338_191.0167 : Citric acid



Markers identification - 2

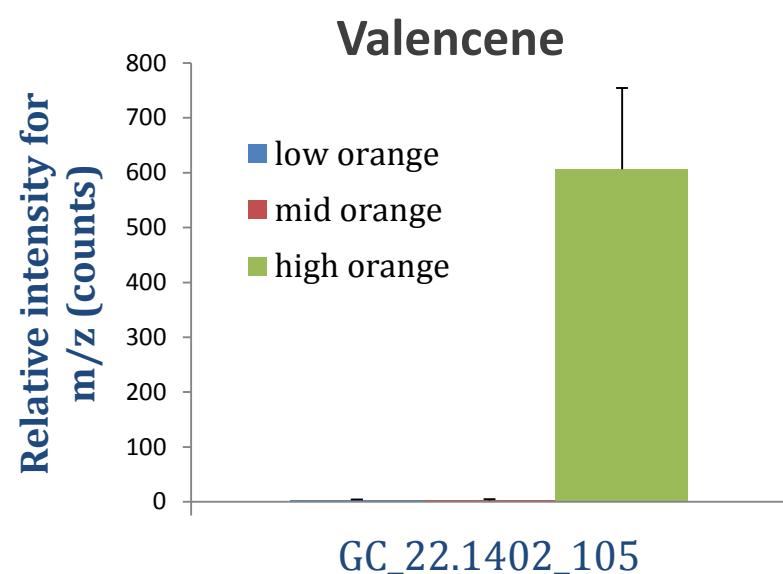
Orange

Ethyl butanoate

Sesquiterpenes

(Selinene, Valencene, α -Panasinsene)

Multivariate marker



However ...

- Correlations established are **NOT causative**
 - studies are necessary to confirm & to understand better the nature/mechanisms of their contribution to flavor
 - Critical evaluation – selection of likely candidates
- Reported markers might have :
 - have synergetic/masking effects with other compounds

Challenges

- Much room for improvement
 - Instrumental
 - Sensitivity (all chemical composition data needed?), MS drift, accurate quantification, problematic identification, determination of importance (?)
 - data processing
 - Used linear models (other approaches – Random Forest or? – Ian Ronnigen)
 - Sensory
 - Difficult to separate liking from scores, dumping effect, individual sensitivity to stimulus, trained panel
 - need for high throughout

What do we gain?

- Improvement in flavor quality – incremental
- Definition of less characterizable attributes e.g. freshness
- Long term - definition of product flavor and acceptable and unacceptable profiles
- Drivers of individual sensory attributes – may link to plant sources, processing and storage (process flavorings)
 - Pathways and therefore methods to potentially control flavor attributes.

What cost?

- Much
 - Special room for instrumentation (humidity, temperature and “noise” control.)
 - Investment in state of the art equipment and columns
 - Multidisciplinary people – talented and dedicated!

- Use
 - FREC – Three dedicated projects in this area
 - Being applied in two major food/flavor corporations .