

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	

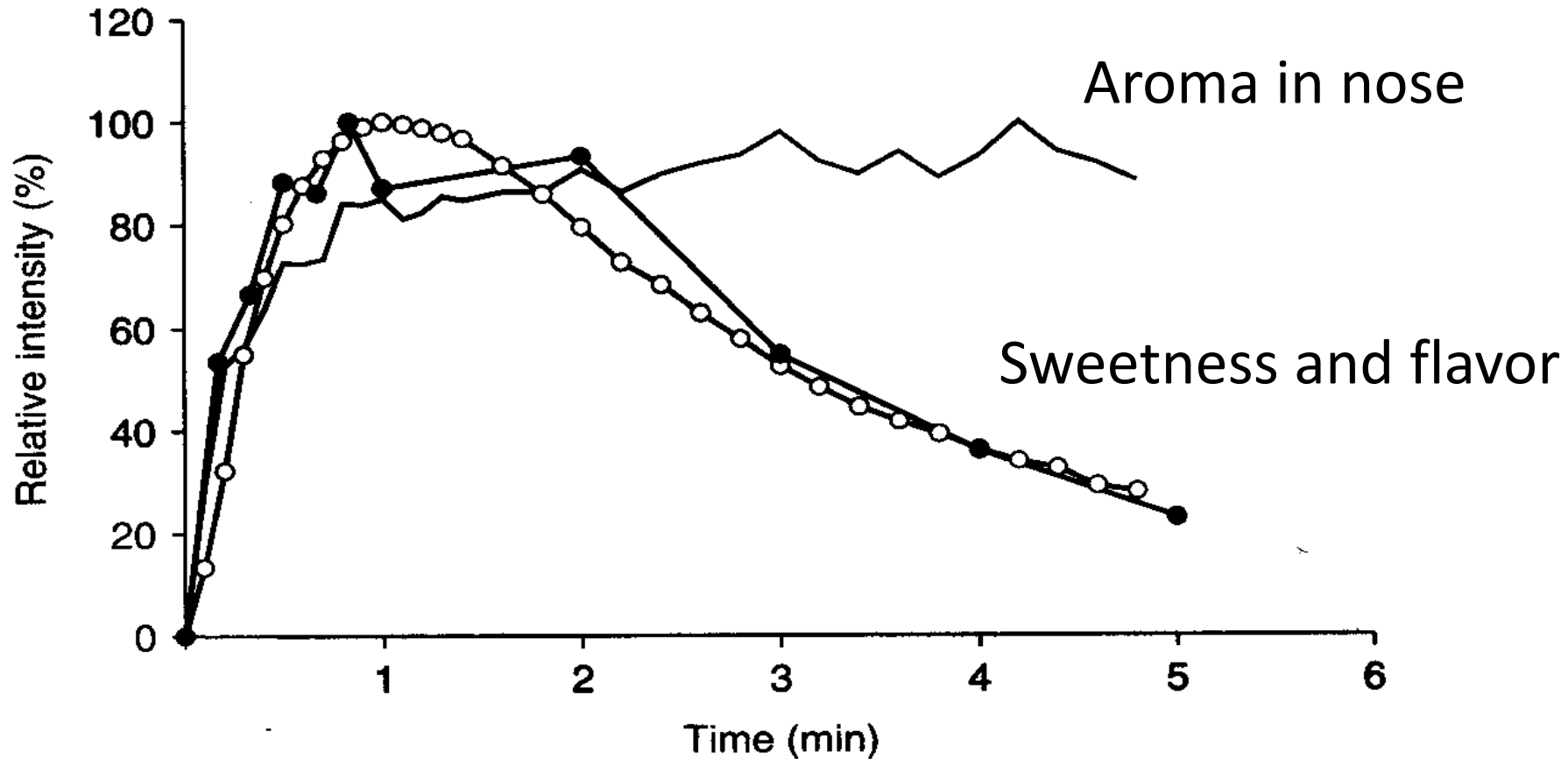
Add the selected compounds to an unflavored base (deodorized base).

- ❑ 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

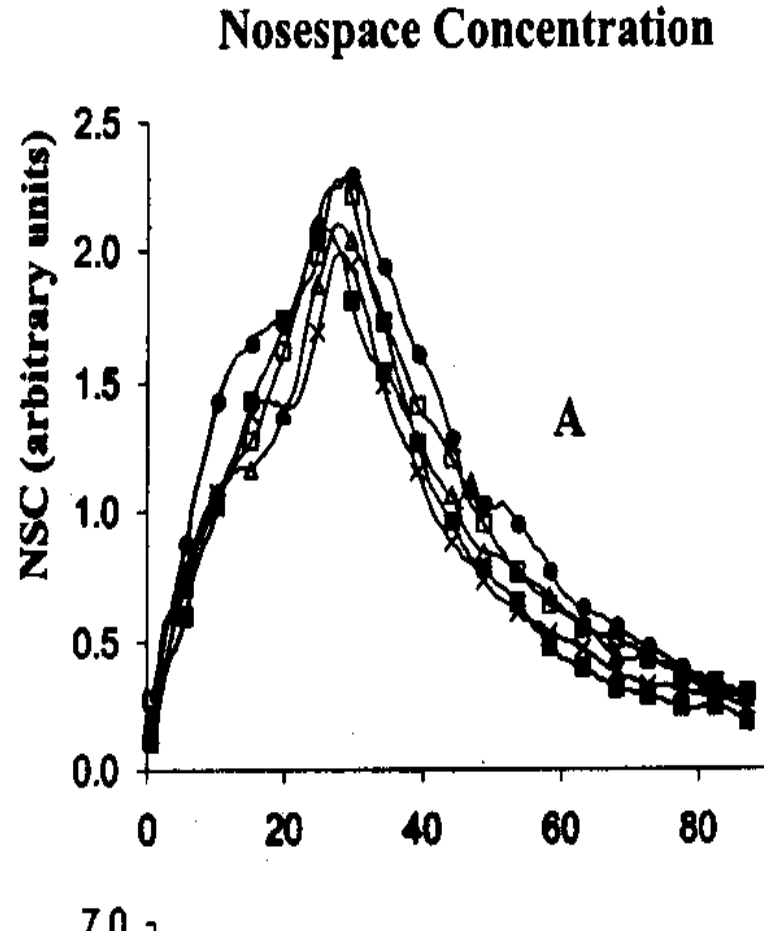
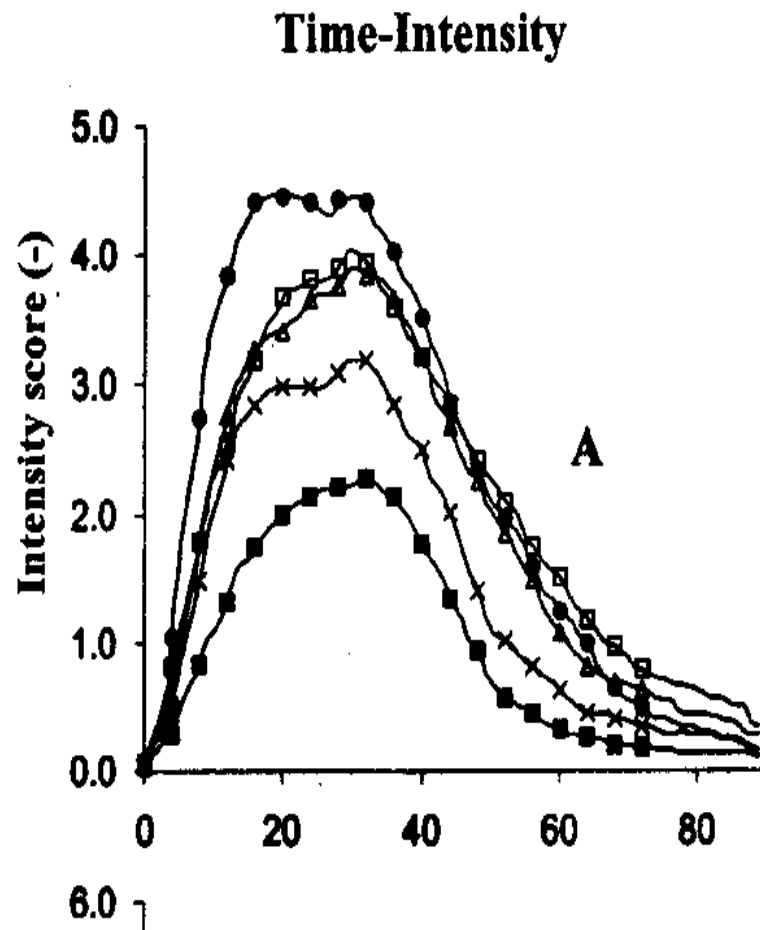
Mid 1990s – Aroma release

- ❑ 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

Olfaction

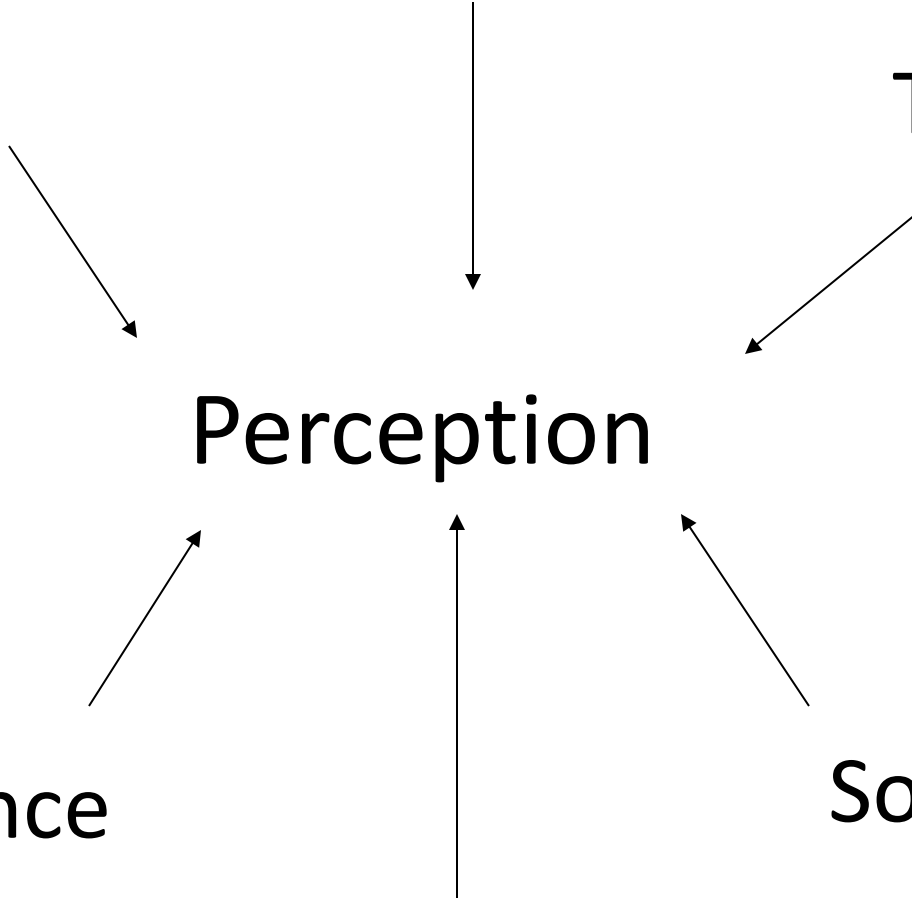
Taste

Perception

Appearance

Sound

Chemesthesis



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

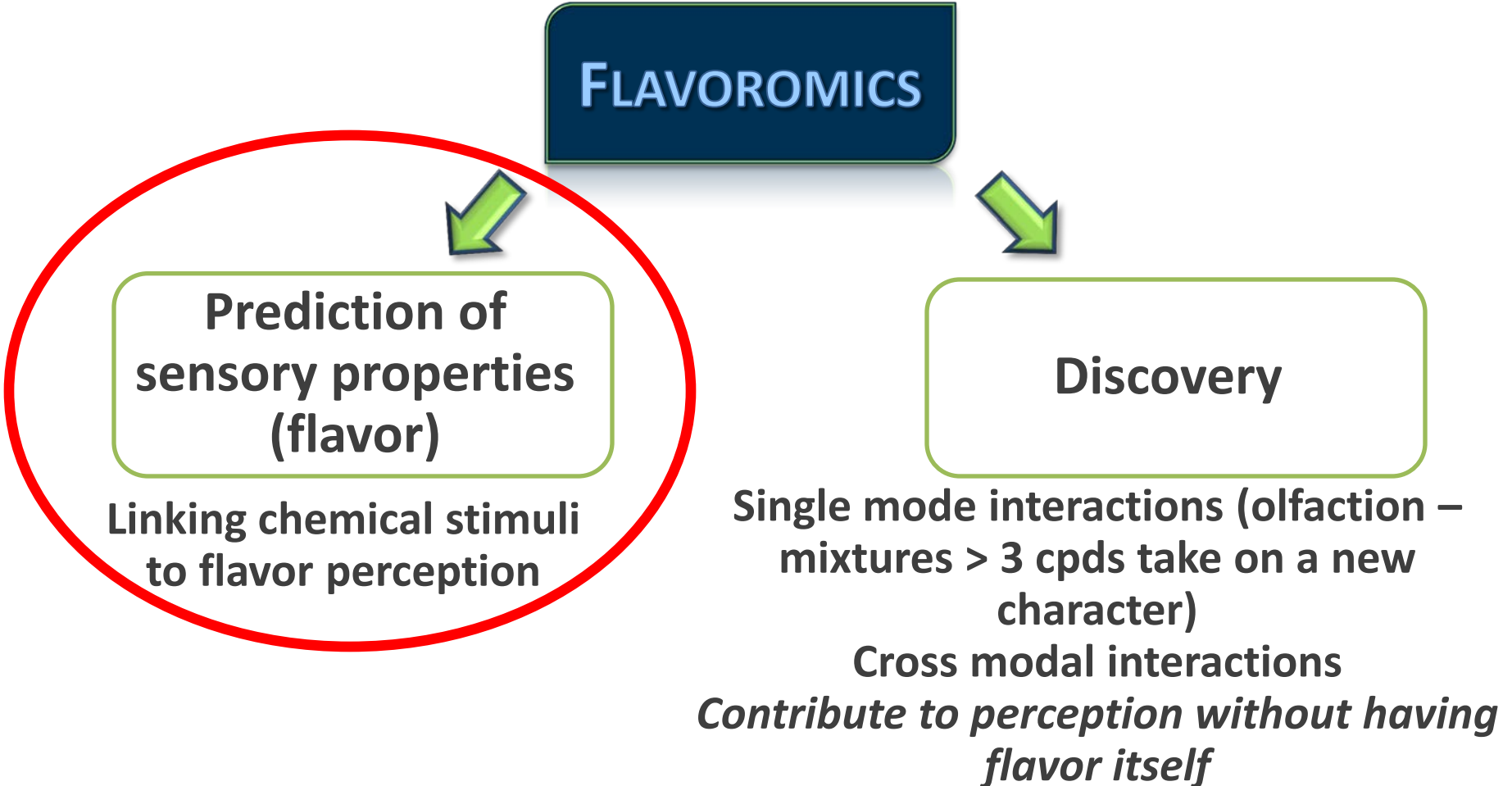
Why needed?

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

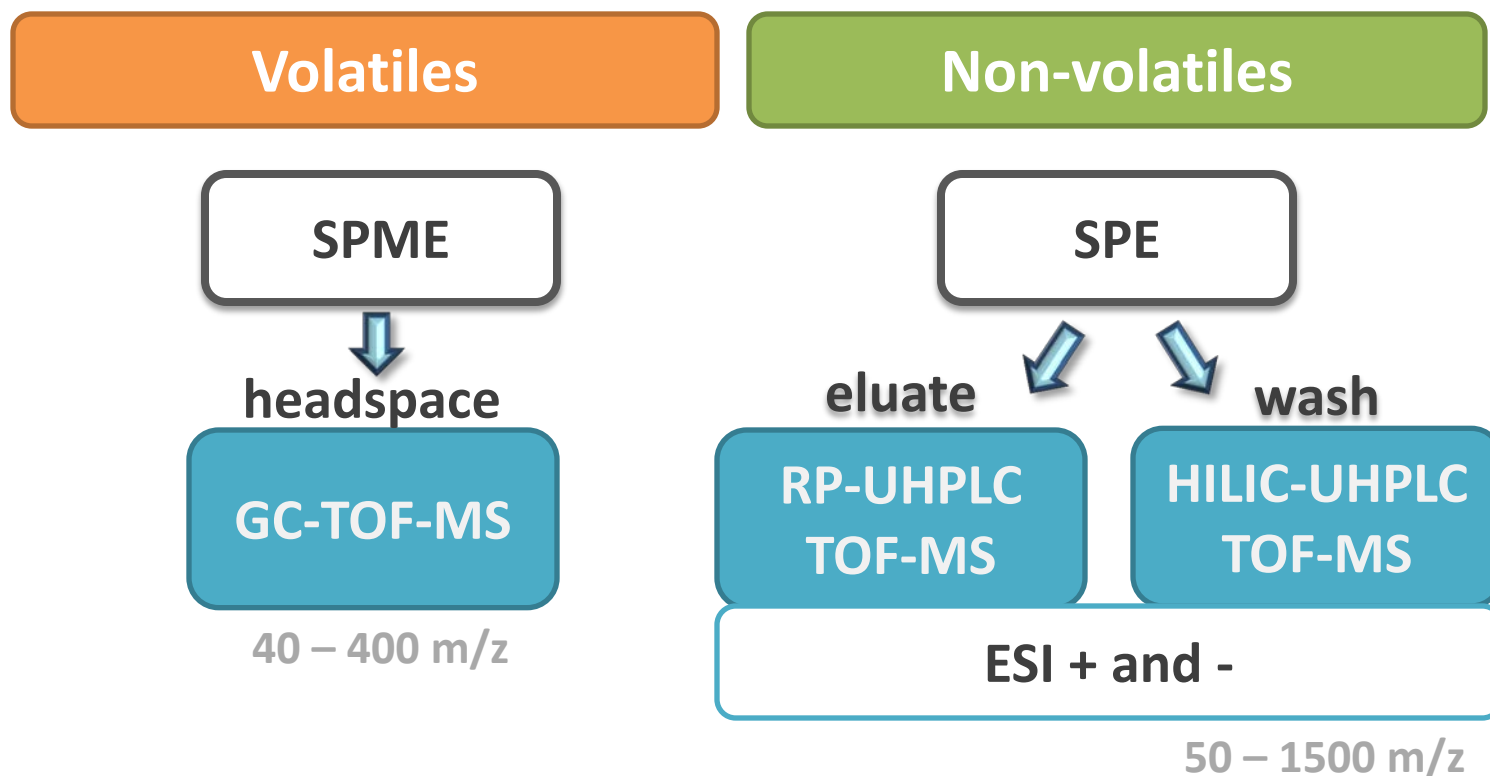
Flavoromics & mandarin fruits

- ❑ 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

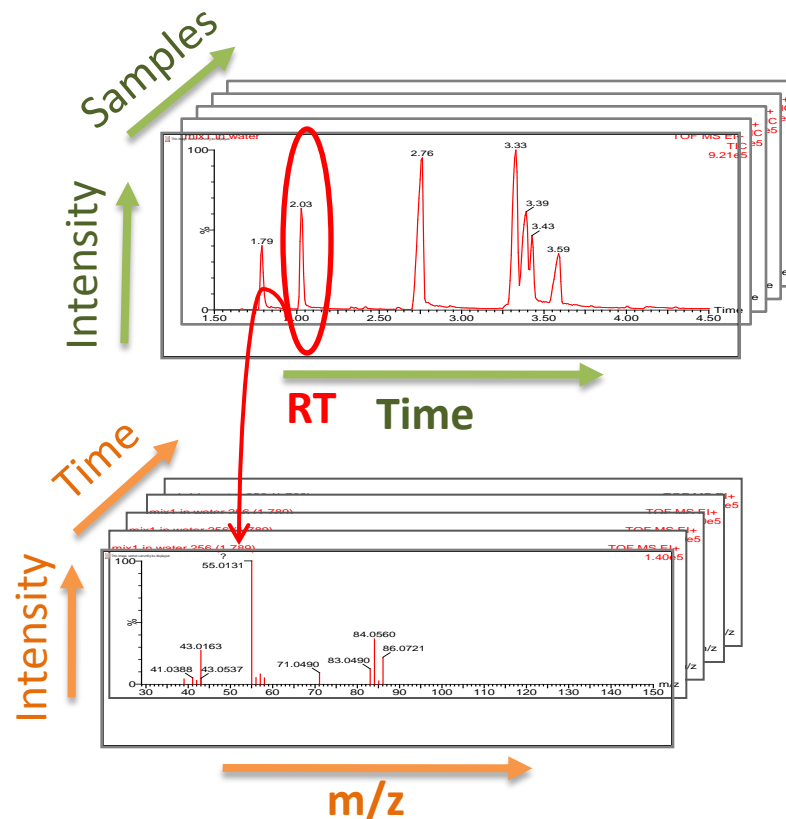
GC: 134 (9420)

RP neg: 439 (1116)

RP pos: 309 (1143)

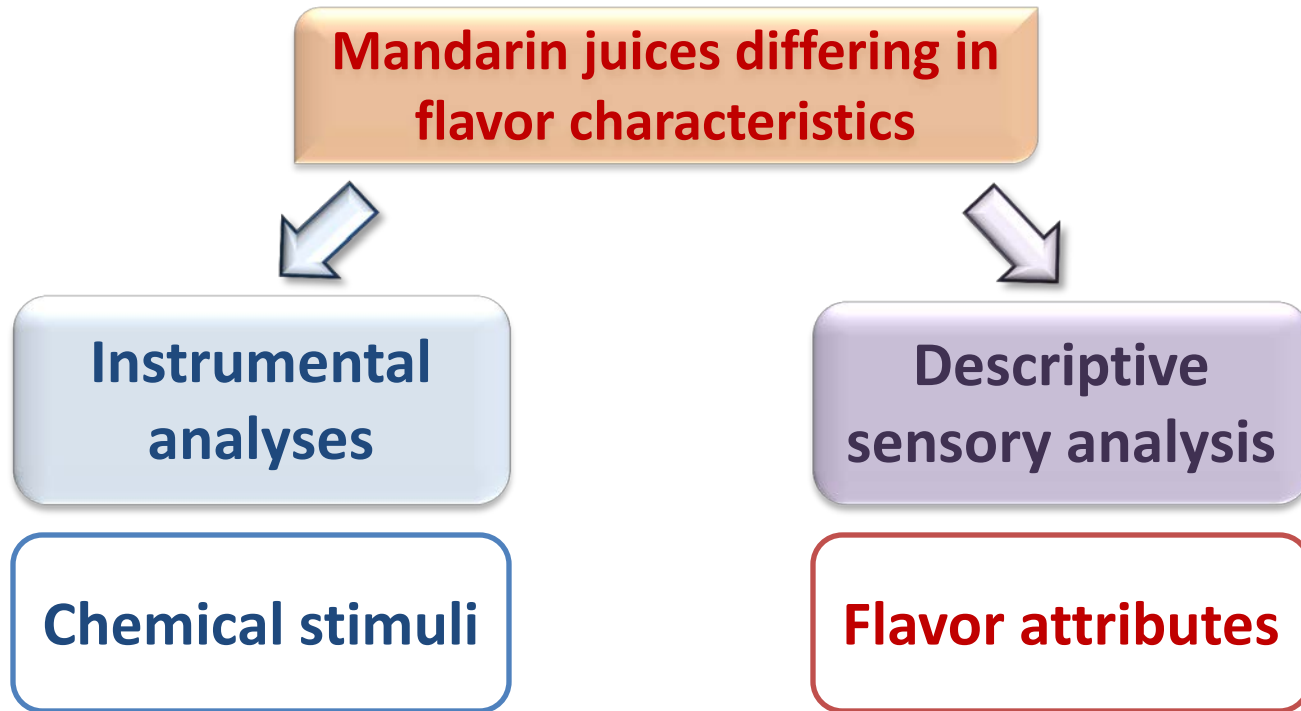
Amide neg: 300 (1002)

Amide pos: 255 (1249)

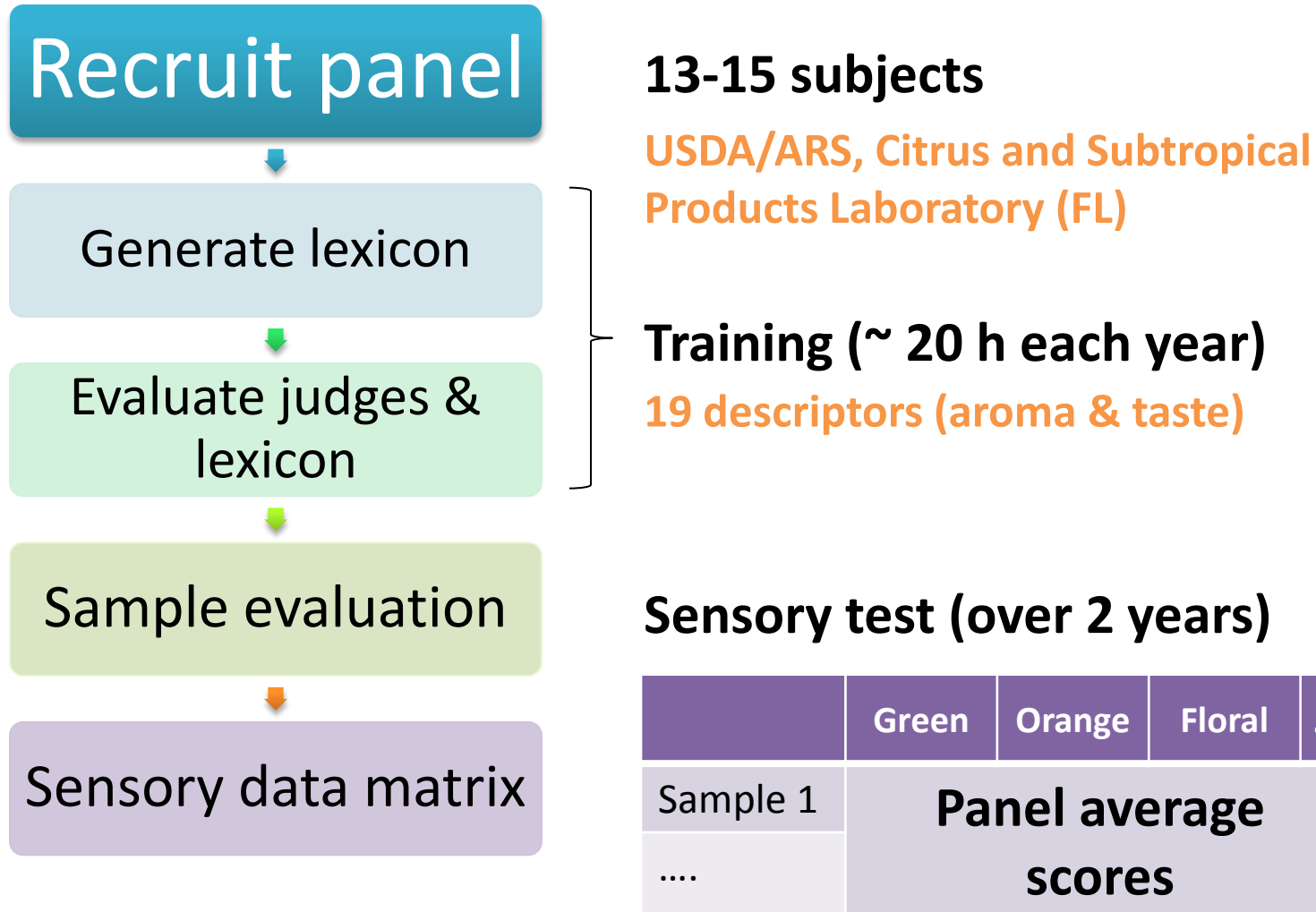


	RT- m/z		
	2.0210 - 146.0591	6.9022- 425.1216 7.0587 - 373.1280
Juice A	Normalized relative intensities		
....			
Juice D			

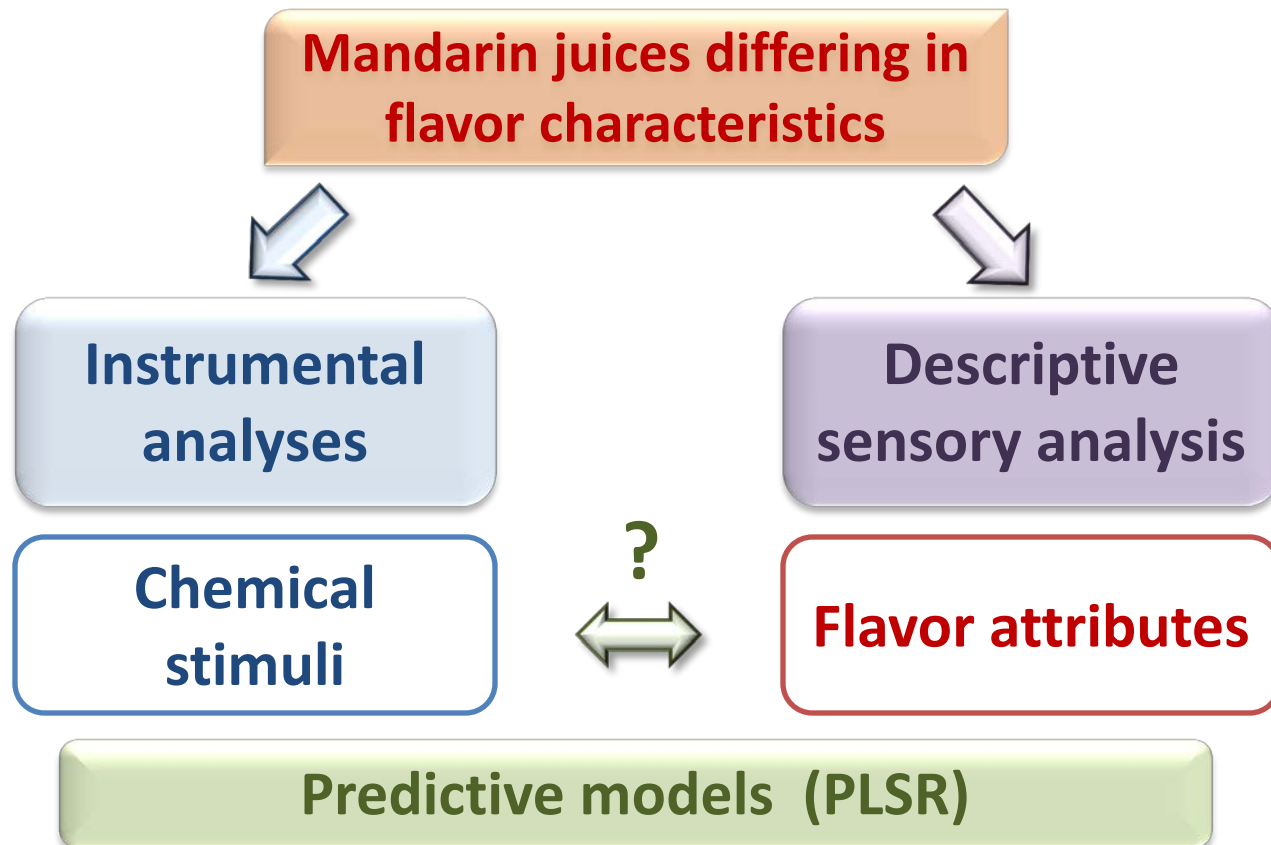
Project workflow



Descriptive sensory analysis



Project workflow



Partial Least Squares regression – 1

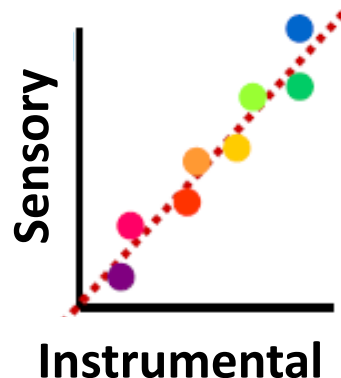
Modeling method that finds a linear multivariate model to link two data matrices

INSTRUMENTAL

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

SENSORY

	citrus	floral	bitter
Juice 1	Y Responses			
...				
Juice i				



Find directions in X
which are predictive
of directions in Y

Maximizes correlation

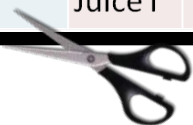
CORRELATION \neq CAUSATION

PLSR – model selection 2

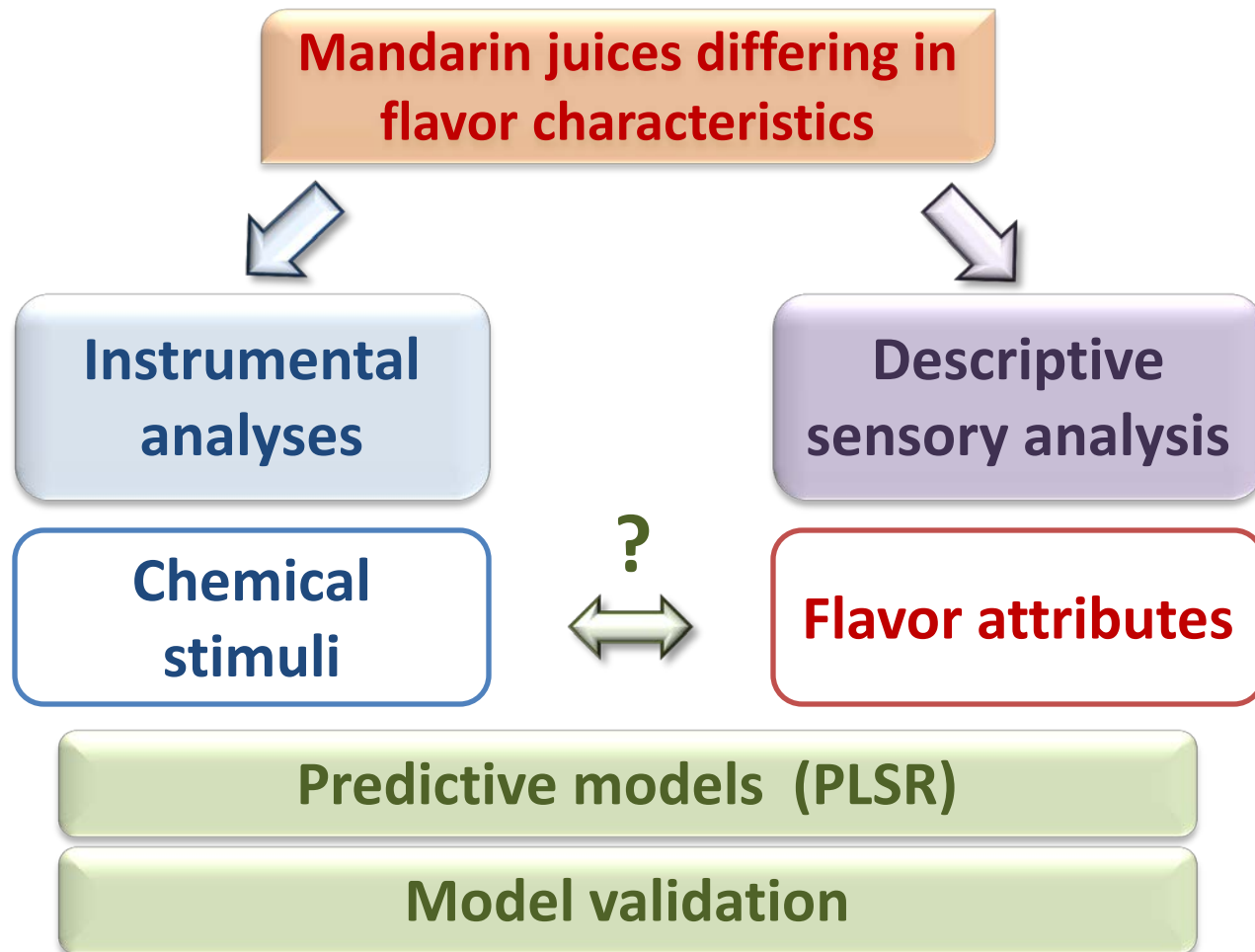
- Improved explanatory & predictive performances of the models when merging all instrumental data into one single data set (as opposed to individual ones)
- Best model with combined data & variable selection
 - 576 instrumental variables

Combined data & variable selection

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



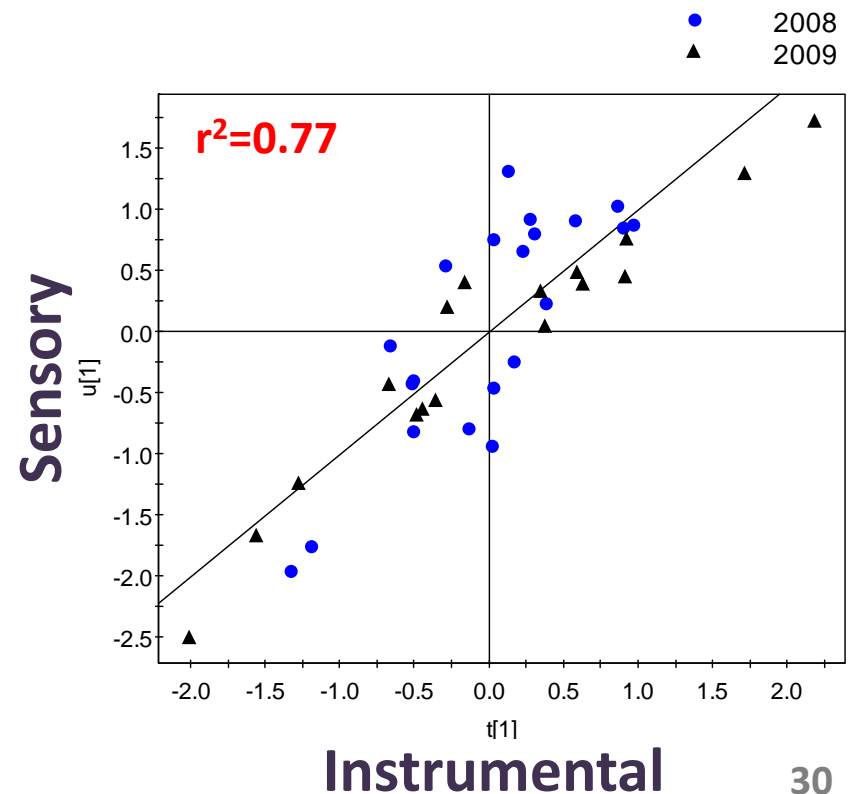
Project workflow



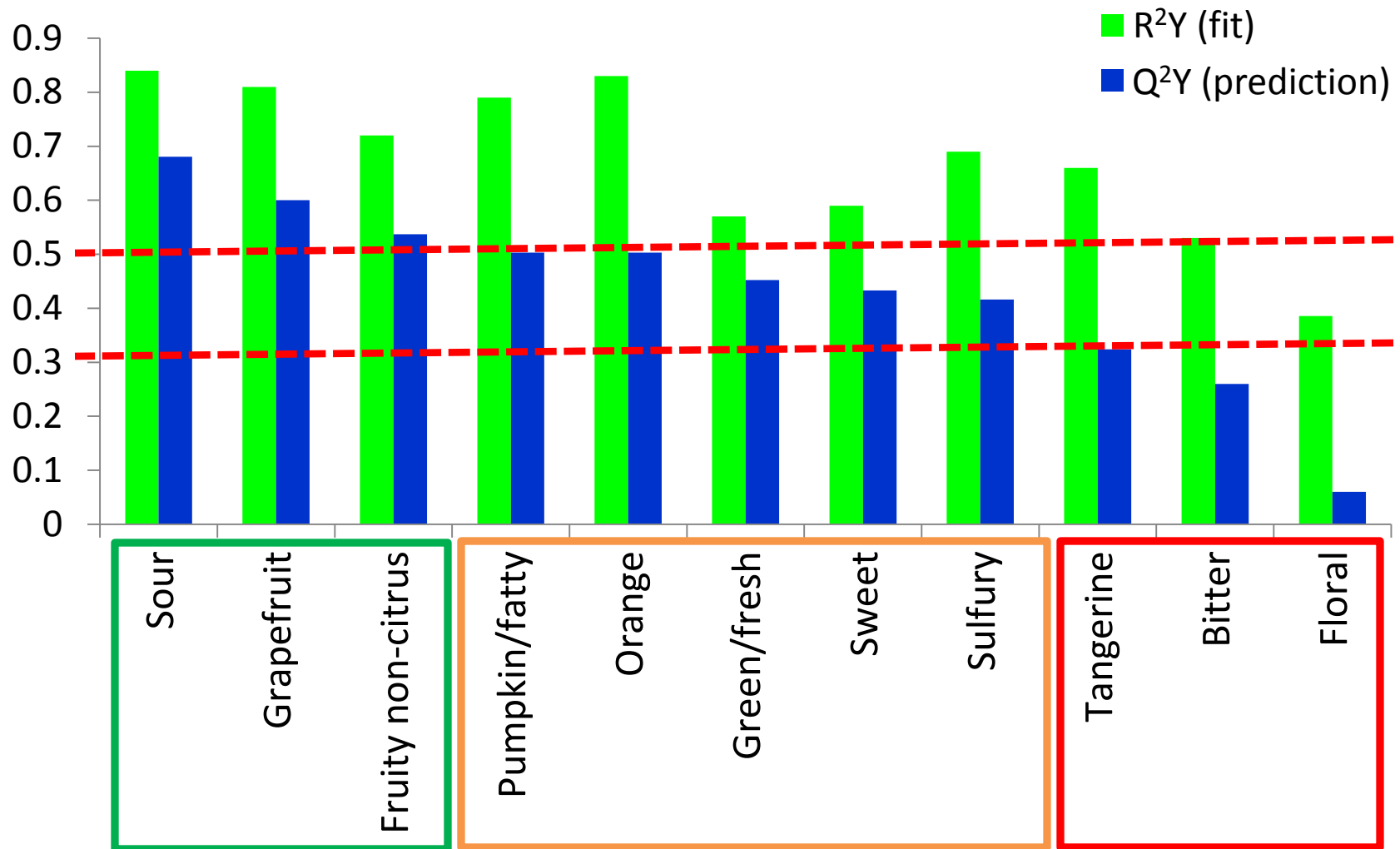
PLSR – model validation 1

- 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

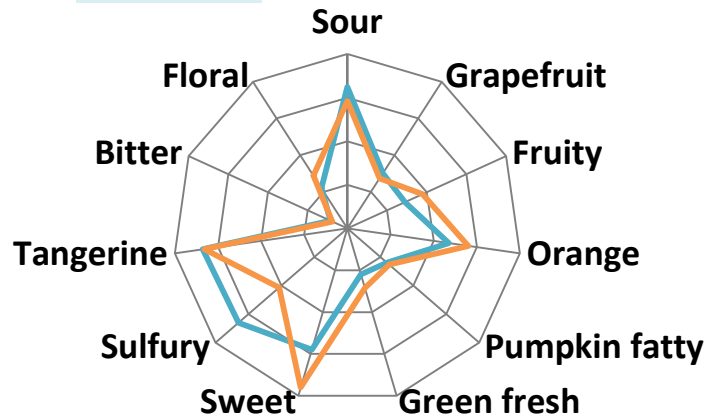


PLSR – model validation 3

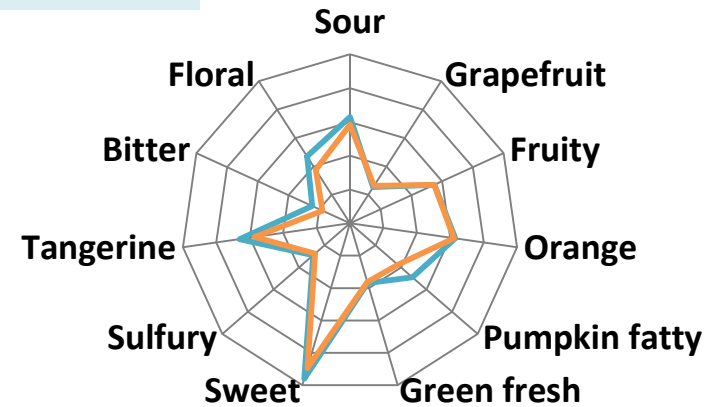
Sensory profile by trained panel

Predicted by PLS model

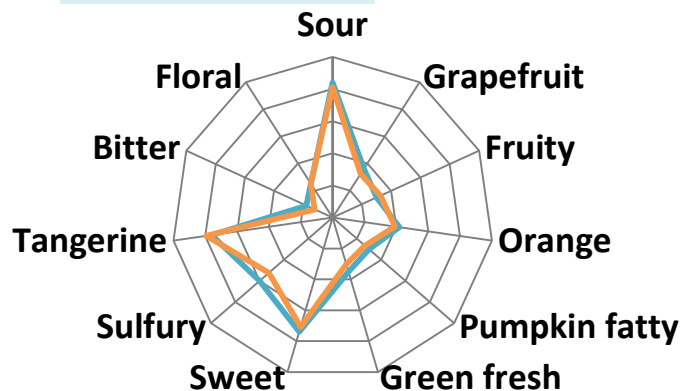
Minneola



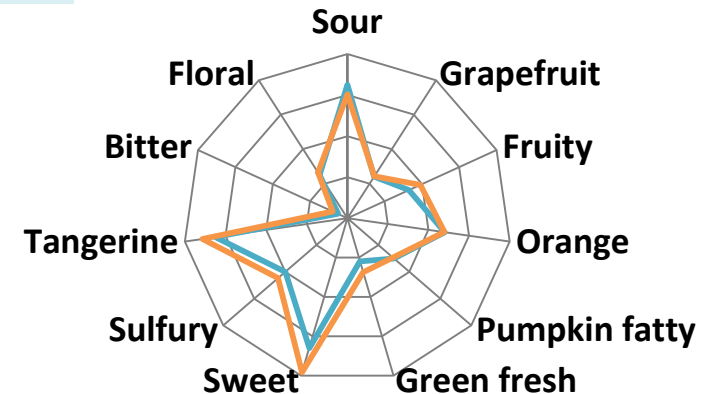
9-4 x Blood4x



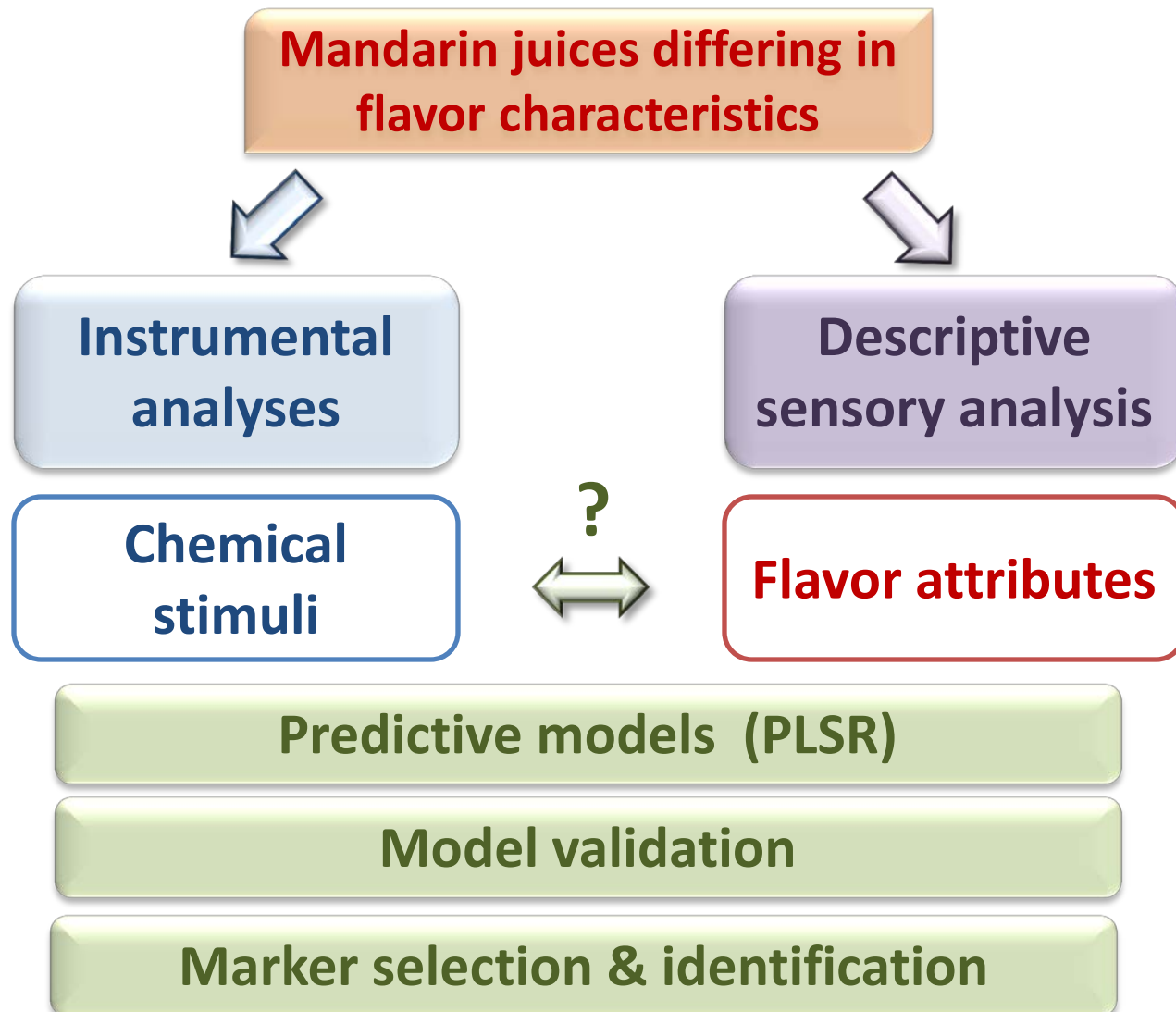
8-9 x Murcott (d)



Fallglo



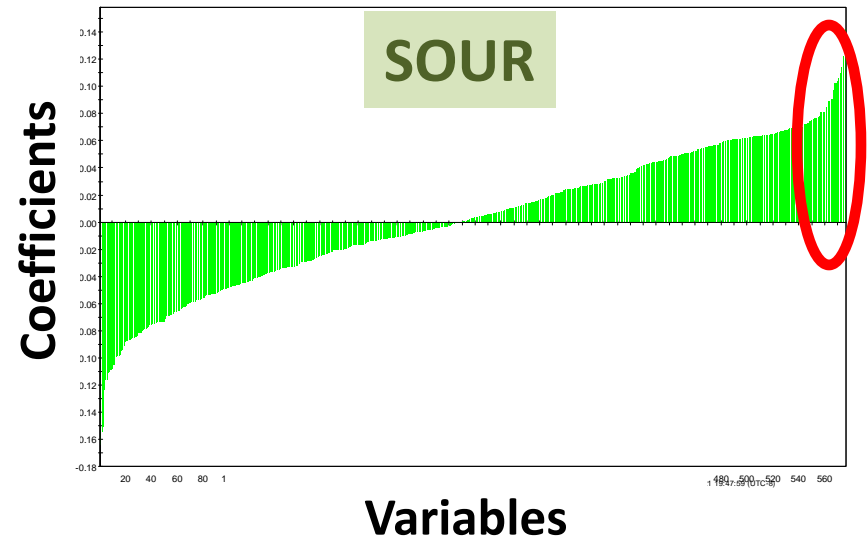
Project workflow



“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

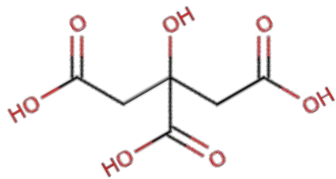
RT – m/z ???

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

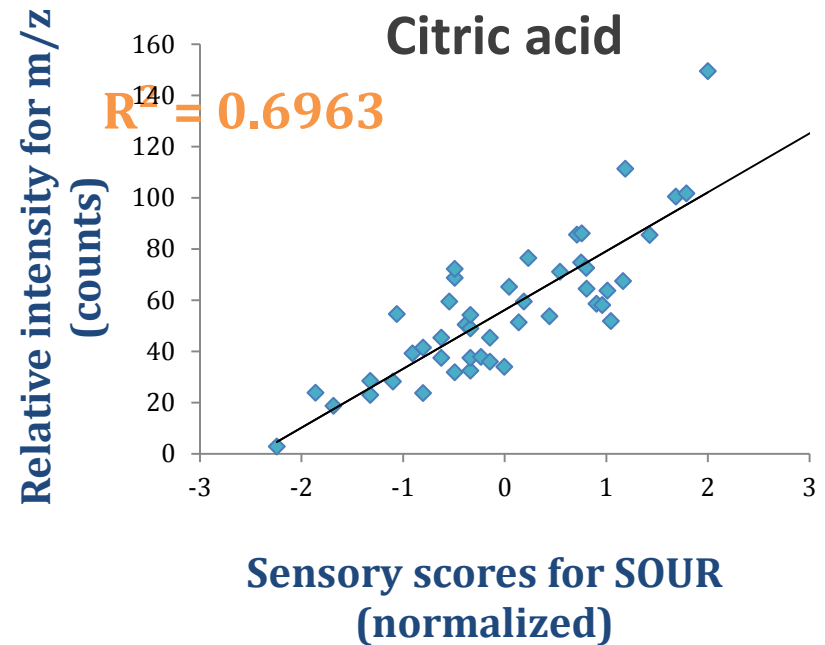
Markers identification - 1

Sour

An_6.2338_191.0167 : Citric acid



Univariate marker



Markers identification - 2

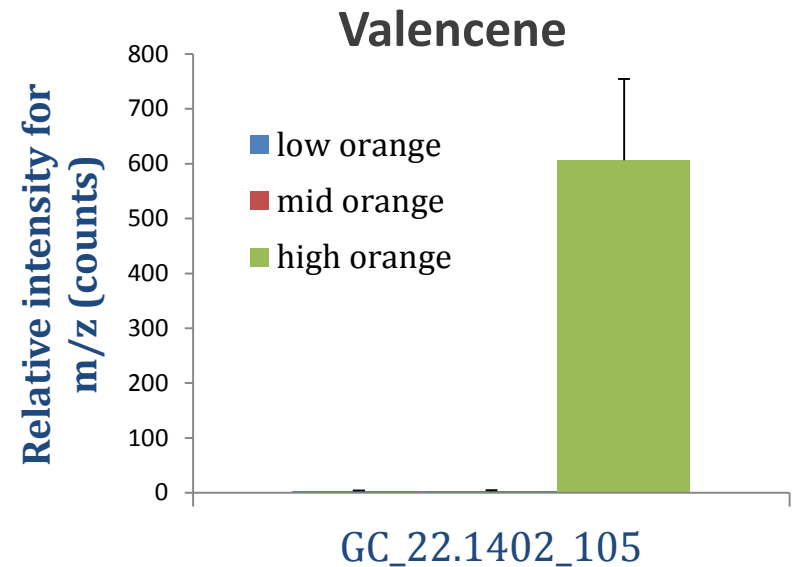
Orange

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 throughput

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 .