

# Proposal of statistical procedure to relate aroma chemistry data to aroma sensory data

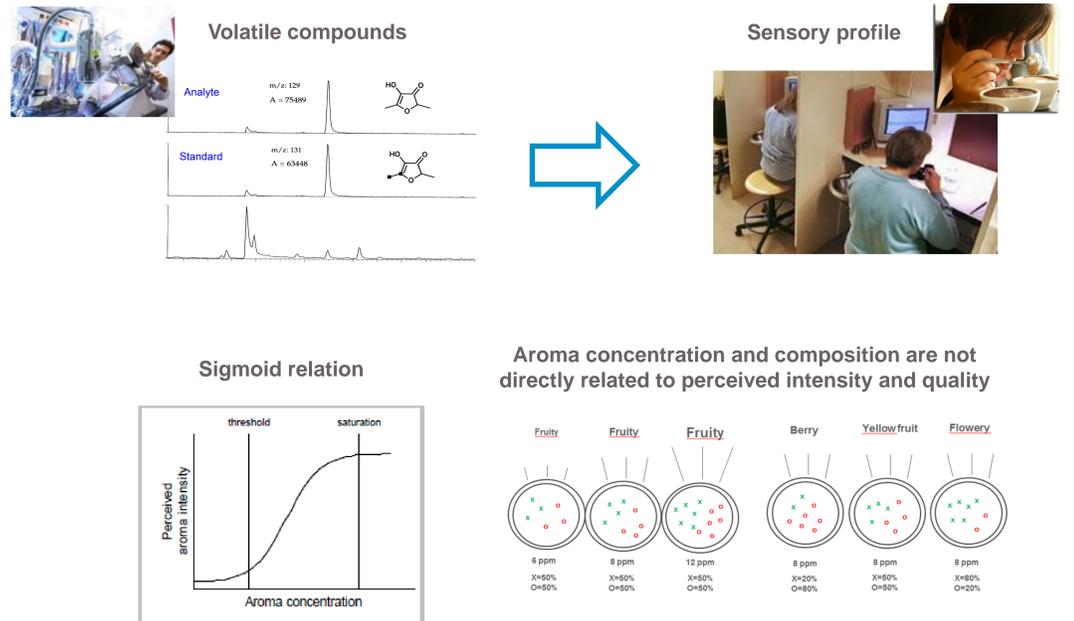
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## 1. Introduction

Directly relating aroma composition and concentration data to the perceived aroma quality and intensity present several challenges:

- The aroma concentration is not linearly related to the perceived aroma intensity (sigmoid relation).
- The absolute abundance of each volatile compound can impact differently the perceived aroma quality and intensity
- Human nose seems more able to identify volatile molecule patterns due to relative abundance than aromas through direct relations to absolute abundances (Menini, 2009)



## 2. Transformation procedure

A transformation of the raw analytical data is required to obtain an estimate of the aroma pseudo-composition which is meaningful to relate with the sensory profiles.

The transformation procedure is described in 4 steps.

### Step 0: Data structure

Consider analytical data of the form  $x_{ij}$  with:

- $i = 1, \dots, n$  rows (products)
- $j = 1, \dots, p$  columns (compounds)

Raw data	Compound 1				Compound p
Product 1					
...					
Product n					

$x_{ij}$   
for  $i = 1, \dots, n$  &  $j = 1, \dots, p$

### Step 1: Log-transformation

Log-transformation of the raw data +1 to avoid that values inferior to 1 get an overestimated weight

Log data	Compound 1				Compound p
Product 1					
...					
Product n					

$$y_{ij} = \ln(x_{ij} + 1)$$

for  $i = 1, \dots, n$  &  $j = 1, \dots, p$

Mean	$\mu(y_1)$				$\mu(y_p)$
SD	$\sigma(y_1)$				$\sigma(y_p)$

### Step 2: Normalization

Standardization of the log-transformed data in order to put all compounds on a similar scale of values

Normalized data	Compound 1				Compound p
Product 1					
...					
Product n					

$$z_{ij} = \frac{y_{ij} - \mu(y_j)}{\sigma(y_j)}$$

for  $i = 1, \dots, n$  &  $j = 1, \dots, p$

### Step 3: Pseudo-concentration

Aroma pseudo-concentration estimate to compare the relative abundance of volatile compounds between products

Normalized data	Compound 1				Compound p
Product 1					
...					
Product n					

$$z_{ij}$$

for  $i = 1, \dots, n$  &  $j = 1, \dots, p$

Pseudo-concentration
$v_1 = \text{mean}(z_{1j})$
...
$v_n = \text{mean}(z_{nj})$

### Step 4: Pseudo-composition

Aroma pseudo-composition estimate to express aroma compounds in terms of relative abundance with respect to the overall load of aroma molecules

Pseudo-composition	Compound 1				Compound p
Product 1					
...					
Product n					

$$w_{ij} = z_{ij} - v_i$$

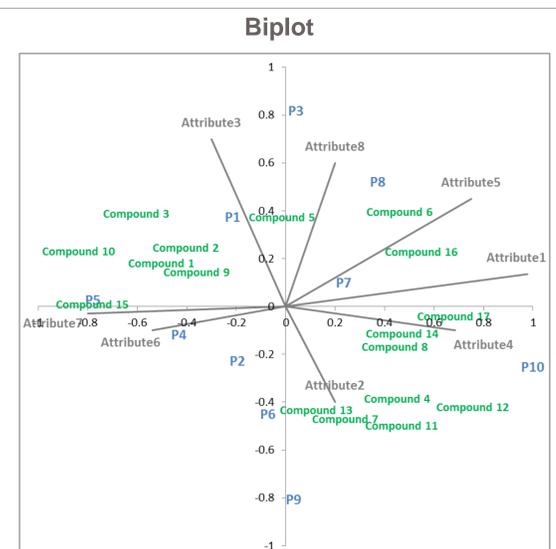
for  $i = 1, \dots, n$  &  $j = 1, \dots, p$

## 3. Procedure to relate transformed data and sensory

The sensory profiles are mapped on a biplot by using the Principal Component Analysis (PCA) with the sensory data. The aroma pseudo-compounds are then overlapped on the map as supplementary variables.

- Products (in blue) are mapped on the biplot based on their sensory profile
- Sensory attributes (in grey) are represented by arrows on the map
- Pseudo-compounds (in green) are overlapped on the map as supplementary variables

This approach has the advantage to give more importance to the sensory space enabling to identify which aroma pseudo-compounds correlate the best with key sensory dimensions.



## References

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