# Hyperspectral imaging for compositional information of sandwiches

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## Goal

Contribute to determining food intake:

- Non-destructive determination of macronutrients in a real-life setting
- (Complex) sandwiches

## $\rightarrow$ Hyperspectral imaging





# Hyperspectral imaging (HSI)



Yields a (sw)NIR spectrum for each pixel of ~0.1 mm<sup>2</sup>



## Information from NIR



Adapted from:

https://doi.org/10.1016/j.jfoodeng.2020.109954



- Diverse foods no calibration for all matrices!
- Complex sandwiches: unknown number and type of foods
- Uneven surfaces (reflectance!) + shadows
- Penetration depth NIR: couple of mm
  - Underlying food might influence the spectrum
  - Underlying food might be hidden





- IMEC Snapscan SWIR C100U-0007
- Range 1116-1670 nm (108 bands)
- 640x512 pixels
- 15 seconds acquisition time





## Raw data

640×512×108 (w×h×λ)
 = 135MB ENVI file

- Example: bread with butter and cheese as RGB
  - r = slice @ 1121 nm
  - g = slice @ 1374 nm
  - b = slice @ 1660 nm







wavelength (nm)

# Multivariate modelling

- No suitable univariate markers
- Relatively few (independent) observations
- Need for a robust, interpretable algorithm

Fat

Sugars

- 13 selected food types from actual consumption profiles
- Covering range of interest
- Total 104 scans selected
- **PLS**-modelling





7

0

55

## Multivariate modelling – fitting statistics



Nutrient	RMSEP	R <sup>2</sup>
Fat	5.0	0.95
Protein	6.6	0.55
Moisture	8.3	0.86
Carbohydrates	10.9	0.73
Sugars	8.1	0.72

- Room for improvement...
- Is this sufficient for our goal?

## Predictions

#### Processing times: (on Intel<sup>®</sup> Xeon<sup>®</sup> E-2176M 6Core, 2.70GHz, Win-10, R 3.6)

Step	time (s)
Raw file loading (135 MB)	35
Data preprocessing (mean-center + spectral stdev-scaling)	3
Predicting composition (327,680 spectra x 5 nutrients)	0.3
(Re)organising and saving results (8 MB)	12
Total (per file)	50.3

Yields fat-, protein-, moisture-, carbs-, sugar- concentrations... as function of x- and y- position of the image...



## Clustering

- Ideally: based on colour (d)RGB image analysis
- Number of foodstuffs visible unknown
- Many tools available (kMeans, HC, metaheuristic,...)

- For now: clustering based on detecting groups in each nutrient's distribution
- Background removal





Observed protein distribution





## **Results** (image processing for bread/butter/cheese example)

### 1. Raw result

3 selected wavelengths as RGB-image





## **Results** (image processing for bread/butter/cheese example)

- Raw result
  3 selected wavelengths
  as RGB-image
- 2. False-colour image

 $R \propto fat\%$ G  $\propto$  protein%

 $B \propto moisture\%$ 





## Results (image processing for bread/butter/cheese example)

- Raw result
  3 selected wavelengths
  as RGB-image
- 2. False-colour image  $R \propto fat\%$   $G \propto protein\%$  $B \propto moisture\%$
- 3. Nutrient-density based clustering random colour per cluster





## Validation set – reference vs medians from auto-segmented HSI scans





- A one-step macronutrient quantification from HSI data is feasible Suitable to discriminate foods & tentative identification
- Subsequent matrix-specific models might increase performance
  Prediction of butter fat% (<±10%) and layer thickness (<±60mg/cm<sup>2</sup>)
- Data quantity is a potential limitation for practical use
- Needs additional data for food intake estimation...



# Thank you for your attention!

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