Nutrition and non-communicable disease risk factors

A talk by Simon Müller for the lecture Introduction to Data Science

Introduction

- Data & Methods
- Data Sources
- Principal Component Analysis
- Clustering
- Results
- Scatter plots
- K-Means clustering
- Correlations
- Conclusion

Introduction and Goals

- Combination of world nutrition, health indicator and GNI per capita datasets
- Apply Principal Component Analysis and clustering to find interesting patterns in the data
- Quantify the correlation between different food groups, health indicators and income





Nutrition database:

- FAO.org
- Food Balance Sheets (1961 2013)
- Item of interest: Food supply (kcal/capita/day)

The data is aggregated by food groups.



Health indicator database:

- WHO.int
- Global Health Observatory data
- Non-communicable diseases
- Common risk factors include unhealthy diet, overweight, raised blood pressure, blood sugar and cholesterol



Given percentage of population that is affected by:

- Raised fasting blood glucose (≥ 7 mmol/L or on medication)
- Raised total cholesterol (≥ 5 mmol/L)
- Raised blood pressure (SBP ≥ 140 or DBP ≥ 90), agestandardized
- Overweight (body mass index ≥ 25), age-standardized

Also:

• HALE: Healthy Life Expectancy Index



GNI per capita to estimate income:

- data.worldbank.org
- Latest (2015) values in USD
- This dataset is only used for the correlation part of this talk

Principal Component Analysis

Reminders:

- PCA emphasizes variation and can be used to reduce the dimensionality of a dataset
- It involves finding linear combinations of a set of variables that has maximum variance and removing its effect

In my case:

- Up to 20 dimensional dataset
- PCA allows to visualise the data in two or three dimensions

Clustering

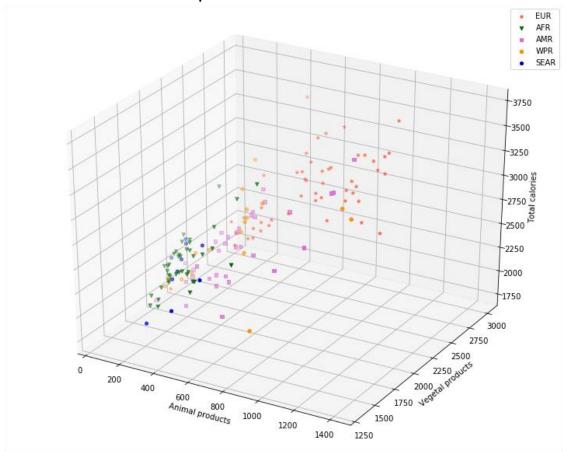
- Flat clustering with the K-Means algorithm
- Metric used: Distance between points
- Parameter: Number of clusters (n = 6)
- Six regions: Africa, Americas, Europe, Eastern Mediterranean, South-East Asia, Western Pacific

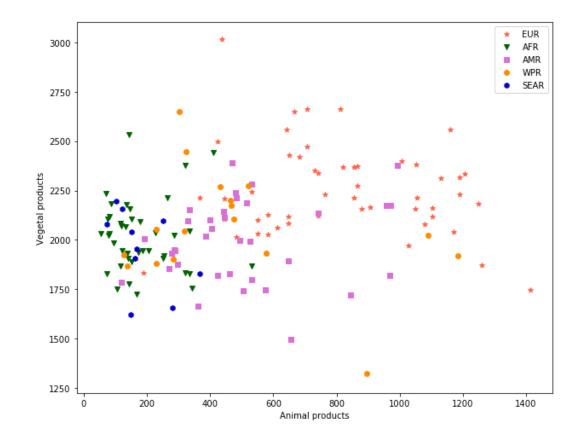
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Scatter plots

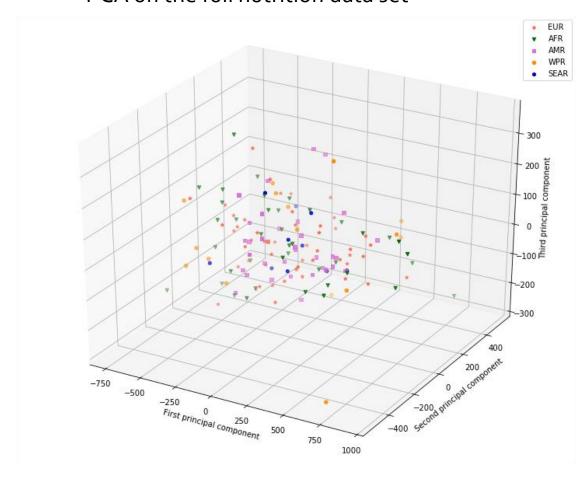
Let's start with something we can plot without manipulation

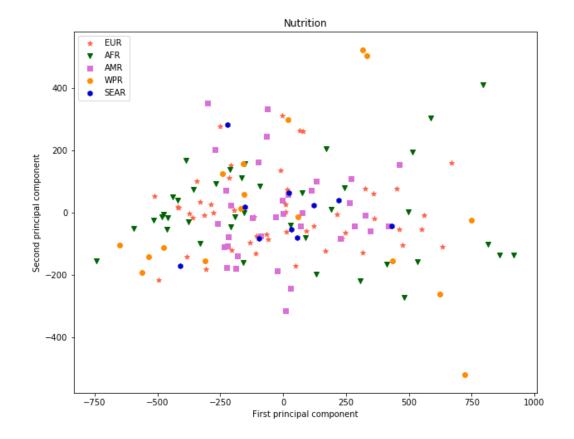




Scatter plots

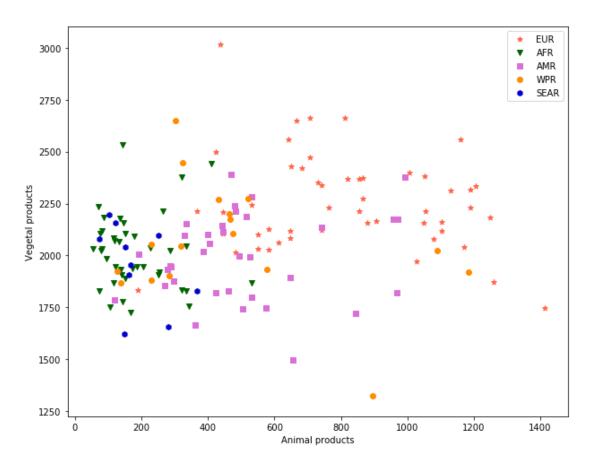
• PCA on the full nutrition data set

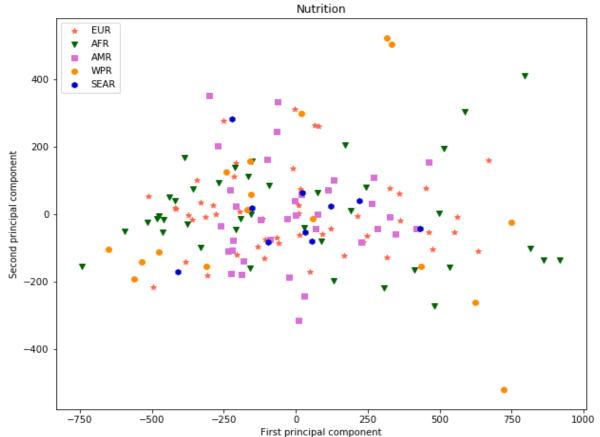




Scatter plots

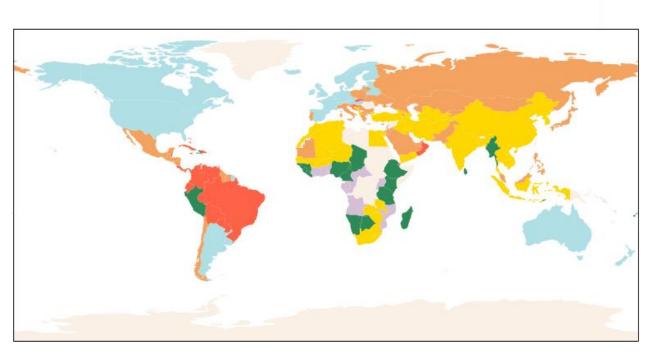
Comparison

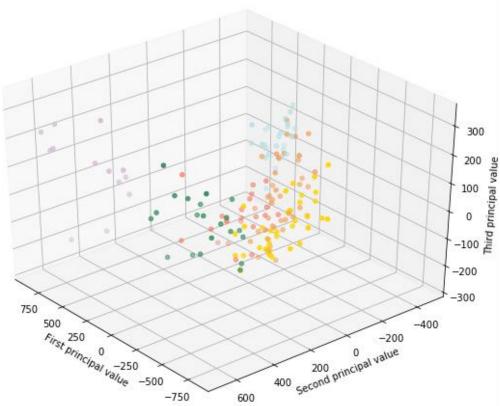




K-Means clustering

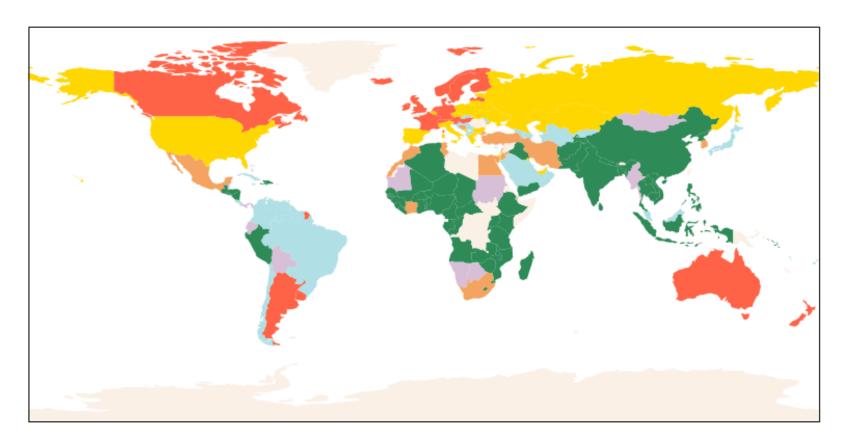
- Initiate K-Means clustering with n = 6 clusters (number of world regions)
- Use PCA on the full data to reduce it to three dimensions





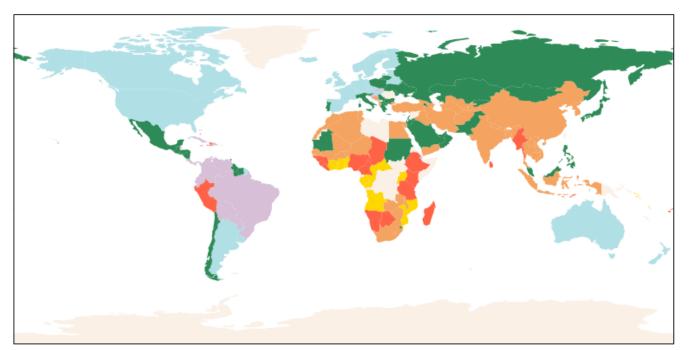
K-Means clustering

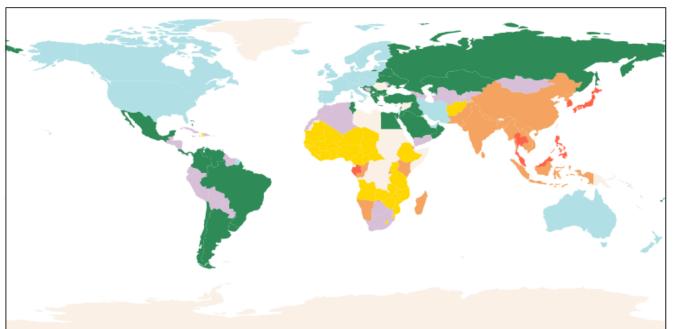
• Clustering on the coarsely aggregated nutrition data



K-Means clustering

- We can use clustering on n-dimensional data and then use the world map to visualise the result
- Next slide: Clustering based on only the nutrition data versus the health indicator data





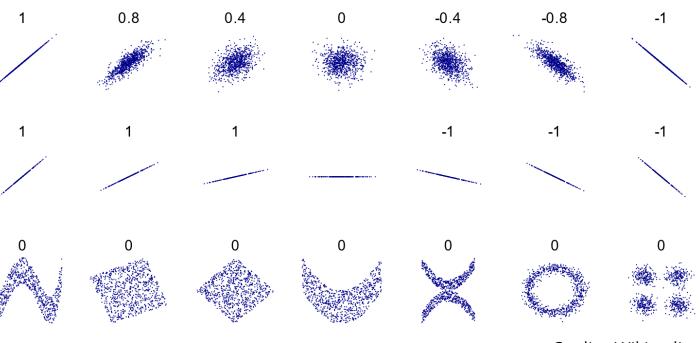
Correlations

We can use the Pearson Correlation Coefficient (PCC) to try and quantify correlations.

Generally used interpretation:

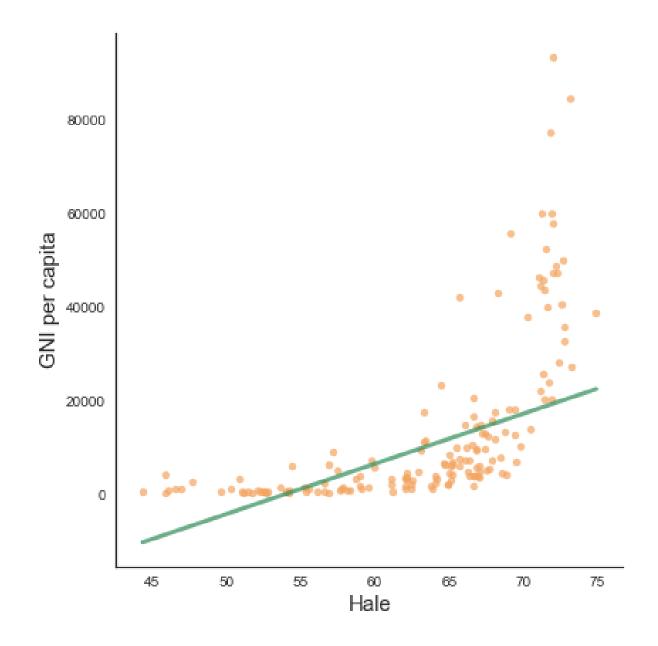
- 0.1 < r < 0.3: Small
- o.3 < r < o.5 : Moderate
- r > 0.5: Strong

Caution: Only valid when considering linear relations

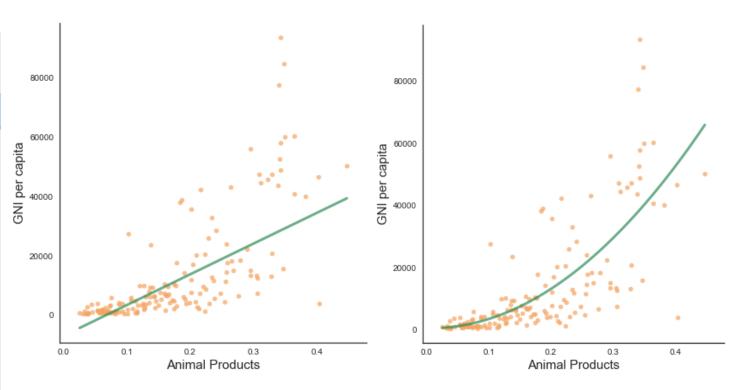


Credits: Wikipedia

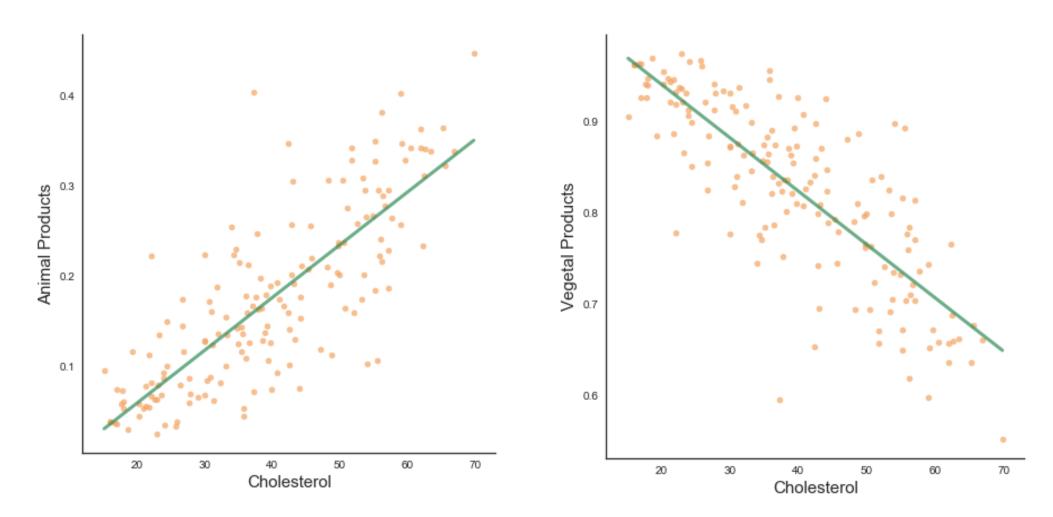
	GNI per capita
Blood pressure	0.05
Blood glucose	-0.09
Cholesterol	0.75
BMI	0.48
HALE	0.63



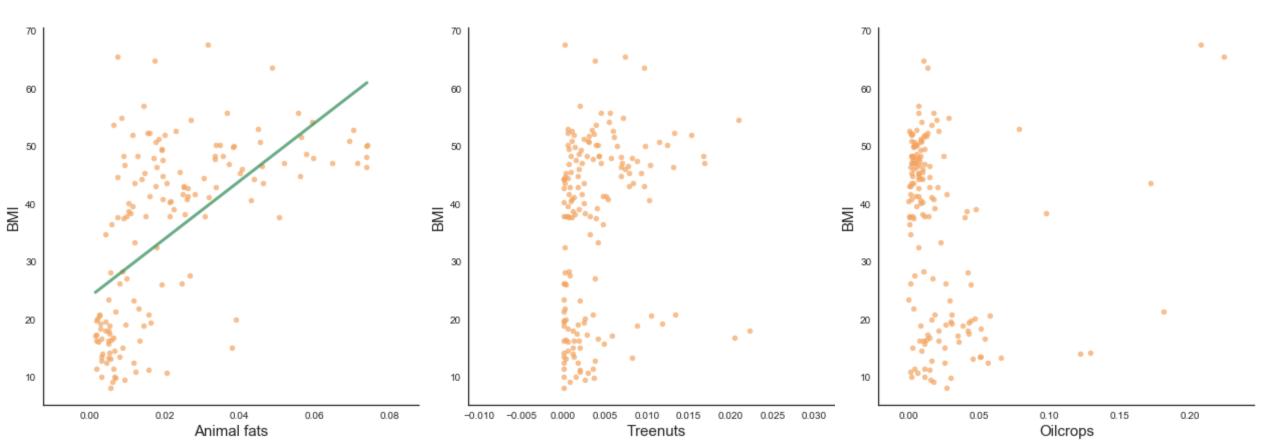
	GNI per capita
Total calories	0.69
Animal products	0.78
Vegetal products	0.22
Animal fats	0.63
Meat	0.56
Offals	0.35
Dairy products	0.61
Eggs	0.62
Fish & Seafood	0.19
Vegetables	0.19
Fruits	-0.04
Cereals	-0.52
Pulses	-0.36
Starchy Roots	-0.26
Sugar & Sweeteners	0.39
Treenuts	0.28
Oilcrops	-0.20
Vegetable oils	0.34



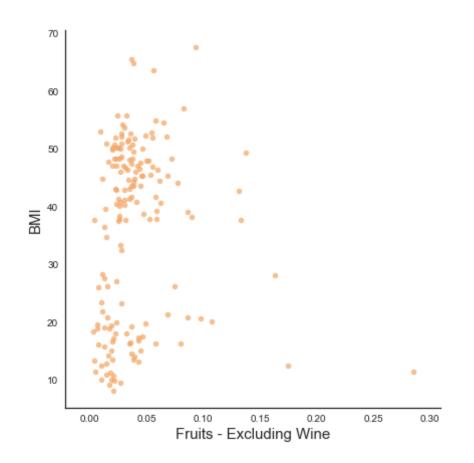
	Animal Products	Vegetal Products
Blood pressure	0.30	-0.30
Blood glucose	0.12	-0.12
Cholesterol	0.80	-0.80
BMI	0.69	-0.69
HALE	0.72	-0.72

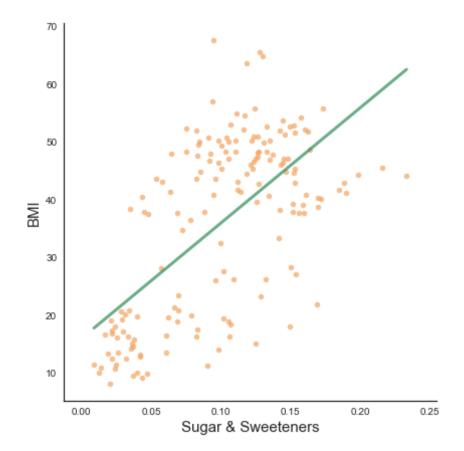


	Animal			Dairy		Fish &					Starchy	Sugar &			Vegetable
	fats	Meat	Offals	products	Eggs	Seafood	Vegetables	Fruits	Cereals	Pulses	Roots	Sweeteners	Treenuts	Oilcrops	oils
Blood pressure	0.27	0.22	0.36	0.36	0.18	-0.24	0.11	-0.21	-0.12	-0.20	-0.05	-0.02	0.07	-0.30	0.19
Blood glucose	0.00	0.17	-0.05	0.04	0.07	0.17	0.18	-0.02	-0.04	-0.28	-0.25	0.34	-0.03	0.27	0.02
Cholesterol	0.67	0.65	0.44	0.65	0.79	0.18	0.36	-0.05	-0.51	-0.48	-0.38	0.53	0.26	-0.27	0.29
вмі	0.54	0.59	0.36	0.61	0.62	-0.01	0.37	0.05	-0.53	-0.42	-0.35	0.63	0.22	-0.11	0.33
HALE	0.56	0.56	0.30	0.61	0.76	0.21	0.42	0.00	-0.41	-0.40	-0.48	0.55	0.23	-0.22	0.22



	Animal fats	Meat	Offals	Dairy products		Fish & Seafood	Vegetables	Fruits	Cereals	Pulses	Starchy Roots	Sugar & Sweeteners	Treenuts	Oilcrops	Vegetable oils
Blood pressure	0.27	0.22	0.36	0.36	0.18	-0.24	0.11	-0.21	-0.12	-0.20	-0.05	-0.02	0.07	-0.30	0.19
Blood glucose	0.00	0.17	-0.05	0.04	0.07	0.17	0.18	-0.02	-0.04	-0.28	-0.25	0.34	-0.03	0.27	0.02
Cholesterol	0.67	0.65	0.44	0.65	0.79	0.18	0.36	-0.05	-0.51	-0.48	-0.38	0.53	0.26	-0.27	0.29
вмі	0.54	0.59	0.36	0.61	0.62	-0.01	0.37	0.05	-0.53	-0.42	-0.35	0.63	0.22	-0.11	0.33
HALE	0.56	0.56	0.30	0.61	0.76	0.21	0.42	0.00	-0.41	-0.40	-0.48	0.55	0.23	-0.22	0.22





Conclusion

- PCA reduced data managed to preserve clustering of original data.
- Clustering when looking at nutrition and health indicator separately is consistent.
- Be aware of deadly vegetables: Apply caution when using correlation coefficients.
- Nonetheless: Interesting trends may be found
- Let the data tell its story don't get too caught up in your own narrative.



Thank you for listening!

Sources:

- Food balance sheets: fao.org/faostat
- Global Health Observatory: who.int/gho/database
- GNI per capita: data.worldbank.org

Python libraries used: numpy, pylab, pandas, cartopy, seaborn, scipy, sklearn

Interesting visualisation to toy around with:
http://www.nationalgeographic.com/what-the-world-eats/

