

## Chapter 3

# Descriptive multivariate analysis

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Printer-friendly slides to the book

A General Introduction to Data Analytics

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# Descriptive Multivariate Analysis

Friend	Max temp	Weight	Height	Years	Gender	Company
Andrew	25	77	175	10	M	Good
Bernhard	31	110	195	12	M	Good
Carolina	15	70	172	2	F	Bad
Dennis	20	85	180	16	M	Good
Eve	10	65	168	0	F	Bad
Fred	12	75	173	6	M	Good
Gwyneth	16	75	180	3	F	Bad
Hayden	26	63	165	2	F	Bad
Irene	15	55	158	5	F	Bad
James	21	66	163	14	M	Good
Kevin	30	95	190	1	M	Bad
Lea	13	72	172	11	F	Good
Marcus	8	83	185	3	F	Bad
Nigel	12	115	192	15	M	Good



# Summary

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- Multivariate frequencies
- Multivariate data visualization
- Multivariate statistics
  - Location multivariate statistics
  - Dispersion multivariate statistics
- Final remarks



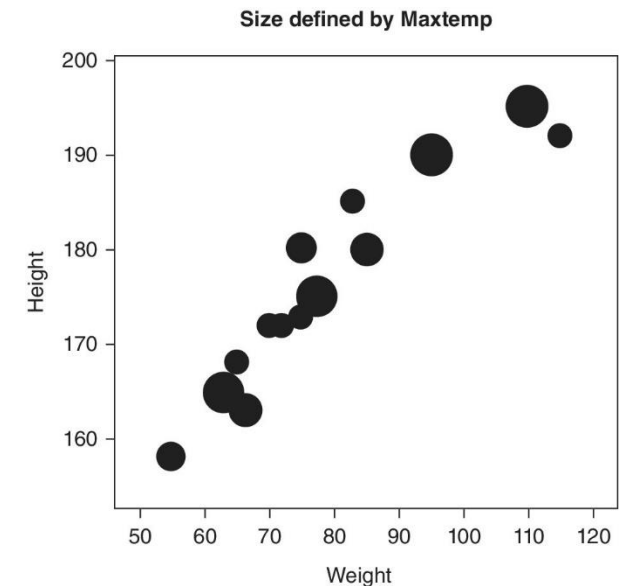
# Multivariate frequencies

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- The multivariate frequency values can be computed independently for each attribute
  - Thus, we can represent the frequency values for each attribute presenting them in a matrix like structure

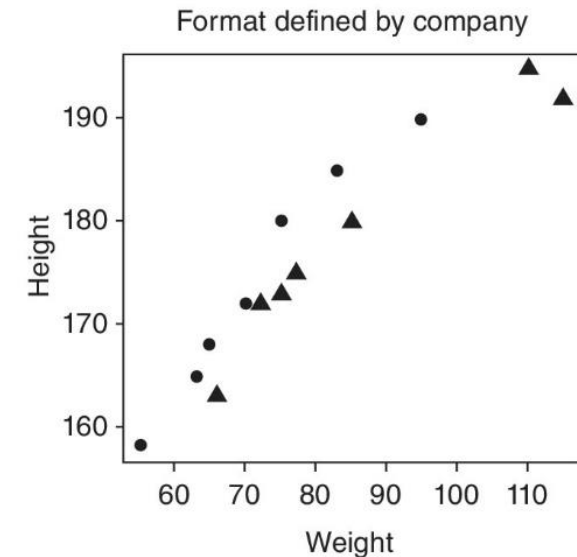
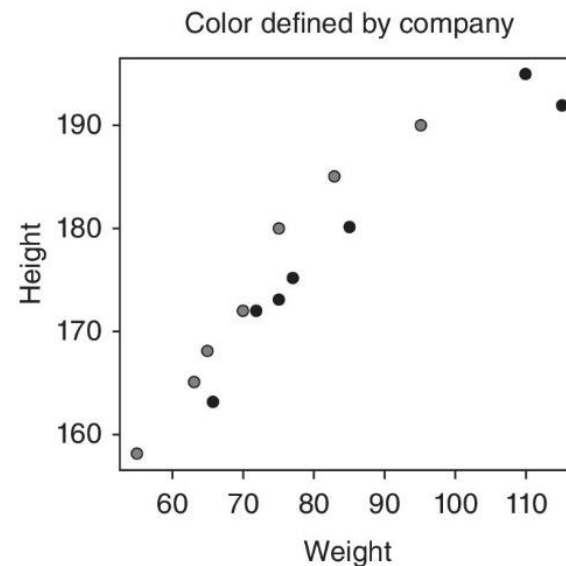
# Multivariate data visualization

- When the multivariate data has **three attributes**, at least two of them quantitative, the data can still be visualized by a bivariate plot
  - This is done by associating the scale types of the values of the third attribute to how each data object is represented in the plot



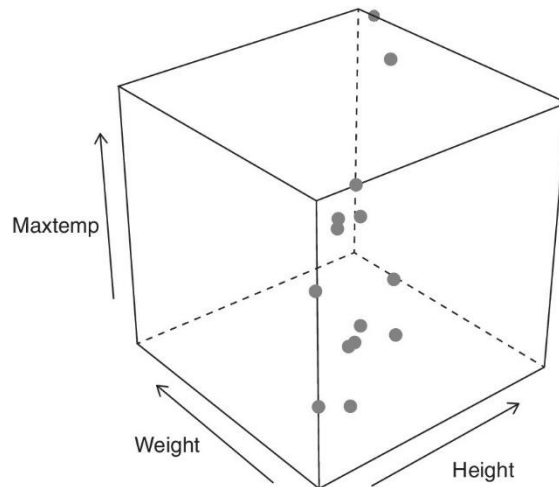
# Multivariate data visualization

- If the third attribute is qualitative, its value can be represented in the plot by either the color or by the shape of the object in the plot
  - The number of colors or shapes will be the number of values the attribute can assume

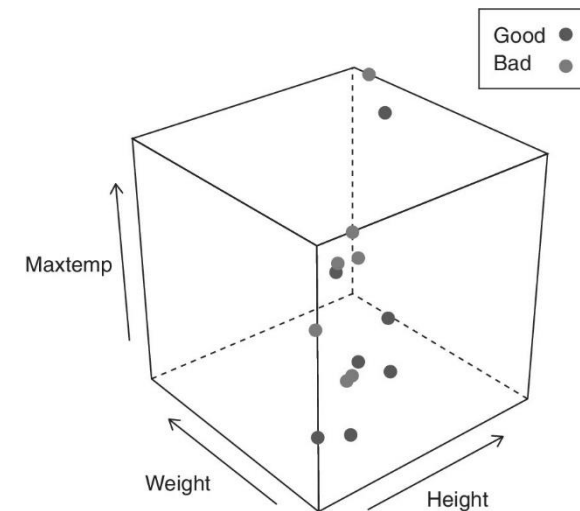


# Multivariate data visualization

- Another approach to represent three attributes is to use a 3-dimensional plot



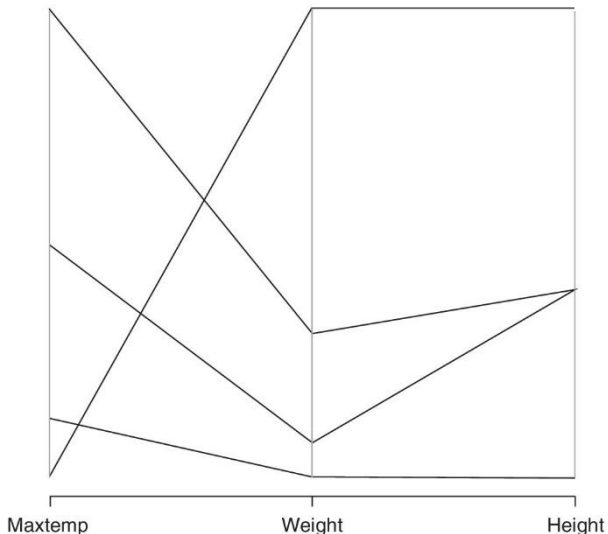
- A fourth attribute can be represented the same way a third attribute was represented in a bi-dimensional space



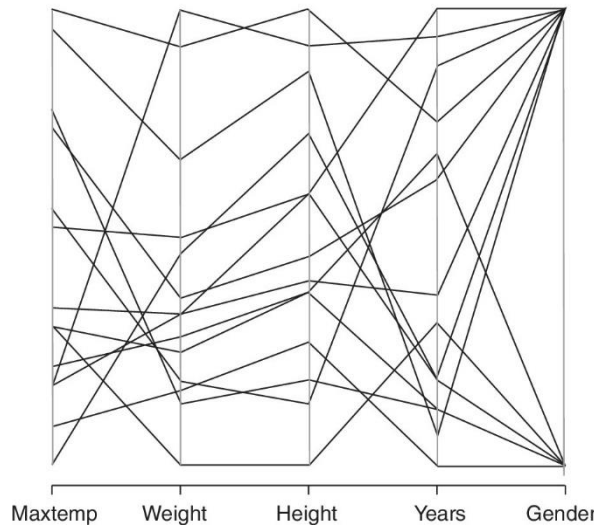
# Multivariate data visualization

- Methods to visualize more than four attributes
  - One of the most popular of them is the **parallel coordinates**, also known as **profile plots**

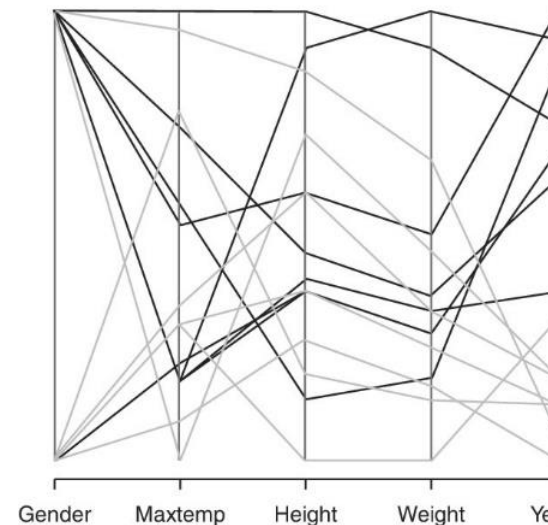
Simple Parallel coordinates



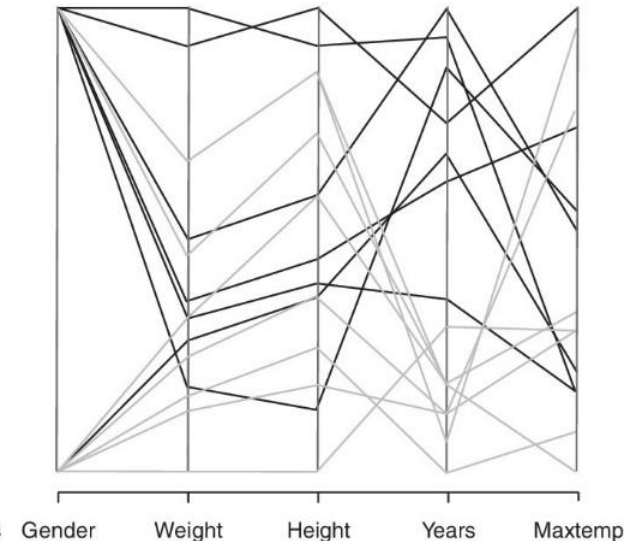
More complex parallel coordinates



Parallel coordinates with colors



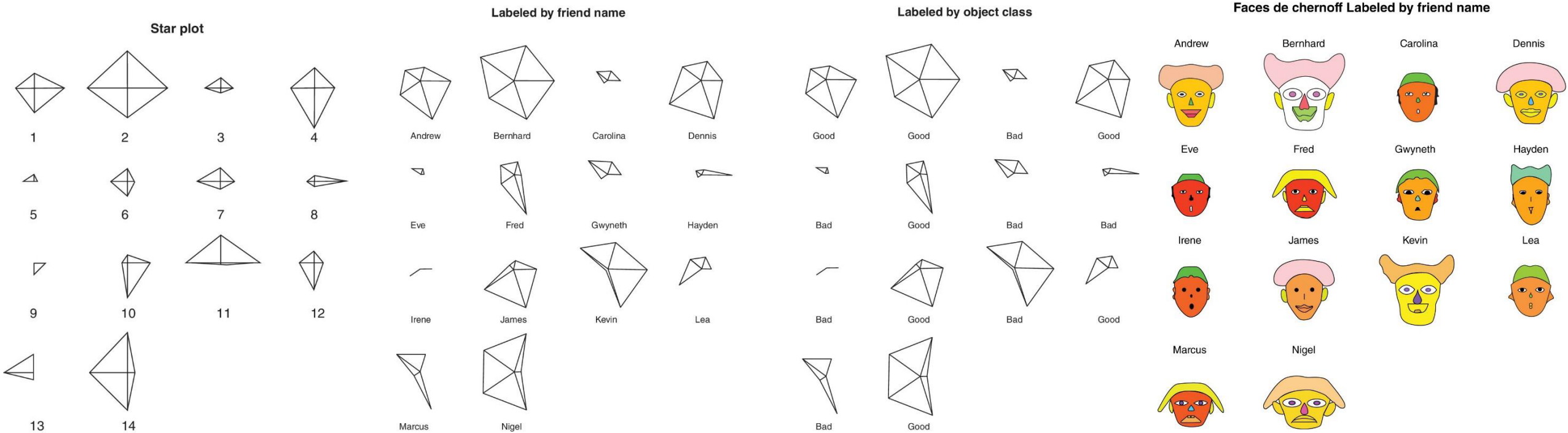
Parallel coordinates re-ordering variables





# Multivariate data visualization

- Methods to visualize more than four attributes
  - **Star plots** and **Chernoff faces**





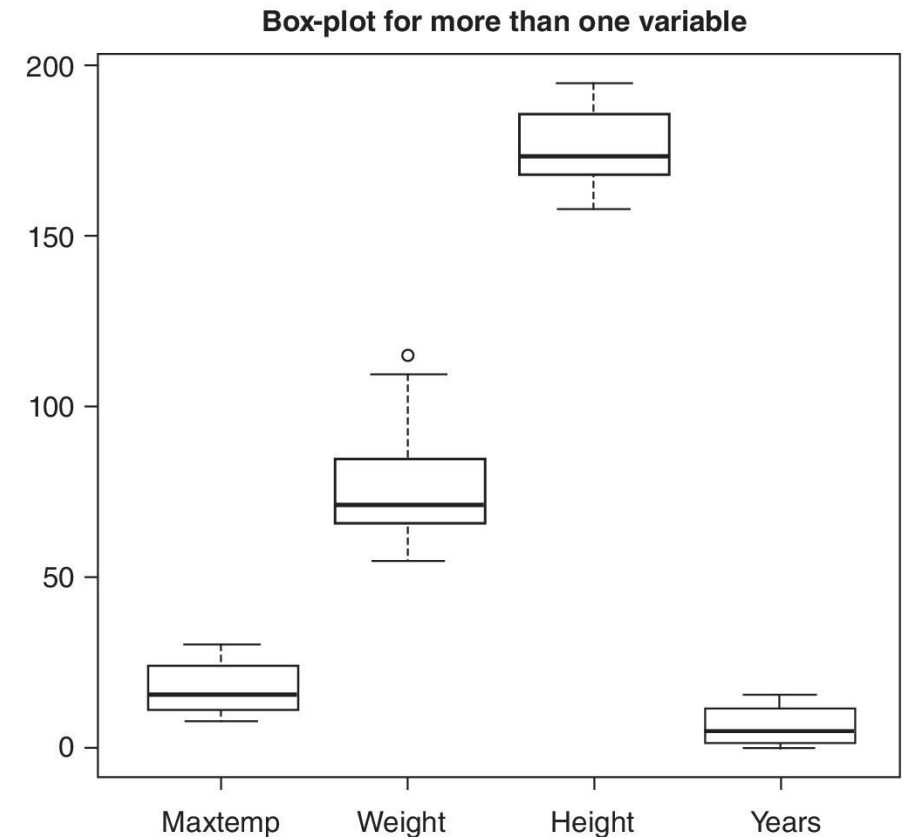
# Location multivariate statistics

- To measure the location statistics of several attributes we just measure the location value for each attribute
  - Thus, we can represent the frequency values for each attribute presenting them in a matrix like structure

Location statistics	Max temp	Weight	Height	Years
min	8.00	55.00	158.00	0.00
max	31.00	115.00	195.00	16.00
average	18.14	79.00	176.29	7.14
mode	15.00	75.00	172.00	2.00
1 <sup>st</sup> quartile	12.25	67.00	169.00	2.25
Median or 2 <sup>nd</sup> quartile	15.50	75.00	174.00	5.50
3 <sup>rd</sup> quartile	24.00	84.50	183.75	11.75

# Location multivariate statistics

- To measure the location statistics of several attributes we just measure the location value for each attribute
  - If the number of attributes is not too large, a set of **box-plots**, one for each attribute, can be used





# Dispersion multivariate statistics

- The extraction of some of the dispersion values for multivariate statistics, like amplitude, interquartile range, mean absolute deviation and standard deviation, can be also independently performed for each attribute

Dispersion statistics	Max temp	Weight	Height	Years
Amplitude	23.00	60.00	37.00	16.00
Interquartile range	11.75	17.50	14.75	9.50
$\overline{MAD}$	7.41	14.09	11.12	6.67
s	7.45	17.38	11.25	5.66



# Dispersion multivariate statistics

- The relation between two attributes is evaluated using covariance or correlation measures
  - The main diagonal of the **covariance matrix** shows the variance of each attribute
  - The matrices are symmetric: the values above the main diagonal are the same as the value below the main diagonal

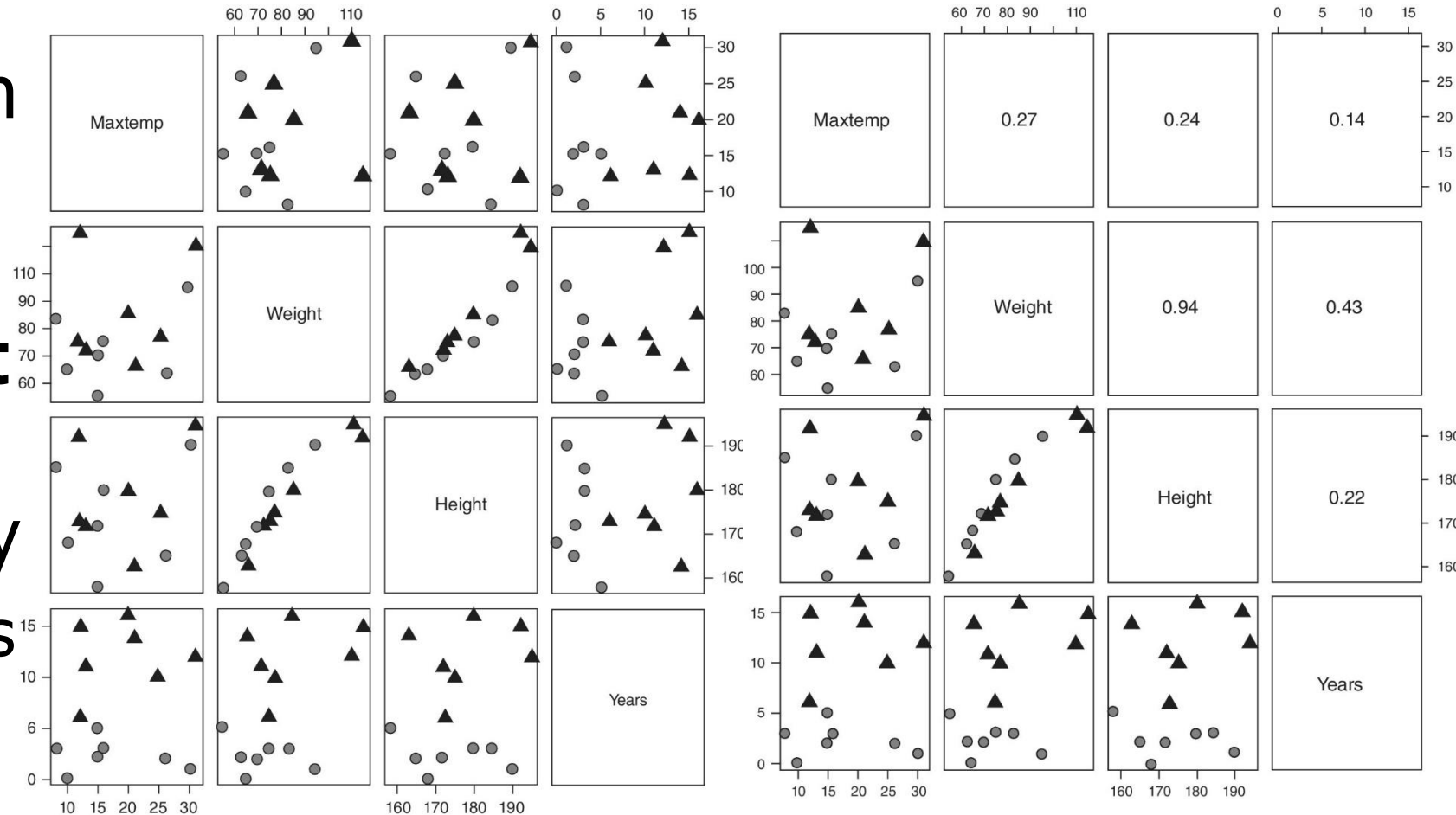
Covariance					Pearson correlation				
Max temp	Weight	Height	Years		Max temp	Weight	Height	Years	
Max temp	55.52	34.46	20.19	5.82	Max temp	1.00	0.27	0.24	0.14
Weight	34.46	302.15	184.62	42.39	Weight	0.27	1.00	0.94	0.43
Height	20.19	184.62	126.53	14.03	Height	0.24	0.94	1.00	0.22
Years	5.82	42.39	14.03	31.98	Years	0.14	0.43	0.22	1.00

# Dispersion multivariate statistics

■ The relation between several attributes can be evaluated using a **scatter plot matrix** aka

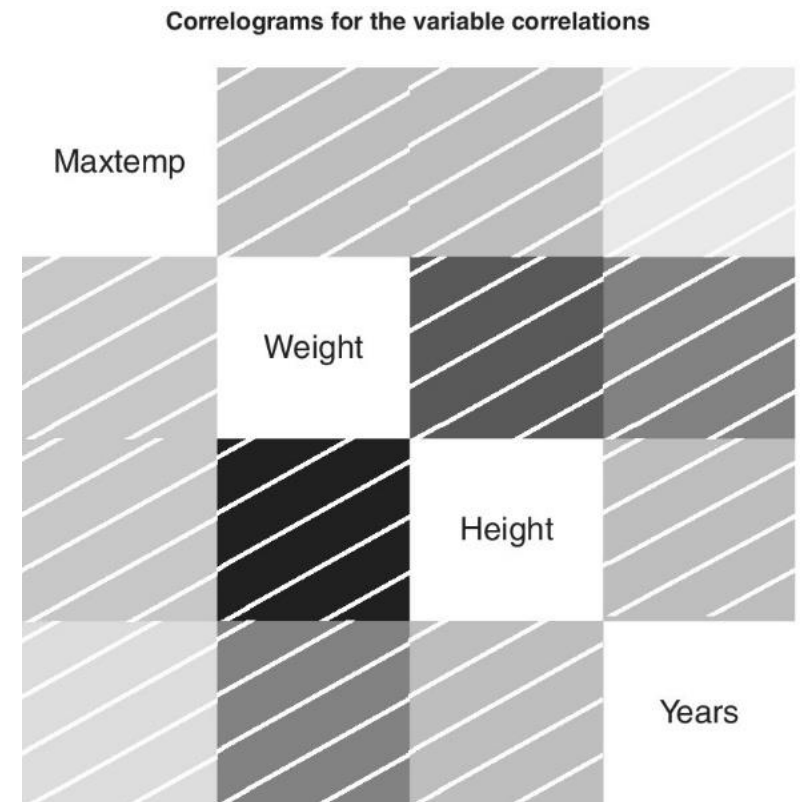
**Draftsman's display**

- The shape represents the class label



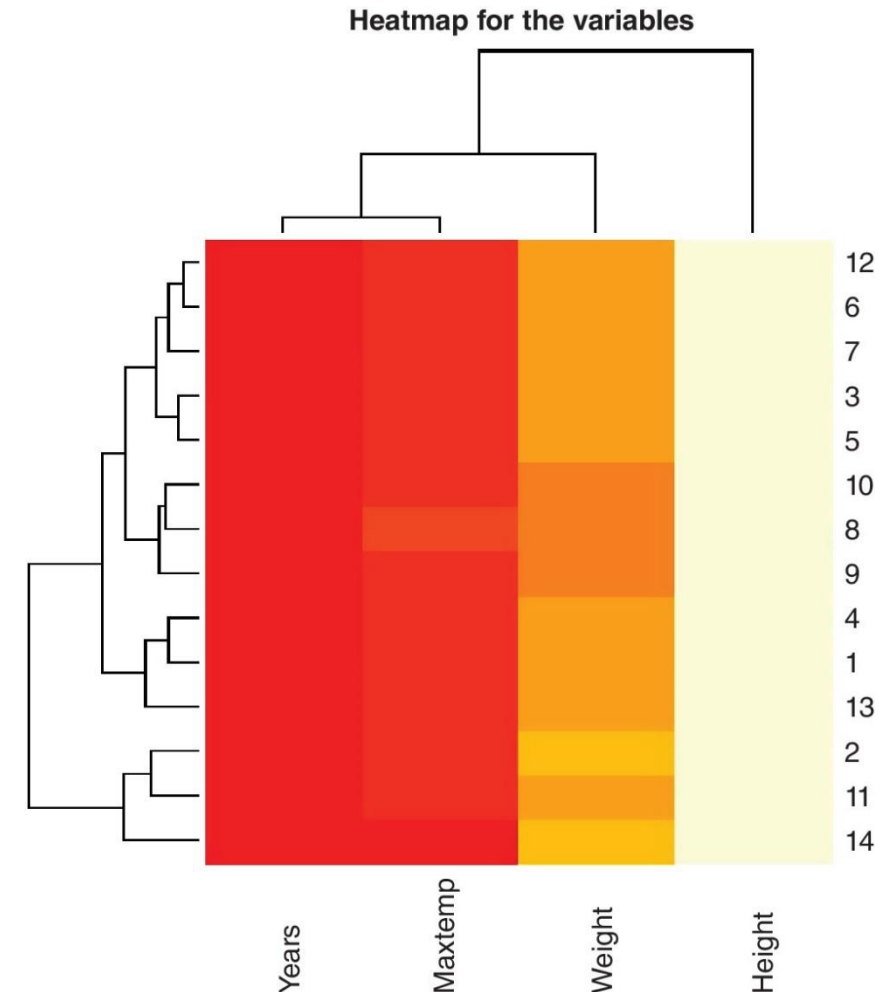
# Dispersion multivariate statistics

- We can use a simpler plot to illustrate a summary of the information present in the scatter plot matrix
- For such, we can plot the linear correlation matrix using a **correlogram**
  - In this figure, the darker the square associated with two attributes, the more correlated they are



# Dispersion multivariate statistics

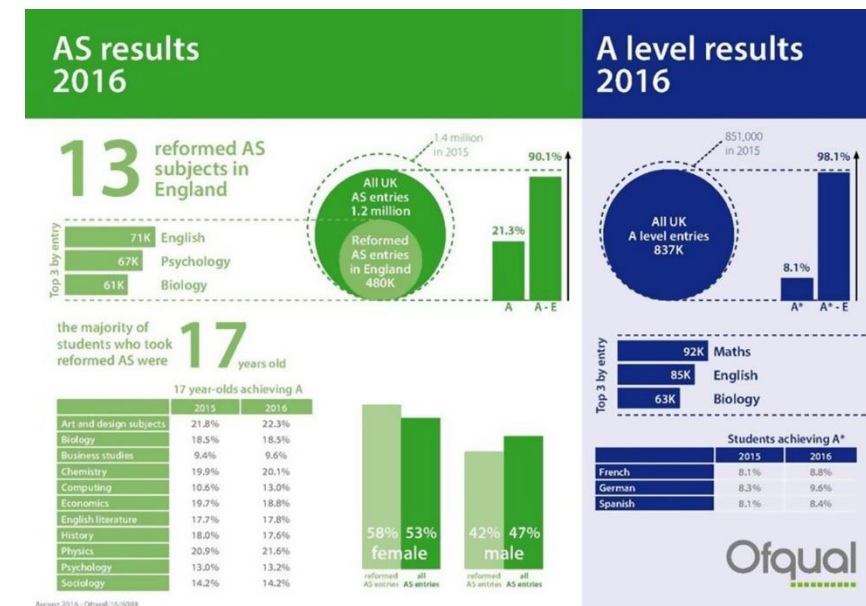
- Another common plot for multivariate data is the **heatmap**, which can show how close the objects of a data set are considering the predictive attribute values
  - Each column (attribute) has boxes of a particular color. Although the boxes in each column have the same color, they can have different color tones
  - The tone of the color in a box is defined by the value of its corresponding predictive attribute in the object associated with its row





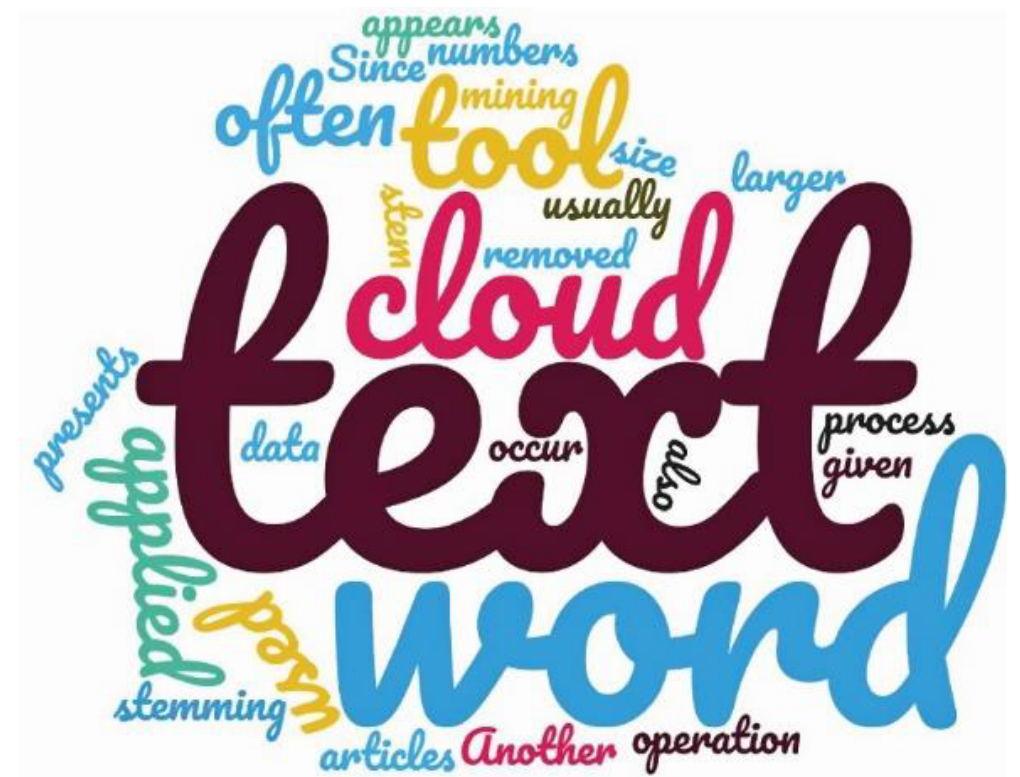
# Dispersion multivariate statistics

- Currently, several studies graphically highlight important facts by using **infographics**
  - While data visualization is objective, automatically produced and can be applied to several data sets
  - Infographics are subjective, manually produced and customized for a particular data set



Infographic illustrating AS and A level results in England in 2016 (Contains public sector information licensed under the Open Government License v3.0.)

- The higher the frequency of a word in the text, the larger its size in the word cloud





# Dispersion multivariate statistics

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- Since articles and prepositions occur very often in a text, and numbers are not text, they are usually removed before the word cloud tool is applied to a text
- Another text process operation, stemming, which substitutes a word in a text by its stem, is also applied to the text before the word cloud tool is used



# Final remarks

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- Descriptive multivariate analysis is more complex as the number of attributes increases
- It extends naturally from univariate and bivariate descriptive statistics
- The area of multivariate data visualization is an active research area



# Questions?

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