

Statistics for data science

Tuesday 22nd November, 2016

- 1 Introduction - Statistics, what is it good for?
- 2 Data
- 3 Relations between data elements
- 4 Conclusion

Statistics, what is it good for?

A data science perspective

Data → Knowledge

Statistics, what is it good for?

A data science perspective

Data → **Knowledge**

Knowledge is a statistical entity

Statistics, what is it good for?

A data science perspective

Data → **Knowledge**

Knowledge is a statistical entity

- ▶ there is randomness in the data
 - a telephonic survey

Statistics, what is it good for?

A data science perspective

Data → **Knowledge**

Knowledge is a statistical entity

- ▶ there is randomness in the data
 - a telephonic survey
- ▶ not enough data to achieve certainty
 - choosing the best restaurant when there are very few reviews

Statistics, what is it good for?

A data science perspective

Data → Knowledge

Knowledge is a statistical entity

- ▶ there is randomness in the data
 - a telephonic survey
- ▶ not enough data to achieve certainty
 - choosing the best restaurant when there are very few reviews
- ▶ the knowledge required is of a statistical nature
 - What are my odds when betting on a soccer match

Statistics, what is it good for?



Statistics is an important component of a data scientist's toolbox

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Statistics is an important component of a data scientist's toolbox

- ▶ "raw" statistics - "getting a feel of the data"
data exploration

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- ▶ "raw" statistics - "getting a feel of the data"
data exploration
- ▶ many data science tools are of a statistical nature

Statistics, what is it good for?

Statistics is an important component of a data scientist's toolbox

- ▶ "raw" statistics - "getting a feel of the data"
data exploration
- ▶ many data science tools are of a statistical nature
- ▶ a knowledge of statistics is essential
 - choose the right tool
 - correctly interpret results

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Structured Vs. Unstructured data

Structured Data

data is organized in well defined fields ...

Structured Vs. Unstructured data

Structured Data

data easily fits in a relational database - tables

Examples?

Structured Vs. Unstructured data

Structured Data

Name	Height (H)	Weight (W)
Prince Humperdinck	6'	166lb
Inigo Montoya	5'3"	142lb
Princess Buttercup	5'4"	136lb

⋮

Structured Vs. Unstructured data

Unstructured Data

data is not organized in well defined fields...

Structured Vs. Unstructured data

Unstructured Data

data does not easily fit in a relational database - tables
what's easily? examples?

Structured Vs. Unstructured data

Israel^{tëch}
challenge

Unstructured Data



Structured Vs. Unstructured data



Semi-structured Data

Data is partially structured and partially unstructured
examples?

Structured Vs. Unstructured data

Semi-structured Data

The image shows a screenshot of an email composition window, illustrating semi-structured data. The window has a light blue header bar with buttons for Send, Attach, Save Draft, Spelling, and Cancel. Below the header are three text input fields: To:, Cc:, and Subject:. To the right of the To: field is a 'Show BCC' link, and to the right of the Subject: field is a 'Plain Text' link. Below the input fields is a rich text editor toolbar with options for font face (Arial), font size (12), bold (B), italic (I), underline (U), text color, background color, emoji, link, and unlink. The main body of the email is a large, empty white text area.

Data Types

- ▶ Numerical - discrete/continuous

Data Types

- ▶ Numerical - discrete/continuous
 - Height
 - Temperature
 - Number of children

Data Types

- ▶ Numerical - discrete/continuous
 - Height
 - Temperature
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- ▶ Categorical

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Data Types

- ▶ Numerical - discrete/continuous
 - Height
 - Temperature
 - Number of children
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 - Gender
 - Nationality
- ▶ Ordinal
 - 1st, 2nd, 3rd...
 - low, medium, high

Data Types

- ▶ Arbitrary (string)
 - natural language - quotes, reviews, phrases
 - hashes

Data Types

- ▶ Arbitrary (string)
 - natural language - quotes, reviews, phrases
 - hashes
- ▶ Time series
 - Navigation data (Waze)
 - Stock market data

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A "practical" example

Physical measurements of N test subjects

Name	Height (H)	Weight (W)
Prince Humperdinck	6'	166lb
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Princess Buttercup	5'4"	136lb

⋮

A "practical" example

Physical measurements of N test subjects

- ▶ What can you say about height?
- ▶ What can you say about weight?

A "practical" example

Physical measurements of N test subjects

- ▶ mean

$$\mu_X$$

$$E[X] = \sum_x x \cdot Pr(x)$$

$$E[X] = \int_x x \cdot f(x) dx$$

- ▶ variance

$$\sigma_X^2$$

$$V(X) = E\left[(X - E[X])^2\right]$$

A "practical" example

Physical measurements of N test subjects

- ▶ What can you say about height?
- ▶ What can you say about weight?
- ▶ How would you assess the connection between the two?

A "practical" example

Physical measurements of N test subjects

► Covariance

$$\text{cov}(X, Y) = E \left[(X - E[X]) (Y - E[Y]) \right]$$

A "practical" example

Physical measurements of N test subjects

- Covariance

$$\text{cov}(X, Y) = E \left[(X - E[X]) (Y - E[Y]) \right]$$

- Covariance of height and weight

$$\text{cov}(H, W) = 6.27$$

A "practical" example

Physical measurements of N test subjects

News from the citadel...

they've invented the metric system

A "practical" example

Physical measurements of N test subjects

News from the citadel...

Name	Height (H)	Weight (W)
Prince Humperdinck	183cm	75.3kg
Inigo Montoya	160cm	64.41kg
Princess Buttercup	162.5cm	61.69kg

⋮

What happens to the connection between height and weight?

A "practical" example

Physical measurements of N test subjects

- Covariance of height and weight

$$\text{cov}(H, W) = 87.1125$$

Discussion

- ▶ What's the problem with using covariance to assess the connection between height and weight?

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- ▶ We need a measure that is invariant to a change of units (scale)

Discussion

- ▶ What's the problem with using covariance to assess the connection between height and weight?
- ▶ We need a measure that is invariant to a change of units (scale)
- ▶ How about normalizing the covariance?

Pearson Correlation

$$\rho_{X,Y}$$

informally: normalized covariance

$$\begin{aligned} \text{corr}(X, Y) &= \frac{\text{cov}(X, Y)}{\sigma_X \cdot \sigma_Y} \\ &= \frac{E[(X - E[X])(Y - E[Y])]}{\sqrt{E[(X - E[X])^2]} \sqrt{E[(Y - E[Y])^2]}} \end{aligned}$$

Pearson Correlation

$$\rho_{X,Y}$$

- ▶ $-1 \leq \rho_{X,Y} \leq 1$
- ▶ high direct correlation $\rho_{X,Y} \approx 1$
- ▶ high inverse correlation $\rho_{X,Y} \approx -1$

Pearson Correlation

$$\rho_{X,Y}$$

Examples

▶ direct correlation

- X number of days it *rains* in the morning
- Y number of days sidewalks are *wet* at noon

▶ inverse correlation

- X number of days it *rains* in the morning
- Y number of days sidewalks are *dry* at noon

A word about correlation and dependence

statistical dependence...

A word about correlation and dependence

Does a lack of correlation imply independence

$$\rho_{X,Y} = 0 \stackrel{?}{\Rightarrow} X \perp\!\!\!\perp Y$$

A word about correlation and dependence

Does a lack of correlation imply independence

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No , Correlation = Linear dependence

$$X, X^2$$

A word about correlation and dependence

Does a lack of correlation imply independence

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No , Correlation = Linear dependence

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A word about correlation and dependence

Does correlation imply dependence

$$\rho_{X,Y} \in \{-1, 1\} \stackrel{?}{\Rightarrow} X \not\perp Y$$

A word about correlation and dependence

Does correlation imply dependence

$$\rho_{X,Y} \in \{-1, 1\} \stackrel{?}{\Rightarrow} X \not\perp Y$$

Yes

A word about correlation and dependence

Does correlation imply dependence

$$\rho_{X,Y} \in \{-1, 1\} \stackrel{?}{\Rightarrow} X \not\perp Y$$

Yes , Linear dependence = a private case of dependence

A word about samples

- ▶ statistical operators - mean, variance,...

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$$E[X] = \sum_x x \cdot Pr(x)$$

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- ▶ how are these computed from a sample?

A word about samples

- ▶ statistical operators - mean, variance,...

$$E[X] = \sum_x x \cdot Pr(x)$$

- ▶ how are these computed from a sample?
- ▶ sample operators - sample mean, sample variance, ...
- ▶ all based on averaging

A word about samples

$$X = \{x_1, x_2, \dots, x_n\}$$

Sample Mean

$$\overline{X} = \frac{1}{n} \sum_{i=1}^n x_i$$

A word about samples

$$X = \{x_1, x_2, \dots, x_n\}$$

Sample Variance

$$\sigma_X^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \overline{X})^2$$

A word about samples

$$X = \{x_1, x_2, \dots, x_n\}$$

Sample covariance

$$\begin{aligned} cov(X, Y) &= \frac{1}{n} \sum_{i=1}^n (x_i - \bar{X}) (y_i - \bar{Y}) \\ &= \frac{1}{n} (X - \bar{X})^T (Y - \bar{Y}) \end{aligned}$$

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recap

- ▶ statistics is important
- ▶ knowledge is statistical
- ▶ data comes in diverse forms
- ▶ simple relations between data elements - covariance, correlation