

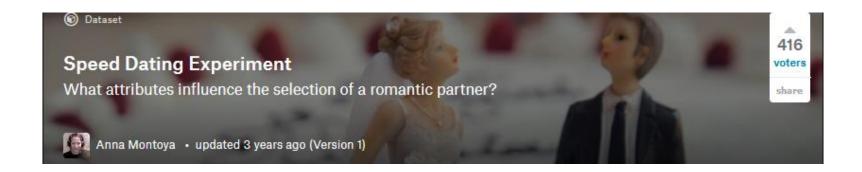
Fundamentals of Machine Learning Week 4: recap linear regression Introduction machine learning

Jonas Moons

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Intro





Feedback assignment #2



- Keep on documenting your analysis:
 - Use Markdown
 - Headings
 - Introductions
 - Intermediate steps / processing
 - Conclusions
- Don't use .fillna(0). Use .dropna() instead!



Research project with Al group

- Building on a project to predict fake news tweets
- Lecture and Q&A from Stefan Leijnen, professor of the research group.
 8 December 11:00-11:45
- Work individually but also as a team
- More (advanced) work on feature engineering & model building
- Some more supervision from me
- Present your work to AI research group at the end
- Details on Canvas tomorrow. You can indicate if you want to join in your proposal.

Topics



- Variable transformations
- Recap and recap exercise
- Machine learning
- k-nearest neighbor algorithm

Variable transformations



- Variable transformations are a kind of feature engineering
- You transform information the algorithm can't use very well into something it can
- Example: in a linear regression, how would you predict traffic jams (km/year) in a country from:
 - Population (millions of inhabitants)
 - Land area (km²)

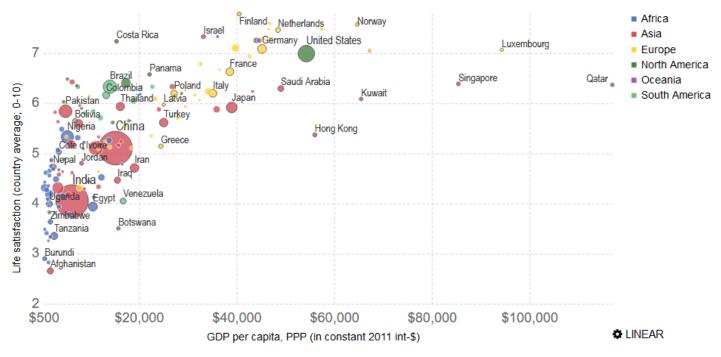
Log transformation



GDP per capita vs Self-reported Life Satisfaction, 2017



Vertical axis shows national average self-reported life satisfaction in the Cantril Ladder (a scale ranging from 0-10 where 10 is the highest possible life satisfaction). Horizontal axis shows GDP per capita based on purchasing power parity (i.e. GDP per head after adjusting for inflation and cross-country price differences).



Source: World Bank, World Happiness Report (2018)

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GDP/capita and self-reported life satisfaction don't show a linear relation – instead, logarithmic

Ourworldindata.org, Max Roser (CC-BY-SA)

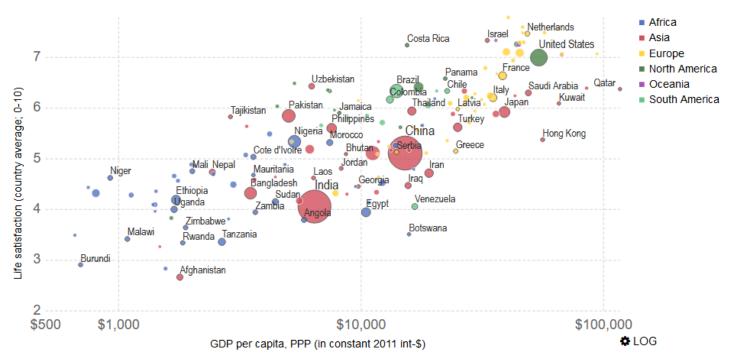
Log transformation



GDP per capita vs Self-reported Life Satisfaction, 2017



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Source: World Bank, World Happiness Report (2018)

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If we take the logarithm of GDP/capita, we get a linear relationship Ourworldindata.org, Max Roser (CC-BY-SA)

Topics



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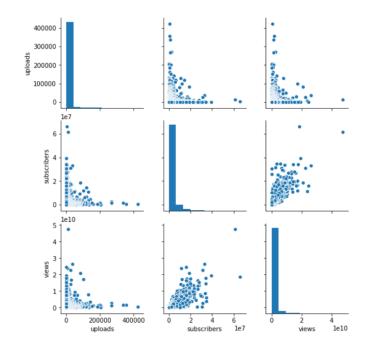


- Clean and explore data
- Look for patterns in data, investigate relations, e.g.
 - Find types of users that use your website differently
- Build statistical models to make predictions to add value to your organization, e.g.
 - Predict the success of a new video ad
 - Approach the users most likely to make a purchase





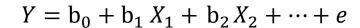




Plot the distributions of qualitative variables to understand the differences

Scatterplots and scatterplot matrices are a great way to understand the relations between quantitative variables

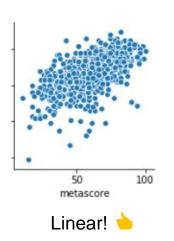
Linear model

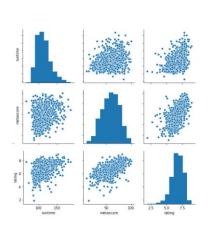


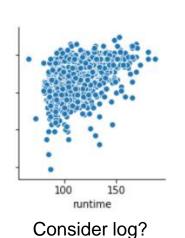


Build

- Select variables
- Make dummy variables for qualitative variables
- Check linearity, consider transformations

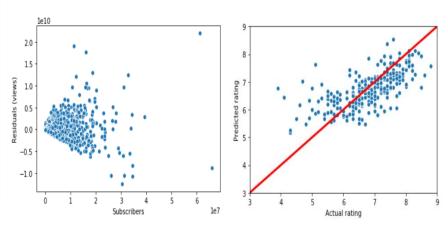






Evaluate

- Create train and test data set.
 Evaluate on the test set.
- Calculate:
 - R²: proportion of variance explained
 - RMSE: typical error in prediction
- Plots:



Residual plots

Y-Y' (actual vs. predicted)

Exercise 1: recap exercise



This is a recap exercise to get you working on your own as a data scientist.

Use the example Notebooks from previous weeks and/or the *Cookbook* Notebook in the main folder.

A Notebook to start with can be found in exercises/recap_linear_regression. The data can be found in the same folder. The data are artificial, but some relations have been built in. The data are on frequent bol.com shoppers.

Your task is to create a linear model that predicts the money spent on bol.com per year. Try and find the optimal model.

Then, predict the variable *spent_bol* for the hold-out test in the folder. You are no longer allowed to change your model, so do this at the end!

Some tips:

- Use the example Notebooks and Cookbook to get snippets of code that you need
- Use graphs to explore the data
- Think of possible transformations
- Calculate the performance of your model with RMSE
- Beware of overfitting...
- Best performance wins a prize!

Topics



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Machine learning



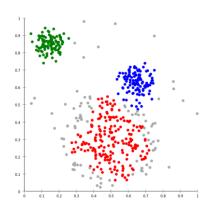
- 'the study of algorithms and statistical models that computer systems use to progressively improve their performance on a specific task.' (Wikipedia)
- Like traditional statistics, but...
 - More variables and cases
 - Complicated and expensive (in terms of computing power) models/algorithms
 - Fitting every 'nook and cranny' of the data



Supervised vs. unsupervised learning

- Supervised: use known patterns to predict new cases
 - Handwriting recognition
- Unsupervised: you let the algorithm discover patterns/clusters on its own
 - Spotify Radio / Discover Weekly

MNIST data set



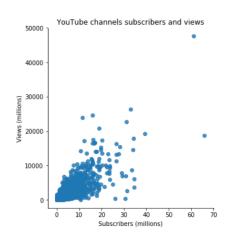




Classification vs. regression

- Within supervised learning, there are two types:
- Classification: categorical dependent
- Regression: numerical dependent

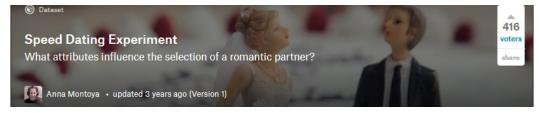




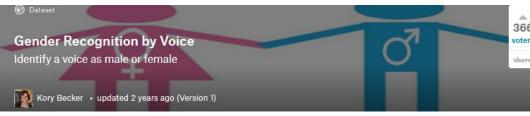
Weekly assignment



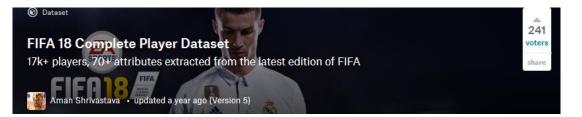












Excercise 2: Kaggle



Go to www.kaggle.com:

- 1. Have a look at the possible data sets for the weekly assignment
- 2. Discuss with your neighbor which data set you would like to try.

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Who is your favorite?

Cristiano Ronaldo or Lionel Messi





- k-nearest neighbor is one of the simplest algorithms in machine learning
- From the data set, pick the k nearest neighbors (k = 3, 5, 7, etc.) of the individual you want to predict for
 - Classification: pick the most frequent answer (e.g., Ronaldo or Messi)
 - Regression: take the mean of the neighbors (e.g., apps downloaded last year)
- What is the obvious 'problem' with this algorithm?

Distance



- How do you calculate a 'distance' between individuals?
- Take 'Euclidean distance'
 - A 'distance' between two individuals can be calculated for any number of dimensions/variables
- Treat all variables the same
 - Normalize all variables so that they are all on the same scale (mean = 0, sd = 1)

ld	nr phones	nr tablets	Apps donwloaded
1	2	2	50
2	1	2	42
3	1	1	23

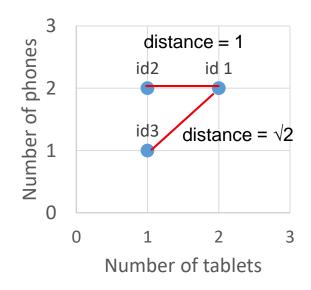


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- Cristiano Ronaldo by Ruben Ortega (CC-BY-SA)
- Lionel Messi by Кирилл Венедиктов (GNU license)