

DATA SCIENCE: REGRESSION MODELS

André R. Brodtkorb

1

TODAYS TOPIC

- Interpolating data with polynomial interpolation
- Approximating data with regression models
- Training and test datasets

MOTIVATION



What is a (my) home worth today?



- All other similar estimates
- electricity price
 - How effective will a drug be for a patient?
 - ...

samsolgt

by DNB Eiendom

Fortell oss litt om boligen din

Adresse
Eksempel: Gatenavn 1A, 0000 Sted...

Størrelse
Eksempel: 40

Antall rom
- 2-roms +

Etasje
- 1. etasje +

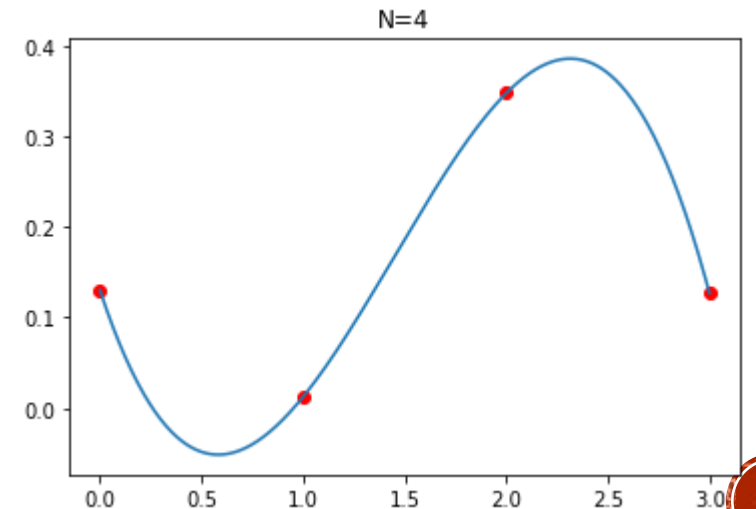
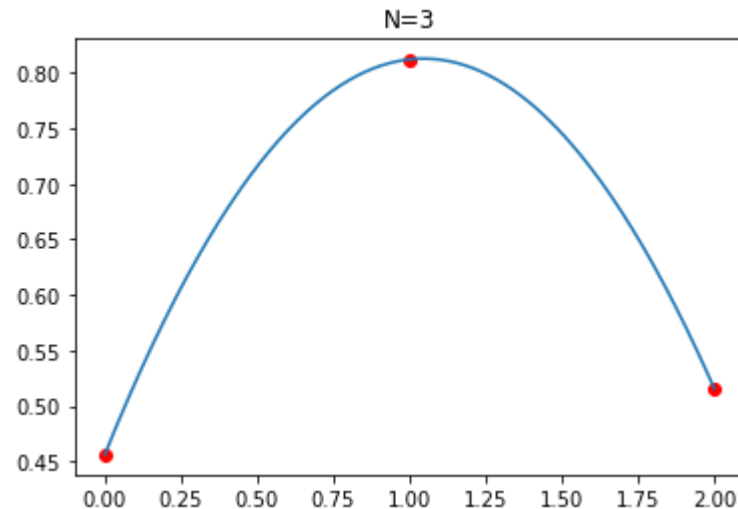
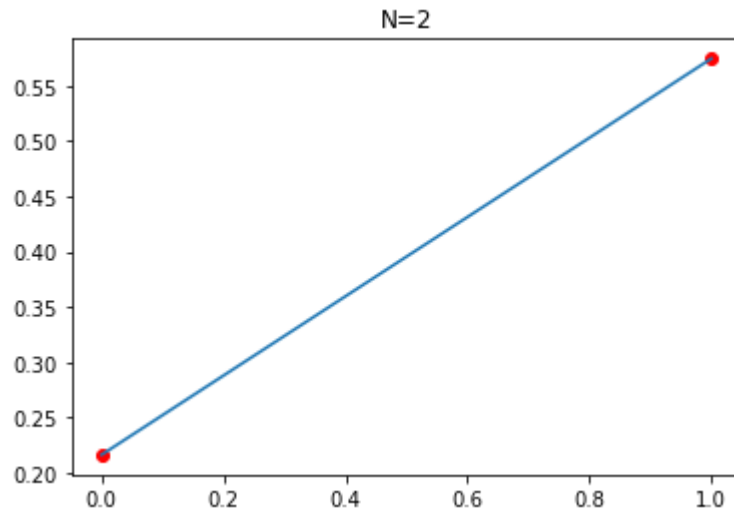
Fasiliteter

Garasje/P-plass	Ildsted
Balkong/Terrasse	Utsikt
Sentralt	Heis

Beregn verdiestimat

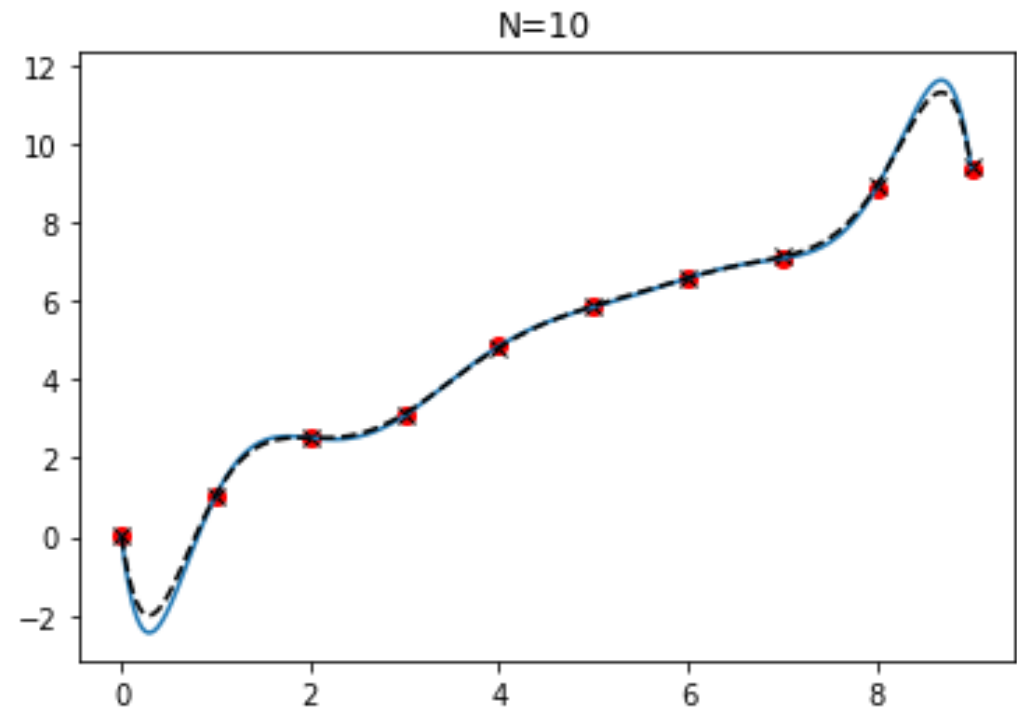
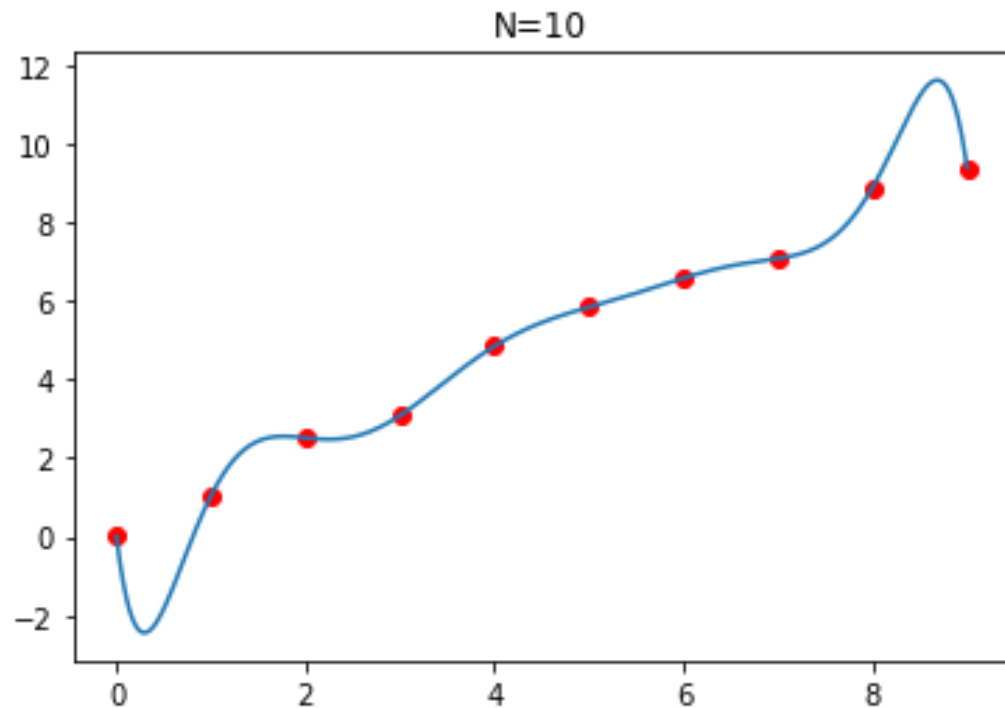
POLYNOMIAL INTERPOLATION

- For n data points, we can find a degree $n-1$ polynomial that interpolates all data points
 - Two points: line ($f(x) = ax + b$)
 - Three points: parabola ($f(x) = ax^2 + bx + c$)
 - Four points: cubic function ($f(x) = ax^3 + bx^2 + cx + d$)
 - ...



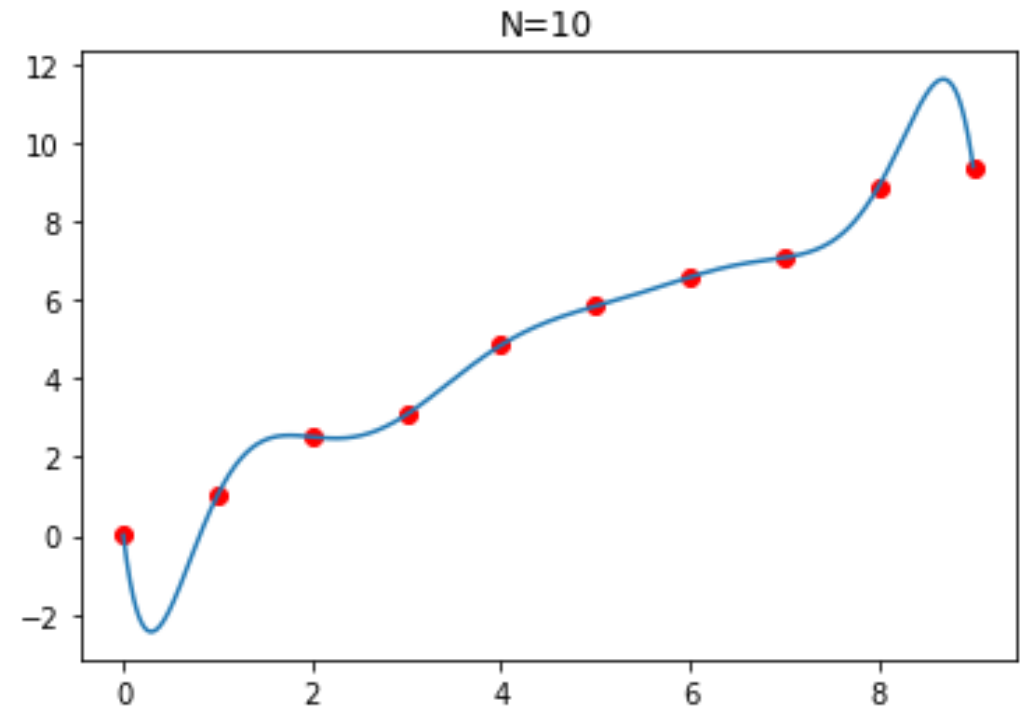
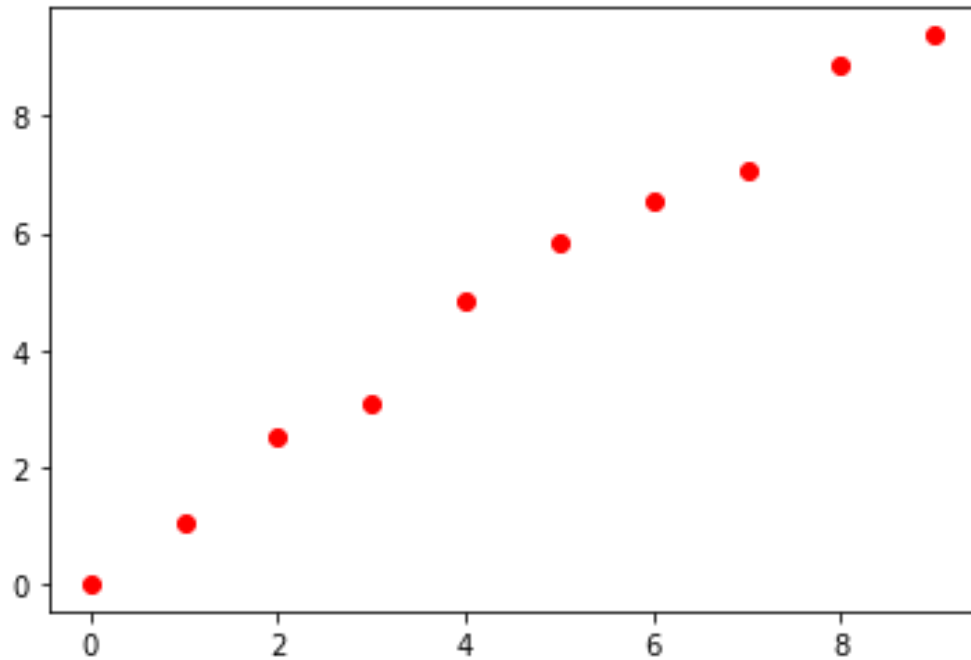
PROBLEMS

- Polynomial interpolation is unstable for large n
- Sensitive to noise



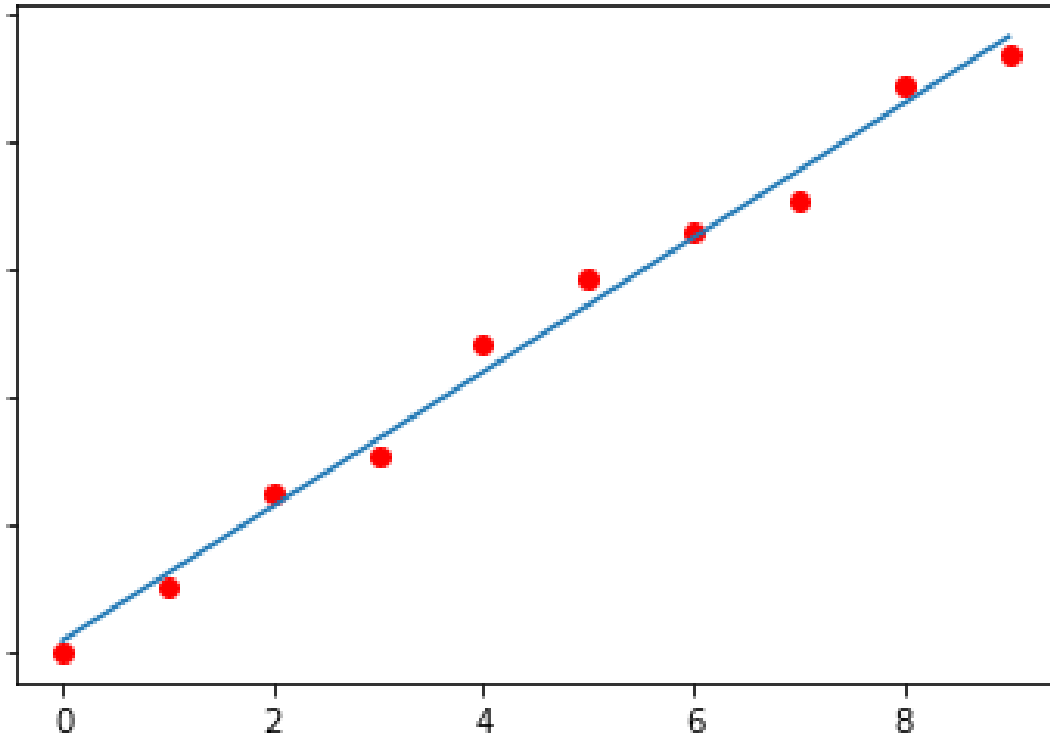
IS IT A GOOD REPRESENTATION?

- Is a degree 10 polynomial a good fit for our data?



APPROXIMATING DATA

Linear regression



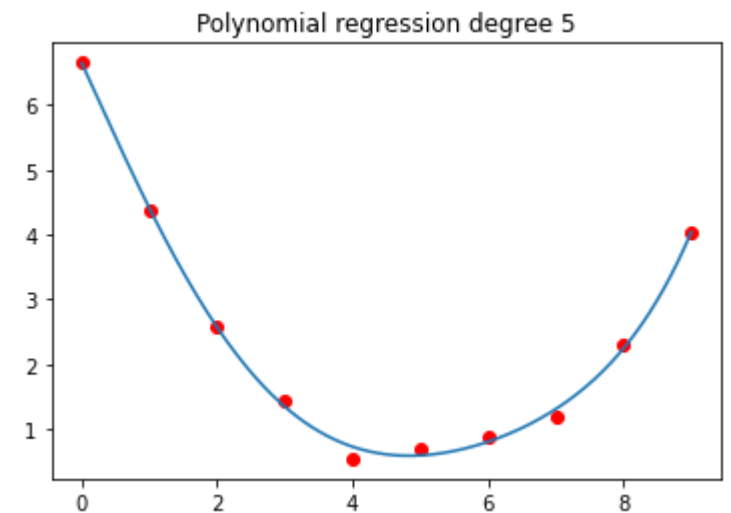
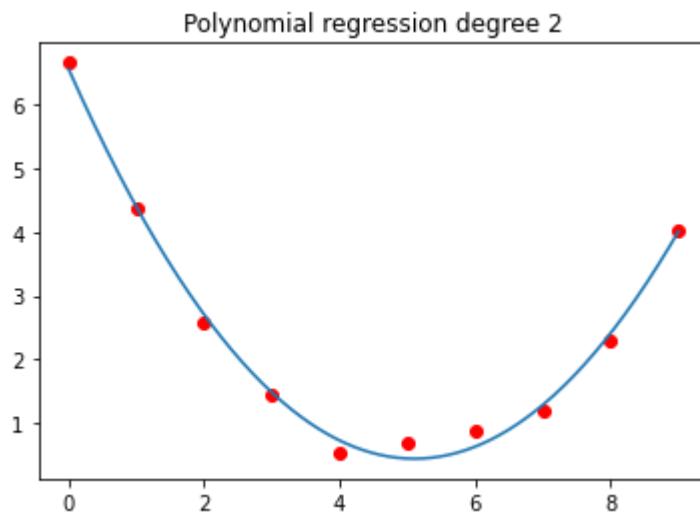
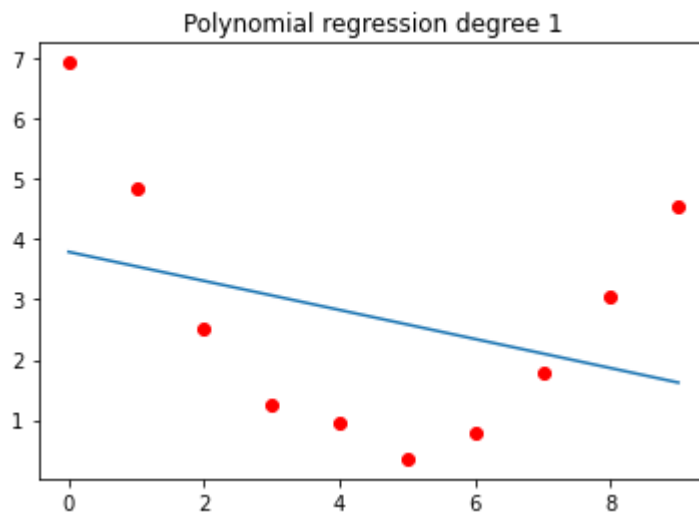
- Instead of interpolating all values, we can find a function that approximates our data (also called regression analysis)
- Our data "looks" linear, lets try linear regression

JUPYTER NOTEBOOK

- Linear regression in Python

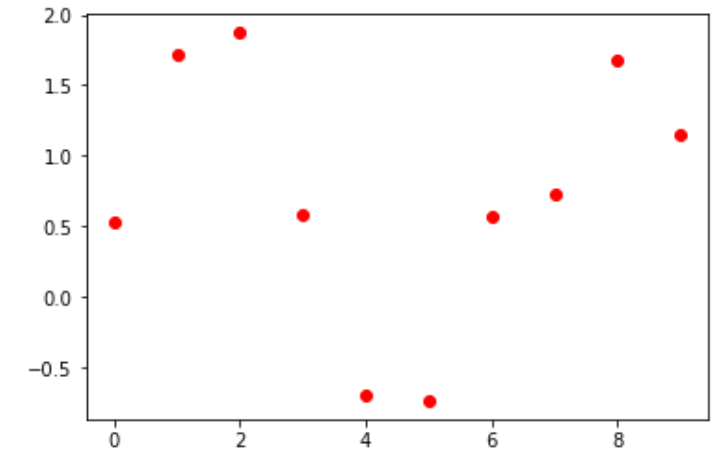
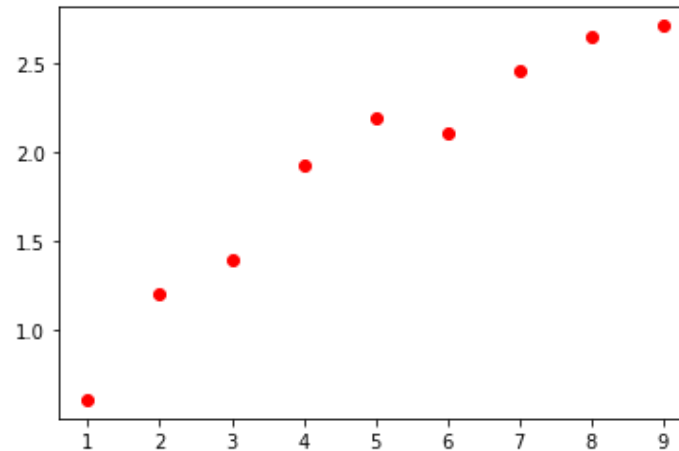
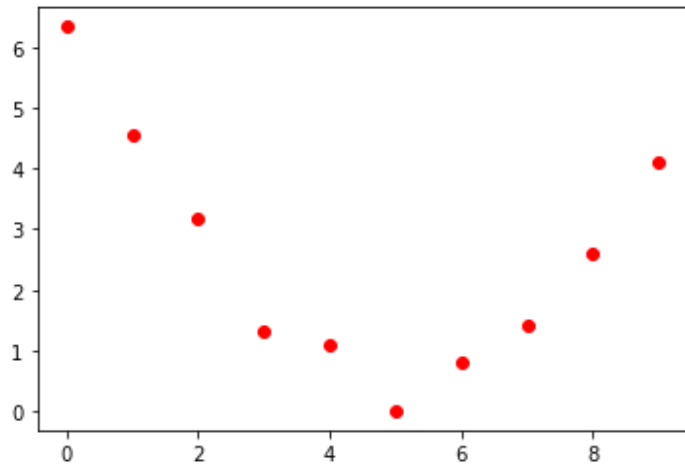
CHOOSING A GOOD MODEL FOR OUR DATA

- Need to inspect data
- Need an educated guess on what type of model should fit our data "well"
- "Easy" for one-dimensional data, very difficult for 4D or higher.



WHAT DOES OUR DATA LOOK LIKE?

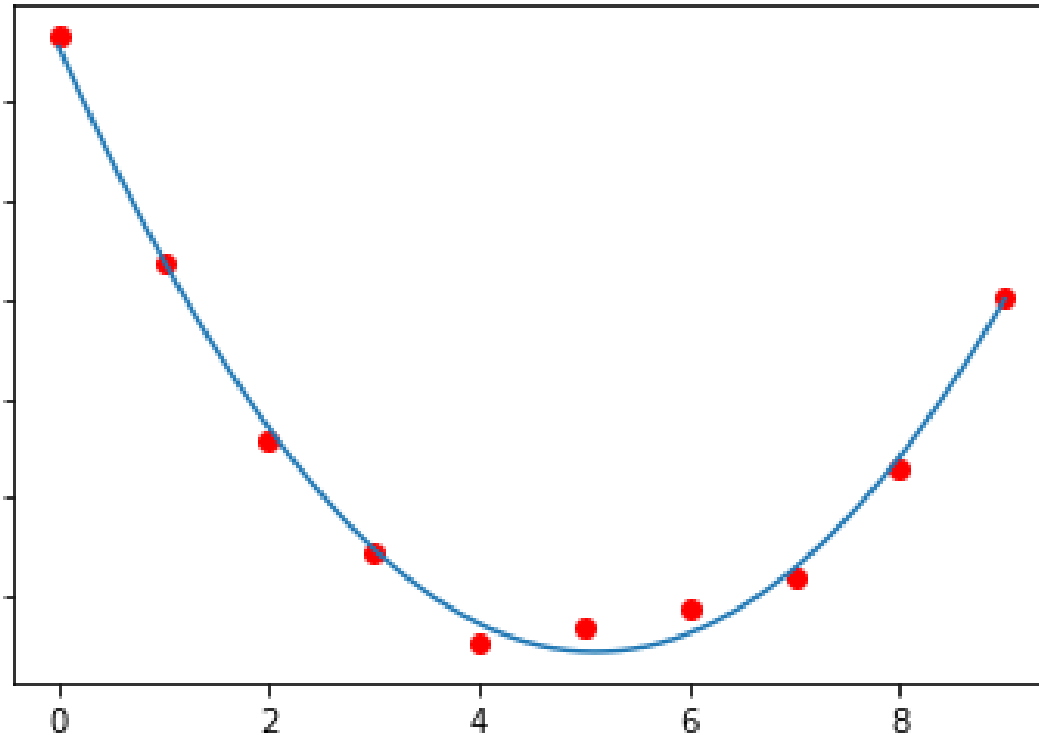
- X^2 , $\log(x)$, $\sin(x)$, ...?



- Sometimes it is difficult to determine or unknown!

QUANTIFYING THE ERROR

Polynomial regression degree 2



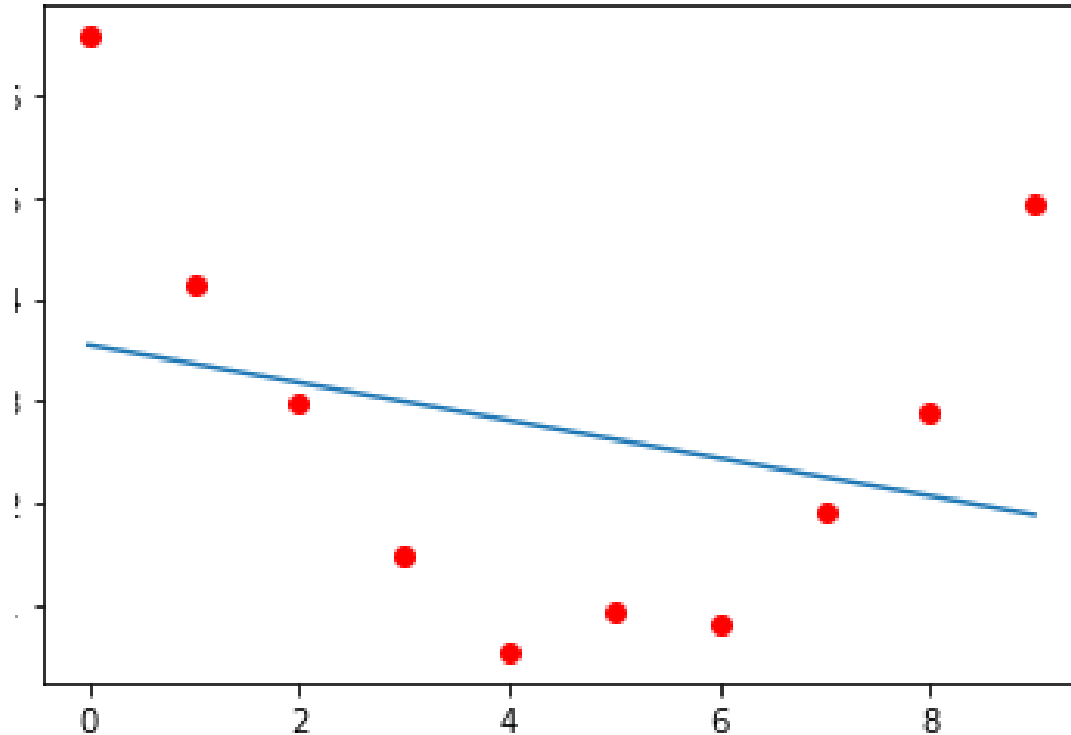
- Mean average error (MAE)
 - Average of absolute difference between prediction and observation
- Mean squared error (MSE)
 - Average of square of difference between prediction and observation
- Root mean squared error (RMSE)
 - Square root of mean squared error
- (more as well, see `scikit.learn` model evaluation for example)

JUPYTER NOTEBOOK

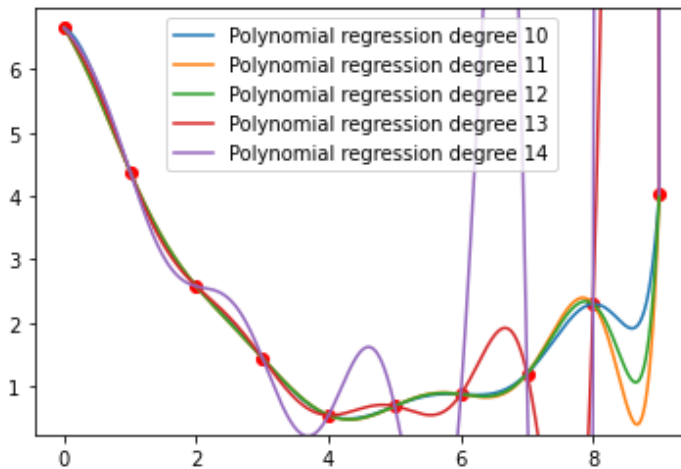
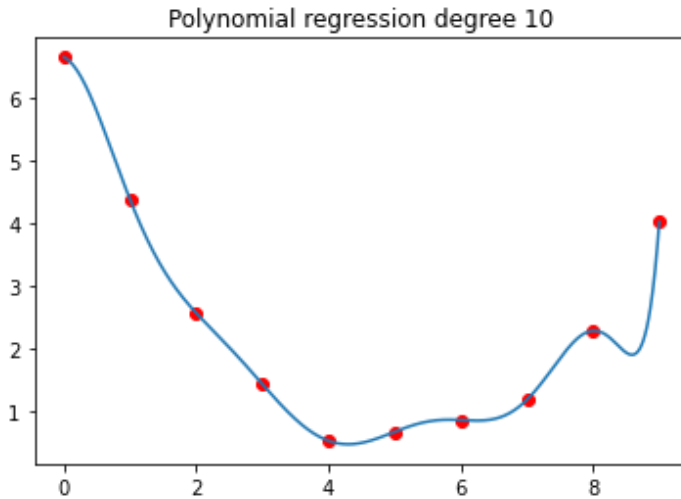
- Score of our linear regression example

UNDERFITTING

Linear regression



- Underfitting happens when we have a too simple model
- Example: Using a linear model to predict nonlinear behaviour
- Symptoms: poor predictive skill, even on the data we try to fit

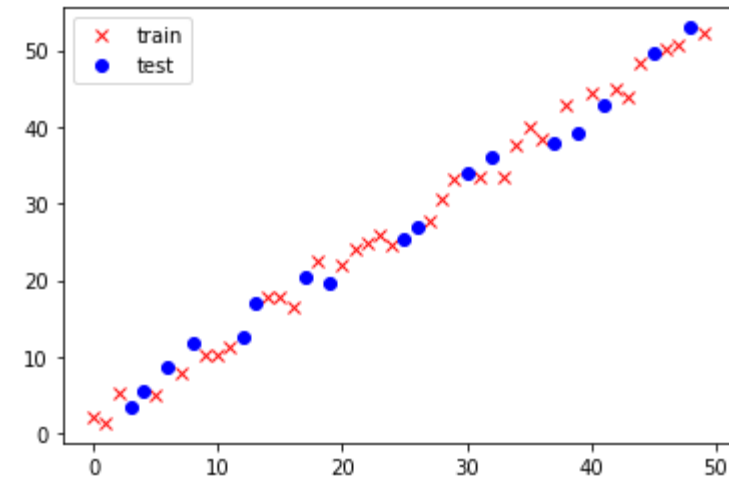
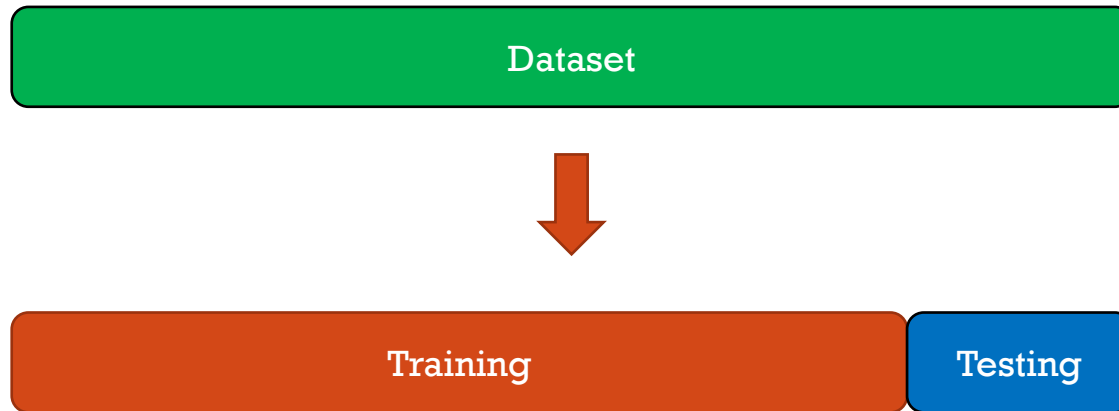


OVERFITTING

- Overfitting is when we have too much freedom in our model
- Example: Using a polynomial of degree $n-1$ for n data points (interpolation)
- Symptoms: Model is extremely good at predicting known data, but terrible at predicting new data

TESTING THE MODEL

- So far, we have tested the model on data that it's already "seen"
- This is not a very good way to quantify model performance
- In machine learning, the dataset is usually divided into train and test subsets

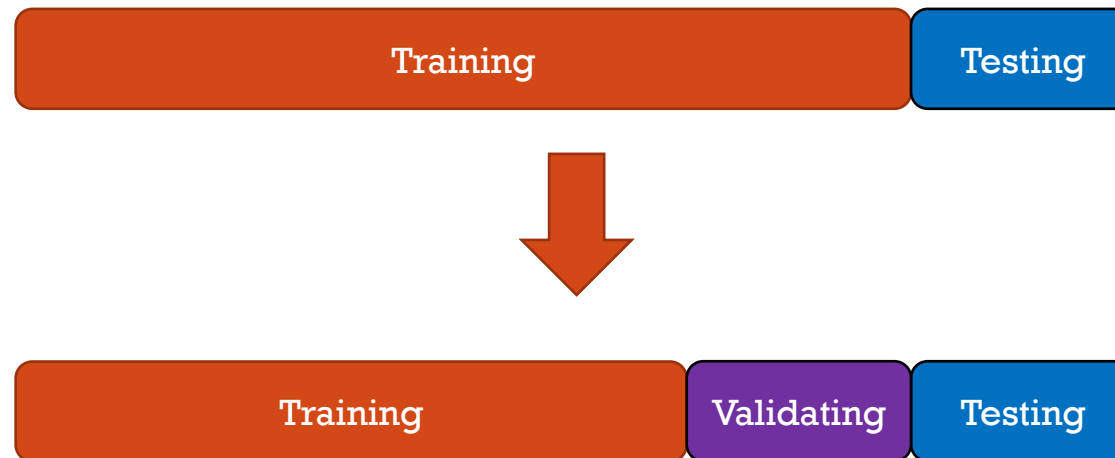


JUPYTER NOTEBOOK

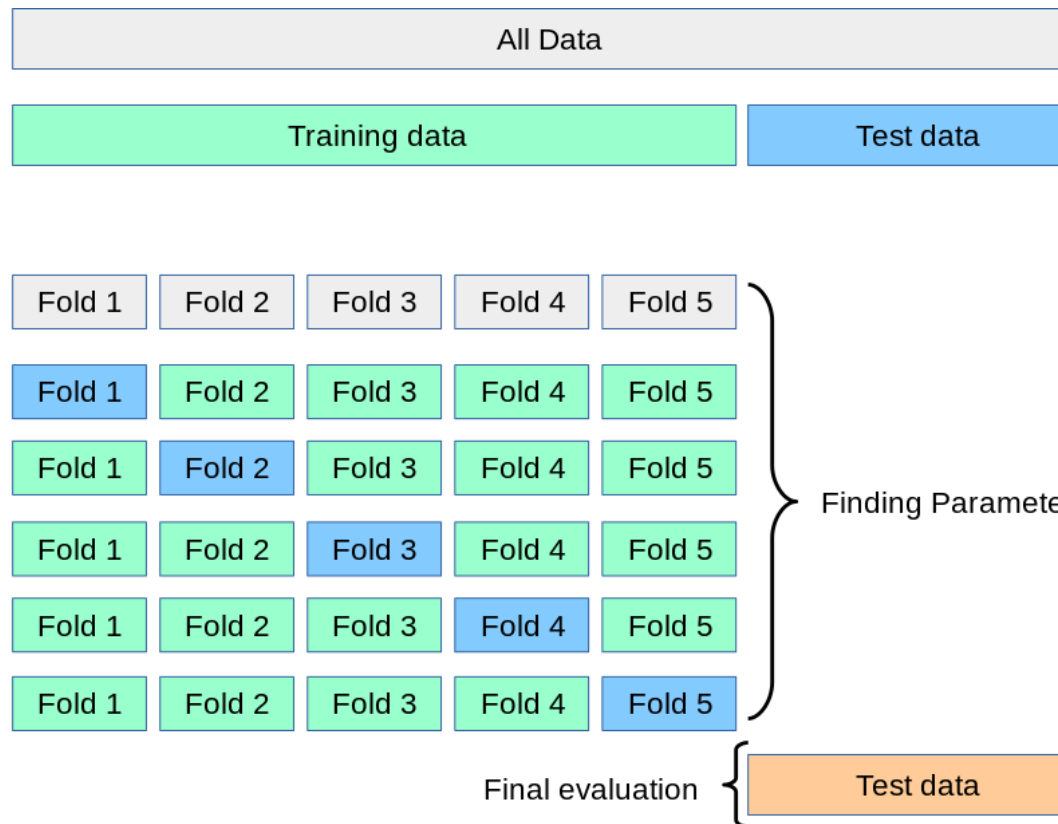
- Testing model performance on test dataset

VALIDATION

- Validation data is used to check model performance and set hyperparameters
- Model may "see" the validation data through performance feedback
- Testing data is still not part of training



K-FOLD CROSS-VALIDATION



- Divide data into k subsets
- Train k models, using a different subset as test data for each model
- Use the rest of the data for training
- Evaluate on separate test dataset

SUMMARY

- Polynomial interpolation does not scale
 - Sensitive to noise and high order
- Regression models approximate data
 - Check for underfitting and overfitting and find the sweetspot in between
- Testing and training datasets
 - K-fold cross-validation
- Source code on github: <https://github.com/babrodtk/>
- Slides on webpage: <https://brodtkorb.org/>

BONUS: BOOTSTRAPPING

- Assume you have population you want to model
- Create a "sample" (subset) of size n
- Pick n data points (with replacement) from your population to create a "bootstrap sample"
- Fit a model to each bootstrap sample
- Average models for prediction

