2025

4Geeks Academy: data science cohort 12

# DAY 19: LINEAR REGRESSION

### TODO

# LINEAR REGRESSION

Model details, evaluation, overfitting and regularization

# LINEAR REGRESSION PROJECTS

- Health insurance cost project 'basic' linear regression
- Health demographics project regularized linear regression

**EDA PROJECT** 

Submit Data Preprocessing Project Tutorial (Exploratory data analysis project) if you haven't done so already.

# **TOPICS**

O1 LINEAR REGRESSION MODELS

**02** REGRESSION METRICS

O3 OVERFITTING

# LINEAR REGRESSION

### **WHAT**

Predicts a continuous label from input feature or features

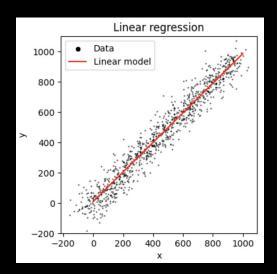
### WHY

- Simple
- Easy to train does not require large amounts of data, time or compute resources
- Interpretable can tell you about the relative importance of features

### HOW

$$Y = mX + b$$

Minimizes error by finding 'best' values for m and b.



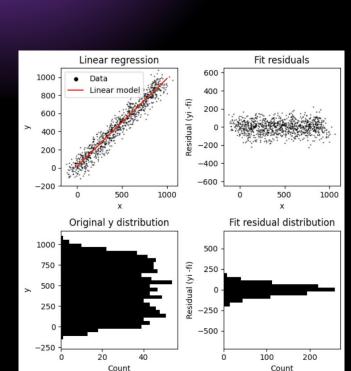
- Simple linear regression: one input variable
- Multiple linear regression: two or more input variables

Available via Scikit-learn LinearRegression()

# REGRESSION METRICS

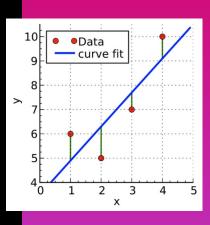
### **RMSE**

- Quantifies on average, how far predicted value is from the true value
- Lower is better
- Has same units as target variable



# R-squared

- Quantifies fraction of variance in the label explained by the model
- Higher is better
  - Zero -> model has no explanatory power
  - o One -> model makes perfect predictions



### **OVERFITTING**

# BIAS-VARIANCE TRADEOFF

Very good fit on training data often leads to bad generalization

### **OVERFITTING**

The model is too tuned to the training data, it can't predict the new/different test data

### **REGULARIZATION**

Set of techniques to mitigate overfitting

- Don't use a over-powered model
- Don't use too many features

### **Linear regression:**

- Ridge regression adds a penalty to constrain size of coefficients
- Lasso regression adds a penalty to force coefficients to zero

