# Machine Learning

Week 3 - Logistic Regression

## About the course

#### Announcements

- Assignment for this week is already in the Notebooks folder
- Notebooks can be done as group (2-3 ppl)
- Two walkthrough videos for terminal (Mac and Windows) and Conda Navigator (Windows)
- This week's assigned paper is kaggle\_competition\_automated\_essay\_grading\_...
- You can download this week's Notebook folder if you haven't and start Jupyter Notebook
- If you can, camera on is appreciated

### Previously on ML..

- What are the 3 main Machine Learning categories?
- What is the formula for a typical simple linear regression?
- Which expression of the Cost function do we prefer and why?

$$J(\theta) = \frac{1}{m} \sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)})^{2}$$

$$J(\theta) = \frac{1}{2m} \sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)})^{2}$$

- How do we evaluate the performance of a simple linear regression?
- What is the difference between simple, multiple and polynomial regression?
- How can you improve a linear regression model?

# Logistic Regression

#### Definition

**Logistic regression** is a statistical **model** that in its basic form uses a **logistic** function to **model** a binary dependent variable.

## Understanding the odds and log of odds

https://www.youtube.com/watch?v=ARfXDSkQf1Y&ab\_channel=StatQuestwithJoshStarmer

# The Derivative of Cost Function for Logistic Regression

https://medium.com/analytics-vidhya/derivative-of-log-loss-function-for-logistic-regression-9b832f025c2d#: ~:text=Since%20the%20hypothesis%20function%20for,function%20follows%20a%20certain%20pattern.

https://rpubs.com/dnuttle/ml-logistic-cost-func derivative

https://web.stanford.edu/class/archive/cs/cs109/cs109.1178/lectureHandouts/220-logistic-regression.pdf

#### Maximum Likelihood

https://www.youtube.com/watch?v=BfKanl1aSG0&ab\_channel=StatQuestwithJoshStarmer

# Conclusion