## Regression + Assignments 2, 3

(Neural Networks Implementation and Application Tutorial)

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#### Overview

- Assignment 2
- Regression
- Assignment 3

### Assignment 2

- Tutor cue: go through the assignment
- Questions?
- Did it work?
- Were you able to collaborate?

# Regression

- What is the difference between classification and regression?
- What is regression in terms of functions? • Any function  $f: F \to \mathbb{R}$  (from joint feature space to numbers)
- What is *linear* regression? <sup>9</sup>
  - $\hat{\mathbf{v}} = \mathbf{x}^T \cdot \boldsymbol{\beta} + \boldsymbol{\beta}_c$  (parameters  $\boldsymbol{\beta}, \boldsymbol{\beta}_c$ )

Which of the following are regression (and linear/polynomial) models? 9 1. 5

- **2**  $4 \cdot x_1 + 5$
- $\mathbf{6} \ \mathbf{4} \cdot x_1 + 3 \cdot x_2^2 + 5$
- $4 \cdot x_1 + 3 \cdot x_1 \cdot x_2 + 5$
- $\mathbf{6} \ 4 \cdot x_1 + 3 \cdot \sin(x_2^2) + 5$
- $\begin{cases}
  4 \cdot x_1 + 5 & \text{if } x_2 \ge 10 \\
  3 \cdot x_1 + 4 & \text{if } x_2 < 10
  \end{cases}$

### Regression

# Regression to Classification 🤔 🤔



Assume that we have a function that outputs a score for every class, e.g. Predict sentiment into (positive, negative, neutral):

$$(15.0, -2.3, 4.1)$$

- How do we use this for classification?
  - Argmax
- Can we get a probability distribution?
  - Softmax:  $\frac{\exp x_i}{\sum_{i} \exp x_k}$

# Loss & Regularization

#### Loss

- Why  $L_2$  and not  $L_1$ ?
  - We care about points that are drastially mispredicted, e.g.  $L_2(-1, 10)$  and not about almost correctly predicted instances  $L_2(-1, -1.1)$
  - In  $L_1$ , these would have the same weight (gradient)
  - $\sim L_1$  leads to some parameters being 0 (sometimes a good thing)

#### Regularization

- Why do we want to regularize?
  - Prevent overfitting, prevent reliance on noise
- Ridge regression uses  $L_2$  penalty: minimize arg min  $L_2^2(\hat{Y}, Y) + \lambda ||\beta||_2^2$
- Lasso regression uses  $L_1$  penalty: minimize arg min  $L_2^{\bar{2}}(\hat{Y}, Y) + \lambda ||\beta||_1$
- ElasticNet regression uses both: minimize arg min  $L_2^2(\hat{Y}, Y) + \lambda_1 ||\beta||_1 + \lambda_2 ||\beta||_2^2$

## Model Capacity

- What contributes to overfitting?
  - Overfitting: little data, large model capacity, too many optimization steps
  - ▶ Underfitting: not enough optimization steps, too strict regularization

#### Bias-variance tradeoff

- Large bias corresponds to ...?
  - Underfitting/small model capacity
- Large variance corresponds to ...?
  - Overfitting

### Assignment 3

• Any questions?