# Assignment 8,9 + Classifiers (SNLP Tutorial 9)

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# Assignment 8

- Exercise 1: Feature Selection (DF, PMI)
- Exercise 2:  $\chi^2$
- Exercise 3: Author identification
- Bonus: Features for clustering

#### **Decision Trees**

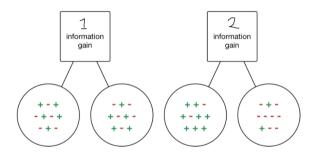
- Can be used for classification as well as regression
- Algorithm

```
function Decision-Tree-Learning(examples, attributes, parent_examples) returns a tree if examples is empty then return Plurality-Value(parent_examples) else if all examples have the same classification then return the classification else if attributes is empty then return Plurality-Value(examples) else A \leftarrow \underset{a \in attributes}{\operatorname{argmax}} \quad \text{Importance}(examples) \\ tree \leftarrow \text{a new decision tree with root test } A \\ \text{for each value } v_k \text{ of } A \text{ do} \\ exs \leftarrow \{e: e \in examples \text{ and } e.A = v_k\} \\ subtree \leftarrow \text{Decision-Tree-Learning}(exs, attributes - A, examples) \\ \text{add a branch to } tree \text{ with label } (A = v_k) \text{ and subtree } subtree \\ \text{return } tree
```

What is plurality value? What is importance?

## Decision Trees - Questions

• Which of the 2 splits has a better information gain?



- What are the pros and cons of decision trees?
- How to avoid overfitting?

# Naïve Bayes

TODO

#### kNN

#### Algorithm

```
k-Nearest Neighbor Classify (\mathbf{X},\mathbf{Y},x) // \mathbf{X}: training data, \mathbf{Y}: class labels of \mathbf{X}, x: unknown sample for i=1 to m do Compute distance d(\mathbf{X}_i,x) end for Compute set I containing indices for the k smallest distances d(\mathbf{X}_i,x). return majority label for \{\mathbf{Y}_i \text{ where } i \in I\}
```

#### Questions

- What are the training and test computation times for kNN?
- What are the pros and cons of kNN classifiers?
- Can kNN be used for regression?
- How will the value of k affect the bias?

## **SVM**

- Find a boundary that maximizes the distance to closest vectors
- If not possible, find one that minimizes the error
- Add the kernel trick

# Perceptron

- Binary classification
- Linear boundary in feature space
- $\hat{y} = sign(wx + b)$

## Algorithm:

• 
$$w_0 = \overrightarrow{0}$$

- For every data point  $x_i$
- $\qquad \qquad \triangleright \ \hat{y_i} = \operatorname{sign}(w_k x_i + b)$
- if  $\hat{y_i} \neq y_i$ :
- $\bullet \qquad \star \quad w_{k+1} = w_k \hat{y}_i \cdot x$
- else:
- $\bullet \qquad \star \quad w_{k+1} = w_k$
- TODO: illustration

# Useful Python Implementations

#### Resources

- UdS SNLP Class, WSD: https://teaching.lsv.uni-saarland.de/snlp/
- Decision Trees: https://www.kdnuggets.com/2020/01/decision-tree-algorithm-explained.html
- **1** kNN Example: https://iq.opengenus.org/text-classification-using-k-nearest-neighbors/