

## **Lecture 10**

# Multinomial Logistic Regression

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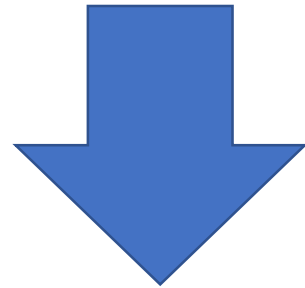
## What we discussed so far...

- Logistic regression model → Probabilistic model
- Where from the name 'logit' comes from?
- What are the important aspects to train a LR model?



# Logistic Regression

- Binary classification model  $\rightarrow$  probabilistic
- It can be extended to multiclass classification problem in different ways.



## Softmax Regression (Multinomial Logistic Regression)

# *Multinomial Logistic Regression*

- Labeled dataset with  $N$  classes ( $N > 2$ )
- The Softmax Regression model first computes a score  $s_k(x)$  for each class  $k$
- Then it estimates the probability of each class by applying the *softmax function* (also called the *normalized exponential*) to the scores.

*Softmax score for class  $k$*

$$s_k(\mathbf{x}) = \mathbf{x}^T \boldsymbol{\theta}^{(k)}$$

The scores are generally called logits or log-odds.

Note that each class has its own dedicated parameter vector  $\boldsymbol{\theta}^{(k)}$

$\boldsymbol{\theta}^{(k)}$   $\rightarrow$  Column vector with  $(n+1)$  elements

$\mathbf{x}^T$   $\rightarrow$  Row vector with  $(n+1)$  elements

## *Softmax function*

$$\hat{p}_k = \sigma(\mathbf{s}(\mathbf{x}))_k = \frac{\exp(s_k(\mathbf{x}))}{\sum_{j=1}^K \exp(s_j(\mathbf{x}))}$$



the estimated probability

# Classes  $\rightarrow K$

## sklearn.linear\_model.LogisticRegression

```
class sklearn.linear_model.LogisticRegression(penalty='l2', *, dual=False, tol=0.0001, C=1.0, fit_intercept=True,
intercept_scaling=1, class_weight=None, random_state=None, solver='lbfgs', max_iter=100, multi_class='auto', verbose=0,
warm_start=False, n_jobs=None, l1_ratio=None)
```