

Tutorial for Project 2

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- Multiclass Classification
- Class imbalance
- Task description

Multiclass Classification

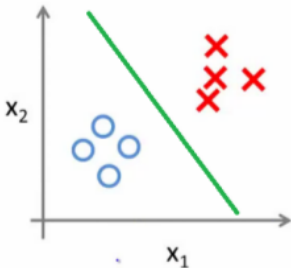
Multiclass Classification

- ▶ Classification when there are more than two classes.
- ▶ Several ways to approach this:
 1. one-vs-all strategy
 2. one-vs-one strategy
 3. multinomial logistic regression
 4. ...

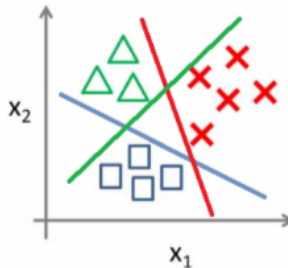
Multiclass Classification – One-vs-All

- Create C *binary* classifiers.
- Each classifier separates one of the classes with respect to *the union of all other classes*.
- Classifier with highest value / confidence is chosen in test time.

Binary classification

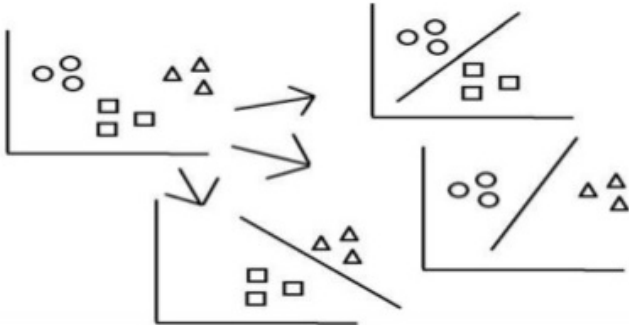


one-vs-all



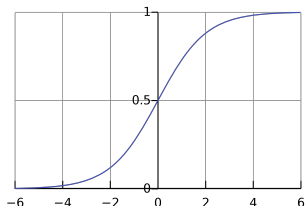
Multiclass Classification – One-vs-One

- ▶ Create $\binom{C}{2}$ binary classifiers, one for every pair of classes.
- ▶ In test time, the class with "highest number of votes wins", according to some voting schema and some tie-breaking rule.



Multiclass Classification – Multinomial Logistic Regression

Logistic function is given as $p_{\theta}(x) = \frac{1}{1+e^{-\theta^T x}}$. For $\theta = \vec{1}$ we have:



- Binary classification loss is defined as follows:

- $$L(\theta) = - \sum_{i=1}^N y_i \log(p_{\theta}(x_i)) + (1 - y_i) \log(1 - p_{\theta}(x_i))$$

where (x_i, y_i) are N pairs of type (data point, binary label).

Multiclass Classification – Multinomial Logistic Regression

Logistic regression for multiclass classification:

$$L(\theta) = - \sum_{i=1}^N \sum_{c=1}^C \mathbb{1}_{\{y_i=c\}} \log \left(\frac{e^{\theta_c^T x_i}}{\sum_{k=1}^C e^{\theta_k^T x_i}} \right)$$

Class imbalance

Class imbalance

- ▶ Refers to the case when the classes are unequally distributed across a data set.
- ▶ We may have a situation in which certain classes are under- or over-represented.
- ▶ Can be more or less important depending on what we optimize for.
- ▶ Example:

		Predicted	
True	Classifier I	Cancer	Normal
	Cancer	0	5
	Normal	0	1000

		Predicted	
True	Classifier II	Cancer	Normal
	Cancer	5	0
	Normal	95	900

Which classifier performs better?

Class imbalance

Possible solutions:

1. Oversampling.
2. Downsampling.
3. Balanced training with weighed loss function.

Task description

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Data:

- ▶ Pre-extracted features from medical images.
- ▶ Each image given as a continuous feature vector of size 1000.
- ▶ 4800 training and 4100 test feature vectors.
- ▶ 3 classes labeled as $\{0, 1, 2\}$.
- ▶ The same class distribution for training and test data.

Task:

- ▶ Multiclass image classification.
- ▶ Learn to map feature vectors into labels.

Evaluation:

- ▶ Class balance-sensitive metric.

- ▶
$$BMAC = \frac{1}{C} \sum_{c=1}^C TPR_c$$
 where $C \in \{0, 1, 2\}$ and $TPR_c = \frac{\#correct_c}{\#all_c}$

Questions?