



# Lab2

Machine Learning 2019  
(P. Zanuttigh – ICT+Physics of Data)

## 2 Notebooks



- Regularized classification on student alcohol dataset
- SVM classification of clothes images

# Regularized Classification

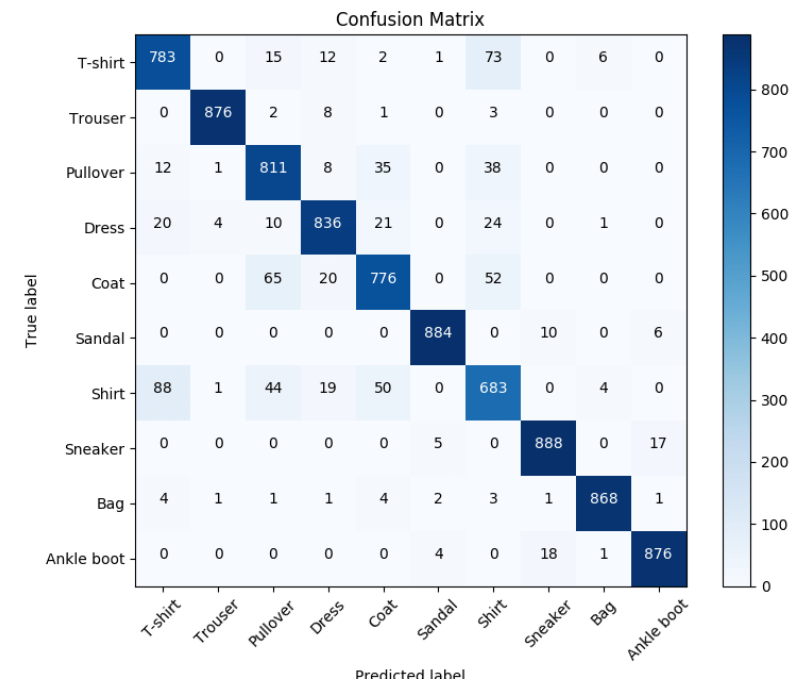


- Estimate if a student drinks based on school and personal data
- Perform logistic regression with and without regularization
- Estimate the optimal  $\lambda$  parameter with cross validation
- See the impact of regularization
- Perform a new test with more training data
- Overall test accuracy is relatively low (65-70%) but the impact of regularization can be seen

# Classification of Clothes Images



- Dataset of small pictures of clothes: multi-class classification
- Use Support Vector Machines
- Try different Kernels
- Estimate parameters with cross validation
- Visualize the results with confusion matrices



# One-hot Encoding

Label Encoding

| Food Name | Categorical # | Calories |
|-----------|---------------|----------|
| Apple     | 1             | 95       |
| Chicken   | 2             | 231      |
| Broccoli  | 3             | 50       |



One Hot Encoding

| Apple | Chicken | Broccoli | Calories |
|-------|---------|----------|----------|
| 1     | 0       | 0        | 95       |
| 0     | 1       | 0        | 231      |
| 0     | 0       | 1        | 50       |



| state |
|-------|
| NY    |
| WA    |
| CA    |

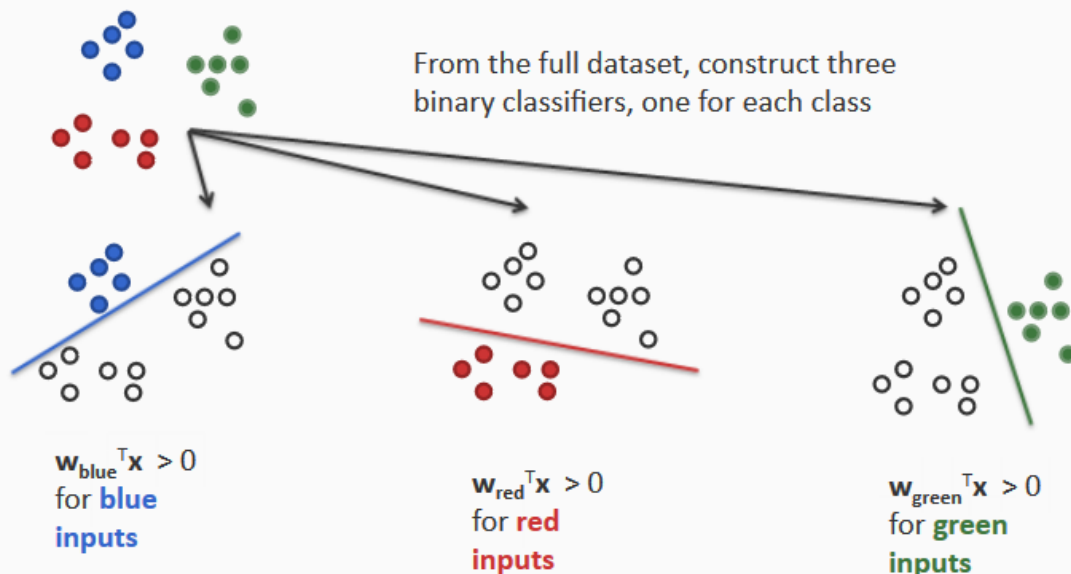
| AL | ... | CA | ... | NY | ... | WA | ... | WY |
|----|-----|----|-----|----|-----|----|-----|----|
| 0  | ... | 0  | ... | 1  | ... | 0  | ... | 0  |
| 0  | ... | 0  | ... | 0  | ... | 1  | ... | 0  |
| 0  | ... | 1  | ... | 0  | ... | 0  | ... | 0  |

- One variable for each class (=1 if sample in class, 0 otherwise)
- Avoid having some classes "closer" to others
- Increases data dimensionality



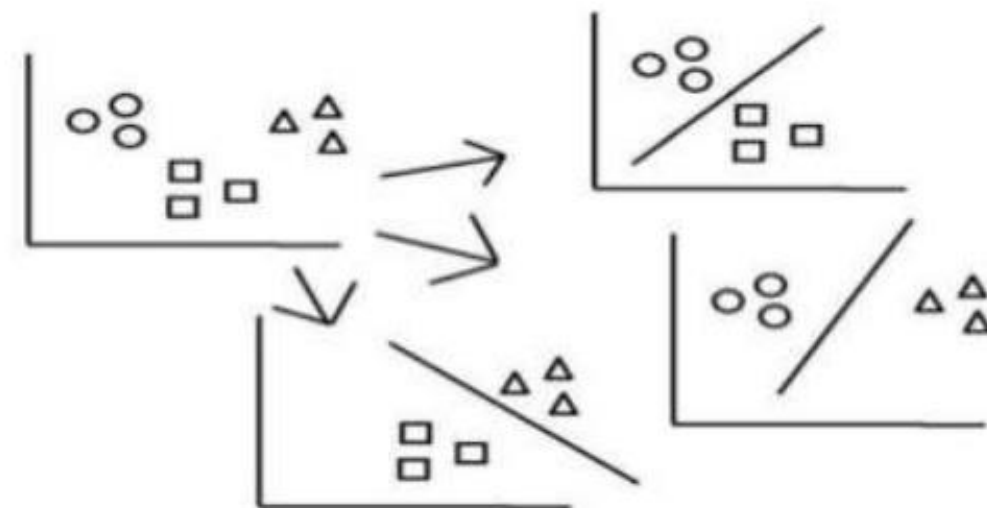
# Multi-class classification

## Visualizing One-vs-all



- Classify each class vs the union of the others
- For each sample select the class with highest classification score, i.e.  $\text{argmax}_i \langle w_i, x \rangle$
- Requires  $n_{\text{classes}}$  comparisons

## One-vs-One (OVO)



- Classify each class vs each other class
- For each sample select the class that has "won" the largest number of classifications
- Requires  $\left(\frac{n_{\text{classes}}-1}{2}\right)^2$  comparisons
- Used by sklearn