

# Basics of ML Assignments



Dmitry Ryabokon, [github.com/dryabokon](https://github.com/dryabokon)


# Assignments

## Scoring

A	90%
B	85%
C	75%
D	65%
E	60%
<hr/>	
FX	<60%

# Assignments

## Scoring

<b>Assignment 1</b>  Classic Machine Learning	<b>Assignment 2</b>  Computer Vision and Deep Learning	<b>Tech Talk</b>
<b>50%</b>	<b>50%</b>	<b>+30%</b> 

# Classic ML

1. Supervised Learning
2. Naive Bayesian Classifier
3. Gaussian Classification Model
4. Linear Discriminant Functions
5. Decision strategies
6. Benchmarking of Supervised Learning Algorithms
7. Unsupervised Learning

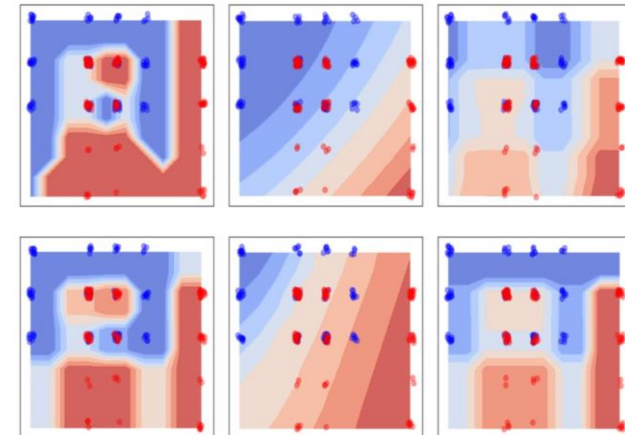
# Supervised learning

## Description

- Get train/test data sample from open sources or generate it by yourself. Focus on low-dimension data, preferably 2 dimensions, 2 classes
- Visualize the distribution of sample data
- Estimate accuracy of data classification for 3-4 classifiers studied over lecture course
- Build ROC curves and hit maps

## Assessment criteria

- In-house classifier: 40%
- Out-of-box classifier: 20%
- Data visualization charts: 20%
- ROC curve charts: 20%
- \* Extra: multi class championship 30%



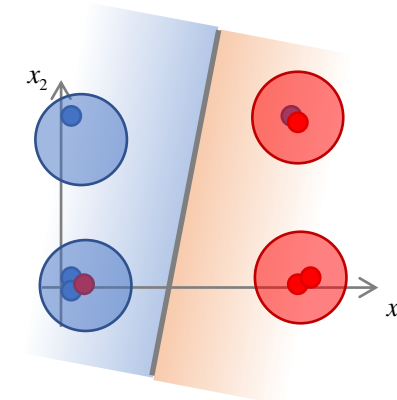
# Naive Bayesian Classifier

## Description

- Get train/test data sample from open [sources](#), e.g. GBP/USD index. Use daily change (↑, ↓) as a target.
- Visualize the distribution of sample data
- Estimate accuracy of data classification
- Build ROC curves and calculate AUC

## Assessment criteria

- In-house classifier: 40%
- Out-of-box classifier: 20%
- Data visualization charts: 20%
- ROC curve charts: 20%
- \* Extra: multi class championship 30%



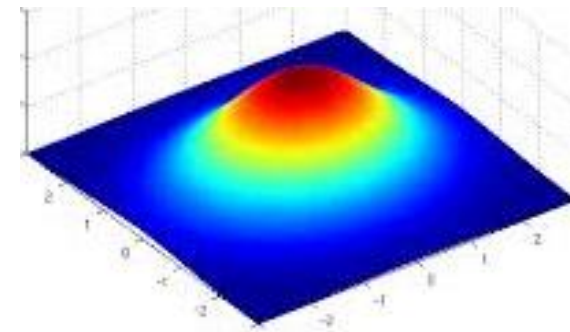
# Gaussian Classification Model

## Description

- Get train/test data sample from open [sources](#), e.g. GBP/USD index. Use daily change (↑, ↓) as a target.
- Visualize the distribution of sample data
- Estimate accuracy of data classification
- Build ROC curves and calculate AUC

## Assessment criteria

- In-house classifier: 40%
- Out-of-box classifier: 20%
- Data visualization charts: 20%
- ROC curve charts: 20%
- \* Extra: multi class championship 30%



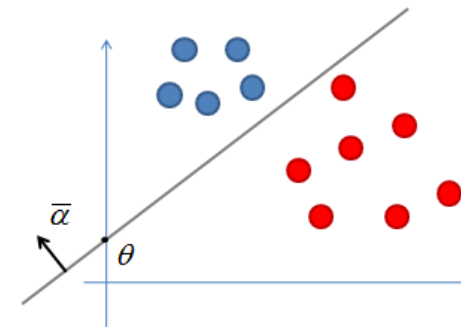
# Linear Discrimination with SVM

## Description

- Get train/test data sample from open [sources](#), e.g. GBP/USD index. Use daily change (↑, ↓) as a target.
- Visualize the distribution of sample data
- Estimate accuracy of data classification
- Build ROC curves and calculate AUC

## Assessment criteria

- In-house classifier: 40%
- Out-of-box classifier: 20%
- Data visualization charts: 20%
- ROC curve charts: 20%
- \* Extra: multi class championship 30%





# Decision strategies

## Description

- Get bet data from open sources
- Visualize the distribution of sample data
- Estimate accuracy of random, deterministic (host, visitor, challenger) and Bayesian strategies
- Build ROC curves and calculate AUC
- Engineer the predictive features and estimate their predictive power

## Assessment criteria

- Deterministic strategy: 20%
- Random Strategy: 20%
- Bayesian Strategy: 40%
- ROC curve charts: 20%
- Show predictive features: 30%
- A
- 0:2
- 0:1
- B
- 1:1



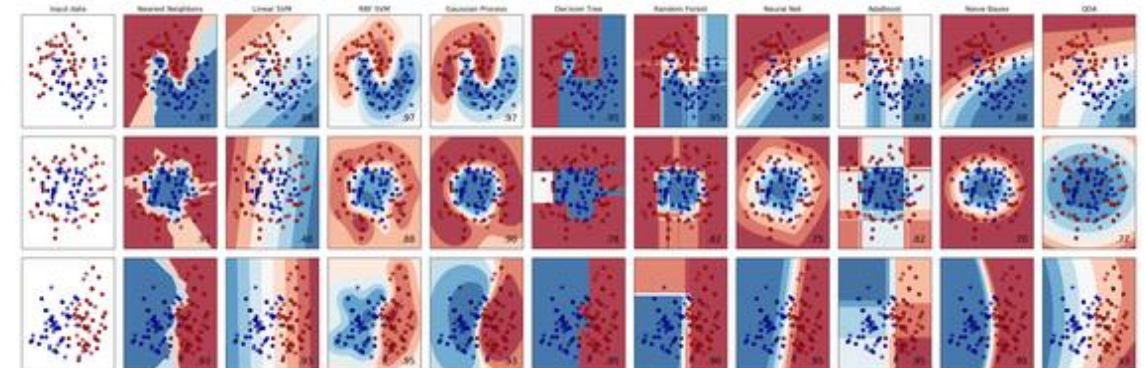
# Benchmarking the Supervised Learning Algorithms

## Description

- Get data from open sources
- Visualize the distribution of sample data
- Estimate accuracy of different classifiers
- Build ROC curves and calculate AUC
- Engineer the predictive features and estimate their predictive power

## Assessment criteria

- Linear discriminant: 20%
- Gaussian model: 20%
- Other models: 20%
- Data visualization charts: 20%
- ROC curve charts: 20%
- \* Extra: multi class championship 30%



# Unsupervised learning

## Description

- Get train/test data sample from open sources or generate it by yourself. Focus on low-dimension data, preferably 2 dimensions, 3-5 classes
- Visualize the distribution of sample data
- Visualize iteration of EM/K-Means algorithms
- Estimate the accuracy of classification

## Assessment criteria

- In-house unsupervised classifier: 40%
- Out-of-box classifier: 20%
- Data visualization charts: 20%
- Visualize the learning iterations: 20%
- \* Benchmark results with some clustering algorithms 30%



# Deep Learning

1. Feature extraction
2. Transfer learning
3. Fine-tuning
4. Object Detection
5. Texture Segmentation
6. Stereo Vision
7. OCR
8. Feature Visualization

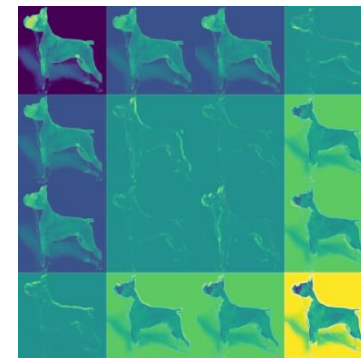
# Feature extraction

## Description

- Get train/test data sample from open sources. Try both scenarios when classless are well separable and are not.
- Run data thru various feature extractors and visualize the results, estimate the quality of clustering
- Visualize the intermediate data of CNN: filters, layers

## Assessment criteria

- Build own extractor : 40%
- Use existing feature extractors: 20%
- PCA and TSNE charts: 20%
- Benchmark quality of extractors: 20%
- \* Illustrate de-convolutions: 30%



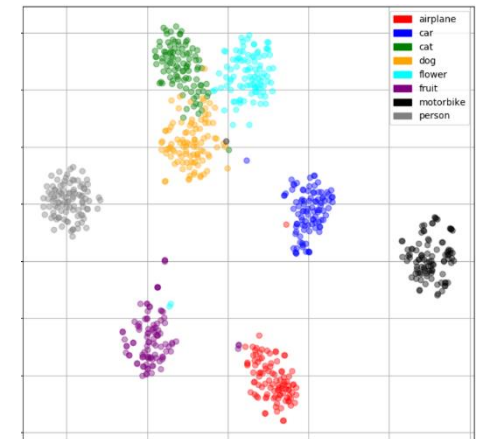
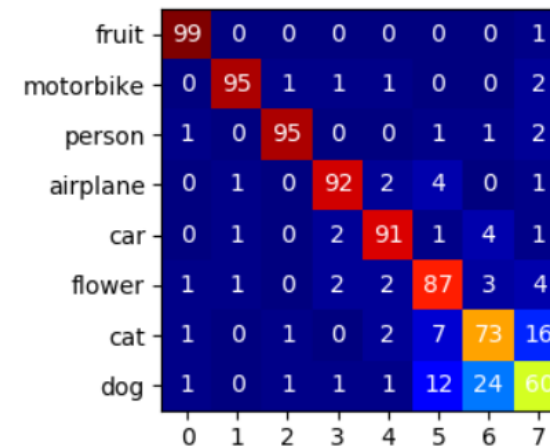
# Transfer learning

## Description

- Benchmark accuracy of multiple classifiers (SVM, RF, LM, etc) applied on top of several feature extractors (AlexNet, VGG, ImageNet etc)
- Research how accuracy very depending on the depth of the frozen layers, visualize this dependency
- Build/visualize confusion matrices
- Illustrate most typical classification errors

## Assessment criteria

- Build confusion matrices: 60%
- Build the accuracy trend: 20%
- Illustrate typical errors: 20%



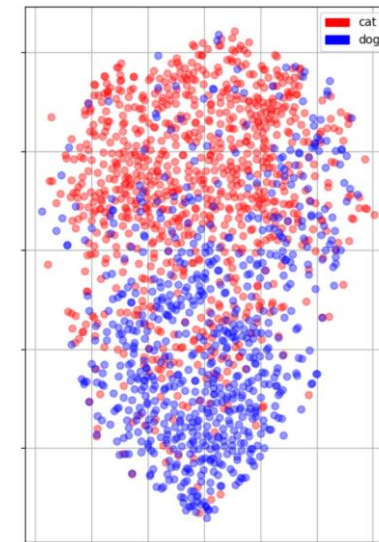
# Fine-tuning

## Description

- Benchmark accuracy of multiple CNNs finetuned on specific classes of images (e.g faces, cars)
- Illustrate how internal parameters were optimized to maximize the classification accuracy
- Build/visualize confusion matrices
- Illustrate most typical classification errors

## Assessment criteria

- Build confusion matrices: 60%
- Illustrate internal CNN data: 20%
- Illustrate typical errors: 20%





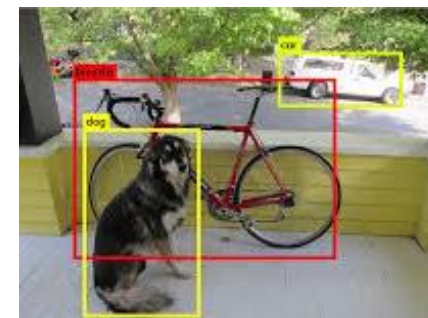
# Object detection

## Description

- Get train/test data sample from open sources or generate it yourself. Try on human faces, human shapes, cars, etc. Any type of object that you can capture with your laptop camera is preferable.
- Train the detector and estimate the accuracy against the test set
- Run detector in live mode, detect objects in video stream captured by a laptop camera

## Assessment criteria

- Build own detector: 40%
- Use pre-trained detector: 20%
- Precision-recall chart: 20%
- Detecting objects in live mode: 20%
- \* Benchmark results with some alternative detectors 30%





# Texture Segmentation

## Description

- Get train/test data sample from open sources or generate it yourself. Try on human faces, human shapes, cars, etc. Any type of object that you can capture with your laptop camera is preferable.
- Train the detector and estimate the accuracy against the test set
- Run detector in live mode, detect objects in video stream captured by a laptop camera

## Assessment criteria

- Build own detector: 40%
- Use pre-trained detector: 20%
- ROC chart for 2 classes problem: 20%
- Processing images in live mode: 20%
- \* Benchmark results with some alternative detectors 30%





## Assessment criteria

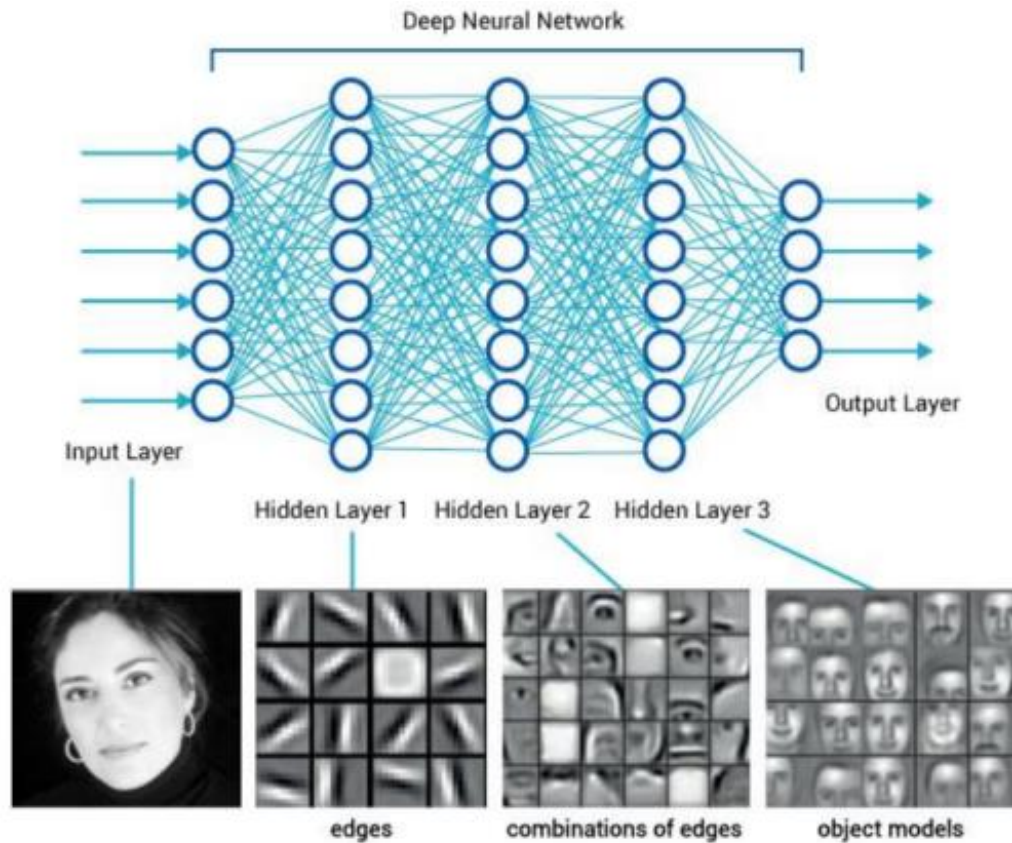
- BM approach: 25%
- SGBM: 25%
- SIFT approach: 25%
- Template Matching approach: 25%



## Assessment criteria

- Build confusion matrices: 60%
- Build the accuracy trend: 20%
- Illustrate typical errors: 20%

# Feature Visualization



## Assessment criteria

- Visualize convolution filters: 50%
- Visualize layers: 50%
- \* Overview existing frameworks for feature visualization 30%

# Tech Talks

Object detection

Style Transfer

GAN

3D object reconstruction

Image to text

Segmentation

Colorization

Denoising

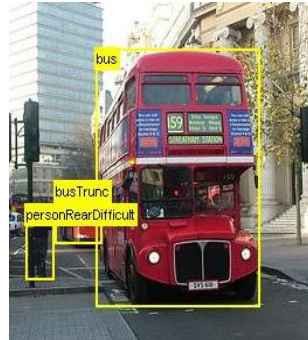
DeepFake

Pose Estimation

NLP

# Tech Talks

Object  
detection



Style  
transfer



GAN



3D object  
reconstruction

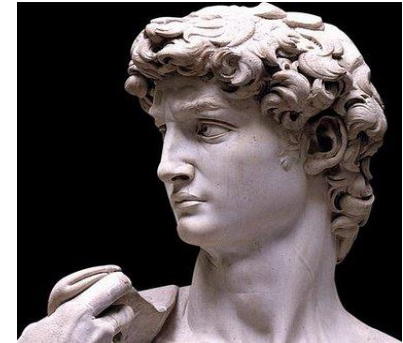
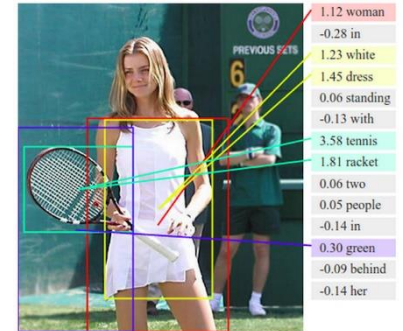
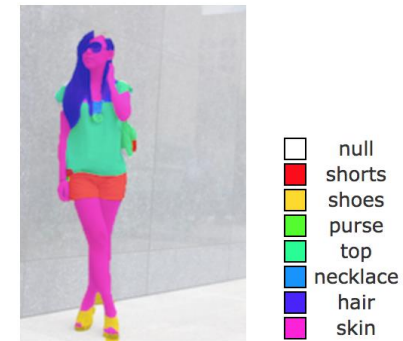


Image  
to text



Segmentation



# Tech Talks

Colorization



Denoising



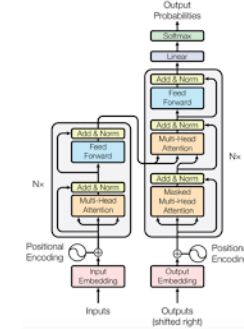
DeepFake



Pose estimation



NLP



More...

