

Practical AI: classification practicum

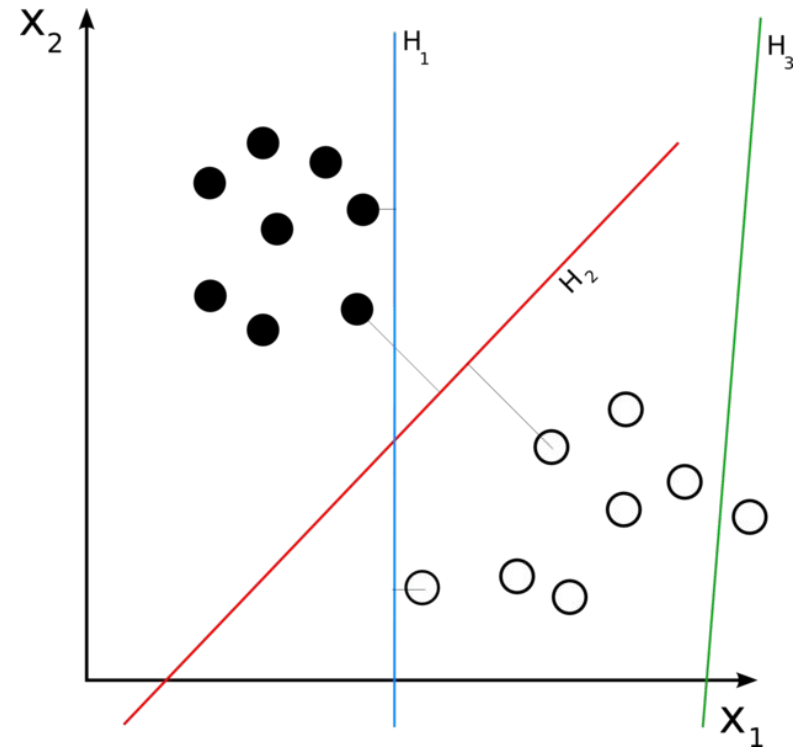
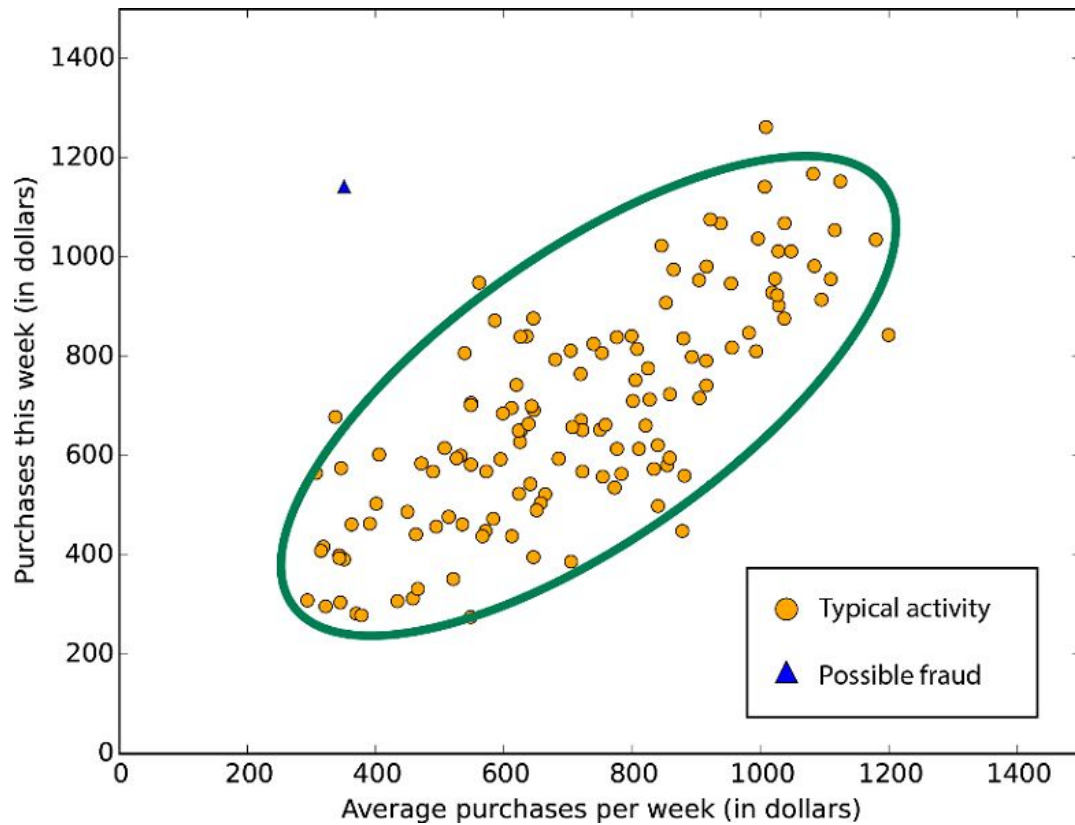
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Agenda

- One-class, binary classification
- Measuring quality:
 - accuracy,
 - precision,
 - recall,
 - F1,
 - ROC/AuC
- Linear model
 - Logistic Regression
- SVM
- ANN
- Multiclass approaches
 - One-vs-all
 - kNN classifier
 - ANN classifier

One-class and binary classification



How to measure

Accuracy - success / total

Precision - we care about **correct detections** (how many of detections are correct). successfully detected / total detections

Recall = TPR - FOMO :) We care about **missing** (we are penalized for missing). successfully detected / total positive

FPR - “panic rate”. How many of predictions made are done for “panic”.

false positive / total detections

F₁-score is an integral metric for precision-recall (harmonic mean).

$$F_1 = 2 * Pr * Re / (Pr + Re), \text{ lays in } [0..1]$$

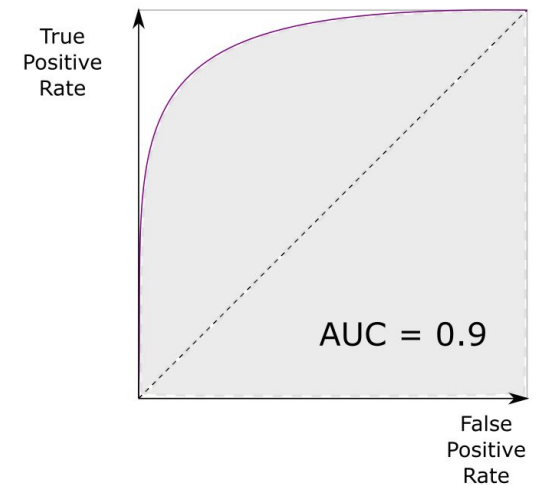
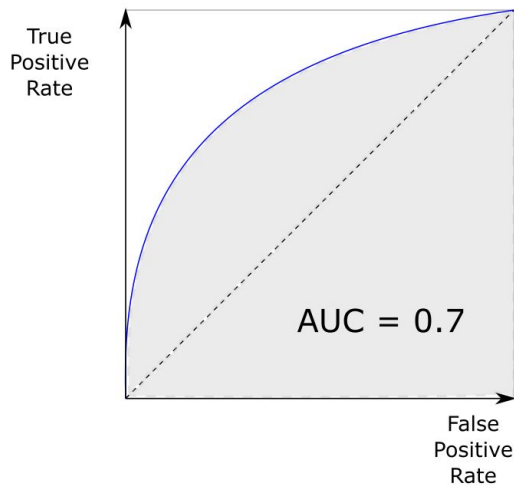
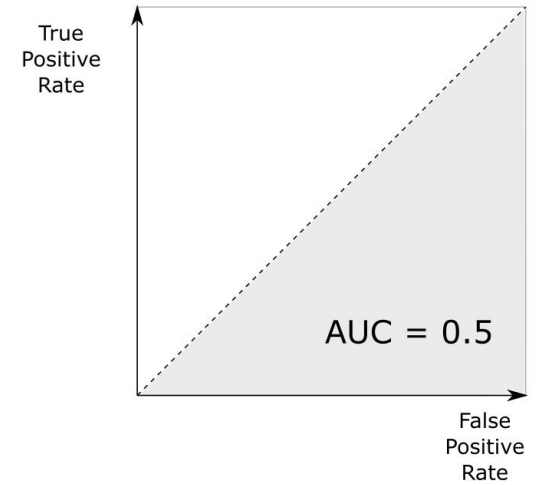
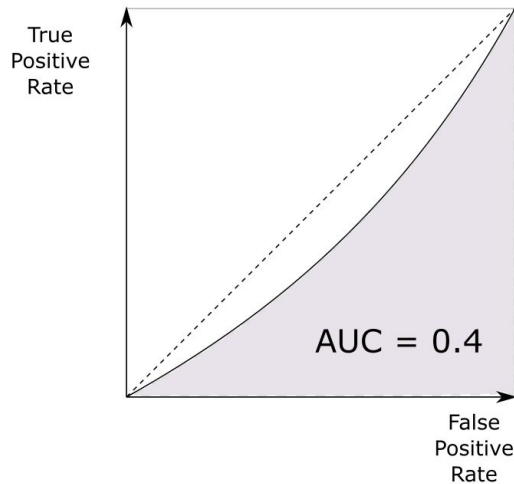
ROC, AuC

ROC - receiver operating characteristic.

A plot **TPR** vs **FPR**.

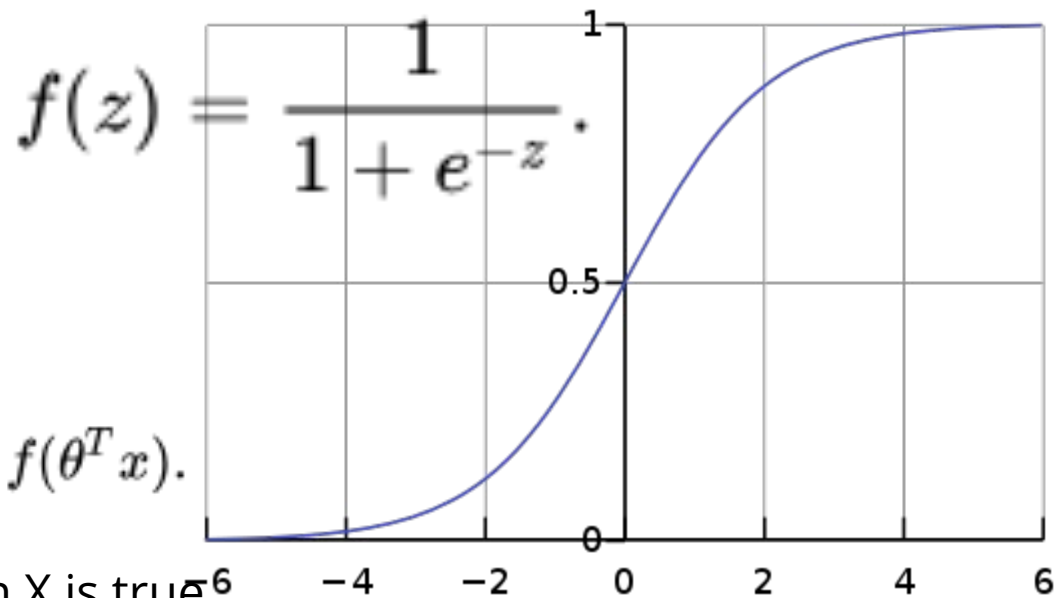
Idea: if we **tune a border** of classifier to bring **more true positive** detections, **how many false positives** it will bring along with?

Area under Curve - single number to describe ROC.



Logistic Regression

$$\mathbb{P}\{y = 1 \mid x\} = f(z),$$
$$\mathbb{P}\{y = 0 \mid x\} = 1 - f(z) = 1 - f(\theta^T x).$$



\mathbf{P} stands for probability that event X is true.

$f(z)$ stands for prediction of probability.

$$\mathbb{P}\{y \mid x\} = f(\theta^T x)^y (1 - f(\theta^T x))^{1-y}, \quad y \in \{0, 1\}.$$
$$\hat{\theta} = \operatorname{argmax}_{\theta} L(\theta) = \operatorname{argmax}_{\theta} \prod_{i=1}^m \mathbb{P}\{y = y^{(i)} \mid x = x^{(i)}\}.$$

We can, again train Logistic Regression with different algorithms.

[LR with X-val](#), [LR with SGD](#), [just LR](#)

Andrew Ng is cool:

<https://see.stanford.edu/materials/aimlcs229/cs229-notes1.pdf>

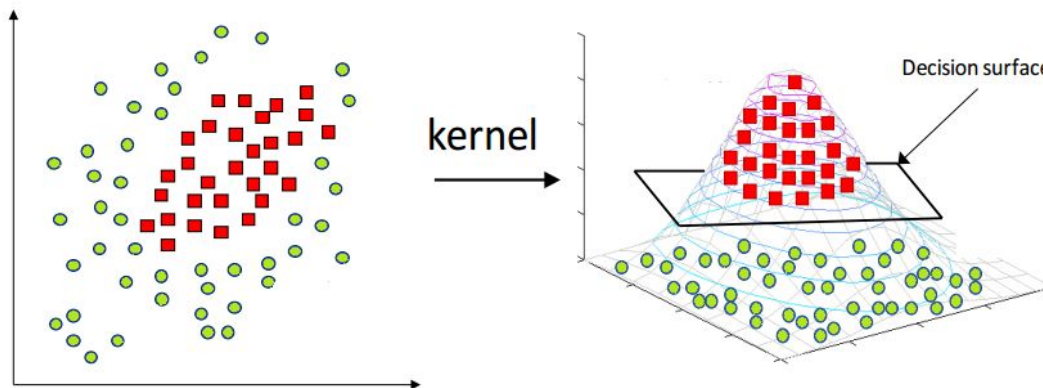
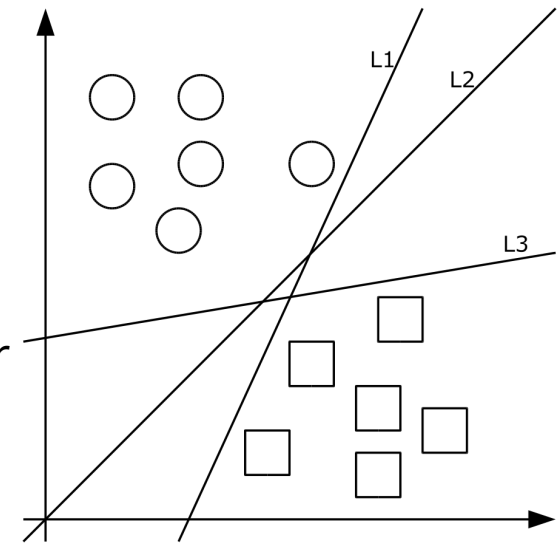
Support vector machine

General idea: classifier has better generalization power if it preserves biggest possible GAP between classes

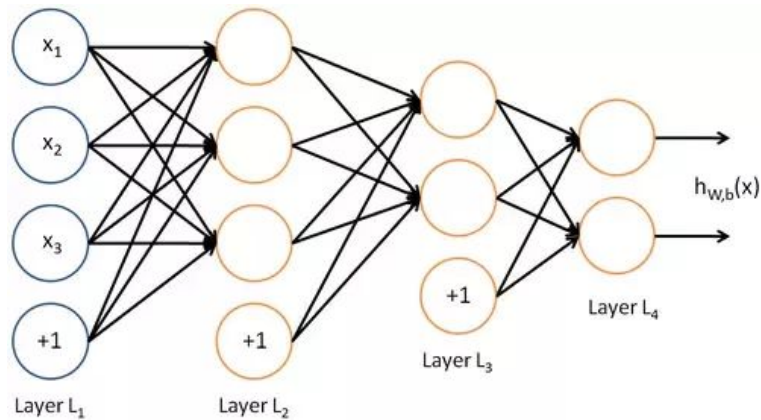
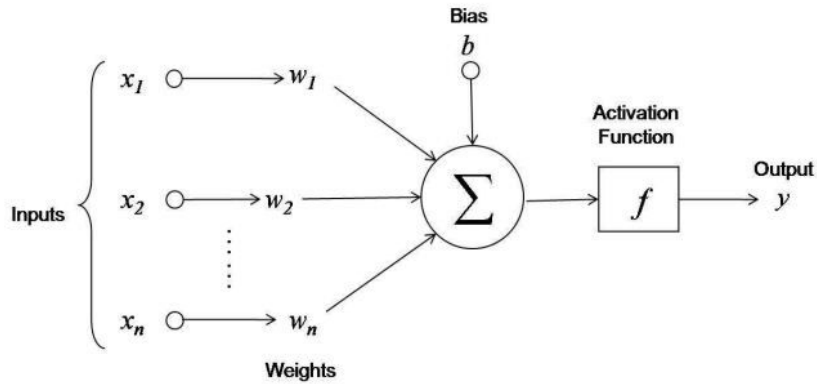
SVM itself is well known for involving **kernel trick**.

Idea: we have raw data, and instead of **computing non-linear features** we can introduce KERNEL - a **non-linear** function to compute sample **distance**. Thus, there IS another space, but it is IMPLICIT.

Then model builds borders for a class by summing kernelized distances.

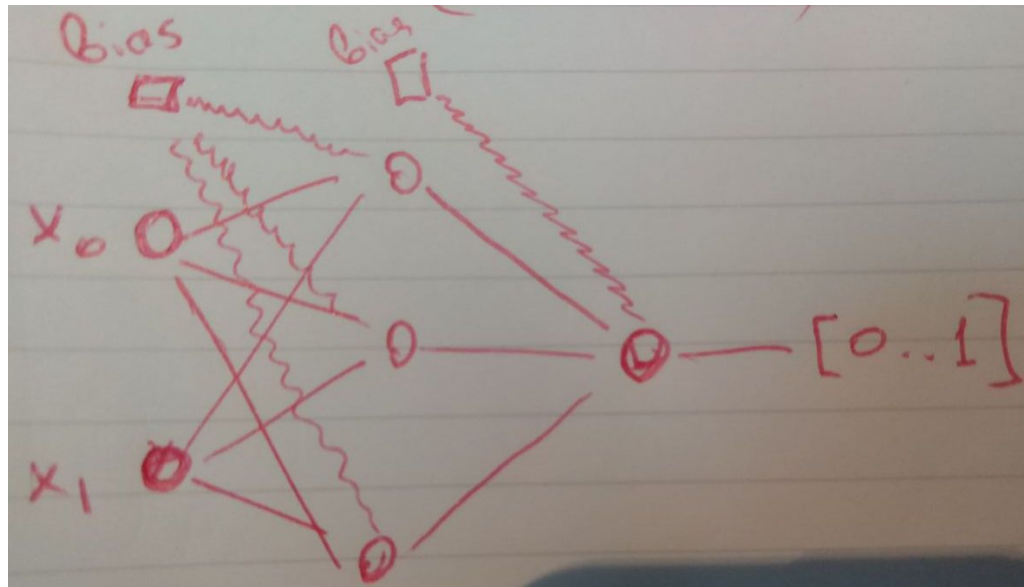


ANN: whiteboard time



Lab #1. ANN classifier lab

1. Consider and run (?) an example on cats-vs-dogs classifier
 - a. Study metrics. Do you understand them all?
 - b. Study parameters of MLP Classifier
2. Implement MLP Classifier for XOR function with given topology:
 - a. Generate dataset
 - b. Use ReLU activation function
 - c. Train
 - d. Measure accuracy, study quality report



Detectors

General idea: for a stream of data run unary or binary classifier.

For image:

- Build image pyramid
- Run for each patch of accepted size a classifier
- Where detects, create bounding box

Run [this example](#) to understand the idea. Try different scale factor





Multiclass classifiers

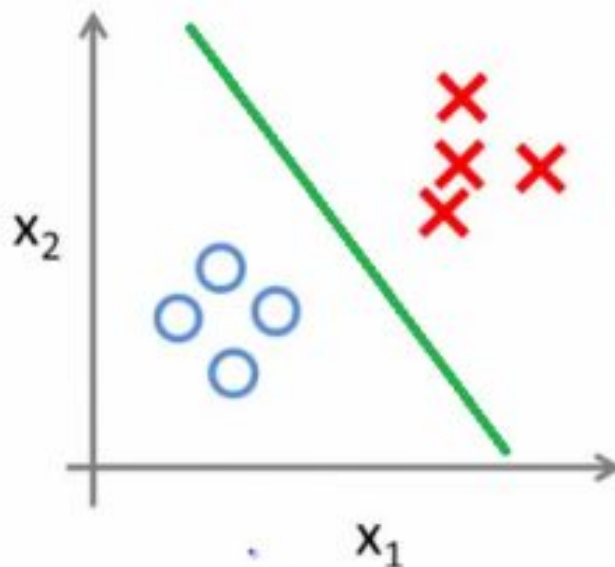


One-vs-all

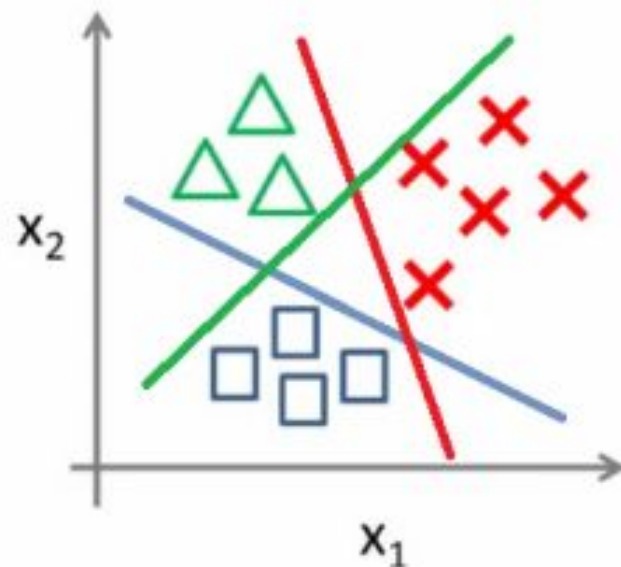
Use any binary classifier.

E.g. this is how it works for [LogisticRegression](#).

Binary classification:

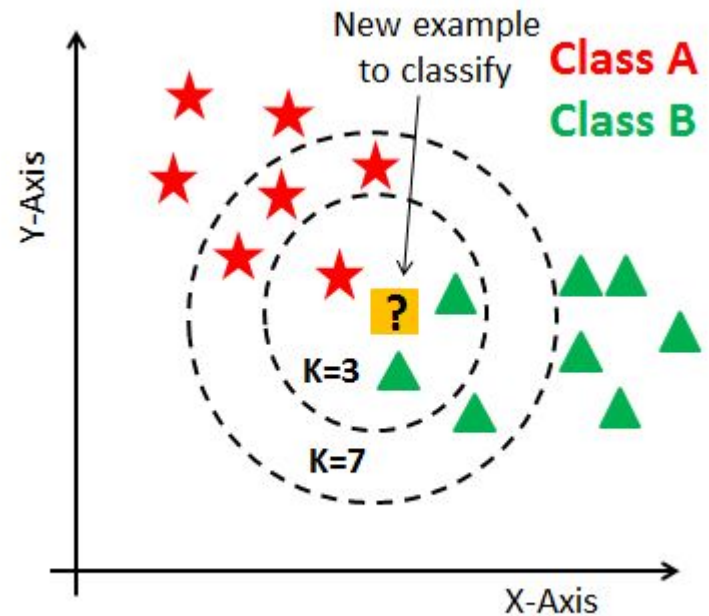


Multi-class classification:



kNN classifier

Distance function is important. Result will depend on data normalization. Use “pyfunc” if you want to use your own.



Lab #2

Study example with grid search for kNN.

Train and save the best **ANN classifier**.

Homework

Recognize hand-written number

[https://github.com/hsu-ai-course/hsu.ai/tree/
master/homeworks/12](https://github.com/hsu-ai-course/hsu.ai/tree/master/homeworks/12)