

Classical Data Analysis



Master in Big Data Solutions 2017-2018

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Session 2 - Classification

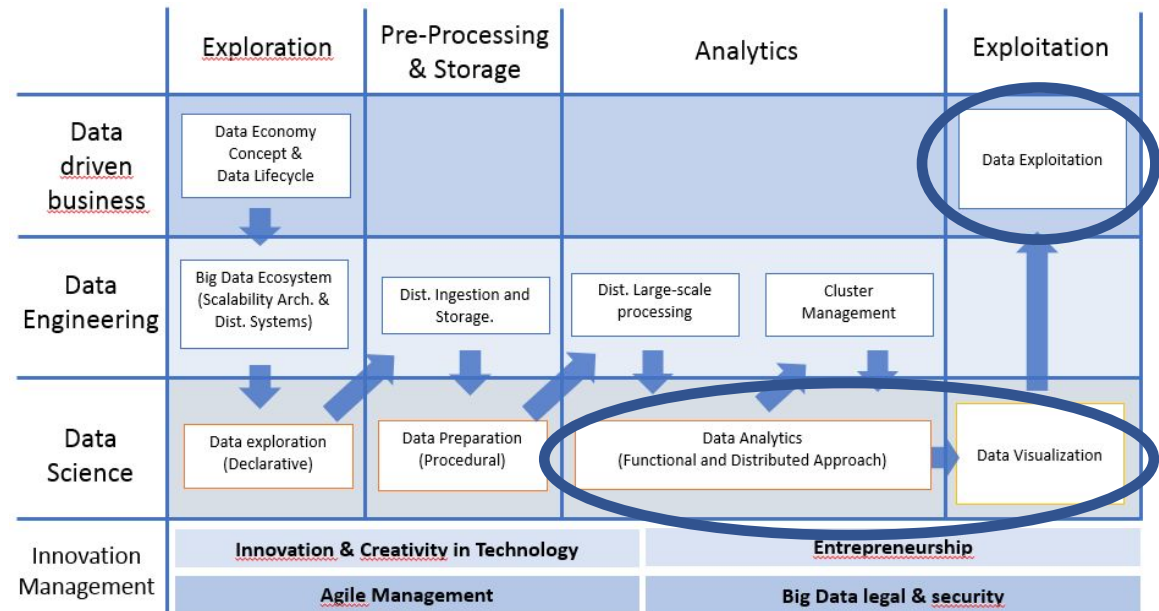
Support Vector Machine (SVM)

Sara Hajian

What we will learn

Session1: Classification

Support Vector Machine (SVM)

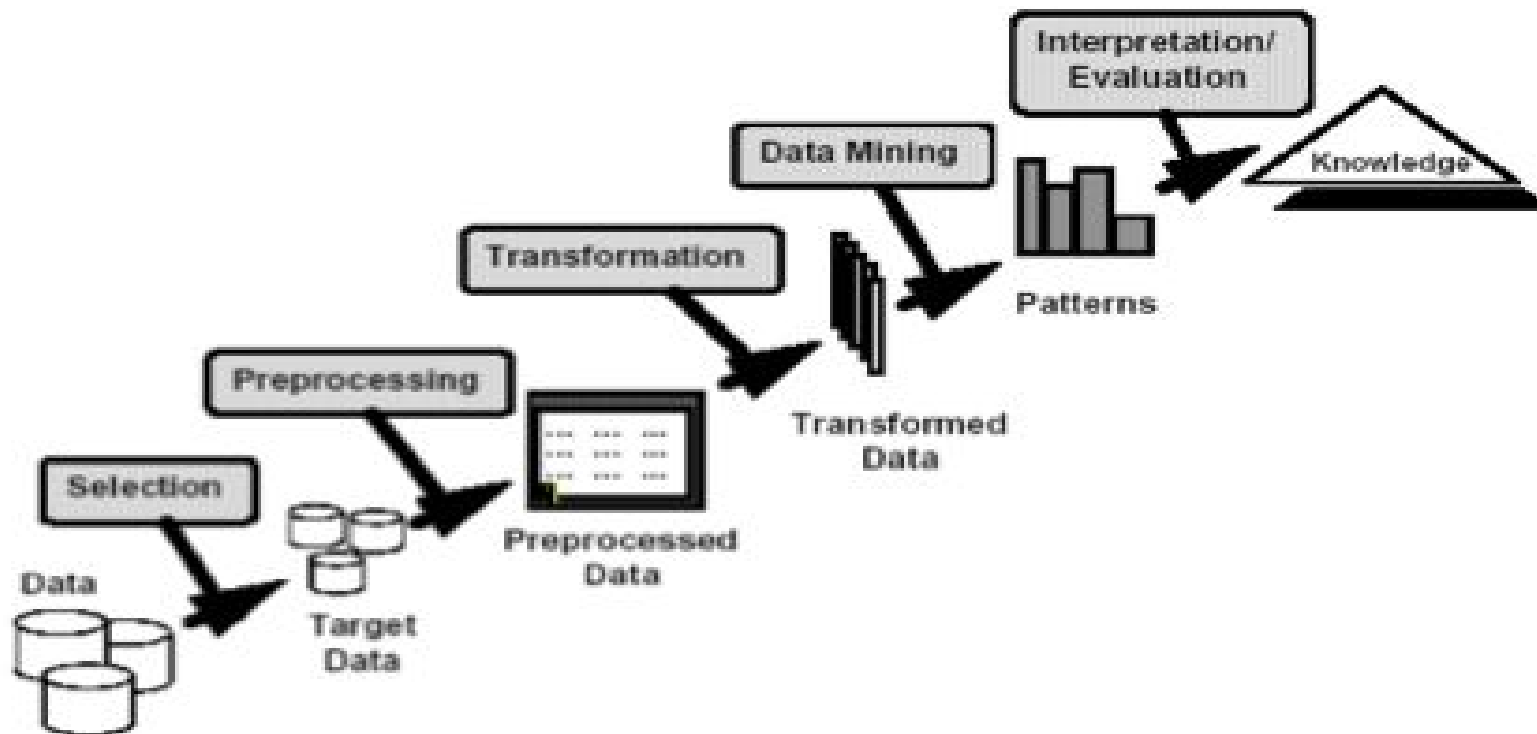


We will learn how to:

- What is data mining, its origin, different types of data mining algorithms
- What is classification, and know about its practical applications
- What is Support Vector Machines and how it works

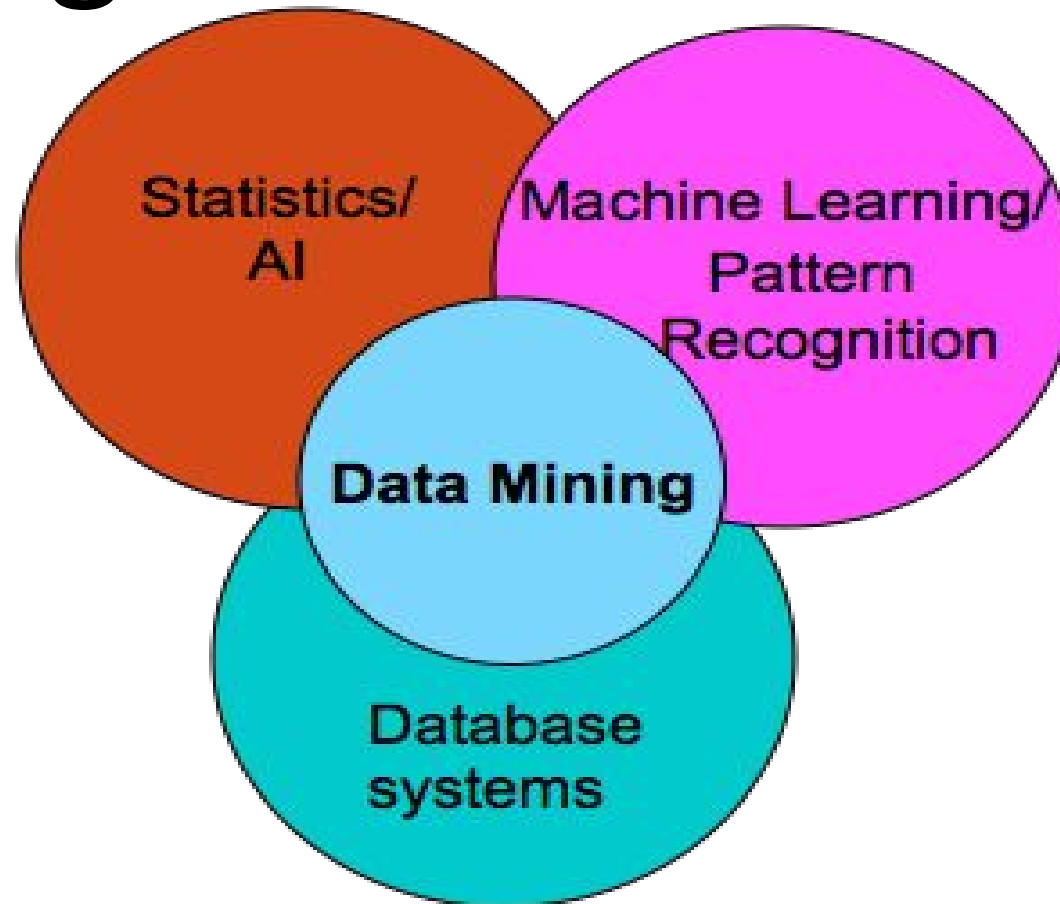
What is data mining

Non-trivial extraction of implicit, previously unknown and potentially useful information from data (i.e., **discovery of meaningful patterns**)



Tan, Pang-Ning. *Introduction to data mining*. Pearson Education India, 2006.

Origins of data mining



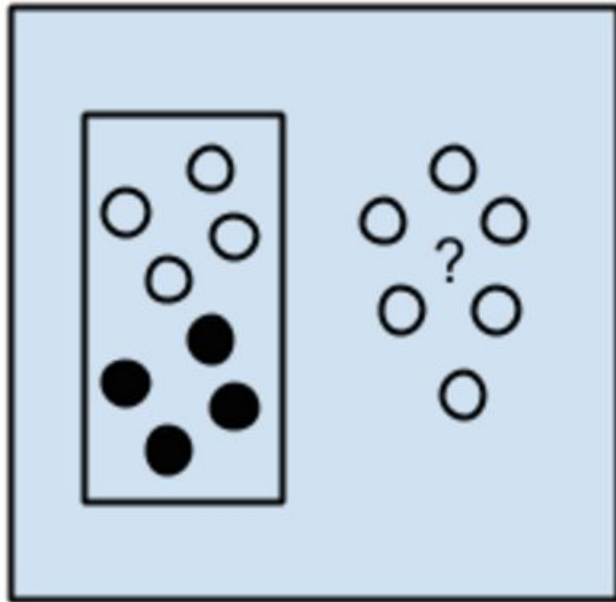
Tan, Pang-Ning. *Introduction to data mining*. Pearson Education India, 2006.

Data mining algorithms

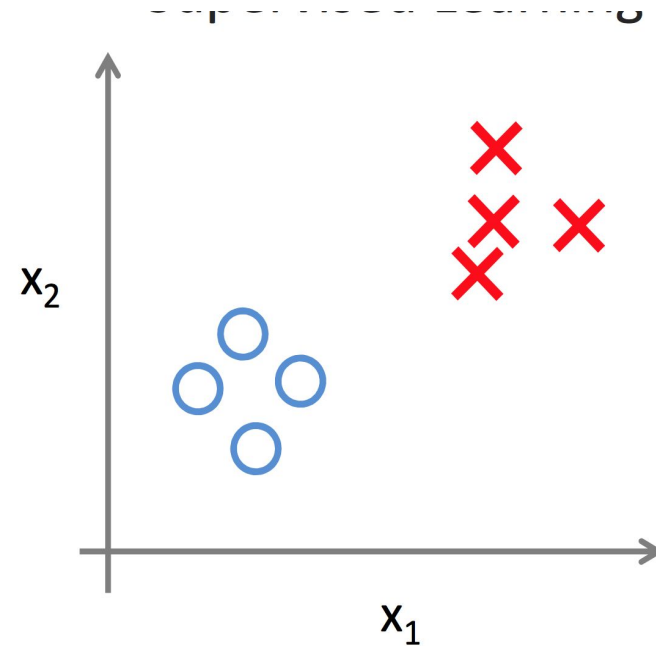
Supervised: Classification and regression

Unsupervised: Clustering

Data mining algorithms

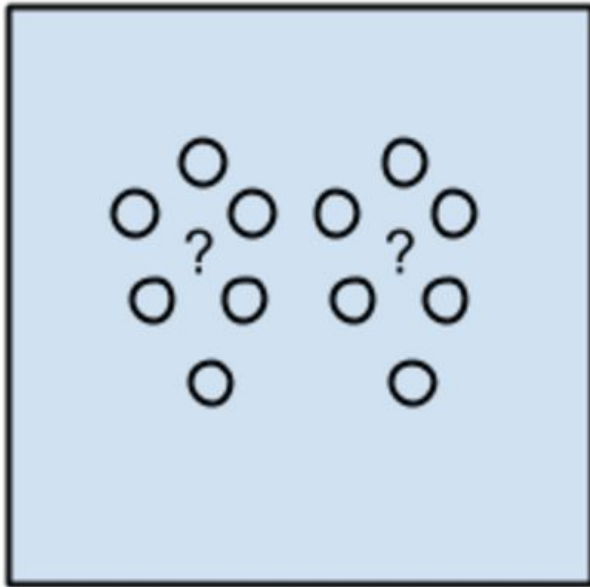


Supervised Learning
Algorithms

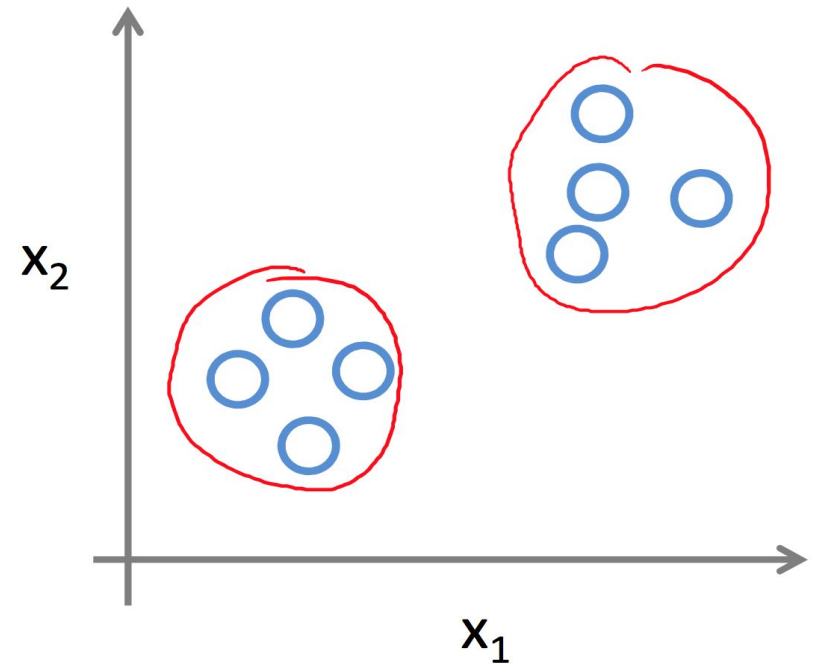


Examples: regression and classification

Data mining algorithms



Unsupervised Learning
Algorithms



Examples: clustering

Classical Data analysis: Curriculum

Part I: Regression

Linear regression

Logistic regression

Part II: Classification

Support vector machines (SVM)

Decision trees

Ensemble methods

Clustering

Supervised learning algorithms

Classical Data analysis: Curriculum

Part I: Regression

Linear regression

Logistic regression

Part II: Classification

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Classical Data analysis: Curriculum

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Unsupervised learning
algorithms

Classification: definition

Given a collection of records (*training set*)

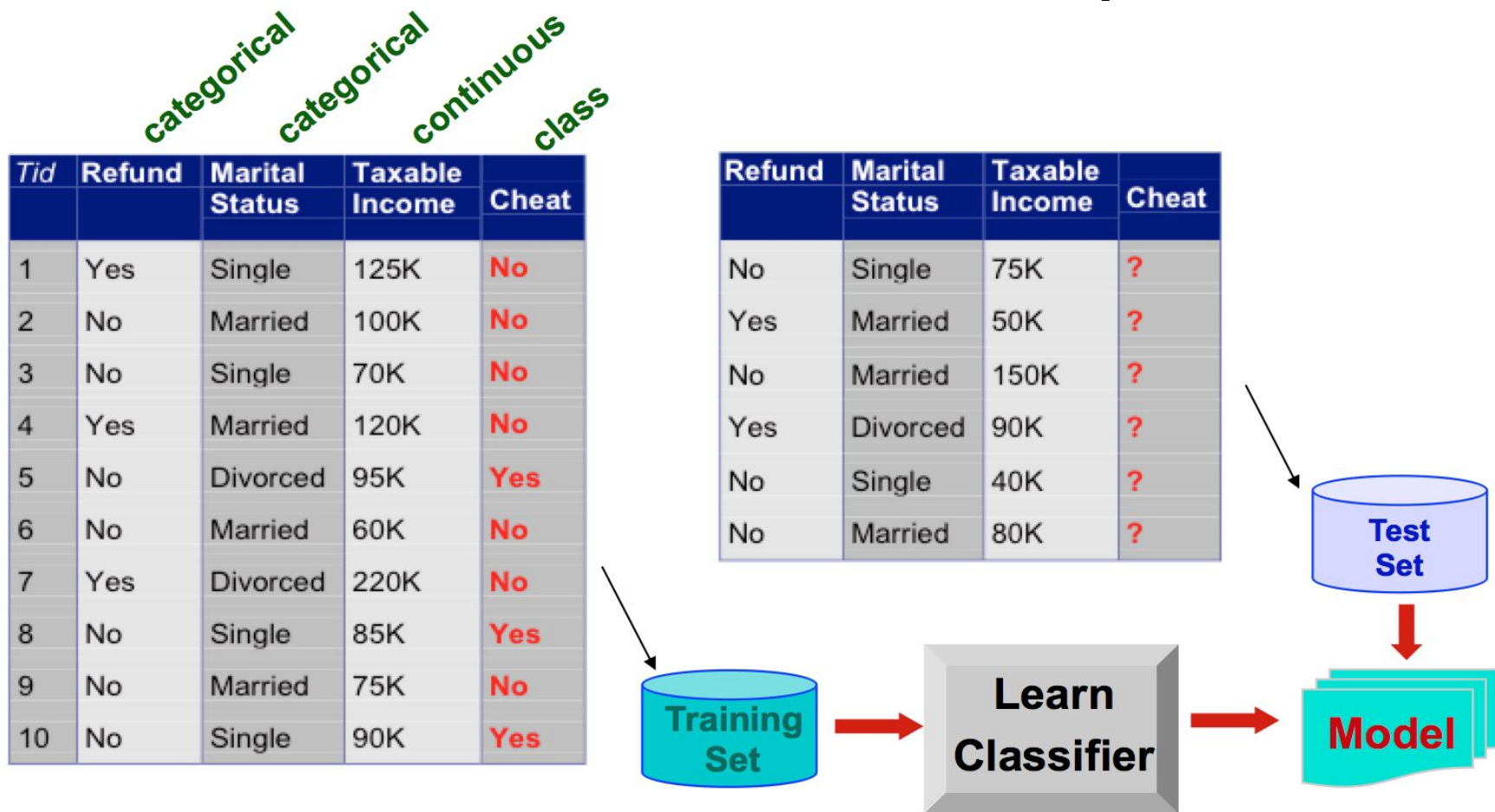
- Each record contains a set of *attributes*, one of the attributes is the *class*.

Find a *model* for class attribute as a function of the values of other attributes.

Goal: previously unseen records should be assigned a class as accurately as possible.

- A *test set* is used to determine the accuracy of the model.
Usually, the given data set is divided into training and test sets, with training set used to build the model and test set used to validate it.

Classification: example



Tan, Pang-Ning. *Introduction to data mining*. Pearson Education India, 2006.

Classification: applications

Direct Marketing

- Goal: Reduce cost of mailing by *targeting* a set of consumers likely to buy a new cell-phone product.

Fraud Detection

- Goal: Predict fraudulent cases in credit card transactions.

Customer Attrition/Churn:

- Goal: To predict whether a customer is likely to be lost to a competitor.

Classification: algorithms

Support vector machines (SVM)

Decision trees

Ensemble methods

Rule-based Methods

Neural Networks

Bayesian algorithms such as Naïve Bayes

Instance-based algorithms such kNN

Deep learning

Classification: algorithms

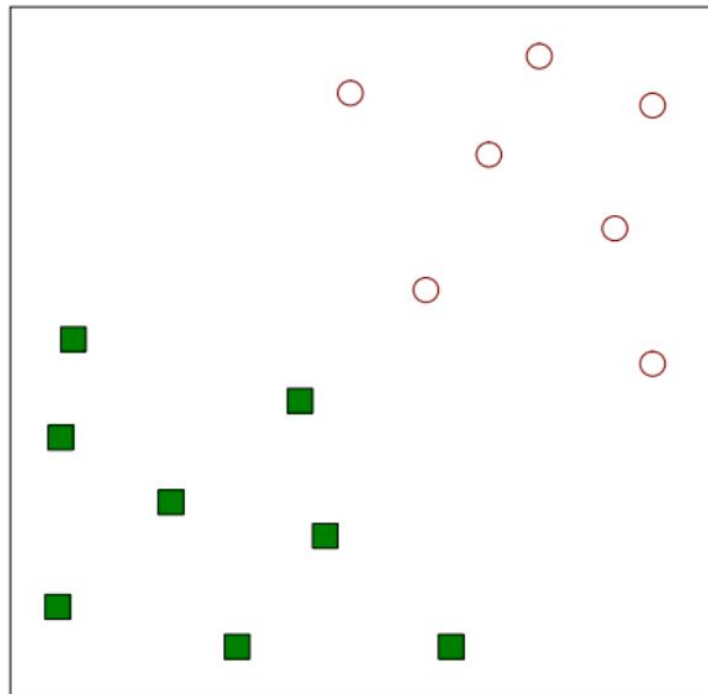
Support vector machines (SVM)

Applications: handwritten digit recognition, text categorization, image classification

Advantages: works well with high-dimensional data
works both with numerical and categorical data

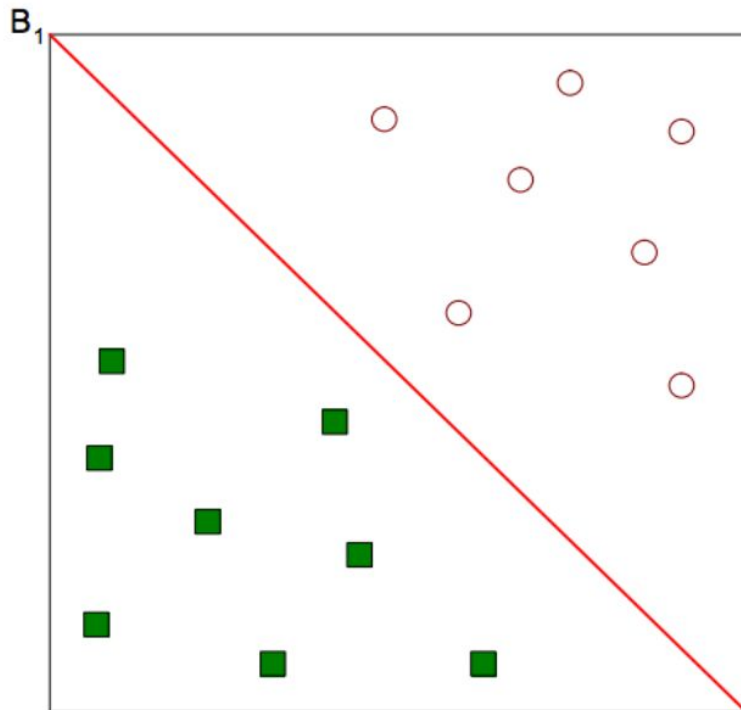
Support Vector Machines (SVM)

Find a linear hyperplane (**decision boundary**) that will separate the data

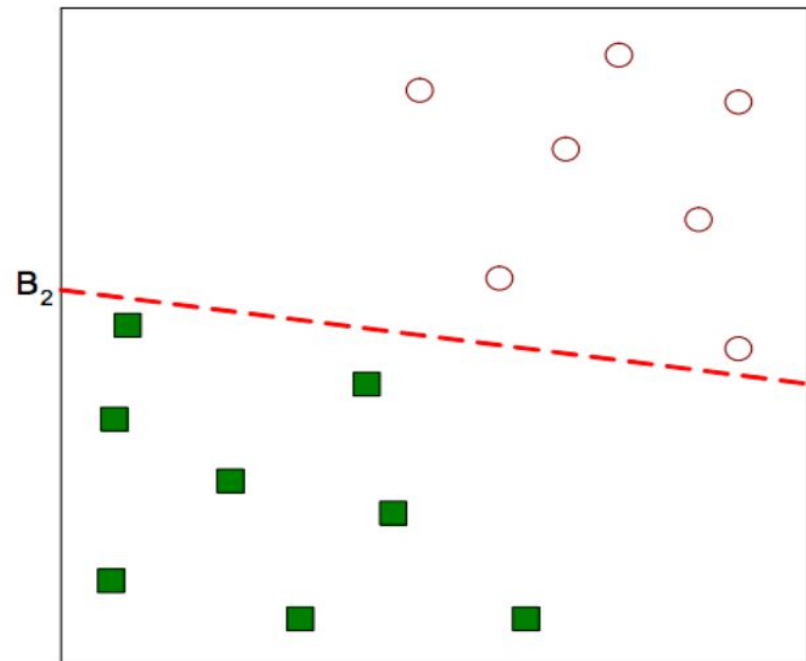


Support Vector Machines (SVM)

One possible solution

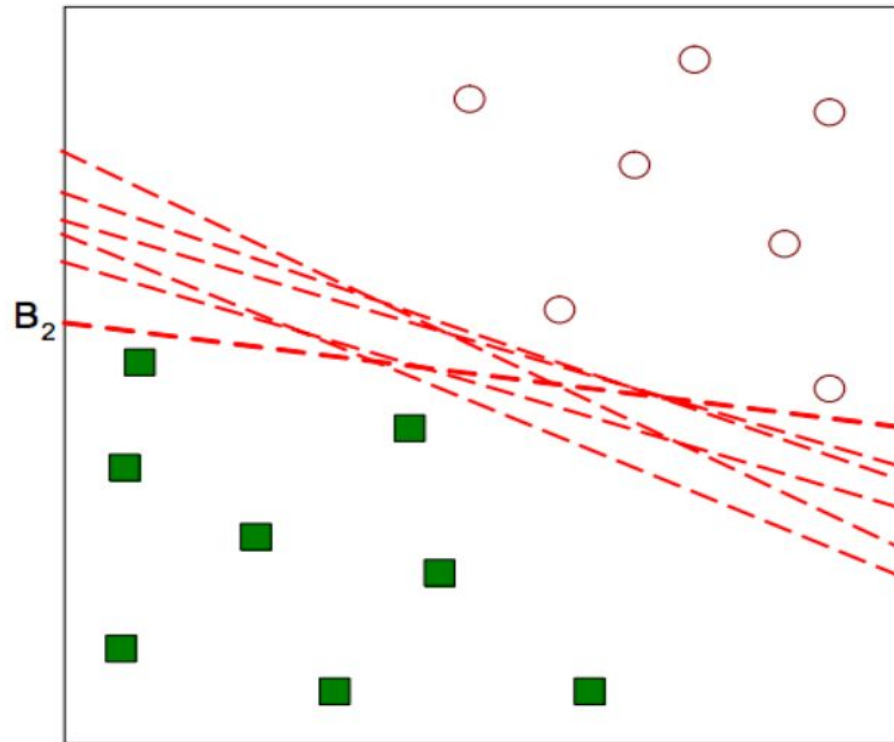


Another possible solution



Support Vector Machines (SVM)

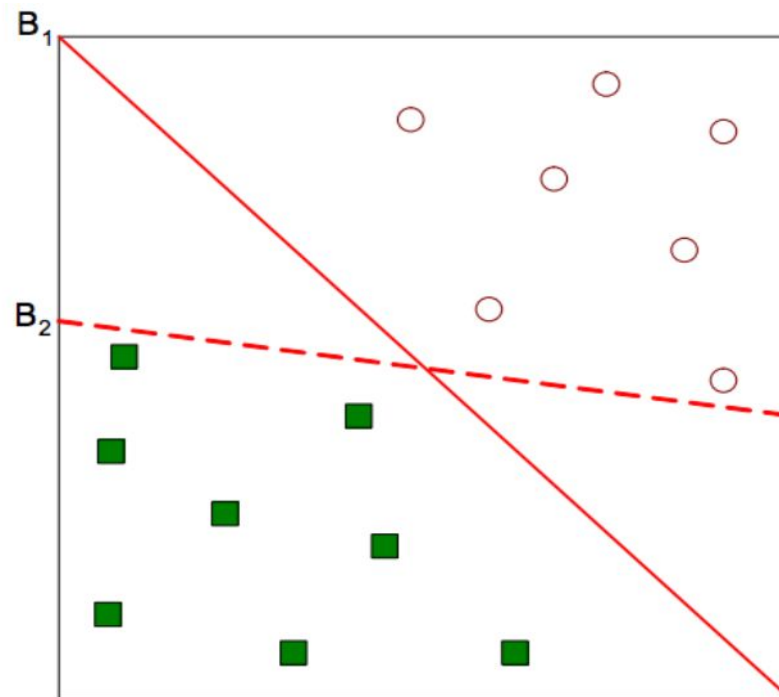
Other possible solutions



Support Vector Machines (SVM)

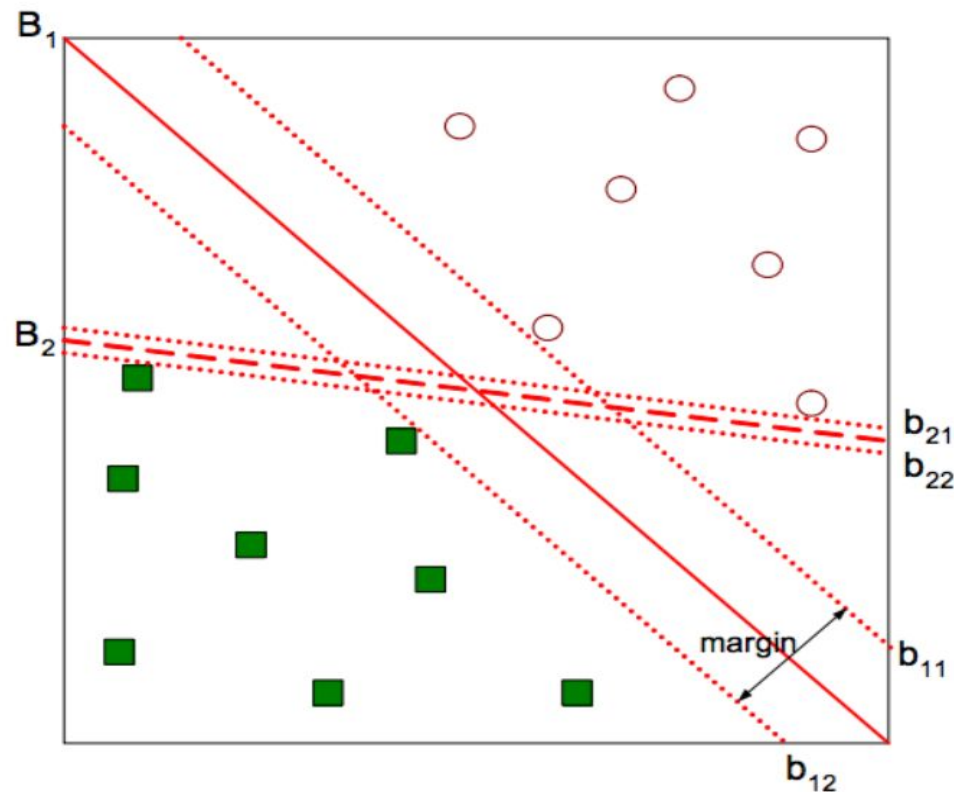
Which one is better? B1 or B2?

How do you define better?



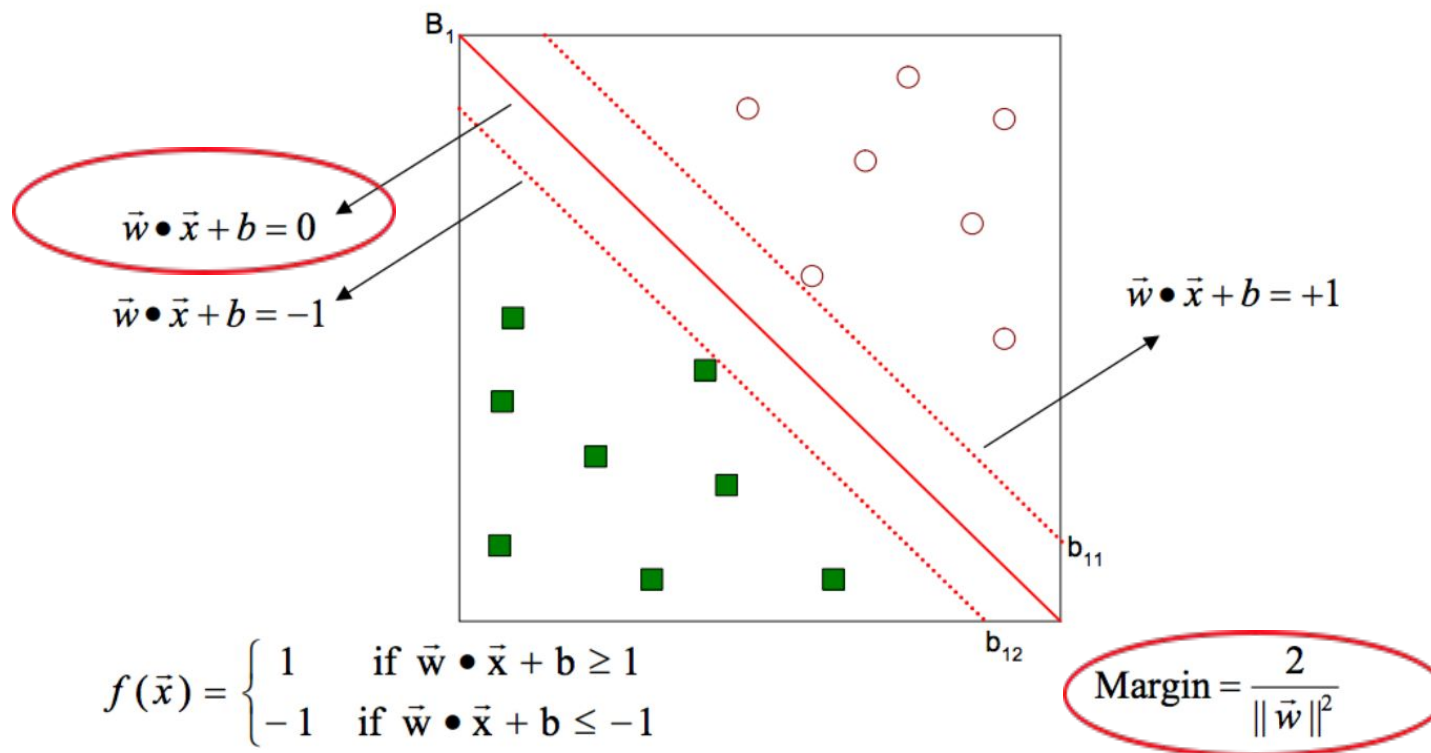
Support Vector Machines (SVM)

Find hyperplane **maximizes the margin** => B1 is better than B2



Support Vector Machines (SVM)

linear decision boundary and margin of a linear classifier



Support Vector Machines (SVM)

Learning the parameters of the decision boundary

We want to maximize: $\text{Margin} = \frac{2}{\|\vec{w}\|^2}$

– Which is equivalent to minimizing: $L(w) = \frac{\|\vec{w}\|^2}{2}$

– But subjected to the following constraints:

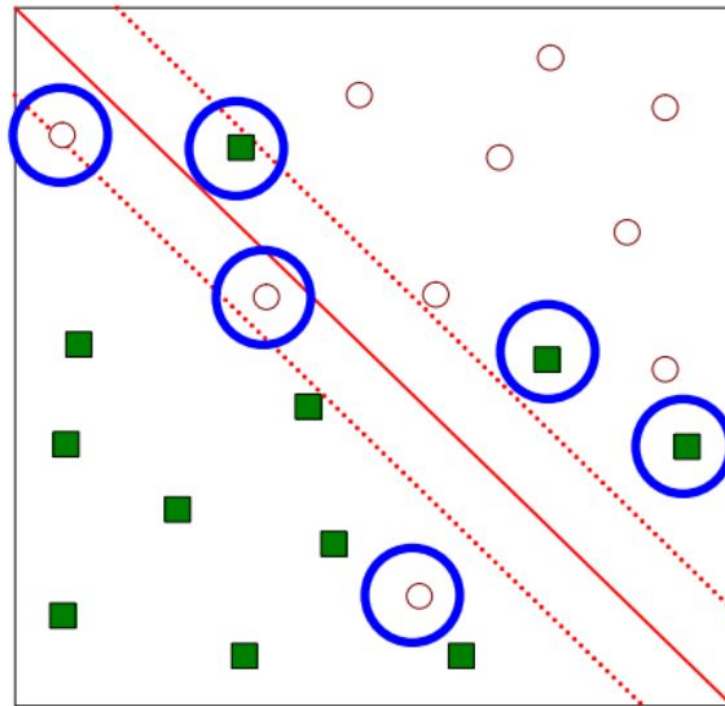
$$f(\vec{x}_i) = \begin{cases} 1 & \text{if } \vec{w} \bullet \vec{x}_i + b \geq 1 \\ -1 & \text{if } \vec{w} \bullet \vec{x}_i + b \leq -1 \end{cases}$$

◆ This is a constrained optimization problem

– Numerical approaches to solve it (e.g., quadratic programming)

Support Vector Machines (SVM)

What is the problem is not linearly separable?



Support Vector Machines (SVM)

The learning algorithm must consider the **trade-off between the width of margin and the number of training errors**

– Introduce slack variables

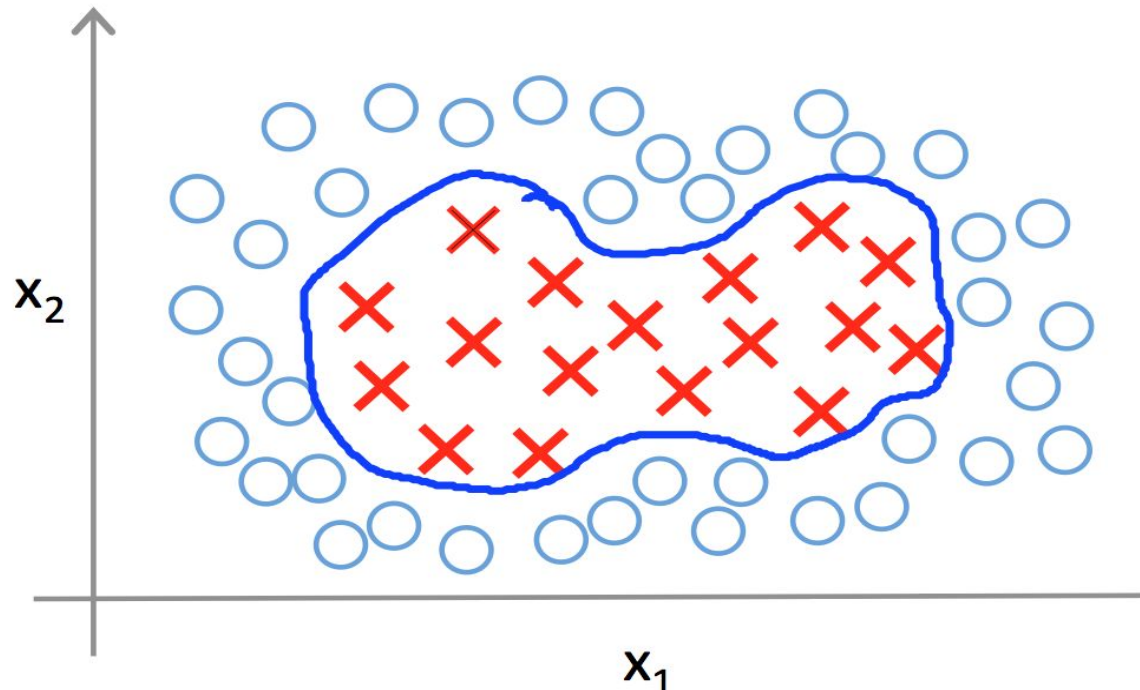
◆ Need to minimize:
$$L(w) = \frac{\|\vec{w}\|^2}{2} + C \left(\sum_{i=1}^N \xi_i^k \right)$$

◆ Subject to:

$$f(\vec{x}_i) = \begin{cases} 1 & \text{if } \vec{w} \bullet \vec{x}_i + b \geq 1 - \xi_i \\ -1 & \text{if } \vec{w} \bullet \vec{x}_i + b \leq -1 + \xi_i \end{cases}$$

Support Vector Machines (SVM)

What if the decision boundary is not linear?



Non-linear SVM

Support Vector Machines (SVM)

Key points and characteristics

Individual assignment

Choose one of the following problem scenarios and describe your approach to solve them:

Scenario 1 : Direct Marketing

—Goal: Reduce cost of mailing by *targeting* a set of consumers likely to buy a new cell-phone product.

Scenario 2: Fraud Detection

—Goal: Predict fraudulent cases in credit card transactions.

Scenario 3: Customer Attrition/Churn

—Goal: To predict whether a customer is likely to be lost to a competitor.

Next class

Support vector machines with Python:

- load the breast cancer dataset from Scikit Learn
- split the data into train and test set
- train a support vector classifier, use the trained model for prediction of test set
- evaluation of the SVM model and predictions

Resources

- Tan, Pang-Ning. *Introduction to data mining*. Pearson Education India, 2006.
- Friedman, Jerome, Trevor Hastie, and Robert Tibshirani. *The elements of statistical learning*. Vol. 1. New York: Springer series in statistics, 2001.



Thank you
Barcelona, 2017