



AARHUS  
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# Optimization and Data Analytics

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# Optimization and Data Analytics

## Course outline, formal stuff

- › Prerequisite
- › Lectures
- › Homework
- › Textbook
- › Blackboard

# What is this and Goals?

## Some definitions

- › **Data Analytics** is the discovery, interpretation and communication of meaningful patterns in data
- › **Optimization** is the process for *selecting the best element* (with regard to some criterion) from some set of available alternatives

## Some course goals

- › Increased mathematical skills
- › A broad knowledge of the classical and modern mathematical optimization techniques
- › Understand the basic Machine Learning models
- › Get experience with Data Analysis through project work
- › Understand data visualization concepts
- › Get introduced to Scientific Writing

machine learning

optimization

communication,  
visualization

# today: data overload





## TIAA-CREF-Stock repaired.xlsm - Microsoft Excel

anything interesting happening here?

---

“interpretation” = analysis  
= finding patterns

what are examples of **data**?

what are examples of **patterns**?





video



audio



satellite



images



stocks



movement

computers can help!

with lots of examples computers can build “pattern finder”





machine learning

optimization

communication,  
visualization

models can “find patterns”

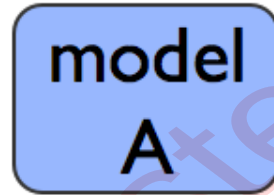
machine learning: building models from lots of examples

model  
A

models can “find patterns”



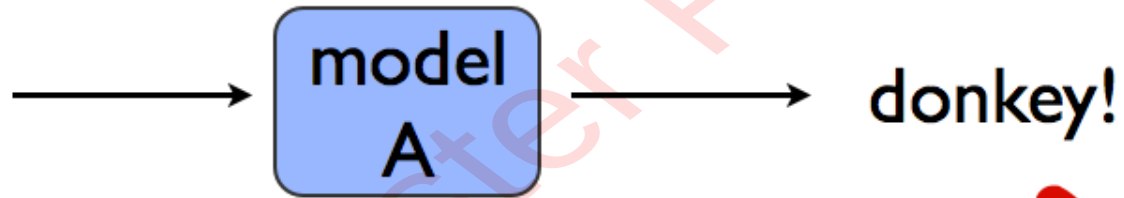
image



models can “find patterns”



image





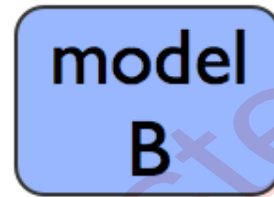
models can “find patterns”

model  
B

models can “find patterns”



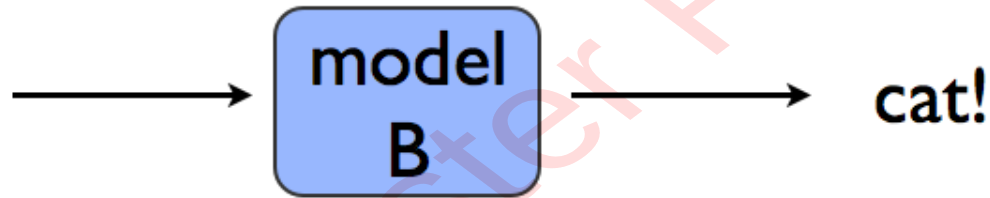
image



models can “find patterns”



image



machine learning

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imagine set of **all possible** models

model  
A

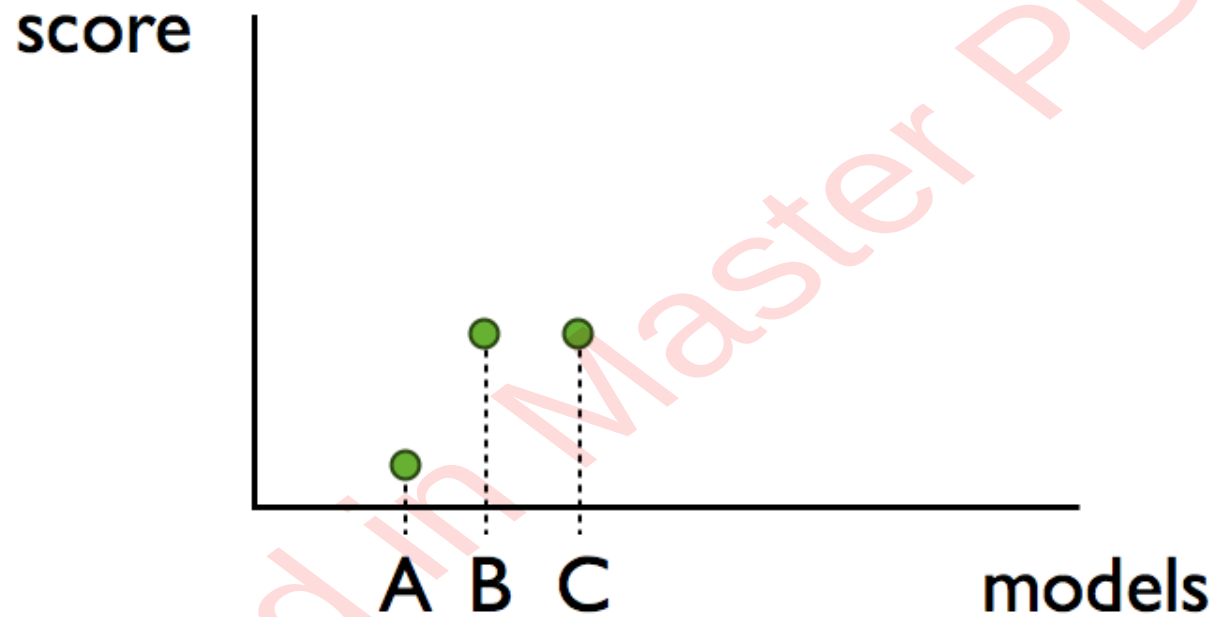
model  
B

model  
C

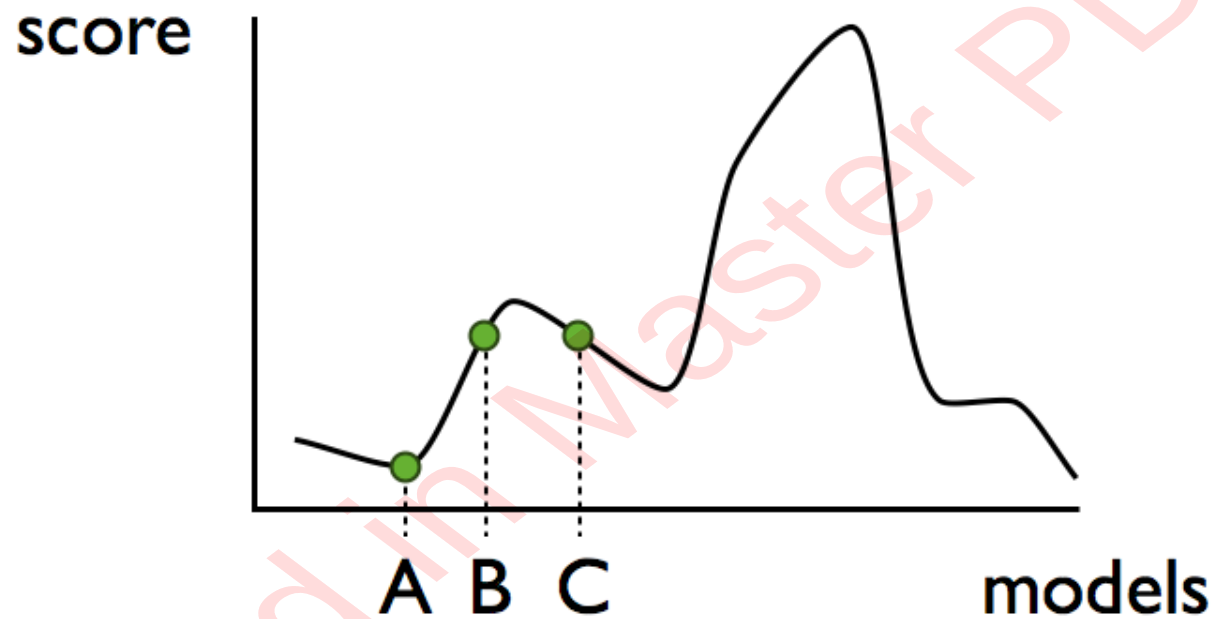
...

let's give each model a **score**

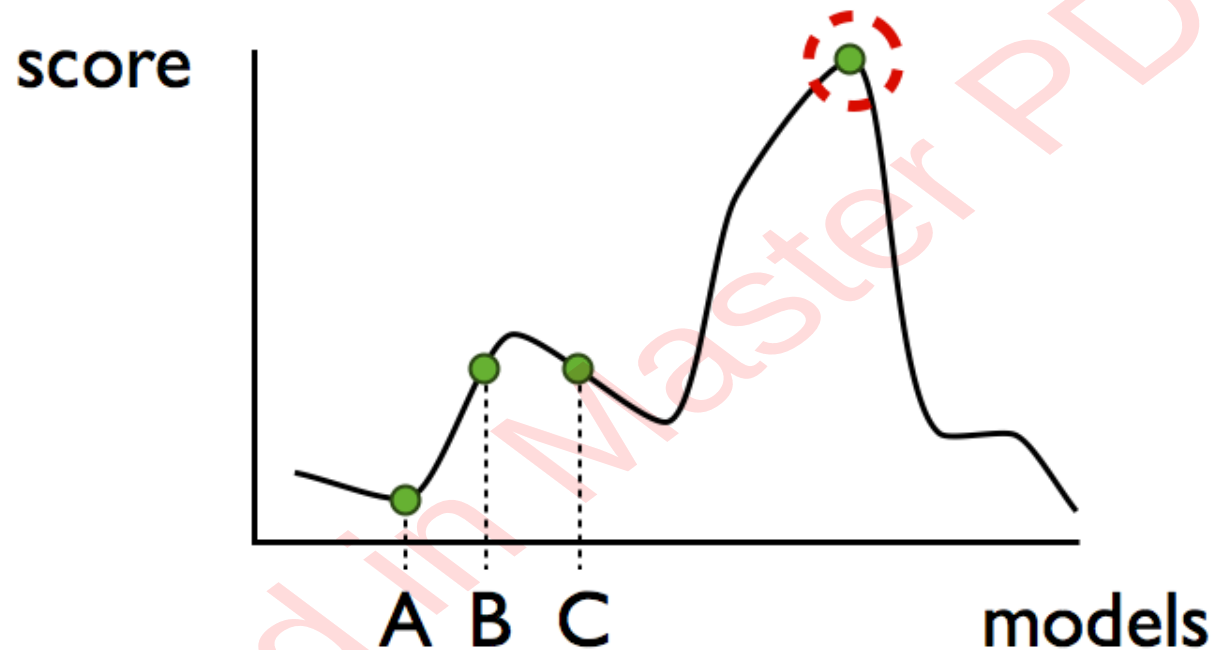
we want **best model** we can find



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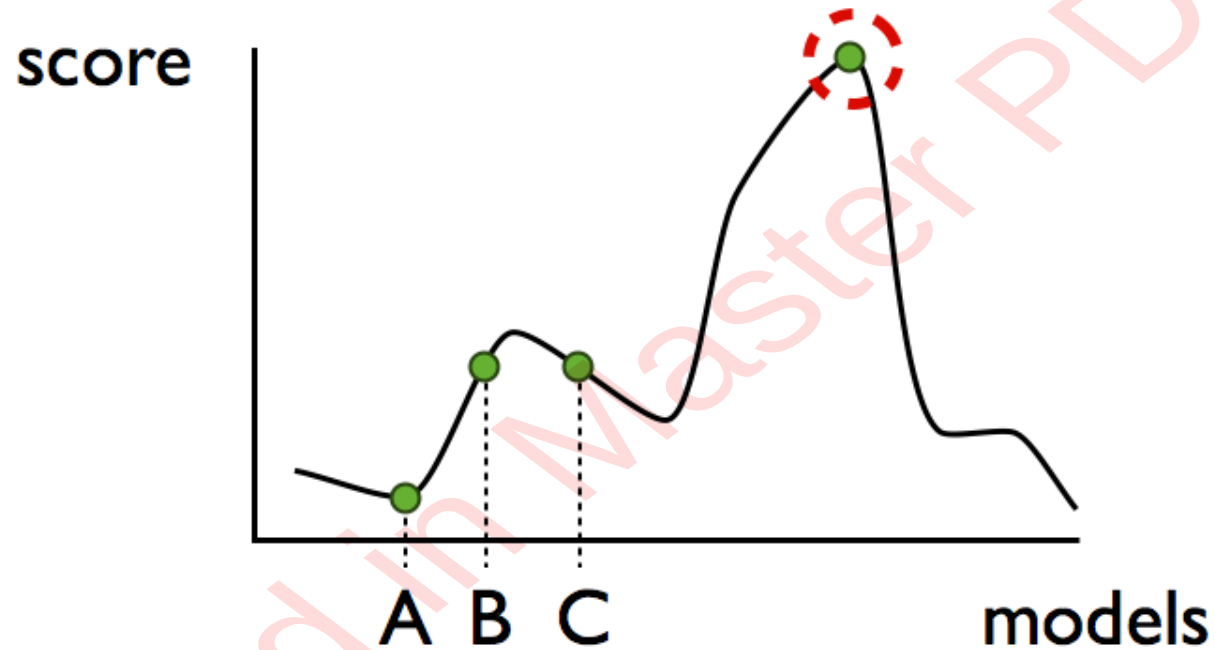
we want **best model** we can find



optimization: find model that maximises score

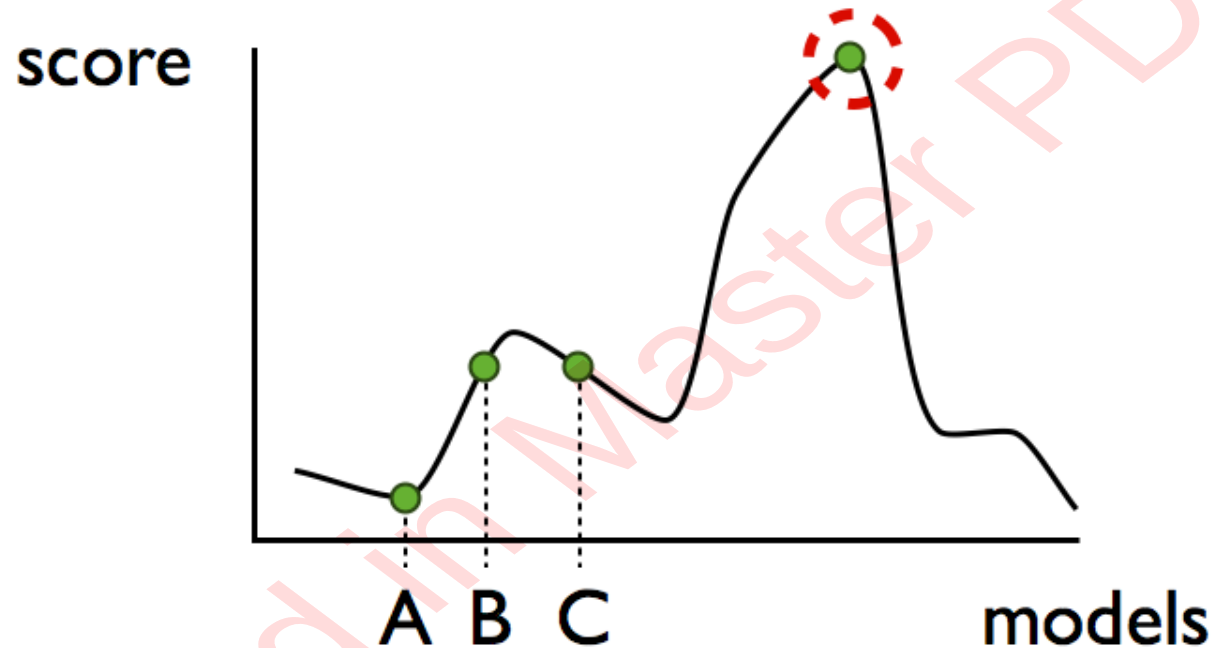


how did I explain this concept to you?



how did I explain this concept to you?

I **visualized** set of models as a **chart**

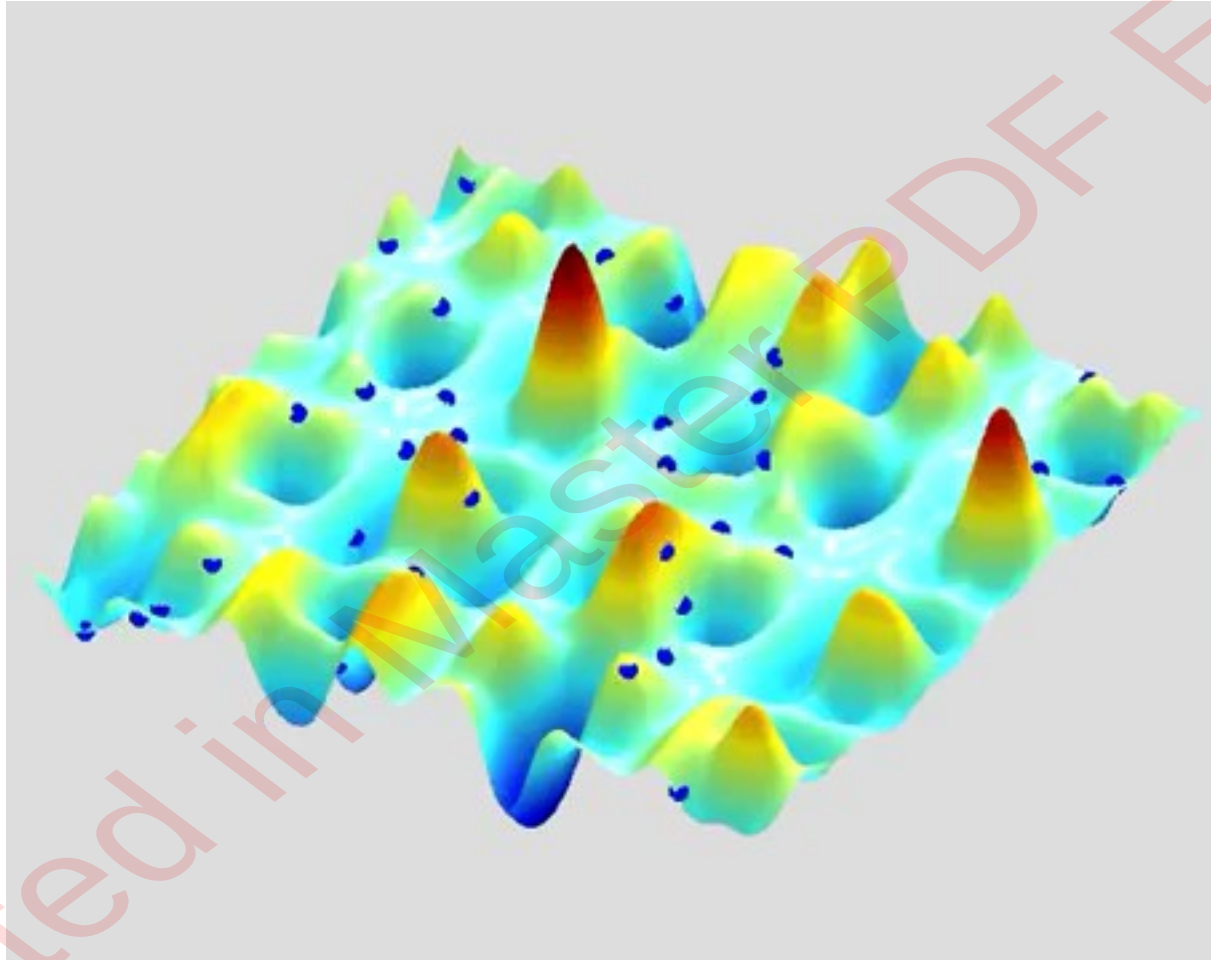


machine learning

optimization

communication,  
visualization

tools for thinking, understanding, communicating

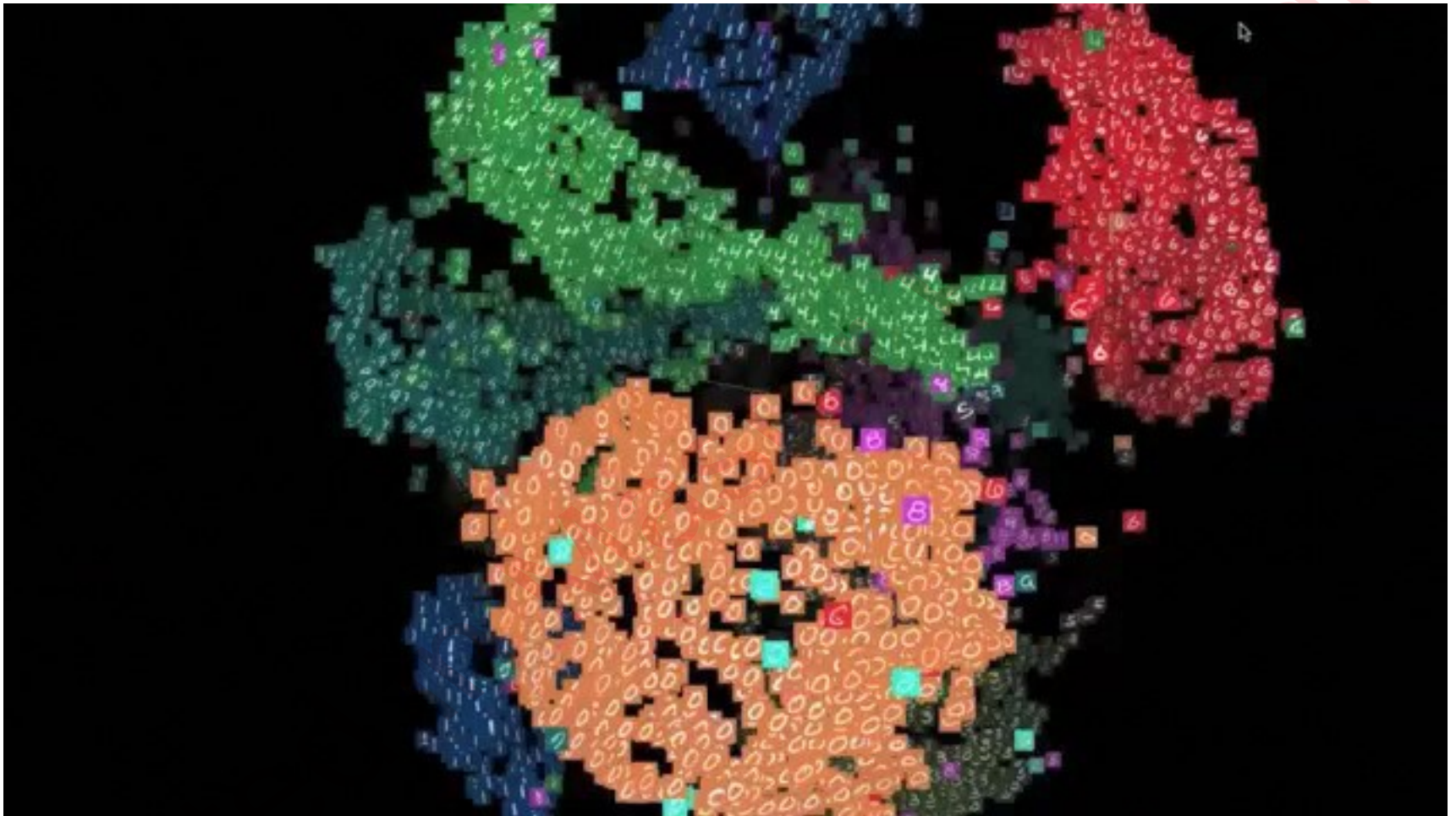


learning process

<https://www.youtube.com/watch?v=VAASmSSsFaY>



tools for thinking, understanding, communicating



learnt classification (from Google)

<https://www.youtube.com/watch?v=wvsE8jmIGzE>

part 3

machine learning

part 1

optimization

part 2

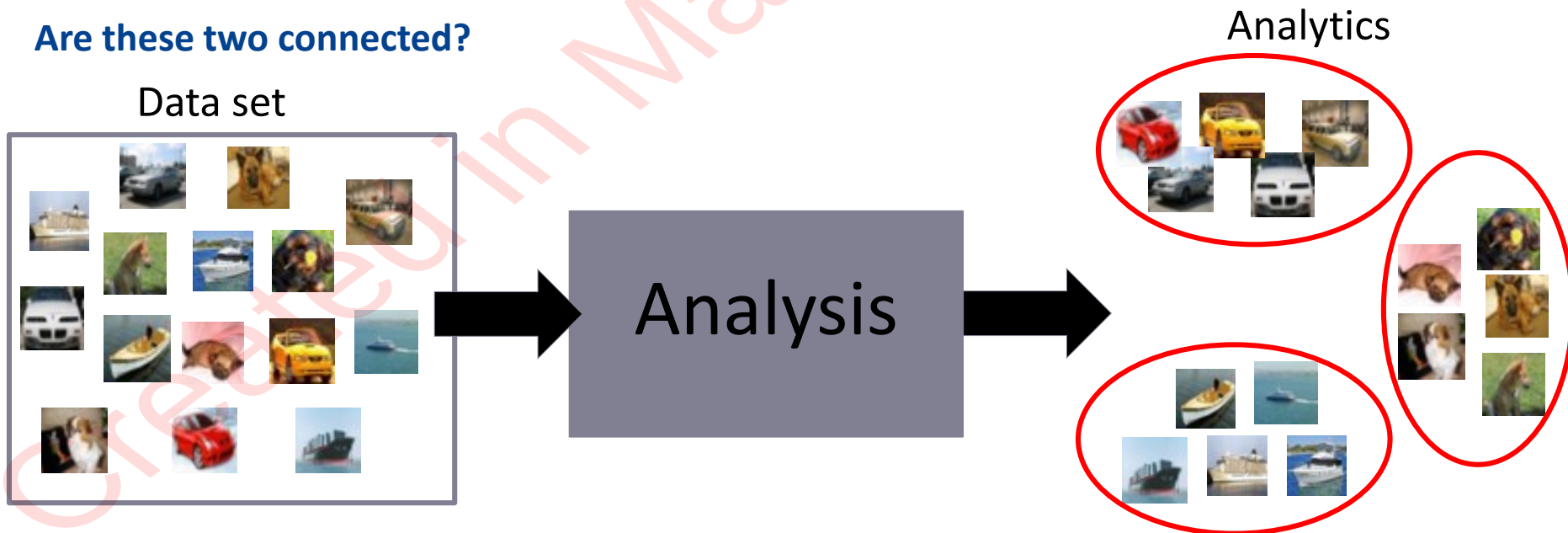
communication,  
visualization

# What is this?

## Some definitions

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## Are these two connected?

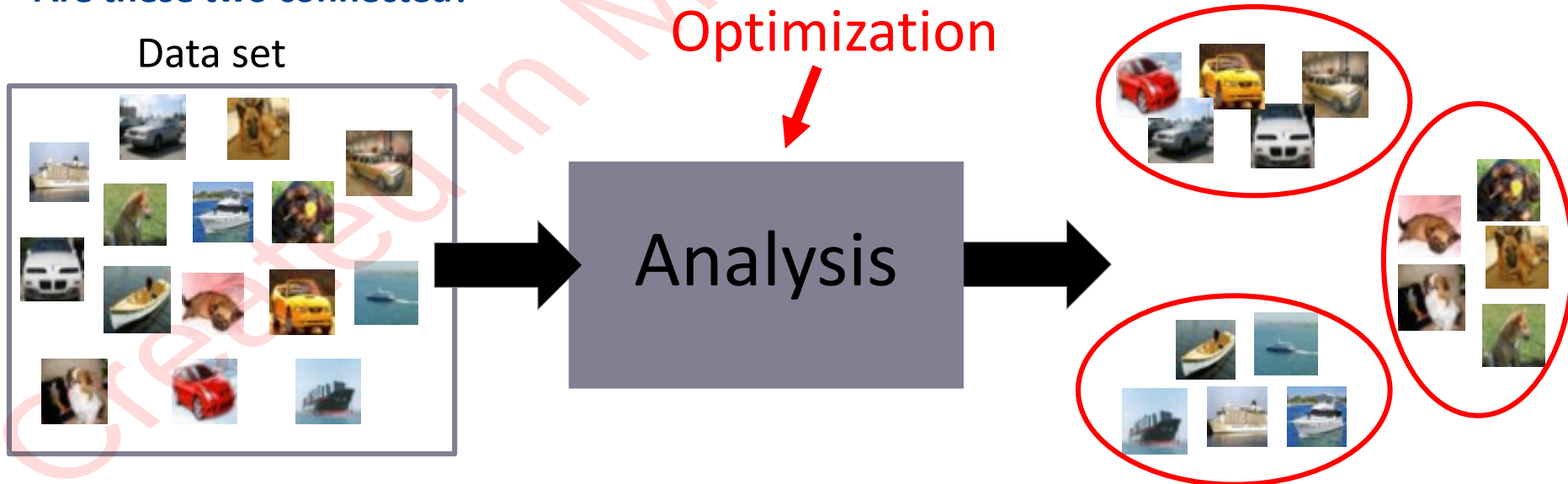


# What is this?

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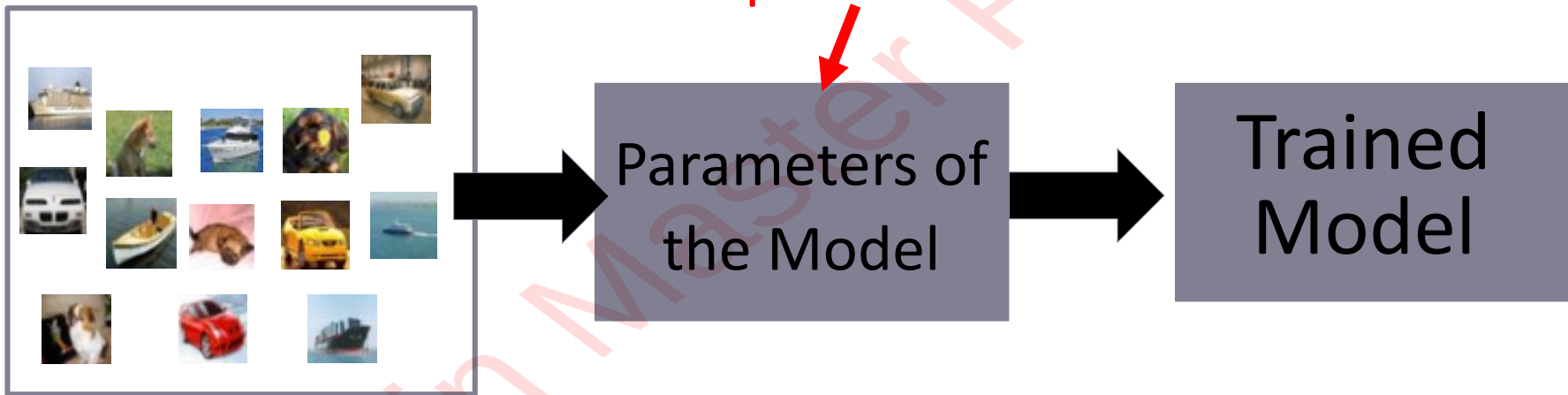
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## Are these two connected?



# Optimization for parameter estimation

## Training phase

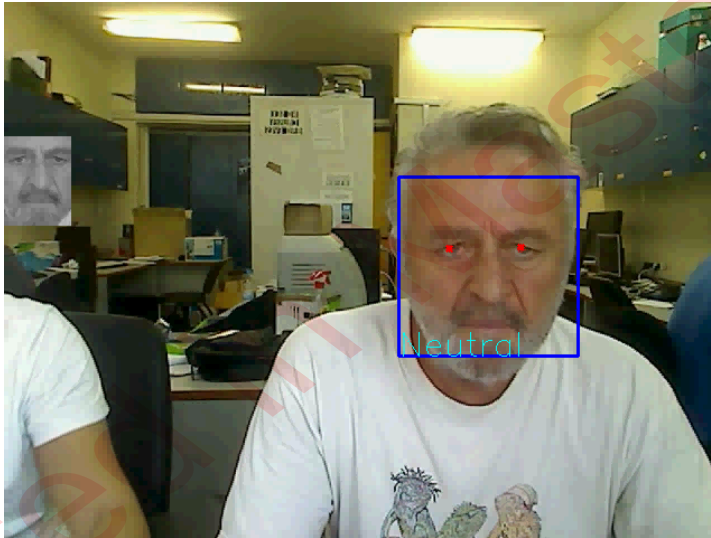


## Test phase/Evaluation/Online process



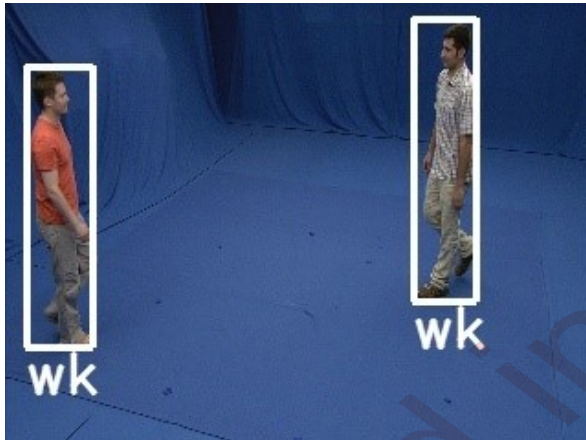
# Examples

## Face Analysis



# Examples

## Action recognition/analysis



A Hough Transform-Based Voting  
Framework for Action Recognition

CVPR 2010

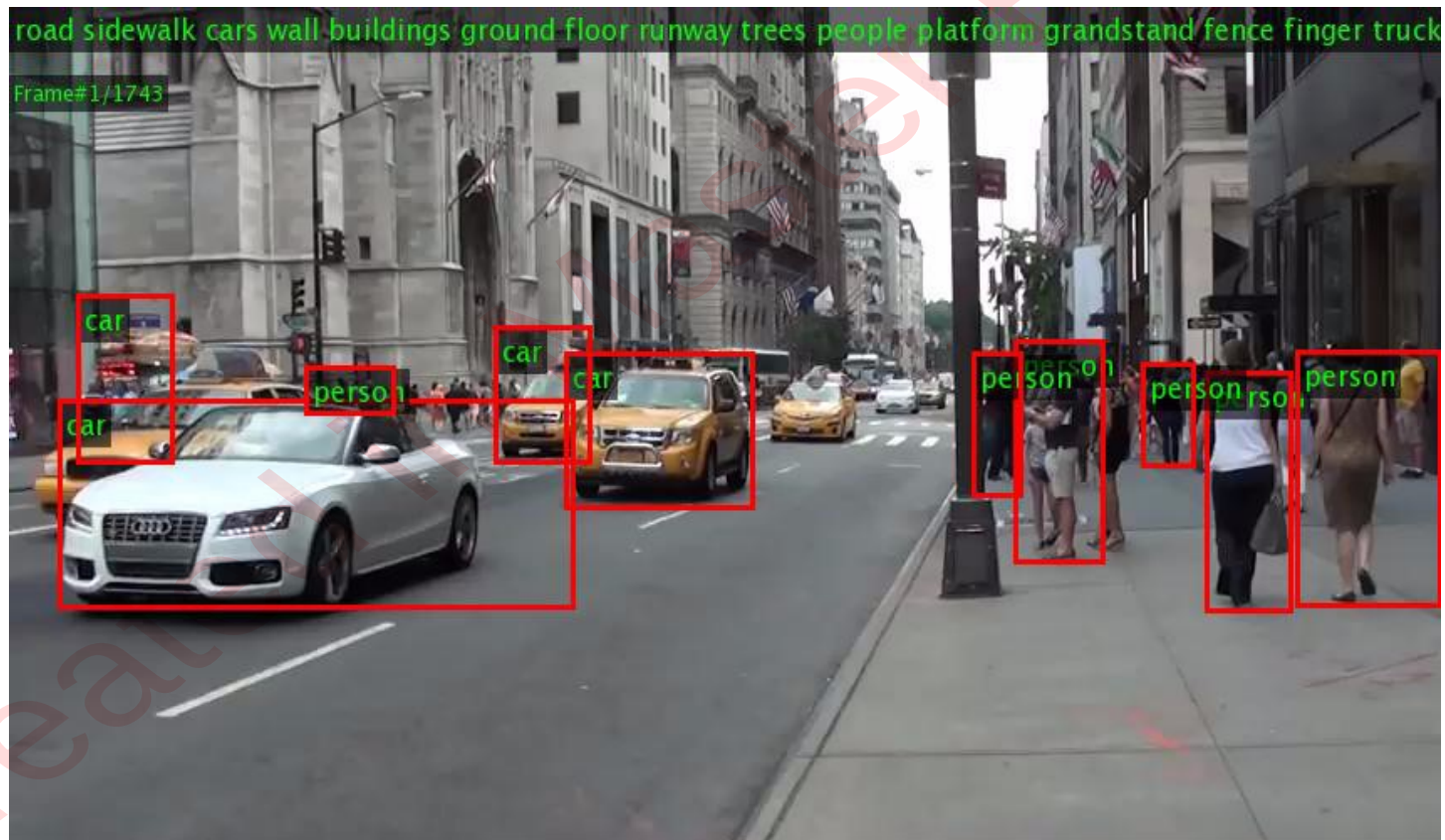
Angela Yao, Juergen Gall  
and Luc Van Gool

Source: <https://www.youtube.com/watch?v=sgcVOOZI8bw>



# Examples

## Object localization/recognition, scene understanding





# Examples

## Image/video captioning



Source: <https://www.youtube.com/watch?v=FmSsek5luHk>

# Examples

## Time-series forecasting



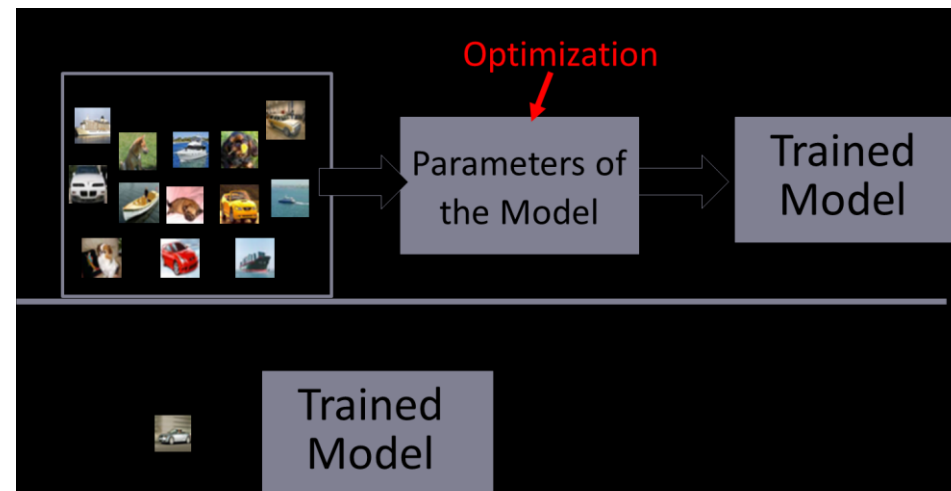
# Examples

## Everyday examples

- › Natural language processing
- › Affective computing
- › Medical diagnosis
- › Robotics
- › Speech and handwritten recognition
- › Translation
- › ...

# Why should I learn about Optimization?

Machine learning models are in fact formed by a set of parameters which need to be 'optimized', i.e. to find good values for these parameters in order to achieve our goal.



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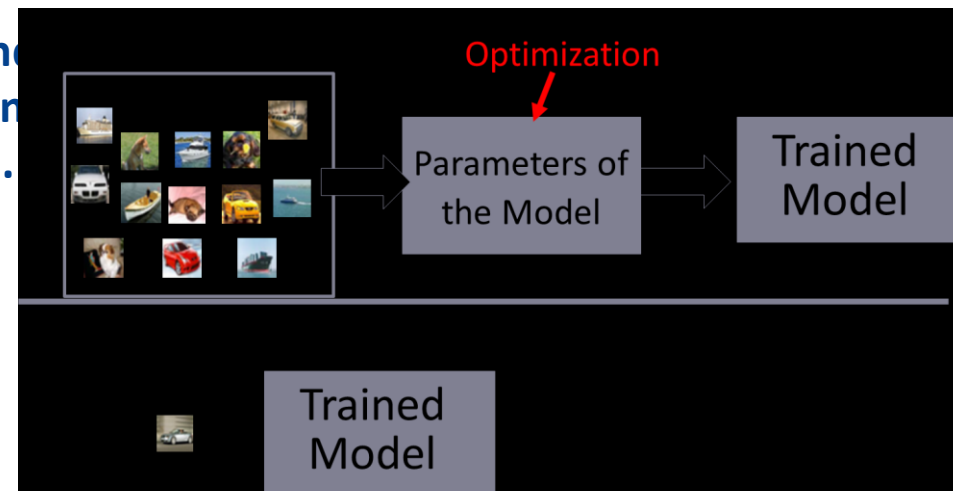
These parameters are usually stored in vectors and matrices. For example a linear model has the form:

$$y_i = W^T x_i$$

W can be random values if x is high dimensions and y is low dimensions. For classifier, W cannot be random.

where the parameters of the linear model are stored in the matrix W.

In this case, the goal of optimization is to find an 'optimal W', which means a W that, given the input  $x_i$  can generate a good (desired)  $y_i$ .



# Optimization topics

**In the course, we will cover the following topics:**

- › Linear programming
- › Unconstrained optimization
- › Nonlinear constrained optimization
- › Solving linear equations
- › Unconstrained optimization: 1D search methods
- › Unconstrained optimization: Gradient methods
- › Global search methods

# Mathematical optimization

A large number of optimization problems can be written in the following form:

$$\begin{aligned} & \text{minimize } f_0(x) \\ & \text{subject to } f_i(x) \leq b_i, \text{ where } i = 1, \dots, m \end{aligned}$$

Constraints are used to define a solution that is practical.

- >  $x = (x_1, \dots, x_n)$ , Optimization variables, or decision variables
- >  $f_0: R^n \rightarrow R$ , objective function
- >  $f_i(x) R^n \rightarrow R$ , where  $i = 1, \dots, m$ : constraint functions
- >  $x^*$  optimal solution, smallest value of  $f_0$  among  $x = (x_1, \dots, x_n)$ , satisfying the constraint

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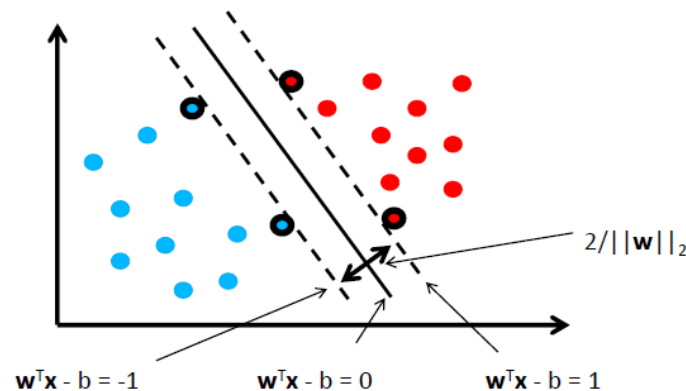
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Example of ML model: SVM



# Linear Programming

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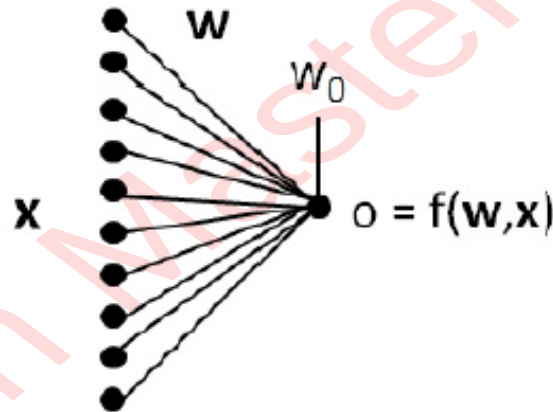
$$\begin{aligned} & \text{minimize } c^T x \\ & \text{subject to } a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

if we had  $c^T c$  then we would not be able to use linear programming

- › *no* analytic solution
- › reliable and efficient algorithms and software
- › computational time proportional to  $n^2 m$ , where  $A \in R^{m \times n}$ , in some case less than this
- › mature technology

# Linear Programming

## Example of ML model: Perceptron



# Unconstrained optimization

Unconstrained optimization is used in cases where there are no constraints on the model's parameters

For example a linear unconstrained problem has the form

$$\text{minimize } c^T x$$

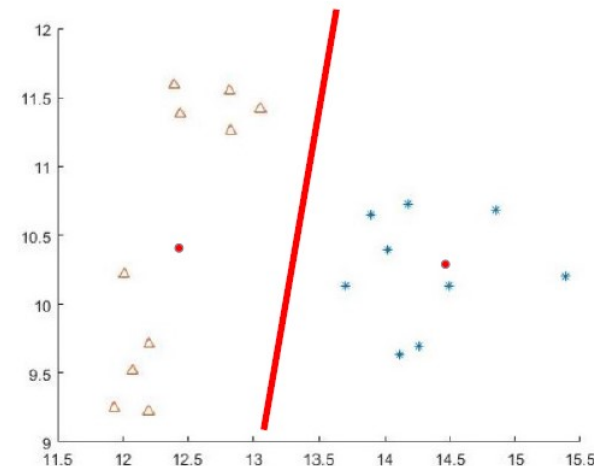
# Unconstrained optimization

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$$\text{minimize } c^T x$$

Example of ML model: Linear classifiers



# Solving linear equations

A system of linear equations has the form:

$$\mathbf{Ax} = \mathbf{b}$$

Example of ML model: Linear/nonlinear Regression

$$\mathcal{J}_s = \sum_{i=1}^N (\mathbf{w}^T \mathbf{x}_i - b_i)^2 = \|\mathbf{X}^T \mathbf{w} - \mathbf{b}\|_2^2$$

# 1D search methods

Very often, optimization cannot be obtained using a closed form solution. Then we need to follow an iterative process which at every step gives a set of parameters which are better than before

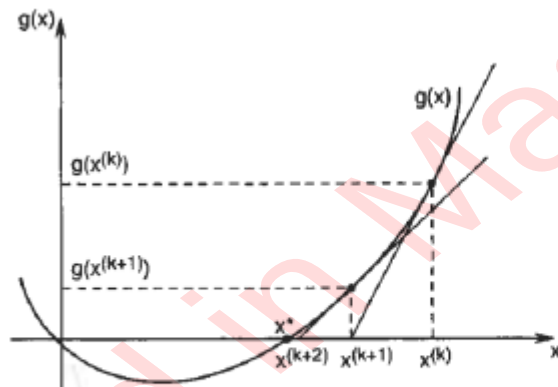
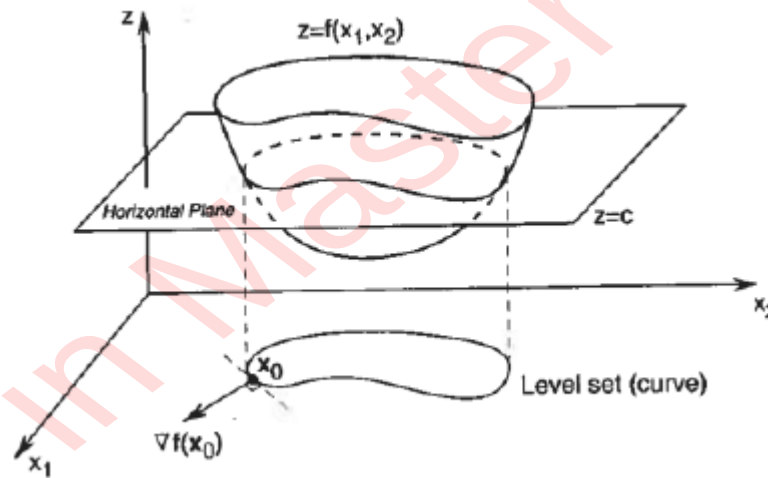


Figure 7.8 Newton's method of tangents.

$$x^{(k+1)} = x^{(k)} - \frac{g(x^{(k)})}{g'(x^{(k)})}$$

# Gradient methods

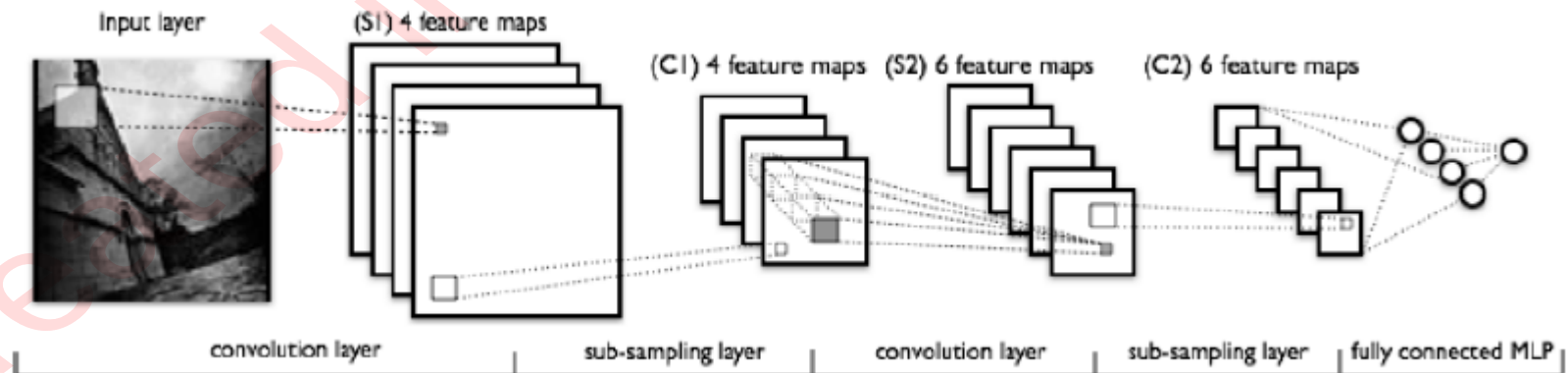
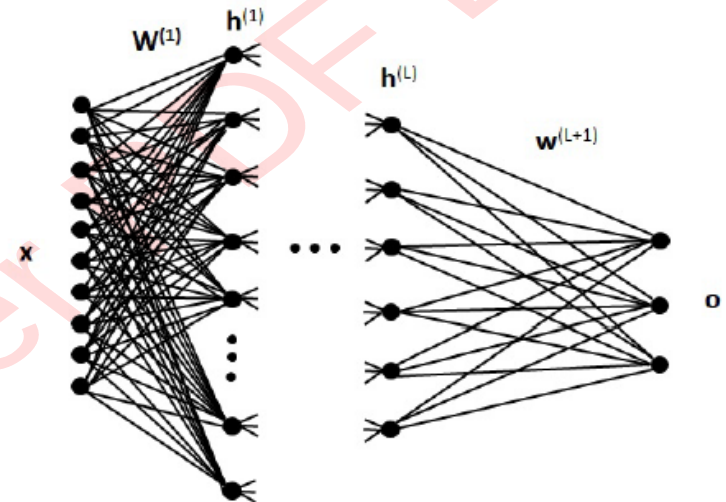
When the number of parameters is higher than one, the iterative optimization process uses the gradient of the function w.r.t. the parameters for updating:





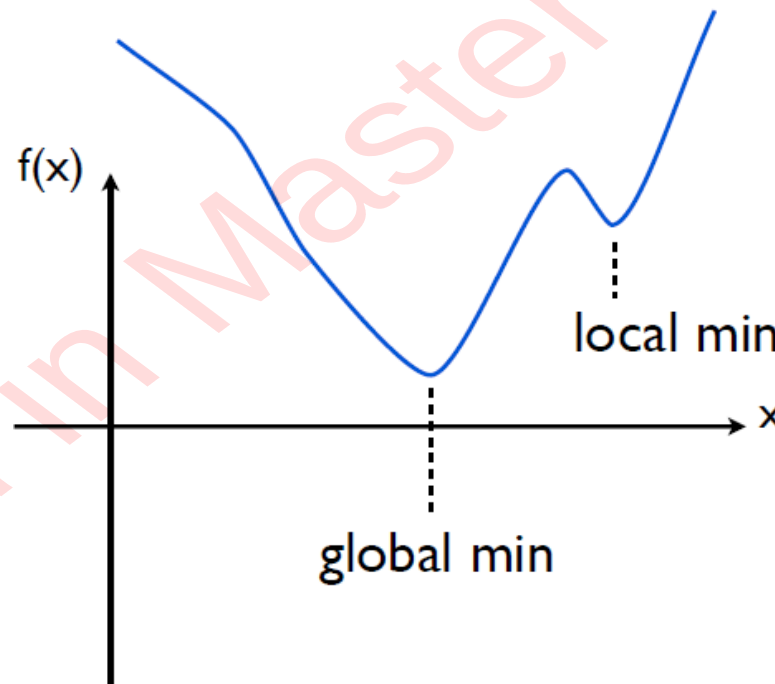
# Gradient methods

## Example of ML model: Neural networks



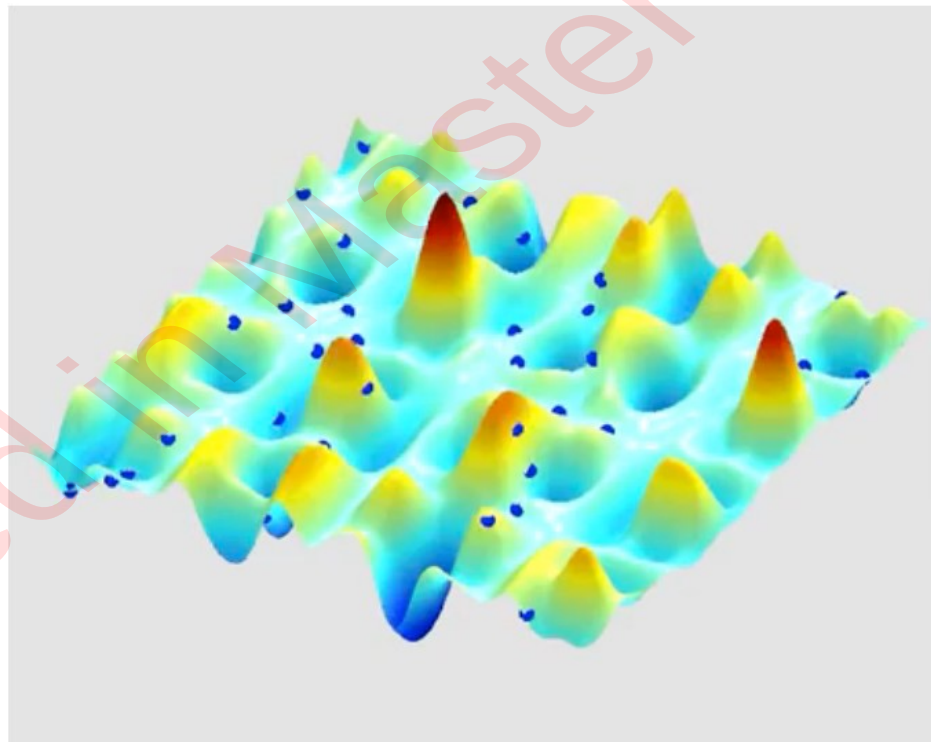
# Global search methods

One of the problems of iterative optimization methods is that their performance depends on the starting point:



# Global search methods

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<https://www.youtube.com/watch?v=VAASmSSsFaY>