

# MACHINE LEARNING

## 01 SUPERVISED MACHINE LEARNING : REGRESSION AND CLASSIFICATION



# UNE SÉRIE DE FORMATIONS

## PARTIE 1 : SUPERVISED MACHINE LEARNING : REGRESSION AND CLASSIFICATION

Introduction to Machine Learning  
Regression with multiple input  
Classification

## PARTIE 2 : ADVANCED LEARNING ALGORITHMS

Neural Networks  
Neural Networks training  
Advice for applying Machine Learning  
Decision Trees

## PARTIE 3 : UNSUPERVISED LEARNING, RECOMMENDERS, REINFORCEMENT LEARNING

Unsupervised Learning  
Recommender Systems  
Reinforcement Learning



# 01 SUPERVISED MACHINE LEARNING : REGRESSION AND CLASSIFICATION

## SUPERVISED VS. UNSUPERVISED ML

What is Machine Learning ?

Supervised Learning

Unsupervised Learning

Jupyter Notebooks

**Lab** : Python and Jupyter Notebooks

## REGRESSION MODEL

Linear regression model

**Lab** : Model representation

Cost function formula

Cost function intuition

Visualizing the cost function

Visualization examples

**Lab** : Cost function

## TRAIN THE MODEL WITH GRADIENT DESCENT

Gradient descent

Implementing gradient descent

Gradient descent intuition

Learning rate

Gradient descent for linear regression

Running gradient descent

**Lab** : Gradient descent

# 101

## SUPERVISED VS. UNSUPERVISED MACHINE LEARNING



# 1 / SUPERVISED VS. UNSUPERVISED MACHINE LEARNING

## Supervised learning

$x$   
input

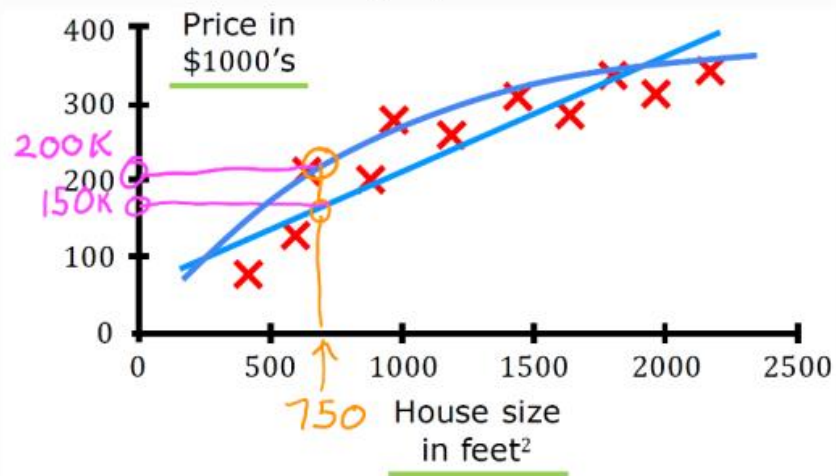
$y$   
output label

learns from being given data labeled with the "right answers"

### Regression

Predict a number

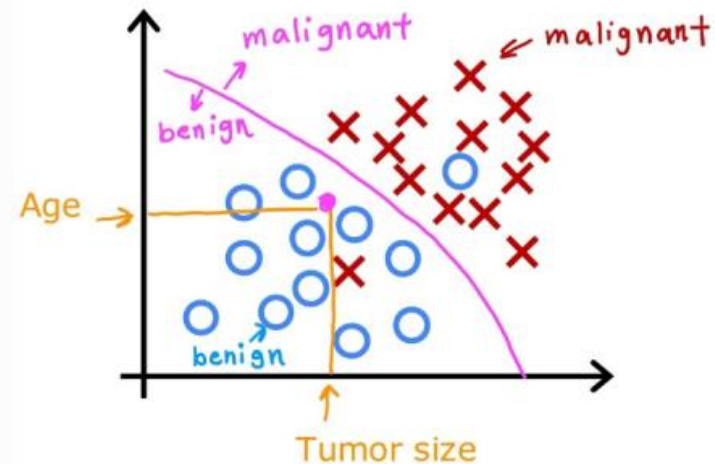
Infinitely many possible outcomes



### Classification

Predict categories

Small number of possible outputs



# 1 / SUPERVISED VS. UNSUPERVISED MACHINE LEARNING

## Unsupervised Learning

Data only comes with inputs  $x$ , but not output labels  $y$   
Algorithm has to find *structure* (= something interesting in *unlabeled* data)

### Clustering

Group similar data points together.

### Anomaly detection

Find unusual data points.

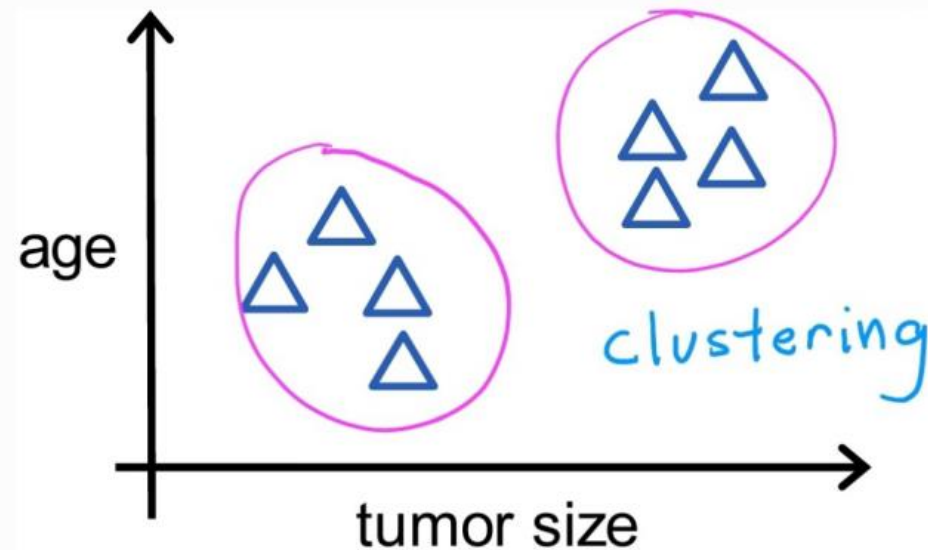
### Dimensionality reduction

Compress data using fewer numbers.

## Clustering



Example: Google News



# LAB-01

## INTRODUCTION TO PYTHON AND JUPYTER NOTEBOOKS



**QUESTIONS ?**  
**SUR UN CONCEPT ? UNE IDÉE ?**  
**SUR UN DÉTAIL DU CODE ?**  
**(ENVIE D'UNE PAUSE ?)**

**N'HÉSITEZ PAS !**

**IL N'Y A PAS DE QUESTION BÊTE, SI VOUS AVEZ UN DOUTE, D'AUTRES ONT SÛREMENT LE MÊME**



# 102

## REGRESSION MODEL

# Linear Regression Model

## Terminology

| $x$            | $y$       |
|----------------|-----------|
| $x^{(1)}$      | $y^{(1)}$ |
| $\vdots$       | $\vdots$  |
| $x^{(m)}$      | $y^{(m)}$ |
| (training set) |           |

$x$  = "input" variable / feature

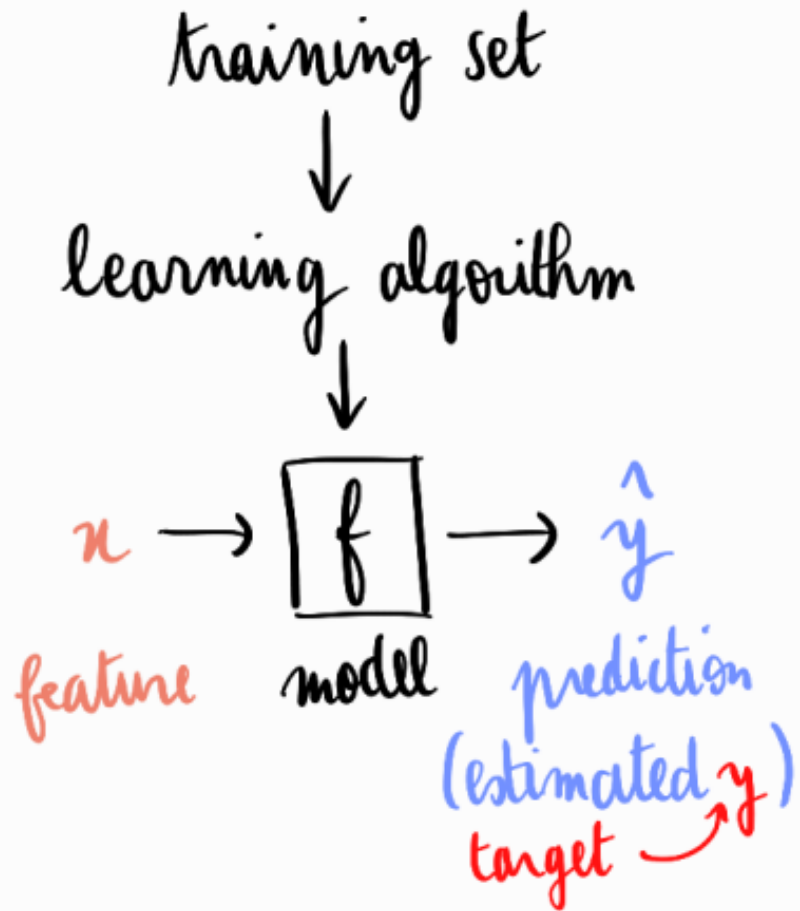
$y$  = "output" variable / "target" variable

$m$  = number of variables in the training set

$w$  = weight /  $b$  = bias

$(x, y)$  = single training example

$(x^{(i)}, y^{(i)})$  =  $i^{\text{th}}$  training example



$$f_{w,b}(x) = wx + b \quad (\text{written } f(x))$$



Linear regression with one variable.

Univariate linear regression.  
one variable



# LAB-02

## MODEL REPRESENTATION

# QUESTIONS ?

SUR UN CONCEPT ? UNE IDÉE ?  
SUR UN DÉTAIL DU CODE ?  
(ENVIE D'UNE PAUSE ?)

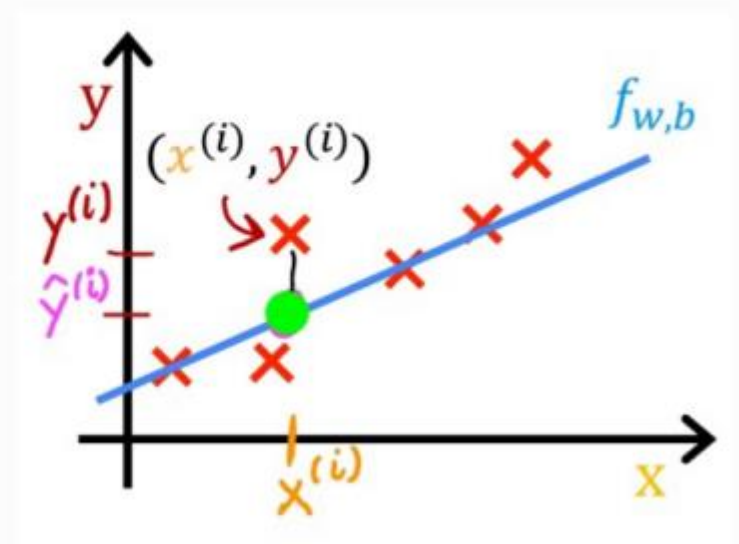
N'HÉSITEZ PAS !

IL N'Y A PAS DE QUESTION BÊTE, SI VOUS AVEZ UN DOUTE, D'AUTRES ONT SÛREMENT LE MÊME

## Cost Function

$$\hat{y}^{(i)} = f_{w,b}(x^{(i)}) = wx^{(i)} + b$$

Find  $(w, b)$  such as  $\hat{y}^{(i)} \approx y^{(i)} \forall (x, y)$



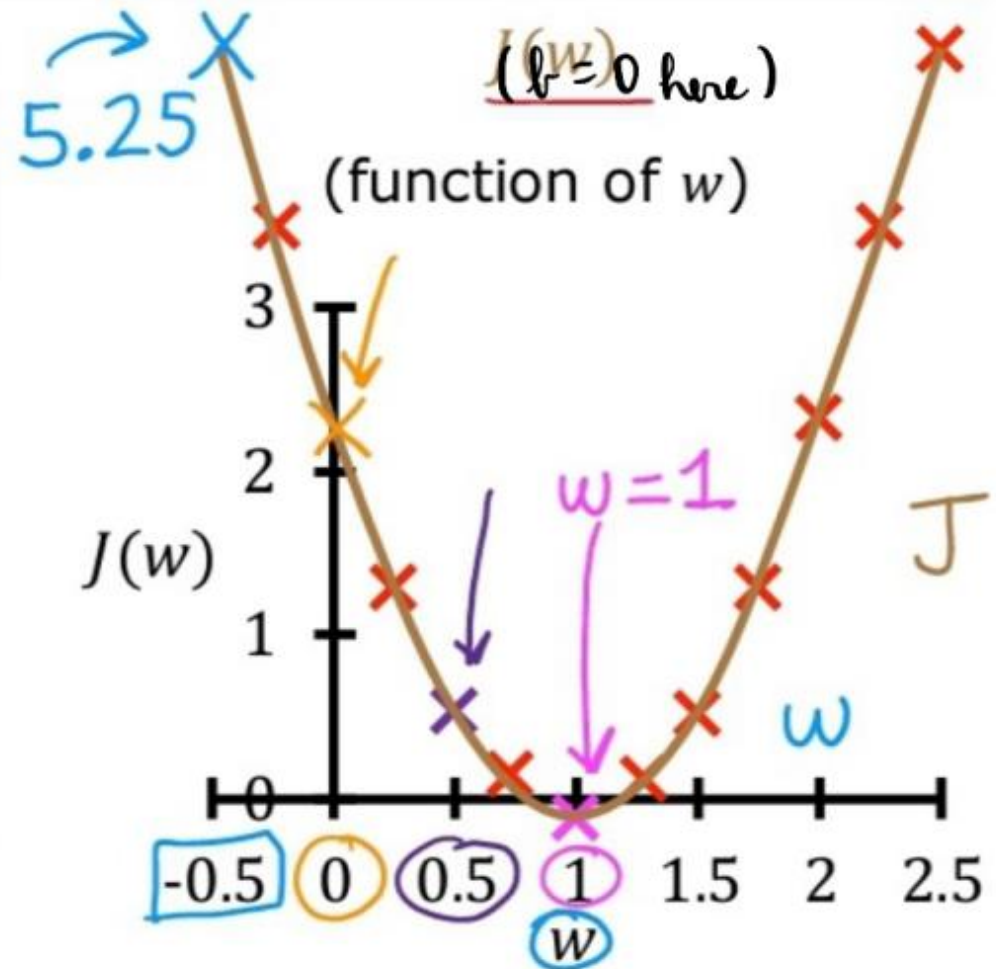
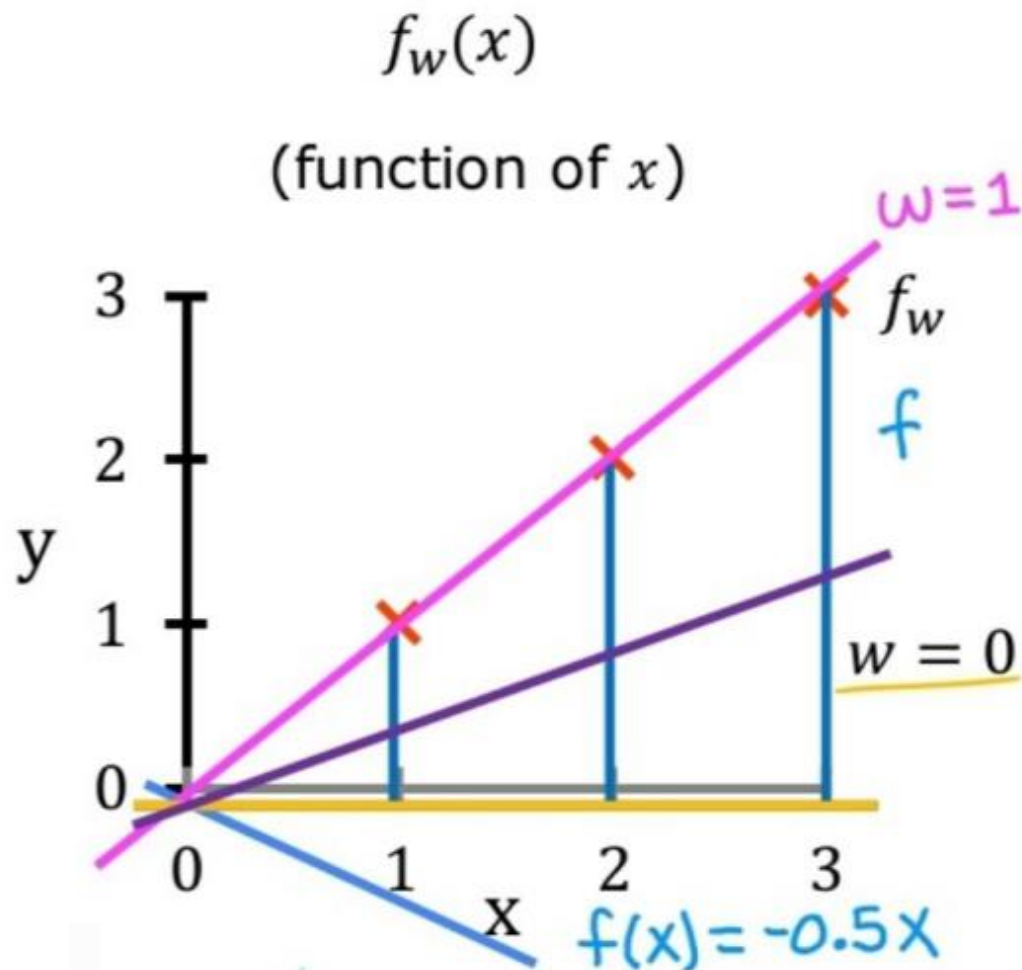
## Squared error cost function

$$J(w, b) = \frac{1}{2m} \sum_{i=1}^m \left( \underbrace{f_{w,b}(x^{(i)})}_{\hat{y}^{(i)}} - y^{(i)} \right)^2$$

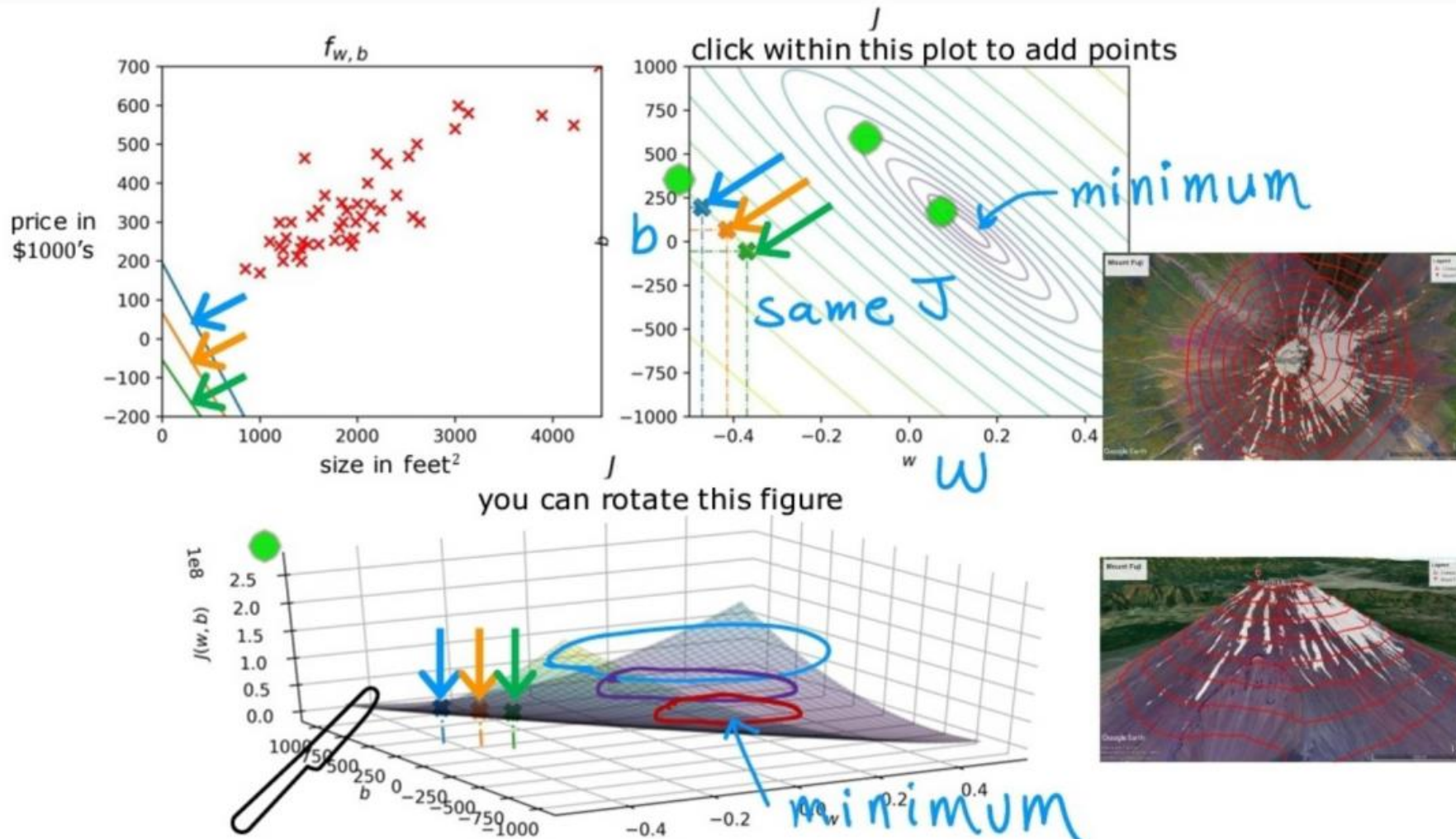
goal: to minimize  $J(w, b)$



## Cost Function Intuition



# Visualizing the Cost Function → 3D Surface Plot vs. Contour Plot



# LAB-03

## COST FUNCTION



# QUESTIONS ?

SUR UN CONCEPT ? UNE IDÉE ?  
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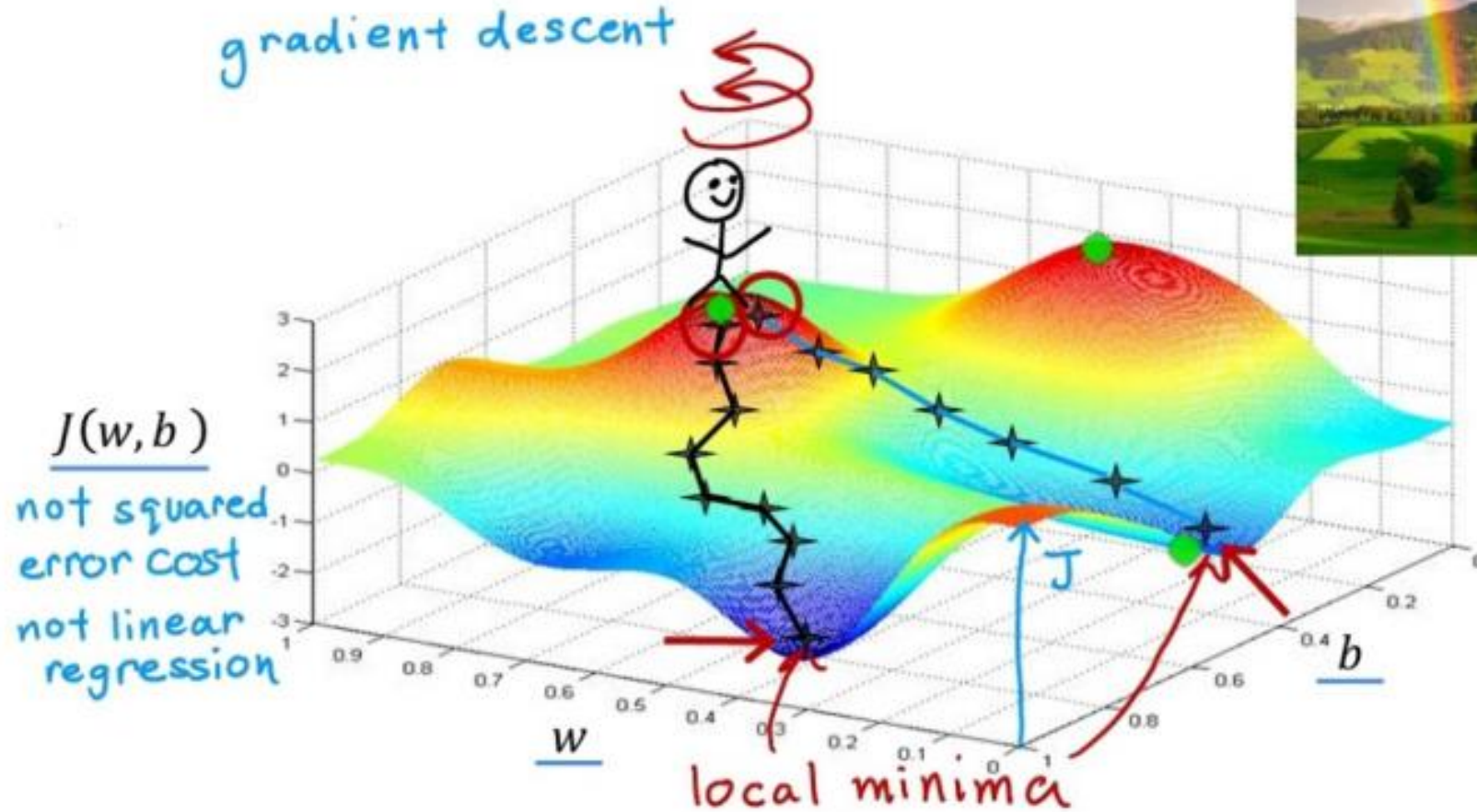
N'HÉSITEZ PAS !

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# 103

## TRAIN THE MODEL WITH GRADIENT DESCENT

## Gradient Descent





## Gradient Descent Algorithm

Repeat until convergence:

$$w = w - \alpha \frac{\partial}{\partial w} J(w, b)$$

$$b = b - \underbrace{\alpha}_{\text{learning rate}} \frac{\partial}{\partial b} J(w, b)$$

Batch = each step of gradient descent uses all the training examples

Simultaneous update:



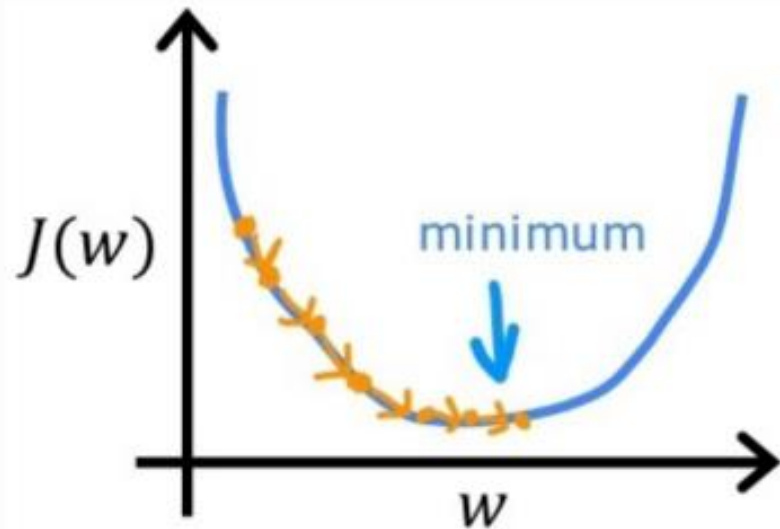
$$\text{tmp-}w = w - \alpha \frac{\partial}{\partial w} J(w, b)$$

$$\text{tmp-}b = b - \alpha \frac{\partial}{\partial b} J(w, b)$$

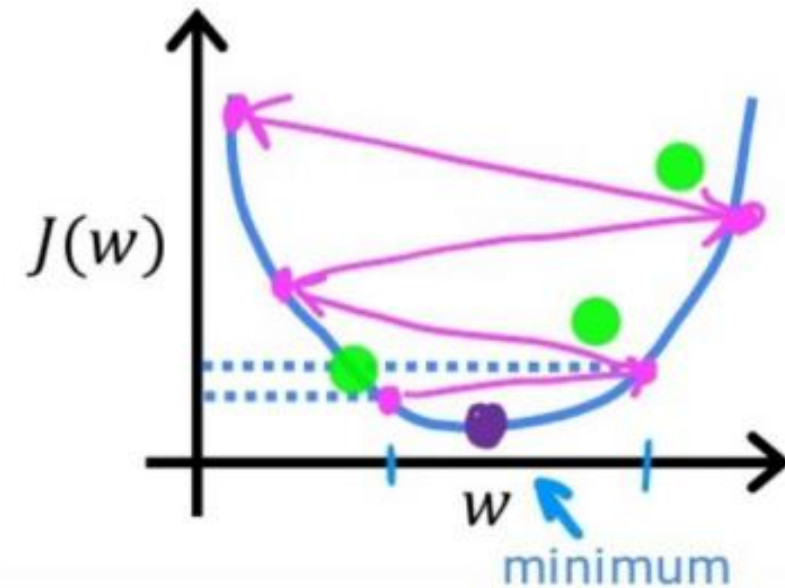
$$w = \text{tmp-}w$$

$$b = \text{tmp-}b$$

## Learning Rate



If  $\alpha$  is too small...  
Gradient descent may be slow.



If  $\alpha$  is too large...

Gradient descent may:

- Overshoot, never reach minimum
- Fail to converge, diverge

# LAB-04

## GRADIENT DESCENT FOR LINEAR REGRESSION



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