

# Data Mining - Mathematics

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# Outline

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1. Introduction
2. Matrices and vectors
3. Derivative
4. Gradient
5. Lagrangian
6. Exercises



# 1. Introduction

## What we will cover

Mathematical background needed for the understanding of Machine Learning techniques:

- matrix and vector operations
- derivative
- gradient
- lagrangian

## Difficulty of mathematics

- use of **symbols** that represent expressions
- some symbols have different meaning depending on the **context**
- sometimes strict syntax and sometimes not

## Example of a prime number

$n$  is a prime number if it has only two divisors 1 and itself (but with restriction that  $x \neq 1$ )

- implies that we deal with integers
- implies the notion of **divisibility**:

$$\forall n, p, q, r \in \mathbb{N}$$

$$n = p \times q + r$$

- $n$  is divisible by  $q$  if (and only if)  $r = 0$  and  $p \geq 1$



## 2. Matrices and vectors

## Vector

- a series of values that can be identified by their index in an array of length  $p$
- $x(p)$  or simply  $x \in \mathbb{R}^p$
- for computer scientists a 1D array of length  $p$
- $x(p) = (x_1, x_2, \dots, x_p) = [x_1, x_2, \dots, x_p]$ ,  $x_i \in \mathbb{R}$
- can also be represented vertically (for mathematicians):

$$x(p) = x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}, \quad x^T = [x_1, x_2, \dots, x_p]$$

in this case  $x^T$  (T for Transposition) will be the horizontal representation



## Operations on vectors

- vectors must have the same length
- $+$ ,  $-$ ,  $\times$ ,  $/$
- the dot (or scalar) product of two vectors

$$xy = x \cdot y = x \odot y = x^T y = \sum_{i=1}^p x_i \times y_i$$

- norm (or length) of a vector  $\|x\| = \sqrt{x \cdot x}$

## Examples of operations on vectors

$$x = [1, -2, 3]$$

$$y = [-1, 4, -7]$$

$$x + y = [1 + (-1), -2 + 4, 3 + (-7)] = [0, 2, -4]$$

$$x \times y = [1 \times -1, -2 \times 4, 3 \times -7] = [-1, -8, -21]$$

$$xy = (1 \times -1) + (-2 \times 4) + (3 \times -7) = -30$$

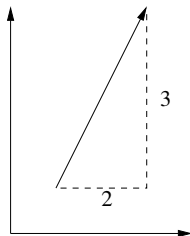
$$x^T y = [1, -2, 3] \cdot \begin{bmatrix} -1 \\ 4 \\ -7 \end{bmatrix} = -30$$

## Norm of a vector in 2D

$$x \cdot x = (2 \times 2) + (3 \times 3)$$

$$= 4 + 9 = 13$$

$$\|x\| = \sqrt{13} = 3.6055$$



## Matrix

- $X(p) = (x_1, x_2, \dots, x_p)$  where  $x_i \in \mathbb{R}^n$ , notation that can lead to confusion
- for computer scientists a 2D array defined by :
  - ▶  $n$  rows
  - ▶ and  $p$  columns
- can be seen as an array of vectors or vector of vectors

$$X(n, p) = \begin{bmatrix} x_1^1 & x_1^2 & \dots & x_1^p \\ x_2^1 & x_2^2 & \dots & x_2^p \\ \vdots & \ddots & \ddots & \vdots \\ x_n^1 & x_n^2 & \dots & x_n^p \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} x_1^1 \\ x_2^1 \\ \vdots \\ x_n^1 \end{bmatrix} & \begin{bmatrix} x_1^2 \\ x_2^2 \\ \vdots \\ x_n^2 \end{bmatrix} & \dots & \begin{bmatrix} x_1^p \\ x_2^p \\ \vdots \\ x_n^p \end{bmatrix} \end{bmatrix}$$

$x_i^j$  is the element in row  $i$  and column  $j$

## Square matrices

- a square matrix is such that  $n = p$ 
  - ▶ the diagonal is a separation line
  - ▶ called **lower triangular** if all the entries above the main diagonal are zero
  - ▶ called **upper triangular** if all the entries under the main diagonal are zero
- $I$  or  $I_n$  is the identity matrix

$$I_n = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \ddots & 1 & 0 \\ 0 & \dots & 0 & 1 \end{bmatrix}$$

## Matrix sum

- matrices must have the same dimensions

$$\begin{aligned} X(n,p) &= \begin{bmatrix} x_1^1 & x_1^2 & \dots & x_1^p \\ x_1^2 & x_2^2 & \dots & x_2^p \\ \vdots & \vdots & \ddots & \vdots \\ x_n^1 & x_n^2 & \dots & x_n^p \end{bmatrix} + Y(n,p) = \begin{bmatrix} y_1^1 & y_1^2 & \dots & y_1^p \\ y_2^1 & y_2^2 & \dots & y_2^p \\ \vdots & \vdots & \ddots & \vdots \\ y_n^1 & y_n^2 & \dots & y_n^p \end{bmatrix} \\ &= Z(n,p) = \begin{bmatrix} x_1^1 + y_1^1 & x_1^2 + y_1^2 & \dots & x_1^p + y_1^p \\ x_2^1 + y_2^1 & x_2^2 + y_2^2 & \dots & x_2^p + y_2^p \\ \vdots & \vdots & \ddots & \vdots \\ x_n^1 + y_n^1 & x_n^2 + y_n^2 & \dots & x_n^p + y_n^p \end{bmatrix} \end{aligned}$$

## Matrix product

- $X(n, p) \times Y(p, q) = Z(n, q)$
- number of columns of  $X$  = number of rows of  $Y$

$$z_i^j = \sum_{k=1}^p a_i^k \times b_k^j$$

```
for (int i = 0; i < n; ++i)
  for (int j = 0; j < q; ++j) {
    double sum = 0;
    for (int k = 0; k < p; ++k) {
      sum += a[i][k] * b[k][j];
    }
    c[i][j] = sum;
  }
```

## Matrix product

Note that generally:

$$AB \neq BA$$



## Example of matrix product

$$X(3,2) = \begin{bmatrix} 11 & 12 \\ 21 & 22 \\ 31 & 32 \end{bmatrix} \times Y(2,3) = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

## Example of matrix product

$$X(3, 2) = \begin{bmatrix} 11 & 12 \\ 21 & 22 \\ 31 & 32 \end{bmatrix} \times Y(2, 3) = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

$$Z(3, 3) = \begin{bmatrix} 11 \times 1 + 12 \times 4 & 11 \times 2 + 12 \times 5 & 11 \times 3 + 12 \times 6 \\ 21 \times 1 + 22 \times 4 & 21 \times 2 + 22 \times 5 & 21 \times 3 + 22 \times 6 \\ 31 \times 1 + 32 \times 4 & 31 \times 2 + 32 \times 5 & 31 \times 3 + 32 \times 6 \end{bmatrix}$$

## Example of matrix and vector product

$$X(3,2) = \begin{bmatrix} 11 & 12 \\ 21 & 22 \\ 31 & 32 \end{bmatrix} \times y(2) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

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$$Z(3,3) = \begin{bmatrix} 11 \times 1 + 12 \times 2 \\ 21 \times 1 + 22 \times 2 \\ 31 \times 1 + 32 \times 2 \end{bmatrix} = \begin{bmatrix} 35 \\ 65 \\ 95 \end{bmatrix}$$

## Matrix transpose

- to make it simple: exchange of values from both sides of the diagonal of the matrix

$$X(n, p) = \begin{bmatrix} x_1^1 & x_1^2 & \dots & x_1^p \\ x_1^2 & x_2^2 & \dots & x_2^p \\ \vdots & \ddots & \ddots & \vdots \\ x_n^1 & x_n^2 & \dots & x_n^p \end{bmatrix} \quad X^T(p, n) = \begin{bmatrix} x_1^1 & x_2^1 & \dots & x_n^1 \\ x_1^2 & x_2^2 & \dots & x_n^2 \\ \vdots & \ddots & \ddots & \vdots \\ x_1^p & x_2^p & \dots & x_n^p \end{bmatrix}$$

## Matrix transpose

$$X(3,3) = \begin{bmatrix} 11 & 12 & 13 \\ 21 & 22 & 23 \\ 31 & 32 & 33 \end{bmatrix} \quad X^T(3,3) = \begin{bmatrix} 11 & 21 & 31 \\ 12 & 22 & 32 \\ 13 & 23 & 33 \end{bmatrix}$$

$$X(3,5) = \begin{bmatrix} 11 & 12 & 13 & 14 & 15 \\ 21 & 22 & 23 & 24 & 25 \\ 31 & 32 & 33 & 34 & 35 \end{bmatrix} \quad X^T(5,3) = \begin{bmatrix} 11 & 21 & 31 \\ 12 & 22 & 32 \\ 13 & 23 & 33 \\ 14 & 24 & 34 \\ 15 & 25 & 35 \end{bmatrix}$$

## Inverse of a matrix

The inverse matrix  $A^{-1}$  of a matrix  $A$  is such that

$$A^{-1} \times A = I$$

and is used for example to solve linear equation systems:

$$A \times x = b \quad \text{then} \quad x = A^{-1} \times b$$

where  $x$  and  $b$  are vectors

## Inverse of a matrix

$$A(3,3) = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 2 & -3 \\ -7 & 6 & -5 \end{bmatrix}$$

$$A^{-1}(3,3) = \begin{bmatrix} 0.125 & 0.4375 & -0.1875 \\ 0.25 & 0.25 & 0 \\ 0.125 & -0.3125 & 0.0625 \end{bmatrix}$$

$$A \times \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = b = \begin{bmatrix} 14 \\ -6 \\ -10 \end{bmatrix} \quad A^{-1}b = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$



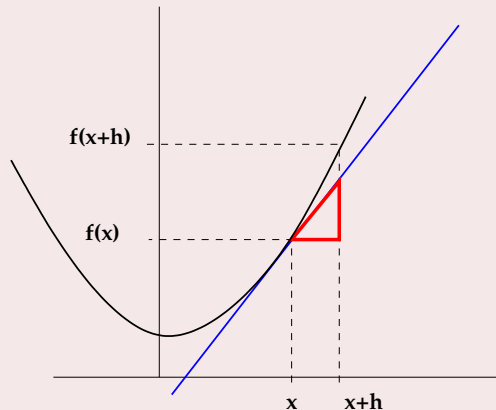


### 3. Derivative

## Derivative of $f(x)$

- the slope of the tangent of a curve in a given point
- if positive: the curve will increase
- if negative: the curve will decrease
- if zero: won't increase or decrease

## Derivative of $f(x)$



## Derivative

More formally the derivative can be defined as

$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{(x+h) - x} = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

Notations:

$$f'(x) \quad \text{or} \quad \frac{df(x)}{dx}$$

$\frac{df(x)}{dx}$  means the variation of  $f(x)$  if we increase  $x$  by a small value

## Property of addition and product of the derivative

$$(f + g)' = f' + g'$$

$$(f \times g)' = f' \times g + f \times g'$$

As an exercise you could try to prove the result of  $(f \times g)'$

## Property of the composition

$$(g \circ f)' = (g' \circ f) \times f'$$

in other words:

$$g(f(x))' = g'(f(x)) \times f'(x) \quad (\text{DerivComp})$$

## Property of the composition

Find the derivative of

$$h(x) = \sin(3x^2 + 2)$$

- let  $g(x) = \sin(x)$ , then  $g'(x) = \cos(x)$
- let  $f(x) = 3x^2 + 2$ , then  $f'(x) = 6x$

So

$$h'(x) = g(f(x))' = g'(f(x)) \times f'(x)$$

$$h'(x) = \cos(3x^2 + 2) \times 6x$$

## Property of the inverse function

Let  $f^{-1}(x)$  be the inverse function of  $f(x)$ , i.e.  $f^{-1}(f(x)) = x$

$$[f^{-1}(f(x))]'$$
                       $=$                        $(x)' = 1$

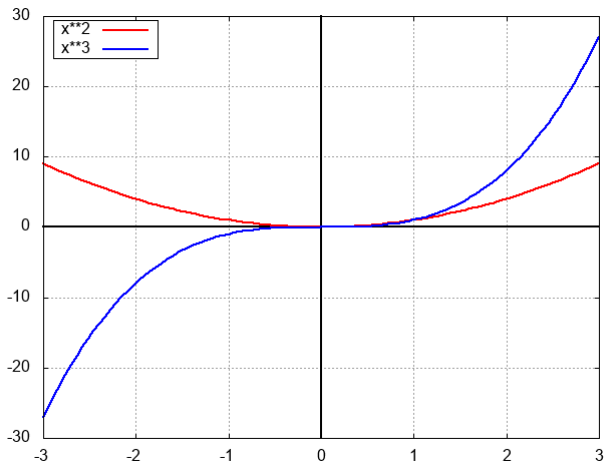
$$f^{-1}'(f(x)) \times f'(x) = 1$$
                      from (DerivComp)

$$f^{-1}'(f(x)) = \frac{1}{f'(x)}$$
                      (DerivInv)



# Derivative of $x^n$

$$f(x) = x^n$$



# Derivative of $x^n$

$$f(x) = x^n$$

The derivative should have the following behaviour

$X$	$-\infty$	$0$	$\infty$
$x^{2k}$	$-$	$0$	$+$
$x^{2k+1}$	$+$	$0$	$+$

with  $n$  even ( $2k$ ) or odd ( $2k + 1$ )

$$(x + a)^n$$

Remember that  $x^0 = 1$

$$(x + a)^2 = x^2 + 2ax + a^2$$

$$(x + a)^3 = x^3 + 3ax^2 + 3a^2x + a^3$$

$$(x + a)^4 = x^4 + 4ax^3 + 6a^2x^2 + 3a^3x + a^4$$

...

$$(x + a)^n = \alpha_{0,n}a^0x^n + \dots + \alpha_{i,j}a^i x^j + \dots + \alpha_{n,0}a^n x^0$$

$$(x + a)^n = \sum_{i=0, j=n-i}^{i=n} \alpha_{i,j} a^i x^j$$

the coefficients  $\alpha_{i,j}$  of  $a^i x^j$  are given by Pascal's triangle

## Blaise Pascal (fr) (1623-1662)

- was a French mathematician, physicist, inventor, writer and catholic theologian
- wrote a significant treatise on the subject of projective geometry at the age of 16
- work on the principles of hydraulic fluids (hydraulic press and the syringe)
- theological work, referred to posthumously as the *Pensées* (*Thoughts*)



## Pascal's triangle

$n$	$x^n$	$ax^{n-1}$	...		
0	1	0			
1	1	<b>1</b>	0		
2	1	<b>2</b>	1	0	
3	1	<b>3</b>	3	1	0
4	1	<b>4</b>	6	4	1

Obviously, the coefficient  $\alpha_{1,n-1}$  of  $ax^{n-1}$  is  $n$

## Derivative of $x^n$

$$f(x+h) - f(x) = (x+h)^n - x^n$$

## Derivative of $x^n$

$$\begin{aligned}f(x+h) - f(x) &= (x+h)^n - x^n \\ &= (x^n + nhx^{n-1} + \alpha_{2,n-2}h^2x^{n-2} + \dots) - x^n\end{aligned}$$

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## Derivative of $x^n$

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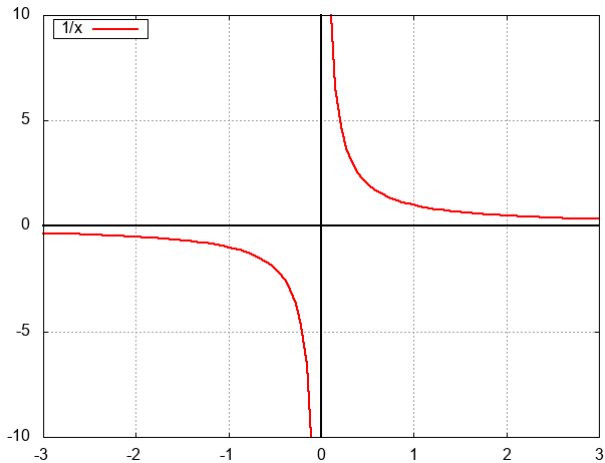
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$$\begin{aligned}\frac{f(x+h)-f(x)}{h} &= \frac{nhx^{n-1} + \alpha_{2,n-2}h^2x^{n-2} + \dots}{h} \\&= nx^{n-1} + \underbrace{\alpha_{2,n-2}hx^{n-2} + \dots}_0\end{aligned}$$

$$\lim_{h \rightarrow 0} \frac{f(x+h)-f(x)}{h} = nx^{n-1}$$

## Function $1/x$



# Derivative of $1/x$

$$f(x) = 1/x$$

The derivative should have the following behaviour

X	$-\infty$	0	$\infty$
x	-	NA	-

## Derivative of $1/x$

$$f(x+h) - f(x) = \frac{1}{x+h} - \frac{1}{x}$$

$$= \frac{x - (x+h)}{x(x+h)}$$

$$= \frac{-h}{x^2 + hx}$$

$$\frac{f(x+h) - f(x)}{h} = \frac{\frac{-h}{x^2 + hx}}{h}$$

$$= \frac{-h}{h(x^2 + hx)}$$

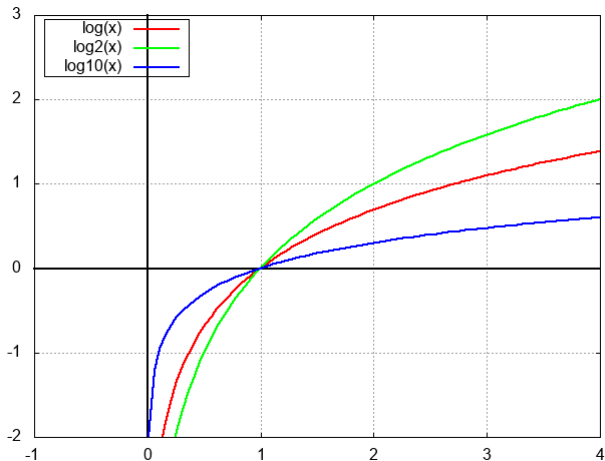
$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} = -\frac{1}{x^2}$$

## Function $\log(x)$

- $\ln(x)$  or  $\log(x)$
- the natural logarithm of  $x$  is the power to which  $e = 2.718281 \dots$  would have to be raised to equal  $x$
- for example  $\ln(7.5) = 2.0149 \dots$ , because  $e^{2.0149 \dots} = 7.5$
- used to replace products by sums
- other functions:

$$\log_n(x) = \frac{\ln(x)}{\ln(n)}$$

## Function $\log(x)$





# Derivative of $\log(x)$

$$f(x) = \log(x)$$

The derivative should have the following behaviour

X	$-\infty$	0	$\infty$
x	NA	$-\infty$	+

# Properties of the function $\log(x)$

## Properties of the function $\log(x)$

$$\log(1) = 0$$

$$\log(e) = 1$$

$$\log(x \times y) = \log(x) + \log(y)$$

$$\log(x/y) = \log(x) - \log(y)$$

$$\log(x^n) = n \times \log(x)$$

## Derivative of $\log(x)$

By definition

$$\log(a) = \int_1^a \frac{1}{x} dx$$

so the derivative of  $\log(x)$  is  $\frac{1}{x}$

## Derivative of $\log(x)$

$$f(x+h) - f(x) = \log(x+h) - \log(x)$$

## Derivative of $\log(x)$

$$\begin{aligned}f(x+h) - f(x) &= \log(x+h) - \log(x) \\ &= \log\left(\frac{x+h}{x}\right)\end{aligned}$$

## Derivative of $\log(x)$

$$\begin{aligned}f(x+h) - f(x) &= \log(x+h) - \log(x) \\&= \log\left(\frac{x+h}{x}\right) \\&= \log\left(1 + \frac{h}{x}\right)\end{aligned}$$

## Derivative of $\log(x)$

$$f(x+h) - f(x) = \log(x+h) - \log(x)$$

$$= \log\left(\frac{x+h}{x}\right)$$

$$= \log\left(1 + \frac{h}{x}\right)$$

$$\frac{f(x+h) - f(x)}{h} = \frac{1}{h} \times \log\left(1 + \frac{h}{x}\right)$$

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$$= \log\left(\left(1 + \frac{h}{x}\right)^{\frac{1}{h}}\right)$$



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$$= \log\left(\lim_{h \rightarrow 0} \left(1 + \frac{h}{x}\right)^{\frac{1}{h}}\right) = \log\left(e^{\frac{1}{x}}\right)$$

## Function $\exp(x)$ (Jakob Bernoulli (ch), 1654-1705)

- the exponential function aka the antilogarithm
- $\exp(x) = e^x = \lim_{n \rightarrow \infty} (1 + \frac{x}{n})^n$
- $e^1 = 2.718281 \dots$

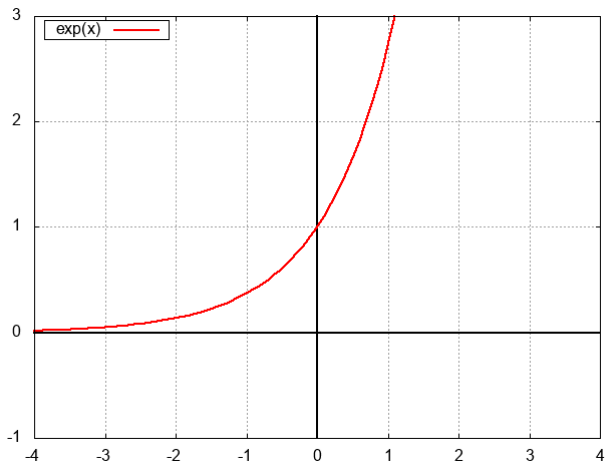
## Jakob Bernoulli (ch), 1654-1705)

- family, of Belgium origin, were refugees fleeing from persecution by the Spanish rulers of the Netherlands
- swiss mathematician and astronomer
- theory of permutations and combinations (Bernoulli numbers), by which he derived the exponential series
- Law of large numbers, in statistics, 1713



# Function $\exp(x)$

## Function $\exp(x)$



# Derivative of $\exp(x)$

$$f(x) = \exp(x)$$

The derivative should have the following behaviour

X	$-\infty$	0	$\infty$
x	+	1	+

# Properties of the function $e^x$

## Properties of the function $e^x$

$$e^0 = 1$$

$$e^x > 0 \quad \forall x$$

$$e^{-x} = \frac{1}{e^x}$$

$$e^{x+y} = e^x \times e^y$$

$$e^{x-y} = \frac{e^x}{e^y}$$

$$e^{x \times y} = (e^x)^y$$

## Derivative of $e^x$

By definition  $\log(e^x) = x$ . So we compute the derivative of this last expression:

$$\log(e^x) = x$$



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$$(\log(e^x))' = (x)'$$

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$$\frac{1}{e^x} \times e'(x) = 1 \quad \text{by(DerivInv)}$$

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$$(\log(e^x))' = (x)'$$

$$\frac{1}{e^x} \times e'(x) = 1 \quad \text{by(DerivInv)}$$

$$e'(x) = e(x)$$

## Derivatives

$$(x^n)' = nx^{n-1}$$

$$\left(\frac{1}{x}\right)' = -\frac{1}{x^2}$$

$$(\log(x))' = \frac{1}{x}$$

$$(e(x))' = e^x$$



## 4. Gradient

## Definition of the gradient

Given a function of several variables  $f(x, y, z)$ , the gradient is the **vector of the partial derivatives** of  $f$

$$\nabla f(x, y, z) = \nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right]$$

The **partial derivative**  $\frac{\partial f}{\partial x}$  is the derivative of  $f(x, y, z)$  when  $y$  and  $z$  are considered as constants:

$$\frac{\partial f}{\partial x} = \frac{\partial f(x, y, z)}{\partial x} = \frac{df(x, y, z)|_{y,z}}{dx}$$

## Property of the gradient

- the gradient  $\nabla f$  gives the direction toward which you can increase the value of the function
- conversly  $-\nabla f$  gives the direction toward which you can decrease the value of the function

## Finding the minimum of a function

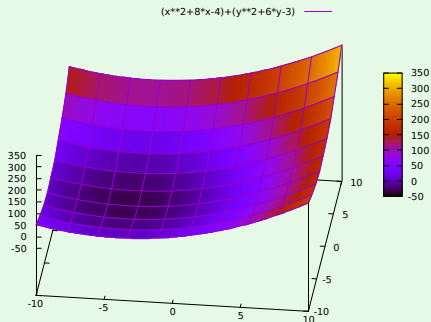
- the gradient can be used to find the minimum of a function by progressively decreasing the coordinates by subtracting
- a fraction of the value of the gradient



## A convex quadratic function

Consider the following function:

$$f(x, y) = (x^2 + 8 \times x - 4) + (y^2 + 6 \times y - 3)$$



## A convex quadratic function

The gradient of the function is:

$$\frac{\partial f}{\partial x} = 2 \times x + 8$$

$$\frac{\partial f}{\partial y} = 2 \times y + 6$$

The minimum is found for  $\frac{\partial f}{\partial x} = 0$  and  $\frac{\partial f}{\partial y} = 0$

$$\frac{\partial f}{\partial x} = 0 \Rightarrow x = -4$$

$$\frac{\partial f}{\partial y} = 0 \Rightarrow y = -3$$

## Gradient descent algorithm

**Data:**  $f(x)$

**Result:**  $x^*$ : the minimum of the function

initialise vector  $x$ ;

**while not** *terminate\_condition* **do**

    compute gradient  $\nabla f$ ;

$x = x - \alpha \times \nabla f$ ;

**end**

**Algorithm 1:** A very simple descent algorithm

- note that the *Terminate Condition* can be defined in different ways (improvement, number of iterations)
- $\alpha = 0.1$  for example, if too big the algorithm won't find the solution

## Descent for a convex quadratic function

For the previous convex function we obtain this:

$x_0 = 3, y_0 = 5, \alpha = 0.1$

gradient = (14, 16)

$x_1 = 1.5999, y_1 = 3.4$

gradient = (11.2, 12.8)

$x_2 = 0.48, y_2 = 2.1199$

gradient = (8.96, 10.2399)

...

$x_{48} = -3.99984389478361, y_{48} = -2.999821594038412$

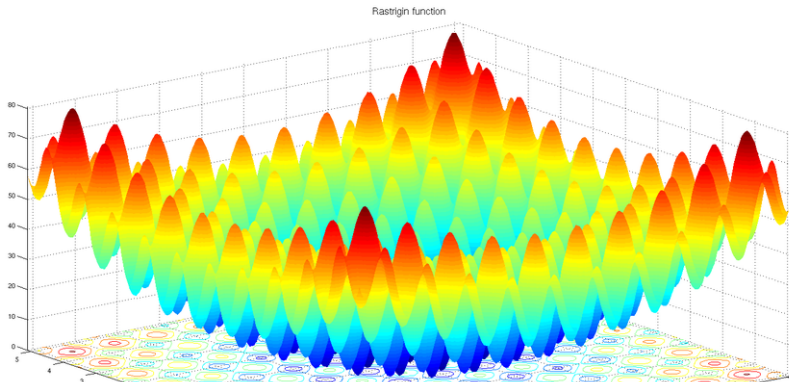
gradient = (0.0003122104327797359, 0.000356811923175826)

$x_{49} = -3.999875115826888, y_{49} = -2.9998572752307298$

## Difficulty of finding the minimum

it becomes more difficult to find the minimum if

- the function is not convex
- the function has many minima (Rastrigin or Himmelblau functions)



## In Python

```
from scipy import optimize

def f(x):
    return x[0]**2+8*x[0]-4+x[1]**2+6*x[1]-3

def fprime(x):
    return np.array([(2*x[0]+8), (2*x[1]+6)])

z = optimize.fmin_bfgs(f, [3, 5], fprime=fprime)
print(z)
```

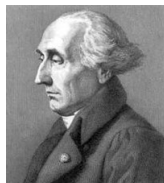
```
Optimization terminated successfully.
    Current function value: -32.000000
    Iterations: 2
    Function evaluations: 4
    Gradient evaluations: 4
[-4. -3.]
```



## 5. Lagrangian

## Joseph-Louis Lagrange

- born **Giuseppe Lodovico Lagrangia** (it,fr) (1736 - 1813) was a franco-italian mathematician and astronomer
- made significant contributions to the fields of analysis, number theory, and both classical and celestial mechanics
- in **1787**, at age 51, moved from Berlin to Paris and became a member of the **French Academy of Sciences**
- remained in France until the end of his life





## Principle

- you want to minimize or maximize  $f(x)$  subject to  $g(x) = 0$
- under certain conditions
  - ▶  $f(x)$  is a quadratic function
  - ▶  $g(x)$  are linear constraints
- define the function

$$\mathcal{L}(x, \alpha) = f(x) + \alpha g(x)$$

where  $\alpha \geq 0 \in \mathbb{R}$  is called the **lagrangian multiplier**

## Resolution

- a solution of  $\mathcal{L}(x, \alpha)$  is a point of gradient 0
- so compute and solve

$$\frac{\partial \mathcal{L}(x, \alpha)}{\partial x} = 0$$

$$\frac{\partial \mathcal{L}(x, \alpha)}{\partial \alpha} = g(x) = 0$$

- or reuse in  $\mathcal{L}(x, \alpha)$

## Statement of the example

Suppose you want to put a fence around some field which as a form of a rectangle  $(x, y)$  and you want to maximize the area knowing that you have  $P$  meters of fence:

$$\begin{cases} \text{Max} & x \times y \\ \text{such that} & P = 2x + 2y \end{cases}$$

then  $f(x, y) = xy$  and  $g(x, y) = P - 2x - 2y = 0$

$$\begin{cases} \text{Max} & xy \\ \text{such that} & P - 2x - 2y = 0 \end{cases}$$

## Lagrange formulation

$$\mathcal{L}(x, y, \alpha) = xy + \alpha(P - 2x - 2y)$$

the derivatives give us

$$\frac{\partial \mathcal{L}(x, y, \alpha)}{\partial x} = y - 2\alpha = 0$$

$$\frac{\partial \mathcal{L}(x, y, \alpha)}{\partial y} = x - 2\alpha = 0$$

$$\frac{\partial \mathcal{L}(x, y, \alpha)}{\partial \alpha} = P - 2x - 2y = 0$$

## Resolution

- the first two constraints give us  $y = 2\alpha = x$ , so  $x = y$
- in other words, the area is a square
- and the last one that  $P = 4x = 4y$
- consequently  $\alpha = P/8$  because  $\alpha = x/2 = y/2$



## 6. Exercises

## Matrix - Matrix multiplication

Consider the following matrices:

$$A = \begin{bmatrix} 1 & 0 & -2 \\ 4 & 5 & -3 \\ 6 & -3 & -2 \end{bmatrix} \quad B = \begin{bmatrix} 0.5 & 7 & 3 \\ 8 & -6 & 2 \\ 1 & -2 & 3 \end{bmatrix}$$

Perform the products by hand and compare:

$$A \times B \stackrel{?}{=} B \times A$$

## Matrix - Vector multiplication

Consider the following matrix and vector:

$$A = \begin{bmatrix} 1 & 0 & -2 \\ 4 & 5 & -3 \\ 6 & -3 & -2 \end{bmatrix} \quad x = \begin{bmatrix} 0.5 \\ 8 \\ 1 \end{bmatrix}$$

Perform the products by hand and compare:

$$A \times x \stackrel{?}{=} x^T \times A$$



## Vector - Vector multiplication

Consider the following vectors:

$$x = \begin{bmatrix} 1 \\ 4 \\ 6 \end{bmatrix} \quad y = \begin{bmatrix} -3 \\ 8 \\ 1 \end{bmatrix}$$

Perform the following operations:

$$x \times y \quad \text{and} \quad x \odot y$$

## Check results in python

Write a program in python to check the results of the different matrix and vector products

## Derivative rules

Remember to use the following rules:

$$(f + g)' = f' + g' \quad (dSum)$$

$$(f \times g)' = f' \times g + f \times g' \quad (dProd)$$

$$f(g(x))' = f'(g(x)) \times g'(x) \quad (dComp)$$

## Derivatives

write a python program to draw the following functions (use matplotlib) and compute their derivatives by hand:

$$f_1(x) = \frac{1 + \frac{1}{x}}{x-3}$$

$$f_2(x) = \frac{1}{x^2 + e^x}$$

$$f_3(x) = \frac{x \times e^x}{1 + e^x}$$

## Derivatives

The results are the following

$$f'_1(x) = -\frac{x^2 + 2x - 3}{(x - 3)^2 x^2}$$

$$f'_2(x) = -\frac{e^x + 2x}{(e^x + x^2)^2}$$

$$f'_3(x) = \frac{e^x (e^x + x + 1)}{(e^x + 1)^2}$$

## Derivative of $f_1(x)$

$$f_1(x) = \left(1 + \frac{1}{x}\right) \times \frac{1}{x-3}$$

$$f_1(x) = F(x) \times G(x)$$

So we need to apply the formula (*dProd*)

## Derivative of $f_1(x)$

$$F(x) = \left(1 + \frac{1}{x}\right)$$

$$F'(x) = -\frac{1}{x^2}$$

$$G(x) = \frac{1}{x-3} = H(K(x))$$

with

$$H(z) = \frac{1}{z}$$

$$K(z) = z - 3$$

## Derivative of $f_1(x)$

$$G'(x) = H'(K(x)) \times K'(x)$$

$$G'(x) = -\frac{1}{(x-3)^2} \times 1$$



## Derivative of $f_1(x)$

Finally

$$f_1'(x) = -\frac{1}{x^2} \times \frac{1}{x-3} + \left(1 + \frac{1}{x}\right) \times -\frac{1}{(x-3)^2}$$

$$f_1'(x) = -\frac{1}{x^2} \times \frac{x-3}{x-3} + 1 + \frac{1}{x} \times -\frac{x}{(x-3)^2}$$

$$f_1'(x) = -\frac{(x-3) + (x+1)x}{x^2(x-3)^2}$$

$$f_1'(x) = -\frac{x^2 + 2x - 3}{x^2(x-3)^2}$$

## Derivative of $f_2(x)$

$$f_2(x) = \frac{1}{x^2 + e^x}$$

$$f_2(x) = F(G(x))$$

So we need to apply the formula (*dComp*) with

$$F(z) = \frac{1}{z}$$

$$G(z) = z^2 + e^z$$

## Derivative of $f_2(x)$

$$F'(z) = -\frac{1}{z^2}$$

$$G'(z) = 2z + e^z \quad (dSum)$$

## Derivative of $f_2(x)$

Finally

$$f_2'(x) = -\frac{1}{x^2 + e^x} \times (2x + e^x)$$

$$f_2'(x) = -\frac{2x + e^x}{x^2 + e^x}$$

## Derivatives

- check the results with the **derivative calculator**
- write a program in python using sympy to compute the derivatives of the functions

## Gradient

- determine where  $f_1(x)$  is minimum
- using the gradient method try to determine where  $f_2(x)$  and  $f_3(x)$  are maximum or minimum

## Gradient/derivative of $f_1$

The derivative of  $f_1$  is:

$$f_1'(x) = -\frac{x^2 + 2x - 3}{x^2(x - 3)^2}$$

There are extremum (maximum or minimum) where

$$f_1'(x) = 0$$

The function is not defined if the denominator is equal to  $x^2(x - 3)^2 = 0$ :

$$\begin{cases} x^2 = 0 & \Rightarrow x = 0 \\ (x - 3)^2 = 0 & \Rightarrow x = 3 \end{cases}$$

## Gradient/derivative of $f_1$

The derivative is equal to 0 if:

$$x^2 + 2x - 3 = 0$$

$$\left\{ \begin{array}{l} \Delta = b^2 - 4ac = 2^2 - 4 \times 1 \times -3 = 16 \\ x_1 = \frac{-b - \sqrt{\Delta}}{2a} = \frac{-2 - 4}{2 \times 1} = -3 \\ x_2 = \frac{-b + \sqrt{\Delta}}{2a} = \frac{-2 + 4}{2 \times 1} = +1 \end{array} \right.$$

then

$$x^2 + 2x - 3 = (x - 1)(x + 3)$$



## Gradient/derivative of $f_2$

The derivative of  $f_2$  is:

$$f_2'(x) = -\frac{2x + e^x}{x^2 + e^x}$$

The denominator  $x^2 + e$  is always positive so we need to solve:

$$2x + e^x = 0$$

which is not possible by analytical methods we need to find the root by using an approximation method

## Gradient/derivative of $f_2$ (1/2)

```
import numpy as np

def gradient_ascent(x, df):
    delta = 0.0000001
    alpha = 0.1
    while True:
        x_n = x + alpha * df(x)
        if math.fabs(x-x_n) < delta:
            break
        x = x_n
    return x
```

## Gradient/derivative of $f_2$ (2/2)

```
def f2(x):  
    return 1/(x*x + np.exp(x))  
  
def df2(x):  
    return (-2*x - math.exp(x))/(x**2 + math.exp(x))**2  
  
x2_star = gradient_ascent2(-0.5, df2)  
print("f2:␣", x2_star, "␣=>␣", f2(np.asarray([x2_star  
    ])))
```

## Lagrangian

Consider the following problem

$$\left\{ \begin{array}{l} \text{Max } x^2 + y^2 + z^2 \\ \text{such that} \\ x + 2y + z = 1 \\ 2x - y - 3z = 4 \end{array} \right.$$

- use the method of Lagrange to solve it to determine  $x$ ,  $y$  and  $z$

## Lagrangian resolution

We define

$$\begin{aligned}\mathcal{L}(x, y, z, \alpha_1, \alpha_2) = & f(x, y, z) && + \\ & \alpha_1(x + 2y + z - 1) && + \\ & \alpha_2(2x - y - 3z - 4)\end{aligned}$$

with  $f(x, y, z) = x^2 + y^2 + z^2$

## Lagrangian resolution

We need to compute the partial derivatives of  $\mathcal{L}$  for each variable :

$$\frac{\partial \mathcal{L}(x,y,z,\alpha_1,\alpha_2)}{\partial x} = 2x + \alpha_1 + 2\alpha_2 = 0 \quad (1)$$

$$\frac{\partial \mathcal{L}(x,y,z,\alpha_1,\alpha_2)}{\partial y} = 2y + 2\alpha_1 - \alpha_2 = 0 \quad (2)$$

$$\frac{\partial \mathcal{L}(x,y,z,\alpha_1,\alpha_2)}{\partial z} = 2z + \alpha_1 - 3\alpha_2 = 0 \quad (3)$$

$$\frac{\partial \mathcal{L}(x,y,z,\alpha_1,\alpha_2)}{\partial \alpha_1} = x + 2y + z - 1 = 0 \quad (4)$$

$$\frac{\partial \mathcal{L}(x,y,z,\alpha_1,\alpha_2)}{\partial \alpha_2} = 2x - y - 3z - 4 = 0 \quad (5)$$

## Lagrangian resolution

Express  $\alpha_1$  from  $x$  and  $\alpha_2$  in (1)

$$-2x - 2\alpha_2 = \alpha_1 \quad (1)$$

$$2y + 2\alpha_1 - \alpha_2 = 0 \quad (2)$$

$$2z + \alpha_1 - 3\alpha_2 = 0 \quad (3)$$

$$x + 2y + z - 1 = 0 \quad (4)$$

$$2x - y - 3z - 4 = 0 \quad (5)$$

## Lagrangian resolution

Then replace in (2) and (3)

$$-2x - 2\alpha_2 = \alpha_1 \quad (1)$$

$$2y + 2(-2x - 2\alpha_2) - \alpha_2 = 0 \Rightarrow 2y - 4x - 5\alpha_2 = 0 \quad (2)$$

$$2z + (-2x - 2\alpha_2) - 3\alpha_2 = 0 \Rightarrow 2z - 2x - 5\alpha_2 = 0 \quad (3)$$

$$x + 2y + z - 1 = 0 \quad (4)$$

$$2x - y - 3z - 4 = 0 \quad (5)$$



## Lagrangian resolution

By subtracting (2) and (3) we get

$-2x + 2y - 2z = x - y + z = 0$ . And finally we have a system of 3 equations with 3 variables:

$$x - y + z = 0 \quad (6) = (2) - (3)$$

$$x + 2y + z - 1 = 0 \quad (4)$$

$$2x - y - 3z - 4 = 0 \quad (5)$$

## Lagrangian resolution

Then we compute (6) – (4) and obtain  $y = \frac{1}{3}$ .

The rest of the resolution is obvious and we should get

$$x = \frac{16}{15}$$

$$y = \frac{1}{3}$$

$$z = \frac{-11}{15}$$

$$\alpha_1 = -\frac{52}{75}$$

$$\alpha_2 = -\frac{54}{75}$$

## Lagrangian resolution

In Python:

```
import numpy as np
A = np.asarray([[2,0,0,-1,-2], [0,2,0,-2,1],
               [0,0,2,-1,3], [1,2,1,0,0], [2,-1,-3,0,0]])
b = np.asarray([0,0,0,1,4])
x = np.linalg.solve(A, b)
print(x)
```



6. End



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