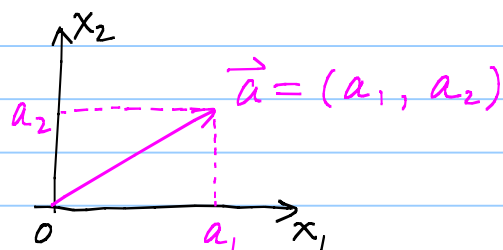


Math Background :

1. From Linear Algebra.

- Vectors



Magnitude (length) :

$$\|\vec{a}\| = \sqrt{a_1^2 + a_2^2}$$

In n -D, $\vec{a} = (a_1, a_2, \dots, a_n)$.

$$\|\vec{a}\| = \sqrt{a_1^2 + a_2^2 + \dots + a_n^2}$$

Given 2 vectors, $\vec{a} = (a_1, a_2, \dots, a_n)$,

$$\vec{b} = (b_1, b_2, \dots, b_n)$$

The DOT PRODUCT between them is

$$\vec{a} \cdot \vec{b} = a_1 b_1 + a_2 b_2 + \dots + a_n b_n$$

An application of Dot Product:

A diagram showing two vectors \vec{a} and \vec{b} originating from the same point. The angle between them is labeled θ . The formula for the cosine of the angle is given as $\cos \theta = \frac{\vec{a} \cdot \vec{b}}{\|\vec{a}\| \cdot \|\vec{b}\|}$.

- Matrix:

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}_{2 \times 3}$$

$$B = \begin{bmatrix} 1 & 2 & 1 \\ 3 & 2 & 1 \\ 4 & 5 & 2 \\ 6 & 7 & 10 \end{bmatrix}_{4 \times 3}$$

- Matrix Addition (& Subtraction):

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$A + B = \begin{bmatrix} a_{11} + b_{11} & a_{12} + b_{12} \\ a_{21} + b_{21} & a_{22} + b_{22} \end{bmatrix}$$

- Matrix Multiplication

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 0 & 6 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 2 & 0 \\ 1 & 3 & 1 \\ 1 & 4 & 2 \end{bmatrix}$$

$$AB = ?$$

Key Idea:

Take each row of A, > dot product!
each column of B,

First, a matrix times a column vector:

$$A \vec{b} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} a_{11}b_1 + a_{12}b_2 + a_{13}b_3 \\ a_{21}b_1 + a_{22}b_2 + a_{23}b_3 \end{bmatrix}$$

2×3 3×1 2×1

Note: A matrix times a vector
is a vector.

Next, a matrix times a matrix:

$$AB = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \\ b_{31} & b_{33} \end{bmatrix} = C$$

2×3 3×2 2×2

Must match

$$C_{11} = [a_{11} \ a_{12} \ a_{13}] \begin{bmatrix} b_{11} \\ b_{21} \\ b_{31} \end{bmatrix}$$

$$= a_{11}b_{11} + a_{12}b_{21} + a_{13}b_{31}$$

Pick 2nd column from B

$$c_{12} = [a_{11} \ a_{12} \ a_{13}] \begin{bmatrix} b_{12} \\ b_{22} \\ b_{32} \end{bmatrix}$$

Pick 1st row from A

$$= a_{11}b_{12} + a_{12}b_{22} + a_{13}b_{32}$$

$$c_{21} = [a_{21} \ a_{22} \ a_{23}] \begin{bmatrix} b_{11} \\ b_{21} \\ b_{31} \end{bmatrix}$$

$$= a_{21}b_{11} + a_{22}b_{21} + a_{23}b_{31}$$

$$c_{22} = [a_{21} \ a_{22} \ a_{23}] \begin{bmatrix} b_{12} \\ b_{22} \\ b_{32} \end{bmatrix}$$

$$c_{22} = a_{21}b_{12} + a_{22}b_{22} + a_{23}b_{32}$$

eg.

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 0 & 6 \end{bmatrix}$$

2x3

$$B = \begin{bmatrix} 1 & 2 & 0 \\ 1 & 3 & 1 \\ 1 & 4 & 2 \end{bmatrix}$$

3x3

$$AB = \begin{bmatrix} 6 & 20 & 8 \\ 10 & 32 & 12 \end{bmatrix}$$

$$BA = ?$$

$$= \begin{bmatrix} 1 & 2 & 0 \\ 1 & 3 & 1 \\ 1 & 4 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 4 & 0 & 6 \end{bmatrix} = \text{nonsense!}$$

3x3

2x3

Must match!

Q: How about $\frac{A}{B} = ?$

In the world of numbers:

for any number $a \neq 0$,

$$a \cdot a^{-1} = a^{-1} \cdot a = 1.$$

In the world of matrices,

Given an $n \times n$ matrix A ,

if you can find another matrix B ,

such that

$$AB = BA = I = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}_{n \times n}.$$

Identity matrix.

then B is the inverse matrix of A .

Notation: $B = A^{-1}$.

For 2×2 matrices, a very quick way—

eg. $A = \begin{bmatrix} 3 & 1 \\ 7 & 9 \end{bmatrix}$, $A^{-1} = ?$

$$A^{-1} = \begin{bmatrix} 9/20 & -1/20 \\ -7/20 & 3/20 \end{bmatrix}.$$

$$\text{Check: } \frac{1}{20} \begin{bmatrix} 9 & -1 \\ -7 & 3 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 7 & 9 \end{bmatrix}$$

$$= \frac{1}{20} \begin{bmatrix} 27-7 & 9-9 \\ -21+21 & -7+27 \end{bmatrix}$$

$$= \frac{1}{20} \begin{bmatrix} 20 & 0 \\ 0 & 20 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

Secret ?

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad A^{-1} = ?$$

$$A^{-1} = \frac{1}{ad-cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Concept: determinant

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - cb$$

How about the 3×3 case ?

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = ?$$

$$= a_{11} a_{22} a_{33} + a_{21} a_{32} a_{13} + a_{31} a_{12} a_{23}$$

$$- a_{31} a_{22} a_{13} - a_{11} a_{32} a_{23} - a_{21} a_{12} a_{33}$$

A better way: expand along any row
or any column.

$$\det(A) = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$= (-1)^{1+1} a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} + (-1)^{1+2} a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}$$

$$+ (-1)^{1+3} a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

Now, matrix inverse for 3×3 case:

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$$

$$A^{-1} = \frac{1}{\det(A)} A^*$$

Concept: minor

Pick a_{ij} , the minor of a_{ij}

is the determinant consisting of the remaining numbers by deleting the i^{th} row and j^{th} column where a_{ij} lies.

$$\text{The minor of } a_{11} \text{ is } A_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix},$$

$$\therefore a_{22} \text{ " } A_{22} = \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix}.$$

The Cofactor :

$$C_{ij} = (-1)^{i+j} A_{ij}$$

$$A^* = \begin{pmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{pmatrix}$$

In matrix jargon : transposed.

$$\text{If } A = \begin{pmatrix} a & b & c \\ d & e & f \end{pmatrix}$$

$$A^T = \begin{pmatrix} a & d \\ b & e \\ c & f \end{pmatrix}.$$

Here, in our inverse matrix,

$$A^* = \text{Cof}(A)^T$$

cofactor

$$A^{-1} = \frac{1}{\det} \text{Cof}(A)^T$$

eg. $A = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 0 & 3 \\ 1 & 4 & 5 \end{pmatrix}$ $A^{-1} = ?$

Soln: $\det(A) = (-1)^{1+1} \cdot 1 \cdot \begin{vmatrix} 0 & 3 \\ 4 & 5 \end{vmatrix}$

$$+ (-1)^{1+2} \cdot 1 \cdot \begin{vmatrix} 2 & 3 \\ 1 & 5 \end{vmatrix}$$

$$+ (-1)^{1+3} \cdot 1 \cdot \begin{vmatrix} 2 & 0 \\ 1 & 4 \end{vmatrix}$$

$$= -12 - (10 - 3) + 8 = -11.$$

$$C_{11} = (-1)^{1+1} \begin{vmatrix} 0 & 3 \\ 4 & 5 \end{vmatrix} = -12$$

$$C_{12} = (-1)^{1+2} \begin{vmatrix} 2 & 3 \\ 1 & 5 \end{vmatrix} = -7$$

$$C_{13} = (-1)^{1+3} \begin{vmatrix} 2 & 0 \\ 1 & 4 \end{vmatrix} = 8$$

$$C_{21} = (-1)^{2+1} \begin{vmatrix} 1 & 1 \\ 4 & 5 \end{vmatrix} = -1$$

$$C_{22} = 4 \quad C_{23} = -3$$

$$C_{31} = 3 \quad C_{32} = -1$$

$$C_{33} = -2$$

$$A^{-1} = \frac{1}{\det(A)} \text{Cof}(A)^T$$

$$= \frac{1}{\det(A)} \begin{pmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{pmatrix}$$

$$= \frac{1}{-11} \begin{pmatrix} -12 & -1 & 3 \\ -7 & 4 & -1 \\ 8 & -3 & -2 \end{pmatrix} .$$

2. From Calculus :

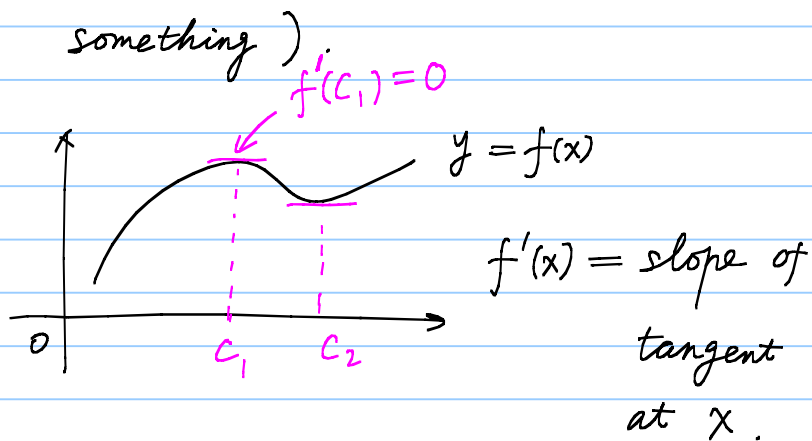
- Derivative :

Given a function $y = f(x)$,

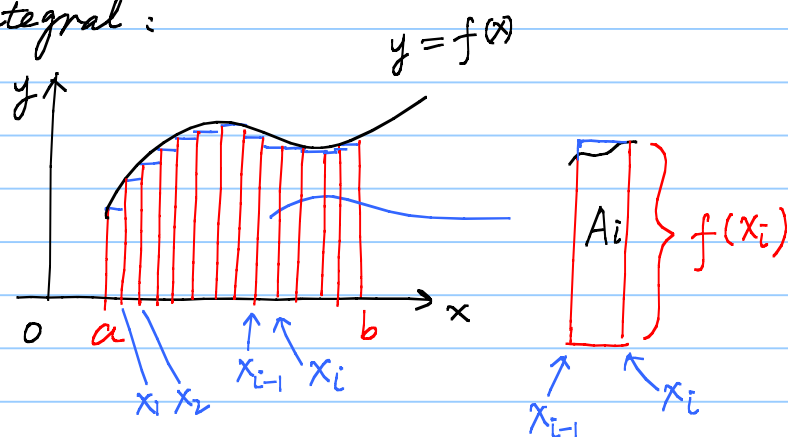
the derivative of $f(x)$ is

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

Application: to solve optimization problems. (to maximize or minimize something)



— Integral:



$$A_i \approx \text{height} \times \text{base}$$

$$\Delta x = x_i - x_{i-1}$$

$$= f(x_i) \Delta x,$$

$$A \approx A_1 + A_2 + \dots + A_n$$

$$= \sum_{i=1}^n A_i$$

$$= \sum_{i=1}^n f(x_i) \Delta x$$

How to make the approximation exact?

Take the limit, let $n \rightarrow \infty$.

$$A = \lim_{n \rightarrow \infty} \sum_{i=1}^n f(x_i) \Delta x$$
$$= \int_a^b f(x) dx$$

3. From Probability:

Sample Point = any possible outcome.

S = Sample Space = ALL possible outcome.
(a set).

E = Event = Any subset of S .

$$P(E) = \frac{\# \text{ of sample point in } E}{\# \text{ of } \cdot \cdot \cdot \text{ in } S}.$$

eg. Toss a die: $S = \{1, 2, 3, 4, 5, 6\}$

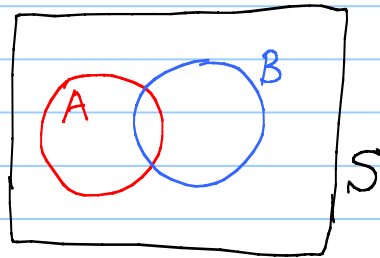
$E = \text{"less than 3"} = \{1, 2\}$

$$P(E) = \frac{2}{6} = \frac{1}{3}.$$

Conditional Probability:

$P(A|B)$ = Probability of the event A
occurs, given the condition

B has already occurred.

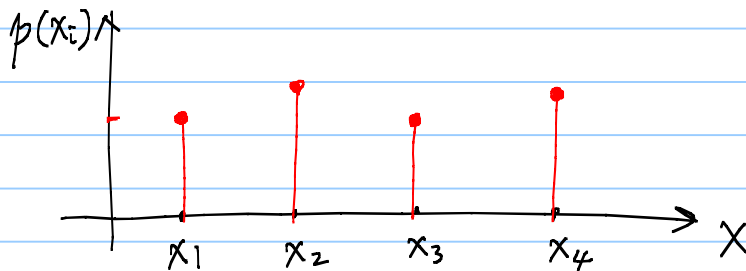


$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Key Concept: Mean & Variance.

constant \leftrightarrow variable
fixed \leftrightarrow changing.

Random Variable: a variable which can take some possible values with certain probability.



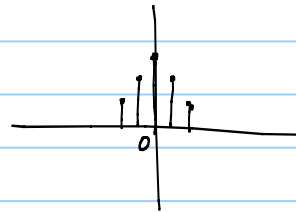
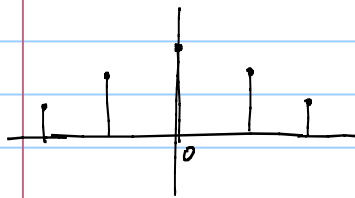
a random variable X can take possible values x_1, x_2, \dots, x_n with probability p_1, p_2, \dots, p_n .

The mean (expected value) of X is

$$\begin{aligned}\mu = E[X] &= x_1 p_1 + x_2 p_2 + \dots + x_n p_n \\ &= \sum_{i=1}^n x_i p_i\end{aligned}$$

Variance:

$$\begin{aligned}\sigma^2 = \text{Var}(X) &= E[(X - \mu)^2] \\ &= \sum_{i=1}^n (x_i - \mu)^2 p_i\end{aligned}$$



Given 2 random variables X and Y ,
How to describe their relative
relationship?

Covariance

$$\sigma_{XY} = E[(X - \mu_X)(Y - \mu_Y)]$$

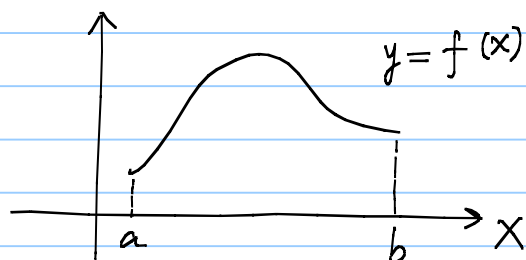
For discrete random variables,

$$E[X] = \sum_{i=1}^n x_i p_i$$

For Continuous random variables

Need to know the

probability density function (PDF)

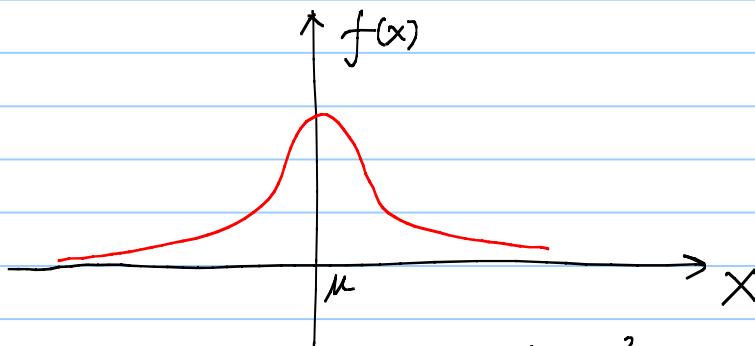


$$\mu = E[X] = \int_a^b x f(x) dx$$

$$\sigma^2 = E[(X - \mu)^2] = \int_a^b (x - \mu)^2 f(x) dx.$$

- Gaussian Distribution

(Normal Distribution)



$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

A normal distributed random variable

X with mean μ , variance σ^2 .

Notation: $X \sim N(\mu, \sigma^2)$,

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