

# REVIEW OF PROBABILITIES

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# TOPICS

- Motivation
- Notations and definitions
- Review of probabilities

# THE CONCEPT OF PROBABILITY

- Actual repeated experiments
  - Example: You record the color of 1000 cars driving by 57 of them are green. You estimate the probability of a car being green as  $\frac{57}{1000} = 0.057$
- In idealized conceptions of repeated process
  - Example: You consider the behavior of an unbiased six-sided die. The expected probability of rolling a 5 is  $\frac{1}{6} = 0.1667$
  - Example: You need a model for how people's heights are distributed. You choose a normal distribution (bell-shaped curve) to represent the expected relative probabilities

# NOTATION

- $a \in A$  : Set membership:  $a$  is a member of set  $A$
- $|B|$  : Cardinality: number of items in set  $B$
- $\|\nu\|$  : Norm: length of a vector  $\nu$
- $\sum$  : Summation
- $\int$ : Integral

# NOTATION

- $\mathbf{x}, \mathbf{y}, \mathbf{z}$  : Vector (bold, lower case)
- $\mathbf{A}, \mathbf{B}, \mathbf{X}$  : Matrix (bold, upper case)
- $y = f(x)$  : Function (map)
- $\frac{dy}{dx}$  : Derivative of  $y$  with respect to single variable  $x$
- $\frac{\partial y}{\partial x_i}$  : Partial derivative of  $y$  with respect to element  $i$  of vector  $x$

# AXIOMS OF PROBABILITY

- Non-negativity:

for any event  $E \in F$ ,  $p(E) \geq 0$

- All possible outcomes, where  $\Omega$  is sample space:

$$p(\Omega) = 1$$

- Additivity of disjoint events:

for all events  $E, G \in F$  where  $E \cap G = \emptyset$ ,

$$p(E \cup G) = p(E) + p(G)$$

# TYPES OF PROBABILITY SPACES

Define  $|\Omega| = \text{number of possible outcomes}$

- Discrete space  $|\Omega|$  is finite
  - Analysis involves summations:  $\sum$
- Continuous space  $|\Omega|$  is infinite
  - Analysis involves integrals:  $\int$

# EXAMPLE: DISCRETE PROBABILITY SPACE

Single roll of a six-sided die

- 6 possible outcomes:  $O = \{1, 2, 3, 4, 5, 6\}$
- If die is fair, then probability of outcomes are equal

$$p[i] = \frac{1}{6} \text{ for } i \in \{1, 2, 3, 4, 5, 6\}$$

- What is the probability of event  $E = (\text{outcome is odd})$ ?

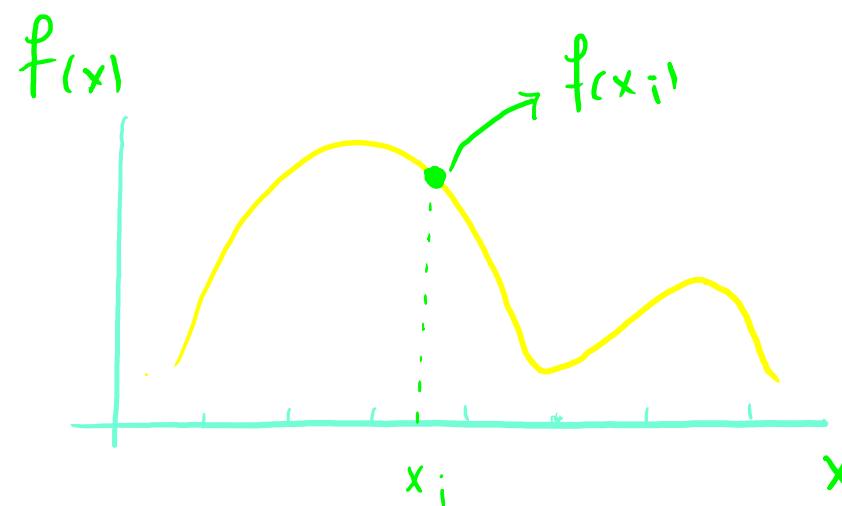
# EXAMPLE 2

Consider 3 consecutive flips of a coin

- How many possible outcomes?
- What is the probability that we see exactly two heads?

# CONTINUOUS PROBABILITY SPACE

- Infinite number of possible outcomes
- Probabilities of outcomes are not equal, and are described by a continuous function.

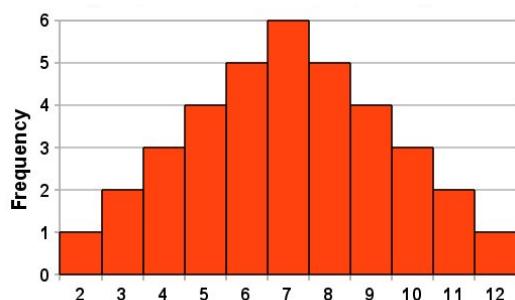


# PROBABILITY FUNCTIONS

## DISCRETE

Example: sum of two fair dice

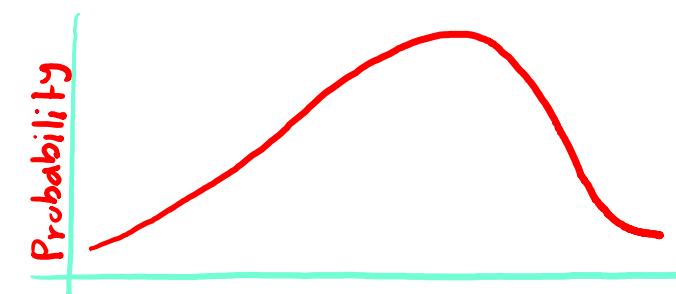
Probability mass function (pmf)



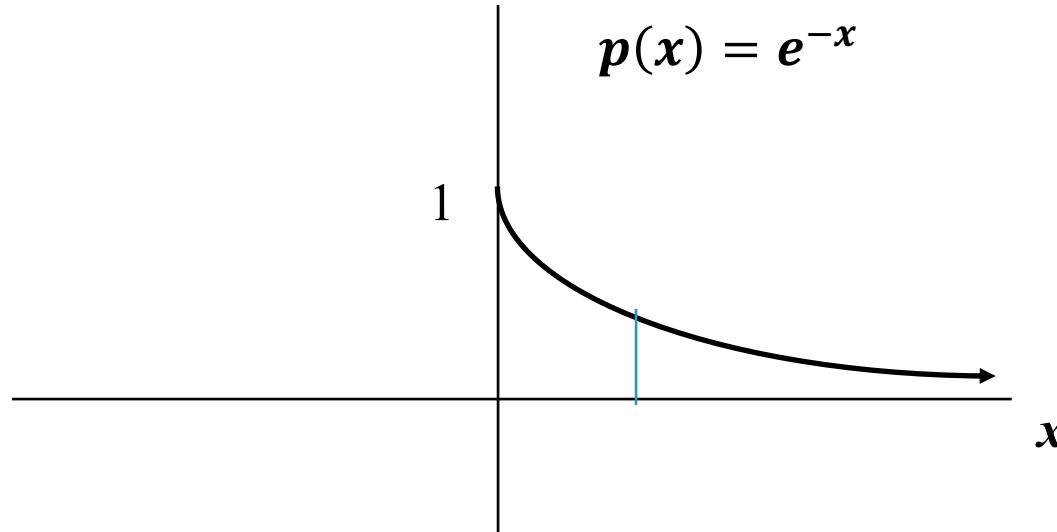
## CONTINUOUS

Example: waiting time for doctor appointment

Probability density function (pdf)

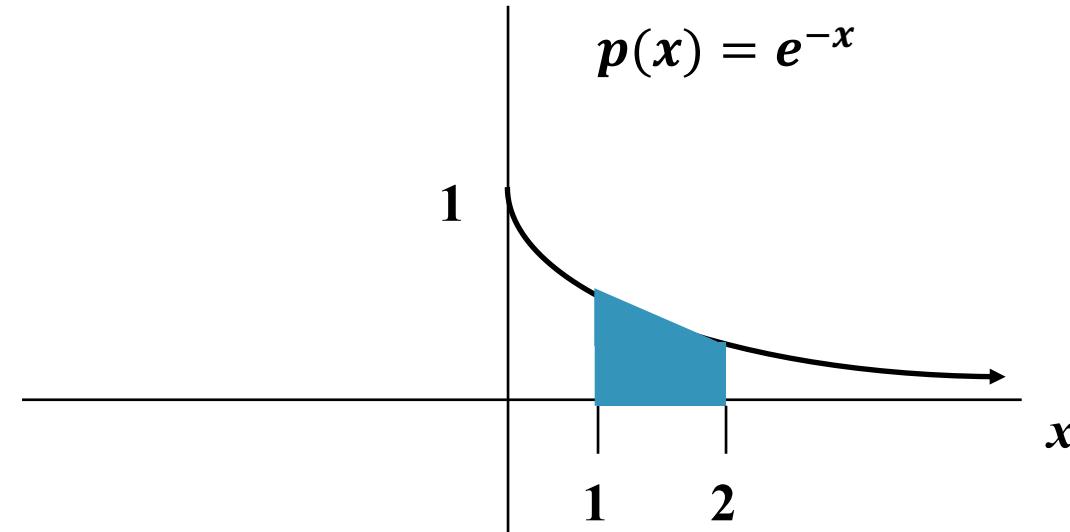


# CONTINUOUS CASE: “PROBABILITY DENSITY FUNCTION” (PDF)



The probability that  $x$  is any exact particular value (such as 1.9976) is 0; we can only assign probabilities to possible ranges of  $x$ .

# THE PROBABILITY OF $x$ FALLING WITHIN 1 TO 2:



$$P(1 \leq x \leq 2) = \int_1^2 e^{-x} dx = -e^{-x} \Big|_1^2 = -e^{-2} - -e^{-1} = -.135 + .368 = .23$$

# CUMULATIVE DISTRIBUTION

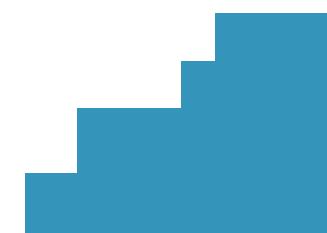
- Cumulative distribution function  $F(x)$  is the probability of getting at most  $x$ .

When playing two dice the sum of outcomes lies between  $2 - 12$ . Using cumulative distribution we can easily find probabilities for different events:

$$p(X < 7) = \frac{15}{36}$$

$$p(X > 9) = 1 - \frac{30}{36} = \frac{6}{36}$$

$$p(4 < X < 9) = \frac{26}{36} - \frac{6}{36} = \frac{20}{36}$$



x	F(x)
2	1/36
3	3/36
4	6/36
5	10/36
6	15/36
7	21/36
8	26/36
9	30/36
10	33/36
11	35/36
12	36/36

# CUMULATIVE DISTRIBUTION FUNCTION

- As in the discrete case, we can specify the “cumulative distribution function” (CDF)
- The CDF is  $p(x) = e^{-x}$

$$p(x \leq A) = \int_0^A e^{-x} = -e^{-x} \Big|_0^A = -e^{-A} - -e^0 = -e^{-A} + 1 = 1 - e^{-A}$$

# PRACTICE PROBLEM

Which of the following are probability functions?

- a)  $f(x) = \frac{1}{4}$  for  $x = 9, 10, 11, 12$
- b)  $f(x) = \frac{3-x}{2}$  for  $x = 1, 2, 3, 4$
- c)  $f(x) = \frac{x^2+x+1}{25}$  for  $x = 0, 1, 2, 3$

# PROBABILITY DISTRIBUTION

Joint Probability distribution

*Probability of  $X = x \text{ & } Y = y$*

$\text{Prob}(X = x, Y = y)$

$p(x, y)$

Conditional probability Distribution

*probability of  $X = x$  given  $Y = y$*

$\text{prob}(X = x|Y = y)$

$p(x|y) = \frac{p(x, y)}{p(y)}$

# EXAMPLE

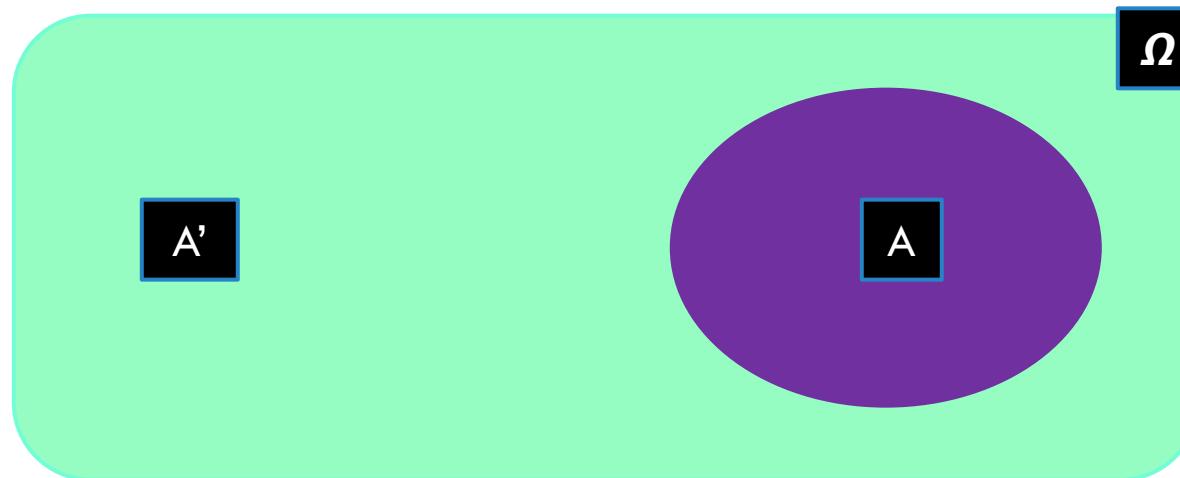
Suppose that we know that a die throw was odd.

What is the probability that a “three” has been thrown.

# COMPLEMENT RULE

Given: event A, which can occur or not

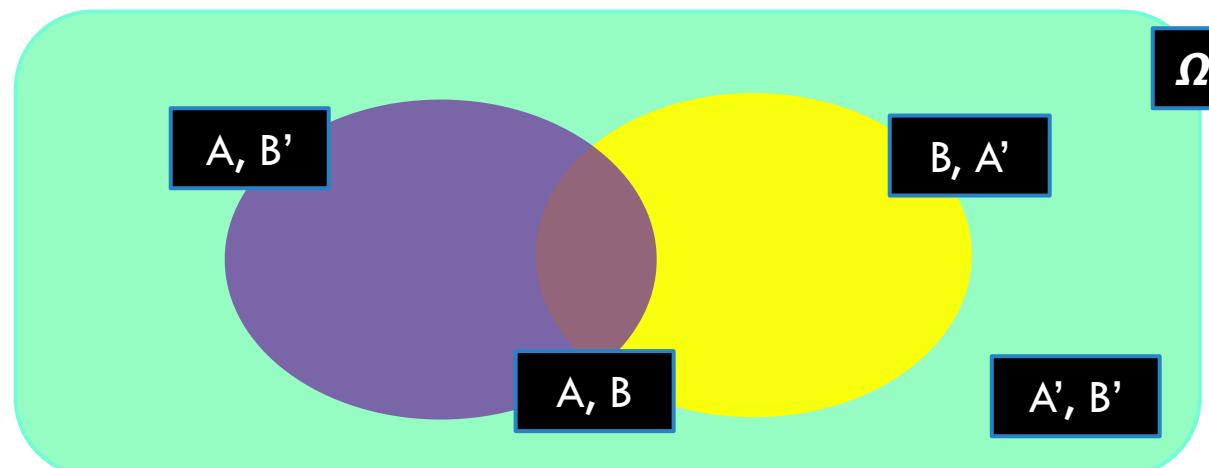
$$P(A') = 1 - p(A)$$



# RULE OF TOTAL PROBABILITY

Given: events A and B, which can co-occur (or not)

$$P(A) = p(A, B) + p(A, B')$$



# OTHER RULES OF PROBABILITY

- Sum Rule (marginalization/summing out):

$$p(x) = \sum_y p(x, y)$$

$$p(x_1) = \sum_{x_2} \dots \sum_{x_N} p(x_1, x_2, \dots, x_N)$$

- Product/Chain Rule

$$p(x, y) = p(y|x)p(x)$$

$$p(x_1, x_2, \dots, x_N) = p(x_1)p(x_2|x_1) \dots p(x_N|x_1, x_2, \dots, x_{N-1})$$

# EXAMPLE OF PRODUCT RULE

- Calculate the probability that a man has gray hair and is over 65 if:
  - Probability of a chosen random man being over 65: 20%
  - Probability of having gray hair given that a randomly chosen man is over 65: 80%

# INDEPENDENCE

- Two random variables are said to be independent if and only if their joint distribution factors

$$\begin{aligned} p(A, B) &= p(A) \cdot p(B) = p(A|B)p(B) = p(B|A)p(A) \\ p(A|B) &= p(A) \\ p(B|A) &= p(B) \end{aligned}$$

- Two random variables are conditionally independent given a third if they are independent after conditioning on the third

$$p(A, B|C) = p(A|C) \cdot p(B|C) = p(A|B, C)p(B|C) = p(B|A, C)p(A|C)$$

# LET'S TEST SOME EVENTS

- Outcomes on multiple rolls of a die
- Outcomes on multiple flips of a coin
- Height of two related individuals
- Probability of getting a king on successive draws from a deck of cards, without replacement
- Probability of getting a king on successive draws from a deck of cards, with replacement

# BAYES' RULE

A way to find conditional probabilities for one variable when conditional probability for another variable is known.

$$p(B|A) = \frac{p(A|B) \cdot p(B)}{p(A)}$$

Where

$$p(A) = p(A, B) + p(A, B')$$

# BAYES' RULE

$$p(B|A) = \frac{p(A|B) \cdot p(B)}{p(A)}$$

The diagram illustrates the components of Bayes' Rule. At the top, there are three colored boxes: a green box labeled "Likelihood", a pink box labeled "prior probability", and a dark grey box labeled "Posterior probability". Below these boxes is the Bayes' Rule formula. The term  $p(A|B)$  is enclosed in a grey oval and has a grey arrow pointing to it from the "Posterior probability" box. The term  $p(B)$  is enclosed in a pink oval and has a pink arrow pointing to it from the "prior probability" box. The term  $p(A)$  is at the bottom right of the formula and has no arrow pointing to it.

# EXAMPLE OF BAYES' RULE

A desk lamp produced by a company was found to be defective ( $D$ ). There are three factories ( $A, B, C$ ) where such desk lamps are manufactured. This is what we know about the company's desk lamp production and the possible source of defects:

Factory	% of total production	Probability of defective lamps
A	$0.35 = p(A)$	$0.015 = p(D   A)$
B	$0.35 = p(B)$	$0.010 = p(D   B)$
C	$0.30 = p(C)$	$0.020 = p(D   C)$

If a randomly selected lamp is defective, what is the probability that the lamp was manufactured in factory C?

# EXPECTATION & VARIANCE

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# EXAMPLE: THE LOTTERY

- The Lottery (also known as a tax on people who are bad at math...)
- A certain lottery works by picking 6 numbers from 1 to 49. It costs \$1.00 to play the lottery, and if you win, you win *\$2 million* after taxes.
- *If you play the lottery once, what are your expected winnings or losses?*

# LOTTERY

Calculate the probability of winning in 1 try:

$$\frac{1}{\binom{49}{6}} = \frac{1}{\frac{49!}{43!6!}} = \frac{1}{13,983,816} = 7.2 \times 10^{-8}$$

“49 choose 6”

Out of 49 numbers, this  
is the number of distinct  
combinations of 6.

The probability function (note, sums to 1.0):

$x$	$p(x)$
-1	0.999999928
+ 2 million	$7.2 \times 10^{-8}$

# EXPECTED VALUE

The probability function (note, sums to 1.0):

$x$	$p(x)$
-1	0.999999928
+ 2 million	$7.2 \times 10^{-8}$

Expected Value:

$$\begin{aligned} E(X) &= p(\text{win}) * \$2,000,000 + p(\text{lose}) * -\$1.00 \\ &= 2.0 \times 10^6 * 7.2 \times 10^{-8} + 0.999999928 (-1) \\ &= 0.144 - 0.999999928 = -\$0.86 \end{aligned}$$

Negative expected value is never good!

You shouldn't play if you expect to lose money!

# EXPECTED VALUE AS A MATHEMATICAL OPERATOR

If  $c$  = a constant number (i.e., not a variable) and  $X$  and  $Y$  are any random variables...

- $E(c) = c$
- $E(cX) = cE(X)$
- $E(c + X) = c + E(X)$
- $E(X + Y) = E(X) + E(Y)$

# EXPECTED VALUE ISN'T EVERYTHING THOUGH...

- Take the show “Deal or No Deal”
- Let's say you are down to two cases left. \$0 and \$400,000. The banker offers you \$200,000.
- So, Deal or No Deal?

# DEAL OR NO DEAL...

- This could really be represented as a probability distribution and a non-random variable:

$x$	$p(x)$
+0	0.50
+\$400,000	0.50

$x$	$p(x)$
+\$200,000	1.0

# EXPECTED VALUE DOESN'T HELP...

- This could really be represented as a probability distribution and a non-random variable:

$x$	$p(x)$
+0	0.50
+\$400,000	0.50

$$\mu = E(X) = \sum_{\text{all } x} x_i p(x_i) = 0 (.50) + 400,000(.50) = 200,000$$

$x$	$p(x)$
+\$200,000	1.0

$$\mu = E(X) = 200,000$$

# HOW TO DECIDE?

- Variance!
- If you take the deal, the variance/standard deviation is 0.
- If you don't take the deal, what is average deviation from the mean?

# VARIANCE

“The average (expected) squared distance (or deviation) from the mean”

$$\sigma^2 = \text{Var}(x) = E[(x - \mu)^2] = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i)$$

\*\*We square because squaring has better properties than absolute value. Take square root to get back linear average distance from the mean (= "standard deviation").

# SUMMARIZING PROBABILITY DISTRIBUTIONS

- It is often useful to give summaries of distributions without defining the whole distribution
- Mean or Expected value:

$$E[x] = \sum x.p[x] \quad \text{or} \quad E(x) = \int_x x.p(x)dx$$

- Variance:

$$\text{var}(x) = \int_x (x - E(x))^2.p(x)dx = E(x^2) - E(x)^2 \quad \text{or} \quad \text{var}[x] = E[x^2] - E[x]^2$$

# COVARIANCE: JOINT PROBABILITY

- The covariance measures the strength of the linear relationship between two variables
- The covariance:

$$E[(x - \mu_x)(y - \mu_y)]$$

$$\sigma_{xy} = \sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y) P(x_i, y_i)$$

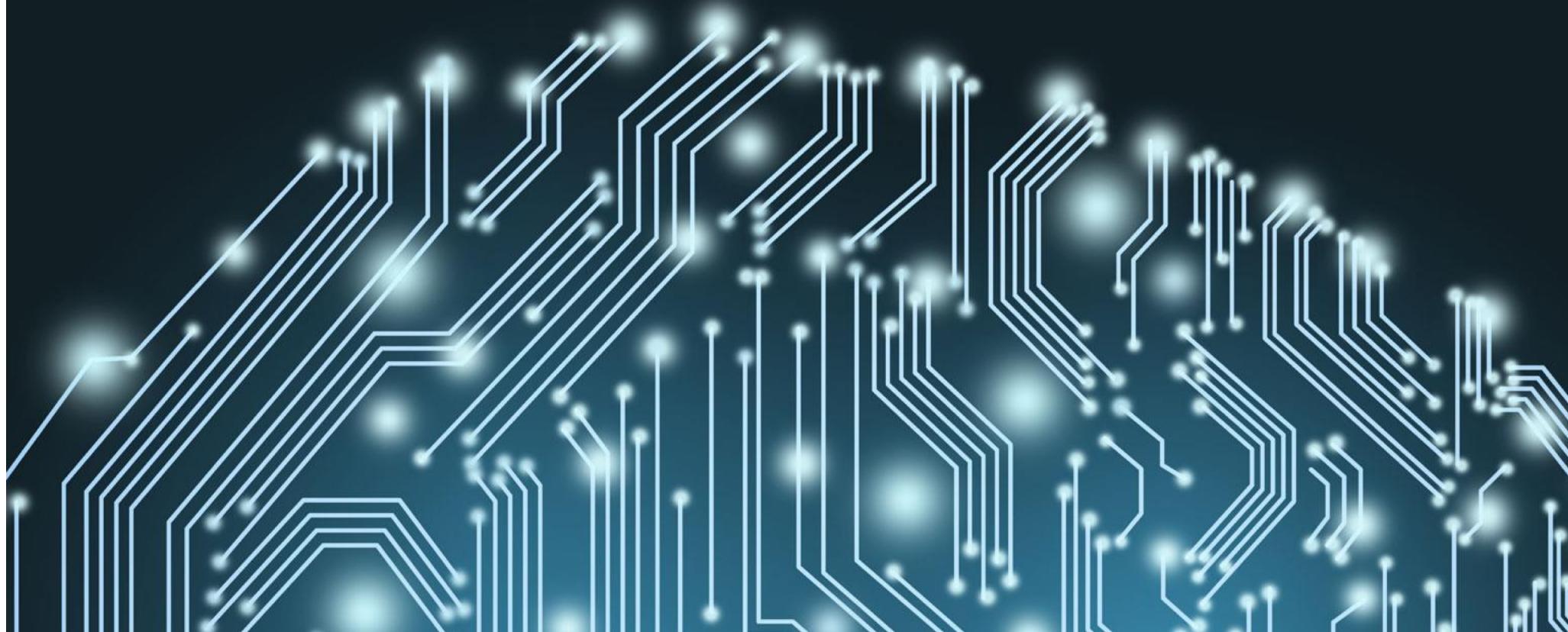
# INTERPRETING COVARIANCE

- **Covariance** between two random variables:

$cov(X, Y) > 0$      $X$  and  $Y$  are positively correlated

$cov(X, Y) < 0$      $X$  and  $Y$  are inversely correlated

$cov(X, Y) = 0$      $X$  and  $Y$  are independent (if relationship is linear)



# QUESTIONS?