

19.09.2022

YZV 231E

Probability Theory & Stats

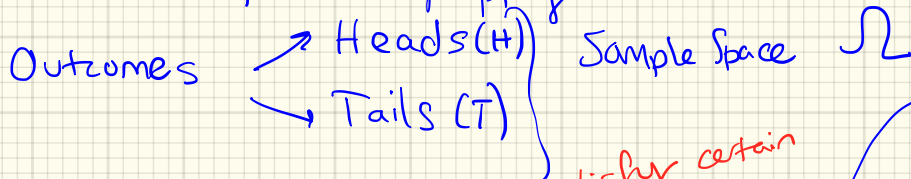
Week 1

Gü.

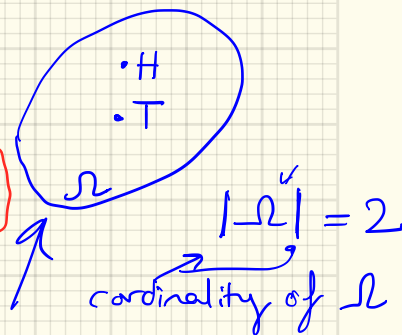
We learn what it takes to set up a probabilistic model

① **Sample Space** : contains all the outcomes of a random experiment

ex: Random experiment: Flipping a coin:



② **Probability Laws** :  $\rightarrow$  should satisfy certain properties  
 $\rightarrow$  **AXIOMS of PROBABILITY**  
describe my beliefs about which outcomes are more likely to occur compared to other outcomes.



③ **Define Events** : sets of outcomes : subsets of the sample space  $\Omega$ .

④ **Calculate probability of Events** :

Sample Space: We execute a particular experiment  $\hookleftarrow$

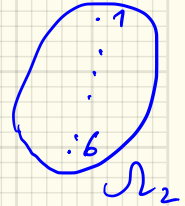
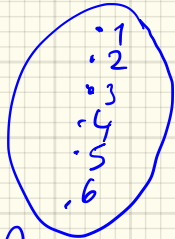
We list all the outcomes

ex: Experiment: rolling a  $\frac{1}{6}$  dice

$A = \{ \text{getting "4" on the dice} \}$

$$P(A) = \frac{1}{6}$$

$$|\Omega| = 6$$



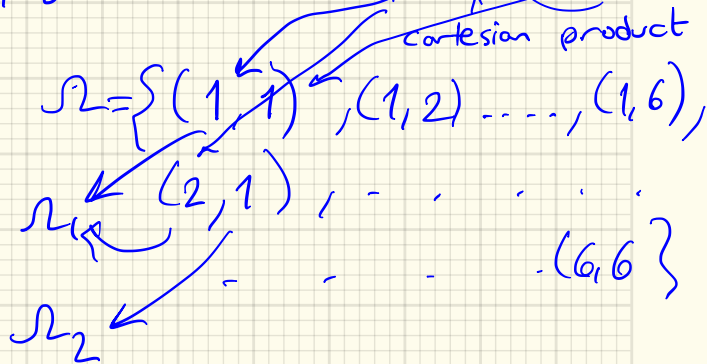
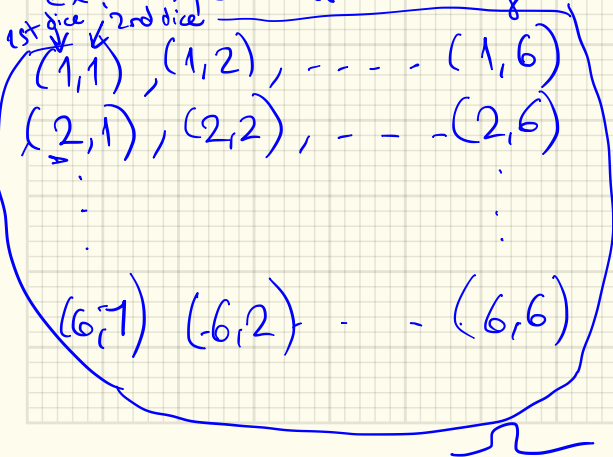
$\Omega$

ex: Experiment: Rolling two die:

$$|\Omega| = 36$$

$$\Omega = \Omega_1 \times \Omega_2$$

Cartesian product

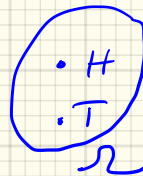
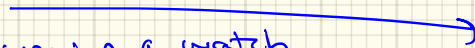
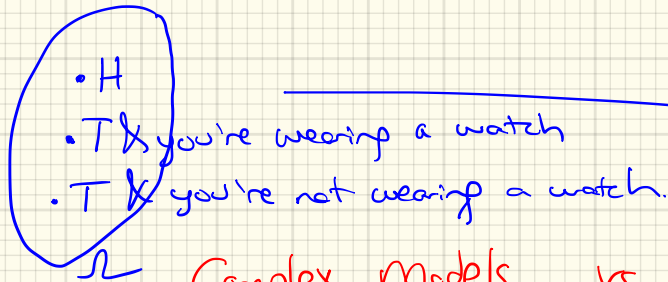


Sample Space (List of all outcomes) should be:

- ① Mutually Exclusive ( $\equiv$  Disjoint) From the experiment we get only one of the outcomes
- ② Collectively Exhaustive: all the outcomes that may happen in the experiment are in  $\Omega$ .

Ex: Flip a coin; you believe wearing a watch affects the outcome:

③ How much "granularity" you need in defining  $\Omega$



Complex Models vs Simplified Models

needed  
granularity

{ Einstein's (Occam's Razor):  
"Everything should be made as simple as possible,  
but no simpler".



→ Picking "right" amount of granularity is an art of engineering

Ex: Rolling the dice twice: Simple experiment, sequentially executed experiment.

Hypothetical dice w/ 4 faces

Single Roll outcomes:  $\{1\}, \{2\}, \{3\}, \{4\}$

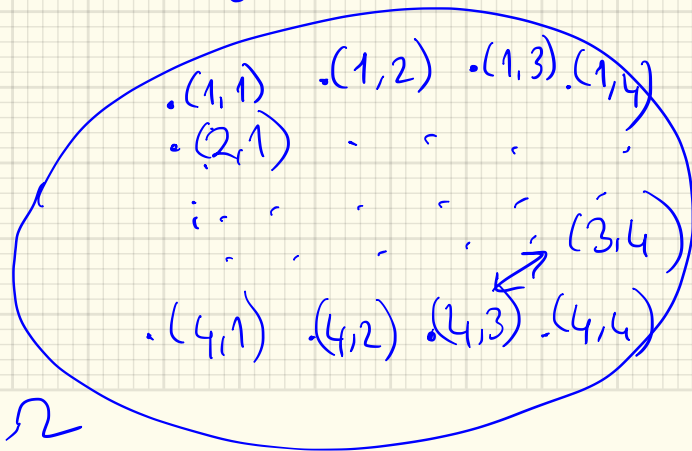
Q. What is the sample space of this experiment?

↑ Roll 2

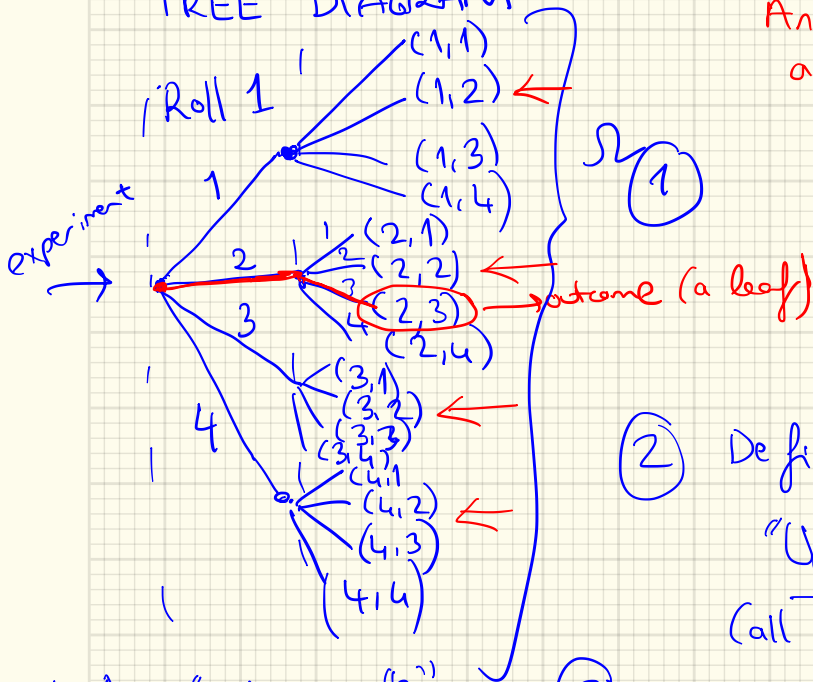
4			(3,4)	
3				(4,3)
2				
1				
	1	2	3	4

← Visualization of the sample space

Roll 1:  $|\Omega| = 16$



For a sequential description of the experiment, we can use a  
**TREE DIAGRAM**



Any path in the tree is associated to a particular outcome & vice versa

# leaves = 16 leaves

$$|\Omega| = 16$$

**Finite** Sample Space

② Define probability law:

"Uniform" probability law:  
 (all outcomes are equally likely)

③ Event  $E = \{(2, 3)\}$

④  $P(E) = ? = \frac{|E|}{|\Omega|} = \frac{1}{16}$

ex:  $A =$  "getting a '2' in the 2nd roll".

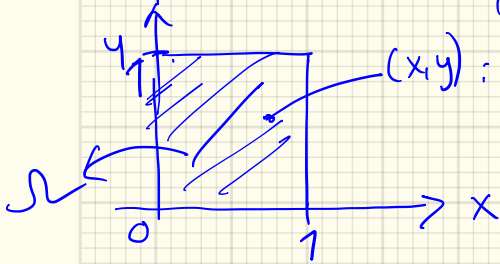
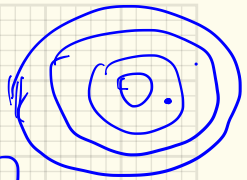
$$|A| = 4$$

$$P(A) = \frac{4}{16} = \frac{1}{4}$$

# Infinite Sample Space (Continuous example)

Random Experiment: Throwing a dart into the square

$$[0,1] \times [0,1]$$



$(x,y)$ : a possible outcome  $(x,y) \in [0,1] \times [0,1]$ .

Q. Which outcome is more likely to occur?

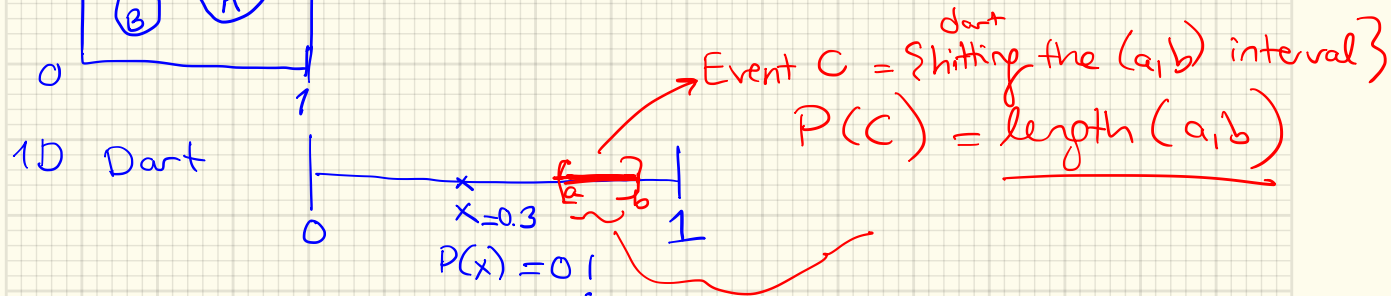
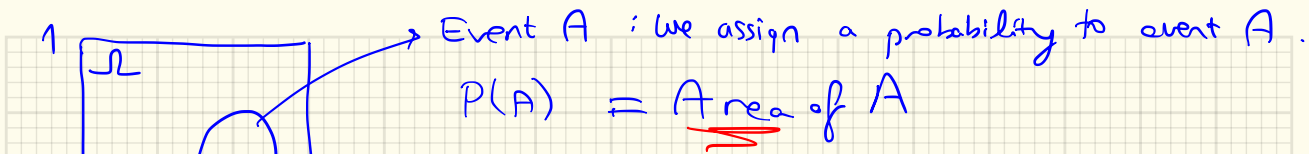
$$P((x,y) = (0.7, 0.6)) = ? \quad 0!$$

Any individual point (outcome) has zero probability  
b/c  $\exists$  only many real numbers  
(there exists)

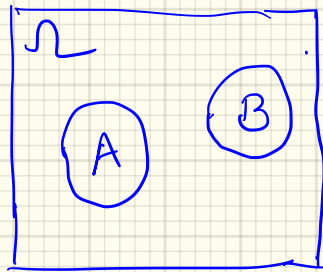
Q  $\rightarrow$  How do we work w/ this?

A: We assign probabilities to SUBSETS of the sample space.

Subsets of  $\Omega$  are called EVENTS



Probabilities  $\propto$  lengths of the intervals in 1D.



$P(A \cup B) =$  Total probability is the sum of individual probabilities of the two events A & B

→ Total mass = Sum of the 2 masses.  
 Analogy

How should we make probability assignments?

→ 1) want probabilities to be between  $[0, 1]$  →

b/c (because) prob = zero → smth. is not going to happen  
event/outcome  
prob = ONE → we're certain that smth. is going to happen.

→ Want to satisfy other rules → all are summarized in  
AXIOMS of PROBABILITY: all legitimate prob. models should obey

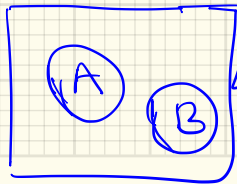
1)  $P(A) \geq 0$  (nonnegativity)

2)  $P(\Omega) = 1$  (normalization)

3) if  $A \cap B = \emptyset$  then  $P(A \cup B) = P(A) + P(B)$

2) is b/c  $\Omega$  is collectively exhaustive

3)



Union event:  $A \cup B$ : event that

A occurred or B occurred,

prob is the sum of their probabilities when they are mutually exclusive.

Refresh SETS :  $A = \{ \omega_1, \omega_2, \dots, \omega_n \}$

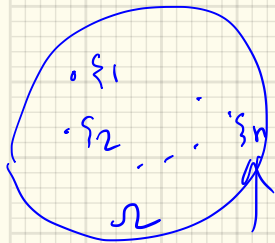
collections of objects  $\omega_i \in A$

$\omega_i \notin A$

← elements of A

-  $\emptyset$ : empty set (impossible event)

- If A is an event  $\rightarrow$  then  $A^c$  (its complement) has to be an event.



$\Omega$ : (sample space) is an event : Certain Event

$\Omega^c = \emptyset$  is an event.

If a set has  $n$  elements,  $\exists 2^n$  subsets including  $\emptyset$  & itself.

Set Operations : - Union :  $A \cup B = \{ x : x \in A \text{ or } x \in B \}$

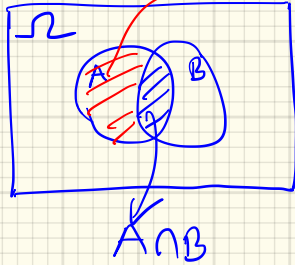
- Intersection :  $A \cap B$

- Complement :  $A^c = \{ x : x \in \Omega \text{ but } x \notin A \}$

- Difference :  $A - B = \{ x : x \in A \text{ and } x \notin B \}$

$$A - B = A \cap B^c$$

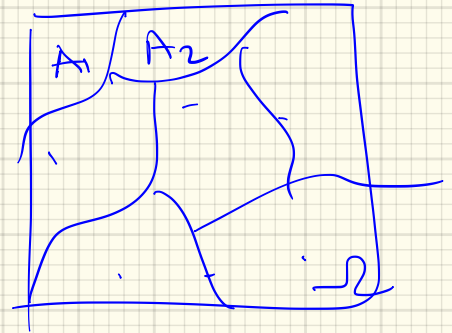
Def: Sets  $A$  &  $B$  are disjoint if  $A \cap B = \emptyset$



\*  $A_i$  are called a PARTITION of  $\Omega$  iff

1)  $A_i$ 's are disjoint

2) 
$$\bigcup_{i=1}^{\infty} A_i = \Omega$$



For any subsets of  $\Omega$

Commutative Law:  $A \cup B = B \cup A$  ;  $A \cap B = B \cap A$

Associative Law:  $A \cup (B \cup C) = (A \cup B) \cup C$

$$A \cap (B \cap C) = (A \cap B) \cap C$$

Distributive Law:  $(A \cup B) \cap C = (A \cap C) \cup (B \cap C)$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$$

De Morgan Law:  $(\bigcup_i A_i)^c = \bigcap_i A_i^c$

$$(\bigcap_i A_i)^c = \bigcup_i A_i^c$$

$A \subseteq B$  :  $A$  is a subset of  $B$  ) <sup>ex:</sup>  $[A \cap (B \cup C)]^c = ?$   
 $\equiv B \supseteq A$  )  $- A^c \cup (B^c \cap C^c)$   
 $\Rightarrow$  IF  $B \subseteq A$  &  $A \subseteq B$  then  $A = B$ .



AXIOMS of PROBABILITY:

- 1)  $P(A) \geq 0$  ←
- 2)  $P(\Omega) = 1$
- 3) If  $A \cap B = \emptyset \rightarrow$   
 $P(A \cup B) = P(A) + P(B)$  ) ←

ex: Rolling two dice (w/ 4 faces)

1) Defined the sample space ✓

② Assign a probability law:

"Discrete" Uniform Probability law (1) Axiom

each outcome has a probability =  $\frac{1}{16}$  (1)  $P(A) \geq 0$

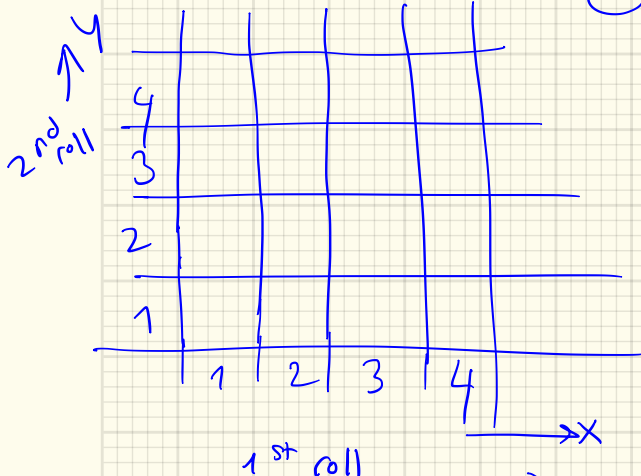
Q. Prob. that the 2<sup>nd</sup> roll gives "1" on the dice?  $A = \bigcup_i A_i$  (2)  $P(\Omega) = 1$

$\rightarrow A_1 \quad A_2 \quad A_3 \quad A_4$  (2) Axiom

(1,1) (2,1) (3,1) (4,1)

↑ "singleton" event

$\bigcap_i A_i = \emptyset$



3<sup>rd</sup> axiom:  $P(A) = P(\bigcup_i A_i) = \sum_i P(A_i) = 4 \cdot \frac{1}{16} = \frac{1}{4}$

# Discrete Uniform Probability Law:

All outcomes are equally likely:

→ Computing probability →

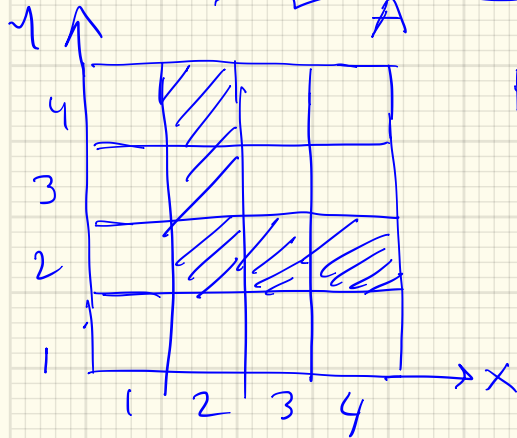
Counting

- Some examples
- fair coins
  - dice
  - well-shuffled decks.

$$P(A) = \frac{|A|}{|\Omega|}$$

Q.  $P(\min(X, Y) = 2) = ?$

A event: minimum of the 2 rolls is 2



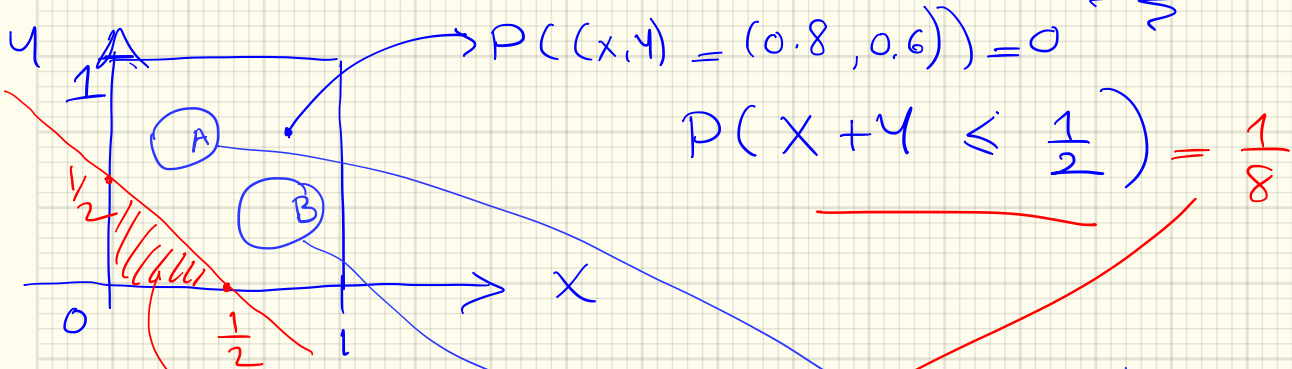
$$P(A) = \frac{5}{16}$$

Q.  $P(X+Y \geq 5)$  calculate. = ?

Q.  $P(X+Y \text{ is even}) = \frac{1}{2}$

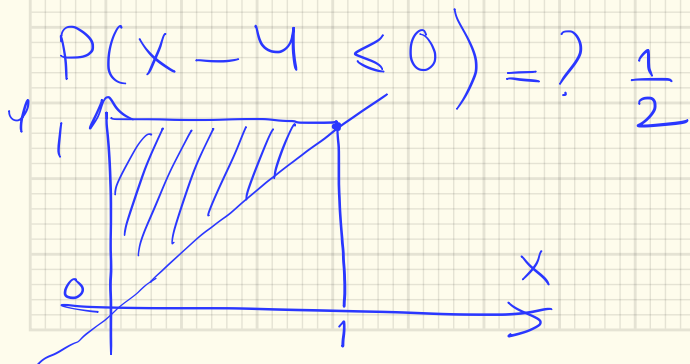
# Continuous Uniform Probability Law :

Recall the dart problem  $\rightarrow$  we measured areas or lengths  $\rightarrow$



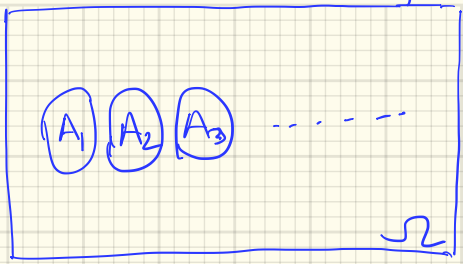
If B has the double  
the area of A

then  
 $P(B) = 2 \times P(A)$



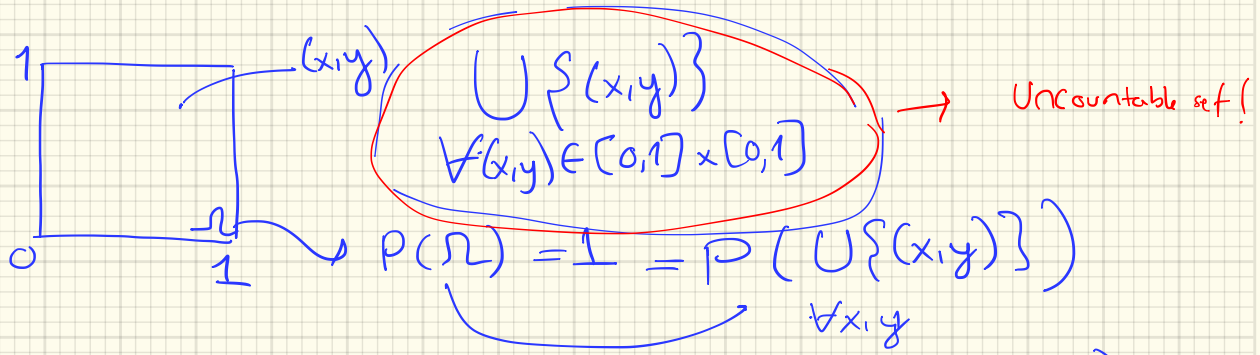
$A_i$  are  $\infty$ -sequence of disjoint events;

due Axiom 3  $\sum_{i=1}^{\infty}$



$$P(A_1 \cup A_2 \cup A_3 \cup \dots) = \sum_{i=1}^{\infty} P(A_i)$$

$$= P(A_1) + P(A_2) + P(A_3) + \dots$$



$$P(\Omega) = 1 = P(\bigcup_{\forall x,y} \{ (x,y) \})$$

$$\underline{1} \neq \sum_{\forall x,y} P(\{ (x,y) \}) = \underline{0}!$$

→ Additivity Axiom → Countable Additive

3<sup>rd</sup> Axiom does not apply to uncountable sets.

$\bigcup_{i=1}^{\infty} A_i$  : countable seq. of events!

$\{A_i\}_{i=1}^{\infty}$  
 $\{1, 2, 3, 4, \dots\}$ 
 
 $\}$ 
 
 $\rightarrow \infty$ 
  
 countable set of events 
 $\{A_1, A_2, A_3, \dots\}$ 
 
 $\}$ 
 
 $\rightarrow \infty$

1-1 correspondence w/ +ve integers  $\mathbb{Z}^+$ .

Summarize: Probabilistic modeling: Random experiment is performed

- 1) Specify Sample Space:
  - $\rightarrow$  mutually exclusive
  - $\rightarrow$  collectively exhaustive
- 2) Define a probability law:  $\rightarrow$  "right" amount of granularity
- 3) Identify an Event of Interest:
- 4) Calculate probability of the event

+ (Kolmogorov's) Axioms of Probability

- 1)  $P(\Omega) = 1$
- 2)  $P(A) \geq 0$
- 3)  $\{A_i\}_{i=1}^{\infty}$  disjoint  $\rightarrow P(\cup_i A_i) = \sum_{i=1}^{\infty} P(A_i)$