

Defn: Sample Space

The sample space  $S$  is the set of possible outcome.

Ex. Flip a coin.

$$S = \{H, T\}$$

Ex. Roll a six-sided die.

$$S = \{1, 2, 3, 4, 5, 6\}$$

Ex. Roll two dice.

$$\begin{aligned} S &= \{(1,1), (1,2), (1,3), \dots \\ &\quad (2,1), (2,2), \dots\} \\ &= \{(i,j) \mid 1 \leq i, j \leq 6\} \end{aligned}$$

Ex. Waiting for a bus to arrive.

$$S = [0, \infty)$$

Ex. Look at number of customers arriving in my restaurant.

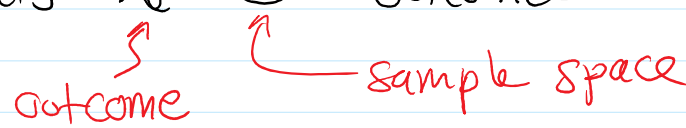
$$S = \mathbb{N}_0 = \{0, 1, 2, 3, 4, \dots\}$$

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Types of sample spaces:

- ① finite ( $|S| < \infty$ )
  - ② infinite ( $|S| \geq \infty$ )
    - ↳ (i) countable (e.g.  $\mathbb{N}_0$ )
    - ↳ (ii) uncountable (e.g.  $[0, \infty)$ )
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Defn: Outcome:

We call elements  $s \in S$  "outcomes"  


Ex.  $S = \{1, 2, 3, 4, 5, 6\}$

then  $1 \in S$  so "1" is an outcome.

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Defn: Event

An event  $E$  is a subset of the sample space  $S$ .

$$(E \subset S)$$

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ex. Roll a die :  $S = \{1, 2, 3, 4, 5, 6\}$  and

so  $E = \{1, 2\} \subset S$

↑ rolling a 1 or a 2.

→ Say an event "happens" if the observed outcome of the experiment is an element of  $E$ .

ex.  $S \subset S$ , here  $S$  is an event.

↑ event that something happens/occurs

ex.  $\emptyset \subset S$ , so  $\emptyset$  is an event.

↑ ??? nothing happens?

## Axiomatic Probability

Given: an experiment (a sample space  $S$ )

want: for any event ( $E \subset S$ ) want to assign some measure of the likelihood that  $E$  occurs.

↑  
probability

prob. fn.

Mathematically:

For each  $E \subset S$  assign a probability  $P(E)$

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prob. of  $E$  occurring.

What are the rules for building  $P$ ?

- ① mathematically consistent
- ② preserve/encode (some) of our intuitions about probability.

Defn: Probability Function  $P$

Given a sample space  $S$  a prob. function  $P$  is a function

$$\rightarrow P: \underbrace{P(S)}_{\text{domain}} \rightarrow \underbrace{\mathbb{R}}_{\text{co-domain}}$$

that satisfies Kolmogorov Axioms

$$\rightarrow \textcircled{1} \text{ non-negativity } P(E) \geq 0 \quad \forall E \subset S$$

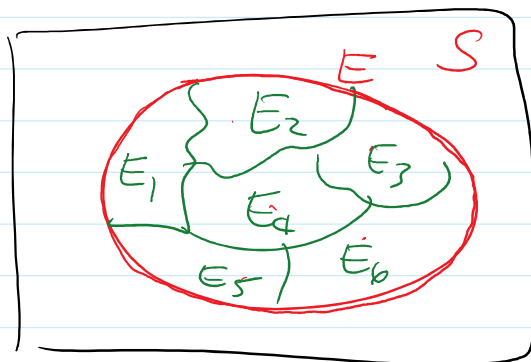
$$\rightarrow \textcircled{2} \text{ unit-measure } P(S) = 1.$$

### (3) Countable additivity

If  $(E_i)_{i=1}^{\infty}$  is a partition of  $E$

$$\left( \begin{array}{l} E_i E_j = \emptyset \quad \forall i \neq j \\ \bigcup_{i=1}^{\infty} E_i = E \end{array} \right)$$

$$P(E) = \sum_{i=1}^{\infty} P(E_i).$$



Countable additivity gives us distributive law

$$\begin{aligned} P(E) &= P\left(\bigcup_{i=1}^{\infty} E_i\right) \\ &= \sum_{i=1}^{\infty} P(E_i) \end{aligned}$$

↓  
for disjoint  
events.

Area of  $E$   
=  
sum of areas of  $E_i$

Ex. Flip a coin.

$$S = \{H, T\}.$$

What is a valid prob. fn on  $S$ ?

$$\begin{array}{ll} P(\{H\}) = \underline{\underline{1/2}} & P(\{H, T\}) = \underline{\underline{1}} \\ P(\{T\}) = \underline{\underline{1/2}} & P(\underline{\underline{\emptyset}}) = \underline{\underline{0}} \end{array}$$

check if this is a valid prob. fn.

✓ ①  $P(E) \geq 0 \quad \forall E \in \mathcal{S}$

✓ ②  $P(S) = 1$

③  $P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i)$  for any partition  $(E_i)_{i=1}^{\infty}$  of any event  $E$

$E_1 = \{H\}$   
 $E_2 = \{T\}$  ) disjoint

$S$  is partitioned into  $E_1 \cup E_2$

$$1 = P(S) = P(E_1 \cup E_2) = \underbrace{P(E_1)}_{1/2} + \underbrace{P(E_2)}_{1/2}$$

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Ex. Other ways of defining  $P$  on  $S$ ?

$$P(S) = 1 \quad P(\{H\}) = \alpha \quad \alpha \in [0, 1]$$

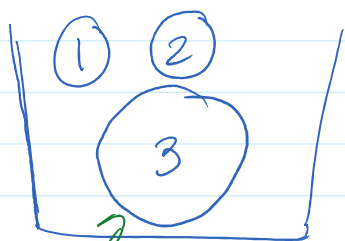
$$P(\emptyset) = 0 \quad P(\{T\}) = 1 - \alpha$$

Valid? Yes. Why?

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Ex.

|| ① ② /  $S = \{1, 2, 3\}$



2x as likely to choose ball 3

$$S = \{1, 2, 3\}$$

$$P_1 = \frac{1}{4} \quad P_2 = \frac{1}{4} \quad P_3 = \frac{1}{2}$$

$$P(\{1, 2\}) = \frac{1}{4} + \frac{1}{4} = P_1 + P_2$$

$$P(\{1, 3\}) = P_1 + P_3 = \frac{1}{4} + \frac{1}{2} = \frac{3}{4}$$

## Theorem: Finite Sample Space Theorem

If  $S = \{s_1, s_2, s_3, \dots, s_n\}$ ,  $|S| = n < \infty$   
to build up IP choose correspondingly

$$P_1, P_2, P_3, \dots, P_n$$

where

$$\textcircled{1} P_i \geq 0 \quad \text{and} \quad \textcircled{2} \sum_{i=1}^n P_i = 1$$

then for ECS,

$$P(E) = \sum_{i: s_i \in E} P_i$$

← valid!

EX.  $P(\{s_1, s_5\}) = P_1 + P_5$

$P(\{s_2, s_7, s_{11}\}) = P_2 + P_7 + P_{11}$

E

Then  $P$  is a valid prob. fn.

Pf. Check that  $P$  satisfies the Kolmogorov Axioms.

✓ ①  $P(E) \geq 0 \quad \forall E \in \mathcal{S}$

$$P(E) = \sum_i \hat{p}_i \geq 0$$

②  $P(S) = 1$

✓  $P(S) = \sum_{i: s_i \in S} \hat{p}_i = \sum_{i=1}^n \hat{p}_i = 1$

③ If  $E \in \mathcal{S}$  and  $(E_i)$  partition  $E$

$(E_i)$  disjoint and  $\bigcup_i E_i = E$

then  $P(E) = P(\bigcup_i E_i) = \sum_i P(E_i)$

WLOG assume  $(E_i)$  is countably infinite.

Then  $E = \bigcup_{i=1}^{\infty} E_i$

So if  $s_j \in S$  and  $s_j \in E$  then

$$s_j \in \bigcup_{i=1}^{\infty} E_i \Rightarrow s_j \text{ is in}$$



$$\Rightarrow \boxed{a_j \in \bigcup_{i=1}^{\infty} E_i} \Leftrightarrow a_j \text{ is in at least one } E_i \text{ exactly}$$

$$P(E) = P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i)$$

Defn:

$$P(E) = \sum_{i: a_i \in E} p_i$$

$$= \sum_{j: a_j \in \bigcup_{i=1}^{\infty} E_i} p_j$$

$$= \sum_{i=1}^{\infty} \sum_{j: a_j \in E_i} p_j$$

$P(E_i)$

Ex.

$$P_1 + P_3 + P_5 + P_6 + P_7 + P_{11}$$

$\uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow$   
 $a_1 \quad a_3 \quad a_5 \quad a_6 \quad a_7 \quad a_{11}$   
 $\uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow$   
 $P(E_1) + P(E_2) + P(E_3)$   
 $P_1 + P_5 + P_3 + P_6 + P_7 + P_{11}$

$$= \sum_{i=1}^{\infty} P(E_i)$$