

Probability Spaces

Sample Space Ω \Rightarrow set of all **outcomes** (mutually

exclusive) of **random experiment**

Event $E \subseteq \Omega \Rightarrow$ any subset of sample space

Extreme Events \Rightarrow null event \emptyset & universal event \mathcal{S}

Elementary Event \Rightarrow singleton subsets of \mathcal{S}

If $s \in E$ is **experiment outcome**, then E **occurred**

Null event \emptyset **never** & **universal event** \mathcal{S} **always** - occurs

For events E_1, E_2, \dots

$E_1 \text{ and } E_2 \text{ and } \dots \Rightarrow \bigcup_i E_i$

$E_1 \text{ and } E_2 \text{ and } \dots \Rightarrow \bigcap_i E_i$

E_1, E_2, \dots are **Mutually Exclusive** $\Rightarrow \forall i, j, E_i \cap E_j = \emptyset$ i.e.

they're **pairwise-disjoint**

E_1, E_2, \dots are **Independent** \Rightarrow

$P(\bigcap_{i=1}^n E_i) = \prod_{i=1}^n P(E_i)$ for any finite subset

$\{E_1, E_2, \dots, E_n\}$

If events A, B are **independent**, then \bar{A}, \bar{B} are **also**

independent

σ -algebra $\mathcal{F} \subseteq \mathcal{P}(\Omega) \Rightarrow$ family of subsets of Ω s.t.

nonempty: $\emptyset \in \mathcal{F}$

closed under **complements**: $E \in \mathcal{F} \Rightarrow \bar{E} \in \mathcal{F}$

closed under **countable union**:

$E_1, E_2, \dots \in \mathcal{F} \Rightarrow \bigcup_i E_i \in \mathcal{F}$

Immediate Basic Results:

$\emptyset \in \mathcal{F}$

closed under **countable intersection**:

$E_1, E_2, \dots \in \mathcal{F} \Rightarrow \bigcap_i E_i \in \mathcal{F}$

$\{\emptyset, \Omega\}$ is smallest & $\mathcal{P}(\Omega)$ is largest **σ -algebras**

Generated σ -algebras:

$\sigma(\mathcal{G}) \Rightarrow$ for family of subsets $\mathcal{G} \subseteq \mathcal{P}(\Omega)$, its **smallest**

σ -algebra to contain \mathcal{G} (exists & unique)

$\sigma(f) \Rightarrow$ for $f: \Omega \rightarrow E$ where (E, \mathcal{E}) is **measurable space**,

$\sigma(f) = \{f^{-1}(F) | F \in \mathcal{E}\}$ i.e. all **pre-images**

trace σ -algebra of $B \in \mathcal{F} \Rightarrow \mathcal{F}_B = \{B \cap A | A \in \mathcal{F}\}$

Probability Measure $P: \mathcal{F} \rightarrow [0, 1]$ on (Ω, \mathcal{F})

$\forall E \in \mathcal{F}, 0 \leq P(E) \leq 1$ i.e. between 0 and 1

$P(\Omega) = 1$ i.e. **universal event** \mathcal{S} **always** occurs

σ -additive (**countably additive**) $\Rightarrow P(\bigcup_i E_i) = \sum_i P(E_i)$

for **pairwise-disjoint** events $E_1, E_2, \dots \in \mathcal{F}$

Immediate Basic Results:

$P(\bar{E}) = 1 - P(E)$

$P(\emptyset) = 0$

$P(A \cup B) = P(A) + P(B) - P(A \cap B)$

Measurable Space $(\Omega, \mathcal{F}) \Rightarrow$ **sample space** Ω with

σ -algebra \mathcal{F} on it

Probability Space $(\Omega, \mathcal{F}, P) \Rightarrow$ **measurable space**

(Ω, \mathcal{F}) with **probability measure** P on it

Conditional Probability

Conditional Probability $\Rightarrow P(A | B) = \frac{P(A \cap B)}{P(B)}$ where

$A, B \subseteq \Omega$ and $P(B) \neq 0$

The **conditional probability space** is $(B, \mathcal{F}_B, P(\cdot | B))$

Sample space $B \subseteq \Omega$

Trace σ -algebra $\mathcal{F}_B = \{B \cap A | A \in \mathcal{F}\}$

Probability measure $P(\cdot | B)$

if A, B are **independent** then $P(A | B) = P(A)$

A_1, A_2 are **Conditionally Independent** given B iff

$P(A_1 \cap A_2 | B) = P(A_1 | B)P(A_2 | B)$

Law Of Total Probability \Rightarrow for any events $\{B_1, B_2, \dots\}$

which **partition** Ω , $P(A) = \sum_i P(A | B_i)P(B_i) = \sum_i P(A \cap B_i)$

Special Case \Rightarrow

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

Bayes Theorem $\Rightarrow P(A | B) = \frac{P(B | A)P(A)}{P(B)}$

General Random Variables

Random Variable \Rightarrow **measurable function** $X: \Omega \rightarrow E$

(Ω, \mathcal{F}, P) is a **probability space**, (E, \mathcal{E}) is a **measurable**

space

For every $B \in \mathcal{E}$ the pre-image of B under X is in \mathcal{F}

i.e. $X^{-1}(B) = \{\omega \in \Omega | X(\omega) \in B\} \in \mathcal{F}$

i.e. $\sigma(X) \subseteq \mathcal{F}$ where $\sigma(X)$ is generated by **function** X

Induced Probability $P_X(X \in B) \Rightarrow$ probability that X

takes on value in $B \in \mathcal{E}$

$P_X(X \in B) = P(X^{-1}(B)) = P(\{\omega \in \Omega | X(\omega) \in B\})$

Also called **Pushforward Measure** of P onto (E, \mathcal{E})

induced by $X \Rightarrow (E, \mathcal{E}, P_X)$ is a **probability space**

Also called the **Probability Distribution** of X

Real Random Variables

Real Random Variable is \mathbb{R} whose co-domain is $E = \mathbb{R}$

Support $\text{supp}(X) \Rightarrow$ is range of X i.e. $\text{supp}(X) = X(\Omega)$

Simple RRV \Rightarrow **finite** $\text{supp}(X)$

Discrete RRV \Rightarrow **countable** $\text{supp}(X)$

Continuous RRV \Rightarrow **uncountable** $\text{supp}(X)$

Induced Probability $\Rightarrow P_X(X \leq x) = P(\{\omega \in \Omega | X(\omega) \leq x\})$

Cumulative Distribution Func. (CDF) $F_X(x) = P_X(X \leq x)$

$F_X(x)$ is **right-continuous** \Rightarrow for any decreasing $\{x_n\}$

$\lim_{n \rightarrow \infty} x_n = x_L \Rightarrow \lim_{n \rightarrow \infty} F_X(x_n) = F_X(x_L)$

To check that function is valid CDF, must obey:

Monotonicity $\forall x_1, x_2 \in \mathbb{R}, x_1 < x_2 \Rightarrow F_X(x_1) \leq F_X(x_2)$

$F_X(-\infty) = 0, F_X(\infty) = 1$

F_X is right-continuous

Simple Properties

$0 \leq F_X(x) \leq 1, \forall x \in \mathbb{R}$

$P_X(a < X \leq b) = F_X(b) - F_X(a)$ for finite intervals $(a, b] \subseteq \mathbb{R}$