

# PROBABILITY SPACE

AN IMPORTANT AND USELESS CONCEPT

COIN FLIP:

Sample space:  $\Omega = \{H, T\}$

Event space:

$$\mathcal{F} = \{ \{H, T\}, \{H\}, \{T\}, \emptyset \}$$

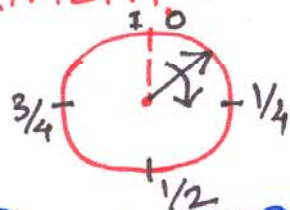
Probability Measure:

$$P_{\mathcal{F}}: \mathcal{F} \rightarrow [0, 1]$$

$$\left[ \begin{array}{l} P_{\mathcal{F}}(\{H\}) \stackrel{\Delta}{=} \frac{1}{2}, P_{\mathcal{F}}(\{T\}) \stackrel{\Delta}{=} \frac{1}{2} \\ P_{\mathcal{F}}(\{H, T\}) \stackrel{\Delta}{=} 1, P_{\mathcal{F}}(\emptyset) \stackrel{\Delta}{=} 0 \end{array} \right]$$

TYCHE'S EXPERIMENT:

$$\Omega = (0, 1]$$



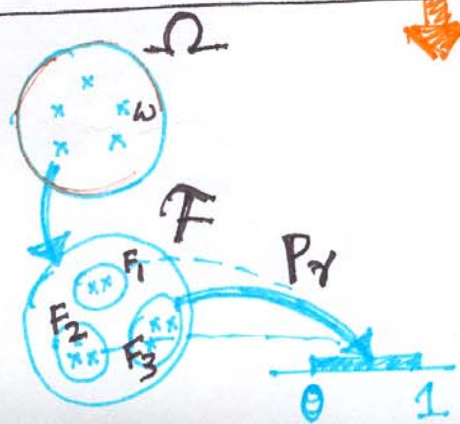
$$\mathcal{F} = \{ \text{all } (a, b) \}, \text{ where } a, b \in \Omega$$

$$P_{\mathcal{F}}^2: \mathcal{F} \rightarrow [0, 1]$$

$$\left[ P_{\mathcal{F}}^2(\{(a, b)\}) \stackrel{\Delta}{=} \int_a^b f(\sigma) d\sigma \right]$$

where  $f(\sigma) = 1, \forall \sigma \in [0, 1]$

(Borel)



Axioms (EVENT SPACE) | (Probability Measure)

- |   |  |
|---|--|
| <p>① if <math>F \in \mathcal{F} \Rightarrow F^c \in \mathcal{F}</math></p> <p>② <math>F_1 \in \mathcal{F}, F_2 \in \mathcal{F} \Rightarrow F_1 \cup F_2 \in \mathcal{F}</math></p> <p>③ <math>\bigcup_{i=1}^{\infty} F_i \in \mathcal{F}</math></p> | <p>① <math>P_{\mathcal{F}}(F) \geq 0</math></p> <p>② <math>P_{\mathcal{F}}(\Omega) = 1</math></p> <p>③ <math>P_{\mathcal{F}}\left(\bigcup_{i=1}^{\infty} F_i\right) = \sum_{i=1}^{\infty} P(F_i)</math><br/>if <math>\bigcap_{i=1}^{\infty} F_i = \emptyset</math></p> |
|---|--|

# INDEPENDENT EVENTS & CONDITIONAL PR.

↳  $F, G \in \mathcal{F}$  are independent events:  $\Pr(F \cap G) = \Pr(F) \Pr(G)$

Different from mutually-exclusive

$$F \cap G = \emptyset \Rightarrow \Pr(F \cup G) = \Pr(F) + \Pr(G)$$

~~$$\Pr(F \cap G)$$~~

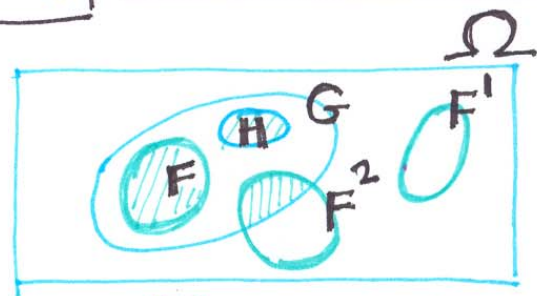
$$\Pr\left(\bigcap_{i=1}^n F_i\right) = \prod_{i=1}^n \Pr(F_i)$$

Conditional Probability  
(A different measure)

$$\Pr(F/G)$$

"Probability that event F occurs given that the event G occurred"

$$\Pr(F/G) = \frac{\Pr(F \cap G)}{\Pr(G)}$$



\*  $\Pr(F'/G) \stackrel{\Delta}{=} 0 \Rightarrow \Pr(G^c/G) = 0$

$$\Pr(F/G) = \Pr(F \cap (G \cup G^c) / G)$$

$$= \Pr(F \cap G / G) + \Pr(F \cap G^c / G)$$

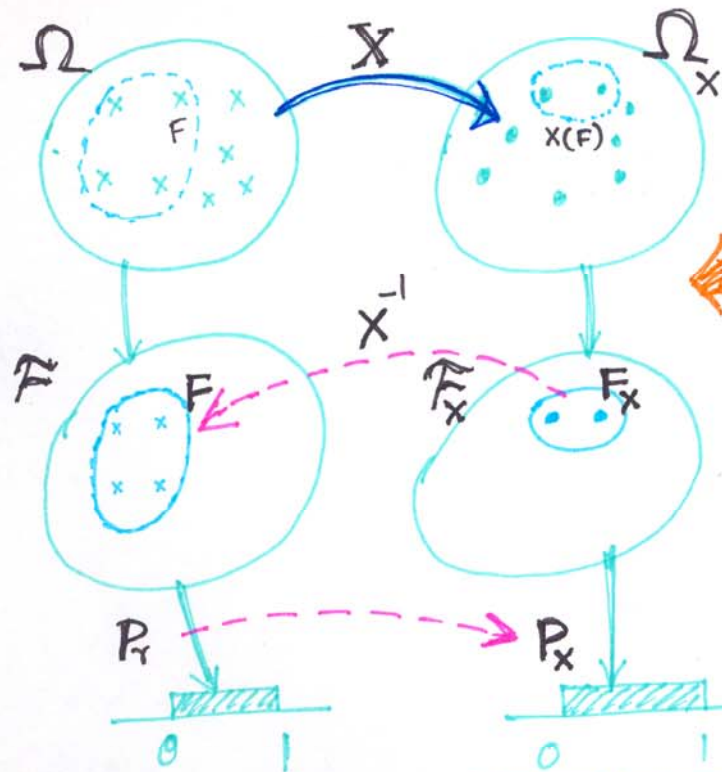
\*  $\frac{\Pr(F \cap G)}{\Pr(\Omega \cap G)} \stackrel{\Delta}{=} \frac{\Pr(F \cap G / G)}{\Pr(\Omega \cap G / G)}$

$$\frac{\Pr(F \cap G)}{\Pr(G)} = \Pr(F \cap G / G)$$



# RANDOM VARIABLES & DISTRIBUTION

its probability.



A random variable is a function  $X: \Omega \rightarrow \Omega_X$ , such that  $X^{-1}(F_X) \in \mathcal{K}$ . i.e. every event is "preserved".

DERIVED DISTRIBUTION

$$P_X(F_X) \triangleq P_\Omega(X^{-1}(F_X)) = P_\Omega(\{\omega: X(\omega) \in F_X\})$$

Discrete r.v.      Continuous r.v.

Cumulative Distribution Function

$$F_X(\alpha) = P_X(\{x: x \leq \alpha\}) \leq 1$$

$$P_X(x_i) = P_X(\{X = x_i\})$$

$\{\omega: X(\omega) = x\}$

Probability mass fn

$$(1) \sum_{x \in \Omega_X} P_X(x) = 1$$

$$(2) P_X(x) \geq 0$$

$$f_X(x) = \lim_{\delta \rightarrow 0} \frac{1}{\delta} P_X(\{x \in (x-\delta, x+\delta)\})$$

Probability density fn

$$(1) \int_{\Omega_X} f_X(x) = 1$$

$$(2) f_X(x) \geq 0$$

# EXAMPLE

Discrete

## BINARY

$$\Omega_x = \{0, 1\},$$

$$p(0) = p, \quad p(1) = 1 - p$$

## UNIFORM

$$\Omega_x = \{0, 1, \dots, m-1\}$$

$$p(k) = 1/m, \quad k \in \Omega_x$$

## BINOMIAL

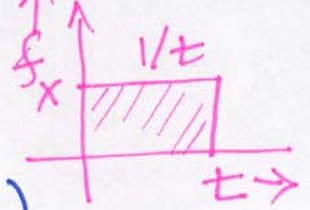
$$\Omega_x = \{0, 1, 2, \dots, n\}, \quad X = Y + Z$$

$$p(k) = \binom{n}{k} p^k (1-p)^{n-k}$$

Binary

$$p + (1-p) = 1 = \begin{cases} 0+1 \\ 1+0 \end{cases}$$

Continuous



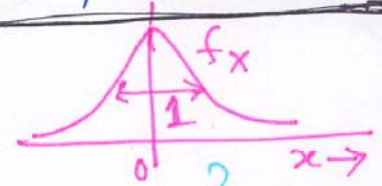
## UNIFORM

$$\Omega_x = (0, t)$$

$$f_x(x) = \begin{cases} 1/t, & 0 \leq x \leq t \\ 0, & \text{otherwise} \end{cases}$$

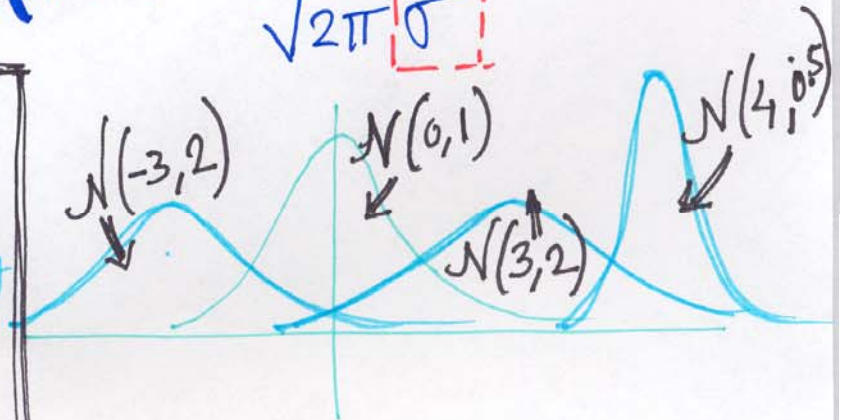
## Gaussian

$$\Omega_x = \mathbb{R}$$



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

$$N(m, \sigma^2) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-m)^2}{2\sigma^2}}$$



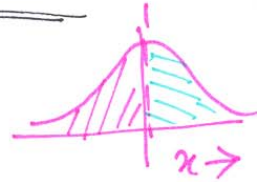


# EXPE<sup>TA</sup>CTION<sup>A</sup> IS GOOD....

$$E_x[g(x)] = \begin{cases} \sum_{x \in \Omega_x} g(x) \cdot p_x(x) \\ \int_{\Omega_x} g(x) \cdot f_x(x) dx \end{cases}$$

Mean ( $m$ )

$$E_x(x) = \begin{cases} \sum x p_x(x) \\ \int x f_x(x) dx \end{cases}$$



Variance ( $\sigma^2$ )

$$E_x[(x-m)^2] = \int_{\Omega_x} (x-m)^2 f_x(x) dx$$

$$F_x(x) \stackrel{\text{or}}{\sim} M_x(j\omega) = 1 + j\omega E_x(x) + \frac{j^2 \omega^2}{2} E_x(x^2) + \dots \downarrow \vdots$$

Gaussian.      ← Higher "moments"

**WEAK**

1.  $\Pr(x \geq 0) = 1 \Rightarrow E_x \geq 0$
2.  $\Pr(x = c) = 1 \Rightarrow E_x = c$
3.  $E(ax + bY) = aE_x + bE_y$
4.  $E\left(\lim_{n \rightarrow \infty} x_n\right) = \lim_{n \rightarrow \infty} E_x x_n$
5.  $E(xy) = E(x) \cdot E(y)$   
(uncorrelated)

**STRONG**

1.  $\Pr(F) \geq 0 \quad \forall F$
2.  $\Pr(\Omega) = 1$
3.  $P(A \cup B) = P(A) + P(B)$
4.  $P\left(\lim_{n \rightarrow \infty} F_n\right) = \lim_{n \rightarrow \infty} P(F_n)$
5.  $P(A \cap B) = P(A) \cdot P(B)$

# TWO RESULTS

## A WEAK LAW OF LARGE NUMBER

$x_0 x_1 x_2 x_3 \dots$

Uncorrelated

Each  $m, \sigma^2$

$$\bar{X}_K = \frac{1}{n} \sum_{n=1}^K x_n$$

As  $K \rightarrow \infty$

$$\begin{aligned} E(\bar{X}_K) &\rightarrow m \\ \sigma_{\bar{X}_K}^2 &\rightarrow 0 \end{aligned}$$

{ statistics! }

## A (STRONG) CENTRAL LIMIT THEOREM

$x_0 x_1 x_2 x_3 \dots$

Independent

Each  $m, \sigma^2$

$$\tilde{X}_K = \frac{1}{\sqrt{n}} \sum_{n=1}^K (x_n - m)$$

As  $K \rightarrow \infty$

$$\begin{aligned} E(\tilde{X}_K) &\rightarrow m \\ F_{\tilde{X}_K} &\rightarrow \mathcal{N}(m, \sigma^2) \end{aligned}$$

{ Thermal Noise  
Brownian Motion }