

Lecture Notes 2: Random Variables

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1 Random Variables

Definition 1.1. A random variable X is a measure map $X : \Omega \rightarrow \mathbb{R}$ that assigns a real number $X(\omega)$ to each and come as "measurable" means that for every X , $\{\omega : X(\omega) \leq x\} \in \mathcal{A}$.

Example 1.1. Flip a coin ten times. Let $X(\omega)$ be a number of heads in the sequence ω .

Example 1.2. Let $\Omega = \{(x, y) | x^2 + y^2 \leq 1\}$. Consider drawing a point at random from Ω .

Definition 1.2. Let $A \subset \mathbb{R}$, $X^{-1} = \{\omega \in \Omega; X(\omega) \in A\} \in \mathcal{A}$. $P(X \in A) \triangleq P(X^{-1}(A)) = P(\{\omega \in \Omega | X(\omega) \in A\})$. $P(X = x) = P(X^{-1}(x)) = P(\{\omega \in \Omega | X(\omega) = x\})$

Example 1.3. Flip a coin twice and let X be the number of heads.

ω	$P(\{\omega\})$	$X(\omega)$
TT	$1/4$	0
TH	$1/4$	1
HT	$1/4$	1
HH	$1/4$	2

X	$P(X)$
0	$1/4$
1	$1/2$
2	$1/4$

1.1 Distribution Function

Cumulative distribution function (or distribution function). CDF is the function $F_X : \mathbb{R} \rightarrow [0, 1]$.

$$F_X(x) = P(X \leq x).$$

Example 1.4. From example 1.3, we can get

$$F_X(x) = \begin{cases} 0 & x < 0 \\ 1/4 & 0 \leq x < 1 \\ 3/4 & 1 \leq x < 2 \\ 1 & x \geq 2 \end{cases}$$

Theorem 1.1. Let X have CDF F , Y have CDF G . If $F(x) = G(x)$ for all x , then $P(X \in A) = P(Y \in A)$ for all measurable A .

Theorem 1.2. A function F mapping $\mathbb{R} \rightarrow [0, 1]$ is a CDF for probability iff

1. F is non-decreasing, $x_1 < x_2 \implies F(x_1) \leq F(x_2)$.

2. F is normalized, i.e. $\lim_{x \rightarrow -\infty} F(x) = 0$, $\lim_{x \rightarrow +\infty} F(x) = 1$.

3. F is right-continuous. $F(x) = F(x^+)$, where $F(x^+) = \lim_{y \rightarrow x, y > x} F(y)$.

Now we will get the proof of right-continuous.

Proof: Let $F(x_1) = P(X \leq x_1)$, $F(x_2) = P(X \leq x_2)$.

Let $X \in \mathbb{R}$, $y_1 > y_2 > \dots$, and $\lim_{n \rightarrow +\infty} y_n = x$.

Let $A_i = (-\infty, y_i]$ and $A = (-\infty, x]$.

Note that $A = \bigcap_{i=1}^{\infty} A_i$ and $A_1 \supset A_2 \supset \dots$

$\lim_{i \rightarrow \infty} P(A_i) = P(\bigcap_{i=1}^{\infty} A_i)$.

$F(x) = P(A) = P(\bigcap_{i=1}^{\infty} A_i) = \lim_{i \rightarrow \infty} P(A_i) = \lim_{i \rightarrow \infty} F(y_i) = F(x^+)$.