

# MATH 180A: INTRO TO PROBABILITY (FOR DATA SCIENCE)

[www.math.ucsd.edu/~tkemp/180A](http://www.math.ucsd.edu/~tkemp/180A)

Today: § 3.1-3.2

Next: § 2.4-2.5

HW2 due **TONIGHT!**

Lab 2.2 due Monday, 10/14.

- \* Regrade requests for HW1: window **Tuesday 10/15**  
**8am - 11pm**
  - ↳ separate request for each problem
  - ↳ detailed, polite responses please.
- \* Numerical answers for HW in rubric on Gradescope
- \* Lab 1 solutions posted on [datahub.ucsd.edu](http://datahub.ucsd.edu).

# Random Variables

3.1

Given a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  a random variable is a function

$$X: \Omega \rightarrow \mathbb{R}$$

This is a bad, old-fashioned name. Would be better to call it a random function or random measurement.

E.g. Toss a fair coin 4 times. Let  $X =$  number of tails.

E.g. Shoot an arrow at a circular target.  $Y =$  distance from center.

E.g. Your car is in a minor accident; the damage repair cost is a random number between \$100 and \$1500. Your insurance deductible is \$500.  $Z =$  your out of pocket expenses.

In all these examples, think about what you observe. You can't really see a formula for the function. By repeating the experiment over and over, all you can learn is the

# Probability Distribution

3.2

Given a probability space  $(\Omega, \mathcal{F}, P)$  and a random variable

$$X: \Omega \rightarrow \mathbb{R}$$

the **probability distribution** or **law** of  $X$  is a probability measure  $\mu_X$  ON  $\mathbb{R}$ .

$$A \subseteq \mathbb{R} \rightsquigarrow \mu_X(A) =$$

[Caution: for this to make sense, we need to have a designated set of allowed "events" in  $\mathbb{R}$ ; call this collection  $\mathcal{B}(\mathbb{R})$   
Then, we must have



This is a condition on  $X$ ; we call such functions **measurable**.

We will ignore these technicalities in this course; all our random variables are indeed measurable.

Eg. Toss a fair coin 4 times. Let  $X =$  number of tails.

$$\Omega = \{(x_1, x_2, x_3, x_4) \in \{H, T\}^4\}$$

$\mathbb{P} =$  uniform on  $\Omega$ ;

$$\mathbb{P}\{(x_1, x_2, x_3, x_4)\} = \frac{1}{2^4} = \frac{1}{16}$$

$$\downarrow$$
$$X \in \{0, 1, 2, 3, 4\} \subset \mathbb{R}$$

if  $A \subseteq \mathbb{R}$  does not contain one of these numbers,  $\mu_X(A) = 0$ .

By additivity of probability measures, to understand  $\mu_X$ , just need to know

$$\mu_X(\{k\}) = ? \quad 0 \leq k \leq 4$$

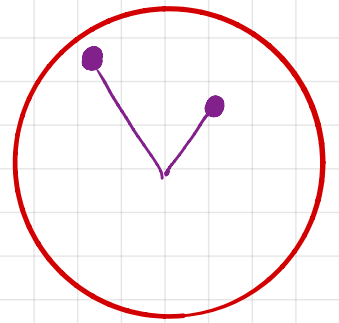
$$\parallel$$
$$\mathbb{P}(X=k) = p_X(k)$$

k	0	1	2	3	4
$p_X(k)$					



Eg. Shoot an arrow at a circular target of radius 1.

$Y =$  distance from center.



Last lecture, you calculated that:

- If  $r \leq 0$ ,  $P(Y \leq r) = 0$
- If  $r \geq 1$ ,  $P(Y \leq r) = 1$ .
- If  $r \in [0, 1]$ ,  $P(Y \leq r) = \frac{\text{Area}(D_r)}{\text{Area}(D_1)} = r^2$ .

$P(\{Y \in (-\infty, r]\})$  ← Can get information about other sets  $A \subseteq \mathbb{R}$  using the properties of  $P$ .

What can we say about  $P(Y = 0.4)$ ?

We will focus mostly on two kinds of random variables:

discrete: There are finitely (or countably) many possible values  $\{k_1, k_2, k_3, \dots\}$  for  $X$ .

↳  $\mu_X$  is described by the **probability mass function**  $p_X(k) = P(X=k)$   
 $k \in \{k_1, k_2, k_3, \dots\}$

In this case, by the laws of probability,

continuous: For any real number  $t \in \mathbb{R}$ ,  $P(X=t) = 0$ .

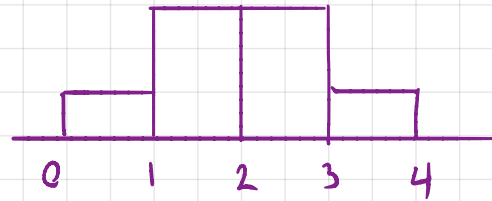
↳  $\mu_X$  is captured by understanding  $P(X \leq r)$  as a function of  $r$ .

# Cumulative Distribution Function (CDF)

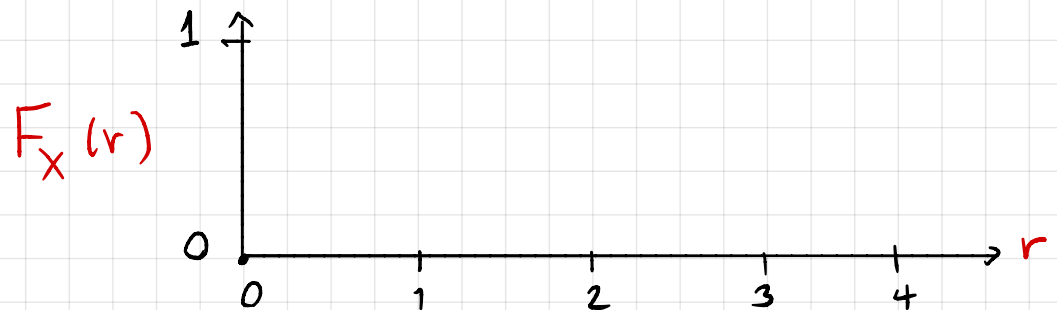
For any random variable  $X$ ,  $F_X(r) = \mathbb{P}(X \leq r)$ . } Actually works in all cases - including discrete.

Ex.  $\mu_X = \text{Bin}(3, \frac{1}{2})$

$k$	0	1	2	3
$P_X(k)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$



- $r < 0$ ,  $\{X \leq r\} =$
- $0 \leq r < 1$ ,  $\{X \leq r\} =$
- $1 \leq r < 2$ ,  $\{X \leq r\} =$
- $2 \leq r < 3$ ,  $\{X \leq r\} =$
- $r \geq 3$ ,  $\{X \leq r\} =$



## Properties of the CDF $F_X(r) = P(X \leq r)$

(1) Monotone increasing:  $s \leq t \Rightarrow F_X(s) \leq F_X(t)$

(2)  $\lim_{r \rightarrow -\infty} F_X(r) = 0$ ,  $\lim_{r \rightarrow +\infty} F_X(r) = 1$ .

(3) The function  $F_X$  is right-continuous:  $\lim_{t \rightarrow r^+} F_X(t) = F_X(r)$ .

Corollary: If  $X$  is a continuous random variable,  $F_X$  is a continuous function.

# Densities

Some continuous random variables have **probability densities**.  
This is an infinitesimal version of a probability mass function.

$X$  discrete,  $\in \{k_1, k_2, k_3, \dots\}$

$$p_X(k) = P(X=k)$$

$$P(X \in A) = \sum_{k \in A} P(X=k)$$

$$= \sum_{k \in A} p_X(k)$$

$$p_X(k) \geq 0, \quad \sum_k p_X(k) = 1.$$

$X$  continuous

$$P(X=t) = 0 \text{ for all } t \in \mathbb{R}.$$

Eg. Shoot an arrow at a circular target of radius 1.

$Y$  = distance from center.

$$\int_{-\infty}^r f(t) dt$$

$$\stackrel{?}{=} \mathbb{P}(Y \in (-\infty, r]) = F_Y(r) = \begin{cases} 0, & r \leq 0 \\ r^2, & 0 \leq r \leq 1 \\ 1, & r \geq 1 \end{cases}$$

Theorem: If  $F_X$  is continuous and piecewise differentiable, then  $X$  has a density  $f_X = F_X'$ .

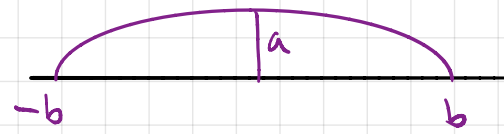
Proof:

Eg. Let  $X =$  a uniformly random number in  $[0,1]$ .

As we discussed in lecture 2, this means

$$P(X \in [s, t]) = t - s \quad \text{if} \quad 0 \leq s < t \leq 1.$$

Eg. Let  $f(t) = c\sqrt{b^2 - t^2}$  for  $|t| \leq b$ , 0 otherwise  
(for some positive constants  $b, c > 0$ ).



Is  $f$  a probability density?



E.g. Your car is in a minor accident; the damage repair cost is a random number between \$100 and \$1500. Your insurance deductible is \$500.  $Z$  = your out of pocket expenses.

The random variable  $Z$  is

- (a) continuous
- (b) discrete
- (c) neither
- (d) both