

# Random variables

March 6, 2019

# Intro

- Sample space and events are basic components of probability
- Similar to numbers in calculus
- Study the relations between numbers we use functions
- In probability we use *random variables*



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# Definition

Let  $(\Omega, P)$  be a probability space.  
A **random variable**  $X$  defined on  
 $(\Omega, P)$  is a quantity that is calculated  
by the outcomes in  $\Omega$ .

- Can think of random variable (RV) as function of outcomes
- RV can take different values with different outcomes
- Can calculate probability of each value of RV

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# Example

Roll 2 dice.  $X = \text{sum of 2 dice}$

- $X$  takes values from 2 to 12
- Can calculate

$$P(X = j), \quad j = 2, 3, \dots, 12$$



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# Distribution

- $P(X = 2) = P(X = 12) = 1/36$
- $P(X = 3) = P(X = 11) = 2/36$
- $P(X = 4) = P(X = 10) = 3/36$
- $P(X = 5) = P(X = 9) = 4/36$
- $P(X = 6) = P(X = 8) = 5/36$
- $P(X = 7) = 6/36$

# Distribution

- $X$  has its own set of all possible values called the range of  $X$
- For each value  $x$  there is a probability that  $X = x$ .
- $X$  creates a new probability space with

$$\sum_x P(X = x) = 1$$

- This probability is called the distribution of  $X$

# Example

Suppose that an individual purchases two electronic components each of which may be either defective or acceptable. Suppose that the four possible results -

$(d, d), (d, a), (a, d), (a, a)$  - have respective probabilities .09, .21, .21, .49 [ $d$  means defective,  $a$  means acceptable, and so on].

Let  $X$  denote the number of acceptable components obtained in the purchase. Find the distribution of  $X$ .

# Solution

- Values of  $X$ : 0,1,2
- $P(X = 0) = .09$
- $P(X = 1) = .21 + .21 = .42$
- $P(X = 2) = .49$



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# Example

- Same example, want to know if there is at least one acceptable component?
- Define new RV

$$I = \begin{cases} 1 & \text{if } X = 1 \text{ or } 2 \\ 0 & \text{if } X = 0 \end{cases}$$

- $I$  : the *indicator RV* of the event “at least one acceptable”



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# Example

$$\begin{aligned}P(\text{at least one acceptable}) \\= P(I = 1) = 0.91\end{aligned}$$



# Types of RV

- If the set of values of  $X$   $\text{Range}(X)$  is finite or countable then  $X$  is called *discrete RV*
- If the set of values of  $X$  is uncountable (like the interval  $[a, b]$ ) then  $X$  is called *continuous RV*

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# Distribution function

The **cumulative distribution function** (cdf)  $F$  of a random variable  $X$  is defined as

$$F(x) = P(X \leq x)$$

for any real value  $x$ .

Each RV has only one cdf.

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# Example

$X$  is the number of acceptable components as in previous example.

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ .09 & \text{if } 0 \leq x < 1 \\ .51 & \text{if } 1 \leq x < 2 \\ 1 & \text{if } 2 \leq x \end{cases}$$



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# Using cdf

$$P(a < X \leq b) = F(b) - F(a)$$



# Example

Suppose  $X$  has cdf

$$F(x) = \begin{cases} 0 & \text{if } x \leq 0 \\ 1 - e^{-x^2} & \text{if } x > 0 \end{cases}$$

What is the probability of  $X > 1$ ?



# Solution

$$\begin{aligned}P(X > 1) &= 1 - P(X \leq 1) \\&= 1 - F(1) \\&= e^{-1} \\&= .368\end{aligned}$$



# Probability mass function

Let  $X$  be a discrete RV. The *probability mass function* (pmf) of  $X$  is defined as

$$p(x) = P(X = x) \text{ for all } x \in \Omega$$



# Properties

- $X$  is discrete  
 $\rightarrow Range(X) = \{x_1, \dots, x_n \dots\}$
- $p(x_i) = P(X = x_i) \geq 0$
- $\sum_{i=1}^{\infty} p(x_i) = 1$
- cdf:  $F(x) = \sum_{x_i < x} p(x)$

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# Example

Suppose  $X$  has 3 values 1, 2, 3 and

$$p(1) = \frac{1}{2}, p(2) = \frac{1}{3}$$

then what is  $p(3)$ ?

$$p(3) = 1 - p(1) - p(2) = 1/6.$$



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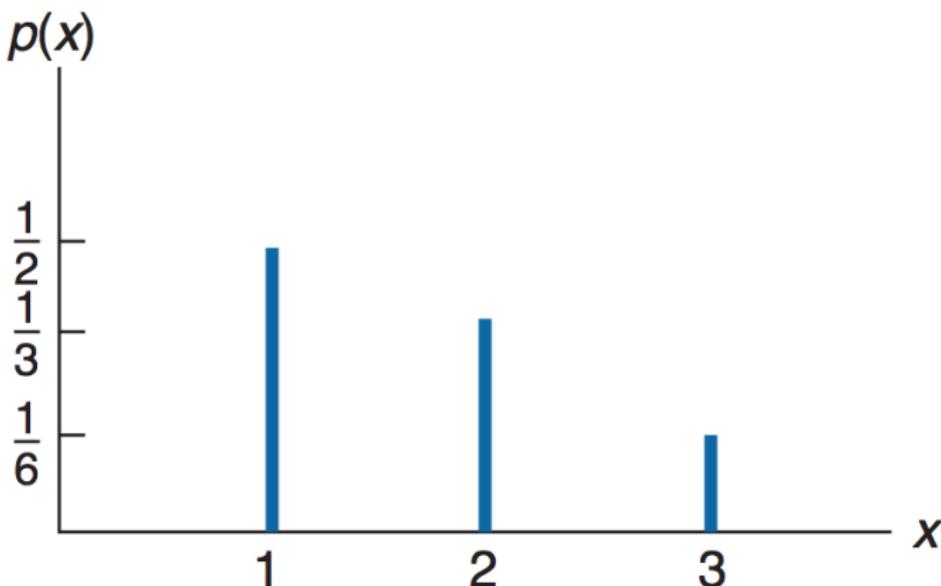
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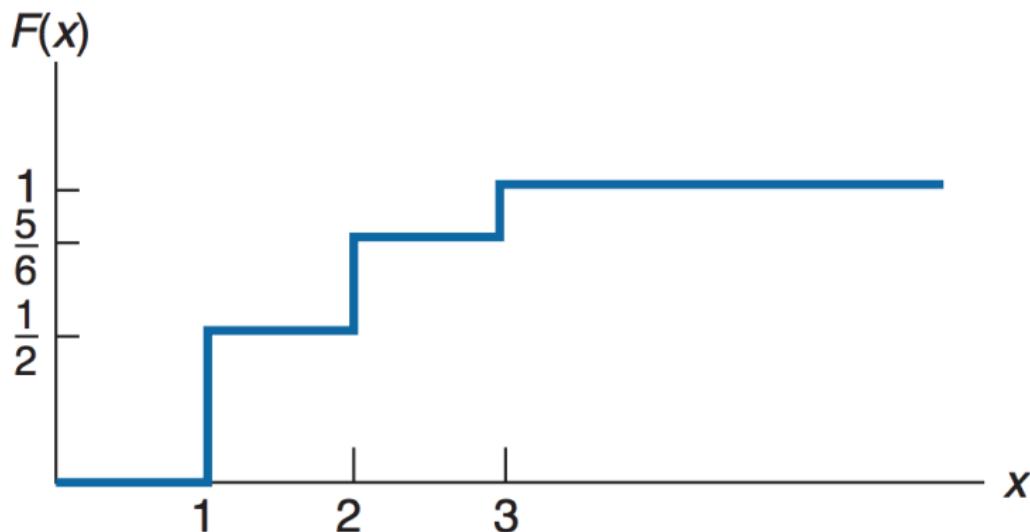
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# Graph of $p(x)$



# Graph of $F(x)$



# Probability density functions

Suppose  $\text{Range}(X)$  is uncountable.  $X$  is *continuous* if there is a non negative function  $f(x)$  so that

$$P(X \leq a) = \int_{-\infty}^a f(x)dx$$

$f(x)$  is called the *probability density function* of  $X$ .

If  $F(x)$  is differentiable then  $f(x) = F'(x)$



# Properties

- $1 = \int_{-\infty}^{\infty} f(x)dx$
- $P(a < X < b) = \int_a^b f(x)dx$
- $P(a \leq X \leq b) = \int_a^b f(x)dx$
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# Example

Suppose the pdf of  $X$  is

$$f(x) = \begin{cases} C(4x - 2x^2) & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

What is the value of  $C$ ?

Find  $P(X > 1)$ ?



# Solution

By property of pdf:

$\int_0^2 C(4x - 2x^2)dx = 1$ , which  
implies  $C = 3/8$ .

Then

$$P(X > 1) = \int_1^2 \frac{3}{8}(4x - 2x^2)dx = \frac{1}{2}$$

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# Practice

Let  $X$  represent the difference between the number of heads and the number of tails obtained when a coin is tossed  $n$  times. What are the possible values of  $X$  ?

# Practice

7. The lifetime in hours of a certain kind of radio tube is a random variable having a probability density function given by

$$f(x) = \begin{cases} 0 & x \leq 100 \\ \frac{100}{x^2} & x > 100 \end{cases}$$

What is the probability that exactly 2 of 5 such tubes in a radio set will have to be replaced within the first 150 hours of operation? Assume that the events  $E_i, i = 1, 2, 3, 4, 5$ , that the  $i$ th such tube will have to be replaced within this time are independent.



# Jointly distributed random variables

# Joint cdf

- Suppose  $X$  and  $Y$  are 2 RVs based on the same experiment.
- We wish to study the relation between them.
- Joint cumulative distribution function

$$F_{X,Y}(x, y) = P(X \leq x, Y \leq y)$$

# Joint cdf

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# cdf from joint cdf

- $P(X \leq x) = F_X(x)$   
 $= F_{X,Y}(X \leq x, Y \leq \infty)$
- $P(Y \leq y) = F_Y(y)$   
 $= F_{X,Y}(X \leq \infty, Y \leq y)$



# cdf from joint cdf

- $P(X \leq x) = F_X(x)$   
 $= F_{X,Y}(X \leq x, Y \leq \infty)$
- $P(Y \leq y) = F_Y(y)$   
 $= F_{X,Y}(X \leq \infty, Y \leq y)$



# Joint pmf

$X$  and  $Y$  are discrete: joint pmf

$$p_{X,Y}(x, y) = P(X = x, Y = y)$$

pmf from joint pmf

$$p_X(x) = \sum_y p_{X,Y}(x, y)$$



# Example

Suppose that 3 batteries are randomly chosen from a group of 3 new, 4 used but still working, and 5 defective batteries. If we let  $X$  and  $Y$  denote, respectively, the number of new and used but still working batteries that are chosen, find the joint probability mass function of  $X$  and  $Y$ .

# Solution

TABLE 4.1  $P\{X = i, Y = j\}$

$i \backslash j$	0	1	2	3	Row Sum $= P\{X = i\}$
0	$\frac{10}{220}$	$\frac{40}{220}$	$\frac{30}{220}$	$\frac{4}{220}$	$\frac{84}{220}$
1	$\frac{30}{220}$	$\frac{60}{220}$	$\frac{18}{220}$	0	$\frac{108}{220}$
2	$\frac{15}{220}$	$\frac{12}{220}$	0	0	$\frac{27}{220}$
3	$\frac{1}{220}$	0	0	0	$\frac{1}{220}$
Column Sums =	$P\{Y = j\}$	$\frac{56}{220}$	$\frac{112}{220}$	$\frac{48}{220}$	$\frac{4}{220}$

# Joint pdf

$X$  and  $Y$  are continuous: joint pdf  
 $f_{X,Y}(x, y)$  so that

$$F_{X,Y}(a, b) = \int_{-\infty}^a \int_{-\infty}^b f_{X,Y}(x, y) dx dy$$

and

$$f_{X,Y}(x, y) = \frac{\partial^2}{\partial x \partial y} F_{X,Y}(x, y)$$

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# Example

$X$  and  $Y$  have joint pdf

$$f_{X,Y}(x,y) = \begin{cases} 2e^{-x}e^{-2y} & \text{if } 0 < x, y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Find  $P(X < Y)$ .



# Solution

$$\begin{aligned} P\{X < Y\} &= \iint_{(x,y):x < y} 2e^{-x}e^{-2y} dx dy \\ &= \int_0^{\infty} \int_0^y 2e^{-x}e^{-2y} dx dy \\ &= \int_0^{\infty} 2e^{-2y}(1 - e^{-y}) dy \\ &= \int_0^{\infty} 2e^{-2y} dy - \int_0^{\infty} 2e^{-3y} dy \\ &= 1 - \frac{2}{3} \\ &= \frac{1}{3} \end{aligned}$$



# pdf from joint pdf

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx$$



# Practice

**10.** The joint probability density function of  $X$  and  $Y$  is given by

$$f(x, y) = \frac{6}{7} \left( x^2 + \frac{xy}{2} \right), \quad 0 < x < 1, \quad 0 < y < 2$$

- (a)** Verify that this is indeed a joint density function.
- (b)** Compute the density function of  $X$ .
- (c)** Find  $P\{X > Y\}$ .



# Independence

$X$  and  $Y$  are independent if for any  $x, y$

$$P(X \leq x, Y \leq y) = P(X \leq x)P(Y \leq y)$$

Can check independency by looking at cdf, pmf and pdf



If  $X$  and  $Y$  are independent then

- $F_{X,Y}(x, y) = F_X(x)F_Y(y)$
- $p_{X,Y}(x, y) = p_X(x)p_Y(y)$
- $f_{X,Y}(x, y) = f_X(x)f_Y(y)$

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# Practice

A bin of 5 transistors is known to contain 3 that are defective. The transistors are to be tested, one at a time, until the defective ones are identified. Denote by  $N_1$  the number of tests made until the first defective is spotted and by  $N_2$  the number of additional tests until the second defective is spotted; find the joint probability mass function of  $N_1$  and  $N_2$ .



# Homework 4

Chapter 4: 1, 3, 4, 6, 8, 10, 12