# 3 Probability Distributions

(Ch 3.4.1, 3.4.2, 4.1, 4.2, 4.3)

# **Probability Distribution Functions**

#### **Probability distribution function (pdf):**

Function for mapping random variables to real numbers.

#### Discrete random variable:

Values constitute a finite or countably infinite set.

#### Continuous random variable:

Set of possible values is the set of real numbers **R**, one interval, or a disjoint union of intervals on the real line.

# Random Variables

#### **Notation!**

- 1. Random variables usually denoted by uppercase letters near the end of our alphabet (e.g. *X, Y*).
- 2. Particular value now use lowercase letters, such as *x*, which correspond to the r.v. *X*.

# **Properties of PDFs**

For f(x) to be a legitimate pdf, it must satisfy the following two conditions:

- 1.  $f(x) \ge 0$  for all x.
- 2.  $\sum_{Q|X \in X} P(X = X) = 1$  for discrete distributions.

$$\int_{all\ x\in X} f(x)dx = 1$$
 for continuous distributions.

# Discrete Random Variables

# PDFs for Discrete RVs

The pdf of a discrete r.v. X describes how the total probability is distributed among all the possible range values of the r.v. X:

f(x) = p(X=x), for each value x in the range of X

# Example

A lab has 6 computers.

Let X denote the number of these computers that are in use during lunch hour -- {0, 1, 2... 6}.

Suppose that the probability distribution of *X* is as given in the following table:

X	0	1	2	3	4	5	6
p(x)	.05	.10	.15	.25	.20	.15	.10

# Example

From here, we can find many things:

- 1. Probability that at most 2 computers are in use
- 2. Probability that at least half of the computers are in use
- 3. Probability that there are 3 or 4 computers free

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p(x)	.05	.10	.15	.25	.20	.15	.10

# **Bernoulli Distribution**

**Bernoulli random variable:** Any random variable whose only possible values are 0 or 1.

This is a discrete random variable – why?

This distribution is specified with a single parameter:

$$P(X = x) = \pi^{x}(1 - \pi)^{(1-x)}; x = 0, 1$$

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Examples?

# **Geometric Distribution**

A patient is waiting for a suitable matching kidney donor for a transplant. The probability that a randomly selected donor is a suitable match is 0.1.

What is the probability the first donor tested is the first matching donor? Second? Third?

# **Geometric Distribution**

Continuing in this way, a general formula for the pmf emerges:

$$P(X = x) = (1 - \pi)^x \pi; \quad x = 0,1,2,...$$

The parameter  $\pi$  can assume any value between 0 and 1. Depending on what parameter  $\pi$  is, we get different members of the **geometric** distribution.

**NOTATION**: We write  $X \sim G(\pi)$  to indicate that X is a geometric rv with success probability  $\pi$ .

The Binomial r.v. counts the **total number of successes** out of *n* trials, where *X* is the number of successes.

- Each trial must be independent of the previous experiment.
- The probability of success must be the same for each trial.

Example: A dice is tossed four times. A "success" is defined as rolling a 1 or a 6.

- The probability of success is 1/3.
- What is P(X = 2)?
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Let's use the probabilities we calculated above to derive the binomial pdf.

**NOTATION**: We write  $X \sim \text{Bin}(n, \pi)$  to indicate that X is a binomial rv based on n Bernoulli trials with success probability  $\pi$ .

# The Negative Binomial Distribution

Consider the dice example for the binomial distribution.

Now we instead want to find the probability that we roll 3 "failures" (i.e. a 2, 3, 4, or 5) before the 2<sup>nd</sup> success.

How is this related to the binomial distribution?

# The Negative Binomial Distribution

Consider the dice example for the binomial distribution. What is the probability that exactly 3 successes occur before 2 failures occur?

**NOTATION**: We write  $X \sim \text{NB}(r, \pi)$  to indicate that X is a negative binomial r.v., with x failures occurring before r successes, where the probability of success is equal to  $\pi$ .

# The Poisson Probability Distribution

A Poisson r.v. describes the total number of events that happen in a certain <u>time period</u>.

#### Examples:

- # of vehicles arriving at a parking lot in one week
- # of gamma rays hitting a satellite per hour
- # of cookies chips in a length of cookie dough

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- # of gamma rays hitting a satellite per hour
- # of cookies sold at a bake sale in 1 hour

# The Poisson Probability Distribution

A Poisson r.v. describes the total number of events that happen in a certain <u>time period</u>.

A discrete random variable X is said to have a **Poisson distribution** with parameter  $\lambda$  ( $\lambda$  > 0) if the pdf of X is

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}; \qquad x = 0, 1, 2, ....$$

**NOTATION**: We write  $X \sim P(\lambda)$  to indicate that X is a Poisson r.v. with parameter  $\lambda$ .

# Example

Let X denote the number of mosquitoes captured in a trap during a given time period.

Suppose that X has a Poisson distribution with  $\lambda = 4.5$ .

What is the probability that the trap contains 5 mosquitoes?

# Example problem

# **Cumulative Distribution Functions**

<u>Definition:</u> The cumulative distribution function ( $\underline{cdf}$ ) is denoted with F(x).

For a discrete r.v. X with pdf f(x) = P(X = x), F(x) is defined for every real number x by

$$F(x) = P(X \le x) = \sum_{y:y \le x} P(X = y)$$

For any number x, the cdf F(x) is the <u>probability</u> that the observed value of X will be <u>at most</u> x.

# Example

Suppose we are given the following pmf:

$$p(x) = \begin{cases} .500 & x = 0 \\ .167 & x = 1 \\ .333 & x = 2 \\ 0 & \text{otherwise} \end{cases}$$

Then, calculate:

F(0), F(1), F(2)

What about F(1.5)? F(20.5)?

Is 
$$P(X < 1) = P(X <= 1)$$
?

# Continuous Random Variables

# Continuous Random Variables

A random variable *X* is **continuous** if possible values comprise either a single interval on the number line or a union of disjoint intervals.

Example: If in the study of the ecology of a lake, *X*, the r.v. may be depth measurements at randomly chosen locations.

# **Cumulative Distribution Functions**

<u>Definition:</u> The cumulative distribution function ( $\underline{cdf}$ ) is denoted with F(x).

For a discrete r.v. X with pdf f(x), F(x) is defined for every real number x by

$$F(x) = \int_{y:y \le x} f(y)dy$$

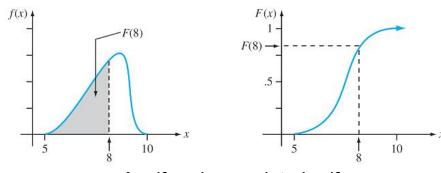
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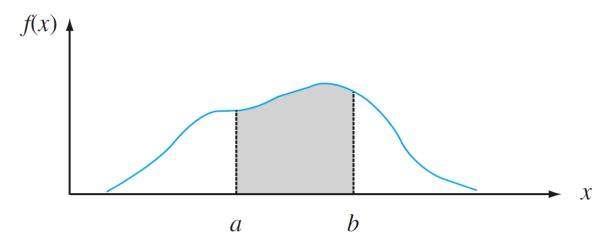
This is illustrated below, where F(x) increases smoothly as x increases.



A pdf and associated cdf

# PDFs for Continuous RVs

The probability that *X* takes on a value in the interval [*a*, *b*] is the area above this interval and under the graph of the density function:



 $P(a \le X \le b)$  = the area under the density curve between a and b

# Example

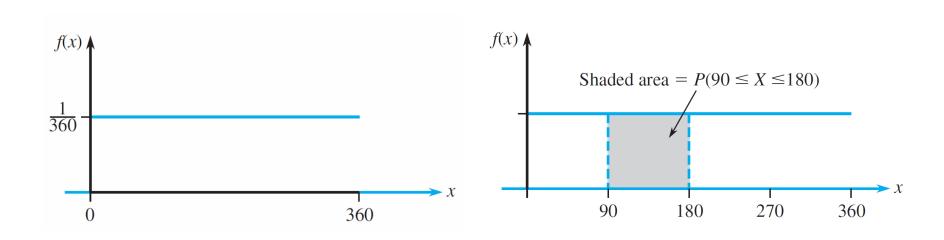
Consider the reference line connecting the valve stem on a tire to the center point.

Let X be the angle measured clockwise to the location of an imperfection. The pdf for X is

$$f(x) = \begin{cases} \frac{1}{360} & 0 \le x < 360\\ 0 & \text{otherwise} \end{cases}$$

# Example

#### The pdf is shown graphically below:



The pdf and probability from example on previous slide.

Clearly  $f(x) \ge 0$ . How can we show that the area of this pdf is equal to 1?

How do we calculate  $P(90 \le X \le 180)$ ?

What is the probability that the angle of occurrence is within 90° of the reference line? (The reference line is at 0 degrees.)

# **Uniform Distribution**

The previous problem was an example of the **uniform distribution**.

**Definition:** A continuous rv X is said to have a **uniform distribution** on the interval [a, b] if the pdf of X is

$$f(x;a,b) = \frac{1}{b-a}, \quad a \le x \le b$$

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**NOTATION**: We write  $X \sim U(a, b)$  to indicate that X is a uniform rv with a lower bound equal to a and an upper bound equal to b.

# **Exponential Distribution**

The family of exponential distributions provides probability models that are very widely used in engineering and science disciplines to describe **time-to-event data**.

Examples?

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**Definition:** X is said to have an exponential distribution with the rate parameter  $\lambda$  ( $\lambda$  > 0) if the pdf of X is

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# **Exponential Distribution**

The family of exponential distributions provides probability models that are very widely used in engineering and science disciplines to describe **time-to-event data**.

A partial reason for the popularity of such applications is the **memoryless property** of the Exponential distribution.

# The Exponential Distributions

Suppose a light bulb's lifetime is exponentially distributed with parameter  $\lambda$ .

What is the probability that the lifetime of the light bulb lasts less than *t* hours?

What is the probability that the lifetime of the light bulb lasts more than *t* hours?

# The Exponential Distributions

Suppose a light bulb's lifetime is exponentially distributed with parameter  $\lambda$ .

Now say you turn the light bulb on and then leave. You come back after  $t_0$  hours to find it still on. What is the probability that the light bulb will last for <u>at least additional that</u> hours?

In symbols, we are looking for  $P(X \ge t + t_0 \mid X \ge t_0)$ .

How would we calculate this?

The Weibull distribution is used similarly to the exponential distribution to model times to an event, but with an extra parameter included for flexibility.

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#### **Definition**

A random variable X is said to have a **Weibull distribution** with parameters  $\alpha$  and  $\beta$  ( $\alpha$  > 0,  $\beta$  > 0) if the pdf of X is

$$f(x; \alpha, \beta) = \alpha \beta^{-\alpha} x^{\alpha - 1} e^{-(\frac{x}{\beta})^{\alpha}}, \qquad x \ge 0$$

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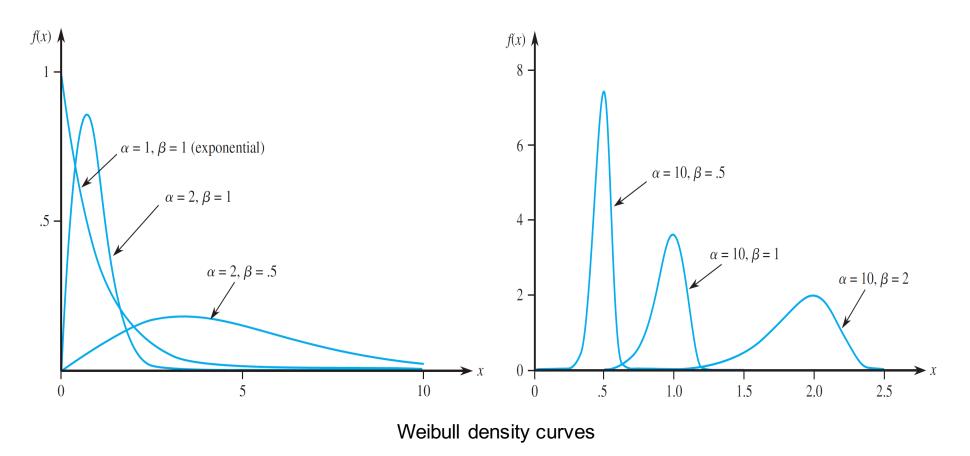
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What is this distribution if alpha = 1?

Both  $\alpha$  and  $\beta$  can be varied to obtain a number of different-looking density curves, as illustrated below.



#### The Beta Distribution

So far, all families of continuous distributions (except for the uniform distribution) had positive density over an infinite interval.

The beta distribution provides positive density only for *X* in an interval from 0 to 1.

The beta distribution is commonly used to model variation in the **proportion or percentage** of a quantity occurring in different samples.

Examples?

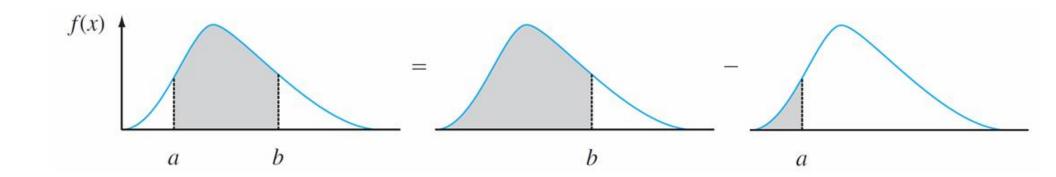
#### The Beta Distribution

#### **Definition**

A random variable X is said to have a **beta distribution** with parameters  $\alpha$ ,  $\beta$  (both positive), if the pdf of X is:

$$f(x; \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\alpha)} x^{\alpha - 1} (1 - x)^{\beta - 1}; 0 \le x \le 1$$

# Using F(x) to Compute Probabilities



# Example

The distribution of the amount of gravel (in tons) sold by a particular construction supply company in a given week is a continuous rv X with pdf

$$f(x) = \begin{cases} \frac{3}{2}(1 - x^2) & 0 \le x \le 1\\ 0 & \text{otherwise} \end{cases}$$

What is the cdf of sales for any x? How do you use this to find the probability that X is less than .25? What about the probability that X is greater than .75? What about

$$P(.25 < X < .75)$$
?

### Percentiles of a Distribution

#### **Definition**

The **median** of a continuous distribution is the 50th percentile of the distribution.

How can we express this in terms of F(x)?

# Example

X is a r.v. such that:

$$f(x) = a^* exp(-ax), x > 0$$

Calculate F(x).

How would you find the median of this distribution?

How do we find the 40th percentile of this distribution?

# Example

You flip and unfair coin four times, where the probability of flipping a head is 60%.

What is the median number of heads for this distribution?