

# STAT 111

## Recitation 3

Mo Huang

Email: [mohuang@wharton.upenn.edu](mailto:mohuang@wharton.upenn.edu)

Office Hours: Wednesdays 3:00 - 4:00 pm, JMHH F96

Slides (adapted from Gemma Moran): [github.com/mohuangx/STAT111-Fall2018](https://github.com/mohuangx/STAT111-Fall2018)

September 21, 2018

## Random Variables: Mean

- ▶ The **mean of a random variable** (expected value) is the long-run average of realizations of the random variable over repeated experiments.

# Random Variables: Mean

- ▶ The **mean of a random variable** (expected value) is the long-run average of realizations of the random variable over repeated experiments.
- ▶ This mean of random variable  $X$  is different from the sample average, which is the average of a *finite* number of observations ( $\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$ , where  $x_1, \dots, x_n$  are observed data).

## Random Variables: Mean

- ▶ The **mean of a random variable** (expected value) is the long-run average of realizations of the random variable over repeated experiments.
- ▶ This mean of random variable  $X$  is different from the sample average, which is the average of a *finite* number of observations ( $\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$ , where  $x_1, \dots, x_n$  are observed data).
- ▶ Let  $X$  be a discrete random variable that can take values  $\{v_1, v_2, \dots, v_k\}$ . Then the mean of  $X$  is given by:

$$\begin{aligned}\mu &= \sum_{i=1}^k v_i P(X = v_i) \\ &= v_1 P(X = v_1) + v_2 P(X = v_2) + \dots + v_k P(X = v_k).\end{aligned}$$

# Random Variables: Mean

- ▶ The **mean of a random variable** (expected value) is the long-run average of realizations of the random variable over repeated experiments.
- ▶ This mean of random variable  $X$  is different from the sample average, which is the average of a *finite* number of observations ( $\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$ , where  $x_1, \dots, x_n$  are observed data).
- ▶ Let  $X$  be a discrete random variable that can take values  $\{v_1, v_2, \dots, v_k\}$ . Then the mean of  $X$  is given by:

$$\begin{aligned}\mu &= \sum_{i=1}^k v_i P(X = v_i) \\ &= v_1 P(X = v_1) + v_2 P(X = v_2) + \dots + v_k P(X = v_k).\end{aligned}$$

- ▶ If  $X$  is binomial with parameters  $n$  and  $\theta$ , then the mean of  $X$  is

$$\mu = n\theta$$

# Random Variables: Variance

- ▶ The **variance of a random variable** is a measure of the *spread* of a distribution - that is, how far away values are from the mean.

# Random Variables: Variance

- ▶ The **variance of a random variable** is a measure of the *spread* of a distribution - that is, how far away values are from the mean.
- ▶ Let  $X$  be a discrete random variable that can take values  $\{v_1, v_2, \dots, v_k\}$ . Then the variance of  $X$  is given by:

$$\sigma^2 = (v_1 - \mu)^2 P(X = v_1) + \dots + (v_k - \mu)^2 P(X = v_k) \quad (1)$$

*or*

$$\sigma^2 = v_1^2 P(X = v_1) + \dots + v_k^2 P(X = v_k) - \mu^2 \quad (2)$$

# Random Variables: Variance

- ▶ The **variance of a random variable** is a measure of the *spread* of a distribution - that is, how far away values are from the mean.
- ▶ Let  $X$  be a discrete random variable that can take values  $\{v_1, v_2, \dots, v_k\}$ . Then the variance of  $X$  is given by:

$$\sigma^2 = (v_1 - \mu)^2 P(X = v_1) + \dots + (v_k - \mu)^2 P(X = v_k) \quad (1)$$

or

$$\sigma^2 = v_1^2 P(X = v_1) + \dots + v_k^2 P(X = v_k) - \mu^2 \quad (2)$$

- ▶ The **standard deviation of a random variable** is the square root of the variance and is denoted by  $\sigma$ .



# Random Variables: Variance

- ▶ The **variance of a random variable** is a measure of the *spread* of a distribution - that is, how far away values are from the mean.
- ▶ Let  $X$  be a discrete random variable that can take values  $\{v_1, v_2, \dots, v_k\}$ . Then the variance of  $X$  is given by:

$$\sigma^2 = (v_1 - \mu)^2 P(X = v_1) + \dots + (v_k - \mu)^2 P(X = v_k) \quad (1)$$

or

$$\sigma^2 = v_1^2 P(X = v_1) + \dots + v_k^2 P(X = v_k) - \mu^2 \quad (2)$$

- ▶ The **standard deviation of a random variable** is the square root of the variance and is denoted by  $\sigma$ .
- ▶ For a **binomial** random variable  $X \sim \mathcal{B}(n, \theta)$ :

$$\sigma^2 = n\theta(1 - \theta)$$

# Random Variables: Questions

Q1: Let  $X$  be a random variable with the below distribution. Find the mean and the variance of  $X$  using formula (1) and then formula (2).

$x$	-3	-1	4	5
$P(X = x)$	0.1	0.3	0.4	0.2

Table: Probability distribution of  $X$ .

## Random Variables: Questions

Q1: Let  $X$  be a random variable with the below distribution. Find the mean and the variance of  $X$  using formula (1) and then formula (2).

$x$	-3	-1	4	5
$P(X = x)$	0.1	0.3	0.4	0.2

Table: Probability distribution of  $X$ .

A1:  $\mu = -3 \times 0.1 - 1 \times 0.3 + 4 \times 0.4 + 5 \times 0.2 = 2.$

$$\begin{aligned}\sigma^2 &= (-3 - 2)^2(0.1) + (-1 - 2)^2(0.3) + (4 - 2)^2(0.4) + (5 - 2)^2(0.2) \\ &= 8.6.\end{aligned}$$

$$\begin{aligned}\sigma^2 &= (-3)^2(0.1) + (-1)^2(0.3) + 4^2(0.4) + 5^2(0.2) - 2^2 \\ &= 8.6.\end{aligned}$$

## Random Variables: Questions

Q2: Let  $X$  be a random variable with the below distribution. Write the probability distribution of  $Y = 2X$  in tableau form.

$x$	-3	-1	4	5
$P(X = x)$	0.1	0.3	0.4	0.2

Table: Probability distribution of  $X$ .

## Random Variables: Questions

Q2: Let  $X$  be a random variable with the below distribution. Write the probability distribution of  $Y = 2X$  in tableau form.

$x$	-3	-1	4	5
$P(X = x)$	0.1	0.3	0.4	0.2

Table: Probability distribution of  $X$ .

A2: Probability distribution of  $Y$ :

$y$	-6	-2	8	10
$P(Y = y)$	0.1	0.3	0.4	0.2

Table: Probability distribution of  $Y$ .

# Properties of mean and variance

- ▶ Suppose  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$ . Consider fixed number  $c$ .

# Properties of mean and variance

- ▶ Suppose  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$ . Consider fixed number  $c$ .
- ▶ Mean properties
  - ▶ Mean of  $X + c \implies \mu + c$ .
  - ▶ Mean of  $cX \implies c\mu$ .

# Properties of mean and variance

- ▶ Suppose  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$ . Consider fixed number  $c$ .
- ▶ Mean properties
  - ▶ Mean of  $X + c \implies \mu + c$ .
  - ▶ Mean of  $cX \implies c\mu$ .
- ▶ Variance properties
  - ▶ Variance of  $X + c \implies \sigma^2$ .
  - ▶ Variance of  $cX \implies c^2\sigma^2$ .



# Properties of mean and variance

- ▶ Suppose  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$ . Consider fixed number  $c$ .

- ▶ Mean properties

- ▶ Mean of  $X + c \implies \mu + c$ .
- ▶ Mean of  $cX \implies c\mu$ .

- ▶ Variance properties

- ▶ Variance of  $X + c \implies \sigma^2$ .
- ▶ Variance of  $cX \implies c^2\sigma^2$ .

Q3: Suppose  $X$  has mean  $\mu_X = 3$  and variance  $\sigma_X^2 = 4$ . What is the mean and variance of  $Y = 2 + 5X$ ?

# Properties of mean and variance

- ▶ Suppose  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$ . Consider fixed number  $c$ .
- ▶ Mean properties
  - ▶ Mean of  $X + c \implies \mu + c$ .
  - ▶ Mean of  $cX \implies c\mu$ .
- ▶ Variance properties
  - ▶ Variance of  $X + c \implies \sigma^2$ .
  - ▶ Variance of  $cX \implies c^2\sigma^2$ .

Q3: Suppose  $X$  has mean  $\mu_X = 3$  and variance  $\sigma_X^2 = 4$ . What is the mean and variance of  $Y = 2 + 5X$ ?

A3:  $\mu_Y = 2 + 5\mu_X = 2 + 5(3) = 17$

$$\sigma_Y^2 = 5^2\sigma_X^2 = 5^2(4) = 100$$

# Many Random Variables

- ▶ Suppose we plan an experiment with a sample size of  $n$ .

# Many Random Variables

- ▶ Suppose we plan an experiment with a **sample size** of  $n$ .
- ▶ We have  $n$  future outcomes, or random variables, denoted by:  
 $X_1, X_2, \dots, X_n$ .

# Many Random Variables

- ▶ Suppose we plan an experiment with a **sample size** of  $n$ .
- ▶ We have  $n$  future outcomes, or random variables, denoted by:  
 $X_1, X_2, \dots, X_n$ .
- ▶ We say  $\{X_1, \dots, X_n\}$  are **independently and identically distributed (or i.i.d)** if:
  - ▶ All the  $X_i$ s are independent of each other.
  - ▶ Each  $X_i$  has the same probability distribution.

# Many Random Variables

- ▶ Suppose we plan an experiment with a **sample size** of  $n$ .
- ▶ We have  $n$  future outcomes, or random variables, denoted by:  
 $X_1, X_2, \dots, X_n$ .
- ▶ We say  $\{X_1, \dots, X_n\}$  are **independently and identically distributed (or i.i.d)** if:
  - ▶ All the  $X_i$ s are independent of each other.
  - ▶ Each  $X_i$  has the same probability distribution.
- ▶ For example:
  - ▶ I plan to roll a dice  $n$  times.  $X_i$  represents the future outcome of the  $i$ th roll.  $X_i$  is *independent* of the other rolls of the dice, and it has the same probability of getting a 1, 2, 3, 4, 5, 6 as the other rolls.

# Many Random Variables

- Recall the sample average is given by:

$$\bar{x} = \frac{x_1 + \cdots + x_n}{n}.$$

# Many Random Variables

- ▶ Recall the sample average is given by:

$$\bar{x} = \frac{x_1 + \cdots + x_n}{n}.$$

- ▶ What if we haven't observed the data  $x_1, \dots, x_n$  yet?



# Many Random Variables

- ▶ Recall the sample average is given by:

$$\bar{x} = \frac{x_1 + \cdots + x_n}{n}.$$

- ▶ What if we haven't observed the data  $x_1, \dots, x_n$  yet?
- ▶ Before an experiment, the average is *also* a random variable:

$$\bar{X} = \frac{X_1 + \cdots + X_n}{n}$$

# Many Random Variables

- ▶ Recall the sample average is given by:

$$\bar{x} = \frac{x_1 + \cdots + x_n}{n}.$$

- ▶ What if we haven't observed the data  $x_1, \dots, x_n$  yet?
- ▶ Before an experiment, the average is *also* a random variable:

$$\bar{X} = \frac{X_1 + \cdots + X_n}{n}$$

- ▶ The *sum* is also a random variable:

$$T_n = X_1 + \cdots + X_n$$

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

$$\text{mean}(X + Y) = \text{mean}(X) + \text{mean}(Y).$$

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

$$\text{mean}(X + Y) = \text{mean}(X) + \text{mean}(Y).$$

- ▶ If  $X$ ,  $Y$  are also independent,

$$\text{variance}(X + Y) = \text{variance}(X) + \text{variance}(Y).$$

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

$$\text{mean}(X + Y) = \text{mean}(X) + \text{mean}(Y).$$

- ▶ If  $X, Y$  are also independent,

$$\text{variance}(X + Y) = \text{variance}(X) + \text{variance}(Y).$$

- ▶ For constants  $a, b$ , we have

$$\text{mean}(aX + bY) = a \times \text{mean}(X) + b \times \text{mean}(Y)$$

$$\text{variance}(aX + bY) = a^2 \times \text{variance}(X) + b^2 \times \text{variance}(Y).$$

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

$$\text{mean}(X + Y) = \text{mean}(X) + \text{mean}(Y).$$

- ▶ If  $X, Y$  are also independent,

$$\text{variance}(X + Y) = \text{variance}(X) + \text{variance}(Y).$$

- ▶ For constants  $a, b$ , we have

$$\text{mean}(aX + bY) = a \times \text{mean}(X) + b \times \text{mean}(Y)$$

$$\text{variance}(aX + bY) = a^2 \times \text{variance}(X) + b^2 \times \text{variance}(Y).$$

- ▶ Let  $D = X - Y$ . What is the variance of  $D$ ?

# Many Random Variables

- ▶ Let  $X$  and  $Y$  be two random variables. Then,

$$\text{mean}(X + Y) = \text{mean}(X) + \text{mean}(Y).$$

- ▶ If  $X, Y$  are also independent,

$$\text{variance}(X + Y) = \text{variance}(X) + \text{variance}(Y).$$

- ▶ For constants  $a, b$ , we have

$$\text{mean}(aX + bY) = a \times \text{mean}(X) + b \times \text{mean}(Y)$$

$$\text{variance}(aX + bY) = a^2 \times \text{variance}(X) + b^2 \times \text{variance}(Y).$$

- ▶ Let  $D = X - Y$ . What is the variance of  $D$ ?

$$\text{variance}(D) = \text{variance}(X) + \text{variance}(Y).$$



# Many Random Variables

- ▶ Let  $X_1, \dots, X_n$  be i.i.d. random variables, each with mean  $\mu$  and variance  $\sigma^2$ .
- ▶ Then for the sum,  $T_n$ :

$$\text{mean of } T_n = n\mu, \quad \text{variance of } T_n = n\sigma^2$$

- ▶ For the average,  $\bar{X}$ :

$$\text{mean of } \bar{X} = \mu, \quad \text{variance of } \bar{X} = \frac{\sigma^2}{n}.$$

# Questions

- Q4: Suppose we have a company producing a medicine. Each day the *mean* amount of medicine produced is 500 mg and the *variance* is 900  $\text{mg}^2$ . Assume the amount produced each day is independent.
- (i) Let  $T_n$  be the total amount of medicine produced in a week. Find the mean and variance of  $T_n$ .

# Questions

Q4: Suppose we have a company producing a medicine. Each day the *mean* amount of medicine produced is 500 mg and the *variance* is 900 mg<sup>2</sup>. Assume the amount produced each day is independent.

- (i) Let  $T_n$  be the total amount of medicine produced in a week. Find the mean and variance of  $T_n$ .

$$\text{Mean}(T_n) = 5 \times 500 = 2500$$

$$\text{Var}(T_n) = 5 \times 900 = 4500$$

# Questions

Q4: Suppose we have a company producing a medicine. Each day the *mean* amount of medicine produced is 500 mg and the *variance* is 900 mg<sup>2</sup>. Assume the amount produced each day is independent.

- (i) Let  $T_n$  be the total amount of medicine produced in a week. Find the mean and variance of  $T_n$ .

$$\text{Mean}(T_n) = 5 \times 500 = 2500$$

$$\text{Var}(T_n) = 5 \times 900 = 4500$$

- (ii) Let  $\bar{X}$  be the average amount of medicine produced in a 5-day week. Find the mean and variance of  $\bar{X}$ .

# Questions

Q4: Suppose we have a company producing a medicine. Each day the *mean* amount of medicine produced is 500 mg and the *variance* is 900 mg<sup>2</sup>. Assume the amount produced each day is independent.

- (i) Let  $T_n$  be the total amount of medicine produced in a week. Find the mean and variance of  $T_n$ .

$$\text{Mean}(T_n) = 5 \times 500 = 2500$$

$$\text{Var}(T_n) = 5 \times 900 = 4500$$

- (ii) Let  $\bar{X}$  be the average amount of medicine produced in a 5-day week. Find the mean and variance of  $\bar{X}$ .

$$\text{Mean}(\bar{X}) = 500$$

$$\text{Var}(\bar{X}) = 900/5 = 180$$

# Proportions

- ▶ Sometimes it is necessary to consider the *proportion* of “successes” in a Binomial trial (instead of the total number of successes)

# Proportions

- ▶ Sometimes it is necessary to consider the *proportion* of “successes” in a Binomial trial (instead of the total number of successes)
- ▶ Proportions are a type of average!

# Proportions

- ▶ Sometimes it is necessary to consider the *proportion* of “successes” in a Binomial trial (instead of the total number of successes)
- ▶ Proportions are a type of average!
- ▶ Let

$$Y_i = \begin{cases} 1 & \text{if the } i\text{th trial is a “success”} \\ 0 & \text{if the } i\text{th trial is a “failure”} \end{cases}$$



# Proportions

- ▶ Sometimes it is necessary to consider the *proportion* of “successes” in a Binomial trial (instead of the total number of successes)
- ▶ Proportions are a type of average!
- ▶ Let

$$Y_i = \begin{cases} 1 & \text{if the } i\text{th trial is a “success”} \\ 0 & \text{if the } i\text{th trial is a “failure”} \end{cases}$$

- ▶ Then we have  $Y_1, \dots, Y_n$  where  $Y_i \sim \text{Binomial}(1, \theta)$ .

# Proportions

- ▶ Sometimes it is necessary to consider the *proportion* of “successes” in a Binomial trial (instead of the total number of successes)
- ▶ Proportions are a type of average!
- ▶ Let

$$Y_i = \begin{cases} 1 & \text{if the } i\text{th trial is a “success”} \\ 0 & \text{if the } i\text{th trial is a “failure”} \end{cases}$$

- ▶ Then we have  $Y_1, \dots, Y_n$  where  $Y_i \sim \text{Binomial}(1, \theta)$ .
- ▶ The proportion of successes is the average:

$$P = \frac{Y_1 + \dots + Y_n}{n}$$

# Proportions

- $Y_i \sim \text{Binomial}(1, \theta)$ . Recall:

$$\text{Mean}(Y_i) = \theta, \quad \text{Var}(Y_i) = \theta(1 - \theta).$$

# Proportions

- ▶  $Y_i \sim \text{Binomial}(1, \theta)$ . Recall:

$$\text{Mean}(Y_i) = \theta, \quad \text{Var}(Y_i) = \theta(1 - \theta).$$

- ▶ The proportion of successes is the average:

$$P = \frac{Y_1 + \cdots + Y_n}{n}$$

# Proportions

- ▶  $Y_i \sim \text{Binomial}(1, \theta)$ . Recall:

$$\text{Mean}(Y_i) = \theta, \quad \text{Var}(Y_i) = \theta(1 - \theta).$$

- ▶ The proportion of successes is the average:

$$P = \frac{Y_1 + \cdots + Y_n}{n}$$

- ▶ Then,

$$\text{Mean}(P) = \theta, \quad \text{Var}(P) = \frac{\theta(1 - \theta)}{n}$$

.

# Proportions

- ▶  $Y_i \sim \text{Binomial}(1, \theta)$ . Recall:

$$\text{Mean}(Y_i) = \theta, \quad \text{Var}(Y_i) = \theta(1 - \theta).$$

- ▶ The proportion of successes is the average:

$$P = \frac{Y_1 + \cdots + Y_n}{n}$$

- ▶ Then,

$$\text{Mean}(P) = \theta, \quad \text{Var}(P) = \frac{\theta(1 - \theta)}{n}$$

- Q5: Suppose we plan to toss a coin 20 times and  $\text{Prob}(H) = 0.7$ . Let  $P$  be the *proportion* of heads that we toss. Find the mean and variance of  $P$ .

# Proportions

- ▶  $Y_i \sim \text{Binomial}(1, \theta)$ . Recall:

$$\text{Mean}(Y_i) = \theta, \quad \text{Var}(Y_i) = \theta(1 - \theta).$$

- ▶ The proportion of successes is the average:

$$P = \frac{Y_1 + \cdots + Y_n}{n}$$

- ▶ Then,

$$\text{Mean}(P) = \theta, \quad \text{Var}(P) = \frac{\theta(1 - \theta)}{n}$$

.

Q5: Suppose we plan to toss a coin 20 times and  $\text{Prob}(H) = 0.7$ . Let  $P$  be the *proportion* of heads that we toss. Find the mean and variance of  $P$ .

A5:

$$\text{Mean}(P) = 0.7, \quad \text{Var}(P) = 0.7 \times 0.3/20 = 0.0105.$$

## Questions

Q6: Suppose the company producing a medicine has different means and variances the amount produced on each day of the week:

Day	Mean	Variance
Monday ( $X_1$ )	450	1200
Tuesday ( $X_2$ )	550	800
Wednesday ( $X_3$ )	600	500
Thursday ( $X_4$ )	550	800
Friday ( $X_5$ )	350	1200



## Questions

Q6: Suppose the company producing a medicine has different means and variances the amount produced on each day of the week:

Day	Mean	Variance
Monday ( $X_1$ )	450	1200
Tuesday ( $X_2$ )	550	800
Wednesday ( $X_3$ )	600	500
Thursday ( $X_4$ )	550	800
Friday ( $X_5$ )	350	1200

Find mean and variance of both the sum  $T_n$  and the average  $\bar{X}$ .

## Questions

Q6: Suppose the company producing a medicine has different means and variances the amount produced on each day of the week:

Day	Mean	Variance
Monday ( $X_1$ )	450	1200
Tuesday ( $X_2$ )	550	800
Wednesday ( $X_3$ )	600	500
Thursday ( $X_4$ )	550	800
Friday ( $X_5$ )	350	1200

Find mean and variance of both the sum  $T_n$  and the average  $\bar{X}$ .

A6:  $X_1, X_2, X_3, X_4$ , and  $X_5$  are no longer *i.i.d.*!

$$\text{Mean}(T_n) = 450 + 550 + 600 + 550 + 350 = 2500$$

$$\text{Var}(T_n) = 1200 + 800 + 500 + 800 + 1200 = 4500$$

## Questions

Q6: Suppose the company producing a medicine has different means and variances the amount produced on each day of the week:

Day	Mean	Variance
Monday ( $X_1$ )	450	1200
Tuesday ( $X_2$ )	550	800
Wednesday ( $X_3$ )	600	500
Thursday ( $X_4$ )	550	800
Friday ( $X_5$ )	350	1200

Find mean and variance of both the sum  $T_n$  and the average  $\bar{X}$ .

A6:  $X_1, X_2, X_3, X_4$ , and  $X_5$  are no longer *i.i.d.*!

$$\text{Mean}(T_n) = 450 + 550 + 600 + 550 + 350 = 2500$$

$$\text{Var}(T_n) = 1200 + 800 + 500 + 800 + 1200 = 4500$$

$$\text{Mean}(\bar{X}) = 1/n \times \text{Mean}(T_n) = 500$$

$$\text{Var}(\bar{X}) = 1/n^2 \times \text{Var}(T_n) = 4500/25 = 180$$

# Questions

Q7: Let  $P_1$  be the proportion of heads in 50 coin tosses, where  $P(H) = 0.6$ . Find  $Mean(P_1)$  and  $Var(P_1)$ .

# Questions

Q7: Let  $P_1$  be the proportion of heads in 50 coin tosses, where  $P(H) = 0.6$ . Find  $Mean(P_1)$  and  $Var(P_1)$ .

A7:  $Mean(P_1) = 0.6$  and  $Var(P_1) = 0.6 \times 0.4/50 = 0.0048$ .

# Questions

Q7: Let  $P_1$  be the proportion of heads in 50 coin tosses, where  $P(H) = 0.6$ . Find  $Mean(P_1)$  and  $Var(P_1)$ .

A7:  $Mean(P_1) = 0.6$  and  $Var(P_1) = 0.6 \times 0.4/50 = 0.0048$ .

Q8: Let  $P_2$  be the proportion of heads in 20 coin tosses, where  $P(H) = 0.7$ . From earlier,  $Mean(P_2) = 0.7$  and  $Var(P_2) = 0.0105$ . Let  $D = P_1 - P_2$ . Find the mean and variance of  $D$ .

# Questions

Q7: Let  $P_1$  be the proportion of heads in 50 coin tosses, where  $P(H) = 0.6$ . Find  $Mean(P_1)$  and  $Var(P_1)$ .

A7:  $Mean(P_1) = 0.6$  and  $Var(P_1) = 0.6 \times 0.4/50 = 0.0048$ .

Q8: Let  $P_2$  be the proportion of heads in 20 coin tosses, where  $P(H) = 0.7$ . From earlier,  $Mean(P_2) = 0.7$  and  $Var(P_2) = 0.0105$ . Let  $D = P_1 - P_2$ . Find the mean and variance of  $D$ .

A8:  $Mean(D) = 0.6 - 0.7 = -0.1$   
 $Var(D) = 0.0048 + 0.0105 = 0.0153$

# Continuous Random Variables

- ▶ So far, we have just considered discrete random variables; those whose possible values are countable.
- ▶ A **continuous random variable** can take continuous values in a future experiment.
- ▶ Every continuous random variable  $X$  has an associated **density function**  $f(x)$ .

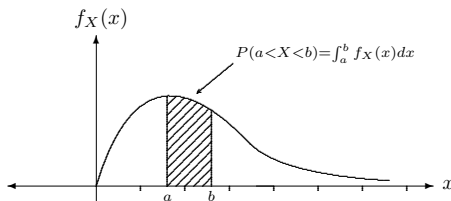
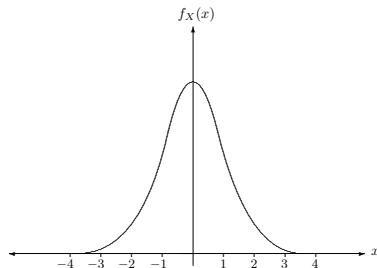


Figure 5:  $P(a < X < b) = \int_a^b f_X(x) dx$ .



# The Normal Distribution

- ▶ A **normal** random variable is a continuous random variable.



The density function for the standard normal distribution with  $\mu = 0$ ,  $\sigma = 1$ .

- ▶ We call a normal random variable with  $\mu = 0$  and  $\sigma^2 = 1$  a **standard normal** random variable.
- ▶ For standard normal random variables, we can use charts (or a computer) to find the area under the density function (i.e. the probabilities).

# Questions

►  $P(Z < -1.75)$

# Questions

►  $P(Z < -1.75)$

$$P(Z < -1.75) = 0.0401$$

# Questions

►  $P(Z < -1.75)$

$$P(Z < -1.75) = 0.0401$$

►  $P(Z > 0.85)$

# Questions

►  $P(Z < -1.75)$

$$P(Z < -1.75) = 0.0401$$

►  $P(Z > 0.85)$

$$P(Z > 0.85) = 1 - 0.8023 = 0.1977$$

# Questions

►  $P(Z < -1.75)$

$$P(Z < -1.75) = 0.0401$$

►  $P(Z > 0.85)$

$$P(Z > 0.85) = 1 - 0.8023 = 0.1977$$

►  $P(-1.43 < Z < 0.92)$

# Questions

►  $P(Z < -1.75)$

$$P(Z < -1.75) = 0.0401$$

►  $P(Z > 0.85)$

$$P(Z > 0.85) = 1 - 0.8023 = 0.1977$$

►  $P(-1.43 < Z < 0.92)$

$$P(-1.43 < Z < 0.92) = 0.8212 - .0764 = .7448$$