STAT 111

Recitation 4

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

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Many Random Variables

Let X and Y be two random variables. Then,

$$mean(X + Y) = mean(X) + mean(Y)$$
.

► If X, Y are also independent,

$$variance(X + Y) = variance(X) + variance(Y).$$

For constants a, b, we have

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

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

$$variance(D) = variance(X) + variance(Y).$$

Many Random Variables

- Let X_1, \ldots, X_n be i.i.d. random variables, each with mean μ and variance σ^2 .
- ▶ Then for the sum, $T_n = X_1 + \cdots + X_n$:

mean of
$$T_n = n\mu$$
, variance of $T_n = n\sigma^2$

▶ For the average, $\bar{X} = \frac{X_1 + \cdots + X_n}{n}$:

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

Q1: 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} .

A1: X_1, X_2, X_3, X_4 , and X_5 are no longer i.i.d.!

$$Mean(T_n) = 450 + 550 + 600 + 550 + 350 = 2500$$

 $Var(T_n) = 1200 + 800 + 500 + 800 + 1200 = 4500$

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

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

- Q2: Let P_1 be the proportion of heads in 50 coin tosses, where P(H) = 0.6. Find $Mean(P_1)$ and $Var(P_1)$.
- A2: $Mean(P_1) = 0.6$ and $Var(P_1) = 0.6 \times 0.4/50 = 0.0048$.
- Q3: 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.
- A3: Mean(D) = 0.6 0.7 = -0.1Var(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).

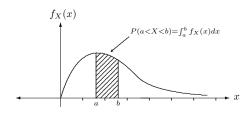
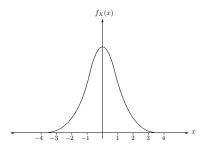


Figure 5: $P(a < X < b) = \int_a^b f_X(x) dx$.

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).

▶
$$P(Z < -1.75)$$

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

P(Z > 0.85)

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

P(-1.43 < Z < 0.92)

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

Standardization

- ▶ What if a normal random variable has a different mean and variance?
- ▶ We need to standardize it.
- Let $X \sim N(\mu, \sigma^2)$: that is, X is a normal random variable with mean μ , variance σ^2 . Let

$$Z = \frac{X - \mu}{\sigma}$$

▶ Then Z is a *standard* normal random variable.

Example

▶ X is a normal random variable with $\mu = 5$ and $\sigma^2 = 9$. Find P(X > 8).

$$P(X > 8) = P\left(\frac{X - 5}{3} > \frac{8 - 5}{3}\right)$$

= $P(Z > 1)$
= 0.16

 $X \sim N(2,16)$. Find P(-1 < X < 6).

$$P(-1 < X < 6) = P\left(\frac{-1-2}{4} < \frac{X-2}{4} < \frac{6-2}{4}\right)$$

$$= P(-0.75 < Z < 1)$$

$$= 0.8413 - 0.2266$$

$$= 0.6147$$

▶ $Y \sim N(7,25)$. Find P(8 < Y < 13).

$$P(8 < Y < 13) = P\left(\frac{8-7}{5} < \frac{Y-7}{5} < \frac{13-7}{5}\right)$$
$$= P(0.2 < Z < 1.2)$$
$$= 0.8849 - 0.5793$$
$$= 0.3056$$

Two-Standard-Deviation Rule

From the chart:

$$P(Z < -1.96) = 0.025, \quad P(Z > 1.96) = 0.025.$$

► Then:

$$P(-1.96 < Z < 1.96) = 0.95.$$

▶ Approximate $1.96 \approx 2$ and "unstandardize":

$$P\left(-2 < \frac{X - \mu}{\sigma} < 2\right) = 0.95$$

 \Rightarrow

$$P(\mu - 2\sigma < X < \mu + 2\sigma) = 0.95.$$

► The probability that a normal random variable is within 2 standard deviations of the mean is 95%.

Normal Distribution: Sums and Averages

 \blacktriangleright Let X_1, \ldots, X_n be independent and normally distributed. Let

$$T_n = X_1 + \cdots + X_n, \qquad \bar{X} = \frac{X_1 + \cdots + X_n}{n}, \qquad D = X_2 - X_1.$$

- ▶ Then T_n , \bar{X} and D are also normal random variables.
- Let $X_1, \ldots, X_n \stackrel{i.i.d}{\sim} N(\mu, \sigma^2)$. Then:

$$T_n \sim N(n\mu, n\sigma^2)$$

$$\overline{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$D \sim N(0, 2\sigma^2)$$

Example

- ➤ Suppose we know the weight *X* of an adult man chosen at random is normally distributed with mean 160 pounds and variance 64 pounds².
- a) Find P(156 < X < 164).

$$P(156 < X < 164) = P\left(\frac{156 - 160}{8} < \frac{X - 160}{8} < \frac{164 - 160}{8}\right)$$
$$= P(-0.5 < Z < 0.5)$$
$$= 0.3830$$

b) Find the probability that the average weight of 16 men chosen at random is between 156 and 164 pounds.

$$\overline{X} \sim N(160, 64/16 = 4)$$
 $P(156 < \overline{X} < 164) = P(-2 < Z < 2)$
 $= 0.95$

Example

- ▶ Suppose we know the weight *X* of an adult man chosen at random is normally distributed with mean 160 pounds and variance 64 pounds².
- c) Calculate the numbers A and B such that $P(A < X < B) \approx 0.95$.

$$0.95 \approx P\left(-2 < \frac{X - \mu}{\sigma} < 2\right) = P(-2\sigma + \mu < X < 2\sigma + \mu)$$

 $A = -2(8) + 160 = 144, \quad B = 2(8) + 160 = 176$

d) Calculate the numbers C and D such that the average of 256 randomly chosen adults is between C and D with probability approximately 0.95.

$$\overline{X}_{256} \sim N(160, 64/256 = 1/4)$$
 $C = -2\sigma + \mu = -2(1/2) + 160 = 159$
 $D = 2\sigma + \mu = 2(1/2) + 160 = 161$