

Lecture 11: Chapter 7, Secs. 2-3

Binomial and Normal Random Variables

- Definition
- What if Events are Dependent?
- Center, Spread, Shape of Counts, Proportions
- Normal Approximation
- Normal Random Variables

1

Looking Back: Review

- 4 Stages of Statistics
 - Data Production (discussed in Lectures 1-3)
 - Displaying and Summarizing (Lectures 3-8)
 - Probability
 - Finding Probabilities (discussed in Lectures 9-10)
 - Random Variables (introduced in Lecture 10)
 - Binomial
 - Normal
 - Sampling Distributions
 - Statistical Inference

2

Definition (Review)

- **Discrete Random Variable:** one whose possible values are finite or countably infinite (like the numbers 1, 2, 3, ...)

Looking Ahead: To perform inference about categorical variables, need to understand behavior of sample proportion. A first step is to understand behavior of sample counts. We will eventually shift from discrete counts to a normal approximation, which is continuous.

3

Definition

- Binomial Random Variable** counts sampled individuals falling into particular category;
- Sample size n is fixed
 - Each selection independent of others
 - Just 2 possible values for each individual
 - Each has same probability p of falling in category of interest

4

Example: *A Simple Binomial Random Variable*

- **Background:** The random variable X is the count of tails in two flips of a coin.
- **Questions:** Why is X binomial? What are n and p ?
- **Responses:**
 - Sample size n fixed?
 - Each selection **independent** of others?
 - Just **2 possible values** for each?
 - Each has **same probability p** ?

Example: *A Simple Binomial Random Variable*

- **Background:** The random variable X is the count of tails in two flips of a coin.
- **Question:** How do we display X ?
- **Response:**

Looking Back: We already discussed and displayed this random variable when learning about probability distributions.

Example: *Determining if R.V. is Binomial*

- **Background:** Consider following R.V.:
 - Pick card from deck of 52, replace, pick another.
 X =no. of cards picked until you get ace.
- **Question:** Is X binomial?
- **Response:**

Example: *Determining if R.V. is Binomial*

- **Background:** Consider following R.V.:
 - Pick 16 cards without replacement from deck of 52. X =no. of red cards picked.
- **Question:** Is X binomial?
- **Response:**

Example: Determining if R.V. is Binomial

- **Background:** Consider following R.V.:
 - Pick 16 cards with replacement from deck of 52. W =no. of clubs, X =no. of diamonds, Y =no. of hearts, Z =no. of spades. Goal is to report how frequently each suit is picked.
- **Question:** Are W , X , Y , Z binomial?
- **Response:**

Example: Determining if R.V. is Binomial

- **Background:** Consider following R.V.:
 - Pick with replacement from German deck of 32 (doesn't include numbers 2-6), then from deck of 52, back to deck of 32, etc. for 16 selections altogether. X =no. of aces picked.
- **Question:** Is X binomial?
- **Response:**

Example: Determining if R.V. is Binomial

- **Background:** Consider following R.V.:
 - Pick 16 cards with replacement from deck of 52. X =no. of hearts picked.
- **Question:** Is X binomial?
- **Response:**
 - fixed $n = 16$
 - selections independent (with replacement)
 - just 2 possible values (heart or not)
 - same $p = 0.25$ for all selections

→ _____

Requirement of Independence

Snag:

- Binomial theory requires independence
- Actual sampling done without replacement so selections are dependent

Resolution: When sampling without replacement, selections are approximately independent if population is at least $10n$.

Example: *A Binomial Probability Problem*

- **Background:** The proportion of Americans who are left-handed is 0.10. Of 46 presidents, 8 have been left-handed (proportion 0.17).
- **Question:** How can we establish if being left-handed predisposes someone to be president?
- **Response:** Determine if 8 out of 46 (0.17) is _____ when sampling at random from a population where 0.10 fall in the category of interest.

Solving Binomial Probability Problems

- Use binomial formula or tables
Only practical for small sample sizes
- Use software
Won't take this approach until later
- Use normal approximation for count X
Not quite: more interested in proportions
- Use normal approximation for proportion
Need mean and standard deviation...

Example: *Mean of Binomial Count, Proportion*

- **Background:** Based on long-run observed outcomes, probability of being left-handed is approx. 0.1. Randomly sample 100 people.
- **Questions:** On average, what should be the
 - count of lefties?
 - proportion of lefties?
- **Responses:** On average, we should get
 - count of lefties _____
 - proportion of lefties _____

Mean and S.D. of Counts, Proportions

Count X binomial with parameters n, p has:

- **Mean** np
- **Standard deviation** $\sqrt{np(1-p)}$

Sample **proportion** $\hat{p} = \frac{X}{n}$ has:

- **Mean** p
- **Standard deviation** $\sqrt{\frac{p(1-p)}{n}}$

Looking Back: Formulas for s.d. require independence: population at least 10n.

Example: Standard Deviation of Sample *Count*

- **Background:** Probability of being left-handed is approx. 0.1. Randomly sample 100 people. Sample **count** has mean $100(0.1) = 10$, standard deviation $\sqrt{100(0.1)(1 - 0.1)} = 3$
- **Question:** How do we interpret these?
- **Response:** On average, expect sample count = ____ lefties.
Counts vary; typical distance from 10 is ____.

Example: S.D. of Sample *Proportion*

- **Background:** Probability of being left-handed is approx. 0.1. Randomly sample 100 people. Sample **proportion** has mean 0.1, standard deviation $\sqrt{\frac{0.1(1-0.1)}{100}} = 0.03$
- **Question:** How do we interpret these?
- **Response:** On average, expect sample proportion = ____ lefties.
Proportions vary; typical distance from 0.1 is ____.

Example: Role of Sample Size in Spread

- **Background:** Consider proportion of tails in various sample sizes n of coinflips.
- **Questions:** What is the standard deviation for
 - $n=1$? $n=4$? $n=16$?
- **Responses:**
 - $n=1$: s.d.=
 - $n=4$: s.d.=
 - $n=16$: s.d.=

A Closer Look: Due to n in the denominator of formula for standard deviation, spread of sample proportion _____ as n increases.

Shape of Distribution of Count, Proportion

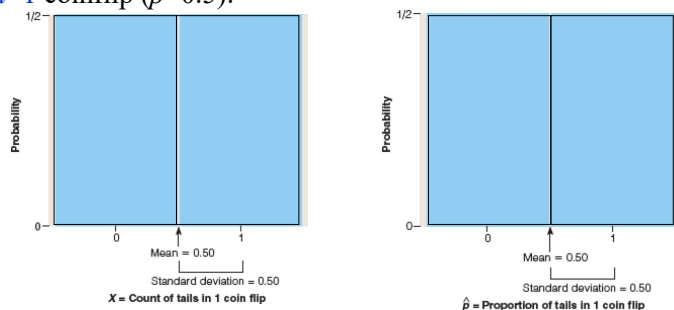
Binomial count X or proportion $\hat{p} = \frac{X}{n}$ for repeated random samples has **shape approximately normal** if samples are large enough to offset underlying skewness.

(Central Limit Theorem)

For a given sample size n , shapes are identical for count and proportion.

Example: Underlying Coinflip Distribution

- Background: Distribution of count or proportion of tails in $n=1$ coinflip ($p=0.5$):

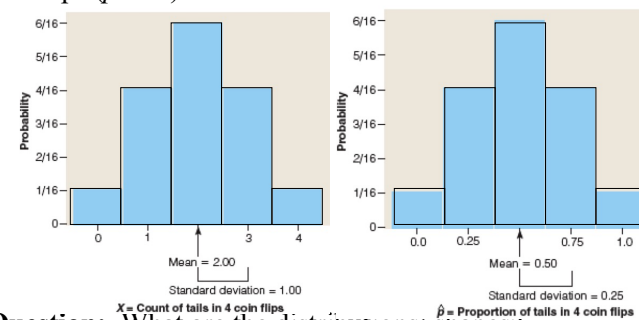


- Question: What are the distributions' shapes?

- Response:

Example: Distribution for 4 Coinflips

- Background: Distribution of count or proportion of tails in $n=4$ coinflips ($p=0.5$):



- Question: What are the distributions' shapes?

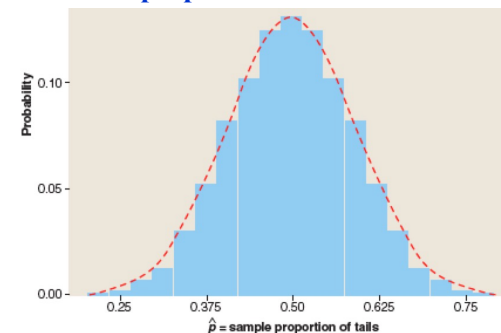
- Response:

Shift from Counts to Proportions

- Binomial Theory begins with counts
- Inference will be about proportions

Example: Distribution of \hat{p} for 16 Coinflips

- Background: Distribution of **proportion** of tails in $n=16$ coinflips ($p=0.5$):

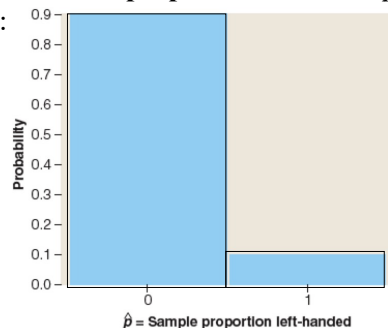


- Question: What is the shape?

- Response:

Example: Underlying Distribution of Lefties

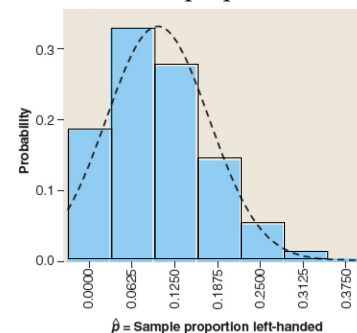
- Background: Distribution of **proportion** of lefties ($p=0.1$) for samples of $n=1$:



- Question: What is the shape?
- Response:

Example: Dist of \hat{p} of Lefties for $n=16$

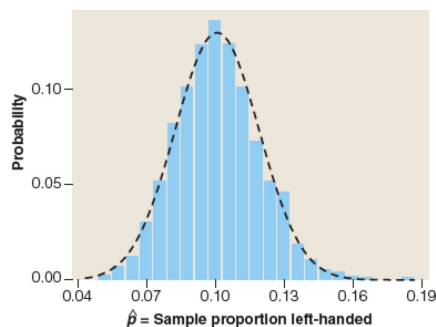
- Background: Distribution of proportion of lefties ($p=0.1$) for $n=16$:



- Question: What is the shape?
- Response:

Example: Dist of \hat{p} of Lefties for $n=100$

- Background: Distribution of proportion of lefties ($p=0.1$) for $n=100$:



- Question: What is the shape?
- Response:

Rule of Thumb: Sample Proportion Approximately Normal

Distribution of \hat{p} is approximately normal if sample size n is large enough relative to shape, determined by population proportion p .

Require $np \geq 10$ and $n(1 - p) \geq 10$

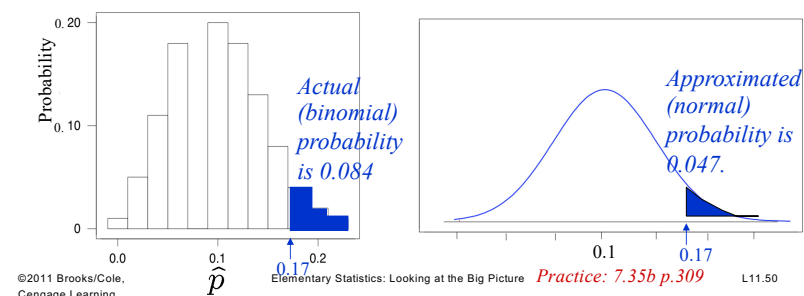
Together, these require us to have larger n for p close to 0 or 1 (underlying distribution skewed right or left).

Example: Applying Rule of Thumb

- **Background:** Consider distribution of sample proportion for various n and p :
 $n=4, p=0.5$; $n=20, p=0.5$; $n=20, p=0.1$; $n=20, p=0.9$; $n=100, p=0.1$
- **Question:** Is shape approximately normal?
- **Response:** Normal?
 - $n=4, p=0.5$ _____ [$np=4(0.5)=2 < 10$]
 - $n=20, p=0.5$ _____ [$np=20(0.5)=10=n(1-p)$]
 - $n=20, p=0.1$ No [_____]
 - $n=20, p=0.9$ No [_____]
 - $n=100, p=0.1$ _____
 $[np=100(0.1)=10, n(1-p)=100(0.9)=90 \text{ both } \geq 10]$

Example: Solving the Left-handed Problem

- **Background:** The proportion of Americans who are lefties is 0.1. Consider $P(\hat{p} \geq 8/46=0.17)$ for a sample of 46 presidents.
- **Question:** Can we use a normal approximation to find the probability that at least 8 of 46 (0.17) are left-handed?
- **Response:**



Example: From Count to Proportion and Vice Versa

- **Background:** Consider these reports:
 - In a sample of 87 assaults on police, 23 used weapons.
 - 0.44 in sample of 25 bankruptcies were due to med. bills
- **Question:** In each case, what are n , X , and \hat{p} ?
- **Response:**
 - First has $n =$ _____, $X =$ _____, $\hat{p} =$ _____
 - Second has $n =$ _____, $\hat{p} =$ _____, $X =$ _____

Looking Back: Review

- **4 Stages of Statistics**
 - Data Production (discussed in Lectures 1-3)
 - Displaying and Summarizing (Lectures 3-8)
 - Probability
 - Finding Probabilities (discussed in Lectures 9-10)
 - Random Variables (introduced in Lecture 10)
 - Binomial (just discussed in Lecture 11)
 - Normal
 - Sampling Distributions
 - Statistical Inference

Role of Normal Distribution in Inference

- **Goal:** Perform inference about unknown **population proportion**, based on **sample proportion**
- **Strategy:** Determine **behavior of sample proportion** in random samples with known population proportion
- **Key Result:** Sample proportion follows **normal** curve for large enough samples.

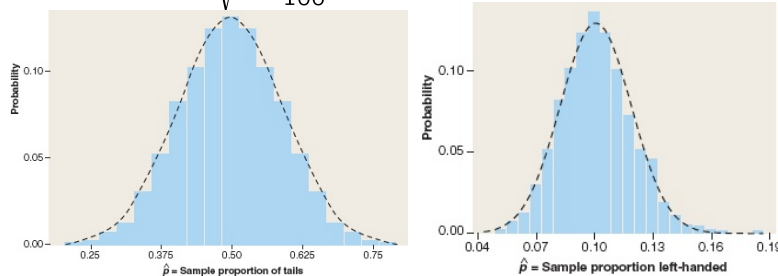
Looking Ahead: Similar approach will be taken with means.

Discrete vs. Continuous Distributions

- **Binomial Count X**
 - **discrete** (distinct possible values like numbers 1, 2, 3, ...)
- **Sample Proportion $\hat{p} = \frac{X}{n}$**
 - **also discrete** (distinct values like count)
- **Normal Approx. to Sample Proportion**
 - **continuous** (follows normal curve)
 - Mean p , standard deviation $\sqrt{\frac{p(1-p)}{n}}$

Sample Proportions Approx. Normal (Review)

- Proportion of **tails** in $n=16$ coinflips ($p=0.5$) has $\mu = 0.5, \sigma = \sqrt{\frac{0.5(1-0.5)}{16}} = 0.125$, shape approx normal
- Proportion of **lefties** ($p=0.1$) in $n=100$ people has $\mu = 0.1, \sigma = \sqrt{\frac{0.1(1-0.1)}{100}} = 0.03$, shape approx normal



Example: Variable Types

- **Background:** Variables in survey excerpt:

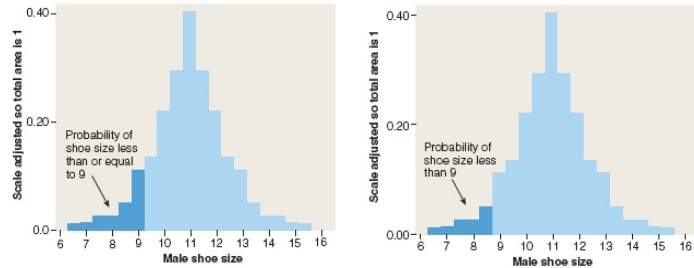
age	breakfast?	comp	credits	...
19.67	no	120	15	
20.08	no	120	16	
19.08	yes	40	14	
...	

- **Question:** Identify type (cat, discrete quan, continuous quan)
 - Age? Breakfast? Comp (daily min. on computer)? Credits?
- **Response:**
 - Age:
 - Breakfast:
 - Comp (daily time in min. on computer):
 - Credits:

Probability Histogram for Discrete R.V.

Histogram for male shoe size X represents probability by area of bars

- $P(X \leq 9)$ (on left)
- $P(X < 9)$ (on right)



For **discrete** R.V., strict inequality or not matters.

©2011 Brooks/Cole,
Cengage Learning

Elementary Statistics: Looking at the Big Picture

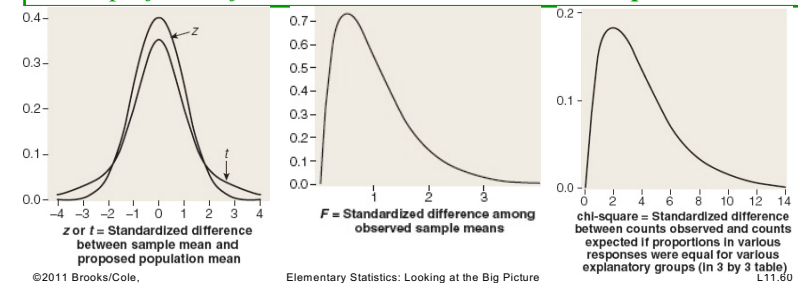
L11.59

59

Definition

Density curve: smooth curve showing prob. dist. of continuous R.V. Area under curve shows prob. that R.V. takes value in given interval.

Looking Ahead: Most commonly used density curve is normal z but to perform inference we also use t , F , and chi-square curves.



©2011 Brooks/Cole,
Cengage Learning

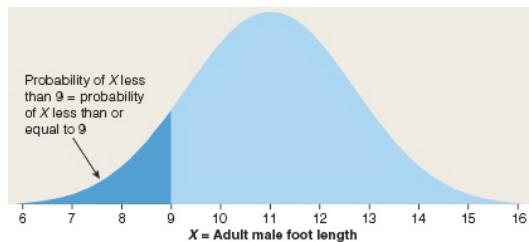
Elementary Statistics: Looking at the Big Picture

L11.60

60

Density Curve for Continuous R.V.

Density curve for male foot length X represents probability by area under curve.



$P(X \leq 9) = P(X < 9)$
Continuous RV: strict inequality or not doesn't matter.

A Closer Look: Shoe sizes are discrete; foot lengths are continuous.

©2011 Brooks/Cole,
Cengage Learning

Elementary Statistics: Looking at the Big Picture

L11.62

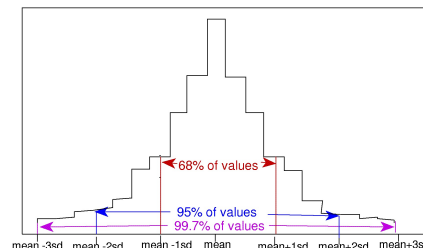
62

68-95-99.7 Rule for Normal Data (Review)

Values of a normal **data set** have

- 68% within 1 standard deviation of mean
- 95% within 2 standard deviations of mean
- 99.7% within 3 standard deviations of mean

68-95-99.7 Rule for Normal Distributions



©2011 Brooks/Cole,
Cengage Learning

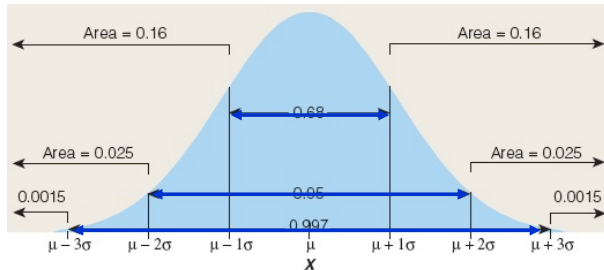
L11.64

64

68-95-99.7 Rule: Normal Random Variable

Sample at **random** from normal **population**; for sampled value X (a R.V.), probability is

- 68% that X is within 1 standard deviation of mean
- 95% that X is within 2 standard deviations of mean
- 99.7% that X is within 3 standard deviations of mean



©2011 Brooks/Cole,
Cengage Learning

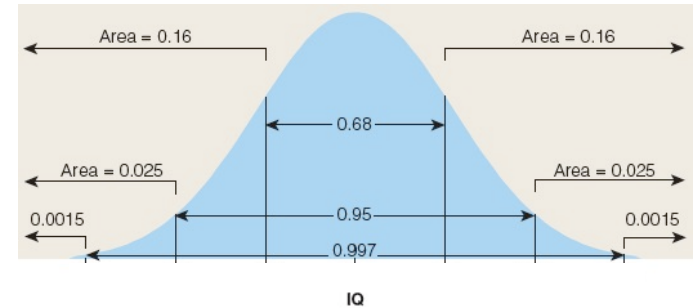
Elementary Statistics: Looking at the Big Picture

L11.65

65

Example: 68-95-99.7 Rule for Normal R.V.

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** What does Rule tell us about distribution of X ?
- **Response:** We can sketch distribution of X :



©2011 Brooks/Cole,
Cengage Learning

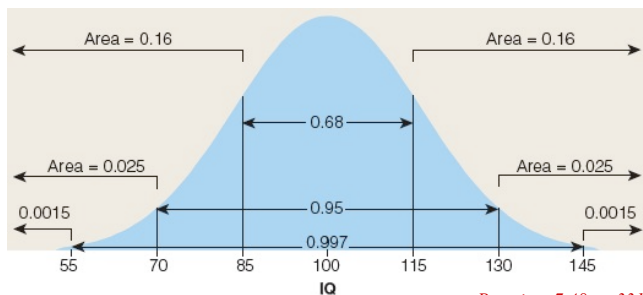
Elementary Statistics: Looking at the Big Picture *Practice: 7.47 p.331*

L11.68

68

Example: Finding Probabilities with Rule

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** Prob. of IQ between 70 and 130 = ?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

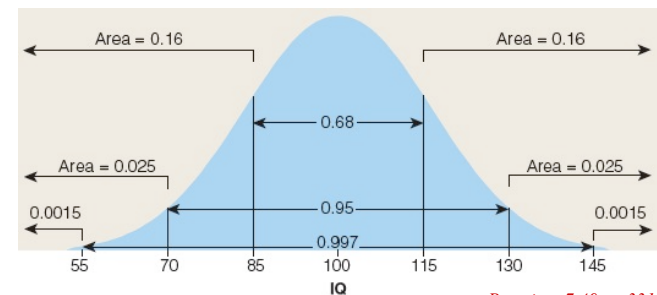
Elementary Statistics: Looking at the Big Picture *Practice: 7.48a p.331*

L11.70

70

Example: Finding Probabilities with Rule

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** Prob. of IQ less than 70 = ?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

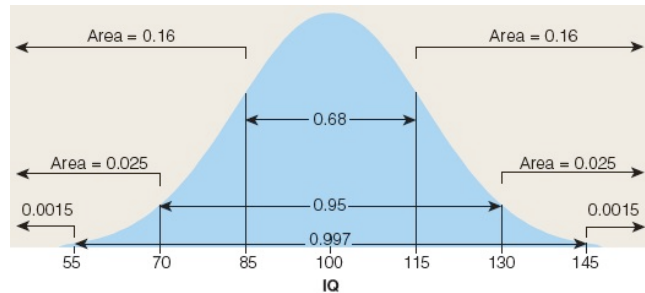
Elementary Statistics: Looking at the Big Picture *Practice: 7.49a p.331*

L11.72

72

Example: Finding *Probabilities* with Rule

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** Prob. of IQ less than 100 = ?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

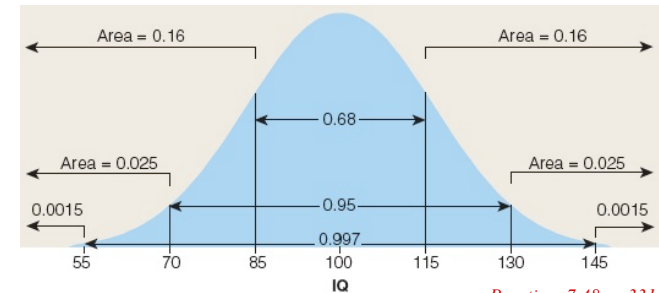
Elementary Statistics: Looking at the Big Picture

L11.74

74

Example: Finding *Values of X* with Rule

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** Prob. is 0.997 that IQ is between...?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

Elementary Statistics: Looking at the Big Picture

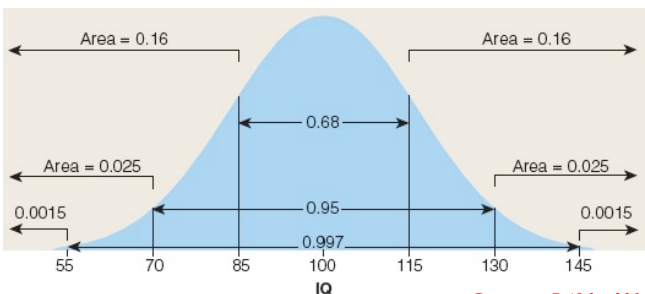
Practice: 7.48c p.331

L11.76

76

Example: Finding *Values of X* with Rule

- **Background:** IQ for randomly chosen adult is normal R.V. X with $\mu = 100$, $\sigma = 15$
- **Question:** Prob. is 0.025 that IQ is above...?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

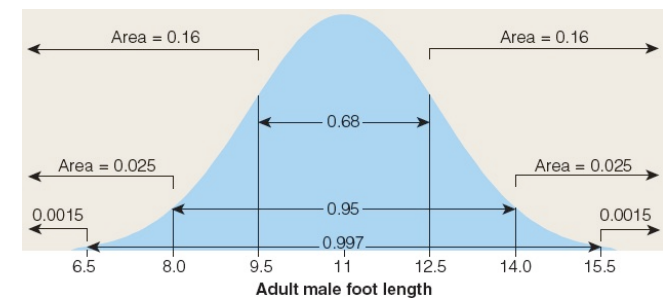
Elementary Statistics: Looking at the Big Picture Practice: 7.48d p.331

L11.78

78

Example: Using Rule to *Evaluate Probabilities*

- **Background:** Foot length of randomly chosen adult male is normal R.V. X with $\mu = 11$, $\sigma = 1.5$ (in.)
- **Question:** How unusual is foot less than 6.5 inches?
- **Response:**



©2011 Brooks/Cole,
Cengage Learning

Elementary Statistics: Looking at the Big Picture

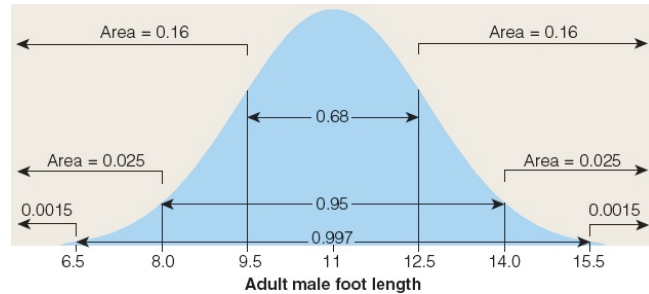
Practice: 7.49b p.331

L11.80

80

Example: Using Rule to *Estimate* Probabilities

- **Background:** Foot length of randomly chosen adult male is normal R.V. X with $\mu = 11$, $\sigma = 1.5$ (in.)
- **Question:** How unusual is foot more than 13 inches?
- **Response:**



©2011 Brooks/Cole, Cengage Learning Elementary Statistics: Looking at the Big Picture Practice: 7.49c-d p.331 L11.82

82

Lecture Summary (Binomial Random Variables)

- Definition; 4 requirements for binomial
- R.V.s that do or don't conform to requirements
- Relaxing requirement of independence
- Binomial counts, proportions
 - Mean
 - Standard deviation
 - Shape
- Normal approximation to binomial

©2011 Brooks/Cole, Cengage Learning

Elementary Statistics: Looking at the Big Picture

L11.83

83

Lecture Summary (Normal Random Variables)

- Relevance of normal distribution
- Continuous random variables; density curves
- 68-95-99.7 Rule for normal R.V.s

©2011 Brooks/Cole, Cengage Learning

Elementary Statistics: Looking at the Big Picture

L11.84

84