

# UNIT 1 FORMULA SHEET

for Sullivan, Statistics: Informed Decisions Using Data@2017 Pearson Education, Inc.

## Chapter 2 Organizing and Summarizing Data

$$\text{Relative Frequency} = \frac{\text{frequency}}{\text{sum of all frequencies}}$$

Class midpoint: The sum of consecutive lower class limits, divided by 2.

## Chapter 3 Numerically Summarizing Data

### Listed Data

Mean

$$\text{Population: } \mu = \frac{\sum x_i}{N}$$

$$\text{Sample: } \bar{x} = \frac{\sum x_i}{n}$$

Standard Deviation

$$\text{Population: } \sigma = \sqrt{\frac{\sum (x_i - \mu)^2}{N}}$$

$$\text{Sample: } S = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n-1}}$$

Variance = standard deviation squared  $\sigma^2$  or  $S^2$

Range = Largest Data Value – Smallest Data Value

### Grouped Data (frequency distribution)

Mean

$$\text{Population: } \mu = \frac{\sum x_i f_i}{\sum f_i}$$

$$\text{Sample: } \bar{x} = \frac{\sum x_i f_i}{\sum f_i}$$

$$\text{Weighted: } \bar{x}_w = \frac{\sum w_i x_i}{\sum w_i}$$

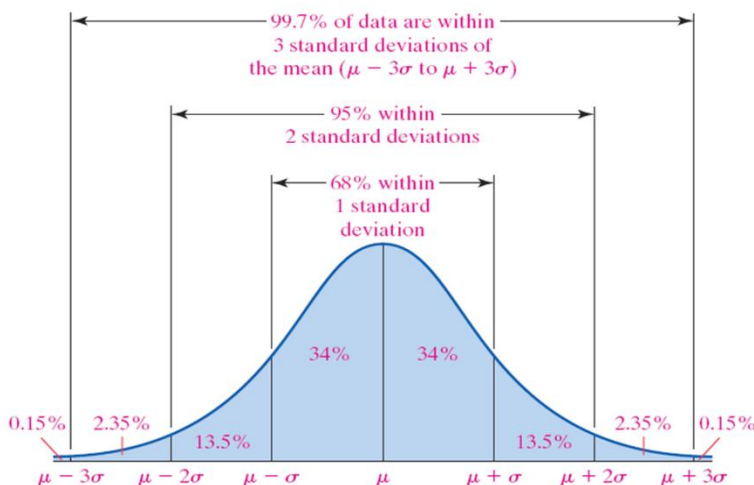
Standard Deviation

$$\text{Population: } \sigma = \sqrt{\frac{\sum (x_i - \mu)^2 f_i}{\sum f_i}}$$

$$\text{Sample: } S = \sqrt{\frac{\sum (x_i - \bar{x})^2 f_i}{(\sum f_i) - 1}}$$

$$\text{Class width} = \frac{\text{range}}{\text{number of classes}} \text{ (round up)}$$

Empirical Rule: If the shape of the distribution is bell-shaped, then:



z-score (standardized test statistic):

$$\text{Population: } Z = \frac{x - \mu}{\sigma}$$

$$\text{Sample: } Z = \frac{x - \bar{x}}{s}$$

Interquartile Range:  $IQR = Q_3 - Q_1$

Lower and Upper Fences:

Lower:  $Q_1 - 1.5(IQR)$

Upper:  $Q_3 + 1.5(IQR)$

Five-Number Summary:

Minimum,  $Q_1$ ,  $M$ ,  $Q_3$ , Maximum

## Chapter 4 Describing the Relation between Two Variables

Residual = observed  $y$  – predicted  $y$ :  $y - \hat{y}$

$$\text{Correlation Coefficient: } r = \frac{\sum \left( \frac{x_i - \bar{x}}{s_x} \right) \left( \frac{y_i - \bar{y}}{s_y} \right)}{n-1}$$

The equation of the least-squares regression line is  $\hat{y} = b_1 x + b_0$ , where  $\hat{y}$  is the predicted value,

$b_1 = r \cdot \frac{s_y}{s_x}$ , is the slope, and  $b_0 = \bar{y} - b_1 \bar{x}$  is the intercept.

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### Chapter 5 Probability

Empirical Probability:

$$P(E) = \frac{\text{frequency of } E}{\text{number of trials of the experiment}}$$

Classical Probability:

$$P(E) = \frac{N(E)}{N(S)} \\ = \frac{\text{number of ways that } E \text{ can occur}}{\text{number of possible outcomes}}$$

Compliment Rule:  $P(E^c) = 1 - P(E)$

Multiplication Rule of Counting: the number of ways for a series of events ( $E_1, E_2, E_3, \dots$ ) to occur in order  
 $= (n(E_1))(n(E_2))(n(E_3)) \dots$

Factorial:  $n! = n(n-1)(n-2) \dots (3)(2)(1)$

Permutation of  $n$  objects taken  $r$  at a time:

$${}_nP_r = \frac{n!}{(n-r)!}$$

Combination of  $n$  objects taken  $r$  at a time:

$${}_nC_r = \frac{n!}{r!(n-r)!}$$

Permutations with Repetition:

$$\frac{n!}{n_1! n_2! \dots n_k!}$$

Addition Rule for Disjoint Events:

$$P(E \text{ or } F) = P(E) + P(F)$$

Addition Rule for  $n$  Disjoint Events:

$$P(E \text{ or } F \text{ or } G \text{ or } \dots) = P(E) + P(F) + P(G) + \dots$$

General Addition Rule:

$$P(E \text{ or } F) = P(E) + P(F) - P(E \text{ and } F)$$

Multiplication Rule for Independent Events:

$$P(E \text{ and } F) = P(E) \bullet P(F)$$

Multiplication Rule for  $n$  Independent Events:

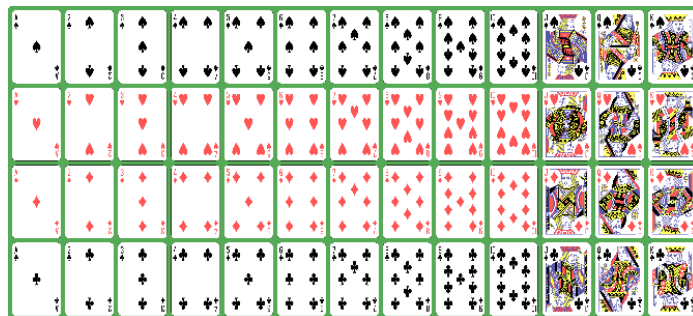
$$P(E \text{ and } F \text{ and } G \dots) = \\ P(E) \bullet P(F) \bullet P(G) \bullet \dots$$

General Multiplication Rule:

$$P(E \text{ and } F) = P(E) \bullet P(F/E)$$

Conditional Probability Rule:

$$P(F / E) = \frac{P(E \text{ and } F)}{P(E)} = \frac{N(E \text{ and } F)}{n(E)}$$



### Chapter 6 Discrete Probability Distributions

Mean (Expected Value) of a Discrete Random Variable:

$$\mu_x = \Sigma [xP(x)]$$

Standard Deviation of a Discrete Random Variable:

$$\sigma_x = \sqrt{\Sigma [(x - \mu)^2 P(x)]}$$

### Chapter 7 The Normal Distribution

Standardizing a Normal Random Variable:

$$z = \frac{x - \mu}{\sigma}$$

Finding the score (cut off for a given percent):

$$x = \mu + z\sigma$$

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### Chapter 8 Sampling Distributions

#### Sampling distribution of $\hat{p}$

Sample proportion  $\hat{p} = \frac{x}{n}$

$$\mu_{\hat{p}} = p$$

$$\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

$$Z = \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}}$$

$$n \leq 0.05N \text{ and } np(1-p) \geq 10$$

#### Sampling distribution of $\bar{x}$

$$\mu_{\bar{x}} = \mu$$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

$$Z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$$

Underlying distribution is normal OR  $n \geq 30$

$$\bar{x} = \mu_{\bar{x}} + Z \sigma_{\bar{x}}$$

### Chapter 9 Estimating the Value of a Parameter

#### Confidence Intervals

##### Proportion

$(1-\alpha)*100\%$  confidence interval about  $p$  is:

$$\hat{p} \pm E$$
$$E = z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

Sample size to estimate the population proportion with a margin of error  $E$

- $\hat{p}$  is a prior estimate of the population proportion:

$$n = \hat{p}(1-\hat{p}) \left( \frac{z_{\alpha/2}}{E} \right)^2 \text{ rounded up to the next integer}$$

- No prior estimate is available:

$$n = 0.25 \left( \frac{z_{\alpha/2}}{E} \right)^2 \text{ rounded up to the next integer}$$

#### Confidence Intervals

##### Mean

$(1-\alpha)*100\%$  confidence interval about  $\mu$  is:

$$\bar{x} \pm E$$
$$E = t_{\alpha/2} \frac{s}{\sqrt{n}}$$

Sample size to estimate the population mean with a margin of error  $E$

- $n = \left( \frac{z_{\alpha/2} * s}{E} \right)^2$  rounded up to the next integer

### Chapter 10 Hypothesis Tests Regarding a Parameter

#### Proportion

$$Z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

#### Mean

$$t = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}}$$