Class 8

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Be The Difference.

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Agenda:

Review Chapters 1 – 5 (Exam 1 Chapters)

Just the highlights!

1. Summation Notation

$$
\sum_{i=1}^{n} f(x_i) = f(x_1) + f(x_2) + \dots + f(x_n)
$$

2. Factorials $n! = n \times (n-1) \times (n-2) \times \cdots \times 2 \times 1$

3. Computations

x=20, y=14, s=16, w=-2, m=15, n=10
Compute
$$
x+y \cdot \frac{\sqrt{s}}{n} = 25.6
$$

4. Simple Linear Equations

$$
2 - 2x = 3x + 3 \qquad x = -1/5
$$

1.1 Americans Here's Looking at you

Statistics is all around us!

How much time between Internet usage?

Fretting Over Messages

Figure from Johnson & Kuby, 2012.

Source: Impulse Research for Qwest Communications online survey of 1,063 adult Wi-Fi users in April 2009.

1.1 What is Statistics?

Population: A collection, or set, of individuals, objects, or events whose properties are to be analyzed.

Sample: Subset of the population.

Variable: A characteristic of interest about each individual element of a population or sample.

Data value: The value of the variable associated with one element of a population or sample.

Parameter: A numerical value summarizing all the data of an entire population.

Statistic: A numerical value summarizing the sample data.

1.1 What is Statistics?

Data: The set of values collected from the variable from each of the elements that belong to the sample.

- **1.1 What is Statistics?**
	- **Qualitative variable:** A variable that describes or categorizes an element of a population.
	- **Nominal variable:** A qualitative variable that characterizes an element of a population. No ordering. No arithmetic.
	- **Ordinal variable:** A qualitative variable that incorporates an ordered position, or ranking.
	- **Quantitative variable:** A variable that quantifies an element of a population.
	- **Discrete variable:** A quantitative variable that can assume a countable number of values. Gap between successive values.
	- **Continuous variable:** A quantitative variable that can assume an uncountable number of values. Continuum of values.

2: Descriptive Analysis and Single Variable Data 2.1 Graphs - Qualitative Data

Circle (pie) graphs and bar graphs:

Circle is parts to whole as angle.

Bar graph is amount in each category as rectangular areas.

Figures from Johnson & Kuby, 2012.

2: Descriptive Analysis and Single Variable Data 2.2 Frequency Distributions and Histograms

Statistics Exam Scores [TA02-06]

Figures from Johnson & Kuby, 2012.

2: Descriptive Analysis and Single Variable Data 2.3 Measures of Central Tendency

Sample Mean: Usual average, p. 63 **Sample Median:** Middle value, p. 64 *n* odd, $\tilde{x} = \frac{\tilde{x} - \tilde{x}}{2}$ value $n \text{ even, avg } \frac{-}{2}$ & values **Sample Mode:** Most often, p. 66 1 1 *n i i* $x = - \sum x$ $n_{\;\;\overline{i=}}$ = $=\frac{1}{\sqrt{2}}\sum$ 1 2 *n* + 2 *n* 1 2 *n x* $\, +$ = $\hat{\mathcal{X}}$ $\hat{\kappa} = \text{most often}$

Measures of central tendency characterize center of distribution.

Measures of dispersion characterize the variability in the data.

2: Descriptive Analysis and Single Variable Data 2.4 Measures of Dispersion

Range: $H-L$, p. 74

Deviation from mean: value minus sample mean, p. 74

$$
i^{th}
$$
 deviation from mean = $x_i - \overline{x}$

Sample Variance: avg squared dev using *n*-1 in den, p. 76

$$
s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \overline{x})^{2} \qquad s^{2} = \frac{1}{n-1} \left\{ \sum_{i=1}^{n} x_{i}^{2} - \left[\left(\sum_{i=1}^{n} x_{i} \right)^{2} / n \right] \right\}
$$

Sample Standard Deviation: $s = \sqrt{s^2}$ $s = \sqrt{s}$

2: Descriptive Analysis and Single Variable Data 2.3, 2.4 Measures of Central Tendency and Dispersion

Example: Data values: 1,2,2,3,4

$$
\overline{x} = 2.4 \qquad \hat{x} = 2 \qquad \quad \tilde{x} = 2
$$

$$
s^2 = 1.3 \t s = 1.1
$$

$$
\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i
$$
 $\hat{x} = \text{most often value}$ $\tilde{x} = \text{middle value}$

$$
s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \overline{x})^{2}
$$
 $s = \sqrt{s^{2}}$

 \blacktriangleleft

2: Descriptive Analysis and Single Variable Data 2.5 Measures of Position

Measures of Position: Quartiles - ranked data into quarters

L = lowest value $H =$ highest value Q_2 = median Q_1 = 25% smaller *Q³* = 75% smaller $IQR = Q_3 - Q_1$

2: Descriptive Analysis and Single Variable Data 2.5 Measures of Position

Measures of Position: percentiles - rank data into 100ths

L = lowest value $H =$ highest value P_k = value where $k\%$ are smaller

rank data

p^k halfway between value and next one average of A^{th} and $(A+1)^{th}$ values

 $p_k^{}$ is value in next largest position, $B\texttt{+1}$ value

Figure from Johnson & Kuby, 2012.

2: Descriptive Analysis and Single Variable Data 2.5 Measures of Position

Standard score, or z-score: The position a particular value of *x* has relative to the mean, measured in standard deviations.

$$
z_i = \frac{i^{\text{th}} \text{ value - mean}}{\text{std. dev.}} = \frac{x_i - \overline{x}}{s}
$$

There can be *n* of these because we have $x_1, x_2, ..., x_n$.

3: Descriptive Analysis and Bivariate Data

3.1 Bivariate Data: two qualitative

Cross-tabulation tables or **contingency tables**

Example: Construct a 2×3 table. Know different %ages.

 $M = ma$ le $F =$ female $LA = liberal$ arts $BA =$ business admin $T = 1$ technology

Figures from Johnson & Kuby, 2012.

3: Descriptive Analysis and Bivariate Data

3.1 Bivariate Data: one qualitative and one quantitative

Example:

Figures from Johnson & Kuby, 2012.

- **3: Descriptive Analysis and Bivariate Data**
- **3.1 Bivariate Data: two quantitative, Scatter Diagram**

Example: Push-ups

Input variable: independent variable, *x*. **Output variable:** dependent variable, *y*.

Scatter Diagram: A plot of all the ordered pairs of bivariate data on a coordinate axis system.

(*x*,*y*) ordered pairs.

Figures from Johnson & Kuby, 2012.

3: Descriptive Analysis and Bivariate Data 3.2 Linear Correlation

Linear Correlation, *r*, is a measure of the strength of a linear relationship between two variables *x* and *y*.

Figure from Johnson & Kuby, 2012.

3: Descriptive Analysis and Bivariate Data

3.2 Linear Correlation

Figure from Johnson & Kuby, 2012.

351

n

 $\sum x_i =$

3: Descriptive Analysis and Bivariate Data

3.2 Linear Correlation

Example:
$$
\frac{\text{Fushups, } \times 27}{\text{Sibups, } \times 30} \times \frac{22}{26} = \frac{15}{25} = \frac{35}{42} = \frac{30}{38} = \frac{55}{40} = \frac{40}{40} = \frac{40}{32} = \frac{1}{27} = 13717
$$
\n
$$
SS(x) = \sum_{i=1}^{n} x_i^2 - \frac{1}{n} \left(\sum_{i=1}^{n} x_i \right)^2 = 13717 - \frac{(351)^2}{10} = 1396.9 \qquad \sum_{i=1}^{n} y_i = 380
$$
\n
$$
SS(y) = \sum_{i=1}^{n} y_i^2 - \frac{1}{n} \left(\sum_{i=1}^{n} y_i \right)^2 = 15298 - \frac{(380)^2}{10} = 858.0 \qquad \sum_{i=1}^{n} y_i^2 = 15298
$$
\n
$$
\sum_{i=1}^{n} x_i y_i = 14257
$$
\n
$$
SS(xy) = \sum_{i=1}^{n} x_i y_i - \frac{1}{n} \left(\sum_{i=1}^{n} x_i \right) \left(\sum_{i=1}^{n} y_i \right) = 14257 - \frac{(351)(380)}{10} = 919.0
$$
\n
$$
r = \frac{SS(xy)}{\sqrt{SS(x)SS(y)}} = \frac{919.0}{\sqrt{(1396.9)(858.0)}} = 0.84
$$
\nFigure from Johnson & Kuby, 2012.

3: Descriptive Analysis and Bivariate Data 3.3 Linear Regression

 $b_{\rm o}$ is estimated y-intercept $b_{\!\scriptscriptstyle 1}$ is estimated slope.

Move line until sum of the squared residuals is a minimum.

3: Descriptive Analysis and Bivariate Data 3.3 Linear Regression 1 *n* **1** *n n n n*

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4: Probability 4.1 Probability of Events

An **experiment** is a process by which a measurement is taken or observations is made. i.e. flip coin or roll die

An **outcome** is the result of an experiment. i.e. Heads, or 3

Sample space is a listing of possible outcomes. i.e. *S*={*H*,*T*}

An **event** is an outcome or a combination of outcomes. i.e. even number when rolling a die

4: Probability

4.1 Probability of Events *Aⁱ* are events

Property 1:
$$
0 \le P(A_i) \le 1
$$

i = 1,...,*n*
Property 2: $\sum_{i=1}^{n} P(O_i) = 1$

Approaches to probability. *Oⁱ* are outcomes

(1) Empirical (AKA experimental)

number of times A occured empirical probability of number of trials *A* $A =$

(2) Theoretical (AKA classical or equally likely)

theoretical probability of $A = \frac{\text{number of times } A \text{ occurs in sample space}}{A \cdot A \cdot A}$ number of elements in the sample space $A =$

4: Probability - Empirical

4.1 Probability of Events – Law of large numbers

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4: Probability 4.1 Probability of Events So let's flip a coin three times.

Can flip three times.

${S} = \{HHH, HHT, HTH, HTT,$, , , } *THH THT TTH TTT* **Sample space:** listing of outcomes for 3 flips

$$
P(HHH) = \frac{\text{\# times }HHH \text{ occurs in } S}{\text{\# elements in } S}
$$

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P(Heart and Ace) = $1/52$

Figure from Johnson & Kuby, 2012.

P(Heart or Ace) = P (Heart) + P (Ace) – P (Heart and Ace)

 \overline{r} *P*(Heart or Ace) = 13 / 52 + 4 / 52 – 1 / 52 = 16 / 52

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4: Probability 4.3 Rules of Probability

Example: Pick Card, *A*=Heart, *B*=Ace $(A | B) = \frac{P(A \text{ and } B)}{P(B)}$ $P(B)$ $P(A | B) = \frac{P(A \text{ and } B)}{P(B | B)}$ *B* = $P(\overline{A}) = 1 - P(A)$ $P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$ $P(A \text{ and } B) = P(A) \cdot P(B | A)$

Figure from Johnson & Kuby, 2012.

4: Probability 4.5 Independent Events

Independent events: … the occurrence or nonoccurence of one gives no information about … occurrence for the other. $P(A) = P(A | B) = P(A | \text{not } B)$

Dependent events: … occurrence of one event does have an effect on the probability of occurrence of the other event. $P(A) \neq P(A | B)$

Special multiplication rule:

Let *A* and *B* … independent events … in a sample space *S*. $P(A \text{ and } B) = P(A) \cdot P(B)$

5: Probability Distributions (Discrete Variables) 5.1 Random Variables

Random Variables: A variable that assumes a unique numerical value for each of the outcomes in the sample space of a probability experiment.

Example:

Let $x =$ the number of heads when we flip a coin twice.

 $x=[0,1,2]$ \longleftarrow numerical values $\{TT,TH,HT,HH\}$ \longleftarrow outcomes in sample space for

5: Probability Distributions (Discrete Variables) 5.1 Random Variables

Discrete Random Variables: A quantitative random variable that can assume a countable number of values.

Continuous Random Variable: A quantitative random variable that can assume an uncountable number of values.

Continuum of values.

Examples: Discrete: Number of heads when we flip a coin ten times.

Continuous: Distance from earth center to sun center.

5: Probability Distributions (Discrete Variables) 5.2 Probability Distributions of a Discrete Random Variable

Random Variables: … assumes a unique … value for each of the outcomes in the sample space … .

Probability Function: A rule *P*(*x*) that assigns probabilities to the values of the random variable *x*.

n

5: Probability Distributions (Discrete Variables) 5.2 Mean and Variance of a Discrete Random Variable

$$
\mu = \sum_{i=1}^{n} [x_i P(x_i)] = x_1 P(x_1) + x_2 P(x_2) + ... + x_n P(x_n)
$$

For the # of *H* when we flip a coin twice discrete distribution:

$$
\mu = (x_1) \cdot P(x_1) + (x_2) \cdot P(x_2) + (x_3) \cdot P(x_3)
$$

\n
$$
\mu = (0) \cdot P(0) + (1) \cdot P(1) + (2) \cdot P(2)
$$

\n
$$
\mu = (0) \cdot (1/4) + (1) \cdot (1/2) + (2) \cdot (1/4)
$$

\n
$$
\mu = 0 + 1/2 + 1/2
$$

$$
\mu = 1
$$

5: Probability Distributions (Discrete Variables) 5.2 Mean and Variance of a Discrete Random Variable

$$
\sigma^2 = \sum_{i=1}^n \left[(x_i - \mu)^2 P(x_i) \right] = (x_1 - \mu)^2 P(x_1) + (x_2 - \mu)^2 P(x_2) + \dots + (x_n - \mu)^2 P(x_n)
$$

For the # of *H* when we flip a coin twice discrete distribution:

$$
\sigma^{2} = (x_{1} - \mu)^{2} \cdot P(x_{1}) + (x_{2} - \mu)^{2} \cdot P(x_{2}) + (x_{3} - \mu)^{2} \cdot P(x_{3}) \xrightarrow[\alpha]{\alpha} \frac{\mu}{x} = 1
$$

\n
$$
\sigma^{2} = (0 - 1)^{2} \cdot P(0) + (1 - 1)^{2} \cdot P(1) + (2 - 1)^{2} \cdot P(2) \xrightarrow[\alpha]{\alpha} \frac{1}{4}
$$

\n
$$
\sigma^{2} = (-1)^{2} \cdot (1/4) + (0)^{2} \cdot (1/2) + (1)^{2} \cdot (1/4) \xrightarrow[\alpha]{\alpha} \frac{1}{2}
$$

\n
$$
\sigma^{2} = 1/4 + 0 + 1/4 \xrightarrow[\alpha]{\alpha} \frac{1}{2} \cdot P(x_{2})
$$

\n
$$
\sigma^{2} = 1/2 \rightarrow \sigma \approx 0.7071 \xrightarrow[\alpha_{3}]{\alpha} \frac{1}{2} \begin{vmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{vmatrix}
$$

\n
$$
\sigma_{1} = \frac{1}{2}
$$

n

5: Probability Distributions (Discrete Variables) 5.2 Mean and Variance of a Discrete Random Variable

$$
\sigma^{2} = \sum_{i=1}^{n} [x_{i}^{2}P(x_{i})] - \mu^{2} = [x_{1}^{2}P(x_{1}) + x_{2}^{2}P(x_{2}) + ... + x_{n}P(x_{n})] - \mu^{2}
$$

Alternate Formula

For the # of *H* when we flip a coin twice discrete distribution:

$$
\sigma^{2} = x_{1}^{2} \cdot P(x_{1}) + x_{2}^{2} \cdot P(x_{2}) + x_{3}^{2} \cdot P(x_{3}) - \mu^{2}
$$

\n
$$
\sigma^{2} = [0^{2} \cdot P(0) + 1^{2} \cdot P(1) + 2^{2} \cdot P(2)] - 1^{2}
$$

\n
$$
\sigma^{2} = [0 \cdot (1/4) + 1^{2} \cdot (1/2) + 2^{2} \cdot (1/4)] - 1
$$

\n
$$
\sigma^{2} = 0 + 1/2 + 4/4 - 1
$$

 $\sigma^2 = 1/2 \rightarrow \sigma \approx 0.7071$

5: Probability Distributions (Discrete Variables) 5.3 The Binomial Probability Distribution

Bi means two like bicycle

An experiment with only two outcomes is called a Binomial exp. Call one outcome *Success* and the other *Failure*.

Each performance of expt. is called a trial and are independent.

n = number of trials or times we repeat the experiment. *x* = the number of successes out of *n* trials. $p =$ the probability of success on an individual trial. $\binom{n}{x} = \frac{n!}{x!(n-x)}$ $\left(x\right) \quad x!(n-$

Rowe, D.B.

!

n n

 $!(n - x)!$

5: Probability Distributions (Discrete Variables) 5.3 The Binomial Probability Distribution

Flip coin ten times.

5: Probability Distributions (Discrete Variables) 5.3 The Binomial Probability Distribution

page 660 page 713

$$
n=10, x=7, p=1/2
$$

$$
P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}
$$

Binomial Probabilities $[(\begin{smallmatrix} n \\ X \end{smallmatrix}) \cdot p^x \cdot q^{n-x}]$ (continued)

Figure from Johnson & Kuby, 2012.

Questions?

Exam 1 on Tuesday! Bring pencil, calculator, scratch paper. (And caffeinated beverage!)

Will be given exam, formula sheet, binomial tables.