### **STAT206**

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- 1. Descriptive and Inferential Statistics Terminology
- (1) Statistics: Science of conducting studies to collect, organize, summarize, analyze and draw conclusions from data.
- (2) variable:attribute can be diff values
- (3) Categorical variables (Gender), Quantitative variables (age weight)
- (4) Discrete var: values that can be counted
- (5) continuous var: assume all values between any two specific values, measured
- (6) random: vals determined by chance
- (7) Data set:collection of data values
- (8) descriptive statistics: describe situation in data ( collect, organize summarize, present data)
- (9) inferential statistics: make inference from samples to population
- (10) population: all indviduals under study
- (11) sample group selected
- (12) statistical hyp: claim
- (13) hypothesis testing: decision making process for evaluating a claim about population from samples

study samples of pop and use inferential stats to make inference about pop

1.1. **Freq distribution and graphs.** raw data in original form organized in some way (i.e. freq distribution). can then be presented in graphs or charts

freq. dist organizes raw data into classes and frequencies.

range of data is large (lower class limit, upper class limit, class boundaries, class width, class midpoint, open-ended distribution)

1.2. General rules for grouped freq. dist. 5-20 classes, class width should be odd to ensure midpoints are integers, mutually exclusive(non overlapping), continuos, exhaustive, equal in width

### 1.3. general procedure.

- (1) determine classes
  - (a) find highest(H) and lowest(L) values,
  - (b) range = H L
  - (c) number(N) of classes desired
  - (d) width = R/N (round to whole)

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- (e) lowest class limit, convenient value less than or equal to the smallest data value.
- (f) add W to get lowr limit of next class
- (g) class boundaries
- (2) tally data
- (3) find freq. of tallies
- (4) cumulative freq
- 1.4. **Histogram.** histogram uses vertical bars to rep. freq of classes

different shapes bell shaped, uniform, j-shaped reverse j-shaped, right and left skewed, bimodal, u-shaped

1.5. **Stem and leaf plots.** uses part of data value as stem and part as leaf to form groups more informative because shows values

first arrange data in order, separate into classes then plot look for peaks and gaps, shape of distribution, variability by looking at spread can put back to back to compare with other data set

1.6. **Descriptive Measures.** describe data with numerical measures

characteristic or measure from population is a parameter, from sample data is called a statistic

Measures of central tendency(mean median, mode) don't round until done

1.6.1. Average. sum divided by total

$$\mu = \bar{X} = \frac{\sum X}{N}$$

rounding rule: final answer should be rounded to one or more decimal place than original finding mean from grouped frequency distribution

(1) 
$$\bar{X} = \frac{\sum f \cdot X_m}{n}$$

f is frequency and  $X_m$  are midpoints of class

- 1.6.2. *Median.* ordered so called data array half smaller half larger, position  $\frac{n+1}{2}$  when even number of values, b/w two given data values
- 1.6.3. *Mode.* value that appears the most frequently data set can have multiple or no modes

## 1.6.4. properties of central tendency.

- (1) mean needs all values of data, varies less than media and mode, unique, cannot be computed for open-ended freq. distribution, highly affected by outliers
- (2) median find middle, whether data falls into upper or lower half of the distribution, can be calc for open data dist. less affected by outlier
- (3) most typical case req., easier to compute, can be computed for categorical or nominal data, not unique

### 2. Variation

## 2.1. Measure of Variation. range is simple measure of variability

**Variance** is measure of variability that uses all the data points, avg. deviation of values from mean.

population variance  $\sigma^2$  is avg. of squared deviations of values from population mean  $\mu$ 

(2) 
$$\sigma^2 = \frac{\sum (X - \mu)^2}{N}$$

where N represents total no. of values in pop

**sample variance** denoted by  $s^2$  sum of squared deviations of the values from sample mean  $\bar{X}$  divided by (n-1)

(3) 
$$s^2 = \frac{\sum (X - \bar{X})^2}{n - 1}$$

where n is no. values in sample

divided by n-1 b/c unbiased estimate of  $\sigma^2$ . standard deviation is square root of variance poulation standard deviation

(4) 
$$\sigma = \sqrt{\sigma^2} = \sqrt{\frac{\sum (X - \bar{\mu})^2}{n - 1}}$$

sample standard deviation

(5) 
$$s = \sqrt{s^2} = \sqrt{\frac{\sum (X - \bar{X})^2}{n - 1}}$$
$$s^2 = \frac{\sum X^2 - \frac{(\sum X)^2}{n}}{n - 1}$$
$$s^2 = \frac{\sum f \cdot X_m^2 - \frac{(\sum f \dot{X}_m)^2}{n}}{n - 1}$$

### 2.2. **Note.**

- (1) value of  $s^2$  always +
- (2) sum of  $(X \bar{X})$  is always 0
- (3) larger value of  $s^2$  or s, larger variability of data (same with  $\sigma$ )

# 3. Empirical(Normal) Rule

distro. bell shaped

some % of vals fall within 1 standard deviation of mean (X-nS, X+nS)

#### 4. Chebyshev's Theorem

in an dist regardl of shape, proportion of value that fall within k standard deviations will be at least  $1 - \frac{1}{k^2}$  where k > 1

subtract average from larger value, divide diff by standard deviation to get k, use cheb to get %

## 5. Measure of Position

standard score / z-score subtract mean from observation and divide result by standard deviation  $z = \frac{X - \dot{\bar{X}}}{s} or \frac{X - \mu}{\sigma}$  number of standard deviations falls above or below mean

## 5.1. Partition Values.

- 5.1.1. percentile. 100 equal groups
- 5.1.2. decile. 10 equal groups
- 5.1.3. quartiles. 4 equal groups
- 5.2. **Percentile Equations.** find the percentile of a piece of data

(6) 
$$Percentile = \frac{number of datavalue below X + 0.5}{n} * 100\%$$

find the position of a percentil

$$(7) c = \frac{n \cdot p}{100}$$

if c is not a whole number round to next whole number, and position is the number. if c is a whole number average c and c+1. this is position of X

### 6. Box-Plots

show distrivution of data

helpful for finding outliers in the data based on the min and max relative the  $Q_1$  and  $Q_3$  measurements

modified box plots dont show outliers

### 7. Product rule

total number of possibilities are  $k_1 \cdot k_2 \cdot \dots k_n$ 

### 8. PERMUTATION RULE

permutation: arrangement of n objects in specific order

Rule 1: arrangement of n objects in specific order using r object of time, denoted by nPr:

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

arrangment of n objects in specific order taking all n at a time is n!

### 9. Combination Rule

Combination: selection of objects with no order the selection of r objects out of n objects can be done in  $nC_r$  ways

(9) 
$$\binom{\binom{n}{n}}{r} = \frac{n!}{(n-r)!r!}$$

## 10. Combinatorics Ex

(10) 
$$n = \frac{8!}{5! * 3!} = \frac{5! * 6 * 7 * 8}{1 * 2 * 3} = \frac{\frac{6!}{3!3!}}{56} + \frac{\frac{6!}{5!}}{56} = \frac{13}{28}$$

#### 11. Bayes Theorem

(11) 
$$P(A \cap B) = P(A) \cdot P(B|A)$$
$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$
$$P(A|B) = \frac{P(Ai * P(D|Ai))}{P(Ai)P(D|Ai) + \dots + P(Ak)P(D|Ak)}$$

### 12. Probability Distributions

- (1) variable: characteristic or attribute that can assume different values
- (2) Random Variable: variable whose value is determined by chance
- (3) have requirement that  $\sum P(X) = 1, 0 \le P(X) \le 1$
- (4) Expectation:  $E(X) = \mu = \sum X \cdot P(X)$

## 13. The Binomial Distribution

binomial experiment: probability experiment that satisfies:

- (1) each trial can have only two outcomes or can be reduced to success or failure
- (2) fixed # of trials
- (3) outcomes independent of each other
- (4) probability of success must remain same for each trial

$$\begin{split} Mean: \mu &= \sum (X \cdot P(X)) \\ Variance: \sigma^2 &= \sum [X^2 \cdot P(X)] - \mu^2 \\ StandardDeviation: \sigma &= \sqrt{\sigma^2} \\ Expectation: E(X) &= \mu = \sum (X \cdot P(X)) \end{split}$$

#### 14. Poisson distribution

discrete probability distribution that is useful in large numbe of trials with small success rate

(12) 
$$P(X;\lambda) = \frac{e^{-\lambda}\lambda^x}{X!} where X = 0, 1, 2$$

where  $\lambda = n \cdot p$  is the parameter of the poisson distribution

- (1) Mean:  $\mu = \lambda$
- (2) Variance:  $\sigma^2 = \lambda$
- (3) Standard Deviation:  $\sigma = \sqrt{\lambda}$

#### 15. Poisson Process

events occur randomly in time and space

- (1) Independence: number of occurrences in disjoint intervals are independent
- (2) individuality: events occur singly P(two or more events occur simultaneously) = 0
- (3) Homogeneity: events occur according to a uniform rate of intensity

if events occur with average rate of  $\lambda$  per unit of time, and X is num of events which occur in t units of time, the X Poisson( $\lambda \cdot t$ )

(13) 
$$f(x) = \frac{e^{-\lambda t}(\lambda t)^x}{x!}$$

$$(14) E(x) = \lambda t$$

$$(15) Var(x) = \lambda t$$

### 16. Geometric Distribution

Bernoulli trials completed until successful.

let X be no of independant Bernoulli trials until the first success including first success then X follow geo dist. X Geom(p)

(16) 
$$f(x) = p(1-p)^{x-1}, x = 1, 2, \dots$$
$$E(X) = \frac{1}{p}$$
$$Var(X) = \frac{1-p}{p^2}$$

16.1. Memoryless Property of Geometric Random Variable. let X be a Geometric (p) R.V. and t1, t2  $\in$  R. then  $P(X = t1 + t2 \mid x = t1) = P(X > t2)$ 

buy a car and travel x kilometers. reliability does not depend on previous examples

(17) 
$$LS = \frac{P(X \ge t_1 + t_2 and x \ge t_1)}{P(X \ge t_1)}$$

17. Chebyshev's Theorem

Let X be a R.V with E(X) and  $Var(x) = \sigma^2$  Then for any  $\epsilon > 0$ .

(18) 
$$P(|X - E(X)| \ge \epsilon) \le \frac{Var(x)}{\epsilon^2} \forall \epsilon > 0$$

(19) 
$$P(|X - E(X) < \epsilon) = 1 - P(|X - E(X) \ge \epsilon) \ge 1 - \frac{VAR(X)}{\epsilon^2}$$

Let  $\epsilon = k\sigma$ , where  $k \geq 0$ ,  $\sigma > 0 - stdeviation$ 

(20) 
$$P(|X - E(X)| < k\sigma) \ge 1 - \frac{\sigma^2}{k^2 \sigma^2}$$

(21) 
$$P(|X - E(X)| < k\sigma) \ge 1 - \frac{1}{k^2}$$

17.1. **Example.** The # of students who miss a friday class is a Random varible X with mean 15 and st. deviation 2. Find/estimate the probability.

(22) 
$$P(9 < x < 21) = P(-6 < x - 15 < 6) = P(|X - 15| \le 6) = 1 - \frac{1}{3^2}$$

17.2. Example 2.

(23) 
$$X Bin(5,1/2); FindP(|X - E(X)| < 2\sigma)$$

$$E[X] = n.p = 5 \cdot 1/2 = 2.5$$

$$Var[X] = np(1-p) = 5 \cdot 1/21/2 = 5/4 = 1.25$$

$$P(|X - 2.5| < 2 \cdot \sqrt{1.25}) \ge 1 - 1/4 = .75$$

actual probability

$$P(|x-2.5| < 2\sqrt{1.25} >) = P(-2.236 < x - 2.5 < 2.236)P(.264 < X < 4.37) = P(1 \le x \le 4) = F(4) - F(0)$$

#### 8

### 18. Continuous Random variables and distributions

function from sample space to real numbers

(27) 
$$def: X: S \to R[a, b] \in R$$

where R(X) is continuous and individual points in R must have 0 probability  $P(X = x) = 0 for any x \in \Re$ 

$$P(a \le X \le b) = P(a < X < b)$$

18.1. **P.** .d.f probability density function

for random variable X, denoted f(x), assigns probability to  $x \in \Re(X)$ 

- (1)  $P(a < X < b) = \int_a^b f(x)dx, (a,b) \subseteq \Re(x)$
- (2)  $f(x)0, \forall x \in \Re(x)$ (3)  $\int_{\infty}^{\infty} f(x)dx = 1$
- 18.2. Cumulative distribution function(cdf). cdf of continuous random variale X, denoted F(x) gives prob that X takes on value less than or equal to x.

(28) 
$$F(x) = P(X \le x) = P(X < x)$$

- 18.2.1. Properties.
  - (1)  $F(-\infty) = 0$
  - (2)  $F(\infty) = 1$
  - (3) F(x) non decreasing
- 18.3. Relationship b/w pdf and cdf.

(29) 
$$\int_{a}^{b} f(x)dx = P(a < X < b)$$
$$= P(X < b) - P(X < a)$$
$$= F(b) - F(a)$$

fundamental theorem of calc  $\frac{dF(x)}{dx} = f(x)$ 

# 19. Continuous uniform distribution

the probability of any subinterval of the range is proportional to the length of the interval(two sub-interv. with same length must have same probability)

(30) 
$$f(X) = \frac{1}{b-a}, a \le x \le b$$

$$F(x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \le x \le b \\ 1 & b < x \end{cases}$$

19.1. Mean/expected value.

$$E(X) = \int_{x} x f(x) dx = \mu$$
$$E[g(x)] = \int_{x} g(x) f(x) dx$$
(31)

E[aX+bY] = aE[x]+bx

19.2. Variance.

(32) 
$$Var(X) = E[(X - E(X))^{2}]$$
$$Var[aX1 + bX2] = a^{2}Var(X1) + b^{2}Var(X2)$$

19.3. **Exponential Distribution.** Events occur according to a Poisson process, measure the inter arrival time s b/w events. If X is the amount of time until next event in a posisson process, X  $Exp(\theta)$  where  $\theta = \frac{1}{\lambda}$ 

(33) 
$$f(x) = (1/\theta)e^{\frac{x}{\theta}}F(x) = 1 - e^{-\frac{x}{\theta}}E(x) = \theta Var(X) = \theta^2$$