Graphing

Class Width = $\frac{high\ value - low\ value}{number\ of\ classes}$

Measures of Central Tendency

Sample mean: $\bar{X} = \frac{\Sigma x}{n}$

Population mean: $\mu = \frac{\Sigma x}{N}$

Class Midpoint:

lower class limit + next lower class limit

Range = largest value - smallest value

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

Computation formula: $S = \sqrt{\frac{\sum x^2 - [(\sum x)^2/n]}{n-1}}$

Population standard deviation: $\sigma = \sqrt{\frac{\Sigma(x-\mu)^2}{N}}$

Sample variance: s^2 Population variance: σ^2

Sample mean for grouped data: $\bar{x} = \frac{\sum xf}{n}$

Sample standard deviation for grouped data:

$$s = \sqrt{\frac{\Sigma(x - \bar{x})^2 \cdot f}{n - 1}} = \sqrt{\frac{\Sigma x^2 f - [(\Sigma x f)^2 / n]}{n - 1}}$$

Coefficient of Variation: $CV = \frac{\sigma}{\mu}$ Interquartile range: $IQR = Q_3 - Q_1$

Lower outlier boundary: $Q_1 - 1.5(IQR)$

Upper outlier boundary: $Q_3 + 1.5(IQR)$

Location corresponding to a given percentile: $L = \frac{P}{100} \cdot n$

Percentile for a given data value: $\% = 100 \cdot \frac{(Location \ of \ x) - 0.5}{n}$

Correlation and Regression

Correlation coefficient:

$$r = \frac{1}{n-1} \sum \left(\frac{x - \bar{X}}{S_x} \right) \left(\frac{y - \bar{y}}{S_y} \right)$$

$$r = \frac{n\Sigma xy - (\Sigma x)(\Sigma y)}{\sqrt{n\Sigma x^2 - (\Sigma x)^2}\sqrt{n\Sigma y^2 - (\Sigma y)^2}}$$

Coefficient of determination: r^2

Equation of least-squares regression line:

$$\hat{y} = a + bx$$

Slope of least-squares regression line:

$$b = r \frac{S_y}{S_x} = \frac{n\Sigma xy - (\Sigma x)(\Sigma y)}{n\Sigma x^2 - (\Sigma x)^2}$$

y-intercept of least-squares regression line:

$$a = \bar{y} - b\bar{x}$$

Residual: $y - \hat{y}$

Probability

Rule of Complements: $P(A^C) = 1 - P(A)$

General Addition Rule:

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

Mutually Exclusive Events:

$$P(A \text{ or } B) = P(A) + P(B)$$

Conditional Probability:

$$P(B|A) = \frac{P(A \text{ and } B)}{P(A)}$$

General Multiplication Rule:

$$P(A \text{ and } B) = P(A)P(B|A) = P(B)P(A|B)$$

Independent Events:

$$P(A \text{ and } B) = P(A)P(B)$$

Combination rule: ${}_{n}C_{r} = \frac{n!}{r!(n-r)!}$

Permutation rule:
$${}_{n}P_{r} = \frac{n!}{(n-r)!}$$

IMPORTANT FORMULAS FOR MA110

Discrete Probability

Mean of a discrete probability distribution: $\mu = \Sigma x P(x)$

Variance of a discrete random variable: $\sigma_X^2 = \Sigma[(x - \mu_X)^2 \cdot P(x)] = \Sigma[x^2 \cdot P(x)] - \mu_X^2$

Standard deviation of a discrete random variable:

$$\sigma_X = \sqrt{\sigma_X^2}$$

Binomial probability distribution:

$$P(x) = C_{n,x}p^x(1-p)^{n-x}$$

x = number of successesp = probability of success

Mean of a binomial random variable: $\mu_{\rm X}=np$ Standard deviation of a binomial random variable:

$$\sigma_X = \sqrt{np(1-p)}$$

Normal Distribution

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

Convert z-score to raw score: $x = z\sigma + \mu$

Mean for a sample mean: $\mu_{\bar{\chi}} = \mu$ Standard error:

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

z-score for a sample mean: $Z=rac{ar{x}-\mu}{\sigma_{\overline{x}}}$

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

Mean of the sample proportion: $\mu_{\hat{p}} = p$ Standard deviation of the sample proportion:

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

z-score for a sample proportion:

$$z = \frac{\hat{p} - \mu_{\hat{p}}}{\sigma_{\hat{p}}}$$

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Confidence Intervals

Confidence intervals for a mean:

$$\bar{x} - E < \mu < \bar{x} + E$$

When σ is known:

Margin of Error: $E = z_c \cdot \sigma_{\bar{x}}$

When σ is unknown:

Margin of Error: $E = t_c \cdot S_{\bar{x}}$

Determining Sample Size:

$$n = \left(\frac{z_c \cdot \sigma}{E}\right)^2$$

Confidence intervals for a proportion:

$$\hat{p} - E$$

Margin of Error: $E = z_c \cdot \sigma_{\widehat{p}}$

Where:
$$\sigma_{\widehat{p}} = \sqrt{\frac{\widehat{p}(1-\widehat{p})}{n}}$$

Sample Sizes for proportions:

 $n = \hat{p}(1 - \hat{p}) \left(\frac{z_c}{E}\right)^2$ if a value for \hat{p} is available

 $n=0.25\left(rac{z_c}{E}
ight)^2$ if no value for \hat{p} is available

Hypothesis Testing

When σ is known: $Z=rac{ar{x}-\mu_{\overline{x}}}{\sigma_{\overline{x}}}$

When σ is unknown: $t = \frac{\bar{x} - \mu_{\bar{x}}}{S_{\bar{x}}}$

For a proportion: $z = \frac{\hat{p} - \mu_{\hat{p}}}{\sigma_{\hat{n}}}$

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