

MATH-338 Midterm 2 Cheat Sheet

THEORY

Lecture 13-14: PDF Curve: total area of one. Computed area = probability. Z-Scores: universal standard for density curves with different scales. 68-95-99.7 rule: the probabilities of 2, 4, & 6 σ away from μ . Parameter (μ): average age in class is 21 years. Parameter (σ): I scored 2 σ from μ , which means I did better than average. Skewed left: long left tail (sloping \rightarrow). Skewed right: long right tail (sloping \leftarrow). Symmetric: perfectly about μ . Unimodal: one hump. Bimodal: two humps. Linear transformation: changing the base function by appending or removing a constant value. Linear Combination: two separate σ or μ are combined ($\mu_{x+y} = \mu_x + \mu_y$). Whisker plots are an effective method to determine if a data set contains outliers (data points not belonging to the sample set).

Lecture 15: Sampling error: when the sample is a misrepresentation of the population. Biased: when there is a difference between the expected and the observed value and bias can cause variability in the sampling distribution. CLT: independent random variables will eventually converge to a normal distribution.

Lecture 17-19: Type I Error: convict an innocent man (μ is not in the interval). Type II Error: we let a guilty man go. P-Value: the probability of obtaining test results at least as extreme as the results actually observed during the test, assuming that the null hypothesis is correct.

Lecture 20: Confidence Interval: the range of values within which reasonably would believe μ to be. Confidence is our estimate of μ . We will never know the value of μ . Between: variation due to changes in μ . Within: variation due to individual differences. Symmetric data $\implies \sigma$ is small. Posthoc Procedure: The decision to reject the null hypothesis at a significance level of $\alpha = 0.01$. There is sufficient evidence to conclude that at least one of the population means different from at least one other population mean.

FRAMEWORK FLOW CHART

NPHT

- Parameter is μ [population mean]. $\mu_0 = \mu_1$
- \bar{X} is sample mean. Under CLT, normal distribution at μ_0 for H_0 and μ_1 for H_1 .
- We accept H_0 if not in CR.

N-P Power Analysis

- Define parameter and its value under H_0 and H_1
- Define a test statistic and its sampling distribution under both hypotheses.
- Use α to compute critical region
- Compute power and compare to 80

One-Sample T-Statistic [NP]

- If t_{observed} in CR, then accept H_1 : $\mu = \mu_1$. Else accept H_0 : $\mu = \mu_0$

Two-Tailed Test

- Take the upper and lower limit of the curve and the significance level (α) is the cut off point of being *statistically significant*. Treat as critical region. If in CR, then accept H_1 . Else accept H_0 .

ANOVA

- Null Hypothesis: $H_0 : \mu_1 = \mu_2 = \mu_3 = \mu_1$
- If the variability BETWEEN the means (Δx) in the numerator is relatively large compared to the variance within the samples (internal spread) in the denominator, the ratio will be much larger than 1.
- The samples then most likely do NOT come from a common population REJECT H_0 . (if at least one of the μ is not equal.)
- ANOVA tests CANNOT determine/make conclusions about all populations means (\forall), only at least one element in the set ($\mu \in \forall$)
- Usage: compare control group and observational studies of more than three populations.

NHST

- Define a parameter and its value under H_0 .
- Define an interval representing an inequality
- Define a test statistic and its sampling distribution under H_0
- Compute p-value. P-Value \leq sig level \implies reject H_0 & accept H_1 . P-Value $>$ sig level \implies fail to reject H_0 . Can only be $>$, $<$ \neq .

Two-Sided Test

- Neyman-Pearson**
- Critical region is $\frac{1}{2}$ left tail and $\frac{1}{2}$ right tail of sampling distribution under H_0 . Power will \downarrow .
- NHST**
- Find the "one-sided" p-value and double it.

Matched Pairs t-Test

- Paired subjects receives their respective treatment or an individual gets two treatments. Also a subset of block design.
- $H_0: \mu_d = 0$ (no difference) and $H_a: \mu_d \neq 0$ (difference).
- If p-value $\leq \alpha$, we reject H_0 & accept H_a conclude there is a difference.
- If p-value $>$ significance level, we fail to reject H_0 cannot claim there is a difference. (We do not have any definitive truth to accept the null hypothesis)
- Requirements: large population, normal distribution, σ is unknown.

FORMULAS

- $\square = \text{width} \times \frac{1}{\text{width}}$ (finite curve)
- $Z = \frac{\bar{x} - \mu}{\sigma}$ (z-score)
- $X \sim N(\mu, \sigma)$
- $\bar{X} \sim N(\mu, \frac{\sigma}{\sqrt{n}})$
- $SEM = \frac{s}{\sqrt{n}}$ (compute standard error)
- $t = \frac{\bar{X} - \mu_0}{\frac{s}{\sqrt{n}}}$ [NP]
- $t = \frac{\frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}}{\frac{s}{\sqrt{n}}}$ [matched pair]
- $t = \frac{(\bar{x}_1 - \bar{x}_2)}{\sqrt{\frac{(s_1)^2}{n_1} + \frac{(s_2)^2}{n_2}}} \sim t(K)$ [NHST]
- $(\bar{x} \pm t^{**} \times \frac{s}{\sqrt{n}})$ [confidence interval]
- $\bar{x} - t^{**} \times \frac{s}{\sqrt{n}} < \mu < \bar{x} + t^{**} \times \frac{s}{\sqrt{n}}$ [confidence interval when sample mean given]

- $IQR = Q_3 - Q_1$
- $K = 1.5$
- Lower fence: $Q_1 - K \times IQR$
- Upper fence: $Q_3 + K \times IQR$
- $t = \frac{\Delta \bar{x} - \Delta \mu}{\frac{\Delta s}{\sqrt{n}}}$
- $df = n - 1$
- $df(\text{treatment}) = k - 1$ ($k \leftarrow$ number of categories)
- $df(\text{error}) = N - k$ ($N \leftarrow$ total sample size).
- $MSTr = SSTr / (k - 1)$ $SSTr \leftarrow$ sum of treatment
- $MSE = SSE / (N - k)$ $SSE \leftarrow$ sum of error
- $F = \frac{MSTr}{MSE}$
- $C = 1 - \alpha$ [confidence level]
- $((\bar{x}_1 - \bar{x}_2) - t^{**} \times \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}), ((\bar{x}_1 - \bar{x}_2) + t^{**} \times \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}})$ [qt($\frac{\alpha}{2}$, 347.41, lower.tail = F) $\implies t^{**}$]