

# MATH-338 Midterm 2 Cheat Sheet

## THEORY

**Day 14:** probability density function is represented an integral with function  $f(x)$ . Our probability lies within the curve and is always 1. Density curve  $\rightarrow$  bell curve. Z-Score allows us to have a universal standard for density curves with different scales. They are directly proportional to the standard deviation and the delta from the mean of the graph.

**Day 15:** unimodal: one hump, bimodal: two humps. Mean is resistant whereas the mean is subject to change. Density curves decay to histograms (integral  $\rightarrow$  to Reimann Sum). Whisker plots are an effective method to determine if a data set contains outliers (data points not belonging to the sample set). Left skew: long left tail. Sloping  $\rightarrow$ . Right skew: long right tail. Sloping  $\leftarrow$

**Day 16:** error: since there is some error while taking sample data, we do allow for some buffer. We also do not measure exact but to a tolerance which is influenced by the buffer above. Central Limit Theorem: when population size is "large enough"  $\bar{x}$  is an approximation. Higher skew and outliers suggest a larger  $n$  value.

**Day 18:** As  $n \uparrow$ ,  $SEM \downarrow$ .

**Day 19:**  $H_1: \mu < \mu_0 \leftarrow$  left side.  $H_1: \mu \neq \mu_0 \leftarrow$  n  $\sigma$  on both side but no middle.  $H_1: \mu > \mu_0 \leftarrow$  lower.tail = TRUE. Population distribution normality  $\implies$  sample population distribution normality. Matched pairs design:

**Day 20:** Two Independent Samples t-Test: two unrelated treatments into one numerical response variable measured in two independent groups. Two different  $\mu_1$  and  $\mu_2$ . NHST approach; identify  $\mu_i$

## FORMULAS

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| <ul style="list-style-type: none"><li><math>\square = width \times \frac{1}{width} \text{ (finite curve)}</math></li><li><math>Z = \frac{x-\mu}{\sigma} \text{ (z-score)}</math></li><li><math>X \sim N(\mu, \sigma)</math></li><li><math>\bar{X} \sim N(\mu, \frac{\sigma}{\sqrt{n}})</math></li><li><math>SEM = \frac{s}{\sqrt{n}}</math></li><li><math>t = \frac{\bar{X}-\mu}{\frac{s}{\sqrt{n}}}</math></li><li><math>t = \frac{\bar{x}}{\sigma_{\bar{x}}}</math></li><li><math>t = \frac{(\bar{x}_1 - \bar{x}_2)}{\sqrt{\frac{(s_1)^2}{n_1} + \frac{(s_2)^2}{n_2}}} \sim t(K) \text{ [NHST]}</math></li></ul> | <ul style="list-style-type: none"><li><math>IQR = Q_3 - Q_1</math></li><li><math>K = 1.5</math></li><li>Lower fence: <math>Q_1 - K \times IQR</math></li><li>Upper fence: <math>Q_3 + K \times IQR</math></li><li><math>t = \frac{\Delta \bar{x} - \Delta \mu}{\frac{\Delta s}{\sqrt{n}}}</math></li><li><math>df = n - 1</math></li><li><math>df(\text{treatment}) = k - 1</math> k <math>\leftarrow</math> number of categories</li><li><math>df(\text{error}) = N - k</math> N <math>\leftarrow</math> total sample size.</li><li><math>MSTR = SSTR / (k - 1)</math> SSTR <math>\leftarrow</math> sum of treatment</li><li><math>MSE = SSE / (N - k)</math> SSE <math>\leftarrow</math> sum of error</li><li><math>F = \frac{MSTR}{MSE}</math></li></ul> |
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## FRAMEWORK FLOW CHART

### NPHT

- Parameter is  $\mu$  [population mean].  $\mu_0 = \mu_1$
- $\bar{X}$  is sample mean. Under CLT, normal distribution at  $\mu_0$  for  $H_0$  and  $\mu_1$  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  $\alpha$  to compute critical region
- Compute power and compare to 80

### One-Sample T-Statistic [NP]

- If  $t_{\text{observed}}$  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 ( $\alpha$ ) is the cut off point of being *statistically significant*. Treat as critical region. If in CR, then accept  $H_1$ . Else accept  $H_0$ .

### 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.

### ANOVA

- Null Hypothesis:  $H_0: \mu_1 = \mu_2 = \mu_3$
- If the variability BETWEEN the means ( $\Delta x$ ) 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  $H_0$ .

### 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).
- Requirements: large population, normal distribution,  $\sigma$  is unknown.