

Statistics One

Lecture 16
Analysis of Variance (ANOVA)

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Two segments

- One-way ANOVA
- Post-hoc tests

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Lecture 16 ~ Segment 1

One-way ANOVA

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Analysis of Variance (ANOVA)

- Appropriate when the predictors (IVs) are all categorical and the outcome (DV) is continuous
 - Most common application is to analyze data from randomized controlled experiments

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Analysis of Variance (ANOVA)

- More specifically, randomized controlled experiments that generate more than two group means
 - If only two group means then use:
 - Independent t-test
 - Dependent t-test

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Analysis of Variance (ANOVA)

- If more than two group means then use:
 - Between groups ANOVA
 - Repeated measures ANOVA

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Example

- Working memory training
 - Four independent groups (8, 12, 17, 19)
 - IV: Number of training sessions
 - DV: IQ gain
 - Null hypothesis: All groups are equal

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Working memory training



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Analysis of Variance (ANOVA)

- ANOVA typically involves NHST
- The test statistic is the F-test (F-ratio)
 - $F = (\text{Variance between groups}) / (\text{Variance within groups})$

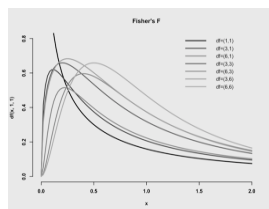
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Analysis of Variance (ANOVA)

- Like the t-test and family of t-distributions
- The F-test has a family of F-distributions
 - The distribution to assume depends on
 - Number of subjects per group
 - Number of groups

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Analysis of Variance (ANOVA)



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One-way ANOVA

- F-ratio
 - $F = \text{between-groups variance} / \text{within-groups variance}$
 - $F = MS_{\text{Between}} / MS_{\text{Within}}$
 - $F = MS_A / MS_{S/A}$

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One-way ANOVA

- $F = MS_A / MS_{S/A}$
- $MS_A = SS_A / df_A$
- $MS_{S/A} = SS_{S/A} / df_{S/A}$

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One-way ANOVA

- $SS_A = n \sum (Y_j - Y_T)^2$
 - Y_j are the group means
 - Y_T is the grand mean

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One-way ANOVA

- $SS_{S/A} = \sum (Y_{ij} - Y_j)^2$
 - Y_{ij} are individual scores
 - Y_j are the group means

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One-way ANOVA

- $df_A = a - 1$
- $df_{S/A} = a(n - 1)$
- $df_{TOTAL} = N - 1$

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Summary Table

| Source | SS | df | MS | F |
|--------|-------------------------|------------|-----------------------|-------------------|
| A | $n \sum (Y_j - Y_T)^2$ | $a - 1$ | SS_A / df_A | $MS_A / MS_{S/A}$ |
| S/A | $\sum (Y_{ij} - Y_j)^2$ | $a(n - 1)$ | $SS_{S/A} / df_{S/A}$ | ----- |
| Total | $\sum (Y_{ij} - Y_T)^2$ | $N - 1$ | ----- | ----- |

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Effect size

- $R^2 = \eta^2$ (eta-squared)
- $\eta^2 = SS_A / SS_{Total}$

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Assumptions

- DV is continuous (interval or ratio variable)
- DV is normally distributed
- Homogeneity of variance
 - Within-groups variance is equivalent for all groups
 - Levene's test

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Homogeneity of variance

- If Levene's test is significant then homogeneity of variance assumption has been violated
 - Conduct pairwise comparisons using a restricted error term

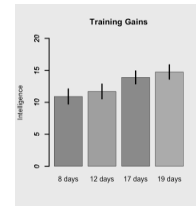
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Example

- Working memory training
 - Four independent groups (8, 12, 17, 19)
 - IV: Number of training sessions
 - DV: IQ gain
 - Null hypothesis: All groups are equal

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Working memory training



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Working memory training

```
> anova.WMT <- aov(WMT$IQ ~ WMT$condition)
> summary(anova.WMT)
              Df Sum Sq Mean Sq F value    Pr(>F)
WMT$condition 3  196.1    65.36   10.49 7.47e-06 ***
Residuals    76  473.4     6.23
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> # Post hoc (Tukey)
> TukeyHSD(anova.WMT)
  Tukey multiple comparisons of means
    95% family-wise confidence level

Fit: aov(formula = WMT$IQ ~ WMT$condition)
```

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Working memory training

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  Tukey multiple comparisons of means
    95% family-wise confidence level

Fit: aov(formula = WMT$IQ ~ WMT$condition)

$WMT$condition
      diff      low      up    p.adj
17 days-12 days  2.285  0.131797  4.278283  0.832686
19 days-12 days  3.850  0.976797  5.223283  0.8813809
8 days-12 days   -0.790 -2.863283  1.283283  0.7452958
19 days-17 days  0.845 -1.228283  2.918283  0.7883557
8 days-17 days  -2.395 -5.868283 -0.927297  0.8816487
8 days-19 days  -3.848 -5.913283 -1.767297  0.8880354
```

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Results from t-test: 12 vs. 17

```
> t.test(Days12, Days17, var.equal=T)

Two Sample t-test

data: Days12 and Days17
t = -2.868, df = 38, p-value = 0.006706
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -3.7534458 -0.6471965
sample estimates:
mean of x mean of y
11.70287 13.90319

> cohensD(Days12, Days17, method='pooled')
[1] 0.9069322
```

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Segment summary

- ANOVA is used to compare means, typically in experimental research
 - Categorical IV
 - Continuous DV

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Segment summary

- ANOVA assumes homogeneity of variance
 - Evaluate with Levene's test

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Segment summary

- Post-hoc tests, such as Tukey's procedure, allow for multiple pairwise comparisons without an increase in the probability of a Type I error

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END SEGMENT

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Lecture 16 ~ Segment 2

Post-hoc tests

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Post-hoc tests

- Post-hoc tests, such as Tukey's procedure, allow for multiple pairwise comparisons without an increase in the probability of a Type I error

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Post-hoc tests

- Many procedures are available; the degree to which p-values are adjusted varies according to procedure
 - Most liberal: No adjustment
 - Most conservative: Bonferroni procedure

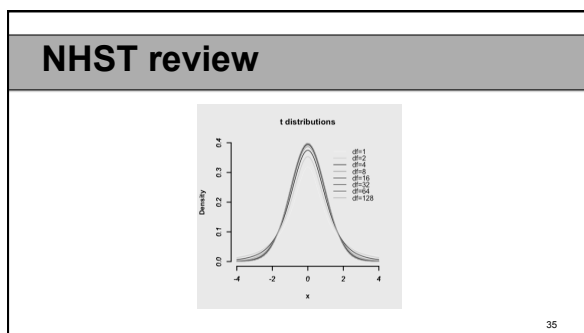
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| NHST review | | | |
|-------------|-------------|-----------------------|----------------------------|
| | | Experimenter Decision | |
| | | Retain H_0 | Reject H_0 |
| Truth | H_0 true | Correct Decision | Type I error (False alarm) |
| | H_0 false | Type II error (Miss) | Correct Decision |

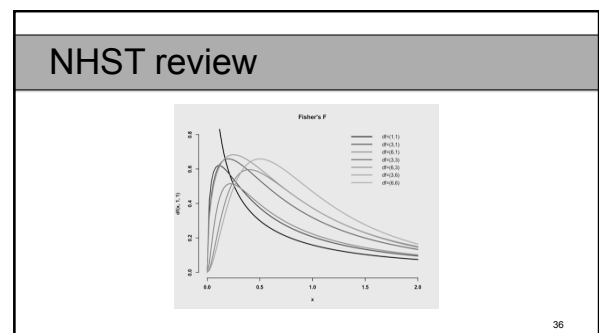
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| NHST review | | | |
|-------------|-------------|-----------------------|---------------------------|
| | | Experimenter Decision | |
| | | Retain H_0 | Reject H_0 |
| Truth | H_0 true | Correct Decision | Type I error $p = .05$ |
| | H_0 false | Type II error (Miss) | Correct Decision |

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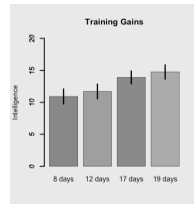


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Working memory training



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Tukey's procedure

```

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 8 days-12 days  -0.798 -2.863203  1.263203  0.7492938
19 days-17 days  0.845  -1.23203  2.916203  0.7883557
 8 days-17 days  -2.995 -5.068203 -0.921797  0.0016467
 6 days-19 days  -3.848 -5.913203 -1.782797  0.0009354

```

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[1] 0.9069322

```

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Bonferroni procedure

```

> p.adjust(.006706, method="bonferroni", 6)
[1] 0.040236

```

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Comparison of procedures

| Procedure | p-value for 12 vs. 17 |
|--------------------|-----------------------|
| Independent t-test | 0.0067 |
| Tukey | 0.0327 |
| Bonferroni | 0.0402 |

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Post-hoc tests

- Post-hoc tests, such as Tukey's procedure, allow for multiple pairwise comparisons without an increase in the probability of a Type I error
- Procedures vary from liberal to conservative

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END SEGMENT

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END LECTURE 16

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