

ANOVA

- $J = N$ OBS X SAMPLE
- $I = N$ TREATMENTS / POPULATION

SINGLE FACTOR

COMPARISON OF MORE THAN TWO POPULATIONS (TREATMENTS) WITH SAME SAMPLES

$H_0: \mu_1, \dots, \mu_N, \mu_1 = \mu_2 = \dots = \mu_N$

ASSUME: POPULATIONS ARE ALL N.O WITH $VAR = \sigma^2 \mid E(X_{ij}) = \mu_i, VAR(X_{ij}) = \sigma^2$

IDEA: COMPARE DIFF BETWEEN SAMPLE TO DIFF WITHIN SAMPLE

TEST STATISTIC: $F = \frac{MSTR}{MSE}$, H_0 TRUE $\rightarrow E(MSTR) = E(MSE)$

• SUMS OF SQUARES

TOTAL $SST = \sum_i \sum_j (x_{ij} - \bar{x}_{..})^2 = \sum_i \sum_j x_{ij}^2 - \frac{1}{J} \sum_i x_{i.}^2$

TREATMENT $SSTR = \sum_i \sum_j (\bar{x}_{i.} - \bar{x}_{..})^2 = \frac{1}{J} \sum_i x_{i.}^2 - \frac{1}{J} \sum_i x_{i.}^2$

ERROR $SSE = \sum_i \sum_j (x_{ij} - \bar{x}_{i.})^2 =$

POP. SAMPLE MEAN: $\bar{x}_i = \frac{\sum_j x_{ij}}{J}$

GRAND MEAN: $\bar{x} = \frac{\sum_i \sum_j x_{ij}}{I \cdot J}$

POP. SAMPLE VARIANCE: $s_i^2 = \frac{\sum_j (x_{ij} - \bar{x}_i)^2}{J - 1}$

$MSTR = \frac{J}{I - 1} \sum_i (\bar{x}_{i.} - \bar{x}_{..})^2$ (BETWEEN SAMPLES) $MSE = \frac{s_1^2 + \dots + s_I^2}{I}$ (WITHIN SAMPLES)

REJECT WHEN $F \geq C = \alpha$ or $F \geq F_{\alpha, I-1, I(J-1)}$

$x_{i.} = \sum_j x_{ij}, x_{..} = \sum_i \sum_j x_{ij}$

$SST = \overset{\text{BETW}}{SSTR} + \overset{\text{WTH}}{SSE}$
VARIANCE BIAS

$MSTR = \frac{SSTR}{I - 1}$

$MSE = \frac{SSE}{I(J - 1)}$

$F = \frac{MSTR}{MSE}$

TUKEY'S PROCEDURE

USES STUDENTIZED RANGE DISTRIBUTION $Q_{\alpha, I, N}$
IDENTIFIES SIGNIFICANTLY DIFFERENT μ_i WHEN ANOVA IS INCONCLUSIVE

- SORT μ_i s IN INCREASING ORDER
- COMPUTE $w = Q_{\alpha, I, I(J-1)} \sqrt{MSE/J}$
- UNDERLINE CONNECT THOSE DIFFERING BY LESS THAN w
- NO SAME LINE \rightarrow SIGNIFANT DIFFERENCE

FUNCTION IS NOT $\mu_i - \mu_j$
CONSTANTS c_i, c_j
 $\left[\sum c_i \bar{x}_i \pm t_{\alpha/2, I(J-1)} \sqrt{\frac{MSE \sum c_i^2}{J}} \right]$

- POSSIBLE TO ADAPT TO DIFF. N. SAMPLES X POP
- POSSIBLE TO TRANSFORM RVs TO EQUALIZE VARIANCES

ANOVA MODEL EQUATIONS

$x_{ij} = \mu_i + \epsilon_{ij}$, ϵ IS RANDOM DEVIATION FROM POP.

REFORMULATE

$x_{ij} = \mu_i + \epsilon$ to $\mu = \frac{1}{I} \sum_i \mu_i, \alpha_1 = \mu_1 - \mu, \dots, \alpha_I = \mu_I - \mu \rightarrow \mu_i = \mu + \alpha_i$ EFFECT DUE TO i TH TREATMENT

$\rightarrow x_{ij} = \mu + \alpha_i + \epsilon_{ij}, H_0: \alpha_1 = \alpha_2 = \dots = \alpha_I = 0$ (FIXED EFFECTS MODEL)

- CAN BE MADE INTO RANDOM EFFECTS MODEL
 $\alpha_i \rightarrow A_i, A_i$ IS RV NORMALLY INDEPENDENT.
 $E(A_i) = E(\epsilon_{ij}) = 0$

TWO FACTOR ANOVA

MODEL EQUATIONS

$X_{ij} = \mu_{ij} + \epsilon_{ij}$, ϵ_{ij} NORMAL WITH COMMON VARIANCE σ^2 , JOINT EFFECTS, PROBLEMATIC

→ $X_{ij} = \alpha_i + \beta_j + \epsilon_{ij}$, $\mu_{ij} = \alpha_i + \beta_j$, **ADDITIVE MODEL**, SUM OF EFFECT DUE TO FACTOR A AND EFFECT DUE TO B
 $\sum \alpha_i = 0$, $\sum \beta_j = 0$, COMMON VARIANCE σ^2

HYPOTHESES

$H_{0A}: \alpha_1 = \alpha_2 = \dots = \alpha_I = 0$ NO EFFECT DUE TO A

$H_{0B}: \beta_1 = \beta_2 = \dots = \beta_J = 0$ NO EFFECT DUE TO B

$$\bullet \hat{\mu} = \bar{X}_{..}, \hat{\alpha}_i = \bar{X}_{i.} - \bar{X}_{..}, \hat{\beta}_j = \bar{X}_{.j} - \bar{X}_{..}$$

$$\bullet SST = SSA + SSB + SSE$$

STATISTICS

$$H_{0A} \text{ vs } H_{0A} \longrightarrow f = \frac{MSA}{MSE} \xrightarrow{\text{REJECT}} f_A \geq F_{\alpha, I-1, (I-1)(J-1)}$$

$$H_{0B} \text{ vs } H_{0B} \longrightarrow f = \frac{MSB}{MSE} \xrightarrow{\text{REJECT}} f_B \geq F_{\alpha, J-1, (I-1)(J-1)}$$

TUKEY: JUST LIKE SINGLE FACTOR, BUT SPLIT

NON ADDITIVE STRUCTURE

INTERACTION BETWEEN FACTORS

$$\bullet \alpha_i = \mu_{i.} - \mu$$

$$\bullet \beta_j = \mu_{.j} - \mu$$

$$\bullet \gamma = \mu_{ij} - (\mu + \alpha_i + \beta_j)$$

$$\bullet \mu_{ij} = \mu + \alpha_i + \beta_j + \gamma_{ij}$$

HYPOTHESES: H_{0A}, H_{0B}, H_{0AB}

$$\longrightarrow X_{ijk} = \mu + \alpha_i + \beta_j + \gamma_{ij} + \epsilon_{ijk}$$

$$\bullet SST = SSA + SSB + SSAB + SSE$$

STATISTICS

$$f_A, f_B$$

$$f_{AB} = \frac{MSAB}{MSE}$$

RANDOM EFFECTS: UFE IF COULD HAVE THEM $\gamma \rightarrow G$

THREE FACTOR ANOVA

$$\mu_{ijk} = \mu + \alpha_i + \beta_j + \gamma_k + \delta_{ij}^{AB} + \gamma_{ik}^{AC} + \gamma_{jk}^{BC} + \gamma_{ijk}$$

$$f_A, f_B, f_{ABC}$$

SAME STUFF

2¹ FACTORIAL EXPERIMENTS: FULL GENERALIZATION