# ANOVA: (univoriate)

we reparameterize

$$\mu_{\ell} = \mu + (\mu_{\ell} - \mu) = \mu + \tau_{\ell}$$

Common/overall treatment

effect effect

The Ho can be rewritten as

$$H_0: T_1 = T_2 = ... = T_g = 0$$

$$X = \mu + \tau_{l} + \varepsilon_{lj} \rightarrow \tau \text{ and om error}$$

$$\int_{\text{notice}} \varepsilon_{lj} \sim N(0, \varepsilon^{2})$$

Similarly observation can be written as

$$\chi_{lj} = \bar{\chi} + (\bar{\chi}_{l} - \bar{\chi}_{l}) + (\chi_{l} - \bar{\chi}_{l})$$

estimated

estimated

estimated

estimated

estimated

every limit mean

- Under Ho, the corrected sum of squares

$$\sum_{\ell=1}^{3} \sum_{j=1}^{n_{\ell}} \left( x_{\ell j} - \bar{x} \right)^{2}$$

Observe that

$$\sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi})^{2} = n_{\ell} (\bar{\pi}_{\ell} - \bar{\pi})^{2} + \sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2}$$

$$\Rightarrow \sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi})^{2} = \sum_{\ell=1}^{g} n_{\ell} (\bar{\pi}_{\ell} - \bar{\pi})^{2} + \sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2}$$

$$= \sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2} + \sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2}$$

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$$= \sum_{\ell=1}^{g} \sum_{\ell=1}^{g} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2} + \sum_{\ell=1}^{g} \sum_{\ell=1}^{g} (\pi_{\ell j} - \bar{\pi}_{\ell})^{2}$$

$$= \sum_{\ell=1}^{g} \sum_{\ell=1}^{g}$$

Remark: For estimating  $T_i$ 's we need a contraint that  $\sum_{i=1}^{8} n_i T_i = 0$ 

#### ANOVA Table

Source	Sum of Squares	Degrees of freedom
Treatment	SS <sub>tr</sub>	9-1
Residual	Ssres	n-9
Total	SS cor	n -1

$$SS_{ts}$$
 ~  $\chi^2_{g-1}$  indep  $SS_{tes}$  ~  $\chi^2_{n-g}$ 

$$\frac{SS_{tr}/(g-1)}{SS_{res}/(n-g)} \sim F$$

Remark: For large Ne # l=1,2,..., g, the assumption of normality can be dropped, due to CLT.

# Multivariate ANOVA: (MANOVA)

$$\underline{\rho_{op^n 1}} : \times_{i_n} \times_{i_{n_e}} \times_{i_{n_e}} \xrightarrow{iid} N_{\rho}(\mu_{i_1}, \Sigma)$$

The samples from different population are indep. We want to test

$$\mu_{6}: \mu_{1} = \mu_{2} = \dots = \mu_{g}$$

against HA: Ho not true

### Reparameterization:

$$x_{ej} = \mu + (\mu_e - \mu) + \mathcal{E}_{ej}$$
 $verall$ 
 $treatment$ 
 $effect$ 
 $effect$ 
 $effect$ 
 $treatment$ 
 $treatment$ 
 $treatment$ 
 $treatment$ 
 $treatment$ 
 $treatment$ 

- For estimating Te's, we need a constraint.

$$\overline{X}_{\ell} = \frac{1}{n_{\ell}} \sum_{j=1}^{n_{\ell}} X_{\ell j}$$
 ;  $\overline{X} = \frac{1}{n} \sum_{\ell=1}^{g} n_{\ell} \overline{X}_{\ell}$ 

Equivalent de composition for sample observation:

$$\mathcal{H}_{lj} = \overline{\mathcal{H}} + (\overline{\mathcal{H}}_{\ell} - \overline{\mathcal{H}}) + (\mathcal{H}_{lj} - \overline{\mathcal{H}}_{\ell})$$

$$= \underbrace{\sum_{\substack{l \in l \text{ stimated} \\ l \in l \text{ stimated}}}_{\text{estimated}} + \underbrace{\sum_{\substack{l \in l \text{ stimated} \\ l \in l \text{ stimated}}}_{\text{effect}}$$

- Similar algebra as in ANOVA yields:

$$\sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} \left( \chi_{\ell j} - \bar{\chi} \right) \left( \chi_{\ell j} - \bar{\chi} \right)^{T} = \sum_{\ell=1}^{g} \eta_{\ell} \left( \bar{\chi}_{\ell} - \bar{\chi} \right) \left( \bar{\chi}_{\ell} - \bar{\chi} \right)^{T} \right) SSCP$$
SSCP

Total (orrected SS + 
$$\sum_{\ell=1}^{g} \sum_{j=1}^{n_{\ell}} (x_{\ell j} - \bar{\chi}_{\ell}) (\chi_{\ell j} - \bar{\chi}_{\ell})^{T}$$
  
and cross products (SSCP)

Residual (within)
SSCP

### Observation:

$$W = (n_1-1) S_1 + (n_2-1) S_2 + ... + (n_g-1) S_g \sim Wishart$$

The Ho can be rewritten as

Ho: 
$$T_1 = T_2 = \dots = T_g = 0$$

#### MANOVA Table:

Source	SSCP	degrees of freedom	
Treatment	R	9-1	-
	W	n - g	
Residual	T	0	

## Wilk's Test:

$$\Lambda^* = \frac{|W|}{|B+W|}$$
, known as Wilk's lambda

1B+W)

Distribution: Pg 300 (AMSA) (Table 6.3)

Also see Baytlett test on same page

Two-way treatment effect:

$$\mu_{\ell K} = \mu + (\mu_{\ell} - \mu) + (\mu_{\ell K} - \mu)$$
 $+ (\mu_{\ell K} - \mu_{\ell} - \mu_{K} + \mu)$ 
 $+ (\mu_{\ell K} - \mu_{\ell} - \mu_{K} + \mu)$ 

