

1.

- (a) Write the 5 plausible models (b) For each Model, Propose a discrete Probability model of how the data Arose

- Model 0:  $\mu_1 = \mu_2 = \mu_3 = \mu \longrightarrow X_{i,j} \sim \text{Poisson}(\mu)$   
 Model 1:  $\mu_1 \neq \mu_2 \neq \mu_3 \longrightarrow X_{i,j} \sim \text{Poisson}(\mu_i)$   
 Model 2:  $(\mu_1 = \mu_2) \neq \mu_3 \longrightarrow X_{i,j} \sim \text{Poisson}(\mu_{12})$  for  $i=1,2 \in X_{3,j} \sim \text{Poisson}(\mu_1)$   
 Model 3:  $(\mu_1 = \mu_3) \neq \mu_2 \longrightarrow X_{i,j} \sim \text{Poisson}(\mu_{13})$  for  $i=1,3 \in X_{2,j} \sim \text{Poisson}(\mu_2)$   
 Model 4:  $(\mu_2 = \mu_3) \neq \mu_1 \longrightarrow X_{i,j} \sim \text{Poisson}(\mu_{23})$  for  $i=2,3 \in X_{1,j} \sim \text{Poisson}(\mu_1)$

- (b) Write down the likelihood function for each model, then solve for the Maximum Likelihood Estimate (MLEs) of the model Parameters

Model 1:  $\mathcal{L}(\mu_1, \mu_2, \mu_3) = \text{the joint Prob of obs} \rightarrow P(A) \cdot P(B) \dots P(Z)$   
 $= P(X_{11} = x_{11}, X_{12} = x_{12}, \dots, X_{3n_3} = x_{3j})$   
 $= P(X_{11} = x_{11}) \cdot P(X_{12} = x_{12}) \dots P(X_{n_j} = x_{3j})$   
 $= \prod_{i=1}^3 \prod_{j=1}^{n_i} P(X_{i,j} = x_{i,j}) = \prod_{j=1}^{n_1} P(X_{1j} = x_{1j}) \prod_{j=2}^{n_2} P(X_{2j} = x_{2j}) \prod_{j=3}^{n_3} P(X_{3j} = x_{3j})$

$$\mathcal{L}(\mu_1, \mu_2, \mu_3) = \left( \prod_{j=1}^{n_1} \frac{e^{-\mu_1} \mu_1^{x_{1j}}}{x_{1j}!} \right) \left( \prod_{j=2}^{n_2} \frac{e^{-\mu_2} \mu_2^{x_{2j}}}{x_{2j}!} \right) \left( \prod_{j=3}^{n_3} \frac{e^{-\mu_3} \mu_3^{x_{3j}}}{x_{3j}!} \right)$$

Take the log likelihood

$$\ln \mathcal{L}(\mu_1, \mu_2, \mu_3) = \sum_{j=1}^{n_1} [-\mu_1 + x_{1j} \ln \mu_1 - \ln(x_{1j}!)] + \sum_{j=2}^{n_2} [-\mu_2 + x_{2j} \ln \mu_2 - \ln(x_{2j}!)] + \sum_{j=3}^{n_3} [-\mu_3 + x_{3j} \ln \mu_3 - \ln(x_{3j}!)]$$

Take the first derivative

$$\frac{d \ln \mathcal{L}(\mu_1, \mu_2, \mu_3)}{d \mu_1} = \frac{d}{d \mu_1} \left[ -n_1 \mu_1 + (\ln \mu_1) \left( \sum_{j=1}^{n_1} x_{1j} \right) - \sum_{j=1}^{n_1} \ln(x_{1j}!) \right]$$

$$= -n_1 + \frac{1}{\mu_1} \cdot \sum_{j=1}^{n_1} x_{1j}$$

$$0 = -n_1 + \frac{1}{\mu_1} \sum_{j=1}^{n_1} x_{1j}$$

$$n_1 = \sum_{j=1}^{n_1} x_{1j}$$

or Rewritten

$$\hat{\mu}_1 = \bar{x}_1$$

MLE for group 1

& by same logic for other Params

$\vdots$

$\hat{\mu}_2 = \bar{x}_2$  MLE for G2

$\hat{\mu}_3 = \bar{x}_3$  MLE for G3



Model 2

Likelihood Fxn:  $\mathcal{L}(\mu_{12}, \mu_3)$  MLE's:  $\hat{\mu}_{12} = \bar{X}_{12}$ ,  $\hat{\mu}_3 = \bar{X}_3$

Model 3

Likelihood Fxn:  $\mathcal{L}(\mu_{13}, \mu_2)$  MLE's:  $\hat{\mu}_{13} = \bar{X}_{13}$ ,  $\hat{\mu}_2 = \bar{X}_2$

Model 4

Likelihood Fxn:  $\mathcal{L}(\mu_{23}, \mu_1)$  MLE's:  $\hat{\mu}_{23} = \bar{X}_{23}$ ,  $\hat{\mu}_1 = \bar{X}_1$

Model 0

Likelihood Fxn:  $\mathcal{L}(\mu)$  MLE:  $\hat{\mu} = \bar{X}$

c) For each model, evaluate the likelihood Fxn at the MLE's & the data at hand. That quantity is the maximized likelihood Fxn.

~ See Code ~

d) Do a Likelihood Ratio Test Using Samuel Wilk's  $\chi^2$  approximation to test the Null Hypothesis Vs each one of the Alternatives Models (Including Model 1). Because we are doing mult. hyp. tests, we need to do a Bonferroni type I error correction.

If the # of tests you are doing is  $m$  and you want to do an overall test with Type I error rate  $\alpha$ ,

then the Bonferroni-Corrected  $\alpha_B$  to be used for each of the tests is equal to  $\alpha/m$

~ See Code ~

e) Parametric Bootstrap Likelihood Ratio Tests (PBLRT)

Step 1: Compute the Maximum Likelihood model Parameters under the null for a given data set.

$\hat{\theta}_0$  for  $M_0$  &  $\hat{\theta}_1$  for  $M_1$

$$\hat{\theta}_0 = [\hat{\mu}, \hat{\mu}, \hat{\mu}]$$

$$L(\theta_0) =$$

$$\hat{\theta}_1 = [\hat{\mu}_1, \hat{\mu}_2, \hat{\mu}_3]$$