

STAT475/675 TUT11

<http://www.sfu.ca/~zza115/teaching.html>
<http://people.stat.sfu.ca/~joanh/stat475-675web.html>

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The lady testing tea

- 8 randomly ordered cups of tea: 4 prepared by first adding milk and 4 prepared by first adding the tea
 - Select the 4 cups prepared by each method
 - The lady got all eight cups correct (according to “The Lady Testing Tea” written by David Salsburg)
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- Tabulated as a 2×2 contingency table
 - Fixed marginals n_{1+} , n_{2+} , n_{+1} , n_{+2} and n_{++} (this happens rarely)
 - As long as one element of $\{n_{11}, n_{12}, n_{21}, n_{22}\}$ is known, all the others would be determined.

Fisher’s exact test

- One-sided test: $H_0 : \text{OR} = 1$ vs $H_1 : \text{OR} > 1$ (the success rate of the speculation is higher than that of a random guess)
- Take n_{ij} as the realization (observed value) of random variable N_{ij}
- Test statistic: n_{11} , the number of successes
- Under H_0 , $N_{11} \sim$ Hypergeometric Distribution, i.e., for $m = 0, \dots, n_{++}$,

$$\Pr_{H_0}(N_{11} = m) = \frac{\binom{n_{1+}}{m} \binom{n_{2+}}{n_{+1}-m}}{\binom{n_{++}}{n_{+1}}}$$

- p -value = $\Pr_{H_0}(N_{11} \geq n_{11})$
 - = $1/70$ for $n_{11} = 4$ in the lady tasting tea experiment

Permutation test for independence

- $H_0 : \text{OR} = 1$ vs $H_1 : \text{OR} \neq 1$
- The null distribution of

$$\chi_{\text{obs}}^2 = \sum \frac{(\text{observed} - \text{fitted})^2}{\text{fitted}}$$

is not well-approximated by χ^2 -distribution when some n_{ij} is not large enough

- Estimate the permutation distribution by simulation
 - randomly permute the “guess”
 - repeat the permutation for $B \gg 1$ times and obtain $\chi_{\text{obs},b}^2, b = 1, \dots, B$
 - approximate the null distribution by the permutation distribution
 - p -value $\approx \#\{\chi_{\text{obs},b}^2 : \chi_{\text{obs},b}^2 \geq \chi_{\text{obs}}^2\} / B$

Demo I

The Lady-Tasting-Tea experiment with different digits.

	Lady said		
Truth	Tea First	Milk First	Row Total
Tea First	10	2	12
Milk First	2	10	12
Column Total	12	12	24

Loglinear-logit connection

- loglinear(XYZ) $\Leftrightarrow Y \sim \text{logit}(XZ)$ or multi-logit(XZ)
 - $\ln \mu_{ijk} = \lambda_0 + \lambda_i^X + \lambda_j^Y + \lambda_k^Z + \lambda_{ij}^{XY} + \lambda_{ik}^{XZ} + \lambda_{jk}^{YZ} + \lambda_{ijk}^{XYZ}$
 - $\ln \frac{\pi_{ijk}}{\pi_{ij'k}} = \beta_{j0} + \beta_{ji}^X + \beta_{jk}^Z + \beta_{jik}^{XZ}$
 - $\beta_{j0} = \lambda_j^Y - \lambda_{j'}^Y$
 - $\beta_{ji}^X = \lambda_{ij}^{XY} - \lambda_{ij'}^{XY}$
 - $\beta_{jk}^Z = \lambda_{jk}^{YZ} - \lambda_{j'k}^{YZ}$
 - $\beta_{jik}^{XZ} = \lambda_{ijk}^{XZ} - \lambda_{ij'k}^{XZ}$
- loglinear(XY, YZ, XZ) $\Leftrightarrow Y \sim \text{logit}(X, Z)$ or multi-logit(X, Z)
 - $\ln \mu_{ijk} = \lambda_0 + \lambda_i^X + \lambda_j^Y + \lambda_k^Z + \lambda_{ij}^{XY} + \lambda_{ik}^{XZ} + \lambda_{jk}^{YZ}$
 - $\ln \frac{\pi_{ijk}}{\pi_{ij'k}} = \beta_{j0} + \beta_{ji}^X + \beta_{jk}^Z$
 - $\beta_{j0} = \lambda_j^Y - \lambda_{j'}^Y$
 - $\beta_{ji}^X = \lambda_{ij}^{XY} - \lambda_{ij'}^{XY}$
 - $\beta_{jk}^Z = \lambda_{jk}^{YZ} - \lambda_{j'k}^{YZ}$
- Given $X \perp\!\!\!\perp Z|Y$, loglinear(XY, YZ) $\Leftrightarrow Y \sim \text{logit}(X, Z)$ or multi-logit(X, Z)
 - $\ln \mu_{ijk} = \lambda_0 + \lambda_i^X + \lambda_j^Y + \lambda_k^Z + \lambda_{ij}^{XY} + \lambda_{jk}^{YZ}$
 - $\ln \frac{\pi_{ijk}}{\pi_{ij'k}} = \beta_{j0} + \beta_{ji}^X + \beta_{jk}^Z$
 - $\beta_{j0} = \lambda_j^Y - \lambda_{j'}^Y$
 - $\beta_{ji}^X = \lambda_{ij}^{XY} - \lambda_{ij'}^{XY}$
 - $\beta_{jk}^Z = \lambda_{jk}^{YZ} - \lambda_{j'k}^{YZ}$

Demo II

- Data “UCBAdmissions” (included in R default Package “datasets”) is on applicants to graduate school at Berkeley for the six largest departments in 1973 classified by admission and sex.
 - Admit: Admitted, Rejected
 - Gender: Male, Female
 - Dept: A, B, C, D, E, F