

Review of Tuesday

The Poisson distⁿ

- applied to counts - X

$$\mu = E(X) \quad P(X=x) = \frac{\mu^x e^{-\mu}}{x!}$$

$$\text{Var}(X) = \mu \quad \hat{\mu} = \bar{X} \quad \text{- sample } n$$

$$\text{s.e.}(\hat{\mu}) = \sqrt{\frac{\bar{X}}{n}} \quad \text{note sometimes } n=1$$

usual asymptotic procedures

$$\text{C.I.} \quad \hat{\mu} \pm 1.96 \text{ se}(\hat{\mu})$$

Large sample Test : $Z = \frac{\hat{\mu} - \mu_0}{\text{se}(\hat{\mu})} \rightarrow \text{use } \mu_0 \text{ not } \bar{X}$

\rightarrow need total count
 $= S$

exact. fla. CI

exact test ...

Inference about rates -

- Usually case that rates are more interpretable scientifically
- denominator
 - span of time
 - count of persons
 - person. time

Goodness of fit

→ set of counts $X_1 \dots X_k$

$$p_1 \dots p_k = 1 \text{ sum}$$

informal is compare $\hat{p}_i = \frac{X_i}{n}$ $n = \sum X_i$

or X_i vs $n p_i =$
obs'd exp'd
frequencies

discrepancy measure $\chi^2 = \sum \frac{(\overset{\text{obs.}}{X_i} - \overset{\text{exp}}{n p_i})^2}{n p_i}$
- Pearson's G.S.F.

If p_i are "right", χ^2
follow χ^2 distn

$$= k - V - 1 \text{ df}$$

↓
* parameters
estimated
to form p_i 's

to apply to
Poisson or variable
Poisson with infinite range → collapse
the tail
to single
category

$$G^2 = 2 \sum \text{obs} \times \log\left(\frac{\text{obs}}{\text{exp}}\right)$$

• likelihood ratio test

Comparing Rates

In example - $j = 1, 4$ age categories,

rates $\lambda_j = 1, 4$

? $\lambda_2 > \lambda_3$ - consider λ_2 / λ_3
 - relative risk . R.R.

$$\lambda_2 = \mu_2 / T_2 \quad \text{person years}$$

5×5.15

$$\lambda_3 = \mu_3 / T_3$$

5×2.67

$$\text{R.R.} = \mu_2 / \mu_3 \times \left(\frac{T_3}{T_2} \right)$$



$$\frac{x_2 / x_3 = 47 / 30$$

convenient to switch to \log

$$S = \log(\mu_2 / \mu_3) = \log \mu_2 - \log \mu_3$$

$$\hat{\theta} = \log X_2 - \log X_3 \quad \text{Var}(\hat{\theta})$$

$$= \text{Var}(\log(X_2)) + \text{Var}(\log(X_3))$$

→ non-linear

Var(f(y)) ? ← approximation in terms of Var(y)

$$\text{Var}(a+by) = b^2 \text{Var}(y)$$

$$f(y) \approx f(y_0) + f'(y_0) \times (y - y_0)$$

Let $y_0 = E(y)$

$$\text{Var}(f(y)) = \left[\frac{df}{dy} \Big|_{y=E(y)} \right]^2 \text{Var}(y)$$

- delta method

$$\frac{d(\log x)}{dx} = \frac{1}{x} \quad \text{Var}(\log(X_2)) \approx \frac{\mu_2}{(\mu_2)^2} = \frac{1}{\mu_2}$$

est'd by $\frac{1}{X_2}$

$$\text{Var}(\hat{\theta}) \approx \frac{1}{X_2} + \frac{1}{X_3}$$

$$95\% \text{ CI} = (\log 47 - \log 30) \pm 1.96 \sqrt{\frac{1}{47} + \frac{1}{30}}$$

+0.45

.46

95% CI for $\theta = -0.01$ to $.91$

95% CI for $\frac{\mu_2}{\mu_3}$ take $e^{\hat{\theta}}$ = .991

to 2.47

adjust for denominator

- multiply by $T_3/T_2 = \frac{5 \times 2.29}{5 \times 1.82}$

Conf. Int. to RR - 1.24 to 3.1

indicates the risk is at least 25% higher
w/ < 4 & possibly 3x higher

point estimate $e^{\hat{\theta}} \times T_3/T_2 = 1.96$

- neglect - hypothesis testing Test RR = 1 exact inference for RR...

I need to give references

Random Events in a Human Population

The following data gives incidence of leukemia in males 0-14 during in a five year period in Birmingham, U.K. (source: Breslow and Day, Vol. I).

T

$\lambda = 2$

denominator - Person years

Ages	Cases	Pop. (100k)	Years	Rate (per 100k person-years)
<1	2	0.45	5	0.88
1 thru 4	47	1.82	5	5.15
5 thru 9	30	2.28	5	2.63
10 thru 14	13	2.03	5	1.28

5x.4.

2.2!

100k

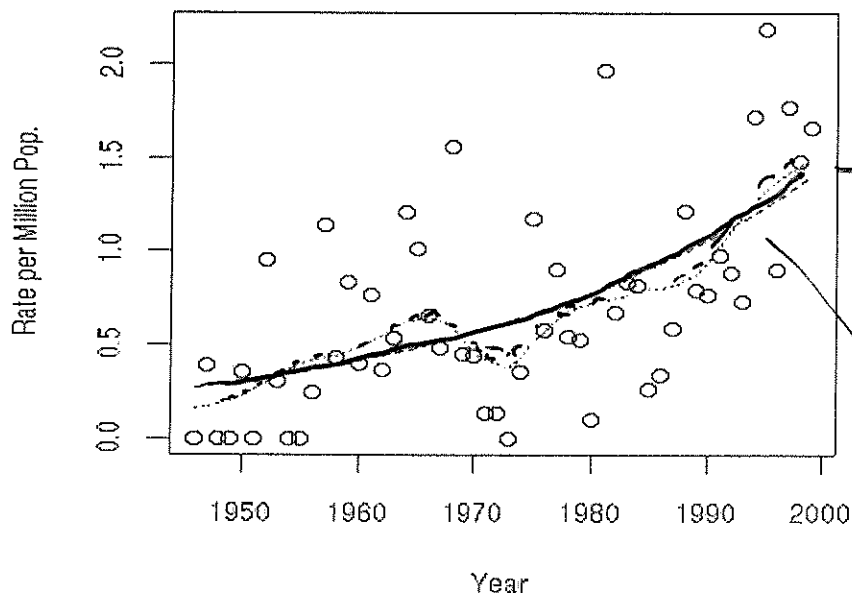
$$\lambda = \frac{\lambda}{T}$$

Poisson Regression

The following plot portrays the rate of shark attacks (per million population) recorded in Florida from 1945-2000. A non-parametric smoother has been added, as well as the fitted line from a Poisson Regression model.

$$\lambda = \frac{\mu}{T}$$

Shark Attacks



scatter plot smoother - supsmu

poisson regression line

→ describes a pattern of
consistent pattern of proportion
or (%) increase year to year
(exponential)