

$$= n \left( \frac{\frac{1-\theta}{\theta}}{(1-\theta)^2} + \frac{1}{\theta^2} \right) = n \left( \frac{1}{\theta(1-\theta)} + \frac{1}{\theta^2} \right)$$

$$= n \left( \frac{1}{\theta^2(1-\theta)} \right) \propto \text{Beta}(0, \frac{1}{2})$$

$$p(\theta) \propto \sqrt{J(\theta)} = \sqrt{n \frac{1}{\theta^2(1-\theta)}} \propto \theta^{-1} (1-\theta)^{-1/2}$$

Jeffrey's:  $\theta \sim \text{Beta}(0, \frac{1}{2}) \Rightarrow \alpha=0, \beta=\frac{1}{2}$   
(improper)

Similar to Wilson estimate  
3/21/17

$$X_1, \dots, X_n | \theta \stackrel{iid}{\sim} \text{Geom}(\theta) = (1-\theta)^x \theta$$

$$\theta \sim \text{Beta}(\alpha, \beta)$$

$$\Rightarrow \theta | X_1, \dots, X_n \sim \text{Beta}(n+\alpha, \sum X_i + \beta)$$

$\alpha$ : # of pseudotrials  
 $\beta$ : # of pseudofailures

$\alpha, \beta$  have Haldane  
(prior of complete ignorance)

$$\hat{\theta}_{MSE} = \frac{n+\alpha}{n+\alpha+\sum X_i+\beta} \quad \left\{ \text{Is there a shrinkage interpretation?} \right.$$

Haldane  $\theta \sim \text{Beta}(0, 0) \Rightarrow \hat{\theta}_{MSE} = \frac{n}{n+\sum X_i} = \frac{1}{1+\frac{\sum X_i}{n}} = \frac{1}{1+\bar{x}} = \hat{\theta}_{MLE}$

Laplace  $\theta \sim \text{Beta}(1, 1) \Rightarrow \hat{\theta}_{MSE} = \frac{n+1}{n+1+\sum X_i+1} = \frac{1}{1+\frac{\sum X_i+1}{n+1}}$

Jeffrey's  $\theta \sim \text{Beta}(0, \frac{1}{2}) \Rightarrow \hat{\theta}_{MSE} = \frac{n}{n+\sum X_i+\frac{1}{2}} = \frac{1}{1+\frac{\sum X_i+\frac{1}{2}}{n}}$

Weighted arithmetic average

$$\begin{aligned} \frac{1}{\hat{\theta}_{MSE}} &= \frac{n + \alpha + \sum x_i + \beta}{n + \alpha} = \frac{\alpha + \beta}{n + \alpha} \cdot \frac{\alpha}{\alpha} + \frac{n + \sum x_i}{n + \alpha} \cdot \frac{n}{n} \\ &= \underbrace{\frac{\alpha + \beta}{n + \alpha}}_{\frac{1}{E(\theta)}} \cdot \underbrace{\frac{\alpha}{n + \alpha}}_p + \underbrace{\frac{n + \sum x_i}{n + \alpha}}_{\frac{1}{\hat{\theta}_{MLE}}} \cdot \underbrace{\frac{n}{n}}_{(1-p)} \quad \left. \vphantom{\frac{1}{\hat{\theta}_{MLE}}} \right\} \text{weighted harmonic average} \\ \bar{x} &= \frac{1}{n} \sum x_i \end{aligned}$$

$$\frac{1}{\bar{x}_H} = \frac{1}{n} \sum \frac{1}{x_i} \quad \left\{ \text{harmonic sample average} \right.$$

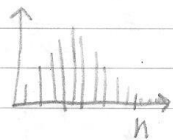
No credible regions + hypothesis test b/c same thing

$$P(X^* | X) = \int_{\Theta} p(x^* | \theta) \underbrace{p(\theta | X)}_{dP(\theta | X)} d\theta$$

$$= \int_0^1 ((1-\theta)^{x^*} \theta) \frac{1}{B(n + \alpha, \sum x_i + \beta)} \theta^{n + \alpha - 1} (1-\theta)^{\sum x_i + \beta - 1} d\theta$$

$\int f(x) dx$  Riemann  
stieljes integral

$$\begin{aligned} &= \frac{1}{B(n + \alpha, \sum x_i + \beta)} \int_0^1 \theta^{n + \alpha + 1 - 1} (1-\theta)^{x^* + \sum x_i + \beta - 1} d\theta \\ &= \frac{B(n + \alpha + 1, x^* + \sum x_i + \beta)}{B(n + \alpha, \sum x_i + \beta)} := \text{Beta Geometric}(n + \alpha, \sum x_i + \beta) \end{aligned}$$



Over-dispersed binomial  
(Beta-Binomial)



(Beta-Geom)

Looks different

Ex-log,  $X^2$  Gamma

$$X_1, \dots, X_n | \theta \stackrel{i.i.d.}{\sim} \text{Bin}(r, \theta) = \binom{x+r-1}{x} (1-\theta)^x \theta^r$$

$\theta \sim \text{Beta}(\alpha, \beta)$   
Everything is same except extra  $r$  term

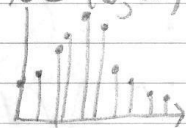
$$X \sim \text{Bin}(n, \theta) := \binom{n}{x} \theta^x (1-\theta)^{n-x}$$

If  $n$  is large,  $\theta$  is small but they're "pushed" to be  $\lambda = n\theta$

$$\lim_{n \rightarrow \infty} \frac{n!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} = \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} \frac{n \cdot n \cdot \dots \cdot n}{(n-x)(n-x-1) \dots (n-x+1)} \cdot \frac{1}{e^{-\lambda}}$$

$$= \frac{\lambda^x}{x!} e^{-\lambda} = \text{Poisson}(\lambda)$$

$$\text{Supp}(X) = \{0, 1, \dots\} = \mathbb{N}_0$$



$$E(X) = \lambda$$

$$\text{Var}(X) = \lambda$$

$$X | \theta \sim \text{Poisson}(\theta) = \frac{e^{-\theta} \theta^x}{x!}$$

Conjugacy

$$\theta | X \sim$$

$$P(\theta | X) \propto P(X | \theta) P(\theta) = \frac{e^{-\theta} \theta^x}{x!} P(\theta) \propto e^{-\theta} \theta^x P(\theta)$$

$$P(\theta) \propto e^{-b\theta} \theta^a$$

$$\Rightarrow P(\theta) = \frac{b^{a+1}}{\Gamma(a+1)} e^{-b\theta} \theta^a$$

$$\theta \sim \text{Gamma}(\alpha, \beta) := \frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\beta\theta} \theta^{\alpha-1}$$

$$\text{Supp}(\theta) = (0, \infty)$$

Support of prior has to equal sample space of likelihood

Param. Space  $\alpha > 0, \beta > 0$



$$E(\theta) = \alpha \quad \text{Var}(\theta) = \frac{\alpha}{\beta^2}$$

$$\text{Mode}(\theta) = \frac{\alpha-1}{\beta} \quad \text{if } \alpha \geq 1$$

TEST Med[ $\theta$ ] = gamma(0.5,  $\alpha, \beta$ )

$$P(\theta|X) \propto P(X|\theta) P(\theta) \propto e^{-\theta} \theta^x$$

$$X|\theta \sim \text{Poisson}(\theta) = \frac{e^{-\theta} \theta^x}{x!}$$

$$\theta \sim \text{Gamma}(\alpha, \beta)$$

$$\theta|X \sim \text{Gamma}(x+\alpha, \beta+1)$$

$$P(\theta|x) \propto P(X|\theta) P(\theta) \propto \left( \frac{e^{-\theta} \theta^x}{x!} \right) \left( \frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\beta\theta} \theta^{\alpha-1} \right)$$

$$\propto e^{-\theta} \theta^x e^{-\beta\theta} \theta^{\alpha-1} = e^{-(\beta+1)\theta} \theta^{x+\alpha-1}$$

$$X_1, \dots, X_n | \theta \stackrel{\text{i.i.d.}}{\sim} \text{Poisson}(\theta)$$

$$\theta | X_1, \dots, X_n \sim \text{Gamma}(x+\alpha, \beta+1)$$

$$P(\theta|x) \propto P(X|\theta) P(\theta) \propto \left( \prod_{i=1}^n \frac{e^{-\theta} \theta^{x_i}}{x_i!} \right) \left( \frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\beta\theta} \theta^{\alpha-1} \right)$$

$$= \frac{e^{-n\theta} \theta^{\sum x_i}}{\prod x_i!} \cdot \frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\beta\theta} \theta^{\alpha-1} \propto e^{-(\beta+n)\theta} \theta^{\sum x_i + \alpha - 1}$$

$$\propto \text{Gamma}(\sum x_i + \alpha, \beta + n)$$

$$\theta | X_1, \dots, X_n \sim \text{Gamma}(\sum x_i + \alpha, \beta + n)$$

$\alpha = \# \text{ of pen tests}$

$\beta = \# \text{ of pen tests}$

# Prior of indifference - Final

$$\Rightarrow \hat{\theta}_{MMSE} = \frac{\sum x_i + \alpha}{\beta + n}, \quad \hat{\theta}_{MAE} = \text{qgamma}(0.5, \sum x_i + \alpha, \beta + n)$$

$$\hat{\theta}_{MAP} = \frac{\sum x_i + \alpha - 1}{\beta + n} \quad \text{if } \sum x_i + \alpha \geq 1$$

Laplace  $\theta \sim U, \dots$  Can we do this?  
 Can't put uniform on infinite support  
 $\int_0^\infty c d\theta \neq 1$

$P(\theta) \propto 1$  Clearly improper

$$P(\theta|X) \propto P(X|\theta)P(\theta) \propto e^{-n\theta} \theta^{\sum x_i} P(\theta) \propto e^{-n\theta} \theta^{\sum x_i}$$

$$\propto \text{Gamma}(\sum x_i + n, 1) \Rightarrow \theta \sim \text{Gamma}(0, 0)$$

"really bad"  $\rightarrow$  "Haldane" = "Laplace"  $\leftarrow$   
 $\rightarrow$  This is improper when  $\sum x_i = 0$

$$L(\theta; x) = \prod_{i=1}^n \frac{e^{-\theta} \theta^{x_i}}{x_i!} = \frac{e^{-n\theta} \theta^{\sum x_i}}{n! x_i!}$$

$$l(\theta; x) = -n\theta + \sum x_i \ln(\theta) - \ln\left(\prod_{i=1}^n x_i!\right)$$

$$l'(\theta; x) = -n + \frac{\sum x_i}{\theta} \stackrel{\text{set}}{=} 0 \Rightarrow \frac{\sum x_i}{\theta} = n$$

$$\Rightarrow \hat{\theta}_{MLE} = \frac{\sum x_i}{n} = \bar{x}$$

$$l''(\theta; x) = -\frac{\sum x_i}{\theta^2}, \quad I(\theta) = E[-l''(\theta; x)] = \frac{E(\sum x_i)}{E(\theta^2)}$$

Jeffrey's Prior

$$\propto \sqrt{I(\theta)} = \sqrt{\frac{n}{\theta}}$$

$$= \frac{E(\sum x_i)}{\theta^2} = \frac{\sum E(x_i)}{\theta^2}$$

$$\propto \sqrt{\frac{1}{\theta}} = \theta^{-1/2}$$

$$= \frac{\sum \theta}{\theta^2} = \frac{n\theta}{\theta^2} = \frac{n}{\theta}$$

$$\propto \text{Gamma}(\frac{1}{2}, 0)$$

(improper)

Can we do this?

$$I(\theta) = \text{Var}_\theta[S(\theta; x)]$$

$$\hat{\theta}_{MMSE} = \frac{\sum x_i}{\beta + n} \cdot \frac{n}{n} + \frac{\alpha}{\beta + n} \cdot \frac{\beta}{\beta} = \frac{\sum x_i \cdot n}{\underbrace{n}_{\theta_{MLE} \text{ 1-p}} \underbrace{\beta + n}_{E(\theta) \text{ p}}} + \frac{\alpha \cdot \beta}{\underbrace{\beta}_{\theta_{MLE} \text{ 1-p}} \underbrace{\beta + n}_{E(\theta) \text{ p}}}$$

$$P(X^*|X) = \int_{\Theta} P(X^*|\theta) P(\theta|X) d\theta$$

$$n^* = 1 \quad = \int_0^\infty \left( \frac{e^{-\theta} \theta^{x^*}}{x^*!} \right) \left( \frac{\beta^{\alpha'}}{\Gamma(\alpha')} e^{-\beta'\theta} \theta^{\alpha'-1} \right) d\theta$$