

Lecture 11•• Uninformative Priors (objective, weak)

indifference /  $\theta \sim U(0,1) = \text{Beta}(1,1) \xrightarrow{X} \theta|X \sim \text{Beta}(1+x, 1+n-x)$

$$\hat{\theta}_{\text{MMSE}} = \frac{x+1}{n+2} \quad \left( \begin{array}{l} \text{Wilson} \\ \text{Estimate} \end{array} \right)$$

zero knowledge / [Haldane, 1932]

$$\theta \sim \text{Beta}(0,0)$$

\* Not a proper prior

Not a legal distribution

$\alpha, \beta = 0 \notin \text{Parameter space}$

$$\xrightarrow{X} \theta|X \sim \text{Beta}(x, n-x)$$

$$\hat{\theta}_{\text{MMSE}} = \frac{x}{n} = \hat{\theta}_{\text{MLE}}$$

\* Posterior is Proper if  $x > 0$  and  $n-x > 0$

•• Informative Priors:  $\theta \sim \text{Beta}(\alpha, \beta)$  where  $\alpha, \beta$  are "large"

ex: Batting Averages:  $BA = \frac{\# \text{ hits}}{\# \text{ at bat}} = \frac{x}{n} \mid X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Bern}(\theta)$

BA is a  $\hat{\theta}_{\text{MLE}} = \frac{x}{n}$ ; weak when  $n$  is small

ex: if  $n=2$ ;  $x=0, 1, 2$  all give unrealistic values

\*\* use  $\hat{\theta}_{\text{MMSE}} = \frac{x+\alpha}{n+\alpha+\beta}$ , Shrinkage towards  $\frac{\alpha}{\alpha+\beta}$  with weight  $\left(\frac{\alpha+\beta}{\alpha+\beta+n}\right)$

• What prior?  $\theta \sim U(0,1)$ ? No b/c then it shrink to 0.5 (not realistic)

• Use "Empirical Bayes"  $\theta$  is based on previous observations.

steps

① def. sample of previous players

② def.  $N_0$ , min. sample for each

③  $\hat{\theta}_{\text{MLE}}$  for all previous players

④ Plot distribution

"fit" data, find  $\alpha, \beta$

ex/

"pseudotrials"

$$N_0 = 500$$

$$\alpha = 78.7 \quad \beta = 224.8$$

$$\hat{\theta}_{\text{MMSE}} = \frac{x+78.2}{n+303.5}$$

$$\hookrightarrow \frac{\alpha+\beta}{\alpha+\beta+n} \cdot \frac{\alpha}{\alpha+\beta}$$

$$e = \frac{\alpha+\beta}{\alpha+\beta+n} = \frac{303.5}{303.5+2} = 99.5\%$$

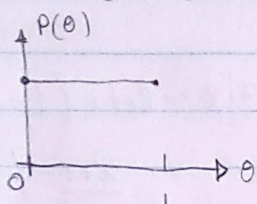
$$\frac{\alpha}{\alpha+\beta} = \frac{78.7}{303.5} = 0.258$$

(Shrinkage Estimate) =  $\rho\%$  towards  $\frac{\alpha}{\alpha+\beta} E(\theta)$   
(for new data - Not shown)



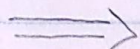
- uninformatives viewed differently

$$\theta \sim U(0,1)$$

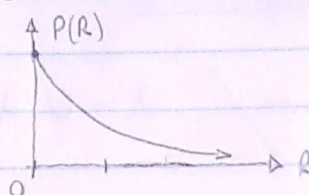


$$\theta \in (0,1)$$

transformation



$$\text{Odds : } R := \frac{\theta}{1-\theta} \in (0, \infty)$$



$$\lim_{\theta \rightarrow 1} \frac{\theta}{1-\theta} = \infty$$

- \* this signifying smaller odds are more likely
- \* prior indifference for R is not possible

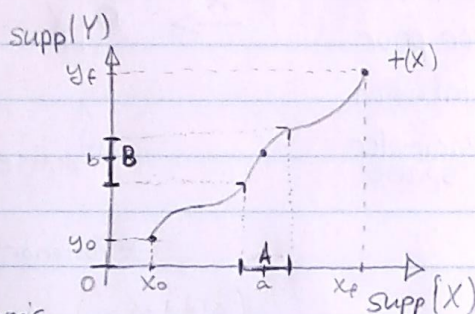
ex 2 R.V.s  $X, Y$

w/ PDF  $f_x, f_y$

- $f_x$  known

$$Y = t(X)$$

$t$  is monotonic  
(invertible)



$$P(X \in A) \approx f_x(a) A$$

$$P(Y \in B) \approx f_y(b) B$$

if  $A > 0, B > 0$  but  
are really small  
 $A = |dx|; B = |dy|$

$$f_y(b) \approx f_x(a) \left| \frac{dx}{dy} \right|$$

ex

for odds for:  $R = t(\theta) = \frac{\theta}{1-\theta} \Rightarrow \theta = t^{-1}(R) = \frac{R}{R+1}$

$$f_R(r) = f_\theta(t^{-1}(r)) \left| \frac{d}{dr} [t^{-1}(r)] \right|$$

$$= f_\theta\left(\frac{r}{r+1}\right) \left| \frac{d}{dr} \left(\frac{r}{r+1}\right) \right|$$

$$= f_\theta(\theta) \cdot \left| \frac{d}{dr} (r(r+1)^{-1}) \right|$$

$$= 1 \cdot \frac{1}{(r+1)^2}$$

since  $\theta \sim (0,1)$

$$f_R(r) = 1/(r+1)^2$$

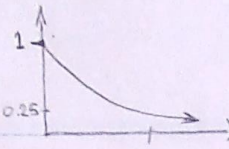
\* Checkable:  $\int_0^\infty \frac{1}{(r+1)^2} dr = \left. \frac{r}{r+1} \right|_0^\infty = 1 - 0 = 1$

$$f_y(y) = f_x(t^{-1}(y)) \left| \frac{d}{dy} [t^{-1}(y)] \right|$$

\* transformation of Variable  
formula



• Odds  $f_R(r) = \frac{1}{(r+1)^2}$  : 0.25



- Protocol to Choose Uniform prior that's the same under transformation?

let  $\phi$  be the transformation =  $t(\theta)$   
 $1:1$ , monotonic

|               |                   |             |           |
|---------------|-------------------|-------------|-----------|
| $P(X \theta)$ | $\longrightarrow$ | $P(\theta)$ | Transform |
| $P(X \phi)$   | $\longrightarrow$ | $P(\phi)$   |           |

"Protocol"

again:  $p(\phi) = p(t^{-1}(\phi)) \left| \frac{d}{d\phi} [t^{-1}(\phi)] \right|$

**\*\* Jeffery's Prior**: to obtain priors from each likelihood  
 [Requires ① kernels ② Fisher information]

~~Kernels~~  $P(\theta|X) = \frac{P(X|\theta) P(\theta)}{P(X)} \propto P(X|\theta) P(\theta)$

$f(x; \theta) \propto \{g(x; \theta)\}$  The Kernel

• if  $\exists c \in \mathbb{R}$   $f(x; \theta) \propto c g(x; \theta)$

\* finding  $c$ :  $\int_{\text{supp}(X)} f(x) dx = 1$

so  $\int_{\text{supp}(X)} c \cdot g(x) dx = \int_{\text{supp}(X)} f(x) dx \Rightarrow \frac{1}{c} = \int_{\text{supp}(X)} \underbrace{g(x; \theta)}_{\text{finite and positive}} dx$

ex: Binomial  $(n, \theta)$   
 $P(X; \theta) = \frac{n!}{x!(n-x)!} \theta^x (1-\theta)^{n-x} (1-\theta)^n (1-\theta)^{-x}$  at a fixed  $n$

•  $P(X; \theta) \propto \underbrace{\frac{1}{x!(n-x)!}}_{\text{kernel of Binomial}} \left( \frac{\theta}{1-\theta} \right)^x$