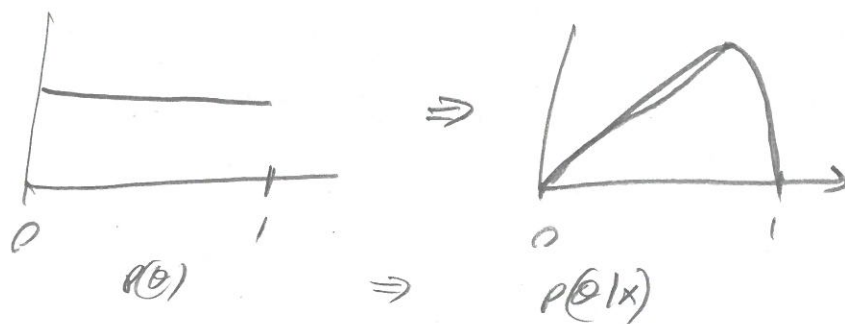


lec 5 2/16/17 prob 3 & 1

$Z = \text{Bernoulli}$, $X = (0, 1, 1)$, $\theta \sim U(0, 1)$

$$P(\theta|x) = 12\theta^2(1-\theta)$$



let's generalize...

$Z = \text{Bernoulli}$ $X = X_1, \dots, X_n$, $\theta \sim U(0, 1)$

$$P(\theta|x) = \frac{P(x|\theta)P(\theta)}{P(x)} = \frac{P(x|\theta)P(\theta)}{\int P(x|\theta)P(\theta)d\theta}$$

$$P(x|\theta) = \prod_{i=1}^n P(x_i|\theta) = \prod_{i=1}^n \theta^{x_i} (1-\theta)^{1-x_i} = \theta^{\sum x_i} (1-\theta)^{n - \sum x_i} = \theta^x (1-\theta)^{n-x}$$

\Rightarrow prob. only dep. on $\sum x_i \dots$ let $x := \sum x_i$ from now on

$$\Rightarrow P(x|\theta) = \frac{\theta^x (1-\theta)^{n-x}}{\int_0^1 \theta^x (1-\theta)^{n-x} d\theta}$$

Gamma function

\uparrow
called β
from β -function

$$B(\alpha, \beta) := \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt$$

$$p(\theta|x) = \frac{1}{\beta(x+1, n-x+1)} \theta^x (1-\theta)^{n-x} = \text{Beta}(x+1, n-x+1)$$

A new r.v.!

$$X \sim \text{Beta}(\alpha, \beta) := \frac{1}{\beta(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} \quad \text{supp}[X] = [0, 1]$$

$f(x)$

Valid PDF?

$$\int_{\text{supp}(X)} f(x) dx = 1 \Rightarrow \int_0^1 \frac{1}{\beta(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx = \frac{1}{\beta(\alpha, \beta)} \beta(\alpha, \beta) = 1 \quad \checkmark$$

For what values of α, β does this hold? I.e. ... what is the parameter space?

What is $\beta(\alpha, \beta) = \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt$ finite?

only if $\alpha > 0$ and $\beta > 0$ \Rightarrow param space

It can be shown that

$$\beta(\alpha, \beta) = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha+\beta)} \quad \text{where} \quad \Gamma(x) := \int_0^{\infty} e^{-t} t^{x-1} dt$$

the gamma function

where the gamma function is the extension of the factorial function to all \mathbb{R}^+ 's

Property $\Gamma(x) = (x-1)! = (x-1)(x-2)! = (x-1)\Gamma(x-1)$

Using this property...

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

$$E(X) = \int_0^1 x \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx$$

$$= \int_0^1 \frac{1}{B(\alpha, \beta)} x^{\alpha} (1-x)^{\beta-1} dx$$

$$= \frac{B(\alpha+1, \beta)}{B(\alpha, \beta)} = \frac{\frac{\Gamma(\alpha+1)\Gamma(\beta)}{\Gamma(\alpha+\beta+1)}}{\frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}} = \frac{\alpha \Gamma(\alpha)\Gamma(\beta)}{(\alpha+\beta) \Gamma(\alpha+\beta)} = \boxed{\frac{\alpha}{\alpha+\beta}}$$

Var(X) ... calc on the (similar)

$$\text{Mode}(X) = \arg\max \left\{ \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} \right\}$$

$$= \arg\max \left\{ x^{\alpha-1} (1-x)^{\beta-1} \right\} = \arg\max \left\{ (\alpha-1) \ln(x) + (\beta-1) \ln(1-x) \right\}$$

take derivative - set = 0

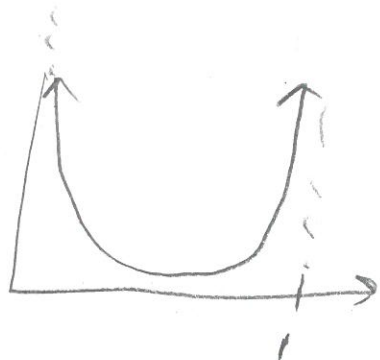
$$\begin{aligned} x^{\alpha-1} (\beta-1) (1-x)^{\beta-2} + (1-x)^{\beta-1} (\alpha-1) x^{\alpha-2} &= 0 \\ -x^{\alpha+1} (\beta-1) (1-x)^{\beta-2} + (1-x)^{\beta-1} (\alpha-1) &= 0 \\ -x^{\alpha+1} (\beta-1) + (1-x)^{\beta-1} (\alpha-1) &= 0 \Rightarrow \frac{x^{\alpha+1}}{(1-x)^{\beta+1}} = \frac{\beta-1}{\alpha-1} \end{aligned}$$

take deriv set = 0

$$\frac{\alpha-1}{x} + \frac{\beta-1}{1-x} = 0 \Rightarrow \frac{\alpha-1}{x} = \frac{\beta-1}{1-x} \Rightarrow \frac{1}{x} - 1 = \frac{\beta-1}{\alpha-1}$$

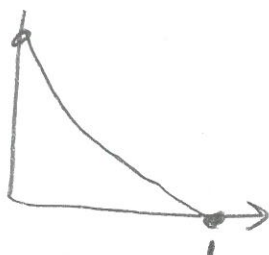
$$\Rightarrow \frac{1}{x} = \frac{\beta+\alpha-2}{\alpha-1} \Rightarrow x = \frac{\alpha-1}{\alpha+\beta-2} \quad \text{only for } \alpha > 1, \beta > 1 \text{ and local min!}$$

Shapes of the beta r.v. density:

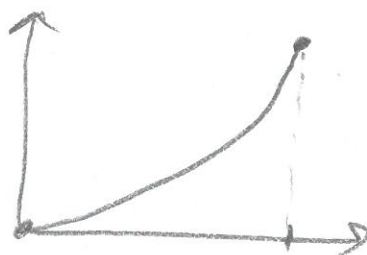


$$\alpha = \beta = 0.5$$

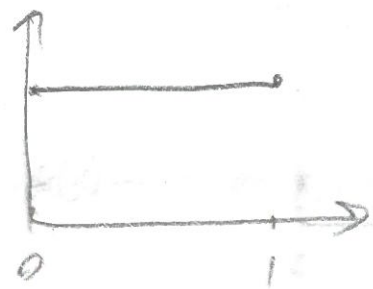
"arcsin distr"



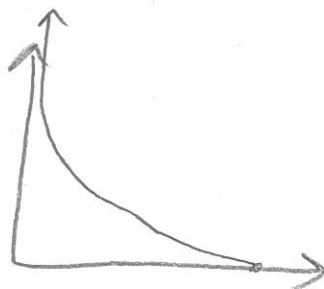
$$\alpha = 1, \beta = 3$$



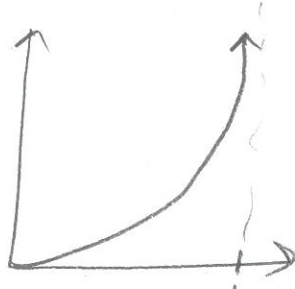
$$\alpha = 5, \beta = 1$$



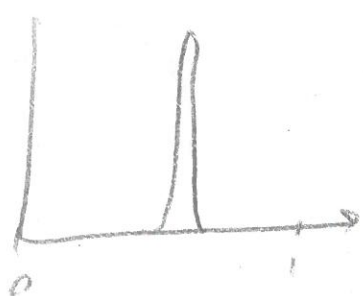
$$\alpha = \beta = 1$$



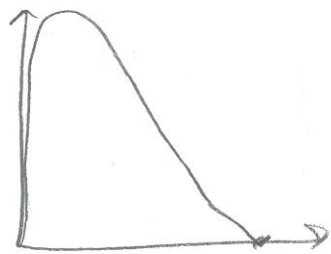
$$\alpha = 0.99, \beta = 3$$



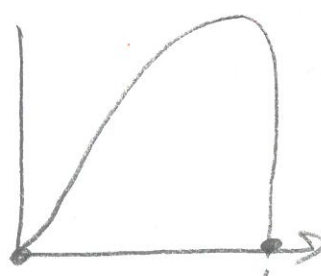
$$\alpha = 5, \beta = 0.99$$



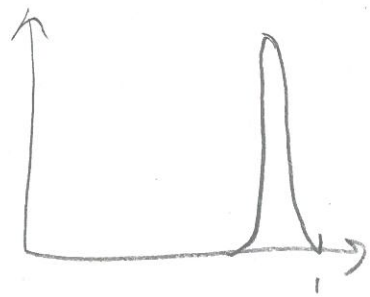
$$\alpha = \beta = 100$$



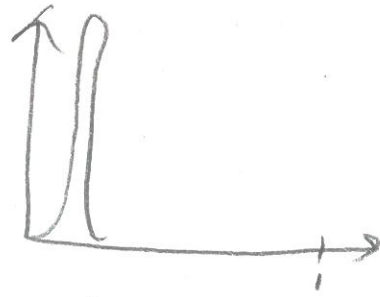
$$\alpha = 1.01, \beta = 3$$



$$\alpha = 5, \beta = 1.01$$



$$\alpha = 100, \beta = 1$$



$$\alpha = 1, \beta = 100$$

Let's return to Bayesian...

We saw before for X_1, \dots, X_n from $\mathcal{K} = \text{Bernoulli}$, only the sum of the x 's mattered. Let's consider just X as a binomial

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

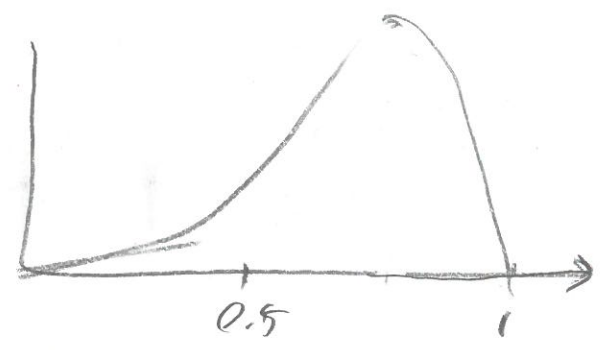
$$\text{let } \theta \sim U(0,1)$$

$$p(\theta|x) = \frac{p(x|\theta)p(\theta)}{p(x)} = \frac{\binom{n}{x} \theta^x (1-\theta)^{n-x}}{\int_0^1 \binom{n}{x} \theta^x (1-\theta)^{n-x}} = \text{Beta}(\underbrace{x+1}_{\alpha'}, \underbrace{n-x+1}_{\beta'})$$

Syme!

$$\text{if } n=10, x=7$$

$$\theta|x \sim \text{Beta}(8, 4)$$



$$\hat{\theta}_{\text{MAP}} = \frac{\alpha'-1}{\alpha'+\beta'-2} = \frac{8-1}{8+4-2} = \frac{7}{10} = 0.7$$

What about the average $\theta|x$?

Ag

$$E[\theta|x] = \frac{\alpha'}{\alpha'+\beta'} = \frac{8}{8+4} = 0.66$$

$$E[\theta] = \frac{\alpha}{\alpha+\beta} = 0.5$$

↑ what's this? Prior mean!!

What about the median $\theta|x$?

$$\text{Med}[\theta|x] \dots \text{no closed form in R} \dots q_{\text{beta}}(0.5, 8, 4) \approx 0.676 \dots$$

Three different ways of pt. estimation.

$$\hat{\theta}_{MAP} = \arg\max\{P(\theta|x)\} \quad \text{posterior mode}$$

$$\hat{\theta}_{MSE} = E[\theta|x] \quad \text{posterior expectation / mean}$$

$$\hat{\theta}_{MAE} = \text{med}[\theta|x] \quad \text{posterior median}$$

Turns out... $\hat{\theta}_{MSE}$ minimizes squared error loss

$$\text{" } \arg\min E[(\theta - \hat{\theta}_{MSE})^2]$$

$\hat{\theta}_{MAE}$ minimizes absolute error loss

$$\text{" } \arg\min E[|\theta - \hat{\theta}_{MAE}|]$$

We will be using all 3. Default is $\hat{\theta}_{MSE} = E[\theta|x]$

but $\hat{\theta}_{MAE}$ is easiest to get! No need to compute $P(x)$!

New idea... priors...

$$\theta \sim U(0,1) = \text{Beta}(1,1)$$

Why not let

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

where I choose α, β to reflect my prior information!

Forget how I choose α, β for now...

$$\begin{aligned}
 p(\theta|x) &= \frac{p(x|\theta)p(\theta)}{p(x)} = \frac{\binom{n}{x} \theta^x (1-\theta)^{n-x} \left(\frac{1}{B(\alpha, \beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1} \right)}{\int_0^1 \binom{n}{x} \theta^x (1-\theta)^{n-x} \left(\frac{1}{B(\alpha, \beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1} \right) d\theta} \\
 &= \frac{\theta^{x+\alpha-1} (1-\theta)^{n-x+\beta-1}}{\int_0^1 \theta^{x+\alpha-1} (1-\theta)^{n-x+\beta-1} d\theta} \\
 &= \frac{1}{B(x+\alpha, n-x+\beta)} \theta^{(x+\alpha)-1} (1-\theta)^{(n-x+\beta)-1} \\
 &= \text{Beta}(x+\alpha, n-x+\beta) \quad \text{Another Beta!}
 \end{aligned}$$

prior beta, posterior beta \Rightarrow conjugacy seems like a "natural" choice!

Def: p is the "conjugate prior" for F if the posterior also has same distribution

\Rightarrow Beta is the conjugate prior for the binomial (or the Bernoulli)

$$\hat{\theta}_{MAP} = \frac{x+\alpha-1}{\alpha+\beta+n-2}, \quad \hat{\theta}_{MLE} = \frac{x+\alpha}{n+\alpha+\beta}, \quad \hat{\theta}_{MAP} = \text{beta}(0.5, x+\alpha, n-x+\beta) \quad \text{use log prior}$$

prediction distr.

After seeing n Bernoulli trials / one Bernoulli trial. What is the distr. of the next observation (the only)?

$$p(x^*|x)?$$