

Example 1

$$X \sim \text{Beta}(a, b) \quad X \in [0, 1] \quad p(x) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1}$$

$$E(X) = \frac{\alpha}{\alpha+\beta} \quad \text{Var}(X) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$$

$$\theta \sim \text{Beta}(a=2, b=20) \quad \text{Mode}(X) = \frac{\alpha-1}{\alpha+\beta-2}$$

$$\begin{aligned} p(\theta|y) &= \frac{p(\theta, y)}{p(y)} \propto p(\theta, y) \quad (\text{only solve up to a proportionality constant}) \\ &= \theta^{a-1} (1-\theta)^{b-1} \theta^y (1-\theta)^{n-y} \\ &= \theta^{a+y-1} (1-\theta)^{n-y+b-1} \end{aligned}$$

$$\propto \text{Beta}(a+y, b+(n-y))$$

• more uncertainty in my prior, model(θ) shifted to left, increased belief that $\theta < 0.10$; data has provided evidence that value of θ is smaller than previously thought.

Exercise 1

$$\theta \sim \text{Unif}(0.05, 0.20) \quad p(\theta) = \frac{1}{0.15} \quad (\text{flat Beta}(1,1) \text{ truncated on interval } (0.05, 0.20))$$

$$\begin{aligned} p(\theta|y) &\propto p(\theta, y) = \theta^y (1-\theta)^{n-y} \propto \text{Beta}(1+y, 1+n-y) \\ &= \theta^{1+y-1} (1-\theta)^{1+n-y-1} \end{aligned}$$

proportional to a Beta distribution truncated at this interval

can express posterior mean as a linear combination of prior and sample mean, (quantifies shrinkage towards prior mean) when n is small?

shrinkage factor: $\frac{a+b}{a+b+n}$

prior assumptions can have an influence on posterior inference

• contour graphs - as n increases, posterior summary becomes more certain. high degree of belief. low prior belief, generally 90% or more certain. $n \rightarrow 100$ → reliability when n is small? → increasing w means more weight on prior, lines do not go to $y=0$