

lec 14 3/30/17 Math 391

$$X_1, \dots, X_n | \theta, \sigma^2 \sim N(\theta, \sigma^2)$$

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θ known, σ^2 unknown. $\Rightarrow \hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \theta)^2 = \frac{SSE}{n}$. Bayesian inference?

$$P(\sigma^2 | X, \theta) \propto P(X | \theta, \sigma^2) P(\sigma^2 | \theta)$$

$$= \left(\prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(X_i - \theta)^2} \right) P(\sigma^2 | \theta)$$

$$= \left(\frac{1}{\sqrt{2\pi}} \right)^n (\sigma^2)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \theta)^2} P(\sigma^2 | \theta)$$

$$\propto (\sigma^2)^{-\frac{n}{2}} e^{-\frac{\frac{1}{2}\hat{\sigma}^2}{\sigma^2}} P(\sigma^2 | \theta)$$

kernel of InverseGamma $\left(\frac{n}{2} - 1, \frac{\frac{1}{2}\hat{\sigma}^2}{2} \right)$.

matches kernel...

$$\Rightarrow \sigma^2 | \theta \sim \text{InverseGamma}(\alpha, \beta)$$

$$P(\sigma^2 | X, \theta) \propto (\sigma^2)^{-\frac{n}{2}} e^{-\frac{\frac{1}{2}\hat{\sigma}^2}{\sigma^2}} \left(\frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\frac{\beta}{\sigma^2}} (\sigma^2)^{-\alpha-1} \right)$$

$$\propto (\sigma^2)^{-\frac{n}{2} - \alpha - 1} e^{-\frac{(\frac{1}{2}\hat{\sigma}^2 + \beta)}{\sigma^2}}$$

$$\propto \text{InverseGamma}\left(\frac{n}{2} + \alpha, \frac{\frac{1}{2}\hat{\sigma}^2 + \beta}{2}\right)$$

Recall:

$$Y \sim \text{InverseGamma}(\alpha, \beta) :=$$

$$\frac{\beta^\alpha}{\Gamma(\alpha)} e^{-\frac{\beta}{y}} y^{-\alpha-1}$$

However... we usually don't use α, β . We use a parameterization that mirrors the ^{empirical} prior.

if $\alpha = \frac{n_0}{2}$, $\beta = \frac{n_0 \sigma_0^2}{2} \Rightarrow \sigma^2 | \theta \sim \text{InverseGamma}\left(\frac{n_0}{2}, \frac{n_0 \sigma_0^2}{2}\right)$

$$\Rightarrow P(\sigma^2 | X, \theta) = \text{InverseGamma}\left(\frac{n+n_0}{2}, \frac{\frac{1}{2}\hat{\sigma}^2 + n_0 \sigma_0^2}{2}\right)$$

Inference

n_0 : # prior trials f... $Y_1, \dots, Y_{n_0} \stackrel{\text{iid}}{\sim} N(\theta, \sigma^2)$

σ_0^2 : prems $\hat{\sigma}_{MLE}^2$ for σ^2 , $\sigma_0^2 := \frac{1}{n_0} \sum_{i=1}^{n_0} (Y_i - \theta)^2$

$$\Rightarrow n_0 \sigma_0^2 = \sum_{i=1}^{n_0} (Y_i - \theta)^2 = SSE_0$$

$$\Rightarrow \sigma^2 | X, \theta \sim \text{Invgamma} \left(\underbrace{\frac{n + n_0}{2}}_{\alpha'}, \underbrace{\frac{SSE + SSE_0}{2}}_{\beta'} \right)$$

$$\hat{\sigma}_{MLE}^2 = \frac{\frac{n \hat{\sigma}_{MLE}^2 + n_0 \sigma_0^2}{2}}{\frac{n + n_0}{2} - 1} = \frac{n \hat{\sigma}_{MLE}^2 + n_0 \sigma_0^2}{n + n_0 - 2}$$

Shrinkage ... easy ... Hw...

$$\hat{\sigma}_{MAP}^2 = \frac{n \hat{\sigma}_{MLE}^2 + n_0 \sigma_0^2}{n + n_0 + 2}$$

$$\hat{\sigma}_{MAP}^2 = \text{Jitgamma} \left(0.5, \frac{n + n_0}{2}, \frac{n \hat{\sigma}_{MLE}^2 + n_0 \sigma_0^2}{2} \right)$$

CR's, hypothesis tests follow...

Uniform prior? If $n_0 = 0$ as if saying "nothing"

$$\sigma^2 | \theta \sim \text{Invgamma}(0, 0) \quad \text{Improper...}$$

But...

$$P(\sigma^2 | X, \theta) = \text{Invgamma} \left(\frac{n}{2}, \frac{n \hat{\sigma}^2}{2} \right) \quad \text{always proper...}$$

$$\hat{\sigma}^2_{nmSE} = \frac{\frac{n \hat{\sigma}^2}{2}}{\frac{n}{2} - 1} = \frac{\frac{n \hat{\sigma}^2}{2}}{n-2} = \frac{1}{n-2} \sum (x_i - \theta)^2 \approx \hat{\sigma}^2_{MLE}$$

What is another uniform prior?

$$\sigma^2 | \theta \sim \text{InverseGamma}(-2, 0) \quad n_0 = 2, \sigma_0^2 = 0, \text{ weird...}$$

$$\Rightarrow \sigma^2 | x_1, \dots, x_n, \theta \sim \text{InverseGamma}\left(\frac{n+2}{2}, \frac{n \hat{\sigma}^2}{2}\right)$$

$$\hat{\sigma}^2_{nmSE} = \frac{\frac{n \hat{\sigma}^2}{2}}{\frac{n+2}{2} - 1} = \hat{\sigma}^2_{MLE}$$

How about Jeffreys prior?

$$P(\sigma^2 | \theta) \propto \sqrt{I(\sigma^2)}$$

$$\ell'(\sigma^2; X, \theta) = -\frac{n}{2\sigma^2} + \frac{1}{2(\sigma^2)^2} \text{SSE} = -\frac{n}{2}(\sigma^2)^{-1} + \frac{\text{SSE}}{2}(\sigma^2)^{-2}$$

$$\ell''(\sigma^2; X, \theta) = \frac{n}{2}(\sigma^2)^{-2} - \text{SSE}(\sigma^2)^{-3}$$

$$I(\sigma^2) = E[-\ell''(\sigma^2; X, \theta)] = E\left[-\frac{n}{2}(\sigma^2)^{-2} + \text{SSE}(\sigma^2)^{-3}\right] = -\frac{n}{2}(\sigma^2)^{-2} + (\sigma^2)^{-3} E[\text{SSE}]$$

$$E[\text{SSE}] = E\left[\sum_{i=1}^n (x_i - \theta)^2\right] = \sum_{i=1}^n E[(x_i - \theta)^2] \overset{\text{due to iid}}{=} n E[(X - \theta)^2] \overset{\text{def of Var}}{=} n \text{Var}[X] = n \sigma^2$$

$$\text{Recall: } X \sim N(\theta, \sigma^2)$$

$$\Rightarrow I(\sigma^2) = -\frac{n}{2}(\sigma^2)^{-2} + (\sigma^2)^{-3} (n \sigma^2) = -\frac{n}{2}(\sigma^2)^{-2} + n(\sigma^2)^{-2} = \left(n - \frac{n}{2}\right)(\sigma^2)^{-2} \quad \frac{n}{2}$$

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$$\Rightarrow \text{Jeffreys } P(\sigma^2 | \theta) \propto \sqrt{\frac{1}{2}(\sigma^2)^{-2}} \propto (\sigma^2)^{-1} = \text{InverseGamma}(0, 0)$$

$\Rightarrow \psi_0 = 0$

Proper? $\int_0^{\infty} \frac{1}{\sigma^2} d\sigma^2 = \infty \dots \text{NO!}$

not prop. distr. ... not covered

Problem 2 T

Find \downarrow

$$X_1, \dots, X_n | \theta, \sigma^2 \stackrel{\text{iid}}{\sim} \mathcal{N}(\theta, \sigma^2)$$

but now both θ, σ^2 unknown...

$$\underbrace{P(\theta, \sigma^2 | X)}_{\text{2-dim posterior}} \propto \underbrace{P(X | \theta, \sigma^2)}_{\text{sam. likelihood}} \underbrace{P(\theta, \sigma^2)}_{\text{2-dim prior}}$$

$$= \left(\prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(X_i - \theta)^2} \right) P(\theta, \sigma^2)$$

$$= \left(\frac{1}{\sqrt{2\pi}} \right)^n (\sigma^2)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \theta)^2} P(\theta, \sigma^2)$$

$$\propto (\sigma^2)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \theta)^2} P(\theta, \sigma^2)$$

Merge gamma? No... θ is a free variable!

This is a 2-dim distr. ... you have seen those before...

Consider

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$$\sum_{i=1}^n (x_i - \theta)^2 = \sum_{i=1}^n (x_i - \bar{x} + \bar{x} - \theta)^2$$

$$= \sum_{i=1}^n (x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \theta) + (\bar{x} - \theta)^2 = \sum_{i=1}^n (x_i - \bar{x})^2 + 2 \sum_{i=1}^n (x_i \bar{x} - x_i \theta - \bar{x}^2 + \bar{x} \theta) + n(\bar{x} - \theta)^2$$

Note: $s^2 := \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$ Variance estimator if both θ, σ^2 unknown

$$= (n-1)s^2 + 2 \left(\bar{x} \sum_{i=1}^n x_i - \theta \sum_{i=1}^n x_i - \sum_{i=1}^n \bar{x}^2 + \theta \sum_{i=1}^n x_i \right) + n(\bar{x} - \theta)^2$$

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

$$= (n-1)s^2 + n(\bar{x} - \theta)^2$$

$$\propto p(x|\theta, \sigma^2) p(\theta, \sigma^2)$$

$$p(\sigma^2, \theta | x) \propto (\sigma^2)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \left((n-1)s^2 + n(\bar{x} - \theta)^2 \right)} p(\theta, \sigma^2)$$

$$= (\sigma^2)^{-\frac{n}{2}} e^{-\frac{(n-1)s^2/2}{\sigma^2}} e^{-\frac{1}{2\frac{\sigma^2}{n}} (\bar{x} - \theta)^2} p(\theta, \sigma^2)$$

$$\propto \text{Normal Inv Gamma} \left(n = \bar{x}, \lambda = n, \alpha = \frac{n}{2} + 1, \beta = \frac{(n-1)s^2}{2} \right)$$

4 param dist!

$p(\theta, \sigma^2)$ should be Normal Inv Gamma as well

However... we will only consider the uninformative Jeffreys prior

$$p(\theta, \sigma^2) = p(\theta | \sigma^2) p(\sigma^2) \propto (1) \left(\frac{1}{\sigma^2} \right) = \frac{1}{\sigma^2}$$

$$\begin{aligned}
 P(\theta, \sigma^2 | X) &\propto (\sigma^2)^{-\frac{n}{2}} e^{-\frac{(n-1)s^2/2}{\sigma^2}} e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2} ((\sigma^2)^{-1}) \\
 &= (\sigma^2)^{-\frac{n}{2} - 1} e^{-\frac{(n-1)s^2/2}{\sigma^2}} e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2}
 \end{aligned}$$

$$\propto \text{InvGamma}(\bar{X}, n, \frac{n}{2}, \frac{(n-1)s^2}{2})$$

How to simulate from this distribution?

$$P(\theta, \sigma^2 | X) = P(\theta | X, \sigma^2) P(\sigma^2 | X) \quad \text{Bayes Rule}$$

$$\begin{aligned}
 \Rightarrow P(\theta | X, \sigma^2) &= \frac{P(\theta, \sigma^2 | X)}{P(\sigma^2 | X)} \propto P(\theta, \sigma^2 | X) \\
 &= (\sigma^2)^{-\frac{n}{2} - 1} e^{-\frac{(n-1)s^2/2}{\sigma^2}} e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2} \\
 &\propto e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2} \\
 &\propto N(\bar{X}, \frac{\sigma^2}{n})
 \end{aligned}$$

$$\begin{aligned}
 \Rightarrow P(\sigma^2 | X) &= \frac{P(\theta, \sigma^2 | X)}{P(\theta | X, \sigma^2)} \propto \frac{(\sigma^2)^{-\frac{n}{2} - 1} e^{-\frac{(n-1)s^2/2}{\sigma^2}} e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2}}{\frac{1}{\sqrt{2\pi} \sigma^2} e^{-\frac{1}{2\sigma^2} (\bar{X} - \theta)^2} \cdot \frac{(\sigma^2)^{-\frac{n}{2}}}{-\frac{n}{2} + \frac{1}{2} - 1}} \\
 &\propto \frac{(\sigma^2)^{-\frac{n}{2} - 1} e^{-\frac{(n-1)s^2/2}{\sigma^2}}}{(\sigma^2)^{-\frac{1}{2}}} = (\sigma^2)^{-\frac{n}{2} - \frac{1}{2}} e^{-\frac{(n-1)s^2/2}{\sigma^2}}
 \end{aligned}$$

$$\propto \text{InvGamma}(\frac{n-1}{2}, \frac{(n-1)s^2}{2})$$