

The t-distribution (Student)

- $Z \sim N(0,1)$
- $V \sim \chi^2_k$
- Z and V are independent

Then $t = \frac{Z}{\sqrt{V/k}}$ satisfies $f(t) \propto \left(1 + \frac{t^2}{k}\right)^{-\frac{k+1}{2}}$ [look up constant]

Observation: when k is large, this is almost Normal.

$$\lim_{k \rightarrow \infty} \left(1 + \underbrace{\frac{k+1}{k}}_1 \cdot \frac{t^2}{k+1}\right)^{-\frac{k+1}{2}} = e^{-t^2/2}$$

$\lim_{n \rightarrow \infty} \left(1 + \frac{x}{n}\right)^n \rightarrow e^x$

$$E[t] = 0 \quad \text{if } k > 1 \quad \text{Var}[t] = \begin{cases} \frac{k}{k-2} & k > 2 \\ \infty & 1 < k \leq 2 \end{cases}$$

Fisher proved the following for Normal samples

- $(n-1) \frac{s^2}{\sigma^2} \sim \chi_{n-1}^2$ $\frac{S}{\sigma^2} \sim \chi_{n-1}^2$ where $S = \sum (x_i - \bar{x})^2$ (*)
- \bar{x} and s^2 are independent (only Normal satisfies this)
- (we know) $\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$

So $\frac{\bar{x} - \mu}{s/\sqrt{n}} \sim t_{n-1}$ (t -distributed with degree $n-1$)

Remark: The text in Note 8 contains proofs for the first two facts when $n=2$.

(*) This is different than previous definition $S = \sum (x_i - \mu)^2$ when μ was known.

Two teams revisited (unknown σ^2)

$$\frac{(n-1)s_x^2}{\sigma_x^2} \sim \chi_{n-1}^2$$

$$\frac{(m-1)s_y^2}{\sigma_y^2} \sim \chi_{m-1}^2$$

$$\text{So } \frac{(n-1)s_x^2}{\sigma_x^2} + \frac{(m-1)s_y^2}{\sigma_y^2} \sim \chi_{n+m-2}^2 \quad [\text{sum of ind. } \chi^2]$$

$$\text{From before } \frac{\bar{x} - \bar{y} - (\mu_x - \mu_y)}{\sqrt{\sigma_x^2/n + \sigma_y^2/m}} \sim N(0,1)$$

Assuming $\sigma_x^2 = \sigma_y^2$ [though both are unknown], testing $\mu_x = \mu_y$

we get

$$\frac{\bar{x} - \bar{y}}{\sqrt{\left(\frac{1}{n} + \frac{1}{m}\right) \frac{(n-1)s_x^2 + (m-1)s_y^2}{n+m-2}}} \sim t_{n+m-2}$$

[Note 8 has an example]

Conjugate Prior

$$x_i | \mu, \sigma^2 \sim N(\mu, \sigma^2) \text{ are i.i.d.}$$

$$\frac{S_0}{\sigma^2} \sim \chi^2_{k_0}$$

$$\mu | \sigma^2 \sim N\left(\beta, \frac{\sigma^2}{n_0}\right)$$

* Making μ & σ independent does not lead to conjugate

We want to prove posterior has the same form.

$$\frac{S'}{\sigma^2} \Big| x_1, \dots, x_n \sim \chi^2_K \quad (\text{Find } S' \text{ and } K)$$

$$\mu | \sigma^2, x_1, \dots, x_n \sim N\left(\beta', \frac{\sigma^2}{n'}\right) \quad (\text{Find } \beta' \text{ and } n').$$

First, let's find $\mu | \sigma^2, x_1, \dots, x_n$

$$\underbrace{f(\mu | x_1, \dots, x_n, \sigma^2)}_{?} \propto \underbrace{f(x_1, \dots, x_n | \mu, \sigma^2)}_{\text{Normal}} \cdot \underbrace{f(\mu | \sigma^2)}_{\text{Normal}}$$

This is the same setting where μ is unknown but σ^2 known

$$\mu | \sigma^2, x_1, \dots, x_n \sim N \left(\frac{\sigma^2 \beta / n + \frac{\sigma^2}{n_0} \bar{x}}{\sigma^2 / n + \sigma^2 / n_0}, \frac{\sigma^2 / n \cdot \sigma^2 / n_0}{\sigma^2 / n + \sigma^2 / n_0} \right)$$

$$\mu | \sigma^2, x_1, \dots, x_n \sim N \left(\frac{n_0 \beta + n \bar{x}}{n + n_0}, \frac{\sigma^2}{n + n_0} \right) = N \left(\beta', \frac{\sigma^2}{n'} \right)$$

- β' is weighted average of β and \bar{x}
- $n = \#$ observations
- $n_0 = \#$ "observations" in prior.

What about σ^2 . Let's start with

$$\begin{aligned} f(\mu, \frac{1}{\sigma^2}) &\propto \underbrace{\frac{1}{\sigma} e^{-\frac{n_0(\mu-\beta)^2}{2\sigma^2}}}_{f(\mu|\sigma^2)} \cdot \underbrace{\left(\frac{1}{\sigma^2}\right)^{\frac{k_0}{2}-1} e^{-\frac{S_0}{2\sigma^2}}}_{f(\frac{1}{\sigma^2})} \\ &= \left(\frac{1}{\sigma^2}\right)^{\frac{k_0+1}{2}-1} e^{-\frac{n_0(\mu-\beta)^2 + S_0}{2\sigma^2}} \end{aligned}$$

Observe that $f(\frac{1}{\sigma^2} | \mu) \propto f(\mu, \frac{1}{\sigma^2})$

$$\text{so } \frac{S_0 + n_0(\mu-\beta)^2}{\sigma^2} | \mu \sim \chi^2_{k_0+1}$$

(Another way of
looking at prior
for σ^2)

Let's find $\frac{1}{\sigma^2} \mid \mu, x_1, \dots, x_n$

$$\underbrace{f\left(\frac{1}{\sigma^2} \mid x_1, \dots, x_n, \mu\right)}_{?} \propto \underbrace{f(x_1, \dots, x_n \mid \sigma^2, \mu)}_{\text{Normal}} \underbrace{f\left(\frac{1}{\sigma^2} \mid \mu\right)}_{\chi^2}$$

This is similar to the setting where σ^2 unknown but μ known.

$$\frac{\sum (x_i - \mu)^2 + S_0 + n_0(\mu - \beta)^2}{\sigma^2} \mid \mu, x_1, \dots, x_n \sim \chi^2_{k_0 + 1 + n}$$

the χ^2 prior (from previous page)

We showed that the posteriors for $\mu \mid \sigma^2$ and $\sigma^2 \mid \mu$ have the same form as their corresponding prior.

Let's find the unconditional posterior for σ^2 .

$$f\left(\frac{1}{\sigma^2} \mid x_1 \dots x_n\right) = \int_{-\infty}^{+\infty} f\left(\frac{1}{\sigma^2}, \mu \mid x_1 \dots x_n\right) d\mu \propto \int_{-\infty}^{+\infty} f(x_1 \dots x_n \mid \mu, \sigma^2) f\left(\mu, \frac{1}{\sigma^2}\right) d\mu$$

$$\int_{-\infty}^{+\infty} \left(\frac{1}{\sigma^2}\right)^{\frac{k_0 + n + 1}{2} - 1} e^{-\frac{\sum (x_i - \mu)^2 + S_0 + n_0(\mu - \beta)^2}{2\sigma^2}} d\mu$$

Note: $\sum (x_i - \mu)^2 = S + n(\bar{x} - \mu)^2$ where $S = \sum (x_i - \bar{x})^2$

$$\left(\frac{1}{\sigma^2}\right)^{\frac{k_0 + n}{2} - 1} e^{-\frac{S + S_0}{2\sigma^2}} \int_{-\infty}^{+\infty} \frac{1}{\sigma} e^{-\frac{n(\bar{x} - \mu)^2 + n_0(\mu - \beta)^2}{2\sigma^2}} d\mu$$

$\underbrace{\hspace{15em}}_{\text{cte. } e^{-\frac{\square}{2\sigma^2}}}$

Posterior for σ^2 [unconditional]

$$\frac{S'}{\sigma^2} \sim \chi^2_k$$

where:

$$k = n + k_0$$

$$\beta' = \frac{n_0 \beta + n \bar{x}}{n + n_0}$$

(same β' as before)

$$n' = n_0 + n$$

$$S' = S + S_0 + n_0 \beta^2 + n \bar{x}^2 - n' \beta'^2$$

Back to μ . What's the unconditional prior?

$$f(\mu) \propto \int_0^{\infty} \left(\frac{1}{\sigma^2}\right)^{\frac{k_0+1}{2}-1} e^{-\frac{n_0(\mu-\beta)^2 + S_0}{2\sigma^2}} d\frac{1}{\sigma^2}$$

Change of variable: $t = \frac{n_0(\mu-\beta)^2 + S_0}{2\sigma^2}$

$$f(\mu) \propto \int_0^{\infty} \left[\frac{2t}{n_0(\mu-\beta)^2 + S_0} \right]^{\frac{k_0+1}{2}-1} e^{-t} \frac{2}{n_0(\mu-\beta)^2 + S_0} dt$$

$$\propto \Gamma\left(\frac{k_0+1}{2}\right) \left[n_0(\mu-\beta)^2 + S_0 \right]^{-\frac{k_0+1}{2}}$$

$$\propto \left[1 + \left(\frac{\mu-\beta}{\sqrt{S_0/n_0 k_0}} \right)^2 / k_0 \right]^{-\frac{k_0+1}{2}} \equiv \left(1 + \frac{x^2}{k} \right)^{-\frac{k+1}{2}}$$

So $(\mu-\beta) / \sqrt{S_0/n_0 k_0} \sim t_{k_0}$ in the prior

We automatically conclude that.

$$\frac{\mu - \beta'}{\sqrt{S'/n'k}} \sim t_k \text{ in the posterior}$$

Finally  PHEW!

Reference Priors

$$n_0 = 0 \quad S_0 = 0 \quad k_0 = -1$$

we get

$$\left(\beta' = \bar{x}, S' = S = \sum (x_i - \bar{x})^2, s^2 = \frac{S}{n-1} \right)$$

Prior:

$$f(\sigma^2) \propto \frac{1}{\sigma^2}$$

$$f(\mu) \propto 1 \quad (\text{Uniform in } [-\infty, \infty])$$

improper

Posterior:

$$\frac{S}{\sigma^2} \sim \chi^2_{n-1}$$

$$\frac{\mu - \bar{x}}{s/\sqrt{n}} \sim t_{n-1}$$

what we knew

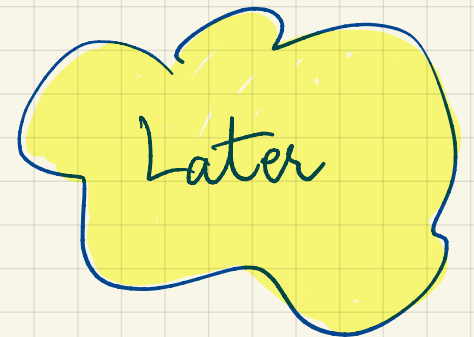
What happens if:

$$x_i | \mu, \sigma^2 \sim N(\mu, \sigma^2)$$

$$\frac{S_0}{\sigma^2} \sim \chi_{k_0}^2$$

$$\mu \sim N(\beta, \tau^2)$$

(Semi-conjugate prior)



Skip last page of Note 8.