

M378K: October 10th, 2025.

The F-Distribution.

Let Y_1 and Y_2 be two independent, χ^2 distributed r.v.s w/ $\boxed{df=1}$.

For both Y_1 and Y_2 the pdf is

$$f_Y(y) = \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}} \cdot \mathbb{1}_{(0,\infty)}(y)$$

Define $W = \frac{Y_2}{Y_1}$, i.e., $W = g(Y_1, Y_2)$ where $g(y_1, y_2) = \frac{y_2}{y_1}$

Goal: Density of W , i.e., f_W !

Start by figuring out the cdf F_W .

$$w > 0: F_W(w) = \mathbb{P}[W \leq w] = \mathbb{P}\left[\frac{Y_2}{Y_1} \leq w\right]$$

$$= \mathbb{P}[Y_2 \leq w \cdot Y_1]$$

$$= \int_0^\infty \int_0^{w \cdot y_1} f_{Y_1, Y_2}(y_1, y_2) dy_2 dy_1$$

$$= \int_0^\infty \int_0^{w \cdot y_1} \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} \cdot \frac{1}{\sqrt{2\pi y_2}} e^{-\frac{y_2}{2}} dy_2 dy_1$$

$$= \int_0^\infty \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} \left(\int_0^{w \cdot y_1} \frac{1}{\sqrt{2\pi y_2}} e^{-\frac{y_2}{2}} dy_2 \right) dy_1$$

$$F_W(w) = \int_0^\infty \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} \cdot \overset{F_{Y_2}(w \cdot y_1)}{F_{Y_2}(w y_1)} dy_1$$

$$f_w(w) = \frac{d}{dw} \int_0^{\infty} \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} \cdot F_{Y_2}(w y_1) dy_1$$

$$\begin{aligned} f_w(w) &= \frac{d}{dw} F_w(w) \\ &= \lim_{h \rightarrow 0} \frac{F_w(w+h) - F_w(w)}{h} \end{aligned}$$

$$f_w(w) = \int_0^{\infty} \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} f_{Y_2}(w \cdot y_1) y_1 dy_1$$

$$f_w(w) = \int_0^{\infty} \frac{1}{\sqrt{2\pi y_1}} e^{-\frac{y_1}{2}} \frac{1}{\sqrt{2\pi w y_1}} e^{-\frac{w y_1}{2}} \cdot y_1 dy_1$$

$$f_w(w) = \frac{1}{2\pi\sqrt{w}} \int_0^{\infty} e^{-\frac{y_1}{2} - \frac{w y_1}{2}} dy_1$$

$$f_w(w) = \frac{1}{2\pi\sqrt{w}} \int_0^{\infty} e^{-\frac{(1+w)y_1}{2}} dy_1$$

$$\left(-\frac{2}{1+w} e^{-\frac{(1+w)y_1}{2}} \right)_{y_1=0}^{\infty} = \frac{2}{1+w}$$

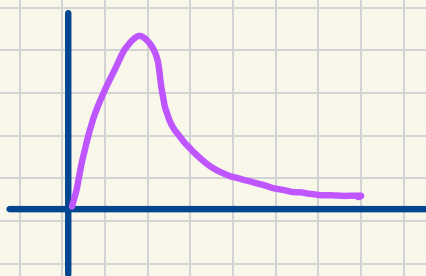
$$f_w(w) = \frac{1}{2\pi\sqrt{w}} \cdot \frac{2}{1+w} = \frac{1}{\pi\sqrt{w}(1+w)} \mathbb{1}_{(0,\infty)}(w)$$

is the density of $F(1,1)$, i.e.,

the F-distribution w/

1 numerator df
and 1 denominator df.

$$F(\nu_1, \nu_2) \stackrel{(d)}{=} \frac{\frac{\chi^2(\nu_1)}{\nu_1}}{\frac{\chi^2(\nu_2)}{\nu_2}}$$



M378K Introduction to Mathematical Statistics

Problem Set #9

Moment generating functions.

Definition 9.1. The k^{th} moment of a random variable Y taken about the origin is defined as $\mathbb{E}[Y^k]$ provided that the expectation exists. We write

$$\mu_k = \mathbb{E}[Y^k]$$

when there is no ambiguity about the random variable in question.

Remark 9.2. μ_k is also referred to as the k^{th} raw moment.

Remark 9.3. In particular, $\mu_1 = \mu$ happens to be the **mean** of the random variable Y .

Definition 9.4. The k^{th} central moment of a random variable Y is defined as $\mathbb{E}[(Y - \mu)^k]$ provided that the expectation exists. We write

$$\mu_k^c = \mathbb{E}[(Y - \mu)^k]$$

when there is no ambiguity about the random variable in question.

Remark 9.5. μ_k is also referred to as the k^{th} moment of a random variable Y taken about its mean.

Definition 9.6. The moment-generating function (mgf) m_Y for a random variable Y is defined as

$$m_Y(t) = \mathbb{E}[e^{tY}]$$

for all t for which the above expectation exists. In fact, we say that the moment-generating function **exists** if there exists a positive number b such that $m_Y(t)$ is finite for all t such that $|t| \leq b$.

Problem 9.1. How much is $m_Y(0)$?

$$m_Y(0) = \mathbb{E}[e^{0 \cdot Y}] = 1$$

$$\begin{aligned} \frac{d}{dt} e^{t \cdot Y} &= e^{t \cdot Y} \cdot \frac{d}{dt} (t \cdot Y) \\ &= e^{t \cdot Y} \cdot Y \end{aligned}$$

chain rule

Remark 9.7. On the choice of terminology ...

Step 1.

$$\begin{aligned} \frac{d}{dt} m_Y(t) &= \frac{d}{dt} \mathbb{E}[e^{t \cdot Y}] = \mathbb{E}\left[\frac{d}{dt} e^{t \cdot Y}\right] \\ &= \mathbb{E}[Y \cdot e^{t \cdot Y}] \end{aligned}$$


Step 2.

$$m_Y'(0) = \mathbb{E}[Y \cdot \underbrace{e^{0 \cdot Y}}_1] = \mathbb{E}[Y] = \mu_Y$$

Step 3.

$$\frac{d}{dt} \left(\frac{d}{dt} m_Y(t) \right) = \frac{d^2}{dt^2} m_Y(t) = \frac{d}{dt} \mathbb{E}[Y e^{t \cdot Y}] = \mathbb{E}[Y^2 e^{t \cdot Y}]$$

Step 4.

$$m_Y''(0) = \mathbb{E}[Y^2], \text{ i.e., the 2nd moment}$$


Step 5. What do you suspect the **generalization** of the above would be?

$$m_Y^{(k)}(0) = \mathbb{E}[Y^k] = \mu_k$$
