

STAT 231

October 26, 2016

Tutorial today

3-30 → Cynthia STP 105

6-00 → Banerjee, DC 1351

Video Review — Friday

Roadmap

- All definitions of interval estimation using sampling techniques.
- Chi-Squared.
- t -distribution • Application.

INTERVAL ESTIMATION

ESTIMATE

ESTIMATOR.

COVERAGE INTERVAL

CONFIDENCE INTERVAL

SAMPLING DISTRIBUTION

PIVOTAL QUANTITY

Underlying model

$$Y = (Y_1, \dots, Y_n) \sim f(y_i; \theta)$$

Sample: $\{y_1, \dots, y_n\}$ $\theta =$ unknown parameter.

ESTIMATE:

An estimate of θ ($\hat{\theta}$) is a function $g(y_1, \dots, y_n)$ which is your "best guess" of θ . ✓

MLEs are estimates:

ESTIMATES are sometimes called
POINT ESTIMATES

ESTIMATOR: is a random variable
of which your estimate is an outcome

$\hat{\theta}$: POINT ESTIMATOR.

Example: $Y_1, \dots, Y_n \sim \mathcal{G}(\mu, \sigma)$
with $\sigma = \text{known}$. We want to
estimate $\mu = \text{unknown}$.

\bar{y} = sample mean = estimate

\bar{Y} = estimator.

An estimator gives us the rule of what to calculate from our sample.

SAMPLING DISTRIBUTION

The sampling distribution of an estimator $\tilde{\theta}$ is the distribution that $\tilde{\theta}$ follows.

Example
 σ known

$$Y_1, \dots, Y_n \sim \mathcal{G}(\mu, \sigma)$$

SAMPLING DISTRIBUTION OF \bar{Y}

$$\bar{Y} \sim \mathcal{G}(\mu, \sigma/\sqrt{n})$$

PIVOTAL QUANTITY

Definition: A pivotal quantity Q is a function of Y_1, \dots, Y_n and θ such that $P(Q \geq a)$, $P(Q \leq b)$ can be computed without knowing what the value of θ is

$$\bar{Y} \sim N(\mu, \sigma/\sqrt{n})$$

$$\left(\frac{\bar{Y} - \mu}{\sigma/\sqrt{n}} \right) = Z$$

$Q =$ pivotal quantity

We use the pivotal quantity
to construct the COVERAGE
INTERVAL.

COVERAGE INTERVAL

A coverage interval is two r.v.s

$[L, U]$ such that

$$P(L \leq \theta \leq U) = 0.95$$

= or some
pre-specified
probability

CONFIDENCE INTERVAL

A confidence interval is $[l, u]$
which are estimates of L, U .
where l, u are calculated from
our sample

THE CHI-SQUARED DISTRIBUTION

1. Properties

- $W \sim \chi_n^2$ if $W = Z_1^2 + \dots + Z_n^2$

$$Z_i \sim N(0, 1)$$

Z_i 's independent.

- W takes values $(0, \infty)$

- $E(W) = n = \text{degrees of freedom}$

$$V(W) = 2n = 2 \times \text{df.}$$

• If $W_1 \sim \chi^2_{n_1}$ and $W_2 \sim \chi^2_{n_2}$

then $W_1 + W_2 \sim \chi^2_{n_1 + n_2}$

if W_1 and W_2 are
independent.

Question 1 $W_1 \sim \chi^2_1$
 $W_2 \sim \chi^2_1$
 \vdots

① $W_i \sim \chi^2(i) \quad i=1, \dots, 5$

W_i 's are independent.

Let $Y = \sum_{i=1}^5 W_i$

What distribution does Y follow?

• (a) $\chi^2(5)$ (c) $N(0, 15)$

(b) $\chi^2(15)$

d) can't say

$Y \sim \chi^2 \quad 1+2+3+4+5 = 15$

Question 2

Suppose $Y \sim G(3, 9)$

Y and
 Z indep.

$Z \sim G(0, 1)$

Let ~~X~~ $W = \left(\frac{Y-3}{9}\right)^2 + Z^2$.

What distribution does W follow?

(a) $\chi^2(1)$

(b) $\chi^2(2)$ ~~77%~~

(c) $N(0, 2)$

(d) none of the
above.

Calculation purposes $k = df$

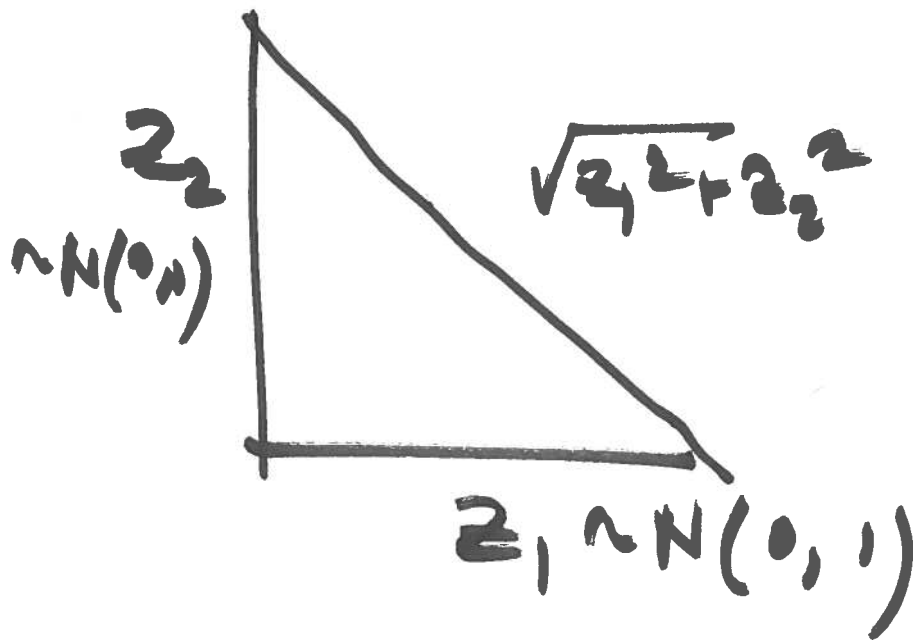
• $k = 1 \longrightarrow W = \chi^2$

• $k = 2 \longrightarrow W \sim \text{Exp}(2)$

• $k > 50 \longrightarrow W \sim G_{\chi}(k, \sqrt{2k})$

• $k \in (2, 50)$

\searrow Consult tables
to get probabilities



$$z_1^2 + z_2^2 \sim \chi^2(2)$$

CHI-SQUARED TABLE

Z-table

Chi-Square.



#s are
probabilities

Quantiles

Row = degrees of freedom.

Column = level of quantile

Row = 5

Column = 0.05

3.656

Table

$$\left. \begin{array}{l} \text{Row} = 5 \\ \text{Column} = 0.4 \end{array} \right\} \\ \text{Value} = 3.656$$

$$P(W \leq 3.656) = 0.4$$

$$\text{where } \chi^2_5 = W$$

Example: Suppose $W \sim \chi^2_{20}$

$$P(W \leq 13) = ? \quad (\text{from the table}).$$

Row = 20 = degrees of
freedom

12.443 \rightarrow 0.1

14.575 \rightarrow 0.2

The probability of $W \leq 13$

lies between 0.1 and 0.2

STUDENT'S T-distribution

Definition: A random variable T is said to follow a Student's T -distribution with n degrees of freedom. if T is a ratio of two independent r.v.s.

$$T = \frac{Z}{Y}$$

where $Z \sim N(0, 1)$

$$Y = \sqrt{\frac{W}{n}}$$

where $W \sim \chi^2_n$

~~$$Y = \sqrt{\frac{\chi^2_n}{n}}$$~~

Properties of the T-distribution

- The T-distribution is symmetric around zero

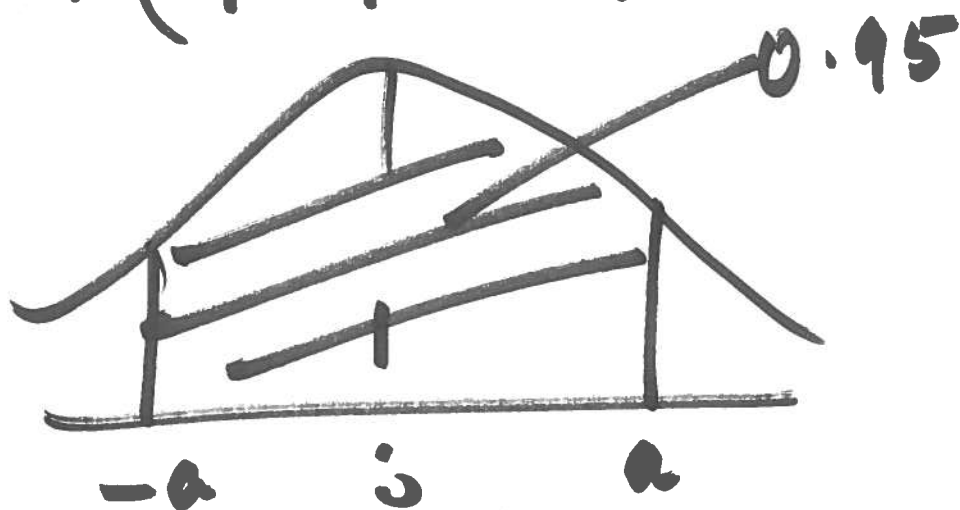
- It looks similar to Z but with fatter tails
(KURTOSIS > 3)

↑ will have more extreme obs. compared to Z .

As the degrees of freedom
 $n \rightarrow \infty$, the T distribution
approaches the Z -distribution

Example : Suppose $T \sim T_{15}$
Find a such that-

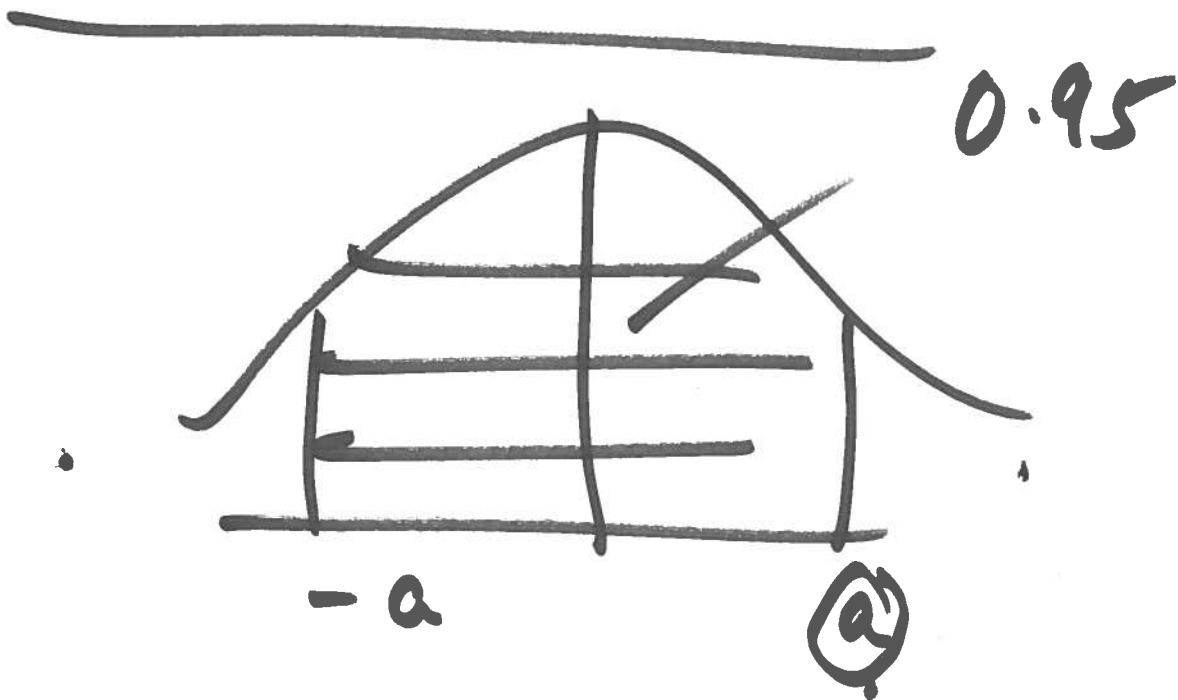
$$P(|T| \leq a) = 0.95$$



T-table

Row = df

Entres: Quantiles



Row : 15 Column = 0.975

$a = 2.1314$