

## 5.2 Useful classical facts

Theorem if  $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$   
Let  $\bar{X} = \frac{1}{n} (X_1 + \dots + X_n)$   
 $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$

Then

1.  $\bar{X}$  and  $S^2$  are independent Random variables
2.  $\bar{X} \sim N(\mu, \sigma^2/n)$
3.  $(n-1) \frac{S^2}{\sigma^2} \sim \chi_{n-1}^2$  [chi square distribution with degree of freedom (d.f.)]

Recall from Section 3-2a.

$\chi_p^2$  denotes chi square distribution with  $p$  d.f.  
the pdf:

$$f(x) = \frac{1}{\Gamma(p/2) 2^{p/2}} x^{p/2-1} e^{-x/2}, \quad x > 0$$

$$\chi_p^2 \equiv \text{Gamma}(p/2, 2).$$

Proof (Sketch)

Suffice to prove for  $\sigma = 1$ .

1. Note

$$\begin{aligned}(n-1)S^2 &= \sum_{i=2}^n (x_i - \bar{x})^2 \\&= \underbrace{(x_1 - \bar{x})^2}_{y_1} + \sum_{i=2}^n \underbrace{(x_i - \bar{x})^2}_{y_i} \\&= \left(\sum_{i=2}^n y_i\right)^2 + \sum_{i=2}^n y_i^2\end{aligned}$$

Use change of variable formula for the mapping  
 $(x_1, \dots, x_n) \mapsto (y_1, y_2, \dots, y_n)$ , where  $\begin{cases} y_1 = \bar{x} \\ y_i = x_i - \bar{x}, i \geq 2 \end{cases}$

to find that

$$f_{Y_1, \dots, Y_n}(y_1, \dots, y_n) = \left(\frac{n}{2\pi}\right)^{1/2} e^{-ny_1^2/2} \frac{n^{1/2}}{(2\pi)^{(n-1)/2}} e^{-\frac{1}{2}\left(\left(\sum_{i=2}^n y_i\right)^2 + \sum_{i=2}^n y_i^2\right)}$$

which factorizes.

Hence  $Y_1 \perp\!\!\!\perp (Y_2, \dots, Y_n)$

So  $(n-1)S^2 \perp\!\!\!\perp Y_1 = \bar{x}$ .

2. Done in previous section

$$3. \quad \text{Let } \bar{X}_k = \frac{1}{k} (X_1 + \dots + X_k) \\ S_k^2 = \frac{1}{k-1} \sum_{i=1}^k (X_i - \bar{X}_k)^2$$

Prove by induction that  $(k-1) S_k^2 \sim \chi_{k-1}^2 \quad \forall k \geq 2.$

• For  $k=2$ , 
$$S_2^2 = (X_1 - \bar{X}_2)^2 + (X_2 - \bar{X}_2)^2 \\ = \frac{1}{2} (X_1 - X_2)^2$$

$$X_1 - X_2 \sim N(0, 2) \Rightarrow \left( \frac{1}{\sqrt{2}} (X_1 - X_2) \right)^2 \sim N(0, 1)^2 \text{ which is } \chi_1^2$$

• Suppose now  $(k-2) S_{k-1}^2 \sim \chi_{k-1}^2$   
We'll be done by proving that  $(k-1) S_k^2 \sim \chi_k^2$ .  
The key is the recursive relation, which can be verified

$$(k-1) S_k^2 = (k-2) S_{k-1}^2 + \left( \frac{k-1}{k} \right) (X_k - \bar{X}_{k-1})^2$$

Now we proceed by the following sequence of arguments

$$* \quad X_k - \bar{X}_{k-1} \sim N\left(0, \frac{k}{k-1}\right) \quad \text{for } k \geq 2$$

$$\Rightarrow \sqrt{\frac{k-1}{k}} (X_k - \bar{X}_{k-1}) \sim N(0, 1)$$

$$\Rightarrow \frac{k-1}{k} (X_k - \bar{X}_{k-1})^2 \sim \chi_1^2 \quad (\text{chi-square df 1})$$

See Sec 2-1.

\* Check  $(X_k - \bar{X}_{k-1}) \perp\!\!\!\perp S_{k-1}$  (by checking joint pdf)

\* By inductive hypothesis,  $(k-2) S_{k-1}^2 \sim \chi_{k-1}^2$

\* Now  $\left\{ \begin{array}{l} \chi_{k-1}^2 \equiv \text{Gamma} \left( \frac{k-1}{2}, 2 \right) \\ \chi_1^2 \equiv \text{Gamma} \left( \frac{1}{2}, 2 \right) \end{array} \right\}$  Sec 3.2a

\* Now sum of 2 indep Gamma variables  $\left( \frac{k-1}{2}, 2 \right)$  and  $\left( \frac{1}{2}, 2 \right)$  is another Gamma  $\left( \frac{k}{2}, 2 \right)$ , (via MGF argument) which is  $\chi_k^2$ .

Hence  $(k-1) S_k^2 \sim \chi_k^2$  as we need to show.  $\square$

## Other Statistics and their distributions

- if  $X_1, X_2, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$ ,  $\mu$  unknown  
 $\bar{X}$  is a statistic that tells us about  $\mu$ .

How so? we know  $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$

this gives us a way to quantify the uncertainty about  $\mu$ , using statistic  $\bar{X}$  as an estimate.

- what if  $\sigma$  is unknown too

we may want to consider  $\frac{\bar{X} - \mu}{S/\sqrt{n}}$

since  $S$  is a statistic that can be obtained from the sample.

To be useful, we need to know the distribution of

$$T := \frac{\bar{X} - \mu}{S/\sqrt{n}}$$

Theorem  $T \sim t_{n-1}$  (Student's t distribution)  
with  $n-1$  degree of freedom

$t_p$  has the pdf

$$f_T(t) = \frac{\Gamma((p+1)/2)}{\Gamma(p/2)} \frac{1}{(pn)^{1/2}} \frac{1}{\left(1 + \frac{t^2}{p}\right)^{(p+1)/2}} \quad -\infty < t < \infty.$$

Remark

- $T$  is the Ratio of two indep. Random variables, as  
 $\bar{X} - \mu \sim \text{Normal}$ ,  $\sqrt{(n-1)S^2}$  square root of  $\chi^2_{n-1}$
- if  $n=2$  then  $T \sim$  Ratio of 2 indep normal variables  $\Rightarrow T \sim \text{Cauchy}$ .
- Proof is by change of variable formula.

For more, see Section 5.3 in the text book.