

## 5.2 Useful classical facts

**Theorem** if  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$   
 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.  $\frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}$  [ chi square distribution with  $(n-1)$  degree of freedom (d.f.) ]

Recall from Section 3-2a.

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

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

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

Proof (Sketch)

Suffice to prove for  $\sigma = 1$ .

1. Note

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

$$= \underbrace{(x_1 - \bar{x})^2}_{\text{y}_1} + \sum_{i=2}^n \underbrace{(x_i - \bar{x})^2}_{\text{y}_i}$$

$$= \left( \sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n y_i^2$$

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. Let  $\bar{X}_k = \frac{1}{k} (x_1 + \dots + x_k)$   
 $S_k^2 = \frac{1}{k-1} \sum_{i=1}^{k-1} (x_i - \bar{x}_k)^2$

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

- For  $k=2$ ,  $S_2^2 = (x_1 - \bar{x}_2)^2 + (x_2 - \bar{x}_2)^2 = \frac{1}{2} (x_1 - x_2)^2$   
 $x_1 - x_2 \sim N(0, 2) \Rightarrow \left(\frac{1}{\sqrt{2}}(x_1 - x_2)\right)^2 \sim N(0, 1)^2$  which is  $\chi_1^2$

- Suppose now  $(k-1) S_{k-1}^2 \sim \chi_{k-1}^2$ .  
 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

- \*  $x_k - \bar{x}_{k-1} \sim N(0, \frac{1}{k-1})$  for  $k \geq 2$
- $\Rightarrow \sqrt{\frac{k-1}{k}} (x_k - \bar{x}_{k-1}) \sim N(0, 1)$
- $\Rightarrow \frac{k-1}{k} (x_k - \bar{x}_{k-1})^2 \sim \chi_1^2$  (chi-square df 1)  
 See Sec 2-1.

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

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

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

\* Now sum of 2 indep Gamma variables  $(\frac{k-1}{2}, 2)$  and  $(\frac{1}{2}, 2)$  is another Gamma  $(\frac{k}{2}, 2)$ , (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  $X_1, X_2, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ ,  $\mu$  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

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

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.