

Recall  $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$

Stat 562

1-22-19

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$$= \frac{1}{n-1} \left[ (X_1 - \bar{X})^2 + \sum_{i=2}^n (X_i - \bar{X})^2 \right]$$

Also  $\sum_{i=1}^n (X_i - \bar{X}) = \sum_{i=1}^n X_i - n\bar{X} = 0$

$$\hookrightarrow X_1 - \bar{X} = - \sum_{i=2}^n (X_i - \bar{X})$$

$$\hookrightarrow S^2 = \frac{1}{n-1} \left[ \left( \sum_{i=2}^n (X_i - \bar{X}) \right)^2 + \sum_{i=2}^n (X_i - \bar{X})^2 \right]$$

That is,  $S^2$  is a function of

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$$X_2 - \bar{X}, X_3 - \bar{X}, \dots, X_n - \bar{X}$$

Assume  $X_1, \dots, X_n \sim \text{iid } N(0,1)$

Let  $Y_1 = \bar{X}$

$Y_2 = X_2 - \bar{X}$

$\vdots$

$Y_n = X_n - \bar{X}$

$X_1 = Y_1 - \sum_{i=2}^n Y_i$

$X_2 = Y_1 + Y_2$

$\vdots$

$X_n = Y_1 + Y_n$

$Y_1 = \frac{1}{n} \sum X_i$

$nY_1 = \sum X_i$

$X_i = nY_1$

$-\sum_{i=2}^n X_i$

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$$J = \begin{vmatrix} 1 & -1 & \dots & -1 \\ 1 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 1 & 0 & \dots & 0 & 1 \end{vmatrix} = n$$

$$\begin{aligned} g(y_1, \dots, y_n) &= f(x_1, \dots, x_n) |J| \\ &= n \prod_{i=1}^n \left( \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} x_i^2} \right) \\ &= \frac{n}{(2\pi)^{n/2}} e^{-\frac{1}{2} \sum_{i=1}^n x_i^2} \end{aligned}$$

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$$\begin{aligned} \sum_{i=1}^n x_i^2 &= \left( y_1 - \sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n (y_1 + y_i)^2 \\ &= y_1^2 + \left( \sum_{i=2}^n y_i \right)^2 - 2y_1 \sum_{i=2}^n y_i \\ &\quad + (n-1)y_1^2 + \sum_{i=2}^n y_i^2 + 2y_1 \sum_{i=2}^n y_i \\ &= ny_1^2 + \left( \sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n y_i^2 \end{aligned}$$

$$\text{Now } g(y_1, \dots, y_n) = \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2}(ny_1^2 + (\sum_2^n y_i)^2 + \sum_2^n y_i^2)} \quad (5)$$

$$= h(y_1) k(y_2, \dots, y_n)$$

By our factorization theorem,

$Y_1$  is independent of  $Y_2, \dots, Y_n$

$\therefore \bar{X}$  is independent of  $S^2$

Things we know about the  $\chi^2$  distribution (6)

① If  $Z \sim N(0,1)$  then  $Z^2 \sim \chi_1^2$

②  $\chi_r^2 \sim \text{Gamma}(\alpha = \frac{r}{2}, \beta = 2)$

③  $W_1, \dots, W_n \sim \text{indep } \chi_{r_i}^2 \Rightarrow \sum_{i=1}^n W_i \sim \chi_{\sum r_i}^2$

Let  $Z_1, \dots, Z_n \sim \text{iid } N(0,1)$

Fact:  $(n-1)S^2 = (n-2) \sum_{i=1}^{n-2} \underbrace{Z_i^2}_{\text{based on } Z_1, \dots, Z_{n-1}} + \frac{1}{n} (Z_n - \bar{Z}_{n-1})^2$  This is HW #15

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Theorem. If  $Z_1, \dots, Z_n \sim \text{iid } N(0,1)$ ,  
then  $(n-1)S^2 \sim \chi^2_{n-1}$

Proof: Start with  $n=2$ .

$$S^2 = \frac{1}{2}(Z_2 - Z_1)^2$$

$$Z_1 \sim N(0,1), Z_2 \sim N(0,1)$$

$$Z_2 - Z_1 \sim N(0,2)$$

$$\frac{Z_2 - Z_1}{\sqrt{2}} \sim N(0,1) \quad \therefore \frac{(Z_2 - Z_1)^2}{2} \sim \chi^2_1$$

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Induction step: Assume true for  $n$ , show true for  $n+1$

Our task states that

$$n S^2 = \underbrace{(n-1)S_n^2}_{\sim \chi^2_{n-1}} + \underbrace{\frac{1}{n+1}(Z_{n+1} - \bar{Z}_n)^2}_{\chi^2_1}$$

$Z_{n+1} \sim N(0,1)$ , indep of  $Z_1, \dots, Z_n$ , so it is indep of  $\bar{Z}_n$

$$\text{So } Z_{n+1} - \bar{Z}_n \sim N(0, 1 + \frac{1}{n})$$

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$$\frac{Z_{n+1} - \bar{Z}_n}{\sqrt{1 + \frac{1}{n}}} \sim N(0,1)$$

$$\frac{(Z_{n+1} - \bar{Z}_n)^2}{1 + \frac{1}{n}} = \frac{n(Z_{n+1} - \bar{Z}_n)^2}{n+1} \sim \chi^2_1$$

Also  $S_n^2$  is indep of  $Z_{n+1}$  and  $\bar{Z}_n$ , so

$$n S_n^2 \sim \chi^2_n$$

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Generalization:

Assume  $X_1, \dots, X_n \sim \text{iid } N(\mu, \sigma^2)$

$$\text{Then if } Z_i = \frac{X_i - \mu}{\sigma},$$

$$Z_1, \dots, Z_n \sim \text{iid } N(0,1)$$

$$\text{Our theorem} \Rightarrow (n-1)S_z^2 \sim \chi^2_{n-1}$$

$$\begin{aligned} \text{But } S_z^2 &= \frac{1}{n-1} \left( \sum_{i=1}^n Z_i^2 - n \bar{Z}^2 \right) \\ &= \frac{1}{n-1} \sum_{i=1}^n (Z_i - \bar{Z})^2 \end{aligned}$$

$$\begin{aligned} S^2 &= \frac{1}{n-1} \sum_{i=1}^n \left( \frac{X_i - \mu}{\sigma} - \frac{\bar{X} - \mu}{\sigma} \right)^2 \\ &= \frac{1}{n-1} \frac{\sum (X_i - \bar{X})^2}{\sigma^2} = \frac{S^2}{\sigma^2} \end{aligned}$$

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$$\text{So } \frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}$$

Review: With the normality assumption of  
 $X_1, \dots, X_n \sim \text{iid } N(\mu, \sigma^2),$

$$\textcircled{1} \bar{X} \sim N(\mu, \frac{\sigma^2}{n})$$

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$$\textcircled{2} \bar{X} \text{ \& } S^2 \text{ are independent}$$

$$\textcircled{3} \frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}$$

Without the normality assumption,

$$\textcircled{1} E(\bar{X}) = \mu$$

$$\textcircled{2} V(\bar{X}) = \sigma^2/n$$

$$\textcircled{3} E[S^2] = \sigma^2$$