### Lecture 20 Confidence Interval For The Variance

BIO210 Biostatistics

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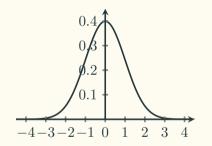


### **Interval Estimation For The Variance**

$$oldsymbol{S}^2 = rac{1}{n-1} \sum_{i=1}^n (oldsymbol{X}_i - ar{oldsymbol{X}})^2$$
 Not normally distributed!

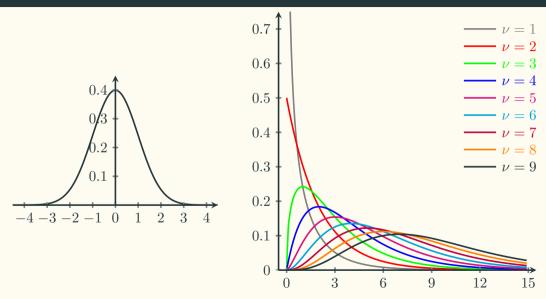
## **Chi-squared Distribution**

$$Z_1, Z_2, Z_3, ..., Z_n \sim \mathcal{N}(0, 1)$$



$$egin{aligned} m{U}_1 &= m{Z}_1^2 & m{U}_1 \sim m{\chi^2}(1) \\ m{U}_2 &= m{Z}_1^2 + m{Z}_2^2 & m{U}_2 \sim m{\chi^2}(2) \\ m{U}_3 &= m{Z}_1^2 + m{Z}_2^2 + m{Z}_3^2 & m{U}_3 \sim m{\chi^2}(3) \\ &\vdots & \vdots & & \vdots \\ m{U}_n &= m{Z}_1^2 + m{Z}_2^2 + m{Z}_2^2 + \cdots m{Z}_n^2 & m{U}_n \sim m{\chi^2}(n) \end{aligned}$$

# Chi-squared Distributions of Different $\boldsymbol{\nu}$



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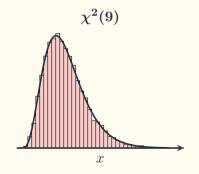
## Sampling Distribution of The Sample Variance

Let  $X_1, X_2, ..., X_n$  be a sample independently drawn from a normal distribution  $\mathcal{N}(\mu, \sigma^2)$ , and we define the sample mean and variance as:

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

Then we have:

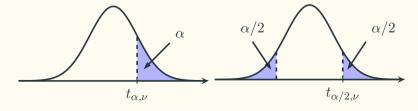
$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$$



### Confidence Interval For The Mean

Recall: construct  $(1 - \alpha) \times 100\%$  confidence interval for the mean.

- One-sided (lower bound):  $\mathbb{P}\left(\mu \geqslant \bar{X} t_{\alpha,\nu} \cdot \frac{s}{\sqrt{n}}\right) = 1 \alpha$
- One-sided (upper bound):  $\mathbb{P}\left(\mu \leqslant \bar{X} + t_{\alpha,\nu} \cdot \frac{s}{\sqrt{n}}\right) = 1 \alpha$
- Two-sided:  $\mathbb{P}\left(\bar{X} t_{\alpha/2,\nu} \cdot \frac{s}{\sqrt{n}} \leqslant \mu \leqslant \bar{X} + t_{\alpha/2,\nu} \cdot \frac{s}{\sqrt{n}}\right) = 1 \alpha$



#### **Confidence Interval For Variance**

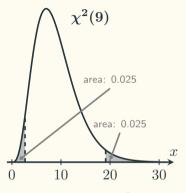
**Goal**: for a population with unknown variance  $\sigma^2$ , find a and b, such that  $\mathbb{P}\left(a\leqslant\sigma^2\leqslant b\right)=0.95$ 

$$\mathbb{P}\left(\chi_{0.975,\nu}^{2} \leqslant \chi^{2} \leqslant \chi_{0.025,\nu}^{2}\right) = 0.95$$

$$\mathbb{P}\left[\chi_{0.975,\nu}^{2} \leqslant \frac{(n-1)S^{2}}{\sigma^{2}} \leqslant \chi_{0.025,\nu}^{2}\right] = 0.95$$

$$\mathbb{P}\left[\frac{1}{\chi_{0.025,\nu}^{2}} \leqslant \frac{\sigma^{2}}{(n-1)S^{2}} \leqslant \frac{1}{\chi_{0.975,\nu}^{2}}\right] = 0.95$$

$$\mathbb{P}\left(\frac{n-1}{\chi_{0.025,\nu}^{2}} \cdot S^{2} \leqslant \sigma^{2} \leqslant \frac{n-1}{\chi_{0.975,\nu}^{2}} \cdot S^{2}\right) = 0.95$$



$$\begin{array}{c} 95\% \; \mathsf{CI} \; \mathsf{for} \; \sigma^2 : \\ \left[ \frac{n-1}{\chi^2_{0.025,\nu}} \cdot \boldsymbol{S}^2, \frac{n-1}{\chi^2_{0.975,\nu}} \cdot \boldsymbol{S}^2 \right] \end{array}$$

## Sphygmomanometer

STETHOSCOPE

A stethoscope is used to hear the sound of blood rushing back through the artery. The first thumping sound is the systolic blood pressure. When the thumping sound is no longer heard, that's the diastolic pressure.



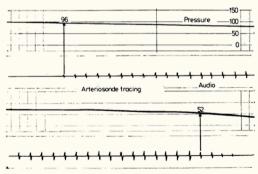




#### Confidence interval for variance

**Hypertension**: An Arteriosonde machine "prints" blood-pressure readings on a tape so that the measurement can be read rather than heard. A major argument for using such a machine is that the variability of measurements obtained by different observers on the same person will be lower than with a standard blood-pressure cuff.





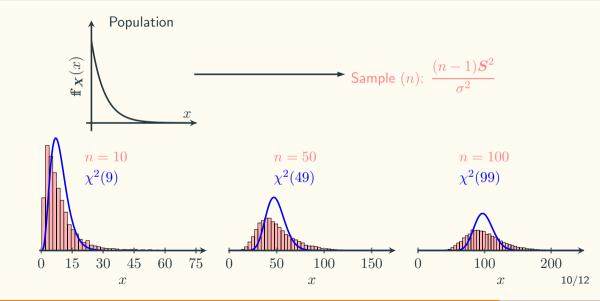
## Confidence interval for variance

Persion	Observer	Observer	Difference
(i)	#1	#2	$(d_i)$
1	194	200	-6
2	126	123	3
3	130	128	2
4	98	101	-3
5	136	135	1
6	145	145	0
7	110	111	-1
8	108	107	1
9	102	99	3
10	126	128	-2

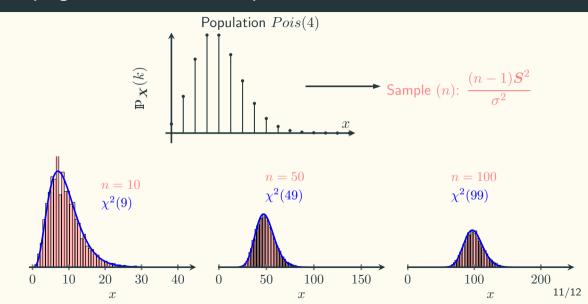
Does it make sense to construct a 95% CI for the variance?

$$\begin{array}{l} 95\% \; \text{CI for } \sigma^2 : \\ \left[ \frac{n-1}{\chi^2_{0.025,\nu}} \cdot s^2 \, , \frac{n-1}{\chi^2_{0.975,\nu}} \cdot s^2 \right] \end{array}$$

# Sampling Distribution of The Sample Variance



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## Conditions For Valid Confidence Intervals For The Variance

- 1. Random Samples
- 2. Independence ( n < 10% population size)
- 3. Original population distribution must be normal