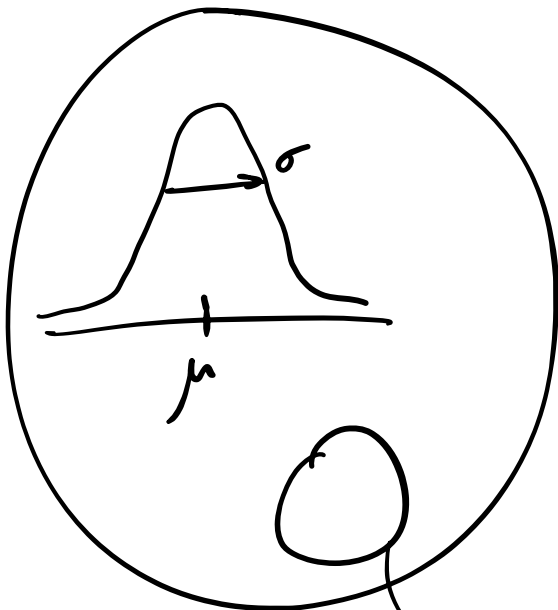


pop: μ unknown
 σ^2 unknown



STAT 515 Lec 11 slides

Variance estimation

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Use S_n^2 to estimate the value of σ^2 .

X_1, \dots, X_n

$$\bar{X}_n = \frac{1}{n} (X_1 + \dots + X_n)$$

Sample variance

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

These slides are an instructional aid; their sole purpose is to display, during the lecture, definitions, plots, results, etc. which take too much time to write by hand on the blackboard. They are not intended to explain or expound on any material.

For a rs X_1, \dots, X_n the *sample variance* is the quantity

$$S_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2.$$

$[S_n^2$ always nonnegative]

↓

	X	Xbar	X - Xbar	(X - Xbar) ²
1	28.72	27.442	1.278	1.633
2	29.07	27.442	1.628	2.65
3	30.21	27.442	2.768	7.661
4	25.66	27.442	-1.782	3.176
5	28.66	27.442	1.218	1.483
6	25.99	27.442	-1.452	2.109
7	27.16	27.442	-0.282	0.08
8	27.9	27.442	0.458	0.21
9	25.45	27.442	-1.992	3.969
10	28.97	27.442	1.528	2.334
11	27.93	27.442	0.488	0.238
12	26.54	27.442	-0.902	0.814
13	26.08	27.442	-1.362	1.855
14	26.08	27.442	-1.592	2.535

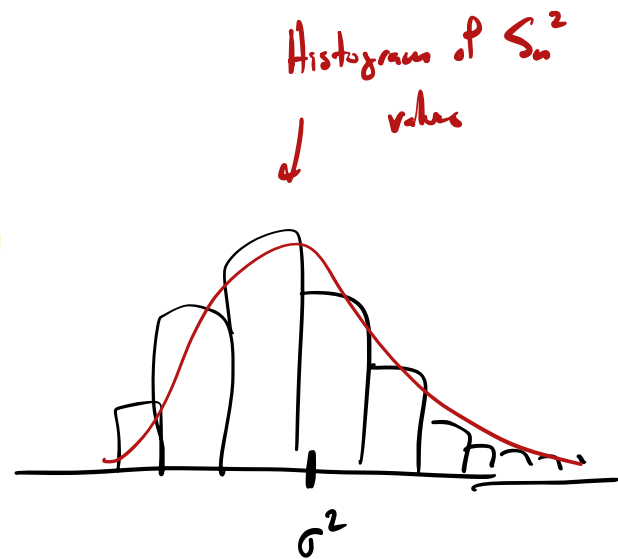
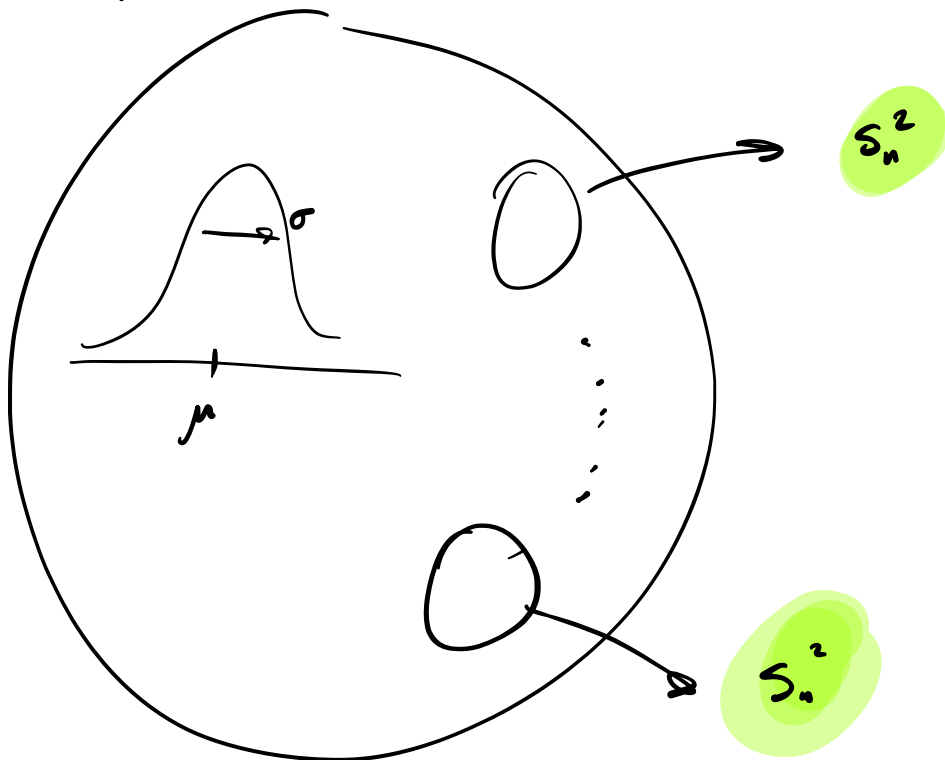
n=14

$$\begin{array}{r} 28.72 \\ - 27.442 \\ \hline 1.278 \end{array}$$

add these up,
divide by n-1.

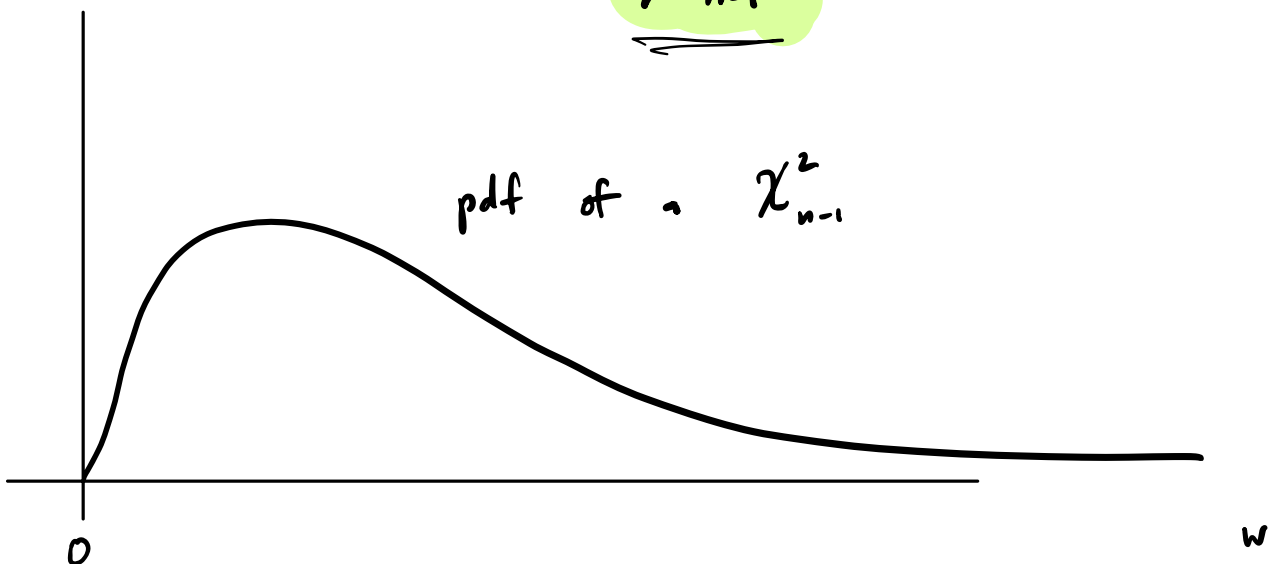
We call S_n the *sample standard deviation*.

pop : $\left. \begin{matrix} \mu \\ \sigma^2 \end{matrix} \right\} \text{ unknown}$



We find

$$W = \frac{(n-1)S_n^2}{\sigma^2} \sim \text{chi-squared dist. with "degrees of freedom" } n-1.$$
$$\sim \underline{\underline{\chi_{n-1}^2}}$$



Sampling distribution of the sample variance

If $X_1, \dots, X_n \stackrel{\text{ind}}{\sim} \text{Normal}(\mu, \sigma^2)$, then

$$\frac{(n-1)S_n^2}{\sigma^2} \sim \chi_{n-1}^2.$$

In the above χ_{n-1}^2 denotes the chi-squared dist. with $n-1$ degrees of freedom...

The chi-squared distributions

The probability distribution with pdf given by

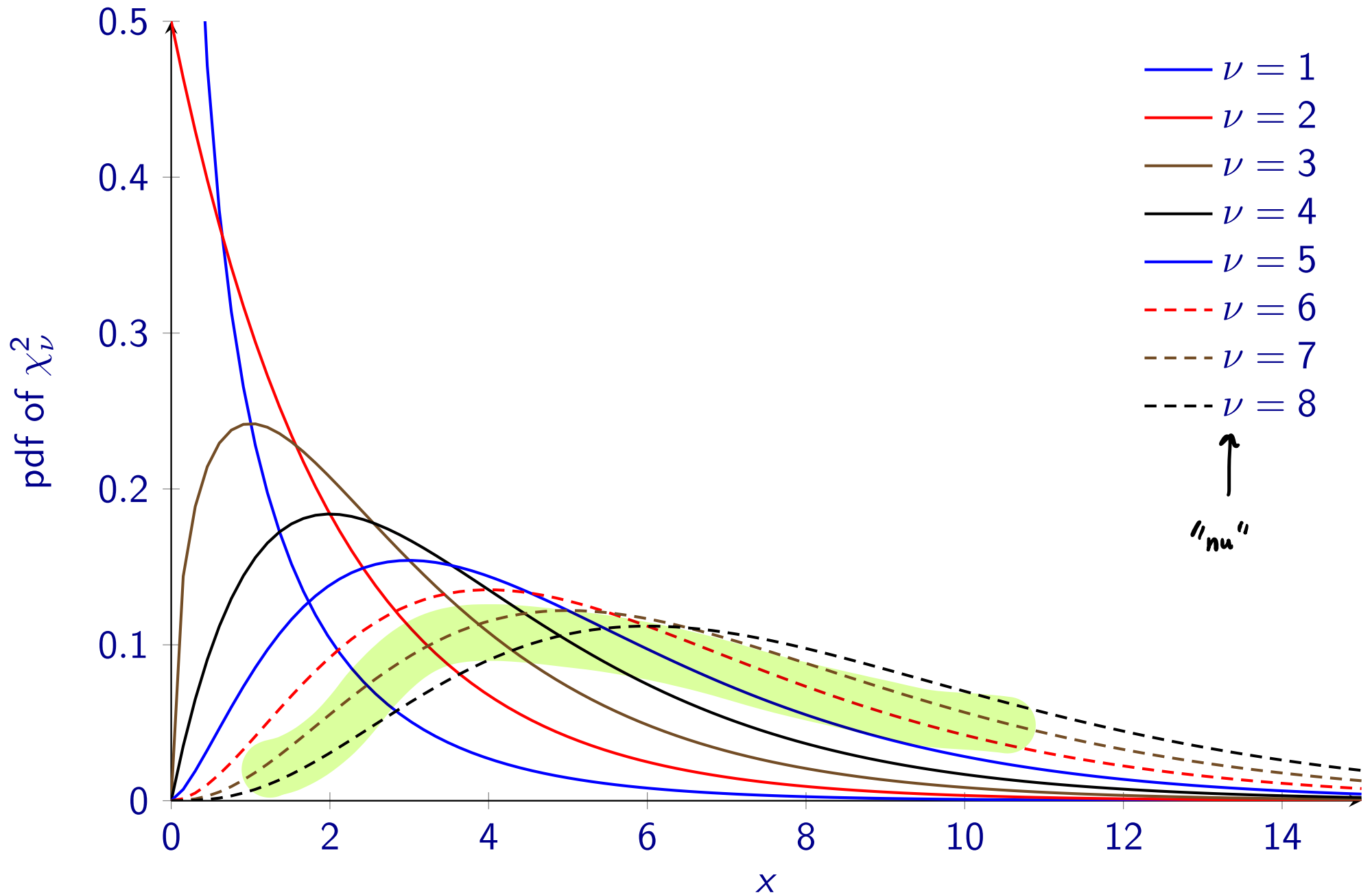
$$f(x) = \frac{1}{\Gamma(\nu/2)2^{\nu/2}} x^{\nu/2-1} \exp\left(-\frac{x}{2}\right), \quad x > 0,$$

is called the *chi-squared distribution* with *degrees of freedom* $\nu > 0$.

For a random variable W with this distribution, we write $W \sim \chi_\nu^2$.

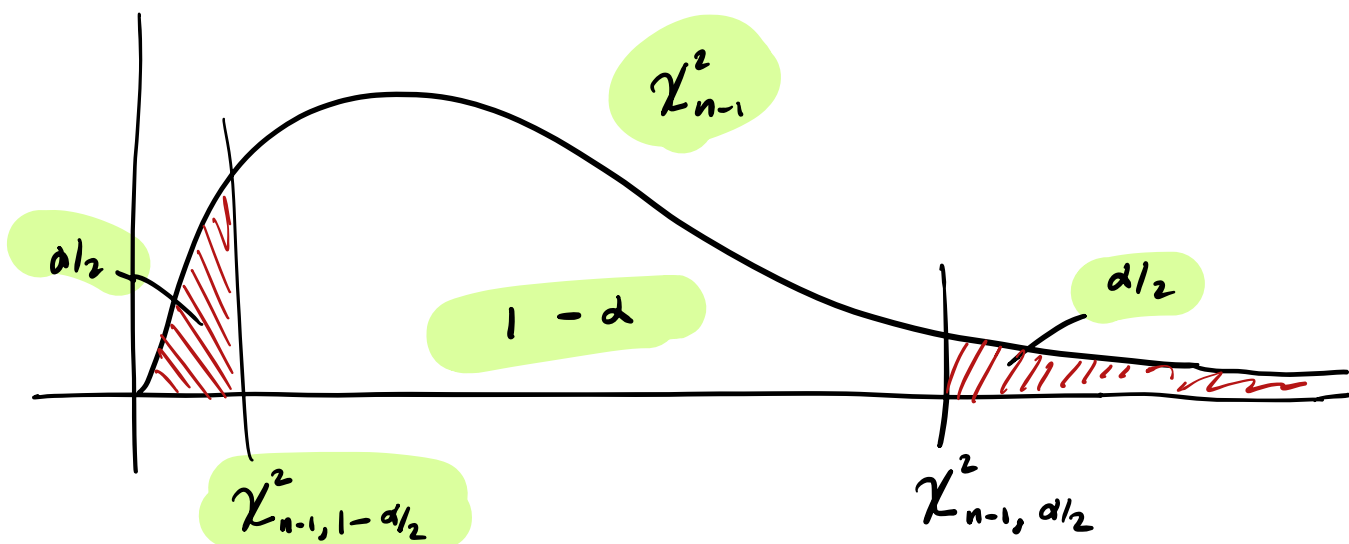
How to build a χ_k^2 random variable

If $W = Z_1^2 + \cdots + Z_k^2$, where $Z_1, \dots, Z_k \stackrel{\text{ind}}{\sim} \text{Normal}(0, 1)$, then $W \sim \chi_k^2$.



Build a $(1-\alpha) \cdot 100\%$ C.I. for σ^2 .

We have $\frac{(n-1)S_n^2}{\sigma^2} \sim \chi_{n-1}^2$.



Can write

$$P\left(\chi_{n-1, 1-\alpha/2}^2 \leq \frac{(n-1)S_n^2}{\sigma^2} \leq \chi_{n-1, \alpha/2}^2\right) = 1-\alpha$$

\Leftrightarrow

$$P\left(\frac{1}{\chi_{n-1, 1-\alpha/2}^2} \geq \frac{\sigma^2}{(n-1)S_n^2} \geq \frac{1}{\chi_{n-1, \alpha/2}^2}\right) = 1-\alpha$$

\Leftrightarrow

$$P\left(\underbrace{\frac{(n-1)S_n^2}{\chi_{n-1, 1-\alpha/2}^2}}_{\text{upper}} \geq \sigma^2 \geq \underbrace{\frac{(n-1)S_n^2}{\chi_{n-1, \alpha/2}^2}}_{\text{lower}}\right) = 1-\alpha$$

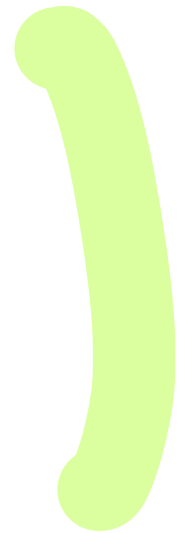
S_n

$$\left[\frac{(n-1)S_n^2}{\chi_{n-1, \alpha/2}^2}, \frac{(n-1)S_n^2}{\chi_{n-1, 1-\alpha/2}^2} \right] \text{ contains } \sigma^2 \text{ w/prob. } 1-\alpha.$$

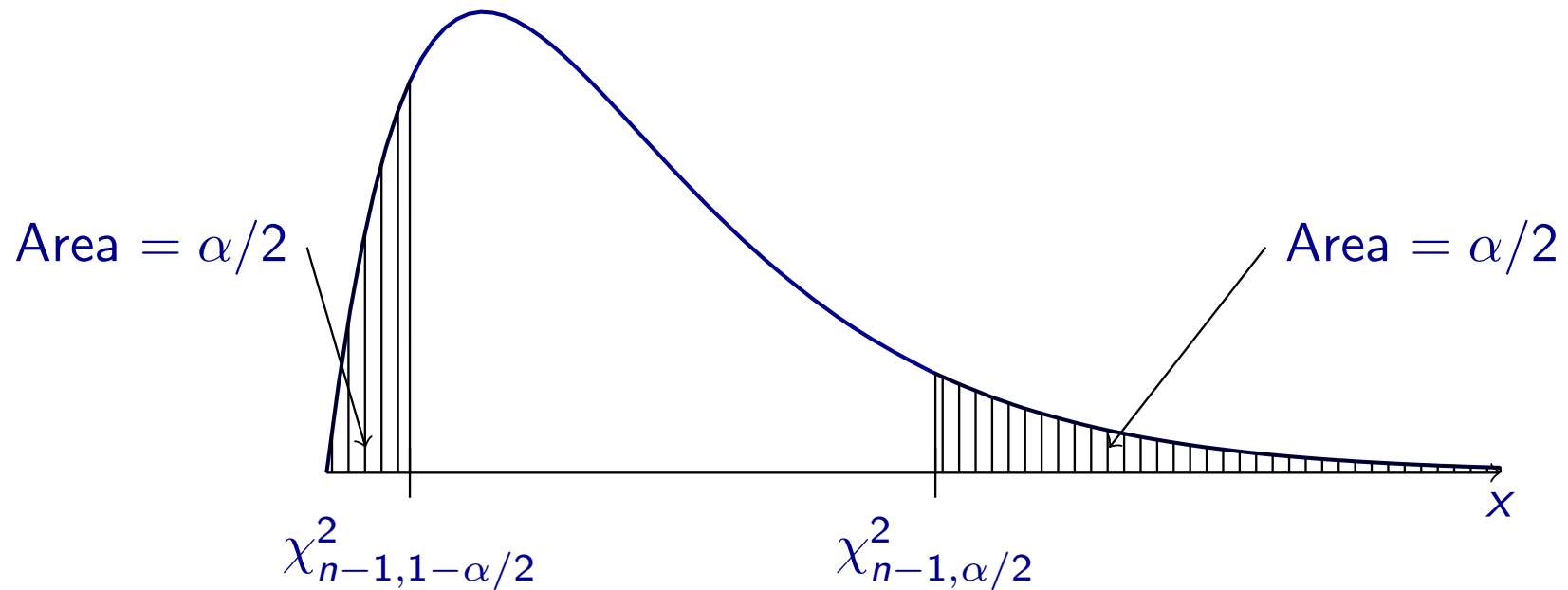
Exercise: Let $X_1, \dots, X_9 \stackrel{\text{ind}}{\sim} \text{Normal}(\mu, \sigma^2)$ and use the fact that

$$P\left(2.18 < \frac{(9-1)S_9^2}{\sigma^2} < 17.53\right) = 0.95$$

to build a 95% confidence interval for σ^2 .



What about a general $(1 - \alpha) \times 100\%$ CI for any $\alpha \in (0, 1)$?



Example: For $\alpha = 0.05$, $n = 9$, can get

$$\chi^2_{9-1, 0.975} = \text{qchisq}(0.025, 9-1) = 2.179731$$

$$\chi^2_{9-1, 0.025} = \text{qchisq}(0.975, 9-1) = 17.53455$$

Can also use [chi-square-table](#).

Confidence interval for variance of a Normal population

Let $X_1, \dots, X_n \stackrel{\text{ind}}{\sim} \text{Normal}(\mu, \sigma^2)$. Then

$$\left[\frac{(n-1)S_n^2}{\chi_{n-1, \alpha/2}^2}, \frac{(n-1)S_n^2}{\chi_{n-1, 1-\alpha/2}^2} \right]$$

is a $(1 - \alpha) \times 100\%$ confidence interval for σ^2 .

Exercise: Run `data(Loblolly)` in R and consider the heights of 10-yr-old trees.

- 1 Make a Normal Q-Q plot of the data.
- 2 Build a 95% CI for the variance of the tree heights.
- 3 Build a 99% CI for the variance of the tree heights.

①

$$\alpha = \underline{0.05}$$

$$n = 14$$

$$S_n = 1.538$$

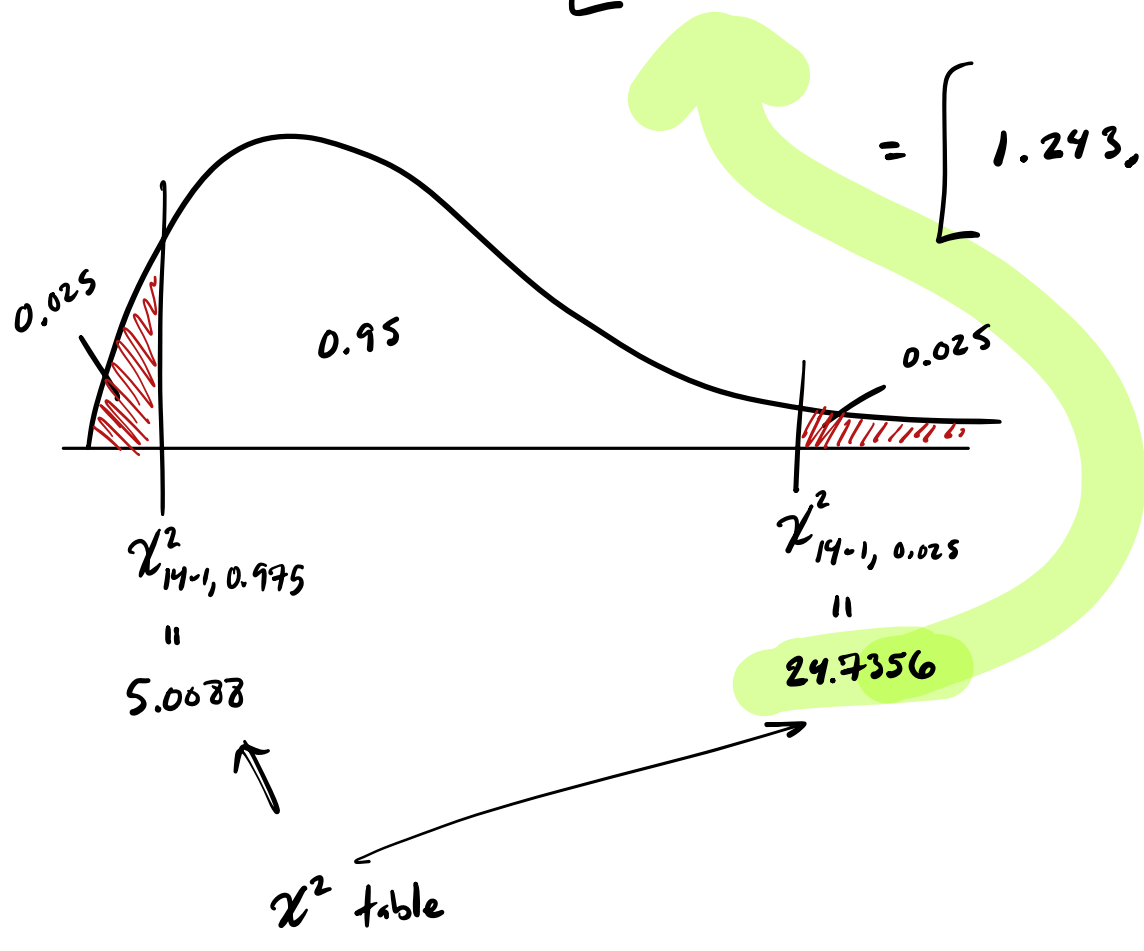
$$S_n^2 = 2.365$$

$$\left[\frac{(n-1)S_n^2}{\chi_{n-1, \alpha/2}^2}, \frac{(n-1)S_n^2}{\chi_{n-1, 1-\alpha/2}^2} \right]$$

$$\left[\frac{(14-1)(1.538)^2}{\chi_{14-1, 0.025}^2}, \frac{(14-1)(1.538)^2}{\chi_{14-1, 0.975}^2} \right]$$

$$= \left[\frac{(14-1)(1.538)^2}{24.7356}, \frac{(14-1)(1.538)^2}{5.0088} \right]$$

$$= [1.243, 6.138]$$



②

$$d = 0.01$$

$$d/2 = 0.005$$

$$1 - d/2 = 0.995$$

$$\left[\frac{(14-1)(1.538)^2}{\underbrace{\chi^2_{14-1, 0.995}}_{4.0747}}, \frac{(14-1)(1.538)^2}{\underbrace{\chi^2_{14-1, 0.005}}_{31.3193}} \right]$$

$$= [0.982, 7.546]$$

```
data("Loblolly")
X <- Loblolly$height[Loblolly$age == 10]
Sn <- sd(X)
n <- length(X)
alpha <- 0.05
lo <- (n-1)*Sn^2 / qchisq(1-alpha/2,n-1)
up <- (n-1)*Sn^2 / qchisq(alpha/2,n-1)
```