

# Lecture 19

$$X \sim \text{Cauchy}(0, 1) = \frac{1}{\pi} \cdot \frac{1}{x^2 + 1}$$

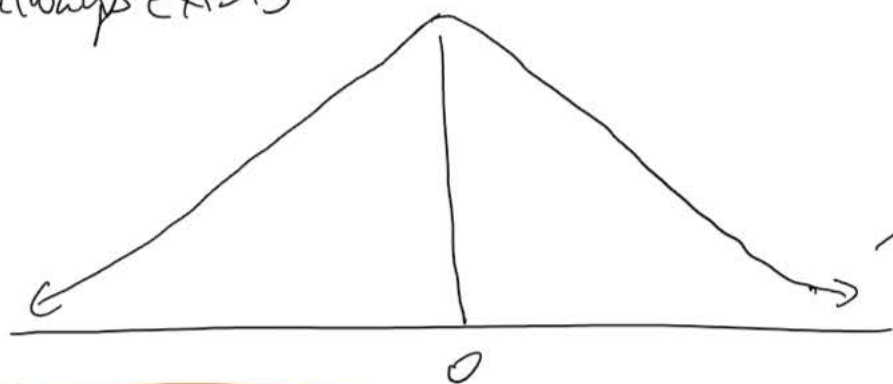
$$E[X] = \int_{\mathbb{R}} x \frac{1}{\pi} \frac{1}{x^2 + 1} dx = \infty \rightarrow \text{the expectation does not exist}$$

$$M_X(t) = E[e^{tx}] = \int_{\mathbb{R}} e^{tx} \frac{1}{\pi} \frac{1}{x^2 + 1} dx = \infty \rightarrow \text{D.N.E.}$$

$$\phi_X(t) = \int_{\mathbb{R}} e^{itx} \frac{1}{\pi} \frac{1}{x^2 + 1} dx = \dots = e^{-|t|}, \quad \phi_X'(t) = -\frac{t}{|t|} e^{-|t|}$$

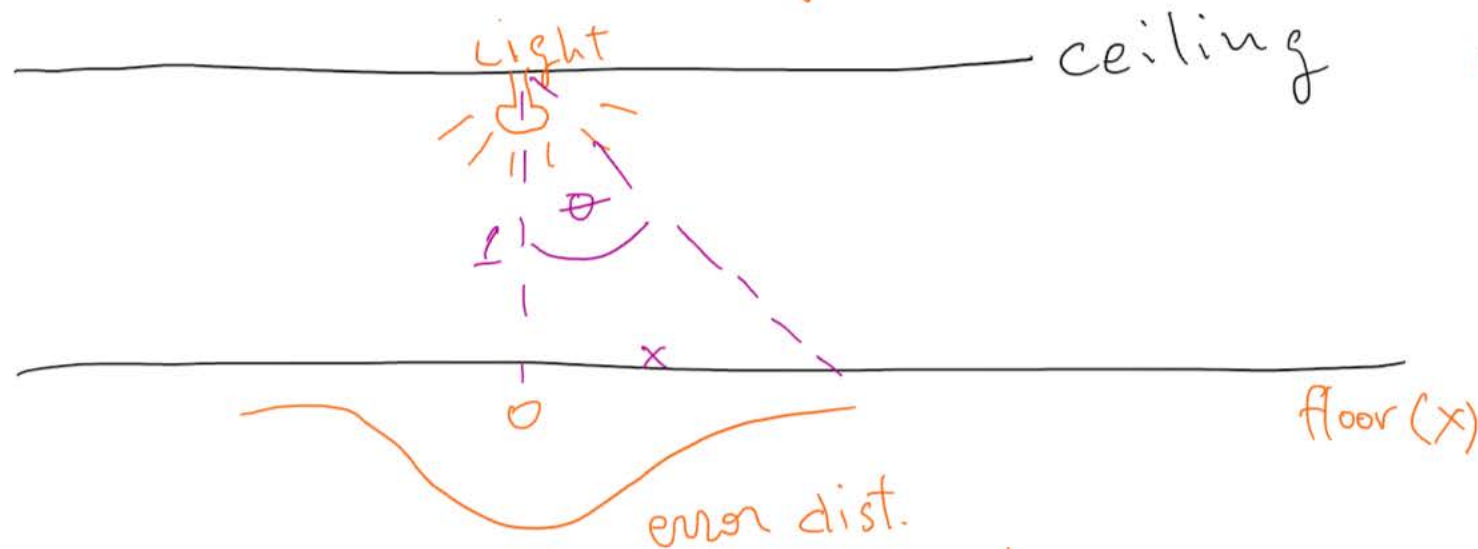
$\hookrightarrow$  always exists

$\phi_X'(0) = \text{D.N.E.}$



tails are fat enough so  
 $\rightarrow$  that the integral of this curve  
 weighted by  $x \in \mathbb{R}$  does not  
 converge.

Let's derive the Cauchy distribution like the physicists found it.



$$\theta \sim U(-\pi/2, \pi/2) = \frac{1}{\pi} \mathbb{I}_{\theta \in [-\pi/2, \pi/2]}$$

$$x = g(\theta)$$

$$\theta = g^{-1}(x) = \arctan x - \text{tangent}$$

is invertible between  $-\pi/2$  and  $\pi/2$

$$f_X(x) = f_{\theta}(g^{-1}(x)) \left| \frac{d}{dx} [g^{-1}(x)] \right| = \frac{1}{\pi} \mathbb{I}_{\underbrace{\arctan(x) \in [-\pi/2, \pi/2]}_{x \in \mathbb{R}}} \cdot \frac{1}{x^2 + 1} = \text{Cauchy}(0, 1)$$

Let  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2) \Rightarrow \frac{X_i - \mu}{\sigma} = Z_i \sim N(0, 1)$  Sample Var.

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 $T_n \sim N(n\mu, n\sigma^2)$ ,  $\bar{X}_n \sim N(\mu, \frac{\sigma^2}{n})$ ,  $S_n^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2 \sim f_{S^2}(s^2) = ?$

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

$$\vec{Z}^T \vec{Z} = \sum_{i=1}^n Z_i^2 = \chi_n^2 = \sum_{i=1}^n \left( \frac{X_i - \mu}{\sigma} \right)^2 = \sum_{i=1}^n \frac{(X_i - \mu)^2}{\sigma^2} =$$

$$\sum (X_i - \mu)^2 = \sum \left( (X_i - \bar{X}) + (\bar{X} - \mu) \right)^2 = \sum (X_i - \bar{X})^2 + \underbrace{\sum 2(X_i - \bar{X})(\bar{X} - \mu)}_0 + \sum (\bar{X} - \mu)^2$$

$$= \sum (X_i - \bar{X})^2 + n(\bar{X} - \mu)^2$$

$$\Rightarrow \frac{\sum (X_i - \mu)^2}{\sigma^2} = \underbrace{\frac{n-1}{\sigma^2} S^2}_{\chi_{n-1}^2} + \underbrace{\left( \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2}_{Z^2 \sim \chi_1^2} \sim \chi_n^2$$

Then maybe  $\chi_{n-1}^2$

recall from last class:  
 $U_1 \sim \chi_{k_1}^2$  indep. of  $U_2 \sim \chi_{k_2}^2$   
 $\Rightarrow U_1 + U_2 \sim \chi_{k_1 + k_2}^2$

In order for this "maybe" to be true, we need independence of those two terms i.e. we need  $S^2$  and  $\bar{X}$  to be independent.

We need Cochran's Theorem to prove this.

$\vec{Z}^T \vec{Z} = \vec{Z}^T \mathbf{I} \vec{Z} \sim \chi_n^2$  This scalar is called a "quadratic form"

Consider  $\vec{Z}^T \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{bmatrix} \vec{Z} = Z_1^2 \sim \chi_1^2$

$$\vec{Z}^T \mathbf{I} \vec{Z} = \vec{Z}^T (B_1 + \dots + B_n) \vec{Z} = \vec{Z}^T B_1 \vec{Z} + \dots + \vec{Z}^T B_n \vec{Z} \sim \chi_n^2$$

$$\chi_1^2 + \dots + \chi_1^2 + \dots + \chi_1^2$$

$\text{rank}[B_i] = 1$



$$\vec{z}^T \mathbf{I} \vec{z} = \vec{z}^T (B_1 + \dots + B_n) \vec{z} = \vec{z}^T B_1 \vec{z} + \dots + \vec{z}^T B_n \vec{z} \sim \chi_n^2$$

$$\chi_1^2 + \dots + \chi_1^2 + \dots + \chi_1^2$$

$$\text{rank}[B_i] = 1$$

$$\sum_{i=1}^n \text{rank}[B_i] = n$$

Conjecture: Each of these quadratic forms is independent.

Cochran's Thm: If  $B_1 + \dots + B_k = \mathbf{I}$ ,  $k \leq n$  and the sum of their ranks is  $n$  then you have two powerful results:

(a)  $\vec{z}^T B_j \vec{z} \sim \chi_{\text{rank}[B_j]}^2$  and (b)  $\vec{z}^T B_{j_1} \vec{z}$  is indep. of  $\vec{z}^T B_{j_2} \vec{z} \forall j_1 \neq j_2$

$$\text{Consider } \sum (z_i - \bar{z})^2 = \sum z_i^2 - 2 \underbrace{\sum z_i}_{n\bar{z}} \bar{z} + \sum \bar{z}^2 = \sum z_i^2 - 2n\bar{z}^2 + n\bar{z}^2$$

$$= \sum z_i^2 - n\bar{z}^2$$

let  $\vec{1}_n = n$ -dim column vector of all ones  $\bar{z} = \frac{1}{n} \vec{1}^T \vec{z} = \frac{1}{n} \vec{z}^T \vec{1}$

$$n\bar{z}^2 = n\bar{z}\bar{z} = \cancel{\frac{1}{n}} \vec{z}^T \vec{1} \frac{1}{n} \vec{1}^T \vec{z} = \frac{1}{n} \vec{z}^T \vec{1} \vec{1}^T \vec{z} = \vec{z}^T \underbrace{\left(\frac{1}{n} \mathbf{J}_n\right)}_{B_2} \vec{z}$$

let  $\mathbf{J}_n = \vec{1}\vec{1}^T$ , which is an  $n \times n$  matrix of all ones.

$$\sum (z_i - \bar{z})^2 = \vec{z}^T \mathbf{I} \vec{z} - \vec{z}^T \left(\frac{1}{n} \mathbf{J}_n\right) \vec{z} = \vec{z}^T \left(\mathbf{I} - \frac{1}{n} \mathbf{J}_n\right) \vec{z}$$

$$\vec{z}^T \vec{z} = \sum (z_i - \bar{z})^2 + n\bar{z}^2 = \vec{z}^T B_1 \vec{z} + \vec{z}^T B_2 \vec{z}$$

I want to use Cochran's thm on the above expression. So I need to make sure  $B_1 + B_2 = \mathbf{I}$  and  $\text{rank}[B_1] + \text{rank}[B_2] = n$

$$B_1 + B_2 = \left(\mathbf{I} - \frac{1}{n} \mathbf{J}_n\right) + \frac{1}{n} \mathbf{J}_n = \mathbf{I} \quad \checkmark$$

$$\text{rank}[B_2] = \text{rank}\left[\frac{1}{n} \mathbf{J}_n\right] = \text{rank}[\mathbf{J}_n] = 1$$

$$\text{rank}[\mathbf{I}] = \text{rank}[\mathbf{I} - \frac{1}{n} \mathbf{J}_n] = n - 1 \Rightarrow \text{rank}[B_1] = 1 + n - 1 = n \quad \checkmark$$



$$B_1 + B_2 = (I - \frac{1}{n} J_n) + \frac{1}{n} J_n = I \quad \checkmark$$

$$\text{rank}[B_2] = \text{rank}\left[\frac{1}{n} J_n\right] = \text{rank}[J_n] = 1$$

$$\text{rank}[B_1] = \text{rank}\left[I - \frac{1}{n} J_n\right] = n-1 \quad \Rightarrow \sum \text{rank}[B_j] = 1 + n-1 = n \quad \checkmark$$

need a Thm for this: If  $A$  is symmetric and idempotent (i.e.  $AA=A$ ) then  $\text{rank}[A] = \text{tr}[A] = \text{sum of } A\text{'s diagonal entries.}$

$$(I - \frac{1}{n} J)^T = I^T - \frac{1}{n} J^T = I - \frac{1}{n} J \quad \checkmark$$

$$\begin{aligned} (I - \frac{1}{n} J)(I - \frac{1}{n} J) &= I I - \frac{1}{n} J I - \frac{1}{n} I J + \frac{1}{n^2} J J \\ &= I - \cancel{\frac{1}{n} J} + \cancel{\frac{1}{n^2} n J} = I - \frac{1}{n} J \end{aligned}$$

$$\text{tr}\left[I - \frac{1}{n} J\right] = (1 - \frac{1}{n}) + (1 - \frac{1}{n}) + \dots + (1 - \frac{1}{n}) = n-1 = \text{rank}[B_1]$$

Since the two conditions of Cochran's Thm are satisfied, we can apply it to get the two results:

$$\Rightarrow \vec{z}^T B_1 \vec{z} = \sum (z_i - \bar{z})^2 \sim \chi_{n-1}^2 \quad \text{indep. of } \vec{z}^T B_2 \vec{z} = n \bar{z}^2 \sim \chi_1^2$$

What does this have to do with our goal? Well, it's the same thing:

$$\bar{z} = \frac{z_1 + \dots + z_n}{n} = \frac{\frac{x_1 - \mu}{\sigma} + \dots + \frac{x_n - \mu}{\sigma}}{n} = \frac{\sum x_i - n\mu}{\sigma n} = \frac{\bar{x} - \mu}{\sigma}$$

$$\begin{aligned} \sum (z_i - \bar{z})^2 &= \sum \left( \frac{x_i - \mu}{\sigma} - \frac{\bar{x} - \mu}{\sigma} \right)^2 = \sum \left( \frac{x_i - \bar{x}}{\sigma} \right)^2 = \frac{1}{\sigma^2} \sum (x_i - \bar{x})^2 \\ &= \frac{n-1}{\sigma^2} S^2 \end{aligned}$$

$$n \bar{z}^2 = n \left( \frac{\bar{x} - \mu}{\sigma} \right)^2 = \left( \frac{\sqrt{n} (\bar{x} - \mu)}{\sigma} \right)^2 = \left( \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right)^2$$

Since the two conditions of Cochran's Theorem are satisfied, we can apply it to get the two results:

$$\Rightarrow \vec{z}^T B_1 \vec{z} = \sum (z_i - \bar{z})^2 \sim \chi_{n-1}^2 \quad \text{indep. of} \quad \vec{z}^T B_2 \vec{z} = n \bar{z}^2 \sim \chi_1^2$$

What does this have to do with our goal? Well, it's the same thing:

$$\bar{z} = \frac{z_1 + \dots + z_n}{n} = \frac{\frac{x_1 - \mu}{\sigma} + \dots + \frac{x_n - \mu}{\sigma}}{n} = \frac{\sum x_i - n\mu}{\sigma n} = \frac{\bar{x} - \mu}{\sigma}$$

$$\sum (z_i - \bar{z})^2 = \sum \left( \frac{x_i - \mu}{\sigma} - \frac{\bar{x} - \mu}{\sigma} \right)^2 = \sum \left( \frac{x_i - \bar{x}}{\sigma} \right)^2 = \frac{1}{\sigma^2} \sum (x_i - \bar{x})^2 = \frac{n-1}{\sigma^2} S^2$$

$$n \bar{z}^2 = n \left( \frac{\bar{x} - \mu}{\sigma} \right)^2 = \left( \frac{\sqrt{n} (\bar{x} - \mu)}{\sigma} \right)^2 = \left( \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right)^2$$

$$\underbrace{\frac{n-1}{\sigma^2} S^2}_{\sim \chi_{n-1}^2} + \underbrace{\left( \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right)^2}_{\sim \chi_1^2} \sim \chi_n^2$$

↑ independent ↑

Fisher proved this without Cochran's theorem in 1925 and Geary proved in 1936 that this decomposition is exclusive to the iid normal r.v. model.

$$\frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1), \quad \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \sim ? \quad \text{Not } N(0, 1)$$