

CLT: $\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$

when X_i are i.i.d.

w/ $E(X) = \mu, \text{Var}(X) = \sigma^2$

If X_i are normally distr.
then CLT is exact for
any sample size n .

If non-normal but not
too skewed $n \geq 30$ ok.

Also by CLT get that

$$\sum_{i=1}^n X_i \sim N(n\mu, n\sigma^2)$$

ex) $X_i \sim \text{Pois}(\lambda=3)$ w/ $n=40$

estimate $P(\bar{X} > 3.1)$.

by CLT: $\bar{X} \sim N\left(3, \frac{3}{40}\right)$ so:

$$P(\bar{X} > 3.1) = 1 - \text{pnorm}(3.1, \sqrt{\frac{3}{40}})$$
$$\approx 0.357$$

ex, con't: Can also calc. w/ sum $\sum X_i$

$$\begin{aligned}P(\bar{X} > 3.1) &= P(n\bar{X} > n \cdot 3.1) \\&= P(\sum X_i > 40 \cdot 3.1) \\&= P(\sum X_i > 124)\end{aligned}$$

$$\begin{aligned}&\approx 1 - \text{pnorm}(124, 120, \sqrt{120}) \\&= 0.357\end{aligned}$$

Note that $\sum X_i \sim \text{Pois}(\lambda = 3 \cdot 40)$
(see: scaling pois. param.)

$$\begin{aligned}\text{so } P(\sum X_i > 124) \\&= 1 - \text{ppois}(124, 120) \\&= \underline{0.336}, \text{ so the}\end{aligned}$$

normal approx. is
reasonable.

(Student's) t-distribution:

$$X_i \sim N(\mu, \sigma^2)$$

$$\text{then } \frac{\bar{X} - \mu}{s/\sqrt{n}} \sim T_{n-1}$$

is t-distributed w/ $n-1$
"degrees of freedom."

ex:

Strength of rope is normally distr. Assume we don't know σ^2 but know $\mu = 120$ lb.

If we have a sample of 4 pieces of rope w/ $\bar{x} =$

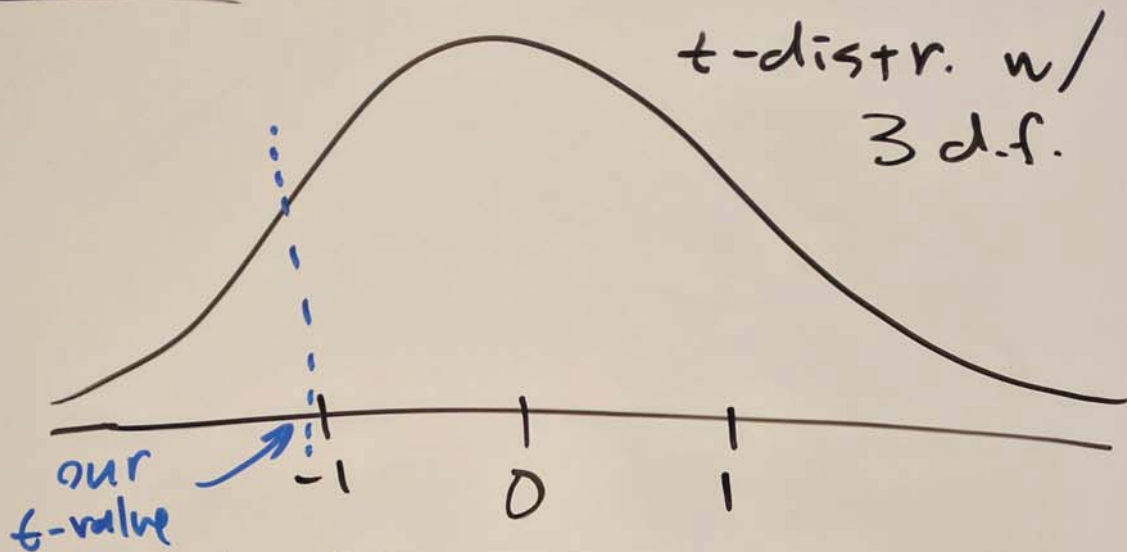
117.5

and $s = 5$, we can

calc. a t-statistic:

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}} = \frac{117.5 - 120}{5/\sqrt{4}} = -1$$

ex cont. $t = -1$



Recall 68-95-99.7 rule.

t-distr. is "wider" than
std. normal, so our ± 1
interval captures less data.

$$P(-1 < T_3 < 1) = P(t(1,3) - P(t(-1,3))) \\ = 0.609$$

Our t-statistic is typical.

What if $\bar{x} = 80$, $s = 30$, $n = 4$

$\bar{x} = 80$, $s = 5$, $n = 4$

$$t = \frac{80 - 120}{30/\sqrt{4}} = \frac{-40}{15} = -2.67 \leftarrow \text{still fairly typical}$$

$$t = \frac{80 - 120}{5/\sqrt{4}} = \frac{-40}{2.5} = -16 \leftarrow \text{unusually large!!}$$

Confidence Intervals:

$$X_i \sim N(\mu, \sigma^2)$$

$$\Rightarrow \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \overset{\text{std. normal:}}{\sim} N(0, 1)$$

$$\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} \overset{\text{t-distr.}}{\sim} T_{n-1}$$

IDEA: if we don't know μ ,
we can use this to
predict a range of
estimates for μ
i.e. μ is likely in interval:

$$(\text{est. } \mu) \pm (\text{error})$$

$(1-\alpha)100\%$ C.I.

$1-\alpha =$ confidence level
usually $\alpha=0.05$ to give 95% C.I.

ex: Let's say we have our
rope sample w/ $\bar{x}=90$,
 $s=5$, $n=4$. Predict
 μ w/ a 95% C.I.

sol'n: we know $\frac{\bar{x}-\mu}{s/\sqrt{n}} \sim T_{n-1}$

$$\text{so: } P\left(-t^* < \frac{\bar{x}-\mu}{s/\sqrt{n}} < t^*\right) = 95\%$$

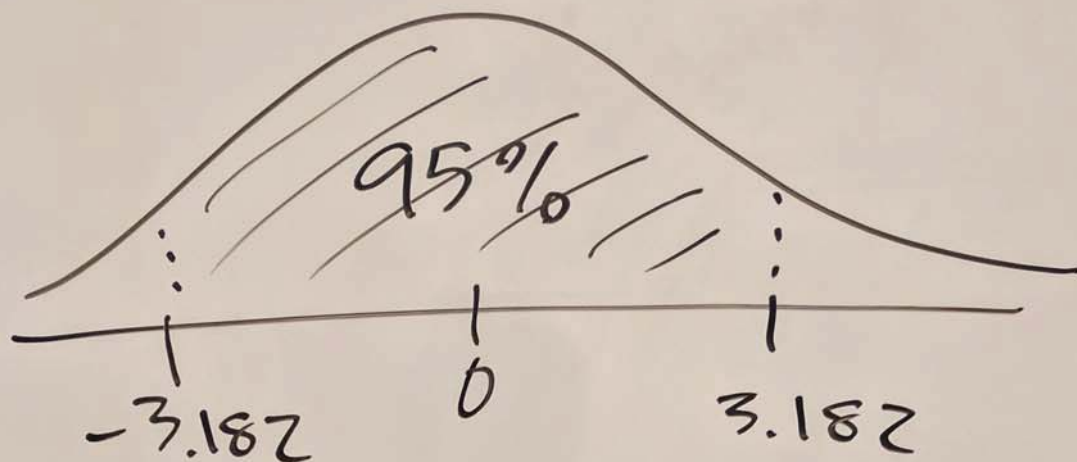
rearrange: $w/ t^* = t_{1-\alpha/2, n-1} = qt(1-\frac{\alpha}{2}, n-1)$
quantile

$$-t^* \frac{s}{\sqrt{n}} < \bar{x} - \mu < t^* \frac{s}{\sqrt{n}}$$

$$\boxed{\bar{x} - t^* \frac{s}{\sqrt{n}} < \mu < \bar{x} + t^* \frac{s}{\sqrt{n}}} \text{ our C.I.}$$

$$t_{0.975, 3} = qt(0.975, 3)$$

$$= 3.182$$



Recall for std. normal ± 1.96 std. dev's.
gives 95%.

We can see that t-distr. is wider.

So our 95% CI for μ is:

$$\bar{X} \pm t^* \frac{s}{\sqrt{n}} \rightarrow 90 \pm 3.182 \cdot \frac{5}{\sqrt{4}}$$

$$= 90 \pm 7.955$$

$$= (82.045, 97.955)$$

Consider this: If we wanted the mean strength of the rope to be $\mu = 120$, then our sample suggests $\mu \neq 120$ b/c it's not in our interval.

★ This is really what STATISTICS is about...

Estimating μ w/ C.I.:

General rules:

σ -known, data normal

$$\bar{X} \pm z_{1-\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

$$z_{1-\alpha/2} = q_{\text{norm}}(1-\alpha/2)$$

σ -unknown, but $n \geq 30$

$$\bar{X} \pm z_{1-\alpha/2} \cdot \frac{s}{\sqrt{n}}$$

(we are plugging in s for σ)

σ -unknown, $n < 30$

$$\bar{X} \pm t_{1-\alpha/2, n-1} \cdot \frac{s}{\sqrt{n}}$$

$$t_{1-\alpha/2, n-1} = q_{t}(1-\alpha/2, n-1)$$

★★ All assume $X_i \sim N(\mu, \sigma^2)$