Announcements

- Homework assignment going up today
- Next in-class lab will be next Wednesday
- I'm trying to get caught up on grading
- I'll be sending out more information this weekend about the final projects
- Starting Ch 4 today. We'll double back to the last bits of 3 later if we have time
- Polling: rembold-class.ddns.net

Friday the 13th

Between 1990 - 1992 researchers in the UK collected data on traffic flow, accidents, and hospital admissions on Friday 13th and the previous Friday, Friday 6th. Below is an excerpt from this data set on traffic flow. We can assume that traffic flow on given day at locations 1 and 2 are independent.

	type	date	6 th	13 th	diff	location
1	traffic	1990, July	139246	138548	698	loc 1
2	traffic	1990, July	134012	132908	1104	loc 2
3	traffic	1991, September	137055	136018	1037	loc 1
4	traffic	1991, September	133732	131843	1889	loc 2
5	traffic	1991, December	123552	121641	1911	loc 1
6	traffic	1991, December	121139	118723	2416	loc 2
7	traffic	1992, March	128293	125532	2761	loc 1
8	traffic	1992, March	124631	120249	4382	loc 2
9	traffic	1992, November	124609	122770	1839	loc 1
10	traffic	1992, November	117584	117263	321	loc 2

Friday the 13th

- We want to investigate if people's behavior is different on Friday 13th compared to Friday 6th.
- One approach is to compare the traffic flow on these two days.
- H_0 : Average traffic flow on Friday 6th and 13th are equal.
 - H_{A} : Average traffic flow on Friday 6^{th} and 13^{th} are different.

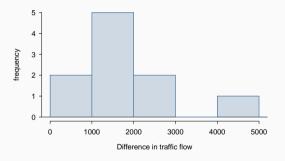
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Each case in the data set represents traffic flow recorded at the same location in the same month of the same year: one count from Friday 6th and the other Friday 13th. Are these two counts independent?

Conditions

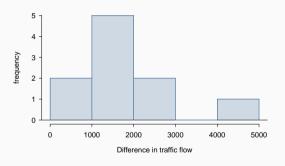
- Independence: We are told to assume that cases (rows) are independent.
- Sample size / skew: The sample distribution does not appear to be extremely skewed, but it's very difficult to assess with such a small sample size.
- We do not know σ and n is too small to assume s is a reliable estimate for σ.



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 σ . So what do we do when the sample size is small?



4

Review: what purpose does a large sample serve?

As long as observations are independent, and the population distribution is not extremely skewed, a large sample would ensure that...

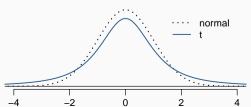
- the sampling distribution of the mean is nearly normal
- the estimate of the standard error, as $\frac{s}{\sqrt{n}}$, is reliable

The normality condition

- The CLT, which states that sampling distributions will be nearly normal, holds true for any sample size as long as the population distribution is nearly normal.
- While this is a helpful special case, it's inherently difficult to verify normality in small data sets.
- We should exercise caution when verifying the normality condition for small samples. It is important to not only examine the data but also think about where the data come from.
 - For example, ask: would I expect this distribution to be symmetric, and am I confident that outliers are rare?

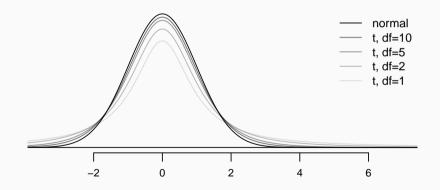
The t distribution

- When the population standard deviation is unknown (almost always), the uncertainty of the standard error estimate is addressed by using a new distribution: the t distribution.
- This distribution also has a bell shape, but its tails are thicker than the normal model's.
- Therefore observations are more likely to fall beyond two SDs from the mean than under the normal distribution.
- These extra thick tails are helpful for resolving our problem with a less reliable estimate the standard error (since n is small)



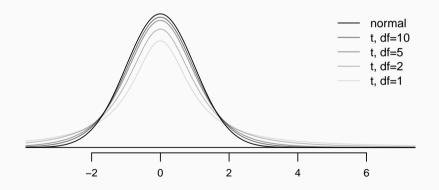
The *t* distribution (cont.)

- Always centered at zero, like the standard normal (z) distribution.
- Has a single parameter: *degrees of freedom* (*df*).



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What happens to shape of the t distribution as df increases?

Back to Friday the 13th

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Test statistic for inference on a small sample mean

The test statistic for inference on a small sample (n < 50) mean is the T statistic with df = n - 1.

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Test statistic for inference on a small sample mean

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 $T = \frac{1836 - 0}{372} = 4.94$
 $df = 10 - 1 = 9$

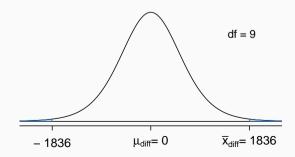
Finding the p-value

- The p-value is, once again, calculated as the tail area under the *t* distribution.
- Using R:

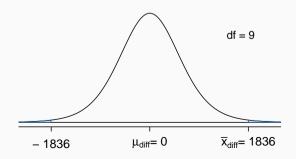
```
> 2 * pt(4.94, df = 9, lower.tail = FALSE)
[1] 0.0008022394
```

- Using a web app: https://gallery.shinyapps.io/dist_calc/
- Or when these aren't available, we can use a *t*-table.

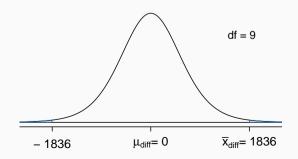
one tail		0.100	0.050	0.025	0.010	0.005
two tails		0.200	0.100	0.050	0.020	0.010
df	6	1.44	1.94	2.45	3.14	3.71
	7	1.41	1.89	2.36	3.00	3.50
	8	1.40	1.86	2.31	2.90	3.36
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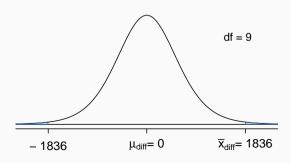
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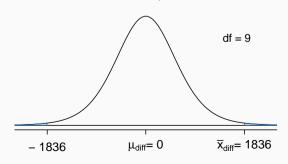


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What is the conclusion of the hypothesis test?

The data provide convincing evidence of a difference between traffic flow on Friday 6th and 13th.

What is the difference?

- We concluded that there is a difference in the traffic flow between Friday 6th and 13th.
- But it would be more interesting to find out what exactly this difference is.
- We can use a confidence interval to estimate this difference.

Confidence interval for a small sample mean

• Confidence intervals are always of the form

point estimate \pm *ME*

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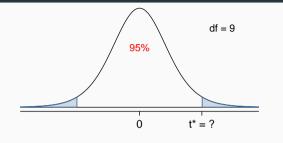
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- ME is always calculated as the product of a critical value and SE.
- Since small sample means follow a t distribution (and not a z distribution), the critical value is a t* (as opposed to a z*).

point estimate
$$\pm t^* \times SE$$

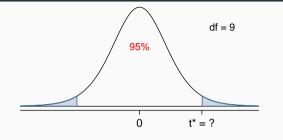
Finding the critical t (t^*)



n=10, df=10-1=9, t^* is at the intersection of row df=9 and two tail probability 0.05.

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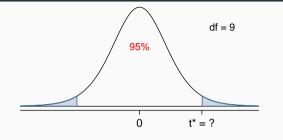
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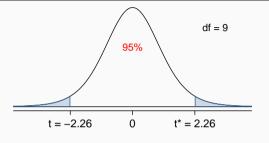
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Constructing a CI for a small sample mean

Which of the following is the correct calculation of a 95% confidence interval for the difference between the traffic flow between Friday 6th and 13th?

$$\bar{x}_{diff} = 1836$$
 $s_{diff} = 1176$ $n = 10$ $SE = 372$

- A) $1836 \pm 1.96 \times 372$
- B) $1836 \pm 2.26 \times 372$
- C) $1836 \pm -2.26 \times 372$
- D) $1836 \pm 2.26 \times 1176$

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Interpreting the CI

Which of the following is the <u>best</u> interpretation for the confidence interval we just calculated?

$$\mu_{diff:6th-13th} = (995, 2677)$$

We are 95% confident that ...

- A) the difference between the average number of cars on the road on Friday 6th and 13th is between 995 and 2,677.
- B) on Friday 6th there are 995 to 2,677 fewer cars on the road than on the Friday 13th, on average.
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 No, this is an observational study. We have just observed a significant difference between the number of cars on the road on these two days. We have not tested for people's beliefs.

Recap: Inference using the *t*-distribution

- If σ is unknown, use the *t*-distribution with $SE = \frac{s}{\sqrt{n}}$.
- Conditions:
 - independence of observations (often verified by a random sample, and if sampling without replacement, n < 10% of population)
 - no extreme skew
- Hypothesis testing:

$$T_{df} = \frac{\text{point estimate - null value}}{SE}$$
, where $df = n - 1$

• Confidence interval:

point estimate
$$\pm t_{df}^{\star} \times SE$$

Note: The example we used was for paired means (difference between dependent groups). We took the difference between the observations and used only these differences (one sample) in our analysis, therefore the mechanics are the same as when we are working with just one sample.