

## Chapter 6: Inference for categorical data

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OpenIntro Statistics, 4th Edition

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## Inference for a single proportion

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Two scientists want to know if a certain drug is effective against high blood pressure. The first scientist wants to give the drug to 1000 people with high blood pressure and see how many of them experience lower blood pressure levels. The second scientist wants to give the drug to 500 people with high blood pressure, and not give the drug to another 500 people with high blood pressure, and see how many in both groups experience lower blood pressure levels. Which is the better way to test this drug?

- (a) All 1000 get the drug
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- (a) All 1000 get the drug
- (b) *500 get the drug, 500 don't*

## Results from the GSS

The GSS asks the same question, below is the distribution of responses from the 2010 survey:

All 1000 get the drug	99
500 get the drug 500 don't	571
Total	670

## Parameter and point estimate

We would like to estimate the proportion of all Americans who have good intuition about experimental design, i.e. would answer “500 get the drug 500 don’t”? What are the parameter of interest and the point estimate?

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- *Parameter of interest:* Proportion of *all* Americans who have good intuition about experimental design.

$p$  (a population proportion)

- *Point estimate:* Proportion of *sampled* Americans who have good intuition about experimental design.

$\hat{p}$  (a sample proportion)

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Standard error of a sample proportion

$$SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

## Sample proportions are also nearly normally distributed

### Central limit theorem for proportions

Sample proportions will be nearly normally distributed with mean equal to the population mean,  $p$ , and standard error equal to

$$\sqrt{\frac{p(1-p)}{n}}.$$

$$\hat{p} \sim N\left(\text{mean} = p, SE = \sqrt{\frac{p(1-p)}{n}}\right)$$

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any guesses?

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**Note:** If  $p$  is unknown (most cases), we use  $\hat{p}$  in the calculation of the standard error.

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1. *Independence*: The sample is random, and  $670 < 10\%$  of all Americans, therefore we can assume that one respondent's response is independent of another.
2. *Success-failure*: 571 people answered correctly (successes) and 99 answered incorrectly (failures), both are greater than 10.

We are given that  $n = 670$ ,  $\hat{p} = 0.85$ , we also just learned that the standard error of the sample proportion is  $SE = \sqrt{\frac{p(1-p)}{n}}$ . Which of the below is the correct calculation of the 95% confidence interval?

- (a)  $0.85 \pm 1.96 \times \sqrt{\frac{0.85 \times 0.15}{670}}$
- (b)  $0.85 \pm 1.65 \times \sqrt{\frac{0.85 \times 0.15}{670}}$
- (c)  $0.85 \pm 1.96 \times \frac{0.85 \times 0.15}{\sqrt{670}}$
- (d)  $571 \pm 1.96 \times \sqrt{\frac{571 \times 99}{670}}$

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- (a)  $0.85 \pm 1.96 \times \sqrt{\frac{0.85 \times 0.15}{670}} \rightarrow (0.82, 0.88)$
- (b)  $0.85 \pm 1.65 \times \sqrt{\frac{0.85 \times 0.15}{670}}$
- (c)  $0.85 \pm 1.96 \times \frac{0.85 \times 0.15}{\sqrt{670}}$
- (d)  $571 \pm 1.96 \times \sqrt{\frac{571 \times 99}{670}}$

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$$n \geq \frac{1.96^2 \times 0.85 \times 0.15}{0.01^2}$$

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## What if there isn't a previous study?

... use  $\hat{p} = 0.5$

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- if you don't know any better, 50-50 is a good guess
- $\hat{p} = 0.5$  gives the most conservative estimate – highest possible sample size

## CI vs. HT for proportions

- Success-failure condition:
  - CI: At least 10 *observed* successes and failures
  - HT: At least 10 *expected* successes and failures, calculated using the null value
- Standard error:
  - CI: calculate using observed sample proportion:  $SE = \sqrt{\frac{p(1-p)}{n}}$
  - HT: calculate using the null value:  $SE = \sqrt{\frac{p_0(1-p_0)}{n}}$

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$$Z = \frac{0.85 - 0.80}{0.0154} = 3.25$$

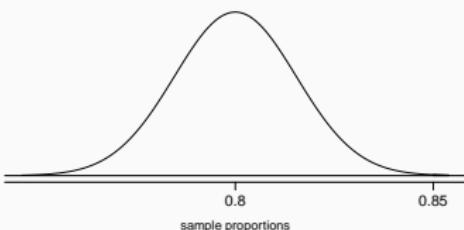
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$$p-value = 1 - 0.9994 = 0.0006$$



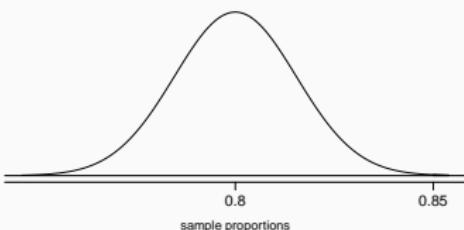
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$$p-value = 1 - 0.9994 = 0.0006$$



Since the p-value is low, we reject  $H_0$ . The data provide convincing evidence that more than 80% of Americans have a good intuition on experimental design.

11% of 1,001 Americans responding to a 2006 Gallup survey stated that they have objections to celebrating Halloween on religious grounds. At 95% confidence level, the margin of error for this survey is  $\pm 3\%$ . A news piece on this study's findings states: "More than 10% of all Americans have objections on religious grounds to celebrating Halloween." At 95% confidence level, is this news piece's statement justified?

- (a) Yes
- (b) No
- (c) Cannot tell

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- Conditions:
  - independence
    - random sample and 10% condition
  - at least 10 successes and failures
    - if not → randomization
- Standard error:  $SE = \sqrt{\frac{p(1-p)}{n}}$ 
  - for CI: use  $\hat{p}$
  - for HT: use  $p_0$

## Difference of two proportions

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## Melting ice cap

Scientists predict that global warming may have big effects on the polar regions within the next 100 years. One of the possible effects is that the northern ice cap may completely melt. Would this bother you a great deal, some, a little, or not at all if it actually happened?

- (a) A great deal
- (b) Some
- (c) A little
- (d) Not at all

## Results from the GSS

The GSS asks the same question, below are the distributions of responses from the 2010 GSS as well as from a group of introductory statistics students at Duke University:

	GSS	Duke
A great deal	454	69
Some	124	30
A little	52	4
Not at all	50	2
Total	680	105

## Parameter and point estimate

- *Parameter of interest:* Difference between the proportions of ***all*** Duke students and ***all*** Americans who would be bothered a great deal by the northern ice cap completely melting.

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Standard error of the difference between two sample proportions

$$SE_{(\hat{p}_1 - \hat{p}_2)} = \sqrt{\frac{p_1(1 - p_1)}{n_1} + \frac{p_2(1 - p_2)}{n_2}}$$

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We can assume that the attitudes of Duke students in the sample are independent of each other, and attitudes of US residents in the sample are independent of each other as well.

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### 2. *Independence between groups:* The sampled Duke students and the US residents are independent of each other.

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### 1. *Independence within groups:*

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We can assume that the attitudes of Duke students in the sample are independent of each other, and attitudes of US residents in the sample are independent of each other as well.

### 2. *Independence between groups:* The sampled Duke students and the US residents are independent of each other.

### 3. *Success-failure:*

At least 10 observed successes and 10 observed failures in the two groups.

Construct a 95% confidence interval for the difference between the proportions of Duke students and Americans who would be bothered a great deal by the melting of the northern ice cap ( $p_{Duke} - p_{US}$ ).

Data	Duke	US
A great deal	69	454
Not a great deal	36	226
Total	105	680

Construct a 95% confidence interval for the difference between the proportions of Duke students and Americans who would be bothered a great deal by the melting of the northern ice cap ( $p_{Duke} - p_{US}$ ).

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$$(\hat{p}_{Duke} - \hat{p}_{US}) \pm z^* \times \sqrt{\frac{\hat{p}_{Duke}(1 - \hat{p}_{Duke})}{n_{Duke}} + \frac{\hat{p}_{US}(1 - \hat{p}_{US})}{n_{US}}}$$

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 = & -0.011 \pm 1.96 \times 0.0497 \\
 = & -0.011 \pm 0.097 \\
 = & (-0.108, 0.086)
 \end{aligned}$$

Which of the following is the correct set of hypotheses for testing if the proportion of all Duke students who would be bothered a great deal by the melting of the northern ice cap differs from the proportion of all Americans who do?

(a)  $H_0 : p_{Duke} = p_{US}$

$$H_A : p_{Duke} \neq p_{US}$$

(b)  $H_0 : \hat{p}_{Duke} = \hat{p}_{US}$

$$H_A : \hat{p}_{Duke} \neq \hat{p}_{US}$$

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$$H_A : p_{Duke} - p_{US} \neq 0$$

(d)  $H_0 : p_{Duke} = p_{US}$

$$H_A : p_{Duke} < p_{US}$$

Which of the following is the correct set of hypotheses for testing if the proportion of all Duke students who would be bothered a great deal by the melting of the northern ice cap differs from the proportion of all Americans who do?

(a)  $H_0 : p_{Duke} = p_{US}$

$$H_A : p_{Duke} \neq p_{US}$$

(b)  $H_0 : \hat{p}_{Duke} = \hat{p}_{US}$

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*Both (a) and (c) are correct.*

## Flashback to working with one proportion

- When constructing a confidence interval for a population proportion, we check if the *observed* number of successes and failures are at least 10.

$$n\hat{p} \geq 10$$

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- When conducting a hypothesis test for a population proportion, we check if the *expected* number of successes and failures are at least 10.

$$np_0 \geq 10 \quad n(1 - p_0) \geq 10$$

## Pooled estimate of a proportion

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- Therefore, we need to first find a common (**pooled**) proportion for the two groups, and use that in our analysis.
- This simply means finding the proportion of total successes among the total number of observations.

Pooled estimate of a proportion

$$\hat{p} = \frac{\# \text{ of successes}_1 + \# \text{ of successes}_2}{n_1 + n_2}$$

Calculate the estimated pooled proportion of Duke students and Americans who would be bothered a great deal by the melting of the northern ice cap. Which sample proportion ( $\hat{p}_{Duke}$  or  $\hat{p}_{US}$ ) the pooled estimate is closer to? Why?

Data	Duke	US
A great deal	69	454
Not a great deal	36	226
Total	105	680
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$$p-value = 2 \times P(Z < -0.22) = 2 \times 0.41 = 0.82$$

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- $SE(\hat{p}_1 - \hat{p}_2) = \sqrt{\frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}}$ 
  - for CI: use  $\hat{p}_1$  and  $\hat{p}_2$
  - for HT:
    - when  $H_0 : p_1 = p_2$ : use  $\hat{p}_{pool} = \frac{\# suc_1 + \# suc_2}{n_1 + n_2}$
    - when  $H_0 : p_1 - p_2 = (\text{some value other than } 0)$ : use  $\hat{p}_1$  and  $\hat{p}_2$ 
      - this is pretty rare

## Reference - standard error calculations

	one sample	two samples
mean	$SE = \frac{s}{\sqrt{n}}$	$SE = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$
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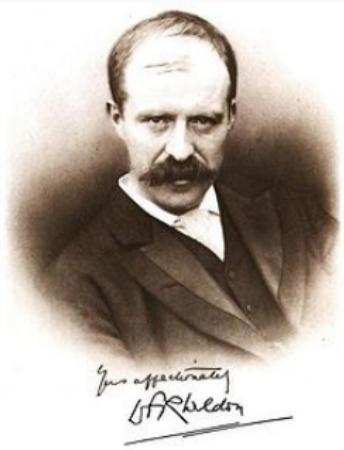
- When working with means, it's very rare that  $\sigma$  is known, so we usually use  $s$ .
- When working with proportions,
  - if doing a hypothesis test,  $p$  comes from the null hypothesis
  - if constructing a confidence interval, use  $\hat{p}$  instead

## **Chi-square test of GOF**

---

## Weldon's dice

- Walter Frank Raphael Weldon (1860 - 1906), was an English evolutionary biologist and a founder of biometry. He was the joint founding editor of Biometrika, with Francis Galton and Karl Pearson.
- In 1894, he rolled 12 dice 26,306 times, and recorded the number of 5s or 6s (which he considered to be a success).
- It was observed that 5s or 6s occurred more often than expected, and Pearson hypothesized that this was probably due to the construction of the dice. Most inexpensive dice have hollowed-out pips, and since opposite sides add to 7, the face with 6 pips is lighter than its opposing face, which has only 1 pip.

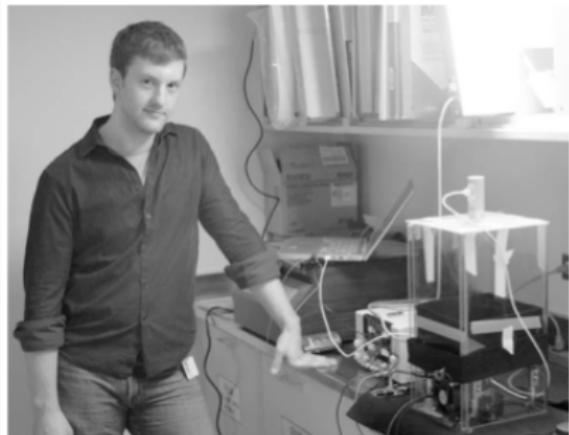


## Labby's dice

- In 2009, Zacariah Labby (U of Chicago), repeated Weldon's experiment using a homemade dice-throwing, pip counting machine.

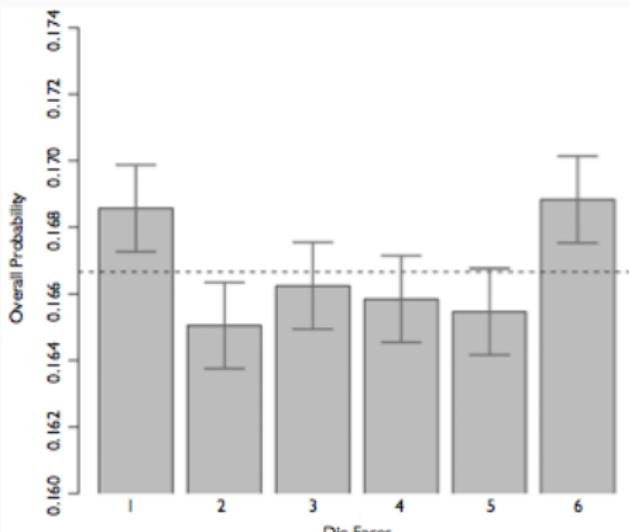
[http://www.youtube.com/  
watch?v=95EErdouO2w](http://www.youtube.com/watch?v=95EErdouO2w)

- The rolling-imaging process took about 20 seconds per roll.
  - Each day there were ~150 images to process manually.
  - At this rate Weldon's experiment was repeated in a little more than six full days.
  - Recommended reading:



## Labby's dice (cont.)

- Labby did not actually observe the same phenomenon that Weldon observed (higher frequency of 5s and 6s).
- Automation allowed Labby to collect more data than Weldon did in 1894, instead of recording “successes” and “failures”, Labby recorded the individual number of pips on each die.



## Expected counts

Labby rolled 12 dice 26,306 times. If each side is equally likely to come up, how many 1s, 2s, ⋯, 6s would he expect to have observed?

- (a)  $\frac{1}{6}$
- (b)  $\frac{12}{6}$
- (c)  $\frac{26,306}{6}$
- (d)  $\frac{12 \times 26,306}{6}$

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- (c)  $\frac{26,306}{6}$
- (d)  $\frac{12 \times 26,306}{6} = 52,612$

## Summarizing Labby's results

The table below shows the observed and expected counts from Labby's experiment.

Outcome	Observed	Expected
1	53,222	52,612
2	52,118	52,612
3	52,465	52,612
4	52,338	52,612
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6	53,285	52,612
Total	315,672	315,672

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Total	315,672	315,672

Why are the expected counts the same for all outcomes but the observed counts are different? At a first glance, does there appear to be an inconsistency between the observed and expected counts?

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$H_0$ : There is no inconsistency between the observed and the expected counts. *The observed counts follow the same distribution as the expected counts.*

$H_A$ : There is an inconsistency between the observed and the expected counts. *The observed counts do not follow the same distribution as the expected counts.* There is a bias in which side comes up on the roll of a die.

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- Large deviations from what would be expected based on sampling variation (chance) alone provide strong evidence for the alternative hypothesis.
- This is called a *goodness of fit* test since we're evaluating how well the observed data fit the expected distribution.

## Anatomy of a test statistic

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These two ideas will help in the construction of an appropriate test statistic for count data.

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$\chi^2$  statistic

$$\chi^2 = \sum_{i=1}^k \frac{(O - E)^2}{E} \quad \text{where } k = \text{total number of cells}$$

## Calculating the chi-square statistic

Outcome	Observed	Expected	$\frac{(O-E)^2}{E}$
1	53,222	52,612	$\frac{(53,222-52,612)^2}{52,612} = 7.07$

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When have we seen this before?

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- The chi-square distribution has just one parameter called *degrees of freedom (df)*, which influences the shape, center, and spread of the distribution.

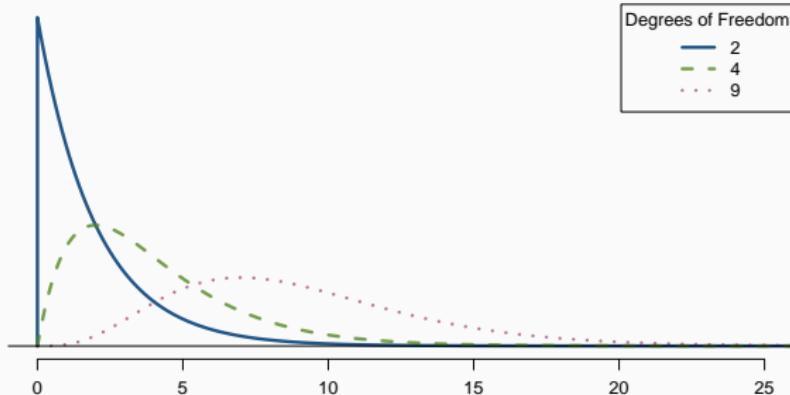
# The chi-square distribution

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*Remember:* So far we've seen three other continuous distributions:

- *normal distribution: unimodal and symmetric with two parameters: mean and standard deviation*
- *T distribution: unimodal and symmetric with one parameter: degrees of freedom*
- *F distribution: unimodal and right skewed with two parameters: degrees of freedom or numerator (between group variance) and denominator (within group variance)*

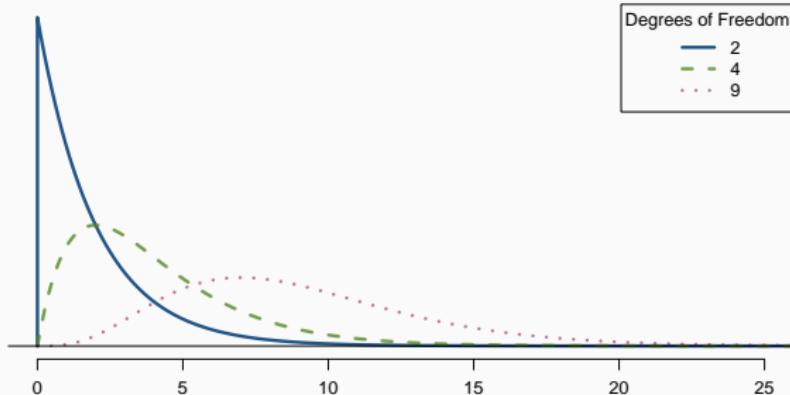
Which of the following is false?



As the df increases,

- (a) the center of the  $\chi^2$  distribution increases as well
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## Finding areas under the chi-square curve

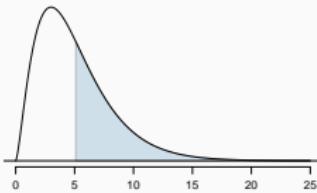
- p-value = tail area under the chi-square distribution (as usual)

## Finding areas under the chi-square curve

- p-value = tail area under the chi-square distribution (as usual)
- For this we can use technology, or a *chi-square probability table*.

## Finding areas under the chi-square curve

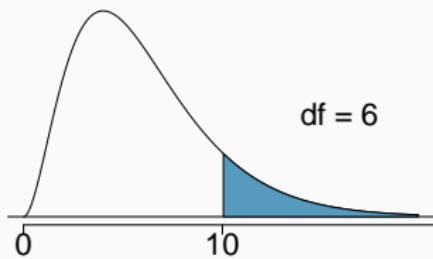
- p-value = tail area under the chi-square distribution (as usual)
- For this we can use technology, or a *chi-square probability table*.
- This table works a lot like the  $t$  table, but only provides upper tail values.



Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
	3	3.66	4.64	6.25	7.81	9.84	11.34	12.84	16.27
	4	4.88	5.99	7.78	9.49	11.67	13.28	14.86	18.47
	5	6.06	7.29	9.24	11.07	13.39	15.09	16.75	20.52
	6	7.23	8.56	10.64	12.59	15.03	16.81	18.55	22.46

## Finding areas under the chi-square curve (cont.)

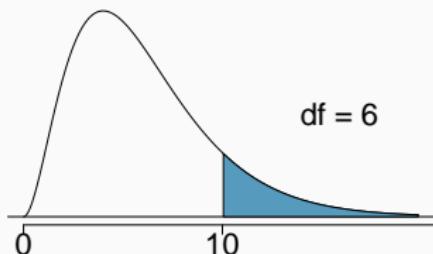
Estimate the shaded area under the chi-square curve with  $df = 6$ .



Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
	3	3.66	4.64	6.25	7.81	9.84	11.34	12.84	16.27
	4	4.88	5.99	7.78	9.49	11.67	13.28	14.86	18.47
	5	6.06	7.29	9.24	11.07	13.39	15.09	16.75	20.52
	6	7.23	8.56	10.64	12.59	15.03	16.81	18.55	22.46
	7	8.38	9.80	12.02	14.07	16.62	18.48	20.28	24.32

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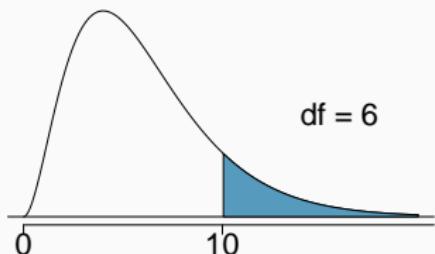
Estimate the shaded area under the chi-square curve with  $df = 6$ .



Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
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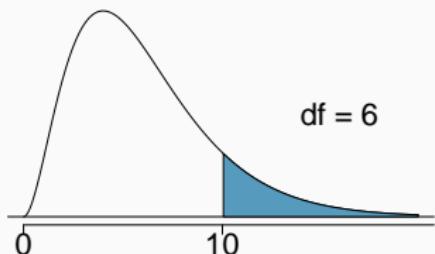
Estimate the shaded area under the chi-square curve with  $df = 6$ .



Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
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## Finding areas under the chi-square curve (cont.)

Estimate the shaded area under the chi-square curve with  $df = 6$ .

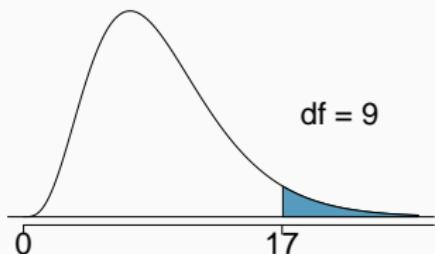


$P(\chi^2_{df=6} > 10)$   
is between 0.1 and 0.2

Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
	3	3.66	4.64	6.25	7.81	9.84	11.34	12.84	16.27
	4	4.88	5.99	7.78	9.49	11.67	13.28	14.86	18.47
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## Finding areas under the chi-square curve (cont.)

Estimate the shaded area (above 17) under the  $\chi^2$  curve with  $df = 9$ .

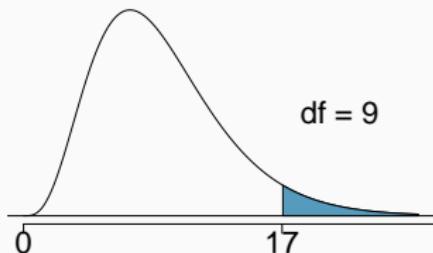


- (a) 0.05
- (b) 0.02
- (c) between 0.02 and 0.05
- (d) between 0.05 and 0.1
- (e) between 0.01 and 0.02

Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	7	8.38	9.80	12.02	14.07	16.62	18.48	20.28	24.32
	8	9.52	11.03	13.36	15.51	18.17	20.09	21.95	26.12
	9	10.66	12.24	14.68	16.92	19.68	21.67	23.59	27.88
	10	11.78	13.44	15.99	18.31	21.16	23.21	25.19	29.59
	11	12.90	14.63	17.28	19.68	22.62	24.72	26.76	31.26

## Finding areas under the chi-square curve (cont.)

Estimate the shaded area (above 17) under the  $\chi^2$  curve with  $df = 9$ .

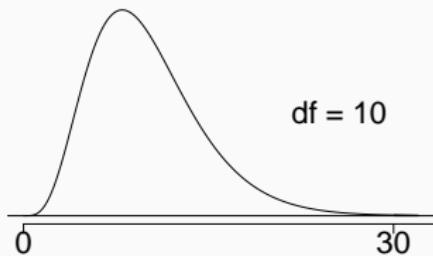


- (a) 0.05
- (b) 0.02
- (c) between 0.02 and 0.05
- (d) between 0.05 and 0.1
- (e) between 0.01 and 0.02

Upper tail	0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001	
df	7	8.38	9.80	12.02	14.07	16.62	18.48	20.28	24.32
	8	9.52	11.03	13.36	15.51	18.17	20.09	21.95	26.12
	9	10.66	12.24	14.68	16.92	19.68	21.67	23.59	27.88
	10	11.78	13.44	15.99	18.31	21.16	23.21	25.19	29.59
	11	12.90	14.63	17.28	19.68	22.62	24.72	26.76	31.26

## Finding areas under the chi-square curve (one more)

Estimate the shaded area (above 30) under the  $\chi^2$  curve with  $df = 10$ .

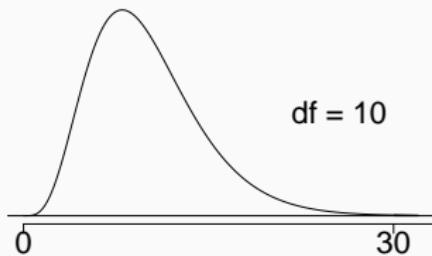


- (a) greater than 0.3
- (b) between 0.005 and 0.001
- (c) less than 0.001
- (d) greater than 0.001
- (e) cannot tell using this table

Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	7	8.38	9.80	12.02	14.07	16.62	18.48	20.28	24.32
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Estimate the shaded area (above 30) under the  $\chi^2$  curve with  $df = 10$ .



- (a) greater than 0.3
- (b) between 0.005 and 0.001
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Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001	→
df	7	8.38	9.80	12.02	14.07	16.62	18.48	20.28	24.32	
	8	9.52	11.03	13.36	15.51	18.17	20.09	21.95	26.12	
	9	10.66	12.24	14.68	16.92	19.68	21.67	23.59	27.88	
	10	11.78	13.44	15.99	18.31	21.16	23.21	25.19	29.59	→
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- While probability tables are very helpful in understanding how probability distributions work, and provide quick reference when computational resources are not available, they are somewhat archaic.

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pchisq(q = 30, df = 10, lower.tail = FALSE)  
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```

- Using a web applet:  
[http://bitly.com/dist\\_calc](http://bitly.com/dist_calc)

## Back to Labby's dice

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- We had calculated a test statistic of  $\chi^2 = 24.67$ .
- All we need is the  $df$  and we can calculate the tail area (the p-value) and make a decision on the hypotheses.

## Degrees of freedom for a goodness of fit test

- When conducting a goodness of fit test to evaluate how well the observed data follow an expected distribution, the degrees of freedom are calculated as the number of cells ( $k$ ) minus 1.

$$df = k - 1$$

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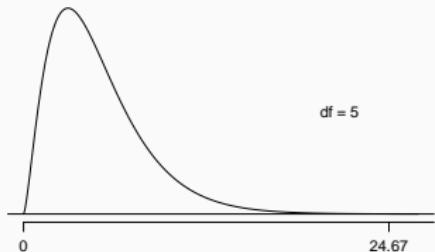
$$df = k - 1$$

- For dice outcomes,  $k = 6$ , therefore

$$df = 6 - 1 = 5$$

## Finding a p-value for a chi-square test

The *p-value* for a chi-square test is defined as the *tail area above the calculated test statistic*.



$$\text{p-value} = P(\chi^2_{df=5} > 24.67)$$

is less than 0.001

Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001	→
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83	
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82	
	3	3.66	4.64	6.25	7.81	9.84	11.34	12.84	16.27	
	4	4.88	5.99	7.78	9.49	11.67	13.28	14.86	18.47	
	5	6.06	7.29	9.24	11.07	13.39	15.09	16.75	20.52	→

## Conclusion of the hypothesis test

We calculated a p-value less than 0.001. At 5% significance level, what is the conclusion of the hypothesis test?

- (a) Reject  $H_0$ , the data provide convincing evidence that the dice are fair.
- (b) Reject  $H_0$ , the data provide convincing evidence that the dice are biased.
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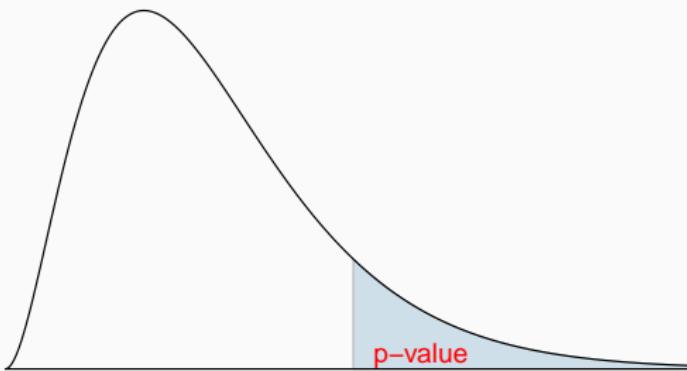
## Turns out...

- The 1-6 axis is consistently shorter than the other two (2-5 and 3-4), thereby supporting the hypothesis that the faces with one and six pips are larger than the other faces.
- Pearson's claim that 5s and 6s appear more often due to the carved-out pips is not supported by these data.
- Dice used in casinos have flush faces, where the pips are filled in with a plastic of the same density as the surrounding material and are precisely balanced.



## Recap: p-value for a chi-square test

- The p-value for a chi-square test is defined as the tail area *above* the calculated test statistic.
- This is because the test statistic is always positive, and a higher test statistic means a stronger deviation from the null hypothesis.



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Failing to check conditions may unintentionally affect the test's error rates.

## 2009 Iran Election

There was lots of talk of election fraud in the 2009 Iran election. We'll compare the data from a poll conducted before the election (observed data) to the reported votes in the election to see if the two follow the same distribution.

Candidate	Observed # of voters in poll	Reported % of votes in election
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## Hypotheses

What are the hypotheses for testing if the distributions of reported and polled votes are different?

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$H_A$ : *The observed counts from the poll do not follow the same distribution as the reported votes.*

## Calculation of the test statistic

Candidate	Observed # of voters in poll	Reported % of votes in election	Expected # of votes in poll
(1) Ahmedinajad	338	63.29%	$504 \times 0.6329 = 319$
(2) Mousavi	136	34.10%	$504 \times 0.3410 = 172$
(3) Minor candidates	30	2.61%	$504 \times 0.0261 = 13$
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$$\chi^2_{df=3-1=2} = 30.89$$

## Conclusion

Based on these calculations what is the conclusion of the hypothesis test?

- (a) p-value is low,  $H_0$  is rejected. The observed counts from the poll do not follow the same distribution as the reported votes.
- (b) p-value is high,  $H_0$  is not rejected. The observed counts from the poll follow the same distribution as the reported votes.
- (c) p-value is low,  $H_0$  is rejected. The observed counts from the poll follow the same distribution as the reported votes
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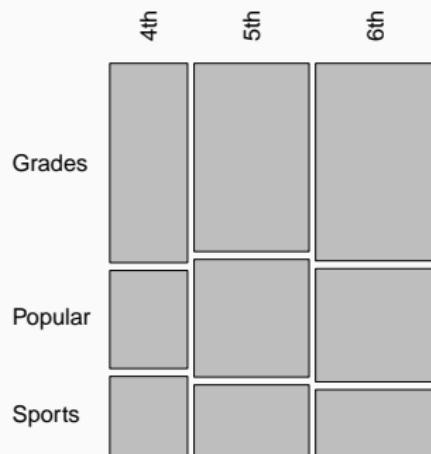
## **Chi-square test of independence**

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## Popular kids

In the dataset `popular`, students in grades 4-6 were asked whether good grades, athletic ability, or popularity was most important to them. A two-way table separating the students by grade and by choice of most important factor is shown below. Do these data provide evidence to suggest that goals vary by grade?

	Grades	Popular	Sports
4 <sup>th</sup>	63	31	25
5 <sup>th</sup>	88	55	33
6 <sup>th</sup>	96	55	32



## Chi-square test of independence

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$$\chi^2_{df} = \sum_{i=1}^k \frac{(O - E)^2}{E} \quad \text{where} \quad df = (R - 1) \times (C - 1),$$

where  $k$  is the number of cells,  $R$  is the number of rows, and  $C$  is the number of columns.

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**Note:** We calculate  $df$  differently for one-way and two-way tables.

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*Note:* We calculate  $df$  differently for one-way and two-way tables.

- The p-value is the area under the  $\chi^2_{df}$  curve, above the calculated test statistic.

## Expected counts in two-way tables

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$$E_{\text{row } 1, \text{col } 1} = \frac{119 \times 247}{478} = 61 \quad E_{\text{row } 1, \text{col } 2} = \frac{119 \times 141}{478} = 35$$

## Expected counts in two-way tables

What is the expected count for the highlighted cell?

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(b)  $\frac{119 \times 141}{478}$

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$$\rightarrow 52$$

*more than expected # of 5th graders  
have a goal of being popular*

## Calculating the test statistic in two-way tables

Expected counts are shown in *blue* next to the observed counts.

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$$\chi^2 = \sum \frac{(63 - 61)^2}{61} + \frac{(31 - 35)^2}{35} + \dots + \frac{(32 - 34)^2}{34} = 1.3121$$

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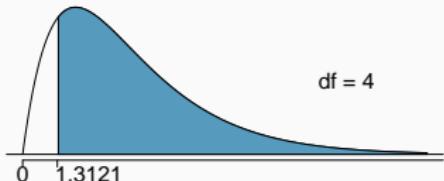
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$$\begin{aligned}\chi^2 &= \sum \frac{(63 - 61)^2}{61} + \frac{(31 - 35)^2}{35} + \dots + \frac{(32 - 34)^2}{34} = 1.3121 \\ df &= (R - 1) \times (C - 1) = (3 - 1) \times (3 - 1) = 2 \times 2 = 4\end{aligned}$$

## Calculating the p-value

Which of the following is the correct p-value for this hypothesis test?

$$\chi^2 = 1.3121 \quad df = 4$$



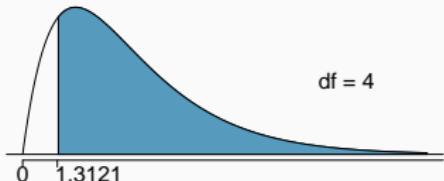
- (a) more than 0.3
- (b) between 0.3 and 0.2
- (c) between 0.2 and 0.1
- (d) between 0.1 and 0.05
- (e) less than 0.001

Upper tail		0.3	0.2	0.1	0.05	0.02	0.01	0.005	0.001
df	1	1.07	1.64	2.71	3.84	5.41	6.63	7.88	10.83
	2	2.41	3.22	4.61	5.99	7.82	9.21	10.60	13.82
	3	3.66	4.64	6.25	7.81	9.84	11.34	12.84	16.27
	4	4.88	5.99	7.78	9.49	11.67	13.28	14.86	18.47
	5	6.06	7.29	9.24	11.07	13.39	15.09	16.75	20.52

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## Conclusion

Do these data provide evidence to suggest that goals vary by grade?

$H_0$ : Grade and goals are independent. Goals do not vary by grade.

$H_A$ : Grade and goals are dependent. Goals vary by grade.

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*Since p-value is high, we fail to reject  $H_0$ . The data do not provide convincing evidence that grade and goals are dependent. It doesn't appear that goals vary by grade.*