#### Data Analytics with R

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Inference for Categorical Data

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Inference for a single proportion

#### Access to Toilet

A survey asks the question about access to toilet. Below is the distribution of responses from a 2018 survey:

Households with toilet	320
Households without toilet	350
Total	670

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p (a population proportion)

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

 $\hat{\rho}$  (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 proportion, p, and standard error equal to  $\sqrt{\frac{p(1-p)}{n}}$ .

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

But of course this is true only under certain conditions...

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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 Indian households, therefore we can assume that one respondent's response is independent of another.
- 2. Success-failure: 320 households have access to toilet (successes) and 350 failures, both are greater than 10.

We are given that n = 670,  $\hat{p} = 0.48$ , 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) 
$$0.48 \pm 1.96 \times \sqrt{\frac{0.48 \times 0.52}{670}}$$

(b) 
$$0.48 \pm 1.65 \times \sqrt{\frac{0.48 \times 0.52}{670}}$$

(c) 
$$0.48 \pm 1.96 \times \frac{0.48 \times 0.52}{\sqrt{670}}$$

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(a) 
$$0.48 \pm 1.96 \times \sqrt{\frac{0.48 \times 0.52}{670}} \rightarrow (0.44, 0.52)$$

(b) 
$$0.48 \pm 1.65 \times \sqrt{\frac{0.48 \times 0.52}{670}}$$

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$$\begin{array}{ll} 0.01 & \geq & 1.96 \times \sqrt{\frac{0.48 \times 0.52}{n}} \rightarrow \textit{Use $\hat{\rho}$ from previous study} \\ 0.01^2 & \geq & 1.96^2 \times \frac{0.48 \times 0.52}{n} \\ n & \geq & \frac{1.96^2 \times 0.48 \times 0.52}{0.01^2} \\ n & \geq & 9584.7 \end{array}$$

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$$n \geq 9584.7 \rightarrow n \text{ should be at least } 9585$$

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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:  $SE = \sqrt{rac{\hat{
    ho}(1-\hat{
    ho})}{n}}$
  - HT: calculate using the null value:  $SE = \sqrt{\frac{p_0(1-p_0)}{n}}$

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Since the p-value is low, we reject  $H_0$ . The data provide convincing evidence that more than 40% of Indian households have access to toilet.

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
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- Conditions:
  - independence
    - random sample and 10% condition
  - at least 10 successes and failures
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- Standard error:  $SE = \sqrt{\frac{p(1-p)}{n}}$ 
  - for CI: use  $\hat{p}$
  - for HT: use  $p_0$

Difference of two proportions

# Results from a Survey

A small pan-India survey asks a question about open defecation in India. Below are the distributions of responses from the survey as well as from a small survey conducted in villages in Tada:

	National Survey	Village Survey
Frequently	454	69
Sometimes	124	30
A little	52	4
Not at all	50	2
Total	680	105

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- *Point estimate*: Difference between the proportions of *sampled* Tada residents and *sampled* Indians who practice open defecation frequently.

$$\hat{p}_{Tada} - \hat{p}_{Ind}$$

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- We just need the appropriate standard error of the point estimate  $(SE_{\hat{p}_{Tada}-\hat{p}_{Ind}})$ , which is the only new concept.

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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- 2. *Independence between groups:* The sampled respondents from Tada and the Indian residents are independent of each other.
- 3. Success-failure:

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

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$$= (-0.108, 0.086)$$

Which of the following is the correct set of hypotheses for testing if the proportion of all Tada residents who practice open defecation frequently differs from the proportion of all Indians who do?

- (a)  $H_0: p_{Tada} = p_{Ind}$  $H_A: p_{Tada} \neq p_{Ind}$
- (b)  $H_0: \hat{p}_{Tada} = \hat{p}_{Ind}$  $H_A: \hat{p}_{Tada} \neq \hat{p}_{Ind}$
- (c)  $H_0: p_{Tada} p_{Ind} = 0$  $H_A: p_{Tada} - p_{Ind} \neq 0$
- (d)  $H_0: p_{Tada} = p_{Ind}$  $H_A: p_{Tada} < p_{Ind}$

Which of the following is the correct set of hypotheses for testing if the proportion of all Tada residents who practice open defecation frequently differs from the proportion of all Indians who do?

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Both (a) and (c) are correct.

# Flashback to working with one proportion

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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.

$$np_0 \ge 10$$
  $n(1-p_0) \ge 10$ 

## 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

$$\hat{\rho} = \frac{\# \ \textit{of} \ \textit{successes}_1 + \# \ \textit{of} \ \textit{successes}_2}{\textit{n}_1 + \textit{n}_2}$$

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$$p - value = 2 \times P(Z < -0.22)$$

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Total	105	680
ρ̂	0.657	0.668

$$Z = \frac{(\hat{\rho}_{Tada} - \hat{\rho}_{Ind})}{\sqrt{\frac{\hat{\rho}(1-\hat{\rho})}{n_{Tada}} + \frac{\hat{\rho}(1-\hat{\rho})}{n_{Ind}}}}$$

$$= \frac{(0.657 - 0.668)}{\sqrt{\frac{0.666 \times 0.334}{105} + \frac{0.666 \times 0.334}{680}}} = \frac{-0.011}{0.0495} = -0.22$$

$$p - value = 2 \times P(Z < -0.22) = 2 \times 0.41 = 0.82$$

- Population parameter:  $(p_1-p_2)$ , point estimate:  $(\hat{p}_1-\hat{p}_2)$ 

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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  $H_0: p_1 p_2 =$  (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	$SE = \frac{s}{\sqrt{n}}$	$SE = \sqrt{rac{s_1^2}{n_1} + rac{s_2^2}{n_2}}$
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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).



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- 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

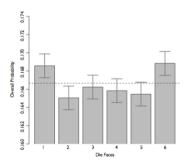
- The rolling-imaging process took about 20 seconds per roll.



- Each day there were  $\sim$ 150 images to process manually.
- At this rate Weldon's experiment was repeated in a little more than six full days.
- Recommended reading: http://galton.uchicago.edu/about/docs/labby09dice.pdf

## 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,  $2s, \dots, 6s$  would he expect to have observed?

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

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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
5	52,244	52,612
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Total	315,672	315,672

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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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- 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

- The general form of a test statistic is

point estimate — null value
SE of point estimate

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

#### Chi-square statistic

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 $\chi^2$  statistic

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

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When have we seen this before?

#### The chi-square distribution

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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.

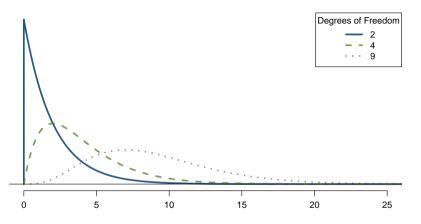
# The chi-square distribution

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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)

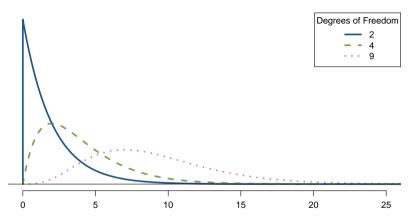
#### Which of the following is false?



As the df increases,

- (a) the center of the  $\chi^2$  distribution increases as well
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- (c) the shape of the  $\chi^2$  distribution becomes more skewed (less like a normal)

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- For this we can use technology, or a chi-square probability table.

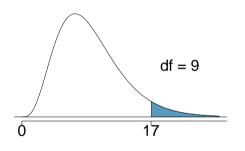
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```
> pchisq(q = 10, df = 6, lower.tail = FALSE)
```

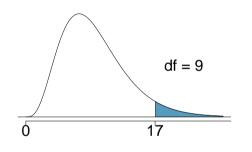
[1] 0.124652

Estimate the shaded area (above the cutoff value of 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

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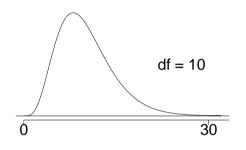


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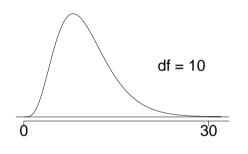
[1] 0.04871598

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
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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

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$$df = k - 1$$

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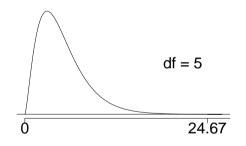
$$df = k - 1$$

- 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*.



p-value =  $P(\chi^2_{df=5}>24.67)$  is less than 0.001

# 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.
- (c) Fail to reject  $H_0$ , the data provide convincing evidence that the dice are fair.
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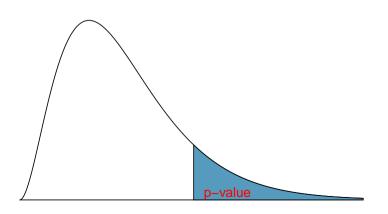
#### Turns out...

- 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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- 3. df > 1: Degrees of freedom must be greater than 1.

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.

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

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(1) Ahmedinajad	338	63.29%	504 × 0.6329 = 319
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	Observed # of Reported % of		Expected # of
Candidate	voters in poll	votes in election	votes in poll
(1) Ahmedinajad	338	63.29%	504 × 0.6329 = 319
(2) Mousavi	136	34.10%	$504 \times 0.3410 = 172$
(3) Minor candidates	30	2.61%	$504 \times 0.0261 = 13$
Total	504	100%	504

$$\frac{(O_1 - E_1)^2}{E_1} = \frac{(338 - 319)^2}{319} = 1.13$$

$$\frac{(O_2 - E_2)^2}{E_2} = \frac{(136 - 172)^2}{172} = 7.53$$

$$\frac{(O_2 - E_2)^2}{E_2} = \frac{(30 - 13)^2}{13} = 22.23$$

$$\chi^2_{df=3-1=2} = 30.89$$

#### 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 <u>not</u> 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
- (d) p-value is low,  $H_0$  is not rejected. The observed counts from the poll do <u>not</u> follow the same distribution as the reported votes.

#### Conclusion

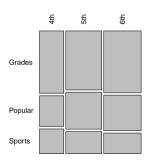
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## Popular kids

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



- The hypotheses are:

 $H_0$ : 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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- The test statistic is calculated as

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

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

*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.

$$\mathsf{Expected} \ \mathsf{Count} = \frac{(\mathsf{row} \ \mathsf{total}) \times (\mathsf{column} \ \mathsf{total})}{\mathsf{table} \ \mathsf{total}}$$

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	Grades	Popular	Sports	Total
4 <sup>th</sup>	63	31	25	119
5 <sup>th</sup>	88	55	33	176
6 <sup>th</sup>	96	55	32	183
Total	247	141	90	478

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

#### What is the expected count for the highlighted cell?

	Grades	Popular	Sports	Total
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$6^{th}$	96	55	32	183
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- (a)  $\frac{176 \times 141}{478}$
- (b)  $\frac{119\times141}{478}$
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ightarrow 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.

	Grades	Popular	Sports	Total
4 <sup>th</sup>	63 <mark>61</mark>	31 35	25 <mark>23</mark>	119
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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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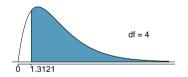
$$\chi^{2} = \sum \frac{(63-61)^{2}}{61} + \frac{(31-35)^{2}}{35} + \dots + \frac{(32-34)^{2}}{34} = 1.3121$$

$$df = (R-1) \times (C-1) = (3-1) \times (3-1) = 2 \times 2 = 4$$

# Calculating the p-value

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

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

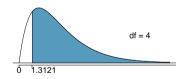


- (a) more than 0.3
- (b) between 0.3 and 0.2
- (c) between 0.2 and 0.1
- (d) between 0.1 and 0.05
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- (e) less than 0.001

#### 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.

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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.