# Chapter 6 Inference for categorical data<sup>1</sup>

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 $<sup>^{1}\</sup>mbox{These}$  notes use content from OpenIntro Statistics Slides by Mine Cetinkaya-Rundel.

# 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

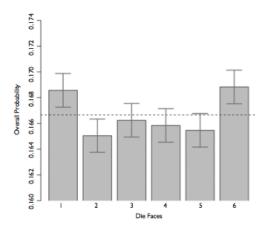
- In 2009, Zacariah Labby (U of Chicago), repeated Weldon's experiment using a homemade dice-throwing, pip counting machine. http://www.youtube.com/watch?v=95EErdouO2w
- The rolling-imaging process took about 20 seconds per roll.



- $\blacktriangleright$  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.

#### Labby's dice

- 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

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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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

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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

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

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- Any standardized difference that is squared will now be positive.
- ▶ Differences that already looked unusual will become much larger after being squared.

When have we seen this before?

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

#### The chi-square distribution

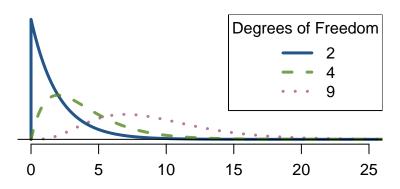
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#### Remember: So far we've seen three other continuous distributions:

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

#### **Practice**

Which of the following is false?

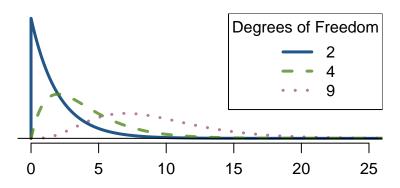


As the df increases,

- A) The center of the  $\chi^2$  distribution increases as well.
- B) The variability of the  $\chi^2$  distribution increases as well.
- C) The shape of the  $\chi^2$  distribution becomes more skewed (less like a normal).

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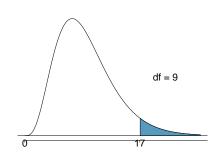
- A) The center of the  $\chi^2$  distribution increases as well.
- B) The variability of the  $\chi^2$  distribution increases as well.
- C) The shape of the <sup>2</sup> distribution becomes more skewed (less like a normal).

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

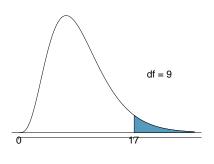
- p-value = tail area under the chi-square distribution (as usual).
- For this we can use technology, or chi-square probability table.

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

```
## [1] 0.124652
```



- 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

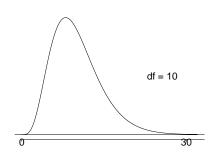


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```
pchisq(q = 17, df = 9, lower.tail = FALSE)
```

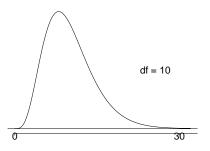
```
## [1] 0.04871598
```

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
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pchisq(q = 30, df = 10, lower.tail = FALSE)
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## [1] 0.0008566412
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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.

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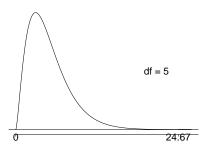
$$df = k - 1$$

 $\triangleright$  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.



 $\label{eq:p-value} \begin{aligned} \text{p-value} &= P(\chi^2_{df=5} > 24.67) \text{ is} \\ \text{less than } 0.001 \end{aligned}$ 

# 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  ${\cal H}_0$ , 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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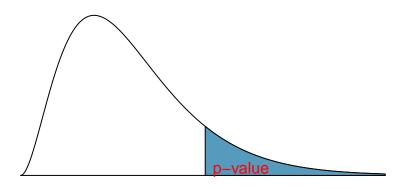
#### Turns out...

- ▶ 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 balance.



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

- ➤ The p-value for a chi-square test is define 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.

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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 <u>not</u> follow the same distribution as the reported votes.
- B) p-value is high,  ${\cal H}_0$  is not rejected. The observed counts from the poll follow the same distribution as the reported votes.
- C) p-value is low,  ${\cal H}_0$  is rejected. The observed counts from the poll follow the same distribution as the reported votes.
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