Brian Caffo

Chi-squared testing

Testing independence

equality of several proportions

Generalization

Independenc

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Goodness of

Mathematical Biostatistics Boot Camp 2: Lecture 8, Chi-Squared Tests

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Goodness of fit testing

Chi-squared testing

 An alternative approach to testing equality of proportions uses the chi-squared statistic

$$\sum \frac{(\mathsf{Observed} - \mathsf{Expected})^2}{\mathsf{Expected}}$$

- "Observed" are the observed counts
- "Expected" are the expected counts under the null hypothesis
- The sum is over all four cells
- This statistic follows a Chi-squared distribution with 1 df
- The Chi-squared statistic is exactly the square of the difference in proportions Score statistic

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Trt	Side Effects	None	Total
X	44	56	100
Y	77	43	120
	121	99	220

- p_1 and p_2 are the rates of side effects.
- $H_0: p_1 = p_2$

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• The χ^2 statistic is $\sum \frac{(O-E)^2}{E}$

•
$$O_{11} = 44$$
, $E_{11} = \frac{121}{220} \times 100 = 55$

•
$$O_{21} = 77$$
, $E_{21} = \frac{121}{220} \times 120 = 66$

•
$$O_{12} = 56$$
, $E_{12} = \frac{99}{220} \times 100 = 45$

•
$$O_{22} = 43$$
, $E_{22} = \frac{99}{220} \times 120 = 54$

$$\chi^2 = \frac{(44 - 55)^2}{55} + \frac{(77 - 66)^2}{666} + \frac{(56 - 45)^2}{45} + \frac{(43 - 54)^2}{54}$$

Which turns out to be 8.96. Compare to a χ^2 with one degree of freedom (reject for large values).

pchisq(8.96, 1, lower.tail = FALSE)
#result is 0.002

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```
dat <- matrix(c(44, 77, 56, 43), 2)
chisq.test(dat)
chisq.test(dat, correct = FALSE)</pre>
```

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Notation reminder

$n_{11}=x$	$n_{12}=n_1-x$	$n_1=n_{1+}$
$n_{21} = y$	$n_{22}=n_2-y$	$n_2 = n_{2+}$
n_{+1}	n_{+2}	n

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- Reject if the statistic is too large
- Alternative is two sided
- Do not divide α by 2
- A small χ^2 statistic implies little difference between the observed values and those expected under H_0
- The χ^2 statistic and approach generalizes to other kinds of tests and larger contingency tables
- Alternative computational form for the χ^2 statistic

$$\chi^2 = \frac{n(n_{11}n_{22} - n_{12}n_{21})^2}{n_{+1}n_{+2}n_{1+}n_{2+}}$$

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Notice that the statistic:

$$\chi^2 = \frac{n(n_{11}n_{22} - n_{12}n_{21})^2}{n_{+1}n_{+2}n_{1+}n_{2+}}$$

does not change if you transpose the rows and the columns of the table

- Surprisingly, the χ^2 statistic can be used
 - the rows are fixed (binomial)
 - the colums are fixed (binomial)
 - the total sample size is fixed (multinomial)
 - none are fixed (Poisson)
- For a given set of data, any of these assumptions results in the same value for the statistic

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- Maternal age versus birthweight¹
- Cross-sectional sample, only the total sample size is fixed

	Birthweight				
Mat. Age	$ e < 2500g \mid \geq 2,500g \mid Tot$				
< 20 <i>y</i>	20	80	100		
$\geq 20y$	30	270	300		
Total	50	350	400		

- H_0 : MA is independent of BW
- H_a: MA is not independent of BW



¹From Agresti Categorical Data Analysis second edition

Continued

- Estimated marginal probability of younger maternal age $P(MA < 20) = \frac{100}{400} = .25$
- Estimated marginal probability of low birth weight $P(BW < 2500) = \frac{50}{400} = .125$
- Under H_0 estimated cell probability of younger and low birth weight

$$P(MA < 20 \text{ and } BW < 2,500) = .25 \times .125$$

Therefore

•
$$E_{11} = \frac{100}{400} \times \frac{50}{400} \times 400 = 12.5$$

•
$$E_{12} = \frac{100}{400} \times \frac{350}{400} \times 400 = 87.5$$

$$E_{21} = \frac{300}{400} \times \frac{500}{400} \times 400 = 37.5$$

•
$$E_{11} = \frac{100}{400} \times \frac{50}{490} \times 400 = 12.5$$

• $E_{12} = \frac{100}{400} \times \frac{380}{400} \times 400 = 87.5$
• $E_{21} = \frac{380}{400} \times \frac{50}{400} \times 400 = 37.5$
• $E_{22} = \frac{300}{400} \times \frac{350}{400} \times 400 = 262.5$

nchisq(6.86.1.1ower.tail = F) = 0.09

•
$$\chi^2 = \frac{(20-12.5)^2}{12.5} + \frac{(80-87.5)^2}{87.5} \frac{(30-37.5)^2}{37.5} + \frac{(270-262.5)^2}{262.5} = 6.86$$

- Compare to critical value qchisq(.95, 1)=3.84
- Or calculate P-value

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Chi-squared testing cont'd

	Alcohol use			
Group	High	Low	Total	
Clergy	32	268	300	
Educators	51	199	250	
Executives	67	233	300	
Retailers	83	267	350	
Total	233	967	1,200	

2



²From Agresti's Categorical Data Analysis second edition

Testing equality of several proportions

 Interest lies in testing whether or not the proportion of high alcohol use is the same in the four occupations

•
$$H_0: p_1 = p_2 = p_3 = p_4 = p$$

• H_a : at least two of the p_i are unequal

•
$$O_{11} = 32$$
, $E_{11} = 300 \times \frac{233}{1200}$

•
$$O_{12} = 268$$
, $E_{12} = 300 \times \frac{967}{1200}$

• Chi-squared statistic
$$\sum \frac{(0-E)^2}{E} = 20.59$$

•
$$df = (Rows - 1)(Columns - 1) = 3$$

• Pvalue pchisq(20.59, 3, lower.tail = FALSE) ≈ 0

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Word distributions

	Book			
Word	1	2	3	Total
а	147	186	101	434
an	25	26	11	62
this	32	39	15	86
that	94	105	37	236
with	59	74	28	161
without	18	10	10	38
Total	375	440	202	1017

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- H₀: The probabilities of each word are the same for every book
- H_a : At least two are different
- $O_{11} = 147 \ E_{11} = 375 \times \frac{434}{1017}$
- $O_{12} = 186 \ E_{12} = 440 \times \frac{434}{1017}$
- ..
- $\sum \frac{(O-E)^2}{E} = 12.27$
- df = (6-1)(3-1) = 10

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Goodness of fit testing

Wife's Rating						
Husband	N	F	V	Α	Tot	
N	7	7	2	3	19	
F	2	8	3	7	20	
V	1	5	4	9	19	
Α	2	8	9	14	33	
	12	28	18	33	91	

N=never, F=fairly often, V=very often, A=almost always 4



⁴From Agresti's Categorical Data Analysis second edition

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H₀: H and W ratings are independent

• H_a : not independent

•
$$P(H = N \& W = A) = P(H = N)P(W = A)$$

•
$$stat = \sum \frac{(O-E)^2}{F}$$

•
$$O_{11} = 7 \ E_{11} = 91 \times \frac{19}{91} \times \frac{12}{91} = 2.51$$

•
$$E_{ij} = n_{i+}n_{+j}/n$$

•
$$df = (Rows - 1)(Cols - 1)$$

Independence

x < -matrix(c(7,7,2,3,2,8,3,7, 1,5,4,9, 2.8.9.14).4chisq.test(x)

$$\bullet \sum \frac{(O-E)^2}{E} = 16.96$$

•
$$df = (4-1)(4-1) = 9$$

•
$$p - value = .049$$

• Cell counts might be too small to use large sample approximation

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Goodness of

- Equal distribution and independence test yield the same results
- Same test results if
 - row totals are fixed
 - column totals are fixed
 - total ss is fixed
 - none are fixed
- Note that this is common in statistics; mathematically equivalent results are applied in different settings, but result in different interpretations

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- Chi-squared result requires large cell counts
- df is always (Rows 1)(Columns 1)
- Generalizations of Fishers exact test can be used or continuity corrections can be employed

Monte Carlo

Exact permutation test

 Reconstruct the individual data W: NNNNNNFFFFFFFVVAAANNFFFFFFF H: NNNNNNNNNNNNNNNNNFFFFFFFFF

- Permute either the W or H row
- Recalculate the contingency table
- Calculate the χ^2 statistic for each permutation
- Percentage of times it is larger than the observed value is an exact P-value chisq.test(x, simulate.p.value = TRUE)

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Results from R's RNG

	[0, .25)	[.25, .5)	[.5, .75)	[.75, 1)	Total
Count	254	235	267	244	1000
TP	. 25	. 25	. 25	.25	1

• H_0 : $p_1 = .25$, $p_2 = .25$, $p_3 = .25$, $p_4 = .25$

• H_a : any $p_i \neq i$ t's hypothesized value

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•
$$O_1 = 254 E_1 = 1000 \times .25 = 250$$

•
$$O_2 = 235 E_2 = 1000 \times .25 = 250$$

•
$$O_3 = 267 E_3 = 1000 \times .25 = 250$$

•
$$O_4 = 244 \ E_4 = 1000 \times .25 = 250$$

•
$$\sum \frac{(O-E)^2}{E} = 2.264$$

•
$$df = 3$$

•
$$P - value = .52$$

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Testing Mendel's hypothesis

	Phenotype			
	Yellow	Green	Total	
Observed	6022	2001	8023	
TP	.75	.25	1	
Expected	6017.25	2005.75	8023	

•
$$H_0: p_1 = .75, p_2 = .25$$

•
$$\sum \frac{(0-E)^2}{E} = \frac{(6022-6017.25)^2}{6017.25} + \frac{(2001-2005.75)^2}{2005.75} = .015$$

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- df = 1
- P-value = .90
- Fisher combined several of Mendel's tables
- $\sum \chi_{v_i}^2 \sim \chi_{\sum v_i}^2$
- Statistic 42, *df* = 84, P-value = .99996
- Agreement with theoretical counts is perhaps too good?

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Goodness of fit testing

- Test of whether or not observed counts equal theoretical values
- Test statistic is $\sum \frac{(0-E)^2}{E}$
- TS follows χ^2 distribution for large n
- df is the number of cells minus 1
- Especially useful for testing RNGs
- Kolmogorov/Smirnov test is an alternative test that does not require discretization