

6. Contingency tables – association of two (or more) categorical variables

Contingency tables – introduction

Contingency tables are tables that summarize frequencies (counts) of two (or more) categorical variables. Their analysis allows testing (in)dependence between the two variables. Table 6.1 is a contingency table summarizing frequencies of people of different eye and hair colors.

Table 6.1. Contingency table of two variables: eye and hair color with basic frequency statistics (marginal sums and grand total).

		Hair color			marginal sums
		black	brown	blonde	
Eye color	blue	12	45	14	71
	brown	51	256	84	391
marginal sums		63	301	98	grand total: 462

Basic analysis by goodness-of-fit test

Association between the variables (i.e. the **null hypothesis which states that the variables are independent**) can be tested by a goodness-of-fit test. This is a universal approach suitable for tables of any size and dimensions, but its explanatory power is limited.

For a goodness-of-fit test, we need expected frequencies under the null hypothesis, which are calculated on the basis of probability theory: $P(\text{event 1 and event 2}) = P(\text{event 1}) \times P(\text{event 2})$, if the two events are independent. In contingency tables, this can be used to calculate expected frequencies as the product of ratios of corresponding marginal totals and the grand total.

For instance, expected probability of observing a blue-eyed and black-haired person in Table 6.1 can be calculated as $P(\text{blueE and blackH}) = 63/462 \times 71/462 = 0.02096$. Multiplication of the probability then gives the expected frequency $\text{Freq}(e) = 0.02096 \times 462 = 9.68$.

The same approach can be used to calculate expected frequencies in all cells but is made automatically by software nowadays. The goodness-of-fit test can consequently be computed (in the same way as described in chapter 5). Note, however, that the number of degrees of freedom is determined as $DF = (\text{number of rows} - 1) \times (\text{number of columns} - 1)$

In our example (Table 6.1): **We did not find a significant association between eye and hair color ($\chi^2 = 0.785$, $DF = 2$, $p = 0.6755$).**

The goodness-of-fit test does not provide much more information on the result than the significance of the association. Still, in the case of a significant result, it may make sense to report also the difference between observed-expected frequencies (i.e. the residuals) or

their standardized values (residuals divided by square root of corresponding expected frequencies) as supplementary information. In particular, standardized residuals are helpful as they indicate excess or deficiency of which combinations cause the association between the variables.

2x2 tables and their analysis

These tables represent particular and the simplest cases of contingency tables (Table 6.2).

Table 6.2. Structure of a 2x2 table.

		Var2		
		level 1	level 2	
Var 1	level 1	f11	f12	R1
	level 2	f21	f22	R2
		C1	C2	<i>n</i>

Their simplicity allows additional statistics to be computed to express how tight the association between the two variables is. Most important of these is the **phi-coefficient**:

$$\phi = \frac{f_{11}f_{22} - f_{12}f_{21}}{\sqrt{R_1R_2C_1C_2}} = \pm \sqrt{\frac{\chi^2}{n}}$$

where *f*, *R*, and *C* symbols correspond to the cells in Table 6.2, and χ^2 is the χ^2 statistics of the table, and *n* is the grand total.

The phi-coefficient can thus be viewed as an average contribution of each observation to the association between the variables. This implies its advantage, which lies in the comparability of the phi coefficients between datasets with unequal numbers of observations.

The 2x2 tables may seem trivial and not of much use. However, they, and especially the phi-coefficient, are frequently used in vegetation ecology to measure the association between occurrences of two species or as a fidelity measure of a species with a vegetation unit. In that case, Var1 (as in Table 6.2) describes the frequency of given species and Var2 frequency of the vegetation unit in the dataset.

Advanced analysis of contingency tables – odds and odds ratios

Odds and odds ratios are additional important statistics that can be used to analyze contingency tables. They are defined for 2x2 tables only but can also be used in larger (in particular *n* x 2) tables subdivided into a series of 2x2 tables. For table 6.1, we can calculate the odds for level 1 of Var1 as:

$$\text{odds}_1 = p/(1-p) = (f_{11}/R_1)/(f_{12}/R_1)$$

where *p* is the probability of one outcome of the second variable and 1-*p* is the probability of the second outcome of the second variable. We can do the same for the second level of Var1 to get odds_2 . Odds ratio then equals:

$$OR = \text{odds}_1 / \text{odds}_2$$

The odds ratio directly indicates how the probability of observing level 1 of Var1 changes with respect to the levels of Var2.

OR values range between 0 and infinity, with $OR < 1$ indicating negative association, $OR = 1$ independence, and $OR > 1$ positive association.

OR is a population parameter. The computation summarized above is actually its maximum-likelihood estimation procedure. As a result, an OR estimate has associated standard error and confidence intervals (i.e. intervals within which the population OR lies with 95% probability). A confidence interval directly indicates significance – if a confidence interval of OR contains 1, the OR is not significantly different from 1, and thus, independence between the two variables cannot be rejected.

A worked example

Malaria is a dangerous disease widespread in tropical areas. It is caused by protozoans of the genus *Plasmodium* and transmitted by mosquitos. Preventing the infection is possible by taking prophylaxis, i.e. a treatment which blocks the disease after a mosquito bite. This is only possible for short-time journeys to malaria areas since the prophylaxis drugs are not safe for long-term use. Here we asked whether the prophylaxis is efficient and whether there is a significant difference between two prophylaxis types. The data are summarized in Table 6.3.

Table 6.3. Table summarizing frequencies of travelers to the tropics infected by malaria (or not) and anti-malaria prophylaxis they used.

Prophylaxis	Infected by malaria	Frequency
none (control)	0	40
none (control)	1	94
doxycycline	0	130
doxycycline	1	80
lariam	0	180
lariam	1	15

Note here that a contingency table can also have a form of a table with individual factor combinations and corresponding frequencies. This is actually a bit better for computation than the cross-tabulated form.

The goodness-of-fit test demonstrates that there is a significant association between the two variables:

$$\text{Chisq} = 137.45, \text{ df} = 2, \text{ p-value} = 1.42\text{e-}30$$

Odds ratios summary then follows. Two odds ratios are produced comparing the second and third levels to the first one (here control). The “lower” and “upper” values indicate limits of confidence intervals. We can see that both types of prophylaxis are associated with significantly decreased infection rates.

```

      infected
prophylax  0      p0  1      p1  oddsratio      lower      upper      p.value
control   40 0.1142857 94 0.49735450 1.00000000      NA      NA      NA
doxy     130 0.3714286 80 0.42328042 0.26186579 0.16479825 0.41610692 6.790312e-09
lariam   180 0.5142857 15 0.07936508 0.03546099 0.01862937 0.06749997 8.847446e-34

```

To compare just the two prophylaxis types, we can select just the corresponding part of the data for analysis (specifying this by square brackets in R). The result shows that taking Lariam is associated with a significantly lower infection rate than taking doxycycline.

```

      infected
prophylax  0      p0  1      p1  oddsratio      lower      upper      p.value
doxy     130 0.4193548 80 0.8421053 1.0000000      NA      NA      NA
lariam   180 0.5806452 15 0.1578947 0.1354167 0.07462922 0.2457171 1.531487e-13

```

In a paper/thesis, the result can be summarized as Table 6.4

Table 6.4. Summary of a contingency table analysis testing the association between malaria prophylaxis and infection. Overall test of independence $\chi^2 = 137.45$, $df = 2$, $p < 10^{-6}$.

	Odds ratio	lower 95% conf. limit	upper 95% conf. limit	p
Lariam vs. none	0.035	0.019	0.067	$< 10^{-6}$
doxycycline vs. none	0.262	0.165	0.416	$< 10^{-6}$
Lariam vs. doxycycline	0.135	0.075	0.246	$< 10^{-6}$

Coincidence and causality

Note here that significant results of a contingency table analysis indicate a significant association. This can be caused either by coincidence or causality. Causality means that if we manipulate one variable, the other also changes, i.e. one variable has a direct effect on the other. By contrast, coincidence may happen due to another variable affecting the two ones analyzed. In such a case, manipulation of one variable does not induce a change in the other variable.

In the malaria example, the travelers using prophylaxis are simultaneously more likely to use mosquito repellents, which is known to decrease infection risk strongly. Therefore, if somebody from the no-prophylaxis travelers decided to take prophylaxis, it may have a much lower effect than our analysis suggests.

People in general like causal explanations (and expect them). As a result, an association is frequently interpreted as a causal relationship, which is inappropriate. An association may only suggest causality at best, which can be consequently demonstrated by a **manipulative**

experiment. In our case, this would mean selecting a group of people, assign them randomly into three groups according to prophylaxis, send them to the tropics and see what happens. In this particular case, however, such research would not be approved by an ethics committee.

How to do in R

1. Chisq analysis of contingency tables

Option 1: apply **chisq.test** on matrix containing frequencies

Option 2: If the data are formatted in the data frame as in Table 6.3, they can be converted to contingency table by function **xtabs**

```
data.table<-xtabs(freq~var1+var2, data=data.frame)
```

chisq.test can then be applied to the contingency table. If its result is saved in an object:

```
test.res<-chisq.test(data.table)
```

running **test.res\$std.resid** can then be used to display standardized residuals.

2. Phi - coefficient

function **phi** (package psych) applied on a 2x2 matrix

3. Odds ratios

function **epitab** (package epitools) applied on contingency table produced by **xtabs**. Square brackets can be used to select the levels to compare.