

# **Module 4. Bivariate analysis: qualitative variables**

**Data Science & AI**

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# Learning goals

- Dependent/independent variable
- Apply suitable analysis techniques for each combination of measurement levels
- Contingency tables and Cramér's  $V$
- Visualization

# Bivariate analysis: overview

Independent	Dependent	Test/Metric
Qualitative	Qualitative	$\chi^2$ -test Cramér's $V$
Qualitative	Quantitative	two-sample $t$ -test Cohen's $d$
Quantitative	Quantitative	— Regression, correlation

# Bivariate analysis

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# Bivariate analysis

- ...is determining whether there is an association between two stochastic variables ( $X$  and  $Y$ ).
- **Association** = you can predict (to some extent) the value of  $Y$  from the value of  $X$ 
  - $X$  — Independent variable
  - $Y$  — Dependent variable
- **Important!** Finding an association does **NOT** imply a causal relation!

# Example research questions

- Is there a difference in taste preference between two beverage brands?
- Is there a difference in spending at the campus restaurant between students and staff?
- Do smokers die more often of lung cancer than non-smokers?
- Do men and women have a different opinion on a survey question?
- ...

We will use `data/rlanders.csv` from the Github repo for lab assignments to explore the last question.

# Contingency tables

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# Contingency tables

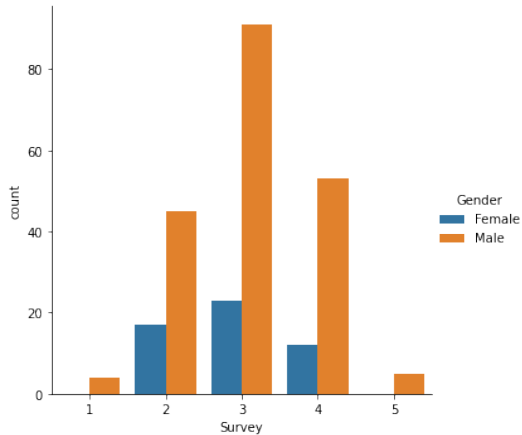
(also: crosstab)

See Python example code in `demo-chi-squared`

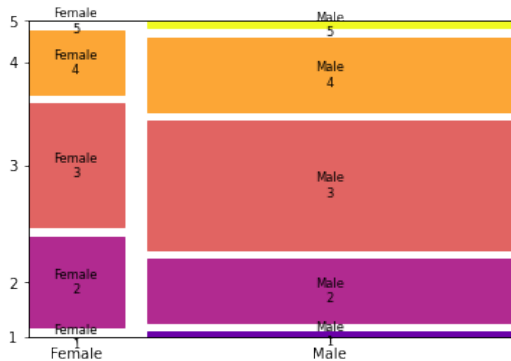
<b>Gender Survey</b>	<b>Female</b>	<b>Male</b>
Strongly disagree	0	4
Disagree	17	45
Neutral	23	91
Agree	12	53
Strongly agree	0	5

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# Visualization



A clustered bar chart



A mosaic plot

# Marginal totals

<b>Gender Survey</b>	<b>Female</b>	<b>Male</b>	<b>Total</b>
Strongly disagree	0	4	4
Disagree	17	45	62
Neutral	23	91	114
Agree	12	53	65
Strongly agree	0	5	5
<b>Total</b>	<b>52</b>	<b>198</b>	<b>250</b>

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# Expected values

If there is no difference (association), we expect the same ratios in each column of the table!

<b>Gender Survey</b>	<b>Female</b>	<b>Male</b>	<b>Total</b>
Strongly disagree	0.832	3.168	4
Disagree	12.896	49.104	62
Neutral	23.712	90.288	114
Agree	13.520	51.480	65
Strongly agree	1.040	3.960	5
<b>Total</b>	<b>52</b>	<b>198</b>	<b>250</b>

In each cell: (row total  $\times$  column total) /  $n$

# Measuring dispersion

How far is the observed value  $o$  from the expected  $e$ ?

$$\frac{(o - e)^2}{e}$$

<b>Gender Survey</b>	<b>Female</b>	<b>Male</b>
Strongly disagree	0.832	0.219
Disagree	1.306	0.343
Neutral	0.021	0.006
Agree	0.171	0.045
Strongly agree	1.040	0.273

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# The chi-squared statistic

The sum of all these values is notated:

$$\chi^2 = \sum_i \frac{(o_i - e_i)^2}{e_i} \approx 4.255$$

- $\chi$  is the Greek letter *chi*
- $o_i$  = number of observations in the  $i$ 'th cell of the contingency table
- $e_i$  = expected frequency
- Small value  $\Rightarrow$  no association
- Large value  $\Rightarrow$  association

# When is $\chi^2$ large enough?

- $2 \times 2$ -table with  $\chi^2 = 10$ 
  - Relatively large difference
  - Indicates association
- $5 \times 5$ -table with  $\chi^2 = 10$ 
  - Relatively small difference
  - Does NOT indicate association

We need a metric independent of table size!

# Cramér's V

$$V = \sqrt{\frac{\chi^2}{n(k-1)}} = \sqrt{\frac{4.255}{250(2-1)}} \approx 0.130$$

with  $n$  the number of observations,  $k = \min(\text{numRows}, \text{numCols})$

Cramér's V	Interpretation
$\approx 0$	no association
$\approx 0.1$	weak association
$\approx 0.25$	moderate association
$\approx 0.5$	strong association
$\approx 0.75$	very strong association
$\approx 1$	complete association

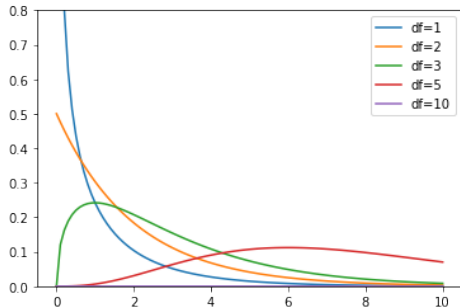


# Chi-squared test for independence

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# $\chi^2$ test for independence

- = Alternative to Cramér's V to investigate association between qualitative variables.
- Value of  $\chi^2$  distributed according to the  $\chi^2$  distribution



# $\chi^2$ -distribution in Python

Import `scipy.stats`

For a  $\chi^2$ -distribution with `df` degrees of freedom:

Function	Purpose
<code>chi2.pdf(x, df=d)</code>	Probability density for $x$
<code>chi2.cdf(x, df=d)</code>	Left-tail probability $P(X < x)$
<code>chi2.sf(x, df=d)</code>	Right-tail probability $P(X > x)$
<code>chi2.isf(1-p, df=d)</code>	$p\%$ of observations is expected to be lower than this value

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# Test procedure

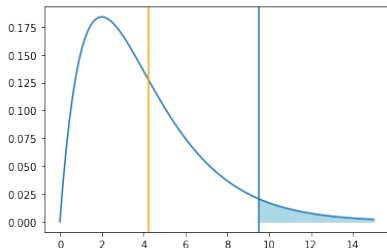
- **Step 1.** Formulate hypotheses:
  - $H_0$ : there is no association ( $\chi^2$  is “small”)
  - $H_1$ : there is an association ( $\chi^2$  is “large”)
- **Step 2.** Choose significance level, e.g.  $\alpha = 0.05$
- **Step 3.** Calculate the test statistic,  $\chi^2 = 4.255$

# Test procedure (cont.)

- **Step 4.** Use  $df = (numRow - 1) \times (numCol - 1)$  and either:
  - Determine critical value  $g$  so  $P(\chi^2 > g) = \alpha$
  - Calculate the  $p$ -value
- **Step 5.** Draw conclusion:
  - $\chi^2 < g$ : do not reject  $H_0$ ;  $\chi^2 > g$ : reject  $H_0$
  - $p > \alpha$ : do not reject  $H_0$ ;  $p < \alpha$ : reject  $H_0$

# Example (Gender vs Survey)

- `g = stats.chi2.isf(0.05, df=4)` (result: 9.4877)
- `p = stats.chi2.sf(4.2555, df=4)` (result: 0.3725)



**Conclusion:** we do not reject the null hypothesis. Differences between expected and observed values are not significantly large.  
We found no association between variables *Gender* and *Survey*

# Test for independence in Python

SciPy has a function to calculate  $\chi^2$  and  $p$ -value from a contingency table:

```
1 observed = pd.crosstab(rlanders.Survey, rlanders.Gender)
2 chi2, p, df, expected = stats.chi2_contingency(observed)
3
4 print("Chi-squared           : %.4f" % chi2)
5 print("Degrees of freedom: %d" % df)
6 print("P-value             : %.4f" % p)
```

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# Goodness-of-fit test

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# Goodness-of-fit test

The  $\chi^2$  test can also be used to determine whether a sample is **representative** for the population.

## Goodness-of-fit test

This test indicates to what degree a sample corresponds to a null hypothesis regarding the distribution of a qualitative variable over mutually exclusive classes.

# Example



Type	# sample	# population
Mutant	127	35%
Human	75	17%
Alien	98	23%
God	27	8%
Demon	73	17%
<b>Total</b>	400	100%

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# Example

Is the distribution of the sample ( $n = 400$ ) representative for the full population (all superheroes)?

- What numbers would you *expect* if the sample is representative?
- How large are the differences from the *observed* numbers?
  - small  $\Rightarrow$  distribution is representative
  - large  $\Rightarrow$  distribution is **not** representative

# Example

Is the distribution of the sample ( $n = 400$ ) representative for the full population (all superheroes)?

- What numbers would you *expect* if the sample is representative?
- How large are the differences from the *observed* numbers?
  - small  $\Rightarrow$  distribution is representative
  - large  $\Rightarrow$  distribution is **not** representative

Can you see the link with contingency tables and Cramer's V?

# Goodness of fit test

- Exactly representative  $\Rightarrow$  35% of superheroes in the sample is a mutant
- The expected number therefore is  $e = 0.35 \times 400 = 140$ .

Therefore:

$$e = n \times \pi$$

If the differences  $o - e$  are relatively small they can be attributed to random sampling errors.

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# Goodness of fit test

Consider  $\chi^2$ :

$$\chi^2 = \sum_{i=1}^n \frac{(o_i - e_i)^2}{e_i}$$

Draw a conclusion based on the value of  $\chi^2$ :

- small  $\Rightarrow$  distribution is representative
- large  $\Rightarrow$  distribution is **not** representative

$\chi^2$  measures the degree of conflict with the null hypothesis

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# Goodness of fit test

Superhero type	$o$	$\pi$	$e$	$\frac{(o-e)^2}{e}$
Mutant	127	35%	140	1.21
Human	75	17%	68	0.72
Alien	98	23%	92	0.39
God	27	8%	32	0.78
Demon	73	17%	68	0.37
			$\chi^2 =$	3.47

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# Goodness of fit test

- The test statistic  $\chi^2$  follows the  $\chi^2$  distribution.
- Critical value  $g$  from the  $\chi^2$  distribution: this is dependent on the number of degrees of freedom ( $df$ ). In general:

$$df = k - 1$$

with  $k$  the number of categories.

- The critical value  $g$  for a given significance level  $\alpha$  and number of degrees of freedom  $df$  can be calculated in Python using the function `isf()`.

$$P(\chi^2 < g) = 1 - \alpha$$



# Goodness of fit test

## Testing Procedure

### 1. **Formulate hypotheses**

- o  $H_0$ : sample is representative for the population
- o  $H_1$ : sample is not representative for the population

### 2. **Choose significance level:** $\alpha = 0.05$

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# Goodness of fit test

## Testing Procedure

### 1. Calculate test statistic:

$$\chi^2 = \sum_{i=1}^n \frac{(o_i - e_i)^2}{e_i}$$

1.1 **Critical area:** Calculate  $g$  so that  $P(\chi^2 < g) = 1 - \alpha$

1.2 **Probability value:** Calculate  $p = 1 - P(X < \chi^2)$

### 2. Conclusion (the test is always right-tailed):

2.1  $\chi^2 < g \Rightarrow$  do not reject  $H_0$ ,  $\chi^2 > g \Rightarrow$  reject  $H_0$

2.2  $p > \alpha \Rightarrow$  do not reject  $H_0$ ,  $p < \alpha \Rightarrow$  reject  $H_0$

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# Example

- `g = stats.chi2.isf(0.05, df=4)` (result: 9.4877)
- `p = stats.chi2.sf(3.4679, df=4)` (result: 0.4828)

**Conclusion.**  $\chi^2 \approx 3.47 < g \approx 9.4877$ , so we don't reject the null hypothesis.  
This sample is representative for the population.

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# Goodness-of-fit test in Python

```
1 observed = np.array([127, 75, 98, 27, 73])
2 expected_p = np.array([.35, .17, .23, .08, .17])
3 expected = expected_p * sum(observed)
4 chi2, p = stats.chisquare(f_obs=observed, f_exp=expected)
5
6 print("χ² = %.4f" % chi2)
7 print("p = %.4f" % p)
```

# Standardized residuals

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# Example: families

Consider all families with exactly 5 children in a given community.

# Example: families

Consider all families with exactly 5 children in a given community. When we look at the number of boys/girls, there are 6 possible combinations:

1. 5 boys
2. 4 boys, 1 girl
3. 3 boys, 2 girls
4. 2 boys, 3 girls
5. 1 boy, 4 girls
6. 5 girls

A survey was conducted regarding 1022 families with exactly 5 children

Are the observed numbers in the 6 classes representative for a population in which the probability of having a boy is equal to the probability of having a girl, or more concrete 0.5?

# Example

i	0	1	2	3	4	5
$o_i$	58	149	305	303	162	45



# Example

$i$	0	1	2	3	4	5
$o_i$	58	149	305	303	162	45

If the assumption is true, the probability  $\pi_i$  to have  $i$  boys is determined by a binomial distribution with parameters  $n = 5$  and  $p = 0.5$ . For example, the probability to have 2 boys out of 5 children is equal to:

$$(0.5)^2 \times (1 - 0.5)^{5-2} \times \binom{5}{2}$$

In general (you don't have to know why):

$$\pi_i = \binom{5}{i} \times 0.5^i \times 0.5^{5-i} = \frac{5!}{i!(5-i)!} \times 0.5^i$$

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# Example

$i$	0	1	2	3	4	5	Total
$o_i$	58	149	305	303	162	45	1022
$\pi_i$	0,031	0,156	0,313	0,313	0,156	0,031	1
$e_i$	31,9	159,7	319,4	319,4	159,7	31,9	1022
$\frac{(o-e)^2}{e}$	21,268	0,715	0,647	0,840	0,033	5,343	28,846
$r_i$	4,686	-0,921	-0,970	-1,105	0,199	2,348	

# Example

1. **Formulate both hypotheses**

- o  $H_0$ : the sample is representative for the population
- o  $H_1$ : the sample is not representative for the population

2. **Determine  $\alpha$  and  $n$ :**  $\alpha = 0.01$  and  $n = 1022$ .

3. **Value of the test statistic in the sample:**

$$\chi^2 = \sum_{i=1}^n \frac{(o_i - e_i)^2}{e_i} \approx 28.846$$

4. **Calculate and plot critical area:** The critical value is 15.0863. Our test statistic is inside the critical area, so we can reject  $H_0$ .

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# Standardized Residuals

## Standardized Residuals

The **standardized residuals** indicate which classes make the greatest contribution to the value of  $\chi^2$ .

$$r_i = \frac{o_i - n\pi_i}{\sqrt{n\pi_i(1 - \pi_i)}}$$

- $r_i \in [-2, 2] \Rightarrow$  “acceptable” values
- $r_i < -2 \Rightarrow$  underrepresented
- $r_i > 2 \Rightarrow$  overrepresented

**Conclusion:** families in which all children are of the same gender are overrepresented.

# Cochran's rules

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# Conditions

In order to apply the  $\chi^2$ -test, the following conditions must be met (Rule of Cochran)

1. For all categories, the expected frequency  $e$  must be greater than 1.
2. In a maximum of 20 % of the categories, the expected frequency  $e$  may be less than 5.