

# MS-E2112 Multivariate Statistical Analysis (5cr)

## Lecture 5: Bivariate Correspondence Analysis

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# Correspondence Analysis

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Correspondence analysis is a PCA-type method appropriate for analyzing **categorical variables**. The aim in bivariate correspondence analysis is to describe dependencies (correspondences) in a two-way contingency table.

# Example, Education and Salary

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In this lecture, we consider an example where we examine dependencies of categorical variables **education** and **salary**.

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## Frequency Tables

# Contingency Tables

We consider a sample of size  $n$  described by two qualitative variables,  $x$  with categories  $A_1, \dots, A_J$  and  $y$  with categories  $B_1, \dots, B_K$ . The number of individuals having the modality (category)  $A_j$  for the variable  $x$  and the modality  $B_k$  for the variable  $y$  is denoted by  $n_{jk}$ . Now the number of individuals having the modality  $A_j$  for the variable  $x$  is given by

$$n_{j.} = \sum_{k=1}^K n_{jk},$$

the number of individuals having the modality  $B_k$  for the variable  $y$  is given by

$$n_{.k} = \sum_{j=1}^J n_{jk},$$

and

$$n = \sum_{j=1}^J \sum_{k=1}^K n_{jk}.$$

# Contingency Tables

The data is often displayed as a two-way contingency table.

	$B_1$	$B_2$	$\cdots$	$B_K$	
$A_1$	$n_{11}$	$n_{12}$	$\cdots$	$n_{1K}$	$n_{1.}$
$A_2$	$n_{21}$	$n_{22}$	$\cdots$	$n_{2K}$	$n_{2.}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$A_J$	$n_{J1}$	$n_{J2}$	$\cdots$	$n_{JK}$	$n_{J.}$
	$n_{.1}$	$n_{.2}$	$\cdots$	$n_{.K}$	$n$

Table: Contingency table

# Example, Education and Salary

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We consider size 1000 sample of two categorical variables. Variable  $x$  Education is divided to categories  $A_1$  Primary School,  $A_2$  High School, and  $A_3$  University, and variable  $y$  Salary is divided to categories  $B_1$  low,  $B_2$  average, and  $B_3$  high.

# Example, Education and Salary

We display the Education and Salary data as a two-way contingency table.

	$B_1$	$B_2$	$B_3$	
$A_1$	150	40	10	200
$A_2$	190	350	60	600
$A_3$	10	110	80	200
	350	500	150	1000

Table: Contingency table

- In this sample of 1000 observations, there are 150 individuals that have Primary School education and low salary.
- In this sample of 1000 observations, there are 10 individuals that have Primary School education and high salary.
- In this sample of 1000 observations, there are 110 individuals that have University education and average salary.
- ...

The value of the numbers  $n_{jk}$  is naturally relative to the total number of observations,  $n$ . Thus it is preferable to analyze the contingency table in the form of joint relative frequencies. From the contingency table, it is straightforward to compute the associated relative frequency table ( $F$ ) where the elements of the contingency table are divided by the number of individuals  $n$  leading to  $f_{jk} = \frac{n_{jk}}{n}$ . The marginal relative frequencies are computed as

$$f_{j.} = \sum_{k=1}^K f_{jk}$$

and

$$f_{.k} = \sum_{j=1}^J f_{jk}.$$

# Contingency Tables

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	$B_1$	$B_2$	$\cdots$	$B_K$	
$A_1$	$f_{11}$	$f_{12}$	$\cdots$	$f_{1K}$	$f_{1.}$
$A_2$	$f_{21}$	$f_{22}$	$\cdots$	$f_{2K}$	$f_{2.}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$A_J$	$f_{J1}$	$f_{J2}$	$\cdots$	$f_{JK}$	$f_{J.}$
	$f_{.1}$	$f_{.2}$	$\cdots$	$f_{.K}$	1

Table: Table of relative frequencies

# Example, Education and Salary

	$B_1$	$B_2$	$B_3$	
$A_1$	0.15	0.04	0.01	0.20
$A_2$	0.19	0.35	0.06	0.60
$A_3$	0.01	0.11	0.08	0.20
	0.35	0.50	0.15	1

**Table:** Table of relative frequencies

- ▶ In this sample 15% of individuals have Primary School education and low salary.
- ▶ In this sample, 1% of individuals have Primary School education and high salary.
- ▶ In this sample, 11% of individuals have University education and average salary.
- ▶ ...

The frequency  $f_{jk}$  is the estimate of

$$p_{jk} = P(x \in A_j, y \in B_k),$$

and  $f_{.j}$  and  $f_{.k}$  are the estimates of

$$p_{.j} = P(x \in A_j),$$

and

$$p_{.k} = P(y \in B_k),$$

respectively.

## Row Profiles

The proportion of individuals that belong to category  $B_k$  for the variable  $y$  among the individuals that have the modality  $A_j$  for the variable  $x$  form the so called table of row profiles. The conditional frequencies for fixed  $j$  and all  $k$  are

$$f_{k|j} = \frac{n_{jk}}{n_{j.}} = \frac{n_{jk}/n}{n_{j.}/n} = \frac{f_{jk}}{f_{j.}}.$$

The frequency  $f_{k|j}$  is the estimate of

$$p_{k|j} = P(y \in B_k | x \in A_j).$$

# Row Profiles

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	$B_1$	$B_2$	$\dots$	$B_K$	
$A_1$	$\frac{f_{11}}{f_{1.}}$	$\frac{f_{12}}{f_{1.}}$	$\dots$	$\frac{f_{1K}}{f_{1.}}$	1
$A_2$	$\frac{f_{21}}{f_{2.}}$	$\frac{f_{22}}{f_{2.}}$	$\dots$	$\frac{f_{2K}}{f_{2.}}$	1
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$A_J$	$\frac{f_{J1}}{f_{J.}}$	$\frac{f_{J2}}{f_{J.}}$	$\dots$	$\frac{f_{JK}}{f_{J.}}$	1

Table: Row profiles

# Example, Education and Salary

	$B_1$	$B_2$	$B_3$	
$A_1$	0.75	0.20	0.05	1
$A_2$	0.32	0.58	0.10	1
$A_3$	0.05	0.55	0.40	1

Table: Row profiles

- ▶ In this sample 75% of the individuals that have Primary School education, have low salary.
- ▶ In this sample, 5% of the individuals that have Primary School education, have high salary.
- ▶ In this sample, 55% of the individuals that have University education, have average salary.
- ▶ ...

## Column Profiles

The proportion of individuals that belong to category  $A_j$  for the variable  $x$  among the individuals that have the modality  $B_k$  for the variable  $y$  form the table of column profiles. The conditional frequencies for fixed  $k$  and all  $j$  are

$$f_{j|k} = \frac{n_{jk}}{n_{.k}} = \frac{n_{jk}/n}{n_{.k}/n} = \frac{f_{jk}}{f_{.k}}.$$

The frequency  $f_{j|k}$  is the estimate of

$$p_{j|k} = P(x \in A_j | y \in B_k).$$

	$B_1$	$B_2$	$\dots$	$B_K$
$A_1$	$\frac{f_{11}}{f_{.1}}$	$\frac{f_{12}}{f_{.2}}$	$\dots$	$\frac{f_{1K}}{f_{.K}}$
$A_2$	$\frac{f_{21}}{f_{.1}}$	$\frac{f_{22}}{f_{.2}}$	$\dots$	$\frac{f_{2K}}{f_{.K}}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$A_J$	$\frac{f_{J1}}{f_{.1}}$	$\frac{f_{J2}}{f_{.2}}$	$\dots$	$\frac{f_{JK}}{f_{.K}}$
	$1$	$1$	$\dots$	$1$

Table: Column profiles

# Example, Education and Salary

	$B_1$	$B_2$	$B_3$
$A_1$	0.43	0.08	0.07
$A_2$	0.54	0.70	0.40
$A_3$	0.03	0.22	0.53
	1	1	1

Table: Column profiles

- ▶ In this sample 43% of the individuals that have low salary, have Primary School education.
- ▶ In this sample, 7% of the individuals that have high salary, have Primary School education.
- ▶ In this sample, 22% of the individuals that have average salary, have University education.
- ▶ ...

# Independence

# Independence

The variables  $x$  and  $y$  are independent if and only if for all  $j, k$  it holds that

$$P(x \in A_j, y \in B_k) = P(x \in A_j)P(y \in B_k),$$

$$P(x \in A_j | y \in B_k) = P(x \in A_j),$$

and

$$P(y \in B_k | x \in A_j) = P(y \in B_k).$$

These equalities can be estimated by

$$f_{jk} \approx f_{j.} f_{.k},$$

$$f_{j|k} = \frac{f_{jk}}{f_{.k}} \approx f_{j.},$$

and

$$f_{k|j} = \frac{f_{jk}}{f_{j.}} \approx f_{.k},$$

respectively.

We can now define the theoretical relative frequencies and theoretical frequencies under the assumption of independence as follows:

$$f_{jk}^* = f_{j.} f_{.k}$$

and

$$n_{jk}^* = \frac{n_{j.} n_{.k}}{n} = f_{jk}^* n.$$

# Example, Education and Salary

	$B_1$	$B_2$	$B_3$	
$A_1$	150	40	10	200
$A_2$	190	350	60	600
$A_3$	10	110	80	200
	350	500	150	1000

Table: Observed frequencies

	$B_1$	$B_2$	$B_3$	
$A_1$	70	100	30	200
$A_2$	210	300	90	600
$A_3$	70	100	30	200
	350	500	150	1000

Table: Theoretical frequencies under independence

# Example, Education and Salary

	$B_1$	$B_2$	$B_3$	
$A_1$	0.15	0.04	0.01	0.20
$A_2$	0.19	0.35	0.06	0.60
$A_3$	0.01	0.11	0.08	0.20
	0.35	0.50	0.15	1

Table: Observed relative frequencies

	$B_1$	$B_2$	$B_3$	
$A_1$	0.07	0.10	0.03	0.20
$A_2$	0.21	0.30	0.09	0.60
$A_3$	0.07	0.10	0.03	0.20
	0.35	0.50	0.15	1

Table: Theoretical relative frequencies under independence

## Attraction Repulsion Matrix

# Attraction Repulsion Matrix

The elements of the attraction repulsion matrix  $D$  are given by

$$d_{jk} = \frac{n_{jk}}{n_{jk}^*} = \frac{f_{jk}}{f_{jk}^*} = \frac{f_{jk}}{f_{j.} \cdot f_{.k}}.$$

Note that

$$d_{jk} > 1 \Leftrightarrow f_{jk} > f_{j.} \cdot f_{.k} \Leftrightarrow \\ f_{j|k} > f_{j.} \text{ and } f_{k|j} > f_{.k}.$$

and

$$d_{jk} < 1 \Leftrightarrow f_{jk} < f_{j.} \cdot f_{.k} \Leftrightarrow \\ f_{j|k} < f_{j.} \text{ and } f_{k|j} < f_{.k}.$$

If  $d_{jk} > 1$ , then the modalities (categories)  $A_j$  and  $B_k$  are said to be attracted to each other. If  $d_{jk} < 1$ , then the modalities  $A_j$  and  $B_k$  are said to repulse each other.

	$B_1$	$B_2$	$B_3$
$A_1$	2.14	0.40	0.33
$A_2$	0.90	1.16	0.67
$A_3$	0.14	1.10	2.67

Table: Attraction repulsion indices

- High salary is more frequent for people with University education.
- High salary is less frequent for people with a Primary School education.
- Low salary is less frequent for people with University education.
- ...

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Next week we will continue discussion about correspondence analysis.

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