

# Tutorial – Mathematics for Social Scientists

Winter semester 2024/25

Eigenvalues and Eigenvectors

# To do

- Weekly recap
- Real world applications
- Hands on practice
- Questions

# Chapter 14 | Eigenvectors & Eigenvalues

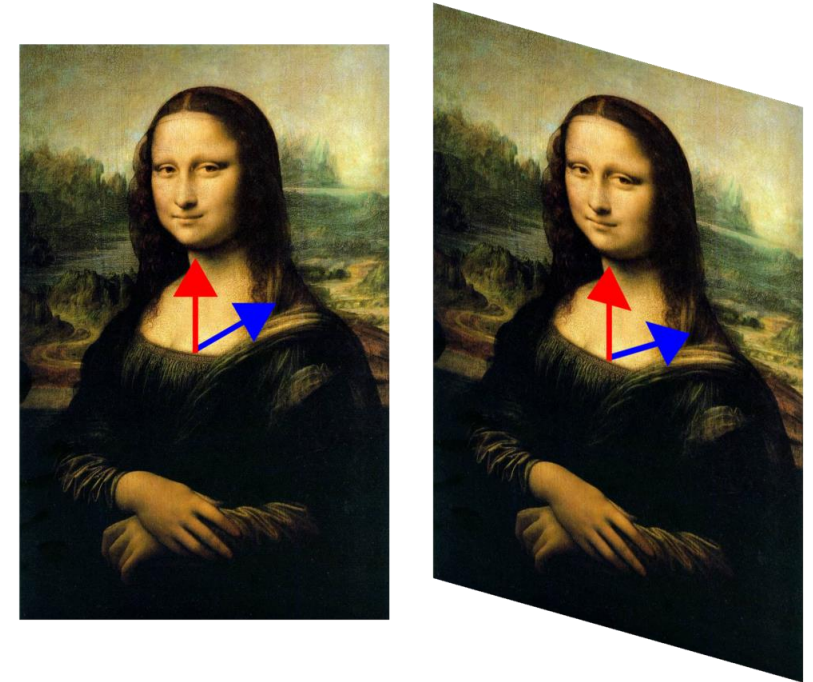
# Eigenpairs in the social sciences

For a square matrix  $A$ , **eigenvalues**  $\lambda$  and **eigenvectors**  $\vec{v}$  make the following equation true:

$$A\vec{v} = \lambda\vec{v}$$

... but **intuitively speaking**, what are they?

- **eigenvalues** are **characteristic** for matrix  $A$
  - **eigenvectors** are **fixed** in **direction** when  $A$  is applied to them
- this is why scalar multiplication must be defined for vector spaces! A 2D graphic is defined on a plane using x- and y-coordinates



Source picture: [https://de.wikipedia.org/wiki/Eigenwerte\\_und\\_Eigenvektoren#/media/Datei:Mona\\_Lisa\\_with\\_eigenvector.png](https://de.wikipedia.org/wiki/Eigenwerte_und_Eigenvektoren#/media/Datei:Mona_Lisa_with_eigenvector.png), 16.20.2023

# Eigenpairs in the social sciences

Let's recall our **calculus** and **analysis** sessions

- why do we study **minima** and **maxima** of **functions**?
  - why do we try to solve for **roots** of **polynomials**?
- they offer **information** about our **system**! This is what we want to do to gain **inference** 😊

**Eigenvectors** (and **–values**) are important, because they **describe** and **limit** the **shape** of our **system**!

- they bound a line (or plane) into shape by restraining its linear transformation!
- we are interested in eigenpairs because they describe the shape and behaviour of our system based on changes in data (our matrix  $A$ ) → **linear transformation**

# Eigenvalues

Eigenvalues  $\lambda$  are specific scalars to a quadratic matrix  $A$  and define, how certain eigenvectors  $\vec{v}$  change under linear transformation:

$$A\vec{v} = \lambda\vec{v}$$

- ‘eigen’ is German for ‘own’ or in this context ‘characteristic
- eigenvalues can be understood as ‘characteristic values’
- or to stick with our previous example:
  - characteristic ‘roots’ that are found when we solve  $A$ ’s characteristic polynomial

$$\lambda^2 + \lambda + \beta = 0 \leftarrow \text{for } A_{2 \times 2}$$

# Eigenvalues – Algorithm

## Algorithm:

- 1) multiply RHS by identity matrix

$$A\vec{v} = \lambda\vec{v} \rightarrow A\vec{v} = \lambda I\vec{v}$$

- 2) factor by  $\vec{v}$  and set equal to zero

$$\vec{v}(A - \lambda I) = 0$$

- 3) compute determinant

$$|A - \lambda I| = [(a_{11} - \lambda) \cdot (a_{22} - \lambda)] - [(a_{12} \cdot a_{21})]$$

- 4) solve for  $\lambda$

$$\lambda^2 + \lambda + \beta = 0$$

# Computing Eigenvalues

**Example:**  $A = \begin{bmatrix} 6 & 1 \\ -2 & 3 \end{bmatrix}$

2) factor by  $\vec{v}$  and set equal to zero:  $\vec{v} (A - \lambda I) = 0$

1)  $A\vec{v} = \lambda\vec{v} \rightarrow A\vec{v} = \lambda I\vec{v}$

$$\begin{bmatrix} 6 & 1 \\ -2 & 3 \end{bmatrix} \vec{v} = \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \vec{v}$$

$$\vec{v} \left( \begin{bmatrix} 6 & 1 \\ -2 & 3 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) = 0$$

$$\vec{v} \left( \begin{bmatrix} 6 - \lambda & 1 \\ -2 & 3 - \lambda \end{bmatrix} \right) = 0$$



# Computing Eigenvalues

## Example:

3) compute determinant  $|A - \lambda I|$

$$|A - \lambda I| = [(6 - \lambda) \cdot (3 - \lambda)] - [1 \cdot (-2)] = \lambda^2 - 9\lambda + 20$$

4) solve for  $\lambda$

$$\begin{aligned}\lambda^2 - 9\lambda + 18 &= 0 \\ \lambda_{1,2} &= \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} = \frac{-(-9) \pm \sqrt{(-9)^2 - 4 \cdot 1 \cdot 20}}{2} = \frac{9 \pm \sqrt{1}}{2} \\ \lambda_1 &= 5 \\ \lambda_2 &= 4\end{aligned}$$

# Hands on – Eigenvalues

**Task:** Compute the eigenvalues for the respective matrices below!

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix}$$

$$2) B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix}$$



# Hands on – Eigenvalues

**Solution:**

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix} \vec{v} = \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \vec{v}$$

$$\vec{v} \left( \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) = 0$$

$$\vec{v} \left( \begin{bmatrix} 4 - \lambda & 3 \\ 6 & 1 - \lambda \end{bmatrix} \right) = 0$$

$$\begin{aligned} |A - \lambda I| &= [(4 - \lambda)(1 - \lambda)] - [3 \cdot 6] \\ &= \lambda^2 - 5\lambda - 14 \end{aligned}$$

$$\lambda_{1,2} = \frac{-(-5) \pm \sqrt{(-5)^2 - 4 \cdot 1 \cdot (-14)}}{2}$$

$$\lambda_1 = 7$$

$$\lambda_2 = (-2)$$

# Hands on – Eigenvalues

**Solution:**

$$2) B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix}$$

$$\begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix} \vec{v} = \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \vec{v}$$

$$\vec{v} \left( \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) = 0$$

$$\vec{v} \left( \begin{bmatrix} (-8) - \lambda & -4 \\ 6 & 2 - \lambda \end{bmatrix} \right) = 0$$

$$\begin{aligned} |A - \lambda I| &= [((-8) - \lambda)(2 - \lambda)] - [6 \cdot (-4)] \\ &= \lambda^2 + 6\lambda + 8 \end{aligned}$$

$$\lambda_{1,2} = \frac{-6 \pm \sqrt{6^2 - 4 \cdot 1 \cdot 8}}{2}$$

$$\lambda_1 = (-2)$$

$$\lambda_2 = (-4)$$

# Eigenvectors

An **eigenvector**  $\vec{v}$  to a **quadratic matrix**  $A$  over an **object**  $K$  characterizes its linear transformation:

- any eigenvector  $\vec{v}$  must be **different** from the **Nullvector**  $\vec{0}$
- an eigenvector  $\vec{v}$  is **NOT 'definitely' defined** as it can be **scaled** by  $A$ 's **eigenvalues**  $\lambda$
- iff  $\vec{v}$  is an eigenvector, every  $\alpha\vec{v} \in K$  will be an eigenvector, too
- such a **scaling / linear transformation** does **not change** the **direction** of  $\vec{v}$ 
  - an **eigenvalue**  $\lambda$  has an **infinite** number of **eigenvectors**  $\vec{v}$
  - an **eigenvector**  $\vec{v}$  only belongs to **ONE eigenvalue**  $\lambda$
- **Note:** Only **quadratic matrices** have eigenvectors/values

# Eigenvectors – Algorithm

## Algorithm:

- 1) start with  $A\vec{v} = \lambda\vec{v}$  and plug in the eigenvalues

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \lambda_1 \begin{bmatrix} x \\ y \end{bmatrix}$$

- 2) multiply to get system of equations

$$\begin{aligned} a_{11}x + a_{12}y &= \lambda_1 x \\ a_{21}x + a_{22}y &= \lambda_1 y \end{aligned}$$

- 3) plug in  $x = 1$ ...

- 4) ... and solve for variables  $y, \dots, n$  to find eigenvector  $\begin{pmatrix} x \\ y \\ \vdots \\ n \end{pmatrix}$

# Computing Eigenvectors

**Example:**  $A = \begin{bmatrix} 6 & 1 \\ -2 & 3 \end{bmatrix}$  with eigenvalues  $\lambda_1 = 5, \lambda_2 = 4$

1) and 2) start with  $A\vec{v} = \lambda\vec{v}$ , plug in eigenvalues, multiply & simplify

$$6x + y = 5x$$

$$-2x + 3y = 5y$$

3) and 4) plug in  $x = 1$  and solve for  $y$

**starting with  $\lambda_1 = 5$**

$$6 \cdot 1 + y = 5 \cdot 1 \rightarrow y = (-1)$$

$$(-2) \cdot 1 + 3y = 5y \rightarrow (-2) = 2y \rightarrow y = (-1)$$



# Computing Eigenvectors

**Example:**  $A = \begin{bmatrix} 6 & 1 \\ -2 & 3 \end{bmatrix}$  with eigenvalues  $\lambda_1 = 5, \lambda_2 = 4$

3) and 4) plug in  $x = 1$  and solve for  $y$

starting with  $\lambda_2 = 4$

$$6 \cdot 1 + y = 4 \cdot 1 \rightarrow y = (-2)$$

$$(-2) \cdot 1 + 3y = 4y \rightarrow y = (-2)$$

→ our eigenvectors for  $x = 1$  are  $\vec{v}_1 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$  and  $\vec{v}_2 = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$

# Hands on – Eigenvectors

**Task:** Compute the eigenvectors for the following matrices!

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix}$$

$$2) B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix}$$

# Hands on – Eigenvectors

**Solution:**

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix}$$

•  **$A\vec{v} = \lambda\vec{v}$  and set up SoE**

$$\begin{cases} 4x + 3y = 7x \\ 6x + y = 7y \end{cases}$$

• **plug in  $x = 1$  and  $\lambda = 7$  solve for  $y$**

$$\begin{cases} 4 \cdot 1 + 3y = 7 \cdot 1 \\ 6 \cdot 1 + y = 7y \end{cases} \Rightarrow \begin{cases} 3y = 3 \rightarrow y = 1 \\ 6y = 6 \rightarrow y = 1 \end{cases} \rightarrow \vec{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

• **plug in  $x = 1$  and  $\lambda = (-2)$  solve for  $y$**

$$\begin{cases} 4 \cdot 1 + 3y = (-2) \cdot 1 \\ 6 \cdot 1 + y = (-2)y \end{cases} \Rightarrow \begin{cases} 3y = (-6) \rightarrow y = (-2) \\ (-3)y = 6 \rightarrow y = (-2) \end{cases} \rightarrow \vec{v}_2 = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$$

# Hands on – Eigenvectors

**Solution:**

$$2) \ B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix}$$

- **$A\vec{v} = \lambda\vec{v}$  and set up SoE**

$$\begin{cases} (-8)x + (-4)y = (-2)x \\ 6x + 2y = (-2)y \end{cases}$$

- **plug in  $x = 1$  and  $\lambda = (-2)$  solve for  $y$**

$$\begin{cases} (-8) \cdot 1 + (-4)y = (-2) \cdot 1 \\ 6 \cdot 1 + 2y = (-2)y \end{cases} \rightarrow \begin{cases} (-4)y = 6 \rightarrow y = -1.5 \\ (-4)y = 6 \rightarrow y = -1.5 \end{cases} \rightarrow \vec{v}_1 = \begin{pmatrix} 1 \\ -1.5 \end{pmatrix}$$

- **plug in  $x = 1$  and  $\lambda = (-4)$  solve for  $y$**

$$\begin{cases} (-8) \cdot 1 + (-4)y = (-4) \cdot 1 \\ 6 \cdot 1 + 2y = (-4)y \end{cases} \rightarrow \begin{cases} (-4)y = (-4) \rightarrow y = (-1) \\ (-6)y = 6 \rightarrow y = (-1) \end{cases} \rightarrow \vec{v}_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

# Real world applications – Eigenvalues & -vectors

## Quantum mechanics and Wave Theory

- incredibly important in quantum mechanics, physics, maths and chemistry... basically all the nature sciences
- anything that can be described as an oscillating system (musical instrument, particles, molecules, bridges...) relies on eigenpairs
  - these systems have ‘favourite’ frequencies – so called resonance frequencies – which are described using eigenfunctions, -values, -vectors
  - we WANT to encourage oscillation in musical instruments
  - we DO NOT want to encourage oscillation in bridges... (see below!)
- <https://www.youtube.com/watch?v=j-zczJXSxnw>

# Matrix diagonalization

**A quadratic matrix  $A$  is called diagonalizable, if:**

- there is an inverse  $Q$  with which the result of  $Q^{-1}AQ$  is a diagonal matrix
- geometric and algebraic multiplicity of eigenvalues (and eigenvectors) are equal
- $A$ 's characteristic polynomial can be expressed entirely using linear factors
- $A$ 's characteristic polynomial has as many roots as  $A$  has dimensions

$$A = \begin{bmatrix} 3 & 2 & 2 \\ 1 & 6 & 5 \\ 1 & 0 & 1 \end{bmatrix} \rightarrow D_A \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 7 \end{bmatrix}$$

# Matrix diagonalization – Algorithm

## Algorithm:

- 1) Find  $A$ 's eigenvalues
- 2) Check, if  $A$  is diagonalizable:

→ Find  $A$ 's eigenvectors

→ Find  $A$ 's characteristic polynomial and its roots

← if you're tasked with finding  $D_A$ ,  
you can most likely expect  $A$  to be  
diagonalizable!

3) Find diagonal matrix  $D_A = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}$

# Matrix diagonalization – Inverse matrix

## How to find the inverse of a diagonal matrix?

→ if you multiply a diagonal matrix with its inverse, the result will be the identity matrix!

**Note:** the inverse of  $D_A$  is composed of  $D_A$ 's reciprocal values!

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 7 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{7} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



# Hands on – Matrix diagonalization

**Task:** Diagonalize the following matrices!

→ **Hint:** Solution will be of form  $D_A = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix}$$

$$2) B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix}$$

# Hands on – Matrix diagonalization

**Solution:**

$$1) A = \begin{bmatrix} 4 & 3 \\ 6 & 1 \end{bmatrix} \rightarrow D_A = \begin{bmatrix} 7 & 0 \\ 0 & (-2) \end{bmatrix}$$

$$2) B = \begin{bmatrix} -8 & -4 \\ 6 & 2 \end{bmatrix} \rightarrow D_B = \begin{bmatrix} (-2) & 0 \\ 0 & (-4) \end{bmatrix}$$

# Real world applications – Matrix diagonalization

## Why do we diagonalize matrices?

→ because it makes our lives **easier!**

- matrix multiplication
- scalar multiplication
- addition and subtraction
- inverses...
- transpose matrices...
- you name it

→ diagonal matrices are of a ‘very-easy-to-work-with’ form ;)

# Matrices and OLS

Linear model  $\rightarrow$  OLS

you should know this:  $y = \beta_0 + \beta_1 x_1 + \varepsilon$

but do you know this:  $\vec{y} = X\vec{b} + \vec{e}$

$$\vec{e} = \vec{y} - X\vec{b}$$

minimized sum of squared  
residuals  $\vec{e}^T \vec{e}$

$$\vec{e}^T \vec{e} = (\vec{y} - X\vec{b})^T (\vec{y} - X\vec{b})$$

$$X^T X \vec{b} = X^T \vec{y}$$

after minimization:  
if  $X^T X$  is non-singular

$$\hat{\beta} = (X^T X)^{-1} X^T \vec{y}$$

# Matrices and OLS – Epidemiology

**Source:** Mathematics II for Chemistry, Life Science and Nano Science | Dr. Stefan Frei

$t_k$	0	1	2	3
$i_k$	58 716	56 998	54 984	53 450
$j_k := \ln(i_k)$	10.98	10.95	10.91	10.89

Source: RKI

let's estimate the 'Reproduction Figure  $R_0$ '  
for Germany back in spring 2020...

Assumption: Infektionszahl follows the relationship:  $i(t) = i_0 \exp\left(\frac{(R_0 - 1)t}{t_{int}}\right)$

→ let's scale the exp.  
growth!

$$j(t) := \ln(i(t)) = \underbrace{\ln(i_0)}_{j_0} + \frac{(R_0 - 1)t}{t_{int}}$$

# Matrices and OLS – Epidemiology

we want to model:  $j(t_k) = j_k \Rightarrow j_0 + \frac{t_k}{t_{inf}} R_0 = \frac{t_k}{t_{inf}} j_k$  for  $k=1, \dots, 4$   
(over time)

set  $t_{inf}$  to 6 days  $\Rightarrow t_{inf}=6$

$$\begin{pmatrix} 1 & \frac{t_1}{6} \\ 1 & \frac{t_2}{6} \\ 1 & \frac{t_3}{6} \\ 1 & \frac{t_4}{6} \end{pmatrix} \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} \frac{t_1}{6} + j_1 \\ \frac{t_2}{6} + j_2 \\ \frac{t_3}{6} + j_3 \\ \frac{t_4}{6} + j_4 \end{pmatrix} \Rightarrow \underbrace{\begin{pmatrix} 1 & 0 \\ 1 & \frac{1}{6} \\ 1 & \frac{1}{3} \\ 1 & \frac{1}{2} \end{pmatrix}}_A \underbrace{\begin{pmatrix} j_0 \\ R_0 \end{pmatrix}}_{=\vec{x}} = \underbrace{\begin{pmatrix} 10.98 \\ 11.12 \\ 11.24 \\ 11.39 \end{pmatrix}}_{=\vec{b}}$$

↑  
parameters

# Matrices and OLS – Epidemiology

$$\rightarrow A^T A \bar{x} = A^T \bar{b} : \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 44.73 \\ 11.29 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0 \\ \frac{1}{6} \\ \frac{1}{3} \\ \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 10.98 \\ 11.12 \\ 11.24 \\ 11.39 \end{pmatrix}$$

$$A^T A =$$

$$\begin{pmatrix} (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) & (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) \\ (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) & (0 \cdot 0) + (\frac{1}{6} \cdot \frac{1}{6}) + (\frac{1}{3} \cdot \frac{1}{3}) + (\frac{1}{2} \cdot \frac{1}{2}) \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} = A^T A$$

# Matrices and OLS – Epidemiology

$$\rightarrow A^T A \bar{x} = A^T \bar{b} : \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 44.73 \\ 11.29 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 1 & \frac{1}{6} \\ 1 & \frac{1}{3} \\ 1 & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 10.98 \\ 11.12 \\ 11.24 \\ 11.39 \end{pmatrix}$$

$$A^T A =$$

$$\begin{pmatrix} (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) & (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) \\ (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) & (0 \cdot 0) + (\frac{1}{6} \cdot \frac{1}{6}) + (\frac{1}{3} \cdot \frac{1}{3}) + (\frac{1}{2} \cdot \frac{1}{2}) \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} = A^T A$$



# Matrices and OLS – Epidemiology

$$\rightarrow A^T A \bar{x} = A^T \bar{b} : \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 44.73 \\ 11.29 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 1 & \frac{1}{6} \\ 1 & \frac{1}{3} \\ 1 & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 10.98 \\ 11.12 \\ 11.24 \\ 11.39 \end{pmatrix}$$

$$A^T A =$$

$$\begin{pmatrix} (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) & (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) \\ (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) & (0 \cdot 0) + (\frac{1}{6} \cdot \frac{1}{6}) + (\frac{1}{3} \cdot \frac{1}{3}) + (\frac{1}{2} \cdot \frac{1}{2}) \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} = A^T A$$

# Matrices and OLS – Epidemiology

$$\rightarrow A^T A \bar{x} = A^T \bar{b} : \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 44.73 \\ 11.29 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 1 & \frac{1}{6} \\ 1 & \frac{1}{3} \\ 1 & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} j_0 \\ R_0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & \frac{1}{6} & \frac{1}{3} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 10.98 \\ 11.12 \\ 11.24 \\ 11.39 \end{pmatrix}$$

$$A^T A =$$

$$\begin{pmatrix} (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) + (1 \cdot 1) & (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) \\ (0 \cdot 1) + (\frac{1}{6} \cdot 1) + (\frac{1}{3} \cdot 1) + (\frac{1}{2} \cdot 1) & (0 \cdot 0) + (\frac{1}{6} \cdot \frac{1}{6}) + (\frac{1}{3} \cdot \frac{1}{3}) + (\frac{1}{2} \cdot \frac{1}{2}) \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 1 & \frac{7}{18} \end{pmatrix} = A^T A$$

# Matrices and OLS – Epidemiology

→ apply Gauss-Elimination:  $\sqrt{II + (-\frac{1}{4})I}$

$$\left( \begin{array}{cc|c} 4 & 1 & 44.73 \\ 1 & \frac{7}{18} & 11.29 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 4 & 1 & 44.73 \\ 0 & \frac{5}{36} & 0.1075 \end{array} \right)$$

→ substitute backwards:

$$x_2 = R_0 = 0.1075 \cdot \frac{36}{5} \approx 0.774$$
$$x_1 = j_0 \approx \frac{1}{4} (44.73 - 0.774) = 10.989$$

$$\Rightarrow i_0 = \exp(j_0) \approx 59219$$

$$\Rightarrow i(t) = 59219 \exp\left(\frac{0.774-1}{t_{int}} t\right)$$

# Time for your questions

- Any questions during the week?
  - [joerdis.strack@uni-konstanz.de](mailto:joerdis.strack@uni-konstanz.de)

