



AARHUS
UNIVERSITY

Class 1: Introduction + Python

Computational Analysis of Text, Audio, and Images, Fall 2023

Aarhus University

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Aarhus University

Today's Menu

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“Hello World!”

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Me:

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Me:

- PhD student
- trained political scientists with a special focus on machine learning and 'data science'
- Research focuses on the behavior of political elites (\approx politicians) using unstructured data sources, in particular text and audio
- Lives with my wife and daughter in Copenhagen, a sports enthusiast

You:

- name
- favorite course so far
- level of programming
- expectation(s)

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- Five 'main topics'
- Within each 'main topic', two or three 'topics'
- One 'topic' \approx One 'class'
- Within each 'topic', one or two 'subtopics'

Class	Date	Main Topic		Topic
1	September 6	Python	I	Introduction to course + Python
2	September 7		II	Python lab session
3	September 13	ML Basics	I	Learning from data
4	September 20		II	ML lab session
5	September 27		III	Neural nets
6	October 4	Text	I	Text basics
7	October 11		II	Topic models and dictionaries
8	October 25		III	Embeddings
9	November 1	Audio	I	Audio basics
10	November 1		II	Audio measurement
11	November 8		III	Speech and speaker recognition
12	November 15	Images	I	Image basics
13	November 22		II	Object detection and face recognition
14	November 29		III	Image classification

* Wednesday, October 18 does not feature any class due to the Autumn break in week 42.

Time: Wednesdays 09 – 12

Location: 1330-038

Structure:

- 09.15-10.00: Lecture
- 10.00-10.15: Break
- 10.15-10.45: Coding Tutorial
- 10.45-11.00: Break
- 11.00-12.00: Lab

Material: https://github.com/mraskj/css_fall2023

09.15-10.00: Lecture

- Cover the most relevant parts from the readings – a mixture of methods and applications – sometimes theory-heavy, sometimes not.
- *Try* to balance theory, measurement validation, and implementation
- Usually 45 minutes, sometimes an hour.
- Slides are available as:
 - `class01-slides.pdf`

10.15-10.45: Coding Tutorial

- Class demonstration of how to implement code snippets and methods
- Tutorials are available as:
 - `class01-tutorial.ipynb`

11.00-12.00: Lab

- Hands-on exercises *in* class
- I'll be available throughout the lab session to help you if needed – help each other!
- The exercises and solutions are available as:
 - `class01-exercise.ipynb`
 - `class01-solution.ipynb`

- 7-day take-home exam
- December 12 at 09.00 to December 19 at ??
- Cover topics from all five 'main topics' with various weights.
- Likely to feature one or two papers, which you should engage with

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Main takeaway(s):

1. Interpret measures with care (Grimmer and Stewart, 2013)

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Main takeaway(s):

1. Interpret measures with care (Grimmer and Stewart, 2013)
2. Integrate measures with theory (Theocharis and Jungherr, 2021)

Defining the Scope of Computational Methods

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Question: What is “computational methods and analysis”?

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Question: What is “computational methods and analysis”?

- how does it relate to the *techniques* and *tools* you learned in Methods 1 and 2? (e.g. regression analysis)
- how does data such as *text*, *audio*, and *images* differ from the data used in Methods 1 and 2?

→ Talk to your neighbor for 4-5 minutes

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 - question to you: should we also include the type of data?

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6. Garbage in, garbage out

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- *Not* learning how to convert (audio) signals from the time to frequency domain
- *Instead*: learning how to **derive and measure political** relevant behavior and concepts
- *Instead*: learning how to **process unstructured data** into a format that we can analyze using simple tools (e.g. regression models)

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 - Requires to write commands within the language's accepted rules

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- *Libraries*: Base Python + thousands of wonderful libraries (e.g. Pandas)
- *Versions*: Multiple Python versions – some libraries require specific versions
- *Environments*: A way to keep dependencies required by different projects separate by creating isolated Python environments for them

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- PyCharm (popular to write packages)
- VS Code
- Text editors (Atom, VIM)
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→ We will work with notebooks either locally or using Colab

Table 1: Comparison of Python and R

Task	Python	R
General Purpose Programming	Great	OK
Causal Inference	Poor	Great
Web Scraping	Great	OK
Data Visualisation	OK	Great
Machine Learning	Great	Great
Natural Language Processing	Great	Great
Computer Vision	Great	Poor
Audio	Great	Poor
Industry Demand	Great	OK

Coding Tutorial

Virtual Environments

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Open the terminal and create an environment:

```
conda create --name NAME-OF-ENVIRONMENT python=PYTHON-VERSION
```

```
conda create --name css-fall2023-topic1 python=3.10
```

Commands I

See a list of conda environments you have:

```
conda env list
```

To activate an environment:

```
conda activate css-fall2023-topic1
```

Verify version of Python is correct:

```
python --version
```

See installed packages:

```
pip list
```

Commands II

Deactivate an environment:

```
conda deactivate
```

Remove an environment

```
conda env remove --name css-fall2023-topic1
```


Commands III

Install the latest version of the package

Pip: `pip install PACKAGE`

Conda: `conda install -c conda-forge PACKAGE`

Install a specific version of a package

Pip: `pip install PACKAGE==1.4.3`

Conda: `conda install PACKAGE=1.4.3`

Lab

Jupyter Notebook

Google Colab

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- [1] Y. Theocharis and A. Jungherr, “Computational social science and the study of political communication,” *Political Communication*, vol. 38, no. 1-2, pp. 1–22, 2021.
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- [3] J. Y. Kim and Y. M. M. Ng, “Teaching computational social science for all,” *PS: Political Science & Politics*, vol. 55, no. 3, pp. 605–609, 2022.