

# Introduction to Text Mining

## Organizational

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<https://cs.upb.de/css>

# Organizational

## Meta

- **Course number.** L.079.05534
- **Modules.** Human machine interaction, Computer science 2
- **Instructors.** Henning Wachsmuth (lectures), Milad Alshomary (tutorials)
- **Languages.** English, Python

## Tasks

- **Six assignments.** Bi-weekly; ~50% programming, ~50% written.  
First one published on October 18; to be submitted on October 28, 23:59 (UTC+1).
- **Exam.** Oral. First round tentatively in February.  
50%+ of all assignment points needed to take the exam. General registration on PAUL.

## Latest information

- **Web page.** <http://cs.upb.de/css/teaching/courses/text-mining-w18>
- **PAUL.** <http://paul.upb.de>

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## Lectures and Tutorials

### Dates and locations

- **Lectures.** Thursday 11–14, as of October 11, in O2  
No lecture on November 1 (holiday)
- **Tutorials.** Monday 11–13, as of October 15, in H3  
First tutorial introduces Python and clarifies the assignment concept.

### Three lecture time options

1. **Early bird.** Start 11:00, end 13:30, 15 minutes break  
Mensa-friendly, not FÜ-friendly, attention-friendly
2. **Starvin' Marvin.** Start 11:15, end 13:45, 15 minutes break  
Not Mensa-friendly, FÜ-friendly, attention-friendly
3. **Workhorse.** Start at 11:15, end at 13:30, no break  
Mensa-friendly, FÜ-friendly, not attention-friendly

### Chosen option (based on discussion in the lecture)

- **Early bird.** As of October 18, the lecture will start at 11:00 s.t.

# Goals of the course

## Overall

- Learn major skills needed to approach typical text mining tasks.

## Contents

- Several linguistic and statistical text analysis techniques.
- Several text mining tasks and applications.
- Needed basics of linguistics, empirical research, and machine learning.

## Competences

- Understanding of theory and practice of text mining.
- Design and implementation of text mining approaches for given tasks.
- Scientific experiments and evaluations on large amounts of data.

# Basics this Course Builds upon

## Required basics

- **Models and algorithms.** Concepts and methods from first semesters.
- **Languages.** Understanding of natural and formal languages.
- **Math.** Basic probability theory and linear algebra.
- **Development.** Some experience with software development in any programming language.

## Covered basics

- **Linguistics.** Fundamental language concepts and phenomena.
- **Statistics.** Concepts and methods related to empirical research.
- **Machine learning.** Fundamental concepts and learning methods.
- **Development.** Programming in Python.

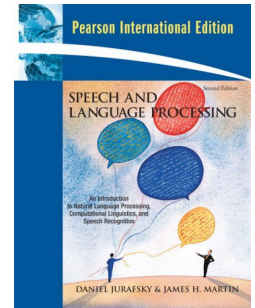
Python mostly covered in the tutorials only.

# Textbooks (Not Mandatory)

Daniel Jurafsky and James H. Martin (2009).

## Speech and Language Processing.

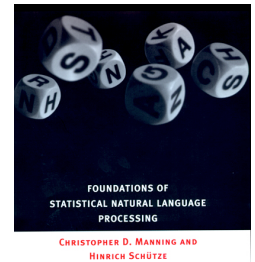
- Oriented towards computational linguistics
- Comprehensive
- Draft of 3rd ed.: <http://web.stanford.edu/~jurafsky/slp3>



Christopher D. Manning and Hinrich Schütze (1999).

## Foundations of Statistical Natural Language Processing.

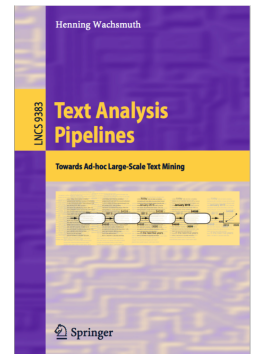
- More oriented towards computer science
- Comprehensive, a bit outdated



Henning Wachsmuth (2015).

## Text Analysis Pipelines.

- Rather oriented towards computer science
- Focused on advanced text mining techniques
- Thesis version: <http://www.arguana.com/publications/wachsmuth15c-lncs.pdf>



# Outline of the Course

- I. Overview
- II. Basics of Linguistics
- III. Text Mining using Rules
- IV. Basics of Empirical Research
- V. Text Mining using Grammars
- VI. Basics of Machine Learning
- VII. Text Mining using Clustering
- VIII. Text Mining using Classification and Regression
- IX. Practical Issues
- X. Text Mining using Sequence Labeling