Course Overview

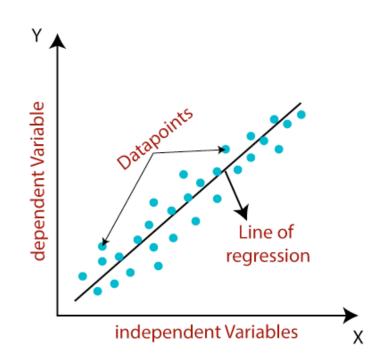
Ling 250/450: Data Science for Linguistics
C.M. Downey
Spring 2025



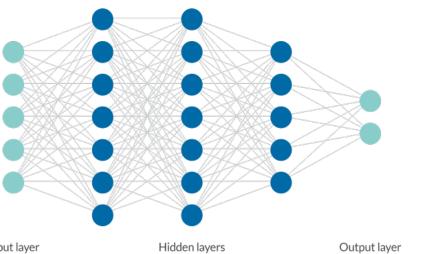
What is Data Science anyway?

What is Data Science?

- Data Science is (in my opinion) somewhat poorly defined
- Things people will often refer to as Data Science:
 - Collecting, manipulating, and cleaning data
 - Running statistical tests and models
 - Training and using machine learning models
 - Interpreting meaning or trends from data
 - Visualizing data in useful and novel ways









- All of these skills are used in many other fields! (And most pre-date "Data Science")
- Data Scientists might only specialize in one or a few of these skills

Is it actually science?

- Again in my opinion, "Data Science" might be a bit of a misnomer
- These techniques are often but not always applied to science
 - An important distinction is that science follows the Scientific Method, and mostly acquires knowledge through hypothesis testing
 - Most of the hallmarks of "Data Science" can be used without following the Scientific Method
- While the name "Data Science" is here to stay, it might (sometimes) be
 more accurate to talk about data {analysis | exploration | engineering}



"Data Science" and Science

- Data Science techniques can be an integral part of the Scientific Method
 - Prior to experimentation: explore, analyze, and visualize patterns in data and previous experimental results in order to form a new testable hypothesis
 - During experimentation: **collect**, **clean**, and **organize** data that allows you to properly **test your hypothesis**, ideally with **statistical methods**
 - After experimentation: analyze results to tell if your hypothesis is supported. Create visualizations to share your results. Organize your data for dissemination and experimental replication
- We will read later about the importance of separating exploratory studies and hypothesis testing (but we will engage in both in this course!)

Data Science and Linguistics

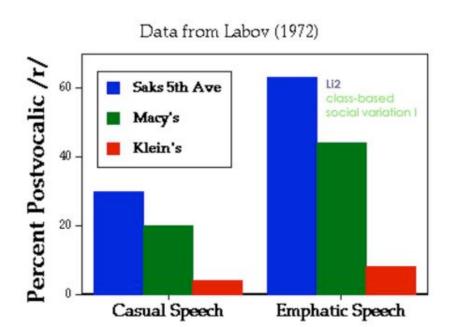
- Linguistics (mostly) acts as the science of language
 - Thus, DS can be used to explore and test hypotheses about language and how it works
- Some linguistic sub-fields have long embraced DS methods
 - e.g. Sociolinguistics, Psycholinguistics, and "Lab"/Experimental Phonetics
- Others have been historically more resistant (especially Chomsky)



• Syntax and Generative Grammar have tended towards methods more similar to math and

philosophy than (experimental) science

```
\langle \text{SENTENCE} \rangle \rightarrow \langle \text{NOUN-PHRASE} \rangle \langle \text{VERB-PHRASE} \rangle
\langle \text{NOUN-PHRASE} \rangle \rightarrow \langle \text{CMPLX-NOUN} \rangle | \langle \text{CMPLX-NOUN} \rangle \langle \text{PREP-PHRASE} \rangle
\langle \text{VERB-PHRASE} \rangle \rightarrow \langle \text{CMPLX-VERB} \rangle | \langle \text{CMPLX-VERB} \rangle \langle \text{PREP-PHRASE} \rangle
\langle \text{PREP-PHRASE} \rangle \rightarrow \langle \text{PREP} \rangle \langle \text{CMPLX-NOUN} \rangle
\langle \text{CMPLX-NOUN} \rangle \rightarrow \langle \text{ARTICLE} \rangle \langle \text{NOUN} \rangle
\langle \text{CMPLX-VERB} \rangle \rightarrow \langle \text{VERB} \rangle | \langle \text{VERB} \rangle \langle \text{NOUN-PHRASE} \rangle
\langle \text{ARTICLE} \rangle \rightarrow \text{a} | \text{the}
\langle \text{NOUN} \rangle \rightarrow \text{boy} | \text{girl} | \text{flower}
\langle \text{VERB} \rangle \rightarrow \text{touches} | \text{likes} | \text{sees}
\langle \text{PREP} \rangle \rightarrow \text{with}
```



Rationalism and Empiricism in Linguistics

- Chomsky's school of Linguistics uses rationalist techniques for insight on language, while experimental and data-driven methods are considered empiricist. (Very) roughly:
 - Rationalism: knowledge is obtained through introspection and logical processes like deduction and induction
 - Empiricism: knowledge is obtained through observation of external experience, especially the results of controlled experiments
- Note that these philosophies are not mutually exclusive, but researchers
 have historically tended to argue that one may be more reliable than the other

Why be an empiricist linguist?

- Rationalist approaches to Linguistics often rely on the introspective judgements from just one or a few people
 - e.g. "is this sentence grammatical?", "do these words mean the same thing?"
- Most linguists eventually have to confront the major insight of Sociolinguistics: language varies extensively
 - (Between people, communities, genres, conversation partners, etc.)
 - The same person might have different judgements at different times!
- This makes it questionable what the introspection of any one person can tell us about language in general
- Empirical techniques are inherently better suited for generalizing from variable data

Ex: Categorical grammar rules

- A key technique of Generative syntax is to use **intuitive judgements** to figure out which sentences are "grammatical" vs. "ungrammatical"
- Examples from Pollard and Sag (1994)
 - We consider Kim to be an acceptable candidate
 - We consider Kim an acceptable candidate
 - We consider Kim quite acceptable
 - We consider Kim among the most acceptable candidates
 - *We consider Kim as an acceptable candidate
 - *We consider Kim as quite acceptable
 - *We consider Kim as among the most acceptable candidates
 - *?We consider Kim as being among the most acceptable candidates



Ex: Categorical grammar rules

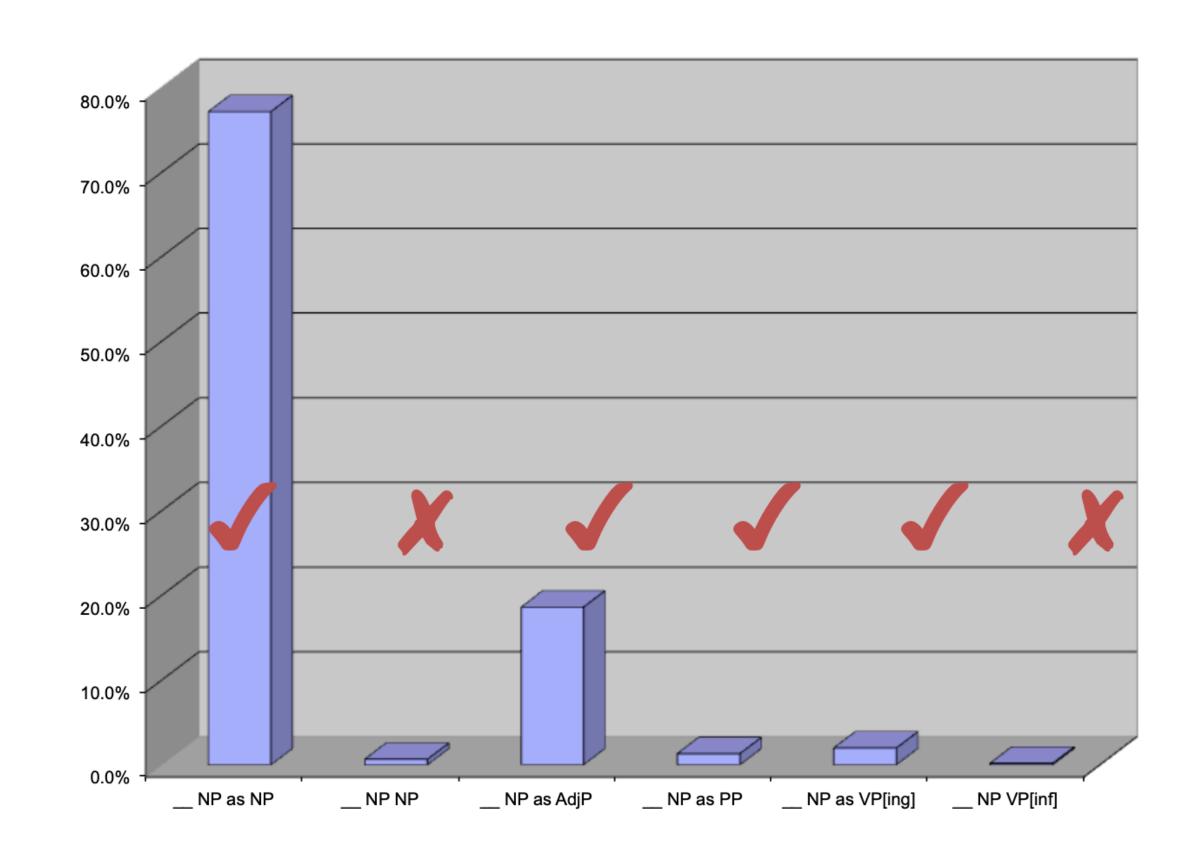
- Actual examples from the New York Times:
 - The boys consider her as family and she participates in everything we do
 - I don't consider it as something that gives me great concern
 - We consider that as part of the job
 - He considers them as having championship potential
 - Culturally, the Croats consider themselves as belonging to the West
- Investigating language from a data-driven approach reveals that it is much more flexible than the introspection-driven account

Problems with categorical theories

- They tend to claim too much
 - They place a hard categorical boundary of grammaticality, where really there is a fuzzy edge, determined by many conflicting constraints
- They tend to explain too little
 - They say very little about the soft constraints that explain how people choose to say things in given situations. These types of soft constraints have long been of interest to e.g. sociolinguists

Probability distribution for regard

- It might be more informative to construct a probability distribution over different variants of a sentence
- This distribution is calculated from a large corpus (dataset) of documented language "in the wild"
- We will discuss various standard corpora and corpus linguistics methods in this course



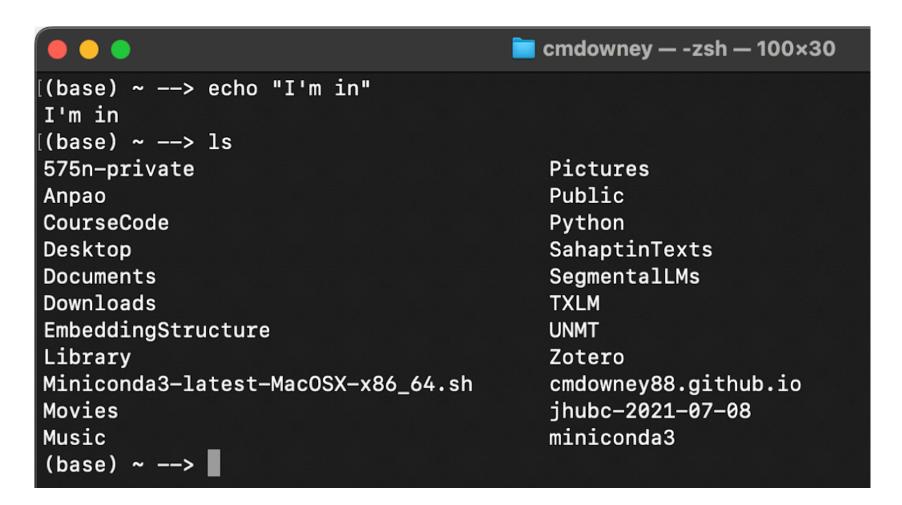
About this course

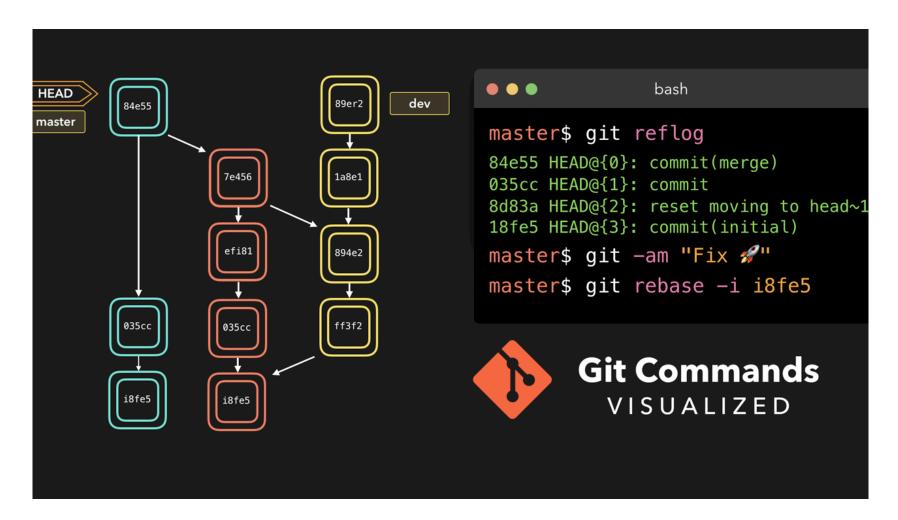
Course goals

- This course is meant to serve as a tools-based introduction to data-driven methods in Linguistics
- The tools we introduce will help you:
 - Collect, clean, organize, and understand linguistic data
 - Gain access to existing corpora and understand annotation conventions
 - Visualize patterns in language, and effectively communicate your findings
 - Test linguistic hypotheses with statistical robustness
- The course will also serve as a gentle introduction to the Python and R programming languages

Larger theme: Basic programming tools

- Covered in the first part of the course
- Using the command line
 - Will learn how to interface with computers in a new and more powerful way
 - Is sometimes the **only way** to interface with powerful servers for data science!
- Version control with Git and Github
 - The standard way to manage collaboration, edits, and differing versions of software





Larger theme: Data manipulation in Python

- Very basic introduction to Python
 - The dominant language for Science and Data Engineering
 - Easy language to get started
- Introduction to relevant Python
 libraries
 - Regular Expressions
 - Natural Language Tool Kit
 - Pandas

```
>>> nltk.corpus.sinica_treebank.tagged_words()
[('ä', 'Neu'), ('åæ', 'Nad'), ('åç', 'Nba'), ...]
>>> nltk.corpus.indian.tagged_words()
[('মহিষের', 'NN'), ('সন্তান', 'NN'), (':', 'SYM'), ...]
>>> nltk.corpus.mac_morpho.tagged_words()
[('Jersei', 'N'), ('atinge', 'V'), ('m\xe9dia', 'N'), ...]
>>> nltk.corpus.conll2002.tagged_words()
[('Sao', 'NC'), ('Paulo', 'VMI'), ('(', 'Fpa'), ...]
>>> nltk.corpus.cess_cat.tagged_words()
[('El', 'da0ms0'), ('Tribunal_Suprem', 'np00000o'), ...]
```

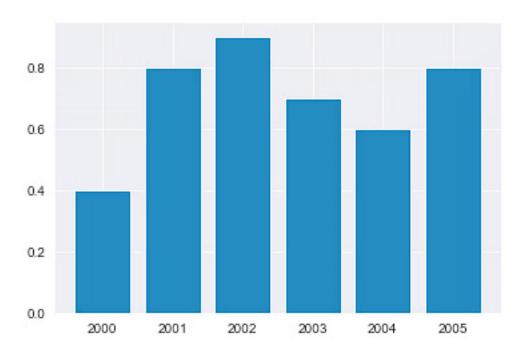
```
years = range(2000, 2006)
apples = [0.35, 0.6, 0.9, 0.8, 0.65, 0.8]
oranges = [0.4, 0.8, 0.9, 0.7, 0.6, 0.8]

plt.bar(years, oranges)

plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')

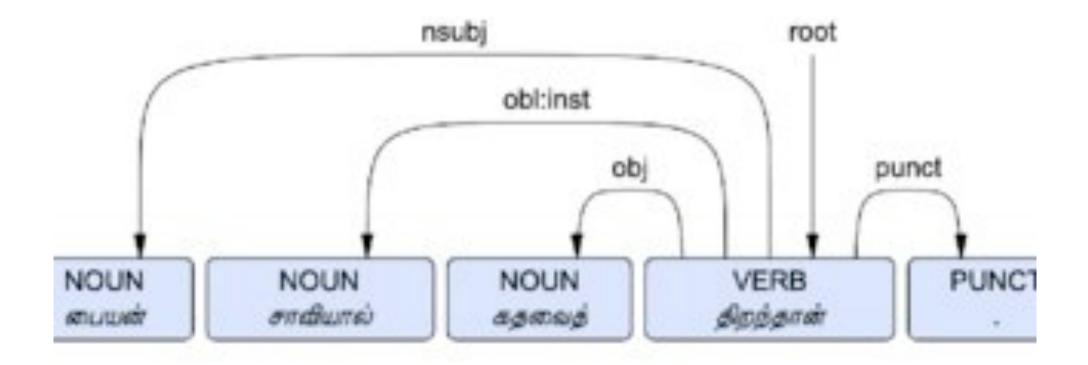
plt.title("Crop Yields in Kanto")
```

<BarContainer object of 6 artists>



Larger theme: Linguistic annotation

- Introduction to common annotation schemes for Linguistic data
- Discussion of why it is important to have machine-readable data
- Discussions of the challenges and reliability of annotation



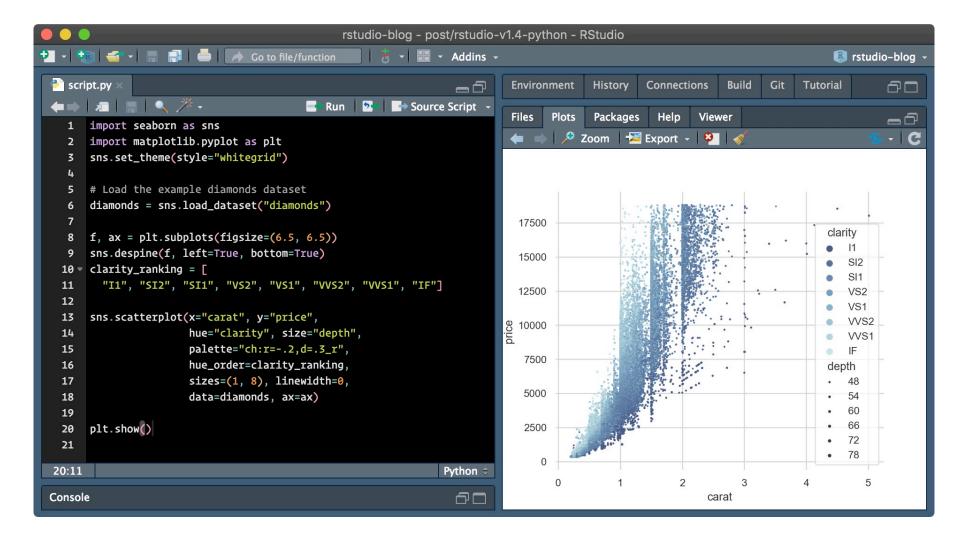
ure 03: A dependency graph for the annotation given in Tabl

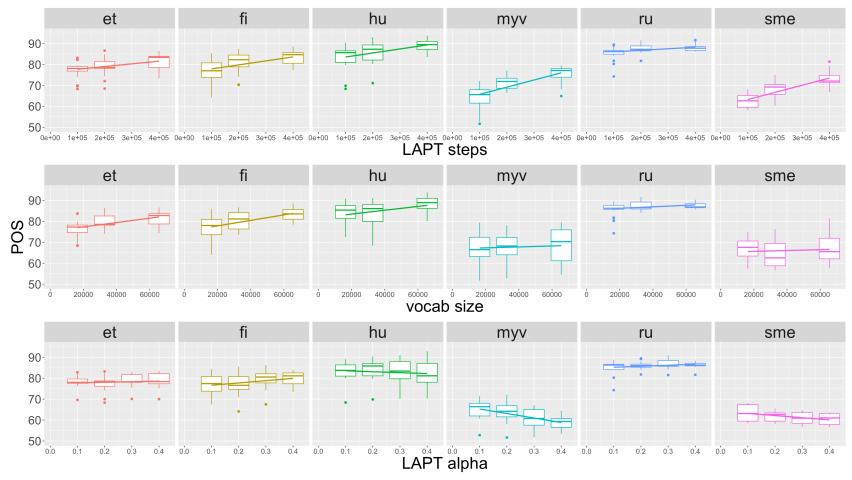
Larger theme: R for testing and visualization

- Introduction to R
 - Most commonly used language for statistical testing
 - Also has the best libraries for data visualization

ggplot

- Library for developing professional data visualizations
- Tricky at first, but gives the best results





What this course is not

- This is not a machine learning course
- This is not a Natural Language Processing (NLP) course
- This is not an algorithms/data-structures course
- This is not a software engineering course
- This course is not about Large Language Models
 - However, you can study the output of LLMs if you want

Policies and Organization

Basic Information

- Instructor
 - C.M. Downey
 - Assistant Professor in Linguistics & Data Science
- Meeting time
 - Monday/Wednesday 9:00-10:15am
 - This week only we meet Wednesday/Friday
- Prerequisites
 - Intro to Linguistics (LING 110)

Online Resources

- Blackboard
 - Announcements
 - Discussion boards
 - Homework submission
- Course website (<u>cmdowney88.github.io/teaching/ling250/spring2025</u>)
 - Detailed course schedule
 - Syllabus
 - Slides (posted after class)
 - Important links and documents

Attendance

- Because of the hands-on nature of this course, attendance is mandatory
- I will keep track of attendance, however:
 - You can be absent from up to four sessions without penalty
 - These absences can be for **any reason** (e.g. illness, travel, catching up on other courses), and you **do not have to approve it with me**
 - Absences beyond these four count against your attendance grade (5%)
- You may request additional excused absences for important exceptions (e.g. religious obligations, interviews, jury duty)

Participation

- Participation is also part of your final grade (5%)
- A large part of class time will be devoted to hands-on demonstrations of various software tools
 - You are expected to follow along with the demo on your own device
 - This means you will need to bring a **laptop** or **other appropriate device** to class (a tablet might work if it is running Windows; an iPad probably won't work)
- Before Friday: follow the setup guide posted on Blackboard and the course website to get your device ready

Course Work

- Homework (40%): between 6 and 8 assignments involving written questions, practice problems, and small coding projects
 - Students may work on these collaboratively, but each must submit their own work
- Midterm Project (20%): a larger coding project, in which everyone will work on the same data/problem
- Final Project (30%): a term research project which will be conducted with data of each student's choosing (within reason)

Deadlines and Late Work

- Unless specified otherwise, all assignments are due 11pm Eastern
- Work submitted after the deadline will incur a penalty
 - Up to 1 hour late: -5%
 - Up to 24 hours late: -10%
 - Up to 48 hours late: -20%
 - > 48 hours late: not graded (0 for the assignment)
- Please feel free to request extensions, but you must request it before the deadline
 - I will be more willing to grant extensions the longer before the deadline it is requested

Academic Honesty

- Homeworks can be completed collaboratively, but the on the Midterm
 Project, only minimal collaboration is allowed
- The use of Large Language Models / Chatbots for this class is allowed, however:
 - It is only allowed for code (not essays or short answers)
 - You are fully responsible for the success our failure of your code
 - We will talk more about responsible practices for using LLMs later

"Required" Textbooks

- You will sometimes be assigned required readings before class, but these will always be drawn from free online sources
 - Some are available online as PDFs, others through the UR Library
- Links to these sources can be found here and on the course website:
 - Natural Language Processing with Python
 - Data Science from Scratch
 - Speech and Language Processing

Questions/Discussion?