CS5740: Natural Language Processing

Introduction

Instructor: Yoav Artzi

TA: Max Grusky

- People:
 - Instructor: Yoav Artzi
 - Office hours: Monday 5pm, Baron
 - TA: Max Grusky:
 - Office hours: Thursday, 1pm, by Skype (coordinate)
- Webpage (everything is there):
 - http://www.cs.cornell.edu/courses/cs5740/2017sp/
- Discussion group on Piazza
- Chat on Slack
- Assignments on CMS
 - Repositories on Github Classroom

Grading:

- 40% assignments, 25% exam, and 30% class review quizzes, 5% participation
- Participation = class + Piazza + Slack
- Enrollment and prerequisites:
 - At least B in CS 5785 (Applied ML) or equivalent Cornell Course
 - Or: instructor permission
 - Audit? Talk to me after class

Quizzes:

- First five minutes of every class, no extensions
- Each quiz: 1.5% of the grade, up to 30%, only top 20 quizzes count
- It is not possible to re-take a missed quiz
- A missed quiz gets zero
- Just like an exam: no copying, chatting, and not taking the quiz remotely -> all Al violations

Quiz practice

- Phones and laptops
- http://socrative.com
- Use NetID to identify
- Today's room: NLP5

- Collaboration:
 - All assignments must be done in pairs
- Use of external code/tools specified in each assignment
 - If have doubt ask!
- Late submissions:
 - 10% off for every 12 hours, rounded up
 - E.g., 25 hours late → grade starts at 70
 - No late submission for final exam
- All assignments should be implemented in Python

- Books (recommended, not required):
 - D. Jurafsky & James H. Martin, Speech and Language Processing
 - C.D. Manning & H. Schuetze, Foundations of Statistical Natural Language Processing
- Other material on the course website

- Come on time
 - Late? Enter quietly and sit at the back
 - Quiz starts on time
- No laptops or phones in class
 - Except during the quiz

WHY ARE YOU HERE?

What is this class?

- Depth-first technical NLP course
- Learn the language of natural language processing
- What this class is not?
 - It is not a tutorial to NLTK, TensorFlow, etc.
 - Stack Overflow already does this well

Class Goals

- Learn about the issues and techniques of modern NLP
- Be able to read current research papers
- Build realistic NLP tools
- Understand the limitation of current techniques

Main Themes

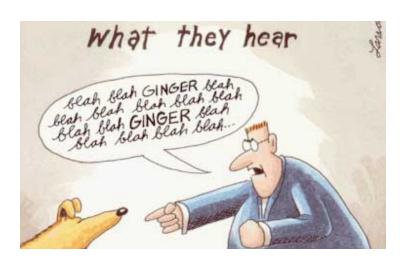
- Linguistic Issues
 - What are the range of language phenomena?
 - What are the knowledge sources that let us make decisions?
 - What representations are appropriate?
- Statistical Modeling Methods
 - Increasingly complex model structures
 - Learning and parameter estimation
 - Efficient inference: dynamic programming, search, sampling
- Engineering Methods
 - Issues of scale
 - Where the theory breaks down (and what to do about it)
- We'll focus on what makes the problems hard, and what works in practice ...

Main Models

- Generative Models
- Discriminative Models
 - Neural Networks
- Graphical Models

What is NLP?

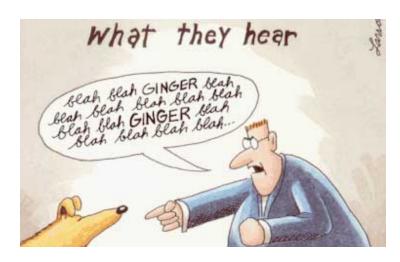




- Fundamental goal: deep understanding of broad language
 - Not just string processing or keyword matching!
- End systems that we want to build:
 - Simple:
 - Complex:

What is NLP?





- Fundamental goal: deep understanding of broad language
 - Not just string processing or keyword matching!
- End systems that we want to build:
 - Simple: spelling correction, text categorization...
 - Complex: speech recognition, machine translation, information extraction, dialog interfaces, question answering...
 - Unknown: human-level comprehension (is this just NLP?)

Today

- Prominent applications
 - Try to imagine approaches
 - What's behind current limitations?
- Some history
- Key problems



L'économie japonaise sort du rouge pour la première fois depuis Fukushima

Après avoir atteint un déficit record en 2014, le Japon dégage un excédent commercial pour la première fois depuis l'accident nucléaire de 2011.



Japan's economy turns red for the first time since Fukushima

After reaching a record deficit in 2014, Japan posted a trade surplus for the first time since the 2011 nuclear accident.

- Translate text from one language to another
- Recombines fragments of example translations
- Challenges:
 - What fragments? How to combine? [learning to translate]
 - How to make efficient? [fast translation search]

Le Monde.fr

La Bourse de Shanghai dégringolait de plus de 6 % mardi 25 août à l'ouverture, après s'être déjà effondrée de presque 8,5 % la veille, dans un marché affolé par l'affaiblissement persistant de l'économie chinoise et miné par des inquiétudes sur la conjoncture mondiale.

Dans les premiers échanges, l'indice composite chutait de 6,41 % soit 205,78 points à 3 004 13 points. La Bourse de Shenzhen plongeait quant à elle de

a ouvert en

The Shanghai Stock Exchange tumbled more than 6% Tuesday, August 25 at the opening, having already collapsed by almost 8.5% yesterday, in a panicked market the persistent weakening of the Chinese economy and undermined by concerns about the global economy.

In early trade, the composite index fell by 6.41% or 205.78 points to 3 004.13 points. The Shenzhen Stock Exchange dived for its 6.97% to 1 751.28 points. The Hong Kong Stock Exchange, meanwhile, opened down 0.67%.

纽约时报中文网 国际纵览

The New York Times Beta

A股跌势蔓延全球

周一美股开盘大跌1000点

NATHANIEL POPPER, NEIL GOUGH 09:54

周一,A股市场下跌8.5%,回吐今年全部涨幅。投资者担心中国经济下滑失控,股市"黑色星期一"波及美欧和亚洲市场,道指开盘数分钟内下跌过千点。

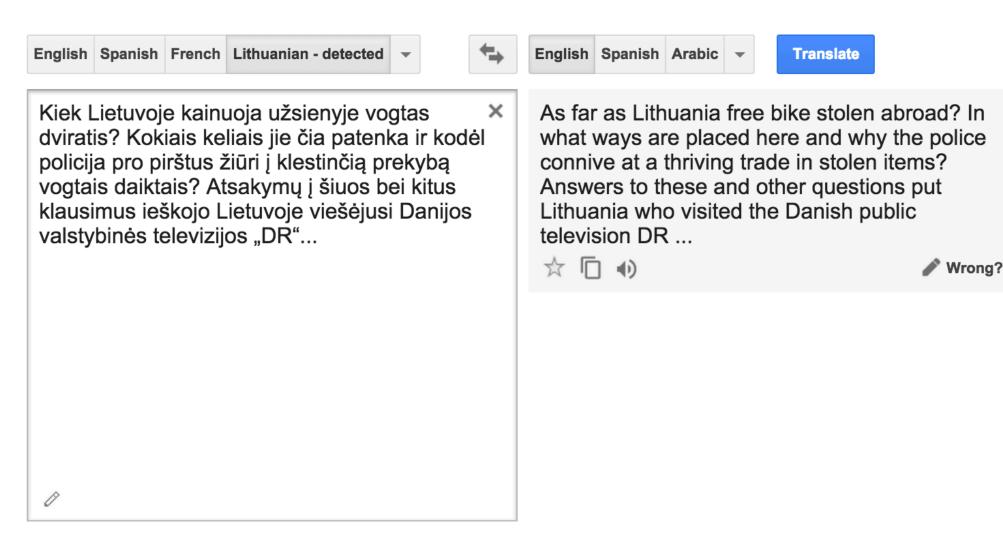
A spread of global stocks decline

US stocks opened Monday fell 1,000 points

NATHANIEL POPPER, NEIL GOUGH 09:54

Monday, A-share market fell 8.5 percent, taking all the gains this year. Investors worried about the economic downturn runaway Chinese stock market "Black Monday" spread to the US and European and Asian markets, the Dow opened down over a thousand points within minutes.





Summarization

- Condensing documents
 - Single or multiple docs
 - Extractive or abstractive
- Very contextdependent!

WASHINGTON (CNN) -- President Obama's inaugural address was cooler, more measured and reas: than that of other presidents making it, perhaps, the right speech for the times.



Some inaugural addresses are known for th soaring, inspirational language. Like John F. Kennedy's in 1961: "Ask not what your count do for you. Ask what you can do for your cour

Obama's address was less stirring, perhapit was also more candid and down-to-earth.

- Schneider: At a time of crisis, president must be reassuring
- · Country has chosen "hope over fear, unity of purpose over ... discord," Obama said
- Obama's speech was a cool speech, not a hot one, Schneider says

President Obama renewed his call for a massive plan to stimulate economic growth.

his first inaugural in 1933, "The only thing we to fear is fear itself." Or Bill Clinton, who took during the economic crisis of the early 1990. "There is nothing wrong with America that ca be fixed by what is right with America," Clinto declared at his first inaugural.

Obama, too, offered reassurance.

"We gather because we have chosen hope over fear, unity of purpose over conflict and discord," Obar said.

Obama's call to unity after decades of political division echoed Abraham Lincoln's first inaugural addr 1861. Even though he delivered it at the onset of a terrible civil war, Lincoln's speech was not a call to It was a call to look beyond the war, toward reconciliation based on what he called "the better angels" nature."

Some presidents used their inaugural address to set out a bold agenda.

Information Extraction

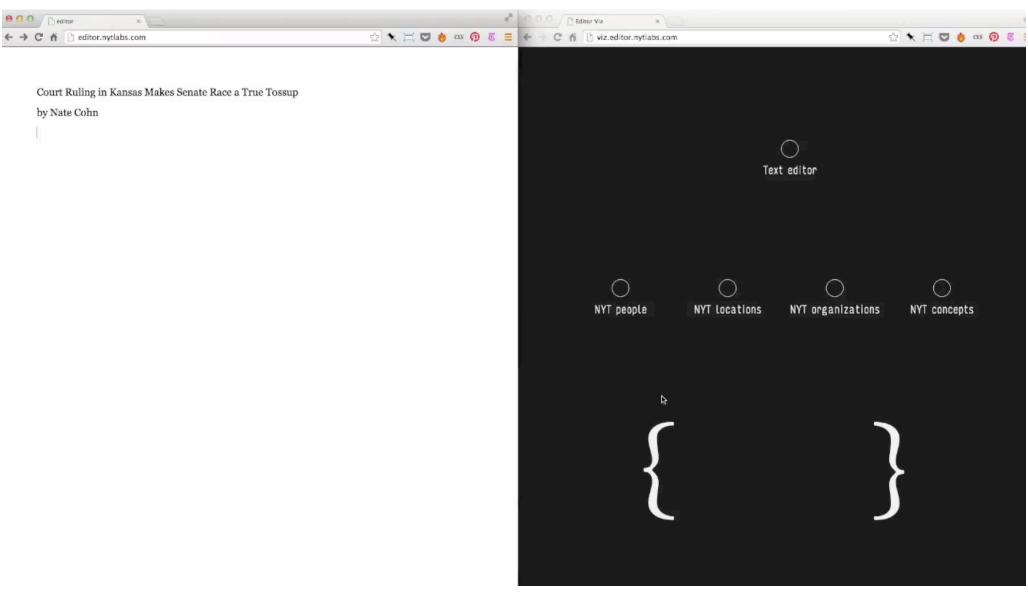
Unstructured text to database entries

New York Times Co. named Russell T. Lewis, 45, president and general manager of its flagship New York Times newspaper, responsible for all business-side activities. He was executive vice president and deputy general manager. He succeeds Lance R. Primis, who in September was named president and chief operating officer of the parent.

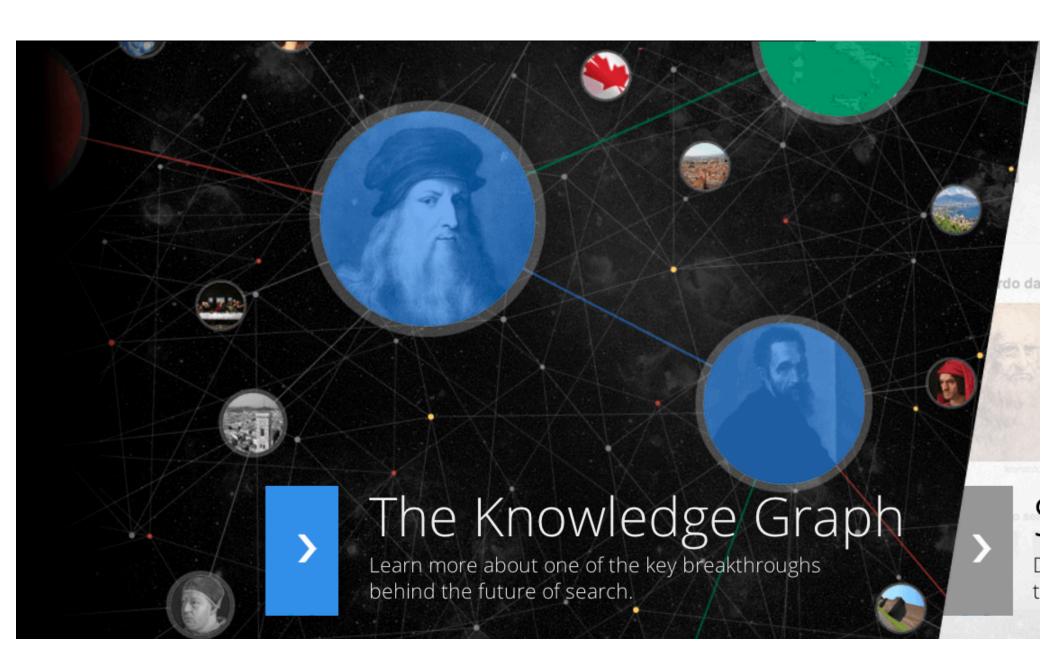
Person	Company	Post	State
Russell T. Lewis	New York Times newspaper	president and general manager	start
Russell T. Lewis	New York Times newspaper	executive vice president	end
Lance R. Primis	New York Times Co.	president and CEO	start

- SOTA: good performance on simple templates (e.g., personrole)
- Harder without defining template

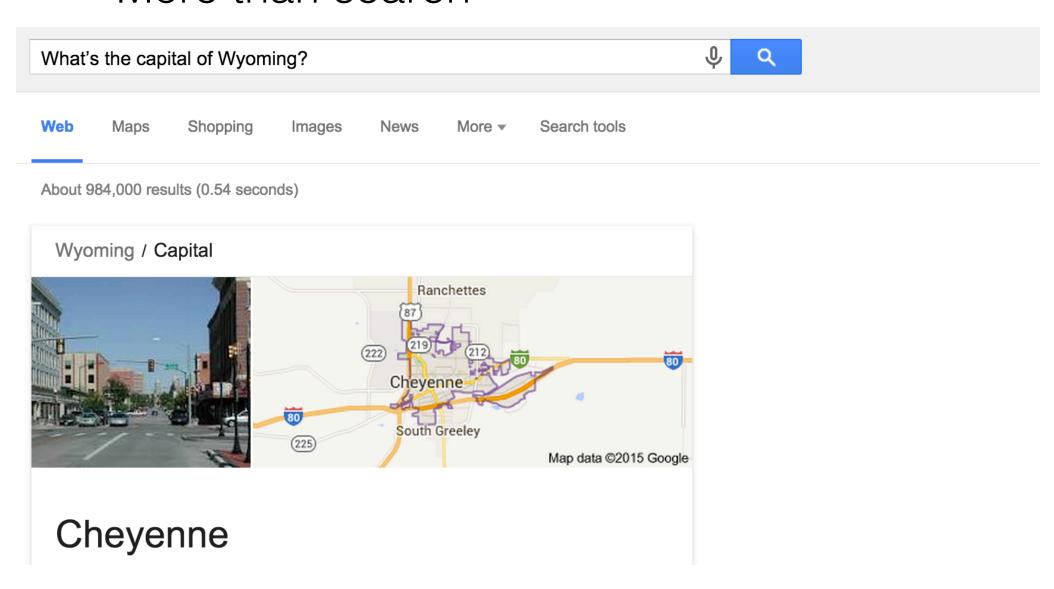
Tagging: Back to Text



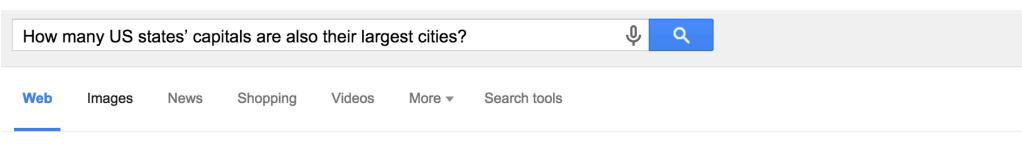
http://nytlabs.com/projects/editor.html



More than search



More than search



About 982,000,000 results (0.67 seconds)

State Capitals and Largest Cities - Infoplease

www.infoplease.com > United States > States ▼
State Capitals and Largest Cities. The following table lists the capital and largest city of every state in the United States. State, Capital, Largest city.

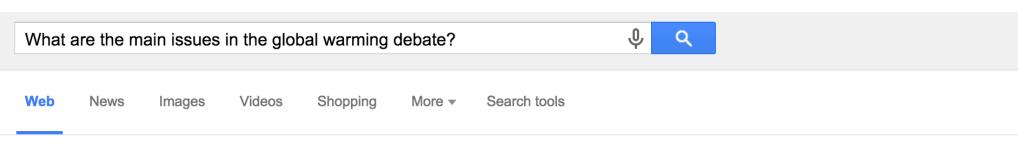
State Capitals and Largest Cities - Fact Monster

www.factmonster.com > United States > States ▼ Fact Monster ▼ State Capitals and Largest Cities. The following table lists the capital and largest city of every state in the United States. State, Capital, Largest city.

List of capitals in the United States - Wikipedia, the free ...

https://en.wikipedia.org/.../List_of_capitals_in_the_United_Sta... ▼ Wikipedia ▼ Austin is the largest state capital that is not also the state's largest city. The Confederate States of America had two capitals during its existence. The first ... In many cases, former capital cities of states are outside the current state borders. State capitals - Insular area capitals - Former national capitals

More than search



About 79,300,000 results (0.36 seconds)

Global warming controversy - Wikipedia, the free encyclopedia

https://en.wikipedia.org/wiki/Global_warming_controversy Wikipedia Jump to Mainstream scientific position, and challenges to it - [edit]. Main article: Scientific opinion on climate change. Summary of opinions from climate ...

Climate Change ProCon.org

climatechange.procon.org/ ▼ ProCon.org ▼

The pro side argues rising levels of atmospheric greenhouse gases are a direct result of human activities such as burning fossil fuels, and that these increases are causing significant and increasingly severe **climate** changes including global warming, loss of sea ice, sea level rise, stronger storms, and more droughts.

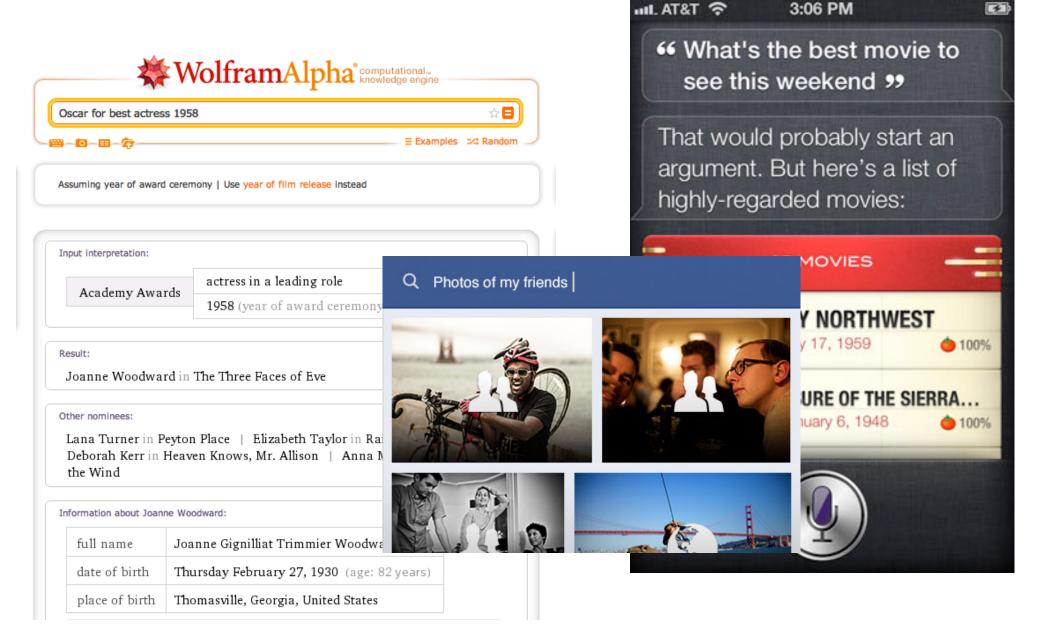
Is human activity a substantial - Footnotes & Sources - Carbon Dioxide (CO2)

Climate Change and Global Warming — Global Issues

www.globalissues.org/issue/178/climate-change-and-global-warming •

Some of the major conferences in recent years are also discussed. 32 articles on

"Climate Change and Global Warming" and 1 related issue:



Natural Language Instruction



 What makes this possible?

• Limitations?



Language Comprehension

"The rock was still wet. The animal was glistening, like it was still swimming," recalls Hou Xianguang. Hou discovered the unusual fossil while surveying rocks as a paleontology graduate student in 1984, near the Chinese town of Chengjiang. "My teachers always talked about the Burgess Shale animals. It looked like one of them. My hands began to shake." Hou had indeed found a Naraoia like those from Canada. However, Hou's animal was 15 million years older than its Canadian relatives.

Language Comprehension

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It can be inferred that Hou Xianguang's "hands began to shake" because he was

- (A) afraid that he might lose the fossil
- (B) worried about the implications of his finding
- (C) concerned that he might not get credit for his work
- (D) uncertain about the authenticity of the fossil
- (E) excited about the magnitude of his discovery

Language Comprehension

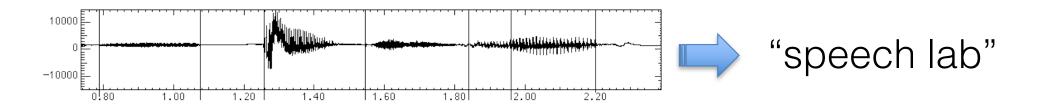
Bang, bang, his silver hammer came down upon her head



Speech Systems

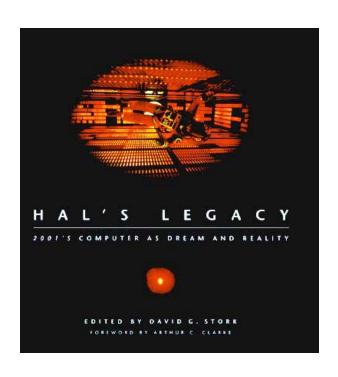


- Automatic Speech Recognition (ASR)
 - Audio in, text out
 - SOTA: 16% PER, Google claims 8% WER



- Text to Speech (TTS)
 - Text in, audio out
 - SOTA: mechanical and monotone

Language and Vision



"Imagine, for example, a computer that could look at an arbitrary scene anything from a sunset over a fishing village to Grand Central Station at rush hour and produce a verbal description. This is a problem of overwhelming difficulty, relying as it does on finding solutions to both vision and language and then integrating them. I suspect that scene analysis will be one of the last cognitive tasks to be performed well by computers"

-- David Stork (HAL's Legacy, 2001) on A. Rosenfeld's vision





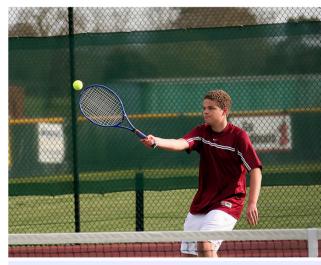
Image Captioning: The Good



a bunch of bananas sitting on top of a wooden table logprob: -8.52



a train traveling down tracks next to a lush green field logprob: -7.65



a man is playing tennis on a tennis court logprob: -6.77



a pizza with toppings on a white plate logprob: -7.40

Image Captioning: The Bad



a young boy is eating a piece of cake logprob: -7.75

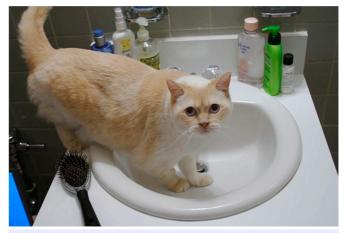


a man is holding a cell phone in his hand logprob: -8.90



a large jetliner flying through a blue sky logprob: -5.79

Image Captioning: The Sitting



a cat is sitting on a toilet seat



a bunch of luggage sitting on top of a hard wood floor logprob: -10.50



a laptop computer sitting on top of a wooden desk logprob: -6.38

a pizza sitting on top of a white plate logprob: -6.15



a group of people sitting around a table with a cake logprob: -8.83

a large airplane sitting on top of an airport runway logprob: -6.70

NLP History: Pre-statistics

- (1) Colorless green ideas sleep furiously.
- (2) Furiously sleep ideas green colorless

NLP History: Pre-statistics

- (1) Colorless green ideas sleep furiously.
- (2) Furiously sleep ideas green colorless

It is fair to assume that neither sentence (1) nor (2) (nor indeed any part of these sentences) had ever occurred in an English discourse. Hence, in any statistical model for grammaticalness, these sentences will be ruled out on identical grounds as equally "remote" from English. Yet (1), though nonsensical, is grammatical, while (2) is not." (Chomsky 1957)

NLP History: Pre-statistics

- 70s and 80s: more linguistic focus
 - Emphasis on deeper models, syntax and semantics
 - Toy domains / manually engineered systems
 - Weak empirical evaluation

NLP History: ML and Empiricism

"Whenever I fire a linguist our system performance improves." –Jelinek, 1988

- 1990s: Empirical Revolution
 - Corpus-based methods produce the first widely used tools
 - Deep linguistic analysis often traded for robust approximations
 - Empirical evaluation is essential

NLP History: ML and Empiricism

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"Of course, we must not go overboard and mistakenly conclude that the successes of statistical NLP render linguistics irrelevant (rash statements to this effect have been made in the past, e.g., the notorious remark, "Every time I fire a linguist, my performance goes up"). The information and insight that linguists, psychologists, and others have gathered about language is invaluable in creating high-performance broad-domain language understanding systems; for instance, in the speech recognition setting described above, a better understanding of language structure can lead to better language models."

- Lillian Lee (2001) http://www.cs.cornell.edu/home/llee/papers/cstb/index.html

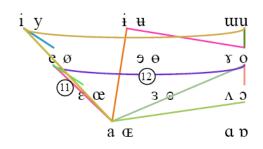
NLP History: ML and Empiricism

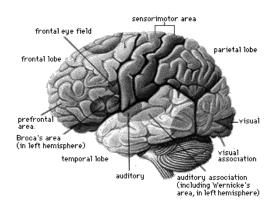
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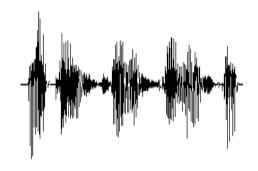
- 1990s: Empirical Revolution
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 - Empirical evaluation is essential
- 2000s: Richer linguistic representations used in statistical approaches, scale to more data!
- 2010s: you decide!

Related Fields

- Computational Linguistics
 - Using computational methods to learn more about how language works
 - We end up doing this and using it
- Cognitive Science
 - Figuring out how the human brain works
 - Includes the bits that do language
 - Humans: the only working NLP prototype!
- Speech?
 - Mapping audio signals to text
 - Traditionally separate from NLP, converging?
 - Two components: acoustic models and language models
 - Language models in the domain of stat NLP







Key Problems

We can understand programming languages. Why is NLP not solved?

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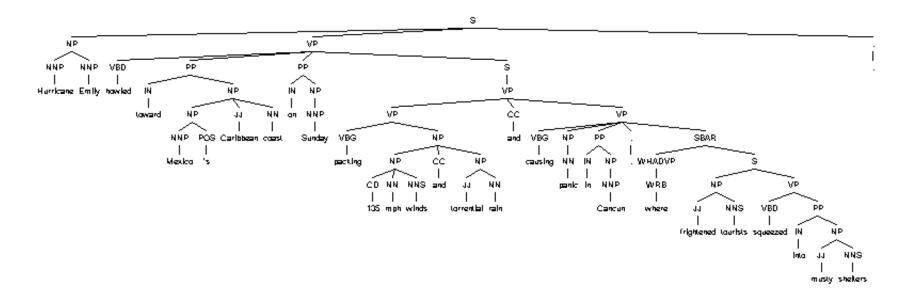
- Ambiguity
- Scale
- Sparsity

Key Problem: Ambiguity

- Some headlines:
 - Enraged Cow Injures Farmer with Ax
 - Ban on Nude Dancing on Governor's Desk
 - Teacher Strikes Idle Kids
 - Hospitals Are Sued by 7 Foot Doctors
 - Iraqi Head Seeks Arms
 - Stolen Painting Found by Tree
 - Kids Make Nutritious Snacks
 - Local HS Dropouts Cut in Half

Syntactic Ambiguity

Hurricane Emily howled toward Mexico 's Caribbean coast on Sunday packing 135 mph winds and torrential rain and causing panic in Cancun, where frightened tourists squeezed into musty shelters.



 SOTA: ~90% accurate for many languages when given many training examples, some progress in analyzing languages given few or no examples

Semantic Ambiguity

At last, a computer that understands you like your mother.

Semantic Ambiguity

At last, a computer that understands you like your mother.

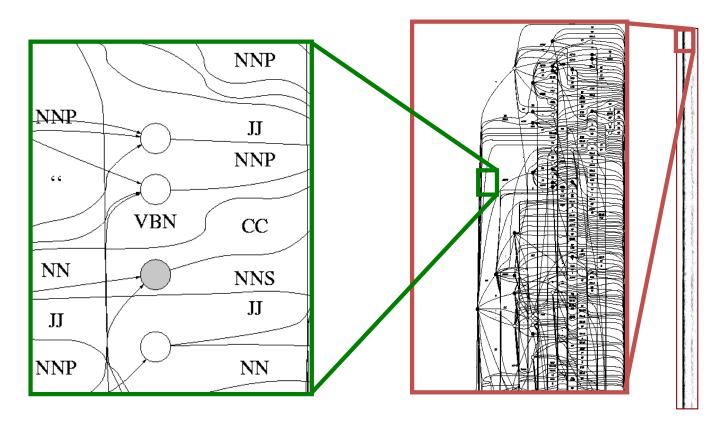
- Direct Meanings:
 - It understands you like your mother (does) [presumably well]
 - It understands (that) you like your mother
 - It understands you like (it understands) your mother
- But there are other possibilities, e.g. mother could mean:
 - a woman who has given birth to a child
 - a stringy slimy substance consisting of yeast cells and bacteria; is added to cider or wine to produce vinegar
- Context matters, e.g. what if previous sentence was:
 - Wow, Amazon predicted that you would need to order a big batch of new vinegar brewing ingredients.

Key Problem: Scale

- People did know that language was ambiguous!
 - ...but they hoped that all interpretations would be "good" ones (or ruled out pragmatically)

Key Problem: Scale

- People did know that language was ambiguous!
 - ...but they hoped that all interpretations would be "good" ones (or ruled out pragmatically)
 - ...they didn't realize how bad it would be



Key Problem: Sparsity



A <u>corpus</u> is a collection of text

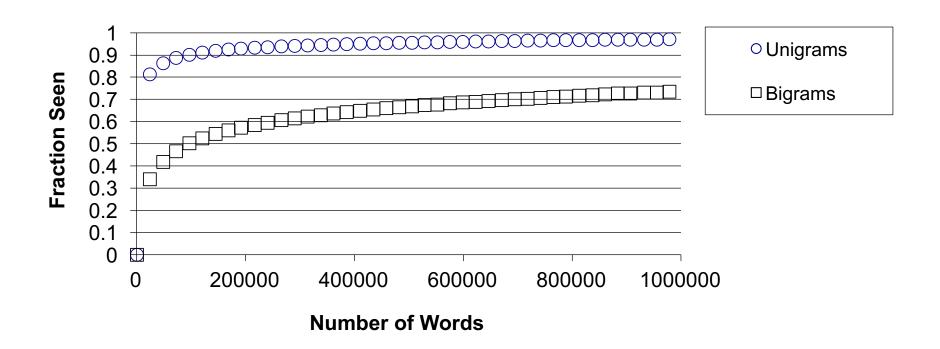
- Often annotated in some way
- Sometimes just lots of text
- Balanced vs. uniform corpora

Examples

- Newswire collections: 500M+ words
- Brown corpus: 1M words of tagged "balanced" text
- Penn Treebank: 1M words of parsed WSJ
- Canadian Hansards: 10M+ words of aligned French / English sentences
- The Web: billions of words of who knows what

Key Problem: Sparsity

- However: sparsity is always a problem
 - New unigram (word), bigram (word pair)



The NLP Community

- Conferences: ACL, NAACL, EMNLP, EACL, CoNNL, COLING, *SEM, LREC, CICLing, ...
- Journals: CL, **TACL**, ...
- Also in AI and ML conferences: AAAI, IJCAI, ICML, NIPS