

Natural Language Processing CS224N/Ling284



Christopher Manning
Spring 2010
Lecture 1



Course logistics in brief

- Instructor: Christopher Manning
- TAs: Mengqiu Wang, Val Spitzkovsky
- Time: MW 11:00–12:15.
 - Section: Fri 11:00–11:50 Skilling 191
- Programming language: Java 1.5+
- Other information: see the webpage.
 - <http://cs224n.stanford.edu/>
- Handouts:



vs.



?



This class

- Assumes you come with some skills...
 - Some basic linear algebra, probability, and statistics; decent programming skills
 - But not everyone has the same skills
 - Assumes some ability to learn missing knowledge
- Teaches key theory and methods for statistical NLP: MT, information extraction, parsing, semantics, etc.
 - Learn techniques which can be used in practical, robust systems that can (partly) understand human language
- But it's something like an "AI Systems" class:
 - A lot of it is hands-on, problem-based learning
 - Often practical issues are as important as theoretical niceties
 - We often combine a bunch of ideas



Natural language: the earliest UI

Dave Bowman: Open the pod bay doors, HAL.
HAL: I'm sorry Dave. I'm afraid I can't do that.



(cf. also false Maria in Metropolis – 1926)



Goals of the field of NLP

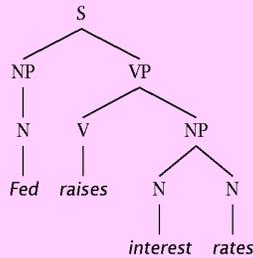
- Computers would be a lot more useful if they could handle our email, do our library research, chat to us ...
- But they are fazed by natural human languages.
 - Or at least their programmers are ... most people just avoid the problem and get into XML, or menus and drop boxes, or ...
- But someone has to work on the hard problems!
 - How can we tell computers about language?
 - Or help them learn it as kids do?
- In this course we seek to identify many of the open research problems in natural language



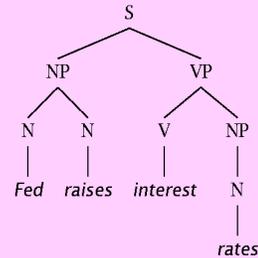
What/where is NLP?

- Goals can be very far reaching ...
 - True text understanding
 - Reasoning about texts
 - Real-time participation in spoken dialogs
- Or very down-to-earth ...
 - Finding the price of products on the web
 - Analyzing reading level or authorship statistically
 - Sentiment detection about products or stocks
 - Extracting facts or relations from documents
- These days, the latter predominate (as NLP becomes increasingly practical, it is increasingly engineering-oriented – also related to changes in approach in AI/NLP)

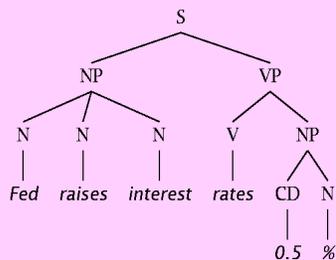
The bad effects of V/N ambiguities (1)



The bad effects of V/N ambiguities (2)



The bad effects of V/N ambiguities (3)



Why NLP is difficult: Newspaper headlines

- Minister Accused Of Having 8 Wives In Jail
- Juvenile Court to Try Shooting Defendant
- Teacher Strikes Idle Kids
- China to Orbit Human on Oct. 15
- Local High School Dropouts Cut in Half
- Red Tape Holds Up New Bridges
- Clinton Wins on Budget, **Minister Accused Of Having 8 Wives In Jail**
- Hospitals Are Sued by 7
- Police: Crack Found in M



ATLANTA (AP) -- A teacher served two years in prison for marrying more women. Bishop Anthony Owens, 44, is in a Gwinnett County jail after being released from prison.



Reference Resolution

U: Where is **A Bug's Life** playing in **Mountain View**?
 S: A Bug's Life is playing at the **Century 16 theater**.
 U: When is **it** playing **there**?
 S: It's playing at 2pm, 5pm, and 8pm.
 U: I'd like 1 **adult** and 2 **children** for **the first show**.
 How much would **that** cost?

- Knowledge sources:
 - Domain knowledge
 - Discourse knowledge
 - World knowledge



Why is natural language computing hard?

- Natural language is:
 - highly ambiguous at all levels
 - complex and subtle use of context to convey meaning
 - fuzzy, probabilistic
 - involves reasoning about the world
 - a key part of people interacting with other people (a social system):
 - persuading, insulting and amusing them
- But NLP can also be surprisingly easy sometimes:
 - rough text features can often do half the job



Making progress on this problem...

- The task is difficult! What tools do we need?
 - Knowledge about language
 - Knowledge about the world
 - A way to combine knowledge sources
- The answer that's been getting traction:
 - **probabilistic models** built from language data
 - P("maison" → "house") **high**
 - P("L'avocat général" → "the general avocado") **low**
- Some computer scientists think this is a new "A.I." idea
 - But really it's an old idea that was stolen from the electrical engineers....



Where do we head?

Look at subproblems, approaches, and applications at different levels

- Statistical machine translation
- Statistical NLP: classification and sequence models (part-of-speech tagging, named entity recognition, information extraction)
- Syntactic (probabilistic) parsing
- Building semantic representations from text. QA.
- (Unfortunately left out: natural language generation, phonology/morphology, speech dialogue systems, more on natural language understanding, There are other classes for some!)



Daily Question!

- What is the ambiguity in this (authentic!) newspaper headline?

**Ban on Nude Dancing
on Governor's Desk**

Machine Translation

美国关岛国际机场及其办公室均接获一名自称沙地阿拉伯富商拉登等发出的电子邮件，威胁将会向机场等公众地方发动生化袭击後，关岛经保持高度戒备。



The U.S. island of Guam is maintaining a high state of alert after the Guam airport and its offices both received an e-mail from someone calling himself the Saudi Arabian Osama bin Laden and threatening a biological/chemical attack against public places such as the airport .

The classic acid test for natural language processing.

Requires capabilities in both interpretation and generation.

About \$10 billion spent annually on human translation.

Scott Klemmer: I learned a surprising fact at our research group lunch today. Google Sketchup releases a version every 18 months, and the primary difficulty of releasing more often is not the difficulty of producing software, but the cost of internationalizing the user manuals!

Mainly slides from Kevin Knight (at ISI)

Translation (human and machine)

对外经济贸易合作部今天提供的数据表明，今年至十一月中国实际利用外资四百六十九点五九亿美元，其中包括外商直接投资四百零七亿美元。

Ref : According to the data provided today by the Ministry of Foreign Trade and Economic Cooperation, as of November this year, China has actually utilized 46.959 billion US dollars of foreign capital, including 40.007 billion US dollars of direct investment from foreign businessmen.

IBM4: the Ministry of Foreign Trade and Economic Cooperation, including foreign direct investment 40.007 billion US dollars today provide data include that year to November china actually using foreign 46.959 billion US dollars and

Yamada/Knight: today's available data of the Ministry of Foreign Trade and Economic Cooperation shows that china's actual utilization of November this year will include 40.007 billion US dollars for the foreign direct investment among 46.959 billion US dollars in foreign capital

Machine Translation History

- 1950s: Intensive research activity in MT
- 1960s: Direct word-for-word replacement
- 1966 (ALPAC): NRC Report on MT
 - Conclusion: MT no longer worthy of serious scientific investigation.
- 1966-1975: 'Recovery period'
- 1975-1985: Resurgence (Europe, Japan)
 - Domain specific rule-based systems
- 1985-1995: Gradual Resurgence (US)
- 1995-2010: Statistical MT surges ahead

<http://ourworld.compuserve.com/homepages/WJHutchins/MTS-93.htm>

What happened between ALPAC and Now?

- Need for MT and other NLP applications confirmed
- Change in expectations
- Computers have become faster, more powerful
- WWW
- Political state of the world
- Maturation of Linguistics
- Hugely increased availability of data
- Development of statistical and hybrid statistical/ symbolic approaches

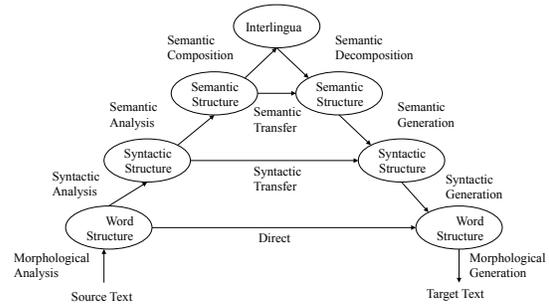


طالبت منظمة هيومن رايتس ووتش السلطات الإسرائيلية بأن ترفع فوراً القيود التي تحرم تلاميذ المدارس العامة في قطاع غزة من الكتب واحتياجات المدارس الأساسية مثل الأوراق والأقلام.

Called on organization Human Rights Watch the Israeli authorities to immediately lift restrictions that deny public school students in the Gaza Strip books

 Called Human Rights Watch, the Israeli authorities to immediately lift restrictions that prohibit public school students in the Gaza Strip of books and basic school needs such as paper and pens.

Three MT Approaches: Direct, Transfer, Interlingual (Vauquois triangle)



Statistical Solution

- Parallel Texts
 - Rosetta Stone

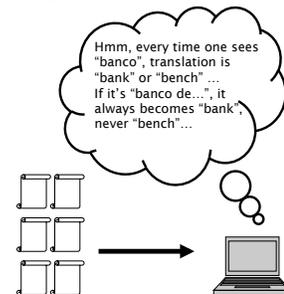
Hieroglyphs
Demotic
Greek



Statistical Solution

- Parallel Texts

- Instruction Manuals
- Hong Kong Legislation
- Macao Legislation
- Canadian Parliament Hansards
- United Nations Reports
- Official Journal of the European Communities





Warren Weaver

- “Also knowing nothing official about, but having guessed and inferred considerable about, the powerful new mechanized methods in cryptography—methods which I believe succeed even when one does not know what language has been coded—one naturally wonders if the problem of translation could conceivably be treated as a problem in cryptography. When I look at an article in Russian, I say: ‘This is really written in English, but it has been coded in some strange symbols. I will now proceed to decode.’ ”
– Warren Weaver (1955:18, quoting a letter he wrote in 1947)



“When I look at an article in Russian, I say: ‘This is really written in English, but it has been coded in some strange symbols. I will now proceed to decode.’ ” – Warren Weaver, March 1947



“... as to the problem of mechanical translation, I frankly am afraid that the [semantic] boundaries of words in different languages are too vague ... to make any quasi-mechanical translation scheme very hopeful.”
– Norbert Wiener, April 1947

Centauri/Arcturan [Knight, 1997]

Your assignment, translate this to Arcturan: farok errok hihok yorok klok kantok ok-yurp

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2a. ok-drubel ok-voon anok plok sprok .	8a. lalok brok anok plok nok .
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process of elimination

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cognate?

Centauri/Arcturan [Knight, 1997]

Your assignment, put these words in order: { **jjat, arrat, mat, bat, oloat, at-yurp** }

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zero fertility

It's Really Spanish/English

Clients do not sell pharmaceuticals in Europe => **Cientes no venden medicinas en Europa**

1a. Garcia and associates .	7a. the clients and the associates are enemies .
1b. Garcia y asociados .	7b. los clientes y los asociados son enemigos .
2a. Carlos Garcia has three associates .	8a. the company has three groups .
2b. Carlos Garcia tiene tres asociados .	8b. la empresa tiene tres grupos .
3a. his associates are not strong .	9a. its groups are in Europe .
3b. sus asociados no son fuertes .	9b. sus grupos estan en Europa .
4a. Garcia has a company also .	10a. the modern groups sell strong pharmaceuticals .
4b. Garcia tambien tiene una empresa .	10b. los grupos modernos venden medicinas fuertes .
5a. its clients are angry .	11a. the groups do not sell zenzanine .
5b. sus clientes estan enfadados .	11b. los grupos no venden zanzanina .
6a. the associates are also angry .	12a. the small groups are not modern .
6b. los asociados tambien estan enfadados .	12b. los grupos pequenos no son modernos .

Speech Recognition: Acoustic Waves

- Human speech generates a wave
 - like a loudspeaker moving
- A wave for the words "speech lab" looks like:

s p ee ch l a b

"l" to "a" transition:

Graphs from Simon Arnfield's web tutorial on speech, Sheffield:
<http://www.psyc.leeds.ac.uk/research/cogn/speech/tutorial/>

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Acoustic Sampling

- 10 ms frame (ms = millisecond = 1/1000 second)
- ~25 ms window around frame [wide band] to allow/smooth signal processing – it let's you see formants

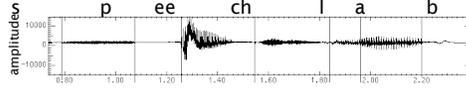
$a_1 \quad a_2 \quad a_3$

Result:
Acoustic Feature Vectors
 (after transformation, numbers in roughly R^{14})

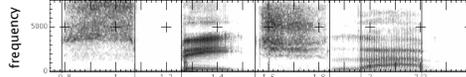
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Spectral Analysis

- Frequency gives pitch; amplitude gives volume
 - sampling at ~8 kHz phone, ~16 kHz mic (kHz=1000 cycles/sec)



- Fourier transform of wave displayed as a spectrogram
 - darkness indicates energy at each frequency
 - hundreds to thousands of frequency samples



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The Speech Recognition Problem

- The **Recognition Problem: Noisy channel model**
 - We started out with English words, they were encoded as an audio signal, and we now wish to decode.
 - Find most likely sequence \mathbf{w} of "words" given the sequence of acoustic observation vectors \mathbf{a}
 - Use Bayes' rule to create a **generative model** and then decode
 - $\text{ArgMax}_{\mathbf{w}} P(\mathbf{w}|\mathbf{a}) = \text{ArgMax}_{\mathbf{w}} P(\mathbf{a}|\mathbf{w}) P(\mathbf{w}) / P(\mathbf{a})$
 $= \text{ArgMax}_{\mathbf{w}} P(\mathbf{a}|\mathbf{w}) P(\mathbf{w})$

• **Acoustic Model:** $P(\mathbf{a}|\mathbf{w})$

• **Language Model:** $P(\mathbf{w})$

A probabilistic theory of a language

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Probabilistic Language Models

- Assign probability $P(\mathbf{w})$ to word sequence $\mathbf{w} = w_1, w_2, \dots, w_k$
- Can't directly compute probability of long sequence – one needs to decompose it
- Chain rule provides a **history-based** model:

$$P(w_1, w_2, \dots, w_k) = P(w_1) P(w_2|w_1) P(w_3|w_1, w_2) \dots P(w_k|w_1, \dots, w_{k-1})$$
- **Cluster** histories to reduce number of parameters
- E.g., just based on the last word (1st order Markov model):

$$P(w_1, w_2, \dots, w_k) = P(w_1|<s>) P(w_2|w_1) P(w_3|w_2) \dots P(w_k|w_{k-1})$$
- How do we estimate these probabilities?
 - We count word sequences in corpora
 - We "smooth" probabilities so as to allow unseen sequences

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