Natural Language Processing

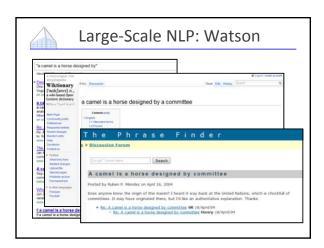


Question Answering

Dan Klein – UC Berkeley

The following slides are largely from Chris Manning, includeing many slides originally from Sanda Harabagiu, ISI, and Nicholas Kushmerick.

Watson



QA vs Search

People want to ask questions? Examples of search queries who invented surf music? how to make stink bombs where are the snowdens of yesteryear? which english translation of the bible is used in official catholic liturgies? how to do clayart how to copy psx how tall is the sears tower? how can i find someone in texas where can i find information on puritan religion? what are the 7 wonders of the world how can i eliminate stress What vacuum cleaner does Consumers Guide recommend

Around 10-15% of query logs



A Brief (Academic) History

- Question answering is not a new research area
- Question answering systems can be found in many areas of NLP research, including:
 - Natural language database systems
 - A lot of early NLP work on these
 - Spoken dialog systems
 - Currently very active and commercially relevant
- The focus on open-domain QA is (relatively) new
 - MURAX (Kupiec 1993): Encyclopedia answers
 - Hirschman: Reading comprehension tests
 - TREC QA competition: 1999-

TREC



Question Answering at TREC

 Question answering competition at TREC consists of answering a set of 500 fact-based questions, e.g.,

"When was Mozart born?".

- For the first three years systems were allowed to return 5 ranked answer snippets (50/250 bytes) to each question.
 - IR thinl
 - Mean Reciprocal Rank (MRR) scoring:
 - 1, 0.5, 0.33, 0.25, 0.2, 0 for 1, 2, 3, 4, 5, 6+ doc
 - Mainly Named Entity answers (person, place, date, ...)
- From 2002+ the systems are only allowed to return a single exact answer and a notion of confidence has been introduced



Sample TREC questions

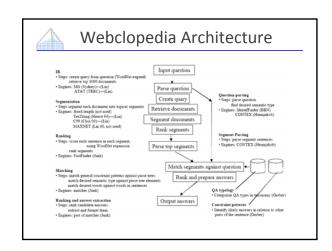
- 1. Who is the author of the book, "The Iron Lady: A Biography of Margaret Thatcher"?
- 2. What was the monetary value of the Nobel Peace Prize in 1989?
- 3. What does the Peugeot company manufacture?
- 4. How much did Mercury spend on advertising in 1993?
- 5. What is the name of the managing director of Apricot Computer?
- 6. Why did David Koresh ask the FBI for a word processor?
- 7. What debts did Qintex group leave?
- 8. What is the name of the rare neurological disease with symptoms such as: involuntary movements (tics), swearing, and incoherent vocalizations (grunts, shouts, etc.)?



Top Performing Systems

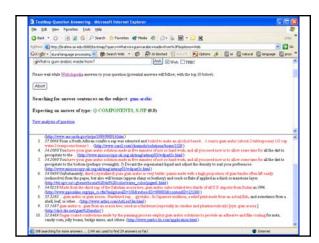
- Currently the best performing systems at TREC can answer approximately 70% of the questions
- Approaches and successes have varied a fair deal
 - Knowledge-rich approaches, using a vast array of NLP techniques stole the show in 2000, 2001, still do well
 - Notably Harabagiu, Moldovan et al. SMU/UTD/LCC
 - AskMSR system stressed how much could be achieved by very simple methods with enough text (and now various copycats)
 - Middle ground is to use large collection of surface matching patterns (ISI)
 - Emerging standard: analysis, soft-matching, abduction

Pattern Induction: ISI











- "When was <person> born"
- - Typical answers
 - "Mozart was born in 1756."
 - "Gandhi (1869-1948)..."
- Suggests phrases like
 - "<NAME> was born in <BIRTHDATE>"
 - "<NAME> (<BIRTHDATE>-"
- Regular expressions



Use Pattern Learning

- Example: Start with "Mozart 1756"
 - Results:
 - "The great composer Mozart (1756-1791) achieved fame at a
 - "Mozart (1756-1791) was a genius"
 - "The whole world would always be indebted to the great music of Mozart (1756-1791)"
 - Longest matching substring for all 3 sentences is "Mozart (1756-1791)"
 - Suffix tree would extract "Mozart (1756-1791)" as an output, with score of 3
- Reminiscent of IE pattern learning



Pattern Learning (cont.)

- Repeat with different examples of same question type
 - "Gandhi 1869", "Newton 1642", etc.
- Some patterns learned for BIRTHDATE
 - a. born in <ANSWER>, <NAME>
 - b. <NAME> was born on <ANSWER> ,
 - . c. <NAME> (<ANSWER> -
 - d. <NAME> (<ANSWER>)



Pattern Precision

BIRTHDATE table:

- 1.0 <NAME> (<ANSWER>)
- 0.85 <NAME> was born on <ANSWER>,
- 0.6 <NAME> was born in <ANSWER>
- 0.59 <NAME> was born <ANSWER>
- 0.53 <ANSWER> <NAME> was born
- 0.50 -<NAME> (<ANSWER>
- 0.36 <NAME> (<ANSWER> -

INVENTOR

- 1.0 <ANSWER> invents <NAME>
- 1.0 the <NAME> was invented by <ANSWER>
- 1.0 <ANSWER> invented the <NAME> in



Pattern Precision

WHY-FAMOUS

- 1.0 <ANSWER> <NAME> called
- 1.0 laureate <ANSWER> <NAME>
- 0.71 <NAME> is the <ANSWER> of

LOCATION

- 1.0 <ANSWER>'s <NAME>
- 1.0 regional : <ANSWER> : <NAME>
- 0.92 near <NAME> in <ANSWER>
- Depending on question type, get high MRR (0.6–0.9), with higher results from use of Web than TREC QA collection



Shortcomings & Extensions

- Need for POS &/or semantic types
 - "Where are the Rocky Mountains?"
 - "Denver's new airport, topped with white fiberglass cones in imitation of the Rocky Mountains in <u>the background</u>, continues to lie empty"
 - <NAME> in <ANSWER>

Long distance dependencies

- "Where is London?"
- "London, which has one of the busiest airports in the world, lies on the banks of the river Thames"
- would require pattern like: <QUESTION>, (<any_word>)*, lies on <ANSWER>
- But: abundance of Web data compensates

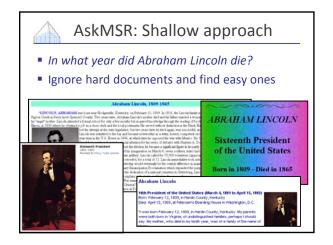
Aggregation: AskMSR

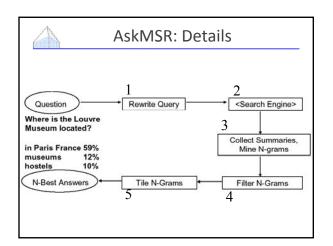


AskMSR

- Web Question Answering: Is More Always Better?
 - Dumais, Banko, Brill, Lin, Ng (Microsoft, MIT, Berkeley)
- Q: "Where is the Louvre located?"
- Want "Paris" or "France" or "75058 Paris Cedex 01" or a map
- Don't just want URLs



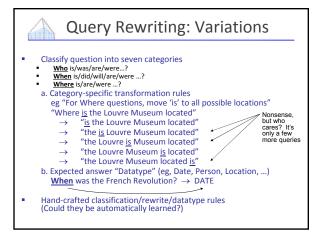


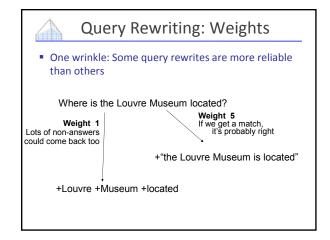




Step 1: Rewrite queries

- Intuition: The user's question is often syntactically quite close to sentences that contain the answer
 - Where is the Louvre Museum located?
 - <u>The Louvre Museum is located</u> in *Paris*
 - Who created the character of Scrooge?
 - Charles Dickens created the character of Scrooge.







Step 2: Query search engine

- Send all rewrites to a search engine
- Retrieve top N answers (100?)
- For speed, rely just on search engine's "snippets", not the full text of the actual document

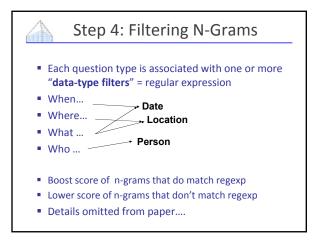


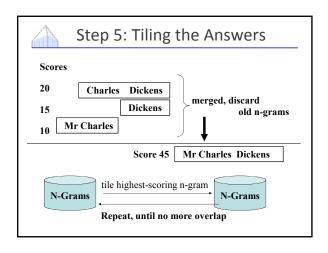
Step 3: Mining N-Grams

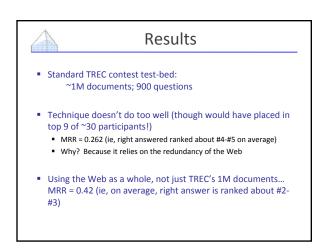
- Simple: Enumerate all N-grams (N=1,2,3 say) in all retrieved snippets
- Weight of an n-gram: occurrence count, each weighted by "reliability" (weight) of rewrite that fetched the document
- Example: "Who created the character of Scrooge?"

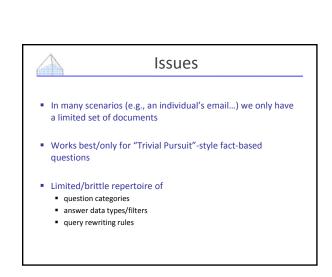
 - Dickens 117Christmas Carol 78
 - Charles Dickens 75 Disney - 72

 - Carl Banks 54A Christmas 41
 - Christmas Carol 45
 - Uncle 31

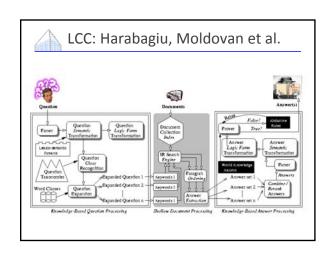








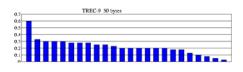
Abduction: LCC





Value from Sophisticated NLP Pasca and Harabagiu (2001)

- Good IR is needed: SMART paragraph retrieval
- Large taxonomy of question types and expected answer types is crucial
- Statistical parser used to parse questions and relevant text for answers, and to build KB
- Query expansion loops (morphological, lexical synonyms, and semantic relations) important
- Answer ranking by simple ML method





Abductive inference

- System attempts inference to justify an answer (often following lexical chains)
- Their inference is a kind of funny middle ground between logic and pattern matching
- But quite effective: 30% improvement
- Q: When was the internal combustion engine
- A: The first internal-combustion engine was built in 1867.
- invent -> create_mentally -> create -> build



Question Answering Example

- How hot does the inside of an active volcano get?
- "lava fragments belched out of the mountain were as hot as 300 degrees Fahrenheit"
 - volcano ISA mountain
 - lava ISPARTOF volcano lava IN volcano
 - fragments of lava HAVEPROPERTIESOF lava
- The needed semantic information is in WordNet definitions, and was successfully translated into a form that was used for rough 'proofs'

Watson

Slides from Ferrucci et al, Al Magazine, 2010



Jeopardy...

Category: General Science
Clue: When hit by electrons, a phosphor gives off
electromagnetic energy in this form.
Arower: Light (or Photons)

Category: Lincoln Blogs
Clue: Secretary Chase just submitted this to me for
the third time; guess what, pal. This time I'm

Category: Head North
Clue: They're the two states you could be reentering
if you're crossing Florida's northern border.
Arrswer: Georgia and Alabama

Category: Decorating

Clue: Though it sounds "harsh," it's just embroidery, often in a floral pattern, done with yarn on cotton cloth.

Auswer: crewel

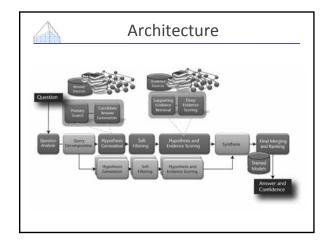
Category: "Rap" Sheet Clue This archaic term for a mischievous or annoy-ing child can also mean a rogue or scamp. Suk-tue 1: This archaic term for a mischievous or annoying child. Sub-tue 2: This term can also mean a rogue or scame.

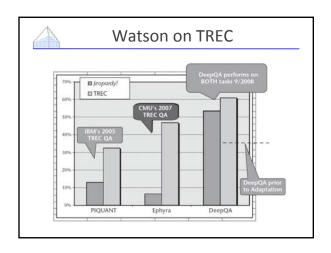
scamp.

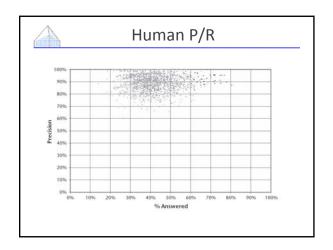
Answer: Rapscallion

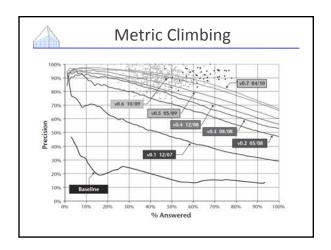
Category: Before and After Goes to the Movies Clue: Film of a typical day in the life of the Beatles, which includes running from bloodthirsty zombie

tles.
Answer I: (A Hard Day's Night)
Salvhae 2: Running from bloodthirsty zombie fans
in a Romero classic.
Answer 2: (Night) of the Living Doah
Answer: A Hard Day's Night of the Living Doah









Complex QA

