CS 6120/CS4120: Natural Language Processing

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Project presentation

- Each team will present for 8 minutes, with 2 minutes for QA.
- After all representations, we will vote for favorite project. Each team has two votes.
- The team that gets the most votes wins. Each team member will get 1% bonus towards final grade.

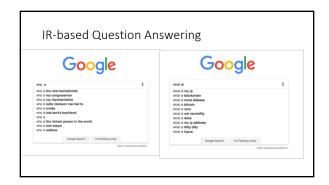
Project presentation

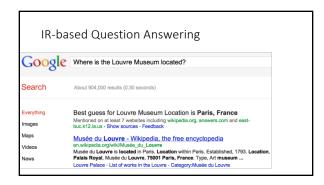
- The presentation order will be posted on piazza.
- Please upload your slides on blackboard after presentation.
- Feedback will be sent to the team through blackboard after the presentation.
- • Final reports are expected to resolve the issues raised in the feedback. Due on Dec 10^{th} , 11:59pm.

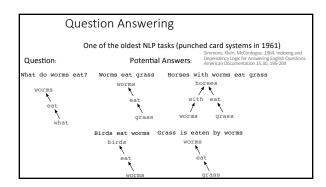
Presentation and final report

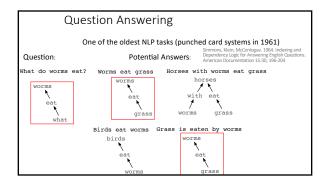
- Problem Description (10%) What is the task? System input and output Examples will be helpful
- Reference/Related work (20%) Put your work in context: what has been done before? You need to have reference! What's new in your work?
- Methodology: What you have done (30%) Preprocessing of the data What are your data? Features used? What are effective, and what are not? What methods do you experiment with? And why do you think they're reasonable and suitable for the task?
- Experiments (40%)
 Datasets size, train/test/development
 Evaluation metrics: what are used and are they
 proper to calibrate system performance?
 Baselines: what are they?
 Results, tables, figures, etc

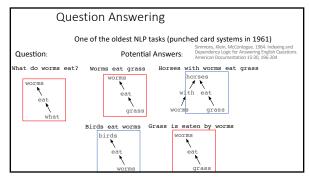
Question Answering











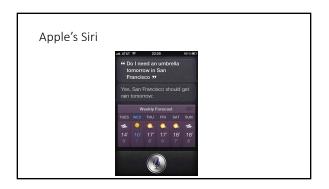
Question Answering: IBM's Watson

• Won Jeopardy on February 16, 2011!

WILLIAM WILKINSON'S

"AN ACCOUNT OF THE PRINCIPALITIES OF
WALLACHIA AND MOLDOVIA"
INSPIRED THIS AUTHOR'S
MOST FAMOUS NOVEL

Bram Stoker







Types of Questions in Modern Systems

• Factoid questions

• Who wrote "The Universal Declaration of Human Rights"?

• How many colories are there in two slices of apple pie?

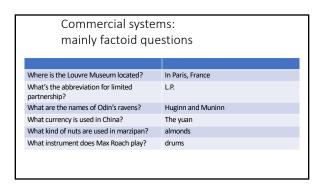
• What is the average age of the onset of autism?

• Where is Apple Computer based?

• Complex (narrative) questions:

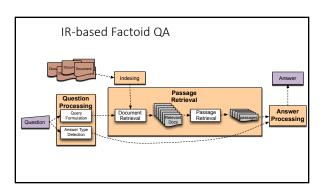
• In children with an acute febrile illness, what is the efficacy of acetaminophen in reducing fever?

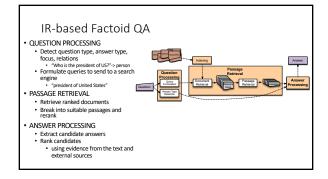
• What do scholars think about Jefferson's position on dealing with pirates?



Paradigms for QA

- •Information Retrieval (IR)-based approaches
 - IBM Watson (some parts); Google
- Knowledge-based and Hybrid approaches
 - IBM Watson; Apple Siri; Wolfram Alpha
- Data-driven, neural network-based approaches





Knowledge-based approaches (Siri)

- Build a semantic representation of the query
- · Times, dates, locations, entities, numeric quantities
- Map from this semantics to query structured data or resources
 - Geospatial databases
 - Ontologies (Wikipedia infoboxes, dbPedia, WordNet, Yago)
 - Restaurant review sources and reservation services
 - Scientific databases

Hybrid approaches (IBM Watson)

- · Build a shallow semantic representation of the query
- Generate answer candidates using IR methods
 - Augmented with ontologies and semi-structured data
- Score each candidate using richer knowledge sources
 - · Geospatial databases
 - Temporal reasoning
 - Taxonomical classification

IR-based Factoid QA QUESTION PROCESSING Detect question type, answer type, focus, relations "Who is the president of US?"-> person Formulate queries to send to a search engine • "president of United States" PASSAGE RETRIEVAL Retrieve ranked documents Break into suitable passages and rerank ANSWER PROCESSING Extract candidate answers Rank candidates using evidence from the text and external sources

Question Processing Things to extract from the question

- Answer Type Detection
 - Decide the named entity type (person, place) of the answer
- Query Formulation
 - $\bullet\,$ Choose $\mbox{\bf query keywords}$ for the IR system
- Question Type classification
 - Is this a definition question, a math question, a list question?
- Focus Detection
 - · Find the guestion words that are replaced by the answer
- Relation Extraction
 - Find relations between entities in the question

Question Processing

Jeopardy!: They're the two states you could be reentering if you're crossing Florida's northern border
You should answer: what are the states of Georgia and Alabama?

- Answer Type: US state
- Query Formulation: two states, border, Florida, north
- Focus: the two states
- Relations: borders(Florida, ?x, north)

Answer Type Detection: Named Entities

• Who founded Virgin Airlines?

Answer Type Detection: Named Entities

- Who founded Virgin Airlines?
 - PERSON
- What Canadian city has the largest population?

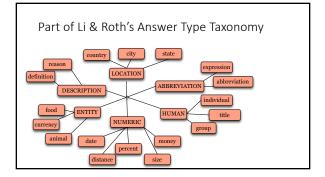
Answer Type Detection: Named Entities

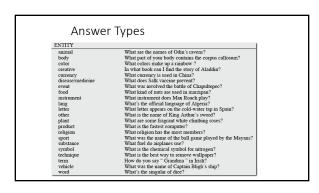
- Who founded Virgin Airlines?
 - PERSON
- What Canadian city has the largest population?
 - CITY

Answer Type Taxonomy

Xin Li, Dan Roth. 2002. Learning Question Classifiers. COLING'02

- 6 coarse classes
 - ABBREVIATION, ENTITY, DESCRIPTION, HUMAN, LOCATION, NUMERIC
- 50 finer classes
 - LOCATION: city, country, mountain...
 - HUMAN: group, individual, title, description
 - ENTITY: animal, body, color, currency...





More Answer Types Who was Confucius? What are the major companies that are part of Dow Jones? Who was the first Russian astronaut to do a spacewalk? What was Queen Victoria's title regarding India? What's the oldest capital city in the Americas Want states to the function to Market School and School and Market School and School and

Answer types in Jeopardy

Ferrucci et al. 2010. Building Watson: An Overview of the DeepQA Project. AI Magazine. Fall 2010. 59-79.

- 2500 answer types in 20,000 Jeopardy question sample
- The most frequent 200 answer types cover ~ 50% of data
- The 40 most frequent Jeopardy answer types

country, city, man, film, state, author, group, here, company, president, capital, star, novel, character, woman, river, island, king, song, part, series, sport, singer, actor, play, team, show, actress, animal, presidential, composer, musical, nation, book, title, leader, game

Answer Type Detection

- •Hand-written rules
- Machine Learning
- Hybrids

Answer Type Detection

- Regular expression-based rules can get some cases:
 - Who {is|was|are|were} PERSON
 PERSON (YEAR YEAR)
- Other rules use the question headword: (the headword of the first noun phrase after the wh-word)
 - Which city in China has the largest number of foreign financial companies?
 - What is the state **flower** of California?

Answer Type Detection

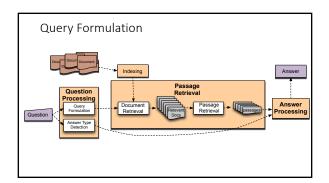
- · Most often, we treat the problem as machine learning classification
 - Define a taxonomy of question types
 - Annotate training data for each question type
 - Train classifiers for each question class using a rich set of features.
 - features include those hand-written rules!

Features for Answer Type Detection

- Question words and phrases
- Part-of-speech tags
- Parse features (headwords)
- Named Entities
- Semantically related words

Which city in China has the largest number of foreign financial companies?

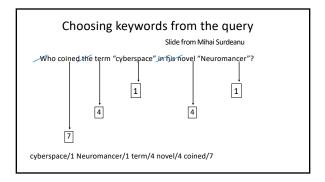
What is the state **flower** of California?

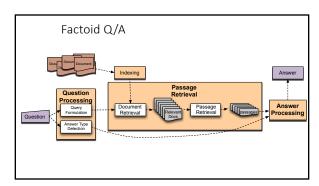


Keyword Selection Algorithm

Dan Moldovan, Sanda Harabagiu, Marius Paca, Rada Mihalcea, Richard Goodrum, Roxana Girju and Vasile Rus. 1999. Proceedings of TREC-8.

- 1. Select all non-stop words in quotations
- 2. Select all NNP words in recognized named entities
- 3. Select all complex nominals with their adjectival modifiers $% \label{eq:complex} % A = \{ (x,y) \in \mathbb{R}^{n} \mid (x,y) \in \mathbb{R}^{n} : (x,y) \in \mathbb{R}^{$
- 4. Select all other complex nominals
- 5. Select all nouns with their adjectival modifiers
- 6. Select all other nouns
- 7. Select all verbs
- 8. Select all adverbs
- 9. Select the question focus word (skipped in all previous steps)
- 10. Select all other words





Passage Retrieval and Answer Extraction

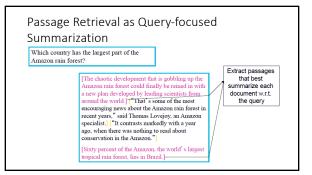
Passage Retrieval

- Step 1: IR engine retrieves documents using query terms
- Step 2: Segment the documents into shorter units
 - something like paragraphs
- Step 3: Passage ranking
 - Use answer type to help rerank passages

Features for Passage Ranking

Either in rule-based classifiers or with supervised machine learning

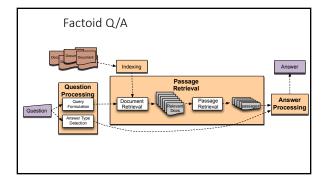
- Number of Named Entities of the right type in passage
- Number of query words in passage
- Number of question N-grams also in passage
- · Proximity of query keywords to each other in passage
- · Longest sequence of question words
- Rank of the document containing passage



Passage Retrieval as Query-focused Summarization

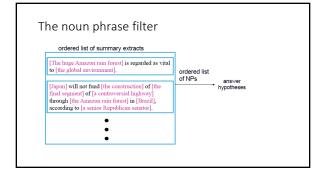
- Decide on a summary length (10% of document length).
- Use standard ad-hoc retrieval algorithm to retrieve top k documents.
- Treat each sentence/paragraph in top N documents as a document itself.
- Use standard document similarity equations to assign a similarity score to the sentence/paragraph.
- Return highest-scoring sentences/paragraphs as the summary, subject to the length constraint.

Passage Retrieval as Query-focused Summarization Which country has the largest part of the Amazon rain forest? [The chaotic development that is gobbling up the Amazon rain forest could finally be reined in with a new plant developed by leading scientists from around the world.]† That's some of the most encouraging news about the Amazon rain forest in recent years," said Thomas Lovejoy, an Amazon specialist.] "It contrasts makedly with a year ago, when there was nothing to read about conservation in the Amazon." [Sixty percent of the Amazon, the world's largest tropical rain forest, lies in Brazil.]



Answer Extraction

- Run an answer-type named-entity tagger on the passages
 - Each answer type requires a named-entity tagger bit the passages
 - If answer type is CITY, tagger has to tag CITY
 - Can be full NER, simple regular expressions, or hybrid
- Return the string with the right type:
 - Who is the prime minister of India (PERSON)
 Manmohan Singh, Prime Minister of India, had told
 left leaders that the deal would not be renegotiated.
 How tall is Mt. Everest? (LENGTH)
 - How tall is Mt. Everest? (LENGIH)
 The official height of Mount Everest is 29035 feet



Adding Analysis Patterns

- "Who is Elvis?"
 - Question type: "who"
 - Named-entity tagging: "Who is <person-name> Elvis</person-name>"
 - Analysis pattern: if question type = "who" and question contains <person-
- · Desired answer probably is a description
- Likely answer extraction patterns
 - "Elvis, the X", e.g., "Elvis, the king of rock and roll!"
 "the X Elvis", e.g., "the legendary entertainer Elvis"

Ranking Candidate Answers

· But what if there are multiple candidate answers!

Q: Who was Queen Victoria's second son?

Answer Type: Person

Passage:

The Marie biscuit is named after Marie Alexandrovna, the daughter of Czar Alexander II of Russia and wife of Alfred, the second son of Queen Victoria and Prince Albert

Ranking Candidate Answers

• But what if there are multiple candidate answers!

Q: Who was Queen Victoria's second son?

Answer Type: Person

Passage:

The Marie biscuit is named after Marie Alexandrovna, the daughter of Czar Alexander II of Russia and wife of Alfred, the second son of Queen Victoria and Prince Albert

Use machine learning:

Features for ranking candidate answers

Answer type match: Candidate contains a phrase with the correct answer type. Pattern match: Regular expression pattern matches the candidate.

Question keywords: # of question keywords in the candidate.

Keyword distance: Distance in words between the candidate and query keywords Novelty factor: A word in the candidate is not in the query.

Apposition features: The candidate is an appositive to question terms

Punctuation location: The candidate is immediately followed by a comma, period, quotation marks, semicolon, or exclamation mark.

Sequences of question terms: The length of the longest sequence of question terms that occurs in the candidate answer.

Candidate Answer scoring in IBM Watson

- Each candidate answer gets scores from >50 components
- (from unstructured text, semi-structured text, triple stores)
- logical form (parse) match between question and candidate
- · passage source reliability
- geospatial location
- California is "southwest of Montana"
- temporal relationships
- · taxonomic classification

Common Evaluation Metrics

- 1. Accuracy (does answer match gold-labeled answer?)
- 2. Mean Reciprocal Rank
 - For each query return a ranked list of M candidate answers.
 - Query score is 1/Rank of the first correct answer

 - If first answer is correct: 1
 else if second answer is correct: ½

 - else if third answer is correct: ½, etc.
 Score is 0 if none of the M answers are correct

$$\bullet$$
 Take the mean over all N queries

$$MRR = \frac{\sum_{i=1}^{N} \frac{1}{rank_i}}{N}$$

Knowledge in QA

Relation Extraction

- Answers: Databases of Relations
 - born-in("Emma Goldman", "June 27 1869")
 - author-of("Cao Xue Qin", "Dream of the Red Chamber")
 - Draw from Wikipedia infoboxes, DBpedia, FreeBase, etc.
- Questions: Extracting Relations in Questions

Whose granddaughter starred in E.T.?

(acted-in ?x "E.T.")

(granddaughter-of ?x ?y)

Temporal Reasoning

- Relation databases
 - (and obituaries, biographical dictionaries, etc.)
- IBM Watson

"In 1594 he took a job as a tax collector in Andalusia" Candidates:

- Thoreau is a bad answer (born in 1817)
- Cervantes is possible (was alive in 1594)

Context and Conversation in Virtual Assistants like Siri

- Coreference helps resolve ambiguities
 - U: "Book a table at II Fornaio at 7:00 with my mom"
 - U: "Also send her an email reminder"
- Clarification questions:
 - U: "Chicago pizza"
 - S: "Did you mean pizza restaurants in Chicago
 - or Chicago-style pizza?"

Limitations of Factoid Q/A

- Question must query a specific fact that is explicitly stated somewhere in the document corpus.
- Does not allow aggregating or accumulating information across multiple information sources.
- Does not require "deep compositional" semantics, nor inferential reasoning to generate answer.

Reading Comprehension Q/A

- Answer questions that test comprehension of a specific document.
- Use standardized tests of reading comprehension to evaluate performance (Hirschman et al. 1999; Rilo & Thelen, 2000; Ng et al. 2000; Charniak et al. 2000).

Sample Reading Comprehension Test

School Edds: Class Up Creat!

GYESPET, WASHINGTON, June, 1985)— Then taken five years of hard work.

Mar props of falshes we cleaned up Topens Creat.

In 1981, the washes washes to the second of the

Q1: Who cleaned up the creek? Q2: What was the name of the creek? Q3: When did the work begin?
Q4: Where is the bed of a creek? Q8: Why did the kid: make booklets?

Large Scale Reading Comprehension Data

- DeepMind's large-scale data for reading comprehension Q/A (Hermann et al., 2015).
 - News articles used as source documents.
 - $\bullet \ \ Questions \ constructed \ automatically \ from \ article \ summary \ sentences.$

	CNN			Daily Mail		
	train	valid	test	train	valid	test
# months	95	- 1	- 1	56	- 1	1
# documents	90,266	1,220	1,093	196,961	12,148	10,397
# queries	380,298	3,924	3,198	879,450	64,835	53,182
Max # entities	527	187	396	371	232	245
Avg # entities	26.4	26.5	24.5	26.5	25.5	26.0
Avg # tokens	762	763	716	813	774	780
Vocab size	118,497			208,045		

Sample DeepMind Reading Comprehension Test



Deep LSTM Reader

 DeepMind uses LSTM recurrent neural net (RNN) to encode document and query into a vector that is then used to predict the answer.



Incorporated various forms of attention to focus the reader on answering the question while reading the document

Visual Question Answering (VQA)

- Answer natural language questions about information in images.
- VaTech/MSR group has put together VQA dataset with ~750K questions over ~250K images (Antol et al., 2016).



