Introd	

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

# brmson (YodaQA) A DeepQA-style Question Answering Pipeline

Petr Baudiš (pasky@ucw.cz)

FEE CTU Prague

January 2015

YodaQA Architecture

Current Performance

Review, Future Work

## Outline

## 1 Introduction

2 YodaQA Architecture

**3** Current Performance

4 Review, Future Work

▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

Introduction ●000	YodaQA Architecture	Current Performance	Review, Future Work 0000
Petr Baudiš			

### Second year **PhD student** at FEE CTU Prague (Petr Pošík + Jan Šedivý), Masters degree in AI from Charles University in Prague

Strong software engineering background: The original Git team, GNU libc development, many open source projects, freelancing

Solid AI, RL background: Computer Go research (MCTS software Pachi — top OSS program, ~4th worldwide)

Basic ML, optimization background: Algorithm portfolios, etc.

Relative newbie in Natural Language Processing!

▲□▶▲□▶▲目▶▲目▶ 目 のへぐ

Introduction ○●○○	YodaQA Architecture 00000	Current Performance	Review, Future Work 0000
brmson			
A Que	estion Answering syste and its DeepQA pi		Vatson

- Practicality
- Primary goals: Extensible design
  - Academic reusability

**Current aims:** Open-domain factoid questions (TREC QA), replicating the DeepQA scheme with 75% recall, 37.5% accuracy-at-1.

Multiple implementations: BlanQA (legacy), YodaQA (current)

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

# brmson: BlanQA (legacy)

- BlanQA: Legacy pipeline based on CMU's OAQA
- Java, UIMA without CAS branching, UIMA-ECD
- Architecture based on OAQA helloqa prototype GitHub branch, but rewritten almost from scratch
- enwiki in solr, Ephyra answer type system,
   Ephyra modules provide the actual algorithms and rules
- Complete setup documentation, fairly clean code
- Interfaces: Interactive and chatbot (IRC)
- Functional OAQA end-to-end pipeline!

YodaQA Architecture

Current Performance

Review, Future Work

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ● ●

# brmson: YodaQA (current)

- YodaQA: "Yet anOther Deep Answering pipeline"
- Designed and implemented from scratch again
- Java, UIMA, DeepQA-style CAS branching, UIMAfit
- Architecture based on simplified DeepQA (as published)
- Every entity (question, retrieved document, answer) == CAS
- NLP analysis: Third-party UIMA annotators via DKpro
- Uses type coercion and general parse tree heuristics; no fixed regex patterns or pre-defined question types

YodaQA Architecture

Current Performance

Review, Future Work

# Outline

## 1 Introduction

### 2 YodaQA Architecture

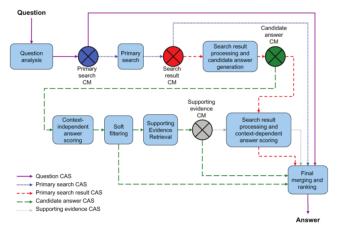
### **3** Current Performance

### 4 Review, Future Work

- ▲日 > ▲ 園 > ▲ 国 > ▲ 国 > 今 Q Q

Introduction	YodaQA Architecture	Current Performance	Review, Future W
0000	●0000		0000

# YodaQA Pipeline



DeepQA architecture (Epstein et al., Making Watson fast). Many modules in YodaQA are obviously much simpler.

・ロト・日本・日本・日本・日本・日本

Introduction	YodaQA Architecture	Current Performance	Review, Future Work
0000	0●000		0000
Question An	alysis		

- Full dependency parse
- Focus generation (hand-crafted dependency, pos rules)
  - What was the first book written by Terry Pratchett?
  - The actor starring in Moon?
- LAT (Lexical Answer Type) generation (from focus)
  - Where is Mount Olympus? location
- Clues (search keywords, keyphrases) generation:
  - POS and constituent token whitelist
  - Named entities
  - Focus and the NSUBJ constituent
  - enwiki article titles

### Outcome: Set of Clue and LAT annotations

Introduction	YodaQA Architecture	Current Performance	Review, Future Work
0000	00●00		0000
Answer Pr	roduction		

Several answer production pipelines run independently in parallel.

- SolrFull: Passage-yielding search
  - *Fulltext:* Full-text + title search for clues, passages containing clues are considered
  - Title-in-clue: Title search for clues, initial passage is considered
  - Passages are parsed, NEs and NPs are answer candidates
- SolrDoc: Full-text search for clues, document titles are answer candidates
- *DBpedia:* Structured data, attributes of clue resources

#### Outcome: Set of candidate answers

・ロト ・ 同 ト ・ ヨ ト ・ ヨ ・ つ へ の

Introduction	YodaQA Architecture	Current Performance	Review, Future Work
0000	000●0		0000
$\Delta$ nswer $\Delta$	nalvsis		

- Each answer is POS-tagged and has dependency tree, Focus generated (dependency root)
- LAT generation named entity type, DBpedia concept type, WordNet instance-of relation, rule for CD POS
- Type coercion of question + answer LAT: Unspecificity is path length in the WordNet (hypernymy, hyponymy) graph
- Answer features (help determine trustworthiness) for:
  - Phrase origin, clue overlaps
  - Generated LATs, type coercion
  - 81 features in total
- Logistic regression generates answer confidences

### Outcome: Ordered set of Answers

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

# Answer Merging and Evidencing

### After initial scoring:

- Take just the top 100 answers (reduce noise for next classifiers)
- Measure overlaps with other answers and diffuse scores
- Re-score

#### After second scoring:

- Further evidence gathering perform full-text search on the top five answers
- Re-score

### Outcome: Ordered set of Answers

YodaQA Architecture 00000 Current Performance

Review, Future Work

# Outline

## 1 Introduction

2 YodaQA Architecture

**3** Current Performance

4 Review, Future Work

▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

Introduction 0000	YodaQA Architecture	Current Performance	Review, Future Work 0000
Testing Da	ataset		

- TREC QA 2002 + 2003 XML datasets converted to plaintext
- Curated (pruned and with revised answer patterns), extended with an IRC BlanQA dataset
- 430 training questions (also used for development),
   430 testing questions (held out)
- 2 × 430 is current practical limit for measurement turn-around (2-3 hour evaluation runs on my home computer)

Matching correct answers with regexes has severe limits

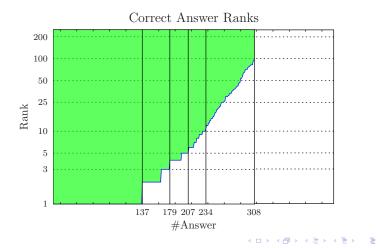
YodaQA Architecture

Current Performance

Review, Future Work

# Experimental Results (Test Set)

Candidate answer binary recall: 74.1% Final answer accuracy-at-one: 31.6%



500

Introduction 0000	YodaQA Architecture	Current Performance	Review, Future Work 0000
Analysis <sup>-</sup>	Tools		

```
$ data/eval/trecnew-single200-measure.sh
```

. . .

```
$ data/eval/tsvout-stats.sh | head -n 5
3b46430 14-08-14 CluesMergeByText: Al... 29/134 14.5%/67.0% as 0.424
fdb239b 14-08-11 Revert "CluesToConce... 23/131 11.5%/65.5% as 0.395
4cd4a09 14-08-10 Clue: Add a label fe... 21/131 10.5%/65.5% as 0.390
e8ad387 14-08-10 SolrFullPrimarySearc... 24/131 12.0%/65.5% as 0.389
acf17eb 14-08-10 ClueConcept, CluesTo... 18/126 9.0%/63.0% as 0.372
```

```
$ data/eval/tsvout-compare.sh 01718ca 8fc9856
----- Gained answer to:
1424 Who wo... for best actor in 1970? George C. Scott 0.00 1.00
----- Improved score for:
1417 Who wa... in less than four minutes? Roger Bannister 0.57 0.77
----- Worsened score for:
1408 Which ... Lionel Jospin a member of? Socialist 0.31 0.30
----- Lost answer to:
1427 What w... spaceship on the moon? Eagle 0.04 0.00 $
```

Introduction	YodaQA Architecture	Current Performance	Review, Future Work
0000		000●	0000

### AdaptWatson-style Analysis Snapshot

- 1407 When did the shootings at Columbine happen? | April 20\s?, 1999 #stopwords/#synsets Columbine High School massacre does not appear; ignore "happen" or add synset that includes "occur"
- 1439 How deep is Crater Lake? | 1\s?,\s?932 feet #wikipieces "Crater Lake" yields "crater lake" matches and never the main article; also, it should be 1,943 feet
- 1666 What is the name of the US military base in Cuba? | Guantanamo #abbrev US -> U.S., then should work (answer Guantánamo!)
- 1606 What is the boiling point of water? | 212 degrees Fahrenheit 100 °C

. . .

#wikipieces (7) - for all NPs/NEs/nouns in question, include sametitled wikipedia articles in primary search; furthermore, do not split such to sub-clues? #synsets (6) - include synsets instead of words #abbrev (4) - acronym generation / expansion; e.g. PC = P.C. = Personal Computer; in expansion, try using #redirects?

YodaQA Architecture 00000 Current Performance

Review, Future Work

# Outline

## 1 Introduction

2 YodaQA Architecture

**3** Current Performance



▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

Review, Future Work

# brmson: YodaQA vs. Primary goals

### Practicality

- Detailed setup instructions (including data sources setup!)
- Detailed design documentation
- Interactive user interface
- Open source (ASL2 licence), clean code and build system

### Extensible design

- UIMA + DeepQA structure: Easy pipeline branching and addition of new modules
- DKPro: Third-party UIMA annotators (tokenizers, parsers, etc.) are freely replaceable
- Internal UIMA components are as finegrained as possible

### Academic platform

- Gold Standard interface, TREC QA based dataset
- All datasets, evaluation tools and measurements published
- AdaptWatson methodology for performance analysis driving development

▲□ > ▲圖 > ▲目 > ▲目 > ▲目 > のへで

Review, Future Work

# brmson: YodaQA vs. Primary goals

### Practicality

- Detailed setup instructions (including data sources setup!)
- Detailed design documentation
- Interactive user interface
- Open source (ASL2 licence), clean code and build system

#### Extensible design

- UIMA + DeepQA structure: Easy pipeline branching and addition of new modules
- DKPro: Third-party UIMA annotators (tokenizers, parsers, etc.) are freely replaceable
- Internal UIMA components are as finegrained as possible

### Academic platform

- Gold Standard interface, TREC QA based dataset
- All datasets, evaluation tools and measurements published
- AdaptWatson methodology for performance analysis driving development

◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへぐ

Review, Future Work

# brmson: YodaQA vs. Primary goals

### Practicality

- Detailed setup instructions (including data sources setup!)
- Detailed design documentation
- Interactive user interface
- Open source (ASL2 licence), clean code and build system

### Extensible design

- UIMA + DeepQA structure: Easy pipeline branching and addition of new modules
- DKPro: Third-party UIMA annotators (tokenizers, parsers, etc.) are freely replaceable
- Internal UIMA components are as finegrained as possible

#### Academic platform

- Gold Standard interface, TREC QA based dataset
- All datasets, evaluation tools and measurements published
- AdaptWatson methodology for performance analysis driving development

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへで

Current Performance

# YodaQA: Future Work

### TODO List for 1.0:

- Extra unstructured answer features (detailed clue-based context)
- Use Freebase as structured answer source (long-term: *solid* enwiki infobox based datasets)
- (?) Try different answer classifiers (random forests or MLP)
- Aiming to publish on CLEF2015

With more contributors:

- Cleaned up testing dataset
- UIMA component unit tests
- Verification dataset runs with human judges
- Insightful web interface
- Scale-out, parallelization and memory usage optimizations

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

 Apply to some real-world projects and domains YodaQA Architecture

Current Performance

Review, Future Work

## Long-term Plans and Goals

• Post-YodaQA architecture reformulation as IE problem:

Latent knowledge graph paradigm

(QA pipeline as on-demand population of semantic network; answer retrieved by path search, scored by edge coercion)

- brmson-based startup: Looking for good business cases
- Disembodied autonomous agent: QA with deduction + goal-setting + planning (maybe in 15 years)
- Personal: Internship at NII Tokyo in Q1 2015 (answering of Physics questions in university entry exams)

Introduction 0000	YodaQA Architecture	Current Performance	Review, Future Work 000●
Conclusion			

#### Practical, open source QA system

- Clean architecture and development methodology
- Reasonably documented!
- Clear path forward, towards reference experimental testbed

baudipet@fel.cvut.cz
petr.baudis@gmail.com
IRC: #brmson at freenode.net

### Thank you for your attention!

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @