

The Computable News Project

Milestone 6

Cassandra, TAC 2012 and Docrep

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Computable News team

- Project leader: **James Curran**
- Project manager: **Candice Loxley**
- Postdoctoral researchers:
 - **Daniel Tse**
 - second position still being filled (2 year position)
- Senior Software Developer: **Will Cannings**
- PhD students:
 - **Joel Nothman**
 - **Will Radford**
 - **Tim O'Keefe**
- PhD students on related work:
 - **Tim Dawborn**
 - **Andrew Naoum**
 - **Glen Pink**

M6: Engineering

- ✘ Battle with Cassandra (to the death?)
- ??? Supported integration with My Masthead
- ✓✓ Refactored NEL to use docrep
- ✓✓ Initiated 4 new developers within a month
- ✓✓ Queue-based parallelism

M6: Research: NEL performance improvement

- TAC 2012
 - ✓✓ Increased performance by 10% points on 2011 data - state-of-the-art
 - ✓✓ 3.1% points off state-of-the art 2012 over kb queries
- SMH dataset
 - ✓✓ Best system performs at 72.44% F-score (61.25% in milestone 3)

M6: Research: Opinions in quotes

- ✓ Presented quote attribution work at EMNLP 2012
- ✓ New annotation tool for marking opinions in quotes
- ✓ Review of existing opinion extraction work
- ✓ Pilot annotation of 700 documents, including some double annotations
- ??? Annotation scheme – has been through several iterations, but is not yet complete

M6: Research: Event linking

- ✓ Presented task description at ACL 2012
- ✓ Built a framework for experimentation
- ✓ Charted and analysed baseline results
- ✓ Used temporal information to constrain event linking queries:
try to determine an event's publication date from language
- ✓ Began comparing Fairfax hyperlinks to manual annotations

Filling gaps in our Fairfax archives

Article published in	We had before	We now have
2009	1 051	1 055
2010	396	51 146
2011	57 657	104 559
2012	58 840	86 018

- Five nights so far
- Fetching 1am-6am, waiting 0.2s between downloads
- 261 124 requests ($\approx 50\%$) returned No Content!
- Much better understanding of DCDS assets
 - and their URLs
 - still uncertain about how branding information is stored

Cassandra

- Goals
 - Move to a faster, more scalable database architecture
 - Reduce the number of databases used in production
- Requires
 - Replacing the existing multiple database architecture with Cassandra
 - Deploying a version of Fizzing Panda that uses Cassandra
- Tasks
 - Design the Cassandra server cluster (FXJ)
 - Design a database schema (CRC)
 - Migrate existing database content (such as the entity store) (CRC)
 - Convert NEL, APIs and the demo to use Cassandra (CRC)

Progress

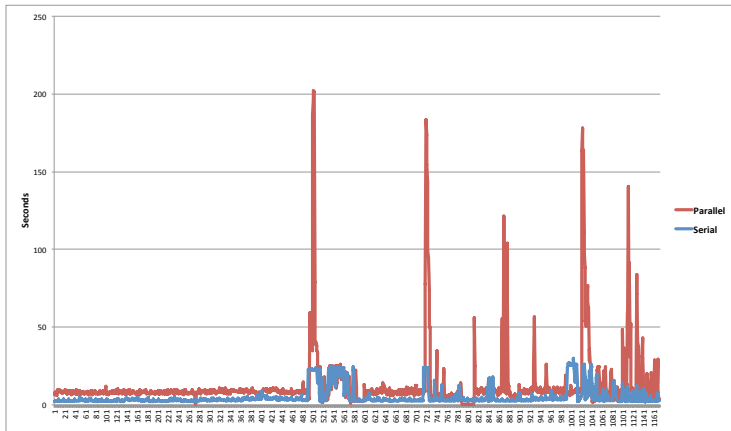
- ✓ An initial database schema has been designed
- ✓ The entity store and data required for linking has been migrated to Cassandra
- ✓ NEL can link end to end using only Cassandra and Solr for full text search
- ✓ Ops have provided a Cassandra cluster for testing
- ✓ Isolated stress tests have been run

Counts 2 Stress Test

Batch Size	1 day of articles (hrs)	1 year of articles (days)
5000	4.04	61.44
2000	1.90	28.87
1000	0.96	14.61
1000 Parallel	2.31	35.2
None	2.42	36.8

- With parallel linkers, it would take over a month to write the counts 2 data for a years worth of articles

Parallel Performance



Gripes

- Counting rows requires reading all counted rows into memory, making counts slow, and placing an upper bound on counts
- Parallel counts on the cluster are affected by spikes in a similar way writes are, making some count operations twice as slow as average
- Mixing reads and writes with multiple clients on one node has caused timeouts and crashes
- At various times Cassandra has lost data, lost indexes, and refused connections with no errors logged

9% improvement on TAC 11 benchmark

System	$B^3 + F1$
Best TAC 11	84.6
CompNews M3	75.4
Median TAC 11	71.6
CompNews M6	84.0

- puts us in the top-3 systems group in TAC 2011

TAC 2012 Results

System	All	In KB	Not in KB
Best	73.0	68.7	84.7
CompNews M6	66.5	65.6	67.5
Median	53.6	49.6	59.4

- KB linking is most aligned with CompNews goals
- NIL-clustering will become important for linking the archive
- Within 3.1 points of state-of-the-art for KB linking
- Well above median performance overall (final results to come)

Improving candidate recall

- Early in pipeline and major limiting factor to NEL performance
- Updated Wikipedia: using April 2012 snapshot
- Manually curated statistical rules for name variation
- Boost title and redirect matches
- In-document coreference for query expansion:
 - Organisational suffixes
 - State abbreviation expansion (e.g. NSW → New South Wales)
 - Bureaucratic name pre-filtering (e.g. Dept of Foreign Affairs)
 - Nicknames (e.g. Christopher/Chris)
- Currently 96% recall for TAC 11 data

Statistical Rules for Name Variation

- Given a large sample of entities and their aliases
- Find common transformations
- Over 500,000 rules extracted from Wikipedia
 - `<1> Corporation` \rightarrow `<1> Corp.`
 - `<1> Inc.` \rightarrow `<1>`
 - `<1> River` \rightarrow `River <1>`
 - `<1> University` \rightarrow `University of <1>`
 - `<1> Texas` \rightarrow `<1> TX`
 - `United States <1>` \rightarrow `US <1>`
 - `Bob <1>` \rightarrow `Robert <1>`

Supervised NEL

- Gives the model the ability to rank candidates by incorporating many sources of evidence in a principled way
- Very common in TAC 11, we implemented features from systems: dmir_inescid and THU:
 - **String features**: which compute functions between the mention text or its background document, and the candidate title or its background document
 - **Topic features**: the similarity of the topic distributions of the mention and its candidate
 - **NER features**: computed over the sets of named entities detected in the mention document and its candidate document
- Results inconclusive so far, but should allow us to fully exploit the SMH annotated dataset

Unsupervised NEL and NIL thresholding

- Higher search recall means a noisier list (many more candidates)
- Simple average of several metrics
- Minimum score threshold, else NIL
- Tuned on development data

11% performance gain on SMH data in 12 months

Milestone	Mention F1
3	61.25
5	69.80
6	72.44

Areas for improvement. . .

Error	Count	%
Span	930	42.5
Wrong KB	546	24.9
NIL as gold	382	17.4
Gold as NIL	333	15.2

NLP is format/data structure hell

- Many layers of annotation:
 - document metadata (e.g. date, author, categories, ...)
 - text: sentence boundaries, tokenisation and word normalisation
 - syntax: part-of-speech tags, parse trees
 - semantics: word senses, named entities, argument roles
 - discourse: topics, coreference, entity links
- ... and headings, paragraphs, hyperlinks, emphasis, etc.
- Each layer has its own legacy text-based format
 - ⇒ docs are stored redundantly ⇒ multiple DB queries
 - ⇒ custom I/O and nasty, inefficient alignment
- Our tools should have access to all layers and talk about them in the same way

Holistic NLP with docrep

- An efficient serialisation format for markup and annotations
 - requirements include pointers, references to text spans
 - layers are deserialised only if an app requests them
- Intuitive, declarative API in C++, Python and Java
- One document == one message/blob
- Communicating over streams or sockets
- Local taggers can exploit document-level context

Tech details: annotation layers in Python

```
1 class Token(dr.Ann):
2     span = dr.Slice()
3     norm = dr.Text()
4     pos = dr.Field()
5
6 class Mention(dr.Ann):
7     span = dr.Slice('Token')
8     type_ = dr.Field()
9     chain = dr.Pointer('Chain')
10
11 class Chain(dr.Ann):
12     concept_id = dr.Field()
```

Tech details: document model in Python

```
1 class Doc(dr.Doc):
2     tokens = dr.Store(Token)
3     mentions = dr.Store(Mention)
4     chains = dr.Store.Chain()
5     asset_id = dr.Field()
6     pub_time = dr.DateTime()
```

And accessing documents with annotations:

```
1 for doc in dr.Reader(open(filename), Doc):
2     for chain in doc.chains:
3         print chain, chain.concepts[0]
```


...in Java

```

1  @dr.Doc public class Doc extends AbstractDoc {
2      @dr.Store public Store<Token> tokens = new Store<Token>();
3      @dr.Store public Store<Mention> mentions = new Store<Mention>();
4      @dr.Store public Store<Chain> chains = new Store<Chain>();
5  }
6
7  @dr.Ann public class Token extends AbstractAnn {
8      @dr.Field public ByteSlice span;
9      @dr.Field public String norm;
10 }
11
12 @dr.Ann public class Mention extends AbstractAnn {
13     @dr.Pointer(store="tokens") public Slice<Token> span;
14     @dr.Field public String type;
15     @dr.Pointer(store="chains") public Chain chain;
16 }

```

...and in C++ (coming soon: Ruby, JavaScript, ...)

```

1  class Mention: public dr::Ann {
2      dr::Slice<Token *> span;
3      std::string type;
4      dr::Pointer<Chain> chain;
5
6      class Schema;
7  };
8
9  class Mention::Schema: public dr::Ann::Schema<Mention> {
10     DR_POINTER(&Mention::span, &Doc::tokens) span;
11     DR_FIELD(&Mention::tag) tag;
12     DR_POINTER(&Mention::chain, &Doc::chains) chain;
13     Schema(void): dr::Ann::Schema<Mention>("Mention"),
14         span(*this, "span", dr::FieldMode::RO),
15         type(*this, "type", dr::FieldMode::RO),
16         chain(*this, "chain", dr::FieldMode::RW){}
17 };

```

Tech details: the wire format

- Based on msgpack
 - A JSON-like binary format
 - But small and fast (before compression!)
 - e.g. each of the following takes 1 byte:
 - integer $-32 \leq n < 128$, true, false, null
 - header for map or array of length < 16
 - header for string of length < 32
- On top of msgpack, docrep:
 - uses small integers where possible
 - but avoids unnecessary data obfuscation
 - stores the data model with each doc \Rightarrow self-describing
 - (un)swizzles pointers
 - deserialises annotation stores only when needed
 - provides lazy dynamic access to other annotations

Implementation status for Computable News

- ✓ Import and tokenisation of RLAY stories
- ✓ Import and tokenisation of syndication feeds
- ✓ C&C tools: POS tags, parse trees, named entities
- ✓ Entity and concept linking
- ✓ Quotation extraction and attribution
- ✓ Event linking experimentation
- ✓ Story rendering to HTML
- ⇒ fewer custom formats
- ⇒ eliminated a large and confusing DB table
- ⇒ more consistent internal code structure
- ⇒ consistent streaming and socket interfaces

Batch correction interface

- Editors can override linking decisions for display
- Annotators can clean up errors, yielding higher-quality data

nelserver batch annotator Annotations will be signed **anonymous** ([change](#))

rome First Previous 1 2 3 4 5 Next Last

Doc id	Concept	Mentions	Missing span	Mixed NE types	Needs review	Annotated by
2489331	Roman Empire	and authorities in Rome are reportedly considering			<input type="checkbox"/>	
2490095	Roman Empire	every socialite and Rome -- pretty much			<input type="checkbox"/>	
4743504	New Rome, Ohio	's fingering in Rome last week of			<input type="checkbox"/>	
4743655	Michael Cowley	Michael Cowley in Rome			<input type="checkbox"/>	

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4743655	Michael Cowley	Michael Cowley in Rome	<input type="checkbox"/>	

New candidate

Rome

Rome (TV series)
Rome, New York
Ancient Rome

Close Save

Identifying opinions in quotes

- We have quotes, topics, and entities
- Can we identify the opinion an entity holds about a topic, via the quotes they're making?

Mr Abbott says that given that change is habit-forming, the best way to stop the “republican extremism” that the conservative republicans say they fear is “to vote down any republic at all – in the same way that deterring major crime requires tackling petty crimes”

Task description

- Identify documents on a topic of interest via keyword search
- Define a “topic statement” which sets out a position on the topic
- Classify quotes as “supporting”, “neutral”, or “opposing”

Progress

- ✓ Annotation tool
- ✓ Review of existing technologies
- ✓ Pilot annotation of 700 documents over 7 topics
 - Republic
 - Carbon tax
 - Reconciliation
 - Work choices
 - Abortion
 - Immigration
 - Same sex marriage

??? Annotation scheme – has been through several iterations

Annotation tool

" Like aristocrats joining a revolution in order to temper its excesses -- useless fools , Lenin called them -- these so-called ' conservatives ' claim that it is necessary to support the Turnbull-Keating republic at the coming referendum in order to stop the far more dangerous Mack-Cleary [direct election] republic emerging just a few years later "

, he says .

But

" anticipating problems is not the same as compromising principles in advance "

, he says , labelling Conservatives for an Australian Head of State as

" mostly lapsed monarchists worried about being on the wrong side of history "

" The fact that the ' conservative ' tag is now being ostentatiously adopted to fight a referendum campaign by people who shunned it when they were running election campaigns shows that this is a marketing ploy rather than a political or philosophical disposition . "

— Tony Abbott

With regards to the topic statement "Australia should cease to be a monarchy with the Queen as head of state and become a republic with an Australian head of state", could you use the quote to convince someone of the speaker's position

assuming the person is not knowledgeable about the topic?

- ☒ Strongly or clearly opposing
 ☐ Opposing
 ☐ Neither supporting nor opposing
 ☐ Supporting
 ☐ Strongly or clearly supporting

assuming the person is knowledgeable about the topic?

- ☒ Strongly or clearly opposing
 ☐ Opposing
 ☐ Neither supporting nor opposing
 ☐ Supporting
 ☐ Strongly or clearly supporting

I can't assign a sentiment to this quote because

- ☐ It is not a quote
 ☐ There is no sensible choice for the sentiment
 ☐ Is not related to the [topic](#)

I can choose a sentiment but this quote has

- ☐ an incorrect [speaker](#)
☐ an incorrect [quote span](#)

Add a comment

sentiment expressed by the speaker from the quote and its immediate context with regards to the following statement:

"Australia should cease to be a monarchy with the Queen as head of state and become a republic with an Australian head of state"

You should have already read the [annotation guide](#) for information on disambiguation, special cases, and other problems that may arise in annotation. Refer back to it if you are uncertain of what to do, or, provide a comment if the guide does not cover an issue.

Article

Article title [Abbott lashes 'lapsed monarchists'](#)

Article topic [republic](#)

Quotes remaining **0**

[Save](#) ☐ Article topic incorrect

Shortcuts

Sentiment scale (no context)	(1, 2, 3, 4, 5)
Sentiment scale (using context)	(Q, W, E, R, T)
Previous quote	J
Next quote	K
Next unannotated	L
Not a quote	Y

Named Entity Linking

- Write TAC 12 system report
- Supervised NEL
- Apposition
- Batch annotation to improve training data quality
- Batch correction to identify specific problem areas
- next steps in live linking editor

NER in noisy text

- Prerequisite for NEL
- Existing algorithms can be fragile when given noisy text
- **CompNews applications:**
 - **ocr:** scanned historical archives, books
 - **transcripts:** video, audio
 - **social text:** comments, blogs
 - **translations:** transliterated names

Slot-filling

- Find the values of specified attributes of an entity (“slots”) in a collection of documents
 - e.g. for a person find their title, age, country of residence, employer, family, etc.
- Currently surveying current approaches and setting up a baseline system for slot-filling
- **CompNews applications:**
 - **nel:** one person cannot have two (correct) dates-of-birth!
 - **editor:** extracting information to insert into news text