Building a DeepDive Application Infrastructure

Ian Ross, University of Wisconsin-Madison Center for High Through Computing

iross@cs.wisc.edu









Key Questions

 Can a machine reading system construct a literature-based data compilation suitable for science?

 How can we build a text data mining (TDM) infrastructure that drives this data compilation (and other literature-heavy applications)?



Key Questions (SPOILERS)

- Can a machine reading system construct a literature-based data compilation suitable for science?
 - Yes DeepDive

CENTER FOR

ROUGHPUT

- How can we build a text data mining (TDM) infrastructure that drives this data compilation (and other literature-heavy applications)?
 - High throughput computing is the heart

Ian Ross — University of Wisconsin - Madison

Table of Contents

- Introduction
 - General overview of DeepDive
 - Overview of PaleoDeepDive results
- Infrastructure Challenges
- Current solutions

IROUGHPUT

- Overview of system
- Next steps

CENTER FOR

Ian Ross — University of Wisconsin - Madison

Introduction — DeepDive

- What is DeepDive?
 - System that leverages machine learning techniques on domainspecific knowledge to drive data analysis
 - Learning at a "distance" experts in a field can build simple rules and let DeepDive do the hard work (i.e. no tedious training)
- <u>http://deepdive.stanford.edu/</u>

HROUGHPUT

CENTER FOR





Deep**Dive**

 Relation: has_spouse

 Subject
 Object
 Prob.

 Barack Obama
 Michelle Obama
 0.97

 ...

Introduction — PaleoDeepDive

- DeepDive technology applied to paleontology
 - Extracts relations between biological taxa, geological formations, geographic locations, and geological time intervals
 - "PaleoDeepDive performs comparably to humans in several complex data extraction and inference tasks and generates congruent synthetic results that describe the geological history of taxonomic diversity and genus-level rates of origination and extinction."
 - <u>"A Machine Reading System for</u> <u>Assembling Synthetic Paleontological</u> <u>Database" (Peters, Zhang, Livny, Re)</u>
- <u>http://deepdive.stanford.edu/doc/</u> <u>paleo.htm</u>

IGH THROUGHPUT

CENTER FOR

COMPUTING



Now what?

 PaleoDeepDive shows that machine reading can build a database that not only compares well to human-built ones, but it also systematically improves as information is added.

So let's build a TDM pipeline to streamline the process of building a DeepDive application!



OUGHPUT

The Infrastructure Challenges

- Sources of documents
 - Legal and responsible access to scientific literature
- Organization
 - Bibliographical information for each document
 - What versions of what tools have we used to process each document? Tracked via a *tag*
- Computing
 - Need resources, automation, and flexibility
 - This framework should be useful for non-DeepDive applications as well!
- Monitoring
- Security
- Discovery Need easy ways to extract documents of interest (currently using ElasticSearch to search the extracted text)

HROUGHPUT Ian Ross — University of Wisconsin - Madison

Challenges — Organization

db.articles.findOne()

```
"_id" : ObjectId("54db884ee138238a47f95eb0")
  "sha1" : "1c918973dcb0a5d2a39ed9afedf955b0b1fe4f7d",
  "filepath" : "/home/iross/DeepDiveEnv/DeepDive/downloads/Journal of Hydrology/0022169465901101.pdf",
  "title" : "Choice of adjustment to floods Gilbert F. White: University of Chicago, Chicago Ill., 1964, 150 p. (Research pape
93)".
   'vol" : "3-4",
  "pubname" : "Journal of Hydrology",
        : "http://api.elsevier.com/content/article/pii/0022169465901101",
  "startinaPaae" : 344
   'authKeywords"
           "2015-02-11 10:49:49.625599",
   cuneiform_processing" :
          "elsevier_002'
                   'harvested" : true
                   'contents" : [].
                   'filename" : [ ]
   'ocr_processing" : {
          "elsevier_002" : {
                   'harvested" : true,
                   "contents" : [],
                   "filename" : [
                           '/home/iaross/elsevier_002/ChtcRun/submit_12Feb_out/job000005/page-1.hocr.html"
   'nlp_processing" : {
           'elsevier 002" :
                   'harvested" : true,
                   'contents" : [],
                            '/home/iaross/elsevier_002/ChtcRun/submit_12Feb_NLP_out_NLP/job000004/input.text"
```

Store information in mongodb

Each articles has the bibliographical information, filelevel info (filepath, SHA1 sum, time fetched), and processing information

Subdocuments tell whether the article has been processed for a certain *type+tag* combination.

This article has had the Cuneiform, Tesseract, and NLP processing done, though the cuneiform failed

New types of processing can easily go within a new subdocument, and new tags can be added to each



The Infrastructure Challenge — Computing



Ian Ross — University of Wisconsin - Madison

COMPUTING

The Infrastructure Challenge — Computing



11

The Infrastructure Challenge — Computing



- At least 10,000 articles a week
- Fairly small/short analysis jobs (5-10 minutes on average for OCR jobs, slightly longer for NLP)
- High throughput computing is exactly what we need!
- Use HTCondor and the UW CHTC resources for all this processing work.

The Infrastructure Challenge — Processing

- Specific needs:
 - Automation and organization 10,000 articles x 4 different processing steps = Potential management nightmare
 - Organizational structure makes it easy to ID articles that need processing.
 - Security
 - Flexibility New tools and document sources should be easy to add to the pipeline
 - New documents are easy if there's an entry in the database, they'll get processed



The Infrastructure Challenge — Processing

- Specific needs:
 - Automation and organization processing steps = Potential m
 - Organizational structure ma processing.
 - Security

CENTER FOR

- Flexibility New tools and doo to the pipeline
 - New documents are easy they'll get processed

HROUGHPUT

- Provided by HTCondor!
 - DAGMan + postscript for organization
 - Encrypted filesystem ensures PDFs won't be left exposed on the execute nodes
- A cron'ed python script satisfies the automation/ flexibility requirements

The Infrastructure Challenge — Monitoring

- Use a round-robin database, make plots of fetched and processed counts for each type of job
 - Have hour/day/week/month views, so we can identify impact at a glance
 - Also provides an estimate of how many CPU hours were used



Week

HIGH THROUGHPUT Ian Ross — University of Wisconsin - Madison

Current Status — System Overview



What lives here: Processing/harvesting. Processed output, databases, web server for monitoring, cronjob for NLP/FontType processing

CENTER FOR HIGH THROUGHPUT

COMPUTING

Ian Ross — University of Wisconsin - Madison

databases.

What lives here: *Fetching*. PDFs, daily backup of

Current Status

- Infrastructure fully in place
- Work on DeepDive application ongoing
- Applications being built off of the existing articles database:
 - Exposing articles and metrics information via API
 - Fulltext search article discovery application

HIGH THROUGHPUT

CENTER FOR



(1 to 10 shown below)

prev next

Brewster-Wingard, G.Lynn; Scott, Thomas M.; Edwards, Lucy E.; Weedman, Suzanne D.; Simmons, Kathleen R., date not known. Reinterpretation of the peninsular Florida Oligocene: an integrated stratigraphic approach, Sedimentary Geology original article

... COSUNA 1966 This Study (southern Florida) 211 Puri and Vernon 1964 Hawthorn Formation Tampa Limestone Suwannee Limestone Peace River Arcadia Formation Suwannee Limestone Peace River Formation imestone Fig. 3. Comparison of historical perspectives on the Oligocene-Pliocene stratigraphic...

...) raised the Hawthorn Formation to group status and delineated the component formations and members. In the study area, the Hawthorn Group consists of, in ascending order, the *Arcadia* Formation with the Nocatee and Tampa Members, and the *Peace River Formation*. Scott (1988) suggested that the Tampa...

... suprajacent *Peace River Formation* (Fig. 3). The *Arcadia* Formation, which straddles the Oligocene-Miocene boundary (Jones et al., 1993; Missimer et al., 1994; Wingard et al., 1994), has two named members with limited area1 distribution, the Tampa Member and the Nocatee Member (Scott, 1988). The terms upper...

... mollusks, echinoderms, bryozoans, corals, and barnacles. The *Arcadia* Formation ranges from absent to more than 200 m thick. The portion of the formation deposited during the Oligocene may exceed 100 m in thickness (W-16814, Fig. 4). The early Miocene to lower Pliocene *Peace River Formation* consists of...

Ian Ross — University of Wisconsin - Madison

Current Status PaleoDeepDive Input Builder

- Use article search API to define a collection of articles to use in a DeepDive application
 - This example is a fulltext query for Phanerozoic
- Preview top search results
- Click "Bundle" to build the ulletDeepDive-ready PostgreSQL databases from the existing NLP and FontType/Layout processes

HROUGHPUT

CENTER FOR

Phanerozoic							
Publication name							
Preview Bundle							
1 .	15497	result	s fou	nd! (Top) 10	0 shown)	
	Geobiology of microbial carbonates: metazoan and seawater saturation state influences on secular trends during the Phanerozoic						
	Pub: Palaeoge	ography, Palaeoclir Palaeoecology	matology,	Auth: Riding, Robert; Liang, Liyuan		Vol.Issue.Pages: 1-2101- 115	
	Variable Phanerozoic thermal history in the Southern Canadian Shield: Evidence from an apatite fission track profile at the Underground Research Laboratory (URL), Manitoba						
	Pub: Tectonophysics	Auth: Feinstein, Shimon; Kohn, Barry; Osadetz, Kirk; Everitt, Richard; Vol.Issue.Pages O'Sullivan, Paul 475.1.190-199				Vol.Issue.Pages: 475.1.190-199	
	Phanerozoic marine biodiversity follows a hyperbolic trend						
	Pub: Palaeoworld	Auth: Mark	v, Alexander V.; Korotayev, Andrey V. V		Vol.I	ol.Issue.Pages: 16.4.311-318	
	Editorial Essay						
	Pub: Precambrian Research		Auth: Simpson, George Gaylord		Vol.Is	Vol.Issue.Pages: 7.2.101-103	
	What, if anything, happened at the transition from the Precambrian to the Phanerozoic?						
	Pub: Precambrian Research		Auth: Lowenstam, Heinz A.		Vol.Issue.Pages: 11.2.89-91		

Conclusions — Key Infrastructure Features

• Automated document fetching at arbitrary maximum rates determined by content providers (e.g., Elsevier 10K/week/API key)

 Secure document storage; encrypted processing methods to protect content owners/providers

• HTC infrastructure to run core tools (e.g., NLP, OCR, table recognition/ parsing, image analysis); tool versioning and success/failure/quality analysis

• API layer with basic capacity to identify documents of potential relevance to a project, with initial results returned as (augmented) bibJSON

 Packaging and delivery of PaleoDeepDive-ready raw data (e.g., PostgreSQL database of NLP results); everything traceable back to specific sources (original URL and locations within documents)

Next Steps

- Currently working on:
 - More modular processing framework
 - Additional tools (e.g. figure extraction)
 - Enhanced article discovery that leverages all possible tools
 - Using the infrastructure as a starting point for a DeepDive application!

Questions?

- <u>deepdive.stanford.edu</u>
- <u>http://deepdive.stanford.edu/doc/paleo.html</u>
- <u>http://journals.plos.org/plosone/article?id=10.1371/journal.pone.0113523</u>

