You Can’t Play 20 Questions with Nature And Win
Oren Etzioni, CEO
Allen Institute for AI (Al2)


## 

rerer


## Big Data Tidal Wave

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## Deep Learning



## Some Skepticism about Deep Learning



## Outline

I. Limits of current paradigm
II. Today's KBs: fact rich but knowledge poor III. Machine Reading
IV. Allen Institute for AI (AI2)


## What are the limits of the classification paradigm?

## Critique applies to both discriminative and generative methods!

## Example of State of the Art

## t Retweets - NYTimes.com - Google Chrome

nes.com/2014/07/03/upshot/a-25-question-twitter-quiz-to-predict-retweets.html
ites

HOME Q SEARCH

## ETo Now Hork Eimes

Consider these two tweets by Al Gore, both promoting the same article:

Sendhil Mullainathan
@m_sendhil

40\% of smartphone users connect to Internet immediately upon awakening, before leaving bed.


## What about creating

 Tweets in the first place?Can you guess whi as retweeted more often?

## Product Recomp



## Prototype in the cloud, then easily move back on-premises

## After a year of medical school, IBM's Watson passes first milestone

Summary: IBM's year-long residency at Memorial Sloan-Kettering Cancer Center and Wellpoint is finally producing cognitive computing breakthroughs (and two new products).


By Andrew Nusca for Between the Lines | February 8, 2013 - 16:45 GMT (08:45 PST)
3 Follow @editorialiste Get the ZDNet Big Data newsletter now

IBM's Watson has made its first commercially-developed cognitive computing breakthroughs, the company announced this morning. They arrive after more than a year of training with healthcare partners WellPoint and Memorial SloanKettering Cancer Center.



Six Windows 8.1 tips for
maximum
performance, reliability, and

ALLEN INSTITUTE for ARTIFICIAL INTELLIGENCE

## More Challenges for Classificaiton

- Learning Arithmetics =
- set of rules \& procedures
- 100\% accuracy
- Distribution insensitive learning
- Playing Chess: move choice requires search Deep Blue
human grand masters


## Reminiscent of...

- System-one: fast, associative, but biased, error prone.
- System two: deliberative, explicit, more reliable.

THVNKING,<br>FAST TAD $\operatorname{SLOW}$<br>DANIEL<br>KAHNEMAN<br>WINNER OF THE NOBEL PRIZE IN ECONOMICS

## Fundamental Claim

Classification, prediction limited
("structured prediction"; "multi-task learning" are steps forward)


Need a process that's constructive, multi-layered,
Knowledge-intensive

Data $\rightarrow$ knowledge $\rightarrow$ reasoning


## The Bottom Line

## "You can't play 20 questions with nature and Win." Allen Newell, 1973

Key Al tasks cannot be reduced to classification problems!
(using realistic feature sets)

## Natural Language Understanding

- Complex syntax, semantics, pragmatics, discourse, reference, context, etc.
- Background knowledge


## Winograd Schema Challenge (Levesque, 2011)

The lara table be World knowledge is necessary for elementary understanding, so do we have it?


## Simple Thought Experiment

How much do you know about apples?

- Edible fruit
- Has peel, stem, flesh
- Consists of mostly water
- Ripens on a tree at least 20 non-trivial facts + inheritance...


## 2014 Knowledge Tour

About 29T:000 000 rapats (0. 37 shetrith)

## Apple

www apple com/ * Apple Inc. .
Apple designs and creates IPod and ITunes, Mac laptop and desktop computers, the OS X operating system, and the revolutionary IPhone and iPad.
$3.8 \star \hbar \# * * 46$ Google reviews Write a review

- 2656 NE University Village St. Seattle, WA 98105 (206) 892-0433


## Apple Store

Shop IPad - Shop IPhone -
Shop Mac - Refurbished \&
Clearance
iPhone
Discover everything
iPhone, including the most advanced ...

## Support

Apple Support is your starting point for help with Apple products .
More results from apple.com *

## News for apple



Another Apple-Samsung Skirmish Heads to Court ABC News - 4 mimutus pro
The fiercest rivaliry in the worid of smartphones is heading back to court this week in the heart of the Silicon Valley, with Apple and Samsung ...

Apple Laop iPhone 6 Scruens Microsef Ofice For Pad, Bieacons For M.B. Secro... Fobbes - by Ewar Spence - Thours aga

Apple plann to create Transparent Phenes with rear camera technology The independent -4 hourt 3 sop


## Apple Store University Village

Dirsectona With s review
Address: 2656 NE Uviversity Vilage St. Seattle, WA 98105
Phone: (206) 852-0433
Prices: $\$ 555$
Hours: Open today $11: 00$ att -6.00 pm

## Reviews

$3.8+t+t+46$ Googie renews
More reviews: Mb/sboak.com yahoo com imsiderpages cam senicearea com

People also search for


## See results about

## Appla inc

Consumer electernics congriny
Apple inc. is an Amerncan miltinational corporaion
hosdquartered in Cupertino. Calfomia that


Web Images Shopping Videos News More * Search tools

Abtui 305000 Iesults 10.27 secunts)

Apple - Wikipedia, the free encyclopedia
en wikipadia ungminlApple * Wikipedia *
The apple is the pomaceous fruit of the apple tree, species Malus domestica in the rose family (Rosaceae), It is one of the most widely cultivated tree fruits, and.
Apple Inc. - Disambiguation - Malus sieversil - List of apple cultivars

Apple fruit nutrition facts and health benefits - Nutrition and ...
 by Umesh Rudrappa - in 70 Google circles Delicious and crunchy apple fruit is one of the popular fruits containing an impressive list of antioxidants and essential nutrients required for good health...
apple (fruit and tree) - Encyclopedia Britannica wmw betannica com/EBcheckedtopi/ lapple * Encyclopaedia Britannica Jul 19, 2013 - Malus domestica fruit of the domesticated tree Malus domestica (family Rosaceae), one of the most widely cultivated tree fruits. The apple is a ...

Cashew-Apple Fruit Growing in the Florida Home Landscape. edia tas.ut edula377 * Institute of Food and Agricuitural Sciences * Cashew-Apple Fruit Growing in the Florida Home Landscape1. John McLaughlin, Carios Balerdi and Jonathan Crane2. Scientific Name: Anacardium ...

The Apple - "King" of the Fruit - Nature's Wonderland - Jewish... www chabad orghlods/article_cdolaid 14763/ IApples himt * Chabad History of the Apple. In honor of Tu-B'Shevat (15th day of Sh'vat), we will now talk about a fruit - a very special fruit - the Apple, "King" of all fruits (in Hebrew it is

Sneaky Green Apple Fruit Leather Recipe - Oh My Veggies
ohmyneggus comisecipe-anuaky-geen-apple-fruit mathei' * 6 hr 10 mins
Aug 15, 2013 - This sugar-free apple fruit leather recipe is hiding a secret--it's made with spinachl But it tastes so good, you'll never know it's in there.


## Apple

Fut
The apple is the porractocus fruit of the apple tree, species Malus domestica in the rose famly. It is one of the most widely cullinated tree fiuts, and tile most mindy furm of the many manbers do grnus thatus that are used by humans Wikparia

## Nutrition Facts <br> Appls =

Amount Per 1 medium ( 3 dia) ( 182 g ) -
Calories 95
s. Daily Value*
Total Fat 0.3 g 0\%

Sacurated tat $0.19 \quad 0 \%$
Polyunsaturated tat 0.1 g
Monounsaturated fat 09
Cholosterol $0 \mathrm{mg} \quad 0 \%$
Sodlum $2 \mathrm{mg} \quad 0 \%$
Potassium 195 mg 5\%
Total Carbohydrate 258 8\%
Dintary 5her $4.4 \mathrm{~g} \quad 17 \%$
Sogar 19 g

| Protain 0.5 g |  | $1 \%$ |
| :--- | :--- | :--- |
| Vitamin A | t\% Vitartin C | $14 \%$ |
| Calcum | 1\% | Iran |
| Vitamin D | 0\% Vitarain B-6 | $1 \%$ |
| Vitamin B-12 | 0\% Magnesum | $5 \%$ |

## WolframAlphà

```
apple fruit
皿 的 䱚
E Examples \(\geqslant\) Random


Indousual mettilion facts
Total nutrition facts
Apple
\begin{tabular}{|lr|}
\hline serving size 1 apple（182 g） & \\
\hline total calories 91 & fat calories 2 \\
\hline & \％daily value \\
\hline total fat 273 mg & \(0 \%\) \\
\hline saturated fat 45 mg & \(0 \%\) \\
\hline trans fat & \(0 \%\) \\
\hline cholesterol 09 & \(0 \%\) \\
\hline sodium \(910 \mu \mathrm{~g}\) & 0 \\
\hline
\end{tabular}

\section*{Wolfram Alpha？}


Take the Tour＞

Take Wolfram Alpha anywhere．


OpenCyc (Curnent); [ http://sw.opencyc.org/concept/Mx8Ngh4rxVipdpwpEbGdrcN5Y29ycB4rvVjBnZwpEbGdrcN5Y29ycA ]
OpenCyc (Versioned): [ http://sw.opencyc.org/2012/05/10/concept/Mx8Ngh4rvVipdpwpEbGdrcN5Y29ycE4rvVjBinZwpEbGdreN5Y29ycA ] OpenCyc (Readable); [http://sw.opencyc.org/2012/05/10/concept/en/Apple ]

\section*{OpenCyc Collection: apple}

Unique ID: [ Mx8Ngh4rvVipdpwpEbGdrcNSY29ycB4rvVjBnZwpEbGdrcNSY29ycA ] English ID: [ (FruitFn AppleTree) ]
English Aliases: [ "apples", "fruit of the apple", "fruit of the apple tree", "fruit of the apple trees", "fruit of the apples", "fruit of the Malus pumila" ]

The collection of individual apples

A Type of: accessory fruit, edible fruit
Instance of: type of food, type of Ife stage, type of object whose instances do not physicaly overlap
Subtypes: Adams Pearmain apple, Ambrosiy apple, Braebum apple, Gala apple, Golden Delicious apple, Granny Smith apple, Jazz apple, Mcintosh apple, red delicious apple Instances:
Same as:
\begin{tabular}{|c|}
\hline apple - RelatedTo \(\rightarrow\) fruit appie is reiated to a frutt \\
\hline apple - RelatedTo \(\rightarrow\) red fruit appic is relatedf to red fruit \\
\hline apple - Relatedie \(\rightarrow\) core apple is related to core \\
\hline \begin{tabular}{l}
apple - Antonym \(\rightarrow\) orange \\
apple is the opposite of orange
\end{tabular} \\
\hline apple - RelatedTo \(\rightarrow\) computer apple is related to a camputer \\
\hline apple - RelatedIo \(\rightarrow\) green apple is related to green. \\
\hline apple - RelatedTo \(\rightarrow\) eve apple is related to eve \\
\hline \begin{tabular}{l}
apple - RelatedTo \(\rightarrow\) mac \\
apple is related to mac
\end{tabular} \\
\hline apple - RelatedTo \(\rightarrow\) pie apple is related to pie \\
\hline apple - HasProperty \(\rightarrow\) green apple can be green \\
\hline
\end{tabular}
\begin{tabular}{|c|}
\hline apple - RelatedTo \(\rightarrow\) a opplie is related to a \\
\hline apple - Relatedio \(\rightarrow\) tree upple is related to tree \\
\hline apple - HasProperty \(\rightarrow\) red upple can be red \\
\hline apple - RelatedTo \(\rightarrow\) seed apple is related to sects \\
\hline apple - RelatedTo \(\rightarrow\) round apple is relatad to round \\
\hline apple - RelatedTo \(\rightarrow\) macintosh uppie is related to matcuntesh \\
\hline \begin{tabular}{l}
apple - Relatedle \(\rightarrow\) pear \\
uppie is related to a par
\end{tabular} \\
\hline apple - RelatedTo \(\rightarrow\) orange apple is related to oranges \\
\hline apple - ReceivesAction \(\rightarrow\) eat an apple can be cater \\
\hline
\end{tabular}

\section*{Observations from the Tour}
1. The truth is "out there"
2. Not specific to Apples
3. Famous KBs have gaping holes

What about obtaining knowledge from text?

\section*{III. Machine Reading}


Auto-Text to Knowledge

\section*{Information Extraction (IE)}
\(\mathrm{IE}(\) sentence \()=\) [tuple, confidence]
"Edison, by all accounts, was the inventor of the light bulb."
\(\rightarrow\) invented(Edison, light bulb), 0.98

Typically, IE requires:
- Pre-specified relations
- Hand-labeled training examples
- Lexicalized features/patterns

\section*{Consider Tom Mitchell's NELL System}

\section*{NELL Lexical Patterns for "haswife"}

\section*{haswife}
(reiation iverse of anteat)
See iemmedinstarces of hasate
Metadata
- antireflexive
- antisymmetric
- description
- pomainon nas a wher
domain
domainvathinfange
extractionPatterns
"apl a bad nhuence or
and atliess Angeina ar
aroz rerpl and tance



\(\qquad\)
committed adituery with
 ange apt has arnou
marned lo aryz 'arpl
 antar min ange argy altar muanayz arg the husband of argz





\author{
A bad influence on \\ A child through \\ Abusive to \\ And actress Angelina
}

Commits adultery with
隼





```

NELL Knowledge Base Browser
CNUR8sothe WeD Fmject

| catoganes | relations | Ask NELL on-demand results: |
| :---: | :---: | :---: |
| everyprometodthing <br> abstroctihing <br> croativework <br> Cbek <br> - peem <br> - lynics <br> - musicalbum <br> - masicaing <br> - teloviaiceshow <br> 4 move <br> - visualatiotm <br> species <br> - arimal <br> 2. verteleste <br> - tind <br> - fish <br> - reptile <br> - mamned <br> amphikian <br> imertebrete <br> - athreped |  | 4 possible entities found <br> Click to change visible ontity <br> - ascle inc lorgarizationd <br> - ascle (migazine) <br> - ande icompany <br> - acpla hutall <br> apple_inc (organization) <br>  Ásple Computers <br> categories <br> - company $100.0 \%$ ) <br> - SEAL B112 (100.0\%) an 03- un-2010 why? using apple <br> - CPL @ 9722 ( $100.0 \%$ ) on $06-a p r \cdot 2013$ yty 2 using apple <br> - Seed <br> - MBL @ 728 ( $99.9 \%$ ) an 29-9pr-2013 utyif using cancapt biotechcompany apple_inc |

```

\section*{NELL Knowledge Base}
categories

\section*{obama（politician）}
literal strings：OBAMA，OBama，Obama，obama
－relatedto
－visualartmoveme
－personinacadem
－generalizationof
－athletessucha
－academicfield
－chemicaltypel
－astronautssuc
－automobileen
－airportsuchas
－amphibiansuc
－aquariumssuc
－arteriessuchas
－animaltypehas －animalsuch
－animalsu
－animalsu －animalsuch －inverseofart
－professiontype
－productinstan
－booksuchasbc
－vehicletypeha
－touristattractio
28elebritysuche
－archaeasucha．

\section*{Help NELL Learn！}

NELL wants to know if these beliefs are correct． If they are or ever were，click thumbs－up．Otherwise，click thumbs－down．
－obama is a politician \(\Omega\)
－obama is a male
－obama is a person who belongs to the organization house（governmentorganization）
－obama is a person born on the date n1961（dateliteral） \(\mathcal{F}\)
－obama is a person who graduated from the university harvard（university）\＆
－obama is a person who graduated from the university harvard＿law＿school（university）

\section*{categories}
－politician（100．0\％）
－SEAL＠9（75．0\％）on 13－jan－2010［ 12 ］using obama
－CPL＠722（88．0\％）on 06－apr－2013［＂＿is a good president＂＂＿defends choice＂＂＿beats McCain＂ massive scope＂＂＿is the only President＂＂＿is the new Hitler＂＂＿is half－white＂＂＿is an empty suit＂＂ President＂＂＿is leading McCain＂＂McCain trails＿＂＂Kennedy endorsed＿＂＂＿＇s Inauguration Day＂＂ camnaian＂＂York Times endorses＂＂＇s Inaugural Address＂＂more＿mō口ular votes than＂＂＇s his

\section*{Information Extraction (2007)}

Question: can we leverage regularities in language to extract information in a relation-independent way?

\section*{Relations often:}
- anchored in verbs.
- exhibit simple syntactic form

\section*{Virtues:}
- No hand-labeled data
- "No sentence left behind"
- Exploit redundancy of Web


\section*{History of Open IE}

2007: 1.0 (TextRunner, auto-labeled examples, CRF) 2011: 2.0 (ReVerb, simple model of verb-based relations) 2012: 3.0 (Ollie, parser, verbs + nouns) 2013: 4.0 (semantic role labeling, n-ary relations) 2014: 5.0 (multiple sentences, complex structures)

©Turing
*uCenter wwCente
W人SHINGTOO
```

Argument 1: ently Apple Relation:
Argument 2:
All .
Q Seaich

```

289 answers from 2008 sentences (results truncated)

\section*{Apple}

The apple is the pomaceous frut of the apple tree, species Malus domestica in the rose family (Rosaceae). It is one of the most widely cultvated tree fruits, and the most widely known of the many members of genus Malus that are used by humans
Apples grow on small decidunuc trane Thenean Malus sieversii, is stil

> Open IE suffers from Attention Deficit Disorder!
al ingredient (37) foodT301
are tender (263)
fall to the ground (68)
are in bloom (68)
grow on Tree (58)
contain Pectin (53)
are a good source of Dietary fiber (40)
produces Apple (34)
are rich in Pectin (33)
are in season (32)
are high in Dietary fiber (30)
fall from Tree (24)
is cut in half (22)

\section*{Knowing is nat enough , We must APPLY. Willing is not enaugh . We must 0 I. \\ }

\section*{IV. The Allen Institute for Al (Al2)}

Non-profit Research Institute:
- Basic \& applied AI research
- Collaborations, internships, open source
- AI Allen Distinguished Investigator (ADI) program

\section*{Small number of High-impact projects}

\section*{Mission + Talent = Impact}

\section*{Time Line}

\section*{Al2 launched}

Jan. 2014
Team of \(30+15\) interns

Team of 50
Oct. 2014


35
AR
ALLEN INSTITUTE
for ARTIFICIAL INTELLIGENCE

\section*{Aristo}

\section*{Project Methodology:}
1. Externally-defined challenge tasks
2. Training data + unseen test data ("as is")
3. Measurable progress, clear focus

Key differences with Watson:
1. Deeper semantics \& inference
2. Open model: publish, collaborate, open source

\section*{Aristo: \(4^{\text {th }}, 5^{\text {th }}, 6^{\text {th }} \ldots\) Grade Science Tests}


THE UNIVERSITY OF THE STATE OF NEW YORK

\section*{GRADE 4}

\section*{ELEMENTARY-LEVEL}

\section*{SCIENCE TEST}

\section*{WRITTEN TEST}

JUNE 3, 2013

12 Which force causes a bicycle to slow down when the brakes are used?
A friction
B electricity
C gravity
D magnetism

\section*{Factual Knowledge for \(4^{\text {th }}\) Grade Science}

\section*{Taxonomy}
"Squirrels are animals"

\section*{Properties}
"Water freezes at 32F"

\section*{Part/whole}
"The lungs are an organ in the body"

\section*{Actions + States}
"Brushing our teeth removes the food and helps keep them strong"

\section*{Language}

Paraphrases;
active/passive
transformations;
apositives;
coreference; idioms; ..

\section*{Processes}
"Photosynthesis is a process by which plants make their own food and give off oxygen and water that they are not using."

\section*{Summary of our Approach}
1. Knowledge + tractable reasoning is necessary
2. Knowledge acquisition has to be highly automated
3. Large bodies of high-quality knowledge can be acquired from text ("machine reading") images, and more.
4. Aristo is tested on unseen, standardized, multiple choice tests in science, arithmetic, and geometry


\section*{Al challenges everywhere...}

\section*{Information Overload}

\section*{THE WEB OF SCHOLARSHIP}

Around 114 million English-language scholarly documents (including papers, books and technical reports) can be found on the web.

"information extraction"
\begin{tabular}{|c|c|}
\hline Articles & \begin{tabular}{l}
[PDF] Maximum Entropy Markov Models for Information Extraction and Segmentation. \\
A McCallum, D Freitag. FCN Pereira - ICML, 2000 - courses ischool.berkeley.edu
\end{tabular} \\
\hline Case law & Page 1.1 Maximum Entropy Markov Models for Information Extraction and Segmentation Andrew \\
\hline & McCallum, Dayne Freitag, and Fernando Pereira ... Named entity recognition: <ORG> \\
\hline My library & Vice President <PRS>John Hime</PRS> - Information extraction: ... \\
\hline & Cited by 1126 Related articles All 50 versions Cite Save More \\
\hline Any time & Incorporating non-local information into information extraction system \\
\hline Since 2014 & JR Finkel, T Grenager, C Manning - ... of the 43rd Annual Meeting on .... 2005 - dl. acm.org \\
\hline Since 2013 & Abstract Most current statistical natural language processing models use only local features \\
\hline Since 2010 & nic programming in inference, but this makes them unable to fully \\
\hline Custom range... & Cited by 1129 Related articles All 23 versions Cite Save \\
\hline by rel & [PDF] Learning dictionaries for information extraction by multi-level bootstrapping \\
\hline Sort by date & E Riloff, R Jones - AAAI/IAAI, 1999 - aaai.org \\
\hline Sort by date & Abstract Information extraction systems usually require two dictionaries: a semantic lexicon \\
\hline & algorithm that generates both the semantic lexicon and extraction patterns simultaneously ... \\
\hline \begin{tabular}{l}
\(\checkmark\) include patents \\
\(\checkmark\) include citations
\end{tabular} & Cited by 753 Related articles All 18 versions Cite Save More \\
\hline & [PDF] Open information extraction for the web \\
\hline & M Banko, MJ Cafarella, S Soderland, M Broadhead... - IJCAI, 2007 - aaai.org \\
\hline \(\checkmark\) Create alert & Abstract Traditionally, Information Extraction (IE) has focused on satisfying precise, narrow, pre-specified requests from small homogeneous corpora (eg, extract the location and time of seminars from a set of announcements). Shifting to a new domain requires the user to ... Cited by 806 Related articles All 31 versions Cite Save More \\
\hline
\end{tabular}

\section*{Leverage Al to combat information overload}


\title{
Cut through the clutter.
}

Home in on key papers, citations, and results.

Q Find it fast

Try: Open information extraction POS tagging Dependency parsing

```

FILTER RESULTS

```
6,929 results
Showing 1-10
1
45
"
Sort by:
Relevance

\title{
On-Demand Information Extraction
}

Satoshi Sekine ACL 2006
Cited by 22 Abstract View PDF/Add to reading list
At present, adapting an Information Extraction system to new topics is an expensive and slowfor each new topic. We propose a new paradigm of Information Extraction which operates 'on demand
Probabilistic Coreference In Information Extraction
Andrew Kehler Conference On Empirical Methods In Natural Language Processing / 1997
Cited by 17 Abstract / View PDF/Add to reading list
Certain applications require that the output of an information extraction system be probabilisticExtraction Abstract Certain applications require that the output of an information...
Multidocument Summarization Via Information Extraction
Michael White, Tanya Korelsky, Claire Cardie, Vincent Ng, David R. Pierce, Kiri Wagstaff / HLT / 2001 Cited by 11 Abstract / View PDF / Add to reading list

    We present and evaluate the initial version of RIPTIDES, a system that combines information

    information extraction, extraction-based summarization, and natural language generation to...
```

Confidence Estimation For Information Extraction
Aron Culotta, Andrew McCallum NAACL }200
Clted by 19 / Abstract / View PDF / Add to reading list

```



Incorporating Non-Local Information Into Information Extraction Systems By Gibbs Sampling
Jenny Rose Finkel, Trond Grenager, Christopher D. Manning / ACL / 2005
Cited by 166 / Abstract View PDF / In reading list / Software
use this technique to augment an existing CRF-based information extraction system with long-distance
inference. We use this technique to augment an existing CRF-based information extraction system with...

\section*{Semantic Role Labeling for Open Information Extraction}

Janara Christensen, undefined Mausam, Stephen Soderland, Oren Etzioni / Proceedings of the NAACL HLT 2010 First International Workshop on Formalisms and Methodology for Learning by Reading / 2010

Cited by 2 Abstract View PDF / Add to reading list/Software
Open Information Extraction is a recent paradigm for machine reading from arbitrary text. In
Abstract Open Information Extraction is a recent paradigm for machine reading from arbitrary text

Multi-Field Information Extraction And Cross-Document Fusion
Gideon S. Mann, David Yarowsky / ACL / 2005
Cited by 9 / Abstract / View PDF / Add to reading list / Software
examples of fact extractions and train Rote, \(\mathrm{Na}{ }^{"}\) Ive Bayes, and Conditional Random Field extraction
Introduction Much recent statistical information extraction research has applied graphical models to...

An Entity-Level Approach to Information Extraction
Aria Haghighi, Dan Klein / ACL / 2010


\section*{216 results}

Showing 1-10
23 \(\qquad\) \(»\) Sort by: Relevance

\section*{Incorporating Non-Local Information Into Information Extraction Systems By Gibbs Sampling}
```

Scholar Rank:

```

Citation rate:
63.8
ince
with..

\section*{TOPICS}

CRF model Conditional Markov Models Monte Carlo algorithm Conditional Random Fields Viterbi algorithm

CITATION CONTEXTS

In this paper, Stanford Named Entity Recognizer (Finkel et al. 2005) is used to classify noun phrases into four semantic categories: PERSON, LOCATION, ORGANIZARION and MISC.

We utilized an offthe-shelf system, Stanford Named Entity Recognizer 4 (Finkel et al., 2005) for detecting entity mentions on the English sentences.

View more citation contexts

Cited by 9 / Abstract View PDF/Add to reading list / Software
examples of fact extractions and train Rote, \(\mathrm{Na}^{-}\)Ive Bayes, and Conditional Random Field extraction Introduction Much recent statistical information extraction research has applied graphical models to..

An Entity-Level Approach to Information Extraction
Aria Haghighi, Dan Klein ACL 2010

Reading list
```

Incorporating Non-
Local Information Into
Information
Extraction Systems By
Gibbs Sampling
O Abstract
7. View PDF
\ In Reading List
O. Software
Scholar Rank:
Top 2%
Citation rate>

## = Citing papers

ㅍ: Cluster map

## Cited by 166 papers 7 are important (4.2\%)

Sort by: Importance *

## IMPORTANT

The Web is not a PERSON, Berners-Lee is not an ORGANIZATION, and African-Americans are not LOCATIONS: An Analysis of the Performance of Named-Entity Recognition
Robert Krovetz, Paul Deane, Nitin Madnani / Proceedings of the Workshop on Multiword Expressions: from Parsing and Generation to the Real World / 2011
Abstract / View PDF / Add to reading list
Citation contexts:
"The Stanford Tagger is based on Conditional Random Fields (Finkel et al., 2005)."
"Although Named Entity Recognition is reported in the literature to have an accuracy rate of 85-95\% (Finkel et al., 2005; Ratinov and Roth, 2009), it was clear by inspection that both the Stanford and the LBJ tagger made a number of mistakes."

## IMPORTANT

Guessing Parts-Of-Speech Of Unknown Words Using Global Information
Tetsuji Nakagawa, Yuji Matsumoto / ACL / 2006
Cited by 6 / Abstract / View PDF / Add to reading list
Citation contexts:
"Global information is known to be useful in other NLP tasks, especially in the named entity recognition task, and several studies successfully used global features (Chieu and Ng , 2002; Finkel et al., 2005)."
"Finkel et al. (2005) used simulated annealing with Gibbs sampling to find a solution in a similar situation."

## TOPICS

Relaxation
Markov random fields

Systems based on perceptron

POS tagging
Dual Decomposition
Maximum Entropy Markov
Models
JNLPBA

F-Score
L1 regularization
CRF models
Entity Recognition
NER system
Conditional Markov Models
Gibbs sampling

Information extraction
POS tags of unknown words
Global Information
AIL
ALLEN INSIIUIE for ARTIFICIAL INTELLIGENCE

## Cut through the clutter.

Home in on key papers, citations, and results.

Q information extraction

Ill information extraction progress summary
...) information extraction using gibbs sampling
146 papers, 41.7 agerage rank
-..) information extraction using dynamic programming 598 papers, 41.7 agerage rank
...) information extraction using maximum entropy model 1,086 papers, 32.4 average rank
-..) information extraction using self-organizing map 36 papers, 30.24 average rank
...) information extraction using conditional random field 358 papers, 24.77 average rank
...) information extraction using preprocessing 1,148 papers, 23.31 average rank

## 1ll Information Extraction Progresssummary

| Accurbey |  |  | 654 | 6754 | 7004 | 15\% <br> Gibbe Sampling: |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 61\% |  |  |  |  |
|  | 58\% |  |  | Maymum Entropy Model | Dynamic Programming |  |
|  | Conditional Random Field | $\begin{aligned} & \text { Selforganizing } \\ & \text { Mop } \end{aligned}$ | Uaticy Function |  |  |  |
|  | 1997 | 2000 | 2003 | 2006 | 2010 | 2012 |




## Cut through the clutter.

Home in on key papers, citations, and results.

Q gibbs sampling
gibbs sampling overview of applications
-.) gibbs sampling in information extraction
162 papers, 41.7 agerage rank
$\rightarrow$ gibbs sampling in dependency parsing
105 papers, 18.71 agerage rank
-.) gibbs sampling in parsing
159 papers, 32.4 average rank
-.) gibbs sampling in machine translation
163 papers, 30.24 average rank
-.) gibbs sampling in POS tagging
87 papers, 23.31 average rank

Reading List
Sign Out


```
HITERHESULTS
CLASSIFICATION
E summo
A Experimental
- Theoretical
<} Software
```

YEAR
yyyy to yyy
venues (15)
ACL
Proceedings of the NAACL
HLT 2010 Workshop on
Creating Speech and
Creating Speech
EMNLP
CONLL
NAACL

- View all venues

TOPICS (213)
Stanford NER Tagger
Stanford Named Entity Recognizer

Krishnan And Manning
Non-Local features

## 429 results

Sart by: Relevance *
| Incorporating Non-Local Information Into Information Extraction | Systems By Gibbs Sampling
Jenny Rose Finket, Trond Grenager, Chirstopher D. Manning ACL 2012
Cited by 166 Abstract View PDF Add to reading list
structure that is prevalent in language use. We show how to solve this dilemma with Gibbs sampling information extraction task. We show 10 runs of Gibbs sampling in the same CRF..-
| Not-So-Latent Dirichlet Allocation: Collapsed Gibbs Sampling Using Human Judgments Jonathan Charg Proceedings of tho NAACL HLT 2010 Workshop on Creating Speech - 2010 Ctied by 1 Abstract View PDF Add to reading list
Probabilistic topic models are a popular tool for the unsupervised anslysis of text, providing both and cluster that annotation, This task simulates the sampling step of the collapsed Gibbs sampler

Sampling Alignment Structure under a Bayesian Translation Model John DeNero, Alexandre Bouchard-Cote, Don Klein EMNNP 2008
Cited by 31 Abstract View POF: Addd to reading lat
We describe the firsx tractable Gibbs sampling procedure for estimating phrase pair frequencies Abstract We describe the first tractable Gibbs sampling procedure for estimating phrase pair
| A Gibbs Sampler for Phrasal Synchronous Grammar Induction Philip Blunsom Trevor Cohn Chris Dyer, MEes Osborne ACL. 2009
Cited by 34 Abstract View POF Add to reading lest
a novel Gibbs sampler over synchronous derivations. This sampler side-steps the intractability scale experiments and the short sampling runs. In this work we also propose a Gibbs sampler but...

Reading List 4 Syn Out


## VENUES (15)

ACL
Proceedings of the NAACL. HLT 2010 Workshop on Creating 5 peech and Language Data

EMNLP
CONLL
NAACL

- View ala vencies

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a novel Gibbs sampler over synchrorious derivations. This sampler side-steps the intractability scale experiments and the short sampling runs. in this work we also propose a Gibbs tampler but...

## Summary

What's Next? (after "big data" \& "deep learning") "You can't play 20 questions with Nature and Win!"

- Focus on data/text Mining systems yielding:
- Knowledge
- Reasoning
- Explanation



## Parting Thought

"The best minds of my generation are thinking about how to make people click ads...
that sucks."


Jeff Hammerbacher

## Our focus: Al for the Common Good

## Thank you from Al2!



## Geometry Domain (Case Study)

Combines vision, NLP, and simple semantics Data set: $1009^{\text {th }}$ grade geometry problems AAAI '14 (Seo, Hajishirzi, Farhadi, Etzioni)

In the diagram, AB intersects circle 0 at D, AC intersects circle 0 at $E, A E=$ 4, $A C=24$, and $A B=$ 16. Find AD.


## Geometry Diagram Alone



## Geometry Text Alone

In the diagram, $A B$ intersects circle 0 at D, AC intersects circle 0 at $E, A E=$ 4, $A C=24$, and $A B=$ 16. Find AD.

## Need Both Diagram and Text

In the diagram, $A B$ intersects circle 0 at D, AC intersects circle 0 at $E, A E=$ 4, $A C=24$, and $A B=$ 16. Find AD.


## Diagram Understanding

## Problem:

1. Identify visual elements
2. Al

## Diagram understanding $\rightarrow$ submodular optimization

In the uring intersects Cl . $\quad-$ secant AC intersects circle 0 at $E$, $A E=4$, $A C=24$, and $A B=16$. Find $A D$.


## Experimental Study



- G-Aligner
$\square$ No C No visual coherence
- No $S$ No textual alignment

■ No S,C

- Baseline


[^0]:    .
    

