

# **Knowledge Harvesting from Web Sources**

Part 1: Knowledge Bases and their Automatic Construction

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# **Acknowledgements**



















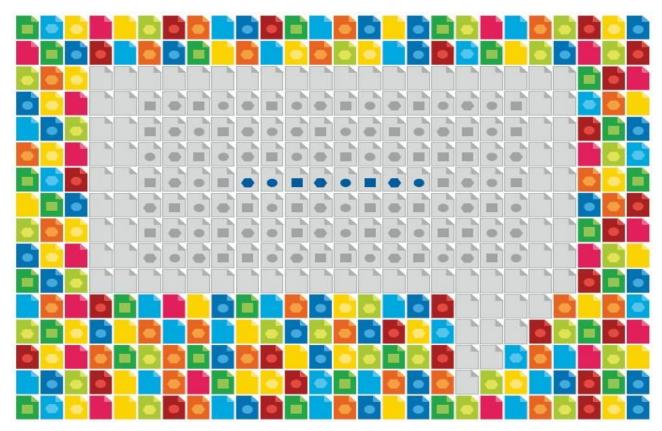








# Goal: Turn Web into Knowledge Base



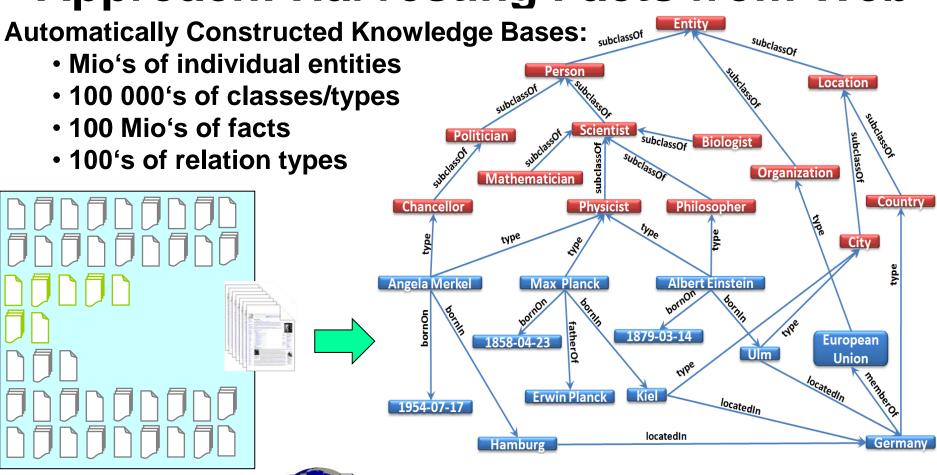
Source: DB & IR methods for knowledge discovery. Communications of the ACM 52(4), 2009

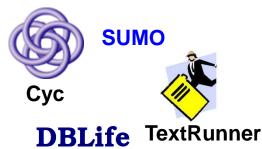
### comprehensive DB of human knowledge

- everything that Wikipedia knows
- everything machine-readable
- capturing entities, classes, relationships



# **Approach: Harvesting Facts from Web**

























# Knowledge for Intelligence

- entity recognition & disambiguation
- understanding natural language & speech
- knowledge services & reasoning for semantic apps (e.g. deep QA)
- semantic search: precise answers to advanced queries (by scientists, students, journalists, analysts, etc.)
- Swedish king's wife when Greta Garbo died?
- ★ FIFA 2010 finalists who played in a Champions League final?
- Politicians who are also scientists?
- Relationships between Max Planck, Angela Merkel, Jim Gray, and the Dalai Lama?
- Enzymes that inhibit HIV?
  Influenza drugs for teens with high blood pressure?





russian composers

Square it

canadian actors maine lighthouses

dog breeds

8000 meter peaks

hotels in chicago

european union countries

canadian prime ministers

tennis champions bars in pacific heights

antibiotics

blueberry varieties

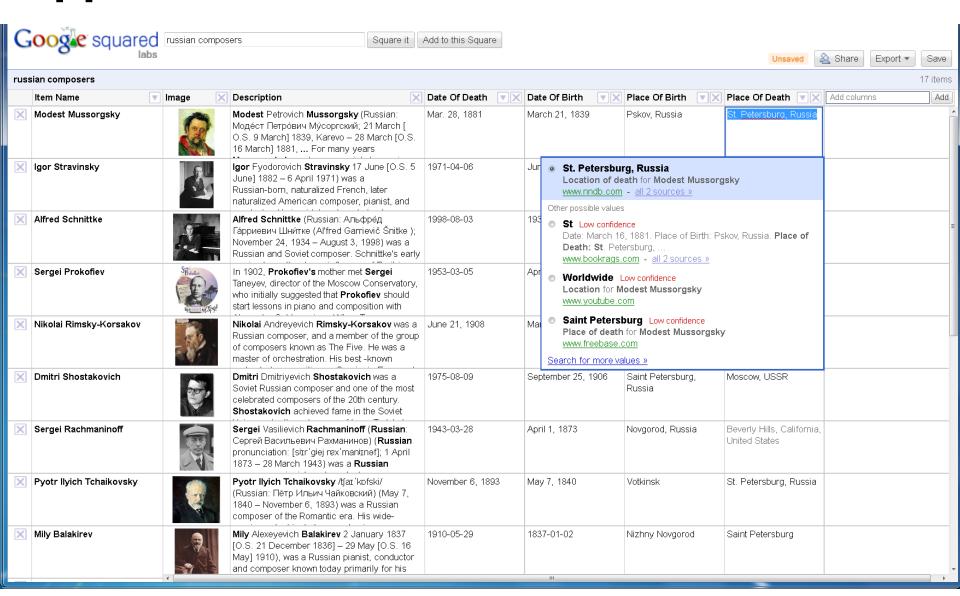
italian provinces

Start with an empty Square

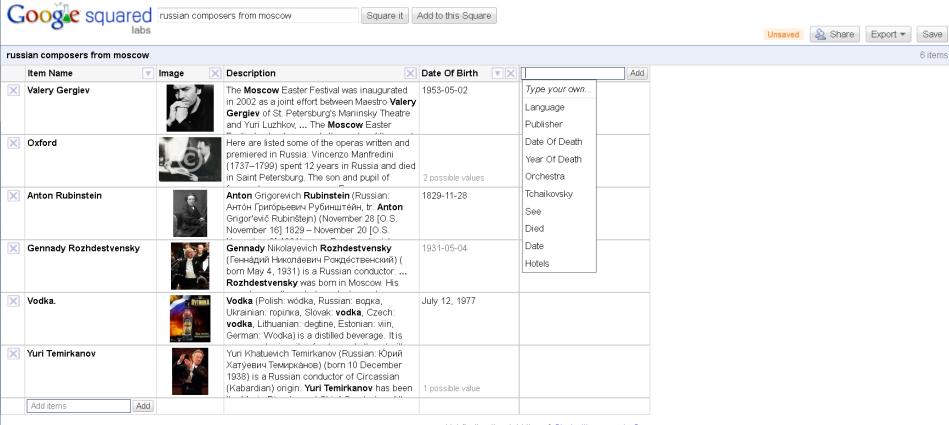
Google Home

© 2011 - Privacy









Not finding the right items? Start with an empty Square.





Square it Add to this Square

### Did you mean: drugs for treating Alzheimer's

Item Name     Image   Description     Cas Number     Formula     Half Life     Pubchen   Variation	bid you mean. drugs for treating Alzheimer s						
Tacrine is the first FDA approved drug for the treatment of Alzheimer's disease as safe and effective. The fear of toxicity has been exaggerated. Liver function testing could be  The Food and Drug Administration (FDA) has deciled to approve Memantine (namenda ) to treat mild Alzheimer's Olanzapine (Zyprexa) Olanzapine (Zyp	drugs for treating Alzheimer						
treatment of Alzheimer's disease as safe and effective. The fear of toxicity has been exaggerated. Liver function testing could be    X	▼×						
✓ Olanzapine       The Food and Drug Administration (FDA) has declied to approve Memantine (namenda) to treat mild Alzheimer's Olanzapine (Zyprexa) Olanzapine (Zyprexa). Atypical Antipsychotic       132539-06-1       C17H20N4S       21–54 hours       CID 4586         ✓ Acetylcholine       Treating the symptoms of Alzheimer's can provide patients with comfort, dignity, and independence for a longer period of time and can encourage Medications called       51-84-3       C7H16NO2       approximately 2 minutes         ✓ Phosphatidylserine       Phosphatidylserine might increase a chemical in the body called acetylcholine. Medications for Alzheimer's disease called acetylcholinesterase inhibitors also increase       8002-43-5       C13H24NO10P       445141         ✓ Reminyl       What should I avoid while taking Reminyl (galantamine)? Galantamine can cause side effects that may impair your thinking or       357-70-0       C17H21NO3       7 hours							
Phosphatidylserine  Phosphatidylserine  Phosphatidylserine might increase a chemical in the body called acetylcholine. Medications for Alzheimer's disease called acetylcholinesterase inhibitors also increase  What should I avoid while taking Reminyl (galantamine)? Galantamine can cause side effects that may impair your thinking or							
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Reminyl  What should I avoid while taking Reminyl (galantamine)? Galantamine can cause side effects that may impair your thinking or							
Teactions Reminy ( galantamine) side	value						
Vitamin E  While current medications cannot stop the damage Alzheimer's causes to brain cells, they may help lessen or stabilize symptoms for a limited time by affecting certain chemicals  C 29 H 50 O 2							
Exelon will not be available in generic form until Novartis' patent expires in 2014. Sources: About Exelon for mild to moderate Alzheimer's dementia. Novartis Pharmaceuticals. 2008.  1.5 hours 77991							
Aricept  It's important to remember that while ARICEPT treats the symptoms of Alzheimer's disease, it is not a cure Before starting on ARICEPT 23 mg/day, patients should be on ARICEPT 10  The symptoms of Alzheimer's disease, it is not a cure Before starting on ARICEPT 10  Www.google.com/squa	<u>'ed/</u>						

# **Application 2: Deep QA in NL**

William Wilkinson's "An Account of the Principalities of Wallachia and Moldavia" inspired this author's most famous novel

This town is known as "Sin City" & its downtown is "Glitter Gulch"

As of 2010, this is the only former Yugoslav republic in the EU

99 cents got me a 4-pack of Ytterlig coasters from this Swedish chain

question classification & decomposition



knowledge back-ends

D. Ferrucci et al.: Building Watson: An Overview of DeepQA Project. Al Magazine, Fall 2010.















# **Application 3: Machine Reading**

It's about the disappearance forty years ago of Harriet Vanger, a young scion of one of the wealthiest families in Sweden and about her uncle, determined to know the truth about what he believes was her murder.

THE GIRL WITH THE DRAGON TATTOO
THE INTERNATIONAL BEST-SELLING NOVEL STIEG LARSSON

Blomkvist visits Henrik Vanger at same to on the same and of Hedeby.

The old man same pend a year writing the Vanger family history as a cover for the real assignment: the disappearance of Vowns niece Harriet some 40 years earlier. Hedeby is home to several generations of Vangers, all part owners in Vanger Enterprises. Blomkvist beco uncleOf inted with the men hires the extended Vanger family, most of whom resent his presence. He does, however, start a short lived affair with Cecilia, the niece of the enemyOf Af Same overing that Salander has backed into his confidential assist him with research. They even affairWith a lovers, but Blomkvist has trouple getting close to Lisbeth who creats virtually everyone she meets with hostility. Ultimately the two

A 24-year-old computer hacker sporting an assertment of tattoos and body piercings supports lerself by doing deep backgrou headOf gations for Dragan Armansky, who, in turns ies that Lisbeth Salander is "the perfect victim for anyone who wished her ill."

discover that Harriet's brother Martin, CEO of Vanger Industries as secretly a serial killer.

O. Etzioni, M. Banko, M.J. Cafarella: Machine Reading, AAAI ,06
T. Mitchell et al.: Populating the Semantic Web by Macro-Reading Internet Text, ISWC'09

### **Outline**

✓ Motivation

- **★** Machine Knowledge
- Knowledge Harvesting
  - Entities and Classes
  - Relational Facts
- Research Challenges
  - Open-Domain Extraction
  - Temporal Knowledge
- ★ Wrap-up



# Spectrum of Machine Knowledge (1)

### factual:

bornIn (GretaGarbo, Stockholm), hasWon (GretaGarbo, AcademyAward), playedRole (GretaGarbo, MataHari), livedIn (GretaGarbo, Klosters)

### taxonomic (ontology):

instanceOf (GretaGarbo, actress), subclassOf (actress, artist)

### lexical (terminology):

means ("Big Apple", NewYorkCity), means ("Apple", AppleComputerCorp) means ("MS", Microsoft), means ("MS", MultipleSclerosis)

### multi-lingual:

meansInChinese ("乔戈里峰", K2), meansInUrdu ("迨", K2) meansInFrench ("école", school (institution)), meansInFrench ("banc", school (of fish))



# Spectrum of Machine Knowledge (2)

### ephemeral (dynamic services):

wsdl:getSongs (musician ?x, song ?y), wsdl:getWeather (city?x, temp ?y)

### common-sense (properties):

```
hasAbility (Fish, swim), hasAbility (Human, write), hasShape (Apple, round), hasProperty (Apple, juicy), hasMaxHeight (Human, 2.5 m)
```

### common-sense (rules):

```
\forall x: human(x) \Rightarrow male(x) \vee female(x) \forall x: (male(x) \Rightarrow \neg female(x)) \wedge (female(x)) \Rightarrow \neg male(x))
```

 $\forall$  x: animal(x)  $\Rightarrow$  (hasLegs(x)  $\Rightarrow$  isEven(numberOfLegs(x))

### temporal (fluents):

hasWon (GretaGarbo, AcademyAward)@1955 marriedTo (AlbertEinstein, MilevaMaric)@[6-Jan-1903, 14-Feb-1919]



# Spectrum of Machine Knowledge (3)

### free-form (open IE):

hasWon (NataliePortman, AcademyAward)

occurs ("Natalie Portman", "celebrated for", "Oscar Award")

occurs ("Jeff Bridges", "nominated for", "Oscar")

### multimodal (photos, videos):

StuartRussell
JamesBruceFalls







### social (opinions):

admires (maleTeen, LadyGaga), supports (AngelaMerkel, HelpForGreece)

### epistemic ((un-)trusted beliefs):

believe(Ptolemy,hasCenter(world,earth)),

believe(Copernicus, has Center(world, sun))

believe (peopleFromTexas, bornIn(BarackObama,Kenya))



# **Knowledge Representation**

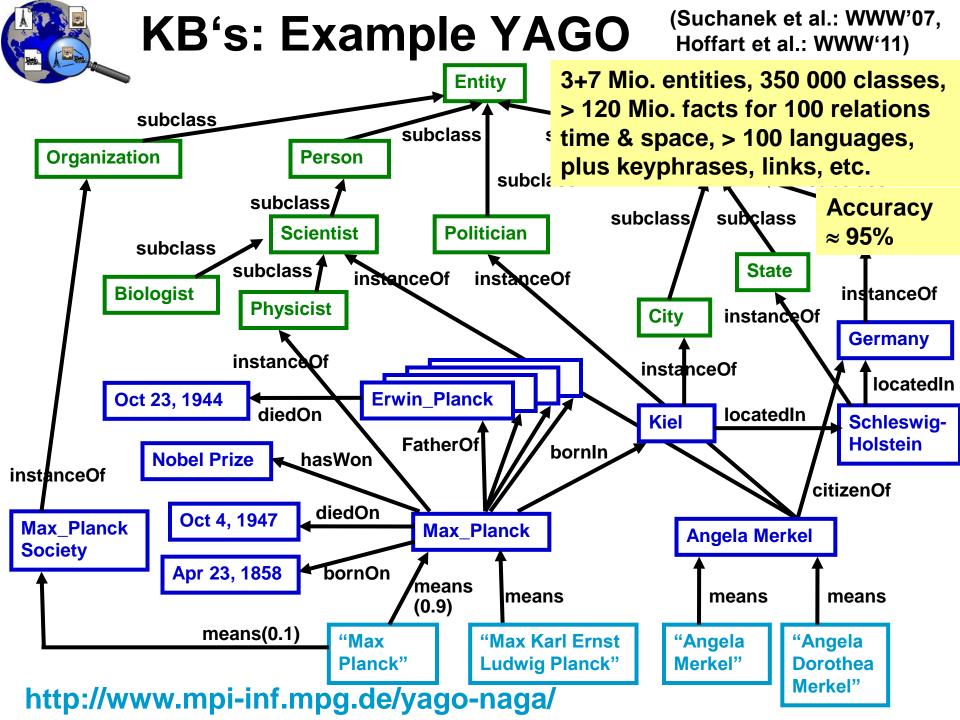
- RDF (Resource Description Framework, W3C): subject-property-object (SPO) triples, binary relations structure, but no (prescriptive) schema
- Relations, frames
- Description logics: OWL, DL-lite
- Higher-order logics, epistemic logics

# facts (RDF triples):facts about facts:1: (JimGray, hasAdvisor, MikeHarrison)5: (1, inYear, 1968)2: (SurajitChaudhuri, hasAdvisor, JeffUllman)6: (2, inYear, 2006)3: (Madonna, marriedTo, GuyRitchie)7: (3, validFrom, 22-Dec-2000)4: (NicolasSarkozy, marriedTo, CarlaBruni)8: (3, validUntil, Nov-2008)9: (4, validFrom, 2-Feb-2008)10: (2, source, SigmodRecord)

# temporal & provenance annotations can refer to reified facts via fact identifiers

(approx. equiv. to RDF quadruples: "Color" × Sub × Prop × Obj)

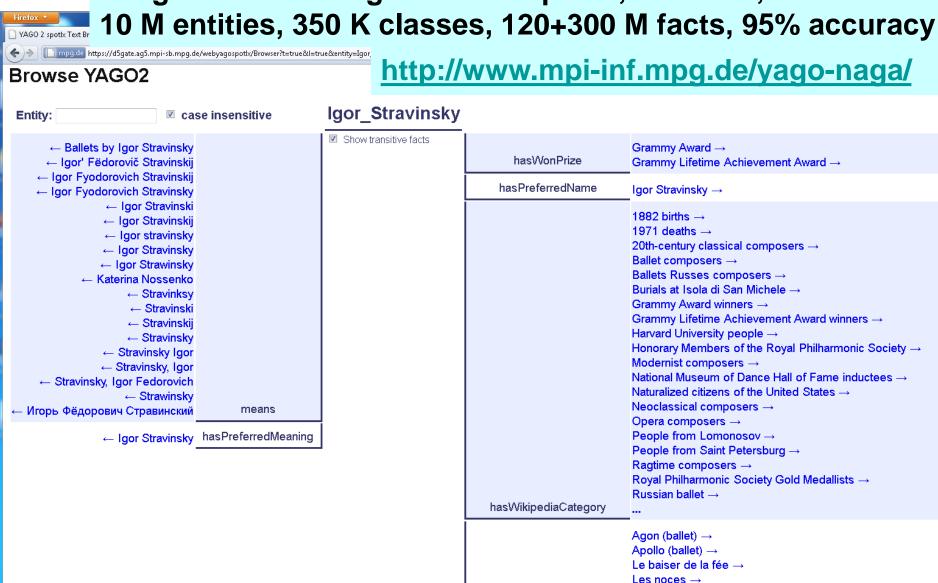






### YAGO2 Knowledge Base (Nov 2010)

integrates knowledge from Wikipedia, WordNet, Geonames:



Petrushka (ballet) → Pulcinella (ballet) → Scènes de hallet (Ashton) -

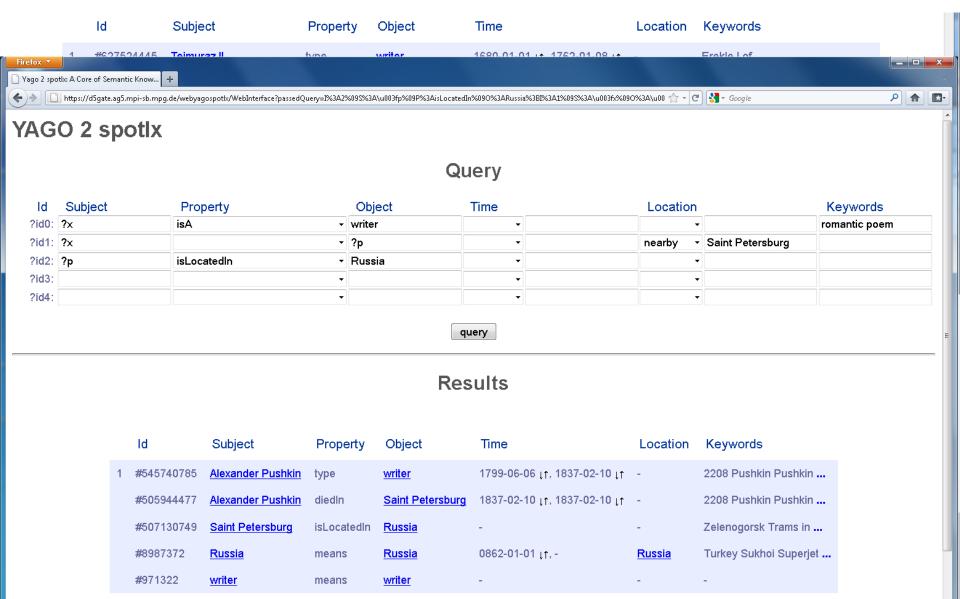


### YAGO2 Knowledge Base (Nov 2010)

integrates knowledge from Wikipedia, WordNet, Geonames: 10 M entities, 350 K classes, 120+300 M facts, 95% accuracy

rue&l=true&entity=Igor_Stravinsky		http://www.mpi-inf.mpg.de/yago-naga/		
	type	20th-century classical composers → composer → musician → artist → creator → person → Ballet composers → composer → musician → artist → creator → person → Ballets Russes composers → composer → musician → artist → creator → person → Harvard University people → person → Modernist composers → composer → musician → artist → creator → person → Naturalized citizens of the United States → citizen → national → person → Neoclassical composers → composer → musician → artist → creator → person → Opera composers → composer → musician → artist → creator → person → People from Lomonosov → person → People from Saint Petersburg → person → Ragtime composers → composer → musician → artist → creator → person → Royal Philharmonic Society Gold Medallists → medalist → winner → contestant → person → Russian composers → composer → musician → artist → creator → person → Soviet immigrants to the United States → immigrant → migrant → traveler → person → artist → creator → person → causal agent → physical entity → entity → citizen → national → person → composer → musician → artist → creator → person → contestant → person → creator → person →	^	
	influences	Carla Lucero → Jono El Grande → Nicholas Lens →		
	hasGivenName	lgor →		
		1913 premiere of → Aaron Copland → Aeolian Company →	Ш	

# Knowledge Querying in Space, Time, Context



Chad

http://www.mpi-inf.mpg.de/yago-naga/

# KB's: Example DBpedia

(Auer, Bizer, et al.: ISWC'07)

About: Igor Stravinsky

An Entity of Type: Neoclassical composers, from Named Graph: http://dbpedia.org, within Data Space:

dbpedia.org



Igor Fyodorovich Stravinsky (17 June 1882 – 6 April 1971) was a Russian composer most important and influential composers of 20th century music. He was a quintesse one of the 100 most influential people of the century. He became a naturalized US ci

•	3.5	Mi	<b>io.</b>	enti	ities,
					,

- 700 Mio. facts (RDF triples)
- 1.5 Mio. entities mapped to

Property	Value	hand-crafted taxonomy of
dbpedia-owl:birthDate	■ 1882-06-17 (xsd:date)	259 classes with 1200 properties
dbpedia-owl:birthPlace	<ul><li>dbpedia:Lomonosov,_Russia</li><li>dbpedia:Russia</li></ul>	• interlinked with Freebase, Yago,
dbpedia-owl:deathDate	■ 1971-04-06 (xsd:date)	
dbpedia-owl:deathPlace	<ul><li>dbpedia:New_York_City</li><li>dbpedia:New_York</li><li>dbpedia:United_States</li></ul>	
dbpedia-owl:thumbnail	<ul><li>http://upload.wikimedia.org/wikipedia/e</li></ul>	n/thumb/3/33/lgor_Stravinsky_LOC_32392u.jpg/200px-lgor_Stravinsky_
dbpprop:alternativeNames	<ul> <li>Stravinskij, Igor Fëdorovič</li> </ul>	
dbpprop:dateOfBirth	■06-17	
dbpprop:dateOfDeath	■ 1971-04-06 (xsd:date)	
dbpprop:filename	<ul> <li>Igor Stravinsky - 3 Pieces for Clarinet.</li> </ul>	Alone.ogg
dbpprop:format	<ul><li>dbpedia:Ogg</li></ul>	
dbpprop:hasPhotoCollection	<ul><li>http://www4.wiwiss.fu-berlin.de/flickrwra</li></ul>	appr/photos/lgor_Stravinsky
dbpprop:name	<ul> <li>Stavinsky, Igor Fyodorovich</li> </ul>	
dbpprop:placeOfBirth	<ul> <li>Lomonosov, Russia, Russia</li> </ul>	
dbpprop:placeOfDeath	<ul> <li>New York City, New York, United States</li> </ul>	
dbpprop:shortDescription	<ul><li>Russian composer</li></ul>	



# **KB's: Example DBpedia**

(Auer, Bizer, et al.: ISWC'07)

### About: Igor Stravinsky

An Entity of Type: Neoclassical composers, from Named Graph: http://dbpedia.org, within Data Space:



dbpedia.org
dcterms:subject

rdf:type

- category:Honorary\_Members\_of\_the\_Royal\_Philharmonic\_Society
- category:People from Saint Petersburg
- category:Russian Orthodox Christians
- category:Grammy\_Award\_winners
- category:Royal\_Philharmonic\_Society\_Gold\_Medallists
- category:People from Lomonosov
- category:Opera\_composers
- category:Burials\_at\_Isola\_di\_San\_Michele
- category:Russian composers
- category:Ragtime composers
- category:National\_Museum\_of\_Dance\_Hall\_of\_Fame\_inductees
- category:Naturalized citizens of the United States
- category:Soviet immigrants to the United States
- category: 20th-century classical composers
- category:Neoclassical composers
- category:Russian\_ballet
- category:Ballets\_Russes\_composers
- category:1882\_births
- category:1971\_deaths
- category:Harvard University people
- category:Grammy\_Lifetime\_Achievement\_Award\_winners
- category:Modernist\_composers
- category:Ballet\_composers
- foaf:Person
  - yago:ModernistComposers
  - yago:RussianComposers
  - yago:BalletComposers
  - yago:BalletsRussesComposers
  - yago:Person100007846
  - yago: OperaComposers
  - yago:RagtimeComposers
  - yago:HarvardUniversityPeople
  - vago:NaturalizedCitizensOfTheUnitedStates
  - yago:PeopleFromSaintPetersburg
  - vago:RovalPhilharmonicSocietyGoldMedallists

# **KB's: Example DBpedia**

(Auer, Bizer, et al.: ISWC'07)

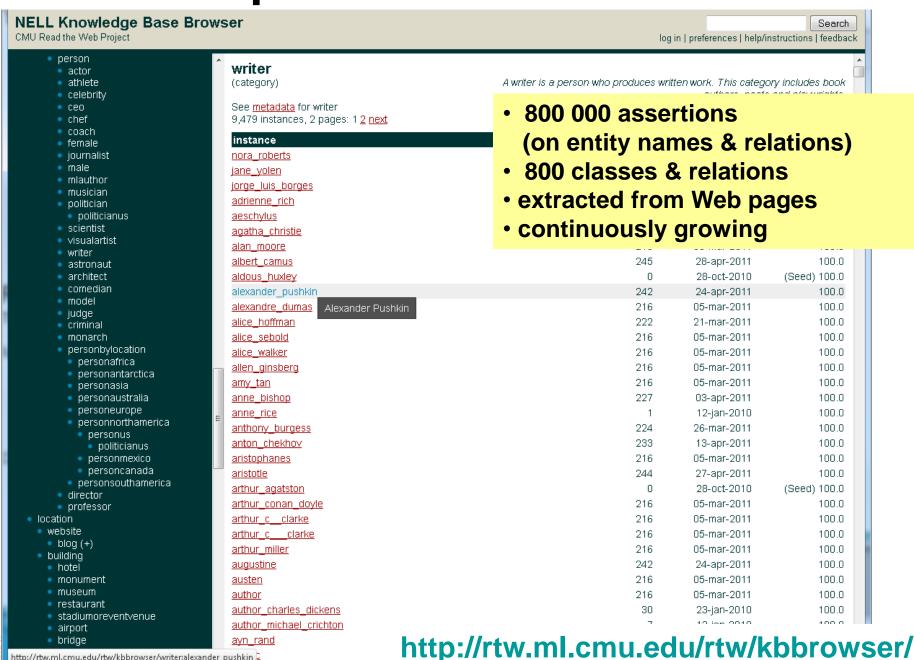
### About: Igor Stravinsky

dbpedia:lgor\_Fyodorovitch\_Stravinsky dbnedia:Stravinsky, Idor Fedorovich

280

About, <u>igoi Stravirisk</u>	<u>·Y</u>	00000
	mposers, from Named Graph: <u>http://dbpedia.org</u> , within Data Space:	<b>DBpedia</b>
dbpedia.org		
rdfs:label	<ul> <li>Igor Stravinsky</li> <li>Ígor Stravinsky</li> <li>Igor Stravinsky</li> <li>Igor Stravinski</li> <li>Igor' Fëdorovič Stravinskij</li> <li>イーゴリ・ストラヴィンスキー</li> <li>Igor Stravinski</li> <li>Igor Stravinski</li> <li>Igor Stravinski</li> <li>「gor Stravinski</li> <li>「gor Stravinski</li> <li>「дог Stravinski</li> <li>Стравинский, Игорь Фёдорович</li> <li>Igor Stravinski</li> <li>日 伊戈尔·费奥多罗维奇·斯特拉文斯基</li> </ul>	
owl:sameAs	<ul> <li>freebase:Igor Stravinsky</li> <li>http://zitgist.com/music/artist/c278de2c-9696-4fdf-a919-0781cd945e2c</li> <li>http://data.nytimes.com/N1686270504070431933</li> </ul>	
foaf:depiction	<ul><li>http://upload.wikimedia.org/wikipedia/en/3/33/lgor_Stravinsky_LOC_32392u.jpg</li></ul>	
foaf:givenName	■ Igor Fyodorovich	
foaf:name	■ Igor Fyodorovich Stavinsky	
foaf:page	■ http://en.wikipedia.org/wiki/lgor_Stravinsky	
foaf:surname	Stavinsky	
is dbpedia-owl:influencedBy of	<ul><li>dbpedia:Nicholas_Lens</li><li>dbpedia:Carla_Lucero</li></ul>	
is dbpedia-owl:musicComposer of	<ul> <li>dbpedia:La_Belle_Noiseuse</li> <li>dbpedia:Melinda_and_Melinda</li> <li>dbpedia:Histoire(s)_du_cinéma</li> </ul>	
is dbpedia-owl:wikiPageDisambiguates (	of dbpedia:lgor ■ dbpedia:Stravinsky_(disambiguation)	
is dbpedia-owl:wikiPageRedirects of	<ul> <li>dbpedia:Stravinsky</li> <li>dbpedia:Igor_Stravinskij</li> <li>dbpedia:Igor_Strawinsky</li> </ul>	

# KB's: Example NELL (Carlson, Mitchell, et al.: WSDM'10, AAAI'10)



http://rtw.ml.cmu.edu/rtw/kbbrowser/writer:alexander\_pushkin 🖟

### KB's: Example NELL (Carlson, Mitchell, et al.: WSDM'10, AAAI'10)

### **NELL Knowledge Base Browser** Search CMU Read the Web Project person alexander pushkin (writer) search results for: 'Pushkin' actor literal strings: Alexander Pushkin, alexander pushkin athlete celebrity pushkin (museum) ceo chef pushkin (writer) coach **Help NELL Learn!** a s pushkin (scientist) female a s pushkin (writer) journalist NELL wants to know if male a pushkin (writer) If they are or ever were, click thum mlauthor alexander pushkin (monarch) musician • alexander pushkin is a writer 🏖 🕏 alexander pushkin (musician) politician • alexander pushkin is a European person 🏖 🕏 politicianus alexander pushkin (writer) scientist the <u>lattitude</u> and <u>longitude</u> of <u>alexander\_pushkin</u> is <u>38.899560000</u> pushkin museum (museum) visualartist aleksandr pushkin (writer) writer astronaut alexander s pushkin (writer) architect categories aleksandr pushkin (personeurope) comedian aleksandr sergeyevich pushkin (scientist) model writer(100.0%) judge poet alexander pushkin (writer) SEAL@190 (75.0%) on 16-jan-2011 [ 1 2 ] using alexande criminal CMC @308 (99.9%) on 20-jun-2011 [ PREFIX=ale FULL\_W kelly lynn pushkin (celebrity) monarch CPL@215 (100.0%) on 26-feb-2011 [ "novel by \_" "novella personbylocation aleksei musin pushkin (monarch) \_" "\_ 's Eugene Onegin" "authors such as \_" "\_ 's novella" ' personafrica alexander pushkin statue (monument) "short story by " "same name by " "favorite writers are " personantarctica • pushkin fine arts museum (museum) personasia person(100.0%) personaustralia the pushkin museum of fine arts (museum) SEAL@227 (75.0%) on 01-apr-2011 [ 1 2 ] using alexande personeurope personeurope(99.9%) pushkin museum of fine arts (museum) personnorthamerica SEAL@180 (93.8%) on 16-dec-2010 [ 1 2 3 4 ] using alexa russian poet alexander pushkin (personeurope) personus CPL@215 (93.8%) on 26-feb-2011 [ "novella by " "narrati politicianus each year mary and alexander pushkin (monarch) alexander pushkin personmexico a s pushkin the complete collection of works (visua CMC @182 (77.7%) on 04-jan-2011 [ WORDSHAPE=aaaa: personcanada alexander pushkin museum of fine arts (museum) alexander pushkin personsouthamerical director pushkin state museum of fine arts (museum) NELL has only weak ev professor pushkinsky bridge (bridge) location scientist katya pushkina (model) CPL@196 (87.5%) on 02-feb-2011 [ "famous pupil was " blog (+) pushkinsky pedestrian bridge (bridge) building relations hotel monument latitudelongitude museum • 38.8995600000000,-77 044090000000 /400 0%\ restaurant

stadiumoreventvenue

airport

bridge

\* LatLong @170 (1 statue@38.900,-1 http://rtw.ml.cmu.edu/rtw/kbbrowser/

## **Outline**

✓ Motivation

- ✓ Machine Knowledge
- Knowledge Harvesting
  - Entities and Classes
  - Relational Facts
- Research Challenges
  - Open-Domain Extraction
  - Temporal Knowledge
- ★ Wrap-up



# WordNet Thesaurus [Miller/Fellbaum 1998]

WordNet Search - 3.0 - WordNet home page - Glossary - Help

Word to search for: partner

Search WordNet

Change

Change

Their synonyms (synset's)

Prelations, "W:" = Show Word (lexical) relations

### Моши

- <u>S: (n) spouse, partner, married person, mate, better half</u> (a person's partner in marriage)
- S: (n) collaborator, cooperator, partner, pardner (an associate in an activity or endeavor or sphere of common interest) "the musician and the librettist were collaborators"; "sexual partners"
- <u>S:</u> (n) partner (a person who is a member of a partnership)

### Verb

- <u>S</u>: (v) partner (provide with a partner)
- S: (v) partner (act as a partner) "Astaire partnered Rogers"



# WordNet Thesaurus [Miller/Fellbaum 1998]

### Noun

- S: (n) spouse, partner, married person, mate, better half (a person's partner in marriage)
  - direct hyponym I full hyponym
    - S: (n) bigamist (someone who marries one person while already legally married to another)
    - S: (n) consort (the husband or wife of a reigning monarch)
    - S: (n) helpmate, helpmeet (a helpful partner)
    - (hyponyms) S: (n) husband, hubby, married man (a married man; a woman's partner in marriage)
    - S: (n) monogamist, monogynist (someone who practices monogamy (one spouse at a time))
    - S: (n) newlywed, honeymooner (someone recently married)
    - S: (n) polygamist (someone who is married to two or more people at the same time)
    - S: (n) wife, married woman (a married woman; a man's partner in marriage)
  - member holonym
  - direct hypernym | inherited hypernym | sister term
    - (hypernyms) S: (n) relative, relation (a person related by blood or marriage) "police are searching for relatives of the deceased"; "he has distant relations back in New Jersey"
    - S: (n) domestic partner, significant other, spousal equivalent, spouse equivalent (a person (not necessarily a spouse) with whom you cohabit and share a long-term sexual relationship)
  - derivationally related form
- S: (n) collaborator, cooperator, partner, pardner (an associate in an activity or endeavor or sphere of common interest) "the musician and the librettist were collaborators"; "sexual partners"
- S: (n) partner (a person who is a member of a partnership)



http://wordnet.princeton.edu/

subclasses

superclasses

# WordNet Thesaurus [Miller & Fellbaum 1998]

- > 100 000 classes and lexical relations; can be cast into
- description logics or
- graph, with weights for relation strengths (derived from co-occurrence statistics)

but:

only few individual entities

(instances of classes)

### Sense 1

scientist, man of science -- (a person with advanced knowledge of => cosmographer, cosmographist -- (a scientist knowledgeable

- => bibliotist -- (someone who engages in bibliotics)
- => biologist, life scientist -- ((biology) a scientist who studies li-
- => chemist -- (a scientist who specializes in chemistry)
- => cognitive scientist -- (a scientist who studies cognitive proce
- => computer scientist -- (a scientist who specializes in the theorem
- => geologist -- (a specialist in geology)
- => linguist, linguistic scientist -- (a specialist in linguistics)
- => mathematician -- (a person skilled in mathematics)
- => medical scientist -- (a scientist who studies disease processe
- => microscopist -- (a scientist who specializes in research with
- => mineralogist -- (a scientist trained in mineralogy)
- => oceanographer -- (a scientist who studies physical and biolo
- => paleontologist, palaeontologist, fossilist -- (a specialist in pale
- => physicist -- (a scientist trained in physics)
- => principal investigator, PI -- (the scientist in charge of an exp
- => psychologist -- (a scientist trained in psychology)
- => radiologic technologist -- (a scientist trained in radiological to
- => research worker, researcher, investigator -- (a scientist who
- => social scientist -- (someone expert in the study of human so

HAS INSTANCE=> Bacon, Roger Bacon -- (English scientist ar combustion and first used lenses to correct vision (122

HAS INSTANCE=> Franklin, Benjamin Franklin -- (printer who the Constitution, he played a major role in the American his research in electricity (1706-1790))

HAS INSTANCE=> Galton, Francis Galton, Sir Francis Galton psychology, anthropology, founder of eugenics and fir HAS INSTANCE=> Harvey, William Harvey -- (English physici ovum produced by the female of the species (1578-165

scientist, man of science

(a person with advanced knowledge)

- => cosmographer, cosmographist
- => biologist, life scientist
- => chemist
- => cognitive scientist
- => computer scientist

•••

=> principal investigator, PI

• • •

**HAS INSTANCE => Bacon, Roger Bacon** 

http://wordnet.princeton.edu/

"Hyponyms [...is a kind of this], brief" search for noun "scientist"

# **Tapping on Wikipedia Categories**

Jim Gray (computer scientist)

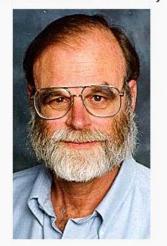
From Wikipedia, the free encyclopedia

James Nicholas "Jim" Gray (born 12 January 1944, lost at sea 28 January 2007) was an American computer scientist who received the Turing Award in 1998 "for seminal contributions to database and transaction processing research and technical leadership in system implementation."

### Contents [hide]

- 1 Family and education
- 2 Work
- 3 Disappearance at sea and search
- 4 Books
- 5 See also
- 6 References
- 7 External links

James Nicholas "Jim" Gray



Born

January 12, 1944<sup>[1]</sup>

San Francisco, California [2]

Died

(lost at sea) January 28, 2007

Nationality American

Fields Come

Computer Science

Institutions

IBM, Tandem Computers, DEC,

licrosoft

Alma mater

University of California, Berkeley

Doctoral advisor

Michael Harrison [2]

Known for

Work on database and transaction

processing systems

Notable awards

Turing Award

Categories: Members of the National Academy of Sciences | American computer scientists | Fellows of the Association for Computing Machinery | Microsoft employees | DEC people | Database researchers | SIGMOD Edgar F. Codd Innovations Award winners | Turing Award laureates | 1944 births | 2007 deaths | People lost at sea | University of California, Berkeley alumni

Tapping on Wikipedia Categories

### Max Planck

From Wikipedia, the free encyclopedia

"Planck" redirects here. For other uses, see Planck (disambiguation).

Max Planck (April 23, 1858 – October 4, 1947) was a German physicist. He is considered to be the founder of the quantum theory, and thus one of the most important physicists of the twentieth century. Planck was awarded the Nobel Prize in Physics in 1918.

### Contents [hide]

- 1 Life and career
  - 1.1 Academic career
  - 1.2 Family
  - 1.3 Professor at Berlin University
  - 1.4 Black-body radiation
  - 1.5 Einstein and the theory of relativity
  - 1.6 World War and Weimar Republic
  - 1.7 Quantum mechanics
  - 1.8 Nazi dictatorship and The Second World War
- 2 Religious view
- 3 Honors and awards



Born

April 23, 1858 Kiel, Holstein

Died

October 4, 1947 (aged 89) Göttingen, West Germany

Nationality

German

Fields

Physics

Institutions

University of Kiel University of Berlin University of Göttingen

University of Göttingen Kaiser-Wilhelm-Gesellschaft

Alma mater

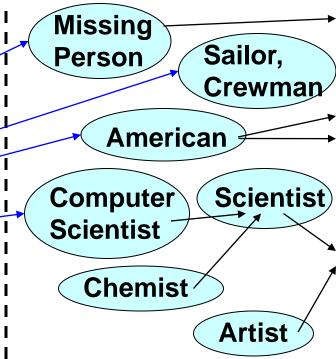
Ludwig Maximilian University of

Categories: German Nobel laureates | German physicists | Members of the Pontifical Academy of Sciences | Members of the Prussian Academy of Sciences | Nobel laureates in Physics | Recipients of the Copley Medal | People from Kiel | People from the Province of Schleswig-Holstein | Quantum physicists | Recipients of the Pour le Mérite (civil class) | Theoretical physicists | Thermodynamicists | University of Munich alumni | University of Munich faculty | Humboldt University of Berlin alumni | Humboldt University of Berlin faculty | University of Kiel faculty | German Christians | Religion and science | Fellows of the Leopoldina | 1858 births | 1947 deaths

Mapping: Wikipedia → WordNet

[Suchanek: WWW'07, Ponzetto&Strube: AAAI'07]

Jim Gray (computer specialist)





Mapping: Wikipedia → WordNet [Suchanek: WWW'07, Ponzetto&Strube: AAAI'07] Missing **Person** þr, **People** vman Lost at Sea Computer **American Scientists** instanceOf by Nation **American Scientist** Computer Computer subclassOf **Scientist Scientists** Jim Gray **Databases** Data-(computer **Database** base specialist) Researcher **Engineering Fellow** (1), **Societies** Comrade **Fellows of** the ACM Fellow (2), **ACM** Colleague name similarity Fellow (3) Members (edit dist., n-gram overlap) ? of Society) of Learned context similarity **Societies** Member (1), (word/phrase level) ? **Fellow** machine learning? Member (2), **Extremity** max planck institut informatik

# Mapping: Wikipedia → WordNet

[Suchanek: WWW'07, Ponzetto & Strube:AAAI'07]

Given: entity e in Wikipedia categories c<sub>1</sub>, ..., c<sub>k</sub>

Wanted: instanceOf(e,c) and subclassOf(c<sub>i</sub>,c) for WN class c

Problem: vagueness & ambiguity of names c<sub>1</sub>, ..., c<sub>k</sub>

### **Analyzing category names** → **noun group parser**:

American Musicians of Italian Descent

pre-modifier head post-modifier

American Folk Music of the 20th Century

pre-modifier head post-modifier

**American Indy 500 Drivers on Pole Positions** 

pre-modifier head post-modifier

**Head word is key, should be in plural for instanceOf** 



## Mapping Wikipedia Entities to WordNet Classes

[Suchanek: WWW'07, Ponzetto & Strube: AAAI'07]

```
Given: entity e in Wikipedia categories c<sub>1</sub>, ..., c<sub>k</sub>
```

Wanted: instanceOf(e,c) and subclassOf(c<sub>i</sub>,c) for WN class c

Problem: vagueness & ambiguity of names c<sub>1</sub>, ..., c<sub>k</sub>

```
\label{eq:heuristic Method:} \begin{split} & \underline{\text{Heuristic Method:}} \\ & \text{for each } c_i \text{ do} \\ & \text{if head word w of category name } c_i \text{ is plural} \\ & \{ \\ & 1) \text{ match w against synsets of WordNet classes} \\ & 2) \text{ choose best fitting class c and set } e \in c \\ & 3) \text{ expand w by pre-modifier and set } c_i \subseteq w^+ \subseteq c \\ & \} \end{split}
```

tuned conservatively: high precision, reduced recall

- can also derive features this way
- feed into supervised classifier



# Learning More Mappings [Wu & Weld: WWW'08]

### **Kylin Ontology Generator (KOG):**

learn classifier for subclassOf across Wikipedia & WordNet using

- YAGO as training data
- advanced ML methods (MLN's, SVM's)
- rich features from various sources
  - category/class name similarity measures
  - category instances and their infobox templates: template names, attribute names (e.g. knownFor)
  - Wikipedia edit history: refinement of categories
  - Hearst patterns:
     C such as X, X and Y and other C's, ...
  - other search-engine statistics: co-occurrence frequencies
    - > 3 Mio. entities
    - > 1 Mio. w/ infoboxes
    - > 500 000 categories



## Long Tail of Class Instances



Automatically create sets of items from a few examples.

Enter a few items from a set of things. (example)

Next, press Large Set or Small Set and we'll try to predict other items in the set.

•	Pushkin	
•	Tolstoy	
•	Pasternak	
•		
•		
	( <u>clear all</u> )	
Large Set	Small Set (15 items or	fewer)

Examples:

green, purple, red chicken dance, macarena, ymca alexander, gladiator, troy hilary duff, kelly clarkson more...

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## **Long Tail of Class Instances**

Predicted Items
tolstoy
pushkin
<u>leo tolstoy</u>
anna karenina
gogol
<u>drama</u>
danielle steel
dostoevsky
maxim gorky
<u>russia</u>
fyodor dostoevsky
anton chekhov
<u>ivan turgenev</u>
paulo coelho
dan brown
ernest hemingway
<u>dostojevski</u>
alexander pushkin

john steinbeck
russian literature
<u>lermontov</u>
stephen king
<u>cs lewis</u>
madame bovary
<u>bible</u>
the idiot
mark twain
mikhail bulgakov
fyodor dostoyevsky
nikolai gogol
susanna tamaro
edward said
dirty dancing
albert camus
<u>shakespeare</u>
romance novel
<u>jack london</u>
george orwell
<u>fiction</u>
authors





## Long Tail of Class Instances

[Etzioni et al. 2004, Cohen et al. 2008, Mitchell et al. 2010]

#### State-of-the-Art Approach (e.g. SEAL):

- Start with seeds: a few class instances
- Find lists, tables, text snippets ("for example: ..."), ...
   that contain one or more seeds
- Extract candidates: noun phrases from vicinity
- Gather co-occurrence stats (seed&cand, cand&className pairs)
- Rank candidates
  - point-wise mutual information, ...
  - random walk (PR-style) on seed-cand graph

#### **But:**

Precision drops for classes with sparse statistics (IR profs, ...)
Harvested items are names, not entities
Canonicalization (de-duplication) unsolved



## **Outline**

✓ Motivation

- ✓ Machine Knowledge
- Knowledge Harvesting
  - Entities and Classes
  - Relational Facts
- Research Challenges
  - Open-Domain Extraction
  - Temporal Knowledge
- ★ Wrap-up



## Tapping on Wikipedia Infoboxes

#### Jim Gray (computer scientist)

From Wikipedia, the free encyclopedia

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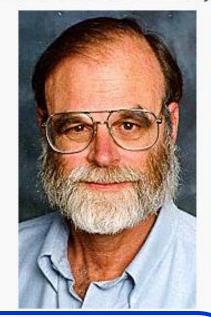
- 1 Family and education
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- 6 References
- 7 External links

# harvest by extraction rules:

- regex matching
- type checking

(?i)IBL\|BEG\s\*awards\s\*=\s\*(.\*?)IBL\|END" => "\$0 hasWonPrize @WikiLink(\$1)

#### James Nicholas "Jim" Gray



Born January 12, 1944<sup>[1]</sup>

San Francisco, California [2]

Died (lost at sea) January 28, 2007

Nationality American

Fields Computer Science

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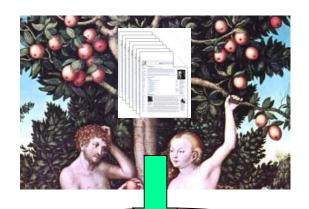
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processing systems

Notable Turing Award

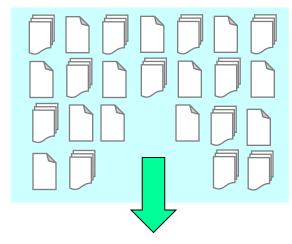


## French Marriage Problem



#### facts in KB:

married
(Hillary, Bill)
married
(Carla, Nicolas)
married
(Angelina, Brad)



#### new facts or fact candidates:

married (Cecilia, Nicolas)
married (Carla, Benjamin)
married (Carla, Mick)
married (Michelle, Barack)
married (Yoko, John)
married (Kate, Leonardo)
married (Carla, Sofie)
married (Larry, Google)

- 1) for recall: pattern-based harvesting
- 2) for precision: consistency reasoning



## **Pattern-Based Harvesting**

(Hearst 92, Brin 98, Agichtein 00, Etzioni 04, ...)

#### Facts & Fact Candidates

#### <u>Patterns</u>

(Hillary, Bill)

(Carla, Nicolas)

(Angelina, Brad)

(Victoria, David)

(Hillary, Bill)

(Carla, Nicolas)

(Yoko, John)

(Kate, Pete)

(Carla, Benjamin)

(Larry, Google)

(Angelina, Brad)

(Victoria, David)

X and her husband Y
X and Y on their honeymoon

X and Y and their children

X has been dating with Y

X loves Y

- good for recall
  - noisy, drifting
  - not robust enough for high precision



## Reasoning about Fact Candidates

Use consistency constraints to prune false candidates

#### **FOL rules (restricted):**

```
spouse(x,y) \land diff(y,z) \Rightarrow \negspouse(x,z)
spouse(x,y) \land diff(w,x) \Rightarrow \negspouse(w,y)
spouse(x,y) \Rightarrow f(x) spouse(x,y) \Rightarrow m(y)
spouse(x,y) \Rightarrow (f(x)\landm(y)) \lor (m(x)\landf(y))
```

Rules reveal inconsistencies Find consistent subset(s) of atoms ("possible world(s)", "the truth")

#### ground atoms:

f(Sofie)

```
spouse(Hillary,Bill)
spouse(Carla,Nicolas)
spouse(Cecilia,Nicolas)
spouse(Carla,Ben)
spouse(Carla,Mick)
spouse(Carla, Sofie)

f(Hillary) m(Bill)
f(Carla) m(Nicolas)
f(Cecilia) m(Ben)
```

m(Mick)

Rules can be weighted (e.g. by fraction of ground atoms that satisfy a rule)

- → uncertain / probabilistic data
- → compute prob. distr. of subset of atoms being the truth



## Markov Logic Networks (MLN's)

(M. Richardson / P. Domingos 2006)

#### Map logical constraints & fact candidates into probabilistic graph model: Markov Random Field (MRF)

$$s(x,y) \wedge diff(y,z) \Rightarrow \neg s(x,z)$$
  
 $s(x,y) \wedge diff(w,y) \Rightarrow \neg s(w,y)$ 

$$\begin{array}{ll} s(x,y) \wedge diff(y,z) \Rightarrow \neg s(x,z) & s(x,y) \Rightarrow f(x) & f(x) \Rightarrow \neg m(x) \\ s(x,y) \wedge diff(w,y) \Rightarrow \neg s(w,y) & s(x,y) \Rightarrow m(y) & m(x) \Rightarrow \neg f(x) \end{array}$$

s(Carla, Nicolas) s(Cecilia, Nicola s(Carla, Ben) s(Carla,Sofie)

#### **Grounding:**

```
Literal → Boolean Var
Literal → binary RV
     –s(Ca,Nic)√ m(Nic)
     ¬s(Ce,Nic) √ m(Nic)
     -s(Ca,Ben) ∨ m(Ben)
     –s(Ca,So) <mark>√</mark>m(So)
```

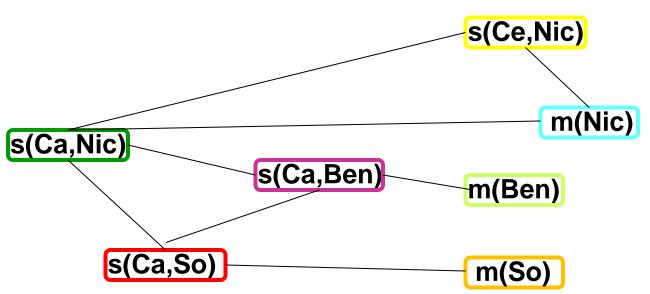
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s(Carla,Nicolas) s(Cecilia,Nicola s(Carla,Ben) s(Carla,Sofie)



RVs coupled by MRF edge if they appear in same clause

Variety of algorithms for joint inference: Gibbs sampling, other MCMC, belief propagation, randomized MaxSat, ... MRF assumption:
P[X<sub>i</sub>|X<sub>1</sub>..X<sub>n</sub>]=P[X<sub>i</sub>|N(X<sub>i</sub>)]
joint distribution
has product form
over all cliques



#### Related Alternative Probabilistic Models

Constrained Conditional Models [D. Roth et al. 2007]

log-linear classifiers with constraint-violation penalty mapped into Integer Linear Programs

#### Factor Graphs with Imperative Variable Coordination

[A. McCallum et al. 2008]

RV's share "factors" (joint feature functions) generalizes MRF, BN, CRF, ... inference via advanced MCMC flexible coupling & constraining of RV's

s(Ca,Nic)
s(Ca,Ben)
m(Ben)
s(Ca,So)
m(So)

software tools: <u>alchemy.cs.washington.edu</u> <u>code.google.com/p/factorie/</u>

research.microsoft.com/en-us/um/cambridge/projects/infernet/

## Reasoning for KB Growth: Direct Route

(F. Suchanek et al.: WWW'09)

facts in KB: married (Hillary, Bill) married (Carla, Nicolas) married (Angelina, Brad)

#### new fact candidates:

married (Cecilia, Nicolas) married (Carla, Benjamin) married (Carla, Mick) married (Carla, Sofie)

married (Larry, Google)

#### patterns:

X and her husband Y

X and Y and their children

X has been dating with Y

X loves Y

#### **Direct approach:**

- 1. facts are true; fact candidates & patterns → hypotheses grounded constraints → clauses with hypotheses as vars
- type signatures of relations greatly reduce #clauses
- cast into Weighted Max-Sat with weights from pattern stats customized approximation algorithm

unifies: fact cand consistency, pattern goodness, entity disambig. www.mpi-inf.mpg.de/yago-naga/sofie/

## Facts & Patterns Consistency with SOFIE

(F. Suchanek et al.: WWW'09)

constraints to connect facts, fact candidates, patterns pattern-fact duality:

```
occurs(p,x,y) \land expresses(p,R) \land type(x)=dom(R) \land type(y)=rng(R) \Rightarrow R(x,y)
occurs(p,x,y) \land R(x,y) \land type(x)=dom(R) \land type(y)=rng(R) \Rightarrow expresses(p,R)
```

```
name(-in-context)-to-entity mapping:
```

```
\neg means(n,e1) \lor \neg means(n,e2) \lor \dots
```

```
functional dependencies: relation properties:
```

spouse(X,Y):  $X \rightarrow Y, Y \rightarrow X$  asymmetry, transitivity, acyclicity, ...

#### type constraints, inclusion dependencies:

spouse ⊆ Person × Person capitalOfCountry ⊆ cityOfCountry

#### domain-specific constraints:

bornInYear(x) + 10years ≤ graduatedInYear(x)

hasAdvisor(x,y)  $\land$  graduatedInYear(x,t)  $\land$  graduatedInYear(y,s)  $\Rightarrow$  s < t



## **Pattern Harvesting Revisited**

(N. Nakashole et al.: WebDB'10,)

#### narrow / nasty / noisy patterns:

- X and his famous advisor Y
- X carried out his doctoral research in math under the supervision of Y
- X jointly developed the method with Y

the supervision of Y using noisy loses precision &

slows down MaxSat

using narrow &

dropping nasty

loses recall!

#### **POS-lifted n-gram itemsets as patterns:**

X { his doctoral research, under the supervision of} Y

X { PRP ADJ advisor } Y

X { PRP doctoral research, IN DET supervision of} Y

#### confidence weights, using seeds and counter-seeds:

seeds: (ThomasHofmann, JoachimBuhmann), (JimGray, MikeHarrison)

counter-seeds: (BernhardSchölkopf, AlexSmola), (AlonHalevy, LarryPage)

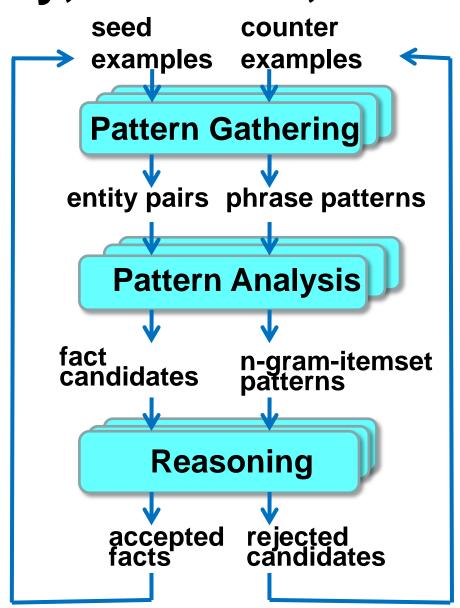
→ confidence of pattern p ~ #p with seeds – #p with counter-seeds



# PROSPERA: Prospering Knowledge with Scalability, Precision, Recall

(N. Nakashole et al.: WSDM'11)

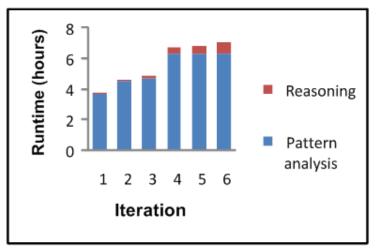
- feedback loop for higher recall
- all stages parallelizable on MapReduce platform

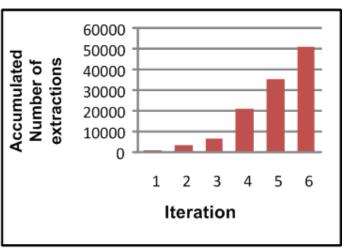




## Web-Scale Experiments [N. Nakashole et al.: WSDM'11]

- on ClueWeb'09 corpus (500 Mio. English Web pages)
- with Hadoop cluster of 10x16 cores and 10x48 GB memory
- 10 seed examples, 5 counter examples for each relation





	PROSPERA		ReadTheWeb [CMU]		
Relation	#Facts	<b>Precision</b>	Prec@1000	#Facts	<b>Precision</b>
AthletePlaysForTeam	14685	82%	100%	456	100%
<b>TeamPlaysAgainstTeam</b>	15170	89%	100%	1068	99%
TeamMate	19666	86%	100%		
FacultyAt	4394	96%	100%		



#### **Outline**

✓ Motivation

✓ Machine Knowledge

- Knowledge Harvesting
  - Entities and Classes
  - Relational Facts
- **Research Challenges** 
  - Open-Domain Extraction
  - Temporal Knowledge
- ★ Wrap-up



## **Discovering New Relation Types**

# Targeted (Domain-Oriented) Gathering of Facts: Entity × Relation × Entity

```
< Carla_Bruni marriedTo Nicolas_Sarkozy>,
< Natalie_Portman wonAward Academy_Award >,
...
```

#### **Explorative (Open-Domain) Gathering of**

"Assertions": Name × Pattern × Name

```
< "Carla Bruni" "had affair with" "Mick Jagger" >,
< "First Lady Carla" "had affair with" "Stones singer Mick" >,
< "Madame Bruni" "happy marriage with" "President Sarkozy" >,
< "Jeff Bridges" "expected to win" "Oscar" >,
< "Coen Brothers" "celebrated for" "Oscar Award" >,
```



# Open-Domain Gathering of Assertions [O. Etzioni et al. 2007, F. Wu et al. 2010]

Analyze verbal phrases between entities for new relation types

unsupervised bootstrapping with short dependency paths



Rumors about Carla indicate there is something between her and Ben

• self-supervised classifier (CRF) for (noun, verb-phrase, noun) triples

```
... seen dating with ... (Carla, Ben), (Carla, Sofie), ...
... partying with ... (Carla, Ben), (Paris, Heidi), ...
```

- build statistics & prune sparse candidates
- group/cluster candidates for new relation types and their facts {datesWith, partiesWith}, {affairWith, flirtsWith}, {romanticRelation}, ...

**But: result is noisy** clusters are not canonicalized relations far from near-human-quality



## Open IE Example: TextRunner / ReVerb



#### ReVerb Search

http://www.cs.washington.edu/research/textrunner/reverbdemo.html

ReVerb took 13.32 seconds

Retrieved 242 results for Predicate containing "has won" and Argument 2 containing "prize"

Grouping results by predicate. Group by: argument 2 | argument 1

#### has won (157 results)

book (28), film (25), no one (25), 50 more... has won prizes

Al Gore (41), IPCC (6), Carter (6), 9 more... has won the Nobel Peace Prize

Al Gore (15), highly successful novelist (6), Chinese writer (5), 11 more... has won the Nobel Prize

book (11), author (3), scholarly work (2), 10 more... has won the Pulitzer Prize

Coetzee (5), book (4), Peter Carey (2), 2 more... has won the Booker Prize

Artist Mark Wallinger (3), Mark Wallinger (2), Simon Starling (2) has won the Turner Prize

James Ehnes (2), Work (2), school (2) has won numerous awards and prizes

recipient (7), Tickets (2) has won a cash prize

Katherine Taylor (2), His poetry (2) has won a Pushcart Prize

Bryan D. Dietrich (3) has won the Paris Review Poetry Prize

Woodward (3) has won the Wynne Prize

British playwright Harold Pinter (2) has won the 2005 Nobel Literature Prize

Al Gore (2) has won the 2007 Nobel Peace Prize

Argentine poet Juan Gelman (3) has won the Cervantes prize

Mr Bean (2) has won a church fete raffle 's top prize

No woman (2) has ever won the Economics Prize

film (2) has already won five international prizes

No other journalist (2) has won the Ford Prize

Alice Munro (3) has won the Giller Prize

Robert Nye (2) has won the Hawthornden Prize

his translator Daniel Hahn (2) have **won** the Independent Foreign Fiction **Prize** 2007

#### Search again: Argument 1

Predicate

has won Argument 2

prize
Search Again

#### Jump to:

has won (157)

will have a chance to win (16) have a chance to win (17)

will have the opportunity to win (15)

has an equal chance of winning

have the opportunity to win (7)

has won a number of (5) should have won (4)

has become the first woman to

<u>win (2)</u>

has a chance at winning (1)

must not have won (2) has not yet won (2)

have gone on to win (1)
have the opportunity to win one of

had a fair shot at winning (1)

# Open IE Example: TextRunner / ReVerb <a href="http://www.cs.washington.edu/research/textrunner/reverbdemo.html">http://www.cs.washington.edu/research/textrunner/reverbdemo.html</a>

ReVerb took 19 89 seconds

Retrieved 296 results for Predicate containing "wrote"

Grouping results by predicate, Group by: argument 2 | argument 1

#### wrote (222 results)

poet (36), Students (31), people (17), **197** more... wrote a poem

Pablo Neruda (2), African-American poet (2), Sappho (2) wrote love poems

Shakespeare (11) wrote plays and poems

Dickinson (5), Emily Dickinson (2) wrote 1,775 poems

hundreds of Asian American men (4) are writing books and poems

students (2), Marion Brown (2), Stuart (2) began writing stories and poems

Keats (2) wrote 150 poems

Students (2) also wrote letters and poems

Shakespeare (2) wrote many poems and plays

Computer hackers (2) had written source code poems

Emily Dickinson (3) wrote 1800 poems

Students (2) write a Fall poem

students (3) then wrote Haiku poems

Scottish poet (2) wrote many many poems and songs

Ginsberg (2) wrote the classic long poem Kaddish

Have students (3) write poems or stories

Composers (2) wrote tone poems

Tennyson (2) wrote two versions of the poem

#### can write (9 results)

Anvone (7), students (5), Children (8), 6 more... can write a poem

#### will write (5 results)

Students (10), someone (2) will write an acrostic **poem** 

Students (5) will write two poems

Students (2) will write a haiku poem

#### Search again: Argument 1 Predicate wrote Argument 2 poem Search Again

#### Jump to:

wrote (222) can write (9) will write (5) is written as (5) wrote in (9) does not write (2) wrote some of (4) has been writing (2) has written thousands of (1) would write (2) write all kinds of (2) wrote more than (2) wrote over (2)

wrote a total of (1) was written to (1) can write the first line of (1)

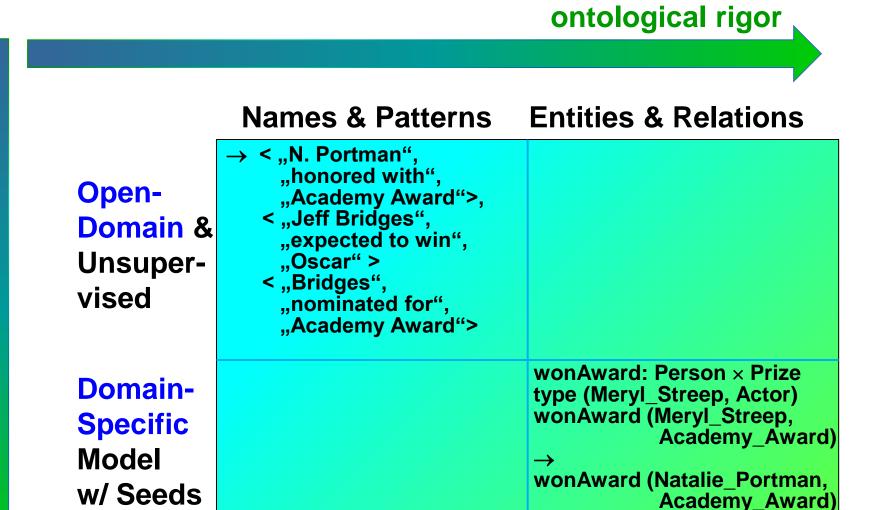
always enjoyed writing (1). have been writing lots of (1) had begun writing (1)

Ve never written (2) started writing (1) will write a variety of (1)

wrote hundreds of (2) wrote scores of (1)

wrote thousands of (1) wrote a variety of (1) were asked to write (1)

#### **Challenge: Unify Targeted & Explorative Methods**

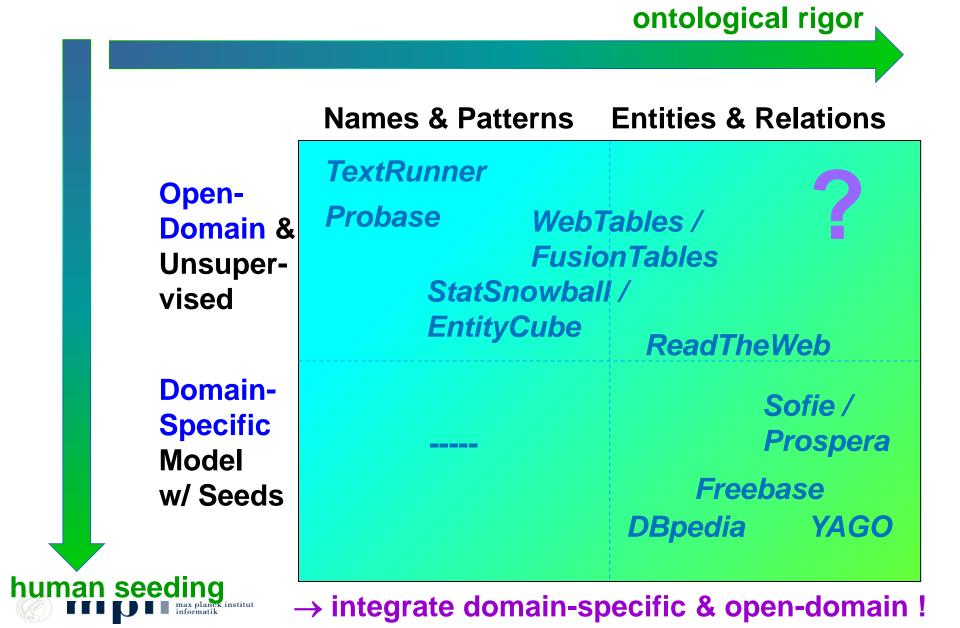


wonAward (Ethan\_Coen,

Palme d'Or)



#### **Challenge: Unify Targeted & Explorative Methods**



## **Outline**

✓ Motivation

✓ Machine Knowledge

- ✓ Knowledge Harvesting
  - Entities and Classes
  - Relational Facts
- Research Challenges
  - Open-Domain Extraction
  - Temporal Knowledge
- ★ Wrap-up



## As Time Goes By: Temporal Knowledge

Which facts for given relations hold at what time point or during which time intervals?

```
marriedTo (Madonna, Guy) [ 22Dec2000, Dec2008 ]
capitalOf (Berlin, Germany) [ 1990, now ]
capitalOf (Bonn, Germany) [ 1949, 1989 ]
hasWonPrize (JimGray, TuringAward) [ 1998 ]
graduatedAt (HectorGarcia-Molina, Stanford) [ 1979 ]
graduatedAt (SusanDavidson, Princeton) [ Oct 1982 ]
hasAdvisor (SusanDavidson, HectorGarcia-Molina) [ Oct 1982, forever ]
```

How can we query & reason on entity-relationship facts in a "time-travel" manner - with uncertain/incomplete KB?

Swedish king's wife when Greta Garbo died? students of Hector Garcia-Molina while he was at Princeton?



## French Marriage Problem

JAN FEE N

MAR

APR MAY JUN JUL AUG SEP OCT NOV DEC

#### facts in KB

1: married

(Hillary, Bill)

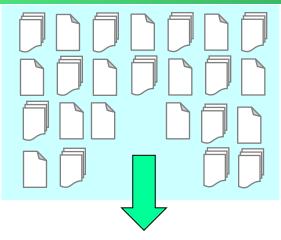
2: married

(Carla, Nicolas)

3: married

(Angelina, Brad)

**validFrom (2, 2008)** 



#### new fact candidates:

4: married (Cecilia, Nicolas)

5: married (Carla, Benjamin)

6: married (Carla, Mick)

7: divorced (Madonna, Guy)

**validUntil (4, 2007)** 

validFrom (4, 1996) validFrom (5, 2010) validFrom (6, 2006)

validFrom (6, 2006) validFrom (7, 2008)





## **Challenge: Temporal Knowledge**

for all people in Wikipedia (300 000) gather all spouses, incl. divorced & widowed, and corresponding time periods! >95% accuracy, >95% coverage, in one night

28 January 1955 (age 53)

Nicolas Paul Stéphane

Paris, France

Sarközv

- 1) recall: gather temporal scopes for base facts
- 2) precision: reason on mutual consistency



#### consistency constraints are potentially helpful:

- functional dependencies: husband, time → wife
- inclusion dependencies: marriedPerson ⊆ adultPerson
- age/time/gender restrictions: birthdate + △ < marriage < divorce

## Difficult Dating

# Nicolas Sarkozy

#### President of France

#### Incumbent Assumed office

28 January 1955 (age 55) Born

Paris, France

Political party Union for a Popular

Movement (2002-present)

Other political Rally for the Republic

affiliations (1976-2002)

Spouse(s) Marie-Dominique Culioli

(1982-1996)

Cécilia Ciganer-Albéniz (1996-2007)

Carla Bruni-Sarkozy (2008-

Children Pierre Sarkozy (by Culioli)

> Jean Sarkozy (by Culioli) Louis Sarkozy (by Ciganer-

Abéniz)

Residence Élysée Palace

Paris X University Nanterre Alma mater

Profession Lawyer

Roman Catholicism Religion

#### Cécilia Attias

#### First Lady of France

In office

16 May 2007 - 10 October 2007

President Nicolas Sarkozy Preceded by Bernadette Chirac

Succeeded by Carla Bruni

Born November 12, 1957 (age 52) Boulogne-Billancourt, France

Spouse(s) Jacques Martin (m. 1984-1989) Nicolas Sarkozy (m. 1996-2007)

Richard Attias (m. 2008-present)

Children Judith Martin (b.1984) Jeanne-Marie Martin (b.1987)

Louis Sarkozy (b.1997)



#### Wife of the President of the French Republic

#### Incumbent

Assumed office

2 February 2008

Nicolas Sarkozy President

Preceded by Cécilia Ciganer-Albéniz

Born 23 December 1967 (age 42)

Turin, Italy

Birth name Carla Gilberta Bruni Tedeschi

Hationality Italian French[1]

Spouse(s) Nicolas Sarkozy Aurélien Enthoven (with Raphaël

Enthoven)

Charles

Prince of Wales; Duke of Rothesay (more)



Lady Diana Spencer Spouse m. 1981; div. 1996 Camilla Parker Bowles m. 2005

Prince William of Wales Prince Harry of Wales

#### **Full name**

Charles Philip Arthur George

Maternal: House of Windsor House

> Paternal: House of Schleswig-Holstein-Sonderburg-Glücksburg

Father Prince Philip, Duke of Edinburgh

Mother Elizabeth II

14 November 1948 (age 61)

Buckingham Palace, London

Signature



Religion Christian (Church of England)

Diana

Princess of Wales; Duchess of Rothesay



Spouse Charles, Prince of Wales (29 July 1981 - 28 August 1996)

Prince William of Wales Prince Henry of Wales

#### **Full name**

Diana Frances Spencer [N 1]

House House of Windsor

Father John Spencer, 8th Earl Spencer

Mother Frances Shand Kydd

Born 1 July 1961

Park House, Sandringham, Norfolk

Died 31 August 1997 (aged 36) Pitié-Salpêtrière Hospital, Paris, France

Althorp, Northamptonshire

Madonna



Madonna at the premiere of I Am Because We Are in 2008.

**Background information** 

Birth name Madonna Louise Ciccone

**Guy Ritchie** 



Guy Ritchie, September 2008

Guy Stuart Ritchie Born

> 10 September 1968 (age 41) Hatfield, Hertfordshire, England

Occupation Filmmaker, Screenwriter

Years 1995-present

active

Spouse(s) Madonna (2000-2008)

(divorced)

## (Even More Difficult) Implicit Dating

#### Nicolas Sarkozy

From Wikipedia, the free encyclopedia

explicit dates vs. implicit dates relative to other dates

"Sarkozy" redirects here. For the sumame, see Sárközi (surname).

Nicolas Sarkozy ((pronounced [nikola saʁkɔzi] ♥ (help·info)), born Nicolas Paul Stéphane Sarkozy de Nagy-Bocsa on 28 January 1955) is the 23rd and current President of the French Republic and ex officio Co-Prince of Andorra. He assumed the office on 16 May 2007 after defeating Socialist Party candidate Ségolène Royal 10 days earlier.

Before his presidency he was leader of the Union for a Popular Movement (UMP). Under Jacques Chirac's presidency he served as Minister of the Interior in Jean-Pierre Raffarin's (UMP) first two governments (from May 2002 to March 2004), then was appointed Minister of Finances in Raffarin's last government (March 2004 to May 2005) and again Minister of the Interior in Dominique de Villepin's government (2005–2007).

Sarkozy was also president of the General council of the Hauts-de-Seine department from 2004 to 2007 and mayor of Neuilly-sur-Seine, one of the wealthiest communes of France from 1983 to 2002. He was Minister of the Budget in the government of Édouard Balladur (RPR, predecessor of the UMP) during François Mitterrand's last term.

Sarkozy is known for wanting to revitalize the French economy. [1][2][3] He has pledged to revive the work ethic, promote new initiatives and fight intolerance. [1] In foreign affairs he has promised a strengthening of the entente cordiale with the United Kingdom [4] and closer cooperation with the United States. [5] He married Carla Bruni-Sarkozy on 2 February 2008 at the Élysée Palace in Paris.



## (Even More Difficult) Relative Dating

#### Early life

# vague dates relative dates

relative order

During Sarkozy's childhood, his father refused to give his wife's family any financial help, even mough ne nau rounded his own advertising agency and had become wealthy. The family lived in a small mansion owned by Sarkozy's grandfather, Benedict Mallah, in the 17th Arrondissement. The family later moved to Neuilly-sur-Seine, one of the wealthiest communes of the Île-de-France région immediately west of the 17th Arrondissement just outside of Paris. According to Sarkozy, his staunchly Gaullist grandfather was more of an influence on him than his father, whom he rarely saw. Sarkozy was, accordingly, raised Catholic. [18]

Sarkozy said that being abandoned by his father shaped much of who he is today. He also has said that, in his early years, he felt inferior in relation to his wealthier classmates. [19] "What made me who I am now is the sum of all the humiliations suffered during childhood", he said later. [19]

narrative text

#### Education

Sarkozy was enrolled in the *Lycée Chaptal* a state-funded public middle and high school in Pans's our arrondissement, where he failed his *sixième*. His family then sent him to the *Cours Saint-Louis de Monceau*, a private Catholic school in the 17th arrondissement, where he was reportedly a mediocre student, <sup>[20]</sup> but where he nonetheless obtained his *baccalauréat* in 1973. He enrolled at the *Université Paris X Nanterre* where he graduated with a Master in Private law, and later with a DEA degree in Business law. Paris X Nanterre had been the starting place for the May '68 student movement and was still a stronghold of leftist students. Described as a quiet student, Sarkozy soon joined the right-wing student organization, in which he was very active. He completed his military service as a part time Air Force cleaner. After graduating, he entered the *Institut d'Études Politiques de Paris* (1979–1981) but failed to graduate due to an insufficient command of the English language. After passing the bar, he became a lawyer specializing in business and family law, <sup>[23]</sup> and was one of Silvio Berlusconi's top French advocates. <sup>[24][25][26]</sup>

#### Framework for T-Fact Extraction

(Y. Wang et al.: EDBT'10, X. Ling et al.: AAAI'10, Y. Wang et al.: CIKM'11)

- 1) represent temporal scopes of facts in the presence of incompleteness and uncertainty
- 2) gather & filter candidates for t-facts: extract base facts R(e1, e2) first; then focus on sentences with e1, e2 and date or temporal phrase
- 3) aggregate & reconcile evidence from observations
- 4) reason on joint constraints about facts and time scopes



(M. Theobald et al.: MUD'10, M. Dylla et al.: BTW'11)

Combine & reconcile t-scopes across different facts

#### constraint:

marriedTo (m) is an injective function at any given point

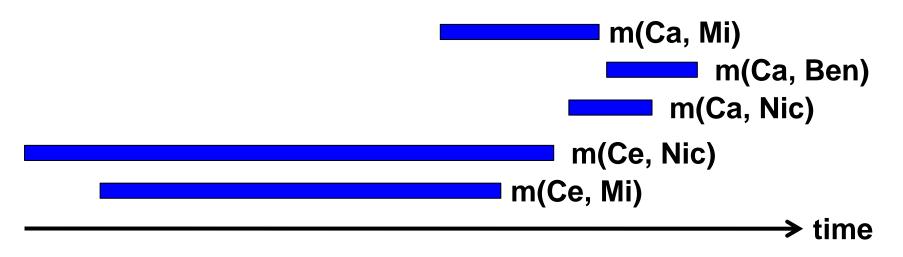
```
∀ X, Y, Z, T1, T2:
    m(X,Y) ∧ m(X,Z) ∧
    validTime(m(X,Y),T1) ∧ validTime(m(X,Z),T2)
    ⇒ ¬ overlaps(T1, T2)
```

#### after grounding:

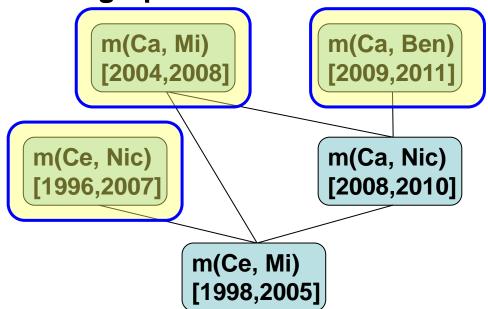
```
m(Carla, Nicolas) \land m(Cecilia, Nicolas) \Rightarrow \neg overlaps ([2008,2010], [1996,2007]) m(Carla, Nicolas) \land m(Carla, Benjamin) \Rightarrow \neg overlaps ([2008,2010], [2009,2011])
```

```
→m(Ca,Nie) ∨
→m(Ce,Nic) ∨
→false

→m(Ca,Nic) ∨
→m(Ca,Nic) ∨
→m(Ca,Ben) ∨
→true
```

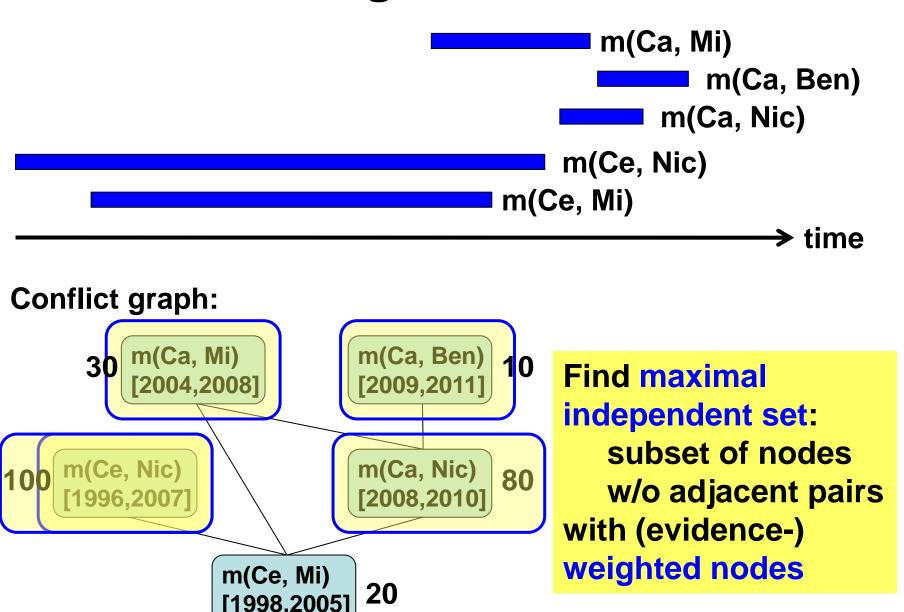


#### **Conflict graph:**



Find maximal independent set:
 subset of nodes
 w/o adjacent pairs with (evidence-)
weighted nodes







alternative approach:
split t-scopes and reason on
consistency of t-fact partitions

m(Ca, Mi)
m(Ca, Ben)
m(Ca, Nic)
m(Ce, Nic)
m(Ce, Mi)

time



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## KB Building: Achievements & Challenges

#### **Entities & Classes**

strong success story, some problems left:

- large taxonomies of classes with individual entities
- long tail calls for new methods
- entity disambiguation remains grand challenge

#### Relationships

good progress, but many challenges left:

- recall & precision by patterns & reasoning
- efficiency & scalability
- soft rules, hard constraints, richer logics, ...
- open-domain discovery of new relation types

#### **Temporal Knowledge**

widely open (fertile) research ground:

- uncertain / incomplete temporal scopes of facts
- joint reasoning on ER facts and time scopes



#### **Overall Take-Home**

Historic opportunity:
revive Cyc vision, make it real & large-scale!
challenging, but high pay-off

Explore & exploit synergies between semantic, statistical, & social Web methods: statistical evidence + logical consistency!

#### For DB / AI / IR / NLP / Web researchers:

- efficiency & scalability
- constraints & reasoning
- killer app for uncertain data management (prob. DB)
- search & ranking for RDF + text
- text (& speech) disambiguation
- knowledge-base life-cycle: growth & maintenance



## Recommended Readings (General)

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- T.M. Mitchell, J.Betteridge, A. Carlson, E.R. Hruschka Jr., R.C. Wang: Populating the Semantic Web by Macro-Reading Internet Text. ISWC 2009



## Recommended Readings (Specific)

- F.M. Suchanek, G. Kasneci, G. Weikum: Yago: a core of semantic knowledge. WWW 2007
- J. Hoffart, F.M. Suchanek, K. Berberich, et al.: YAGO2: exploring and querying world knowledge in time, space, context, and many languages. WWW 2011
- S. Auer, C. Bizer, et al.: DBpedia: A Nucleus for a Web of Open Data. ISWC 2007
- S.P. Ponzetto, M. Strube: Deriving a Large-Scale Taxonomy from Wikipedia. AAAI 2007
- F. Wu, D.S. Weld: Automatically refining the wikipedia infobox ontology. WWW 2008
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- S. Riedel, L. Yao, A. McCallum: Modeling relations and their mentions without labeled text. ECML 2010
- Y.S. Chan, D. Roth: Exploiting Background Knowledge for Relation Extraction. COLING 2010
- M. Banko, M.J. Cafarella, S. Soderland, et al.: Open Information Extraction from the Web. IJCAI 2007
- A. Fader. S. Soderland, O. Etzioni: Identifying Relations for Open Information Extraction, EMNLP 2011
- P.P. Talukdar, F. Pereira: Experiments in Graph-Based Semi-Supervised Learning Methods for Class-Instance Acquisition. ACL 2010
- R. Wang, W.W. Cohen: Language-independent set expansion of named entities using the web. ICDM 2007
- P. Venetis, A. Halevy, et al.: Recovering Semantics of Tables on the Web, VLDB 2011
- F. Niu, C. Re, A. Doan, et al.: Tuffy: Scaling up Statistical Inference in Markov Logic Networks using an RDBMS, VLDB 2011
- X. Ling, D.S. Weld: Temporal Information Extraction. AAAI 2010
- Y. Wang, M. Zhu, L. Qu, M. Spaniol, G. Weikum: Timely YAGO: harvesting, querying, and visualizing temporal knowledge from Wikipedia. EDBT 2010
- Y. Wang, L. Qu, B. Yang, M. Spaniol, G. Weikum: Harvesting Facts from Textual Web Sources by Constrained Label Propagation. CIKM 2011

## **Thank You!**

# **Thank You!**

























