##  informatik

# What Computers Should Know, Shouldn't Know and Shouldn't Believe 

## Gerhard Weikum

Max Planck Institute for Informatics and Saarland University http://mpi-inf.mpg.de/~weikum

## Turn Text \& Data into Knowledge



## Web of Knowledge and Open Data

## > 100 Billion subject-predicate-object facts from > 1000 sources



## Web of Knowledge



## Machine Knowledge for Answer Engines

Precise and concise answers for advanced information needs:

properties of entity
Nobel laureate who outlived two world wars and all his children?
$\star$ Politicians who are also scientists?

sets of entities
relationships between entities
Commonalities \& relationships among:
T.E. Lawrence, Gandhi, Alan Turing, June Carter?


## Machine Knowledge for Answer Engines

Precise and concise answers for advanced information needs:

Proteins that bind to the Zika virus?
real applications
Antidepressants that interfere with thyroid drugs?
Polymer materials for super-capacitators?
German philosophers influenced by William of Ockham?
Green politicians mentioned in Panama Papers?

## Outline

$\checkmark$ Introduction
$\star$ What Computers Should Know
$\star$ What Computers Shouldn't Know
$\star$ What Computers Shouldn't Believe
$\star$ Conclusion

## Missing Predicates \& Salient Facts

## Which interesting facts about an entity are in infoboxes?



Died

Cause of death Resting place

Occupation

Years active Spouse(s)
J. R. Cash

February 26, 1932
Kingsland, Arkansas, U.S.
September 12, 2003
(aged 71)
Nashville, Tennessee, U.S.
Diabetes mellitus
Hendersonville Memory Gardens

Singer-songwriter • guitarist • actor - author

1954-2003
Vivian Liberto (m. 1954;
div. 1966)

June Carter (m. 1968; her
death 2003)

Military career

| Allegiance | America |
| :--- | :--- |
| Service/branch | Force |
| Years of |  |
| service |  |
| Rank | $1950-1954$ |

romanticRelationship(Johnny Cash,JuneCarter, 1961-2003)featuredInMovie(romanticRelationship(...),WalkTheLine))
playedConcertAtLocation (JohnnyCash, SanQuentinStatePrison) covered (JohnnyCash, One by U2)
covered (JohnnyCash, MercySeat by NickCave)

## Missing Predicates \& Salient Facts

## Which interesting facts about an entity are in infoboxes?



Studio album by Nick Cave and the Bad Seeds

Released 20 September 2004
Recorded March-April 2004 at Studio Ferber in Paris, France

Genre Altemative rock
Length
82:30
Label Mute
Producer Nick Launay


Nick Cave and the Bad Seeds chronology

| Nocturama | Abattoir |  |
| :---: | :---: | :---: |
| (2003) | Blues $/$ The | Rarities |
|  | Lyre of | (2005) |
|  | Orpheus |  |
|  | (2004) |  |

Singles from Abattoir Blues / The Lyre of Orpheus

1. "Nature Boy"

Released: 6 September 2004
2. "Breathless / There She Goes, My

Beautiful World"
Released: 15 November 2004
3. "Get Ready for Love"

Released: 14 March 2005
isOnAlbum
(Let the Bells Ring, AbbatoirBlues)
lyricsAbout
(Let the Bells Ring, JohnnyCash) isOnAlbum (O‘Children, AbbatoirBlues playedInMovie (O‘Children, HarryPotter 6)

# Spectrum of Digital Knowledge (1): School Education for Computers 

## taxonomic knowledge:

type (SteveJobs, entrepreneur), subtypeOf (entrepreneur, businessperson) type (SteveJobs, inventor), subtypeOf (inventor, human) type (SteveJobs, YogaPractitioner), type (SteveJobs, GratefulDeadFan)

long-tail classes

## factual knowledge:

hasFounded (SteveJobs, Apple), CEO (SteveJobs, Apple)
long-tail entities hasDaughter (SteveJobs, LisaBrennan), namedAfter (AppleLisa, LisaBrennan) hasFavoriteSong (SteveJobs, Imagine), hasFavoriteSong (SteveJobs, Truckin') dated (SteveJobs, JoanBaez), admired (SteveJobs, BobDylan) composed (JoanBaez, Diamonds\&Rust), IyricsAbout (Diamonds\&Rust, BobDylan) sangAt (JoanBaez, memorialForSteveJobs)

## spatial \& temporal knowledge:

diedOn (SteveJobs, 5-Oct-2011), diedln (SteveJobs, Palo Alto)
happened (hasFounded (SteveJobs, Apple), Cupertino, 1976) validDuring (CEO (SteveJobs,Apple), 1997-2011)

# Spectrum of Digital Knowledge (2): Kindergarten and University 

## commonsense properties:

property (lemon, yellow), property (lemon, juicy), property (lemon, sour), ability (fish, swim), ability (human, speak), usedFor (classroom, teaching), maxHeight (human, 2.5 m ), maxLength (snake, 10 m )
commonsense rules:
$\forall \mathbf{x}$ : human $(\mathrm{x}) \Rightarrow(\exists \mathrm{y}$ mother $(\mathrm{x}, \mathrm{y})) \wedge(\exists \mathrm{z}$ father $(\mathrm{x}, \mathrm{z}))$
$\forall \mathbf{x}, \mathbf{y}, \mathbf{z}$ mother $(\mathbf{x}, \mathbf{y}) \wedge$ mother $(\mathbf{x}, \mathbf{z}) \Rightarrow \mathbf{y}=\mathbf{z}$
domain-specific expert knowledge:
type (Ubiquinone-8, coenzyme), expresses (COQ8, Ubiquinone-8) causes (lack of Ubiquinone-8, mitochondrial disorder)

# Spectrum of Digital Knowledge (3): Learned in Life 

## socio-cultural and social knowledge:

invented (Internet, Cerf and Kahn, techies), invented (Internet, Berners-Lee, common people), invented (Internet, AI Gore, mid-aged democrat voters) drink (beer, Germany), drink (wine, California), drink (lassi, India) alleviates (ice, bruises), alleviates (eucalyptusOil, sinusitis)
belief knowledge:
believe (Ptolemy, center (world,earth), believe (Galileo, center (world, sun)) believe (Chinese, badLuckNumber (4)), believe (Germans, badLuckNumber (13)) believe (AustralianAborigines, taboo (photosOfDeadPeople))

## process knowledge:

type (changeTire, mechanicalTask)
subtask (changeTire, loosenBolts), subtask (changeTire, liftCar), requires (loosenBolts, spiderWrench), requires (liftCar, jack) precedes (loosenBolts, liftCar)

## Knowledge Gaps

Temporal and Spatial Knowledge
Long-Tail Knowledge (on types and entities)
Dynamic Knowledge (events, emerging entities)
Open-Ended Knowledge (relation types)
On-the-Fly Knowledge
Visual Knowledge (on types and long-tail entities)
Cultural Knowledge
Commonsense Knowledge
Social Knowledge
Intensional Knowledge
Negative Knowledge

## Commonsense Knowledge: Not So Common

## Every child knows that

apples are green, red, round, juicy, ...
but not fast, funny, verbose, ...
pots and pans are in the kitchen or cupboard, on the stove, ... but not in in the bedroom, in your pocket, in the sky, ...
children usually live with their parents
But: commonsense is rarely stated explicitly
Plus: web and social media have reporting bias
color of elephants ?
pink elephant: 0.9 Mio on Google grey elephant: 0.4 Mio on Google
rich family: 27.8 Mio on Bing poor family: 3.5 Mio on Bing

singers: 22.8 Mio on Bing workers: 14.5 Mio on Bing

# Pattern-based Harvesting of Commonsense Properties 

Approach: Start with seed facts
hasProperty (apple, round)
hasAbility (dog, bark)
hasLocation (plate, table)
Learn patterns that express these relations, such as $X$ is very $Y, X$ can $Y, X$ put in/on $Y, \ldots$

Apply patterns to Web, books, N-grams corpora, image tags, etc. $\rightarrow$ statistics, semisupervised learning, constraint reasoning hasColor (elephant, grey), hasShape (circle, round) ... hasAbility (fish,swim), hasAbility (human, talk) ... WebChild KB: usedFor (book, learn), usedFor (computer, learn) 5 Mio. assertions partOf (wheel, bike), partOf (wheel, car) ... hasTemperature (oven, hot), hasTaste (chili, hot) sense-disambiguated [WSDM 2014]

## Commonsense \& Visual Contents

[N. Tandon et al.: WWW‘15, CIKM‘15, AAAI'16, WWW‘17]


Refined part-whole relations from web\&books text and image tags
$\rightarrow 6.7$ Mio sense-disambiguated triples for physicalPartOf, visiblePartOf, hasCardinality, memberOf, substanceOf

## driveHerd:

location (ranch, outback) $\because$ time (daytime) scripts, aligned with visual scenes
$\rightarrow 0.5$ Mio activity types with attributes: location, time, participants, prev/next

## Commonsense for Language Understanding

- How-to queries:
repair a bike tire, pitch a tent, cross a river, ...
- Scene search (over videos, books, ): romantic dinner, dramatic climb, ...
- Question disambiguation:
jordan weather forecast
Germany's keeper at the Euro ?
- Sentiment analysis:
the hottest place on earth
the hottest bar in town


## Commonsense for Computer Vision



## woman (01)

? lemon (02)
racket (o3)

+ commonsense knowledge:
hasColor (tennisball, yellow) hasShape (tennisball, round) occursAt (tennisball, tennis court) occursAt (racket, tennis court) $\Rightarrow$ tennisball(o2)


## Commonsense for Image Description


dog, man, woman, motorbike
$\rightarrow$ red dog in front of man and woman on a motorbike

+ commonsense knowledge:
$\rightarrow$ romantic Australia
$\rightarrow$ the wanderer of the Pilbara



## Outline

$\checkmark$ Introduction
$\checkmark$ What Computers Should Know
$\star$ What Computers Shouldn't Know
$\star$ What Computers Shouldn't Believe
$\star$ Conclusion

## What Computers Know About You (But Shouldn't ?)



Nobody interested in your research? We read your papers!


Boyfriend Tracker Free
Android Aplicativos Ponto Com - July 30, 2014
Tools
Install
Add to Wishlist
A. You don't have any devices

*     *         * ( - - 35)



Social Web Analysis


Social Intelligence Risk Score

## Goal: Privacy Risk Advisor

Users are tracked, linked, targeted, rated ...
Algorithmic decisions made with this data



Background Check Report
on marc spaniol

| O Full Name | O Property Records |
| :---: | :---: |
| - Current Addres | O Bankrupries |
| O Prone Numbers | - juggnents 8 Liens |
| - Prior Residences | - Marriage $\frac{\text { Divorce }}{}$ |
| - Relatives | - Birth \& Death Records |
| - Aliases 8 akas | O Misdemeanors |
| - Age s oce | - Criminal Cheok |
| $\bigcirc$ Neightors | - SexOffenser fucors |
| Umited Time Oflerl |  |
| Report - Trial | continue |
| $349.95 \quad 39.95$ |  |

Established privacy models

- Data: single database

- Adversary: computationally powerful, but agnostic
- Goal: anonymity guarantee
- Measures: data coarsening, perturbation, limit queries


## Today's user behavior \& risks

- Data \& User: multimodal, social, agile, longitudinal
- Adversary: world knowledge \& probabilistic inference
- Goal: alert \& advise, bound risk
- Measures: guide user on policies, selective anonymization


## Ranking-based Risk Assessment

Quora
Q Search for questions, people, and topics

FMLA Pregnancy Sex Mesicine and Heathcare Jobs and Careers Career Advice

Should I tell prospective employers that I am pregnant?
I am being laid off from a large col amy63628
in July for the birth of my child. I to know since I just got a good re and telling them I will be available debating whether or not to tell the do not have to tell them, and they honest and simply making sure I

4 Answers

## - Patient

Fluoxetine symptoms??
Hi everyone, So on friday i was dignosed with clincal depression and PTSD and was put on fluoxetine starting at 2.5 ml (to be increased gradually because i'm quite young) So yesterday $i$ was really hyper but $i$ was excited because i was going to see a concert later in the night so i'm going to put

## Asia

 Biega Gummadisocial media, online communities, review forums, query logs,

Exposure Risk [J.A. Biega et al.: SIGIR‘16]:

1. Capture sensitive topics (depression, gambling, debts, ...) with knowledge on vocabulary \& correlations $\rightarrow$ latent topic models (LDA, word2vec skip-gram)
2. Estimate user exposure per topic:
$\rightarrow$ quantitative risk measures
3. Rank users: compute top-k targets
$\rightarrow$ alert user, recommend selective anonymization

## Outline

$\checkmark$ Introduction
$\checkmark$ What Computers Should Know
$\checkmark$ What Computers Shouldn't Know
$\star$ What Computers Shouldn't Believe
$\star$ Conclusion

## Social Knowledge: "Folk Wisdom"

Beliefs become like facts
in specific socio-cultural contexts
invented (computer, Eckert and Mauchley, USA)
invented (computer, KonradZuse, Germany)
invented (computer, AlanTuring, UK)
invented (computer, SteveJobs, young nerds)
invented (computer, DonaldTrump, twitter)
Knowledge distilled from subjective behavior / experience of many people

Need to tap riskier sources (social media posts etc.)

- common habits on food, drinks, social behavior
- experience with electric/diesel/green cars, on the road
- grassroots insight on health, symptoms, therapies, ...


# Credibility and Trust: Where the Truth Lies 



Assess credibility of statements / claims on the Internet and the trustworthiness of their sources

- Search results:
- Biased news:
- KB contents:
- Social media:
- Health forums:
- Press conference:
love affairs of Hillary Clinton?
Merkel hates Greece
Berners-Lee \& AI Gore invented the Internet Obamacare requires microchip implant
Xanax causes hallucinations
Trump proves that CNN has only fake news
facts ......bias ...... post-truth ...... manipulation ...... fake
$\rightarrow$ Polygraph for digital contents: for IE, KB, analytics, users


## Between Facts and Fake

## Post-truth politics

From Wikipedia, the free encyclopedia

Post-truth politics (also called post-factual politics) is a political culture in which debate is framed largely by appeals to emotion disconnected from the details of policy, and by the repeated assertion of talking points to which factual rebuttals are ignored. Post-truth differs from traditional contesting and falsifying of truth by rendering it of "secondary" importance. While this has been described as a contemporary problem, there is a possibility that it has long been a part of political life, but was less notable before the advent of the Internet. In the novel Nineteen Eighty-Four. George Onwell cast a world in which the state changes historic records daily to fit its propaganda goals of the day.

Political commentators have identified post-truth politics as ascendant in Russian, Chinese, American, Australian, British, Indian, Japanese and Turkish politics, as well as in other areas of debate, driven by a combination of the 24hour news cycle, false balance in news reporting, and the increasing ubiquity of social media. ${ }^{[1][2][3][4][5][6]}$ In 2016, "post-truth" was chosen as the Oxford Dictionaries' Word of the Year ${ }^{[7]}$ due to its prevalence in the context of that year's Brexit referendum and U.S. presidential election. ${ }^{[8[9]}$

This article is part of a series on
Misinformation and disinformation


General information
[hide]
Big lie • Black propaganda - Circular sourcing • Deception • Dog-whistling • Doublespeak • Echo chamber • Euphemistic misspeaking • False flag • Filter bubble • Factoid • Fallacy $\cdot$ Fake news (by country • online) • Gaslighting • Half-truth • Hoax • Ideological framing . Internet manipulation • Media manipulation -
Post-truth • Propaganda - Quote mining •
Scientific fabrication $\cdot$ Social bot $\cdot$ Spin

## Credibility \& Trust in Health Communities

[S. Mukherjee et al.: KDD'14, CIKM‘15]

I took the whole med cocktail at once. Xanax gave me wild hallucinations and a demonic feel.

p1


## Credibility \& Trust in Health Communities

[S. Mukherjee et al.: KDD'14, CIKM‘15]

I took the whole med cocktail at once. Xanax gave me wild hallucinations and a demonic feel.

Xanax and Prozac are known to cause drowsiness.

Xanax made me dizzy and sleepless.

Language Objectivity p1

s2
Statement Credibility
joint reasoning with probabilistic graphical model (semi-supervised heterogeneous CRF with EM-style learning)

## Semi-Supervised Heterogeneous MRF

## for Joint Inference on Posts, Users, Statements

Post language style: modals, emotion adject., first person pronouns, inferential conjunctions, etc.

User activities: \#posts, \#replies, \#thanks received, etc.


Statement training labels: true,false

## Experiments with Health Community Data

[S. Mukherjee et al.: KDD‘14]
Data: healthboards.com
850,000 users, 4.5 mio. posts sampled 15,000 users, 2.8 mio. posts

Ground-truth: expert data from MayoClinic portal focused on 6 widely used drug families

Inferring rare side effects of drug families


■Freq. ■SVM $■$ SVM Dist. SV. ■ CRF

Inferring expert users on drug families


- Thanks ■CRF


## Credibility of Textual Claims

## Jack Ma is the new owner of AC Milan

[K. Popat et al.: CIKM‘16, WWW‘17]

## $\square$ True or False ?

## search Web for evidence

GAMEON


Chinese investors want to buy the legendary AC Milan soccer club from Silvio Berlusconi
By zheoping Hanag May 10,2016


## Credibility of Textual Claims


joint reasoning on credibility of claim and
authority of supporting/refuting sources


## Distantly Supervised Model

[K. Popat et al.: WWW 2017]


# Temporal Behavior of Claims \& Evidence 

[K. Popat et al.: CIKM‘16, WWW‘17]

consider gradients

- Model temporal patterns of spreading claims, support and denial
- Exploit for early assessment of credibility


## Temporal Footprint of Claims

[K. Popat et al.: WWW 2017]


The Centers For Disease Control confirmed that a patient in Dallas has tested positive for Ebola.


Actor Macaulay Culkin has died.


The iPhone 6 Plus will bend easily if placed in a pocket.

## Research Opportunities:

- Understand \& incorporate the influence of attribution in sources and dissemination across sources
- Consider anonymous sources
(e.g. user posts in online communities)
- Explain veracity of claims in socio-cultural context


## Outline

## Introduction

$\checkmark$ What Computers Should Know
$\checkmark$ What Computers Shouldn't Know
$\checkmark$ What Computers Shouldn't Believe
$\star$ Conclusion

## Take-Home Message

more knowledge, analytics, insight

Computers shouldn't believe everything

knowledge acquisition
machine reading

# Digital <br> Knowledge 

Computers need socio-cultural knowledge

You should know what computers know about YOU!


