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Beyond Knowledge Graphs: New Frontiers of Machine Knowledge

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Outline

★ **1 Knowledge Graphs ...**

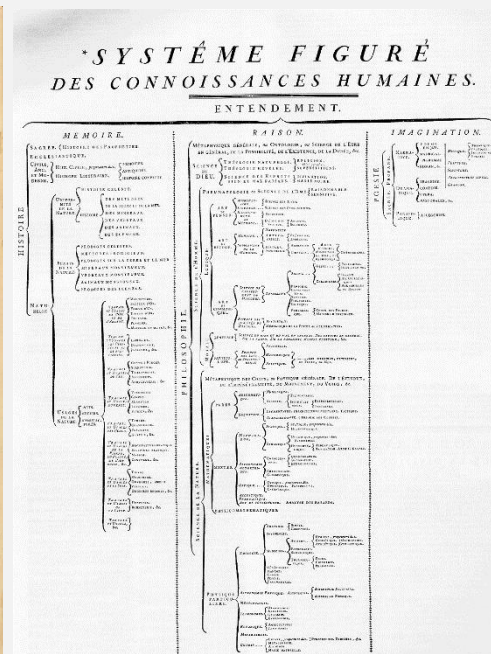
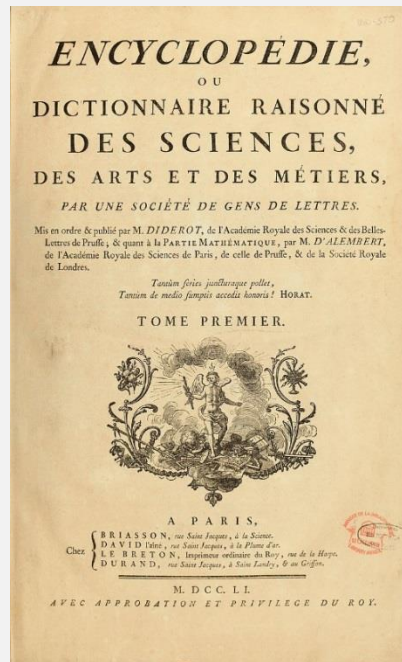
★ **2 ... And Beyond**

★ **Conclusion**

Brief History of Knowledge Bases



1751-1780

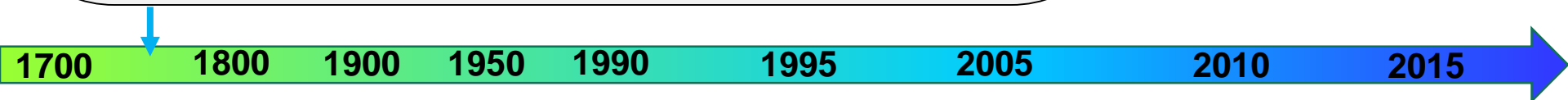


Didier Diderot



Jean Le Rond d'Alembert

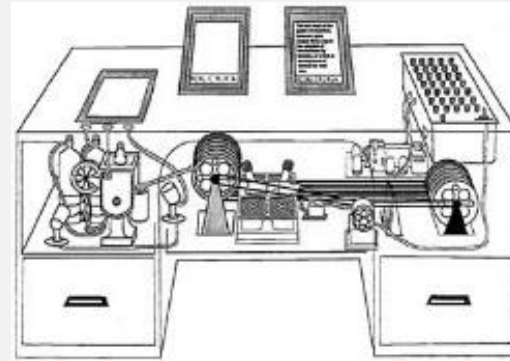
72000 articles, 2250 contributors



Brief History of Knowledge Bases



Vannevar
Bush
1945



... a growing **mountain of research** ...
a **memex** is a ... **supplement to memory**



Denis Diderot Jean d'Alembert



Brief History of Knowledge Bases



Cyc



**Douglas
Lenat
1984**

$\forall x: \text{human}(x) \Rightarrow$
 $(\exists y: \text{mother}(x,y) \wedge$
 $\exists z: \text{father}(x,z))$

$\forall x,u,w: (\text{mother}(x,u) \wedge$
 $\text{mother}(x,w)$
 $\Rightarrow u=w)$

**Vannevar
Bush**



**Denis
Diderot** **Jean
d'Alembert**



Brief History of Knowledge Bases

WordNet



George Miller **Christiane Fellbaum**

1985

guitarist \subset
{player, musician}
 \subset **artist**
{player, footballer}
 \subset **athlete**

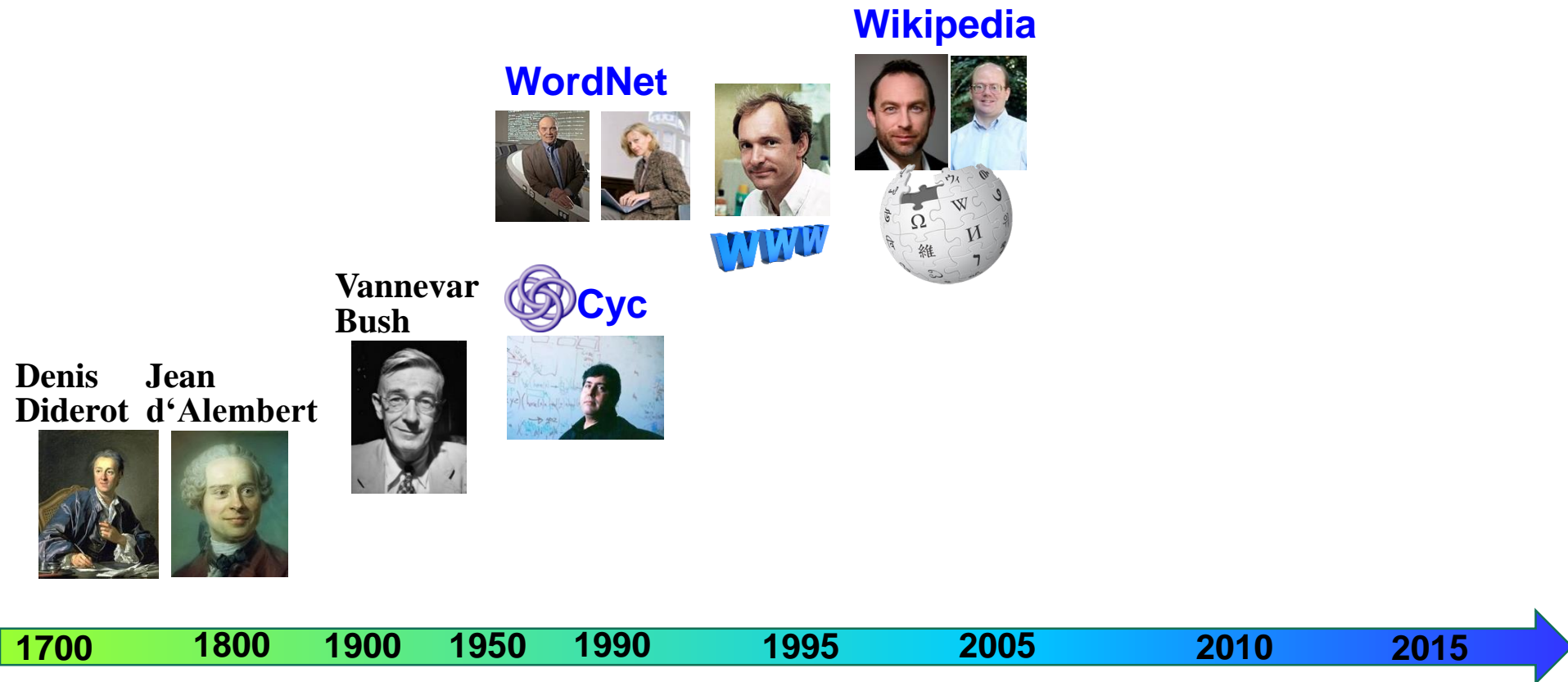
Vannevar Bush



Denis Diderot **Jean d'Alembert**



Brief History of Knowledge Bases

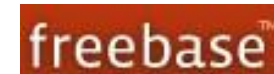


Brief History of Knowledge Bases

WolframAlpha



IBM Watson



Wikipedia



WordNet



Vannevar Bush



Denis Diderot



Jean d'Alembert



Encyclopedic Machine Knowledge

subject-predicate-object triples about entities,
attributes of and relations between entities

plus reification

predicate (subject, object)

type (SteveJobs, entrepreneur)

subtypeOf (entrepreneur, businessperson)

hasFounded (SteveJobs, Apple)

hasDaughter (SteveJobs, LisaBrennan)

diedOf (SteveJobs, pancreatic cancer)

diedOn (SteveJobs, 5-Oct-2011)

hasSymptom (pancreatic cancer, jaundice)

treats (ErlotinibHydrochloride, pancreatic cancer)

foundingEvent (123, blank)

founder (123, SteveJobs)

foundedOrg (123, Apple)

foundedOn (123, 1-April-1976)

foundedAt (123, LosAltos-CA-USA)

taxonomic knowledge

factual knowledge

expert knowledge

spatial & temporal knowledge

Mio's of entities & Bio's of facts

→ semantic search & QA

→ language understanding

→ text analytics, data cleaning ...

Machine Knowledge for Answer Engines

**Precise and concise answers
for advanced information needs:**



properties of entity

★ **Nobel laureate whose daughter also won a Nobel prize?**

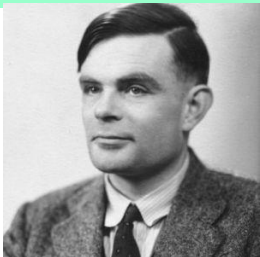


sets of entities

★ **Pop singers who are also poets?**

relationships between entities

★ **Commonalities & relationships among:
Alan Turing, Paul Bocuse, Steve Jobs, Katherine Goble?**



Machine Knowledge for Answer Engines

**Precise and concise answers
for advanced information needs:**

real applications

- ★ **Proteins that bind to the Zika virus?**
- Polymer materials for super-capacitators?**
- European politicians mentioned in Panama Leaks?**

Conversational Chatbots



My name is Jerry, cool to meet you Mitsuku.



Cool is a nice middle name. Why did your parents call you that?



What is the highest mountain on earth?



Mt. Everest. It's 8848 metres high.



Isn't the K2 higher?



A large mountain in the Himalayas.



How high is the Eiffel tower?



It is about 324 metres.



And in feet?



Oh I don't know. We use the metric system now.



Which is higher the Eiffel tower or the Everest?



The one closest to the sky.



What is taller a giraffe or a cat?



Are you serious, a giraffe is a lot bigger than a cat.



What is longer, a king cobra or a worm?



The cobra as it has 8 letters.

+ Language Skills

+ Encyclopedic Knowledge

? Quantities

? Commonsense

<https://pandorabots.com/mitsuku/>

Visual AI: from Perception to Cognition

Computer Vision → Image Description → Question Answering



man in blue wetsuit
is surfing on a wave

Source: <https://cs.stanford.edu/people/karpathy/deepimagesent/>



woman is holding
a bunch of bananas

Who is wearing glasses?



man



woman

Source: <http://www.visualqa.org/>

What is the
polar bear eating?



carrot

? Quantities

? Commonsense



What is behind the giraffe?

→ mountain

What is on the mountain peak?

→ snow

What is the largest object
in this image?

→ giraffe

How many legs does
the giraffe have?

→ 3

Outline

★ 1 Knowledge Graphs ...

★ 2 ... And Beyond

★ Quantities

★ Commonsense

★ Conclusion

Quantities: More than just Numbers

Counts

1200 submissions, 200 accepted

Fractions

5% of the world population, 30% of the guns

Rates

FB increased revenue by 30%

Dates

24-April-2018, today, last weekend, Easter

Money

acquired for 80 Mio USD

Time

for 3 years, new record 9.58 s

Scores

Brazil vs. Germany 1:7

Ranks

4th in the Olympics final

Physical Measures

407 MW renewable energy

3400 kWh annual energy consumption

Medical Measures

15mg Xarelto daily

Potassium low: 2 mmol/L

Quantities: Measure, Value, Unit

Quantity = (Measure, Value, Unit)

(Distance, 200, km)

(Height, 1.80, m)

(Power, 407, MW)

(Energy, 3400, kWh)

(Price, 80,000,000, USD)

normalize: scale & precision

2E5: scale 10^5 prec 1

1.80E0: scale 10^0 prec 3

4.07E8: scale 10^8 prec 3

3.4E6: scale 10^6 prec 2

8.0E7: scale 10^7 prec 2



Quantity meta-properties:

- typical / normal range
- value distribution
- unit conversions

Reference systems:

SI: International System of Units

UNECE: Codes for Units of Measurement used in the International Trade

Kilowatt hour



Residential [electricity meter](#) located in Canada

Unit information	
Unit system	Non-SI metric
Unit of	Energy
Symbol	kW·h

Unit conversions	
1 kW·h in is equal to ...
SI units	3.6 MJ
CGS units	3.6×10^{13} erg
English Engineering units	$\approx 2,655,224$ ft·lbf
British Gravitational units	$\approx 85,429,300$ ft·pdl

Quantities in Knowledge Bases

Usain Bolt (Q1189)

height



195 centimetre

► 1 reference

mass



94 kilogram

► 1 reference

Tesla Model X (Q1634161)

height



1,684 millimetre

► 1 reference

width



1,999 millimetre

► 1 reference



Men's 100 metres

[Event History](#) · [Glossary](#) · [SHARE](#) · [Embed](#) · [CSV](#) · [Export](#) · [PRE](#)

Games	Age	City	Sport	Country	Phase	Unit	Rank	T(A)	RT
2008 Summer	21	Beijing	Athletics	Jamaica	Final		1	WR 9.69	0.165
2008 Summer	21	Beijing	Athletics	Jamaica	Semi-Finals	Heat One	1	Q 9.85	0.161
2008 Summer	21	Beijing	Athletics	Jamaica	Quarter-Finals	Heat Four	1	Q 9.92	0.165
2008 Summer	21	Beijing	Athletics	Jamaica	Round One	Heat One	1	Q 10.20	0.186
2012 Summer	25	London	Athletics	Jamaica	Final		1	9.63	0.165
2012 Summer	25	London	Athletics	Jamaica	Semi-Finals	Heat Two	1	Q 9.87	0.180
2012 Summer	25	London	Athletics	Jamaica	Round One	Heat Four	1	Q 10.09	0.178
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Final		1	9.81	0.155
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Semi-Finals	Heat Two	1	Q 9.86	0.143
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Round One	Heat Seven	1	Q 10.07	0.156

Men's 200 metres

[Event History](#) · [Glossary](#) · [SHARE](#) · [Embed](#) · [CSV](#) · [Export](#) · [PRE](#)

Games	Age	City	Sport	Country	Phase	Unit	Rank	T(A)	RT	L
2004 Summer	17	Athina	Athletics	Jamaica	Round One	Heat Four	5	21.05	0.254	5
2008 Summer	21	Beijing	Athletics	Jamaica	Final		1	WR 19.30	0.182	
2008 Summer	21	Beijing	Athletics	Jamaica	Semi-Finals	Heat Two	1	Q 20.09	0.175	
2008 Summer	21	Beijing	Athletics	Jamaica	Quarter-Finals	Heat One	1	Q 20.29	0.186	
2008 Summer	21	Beijing	Athletics	Jamaica	Round One	Heat Five	2	Q 20.64	0.177	
2012 Summer	25	London	Athletics	Jamaica	Final		1	19.32	0.180	
2012 Summer	25	London	Athletics	Jamaica	Semi-Finals	Heat Two	1	Q 20.18	0.192	
2012 Summer	25	London	Athletics	Jamaica	Round One	Heat One	1	Q 20.39	0.191	
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Final		1	19.78	0.156	
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Semi-Finals	Heat Two	1	Q 19.78	0.156	
2016 Summer	29	Rio de Janeiro	Athletics	Jamaica	Round One	Heat Nine	1	Q 20.28	0.177	



2016 Tesla Model X fuel economy and operating costs

Model	Model year	Fuel efficiency (MPGe)		
		Combined	City	Highway
AWD 90D (90 kWh) ^[12]	2016	92; 34 kWh/100 mi or 21 kWh/100 km	90; 37 kWh/100 mi or 23 kWh/100 km	94; 32 kWh/100 mi or 20 kWh/100 km
AWD P90D (90 kWh) ^{[47][48]}	2016	89; 38 kWh/100 mi or 24 kWh/100 km	89; 38 kWh/100 mi or 24 kWh/100 km	90; 38 kWh/100 mi or 24 kWh/100 km
AWD P100D (100 kWh)	2016	86; 38 kWh/100 mi or 24 kWh/100 km	81; 38 kWh/100 mi or 24 kWh/100 km	92; 38 kWh/100 mi or 24 kWh/100 km

Quantities in Relations

[Sarawagi et al.: AAAI'16,
Mausam et al.: ACL'17,
Alonso et al.: ICWE'15,
Polleres et al.: ISWC'16
Ibrahim et al.: CIKM'16]

Extracting quantities from text & tables

Bolt won the 200m Olympics in 2008 in **19.30s**

Bolt won 100 meters Gold in Rio in **9.81 seconds**

Xarelto is usually applied after thromboses at **20mg twice per day**

Xarelto for children should be limited to a **daily** dosage of **5-15 mg**

- hasWon (UsainBolt, OlympicGold, 2008, 200m, 19.30s)
- hasWon (UsainBolt, OlympicGold, 2016, 100m, 9.81s)
- treatment (Xarelto, thrombosis, adults, 40mg/day)
- treatment (Xarelto, thrombosis, children, 15mg/day)

Problems:

- **n-ary relations** expressed in more complex language
- **partial observations** of facts from multiple sentences

Quantities in Relations



[P. Ernst et al.:
WWW'18]

Extracting tuples with quantities from text & tables

Problems:

- n-ary relations expressed in more complex language
- **partial observations of facts** from multiple sentences

Bolt won the 200m Olympics in 19.30 seconds

Bolt won second Gold in 2008 with 19.30s record

Bolt ran the 100m final in Rio in 9.81

Bolt won 100m Gold in Rio

-
- hasWon (UsainBolt, OlympicGold, ?X, 200m, 19.30s)
 - hasWon (UsainBolt, OlympicGold, 2008, ?Y, 19.30s)
 - hasWon (UsainBolt, ?W, 2016, 100m, 9.81s)
 - hasWon (UsainBolt, OlympicGold, 2016, 100m, ?Z)

Solution: use **Weighted MaxSat** for constraint reasoning
over **partially grounded fact candidates**

Quantity Knowledge: Challenges

Representation:

- more triples, n-ary relations, nested tuples?
emission (TeslaS, California, 2016, measure(0.25, kg/km, CO2))
- meta-properties?
max(height(giraffe),5m)
unit (distance(city,city),km), unit (distance(star,star),lightyears)

Acquisition:

- human experts, crowdsourcing, learning from tables/stats?
- integrate existing reference systems (SI, medicine, finance, ...)

Usage:

- search with quantities
car models with emission below 0.2 kg/km CO2
- aggregation & comparison for analytics
world-wide avg. dosage of beta blockers per age group

Outline

★ 1 Knowledge Graphs ...

★ 2 ... And Beyond

★ Quantities

★ Commonsense

★ Conclusion

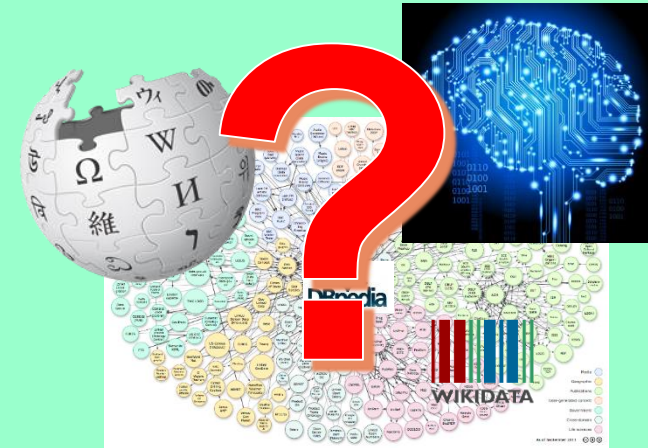
Commonsense Knowledge: Not So Common

Every child knows that

apples are green, red, round, juicy, ...
but not fast, smart, funny, verbose, ...

children live with their parents
programmers need computers

... ..



But: commonsense is rarely stated explicitly
reporting in web and social media is **sparse** and **biased**

spider 3 legs: 7 Mio
spider 8 legs: 1 Mio

singer: 236 Mio
programmer: 40 Mio
manager: 570 Mio

pink elephant: 16 Mio
grey elephant: 9 Mio



Commonsense Knowledge Bases

John McCarthy
(Turing Award
1971)



“Commonsense knowledge
is key to generality in AI”

- **Properties of Everyday Objects**
- **Human Activities / Abilities / Emotions**
- **Plausibility Invariants (incl. Causality ...)**
- **Class/Concept Taxonomy**



Cyc



Douglas Lenat

$$\begin{aligned} \forall x: \text{human}(x) \Rightarrow \\ & (\exists y: \text{mother}(x,y) \wedge \\ & \quad \exists z: \text{father}(x,z)) \\ \forall x,u,w: (\text{mother}(x,u) \wedge \\ & \quad \text{mother}(x,w) \\ & \quad \Rightarrow u=w) \end{aligned}$$

WordNet



George
Miller



Christiane
Fellbaum

guitarist \subset
{player,musician}
 \subset **artist**
{player,footballer}
 \subset **athlete**

Acquiring Commonsense Knowledge

Approach 1: **Knowledge engineers**

→ Cyc, WordNet

Problem: scope and scale

Approach 2: **Crowdsourcing**

→ ConceptNet [Speer&Havasi],
VisualGenome [Krishna et al.], ...

Problem: scale and quality

Approach 3: **Information Extraction**

→ WebChild [Tandon et al.],
AI2 Alexandria [Etzioni et al.]

Problems: noise and bias

many specific approaches for subclassOf (hypernymy)

WebChild

[N. Tandon & G. de Melo et al.: WSDM'14, AAAI'16, ACL'17]



Use seeds

hasProperty (apple,round), **hasAbility** (fish,swim), **hasLoc** (fish,water)
to learn patterns

X is very Y, X can Y, X is in Y, ...

for refined properties **hasLoc**, **hasColor**, **hasShape**, **hasSize**, **hasTaste** ...

Overcome sparseness and bias by
tapping into **books (n-grams)** and **image tags**



hasColor (elephant, grey), **hasShape** (circle, round)
hasAbility (fish, bite), **hasAbility** (human, talk) ...
usedFor (book, learn), **usedFor** (computer, learn) ...
partOf (wheel, bike), **partOf** (wheel, car) ...
hasTemp (weather, crisp), **hasTaste** (cookie, crisp) ...

... ..

2 Mio concepts
18 Mio. assertions
for 30 relations
• semantically typed
• sense disambiguated

<https://gate.d5.mpi-inf.mpg.de/webchild/>

WebChild: Limitations

Highly varying confidence:

hasColor (apple, ...): red, green, yellow, pink, purple, black, white ...

hasTaste (apple, ...): sweet, sour, juicy, chewy, healthy, cheap ...

hasProperty (mountain, ...): high, steep, cold, lethal, far, holy, true ...

Consider **epistemic logic with modalities**:

always, usually, possibly, never

Subjective properties:

hasAppearance (kangaroo, cute)

hasProperty (spider, dangerous)

hasProperty (grasshopper, edible), hasProperty (FoieGras, edible)

Consider **belief logics**:

fromFrance(x) \Rightarrow belief (x, hasProperty (FoieGras, edible))

\neg fromAustralia(x) \Rightarrow belief (x, hasAppearance (kangaroo, cute))

Human Activities & Behaviors



climb mountain	
Agent:	...
Participant:	...
Object:	...
Location:	...
Time:	...

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

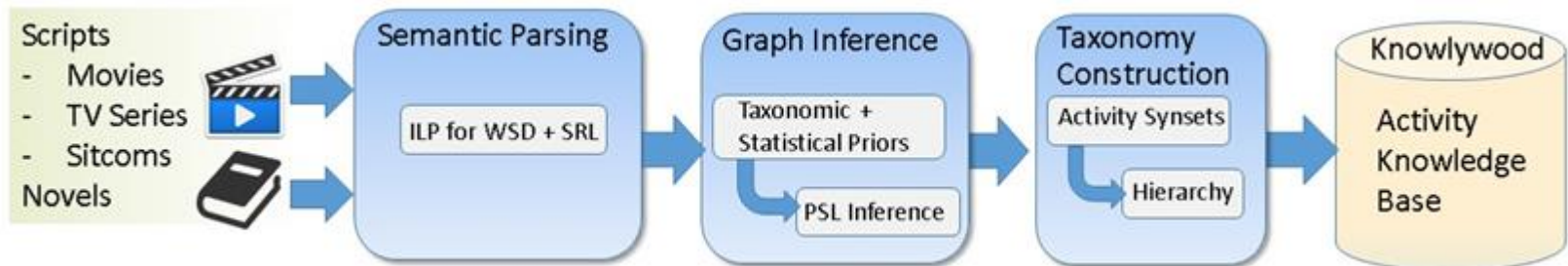
reach summit	
Agent:	...
Participant:	...
Object:	...
Location:	...
Time:	...

Type

climb on ice	
Agent:	mountaineer, man, woman
Participant:	guide, sherpa
Object:	rope, crampons, oxygen mask
Location:	outdoor, mountain peak
Time:	daytime, sunshine

Next

Tap narratives & movie scripts → Knowlywood project



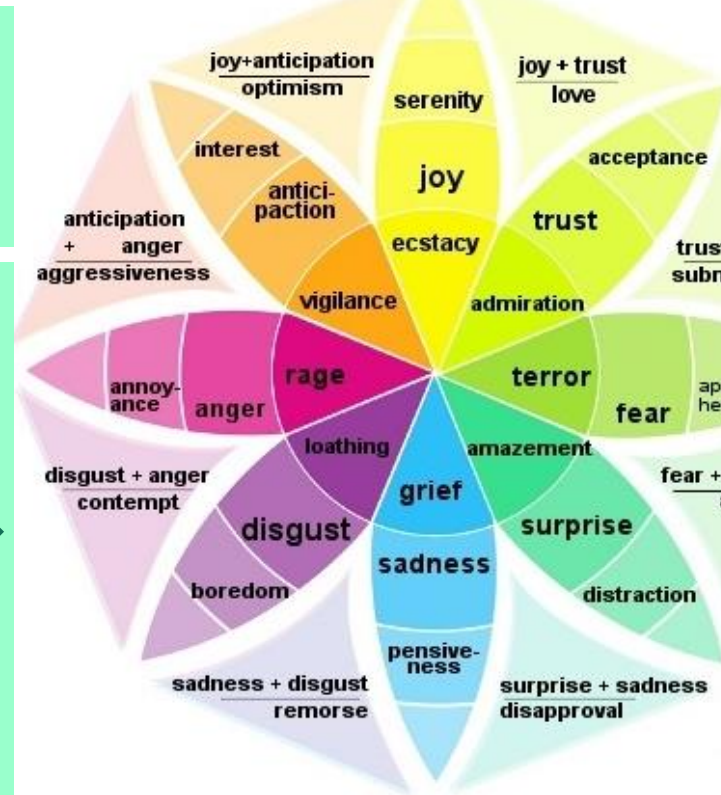
Human Emotions

[H. Jhavar & Paramita: WWW'18]



I got **invited to** give a talk at the Wiki workshop

I got **invited to** this stupid dinner with my boss



HappyDB:
100,000 happy moments [Asai et al.: LREC'18] :
I got invited to a friend's wedding.
I went to the park with the kids.
Morning started with chirping of birds.

**Plutchik's wheel
of emotions**

Source:

<https://www.cheatography.com/davidpol/cheat-sheets/plutchik-s-wheel-of-emotions/>

Socio-Cultural Knowledge

Problem:

Commonsense often seems subjective

Varies with region, age, social group, cultural background

dangerous (snake), dangerous (biking)
dangerous (kangaroo), cute (kangaroo)
edible (grasshopper), edible (FoieGras)
greeting (handshake), greeting (kiss), greeting (wai)

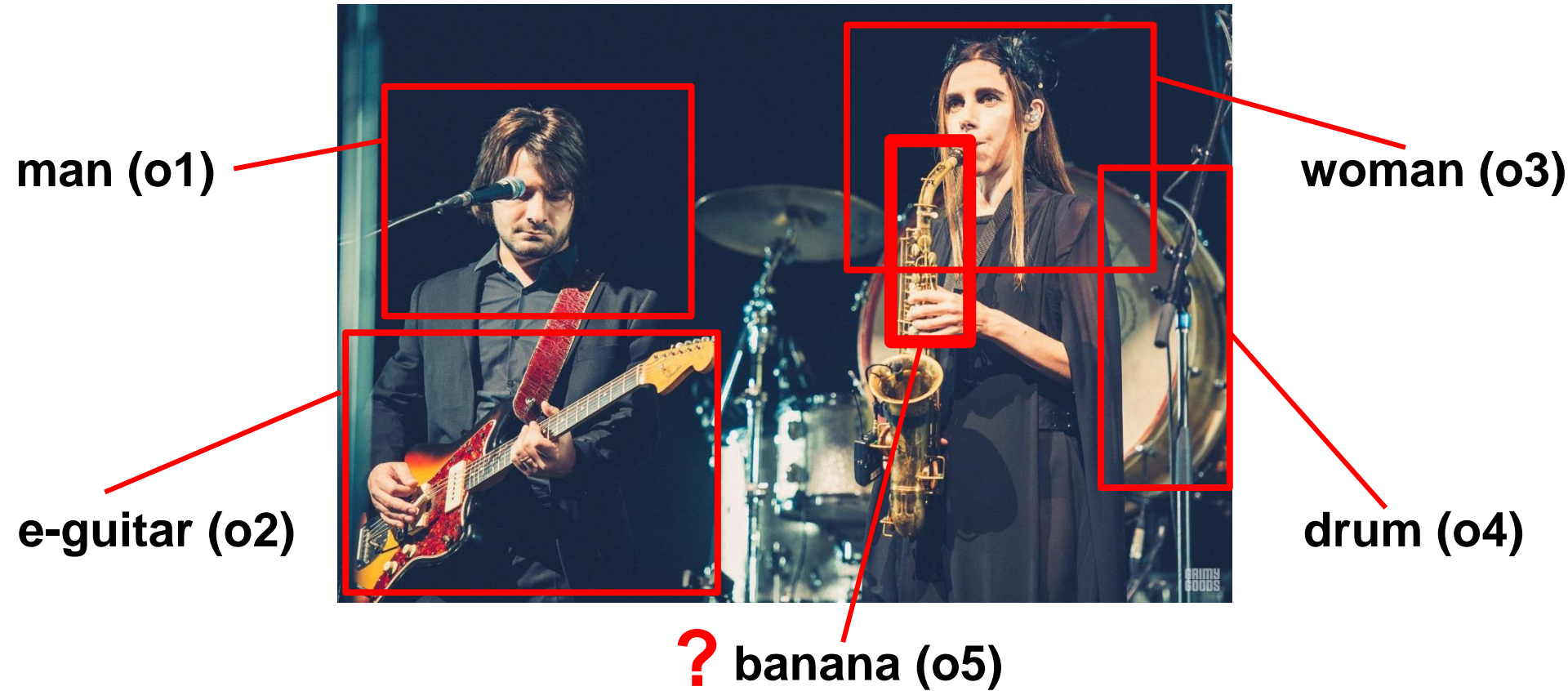


You see a dead cat on the street – what do you do?

- A: call the police**
- B: walk on**
- C: take a selfie**
- D: think about Schroedinger's cat**



Use Case: Visual Understanding



+ commonsense
knowledge:

hasColor (saxophone, yellow)
hasShape (saxophone, curved)
occursWith (saxophone, guitar)
usedFor (saxophone, music concert)

⇒ **saxophone(o5)**

Use Case: Visual Understanding



man, guitar, woman, saxophone, drums, ...

→ *man playing guitar and woman holding sax*

+ commonsense
knowledge:

→ **groovy rock concert**

Commonsense Knowledge: Challenges

Representation:

- triples, n-ary relations, modalities, fuzzy, higher-order?
`hasTaste(chocolate,sweet,mostly)`, `hasTaste(chocolate, bitter, possibly)`
`belief (x,edible(grasshoppers))`, `belief (y,edible(FoieGras))`
- plausibility invariants (rules)
 $\forall x: \text{human}(x) \Rightarrow ((\exists y: \text{mother}(x,y)) \Rightarrow \text{wasPregnant}(x))$
 $\forall x: (\text{animal}(x) \wedge \text{hasLegs}(x)) \Rightarrow \text{isEven}(\text{numLegs}(x))$

Acquisition:

- crowdsourcing, learning from online contents?
- tap books, movies, videos, social media, games?

Usage:

- from narrow chatbots to versatile digital companions
(telepresence, games, smart homes ...)
- from visual perception to visual cognition
(virtual humans, drones/robots ...)



Outline

★ 1 Knowledge Graphs ...

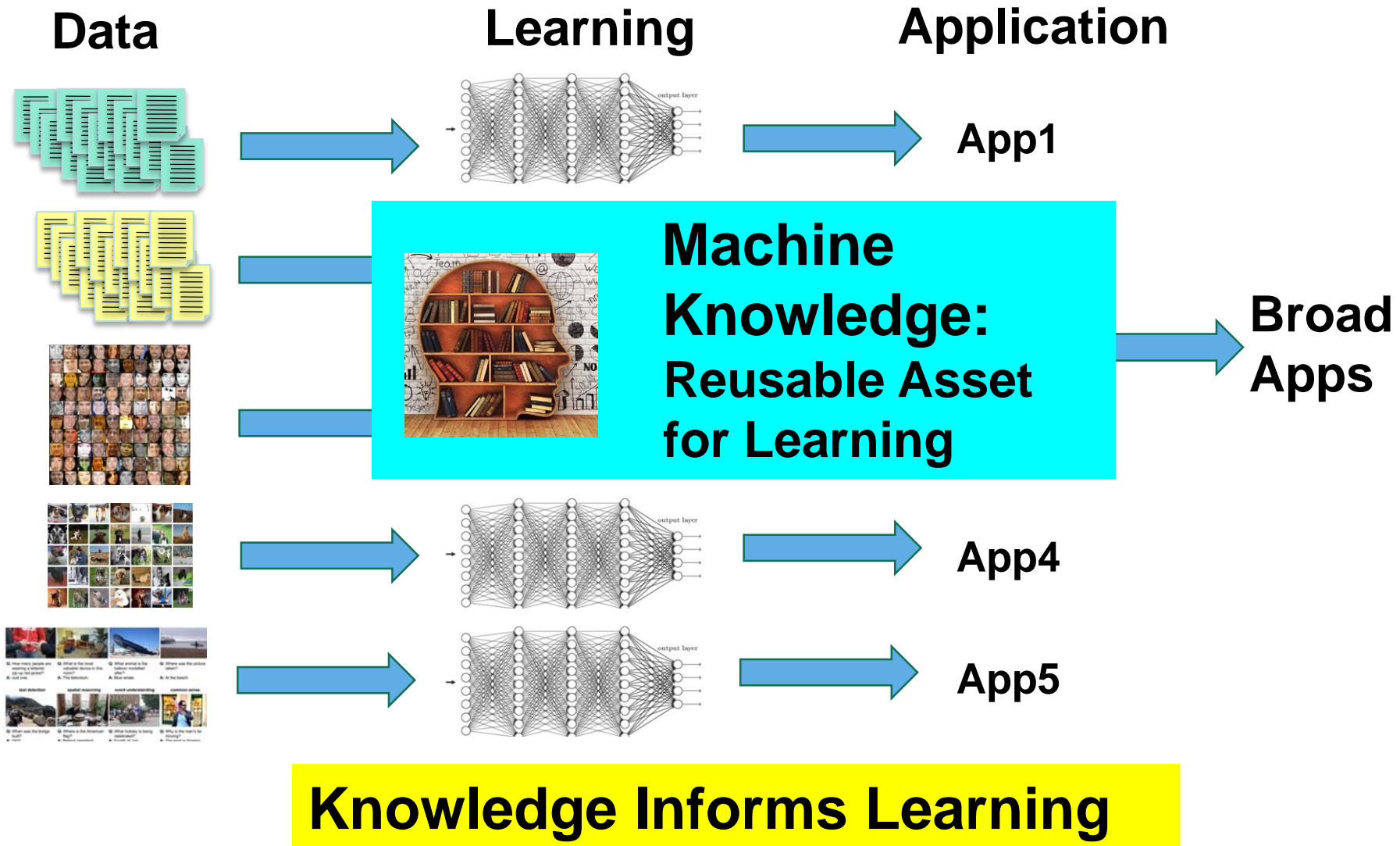
★ 2 ... And Beyond

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Deep Learning End-to-End: No Need for Machine Knowledge ?



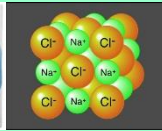
Take-Home Message

**Thank You !
Questions ?**



Encyclopedic Knowledge

Entities ...



Expert Knowledge

Quantities ...



Commonsense Knowledge

Properties ...



Socio-Cultural Knowledge

Activities ...

Advanced Machine Knowledge: Key Asset for
Robust, Interpretable, Versatile AI Applications
(cognitive assistants, visual understanding, ...)

Challenges: Representation, Acquisition, Usage