

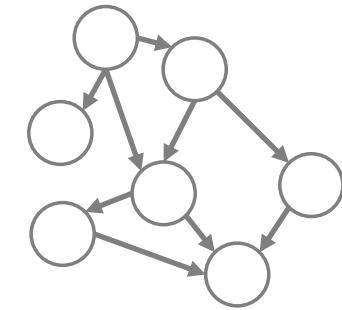
**Part 2:
Knowledge
Extraction**



**Part 3:
Graph
Construction**



Part 1: Knowledge Graphs



Part 4: Critical Analysis

Tutorial Outline

1. Knowledge Graph Primer [Jay] 
2. Knowledge Extraction from Text
 - a. NLP Fundamentals [Sameer] 
 - b. Information Extraction [Bhavana] 
- Coffee Break 
3. Knowledge Graph Construction
 - a. Probabilistic Models [Jay] 
 - b. Embedding Techniques [Sameer] 
4. Critical Overview and Conclusion [Bhavana] 

Critical Overview

SUMMARY

SUCCESS STORIES

DATASETS, TASKS, SOFTWARES

EXCITING ACTIVE RESEARCH

FUTURE RESEARCH DIRECTIONS

Critical Overview

SUMMARY

SUCCESS STORIES

DATASETS, TASKS, SOFTWARES

EXCITING ACTIVE RESEARCH

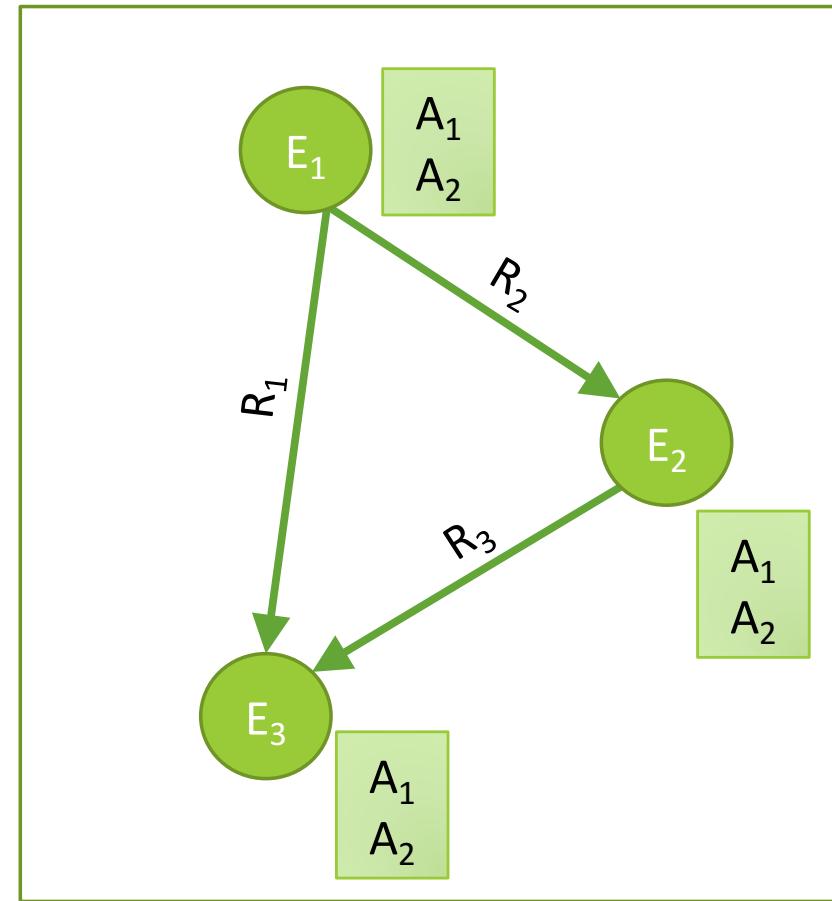
FUTURE RESEARCH DIRECTIONS

Why do we need Knowledge graphs?

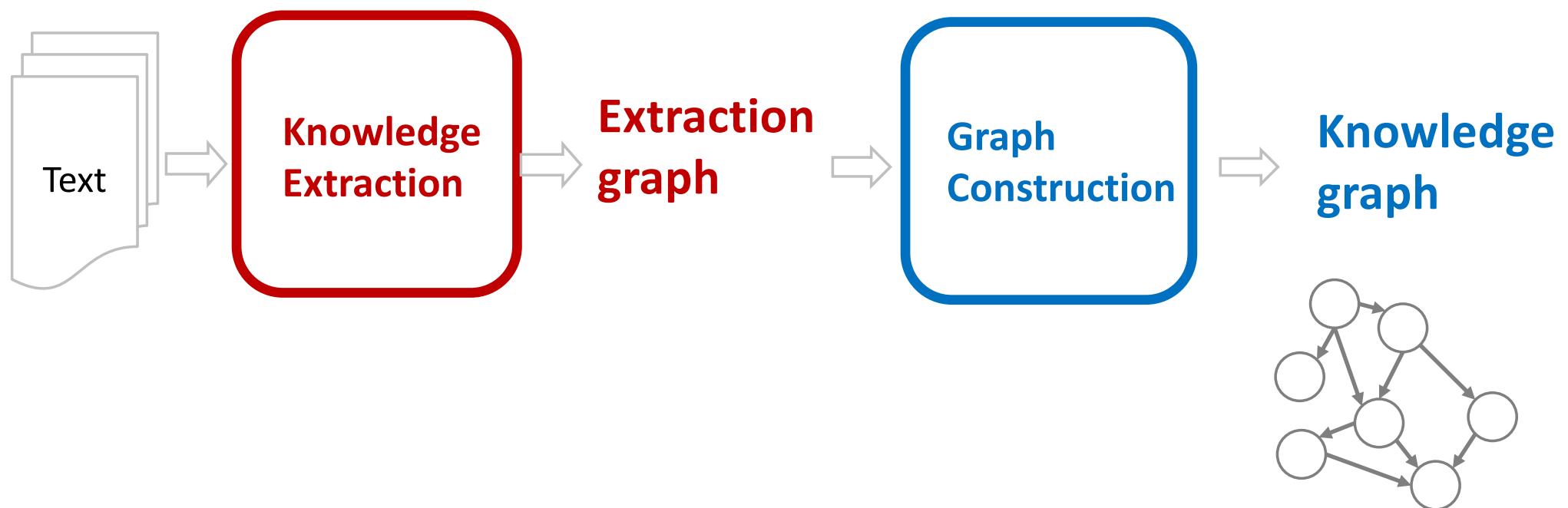
- Humans can explore large database in intuitive ways
- AI agents get access to human common sense knowledge

Knowledge graph construction

- **Who** are the entities (nodes) in the graph?
- **What** are their attributes and types (labels)?
- **How** are they related (edges)?



Knowledge Graph Construction



Two perspectives

	Extraction graph	Knowledge graph
Who are the entities? (nodes)		
What are their attributes? (labels)		
How are they related? (edges)		

Two perspectives

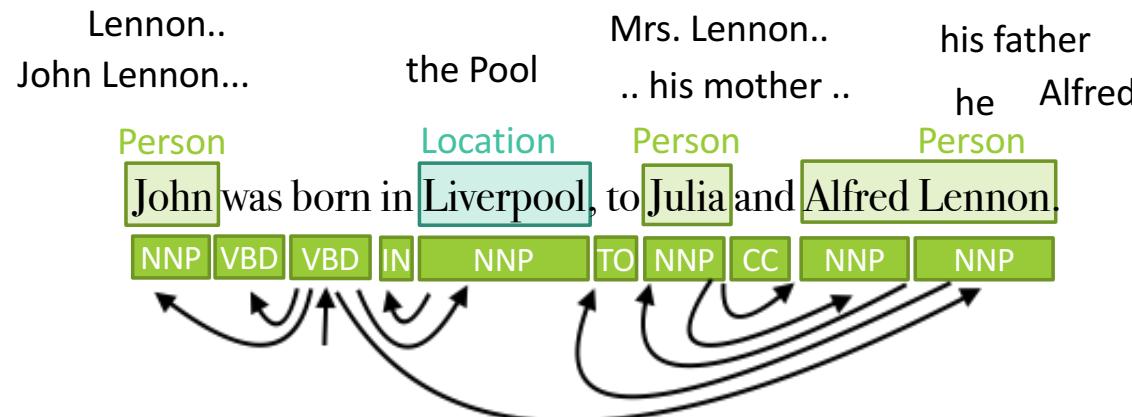
	Extraction graph	Knowledge graph
Who are the entities? (nodes)	<ul style="list-style-type: none">• Named Entity Recognition• Entity Coreference	<ul style="list-style-type: none">• Entity Linking• Entity Resolution
What are their attributes? (labels)	<ul style="list-style-type: none">• Entity Typing	<ul style="list-style-type: none">• Collective classification
How are they related? (edges)	<ul style="list-style-type: none">• Semantic role labeling• Relation Extraction	<ul style="list-style-type: none">• Link prediction

Knowledge Extraction

John was born in Liverpool, to Julia and Alfred Lennon.

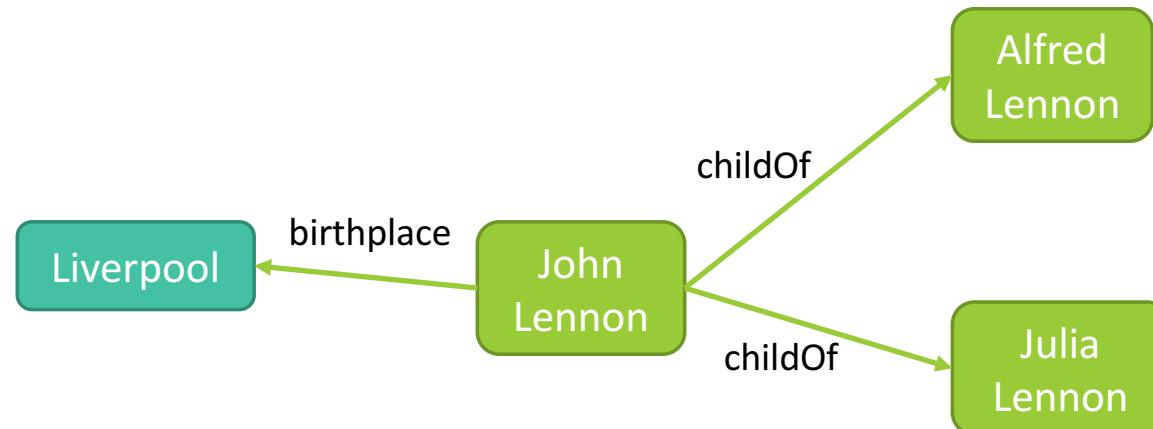
Text

NLP



Annotated text

Information Extraction



Extraction graph

NLP

Document

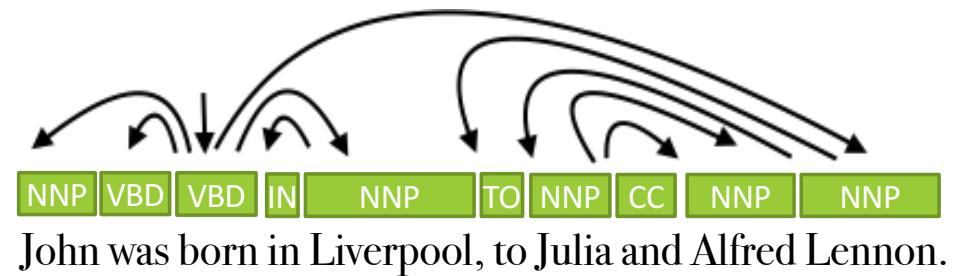
Within-doc Coreference...

Lennon..
John Lennon...
the Pool
Mrs. Lennon..
.. his mother ..
his father
he Alfred

Person Location Person Person
John was born in Liverpool, to Julia and Alfred Lennon.

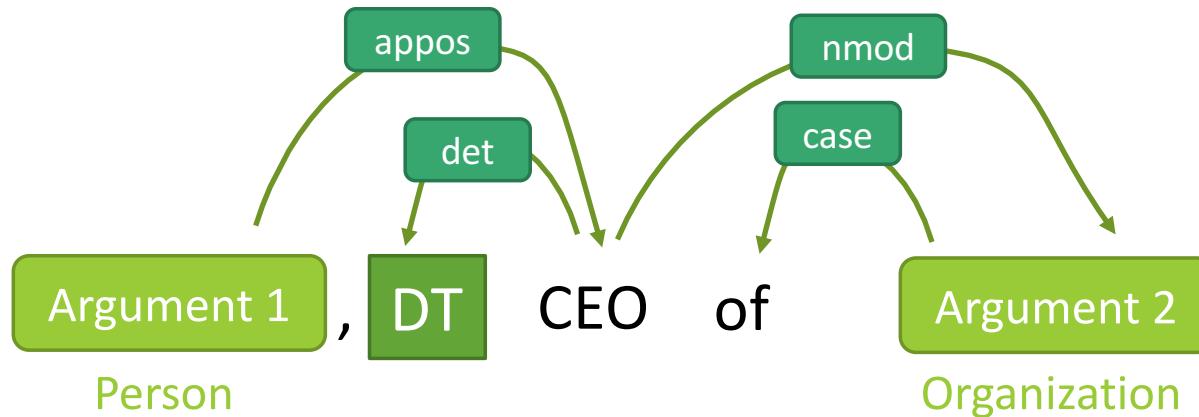
Sentence

Dependency Parsing,
Part of speech tagging,
Named entity recognition...



Dependency Parse → Surface Patterns

Combine tokens, dependency paths, and entity types to define rules.



Bill Gates, the CEO of Microsoft, said ...

Mr. Jobs, the brilliant and charming CEO of Apple Inc., said ...

... announced by Steve Jobs, the CEO of Apple.

... announced by Bill Gates, the director and CEO of Microsoft.

... mused Bill, a former CEO of Microsoft.

and many other possible instantiations...

Information Extraction

Single extractor

Defining domain

Learning extractors

Scoring candidate facts



Supervised



Semi-supervised



Unsupervised

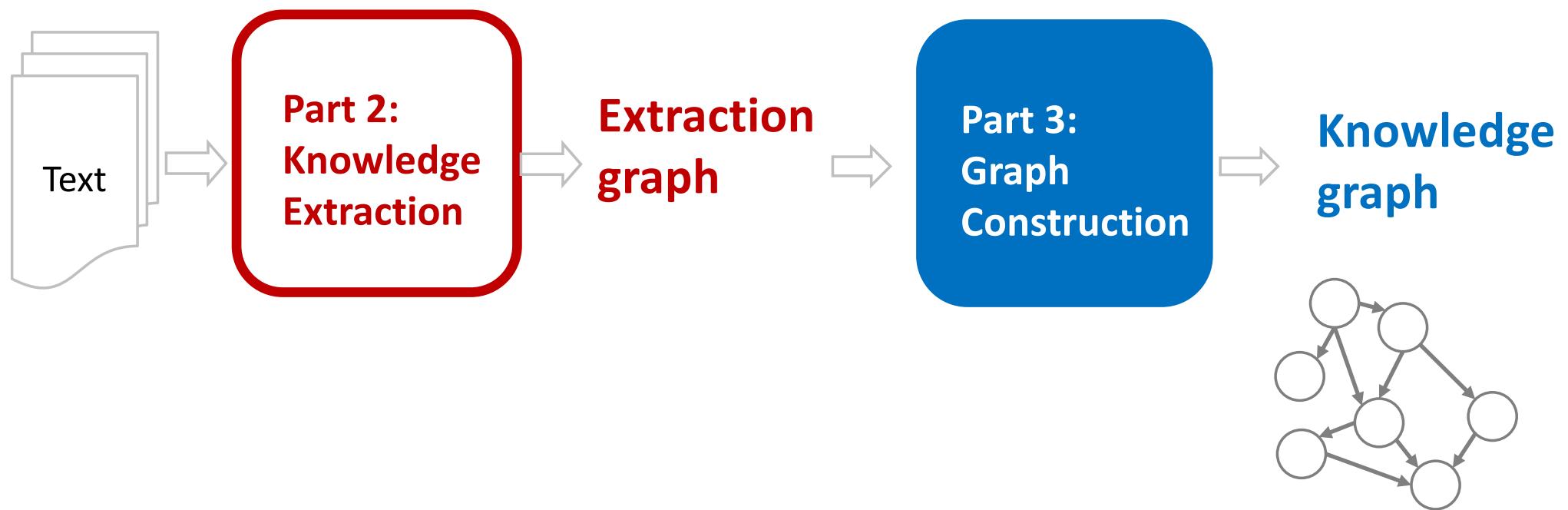


Fusing multiple extractors

IE systems in practice

	Defining domain	Learning extractors	Scoring candidate facts	Fusing extractors
ConceptNet				
NELL				Heuristic rules
Knowledge Vault				Classifier
OpenIE				

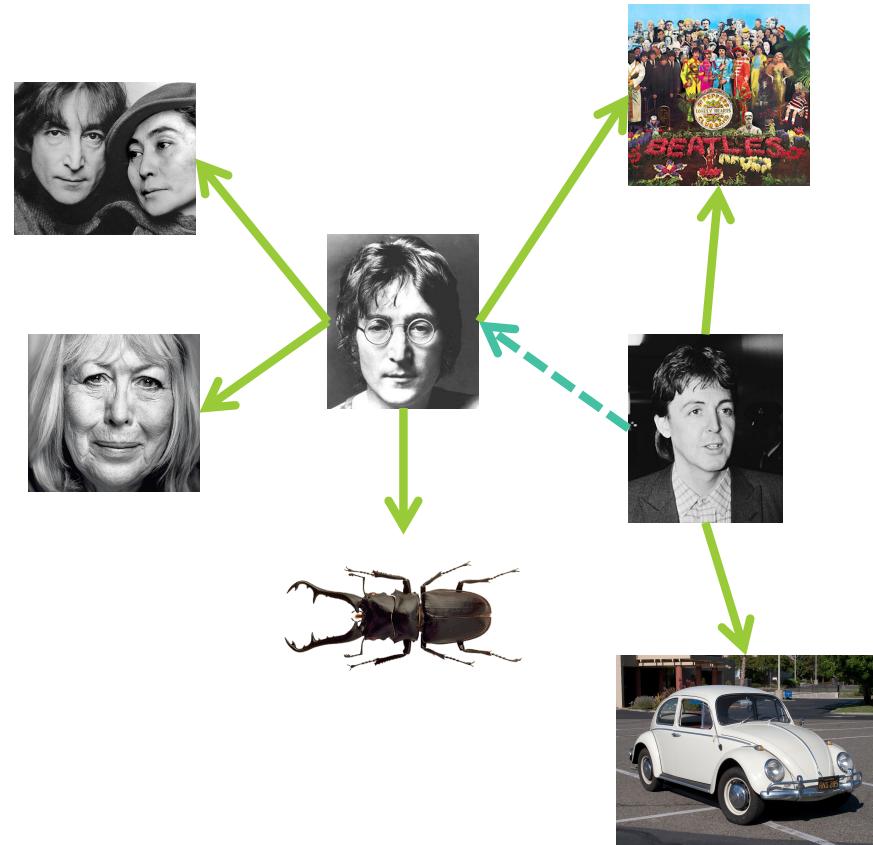
Knowledge Graph Construction



Issues with Extraction Graph

Extracted knowledge could be:

- ambiguous



- incomplete

- inconsistent

Two approaches for KG construction

PROBABILISTIC MODELS

EMBEDDING BASED MODELS

Two approaches for KG construction

PROBABILISTIC MODELS

EMBEDDING BASED MODELS

Two classes of Probabilistic Models

GRAPHICAL MODEL BASED

- Possible facts in KG are variables
- Logical rules relate facts
- Probability \propto satisfied rules
- Universal-quantification

RANDOM WALK BASED

- Possible facts posed as queries
- Random walks of the KG constitute “proofs”
- Probability \propto path lengths/transitions
- Local grounding

Two approaches for KG construction

PROBABILISTIC MODELS

EMBEDDING BASED MODELS

Why embeddings?

Limitations of
probabilistic models

Limitation to Logical Relations

- Representation restricted by manual design
 - Clustering? Asymmetric implications?
 - Information flows through these relations
- Difficult to generalize to unseen entities/relations

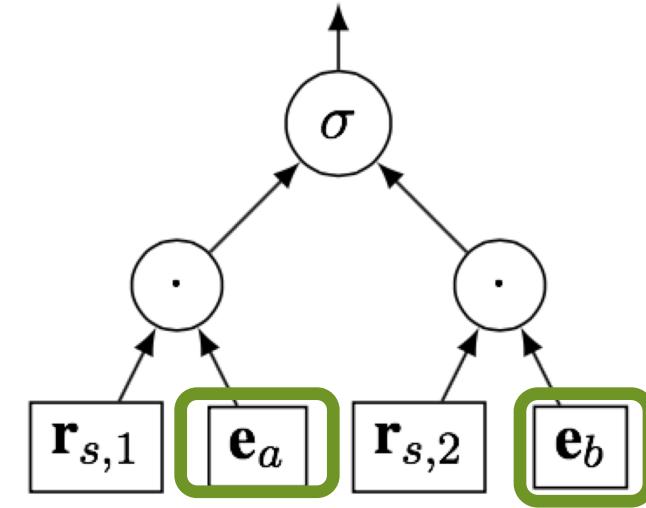
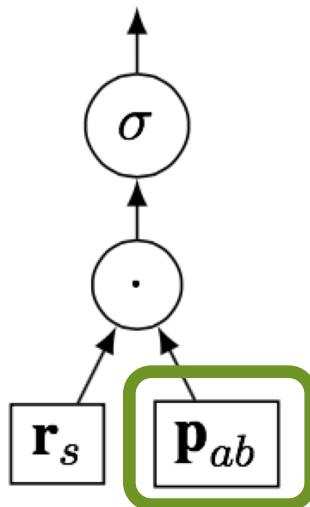
Computational Complexity of Algorithms

- Learning is NP-Hard, difficult to approximate
- Query-time inference is also NP-Hard
- Not easy to parallelize, or use GPUs
- Scalability is badly affected by representation

Embedding
based models

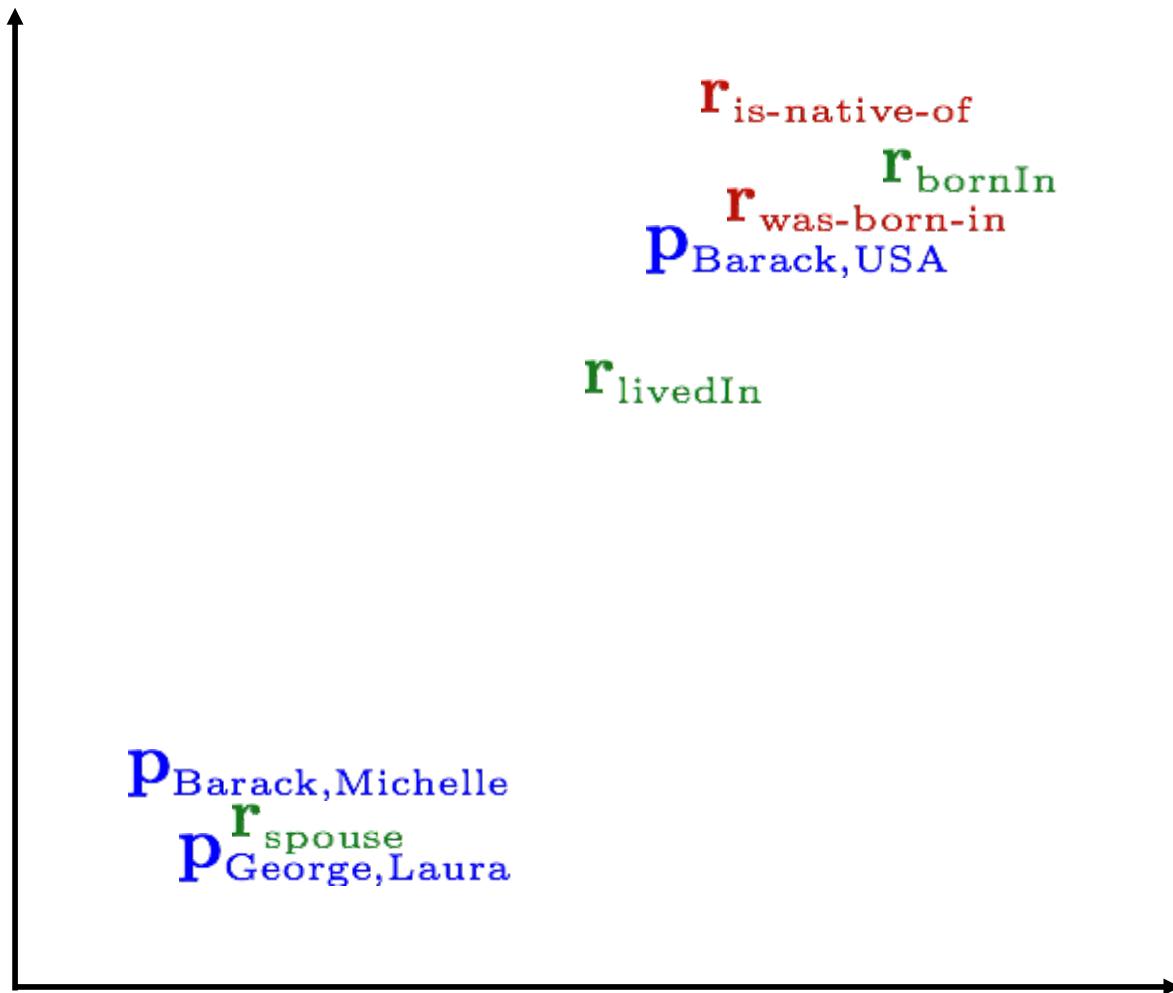
- Can generalize to unseen entities and relations
- Efficient inference at large scale

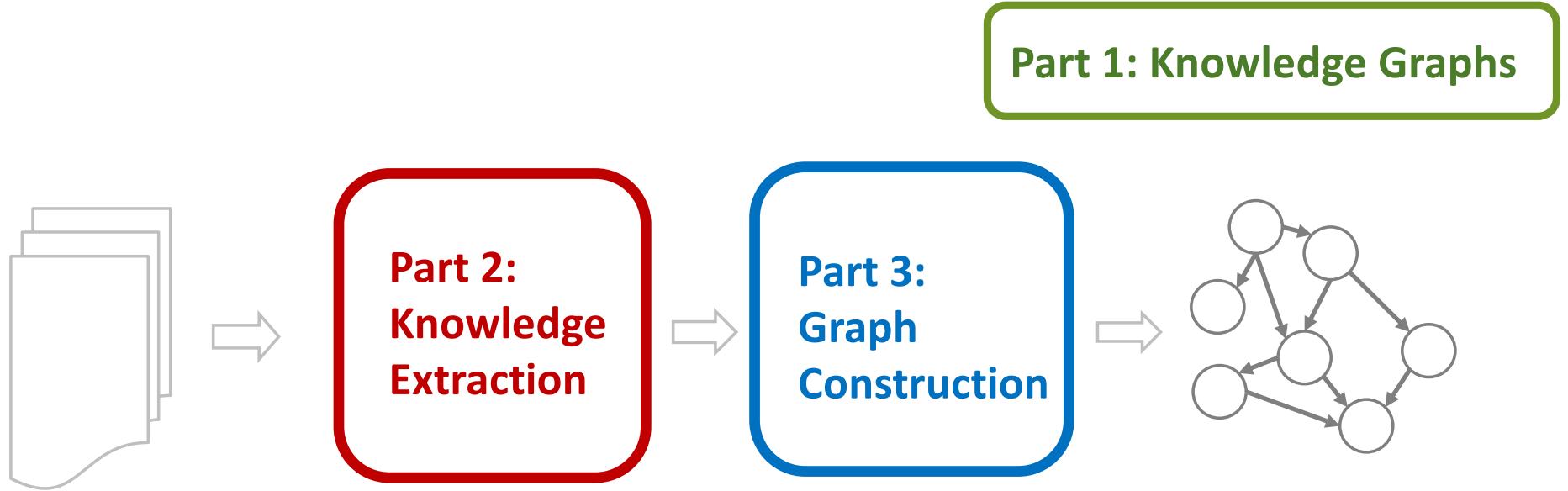
Matrix vs Tensor Factorization



- No encoding of type information
- Can only predict for entity pairs that appear in text together
- Sufficient evidence has to be seen for each entity pair
- Assume low-rank for pairs
- But many relations are not!
- Spouse: you can have only ~ 1
- Cannot learn pair specific information

Relation Embeddings





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Success story: OpenIE

- **Key contributions:**

- No need for human defined relation schemas
- First ever successful open-source open domain IE system

- **ReVerb**

- Input = Clueweb09 corpus (1B web pages)
- Output = 15M high-precision extractions

Success story: OpenIE (ReVerb)



Open Information Extraction

openie.allenai.org

Hosted by

Created at



Argument 1: entity:The Beatles

Relation:

Argument 2:

All

Search

all

[location \(21\)](#)

[film location \(18\)](#)

[statistical region \(16\)](#)

[name source \(15\)](#)

[travel destination \(14\)](#)

[misc.](#)

[more types ▾](#)

[were bigger than Jesus \(100\)](#)

[came to America \(95\)](#)

[appeared on The Ed Sullivan Show \(88\)](#)

[broke up in 1970 \(56\)](#)

[Here Comes the Sun \(46\)](#)

[came to America \(45\)](#)

[is for the future \(44\)](#)

[are a great band \(42\)](#)

[perform on The Ed Sullivan Show \(39\)](#)

[were Musical ensemble \(36\)](#)

are a great band ►

Extracted Synonyms:

were
is
was

Extracted from these sentences:

are The Beatles are the best band , hands down but Oasis did make a great cover . (via ClueWeb12)

The Beatles are a great band . (via ClueWeb12)

The Beatles are the best band . (via ClueWeb12)

The Beatles are the greatest band ... Started 1 month ago by georgedcc Yeah , Songs in the Key of Life is a bit much for 1 listen . (via ClueWeb12)

The Beatles , arguably , are the greatest band , and may or may not have the greatest name . (via ClueWeb12)

The point is , from my view , The Beatles are a good band , but way behind the greatest artists to ever grace rock . (via ClueWeb12)

Open IE Systems

2007

2010

2012

2014

2016



OpenIE v 1.0

TextRunner

CRF

Self-training

v 2.0

ReVerb

POS-tag
based
relation
extraction

v 3.0

OLLIE

Dependency
parse based
extraction

OpenIE 4.0

SRL-based
extraction;
temporal,
spatial
extractions

OpenIE 5.0

Supports
compound
noun
phrases;
numbers;
lists

Increase in precision, recall, expressiveness

Success story: NELL

- **Key technical contributions:**
 - “Never ending learning” paradigm
 - “Coupled bootstrap learning” to reduce semantic drift
- Input: Clueweb09 corpus (1B web pages)
- Ontology: ~2K predicates
 - ~700K constraints between predicates
- Output: 50 million candidate facts
 - 3 million high-confidence facts

Success story: NELL

NELL Knowledge Base Browser

CMU Read the Web Project

log in | preferences | help/instructions | feedback | Search

categories relations

everypromotedthing
abstractthing
event
convention
musicfestival
protestevent
meetingeventtitle
conference
mlconference
weatherphenomenon
sportsevent
sportsgame
race
olympics
grandprix
crimeorcharge
earthquakeevent
election
bombingevevt
militaryeventtype
militaryconflict
productlaunchevent
filmfestival
roadaccidentevevt
meetingeventtype
eventoutcome
mlalgorithm
physiologicalcondition
disease

beatles (musicartist)
literal strings: [BEATLES](#), [Beatles](#), [beatles](#)

Help NELL Learn!

NELL wants to know if these beliefs are correct.
If they are or ever were, click thumbs-up. Otherwise, click thumbs-down.

- [beatles](#) is a [musical artist](#)  
- [beatles](#) is a musician in the [genre classic_pop](#) (musicgenre)  
- [beatles](#) is a musician in the [genre pop](#) (musicgenre)  
- [beatles](#) is a musician in the [genre rock](#) (musicgenre)  
- [beatles](#) is a musician in the [genre classic_rock](#) (musicgenre)  

categories

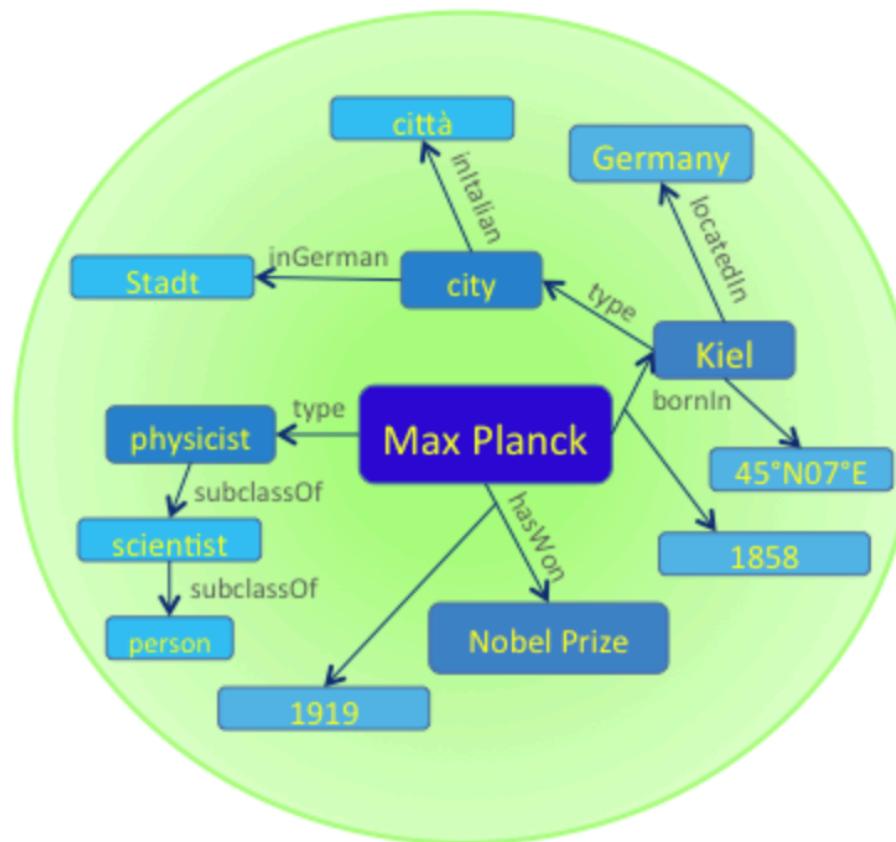
- [musicartist](#)(100.0%)
 - MBL @198 (100.0%) on 07-feb-2011 [Promotion of musicartist:beatles musicartistgenre musicgenre:classic_rock]
 - CPL @1021 (80.9%) on 14-oct-2016 ["numerous other artists including _ " "traducidas de _ " "incluidas en _ " "had a guitar player" "early pioneers such as _ " "controversial photo of _ " "distressed image of _ " "D-tracks of _ " "Beatles Come Together" "ohne die _ " "opening band for _ " "American acts like _ " "classic acts like _ " "performance footage of _ " "were the perfect band" " " "record label" "record album by _ " "les paroles de _ " "never recorded the song" "such renowned artists as _ " "did a few songs" "Top artists include _ " "crazy lives of _ " "UK artists such as _ " "Lennon started _ " " 'musical talent" " " 'Birthplace" " " 'harmonies" "Tour , starring _ " " 'last days" " " 'fourth album" " " sixth studio album" " " original recordings" "They were also pushing _ " "She Said by _ " "Other artists featured include _ " "Post general comments related to _ " "track also shows _ " "such major artists as _ " "time favorite band is _ " "past masters such as _ " "pop hooks of _ " "popular musicians like _ " "pop icons such as _ " "music artists like _ " "music bands like _ " "pop stars such as _ " "pop influenced by _ "

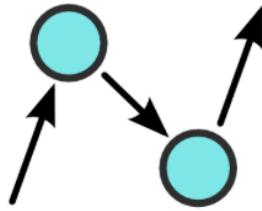
Success story: YAGO

- **Key contributions:**
 - **Rich Ontology:** Linking Wikipedia categories to WordNet, providing a rich taxonomy
 - **High Quality:** High precision extractions (accuracy ~95%)

Success story: YAGO

- **Input:** Wikipedia infoboxes, WordNet and GeoNames
- **Output:** KG with 350K entity types, 10M entities, 120M facts
- Temporal and spatial information





ConceptNet

An open, multilingual knowledge graph

Link

en

beatles

An English term in ConceptNet 5.5

Derived terms

- en** beatle →
- en** beatledom →
- en** beatlemania →
- en** beatlesque →
- en** fourth beatle →

beatles is a type of...

- en** a British band →
- en** man (n) →
- en** band (n) →
- en** musician (n) →
- en** album (n) →

Links to other sites

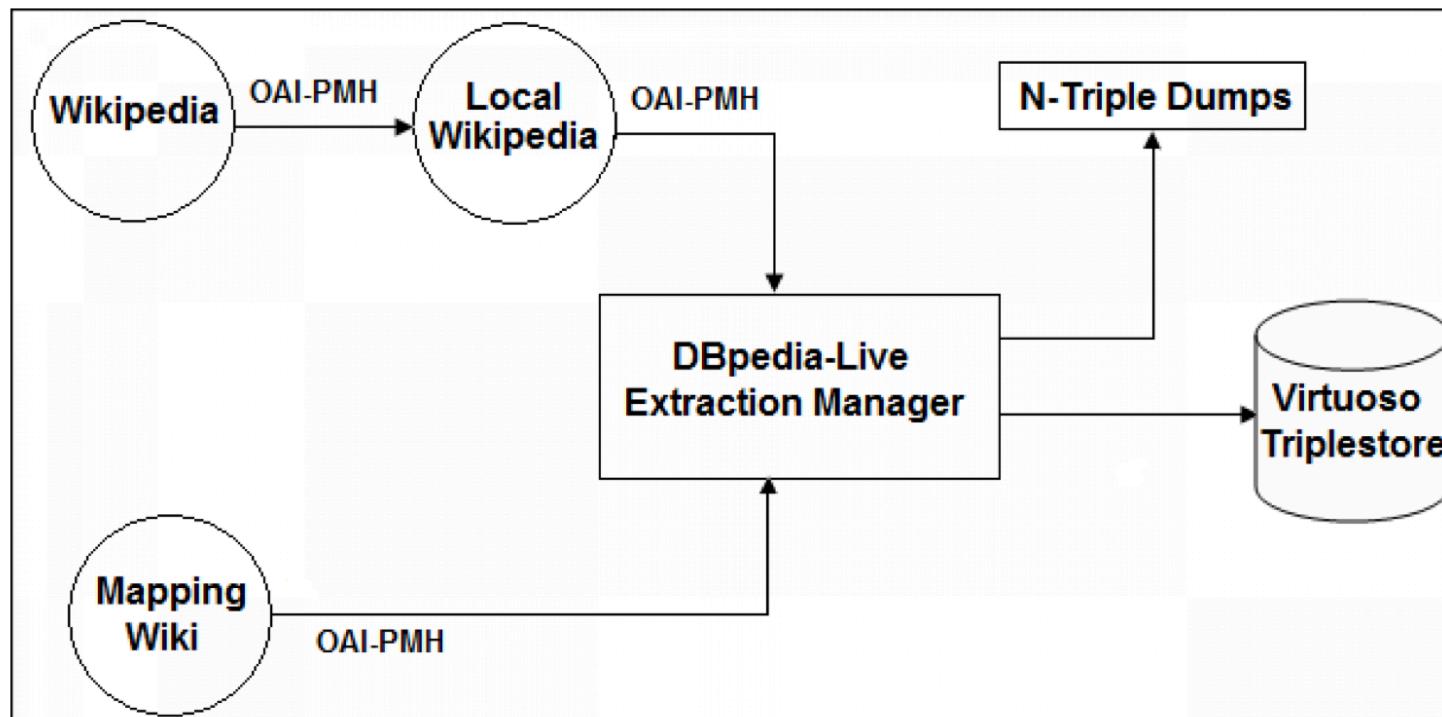
- dbpedia.org** The Beatles →
- sw.opencyc.org** Beatle →
- umbel.org** Beatle →
- wordnet-rdf.princeton.edu** 400520405-N →
- wordnet-rdf.princeton.edu** 108386847-n →
- wikidata.dbpedia.org** Q1299 →
- en.wiktionary.org** Beatles →
- dbpedia.org** The Beatles (No. 1) →
- wikidata.dbpedia.org** Q738260 →
- fr.wiktionary.org** Beatles →
- dbpedia.org** The Beatles (The Original Studio Recordings) →
- wikidata.dbpedia.org** Q603122 →

Success story: ConceptNet

- Commonsense knowledge base
- **Key contributions:**
 - **Freely available resource:** covers wide range of common sense concepts and relations organized in a easy-to-use semantic network
 - **NLP toolkit based on this resource:** supports analogy, text summarization, context dependent inferences
- ConceptNet4 was manually built using inputs from thousands of people
 - 28 million facts expressed in natural language
 - spanning 304 different languages

Success story

- Wikipedia is manually built encyclopedia project
- DBpedia is automatically extracted structured data from Wikipedia
 - 17M canonical instances
 - 88M type statements
 - 72M infobox statements

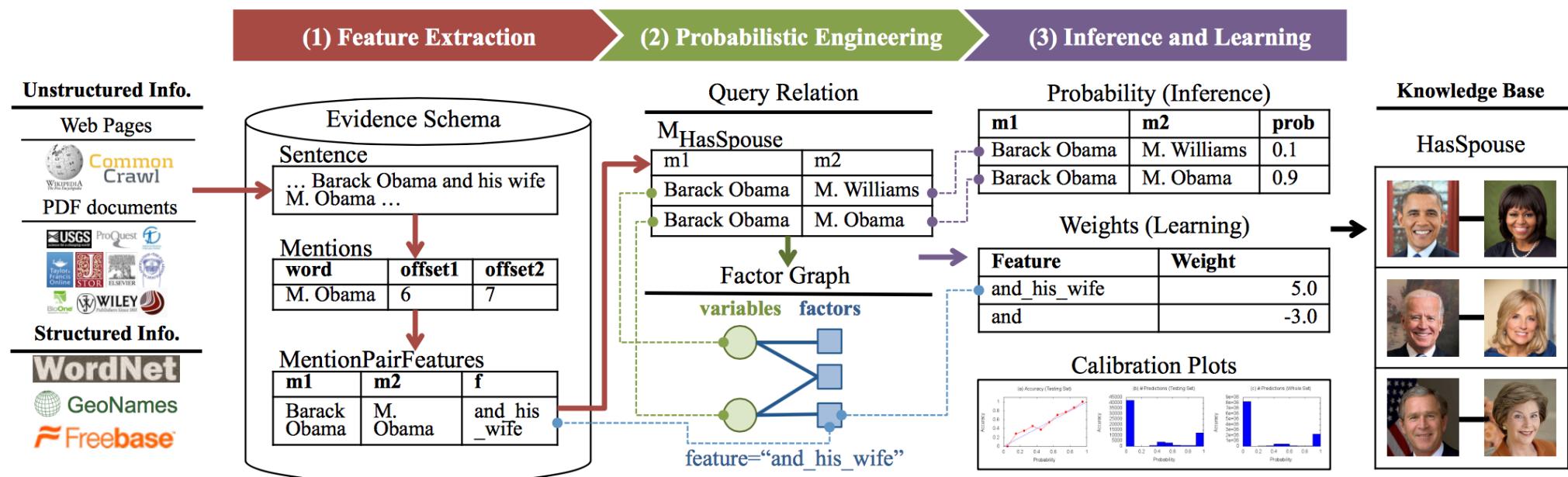


DeepDive



- Machine learning based extraction system
- Key contributions:
 - **scalable, high-performance inference and learning engine**
 - **Developers contribute features (rules) not algorithms**
 - **Combines data from variety of sources (webpages, pdf, figures, tables)**

DeepDive



- Best Precision/recall/F1 in KBP-slot filling task 2014 evaluations (31 teams participated)

Critical Overview

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EXCITING ACTIVE RESEARCH

FUTURE RESEARCH DIRECTIONS

Datasets

- KG as datasets
 - [FB15K-237](#) Knowledge base completion dataset based on Freebase
 - [DBpedia](#) Structured data extracted from Wikipedia
 - [NELL](#) Read the web datasets
 - [AristoKB](#) Tuple knowledge base for Science domain
- Text datasets
 - [Clueweb09](#): 1 billion webpages (sample of Web)
 - [FACC1](#): Freebase Annotations of the Clueweb09 Corpora
 - [Gigaword](#): automatically-generated syntactic and discourse structure
 - [NYTimes](#): The New York Times Annotated Corpus
- Datasets related to Semi-supervised learning for information extraction
[Link](#): entity typing, concept discovery, aligning glosses to KB, multi-view learning

Shared tasks

- Text Analysis Conference on Knowledge Base Population (TAC KBP)
 - Slot filling task
 - Cold Start KBP Track
 - Tri-Lingual Entity Discovery and Linking Track (EDL)
 - Event Track
 - Validation/Ensembling Track

Softwares: NLP

- Stanford CoreNLP: a suite of core NLP tools
[\[link\]](#) (Java code)

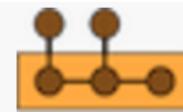


Stanford CoreNLP

- FIGER: fine-grained entity recognizer
assigns over 100 semantic types
[link](#) (Java code)

UNIVERSITY *of* WASHINGTON

- FACTORIE: out-of-the-box tools for NLP and
information integration
[link](#) (Scala code)



FACTORIE

- EasySRL: Semantic role labeling
[link](#) (Java code)

UNIVERSITY *of* WASHINGTON

Softwares: Information Extraction

- Open IE
(University of Washington)
Open IE 4.2 [link](#) (Scala code)
- Stanford Open IE [link](#) (Java code)



- Interactive Knowledge Extraction (IKE)
(Allen Institute for Artificial Intelligence)
[link](#) (Scala code)



- Universal schema
[link](#) (Java + Scala code)

UMASS
AMHERST

Softwares: Knowledge Graphs

- **AMIE:** Association Rule Mining under Incomplete Evidence in Ontological Knowledge Bases
[link](#) (Java code)



- **PSL:** Probabilistic soft logic
[link](#) (Java code)



- **ProPPR:** Programming with Personalized PageRank
[link](#) (Java code)

Carnegie Mellon University

- **Junto:** graph based semi-supervised learning methods
[link](#) (Scala code)



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Exciting Active Research

- INTERESTING APPLICATIONS OF KG
- MULTI-MODAL INFORMATION EXTRACTION
- AMBITIOUS PROJECTS

Exciting Active Research

- INTERESTING APPLICATIONS OF KG
- MULTI-MODAL INFORMATION EXTRACTION
- AMBITIOUS PROJECTS

(1) Exciting active research: Interesting application of Knowledge Graphs

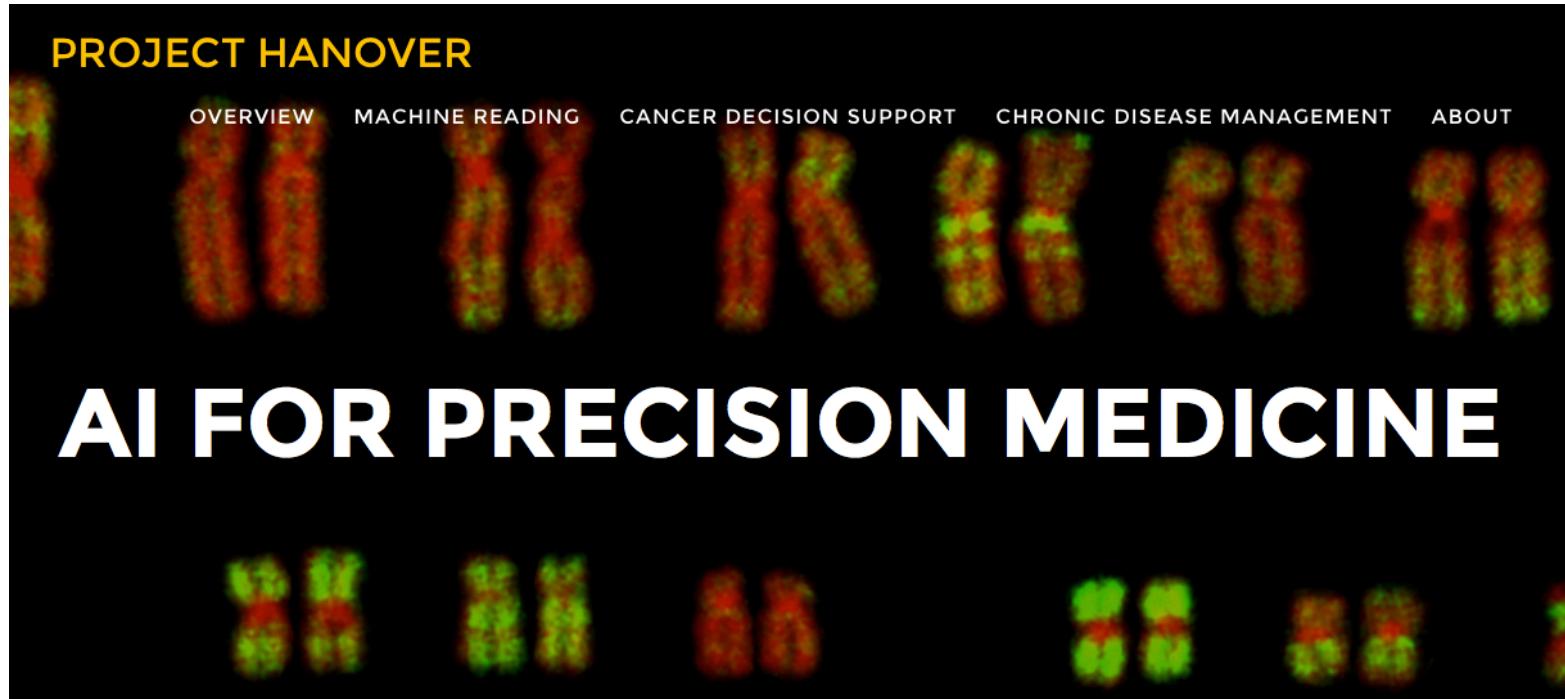
The Literome Project [[link](#)]

- Automatic system for extracting genomic knowledge from PubMed articles
- Web-accessible knowledge base

Search for directed genic interactions: [help](#)

Search for genotype-phenotype associations: [help](#)

(1) Exciting active research: Interesting application of Knowledge Graphs



Microsoft®
Research

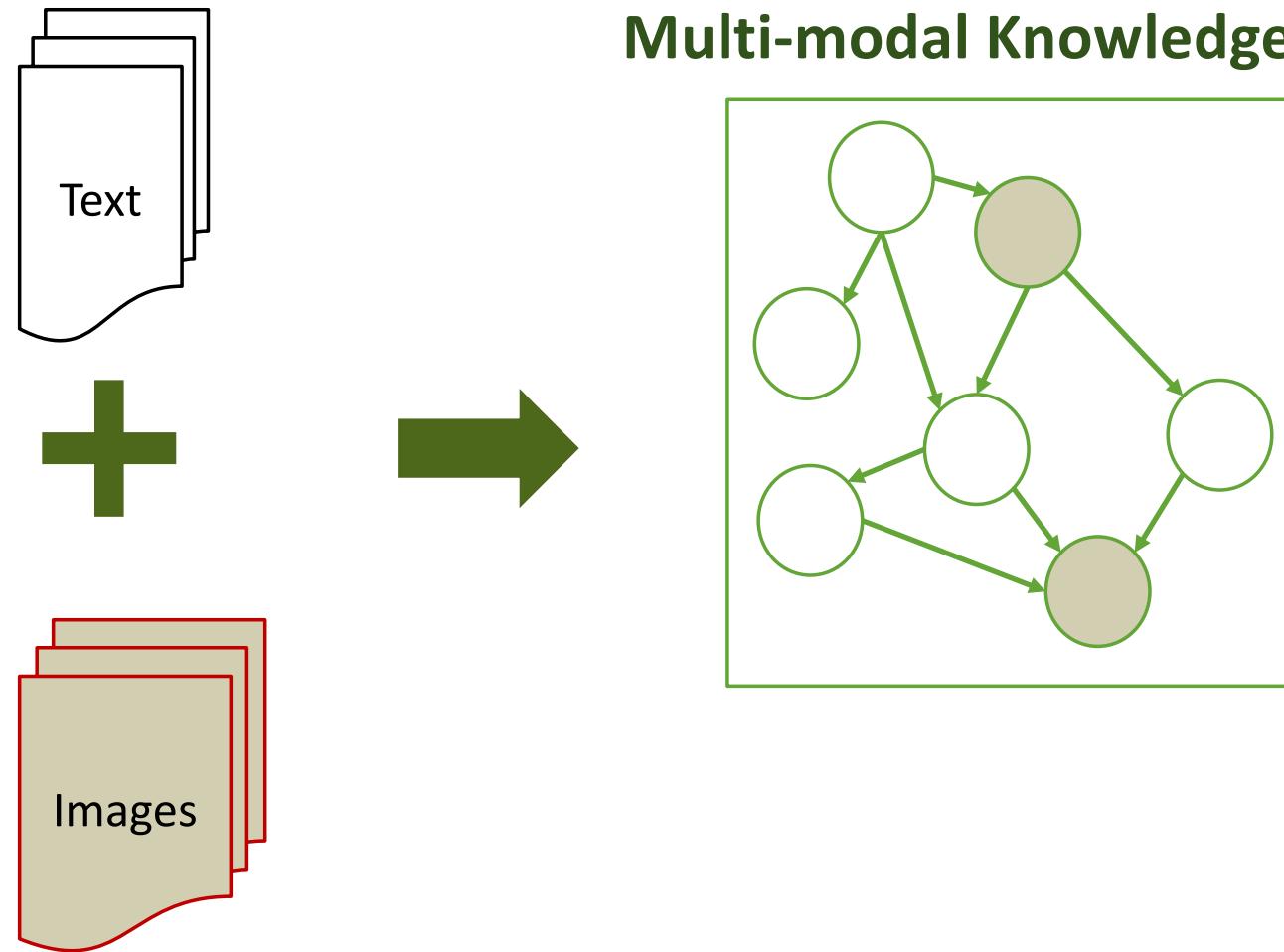
Chronic disease management:

develop AI technology for predictive and preventive personalized medicine to reduce the national healthcare expenditure on chronic diseases (90% of total cost)

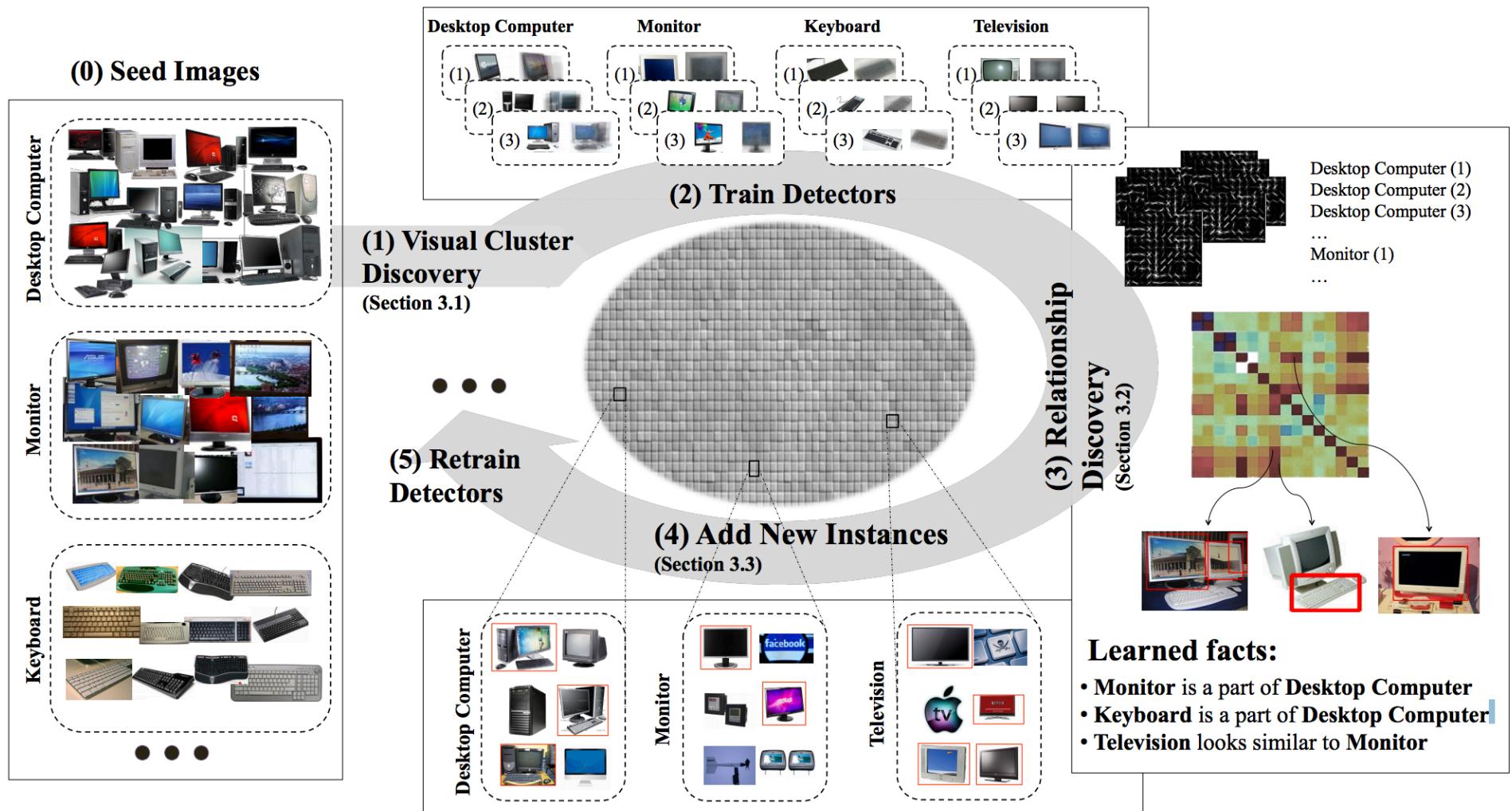
Exciting Active Research

- INTERESTING APPLICATIONS OF KG
- MULTI-MODAL INFORMATION EXTRACTION
- AMBITIOUS PROJECTS

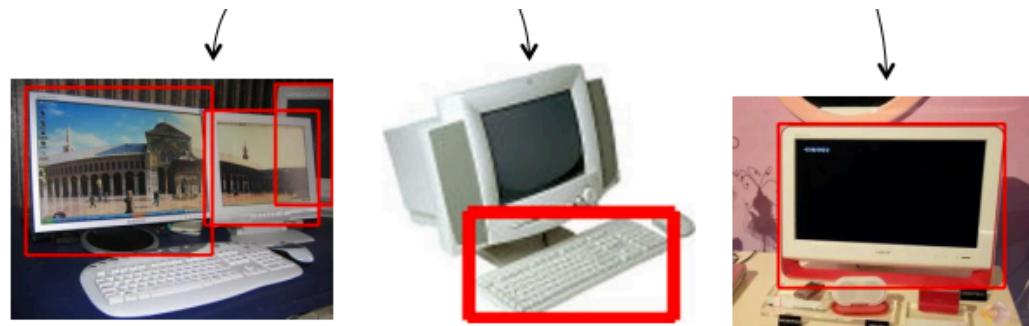
(2) Exciting active research: Multi-modal information extraction



NEIL: Extracting Visual Knowledge from Web Data



NEIL: Extracting Visual Knowledge from Web Data



Learned facts:

- **Monitor** is a part of **Desktop Computer**
- **Keyboard** is a part of **Desktop Computer**
- **Television** looks similar to **Monitor**

WebChild: Text + Images

WEBCHILD Commonsense Browser

e.g. car,bicycle OR car OR a:fix bicycle 

Guess the concept

Domain  mouse

Comparable  keyboard

Physical Part  a hand-operated electronic device that controls the coordinates of a cursor on your computer screen as you move it around on a pad; on the bottom of the device is a ball that rolls on the surface of the pad; 'a mouse takes much more room than a trackball'

Activity 

Property 

Location 

Ask me!


a hand-operated electronic device that controls the coordinates of a cursor on your computer screen as you move it around on a pad; on the bottom of the device is a ball that rolls on the surface of the pad; 'a mouse takes much more room than a trackball'


device consisting of a set of keys on a piano or organ or typewriter or typesetting machine or computer or the like

Exciting Active Research

- INTERESTING APPLICATIONS OF KG
- MULTI-MODAL INFORMATION EXTRACTION
- AMBITIOUS PROJECTS

(3) Exciting active research: Ambitious Project



The Allen AI Science Challenge

Is your model smarter than an 8th grader?

\$80,000 · a year ago

#	△1w	Team Name * in the money	Kernel	Team Members	Score ⓘ	Entries	Last
1	—	* Alejandro Mosquera		 	0.59375	2	1y
2	new	* Cardal		 	0.59000	2	1y
3	new	* poweredByTalkwalker		 	0.59000	4	1y

Aristo Science QA challenge

- Science questions dataset

~5K 4-way multiple choice questions

Frogs lay eggs that develop into tadpoles and then into adult frogs. This sequence of changes is an example of how living things _____

- (A) go through a life cycle
- (B) form a food web
- (C) act as a source of food
- (D) affect other parts of the ecosystem

Science knowledge

frog's life cycle,
metamorphosis



Common sense
knowledge

frog is an animal,
animals have life cycle

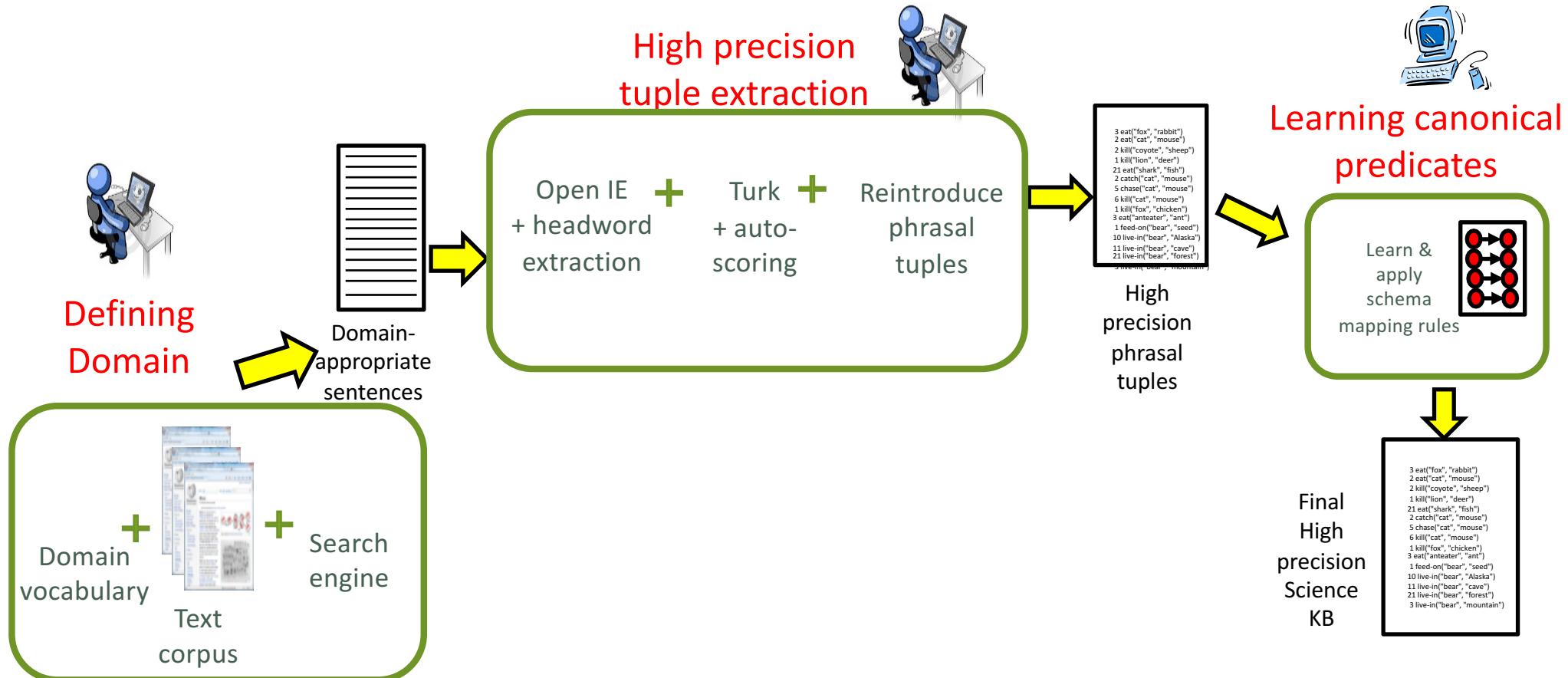
Existing knowledge graphs

- Too named entity centric (no domain relevance)
- Too noisy (not directly usable by inference systems)

AI2's ScienceKB

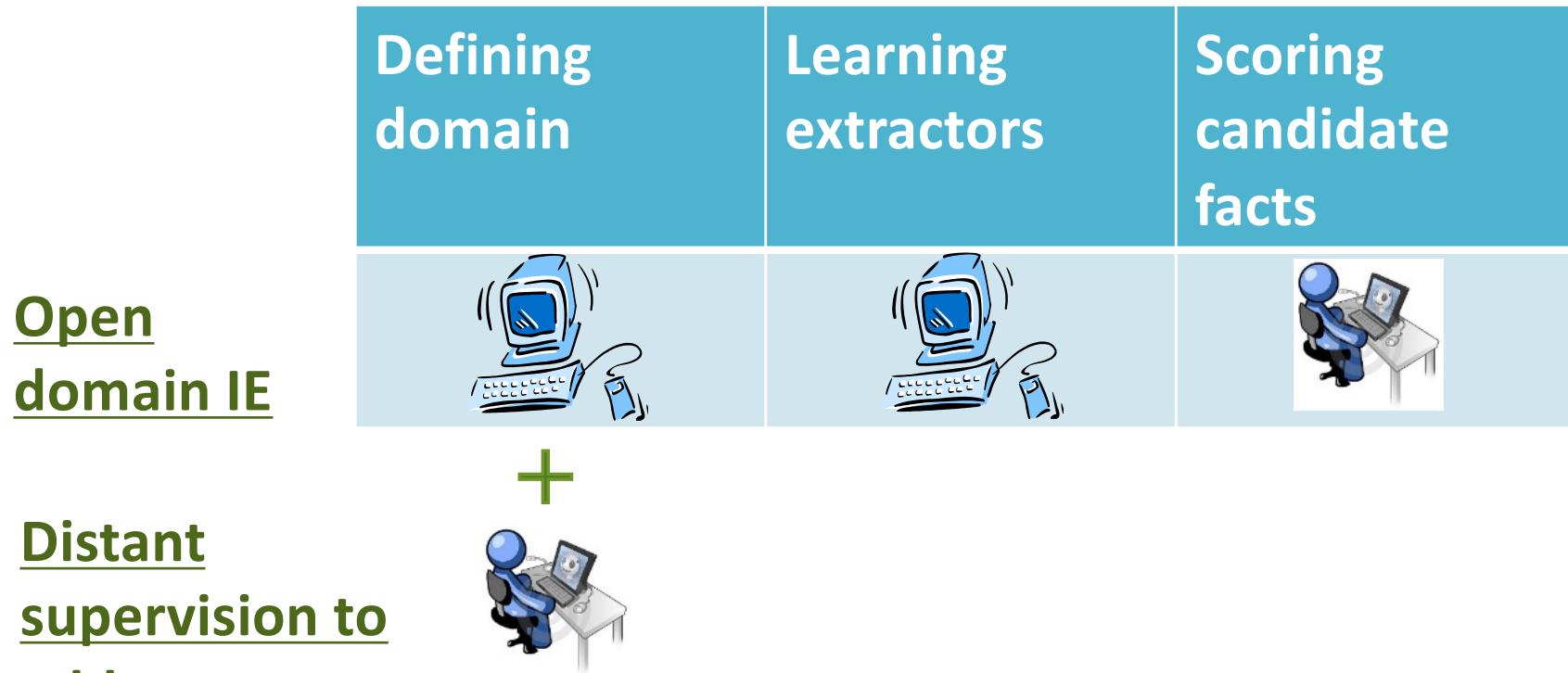


ALLEN INSTITUTE
for ARTIFICIAL INTELLIGENCE



**Upcoming article on ``High Precision Knowledge Extraction for Science domain''

Hybrid Approach: Adding structure to Open domain IE



AI2's TupleKB dataset: [link](#)

> 300K common-sense and science facts
> 80% precision

Aristo ScienceKB

- AI2's TupleKB dataset: [link](#)
- **Open problems**
 - Best KR for Science domain
 - Domain targeted KB completion
 - Measuring recall w.r.t. end task

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Future Research Directions

GOING BEYOND FACTS

CONTINUOUSLY LEARNING AND
SELF-CORRECTING SYSTEMS

(1) Future research directions:

Going beyond facts

- Most of the existing KGs are designed to represent and extract binary relations → good enough for search engines
- Applications like QA demand in depth knowledge about higher level structures like activities, events, processes

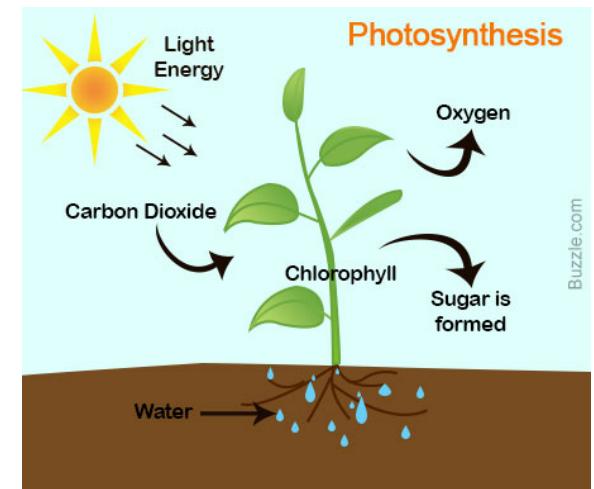
Future research directions: Going beyond facts

- **Fact:** Individual knowledge tuples
(plant, take in, CO₂)

- **Event frame:**
more context how, when, where?

subject	plant
predicate	Take in
object	CO ₂
time	daytime

- **Processes:**
representing larger structures, sequence of events
e.g. Photosynthesis



(2) Future research directions: Online KG Construction

- One shot KG construction → Online KG construction
 - Consume online stream of data
 - Temporal scoping of facts
 - Discovering new concepts automatically
 - Self-correcting systems

(2) Future research directions: Online KG Construction

- **Continuously learning and self-correcting systems**
 - *[Selecting Actions for Resource-bounded Information Extraction using Reinforcement Learning, Kanani and McCallum, WSDM 2012]*
 - Presented a reinforcement learning framework for budget constrained information extraction
 - *[Never-Ending Learning, Mitchell et al. AAAI 2015]*
 - Tom Mitchell says “Self reflection and an explicit agenda of learning subgoals” is an important direction of future research for continuously learning systems.

Future.....



- Knowledge graphs construction systems

Can consume online stream of data at Web scale,

Represent context beyond facts,

Support domains like medicine, science to help humanity

Can correct its own mistakes over time

Thank You

