



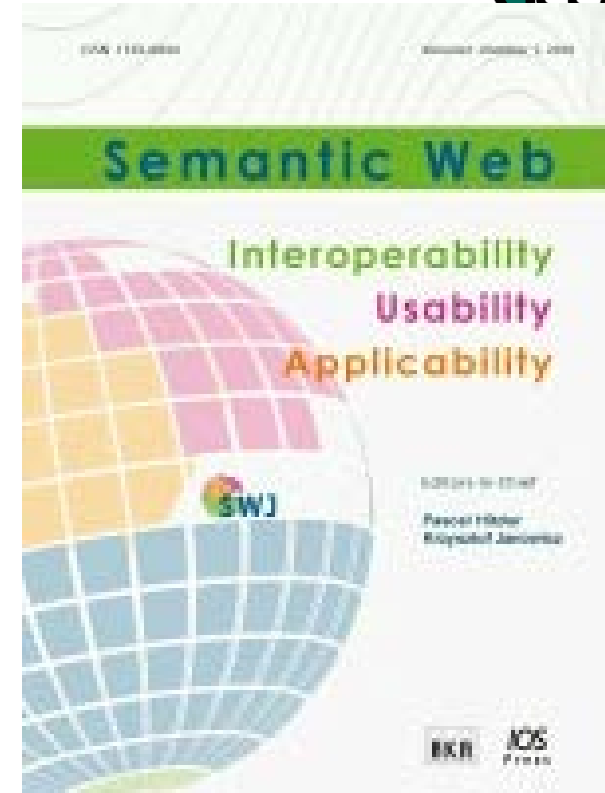
Towards Explaining Neural Networks Through Background Knowledge

Pascal Hitzler

Data Semantics Laboratory (DaSe Lab)
Data Science and Security Cluster (DSSC)
Wright State University
<http://www.pascal-hitzler.de>



- **EiCs:** Pascal Hitzler
Krzysztof Janowicz
- **Funded 2010**
- **2016 Impact factor of 1.786, top of all journals with “Web” in the title**
- **We very much welcome contributions at the “rim” of traditional Semantic Web research – e.g., work which is strongly inspired by a different field.**
- **Non-standard (open & transparent) review process.**



- **<http://www.semantic-web-journal.net/>**

Two faculty (Michelle Cheatham, Pascal Hitzler)

1 postdoc (Adila Krisnadhi)

10 graduate students



See <http://daselab.org/>

Data Science and Security Cluster (DSSC):

Combines 8 faculty, 6 labs (data semantics, bioinformatics, visualization, data science in healthcare, cybersecurity, web and complex systems)



Membership includes 8 faculty and over 40 graduate and undergraduate students across 6 distinct research labs:

- [Advanced Visual Data Analysis \(AVIDA\)](#), directed by [Thomas Wischgoll](#).
- [Bioinformatics Research Group \(BiRG\)](#), directed by [Travis Doom](#) and [Mike Raymer](#)
- [Cybersecurity Lab](#), directed by [Junjie Zhang](#)
- [Data Science for Healthcare Lab](#), directed by [Tanvi Banerjee](#)
- [Data Semantics \(DaSe\) Lab](#), directed by [Michelle Cheatham](#) and [Pascal Hitzler](#)
- [Web and Complex Systems \(WaCS\) Lab](#), directed by [Derek Doran](#)



- A research field about methods for:

Data and Information sharing, discovery, integration, and reuse.

Key paradigms:

- Representation of information via knowledge graphs in standardized formats (e.g., W3C's RDF).
- Typing of the knowledge graphs together with a type logic a.k.a. ontology or schema, represented in standardized/sharable formats (e.g., W3C's OWL)



Two major examples of semantic web technologies at work:

- **Google knowledge graph**
You see a glimpse of it in the boxes to the right of your search results.
- **Schema.org**
Joint effort by major search engine providers.
Schema/ontology for annotating Web page content, so that search engines can provide better results.
In the meantime, schema.org annotations are ubiquitous on the Web.

Google Knowledge Graph



Angela Merkel



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Angela Merkel - Wikipedia

https://en.wikipedia.org/wiki/Angela_Merkel ▼

Angela Dorothea Merkel is a German politician who is currently Chancellor of Germany. She is also the leader of the Christian Democratic Union (CDU). [Merkel ...](#)

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Angela Merkel – Wikipedia

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Angela Dorothea Merkel (* 17. Juli 1954 in Hamburg als [Angela Dorothea Kasner](#)) ist eine deutsche Politikerin (CDU) und seit dem 22. November 2005 ...

[Joachim Sauer](#) · [Horst Kasner](#) · [Angela Merkel](#) · [Liste der Auslandsreisen](#)

Top stories



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Angela Merkel - Forbes

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Angela Merkel on Forbes. ... Merkel, who faces a challenging reelection bid in 2017, has been tasked with maintaining a united European front in the wake of



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Angela Merkel

Chancellor of Germany

angela-merkel.de

Angela Dorothea Merkel is a German politician who is currently Chancellor of Germany. She is also the leader of the Christian Democratic Union. [Wikipedia](#)

Born: July 17, 1954 (age 62), Hamburg

Height: 1.65 m

Party: Christian Democratic Union of Germany

President: Horst Köhler; Christian Wulff; Joachim Gauck; Frank-Walter Steinmeier

Spouse: Joachim Sauer (m. 1998), Ulrich Merkel (m. 1977–1982)

Siblings: Marcus Kasner, Irene Kasner

Profiles



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Angela Merkel > Spouse > Joachim Sauer

Joachim Sauer
m. 1998



Ulrich Merkel
m. 1977–1982

Joachim Sauer - Wikipedia

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Joachim Sauer (born 19 April 1949) is a German quantum chemist and full professor at the Humboldt University of Berlin. He is the husband of the Chancellor of ...
Scientific career · Personal life · Public visibility as husband ... · References

Joachim Sauer – Wikipedia

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Joachim Sauer (* 19. April 1949 in Hosena, Landkreis Oberspreewald-Lausitz, Brandenburg) ist ein deutscher Quantenchemiker sowie Physikochemiker.
Berufliche Laufbahn · Ehrungen und Auszeichnungen · Privates · Werke (Auswahl)

Prof. Dr. Dr. h.c. Joachim Sauer — Quantenchemie der Festkörper ...

<https://www.chemie.hu-berlin.de/de/forschung/quantenchemie/Group/js-1> ▼

Year of Birth, 1949. Education, 1967 - 1972, Humboldt University, Berlin. Academic degrees, 1972, Diploma in Chemistry. 1974, Dr. rer. nat. (summa cum laude)

Joachim Sauer wird 65: „Der Professor“ statt der Kanzlerin-Gatte

www.handelsblatt.com > Politik > Deutschland ▼ Translate this page

Apr 19, 2014 - Er ist Deutschlands „First Husband“ – aber er tritt nicht als Ehemann der Kanzlerin in Erscheinung: Joachim Sauer lebt sein eigenes Leben als ...

Joachim Sauer, Angela Merkel's Husband: 5 Fast Facts | Heavy.com

heavy.com/.../joachim-sauer-angela-merkel-husband-married-germany-chancellor-sci... ▼

Mar 17, 2017 - Joachim Sauer is the media-shy husband of German Chancellor Angela Merkel, who is meeting President Donald Trump in Washington today.

Rare sighting of Angela Merkel's publicity-shy husband at G7 summit

www.telegraph.co.uk > News ▼



Joachim Sauer

Chemist

Joachim Sauer is a German quantum chemist and full professor at the Humboldt University of Berlin. He is the husband of the Chancellor of Germany, Angela Merkel.
Wikipedia

Born: April 19, 1949 (age 68), Hosena, Senftenberg

Spouse: Angela Merkel (m. 1998)

Education: Humboldt University of Berlin

Children: Adrian Sauer, Daniel Sauer

Parents: Richard Sauer, Elfriede Sauer

Awards: Kołos Medal

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Schema.org



- Collaboratively launched in 2011 by Google, Microsoft, Yahoo, Yandex.
2011: 297 classes, 187 relations
2015: 638 classes, 965 relations
- Simple schema, request to web site providers to annotate their content with schema.org markup. Promise: They will make better searches based on this.
- 2015: 31.3% of Web pages have schema.org markup, on average 26 assertions per page.

Ramanathan V. Guha, Dan Brickley, Steve Macbeth:
Schema.org: Evolution of Structured Data on the
Web. ACM Queue 13(9): 10 (2015)

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Resource Description Framework (RDF), W3C Standard



- Use URIs to make entities and vocabularies referenceable on the Web.
- Essentially, a labelled and typed graph.

```
@prefix : <https://w3id.org/rdfchess/id/> .
@prefix xsd: <http://www.w3.org/2001/XMLSchema#> .
@prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix chon: <https://w3id.org/rdfchess/chessonto#> .
```

```
:gam19e02 rdf:type chon:ChessGame ;
  chon:subEventOf :tou23as5 ;
  chon:atPlace :pla537es8 ;
  chon:atTime "2013-11-09T00:00:00"^^xsd:dateTime ;
  chon:providesAgentRole :rol88y76c, :rol92z01m ;
  chon:hasResult :res23tu77h ;
  chon:hasOpening :ope662vn2 .
```

```
:tou23as5 rdf:type chon:ChessTournament ;
  chon:hasName "WCh 2013"^^xsd:string ;
  chon:atPlace :pla537es8 ;
```

```
@prefix : <https://w3id.org/rdfchess/id/> .
@prefix xsd: <http://www.w3.org/2001/XMLSchema#> .
@prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix chon: <https://w3id.org/rdfchess/chessonto#> .
```

```
:gam19e02 rdf:type chon:ChessGame ;
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  chon:atTime "2013-11-09T00:00:00"^^xsd:dateTime ;
  chon:providesAgentRole :rol88y76c, :rol92z01m ;
  chon:hasResult :res23tu77h ;
  chon:hasOpening :ope662vn2 .
```

```
:tou23as5 rdf:type chon:ChessTournament ;
  chon:hasName "WCh 2013"^^xsd:string ;
  chon:atPlace :pla537es8 ;
  chon:atTime "2013-11-09T00:00:00"^^xsd:dateTime .
```

```
:pla537es8 rdf:type chon:Place ;
  chon:hasName "Chennai"^^xsd:string .
```

```
:rol88y76c rdf:type chon:WhitePlayerRole ; chon:performedBy :ag422yt6 .
```

```
:ag422yt6 rdf:type chon:Agent ;
  chon:hasName "Carlsen, Magnus"^^xsd:string .
```

```
:rol92z01m rdf:type chon:BlackPlayerRole ; chon:performedBy :ag79yy12 .
```

```
:ag79yy12 rdf:type chon:Agent ;
  chon:hasName "Anand, Viswanathan"^^xsd:string .
```

```
:res23tu77h rdf:type chon:ChessGameResult ;
  chon:encodedAsSAN "1/2-1/2"^^xsd:string .
```

```
:ope662vn2 rdf:type chon:ChessOpening ;
  chon:hasECOCode "A07"^^xsd:string .
```

```
:cgr448uy6 rdf:type chon:ChessGameReport ;
  chon:providesAgentRole :rol08ii2a ;
```

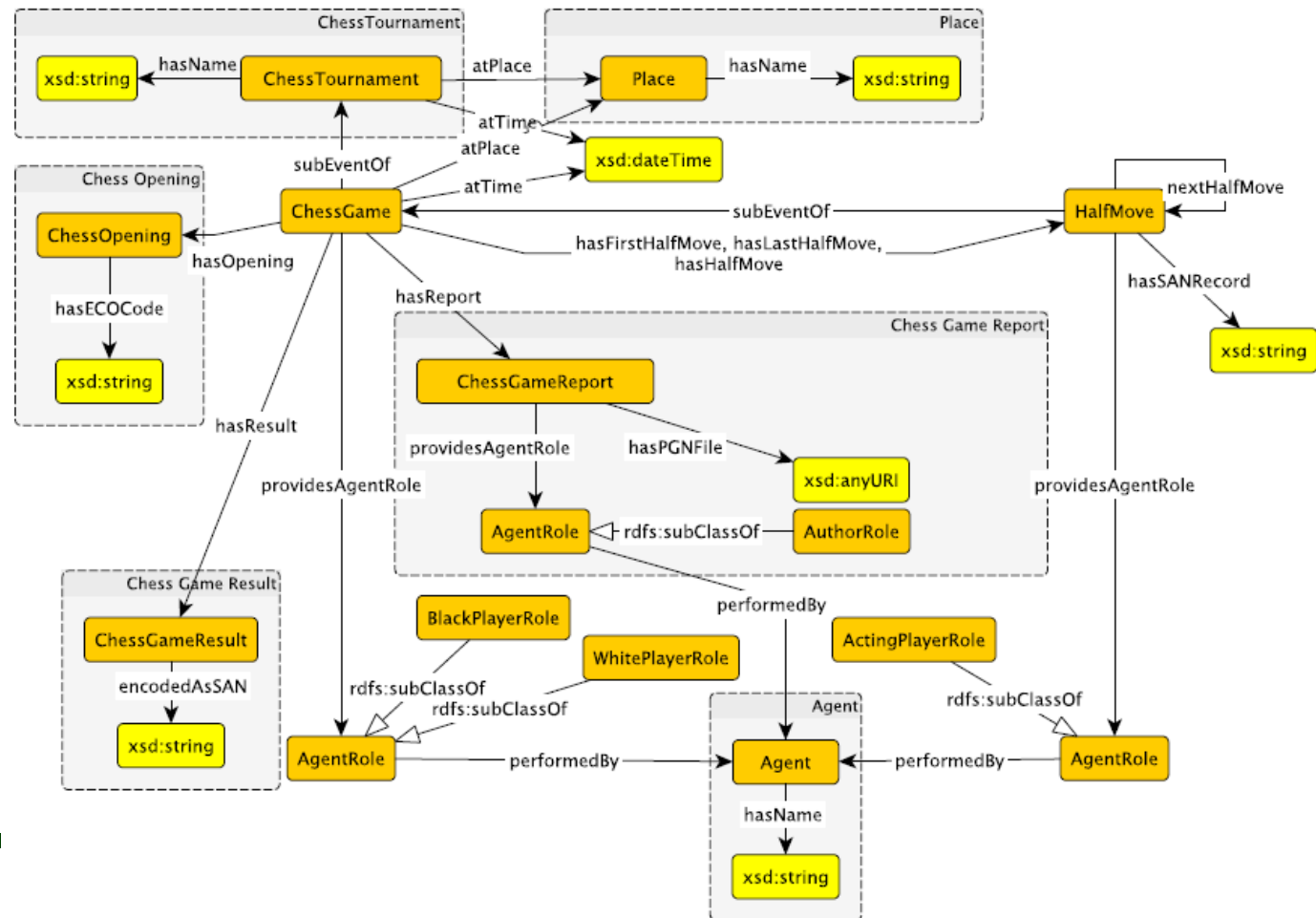
Se Lab



Representation format

Web Ontology Language (OWL), W3C Standard

- Essentially, a type logic over the types in the RDF graph
- Based on description logics, enables automated reasoning.



AgentRole \sqsubseteq ($=1$ performedBy.Agent) \sqcap \forall performedBy.Agent (10.1)

\exists performedBy.Agent \sqsubseteq AgentRole (10.2)

\top \sqsubseteq \forall pAR.AgentRole (10.3)

ChessGame \sqsubseteq \exists atPlace.Place \sqcap \forall atPlace.Place (10.4)

ChessGame \sqsubseteq \exists atTime.xsd:dateTime \sqcap \forall atTime.xsd:dateTime (10.5)

ChessGame \sqsubseteq \exists pAR.BlackPlayerRole \sqcap \exists pAR.WhitePlayerRole (10.6)

\exists subEventOf.ChessTournament \sqcup \exists hasOpening.ChessOpening \sqsubseteq ChessGame (10.7)

\exists hasResult.ChessGameResult \sqcup \exists hasReport.ChessGameReport \sqsubseteq ChessGame (10.8)

ChessGame \sqsubseteq \forall subEventOf.ChessTournament \sqcap \forall hasOpening.ChessOpening (10.9)

ChessGame \sqsubseteq \forall hasResult.ChessGameResult \sqcap \forall hasReport.ChessGameReport (10.10)

BlackPlayerRole \sqcup WhitePlayerRole \sqsubseteq AgentRole \sqcap ($=1$ pAR⁻.ChessGame) (10.11)

ChessGame \sqsubseteq ($=1$ hasFirstHalfMove.HalfMove) \sqcap ($=1$ hasLastHalfMove.HalfMove) (10.12)

ChessGame \sqsubseteq ($=1$ hasLastHalfMove.HalfMove) (10.13)

hasHalfMove \sqsubseteq subEventOf⁻ (10.14)

hasFirstHalfMove \sqsubseteq hasHalfMove (10.15)

hasLastHalfMove \sqsubseteq hasHalfMove (10.16)

HalfMove \sqsubseteq Event \sqcap \exists pAR.ActingPlayerRole \sqcap ($=1$ hasHalfMove⁻.ChessGame) (10.17)

ActingPlayerRole \sqsubseteq AgentRole \sqcap ($=1$ pAR⁻.HalfMove) (10.18)

HalfMove \sqsubseteq (≤ 1 nextHalfMove.HalfMove) \sqcap $\neg \exists$ nextHalfMove.Self (10.19)

\exists subEventOf.ChessGame \sqcup \exists nextHalfMove.HalfMove \sqsubseteq HalfMove (10.20)

\exists hasSANRecord.xsd:string \sqsubseteq HalfMove (10.21)

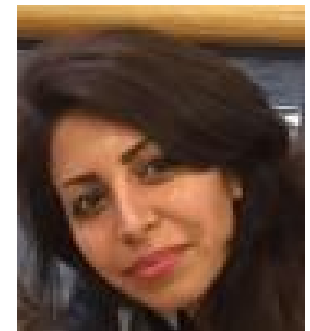
HalfMove \sqcap \forall subEventOf.ChessGame \sqcap \forall nextHalfMove.HalfMove (10.22)

- **Given an ontology and some data (or text), how to automatically transform the data such that it will adhere to the schema defined by the ontology? (Ontology/Knowledge base population)**
- **Given two ontologies, how to map between the ontologies such that data can be integrated? (Ontology alignment)**
- **How to make ontologies, which make these (and other) data management tasks as simple as possible?**



Propositional rule extraction from trained neural networks under background knowledge

(work with Maryam Labaf)



Neural-Symbolic Methodology

high-level symbolic representations
(abstraction, recursion, relations, modalities)



translations



low level, efficient neural structures
(with the same, simple architecture throughout)

Analogy: low-level implementation (machine code) of
high-level representations (e.g. java, requirement specs)





In this case: extracting propositional rules.

General idea:

- Input value 1 interpreted as “true”, value 0 as “false”
- Outputs interpreted as true or false according to a threshold
- I.e. network function maps binary vectors.

Garcez et al, 2001: By weight analysis (layer by layer) under differentiable activation functions. Possible in principle but intricate and, arguably, the resulting rule sets are usually rather difficult to understand.

Lehmann, Bader, Hitzler, 2010: Black-box approach (looking at inputs and outputs only).

For every monotonic function

$$f : \{0, 1\}^n \rightarrow \{0, 1\}^k$$

there is a unique reduced set of positive propositional rules which capture exactly the function f .

Reduced means: no redundancies, and as small as possible.

Problem: Rule sets can get large and messy, i.e. still very difficult to understand.





Can we lift the result just given to include background knowledge?

Given:

- A (reduced) propositional logic program P (extracted from an ANN as above).
- Set I of prop. variables representing ANN inputs.
- Set O of prop. variables representing ANN outputs.
- A background knowledge base K (a propositional logic program).

We then seek a logic program P' (simpler than P)
s.t. for all subsets i in I and each o in O we have

$$P \wedge i \models o \quad \text{iff} \quad P' \wedge K \wedge i \models o.$$



It turns out that

- **P'** is no longer unique in general (even under reduction).
- **P'** may not even exist (unless **I** is restricted to the left-hand side of rules in **K**).
- But with suitable **K** you can get **P'** which are simpler than **P**.
Typical case:

$$\begin{array}{lll} \mathbf{P}: & p_1 \wedge q \rightarrow o & \mathbf{K}: \quad p_1 \rightarrow r \\ & p_2 \wedge q \rightarrow o & \mathbf{P}': \quad r \wedge q \rightarrow o \\ & & \quad \quad p_2 \rightarrow r \end{array}$$



$$\mathbf{P}: \quad p_1 \wedge q \rightarrow o \qquad \mathbf{K}: \quad p_1 \rightarrow r$$
$$p_2 \wedge q \rightarrow o \qquad p_2 \rightarrow r$$

$$\mathbf{P}': \quad r \wedge q \rightarrow o$$

Note that K essentially groups input variables. One could think of r being a “more general concept” than either p1 and p2.

Of course, we have only discussed the propositional case so far, but in order to obtain strong explanations for the input-output behavior of ANNs we need to go beyond propositional.

Comprehensibility of ILP-learned Programs

Inductive Logic Programming and Predicate Invention:

: grandparent without PI

```
p(X,Y) :- father(X,Z), father(Z,Y).  
p(X,Y) :- father(X,Z), mother(Z,Y).  
p(X,Y) :- mother(X,Z), mother(Z,Y).  
p(X,Y) :- mother(X,Z), father(Z,Y).
```

: grandparent with PI

```
p(X,Y) :- p1(X,Z), p1(Z,Y).  
p1(X,Y) :- father(X,Y).  
p1(X,Y) :- mother(X,Y).
```

: ancestor without PI

```
p(X,Y) :- father(X,Y).  
p(X,Y) :- mother(X,Y).  
p(X,Y) :- father(X,Z), p(Z,Y).  
p(X,Y) :- mother(X,Z), p(Z,Y).
```

: greatgrandparent without PI

```
p(X,Y) :- father(X,U), father(U,Z), father(Z,Y).  
p(X,Y) :- father(X,U), father(U,Z), mother(Z,Y).  
p(X,Y) :- father(X,U), mother(U,Z), father(Z,Y).  
p(X,Y) :- father(X,U), mother(U,Z), mother(Z,Y).  
p(X,Y) :- mother(X,U), father(U,Z), mother(Z,Y).  
p(X,Y) :- mother(X,U), father(U,Z), father(Z,Y).  
p(X,Y) :- mother(X,U), mother(U,Z), mother(Z,Y).  
p(X,Y) :- mother(X,U), mother(U,Z), father(Z,Y).
```

: greatgrandparent with PI

```
p(X,Y) :- p1(X,U), p1(U,Z), p1(Z,Y).  
p1(X,Y) :- father(X,Y).  
p1(X,Y) :- mother(X,Y).
```

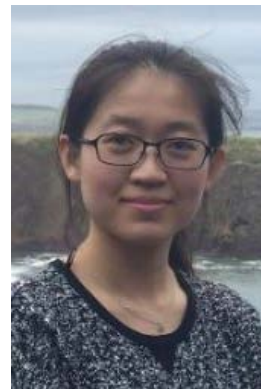
: ancestor with PI

```
p(X,Y) :- p1(X,Y).  
p(X,Y) :- p1(X,Z), p(Z,Y).  
p1(X,Y) :- father(X,Y).  
p1(X,Y) :- mother(X,Y).
```

Example Prolog programs for family relations (with and without the use of Predicate Invention).

Description Logic extraction from trained neural networks under background knowledge

(work with Md Kamruzzaman Sarker, Derek Doran, Ning Xie, Mike Raymer)





- Explain input-output behavior of trained (deep) NNs.
- Idea:
 - Use background knowledge in the form of linked data and ontologies to help explain.
 - Link inputs and outputs to background knowledge.
 - Use a symbolic learning system (e.g., DL-Learner) to generate an explanatory theory.
- We're just starting on this, experiments (below) just came out.

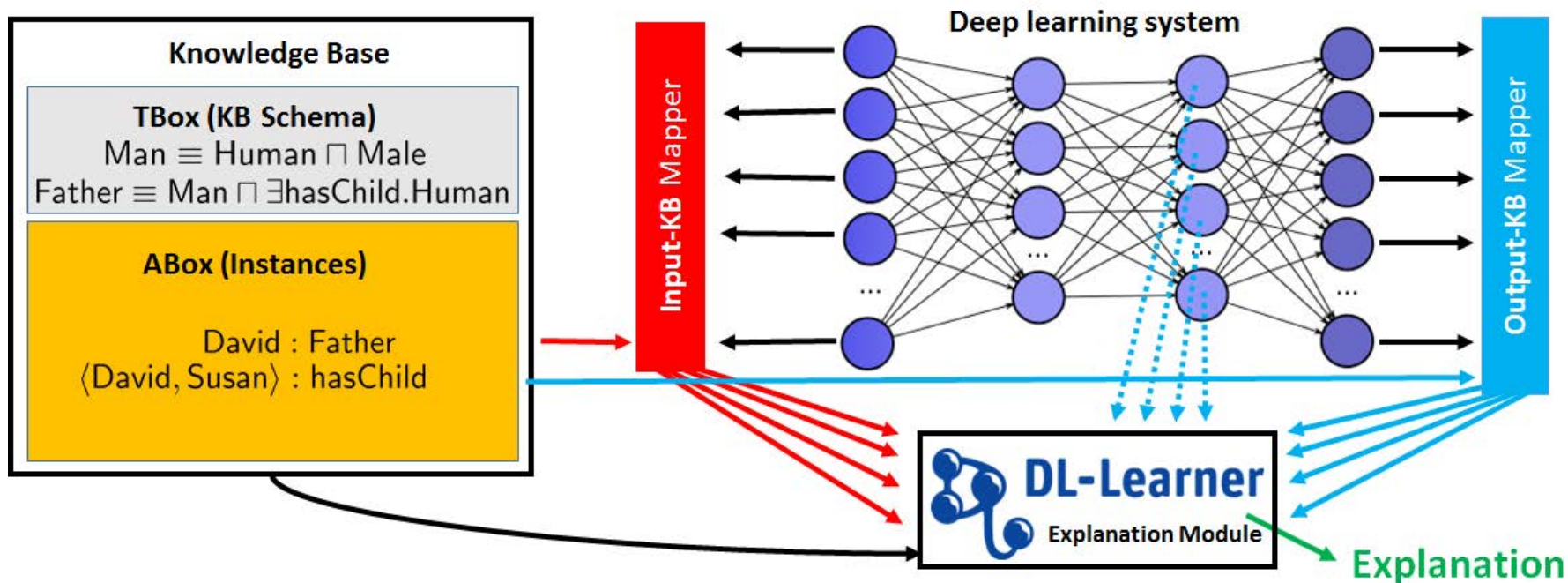


Possible data sources:

- **Linked data / semantic web data**
 - I.e. structured data on the web, organized in so-called RDF graphs.
- **Cross-domain ontologies (e.g., SUMO, Proton)**
- **Wikidata**
- **schema.org**

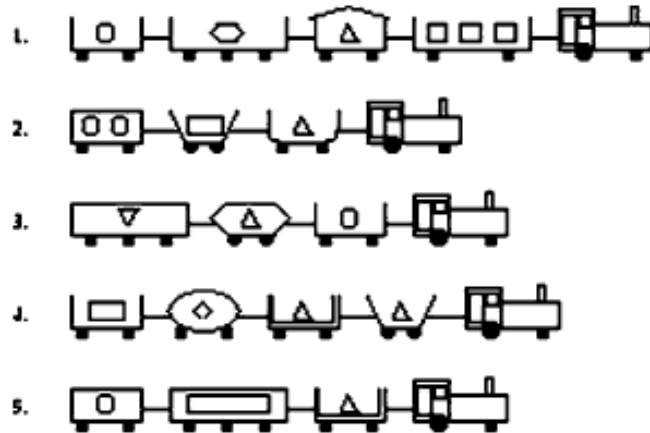
Essentially, all content already readily and publicly available in structured form.

If further domain knowledge is needed: use state-of-the-art approaches for knowledge graph generation in order to obtain structured data from suitable text corpora.

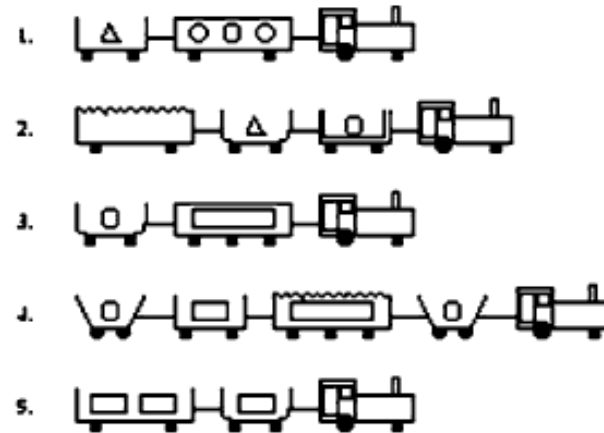


Approach similar to inductive logic programming, but using Description Logics (the logic underlying OWL).

Positive examples:



negative examples:



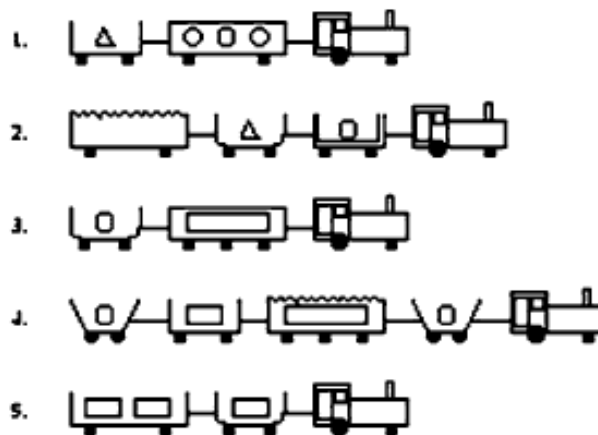
Task: find a class description (logical formula) which separates positive and negative examples.



Positive examples:



negative examples:



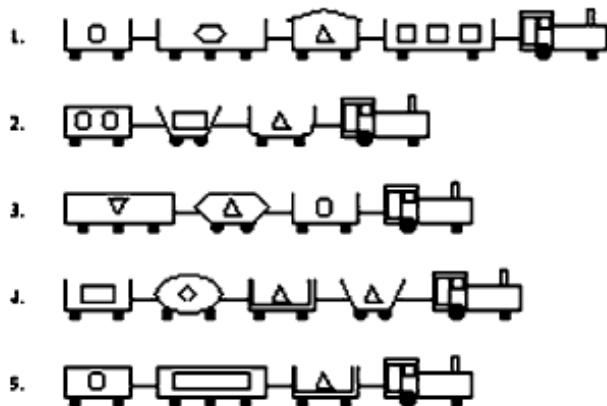
DL-Learner result:

$\exists \text{hasCar} . (\text{Closed} \sqcap \text{Short})$

In FOL:

$$\{x \mid \exists y (\text{hasCar}(x, y) \wedge \text{Closed}(y) \wedge \text{Short}(y))\}$$

DL-Learner uses refinement operators to construct ever better approximations of a solution.



\top

Train – covers all examples.

$\exists \text{hasCar}.\top$

$\exists \text{hasCar}.\text{Closed}$ – covers all positives, two negatives

$\exists \text{hasCar}(\text{Closed} \sqcap \text{Short})$ – solution

Proof of Concept Experiment

Positive:



Negative:



Come from the MIT ADE20k dataset

<http://groups.csail.mit.edu/vision/datasets/ADE20K/>

They come with annotations of objects in the picture:

```
001 # 0 # 0 # sky # sky # ""
002 # 0 # 0 # road, route # road # ""
005 # 0 # 0 # sidewalk, pavement # sidewalk # ""
006 # 0 # 0 # building, edifice # building # ""
007 # 0 # 0 # truck, motortruck # truck # ""
008 # 0 # 0 # hovel, hut, hutch, shack, shanty # hut # ""
009 # 0 # 0 # pallet # pallet # ""
011 # 0 # 0 # box # boxes # ""
001 # 1 # 0 # door # door # ""
002 # 1 # 0 # window # window # ""
009 # 1 # 0 # wheel # wheel # ""
```



Mapping to SUMO

Simple approach: for each known object in image, create an individual for the ontology which is in the appropriate SUMO class:

- contains road1**
- contains window1**
- contains door1**
- contains wheel1**
- contains sidewalk1**
- contains truck1**
- contains box1**
- contains building1**





- Suggested Merged Upper Ontology
<http://www.adampease.org/OP/>
- Approx. 25,000 common terms covering a wide range of domains
- Centrally, a relatively naïve class hierarchy.
- Objects in image annotations became individuals (constants), which were then typed using SUMO classes.



Positive:

img1: road, window, door, wheel, sidewalk, truck, box, building

img2: tree, road, window, timber, building, lumber

img3: hand, sidewalk, clock, steps, door, face, building, window, road

Negative:

img4: shelf, ceiling, floor

img5: box, floor, wall, ceiling, product

img6: ceiling, wall, shelf, floor, product

DL-Learner results include:

\exists contains.Transitway

\exists contains.LandArea

Proof of Concept Experiment

Positive:



Negative:



\exists contains.Transitway

\exists contains.LandArea



- | | | | |
|--|-----|------------------------------------|------|
| \exists contains.Window | (1) | \exists contains.LandTransitway | (6) |
| \exists contains.Transitway | (2) | \exists contains.LandArea | (7) |
| \exists contains.SelfConnectedObject | (3) | \exists contains.Building | (8) |
| \exists contains.Roadway | (4) | \forall contains. \neg Floor | (9) |
| \exists contains.Road | (5) | \forall contains. \neg Ceiling | (10) |

Experiment 2

Positive (selection):



Negative (selection):



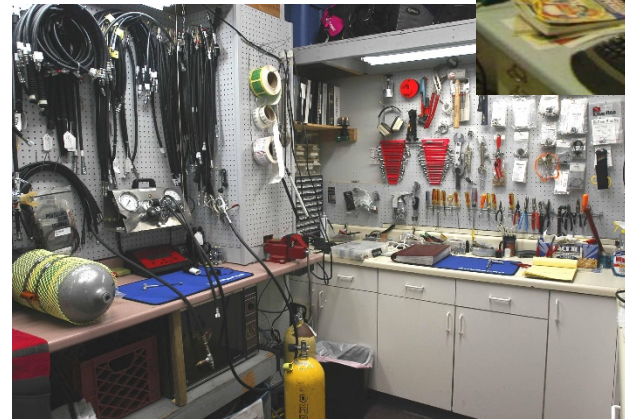
∃contains. (DurableGood \sqcap \neg ForestProduct)

Experiment 3

Positive:



Negative:



$\forall \text{contains.} (\neg \text{Furniture} \sqcap \neg \text{IndustrialSupply})$

Experiment 4

Positive (selection):



Negative (selection):



Does not contain a Sentient Agent

Experiment 5

Positive:



Negative (selection):



\exists contains.BodyOfWater



- Utilize more sophisticated ontology.
- Utilize more sophisticated mappings.
- Explain hidden neurons.
- Tune DL-Learner better to the specific task.

Collaborators Derek Doran and Ning Xie (Web and Complex Systems Lab)



They explore how to determine groups of hidden neurons which often fire together and thus may indicate the “detection” of certain features.

We plan to apply the above mentioned DL-Learner approach also to these groups of hidden neurons, in order to determine which features they detect.

Thanks!



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