


Extracting Literal Assertions for DBpedia from Wikipedia Abstracts



Florian Schrage, Nicolas Heist, **Heiko Paulheim**

Flashback: ISWC 2017

- Heist, Paulheim (2017): “Language-agnostic relation extraction from Wikipedia abstracts”
- Main idea:
 - Find recurring patterns in abstracts

Reinheim 

From Wikipedia, the free encyclopedia

For Reinheim in Saarland, see [Gersheim](#).

Reinheim is a town in the [Darmstadt-Dieburg](#) district, in [Hesse](#), [Germany](#). It is situated 14 kilometres (9 miles) southeast of [Darmstadt](#).

municipality


state

country

Groß-Bieberau 


From Wikipedia, the free encyclopedia

Groß-Bieberau is a town in the [Darmstadt-Dieburg](#) district, in [Hesse](#), [Germany](#). It is situated 15 km southeast of [Darmstadt](#). It has several sister cities.

Hofgeismar 

From Wikipedia, the free encyclopedia

Hofgeismar is a town in the [district of Kassel](#), in northern [Hesse](#), [Germany](#). It is located 25 km north of [Kassel](#) on the [German Timber-Frame Road](#). In 1978, the town hosted the 18th [Hessentag](#) state festival.

Modoc, Illinois 

From Wikipedia, the free encyclopedia

Modoc is an [unincorporated community](#) in [Randolph County](#), [Illinois](#), [United States](#), located four miles southeast of [Prairie du Rocher](#) under the bluffs of the [Mississippi River](#).

Heist, Paulheim: *Language-agnostic relation extraction from Wikipedia Abstracts*. In: *ISWC 2017*

Observation: Typical Patterns

- The first three populated places linked in an abstract about a town are that town's *municipality*, *state*, and *country*
- All genres linked in an abstract about a writer are that writer's *genres*
- The first place linked in an abstract about a person is that person's *birthplace*

- Automatically finding those patterns:
We can use existing relations as training data
 - Using a *local closed world assumption* for creating negative examples
- Training data:
 - Linked instances in an abstract, explicit relations extracted from infobox

From Entities to Numbers and Dates

- Key assumption: such patterns also exist for numbers and dates
- Examples:
 - First date in an abstract about a *person* is the person's birthdate
 - First number in an abstract about a *city* is the city's population
- Differences to entity-based extraction (aka: challenges)
 1. numbers/dates are neither tagged nor typed
 2. numbers/dates come in different formats
 3. infobox value and value in abstract may use a different format and/or unit of measure and/or rounding

Trent Reznor

From Wikipedia, the free encyclopedia

Michael Trent Reznor (born May 17, 1965) is an American singer, : *Inch Nails*, which he founded in 1988 and of which he was the sole of album *Pretty Hate Machine*, was a commercial and critical success. F *Columbia Records* in 2012.

Mannheim

From Wikipedia, the free encyclopedia

This article is about the city in Germany. For other uses, see [Mannh](#)

Mannheim (German pronunciation: [ˈmanhaim] ⓘ[ⓘ] listen); *Palatine German* and *Karlsruhe* with a 2015 population of approximately 305,000 inhabita Germany's eighth-largest metropolitan region.

Challenge: Number/Date Formats

- Sometimes even inconsistent *within* a single Wikipedia page

Hohwart

From Wikipedia, the free encyclopedia

Coordinates:  47°56′40.49″N 8°3′36.68″E

The **Hohwart** is a mountain near the village of [Breitnau](#) in the [Black Forest](#) in the German state of [Baden-Wurtemberg](#). It is **1,123** metres high.

The Hohwart lies on an east-west ridge behind Breitnau. To the south the ridge slopes down over cattle pasture to the Breitnau bowl. To the north the slopes are wooded and descend towards the valley of the Griesdobelbach. The ridge continues northwest to the Otten (**1040** m). Just over 1 km to the east is the [Roßberg](#) (**1125** m) and 3 km to the east, across the Oberbach valley, is the highest mountain in Breitnau municipality: the [Weißtannenhöhe](#) (**1190** m).



Paulheim: *A Robust Number Parser based on Conditional Random Fields*. In: KI 2017

Challenge: Infobox vs. Text Mismatch

Baden-Württemberg (/ˌbɑːdən ˈvɜːrtɛmbɜːrɡ/,^[5] German: [ˌbaːdŋ ˈvʏʁtɛmbɛʁk] ^(listen)) is a [state](#) in southwest [Germany](#), east of the [Rhine](#), which forms the border with France. It is Germany's third-largest state, with an area of 35,751 km² (13,804 sq mi) and [11 million](#) inhabitants.^[6] Baden-Württemberg is a [parliamentary](#)

Bergpark Wilhelmshöhe is a landscape park in [Kassel](#), [Germany](#). The area of the park is [2.4 square kilometres](#) (590 acres), making it the largest European hillside park, and second largest park on a hill slope in the world. Construction of the *Bergpark*, or "mountain park", began in 1689 at the behest of the [Landgraves of Hesse-Kassel](#) and took about 150 years. The park is open to the public today. Since 2013, it has been a [UNESCO World Heritage Site](#).

Area ^[1]	
• Total	35,751.46 km ² (13,803.72 sq mi)
Population (2017-12-31) ^[2]	
• Total	11,023,424
• Density	310/km ² (800/sq mi)

Location	Kassel , Hesse , Germany
Criteria	Cultural: (iii), (iv)
Reference	1413
Inscription	2013 (37th Session)
Area	558.7 ha (1,381 acres)
Buffer zone	2,665.7 ha (6,587 acres)
Coordinates	 51°18′57″N 09°23′35″E

Creating Training Data

- First step: spot any character sequences containing numbers
 - Those could be numbers, dates, and others
- Second step:
 - Try to parse sequences with spaCy and dateparser
 - Tolerant, language-independent Python based number and date parsers

The logo for spaCy, featuring the word "spaCy" in a blue, lowercase, sans-serif font. The "C" is significantly larger than the other letters.The logo for dateparser 0.7.1, featuring the text "dateparser 0.7.1" in white, lowercase, sans-serif font on a blue rectangular background.A code block showing the command "pip install dateparser" in white, lowercase, sans-serif font on a blue rectangular background. To the right of the code is a small white icon of a document with a checkmark.

Creating Training Data

- Challenge: abstracts often use rounded values
 - Or there are slight deviations
 - Experimented with 1%, 1.5%, 2% tolerance
 - Precision drops at 2% → we use 1.5%
- Gain: more training data
- Loss: false positives

Kassel

From Wikipedia, the free encyclopedia

For other uses, see [Kassel \(disambiguation\)](#).

Kassel (German pronunciation: [ˈkasl] listen[ⓘ]); spelled **Cassel** until 1928 is a city located on the [Fulda River](#) in northern [Hesse](#), Germany. It is the administrative seat of the [Regierungsbezirk Kassel](#) and the district of the same name and had 200,507 inhabitants in December 2015. The former capital of the state of Hesse-Kassel has many palaces and parks, including the [Bergpark Wilhelmshöhe](#), which is a [UNESCO World Heritage Site](#). Kassel is also known for the *documenta* exhibitions of contemporary art. Kassel has a public university with 25,000 students (2018) and a multicultural population (39% of the citizens in 2017 had a migration background).

Country	 Germany
State	Hesse
Admin. region	Kassel
District	Urban district
Government	
 • Mayor	Christian Geselle (SPD)
Area ^[1]	
 • City	106.8 km ² (41.2 sq mi)
Elevation	167 m (548 ft)
Population (2017-12-31) ^[2]	
 • City	200,736
 • Density	1,900/km ² (4,900/sq mi)
 • Metro	450,000
Time zone	CET/CEST (UTC+1/+2)
Postal codes	34001–34134
Dialling codes	0561
Vehicle registration	KS
Website	www.stadt-kassel.de

Creating Training Data

- Challenge: different units of measure, mixed number-text notation (e.g., “3.4 million”)
- Approach: train a linear ($b=0$) model for context words
 - i.e., context words can be linked to linear factors
 - Accept models with at least 100 examples and R^2 value >0.85

Table 1. Examples for unit conversions learned from the data.

Token	Target Unit	Correct Factor	Inferred Factor	R Squared
<i>km²</i>	<i>m²</i>	1,000,000	997,097	0.9949
<i>km2</i>	<i>m²</i>	1,000,000	999,927	0.9999
<i>ha</i>	<i>m²</i>	10,000	9,467	0.8987
<i>pupils</i>	\$	–	13,613	0.9062
<i>kilometers</i>	<i>m</i>	1,000	973	0.9347
<i>century</i>	<i>m</i>	–	73,453	0.9421

Overall Approach

- Extract numbers from abstract
- Match them to numbers in the infobox
 - Matching: positive example
 - Non-matching: negative example
- Train a classifier
 - Self-assessment: estimate precision
 - Only classifiers above 95% precision are used to produce statements



Experiments

- Training example generation
 - Extracted by identifying matching pairs in abstract and infobox
 - Allowing deviation and linear factors (as above)
 - Negatives: non-matching numbers/dates in the same abstract
- Datasets used for classification (true/false extraction)
 - DBpedia 2016-10 and corresponding Wikipedia dump
 - 120 number and date valued properties
w/ at least 100 positive training examples
 - 120 classifiers trained
 - 75%/25% split to allow self-assessment of trained models
 - 28 reach a precision >95%

Experiments

- Feature set
 - Motivation: patterns such as “The first number in an abstract...”
 - Features used: position in sentence, sentence in abstract, ...
 - Plus: bag of words around literal (e.g., “birth”, “population”, ...)
 - For numbers: deviation from mean
- Classifiers
 - SGD, Naive Bayes, SVM, Decision Trees, Random Forest, Extra Trees, Bagging Decision Trees, XGBoost
 - RandomForest used and fine-tuned after initial experiment



Results

- 28 properties for which a model with 95% precision is trained
 - Those generate 9M facts
 - 7% are not contained in DBpedia
 - Mostly dates, not numbers

Range	Properties	Statements	New Statements
Date	17	5,525,089	621,747
Int	6	224,606	15,326
Float	5	3,185,497	5,955
Total	28	8,955,192	643,030

Table 2. Number of statements extracted at 95% precision according to internal validation.

- Posterior validation on 500 newly generated facts
 - Precision is 94.2%
 - i.e., estimated precision is valid

Take Aways

- Literal-valued relations are challenging
- Tweaks to original entity-based approach
 - Number/date tagging and parsing
 - Tolerance intervals
 - Learned model for unit conversion
- 9M statements could be extracted (600k new)



- Code: <https://github.com/FlorianSchrage/DBpediaLiteralRelations>

Future Challenges

- Deeper analysis of deviations
 - Is the correct value more likely in the abstract or the infobox?
- Better training data and learning
 - Robustly discarding false matches
 - Learning models for smaller datasets
- Learning complex formulae
 - e.g., population density
- Transfer to other datasets
 - e.g., DBkWik

The logo for DBkWik, featuring the text 'DBkWik' in a bold, sans-serif font. The letters 'B', 'k', and 'W' are significantly larger and more prominent than the 'D', 'i', and 'i'.

Extracting Literal Assertions for DBpedia from Wikipedia Abstracts



Florian Schrage, Nicolas Heist, **Heiko Paulheim**