# Text Similarity

Eva Gerlitz

January 31, 2018

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#### Outline

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- 2 DBpedia
- 3 Training
  - Supervised
    - Preprocessing
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  - Unsupervised
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#### Introduction

- Take different classes of wikipedia articles (using DBpedia)
- Look at abstracts
- How well do different models learn to discriminate the classes?

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#### What is DBpedia?



Figure 1: DBpedia<sup>1</sup>

- Linked Data project
- Extracted, structured content from Wikipedia (Infoboxes)
- Possible to query properties and relationships

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<sup>&</sup>lt;sup>1</sup>Pic from https://en.wikipedia.org/wiki/DBpedia

#### DBpedia



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#### DBpedia - get data

• SPARQL queries <sup>2</sup>

```
PREFIX dbpedia0: <a href="http://dbpedia.org/ontology/">http://dbpedia.org/property/</a>
PREFIX dbpedia2: <a href="http://dbpedia.org/resource/">http://dbpedia.org/property/</a>
PREFIX dbpedia: <a href="http://dbpedia.org/resource/">http://dbpedia.org/resource/</a>
select distinct ?name ?abstract where {
    ?instance a dbpedia0: EducationalInstitution.
    ?instance foaf:name ?name.
    ?instance dbpedia0: abstract ?abstract.
    filter(langMatches(lang(?abstract),"en"))
}
```

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<sup>&</sup>lt;sup>2</sup>http://dbpedia.org/snorql

#### DBpedia - Classes

• Use Subclasses that are still large enough

- Organisation (edit)
  - Broadcaster (edit)
    - BroadcastNetwork (edit)
    - RadioStation (edit)
    - TelevisionStation (edit)
  - Company (edit)
    - Bank (edit)
    - Brewery (edit)
    - Caterer (edit)
    - LawFirm (edit)
    - PublicTransitSystem (edit)
      - Airline (edit)
      - BusCompany (edit)
    - Publisher (edit)
    - RecordLabel (edit)
    - Winery (edit)
  - EducationalInstitution (edit)
    - College (edit)
    - Library (edit)
    - School (edit)
    - University (edit)

Figure 2: Classes in DBpedia<sup>3</sup>

• DBpedia Data not perfect (College!)

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<sup>&</sup>lt;sup>3</sup>http://mappings.dbpedia.org/server/ontology/classes/

#### Datasets

- **Actor** (10.000)
- City (10.000)
- Celestial Body (7699)
- Educational Institution (10.000)
- Lake (10.000)
- Athlete (10.000)
- Fictional Character (10.000)
- Musical Artist (10.000)

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## Preprocessing - Hashing Vectorizer

- Turns abstracts into a sparse matrix
- No dictionary, but hashing function.

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## Preprocessing - Tfidf Vectorizer

- tf = term frequency, idf = inverse document frequency
- Term frequency = proportion of occurrences of a specific term to the total number of terms in a document
- Inverse document frequency = inverse of the proportion of documents that contain that word/phrase.

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#### Model 1: Random Forest

- Consists of many decision trees, that are built during training
- When predicting a class, every tree gives a vote
- The class with most votes wins

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## Model 2: K-nearest neighbors

- Training examples in feature space, dimension = number of features
- Look at k neighbors  $\rightarrow$  majority of classes will be predicted.

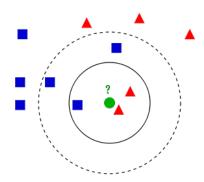


Figure 3: k-nearest neighbors<sup>4</sup>

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<sup>&</sup>lt;sup>4</sup>Pic from https://en.wikipedia.org/wiki/K-nearest\_neighbors\_algorithm

#### Model 3: Stochastic Gradient Descent

- Multiple binary classifiers, "One versus all" scheme
- For classification: compute confidence score

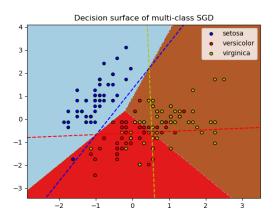


Figure 4: Multiple classes<sup>5</sup>

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<sup>&</sup>lt;sup>5</sup>Pic from http://scikit-learn.org/stable/modules/sgd.html

# Results - 2 Classes

Classes	Accuracy
Actors and Fictional Characters	0.98
Actors and Athletes	0.989
Actors and Cities	0.997
Actor and MusicalArtist	0.954
City and Lake	0.986
City and Educational Institution	0.989
Actor and Celestial Body	1.00

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#### Results - 5 Classes

$Preprocessing \rightarrow$	Hashing	tfidf
Models ↓		
Random Forest	0.986	0.986
k-nearest Neighbors	0.993	0.977
Stochastic Gradient Descent	0.996	0.994

 $\rightarrow$  About 140 mistakes for hashing & random forest.

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# Results - What went wrong?

Classes	Actor	City	Cel.	Edu. In-	Lake
			Body	stitution	
Actor	_	5/4/8	4/2/1	16/14/4	1/2/2
City	*	_	1/0/0	34/19/17	57/26/6
Cel.	*	*	_	0/0/0	2/0/0
Body					
Edu. In-	*	*	*	-	19/4/0
stitution					
Lake	*	*	*	*	-

Table 1: How often were the class objects mistaken? - Hashing Vectorizer, Random Forest, K-nearest neighbors, Stochastic Gradient Descent

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# Results - What went wrong?

Classes	Actor	City	Cel.	Edu. In-	Lake
			Body	stitution	
Actor	_	5/4/8	4/2/1	16/14/4	1/2/2
City	*	-	1/0/0	34/19/17	57/26/6
Cel.	*	*	-	0/0/0	2/0/0
Body					
Edu. In-	*	*	*	-	19/4/0
stitution					
Lake	*	*	*	*	-

Table 2: How often were the class objects mistaken? - Hashing Vectorizer, Random Forest, K-nearest neighbors, Stochastic Gradient Descent

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# Results - What went wrong?

Classes	Actor	City	Cel.	Edu. In-	Lake
			Body	stitution	
Actor	_	5/4/8	4/2/1	16/14/4	1/2/2
City	*	_	1/0/0	34/19/17	57/26/6
Cel.	*	*	_	0/0/0	2/0/0
Body					
Edu. In-	*	*	*	-	19/4/0
stitution					
Lake	*	*	*	*	-

Table 3: How often were the class objects mistaken? - Hashing Vectorizer, Random Forest, K-nearest neighbors, Stochastic Gradient Descent

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• "Pantnagar is a town and a university campus in Udham Singh Nagar district, Uttarakhand. Nainital, Kashipur, Rudrapur and Kiccha, Haldwani are the major cities surrounding Pantnagar.

The town is famous for having the first agricultural university of India which was established on 17 November 1960. Previously the university was called the Uttar Pradesh Agricultural University or Pantnagar University. It was rechristened G. B. Pant University of Agriculture and Technology. keeping in view the contributions of Pt. Govind Ballabh Pant, the then Chief Minister of UP.

In recent years, an integrated industrial estate has been established near the campus which houses companies such as Tata motor, Bajaj, Britannia, HP, HCL, Voltas, Schneider Electric, Nestle, Dabur, Vedanta Resources etc., as a part of SIDCUL industrial area developed by government owned State Industrial Development Corporation of Uttarakhand Limited."

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• "Pantnagar is a town and a university campus in Udham Singh Nagar district, Uttarakhand. Nainital, Kashipur, Rudrapur and Kiccha, Haldwani are the major cities surrounding Pantnagar.

The town is famous for having the first agricultural university of India which was established on 17 November 1960. Previously the university was called the Uttar Pradesh Agricultural University or Pantnagar University. It was rechristened G. B. Pant University of Agriculture and Technology. keeping in view the contributions of Pt. Govind Ballabh Pant, the then Chief Minister of UP.

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• Belongs to class **City** 

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• "Pantnagar is a town and a university campus in Udham Singh Nagar district, Uttarakhand. Nainital, Kashipur, Rudrapur and Kiccha, Haldwani are the major cities surrounding Pantnagar.

The town is famous for having the first agricultural university of India which was established on 17 November 1960. Previously the university was called the Uttar Pradesh Agricultural University or Pantnagar University. It was rechristened G. B. Pant University of Agriculture and Technology. keeping in view the contributions of Pt. Govind Ballabh Pant, the then Chief Minister of UP.

In recent years, an integrated industrial estate has been established near the campus which houses companies such as Tata motor, Bajaj, Britannia, HP, HCL, Voltas, Schneider Electric, Nestle, Dabur, Vedanta Resources etc., as a part of SIDCUL industrial area developed by government owned State Industrial Development Corporation of Uttarakhand Limited."

- Belongs to class City
- Labeled as: Educational Institution
- Problem: Overlapping of different classes

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• "Rangsit is a city in Pathum Thani Province, Thailand. Rangsit is the home of Rangsit University."

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- "Rangsit is a city in Pathum Thani Province, Thailand. Rangsit is the home of Rangsit University."
- Belongs to class **City**

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- "Rangsit is a city in Pathum Thani Province, Thailand. Rangsit is the home of Rangsit University."
- Belongs to class **City**
- Labeled as: **Educational Institution**
- Problem: Short tetxts

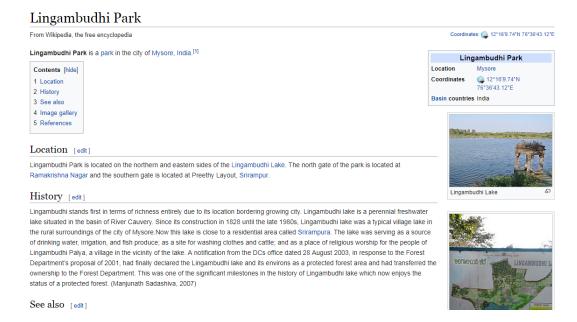
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• "Lingambudhi Park is a park in the city of Mysore, India."

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- "Lingambudhi Park is a park in the city of Mysore, India."
- Belongs to class Lake
- Labeled as: City

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• Problem: Wrong class in DBpedia

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## Topic Modeling

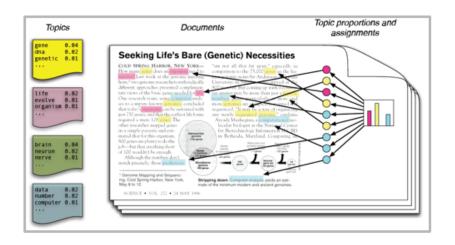


Figure 5: Topic Modeling<sup>6</sup>

https://www.analyticsvidhya.com/wp-content/uploads/2016/08/Modeling1.png

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<sup>&</sup>lt;sup>6</sup>Pic from

# Preprocessing

- 1. Remove punctuation
- 2. Remove Stop words
- 3. Lemmatize all words
- 4. Build dictionary
- 5. Build a matrix: For every abstract: Tupel (word id, word frequency)

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#### LDA - Latent Dirichlet Allocation

- Statistical model
- What topics could have created this text?
- Which words are the most probable for the topics?

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#### Topic Modeling - 1. Results

#### 5 topis, 5 words:

- school, high, university, college, student
- lake, river, located, water, area
- star, asteroid, galaxy, year, constellation
- loch, also, ontario, film, john
- chinese, hong, china, kong, quechua

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#### Topic Modeling - 1. Results

#### 5 topis, 5 words:

- school, high, university, college, student → Educational Institution
- lake, river, located, water, area
- star, asteroid, galaxy, year, constellation
- loch, also, ontario, film, john
- chinese, hong, china, kong, quechua

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## Topic Modeling - 1. Results

#### 5 topis, 5 words:

- $\bullet$  school, high, university, college, student  $\to$  Educational Institution
- lake, river, located, water, area  $\rightarrow$  Lake
- star, asteroid, galaxy, year, constellation
- loch, also, ontario, film, john
- chinese, hong, china, kong, quechua

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#### 5 topis, 5 words:

- ullet school, high, university, college, student  $\to$  Educational Institution
- lake, river, located, water, area  $\rightarrow$  Lake
- star, asteroid, galaxy, year, constellation  $\rightarrow$  Celestial Body
- loch, also, ontario, film, john
- chinese, hong, china, kong, quechua

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#### 5 topis, 5 words:

- $\bullet$  school, high, university, college, student  $\to$  Educational Institution
- lake, river, located, water, area  $\rightarrow$  Lake
- star, asteroid, galaxy, year, constellation  $\rightarrow$  Celestial Body
- loch, also, ontario, film, john  $\rightarrow$  Actor?
- chinese, hong, china, kong, quechua

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#### 5 topis, 5 words:

- $\bullet$  school, high, university, college, student  $\to$  Educational Institution
- lake, river, located, water, area  $\rightarrow$  Lake
- star, asteroid, galaxy, year, constellation  $\rightarrow$  Celestial Body
- loch, also, ontario, film, john  $\rightarrow$  Actor?
- chinese, hong, china, kong, quechua  $\rightarrow$  City??

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# Preprocessing

- Remove all parts of names as well
- $\bullet$   $\rightarrow$  stopwords = stopwords + "also" + all names

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#### 5 topics, 5 words:

- school, high, university, college, student
- river, located, area, reservoir, water
- asteroid, year, galaxy, constellation, approximately
- chinese, film, actor, actress, known
- province, county, norway, lough, district

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#### 5 topics, 5 words:

- school, high, university, college, student → Educational Institution
- river, located, area, reservoir, water
- asteroid, year, galaxy, constellation, approximately
- chinese, film, actor, actress, known
- province, county, norway, lough, district

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#### 5 topics, 5 words:

- $\bullet$  school, high, university, college, student  $\to$  Educational Institution
- river, located, area, reservoir, water  $\rightarrow$  Lake
- asteroid, year, galaxy, constellation, approximately
- chinese, film, actor, actress, known
- province, county, norway, lough, district

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#### 5 topics, 5 words:

- $\bullet$  school, high, university, college, student  $\to$  Educational Institution
- river, located, area, reservoir, water  $\rightarrow$  Lake
- ullet asteroid, year, galaxy, constellation, approximately ullet Celestial Body
- chinese, film, actor, actress, known
- province, county, norway, lough, district

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#### 5 topics, 5 words:

- school, high, university, college, student  $\rightarrow$  Educational Institution
- river, located, area, reservoir, water  $\rightarrow$  Lake
- $\bullet$  asteroid, year, galaxy, constellation, approximately  $\to$  Celestial Body
- chinese, film, actor, actress, known  $\rightarrow$  **Actor**
- province, county, norway, lough, district

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#### 5 topics, 5 words:

- school, high, university, college, student  $\rightarrow$  Educational Institution
- river, located, area, reservoir, water  $\rightarrow$  Lake
- $\bullet$  asteroid, year, galaxy, constellation, approximately  $\to$  Celestial Body
- chinese, film, actor, actress, known  $\rightarrow$  Actor
- province, county, norway, lough, district  $\rightarrow$  City

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#### Conclusion

- Supervised and unsupervised learning for wikipedia abstracts
- 5 classes  $\rightarrow$  How well can they be discriminated?
- Supervised learning: Problems with overlapping classes, short texts, false classes

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