Training Camp on "Knowledge Graph Completion"

— Sapienza University, M.Sc. Degree in Data Science —

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https://www.kaggle.com/c/unicredittrainingcamp/overview/lectures

June 30th - July 2nd, 2021

Day 3: Combining Rule Mining and Embedding Learning for Knowledge Graph Completion

Schedule

- Introduction to rule mining
 - Mining Rules from Knowledge Graph Data
 - Amie
- Employing rule-learning and embedding learning for Knowledge Graph Reasoning
 - Advantages and Limitations of Rule-Based learning
 - Advantages and Limitations of Embedding-Based learning
- Combining rules and embeddings for Knowledge Graph Reasoning
 - IterE: Iteratively Learning Embeddings and Rules for Knowledge Graph Reasoning
- Lab
 - Extracting rules from KG data with AMIE
 - Exploiting rules for knowledge graph completion
 - Combining rule mining and embedding learning

Expected outcome at the end of Day 3

Capability of:

- Extracting rules from knowledge graph data exploiting existing tools (AMIE)
- Instantiate rules and exploit them to produce new triples and/or new features for Knowledge Graph Completion
- Combining embedding learning with rule mining to achieve improved prediction performance

References

- [amie paper] AMIE: Association Rule Mining under Incomplete Evidence in Ontological Knowledge Bases
- [amie repo] https://github.com/lajus/amie
- [itere paper] Iteratively Learning Embeddings and Rules for Knowledge Graph Reasoning
- [itere repo] https://github.com/wencolani/IterE

Natural Language vs Knowledge Bases (KBs)

Natural Language

Shakira

From Wikipedia, the free encyclopedia

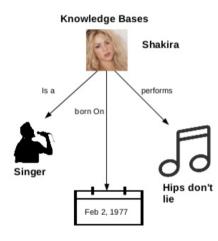
This article is about the musician. For her self-titled album, see Shakira (album) (disambiguation).

This name uses Spanish naming customs; the first or paternal family name is M family name is Ripoli.

Shakira Isabel Mebarak Ripoll (pronounce [[a'kira isa']lel meţla'rak ri'pol]; born February 2, 1977), [1] known professionally as Shakira

towin (Nixara/¹⁰ sweet [Sixtra]), is a Colorbian singer-songeriter, discerrecord producer, choreographer and model. Born and raised in Burranquilla, she began performing in school, demonstrating Latin, Arabic, and rock and roll influences and beily discrizing slattices. Shakir is released her first studio altums, Majar and Poliny, in the early 1900s, failing to statin commercial success, however, the rose to prominence in Latin America with her majar-label debut, Pas Democales (1906), and her forunt abum, Olivide Edini sits Ladinovair (1998).

Stakins entered the English Jangsage market with her fifth allows, Laurstry, Service 2001, Mech has said over 2 million opinis with stakes. ³⁰ It is load single, "Whenever, Wherever", became the best-selling single of 2002, Ner success was solidard with her sish and ensewth abserts [Escrish Cred. 16.1. 2 million 10° La Francisco, 16.1. 2 (2005), the latter of which spawmed the heat-selling song life 10° La Francisco, 16.1. 2 (2005), the latter of which spawmed the heat-selling song life (2009) and Sale of Sel (2011), received critical praise but suffered from limited (2009) and Sale of Sel (2011), received critical praise but suffered from limited (2009) and Sale of Sel (2011), received critical praise but suffered from limited (2009) and Sale of Sel (2011), "Sel (2011), "Sel



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Shakira Isabel M born February

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Some popular KBs















Social graphs are KBs

They both share

- A natural graph-like structure
- Incompleteness
- Opportunities for data description and prediction
- E.g., If you like Shakira, you are likely to buy her latest song



Rule Mining and KBs

- Data mining is about finding interesting and non-obvious correlations in the data
- Correlations may be seen as rules that hold often
 - You probably live in the same city of your spouse
 - If you like an artist, you probably like her songs
- Correlations can be represented as logical rules:
 - $isMarriedTo(x, y) \land livesIn(x, city) \Rightarrow livesIn(y, city)$
 - $likes(x, artist) \land performs(artist, song) \Rightarrow likes(x, song)$
- Rules allow to exploit social data for real-world applications (e.g., product recommendation)

Applications for social data

- Market basket analysis
 - People who buy laptops also buy laptop cases
- Link and Event Prediction
 - Two people who attended the same high school the same year might know each other
 - If you registered for a conference in Rome, then you are coming to Rome (and you need to book a flight and a hotel)
- Dealing with Incompleteneess
 - If you like German newspapers, fluency in German is probably missing in your profile

AMIE: Association Rule Mining under Incomplete Evidence

- AMIE is a system that learns **Horn rules** such as:
 - $livesIn(x, city) \land isMarriedTo(x, y) \Rightarrow livesIn(y, city)$
- Starting with all possible head relations r(x, y) and a minimum support threshold:
 - The system explores the search space by means of carefully designed mining operators
 - Search space is restricted to closed Horn rules
 - Head coverage is used for pruning: we are not interested in rules that cover only very few facts of the head relation
 - E.g., Rules that cover, for example, less than 1% of the facts of the head relation can safely assumed to be marginal
 - Head coverage decreases monotonically as we add more atoms. This allows us to safely discard any rule that trespasses the threshold
 - If a rule $B_1 \wedge B_2 \wedge \cdots \wedge B_n \wedge B_{n+1} \Rightarrow H$ does not have larger confidence than the rule $B_1 \wedge B_2 \wedge \cdots \wedge B_n \Rightarrow H$, then we do not output the longer rule: both confidence and head coverage of the longer rule are necessarily dominated by the shorter rule.

Challenges of rule mining on KBs

- Incompleteness: graph data often contains gaps
- Open World Assumption (OWA):
 absence of evidence is not evidence of absence
- Problems to estimate the **confidence** of a rule

likes(x, Shakira) => isCitizenOf(x, Ecuador)

Standard confidence uses a CWA and counts Shamira as counterexample. Score = 0.5



Challenges of rule mining on KBs

- AMIE uses the Partial Completeness Assumption (PCA) to estimate the confidence of rules under OWA
- A KB knows all or none of the nationalities of a person

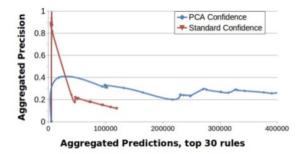
likes(x, Shakira) => isCitizenOf(x, Ecuador)

PCA confidence considers as counterexamples only those people whose nationality is known to be different from Ecuador. Score = 1.0



AMIE: Predictive Behavior

PCA confidence has better predictive behaviour than standard confidence



Examples of rules mined by AMIE on YAGO

```
\label{eq:linear_continuous_continuous} \hline \text{isMarriedTo}(x, y) \land \text{livesIn}(x, z) => \text{livesIn}(y, z) \\ \text{isCitizenOf}(x, y) => \text{livesIn}(x, y) \\ \text{hasAdvisor}(x, y) \land \text{graduatedFrom}(x, z) => \text{worksAt}(y, z) \\ \text{hasWonPrize}(x, \text{Gottfried Wilhelm Leibniz Prize}) => \text{livesIn}(x, \text{Germany}) \\ \hline \\
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Knowledge Graph Reasoning (KGR)

- Knowledge graph reasoning (KGR) can infer new knowledge based on existing ones and check knowledge consistency
- Applications: Knowledge graph cleaning and completion
- Two main learning methods for KGR:
- Embedding-based reasoning
 - learns latent representations of entities and relations in continuous vector spaces, called embeddings, so as to preserve the information and semantics in KGs
 - more efficient when there are a large number of relations or triples to reason over
- Rule-based reasoning
 - aims to learn deductive and interpretable inference rules
 - precise and can provide insights for inference results

Limitations of learning methods for KGR

- Sparsity Problem for Embedding Learning
 - Poor capability of encoding sparse entities (those with only a few triples)
 - Prediction results of entities are highly related to their frequency
- Efficiency Problem for Rule Learning
 - Search space exponential to the number of relations

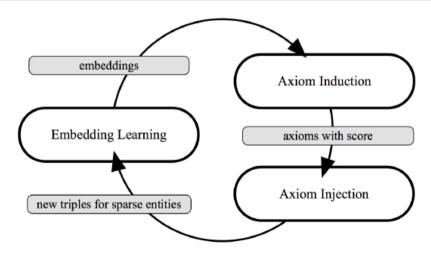
Combining embedding learning and rule learning

- With different advantages and difficulties, embedding learning and rule learning can benefit and complement each other
- Deductive rules can infer additional triples for sparse entities and help embedding learning methods encode them better.
- Embeddings encoded with rich semantics can turn rule learning from discrete graph search into vector space calculation, so that reduce the search space significantly

IterE: Iteratively Learning Embeddings and Rules for Knowledge Graph Reasoning Zhang et al., 2019

- A framework that iteratively learns embeddings and rules
- Can combine different embedding methods and kinds of rules
- 3 main parts:
 - Embedding learning: learns embeddings for entities and relations, with input including triples existing in KG and those inferred by rules.
 - Axiom induction: generates a pool of possible axioms with an effective pruning strategy, then assigns a score to each axiom based on calculation between relation embeddings
 - Axiom injection: utilizes axioms' deductive capability to infer new triples for sparse entities to be injected into KG
- The three parts are conducted iteratively during training.

IterE: Iteratively Learning Embeddings and Rules for Knowledge Graph Reasoning Zhang et al., 2019



How to exploit rule mining for knowledge graph completion?

- Mine rules on our training set
- Instantiate rules: either all of them, or setting a threshold on PCA confidence
- Derive new KG triples (with a confidence label)
- Use instantiated rules for direct prediction: given a test triple, predict it as true
 - if I can find it as output of a rule
 - if I can find it as output of a rule with confidence above a threshold
 - if I can find it as output of many rules
- Use instantiated rules to **add new features to the classifier** that predicts whether a triple is *true* or *false*: e.g.,
 - Number of rules which a triple is involved in
 - Min, avg, max confidence of the rules which a triple is involved in
- Use rules to improve the training of triple embeddings:
 - Instantiate all rules (with a min confidence) and add new triples to the training set
 - Retrain embeddings and check if a better model can be learnt (Once, or iteratively)
 - Try of the existing models that iteratively combine rule mining and embedding learning: [itere] Iteratively Learning Embeddings and Rules for Knowledge Graph Reasoning