

Introductory Presentation

"Graph Kernels for RDF Data"* - Summary and implementation ideas

Dennis Oliver Kubitza

5. Dezember 2017

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- * **Graph kernels for RDF data.**

Lösch, U., Bloehdorn, S., & Rettinger, A. (2012).

The Semantic Web: Research and Applications, 134-148

[10.1007/978-3-642-30284-8_16](https://doi.org/10.1007/978-3-642-30284-8_16)

2012

Content Overview

- 1 RDF-Kernels
 - Motivation
 - Kernels for RDF
 - Relevance

- 2 Implementation

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Combining RDF with Machine learning

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We want to apply Machine learning algorithms on RDF.

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We want to apply Machine learning algorithms on RDF.

- Resource Description Frameworks (RDF) impose a Graph structure on the Data
- Machine learning algorithms are optimized on certain Data Structures (Tables etc.)

Kernel based machine learning

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Definition

Let D be the Space of Data, $\psi : D \rightarrow F$ a mapping representing the Data as $F \subset \mathbb{R}^n$. A function with a representation

$$K : D \times D \rightarrow \mathbb{R}$$

$$K(d_1, d_2) = \langle \psi(d_1), \psi(d_2) \rangle$$

is called Kernel.

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Kernels for RDF

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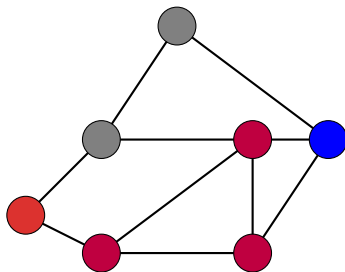
$$\blacksquare D = \{G', G' \subset G\}$$

What Kernels exists on can we define for Graphs ?

- 1 Walk Kernel (of maximum path length l)
- 2 Path Kernel (of maximum path length l)
- 3 Full Subtree Kernel (of depth d)
- 4 Partial Subtree Kernel (of depth d)

These are the examples provided by the authors. Of course infinite many definitions can be made.

Example



$$\kappa_{l,\lambda} = \sum_{i=1}^l |\{p | p \in paths_i(G_1 \cap G_2)\}|$$

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Example

Why this approach ?

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Lösch U., Bloehdorn, S., Rettinger, A. showed that this approach of applying Graph Kernels to RDF show almost the same performance as retailoring ML algorithms to RDF structure.

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Basic Idea

- Build an Interface for the mangement of kernel information / values.

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 - Give the possibility to use User defined Update/Deletation algorithms and kernels.
 - Provide the defined example Kernels with efficient implementation of Update/Deletation

Computation of Predefined Kernels

- Use Sansa's Graph Querying for Calculation of Trees, Paths.

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Computation of Predefined Kernels

- Use Sansa's Graph Querying for Calculation of Trees, Paths.
- + Faster than Sparks RDD querying.
- + Computational Performance will scale up with SANSA's performance.

Value Storage

Two ideas:

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 - + All information is provided in Consistent Format.
 - + Induces a Distance Graph on Original RDF.
 - + Can be used for further analysis outside of Sansa

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- 2 Store Kernel-Values in seperate Structure.
 - + No Changes on original RDF.
 - + Only relevant Information for ML/DM.

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Two ideas:

1 Store Kernel-Values directly in original RDF

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2 Store Kernel-Values in separate Structure.

- + No Changes on original RDF.
- + Only relevant Information for ML/DM.
 - What size ? (Has to be Scalable, e.g. updates)
 - What structure ? (f.e. Distance Matrix: Inefficient if Sparse)
 - Has to be a known format ! Otherwise not user friendly.

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- Offer both possibilities:
 - 1 Store Kernels in original RDF
 - 2 Store Kernels in separate reduced RDF (Original RDF without any Predicates and Attributes, except Kernel Oriented information)
- + Same implementation for both variants. Only different target for saving the values
- + User can pick which one he favours.
- + In each way the user is familiar with the storage format.
- + A posteriori the RDFs can still be merged.