

# Introductory Presentation

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

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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

- General Concept
- Ideas for Implementation

# Content Overview

## 1 RDF-Kernels

### ■ Motivation

### ■ Kernels for RDF

### ■ Relevance

# Combining RDF with Machine learning

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We want to apply Machine learning algorithms on RDF.

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- Machine learning algorithms are optimized on certain Data Structures (Tables etc.)

## Problem!



# Combining RDF with Machine learning

## Goal

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.)

## Problem!

## Solutions:

- 1 Rewrite the ML-Algorithms for RDF Data
- 2 Try to transform the RDF Data that it suits ML-Algorithms

# Kernels mathematically

## 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  $K$  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.

Short version:

- 1 We take our Data  $d_1, d_2$
- 2 We transfer them to real-valued vectors  $\psi(d_1), \psi(d_2)$
- 3 We apply a scalar product.

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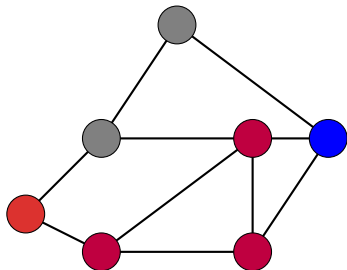
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What Kernels can we define for Graphs? Lösch et al. proposed for different suitable kernels.

- 1 Walk Kernel
- 2 Path Kernel
- 3 Full Subtree Kernel
- 4 Partial Subtree Kernel

# Example: The path kernel



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



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# Example

Why this approach ?

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



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# Computation of Predefined Kernels

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

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

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  - + All information is provided in Consistent Format.
  - + Induces a Distance Graph on Original RDF.
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- 2** Store Kernel-Values in separate Structure.
  - + No Changes on original RDF.
  - + Only relevant Information for ML/DM.



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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.