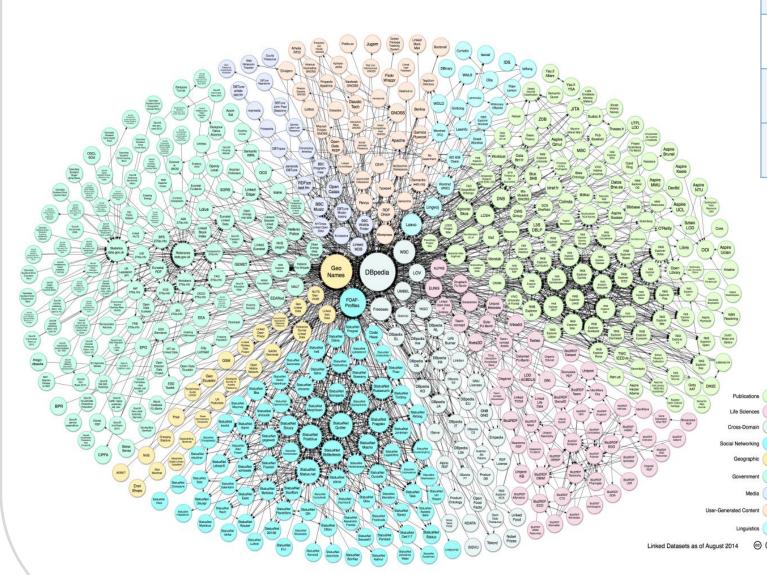


# WebER: Resolving Entities in the Web



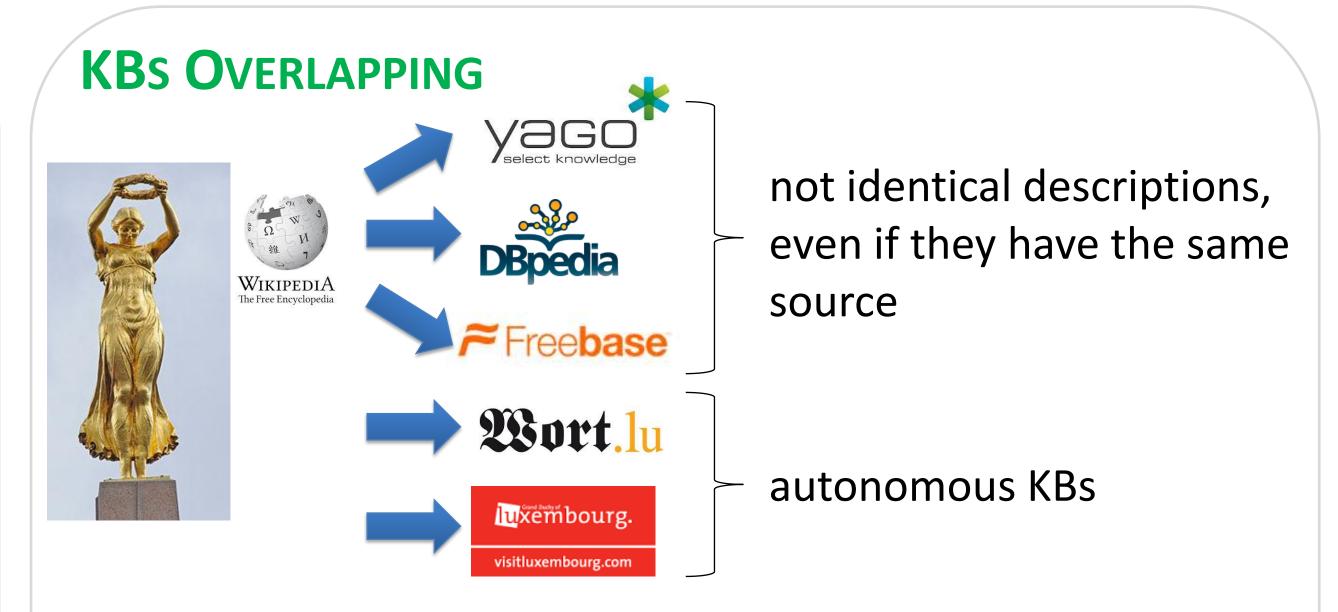
### THE LOD CLOUD



KB examples	entities	entity types	properties
DBpedia(en)	4.58M	685	2,795
Freebase	46.3M	1.5K	4K
YAGO2	10M	350K	100
Uniprot	1.9B	123	112

Main characteristics of LOD: Important number of KBs (~ hundreds). Large number of entity types & properties (~ thousands).

Massive volume of entities (~billions).

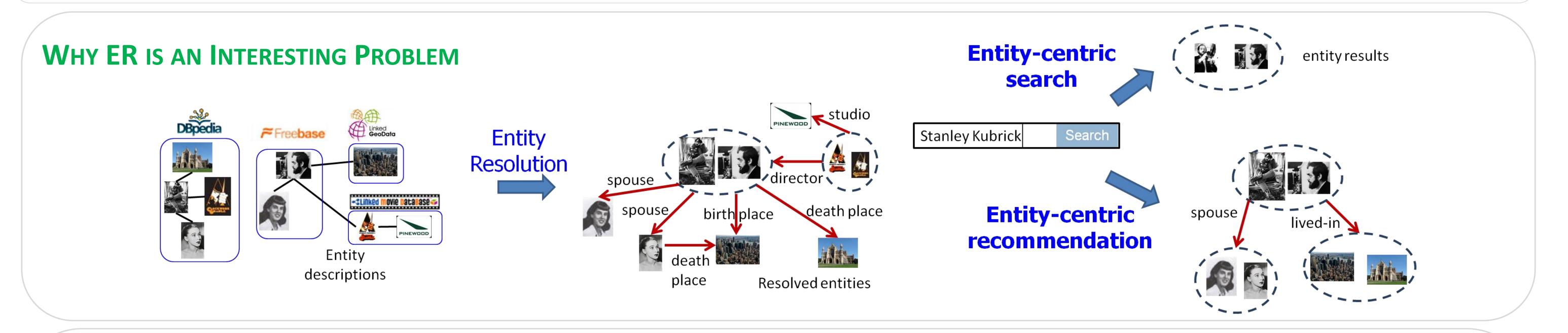


Highly and somehow similar descriptions of the same entity. ▶ 3-4 common tokens in central KBs.

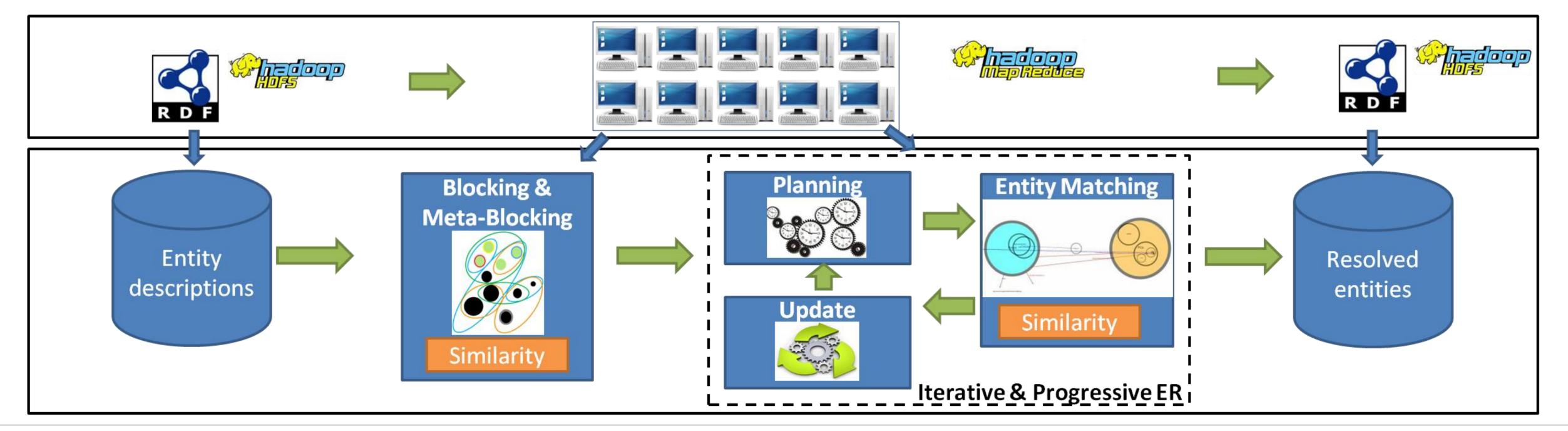
▶ 1-2 common tokens in peripheral KBs.

#### **ENTITY RESOLUTION (ER):** The problem of identifying descriptions of the same real-world entity

**WEB-SCALE ER:** Large-scale, multi-type, and cross-domain ER



THE WORKFLOW OF ER



#### BLOCKING

Avoid the quadratic complexity of ER:

Split descriptions into blocks.

Suggest the comparison of descriptions placed in a common block.

**Examples**: token blocking, PPJoin+, and Locality-sensitive Hashing (LSH). [Christophides et al., 2015, Efthymiou et al., IEEE BigData 2015a]

### **META-BLOCKING**

Blocking post-processing:

Prune the redundant comparisons generated by blocking. Prune the least promising comparisons generated by blocking. [Efthymiou et al., IEEE BigData 2015b]

## **ITERATIVE & PROGRESSIVE ER**

**Iterative ER**: identify new matches based on previous results and similarity propagation.

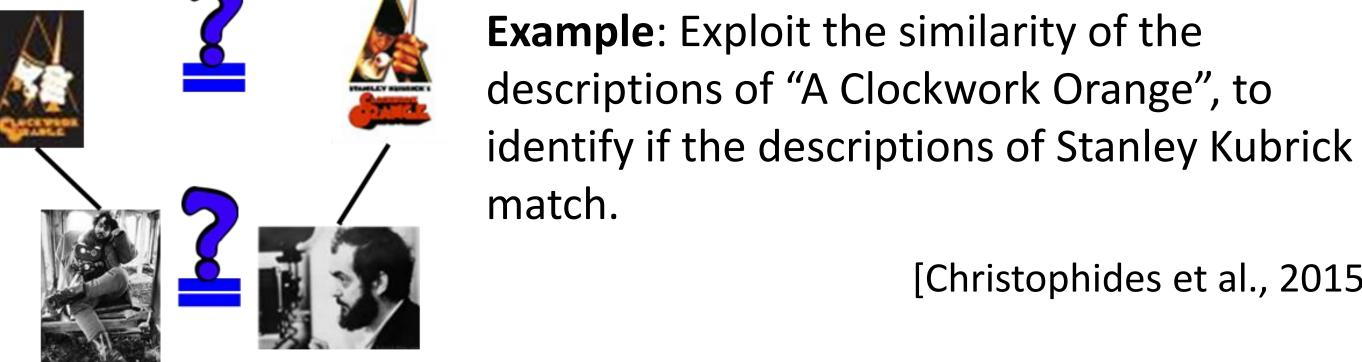
**Progressive ER**  $\rightarrow$  **Optimization**: maximize benefit (number of matches | type of matches) for a given cost (#comparisons | disk access). Planning: select which pairs will be compared next and in what order. Update: propagate the results of matching, such that a new planning phase will promote the comparison of pairs influenced by the previous matches.

Exploit the parallel processing power of a computer cluster via Hadoop MapReduce



#### SIMILARITY

There is no single ideal similarity measure to identify all matches. Indications for matches can be provided by similarity in: Content, i.e., the values of the descriptions (e.g., Jaccard) Structure, i.e., the attributes of the descriptions (e.g., SimRank) Neighborhoods, i.e., related descriptions (e.g., LINDA)



[Christophides et al., 2015]

More details, source code and datasets available at: <u>http://www.csd.uoc.gr/~vefthym/minoanER/</u>

V. Christophides, V. Efthymiou, K. Stefanidis. Entity Resolution in the Web of Data. Synthesis Lectures on the Semantic Web: Theory and Technology, Morgan & Claypool, August 2015 V. Efthymiou, K. Stefanidis, V. Christophides. Big Data Entity Resolution: From Highly to Somehow Similar Entity Descriptions in the Web. In IEEE Big Data, 2015a V. Efthymiou, G. Papadakis, G. Papastefanatos, K. Stefanidis, T. Palpanas. Parallel Meta-blocking: Realizing Scalable Entity Resolution over Large, Heterogeneous Data. In IEEE Big Data, 2015b

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