

Efficient Semantic-Aware Detection of Near Duplicate Resources ^{7th} Extended Semantic Web Conference

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Outline

I. Motivation

2. RDFsim Approach

Resource representation

- Indexing structure
- Querying for near duplicate resources
- **Experimental Evaluation** 3.
- Conclusions 4.



Motivation

- Plethora of current Semantic and Social Web applications that integrate data from various sources
- BUT data is overlapping or complementary
- Detect near duplicate data:
 - group, merge, remove resources
 - avoid repetition and redundancy



Motivation - News Aggregation Service

- Aggregate articles from a large number of news agencies
- Republish same articles, include slight changes, spelling mistakes, an additional image, or some new information
- RDF data from extractors \rightarrow entities, relationships, ...

Intel upgrades Atom chip platform

Published: Dec. 21, 2009 at 3:52 PM

SANTA CLARA, Calif., Dec. 21 (UPI) -- U.S. microchip maker Intel said Monday its next generation Atom chip platform would make its debut in netbooks and laptops in January 2010.

The latest improvements create a platform with increased energy efficiency with "integrated graphics capabilities and an on-board memory controller," eWeek reported Monday.

The Atom chip has been an integral component in netbooks, which have ...



Netbooks to get smaller, faster and cheaper

8:16 AM Tuesday Dec 22, 2009

Intel plans to shrink netbooks even further with its latest range of Atom processors, which feature built in graphics as well as a smaller, more energy efficient design.

Previously codenamed Pine Trail, the new Atom processor is primarily designed for use in netbooks and entry-level desktop PCs. It is now officially Intel's smallest chip.



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Intel's new Atom release can expect a new crop of efficient and cheaper net months.



Detecting Near Duplicate RDF Resources: \rightarrow compute similarity and select based on requirements

Two main issues:

- How to compute the similarity between a pair of a) **RDF** Resources ?
- How to efficiently compare resources ? b)
 - Avoid pairwise comparisons
 - Allow on-the-fly operation



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RDFsim Approach

- Each resource R is an RDF graph Set of RDF triples
- \mathcal{R} is the set of all available resources
- Function computing similarity sim: $\mathcal{R} \times \mathcal{R} \rightarrow [0, 1]$

$R_1 \& R_2$ are near duplicates: $sim(R_1, R_2) \ge minSim$

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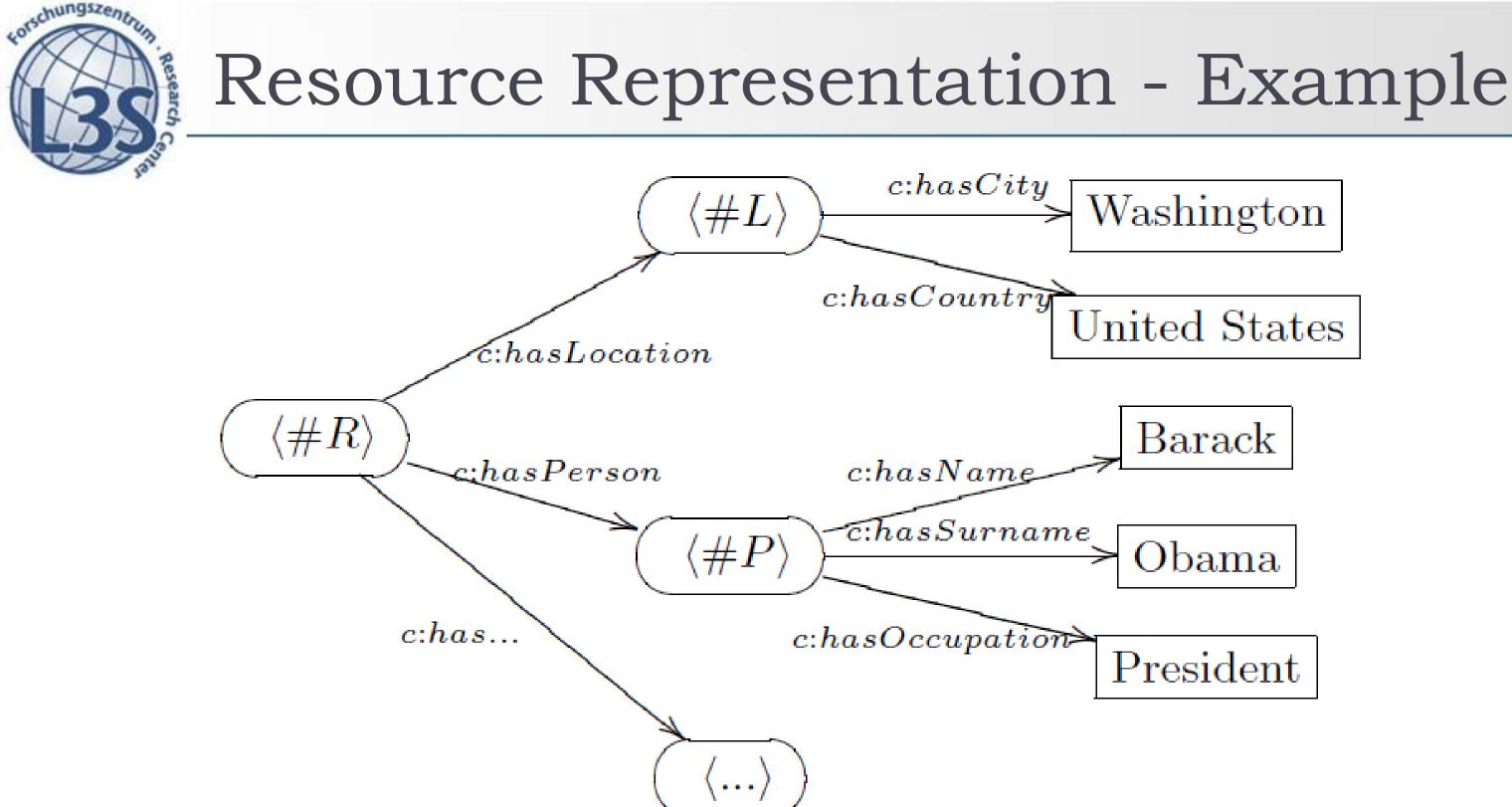


Representation is denoted with rep(R)

RDFsim applies a transformation of the RDF graph:

- for each triple \rightarrow concatenate predicate with object
- if object is another RDF triple Ry \rightarrow union with *rep*(Ry)





rep(L) = { "c:hasCity, Washington", "c:hasCountry, United States" } rep(P) = { "c:hasName, Barack", "c:hasSurname, Obama", "c:hasOccupation, President" }

 $rep(R) = \{$ "c:hasLocation, L", "c:hasPerson, P", ... $\} \cup rep(L) \cup rep(P)$

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- Based on the Locality Sensitive Hashing (LSH)
- Indexing structure \mathcal{I} that consists of l binary trees: $\bullet \quad \mathcal{T}_1, \mathcal{T}_2, \ldots, \mathcal{T}_l$
- Each tree is bound to k hash function: $\mathcal{T}_i \rightarrow h_{1,i}, h_{2,i}, \dots, h_{k,i}$



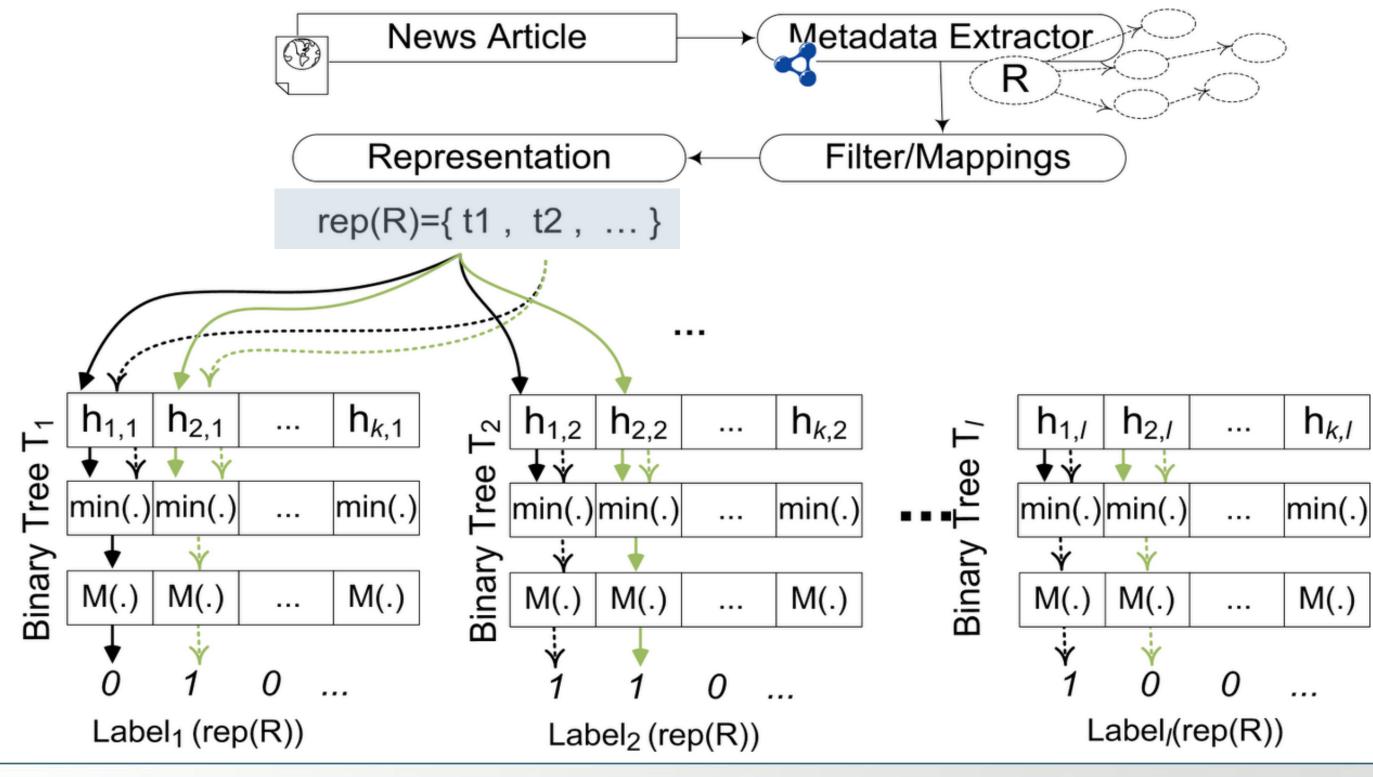
- A. Extract rep(Rx)
- Compute l labels of length k for each binary tree Β.
 - B.I. Hash all terms in rep(Rx) using each hash function $h_{i,j}(.)$
 - B.2. Detect the minimum hash value produced by $h_{i,j}(.)$
 - B.3. Map min(hi, j(.)) to a bit
 - B.4. Use result as the *i*'th bit of the label of rep(Rx)
- C. Insert labels in the trees

example in next slides ...



Adding new resource

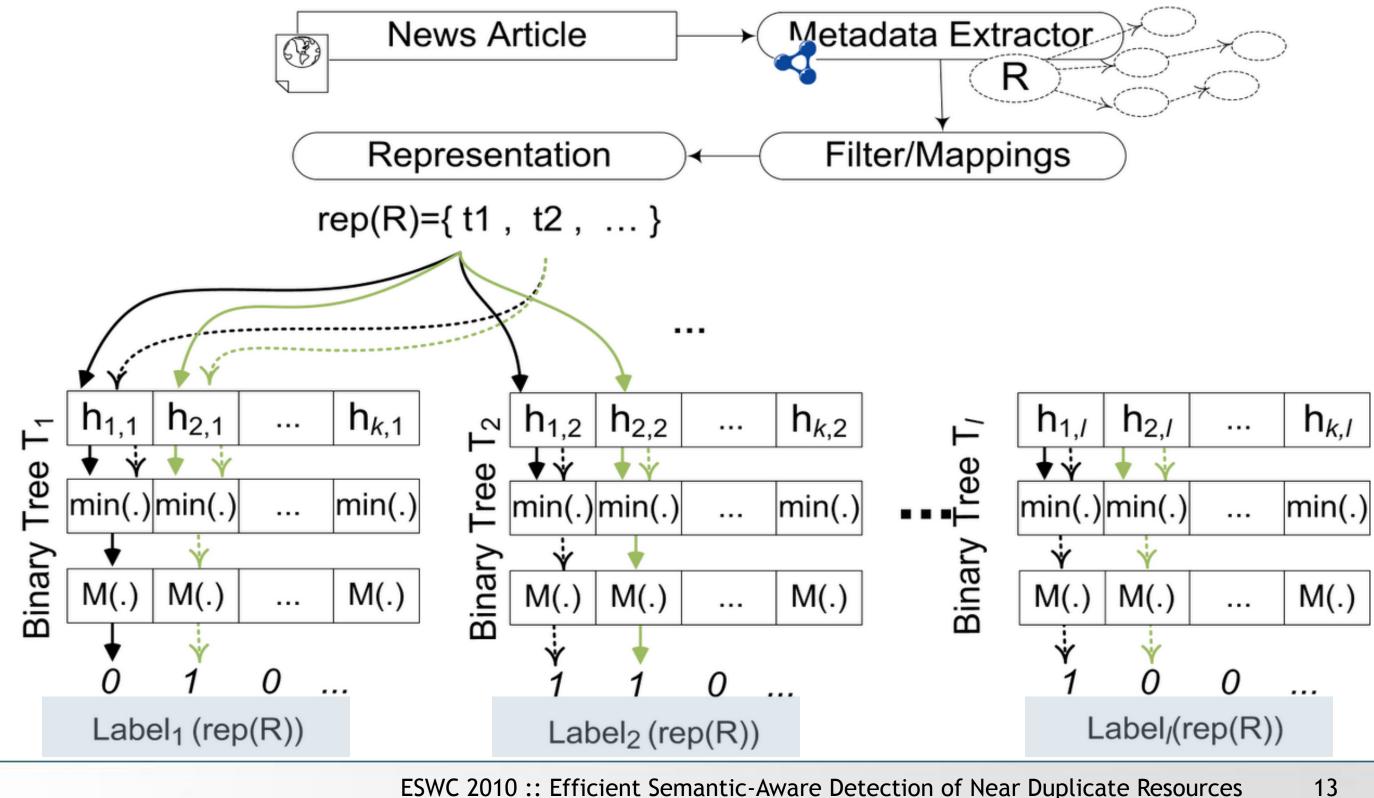
A. Extract rep(Rx)





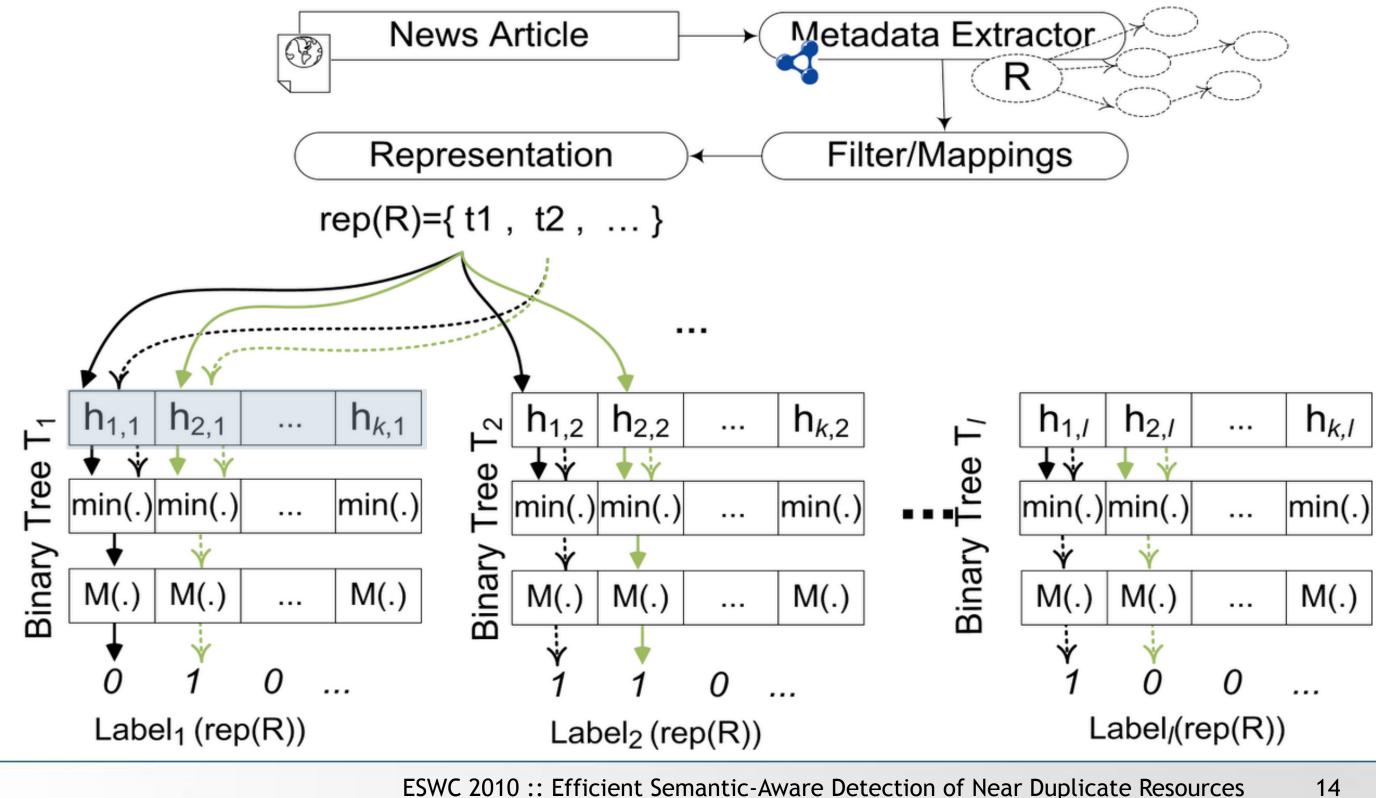
Adding new resource

B. Compute l labels of length k for each binary tree



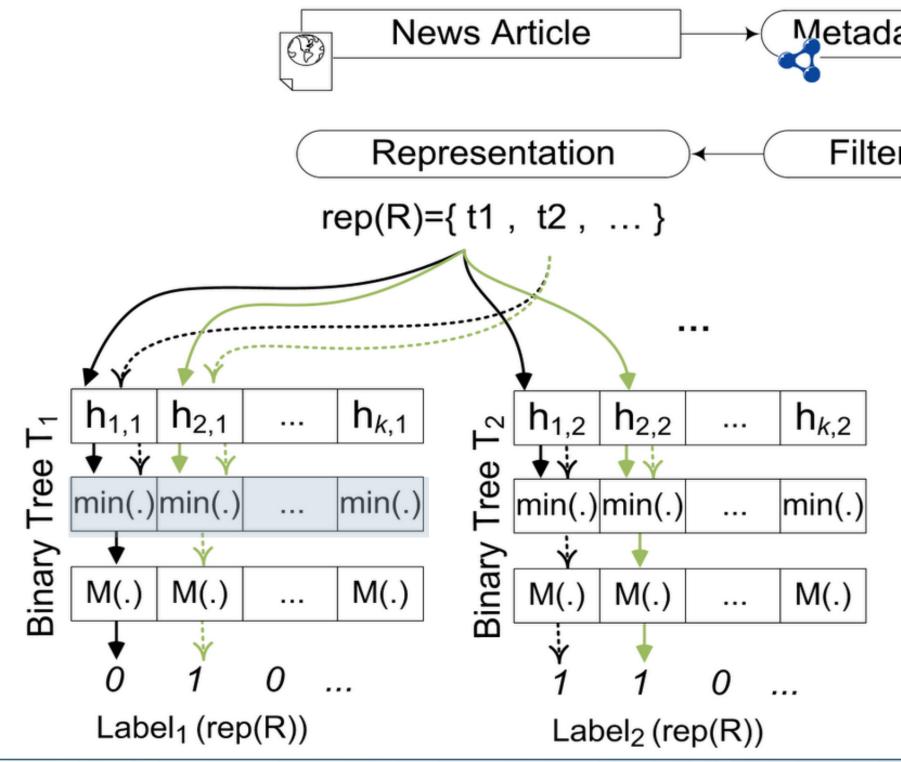


B.I. Hash all terms in rep(Rx) using each hash function $h_{i,j}(.)$

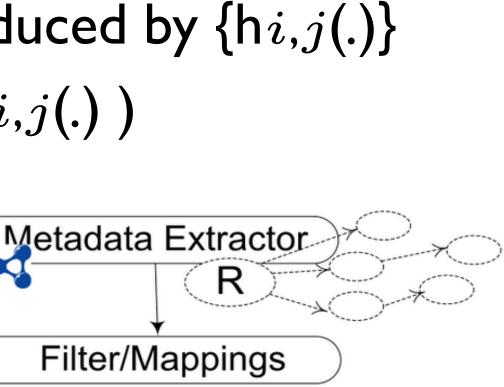


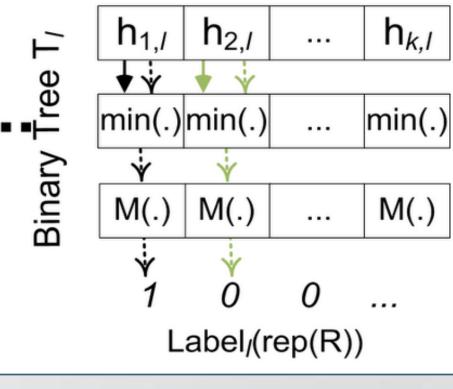


B.2. Detect the minimum hash value produced by $\{h_{i,j}(.)\}$ for all i=1...k, $j=1...l \rightarrow min(h_{i,j}(.))$



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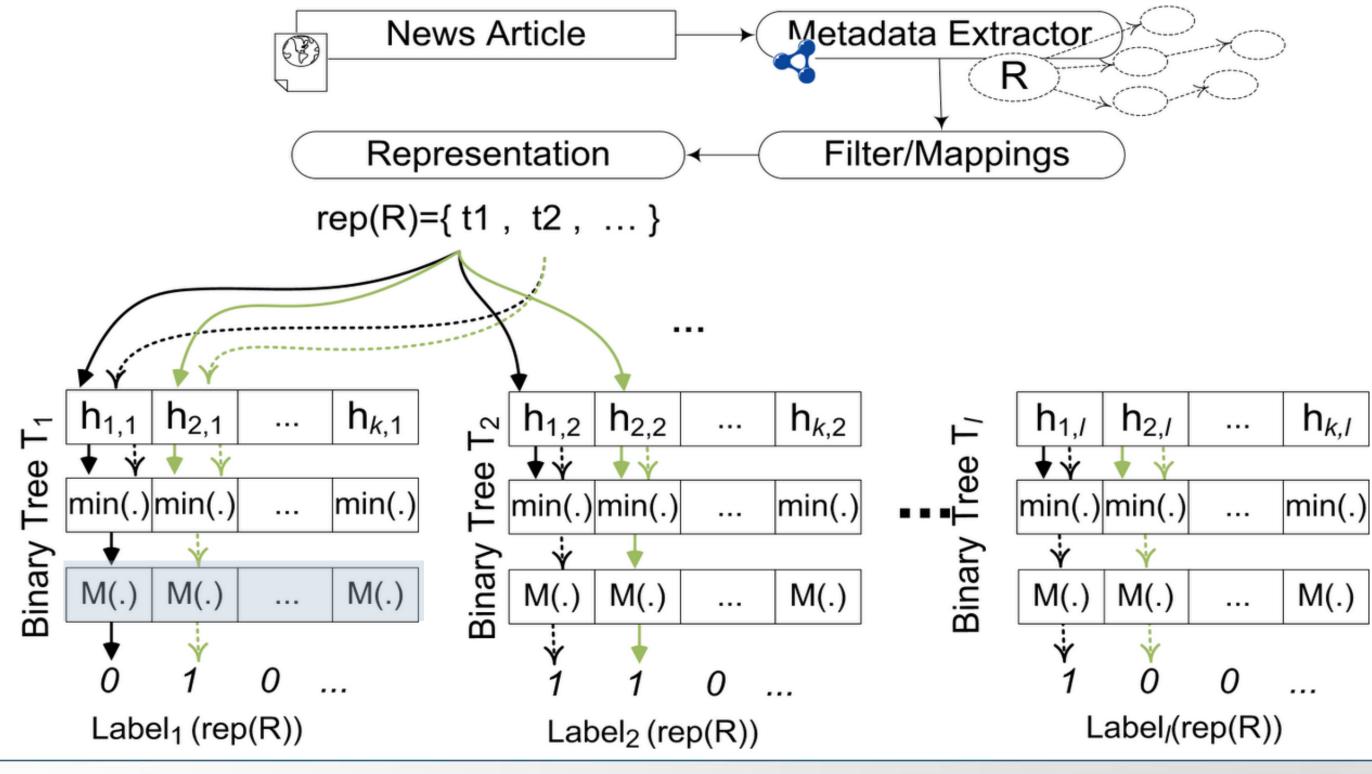




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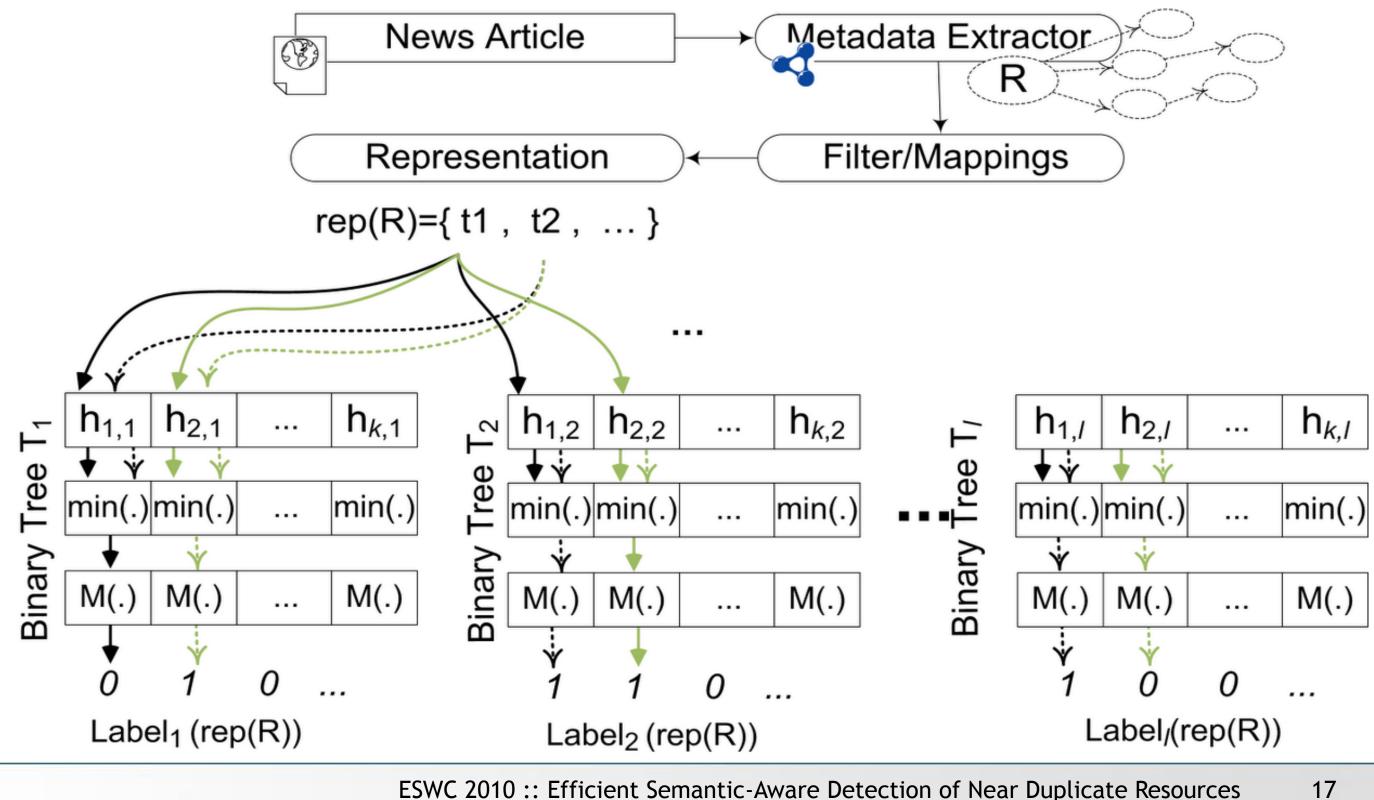
B.3. Map min(hi, j(.)) to a bit 0 or 1

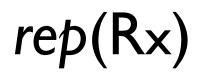




Adding new resource

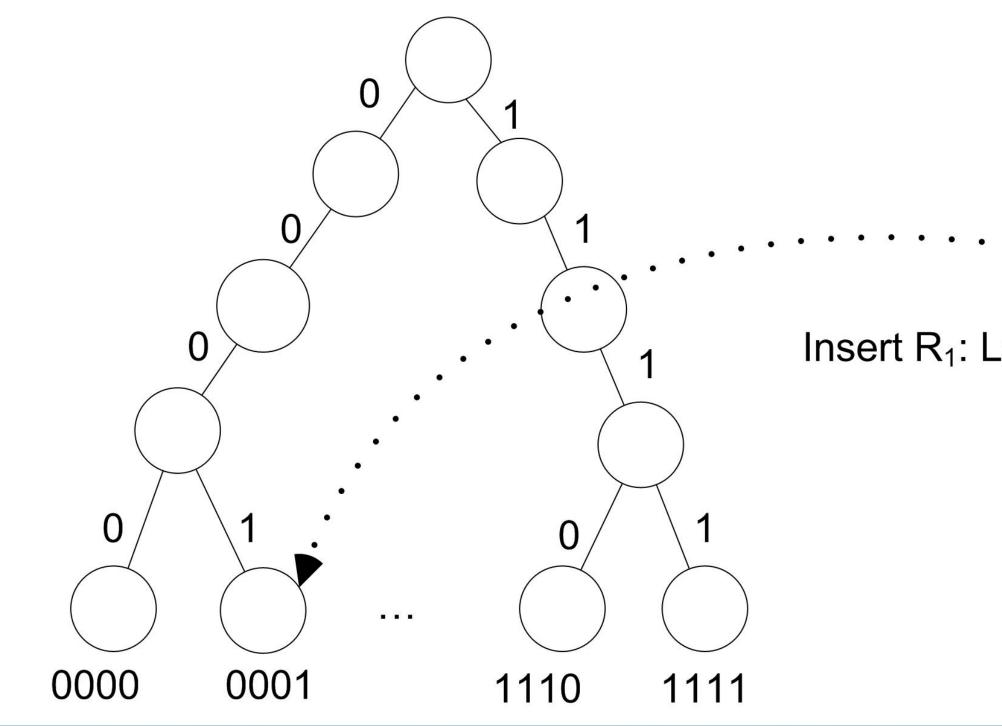
B.4. Use result as the i'th bit of the label of rep(Rx)







C. Insert labels in the trees i.e., Label $i(rep(Rx)) \rightarrow binary$ label for Ti



ESWC 2010 :: Efficient Semantic-Aware Detection of Near Duplicate Resources

Insert R_1 : Label(rep(R_1)) = 0001



Querying for near duplicate resources

- Create the labels for each tree $\mathcal{T}_1, \mathcal{T}_2, \ldots, \mathcal{T}_l$
- Similar resources are indexed at nearby nodes in the tree with high probability

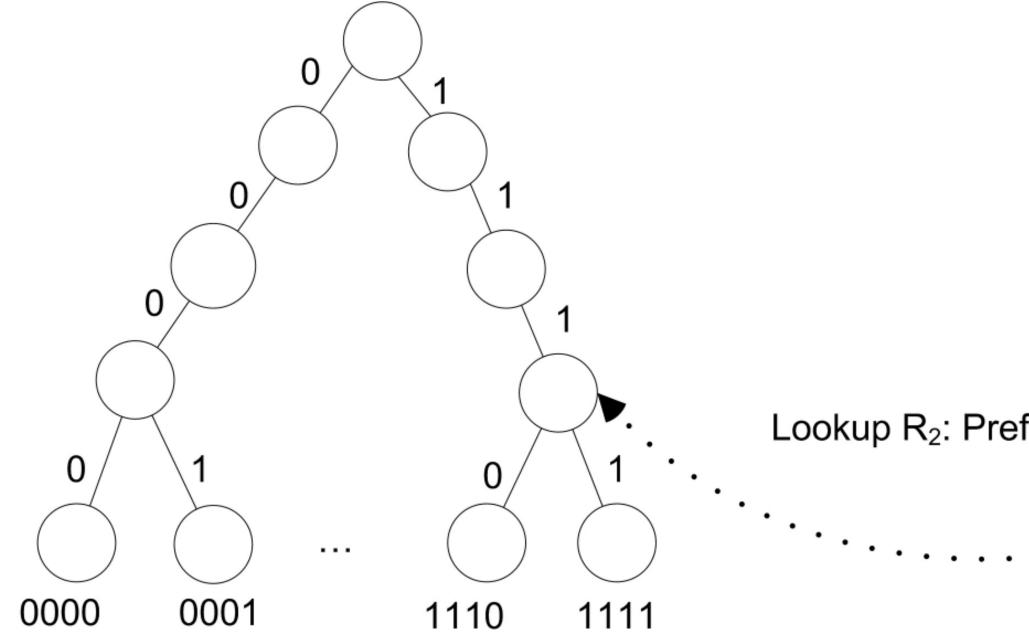
 \rightarrow selection criterion can be relaxed

i.e., prexfix lookup with length k'

- We set k' that allows detection with probability equal or higher to the requested minProb (see paper)
- We retrieve from each tree the resources
- Return the union



Example: retrieve resource from tree



Lookup R₂: Prefix(Label(rep(R_2))) = 111



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Dataset (available online) :

- Crawled news articles from the Google News Web site (e.g., BBC, Reuters, and CNN)
- RDF statements using the Open Calais Web service
- 94.829 news articles with 2.711.217 entities (RDF data)

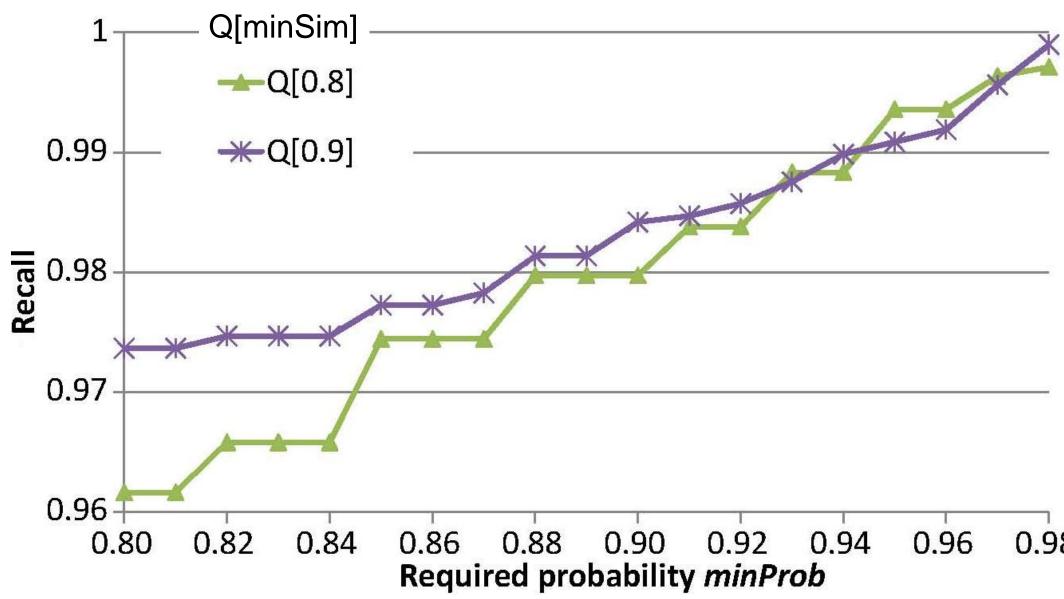
Methodology:

- Detect near duplicate for each articles
- Different required probabilistic guarantees, i.e., minProb
- Two approaches:
 - Searching using the RDFsim approach
 - Detecting near duplicates with pairwise comparison



Probabilistic guarantees vs. recall:

- Recall increases with the required minProb
- Recall is always higher than the value of *minProb* (verifies that the probabilistic guarantees are satisfied)

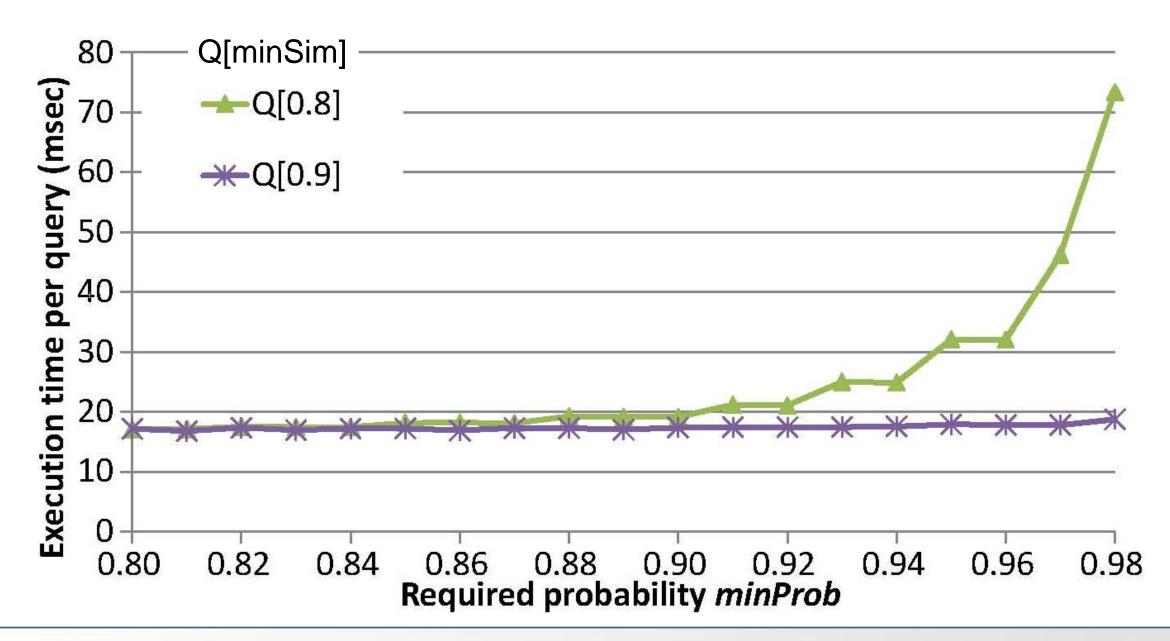


0.98



Probabilistic guarantees vs. average query execution time:

- Small avg execution time for all configurations
- Time increases as the requested minProb increases





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Conclusions

- Efficiently detect near duplicate resources on the Semantic Web
- Utilize the RDF representations of resources
- Consider semantics and structure of descriptions