

# Pairwise Webpage Coreference Classification using Distant Supervision



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#### Introduction

# A person or other entity is often associated with multiple URL endpoints on the web

- ▶ Barack Obama ⇔ https://barackobama.com/ and https://en.wikipedia.org/wiki/Barack\_Obama
- ► Donald Trump https://twitter.com/realDonaldTrump and https://www.instagram.com/ivankatrump

Motivates the task of webpage coreference classification for a given entity!

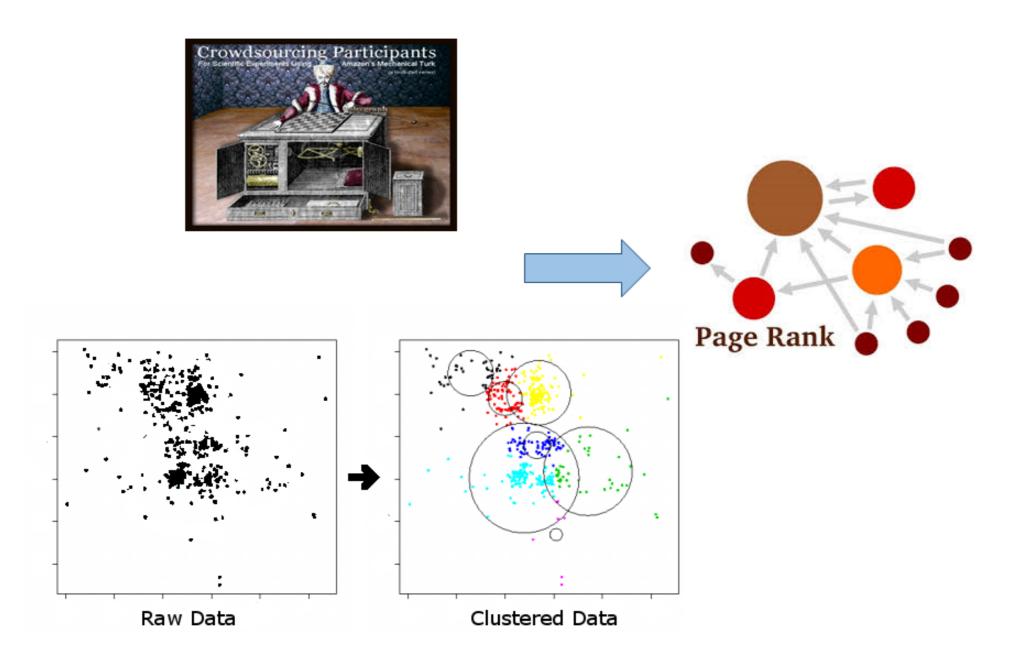
#### **Problem Setup**

- ► We assume access to web KBs based on automatic crawling
- > social networks (e.g. facebook.com/\*), news aggregation endpoints (e.g.
  nytimes.com/topic/person/\*) and organisation directories (e.g.
  www.gtlaw.com/People/\*)
- ► Given a training dataset *D* with pairs of web URLs
  - ▶ Initially all the pairs are unlabeled  $(D_u \leftarrow D, D_p, D_n = \phi)$
  - Learn a model  $f(\phi(U_i, U_i)) \rightarrow y$ , for URL pair  $U_i$  and  $U_i$
  - ► Target  $y \in \{0, 1\}$

## **Distant Supervision**

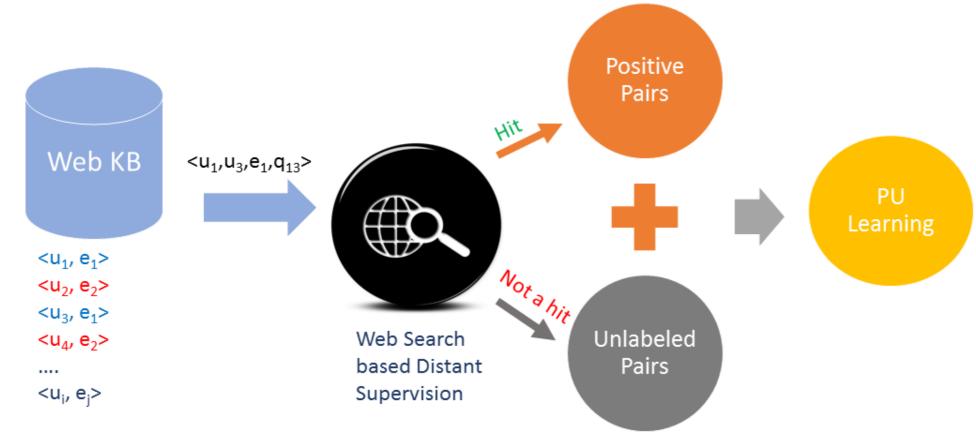
To strike a balance between unsupervised and supervised methods that require annotated data

- ► We obtain positive examples using web-search-based distant supervision
- ► Search query George Clinton AND P-Funk fetches https://en.wikipedia.org/wiki/George\_Clinton\_(musician)
- ► But not http://www.biography.com/people/george-clinton-537674
- ► We build a positive and unlabelled (PU) learning model



## **Proposed Approach**

- ► We generate queries for URL pairs that share same entity name
- ► Employ a label propagation technique to expand the set of positive examples



► Any binary classifier can be trained on the expanded labeled set

#### **Query Generation**

- ► We construct web search queries for distant supervision as follows:
- $\triangleright Q_i$ : Using the target entity name and context information from  $U_i$
- $\triangleright Q_j$ : Similar to the above, we generate context information from  $U_j$ .
- ► E.g., for URL pairs: www.imperial.ac.uk/people/f.allen and https://www.linkedin.com/in/franklin-allen-0557906 a query constructed is "Franklin Allen Brevan Howard Centre"

#### **Initialize Labels**

- For each query in  $Q_i$ , we check to see if  $U_j$  is present in the top-K search results  $S_{ij}$
- ightharpoonup Conversely if  $U_i$  is present in the top-K results,  $S_{ji}$  for each query in  $Q_j$ 

  - $ightharpoonup D_p \leftarrow D[\hat{y_{ij}}=1], \ D_u \leftarrow D_u \setminus D[\hat{y_{ij}}=1].$

# **Labelling Unlabelled Pairs**

- ▶ Step 1: Randomly select N instances from  $D_p$ , and hold them out in  $S_p$ .
- Step 2: Train a binary classifier  $\theta$ , taking  $D'_p = D_p \setminus S_p$  as positive instances and  $D_u$  as negative instances.
- Step 3:  $\mu_p = \frac{1}{|S_p|} \sum_{i:S_p} p(x_i = 1|\theta)$ , (using Platt scaling)
  - $D_p^* = x_u \in D_u : p(x_u = 1) > \mu_p.$
  - $ightharpoonup D_p \leftarrow D_p \cup D_p^*, \ D_n \leftarrow D_u \setminus D_p^*$

#### **Datasets**

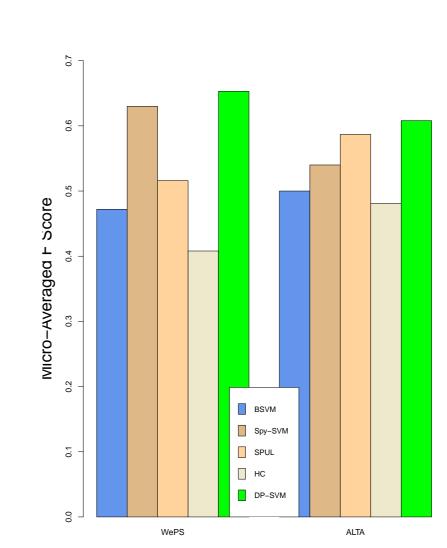
- ► SemEval-2007 WePS development set
  - ▶ Balanced 1000 end-point URL-pairs dataset from webpages for 49 people.
- ► ALTA-2016 shared task dataset
  - ▶ Balanced 400 end-point URL-pairs dataset that can refer to any entity

#### **Feature Representation**

- ➤ Structural features such as document length difference, URL path length difference
- Semantic features such as unigram cosine similarity, cosine similarity over an average word-level word2vec representation, machine translation scores (BLEU, METEOR, TER)

# **Experimental Results**

- Baselines
  - ► Hierarchical Clustering (HC) Unsupervised Approach
  - ▶ Biased SVM (BSVM) with costs for positive and negative classes
  - Spy-SVM (B. Liu et. al., ICML 2002)
  - SPUL (C. Elkan et. al., KDD 2008)
- Proposed Approach
  - ► DP-SVM (Linear Kernel SVM built using propagated distant labels)



#### Conclusions

- Approach to determining whether two endpoint URLs refer to the same entity.
- ► Two key contributions:
- use of distant supervision
- application of PU Learning to the task