



Burning Glass Technologies: Skill Extraction and Building a Skill Knowledge Graph

Bhuvana Surapaneni
Ly (Harriet) Bui
Rishi Mody
Rishikesh Jha

OUTLINE

- Extract new skills using Wikipedia
- Knowledge base population and visualization using Wikipedia and Wikidata

METHODS

I. Extracting New Skills:

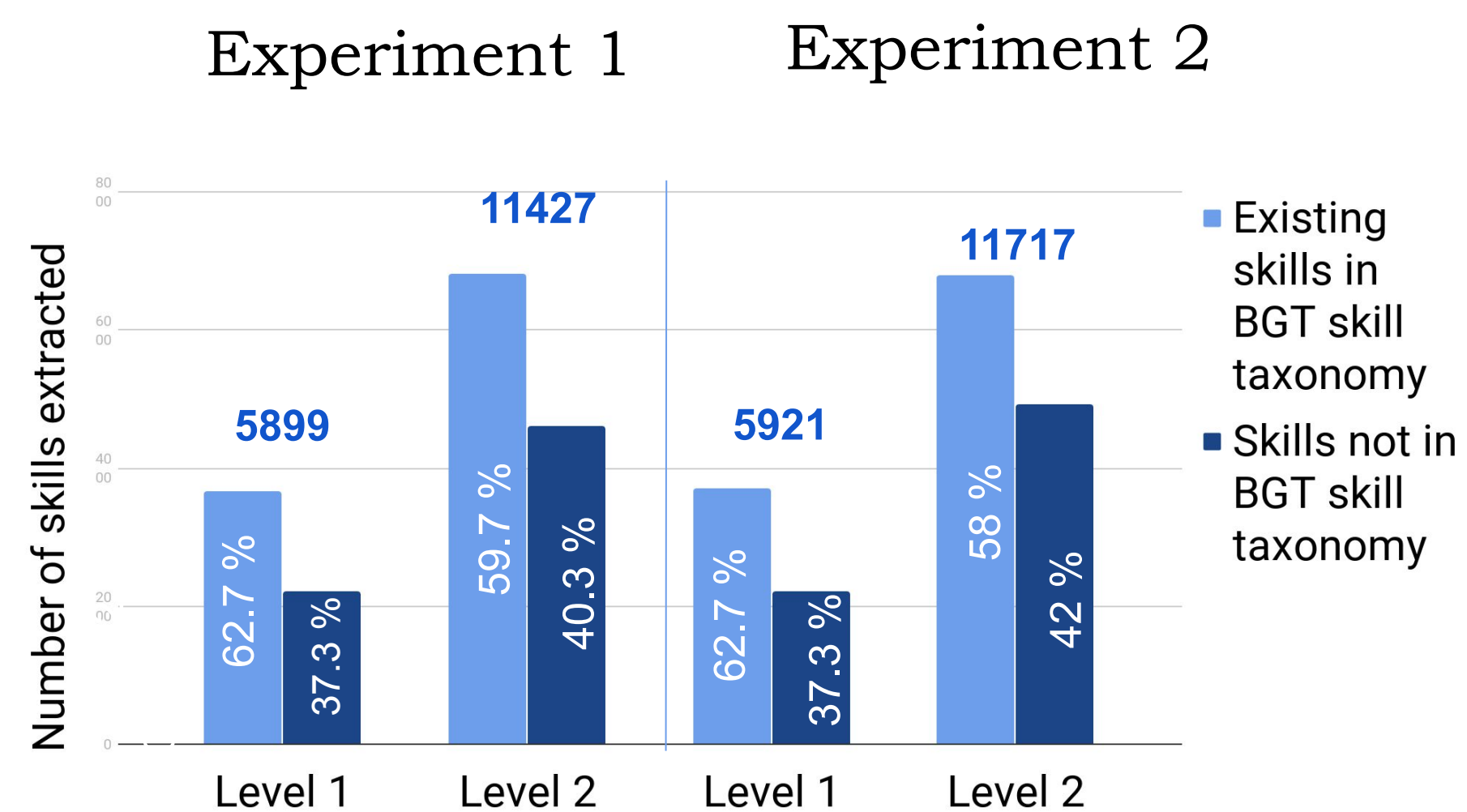
SpaCy Named Entity Recognition Model

Convolutional layers with residual connections, layer normalization and maxout non-linearity, use an imitation learning objective

Experiment 1 Trained on job posting data

Experiment 2 Updated model with Wikipedia training data: 36772 Wikipedia articles

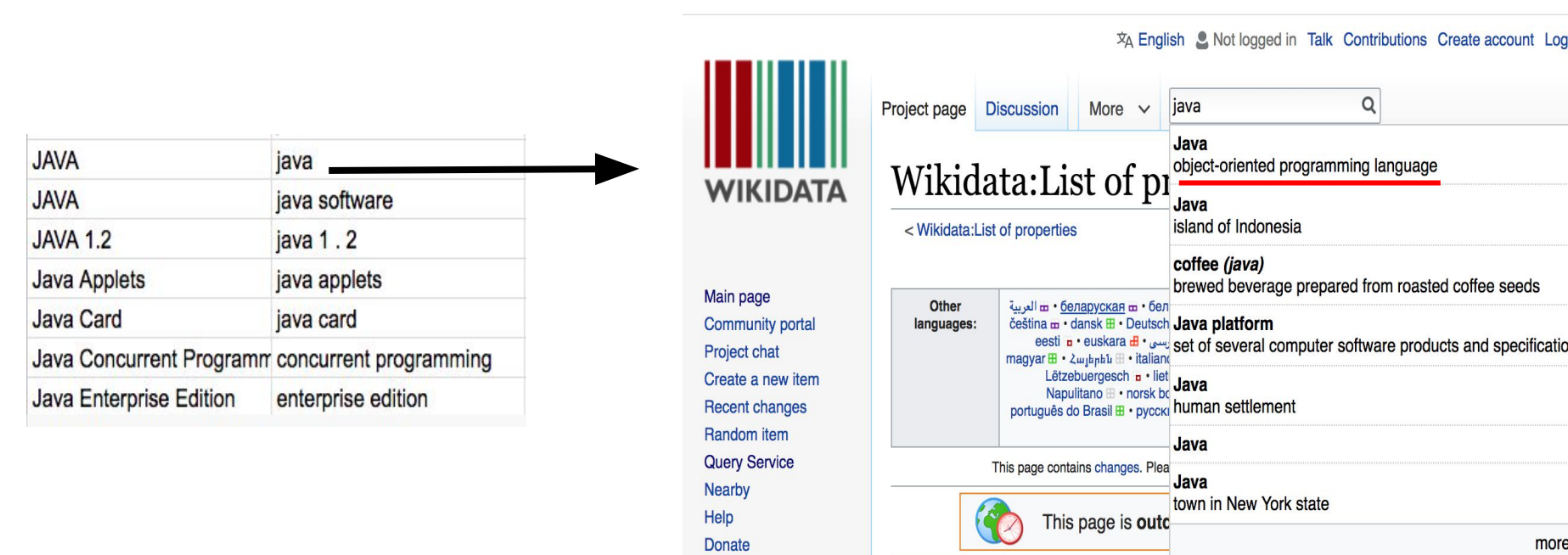
Results



| Accuracy for skills extracted | Experiment 1 | Experiment 2 |
|-------------------------------|--------------|--------------|
| Existing skill extracted | 82.3% | 83.1% |
| New skill extracted | 75.1% | 76% |

II. Constructing Knowledge Graph:

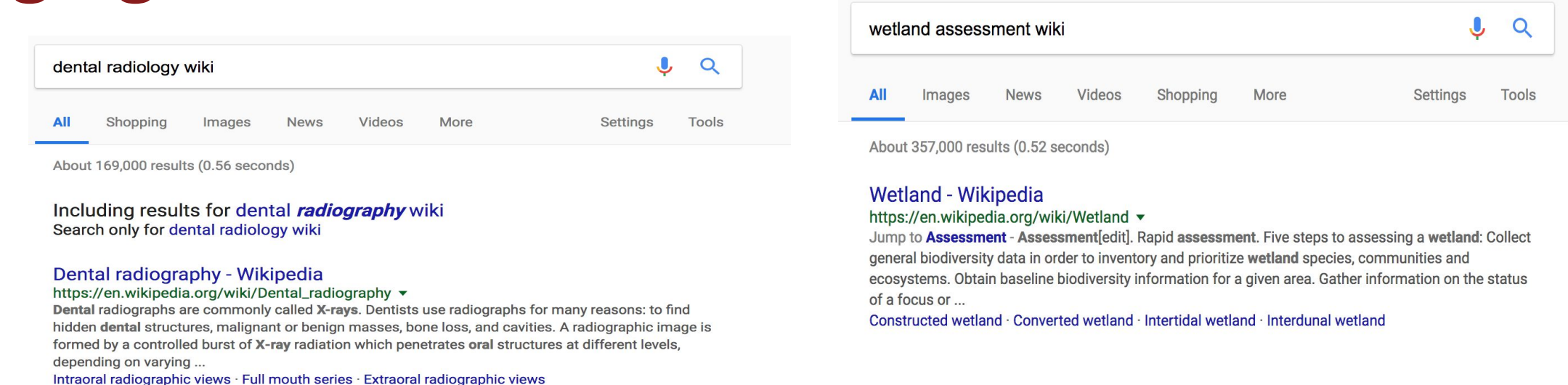
Dataset: Wikidata RDF triplets of selected properties



BGT Taxonomy and Wikidata intersection

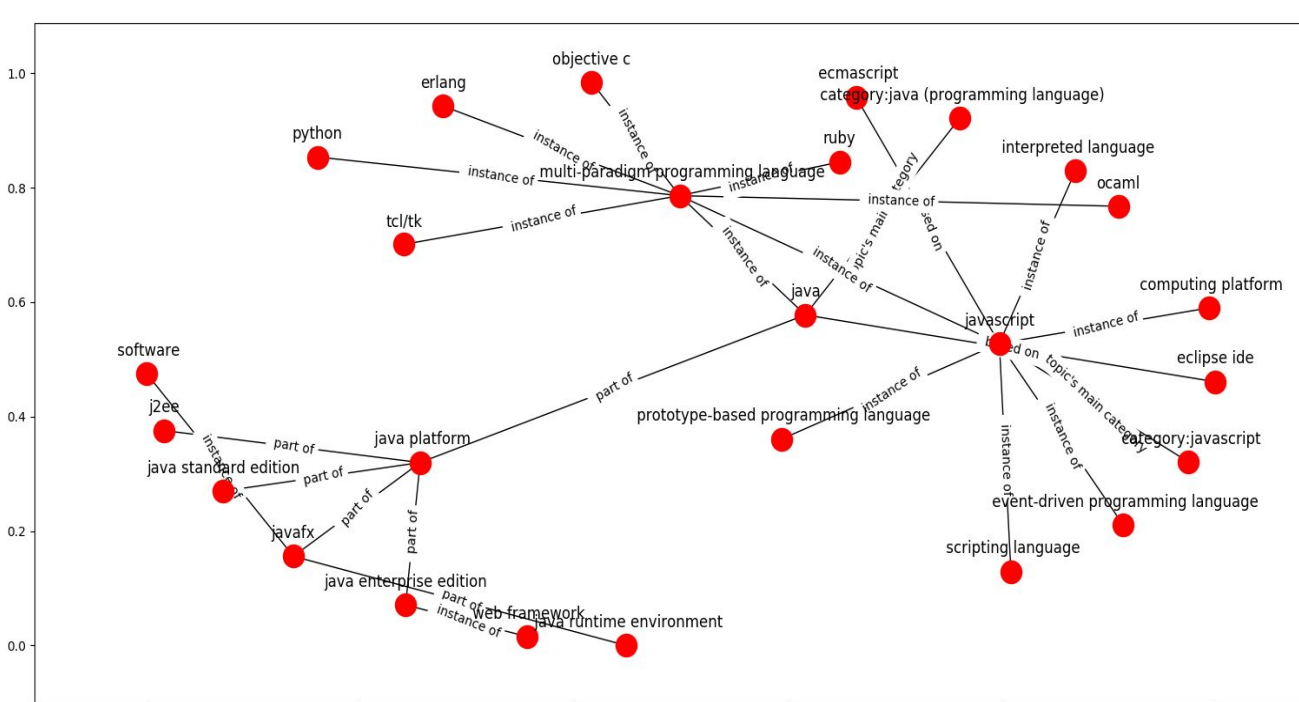
- Found Wikidata entries for 7121(out of 16K) 43.67% coverage on Canonical skills
- 1814 matches are scientific journals

Improving and increasing intersection using google search API



Building Knowledge Base

- Extract RDF triplets of shortlisted properties corresponding to the mapped skills
- Used Apache Jena to create a database of RDF Triples using its Graph interface
- Created API to find relationship between two skills



III. Knowledge Base Population:

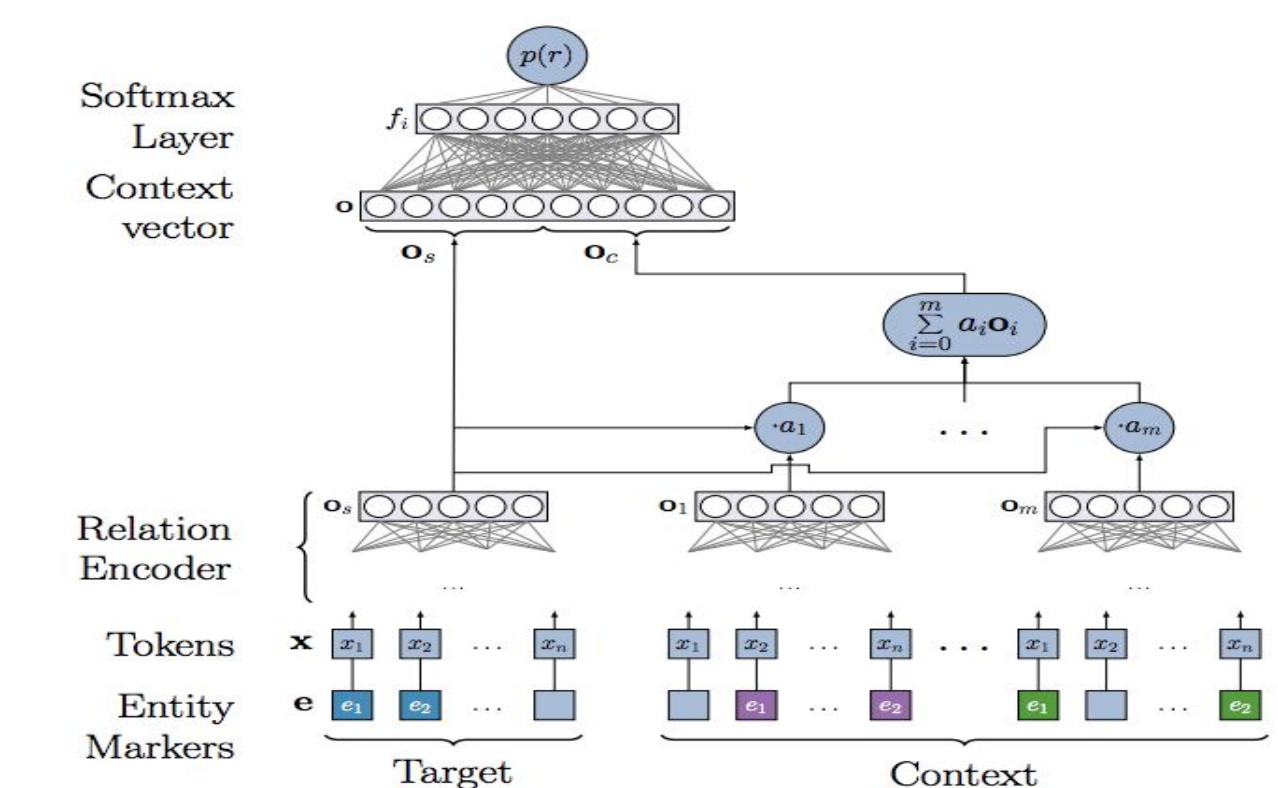
Dataset: Wikidata and Wikipedia

- Used Distant Supervision to create dataset for relation extraction
- Created feature vector by concatenating GloVe word embeddings with marker embeddings

Results using LSTM(Baseline)

| Relation Type | Precision | Recall |
|---------------|-----------|--------|
| Part of | 0.589 | 0.503 |
| Instance of | 0.761 | 0.837 |
| Subclass of | 0.513 | 0.468 |

Context Aware LSTM Model



Future Work

- Deduplication of the extracted skills
- Creating interactive UI for exploring skill relations from knowledge graph
- Exploring different approaches for improving relation extraction model

REFERENCES

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- <https://engineering.linkedin.com/blog/2016/10/building-the-linkedin-knowledge-graph>
- Sorokin et al. Context-Aware Representations for Knowledge Base Relation Extraction. Proceedings of the 2017 Conference on Empirical Methods in NLP, pages 1784-1789
- Traylor et al. Learning String Alignments for Entity Aliases. NIPS 2017