Entity Retrieval in the Knowledge Graph with Hierarchical Entity Type and Content

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Introduction

- Entity: broadly defined as a thing or concept that exists in the world or fiction
 - such as a person, a company or a color.
- Entity Retrieval: answering entity-targeted queries
 - e.g. 'give me all US food company stocks'
 - by returning a ranked list of entities from a structured knowledge base or knowledge graph.
 - e.g. McDonald's(NYSE:MCD), Burger King(NYSE:BKC), Kellogg's(NYSE:K) ...



who invented the light bulb





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Incandescent light bulb > Inventors





Hiram Maxim



Joseph Swan





Entities

Who Invented the Light Bulb? - Live Science

https://www.livescience.com > History ▼

Aug 16, 2017 - **Edison** and his team of researchers in Edison's laboratory in Menlo Park, N.J., tested more than 3,000 designs for bulbs between 1878 and 1880. In November 1879, **Edison** filed a patent for an electric lamp with a carbon filament.

Who really invented the light bulb? - Science Focus - BBC Focus ...

https://www.sciencefocus.com > Everyday science ▼

US **inventor Thomas Edison** often gets all the credit, but was he really the ... In 1878, another British chemist, **Joseph Swan**, publicly demonstrated the first **light** ...

Videos



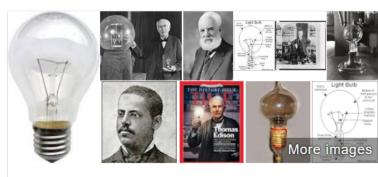
Ask History: Who Really Invented the Light Bulb? | History



Who invented the Lightbulb first?



Who Really Invented
The Light Bulb - History
of Everything



Incandescent light bulb



An incandescent light bulb, incandescent lamp or incandescent light globe is an electric light with a wire filament heated to such a high temperature that it glows with visible light. The filament is protected from oxidation with a glass or fused quartz bulb that is filled with inert gas or a vacuum. Wikipedia

People also search for



diode

LED lamp





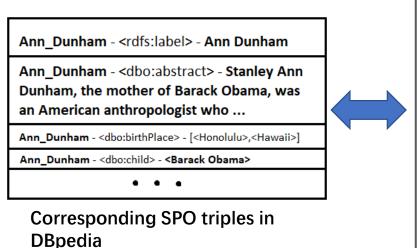
escent Compact fluorescent

View 10+ more

Introduction – Entity Representation

Entities are usually represented by subject-predicate-object(SPO)

triples in the knowledge graph



Knowledge base entry for ANN DUNHAM <rdfs:label>: Ann Dunham <dbo:abstract>: Stanley Ann Dunham, the mother of Barack Obama, was an American anthropologist who ... <dbo:birthPlace>: [<Honolulu>, <Hawaii>] <dbo:child>: <Barack_Obama> <dbo:wikiPageWikiLink>: [<United_States>, <Family_of_Barack_Obama>, ...]



Introduction – Previous Approaches

- Employ standard document retrieval methods
- by converting SPO triples into pseudo documents
 - e.g. "Steve_Jobs-birthYear-1955" to "steve jobs birth year 1955"

Knowledge base entry for ANN DUNHAM

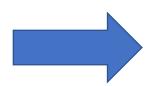
```
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<dbo:abstract>:
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    anthropologist who ...

<dbo:birthPlace>:
    [<Honolulu>, <Hawaii>]

<dbo:child>:
    <Barack_Obama>

<dbo:wikiPageWikiLink>:
    [<United_States>,
        <Family_of_Barack_Obama>, ...]
```



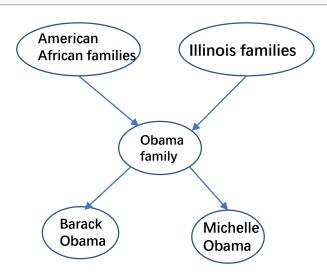
multi-fielded pseudo document

Introduction

 However, there is more structure available than in standard document retrieval

Entity Types in the **type taxonomy**

Categories: Obama family | American anthropologists | American women People from Honolulu | People from Mercer Island, Washington | People American people of Scottish descent | American people of Swiss descent Deaths from uterine cancer | Mothers of Presidents of the United States



Entity Descriptions in the knowledge source

Ann Dunham From Wikipedia, the free encyclopedia Not to be confused with the British equestrian Anne Dunham Stanley Ann Dunham (November 29, 1942 - November 7, 1995) was an American anthropologist who specialized in the economic anthropology and rural development of Ann Dunham Indonesia.[1] She was the mother of Barack Obama, the 44th President of the United States. Dunham was known as Stanley Ann Dunham through high school, then as Ann Dunham, Ann Obama, Ann Soetoro, Ann Sutoro, and finally after her second divorce as Ann Dunham [2] Born in Wichita Kansas, Dunham spent her childhood in California, Oklahoma, Texas and Kansas, her teenage years in Mercer Island, Washington, and most of her adult received master of arts (1974) and PhD (1992) degrees, also in anthropology. [5] She also attended University of Washington at Seattle in 1961–1962. Interested in craftsmanship, weaving, and the role of women in cottage industries. Dunham's research focused on women's work on the island of Java and blacksmithing in Indonesia. To address the problem of poverty in rural villages, she created microcredit programs while working as a consultant for the United States Agency for International Development, Dunham was also employed by the Ford Foundation in Jakarta and she consulted with the Asian Development Bank in Gujranwala, Pakistan. Towards the latter part of her life, she worked with Bank Rakyat Indonesia, where she helped apply her research to the largest microfinance program in the world.^[5] After her son was elected President, interest renewed in Dunham's work: the University of Hawaii held a symposium about her research; an exhibition of Dunham's Indonesian batik textile collection toured the United States; and in December 2009, Duke University Press published Surviving against the Odds: Village Industry in Indonesia, a book based on Stanley Ann Dunhan Dunham's original 1992 dissertation. Janny Scott, an author and former New York Times reporter, published a biography about Ann Dunham's life titled A Singular Woman in 2011. November 29, 1942 Posthumous interest has also led to the creation of The Ann Dunham Soetoro Endowment in the Anthropology Department at the University of Hawai'i at Mānoa, as well as the Ann Dunham Soetoro Graduate Fellowships, intended to fund students associated with the East-West Center (EWC) in Honolulu, Hawaii. [1] November 7, 1995 (aged 52 Honolulu, Hawaii, U.S. In an interview, Barack Obama referred to his mother as "the dominant figure in my formative years ... The values she taught me continue to be my touchstone when it comes to how I go about the world of politics."[7]

Contents [hide] 1 Early life

2 Family life and marriage: 2.1 First marriage 2.2 Second marriag 3 Professional life

4 Illness and death 5 Posthumous interes

6 Personal beliefs

Cause of Complications from uterir cancer and ovarian cancer

Education University of Washington University of Hawaii, Manoa (B MA, PhD)

Spouse(s) Barack Obama Sr. (m. 1961; div. 1964) (m. 1965; div. 1980)

hildren Barack Obama Maya Soetoro-No Parent(s) Stanley Armour Dunhar

Our proposal

- Improve entity retrieval by incorporating
 - hierarchical entity type information
 - entity descriptions.

Model Description

- Given a query Q
- A candidate entity *E*
- $D = \{D_f\}$: multi-fielded pseudo documents of E.

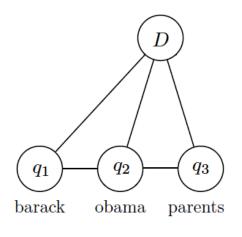
The overall scoring function

$$g(Q, D, E) = \lambda_W \tilde{g}(Q, D, E) + (1 - \lambda_W) \tilde{h}(Q, E)$$

 $\tilde{g}(Q,D,E)$: scoring function incorporating entity type information

 $\tilde{h}(Q,E)$: scoring function incorporating entity descriptions

 λ_w : model parameter



Markov random field based framework

Example: query='barack Obama parents'

unigrams={'barack','Obama','parents'}

bigrams={'barack Obama', 'Obama parents'}

$$ilde{g}(Q,D,E) = \max_{c \,\in\, types(E)} \max_{p \,\in\, Paths(c)} \{\, \lambda_T \, \sum_{q_i \in \,Q} ilde{f}_T \, (q_i,D,p) \,+ \,$$

 $\lambda_O \sum_{q_i,q_i \in O} ilde{f}_O \left(q_i,q_{i+1},D,p
ight) +$

$$\lambda_{U}\sum_{q_{i},q_{i+1}\in Q} ilde{f}_{U}(q_{i},q_{i+1},D,p)\}$$

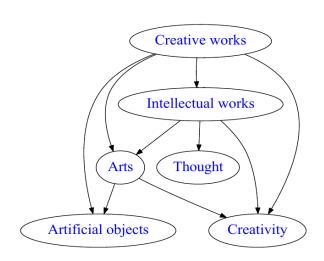
feature function for unigrams

feature function for ordered bigram occurrences

feature function for unordered bigrams occurrences

- Consider path as a variable
- Enumerate each path p in the type taxonomy
 - starts from an entity type *c*

$$egin{aligned} ilde{g}(Q,D,E) &= \max_{c \,\in\, types(E)} \max_{p \,\in\, Paths(c)} \{\, \lambda_T \sum_{q_i \in Q} ilde{f}_T(q_i,D,p) \,+ \ & \lambda_O \sum_{q_i,q_{i+1} \in Q} ilde{f}_O(q_i,q_{i+1},D,p) \,+ \ & \lambda_U \sum_{q_i,q_{i+1} \in Q} ilde{f}_U(q_i,q_{i+1},D,p) \} \end{aligned}$$



Dirichlet prior smoothed feature function

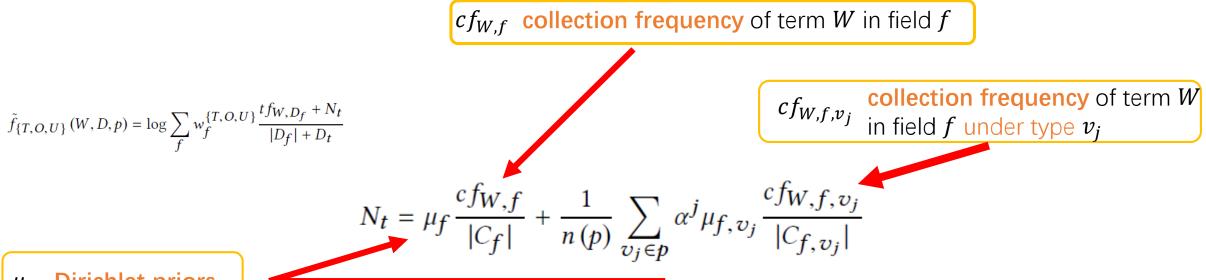
$$\tilde{f}_{\{T,O,U\}}(W,D,p) = \log \sum_{f} w_{f}^{\{T,O,U\}} \frac{t f_{W,D_{f}} + N_{t}}{|D_{f}| + D_{t}}$$

 tf_{W,D_f} : term frequency of W in the pseudo document of f

W generalizes to unigrams and bigrams

 N_t , D_t : path-aware smoothing components

 Specifically, we add a path-aware smoothing component in the feature function



 μ_f . Dirichlet priors

$$D_{t} = \mu_{f} + \frac{1}{n(p)} \sum_{v_{j} \in p} \alpha^{j} \mu_{f, v_{j}}$$

$$n(p), \alpha \text{ model parameters}$$

Can derive many existing models by taking particular combinations of parameters (LM,MLM,SDM,PRMS...) 12

Model Description – Time Complexity Analysis

```
T = O(|types(E)| \cdot |paths(c)| \cdot |p| \cdot |Q| \cdot |F|)
```

```
|types(E)| number of entity types considered |paths(c)| number of paths explored in the type taxonomy |p| average length of a path |Q| average length of a query |F| number of fields
```

- Given a query and a candidate entity
- assuming the operation of computing term frequencies and collection frequencies takes constant time
- Path exploration implemented in a recursive way

Model Description - Incorporating entity descriptions

- Similar formulation
- Enumerate each path p from lead section to a leaf section

$$\begin{split} \tilde{h}\left(Q,E\right) &= \max_{p \in T(E)} \left\{ \lambda_T \sum_{q_i \in Q} \tilde{h}_T\left(q_i, E, p\right) + \right. \\ &\left. \lambda_O \sum_{q_i, \, q_{i+1} \in Q} \tilde{h}_O\left(q_i, q_{i+1}, E, p\right) + \right. \\ &\left. \lambda_U \sum_{q_i, \, q_{i+1} \in Q} \tilde{h}_U\left(q_i, q_{i+1}, E, p\right) \right\} \end{split}$$

$$\tilde{h}_{\left\{T,O,U\right\}}\left(W,E,p\right) = \log \frac{\sum_{s_{j} \in p} \beta^{j} \cdot t f_{W,s_{j}} + \mu_{d} \frac{c f_{W,C_{d}}}{|C_{d}|}}{\sum_{s_{j} \in p} \beta^{j} \cdot |s_{j}| + \mu_{d}}$$

Food sources Animals
Sweet
Production
Sour
Taste
perception
Bitter
Umami

 s_i : the j-th section in the path

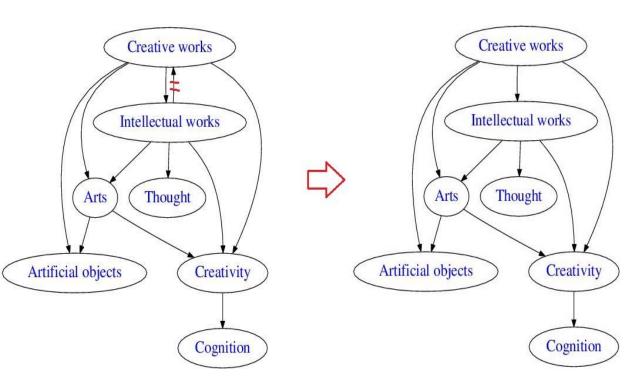
• Knowledge graph: DBpedia 2015-10

Knowledge source: Wikipedia 2015-10

Type taxonomy: Wikipedia Category System

- Test collection: DBpedia-Entity v2
 - Four benchmark datasets with 467 queries in total
 - INEX-LD: keyword based queries
 - SemSearch_ES: named entity targeted queries
 - ListSearch: queries that seek a particular list of entities
 - QALD2: natural language questions
 - Metric: NDCG@10, MAP@100
 - "DBpedia-Entity v2: A Test Collection for Entity Search", Hasibi et.al, SIGIR 2017

Wikipedia Category Processing



Step 1 Construct Wikipedia Category Graph

Each node represents a Wikipedia category and connects to its parent categories.

Step 2 Remove Strongly Connected Components in the Graph

Divide the graph into SCCs. A SCC is then reduced by removing common edges shared by intersected elementary cycles or one edge of a sole circle.

Baselines

 BM25F: the BM25 Model with extension to multiple weighted fields

Stephen Robertson, Hugo Zaragoza, et al. (2009)

 PRMS: the Probabilistic Retrieval Model for Semi structured Data

Jinyoung Kim, Xiaobing Xue, and W Bruce Croft. (2009)

- **LM**: standard language modelling Zhai Chengxiang, and John Lafferty. (2004)
- MLM: Mixture of Language models

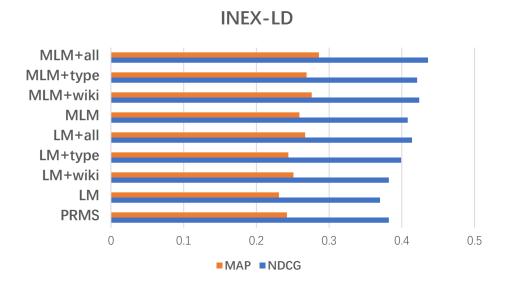
 Paul Ogilvie and Jamie Callan. (2003)
- SDM: Sequential Dependence Model Donald Metzler and W Bruce Croft. (2005)
- FSDM: Fielded Sequential Dependence Model

Nikita Zhiltsov, Alexander Kotov, and Fedor Nikolaev. (2015) Faegheh Hasibi, Krisztian Balog, and Svein Erik Bratsberg. (2016)

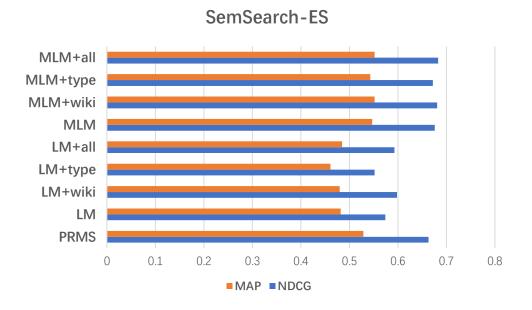
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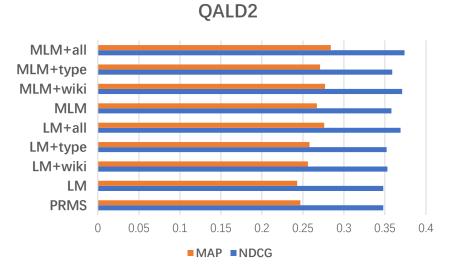
| Model | INEX-LD | | SemSearch ES | | ListSearch | | QALD2 | |
|-----------|---------|------------|--------------|-------|------------|-------|------------|------------|
| | NDCG | MAP | NDCG | MAP | NDCG | MAP | NDCG | MAP |
| BM25F | .413 | .253 | .624 | .488 | .380 | .279 | .337 | .248 |
| PRMS | .382 | .242 | .663 | .529 | .378 | .267 | .348 | .247 |
| LM | .370 | .231 | .574 | .482 | .389 | .284 | .348 | .243 |
| LM+wiki | .392* | .251* | .598* | .480 | .397 | .297* | .353 | .256* |
| LM+type | .399* | $.244^{*}$ | .552 | .461 | .398* | .301* | .352 | $.258^{*}$ |
| LM+all | .414* | .267* | .593* | .485 | .410* | .323* | .369* | .276* |
| MLM | .408 | .259 | .676 | .547 | .372 | .278 | .358 | .267 |
| MLM+wiki | .424* | .276* | .681 | .552 | .380 | .289* | .371* | .277* |
| MLM+type | .421* | .269* | .672 | .543 | .374 | .283 | .359 | .271 |
| MLM+all | .436* | .286* | .683* | .552* | .388* | .298* | $.374^{*}$ | .284* |
| SDM | .373 | .233 | .604 | .510 | .394 | .288 | .354 | .251 |
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| FSDM+wiki | .412* | .270* | .671 | .543 | .407 | .304* | .388* | .276* |
| FSDM+type | .402* | .247 | .662 | .539 | .402 | .282 | .372* | .266 |
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Baselines – LM based ones

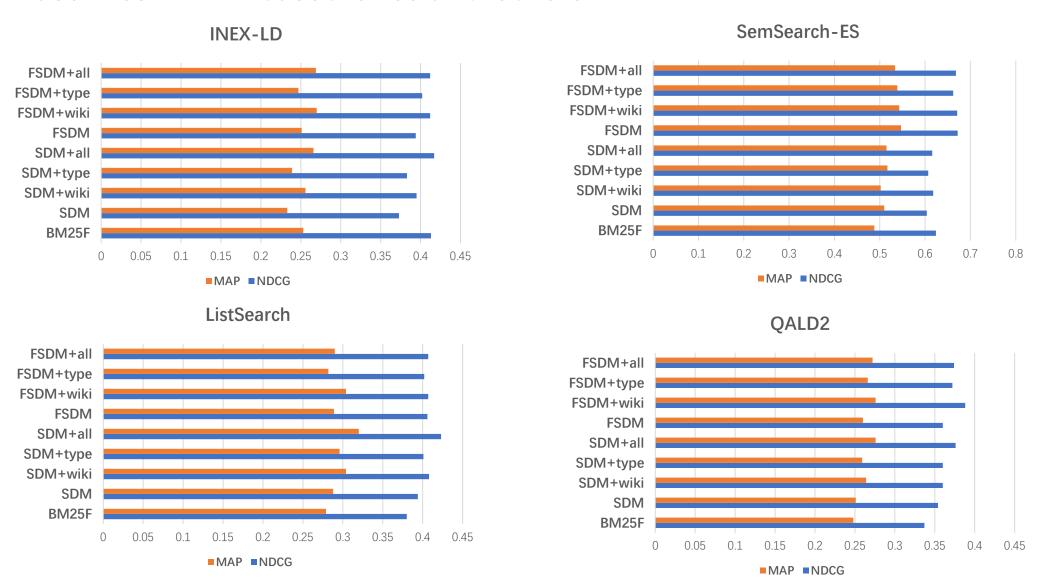






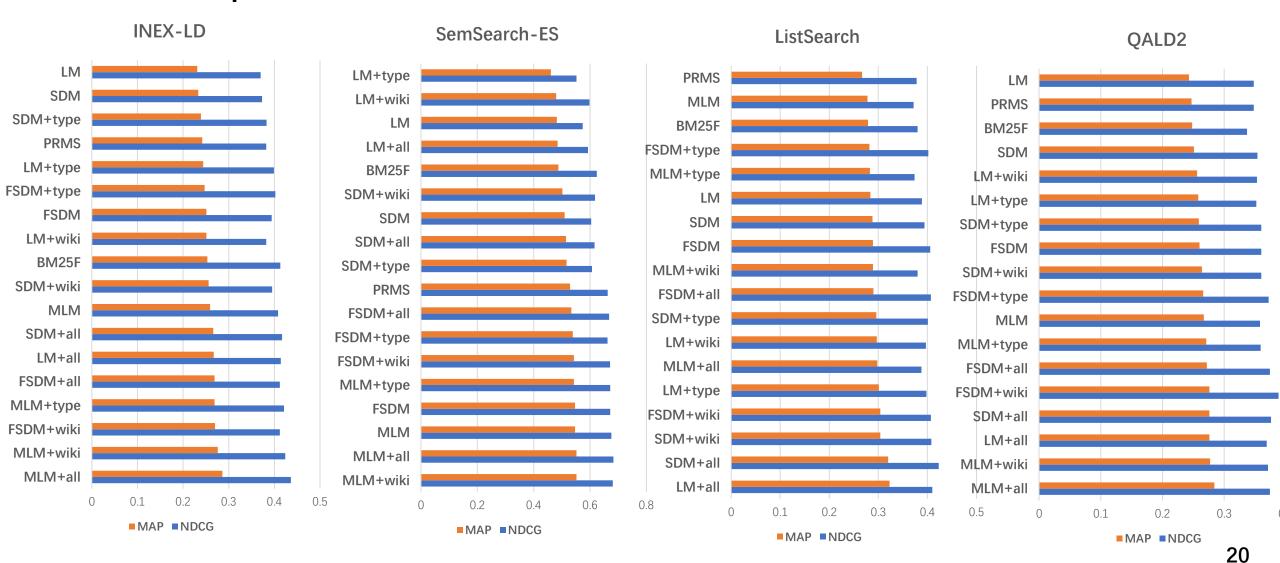


Baselines - MRF based ones and others



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Dataset specific results



Analysis

- Incorporating both entity documents and type information brings the largest improvements than with either of them.
- For all dataset except SemSearch ES, the entity descriptions and type information contribute roughly equally
- On SemSearch ES, most improvements comes from exploiting the entity descriptions
- Due to different query characteristics/intents in each dataset.

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Summary

- Propose a Markov random field based framework incorporating both hierarchical entity type information and entity descriptions
 - Add a path aware smoothing component in the feature functions
- Propose a graph-based approach to process Wikipedia category system
- Comparing to existing type-aware entity retrieval models, our model does not require explicit inference of query type
- Future work includes investigation of incorporating more knowledge graph structures and query characteristics/intents

Thank you

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