

Context- and Content-aware Embeddings for Query Rewriting in Sponsored Search

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Yahoo Labs - US & London

Query Rewriting

Try our state-of-the-art system for query rewriting.

bank of america loans

| Result | Relevance | + |
|---------------------------------------|-----------|---|
| bank_of_america_loans_personal | 73.3375% | |
| bank_of_america_personal_loans | 70.0765% | |
| bank_of_america_car_loans | 68.5054% | |
| bank_of_america_home_loans | 68.2654% | |
| bank_of_america_auto_loans | 65.7220% | |
| bank_of_america_mortgage | 65.6517% | |
| bank_of_america_personal_loans_online | 64.9183% | |
| bank_of_america_loan | 64.1622% | |
| bank_of_america_personal_loan | 62.3031% | |
| chase_personal_loans | 62.23% | |
| bank_of_america_student_loans | 61.3762% | |

Input Type:

Search Query

Output Type:

Search Query

Search

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Query Rewriting for Advertising



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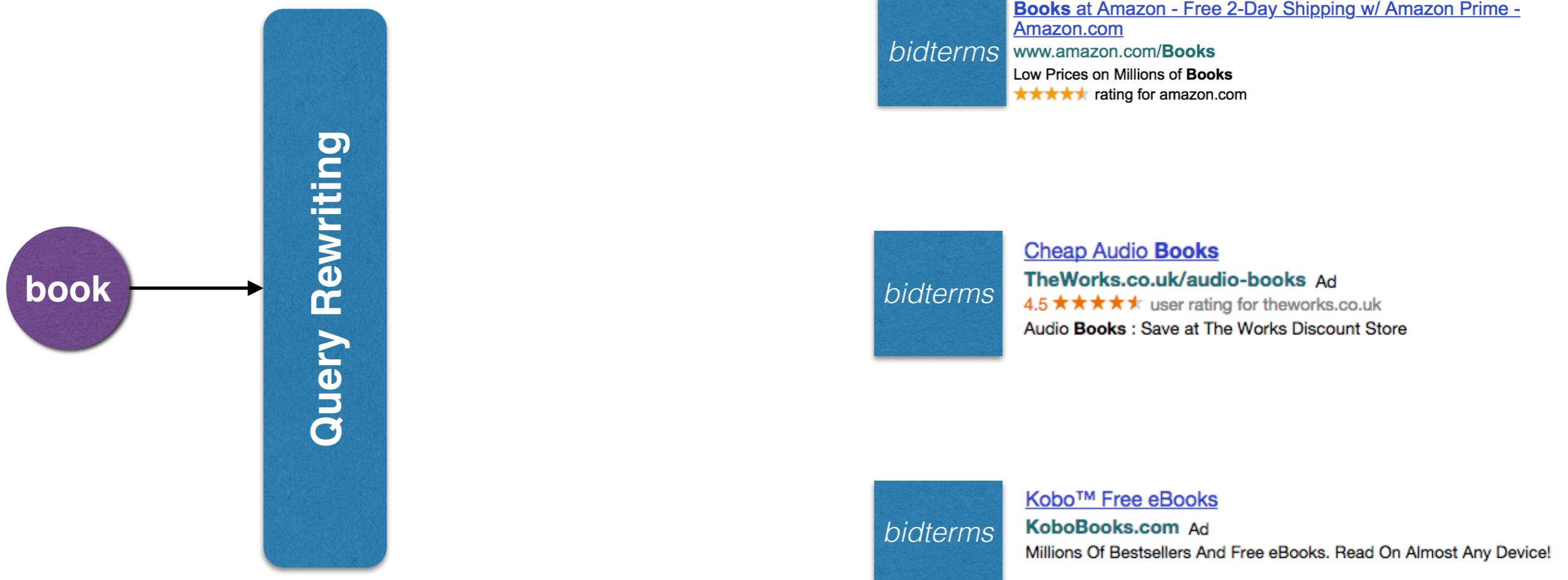


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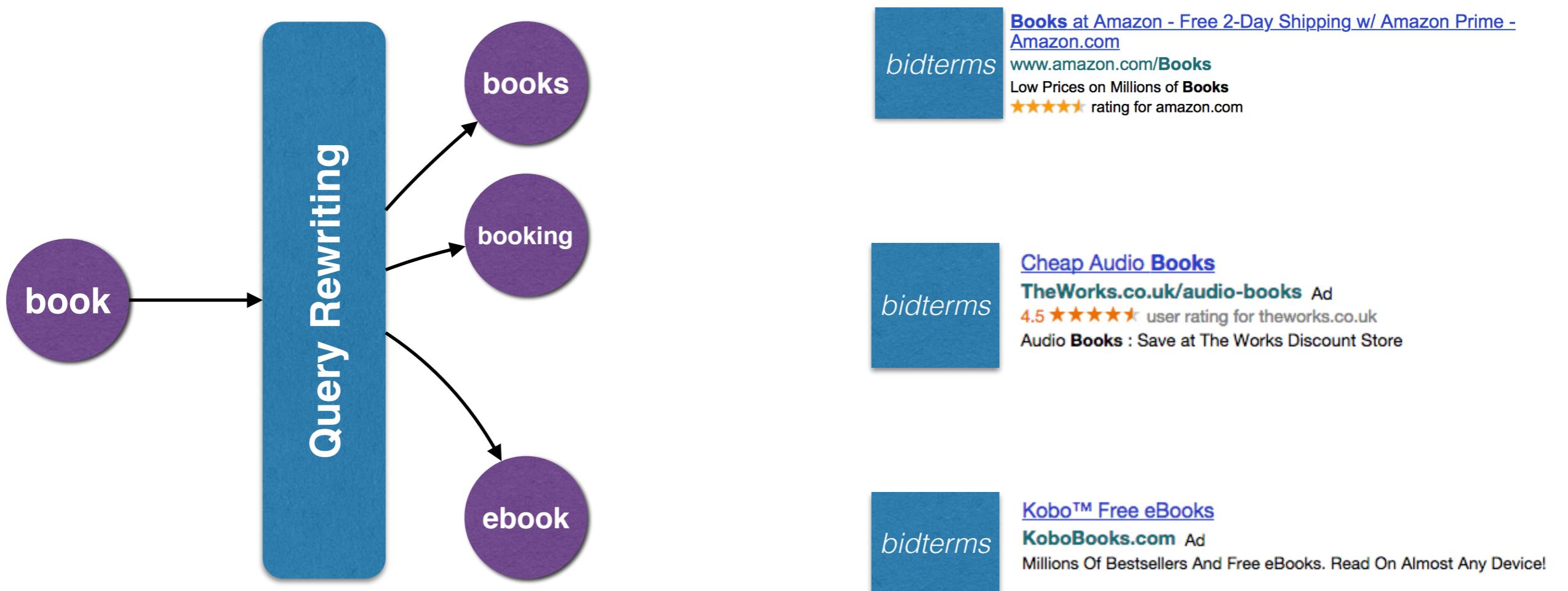


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Query Rewriting for Advertising



Query Rewriting for Advertising

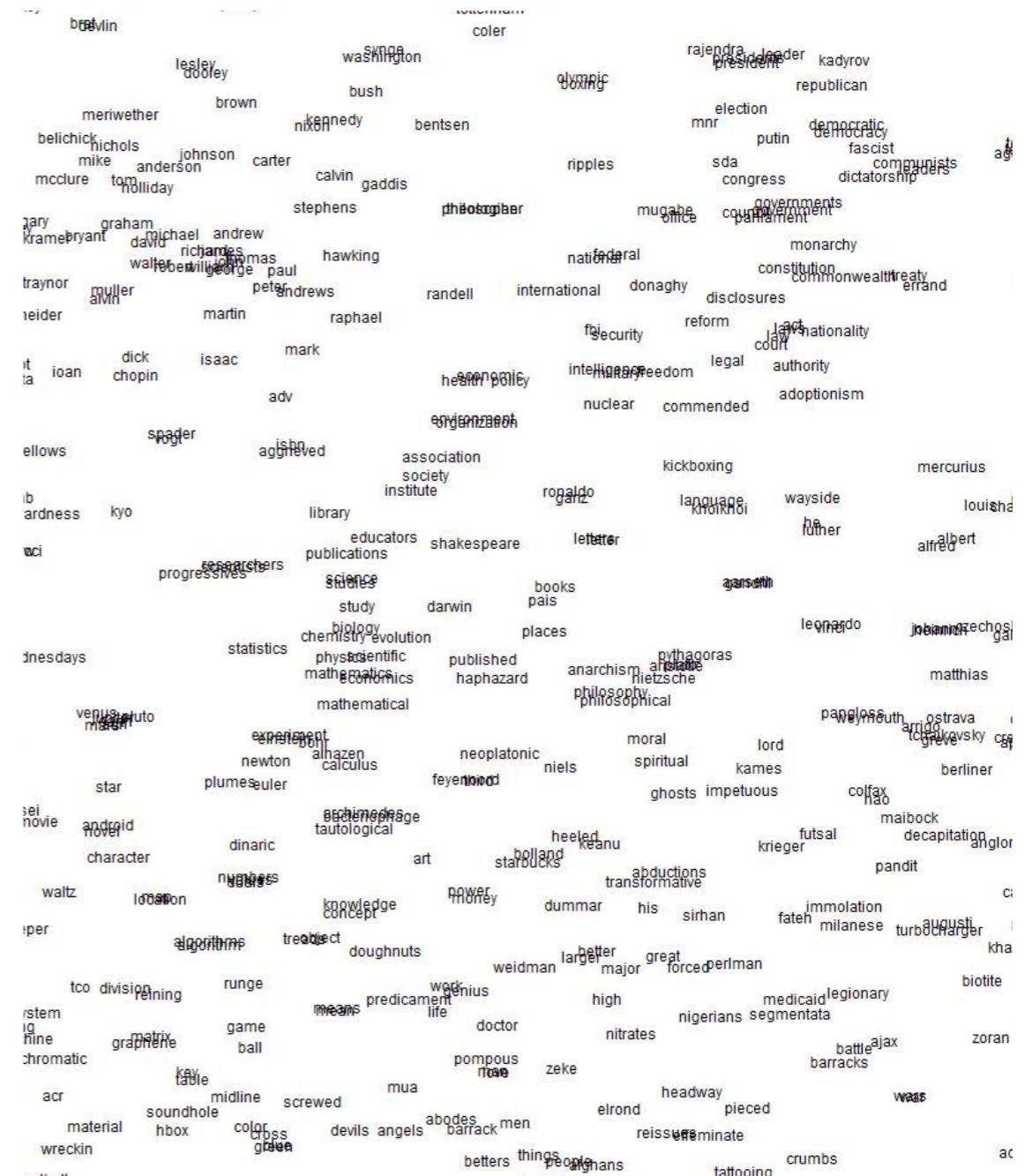


(Lots of) Previous Work

- Graph-based Methods
 - Click Graph
 - Query-Flow Graph
 - Term-Query Graph
 - ...
 - Clustering queries from history
 - Supervised Learning-based methods
 - ...
- See more ads for:
- [nike shoes](#)
- [cheap nike shoes](#)
- [nike shoes online](#)
- [nike shoes sale](#)
- [kids nike shoes](#)
- Ads
- [Nike Shoes](#)
- [lowpriceshopper.co](#)
- Best **Nike Shoes** Deals

Using Word Embeddings

- Sessions represent well the interactions of users with a search engines:
 - *queries*
 - *clicks*: algo results and ads
 - We generate a distributed representation of actions in a common vector space:
 - inspired by *word2vec*



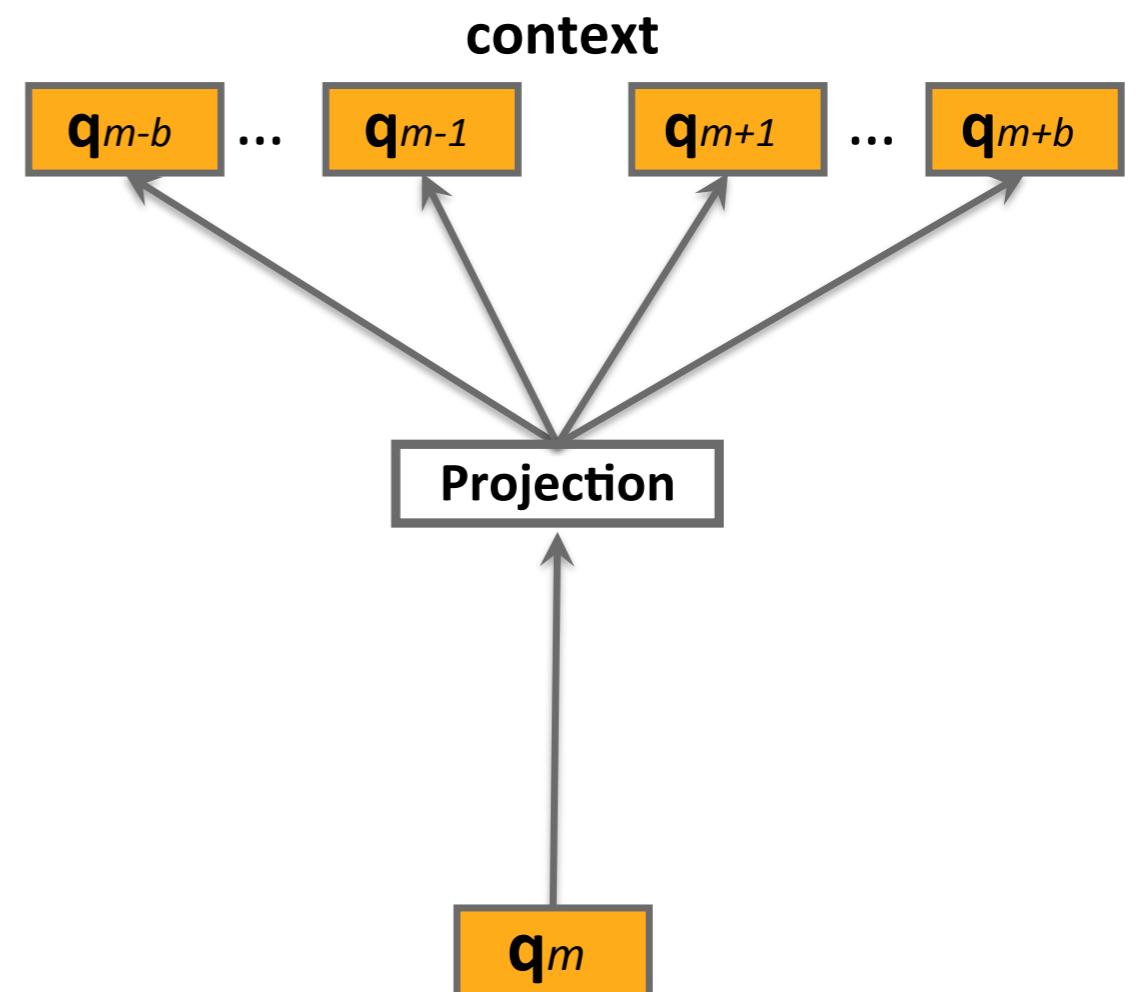
Models for Embedding Interactions

- Context2Vec
- Content2Vec
- Context-Content2Vec



Model 1: Context2vec

- Session as a sentence:
 - actions are words (e.g., ‘nike shoes’)
 - vectors are learned on each action/word



$$\mathcal{L} = \sum_{s \in \mathcal{S}} \sum_{q_m \in s} \sum_{-b \leq i \leq b, i \neq 0} \log \mathbb{P}(q_{m+i} | q_m).$$

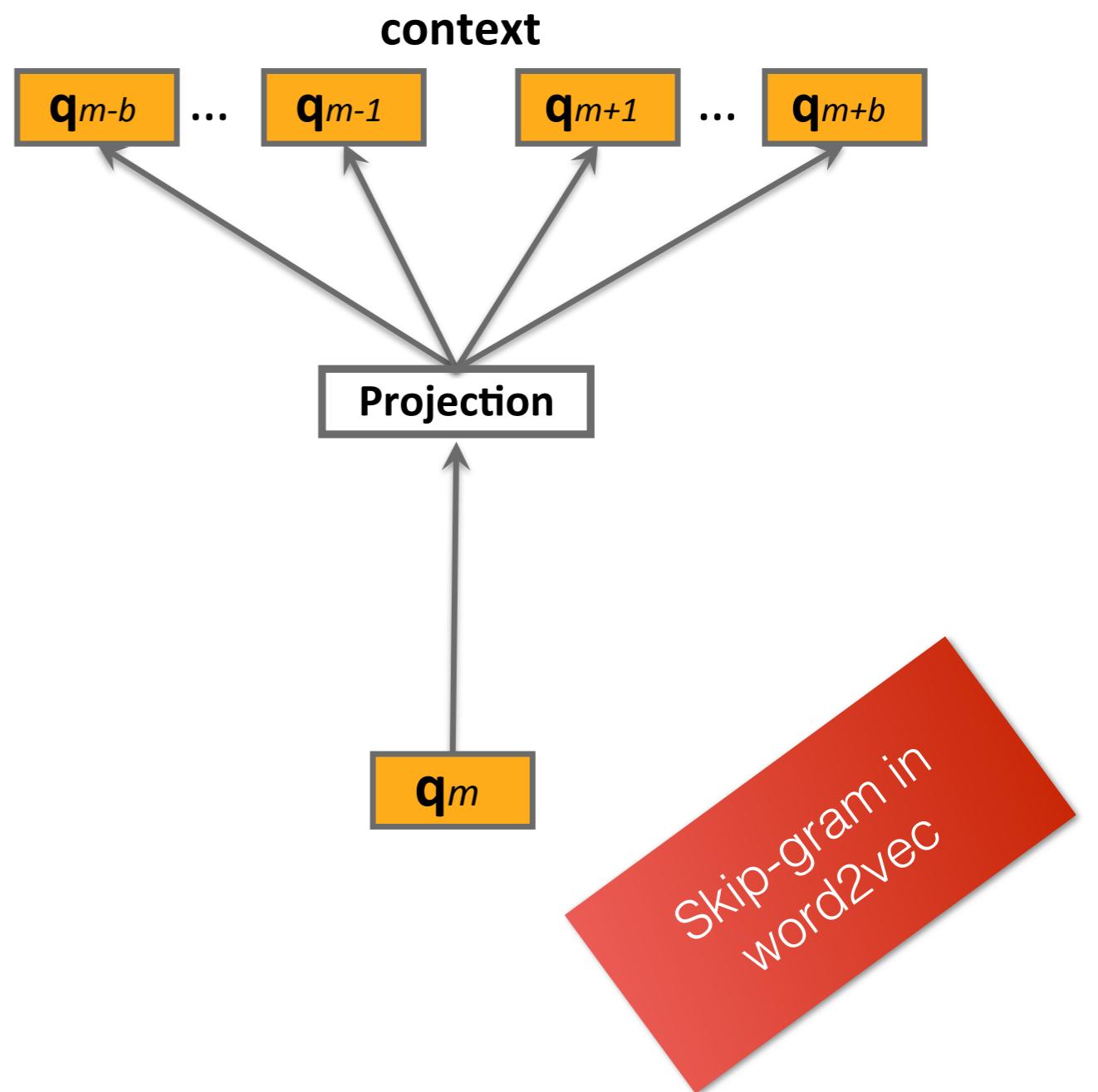
$$\mathbb{P}(q_{m+i} | q_m) = \frac{\exp(\mathbf{v}_{q_m}^\top \mathbf{v}'_{q_{m+i}})}{\sum_{q=1}^V \exp(\mathbf{v}_{q_m}^\top \mathbf{v}'_q)}$$

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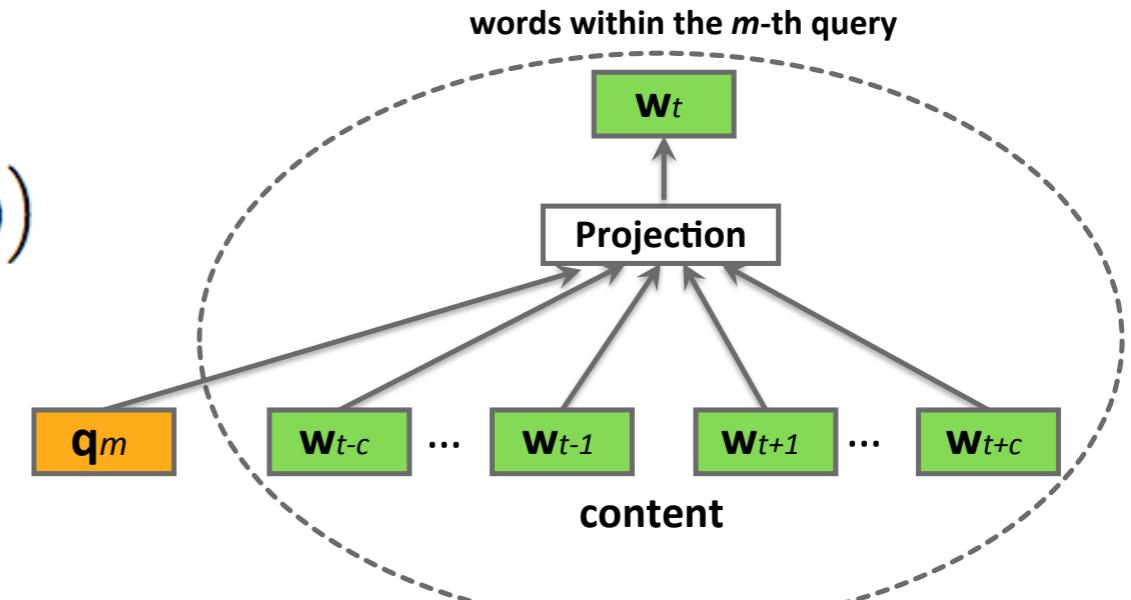
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Model 2: Content2Vec

- Consider the “content” of the action

$$\begin{aligned}\mathcal{L} = & \sum_{s \in \mathcal{S}} \left(\sum_{q_m \in s} \log \mathbb{P}(q_m | w_{m1} : w_{mT_m}) \right. \\ & \left. + \sum_{w_{mt} \in q_m} \log \mathbb{P}(w_{mt} | w_{m,t-c} : w_{m,t+c}, q_m) \right)\end{aligned}$$



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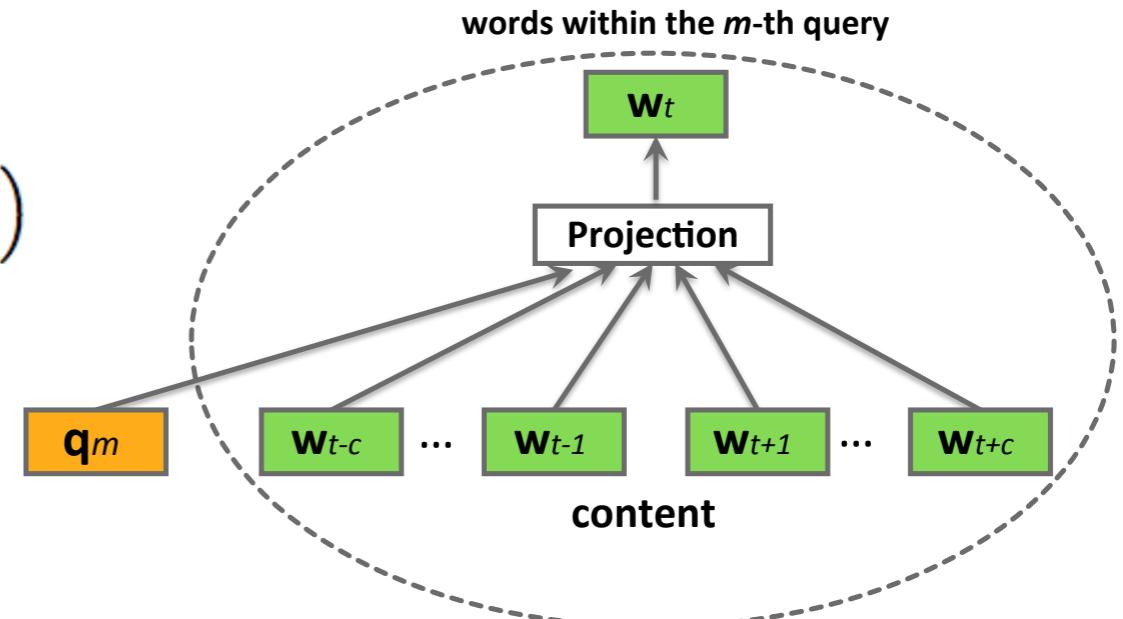
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$$\bar{\mathbf{v}} = \frac{1}{2c+1} (\mathbf{v}_{q_m} + \sum_{-c \leq j \leq c, j \neq 0} \mathbf{v}_{w_{m,t+j}})$$

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$$\bar{\mathbf{v}}_m = \frac{1}{T_m} \sum_{t=1}^{T_m} \mathbf{v}_{w_{mt}}$$

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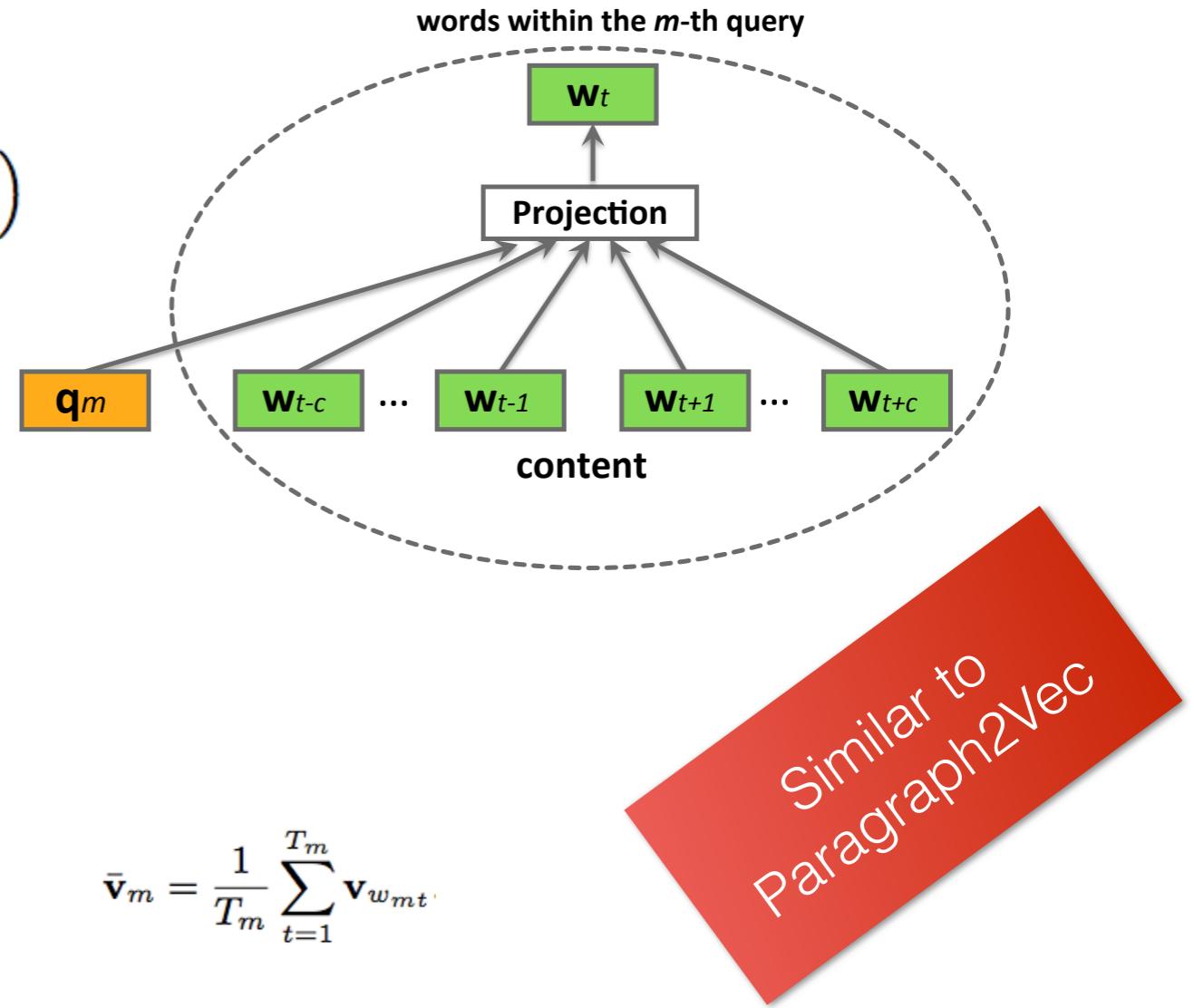
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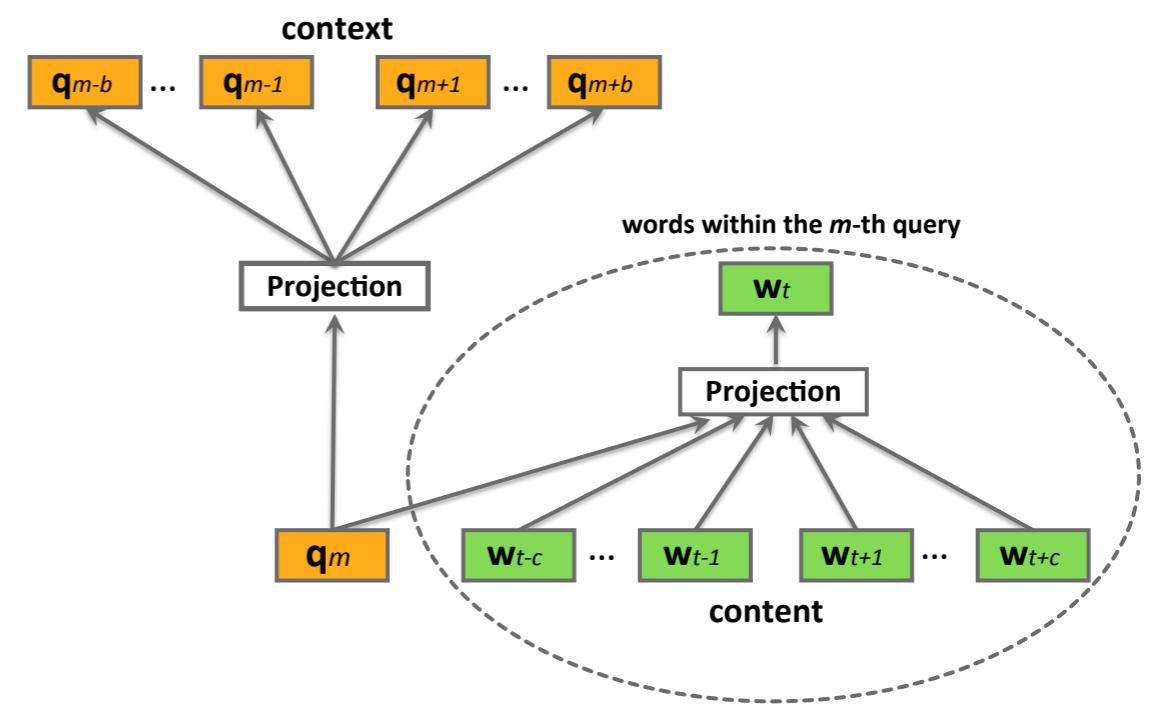
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Model 3: Context-Content2Vec

- A combination of the previous two models
- When an action is not popular we give more importance to its content

$$\begin{aligned} \mathcal{L} = & \sum_{s \in \mathcal{S}} \sum_{q_m \in s} \left(\sum_{-b \leq i \leq b, i \neq 0} \log \mathbb{P}(q_{m+i} | q_m) \right. \\ & + \alpha_m \log \mathbb{P}(q_m | w_{m1} : w_{mT_m}) \\ & \left. + \sum_{w_{mt} \in q_m} \log \mathbb{P}(w_{mt} | w_{m,t-c} : w_{m,t+c}, q_m) \right) \end{aligned}$$

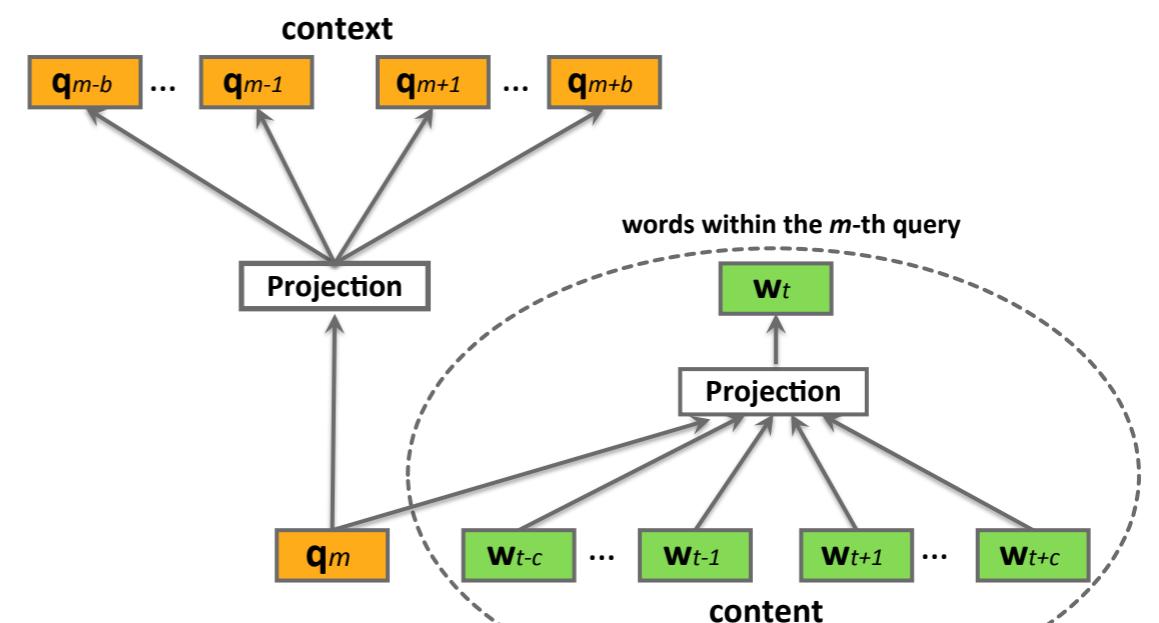


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$$\alpha_m = \frac{1}{\log(1+K_m)}$$



K_m is the frequency of the K-th actions

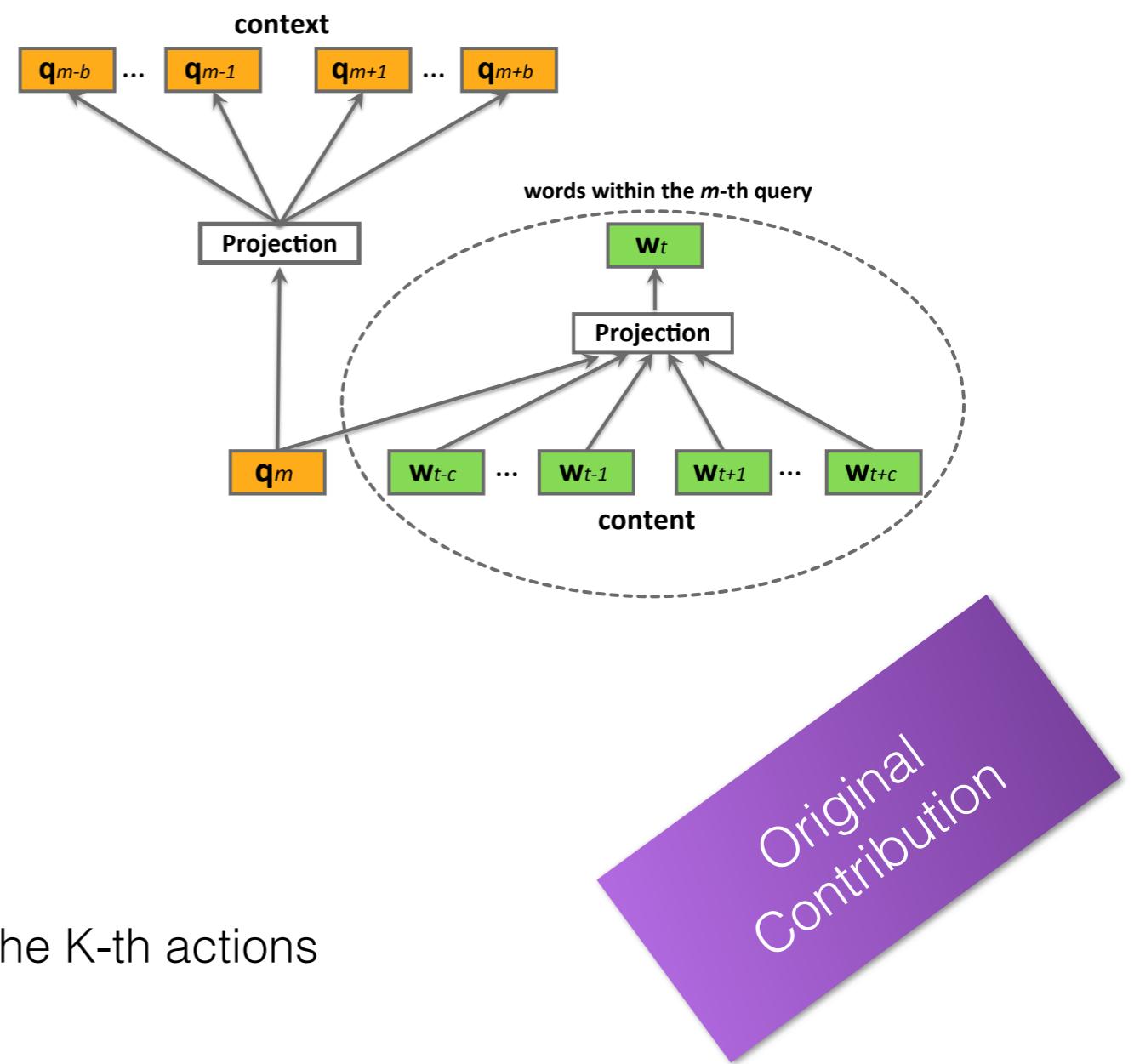
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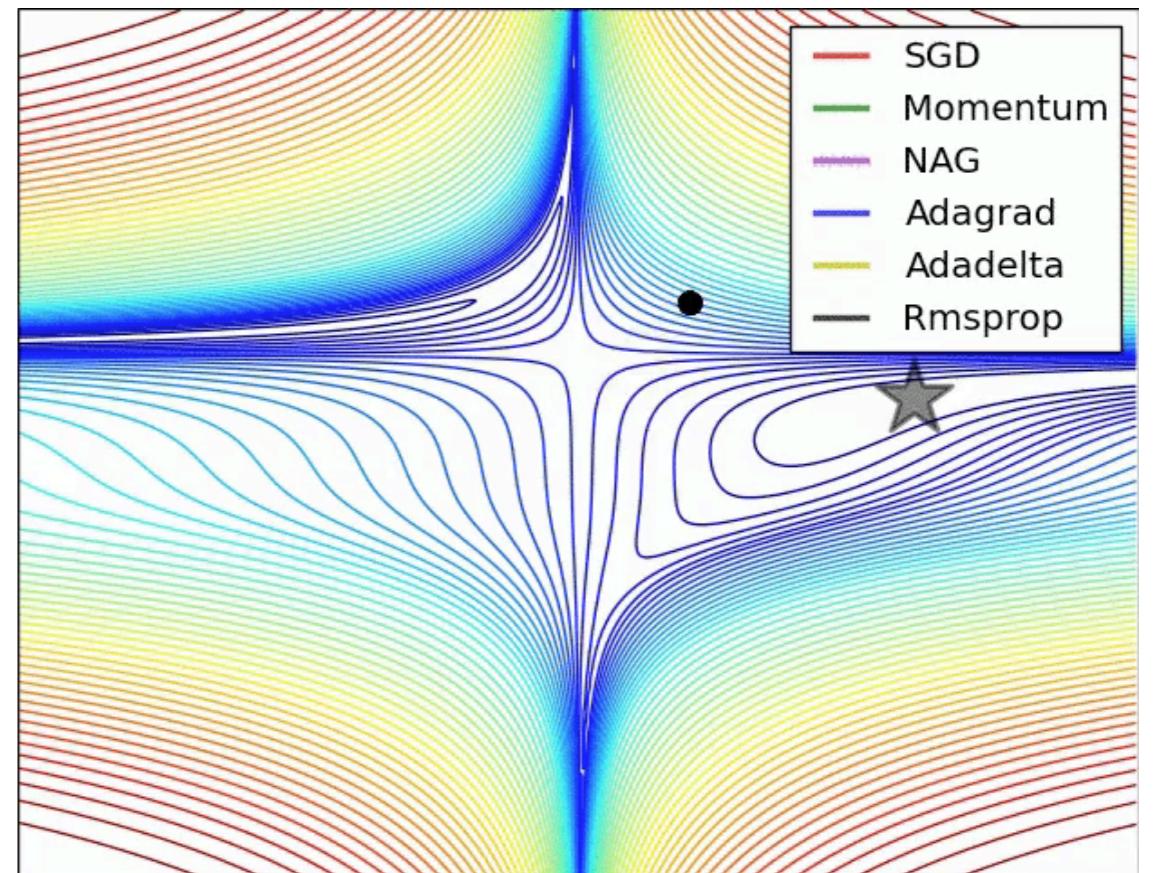
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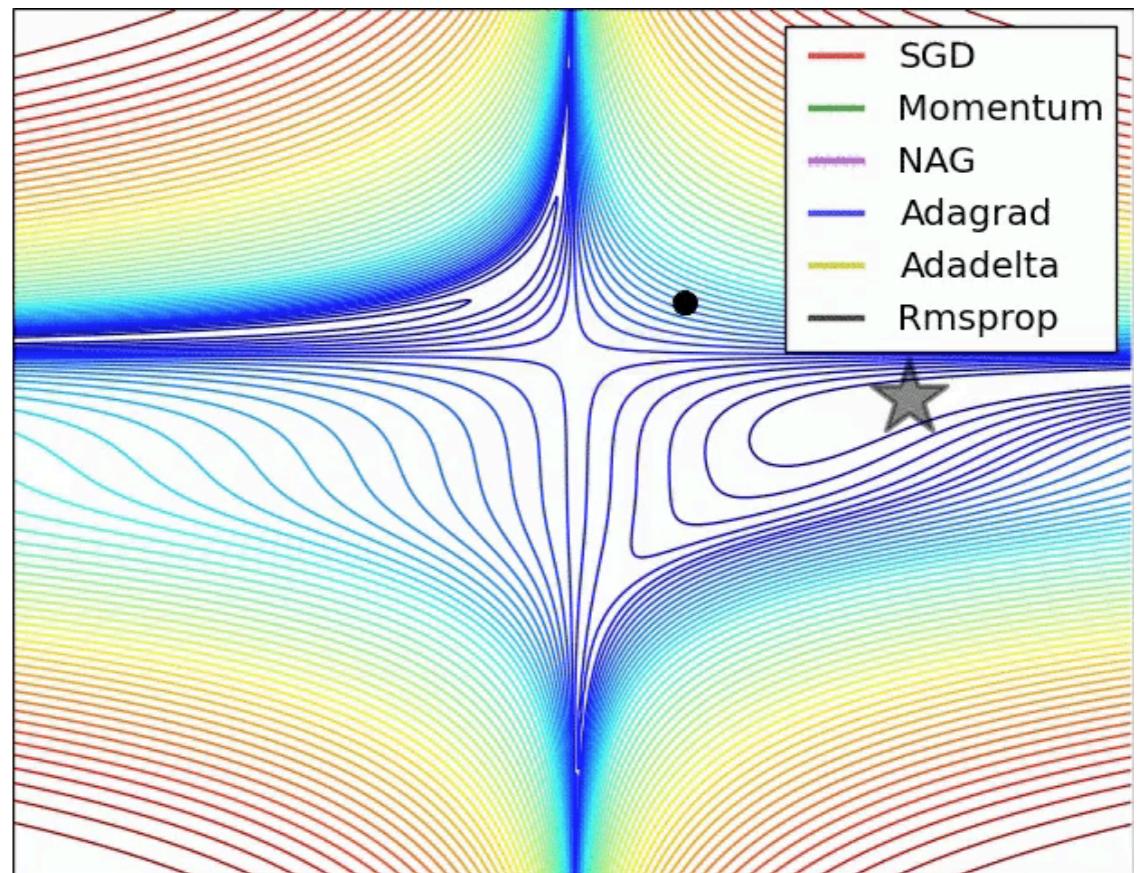
Optimization

- Stochastic Gradient Ascent
- Negative Sampling:
 - 10 Negative samples
 - Negatives chosen in order to deal with clicks and skips
- Dimension D = 300
- Context neighborhood = 5
- Content neighborhood = 7



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Experiments

- Session Data
 - *12 billion* sessions collected on the US website of Yahoo Search
- Query Content Data
 - from the *45 million* most frequent queries
- Click Data
 - For all the sessions we collected the clicks on algo and sponsored results
- Query Flow Graph (QFG) as baseline



Anecdotes

| Query | cx2vec | cx-cn2vec |
|--|---|--|
| minnesota insurance exam crossword puzzles | satellite tv otego satellite tv menominee satellite tv west end satellite tv townsend satellite tv lake sara | minnesota insurance minnesota insurance license practice exams online insurance exam crossword puzzles colorado insurance exam crossword puzzles online minnesota insurance exam crossword puzzles |
| microwave food safety | staphylococcal enteritis definition salmonella enteritis definition listeria monocytogenes prevention e coli cdc preventing cross contamination | microwave oven food safety microwave baby food safety microwave food safety studies microwave food safety issues foodsafety.com |
| what to cook in cast iron skillet | steak sauce substitute montreal seasoning ingredients ground turkey breakfast sausage how to cook steak on cast iron skillet reseason cast iron | cast iron skillet recipes how to cook with cast iron skillet how to cook in a cast iron skillet how to cook with a cast iron skillet chicken in cast iron skillet |
| iphone 6 repair services | mp3attic music donar ovulos en elche credit a la consommation rapide smart phone repair service social security disability bronx ny | at&t iphone repair service iphone 5c repair service iphone repair iphone service repair iphone repair services |

The Effect of Ad Clicks

| <u>makeup (no ads)</u> | <u>makeup (with ads)</u> |
|-------------------------------|---------------------------------|
| makeup tips | lipstick |
| fashion makeup | mac makeup |
| make up | makeup sets |
| makeup pictures | eye shadow |
| makeup images | makeup covergirl |
| makeup tutorial | makeup items |

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| makeup pictures | eye shadow |
| <u>snowboarding (no ads)</u> | <u>snowboarding (with ads)</u> |
| make | |
| snowbaording | snowboards |
| snow boardding | snowboarding gear |
| snowboarding information | burton snowboarding |
| snowboarding jumps | snowboard deals |
| snowboard pics | snowboards on sale |
| shaun white snowboarding | snowboarding mountains |

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| make | | |
| snowbaording | snowboards | |
| snow boarding | snowboarding gear | |
| snowboarding information | burton snowboarding | |
| snowboarding jump | <u>seafood (no ads)</u> | <u>seafood (with ads)</u> |
| snowboard pics | sea food | seafood restaurant |
| shaun white snowbo | crab legs | seafood restaurants |
| | best seafood | crab shack |
| | oysters | seafood market |
| | lobster recipes | sea food |
| | seafood market | sea food menu |

A Broader Comparison

| Original | QFG_{ad+link} | cn2vec | cx-cn2vec | cx-cn2vec_{ad} |
|---------------------------|--|--|---|---|
| wedding budget calculator | wedding budget wedding cost calculator wedding calculator wedding cost breakdown wedding budget worksheet | monthly budget calculator online budget calculator budget wedding budget calculator free average wedding budget | wedding planning checklist wedding budget template wedding budget worksheet wedding checklist printable wedding costs average | wedding budget worksheet wedding vendors the knot wedding planning checklist wedding wire |
| gmat prep classes | gmat prep gmat classes gmat kaplan gmat course kaplan gmat | gmat prep online gmat prep courses gmat online prep online gmat prep best gmat prep courses | gmat preparation courses sample gmat tests gmat prep class which is easier gre or gmat how much is the gmat | gmat study books gmat test prep classes gmat prep class kaplan gre courses free gmat sample tests |
| how to build a fence | building a fence build your own fence how to build a wood fence do it yourself fence how to build a privacy fence | how to build fence build fence build a fence how to build a cheap fence build your own fence | how to build a fence on a hill how to fence a yard how to build a fence gate how to build a brick wall fence how to build a fence video | how to build a fence minecraft fancy fences and gates how to build a metal fence home depot com fencing back yard fences |
| solar panels | solar panels for homes solar electric panels solar power ebay solar panels how to make solar panels | solar panels for your home solar panels for residential homes solar panels for solar panels on ebay davis solar | solar power solar energy solar panels for homes solar power systems solar panel | solar power solar panels for homes solar panels on sale solar panel kits solar panels for sale |

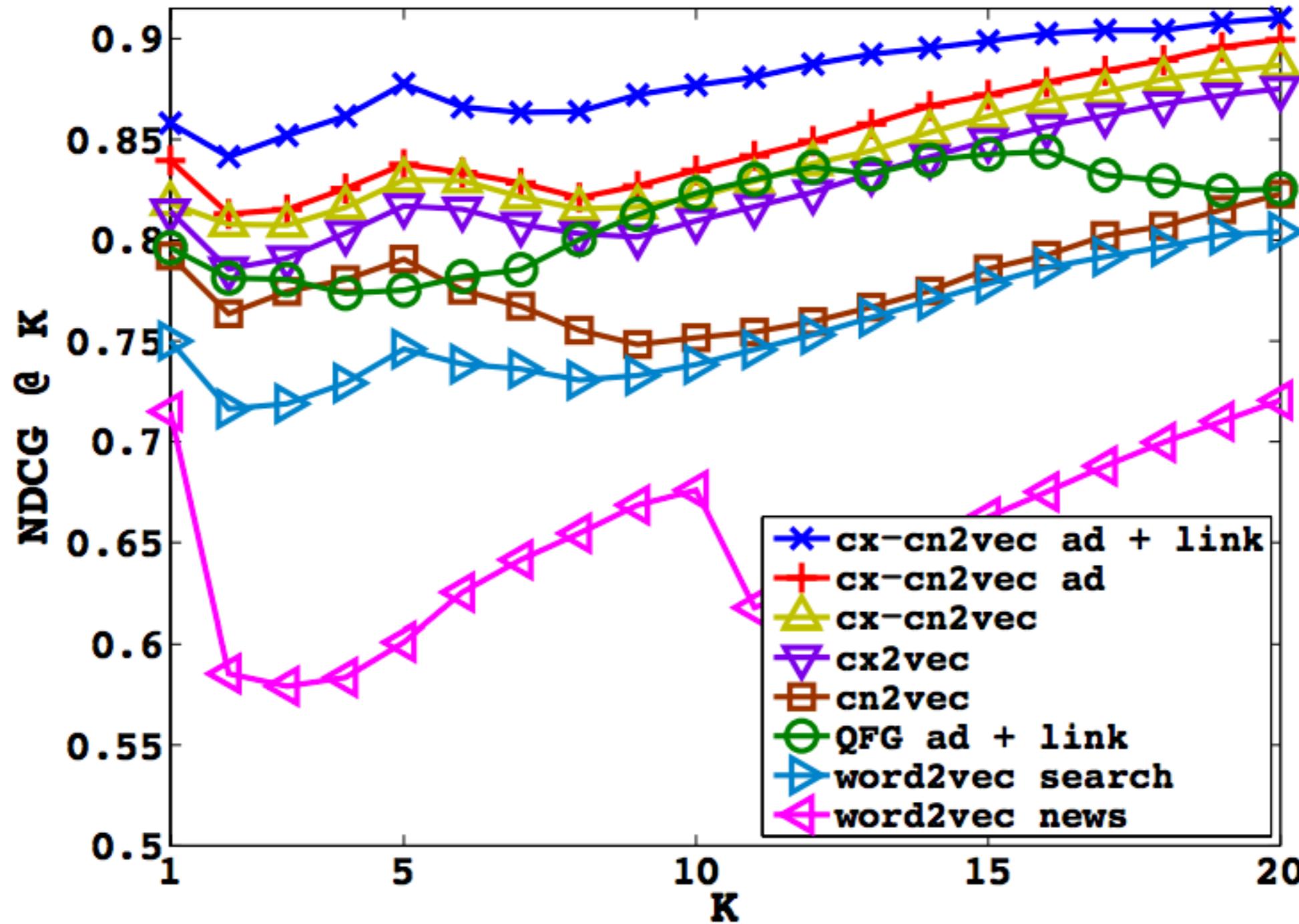
Editorial Evaluation

In-dictionary

| grade | pairs | cn2vec | cx2vec | cx-cn2vec | cx-cn2vec _{ad} | cx-cn2vec_{ad+link} | word2vec _{news} | word2vec _{search} | QFG _{ad+link} |
|------------------|--------|---------------|---------------|---------------|-------------------------|------------------------------------|--------------------------|----------------------------|------------------------|
| Excellent | 1,518 | 0.630 (0.136) | 0.658 (0.107) | 0.669 (0.107) | 0.668 (0.114) | 0.733 (0.094) | 0.818 (0.146) | 0.648 (0.151) | 0.329 (0.667) |
| Good | 5,531 | 0.599 (0.136) | 0.621 (0.125) | 0.637 (0.100) | 0.632 (0.104) | 0.683 (0.097) | 0.770 (0.152) | 0.614 (0.155) | 0.205 (0.574) |
| Fair | 4,021 | 0.550 (0.167) | 0.565 (0.129) | 0.577 (0.124) | 0.566 (0.130) | 0.605 (0.132) | 0.749 (0.190) | 0.567 (0.173) | 0.114 (0.366) |
| Bad | 4,229 | 0.398 (0.196) | 0.363 (0.170) | 0.349 (0.184) | 0.336 (0.187) | 0.425 (0.179) | 0.517 (0.280) | 0.395 (0.201) | 0.166 (0.584) |
| avg. p-value | - | 1.39e-15 | 5.44e-26 | 8.27e-28 | 2.99e-30 | 1e-100 | 4.121e-07 | 1.014e-14 | 0.013 |
| Excellent | 2,119 | 0.791 (0.166) | 0.623 (0.141) | 0.628 (0.134) | 0.623 (0.143) | 0.668 (0.147) | 0.824 (0.145) | 0.790 (0.157) | - |
| Good | 11,305 | 0.752 (0.155) | 0.587 (0.135) | 0.592 (0.130) | 0.584 (0.137) | 0.612 (0.141) | 0.796 (0.145) | 0.756 (0.156) | - |
| Fair | 11,146 | 0.715 (0.136) | 0.561 (0.145) | 0.565 (0.139) | 0.558 (0.146) | 0.584 (0.147) | 0.769 (0.175) | 0.707 (0.141) | - |
| Bad | 7,849 | 0.635 (0.199) | 0.383 (0.211) | 0.387 (0.209) | 0.382 (0.212) | 0.410 (0.208) | 0.509 (0.311) | 0.602 (0.208) | - |
| avg. p-value | - | 4.99e-26 | 1.137e-27 | 4.0423e-32 | 1.196e-30 | 2.926e-40 | 2.038e-16 | 9.388e-22 | - |

Out-of-dictionary

NDCG@K Values



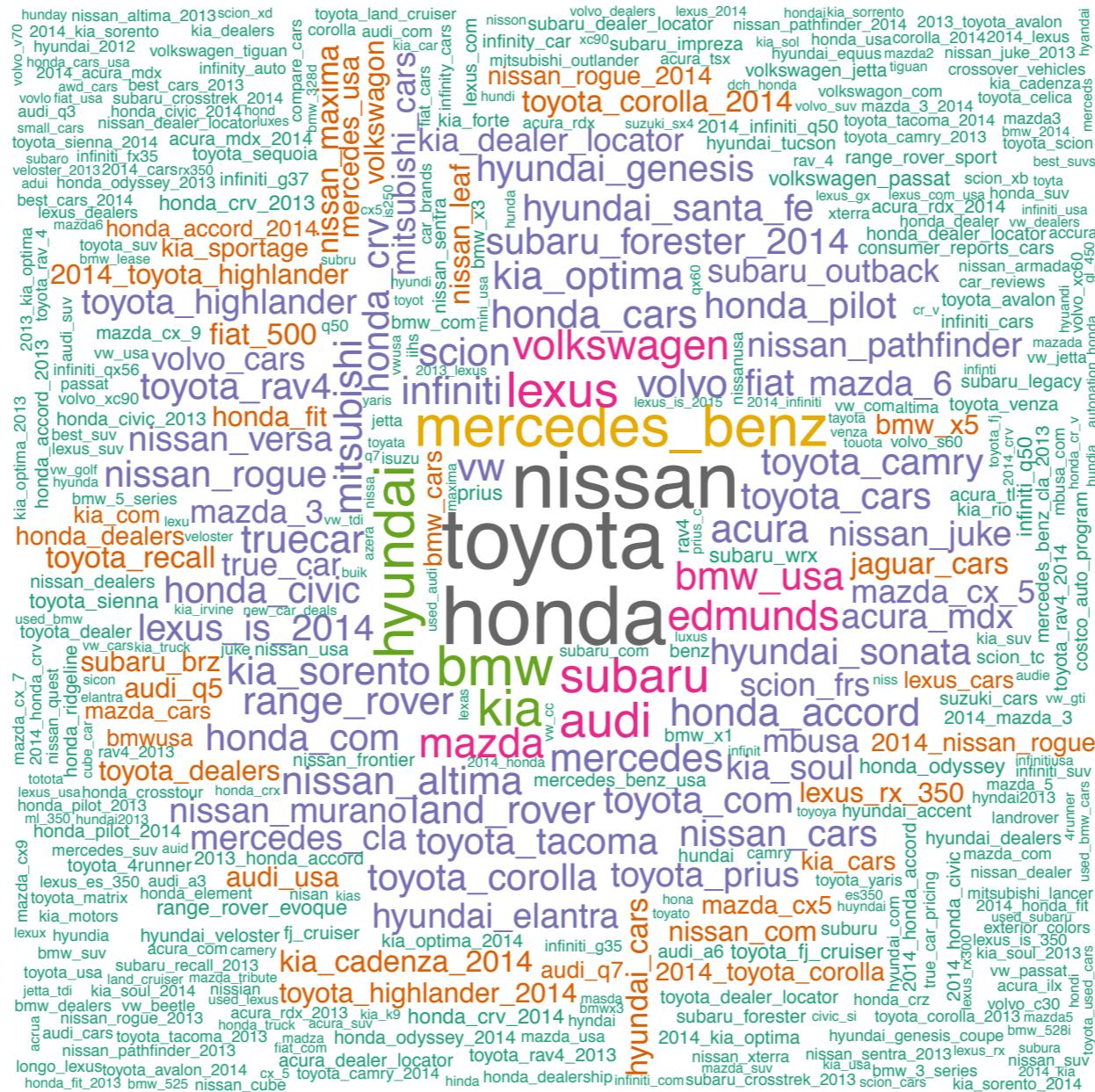
Experiments on TREC Data

| Method | Editorial grade | Levenshtein dist. |
|------------------------------------|-----------------|-------------------|
| QFG _{ad+link} | 1.0441 | 11.70 |
| word2vec _{news} | 0.9189 | 10.91 |
| word2vec _{search} | 0.9492 | 11.32 |
| cn2vec | 0.9571 | 11.37 |
| cx2vec | 1.1273 | 13.79 |
| cx-cn2vec | 1.1343 | 13.13 |
| cx-cn2vec _{ad} | 1.2281 | 13.62 |
| cx-cn2vec_{ad+link} | 1.2457 | 13.25 |

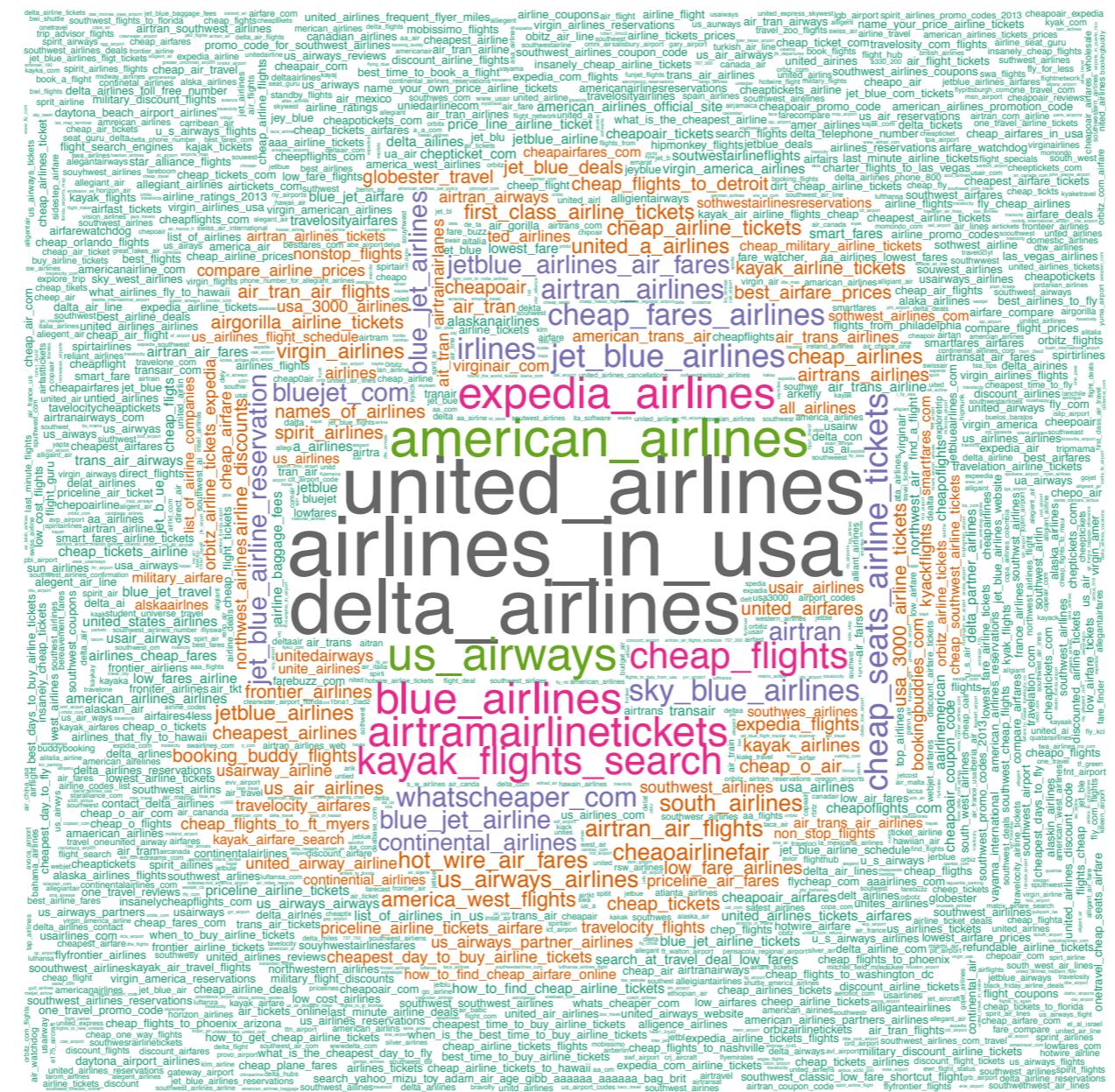
Coverage and Estimated Improvements

| Method | In-house data | | | TREC data | | |
|-------------------------------|---------------|-------------------|-------------|-------------|-------------------|-------------|
| | Coverage | Revenue potential | eCPM | Coverage | Revenue potential | eCPM |
| QFG _{ad+link} | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 |
| word2vec _{news} | 0.76 | 0.39 | 0.46 | 0.32 | 0.66 | 0.73 |
| word2vec _{search} | 0.87 | 0.58 | 0.77 | 0.57 | 0.84 | 0.75 |
| cn2vec | 0.89 | 0.62 | 0.84 | 0.59 | 0.84 | 0.74 |
| cx2vec | 1.16 | 1.80 | 1.41 | 1.41 | 1.16 | 1.20 |
| cx-cn2vec | 1.18 | 1.86 | 1.38 | 1.44 | 1.21 | 1.19 |
| cx-cn2vec_{ad} | 1.20 | 1.89 | 1.60 | 1.52 | 1.35 | 1.31 |
| cx-cn2vec _{ad+link} | 1.18 | 1.88 | 1.45 | 1.50 | 1.28 | 1.22 |

Queries similar to ads (ads category = 'cars')



Queries similar to ads
(ads category = 'air travel')



Conclusions and Future Work

- We presented the first query rewriting mechanism using word embeddings
 - We evaluated the proposed methods using both in-house and publicly available TREC data sets
 - When compared to the current state-of-the-art approaches, we showed that context-content2vec generates the most relevant query rewrites, while at the same time maintains high level of ad coverage
- Future:
 - Better sessionization algorithms
 - Navigational query detection to bias our method less towards generating those queries



Gracias

