

Embedding-based Query Spelling Correction

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Gustav Lahmann¹

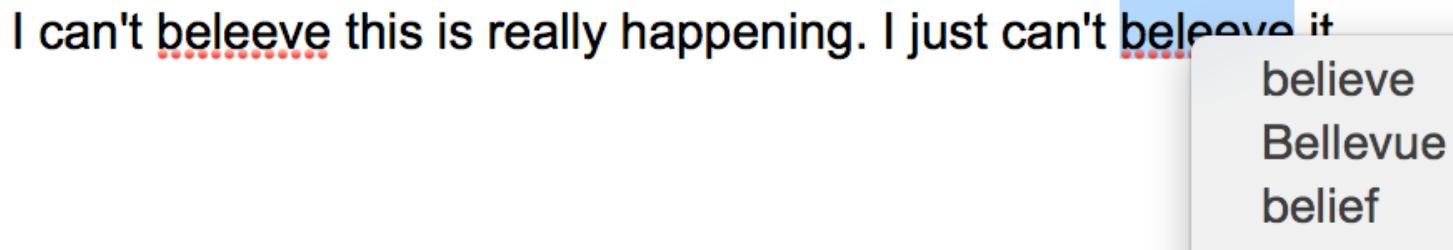
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webis.de

Query Spelling Correction



[<https://i.stack.imgur.com/DCUcD.png>]

Query Spelling Correction

Type	Misspelling	Correction
Deletion	corosion	→ corrosion
Insertion	occurr	→ occur
Space	abouta	→ about a
Special character	isnt	→ isn't
Substitution	grammer	→ grammar
Transposition	rewriet	→ rewrite

[Hagen et al. 2017, SIGIR]

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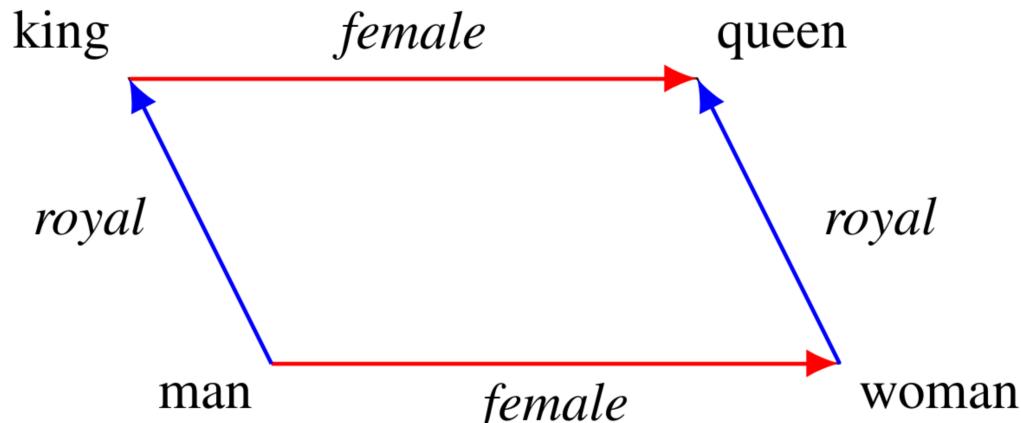
- real-word errors (e.g.: their instead of there)
- non-word errors (e.g.: grammer)

[Hagen et al. 2017, SIGIR]

Embeddings

- Blog post: embedding-based correction [Rushton, 2018]

<https://edrushton.medium.com/a-simple-spell-checker-built-from-word-vectors-9f28452b6f26>

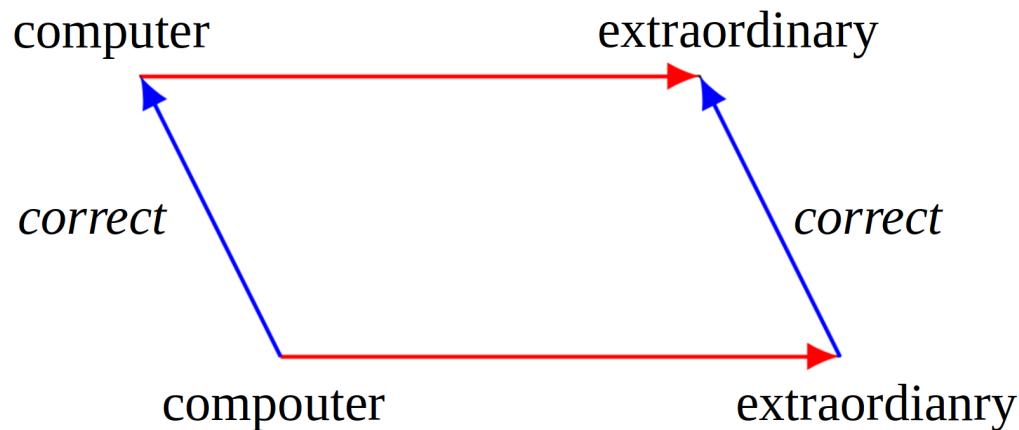


$$\overrightarrow{\text{king}} - \overrightarrow{\text{man}} + \overrightarrow{\text{woman}} \approx \overrightarrow{\text{queen}}$$

[Ethayarajh et al. 2019, ACL]

Query Spelling Correction with Embeddings

- Misspelling property encoded in embeddings



$$\overrightarrow{\text{computer}} - \overrightarrow{\text{compouter}} + \overrightarrow{\text{extraordianry}} \approx \overrightarrow{\text{extraordinary}}$$

[Rushton, 2018]

Query Spelling Correction with Embeddings

- 300-dim GloVe embeddings nlp.stanford.edu/projects/glove/
- Compute correction vector $\overrightarrow{v_{correct}}$ (Oxford dict of common misspellings)
- Add $\overrightarrow{v_{correct}}$ to each GloVe embedding
- Save nearest neighbor
- Look up correction for each query word

Evaluation – Retrieval Effectiveness

None

Rushton

GPT-3.5

Hunspell

pyspellchecker

Evaluation – Retrieval Effectiveness

	ANTIQUE	TREC DL '19	TREC DL '20
Spell Correction	BM25 PL2	BM25 PL2	BM25 PL2
None			
Rushton			
GPT-3.5			
Hunspell			
pyspellchecker			

nDCG@10

Evaluation – Retrieval Effectiveness

Spell Correction	ANTIQUE		TREC DL '19		TREC DL '20	
	BM25	PL2	BM25	PL2	BM25	PL2
None	0.51	0.49	0.48	0.47	0.49	0.48
Rushton	0.51	0.49	0.48	0.47	0.49	0.48
GPT-3.5						
Hunspell						
pyspellchecker						

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Evaluation – Retrieval Effectiveness

Spell Correction	ANTIQUE		TREC DL '19		TREC DL '20	
	BM25	PL2	BM25	PL2	BM25	PL2
None	0.51	0.49	0.48	0.47	0.49	0.48
Rushton	0.51	0.49	0.48	0.47	0.49	0.48
GPT-3.5	0.49	0.47	0.48	0.47	0.48	0.47
Hunspell	0.48	0.46	0.36	0.34	0.42	0.41
pyspellchecker	0.43	0.41	0.31	0.30	0.33	0.32

nDCG@10

Evaluation – Corrections

Error Type	ANTIQUE				
	#err.	corrected			
	R	H	P	G	
Deletion	14	4	6	5	12
Insertion	5	1	1	1	4
Space	13	1	1	2	11
Special char.	19	-	2	6	9
Substitution	8	1	2	2	8
Transposition	4	2	2	1	4
Incorr. changes	0	52	75	42	

R – Rushton

H – Hunspell

P – pyspellchecker

G – GPT3.5

Evaluation – Corrections Examples

		ANTIQUE					
		#err.	corrected	R	H	P	G
Error Type							
Deletion		14	4	6	5	12	
Insertion		5	1	1	1	4	
Space		13	1	1	2	11	
Special char.		19	-	2	6	9	
Substitution		8	1	2	2	8	
Transposition		4	2	2	1	4	
Incorr. changes		0	52	75	42		

Original query:

why is gordon **ramsey** so popular

Rushton: —

Hunspell: —

pyspellchecker:

why is **cordon raise** so popular

GPT-3.5:

why is gordon **ramsay** so popular?

Evaluation – Corrections Examples

		ANTIQUE					
		#err.	corrected	R	H	P	G
Error Type							
Deletion		14	4	6	5	12	
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GPT-3.5:

why is gordon **ramsay** so popular?

Limitation Rushton: GloVe vocabulary

Evaluation – Corrections Examples

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Error Type		R	H	P	G
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Original query:

[...] copies of letters of
commendation **prsent**ed [...]

Rushton, Hunspell, pyspell., GPT-3.5:

[...] copies of letters of
commendation **present**ed [...]

Evaluation – Corrections Examples

Error Type	ANTIQUE				
	#err.	corrected	R	H	P
Deletion	14	4	6	5	12
Insertion	5	1	1	1	4
Space	13	1	1	2	11
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Original query:

how can i keep my **rabit** indoors

Rushton:

how can i keep my **rabbit** indoors

Hunspell:

how can i keep my **habit** indoors

pyspellchecker:

how can i keep my **barit** indoors

GPT-3.5:

how can i keep my **rabbit** indoors

Evaluation – Corrections Examples

ANTIQUE					
	#err.	corrected			
Error Type		R	H	P	G
Deletion	14	4	6	5	12
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Special char.	19	-	2	6	9
Substitution	8	1	2	2	8
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Incorr. changes	0	52	75	42	

Original query:

what is wifi vs bluetooth

Rushton, GPT-3.5: —

Hunspell, pyspellchecker:

what is **wife** vs bluetooth

Evaluation – Corrections Examples

Error Type	ANTIQUE				
	#err.	corrected	R	H	P
Deletion	14	4	6	5	12
Insertion	5	1	1	1	4
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Special char.	19	-	2	6	9
Substitution	8	1	2	2	8
Transposition	4	2	2	1	4
Incorr. changes	0	52	75	42	

Original query:

why hot air raise up

GPT-3.5:

why does hot air raise up?

Original query:

is cdg airport in main paris

GPT-3.5:

[...] corrected search query:
is cdg airport in main paris?

Evaluation – Corrections Examples

Error Type	ANTIQUE				
	#err.	corrected	R	H	P
Deletion	14	4	6	5	12
Insertion	5	1	1	1	4
Space	13	1	1	2	11
Special char.	19	-	2	6	9
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Incorr. changes	0	52	75	42	

Original query:

how can we get concentration
on something

Rushton: —

Hunspell:

how can we get concentration
something

pyspellchecker:

how can we get concentration

GPT-3.5:

how can we get concentration
on something?

Limitation Rushton: Spaces, real-word errors

Evaluation – Corrections Examples

Error Type	ANTIQUE					TREC DL '19					TREC DL '20				
	#err.	corrected				#err.	corrected				#err.	corrected			
	R	H	P	G		R	H	P	G		R	H	P	G	
Deletion	14	4	6	5	12	-	-	-	-	-	-	-	-	-	-
Insertion	5	1	1	1	4	1	-	1	1	1	-	-	-	-	-
Space	13	1	1	2	11	-	-	-	-	-	-	-	-	-	-
Special char.	19	-	2	6	9	1	-	-	-	-	3	-	-	-	-
Substitution	8	1	2	2	8	1	-	1	1	1	1	-	-	-	-
Transposition	4	2	2	1	4	-	-	-	-	-	-	-	-	-	-
Incorr. changes	0	52	75	42		1	12	15	5		0	11	23	15	

Wrap-Up

- ❑ Considered spell correctors worsen retrieval results (on average)
- ❑ Embedding-based approach better than common spell correctors

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Limitations

- ❑ GloVe vocabulary
- ❑ Single words
- ❑ No context

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Future Work

- ❑ Other correction approaches
- ❑ Other embeddings
- ❑ Create misspelling query variants