

CS4051

Information Retrieval

Week 03

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February 07, 2023

Spelling & Phonetic  
Corrections

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## Spelling Corrections

- Two principal uses
  - Correcting document(s) being indexed
  - Correcting user queries to retrieve “right” answers
- Two main flavors:
  - Isolated word
    - Check each word on its own for misspelling
    - Will not catch typos resulting in correctly spelled words
    - e.g., **from** → **form**
  - Context-sensitive
    - Look at surrounding words,
    - e.g., **I flew form Heathrow to Narita.**

## Spelling Corrections

[Return to Google's jobs pages](#)

488941 britney spears	29 britent spears	9 brinttany spears	5 brney spears	3 britly spears	2 brirreny spears
40134 brittany spears	29 britttany spears	9 britanay spears	5 broitney spears	3 britmeny spears	2 brittany spears
36315 brittney spears	29 britttany spears	9 brittany spears	5 brotney spears	3 britneey spears	2 britttany spears
24342 brittany spears	29 btiney spears	9 brits spears	5 bruteny spears	3 britnehy spears	2 britttney spears
7331 britny spears	26 brittney spears	9 britnew spears	5 btiuey spears	3 britnely spears	2 brittain spears
6633 briteny spears	26 breitney spears	9 britneyn spears	5 brittney spears	3 britney spears	2 britane spears
2696 brittney spears	26 britnity spears	9 brittney spears	5 gritney spears	3 britnetty spears	2 britaneny spears
1807 briney spears	26 britteny spears	9 brtiny spears	5 spritney spears	3 britnex spears	2 britania spears
1655 britny spears	26 brittneyt spears	9 britttney spears	4 bitny spears	3 britneyxax spears	2 britann spears
1479 brintey spears	26 brittan spears	9 brtny spears	4 brritney spears	3 britnity spears	2 britanna spears
1479 britanny spears	26 brittne spears	9 brytny spears	4 brandy spears	3 britntey spears	2 britannie spears
1338 britiny spears	26 britttany spears	9 rbitney spears	4 brritney spears	3 britnyey spears	2 britannt spears
1211 britnet spears	24 beitney spears	8 birtiny spears	4 brraitny spears	3 britttany spears	2 britannu spears
1096 brittney spears	24 birtney spears	8 birtney spears	4 brraitney spears	3 britttney spears	2 britanny spears
991 britaney spears	24 brightiny spears	8 brattany spears	4 brritney spears	3 britttney spears	2 britanny spears
991 britnay spears	24 brintiny spears	8 braitny spears	4 brritney spears	3 britttney spears	2 britanny spears
811 brittney spears	24 brittany spears	8 brritney spears	4 brraitny spears	3 brittyn spears	2 brittany spears
811 britney spears	24 brittney spears	8 brightny spears	4 brritney spears	3 brritney spears	2 brittney spears
664 britney spears	24 brittini spears	8 brintay spears	4 brritney spears	3 britney spears	2 brittney spears
664 brittney spears	24 brittney spears	8 brritney spears	4 brritney spears	3 brotney spears	2 brittney spears
664 britney spears	24 brittini spears	8 brritney spears	4 brritney spears	3 brtney spears	2 brittney spears
601 britney spears	24 brittne spears	8 brittany spears	4 brritney spears	3 brittany spears	2 brittany spears
601 britny spears	21 brritney spears	8 brittley spears	4 brritney spears	3 brittney spears	2 brittany spears
544 brittany spears	21 brritany spears	8 brritney spears	4 brritney spears	3 brittney spears	2 brittany spears
544 brittany spears	21 brritany spears	8 brritney spears	4 brritney spears	3 brittney spears	2 brittany spears
364 britney spears	21 brittney spears	8 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
364 brittney spears	21 brittany spears	8 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
329 britney spears	21 brittany spears	8 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
269 britney spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
269 brittney spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
244 brittne spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
244 brittney spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
220 brittany spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
220 brittany spears	21 brittany spears	7 brritney spears	4 brittney spears	3 brittney spears	2 brittany spears
199 brittney spears	19 brittney spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
163 brittney spears	19 brittney spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
147 brittney spears	19 brittney spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
147 brittney spears	19 brittney spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
147 brittney spears	19 brittany spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears
147 brittney spears	19 brittany spears	7 brittney spears	4 brittney spears	3 brittney spears	2 brittany spears

## Document Correction

- Especially needed for OCR'ed documents
    - Correction algorithms are tuned for this: “rn” / ”m”
    - Can use domain-specific knowledge
      - E.g., OCR can confuse O and D more often than it would confuse O and I (adjacent on the QWERTY keyboard, so more likely interchanged in typing).
  - But also: web pages and even printed material has typos
  - Goal: the dictionary contains fewer misspellings
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## Isolated Word Correction

- Fundamental premise – there is a lexicon from which the correct spellings come
  - Two basic choices for this
    - A standard lexicon such as
      - Webster's English Dictionary
      - An “industry-specific” lexicon – hand-maintained
    - The lexicon of the indexed corpus
      - E.g., all words on the web
      - All names, acronyms etc.
      - (Including the mis-spellings)
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## Isolated Word Correction

- Given a lexicon and a character sequence  $Q$ , return the words in the lexicon closest to  $Q$
  - What's "closest"?
  - We'll study several alternatives
    - Edit distance (Levenshtein distance)
    - Weighted edit distance
    - $n$ -gram overlap
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## Edit Distance

- Given two strings  $S_1$  and  $S_2$ , the minimum number of operations to convert one to the other
  - Operations are typically character-level
    - Insert, Delete, Replace, (Transposition)
  - E.g., the edit distance from **dof** to **dog** is 1
    - From **cat** to **act** is 2 (Just 1 with transpose.)
    - from **cat** to **dog** is 3.
-

## Edit Distance – Levenshtein

```

EDITDISTANCE( $s_1, s_2$ )
1   $int\ m[i, j] = 0$ 
2  for  $i \leftarrow 1$  to  $|s_1|$ 
3  do  $m[i, 0] = i$ 
4  for  $j \leftarrow 1$  to  $|s_2|$ 
5  do  $m[0, j] = j$ 
6  for  $i \leftarrow 1$  to  $|s_1|$ 
7  do for  $j \leftarrow 1$  to  $|s_2|$ 
8      do  $m[i, j] = \min\{m[i-1, j-1] + \text{if } (s_1[i] = s_2[j]) \text{ then } 0 \text{ else } 1, \\$ 
9           $m[i-1, j] + 1, \\$ 
10          $m[i, j-1] + 1\}$ 
11 return  $m[|s_1|, |s_2|]$ 

```

► **Figure 3.5** Dynamic programming algorithm for computing the edit distance between strings  $s_1$  and  $s_2$ .

## Edit Distance – Levenshtein

		f	a	s	t
	0	1 1	2 2	3 3	4 4
c	1 1	1 2 2 1	2 3 2 2	3 4 3 3	4 5 4 4
a	2 2	2 2 3 2	1 3 3 1	3 4 2 2	4 5 3 3
t	3 3	3 3 4 3	3 2 4 2	2 3 3 2	2 4 3 2
s	4 4	4 4 5 4	4 3 5 3	2 3 4 2	3 3 3 3

► **Figure 3.6** Example Levenshtein distance computation. The  $2 \times 2$  cell in the  $[i, j]$  entry of the table shows the three numbers whose minimum yields the fourth. The cells in *italics* determine the edit distance in this example.

## Using Edit Distance

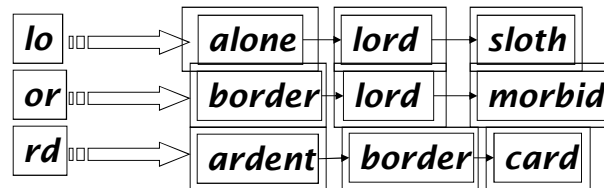
- Given query, first enumerate all character sequences within a preset (weighted) edit distance (e.g., 2)
  - Intersect this set with list of “correct” words
  - Show terms you found to user as suggestions
  - Alternatively,
    - We can look up all possible corrections in our inverted index and return all docs ... slow
    - We can run with a single most likely correction
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## n-gram Overlaps

- Enumerate all the  $n$ -grams in the query string as well as in the lexicon
  - Use the  $n$ -gram index (recall wild-card search) to retrieve all lexicon terms matching any of the query  $n$ -grams
  - Threshold by number of matching  $n$ -grams
    - Variants – weight by keyboard layout, etc.
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## 2-grams for match

- Consider the query **lord** – we wish to identify words matching 2 of its 3 bigrams (**lo**, **or**, **rd**)



Standard postings “merge” will enumerate ...

## Context-Sensitive Spelling Corrections

- Text: ***I flew from Heathrow to Narita.***
- Consider the phrase query “***flew form Heathrow***”
- We’d like to respond  
Did you mean “***flew from Heathrow***”?  
because no docs matched the query phrase.

## Context-Sensitive Spelling Corrections

- Need surrounding context to catch this.
  - First idea: retrieve dictionary terms close (in weighted edit distance) to each query term
  - Now try all possible resulting phrases with one word “fixed” at a time
    - *flew from heathrow*
    - *fled form heathrow*
    - *flea form heathrow*
  - **Hit-based spelling correction:** Suggest the alternative that has lots of hits.
- 

## Issues in Spelling Corrections

- We enumerate multiple alternatives for “Did you mean?”
  - Need to figure out which to present to the user
  - Use heuristics
    - The alternative hitting most docs
    - Query log analysis + tweaking
      - For especially popular, topical queries
  - Spell-correction is computationally expensive
    - Avoid running routinely on every query?
    - Run only on queries that matched few docs
-



## Soundex

- Class of heuristics to expand a query into phonetic equivalents
    - Language specific – mainly for names
    - E.g., ***chebyshev*** → ***tchebycheff***
  - Invented for the U.S. census ... in 1918
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## Soundex Algorithm

1. Retain the first letter of the word.
  2. Change all occurrences of the following letters to '0' (zero):  
'A', 'E', 'I', 'O', 'U', 'H', 'W', 'Y'.
  3. Change letters to digits as follows:
    - B, F, P, V → 1
    - C, G, J, K, Q, S, X, Z → 2
    - D, T → 3
    - L → 4
    - M, N → 5
    - R → 6
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## Soundex Algorithm

4. Remove all pairs of consecutive digits.
5. Remove all zeros from the resulting string.
6. Pad the resulting string with trailing zeros and return the first four positions, which will be of the form <uppercase letter> <digit> <digit> <digit>.

E.g., **Herman** becomes H655.

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## Soundex

- Soundex is the classic algorithm, provided by most databases (Oracle, Microsoft, ...)
  - How useful is soundex?
    - Not very – for information retrieval
  - Zobel and Dart (1996) show that other algorithms for phonetic matching perform much better in the context of IR
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## Soundex Exercise

- Find two differently spelled proper nouns (different to the course example) whose soundex codes are the same and give their soundex code.
    - Mary, Nira (Soundex code = 5600).
  - Find two phonetically similar proper nouns whose soundex codes are different.
    - Chebyshev, Tchebycheff
    - Rafi, Rafee
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