How to Write a Spelling Corrector

Dongqing Zhu

Based on Peter Novig's Online Tutorial http://norvig.com/spell-correct.html

Introduction

- Peter Novig's Online Tutorial
 - Exellent material for HW3
 - Simple probability methods
 - 21 lines of python code
 - ~ 6.5 MB training data
 - ~ 70% correctness on two test datasets

Outline

- Problem Formulation
- Solving the Problem
- Evaluation
- Improvement
- Useful Links

Problem Formulation

- Problem Statement
 - Find the most likely spelling correction c for a given word w among all possible corrections

- Probability Theory
 - argmax_c P(c|w)
 - => argmax_c P(w|c) P(c) / P(w) -- by Bayes' Theorem
 - $=> \operatorname{argmax}_{c} P(w|c) P(c)$ -- by ignoring P(w)

Solving the Problem

- argmax_c P(w|c) P(c)
 - argmaxc
 - cover all possible corrections ideally
 - P(c)
 - Language Model
 - capturing how likely c would stand on its own
 - P(w|c)
 - Error model
 - capturing how like the change c => w will happen

argmax_c

- Enumerate possible corrections by edit distance
 - words of edit distance 1
 cover 80% to 95% of spelling errors

 words of edit distance 2 cover 98.9% of spelling errors

argmax_c

Get possible corrections by edit distance

```
alphabet = 'abcdefghijklmnopqrstuvwxyz'
    def edits1(word):
       splits = [(word[:i], word[i:]) for i in range(len(word) + 1)]
       deletes = [a + b[1:]  for a, b in splits if b]
       transposes = [a + b[1] + b[0] + b[2:] for a, b in splits if len(b)>1
      replaces = [a + c + b[1:] for a, b in splits for c in alphabet if b]
8
       inserts = [a + c + b] for a, b in splits for c in alphabet]
       return set(deletes + transposes + replaces + inserts)
10
    \# n dels, n-1 trans, 26n reps, and 26(n+1) ins, totally 54n+25
    def known edits2(word):
13
        return set(e2 for e1 in edits1(word) for e2 in edits1(e1) if e2 in
    NWORDS)
    # only consider those are known in the training set
```

Solving the Problem

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 - capturing how like the change **c** => **w** will happen

Building Language Model P(c)

- Training data
 - counts of each word in the training data
 - smoothing for novel words

Building Language Model P(c)

Train the languge model

```
def words(text): return re.findall('[a-z]+', text.lower())
    # return a long list of words

def train(features):
    model = collections.defaultdict(lambda: 1) # smoothing
    for f in features:
        model[f] += 1 # counting word occurences
    return model

NWORDS = train(words(file('big.txt').read()))
# NWORDS stores word-count pairs
```

Solving the Problem

- argmax_c P(w|c) P(c)
 - argmaxc
 - cover all possible corrections ideally
 - P(c)
 - Language Model
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Building Error Model P(w|c)

- Training data
- Trivial model

P(any known edit0 word) >> P(any known edit1 word) >> P(any known edit2 word)

```
def known(words): return set(w for w in words if w in NWORDS)

def correct(word):
    candidates = known([word]) or known(edits1(word)) or known_edits2(word) or [word]
    return max(candidates, key=lambda w: NWORDS[w])
```

Solving the Problem

- argmax_c P(w|c) P(c)
 - argmaxc
 - cover all possible corrections ideally
 - P(c)
 - Language Model
 - capturing how likely c would stand on its own
 - P(w|c)
 - Error model
 - capturing how like the change c => w will happen

Evaluation

```
>>> correct('speling')
    'spelling'
    >>> correct('korrecter')
    'corrector'
```

- Testing data
 - Roger Mitton's Birkbeck spelling error corpus
 - ~70% correctness on two datasets

- Places can we improve
 - argmax_c P(w|c) P(c)
 - argmaxc
 - P(c)
 - -P(w|c)

- Improving coverage of possible corrections (argmax_c)
 - words beyond edit distance 2
 - Examples:
 - successful sucssuful
 - hierarchy heiarky
 - profession preffeson
 - allowing a limited set of edits at edit distance 3

- Improving the language model P(c)
 - deal with unknown words
 - bad case1: correct('economtric') => 'economic' (121); expected 'econometric' (1)
 - bad case 2: correct('generataed') => 'generate' (2);
 expected 'generated' (1)
 - Possible solutions
 - more training data
 - add different forms of a word, e.g. -s to a noun, -ed to a verb
 - sequences of characters, e.g. suffix "-ally"

- Improving the error model P(w|c)
 - P(any known edit_0 word) >> P(any known edit1 word) >> P(any known edit2 word)
 - bad cases:
 - correct('reciet') => 'recite' (5); expected 'receipt' (14)
 - correct('thay') => 'that' (12513); expected 'they' (4939)
 - correct('wonted') => 'wonted' (2); expected 'wanted' (214)
 - possible solution
 - alternative similarity metrics
 - a corpus of spelling errors

- Beyond the three major factors
 - Context
 - examples
 - correct('carrers') => 'carriers' (7); expected 'careers' (2)
 - correct('quies') => 'quiet' (119); expected 'queries' (1)
 - Google n-gram will be very useful

- Beyond the three major factors
 - Fixing errors in testing data
 - examples
 - correct('sumarys') => 'summary' (17); expected 'summarys' (1)
 - correct('humor') => 'humor' (17); expected 'humour' (5)
 - Improve response time
 - results caching

Useful Links

• Original tutorial "How to write a spelling corrector":

http://norvig.com/spell-correct.html

Other versions:

Perl: http://www.riffraff.info/2007/5/20/a-spell-corrector-in-perl6-part-3

Java: http://raelcunha.com/spell-correct.php

C:http://scarvenger.wordpress.com/2007/12/11/how-to-write-a-spelling-corrector/

aspell project:

http://aspell.net/test/ (more testing data)

Toolkit: LingPipe:

http://alias-i.com/lingpipe/demos/tutorial/querySpellChecker/read-me.html