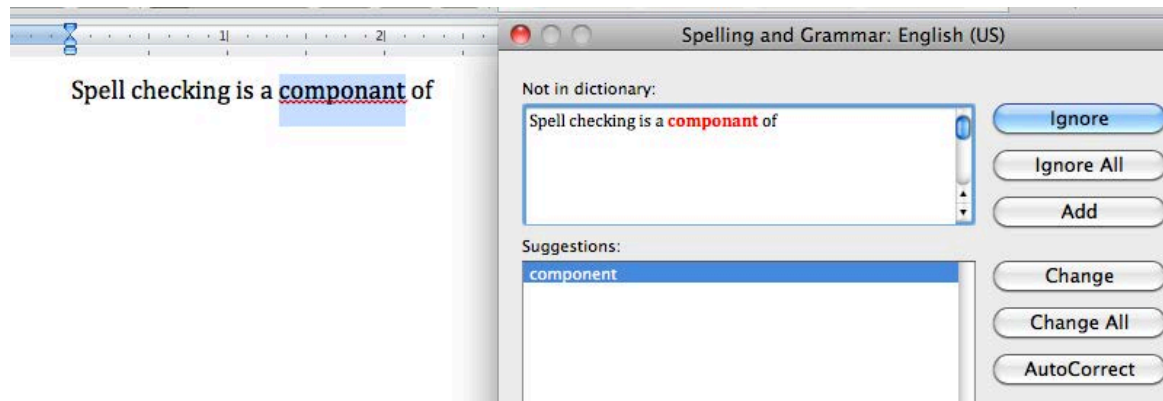


Applications for spelling correction

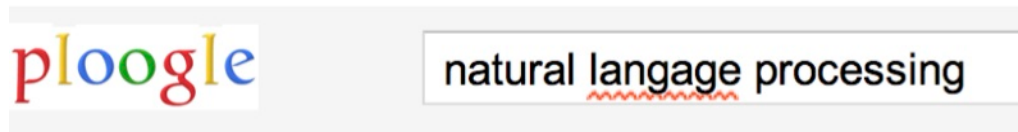
Word processing



Phones



Web search



Showing results for [natural language processing](#)
Search instead for [natural language processing](#)

Spelling Tasks

- Spelling Error Detection
- Spelling Error Correction:
 - Autocorrect
 - hte→the
 - Suggest a correction
 - Suggestion lists

Types of spelling errors

- Non-word Errors
 - *graffe* → *giraffe*
- Real-word Errors
 - Typographical errors
 - *three* → *there*
 - Cognitive Errors (homophones)
 - *piece* → *peace*,
 - *too* → *two*

Rates of spelling errors

26%: Web queries [Wang et al. 2003](#)

13%: Retyping, no backspace: [Whitelaw et al. English&German](#)

7%: Words corrected retyping on phone-sized organizer

2%: Words uncorrected on organizer [Soukoreff & MacKenzie 2003](#)

1-2%: Retyping: [Kane and Wobbrock 2007](#), [Gruden et al. 1983](#)

Non-word spelling errors

- Non-word spelling error detection:
 - Any word not in a ***dictionary*** is an error
 - The larger the dictionary the better
- Non-word spelling error correction:
 - Generate ***candidates***: real words that are similar to error
 - Choose the one which is best:
 - Shortest weighted edit distance
 - Highest noisy channel probability

Real word spelling errors

- For each word w , generate candidate set:
 - Find candidate words with similar ***pronunciations***
 - Find candidate words with similar ***spelling***
 - Include w in candidate set
- Choose best candidate
 - Noisy Channel
 - Classifier

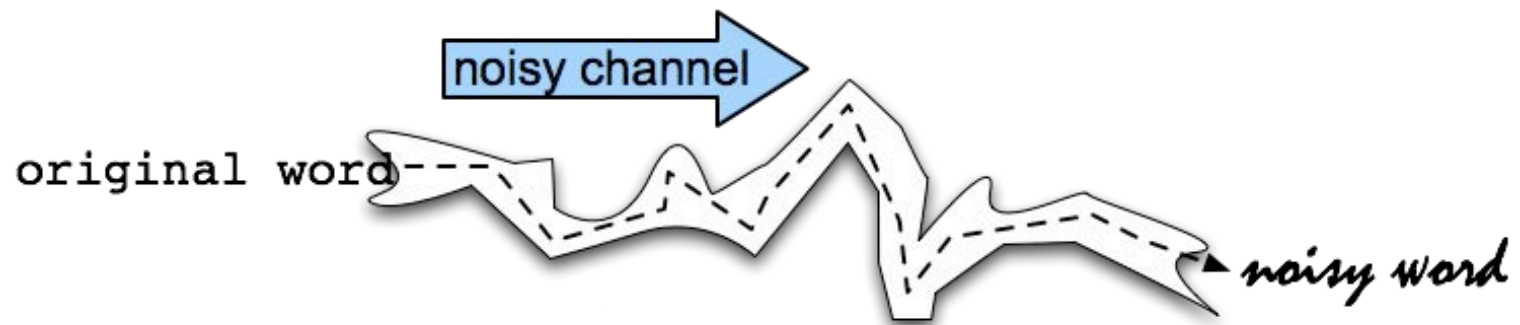
Spelling Correction and the Noisy Channel

The Spelling Correction Task

Spelling Correction and the Noisy Channel

The Noisy Channel Model of Spelling

Noisy Channel Intuition



Noisy Channel

- We see an observation x of a misspelled word
- Find the correct word w

$$\begin{aligned}\hat{w} &= \operatorname{argmax}_{w \in V} P(w \mid x) \\ &= \operatorname{argmax}_{w \in V} \frac{P(x \mid w)P(w)}{P(x)} \\ &= \operatorname{argmax}_{w \in V} P(x \mid w)P(w)\end{aligned}$$

History: Noisy channel for spelling proposed around 1990

- **IBM**

- Mays, Eric, Fred J. Damerau and Robert L. Mercer. 1991. Context based spelling correction. *Information Processing and Management*, 23(5), 517–522

- **AT&T Bell Labs**

- Kernighan, Mark D., Kenneth W. Church, and William A. Gale. 1990. A spelling correction program based on a noisy channel model. Proceedings of COLING 1990, 205-210

Non-word spelling error example

acress

Candidate generation

- Words with similar spelling
 - Small edit distance to error
- Words with similar pronunciation
 - Small edit distance of pronunciation to error

Damerau-Levenshtein edit distance

- Minimal edit distance between two strings, where edits are:
 - Insertion
 - Deletion
 - Substitution
 - Transposition of two adjacent letters

Words within 1 of across

Error	Candidate Correction	Correct Letter	Error Letter	Type
acress	actress	t	-	deletion
acress	cress	-	a	insertion
acress	caress	ca	ac	transposition
acress	access	c	r	substitution
acress	across	o	e	substitution
acress	acres	-	s	insertion
acress	acres	-	s	insertion

Candidate generation

- 80% of errors are within edit distance 1
- Almost all errors within edit distance 2
- Also allow insertion of **space** or **hyphen**
 - `thisidea` → `this idea`
 - `inlaw` → `in-law`

Language Model

- Use any of the language modeling algorithms we've learned
- Unigram, bigram, trigram
- Web-scale spelling correction
 - Stupid backoff

Unigram Prior probability

Counts from 404,253,213 words in Corpus of Contemporary English (COCA)

word	Frequency of word	P(word)
actress	9 , 321	.0000230573
cress	220	.0000005442
caress	686	.0000016969
access	37 , 038	.0000916207
across	120 , 844	.0002989314
acres	12 , 874	.0000318463

Channel model probability

- Error model probability, Edit probability
- *Kernighan, Church, Gale 1990*
- *Misspelled word $x = x_1, x_2, x_3 \dots x_m$*
- *Correct word $w = w_1, w_2, w_3, \dots, w_n$*
- $P(x|w)$ = probability of the edit
 - (deletion/insertion/substitution/transposition)

Computing error probability: confusion matrix

```
del[x,y]:      count(xy typed as x)
ins[x,y]:      count(x typed as xy)
sub[x,y]:      count(x typed as y)
trans[x,y]:    count(xy typed as yx)
```

Insertion and deletion conditioned on previous character

Confusion matrix for spelling errors

sub[X, Y] = Substitution of X (incorrect) for Y (correct)																											
X	Y (correct)																										
	a	b	c	d	e	f	g	h	i	j	k	l	m	n	o	p	q	r	s	t	u	v	w	x	y	z	
a	0	0	7	1	342	0	0	2	118	0	1	0	0	3	76	0	0	1	35	9	9	0	1	0	5	0	
b	0	0	9	9	2	2	3	1	0	0	0	5	11	5	0	10	0	0	2	1	0	0	8	0	0	0	
c	6	5	0	16	0	9	5	0	0	0	1	0	7	9	1	10	2	5	39	40	1	3	7	1	1	0	
d	1	10	13	0	12	0	5	5	0	0	2	3	7	3	0	1	0	43	30	22	0	0	4	0	2	0	
e	388	0	3	11	0	2	2	0	89	0	0	3	0	5	93	0	0	14	12	6	15	0	1	0	18	0	
f	0	15	0	3	1	0	5	2	0	0	0	3	4	1	0	0	0	6	4	12	0	0	2	0	0	0	
g	4	1	11	11	9	2	0	0	0	1	1	3	0	0	2	1	3	5	13	21	0	0	1	0	3	0	
h	1	8	0	3	0	0	0	0	0	0	2	0	12	14	2	3	0	3	1	11	0	0	2	0	0	0	
i	103	0	0	0	146	0	1	0	0	0	0	6	0	0	49	0	0	0	2	1	47	0	2	1	15	0	
j	0	1	1	9	0	0	1	0	0	0	0	2	1	0	0	0	0	0	5	0	0	0	0	0	0	0	
k	1	2	8	4	1	1	2	5	0	0	0	0	5	0	2	0	0	0	6	0	0	0	4	0	0	3	
l	2	10	1	4	0	4	5	6	13	0	1	0	0	14	2	5	0	11	10	2	0	0	0	0	0	0	
m	1	3	7	8	0	2	0	6	0	0	4	4	0	180	0	6	0	0	9	15	13	3	2	2	3	0	
n	2	7	6	5	3	0	1	19	1	0	4	35	78	0	0	7	0	28	5	7	0	0	1	2	0	2	
o	91	1	1	3	116	0	0	0	25	0	2	0	0	0	14	0	2	4	14	39	0	0	0	0	18	0	
p	0	11	1	2	0	6	5	0	2	9	0	2	7	6	15	0	0	1	3	6	0	4	1	0	0	0	
q	0	0	1	0	0	0	27	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
r	0	14	0	30	12	2	2	8	2	0	5	8	4	20	1	14	0	0	12	22	4	0	0	1	0	0	
s	11	8	27	33	35	4	0	1	0	1	0	27	0	6	1	7	0	14	0	15	0	0	5	3	20	1	
t	3	4	9	42	7	5	19	5	0	1	0	14	9	5	5	6	0	11	37	0	0	2	19	0	7	6	
u	20	0	0	0	44	0	0	0	64	0	0	0	0	2	43	0	0	4	0	0	0	0	2	0	8	0	
v	0	0	7	0	0	3	0	0	0	0	0	1	0	0	1	0	0	0	8	3	0	0	0	0	0	0	
w	2	2	1	0	1	0	0	2	0	0	1	0	0	0	0	7	0	6	3	3	1	0	0	0	0	0	
x	0	0	0	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	
y	0	0	2	0	15	0	1	7	15	0	0	0	2	0	6	1	0	7	36	8	5	0	0	1	0	0	
z	0	0	0	7	0	0	0	0	0	0	0	7	5	0	0	0	0	2	21	3	0	0	0	0	3	0	

Generating the confusion matrix

- Peter Norvig's list of errors
- Peter Norvig's list of counts of single-edit errors

Channel model

Kernighan, Church, Gale 1990

$$P(x|w) = \begin{cases} \frac{\text{del}[w_{i-1}, w_i]}{\text{count}[w_{i-1} w_i]}, & \text{if deletion} \\ \frac{\text{ins}[w_{i-1}, x_i]}{\text{count}[w_{i-1}]}, & \text{if insertion} \\ \frac{\text{sub}[x_i, w_i]}{\text{count}[w_i]}, & \text{if substitution} \\ \frac{\text{trans}[w_i, w_{i+1}]}{\text{count}[w_i w_{i+1}]}, & \text{if transposition} \end{cases}$$

Channel model for across

Candidate Correction	Correct Letter	Error Letter	x w	P(x word)
actress	t	-	c ct	.000117
cress	-	a	a #	.00000144
caress	ca	ac	ac ca	.00000164
access	c	r	r c	.000000209
across	o	e	e o	.0000093
acres	-	s	es e	.0000321
acres	-	s	ss s	.0000342

Noisy channel probability for across

Candidate Correction	Correct Letter	Error Letter	$x w$	$P(x word)$	$P(word)$	$10^9 * P(x w)P(w)$
actress	t	-	c ct	.000117	.0000231	2.7
cress	-	a	a #	.00000144	.000000544	.00078
caress	ca	ac	ac ca	.00000164	.00000170	.0028
access	c	r	r c	.000000209	.0000916	.019
across	o	e	e o	.0000093	.000299	2.8
acres	-	s	es e	.0000321	.0000318	1.0
acres	-	s	ss s	.0000342	.0000318	1.0

Noisy channel probability for across

Candidate Correction	Correct Letter	Error Letter	$x w$	$P(x word)$	$P(word)$	$10^9 * P(x w)P(w)$
actress	t	-	c ct	.000117	.0000231	2.7
cress	-	a	a #	.00000144	.000000544	.00078
caress	ca	ac	ac ca	.00000164	.00000170	.0028
access	c	r	r c	.000000209	.0000916	.019
across	o	e	e o	.0000093	.000299	2.8
acres	-	s	es e	.0000321	.0000318	1.0
acres	-	s	ss s	.0000342	.0000318	1.0

Using a bigram language model

- "a stellar and versatile **acress** whose combination of sass and glamour..."
- Counts from the Corpus of Contemporary American English with add-1 smoothing
- $P(\text{actress}|\text{versatile}) = .000021$ $P(\text{whose}|\text{actress}) = .0010$
- $P(\text{across}|\text{versatile}) = .000021$ $P(\text{whose}|\text{across}) = .000006$
- $P(\text{"versatile actress whose"}) = .000021 * .0010 = 210 \times 10^{-10}$
- $P(\text{"versatile across whose"}) = .000021 * .000006 = 1 \times 10^{-10}$

Using a bigram language model

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- $P(\text{"versatile across whose"}) = .000021 * .000006 = 1 \times 10^{-10}$

Evaluation

- Some spelling error test sets
 - [Wikipedia's list of common English misspelling](#)
 - [Aspell filtered version of that list](#)
 - [Birkbeck spelling error corpus](#)
 - [Peter Norvig's list of errors \(includes Wikipedia and Birkbeck, for training or testing\)](#)

Spelling Correction and the Noisy Channel

Real-Word Spelling Correction

Real-word spelling errors

- ...leaving in about fifteen **minuets** to go to her house.
- The design **an** construction of the system...
- Can they **lave** him my messages?
- The study was conducted mainly **be** John Black.
- 25-40% of spelling errors are real words [Kukich 1992](#)

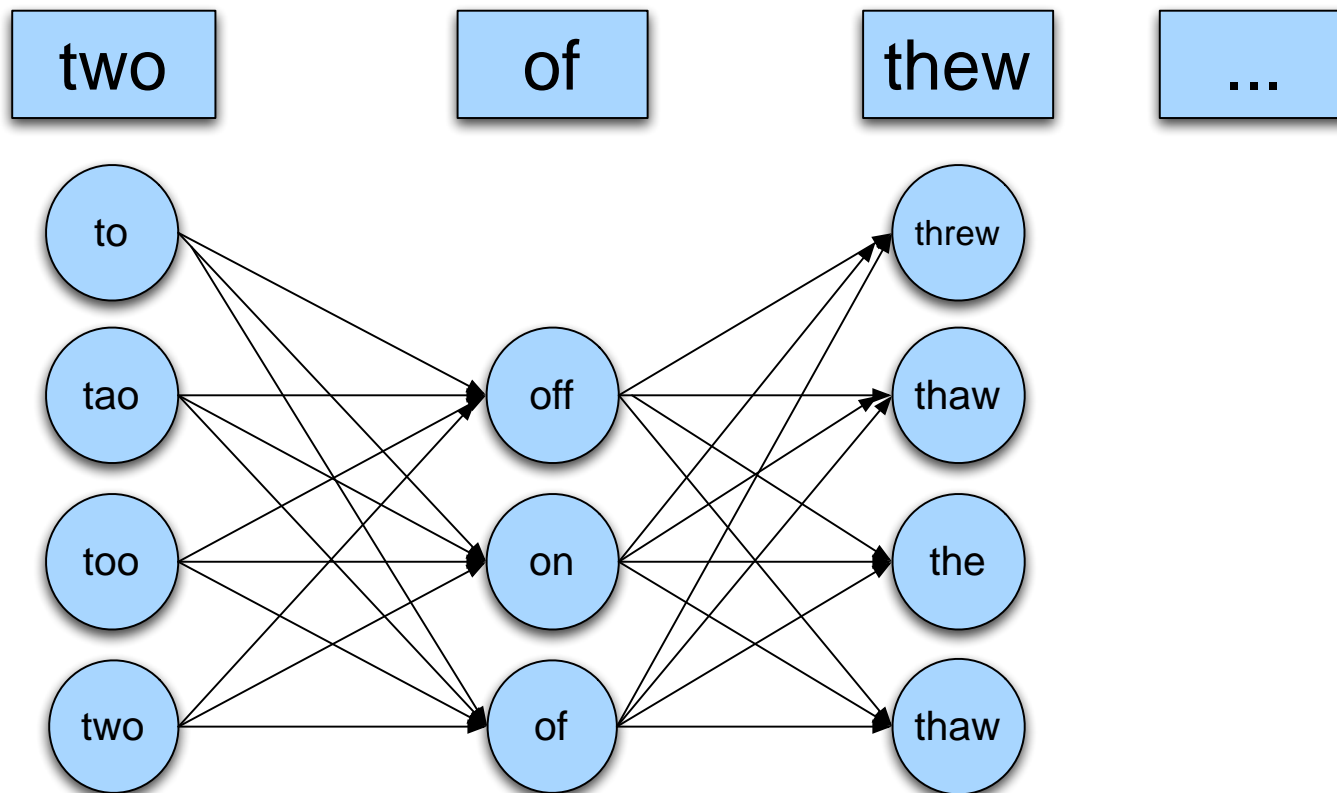
Solving real-world spelling errors

- For each word in sentence
 - Generate *candidate set*
 - the word itself
 - all single-letter edits that are English words
 - words that are homophones
- Choose best candidates
 - Noisy channel model
 - Task-specific classifier

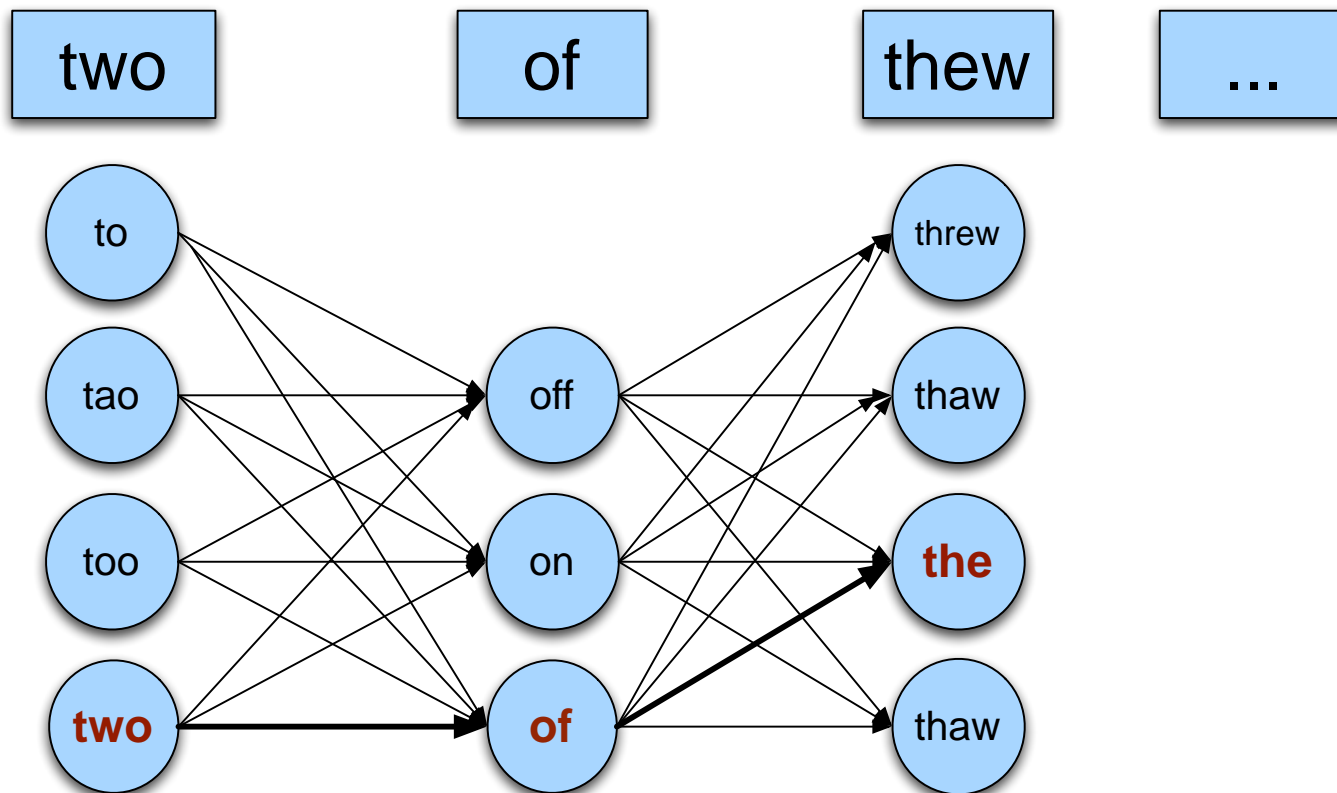
Noisy channel for real-word spell correction

- Given a sentence $w_1, w_2, w_3, \dots, w_n$
- Generate a set of candidates for each word w_i
 - $\text{Candidate}(w_1) = \{w_1, w'_1, w''_1, w'''_1, \dots\}$
 - $\text{Candidate}(w_2) = \{w_2, w'_2, w''_2, w'''_2, \dots\}$
 - $\text{Candidate}(w_n) = \{w_n, w'_n, w''_n, w'''_n, \dots\}$
- Choose the sequence W that maximizes $P(W)$

Noisy channel for real-word spell correction



Noisy channel for real-word spell correction



Simplification: One error per sentence

- Out of all possible sentences with one word replaced
 - w_1, w''_2, w_3, w_4 **two off thew**
 - w_1, w_2, w'_3, w_4 **two of the**
 - w'''_1, w_2, w_3, w_4 **too of thew**
 - ...
- Choose the sequence W that maximizes $P(W)$

Where to get the probabilities

- Language model
 - Unigram
 - Bigram
 - Etc
- Channel model
 - Same as for non-word spelling correction
 - Plus need probability for no error, $P(w|w)$

Probability of no error

- What is the channel probability for a correctly typed word?
- $P(\text{"the"} | \text{"the"})$
- Obviously this depends on the application
 - .90 (1 error in 10 words)
 - .95 (1 error in 20 words)
 - .99 (1 error in 100 words)
 - .995 (1 error in 200 words)

Peter Norvig's “thew” example

x	w	x w	$P(x w)$	$P(w)$	$10^9 P(x w)P(w)$
thew	the	ew e	0.0000007	0.02	144
thew	thew		0.95	0.000000009	90
thew	thaw	e a	0.001	0.00000007	0.7
thew	threw	h hr	0.0000008	0.0000004	0.03
thew	thwe	ew we	0.0000003	0.000000004	0.0001

Spelling Correction and the Noisy Channel

State-of-the-art Systems

HCI issues in spelling

- If very confident in correction
 - Autocorrect
- Less confident
 - Give the best correction
- Less confident
 - Give a correction list
- Unconfident
 - Just flag as an error

State of the art noisy channel

- We never just multiply the prior and the error model
- Independence assumptions \rightarrow probabilities not commensurate
- Instead: Weigh them

$$\hat{w} = \operatorname{argmax}_{w \in V} P(x | w) P(w)^\lambda$$

- Learn λ from a development test set

Phonetic error model

- Metaphone, used in GNU aspell
 - Convert misspelling to metaphone pronunciation
 - “Drop duplicate adjacent letters, except for C.”
 - “If the word begins with 'KN', 'GN', 'PN', 'AE', 'WR', drop the first letter.”
 - “Drop 'B' if after 'M' and if it is at the end of the word”
 - ...
 - Find words whose pronunciation is 1-2 edit distance from misspelling's
 - Score result list
 - Weighted edit distance of candidate to misspelling
 - Edit distance of candidate pronunciation to misspelling pronunciation

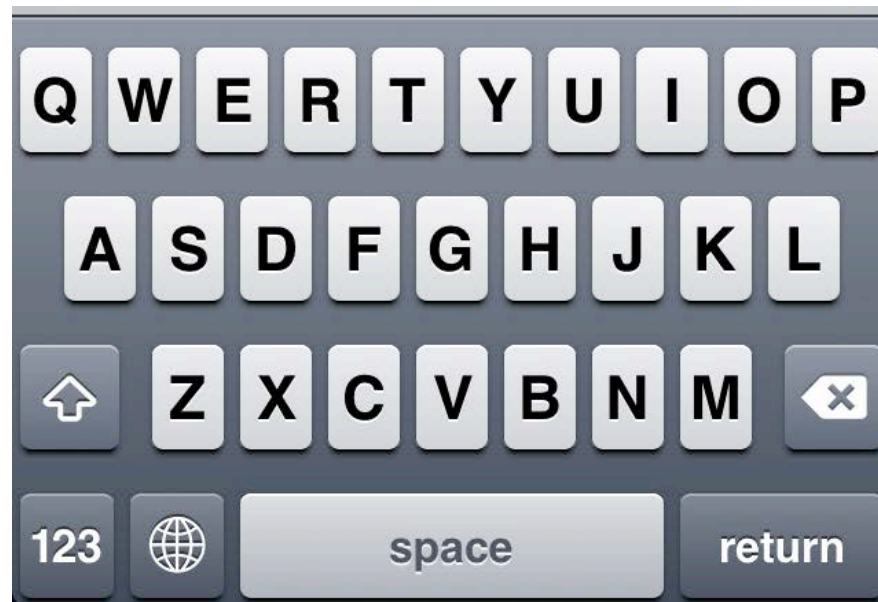
Improvements to channel model

- Allow richer edits (Brill and Moore 2000)
 - ent→ant
 - ph→f
 - le→al
- Incorporate pronunciation into channel (Toutanova and Moore 2002)

Channel model

- Factors that could influence $p(\text{misspelling} | \text{word})$
 - The source letter
 - The target letter
 - Surrounding letters
 - The position in the word
 - Nearby keys on the keyboard
 - Homology on the keyboard
 - Pronunciations
 - Likely morpheme transformations

Nearby keys



Classifier-based methods for real-word spelling correction

- Instead of just channel model and language model
- Use many features in a classifier.
- Build a classifier for a specific pair like:

whether/weather

- “cloudy” within +- 10 words
- ____ to VERB
- ____ or not

