

# Word Pronunciation

Julia Hirschberg

CS 4706

# Today

- Motivation
- Challenges for automatic word pronunciation
- Standard methods
- Innovative solutions

- TTS demos:
  - ScanSoft/Nuance
  - AT&T
  - IBM
  - Cepstral
- SNL Robot Repair

# Motivation

- Intelligibility
- Naturalness
- Applications to language learning
  - Unlimited vocabulary
  - Type a word or phrase and hear it spoken in your target language
    - To imitate
    - To learn to recognize
- Speech therapy

# Word Pronunciation

- What determines how a word is pronounced?
  - History/Language Origin/Dictionaries:
    - shoe (ME shoo), phoenix (Gr)
    - mole, attaches, resume
  - Part-of-speech:
    - use, close, dove, multiply, coax
  - Morphology:
    - ferryboat, ferryboats
    - Popemobile (pope+mobile)

# Letter-to-Sound Rules

- Define correspondences between orthography and phonemic representation, e.g.
  - $i \_ \{C\} e \$ \rightarrow /ai/$       **rise**
  - Else  $i \rightarrow /ih/$       **rip**
- Deals with any input

# Problems

- Must be built by hand
- Many exceptions, e.g.
  - $i_{\{C\}}e \rightarrow /ai/$  matches **ripen/risen/riser/river/ripper**
  - Proper names: **Nice, Ramirez, Ribeiro, Rise, Infiniti**
- Symbols and abbreviations: &c, evalu8, cu, tsp
- Assigning **lexical stress**
- Solutions
  - More complex rules
  - Exceptions dictionary
    - Consulted first
    - But how do we handle morphological variation? E.g.
      - **Rise's hat**

# Dictionary-based Approaches

- Rely on very large dictionary with orthography and pronunciation for each word
- Typically created by hand or by expansion of online pronouncing dictionary



# Problems

- Redundancy of representation
  - Cat, cats, cat's, cats'
- Out-of-vocabulary (OOV) items
  - Proper names: covering all U.K. surnames would require >5,000,000 entries
  - New words: ...
    - Technical terms: liposuction, anova, bernaise
    - Foreign borrowings: frappe, ciao, louche

- Solutions
  - Larger dictionary
  - Morphological preprocessing before dictionary look-up
  - Fall back to L2Sound rules if no dictionary 'hit'

# Major Challenges for TTS

- Disambiguating homographs
  - bass/bass
- Pronouncing new words
  - New names in the news:
  - New words: iPad, Kindle
- Expanding abbreviations and acronyms correctly

# Homograph Disambiguation by Decision List Classifiers (Yarowsky '97)

- E.g., **bass/bass, nice/Nice, live/live, desert/desert, lead/lead**

Pronunciation	Context
(1) led	.. it monitors the <i>lead</i> levels in drinking
(1) led	... median blood <i>lead</i> concentration was
(1) led	.. found layers of <i>lead</i> telluride inside ..
(1) led	... conference on <i>lead</i> poisoning in ...
(1) led	.. strontium and <i>lead</i> isotope zonation ..
(2) lid	maintained their <i>lead</i> Thursday over ...
(2) lid	.. to Boston and <i>lead</i> singer for Purple
(2) lid	Bush a 17-point <i>lead</i> in Texas , only 3
(2) lid	his double digit <i>lead</i> nationwide . The
(2) lid	the fairly short <i>lead</i> time allowed on ..

- Rank by  $Abs(\text{Log} \left( \frac{P(\text{Sense}_1 | f_i = v_j)}{P(\text{Sense}_2 | f_i = v_j)} \right))$

# Pronouncing OOV Words

- Techniques for handling OOVs
  - Inferring country of origin:
    - Takashita, Leroy, Kirov, Lima, Infiniti
  - Pronunciation by analogy
    - Analog/dialog
    - Risible/visible
    - Proper names: Alifano/Califano

# Bootstrapping Phonetic Lexicons (Maskey et al '04)

- For some languages, online pronouncing lexicons exist – but for others....e.g. Nepali
  - How to minimize effort in creating lexicons?
- Approach
  - Given a native speaker and a large amount of online text in the language...
    - Native speaker builds small lexicon by hand for seed set of N most common words in text, e.g.
      - is: /izh/
      - the: /dhax/

- Derive L2S rules from lexicon automatically, e.g.
  - is → ih{zh}
  - the → {dh}ax ...
- Loop: Choose the next N most common set of words from the text and use the lexicon + L2S rules to predict pronunciations, e.g.
  - telephone -> /telaxfown/
  - He -> /hax/?
  - Rise -> /rihzhax/?
- Assign a confidence score to each prediction by comparing each word to all words in lexicon
  - If is -> /ihzh} in lexicon and no other orthographically similar words are pronounced differently, new rule his -> /hihzh/ scores high

- For low confidence pronunciations, Active Learning step:
  - Inspect and calculate error rate
  - Hand correct errors and add all to lexicon
- Iterate from Loop until performance stabilizes
  - Build a new set of L2S rules from augmented lexicon
- Results
  - English:
    - 94% success on test set after 23 iterations, 16K entry lexicon
    - Performance comparable to CMUDict and 1/7 the size
  - German:
    - 90% accuracy after 13 iterations, 28K lexicon
  - Nepali
    - 94.6% accuracy after 16 iterations, 5K lexicon



# Improving Pronunciation Dictionary Coverage (Fackrell and Skut '04)

- Idea: Many proper names have more than one spelling (e.g. More/Moore; Smith/Smythe)
  - Homophones
  - Find a ‘fuzzy’ mapping between OOV (Out of Vocabulary) words and words already in the lexicon
  - Identify spelling alternations that are ‘pronunciation-neutral’ in an existing lexicon to produce rewrite rules for OOVs

- Pros?
- Cons?

# Deriving Pronunciations from the Web (Ghoshal et al '09)

- Extract candidate orthography/pronunciation pairs (ad-hoc and IPA)
  - E.g. **bruschetta (pronounced broo-SKET-uh)**
- Validate the candidates: how likely are these pairs to represent a word and its pronunciation
- Normalize ad-hoc and IPA pronunciations

- Pros?
- Cons?

# Pronunciation Evaluation

- How would you evaluate the pronunciation module of a TTS system?

# Next Class

- Readings
- Download the ToBI cardinal examples (see <http://www1.cs.columbia.edu/~agus/tobi/>)
  - You will first need to download WaveSurfer
    - <http://www.speech.kth.se/wavesurfer/>
  - Then download the cardinal examples
    - <http://www1.cs.columbia.edu/~agus/tobi/cardinals/manual.php>
- Listen to each of the cardinal examples
  - Try to imitate each one and to decide what it ‘means’