Automatic Speech Recognition: Introduction

Steve Renals & Hiroshi Shimodaira

Automatic Speech Recognition— ASR Lecture 1 14 January 2019

Image: A Image: A

Automatic Speech Recognition — ASR

Course details

- Lectures: About 18 lectures
- Labs: Weekly lab sessions using Kaldi (kaldi-asr.org)
 - Lab sessions in AT-4.12: Tuesdays 10:00, Wednesdays 10:00, Wednesdays 15:10, start week 2 (22/23 January)
 - Select one lab session on Learn

Assessment:

- Exam in April or May (worth 70%)
- Coursework (worth 30%, building on the lab sessions) (out on Thurday 14 February; in by Wednesday 20 March)

• People:

- Lecturers: Steve Renals and Hiroshi Shimodaira
- TAs: Joachim Fainberg and Ondrej Klejch

http://www.inf.ed.ac.uk/teaching/courses/asr/

(日本) (日本) (日本)

- If you have taken:
 - Speech Processing and either of (MLPR or MLP)
 - Perfect!
 - either of (MLPR or MLP) *but not* Speech Processing (probably you are from Informatics)
 - You'll require some speech background:
 - A couple of the lectures will cover material that was in Speech Processing
 - Some additional background study (including material from Speech Processing)
 - Speech Processing *but neither of* (MLPR or MLP) (probably you are from SLP)
 - You'll require some machine learning background (especially neural networks)
 - A couple of introductory lectures on neural networks provided for SLP students
 - Some additional background study

イロト イポト イヨト イヨト

- Series of weekly labs using Kaldi.
 - Sign up for one lab session on Learn
- Labs start week 2 (next week)
- Note: Training speech recognisers can take time
 - ASR training in some labs will not finish in an hour...
 - Give yourself plenty of time to complete the coursework, don't leave it until the last couple of days

高 ト イ ヨ ト イ ヨ ト

Speech-to-text transcription

- Transform recorded audio into a sequence of words
- Just the words, no meaning.... But do need to deal with acoustic ambiguity: "Recognise speech?" or "Wreck a nice beach?"
- Speaker diarization: Who spoke when?
- Speech recognition: what did they say?
- Paralinguistic aspects: how did they say it? (timing, intonation, voice quality)
- Speech understanding: what does it mean?

A (1) < A (2) < A (2) </p>

Why is speech recognition difficult?

高 ト イ ヨ ト イ ヨ ト

크

Variability in speech recognition

Several sources of variation

Size Number of word types in vocabulary, perplexity

크

Variability in speech recognition

Several sources of variation

Size Number of word types in vocabulary, perplexity Speaker Tuned for a particular speaker, or speaker-independent? Adaptation to speaker characteristics

Size Number of word types in vocabulary, perplexity

Speaker Tuned for a particular speaker, or speaker-independent? Adaptation to speaker characteristics

Acoustic environment Noise, competing speakers, channel conditions (microphone, phone line, room acoustics)

Size Number of word types in vocabulary, perplexity Speaker Tuned for a particular speaker, or speaker-independent? Adaptation to speaker characteristics

Acoustic environment Noise, competing speakers, channel conditions (microphone, phone line, room acoustics) Style Continuously spoken or isolated? Planned monologue or spontaneous conversation?

Size Number of word types in vocabulary, perplexity Speaker Tuned for a particular speaker, or speaker-independent? Adaptation to speaker characteristics

Acoustic environment Noise, competing speakers, channel conditions (microphone, phone line, room acoustics) Style Continuously spoken or isolated? Planned monologue or spontaneous conversation?

Accent/dialect Recognise the speech of all speakers who speak a particular language

向下 イヨト イヨト

Size Number of word types in vocabulary, perplexity Speaker Tuned for a particular speaker, or speaker-independent? Adaptation to speaker characteristics

Acoustic environment Noise, competing speakers, channel conditions (microphone, phone line, room acoustics)

Style Continuously spoken or isolated? Planned monologue or spontaneous conversation?

Accent/dialect Recognise the speech of all speakers who speak a particular language

Language spoken There are many languages beyond English, Mandarin Chinese, Spanish, ...

What is the difference between a dialect and a language?

・ロト ・回 ト ・ヨト ・ヨト

Hierarchical modelling of speech



・ロト ・回ト ・ヨト ・ヨト

æ

Hierarchical modelling of speech



æ

"Fundamental Equation of Statistical Speech Recognition"

If **X** is the sequence of acoustic feature vectors (observations) and **W** denotes a word sequence, the most likely word sequence \mathbf{W}^* is given by

$$\mathbf{W}^* = rg\max_{\mathbf{W}} P(\mathbf{W} \mid \mathbf{X})$$

"Fundamental Equation of Statistical Speech Recognition"

If **X** is the sequence of acoustic feature vectors (observations) and **W** denotes a word sequence, the most likely word sequence \mathbf{W}^* is given by

$$\mathbf{W}^* = rg \max_{\mathbf{W}} P(\mathbf{W} \mid \mathbf{X})$$

Applying Bayes' Theorem:

$$P(\mathbf{W} \mid \mathbf{X}) = \frac{p(\mathbf{X} \mid \mathbf{W})P(\mathbf{W})}{p(\mathbf{X})}$$

$$\propto p(\mathbf{X} \mid \mathbf{W})P(\mathbf{W})$$

$$\mathbf{W}^* = \arg \max_{\mathbf{W}} \underbrace{p(\mathbf{X} \mid \mathbf{W})}_{\text{Acoustic}} \underbrace{P(\mathbf{W})}_{\text{Language}}$$

model model

$$\mathbf{W}^* = \arg\max_{\mathbf{W}} p(\mathbf{X} \mid \mathbf{W}) P(\mathbf{W})$$

Use an acoustic model, language model, and lexicon to obtain the most probable word sequence \bm{W}^* given the observed acoustics \bm{X}









Directly model transforming an input acoustic sequence into an output word or character sequence "No right" Utterance W Direct mapping: acoustics - transcription 4000 Acoustics X 3000 2000 1500 1000 ASR Lecture 1 Automatic Speech Recognition: Introduction





Representing recorded speech (X)



Represent a recorded utterance as a sequence of *feature vectors*

Reading: Jurafsky & Martin section 9.3

(日)

Labelling speech (W)



Labels may be at different levels: words, phones, etc. Labels may be *time-aligned* – i.e. the start and end times of an acoustic segment corresponding to a label are known

Reading: Jurafsky & Martin chapter 7 (especially sections 7.4, 7.5)

Phonemes

- abstract unit defined by linguists based on contrastive role in word meanings (eg "cat" vs "bat")
- 40-50 phonemes in English
- Phones
 - speech sounds defined by the acoustics
 - many allophones of the same phoneme (eg /p/ in "pit" and "spit")
 - limitless in number
- Phones are usually used in speech recognition but no conclusive evidence that they are the basic units in speech recognition
- Possible alternatives: syllables, automatically derived units, ...

イロト イポト イヨト イヨト

⁽Slide taken from Martin Cooke from long ago)

Example: TIMIT Corpus

- TIMIT corpus (1986)—first widely used corpus, still in use
 - Utterances from 630 North American speakers
 - Phonetically transcribed, time-aligned
 - Standard training and test sets, agreed evaluation metric (phone error rate)
- TIMIT phone recognition label the audio of a recorded utterance using a sequence of phone symbols
 - Frame classification attach a phone label to each frame data
 - Phone classification given a segmentation of the audio, attach a phone label to each (multi-frame) segment
 - Phone recognition supply the sequence of labels corresponding to the recorded utterance

・ 同 ト ・ ヨ ト ・ ヨ ト

16

Basic speech recognition on TIMIT

- Train a classifier of some sort to associate each feature vector with its corresponding label. Classifier could be
 - Neural network
 - Gaussian mixture model
 - ...

Then at run time, a label is assigned to each frame

- Questions
 - What's good about this approach?
 - What the limitations? How might we address them?

Image: A Image: A

Evaluation

- How accurate is a speech recognizer?
- String edit distance
 - Use dynamic programming to align the ASR output with a reference transcription
 - Three type of error: insertion, deletion, substitutions
- Word error rate (WER) sums the three types of error. If there are *N* words in the reference transcript, and the ASR output has *S* substitutions, *D* deletions and *I* insertions, then:

$$\mathsf{WER} = 100 \cdot \frac{\mathsf{S} + \mathsf{D} + \mathsf{I}}{\mathsf{N}}$$
 Accuracy $= 100 - \mathsf{WER}$ %

- For TIMIT, define phone error error rate analagously to word error rate
- Speech recognition evaluations: common training and development data, release of new test sets on which different systems may be evaluated using word error rate



イロト イヨト イヨト イヨト

æ

- Jurafsky and Martin (2008). *Speech and Language Processing* (2nd ed.): Chapter 7 (esp 7.4, 7.5) and Section 9.3.
- General interest:
 - The Economist Technology Quarterly, "Language: Finding a Voice", Jan 2017. http://www.economist.com/technology-quarterly/2017-05-01/language
 - The State of Automatic Speech Recognition: Q&A with Kaldi's Dan Povey, Jul 2018. https://medium.com/descript/the-state-of-automaticspeech-recognition-q-a-with-kaldis-dan-poveyc860aada9b85

・ロト ・回ト ・ヨト ・ヨト