# 語音事件偵測器的設計與估算 On Designing and Evaluating Speech Event Detectors

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#### Introduction

 Event detection is a critical component of a recently proposed automatic speech attribute transcription (ASAT) paradigm.

 A good detector needs to effectively detect speech attributes of interest while rejecting extraneous events.

# 簡介

• <u>事件偵測</u>在最近提出的ASAT中是一個很重要的元件

• 一個好的偵測器必須能夠偵測出有效的語音屬性並拒絕無關的事件

#### Introduction

 These "events" are usually low level speech attributes related to information required to form higher level "evidences", so they can then be combined to detect phones, words and sentences.

# 簡介

• 這邊的"事件"通常都是指低階的語音屬性, 透過相關資料可以做轉換形成高階的"證據", 並將多個證據作結合成為偵測音素、字、 和句子

#### Introduction

 We compare <u>frame</u> and <u>segment</u> based detector by their properties in detecting manners of articulation.

 Proposed new mechanism to evaluate detector performance.

## 簡介

 我們由音框基礎偵測器與音段基礎偵測器 兩者在發音行為上的特性作為比較之依據

• 並提出一個新的估算偵測器效能的方法

> Frame and segment based detector:

#### (1) frame based detector:

realized with artificial neural networks (ANNs).

#### (2) Segment based detector:

built by combining frame based detectors, or with segment models, such as hidden Markov models (HMMs)

>音框基礎與音段基礎偵測器:

(1)音框基礎偵測器 實現在ANN上

(2)音段基礎偵測器

可以由結合音框基礎偵測器或者使用音段 模型,例如HMM

#### (1)Frame based detector

• ANN based detectors is that the output scores can simulate the *a posteriori probabilities of an attribute given the* speech signal.

 Strictly speaking, these event "detectors" can also categorize each speech frame into one of the competing attributes.

#### (1)音框基礎偵測器

 ANN為基礎的偵測器,主要會將語音訊號 的提供的屬性轉換成為一個事後機率

嚴格來說,這些事件偵測器可將每個音框 分類到對應的屬性

#### (2) Segment based detector

 We train two HMMs, one for a target event and the other for all other competing events to decode the utterance.

 The segments that are recognized with the target label mapping to the detected target events.

#### (2)音段基礎偵測器

• 我們訓練兩個HMMS,一個是用在處理<u>目標</u> 事件,而另外一個是用在其他的競爭事件, 並利用他們來對句子做解碼

• 根據目標纜籤對應的偵測目標事件來作為 音段的辨識

#### > Performance measurement

 For the fricative event, the reference and detected strings are listed as <u>ref.mlf</u> and <u>det.mlf</u>

```
ref.mlf
                              det.mlf
   #!MLF!#
                                  #!MLF!#
   "*/si1039.lab"
                                  "*/si1039.rec"
   0 8 fricative -3.1
                                  0 13 fricative -13.1
   8 13 fricative -10.0
                                  13 21 non -12.1
                                  21 24 fricative -16.0
   13 21 non -12.1
   21 30 fricative -34.2
                                  24 30 fricative -18.2
   30 41 non -19.0
                                  30 41 non -19.0
                                  41 45 fricative -3.0
   41 45 fricative -3.0
```

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#### > 效能偵測

• 在摩擦音事件中,參照與偵測字串為 ref.mlf 和det.mlf

```
ref.mlf
                              det.mlf
   #!MLF!#
                                  #!MLF!#
   "*/si1039.lab"
                                  "*/si1039.rec"
   0.8 fricative -3.1
                                  0 13 fricative -13.1
   8 13 fricative -10.0
                                  13 21 non -12.1
                                  21 24 fricative -16.0
    13 21 non -12.1
                                  24 30 fricative -18.2
   21 30 fricative -34.2
   30 41 non -19.0
                                  30 41 non -19.0
   41 45 fricative -3.0
                                  41 45 fricative -3.0
```

- In HTK, the detection false alarms (FA) and hits can be obtained with its HResults tool.
- If the reference label lies between the start and end times of a recognized segment with the same label, we regard it as a hit.
- For every detected segment, we consider it as a FA only if the center of the recognized segment falls into the reference segment with a different label.

• 在HTK中,我們可從HResult tool中獲得誤警報(FA)和擊中(hit)

• 如果辨識音段在開始到結束這段時間內的標籤與參照標籤相同時,我們稱為擊中(hit)

• 若每個偵測音段,其中間辨識音段與落在 參照音段的標籤不同時,稱為FA

#### > Detector optimization

 In order to enhance detector performance, we can use discriminative speech parameters and models with detailed acoustic resolution and refined content dependency.

For MCE training, the LLR is defined as:

$$LLR(O) = \log L(O|\Lambda_0) - \log L(O|\Lambda_1)$$

- ▶偵測器最佳化
- 我們可以利用含有詳細聲學分析與內容相 依的個別的語音參數和模型來加強偵測器 效能

• 針對MCE(minimum classification error)訓練 定義LLR為:

$$LLR(O) = \log L(O|\Lambda_0) - \log L(O|\Lambda_1)$$

## Experiment

Use the TIMIT database

 We design a experiment for comparing <u>frame</u> <u>based ANN</u> and <u>segment HMM</u> detectors for manner of articulation attributes.

# 實驗

• 使用TIMIT資料庫

• 我們設計一個實驗,比較以音框為基礎的 ANN和音段HMM的偵測器

## Experiment

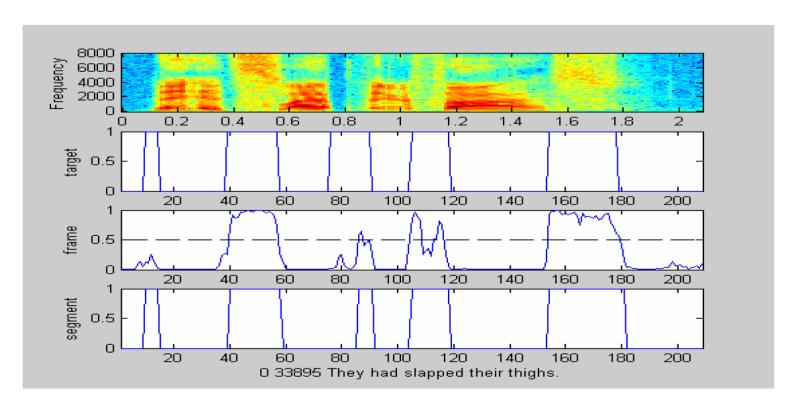


Figure 1 Detection curves of the ANN frame and HMM segment based detectors for the fricative attribute

### Experiment

Detection with known segment boundaries
 -sum over all frame scores of the ANN detector to smooth noisy outputs to some extent.

Detection with unknown segment boundaries

 use the two competing HMMs to decode each utterance into the target and non-target segments, and compute the FA, FR and error rates.

# 實驗

在音段界線已知情況下測量偵測器效能
 -將ANN產生的所有音框分數作加總,來做平移到一個程度

在音段界線未知情況下測量偵測器效能
 使用兩個HMM來對句子做解碼,而分成目標音段與非目標音段,並計算FA、FR和錯誤率