

Forced Alignment and Speech Recognition Systems

Overview

- Uses of automatic speech recognition technology
- Principles of forced alignment and speech recognition systems
- Some practicalities
- Evaluating alignment quality

ASR technology - Existing uses

As a toolbox:

Pre-built generic application used as a tool

- > speech recognition
- > forced alignment

for lexical transcription or time stamps

As a methodology:

Forms an integral part of the experimental procedure

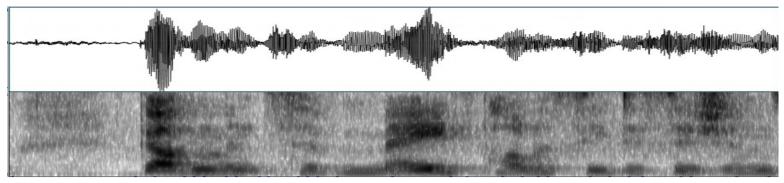
> tune for unusual pronunciations

> extract probability of words/phonemes matching models

> detect assimilation, deletion, insertion

Forced alignment

With transcription: Already know exactly what is in the audio.

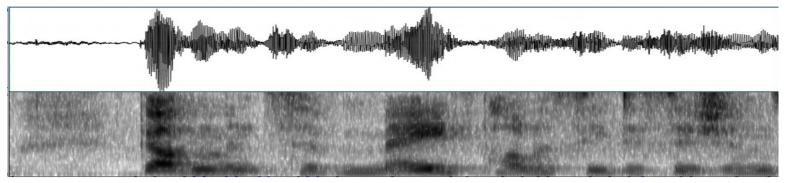


Align

sp	g ahi m n s	v	M EY1 D	$P \stackrel{A}{A1} L \qquad S \begin{matrix} IY \\ 0 \end{matrix}$	D _H ^I S ^I ZHNZ
sp	GOVERNMENTS	HAVE	MADE	POLICY	DECISIONS

Forced alignment

With some transcription: Know what is in some of the audio.

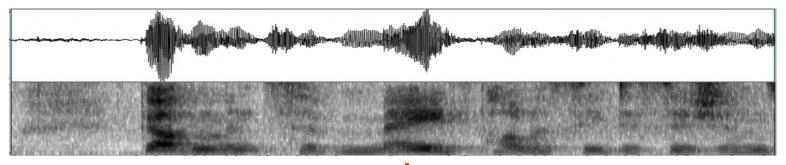


Align

Noise	M EY1			P A L S IY 0			Nicion			
Noise	N	IADE			PC	OLI	CY		Noise	

Automatic speech recognition

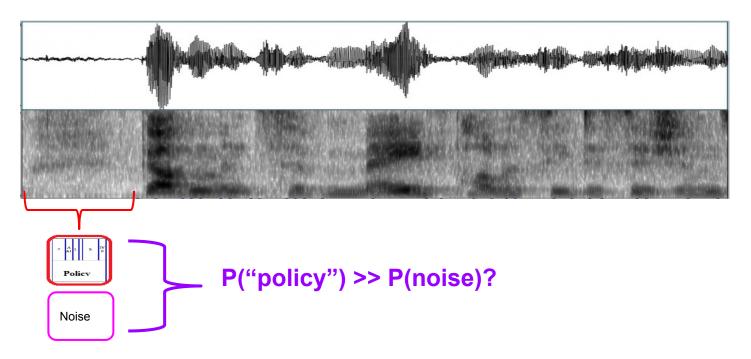
No transcription: Don't know what's in the audio

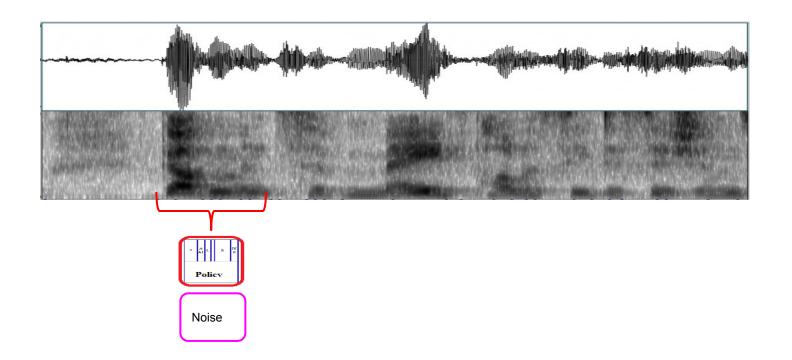


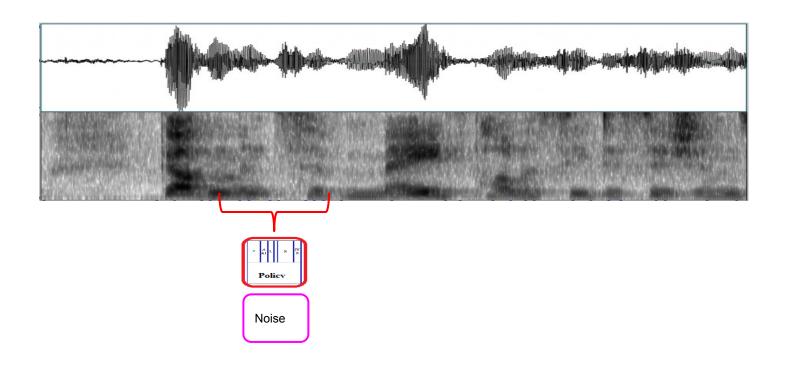
Estimate

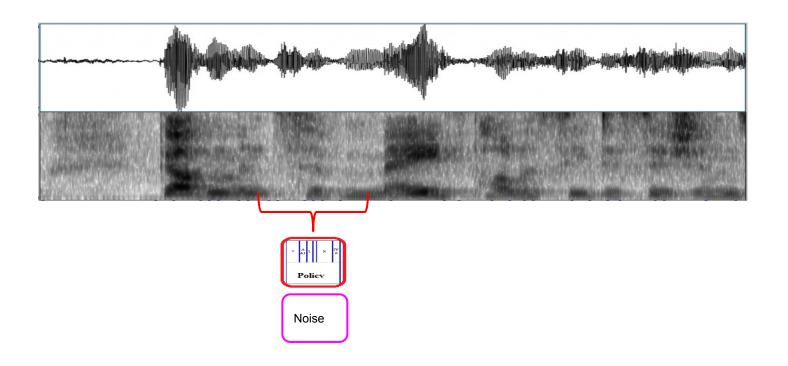
sp	g ahi m n s	v	M EY1 D	$P \stackrel{A}{A1} L \qquad S \stackrel{IY}{0}$	D _H ^I S ^I ZHN	Z phone (148/1931)
sp	GOVERNMENTS	HAVE	MADE	POLICY	DECISIONS	word (531)

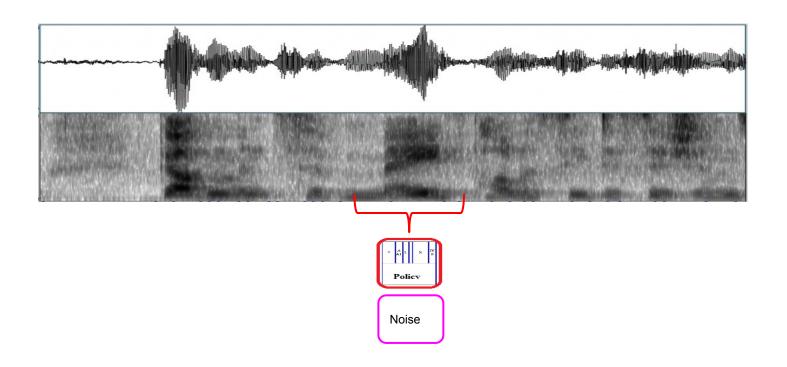
Possible transcription: Looking for a word/phrase



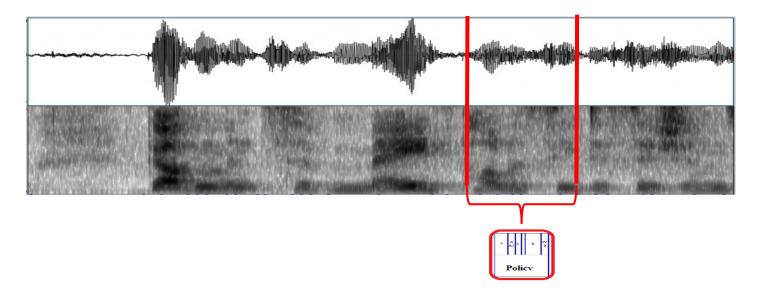






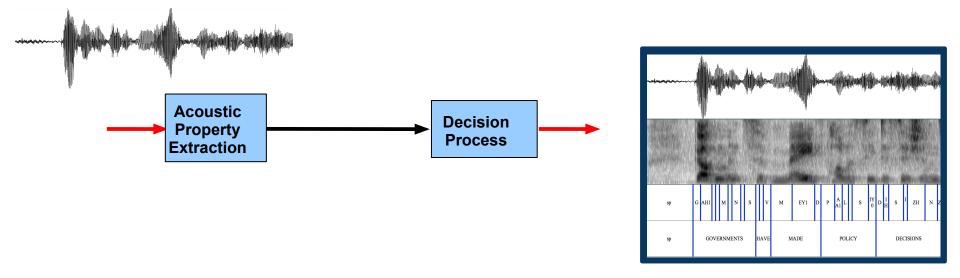


Possible transcription:

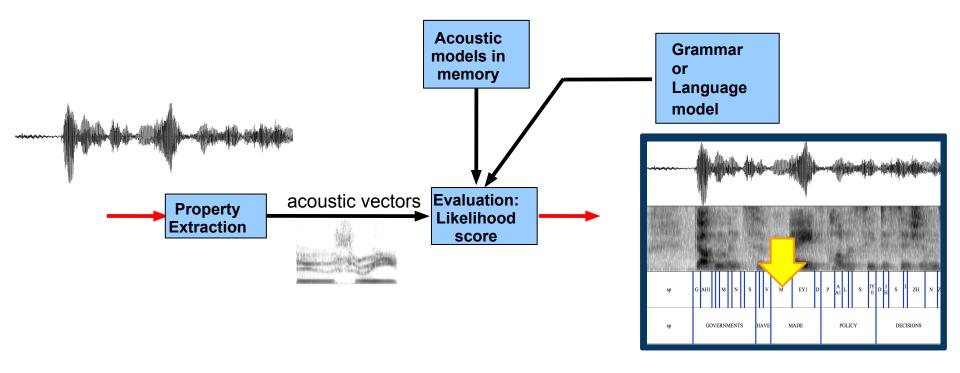


Yes! P("policy") >> P(noise)

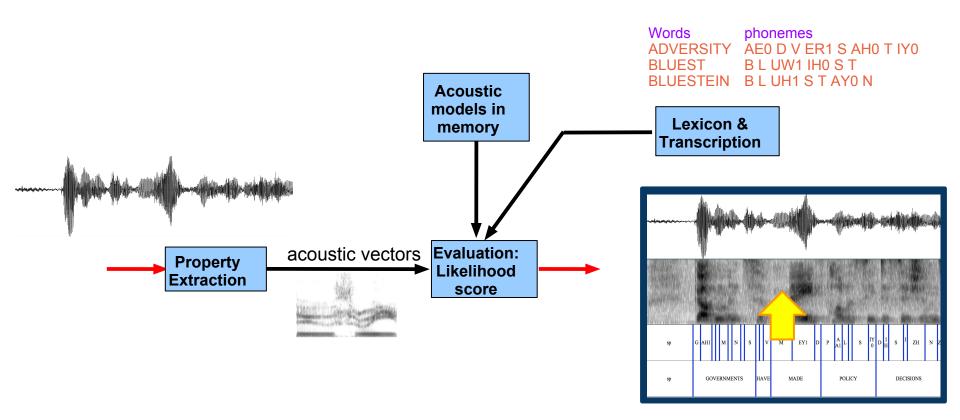
Early Automatic Speech Recognition (ASR) Systems



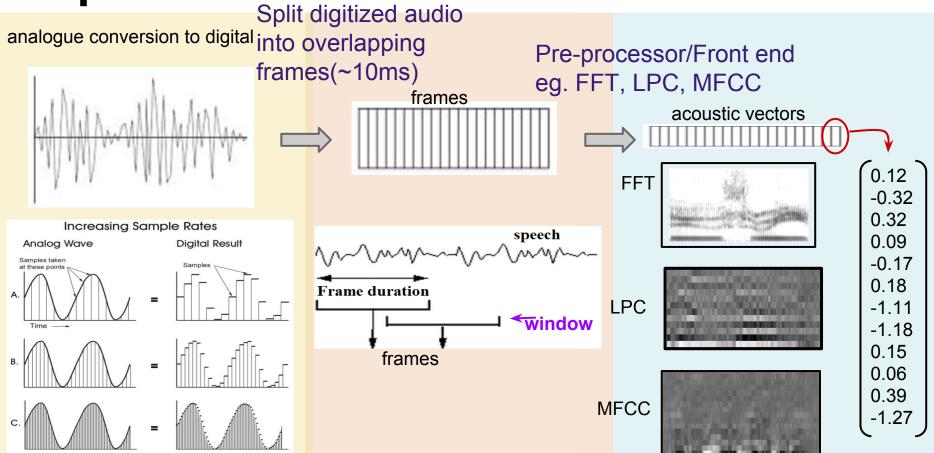
Automatic Speech Recognition System



Forced Alignment system



Speech transformation:



Speech transformation for ASR:



Split digitized audio into overlapping segments frames

Pre-processor/Front end eg. FFT, LPC, MFCC



Sampling rate

Window duration, shape and overlap **Pre-processor**

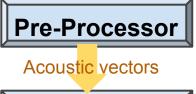
Forced Alignment can be based on:

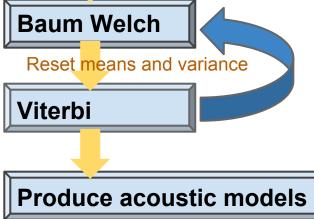
- (Mono)phones
- Diphones
- Triphones
- Words
- Syllables
- Broad phonetic classes (eg. stop, sonorant, vowel)

Training the system Using Hidden Markov Models (HMM)

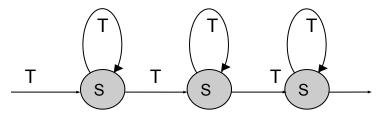
Digitised speech for training

Dat<mark>a</mark> set A

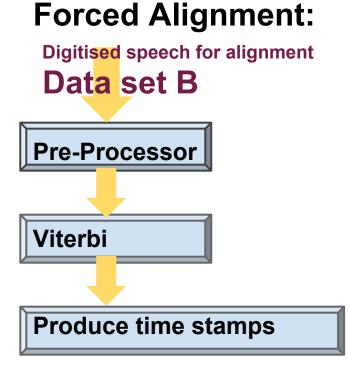




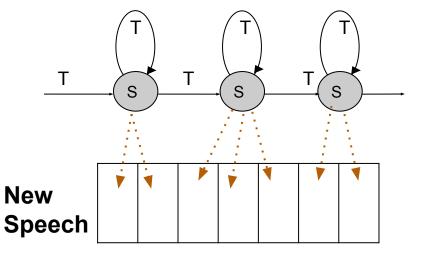
Calculate the acoustic models



State S = Probability of being in a state Transition T = Probability of moving to a state



Acoustic Model



State S = Probability of being in a state Transition T = Probability of moving to a state

Open Source Alignment/Recognition Systems:

Toolkit for development:

HTK htk.eng.cam.ac.uk

Kaldi kaldi.sourceforge.net

Ready systems:

P2FA www.ling.upenn.edu/phonetics/p2fa

Julius julius.sourceforge.jp/en_index.php SPPAS <u>aune.lpl-aix.fr/~bigi/sppas/</u>

Sphinx sourceforge.net/projects/cmusphinx

HTK / P2FA prerequisites

INPUT TO HTK / P2FA

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- Audio sampling rate (16kHz best)
- Lists of phone and silence model names
- Dictionary/Lexicon (ARPAbet)

Orthographic transcription

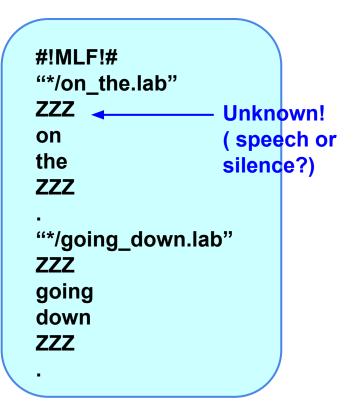
```
ADVERSE AH0 D V ER1 S
ADVERSELY AE0 D V ER1 S L IH0
ADVERSELY AE0 D V ER1 S L IY0
ADVERSITIES AHO D V ER1 S IHO T IHO Z
ADVERSITY AE0 D V ER1 S AH0 T IY0
            B L UW1 IH0 S T
BLUEST
BLUESTEIN
          B L UH1 S T AY0 N
BLUESTEIN
          B L UH1 S T IY0 N
BLUESTINE
              BIUW1STAY2N
ZZZ
              sp
777
              ns
```

Enter all likely pronunciations (BNC_dict.txt)

HTK / P2FA prerequisites

Orthographic transcription:

- Only use subset of ASCII best to keep to letters, numbers, underscore
- "+", "-", "\" etc. have special meanings in HTK, and P2FA
- Make one big file with transcriptions for all the wav file names: "Master Label File" (extension .MLF)



HTK result formats (also .mlf file)

#!MLF!#

.

"*/on_the.rec"

0 200000 sp 62.947620 ZZZ

200000 1100000 AA1 137.138046 ON

1100000 1500000 N 110.194252

1500000 1500000 sp -0.156736 sp

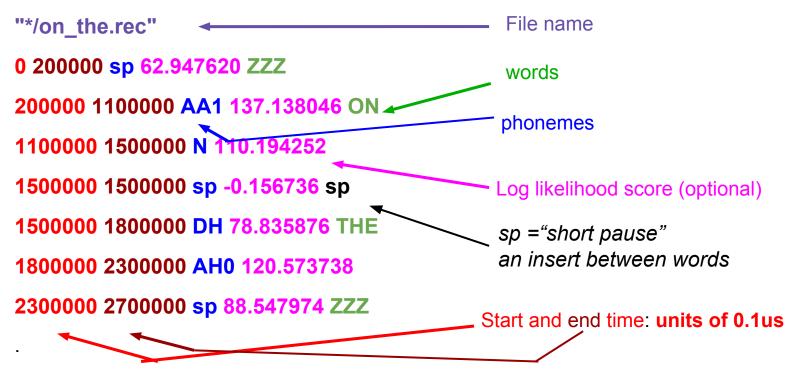
1500000 1800000 DH 78.835876 THE

1800000 2300000 AH0 120.573738

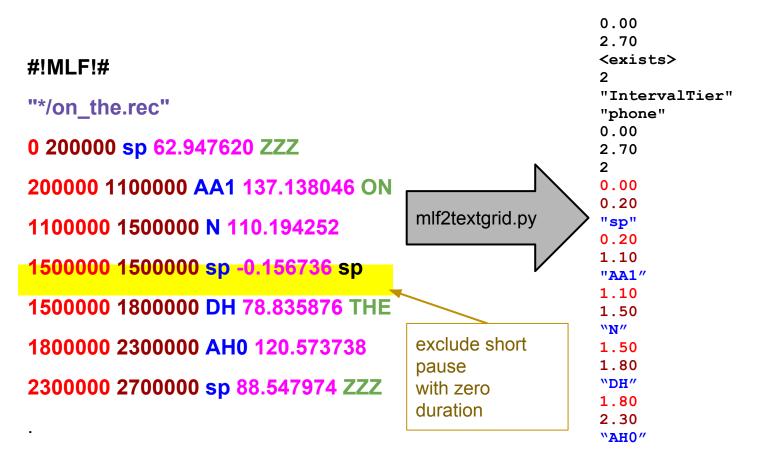
2300000 2700000 sp 88.547974 ZZZ

HTK result formats (*.mlf)





Converting .mlf to .TextGrid



File type = "ooTextFile short"
"TextGrid"

0.01 1.01 <exists> 2 "IntervalTier" "phone" 0.01 1.01 2 0.01 0.055 "ih1" 0.055 1.01 "n" "IntervalTier" "words" 0.01 1.01 1 0.01 1.01 "IN"

0.01 1.01	Audio start and end time
<exists></exists>	
2	Number of tiers
"IntervalTier"	Number of tiers
"phone"	Tior nome
0.01	Tier name
1.01	- Tier start and end time
2	- Number of labels
0.01	- NUMBER OF IADEIS
0.055	Label start and end time
"ih1"	
0.055	- Label
1.01	
"n"	
"IntervalTier"	
"words"	
0.01	
1.01	
1	
0.01	
1.01	
"IN"	

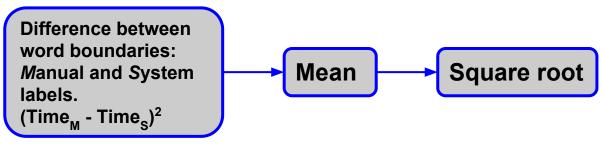
0.01	Audio start and end time
<exists></exists>	
2	Number of tiers
"IntervalTier"	
"phone"	Tier name
0.01	
1.01 -	 Tier start and end time
2	– Number of labels
0.01	
0.055	_ Label start and end time
"ih1"	
0.055	– Label
1.01	
"n"	
"IntervalTier"	
"words"	
0.01	
1.01	
1	
0.01	
1.01	
"IN"	

0.01 1.01	Audio start and end time
<exists> 2 "IntervalTier" "phone" 0.01 1.01 2</exists>	Number of tiers Tier name - Tier start and end time - Number of labels
0.01 0.055 "ih1" 0.055 1.01 "n" "IntervalTier"	Label start and end time Label
"words" 0.01 1.01 1 0.01 1.01 "IN"	

Evaluating alignment quality

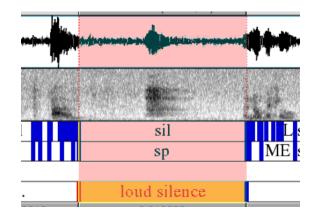
- Listening test (results can be as a percentage)
- Gold standard manual hand labels

Root mean square (rms) difference



• Speech recognition labels - select correct words and compare boundaries

Evaluating alignment quality



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			144 (j)			droberg)			
,	D	0	Ν	0		AH1	Т	AH P n	
,	DON'T		K	NO		SHUT		UP {	
٦	long phonemes								
	1.530000								

- Gross Error
 - \circ Speech in Silence
 - Silence in Speech
 - phoneme/word durations (too long or too short)

