# TOWARDS AUTOMATIC VALIDATION OF PRONUNCIATION VARIANTS FOR SPEECH RECOGNITION AND SYNTHESIS Colleen BEAUMARD, Nicolas PETITJEAN, Tom WYSOCKI Supervisor: Denis JOUVET

## Problematic

Speech synthesis and recognition have become important challenges for the future. They allow us to use voice commands on devices, but also to transcribe speech into text.

Objective: validation of word pronunciation variants by  $\rightarrow$  alignment rules, inconclusive results, research of grapheme-to-phoneme conversion models to improve word prediction performance.



Upside down

# **English pronunciation corpus**

### **French pronunciation corpus**

**CMUdict** - 134 304 words with their phonetic transcription (ARPABET symbols):

CATEGORIZE	K AE1 T AHØ G ERØ AY2 Z	
CATEGORIZED	K AE1 T AHØ G ERØ AY2 Z D	
CATEGORIZES	K AE1 T AHØ G ERØ AY2 Z IHØ Z	
CATEGORIZING	K AE1 T AHØ G ERØ AY2 Z IHØ NO	G
CATEGORY	K AE1 T AHØ G AO2 R IYØ	
CATELLI	K AHØ T EH1 L IYØ	
CATENA	K AHØ T IY1 N AHØ	
CATER	K EY1 T ERØ	
CATERED	K EY1 T ERØ D	
CATERER	K EY1 T ERØ ERØ	

0 : no stress - 1 : primary stress - 2 : secondary stress

**BDLex** - 337 550 words with their phonetic transcription (in SAMPA):

#### Right side

ballottement	bal0/t@ma~	tnemettollab	a∼ m@t0/lab
ballottements	b a l O∕ t @ m a~	stnemettollab	a~ m@t0/lab
balsa	balza	aslab	azlab
balsas	balza	saslab	azlab
bambochade	ba∼bO/Sad	edahcobmab	d a S 0/ b a~ b
bambochades	ba∼bO/Sad	sedahcobmab	daSO/ba~b
bambocheur	b a~ b O/ S 9 R	ruehcobmab	R 9 S 0/ b a~ b
bambocheurs	b a~ b O/ S 9 R	sruehcobmab	R 9 S 0/ b a~ b
bambocheuse	b a∼ b 0/ S 2 z	esuehcobmab	z 2 S 0/ b a~ b
bambocheuses	b a~ b 0/ S 2 z	sesuehcobmab	z 2 S 0/ b a~ b

#### **Processing of the corpus**

Train Development Test 15% 15% 70%

For the English lexicon, *test* (with stress) and *test bis* (without stress)

For the French lexicon, each file in right side and upside down versions.

For the models of the **neural network** approach and the **statistical** approach, we evaluate the prediction performances of the files with 2 scores: WER = Word Error Rate, the error rate per word and PER = Phoneme Error Rate, the error rate per phoneme.

## Alignment rule approach

Alignment of graphemes and phonemes using predefined rules. If insufficient, add new rules: columbarium  $\rightarrow k O/l o \sim b a R j O m$  $\rightarrow$   $\sim$ um breitschwanz  $\rightarrow$  b R a j t S v a t s  $z \rightarrow ts$ fuel  $\rightarrow$  f j u l ue  $\rightarrow$  ju

#### Résultats

Do not detect mispronunciations  $\rightarrow$  even if incorrect predictions, alignment possible: abstin abstinent

Neural network approach

It is given a sequence of letters as input and the network provides a sequence of phonemes as output.



#### Software : OpenNMT

- Free software for neural networks;
- Use of a YAML configuration file (where to find the files needed for training).

### **Statistical approach**

Statistical method for predicting a phoneme from a sequence of letters.

$\begin{array}{l} \text{``mixing''} \\ [m1ks10] \end{array} =$	m [m]	i [I]	x [ks]	i [I]	n [ŋ]	g
----------------------------------------------------------------	----------	----------	-----------	----------	----------	---

#### Software : Sequitur

• Each element is linked to a sequence of letters and a sequence of phonemes, sequences that allow the reconstruction of the word with its pronunciation  $\rightarrow$  Sequence Joined;

• Based on a grapheme-phoneme alignment.

### **English and French lexicon results**

#### English Lexicon

# Models combination results (French lexicon)

#### **Combination of the models**

Modele	PER	WER
With consideration of stress	10.93%	41.58%
Without taking stress into consideration	8.71%	36.86%

#### French Lexicon

Modele	PER	WER
OpenNMT right side	0.59%	3.05%
OpenNMT upside down	0.76%	4.24%
Sequitur right side	0.53%	3.15%
Sequitur upside dow	0.50%	2.92%

	Number of identical	Number of	<b>Proportion of</b>	
	predictions among	predictions out	configurations	
	the models	of 51612	of a total in percent	
	4	48 165	93.32%	
	3-1	2098	4.06%	
	2-2	917	1.78%	
	2-1-1	424	0.82%	
	1-1-1-1	8	0.02%	
Examples:		1		
<b>3 - 1</b> : a	blytjo∼ ably	sjo∼ ably	sjo∼ ablysjo	
<b>2 - 1 - 1</b> : a	$\sim$ m 9 l a m 9	$1 a \sim m$	21 a~ m $21$	

#### Conclusion

• Rules not sufficient to validate pronunciation variants;

• Neural network and statistical models perform well, slightly better when combined:  $(\mathbf{PER} : 0.47\% \text{ and } \mathbf{WER} : 2.75\%).$