

# Articulatory normalization via imitation strategy in phone classification task



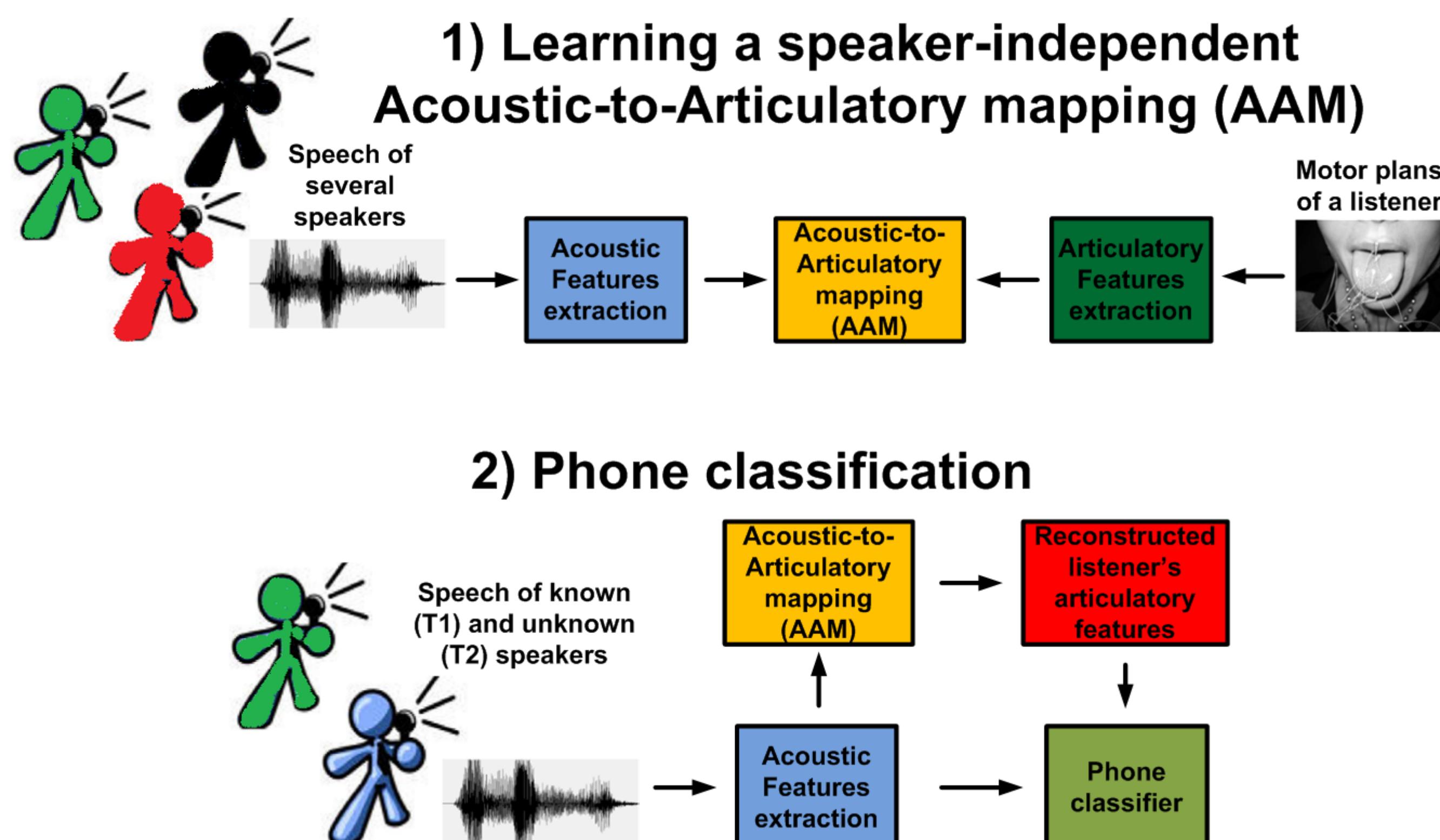
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## Introduction

Automatic Speech Recognition (ASR) systems classically apply normalization strategies that minimize inter-speaker variability by explicitly removing speakers' peculiarities ([1], [5]) or adapting different speakers to a reference model ([7]) or creating a compact adaptable and robust models ([2]) without exploiting speakers' variations in learning and recognition processes.

## Approach

We propose a speaker normalization strategy that uses measured articulatory information and test it in a phone classification task ([4]).



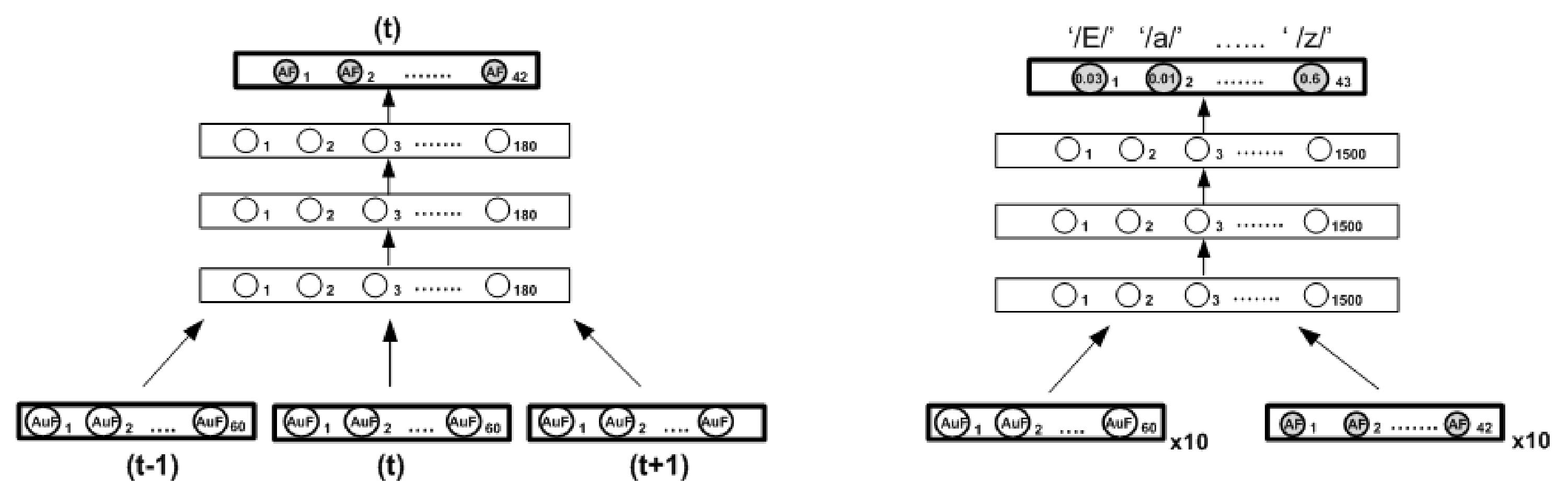
## Questions

- 1)** Does the Articulatory normalization produce a more accurate phone classification w.r.t.
  - no normalization strategy ?
  - a corresponding acoustic normalization strategy?
- 2)** How can articulatory similarities affect the phone discrimination?

## Material and Methods

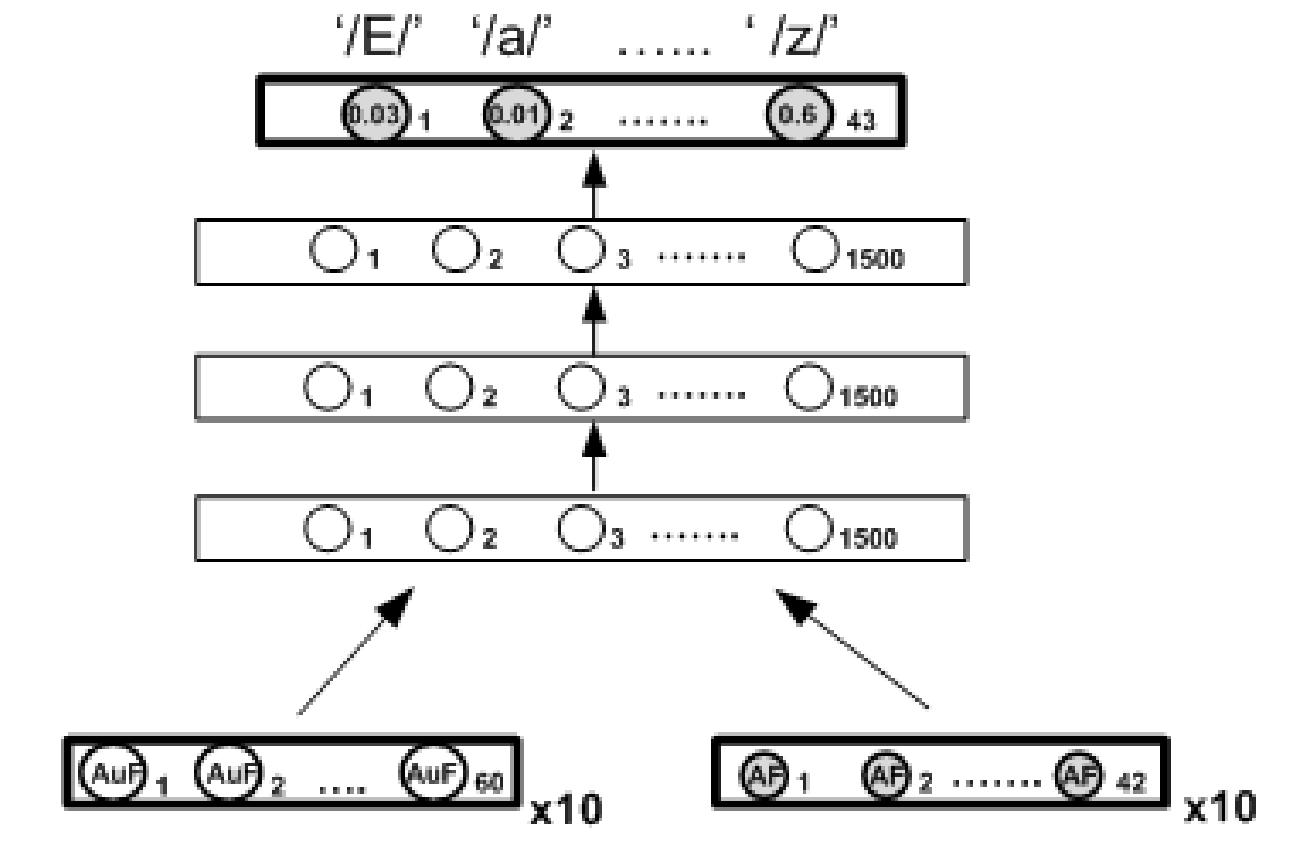
**Dataset:** Italian corpus of simultaneous recording of speech and the trajectories of 7 articulators (UI, LI, LL, UL, TT, TB, TD) tracked by electromagnetic articulography. It includes 3120 single word utterances pronounced by 5 female subjects ([6]).

## Acoustic-to-Articulatory Mapping\* 4-layer DNN ([3])



\*An identical DNN is also used during acoustic normalization to perform Acoustic-to-Acoustic Mapping.

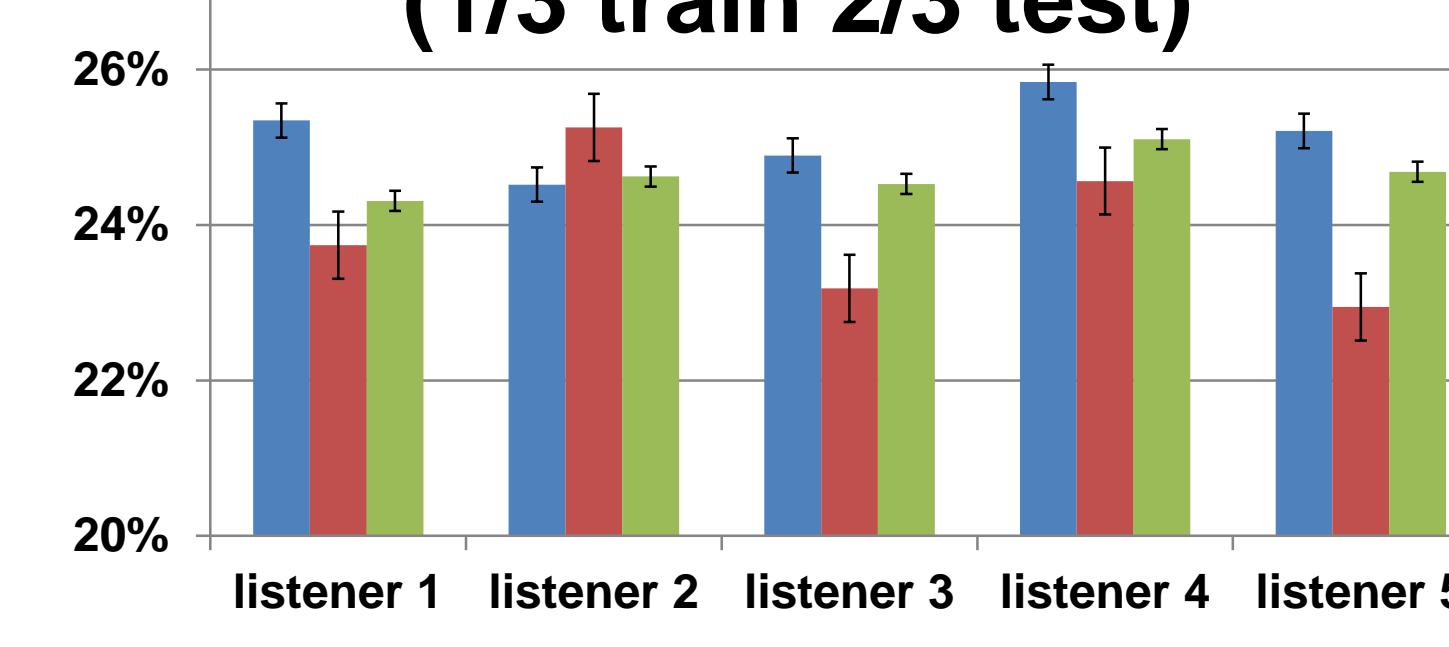
## Phone classifier



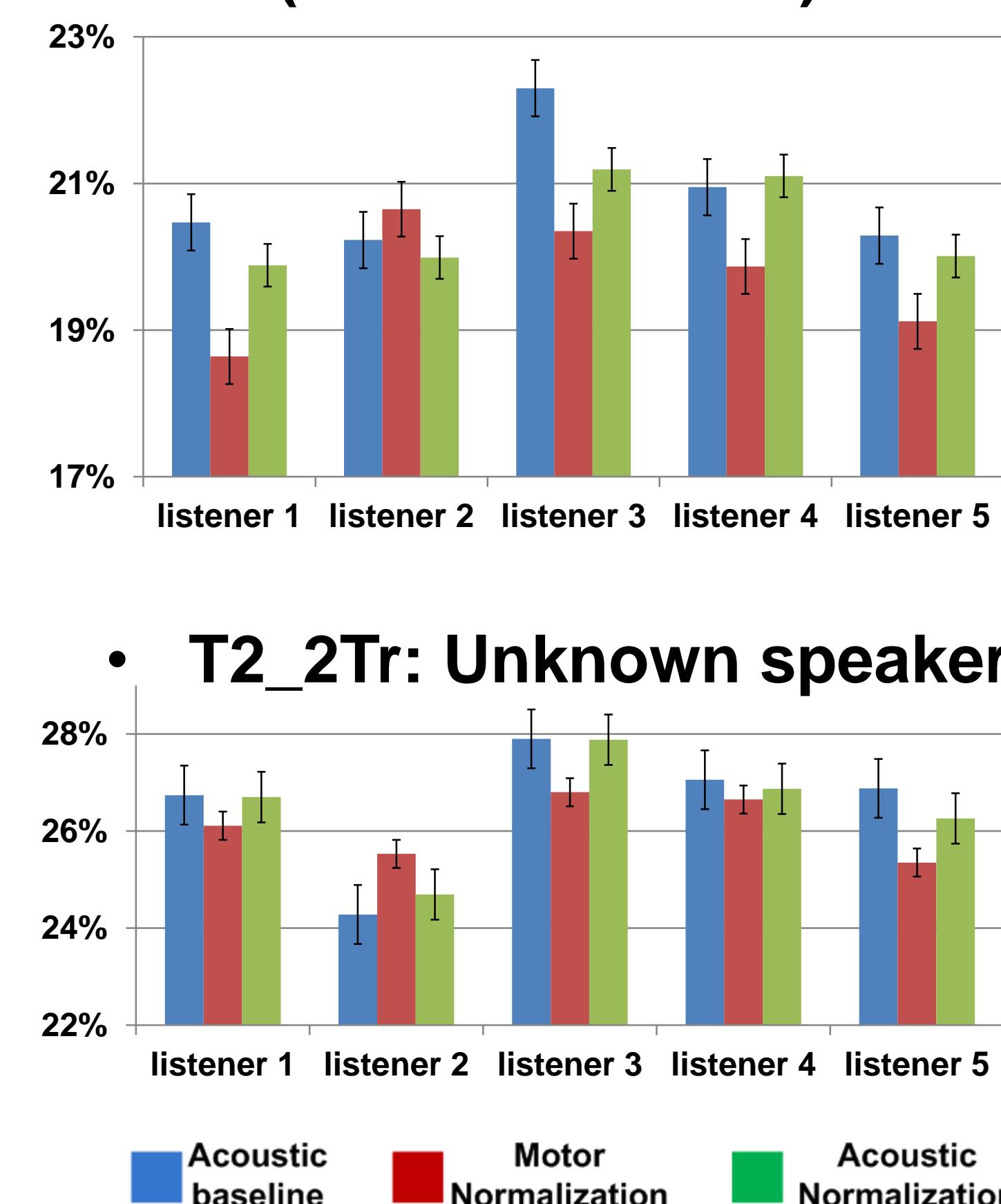
## Results

### 1) Phone Classification Error

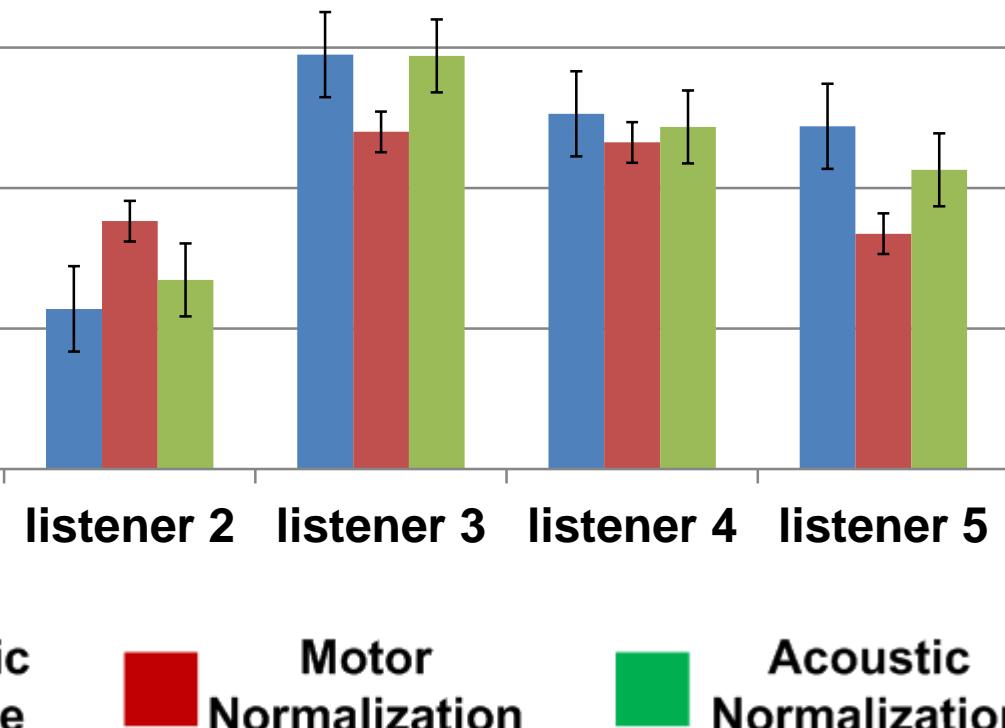
- T1\_1Tr: Known speaker (1/3 train 2/3 test)



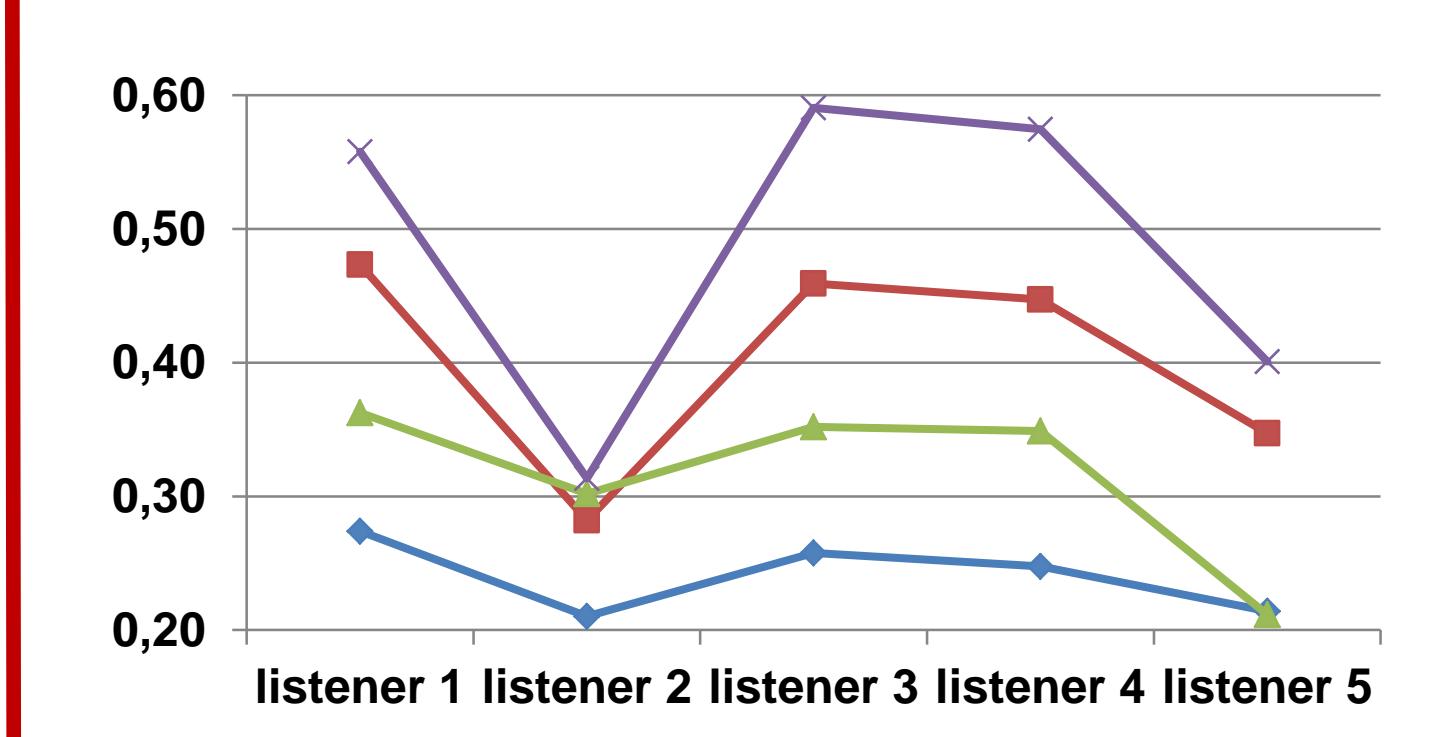
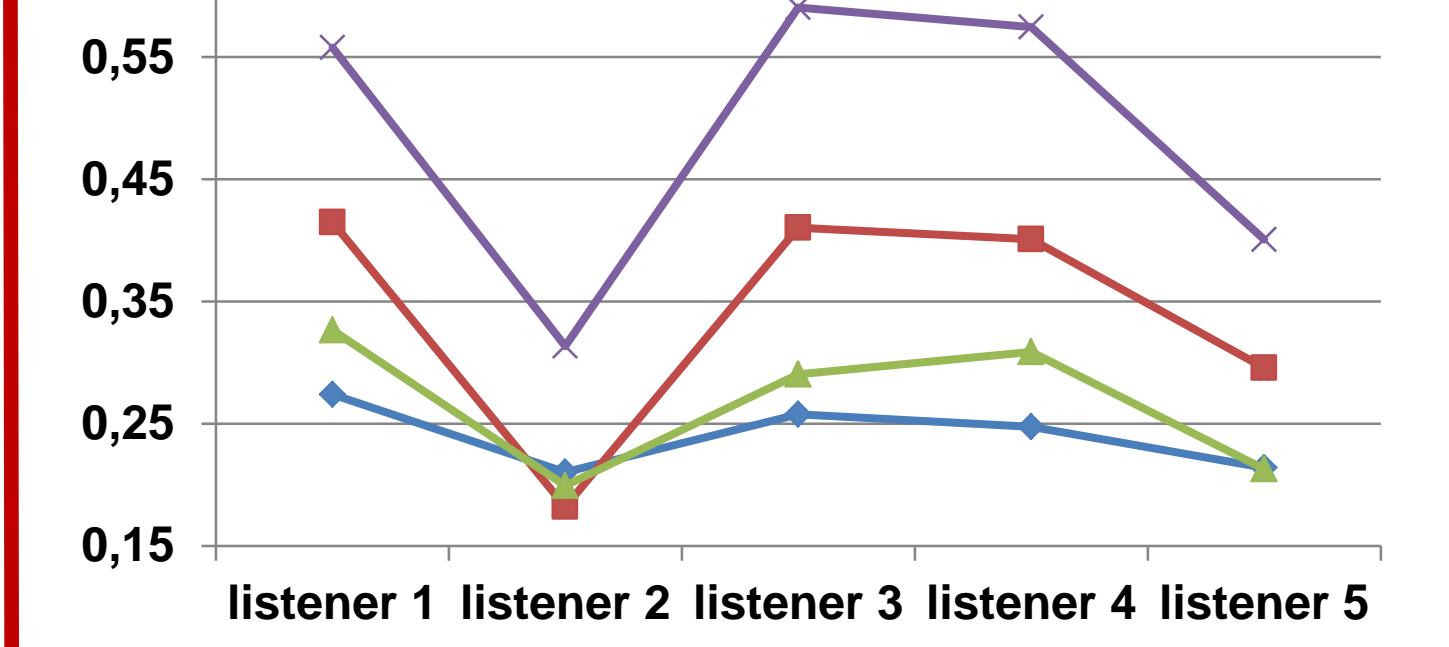
- T1\_2Tr: Known speaker (2/3 train 1/3 test)



- T2\_2Tr: Unknown speaker



## Pearson Correlations



Legend:  
X Intra-Listener AFs  
■ Listener actual AFs - Listener reconstructed AFs  
▲ Listener - speaker actual AFs  
◆ Speaker actual AFs - Listener reconstructed AFs

### 2) Articulatory similarities

PER reduction/	T1_1Tr	T1_2Tr
Pearson corr X	48% (0.032)	42.5% (0.06)
Pearson corr ■	54.5% (0.013)	41.9% (0.07)
Pearson corr ▲	28% (0.23)	4% (0.8)

## References

- [1] Anastasakos, T., Kubale, F., Makhoul, J., Schwartz, R., "Adaptation to new microphones using tied-mixture normalization", in Proc. of IEEE International Conference on Acoustics, Speech and Signal Processing, Adelaide, SA, 1994, pp. 433-436.
- [2] Anastasakos, T., McDonough, J., Schwartz, R., Makhoul, J., "A Compact Model for Speaker-Adaptive Training", in Proc. of Spoken Language Processing, Philadelphia, PA, 1996, pp. 1137-1140.
- [3] Badino, L., Canevari, C., Fadiga, L., Metta, G., "Deep-Level Acoustic-to-Articulatory Mapping for DBN-HMM Based Phone Recognition", in Proc. of IEEE spoken language technology workshop, 2012, Miami.
- [4] Canevari, C., Badino, L., D'Ausilio, A., Fadiga, L., Metta, G., "Modeling speech imitation and ecological learning of auditory-motor maps", Frontiers in psychology, June 2013, 4.
- [5] Eide, H., Gish, H., "A parametric approach to vocal tract length normalization", in Proc. of IEEE International Conference on Acoustics, Speech and Signal Processing, Atlanta, GA, 1996, pp. 346-349.
- [6] Grimaldi, M., Gili Fivela, B., Sigona, F., Tavella, M., Fitzpatrick, P., Craighero, L., et al., 'New technologies for simultaneous acquisition of speech articulatory data : 3D articulograph, ultrasound, and electroglottograph', in Proc. of Language Teching, 2008, Rome, Italy.
- [7] Huang, X., Acero, A., Hon, H., W., "Spoken Language Processing", Upper Saddle River, New Jersey, NJ: Prentice-Hall.