

# Speech Recognition Israel

**Meetups**

**Technical Facebook group!**

**Community**

# Diarization in practice

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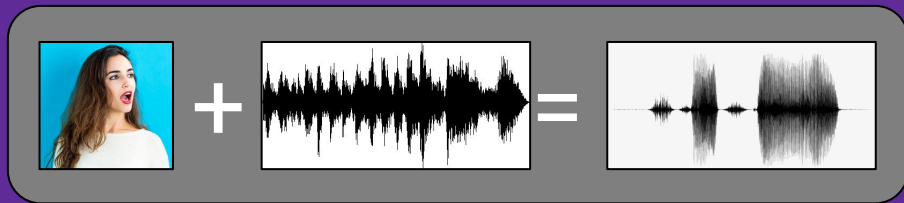
***@YoavR7***



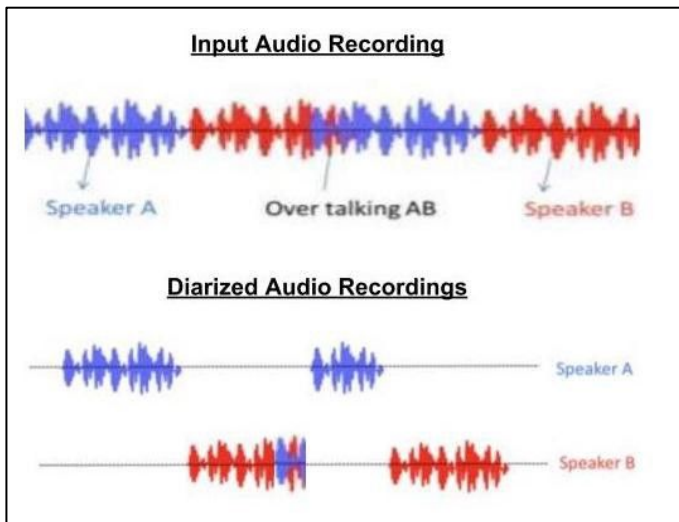
***@yoavramon***



## Audio-Visual Robust Speech Solution



# The Problem



## Why it's important?

- Speaker identification
- Speech Recognition
- Real-world data analysis

## “Who spoke when”

Diarization  $\neq$  Speaker Separation

# How good are we?

## Diarization with X-vectors (Snyder, 2018)

	DER
Without Oracle	8.39
With Oracle	7.12



## Callhome Challenge



1997, 120 Calls, 2 Channels, Telephony  
60 Hours +-

# So... Is it solved?

“While state-of-the-art diarization systems perform remarkably well for some domains (e.g., conversational telephone speech such as CallHome), **as was discovered at the 2017 JSALT Summer Workshop at CMU, this success does not transfer to more challenging corpora** such as child language recordings, clinical interviews, speech in reverberant environments, web video, and **speech in the wild**” (Church et al., Feb. 2018)

## DIHARD Challenge

Interspeech 2018, September

### Development Test

- Only 19 Hours
- 9 Domains  
(*Child language, Supreme Court, Clinical interviews, Radio interviews, Map tasks, Sociolinguistic interviews, Meeting speech, Audiobooks, YouTube videos*)
- Single Channel
- 5 Minutes per sample

### Evaluation Test

- 21 Hours
- 3 Different domains  
(*Sociolinguistic interviews, Meeting speech, Restaurant conversation*)
- Single Channel
- 5 Minutes per sample

# So... Is it solved?

## **Diarization is Hard: Some Experiences and Lessons Learned for the JHU Team in the Inaugural DIHARD Challenge**

*Gregory Sell, David Snyder, Alan McCree, Daniel Garcia-Romero, Jesús Villalba, Matthew Maciejewski, Vimal Manohar, Najim Dehak, Daniel Povey, Shinji Watanabe, Sanjeev Khudanpur*

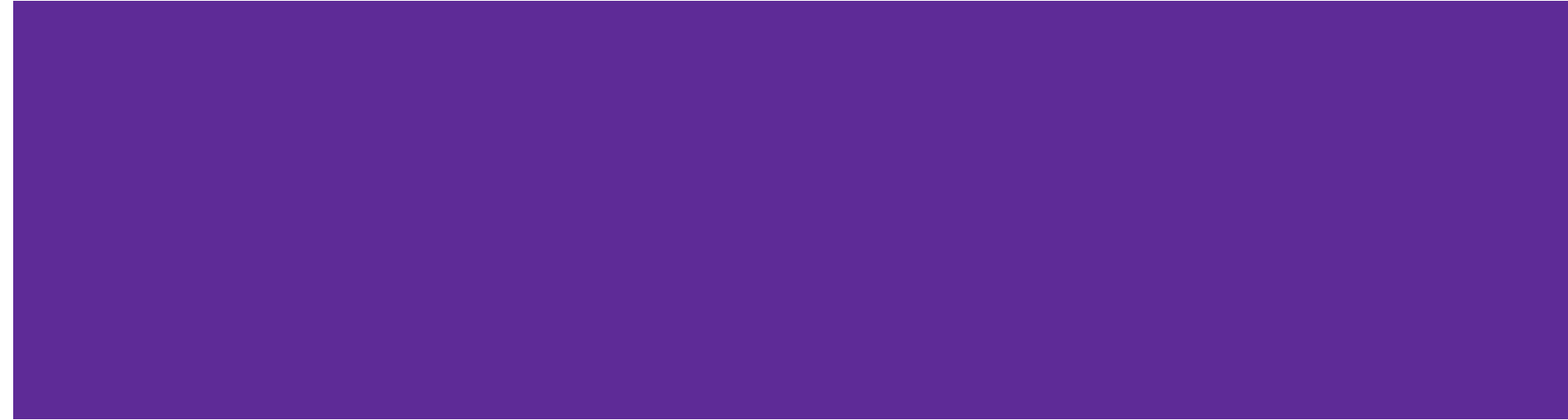
Center for Language and Speech Processing & Human Language Technology Center of Excellence  
Johns Hopkins University, USA

System	Track 1	
	Dev DER	Eval DER
All same speaker	35.97	39.01
Initial System	26.58	31.56
i-vector, no VB	21.74	28.06
x-vector, no VB	20.03	25.94
Fusion, no VB	19.54	25.50
i-vector, with VB*	19.69	25.06
x-vector, with VB*	18.20	23.73
Fusion, with VB*	18.17	23.99

***Far from being solved...***

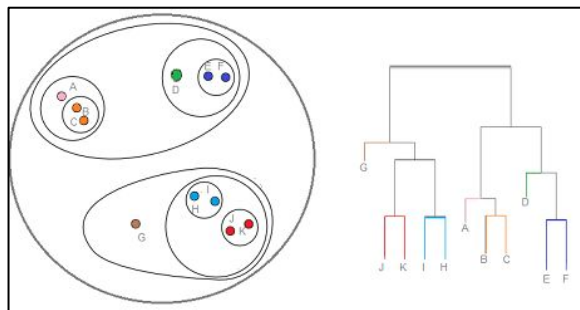
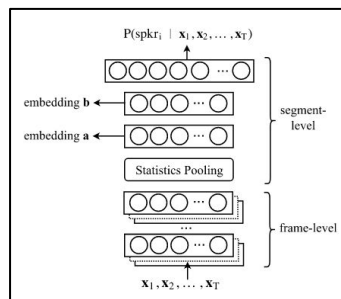
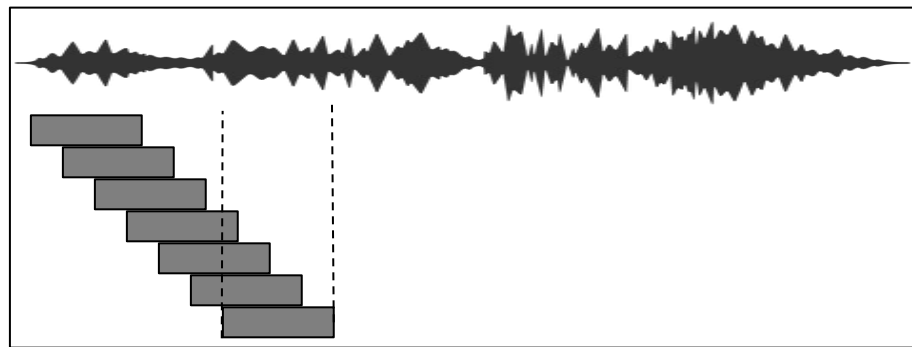
# Implementation

(So.... How do I do it?)





# The framework:

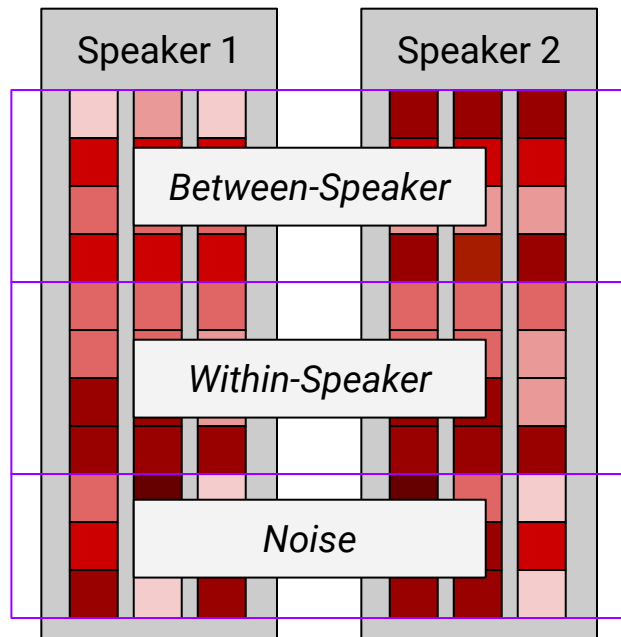


Next Talk

Normalization

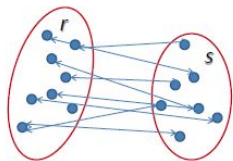
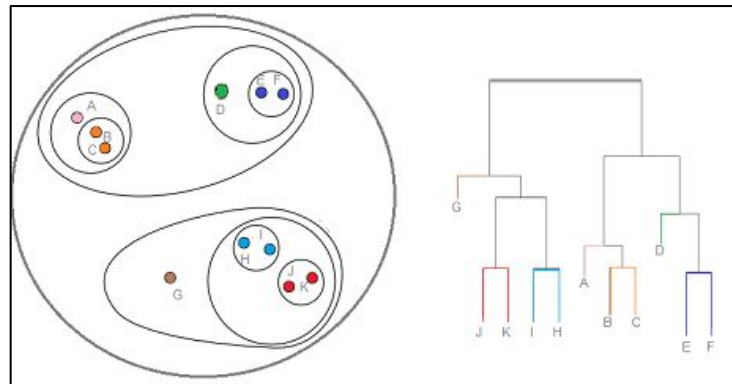
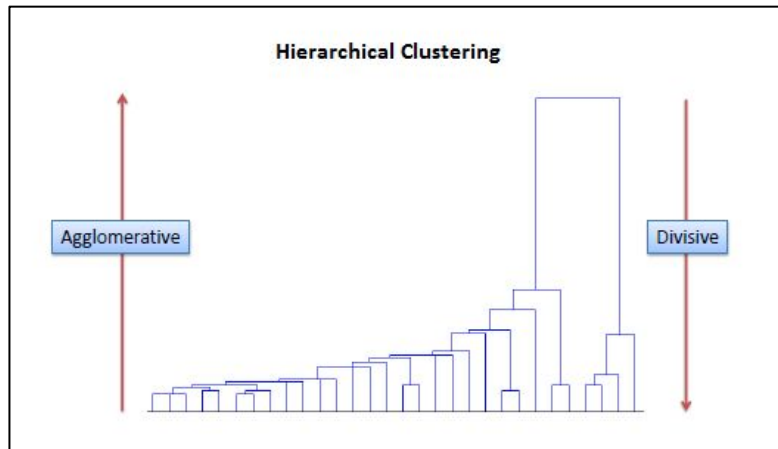
Clustering

# PLDA Normalization



$$\Phi = \mu + Vy + Ux + \epsilon$$

# Agglomerative Clustering



$$L(r, s) = \frac{1}{n_r n_s} \sum_{i=1}^{n_r} \sum_{j=1}^{n_s} D(x_{ri}, x_{sj})$$

**With Oracle: Until we have enough speakers**

**Without Oracle: With Threshold**

# You can do it with Kaldi!



- It's easy to use
- There are pretrained models
  - Although you'll want to adapt them
- There is support in clustering with oracle.
- It's easy to integrate inside speech processing pipeline

<https://towardsdatascience.com/speaker-diarization-with-kaldi-e30301b05cc8>

# Challenges

- Real Time
- Not good enough vectorization
- Cross-Domain robustness (20% DER)
- Estimating the number of speakers might be hard
- Dimensionality reduction might create too-sparse vectors

# Advantages

- Really Active research...
- Not so hard to implement.
- “Building Block” architecture leaves place to innovation.
- The value to ASR Systems is huge, even with relatively poor DER
- “Transfer Learning” on PLDA is easy



# Thank you!

Questions?

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