

# USC Viterbi School of Engineering

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## Assessment of a child's engagement using sequence model based features

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## Motivation & Introduction

#### **Hypotheses**

- Engagement behavior of children in reflected in the speech of the participants in conversation
- Local patterns of features also carries information about the engagement behavior apart from global statistics

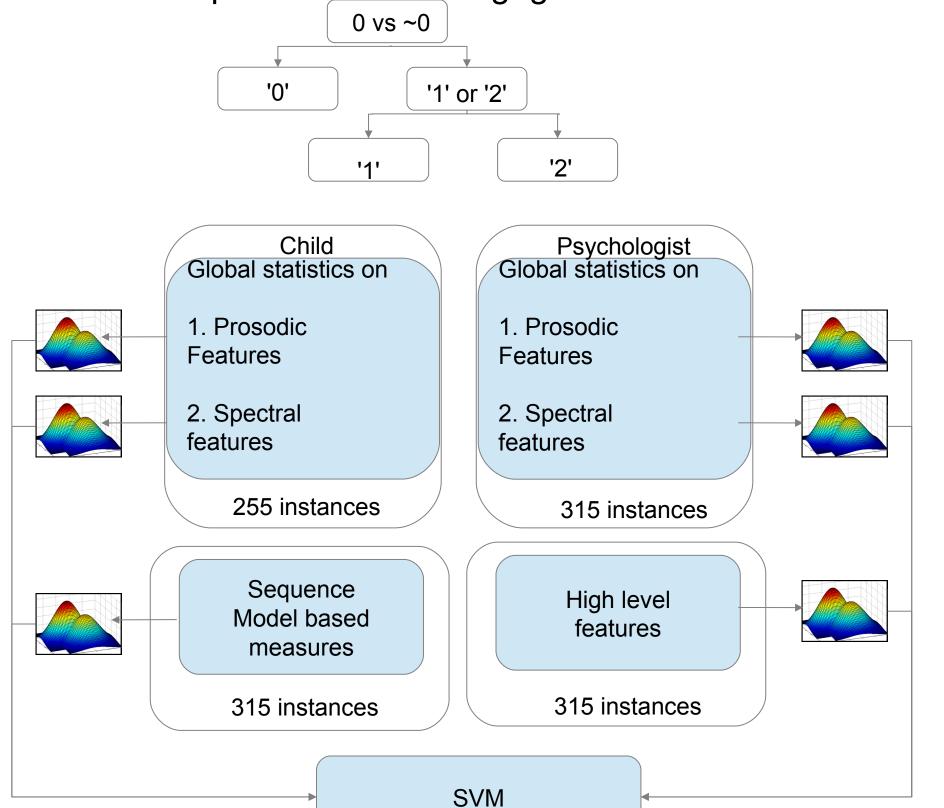
#### Goal

- Develop a purely speech based system to predict the engagement level of children in their interaction with a psychologist
- -Exploring the phenomenon of joint attention and other developmental aspects in children using engagement

## Experiment

#### Database:

- 65 sessions from R-abc database
- Predict the perceived child engagement from audio stream



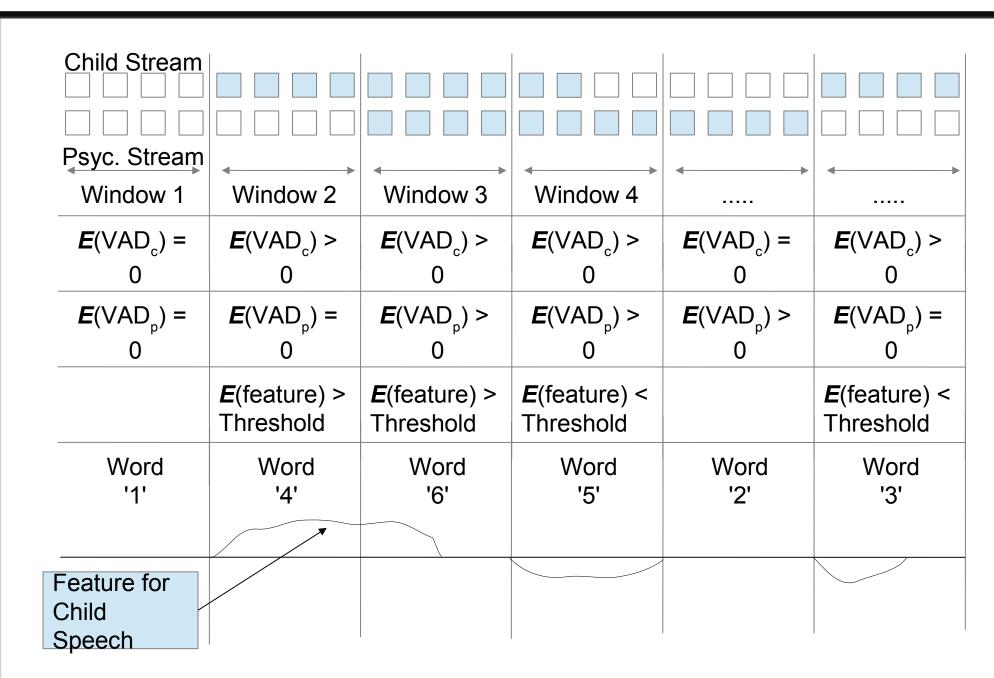
### **Feature sets**

- Global Statistics

Psychologist speech	Prosody(F0, jitter, shimmer, intensity)	Mean, std, kurtosis, Quantiles				
	Spectral (MFCC)					
Child speech	-do-	-do-				
High level features	Sub-session length (normalized task-wise), Child speech length, Number of (#) overlaps, Total Speech activity, #Psychologist utterance					

- Sequence based features
  - Capture local patterns in speech

## Sequence Features



- Train a sequence model on the training set
  - Normalized count of an "n-gram" over the training set

$$\frac{\mathrm{SM}(w_k/w_{k-1},..,w_{k-n+1}) =}{\sum_{t \in \mathrm{Train}} \sum_{\mathrm{Set}} \#(w_k^{S_t}/w_{k-1}^{S_t},..,w_{k-n+1}^{S_t})}{\sum_{t \in \mathrm{Train}} \sum_{\mathrm{Set}} N_t}$$

- Get the propability on the test set
  - # of occurrence of an "n-gram" X above normalized count

$$\begin{aligned} \mathbf{M}(w_k^S/w_{k-1}^S,..,w_{k-n+1}^S) = \\ \mathbf{SM}(w_k/w_{k-1},..,w_{k-n+1}) \times \#(w_k^S/w_{k-1}^S,..,w_{k-n+1}^S) \end{aligned}$$

### Results

Feature Source	Selected model	Class-wise accuracy			Uw. accura -cy(2 class)	Uw accur -acy (3 class)	
		'0' (%)	~'0' (%)	'1' (%)	'2'		
Global statistics on features	High level features	78.6	63.7			71.1	
		78.6		30.0	38.7		49.1
based on SM	None,child	77.2	69.2			73.2	
	pitch, psyc.pitch, child intensity	77.2		33.3	32.3		47.6
Fused Model	None,child pitch, psyc.pitch, child intensity, high level features	77.2	69.2			73.2	
		79.0		33.3	41.9		51.4

- Apart from global statistics, local patterning of features also informative in capturing engagement levels

#### Future Work

- Incorporate visual features
  - Better sequence models
    - Smoothing techniques