

Multimodal Sequential Modeling of Task-Mediated Frustration

*Intermediate Fusion for
High and Low Level Features*

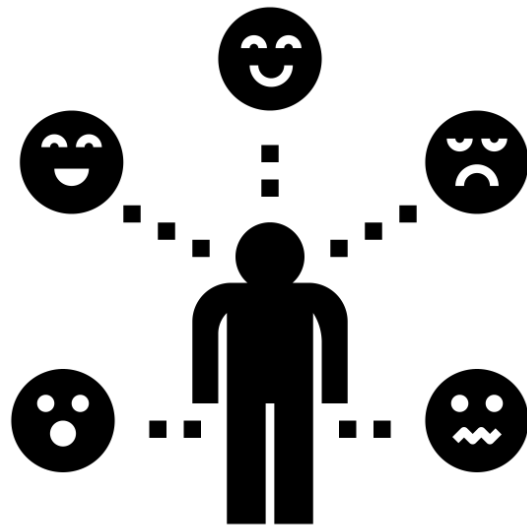
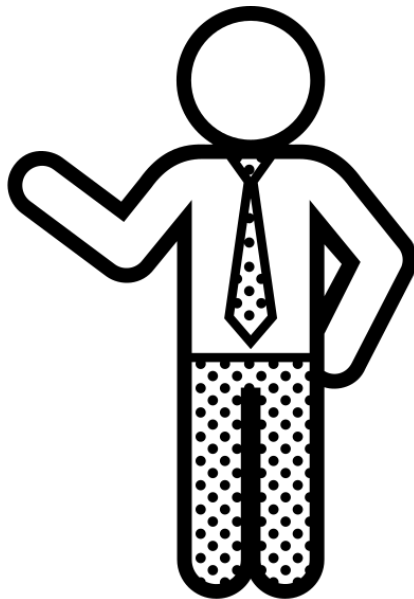
Michael Peechatt

ENGL.584.01/684.01 - Speech Processing
II

CLaSP Lab + Graphics Lab @ **RIT**



What is Multimodality?



Motivation

- ❑ **Cooperation is under-explored in affective computing**
 - ❑ The pandemic increased Zoom usage in an collaborative context
- ❑ **Human perception combines low and high level features**
 - ❑ Can our machine learning models reflect this?
- ❑ **RQ1:** *Is or high or low level features better for identifying frustration?*
- ❑ **RQ2:** *Does, on average, fusing predictions improve overall performance?*

Builder *in CLaSP Lab*



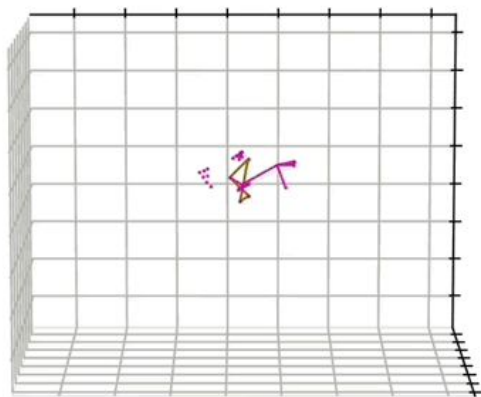
Setup

Instructor *in Whisper Room*

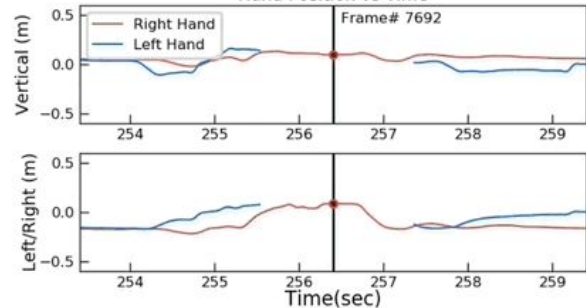


Builder Modalities

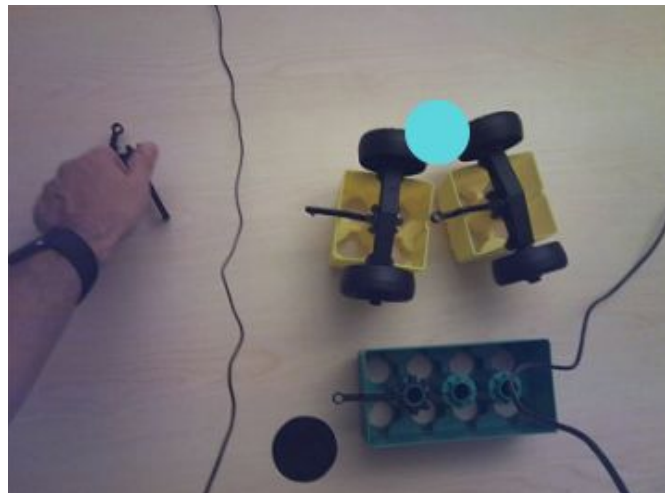
Session: sesh_2022-10-14_18_02_51



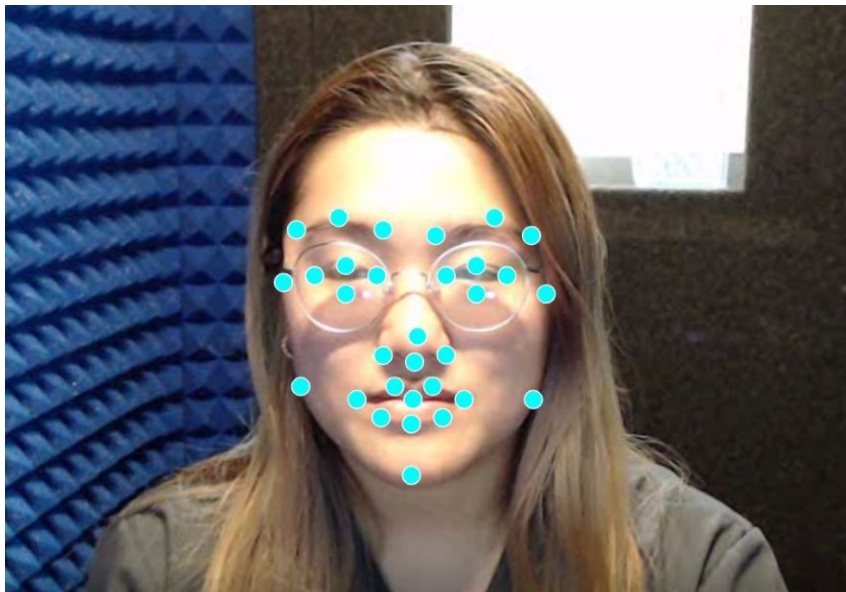
Hand Position vs Time



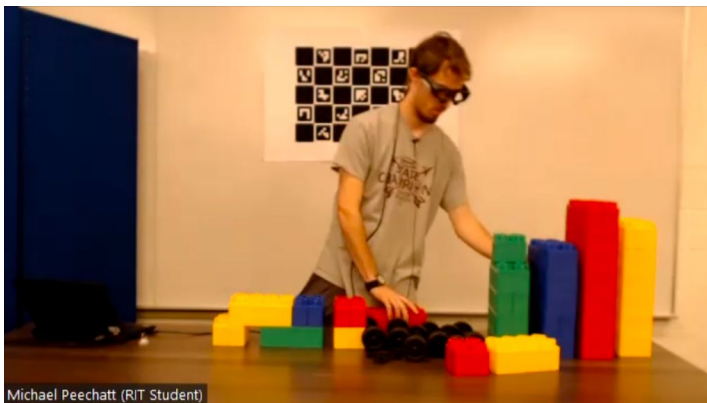
github.com/jonmatthis/freemocap || freemocap.org



Instructor Modalities



Common Modalities



Annotation

Michael Peechatt

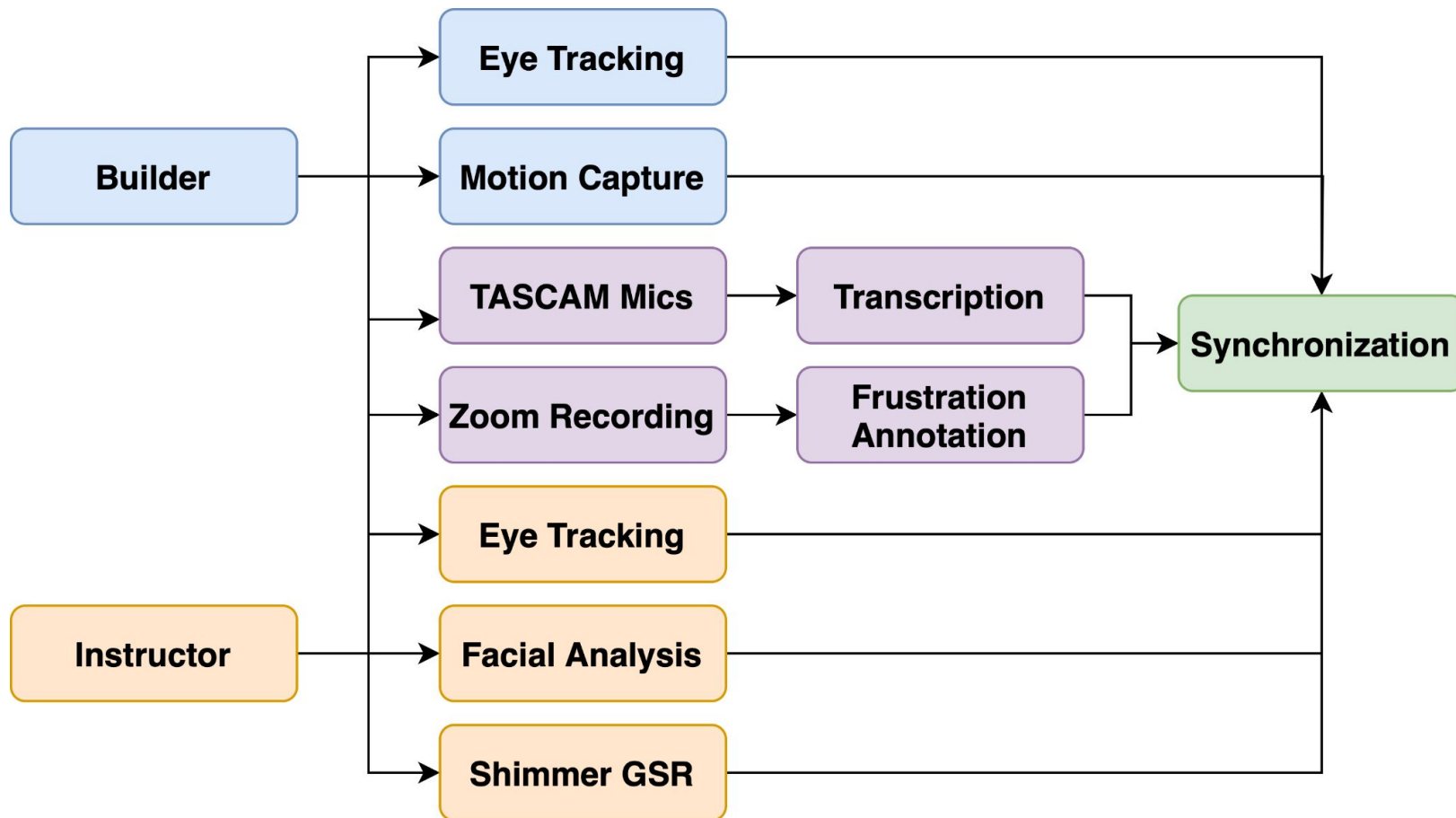
Michael Peechatt (RIT Student)

Not At All Frustrated

Slightly Frustrated

Very Frustrated

Extremely Frustrated



MULTICOLLAB Dataset

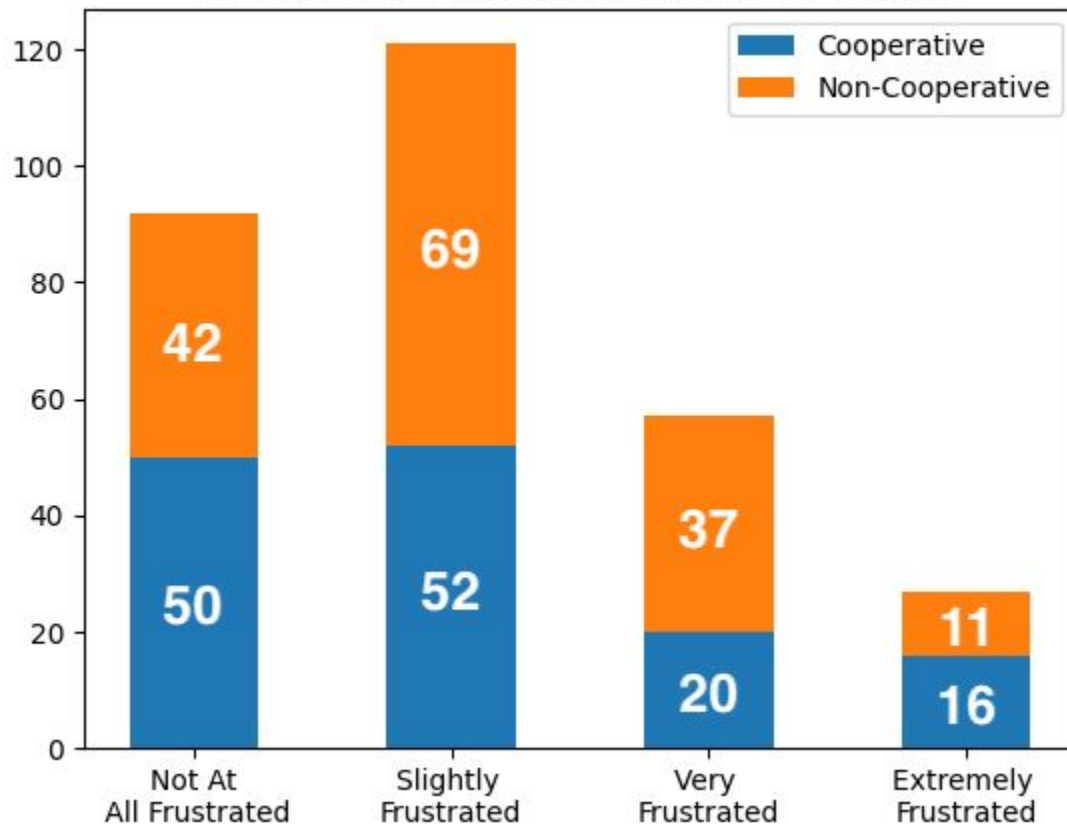
❑ 48 subjects (24 builder-instructor groups)

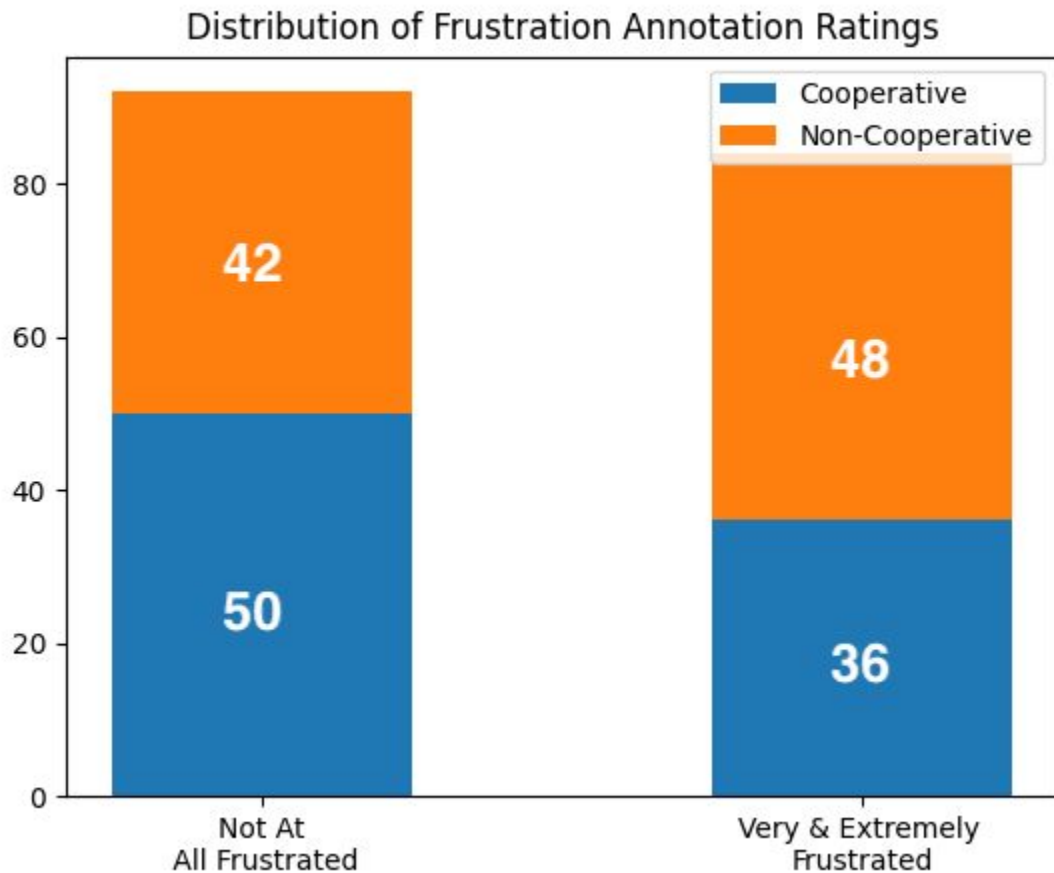
- ❑ 42% Female, 56% Male, 2% undisclosed
- ❑ 8 groups had different gendered interactions
- ❑ 16 groups had same gendered interactions

❑ Ethnicity Distribution

- ❑ 2.1% Southeast Asian, 8.4% African-American, 10.5% Hispanic, 39.6% Asian, and 37.5% Caucasian (1.8% Undisclosed)
- ❑ 20.8% ESL speakers, 79.2% native English speakers
- ❑ 3 of 24 groups were mix of non-native and native interactions

Distribution of Frustration Annotation Ratings





Feature Extraction

Voice Features

Intensity (dB)

F0 (Hz)

Facial Features

Brow Furrow

Chin Raise

Lid Tighten

Lip Corner Depressor

Eye Gaze Features

Saccade Duration

Saccade Peak Velocity

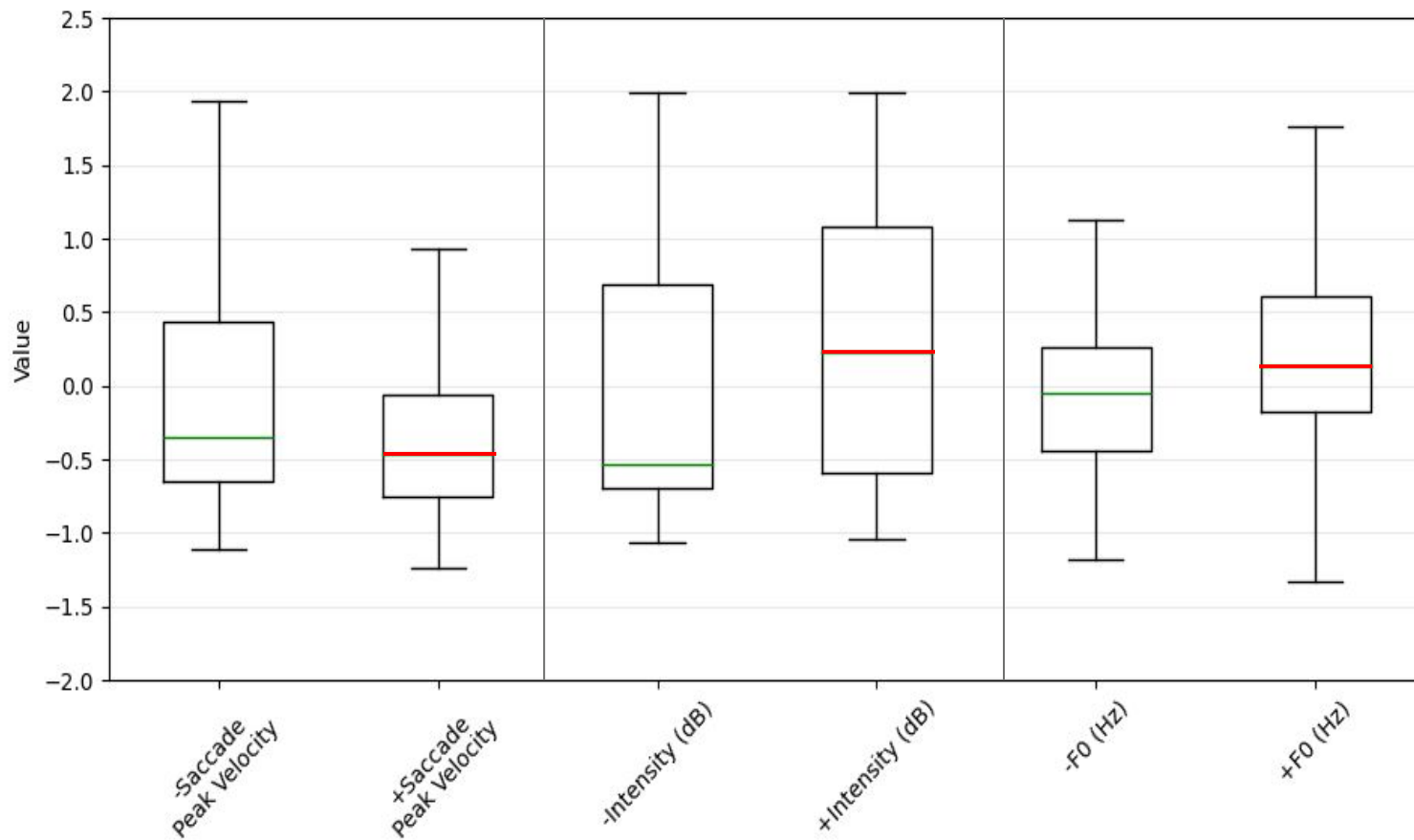
Fixation Dispersion

Fixation Duration

Gaze Velocity

Biophysical Features

GSR Conductance



Dataset Shape

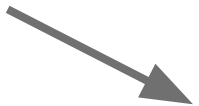
$$\begin{matrix} & m_1 & m_2 & \dots & m_n \\ \left[\begin{array}{c} t_1 \\ t_2 \\ \vdots \\ t_n \end{array} \right] & & & & \end{matrix} \quad \begin{matrix} t_1 \\ t_2 \\ \vdots \\ t_n \end{matrix}$$

Dataset Shape

$$\begin{array}{cccc}
 m_1 & m_2 & \dots & m_n \\
 \left[\begin{array}{c} t_1 \\ t_2 \\ \vdots \\ t_n \end{array} \right] & & & \left[\begin{array}{c} t_1 \\ t_2 \\ \vdots \\ t_n \end{array} \right]
 \end{array}$$

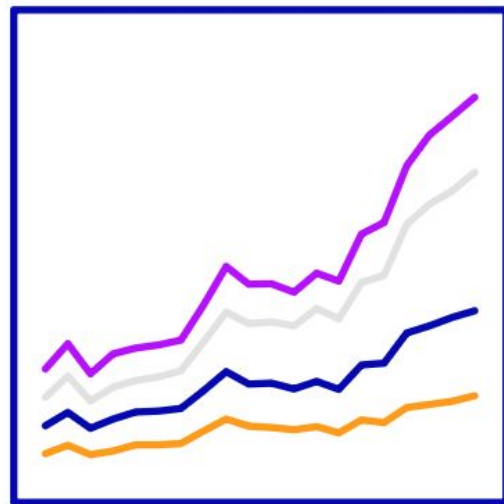
$$||\{m_1, m_2, \dots, m_n\}|| \cdot t_n$$

↓ flatten

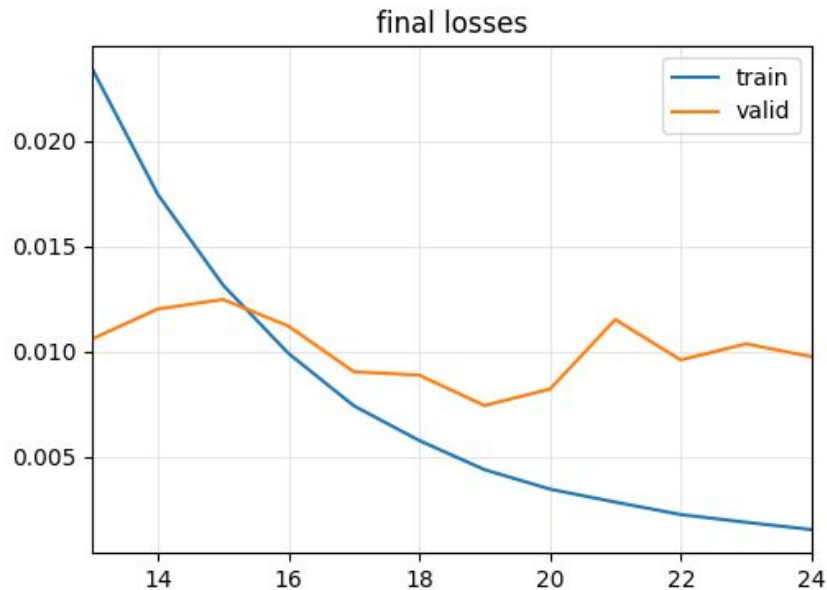
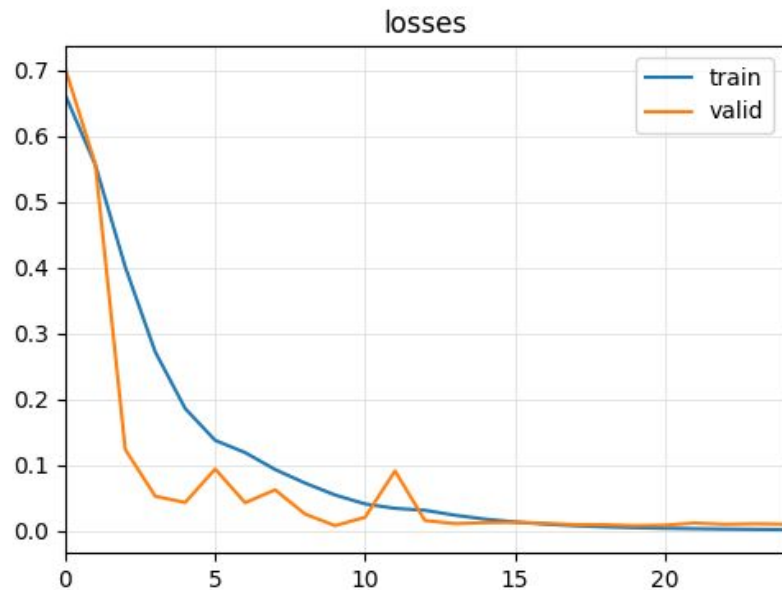


$$\begin{array}{cccc}
 m_{1_t_1} & m_{2_t_1} & \dots & m_{n_t_n} \\
 \left[\begin{array}{c} \\ \\ \\ \end{array} \right] & & & \left[\begin{array}{c} \\ \\ \\ \end{array} \right]
 \end{array}$$

tsai



state-of-the-art deep learning
for time series and sequences



$t_{\text{win}} = 10000$ milliseconds, $t_n = 20$
avg. accuracy = 0.523

30NI_transcript

start	end	word
1.28	1.62	okay
2.52	2.78	so
3.86	3.96	do
4.0	4.04	i
4.14	4.48	start
5.96	6.08	all
6.08	6.34	right
7.68	7.9	so
9.04	9.1	the
9.18	9.4	first
9.46	9.58	thing
9.62	9.74	we're

27CI_transcript

start	end	word
0.48	0.48	yeah
7.98	8.1	oh
8.32	8.52	okay
9.52	9.62	um
10.12	10.24	to
10.36	10.6	take
11.06	11.2	the
11.68	11.92	four
11.96	12.08	by
12.14	12.28	two
12.36	12.68	yellow
12.7	12.9	one

*fast*Text

$$\frac{\sum_{w \in W} f_t(w)}{|W|}$$

Rating	Timestamp	Group	Utterance
3	315680.0	18N	those not connectors beside the yellow block
3	435720.0	19C	color in another way you are you are making yeah
3	159650.0	22N	on small blue one no select
3	403370.0	39C	
3	257850.0	22N	it vertically not horizontally
3	355950.0	24N	no no no no the other rectangle yep go back to that one yep put
3	318980.0	26N	no you want it to
3	431780.0	26N	between the two blocks there you go that well you want it like back
3	431540.0	28N	all
3	39950.0	31C	no no this too yeah can you show me no no
3	44350.0	31C	no no not in right away remove that one do you have just
3	87450.0	31C	middle no not that way not that way
3	182650.0	31C	do we have a hook

Averaging TSAI Inferences with XGBoost Word Inferences

$t_w(ms)$	t_s	L_{acc}	H_{acc}	F_{acc}	Precision	Recall	F1
4500	5	56.8	72.0	74.2 \pm 2.8	100.0	50.7	67.1
5000	15	56.8	66.7	68.9 \pm 3.9	87.1	47.8	61.1
4500	15	40.9	72.0	66.7 \pm 4.3	80.6	47.8	59.8
4500	10	59.1	72.0	65.2 \pm 1.1	71.3	56.5	62.7
5000	5	61.4	66.7	65.2 \pm 5.7	75.2	49.3	59.0
4000	5	59.1	62.9	65.2 \pm 3.9	80.7	44.9	57.4
2500	15	65.9	58.3	64.4 \pm 2.1	74.0	50.7	59.8
3000	20	65.9	55.3	64.4 \pm 3.9	69.6	56.5	61.6

Research Question Answers

❑ Low Level Average

❑ TSAI Accuracy = 0.545

❑ High Level Average

❑ XGBoost Accuracy = **0.609**

❑ Fusion Average

❑ TSAI + XGBoost Accuracy = 0.595

Future Work

- ❑ **Perform ablation study on audio features**

- ❑ Consider into including jitter and shimmer features

- ❑ **Explore other word embeddings**

- ❑ FastText is not optimized for *spoken dialogue*

- ❑ **Look into clustering data for generating labels**

- ❑ Rather than relying on human annotation

Questions?