

ALFORM: Articulatory Labeling For Optimal Real-time MRI Experiments

18-844 Final Slides

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Investment Thesis

(The Elevator Pitch)

- *FOR* MRI technologists and researchers
- *WHO* need reliable articulatory data without wasting scanner time
- *THE* ALFORM system *IS A* real-time articulatory tracking tool
- *THAT* shows which articulators are being used and how often, while you scan
- *UNLIKE* expensive trial-and-error experiments that only discover missing articulators after the fact
- *OUR PRODUCT* gives instant feedback so you can focus on the right movements and collect better data, faster

PROBLEM

(Before ALFORM)

During the scan:

We guess which tasks will engage target articulators.

We collect RT-MRI data with no real-time feedback on articulator usage.



After the scan:

We manually label articulators offline.

We discover that key articulators were under-sampled or missed.

Wasted scanner time

Expensive RT-MRI sessions that don't capture what we need

Incomplete biased datasets

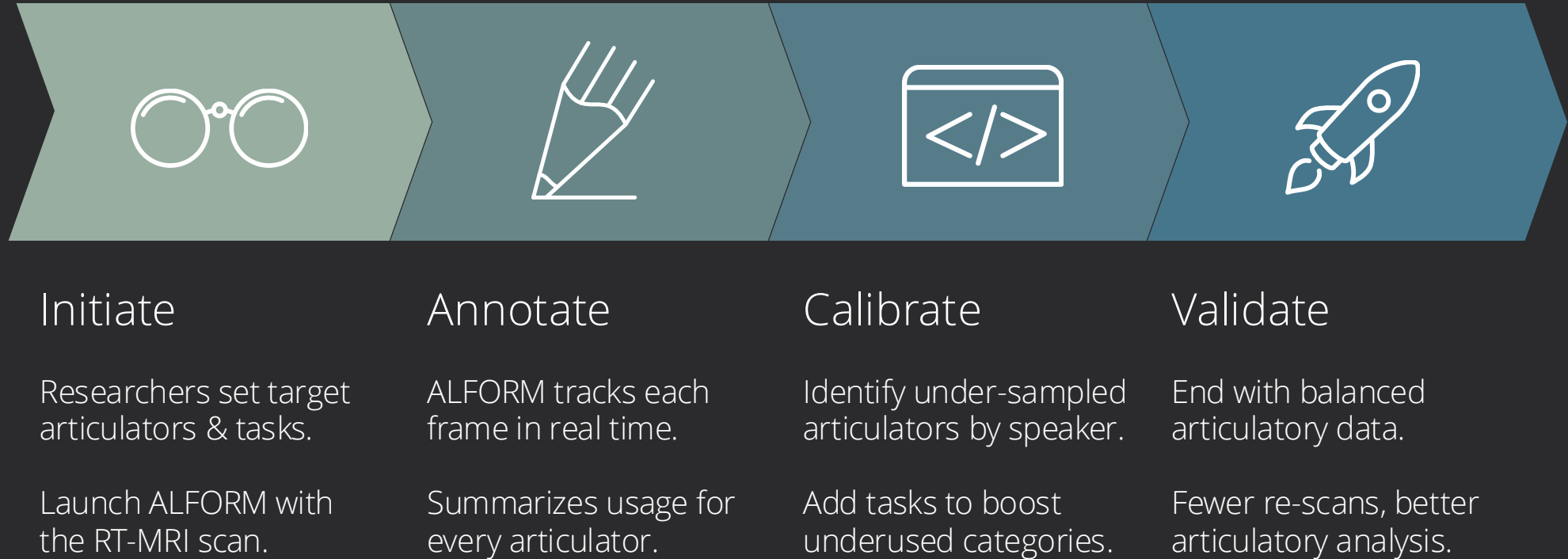
Important articulators (e.g., velum, tongue root) under-represented.

Slow, non-scalable science

Manual labeling + re-scans limit sample sizes & slow down progress

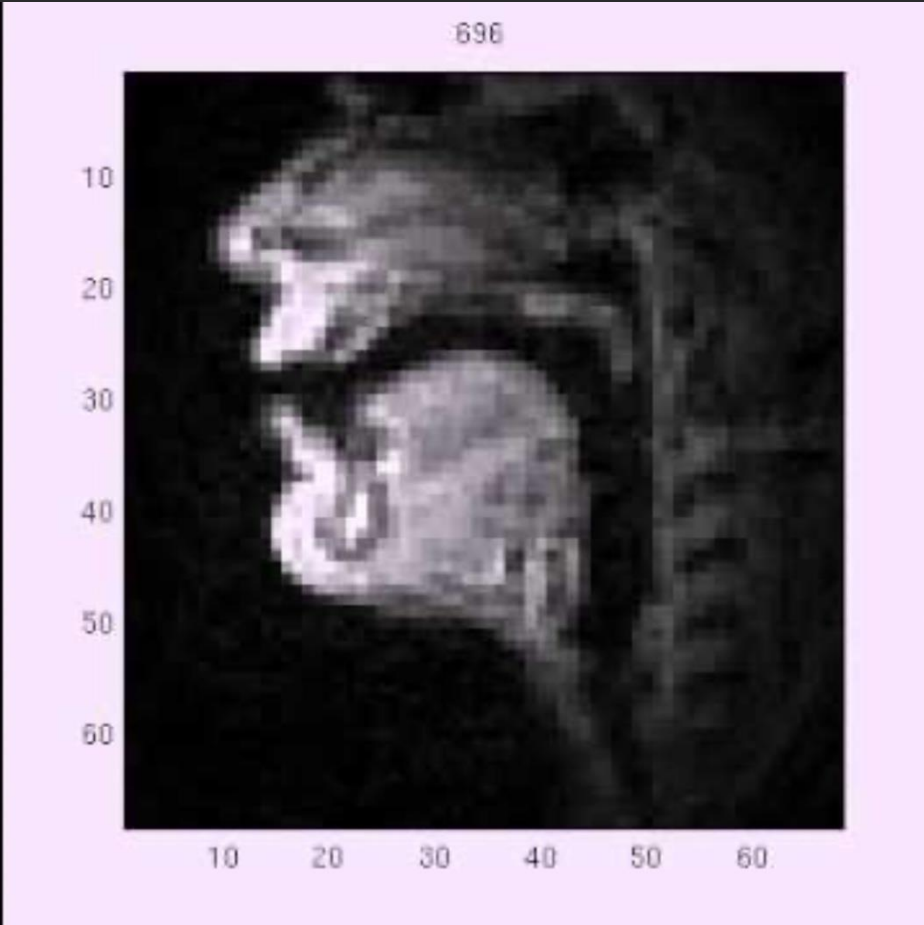
Here's How We Enhance Productivity

By Improving Data Collection Quality



Unique Real-Time Challenge

Very Challenging For Human Annotators



Articulatory Labels
About 70 to 140 per second



Articulators
About 7 per speech sound

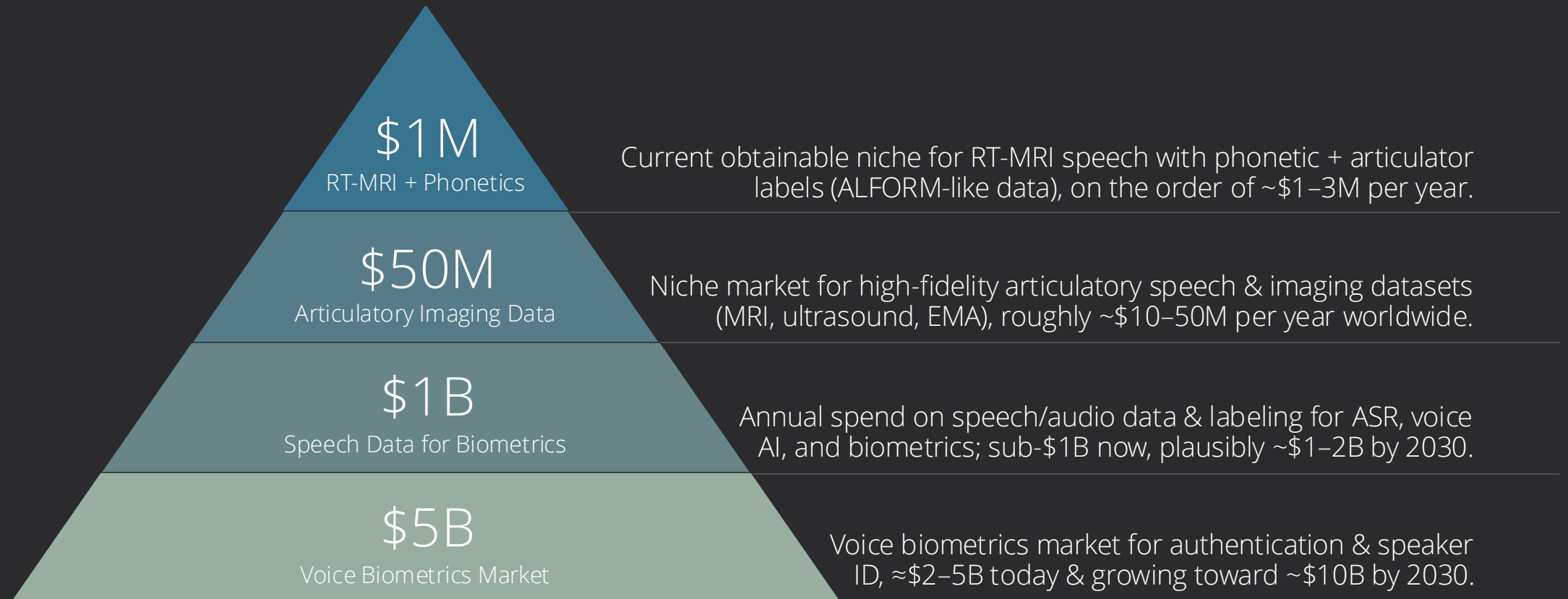


Speech Sounds
About 10-20 per second



A Blue Ocean Market

With Critical Need For Real-Time MRI Speech Data



How Customers Use Our Product

In Medical Applications



A Jetson AGX Orin is used for development, with the compute + monitor + interface under 100W.



Stricter medical certification is required later on. A medical cart like this would be the next phase.

How Our Product Works

With Deep Learning Technology



Enhanced Phonetic
Annotated Datasets
With Articulations

- Have phoneticians expertly annotate speech recordings
- Map phonetic labels to physiological articulations
- Final annotations are speech data & articulatory label pairs



Optimized Recurrent
And Convolutional
Neural Architectures

- Recurrence enables efficient temporal processing
- Convolutions enable hierarchical representation
- Minimize cross entropy loss of predicted alignments



Inference Hardware
Optimization Of
Performance-Per-Watt

- Models are quantized from 32bit to 8bit
- Different inference hardware are experimented with
- Different wattages are experimented with

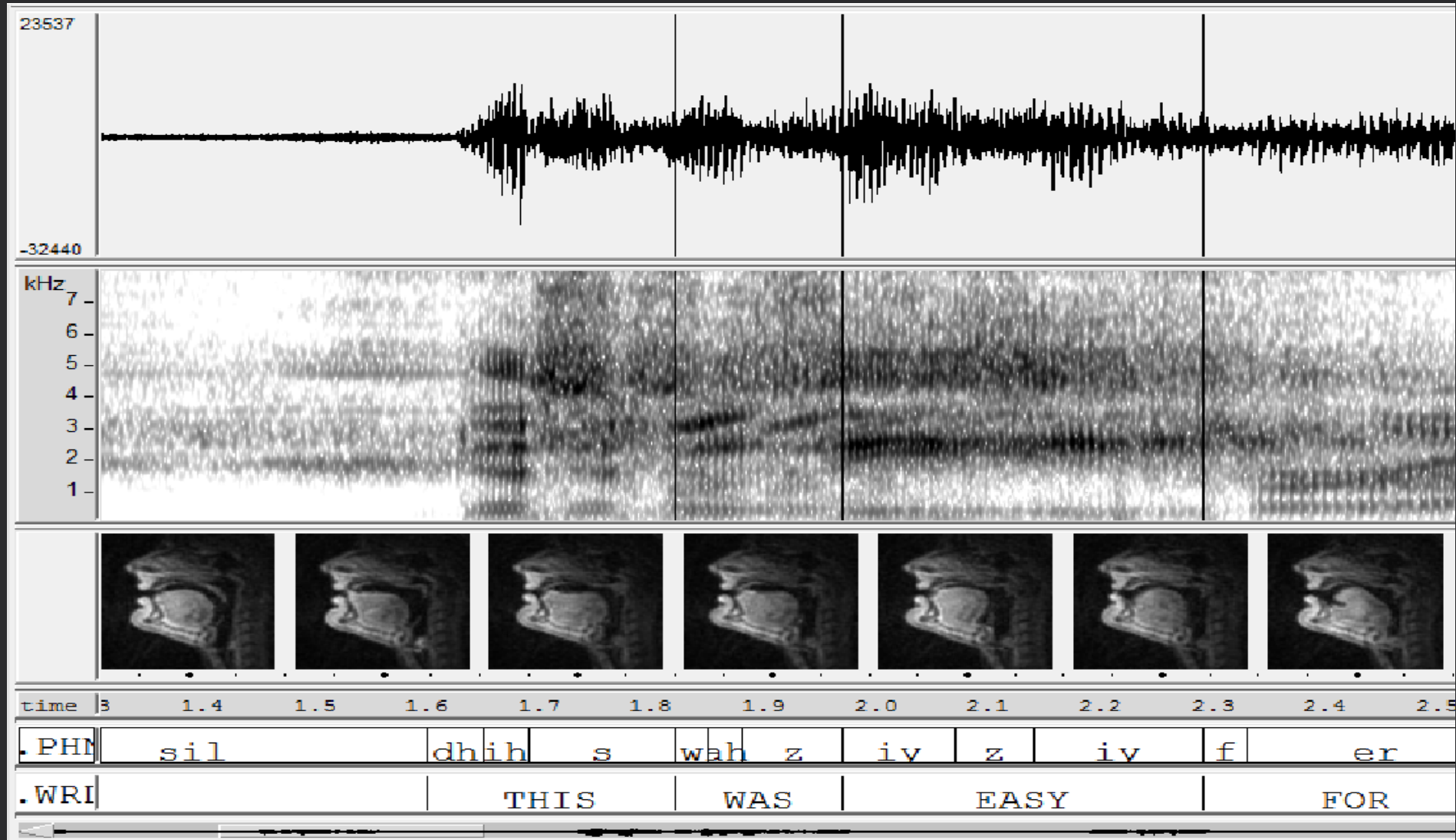
What Makes Our Product Different

Efficiency, Efficacy, and Economics



Our Primary Modality Is Speech

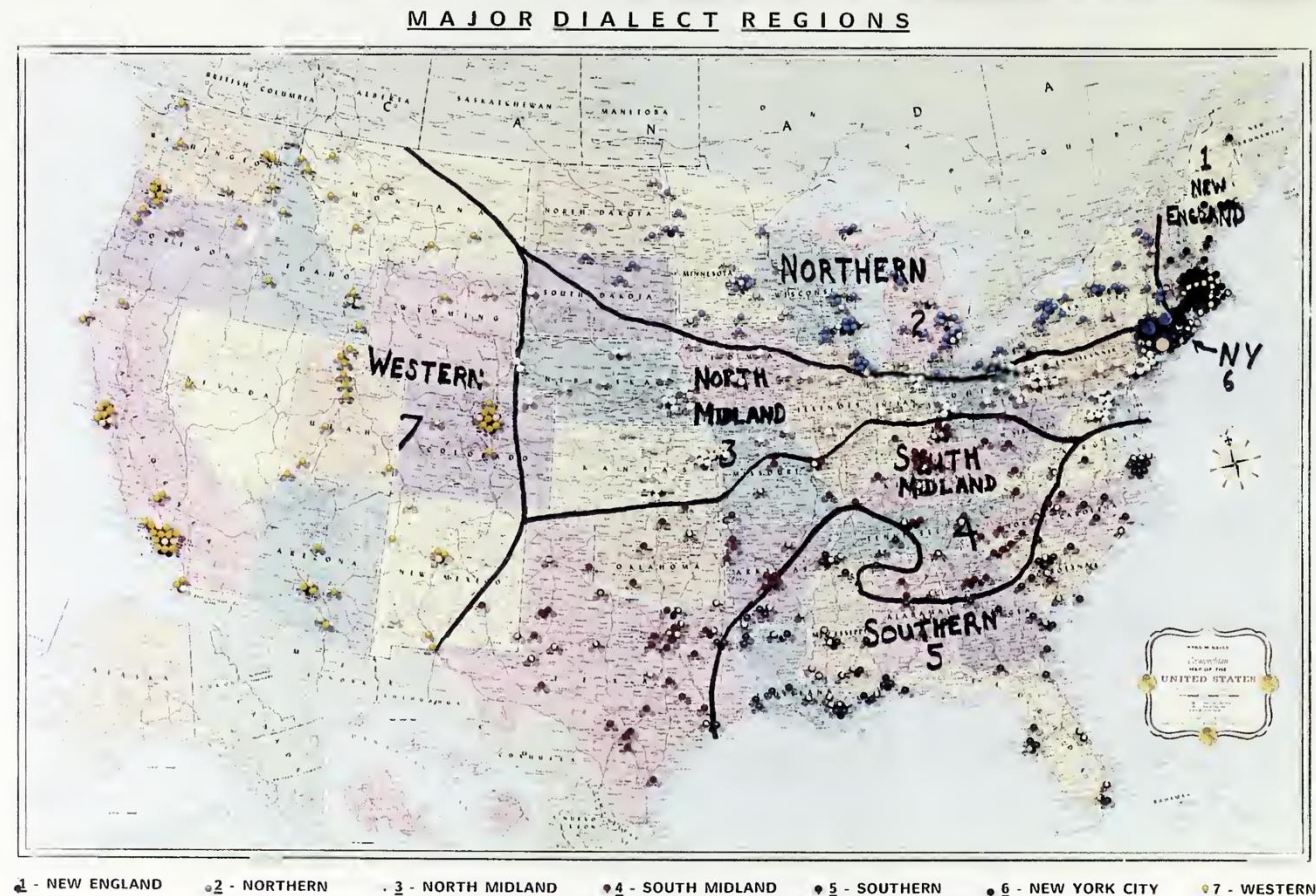
Each frame has 1 phoneme label and 7 articulatory labels



Here Are Our Demographics

Focused On 8 Dialect Regions For American English

Figure 3.1: Map of TIMIT Dialect Regions



Courtesy of Texas Instruments, Inc.

Table 3.1: Dialect distribution of speakers

Dialect Region		# Male Speakers	# Female Speakers	Total # Speakers
Name	Code (dr)			
New England	1	31 (63%)	18 (27%)	49 (8%)
Northern	2	71 (70%)	31 (30%)	102 (16%)
North Midland	3	79 (67%)	23 (23%)	102 (16%)
South Midland	4	69 (69%)	31 (31%)	100 (16%)
Southern	5	62 (63%)	36 (37%)	98 (16%)
New York City	6	30 (65%)	16 (35%)	46 (7%)
Western	7	74 (74%)	26 (26%)	100 (16%)
Army Brat (moved around)	8	22 (67%)	11 (33%)	33 (5%)
Total # Speakers:		438 (70%)	192 (30%)	630 (100%)

Tradition Phoneme Recognition Has One Classification Task

Phoneme Symbol Logits

AA	AE	AH	AO	AW	AY	B	CH	D	DH	EH	ER	EY	F	G	HH	IH	IY	JH	K	L	M	N	NG	OW	OY	P	R	S	SH	T	TH	UH	UW	V	W	Y	Z	ZH	SIL
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Articulation Involves Seven Separate Classifications

Vowel Logits

True

False

Voiced Logits

True

False

Manner Logits

Low

Mid

High

Nasal

Shut

Primary

Divided

Glide

Place Logits

Front

Mixed

Lip

Point

Top

Back

Throat

Round Logits

True

False

Wide Logits

True

False

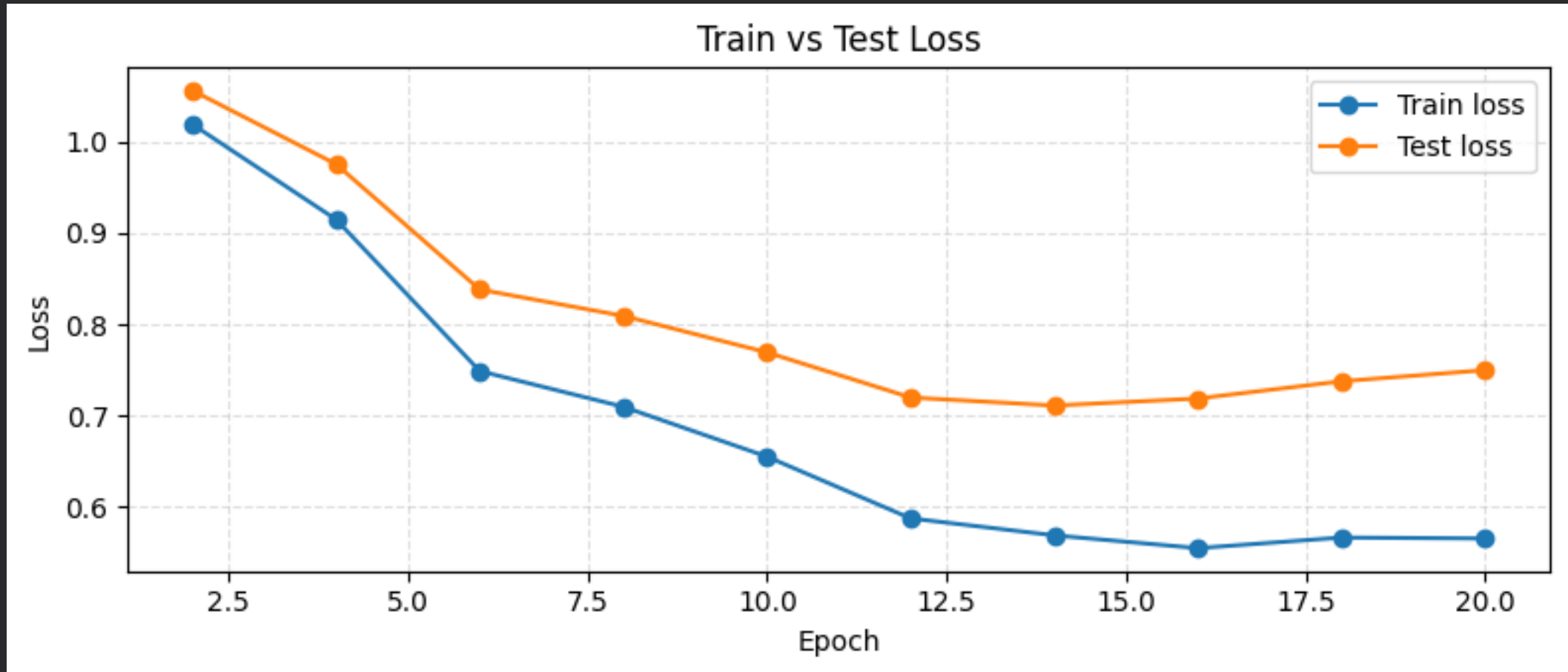
Here Are Our Metrics

On Articulator Recognition By Gender & Dialect

	Vowel	Voiced	Manner	Place	Modified	Round	Wide
FEMALE	93%	97%	87%	85%	96%	93%	87%
MALE	92%	97%	86%	85%	96%	94%	89%
DR1	92%	97%	85%	85%	95%	92%	87%
DR2	93%	97%	87%	86%	97%	95%	89%
DR3	92%	97%	87%	85%	97%	95%	89%
DR4	92%	96%	86%	84%	96%	94%	88%
DR5	92%	97%	85%	84%	96%	93%	87%
DR6	92%	97%	87%	85%	96%	94%	87%
DR7	92%	97%	86%	85%	97%	93%	89%
DR8	92%	96%	86%	85%	97%	94%	88%
ALL	92%	97%	86%	85%	96%	94%	88%

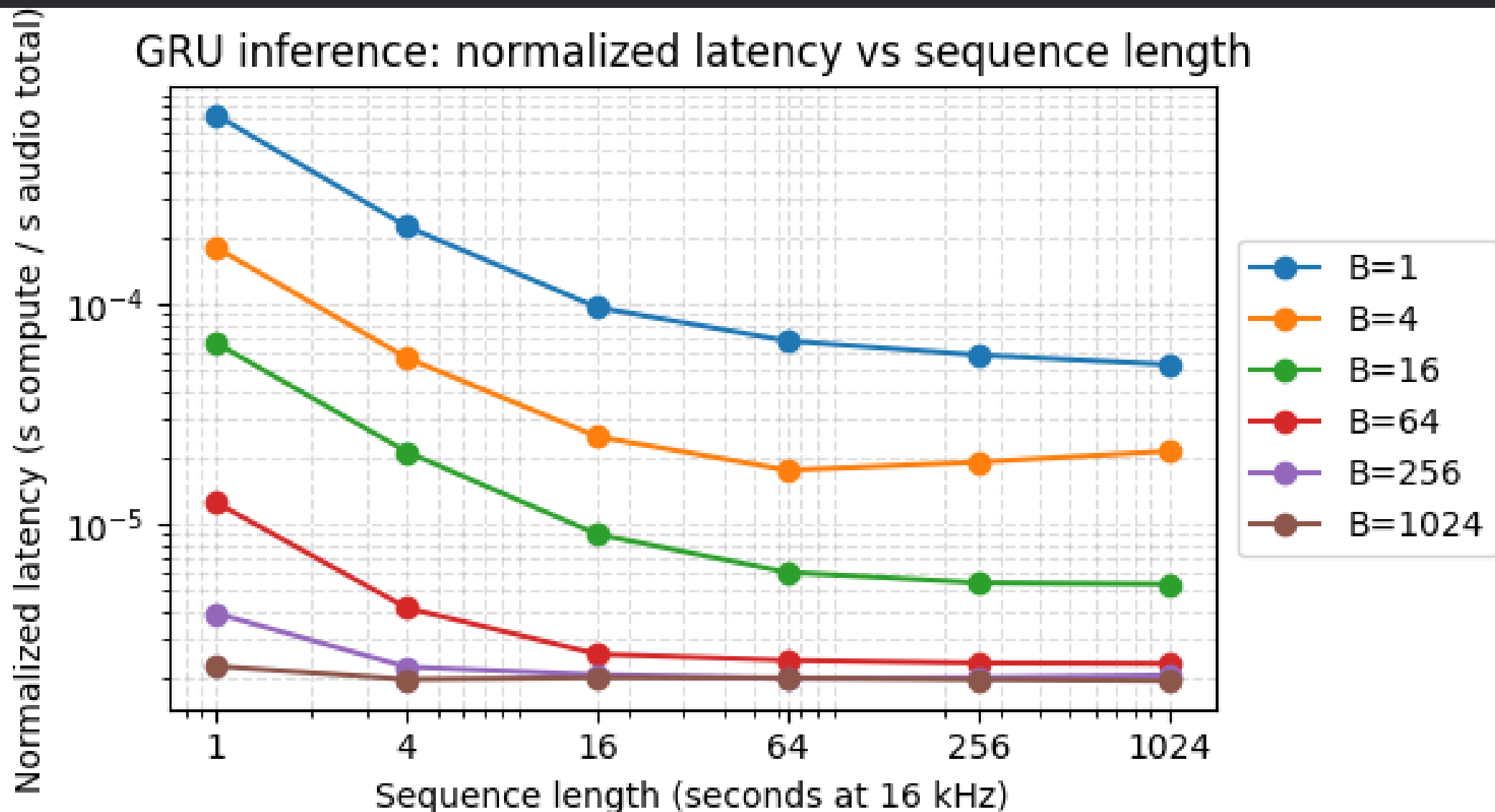
Here Are Our Metrics

On Model Training Optimization



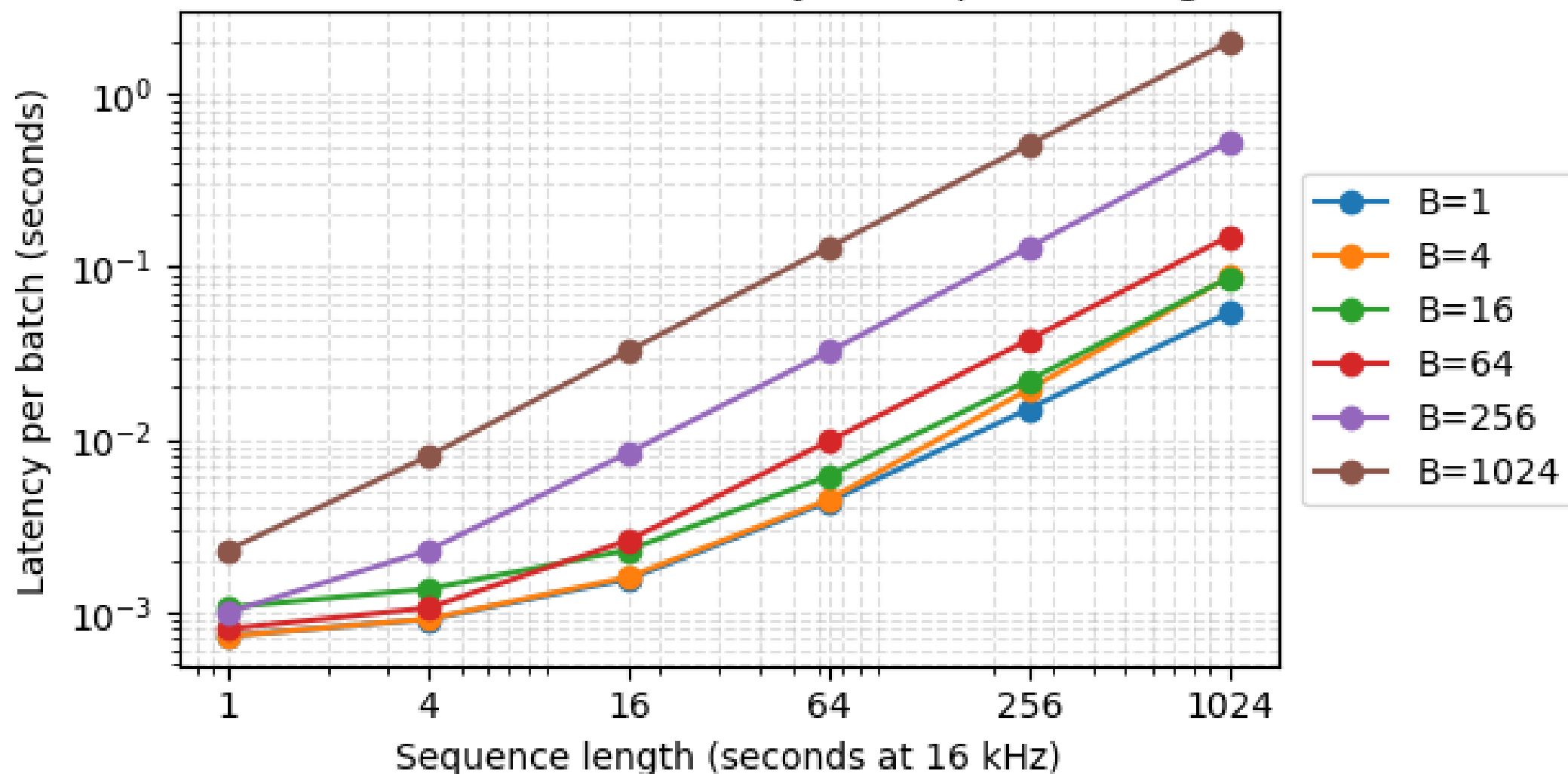
Model overfits, but shows capability of learning more-complex representations. Future work will look into closing the gap using different loss functions as regularizers.

Here Are Our Metrics



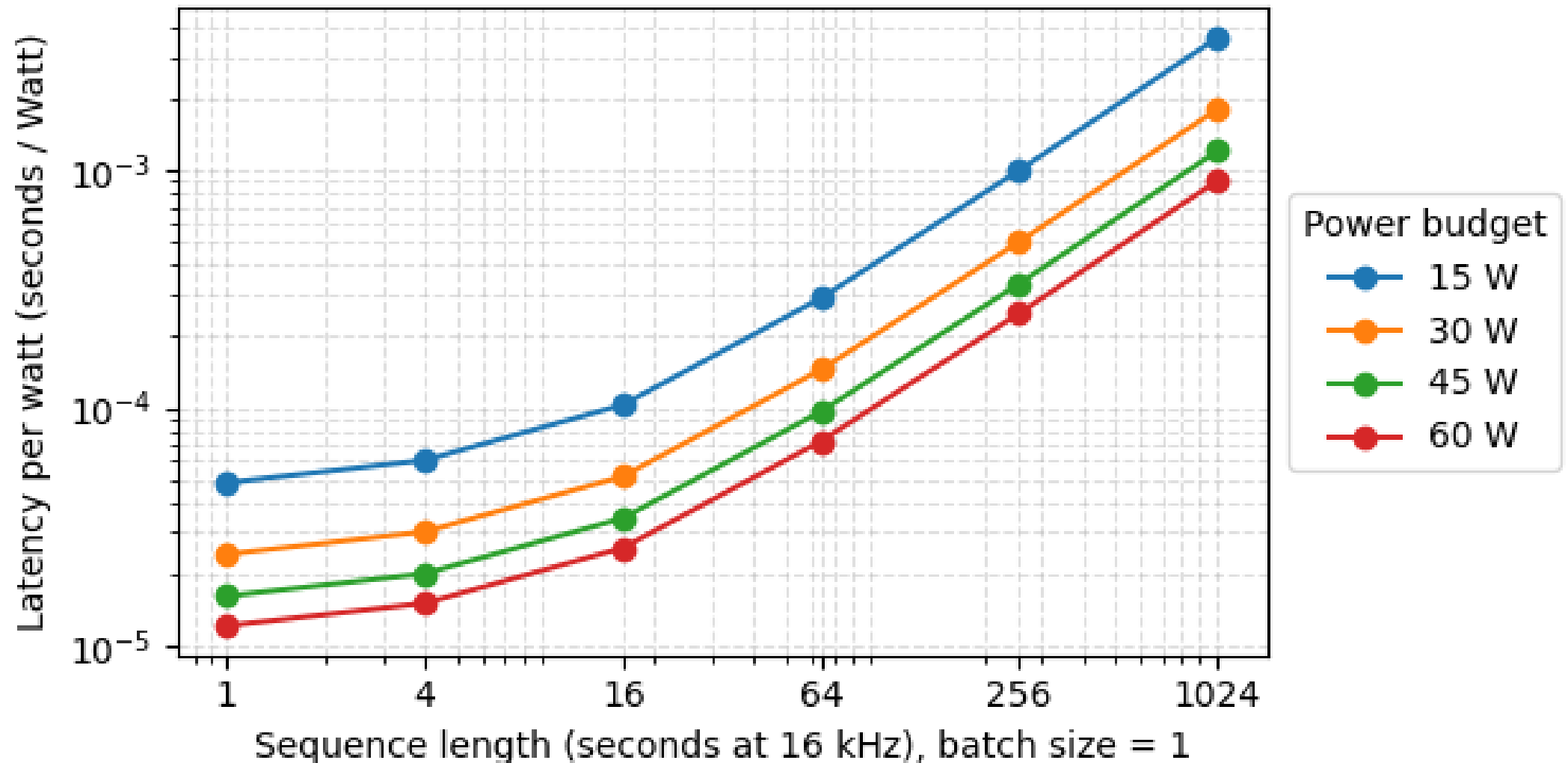
Here Are Our Metrics

GRU inference: total latency vs sequence length



Here Are Our Metrics

Latency per Watt vs sequence length (B=1)



The Next Phase Needs \$10K

To Prototype On Medical-Grade Hardware (IEC 60601-certified)

01

Computer

Rugged AI computer with shielded medical I/O and ESD-safe chassis so plugging in cameras and sensors is safe for staff and patients.

\$5,800

02

Monitor

High-brightness, disinfectant-safe screen that stays readable under hospital lighting and survives constant cleaning without damage.

\$1,200

03

Chassis

Stable, low-tip medical cart with smooth hospital grade casters and cable routing so the whole system moves safely between patients.

\$1,500

04

Battery

Fault-tolerant medical battery pack with thermal and overload protection, built to power the cart for hours without risky DIY wiring.

\$1,500

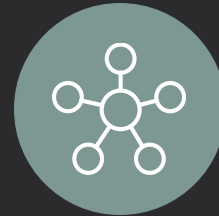
Considerations TODO

Many practical challenges need to be considered.



RNNs on Edge Impulse?

RNNs aren't supported, so it needs to be hacked or CNN is needed instead.



Propose HW Configs

Give some options for low, medium, and high end recommended configs.



Testing On Jetson AGX Orin

Hardware optimization needs to be re-run on the different AGX constraints.



Real-Time Demonstration

Optimal hardware usage is likely not real-time, but RT would improve demo.



Applied Example

An example articulatory report needs to be made for the medical technician.



Training Improvement

Close the overfitting gap and consider a contrastive loss for robustness.

Investment Thesis Recap

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Thank You!



Acknowledgements

- MRI images are from USC 75-Speaker Speech MRI Database
- Gender and Dialect region data are from TIMIT
- Source link to the articulation video is embedded
- Financial estimates are approximate
- A language model was used to facilitate drafting