

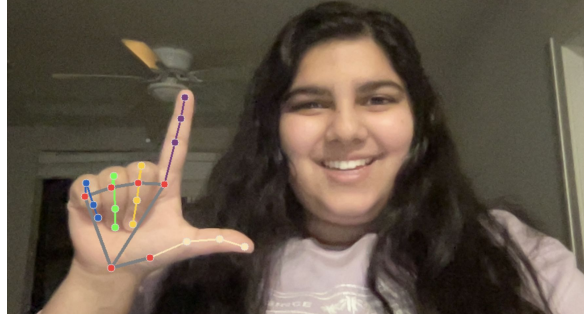
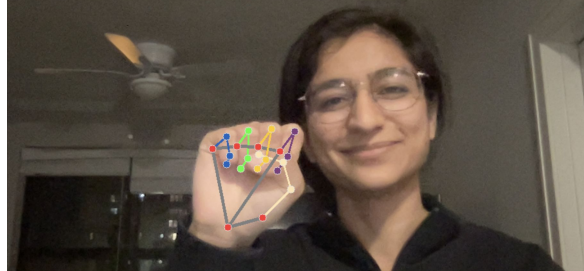
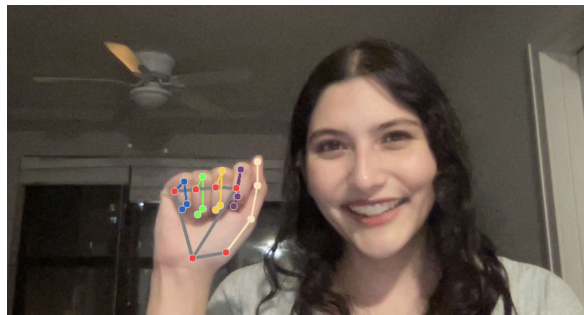


# Team E5: ASLearn

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# Creating ASLearn: an ASL Learning Platform

- Facilitate ASL remote learning
- Users able to learn signs and test their knowledge
  - Signing into a camera
  - Getting immediate feedback on sign correctness
  - Flexibility in learning pace and schedule

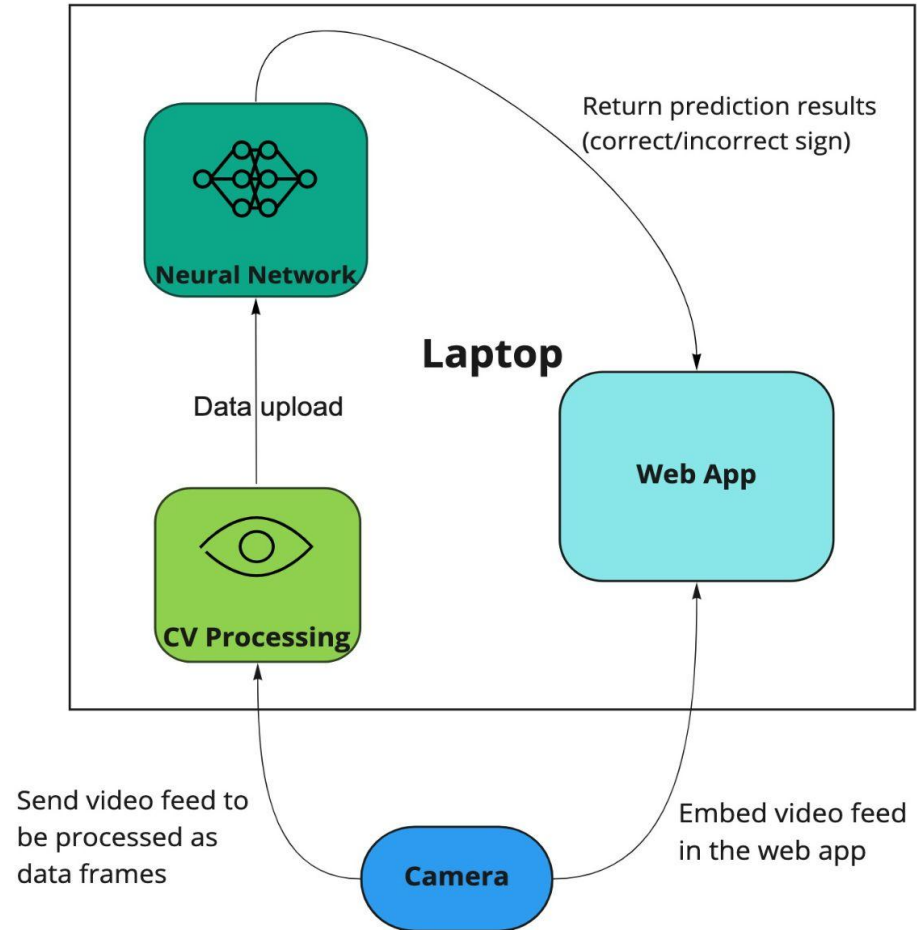


# Use-Case Requirements

<u>Requirement</u>	<u>Metric</u>
Use computer vision to recognize when user is making ASL signs	<b>15</b> communicative signs + <b>26</b> alphabet letters + <b>10</b> numbers (0-9)
Model continues detecting user signs at a reasonable distance	Within <b>3 feet or less</b> facing camera head on
Correct detections occur in a timely manner	Within <b>2 seconds</b> of visual input being received
Accurate feedback on sign correctness	<b>90%</b> accuracy based on model evaluation with test data
Easy to navigate course page	<b>90%</b> user satisfaction based on user survey
Easy to understand feedback of signage	<b>90%</b> user satisfaction based on user survey

# Solution Approach

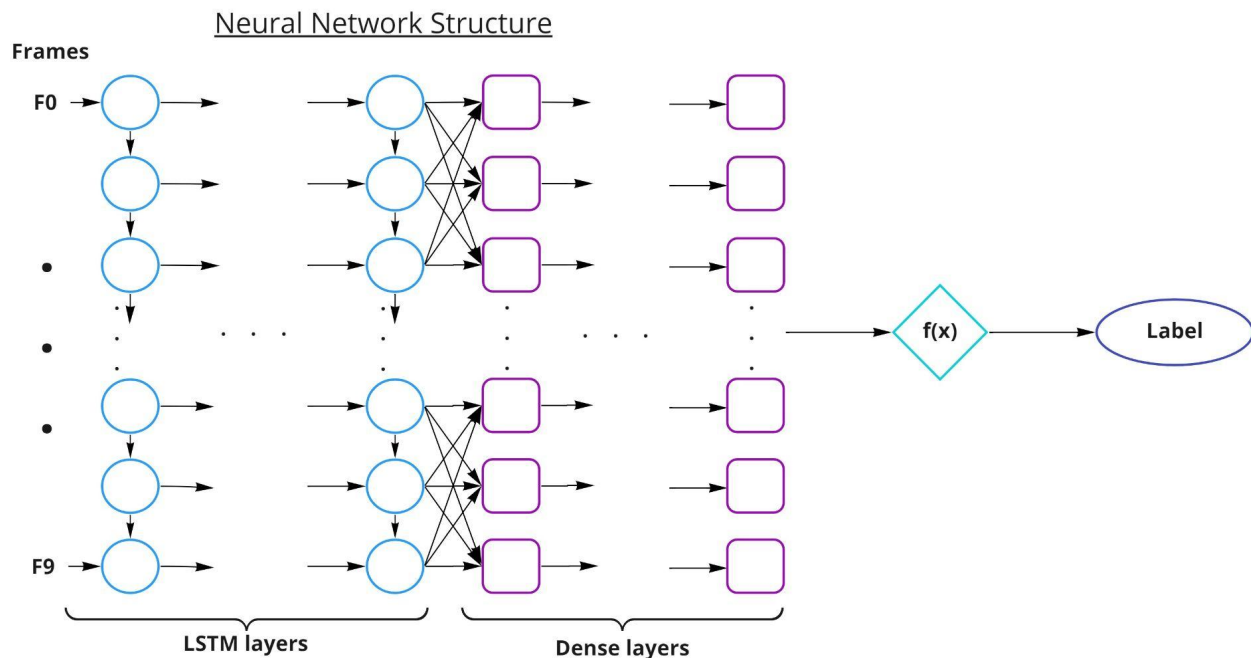
- Video Input
- Computer Vision
- Machine Learning/Neural Network
- Web Application



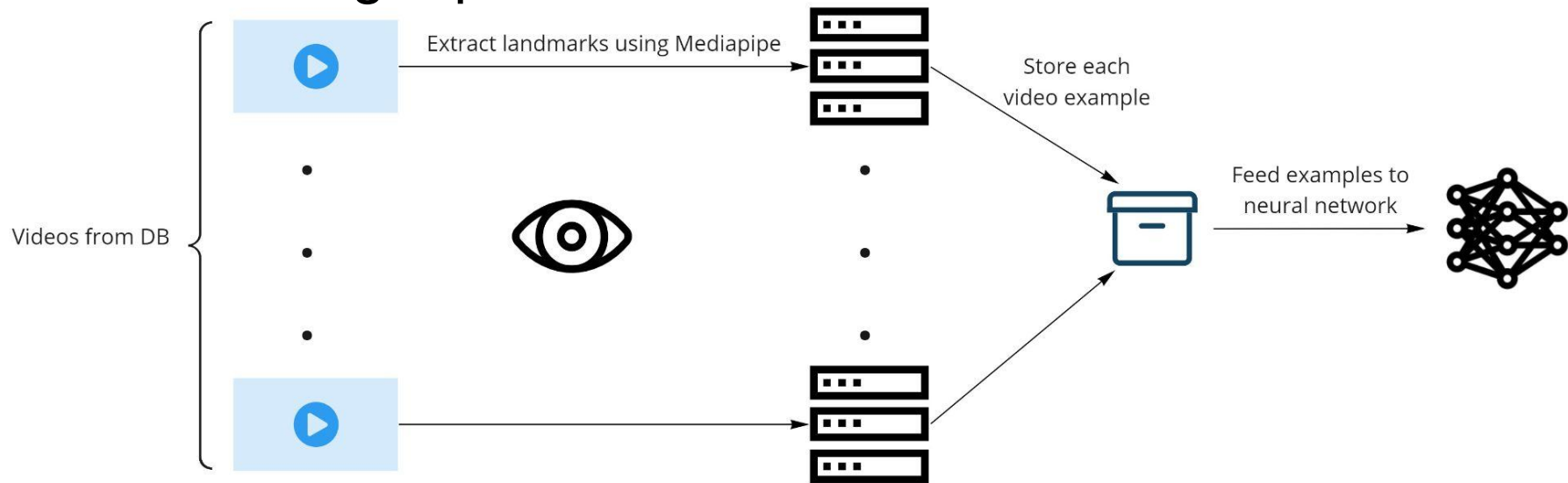
# Neural Network

Final layer will output probabilities of each label being the detected gesture.

$f(x)$  will apply  $\text{argmax}$  to compute the label with the highest probability



# Model Training Pipeline



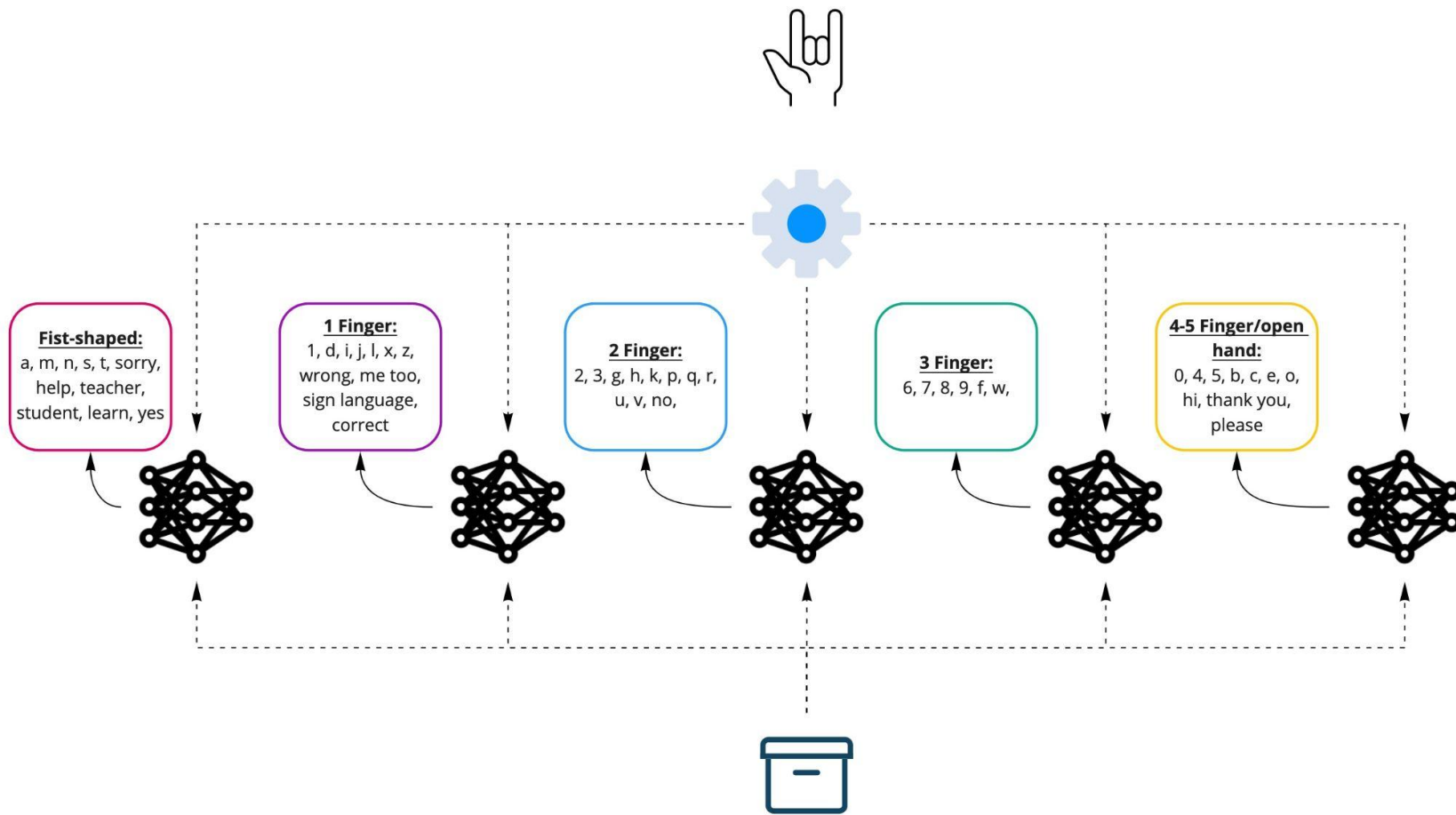
Training data sources:

[Mendeley](#)  
[Alphabet](#)

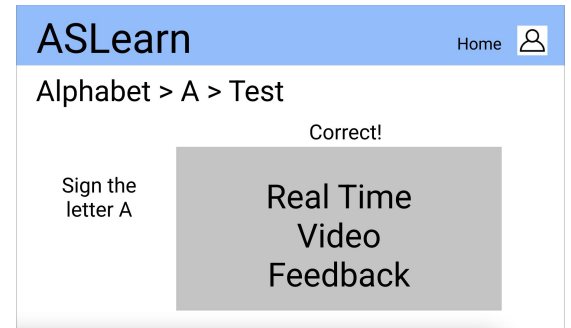
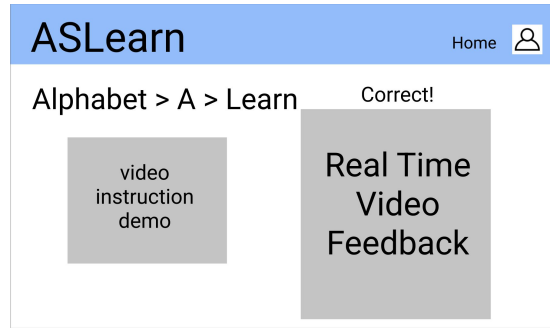
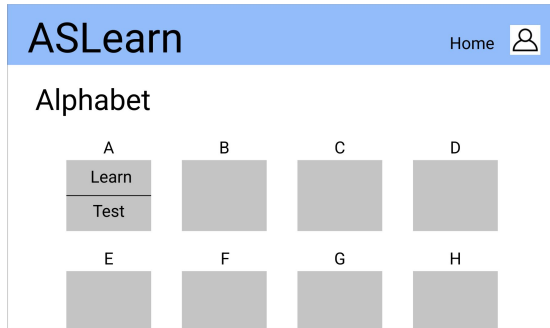
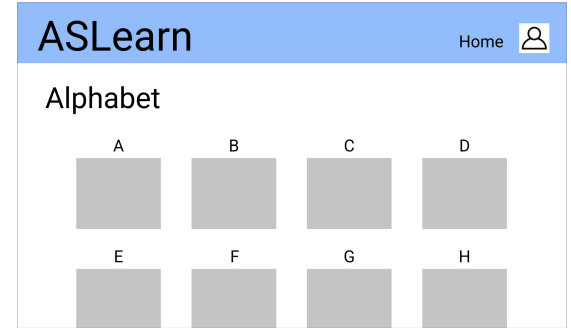
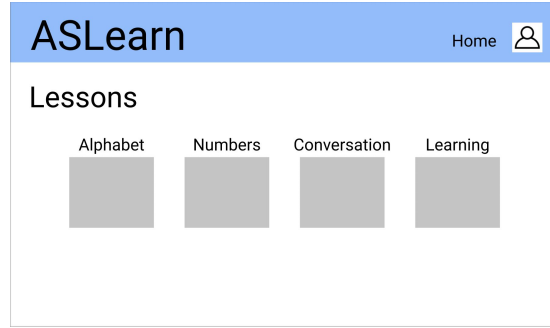
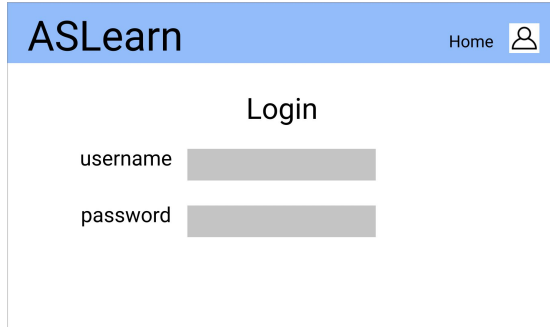
[WASL](#)  
[Numbers](#)

[Letters j.z](#)  
[ASLLVD](#)

# Neural Network as Microservice Architecture



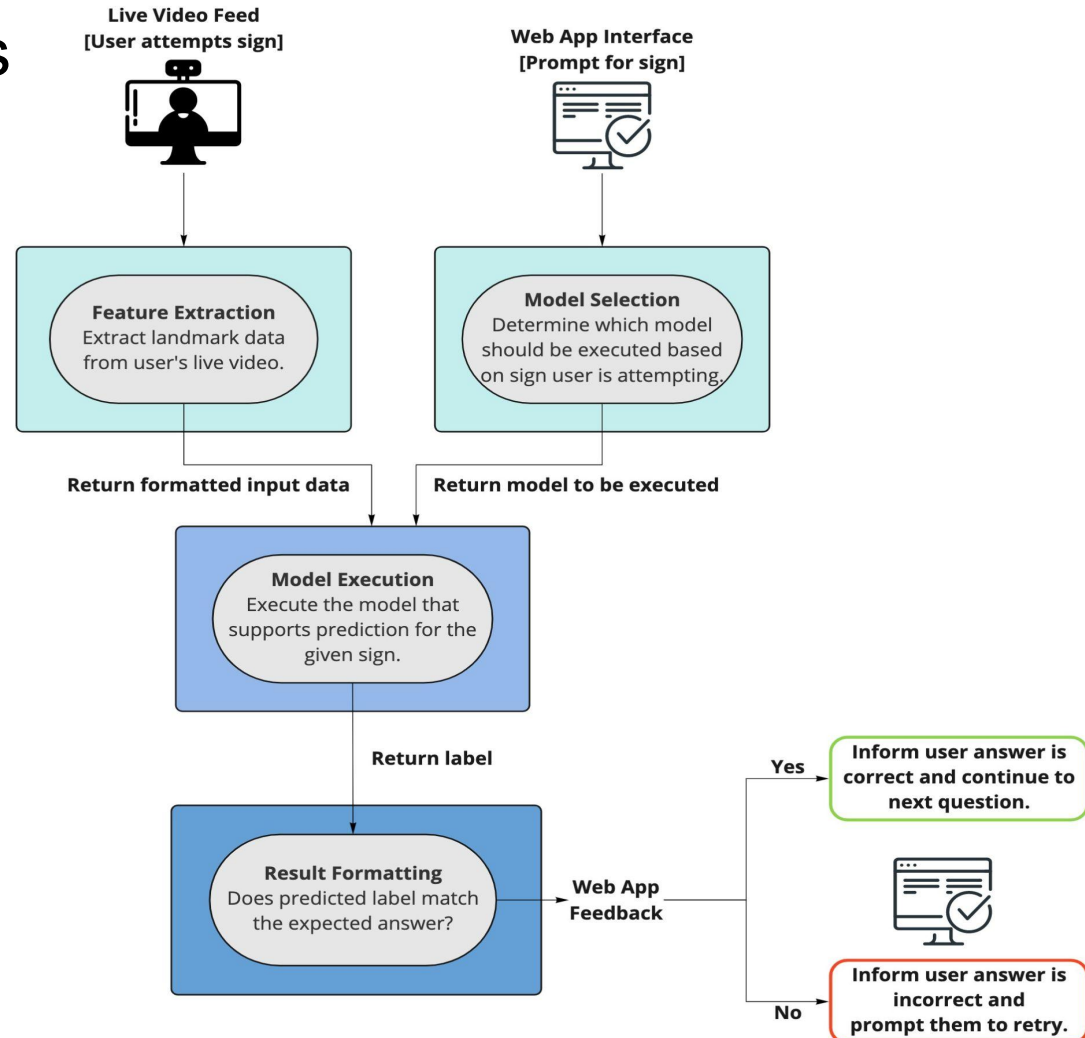
# Webapp UI Design





# Responding to User Signs

This sequence of steps will determine correctness of user's answer and what feedback to display in the web app.



# Testing, Verification, Metrics

<u>Requirement</u>	<u>How We Will test</u>	<u>Passing Metrics</u>	<u>Risk Mitigation</u>
Distance from camera	Do signs at varying distances from camera (up to 3ft)	Sign recognition should be 90% accurate at <b>&lt;3 feet</b>	Reduce distance requirement, use bounding box.
Platform latency	Have users sign in platform and record feedback time	Feedback will be given to user within <b>2 seconds</b>	Use alternative prediction generation method
Webapp User Interface	Conduct user testing and surveys (with at least 5 users)	<b>90%</b> overall user satisfaction based on survey	Modify UI according to user feedback

# Testing, Verification, Metrics (cont)

<u>Requirement</u>	<u>How We Will test</u>	<u>Passing Metrics</u>	<u>Risk Mitigation</u>
Accuracy (i.e. jewelry)	Test with various user impediments (i.e users wearing bracelets, rings, etc)	Sign recognition should be <b>90% accurate</b> regardless of these impediments	Tell user not to wear jewelry, and remove other potential impediments
Left/Right hand dominance	Test platform with right and left handed users	Sign recognition should be <b>90% accurate</b> in both left / right handed cases	Tell user to make signs using only right hand (as that is our testing dominance currently)

# Specifications

## Initial Specifications (experimenting with toy data)

Raw input video specs	5 seconds, 30 FPS
Sampling specs	10 frames, 2 FPS
Model execution time (1 LSTM Layer, 1 Dense Layer)	0.07-0.12 seconds
<b><u>Tools</u></b>	
MediaPipe (Hand, MediaSequence)	Extracting landmark data from video data
Tensorflow	Creating, training, and saving neural network
NumPy	Formatting data to feed into neural network
AWS	Training models and deploying web app within EC2 instances

