

Exploring Sign Language Recognition Through LSTM and Cosine Similarity Modeling

Github Link



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O1 Problem Statement



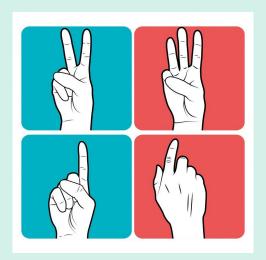


Human Translator Expensive











Communication Barrier



02 Solution



Word Level Translation

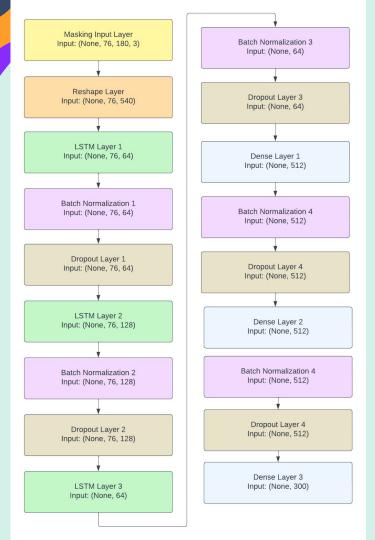
Deep Learning, MediaPipe, Cosine Similarity

Focus on Interpretation of Fine-Grained Gestures

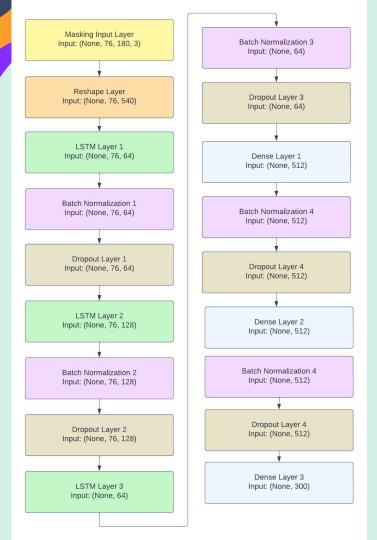




O3 Model Architecture



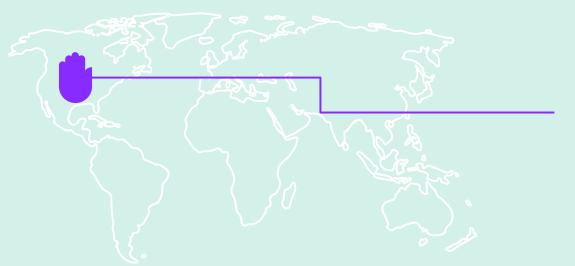
- Input Layer:
 - Masking Layer = ignores padded values
- Reshape Layer:
 - Convert to 3D format
- LSTM Layers:
 - 3 Layers
 - ReLu Activation
 - First 2 Layers have return sequences



- Dense Layers:
 - 2 Layers
 - ReLu Activation
- Output Layer:
 - Dense Layer
 - Linear activation
- Techniques:
 - Batch Normalization
 - Dropout (rate=0.5) = Regularization
 - Applied after each hidden layer



Dataset



WLASL

- 21,083 videos corresponding to 2,000 common ASL words
- Metadata provided



Feature 05 Engineering





Landmark Extraction

- Hands:
 - 42 (21 on each hand)
- Pose:
 - 6 landmarks
 - For the upper body
 - Excluding the face
- Face:
 - 132 out of the 478
 - Focusing on the lips, eyes, eyebrows, and the outline of the face

Total Landmarks: 180 Landmarks Centered





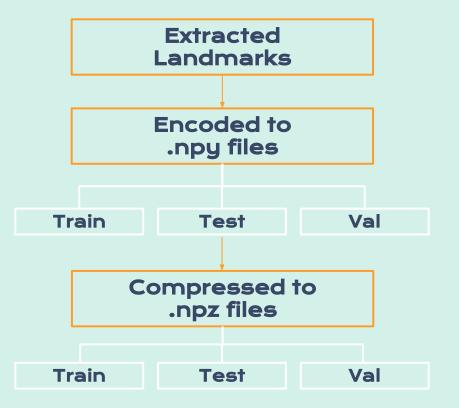
Landmark Extraction

| Body | Hands (Left and Right) | Face |
|---|---|--|
| Left Shoulder Right Shoulder Left Elbow Right Elbow Left Wrist Right Wrist | Wrist Thumb CMC Thumb MCP Thumb IP Thumb Tip Index Finger MCP Index Finger DIP Index Finger Tip Middle Finger MCP Middle Finger DIP Middle Finger DIP Middle Finger DIP Middle Finger DIP Middle Finger Tip Ring Finger MCP Ring Finger DIP Ring Finger Tip Pinky MCP Pinky DIP Pinky DIP Pinky Tip | Face Outline Lips Eyes Eyebrows |





Preprocessing





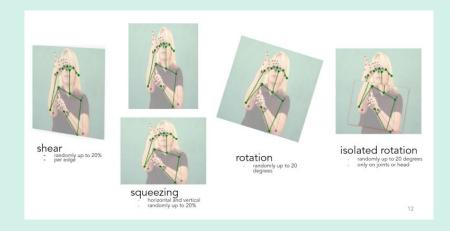


Augmentation

Rotation

Zoom

Shift



Mask

HFlip

Speedup





Padding

X=74 (example)



X=76 (expected)

Padding
value = -80
for (76-74=)
2 frames
each with
180
landmarks





Label Encoding

Labels

FastText

Encoded to word vectors



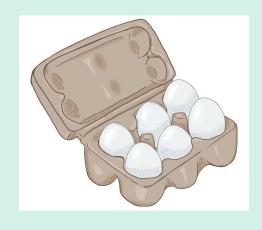
06 Training

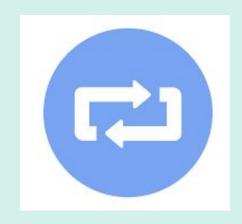




Epochs = 100

Validation





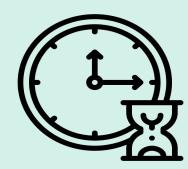




Early Stopping

```
early_stopping = tf.keras.callbacks.EarlyStopping(
    monitor='val_accuracy',
    patience=10,
    restore_best_weights=True
)
```

Patience = 10





Model Checkpoint



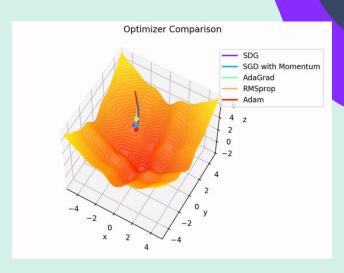
```
model_checkpoint = tf.keras.callbacks.ModelCheckpoint(
    filepath=os.path.join(checkpoint_filepath, 'model_{epoch:02d}.h5'),
    save_weights_only=True,
    save_best_only=True,
    monitor='val_accuracy',
    mode='max',
)
```



Optimizer

Adam Optimizer (∝ = 0.05, 0.9 decay every 1000 steps)

```
lr_scheduler = tf.keras.optimizers.schedules.ExponentialDecay(
    initial_learning_rate=0.05,
    decay_steps=1000, # decrease learning rate at an expotential rate
    decay_rate=0.9
)
```

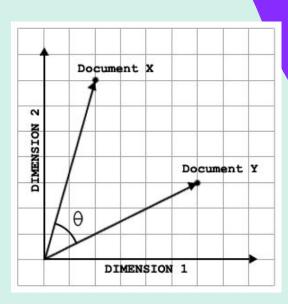




Loss

Cosine Similarity Loss

```
model.compile(
    loss=tf.keras.losses.cosine_similarity,
    optimizer=tf.keras.optimizers.Adam(learning_rate=lr_scheduler),
    metrics=['accuracy']
)
```







16.68%

Test Accuracy on

2000

labels i.e. the entire dataset







Prediction

Confusion Loss ROC Accuracy **Matrix** Total Loss VS Total For each unique label Total Accuracy VS Total For first 50 samples Validation Accuracy Validation Loss with AUC





| Subset Sample Size | Test Accuracy (%) | |
|--------------------|-------------------|--|
| 50 | 22.378 | |
| 200 | 21.154 | |
| 1000 | 17.964 | |
| 2000 | 16.678 | |

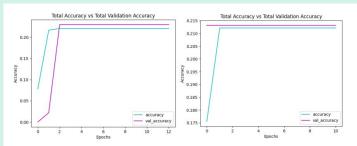


Fig. 5: For 50 samples

Fig. 6: For 200 samples

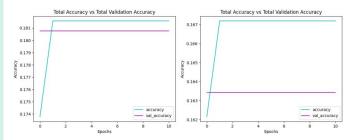


Fig. 7: For 1000 samples

Fig. 8: For 2000 samples

Fig. 9: Total Accuracy vs Total Validation Accuracy

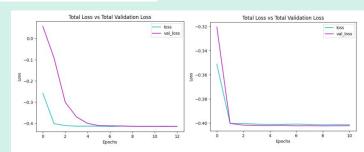


Fig. 10: For 50 samples

Fig. 11: For 200 samples

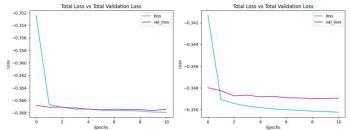


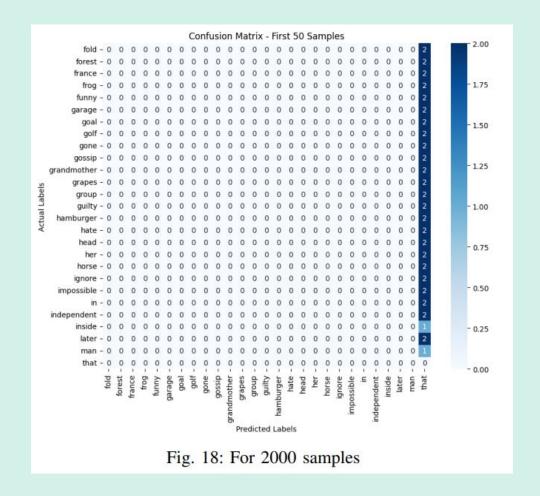
Fig. 12: For 1000 samples

Fig. 13: For 2000 samples

Fig. 14: Total Loss vs Total Validation Loss

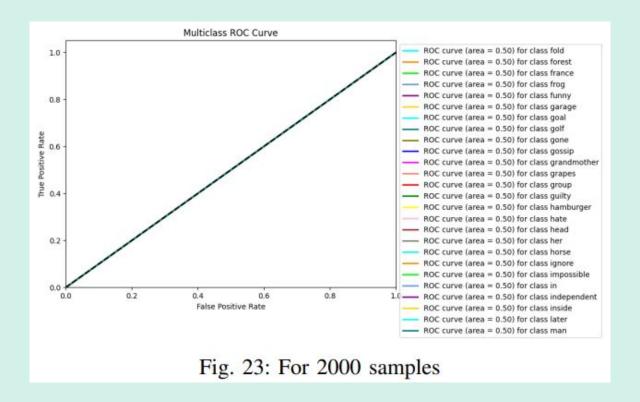
















Challenges 09





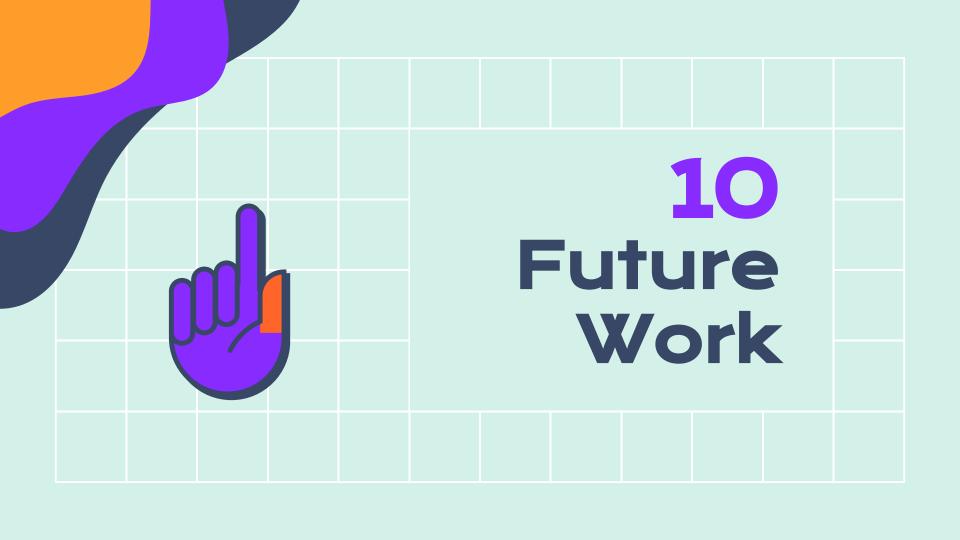
Small epoch

Computational resource constraint

Time constraint

Padding & Augmentation





Larger Epoch

More Augmentation Real-Time Translation

Diverse & Larger Dataset

Sentence Level Translation

Thank You For Your Patience

