

American Sign Language Recognition using CNNs

This presentation outlines our approach to American Sign Language (ASL) recognition. We utilised Convolutional Neural Networks (CNNs). We aimed to build a model capable of translating hand gestures into text. Our project involved data collection, model training, and real-time detection.

Project Methodology

Data Preparation

- Images resized to 64×64 pixels.
- Normalization applied.
- Data augmentation techniques employed.

Model Architecture

- CNN with Conv2D layers.
- Max Pooling layers for feature extraction.
- Dense layers for classification.

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Training and Detection

Training

The model was trained over 10 epochs. We utilized the Adam optimizer.

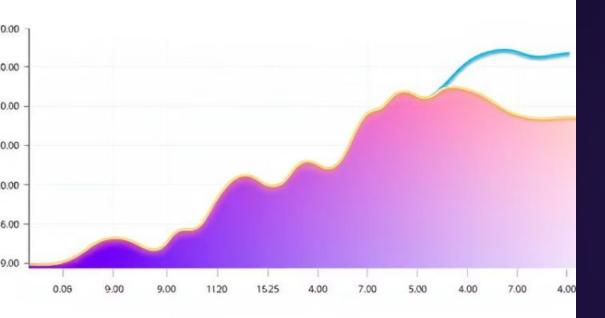
ASL Alphabet dataset
(~3000 images per letter
A-Z) was used.

Dataset

3 Detection

The test dataset was sourced from

/content/asl_dataset/archive/asl_alphabet_test/asl_alphabet_t est.



Key Results



Training
Accuracy
Achieved

approximately

89.97%.



Validation

Accuracy

Reached around

78.73%.



Real-time

Detection

Observed about

38.46% accuracy.(10

out of 26 correct)



Confusion Matrix Analysis

Diagonal Dominance

Most correct predictions reside along the diagonal.

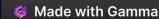
3

Common Confusions

Some letters are frequently confused (e.g., 'I' and 'J').

Insights

Highlights areas needing model improvement.





Real-Time Detection Demonstration

1

Webcam Input

Capturing live hand gestures.

2

Prediction

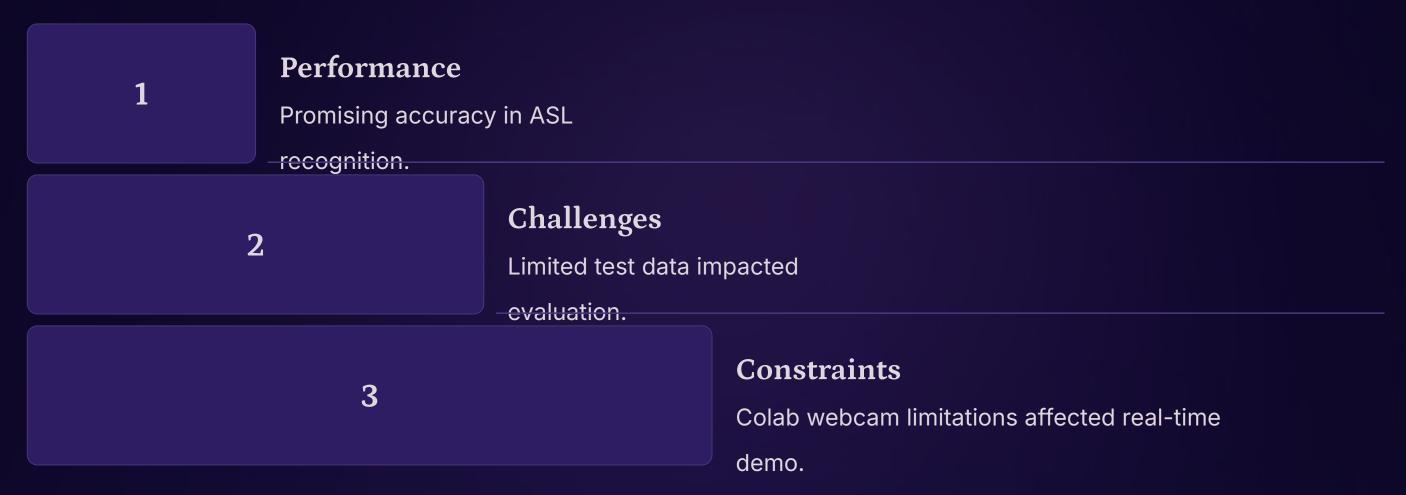
CNN outputs the recognized letter.

3

Display

Detected ASL letter shown on screen.

Project Conclusion



Despite challenges, the CNN model demonstrates potential for ASL translation. Future work should address data limitations.





Future

More Epochs

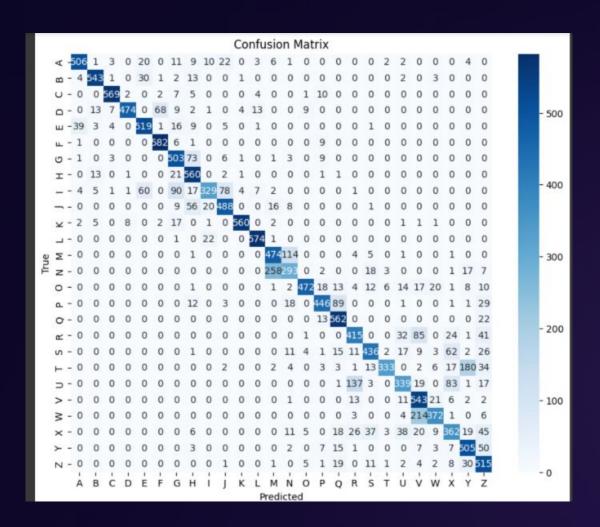
Train the model for a greater number of epochs.

Larger Images

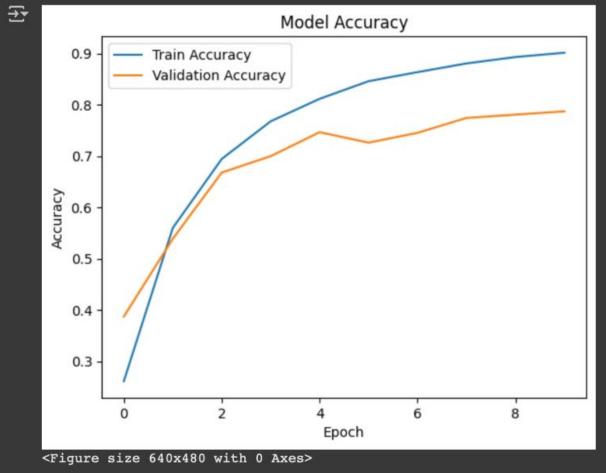
Increase the image resolution for better feature extraction.

Expanded Dataset

Incorporate more diverse and extensive training data.



```
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Model Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
plt.show()
plt.savefig('/content/accuracy_plot.png') # Save for report
```



Thank You

We appreciate your time and interest in our presentation on American Sign Language Recognition using CNNs. We hope you found it insightful We encourage you to explore the potential of this technology and contribute to its development!

