

Configuration Manual

MSc Research Project MSc in Data Analytics

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Project Title: SIGN LANGUAGE RECOGNITION: A COMPARATIVE STUDY

OF DEEP LEARNING MODELS USING YOLOV8, RT-DETR, and Faster

R-CNN

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Date: 29/01/2025

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Configuration Manual

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1 Introduction

1.1 Brief Overview of the Project

This project implements an Indian Sign Language (ISL) Recognition system that compares three deep learning architectures:

- RT-DETR (Real-Time Detection Transformer)
- YOLOv8
- Faster R-CNN

The system processes a dataset of 4,410 ISL images spanning 35 classes (numbers 1-9 and letters A-Z) and evaluates each architecture's performance across different configurations and hyperparameters.

1.2 Purpose of the Configuration Manual

This manual is designed to:

- 1. Guide users through the complete setup and implementation process
- Provide detailed configuration instructions for each architecture
 Help users understand and optimize model performance
- 4. Serve as a reference for troubleshooting common issues

The manual covers:

- Environment setup and requirements
- Dataset preparation and organization
- Model configuration and training
- Hyperparameter tuning procedures
- Evaluation methods
- Inference and testing processes

This documentation ensures reproducible results and consistent implementation across different setups and configurations.

System Requirements

2.1 Hardware Requirements

Minimum Specifications

- GPU: NVIDIA A100 GPU with 40GB VRAM
- o **Storage**: 50GB free space on Google Drive
- **Internet**: Stable broadband connection (10 Mbps minimum)
- RAM: 32GB system memory recommended
- **CPU**: Multi-core processor supporting CUDA operations

Recommended Specifications

- **GPU**: NVIDIA A100 GPU with 80GB VRAM for larger batch sizes
- **Storage**: 100GB+ free space on Google Drive
- **Internet**: High-speed connection (50+ Mbps)

- **RAM**: 64GB system memory for optimal performance
- CPU: Latest generation multi-core processor

2.2 Software Requirements

Operating System

- Any OS compatible with Google Chrome browser
- Google Chrome (Version 90+) recommended

Development Environment

- Google Colab Pro+ subscription
- Google Drive account with sufficient storage
- Python 3.8 or higher

Required Libraries and Versions

- Core Libraries
 - o torch==2.1.0
 - o torchvision==0.16.0
 - o ultralytics==8.3.40
 - o PyYAML==6.0.1
- mage Processing
 - o opency-python==4.8.0
 - o Pillow==9.5.0
- Data Processing
 - \circ pandas==2.0.3
 - o numpy==1.24.3
- Visualization
 - o matplotlib==3.7.1
 - o seaborn==0.12.2
- Additional Tools
 - o tensorboard==2.14.0
 - o GPUtil==1.4.0

Additional Tools

- Google Drive File Stream for local access
- Jupyter Notebook environment (provided by Colab)

2.3 Environment Setup Requirements

- 1. Google Account with:
 - o Access to Google Colab Pro+
 - o Sufficient Google Drive storage

- o Permission to mount Drive in Colab
- 2. Browser Requirements:
 - o JavaScript enabled
 - Cookie permissions for Colab
- 3. Network Requirements:
 - Stable internet connection
 - o Ability to maintain long-running Colab sessions
 - o Access to Google services

2.4 Compatibility Notes

- Models tested on NVIDIA A100 GPU
- Code compatible with PyTorch 2.0+
- CUDA 11.0 or higher required for GPU operations
- System designed for Google Colab environment

3 Connecting to Google Colab and Drive Setup for ISL Recognition System

3.1 Initial Setup

- 3.1.1 Colab Access
 - 1. Visit Google Colab by navigating to colab.research.google.com
 - 2. Sign in with your Google Account (Colab Pro+ subscription required)
 - 3. Upload the notebook file of the
- 3.1.2 GPU Configuration
 - 1. For each notebook, configure GPU settings:
 - o From the top menu, select Runtime, then Change runtime type
 - o In the hardware accelerator dropdown, select GPU
 - o Choose A100 GPU with High-RAM configuration
 - Save your selection

3.2 Project Structure Setup

3.2.1 Google Drive Organization

Create a main project folder called "NCI_Sep_Thesis_2024" in your Google Drive. Inside this, create a subfolder "Devaki_NCI_RIC". Within this folder, create four essential subfolders:

- ISL Devaki Dataset: For storing all dataset files
- training_results: For storing training outputs
- trained_models: For storing trained model weights
- hyperparameter_tuning: For tuning results

3.2.2 Model File Organization

In the training_results folder, organize your model files in onen separate folders:

- RT-DETR folder containing best.pt
- YOLO folder containing best.pt
- Faster R-CNN folder containing best_model_epoch_10.pth

3.3 Dataset Organization

3.3.1 Dataset Structure

Create three main subfolders within ISL_Devaki_Dataset:

- train folder: Contains training images and labels
- valid folder: Contains validation images and labels
- test folder: Contains test images and labels Each subfolder should have an 'images' and 'labels' directory. Place a data.yaml file in the root of ISL_Devaki_Dataset.

3.3.2 System Verification

Essential Checks

- 1. Verify GPU availability:
 - o Confirm A100 GPU access
 - Check 40GB VRAM availability
- 2. Verify storage space:
 - o Ensure minimum 50GB free space
 - o Confirm access to all model weights
 - Verify dataset accessibility
- 3. Verify drive mounting:
 - o Check proper authentication
 - o Confirm read/write permissions
 - Test path accessibility

3.3.3 Troubleshooting Guide

Common Issues and Solutions

- 1. If drive mounting fails:
 - Try re-authenticating
 - Check your internet connection
 - Verify your drive permissions
- 2. If GPU is not available:
 - Verify your Colab Pro+ subscription
 - o Try switching the runtime type
 - o Restart the runtime
- 3. For memory issues:
 - Clear your runtime
 - Try reducing batch sizes
 - Verify file integrity

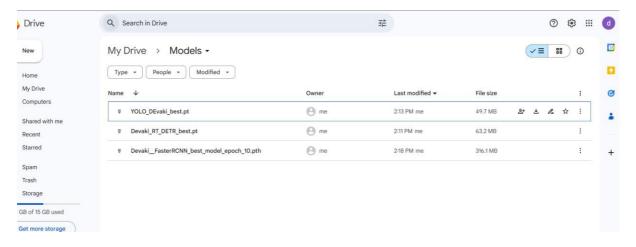
4 Update the Model path in the Demo.ipynb File

4.1 Running the Experiment

- Step 1: Model File Setup
- Required Model Files

Place the trained models in your Google Drive:

- o RT-DETR model: Models/Devaki_RT_DETR_best.pt
- YOLOv8 model: Models/YOLO_DEvaki_best.pt
- Faster R-CNN model: Models/Devaki FasterRCNN_best_model_epoch_10.pth



4.2 Model Path Configuration

Update the following paths in the code as per the directory placed.

For RT-DETR: model_path = "/content/drive/MyDrive/Models/Devaki_RT_DETR_best.pt"
For YOLOv8: model_path = "/content/drive/MyDrive/Models/YOLO_DEvaki_best.pt"
For Faster R-CNN: model_path =

"/content/drive/MyDrive/Models/Devaki__FasterRCNN_best_model_epoch_10.pth"

update the model_path as per the path if changed above directory names.

4.3 Image Input Setup

Individual Images

- 1. Create a folder named 'ISL' in your Google Drive
- 2. Upload test images to this folder
- 3. Update image paths in the code:

folder_path = '/content/drive/MyDrive/ISL'{sample Path}

same here, chang ethe model path and image directory path as per their location, if same names are not followed.

```
△ DEMO.ipynb ☆
        File Edit View Insert Runtime Tools Help Last saved at 5:49 PM
≣
                     ax.set_title(f'Image {idx}: {image_name}', fontsize=14)
        0
                     ax.axis('off')
Q
\{x\}
                     print("Detections:")
                      for score, label in zip(scores, labels):
                     print(f"{detector.class_names[label.item()]}: {score:.4f}")
print("-" * 50)
⊙
plt.tight layout()
                 plt.show()
             def main():
                 model_path = '/content/drive/MyDrive/Models/Devaki__FasterRCNN_best_model_epoch_10.pth'
test_folder = '/content/drive/MyDrive/ISL'
                 print(f"Test folder: {test_folder}")
                 print(f"Number of test images: {len(os.listdir(test_folder))}")
                 process random images(test folder, model path, num images=37)
                 __name
                 main()
<>
```

Video Input

- 1. Upload your test video to Google Drive
- 2. Update video path:

video_path = '/content/drive/MyDrive/NCI/DevakiISL.mp4'{sample Path}

Update the Three model paths corresponding to the directories they are available and vidoeo

path as per the locaiton present.

```
Tools Help Last saved at 5:49 PM
        + Code + Text
≣
                 import GPUT11
           0
Q
                 class CombinedSignLanguageDetector:
                      def __init__(self):
    self.device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
    print(f"Using device: {self.device}")
{x}
☞
                            self.yolo_path = '_content/drive/MyDrive/Models/YOLO_DEvaki_best.pt'
self.rtdetr_path = '_content/drive/MyDrive/Models/Devaki_RT_DETR_best.pt'
self.frcnn_path = '_content/drive/MyDrive/Models/Devaki_FasterRCNN_best_model_epoch_10.pth'
self.yolo_model = YOLO(self.yolo_path)
                            self.rtdetr_model = RTDETR(self.rtdetr_path)
                            self.frcnn_model = self._load_frcnn_model()
                            print("All models loaded successfully!")
                            # Output directory setup
self.output_dir = '/content/drive/MyDrive/predictions
                            Path(self.output dir).mkdir(parents=True, exist ok=True)
                            self.frcnn_classes = {
                                 0: "background", 1: "1", 2: "2", 3: "3", 4: "4", 5: "5", 6: "6", 7: "7", 8: "8", 9: "9", 10: "A", 11: "B", 12: "C", 13: "D", 14: "E", 15: "F", 16: "G", 17: "H", 18: "I", 19: "J", 20: "K", 21: "L", 22: "M", 23: "N", 24: "0", 25: "P", 26: "Q", 27: "R", 28: "S", 29: "T", 30: "U", 31: "V", 32: "W", 33: "X", 24: "V", 35: "7"
<>
>_
           △ DEMO.ipynb ☆
 File Edit View Insert Runtime Tools Help Last saved at 5:49 PM
         + Code + Text
                            return img
Q
                 def main():
{x}
                       detector = CombinedSignLanguageDetector()
                       video_path = '/content/drive/MyDrive/NCI_Sep_Thesis_2024/DevakiISL.mp4' # Replace with your video path
©⊋
detector.process_video(video_path, conf_threshold=0.25)
                      except Exception as e:
                           print(f"Error processing video: {e}")
                 if __name__ == "__main__":
    main()
           Streaming output truncated to the last 5000 lines.

Speed: 2.8ms preprocess, 59.9ms inference, 0.7ms postprocess per image at shape (1, 3, 480, 480)
                 Processing frame 9895
                 0: 640x640 (no detections), 25.5ms
```

4.4 Output Directory Setup

4.4.1 Create directories for results:

/content/drive/MyDrive/predictions/

- Step 4: Running Inference
- For Images
 - Run RT-DETR inference:
 - o Execute RT-DETR section
 - o Expected output: Visualizations with bounding boxes and class labels
 - Run YOLOv8 inference:
 - Execute YOLOv8 section

- o Expected output: Grid visualization of detections
- Run Faster R-CNN inference:
 - o Execute Faster R-CNN section
 - o Expected output: Detection visualizations with confidence scores

For Video

- Run combined video inference:
 - o Execute video inference section
 - o Expected outputs:
 - Individual model predictions
 - Combined visualization grid
 - Performance metrics

4.5 Verification

- Check Output Files
 - Image results:
 - Look for detection visualizations
 - Verify bounding boxes and labels
 - Check confidence scores
 - Video results:
 - o Verify four output videos:
 - YOLO predictions
 - RT-DETR predictions
 - Faster R-CNN predictions
 - Combined grid view

• Verify Performance

- Check console output for:
 - o Processing time
 - o FPS (Frames Per Second)
 - o GPU usage metrics
 - Memory consumption

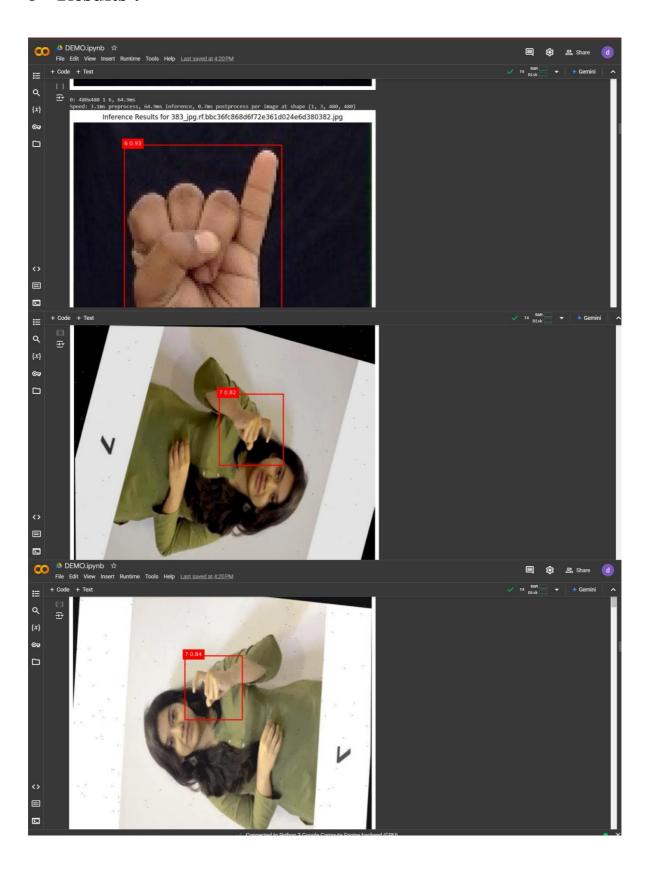
4.6 Troubleshooting

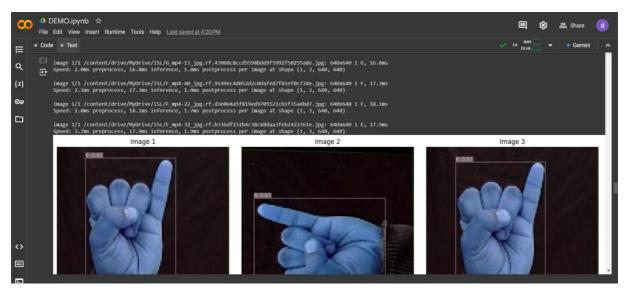
Common Issues:

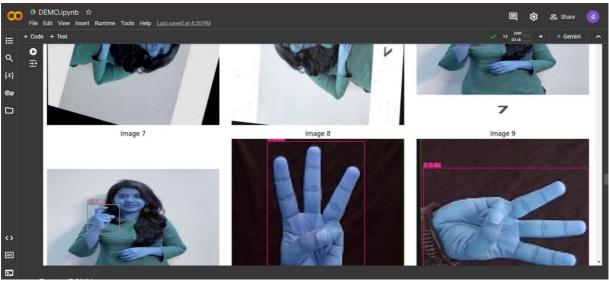
- 1. File not found errors:
 - Verify model paths
 - o Check image/video paths
 - o Confirm directory structure
- 2. Memory issues:
 - o Reduce batch size
 - Process fewer images
 - Clear runtime memory
- 3. GPU errors:
 - o Verify GPU availability
 - Monitor GPU memory
 - Restart runtime if needed

Follow these steps sequentially to ensure proper execution of the ISL recognition system.

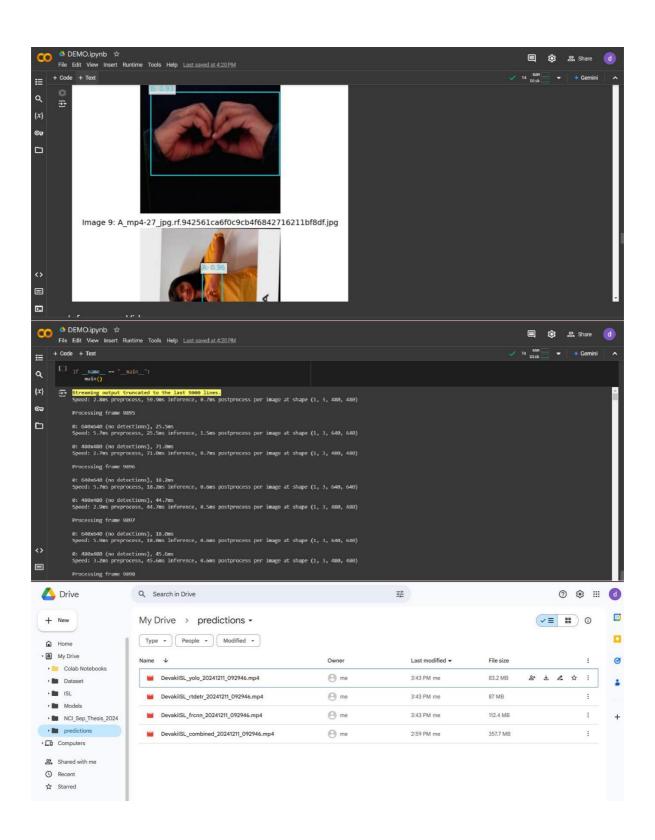
5 Results:

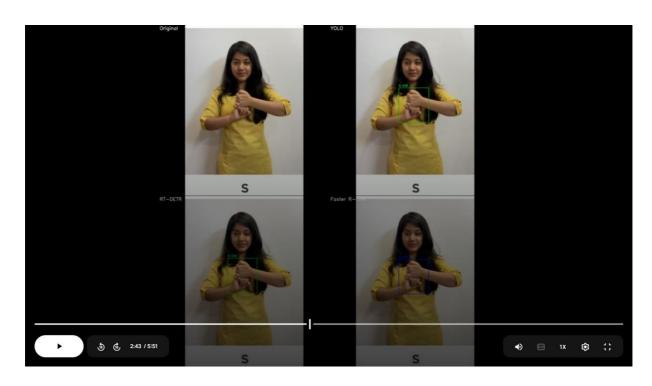


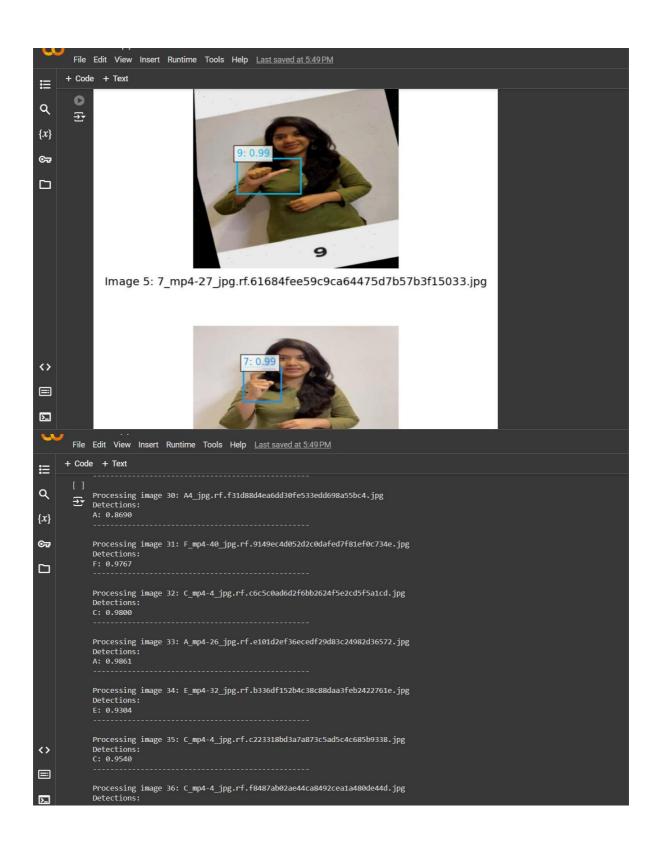


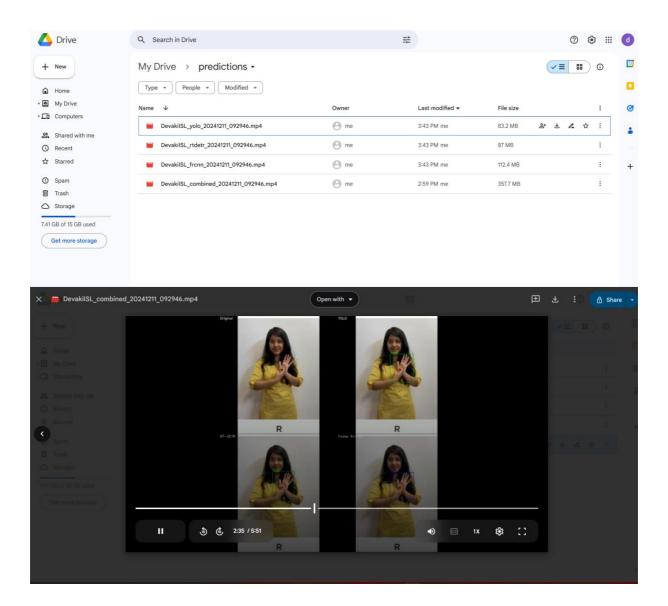












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