## cnn+transformer

## July 22, 2025

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[]: import os
     import cv2
     import json
     video_dir = r"C:\Users\Harsimran⊔
      →Singh\Downloads\WLASL-master\start_kit\raw_videos"
     output_dir = r"C:\Users\Harsimran_\(\text{\text{}}\)
      Singh\Downloads\WLASL-master\start_kit\wlasl_frames"
     sequence_length = 16
     with open(r"C:\Users\Harsimran Singh\Downloads\WLASL-master\start_kit\WLASL_v0.
      ⇒3.json") as f:
         data = json.load(f)
     for entry in data:
         label = entry["gloss"]
         for instance in entry["instances"]:
             video_id = instance["video_id"]
             video_path = os.path.join(video_dir, video_id + ".mp4")
             if not os.path.exists(video_path):
                 continue
             cap = cv2.VideoCapture(video_path)
             frames = []
             while True:
                 ret, frame = cap.read()
                 if not ret:
                     break
                 frames.append(frame)
             cap.release()
             if len(frames) < sequence_length:</pre>
                 continue
             # Sample SEQUENCE_LENGTH evenly spaced frames
             step = len(frames) // sequence_length
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sampled_frames = [frames[i] for i in range(0, len(frames), step)[:
      ⇒sequence_length]]
             label_path = os.path.join(output_dir, label)
             os.makedirs(label_path, exist_ok=True)
             video folder = os.path.join(label path, video id)
             os.makedirs(video_folder, exist_ok=True)
             for i, frame in enumerate(sampled_frames):
                 out_path = os.path.join(video_folder, f"frame_{i:03d}.jpg")
                 cv2.imwrite(out_path, frame)
     print(" Frame extraction complete.")
[1]: import os
     import cv2
     import numpy as np
     import tensorflow as tf
     from tensorflow.keras.models import Model, load_model
     from tensorflow.keras.layers import (Input, TimeDistributed,
      GlobalAveragePooling1D, Dropout, Dense, MultiHeadAttention, Add,
      →LayerNormalization, Embedding)
     from tensorflow.keras.applications import EfficientNetV2B0
     from tensorflow.keras.callbacks import EarlyStopping, LearningRateScheduler, u
      \hookrightarrowModelCheckpoint
     from tensorflow.keras.utils import to_categorical
     from sklearn.model selection import train test split
     from collections import Counter, deque
     import json
     from collections import Counter
[2]: tf.random.set_seed(42)
     np.random.seed(42)
[3]: | IMG_HEIGHT, IMG_WIDTH = 64, 64
     SEQUENCE_LENGTH = 16
     BATCH SIZE = 16
     EPOCHS = 40
[4]: def augment_frame(img):
         if np.random.rand() < 0.3:</pre>
             img = cv2.flip(img, 1)
         if np.random.rand() < 0.3:</pre>
             img = cv2.GaussianBlur(img, (3,3), 0)
         if np.random.rand() < 0.3:</pre>
             hsv = cv2.cvtColor(img, cv2.COLOR_BGR2HSV)
             hsv[...,1] = hsv[...,1] * (0.8 + np.random.rand() * 0.4)
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img = cv2.cvtColor(hsv, cv2.COLOR_HSV2BGR)
if np.random.rand() < 0.3:
    brightness = 0.7 + np.random.rand() * 0.6
    img = np.clip(img * brightness, 0, 255).astype(np.uint8)
return img</pre>
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[5]: def load_dataset(path):
         X, y = [], []
         label_list = sorted(os.listdir(path))
         label_map = {label: idx for idx, label in enumerate(label_list)}
         for label in label_list:
             label_path = os.path.join(path, label)
             for sample_folder in os.listdir(label_path):
                 sample_path = os.path.join(label_path, sample_folder)
                 frames = []
                 frame_files = sorted(os.listdir(sample_path))[:SEQUENCE_LENGTH]
                 for frame_name in frame_files:
                     frame = cv2.imread(os.path.join(sample_path, frame_name))
                     frame = cv2.resize(frame, (IMG_WIDTH, IMG_HEIGHT))
                     frame = augment_frame(frame)
                     frames.append(frame)
                 if len(frames) == SEQUENCE_LENGTH:
                     X.append(frames)
                     y.append(label_map[label])
         class counts = Counter(y)
         valid_classes = {cls for cls, count in class_counts.items() if count >= 2}
         X_filtered, y_filtered = [], []
         for xi, yi in zip(X, y):
             if yi in valid_classes:
                 X_filtered.append(xi)
                 y_filtered.append(yi)
         X = np.array(X_filtered)
         X = X.astype(np.float32)
         mean = np.mean(X)
         std = np.std(X)
         X = (X - mean) / (std + 1e-7)
         print(f" Filtered dataset: {len(valid_classes)} classes retained.")
         return X, to_categorical(y_filtered), label_map
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[6]: def transformer_encoder(inputs, head_size=64, num_heads=4, ff_dim=128, udropout=0.2):
    x = MultiHeadAttention(num_heads=num_heads, key_dim=head_size)(inputs, udinputs)
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x = Dropout(dropout)(x)
         x = Add()([x, inputs])
         x = LayerNormalization()(x)
         ff = Dense(ff_dim, activation='relu')(x)
         ff = Dropout(dropout)(ff)
         ff = Dense(inputs.shape[-1])(ff)
         x = Add()([x, ff])
         return LayerNormalization()(x)
[7]: def add learned positional encoding(x):
         pos = tf.range(start=0, limit=SEQUENCE_LENGTH, delta=1)
         pos_emb = Embedding(input_dim=SEQUENCE_LENGTH, output_dim=x.shape[-1])(pos)
         return x + pos_emb
[8]: def build_model(input_shape, num_classes):
         inputs = Input(shape=input_shape)
         base_cnn = EfficientNetV2BO(include_top=False, input_shape=(IMG_HEIGHT,__
      →IMG_WIDTH, 3), pooling='avg', weights='imagenet')
         base cnn.trainable = True
         x = TimeDistributed(base cnn)(inputs)
         x = Dropout(0.2)(x)
         x = add_learned_positional_encoding(x)
         for _ in range(4):
             x = transformer_encoder(x, head_size=128, num_heads=4, ff_dim=256)
         x = GlobalAveragePooling1D()(x)
         x = Dropout(0.4)(x)
         x = Dense(128, activation='relu')(x)
         outputs = Dense(num_classes, activation='softmax')(x)
         return Model(inputs, outputs)
[9]: X, y, label_map = load_dataset("C:/Users/Harsimran Singh/Downloads/WLASL-master/
      ⇔start_kit/wlasl_frames")
     X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.2,_
      ⇔stratify=y)
     model = build_model((SEQUENCE_LENGTH, IMG_HEIGHT, IMG_WIDTH, 3), len(label_map))
     model.compile(optimizer=tf.keras.optimizers.AdamW(learning rate=1e-4,__
      ⇒weight_decay=1e-5),
                   loss=tf.keras.losses.CategoricalCrossentropy(label_smoothing=0.
      ⇔05),
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metrics=['accuracy', tf.keras.metrics.
 →TopKCategoricalAccuracy(k=3)])
def scheduler(epoch, lr):
   if epoch < 10:
      return lr
   else:
      return lr * tf.math.exp(-0.1)
callbacks = [
   EarlyStopping(patience=7, restore_best_weights=True),
   LearningRateScheduler(scheduler),
   ModelCheckpoint("best_model.h5", save_best_only=True,_
 →monitor='val_accuracy', mode='max')
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model.fit(X_train, y_train, validation_data=(X_val, y_val),
       epochs=EPOCHS, batch_size=BATCH_SIZE, callbacks=callbacks)
 Filtered dataset: 38 classes retained.
Epoch 1/40
0.0338 - top_k_categorical_accuracy: 0.0977 - val_loss: 3.9003 - val_accuracy:
0.0149 - val_top_k_categorical_accuracy: 0.0597 - lr: 1.0000e-04
Epoch 2/40
0.0602 - top_k_categorical_accuracy: 0.1203 - val_loss: 3.7045 - val_accuracy:
0.0299 - val_top_k_categorical_accuracy: 0.1343 - lr: 1.0000e-04
Epoch 3/40
0.0639 - top k_categorical_accuracy: 0.1692 - val_loss: 3.7474 - val_accuracy:
0.0299 - val_top_k_categorical_accuracy: 0.1343 - lr: 1.0000e-04
Epoch 4/40
0.0865 - top k_categorical_accuracy: 0.2105 - val_loss: 3.7414 - val_accuracy:
0.0597 - val_top_k_categorical_accuracy: 0.1045 - lr: 1.0000e-04
Epoch 5/40
0.1316 - top k categorical accuracy: 0.2857 - val loss: 3.7624 - val accuracy:
0.0597 - val_top_k_categorical_accuracy: 0.1194 - lr: 1.0000e-04
Epoch 6/40
0.1805 - top_k_categorical_accuracy: 0.3233 - val_loss: 3.7869 - val_accuracy:
0.1343 - val_top_k_categorical_accuracy: 0.2239 - lr: 1.0000e-04
Epoch 7/40
0.2331 - top_k_categorical_accuracy: 0.4436 - val_loss: 3.6938 - val_accuracy:
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0.2519 - top_k_categorical_accuracy: 0.4812 - val_loss: 4.0741 - val_accuracy:
    0.0746 - val_top_k_categorical_accuracy: 0.1791 - lr: 1.0000e-04
    Epoch 9/40
    0.3722 - top_k_categorical_accuracy: 0.5526 - val_loss: 4.0608 - val_accuracy:
    0.1791 - val_top_k_categorical_accuracy: 0.2239 - lr: 1.0000e-04
    Epoch 10/40
    0.4737 - top k_categorical_accuracy: 0.6353 - val_loss: 4.4786 - val_accuracy:
    0.1045 - val_top_k_categorical_accuracy: 0.2239 - lr: 1.0000e-04
    Epoch 11/40
    0.5526 - top_k_categorical_accuracy: 0.7707 - val_loss: 4.3516 - val_accuracy:
    0.0597 - val_top_k_categorical_accuracy: 0.1045 - lr: 9.0484e-05
    Epoch 12/40
    0.7218 - top_k_categorical_accuracy: 0.8872 - val_loss: 4.8396 - val_accuracy:
    0.1493 - val_top_k_categorical_accuracy: 0.2388 - lr: 8.1873e-05
    Epoch 13/40
    0.7406 - top_k_categorical_accuracy: 0.9135 - val_loss: 4.7247 - val_accuracy:
    0.1642 - val_top_k_categorical_accuracy: 0.2537 - lr: 7.4082e-05
    Epoch 14/40
    0.7744 - top k_categorical_accuracy: 0.9286 - val_loss: 4.5940 - val_accuracy:
    0.1045 - val_top_k_categorical_accuracy: 0.2090 - 1r: 6.7032e-05
[9]: <keras.callbacks.History at 0x1e867a23310>
[12]: model.save(r"C:\Users\Harsimran Singh\Downloads\Untitled_
     →Folder\cnn_transformer_sign_model.h5")
    with open("C:/Users/Harsimran Singh/Downloads/WLASL-master/start kit/label map.

yjson", "w") as f:

       json.dump({k: int(v) for k, v in label_map.items()}, f)
    print("Training complete and model saved.")
    Training complete and model saved.
[11]: print("\n\ud83c\udfa5 Starting real-time prediction. Press 'q' to quit.")
    model = load model("C:/Users/Harsimran Singh/Downloads/WLASL-master/start_kit/

¬cnn_transformer_sign_model.h5")
    with open("C:/Users/Harsimran Singh/Downloads/WLASL-master/start kit/label map.

    json") as f:

       label_map = json.load(f)
```

0.0896 - val\_top\_k\_categorical\_accuracy: 0.1940 - lr: 1.0000e-04

Epoch 8/40

```
inv_label_map = {int(v): k for k, v in label_map.items()}
cap = cv2.VideoCapture(0)
def capture_sequence(cap, seq_len=SEQUENCE_LENGTH):
    frames = []
    for _ in range(seq_len):
        ret, frame = cap.read()
        if not ret:
            break
        frame = cv2.flip(frame, 1)
        resized = cv2.resize(frame, (IMG_WIDTH, IMG_HEIGHT))
        frames.append(resized.astype(np.float32))
    X_seq = np.array(frames)
    X_{seq} = (X_{seq} - np.mean(X_{seq})) / (np.std(X_{seq}) + 1e-7)
    return X_seq
pred_queue = deque(maxlen=5)
while True:
    seq = capture_sequence(cap)
    if seq.shape[0] != SEQUENCE_LENGTH:
        continue
    input_tensor = np.expand_dims(seq, axis=0)
    preds = model.predict(input_tensor, verbose=0)
    label_idx = int(np.argmax(preds))
    pred_queue.append(label_idx)
    if len(pred_queue) == pred_queue.maxlen:
        most_common = Counter(pred_queue).most_common(1)[0][0]
        predicted_word = inv_label_map.get(most_common, "Unknown")
    else:
        predicted_word = inv_label_map.get(label_idx, "Unknown")
    for _ in range(5):
        ret, frame = cap.read()
        if not ret:
            break
        frame = cv2.putText(frame, f"Prediction: {predicted_word}", (20, 50),
                            cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 255, 0), 2)
        cv2.imshow("Sign Language Predictor", frame)
        if cv2.waitKey(30) & OxFF == ord('q'):
            break
    if cv2.waitKey(1) & OxFF == ord('q'):
        break
```

cap.release()
cv2.destroyAllWindows()