PreprocessParquet

December 11, 2023

```
[]: PATH = '../Dataset_GISLR/asl-signs/'
     PROCESSED_OUTPUT_PATH = './Dataset_GISLR_Processed/'
[]: import cv2
     import numpy as np
     import pandas as pd
     from PIL import Image
     from ipywidgets import interact, interactive, fixed, interact_manual
     import ipywidgets as widgets
     import mediapipe as mp
     from mediapipe.framework.formats import landmark_pb2
     mp_drawing = mp.solutions.drawing_utils
     mp_drawing_styles = mp.solutions.drawing_styles
     mp_holistic = mp.solutions.holistic
     import numpy as np
     import matplotlib.pyplot as plt
     import math
     import json
     import os
     import concurrent.futures
```

Iterate over every parquet and do the following: - Find the first 10 frames that contain a consequtive left or right hand - Convert the 10 frames into a mediapipe result structure - Convert each mediapipe result into a numpy array that will be used for training - Store the numpy array into the folder with the matching label/category/sign

```
mp_holistic.FACEMESH_TESSELATION,
                 landmark_drawing_spec=None,
                 connection_drawing_spec=mp_drawing_styles
                 .get_default_face_mesh_tesselation_style())
         if show_face_contour:
             mp_drawing.draw_landmarks(
                 annotated_image,
                 results.face_landmarks,
                 mp holistic.FACEMESH CONTOURS,
                 landmark_drawing_spec=None,
                 connection_drawing_spec=mp_drawing_styles
                 .get_default_face_mesh_contours_style())
         if show pose:
             mp_drawing.draw_landmarks(
                 annotated_image,
                 results.pose_landmarks,
                 mp_holistic.POSE_CONNECTIONS,
                 landmark_drawing_spec=mp_drawing_styles.
                 get_default_pose_landmarks_style())
         if show_left_hand:
             mp_drawing.draw_landmarks(
                 annotated_image,
                 results.left_hand_landmarks,
                 mp holistic. HAND CONNECTIONS,
                 landmark_drawing_spec=mp_drawing_styles
                 .get_default_hand_landmarks_style())
         if show_right_hand:
             mp_drawing.draw_landmarks(
                 annotated_image,
                 results.right_hand_landmarks,
                 mp_holistic.HAND_CONNECTIONS,
                 landmark_drawing_spec=mp_drawing_styles
                 .get_default_hand_landmarks_style())
         return annotated_image
     def display_image(img):
         plt.imshow(img)
         plt.axis('off') # Turn off the axis
         plt.show()
[ ]: def get_avg(example_landmark):
         filtered landmarks = example landmark.dropna(subset=["x", "y", "z"])
         filtered_landmarks = filtered_landmarks[(filtered_landmarks[["x", "y", ]
      \rightarrow"z"]] != 0).all(axis=1)]
         # Get the number of landmarks with x, y, z data per type
         landmarks count = filtered landmarks["type"].value counts()
```

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meta = landmarks_count.to_dict()
        meta["frames"] = filtered_landmarks["frame"].nunique()
        # Identify unique frames with left and right hand landmarks
        left_hand_frames = filtered_landmarks[filtered_landmarks['type'] ==__
      right_hand_frames = filtered_landmarks[filtered_landmarks['type'] ==_
      print(f"Left hand frames: {left_hand_frames}")
        print(f"Right hand frames: {right_hand_frames}")
[]: from mediapipe.framework.formats import landmark_pb2
    class Landmarks(object):
        pass
    def get_landmarks_from_parquet(pf,frame):
        f = pf[pf.frame == frame]
        face = landmark_pb2.NormalizedLandmarkList()
        for t in f[f.type=='face'][['x','y','z']].itertuples(index=False):
            face.landmark.add(x=t.x,y=t.y,z=t.z)
        pose = landmark_pb2.NormalizedLandmarkList()
        for t in f[f.type=='pose'][['x','y','z']].itertuples(index=False):
            pose.landmark.add(x=t.x,y=t.y,z=t.z)
        left_hand = landmark_pb2.NormalizedLandmarkList()
        for t in f[f.type=='left_hand'][['x','y','z']].itertuples(index=False):
            left_hand.landmark.add(x=t.x,y=t.y,z=t.z)
        right hand = landmark pb2.NormalizedLandmarkList()
        for t in f[f.type=='right_hand'][['x','y','z']].itertuples(index=False):
            right_hand.landmark.add(x=t.x,y=t.y,z=t.z)
        result = Landmarks()
        result.face landmarks = face
        result.pose_landmarks = pose
        result.left_hand_landmarks = left_hand
        result.right_hand_landmarks = right_hand
        return result
[]: df = pd.read_csv(PATH + 'train.csv')
    df.head()
[]:
                                               path participant_id sequence_id \
    0 train_landmark_files/26734/1000035562.parquet
                                                              26734
                                                                     1000035562
    1 train_landmark_files/28656/1000106739.parquet
                                                             28656
                                                                     1000106739
    2 train_landmark_files/16069/100015657.parquet
                                                             16069
                                                                     100015657
    3 train_landmark_files/25571/1000210073.parquet
                                                             25571
                                                                     1000210073
```

62590

1000240708

4 train_landmark_files/62590/1000240708.parquet

```
1
       wait
     2 cloud
     3 bird
        owie
[]: def decide_which_array_to_use(left_hand_frames, right_hand_frames):
         if left_hand_frames and right_hand_frames:
             # right hand is more important
             return right_hand_frames
         elif left_hand_frames:
             return left_hand_frames
         elif right_hand_frames:
             return right_hand_frames
         else:
             return None
     def are_landmarks_valid(landmarks):
         return all(not math.isnan(landmark.x) and not math.isnan(landmark.y) and
      onot math.isnan(landmark.z) for landmark in landmarks.landmark)
     def display_all_frames(landmarks_array):
         for i in range(len(landmarks_array)):
             if are_landmarks_valid(landmarks_array[i].pose_landmarks):
                 annotated image = np.zeros((1024,1024,3),dtype=np.uint8)
                 annotated_image = draw_landmarks(landmarks_array[i],annotated_image)
                 display_image(annotated_image)
     def find_first_10_consecutive_frames(landmarks_array, hand_type):
         consecutive_frames = []
         first_10_frames = []
         for landmarks in landmarks_array:
             if hand_type == "left":
                 has_valid_hand = are_landmarks_valid(landmarks.left_hand_landmarks)
             elif hand_type == "right":
                 has_valid_hand = are_landmarks_valid(landmarks.right_hand_landmarks)
             else:
                 raise ValueError("Hand type must be 'left' or 'right'")
             if has_valid_hand:
                 consecutive_frames.append(landmarks)
                 if len(consecutive_frames) == 10:
                     first_10_frames = consecutive_frames.copy()
                     break
             else:
```

sign

blow

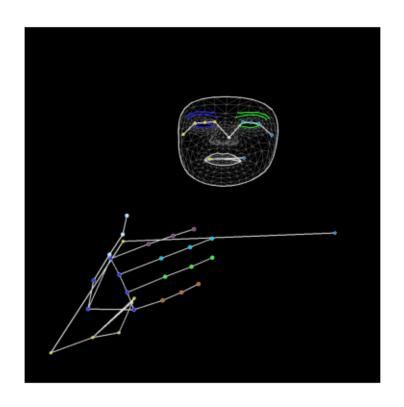
0

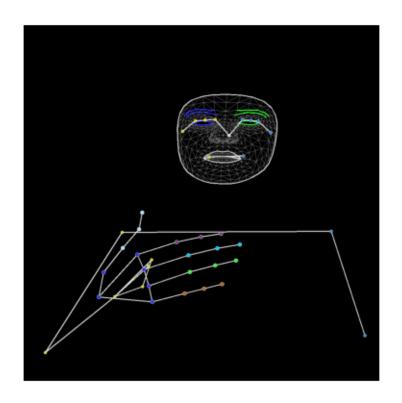
```
consecutive_frames = []
         return first_10_frames
     def process_parquet_file(parquet_file):
         pf = pd.read_parquet(PATH + parquet_file)
         frame_numbers = pf.frame.unique()
         landmarks_array = [get_landmarks_from_parquet(pf, frame) for frame in_
      →frame numbers]
         first_10_frames_with_left_hand =_u
      find_first_10_consecutive_frames(landmarks_array, "left")
         first_10_frames_with_right_hand =_

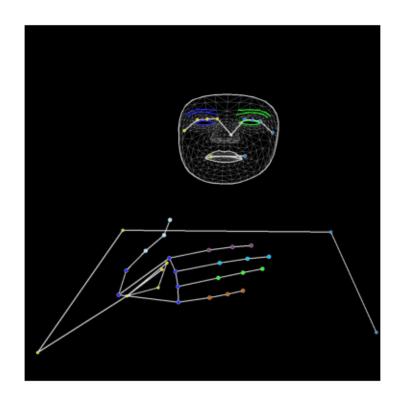
→find_first_10_consecutive_frames(landmarks_array, "right")

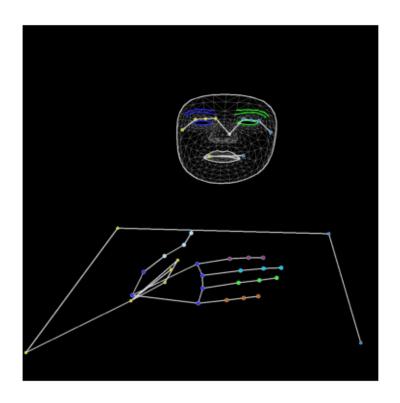
         chosen_frames = decide_which_array_to_use(first_10_frames_with_left_hand,__
      →first_10_frames_with_right_hand)
         return chosen_frames
[]: def display_index(index):
         parquet_file = df.iloc[index].path
         print(f"Sign: {df.iloc[index].sign}")
         chosen_frames = process_parquet_file(parquet_file)
         if chosen_frames is None:
             return None
         display_all_frames(chosen_frames)
     display_index(6)
```

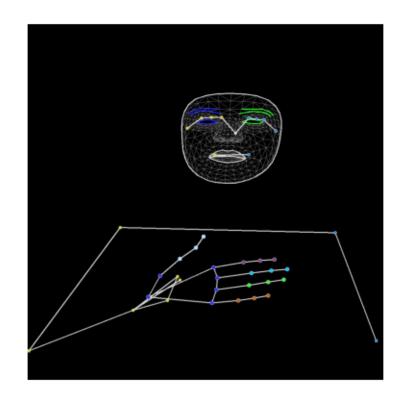
Sign: minemy

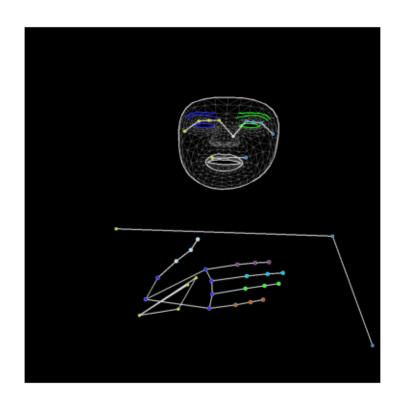


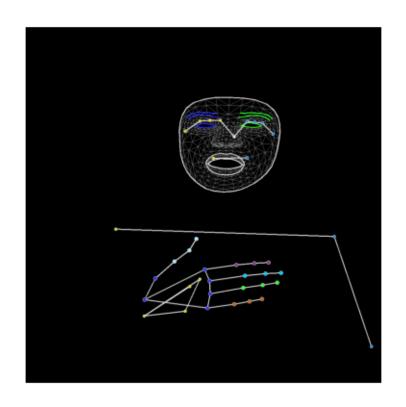


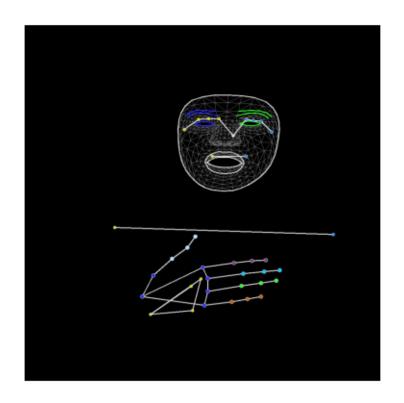


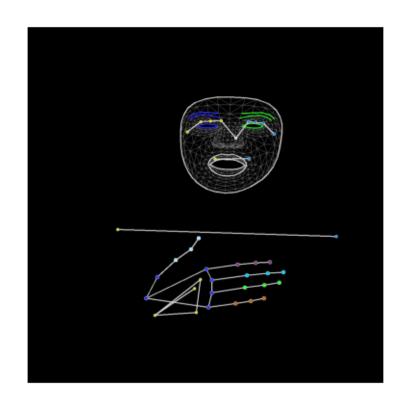


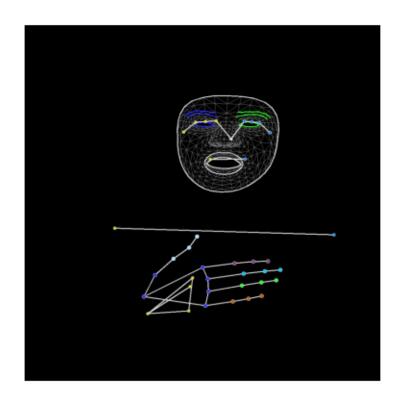












```
[]: def load_json_file(file_path):
    with open(file_path, 'r') as file:
        data = json.load(file)
    return data
    json_file_path = PATH + 'sign_to_prediction_index_map.json'
    category_dict = load_json_file(json_file_path)
    print(category_dict)
```

{'TV': 0, 'after': 1, 'airplane': 2, 'all': 3, 'alligator': 4, 'animal': 5, 'another': 6, 'any': 7, 'apple': 8, 'arm': 9, 'aunt': 10, 'awake': 11, 'backyard': 12, 'bad': 13, 'balloon': 14, 'bath': 15, 'because': 16, 'bed': 17, 'bedroom': 18, 'bee': 19, 'before': 20, 'beside': 21, 'better': 22, 'bird': 23, 'black': 24, 'blow': 25, 'blue': 26, 'boat': 27, 'book': 28, 'boy': 29, 'brother': 30, 'brown': 31, 'bug': 32, 'bye': 33, 'callonphone': 34, 'can': 35, 'car': 36, 'carrot': 37, 'cat': 38, 'cereal': 39, 'chair': 40, 'cheek': 41, 'child': 42, 'chin': 43, 'chocolate': 44, 'clean': 45, 'close': 46, 'closet': 47, 'cloud': 48, 'clown': 49, 'cow': 50, 'cowboy': 51, 'cry': 52, 'cut': 53, 'cute': 54, 'dad': 55, 'dance': 56, 'dirty': 57, 'dog': 58, 'doll': 59, 'donkey': 60, 'down': 61, 'drawer': 62, 'drink': 63, 'drop': 64, 'dry': 65, 'dryer': 66, 'duck': 67, 'ear': 68, 'elephant': 69, 'empty': 70, 'every': 71, 'eye': 72, 'face': 73, 'fall': 74, 'farm': 75, 'fast': 76, 'feet': 77, 'find': 78, 'fine': 79, 'finger': 80, 'finish': 81, 'fireman': 82, 'first': 83, 'fish': 84, 'flag': 85, 'flower': 86, 'food': 87, 'for': 88, 'frenchfries': 89, 'frog': 90, 'garbage': 91, 'gift': 92, 'giraffe': 93, 'girl': 94, 'give': 95, 'glasswindow': 96, 'go': 97, 'goose': 98, 'grandma': 99, 'grandpa': 100, 'grass': 101, 'green': 102, 'gum': 103, 'hair': 104, 'happy': 105, 'hat': 106, 'hate': 107, 'have': 108, 'haveto': 109, 'head': 110, 'hear': 111, 'helicopter': 112, 'hello': 113, 'hen': 114, 'hesheit': 115, 'hide': 116, 'high': 117, 'home': 118, 'horse': 119, 'hot': 120, 'hungry': 121, 'icecream': 122, 'if': 123, 'into': 124, 'jacket': 125, 'jeans': 126, 'jump': 127, 'kiss': 128, 'kitty': 129, 'lamp': 130, 'later': 131, 'like': 132, 'lion': 133, 'lips': 134, 'listen': 135, 'look': 136, 'loud': 137, 'mad': 138, 'make': 139, 'man': 140, 'many': 141, 'milk': 142, 'minemy': 143, 'mitten': 144, 'mom': 145, 'moon': 146, 'morning': 147, 'mouse': 148, 'mouth': 149, 'nap': 150, 'napkin': 151, 'night': 152, 'no': 153, 'noisy': 154, 'nose': 155, 'not': 156, 'now': 157, 'nuts': 158, 'old': 159, 'on': 160, 'open': 161, 'orange': 162, 'outside': 163, 'owie': 164, 'owl': 165, 'pajamas': 166, 'pen': 167, 'pencil': 168, 'penny': 169, 'person': 170, 'pig': 171, 'pizza': 172, 'please': 173, 'police': 174, 'pool': 175, 'potty': 176, 'pretend': 177, 'pretty': 178, 'puppy': 179, 'puzzle': 180, 'quiet': 181, 'radio': 182, 'rain': 183, 'read': 184, 'red': 185, 'refrigerator': 186, 'ride': 187, 'room': 188, 'sad': 189, 'same': 190, 'say': 191, 'scissors': 192, 'see': 193, 'shhh': 194, 'shirt': 195, 'shoe': 196, 'shower': 197, 'sick': 198, 'sleep': 199, 'sleepy': 200, 'smile': 201, 'snack': 202, 'snow': 203, 'stairs': 204, 'stay': 205, 'sticky': 206, 'store': 207, 'story': 208, 'stuck': 209, 'sun': 210, 'table': 211, 'talk': 212, 'taste': 213, 'thankyou': 214, 'that': 215, 'there': 216, 'think': 217, 'thirsty': 218, 'tiger': 219, 'time': 220, 'tomorrow': 221, 'tongue': 222, 'tooth': 223, 'toothbrush': 224, 'touch': 225, 'toy': 226, 'tree': 227, 'uncle': 228, 'underwear': 229, 'up': 230, 'vacuum':

```
231, 'wait': 232, 'wake': 233, 'water': 234, 'wet': 235, 'weus': 236, 'where':
    237, 'white': 238, 'who': 239, 'why': 240, 'will': 241, 'wolf': 242, 'yellow':
    243, 'yes': 244, 'yesterday': 245, 'yourself': 246, 'yucky': 247, 'zebra': 248,
    'zipper': 249}
[]:  # size of df
     print(len(df))
     NUM_TO_TRAIN = len(df)
     print(NUM_TO_TRAIN)
    94477
    94477
[ ]: def extract_keypoints(results):
         pose = np.array([[res.x, res.y, res.z, res.visibility] for res in results.
      pose_landmarks.landmark]).flatten() if results.pose_landmarks else np.
      ⇒zeros(33*4)
         face = np.array([[res.x, res.y, res.z] for res in results.face_landmarks.
      →landmark]).flatten() if results.face_landmarks else np.zeros(468*3)
         lh = np.array([[res.x, res.y, res.z] for res in results.left_hand_landmarks.
      alandmark]).flatten() if results.left_hand_landmarks else np.zeros(21*3)
         rh = np.array([[res.x, res.y, res.z] for res in results.
      oright_hand_landmarks.landmark]).flatten() if results.right_hand_landmarks_□
      ⇔else np.zeros(21*3)
         return np.concatenate([pose, face, lh, rh])
     def process_and_save_sequences(index, output_folder):
         parquet_file = df.iloc[index].path
         label = df.iloc[index].sign
         name = df.iloc[index].sequence_id
         label_folder = os.path.join(output_folder, label)
         os.makedirs(label folder, exist ok=True)
         # make file name with name
         file name = str(name) + ".npy"
         seq_file = os.path.join(label_folder, file_name)
         # print(f"Processing {seq_file}...")
         # check if file exists
         if os.path.isfile(seq_file):
             # print(f"File {seq_file} already exists. Skipping...")
             return None
         chosen_frames = process_parquet_file(parquet_file)
         if chosen_frames is None:
             return None
```

Using 16 threads