

```
In [1]: import cv2
import mediapipe as mp
import matplotlib.pyplot as plt
import numpy as np
import os
import time
```

```
In [2]: mp_holistic= mp.solutions.holistic #Holistic model
mp_drawing= mp.solutions.drawing_utils #Drawing Skeleton in feed
```

```
In [3]: # we are use the model to get holistic skeleton from each frame or image
def mediapipe_holistic_detection(image,model):
    image=cv2.cvtColor(image,cv2.COLOR_BGR2RGB) # Convert BGR(cv2 feed) to RGB
    image.flags.writeable= False
    results = model.process(image)
    image.flags.writeable= True
    image=cv2.cvtColor(image,cv2.COLOR_RGB2BGR) # Convert RGB to BGR
    return image, results
```

```
In [4]: mp_face_mesh = mp.solutions.face_mesh
def draw_landmarks(image, results):
    mp_drawing.draw_landmarks(image, results.pose_landmarks, mp_holistic.POSE_CONNE
    mp_drawing.draw_landmarks(image, results.face_landmarks,mp_face_mesh.FACEMESH_T
    mp_drawing.draw_landmarks(image, results.left_hand_landmarks, mp_holistic.HAND_
    mp_drawing.draw_landmarks(image, results.right_hand_landmarks, mp_holistic.HAND
```

```
In [5]: mp_face_mesh = mp.solutions.face_mesh
DrawingSpec = mp_drawing.DrawingSpec
def draw_design_landmarks(image, results):
    pose_style = DrawingSpec(color=(255, 255, 255), thickness=3, circle_radius=2)
    face_style = DrawingSpec(color=(255, 0, 0), thickness=1, circle_radius=2) # Sm
    hand_style = DrawingSpec(color=(255, 255, 255), thickness=3, circle_radius=2)
    mp_drawing.draw_landmarks(image, results.pose_landmarks, mp_holistic.POSE_CONNE
    mp_drawing.draw_landmarks(image, results.face_landmarks,mp_face_mesh.FACEMESH_T
    mp_drawing.draw_landmarks(image, results.left_hand_landmarks, mp_holistic.HAND_
    mp_drawing.draw_landmarks(image, results.right_hand_landmarks, mp_holistic.HAND
```

```
In [6]: def extract_keypoints(results):
    #pose
    pose=[] #List
    if results.pose_landmarks:
        for res in results.pose_landmarks.landmark: #extracting each list from
            temp = np.array([res.x,res.y,res.z,res.visibility])# coverting each
            pose.append(temp)# appending all np array to pose list
        pose=np.array(pose) # List to np array
        pose = pose.flatten() # from (33,4) to (132)
    else:
        pose= np.zeros(33*4) # handelling the empty values

    #face
    face=[]
    if results.face_landmarks:
        for res in results.face_landmarks.landmark:
```

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        temp = np.array([res.x,res.y,res.z])
        face.append(temp)
        face = np.array(face)
        face = face.flatten()
    else:
        face= np.zeros(468*3)

    #Left hand
    left_hand=[]
    if results.left_hand_landmarks:
        for res in results.left_hand_landmarks.landmark:
            temp = np.array([res.x,res.y,res.z])
            left_hand.append(temp)
        left_hand = np.array(left_hand)
        left_hand = left_hand.flatten()
    else:
        left_hand= np.zeros(21*3)

    #right hand
    right_hand=[]
    if results.right_hand_landmarks:
        for res in results.right_hand_landmarks.landmark:
            temp = np.array([res.x,res.y,res.z])
            right_hand.append(temp)
        right_hand = np.array(right_hand)
        right_hand = right_hand.flatten()
    else:
        right_hand= np.zeros(21*3)

    return np.concatenate([pose,face, left_hand, right_hand])

```

```

In [7]: keypoints_data = os.path.join('Keypoints_Data')

# Create the folder if it doesn't exist
if not os.path.exists(keypoints_data):
    os.makedirs(keypoints_data)

# Actions
actions = ['Hello', 'Thanks', 'I_Love_You', 'Yes', 'No', 'Help', 'Please']
no_sequences = 40 # Number of videos
sequence_length = 30 # Frames per video

```

```

In [9]: for action in actions:
        for sequence in range(no_sequences):
            folder_path = os.path.join(keypoints_data, action, str(sequence))
            if not os.path.exists(folder_path):
                os.makedirs(folder_path)

```

```

In [11]: cap = cv2.VideoCapture(0)

#setting the size of the feed video
cap.set(cv2.CAP_PROP_FRAME_WIDTH, 1000)
cap.set(cv2.CAP_PROP_FRAME_HEIGHT, 600)

```

```

# set up the holistic model with the function
with mp_holistic.Holistic(min_detection_confidence=0.5,min_tracking_confidence=0.5)
    for action in actions:
        for sequence in range(no_sequences):
            for frame_no in range(sequence_length):
                ret, frame= cap.read()# read frame
                image,results=mediapipe_holistic_detection(frame,holistic) #holisti
                draw_landmarks(image,results)# drawing Landmarks

            # Wait Logic
            if frame_no == 0:
                cv2.putText(image, 'Start Collection', (20, 40),
                            cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 0, 0), 2, cv2.LINE_AA)
                cv2.putText(image, f'Collecting frames for {action} Video no {s
                            cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 0, 0), 2, cv2.LINE_AA)
                cv2.waitKey(2000)
            else:
                cv2.putText(image, f'Collecting frames for {action} Video no {s
                            cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 0, 0), 2, cv2.LINE_AA)

            #export keypoints
            keypoints = extract_keypoints(results)
            npy_path= os.path.join(keypoints_data,action,str(sequence),str(frame_no))
            np.save(npy_path,keypoints)

            cv2.imshow("OpenCV", image)
            if cv2.waitKey(1) & 0xFF == ord('q'):
                break

            print(f"Completed sequence {sequence} for action {action}. Waiting for 5 seconds...")
            cv2.putText(image, 'Sequence Completed. Waiting...', (20, 120),
                        cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 0, 0), 2, cv2.LINE_AA)
            cv2.imshow("OpenCV", image)
            cv2.waitKey(5000)

cap.release()
cv2.destroyAllWindows()

```

Completed sequence 39 for action Hello. Waiting for 5 seconds...  
 Completed sequence 39 for action Thanks. Waiting for 5 seconds...  
 Completed sequence 39 for action I\_Love\_You. Waiting for 5 seconds...  
 Completed sequence 39 for action Yes. Waiting for 5 seconds...  
 Completed sequence 39 for action No. Waiting for 5 seconds...  
 Completed sequence 39 for action Help. Waiting for 5 seconds...  
 Completed sequence 39 for action Please. Waiting for 5 seconds...

```

In [8]: from sklearn.model_selection import train_test_split
        from tensorflow.keras.utils import to_categorical

```

```

In [9]: label_map = {}
        for num, label in enumerate(actions):#enumerate provides both index & item in the list
            label_map[label] = num # Each action is added to the dictionary as a key & value

```

```

In [10]: sequences, labels= [], []
         for action in actions:
             for sequence in range(no_sequences):

```

```

window=[]
for frame_no in range(sequence_length):
    res= np.load(os.path.join(keypoints_data,action,str(sequence),"{}.npy").
    window.append(res)
sequences.append(window)
labels.append(label_map[action])

```

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In [11]: X=np.array(sequences)
y= to_categorical(labels).astype(int)

```

```

In [25]: X_train,X_test,y_train,y_test= train_test_split(X,y,test_size=0.05)

```

```

In [26]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dropout, BatchNormalization, Dense, Bidir
from tensorflow.keras.callbacks import TensorBoard

```

```

In [27]: log_dir=os.path.join('Logs')
tb_callback= TensorBoard(log_dir=log_dir)

```

```

In [42]: model = Sequential([
    Bidirectional(LSTM(128, return_sequences=True), input_shape=(30, 1662)),
    Dropout(0.3),
    BatchNormalization(),

    Bidirectional(LSTM(64, return_sequences=True)),
    Dropout(0.3),
    BatchNormalization(),

    GlobalAveragePooling1D(),
    Dense(128, activation='relu'),
    Dropout(0.4),
    Dense(64, activation='relu'),
    Dropout(0.4),
    Dense(len(actions), activation='softmax')
])

```

C:\Users\KIIT\anaconda3\envs\Sign\_Langage\_Detection\Lib\site-packages\keras\src\layers\rnn\bidirectional.py:107: UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```

super().__init__(**kwargs)

```

```

In [29]: model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['categori
model.summary()

```

**Model: "sequential\_1"**


Layer (type)	Output Shape	
bidirectional_2 (Bidirectional)	(None, 30, 256)	
dropout_4 (Dropout)	(None, 30, 256)	
batch_normalization_2 (BatchNormalization)	(None, 30, 256)	
bidirectional_3 (Bidirectional)	(None, 30, 128)	
dropout_5 (Dropout)	(None, 30, 128)	
batch_normalization_3 (BatchNormalization)	(None, 30, 128)	
global_average_pooling1d_1 (GlobalAveragePooling1D)	(None, 128)	
dense_3 (Dense)	(None, 128)	
dropout_6 (Dropout)	(None, 128)	
dense_4 (Dense)	(None, 64)	
dropout_7 (Dropout)	(None, 64)	
dense_5 (Dense)	(None, 7)	


**Total params:** 2,025,095 (7.73 MB)


**Trainable params:** 2,024,327 (7.72 MB)


**Non-trainable params:** 768 (3.00 KB)


```
In [17]: model.fit(X_train,y_train,epochs=35,callbacks=[tb_callback])
```


Epoch 1/35  
9/9  22s 189ms/step - categorical\_accuracy: 0.3257 - loss: 1.6952


Epoch 2/35  
9/9  2s 188ms/step - categorical\_accuracy: 0.5538 - loss: 1.1789


Epoch 3/35  
9/9  2s 186ms/step - categorical\_accuracy: 0.7139 - loss: 0.8095


Epoch 4/35  
9/9  2s 187ms/step - categorical\_accuracy: 0.7848 - loss: 0.6430


Epoch 5/35  
9/9  2s 184ms/step - categorical\_accuracy: 0.8245 - loss: 0.5009


Epoch 6/35  
9/9  2s 180ms/step - categorical\_accuracy: 0.8489 - loss: 0.4114


Epoch 7/35  
9/9  2s 165ms/step - categorical\_accuracy: 0.8792 - loss: 0.3602


Epoch 8/35  
9/9  2s 162ms/step - categorical\_accuracy: 0.8977 - loss: 0.2882


Epoch 9/35  
9/9  2s 168ms/step - categorical\_accuracy: 0.8942 - loss: 0.2919


Epoch 10/35  
9/9  2s 166ms/step - categorical\_accuracy: 0.9121 - loss: 0.2856


Epoch 11/35  
9/9  2s 165ms/step - categorical\_accuracy: 0.9155 - loss: 0.2479


Epoch 12/35  
9/9  2s 172ms/step - categorical\_accuracy: 0.9211 - loss: 0.2377


Epoch 13/35  
9/9  3s 166ms/step - categorical\_accuracy: 0.9051 - loss: 0.2832


Epoch 14/35  
9/9  2s 167ms/step - categorical\_accuracy: 0.9471 - loss: 0.1718


Epoch 15/35  
9/9  2s 165ms/step - categorical\_accuracy: 0.9069 - loss: 0.2248


Epoch 16/35  
9/9  2s 164ms/step - categorical\_accuracy: 0.9207 - loss: 0.1921


Epoch 17/35  
9/9  2s 169ms/step - categorical\_accuracy: 0.9520 - loss: 0.1526


Epoch 18/35  
9/9  2s 168ms/step - categorical\_accuracy: 0.9140 - loss: 0.1828


Epoch 19/35  
9/9  2s 163ms/step - categorical\_accuracy: 0.8833 - loss: 0.3096


Epoch 20/35  
9/9  2s 166ms/step - categorical\_accuracy: 0.9206 - loss: 0.1778


Epoch 21/35  
9/9  2s 171ms/step - categorical\_accuracy: 0.9600 - loss: 0.1289


Epoch 22/35  
9/9  2s 166ms/step - categorical\_accuracy: 0.9718 - loss: 0.0860

Epoch 23/35  
9/9  2s 165ms/step - categorical\_accuracy: 0.9687 - loss: 0.0862

Epoch 24/35  
9/9  2s 165ms/step - categorical\_accuracy: 0.9447 - loss: 0.1576

Epoch 25/35  
9/9  2s 171ms/step - categorical\_accuracy: 0.9861 - loss: 0.0690

Epoch 26/35  
9/9  2s 168ms/step - categorical\_accuracy: 0.9806 - loss: 0.0686

Epoch 27/35  
9/9  2s 164ms/step - categorical\_accuracy: 0.9666 - loss: 0.0983

Epoch 28/35

```

9/9 ————— 2s 166ms/step - categorical_accuracy: 0.9738 - loss: 0.0685
Epoch 29/35
9/9 ————— 2s 170ms/step - categorical_accuracy: 0.9780 - loss: 0.0697
Epoch 30/35
9/9 ————— 2s 167ms/step - categorical_accuracy: 0.9944 - loss: 0.0278
Epoch 31/35
9/9 ————— 2s 162ms/step - categorical_accuracy: 0.9865 - loss: 0.0467
Epoch 32/35
9/9 ————— 2s 166ms/step - categorical_accuracy: 0.9906 - loss: 0.0396
Epoch 33/35
9/9 ————— 2s 165ms/step - categorical_accuracy: 0.9575 - loss: 0.1614
Epoch 34/35
9/9 ————— 2s 167ms/step - categorical_accuracy: 0.9764 - loss: 0.1075
Epoch 35/35
9/9 ————— 2s 164ms/step - categorical_accuracy: 1.0000 - loss: 0.0207

```

Out[17]: <keras.src.callbacks.history.History at 0x2a0fdc76810>

In [24]: `del model`

In [30]: `model.load_weights('my_model.keras')`

In [31]: `res=model.predict(X_test)`

```

1/1 ————— 2s 2s/step

```

In [32]: `print(actions[np.argmax(res[1])])`  
`print(actions[np.argmax(res[6])])`

Yes  
Help

In [33]: `print(actions[np.argmax(y_test[1])])`  
`print(actions[np.argmax(y_test[6])])`

Yes  
Help

In [34]: `from sklearn.metrics import multilabel_confusion_matrix,accuracy_score`  
`res= np.argmax(res, axis=1).tolist() # Convert predicted results to a list of indi`  
`true = np.argmax(y_test, axis=1).tolist()`

In [35]: `multilabel_confusion_matrix(res,true)`

```
Out[35]: array([[12, 0],
               [ 0, 2]],

               [[13, 1],
               [ 0, 0]],

               [[13, 0],
               [ 0, 1]],

               [[11, 0],
               [ 0, 3]],

               [[13, 0],
               [ 0, 1]],

               [[10, 0],
               [ 0, 4]],

               [[11, 0],
               [ 1, 2]]], dtype=int64)
```

```
In [36]: accuracy_score(res,true)
```

```
Out[36]: 0.9285714285714286
```

```
In [25]: model.save('my_model1.keras')
```

```
In [41]: sequence = []
sentence = []
predictions = []
threshold = 0.5

colors = [
    (255, 0, 0), (0, 255, 0), (0, 0, 255),
    (255, 255, 0), (255, 0, 255), (0, 255, 255),
    (128, 128, 128)
]

cap = cv2.VideoCapture(0)

#setting the size of the feed video
cap.set(cv2.CAP_PROP_FRAME_WIDTH, 1000)
cap.set(cv2.CAP_PROP_FRAME_HEIGHT, 600)

with mp.solutions.holistic.Holistic(min_detection_confidence=0.5, min_tracking_confidence=0.5) as holistic:
    while cap.isOpened():
        ret, frame = cap.read()
        if not ret:
            print("Error: Unable to read from camera")
            break

        image, results = mediapipe_holistic_detection(frame, holistic)
        draw_landmarks(image, results)

        keypoints = extract_keypoints(results)
```



```

sequence.append(keypoints)
sequence = sequence[-30:]

if len(sequence) == 30:
    res = model.predict(np.expand_dims(sequence, axis=0))[0]
    predictions.append(np.argmax(res))

    if np.unique(predictions[-10:])[0] == np.argmax(res):
        if res[np.argmax(res)] > threshold:
            if len(sentence) == 0 or actions[np.argmax(res)] != sentence[-1]:
                sentence.append(actions[np.argmax(res)])

    if len(sentence) > 5:
        sentence = sentence[-5:]

    for idx, prob in enumerate(res):
        cv2.rectangle(image, (0, 60 + idx * 40), (int(prob * 200), 90 + idx * 40),
                       (255, 255, 255), 2)
        cv2.putText(
            image, f"{actions[idx]}: {prob:.2f}",
            (10, 85 + idx * 40), cv2.FONT_HERSHEY_SIMPLEX, 0.5, (255, 255, 255)
        )

cv2.putText(
    image, ' '.join(sentence),
    (image.shape[1] // 2 - len(' '.join(sentence)) * 7, image.shape[0] - 20),
    cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2, cv2.LINE_AA
)

# Show the feed
cv2.imshow('Sign Language Detection', image)

# Graceful exit
if cv2.waitKey(10) & 0xFF == ord('q'):
    break

cap.release()
cv2.destroyAllWindows()

```

1/1	—————	0s	48ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	92ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	70ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	91ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	65ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	82ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	54ms/step

1/1	—————	0s	47ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	88ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	63ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	84ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	58ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	34ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	61ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	57ms/step
1/1	—————	0s	62ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	63ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	80ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	50ms/step

1/1	—————	0s	34ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	36ms/step
1/1	—————	0s	78ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	78ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	75ms/step
1/1	—————	0s	58ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	79ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	61ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	84ms/step

1/1	—————	0s	35ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	61ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	34ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	70ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	61ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	32ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	68ms/step
1/1	—————	0s	36ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	33ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	37ms/step
1/1	—————	0s	66ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	64ms/step

1/1	—————	0s	52ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	76ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	43ms/step
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1/1	—————	0s	60ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	68ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	61ms/step
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1/1	—————	0s	42ms/step
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1/1	—————	0s	47ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	53ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	36ms/step
1/1	—————	0s	37ms/step

1/1	—————	0s	50ms/step
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1/1	—————	0s	51ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	58ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	51ms/step
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1/1	—————	0s	50ms/step
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1/1	—————	0s	63ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	48ms/step

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1/1	—————	0s	44ms/step
1/1	—————	0s	74ms/step
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1/1	—————	0s	49ms/step
1/1	—————	0s	58ms/step
1/1	—————	0s	67ms/step
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1/1	—————	0s	49ms/step
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1/1	—————	0s	43ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	65ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	62ms/step
1/1	—————	0s	123ms/step



1/1	—————	0s	44ms/step
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1/1	—————	0s	50ms/step
1/1	—————	0s	50ms/step
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1/1	—————	0s	45ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	62ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	61ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	64ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	114ms/step
1/1	—————	0s	82ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	42ms/step
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1/1	—————	0s	38ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	82ms/step
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1/1	—————	0s	61ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	66ms/step
1/1	—————	0s	57ms/step
1/1	—————	0s	34ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	36ms/step
1/1	—————	0s	42ms/step
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1/1	—————	0s	50ms/step
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1/1	—————	0s	44ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	35ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	67ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	64ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	57ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	62ms/step
1/1	—————	0s	47ms/step

1/1	_____	0s	61ms/step
1/1	_____	0s	39ms/step
1/1	_____	0s	48ms/step
1/1	_____	0s	54ms/step
1/1	_____	0s	57ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	48ms/step
1/1	_____	0s	57ms/step
1/1	_____	0s	48ms/step
1/1	_____	0s	48ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	42ms/step
1/1	_____	0s	95ms/step
1/1	_____	0s	47ms/step
1/1	_____	0s	45ms/step
1/1	_____	0s	39ms/step
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1/1	_____	0s	51ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	51ms/step
1/1	_____	0s	42ms/step
1/1	_____	0s	54ms/step
1/1	_____	0s	69ms/step
1/1	_____	0s	50ms/step
1/1	_____	0s	50ms/step
1/1	_____	0s	52ms/step
1/1	_____	0s	53ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	43ms/step
1/1	_____	0s	37ms/step
1/1	_____	0s	48ms/step
1/1	_____	0s	55ms/step
1/1	_____	0s	52ms/step
1/1	_____	0s	48ms/step
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1/1	_____	0s	44ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	59ms/step
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1/1	_____	0s	43ms/step
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1/1	_____	0s	43ms/step
1/1	_____	0s	51ms/step
1/1	_____	0s	46ms/step
1/1	_____	0s	37ms/step
1/1	_____	0s	43ms/step
1/1	_____	0s	47ms/step
1/1	_____	0s	35ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	57ms/step
1/1	_____	0s	97ms/step

1/1	—————	0s	39ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	57ms/step
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1/1	—————	0s	54ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	59ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	52ms/step
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1/1	—————	0s	41ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	44ms/step

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1/1	_____	0s	49ms/step
1/1	_____	0s	51ms/step
1/1	_____	0s	45ms/step
1/1	_____	0s	48ms/step
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1/1	_____	0s	51ms/step
1/1	_____	0s	43ms/step
1/1	_____	0s	42ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	73ms/step
1/1	_____	0s	37ms/step
1/1	_____	0s	50ms/step
1/1	_____	0s	40ms/step
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1/1	_____	0s	44ms/step
1/1	_____	0s	86ms/step
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1/1	_____	0s	44ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	43ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	58ms/step
1/1	_____	0s	38ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	33ms/step
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1/1	_____	0s	54ms/step
1/1	_____	0s	56ms/step
1/1	_____	0s	80ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	54ms/step
1/1	_____	0s	37ms/step
1/1	_____	0s	49ms/step
1/1	_____	0s	44ms/step
1/1	_____	0s	40ms/step
1/1	_____	0s	93ms/step
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1/1	_____	0s	52ms/step
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1/1	_____	0s	50ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	42ms/step
1/1	_____	0s	42ms/step
1/1	_____	0s	67ms/step
1/1	_____	0s	57ms/step
1/1	_____	0s	39ms/step
1/1	_____	0s	49ms/step
1/1	_____	0s	52ms/step
1/1	_____	0s	41ms/step
1/1	_____	0s	50ms/step
1/1	_____	0s	43ms/step

1/1	0s	36ms/step
1/1	0s	60ms/step
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1/1	0s	69ms/step
1/1	0s	48ms/step
1/1	0s	45ms/step
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1/1	0s	58ms/step
1/1	0s	47ms/step
1/1	0s	42ms/step
1/1	0s	53ms/step
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1/1	0s	54ms/step
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1/1	0s	54ms/step
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1/1	0s	58ms/step
1/1	0s	40ms/step
1/1	0s	40ms/step
1/1	0s	41ms/step
1/1	0s	48ms/step
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1/1	0s	57ms/step
1/1	0s	43ms/step
1/1	0s	57ms/step
1/1	0s	77ms/step
1/1	0s	50ms/step
1/1	0s	67ms/step
1/1	0s	44ms/step
1/1	0s	57ms/step

1/1	—————	0s	50ms/step
1/1	—————	0s	60ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	55ms/step
1/1	—————	0s	81ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	64ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	35ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	80ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	47ms/step
1/1	—————	0s	106ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	78ms/step
1/1	—————	0s	57ms/step
1/1	—————	0s	54ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	50ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	36ms/step
1/1	—————	0s	62ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	42ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	41ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	37ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	38ms/step

1/1	—————	0s	51ms/step
1/1	—————	0s	39ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	40ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	52ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	49ms/step
1/1	—————	0s	45ms/step
1/1	—————	0s	57ms/step
1/1	—————	0s	53ms/step
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1/1	—————	0s	54ms/step
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1/1	—————	0s	47ms/step
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1/1	—————	0s	42ms/step
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1/1	—————	0s	55ms/step
1/1	—————	0s	43ms/step
1/1	—————	0s	46ms/step
1/1	—————	0s	38ms/step
1/1	—————	0s	37ms/step
1/1	—————	0s	48ms/step
1/1	—————	0s	56ms/step
1/1	—————	0s	35ms/step
1/1	—————	0s	51ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	44ms/step
1/1	—————	0s	50ms/step

In [ ]: