## pytorch-training-visualisation

## November 24, 2024

```
[1]: import os
     import numpy as np
     import cv2
     !pip install mediapipe
     import mediapipe as mp
     import matplotlib.pyplot as plt
     import torch
     import torch.nn as nn
     import torch.optim as optim
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import LabelEncoder
     from sklearn.metrics import accuracy_score, classification_report
     import seaborn as sns
     import pandas as pd
    Requirement already satisfied: mediapipe in /opt/conda/lib/python3.10/site-
    packages (0.10.15)
    Requirement already satisfied: absl-py in /opt/conda/lib/python3.10/site-
    packages (from mediapipe) (1.4.0)
    Requirement already satisfied: attrs>=19.1.0 in /opt/conda/lib/python3.10/site-
    packages (from mediapipe) (23.2.0)
    Requirement already satisfied: flatbuffers>=2.0 in
    /opt/conda/lib/python3.10/site-packages (from mediapipe) (24.3.25)
    Requirement already satisfied: jax in /opt/conda/lib/python3.10/site-packages
    (from mediapipe) (0.4.33)
    Requirement already satisfied: jaxlib in /opt/conda/lib/python3.10/site-packages
    (from mediapipe) (0.4.33)
    Requirement already satisfied: matplotlib in /opt/conda/lib/python3.10/site-
    packages (from mediapipe) (3.7.5)
    Requirement already satisfied: numpy<2 in /opt/conda/lib/python3.10/site-
    packages (from mediapipe) (1.26.4)
    Requirement already satisfied: opency-contrib-python in
    /opt/conda/lib/python3.10/site-packages (from mediapipe) (4.10.0.84)
    Requirement already satisfied: protobuf<5,>=4.25.3 in
    /opt/conda/lib/python3.10/site-packages (from mediapipe) (4.25.5)
    Requirement already satisfied: sounddevice>=0.4.4 in
    /opt/conda/lib/python3.10/site-packages (from mediapipe) (0.5.1)
    Requirement already satisfied: CFFI>=1.0 in /opt/conda/lib/python3.10/site-
```

```
packages (from sounddevice>=0.4.4->mediapipe) (1.16.0)
    Requirement already satisfied: ml-dtypes>=0.2.0 in
    /opt/conda/lib/python3.10/site-packages (from jax->mediapipe) (0.3.2)
    Requirement already satisfied: opt-einsum in /opt/conda/lib/python3.10/site-
    packages (from jax->mediapipe) (3.3.0)
    Requirement already satisfied: scipy>=1.10 in /opt/conda/lib/python3.10/site-
    packages (from jax->mediapipe) (1.14.1)
    Requirement already satisfied: contourpy>=1.0.1 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe) (1.2.1)
    Requirement already satisfied: cycler>=0.10 in /opt/conda/lib/python3.10/site-
    packages (from matplotlib->mediapipe) (0.12.1)
    Requirement already satisfied: fonttools>=4.22.0 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe) (4.53.0)
    Requirement already satisfied: kiwisolver>=1.0.1 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe) (1.4.5)
    Requirement already satisfied: packaging>=20.0 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe) (21.3)
    Requirement already satisfied: pillow>=6.2.0 in /opt/conda/lib/python3.10/site-
    packages (from matplotlib->mediapipe) (10.3.0)
    Requirement already satisfied: pyparsing>=2.3.1 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe) (3.1.2)
    Requirement already satisfied: python-dateutil>=2.7 in
    /opt/conda/lib/python3.10/site-packages (from matplotlib->mediapipe)
    (2.9.0.post0)
    Requirement already satisfied: pycparser in /opt/conda/lib/python3.10/site-
    packages (from CFFI>=1.0->sounddevice>=0.4.4->mediapipe) (2.22)
    Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-
    packages (from python-dateutil>=2.7->matplotlib->mediapipe) (1.16.0)
[2]: # Function to extract pose landmarks from a frame
     def get_landmark(frame, pose):
        rgb_frame = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
        results = pose.process(rgb_frame)
         if results.pose_landmarks:
             landmarks = np.array([[landmark.x, landmark.y, landmark.z] for landmark_u
      →in results.pose_landmarks.landmark])
            return landmarks
        else:
             return None
[3]: # Main function to process video files and assign labels
     def video_process(root_folder):
        mp pose = mp.solutions.pose
        pose = mp_pose.Pose(static_image_mode=False, min_detection_confidence=0.5,__
      →min_tracking_confidence=0.5)
        dataset = []
        labels = []
```

```
sign_folders = os.listdir(root_folder)
for sign_folder in sign_folders:
    sign_path = os.path.join(root_folder, sign_folder)
    if os.path.isdir(sign_path):
        for filename in os.listdir(sign_path):
            if filename.endswith('.MOV'):
                filepath = os.path.join(sign_path, filename)
                cap = cv2.VideoCapture(filepath)
                while cap.isOpened():
                    ret, frame = cap.read()
                    if not ret:
                        hreak
                    landmarks = get_landmark(frame, pose)
                    if landmarks is not None:
                        dataset.append(landmarks)
                        labels.append(sign_folder)
                cap.release()
dataset = np.array(dataset)
labels = np.array(labels)
np.save('pose_landmarks_dataset.npy', dataset)
np.save('pose_landmarks_labels.npy', labels)
pose.close()
```

```
[]: if __name__ == "__main__": video_process('/kaggle/input/include/Adjectives_1of8/Adjectives')
```

INFO: Created TensorFlow Lite XNNPACK delegate for CPU.

WARNING: All log messages before absl::InitializeLog() is called are written to STDERR

W0000 00:00:1729827680.019930 1321 inference\_feedback\_manager.cc:114]

Feedback manager requires a model with a single signature inference. Disabling support for feedback tensors.

W0000 00:00:1729827680.083520 1321 inference feedback manager.cc:114]

Feedback manager requires a model with a single signature inference. Disabling support for feedback tensors.

W0000 00:00:1729827680.128115 1321 landmark\_projection\_calculator.cc:186]

Using NORM\_RECT without IMAGE\_DIMENSIONS is only supported for the square ROI. Provide IMAGE\_DIMENSIONS or use PROJECTION\_MATRIX.

/opt/conda/lib/python3.10/site-packages/google/protobuf/symbol\_database.py:55:

 ${\tt UserWarning: SymbolDatabase.GetPrototype() is deprecated. \ Please \ use}$ 

message\_factory.GetMessageClass() instead. SymbolDatabase.GetPrototype() will be removed soon.

warnings.warn('SymbolDatabase.GetPrototype() is deprecated. Please '

```
[5]: # Load the dataset
     x = np.load('pose_landmarks_dataset.npy')
     y = np.load('pose_landmarks_labels.npy')
[6]: x[0]
                           0.34332821, -0.60569352],
[6]: array([[ 0.48804522,
            [ 0.49639684,
                           0.32307142, -0.58474225],
                           0.32332098, -0.584759 ],
            [ 0.50148481,
            [ 0.50600356,
                           0.32398093, -0.58486313],
                           0.3248412 , -0.58535296],
            [ 0.47754201,
            [ 0.47062844,
                           0.32658514, -0.58544183,
                           0.32879278, -0.58549827,
            [ 0.46449503,
                           0.33910868, -0.46264228],
            [ 0.51446033,
            [ 0.45775276,
                           0.34610149, -0.46666297],
                           0.3737804 , -0.56138498],
            [ 0.49964294,
            [ 0.47681263,
                           0.37546605, -0.56244284],
                           0.50288129, -0.39907786],
            [ 0.55738395,
            [ 0.42038509,
                           0.49544892, -0.41369334],
                           0.69080114, -0.41229066],
            [ 0.56581622,
                           0.70087284, -0.43329164],
            [ 0.41551223,
                           0.88313276, -0.49369031],
            [ 0.57494098,
            [ 0.40359646,
                           0.89318508, -0.51685232],
                           0.92966568, -0.53621918,
            [ 0.5813731 ,
                           0.94439512, -0.56358778,
            [ 0.39398205,
            [ 0.56915253,
                           0.93214804, -0.5644899],
                           0.94479257, -0.59031284,
            [0.40971729,
            [ 0.5647279 ,
                           0.92184383, -0.5084672],
                           0.92987722, -0.53157109],
            [ 0.41569763,
            [ 0.52013367,
                           0.74957937, -0.00393986,
            [ 0.44403163,
                           0.74465775,
                                         0.00370453],
                           0.89091265,
            [ 0.52481782,
                                         0.10332326],
            [ 0.43262118,
                           0.85476977,
                                         0.16239275],
            [ 0.50152683,
                           0.74944651,
                                         0.52990156],
            [ 0.44550464,
                           0.72069395,
                                         0.51354092],
            [ 0.4989064 ,
                           0.71795338,
                                         0.57534951],
            [ 0.44526353,
                           0.70369107,
                                         0.55287141],
            [ 0.4919773 ,
                           0.75878739,
                                         0.58874226],
            [ 0.45748508,
                           0.7270627 ,
                                         0.54836553]])
[7]:
     x.shape
[7]: (5991, 33, 3)
    y[0]
[8]:
[8]: '1. loud'
```

```
[9]: y.shape
 [9]: (5991,)
[11]: # Main function to process multiple videos from a folder and its subfolders
      import random
      def visualize_middle_frames_from_videos(folder_path, num_videos=5):
          video_files = []
          # Traverse the folder and subfolders to find video files
          for root, _, files in os.walk(folder_path):
              for file in files:
                  if file.endswith('.MOV') or file.endswith('.mp4'):
                      video_files.append(os.path.join(root, file))
          if len(video_files) == 0:
              print(f'No video files found in {folder_path}.')
              return
          # Randomly select num videos from the collected video files
          selected_videos = random.sample(video_files, min(num_videos,__
       →len(video_files)))
          # Display the middle frame from each selected video
          for video_file in selected_videos:
              visualize_middle_frame_from_video(video_file)
      # Example usage
      visualize_middle_frames_from_videos('/kaggle/input/include/Adjectives_1of8/
       →Adjectives', num_videos=5)
```

Middle frame index for /kaggle/input/include/Adjectives\_1of8/Adjectives/2.quiet/MVI\_9451.MOV: 28

Middle Frame from MVI\_9451.MOV (Frame 28)



Middle frame index for /kaggle/input/include/Adjectives\_1of8/Adjectives/3. happy/MVI\_9297.MOV: 24

Middle Frame from MVI\_9297.MOV (Frame 24)



Middle frame index for /kaggle/input/include/Adjectives\_1of8/Adjectives/2.quiet/MVI\_9372.MOV: 23

Middle Frame from MVI\_9372.MOV (Frame 23)



Middle frame index for /kaggle/input/include/Adjectives\_1of8/Adjectives/1. loud/MVI\_9369.MOV: 24

Middle Frame from MVI\_9369.MOV (Frame 24)



Middle frame index for /kaggle/input/include/Adjectives\_1of8/Adjectives/5. Beautiful/MVI\_9570.MOV: 39

## Middle Frame from MVI\_9570.MOV (Frame 39)



```
[12]: # Split data into training and testing sets
      x_train, x_val, y_train, y_val = train_test_split(x, y, test_size=0.2,_
       →random_state=42)
[13]: # Initialize label encoder
      label_encoder = LabelEncoder()
      y_train_encoded = label_encoder.fit_transform(y_train)
      y_val_encoded = label_encoder.transform(y_val)
[32]: class LSTMModel(nn.Module):
          def __init__(self, num_classes):
              super(LSTMModel, self).__init__()
              self.lstm1 = nn.LSTM(input_size=x_train.shape[2], hidden_size=64,__
       ⇒batch_first=True, dropout=0.4)
              self.lstm2 = nn.LSTM(64, 32, batch_first=True, dropout=0.4)
              self.lstm3 = nn.LSTM(32, 16, batch_first=True, dropout=0.4)
              self.fc = nn.Linear(16, num_classes)
          def forward(self, x):
              out, _{-} = self.lstm1(x)
              out = out[:, -1, :] # Get last output from the sequence
              out, _ = self.lstm2(out.unsqueeze(1)) # Ensure correct shape
              out = out[:, -1, :]
              out, _ = self.lstm3(out.unsqueeze(1))
              out = out[:, -1, :]
              out = self.fc(out) # Final output
              return out
```

```
[34]: # Get the number of classes
      num_classes = len(np.unique(y_train_encoded))
      # Initialize the model with both input_size and num_classes
      model = LSTMModel(num_classes)
      # Move the model to the appropriate device (CPU/GPU)
      device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
      model.to(device)
     /opt/conda/lib/python3.10/site-packages/torch/nn/modules/rnn.py:88: UserWarning:
     dropout option adds dropout after all but last recurrent layer, so non-zero
     dropout expects num_layers greater than 1, but got dropout=0.4 and num_layers=1
       warnings.warn("dropout option adds dropout after all but last "
[34]: LSTMModel(
        (lstm1): LSTM(3, 64, batch_first=True, dropout=0.4)
        (lstm2): LSTM(64, 32, batch_first=True, dropout=0.4)
        (lstm3): LSTM(32, 16, batch_first=True, dropout=0.4)
        (fc): Linear(in_features=16, out_features=8, bias=True)
      )
[35]: # Convert training and testing data to tensors
      x_train_tensor = torch.Tensor(x_train).to(device)
      y train tensor = torch. Tensor(y train encoded).long().to(device)
      x_val_tensor = torch.Tensor(x_val).to(device)
      y_val_tensor = torch.Tensor(y_val_encoded).long().to(device)
[36]: from torch.utils.data import DataLoader, TensorDataset
      train_dataset = TensorDataset(x_train_tensor, y_train_tensor)
      val_dataset = TensorDataset(x_val_tensor, y_val_tensor)
[37]: batch_size = 32
      train loader = DataLoader(train dataset, batch size=batch size, shuffle=True)
      val_loader = DataLoader(val_dataset, batch_size=batch_size, shuffle=False)
      # Training Loop
      criterion = nn.CrossEntropyLoss()
      optimizer = optim.Adam(model.parameters(), lr=0.005)
[38]: # Initialize arrays to store training and validation loss and accuracy for each
       ⇔epoch
      train_losses = []
      val losses = []
      train_accuracies = []
      val_accuracies = []
```

```
num_epochs = 500
for epoch in range(num_epochs):
    # Training Phase
   model.train()
   running_train_loss = 0.0
   correct_train = 0
   total_train = 0
   for batch_x, batch_y in train_loader:
        optimizer.zero_grad()
       outputs = model(batch_x)
       loss = criterion(outputs, batch_y)
       loss.backward()
       optimizer.step()
       running_train_loss += loss.item()
        # Calculate training accuracy
        _, predicted = torch.max(outputs, 1)
        total_train += batch_y.size(0)
        correct_train += (predicted == batch_y).sum().item()
   avg_train_loss = running_train_loss / len(train_loader)
   train_accuracy = 100 * correct_train / total_train
    # Store training loss and accuracy for the current epoch
   train_losses.append(avg_train_loss)
   train_accuracies.append(train_accuracy)
   # Validation Phase
   model.eval() # Put the model in evaluation mode
   running_val_loss = 0.0
   correct_val = 0
   total_val = 0
   with torch.no_grad(): # Disable gradient computation during validation
        for batch_x, batch_y in val_loader:
            outputs = model(batch_x)
            loss = criterion(outputs, batch y)
            running_val_loss += loss.item()
            # Calculate validation accuracy
            _, predicted = torch.max(outputs, 1)
            total_val += batch_y.size(0)
            correct_val += (predicted == batch_y).sum().item()
    avg_val_loss = running_val_loss / len(val_loader)
```

```
val_accuracy = 100 * correct_val / total_val
     # Store validation loss and accuracy for the current epoch
     val_losses.append(avg_val_loss)
     val_accuracies.append(val_accuracy)
     print(f'Epoch [{epoch + 1}/{num_epochs}], '
            f'Train Loss: {avg_train_loss:.4f}, Train Accuracy: {train_accuracy:.

</p
            f'Val Loss: {avg_val_loss:.4f}, Val Accuracy: {val_accuracy:.2f}%')
# After training, you can access `train_losses`, `val_losses`, _
  → `train_accuracies`, and `val_accuracies` arrays.
Epoch [1/500], Train Loss: 1.9958, Train Accuracy: 20.26%, Val Loss: 1.9872, Val
Accuracy: 19.68%
Epoch [2/500], Train Loss: 1.9886, Train Accuracy: 20.83%, Val Loss: 1.9858, Val
Accuracy: 19.68%
Epoch [3/500], Train Loss: 1.9865, Train Accuracy: 20.26%, Val Loss: 1.9839, Val
Accuracy: 19.68%
Epoch [4/500], Train Loss: 1.9866, Train Accuracy: 20.49%, Val Loss: 1.9805, Val
Accuracy: 19.68%
Epoch [5/500], Train Loss: 1.9869, Train Accuracy: 20.18%, Val Loss: 1.9914, Val
Accuracy: 19.68%
Epoch [6/500], Train Loss: 1.9872, Train Accuracy: 21.22%, Val Loss: 1.9841, Val
Accuracy: 19.68%
Epoch [7/500], Train Loss: 1.9871, Train Accuracy: 20.78%, Val Loss: 1.9818, Val
Accuracy: 19.68%
Epoch [8/500], Train Loss: 1.9863, Train Accuracy: 20.41%, Val Loss: 1.9863, Val
Accuracy: 19.68%
Epoch [9/500], Train Loss: 1.9870, Train Accuracy: 20.43%, Val Loss: 1.9801, Val
Accuracy: 19.68%
Epoch [10/500], Train Loss: 1.9872, Train Accuracy: 21.22%, Val Loss: 1.9798,
Val Accuracy: 19.68%
Epoch [11/500], Train Loss: 1.9870, Train Accuracy: 20.47%, Val Loss: 1.9798,
Val Accuracy: 19.68%
Epoch [12/500], Train Loss: 1.9866, Train Accuracy: 20.74%, Val Loss: 1.9816,
Val Accuracy: 19.68%
Epoch [13/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9808,
Val Accuracy: 19.68%
Epoch [14/500], Train Loss: 1.9862, Train Accuracy: 20.89%, Val Loss: 1.9823,
Val Accuracy: 19.68%
Epoch [15/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9793,
Val Accuracy: 19.68%
Epoch [16/500], Train Loss: 1.9863, Train Accuracy: 20.83%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [17/500], Train Loss: 1.9864, Train Accuracy: 21.22%, Val Loss: 1.9824,
```

```
Val Accuracy: 19.68%
Epoch [18/500], Train Loss: 1.9866, Train Accuracy: 20.74%, Val Loss: 1.9842,
Val Accuracy: 19.68%
Epoch [19/500], Train Loss: 1.9860, Train Accuracy: 21.22%, Val Loss: 1.9812,
Val Accuracy: 19.68%
Epoch [20/500], Train Loss: 1.9865, Train Accuracy: 21.22%, Val Loss: 1.9789,
Val Accuracy: 19.68%
Epoch [21/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9800,
Val Accuracy: 19.68%
Epoch [22/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9818,
Val Accuracy: 19.68%
Epoch [23/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [24/500], Train Loss: 1.9861, Train Accuracy: 20.66%, Val Loss: 1.9811,
Val Accuracy: 19.68%
Epoch [25/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9809,
Val Accuracy: 19.68%
Epoch [26/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9806,
Val Accuracy: 19.68%
Epoch [27/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9807,
Val Accuracy: 19.68%
Epoch [28/500], Train Loss: 1.9862, Train Accuracy: 21.22%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [29/500], Train Loss: 1.9858, Train Accuracy: 20.93%, Val Loss: 1.9822,
Val Accuracy: 19.68%
Epoch [30/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9802,
Val Accuracy: 19.68%
Epoch [31/500], Train Loss: 1.9859, Train Accuracy: 20.74%, Val Loss: 1.9803,
Val Accuracy: 19.68%
Epoch [32/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9812,
Val Accuracy: 19.68%
Epoch [33/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9801,
Val Accuracy: 19.68%
Epoch [34/500], Train Loss: 1.9859, Train Accuracy: 21.04%, Val Loss: 1.9814,
Val Accuracy: 19.68%
Epoch [35/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9799,
Val Accuracy: 19.68%
Epoch [36/500], Train Loss: 1.9864, Train Accuracy: 21.22%, Val Loss: 1.9809,
Val Accuracy: 19.68%
Epoch [37/500], Train Loss: 1.9854, Train Accuracy: 21.22%, Val Loss: 1.9812,
Val Accuracy: 19.68%
Epoch [38/500], Train Loss: 1.9865, Train Accuracy: 21.22%, Val Loss: 1.9803,
Val Accuracy: 19.68%
Epoch [39/500], Train Loss: 1.9860, Train Accuracy: 21.22%, Val Loss: 1.9813,
Val Accuracy: 19.68%
Epoch [40/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9792,
Val Accuracy: 19.68%
Epoch [41/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9796,
```

```
Val Accuracy: 19.68%
Epoch [42/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9813,
Val Accuracy: 19.68%
Epoch [43/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9788,
Val Accuracy: 19.68%
Epoch [44/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9784,
Val Accuracy: 19.68%
Epoch [45/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9802,
Val Accuracy: 19.68%
Epoch [46/500], Train Loss: 1.9855, Train Accuracy: 20.99%, Val Loss: 1.9831,
Val Accuracy: 19.68%
Epoch [47/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9793,
Val Accuracy: 19.68%
Epoch [48/500], Train Loss: 1.9862, Train Accuracy: 21.22%, Val Loss: 1.9805,
Val Accuracy: 19.68%
Epoch [49/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9810,
Val Accuracy: 19.68%
Epoch [50/500], Train Loss: 1.9856, Train Accuracy: 21.22%, Val Loss: 1.9822,
Val Accuracy: 19.68%
Epoch [51/500], Train Loss: 1.9863, Train Accuracy: 20.41%, Val Loss: 1.9824,
Val Accuracy: 19.68%
Epoch [52/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9808,
Val Accuracy: 19.68%
Epoch [53/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9810,
Val Accuracy: 19.68%
Epoch [54/500], Train Loss: 1.9858, Train Accuracy: 21.22%, Val Loss: 1.9823,
Val Accuracy: 19.68%
Epoch [55/500], Train Loss: 1.9860, Train Accuracy: 21.22%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [56/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9806,
Val Accuracy: 19.68%
Epoch [57/500], Train Loss: 1.9854, Train Accuracy: 21.22%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [58/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9808,
Val Accuracy: 19.68%
Epoch [59/500], Train Loss: 1.9855, Train Accuracy: 21.22%, Val Loss: 1.9804,
Val Accuracy: 19.68%
Epoch [60/500], Train Loss: 1.9861, Train Accuracy: 20.81%, Val Loss: 1.9814,
Val Accuracy: 19.68%
Epoch [61/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9811,
Val Accuracy: 19.68%
Epoch [62/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9785,
Val Accuracy: 19.68%
Epoch [63/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9807,
Val Accuracy: 19.68%
Epoch [64/500], Train Loss: 1.9863, Train Accuracy: 21.22%, Val Loss: 1.9799,
Val Accuracy: 19.68%
Epoch [65/500], Train Loss: 1.9856, Train Accuracy: 21.22%, Val Loss: 1.9801,
```

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Val Accuracy: 19.68%
Epoch [66/500], Train Loss: 1.9862, Train Accuracy: 21.22%, Val Loss: 1.9796,
Val Accuracy: 19.68%
Epoch [67/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9792,
Val Accuracy: 19.68%
Epoch [68/500], Train Loss: 1.9862, Train Accuracy: 21.22%, Val Loss: 1.9798,
Val Accuracy: 19.68%
Epoch [69/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9805,
Val Accuracy: 19.68%
Epoch [70/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9800,
Val Accuracy: 19.68%
Epoch [71/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9810,
Val Accuracy: 19.68%
Epoch [72/500], Train Loss: 1.9861, Train Accuracy: 21.22%, Val Loss: 1.9798,
Val Accuracy: 19.68%
Epoch [73/500], Train Loss: 1.9859, Train Accuracy: 21.22%, Val Loss: 1.9794,
Val Accuracy: 19.68%
Epoch [74/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9824,
Val Accuracy: 19.68%
Epoch [75/500], Train Loss: 1.9856, Train Accuracy: 21.22%, Val Loss: 1.9790,
Val Accuracy: 19.68%
Epoch [76/500], Train Loss: 1.9860, Train Accuracy: 21.22%, Val Loss: 1.9806,
Val Accuracy: 19.68%
Epoch [77/500], Train Loss: 1.9857, Train Accuracy: 21.22%, Val Loss: 1.9798,
Val Accuracy: 19.68%
Epoch [78/500], Train Loss: 1.9855, Train Accuracy: 21.22%, Val Loss: 1.9622,
Val Accuracy: 19.68%
Epoch [79/500], Train Loss: 1.7463, Train Accuracy: 31.76%, Val Loss: 1.6377,
Val Accuracy: 33.44%
Epoch [80/500], Train Loss: 1.6687, Train Accuracy: 33.43%, Val Loss: 1.6110,
Val Accuracy: 33.03%
Epoch [81/500], Train Loss: 1.6352, Train Accuracy: 33.60%, Val Loss: 1.5926,
Val Accuracy: 33.86%
Epoch [82/500], Train Loss: 1.5945, Train Accuracy: 34.47%, Val Loss: 1.5768,
Val Accuracy: 33.19%
Epoch [83/500], Train Loss: 1.5827, Train Accuracy: 33.95%, Val Loss: 1.5617,
Val Accuracy: 32.86%
Epoch [84/500], Train Loss: 1.5642, Train Accuracy: 34.27%, Val Loss: 1.5296,
Val Accuracy: 32.61%
Epoch [85/500], Train Loss: 1.5746, Train Accuracy: 34.70%, Val Loss: 1.5465,
Val Accuracy: 33.53%
Epoch [86/500], Train Loss: 1.5445, Train Accuracy: 34.68%, Val Loss: 1.4794,
Val Accuracy: 36.11%
Epoch [87/500], Train Loss: 1.5282, Train Accuracy: 34.72%, Val Loss: 1.5358,
Val Accuracy: 34.28%
Epoch [88/500], Train Loss: 1.5011, Train Accuracy: 35.12%, Val Loss: 1.4075,
Val Accuracy: 37.53%
Epoch [89/500], Train Loss: 1.5518, Train Accuracy: 33.72%, Val Loss: 1.5069,
```

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Val Accuracy: 34.11%
Epoch [90/500], Train Loss: 1.5418, Train Accuracy: 34.54%, Val Loss: 1.4738,
Val Accuracy: 35.70%
Epoch [91/500], Train Loss: 1.5157, Train Accuracy: 33.85%, Val Loss: 1.4529,
Val Accuracy: 35.03%
Epoch [92/500], Train Loss: 1.4480, Train Accuracy: 35.91%, Val Loss: 1.4164,
Val Accuracy: 35.28%
Epoch [93/500], Train Loss: 1.4337, Train Accuracy: 35.98%, Val Loss: 1.4173,
Val Accuracy: 37.36%
Epoch [94/500], Train Loss: 1.4161, Train Accuracy: 36.83%, Val Loss: 1.3859,
Val Accuracy: 37.11%
Epoch [95/500], Train Loss: 1.4039, Train Accuracy: 36.77%, Val Loss: 1.4028,
Val Accuracy: 35.36%
Epoch [96/500], Train Loss: 1.4206, Train Accuracy: 35.93%, Val Loss: 1.4185,
Val Accuracy: 36.70%
Epoch [97/500], Train Loss: 1.3975, Train Accuracy: 36.64%, Val Loss: 1.4060,
Val Accuracy: 34.95%
Epoch [98/500], Train Loss: 1.4125, Train Accuracy: 36.44%, Val Loss: 1.3728,
Val Accuracy: 36.61%
Epoch [99/500], Train Loss: 1.3713, Train Accuracy: 37.00%, Val Loss: 1.3596,
Val Accuracy: 37.61%
Epoch [100/500], Train Loss: 1.4222, Train Accuracy: 36.23%, Val Loss: 1.3737,
Val Accuracy: 36.53%
Epoch [101/500], Train Loss: 1.4854, Train Accuracy: 32.93%, Val Loss: 1.4647,
Val Accuracy: 34.61%
Epoch [102/500], Train Loss: 1.4181, Train Accuracy: 36.67%, Val Loss: 1.3667,
Val Accuracy: 37.03%
Epoch [103/500], Train Loss: 1.3588, Train Accuracy: 37.83%, Val Loss: 1.3644,
Val Accuracy: 37.20%
Epoch [104/500], Train Loss: 1.3790, Train Accuracy: 36.58%, Val Loss: 1.3600,
Val Accuracy: 35.86%
Epoch [105/500], Train Loss: 1.3586, Train Accuracy: 37.21%, Val Loss: 1.3208,
Val Accuracy: 38.62%
Epoch [106/500], Train Loss: 1.3837, Train Accuracy: 37.27%, Val Loss: 1.3628,
Val Accuracy: 35.70%
Epoch [107/500], Train Loss: 1.3550, Train Accuracy: 37.67%, Val Loss: 1.3390,
Val Accuracy: 37.53%
Epoch [108/500], Train Loss: 1.3605, Train Accuracy: 38.29%, Val Loss: 1.3160,
Val Accuracy: 37.70%
Epoch [109/500], Train Loss: 1.3382, Train Accuracy: 37.94%, Val Loss: 1.3047,
Val Accuracy: 38.12%
Epoch [110/500], Train Loss: 1.3332, Train Accuracy: 38.25%, Val Loss: 1.3448,
Val Accuracy: 36.45%
Epoch [111/500], Train Loss: 1.3675, Train Accuracy: 38.15%, Val Loss: 1.3568,
Val Accuracy: 41.37%
Epoch [112/500], Train Loss: 1.4121, Train Accuracy: 37.15%, Val Loss: 1.4762,
Val Accuracy: 36.70%
Epoch [113/500], Train Loss: 1.5527, Train Accuracy: 32.01%, Val Loss: 1.5679,
```

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Val Accuracy: 31.78%
Epoch [114/500], Train Loss: 1.5170, Train Accuracy: 32.83%, Val Loss: 1.4795,
Val Accuracy: 30.86%
Epoch [115/500], Train Loss: 1.4138, Train Accuracy: 36.19%, Val Loss: 1.3815,
Val Accuracy: 35.95%
Epoch [116/500], Train Loss: 1.3569, Train Accuracy: 37.85%, Val Loss: 1.3453,
Val Accuracy: 38.62%
Epoch [117/500], Train Loss: 1.3278, Train Accuracy: 40.00%, Val Loss: 1.3156,
Val Accuracy: 39.53%
Epoch [118/500], Train Loss: 1.3183, Train Accuracy: 41.38%, Val Loss: 1.2778,
Val Accuracy: 43.95%
Epoch [119/500], Train Loss: 1.3004, Train Accuracy: 41.61%, Val Loss: 1.3000,
Val Accuracy: 39.62%
Epoch [120/500], Train Loss: 1.2656, Train Accuracy: 42.32%, Val Loss: 1.3350,
Val Accuracy: 38.20%
Epoch [121/500], Train Loss: 1.2403, Train Accuracy: 43.18%, Val Loss: 1.3055,
Val Accuracy: 38.95%
Epoch [122/500], Train Loss: 1.2460, Train Accuracy: 42.40%, Val Loss: 1.2024,
Val Accuracy: 44.45%
Epoch [123/500], Train Loss: 1.2346, Train Accuracy: 43.03%, Val Loss: 1.2206,
Val Accuracy: 42.87%
Epoch [124/500], Train Loss: 1.2093, Train Accuracy: 44.09%, Val Loss: 1.1881,
Val Accuracy: 43.62%
Epoch [125/500], Train Loss: 1.2154, Train Accuracy: 44.20%, Val Loss: 1.1701,
Val Accuracy: 47.54%
Epoch [126/500], Train Loss: 1.1952, Train Accuracy: 45.64%, Val Loss: 1.1838,
Val Accuracy: 46.29%
Epoch [127/500], Train Loss: 1.1983, Train Accuracy: 45.68%, Val Loss: 1.1650,
Val Accuracy: 47.87%
Epoch [128/500], Train Loss: 1.1729, Train Accuracy: 46.49%, Val Loss: 1.1468,
Val Accuracy: 47.96%
Epoch [129/500], Train Loss: 1.1659, Train Accuracy: 47.20%, Val Loss: 1.2856,
Val Accuracy: 39.53%
Epoch [130/500], Train Loss: 1.1864, Train Accuracy: 45.26%, Val Loss: 1.1379,
Val Accuracy: 47.37%
Epoch [131/500], Train Loss: 1.1796, Train Accuracy: 46.16%, Val Loss: 1.1634,
Val Accuracy: 47.12%
Epoch [132/500], Train Loss: 1.1726, Train Accuracy: 46.16%, Val Loss: 1.1094,
Val Accuracy: 49.46%
Epoch [133/500], Train Loss: 1.1416, Train Accuracy: 47.87%, Val Loss: 1.1567,
Val Accuracy: 45.87%
Epoch [134/500], Train Loss: 1.1208, Train Accuracy: 48.94%, Val Loss: 1.1103,
Val Accuracy: 46.12%
Epoch [135/500], Train Loss: 1.2971, Train Accuracy: 43.28%, Val Loss: 1.1911,
Val Accuracy: 44.62%
Epoch [136/500], Train Loss: 1.2399, Train Accuracy: 42.15%, Val Loss: 1.3742,
Val Accuracy: 37.70%
Epoch [137/500], Train Loss: 1.1967, Train Accuracy: 42.97%, Val Loss: 1.0958,
```

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Val Accuracy: 49.04%
Epoch [138/500], Train Loss: 1.1214, Train Accuracy: 48.02%, Val Loss: 1.1071,
Val Accuracy: 48.71%
Epoch [139/500], Train Loss: 1.1192, Train Accuracy: 48.52%, Val Loss: 1.0813,
Val Accuracy: 48.79%
Epoch [140/500], Train Loss: 1.1053, Train Accuracy: 49.19%, Val Loss: 1.0770,
Val Accuracy: 51.38%
Epoch [141/500], Train Loss: 1.1326, Train Accuracy: 47.56%, Val Loss: 1.0743,
Val Accuracy: 50.96%
Epoch [142/500], Train Loss: 1.1249, Train Accuracy: 48.37%, Val Loss: 1.1017,
Val Accuracy: 48.87%
Epoch [143/500], Train Loss: 1.1009, Train Accuracy: 48.96%, Val Loss: 1.0635,
Val Accuracy: 50.79%
Epoch [144/500], Train Loss: 1.1068, Train Accuracy: 49.44%, Val Loss: 1.0764,
Val Accuracy: 50.96%
Epoch [145/500], Train Loss: 1.0996, Train Accuracy: 48.89%, Val Loss: 1.0728,
Val Accuracy: 48.21%
Epoch [146/500], Train Loss: 1.1138, Train Accuracy: 47.56%, Val Loss: 1.0639,
Val Accuracy: 49.87%
Epoch [147/500], Train Loss: 1.0993, Train Accuracy: 49.75%, Val Loss: 1.0833,
Val Accuracy: 50.46%
Epoch [148/500], Train Loss: 1.0832, Train Accuracy: 50.56%, Val Loss: 1.0796,
Val Accuracy: 49.96%
Epoch [149/500], Train Loss: 1.0850, Train Accuracy: 50.38%, Val Loss: 1.0631,
Val Accuracy: 50.38%
Epoch [150/500], Train Loss: 1.0934, Train Accuracy: 49.44%, Val Loss: 1.0451,
Val Accuracy: 50.54%
Epoch [151/500], Train Loss: 1.0888, Train Accuracy: 50.63%, Val Loss: 1.0581,
Val Accuracy: 50.96%
Epoch [152/500], Train Loss: 1.0764, Train Accuracy: 50.04%, Val Loss: 1.0676,
Val Accuracy: 51.38%
Epoch [153/500], Train Loss: 1.0890, Train Accuracy: 49.90%, Val Loss: 1.0961,
Val Accuracy: 48.21%
Epoch [154/500], Train Loss: 1.0799, Train Accuracy: 49.54%, Val Loss: 1.0441,
Val Accuracy: 50.96%
Epoch [155/500], Train Loss: 1.0702, Train Accuracy: 50.83%, Val Loss: 1.1207,
Val Accuracy: 48.54%
Epoch [156/500], Train Loss: 1.0925, Train Accuracy: 49.48%, Val Loss: 1.0486,
Val Accuracy: 52.79%
Epoch [157/500], Train Loss: 1.0621, Train Accuracy: 51.21%, Val Loss: 1.0674,
Val Accuracy: 50.13%
Epoch [158/500], Train Loss: 1.0778, Train Accuracy: 50.42%, Val Loss: 1.0752,
Val Accuracy: 51.29%
Epoch [159/500], Train Loss: 1.1207, Train Accuracy: 47.77%, Val Loss: 1.1817,
Val Accuracy: 43.29%
Epoch [160/500], Train Loss: 1.6539, Train Accuracy: 36.21%, Val Loss: 1.7214,
Val Accuracy: 29.69%
Epoch [161/500], Train Loss: 1.2507, Train Accuracy: 42.53%, Val Loss: 1.1017,
```

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Val Accuracy: 47.87%
Epoch [162/500], Train Loss: 1.0830, Train Accuracy: 49.48%, Val Loss: 1.0459,
Val Accuracy: 52.79%
Epoch [163/500], Train Loss: 1.0637, Train Accuracy: 52.46%, Val Loss: 1.1554,
Val Accuracy: 43.95%
Epoch [164/500], Train Loss: 1.0842, Train Accuracy: 49.85%, Val Loss: 1.0880,
Val Accuracy: 47.04%
Epoch [165/500], Train Loss: 1.0768, Train Accuracy: 50.17%, Val Loss: 1.1009,
Val Accuracy: 46.96%
Epoch [166/500], Train Loss: 1.0585, Train Accuracy: 51.25%, Val Loss: 1.0658,
Val Accuracy: 50.21%
Epoch [167/500], Train Loss: 1.0521, Train Accuracy: 50.27%, Val Loss: 1.0232,
Val Accuracy: 51.46%
Epoch [168/500], Train Loss: 1.0522, Train Accuracy: 50.50%, Val Loss: 1.0267,
Val Accuracy: 51.54%
Epoch [169/500], Train Loss: 1.0274, Train Accuracy: 52.23%, Val Loss: 1.0062,
Val Accuracy: 52.38%
Epoch [170/500], Train Loss: 1.0245, Train Accuracy: 53.40%, Val Loss: 1.0285,
Val Accuracy: 55.30%
Epoch [171/500], Train Loss: 1.0625, Train Accuracy: 50.94%, Val Loss: 1.0376,
Val Accuracy: 52.13%
Epoch [172/500], Train Loss: 1.0380, Train Accuracy: 52.02%, Val Loss: 1.0150,
Val Accuracy: 52.29%
Epoch [173/500], Train Loss: 1.0312, Train Accuracy: 52.27%, Val Loss: 1.0281,
Val Accuracy: 54.21%
Epoch [174/500], Train Loss: 1.1638, Train Accuracy: 44.20%, Val Loss: 1.0789,
Val Accuracy: 48.12%
Epoch [175/500], Train Loss: 1.0490, Train Accuracy: 49.96%, Val Loss: 1.0133,
Val Accuracy: 52.04%
Epoch [176/500], Train Loss: 1.0272, Train Accuracy: 51.82%, Val Loss: 0.9878,
Val Accuracy: 54.30%
Epoch [177/500], Train Loss: 1.0334, Train Accuracy: 52.02%, Val Loss: 1.0008,
Val Accuracy: 50.79%
Epoch [178/500], Train Loss: 1.0165, Train Accuracy: 52.94%, Val Loss: 0.9976,
Val Accuracy: 53.88%
Epoch [179/500], Train Loss: 1.0190, Train Accuracy: 53.09%, Val Loss: 1.0083,
Val Accuracy: 55.05%
Epoch [180/500], Train Loss: 1.0108, Train Accuracy: 54.05%, Val Loss: 0.9753,
Val Accuracy: 55.55%
Epoch [181/500], Train Loss: 1.0131, Train Accuracy: 54.01%, Val Loss: 0.9879,
Val Accuracy: 54.63%
Epoch [182/500], Train Loss: 0.9981, Train Accuracy: 54.26%, Val Loss: 1.0577,
Val Accuracy: 50.46%
Epoch [183/500], Train Loss: 0.9937, Train Accuracy: 55.09%, Val Loss: 1.0229,
Val Accuracy: 52.54%
Epoch [184/500], Train Loss: 0.9868, Train Accuracy: 54.67%, Val Loss: 1.0110,
Val Accuracy: 53.71%
Epoch [185/500], Train Loss: 0.9932, Train Accuracy: 54.51%, Val Loss: 0.9795,
```

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Val Accuracy: 55.63%
Epoch [186/500], Train Loss: 0.9935, Train Accuracy: 55.18%, Val Loss: 1.0516,
Val Accuracy: 50.88%
Epoch [187/500], Train Loss: 1.0177, Train Accuracy: 52.32%, Val Loss: 1.0181,
Val Accuracy: 51.29%
Epoch [188/500], Train Loss: 0.9912, Train Accuracy: 53.80%, Val Loss: 0.9586,
Val Accuracy: 56.63%
Epoch [189/500], Train Loss: 0.9793, Train Accuracy: 55.63%, Val Loss: 0.9543,
Val Accuracy: 58.13%
Epoch [190/500], Train Loss: 0.9651, Train Accuracy: 56.01%, Val Loss: 0.9639,
Val Accuracy: 56.30%
Epoch [191/500], Train Loss: 0.9810, Train Accuracy: 55.63%, Val Loss: 0.9989,
Val Accuracy: 53.29%
Epoch [192/500], Train Loss: 0.9692, Train Accuracy: 56.11%, Val Loss: 1.0806,
Val Accuracy: 48.62%
Epoch [193/500], Train Loss: 0.9596, Train Accuracy: 56.89%, Val Loss: 0.9514,
Val Accuracy: 56.96%
Epoch [194/500], Train Loss: 0.9529, Train Accuracy: 56.80%, Val Loss: 1.0008,
Val Accuracy: 53.96%
Epoch [195/500], Train Loss: 0.9543, Train Accuracy: 56.59%, Val Loss: 0.9649,
Val Accuracy: 54.88%
Epoch [196/500], Train Loss: 0.9682, Train Accuracy: 55.68%, Val Loss: 1.0607,
Val Accuracy: 48.79%
Epoch [197/500], Train Loss: 0.9720, Train Accuracy: 55.70%, Val Loss: 1.0021,
Val Accuracy: 52.21%
Epoch [198/500], Train Loss: 0.9635, Train Accuracy: 56.14%, Val Loss: 0.9480,
Val Accuracy: 57.46%
Epoch [199/500], Train Loss: 0.9626, Train Accuracy: 56.18%, Val Loss: 0.9906,
Val Accuracy: 53.54%
Epoch [200/500], Train Loss: 0.9363, Train Accuracy: 57.10%, Val Loss: 1.0096,
Val Accuracy: 52.54%
Epoch [201/500], Train Loss: 0.9403, Train Accuracy: 57.78%, Val Loss: 0.9358,
Val Accuracy: 56.55%
Epoch [202/500], Train Loss: 0.9295, Train Accuracy: 57.16%, Val Loss: 0.9470,
Val Accuracy: 57.71%
Epoch [203/500], Train Loss: 0.9575, Train Accuracy: 56.26%, Val Loss: 0.9868,
Val Accuracy: 52.88%
Epoch [204/500], Train Loss: 0.9286, Train Accuracy: 59.37%, Val Loss: 0.9021,
Val Accuracy: 61.13%
Epoch [205/500], Train Loss: 0.9252, Train Accuracy: 58.97%, Val Loss: 0.9082,
Val Accuracy: 59.47%
Epoch [206/500], Train Loss: 0.9053, Train Accuracy: 59.37%, Val Loss: 0.9270,
Val Accuracy: 56.80%
Epoch [207/500], Train Loss: 0.9312, Train Accuracy: 57.95%, Val Loss: 0.9548,
Val Accuracy: 58.22%
Epoch [208/500], Train Loss: 0.9156, Train Accuracy: 59.56%, Val Loss: 0.9421,
Val Accuracy: 55.63%
Epoch [209/500], Train Loss: 0.9251, Train Accuracy: 58.18%, Val Loss: 0.8989,
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Val Accuracy: 61.47%
Epoch [210/500], Train Loss: 0.9407, Train Accuracy: 57.60%, Val Loss: 0.8882,
Val Accuracy: 61.88%
Epoch [211/500], Train Loss: 0.9086, Train Accuracy: 59.06%, Val Loss: 0.8920,
Val Accuracy: 61.88%
Epoch [212/500], Train Loss: 0.9201, Train Accuracy: 58.26%, Val Loss: 1.0015,
Val Accuracy: 53.79%
Epoch [213/500], Train Loss: 0.9905, Train Accuracy: 55.78%, Val Loss: 1.1918,
Val Accuracy: 45.20%
Epoch [214/500], Train Loss: 0.9915, Train Accuracy: 55.47%, Val Loss: 0.9512,
Val Accuracy: 55.96%
Epoch [215/500], Train Loss: 0.9587, Train Accuracy: 55.63%, Val Loss: 0.9534,
Val Accuracy: 53.88%
Epoch [216/500], Train Loss: 0.9221, Train Accuracy: 58.16%, Val Loss: 0.8897,
Val Accuracy: 59.38%
Epoch [217/500], Train Loss: 0.9072, Train Accuracy: 59.27%, Val Loss: 0.9039,
Val Accuracy: 58.38%
Epoch [218/500], Train Loss: 0.8955, Train Accuracy: 60.08%, Val Loss: 0.8991,
Val Accuracy: 59.22%
Epoch [219/500], Train Loss: 0.9160, Train Accuracy: 58.51%, Val Loss: 0.9505,
Val Accuracy: 54.71%
Epoch [220/500], Train Loss: 0.8963, Train Accuracy: 59.89%, Val Loss: 0.9640,
Val Accuracy: 56.21%
Epoch [221/500], Train Loss: 0.8866, Train Accuracy: 59.79%, Val Loss: 0.9327,
Val Accuracy: 56.88%
Epoch [222/500], Train Loss: 0.8952, Train Accuracy: 60.29%, Val Loss: 0.8967,
Val Accuracy: 60.13%
Epoch [223/500], Train Loss: 0.8713, Train Accuracy: 61.29%, Val Loss: 0.8487,
Val Accuracy: 62.55%
Epoch [224/500], Train Loss: 0.8713, Train Accuracy: 60.96%, Val Loss: 0.9247,
Val Accuracy: 55.88%
Epoch [225/500], Train Loss: 0.8830, Train Accuracy: 59.77%, Val Loss: 0.9106,
Val Accuracy: 58.22%
Epoch [226/500], Train Loss: 0.8609, Train Accuracy: 61.06%, Val Loss: 0.8654,
Val Accuracy: 59.05%
Epoch [227/500], Train Loss: 0.9105, Train Accuracy: 58.74%, Val Loss: 0.9885,
Val Accuracy: 50.46%
Epoch [228/500], Train Loss: 0.8756, Train Accuracy: 60.50%, Val Loss: 0.8977,
Val Accuracy: 60.05%
Epoch [229/500], Train Loss: 0.8516, Train Accuracy: 61.71%, Val Loss: 0.8925,
Val Accuracy: 59.47%
Epoch [230/500], Train Loss: 0.8670, Train Accuracy: 61.04%, Val Loss: 0.8553,
Val Accuracy: 60.72%
Epoch [231/500], Train Loss: 0.8760, Train Accuracy: 60.52%, Val Loss: 0.9140,
Val Accuracy: 58.80%
Epoch [232/500], Train Loss: 0.8675, Train Accuracy: 61.52%, Val Loss: 0.8792,
Val Accuracy: 59.88%
Epoch [233/500], Train Loss: 0.8467, Train Accuracy: 62.15%, Val Loss: 0.9109,
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Val Accuracy: 59.72%
Epoch [234/500], Train Loss: 0.8441, Train Accuracy: 62.02%, Val Loss: 0.8633,
Val Accuracy: 61.63%
Epoch [235/500], Train Loss: 0.8546, Train Accuracy: 61.52%, Val Loss: 0.8578,
Val Accuracy: 60.72%
Epoch [236/500], Train Loss: 0.8568, Train Accuracy: 61.54%, Val Loss: 0.8797,
Val Accuracy: 61.47%
Epoch [237/500], Train Loss: 0.8658, Train Accuracy: 61.60%, Val Loss: 0.8899,
Val Accuracy: 58.47%
Epoch [238/500], Train Loss: 0.8405, Train Accuracy: 62.33%, Val Loss: 0.8387,
Val Accuracy: 64.05%
Epoch [239/500], Train Loss: 0.8277, Train Accuracy: 62.75%, Val Loss: 0.8606,
Val Accuracy: 60.38%
Epoch [240/500], Train Loss: 0.8202, Train Accuracy: 63.50%, Val Loss: 0.8547,
Val Accuracy: 59.88%
Epoch [241/500], Train Loss: 0.8242, Train Accuracy: 63.88%, Val Loss: 0.9080,
Val Accuracy: 57.80%
Epoch [242/500], Train Loss: 0.8495, Train Accuracy: 62.79%, Val Loss: 0.8669,
Val Accuracy: 59.47%
Epoch [243/500], Train Loss: 0.8250, Train Accuracy: 62.94%, Val Loss: 1.0188,
Val Accuracy: 53.71%
Epoch [244/500], Train Loss: 0.8342, Train Accuracy: 61.92%, Val Loss: 0.8609,
Val Accuracy: 61.80%
Epoch [245/500], Train Loss: 0.8254, Train Accuracy: 63.02%, Val Loss: 0.8371,
Val Accuracy: 60.88%
Epoch [246/500], Train Loss: 0.8736, Train Accuracy: 61.06%, Val Loss: 0.9961,
Val Accuracy: 52.04%
Epoch [247/500], Train Loss: 0.8468, Train Accuracy: 62.25%, Val Loss: 0.8441,
Val Accuracy: 62.22%
Epoch [248/500], Train Loss: 0.8219, Train Accuracy: 63.69%, Val Loss: 0.8728,
Val Accuracy: 57.71%
Epoch [249/500], Train Loss: 0.8096, Train Accuracy: 63.56%, Val Loss: 0.8493,
Val Accuracy: 62.72%
Epoch [250/500], Train Loss: 0.8166, Train Accuracy: 63.75%, Val Loss: 0.8596,
Val Accuracy: 61.22%
Epoch [251/500], Train Loss: 0.8022, Train Accuracy: 64.34%, Val Loss: 0.8922,
Val Accuracy: 57.96%
Epoch [252/500], Train Loss: 0.8135, Train Accuracy: 64.38%, Val Loss: 0.8275,
Val Accuracy: 61.47%
Epoch [253/500], Train Loss: 0.8113, Train Accuracy: 64.07%, Val Loss: 0.9562,
Val Accuracy: 57.38%
Epoch [254/500], Train Loss: 0.8179, Train Accuracy: 64.19%, Val Loss: 0.8446,
Val Accuracy: 62.05%
Epoch [255/500], Train Loss: 0.7935, Train Accuracy: 64.48%, Val Loss: 0.8741,
Val Accuracy: 59.63%
Epoch [256/500], Train Loss: 0.7996, Train Accuracy: 64.61%, Val Loss: 0.9040,
Val Accuracy: 59.72%
Epoch [257/500], Train Loss: 0.8109, Train Accuracy: 64.38%, Val Loss: 0.8739,
```

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Val Accuracy: 59.72%
Epoch [258/500], Train Loss: 0.7945, Train Accuracy: 64.77%, Val Loss: 0.9847,
Val Accuracy: 55.55%
Epoch [259/500], Train Loss: 0.8099, Train Accuracy: 64.34%, Val Loss: 0.8776,
Val Accuracy: 60.63%
Epoch [260/500], Train Loss: 0.7906, Train Accuracy: 64.13%, Val Loss: 0.8831,
Val Accuracy: 59.63%
Epoch [261/500], Train Loss: 0.7964, Train Accuracy: 65.17%, Val Loss: 0.8226,
Val Accuracy: 65.72%
Epoch [262/500], Train Loss: 0.7691, Train Accuracy: 66.67%, Val Loss: 0.8922,
Val Accuracy: 60.30%
Epoch [263/500], Train Loss: 0.8328, Train Accuracy: 62.50%, Val Loss: 0.9188,
Val Accuracy: 59.05%
Epoch [264/500], Train Loss: 0.8263, Train Accuracy: 63.11%, Val Loss: 0.8180,
Val Accuracy: 63.55%
Epoch [265/500], Train Loss: 0.7702, Train Accuracy: 66.59%, Val Loss: 0.8576,
Val Accuracy: 60.13%
Epoch [266/500], Train Loss: 0.7886, Train Accuracy: 64.59%, Val Loss: 0.8476,
Val Accuracy: 63.14%
Epoch [267/500], Train Loss: 0.7558, Train Accuracy: 66.61%, Val Loss: 0.8627,
Val Accuracy: 61.13%
Epoch [268/500], Train Loss: 0.7926, Train Accuracy: 65.23%, Val Loss: 0.8445,
Val Accuracy: 62.80%
Epoch [269/500], Train Loss: 0.7734, Train Accuracy: 66.17%, Val Loss: 0.8412,
Val Accuracy: 62.30%
Epoch [270/500], Train Loss: 0.7432, Train Accuracy: 67.42%, Val Loss: 0.8587,
Val Accuracy: 61.13%
Epoch [271/500], Train Loss: 0.7892, Train Accuracy: 64.61%, Val Loss: 0.8726,
Val Accuracy: 61.30%
Epoch [272/500], Train Loss: 0.7666, Train Accuracy: 66.61%, Val Loss: 0.7842,
Val Accuracy: 64.22%
Epoch [273/500], Train Loss: 1.1479, Train Accuracy: 52.07%, Val Loss: 1.3821,
Val Accuracy: 42.12%
Epoch [274/500], Train Loss: 1.0792, Train Accuracy: 51.63%, Val Loss: 0.9255,
Val Accuracy: 58.80%
Epoch [275/500], Train Loss: 0.8579, Train Accuracy: 62.04%, Val Loss: 0.8505,
Val Accuracy: 61.97%
Epoch [276/500], Train Loss: 0.7940, Train Accuracy: 64.82%, Val Loss: 0.8276,
Val Accuracy: 62.55%
Epoch [277/500], Train Loss: 0.7986, Train Accuracy: 64.98%, Val Loss: 0.8549,
Val Accuracy: 59.80%
Epoch [278/500], Train Loss: 0.7795, Train Accuracy: 65.42%, Val Loss: 0.8235,
Val Accuracy: 62.64%
Epoch [279/500], Train Loss: 0.7617, Train Accuracy: 67.26%, Val Loss: 0.8413,
Val Accuracy: 62.30%
Epoch [280/500], Train Loss: 0.7859, Train Accuracy: 64.80%, Val Loss: 0.7772,
Val Accuracy: 65.47%
Epoch [281/500], Train Loss: 0.7594, Train Accuracy: 66.26%, Val Loss: 0.8467,
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Val Accuracy: 60.97%
Epoch [282/500], Train Loss: 0.7527, Train Accuracy: 66.51%, Val Loss: 0.8176,
Val Accuracy: 63.47%
Epoch [283/500], Train Loss: 0.7687, Train Accuracy: 65.86%, Val Loss: 0.7885,
Val Accuracy: 64.14%
Epoch [284/500], Train Loss: 0.7504, Train Accuracy: 67.49%, Val Loss: 0.7840,
Val Accuracy: 65.47%
Epoch [285/500], Train Loss: 0.7649, Train Accuracy: 66.61%, Val Loss: 0.9189,
Val Accuracy: 58.97%
Epoch [286/500], Train Loss: 0.7602, Train Accuracy: 67.51%, Val Loss: 0.7841,
Val Accuracy: 65.64%
Epoch [287/500], Train Loss: 0.7344, Train Accuracy: 68.66%, Val Loss: 0.7950,
Val Accuracy: 65.64%
Epoch [288/500], Train Loss: 0.7187, Train Accuracy: 69.14%, Val Loss: 0.7697,
Val Accuracy: 66.56%
Epoch [289/500], Train Loss: 0.7739, Train Accuracy: 66.55%, Val Loss: 0.7856,
Val Accuracy: 63.47%
Epoch [290/500], Train Loss: 0.7048, Train Accuracy: 68.89%, Val Loss: 0.7381,
Val Accuracy: 68.31%
Epoch [291/500], Train Loss: 0.7467, Train Accuracy: 67.11%, Val Loss: 0.7934,
Val Accuracy: 64.89%
Epoch [292/500], Train Loss: 0.7041, Train Accuracy: 69.68%, Val Loss: 0.7847,
Val Accuracy: 65.80%
Epoch [293/500], Train Loss: 0.7283, Train Accuracy: 68.24%, Val Loss: 0.7201,
Val Accuracy: 69.72%
Epoch [294/500], Train Loss: 0.7374, Train Accuracy: 67.40%, Val Loss: 0.7483,
Val Accuracy: 67.81%
Epoch [295/500], Train Loss: 0.7217, Train Accuracy: 67.95%, Val Loss: 0.7725,
Val Accuracy: 64.97%
Epoch [296/500], Train Loss: 0.7894, Train Accuracy: 66.26%, Val Loss: 0.8731,
Val Accuracy: 65.30%
Epoch [297/500], Train Loss: 0.7213, Train Accuracy: 68.97%, Val Loss: 0.8376,
Val Accuracy: 61.38%
Epoch [298/500], Train Loss: 0.7150, Train Accuracy: 68.66%, Val Loss: 1.0765,
Val Accuracy: 54.63%
Epoch [299/500], Train Loss: 0.8044, Train Accuracy: 65.42%, Val Loss: 0.7260,
Val Accuracy: 69.39%
Epoch [300/500], Train Loss: 0.7136, Train Accuracy: 68.47%, Val Loss: 0.7786,
Val Accuracy: 66.31%
Epoch [301/500], Train Loss: 0.7328, Train Accuracy: 67.47%, Val Loss: 0.7469,
Val Accuracy: 68.81%
Epoch [302/500], Train Loss: 0.7404, Train Accuracy: 67.47%, Val Loss: 0.7430,
Val Accuracy: 67.47%
Epoch [303/500], Train Loss: 0.8663, Train Accuracy: 62.46%, Val Loss: 0.7916,
Val Accuracy: 64.55%
Epoch [304/500], Train Loss: 0.7327, Train Accuracy: 67.76%, Val Loss: 0.7526,
Val Accuracy: 66.47%
Epoch [305/500], Train Loss: 0.6997, Train Accuracy: 69.32%, Val Loss: 0.7513,
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Val Accuracy: 67.22%
Epoch [306/500], Train Loss: 0.7050, Train Accuracy: 69.24%, Val Loss: 0.7688,
Val Accuracy: 65.30%
Epoch [307/500], Train Loss: 0.7135, Train Accuracy: 68.89%, Val Loss: 0.7801,
Val Accuracy: 65.05%
Epoch [308/500], Train Loss: 0.6813, Train Accuracy: 70.70%, Val Loss: 0.7673,
Val Accuracy: 67.56%
Epoch [309/500], Train Loss: 0.7058, Train Accuracy: 68.72%, Val Loss: 0.7408,
Val Accuracy: 68.14%
Epoch [310/500], Train Loss: 0.6722, Train Accuracy: 70.60%, Val Loss: 0.7052,
Val Accuracy: 69.22%
Epoch [311/500], Train Loss: 0.6921, Train Accuracy: 69.64%, Val Loss: 0.7888,
Val Accuracy: 66.56%
Epoch [312/500], Train Loss: 0.7010, Train Accuracy: 70.05%, Val Loss: 0.7588,
Val Accuracy: 66.97%
Epoch [313/500], Train Loss: 0.6743, Train Accuracy: 70.26%, Val Loss: 0.8839,
Val Accuracy: 62.30%
Epoch [314/500], Train Loss: 0.7110, Train Accuracy: 68.43%, Val Loss: 0.8249,
Val Accuracy: 63.55%
Epoch [315/500], Train Loss: 0.7007, Train Accuracy: 70.62%, Val Loss: 0.7084,
Val Accuracy: 69.72%
Epoch [316/500], Train Loss: 0.6744, Train Accuracy: 70.60%, Val Loss: 0.7760,
Val Accuracy: 66.72%
Epoch [317/500], Train Loss: 0.6712, Train Accuracy: 70.47%, Val Loss: 0.7251,
Val Accuracy: 67.72%
Epoch [318/500], Train Loss: 0.6638, Train Accuracy: 71.26%, Val Loss: 0.7077,
Val Accuracy: 69.31%
Epoch [319/500], Train Loss: 0.6563, Train Accuracy: 70.83%, Val Loss: 0.9178,
Val Accuracy: 59.22%
Epoch [320/500], Train Loss: 0.7152, Train Accuracy: 68.43%, Val Loss: 0.7156,
Val Accuracy: 70.23%
Epoch [321/500], Train Loss: 0.6592, Train Accuracy: 71.60%, Val Loss: 0.7683,
Val Accuracy: 66.31%
Epoch [322/500], Train Loss: 0.6468, Train Accuracy: 71.18%, Val Loss: 0.7289,
Val Accuracy: 69.14%
Epoch [323/500], Train Loss: 0.6538, Train Accuracy: 71.45%, Val Loss: 0.7496,
Val Accuracy: 67.81%
Epoch [324/500], Train Loss: 0.6727, Train Accuracy: 70.81%, Val Loss: 0.7838,
Val Accuracy: 65.22%
Epoch [325/500], Train Loss: 0.6910, Train Accuracy: 69.97%, Val Loss: 0.7212,
Val Accuracy: 68.64%
Epoch [326/500], Train Loss: 0.6895, Train Accuracy: 70.05%, Val Loss: 0.7335,
Val Accuracy: 66.47%
Epoch [327/500], Train Loss: 0.6485, Train Accuracy: 71.91%, Val Loss: 0.7200,
Val Accuracy: 69.31%
Epoch [328/500], Train Loss: 0.6941, Train Accuracy: 70.70%, Val Loss: 0.7033,
Val Accuracy: 69.14%
Epoch [329/500], Train Loss: 0.6566, Train Accuracy: 71.81%, Val Loss: 0.6908,
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Val Accuracy: 69.97%
Epoch [330/500], Train Loss: 0.6603, Train Accuracy: 71.49%, Val Loss: 0.8639,
Val Accuracy: 60.88%
Epoch [331/500], Train Loss: 0.6822, Train Accuracy: 70.74%, Val Loss: 0.7371,
Val Accuracy: 68.31%
Epoch [332/500], Train Loss: 0.6452, Train Accuracy: 71.64%, Val Loss: 0.7174,
Val Accuracy: 68.39%
Epoch [333/500], Train Loss: 0.6196, Train Accuracy: 73.18%, Val Loss: 0.6938,
Val Accuracy: 69.97%
Epoch [334/500], Train Loss: 0.6293, Train Accuracy: 73.04%, Val Loss: 0.8238,
Val Accuracy: 65.14%
Epoch [335/500], Train Loss: 0.6642, Train Accuracy: 71.47%, Val Loss: 0.7659,
Val Accuracy: 66.47%
Epoch [336/500], Train Loss: 0.6246, Train Accuracy: 73.06%, Val Loss: 0.6664,
Val Accuracy: 71.23%
Epoch [337/500], Train Loss: 0.6258, Train Accuracy: 73.25%, Val Loss: 0.7535,
Val Accuracy: 66.06%
Epoch [338/500], Train Loss: 0.6455, Train Accuracy: 72.29%, Val Loss: 0.7420,
Val Accuracy: 68.06%
Epoch [339/500], Train Loss: 0.6435, Train Accuracy: 71.89%, Val Loss: 0.7708,
Val Accuracy: 65.89%
Epoch [340/500], Train Loss: 0.6466, Train Accuracy: 71.31%, Val Loss: 0.7947,
Val Accuracy: 65.72%
Epoch [341/500], Train Loss: 0.6468, Train Accuracy: 72.47%, Val Loss: 0.7673,
Val Accuracy: 66.81%
Epoch [342/500], Train Loss: 0.6544, Train Accuracy: 71.54%, Val Loss: 0.6928,
Val Accuracy: 70.23%
Epoch [343/500], Train Loss: 0.6359, Train Accuracy: 72.54%, Val Loss: 0.6740,
Val Accuracy: 70.56%
Epoch [344/500], Train Loss: 0.6147, Train Accuracy: 74.25%, Val Loss: 0.8389,
Val Accuracy: 67.81%
Epoch [345/500], Train Loss: 0.6490, Train Accuracy: 72.06%, Val Loss: 0.6933,
Val Accuracy: 72.06%
Epoch [346/500], Train Loss: 0.6003, Train Accuracy: 74.00%, Val Loss: 0.6792,
Val Accuracy: 70.31%
Epoch [347/500], Train Loss: 0.6618, Train Accuracy: 71.81%, Val Loss: 0.8645,
Val Accuracy: 65.30%
Epoch [348/500], Train Loss: 0.6313, Train Accuracy: 72.93%, Val Loss: 0.6882,
Val Accuracy: 69.64%
Epoch [349/500], Train Loss: 0.6260, Train Accuracy: 73.04%, Val Loss: 0.6430,
Val Accuracy: 71.81%
Epoch [350/500], Train Loss: 0.6197, Train Accuracy: 73.96%, Val Loss: 0.7667,
Val Accuracy: 69.47%
Epoch [351/500], Train Loss: 0.6978, Train Accuracy: 70.22%, Val Loss: 0.7598,
Val Accuracy: 68.22%
Epoch [352/500], Train Loss: 0.6772, Train Accuracy: 70.58%, Val Loss: 0.6539,
Val Accuracy: 71.89%
Epoch [353/500], Train Loss: 0.6182, Train Accuracy: 73.73%, Val Loss: 0.6676,
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Val Accuracy: 71.23%
Epoch [354/500], Train Loss: 0.6103, Train Accuracy: 73.64%, Val Loss: 0.6842,
Val Accuracy: 70.31%
Epoch [355/500], Train Loss: 0.5933, Train Accuracy: 74.50%, Val Loss: 0.6954,
Val Accuracy: 69.14%
Epoch [356/500], Train Loss: 0.6933, Train Accuracy: 70.45%, Val Loss: 0.7258,
Val Accuracy: 68.22%
Epoch [357/500], Train Loss: 0.6409, Train Accuracy: 72.27%, Val Loss: 0.6673,
Val Accuracy: 71.06%
Epoch [358/500], Train Loss: 0.5968, Train Accuracy: 74.25%, Val Loss: 0.8367,
Val Accuracy: 65.22%
Epoch [359/500], Train Loss: 0.6116, Train Accuracy: 73.66%, Val Loss: 0.7048,
Val Accuracy: 68.64%
Epoch [360/500], Train Loss: 0.5862, Train Accuracy: 74.54%, Val Loss: 0.6675,
Val Accuracy: 72.39%
Epoch [361/500], Train Loss: 0.6664, Train Accuracy: 71.62%, Val Loss: 0.6779,
Val Accuracy: 71.89%
Epoch [362/500], Train Loss: 0.5788, Train Accuracy: 75.06%, Val Loss: 0.6775,
Val Accuracy: 71.14%
Epoch [363/500], Train Loss: 0.6267, Train Accuracy: 73.64%, Val Loss: 0.6974,
Val Accuracy: 71.14%
Epoch [364/500], Train Loss: 0.5903, Train Accuracy: 75.08%, Val Loss: 0.6540,
Val Accuracy: 72.64%
Epoch [365/500], Train Loss: 0.5980, Train Accuracy: 74.73%, Val Loss: 0.7056,
Val Accuracy: 69.64%
Epoch [366/500], Train Loss: 0.5844, Train Accuracy: 74.87%, Val Loss: 0.7948,
Val Accuracy: 67.89%
Epoch [367/500], Train Loss: 0.5908, Train Accuracy: 75.10%, Val Loss: 0.9297,
Val Accuracy: 64.55%
Epoch [368/500], Train Loss: 0.6072, Train Accuracy: 74.27%, Val Loss: 0.6354,
Val Accuracy: 74.15%
Epoch [369/500], Train Loss: 0.5828, Train Accuracy: 74.81%, Val Loss: 0.6629,
Val Accuracy: 71.39%
Epoch [370/500], Train Loss: 0.5893, Train Accuracy: 75.00%, Val Loss: 0.8782,
Val Accuracy: 64.22%
Epoch [371/500], Train Loss: 0.6057, Train Accuracy: 73.94%, Val Loss: 0.7146,
Val Accuracy: 69.81%
Epoch [372/500], Train Loss: 0.5946, Train Accuracy: 74.35%, Val Loss: 0.6940,
Val Accuracy: 69.72%
Epoch [373/500], Train Loss: 0.6203, Train Accuracy: 73.71%, Val Loss: 0.6470,
Val Accuracy: 72.23%
Epoch [374/500], Train Loss: 0.5832, Train Accuracy: 75.52%, Val Loss: 0.6247,
Val Accuracy: 73.89%
Epoch [375/500], Train Loss: 0.5691, Train Accuracy: 75.75%, Val Loss: 0.6942,
Val Accuracy: 70.73%
Epoch [376/500], Train Loss: 0.5822, Train Accuracy: 75.08%, Val Loss: 0.6648,
Val Accuracy: 71.56%
Epoch [377/500], Train Loss: 0.5718, Train Accuracy: 75.58%, Val Loss: 0.6645,
```

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Val Accuracy: 72.31%
Epoch [378/500], Train Loss: 0.5557, Train Accuracy: 76.29%, Val Loss: 0.6311,
Val Accuracy: 72.73%
Epoch [379/500], Train Loss: 0.6174, Train Accuracy: 73.89%, Val Loss: 0.7382,
Val Accuracy: 69.31%
Epoch [380/500], Train Loss: 0.5922, Train Accuracy: 75.19%, Val Loss: 0.6347,
Val Accuracy: 73.23%
Epoch [381/500], Train Loss: 0.5471, Train Accuracy: 76.67%, Val Loss: 0.6345,
Val Accuracy: 73.64%
Epoch [382/500], Train Loss: 0.5747, Train Accuracy: 76.04%, Val Loss: 0.6672,
Val Accuracy: 73.23%
Epoch [383/500], Train Loss: 0.5587, Train Accuracy: 75.50%, Val Loss: 0.6354,
Val Accuracy: 74.06%
Epoch [384/500], Train Loss: 0.5644, Train Accuracy: 76.36%, Val Loss: 0.6297,
Val Accuracy: 73.81%
Epoch [385/500], Train Loss: 0.6309, Train Accuracy: 73.35%, Val Loss: 0.7076,
Val Accuracy: 69.56%
Epoch [386/500], Train Loss: 0.5650, Train Accuracy: 76.13%, Val Loss: 0.7360,
Val Accuracy: 69.72%
Epoch [387/500], Train Loss: 0.5632, Train Accuracy: 75.94%, Val Loss: 0.6721,
Val Accuracy: 71.89%
Epoch [388/500], Train Loss: 0.7346, Train Accuracy: 69.87%, Val Loss: 0.6585,
Val Accuracy: 72.56%
Epoch [389/500], Train Loss: 0.5608, Train Accuracy: 76.52%, Val Loss: 0.6351,
Val Accuracy: 73.31%
Epoch [390/500], Train Loss: 0.5488, Train Accuracy: 76.63%, Val Loss: 0.6081,
Val Accuracy: 73.56%
Epoch [391/500], Train Loss: 0.5402, Train Accuracy: 76.86%, Val Loss: 0.6642,
Val Accuracy: 71.56%
Epoch [392/500], Train Loss: 0.6182, Train Accuracy: 74.33%, Val Loss: 0.8367,
Val Accuracy: 65.30%
Epoch [393/500], Train Loss: 0.6130, Train Accuracy: 73.96%, Val Loss: 0.6524,
Val Accuracy: 71.73%
Epoch [394/500], Train Loss: 0.5361, Train Accuracy: 77.44%, Val Loss: 0.6398,
Val Accuracy: 73.06%
Epoch [395/500], Train Loss: 0.5942, Train Accuracy: 73.79%, Val Loss: 0.7407,
Val Accuracy: 68.31%
Epoch [396/500], Train Loss: 0.5737, Train Accuracy: 74.94%, Val Loss: 0.6790,
Val Accuracy: 70.89%
Epoch [397/500], Train Loss: 0.5741, Train Accuracy: 75.65%, Val Loss: 0.6508,
Val Accuracy: 72.81%
Epoch [398/500], Train Loss: 0.5520, Train Accuracy: 76.79%, Val Loss: 0.9013,
Val Accuracy: 64.05%
Epoch [399/500], Train Loss: 0.5372, Train Accuracy: 76.67%, Val Loss: 0.5907,
Val Accuracy: 74.48%
Epoch [400/500], Train Loss: 0.5267, Train Accuracy: 77.40%, Val Loss: 0.6064,
Val Accuracy: 75.06%
Epoch [401/500], Train Loss: 0.5604, Train Accuracy: 76.61%, Val Loss: 0.6268,
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Val Accuracy: 73.39%
Epoch [402/500], Train Loss: 0.5934, Train Accuracy: 74.56%, Val Loss: 0.6965,
Val Accuracy: 72.14%
Epoch [403/500], Train Loss: 0.6446, Train Accuracy: 73.75%, Val Loss: 0.7626,
Val Accuracy: 69.81%
Epoch [404/500], Train Loss: 0.5556, Train Accuracy: 76.13%, Val Loss: 0.6610,
Val Accuracy: 72.73%
Epoch [405/500], Train Loss: 0.5495, Train Accuracy: 76.73%, Val Loss: 0.6836,
Val Accuracy: 72.39%
Epoch [406/500], Train Loss: 0.5456, Train Accuracy: 77.32%, Val Loss: 0.6039,
Val Accuracy: 75.65%
Epoch [407/500], Train Loss: 0.5600, Train Accuracy: 76.19%, Val Loss: 0.6919,
Val Accuracy: 72.06%
Epoch [408/500], Train Loss: 0.5946, Train Accuracy: 75.48%, Val Loss: 0.6253,
Val Accuracy: 72.64%
Epoch [409/500], Train Loss: 0.5680, Train Accuracy: 75.71%, Val Loss: 0.6630,
Val Accuracy: 72.64%
Epoch [410/500], Train Loss: 0.5481, Train Accuracy: 77.23%, Val Loss: 0.6097,
Val Accuracy: 74.15%
Epoch [411/500], Train Loss: 0.5377, Train Accuracy: 77.05%, Val Loss: 0.6183,
Val Accuracy: 73.73%
Epoch [412/500], Train Loss: 0.5576, Train Accuracy: 75.65%, Val Loss: 0.6766,
Val Accuracy: 71.56%
Epoch [413/500], Train Loss: 0.5170, Train Accuracy: 78.01%, Val Loss: 0.7131,
Val Accuracy: 69.14%
Epoch [414/500], Train Loss: 0.5858, Train Accuracy: 75.75%, Val Loss: 0.6333,
Val Accuracy: 73.23%
Epoch [415/500], Train Loss: 0.5384, Train Accuracy: 77.02%, Val Loss: 0.8677,
Val Accuracy: 65.97%
Epoch [416/500], Train Loss: 0.5813, Train Accuracy: 75.48%, Val Loss: 0.6443,
Val Accuracy: 73.06%
Epoch [417/500], Train Loss: 0.5232, Train Accuracy: 77.80%, Val Loss: 0.6131,
Val Accuracy: 74.23%
Epoch [418/500], Train Loss: 0.4897, Train Accuracy: 79.59%, Val Loss: 0.6444,
Val Accuracy: 73.56%
Epoch [419/500], Train Loss: 0.5351, Train Accuracy: 77.80%, Val Loss: 0.6467,
Val Accuracy: 73.73%
Epoch [420/500], Train Loss: 0.5414, Train Accuracy: 76.77%, Val Loss: 0.6214,
Val Accuracy: 73.56%
Epoch [421/500], Train Loss: 0.5504, Train Accuracy: 76.25%, Val Loss: 0.6143,
Val Accuracy: 75.40%
Epoch [422/500], Train Loss: 0.5391, Train Accuracy: 76.90%, Val Loss: 0.6197,
Val Accuracy: 73.48%
Epoch [423/500], Train Loss: 0.5300, Train Accuracy: 76.94%, Val Loss: 0.6872,
Val Accuracy: 71.98%
Epoch [424/500], Train Loss: 0.5114, Train Accuracy: 78.61%, Val Loss: 0.6611,
Val Accuracy: 73.31%
Epoch [425/500], Train Loss: 0.5257, Train Accuracy: 77.53%, Val Loss: 0.6823,
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Val Accuracy: 72.06%
Epoch [426/500], Train Loss: 0.5136, Train Accuracy: 78.44%, Val Loss: 0.7815,
Val Accuracy: 69.72%
Epoch [427/500], Train Loss: 0.5241, Train Accuracy: 76.79%, Val Loss: 0.5966,
Val Accuracy: 74.98%
Epoch [428/500], Train Loss: 0.5258, Train Accuracy: 78.69%, Val Loss: 1.0236,
Val Accuracy: 62.97%
Epoch [429/500], Train Loss: 0.5551, Train Accuracy: 77.11%, Val Loss: 0.6182,
Val Accuracy: 73.39%
Epoch [430/500], Train Loss: 0.4953, Train Accuracy: 79.09%, Val Loss: 0.6358,
Val Accuracy: 73.89%
Epoch [431/500], Train Loss: 0.5100, Train Accuracy: 78.11%, Val Loss: 0.6860,
Val Accuracy: 73.73%
Epoch [432/500], Train Loss: 0.5424, Train Accuracy: 76.75%, Val Loss: 0.6890,
Val Accuracy: 72.81%
Epoch [433/500], Train Loss: 0.5270, Train Accuracy: 77.65%, Val Loss: 0.7417,
Val Accuracy: 71.98%
Epoch [434/500], Train Loss: 0.5456, Train Accuracy: 76.96%, Val Loss: 0.6157,
Val Accuracy: 75.31%
Epoch [435/500], Train Loss: 0.5149, Train Accuracy: 78.26%, Val Loss: 0.6100,
Val Accuracy: 75.73%
Epoch [436/500], Train Loss: 0.5091, Train Accuracy: 78.46%, Val Loss: 0.6533,
Val Accuracy: 74.73%
Epoch [437/500], Train Loss: 0.5476, Train Accuracy: 76.90%, Val Loss: 0.6613,
Val Accuracy: 72.48%
Epoch [438/500], Train Loss: 0.5512, Train Accuracy: 77.34%, Val Loss: 0.5876,
Val Accuracy: 74.65%
Epoch [439/500], Train Loss: 0.5054, Train Accuracy: 78.92%, Val Loss: 0.6305,
Val Accuracy: 74.31%
Epoch [440/500], Train Loss: 0.5592, Train Accuracy: 75.83%, Val Loss: 0.6695,
Val Accuracy: 73.89%
Epoch [441/500], Train Loss: 0.5425, Train Accuracy: 77.34%, Val Loss: 0.6499,
Val Accuracy: 74.81%
Epoch [442/500], Train Loss: 0.5078, Train Accuracy: 78.21%, Val Loss: 0.6232,
Val Accuracy: 74.48%
Epoch [443/500], Train Loss: 0.4737, Train Accuracy: 79.95%, Val Loss: 0.7843,
Val Accuracy: 69.06%
Epoch [444/500], Train Loss: 0.5121, Train Accuracy: 78.03%, Val Loss: 0.7086,
Val Accuracy: 71.64%
Epoch [445/500], Train Loss: 0.5161, Train Accuracy: 77.92%, Val Loss: 0.7424,
Val Accuracy: 70.64%
Epoch [446/500], Train Loss: 0.5306, Train Accuracy: 78.05%, Val Loss: 0.6287,
Val Accuracy: 75.06%
Epoch [447/500], Train Loss: 0.5119, Train Accuracy: 78.21%, Val Loss: 0.6220,
Val Accuracy: 74.73%
Epoch [448/500], Train Loss: 0.4958, Train Accuracy: 77.94%, Val Loss: 0.6194,
Val Accuracy: 74.98%
Epoch [449/500], Train Loss: 0.5291, Train Accuracy: 78.05%, Val Loss: 0.6658,
```

```
Val Accuracy: 72.48%
Epoch [450/500], Train Loss: 0.4861, Train Accuracy: 79.05%, Val Loss: 0.6089,
Val Accuracy: 75.40%
Epoch [451/500], Train Loss: 0.4851, Train Accuracy: 79.13%, Val Loss: 0.6008,
Val Accuracy: 76.48%
Epoch [452/500], Train Loss: 0.4907, Train Accuracy: 79.17%, Val Loss: 0.6241,
Val Accuracy: 73.64%
Epoch [453/500], Train Loss: 0.4842, Train Accuracy: 79.47%, Val Loss: 0.5929,
Val Accuracy: 76.23%
Epoch [454/500], Train Loss: 0.5381, Train Accuracy: 76.98%, Val Loss: 0.6169,
Val Accuracy: 74.73%
Epoch [455/500], Train Loss: 0.5375, Train Accuracy: 77.44%, Val Loss: 0.5944,
Val Accuracy: 75.31%
Epoch [456/500], Train Loss: 0.5492, Train Accuracy: 76.71%, Val Loss: 0.6234,
Val Accuracy: 74.40%
Epoch [457/500], Train Loss: 0.4726, Train Accuracy: 80.24%, Val Loss: 0.6437,
Val Accuracy: 75.06%
Epoch [458/500], Train Loss: 0.5182, Train Accuracy: 78.44%, Val Loss: 0.6479,
Val Accuracy: 73.23%
Epoch [459/500], Train Loss: 0.4806, Train Accuracy: 79.80%, Val Loss: 0.6688,
Val Accuracy: 72.98%
Epoch [460/500], Train Loss: 0.4939, Train Accuracy: 78.55%, Val Loss: 0.6631,
Val Accuracy: 73.39%
Epoch [461/500], Train Loss: 0.5152, Train Accuracy: 78.82%, Val Loss: 0.6069,
Val Accuracy: 74.65%
Epoch [462/500], Train Loss: 0.4903, Train Accuracy: 79.40%, Val Loss: 0.7342,
Val Accuracy: 71.39%
Epoch [463/500], Train Loss: 0.4963, Train Accuracy: 78.80%, Val Loss: 0.6651,
Val Accuracy: 73.98%
Epoch [464/500], Train Loss: 0.4797, Train Accuracy: 79.78%, Val Loss: 0.6506,
Val Accuracy: 73.14%
Epoch [465/500], Train Loss: 0.4569, Train Accuracy: 79.65%, Val Loss: 0.6286,
Val Accuracy: 75.23%
Epoch [466/500], Train Loss: 0.4697, Train Accuracy: 79.49%, Val Loss: 0.6084,
Val Accuracy: 75.15%
Epoch [467/500], Train Loss: 0.5416, Train Accuracy: 77.23%, Val Loss: 0.6020,
Val Accuracy: 75.06%
Epoch [468/500], Train Loss: 0.4564, Train Accuracy: 80.76%, Val Loss: 0.5543,
Val Accuracy: 77.23%
Epoch [469/500], Train Loss: 0.4934, Train Accuracy: 79.55%, Val Loss: 0.6766,
Val Accuracy: 73.14%
Epoch [470/500], Train Loss: 0.4789, Train Accuracy: 79.26%, Val Loss: 0.5935,
Val Accuracy: 75.40%
Epoch [471/500], Train Loss: 0.5541, Train Accuracy: 76.98%, Val Loss: 0.7019,
Val Accuracy: 71.48%
Epoch [472/500], Train Loss: 0.5056, Train Accuracy: 79.22%, Val Loss: 0.6446,
Val Accuracy: 74.31%
Epoch [473/500], Train Loss: 0.5002, Train Accuracy: 79.03%, Val Loss: 0.5861,
```

```
Val Accuracy: 76.73%
Epoch [474/500], Train Loss: 0.4596, Train Accuracy: 80.07%, Val Loss: 0.5729,
Val Accuracy: 77.56%
Epoch [475/500], Train Loss: 0.4781, Train Accuracy: 80.07%, Val Loss: 0.6030,
Val Accuracy: 76.65%
Epoch [476/500], Train Loss: 0.4757, Train Accuracy: 80.15%, Val Loss: 0.6921,
Val Accuracy: 72.98%
Epoch [477/500], Train Loss: 0.5246, Train Accuracy: 78.03%, Val Loss: 0.8569,
Val Accuracy: 66.89%
Epoch [478/500], Train Loss: 0.4614, Train Accuracy: 80.53%, Val Loss: 0.6196,
Val Accuracy: 74.90%
Epoch [479/500], Train Loss: 0.4992, Train Accuracy: 79.47%, Val Loss: 0.5683,
Val Accuracy: 77.15%
Epoch [480/500], Train Loss: 0.4403, Train Accuracy: 81.30%, Val Loss: 0.5654,
Val Accuracy: 77.31%
Epoch [481/500], Train Loss: 0.4950, Train Accuracy: 78.78%, Val Loss: 0.6424,
Val Accuracy: 73.48%
Epoch [482/500], Train Loss: 0.4591, Train Accuracy: 80.40%, Val Loss: 0.6354,
Val Accuracy: 73.98%
Epoch [483/500], Train Loss: 0.4601, Train Accuracy: 80.24%, Val Loss: 0.6192,
Val Accuracy: 75.81%
Epoch [484/500], Train Loss: 0.4618, Train Accuracy: 80.66%, Val Loss: 0.5780,
Val Accuracy: 76.06%
Epoch [485/500], Train Loss: 0.5092, Train Accuracy: 79.05%, Val Loss: 0.5649,
Val Accuracy: 77.15%
Epoch [486/500], Train Loss: 0.4563, Train Accuracy: 80.97%, Val Loss: 0.6190,
Val Accuracy: 76.06%
Epoch [487/500], Train Loss: 0.5153, Train Accuracy: 78.99%, Val Loss: 0.5630,
Val Accuracy: 77.40%
Epoch [488/500], Train Loss: 0.4470, Train Accuracy: 80.43%, Val Loss: 0.6408,
Val Accuracy: 74.65%
Epoch [489/500], Train Loss: 0.4867, Train Accuracy: 79.15%, Val Loss: 0.5811,
Val Accuracy: 76.56%
Epoch [490/500], Train Loss: 0.4251, Train Accuracy: 81.49%, Val Loss: 0.6002,
Val Accuracy: 75.98%
Epoch [491/500], Train Loss: 0.4890, Train Accuracy: 79.49%, Val Loss: 0.5794,
Val Accuracy: 77.56%
Epoch [492/500], Train Loss: 0.4450, Train Accuracy: 81.57%, Val Loss: 0.6065,
Val Accuracy: 75.90%
Epoch [493/500], Train Loss: 0.4481, Train Accuracy: 81.28%, Val Loss: 0.5946,
Val Accuracy: 76.90%
Epoch [494/500], Train Loss: 0.4380, Train Accuracy: 80.97%, Val Loss: 0.7558,
Val Accuracy: 70.23%
Epoch [495/500], Train Loss: 0.4837, Train Accuracy: 79.65%, Val Loss: 0.6278,
Val Accuracy: 76.73%
Epoch [496/500], Train Loss: 0.4758, Train Accuracy: 80.03%, Val Loss: 0.5978,
Val Accuracy: 77.06%
Epoch [497/500], Train Loss: 0.4237, Train Accuracy: 82.66%, Val Loss: 0.5859,
```

```
Epoch [498/500], Train Loss: 0.4694, Train Accuracy: 79.63%, Val Loss: 0.7873,
     Val Accuracy: 71.39%
     Epoch [499/500], Train Loss: 0.4573, Train Accuracy: 80.70%, Val Loss: 0.5953,
     Val Accuracy: 76.06%
     Epoch [500/500], Train Loss: 0.4842, Train Accuracy: 79.99%, Val Loss: 0.6235,
     Val Accuracy: 76.06%
[44]: num epochs = 100
      for epoch in range(num_epochs):
          # Training Phase
          model.train()
          running_train_loss = 0.0
          correct_train = 0
          total_train = 0
          for batch_x, batch_y in train_loader:
              optimizer.zero_grad()
              outputs = model(batch_x)
              loss = criterion(outputs, batch_y)
              loss.backward()
              optimizer.step()
              running_train_loss += loss.item()
              # Calculate training accuracy
              _, predicted = torch.max(outputs, 1)
              total_train += batch_y.size(0)
              correct_train += (predicted == batch_y).sum().item()
          avg_train_loss = running_train_loss / len(train_loader)
          train_accuracy = 100 * correct_train / total_train
          # Store training loss and accuracy for the current epoch
          train_losses.append(avg_train_loss)
          train_accuracies.append(train_accuracy)
          # Validation Phase
          model.eval() # Put the model in evaluation mode
          running_val_loss = 0.0
          correct_val = 0
          total_val = 0
          with torch.no_grad(): # Disable gradient computation during validation
              for batch_x, batch_y in val_loader:
                  outputs = model(batch_x)
                  loss = criterion(outputs, batch_y)
```

Val Accuracy: 76.90%

```
_, predicted = torch.max(outputs, 1)
            total_val += batch_y.size(0)
            correct_val += (predicted == batch_y).sum().item()
    avg_val_loss = running_val_loss / len(val_loader)
    val_accuracy = 100 * correct_val / total_val
    # Store validation loss and accuracy for the current epoch
    val_losses.append(avg_val_loss)
    val_accuracies.append(val_accuracy)
    print(f'Epoch [{epoch + 1}/{num_epochs}], '
          f'Train Loss: {avg_train_loss:.4f}, Train Accuracy: {train_accuracy:.
  ⇒2f}%, '
          f'Val Loss: {avg_val_loss:.4f}, Val Accuracy: {val_accuracy:.2f}%')
# After training, you can access `train_losses`, `val_losses`, _u
 → `train_accuracies`, and `val_accuracies` arrays.
Epoch [1/100], Train Loss: 0.4467, Train Accuracy: 81.43%, Val Loss: 0.4959, Val
Accuracy: 80.48%
Epoch [2/100], Train Loss: 0.4251, Train Accuracy: 83.14%, Val Loss: 0.5311, Val
Accuracy: 78.07%
Epoch [3/100], Train Loss: 0.4671, Train Accuracy: 79.59%, Val Loss: 0.5473, Val
Accuracy: 78.23%
Epoch [4/100], Train Loss: 0.4474, Train Accuracy: 81.22%, Val Loss: 0.5438, Val
Accuracy: 78.73%
Epoch [5/100], Train Loss: 0.4168, Train Accuracy: 83.16%, Val Loss: 0.6009, Val
Accuracy: 76.15%
Epoch [6/100], Train Loss: 0.5643, Train Accuracy: 76.57%, Val Loss: 0.5946, Val
Accuracy: 76.06%
Epoch [7/100], Train Loss: 0.4499, Train Accuracy: 81.41%, Val Loss: 0.5126, Val
Accuracy: 79.48%
Epoch [8/100], Train Loss: 0.4158, Train Accuracy: 82.35%, Val Loss: 0.5320, Val
Accuracy: 78.32%
Epoch [9/100], Train Loss: 0.4002, Train Accuracy: 83.39%, Val Loss: 0.6825, Val
Accuracy: 73.98%
Epoch [10/100], Train Loss: 0.4391, Train Accuracy: 82.03%, Val Loss: 0.6029,
Val Accuracy: 75.65%
Epoch [11/100], Train Loss: 0.4798, Train Accuracy: 80.93%, Val Loss: 0.6245,
Val Accuracy: 74.40%
Epoch [12/100], Train Loss: 0.4813, Train Accuracy: 80.88%, Val Loss: 0.5475,
Val Accuracy: 77.65%
Epoch [13/100], Train Loss: 0.4163, Train Accuracy: 82.99%, Val Loss: 0.5741,
```

running\_val\_loss += loss.item()

# Calculate validation accuracy

```
Val Accuracy: 77.81%
Epoch [14/100], Train Loss: 0.4259, Train Accuracy: 82.43%, Val Loss: 0.7940,
Val Accuracy: 70.31%
Epoch [15/100], Train Loss: 0.4589, Train Accuracy: 81.01%, Val Loss: 0.5647,
Val Accuracy: 78.15%
Epoch [16/100], Train Loss: 0.4717, Train Accuracy: 80.61%, Val Loss: 0.5468,
Val Accuracy: 76.06%
Epoch [17/100], Train Loss: 0.4457, Train Accuracy: 81.39%, Val Loss: 0.5308,
Val Accuracy: 78.15%
Epoch [18/100], Train Loss: 0.4235, Train Accuracy: 82.55%, Val Loss: 0.5385,
Val Accuracy: 78.73%
Epoch [19/100], Train Loss: 0.4021, Train Accuracy: 83.31%, Val Loss: 0.5420,
Val Accuracy: 78.98%
Epoch [20/100], Train Loss: 0.4231, Train Accuracy: 82.39%, Val Loss: 0.5527,
Val Accuracy: 77.48%
Epoch [21/100], Train Loss: 0.5186, Train Accuracy: 78.71%, Val Loss: 0.6123,
Val Accuracy: 74.31%
Epoch [22/100], Train Loss: 0.4220, Train Accuracy: 82.76%, Val Loss: 0.5644,
Val Accuracy: 76.73%
Epoch [23/100], Train Loss: 0.4747, Train Accuracy: 81.16%, Val Loss: 0.6124,
Val Accuracy: 76.81%
Epoch [24/100], Train Loss: 0.4636, Train Accuracy: 80.45%, Val Loss: 0.5538,
Val Accuracy: 76.65%
Epoch [25/100], Train Loss: 0.4267, Train Accuracy: 82.47%, Val Loss: 0.5831,
Val Accuracy: 77.56%
Epoch [26/100], Train Loss: 0.4240, Train Accuracy: 82.35%, Val Loss: 0.5188,
Val Accuracy: 80.23%
Epoch [27/100], Train Loss: 0.4696, Train Accuracy: 81.09%, Val Loss: 0.5998,
Val Accuracy: 76.23%
Epoch [28/100], Train Loss: 0.4173, Train Accuracy: 82.78%, Val Loss: 0.5278,
Val Accuracy: 80.07%
Epoch [29/100], Train Loss: 0.4635, Train Accuracy: 81.39%, Val Loss: 0.4787,
Val Accuracy: 80.73%
Epoch [30/100], Train Loss: 0.4066, Train Accuracy: 83.16%, Val Loss: 0.6295,
Val Accuracy: 76.73%
Epoch [31/100], Train Loss: 0.4990, Train Accuracy: 79.47%, Val Loss: 0.6731,
Val Accuracy: 73.14%
Epoch [32/100], Train Loss: 0.4774, Train Accuracy: 80.86%, Val Loss: 0.6750,
Val Accuracy: 74.73%
Epoch [33/100], Train Loss: 0.4414, Train Accuracy: 82.26%, Val Loss: 0.5364,
Val Accuracy: 79.73%
Epoch [34/100], Train Loss: 0.3973, Train Accuracy: 83.58%, Val Loss: 0.5658,
Val Accuracy: 76.73%
Epoch [35/100], Train Loss: 0.3949, Train Accuracy: 83.85%, Val Loss: 0.5669,
Val Accuracy: 79.65%
Epoch [36/100], Train Loss: 0.5034, Train Accuracy: 79.63%, Val Loss: 0.6399,
Val Accuracy: 75.40%
Epoch [37/100], Train Loss: 0.4502, Train Accuracy: 81.57%, Val Loss: 0.5710,
```

```
Val Accuracy: 76.81%
Epoch [38/100], Train Loss: 0.4041, Train Accuracy: 83.74%, Val Loss: 0.5336,
Val Accuracy: 78.90%
Epoch [39/100], Train Loss: 0.3968, Train Accuracy: 84.31%, Val Loss: 0.4796,
Val Accuracy: 81.23%
Epoch [40/100], Train Loss: 0.4271, Train Accuracy: 82.51%, Val Loss: 0.5041,
Val Accuracy: 80.90%
Epoch [41/100], Train Loss: 0.4948, Train Accuracy: 80.43%, Val Loss: 0.8611,
Val Accuracy: 66.56%
Epoch [42/100], Train Loss: 0.4547, Train Accuracy: 81.30%, Val Loss: 0.5416,
Val Accuracy: 78.65%
Epoch [43/100], Train Loss: 0.4248, Train Accuracy: 83.37%, Val Loss: 0.4784,
Val Accuracy: 81.48%
Epoch [44/100], Train Loss: 0.4039, Train Accuracy: 83.49%, Val Loss: 0.5190,
Val Accuracy: 79.15%
Epoch [45/100], Train Loss: 0.3979, Train Accuracy: 82.99%, Val Loss: 0.5802,
Val Accuracy: 77.81%
Epoch [46/100], Train Loss: 0.3938, Train Accuracy: 83.58%, Val Loss: 0.5514,
Val Accuracy: 78.48%
Epoch [47/100], Train Loss: 0.4495, Train Accuracy: 81.89%, Val Loss: 0.4863,
Val Accuracy: 80.65%
Epoch [48/100], Train Loss: 0.3821, Train Accuracy: 83.81%, Val Loss: 0.5129,
Val Accuracy: 79.98%
Epoch [49/100], Train Loss: 0.4273, Train Accuracy: 82.83%, Val Loss: 0.5269,
Val Accuracy: 80.15%
Epoch [50/100], Train Loss: 0.4334, Train Accuracy: 82.74%, Val Loss: 0.5259,
Val Accuracy: 77.98%
Epoch [51/100], Train Loss: 0.4079, Train Accuracy: 82.74%, Val Loss: 0.4831,
Val Accuracy: 80.82%
Epoch [52/100], Train Loss: 0.3856, Train Accuracy: 84.37%, Val Loss: 0.5591,
Val Accuracy: 77.31%
Epoch [53/100], Train Loss: 0.3937, Train Accuracy: 83.91%, Val Loss: 0.5208,
Val Accuracy: 79.07%
Epoch [54/100], Train Loss: 0.4193, Train Accuracy: 83.22%, Val Loss: 0.7577,
Val Accuracy: 69.64%
Epoch [55/100], Train Loss: 0.5049, Train Accuracy: 79.11%, Val Loss: 0.6967,
Val Accuracy: 73.39%
Epoch [56/100], Train Loss: 0.4420, Train Accuracy: 82.14%, Val Loss: 0.6020,
Val Accuracy: 76.56%
Epoch [57/100], Train Loss: 0.3573, Train Accuracy: 85.73%, Val Loss: 0.5042,
Val Accuracy: 80.40%
Epoch [58/100], Train Loss: 0.4045, Train Accuracy: 83.56%, Val Loss: 0.5368,
Val Accuracy: 78.98%
Epoch [59/100], Train Loss: 0.4363, Train Accuracy: 82.01%, Val Loss: 0.6103,
Val Accuracy: 76.73%
Epoch [60/100], Train Loss: 0.4076, Train Accuracy: 83.51%, Val Loss: 0.5374,
Val Accuracy: 79.07%
Epoch [61/100], Train Loss: 0.4311, Train Accuracy: 82.41%, Val Loss: 0.5272,
```

```
Val Accuracy: 77.98%
Epoch [62/100], Train Loss: 0.3682, Train Accuracy: 84.91%, Val Loss: 0.5111,
Val Accuracy: 80.57%
Epoch [63/100], Train Loss: 0.3781, Train Accuracy: 84.77%, Val Loss: 0.5435,
Val Accuracy: 78.82%
Epoch [64/100], Train Loss: 0.3790, Train Accuracy: 84.85%, Val Loss: 0.5838,
Val Accuracy: 77.31%
Epoch [65/100], Train Loss: 0.4990, Train Accuracy: 81.05%, Val Loss: 0.5316,
Val Accuracy: 79.82%
Epoch [66/100], Train Loss: 0.4062, Train Accuracy: 83.14%, Val Loss: 0.6246,
Val Accuracy: 76.06%
Epoch [67/100], Train Loss: 0.3678, Train Accuracy: 84.75%, Val Loss: 0.5587,
Val Accuracy: 78.32%
Epoch [68/100], Train Loss: 0.4020, Train Accuracy: 83.51%, Val Loss: 0.5104,
Val Accuracy: 80.32%
Epoch [69/100], Train Loss: 0.3635, Train Accuracy: 85.50%, Val Loss: 0.5313,
Val Accuracy: 80.57%
Epoch [70/100], Train Loss: 0.4015, Train Accuracy: 83.10%, Val Loss: 0.5290,
Val Accuracy: 80.15%
Epoch [71/100], Train Loss: 0.3826, Train Accuracy: 84.77%, Val Loss: 0.4989,
Val Accuracy: 80.90%
Epoch [72/100], Train Loss: 0.4545, Train Accuracy: 81.16%, Val Loss: 0.4989,
Val Accuracy: 79.65%
Epoch [73/100], Train Loss: 0.3791, Train Accuracy: 84.22%, Val Loss: 0.6085,
Val Accuracy: 76.23%
Epoch [74/100], Train Loss: 0.4293, Train Accuracy: 82.78%, Val Loss: 0.4943,
Val Accuracy: 81.15%
Epoch [75/100], Train Loss: 0.3663, Train Accuracy: 85.39%, Val Loss: 0.5819,
Val Accuracy: 78.23%
Epoch [76/100], Train Loss: 0.4172, Train Accuracy: 82.89%, Val Loss: 0.6872,
Val Accuracy: 75.73%
Epoch [77/100], Train Loss: 0.3641, Train Accuracy: 85.54%, Val Loss: 0.5067,
Val Accuracy: 81.23%
Epoch [78/100], Train Loss: 0.4046, Train Accuracy: 83.62%, Val Loss: 0.5791,
Val Accuracy: 79.23%
Epoch [79/100], Train Loss: 0.4335, Train Accuracy: 82.22%, Val Loss: 0.5232,
Val Accuracy: 78.48%
Epoch [80/100], Train Loss: 0.3786, Train Accuracy: 84.85%, Val Loss: 0.4540,
Val Accuracy: 82.24%
Epoch [81/100], Train Loss: 0.3516, Train Accuracy: 85.60%, Val Loss: 0.4755,
Val Accuracy: 80.48%
Epoch [82/100], Train Loss: 0.4051, Train Accuracy: 83.64%, Val Loss: 0.5382,
Val Accuracy: 79.15%
Epoch [83/100], Train Loss: 0.3846, Train Accuracy: 84.45%, Val Loss: 0.5346,
Val Accuracy: 79.40%
Epoch [84/100], Train Loss: 0.3834, Train Accuracy: 84.33%, Val Loss: 0.4912,
Val Accuracy: 81.07%
Epoch [85/100], Train Loss: 0.3709, Train Accuracy: 85.29%, Val Loss: 0.5373,
```

```
Val Accuracy: 77.65%
     Epoch [87/100], Train Loss: 0.6969, Train Accuracy: 73.85%, Val Loss: 0.5671,
     Val Accuracy: 78.32%
     Epoch [88/100], Train Loss: 0.3863, Train Accuracy: 84.35%, Val Loss: 0.5296,
     Val Accuracy: 77.65%
     Epoch [89/100], Train Loss: 0.3806, Train Accuracy: 84.66%, Val Loss: 0.5292,
     Val Accuracy: 79.90%
     Epoch [90/100], Train Loss: 0.3922, Train Accuracy: 84.49%, Val Loss: 0.7298,
     Val Accuracy: 74.98%
     Epoch [91/100], Train Loss: 0.4177, Train Accuracy: 82.87%, Val Loss: 0.6423,
     Val Accuracy: 75.40%
     Epoch [92/100], Train Loss: 0.3743, Train Accuracy: 84.49%, Val Loss: 0.5253,
     Val Accuracy: 80.07%
     Epoch [93/100], Train Loss: 0.5212, Train Accuracy: 79.80%, Val Loss: 0.8720,
     Val Accuracy: 69.81%
     Epoch [94/100], Train Loss: 0.4166, Train Accuracy: 83.22%, Val Loss: 0.4653,
     Val Accuracy: 81.65%
     Epoch [95/100], Train Loss: 0.3349, Train Accuracy: 86.31%, Val Loss: 0.4733,
     Val Accuracy: 81.65%
     Epoch [96/100], Train Loss: 0.3648, Train Accuracy: 85.50%, Val Loss: 0.5815,
     Val Accuracy: 77.81%
     Epoch [97/100], Train Loss: 0.3860, Train Accuracy: 83.99%, Val Loss: 0.5376,
     Val Accuracy: 79.23%
     Epoch [98/100], Train Loss: 0.3805, Train Accuracy: 84.70%, Val Loss: 0.9226,
     Val Accuracy: 65.97%
     Epoch [99/100], Train Loss: 0.5072, Train Accuracy: 79.99%, Val Loss: 0.6451,
     Val Accuracy: 76.90%
     Epoch [100/100], Train Loss: 0.3795, Train Accuracy: 84.68%, Val Loss: 0.4827,
     Val Accuracy: 79.73%
[48]: # Assuming you have trained for 600 epochs, but check lengths first
     num_epochs = min(len(train_accuracies), len(val_accuracies), 600)
     epochs = range(1, num_epochs + 1)
     plt.figure(figsize=(10, 5))
      # Plot training accuracy
     plt.plot(epochs, train_accuracies[:num_epochs], label='Training Accuracy', __

color='blue')
      # Plot validation accuracy
     plt.plot(epochs, val_accuracies[:num_epochs], label='Validation Accuracy',__
       ⇔color='orange')
     # Labeling the graph
```

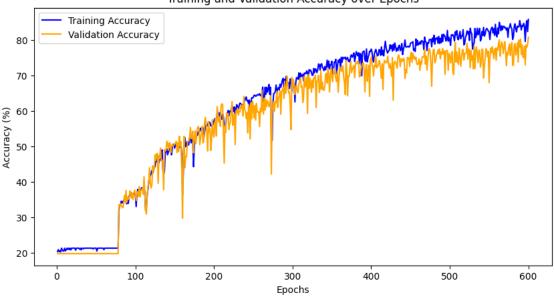
Epoch [86/100], Train Loss: 0.3957, Train Accuracy: 84.16%, Val Loss: 0.6231,

Val Accuracy: 79.65%

```
plt.title('Training and Validation Accuracy over Epochs')
plt.xlabel('Epochs')
plt.ylabel('Accuracy (%)')
plt.legend()

# Display the graph
plt.show()
```

## Training and Validation Accuracy over Epochs



Val Accuracy: 0.7973 precision recall f1-score support

```
1. loud
                        0.80
                                  0.77
                                             0.78
                                                        253
         2. quiet
                        0.91
                                  0.92
                                             0.91
                                                        236
         3. happy
                        0.76
                                  0.78
                                             0.77
                                                        229
           4. sad
                        0.77
                                  0.93
                                             0.84
                                                         95
     5. Beautiful
                        0.75
                                  0.66
                                             0.70
                                                         99
          6. Ugly
                        0.82
                                  0.80
                                             0.81
                                                        119
          7. Deaf
                        0.76
                                  0.66
                                             0.70
                                                         87
         8. Blind
                        0.69
                                  0.75
                                             0.72
                                                         81
                                             0.80
                                                       1199
         accuracy
                        0.78
                                  0.78
                                             0.78
                                                       1199
        macro avg
     weighted avg
                        0.80
                                  0.80
                                             0.80
                                                       1199
[50]: # Save the model
      torch.save(model.state_dict(), 'demo_lstm.pth')
[51]: from tensorflow.keras.models import Sequential
      from tensorflow.keras.layers import LSTM, Dense
      from tensorflow.keras.layers import Dropout
[52]: from keras.models import Sequential
      from keras.layers import LSTM, Dropout, Dense, TimeDistributed
      from keras.optimizers import Adam
[53]: model = Sequential()
      model.add(LSTM(128, return_sequences=True, input_shape=(x_train.shape[1],_
       ⇔x_train.shape[2])))
      model.add(Dropout(0.4))
      model.add(LSTM(128, return_sequences=True))
      model.add(Dropout(0.4))
      model.add(LSTM(64, return_sequences=True))
      model.add(Dropout(0.4))
      model.add(TimeDistributed(Dense(64, activation='relu')))
```

/opt/conda/lib/python3.10/site-packages/keras/src/layers/rnn/rnn.py:204: 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.

model.compile(optimizer=optimizer, loss='sparse categorical crossentropy',

model.add(Dropout(0.4))
model.add(LSTM(32))
model.add(Dropout(0.4))

→metrics=['accuracy'])

model.add(Dense(num\_classes, activation='softmax'))

optimizer = Adam(learning\_rate=0.001)

```
super().__init__(**kwargs)
```

```
[]: # Train the model and store the training history
     history = model.fit(x_train, y_train_encoded,
                         validation_data=(x_val, y_val_encoded),
                         epochs=600, batch_size=32)
     # Extract accuracy and validation accuracy from the history
     train_accuracy = history.history['accuracy']
     val_accuracy = history.history['val_accuracy']
     epochs = range(1, len(train_accuracy) + 1)
     # Plot training and validation accuracy
     plt.figure(figsize=(10, 5))
     plt.plot(epochs, train accuracy, label='Training Accuracy', color='blue')
     plt.plot(epochs, val_accuracy, label='Validation Accuracy', color='orange')
     # Label the graph
     plt.title('Training and Validation Accuracy over Epochs')
     plt.xlabel('Epochs')
     plt.ylabel('Accuracy')
     plt.legend()
     # Show the plot
     plt.show()
[]: # Evaluate the TensorFlow/Keras model on the validation set
     loss, tf_keras_val_accuracy = model.evaluate(x_val, y_val_encoded, verbose=0)
     print(f'TensorFlow/Keras Model Validation Accuracy: {tf_keras_val_accuracy:.

4f}')
     # Generate predictions for the validation set
     y_pred = model.predict(x_val)
     y_pred_classes = np.argmax(y_pred, axis=1)
     # Classification Report for the TensorFlow/Keras model
     print(classification_report(y_val_encoded, y_pred_classes,__
      →target names=label encoder.classes ))
     # Visualization of results
     metrics = {'Model': ['LSTM (PyTorch)', 'LSTM (TensorFlow/Keras)'],
                'Accuracy': [accuracy, tf_keras_val_accuracy]}
     metrics_df = pd.DataFrame(metrics)
     # Plot the results
     plt.figure(figsize=(8, 6))
```

sns.barplot(x='Model', y='Accuracy', data=metrics\_df)

```
plt.title('Model Accuracy Comparison (PyTorch vs TensorFlow)')
plt.ylim(0, 1)
plt.show()

[]: # Save the entire model
model.save('demo2_lstm.h5')
```