effnet_rcnn_genre

March 25, 2025

```
[2]: import torch
     import pandas as pd
     import numpy as np
     from tqdm import tqdm
     import wandb
[3]: wandb.
      oinit(entity="ameyar3103-iiit-hyderabad",project="recurrent_conv_art_effnet",⊔
      ⇔config={
         "epochs": 20,
         "batch_size": 32,
         "learning_rate": 0.001,
         "model": "EFFNET_RCNN"
     })
    wandb: Using wandb-core as the SDK backend. Please refer to
    https://wandb.me/wandb-core for more information.
    wandb: Currently logged in as: ameyar3103
    (ameyar3103-iiit-hyderabad) to https://api.wandb.ai. Use
    `wandb login --relogin` to force relogin
    <IPython.core.display.HTML object>
    <IPython.core.display.HTML object>
    <IPython.core.display.HTML object>
    <IPython.core.display.HTML object>
    <IPython.core.display.HTML object>
[3]: <wandb.sdk.wandb_run.Run at 0x7e8191cc1390>
    0.1 Data loading
[4]: df_train = pd.read_csv('wikiart_csv/genre_train.csv',header=None,_

¬names=["image_path", "genre_id"])
     df_val = pd.read_csv('wikiart_csv/genre_val.csv',header=None,__
      →names=["image_path", "genre_id"])
```

```
[5]: # get the number of classes
   num_classes = 10 # from artist_class.txt

[6]: # Gather input data
   train_images = df_train['image_path'].values
   train_labels = df_train['genre_id'].values

   val_images = df_val['image_path'].values
   val_labels = df_val['genre_id'].values

[7]: from torchvision import transforms
   import cv2
```

0.2 Preprocess data and create test and train dataset

```
[]: def get_image(image_path,image_size=224):
         try:
             img = cv2.imread('./wikiart/' + image_path)
             if img is None:
                 raise ValueError(f"Image not loaded: ./wikiart/{image_path}")
             img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
             h, w, _= img.shape
             scale = 256 / min(h, w)
             new w = int(w * scale)
             new_h = int(h * scale)
             img resized = cv2.resize(img, (new w, new h))
             start_x = (new_w - image_size) // 2
             start_y = (new_h - image_size) // 2
             img_cropped = img_resized[start_y:start_y+image_size, start_x:
      ⇔start_x+image_size]
             img_cropped = img_cropped.astype(np.float32) / 255.0
             img_tensor = torch.from_numpy(img_cropped).permute(2, 0, 1)
             mean = torch.tensor([0.485, 0.456, 0.406]).view(3, 1, 1)
             std = torch.tensor([0.229, 0.224, 0.225]).view(3, 1, 1)
             img_tensor = (img_tensor - mean) / std
             return img_tensor
         except Exception as e:
             print(f"Error processing {image_path}: {e}")
             return torch.zeros(3, image_size, image_size)
     class WikiArtDataset(torch.utils.data.Dataset):
         def __init__(self, images, labels):
             self.images = images
             self.labels = labels
         def __len__(self):
             return len(self.images)
```

```
def __getitem__(self, idx):
        # image_vectors = []
         # for image in self.images:
              image_emb = get_image(image)
               image_vectors.append(image_emb)
        # image = torch.stack(image vectors)
        image = self.images[idx]
         # label should be a one-hot encoded vector
        label = torch.zeros(num classes)
        label[self.labels[idx]] = 1
        return image, label
train_dataset = WikiArtDataset(train_images, train_labels)
train_loader = torch.utils.data.DataLoader(train_dataset, batch_size=32,__
  ⇔shuffle=True)
val_dataset = WikiArtDataset(val_images, val_labels)
val loader = torch.utils.data.DataLoader(val dataset, batch size=32,,,
  ⇔shuffle=False)
for i, (images, labels) in enumerate(train_loader):
    print(images)
    print(labels)
    break
('Impressionism/claude-monet_small-boat-on-the-small-branch-of-the-seine-at-
argenteuil.jpg', 'Art_Nouveau_Modern/raphael-kirchner_lillian-lorraine.jpg',
'Abstract_Expressionism/conrad-marca-relli_untitled-1958.jpg',
'Naive Art_Primitivism/marc-chagall_lovers.jpg', 'Rococo/bernardo-bellotto_the-
kreuzkirche-in-dresden.jpg', 'High_Renaissance/michelangelo_the-prophet-
daniel-1511.jpg', 'High_Renaissance/pietro-perugino_pala-di-monteripido.jpg',
'Symbolism/nicholas-roerich and-we-are-opening-the-gates-1922.jpg',
'Impressionism/claude-monet_vetheuil-afternoon.jpg', 'Romanticism/francisco-
goya_the-repentant-saint-peter.jpg', 'Realism/edgar-degas_portrait-of-rene-de-
gas-1855.jpg', 'Realism/camille-corot_dieppe-end-of-a-pier-and-the-
sea-1822.jpg', 'Romanticism/orest-kiprensky_peasant-boy-1814.jpg',
'Expressionism/richard-gerstl small-street-nu-dorferstra-e-1908.jpg',
'Abstract_Expressionism/rafa-nasiri_untitled-068-2002.jpg',
'Romanticism/gheorghe-tattarescu_laz-r-kalinderu.jpg', 'Realism/ivan-
shishkin_countess-mordvinov-s-forest-1891.jpg', 'Baroque/jan-steen_twelfth-
night-1668.jpg', 'Realism/salvador-dali_cathedral-unfinished.jpg',
'Northern_Renaissance/pieter-bruegel-the-elder_the-fair-at-hoboken-1559.jpg',
'Impressionism/claude-monet_palazzo-contarini-2.jpg',
'Abstract_Expressionism/brice-marden_dragons-2004.jpg',
'Post Impressionism/henri-de-toulouse-lautrec girl-in-a-fur-mademoiselle-jeanne-
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fontaine-1891.jpg', 'Color_Field_Painting/john-hoyland_22-8-74-1974.jpg',

```
'Expressionism/oskar-kokoschka_not_detected_235843.jpg', 'Impressionism/pierre-
    auguste-renoir_the-first-outing-1876.jpg', 'Expressionism/nicolae-
    tonitza_morning-at-balcic.jpg', 'Baroque/jan-steen_prayer-before-meal-1660.jpg',
    'Realism/ivan-shishkin_among-the-open-valley-1883.jpg', 'Baroque/anthony-van-
    dyck margareta-snyders.jpg', 'Ukiyo e/kitagawa-utamaro a-woman-watches-two-
    children.jpg', 'New_Realism/edward-hopper_untitled.jpg')
    tensor([[0., 0., 0., 0., 1., 0., 0., 0., 0., 0.],
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [1., 0., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 1., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 0., 1., 0., 0.],
            [0., 0., 0., 0., 0., 0., 0., 1., 0., 0.],
            [0., 1., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 0., 1., 0., 0.],
            [0., 0., 0., 0., 0., 0., 1., 0., 0., 0.],
            [0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 1., 0., 0., 0.],
            [0., 1., 0., 0., 0., 0., 0., 0., 0., 0.]
            [1., 0., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 1., 0., 0., 0.]
            [0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 0., 0., 1., 0.],
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 1., 0., 0., 0., 0., 0., 0., 0., 0.]
            [1., 0., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 1., 0., 0., 0.]
            [1., 0., 0., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 1., 0., 0., 0., 0., 0., 0.]
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 1., 0., 0., 0.]
            [0., 0., 1., 0., 0., 0., 0., 0., 0., 0.]
            [0., 0., 0., 0., 0., 0., 0., 0., 1., 0.]])
[]: # CNN model
     import torch.nn as nn
     import torch.nn.functional as F
     import torchvision.models as models
     import torch
     class EffNetLSTM(nn.Module):
```

```
def __init__(self, num_classes):
        super().__init__()
        # EfficientNet-BO backbone (outputs 1280 channels)
        effnet = models.efficientnet_b0(weights=models.EfficientNet_B0_Weights.
 →IMAGENET1K_V1)
        self.cnn = effnet.features
        self.channel_reducer = nn.Sequential(
            nn.Conv2d(1280, 512, kernel_size=1),
            nn.BatchNorm2d(512),
            nn.ReLU()
        )
        self.lstm = nn.LSTM(
            input_size=512,
            hidden_size=256,
            num_layers=2,
            bidirectional=True,
            batch_first=True
        )
        self.classifier = nn.Sequential(
            nn.Linear(512, 256),
            nn.ReLU(),
            nn.Dropout(0.5),
            nn.Linear(256, num_classes)
        )
    def forward(self, x):
        features = self.cnn(x)
        x = self.channel_reducer(features)
        bs, c, h, w = x.size()
        x = x.permute(0, 2, 3, 1).reshape(bs, h*w, c)
        lstm_out, (h_n, c_n) = self.lstm(x)
        last_hidden = torch.cat((h_n[-2], h_n[-1]), dim=1)
        return self.classifier(last_hidden)
model = EffNetLSTM(num_classes)
model.to('cuda')
# Loss and optimizer
import torch.optim as optim
```

0.3 Training the model

```
[]: # Train the model
     num_epochs = 20
     for epoch in range(num_epochs):
         model.train()
         running_loss = 0.0
         train_bar = tqdm(train_loader, desc=f"Epoch {epoch+1}/{num_epochs}")
         for image_paths, labels in train_bar:
             image_tensors = torch.stack([get_image(image_path) for image_path in_
      →image_paths])
             images = image_tensors.to('cuda')
             labels = labels.to('cuda')
             # Forward pass
             outputs = model(images)
             loss = criterion(outputs, labels)
             # Backward and optimize
             optimizer.zero_grad()
             loss.backward()
             optimizer.step()
             running_loss += loss.item()
             train_bar.set_postfix(loss=loss.item())
         avg_train_loss = running_loss / len(train_loader)
         wandb.log({"epoch": epoch+1, "train_loss": avg_train_loss})
         # Validation Loop
         model.eval()
         val loss = 0.0
         correct = 0
         total = 0
         with torch.no_grad():
             val_bar = tqdm(val_loader, desc="Validation")
             for image_paths, labels in val_bar:
```

```
image_tensors = torch.stack([get_image(image_path) for image_path_
  →in image_paths])
            image_tensors = image_tensors.to('cuda')
            labels = labels.to('cuda')
            outputs = model(image_tensors)
            loss = criterion(outputs, labels)
            val_loss += loss.item()
            _, predicted = torch.max(outputs.data, 1)
            total += labels.size(0)
            correct += (predicted == labels.argmax(dim=1)).sum().item()
            val_bar.set_postfix(loss=loss.item())
    avg_val_loss = val_loss / len(val_loader)
    val_accuracy = 100 * correct / total
    wandb.log({"val_loss": avg_val_loss, "val_accuracy": val_accuracy})
    print(f"Epoch {epoch+1}/{num_epochs} - Train Loss: {avg_train_loss:.4f},__

¬Val Loss: {avg_val_loss:.4f}, Val Accuracy: {val_accuracy:.2f}%")
    if(epoch%5==0):
        torch.save(model.state_dict(), f"effnet_rcnn_epoch_{epoch+1} genre.pth")
        torch.save(optimizer.state_dict(),_u
  →f"effnet_rcnn_optimizer_epoch_{epoch+1}_genre.pth")
Epoch 1/20: 58%
                        | 824/1422 [08:56<07:01, 1.42it/s, loss=1.18]
Corrupt JPEG data: premature end of data segment
                      | 1164/1422 [12:35<02:51, 1.51it/s, loss=1.35]
Epoch 1/20: 82%
Corrupt JPEG data: bad Huffman code
Epoch 1/20: 100%
                      | 1422/1422 [15:20<00:00, 1.54it/s, loss=1.03]
Validation: 100%
                      | 610/610 [04:51<00:00, 2.09it/s, loss=0.6]
Epoch 1/20 - Train Loss: 1.1836, Val Loss: 0.8882, Val Accuracy: 70.15%
Epoch 2/20: 56%
                        | 791/1422 [08:29<08:09, 1.29it/s, loss=1.07]
Corrupt JPEG data: bad Huffman code
                      | 1386/1422 [14:49<00:21, 1.69it/s,
Epoch 2/20: 97%
loss=0.583]Corrupt JPEG data: premature end of data segment
Epoch 2/20: 100%|
                      | 1422/1422 [15:11<00:00, 1.56it/s, loss=0.774]
Validation: 100%
                      | 610/610 [04:49<00:00, 2.11it/s, loss=0.204]
Epoch 2/20 - Train Loss: 0.8713, Val Loss: 0.7987, Val Accuracy: 72.35%
Epoch 3/20: 10%|
                          | 148/1422 [01:31<13:01, 1.63it/s,
loss=0.905]Corrupt JPEG data: bad Huffman code
Epoch 3/20: 30%|
                         | 424/1422 [04:28<10:34, 1.57it/s,
loss=0.983]Corrupt JPEG data: premature end of data segment
Epoch 3/20: 100%
                      | 1422/1422 [15:05<00:00, 1.57it/s, loss=0.523]
Validation: 100%
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.157]
Epoch 3/20 - Train Loss: 0.7778, Val Loss: 0.7656, Val Accuracy: 73.76%
Epoch 4/20: 38%
                         | 537/1422 [05:42<10:12, 1.45it/s, loss=0.44]
```

```
Corrupt JPEG data: premature end of data segment
                      | 978/1422 [10:22<04:53, 1.51it/s, loss=1.04]
Epoch 4/20: 69%
Corrupt JPEG data: bad Huffman code
Epoch 4/20: 100%|
                      | 1422/1422 [15:07<00:00, 1.57it/s, loss=0.917]
Validation: 100%
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.035]
Epoch 4/20 - Train Loss: 0.7073, Val Loss: 0.7555, Val Accuracy: 74.07%
                       | 812/1422 [08:41<07:30, 1.35it/s, loss=1.04]
Epoch 5/20: 57%
Corrupt JPEG data: premature end of data segment
                      | 1121/1422 [11:55<03:11,
Epoch 5/20: 79%
                                                 1.57it/s,
loss=0.732]Corrupt JPEG data: bad Huffman code
Epoch 5/20: 100%
                      | 1422/1422 [15:08<00:00, 1.56it/s, loss=0.52]
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.079]
Validation: 100%
Epoch 5/20 - Train Loss: 0.6479, Val Loss: 0.7431, Val Accuracy: 74.32%
             6% l
                          | 80/1422 [00:52<17:43, 1.26it/s, loss=0.683]Corrupt
Epoch 6/20:
JPEG data: bad Huffman code
Epoch 6/20: 18%|
                         | 251/1422 [02:39<11:19, 1.72it/s,
loss=0.924]Corrupt JPEG data: premature end of data segment
Epoch 6/20: 100%|
                    | 1422/1422 [15:08<00:00, 1.57it/s, loss=0.751]
Validation: 100%
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.0644]
Epoch 6/20 - Train Loss: 0.5945, Val Loss: 0.7447, Val Accuracy: 74.87%
Epoch 7/20:
                          | 84/1422 [00:53<15:58, 1.40it/s, loss=0.664]Corrupt
             6% l
JPEG data: premature end of data segment
                       | 768/1422 [08:06<07:09, 1.52it/s,
Epoch 7/20: 54%|
loss=0.575]Corrupt JPEG data: bad Huffman code
Epoch 7/20: 100%
                      | 1422/1422 [15:08<00:00, 1.57it/s, loss=0.173]
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.0354]
Validation: 100%
Epoch 7/20 - Train Loss: 0.5459, Val Loss: 0.7553, Val Accuracy: 74.84%
                        | 484/1422 [05:00<09:12, 1.70it/s, loss=0.48]
Epoch 8/20: 34%|
Corrupt JPEG data: bad Huffman code
Epoch 8/20: 77%
                      | 1094/1422 [11:34<03:16, 1.67it/s,
loss=0.685]Corrupt JPEG data: premature end of data segment
                      | 1422/1422 [15:06<00:00, 1.57it/s, loss=0.312]
Epoch 8/20: 100%
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.195]
Validation: 100%
Epoch 8/20 - Train Loss: 0.4907, Val Loss: 0.7706, Val Accuracy: 74.84%
                         | 211/1422 [02:12<12:17, 1.64it/s,
Epoch 9/20: 15%
loss=0.241]Corrupt JPEG data: bad Huffman code
Epoch 9/20: 24%|
                         | 335/1422 [03:30<11:42, 1.55it/s, loss=0.69]
Corrupt JPEG data: premature end of data segment
Epoch 9/20: 100%
                    | 1422/1422 [15:04<00:00, 1.57it/s, loss=0.492]
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.1]
Validation: 100%
```

Epoch 9/20 - Train Loss: 0.4510, Val Loss: 0.7911, Val Accuracy: 75.19%

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2%1
                           | 24/1422 [00:14<14:36, 1.59it/s,
Epoch 10/20:
loss=0.443]Corrupt JPEG data: premature end of data segment
Epoch 10/20: 82%
                       | 1171/1422 [12:23<02:49, 1.48it/s,
loss=0.485]Corrupt JPEG data: bad Huffman code
Epoch 10/20: 100%
                       | 1422/1422 [15:05<00:00, 1.57it/s, loss=0.174]
Validation: 100%
                      | 610/610 [04:49<00:00, 2.11it/s, loss=0.0766]
Epoch 10/20 - Train Loss: 0.4041, Val Loss: 0.8214, Val Accuracy: 74.76%
                           | 48/1422 [00:30<16:54, 1.35it/s,
Epoch 11/20:
              3%1
loss=0.299]Corrupt JPEG data: premature end of data segment
                        | 1019/1422 [10:47<04:36, 1.46it/s, loss=0.76]
Epoch 11/20: 72%
Corrupt JPEG data: bad Huffman code
                      | 1422/1422 [15:06<00:00, 1.57it/s, loss=0.26]
Epoch 11/20: 100%
Validation: 100%
                      | 610/610 [04:50<00:00, 2.10it/s, loss=0.0493]
Epoch 11/20 - Train Loss: 0.3686, Val Loss: 0.8205, Val Accuracy: 74.91%
Epoch 12/20: 56%
                        | 796/1422 [08:31<06:17, 1.66it/s, loss=0.481]
Corrupt JPEG data: premature end of data segment
                       | 1228/1422 [13:06<02:02, 1.58it/s, loss=0.218]
Epoch 12/20: 86%
Corrupt JPEG data: bad Huffman code
Epoch 12/20: 100%
                       | 1422/1422 [15:08<00:00, 1.57it/s, loss=0.485]
Validation: 100%
                      | 610/610 [04:51<00:00, 2.09it/s, loss=0.00808]
Epoch 12/20 - Train Loss: 0.3261, Val Loss: 0.8793, Val Accuracy: 74.82%
Epoch 13/20: 13%
                          | 179/1422 [01:53<14:04, 1.47it/s, loss=0.106]
Corrupt JPEG data: premature end of data segment
Epoch 13/20: 29%|
                          | 419/1422 [04:24<11:09, 1.50it/s, loss=0.545]
Corrupt JPEG data: bad Huffman code
                       | 1422/1422 [15:05<00:00, 1.57it/s, loss=0.205]
Epoch 13/20: 100%
Validation: 100%
                      | 610/610 [04:48<00:00, 2.11it/s, loss=0.024]
Epoch 13/20 - Train Loss: 0.2988, Val Loss: 0.8529, Val Accuracy: 75.08%
                          | 328/1422 [03:25<11:18, 1.61it/s, loss=0.1]
Epoch 14/20: 23%
Corrupt JPEG data: bad Huffman code
Epoch 14/20: 35%|
                         | 500/1422 [05:15<11:12, 1.37it/s, loss=0.261]
Corrupt JPEG data: premature end of data segment
Epoch 14/20: 100%|
                       | 1422/1422 [15:06<00:00, 1.57it/s, loss=0.174]
Validation: 100%
                      | 610/610 [04:49<00:00, 2.10it/s, loss=0.00185]
Epoch 14/20 - Train Loss: 0.2672, Val Loss: 0.9576, Val Accuracy: 74.77%
                        | 781/1422 [08:12<06:12, 1.72it/s, loss=0.558]
Epoch 15/20: 55%
Corrupt JPEG data: bad Huffman code
Epoch 15/20: 68%
                        | 966/1422 [10:12<05:05, 1.49it/s, loss=0.226]
Corrupt JPEG data: premature end of data segment
                      | 1422/1422 [15:07<00:00, 1.57it/s, loss=0.364]
Epoch 15/20: 100%|
Validation: 100%
                      | 610/610 [04:49<00:00, 2.11it/s, loss=0.00409]
Epoch 15/20 - Train Loss: 0.2376, Val Loss: 0.9952, Val Accuracy: 74.33%
```

```
Epoch 16/20: 49% | 694/1422 [07:24<08:46, 1.38it/s, loss=0.245]
Corrupt JPEG data: premature end of data segment
Epoch 16/20: 81% | 1147/1422 [12:10<03:21, 1.37it/s, loss=0.103]
Corrupt JPEG data: bad Huffman code
Epoch 16/20: 100%
                  | 1422/1422 [15:04<00:00, 1.57it/s, loss=0.234]
Validation: 33%
                         | 204/610 [01:51<03:41, 1.83it/s, loss=1.56]
                                           Traceback (most recent call last)
 KeyboardInterrupt
 Cell In[10], line 36
      34 val_bar = tqdm(val_loader, desc="Validation")
      35 for image_paths, labels in val_bar:
             image_tensors = torch.stack([get_image(image_path) for image_path image_path)
 ---> 36
  →image_paths])
             image tensors = image tensors.to('cuda')
      37
      38
             labels = labels.to('cuda')
 Cell In[10], line 36, in <listcomp>(.0)
      34 val_bar = tqdm(val_loader, desc="Validation")
      35 for image_paths, labels in val_bar:
 ---> 36
             image_tensors = torch.stack([get image(image path)] for image_path i:
  →image_paths])
      37
             image_tensors = image_tensors.to('cuda')
      38
             labels = labels.to('cuda')
 Cell In[8], line 5, in get_image(image_path, image_size)
       3 def get_image(image_path,image_size=224):
            try:
 ---> 5
                img = cv2.imread('./wikiart/' + image_path)
                 if img is None:
                     raise ValueError(f"Image not loaded: ./wikiart/{image path})
```

Error in callback <bound method _WandbInit._pause_backend of <wandb.sdk.wandb_init._WandbInit object at 0x7e8191d8c460>> (for post_run_cell):

KeyboardInterrupt:

```
767 def publish_pause(self) -> None:
            pause = pb.PauseRequest()
    768
--> 769
            self._publish_pause(pause)
File ~/.local/lib/python3.10/site-packages/wandb/sdk/interface/interface shared
 ⇒py:289, in InterfaceShared. publish pause(self, pause)
    287 def publish pause(self, pause: pb.PauseRequest) -> None:
            rec = self. make request(pause=pause)
    288
--> 289
            self. publish(rec)
File ~/.local/lib/python3.10/site-packages/wandb/sdk/interface/interface_sock.p
 →39, in InterfaceSock. publish(self, record, local)
     37 def _publish(self, record: "pb.Record", local: Optional[bool] = None) -
 →None:
            self._assign(record)
     38
            self._sock_client.send_record_publish(record)
---> 39
File ~/.local/lib/python3.10/site-packages/wandb/sdk/lib/sock_client.py:174, in
 →SockClient.send_record_publish(self, record)
    172 server req.request id = record.control.mailbox slot
    173 server req.record publish.CopyFrom(record)
--> 174 self.send server request(server req)
File ~/.local/lib/python3.10/site-packages/wandb/sdk/lib/sock_client.py:154, in
 →SockClient.send_server_request(self, msg)
    153 def send server request(self, msg: spb.ServerRequest) -> None:
            self._send_message(msg)
--> 154
File ~/.local/lib/python3.10/site-packages/wandb/sdk/lib/sock_client.py:151, in
 →SockClient._send_message(self, msg)
    149 header = struct.pack("<BI", ord("W"), raw_size)
    150 with self._lock:
            self._sendall_with_error_handle(header + data)
--> 151
File ~/.local/lib/python3.10/site-packages/wandb/sdk/lib/sock client.py:130, in
 →SockClient. sendall with error handle(self, data)
    128 start_time = time.monotonic()
    129 try:
            sent = self._sock.send(data)
--> 130
            # sent equal to 0 indicates a closed socket
    131
    132
            if sent == 0:
BrokenPipeError: [Errno 32] Broken pipe
```

The Kernel crashed while executing code in the current cell or a previous cell.

Please review the code in the cell(s) to identify a possible cause of the $_{\!\sqcup}$ $_{\!\circlearrowleft} failure.$

Click here for more info.

View Jupyter log for further details.