```
In [9]: import torch
        import torch.nn as nn
        import torch.optim as optim
        from torch.utils.data import Dataset, DataLoader
        import torchvision
        import torchvision.transforms as transforms
        from torchvision.datasets import ImageFolder
        import timm # Torch Image Classifier
        import matplotlib.pyplot as plt # For data viz
        import pandas as pd
        import numpy as np
        import sys
        from tqdm.notebook import tqdm
        print('System Version:', sys.version)
        print('PyTorch version', torch. version )
        print('Torchvision version', torchvision.__version__)
        print('Numpy version', np.__version__)
        print('Pandas version', pd.__version__)
        System Version: 3.10.13 | packaged by conda-forge | (main, Dec 23 2023, 15:36:39) [GCC 12.3.0]
        PyTorch version 2.1.2
        Torchvision version 0.16.2
```

Numpy version 1.26.4 Pandas version 2.2.2

```
In [10]: class PlayingCardDataset(Dataset):
             def init (self, data dir, transform = None):
                 self.data = ImageFolder(data dir, transform = transform)
             def len (self):
                 return len(self.data)
             def __getitem__(self, idx):
                 return self.data[idx]
             @property
             def classes(self):
                 return self.data.classes
         data dir = "/kaggle/input/cards-image-datasetclassification/train"
         target to class = {v:k for k, v in ImageFolder(data dir).class to idx.items()}
In [12]: transform = transforms.Compose([
             transforms.Resize((128, 128)),
             transforms.ToTensor(),
         ])
         dataset = PlayingCardDataset(
             data_dir = data_dir, transform = transform
In [13]: dataloader = DataLoader(dataset, batch_size=32, shuffle=True)
In [14]: for images, label in dataloader:
             break
```

```
In [15]: class SimpleCardClassifier(nn.Module):
    def __init__(self, num_classes=53):
        super(SimpleCardClassifier, self).__init__()

        self.base_model = timm.create_model('efficientnet_b0', pretrained=True)
        self.features = nn.Sequential(*list(self.base_model.children())[:-1])
        enet_out_size = 1280

        self.classifier = nn.Linear(enet_out_size, num_classes)

    def forward(self, x):
        x = self.features(x)
        output = self.classifier(x)
        return output
```

```
In [16]: model = SimpleCardClassifier(num_classes=53)
```

```
In [17]: # Testing the model
    example_out = model(images)
    example_out.shape # [batch_size, num_classes]

Out[17]: torch.Size([32, 53])

In [18]: # Loss Function
    criterion = nn.CrossEntropyLoss()
    optimizer = optim.Adam(model.parameters(), lr=0.001)

In [19]: criterion(example_out, label)

Out[19]: tensor(4.0062, grad_fn=<NllLossBackward0>)
```

```
In [21]: device = torch.device('cuda:0' if torch.cuda.is_available() else "cpu")
    print(device)
```

cuda:0

```
In [22]: num epochs = 10
         train_losses, val_losses = [], []
         model = SimpleCardClassifier(num classes=53)
         model.to(device)
         # Loss Function
         criterion = nn.CrossEntropyLoss()
         optimizer = optim.Adam(model.parameters(), lr=0.001)
         for epoch in range(num epochs):
             # Set the model to train
             model.train()
             running loss = 0.0
             for images, labels in tqdm(train dataloader, desc="Training Loop"):
                 images, labels = images.to(device), labels.to(device)
                 optimizer.zero grad()
                 outputs = model(images)
                 loss = criterion(outputs, labels)
                 loss.backward()
                 optimizer.step()
                 running loss += loss.item() * labels.size(0)
             train loss = running loss / len(train dataloader.dataset)
             train losses.append(train loss)
             # Validation Phase
             model.eval()
             running loss = 0.0
             with torch.no grad():
                 for images, labels in tqdm(valid dataloader, desc="Validation Loop"):
                     images, labels = images.to(device), labels.to(device)
                     outputs = model(images)
                     loss = criterion(outputs, labels)
                     running loss += loss.item() * labels.size(0)
             val loss = running loss / len(valid dataloader.dataset)
             val losses.append(val loss)
             # print epoch stats
             print(f"Epoch {epoch+1}/{num epochs} - Train loss: {train loss}, Validation loss: {val loss}")
```

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Epoch 1/10 - Train loss: 1.5781370137445825, Validation loss: 0.3883966752943003

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Epoch 2/10 - Train loss: 0.5411015779624331, Validation loss: 0.2215240403728665

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Epoch 3/10 - Train loss: 0.3300596840624919, Validation loss: 0.1535666601837806

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Epoch 4/10 - Train loss: 0.2291205584424365, Validation loss: 0.18126810570862495

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Epoch 5/10 - Train loss: 0.18844059318429152, Validation loss: 0.22565973403881182

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Epoch 6/10 - Train loss: 0.1521692796877086, Validation loss: 0.16000195920889107

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Epoch 7/10 - Train loss: 0.15477401118213457, Validation loss: 0.16202498696885018

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Epoch 8/10 - Train loss: 0.14209554583060965, Validation loss: 0.25466126299129344

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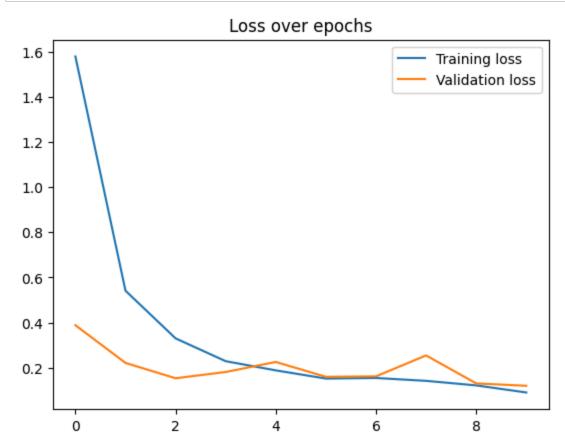
Epoch 9/10 - Train loss: 0.12193363885603384, Validation loss: 0.130748725169391

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Epoch 10/10 - Train loss: 0.09061787257915029, Validation loss: 0.12028200890658036

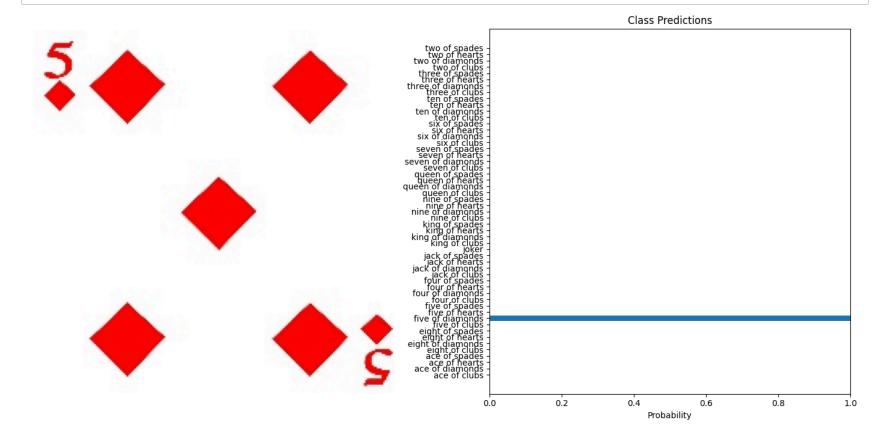
```
In [23]: plt.plot(train_losses, label='Training loss')
    plt.plot(val_losses, label='Validation loss')
    plt.legend()
    plt.title("Loss over epochs")
    plt.show()
```



```
In [24]: import torch
         import torchvision.transforms as transforms
         from PIL import Image
         import matplotlib.pyplot as plt
         import numpy as np
         # Load and preprocess the image
         def preprocess image(image path, transform):
             image = Image.open(image path).convert("RGB")
             return image, transform(image).unsqueeze(0)
         # Predict using the model
         def predict(model, image_tensor, device):
             model.eval()
             with torch.no grad():
                 image tensor = image tensor.to(device)
                 outputs = model(image tensor)
                 probabilities = torch.nn.functional.softmax(outputs, dim=1)
             return probabilities.cpu().numpy().flatten()
         # Visualization
         def visualize predictions(original image, probabilities, class names):
             fig, axarr = plt.subplots(1, 2, figsize=(14, 7))
             # Display image
             axarr[0].imshow(original image)
             axarr[0].axis("off")
             # Display predictions
             axarr[1].barh(class names, probabilities)
             axarr[1].set xlabel("Probability")
             axarr[1].set title("Class Predictions")
             axarr[1].set_xlim(0, 1)
             plt.tight layout()
             plt.show()
         # Example usage
         test image = "/kaggle/input/cards-image-datasetclassification/test/five of diamonds/2.jpg"
         transform = transforms.Compose([
             transforms.Resize((128, 128)),
             transforms.ToTensor()
         ])
```

```
original_image, image_tensor = preprocess_image(test_image, transform)
probabilities = predict(model, image_tensor, device)

# Assuming dataset.classes gives the class names
class_names = dataset.classes
visualize_predictions(original_image, probabilities, class_names)
```

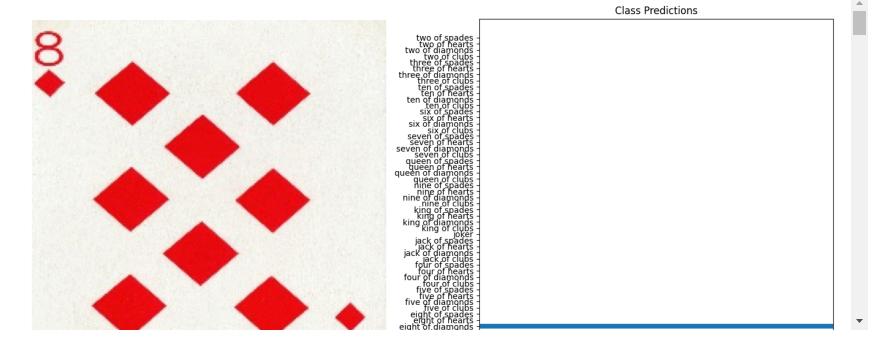


```
In [25]: from glob import glob

test_images = glob('../input/cards-image-datasetclassification/test/*/*')
test_example = np.random.choice(test_images,10)

for example in test_example:
    original_image, image_tensor = preprocess_image(example, transform)
    probabilities = predict(model, image_tensor, device)

class_names = dataset.classes
    visualize_predictions(original_image, probabilities, class_names)
```



```
In [26]: def calculate_accuracy(model, data_loader, device):
    model.eval()
    correct = 0
    total = 0
    with torch.no_grad():
        for images, labels in data_loader:
            images, labels = images.to(device), labels.to(device)
            outputs = model(images)
            _, predicted = torch.max(outputs.data, 1)
            total += labels.size(0)
            correct += (predicted == labels).sum().item()
    return 100 * correct / total

# After training, calculate accuracy on the test set

test_accuracy = calculate_accuracy(model, test_dataloader, device)
    print(f"Test Accuracy: {test_accuracy}%")
```

Test Accuracy: 98.11320754716981%