Lecture 12- Introduction to Python Project

Guiliang Liu

The Chinese University of Hong Kong, Shenzhen

CSC-1004: Computational Laboratory Using Java Course Page: [Click]

Outline

- Python Project Management and Basic Knowledge
- Basic Machine Learning with Python



Main Components in a Machine Learning Project

• Environment Manager (e.g., MiniConda): manage the machine learning environment.

Machine Learning Library (e.g., Pytorch, Scikit-Learn): provide API for machine learning algorithms.

Programming language (e.g., Python): implement the logic for handling the input/output.



Step I: Read the configuration

The main function is the entrance of a Python program.

```
if name == ' main ':
   arg = read args()
   """toad training settings"""
   config = load_config(arg)
   """train model and record results"""
    run(config)
   """plot the mean results"""
   plot_mean()
```

- Run from the command line.
 python main.py ./config/minist.yaml
- Read the input argument.

$$arg = read _args()$$

Load the config and run training.

```
config=load_config()
run(config)
```



Step I: Read the configuration

The main function is the entrance of a Python program.

The "minist.yaml" file.

```
batch_size: 64 # input batch size for training

test_batch_size: 1000 # input batch size for testing

epochs: 15 # number of epochs to train

lr: 0.01 # learning rate

gamma: 0.7 # learning rate step gamma

no_cuda: True # disables CUDA training

no_mps: True # disables macOS GPU training

dry_run: False # quickly check a single pass

seed: 123 # random seeds for the three runs are 123, 321, 666

log_interval: 10 # how many batches to wait before logging training status

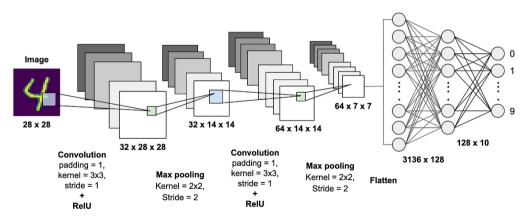
save_model: True # For Saving the current Model
```



Step II: Define the Neural Network Model

Define the neural network structure with Pytorch.

Note that this is the complete structure. We use a simplified one in our code.



Step II: Define the Neural Network Model

Define the neural network structure with Pytorch.

```
class Net(nn.Module):
   def init (self):
        super(Net, self).__init__()
        self.conv1 = nn.Conv2d(1, 32, 3, 1)
        self.conv2 = nn.Conv2d(32, 64, 3, 1)
        self.dropout1 = nn.Dropout(0.25)
        self.dropout2 = nn.Dropout(0.5)
        self.fc1 = nn.Linear(9216. 128)
        self.fc2 = nn.Linear(128, 10)
   def forward(self. x):
        x = self.conv1(x)
        x = F_relu(x)
        x = self.conv2(x)
        x = F.relu(x)
        x = F.max pool2d(x, 2)
```

Define the network layers.

```
def___init___(self):
```

```
香港中文大學(深圳)
The Chinese University of Hong Kong, Shenzhen
```

Step II: Define the Neural Network Model

Define the neural network structure with Pytorch.

```
def forward(self, x):
    x = self.conv1(x)
    x = F.relu(x)
    x = self.conv2(x)
    x = F.relu(x)
    x = F.max_pool2d(x, 2)
    x = self.dropout1(x)
    x = torch.flatten(x. 1)
    x = self.fc1(x)
    x = F.relu(x)
    x = self.dropout2(x)
    x = self.fc2(x)
    output = F.log_softmax(x, dim=1)
    return output
```

Connect these layers to map inputs

 (i.e., x) to outputs.

 def forward(self, x):

```
香港中文大學(深圳)
The Chinese University of Hong Kong, Shenzhen
```

Step III: Run the code

The run() function – Part I.

```
def run(config):
   use cuda = not config.no cuda and torch.cuda.is available()
   use mps = not config.no mps and torch.backends.mps.is available()
   torch.manual seed(config.seed)
    if use cuda:
        device = torch.device("cuda")
   elif use mps:
        device = torch.device("mps")
   else:
        device = torch.device("cpu")
   train_kwargs = {'batch_size': config.batch_size, 'shuffle': True}
   test_kwargs = {'batch_size': config.test_batch_size, 'shuffle': True}
    if use cuda:
        cuda kwargs = {'num workers': 1,
                       'pin memory': True.}
        train kwargs.update(cuda kwargs)
        test kwargs.update(cuda kwargs)
```

 Set up CUDA. CUDA is a parallel computing platform and application programming interface (API). Ignore it if you don't need to use GPU.



Step III: Write the Run() function

The run() function - Part II.

- Download training and testing data from the web.
- Store them in data loader for training and testing.



Step III: Write the Run() function

```
The run() function - Part III.
```

```
model = Net().to(device)
optimizer = optim.Adadelta(model.parameters(), lr=config.lr)
"""record the performance"""
epoches = []
training_accuracies = []
training_loss = []
testing_accuracies = []
testing_loss = []
```

 Initialize the neural network model.

```
model=Net().to(device)
```

Initialize the optimizer.
 optimizer=optim.Adadelta(...



Step III: Write the Run() function

The run() function - Part IV.

```
scheduler = StepLR(optimizer, step_size=1, gamma=config.gamma)
for epoch in range(1, config.epochs + 1):
    train_acc, train_loss = train(config, model, device, train_loader, optimizer, epoch)
    """record training info, Fill your code"""
    test_acc, test_loss = test(model, device, test_loader)
    """record testing info, Fill your code"""
    scheduler.step()
    """update the records, Fill your code"""
```

- Train the model in each iteration.
- Test the model in each iteration.
- Go over the dataset "config epochs" times.



Step IV: Write the train() function

The train() function.

```
model.train()
for batch_idx, (data, target) in enumerate(train_loader):
    data, target = data.to(device), target.to(device)
    optimizer.zero_grad()
    output = model(data)
    loss = F.nll_loss(output, target)
    loss.backward()
    optimizer.step()
'''Fill your code'''
training_acc, training_loss = None, None # replace this line
return training_acc, training_loss
```

- Set the training mode. model.train()
- Obtain output from input (e.g., data, the images).
 output=model(data)
- Calculate Loss due to the estimation error.

```
loss=F.nll loss(output, target)
```



Step IV: Write the train() function

The train() function.

```
model.train()
for batch_idx, (data, target) in enumerate(train_loader):
    data, target = data.to(device), target.to(device)
    optimizer.zero_grad()
    output = model(data)
    loss = F.nll_loss(output, target)
    loss.backward()
    optimizer.step()
'''Fill your code'''
training_acc, training_loss = None, None # replace this line
return training_acc, training_loss
```

- Calculate the gradient.loss.backward()
- Perform the gradient descent for updating the neural network.

```
optimizer.step()
```



Step V: Write the test() function

The test() function.

```
model.eval()
test_loss = 0
correct = 0
with torch.no_grad():
    for data, target in test_loader:
        '''Fill your code'''
        pass
testing_acc, testing_loss = None, None # replace this line
return testing_acc, testing_loss
```

Set the testing mode.

```
\mathsf{model}.\mathsf{eval}()
```

• Stop gradient descent.

```
with torch.no_grad()
```



Question and Answering (Q&A)



