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1. Installation

```
pip install torch torchvision torchaudio
```

Verify Installation:

```
import torch
print(torch.__version__)
```

2. Import PyTorch

```
import torch
import torch.nn as nn
import torch.optim as optim
import torchvision
```

3. Tensors

Create Tensors:

```
x = torch.tensor([5.0, 10.0, 15.0])
y = torch.zeros(3)                # Zero tensor
z = torch.ones(3)                 # Ones tensor
rand = torch.rand(3, 3)           # Random tensor
```

Tensor Shape & Type:

```
print(x.shape)
print(x.dtype)
```

Reshape Tensors:

```
x = torch.rand(4, 4)
x_resaped = x.view(2, 8)
```

4. Tensor Operations

```
a = torch.tensor([1, 2, 3])
b = torch.tensor([4, 5, 6])
```

Basic Operations

```
c = a + b
```

```
d = a * b
```

```
e = torch.dot(a, b)
```

Dot product

```
f = torch.sum(a)
```

Sum of elements

Matrix Multiplication:

```
x = torch.rand(2, 3)
```

```
y = torch.rand(3, 2)
```

```
z = torch.mm(x, y)
```

Element-wise Operations:

```
x = torch.rand(3)
y = torch.exp(x)          # Exponential
z = torch.sqrt(x)         # Square root
```

5. Autograd (Automatic Differentiation)

```
x = torch.tensor(2.0, requires_grad=True)
y = x**3 + 5*x
y.backward()               # Calculate gradients
print(x.grad)             #  $dy/dx = 3x^2 + 5$ 
```

6. Neural Network Basics

Simple Neural Network:

```
class Net(nn.Module):
    def __init__(self):
        super(Net, self).__init__()
        self.fc1 = nn.Linear(4, 3)
        self.fc2 = nn.Linear(3, 1)

    def forward(self, x):
        x = torch.relu(self.fc1(x))
        x = torch.sigmoid(self.fc2(x))
        return x

net = Net()
print(net)
```

7. Loss and Optimizer

```
criterion = nn.MSELoss()                # Mean Squared Error
optimizer = optim.SGD(net.parameters(), lr=0.01)
```

8. Training a Neural Network

```
for epoch in range(100):
    optimizer.zero_grad()                # Zero gradients
    outputs = net(x)                     # Forward pass
    loss = criterion(outputs, y)          # Compute loss
    loss.backward()                       # Backward pass
    optimizer.step()                     # Update weights
```

9. Loading Data (Datasets & Dataloaders)

```
from torchvision import datasets, transforms

transform = transforms.Compose([
    transforms.ToTensor(),
    transforms.Normalize((0.5,), (0.5,))
])

train_data = datasets.MNIST(root='./data', train=True, download=True,
transform=transform)
train_loader = torch.utils.data.DataLoader(dataset=train_data,
batch_size=64, shuffle=True)
```

10. GPU Acceleration

```
device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
```

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

11. Saving and Loading Models

```
# Save Model
torch.save(net.state_dict(), 'model.pth')

# Load Model
model = Net()
model.load_state_dict(torch.load('model.pth'))
model.eval()
```

12. Testing the Model

```
with torch.no_grad():
    for data in train_loader:
        images, labels = data
        outputs = model(images)
        _, predicted = torch.max(outputs, 1)
```

13. Common PyTorch Functions

Function	Description
<code>torch.tensor()</code>	Creates a tensor.
<code>torch.rand()</code>	Random tensor.
<code>torch.zeros()</code>	Zero tensor.
<code>torch.ones()</code>	Ones tensor.
<code>torch.mm()</code>	Matrix multiplication.
<code>torch.sum()</code>	Sum of elements.
<code>torch.cat()</code>	Concatenate tensors.

Function	Description
<code>torch.max()</code>	Maximum value.
<code>torch.mean()</code>	Mean value.
<code>torch.relu()</code>	ReLU activation function.

14. Plotting with PyTorch

```
import matplotlib.pyplot as plt

x = torch.linspace(-10, 10, 100)
y = torch.sin(x)

plt.plot(x.numpy(), y.numpy())
plt.title('Sine Wave')
plt.show()
```

15. Example: Linear Regression

```
# Data
x = torch.rand(100, 1)
y = 3 * x + 2 + torch.randn(100, 1) * 0.1

# Model
model = nn.Linear(1, 1)
criterion = nn.MSELoss()
optimizer = optim.SGD(model.parameters(), lr=0.01)

# Training Loop
for epoch in range(500):
    y_pred = model(x)
    loss = criterion(y_pred, y)
    optimizer.zero_grad()
    loss.backward()
    optimizer.step()
```

```
# Visualization
plt.scatter(x.numpy(), y.numpy())
plt.plot(x.numpy(), y_pred.detach().numpy(), color='red')
plt.title('Linear Regression with PyTorch')
plt.show()
```