```
train_losses.append(running_loss / len(trainloader))
    model.eval()
    correct, total = 0, 0
    with torch.no_grad():
      for inputs, labels in testloader:
         inputs, labels = inputs.to(device), labels.to(device)
         outputs = model(inputs)
         _, predicted = torch.max(outputs, 1)
        total += labels.size(0)
        correct += (predicted == labels).sum().item()
    test accuracies.append(100 * correct / total)
    print(f"Epoch [{epoch+1}/{epochs}] | Loss: {train | losses[-1]:.4f} | Accuracy:
{test_accuracies[-1]:.2f}%")
  return train_losses, test_accuracies
# Train with both optimizers
sgd_losses, sgd_acc = train_model(optimizer_type="sgd", epochs=10, lr=0.01,
momentum=0.9)
adam_losses, adam_acc = train_model(optimizer_type="adam", epochs=10, lr=0.001)
# Plot training loss comparison
plt.figure(figsize=(10,5))
plt.plot(sgd_losses, label='SGD + Momentum')
```

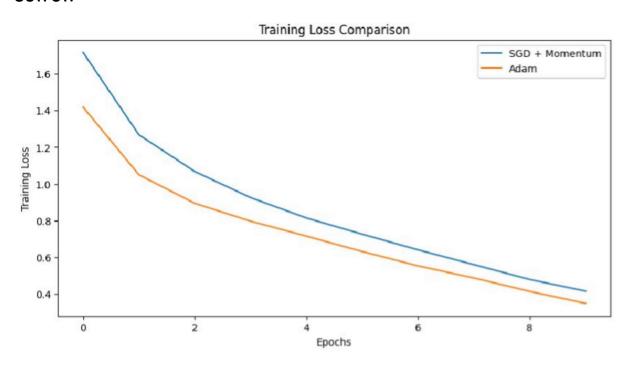
DEEP LEARNING (AI23531)

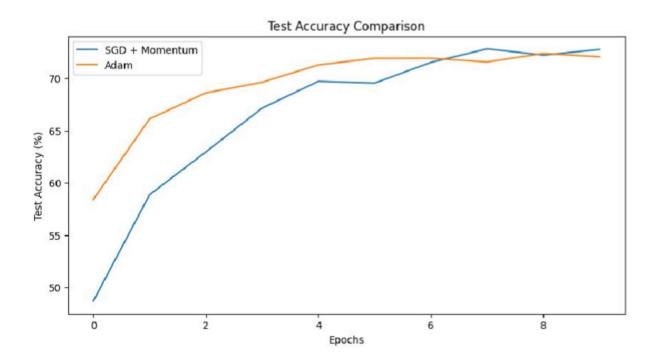
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```
plt.plot(adam_losses, label='Adam')
plt.xlabel('Epochs'); plt.ylabel('Training Loss')
plt.title('Training Loss Comparison')
plt.legend()
plt.show()

# Plot accuracy comparison
plt.figure(figsize=(10,5))
plt.plot(sgd_acc, label='SGD + Momentum')
plt.plot(adam_acc, label='Adam')
plt.xlabel('Epochs'); plt.ylabel('Test Accuracy (%)')
plt.title('Test Accuracy Comparison')
plt.legend()
plt.show()
```

OUTPUT:





RESULT: The Adam optimizer achieved faster convergence and higher accuracy 72.78% compared to SGD with Momentum 72.08% on CIFAR-10.

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EXP NO: 04	IMPLEMENT CNN FROM SCRATCH
DATE: 06/09/2025	

AIM: To implement a Convolutional Neural Network (CNN) from scratch for image classification using the CIFAR-10 dataset, evaluate its performance, and visualize the learned convolution filters.

ALGORITHM:

- Load the CIFAR-10 dataset and apply normalization preprocessing.
- Define a CNN model with convolution, ReLU, pooling, and fully-connected layers.
- Perform forward propagation to generate class predictions.
- Compute the classification loss using Cross-Entropy Loss.
- Perform backpropagation and update model weights using Adam optimizer.
- Evaluate classification accuracy on the test dataset.
- Visualize the learned filters from the first convolution layer.

CODE:

CNN from scratch on CIFAR-10

import torch, torch.nn as nn, torch.optim as optim

import torch.nn.functional as F

from torchvision import datasets, transforms

from torch.utils.data import DataLoader

import matplotlib.pyplot as plt

import numpy as np

import torchvision

Device

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")

Data Preprocessing

transform = transforms.Compose([

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```

```
transforms.ToTensor(),
  transforms.Normalize((0.5,0.5,0.5),(0.5,0.5,0.5))
])
trainset = datasets.CIFAR10(root='./data', train=True, download=True,
transform=transform)
testset = datasets.CIFAR10(root='./data', train=False, download=True,
transform=transform)
trainloader = DataLoader(trainset, batch_size=64, shuffle=True)
testloader = DataLoader(testset, batch size=64, shuffle=False)
# CNN Model
class SimpleCNN(nn.Module):
  def __init__(self):
    super(SimpleCNN, self).__init__()
    self.conv1 = nn.Conv2d(3, 32, 3, padding=1)
    self.conv2 = nn.Conv2d(32, 64, 3, padding=1)
    self.pool = nn.MaxPool2d(2,2)
    self.fc1 = nn.Linear(64*8*8, 128)
    self.fc2 = nn.Linear(128, 10)
  def forward(self, x):
    x = self.pool(F.relu(self.conv1(x))) # 32x16x16
    x = self.pool(F.relu(self.conv2(x))) # 64x8x8
    x = x.view(-1, 64*8*8)
    x = F.relu(self.fc1(x))
    return self.fc2(x)
model = SimpleCNN().to(device)
```

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```

```
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr=0.001)
# Training
for epoch in range(10):
  model.train()
  for imgs, labels in trainloader:
    imgs, labels = imgs.to(device), labels.to(device)
    optimizer.zero_grad()
    outputs = model(imgs)
    loss = criterion(outputs, labels)
    loss.backward()
    optimizer.step()
  print(f"Epoch {epoch+1} complete")
# Evaluation
correct, total = 0, 0
model.eval()
with torch.no_grad():
  for imgs, labels in testloader:
    imgs, labels = imgs.to(device), labels.to(device)
    outputs = model(imgs)
    _, pred = torch.max(outputs, 1)
    total += labels.size(0)
    correct += (pred == labels).sum().item()
accuracy = 100 * correct / total
print(f"Test Accuracy: {accuracy:.2f}%")
```