

13	27/10/25	Architecture of pre-trained models	
14	27/10/25	Pre-trained CNN model as a Feature Extractor using Transfer learning	Eg. ✓
15	27/10/25	Yolo model to Detect the objects.	Eg. ✓

23/10/25

Implement a pre-trained CNN model
lab 14: as a feature extractor using Transfer Learning

Aim:

To use a pre-trained CNN (e.g. VGG16, ResNet50, MobileNet) as a feature extractor for image classification tasks.

Objectives:

- 1) To apply transfer learning for feature extraction
- 2) To fine-tune the pre-trained network for a new dataset.
- 3) To train a classifier on extracted features.

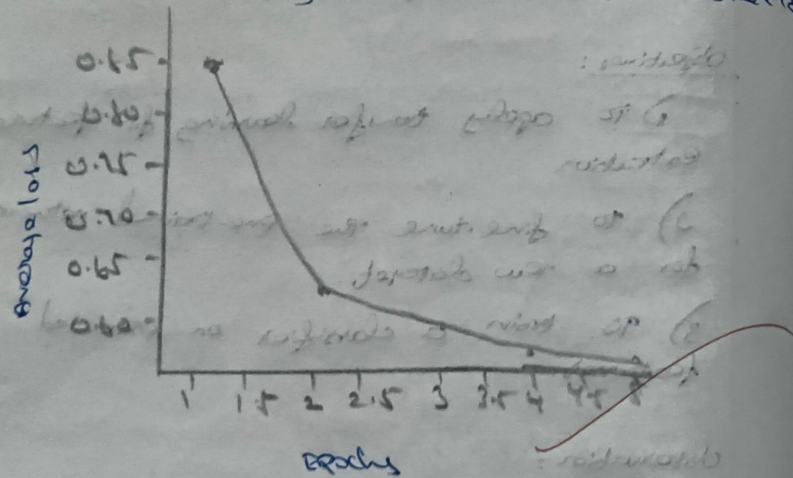
Observation:

- 1) Early layers of pre-trained CNNs provide ~~valuable~~ features.
- 2) ~~Using~~ pre-trained models reduces training time and improves performance.
- 3) The final classification layer can be replaced with a custom classifier.

Pseudocode:

- 1) Import Libraries
- 2) Load pre-trained CNN model
- 3) Freeze base model layers.
- 4) Add new classification layers (Dense + Softmax)
- 5) Compile model (Adam optimizer, Categorical cross entropy)

Training Loss Curve - Transfer Learning with ResNet18



- c) Train on new dataset
- d) Evaluate performance and visualize accuracy/loss.

Output:

Epoch 1, Loss : 0.8348
 Epoch 2, Loss : 0.6229
 Epoch 3, Loss : 0.5932
 Epoch 4, Loss : 0.5997
 Epoch 5, Loss : 0.5699

Result:

Successfully implemented the pre-trained CNN model as a feature Extractor using transfer learning.


```

# Loss and optimizer
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr=0.001)

# Lists to store loss values for graph
loss_values = []

# training for a few epochs
epochs = 5
for epoch in range(epochs):
    running_loss = 0.0
    for images, labels in trainloader:
        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()

    avg_loss = running_loss / len(trainloader)
    loss_values.append(avg_loss)
    print(f"Epoch [{epoch+1}/{epochs}], Loss: {avg_loss:.4f}")

print("\n Training complete using Pre-trained ResNet18 as Feature Extractor.\n")

# Plot the training loss graph
plt.figure(figsize=(7, 4))
plt.plot(range(1, epochs + 1), loss_values, marker='o', linestyle='-', linewidth=2)
plt.title("Training Loss Curve - Transfer Learning with ResNet18")
plt.xlabel("Epochs")
plt.ylabel("Average Loss")
plt.grid(True)
plt.show()

```

using device: cuda

Downloading: "https://download.pytorch.org/models/resnet18-f37072f6.pth" to /root/.cache/torch/hub/checkpoints/resnet18-f37072f6.pth
 /usr/local/lib/python3.12/dist-packages/torchvision/models/_utils.py:206: UserWarning: The parameter 'pretrained' is deprecated since 0.15 and may be removed in the future, please use 'weights' instead.
 warnings.warn(
 /usr/local/lib/python3.12/dist-packages/torchvision/models/_utils.py:225: UserWarning: Arguments other than a weight enum or 'None' for 'weights' are deprecated since 0.15 and may be removed in the future. The current behavior is equivalent to passing 'weights=ResNet18_Weights.DEFAULT'. You can also use 'weights=ResNet18_Weights.DEFAULT' to get the most up-to-date weights.
 warnings.warn(
 100% 44.7M/44.7M [00:00<00:00, 15.9MB/s]
 100% 178M/178M [00:03<00:00, 43.5MB/s]
 Epoch [1/5], Loss: 0.5481
 Epoch [2/5], Loss: 0.6266
 Epoch [3/5], Loss: 0.5927
 Epoch [4/5], Loss: 0.5797
 Epoch [5/5], Loss: 0.5699

Training complete using Pre-trained ResNet18 as Feature Extractor.

