

WEEK - 13

Aim :- To analyze and understand the architecture and working principle of a pretrained deep learning model (such as VGG16, ResNet, Inception) used for image classification and feature extraction.

Algorithm :-

- 1) Import a pre-trained model from keras applications.
- 2) Load model with pretrained weights.
- 3) Display model summary (layers, parameters).
- 4) Analyze layer types (conv, pool, dense).
- 5) Identify trainable vs non-trainable parameters.
- 6) optionally, visualize feature maps.

pseudocode :-

Begin

Import deep learning library (eg: Tensorflow, pytorch)

Load a pretrained model

→ include weights = 'imagenet'

→ exclude top layer if using for feature extraction.

Display model summary

→ print layer names, types, output shapes & parameter counts.

For each layer in model:

→ Identify type (conv, pool, dense, etc...)

→ Note activation function and No. of filters.

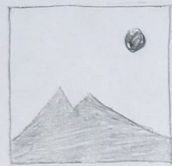
visualize architecture diagram

→ show flow from input image to output class.

Train the model

Freeze lower layers if needed.

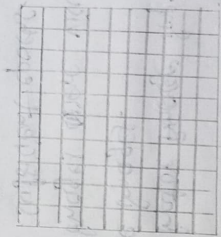
Usage of pre trained Architecture



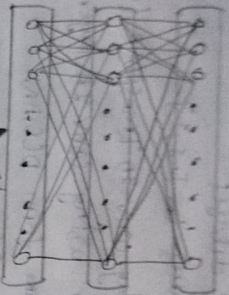
Input Image.

Feature extraction process

pretrained deep learning Architecture



Extracted feature



observation :-

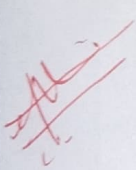
- The pre-trained model has convolutional, pooling and fully connected layers.
- Convolutional layers extract features, pooling layers reduce size.
- pre-trained weights allow faster training and better accuracy.
- Sample images are correctly classified (or) have meaningful features extracted.

conclusion :-

pre trained models save time and computation by using already-learned image features, making them powerful for transfer learning and real world AI applications.

Results :-

The Architecture of pre-trained CNN's was successfully analyzed and visualized.



output :-

Total parameters : 138357544

Trainable parameters : 138357544

Non Trainable parameters : 0

Layer names:

conv0_1 : conv2D

bn_1 : Batch Normal 2D

relu : ReLU

maxpool : max pool 2D

layer_1 : sequential

layer_2 : sequential

layer_3 : sequential

layer_4 : sequential

avg pool : Adaptive Avg pool 2D

f1 : Linear



Lab 13: Understanding the architecture of a pre-trained model (VGG16)

```
import tensorflow as tf
from tensorflow.keras.applications import VGG16

# Load pre-trained VGG16 model with ImageNet weights
model = VGG16(weights="imagenet", include_top=True)

# Display model summary
model.summary()

# Plot model architecture
tf.keras.utils.plot_model(model, to_file="vgg16_architecture.png", show_shapes=True, show_layer_names=True)

print("\n✅ Model architecture plot saved as 'vgg16_architecture.png'")
```



Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467096/553467096 ————— 3s 0us/step
Model: "vgg16"

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467096/553467096 ————— 3s 0us/step
Model: "vgg16"

Layer (type)	Output Shape	Param #
input_layer_13 (InputLayer)	(None, 224, 224, 3)	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1,792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36,928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73,856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147,584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295,168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590,080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590,080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1,180,160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
flatten (Flatten)	(None, 25088)	0
fc1 (Dense)	(None, 4096)	102,764,544
fc2 (Dense)	(None, 4096)	16,781,312
predictions (Dense)	(None, 1000)	4,097,000

Total params: 138,357,544 (527.79 MB)

Trainable params: 138,357,544 (527.79 MB)

Non-trainable params: 0 (0.00 B)

Model architecture plot saved as 'vgg16_architecture.png'