



15/17

Keerthi Vasanth A

212222240048

6B) In a practical DL project, how do you determine the most suitable function for different layers in a neural network and why? Additionally, explain these activation functions and provide PyTorch implementations?

7A) Explain any 3 CNN architecture, discussing their key components, advantages and applications.

8B) VGG, CIFAR-10, PyTorch.

i) Load the CIFAR dataset.

ii) Use a pre-trained VGG model as the base model.

iii) Create a Sequential model with the appropriate number of neurons in the output layer, activation function, and loss function.

iv) Train the model with training data and validation data.



## 6B) Activation Functions:-

\* Activation function is a function that is implemented in the hidden layers of the neuron that is activated when the signals are transmitted from the input layer.

\* These activation functions improve the effect of each input signal so that the loss function is reduced and the accuracy is improved.

\* There are different types of activation function which are used based on the requirements, they are:

\* Sigmoid.

\* Tanh

\* ReLU

\* Leaky ReLU

\* PReLU

\* ELU

\* Softmax.

### Sigmoid:-

\* Sigmoid activation function is used for the Binary Classification process.

\* The range of the output will be





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(2.0, 1.0, 0.5) and ) using an = value  
ranging from 0 to 1

Implementation:

```
sigmoid = nn.Sigmoid(torch.tensor([2.0, 1.0, 0.5]))
output = sigmoid.model().item()
```

Tanh:

\* Tanh is also called as Hyperbolic (Tangent) activation function, which ranges from -1 to 1.

Implementation:

```
tanh = nn.Tanh(torch.tensor([2.0, 1.0, 0.5]))
output = tanh.model().item()
```

ReLU:

ReLU (Rectified Linear Unit). ReLU focuses on zero-centered outputs.

\* ReLU activation function ranges from 0 to infinity as it adds the signals



as list:

```
relu = nn.ReLU(torch(2.0, 1, 0.5))
```

```
output = relu.model()
```

### Leaky ReLU:-

\* Leaky ReLU or Leaking ReLU can be used to avoid the dying neuron problem and fixing it.

### Implementation:

```
lnelu = nn.LReLU(torch(2.0, 1, 0.5))
```

```
output = lnelu.model()
```

### PReLU:

\* PReLU is the short form of Parametric Rectified Linear Unit.

\* PReLU is a very good activation function to avoid the vanishing gradient problem.

```
lnprelu = nn.PReLU(torch(2.0, 1, 0.5))
```

```
output = lnprelu.model()
```





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ELU:

\* ELU is the shortest form of Exponential Linear Unit.

\* The ELU activation function performs well even in initial conditions like vanishing gradient problem and dying neurons.

$$elu = nn.ELU(torch(0, 1, 0.5))$$

output = elu model().

Softmax:

\* Softmax is one among the well known activation function and mostly implemented for smaller threads.

\* Softmax is also good in the classification process and can expect accurate results.



## 7A) CNN Architectures:

\* CNN stands for Convolutional Neural Networks which has convolutional layers with different ranges is present.

\* There are many types of CNN architectures, such as:

\* LeNet-98

\* AlexNet

\* ResNet

### 1) LeNet:

\* The LeNet architecture is the oldest architecture of the convolutional Neural Network.

\* It was invented in the year of 1998.

\* LeNet has 2 convolutional layers and 2 fully connected layers.

\* LeNet uses softmax activation function and can classify zero to nine input signals for a single loop.

\* The inputs are  $32 \times 32$  grayscale images.





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### Advantages of LeNet:

\* LeNet architecture and perform well for small number of inputs and the accuracy will be high.

\* It was the fastest known architecture during its time.

### Applications:

\* It is applied in character recognition from various input files like images and PDFs.

\* OCR (Optical Character Recognition) works using LeNet architecture.



## ii) AlexNet:

\* AlexNet is invented in the year of 2012.

\* AlexNet architecture can have five convolutional layers and three fully connected layers.

\* The input is a  $227 \times 227$  gray scale image.

\* AlexNet uses ReLU for processing and Softmax for output generation.

\* It can classify around 1000 images for a single loop.

### Advantages:

\* It can be used in other image classification method, for higher pixel images.

\* Faster and easy implementation can be seen.





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Application:

\* Alex Net is implemented in the ImageNet.

\* It can be also used for digital image processing and medical images.

iii) ResNet:

\* The ResNet architecture is invented by He et al. in the year of 2015.

\* The input image ranges from 30 to 500 and 500 to 1000.

\* ResNet uses softmax activation function.

\* ResNet consists of convolutional network and can have any number of fully connected layers based on the requirements.



## Advantages and Implementation:

\* ResNet gives faster results and the desired results are very accurate.

\* It is implemented in Medical image processing and Facial recognition.

8b) i) Loading CIFAR-10 Dataset :-

```
import torch
from torch import torchvision, nn, optim
```

```
torch = torchvision (torchvision, Resize (224),
                    torchvision, Replace = True).
```

```
train = torch.dataset.CIFAR10 (train = True, download
                                = True, shuffle = True).
```

```
train_loader = torchvision.DataLoader (train).
```

```
test = torch.dataset.CIFAR10 (train = False,
                                download = True, shuffle = True).
```

```
test_loader = torchvision.DataLoader (test).
```





ii) Using vgg model as "base" model:-

from torchvision import models

vgg = models.vgg11()

for vgg in parameters():

vgg.requires\_grad\_()

iii) Creating Sequential Model:-

criterion = torch.nn.CrossEntropyLoss()

iv) Training

epochs = 5

for vgg in range():

running\_loss = 0.0

for param in parameters:

loss = criterion.loss()



running\_loss += loss

print("Loss = " + running\_loss / len(train))

def train\_model(model, train\_loader, dev\_loader):

# Train model

for epoch in range(1, num\_epochs):

# Train model

# Evaluate model

# Print loss

return model

def main():

# Train model

# Evaluate model

# Print loss

# End of program