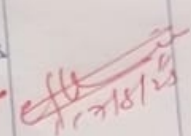

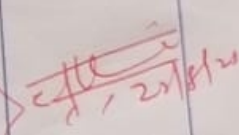
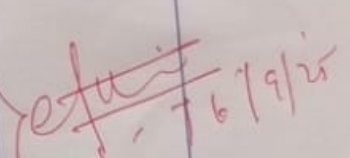
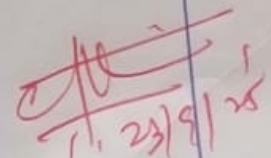


RA231047010016			
S.No.	Date	Title	Signature
1	07.08.2025	Exploring the Deep learning platforms	 effe 12/8/25
2	07.08.2025	Implement a classifier using open source dataset	
3	14.08.2025	Study of classifier with respect to Statistical parameters	 effe 11/4/25
4	22.08.2025	To build and train a simple feed forward Network (FFNN) on MNIST dataset.	 effe 22/8/25
5	22.8.25	Study different activation function used in NN	 effe 76/9/25
6	9.9.25	Implement Gradient Descent & Backpropagation in NN	
7	16.09.2025	Build a CNN model to classify Cat and dog image	 effe 11/23/9/25

Lab-5

Aim: To study different activation functions used in neural networks.

Objective:

→ To explore commonly used activation functions -

→ To analyse their mathematical behaviour and impact on learning.

→ To understand the importance of non-linearity in deep neural networks.

Pseudocode:

→ Define different activation functions: sigmoid, tanh, ReLU, Leaky ReLU, Softmax.

→ Replace activation layers with the current functions.

→ Train the model on a dataset (e.g., MNIST)

→ Record training loss and test accuracy.

→ Compare results across activation functions.

Observation:

Sen

→ Sigmoid: Can cause vanishing gradients; Slow training

→ Tanh: Zero centered; better than sigmoid but still prone to vanishing gradients.

→ ReLU: Fast training, mitigates vanishing gradient; may cause "dead neurons".

→ Leaky ReLU: Fixes dead neuron problem by allowing small gradient when inactive

← → ↻ Not secure 10.1.38.19/user/ra2311047010016/lab/workspaces/auto-K/tree/LAB5.ipynb ☆ 📁 🔒 ⋮

File Edit View Run Kernel Tabs Settings Help

Filter files by name 🔍

Name	Last Modified
normal...	5 months ago
impleme...	6 months ago
impleme...	6 months ago
insertion.c	7 months ago
knapsack...	7 months ago
LAB2.ipynb	8 days ago
LAB3.ipynb	8 days ago
LAB4.ipynb	8 days ago
LAB5.ipynb	8 days ago
LAB6.ipynb	17 minutes ago
largestco...	last year
linear.c	6 months ago
maxmini...	6 months ago
mergeso...	6 months ago
mnist_cla...	last month
multithre...	6 months ago
nqueen.c	last year
nqueens...	7 months ago

Launcher LAB2.ipynb LAB3.ipynb LAB4.ipynb LAB5.ipynb LAB6.ipynb

Notebook Python 3 (ipykernel)

```
[1]: import tensorflow as tf
from tensorflow.keras import datasets, models, layers
import matplotlib.pyplot as plt

# Load MNIST
(x_train, y_train), (x_test, y_test) = datasets.mnist.load_data()
x_train = x_train.reshape(-1, 784).astype("float32") / 255.0
x_test = x_test.reshape(-1, 784).astype("float32") / 255.0

# Function to build model with different activation
def build_model(activation):
    model = models.Sequential([
        layers.Input(shape=(784,)),
        layers.Dense(128, activation=activation),
        layers.Dense(64, activation=activation),
        layers.Dense(10, activation="softmax")
    ])
    model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
    return model

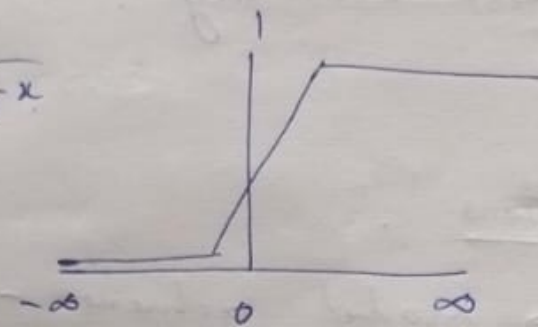
# Test different activations
activations = ["sigmoid", "tanh", "relu", "elu"]
results = {}

for act in activations:
    print(f"\n♦ Training with {act} activation")
    model = build_model(act)
    history = model.fit(x_train, y_train, validation_split=0.1,
```

x	Sigmoid	tanh	ReLU	LIR
5	0.9933	0.999	5	5
7	0.999	0.999	7	7
-1	0.268	-0.761	0	0

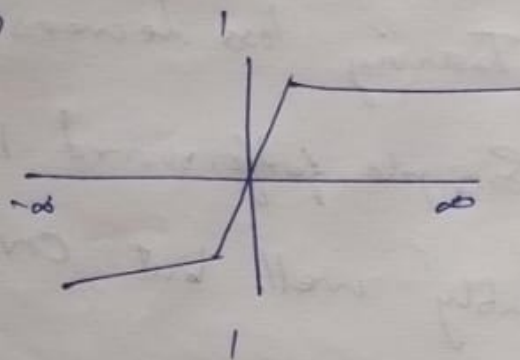
1.) Sigmoid

$$\sigma(x) = \frac{1}{1+e^{-x}}$$



2. tanh

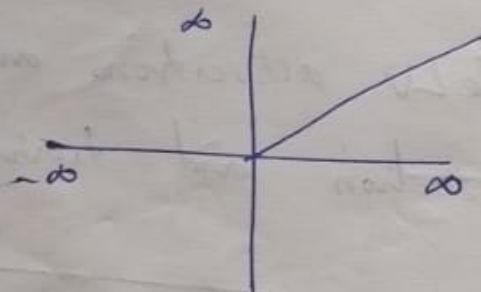
tanh



5. Maxout

$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

3.) ReLU

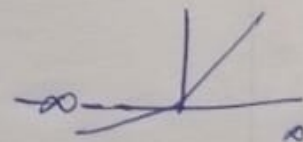
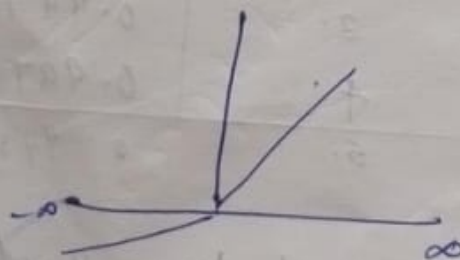


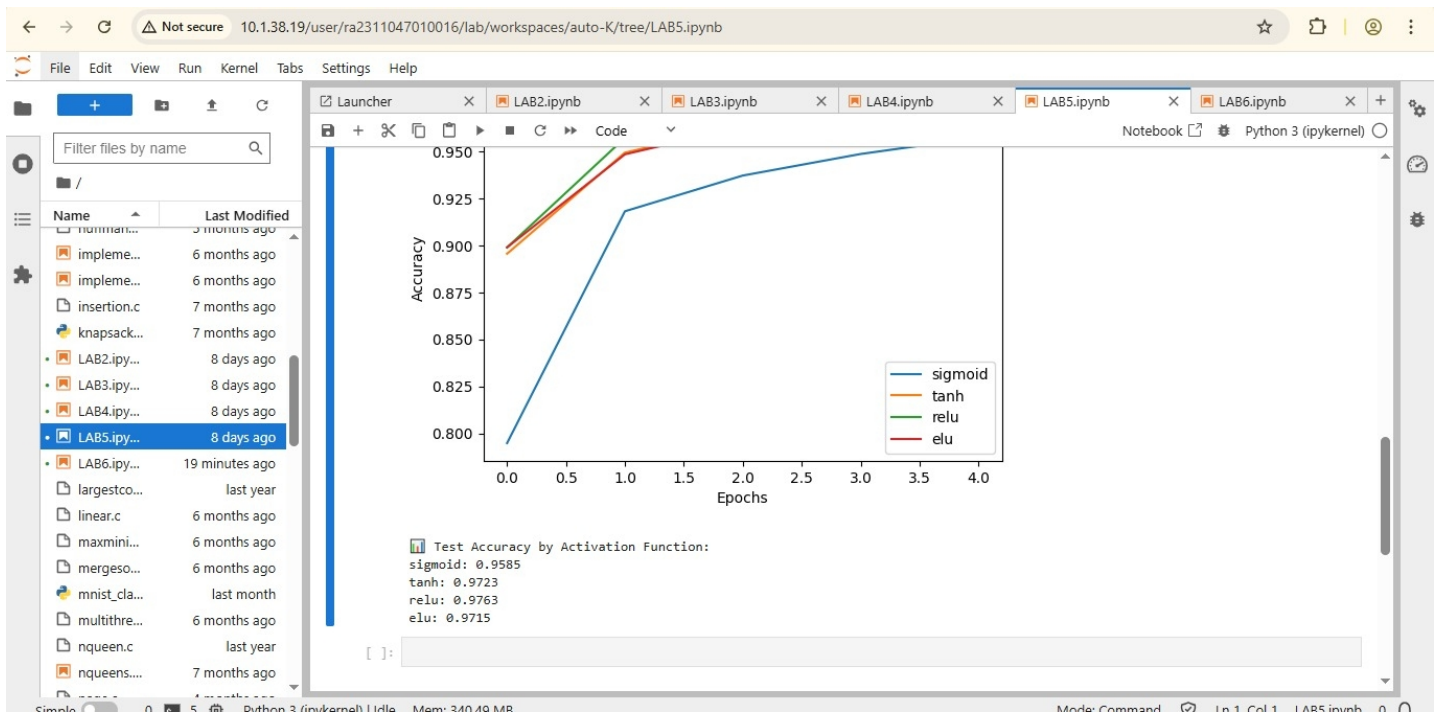
6. ELU

$$\begin{cases} x & x \geq 0 \\ \alpha (e^x - 1) & x < 0 \end{cases}$$

4.) leaky ReLU

$$\max(0, \alpha x, x)$$





→ Activation choice significantly affects
convergent speed and final accuracy.

Result:

Successfully completed the
different activation functions
used in neural networks.

