

S.No.	Date	Title	Signature
1.	07.08.2025	Exploring the Deep learning platforms	efft, 7/8/25
2.	07.08.2025	Implement a classifier using open source dataset	efft, 7/8/25
3.	14.08.2025	Study of classifiers with respect to Statistical parameters	efft, 14/8/25
4.	22.08.2025	To build and train a simple feed forward Network (FFNN) on MNIST dataset	efft, 22/8/25
5.	22.08.25	Study different activation function used in NN	efft, 22/8/25
6.	9.09.25	Implement Gradient Descent & Backpropagation in DNN	efft, 9/9/25
7.	16.09.2025	Build a CNN model to classify Cat and dog image	efft, 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, torch, ReLU, heavy KELU, 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:

Gen

→ 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

The screenshot shows a Jupyter Notebook interface with the following details:

- Header:** The address bar shows the URL: `Not secure 10.1.38.19/user/ra2311047010016/lab/workspaces/auto-K/tree/LAB5.ipynb`.
- File Menu:** File, Edit, View, Run, Kernel, Tabs, Settings, Help.
- Toolbar:** Includes icons for New, Open, Save, Run Cell, Stop, Kernel, Help, and a plus sign for creating new cells.
- File Browser:** On the left, it lists files and folders in the current directory. The files listed are:
 - implement.c (6 months ago)
 - implement.h (6 months ago)
 - insertion.c (7 months ago)
 - knapsack.c (7 months ago)
 - LAB2.ipynb (8 days ago)
 - LAB3.ipynb (8 days ago)
 - LAB4.ipynb (8 days ago)
 - LAB5.ipynb (8 days ago)** - This file is currently selected.
 - 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)
- Notebook Tab:** Shows tabs for various notebooks: Launcher, LAB2.ipynb, LAB3.ipynb, LAB4.ipynb, LAB5.ipynb (selected), and LAB6.ipynb.
- Code Cell:** The main area displays a Python code cell (cell 1) with the following content:

```
[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

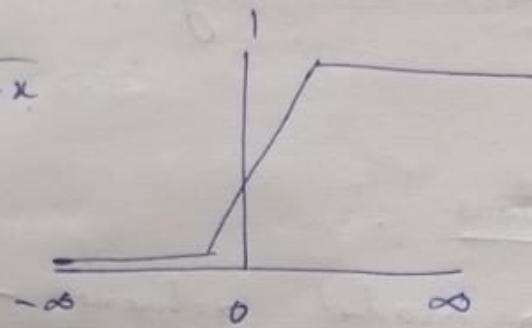
# Test different activations
activations = ["sigmoid", "tanh", "relu", "elu"]
results = {}

for act in activations:
    print(f"\n◆ Training with {act} activation")
    model = build_model(act)
    history = model.fit(x_train, y_train, validation_split=0.2)
```

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

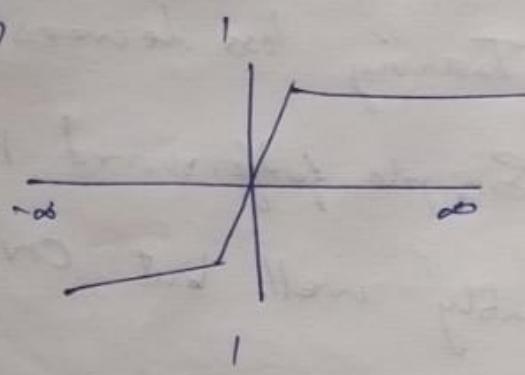
1.) Sigmoid

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



2. tanh

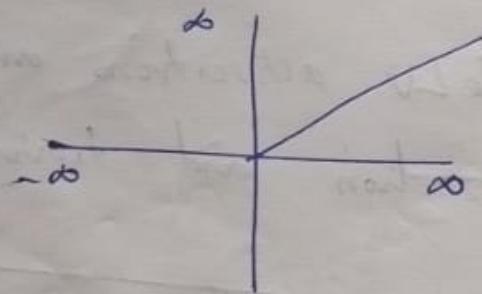
tanh



5. Maxout

$$\max(\omega_1^T x + b_1, \omega_2^T x)$$

3.) ReLU

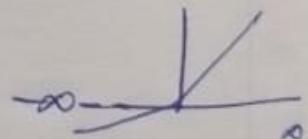
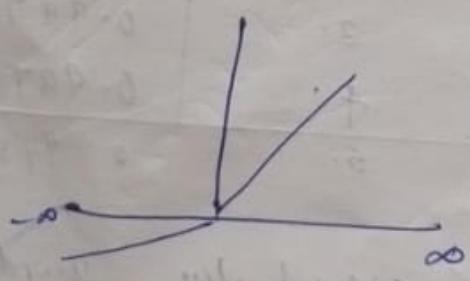


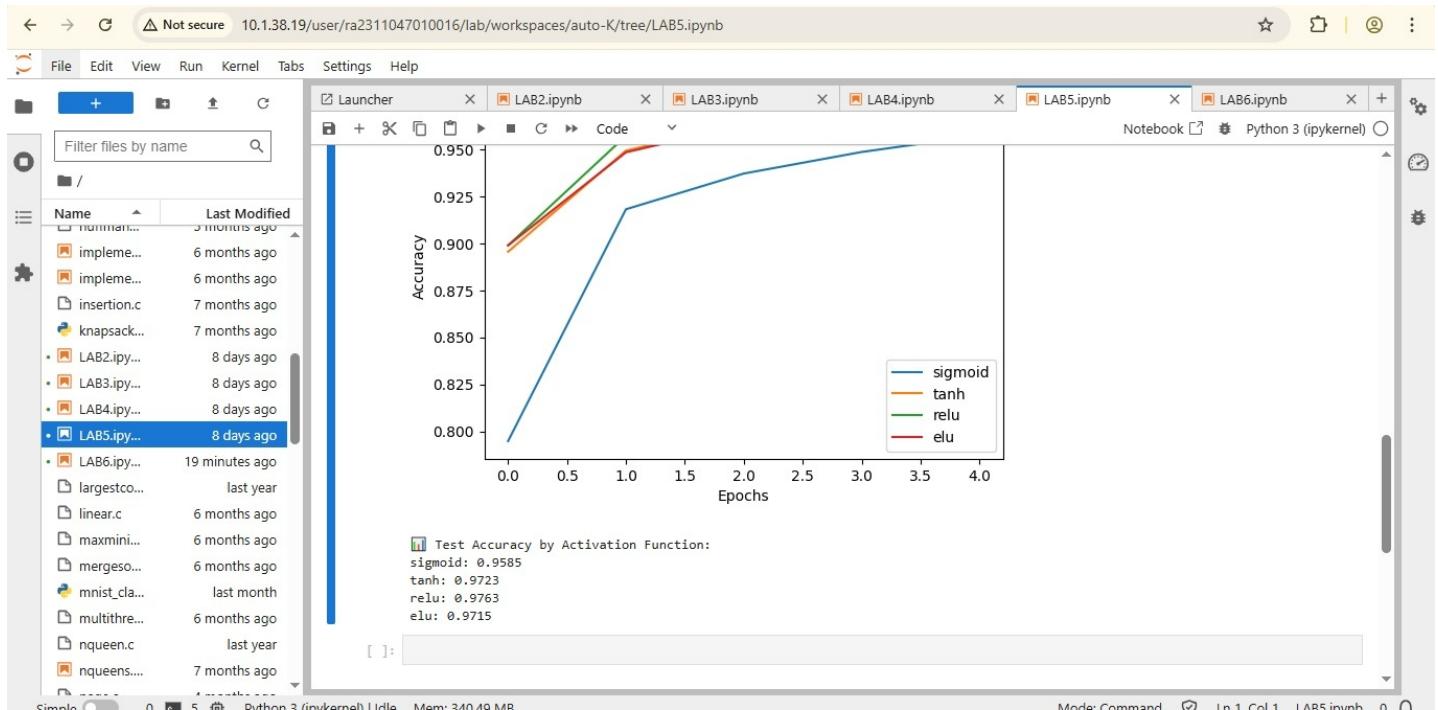
6. ELU

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

4.) leaky ReLU

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





→ Activation choice significantly affects
convergent speed and final accuracy.

Result:

successfully completed the
different activation functions
used in neural networks.

Not secure 10.1.38.19/user/ra2311047010016/lab/workspaces/auto-K/tree/LAB5.ipynb

File Edit View Run Kernel Tabs Settings Help

Launcher LAB2.ipynb LAB3.ipynb LAB4.ipynb LAB5.ipynb LAB6.ipynb Notebook Python 3 (ipykernel)

Filter files by name

Name	Last Modified
nummar...	3 months ago
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LAB2.ipynb	8 days ago
LAB3.ipynb	8 days ago
LAB4.ipynb	8 days ago
LAB5.ipynb	8 days ago
LAB6.ipynb	18 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

```
model = build_model(act)
history = model.fit(x_train, y_train, validation_split=0.1,
                     epochs=5, batch_size=128, verbose=0)

test_loss, test_acc = model.evaluate(x_test, y_test, verbose=0)
results[act] = test_acc

# Plot training accuracy
plt.plot(history.history["accuracy"], label=f"(act)")

plt.title("Training Accuracy per Activation Function")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.show()

print("\nTest Accuracy by Activation Function:")
for act, acc in results.items():
    print(f"(act): {acc:.4f}")

2025-09-01 08:46:11.511532: I tensorflow/core/platform/cpu_feature_guard.cc:210] This TensorFlow binary is optimized to use available CPU instructions in performance-critical operations.
To enable the following instructions: AVX2 FMA, in other operations, rebuild TensorFlow with the appropriate compiler flags.

◆ Training with sigmoid activation
2025-09-01 08:46:18.028356: E external/local_xla/xla/stream_executor/cuda/cuda_platform.cc:51] failed call to cuInit: INTERNAL: CUDA error: Failed call to cuInit: UNKNOWN ERROR (303)
```