

EDGE INTELLIGENCE

LAB - 4

25MML0038

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TASK 1: ARTIFICIAL NEURAL NETWORK USING DL

```
import tensorflow as tf
from tensorflow import keras
import numpy as np
import pickle

# 1. Load Data
# Ensure the path is correct for your local machine
KAGGLE_FILE_PATH = r"C:\Users\batch1\Downloads\archive (3)\mnist.npz"
with np.load(KAGGLE_FILE_PATH) as f:
    x_train, y_train = f['x_train'], f['y_train']
    x_test, y_test = f['x_test'], f['y_test']

# 2. Preprocess
x_train = x_train / 255.0
x_test = x_test / 255.0

# 3. Build Model
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(128, activation='relu'),
    keras.layers.Dense(10, activation='softmax')
])

# 4. Compile Model
model.compile(optimizer='sgd',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

# 5. Train Model
print("Starting training...")
model.fit(x_train, y_train, epochs=2)
```

[6]

```

# 5. Train Model
print("Starting training...")
model.fit(x_train, y_train, epochs=2)

# 6. Evaluate
print("\nStarting evaluation...")
loss, acc = model.evaluate(x_test, y_test, verbose=0)
print(f"Accuracy: {acc*100:.2f}%")

# 7. Save Model as Pickle
# We open the file in 'wb' (write binary) mode
pickle_filename = "mnist_model.pkl"
print(f"\nSaving model as {pickle_filename}...")
with open(pickle_filename, "wb") as f:
    pickle.dump(model, f)

# 8. Demonstrate Loading the Pickled Model
print("Loading model back from pickle...")
with open(pickle_filename, "rb") as f:
    loaded_model = pickle.load(f)

# 9. Verify Loaded Model
print("\nMaking predictions with loaded model for first 5 images:")
pred = loaded_model.predict(x_test[:5])
predicted_digits = np.argmax(pred, axis=1)
print("Predicted digits:", predicted_digits)
print("True labels:", y_test[:5])

```

```

c:\Users\hsuhh\AppData\Local\Programs\Python\Python312\Lib\site-packages\keras\src\layers\reshaping\flatten.py:37: UserWarning: Do not pass an 'input_shape'/'input_dim' argument to a 1
super().__init__(**kwargs)
Starting training...
Epoch 1/2
1875/1875 — 5s 2ms/step - accuracy: 0.8419 - loss: 0.6324
Epoch 2/2
1875/1875 — 5s 2ms/step - accuracy: 0.9078 - loss: 0.3312

Starting evaluation...
Accuracy: 92.08%

Saving model as mnist_model.pkl...
Loading model back from pickle...

Making predictions with loaded model for first 5 images:
1/1 — 0s 85ms/step
Predicted digits: [7 2 1 0 4]
True labels: [7 2 1 0 4]

import os
print(f"The model is saved at: {os.path.join(os.getcwd(), 'mnist_model.pkl')}")

```

The model is saved at: c:\Users\hsuhh\Downloads\mnist_model.pkl

```

import tensorflow as tf
import numpy as np
import pickle

# 1. Load the model from the pickle file
pickle_filename = "mnist_model.pkl"
print(f"Loading model from {pickle_filename}...")

with open(pickle_filename, "rb") as f:
    loaded_model = pickle.load(f)

# 2. Prepare test data (ensure it matches training format)
# Note: In a real scenario, you would load your actual new data here
mnist = tf.keras.datasets.mnist
(_, _), (x_test, y_test) = mnist.load_data()
x_test = x_test / 255.0 # Crucial: Must normalize like training!

# 3. Test the model's accuracy
print("\nEvaluating loaded model performance...")
loss, acc = loaded_model.evaluate(x_test, y_test, verbose=0)
print(f"Restored model accuracy: {acc*100:.2f}%")

# 4. Make individual predictions
print("\nTesting on first 5 images...")
predictions = loaded_model.predict(x_test[:5])
predicted_classes = np.argmax(predictions, axis=1)

print(f"Predicted: {predicted_classes}")
print(f"Actual: {y_test[:5]}")

```

```

... Loading model from mnist_model.pkl...
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 ————— 2s 0us/step

Evaluating loaded model performance...
Restored model accuracy: 92.08%

Testing on first 5 images...
1/1 ————— 0s 82ms/step
Predicted: [7 2 1 0 4]
Actual: [7 2 1 0 4]

```

```

import os

file_path = "mnist_model.pkl"
# getsize returns bytes, divide by 1024 for KB
size_kb = os.path.getsize(file_path) / 1024

print(f"File Size: {size_kb:.2f} KB")

```

[4] ✓ 0.0s

... File Size: 417.60 KB

```
model.summary()
```

[5] ✓ 0.0s

Model: "sequential"

Layer (type)	Output Shape	Param #
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 128)	100,480
dense_1 (Dense)	(None, 10)	1,290

Total params: 101,772 (397.55 KB)

Trainable params: 101,770 (397.54 KB)

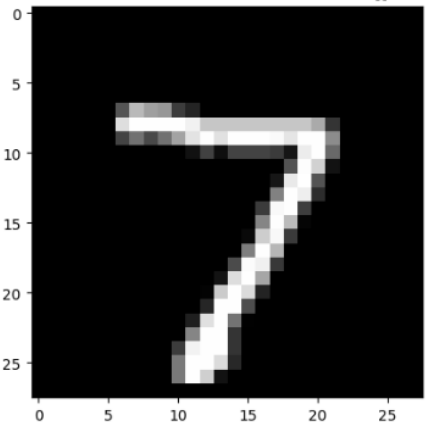
Non-trainable params: 0 (0.00 B)

Optimizer params: 2 (12.00 B)

```
import matplotlib.pyplot as plt
plt.imshow(x_test[0], cmap='gray')
plt.title("predicted:" + str(pred))
```

✓ 0.5s

Text(0.5, 1.0, 'predicted:[[2.75356637e-04 1.69666248e-06 2.37161395e-04 1.73411972e-03\n 2.40992995e-05 8.68187635e-05 9.65907248e-07 9.94599342e-01\n 1.27018560e-04 2.91345967e-03]\n [3.54252290e-03 1.65653692e-04 8.88520598e-01 2.14903187e-02\n 3.64339212e-06 1.29046664e-02 6.48955032e-02 1.60882678e-07\n 8.47490225e-03 2.03188733e-06]\n [1.30446148e-04 9.68027294e-01 1.10185165e-02 3.98947159e-03\n 7.58480746e-04 1.76262611e-03 3.21953674e-03 3.53753963e-03\n 6.44446723e-03 1.11162977e-03]\n [9.98167276e-01 4.62121896e-09 1.48636289e-04 4.60250521e-05\n 2.46783685e-07 1.08933903e-03 3.38261423e-04 6.30978029e-05\n 9.91277120e-05 4.79449554e-05]\n [1.96933770e-03 2.54092865e-05 9.73366201e-03 5.40680077e-04\n 8.95198882e-01 1.14605890e-03 5.85654425e-03 7.27365073e-03\n 9.97465756e-03 6.82810321e-02]]')



TASK 2: EDGE IMPLUSE

STEPS:

1. Create an account on **Edge Impulse**.
2. Login to the Edge Impulse platform.
3. Create a **new project**.
4. Click **Data Acquisition**.
5. Select **Connect to device**.
6. Choose **Use mobile phone** and scan the QR code.
7. Connect the mobile phone as a camera.
8. Capture sample images.
9. Collect images for **training** and **testing**.
10. Rename images with meaningful names.
11. Label all images correctly.

Training -8

The screenshot displays the Edge Impulse web interface. The top navigation bar includes tabs for Dataset, Data sources, Synthetic data, Labeling queue (0), and AI labeling (NEW). The left sidebar contains a menu with options: Dashboard, Devices, Data acquisition, Experiments, EON Tuner, Impulse design (selected), Live classification, Model testing, and Post-processing. Below the menu is an 'Upgrade Plan' section.

The main content area shows the 'Dataset' tab with a table of 14 items. The table has columns for SAMPLE NAME, LABELS, and ADDED. The items are categorized into Training (8), Test (4), and Post-processing (2). The training progress is shown as 57% / 43%.

SAMPLE NAME	LABELS	ADDED
mouse4	MOUSE	Today, 17:15:08
mouse4	MOUSE	Today, 17:12:58
mouse2	MOUSE	Today, 17:11:04
mouse1	MOUSE	Today, 17:10:55
fansample5	FAN	Today, 17:10:37
fansample4	FAN	Today, 17:09:54
fansample2	FAN	Today, 17:07:06
Fan	FAN	Today, 17:03:57

On the right side, there is a 'Collect data' section with a button to 'Connect a device' and a 'Click on a sample to load...' button. A 'Resume tutorial' button is located at the bottom right.

Testing-6

The screenshot displays the Edge Impulse web interface for a project named "subhashin_m / subhashin_m-project-1". The interface is divided into a left sidebar, a top navigation bar, and a main content area.

Left Sidebar: Contains navigation links for Dashboard, Devices, Data acquisition, Experiments, EON Tuner, Impulse design, Create impulse, Live classification, Model testing, and Post-processing. An "Upgrade Plan" button is also present.

Top Navigation Bar: Includes tabs for Dataset, Data sources, Synthetic data, Labeling queue (0), and AI labeling (NEW). The user's target device is listed as "Target: Cortex-M4F 80MHz".

Main Content Area:

- Dataset Overview:** Shows "DATA COLLECTED 14 items" and "TRAIN / TEST SPLIT 57% / 43%".
- Dataset Table:** A table with columns "SAMPLE NAME", "LABELS", and "ADDED". It lists 6 samples with labels "mouse", "fan", and "MOUSE".
- Collect data section:** Includes a "Connect a device" button and a "Click on a sample to load..." instruction.
- Bottom Right:** A "Resume tutorial" button.

SAMPLE NAME	LABELS	ADDED
unknown.6ebbvlcu	mouse	Today, 17:24:46
unknown.6ebbud6q	fan	Today, 17:24:05
unknown.6ebbs56o	fan	Today, 17:22:51
unknown.6ebbbb29	MOUSE	Today, 17:13:40
mouse3	MOUSE	Today, 17:25:33
unknown.6ebb74ql	MOUSE	Today, 17:11:23