

## **LAPORAN AKHIR PRAKTIKUM**

Mata Praktikum : Kecerdasan Buatan  
Kelas : 3IA13  
Praktikum ke- : 1  
Tanggal : 2 Desember 2022  
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**LABORATORIUM TEKNIK INFORMATIKA**

**UNIVERSITAS GUNADARMA**

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## Isi Laporan

1. Blok kode yang digunakan untuk mengimport library yang dibutuhkan

```
In [*]: import numpy as np #komputasi numerik
import matplotlib.pyplot as plt #visualisasi data
import tensorflow as tf #backend
from tensorflow.keras.datasets import mnist #dataset
from sklearn.metrics import confusion_matrix #evaluasi
import itertools #iterator untuk iterasi data
```

2. Load dataset mnist dari tensorflow keras.

```
In [2]: (x_train, y_train), (x_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 18s 2us/step
```

3. Print dimensi data dan array `x_train[3]`

[illegible]

4. Reshape x\_train, ini hanya mengubah bentuk tanpa mengubah data

```
In [7]: x_train=np.reshape(x_train,(x_train.shape[0], x_train.shape[1] * x_train.shape[2]))
x_test=np.reshape(x_test, (x_test.shape[0], 784))
```

5. Print data yang sudah di reshape

[illegible]

## 6. Normalisasi tingkat kecerahan

```
#normalisasi tingkat kecerahan . Diubah dari 0-255 menjadi 0-1
x_train = x_train/255.0
x_test = x_test/255.0
```

7. Array `x_train[0]`

[illegible]

8. Membuat label dataset menjadi kategori atau kelas-kelas dengan menggunakan tensorflow

```
In [ ]: from tensorflow.keras.utils import to_categorical
y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
```

```
In [8]: y_train
```

```
Out[8]: array([5, 0, 4, ..., 5, 6, 8], dtype=uint8)
```

```
In [9]: y_train
```

```
Out[9]: array([[0., 0., 0., ..., 0., 0., 0.],
               [1., 0., 0., ..., 0., 0., 0.],
               [0., 0., 0., ..., 0., 0., 0.],
               ...,
               [0., 0., 0., ..., 0., 0., 0.],
               [0., 0., 0., ..., 0., 0., 0.],
               [0., 0., 0., ..., 0., 1., 0.]], dtype=float32)
```

## 9. Membuat model machine learning

```
In [ ]: num_input = 28 * 28 #jumlah input vektor

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(500, input_dim = num_input, activation='relu'),
    tf.keras.layers.Dense(10, activation='softmax')
])

model.compile(loss='categorical_crossentropy', #mengukur informasi yang hilang
              optimizer='adam', #pengaturan Learning rate
              metrics=['accuracy'])
```

## 10. Training Data (semakin banyak loop akurasi semakin bagus)

```
In [*]: hist = model.fit(x_train, y_train,
                        epochs = 15,
                        batch_size = 200,
                        validation_data = (x_test, y_test))

Epoch 1/15
300/300 [=====] - 4s 9ms/step - loss: 6.1705 - accuracy: 0.8959 - val_loss: 1.5220 - val_accuracy: 0.9
427
Epoch 2/15
300/300 [=====] - 2s 8ms/step - loss: 0.8875 - accuracy: 0.9553 - val_loss: 0.8517 - val_accuracy: 0.9
496
Epoch 3/15
300/300 [=====] - 2s 8ms/step - loss: 0.4660 - accuracy: 0.9680 - val_loss: 0.6045 - val_accuracy: 0.9
641
Epoch 4/15
300/300 [=====] - 2s 8ms/step - loss: 0.2624 - accuracy: 0.9770 - val_loss: 0.7119 - val_accuracy: 0.9
610
Epoch 5/15
300/300 [=====] - 3s 9ms/step - loss: 0.1872 - accuracy: 0.9817 - val_loss: 0.6227 - val_accuracy: 0.9
623
Epoch 6/15
300/300 [=====] - 3s 9ms/step - loss: 0.1764 - accuracy: 0.9817 - val_loss: 0.6220 - val_accuracy: 0.9
655
Epoch 7/15
98/300 [=====>.....] - ETA: 1s - loss: 0.1369 - accuracy: 0.9855
```

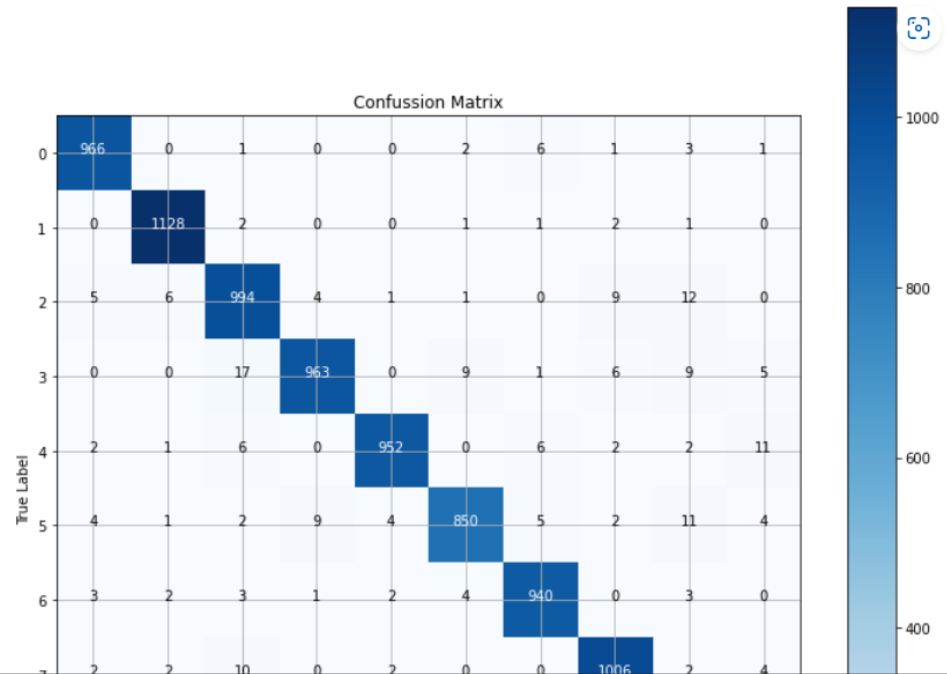
## 11. Menyimpan model dan visualisasi

```
In [13]: model.save_weights('model_weights.h5') #menyimpan model

In [14]: from prompt_toolkit.layout.containers import HorizontalAlign
def plot_confusion_matrix(cm, classes, normalize = False,
                        title = 'Confusion Matrix',
                        cmap=plt.cm.Blues):
    plt.figure(figsize = (10,10))
    plt.imshow(cm, interpolation = 'nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation = 90)
    plt.yticks(tick_marks, classes)
    if normalize:
        cm = cm.astype('float')/cm.sum(axis=1)[:, np.newaxis]
    thresh = cm.max() / 2
    for i,j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j,i, cm[i,j],
                horizontalalignment = "center",
                color = "white" if cm[i,j] > thresh else "black")
    plt.grid('off')
    plt.tight_layout()
    plt.ylabel('True Label')
    plt.xlabel('Predicted Label')
```

```
In [15]: list_classes = [0,1,2,3,4,5,6,7,8,9]
Y_pred = model.predict(x_test)
Y_pred_classes = np.argmax(Y_pred, axis=1)
Y_true = np.argmax(y_test, axis=1)
confusion_mtx= confusion_matrix(Y_true, Y_pred_classes)
plot_confusion_matrix(confusion_mtx, classes= list_classes)
```

313/313 [=====] - 1s 2ms/step



L

```
In [16]: img = x_test[7777]
predicted = model.predict(np.reshape(img,(1,784)))
predicted
```

1/1 [=====] - 0s 27ms/step

```
Out[16]: array([[0.0000000e+00, 5.5085948e-24, 0.0000000e+00, 2.3357012e-23,
0.0000000e+00, 1.0000000e+00, 0.0000000e+00, 0.0000000e+00,
0.0000000e+00, 3.0657754e-37]], dtype=float32)
```

```
In [17]: predictednumber= np.argmax(predicted)
predictednumber
```

Out[17]: 5

```
In [18]: np.argmax(y_test[7777])
```

Out[18]: 5

```
In [ ]: plt.imshow(np.reshape(x_test[7777],(28,28)))
plt.axis('off')
plt.sho
```

In [ ]: |