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In [ ]: #Name :Mane Shivraj Pandurang
        #class: B.E.A.I & D.S.
        #Roll No:37
        #Subject : Deep Learning (CL-IV)
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In [ ]: # Problem Statement : In this practical, we design and implement a CNN model for
        # dataset (e.g., medical imaging, agriculture, or CIFAR-10). The model is optimi
        # hyperparameters such as Learning rate, filter size, number of layers, optimize
        # improve performance.
```

```
In [1]: #Step 1: Import Required Libraries
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
from tensorflow.keras.optimizers import Adam, SGD, RMSprop
from tensorflow.keras.datasets import cifar10
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion_matrix
```

```
In [2]: #Step 2: Load and Preprocess the Dataset
        # Load CIFAR-10 dataset
        (x_train, y_train), (x_test, y_test) = cifar10.load_data()

        # Normalize pixel values (scale between 0 and 1)
        x_train, x_test = x_train / 255.0, x_test / 255.0

        # Convert labels to one-hot encoding
        y_train = tf.keras.utils.to_categorical(y_train, num_classes=10)
        y_test = tf.keras.utils.to_categorical(y_test, num_classes=10)
```

Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>
170498071/170498071 ————— 2s 0us/step

```
In [3]: #Step 3: Define and Optimize the CNN Model
        # Hyperparameter tuning
        learning_rate = 0.001
        dropout_rate = 0.4
        filter_size = (3,3)

        # Build CNN model
        model = Sequential([
            Conv2D(32, filter_size, activation='relu', input_shape=(32, 32, 3)),
            MaxPooling2D(2,2),
            Dropout(dropout_rate),

            Conv2D(64, filter_size, activation='relu'),
            MaxPooling2D(2,2),
            Dropout(dropout_rate),

            Flatten(),
            Dense(128, activation='relu'),
            Dropout(dropout_rate),
            Dense(10, activation='softmax') # 10 classes (CIFAR-10)
        ])

        # Compile the model using Adam optimizer
```

```
optimizer = Adam(learning_rate=learning_rate)
model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=['ac
```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

In [4]: *#Step 4: Train the Model*

```
history = model.fit(x_train, y_train, epochs=10, batch_size=64, validation_data=
```

Epoch 1/10

782/782 ————— 77s 95ms/step - accuracy: 0.3041 - loss: 1.8840 - val_accuracy: 0.5318 - val_loss: 1.3469

Epoch 2/10

782/782 ————— 71s 82ms/step - accuracy: 0.4990 - loss: 1.4021 - val_accuracy: 0.5816 - val_loss: 1.2178

Epoch 3/10

782/782 ————— 81s 80ms/step - accuracy: 0.5471 - loss: 1.2759 - val_accuracy: 0.6114 - val_loss: 1.1170

Epoch 4/10

782/782 ————— 87s 87ms/step - accuracy: 0.5720 - loss: 1.2032 - val_accuracy: 0.6470 - val_loss: 1.0390

Epoch 5/10

782/782 ————— 79s 83ms/step - accuracy: 0.5974 - loss: 1.1414 - val_accuracy: 0.6618 - val_loss: 1.0170

Epoch 6/10

782/782 ————— 82s 83ms/step - accuracy: 0.6044 - loss: 1.1190 - val_accuracy: 0.6472 - val_loss: 1.0089

Epoch 7/10

782/782 ————— 80s 80ms/step - accuracy: 0.6172 - loss: 1.0861 - val_accuracy: 0.6770 - val_loss: 0.9405

Epoch 8/10

782/782 ————— 84s 83ms/step - accuracy: 0.6288 - loss: 1.0567 - val_accuracy: 0.6840 - val_loss: 0.9162

Epoch 9/10

782/782 ————— 80s 80ms/step - accuracy: 0.6389 - loss: 1.0289 - val_accuracy: 0.6887 - val_loss: 0.9307

Epoch 10/10

782/782 ————— 82s 81ms/step - accuracy: 0.6413 - loss: 1.0209 - val_accuracy: 0.6920 - val_loss: 0.8945

In [5]: *#Step 5: Evaluate the Model*

```
# Evaluate on test data
```

```
test_loss, test_acc = model.evaluate(x_test, y_test)
```

```
print("Test Accuracy:", test_acc)
```

313/313 ————— 4s 13ms/step - accuracy: 0.6908 - loss: 0.8927

Test Accuracy: 0.6919999718666077

In [6]: *#Step 6: Generate Confusion Matrix*

```
# Predict classes
```

```
y_pred = np.argmax(model.predict(x_test), axis=1)
```

```
y_true = np.argmax(y_test, axis=1)
```

```
# Compute confusion matrix
```

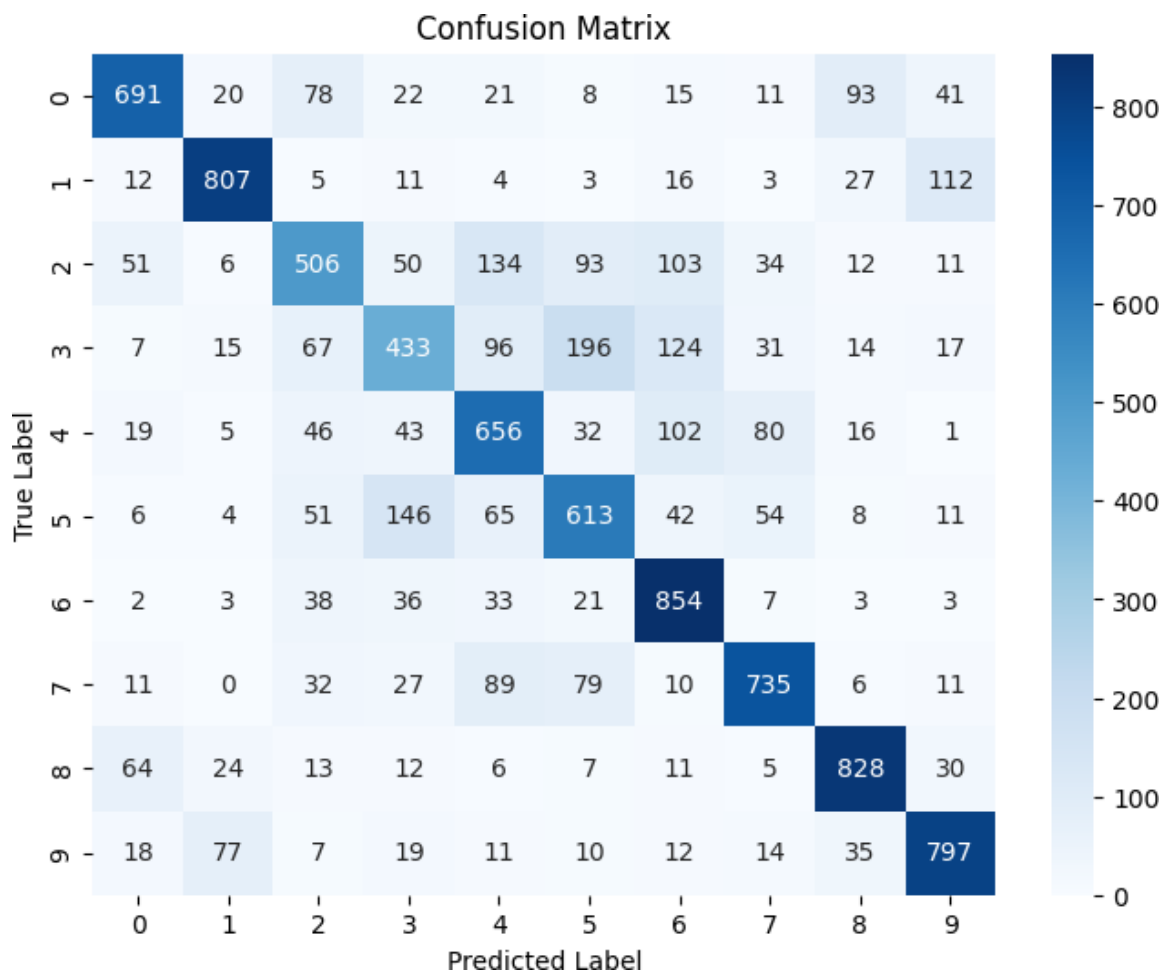
```
cm = confusion_matrix(y_true, y_pred)
```

```
# Plot confusion matrix
```

```
plt.figure(figsize=(8,6))
```

```
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d", xticklabels=range(10), ytickl
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix")
plt.show()
```

313/313 ————— 3s 10ms/step



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