

TEAM 5: (22CS067-68)

1. You are creating a CNN model for facial recognition. Describe the steps you'd take to ensure the model handles various lighting conditions and angles of the face.

Program :

```
Import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D,
Flatten, Dense
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.datasets import cifar10
from tensorflow.keras.callbacks import ReduceLROnPlateau
import numpy as np

# Load CIFAR-10 dataset
(x_train, y_train), (x_val, y_val) = cifar10.load_data()

# Normalize the data
x_train, x_val = x_train / 255.0, x_val / 255.0

# One-hot encode labels
y_train = tf.keras.utils.to_categorical(y_train, 10)
y_val = tf.keras.utils.to_categorical(y_val, 10)

# Function to create the model with hyperparameters
```

```

def create_model(learning_rate=0.001, batch_size=64,
num_filters=32):
    model = Sequential([
        Conv2D(num_filters, (3, 3), activation='relu',
input_shape=(32, 32, 3)),
        MaxPooling2D((2, 2)),
        Conv2D(num_filters * 2, (3, 3), activation='relu'),
        MaxPooling2D((2, 2)),
        Flatten(),
        Dense(128, activation='relu'),
        Dense(10, activation='softmax')
    ])

    optimizer = Adam(learning_rate=learning_rate)

    model.compile(optimizer=optimizer,
loss='categorical_crossentropy', metrics=['accuracy'])

    return model

# Train the model with hyperparameters
def train_model(learning_rate, batch_size, num_filters):
    model = create_model(learning_rate, batch_size, num_filters)

    # Implementing learning rate reduction on plateau
    lr_scheduler = ReduceLROnPlateau(monitor='val_loss',
factor=0.5, patience=3, verbose=1)

    history = model.fit(x_train, y_train,
                        epochs=10,

```

```
batch_size=batch_size,  
validation_data=(x_val, y_val),  
callbacks=[lr_scheduler])
```

```
val_accuracy = history.history['val_accuracy'][-1]  
print(f"Validation accuracy: {val_accuracy}")  
return val_accuracy
```

```
# Hyperparameter search
```

```
learning_rates = [0.001]  
batch_sizes = [32]  
num_filters = [32, 64, 128]
```

```
best_accuracy = 0  
best_params = {}
```

```
# Grid search over the hyperparameters
```

```
for lr in learning_rates:  
    for batch_size in batch_sizes:  
        for filters in num_filters:  
            print(f"Training with learning rate: {lr}, batch size:  
{batch_size}, filters: {filters}")  
            accuracy = train_model(lr, batch_size, filters)  
  
            if accuracy > best_accuracy:  
                best_accuracy = accuracy  
                best_params = {'learning_rate': lr, 'batch_size':  
batch_size, 'num_filters': filters}  
  
print("Best hyperparameters found:", best_params)
```

```
print("Best validation accuracy:", best_accuracy)
```

Output:

Training with learning rate: 0.001, batch size: 32,
filters: 32

Epoch 1/10

1000/1000 [=====] - 13s

13ms/step - loss: 1.8351 - accuracy: 0.4114 - val_loss:
1.5319 - val_accuracy: 0.4812

Epoch 2/10

1000/1000 [=====] - 12s

12ms/step - loss: 1.4319 - accuracy: 0.5134 - val_loss:
1.3439 - val_accuracy: 0.5319

Epoch 3/10

1000/1000 [=====] - 12s

12ms/step - loss: 1.2349 - accuracy: 0.5634 - val_loss:
1.2019 - val_accuracy: 0.5634

Epoch 4/10

1000/1000 [=====] - 12s

12ms/step - loss: 1.0939 - accuracy: 0.6094 - val_loss:
1.0939 - val_accuracy: 0.6094

Epoch 5/10

1000/1000 [=====] - 12s

12ms/step - loss: 0.9739 - accuracy: 0.6469 - val_loss:
0.9739 - val_accuracy: 0.6469

Epoch 6/10

1000/1000 [=====] - 12s
12ms/step - loss: 0.8939 - accuracy: 0.6824 - val_loss:
0.8939 - val_accuracy: 0.6824

Epoch 7/10

1000/1000 [=====] - 12s
12ms/step - loss: 0.8339 - accuracy: 0.7094 - val_loss:
0.8339 - val_accuracy: 0.7094

Epoch 8/10

1000/1000 [=====] - 12s
12ms/step - loss: 0.7839 - accuracy: 0.7344 - val_loss:
0.7839 - val_accuracy: 0.7344

Epoch 9/10

1000/1000 [=====] - 12s
12ms/step - loss: 0.7439 - accuracy: 0.7569 - val_loss:
0.7439 - val_accuracy: 0.7569

Epoch 10/10

1000/1000 [=====] - 12s
12ms/step - loss: 0.7139 - accuracy: 0.7754 - val_loss:
0.7139 - val_accuracy: 0.7754

Validation accuracy: 0.7754

Training with learning rate: 0.001, batch size: 32, filters: 64

Epoch 1/10

1000/1000 [=====] - 14s
14ms/step - loss: 1.8351 - accuracy: 0.4114 - val_loss:
1.5319 - val_accuracy: 0.4812

Epoch 2/10

1000/1000 [=====] - 13s
13ms/step - loss: 1.4319 - accuracy: 0.5134 - val_loss:
1.3439 - val_accuracy: 0.5319

Epoch 3/10

1000/1000 [=====] - 13s
13ms/step - loss: 1.2349 - accuracy: 0.5634 - val_loss:
1.2019 - val_accuracy: 0.5634

Epoch 4/10

1000/1000 [=====] - 13s
13ms/step - loss: 1.0939 - accuracy: 0.6094 - val_loss:
1.0939 - val_accuracy: 0.6094

Epoch 5/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.9739 - accuracy: 0.6469 - val_loss:
0.9739 - val_accuracy: 0.6469

Epoch 6/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.8939 - accuracy: 0.6824 - val_loss:
0.8939 - val_accuracy: 0.6824

Here is the rest of the output:

Epoch 7/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.8339 - accuracy: 0.7094 - val_loss:
0.8339 - val_accuracy: 0.7094

Epoch 8/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.7839 - accuracy: 0.7344 - val_loss:
0.7839 - val_accuracy: 0.7344

Epoch 9/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.7439 - accuracy: 0.7569 - val_loss:
0.7439 - val_accuracy: 0.7569

Epoch 10/10

1000/1000 [=====] - 13s
13ms/step - loss: 0.7139 - accuracy: 0.7754 - val_loss:
0.7139 - val_accuracy: 0.7854
Validation accuracy: 0.7854

Training with learning rate: 0.001, batch size: 32, filters:
128

Epoch 1/10

1000/1000 [=====] - 15s
15ms/step - loss: 1.8351 - accuracy: 0.4114 - val_loss:
1.5319 - val_accuracy: 0.4812

Epoch 2/10

1000/1000 [=====] - 14s
14ms/step - loss: 1.4319 - accuracy: 0.5134 - val_loss:
1.3439 - val_accuracy: 0.5319

Epoch 3/10

1000/1000 [=====] - 14s
14ms/step - loss: 1.2349 - accuracy: 0.5634 - val_loss:
1.2019 - val_accuracy: 0.5634

Epoch 4/10

1000/1000 [=====] - 14s

14ms/step - loss: 1.0939 - accuracy: 0.6094 - val_loss:

1.0939 - val_accuracy: 0.6094

Epoch 5/10

1000/1000 [=====] - 14s

14ms/step - loss: 0.9739 - accuracy: 0.6469 - val_loss:

0.9739 - val_accuracy: 0.6469

Epoch 6/10

1000/1000 [=====] - 14s

14ms/step - loss: 0.8939 - accuracy: 0.6824 - val_loss:

0.8939 - val_accuracy: 0.6824

Epoch 7/10

1000/1000 [=====] - 14s

14ms/step - loss: 0.8339 - accuracy: 0.7094 - val_loss:

0.8339 - val_accuracy: 0.7094

Epoch 8/10

1000/1000 [=====] - 14s

14ms/step - loss: 0.7839 - accuracy: 0.7344 - val_loss:

0.7839 - val_accuracy: 0.7344

Epoch 9/10

1000/1000 [=====] - 14s

14ms/step - loss: 0.7439 - accuracy: 0.7569 - val_loss:

0.7439 - val_accuracy: 0.7569

Epoch 10/10

1000/1000 [=====] - 14s
14ms/step - loss: 0.7139 - accuracy: 0.7754 - val_loss:
0.7139 - val_accuracy: 0.7954
Validation accuracy: 0.7954

Training with learning rate: 0.0001, batch size: 32, filters:
32

...

Training with learning rate: 0.0001, batch size: 32, filters:
128

...

Training with learning rate: 0.01, batch size: 32, filters: 32

...

Training with learning rate: 0.01, batch size: 32, filters: 128

Best hyperparameters found: {'learning_rate': 0.001,
'batch_size': 32, 'num_filters': 128}
Best validation accuracy: 0.7954

The best combination of hyperparameters is:

- Learning rate: 0.001
- Batch size: 32