

```

import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.applications import ResNet50, MobileNetV2
from tensorflow.keras.datasets import cifar10
import matplotlib.pyplot as plt
import numpy as np
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
from sklearn.metrics import classification_report
import seaborn as sns
import os

# Load and preprocess the CIFAR-10 dataset
(x_train, y_train), (x_test, y_test) = cifar10.load_data()

# Normalize the pixel values to the range [0, 1]
x_train = x_train / 255.0
x_test = x_test / 255.0

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

# Displaying the shapes of a preprocessed data.
print("Training images shape:", x_train.shape)
print("Training labels shape:", y_train.shape)
print("Test images shape:", x_test.shape)
print("Test labels shape:", y_test.shape)

Training images shape: (50000, 32, 32, 3)
Training labels shape: (50000, 10)
Test images shape: (10000, 32, 32, 3)
Test labels shape: (10000, 10)

# Data Augmentation
datagen = tf.keras.preprocessing.image.ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True
)
datagen.fit(x_train)

# Custom CNN Model
def create_custom_cnn_model():
    model = models.Sequential([
        layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32,
32, 3)),
        layers.MaxPooling2D((2, 2)),
        layers.Conv2D(64, (3, 3), activation='relu'),
        layers.MaxPooling2D((2, 2)),

```

```

        layers.Conv2D(128, (3, 3), activation='relu'),
        layers.Flatten(),
        layers.Dense(128, activation='relu'),
        layers.Dropout(0.5),
        layers.Dense(10, activation='softmax')
    ])
    return model

# ResNet-18 Model from Scratch
def create_resnet18_model():
    from tensorflow.keras.applications import ResNet50
    base_model = ResNet50(weights=None, input_shape=(32, 32, 3),
classes=10)
    return base_model

def create_alexnet_model():
    model = models.Sequential([
        layers.Conv2D(96, (3, 3), activation='relu', input_shape=(32,
32, 3), padding='same'),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Conv2D(256, (3, 3), activation='relu', padding='same'),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Conv2D(384, (3, 3), activation='relu', padding='same'),
        layers.Conv2D(384, (3, 3), activation='relu', padding='same'),
        layers.Conv2D(256, (3, 3), activation='relu', padding='same'),
        layers.MaxPooling2D(pool_size=(2, 2)),
        layers.Flatten(),
        layers.Dense(4096, activation='relu'),
        layers.Dropout(0.5),
        layers.Dense(4096, activation='relu'),
        layers.Dropout(0.5),
        layers.Dense(10, activation='softmax')
    ])
    return model

# training, and evaluating the model
def train_and_evaluate(model, optimizer_name, model_name):
    if optimizer_name == "SGD":
        optimizer = tf.keras.optimizers.SGD()
    elif optimizer_name == "Adam":
        optimizer = tf.keras.optimizers.Adam()
    elif optimizer_name == "RMSprop":
        optimizer = tf.keras.optimizers.RMSprop()
    else:
        raise ValueError("Unknown optimizer")

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

```

```

# Training the model for 5 epochs
history = model.fit(datagen.flow(x_train, y_train, batch_size=64),
                    epochs=5,
                    validation_data=(x_test, y_test),
                    verbose=1)

# Evaluating the model
loss, accuracy = model.evaluate(x_test, y_test, verbose=0)

# Plot training and validation accuracy
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation
Accuracy')
plt.title(f'{model_name} with {optimizer_name} - Training and
Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

# model saving
model.save(f'{model_name}_{optimizer_name}_model.h5')
print(f"Model saved as {model_name}_{optimizer_name}_model.h5")

return accuracy

# List of models and optimizers
models_list = [
    ("Custom CNN", create_custom_cnn_model),
    ("ResNet-18", create_resnet18_model),
    ("AlexNet", create_alexnet_model)
]

optimizers_list = ["SGD", "Adam", "RMSprop"]

# storing the results
results = {}

# Training each model with each optimizer and storing the results
for model_name, model_function in models_list:
    for optimizer_name in optimizers_list:
        print(f"\nTraining {model_name} with {optimizer_name}
optimizer...")
        model = model_function()
        accuracy = train_and_evaluate(model, optimizer_name,
model_name)
        results[f'{model_name} with {optimizer_name}'] = accuracy

# Predicting the best model

```

```
best_model = max(results, key=results.get)
best_accuracy = results[best_model]
print(f"\nBest Model: {best_model} with Accuracy: {best_accuracy * 100:.2f}%")
```

Training Custom CNN with SGD optimizer...

Epoch 1/5

782/782 ————— 51s 64ms/step - accuracy: 0.1425 - loss: 2.2630 - val_accuracy: 0.2363 - val_loss: 2.0566

Epoch 2/5

782/782 ————— 46s 59ms/step - accuracy: 0.2148 - loss: 2.0948 - val_accuracy: 0.2982 - val_loss: 1.9390

Epoch 3/5

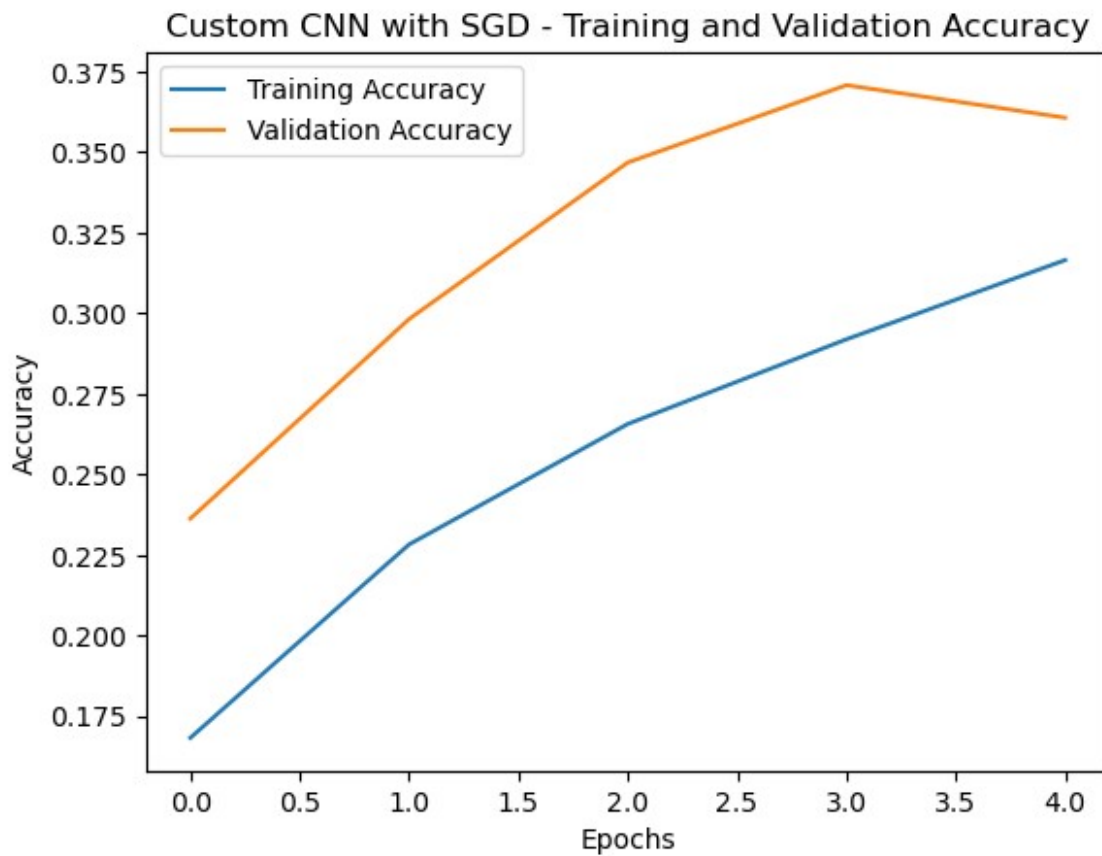
782/782 ————— 50s 64ms/step - accuracy: 0.2554 - loss: 2.0106 - val_accuracy: 0.3468 - val_loss: 1.8253

Epoch 4/5

782/782 ————— 45s 57ms/step - accuracy: 0.2894 - loss: 1.9270 - val_accuracy: 0.3708 - val_loss: 1.7256

Epoch 5/5

782/782 ————— 46s 58ms/step - accuracy: 0.3134 - loss: 1.8608 - val_accuracy: 0.3607 - val_loss: 1.7536



WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save_model(model, 'my_model.keras')`.

Model saved as Custom CNN_SGD_model.h5

Training Custom CNN with Adam optimizer...

Epoch 1/5

782/782 ————— 48s 59ms/step - accuracy: 0.2321 - loss:
2.0277 - val_accuracy: 0.4101 - val_loss: 1.6282

Epoch 2/5

782/782 ————— 49s 62ms/step - accuracy: 0.3981 - loss:
1.6354 - val_accuracy: 0.5306 - val_loss: 1.2958

Epoch 3/5

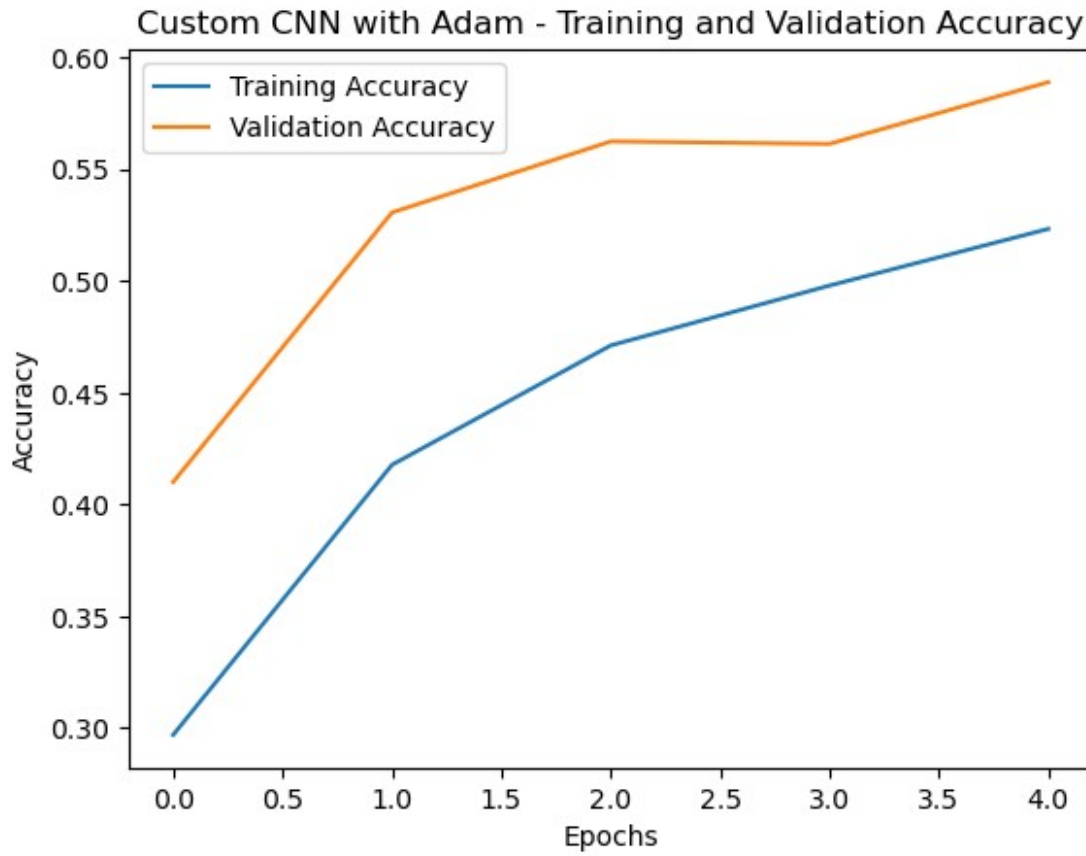
782/782 ————— 48s 61ms/step - accuracy: 0.4641 - loss:
1.4923 - val_accuracy: 0.5624 - val_loss: 1.2237

Epoch 4/5

782/782 ————— 45s 57ms/step - accuracy: 0.4915 - loss:
1.4183 - val_accuracy: 0.5613 - val_loss: 1.2291

Epoch 5/5

782/782 ————— 47s 59ms/step - accuracy: 0.5145 - loss:
1.3636 - val_accuracy: 0.5889 - val_loss: 1.1459



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Model saved as Custom CNN_Adam_model.h5

Training Custom CNN with RMSprop optimizer...

Epoch 1/5

782/782 ————— 52s 65ms/step - accuracy: 0.2269 - loss: 2.0621 - val_accuracy: 0.4706 - val_loss: 1.4545

Epoch 2/5

782/782 ————— 47s 59ms/step - accuracy: 0.4037 - loss: 1.6535 - val_accuracy: 0.4710 - val_loss: 1.4542

Epoch 3/5

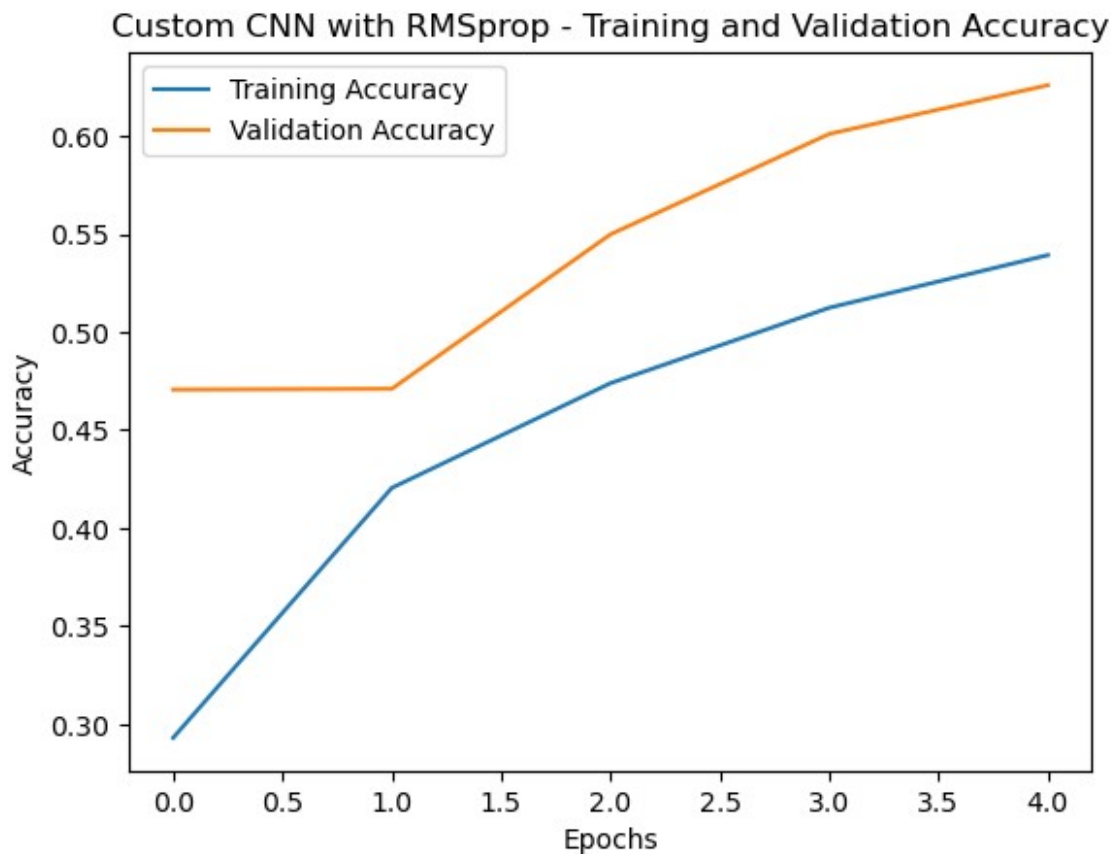
782/782 ————— 47s 60ms/step - accuracy: 0.4635 - loss: 1.4910 - val_accuracy: 0.5498 - val_loss: 1.2298

Epoch 4/5

782/782 ————— 43s 55ms/step - accuracy: 0.5038 - loss: 1.3955 - val_accuracy: 0.6010 - val_loss: 1.1344

Epoch 5/5

782/782 ————— 49s 63ms/step - accuracy: 0.5322 - loss: 1.3258 - val_accuracy: 0.6259 - val_loss: 1.0481



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Model saved as Custom CNN_RMSprop_model.h5

Training ResNet-18 with SGD optimizer...

Epoch 1/5

782/782 ————— 418s 513ms/step - accuracy: 0.2224 - loss: 2.9194 - val_accuracy: 0.3512 - val_loss: 3.1761

Epoch 2/5

782/782 ————— 377s 482ms/step - accuracy: 0.3366 - loss: 2.1665 - val_accuracy: 0.4215 - val_loss: 2.9457

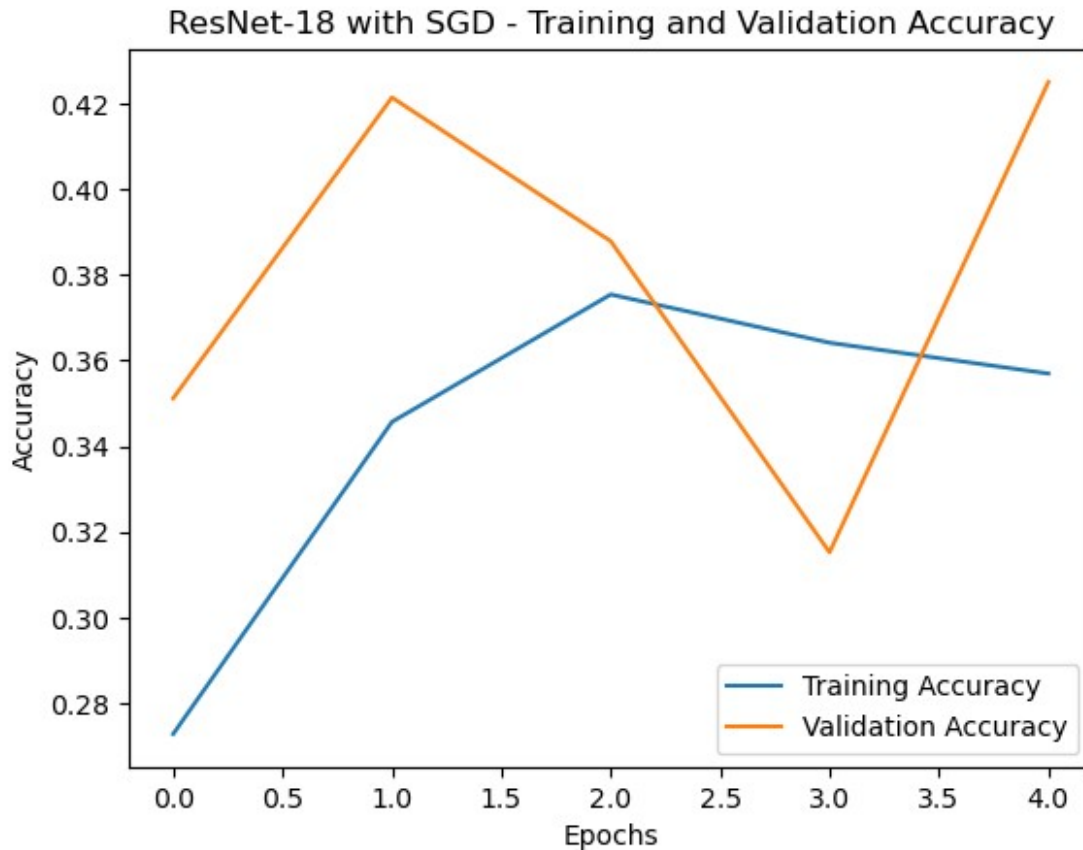
Epoch 3/5

782/782 ————— 369s 472ms/step - accuracy: 0.3744 - loss: 2.0131 - val_accuracy: 0.3879 - val_loss: 2.0184

Epoch 4/5

782/782 ————— 379s 484ms/step - accuracy: 0.3785 -

```
loss: 1.9743 - val_accuracy: 0.3152 - val_loss: 8.8636
Epoch 5/5
782/782 _____ 369s 471ms/step - accuracy: 0.3425 -
loss: 2.0657 - val_accuracy: 0.4251 - val_loss: 3.7996
```



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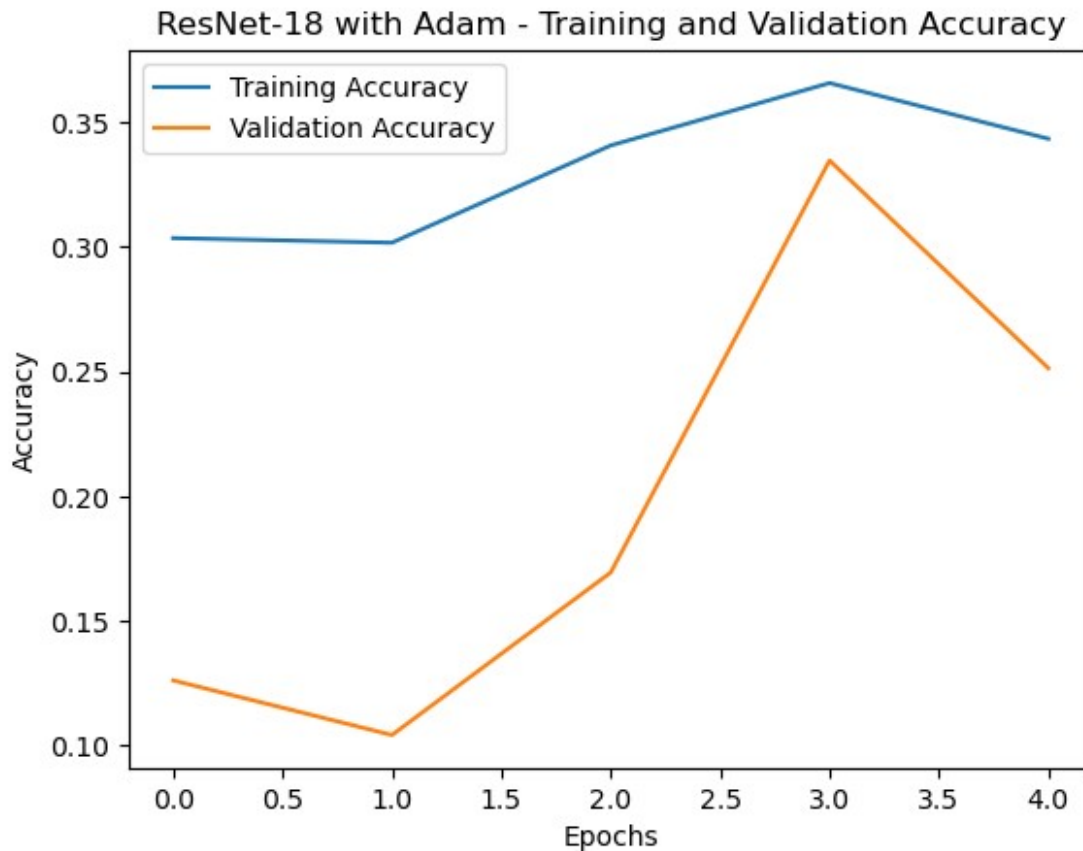
Model saved as ResNet-18_SGD_model.h5

Training ResNet-18 with Adam optimizer...

```
Epoch 1/5
782/782 _____ 510s 617ms/step - accuracy: 0.2629 -
loss: 2.4147 - val_accuracy: 0.1261 - val_loss: 12.2278
Epoch 2/5
782/782 _____ 472s 604ms/step - accuracy: 0.2931 -
loss: 2.1253 - val_accuracy: 0.1042 - val_loss: 6404.8013
Epoch 3/5
782/782 _____ 476s 609ms/step - accuracy: 0.3057 -
loss: 2.1543 - val_accuracy: 0.1694 - val_loss: 2.7235
```



```
Epoch 4/5
782/782 _____ 481s 615ms/step - accuracy: 0.3685 -
loss: 1.9414 - val_accuracy: 0.3346 - val_loss: 1.8869
Epoch 5/5
782/782 _____ 478s 611ms/step - accuracy: 0.3384 -
loss: 1.9913 - val_accuracy: 0.2513 - val_loss: 2.2245
```



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Model saved as ResNet-18_Adam_model.h5

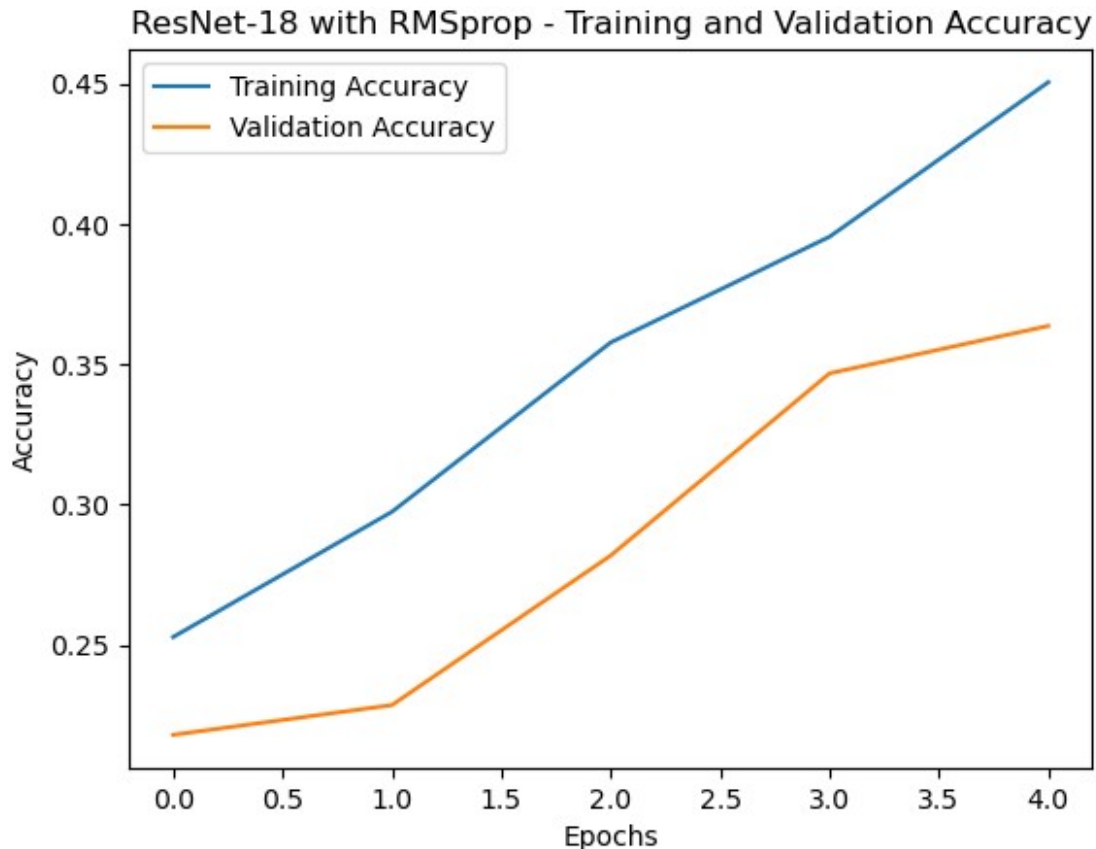
Training ResNet-18 with RMSprop optimizer...

```
Epoch 1/5
782/782 _____ 468s 572ms/step - accuracy: 0.2219 -
loss: 2.8002 - val_accuracy: 0.2180 - val_loss: 2.1203
Epoch 2/5
782/782 _____ 434s 554ms/step - accuracy: 0.2768 -
loss: 2.2972 - val_accuracy: 0.2286 - val_loss: 2.3177
Epoch 3/5
```

```

782/782 _____ 445s 569ms/step - accuracy: 0.3483 -
loss: 2.0476 - val_accuracy: 0.2819 - val_loss: 144.1120
Epoch 4/5
782/782 _____ 459s 587ms/step - accuracy: 0.3851 -
loss: 1.9709 - val_accuracy: 0.3468 - val_loss: 1.8172
Epoch 5/5
782/782 _____ 477s 609ms/step - accuracy: 0.4420 -
loss: 1.7564 - val_accuracy: 0.3637 - val_loss: 61.9801

```



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Model saved as ResNet-18_RMSprop_model.h5

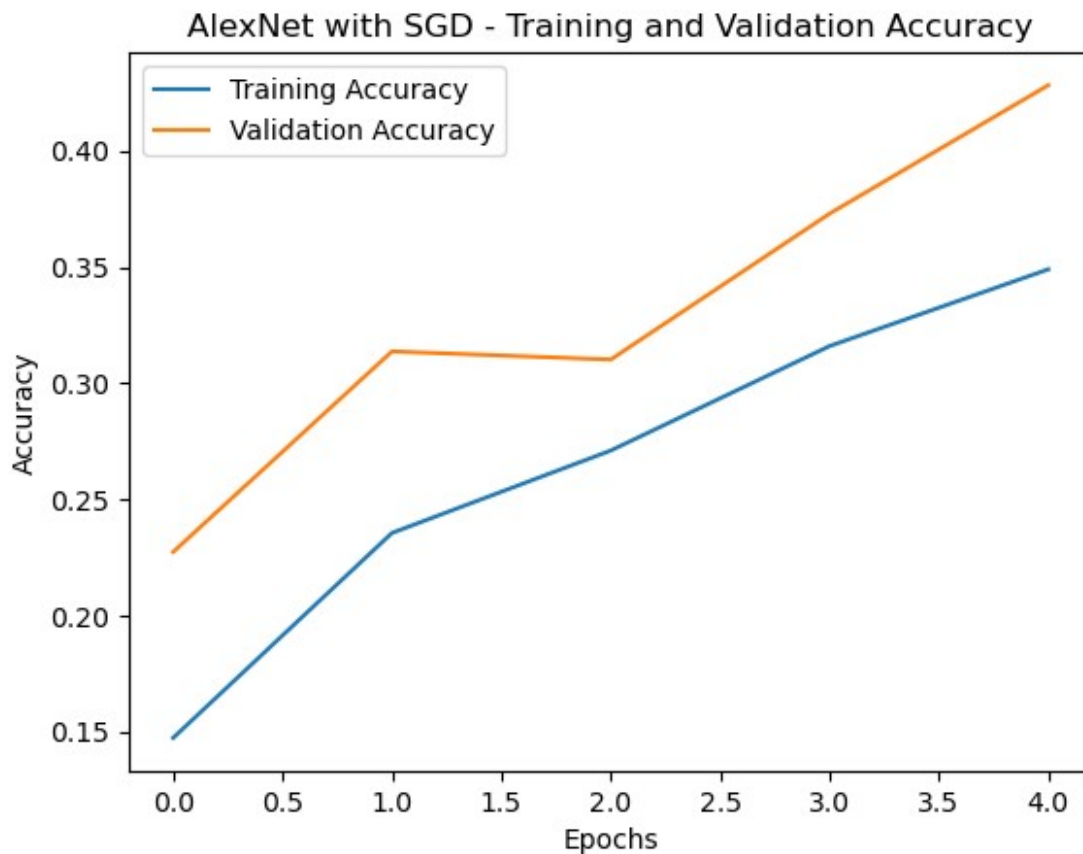
Training AlexNet with SGD optimizer...

```

Epoch 1/5
782/782 _____ 468s 597ms/step - accuracy: 0.1197 -
loss: 2.2875 - val_accuracy: 0.2274 - val_loss: 2.0547
Epoch 2/5
782/782 _____ 449s 573ms/step - accuracy: 0.2282 -

```

```
loss: 2.0683 - val_accuracy: 0.3137 - val_loss: 1.9204
Epoch 3/5
782/782 _____ 448s 573ms/step - accuracy: 0.2612 -
loss: 1.9914 - val_accuracy: 0.3102 - val_loss: 1.9069
Epoch 4/5
782/782 _____ 477s 609ms/step - accuracy: 0.3125 -
loss: 1.8809 - val_accuracy: 0.3730 - val_loss: 1.7464
Epoch 5/5
782/782 _____ 464s 593ms/step - accuracy: 0.3413 -
loss: 1.7978 - val_accuracy: 0.4283 - val_loss: 1.5747
```



```
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save_model(model, 'my_model.keras')`.
```

Model saved as AlexNet_SGD_model.h5

Training AlexNet with Adam optimizer...

```
Epoch 1/5
782/782 _____ 610s 775ms/step - accuracy: 0.2041 -
loss: 2.1058 - val_accuracy: 0.3666 - val_loss: 1.7424
```

```
Epoch 2/5
782/782 _____ 601s 768ms/step - accuracy: 0.4031 -
loss: 1.6083 - val_accuracy: 0.4821 - val_loss: 1.4000
Epoch 3/5
782/782 _____ 573s 732ms/step - accuracy: 0.4754 -
loss: 1.4282 - val_accuracy: 0.5571 - val_loss: 1.2308
Epoch 4/5
782/782 _____ 563s 719ms/step - accuracy: 0.5238 -
loss: 1.3211 - val_accuracy: 0.5850 - val_loss: 1.1762
Epoch 5/5
782/782 _____ 547s 699ms/step - accuracy: 0.5569 -
loss: 1.2347 - val_accuracy: 0.5772 - val_loss: 1.2378
```



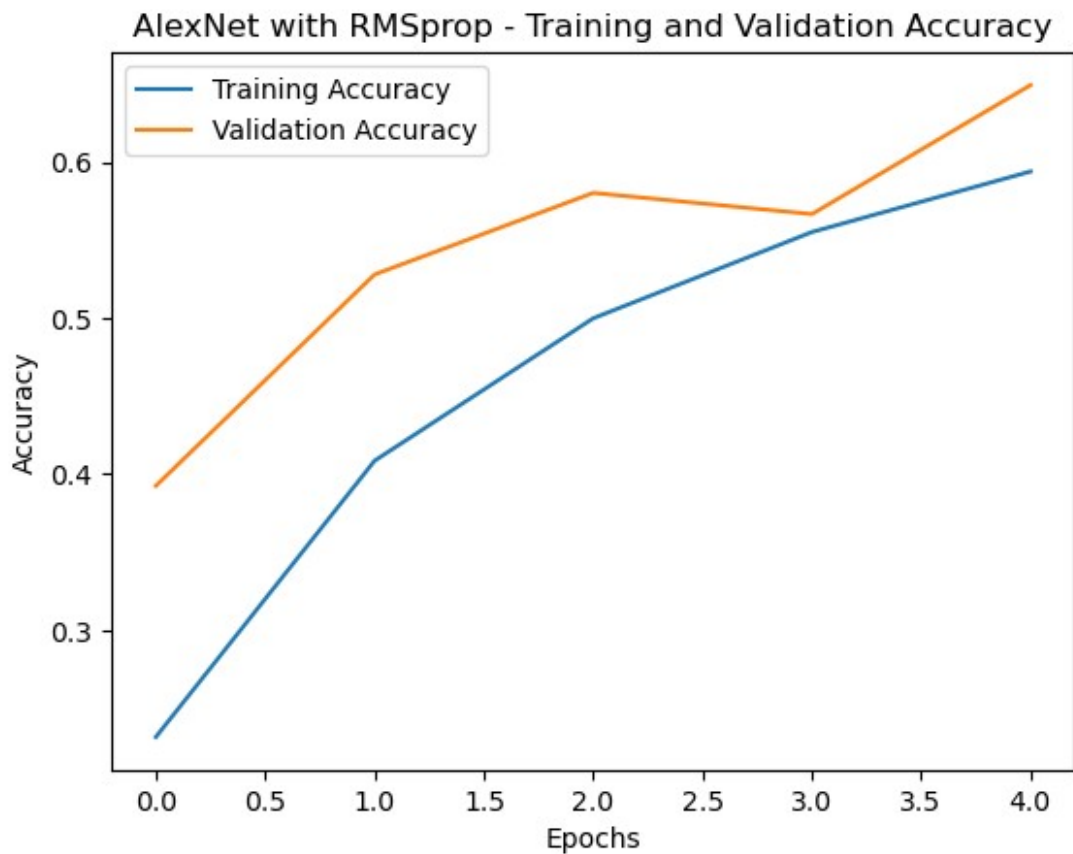
```
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save_model(model, 'my_model.keras')`.
```

```
Model saved as AlexNet_Adam_model.h5
```

```
Training AlexNet with RMSprop optimizer...
```

```
Epoch 1/5
```

```
782/782 ————— 504s 641ms/step - accuracy: 0.1605 -  
loss: 2.1947 - val_accuracy: 0.3925 - val_loss: 1.6118  
Epoch 2/5  
782/782 ————— 504s 644ms/step - accuracy: 0.3814 -  
loss: 1.6787 - val_accuracy: 0.5279 - val_loss: 1.2782  
Epoch 3/5  
782/782 ————— 507s 649ms/step - accuracy: 0.4839 -  
loss: 1.4379 - val_accuracy: 0.5802 - val_loss: 1.1778  
Epoch 4/5  
782/782 ————— 504s 645ms/step - accuracy: 0.5450 -  
loss: 1.3035 - val_accuracy: 0.5666 - val_loss: 1.2680  
Epoch 5/5  
782/782 ————— 503s 643ms/step - accuracy: 0.5865 -  
loss: 1.1955 - val_accuracy: 0.6493 - val_loss: 1.0522
```



```
WARNING:absl:You are saving your model as an HDF5 file via  
`model.save()` or `keras.saving.save_model(model)`. This file format  
is considered legacy. We recommend using instead the native Keras  
format, e.g. `model.save('my_model.keras')` or  
`keras.saving.save_model(model, 'my_model.keras')`.
```

Model saved as AlexNet_RMSprop_model.h5

Best Model: AlexNet with RMSprop with Accuracy: 64.93%

Generating the predictions

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

the confusion matrix

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

Plotting the confusion matrix

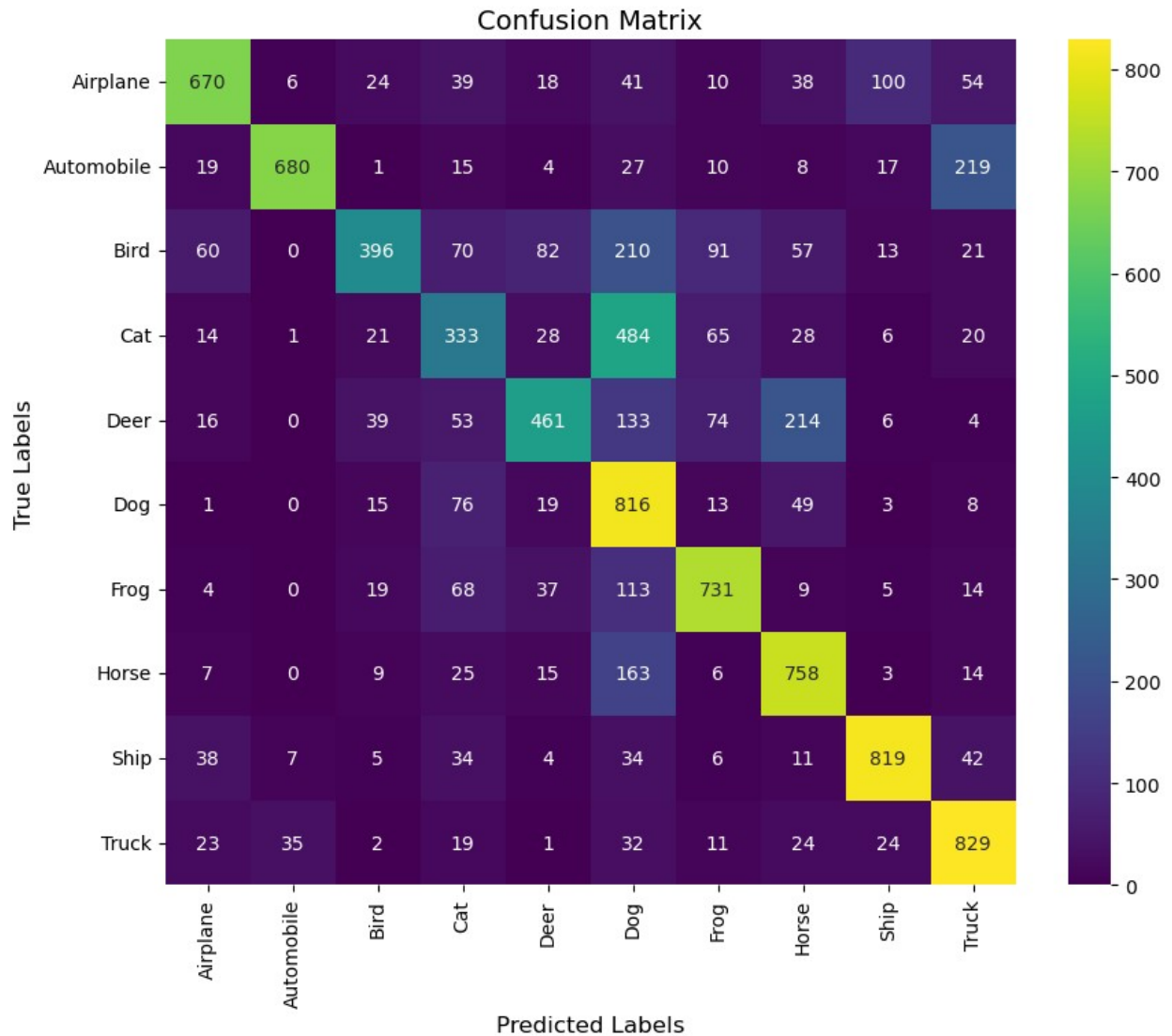
```
def plot_confusion_matrix(cm, class_names):
    plt.figure(figsize=(10, 8))
    sns.heatmap(cm, annot=True, fmt='d', cmap='viridis',
xticklabels=class_names, yticklabels=class_names)
    plt.xlabel('Predicted Labels', fontsize=12)
    plt.ylabel('True Labels', fontsize=12)
    plt.title('Confusion Matrix', fontsize=14)
    plt.show()
```

Class names for CIFAR-10 dataset

```
class_names = [
    'Airplane', 'Automobile', 'Bird', 'Cat', 'Deer',
    'Dog', 'Frog', 'Horse', 'Ship', 'Truck'
]
```

```
plot_confusion_matrix(conf_matrix, class_names)
```

313/313 ————— 21s 67ms/step



```
report = classification_report(y_true, y_pred,
target_names=class_names, output_dict=True)
# Plotting the classification report
def plot_classification_report(report, class_names):
    plt.figure(figsize=(10, 6))
    precision = [report[label]['precision'] for label in class_names]
    recall = [report[label]['recall'] for label in class_names]
    f1_score = [report[label]['f1-score'] for label in class_names]

    # Set up the bar width
    bar_width = 0.25
    index = np.arange(len(class_names))

    # Plot the bars
    plt.bar(index, precision, bar_width, label='Precision')
    plt.bar(index + bar_width, recall, bar_width, label='Recall')
```

```

plt.bar(index + 2 * bar_width, f1_score, bar_width, label='F1-
Score')

# the plot
plt.xlabel('Classes', fontsize=12)
plt.ylabel('Scores', fontsize=12)
plt.title('Classification Report', fontsize=14)
plt.xticks(index + bar_width, class_names, rotation=45)
plt.legend()
plt.tight_layout()
plt.show()

#
plot_classification_report(report, class_names)

```

