

```
In [78]: # Step 1: Import libraries
import os
import numpy as np
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
from tensorflow.keras.preprocessing import image_dataset_from_directory
from tensorflow.keras import layers, models, Sequential
from tensorflow.keras.applications import ResNet50
from tensorflow.keras import layers, models
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
from tensorflow.keras.applications.resnet50 import preprocess_input, decode_predictions
from sklearn.model_selection import GridSearchCV
```

```
In [79]: data_directory = "/Users/navd/Downloads/IndianCurrencyNotesDataset/AllImages"
batch_size = 32
image_size = (224, 224)
```

```
In [80]: # Create a labeled dataset from a directory
train_data = image_dataset_from_directory(
    data_directory,
    labels="inferred",
    label_mode="int",
    image_size=image_size,
    batch_size=batch_size,
    validation_split=.2,
    subset="training",
    seed=123,
    shuffle=True
)

validation_data = image_dataset_from_directory(
    data_directory,
    labels="inferred",
    label_mode="int",
    image_size=image_size,
    batch_size=batch_size,
    validation_split=.2,
    subset="validation",
    seed=123,
    shuffle=True
)
```

```
Found 178 files belonging to 7 classes.
Using 143 files for training.
Found 178 files belonging to 7 classes.
Using 35 files for validation.
```

```
In [81]: categories = train_data.class_names
plt.figure(figsize=(10, 10))
for images, labels in train_data.take(1):
    for i in range(9):
        ax = plt.subplot(3, 3, i + 1)
        plt.imshow(images[i].numpy().astype("uint8"))
        plt.title(categories[labels[i]])
        plt.axis("off")
plt.show()

data_augmentation = Sequential([
    layers.experimental.preprocessing.RandomFlip("horizontal"),
    layers.experimental.preprocessing.RandomRotation(0.2),
    layers.experimental.preprocessing.RandomZoom(0.2),
])
augmented_train_data = train_data.map(lambda x, y: (data_augmentation(x), y))
```

20 Note



10 Note



10 Note



50 Note



10 Note



200 Note





500 Note



50 Note



20 Note



In [82]: *# Step 4: Load pretrained ResNet50 model*

```
resnet_model = ResNet50(
    include_top=False,
    weights="imagenet",
    input_shape=(224,224,3),
    pooling='avg',
    classes=7,
)
resnet_model.summary()
```

Model: "resnet50"

Layer (type)	Output Shape	Param #	Connected to
input_13 (InputLayer)	[ (None, 224, 224, 3) ]	0	[ ]
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3)	0	[ 'input_13[0][0]' ]
conv1_conv (Conv2D)	(None, 112, 112, 64)	9472	[ 'conv1_pad[0][0]' ]
conv1_bn (BatchNormalization)	(None, 112, 112, 64)	256	[ 'conv1_conv[0][0]' ]
conv1_relu (Activation)	(None, 112, 112, 64)	0	[ 'conv1_bn[0][0]' ]
pool1_pad (ZeroPadding2D)	(None, 114, 114, 64)	0	[ 'conv1_relu[0][0]' ]
pool1_pool (MaxPooling2D)	(None, 56, 56, 64)	0	[ 'pool1_pad[0][0]' ]
conv2_block1_1_conv (Conv2D)	(None, 56, 56, 64)	4160	[ 'pool1_pool[0][0]' ]
conv2_block1_1_bn (BatchNormalization)	(None, 56, 56, 64)	256	[ 'conv2_block1_1_conv[0][0]' ]
conv2_block1_1_relu (Activation)	(None, 56, 56, 64)	0	[ 'conv2_block1_1_bn[0][0]' ]

conv2_block1_2_conv (Conv2D)	(None, 56, 56, 64)	36928	['conv2_block1_1_relu[0][0]']
conv2_block1_2_bn (Batch Normalization)	(None, 56, 56, 64)	256	['conv2_block1_2_conv[0][0]']
conv2_block1_2_relu (Activation)	(None, 56, 56, 64)	0	['conv2_block1_2_bn[0][0]']
conv2_block1_0_conv (Conv2D)	(None, 56, 56, 256)	16640	['pool1_pool[0][0]']
conv2_block1_3_conv (Conv2D)	(None, 56, 56, 256)	16640	['conv2_block1_2_relu[0][0]']
conv2_block1_0_bn (Batch Normalization)	(None, 56, 56, 256)	1024	['conv2_block1_0_conv[0][0]']
conv2_block1_3_bn (Batch Normalization)	(None, 56, 56, 256)	1024	['conv2_block1_3_conv[0][0]']
conv2_block1_add (Add)	(None, 56, 56, 256)	0	['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']
conv2_block1_out (Activation)	(None, 56, 56, 256)	0	['conv2_block1_add[0][0]']
conv2_block2_1_conv (Conv2D)	(None, 56, 56, 64)	16448	['conv2_block1_out[0][0]']
conv2_block2_1_bn (Batch Normalization)	(None, 56, 56, 64)	256	['conv2_block2_1_conv[0][0]']
conv2_block2_1_relu (Activation)	(None, 56, 56, 64)	0	['conv2_block2_1_bn[0][0]']
conv2_block2_2_conv (Conv2D)	(None, 56, 56, 64)	36928	['conv2_block2_1_relu[0][0]']

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conv2_block2_2_bn (Batch Normalization)	(None, 56, 56, 64)	256	['conv2_block2_2_conv[0][0]']
conv2_block2_2_relu (Activation)	(None, 56, 56, 64)	0	['conv2_block2_2_bn[0][0]']
conv2_block2_3_conv (Conv2D)	(None, 56, 56, 256)	16640	['conv2_block2_2_relu[0][0]']
conv2_block2_3_bn (Batch Normalization)	(None, 56, 56, 256)	1024	['conv2_block2_3_conv[0][0]']
conv2_block2_add (Add)	(None, 56, 56, 256)	0	['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']
conv2_block2_out (Activation)	(None, 56, 56, 256)	0	['conv2_block2_add[0][0]']
conv2_block3_1_conv (Conv2D)	(None, 56, 56, 64)	16448	['conv2_block2_out[0][0]']
conv2_block3_1_bn (Batch Normalization)	(None, 56, 56, 64)	256	['conv2_block3_1_conv[0][0]']
conv2_block3_1_relu (Activation)	(None, 56, 56, 64)	0	['conv2_block3_1_bn[0][0]']
conv2_block3_2_conv (Conv2D)	(None, 56, 56, 64)	36928	['conv2_block3_1_relu[0][0]']
conv2_block3_2_bn (Batch Normalization)	(None, 56, 56, 64)	256	['conv2_block3_2_conv[0][0]']
conv2_block3_2_relu (Activation)	(None, 56, 56, 64)	0	['conv2_block3_2_bn[0][0]']

conv2_block3_3_conv (Conv2D)	(None, 56, 56, 256)	16640	['conv2_block3_2_relu[0][0]']
conv2_block3_3_bn (Batch Normalization)	(None, 56, 56, 256)	1024	['conv2_block3_3_conv[0][0]']
conv2_block3_add (Add)	(None, 56, 56, 256)	0	['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']
conv2_block3_out (Activation)	(None, 56, 56, 256)	0	['conv2_block3_add[0][0]']
conv3_block1_1_conv (Conv2D)	(None, 28, 28, 128)	32896	['conv2_block3_out[0][0]']
conv3_block1_1_bn (Batch Normalization)	(None, 28, 28, 128)	512	['conv3_block1_1_conv[0][0]']
conv3_block1_1_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block1_1_bn[0][0]']
conv3_block1_2_conv (Conv2D)	(None, 28, 28, 128)	147584	['conv3_block1_1_relu[0][0]']
conv3_block1_2_bn (Batch Normalization)	(None, 28, 28, 128)	512	['conv3_block1_2_conv[0][0]']
conv3_block1_2_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block1_2_bn[0][0]']
conv3_block1_0_conv (Conv2D)	(None, 28, 28, 512)	131584	['conv2_block3_out[0][0]']
conv3_block1_3_conv (Conv2D)	(None, 28, 28, 512)	66048	['conv3_block1_2_relu[0][0]']
conv3_block1_0_bn (Batch Normalization)	(None, 28, 28, 512)	2048	['conv3_block1_0_conv[0][0]']



conv3_block1_3_bn (Batch Normalization)	(None, 28, 28, 512)	2048	['conv3_block1_3_conv[0][0]']
conv3_block1_add (Add)	(None, 28, 28, 512)	0	['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']
conv3_block1_out (Activation)	(None, 28, 28, 512)	0	['conv3_block1_add[0][0]']
conv3_block2_1_conv (Conv2D)	(None, 28, 28, 128)	65664	['conv3_block1_out[0][0]']
conv3_block2_1_bn (Batch Normalization)	(None, 28, 28, 128)	512	['conv3_block2_1_conv[0][0]']
conv3_block2_1_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block2_1_bn[0][0]']
conv3_block2_2_conv (Conv2D)	(None, 28, 28, 128)	147584	['conv3_block2_1_relu[0][0]']
conv3_block2_2_bn (Batch Normalization)	(None, 28, 28, 128)	512	['conv3_block2_2_conv[0][0]']
conv3_block2_2_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block2_2_bn[0][0]']
conv3_block2_3_conv (Conv2D)	(None, 28, 28, 512)	66048	['conv3_block2_2_relu[0][0]']
conv3_block2_3_bn (Batch Normalization)	(None, 28, 28, 512)	2048	['conv3_block2_3_conv[0][0]']
conv3_block2_add (Add)	(None, 28, 28, 512)	0	['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']
conv3_block2_out (Activation)	(None, 28, 28, 512)	0	['conv3_block2_add[0][0]']

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conv3_block3_1_conv (Conv2D)	(None, 28, 28, 128)	65664	['conv3_block2_out[0][0]']
conv3_block3_1_bn (BatchNormalization)	(None, 28, 28, 128)	512	['conv3_block3_1_conv[0][0]']
conv3_block3_1_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block3_1_bn[0][0]']
conv3_block3_2_conv (Conv2D)	(None, 28, 28, 128)	147584	['conv3_block3_1_relu[0][0]']
conv3_block3_2_bn (BatchNormalization)	(None, 28, 28, 128)	512	['conv3_block3_2_conv[0][0]']
conv3_block3_2_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block3_2_bn[0][0]']
conv3_block3_3_conv (Conv2D)	(None, 28, 28, 512)	66048	['conv3_block3_2_relu[0][0]']
conv3_block3_3_bn (BatchNormalization)	(None, 28, 28, 512)	2048	['conv3_block3_3_conv[0][0]']
conv3_block3_add (Add)	(None, 28, 28, 512)	0	['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']
conv3_block3_out (Activation)	(None, 28, 28, 512)	0	['conv3_block3_add[0][0]']
conv3_block4_1_conv (Conv2D)	(None, 28, 28, 128)	65664	['conv3_block3_out[0][0]']
conv3_block4_1_bn (BatchNormalization)	(None, 28, 28, 128)	512	['conv3_block4_1_conv[0][0]']

conv3_block4_1_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block4_1_bn[0][0]']
conv3_block4_2_conv (Conv2D)	(None, 28, 28, 128)	147584	['conv3_block4_1_relu[0][0]']
conv3_block4_2_bn (BatchNormalization)	(None, 28, 28, 128)	512	['conv3_block4_2_conv[0][0]']
conv3_block4_2_relu (Activation)	(None, 28, 28, 128)	0	['conv3_block4_2_bn[0][0]']
conv3_block4_3_conv (Conv2D)	(None, 28, 28, 512)	66048	['conv3_block4_2_relu[0][0]']
conv3_block4_3_bn (BatchNormalization)	(None, 28, 28, 512)	2048	['conv3_block4_3_conv[0][0]']
conv3_block4_add (Add)	(None, 28, 28, 512)	0	['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']
conv3_block4_out (Activation)	(None, 28, 28, 512)	0	['conv3_block4_add[0][0]']
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328	['conv3_block4_out[0][0]']
conv4_block1_1_bn (BatchNormalization)	(None, 14, 14, 256)	1024	['conv4_block1_1_conv[0][0]']
conv4_block1_1_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block1_1_bn[0][0]']
conv4_block1_2_conv (Conv2D)	(None, 14, 14, 256)	590080	['conv4_block1_1_relu[0][0]']
conv4_block1_2_bn (BatchNormalization)	(None, 14, 14, 256)	1024	['conv4_block1_2_conv[0][0]']

conv4_block1_2_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block1_2_bn[0][0]']
conv4_block1_0_conv (Conv2D)	(None, 14, 14, 1024)	525312	['conv3_block4_out[0][0]']
conv4_block1_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	['conv4_block1_2_relu[0][0]']
conv4_block1_0_bn (Batch Normalization)	(None, 14, 14, 1024)	4096	['conv4_block1_0_conv[0][0]']
conv4_block1_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096	['conv4_block1_3_conv[0][0]']
conv4_block1_add (Add)	(None, 14, 14, 1024)	0	['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']
conv4_block1_out (Activation)	(None, 14, 14, 1024)	0	['conv4_block1_add[0][0]']
conv4_block2_1_conv (Conv2D)	(None, 14, 14, 256)	262400	['conv4_block1_out[0][0]']
conv4_block2_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block2_1_conv[0][0]']
conv4_block2_1_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block2_1_bn[0][0]']
conv4_block2_2_conv (Conv2D)	(None, 14, 14, 256)	590080	['conv4_block2_1_relu[0][0]']
conv4_block2_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block2_2_conv[0][0]']
conv4_block2_2_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block2_2_bn[0][0]']

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conv4_block2_3_conv (Conv2D)      (None, 14, 14, 1024)      263168      ['conv4_block2_2_relu[0][0]']

conv4_block2_3_bn (BatchNormalization) (None, 14, 14, 1024)      4096        ['conv4_block2_3_conv[0][0]']

conv4_block2_add (Add)             (None, 14, 14, 1024)      0            ['conv4_block1_out[0][0]',
                                'conv4_block2_3_bn[0][0]']

conv4_block2_out (Activation)      (None, 14, 14, 1024)      0            ['conv4_block2_add[0][0]']

conv4_block3_1_conv (Conv2D)      (None, 14, 14, 256)       262400      ['conv4_block2_out[0][0]']

conv4_block3_1_bn (BatchNormalization) (None, 14, 14, 256)       1024        ['conv4_block3_1_conv[0][0]']

conv4_block3_1_relu (Activation)   (None, 14, 14, 256)       0            ['conv4_block3_1_bn[0][0]']

conv4_block3_2_conv (Conv2D)      (None, 14, 14, 256)       590080      ['conv4_block3_1_relu[0][0]']

conv4_block3_2_bn (BatchNormalization) (None, 14, 14, 256)       1024        ['conv4_block3_2_conv[0][0]']

conv4_block3_2_relu (Activation)   (None, 14, 14, 256)       0            ['conv4_block3_2_bn[0][0]']

conv4_block3_3_conv (Conv2D)      (None, 14, 14, 1024)      263168      ['conv4_block3_2_relu[0][0]']

conv4_block3_3_bn (BatchNormalization) (None, 14, 14, 1024)      4096        ['conv4_block3_3_conv[0][0]']

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conv4_block3_add (Add)	(None, 14, 14, 1024)	0	['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']
conv4_block3_out (Activation)	(None, 14, 14, 1024)	0	['conv4_block3_add[0][0]']
conv4_block4_1_conv (Conv2D)	(None, 14, 14, 256)	262400	['conv4_block3_out[0][0]']
conv4_block4_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block4_1_conv[0][0]']
conv4_block4_1_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block4_1_bn[0][0]']
conv4_block4_2_conv (Conv2D)	(None, 14, 14, 256)	590080	['conv4_block4_1_relu[0][0]']
conv4_block4_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block4_2_conv[0][0]']
conv4_block4_2_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block4_2_bn[0][0]']
conv4_block4_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	['conv4_block4_2_relu[0][0]']
conv4_block4_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096	['conv4_block4_3_conv[0][0]']
conv4_block4_add (Add)	(None, 14, 14, 1024)	0	['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']
conv4_block4_out (Activation)	(None, 14, 14, 1024)	0	['conv4_block4_add[0][0]']
conv4_block5_1_conv (Conv2D)	(None, 14, 14, 256)	262400	['conv4_block4_out[0][0]']

conv4_block5_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block5_1_conv[0][0]']
conv4_block5_1_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block5_1_bn[0][0]']
conv4_block5_2_conv (Conv2D)	(None, 14, 14, 256)	590080	['conv4_block5_1_relu[0][0]']
conv4_block5_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block5_2_conv[0][0]']
conv4_block5_2_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block5_2_bn[0][0]']
conv4_block5_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	['conv4_block5_2_relu[0][0]']
conv4_block5_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096	['conv4_block5_3_conv[0][0]']
conv4_block5_add (Add)	(None, 14, 14, 1024)	0	['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']
conv4_block5_out (Activation)	(None, 14, 14, 1024)	0	['conv4_block5_add[0][0]']
conv4_block6_1_conv (Conv2D)	(None, 14, 14, 256)	262400	['conv4_block5_out[0][0]']
conv4_block6_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block6_1_conv[0][0]']
conv4_block6_1_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block6_1_bn[0][0]']
conv4_block6_2_conv (Conv2D)	(None, 14, 14, 256)	590080	['conv4_block6_1_relu[0][0]']

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conv4_block6_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024	['conv4_block6_2_conv[0][0]']
conv4_block6_2_relu (Activation)	(None, 14, 14, 256)	0	['conv4_block6_2_bn[0][0]']
conv4_block6_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	['conv4_block6_2_relu[0][0]']
conv4_block6_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096	['conv4_block6_3_conv[0][0]']
conv4_block6_add (Add)	(None, 14, 14, 1024)	0	['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']
conv4_block6_out (Activation)	(None, 14, 14, 1024)	0	['conv4_block6_add[0][0]']
conv5_block1_1_conv (Conv2D)	(None, 7, 7, 512)	524800	['conv4_block6_out[0][0]']
conv5_block1_1_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block1_1_conv[0][0]']
conv5_block1_1_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block1_1_bn[0][0]']
conv5_block1_2_conv (Conv2D)	(None, 7, 7, 512)	2359808	['conv5_block1_1_relu[0][0]']
conv5_block1_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block1_2_conv[0][0]']
conv5_block1_2_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block1_2_bn[0][0]']



conv5_block1_0_conv (Conv2D)	(None, 7, 7, 2048)	2099200	['conv4_block6_out[0][0]']
conv5_block1_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624	['conv5_block1_2_relu[0][0]']
conv5_block1_0_bn (Batch Normalization)	(None, 7, 7, 2048)	8192	['conv5_block1_0_conv[0][0]']
conv5_block1_3_bn (Batch Normalization)	(None, 7, 7, 2048)	8192	['conv5_block1_3_conv[0][0]']
conv5_block1_add (Add)	(None, 7, 7, 2048)	0	['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']
conv5_block1_out (Activation)	(None, 7, 7, 2048)	0	['conv5_block1_add[0][0]']
conv5_block2_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	['conv5_block1_out[0][0]']
conv5_block2_1_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block2_1_conv[0][0]']
conv5_block2_1_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block2_1_bn[0][0]']
conv5_block2_2_conv (Conv2D)	(None, 7, 7, 512)	2359808	['conv5_block2_1_relu[0][0]']
conv5_block2_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block2_2_conv[0][0]']
conv5_block2_2_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block2_2_bn[0][0]']
conv5_block2_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624	['conv5_block2_2_relu[0][0]']

conv5_block2_3_bn (Batch Normalization)	(None, 7, 7, 2048)	8192	['conv5_block2_3_conv[0][0]']
conv5_block2_add (Add)	(None, 7, 7, 2048)	0	['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']
conv5_block2_out (Activation)	(None, 7, 7, 2048)	0	['conv5_block2_add[0][0]']
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	['conv5_block2_out[0][0]']
conv5_block3_1_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block3_1_conv[0][0]']
conv5_block3_1_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block3_1_bn[0][0]']
conv5_block3_2_conv (Conv2D)	(None, 7, 7, 512)	2359808	['conv5_block3_1_relu[0][0]']
conv5_block3_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048	['conv5_block3_2_conv[0][0]']
conv5_block3_2_relu (Activation)	(None, 7, 7, 512)	0	['conv5_block3_2_bn[0][0]']
conv5_block3_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624	['conv5_block3_2_relu[0][0]']
conv5_block3_3_bn (Batch Normalization)	(None, 7, 7, 2048)	8192	['conv5_block3_3_conv[0][0]']
conv5_block3_add (Add)	(None, 7, 7, 2048)	0	['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']
conv5_block3_out (Activation)	(None, 7, 7, 2048)	0	['conv5_block3_add[0][0]']

on)

avg\_pool (GlobalAveragePool (None, 2048) 0 ['conv5\_block3\_out[0][0]']  
ling2D)

```
=====
Total params: 23587712 (89.98 MB)
Trainable params: 23534592 (89.78 MB)
Non-trainable params: 53120 (207.50 KB)
```

```
In [83]: features_list = []
labels_list = []

for images, labels in augmented_train_data:
    preprocessed_images = preprocess_input(images)

    features = resnet_model.predict(preprocessed_images)

    features_list.append(features)
    labels_list.append(labels.numpy())
features_array = np.concatenate(features_list, axis=0)
labels_array = np.concatenate(labels_list, axis=0)

print("Shape of features array:", features_array.shape)
print("Shape of labels array:", labels_array.shape)

1/1 [=====] - 1s 1s/step
1/1 [=====] - 1s 839ms/step
1/1 [=====] - 1s 815ms/step
1/1 [=====] - 1s 786ms/step
1/1 [=====] - 1s 641ms/step
Shape of features array: (143, 2048)
Shape of labels array: (143,)
```

```

In [84]: test_features_list = []
test_labels_list = []
# Extract features and labels
for images, labels in validation_data:
    # Preprocess images for ResNet50 model
    preprocessed_images = preprocess_input(images)

    # Extract features using the pre-trained ResNet50 model
    features = resnet_model.predict(preprocessed_images)

    # Append features and labels to the lists
    test_features_list.append(features)
    test_labels_list.append(labels.numpy())

# Convert lists to NumPy arrays
test_features_array = np.concatenate(test_features_list, axis=0)
test_labels_array = np.concatenate(test_labels_list, axis=0)

# Display the shapes of the extracted features and labels
print("Shape of features array:", test_features_array.shape)
print("Shape of labels array:", test_labels_array.shape)

1/1 [=====] - 1s 818ms/step
1/1 [=====] - 0s 110ms/step
Shape of features array: (35, 2048)
Shape of labels array: (35,)

```

```

In [85]: # Define the parameter grid to search
parameter_grid = {
    'C': [0.1, 1, 10, 100], # You can adjust the range of C values
    'kernel': ['linear', 'rbf', 'poly', 'sigmoid'], # You can adjust the list of kernel functions
}

# Create an SVM classifier
svm_classifier = SVC()

# Create the GridSearchCV object

```

```
grid_search_cv = GridSearchCV(svm_classifier, parameter_grid, cv=5, scoring='accuracy')

# Perform the grid search on your training data
grid_search_cv.fit(features_array, labels_array)

# Print the best hyperparameters and corresponding accuracy
print("Best Hyperparameters:", grid_search_cv.best_params_)
print("Best Accuracy:", grid_search_cv.best_score_)

# Get the best model from the grid search
best_svm_classifier = grid_search_cv.best_estimator_

# Make predictions on the test set using the best model
best_predictions = best_svm_classifier.predict(test_features_array)

# Calculate accuracy and print confusion matrix for the best model
best_accuracy = accuracy_score(test_labels_array, best_predictions)
print(f'Best Model Accuracy: {best_accuracy * 100:.2f}%')

best_confusion_matrix = confusion_matrix(test_labels_array, best_predictions)

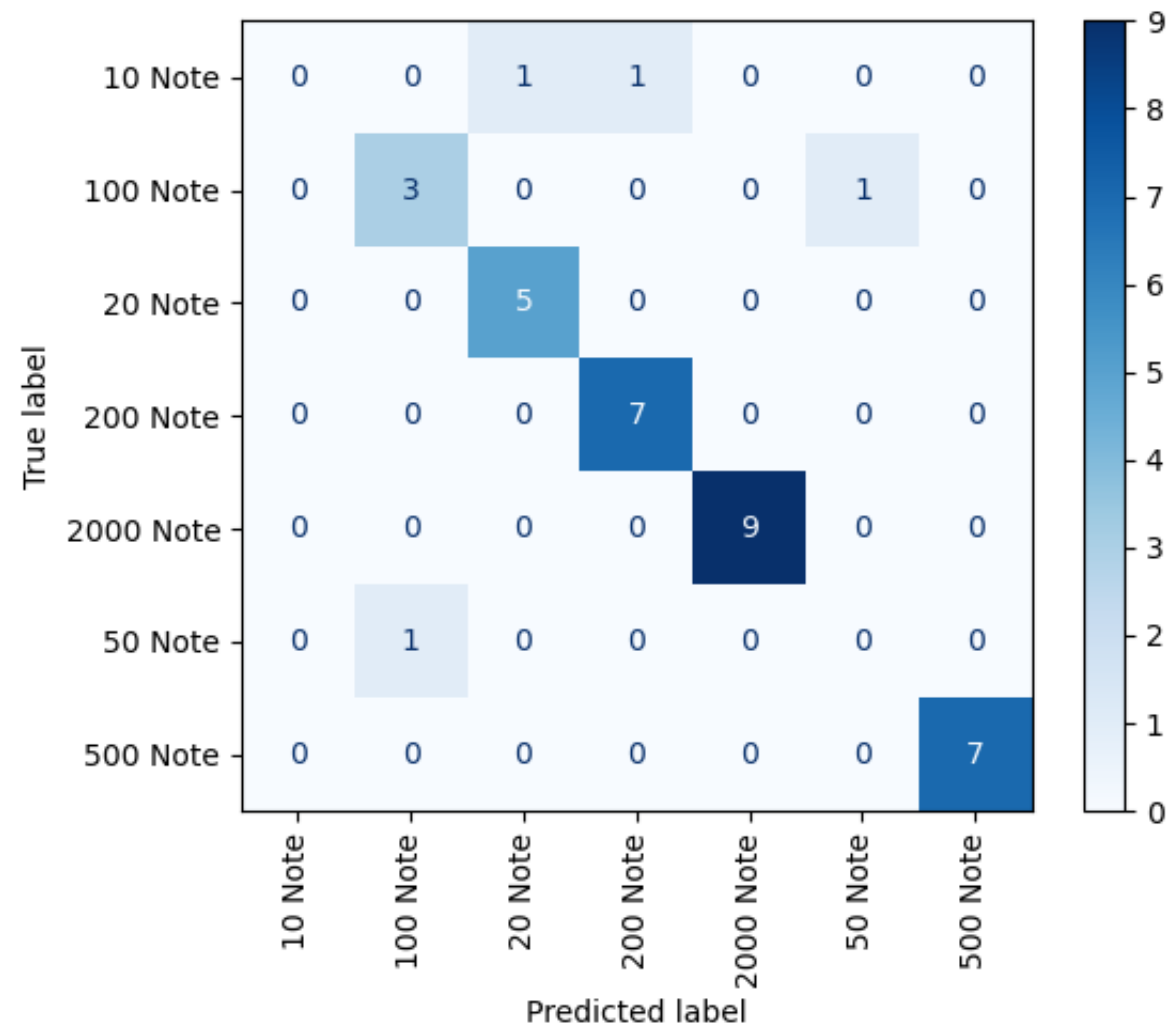
plt.figure(figsize=(8, 6))

best_confusion_matrix_display = ConfusionMatrixDisplay(confusion_matrix=best_confusion_matrix, display_labels=cat

best_confusion_matrix_display.plot(cmap=plt.cm.Blues, xticks_rotation=90)

plt.show()
```

```
Best Hyperparameters: {'C': 0.1, 'kernel': 'linear'}
Best Accuracy: 0.8669950738916257
Best Model Accuracy: 88.57%
<Figure size 800x600 with 0 Axes>
```



```
In [86]: correctly_classified_indices = np.where(test_labels_array == best_predictions)[0]
num_correctly_classified_images = min(9, len(correctly_classified_indices))

plt.figure(figsize=(12, 12))
for i in range(num_correctly_classified_images):
    ax = plt.subplot(3, 3, i + 1)
    index = correctly_classified_indices[i]
    img, actual_label = val_dataset.unbatch().skip(index).take(1).as_numpy_iterator().next()
    img = img.astype("uint8")
    predicted_label = class_names[best_predictions[index]]

    plt.imshow(img)
    plt.title(f'Actual: {class_names[actual_label]}\nPredicted: {predicted_label}')
    plt.axis("off")
plt.suptitle("Some correctly classified images")
plt.show()
```

### Some correctly classified images

Actual: 100 Note  
Predicted: 100 Note



Actual: 20 Note  
Predicted: 200 Note



Actual: 200 Note  
Predicted: 2000 Note





Actual: 2000 Note  
Predicted: 2000 Note



Actual: 10 Note  
Predicted: 100 Note

नए 10₹ के सबसे कीमती नोट



Actual: 500 Note  
Predicted: 500 Note



Actual: 20 Note  
Predicted: 500 Note



Actual: 500 Note  
Predicted: 2000 Note



Actual: 200 Note  
Predicted: 100 Note







```
In [87]: wrongly_classified_indices = np.where(test_labels_array != best_predictions)[0]
num_wrongly_classified_images = min(9, len(wrongly_classified_indices))

plt.figure(figsize=(12, 12))
for i in range(num_wrongly_classified_images):
    ax = plt.subplot(3, 3, i + 1)
    index = wrongly_classified_indices[i]
    img, actual_label = val_dataset.unbatch().skip(index).take(1).as_numpy_iterator().next()
    img = img.astype("uint8")
    predicted_label = class_names[best_predictions[index]]

    plt.imshow(img)
    plt.title(f'Actual: {class_names[actual_label]}\nPredicted: {predicted_label}')
    plt.axis("off")
plt.suptitle("some wrongly classified images")
plt.show()
```

some wrongly classified images

Actual: 2000 Note  
Predicted: 100 Note

Note Ko Pehechanane Ka Sahi Tarika



Actual: 50 Note  
Predicted: 200 Note

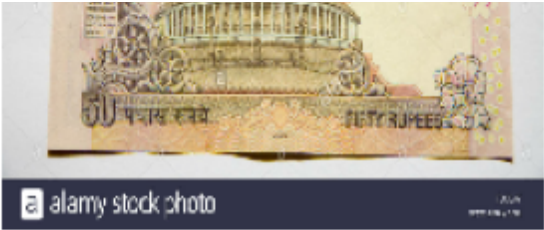


Actual: 2000 Note  
Predicted: 20 Note





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Actual: 2000 Note  
Predicted: 50 Note



In [ ]:

In [ ]:

In [ ]: