

# Convolutional Neural Networks

## Inclass Project 4 - MA4144

This project contains 5 tasks/questions to be completed, some require written answers. Open a markdown cell below the respective question that require written answers and provide (type) your answers. Questions that required written answers are given in blue fonts. Almost all written questions are open ended, they do not have a correct or wrong answer. You are free to give your opinions, but please provide related answers within the context.

After finishing project run the entire notebook once and **save the notebook as a pdf** (File menu -> Save and Export Notebook As -> PDF). You are **required to upload this PDF on moodle**.

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## Outline of the project

The aim of the project is to practically learn and implement about CNN. This project will have two main sections.

Section 1: Build a convolutional layer and pooling layer from scratch. Then test them on a sample image.

Section 2: Use the Keras library to implement a CNN to classify images on the [CIFAR10 dataset](#).

---

Use the below cell to use any include any imports

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import random
from keras.preprocessing.image import load_img
import keras
from PIL import Image
```

## Section 1: Convolution and Pooling

**Q1** In the following cell, implement a method called `create_padding`. The method will take in `input_image` ( $n \times m$ ) and will return a zero-padded image called `output_image` of dimension  $(n + 2d) \times (m + 2d)$  where  $d$  is the padding thickness on either side.

```
In [13]: def create_padding(input_image, d):

    # Create a new image with the desired padding
    output_image = np.zeros((input_image.shape[0] + 2*d, input_image.shape[1] + 2*d
    #place the original image in the center of the new image
    output_image[d:input_image.shape[0]+d, d:input_image.shape[1]+d] = input_image

    return output_image
```

**Q2** In the following cell, implement a method called convolution. The method will take in input\_image ( $n \times m$ ), kernel ( $k \times k$ ) and will return output\_image of dimension  $(n - k + 1) \times (m - k + 1)$ . The output\_image is the result of the convolution between input\_image and kernel. You may assume that the stride is 1.

```
In [14]: def convolution(input_image, kernel):

    # Define the dimensions
    kernel_height, kernel_width = kernel.shape
    #padded_image = create_padding(input_image, d=0)
    #print(padded_image)
    # Extract sliding windows (image patches) with the same shape as the kernel
    patches = np.lib.stride_tricks.sliding_window_view(input_image, (kernel_height,
    #print(patches.shape)
    # Reshape patches to ensure compatible shapes for einsum
    patches = patches.reshape(patches.shape[0], patches.shape[1], -1)
    #print(patches)
    # Flatten the kernel to match the reshaped patches
    flat_kernel = kernel.ravel()

    # Use einsum to perform element-wise multiplication and summing for convolution
    output_image = np.einsum('ijk,k->ij', patches, flat_kernel)
    #patches*flat_kernel)summation
    return output_image
```

```
In [15]: # Test input image (3x3)
input_image = np.array([
    [1, 2, 0],
    [4, 5, 6],
    [7, 8, 9]
])

# Test kernel (2x2)
kernel = np.array([
    [1, 0],
    [0, -1]
])

# Perform the convolution
output_image = convolution(input_image, kernel)
print(output_image.shape)
print("Output Image:\n", output_image)
```

```
(2, 2)
Output Image:
[[-4 -4]
 [-4 -4]]
```

**Q3** In the following cell, implement a method called pooling. The method will take in input\_image ( $n \times m$ ),  $p$  the pooling dimension, pooling\_type (either max\_pooling or avg\_pooling) and will return output\_image of dimension  $(n - p + 1) \times (m - p + 1)$ . The output\_image is the result of performing pooling on input\_image by a window of dimension  $p \times p$ . You may assume that the stride is 1.

```
In [16]: def pooling(input_image, p, pooling_type = "max_pooling"):

    # Extract sliding windows (pooling patches)
    patches = np.lib.stride_tricks.sliding_window_view(input_image, (p, p))

    if pooling_type == "max_pooling":
        # Perform max pooling
        output_image = np.max(patches, axis=(2, 3))

    elif pooling_type == "avg_pooling":
        # Perform average pooling
        output_image = np.mean(patches, axis=(2, 3))

    else:
        print("Error: Invalid pooling type")
        return None

    return(output_image)
```

```
In [17]: # Example usage:
input_image = np.array([
    [1, 2, 3, 4],
    [5, 6, 7, 8],
    [9, 10, 11, 12],
    [13, 14, 15, 16]
])

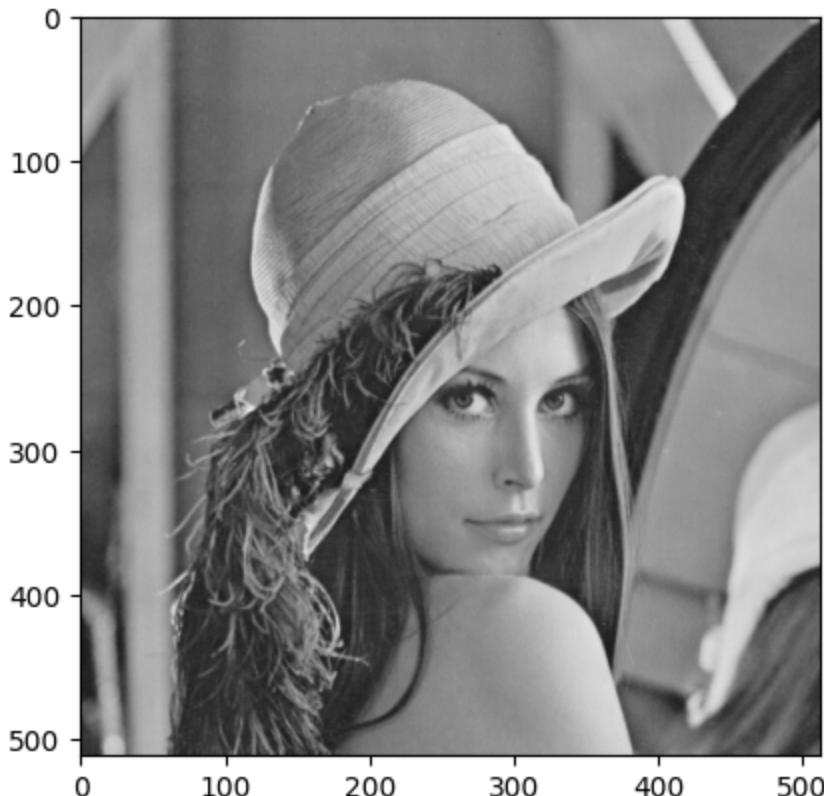
# Perform max pooling with a window size of 2x2
max_pooled_output = pooling(input_image, p=2, pooling_type="max_pooling")
print(max_pooled_output.shape)
print("Max Pooled Output:\n", max_pooled_output)

# Perform average pooling with a window size of 2x2
avg_pooled_output = pooling(input_image, p=2, pooling_type="avg_pooling")
print("Average Pooled Output:\n", avg_pooled_output)
```

```
(3, 3)
Max Pooled Output:
[[ 6  7  8]
 [10 11 12]
 [14 15 16]]
Average Pooled Output:
[[ 3.5  4.5  5.5]
 [ 7.5  8.5  9.5]
 [11.5 12.5 13.5]]
```

The 'lena' image is widely used for image processing experiments and has been a benchmark image until recently. We will use a  $512 \times 512$  grayscale lena sample to test our convolution and pooling implementations.

```
In [18]: lena = load_img('/kaggle/input/lenadat/lena.gif')
plt.imshow(lena)
plt.show()
```



**Q4** In the following perform convolution on lena. Make sure you use padding appropriately to maintain the image size after convolution. However, pooling should be done on an unpadded image and image size may not be preserved after pooling. Use the following kernels to perform convolution separately.

$$1. \begin{bmatrix} +1 & 0 & -1 \\ +1 & 0 & -1 \\ +1 & 0 & -1 \end{bmatrix}$$

2. 
$$\begin{bmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ +1 & +1 & +1 \end{bmatrix}$$

3. 
$$\begin{bmatrix} -1 & 0 & +1 \\ -2 & 0 & +2 \\ -1 & 0 & +1 \end{bmatrix}$$

4. 
$$\begin{bmatrix} +1 & +2 & +1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

5. Any other kernel that you may find interesting.

Explain what the above kernels (including your choice) will do to the image.

### Answer

1.Detects edges in the horizontal direction (i.e., vertical edges).- Sobel X filter

2.Detects edges in the vertical direction (i.e., horizontal edges).- Sobel Y filter

3.Detects edges in all directions and emphasizes regions of rapid intensity change. It is effective for highlighting both vertical and horizontal edges and often produces a more pronounced effect than Sobel filters.

4.Used for edge detection, similar to the 1,2 but with a different weighting scheme. It highlights both vertical and horizontal edges but may produce slightly different results compared to above. It generally enhances edge features while reducing noise.

5.Using this will cause edges to stand out more prominently.Details will be enhanced.Textures within flat regions may become more visible. While it enhances details, it might also amplify noise, especially in low-contrast areas.

6.Blurs the image, reducing noise and detail. It smooths out the pixel intensity variations, which can be useful for preprocessing before edge detection.

```
In [19]: # Convert the image to a numpy array
lena = np.array(lena)
print(lena.shape)
```

```
lena_image_pil = Image.open('/kaggle/input/lenadat/lena.gif')
lena_image = np.array(lena_image_pil.convert('L'))
print(lena_image.shape)
```

```
(512, 512, 3)
(512, 512)
```

```
In [20]: kernels = {
    "Kernel 1: Sobel X": np.array([
        [1, 0, -1],
        [1, 0, -1],
        [1, 0, -1]
    ]),
    "Kernel 2: Sobel Y": np.array([
        [-1, -1, -1],
        [0, 0, 0],
        [1, 1, 1]
    ]),
    "Kernel 3": np.array([
        [-1, 0, 1],
        [-2, 0, 2],
        [-1, 0, 1]
    ]),
    "Kernel 4": np.array([
        [1, 2, 1],
        [0, 0, 0],
        [-1, -2, -1]
    ]),
    "Kernel 5": np.array([
        [0, 1, 0],
        [1, -5, 1],
        [0, 1, 0]
    ]),
    "Gaussian Blur": np.array([
        [1, 2, 1],
        [2, 4, 2],
        [1, 2, 1]
    ]),
}
```

```
In [21]: # Perform convolution for each kernel
output_images = convolution(lena_image, kernels["Kernel 1: Sobel X"]), convolution(
```

Show the resulting image after convolution and pooling separately on two subplots (of the same plot) for each kernel. There should be 5 plots with two sub plots in each.

```
In [22]: for i, (title, kernel) in enumerate(kernels.items()):
    #print(lena_image.shape)
    padded_image = create_padding(lena_image, d=1)
    #print(padded_image.shape)
    # Perform convolution
    output_image_conv = convolution(padded_image, kernel)
    #print(output_image_conv.shape)
    # 2x2 max pooling
    output_image_pool = pooling(output_image_conv, p=2, pooling_type="max_pool")
    #print(output_image_pool.shape)

    plt.figure(figsize=(6, 3))
```

```
plt.subplot(1, 2, 1)
plt.imshow(output_image_conv, cmap='gray')
plt.title(f"{title} Convolution")
plt.axis('off')

# Subplot for pooling result
plt.subplot(1, 2, 2)
plt.imshow(output_image_pool, cmap='gray')
plt.title(f"{title} Pooling")
plt.axis('off')

plt.tight_layout()
plt.show()
```

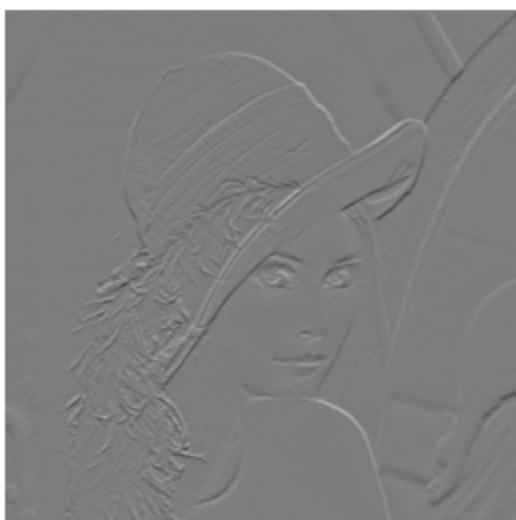
Kernal 1: Sobel X Convolution



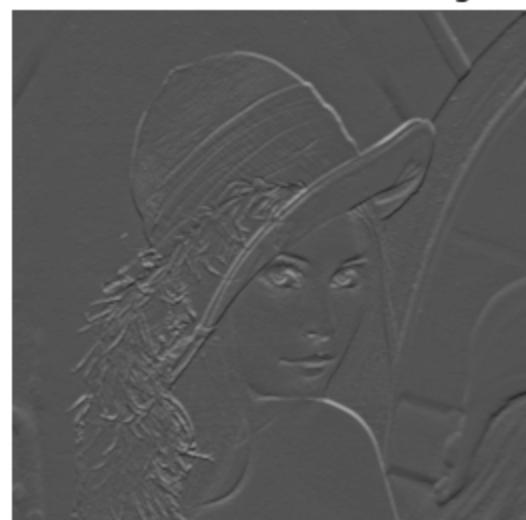
Kernal 1: Sobel X Pooling

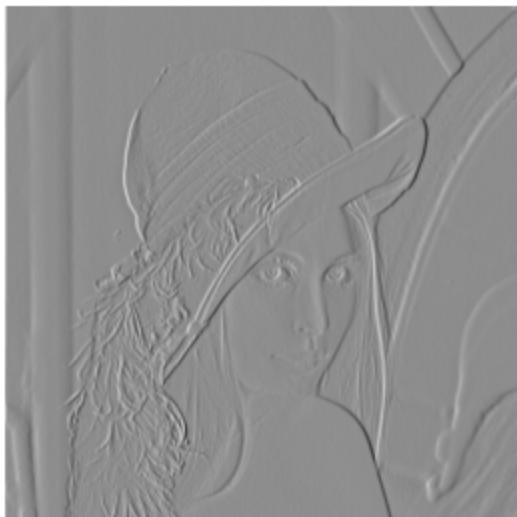
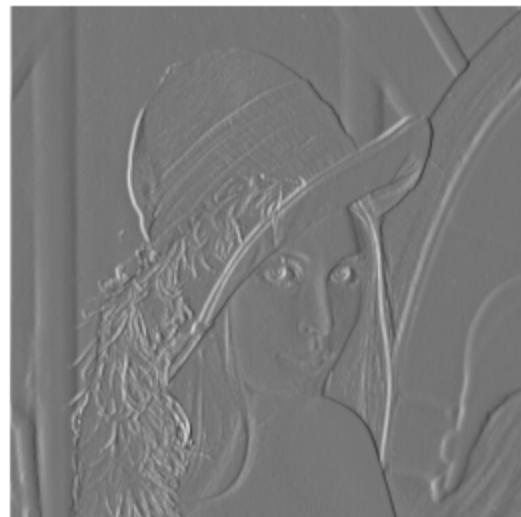
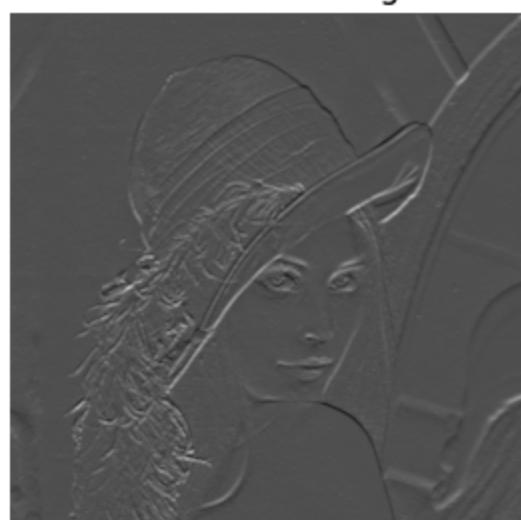
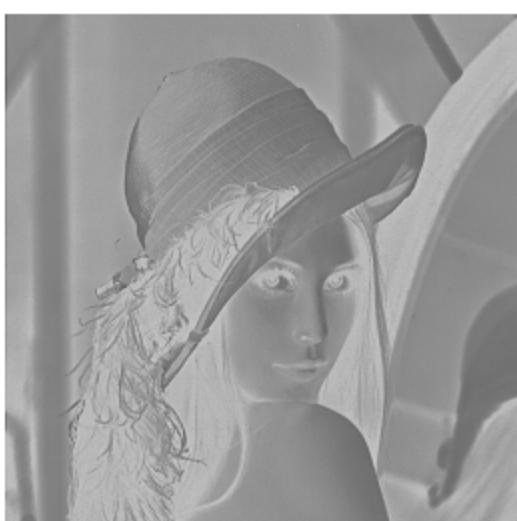


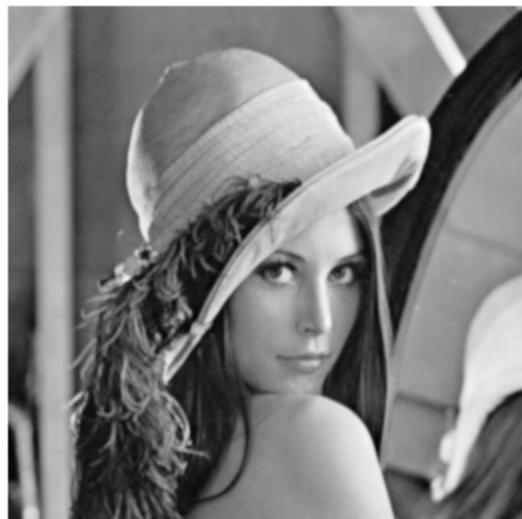
Kernal 2: Sobel Y Convolution



Kernal 2: Sobel Y Pooling



**Kernal 3 Convolution****Kernal 3 Pooling****Kernal 4 Convolution****Kernal 4 Pooling****Kernal 5 Convolution****Kernal 5 Pooling**

**Gaussian Blur Convolution****Gaussian Blur Pooling**

Comment on the results of the above experiment. Mention whether you think the experiment was successful, and what your learnt from it.

### **Answer**

Yes, the experiment was a success. I learnt the how convolution and pooling extract image features, while preserving its spacial dimensions. As in example we did, some kernal extracts vertical features, horizontal feature, diagonal features and so on. Some kernals (Gaussian) smoothes the image, while some other sharpen the image features. Also, I observed how output image sizes vary with kernal sizes, strides, padding, and pooling sizes. Moreover, I learnt about the implementation of numpy einsum function for convolution and pooling operations. I observed how it breaks down the padded input immage into patches, then perform elementwise multiplication with flattened kernal and lastly get the rowwise sum of products, resulting in output convoluted image. The einsum function for max and average pooling was also observed and understood.

## **Section 2: Using Keras to implement CNN for image classification**

This section, unlike the previous projects you are granted full liberty to build the structure of your project appropriately using keras. I have provided only the code to download the cifar10 dataset. After using CNN on the dataset, provide the following. (Note that cifar10 contains rgb images with 3 channels unlike the grayscale image lena we used earlier.)

1. 5-fold cross validation accuracy.
2. Testing accuracy.
3. Confusion matrix of the result.
4. Precision recall for each class.

Note: You are required test on different hyperparameters and network architectures and select decide the best performer based on the cross-validation accuracy.

```
In [2]: from keras.models import Sequential
from keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormal
from keras.regularizers import l2
import keras
from sklearn.model_selection import KFold
import seaborn as sns
from tensorflow.keras.backend import clear_session
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
```

```
In [3]: (x_train, y_train), (x_test, y_test) = keras.datasets.cifar10.load_data()
```

Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>  
170498071/170498071 ————— 3s 0us/step

```
In [4]: print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
```

```
(50000, 32, 32, 3)
(50000, 1)
(10000, 32, 32, 3)
(10000, 1)
```

```
In [5]: class_names = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']

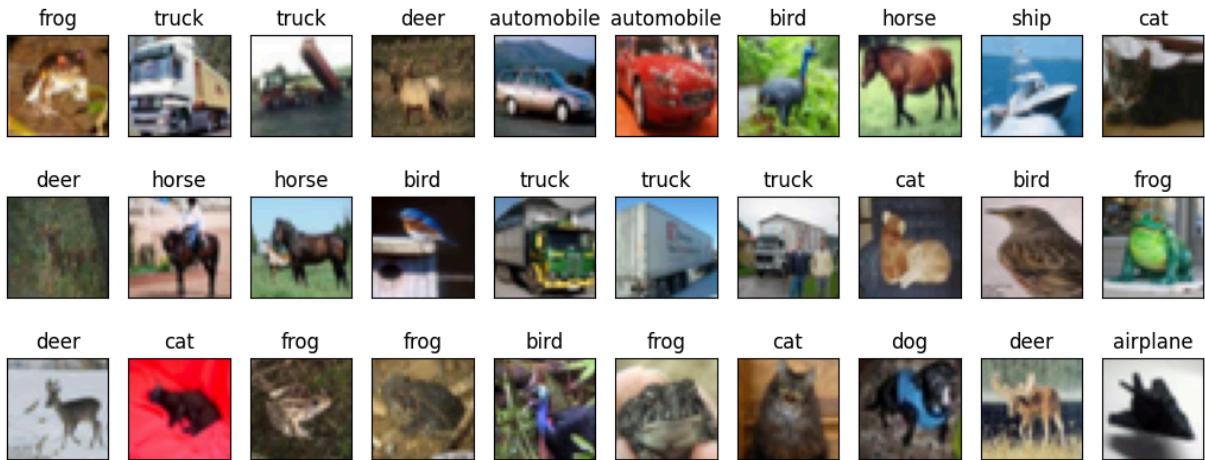
# Create a new figure
plt.figure(figsize=(13,5))

# Loop over the first 25 images
for i in range(30):
    # Create a subplot for each image
    plt.subplot(3, 10, i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)

    # Display the image
    plt.imshow(x_train[i])

    # Set the label as the title
    plt.title(class_names[y_train[i][0]], fontsize=12)

# Display the figure
plt.show()
```



```
In [6]: #identify the unique classes in the dataset
unique_classes = np.unique(y_train)
print(len(unique_classes))
#there are 10 unique classes in the dataset
```

10

```
In [7]: #one hot encode the target values
from keras.utils import to_categorical
# one hot encode target values
y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
```

```
In [8]: #preprocess the input data
def preprocess_input(x):
    #convert the pixel values to float
    x = x.astype('float32')

    #normalize the pixel values
    x_mean = np.mean(x)
    x_std = np.std(x)

    x = (x - x_mean) / x_std
    #resize the images to 32x32
    #x = np.array([np.array(Image.fromarray(image).resize((32, 32))) for image in x])

    return x
```

```
In [9]: def define_model(architecture="baseline", learning_rate=0.0005):

    if architecture == "baseline":
        # VGG-style architecture with 2 blocks
        model = Sequential()
        model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))
        model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))
        model.add(MaxPooling2D((2, 2)))

        model.add(Flatten())
        model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))
```

```
model.add(Dense(10, activation='softmax'))\n\n\nelif architecture == "vgg3block":\n    # VGG-style architecture with 3 blocks\n    # example of a 3-block vgg style architecture\n    model = Sequential()\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n\n    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n\n    model.add(Flatten())\n    model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))\n\n    model.add(Dense(10, activation='softmax'))\n\n\nelif architecture == "with_dropout_0.2":\n    model = Sequential()\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n    model.add(Dropout(0.2))\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n    model.add(Dropout(0.2))\n    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n    model.add(Dropout(0.2))\n    model.add(Flatten())\n    model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))\n    model.add(Dropout(0.2))\n    model.add(Dense(10, activation='softmax'))\n\n\nelif architecture == "with_dropout_increasing":\n    model = Sequential()\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n    model.add(Dropout(0.2))\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))\n    model.add(MaxPooling2D((2, 2)))\n    model.add(Dropout(0.3))
```

```

model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uni'))
model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uni'))
model.add(MaxPooling2D((2, 2)))
model.add(Dropout(0.4))
model.add(Flatten())
model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))
model.add(Dropout(0.5))
model.add(Dense(10, activation='softmax'))

elif architecture == "custom":
    # Custom architecture
    model = Sequential()
    model.add(Conv2D(32, (3, 3), activation='relu', padding='same', input_shape=)
    model.add(MaxPooling2D((2, 2)))

    model.add(Conv2D(64, (3, 3), activation='relu', padding='same'))
    model.add(MaxPooling2D((2, 2)))

    model.add(Flatten())
    model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))

    model.add(Dense(10, activation='softmax'))

elif architecture == "with_batch_norm":
    model = Sequential()
    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2, 2)))
    model.add(Dropout(0.2))
    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2, 2)))
    model.add(Dropout(0.3))
    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uni'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2, 2)))
    model.add(Dropout(0.4))
    model.add(Flatten())
    model.add(Dense(128, activation='relu', kernel_initializer='he_uniform'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(10, activation='softmax'))

else:
    raise ValueError("Architecture not recognized. Choose from 'vgg2block', 'vgg16block' or 'custom'")

# Add the fully connected layers

```

```
# Compile the model with Adam optimizer
optimizer_instance = keras.optimizers.Adam(learning_rate=learning_rate)
model.compile(optimizer=optimizer_instance, loss='categorical_crossentropy', metrics=['accuracy'])

return model
```

In [10]:

```
x_train = preprocess_input(x_train)
x_test = preprocess_input(x_test)
```

In [11]:

```
#evaluate the model

#perform 5 fold cross validation
```

```
def CrossVal_Training(archi,epochs,batch_size):
    # Original data (do not reassign within the loop)
    original_x_train, original_y_train = x_train, y_train
    # Initialize the KFold object with 5 splits
    kf = KFold(n_splits=5, shuffle=True, random_state=42)

    # List to store accuracy for each fold
    fold_accuracies = []
    histories = []
    # Initialize an array to store predictions for each fold
    all_predictions = np.zeros((x_test.shape[0], 10)) # 10 classes for CIFAR-10
    # Perform 5-fold cross-validation
    for train_index, val_index in kf.split(x_train):
        clear_session()

        # Define and train the model for the current fold
        model = define_model(architecture=archi)
        history = model.fit(x_train, y_train, epochs=epochs, batch_size=batch_size,
                             histories.append(history))
        # Evaluate the model on the validation set
        val_loss, val_accuracy = model.evaluate(x_train[val_index], y_train[val_index])
        fold_accuracies.append(val_accuracy)
        print(f"\nFold Validation Accuracy: {val_accuracy * 100:.2f}%")

    # Calculate and print the average accuracy across all folds
    average_accuracy = np.mean(fold_accuracies)
    print(f"\nAverage 5-Fold Cross-Validation Accuracy: {average_accuracy * 100:.2f}%")

    return model,histories
```

In [12]:

```
#evaluate on test set
def evaluate_on_test(model):

    model.fit(x_train, y_train, batch_size=64, epochs=10, verbose=1)

    test_loss, test_acc = model.evaluate(x_test, y_test, verbose=1)

    print('\nTest Accuracy:', test_acc)
```

```
    print('Test Loss:      ', test_loss)

    return test_acc
```

```
In [13]: def PlotDetails(histories):
    plt.figure(figsize=(15, 6))

    # Plot training and validation loss for each fold
    plt.subplot(1, 2, 1)
    for i, history in enumerate(histories):
        plt.plot(history.history['loss'], label=f'Fold {i+1} Train Loss', alpha=0.6)
        plt.plot(history.history['val_loss'], label=f'Fold {i+1} Val Loss', alpha=0.6)
    plt.legend()
    plt.title('Loss Evolution Across Folds')

    # Plot training and validation accuracy for each fold
    plt.subplot(1, 2, 2)
    for i, history in enumerate(histories):
        plt.plot(history.history['accuracy'], label=f'Fold {i+1} Train Accuracy', alpha=0.6)
        plt.plot(history.history['val_accuracy'], label=f'Fold {i+1} Val Accuracy', alpha=0.6)
    plt.legend()
    plt.title('Accuracy Evolution Across Folds')
```

```
In [58]: model,histories= CrossVal_Training("baseline",10,64)
```

Epoch 1/10  
782/782 6s 5ms/step - accuracy: 0.3924 - loss: 1.7637 - val\_accuracy: 0.6179 - val\_loss: 1.1030  
Epoch 2/10  
782/782 2s 3ms/step - accuracy: 0.6256 - loss: 1.0789 - val\_accuracy: 0.6892 - val\_loss: 0.8957  
Epoch 3/10  
782/782 2s 3ms/step - accuracy: 0.6861 - loss: 0.9027 - val\_accuracy: 0.7330 - val\_loss: 0.7775  
Epoch 4/10  
782/782 2s 3ms/step - accuracy: 0.7325 - loss: 0.7755 - val\_accuracy: 0.7720 - val\_loss: 0.6615  
Epoch 5/10  
782/782 3s 3ms/step - accuracy: 0.7651 - loss: 0.6845 - val\_accuracy: 0.8069 - val\_loss: 0.5713  
Epoch 6/10  
782/782 2s 3ms/step - accuracy: 0.7959 - loss: 0.6006 - val\_accuracy: 0.8339 - val\_loss: 0.5019  
Epoch 7/10  
782/782 2s 3ms/step - accuracy: 0.8209 - loss: 0.5249 - val\_accuracy: 0.8701 - val\_loss: 0.4196  
Epoch 8/10  
782/782 2s 3ms/step - accuracy: 0.8459 - loss: 0.4556 - val\_accuracy: 0.8905 - val\_loss: 0.3581  
Epoch 9/10  
782/782 2s 3ms/step - accuracy: 0.8679 - loss: 0.3915 - val\_accuracy: 0.9019 - val\_loss: 0.3202  
Epoch 10/10  
782/782 2s 3ms/step - accuracy: 0.8887 - loss: 0.3346 - val\_accuracy: 0.9140 - val\_loss: 0.2820  
313/313 1s 1ms/step - accuracy: 0.9163 - loss: 0.2826  
Fold Validation Accuracy: 91.40%  
Epoch 1/10  
782/782 5s 5ms/step - accuracy: 0.4342 - loss: 1.6239 - val\_accuracy: 0.6506 - val\_loss: 1.0025  
Epoch 2/10  
782/782 2s 3ms/step - accuracy: 0.6647 - loss: 0.9736 - val\_accuracy: 0.7156 - val\_loss: 0.8189  
Epoch 3/10  
782/782 2s 3ms/step - accuracy: 0.7287 - loss: 0.7894 - val\_accuracy: 0.7672 - val\_loss: 0.6747  
Epoch 4/10  
782/782 2s 3ms/step - accuracy: 0.7725 - loss: 0.6658 - val\_accuracy: 0.8368 - val\_loss: 0.5124  
Epoch 5/10  
782/782 2s 3ms/step - accuracy: 0.8130 - loss: 0.5440 - val\_accuracy: 0.8694 - val\_loss: 0.4226  
Epoch 6/10  
782/782 2s 3ms/step - accuracy: 0.8550 - loss: 0.4318 - val\_accuracy: 0.8964 - val\_loss: 0.3298  
Epoch 7/10  
782/782 2s 3ms/step - accuracy: 0.8899 - loss: 0.3380 - val\_accuracy: 0.9311 - val\_loss: 0.2419  
Epoch 8/10  
782/782 2s 3ms/step - accuracy: 0.9228 - loss: 0.2431 - val\_accuracy: 0.9529 - val\_loss: 0.1733

Epoch 9/10  
**782/782** 2s 3ms/step - accuracy: 0.9434 - loss: 0.1806 - val\_accuracy: 0.9660 - val\_loss: 0.1354  
Epoch 10/10  
**782/782** 2s 3ms/step - accuracy: 0.9634 - loss: 0.1276 - val\_accuracy: 0.9707 - val\_loss: 0.1040  
**313/313** 1s 1ms/step - accuracy: 0.9695 - loss: 0.1056  
Fold Validation Accuracy: 97.07%  
Epoch 1/10  
**782/782** 5s 5ms/step - accuracy: 0.3920 - loss: 1.7884 - val\_accuracy: 0.6109 - val\_loss: 1.1208  
Epoch 2/10  
**782/782** 2s 3ms/step - accuracy: 0.6144 - loss: 1.0984 - val\_accuracy: 0.6795 - val\_loss: 0.9288  
Epoch 3/10  
**782/782** 2s 3ms/step - accuracy: 0.6784 - loss: 0.9261 - val\_accuracy: 0.7055 - val\_loss: 0.8437  
Epoch 4/10  
**782/782** 2s 3ms/step - accuracy: 0.7138 - loss: 0.8180 - val\_accuracy: 0.7371 - val\_loss: 0.7631  
Epoch 5/10  
**782/782** 2s 3ms/step - accuracy: 0.7495 - loss: 0.7247 - val\_accuracy: 0.7794 - val\_loss: 0.6454  
Epoch 6/10  
**782/782** 2s 3ms/step - accuracy: 0.7789 - loss: 0.6441 - val\_accuracy: 0.8036 - val\_loss: 0.5750  
Epoch 7/10  
**782/782** 2s 3ms/step - accuracy: 0.7959 - loss: 0.5859 - val\_accuracy: 0.8326 - val\_loss: 0.4987  
Epoch 8/10  
**782/782** 2s 3ms/step - accuracy: 0.8240 - loss: 0.5191 - val\_accuracy: 0.8535 - val\_loss: 0.4438  
Epoch 9/10  
**782/782** 2s 3ms/step - accuracy: 0.8417 - loss: 0.4626 - val\_accuracy: 0.8647 - val\_loss: 0.4001  
Epoch 10/10  
**782/782** 2s 3ms/step - accuracy: 0.8644 - loss: 0.3965 - val\_accuracy: 0.8816 - val\_loss: 0.3477  
**313/313** 1s 1ms/step - accuracy: 0.8802 - loss: 0.3493  
Fold Validation Accuracy: 88.16%  
Epoch 1/10  
**782/782** 5s 5ms/step - accuracy: 0.3569 - loss: 1.9962 - val\_accuracy: 0.5968 - val\_loss: 1.1526  
Epoch 2/10  
**782/782** 2s 3ms/step - accuracy: 0.6099 - loss: 1.1135 - val\_accuracy: 0.6950 - val\_loss: 0.8933  
Epoch 3/10  
**782/782** 2s 3ms/step - accuracy: 0.6837 - loss: 0.9114 - val\_accuracy: 0.7326 - val\_loss: 0.7892  
Epoch 4/10  
**782/782** 2s 3ms/step - accuracy: 0.7262 - loss: 0.7824 - val\_accuracy: 0.7859 - val\_loss: 0.6507  
Epoch 5/10  
**782/782** 2s 3ms/step - accuracy: 0.7620 - loss: 0.6875 - val\_accuracy: 0.7976 - val\_loss: 0.5822  
Epoch 6/10

```
782/782 2s 3ms/step - accuracy: 0.7903 - loss: 0.6044 - val_accuracy: 0.8372 - val_loss: 0.4893
Epoch 7/10
782/782 2s 3ms/step - accuracy: 0.8229 - loss: 0.5200 - val_accuracy: 0.8572 - val_loss: 0.4314
Epoch 8/10
782/782 2s 3ms/step - accuracy: 0.8491 - loss: 0.4465 - val_accuracy: 0.8720 - val_loss: 0.3787
Epoch 9/10
782/782 2s 3ms/step - accuracy: 0.8730 - loss: 0.3790 - val_accuracy: 0.9133 - val_loss: 0.2879
Epoch 10/10
782/782 3s 2ms/step - accuracy: 0.8948 - loss: 0.3133 - val_accuracy: 0.9381 - val_loss: 0.2198
313/313 1s 2ms/step - accuracy: 0.9444 - loss: 0.2066
Fold Validation Accuracy: 93.81%
Epoch 1/10
782/782 5s 5ms/step - accuracy: 0.3918 - loss: 1.8147 - val_accuracy: 0.6095 - val_loss: 1.1104
Epoch 2/10
782/782 3s 3ms/step - accuracy: 0.6294 - loss: 1.0536 - val_accuracy: 0.6918 - val_loss: 0.8862
Epoch 3/10
782/782 3s 3ms/step - accuracy: 0.6925 - loss: 0.8890 - val_accuracy: 0.7346 - val_loss: 0.7843
Epoch 4/10
782/782 3s 3ms/step - accuracy: 0.7297 - loss: 0.7819 - val_accuracy: 0.7699 - val_loss: 0.6871
Epoch 5/10
782/782 3s 3ms/step - accuracy: 0.7598 - loss: 0.6925 - val_accuracy: 0.7851 - val_loss: 0.6313
Epoch 6/10
782/782 3s 3ms/step - accuracy: 0.7893 - loss: 0.6164 - val_accuracy: 0.8120 - val_loss: 0.5500
Epoch 7/10
782/782 3s 3ms/step - accuracy: 0.8140 - loss: 0.5397 - val_accuracy: 0.8610 - val_loss: 0.4297
Epoch 8/10
782/782 3s 3ms/step - accuracy: 0.8407 - loss: 0.4634 - val_accuracy: 0.8881 - val_loss: 0.3629
Epoch 9/10
782/782 3s 3ms/step - accuracy: 0.8712 - loss: 0.3903 - val_accuracy: 0.9041 - val_loss: 0.3077
Epoch 10/10
782/782 3s 3ms/step - accuracy: 0.8887 - loss: 0.3319 - val_accuracy: 0.9206 - val_loss: 0.2593
313/313 1s 1ms/step - accuracy: 0.9221 - loss: 0.2585
Fold Validation Accuracy: 92.06%
```

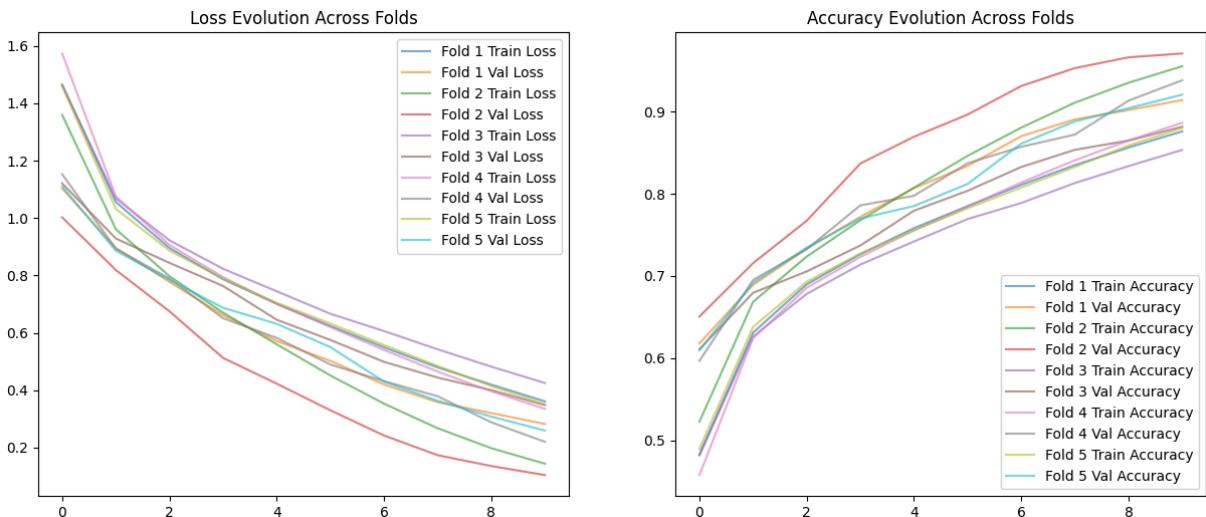
Average 5-Fold Cross-Validation Accuracy: 92.50%

In [59]: `test_accuracy=evaluate_on_test(model)`  
`PlotDetails(histories)`

Epoch 1/10  
**782/782** 2s 3ms/step - accuracy: 0.9147 - loss: 0.2663  
Epoch 2/10  
**782/782** 2s 3ms/step - accuracy: 0.9282 - loss: 0.2252  
Epoch 3/10  
**782/782** 2s 3ms/step - accuracy: 0.9447 - loss: 0.1772  
Epoch 4/10  
**782/782** 2s 3ms/step - accuracy: 0.9564 - loss: 0.1430  
Epoch 5/10  
**782/782** 2s 3ms/step - accuracy: 0.9663 - loss: 0.1140  
Epoch 6/10  
**782/782** 2s 3ms/step - accuracy: 0.9753 - loss: 0.0916  
Epoch 7/10  
**782/782** 2s 3ms/step - accuracy: 0.9741 - loss: 0.0865  
Epoch 8/10  
**782/782** 2s 3ms/step - accuracy: 0.9774 - loss: 0.0730  
Epoch 9/10  
**782/782** 2s 3ms/step - accuracy: 0.9804 - loss: 0.0657  
Epoch 10/10  
**782/782** 2s 3ms/step - accuracy: 0.9840 - loss: 0.0556  
**313/313** 0s 1ms/step - accuracy: 0.6391 - loss: 2.5234

Test Accuracy: 0.6381999850273132

Test Loss: 2.5113935470581055



With the baseline model, achieved an accuracy of 63.8%.

```
In [19]: def plotConfusionMatrix(model,plotTrue=True):
    # Make predictions on the test set
    predictions = model.predict(x_test)
    predicted_classes = np.argmax(predictions, axis=1)
    true_classes = np.argmax(y_test, axis=1)

    # Calculate the confusion matrix
    cm = confusion_matrix(true_classes, predicted_classes)
    if(plotTrue):
        # Plot the confusion matrix
        plt.figure(figsize=(10, 8))
        sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False,
                    xticklabels=np.arange(10), yticklabels=np.arange(10))
```

```
plt.title('Confusion Matrix')
plt.xlabel('Predicted Class')
plt.ylabel('True Class')
plt.show()

return true_classes,predicted_classes,cm
```

```
In [17]: def GenerateClassificationReport(model):
    true_classes,predicted_classes,cm= plotConfusionMatrix(model,plotTrue=True)
    # Generate a classification report
    report = classification_report(true_classes, predicted_classes, target_names=[s
    print(report)

    # Initialize lists to store precision and recall
    precision = []
    recall = []

    # Calculate precision and recall for each class
    for i in range(cm.shape[0]):
        tp = cm[i, i] # True Positives
        fp = np.sum(cm[:, i]) - tp # False Positives
        fn = np.sum(cm[i, :]) - tp # False Negatives

        precision_class = tp / (tp + fp) if (tp + fp) > 0 else 0
        recall_class = tp / (tp + fn) if (tp + fn) > 0 else 0

        precision.append(precision_class)
        recall.append(recall_class)

    # Print precision and recall for each class
    for i in range(len(precision)):
        print(f"Class {i}: Precision = {precision[i]:.2f}, Recall = {recall[i]:.2f}")
```

```
In [60]: GenerateClassificationReport(model)
```

```
313/313 ━━━━━━━━ 1s 2ms/step
```

Confusion Matrix										
True Class	0	1	2	3	4	5	6	7	8	9
	785	24	38	25	12	8	5	7	51	45
	65	758	6	13	3	6	5	1	26	117
	131	13	470	82	95	71	55	46	9	28
	58	31	60	455	70	197	52	41	9	27
	64	11	82	80	573	48	37	85	10	10
	47	13	57	192	39	557	19	54	4	18
	20	24	49	110	77	47	639	6	8	20
	40	5	26	48	67	61	5	709	6	33
	157	65	9	15	5	8	3	8	679	51
	64	106	11	12	7	4	1	12	26	757

	precision	recall	f1-score	support
0	0.55	0.79	0.65	1000
1	0.72	0.76	0.74	1000
2	0.58	0.47	0.52	1000
3	0.44	0.46	0.45	1000
4	0.60	0.57	0.59	1000
5	0.55	0.56	0.56	1000
6	0.78	0.64	0.70	1000
7	0.73	0.71	0.72	1000
8	0.82	0.68	0.74	1000
9	0.68	0.76	0.72	1000
accuracy			0.64	10000
macro avg	0.65	0.64	0.64	10000
weighted avg	0.65	0.64	0.64	10000

Class 0: Precision = 0.55, Recall = 0.79  
 Class 1: Precision = 0.72, Recall = 0.76  
 Class 2: Precision = 0.58, Recall = 0.47  
 Class 3: Precision = 0.44, Recall = 0.46  
 Class 4: Precision = 0.60, Recall = 0.57  
 Class 5: Precision = 0.55, Recall = 0.56  
 Class 6: Precision = 0.78, Recall = 0.64  
 Class 7: Precision = 0.73, Recall = 0.71  
 Class 8: Precision = 0.82, Recall = 0.68  
 Class 9: Precision = 0.68, Recall = 0.76

## Hyperparameter Tuning

In [42]:

```
import itertools
from tqdm import tqdm # Import tqdm for progress bar

save_dir = '/saved_models/'

# Function to perform cross-validation manually
def cross_validate_model(x_train, y_train, params, n_splits=5):
    kfold = KFold(n_splits=n_splits, shuffle=True, random_state=42)
    scores = []

    # Log the start of cross-validation for the current parameter set
    print(f"Starting cross-validation with parameters: {params}")

    for train_index, val_index in tqdm(kfold.split(x_train, y_train), desc="Cross-validation"):
        x_train_fold, x_val_fold = x_train[train_index], x_train[val_index]
        y_train_fold, y_val_fold = y_train[train_index], y_train[val_index]

        model = define_model(
            architecture=params['architecture'],
            learning_rate=params['learning_rate'],
        )
        model.fit(x_train_fold, y_train_fold, epochs=params['epochs'], batch_size=params['batch_size'])

        score = model.evaluate(x_val_fold, y_val_fold, verbose=0)
        scores.append(score[1]) # Append accuracy

    return np.mean(scores)
```

```

# Example parameter grid
param_grid = {
    'architecture': ['vgg3block', 'with_dropout_increasing', 'custom'],
    'batch_size': [64],
    'epochs': [10, 20],
    'learning_rate': [0.001, 0.0005]
}

# To track the best hyperparameters
best_mean_accuracy = 0
best_params = {}
save_model=None

# Generate combinations of hyperparameters
keys, values = zip(*param_grid.items())
param_combinations = [dict(zip(keys, v)) for v in itertools.product(*values)]

# Iterate over all parameter combinations
for params in param_combinations:
    mean_accuracy = cross_validate_model(x_train, y_train, params)
    print(f"Params: {params} -> Mean accuracy: {mean_accuracy:.2f}")

    # Check if this is the best model
    if mean_accuracy > best_mean_accuracy:
        best_mean_accuracy = mean_accuracy
        best_params = params
        save_model=model

    save_model.save('/kaggle/working/last_best_model.h5')
# Print the best hyperparameters found
print(f"Best Hyperparameters: {best_params}")
print(f"Best Mean Accuracy: {best_mean_accuracy:.2f}")

```

Starting cross-validation with parameters: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 10, 'learning\_rate': 0.001}

Params: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 10, 'learning\_rate': 0.001} -> Mean accuracy: 0.74

Starting cross-validation with parameters: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 10, 'learning\_rate': 0.0005}

Params: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 10, 'learning\_rate': 0.0005} -> Mean accuracy: 0.73

Starting cross-validation with parameters: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 20, 'learning\_rate': 0.001}

Params: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 20, 'learning\_rate': 0.001} -> Mean accuracy: 0.74

Starting cross-validation with parameters: {'architecture': 'vgg3block', 'batch\_size': 64, 'epochs': 20, 'learning\_rate': 0.0005}

```
Params: {'architecture': 'vgg3block', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005} -> Mean accuracy: 0.73
```

```
Starting cross-validation with parameters: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.001}
```

```
Params: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.001} -> Mean accuracy: 0.75
```

```
Starting cross-validation with parameters: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.0005}
```

```
Params: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.0005} -> Mean accuracy: 0.75
```

```
Starting cross-validation with parameters: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.001}
```

```
Params: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.001} -> Mean accuracy: 0.79
```

```
Starting cross-validation with parameters: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005}
```

```
Params: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005} -> Mean accuracy: 0.80
```

```
Starting cross-validation with parameters: {'architecture': 'custom', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.001}
```

```
Params: {'architecture': 'custom', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.001} -> Mean accuracy: 0.70
```

```
Starting cross-validation with parameters: {'architecture': 'custom', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.0005}
```

```
Params: {'architecture': 'custom', 'batch_size': 64, 'epochs': 10, 'learning_rate': 0.0005} -> Mean accuracy: 0.71
```

```
Starting cross-validation with parameters: {'architecture': 'custom', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.001}
```

```
Params: {'architecture': 'custom', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.001} -> Mean accuracy: 0.69
```

```
Starting cross-validation with parameters: {'architecture': 'custom', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005}
```

```
Params: {'architecture': 'custom', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005} -> Mean accuracy: 0.70
```

```
Best Hyperparameters: {'architecture': 'with_dropout_increasing', 'batch_size': 64, 'epochs': 20, 'learning_rate': 0.0005}
```

```
Best Mean Accuracy: 0.80
```

```
In [45]: evaluate_on_test(save_model)
```

```
Epoch 1/10
782/782 ----- 2s 3ms/step - accuracy: 0.9888 - loss: 0.0332
Epoch 2/10
782/782 ----- 2s 3ms/step - accuracy: 0.9905 - loss: 0.0307
Epoch 3/10
782/782 ----- 2s 3ms/step - accuracy: 0.9917 - loss: 0.0261
Epoch 4/10
782/782 ----- 2s 3ms/step - accuracy: 0.9926 - loss: 0.0227
Epoch 5/10
782/782 ----- 2s 3ms/step - accuracy: 0.9927 - loss: 0.0222
Epoch 6/10
782/782 ----- 2s 3ms/step - accuracy: 0.9893 - loss: 0.0310
Epoch 7/10
782/782 ----- 2s 3ms/step - accuracy: 0.9935 - loss: 0.0227
Epoch 8/10
782/782 ----- 2s 3ms/step - accuracy: 0.9924 - loss: 0.0237
Epoch 9/10
782/782 ----- 2s 3ms/step - accuracy: 0.9935 - loss: 0.0215
Epoch 10/10
782/782 ----- 2s 3ms/step - accuracy: 0.9942 - loss: 0.0179
313/313 ----- 0s 1ms/step - accuracy: 0.6320 - loss: 4.8893
```

Test Accuracy: 0.6276000142097473

Test Loss: 4.984678745269775

Out[45]: 0.6276000142097473

In [54]: `GenerateClassificationReport(save_model)`

**313/313** ----- **0s** 1ms/step

Confusion Matrix										
	0	1	2	3	4	5	6	7	8	9
True Class	656	27	65	35	31	20	17	12	97	40
0	24	729	9	22	7	14	15	3	53	124
1	80	12	478	76	99	89	89	44	22	11
2	26	14	74	434	78	196	103	35	22	18
3	19	2	117	78	545	57	97	62	16	7
4	14	8	71	180	50	555	52	50	14	6
5	4	8	66	63	53	31	752	6	13	4
6	19	9	46	61	68	105	17	653	7	15
7	78	52	18	22	9	13	6	5	762	35
8	42	103	17	15	9	10	12	21	59	712
Predicted Class	0	1	2	3	4	5	6	7	8	9

	precision	recall	f1-score	support
0	0.68	0.66	0.67	1000
1	0.76	0.73	0.74	1000
2	0.50	0.48	0.49	1000
3	0.44	0.43	0.44	1000
4	0.57	0.55	0.56	1000
5	0.51	0.56	0.53	1000
6	0.65	0.75	0.70	1000
7	0.73	0.65	0.69	1000
8	0.72	0.76	0.74	1000
9	0.73	0.71	0.72	1000
accuracy			0.63	10000
macro avg	0.63	0.63	0.63	10000
weighted avg	0.63	0.63	0.63	10000

Class 0: Precision = 0.68, Recall = 0.66  
 Class 1: Precision = 0.76, Recall = 0.73  
 Class 2: Precision = 0.50, Recall = 0.48  
 Class 3: Precision = 0.44, Recall = 0.43  
 Class 4: Precision = 0.57, Recall = 0.55  
 Class 5: Precision = 0.51, Recall = 0.56  
 Class 6: Precision = 0.65, Recall = 0.75  
 Class 7: Precision = 0.73, Recall = 0.65  
 Class 8: Precision = 0.72, Recall = 0.76  
 Class 9: Precision = 0.73, Recall = 0.71

from the above parameter tuning, we didn't receive a good enough accuracy, therefore we select our best hyperparameter set and conduct training for more epochs and more network architectures.

Best Hyperparameters: {'architecture': 'with\_dropout\_increasing', 'batch\_size': 64, 'epochs': 20, 'learning\_rate': 0.0005}

In [62]: model1,histories= CrossVal\_Training("with\_dropout\_increasing",50,64)

Epoch 1/50  
**782/782** 20s 15ms/step - accuracy: 0.2177 - loss: 2.1992 - val\_accuracy: 0.4668 - val\_loss: 1.4478  
Epoch 2/50  
**782/782** 5s 6ms/step - accuracy: 0.4417 - loss: 1.5235 - val\_accuracy: 0.5657 - val\_loss: 1.1957  
Epoch 3/50  
**782/782** 4s 5ms/step - accuracy: 0.5360 - loss: 1.2896 - val\_accuracy: 0.6480 - val\_loss: 0.9969  
Epoch 4/50  
**782/782** 4s 5ms/step - accuracy: 0.6032 - loss: 1.1283 - val\_accuracy: 0.6749 - val\_loss: 0.9161  
Epoch 5/50  
**782/782** 4s 5ms/step - accuracy: 0.6416 - loss: 1.0202 - val\_accuracy: 0.7207 - val\_loss: 0.7846  
Epoch 6/50  
**782/782** 4s 5ms/step - accuracy: 0.6761 - loss: 0.9244 - val\_accuracy: 0.7547 - val\_loss: 0.6840  
Epoch 7/50  
**782/782** 4s 6ms/step - accuracy: 0.6963 - loss: 0.8717 - val\_accuracy: 0.7600 - val\_loss: 0.6785  
Epoch 8/50  
**782/782** 5s 6ms/step - accuracy: 0.7121 - loss: 0.8276 - val\_accuracy: 0.7793 - val\_loss: 0.6327  
Epoch 9/50  
**782/782** 5s 6ms/step - accuracy: 0.7285 - loss: 0.7862 - val\_accuracy: 0.8114 - val\_loss: 0.5292  
Epoch 10/50  
**782/782** 5s 6ms/step - accuracy: 0.7471 - loss: 0.7355 - val\_accuracy: 0.8135 - val\_loss: 0.5288  
Epoch 11/50  
**782/782** 5s 6ms/step - accuracy: 0.7529 - loss: 0.7103 - val\_accuracy: 0.8189 - val\_loss: 0.5248  
Epoch 12/50  
**782/782** 5s 6ms/step - accuracy: 0.7687 - loss: 0.6725 - val\_accuracy: 0.8310 - val\_loss: 0.4703  
Epoch 13/50  
**782/782** 5s 6ms/step - accuracy: 0.7748 - loss: 0.6577 - val\_accuracy: 0.8387 - val\_loss: 0.4633  
Epoch 14/50  
**782/782** 5s 6ms/step - accuracy: 0.7828 - loss: 0.6286 - val\_accuracy: 0.8541 - val\_loss: 0.4344  
Epoch 15/50  
**782/782** 5s 6ms/step - accuracy: 0.7902 - loss: 0.6123 - val\_accuracy: 0.8479 - val\_loss: 0.4306  
Epoch 16/50  
**782/782** 5s 6ms/step - accuracy: 0.7947 - loss: 0.5905 - val\_accuracy: 0.8601 - val\_loss: 0.3922  
Epoch 17/50  
**782/782** 5s 6ms/step - accuracy: 0.7978 - loss: 0.5782 - val\_accuracy: 0.8764 - val\_loss: 0.3653  
Epoch 18/50  
**782/782** 5s 6ms/step - accuracy: 0.8065 - loss: 0.5567 - val\_accuracy: 0.8780 - val\_loss: 0.3479  
Epoch 19/50  
**782/782** 5s 6ms/step - accuracy: 0.8116 - loss: 0.5428 - val\_accuracy:

```
uracy: 0.8813 - val_loss: 0.3364
Epoch 20/50
782/782 5s 6ms/step - accuracy: 0.8154 - loss: 0.5276 - val_acc
uracy: 0.8848 - val_loss: 0.3322
Epoch 21/50
782/782 5s 6ms/step - accuracy: 0.8164 - loss: 0.5263 - val_acc
uracy: 0.8888 - val_loss: 0.3285
Epoch 22/50
782/782 5s 6ms/step - accuracy: 0.8187 - loss: 0.5191 - val_acc
uracy: 0.8965 - val_loss: 0.3074
Epoch 23/50
782/782 5s 6ms/step - accuracy: 0.8243 - loss: 0.5021 - val_acc
uracy: 0.8997 - val_loss: 0.2937
Epoch 24/50
782/782 5s 6ms/step - accuracy: 0.8287 - loss: 0.4842 - val_acc
uracy: 0.9049 - val_loss: 0.2771
Epoch 25/50
782/782 5s 6ms/step - accuracy: 0.8319 - loss: 0.4838 - val_acc
uracy: 0.8991 - val_loss: 0.2858
Epoch 26/50
782/782 5s 6ms/step - accuracy: 0.8368 - loss: 0.4688 - val_acc
uracy: 0.9105 - val_loss: 0.2582
Epoch 27/50
782/782 5s 6ms/step - accuracy: 0.8361 - loss: 0.4680 - val_acc
uracy: 0.9094 - val_loss: 0.2660
Epoch 28/50
782/782 5s 6ms/step - accuracy: 0.8395 - loss: 0.4616 - val_acc
uracy: 0.9144 - val_loss: 0.2477
Epoch 29/50
782/782 5s 6ms/step - accuracy: 0.8436 - loss: 0.4510 - val_acc
uracy: 0.9207 - val_loss: 0.2367
Epoch 30/50
782/782 5s 6ms/step - accuracy: 0.8446 - loss: 0.4364 - val_acc
uracy: 0.9241 - val_loss: 0.2214
Epoch 31/50
782/782 4s 5ms/step - accuracy: 0.8484 - loss: 0.4306 - val_acc
uracy: 0.9257 - val_loss: 0.2148
Epoch 32/50
782/782 4s 5ms/step - accuracy: 0.8510 - loss: 0.4332 - val_acc
uracy: 0.9252 - val_loss: 0.2178
Epoch 33/50
782/782 5s 6ms/step - accuracy: 0.8523 - loss: 0.4206 - val_acc
uracy: 0.9281 - val_loss: 0.2104
Epoch 34/50
782/782 5s 6ms/step - accuracy: 0.8541 - loss: 0.4203 - val_acc
uracy: 0.9342 - val_loss: 0.2051
Epoch 35/50
782/782 5s 6ms/step - accuracy: 0.8557 - loss: 0.4142 - val_acc
uracy: 0.9211 - val_loss: 0.2273
Epoch 36/50
782/782 5s 6ms/step - accuracy: 0.8583 - loss: 0.4071 - val_acc
uracy: 0.9332 - val_loss: 0.1987
Epoch 37/50
782/782 5s 6ms/step - accuracy: 0.8612 - loss: 0.4004 - val_acc
uracy: 0.9359 - val_loss: 0.1856
Epoch 38/50
```

782/782 ————— 4s 5ms/step - accuracy: 0.8656 - loss: 0.3906 - val\_accuracy: 0.9310 - val\_loss: 0.1894  
Epoch 39/50  
782/782 ————— 4s 6ms/step - accuracy: 0.8634 - loss: 0.3885 - val\_accuracy: 0.9423 - val\_loss: 0.1732  
Epoch 40/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8662 - loss: 0.3847 - val\_accuracy: 0.9389 - val\_loss: 0.1863  
Epoch 41/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8674 - loss: 0.3818 - val\_accuracy: 0.9430 - val\_loss: 0.1732  
Epoch 42/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8701 - loss: 0.3707 - val\_accuracy: 0.9358 - val\_loss: 0.1904  
Epoch 43/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8714 - loss: 0.3677 - val\_accuracy: 0.9414 - val\_loss: 0.1729  
Epoch 44/50  
782/782 ————— 4s 5ms/step - accuracy: 0.8747 - loss: 0.3646 - val\_accuracy: 0.9498 - val\_loss: 0.1522  
Epoch 45/50  
782/782 ————— 4s 5ms/step - accuracy: 0.8736 - loss: 0.3627 - val\_accuracy: 0.9470 - val\_loss: 0.1628  
Epoch 46/50  
782/782 ————— 4s 5ms/step - accuracy: 0.8722 - loss: 0.3628 - val\_accuracy: 0.9476 - val\_loss: 0.1536  
Epoch 47/50  
782/782 ————— 4s 5ms/step - accuracy: 0.8771 - loss: 0.3481 - val\_accuracy: 0.9507 - val\_loss: 0.1491  
Epoch 48/50  
782/782 ————— 4s 6ms/step - accuracy: 0.8755 - loss: 0.3543 - val\_accuracy: 0.9488 - val\_loss: 0.1515  
Epoch 49/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8783 - loss: 0.3472 - val\_accuracy: 0.9417 - val\_loss: 0.1686  
Epoch 50/50  
782/782 ————— 5s 6ms/step - accuracy: 0.8770 - loss: 0.3539 - val\_accuracy: 0.9511 - val\_loss: 0.1557  
313/313 ————— 1s 2ms/step - accuracy: 0.9548 - loss: 0.1513  
Fold Validation Accuracy: 95.11%  
Epoch 1/50  
782/782 ————— 20s 15ms/step - accuracy: 0.2117 - loss: 2.1932 - val\_accuracy: 0.4589 - val\_loss: 1.4947  
Epoch 2/50  
782/782 ————— 5s 6ms/step - accuracy: 0.4397 - loss: 1.5215 - val\_accuracy: 0.5755 - val\_loss: 1.1757  
Epoch 3/50  
782/782 ————— 5s 6ms/step - accuracy: 0.5330 - loss: 1.2931 - val\_accuracy: 0.6332 - val\_loss: 1.0338  
Epoch 4/50  
782/782 ————— 5s 6ms/step - accuracy: 0.6032 - loss: 1.1214 - val\_accuracy: 0.6691 - val\_loss: 0.9162  
Epoch 5/50  
782/782 ————— 5s 6ms/step - accuracy: 0.6497 - loss: 1.0109 - val\_accuracy: 0.7199 - val\_loss: 0.7948  
Epoch 6/50

782/782 5s 6ms/step - accuracy: 0.6760 - loss: 0.9212 - val\_accuracy: 0.7542 - val\_loss: 0.6945  
Epoch 7/50  
782/782 5s 6ms/step - accuracy: 0.7002 - loss: 0.8600 - val\_accuracy: 0.7697 - val\_loss: 0.6515  
Epoch 8/50  
782/782 5s 6ms/step - accuracy: 0.7178 - loss: 0.8063 - val\_accuracy: 0.7641 - val\_loss: 0.6722  
Epoch 9/50  
782/782 5s 6ms/step - accuracy: 0.7400 - loss: 0.7625 - val\_accuracy: 0.8046 - val\_loss: 0.5454  
Epoch 10/50  
782/782 5s 6ms/step - accuracy: 0.7481 - loss: 0.7295 - val\_accuracy: 0.8123 - val\_loss: 0.5213  
Epoch 11/50  
782/782 5s 6ms/step - accuracy: 0.7587 - loss: 0.6977 - val\_accuracy: 0.8301 - val\_loss: 0.4832  
Epoch 12/50  
782/782 5s 6ms/step - accuracy: 0.7703 - loss: 0.6602 - val\_accuracy: 0.8351 - val\_loss: 0.4687  
Epoch 13/50  
782/782 5s 6ms/step - accuracy: 0.7777 - loss: 0.6423 - val\_accuracy: 0.8518 - val\_loss: 0.4281  
Epoch 14/50  
782/782 5s 6ms/step - accuracy: 0.7839 - loss: 0.6158 - val\_accuracy: 0.8477 - val\_loss: 0.4304  
Epoch 15/50  
782/782 5s 6ms/step - accuracy: 0.7956 - loss: 0.5980 - val\_accuracy: 0.8520 - val\_loss: 0.4235  
Epoch 16/50  
782/782 5s 6ms/step - accuracy: 0.7965 - loss: 0.5849 - val\_accuracy: 0.8671 - val\_loss: 0.3801  
Epoch 17/50  
782/782 5s 6ms/step - accuracy: 0.8061 - loss: 0.5617 - val\_accuracy: 0.8760 - val\_loss: 0.3481  
Epoch 18/50  
782/782 4s 5ms/step - accuracy: 0.8104 - loss: 0.5496 - val\_accuracy: 0.8784 - val\_loss: 0.3507  
Epoch 19/50  
782/782 4s 5ms/step - accuracy: 0.8152 - loss: 0.5318 - val\_accuracy: 0.8775 - val\_loss: 0.3505  
Epoch 20/50  
782/782 5s 6ms/step - accuracy: 0.8188 - loss: 0.5243 - val\_accuracy: 0.8773 - val\_loss: 0.3436  
Epoch 21/50  
782/782 5s 6ms/step - accuracy: 0.8215 - loss: 0.5081 - val\_accuracy: 0.8979 - val\_loss: 0.2966  
Epoch 22/50  
782/782 5s 6ms/step - accuracy: 0.8284 - loss: 0.4961 - val\_accuracy: 0.8885 - val\_loss: 0.3256  
Epoch 23/50  
782/782 5s 6ms/step - accuracy: 0.8256 - loss: 0.5014 - val\_accuracy: 0.9051 - val\_loss: 0.2721  
Epoch 24/50  
782/782 4s 6ms/step - accuracy: 0.8317 - loss: 0.4848 - val\_accuracy: 0.9031 - val\_loss: 0.2832

Epoch 25/50  
**782/782** 4s 6ms/step - accuracy: 0.8377 - loss: 0.4721 - val\_accuracy: 0.9137 - val\_loss: 0.2529  
Epoch 26/50  
**782/782** 5s 6ms/step - accuracy: 0.8361 - loss: 0.4697 - val\_accuracy: 0.9159 - val\_loss: 0.2470  
Epoch 27/50  
**782/782** 5s 6ms/step - accuracy: 0.8436 - loss: 0.4546 - val\_accuracy: 0.9074 - val\_loss: 0.2589  
Epoch 28/50  
**782/782** 5s 6ms/step - accuracy: 0.8458 - loss: 0.4461 - val\_accuracy: 0.9065 - val\_loss: 0.2570  
Epoch 29/50  
**782/782** 5s 6ms/step - accuracy: 0.8513 - loss: 0.4366 - val\_accuracy: 0.9105 - val\_loss: 0.2564  
Epoch 30/50  
**782/782** 5s 6ms/step - accuracy: 0.8490 - loss: 0.4362 - val\_accuracy: 0.9177 - val\_loss: 0.2337  
Epoch 31/50  
**782/782** 4s 5ms/step - accuracy: 0.8503 - loss: 0.4289 - val\_accuracy: 0.9245 - val\_loss: 0.2176  
Epoch 32/50  
**782/782** 4s 6ms/step - accuracy: 0.8539 - loss: 0.4212 - val\_accuracy: 0.9308 - val\_loss: 0.2036  
Epoch 33/50  
**782/782** 4s 5ms/step - accuracy: 0.8578 - loss: 0.4044 - val\_accuracy: 0.9249 - val\_loss: 0.2191  
Epoch 34/50  
**782/782** 4s 5ms/step - accuracy: 0.8568 - loss: 0.4078 - val\_accuracy: 0.9347 - val\_loss: 0.1931  
Epoch 35/50  
**782/782** 4s 6ms/step - accuracy: 0.8583 - loss: 0.4085 - val\_accuracy: 0.9226 - val\_loss: 0.2179  
Epoch 36/50  
**782/782** 5s 6ms/step - accuracy: 0.8584 - loss: 0.4011 - val\_accuracy: 0.9282 - val\_loss: 0.2051  
Epoch 37/50  
**782/782** 5s 6ms/step - accuracy: 0.8645 - loss: 0.3929 - val\_accuracy: 0.9405 - val\_loss: 0.1783  
Epoch 38/50  
**782/782** 5s 6ms/step - accuracy: 0.8632 - loss: 0.3891 - val\_accuracy: 0.9392 - val\_loss: 0.1851  
Epoch 39/50  
**782/782** 5s 6ms/step - accuracy: 0.8672 - loss: 0.3806 - val\_accuracy: 0.9378 - val\_loss: 0.1867  
Epoch 40/50  
**782/782** 5s 6ms/step - accuracy: 0.8657 - loss: 0.3843 - val\_accuracy: 0.9447 - val\_loss: 0.1673  
Epoch 41/50  
**782/782** 5s 6ms/step - accuracy: 0.8678 - loss: 0.3823 - val\_accuracy: 0.9444 - val\_loss: 0.1653  
Epoch 42/50  
**782/782** 5s 6ms/step - accuracy: 0.8696 - loss: 0.3776 - val\_accuracy: 0.9406 - val\_loss: 0.1724  
Epoch 43/50  
**782/782** 5s 6ms/step - accuracy: 0.8686 - loss: 0.3742 - val\_accuracy: 0.9444 - val\_loss: 0.1653

```
uracy: 0.9464 - val_loss: 0.1609
Epoch 44/50
782/782 5s 6ms/step - accuracy: 0.8741 - loss: 0.3630 - val_acc
uracy: 0.9435 - val_loss: 0.1693
Epoch 45/50
782/782 5s 6ms/step - accuracy: 0.8733 - loss: 0.3648 - val_acc
uracy: 0.9479 - val_loss: 0.1526
Epoch 46/50
782/782 5s 6ms/step - accuracy: 0.8742 - loss: 0.3617 - val_acc
uracy: 0.9536 - val_loss: 0.1513
Epoch 47/50
782/782 5s 6ms/step - accuracy: 0.8774 - loss: 0.3534 - val_acc
uracy: 0.9562 - val_loss: 0.1374
Epoch 48/50
782/782 5s 6ms/step - accuracy: 0.8753 - loss: 0.3541 - val_acc
uracy: 0.9561 - val_loss: 0.1370
Epoch 49/50
782/782 5s 6ms/step - accuracy: 0.8778 - loss: 0.3480 - val_acc
uracy: 0.9561 - val_loss: 0.1298
Epoch 50/50
782/782 5s 6ms/step - accuracy: 0.8792 - loss: 0.3461 - val_acc
uracy: 0.9565 - val_loss: 0.1330
313/313 1s 2ms/step - accuracy: 0.9566 - loss: 0.1314
Fold Validation Accuracy: 95.65%
Epoch 1/50
782/782 20s 14ms/step - accuracy: 0.1947 - loss: 2.2347 - val_a
ccuracy: 0.4950 - val_loss: 1.4145
Epoch 2/50
782/782 5s 6ms/step - accuracy: 0.4560 - loss: 1.4824 - val_acc
uracy: 0.5846 - val_loss: 1.1573
Epoch 3/50
782/782 5s 6ms/step - accuracy: 0.5522 - loss: 1.2550 - val_acc
uracy: 0.6541 - val_loss: 0.9809
Epoch 4/50
782/782 5s 6ms/step - accuracy: 0.6210 - loss: 1.0813 - val_acc
uracy: 0.7143 - val_loss: 0.8163
Epoch 5/50
782/782 4s 5ms/step - accuracy: 0.6535 - loss: 1.0007 - val_acc
uracy: 0.7229 - val_loss: 0.7784
Epoch 6/50
782/782 5s 6ms/step - accuracy: 0.6869 - loss: 0.9040 - val_acc
uracy: 0.7585 - val_loss: 0.6932
Epoch 7/50
782/782 5s 6ms/step - accuracy: 0.7092 - loss: 0.8295 - val_acc
uracy: 0.7808 - val_loss: 0.6144
Epoch 8/50
782/782 5s 6ms/step - accuracy: 0.7319 - loss: 0.7788 - val_acc
uracy: 0.8032 - val_loss: 0.5623
Epoch 9/50
782/782 5s 6ms/step - accuracy: 0.7407 - loss: 0.7391 - val_acc
uracy: 0.8186 - val_loss: 0.5259
Epoch 10/50
782/782 5s 6ms/step - accuracy: 0.7605 - loss: 0.6997 - val_acc
uracy: 0.8192 - val_loss: 0.5150
Epoch 11/50
782/782 4s 5ms/step - accuracy: 0.7691 - loss: 0.6671 - val_acc
```

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uracy: 0.8357 - val_loss: 0.4586
Epoch 12/50
782/782 4s 6ms/step - accuracy: 0.7723 - loss: 0.6518 - val_acc
uracy: 0.8527 - val_loss: 0.4205
Epoch 13/50
782/782 5s 6ms/step - accuracy: 0.7844 - loss: 0.6248 - val_acc
uracy: 0.8594 - val_loss: 0.4117
Epoch 14/50
782/782 5s 6ms/step - accuracy: 0.7881 - loss: 0.6103 - val_acc
uracy: 0.8584 - val_loss: 0.4001
Epoch 15/50
782/782 5s 6ms/step - accuracy: 0.7951 - loss: 0.5909 - val_acc
uracy: 0.8705 - val_loss: 0.3751
Epoch 16/50
782/782 5s 6ms/step - accuracy: 0.8066 - loss: 0.5715 - val_acc
uracy: 0.8769 - val_loss: 0.3583
Epoch 17/50
782/782 4s 6ms/step - accuracy: 0.8096 - loss: 0.5572 - val_acc
uracy: 0.8749 - val_loss: 0.3463
Epoch 18/50
782/782 4s 5ms/step - accuracy: 0.8129 - loss: 0.5385 - val_acc
uracy: 0.8812 - val_loss: 0.3414
Epoch 19/50
782/782 4s 5ms/step - accuracy: 0.8206 - loss: 0.5187 - val_acc
uracy: 0.8863 - val_loss: 0.3186
Epoch 20/50
782/782 4s 6ms/step - accuracy: 0.8253 - loss: 0.5113 - val_acc
uracy: 0.8971 - val_loss: 0.2947
Epoch 21/50
782/782 4s 6ms/step - accuracy: 0.8235 - loss: 0.5062 - val_acc
uracy: 0.9002 - val_loss: 0.2908
Epoch 22/50
782/782 5s 6ms/step - accuracy: 0.8340 - loss: 0.4740 - val_acc
uracy: 0.9016 - val_loss: 0.2840
Epoch 23/50
782/782 5s 6ms/step - accuracy: 0.8353 - loss: 0.4714 - val_acc
uracy: 0.9124 - val_loss: 0.2572
Epoch 24/50
782/782 5s 6ms/step - accuracy: 0.8361 - loss: 0.4705 - val_acc
uracy: 0.9147 - val_loss: 0.2467
Epoch 25/50
782/782 5s 6ms/step - accuracy: 0.8454 - loss: 0.4537 - val_acc
uracy: 0.9144 - val_loss: 0.2451
Epoch 26/50
782/782 5s 6ms/step - accuracy: 0.8463 - loss: 0.4445 - val_acc
uracy: 0.9253 - val_loss: 0.2302
Epoch 27/50
782/782 5s 6ms/step - accuracy: 0.8476 - loss: 0.4367 - val_acc
uracy: 0.9233 - val_loss: 0.2286
Epoch 28/50
782/782 5s 6ms/step - accuracy: 0.8476 - loss: 0.4340 - val_acc
uracy: 0.9279 - val_loss: 0.2172
Epoch 29/50
782/782 5s 6ms/step - accuracy: 0.8522 - loss: 0.4276 - val_acc
uracy: 0.9233 - val_loss: 0.2181
Epoch 30/50
```

782/782 5s 6ms/step - accuracy: 0.8519 - loss: 0.4248 - val\_accuracy: 0.9299 - val\_loss: 0.2146  
Epoch 31/50  
782/782 5s 6ms/step - accuracy: 0.8567 - loss: 0.4088 - val\_accuracy: 0.9292 - val\_loss: 0.2113  
Epoch 32/50  
782/782 5s 6ms/step - accuracy: 0.8548 - loss: 0.4170 - val\_accuracy: 0.9277 - val\_loss: 0.2013  
Epoch 33/50  
782/782 5s 6ms/step - accuracy: 0.8561 - loss: 0.4138 - val\_accuracy: 0.9388 - val\_loss: 0.1921  
Epoch 34/50  
782/782 5s 6ms/step - accuracy: 0.8631 - loss: 0.3969 - val\_accuracy: 0.9390 - val\_loss: 0.1763  
Epoch 35/50  
782/782 5s 6ms/step - accuracy: 0.8661 - loss: 0.3870 - val\_accuracy: 0.9422 - val\_loss: 0.1803  
Epoch 36/50  
782/782 5s 6ms/step - accuracy: 0.8653 - loss: 0.3879 - val\_accuracy: 0.9442 - val\_loss: 0.1743  
Epoch 37/50  
782/782 5s 6ms/step - accuracy: 0.8670 - loss: 0.3813 - val\_accuracy: 0.9447 - val\_loss: 0.1709  
Epoch 38/50  
782/782 5s 6ms/step - accuracy: 0.8669 - loss: 0.3799 - val\_accuracy: 0.9485 - val\_loss: 0.1600  
Epoch 39/50  
782/782 5s 6ms/step - accuracy: 0.8708 - loss: 0.3725 - val\_accuracy: 0.9425 - val\_loss: 0.1663  
Epoch 40/50  
782/782 5s 6ms/step - accuracy: 0.8704 - loss: 0.3736 - val\_accuracy: 0.9526 - val\_loss: 0.1516  
Epoch 41/50  
782/782 5s 6ms/step - accuracy: 0.8713 - loss: 0.3638 - val\_accuracy: 0.9500 - val\_loss: 0.1592  
Epoch 42/50  
782/782 5s 6ms/step - accuracy: 0.8752 - loss: 0.3566 - val\_accuracy: 0.9505 - val\_loss: 0.1463  
Epoch 43/50  
782/782 5s 6ms/step - accuracy: 0.8771 - loss: 0.3573 - val\_accuracy: 0.9548 - val\_loss: 0.1468  
Epoch 44/50  
782/782 5s 6ms/step - accuracy: 0.8744 - loss: 0.3572 - val\_accuracy: 0.9549 - val\_loss: 0.1487  
Epoch 45/50  
782/782 4s 6ms/step - accuracy: 0.8788 - loss: 0.3478 - val\_accuracy: 0.9550 - val\_loss: 0.1421  
Epoch 46/50  
782/782 4s 5ms/step - accuracy: 0.8777 - loss: 0.3535 - val\_accuracy: 0.9540 - val\_loss: 0.1408  
Epoch 47/50  
782/782 5s 6ms/step - accuracy: 0.8814 - loss: 0.3427 - val\_accuracy: 0.9569 - val\_loss: 0.1354  
Epoch 48/50  
782/782 5s 6ms/step - accuracy: 0.8850 - loss: 0.3330 - val\_accuracy: 0.9581 - val\_loss: 0.1338

Epoch 49/50  
**782/782** 5s 6ms/step - accuracy: 0.8834 - loss: 0.3366 - val\_accuracy: 0.9578 - val\_loss: 0.1348  
Epoch 50/50  
**782/782** 5s 6ms/step - accuracy: 0.8812 - loss: 0.3356 - val\_accuracy: 0.9656 - val\_loss: 0.1220  
**313/313** 1s 2ms/step - accuracy: 0.9647 - loss: 0.1221  
Fold Validation Accuracy: 96.56%  
Epoch 1/50  
**782/782** 20s 15ms/step - accuracy: 0.2015 - loss: 2.3009 - val\_accuracy: 0.4803 - val\_loss: 1.4696  
Epoch 2/50  
**782/782** 5s 6ms/step - accuracy: 0.4419 - loss: 1.5258 - val\_accuracy: 0.5768 - val\_loss: 1.2198  
Epoch 3/50  
**782/782** 4s 6ms/step - accuracy: 0.5243 - loss: 1.3151 - val\_accuracy: 0.6331 - val\_loss: 1.0176  
Epoch 4/50  
**782/782** 4s 5ms/step - accuracy: 0.5898 - loss: 1.1607 - val\_accuracy: 0.6928 - val\_loss: 0.8784  
Epoch 5/50  
**782/782** 4s 5ms/step - accuracy: 0.6317 - loss: 1.0451 - val\_accuracy: 0.7191 - val\_loss: 0.7785  
Epoch 6/50  
**782/782** 4s 5ms/step - accuracy: 0.6683 - loss: 0.9505 - val\_accuracy: 0.7506 - val\_loss: 0.7127  
Epoch 7/50  
**782/782** 4s 6ms/step - accuracy: 0.6940 - loss: 0.8800 - val\_accuracy: 0.7672 - val\_loss: 0.6549  
Epoch 8/50  
**782/782** 5s 6ms/step - accuracy: 0.7145 - loss: 0.8213 - val\_accuracy: 0.7806 - val\_loss: 0.6197  
Epoch 9/50  
**782/782** 5s 6ms/step - accuracy: 0.7313 - loss: 0.7750 - val\_accuracy: 0.8051 - val\_loss: 0.5517  
Epoch 10/50  
**782/782** 5s 6ms/step - accuracy: 0.7463 - loss: 0.7311 - val\_accuracy: 0.8165 - val\_loss: 0.5216  
Epoch 11/50  
**782/782** 5s 6ms/step - accuracy: 0.7604 - loss: 0.6958 - val\_accuracy: 0.8315 - val\_loss: 0.4856  
Epoch 12/50  
**782/782** 5s 6ms/step - accuracy: 0.7592 - loss: 0.6884 - val\_accuracy: 0.8420 - val\_loss: 0.4485  
Epoch 13/50  
**782/782** 5s 6ms/step - accuracy: 0.7816 - loss: 0.6375 - val\_accuracy: 0.8470 - val\_loss: 0.4324  
Epoch 14/50  
**782/782** 5s 6ms/step - accuracy: 0.7845 - loss: 0.6225 - val\_accuracy: 0.8573 - val\_loss: 0.4039  
Epoch 15/50  
**782/782** 5s 6ms/step - accuracy: 0.7895 - loss: 0.6083 - val\_accuracy: 0.8598 - val\_loss: 0.4038  
Epoch 16/50  
**782/782** 5s 6ms/step - accuracy: 0.8011 - loss: 0.5726 - val\_accuracy: 0.8691 - val\_loss: 0.3771

Epoch 17/50  
**782/782** 5s 6ms/step - accuracy: 0.8032 - loss: 0.5660 - val\_accuracy: 0.8802 - val\_loss: 0.3488  
Epoch 18/50  
**782/782** 5s 6ms/step - accuracy: 0.8092 - loss: 0.5490 - val\_accuracy: 0.8778 - val\_loss: 0.3486  
Epoch 19/50  
**782/782** 5s 6ms/step - accuracy: 0.8189 - loss: 0.5241 - val\_accuracy: 0.8859 - val\_loss: 0.3218  
Epoch 20/50  
**782/782** 5s 6ms/step - accuracy: 0.8202 - loss: 0.5172 - val\_accuracy: 0.8964 - val\_loss: 0.3040  
Epoch 21/50  
**782/782** 5s 6ms/step - accuracy: 0.8226 - loss: 0.5110 - val\_accuracy: 0.8925 - val\_loss: 0.3050  
Epoch 22/50  
**782/782** 5s 6ms/step - accuracy: 0.8237 - loss: 0.5108 - val\_accuracy: 0.9048 - val\_loss: 0.2769  
Epoch 23/50  
**782/782** 5s 6ms/step - accuracy: 0.8331 - loss: 0.4822 - val\_accuracy: 0.9012 - val\_loss: 0.2916  
Epoch 24/50  
**782/782** 5s 6ms/step - accuracy: 0.8297 - loss: 0.4845 - val\_accuracy: 0.9118 - val\_loss: 0.2547  
Epoch 25/50  
**782/782** 5s 6ms/step - accuracy: 0.8377 - loss: 0.4737 - val\_accuracy: 0.9148 - val\_loss: 0.2470  
Epoch 26/50  
**782/782** 5s 6ms/step - accuracy: 0.8378 - loss: 0.4621 - val\_accuracy: 0.9191 - val\_loss: 0.2415  
Epoch 27/50  
**782/782** 5s 6ms/step - accuracy: 0.8394 - loss: 0.4606 - val\_accuracy: 0.9177 - val\_loss: 0.2420  
Epoch 28/50  
**782/782** 5s 6ms/step - accuracy: 0.8449 - loss: 0.4475 - val\_accuracy: 0.9136 - val\_loss: 0.2553  
Epoch 29/50  
**782/782** 5s 6ms/step - accuracy: 0.8454 - loss: 0.4408 - val\_accuracy: 0.9252 - val\_loss: 0.2216  
Epoch 30/50  
**782/782** 5s 6ms/step - accuracy: 0.8457 - loss: 0.4424 - val\_accuracy: 0.9260 - val\_loss: 0.2141  
Epoch 31/50  
**782/782** 4s 5ms/step - accuracy: 0.8516 - loss: 0.4246 - val\_accuracy: 0.9264 - val\_loss: 0.2196  
Epoch 32/50  
**782/782** 4s 5ms/step - accuracy: 0.8519 - loss: 0.4133 - val\_accuracy: 0.9264 - val\_loss: 0.2116  
Epoch 33/50  
**782/782** 4s 5ms/step - accuracy: 0.8540 - loss: 0.4162 - val\_accuracy: 0.9281 - val\_loss: 0.2081  
Epoch 34/50  
**782/782** 4s 5ms/step - accuracy: 0.8555 - loss: 0.4188 - val\_accuracy: 0.9256 - val\_loss: 0.2223  
Epoch 35/50  
**782/782** 4s 5ms/step - accuracy: 0.8600 - loss: 0.4038 - val\_accuracy: 0.9300 - val\_loss: 0.2000

```
uracy: 0.9389 - val_loss: 0.1930
Epoch 36/50
782/782 4s 5ms/step - accuracy: 0.8641 - loss: 0.3900 - val_acc
uracy: 0.9332 - val_loss: 0.2006
Epoch 37/50
782/782 4s 5ms/step - accuracy: 0.8636 - loss: 0.3891 - val_acc
uracy: 0.9353 - val_loss: 0.1864
Epoch 38/50
782/782 4s 5ms/step - accuracy: 0.8643 - loss: 0.3867 - val_acc
uracy: 0.9383 - val_loss: 0.1828
Epoch 39/50
782/782 4s 5ms/step - accuracy: 0.8653 - loss: 0.3862 - val_acc
uracy: 0.9373 - val_loss: 0.1820
Epoch 40/50
782/782 4s 5ms/step - accuracy: 0.8700 - loss: 0.3740 - val_acc
uracy: 0.9463 - val_loss: 0.1630
Epoch 41/50
782/782 4s 5ms/step - accuracy: 0.8701 - loss: 0.3777 - val_acc
uracy: 0.9484 - val_loss: 0.1657
Epoch 42/50
782/782 4s 5ms/step - accuracy: 0.8680 - loss: 0.3686 - val_acc
uracy: 0.9468 - val_loss: 0.1556
Epoch 43/50
782/782 4s 5ms/step - accuracy: 0.8738 - loss: 0.3621 - val_acc
uracy: 0.9503 - val_loss: 0.1623
Epoch 44/50
782/782 4s 5ms/step - accuracy: 0.8745 - loss: 0.3568 - val_acc
uracy: 0.9520 - val_loss: 0.1525
Epoch 45/50
782/782 4s 5ms/step - accuracy: 0.8781 - loss: 0.3495 - val_acc
uracy: 0.9503 - val_loss: 0.1532
Epoch 46/50
782/782 4s 5ms/step - accuracy: 0.8776 - loss: 0.3484 - val_acc
uracy: 0.9535 - val_loss: 0.1457
Epoch 47/50
782/782 4s 5ms/step - accuracy: 0.8796 - loss: 0.3447 - val_acc
uracy: 0.9512 - val_loss: 0.1452
Epoch 48/50
782/782 4s 5ms/step - accuracy: 0.8804 - loss: 0.3418 - val_acc
uracy: 0.9574 - val_loss: 0.1316
Epoch 49/50
782/782 4s 5ms/step - accuracy: 0.8759 - loss: 0.3488 - val_acc
uracy: 0.9525 - val_loss: 0.1395
Epoch 50/50
782/782 4s 5ms/step - accuracy: 0.8801 - loss: 0.3414 - val_acc
uracy: 0.9570 - val_loss: 0.1319
313/313 1s 2ms/step - accuracy: 0.9555 - loss: 0.1302
Fold Validation Accuracy: 95.70%
Epoch 1/50
782/782 17s 13ms/step - accuracy: 0.2060 - loss: 2.2383 - val_a
ccuracy: 0.4641 - val_loss: 1.4881
Epoch 2/50
782/782 4s 5ms/step - accuracy: 0.4449 - loss: 1.5142 - val_acc
uracy: 0.5855 - val_loss: 1.1765
Epoch 3/50
782/782 4s 5ms/step - accuracy: 0.5385 - loss: 1.2804 - val_acc
```

```
uracy: 0.6529 - val_loss: 0.9992
Epoch 4/50
782/782 4s 5ms/step - accuracy: 0.6029 - loss: 1.1225 - val_acc
uracy: 0.6673 - val_loss: 0.9278
Epoch 5/50
782/782 4s 5ms/step - accuracy: 0.6462 - loss: 1.0174 - val_acc
uracy: 0.7334 - val_loss: 0.7535
Epoch 6/50
782/782 4s 5ms/step - accuracy: 0.6672 - loss: 0.9478 - val_acc
uracy: 0.7528 - val_loss: 0.6943
Epoch 7/50
782/782 4s 5ms/step - accuracy: 0.6917 - loss: 0.8872 - val_acc
uracy: 0.7629 - val_loss: 0.6791
Epoch 8/50
782/782 4s 5ms/step - accuracy: 0.7119 - loss: 0.8311 - val_acc
uracy: 0.7748 - val_loss: 0.6317
Epoch 9/50
782/782 4s 5ms/step - accuracy: 0.7269 - loss: 0.7896 - val_acc
uracy: 0.7970 - val_loss: 0.5796
Epoch 10/50
782/782 4s 5ms/step - accuracy: 0.7427 - loss: 0.7391 - val_acc
uracy: 0.8144 - val_loss: 0.5244
Epoch 11/50
782/782 4s 5ms/step - accuracy: 0.7500 - loss: 0.7221 - val_acc
uracy: 0.8285 - val_loss: 0.4931
Epoch 12/50
782/782 4s 5ms/step - accuracy: 0.7610 - loss: 0.6806 - val_acc
uracy: 0.8279 - val_loss: 0.4924
Epoch 13/50
782/782 4s 5ms/step - accuracy: 0.7737 - loss: 0.6527 - val_acc
uracy: 0.8327 - val_loss: 0.4707
Epoch 14/50
782/782 4s 5ms/step - accuracy: 0.7752 - loss: 0.6421 - val_acc
uracy: 0.8530 - val_loss: 0.4153
Epoch 15/50
782/782 4s 5ms/step - accuracy: 0.7881 - loss: 0.6171 - val_acc
uracy: 0.8642 - val_loss: 0.3867
Epoch 16/50
782/782 4s 5ms/step - accuracy: 0.7949 - loss: 0.5990 - val_acc
uracy: 0.8470 - val_loss: 0.4225
Epoch 17/50
782/782 4s 5ms/step - accuracy: 0.7980 - loss: 0.5840 - val_acc
uracy: 0.8691 - val_loss: 0.3732
Epoch 18/50
782/782 4s 5ms/step - accuracy: 0.8022 - loss: 0.5708 - val_acc
uracy: 0.8795 - val_loss: 0.3433
Epoch 19/50
782/782 4s 5ms/step - accuracy: 0.8096 - loss: 0.5516 - val_acc
uracy: 0.8849 - val_loss: 0.3376
Epoch 20/50
782/782 4s 5ms/step - accuracy: 0.8155 - loss: 0.5327 - val_acc
uracy: 0.8934 - val_loss: 0.3159
Epoch 21/50
782/782 4s 5ms/step - accuracy: 0.8188 - loss: 0.5212 - val_acc
uracy: 0.8905 - val_loss: 0.3171
Epoch 22/50
```

782/782 4s 5ms/step - accuracy: 0.8235 - loss: 0.5066 - val\_accuracy: 0.8987 - val\_loss: 0.2917  
Epoch 23/50  
782/782 4s 5ms/step - accuracy: 0.8283 - loss: 0.4867 - val\_accuracy: 0.9001 - val\_loss: 0.2899  
Epoch 24/50  
782/782 4s 5ms/step - accuracy: 0.8302 - loss: 0.4864 - val\_accuracy: 0.8980 - val\_loss: 0.2965  
Epoch 25/50  
782/782 4s 5ms/step - accuracy: 0.8344 - loss: 0.4843 - val\_accuracy: 0.9103 - val\_loss: 0.2612  
Epoch 26/50  
782/782 4s 5ms/step - accuracy: 0.8349 - loss: 0.4775 - val\_accuracy: 0.9087 - val\_loss: 0.2580  
Epoch 27/50  
782/782 4s 5ms/step - accuracy: 0.8360 - loss: 0.4677 - val\_accuracy: 0.9208 - val\_loss: 0.2374  
Epoch 28/50  
782/782 4s 5ms/step - accuracy: 0.8420 - loss: 0.4534 - val\_accuracy: 0.9149 - val\_loss: 0.2479  
Epoch 29/50  
782/782 4s 5ms/step - accuracy: 0.8425 - loss: 0.4470 - val\_accuracy: 0.9210 - val\_loss: 0.2347  
Epoch 30/50  
782/782 4s 5ms/step - accuracy: 0.8453 - loss: 0.4412 - val\_accuracy: 0.9102 - val\_loss: 0.2479  
Epoch 31/50  
782/782 4s 5ms/step - accuracy: 0.8471 - loss: 0.4465 - val\_accuracy: 0.9306 - val\_loss: 0.2167  
Epoch 32/50  
782/782 4s 5ms/step - accuracy: 0.8516 - loss: 0.4275 - val\_accuracy: 0.9275 - val\_loss: 0.2054  
Epoch 33/50  
782/782 4s 5ms/step - accuracy: 0.8523 - loss: 0.4227 - val\_accuracy: 0.9246 - val\_loss: 0.2117  
Epoch 34/50  
782/782 4s 5ms/step - accuracy: 0.8530 - loss: 0.4170 - val\_accuracy: 0.9282 - val\_loss: 0.2093  
Epoch 35/50  
782/782 4s 5ms/step - accuracy: 0.8563 - loss: 0.4152 - val\_accuracy: 0.9332 - val\_loss: 0.2039  
Epoch 36/50  
782/782 4s 5ms/step - accuracy: 0.8587 - loss: 0.4027 - val\_accuracy: 0.9355 - val\_loss: 0.1839  
Epoch 37/50  
782/782 4s 5ms/step - accuracy: 0.8652 - loss: 0.3915 - val\_accuracy: 0.9313 - val\_loss: 0.1927  
Epoch 38/50  
782/782 4s 5ms/step - accuracy: 0.8609 - loss: 0.3957 - val\_accuracy: 0.9321 - val\_loss: 0.1875  
Epoch 39/50  
782/782 4s 5ms/step - accuracy: 0.8614 - loss: 0.3935 - val\_accuracy: 0.9451 - val\_loss: 0.1688  
Epoch 40/50  
782/782 4s 5ms/step - accuracy: 0.8632 - loss: 0.3921 - val\_accuracy: 0.9455 - val\_loss: 0.1683

```
Epoch 41/50
782/782 4s 5ms/step - accuracy: 0.8695 - loss: 0.3761 - val_accuracy: 0.9465 - val_loss: 0.1632
Epoch 42/50
782/782 4s 5ms/step - accuracy: 0.8673 - loss: 0.3795 - val_accuracy: 0.9448 - val_loss: 0.1633
Epoch 43/50
782/782 4s 5ms/step - accuracy: 0.8690 - loss: 0.3710 - val_accuracy: 0.9418 - val_loss: 0.1703
Epoch 44/50
782/782 4s 5ms/step - accuracy: 0.8717 - loss: 0.3645 - val_accuracy: 0.9364 - val_loss: 0.1852
Epoch 45/50
782/782 4s 5ms/step - accuracy: 0.8737 - loss: 0.3588 - val_accuracy: 0.9476 - val_loss: 0.1565
Epoch 46/50
782/782 4s 5ms/step - accuracy: 0.8729 - loss: 0.3630 - val_accuracy: 0.9489 - val_loss: 0.1566
Epoch 47/50
782/782 4s 5ms/step - accuracy: 0.8749 - loss: 0.3538 - val_accuracy: 0.9451 - val_loss: 0.1544
Epoch 48/50
782/782 4s 5ms/step - accuracy: 0.8739 - loss: 0.3501 - val_accuracy: 0.9473 - val_loss: 0.1559
Epoch 49/50
782/782 4s 5ms/step - accuracy: 0.8784 - loss: 0.3488 - val_accuracy: 0.9454 - val_loss: 0.1578
Epoch 50/50
782/782 4s 5ms/step - accuracy: 0.8793 - loss: 0.3470 - val_accuracy: 0.9548 - val_loss: 0.1427
313/313 1s 2ms/step - accuracy: 0.9532 - loss: 0.1469
Fold Validation Accuracy: 95.48%
```

Average 5-Fold Cross-Validation Accuracy: 95.70%

```
In [63]: evaluate_on_test(model1)
```

```
Epoch 1/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8817 - loss: 0.3409  
Epoch 2/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8768 - loss: 0.3483  
Epoch 3/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8840 - loss: 0.3355  
Epoch 4/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8815 - loss: 0.3364  
Epoch 5/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8828 - loss: 0.3355  
Epoch 6/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8869 - loss: 0.3259  
Epoch 7/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8838 - loss: 0.3315  
Epoch 8/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8879 - loss: 0.3189  
Epoch 9/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8886 - loss: 0.3198  
Epoch 10/10  
782/782 ━━━━━━━━ 4s 5ms/step - accuracy: 0.8901 - loss: 0.3140  
313/313 ━━━━━━ 1s 2ms/step - accuracy: 0.8432 - loss: 0.5352
```

Test Accuracy: 0.8382999897003174

Test Loss: 0.54093998670578

Out[63]: 0.8382999897003174

In [64]: `GenerateClassificationReport(model1)`

313/313 ━━━━━━ 1s 3ms/step

Confusion Matrix										
True Class	0	1	2	3	4	5	6	7	8	9
	871	4	15	9	8	1	4	4	64	20
	14	918	0	1	0	1	1	0	22	43
	69	1	706	29	76	28	63	12	12	4
	21	2	33	683	51	106	64	18	15	7
	11	1	17	21	873	11	34	26	5	1
	12	1	29	131	46	734	20	22	2	3
	6	0	18	32	26	7	900	3	7	1
	14	0	23	21	49	34	7	845	3	4
	28	2	3	2	4	0	3	0	944	14
	17	37	2	6	0	1	1	3	24	909

	precision	recall	f1-score	support
0	0.82	0.87	0.84	1000
1	0.95	0.92	0.93	1000
2	0.83	0.71	0.76	1000
3	0.73	0.68	0.71	1000
4	0.77	0.87	0.82	1000
5	0.80	0.73	0.76	1000
6	0.82	0.90	0.86	1000
7	0.91	0.84	0.87	1000
8	0.86	0.94	0.90	1000
9	0.90	0.91	0.91	1000
accuracy			0.84	10000
macro avg	0.84	0.84	0.84	10000
weighted avg	0.84	0.84	0.84	10000

Class 0: Precision = 0.82, Recall = 0.87  
 Class 1: Precision = 0.95, Recall = 0.92  
 Class 2: Precision = 0.83, Recall = 0.71  
 Class 3: Precision = 0.73, Recall = 0.68  
 Class 4: Precision = 0.77, Recall = 0.87  
 Class 5: Precision = 0.80, Recall = 0.73  
 Class 6: Precision = 0.82, Recall = 0.90  
 Class 7: Precision = 0.91, Recall = 0.84  
 Class 8: Precision = 0.86, Recall = 0.94  
 Class 9: Precision = 0.90, Recall = 0.91

Training it for 50 epochs gave us an accuracy of 83.8%

In [73]:

```
#save model1 and download
model1.save('/kaggle/working/model1.h5')
!zip -r /kaggle/working/model1.zip /kaggle/working/model1.h5
```

/opt/conda/lib/python3.10/pty.py:89: RuntimeWarning: os.fork() was called. os.fork() is incompatible with multithreaded code, and JAX is multithreaded, so this will likely lead to a deadlock.

```
pid, fd = os.forkpty()
adding: kaggle/working/model1.h5 (deflated 9%)
```

In [14]:

```
model4,histories= CrossVal_Training("with_dropout_increasing",100,64)
```

/opt/conda/lib/python3.10/site-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)
```

Epoch 1/100

WARNING: All log messages before absl::InitializeLog() is called are written to STDE RR

I0000 00:00:1730095480.317605 78 service.cc:145] XLA service 0x7c91c40072a0 initialized for platform CUDA (this does not guarantee that XLA will be used). Devices: I0000 00:00:1730095480.317659 78 service.cc:153] StreamExecutor device (0): Tesla P100-PCIE-16GB, Compute Capability 6.0

31/782 ————— 3s 5ms/step - accuracy: 0.1000 - loss: 4.6740

I00000 00:00:1730095487.795047 78 device\_compiler.h:188] Compiled cluster using XLA! This line is logged at most once for the lifetime of the process.

782/782 21s 14ms/step - accuracy: 0.2085 - loss: 2.2869 - val\_accuracy: 0.4319 - val\_loss: 1.5662  
Epoch 2/100  
782/782 4s 5ms/step - accuracy: 0.4306 - loss: 1.5462 - val\_accuracy: 0.5524 - val\_loss: 1.2162  
Epoch 3/100  
782/782 4s 5ms/step - accuracy: 0.5198 - loss: 1.3242 - val\_accuracy: 0.6422 - val\_loss: 0.9877  
Epoch 4/100  
782/782 4s 5ms/step - accuracy: 0.5905 - loss: 1.1472 - val\_accuracy: 0.6859 - val\_loss: 0.8666  
Epoch 5/100  
782/782 4s 5ms/step - accuracy: 0.6395 - loss: 1.0215 - val\_accuracy: 0.7228 - val\_loss: 0.7795  
Epoch 6/100  
782/782 4s 5ms/step - accuracy: 0.6805 - loss: 0.9284 - val\_accuracy: 0.7454 - val\_loss: 0.7067  
Epoch 7/100  
782/782 4s 5ms/step - accuracy: 0.7002 - loss: 0.8620 - val\_accuracy: 0.7795 - val\_loss: 0.6388  
Epoch 8/100  
782/782 4s 5ms/step - accuracy: 0.7149 - loss: 0.8211 - val\_accuracy: 0.7916 - val\_loss: 0.6049  
Epoch 9/100  
782/782 4s 5ms/step - accuracy: 0.7329 - loss: 0.7702 - val\_accuracy: 0.8091 - val\_loss: 0.5393  
Epoch 10/100  
782/782 4s 5ms/step - accuracy: 0.7474 - loss: 0.7319 - val\_accuracy: 0.8179 - val\_loss: 0.5256  
Epoch 11/100  
782/782 4s 5ms/step - accuracy: 0.7531 - loss: 0.7084 - val\_accuracy: 0.8273 - val\_loss: 0.4912  
Epoch 12/100  
782/782 4s 5ms/step - accuracy: 0.7691 - loss: 0.6716 - val\_accuracy: 0.8469 - val\_loss: 0.4435  
Epoch 13/100  
782/782 4s 5ms/step - accuracy: 0.7744 - loss: 0.6584 - val\_accuracy: 0.8549 - val\_loss: 0.4203  
Epoch 14/100  
782/782 4s 5ms/step - accuracy: 0.7788 - loss: 0.6425 - val\_accuracy: 0.8498 - val\_loss: 0.4405  
Epoch 15/100  
782/782 4s 5ms/step - accuracy: 0.7881 - loss: 0.6060 - val\_accuracy: 0.8657 - val\_loss: 0.3971  
Epoch 16/100  
782/782 4s 5ms/step - accuracy: 0.7938 - loss: 0.5979 - val\_accuracy: 0.8647 - val\_loss: 0.3853  
Epoch 17/100  
782/782 4s 5ms/step - accuracy: 0.8034 - loss: 0.5678 - val\_accuracy: 0.8633 - val\_loss: 0.3982  
Epoch 18/100  
782/782 4s 5ms/step - accuracy: 0.8069 - loss: 0.5584 - val\_accuracy: 0.8815 - val\_loss: 0.3507  
Epoch 19/100  
782/782 4s 5ms/step - accuracy: 0.8091 - loss: 0.5487 - val\_accuracy: 0.8892 - val\_loss: 0.3176

Epoch 20/100  
**782/782** 4s 5ms/step - accuracy: 0.8207 - loss: 0.5233 - val\_accuracy: 0.8962 - val\_loss: 0.3072  
Epoch 21/100  
**782/782** 4s 5ms/step - accuracy: 0.8225 - loss: 0.5078 - val\_accuracy: 0.8922 - val\_loss: 0.3155  
Epoch 22/100  
**782/782** 4s 5ms/step - accuracy: 0.8209 - loss: 0.5184 - val\_accuracy: 0.8967 - val\_loss: 0.3004  
Epoch 23/100  
**782/782** 4s 5ms/step - accuracy: 0.8257 - loss: 0.5019 - val\_accuracy: 0.9018 - val\_loss: 0.2857  
Epoch 24/100  
**782/782** 4s 5ms/step - accuracy: 0.8338 - loss: 0.4862 - val\_accuracy: 0.9014 - val\_loss: 0.2930  
Epoch 25/100  
**782/782** 4s 5ms/step - accuracy: 0.8348 - loss: 0.4782 - val\_accuracy: 0.9079 - val\_loss: 0.2741  
Epoch 26/100  
**782/782** 4s 5ms/step - accuracy: 0.8378 - loss: 0.4657 - val\_accuracy: 0.9055 - val\_loss: 0.2623  
Epoch 27/100  
**782/782** 4s 5ms/step - accuracy: 0.8438 - loss: 0.4497 - val\_accuracy: 0.9063 - val\_loss: 0.2674  
Epoch 28/100  
**782/782** 4s 5ms/step - accuracy: 0.8468 - loss: 0.4460 - val\_accuracy: 0.9113 - val\_loss: 0.2582  
Epoch 29/100  
**782/782** 4s 5ms/step - accuracy: 0.8466 - loss: 0.4386 - val\_accuracy: 0.9120 - val\_loss: 0.2510  
Epoch 30/100  
**782/782** 4s 5ms/step - accuracy: 0.8512 - loss: 0.4312 - val\_accuracy: 0.9197 - val\_loss: 0.2293  
Epoch 31/100  
**782/782** 4s 5ms/step - accuracy: 0.8515 - loss: 0.4302 - val\_accuracy: 0.9228 - val\_loss: 0.2234  
Epoch 32/100  
**782/782** 4s 5ms/step - accuracy: 0.8517 - loss: 0.4223 - val\_accuracy: 0.9255 - val\_loss: 0.2167  
Epoch 33/100  
**782/782** 4s 5ms/step - accuracy: 0.8549 - loss: 0.4220 - val\_accuracy: 0.9304 - val\_loss: 0.2110  
Epoch 34/100  
**782/782** 4s 5ms/step - accuracy: 0.8574 - loss: 0.4090 - val\_accuracy: 0.9326 - val\_loss: 0.1941  
Epoch 35/100  
**782/782** 4s 5ms/step - accuracy: 0.8549 - loss: 0.4097 - val\_accuracy: 0.9320 - val\_loss: 0.1994  
Epoch 36/100  
**782/782** 4s 5ms/step - accuracy: 0.8600 - loss: 0.3984 - val\_accuracy: 0.9358 - val\_loss: 0.1887  
Epoch 37/100  
**782/782** 4s 5ms/step - accuracy: 0.8611 - loss: 0.4009 - val\_accuracy: 0.9344 - val\_loss: 0.1995  
Epoch 38/100  
**782/782** 4s 5ms/step - accuracy: 0.8652 - loss: 0.3858 - val\_accuracy: 0.9380 - val\_loss: 0.1850

```
uracy: 0.9395 - val_loss: 0.1836
Epoch 39/100
782/782 4s 5ms/step - accuracy: 0.8674 - loss: 0.3784 - val_acc
uracy: 0.9441 - val_loss: 0.1663
Epoch 40/100
782/782 5s 5ms/step - accuracy: 0.8676 - loss: 0.3819 - val_acc
uracy: 0.9355 - val_loss: 0.1903
Epoch 41/100
782/782 4s 5ms/step - accuracy: 0.8669 - loss: 0.3806 - val_acc
uracy: 0.9435 - val_loss: 0.1676
Epoch 42/100
782/782 4s 5ms/step - accuracy: 0.8708 - loss: 0.3665 - val_acc
uracy: 0.9444 - val_loss: 0.1685
Epoch 43/100
782/782 4s 5ms/step - accuracy: 0.8738 - loss: 0.3606 - val_acc
uracy: 0.9509 - val_loss: 0.1537
Epoch 44/100
782/782 4s 5ms/step - accuracy: 0.8765 - loss: 0.3582 - val_acc
uracy: 0.9461 - val_loss: 0.1681
Epoch 45/100
782/782 4s 5ms/step - accuracy: 0.8724 - loss: 0.3657 - val_acc
uracy: 0.9513 - val_loss: 0.1543
Epoch 46/100
782/782 4s 5ms/step - accuracy: 0.8762 - loss: 0.3563 - val_acc
uracy: 0.9467 - val_loss: 0.1607
Epoch 47/100
782/782 4s 5ms/step - accuracy: 0.8794 - loss: 0.3502 - val_acc
uracy: 0.9564 - val_loss: 0.1380
Epoch 48/100
782/782 4s 5ms/step - accuracy: 0.8781 - loss: 0.3458 - val_acc
uracy: 0.9595 - val_loss: 0.1341
Epoch 49/100
782/782 4s 5ms/step - accuracy: 0.8805 - loss: 0.3458 - val_acc
uracy: 0.9525 - val_loss: 0.1479
Epoch 50/100
782/782 4s 5ms/step - accuracy: 0.8782 - loss: 0.3451 - val_acc
uracy: 0.9550 - val_loss: 0.1402
Epoch 51/100
782/782 4s 5ms/step - accuracy: 0.8803 - loss: 0.3439 - val_acc
uracy: 0.9501 - val_loss: 0.1436
Epoch 52/100
782/782 4s 5ms/step - accuracy: 0.8798 - loss: 0.3478 - val_acc
uracy: 0.9552 - val_loss: 0.1387
Epoch 53/100
782/782 4s 5ms/step - accuracy: 0.8812 - loss: 0.3414 - val_acc
uracy: 0.9603 - val_loss: 0.1246
Epoch 54/100
782/782 4s 5ms/step - accuracy: 0.8843 - loss: 0.3285 - val_acc
uracy: 0.9619 - val_loss: 0.1313
Epoch 55/100
782/782 4s 5ms/step - accuracy: 0.8863 - loss: 0.3288 - val_acc
uracy: 0.9632 - val_loss: 0.1213
Epoch 56/100
782/782 4s 5ms/step - accuracy: 0.8906 - loss: 0.3186 - val_acc
uracy: 0.9539 - val_loss: 0.1308
Epoch 57/100
```

782/782 4s 5ms/step - accuracy: 0.8872 - loss: 0.3224 - val\_accuracy: 0.9586 - val\_loss: 0.1315  
Epoch 58/100  
782/782 4s 5ms/step - accuracy: 0.8873 - loss: 0.3252 - val\_accuracy: 0.9633 - val\_loss: 0.1180  
Epoch 59/100  
782/782 4s 5ms/step - accuracy: 0.8888 - loss: 0.3239 - val\_accuracy: 0.9627 - val\_loss: 0.1185  
Epoch 60/100  
782/782 4s 5ms/step - accuracy: 0.8892 - loss: 0.3215 - val\_accuracy: 0.9576 - val\_loss: 0.1259  
Epoch 61/100  
782/782 4s 5ms/step - accuracy: 0.8904 - loss: 0.3161 - val\_accuracy: 0.9644 - val\_loss: 0.1142  
Epoch 62/100  
782/782 4s 5ms/step - accuracy: 0.8907 - loss: 0.3134 - val\_accuracy: 0.9684 - val\_loss: 0.1060  
Epoch 63/100  
782/782 4s 5ms/step - accuracy: 0.8934 - loss: 0.3117 - val\_accuracy: 0.9643 - val\_loss: 0.1157  
Epoch 64/100  
782/782 4s 5ms/step - accuracy: 0.8919 - loss: 0.3056 - val\_accuracy: 0.9628 - val\_loss: 0.1188  
Epoch 65/100  
782/782 4s 5ms/step - accuracy: 0.8971 - loss: 0.3014 - val\_accuracy: 0.9694 - val\_loss: 0.1016  
Epoch 66/100  
782/782 4s 5ms/step - accuracy: 0.8970 - loss: 0.2951 - val\_accuracy: 0.9657 - val\_loss: 0.1044  
Epoch 67/100  
782/782 4s 5ms/step - accuracy: 0.8946 - loss: 0.3049 - val\_accuracy: 0.9673 - val\_loss: 0.1052  
Epoch 68/100  
782/782 4s 5ms/step - accuracy: 0.8964 - loss: 0.2996 - val\_accuracy: 0.9734 - val\_loss: 0.0935  
Epoch 69/100  
782/782 4s 5ms/step - accuracy: 0.8989 - loss: 0.2944 - val\_accuracy: 0.9665 - val\_loss: 0.1041  
Epoch 70/100  
782/782 4s 5ms/step - accuracy: 0.8936 - loss: 0.2957 - val\_accuracy: 0.9708 - val\_loss: 0.0922  
Epoch 71/100  
782/782 4s 5ms/step - accuracy: 0.8979 - loss: 0.2899 - val\_accuracy: 0.9715 - val\_loss: 0.0943  
Epoch 72/100  
782/782 4s 5ms/step - accuracy: 0.8955 - loss: 0.2984 - val\_accuracy: 0.9746 - val\_loss: 0.0891  
Epoch 73/100  
782/782 4s 5ms/step - accuracy: 0.8964 - loss: 0.2939 - val\_accuracy: 0.9564 - val\_loss: 0.1295  
Epoch 74/100  
782/782 4s 5ms/step - accuracy: 0.8978 - loss: 0.2976 - val\_accuracy: 0.9714 - val\_loss: 0.0882  
Epoch 75/100  
782/782 4s 5ms/step - accuracy: 0.9023 - loss: 0.2794 - val\_accuracy: 0.9752 - val\_loss: 0.0866

Epoch 76/100  
**782/782** 4s 5ms/step - accuracy: 0.9042 - loss: 0.2719 - val\_accuracy: 0.9755 - val\_loss: 0.0823  
Epoch 77/100  
**782/782** 4s 5ms/step - accuracy: 0.8990 - loss: 0.2887 - val\_accuracy: 0.9657 - val\_loss: 0.1001  
Epoch 78/100  
**782/782** 4s 5ms/step - accuracy: 0.9043 - loss: 0.2752 - val\_accuracy: 0.9702 - val\_loss: 0.0933  
Epoch 79/100  
**782/782** 4s 5ms/step - accuracy: 0.9034 - loss: 0.2721 - val\_accuracy: 0.9737 - val\_loss: 0.0829  
Epoch 80/100  
**782/782** 4s 5ms/step - accuracy: 0.9044 - loss: 0.2791 - val\_accuracy: 0.9748 - val\_loss: 0.0798  
Epoch 81/100  
**782/782** 4s 5ms/step - accuracy: 0.9070 - loss: 0.2677 - val\_accuracy: 0.9763 - val\_loss: 0.0776  
Epoch 82/100  
**782/782** 4s 5ms/step - accuracy: 0.9014 - loss: 0.2850 - val\_accuracy: 0.9761 - val\_loss: 0.0769  
Epoch 83/100  
**782/782** 4s 5ms/step - accuracy: 0.9031 - loss: 0.2763 - val\_accuracy: 0.9769 - val\_loss: 0.0765  
Epoch 84/100  
**782/782** 4s 5ms/step - accuracy: 0.9005 - loss: 0.2856 - val\_accuracy: 0.9707 - val\_loss: 0.0853  
Epoch 85/100  
**782/782** 4s 5ms/step - accuracy: 0.9069 - loss: 0.2699 - val\_accuracy: 0.9803 - val\_loss: 0.0689  
Epoch 86/100  
**782/782** 4s 5ms/step - accuracy: 0.9078 - loss: 0.2663 - val\_accuracy: 0.9778 - val\_loss: 0.0731  
Epoch 87/100  
**782/782** 4s 5ms/step - accuracy: 0.9035 - loss: 0.2785 - val\_accuracy: 0.9726 - val\_loss: 0.0808  
Epoch 88/100  
**782/782** 4s 5ms/step - accuracy: 0.9119 - loss: 0.2620 - val\_accuracy: 0.9749 - val\_loss: 0.0850  
Epoch 89/100  
**782/782** 4s 5ms/step - accuracy: 0.9069 - loss: 0.2742 - val\_accuracy: 0.9824 - val\_loss: 0.0669  
Epoch 90/100  
**782/782** 4s 5ms/step - accuracy: 0.9087 - loss: 0.2600 - val\_accuracy: 0.9779 - val\_loss: 0.0740  
Epoch 91/100  
**782/782** 4s 5ms/step - accuracy: 0.9062 - loss: 0.2705 - val\_accuracy: 0.9788 - val\_loss: 0.0712  
Epoch 92/100  
**782/782** 4s 5ms/step - accuracy: 0.9120 - loss: 0.2554 - val\_accuracy: 0.9808 - val\_loss: 0.0680  
Epoch 93/100  
**782/782** 4s 5ms/step - accuracy: 0.9073 - loss: 0.2670 - val\_accuracy: 0.9772 - val\_loss: 0.0748  
Epoch 94/100  
**782/782** 4s 5ms/step - accuracy: 0.9089 - loss: 0.2603 - val\_accuracy: 0.9788 - val\_loss: 0.0712

```
uracy: 0.9786 - val_loss: 0.0695
Epoch 95/100
782/782 5s 5ms/step - accuracy: 0.9110 - loss: 0.2580 - val_acc
uracy: 0.9804 - val_loss: 0.0672
Epoch 96/100
782/782 4s 5ms/step - accuracy: 0.9104 - loss: 0.2548 - val_acc
uracy: 0.9848 - val_loss: 0.0590
Epoch 97/100
782/782 4s 5ms/step - accuracy: 0.9148 - loss: 0.2519 - val_acc
uracy: 0.9829 - val_loss: 0.0604
Epoch 98/100
782/782 4s 5ms/step - accuracy: 0.9107 - loss: 0.2597 - val_acc
uracy: 0.9811 - val_loss: 0.0673
Epoch 99/100
782/782 4s 5ms/step - accuracy: 0.9124 - loss: 0.2510 - val_acc
uracy: 0.9836 - val_loss: 0.0566
Epoch 100/100
782/782 4s 5ms/step - accuracy: 0.9121 - loss: 0.2554 - val_acc
uracy: 0.9824 - val_loss: 0.0596
313/313 1s 2ms/step - accuracy: 0.9836 - loss: 0.0571
Fold Validation Accuracy: 98.24%
Epoch 1/100
782/782 17s 13ms/step - accuracy: 0.1972 - loss: 2.2754 - val_a
ccuracy: 0.4473 - val_loss: 1.5186
Epoch 2/100
782/782 4s 5ms/step - accuracy: 0.4405 - loss: 1.5301 - val_acc
uracy: 0.5594 - val_loss: 1.2016
Epoch 3/100
782/782 4s 5ms/step - accuracy: 0.5280 - loss: 1.3217 - val_acc
uracy: 0.6030 - val_loss: 1.0962
Epoch 4/100
782/782 4s 6ms/step - accuracy: 0.5887 - loss: 1.1556 - val_acc
uracy: 0.6813 - val_loss: 0.8911
Epoch 5/100
782/782 4s 5ms/step - accuracy: 0.6396 - loss: 1.0330 - val_acc
uracy: 0.7109 - val_loss: 0.8042
Epoch 6/100
782/782 4s 5ms/step - accuracy: 0.6711 - loss: 0.9422 - val_acc
uracy: 0.7410 - val_loss: 0.7218
Epoch 7/100
782/782 4s 5ms/step - accuracy: 0.6906 - loss: 0.8900 - val_acc
uracy: 0.7427 - val_loss: 0.7187
Epoch 8/100
782/782 4s 5ms/step - accuracy: 0.7092 - loss: 0.8418 - val_acc
uracy: 0.7651 - val_loss: 0.6700
Epoch 9/100
782/782 4s 5ms/step - accuracy: 0.7288 - loss: 0.7856 - val_acc
uracy: 0.7975 - val_loss: 0.5770
Epoch 10/100
782/782 4s 5ms/step - accuracy: 0.7424 - loss: 0.7425 - val_acc
uracy: 0.8023 - val_loss: 0.5572
Epoch 11/100
782/782 4s 5ms/step - accuracy: 0.7516 - loss: 0.7227 - val_acc
uracy: 0.8154 - val_loss: 0.5236
Epoch 12/100
782/782 4s 5ms/step - accuracy: 0.7588 - loss: 0.6890 - val_acc
```

```
uracy: 0.8241 - val_loss: 0.4951
Epoch 13/100
782/782 4s 5ms/step - accuracy: 0.7749 - loss: 0.6570 - val_acc
uracy: 0.8370 - val_loss: 0.4583
Epoch 14/100
782/782 4s 5ms/step - accuracy: 0.7753 - loss: 0.6427 - val_acc
uracy: 0.8390 - val_loss: 0.4579
Epoch 15/100
782/782 4s 5ms/step - accuracy: 0.7808 - loss: 0.6265 - val_acc
uracy: 0.8521 - val_loss: 0.4271
Epoch 16/100
782/782 4s 5ms/step - accuracy: 0.7898 - loss: 0.6023 - val_acc
uracy: 0.8455 - val_loss: 0.4250
Epoch 17/100
782/782 4s 5ms/step - accuracy: 0.8028 - loss: 0.5768 - val_acc
uracy: 0.8722 - val_loss: 0.3705
Epoch 18/100
782/782 4s 5ms/step - accuracy: 0.8053 - loss: 0.5702 - val_acc
uracy: 0.8705 - val_loss: 0.3590
Epoch 19/100
782/782 4s 5ms/step - accuracy: 0.8084 - loss: 0.5522 - val_acc
uracy: 0.8721 - val_loss: 0.3678
Epoch 20/100
782/782 4s 5ms/step - accuracy: 0.8129 - loss: 0.5387 - val_acc
uracy: 0.8765 - val_loss: 0.3536
Epoch 21/100
782/782 4s 5ms/step - accuracy: 0.8147 - loss: 0.5294 - val_acc
uracy: 0.8884 - val_loss: 0.3199
Epoch 22/100
782/782 4s 5ms/step - accuracy: 0.8223 - loss: 0.5130 - val_acc
uracy: 0.8823 - val_loss: 0.3361
Epoch 23/100
782/782 4s 5ms/step - accuracy: 0.8191 - loss: 0.5167 - val_acc
uracy: 0.9003 - val_loss: 0.2890
Epoch 24/100
782/782 4s 5ms/step - accuracy: 0.8265 - loss: 0.4940 - val_acc
uracy: 0.8864 - val_loss: 0.3196
Epoch 25/100
782/782 4s 5ms/step - accuracy: 0.8298 - loss: 0.4883 - val_acc
uracy: 0.9011 - val_loss: 0.2808
Epoch 26/100
782/782 4s 5ms/step - accuracy: 0.8358 - loss: 0.4691 - val_acc
uracy: 0.9054 - val_loss: 0.2722
Epoch 27/100
782/782 4s 5ms/step - accuracy: 0.8398 - loss: 0.4710 - val_acc
uracy: 0.9044 - val_loss: 0.2735
Epoch 28/100
782/782 4s 5ms/step - accuracy: 0.8439 - loss: 0.4480 - val_acc
uracy: 0.9163 - val_loss: 0.2444
Epoch 29/100
782/782 4s 5ms/step - accuracy: 0.8405 - loss: 0.4610 - val_acc
uracy: 0.9206 - val_loss: 0.2418
Epoch 30/100
782/782 4s 5ms/step - accuracy: 0.8473 - loss: 0.4378 - val_acc
uracy: 0.9190 - val_loss: 0.2322
Epoch 31/100
```

782/782 4s 5ms/step - accuracy: 0.8453 - loss: 0.4426 - val\_accuracy: 0.9302 - val\_loss: 0.2109  
Epoch 32/100  
782/782 4s 5ms/step - accuracy: 0.8491 - loss: 0.4290 - val\_accuracy: 0.9267 - val\_loss: 0.2214  
Epoch 33/100  
782/782 4s 5ms/step - accuracy: 0.8498 - loss: 0.4264 - val\_accuracy: 0.9225 - val\_loss: 0.2267  
Epoch 34/100  
782/782 4s 5ms/step - accuracy: 0.8565 - loss: 0.4138 - val\_accuracy: 0.9191 - val\_loss: 0.2333  
Epoch 35/100  
782/782 4s 5ms/step - accuracy: 0.8530 - loss: 0.4157 - val\_accuracy: 0.9203 - val\_loss: 0.2189  
Epoch 36/100  
782/782 4s 5ms/step - accuracy: 0.8564 - loss: 0.4079 - val\_accuracy: 0.9322 - val\_loss: 0.1974  
Epoch 37/100  
782/782 4s 5ms/step - accuracy: 0.8573 - loss: 0.4099 - val\_accuracy: 0.9399 - val\_loss: 0.1802  
Epoch 38/100  
782/782 4s 5ms/step - accuracy: 0.8644 - loss: 0.3964 - val\_accuracy: 0.9360 - val\_loss: 0.1870  
Epoch 39/100  
782/782 4s 5ms/step - accuracy: 0.8614 - loss: 0.3952 - val\_accuracy: 0.9420 - val\_loss: 0.1791  
Epoch 40/100  
782/782 4s 5ms/step - accuracy: 0.8666 - loss: 0.3827 - val\_accuracy: 0.9435 - val\_loss: 0.1675  
Epoch 41/100  
782/782 4s 5ms/step - accuracy: 0.8717 - loss: 0.3755 - val\_accuracy: 0.9399 - val\_loss: 0.1764  
Epoch 42/100  
782/782 4s 5ms/step - accuracy: 0.8682 - loss: 0.3796 - val\_accuracy: 0.9451 - val\_loss: 0.1651  
Epoch 43/100  
782/782 4s 5ms/step - accuracy: 0.8669 - loss: 0.3829 - val\_accuracy: 0.9455 - val\_loss: 0.1631  
Epoch 44/100  
782/782 4s 5ms/step - accuracy: 0.8728 - loss: 0.3673 - val\_accuracy: 0.9425 - val\_loss: 0.1742  
Epoch 45/100  
782/782 4s 5ms/step - accuracy: 0.8729 - loss: 0.3649 - val\_accuracy: 0.9461 - val\_loss: 0.1600  
Epoch 46/100  
782/782 4s 5ms/step - accuracy: 0.8691 - loss: 0.3711 - val\_accuracy: 0.9463 - val\_loss: 0.1680  
Epoch 47/100  
782/782 4s 5ms/step - accuracy: 0.8736 - loss: 0.3579 - val\_accuracy: 0.9467 - val\_loss: 0.1591  
Epoch 48/100  
782/782 4s 5ms/step - accuracy: 0.8775 - loss: 0.3519 - val\_accuracy: 0.9498 - val\_loss: 0.1533  
Epoch 49/100  
782/782 4s 5ms/step - accuracy: 0.8770 - loss: 0.3498 - val\_accuracy: 0.9512 - val\_loss: 0.1555

Epoch 50/100  
**782/782** 4s 5ms/step - accuracy: 0.8807 - loss: 0.3416 - val\_accuracy: 0.9529 - val\_loss: 0.1476  
Epoch 51/100  
**782/782** 4s 5ms/step - accuracy: 0.8754 - loss: 0.3544 - val\_accuracy: 0.9446 - val\_loss: 0.1625  
Epoch 52/100  
**782/782** 4s 5ms/step - accuracy: 0.8800 - loss: 0.3437 - val\_accuracy: 0.9529 - val\_loss: 0.1474  
Epoch 53/100  
**782/782** 4s 5ms/step - accuracy: 0.8827 - loss: 0.3391 - val\_accuracy: 0.9577 - val\_loss: 0.1313  
Epoch 54/100  
**782/782** 4s 5ms/step - accuracy: 0.8824 - loss: 0.3451 - val\_accuracy: 0.9562 - val\_loss: 0.1363  
Epoch 55/100  
**782/782** 4s 5ms/step - accuracy: 0.8846 - loss: 0.3325 - val\_accuracy: 0.9603 - val\_loss: 0.1232  
Epoch 56/100  
**782/782** 4s 5ms/step - accuracy: 0.8812 - loss: 0.3396 - val\_accuracy: 0.9590 - val\_loss: 0.1295  
Epoch 57/100  
**782/782** 4s 5ms/step - accuracy: 0.8832 - loss: 0.3305 - val\_accuracy: 0.9449 - val\_loss: 0.1578  
Epoch 58/100  
**782/782** 4s 5ms/step - accuracy: 0.8841 - loss: 0.3343 - val\_accuracy: 0.9667 - val\_loss: 0.1152  
Epoch 59/100  
**782/782** 4s 5ms/step - accuracy: 0.8873 - loss: 0.3277 - val\_accuracy: 0.9601 - val\_loss: 0.1254  
Epoch 60/100  
**782/782** 4s 5ms/step - accuracy: 0.8857 - loss: 0.3259 - val\_accuracy: 0.9596 - val\_loss: 0.1280  
Epoch 61/100  
**782/782** 4s 5ms/step - accuracy: 0.8866 - loss: 0.3257 - val\_accuracy: 0.9568 - val\_loss: 0.1321  
Epoch 62/100  
**782/782** 4s 5ms/step - accuracy: 0.8894 - loss: 0.3173 - val\_accuracy: 0.9675 - val\_loss: 0.1130  
Epoch 63/100  
**782/782** 4s 5ms/step - accuracy: 0.8920 - loss: 0.3126 - val\_accuracy: 0.9638 - val\_loss: 0.1167  
Epoch 64/100  
**782/782** 4s 5ms/step - accuracy: 0.8896 - loss: 0.3158 - val\_accuracy: 0.9651 - val\_loss: 0.1060  
Epoch 65/100  
**782/782** 4s 5ms/step - accuracy: 0.8912 - loss: 0.3115 - val\_accuracy: 0.9694 - val\_loss: 0.1008  
Epoch 66/100  
**782/782** 4s 5ms/step - accuracy: 0.8914 - loss: 0.3126 - val\_accuracy: 0.9708 - val\_loss: 0.0979  
Epoch 67/100  
**782/782** 4s 5ms/step - accuracy: 0.8971 - loss: 0.3017 - val\_accuracy: 0.9646 - val\_loss: 0.1089  
Epoch 68/100  
**782/782** 4s 5ms/step - accuracy: 0.8905 - loss: 0.3131 - val\_accuracy: 0.9694 - val\_loss: 0.1008

```
uracy: 0.9609 - val_loss: 0.1194
Epoch 69/100
782/782 4s 5ms/step - accuracy: 0.8935 - loss: 0.3082 - val_acc
uracy: 0.9634 - val_loss: 0.1138
Epoch 70/100
782/782 4s 5ms/step - accuracy: 0.8957 - loss: 0.2899 - val_acc
uracy: 0.9724 - val_loss: 0.0957
Epoch 71/100
782/782 4s 5ms/step - accuracy: 0.8956 - loss: 0.3018 - val_acc
uracy: 0.9666 - val_loss: 0.1043
Epoch 72/100
782/782 4s 5ms/step - accuracy: 0.8976 - loss: 0.2975 - val_acc
uracy: 0.9742 - val_loss: 0.0898
Epoch 73/100
782/782 4s 5ms/step - accuracy: 0.8915 - loss: 0.3082 - val_acc
uracy: 0.9697 - val_loss: 0.0975
Epoch 74/100
782/782 4s 5ms/step - accuracy: 0.8997 - loss: 0.2922 - val_acc
uracy: 0.9716 - val_loss: 0.0931
Epoch 75/100
782/782 4s 5ms/step - accuracy: 0.8972 - loss: 0.2967 - val_acc
uracy: 0.9735 - val_loss: 0.0908
Epoch 76/100
782/782 4s 5ms/step - accuracy: 0.8973 - loss: 0.2862 - val_acc
uracy: 0.9725 - val_loss: 0.0948
Epoch 77/100
782/782 4s 5ms/step - accuracy: 0.9007 - loss: 0.2884 - val_acc
uracy: 0.9678 - val_loss: 0.0983
Epoch 78/100
782/782 4s 5ms/step - accuracy: 0.8955 - loss: 0.2961 - val_acc
uracy: 0.9759 - val_loss: 0.0850
Epoch 79/100
782/782 4s 5ms/step - accuracy: 0.9023 - loss: 0.2838 - val_acc
uracy: 0.9778 - val_loss: 0.0789
Epoch 80/100
782/782 4s 5ms/step - accuracy: 0.8995 - loss: 0.2868 - val_acc
uracy: 0.9762 - val_loss: 0.0793
Epoch 81/100
782/782 4s 5ms/step - accuracy: 0.9003 - loss: 0.2851 - val_acc
uracy: 0.9756 - val_loss: 0.0860
Epoch 82/100
782/782 4s 5ms/step - accuracy: 0.9015 - loss: 0.2789 - val_acc
uracy: 0.9758 - val_loss: 0.0807
Epoch 83/100
782/782 4s 5ms/step - accuracy: 0.9007 - loss: 0.2835 - val_acc
uracy: 0.9762 - val_loss: 0.0890
Epoch 84/100
782/782 4s 5ms/step - accuracy: 0.9018 - loss: 0.2845 - val_acc
uracy: 0.9792 - val_loss: 0.0757
Epoch 85/100
782/782 4s 5ms/step - accuracy: 0.9044 - loss: 0.2776 - val_acc
uracy: 0.9757 - val_loss: 0.0822
Epoch 86/100
782/782 4s 5ms/step - accuracy: 0.9040 - loss: 0.2747 - val_acc
uracy: 0.9784 - val_loss: 0.0761
Epoch 87/100
```

782/782 4s 5ms/step - accuracy: 0.9076 - loss: 0.2694 - val\_accuracy: 0.9736 - val\_loss: 0.0820  
Epoch 88/100  
782/782 4s 5ms/step - accuracy: 0.9023 - loss: 0.2787 - val\_accuracy: 0.9765 - val\_loss: 0.0807  
Epoch 89/100  
782/782 4s 5ms/step - accuracy: 0.9086 - loss: 0.2653 - val\_accuracy: 0.9785 - val\_loss: 0.0759  
Epoch 90/100  
782/782 4s 5ms/step - accuracy: 0.9033 - loss: 0.2793 - val\_accuracy: 0.9818 - val\_loss: 0.0714  
Epoch 91/100  
782/782 4s 5ms/step - accuracy: 0.9069 - loss: 0.2702 - val\_accuracy: 0.9702 - val\_loss: 0.0985  
Epoch 92/100  
782/782 4s 5ms/step - accuracy: 0.9044 - loss: 0.2743 - val\_accuracy: 0.9813 - val\_loss: 0.0713  
Epoch 93/100  
782/782 4s 5ms/step - accuracy: 0.9097 - loss: 0.2585 - val\_accuracy: 0.9824 - val\_loss: 0.0667  
Epoch 94/100  
782/782 4s 5ms/step - accuracy: 0.9070 - loss: 0.2602 - val\_accuracy: 0.9817 - val\_loss: 0.0722  
Epoch 95/100  
782/782 4s 5ms/step - accuracy: 0.9069 - loss: 0.2663 - val\_accuracy: 0.9804 - val\_loss: 0.0690  
Epoch 96/100  
782/782 4s 5ms/step - accuracy: 0.9088 - loss: 0.2633 - val\_accuracy: 0.9735 - val\_loss: 0.0898  
Epoch 97/100  
782/782 5s 5ms/step - accuracy: 0.9072 - loss: 0.2682 - val\_accuracy: 0.9788 - val\_loss: 0.0736  
Epoch 98/100  
782/782 4s 5ms/step - accuracy: 0.9100 - loss: 0.2645 - val\_accuracy: 0.9779 - val\_loss: 0.0727  
Epoch 99/100  
782/782 4s 5ms/step - accuracy: 0.9098 - loss: 0.2584 - val\_accuracy: 0.9802 - val\_loss: 0.0684  
Epoch 100/100  
782/782 4s 5ms/step - accuracy: 0.9073 - loss: 0.2679 - val\_accuracy: 0.9790 - val\_loss: 0.0781  
313/313 1s 2ms/step - accuracy: 0.9784 - loss: 0.0767  
Fold Validation Accuracy: 97.90%  
Epoch 1/100  
782/782 17s 13ms/step - accuracy: 0.1973 - loss: 2.2422 - val\_accuracy: 0.4699 - val\_loss: 1.5027  
Epoch 2/100  
782/782 4s 5ms/step - accuracy: 0.4499 - loss: 1.5053 - val\_accuracy: 0.5901 - val\_loss: 1.1368  
Epoch 3/100  
782/782 4s 5ms/step - accuracy: 0.5459 - loss: 1.2602 - val\_accuracy: 0.6623 - val\_loss: 0.9809  
Epoch 4/100  
782/782 4s 5ms/step - accuracy: 0.6102 - loss: 1.1087 - val\_accuracy: 0.7005 - val\_loss: 0.8578  
Epoch 5/100

782/782 4s 5ms/step - accuracy: 0.6508 - loss: 1.0026 - val\_accuracy: 0.7400 - val\_loss: 0.7438  
Epoch 6/100  
782/782 4s 5ms/step - accuracy: 0.6755 - loss: 0.9255 - val\_accuracy: 0.7555 - val\_loss: 0.7006  
Epoch 7/100  
782/782 4s 5ms/step - accuracy: 0.7025 - loss: 0.8584 - val\_accuracy: 0.7575 - val\_loss: 0.6763  
Epoch 8/100  
782/782 4s 5ms/step - accuracy: 0.7202 - loss: 0.8032 - val\_accuracy: 0.7945 - val\_loss: 0.5858  
Epoch 9/100  
782/782 4s 5ms/step - accuracy: 0.7372 - loss: 0.7677 - val\_accuracy: 0.7899 - val\_loss: 0.5805  
Epoch 10/100  
782/782 4s 5ms/step - accuracy: 0.7459 - loss: 0.7356 - val\_accuracy: 0.8268 - val\_loss: 0.5073  
Epoch 11/100  
782/782 4s 5ms/step - accuracy: 0.7582 - loss: 0.7034 - val\_accuracy: 0.8249 - val\_loss: 0.5025  
Epoch 12/100  
782/782 4s 5ms/step - accuracy: 0.7658 - loss: 0.6659 - val\_accuracy: 0.8424 - val\_loss: 0.4480  
Epoch 13/100  
782/782 4s 5ms/step - accuracy: 0.7688 - loss: 0.6594 - val\_accuracy: 0.8466 - val\_loss: 0.4330  
Epoch 14/100  
782/782 4s 5ms/step - accuracy: 0.7865 - loss: 0.6171 - val\_accuracy: 0.8590 - val\_loss: 0.4078  
Epoch 15/100  
782/782 4s 5ms/step - accuracy: 0.7954 - loss: 0.5897 - val\_accuracy: 0.8676 - val\_loss: 0.3846  
Epoch 16/100  
782/782 4s 5ms/step - accuracy: 0.7949 - loss: 0.5902 - val\_accuracy: 0.8772 - val\_loss: 0.3542  
Epoch 17/100  
782/782 4s 5ms/step - accuracy: 0.8028 - loss: 0.5704 - val\_accuracy: 0.8822 - val\_loss: 0.3485  
Epoch 18/100  
782/782 4s 5ms/step - accuracy: 0.8096 - loss: 0.5534 - val\_accuracy: 0.8822 - val\_loss: 0.3370  
Epoch 19/100  
782/782 4s 5ms/step - accuracy: 0.8148 - loss: 0.5369 - val\_accuracy: 0.8890 - val\_loss: 0.3255  
Epoch 20/100  
782/782 4s 5ms/step - accuracy: 0.8149 - loss: 0.5342 - val\_accuracy: 0.8915 - val\_loss: 0.3146  
Epoch 21/100  
782/782 4s 5ms/step - accuracy: 0.8204 - loss: 0.5158 - val\_accuracy: 0.8916 - val\_loss: 0.3120  
Epoch 22/100  
782/782 4s 5ms/step - accuracy: 0.8241 - loss: 0.5017 - val\_accuracy: 0.9034 - val\_loss: 0.2831  
Epoch 23/100  
782/782 4s 5ms/step - accuracy: 0.8320 - loss: 0.4926 - val\_accuracy: 0.9058 - val\_loss: 0.2796

Epoch 24/100  
**782/782** 4s 5ms/step - accuracy: 0.8329 - loss: 0.4844 - val\_accuracy: 0.9053 - val\_loss: 0.2642  
Epoch 25/100  
**782/782** 4s 5ms/step - accuracy: 0.8339 - loss: 0.4768 - val\_accuracy: 0.9176 - val\_loss: 0.2527  
Epoch 26/100  
**782/782** 4s 5ms/step - accuracy: 0.8388 - loss: 0.4645 - val\_accuracy: 0.9189 - val\_loss: 0.2474  
Epoch 27/100  
**782/782** 4s 5ms/step - accuracy: 0.8385 - loss: 0.4631 - val\_accuracy: 0.9220 - val\_loss: 0.2292  
Epoch 28/100  
**782/782** 4s 5ms/step - accuracy: 0.8433 - loss: 0.4521 - val\_accuracy: 0.9095 - val\_loss: 0.2507  
Epoch 29/100  
**782/782** 4s 5ms/step - accuracy: 0.8417 - loss: 0.4553 - val\_accuracy: 0.9271 - val\_loss: 0.2177  
Epoch 30/100  
**782/782** 4s 5ms/step - accuracy: 0.8470 - loss: 0.4405 - val\_accuracy: 0.9183 - val\_loss: 0.2405  
Epoch 31/100  
**782/782** 4s 5ms/step - accuracy: 0.8475 - loss: 0.4402 - val\_accuracy: 0.9228 - val\_loss: 0.2235  
Epoch 32/100  
**782/782** 4s 5ms/step - accuracy: 0.8505 - loss: 0.4284 - val\_accuracy: 0.9375 - val\_loss: 0.1939  
Epoch 33/100  
**782/782** 4s 5ms/step - accuracy: 0.8532 - loss: 0.4165 - val\_accuracy: 0.9367 - val\_loss: 0.1964  
Epoch 34/100  
**782/782** 4s 5ms/step - accuracy: 0.8554 - loss: 0.4145 - val\_accuracy: 0.9395 - val\_loss: 0.1834  
Epoch 35/100  
**782/782** 4s 5ms/step - accuracy: 0.8589 - loss: 0.4084 - val\_accuracy: 0.9297 - val\_loss: 0.2053  
Epoch 36/100  
**782/782** 4s 5ms/step - accuracy: 0.8598 - loss: 0.4001 - val\_accuracy: 0.9395 - val\_loss: 0.1835  
Epoch 37/100  
**782/782** 4s 5ms/step - accuracy: 0.8647 - loss: 0.3876 - val\_accuracy: 0.9409 - val\_loss: 0.1870  
Epoch 38/100  
**782/782** 4s 5ms/step - accuracy: 0.8634 - loss: 0.3944 - val\_accuracy: 0.9438 - val\_loss: 0.1668  
Epoch 39/100  
**782/782** 4s 5ms/step - accuracy: 0.8618 - loss: 0.3926 - val\_accuracy: 0.9434 - val\_loss: 0.1700  
Epoch 40/100  
**782/782** 4s 5ms/step - accuracy: 0.8665 - loss: 0.3776 - val\_accuracy: 0.9373 - val\_loss: 0.1804  
Epoch 41/100  
**782/782** 4s 5ms/step - accuracy: 0.8691 - loss: 0.3778 - val\_accuracy: 0.9455 - val\_loss: 0.1616  
Epoch 42/100  
**782/782** 4s 5ms/step - accuracy: 0.8720 - loss: 0.3733 - val\_accuracy: 0.9455 - val\_loss: 0.1616

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uracy: 0.9425 - val_loss: 0.1676
Epoch 43/100
782/782 4s 5ms/step - accuracy: 0.8707 - loss: 0.3703 - val_acc
uracy: 0.9449 - val_loss: 0.1586
Epoch 44/100
782/782 4s 5ms/step - accuracy: 0.8718 - loss: 0.3747 - val_acc
uracy: 0.9564 - val_loss: 0.1413
Epoch 45/100
782/782 4s 5ms/step - accuracy: 0.8743 - loss: 0.3569 - val_acc
uracy: 0.9609 - val_loss: 0.1325
Epoch 46/100
782/782 4s 5ms/step - accuracy: 0.8753 - loss: 0.3596 - val_acc
uracy: 0.9535 - val_loss: 0.1419
Epoch 47/100
782/782 4s 5ms/step - accuracy: 0.8739 - loss: 0.3649 - val_acc
uracy: 0.9545 - val_loss: 0.1467
Epoch 48/100
782/782 4s 5ms/step - accuracy: 0.8780 - loss: 0.3491 - val_acc
uracy: 0.9553 - val_loss: 0.1368
Epoch 49/100
782/782 4s 5ms/step - accuracy: 0.8789 - loss: 0.3482 - val_acc
uracy: 0.9525 - val_loss: 0.1539
Epoch 50/100
782/782 4s 5ms/step - accuracy: 0.8793 - loss: 0.3480 - val_acc
uracy: 0.9501 - val_loss: 0.1510
Epoch 51/100
782/782 4s 5ms/step - accuracy: 0.8787 - loss: 0.3421 - val_acc
uracy: 0.9600 - val_loss: 0.1293
Epoch 52/100
782/782 4s 5ms/step - accuracy: 0.8827 - loss: 0.3346 - val_acc
uracy: 0.9514 - val_loss: 0.1414
Epoch 53/100
782/782 4s 5ms/step - accuracy: 0.8795 - loss: 0.3467 - val_acc
uracy: 0.9559 - val_loss: 0.1341
Epoch 54/100
782/782 4s 5ms/step - accuracy: 0.8826 - loss: 0.3357 - val_acc
uracy: 0.9613 - val_loss: 0.1242
Epoch 55/100
782/782 4s 5ms/step - accuracy: 0.8865 - loss: 0.3306 - val_acc
uracy: 0.9676 - val_loss: 0.1149
Epoch 56/100
782/782 4s 5ms/step - accuracy: 0.8881 - loss: 0.3257 - val_acc
uracy: 0.9658 - val_loss: 0.1123
Epoch 57/100
782/782 4s 5ms/step - accuracy: 0.8820 - loss: 0.3369 - val_acc
uracy: 0.9663 - val_loss: 0.1162
Epoch 58/100
782/782 4s 5ms/step - accuracy: 0.8882 - loss: 0.3275 - val_acc
uracy: 0.9719 - val_loss: 0.0991
Epoch 59/100
782/782 4s 5ms/step - accuracy: 0.8873 - loss: 0.3194 - val_acc
uracy: 0.9660 - val_loss: 0.1070
Epoch 60/100
782/782 4s 5ms/step - accuracy: 0.8880 - loss: 0.3266 - val_acc
uracy: 0.9672 - val_loss: 0.1077
Epoch 61/100
```

782/782 4s 5ms/step - accuracy: 0.8916 - loss: 0.3152 - val\_accuracy: 0.9702 - val\_loss: 0.1020  
Epoch 62/100  
782/782 4s 5ms/step - accuracy: 0.8917 - loss: 0.3128 - val\_accuracy: 0.9652 - val\_loss: 0.1120  
Epoch 63/100  
782/782 4s 5ms/step - accuracy: 0.8903 - loss: 0.3169 - val\_accuracy: 0.9674 - val\_loss: 0.1074  
Epoch 64/100  
782/782 4s 5ms/step - accuracy: 0.8902 - loss: 0.3130 - val\_accuracy: 0.9722 - val\_loss: 0.0970  
Epoch 65/100  
782/782 4s 5ms/step - accuracy: 0.8875 - loss: 0.3186 - val\_accuracy: 0.9719 - val\_loss: 0.1000  
Epoch 66/100  
782/782 4s 5ms/step - accuracy: 0.8936 - loss: 0.3016 - val\_accuracy: 0.9663 - val\_loss: 0.1100  
Epoch 67/100  
782/782 4s 5ms/step - accuracy: 0.8946 - loss: 0.3048 - val\_accuracy: 0.9687 - val\_loss: 0.0999  
Epoch 68/100  
782/782 4s 6ms/step - accuracy: 0.8954 - loss: 0.3041 - val\_accuracy: 0.9754 - val\_loss: 0.0886  
Epoch 69/100  
782/782 5s 5ms/step - accuracy: 0.8970 - loss: 0.2974 - val\_accuracy: 0.9778 - val\_loss: 0.0893  
Epoch 70/100  
782/782 4s 5ms/step - accuracy: 0.8954 - loss: 0.2986 - val\_accuracy: 0.9763 - val\_loss: 0.0871  
Epoch 71/100  
782/782 4s 5ms/step - accuracy: 0.8973 - loss: 0.2934 - val\_accuracy: 0.9785 - val\_loss: 0.0842  
Epoch 72/100  
782/782 4s 5ms/step - accuracy: 0.8973 - loss: 0.2899 - val\_accuracy: 0.9727 - val\_loss: 0.0909  
Epoch 73/100  
782/782 4s 5ms/step - accuracy: 0.8996 - loss: 0.2920 - val\_accuracy: 0.9771 - val\_loss: 0.0848  
Epoch 74/100  
782/782 4s 5ms/step - accuracy: 0.8980 - loss: 0.2914 - val\_accuracy: 0.9780 - val\_loss: 0.0855  
Epoch 75/100  
782/782 4s 5ms/step - accuracy: 0.8986 - loss: 0.2948 - val\_accuracy: 0.9775 - val\_loss: 0.0869  
Epoch 76/100  
782/782 4s 5ms/step - accuracy: 0.8965 - loss: 0.2976 - val\_accuracy: 0.9806 - val\_loss: 0.0746  
Epoch 77/100  
782/782 4s 5ms/step - accuracy: 0.8988 - loss: 0.2881 - val\_accuracy: 0.9753 - val\_loss: 0.0886  
Epoch 78/100  
782/782 4s 5ms/step - accuracy: 0.8992 - loss: 0.2840 - val\_accuracy: 0.9793 - val\_loss: 0.0839  
Epoch 79/100  
782/782 4s 5ms/step - accuracy: 0.9043 - loss: 0.2836 - val\_accuracy: 0.9801 - val\_loss: 0.0770

Epoch 80/100  
**782/782** 4s 5ms/step - accuracy: 0.9030 - loss: 0.2834 - val\_accuracy: 0.9762 - val\_loss: 0.0795  
Epoch 81/100  
**782/782** 4s 5ms/step - accuracy: 0.9029 - loss: 0.2802 - val\_accuracy: 0.9786 - val\_loss: 0.0792  
Epoch 82/100  
**782/782** 4s 5ms/step - accuracy: 0.9051 - loss: 0.2732 - val\_accuracy: 0.9695 - val\_loss: 0.0953  
Epoch 83/100  
**782/782** 4s 5ms/step - accuracy: 0.9021 - loss: 0.2801 - val\_accuracy: 0.9809 - val\_loss: 0.0734  
Epoch 84/100  
**782/782** 4s 5ms/step - accuracy: 0.9069 - loss: 0.2705 - val\_accuracy: 0.9789 - val\_loss: 0.0733  
Epoch 85/100  
**782/782** 4s 5ms/step - accuracy: 0.9047 - loss: 0.2753 - val\_accuracy: 0.9822 - val\_loss: 0.0666  
Epoch 86/100  
**782/782** 4s 5ms/step - accuracy: 0.9046 - loss: 0.2769 - val\_accuracy: 0.9816 - val\_loss: 0.0745  
Epoch 87/100  
**782/782** 4s 5ms/step - accuracy: 0.9038 - loss: 0.2747 - val\_accuracy: 0.9765 - val\_loss: 0.0845  
Epoch 88/100  
**782/782** 4s 5ms/step - accuracy: 0.9077 - loss: 0.2685 - val\_accuracy: 0.9744 - val\_loss: 0.0780  
Epoch 89/100  
**782/782** 4s 5ms/step - accuracy: 0.9049 - loss: 0.2767 - val\_accuracy: 0.9794 - val\_loss: 0.0684  
Epoch 90/100  
**782/782** 4s 5ms/step - accuracy: 0.9077 - loss: 0.2672 - val\_accuracy: 0.9821 - val\_loss: 0.0683  
Epoch 91/100  
**782/782** 4s 5ms/step - accuracy: 0.9048 - loss: 0.2700 - val\_accuracy: 0.9845 - val\_loss: 0.0634  
Epoch 92/100  
**782/782** 4s 5ms/step - accuracy: 0.9078 - loss: 0.2683 - val\_accuracy: 0.9838 - val\_loss: 0.0645  
Epoch 93/100  
**782/782** 4s 5ms/step - accuracy: 0.9074 - loss: 0.2658 - val\_accuracy: 0.9851 - val\_loss: 0.0603  
Epoch 94/100  
**782/782** 4s 5ms/step - accuracy: 0.9068 - loss: 0.2694 - val\_accuracy: 0.9824 - val\_loss: 0.0692  
Epoch 95/100  
**782/782** 4s 5ms/step - accuracy: 0.9061 - loss: 0.2707 - val\_accuracy: 0.9818 - val\_loss: 0.0603  
Epoch 96/100  
**782/782** 4s 5ms/step - accuracy: 0.9102 - loss: 0.2595 - val\_accuracy: 0.9801 - val\_loss: 0.0692  
Epoch 97/100  
**782/782** 4s 5ms/step - accuracy: 0.9087 - loss: 0.2647 - val\_accuracy: 0.9851 - val\_loss: 0.0593  
Epoch 98/100  
**782/782** 4s 5ms/step - accuracy: 0.9068 - loss: 0.2638 - val\_accuracy: 0.9801 - val\_loss: 0.0593

```
uracy: 0.9832 - val_loss: 0.0601
Epoch 99/100
782/782 4s 5ms/step - accuracy: 0.9120 - loss: 0.2523 - val_acc
uracy: 0.9814 - val_loss: 0.0680
Epoch 100/100
782/782 4s 5ms/step - accuracy: 0.9102 - loss: 0.2564 - val_acc
uracy: 0.9860 - val_loss: 0.0572
313/313 1s 2ms/step - accuracy: 0.9869 - loss: 0.0575
Fold Validation Accuracy: 98.60%
Epoch 1/100
782/782 17s 13ms/step - accuracy: 0.1827 - loss: 2.3441 - val_a
ccuracy: 0.4426 - val_loss: 1.5393
Epoch 2/100
782/782 4s 5ms/step - accuracy: 0.4178 - loss: 1.5782 - val_acc
uracy: 0.5305 - val_loss: 1.3009
Epoch 3/100
782/782 4s 5ms/step - accuracy: 0.5131 - loss: 1.3402 - val_acc
uracy: 0.6373 - val_loss: 1.0185
Epoch 4/100
782/782 4s 5ms/step - accuracy: 0.5876 - loss: 1.1522 - val_acc
uracy: 0.6649 - val_loss: 0.9244
Epoch 5/100
782/782 4s 5ms/step - accuracy: 0.6303 - loss: 1.0413 - val_acc
uracy: 0.7160 - val_loss: 0.7905
Epoch 6/100
782/782 4s 5ms/step - accuracy: 0.6659 - loss: 0.9511 - val_acc
uracy: 0.7477 - val_loss: 0.7110
Epoch 7/100
782/782 4s 5ms/step - accuracy: 0.6969 - loss: 0.8690 - val_acc
uracy: 0.7645 - val_loss: 0.6695
Epoch 8/100
782/782 4s 5ms/step - accuracy: 0.7107 - loss: 0.8275 - val_acc
uracy: 0.7921 - val_loss: 0.5957
Epoch 9/100
782/782 4s 5ms/step - accuracy: 0.7324 - loss: 0.7734 - val_acc
uracy: 0.8028 - val_loss: 0.5548
Epoch 10/100
782/782 4s 5ms/step - accuracy: 0.7470 - loss: 0.7339 - val_acc
uracy: 0.8099 - val_loss: 0.5389
Epoch 11/100
782/782 4s 5ms/step - accuracy: 0.7532 - loss: 0.7154 - val_acc
uracy: 0.8351 - val_loss: 0.4700
Epoch 12/100
782/782 4s 5ms/step - accuracy: 0.7620 - loss: 0.6830 - val_acc
uracy: 0.8366 - val_loss: 0.4629
Epoch 13/100
782/782 4s 5ms/step - accuracy: 0.7723 - loss: 0.6522 - val_acc
uracy: 0.8479 - val_loss: 0.4346
Epoch 14/100
782/782 4s 5ms/step - accuracy: 0.7740 - loss: 0.6427 - val_acc
uracy: 0.8562 - val_loss: 0.4096
Epoch 15/100
782/782 4s 5ms/step - accuracy: 0.7885 - loss: 0.6128 - val_acc
uracy: 0.8561 - val_loss: 0.4073
Epoch 16/100
782/782 4s 5ms/step - accuracy: 0.7974 - loss: 0.5813 - val_acc
```

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uracy: 0.8686 - val_loss: 0.3658
Epoch 17/100
782/782 4s 5ms/step - accuracy: 0.8002 - loss: 0.5798 - val_acc
uracy: 0.8703 - val_loss: 0.3689
Epoch 18/100
782/782 4s 5ms/step - accuracy: 0.8090 - loss: 0.5544 - val_acc
uracy: 0.8832 - val_loss: 0.3394
Epoch 19/100
782/782 4s 5ms/step - accuracy: 0.8145 - loss: 0.5385 - val_acc
uracy: 0.8694 - val_loss: 0.3712
Epoch 20/100
782/782 4s 5ms/step - accuracy: 0.8182 - loss: 0.5200 - val_acc
uracy: 0.8913 - val_loss: 0.3104
Epoch 21/100
782/782 4s 5ms/step - accuracy: 0.8199 - loss: 0.5129 - val_acc
uracy: 0.8948 - val_loss: 0.2910
Epoch 22/100
782/782 4s 5ms/step - accuracy: 0.8227 - loss: 0.5097 - val_acc
uracy: 0.8940 - val_loss: 0.3071
Epoch 23/100
782/782 4s 5ms/step - accuracy: 0.8300 - loss: 0.4920 - val_acc
uracy: 0.8926 - val_loss: 0.2984
Epoch 24/100
782/782 4s 5ms/step - accuracy: 0.8281 - loss: 0.4924 - val_acc
uracy: 0.9110 - val_loss: 0.2729
Epoch 25/100
782/782 4s 5ms/step - accuracy: 0.8376 - loss: 0.4670 - val_acc
uracy: 0.9060 - val_loss: 0.2596
Epoch 26/100
782/782 5s 5ms/step - accuracy: 0.8369 - loss: 0.4637 - val_acc
uracy: 0.9159 - val_loss: 0.2482
Epoch 27/100
782/782 4s 5ms/step - accuracy: 0.8403 - loss: 0.4626 - val_acc
uracy: 0.9189 - val_loss: 0.2424
Epoch 28/100
782/782 4s 5ms/step - accuracy: 0.8436 - loss: 0.4437 - val_acc
uracy: 0.9227 - val_loss: 0.2235
Epoch 29/100
782/782 4s 5ms/step - accuracy: 0.8482 - loss: 0.4406 - val_acc
uracy: 0.9243 - val_loss: 0.2353
Epoch 30/100
782/782 4s 5ms/step - accuracy: 0.8468 - loss: 0.4363 - val_acc
uracy: 0.9278 - val_loss: 0.2126
Epoch 31/100
782/782 4s 5ms/step - accuracy: 0.8538 - loss: 0.4199 - val_acc
uracy: 0.9249 - val_loss: 0.2171
Epoch 32/100
782/782 4s 5ms/step - accuracy: 0.8484 - loss: 0.4287 - val_acc
uracy: 0.9238 - val_loss: 0.2138
Epoch 33/100
782/782 4s 5ms/step - accuracy: 0.8580 - loss: 0.4074 - val_acc
uracy: 0.9350 - val_loss: 0.1905
Epoch 34/100
782/782 4s 5ms/step - accuracy: 0.8584 - loss: 0.4000 - val_acc
uracy: 0.9405 - val_loss: 0.1796
Epoch 35/100
```

782/782 4s 5ms/step - accuracy: 0.8603 - loss: 0.3950 - val\_accuracy: 0.9279 - val\_loss: 0.2106  
Epoch 36/100  
782/782 4s 5ms/step - accuracy: 0.8616 - loss: 0.3940 - val\_accuracy: 0.9366 - val\_loss: 0.1954  
Epoch 37/100  
782/782 4s 5ms/step - accuracy: 0.8624 - loss: 0.3916 - val\_accuracy: 0.9376 - val\_loss: 0.1843  
Epoch 38/100  
782/782 4s 5ms/step - accuracy: 0.8646 - loss: 0.3925 - val\_accuracy: 0.9394 - val\_loss: 0.1778  
Epoch 39/100  
782/782 4s 5ms/step - accuracy: 0.8649 - loss: 0.3875 - val\_accuracy: 0.9418 - val\_loss: 0.1679  
Epoch 40/100  
782/782 4s 5ms/step - accuracy: 0.8725 - loss: 0.3667 - val\_accuracy: 0.9454 - val\_loss: 0.1746  
Epoch 41/100  
782/782 4s 5ms/step - accuracy: 0.8694 - loss: 0.3750 - val\_accuracy: 0.9499 - val\_loss: 0.1511  
Epoch 42/100  
782/782 4s 5ms/step - accuracy: 0.8761 - loss: 0.3634 - val\_accuracy: 0.9489 - val\_loss: 0.1459  
Epoch 43/100  
782/782 4s 5ms/step - accuracy: 0.8745 - loss: 0.3623 - val\_accuracy: 0.9523 - val\_loss: 0.1518  
Epoch 44/100  
782/782 4s 5ms/step - accuracy: 0.8759 - loss: 0.3582 - val\_accuracy: 0.9489 - val\_loss: 0.1598  
Epoch 45/100  
782/782 4s 5ms/step - accuracy: 0.8738 - loss: 0.3604 - val\_accuracy: 0.9549 - val\_loss: 0.1445  
Epoch 46/100  
782/782 4s 5ms/step - accuracy: 0.8777 - loss: 0.3480 - val\_accuracy: 0.9553 - val\_loss: 0.1363  
Epoch 47/100  
782/782 4s 5ms/step - accuracy: 0.8786 - loss: 0.3444 - val\_accuracy: 0.9535 - val\_loss: 0.1438  
Epoch 48/100  
782/782 4s 5ms/step - accuracy: 0.8739 - loss: 0.3621 - val\_accuracy: 0.9585 - val\_loss: 0.1259  
Epoch 49/100  
782/782 4s 5ms/step - accuracy: 0.8779 - loss: 0.3442 - val\_accuracy: 0.9611 - val\_loss: 0.1239  
Epoch 50/100  
782/782 4s 5ms/step - accuracy: 0.8838 - loss: 0.3334 - val\_accuracy: 0.9588 - val\_loss: 0.1368  
Epoch 51/100  
782/782 4s 5ms/step - accuracy: 0.8819 - loss: 0.3399 - val\_accuracy: 0.9613 - val\_loss: 0.1168  
Epoch 52/100  
782/782 4s 5ms/step - accuracy: 0.8845 - loss: 0.3325 - val\_accuracy: 0.9571 - val\_loss: 0.1311  
Epoch 53/100  
782/782 4s 5ms/step - accuracy: 0.8868 - loss: 0.3218 - val\_accuracy: 0.9623 - val\_loss: 0.1215

Epoch 54/100  
**782/782** 4s 5ms/step - accuracy: 0.8845 - loss: 0.3292 - val\_accuracy: 0.9542 - val\_loss: 0.1364  
Epoch 55/100  
**782/782** 4s 5ms/step - accuracy: 0.8869 - loss: 0.3266 - val\_accuracy: 0.9616 - val\_loss: 0.1172  
Epoch 56/100  
**782/782** 4s 5ms/step - accuracy: 0.8883 - loss: 0.3157 - val\_accuracy: 0.9663 - val\_loss: 0.1101  
Epoch 57/100  
**782/782** 4s 5ms/step - accuracy: 0.8894 - loss: 0.3172 - val\_accuracy: 0.9665 - val\_loss: 0.1075  
Epoch 58/100  
**782/782** 4s 5ms/step - accuracy: 0.8919 - loss: 0.3107 - val\_accuracy: 0.9611 - val\_loss: 0.1202  
Epoch 59/100  
**782/782** 4s 5ms/step - accuracy: 0.8880 - loss: 0.3126 - val\_accuracy: 0.9688 - val\_loss: 0.1036  
Epoch 60/100  
**782/782** 4s 5ms/step - accuracy: 0.8909 - loss: 0.3050 - val\_accuracy: 0.9688 - val\_loss: 0.1024  
Epoch 61/100  
**782/782** 4s 5ms/step - accuracy: 0.8924 - loss: 0.3027 - val\_accuracy: 0.9698 - val\_loss: 0.1041  
Epoch 62/100  
**782/782** 4s 5ms/step - accuracy: 0.8945 - loss: 0.3022 - val\_accuracy: 0.9666 - val\_loss: 0.1047  
Epoch 63/100  
**782/782** 4s 5ms/step - accuracy: 0.8910 - loss: 0.3108 - val\_accuracy: 0.9724 - val\_loss: 0.0968  
Epoch 64/100  
**782/782** 4s 5ms/step - accuracy: 0.8947 - loss: 0.3033 - val\_accuracy: 0.9663 - val\_loss: 0.1071  
Epoch 65/100  
**782/782** 4s 5ms/step - accuracy: 0.8938 - loss: 0.3045 - val\_accuracy: 0.9723 - val\_loss: 0.0998  
Epoch 66/100  
**782/782** 4s 5ms/step - accuracy: 0.8946 - loss: 0.3021 - val\_accuracy: 0.9738 - val\_loss: 0.0917  
Epoch 67/100  
**782/782** 4s 5ms/step - accuracy: 0.8962 - loss: 0.2969 - val\_accuracy: 0.9716 - val\_loss: 0.0897  
Epoch 68/100  
**782/782** 4s 5ms/step - accuracy: 0.8982 - loss: 0.2960 - val\_accuracy: 0.9726 - val\_loss: 0.0928  
Epoch 69/100  
**782/782** 4s 5ms/step - accuracy: 0.8955 - loss: 0.3006 - val\_accuracy: 0.9773 - val\_loss: 0.0852  
Epoch 70/100  
**782/782** 4s 5ms/step - accuracy: 0.8981 - loss: 0.2931 - val\_accuracy: 0.9741 - val\_loss: 0.0920  
Epoch 71/100  
**782/782** 4s 5ms/step - accuracy: 0.8992 - loss: 0.2907 - val\_accuracy: 0.9796 - val\_loss: 0.0777  
Epoch 72/100  
**782/782** 4s 5ms/step - accuracy: 0.8994 - loss: 0.2907 - val\_accuracy: 0.9796 - val\_loss: 0.0777

```
uracy: 0.9771 - val_loss: 0.0817
Epoch 73/100
782/782 4s 5ms/step - accuracy: 0.9025 - loss: 0.2845 - val_acc
uracy: 0.9794 - val_loss: 0.0769
Epoch 74/100
782/782 4s 5ms/step - accuracy: 0.8991 - loss: 0.2847 - val_acc
uracy: 0.9798 - val_loss: 0.0775
Epoch 75/100
782/782 4s 5ms/step - accuracy: 0.9014 - loss: 0.2841 - val_acc
uracy: 0.9751 - val_loss: 0.0862
Epoch 76/100
782/782 4s 5ms/step - accuracy: 0.9022 - loss: 0.2809 - val_acc
uracy: 0.9758 - val_loss: 0.0825
Epoch 77/100
782/782 4s 5ms/step - accuracy: 0.9071 - loss: 0.2706 - val_acc
uracy: 0.9784 - val_loss: 0.0783
Epoch 78/100
782/782 4s 5ms/step - accuracy: 0.9040 - loss: 0.2752 - val_acc
uracy: 0.9821 - val_loss: 0.0707
Epoch 79/100
782/782 4s 5ms/step - accuracy: 0.9021 - loss: 0.2871 - val_acc
uracy: 0.9787 - val_loss: 0.0765
Epoch 80/100
782/782 4s 5ms/step - accuracy: 0.9034 - loss: 0.2682 - val_acc
uracy: 0.9789 - val_loss: 0.0723
Epoch 81/100
782/782 4s 5ms/step - accuracy: 0.9051 - loss: 0.2742 - val_acc
uracy: 0.9829 - val_loss: 0.0692
Epoch 82/100
782/782 4s 5ms/step - accuracy: 0.9054 - loss: 0.2691 - val_acc
uracy: 0.9782 - val_loss: 0.0744
Epoch 83/100
782/782 4s 5ms/step - accuracy: 0.9073 - loss: 0.2660 - val_acc
uracy: 0.9802 - val_loss: 0.0690
Epoch 84/100
782/782 4s 5ms/step - accuracy: 0.9071 - loss: 0.2718 - val_acc
uracy: 0.9824 - val_loss: 0.0664
Epoch 85/100
782/782 4s 5ms/step - accuracy: 0.9079 - loss: 0.2686 - val_acc
uracy: 0.9839 - val_loss: 0.0648
Epoch 86/100
782/782 4s 5ms/step - accuracy: 0.9049 - loss: 0.2677 - val_acc
uracy: 0.9775 - val_loss: 0.0826
Epoch 87/100
782/782 4s 5ms/step - accuracy: 0.9109 - loss: 0.2562 - val_acc
uracy: 0.9849 - val_loss: 0.0612
Epoch 88/100
782/782 4s 5ms/step - accuracy: 0.9057 - loss: 0.2702 - val_acc
uracy: 0.9852 - val_loss: 0.0559
Epoch 89/100
782/782 4s 5ms/step - accuracy: 0.9101 - loss: 0.2560 - val_acc
uracy: 0.9844 - val_loss: 0.0662
Epoch 90/100
782/782 4s 5ms/step - accuracy: 0.9087 - loss: 0.2591 - val_acc
uracy: 0.9822 - val_loss: 0.0684
Epoch 91/100
```

782/782 4s 5ms/step - accuracy: 0.9107 - loss: 0.2609 - val\_accuracy: 0.9834 - val\_loss: 0.0657  
Epoch 92/100  
782/782 4s 5ms/step - accuracy: 0.9128 - loss: 0.2573 - val\_accuracy: 0.9884 - val\_loss: 0.0571  
Epoch 93/100  
782/782 4s 5ms/step - accuracy: 0.9109 - loss: 0.2603 - val\_accuracy: 0.9836 - val\_loss: 0.0635  
Epoch 94/100  
782/782 4s 5ms/step - accuracy: 0.9097 - loss: 0.2648 - val\_accuracy: 0.9847 - val\_loss: 0.0576  
Epoch 95/100  
782/782 4s 5ms/step - accuracy: 0.9106 - loss: 0.2548 - val\_accuracy: 0.9813 - val\_loss: 0.0671  
Epoch 96/100  
782/782 4s 5ms/step - accuracy: 0.9121 - loss: 0.2502 - val\_accuracy: 0.9841 - val\_loss: 0.0569  
Epoch 97/100  
782/782 4s 5ms/step - accuracy: 0.9119 - loss: 0.2509 - val\_accuracy: 0.9838 - val\_loss: 0.0606  
Epoch 98/100  
782/782 4s 5ms/step - accuracy: 0.9088 - loss: 0.2656 - val\_accuracy: 0.9858 - val\_loss: 0.0592  
Epoch 99/100  
782/782 4s 5ms/step - accuracy: 0.9144 - loss: 0.2489 - val\_accuracy: 0.9797 - val\_loss: 0.0651  
Epoch 100/100  
782/782 4s 5ms/step - accuracy: 0.9110 - loss: 0.2497 - val\_accuracy: 0.9869 - val\_loss: 0.0543  
313/313 1s 2ms/step - accuracy: 0.9873 - loss: 0.0531  
Fold Validation Accuracy: 98.69%  
Epoch 1/100  
782/782 18s 13ms/step - accuracy: 0.2223 - loss: 2.1839 - val\_accuracy: 0.5011 - val\_loss: 1.3754  
Epoch 2/100  
782/782 4s 5ms/step - accuracy: 0.4618 - loss: 1.4759 - val\_accuracy: 0.5775 - val\_loss: 1.1848  
Epoch 3/100  
782/782 4s 5ms/step - accuracy: 0.5517 - loss: 1.2524 - val\_accuracy: 0.6485 - val\_loss: 0.9740  
Epoch 4/100  
782/782 4s 5ms/step - accuracy: 0.6123 - loss: 1.1039 - val\_accuracy: 0.6948 - val\_loss: 0.8521  
Epoch 5/100  
782/782 4s 6ms/step - accuracy: 0.6560 - loss: 0.9824 - val\_accuracy: 0.7190 - val\_loss: 0.7809  
Epoch 6/100  
782/782 4s 5ms/step - accuracy: 0.6846 - loss: 0.9005 - val\_accuracy: 0.7491 - val\_loss: 0.6918  
Epoch 7/100  
782/782 4s 5ms/step - accuracy: 0.7062 - loss: 0.8456 - val\_accuracy: 0.7783 - val\_loss: 0.6207  
Epoch 8/100  
782/782 4s 5ms/step - accuracy: 0.7296 - loss: 0.7887 - val\_accuracy: 0.7907 - val\_loss: 0.5715  
Epoch 9/100

782/782 4s 5ms/step - accuracy: 0.7388 - loss: 0.7614 - val\_accuracy: 0.7989 - val\_loss: 0.5766  
Epoch 10/100  
782/782 4s 5ms/step - accuracy: 0.7514 - loss: 0.7272 - val\_accuracy: 0.8086 - val\_loss: 0.5395  
Epoch 11/100  
782/782 4s 5ms/step - accuracy: 0.7592 - loss: 0.6914 - val\_accuracy: 0.8297 - val\_loss: 0.4764  
Epoch 12/100  
782/782 4s 5ms/step - accuracy: 0.7666 - loss: 0.6737 - val\_accuracy: 0.8490 - val\_loss: 0.4423  
Epoch 13/100  
782/782 4s 5ms/step - accuracy: 0.7755 - loss: 0.6422 - val\_accuracy: 0.8524 - val\_loss: 0.4170  
Epoch 14/100  
782/782 4s 5ms/step - accuracy: 0.7869 - loss: 0.6186 - val\_accuracy: 0.8536 - val\_loss: 0.4068  
Epoch 15/100  
782/782 4s 5ms/step - accuracy: 0.7906 - loss: 0.6034 - val\_accuracy: 0.8627 - val\_loss: 0.3875  
Epoch 16/100  
782/782 4s 5ms/step - accuracy: 0.7995 - loss: 0.5803 - val\_accuracy: 0.8745 - val\_loss: 0.3564  
Epoch 17/100  
782/782 4s 5ms/step - accuracy: 0.8059 - loss: 0.5620 - val\_accuracy: 0.8665 - val\_loss: 0.3632  
Epoch 18/100  
782/782 4s 5ms/step - accuracy: 0.8097 - loss: 0.5479 - val\_accuracy: 0.8818 - val\_loss: 0.3395  
Epoch 19/100  
782/782 4s 5ms/step - accuracy: 0.8174 - loss: 0.5344 - val\_accuracy: 0.8815 - val\_loss: 0.3355  
Epoch 20/100  
782/782 4s 5ms/step - accuracy: 0.8216 - loss: 0.5144 - val\_accuracy: 0.8950 - val\_loss: 0.2947  
Epoch 21/100  
782/782 4s 5ms/step - accuracy: 0.8208 - loss: 0.5150 - val\_accuracy: 0.8963 - val\_loss: 0.2937  
Epoch 22/100  
782/782 4s 5ms/step - accuracy: 0.8261 - loss: 0.4998 - val\_accuracy: 0.8941 - val\_loss: 0.2830  
Epoch 23/100  
782/782 4s 5ms/step - accuracy: 0.8291 - loss: 0.4877 - val\_accuracy: 0.9064 - val\_loss: 0.2714  
Epoch 24/100  
782/782 4s 5ms/step - accuracy: 0.8331 - loss: 0.4762 - val\_accuracy: 0.9089 - val\_loss: 0.2656  
Epoch 25/100  
782/782 4s 5ms/step - accuracy: 0.8343 - loss: 0.4671 - val\_accuracy: 0.9203 - val\_loss: 0.2393  
Epoch 26/100  
782/782 4s 5ms/step - accuracy: 0.8408 - loss: 0.4598 - val\_accuracy: 0.9155 - val\_loss: 0.2541  
Epoch 27/100  
782/782 4s 5ms/step - accuracy: 0.8426 - loss: 0.4554 - val\_accuracy: 0.9125 - val\_loss: 0.2513

Epoch 28/100  
**782/782** 4s 5ms/step - accuracy: 0.8446 - loss: 0.4434 - val\_accuracy: 0.9132 - val\_loss: 0.2373  
Epoch 29/100  
**782/782** 4s 5ms/step - accuracy: 0.8465 - loss: 0.4317 - val\_accuracy: 0.9289 - val\_loss: 0.2195  
Epoch 30/100  
**782/782** 4s 5ms/step - accuracy: 0.8509 - loss: 0.4258 - val\_accuracy: 0.9264 - val\_loss: 0.2189  
Epoch 31/100  
**782/782** 4s 5ms/step - accuracy: 0.8512 - loss: 0.4283 - val\_accuracy: 0.9329 - val\_loss: 0.2073  
Epoch 32/100  
**782/782** 4s 5ms/step - accuracy: 0.8561 - loss: 0.4151 - val\_accuracy: 0.9156 - val\_loss: 0.2380  
Epoch 33/100  
**782/782** 4s 5ms/step - accuracy: 0.8549 - loss: 0.4122 - val\_accuracy: 0.9179 - val\_loss: 0.2304  
Epoch 34/100  
**782/782** 4s 5ms/step - accuracy: 0.8591 - loss: 0.4016 - val\_accuracy: 0.9320 - val\_loss: 0.1909  
Epoch 35/100  
**782/782** 4s 5ms/step - accuracy: 0.8593 - loss: 0.4020 - val\_accuracy: 0.9312 - val\_loss: 0.2063  
Epoch 36/100  
**782/782** 4s 5ms/step - accuracy: 0.8627 - loss: 0.3975 - val\_accuracy: 0.9351 - val\_loss: 0.1926  
Epoch 37/100  
**782/782** 4s 5ms/step - accuracy: 0.8633 - loss: 0.3903 - val\_accuracy: 0.9376 - val\_loss: 0.1859  
Epoch 38/100  
**782/782** 4s 5ms/step - accuracy: 0.8610 - loss: 0.3994 - val\_accuracy: 0.9408 - val\_loss: 0.1734  
Epoch 39/100  
**782/782** 4s 5ms/step - accuracy: 0.8657 - loss: 0.3777 - val\_accuracy: 0.9429 - val\_loss: 0.1761  
Epoch 40/100  
**782/782** 4s 5ms/step - accuracy: 0.8668 - loss: 0.3769 - val\_accuracy: 0.9464 - val\_loss: 0.1626  
Epoch 41/100  
**782/782** 4s 5ms/step - accuracy: 0.8734 - loss: 0.3626 - val\_accuracy: 0.9450 - val\_loss: 0.1616  
Epoch 42/100  
**782/782** 4s 5ms/step - accuracy: 0.8702 - loss: 0.3698 - val\_accuracy: 0.9488 - val\_loss: 0.1551  
Epoch 43/100  
**782/782** 4s 5ms/step - accuracy: 0.8710 - loss: 0.3694 - val\_accuracy: 0.9473 - val\_loss: 0.1570  
Epoch 44/100  
**782/782** 4s 5ms/step - accuracy: 0.8697 - loss: 0.3687 - val\_accuracy: 0.9447 - val\_loss: 0.1627  
Epoch 45/100  
**782/782** 4s 5ms/step - accuracy: 0.8747 - loss: 0.3574 - val\_accuracy: 0.9482 - val\_loss: 0.1576  
Epoch 46/100  
**782/782** 4s 5ms/step - accuracy: 0.8751 - loss: 0.3561 - val\_accuracy: 0.9500 - val\_loss: 0.1550

```
uracy: 0.9488 - val_loss: 0.1563
Epoch 47/100
782/782 4s 5ms/step - accuracy: 0.8776 - loss: 0.3484 - val_acc
uracy: 0.9511 - val_loss: 0.1462
Epoch 48/100
782/782 4s 5ms/step - accuracy: 0.8775 - loss: 0.3506 - val_acc
uracy: 0.9525 - val_loss: 0.1408
Epoch 49/100
782/782 4s 5ms/step - accuracy: 0.8782 - loss: 0.3494 - val_acc
uracy: 0.9607 - val_loss: 0.1271
Epoch 50/100
782/782 4s 5ms/step - accuracy: 0.8798 - loss: 0.3459 - val_acc
uracy: 0.9579 - val_loss: 0.1338
Epoch 51/100
782/782 4s 5ms/step - accuracy: 0.8769 - loss: 0.3464 - val_acc
uracy: 0.9566 - val_loss: 0.1337
Epoch 52/100
782/782 4s 5ms/step - accuracy: 0.8815 - loss: 0.3379 - val_acc
uracy: 0.9587 - val_loss: 0.1287
Epoch 53/100
782/782 4s 5ms/step - accuracy: 0.8855 - loss: 0.3241 - val_acc
uracy: 0.9603 - val_loss: 0.1286
Epoch 54/100
782/782 4s 5ms/step - accuracy: 0.8840 - loss: 0.3231 - val_acc
uracy: 0.9627 - val_loss: 0.1150
Epoch 55/100
782/782 4s 5ms/step - accuracy: 0.8905 - loss: 0.3147 - val_acc
uracy: 0.9534 - val_loss: 0.1404
Epoch 56/100
782/782 4s 5ms/step - accuracy: 0.8889 - loss: 0.3248 - val_acc
uracy: 0.9622 - val_loss: 0.1213
Epoch 57/100
782/782 4s 5ms/step - accuracy: 0.8852 - loss: 0.3289 - val_acc
uracy: 0.9610 - val_loss: 0.1249
Epoch 58/100
782/782 4s 5ms/step - accuracy: 0.8859 - loss: 0.3225 - val_acc
uracy: 0.9648 - val_loss: 0.1075
Epoch 59/100
782/782 4s 5ms/step - accuracy: 0.8919 - loss: 0.3056 - val_acc
uracy: 0.9659 - val_loss: 0.1105
Epoch 60/100
782/782 4s 5ms/step - accuracy: 0.8870 - loss: 0.3231 - val_acc
uracy: 0.9678 - val_loss: 0.1068
Epoch 61/100
782/782 4s 5ms/step - accuracy: 0.8935 - loss: 0.3029 - val_acc
uracy: 0.9697 - val_loss: 0.1032
Epoch 62/100
782/782 4s 5ms/step - accuracy: 0.8931 - loss: 0.3053 - val_acc
uracy: 0.9635 - val_loss: 0.1160
Epoch 63/100
782/782 4s 5ms/step - accuracy: 0.8931 - loss: 0.3068 - val_acc
uracy: 0.9681 - val_loss: 0.1041
Epoch 64/100
782/782 4s 5ms/step - accuracy: 0.8944 - loss: 0.3032 - val_acc
uracy: 0.9684 - val_loss: 0.1057
Epoch 65/100
```

782/782 4s 5ms/step - accuracy: 0.8933 - loss: 0.3067 - val\_accuracy: 0.9688 - val\_loss: 0.1035  
Epoch 66/100  
782/782 4s 5ms/step - accuracy: 0.8900 - loss: 0.3152 - val\_accuracy: 0.9684 - val\_loss: 0.1047  
Epoch 67/100  
782/782 4s 5ms/step - accuracy: 0.8934 - loss: 0.3011 - val\_accuracy: 0.9718 - val\_loss: 0.0914  
Epoch 68/100  
782/782 4s 5ms/step - accuracy: 0.8964 - loss: 0.2968 - val\_accuracy: 0.9709 - val\_loss: 0.0909  
Epoch 69/100  
782/782 4s 5ms/step - accuracy: 0.8965 - loss: 0.2928 - val\_accuracy: 0.9636 - val\_loss: 0.1041  
Epoch 70/100  
782/782 4s 5ms/step - accuracy: 0.8991 - loss: 0.2918 - val\_accuracy: 0.9729 - val\_loss: 0.0939  
Epoch 71/100  
782/782 4s 5ms/step - accuracy: 0.8962 - loss: 0.2974 - val\_accuracy: 0.9679 - val\_loss: 0.1015  
Epoch 72/100  
782/782 4s 5ms/step - accuracy: 0.8988 - loss: 0.2981 - val\_accuracy: 0.9662 - val\_loss: 0.1022  
Epoch 73/100  
782/782 4s 5ms/step - accuracy: 0.8974 - loss: 0.2988 - val\_accuracy: 0.9789 - val\_loss: 0.0809  
Epoch 74/100  
782/782 4s 5ms/step - accuracy: 0.8985 - loss: 0.2929 - val\_accuracy: 0.9718 - val\_loss: 0.0985  
Epoch 75/100  
782/782 4s 5ms/step - accuracy: 0.8989 - loss: 0.2882 - val\_accuracy: 0.9789 - val\_loss: 0.0804  
Epoch 76/100  
782/782 4s 5ms/step - accuracy: 0.9026 - loss: 0.2809 - val\_accuracy: 0.9731 - val\_loss: 0.0908  
Epoch 77/100  
782/782 4s 5ms/step - accuracy: 0.9016 - loss: 0.2839 - val\_accuracy: 0.9741 - val\_loss: 0.0894  
Epoch 78/100  
782/782 4s 5ms/step - accuracy: 0.9032 - loss: 0.2782 - val\_accuracy: 0.9774 - val\_loss: 0.0786  
Epoch 79/100  
782/782 4s 5ms/step - accuracy: 0.9039 - loss: 0.2766 - val\_accuracy: 0.9739 - val\_loss: 0.0863  
Epoch 80/100  
782/782 4s 5ms/step - accuracy: 0.9018 - loss: 0.2819 - val\_accuracy: 0.9756 - val\_loss: 0.0806  
Epoch 81/100  
782/782 4s 5ms/step - accuracy: 0.9049 - loss: 0.2726 - val\_accuracy: 0.9755 - val\_loss: 0.0816  
Epoch 82/100  
782/782 4s 5ms/step - accuracy: 0.9017 - loss: 0.2817 - val\_accuracy: 0.9827 - val\_loss: 0.0673  
Epoch 83/100  
782/782 4s 5ms/step - accuracy: 0.9067 - loss: 0.2705 - val\_accuracy: 0.9793 - val\_loss: 0.0807

Epoch 84/100  
**782/782** 4s 5ms/step - accuracy: 0.9043 - loss: 0.2740 - val\_accuracy: 0.9744 - val\_loss: 0.0819  
Epoch 85/100  
**782/782** 4s 5ms/step - accuracy: 0.9069 - loss: 0.2699 - val\_accuracy: 0.9790 - val\_loss: 0.0765  
Epoch 86/100  
**782/782** 4s 5ms/step - accuracy: 0.9061 - loss: 0.2664 - val\_accuracy: 0.9816 - val\_loss: 0.0692  
Epoch 87/100  
**782/782** 4s 5ms/step - accuracy: 0.9042 - loss: 0.2753 - val\_accuracy: 0.9816 - val\_loss: 0.0704  
Epoch 88/100  
**782/782** 4s 5ms/step - accuracy: 0.9078 - loss: 0.2656 - val\_accuracy: 0.9817 - val\_loss: 0.0663  
Epoch 89/100  
**782/782** 4s 5ms/step - accuracy: 0.9066 - loss: 0.2628 - val\_accuracy: 0.9801 - val\_loss: 0.0670  
Epoch 90/100  
**782/782** 4s 5ms/step - accuracy: 0.9077 - loss: 0.2655 - val\_accuracy: 0.9818 - val\_loss: 0.0627  
Epoch 91/100  
**782/782** 4s 5ms/step - accuracy: 0.9082 - loss: 0.2605 - val\_accuracy: 0.9825 - val\_loss: 0.0628  
Epoch 92/100  
**782/782** 4s 5ms/step - accuracy: 0.9098 - loss: 0.2595 - val\_accuracy: 0.9832 - val\_loss: 0.0654  
Epoch 93/100  
**782/782** 4s 5ms/step - accuracy: 0.9079 - loss: 0.2657 - val\_accuracy: 0.9827 - val\_loss: 0.0648  
Epoch 94/100  
**782/782** 4s 5ms/step - accuracy: 0.9124 - loss: 0.2488 - val\_accuracy: 0.9855 - val\_loss: 0.0592  
Epoch 95/100  
**782/782** 4s 5ms/step - accuracy: 0.9101 - loss: 0.2611 - val\_accuracy: 0.9851 - val\_loss: 0.0560  
Epoch 96/100  
**782/782** 4s 5ms/step - accuracy: 0.9114 - loss: 0.2569 - val\_accuracy: 0.9830 - val\_loss: 0.0658  
Epoch 97/100  
**782/782** 4s 5ms/step - accuracy: 0.9106 - loss: 0.2567 - val\_accuracy: 0.9807 - val\_loss: 0.0702  
Epoch 98/100  
**782/782** 4s 5ms/step - accuracy: 0.9100 - loss: 0.2591 - val\_accuracy: 0.9824 - val\_loss: 0.0680  
Epoch 99/100  
**782/782** 4s 5ms/step - accuracy: 0.9110 - loss: 0.2600 - val\_accuracy: 0.9847 - val\_loss: 0.0581  
Epoch 100/100  
**782/782** 4s 5ms/step - accuracy: 0.9121 - loss: 0.2517 - val\_accuracy: 0.9842 - val\_loss: 0.0622  
**313/313** 1s 2ms/step - accuracy: 0.9848 - loss: 0.0621  
Fold Validation Accuracy: 98.42%

Average 5-Fold Cross-Validation Accuracy: 98.37%

```
In [15]: evaluate_on_test(model4)
```

```
Epoch 1/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9094 - loss: 0.2593  
Epoch 2/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9161 - loss: 0.2425  
Epoch 3/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9135 - loss: 0.2497  
Epoch 4/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9107 - loss: 0.2504  
Epoch 5/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9125 - loss: 0.2525  
Epoch 6/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9145 - loss: 0.2462  
Epoch 7/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9134 - loss: 0.2514  
Epoch 8/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9133 - loss: 0.2517  
Epoch 9/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9136 - loss: 0.2517  
Epoch 10/10  
782/782 ━━━━━━━━━━ 4s 5ms/step - accuracy: 0.9141 - loss: 0.2489  
313/313 ━━━━━━━━ 1s 2ms/step - accuracy: 0.8484 - loss: 0.5952
```

Test Accuracy: 0.8454999923706055

Test Loss: 0.5983885526657104

```
Out[15]: 0.8454999923706055
```

```
In [20]: GenerateClassificationReport(model4)
```

```
313/313 ━━━━━━━━ 1s 2ms/step
```

Confusion Matrix											
True Class	0	1	2	3	4	5	6	7	8	9	
	874	4	30	6	7	1	8	2	55	13	
	8	897	2	1	1	1	4	1	26	59	
	44	0	815	19	38	15	51	9	7	2	
	24	2	70	636	44	81	92	25	18	8	
	14	0	60	14	838	10	40	16	6	2	
	11	2	51	106	40	727	31	23	5	4	
	3	0	28	11	10	3	937	2	5	1	
	16	1	23	18	45	18	2	873	3	1	
	35	4	3	0	3	1	3	2	941	8	
Predicted Class											

	precision	recall	f1-score	support
0	0.83	0.87	0.85	1000
1	0.96	0.90	0.93	1000
2	0.75	0.81	0.78	1000
3	0.78	0.64	0.70	1000
4	0.82	0.84	0.83	1000
5	0.85	0.73	0.78	1000
6	0.80	0.94	0.86	1000
7	0.91	0.87	0.89	1000
8	0.87	0.94	0.90	1000
9	0.90	0.92	0.91	1000
accuracy			0.85	10000
macro avg	0.85	0.85	0.84	10000
weighted avg	0.85	0.85	0.84	10000

Class 0: Precision = 0.83, Recall = 0.87  
 Class 1: Precision = 0.96, Recall = 0.90  
 Class 2: Precision = 0.75, Recall = 0.81  
 Class 3: Precision = 0.78, Recall = 0.64  
 Class 4: Precision = 0.82, Recall = 0.84  
 Class 5: Precision = 0.85, Recall = 0.73  
 Class 6: Precision = 0.80, Recall = 0.94  
 Class 7: Precision = 0.91, Recall = 0.87  
 Class 8: Precision = 0.87, Recall = 0.94  
 Class 9: Precision = 0.90, Recall = 0.92

With it being trained for 100 epochs, the accuracy has improved a little giving it an overall testing accuracy of 84.5%

Now we train the model with new architecture which includes batch normalization in every block along with dropout.

```
In [21]: model2,histories= CrossVal_Training("with_batch_norm",100,64)
```

```
/opt/conda/lib/python3.10/site-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)
```

Epoch 1/100  
**782/782** 24s 17ms/step - accuracy: 0.3068 - loss: 2.2413 - val\_accuracy: 0.5316 - val\_loss: 1.3282  
Epoch 2/100  
**782/782** 5s 6ms/step - accuracy: 0.5157 - loss: 1.3581 - val\_accuracy: 0.6083 - val\_loss: 1.0900  
Epoch 3/100  
**782/782** 5s 6ms/step - accuracy: 0.6087 - loss: 1.1009 - val\_accuracy: 0.6998 - val\_loss: 0.8449  
Epoch 4/100  
**782/782** 5s 6ms/step - accuracy: 0.6558 - loss: 0.9747 - val\_accuracy: 0.7385 - val\_loss: 0.7305  
Epoch 5/100  
**782/782** 5s 6ms/step - accuracy: 0.6960 - loss: 0.8685 - val\_accuracy: 0.7779 - val\_loss: 0.6315  
Epoch 6/100  
**782/782** 5s 6ms/step - accuracy: 0.7239 - loss: 0.7900 - val\_accuracy: 0.7863 - val\_loss: 0.6051  
Epoch 7/100  
**782/782** 5s 6ms/step - accuracy: 0.7435 - loss: 0.7309 - val\_accuracy: 0.8149 - val\_loss: 0.5250  
Epoch 8/100  
**782/782** 5s 6ms/step - accuracy: 0.7657 - loss: 0.6746 - val\_accuracy: 0.8317 - val\_loss: 0.4832  
Epoch 9/100  
**782/782** 5s 6ms/step - accuracy: 0.7763 - loss: 0.6456 - val\_accuracy: 0.8509 - val\_loss: 0.4295  
Epoch 10/100  
**782/782** 5s 6ms/step - accuracy: 0.7963 - loss: 0.5923 - val\_accuracy: 0.8554 - val\_loss: 0.3996  
Epoch 11/100  
**782/782** 5s 6ms/step - accuracy: 0.8063 - loss: 0.5582 - val\_accuracy: 0.8713 - val\_loss: 0.3800  
Epoch 12/100  
**782/782** 5s 6ms/step - accuracy: 0.8172 - loss: 0.5311 - val\_accuracy: 0.8828 - val\_loss: 0.3595  
Epoch 13/100  
**782/782** 5s 6ms/step - accuracy: 0.8227 - loss: 0.5099 - val\_accuracy: 0.8904 - val\_loss: 0.3202  
Epoch 14/100  
**782/782** 5s 6ms/step - accuracy: 0.8302 - loss: 0.4898 - val\_accuracy: 0.9010 - val\_loss: 0.2883  
Epoch 15/100  
**782/782** 5s 6ms/step - accuracy: 0.8383 - loss: 0.4689 - val\_accuracy: 0.9047 - val\_loss: 0.2869  
Epoch 16/100  
**782/782** 5s 6ms/step - accuracy: 0.8476 - loss: 0.4513 - val\_accuracy: 0.9172 - val\_loss: 0.2513  
Epoch 17/100  
**782/782** 5s 6ms/step - accuracy: 0.8520 - loss: 0.4282 - val\_accuracy: 0.9162 - val\_loss: 0.2452  
Epoch 18/100  
**782/782** 5s 6ms/step - accuracy: 0.8577 - loss: 0.4070 - val\_accuracy: 0.9209 - val\_loss: 0.2286  
Epoch 19/100  
**782/782** 5s 6ms/step - accuracy: 0.8619 - loss: 0.4010 - val\_accuracy:

```
uracy: 0.9282 - val_loss: 0.2108
Epoch 20/100
782/782 5s 6ms/step - accuracy: 0.8667 - loss: 0.3851 - val_acc
uracy: 0.9343 - val_loss: 0.1959
Epoch 21/100
782/782 5s 6ms/step - accuracy: 0.8704 - loss: 0.3762 - val_acc
uracy: 0.9402 - val_loss: 0.1805
Epoch 22/100
782/782 5s 6ms/step - accuracy: 0.8752 - loss: 0.3661 - val_acc
uracy: 0.9467 - val_loss: 0.1712
Epoch 23/100
782/782 5s 6ms/step - accuracy: 0.8790 - loss: 0.3510 - val_acc
uracy: 0.9516 - val_loss: 0.1530
Epoch 24/100
782/782 5s 6ms/step - accuracy: 0.8816 - loss: 0.3410 - val_acc
uracy: 0.9534 - val_loss: 0.1461
Epoch 25/100
782/782 5s 6ms/step - accuracy: 0.8864 - loss: 0.3273 - val_acc
uracy: 0.9534 - val_loss: 0.1421
Epoch 26/100
782/782 5s 6ms/step - accuracy: 0.8888 - loss: 0.3196 - val_acc
uracy: 0.9536 - val_loss: 0.1408
Epoch 27/100
782/782 5s 6ms/step - accuracy: 0.8841 - loss: 0.3260 - val_acc
uracy: 0.9577 - val_loss: 0.1310
Epoch 28/100
782/782 5s 6ms/step - accuracy: 0.8920 - loss: 0.3118 - val_acc
uracy: 0.9584 - val_loss: 0.1264
Epoch 29/100
782/782 5s 6ms/step - accuracy: 0.8911 - loss: 0.3095 - val_acc
uracy: 0.9665 - val_loss: 0.1118
Epoch 30/100
782/782 5s 6ms/step - accuracy: 0.8971 - loss: 0.2987 - val_acc
uracy: 0.9645 - val_loss: 0.1130
Epoch 31/100
782/782 5s 6ms/step - accuracy: 0.9038 - loss: 0.2784 - val_acc
uracy: 0.9702 - val_loss: 0.0990
Epoch 32/100
782/782 5s 6ms/step - accuracy: 0.9022 - loss: 0.2812 - val_acc
uracy: 0.9709 - val_loss: 0.0929
Epoch 33/100
782/782 5s 6ms/step - accuracy: 0.9017 - loss: 0.2772 - val_acc
uracy: 0.9727 - val_loss: 0.0941
Epoch 34/100
782/782 5s 6ms/step - accuracy: 0.9102 - loss: 0.2629 - val_acc
uracy: 0.9730 - val_loss: 0.0871
Epoch 35/100
782/782 5s 6ms/step - accuracy: 0.9078 - loss: 0.2677 - val_acc
uracy: 0.9749 - val_loss: 0.0854
Epoch 36/100
782/782 5s 6ms/step - accuracy: 0.9082 - loss: 0.2603 - val_acc
uracy: 0.9776 - val_loss: 0.0752
Epoch 37/100
782/782 5s 6ms/step - accuracy: 0.9097 - loss: 0.2537 - val_acc
uracy: 0.9783 - val_loss: 0.0744
Epoch 38/100
```

782/782 5s 6ms/step - accuracy: 0.9160 - loss: 0.2431 - val\_accuracy: 0.9829 - val\_loss: 0.0635  
Epoch 39/100  
782/782 5s 6ms/step - accuracy: 0.9124 - loss: 0.2470 - val\_accuracy: 0.9805 - val\_loss: 0.0694  
Epoch 40/100  
782/782 5s 6ms/step - accuracy: 0.9153 - loss: 0.2441 - val\_accuracy: 0.9805 - val\_loss: 0.0641  
Epoch 41/100  
782/782 5s 6ms/step - accuracy: 0.9153 - loss: 0.2374 - val\_accuracy: 0.9824 - val\_loss: 0.0652  
Epoch 42/100  
782/782 5s 6ms/step - accuracy: 0.9183 - loss: 0.2353 - val\_accuracy: 0.9804 - val\_loss: 0.0664  
Epoch 43/100  
782/782 5s 6ms/step - accuracy: 0.9217 - loss: 0.2266 - val\_accuracy: 0.9855 - val\_loss: 0.0528  
Epoch 44/100  
782/782 5s 6ms/step - accuracy: 0.9182 - loss: 0.2288 - val\_accuracy: 0.9860 - val\_loss: 0.0513  
Epoch 45/100  
782/782 5s 6ms/step - accuracy: 0.9239 - loss: 0.2186 - val\_accuracy: 0.9865 - val\_loss: 0.0477  
Epoch 46/100  
782/782 5s 6ms/step - accuracy: 0.9256 - loss: 0.2145 - val\_accuracy: 0.9891 - val\_loss: 0.0448  
Epoch 47/100  
782/782 5s 6ms/step - accuracy: 0.9252 - loss: 0.2094 - val\_accuracy: 0.9861 - val\_loss: 0.0482  
Epoch 48/100  
782/782 5s 6ms/step - accuracy: 0.9259 - loss: 0.2132 - val\_accuracy: 0.9872 - val\_loss: 0.0484  
Epoch 49/100  
782/782 5s 6ms/step - accuracy: 0.9251 - loss: 0.2130 - val\_accuracy: 0.9898 - val\_loss: 0.0407  
Epoch 50/100  
782/782 5s 6ms/step - accuracy: 0.9281 - loss: 0.2043 - val\_accuracy: 0.9899 - val\_loss: 0.0394  
Epoch 51/100  
782/782 5s 6ms/step - accuracy: 0.9255 - loss: 0.2102 - val\_accuracy: 0.9904 - val\_loss: 0.0394  
Epoch 52/100  
782/782 5s 6ms/step - accuracy: 0.9270 - loss: 0.2099 - val\_accuracy: 0.9888 - val\_loss: 0.0421  
Epoch 53/100  
782/782 5s 6ms/step - accuracy: 0.9298 - loss: 0.2028 - val\_accuracy: 0.9917 - val\_loss: 0.0357  
Epoch 54/100  
782/782 5s 6ms/step - accuracy: 0.9308 - loss: 0.2007 - val\_accuracy: 0.9918 - val\_loss: 0.0355  
Epoch 55/100  
782/782 5s 6ms/step - accuracy: 0.9305 - loss: 0.1960 - val\_accuracy: 0.9896 - val\_loss: 0.0367  
Epoch 56/100  
782/782 5s 6ms/step - accuracy: 0.9333 - loss: 0.1885 - val\_accuracy: 0.9930 - val\_loss: 0.0322

Epoch 57/100  
**782/782** 5s 6ms/step - accuracy: 0.9345 - loss: 0.1869 - val\_accuracy: 0.9930 - val\_loss: 0.0321  
Epoch 58/100  
**782/782** 5s 6ms/step - accuracy: 0.9312 - loss: 0.1949 - val\_accuracy: 0.9939 - val\_loss: 0.0280  
Epoch 59/100  
**782/782** 5s 6ms/step - accuracy: 0.9339 - loss: 0.1843 - val\_accuracy: 0.9954 - val\_loss: 0.0249  
Epoch 60/100  
**782/782** 5s 6ms/step - accuracy: 0.9337 - loss: 0.1873 - val\_accuracy: 0.9946 - val\_loss: 0.0258  
Epoch 61/100  
**782/782** 5s 6ms/step - accuracy: 0.9357 - loss: 0.1818 - val\_accuracy: 0.9927 - val\_loss: 0.0301  
Epoch 62/100  
**782/782** 5s 6ms/step - accuracy: 0.9362 - loss: 0.1848 - val\_accuracy: 0.9954 - val\_loss: 0.0244  
Epoch 63/100  
**782/782** 5s 6ms/step - accuracy: 0.9362 - loss: 0.1802 - val\_accuracy: 0.9927 - val\_loss: 0.0267  
Epoch 64/100  
**782/782** 5s 6ms/step - accuracy: 0.9369 - loss: 0.1810 - val\_accuracy: 0.9955 - val\_loss: 0.0235  
Epoch 65/100  
**782/782** 5s 6ms/step - accuracy: 0.9410 - loss: 0.1674 - val\_accuracy: 0.9955 - val\_loss: 0.0220  
Epoch 66/100  
**782/782** 5s 6ms/step - accuracy: 0.9394 - loss: 0.1736 - val\_accuracy: 0.9952 - val\_loss: 0.0228  
Epoch 67/100  
**782/782** 5s 6ms/step - accuracy: 0.9406 - loss: 0.1694 - val\_accuracy: 0.9949 - val\_loss: 0.0219  
Epoch 68/100  
**782/782** 5s 6ms/step - accuracy: 0.9401 - loss: 0.1714 - val\_accuracy: 0.9968 - val\_loss: 0.0187  
Epoch 69/100  
**782/782** 5s 6ms/step - accuracy: 0.9405 - loss: 0.1704 - val\_accuracy: 0.9956 - val\_loss: 0.0187  
Epoch 70/100  
**782/782** 5s 6ms/step - accuracy: 0.9401 - loss: 0.1711 - val\_accuracy: 0.9965 - val\_loss: 0.0182  
Epoch 71/100  
**782/782** 5s 6ms/step - accuracy: 0.9417 - loss: 0.1626 - val\_accuracy: 0.9969 - val\_loss: 0.0178  
Epoch 72/100  
**782/782** 5s 6ms/step - accuracy: 0.9436 - loss: 0.1597 - val\_accuracy: 0.9974 - val\_loss: 0.0183  
Epoch 73/100  
**782/782** 5s 6ms/step - accuracy: 0.9417 - loss: 0.1700 - val\_accuracy: 0.9956 - val\_loss: 0.0222  
Epoch 74/100  
**782/782** 5s 6ms/step - accuracy: 0.9434 - loss: 0.1609 - val\_accuracy: 0.9968 - val\_loss: 0.0169  
Epoch 75/100  
**782/782** 5s 6ms/step - accuracy: 0.9408 - loss: 0.1736 - val\_accuracy:

```
uracy: 0.9949 - val_loss: 0.0222
Epoch 76/100
782/782 5s 6ms/step - accuracy: 0.9430 - loss: 0.1617 - val_acc
uracy: 0.9970 - val_loss: 0.0175
Epoch 77/100
782/782 5s 6ms/step - accuracy: 0.9449 - loss: 0.1590 - val_acc
uracy: 0.9974 - val_loss: 0.0147
Epoch 78/100
782/782 5s 6ms/step - accuracy: 0.9440 - loss: 0.1607 - val_acc
uracy: 0.9985 - val_loss: 0.0129
Epoch 79/100
782/782 5s 6ms/step - accuracy: 0.9481 - loss: 0.1504 - val_acc
uracy: 0.9979 - val_loss: 0.0135
Epoch 80/100
782/782 5s 6ms/step - accuracy: 0.9458 - loss: 0.1553 - val_acc
uracy: 0.9982 - val_loss: 0.0135
Epoch 81/100
782/782 5s 6ms/step - accuracy: 0.9439 - loss: 0.1559 - val_acc
uracy: 0.9973 - val_loss: 0.0142
Epoch 82/100
782/782 5s 6ms/step - accuracy: 0.9458 - loss: 0.1547 - val_acc
uracy: 0.9977 - val_loss: 0.0129
Epoch 83/100
782/782 5s 6ms/step - accuracy: 0.9454 - loss: 0.1590 - val_acc
uracy: 0.9978 - val_loss: 0.0135
Epoch 84/100
782/782 5s 6ms/step - accuracy: 0.9464 - loss: 0.1521 - val_acc
uracy: 0.9985 - val_loss: 0.0130
Epoch 85/100
782/782 5s 6ms/step - accuracy: 0.9468 - loss: 0.1516 - val_acc
uracy: 0.9991 - val_loss: 0.0108
Epoch 86/100
782/782 5s 6ms/step - accuracy: 0.9494 - loss: 0.1474 - val_acc
uracy: 0.9982 - val_loss: 0.0109
Epoch 87/100
782/782 5s 6ms/step - accuracy: 0.9469 - loss: 0.1498 - val_acc
uracy: 0.9972 - val_loss: 0.0142
Epoch 88/100
782/782 5s 6ms/step - accuracy: 0.9476 - loss: 0.1516 - val_acc
uracy: 0.9983 - val_loss: 0.0119
Epoch 89/100
782/782 5s 6ms/step - accuracy: 0.9488 - loss: 0.1471 - val_acc
uracy: 0.9981 - val_loss: 0.0111
Epoch 90/100
782/782 5s 6ms/step - accuracy: 0.9490 - loss: 0.1423 - val_acc
uracy: 0.9984 - val_loss: 0.0098
Epoch 91/100
782/782 5s 6ms/step - accuracy: 0.9503 - loss: 0.1428 - val_acc
uracy: 0.9984 - val_loss: 0.0105
Epoch 92/100
782/782 5s 6ms/step - accuracy: 0.9491 - loss: 0.1432 - val_acc
uracy: 0.9988 - val_loss: 0.0095
Epoch 93/100
782/782 5s 6ms/step - accuracy: 0.9516 - loss: 0.1399 - val_acc
uracy: 0.9984 - val_loss: 0.0084
Epoch 94/100
```

```
782/782 5s 6ms/step - accuracy: 0.9484 - loss: 0.1469 - val_accuracy: 0.9988 - val_loss: 0.0111
Epoch 95/100
782/782 5s 6ms/step - accuracy: 0.9509 - loss: 0.1420 - val_accuracy: 0.9992 - val_loss: 0.0094
Epoch 96/100
782/782 5s 6ms/step - accuracy: 0.9514 - loss: 0.1387 - val_accuracy: 0.9988 - val_loss: 0.0091
Epoch 97/100
782/782 5s 6ms/step - accuracy: 0.9528 - loss: 0.1353 - val_accuracy: 0.9986 - val_loss: 0.0094
Epoch 98/100
782/782 5s 6ms/step - accuracy: 0.9527 - loss: 0.1385 - val_accuracy: 0.9990 - val_loss: 0.0088
Epoch 99/100
782/782 5s 6ms/step - accuracy: 0.9535 - loss: 0.1373 - val_accuracy: 0.9990 - val_loss: 0.0080
Epoch 100/100
782/782 5s 6ms/step - accuracy: 0.9518 - loss: 0.1373 - val_accuracy: 0.9992 - val_loss: 0.0080
313/313 1s 2ms/step - accuracy: 0.9991 - loss: 0.0079
Fold Validation Accuracy: 99.92%
Epoch 1/100
782/782 23s 16ms/step - accuracy: 0.3044 - loss: 2.2842 - val_accuracy: 0.5274 - val_loss: 1.3112
Epoch 2/100
782/782 5s 6ms/step - accuracy: 0.5234 - loss: 1.3328 - val_accuracy: 0.6202 - val_loss: 1.0486
Epoch 3/100
782/782 5s 6ms/step - accuracy: 0.6028 - loss: 1.1213 - val_accuracy: 0.6755 - val_loss: 0.9281
Epoch 4/100
782/782 5s 6ms/step - accuracy: 0.6555 - loss: 0.9825 - val_accuracy: 0.7230 - val_loss: 0.7665
Epoch 5/100
782/782 5s 6ms/step - accuracy: 0.6909 - loss: 0.8841 - val_accuracy: 0.7591 - val_loss: 0.6697
Epoch 6/100
782/782 5s 6ms/step - accuracy: 0.7154 - loss: 0.8081 - val_accuracy: 0.7806 - val_loss: 0.6224
Epoch 7/100
782/782 5s 6ms/step - accuracy: 0.7400 - loss: 0.7470 - val_accuracy: 0.8068 - val_loss: 0.5443
Epoch 8/100
782/782 5s 6ms/step - accuracy: 0.7584 - loss: 0.6917 - val_accuracy: 0.8224 - val_loss: 0.4978
Epoch 9/100
782/782 5s 6ms/step - accuracy: 0.7709 - loss: 0.6599 - val_accuracy: 0.8389 - val_loss: 0.4566
Epoch 10/100
782/782 5s 6ms/step - accuracy: 0.7865 - loss: 0.6163 - val_accuracy: 0.8543 - val_loss: 0.4143
Epoch 11/100
782/782 5s 6ms/step - accuracy: 0.8043 - loss: 0.5715 - val_accuracy: 0.8599 - val_loss: 0.3975
Epoch 12/100
```

782/782 5s 6ms/step - accuracy: 0.8041 - loss: 0.5564 - val\_accuracy: 0.8719 - val\_loss: 0.3638  
Epoch 13/100  
782/782 5s 6ms/step - accuracy: 0.8155 - loss: 0.5291 - val\_accuracy: 0.8836 - val\_loss: 0.3337  
Epoch 14/100  
782/782 5s 6ms/step - accuracy: 0.8291 - loss: 0.4987 - val\_accuracy: 0.8807 - val\_loss: 0.3340  
Epoch 15/100  
782/782 5s 6ms/step - accuracy: 0.8296 - loss: 0.4889 - val\_accuracy: 0.8998 - val\_loss: 0.2965  
Epoch 16/100  
782/782 5s 6ms/step - accuracy: 0.8395 - loss: 0.4710 - val\_accuracy: 0.9058 - val\_loss: 0.2772  
Epoch 17/100  
782/782 5s 6ms/step - accuracy: 0.8496 - loss: 0.4387 - val\_accuracy: 0.9082 - val\_loss: 0.2612  
Epoch 18/100  
782/782 5s 6ms/step - accuracy: 0.8513 - loss: 0.4291 - val\_accuracy: 0.9143 - val\_loss: 0.2463  
Epoch 19/100  
782/782 5s 6ms/step - accuracy: 0.8582 - loss: 0.4131 - val\_accuracy: 0.9263 - val\_loss: 0.2187  
Epoch 20/100  
782/782 5s 6ms/step - accuracy: 0.8588 - loss: 0.4017 - val\_accuracy: 0.9324 - val\_loss: 0.2048  
Epoch 21/100  
782/782 5s 6ms/step - accuracy: 0.8666 - loss: 0.3824 - val\_accuracy: 0.9360 - val\_loss: 0.1949  
Epoch 22/100  
782/782 5s 6ms/step - accuracy: 0.8644 - loss: 0.3860 - val\_accuracy: 0.9345 - val\_loss: 0.1882  
Epoch 23/100  
782/782 5s 6ms/step - accuracy: 0.8731 - loss: 0.3642 - val\_accuracy: 0.9434 - val\_loss: 0.1690  
Epoch 24/100  
782/782 5s 6ms/step - accuracy: 0.8780 - loss: 0.3527 - val\_accuracy: 0.9442 - val\_loss: 0.1656  
Epoch 25/100  
782/782 5s 6ms/step - accuracy: 0.8821 - loss: 0.3413 - val\_accuracy: 0.9471 - val\_loss: 0.1512  
Epoch 26/100  
782/782 5s 6ms/step - accuracy: 0.8822 - loss: 0.3399 - val\_accuracy: 0.9523 - val\_loss: 0.1486  
Epoch 27/100  
782/782 5s 6ms/step - accuracy: 0.8870 - loss: 0.3283 - val\_accuracy: 0.9527 - val\_loss: 0.1411  
Epoch 28/100  
782/782 5s 6ms/step - accuracy: 0.8874 - loss: 0.3257 - val\_accuracy: 0.9607 - val\_loss: 0.1263  
Epoch 29/100  
782/782 5s 6ms/step - accuracy: 0.8963 - loss: 0.3013 - val\_accuracy: 0.9631 - val\_loss: 0.1167  
Epoch 30/100  
782/782 5s 6ms/step - accuracy: 0.8943 - loss: 0.3028 - val\_accuracy: 0.9661 - val\_loss: 0.1097

Epoch 31/100  
**782/782** 5s 6ms/step - accuracy: 0.8946 - loss: 0.3039 - val\_accuracy: 0.9663 - val\_loss: 0.1104  
Epoch 32/100  
**782/782** 5s 6ms/step - accuracy: 0.8987 - loss: 0.2906 - val\_accuracy: 0.9696 - val\_loss: 0.0990  
Epoch 33/100  
**782/782** 5s 6ms/step - accuracy: 0.8990 - loss: 0.2872 - val\_accuracy: 0.9674 - val\_loss: 0.1025  
Epoch 34/100  
**782/782** 5s 6ms/step - accuracy: 0.9054 - loss: 0.2712 - val\_accuracy: 0.9705 - val\_loss: 0.0963  
Epoch 35/100  
**782/782** 5s 6ms/step - accuracy: 0.9084 - loss: 0.2680 - val\_accuracy: 0.9721 - val\_loss: 0.0883  
Epoch 36/100  
**782/782** 5s 6ms/step - accuracy: 0.9087 - loss: 0.2591 - val\_accuracy: 0.9782 - val\_loss: 0.0769  
Epoch 37/100  
**782/782** 5s 6ms/step - accuracy: 0.9092 - loss: 0.2606 - val\_accuracy: 0.9714 - val\_loss: 0.0879  
Epoch 38/100  
**782/782** 5s 6ms/step - accuracy: 0.9088 - loss: 0.2626 - val\_accuracy: 0.9782 - val\_loss: 0.0719  
Epoch 39/100  
**782/782** 5s 6ms/step - accuracy: 0.9101 - loss: 0.2549 - val\_accuracy: 0.9820 - val\_loss: 0.0676  
Epoch 40/100  
**782/782** 5s 6ms/step - accuracy: 0.9150 - loss: 0.2466 - val\_accuracy: 0.9814 - val\_loss: 0.0649  
Epoch 41/100  
**782/782** 5s 6ms/step - accuracy: 0.9147 - loss: 0.2450 - val\_accuracy: 0.9797 - val\_loss: 0.0712  
Epoch 42/100  
**782/782** 5s 6ms/step - accuracy: 0.9145 - loss: 0.2484 - val\_accuracy: 0.9836 - val\_loss: 0.0594  
Epoch 43/100  
**782/782** 5s 6ms/step - accuracy: 0.9201 - loss: 0.2331 - val\_accuracy: 0.9838 - val\_loss: 0.0588  
Epoch 44/100  
**782/782** 5s 6ms/step - accuracy: 0.9131 - loss: 0.2441 - val\_accuracy: 0.9837 - val\_loss: 0.0584  
Epoch 45/100  
**782/782** 5s 6ms/step - accuracy: 0.9198 - loss: 0.2274 - val\_accuracy: 0.9836 - val\_loss: 0.0576  
Epoch 46/100  
**782/782** 5s 6ms/step - accuracy: 0.9227 - loss: 0.2215 - val\_accuracy: 0.9873 - val\_loss: 0.0510  
Epoch 47/100  
**782/782** 5s 6ms/step - accuracy: 0.9214 - loss: 0.2294 - val\_accuracy: 0.9856 - val\_loss: 0.0499  
Epoch 48/100  
**782/782** 5s 6ms/step - accuracy: 0.9235 - loss: 0.2218 - val\_accuracy: 0.9864 - val\_loss: 0.0491  
Epoch 49/100  
**782/782** 5s 6ms/step - accuracy: 0.9230 - loss: 0.2210 - val\_accuracy: 0.9864 - val\_loss: 0.0491

```
uracy: 0.9873 - val_loss: 0.0471
Epoch 50/100
782/782 5s 6ms/step - accuracy: 0.9240 - loss: 0.2168 - val_acc
uracy: 0.9894 - val_loss: 0.0415
Epoch 51/100
782/782 5s 6ms/step - accuracy: 0.9265 - loss: 0.2112 - val_acc
uracy: 0.9903 - val_loss: 0.0403
Epoch 52/100
782/782 5s 6ms/step - accuracy: 0.9280 - loss: 0.2079 - val_acc
uracy: 0.9898 - val_loss: 0.0391
Epoch 53/100
782/782 5s 6ms/step - accuracy: 0.9263 - loss: 0.2081 - val_acc
uracy: 0.9921 - val_loss: 0.0340
Epoch 54/100
782/782 5s 6ms/step - accuracy: 0.9276 - loss: 0.2118 - val_acc
uracy: 0.9918 - val_loss: 0.0376
Epoch 55/100
782/782 5s 6ms/step - accuracy: 0.9300 - loss: 0.2019 - val_acc
uracy: 0.9905 - val_loss: 0.0394
Epoch 56/100
782/782 5s 6ms/step - accuracy: 0.9288 - loss: 0.1997 - val_acc
uracy: 0.9903 - val_loss: 0.0342
Epoch 57/100
782/782 5s 6ms/step - accuracy: 0.9323 - loss: 0.1946 - val_acc
uracy: 0.9919 - val_loss: 0.0342
Epoch 58/100
782/782 5s 6ms/step - accuracy: 0.9353 - loss: 0.1866 - val_acc
uracy: 0.9940 - val_loss: 0.0292
Epoch 59/100
782/782 5s 6ms/step - accuracy: 0.9337 - loss: 0.1937 - val_acc
uracy: 0.9931 - val_loss: 0.0304
Epoch 60/100
782/782 5s 6ms/step - accuracy: 0.9322 - loss: 0.1940 - val_acc
uracy: 0.9945 - val_loss: 0.0268
Epoch 61/100
782/782 5s 6ms/step - accuracy: 0.9344 - loss: 0.1886 - val_acc
uracy: 0.9952 - val_loss: 0.0248
Epoch 62/100
782/782 5s 6ms/step - accuracy: 0.9348 - loss: 0.1828 - val_acc
uracy: 0.9941 - val_loss: 0.0266
Epoch 63/100
782/782 5s 6ms/step - accuracy: 0.9331 - loss: 0.1908 - val_acc
uracy: 0.9932 - val_loss: 0.0291
Epoch 64/100
782/782 5s 6ms/step - accuracy: 0.9368 - loss: 0.1828 - val_acc
uracy: 0.9950 - val_loss: 0.0243
Epoch 65/100
782/782 5s 6ms/step - accuracy: 0.9350 - loss: 0.1861 - val_acc
uracy: 0.9948 - val_loss: 0.0257
Epoch 66/100
782/782 5s 6ms/step - accuracy: 0.9373 - loss: 0.1806 - val_acc
uracy: 0.9950 - val_loss: 0.0239
Epoch 67/100
782/782 5s 6ms/step - accuracy: 0.9392 - loss: 0.1715 - val_acc
uracy: 0.9964 - val_loss: 0.0210
Epoch 68/100
```

782/782 5s 6ms/step - accuracy: 0.9349 - loss: 0.1842 - val\_accuracy: 0.9966 - val\_loss: 0.0185  
Epoch 69/100  
782/782 5s 6ms/step - accuracy: 0.9372 - loss: 0.1773 - val\_accuracy: 0.9969 - val\_loss: 0.0192  
Epoch 70/100  
782/782 5s 6ms/step - accuracy: 0.9414 - loss: 0.1732 - val\_accuracy: 0.9957 - val\_loss: 0.0219  
Epoch 71/100  
782/782 5s 6ms/step - accuracy: 0.9387 - loss: 0.1765 - val\_accuracy: 0.9962 - val\_loss: 0.0188  
Epoch 72/100  
782/782 5s 6ms/step - accuracy: 0.9407 - loss: 0.1698 - val\_accuracy: 0.9954 - val\_loss: 0.0205  
Epoch 73/100  
782/782 5s 6ms/step - accuracy: 0.9406 - loss: 0.1678 - val\_accuracy: 0.9970 - val\_loss: 0.0192  
Epoch 74/100  
782/782 5s 6ms/step - accuracy: 0.9432 - loss: 0.1662 - val\_accuracy: 0.9970 - val\_loss: 0.0163  
Epoch 75/100  
782/782 5s 6ms/step - accuracy: 0.9435 - loss: 0.1627 - val\_accuracy: 0.9974 - val\_loss: 0.0164  
Epoch 76/100  
782/782 5s 6ms/step - accuracy: 0.9421 - loss: 0.1633 - val\_accuracy: 0.9984 - val\_loss: 0.0132  
Epoch 77/100  
782/782 5s 6ms/step - accuracy: 0.9463 - loss: 0.1555 - val\_accuracy: 0.9983 - val\_loss: 0.0140  
Epoch 78/100  
782/782 5s 6ms/step - accuracy: 0.9440 - loss: 0.1611 - val\_accuracy: 0.9967 - val\_loss: 0.0168  
Epoch 79/100  
782/782 5s 6ms/step - accuracy: 0.9445 - loss: 0.1568 - val\_accuracy: 0.9971 - val\_loss: 0.0156  
Epoch 80/100  
782/782 5s 6ms/step - accuracy: 0.9455 - loss: 0.1551 - val\_accuracy: 0.9979 - val\_loss: 0.0128  
Epoch 81/100  
782/782 5s 6ms/step - accuracy: 0.9421 - loss: 0.1629 - val\_accuracy: 0.9976 - val\_loss: 0.0130  
Epoch 82/100  
782/782 5s 6ms/step - accuracy: 0.9442 - loss: 0.1594 - val\_accuracy: 0.9970 - val\_loss: 0.0155  
Epoch 83/100  
782/782 5s 6ms/step - accuracy: 0.9455 - loss: 0.1535 - val\_accuracy: 0.9975 - val\_loss: 0.0137  
Epoch 84/100  
782/782 5s 6ms/step - accuracy: 0.9462 - loss: 0.1538 - val\_accuracy: 0.9976 - val\_loss: 0.0137  
Epoch 85/100  
782/782 5s 6ms/step - accuracy: 0.9475 - loss: 0.1504 - val\_accuracy: 0.9960 - val\_loss: 0.0173  
Epoch 86/100  
782/782 5s 6ms/step - accuracy: 0.9449 - loss: 0.1580 - val\_accuracy: 0.9981 - val\_loss: 0.0129

Epoch 87/100  
**782/782** 5s 6ms/step - accuracy: 0.9473 - loss: 0.1474 - val\_accuracy: 0.9984 - val\_loss: 0.0136  
Epoch 88/100  
**782/782** 5s 6ms/step - accuracy: 0.9483 - loss: 0.1477 - val\_accuracy: 0.9976 - val\_loss: 0.0124  
Epoch 89/100  
**782/782** 5s 6ms/step - accuracy: 0.9473 - loss: 0.1536 - val\_accuracy: 0.9980 - val\_loss: 0.0119  
Epoch 90/100  
**782/782** 5s 6ms/step - accuracy: 0.9494 - loss: 0.1441 - val\_accuracy: 0.9984 - val\_loss: 0.0114  
Epoch 91/100  
**782/782** 5s 6ms/step - accuracy: 0.9487 - loss: 0.1458 - val\_accuracy: 0.9978 - val\_loss: 0.0123  
Epoch 92/100  
**782/782** 5s 6ms/step - accuracy: 0.9490 - loss: 0.1462 - val\_accuracy: 0.9985 - val\_loss: 0.0115  
Epoch 93/100  
**782/782** 5s 6ms/step - accuracy: 0.9478 - loss: 0.1454 - val\_accuracy: 0.9993 - val\_loss: 0.0079  
Epoch 94/100  
**782/782** 5s 6ms/step - accuracy: 0.9490 - loss: 0.1464 - val\_accuracy: 0.9989 - val\_loss: 0.0093  
Epoch 95/100  
**782/782** 5s 7ms/step - accuracy: 0.9499 - loss: 0.1424 - val\_accuracy: 0.9988 - val\_loss: 0.0093  
Epoch 96/100  
**782/782** 5s 6ms/step - accuracy: 0.9499 - loss: 0.1434 - val\_accuracy: 0.9992 - val\_loss: 0.0086  
Epoch 97/100  
**782/782** 5s 6ms/step - accuracy: 0.9499 - loss: 0.1454 - val\_accuracy: 0.9987 - val\_loss: 0.0098  
Epoch 98/100  
**782/782** 5s 6ms/step - accuracy: 0.9486 - loss: 0.1437 - val\_accuracy: 0.9996 - val\_loss: 0.0081  
Epoch 99/100  
**782/782** 5s 6ms/step - accuracy: 0.9526 - loss: 0.1392 - val\_accuracy: 0.9989 - val\_loss: 0.0094  
Epoch 100/100  
**782/782** 5s 6ms/step - accuracy: 0.9530 - loss: 0.1329 - val\_accuracy: 0.9992 - val\_loss: 0.0085  
**313/313** 1s 2ms/step - accuracy: 0.9991 - loss: 0.0085  
Fold Validation Accuracy: 99.92%  
Epoch 1/100  
**782/782** 23s 16ms/step - accuracy: 0.3058 - loss: 2.2357 - val\_accuracy: 0.5277 - val\_loss: 1.2992  
Epoch 2/100  
**782/782** 5s 6ms/step - accuracy: 0.5140 - loss: 1.3575 - val\_accuracy: 0.6162 - val\_loss: 1.0737  
Epoch 3/100  
**782/782** 5s 6ms/step - accuracy: 0.5898 - loss: 1.1431 - val\_accuracy: 0.6928 - val\_loss: 0.8667  
Epoch 4/100  
**782/782** 5s 6ms/step - accuracy: 0.6523 - loss: 0.9789 - val\_accuracy: 0.7267 - val\_loss: 0.7791

Epoch 5/100  
**782/782** 5s 6ms/step - accuracy: 0.6916 - loss: 0.8857 - val\_accuracy: 0.7664 - val\_loss: 0.6643  
Epoch 6/100  
**782/782** 5s 6ms/step - accuracy: 0.7216 - loss: 0.7998 - val\_accuracy: 0.7951 - val\_loss: 0.5938  
Epoch 7/100  
**782/782** 5s 6ms/step - accuracy: 0.7427 - loss: 0.7423 - val\_accuracy: 0.8181 - val\_loss: 0.5289  
Epoch 8/100  
**782/782** 5s 6ms/step - accuracy: 0.7610 - loss: 0.6874 - val\_accuracy: 0.8194 - val\_loss: 0.5155  
Epoch 9/100  
**782/782** 5s 6ms/step - accuracy: 0.7728 - loss: 0.6488 - val\_accuracy: 0.8399 - val\_loss: 0.4517  
Epoch 10/100  
**782/782** 5s 6ms/step - accuracy: 0.7920 - loss: 0.6001 - val\_accuracy: 0.8641 - val\_loss: 0.3989  
Epoch 11/100  
**782/782** 5s 6ms/step - accuracy: 0.8000 - loss: 0.5790 - val\_accuracy: 0.8670 - val\_loss: 0.3830  
Epoch 12/100  
**782/782** 5s 6ms/step - accuracy: 0.8114 - loss: 0.5432 - val\_accuracy: 0.8748 - val\_loss: 0.3613  
Epoch 13/100  
**782/782** 5s 6ms/step - accuracy: 0.8172 - loss: 0.5284 - val\_accuracy: 0.8969 - val\_loss: 0.3043  
Epoch 14/100  
**782/782** 5s 6ms/step - accuracy: 0.8317 - loss: 0.4898 - val\_accuracy: 0.8959 - val\_loss: 0.2956  
Epoch 15/100  
**782/782** 5s 6ms/step - accuracy: 0.8374 - loss: 0.4764 - val\_accuracy: 0.8957 - val\_loss: 0.2967  
Epoch 16/100  
**782/782** 5s 6ms/step - accuracy: 0.8424 - loss: 0.4538 - val\_accuracy: 0.9123 - val\_loss: 0.2561  
Epoch 17/100  
**782/782** 5s 6ms/step - accuracy: 0.8508 - loss: 0.4341 - val\_accuracy: 0.9175 - val\_loss: 0.2388  
Epoch 18/100  
**782/782** 5s 6ms/step - accuracy: 0.8532 - loss: 0.4203 - val\_accuracy: 0.9257 - val\_loss: 0.2213  
Epoch 19/100  
**782/782** 5s 6ms/step - accuracy: 0.8624 - loss: 0.4033 - val\_accuracy: 0.9310 - val\_loss: 0.2047  
Epoch 20/100  
**782/782** 5s 6ms/step - accuracy: 0.8633 - loss: 0.3947 - val\_accuracy: 0.9303 - val\_loss: 0.1958  
Epoch 21/100  
**782/782** 5s 6ms/step - accuracy: 0.8680 - loss: 0.3802 - val\_accuracy: 0.9388 - val\_loss: 0.1782  
Epoch 22/100  
**782/782** 5s 6ms/step - accuracy: 0.8732 - loss: 0.3692 - val\_accuracy: 0.9451 - val\_loss: 0.1676  
Epoch 23/100  
**782/782** 5s 6ms/step - accuracy: 0.8744 - loss: 0.3604 - val\_accuracy: 0.9451 - val\_loss: 0.1676

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uracy: 0.9473 - val_loss: 0.1572
Epoch 24/100
782/782 5s 6ms/step - accuracy: 0.8787 - loss: 0.3475 - val_acc
uracy: 0.9533 - val_loss: 0.1457
Epoch 25/100
782/782 5s 6ms/step - accuracy: 0.8853 - loss: 0.3344 - val_acc
uracy: 0.9580 - val_loss: 0.1373
Epoch 26/100
782/782 5s 6ms/step - accuracy: 0.8837 - loss: 0.3367 - val_acc
uracy: 0.9566 - val_loss: 0.1352
Epoch 27/100
782/782 5s 6ms/step - accuracy: 0.8898 - loss: 0.3175 - val_acc
uracy: 0.9562 - val_loss: 0.1352
Epoch 28/100
782/782 5s 6ms/step - accuracy: 0.8928 - loss: 0.3090 - val_acc
uracy: 0.9575 - val_loss: 0.1290
Epoch 29/100
782/782 5s 6ms/step - accuracy: 0.8904 - loss: 0.3074 - val_acc
uracy: 0.9660 - val_loss: 0.1079
Epoch 30/100
782/782 5s 6ms/step - accuracy: 0.8985 - loss: 0.2889 - val_acc
uracy: 0.9677 - val_loss: 0.1060
Epoch 31/100
782/782 5s 6ms/step - accuracy: 0.8975 - loss: 0.2926 - val_acc
uracy: 0.9673 - val_loss: 0.1049
Epoch 32/100
782/782 5s 6ms/step - accuracy: 0.8999 - loss: 0.2867 - val_acc
uracy: 0.9706 - val_loss: 0.0969
Epoch 33/100
782/782 5s 6ms/step - accuracy: 0.9030 - loss: 0.2760 - val_acc
uracy: 0.9715 - val_loss: 0.0938
Epoch 34/100
782/782 5s 6ms/step - accuracy: 0.9059 - loss: 0.2746 - val_acc
uracy: 0.9764 - val_loss: 0.0837
Epoch 35/100
782/782 5s 6ms/step - accuracy: 0.9051 - loss: 0.2716 - val_acc
uracy: 0.9750 - val_loss: 0.0861
Epoch 36/100
782/782 5s 6ms/step - accuracy: 0.9039 - loss: 0.2733 - val_acc
uracy: 0.9763 - val_loss: 0.0825
Epoch 37/100
782/782 5s 6ms/step - accuracy: 0.9091 - loss: 0.2618 - val_acc
uracy: 0.9795 - val_loss: 0.0715
Epoch 38/100
782/782 5s 6ms/step - accuracy: 0.9122 - loss: 0.2520 - val_acc
uracy: 0.9786 - val_loss: 0.0763
Epoch 39/100
782/782 5s 6ms/step - accuracy: 0.9140 - loss: 0.2454 - val_acc
uracy: 0.9839 - val_loss: 0.0582
Epoch 40/100
782/782 5s 6ms/step - accuracy: 0.9125 - loss: 0.2529 - val_acc
uracy: 0.9851 - val_loss: 0.0574
Epoch 41/100
782/782 5s 6ms/step - accuracy: 0.9148 - loss: 0.2401 - val_acc
uracy: 0.9817 - val_loss: 0.0649
Epoch 42/100
```

782/782 5s 6ms/step - accuracy: 0.9173 - loss: 0.2354 - val\_accuracy: 0.9848 - val\_loss: 0.0591  
Epoch 43/100  
782/782 5s 6ms/step - accuracy: 0.9184 - loss: 0.2347 - val\_accuracy: 0.9851 - val\_loss: 0.0544  
Epoch 44/100  
782/782 5s 6ms/step - accuracy: 0.9204 - loss: 0.2254 - val\_accuracy: 0.9840 - val\_loss: 0.0549  
Epoch 45/100  
782/782 5s 6ms/step - accuracy: 0.9183 - loss: 0.2298 - val\_accuracy: 0.9852 - val\_loss: 0.0532  
Epoch 46/100  
782/782 5s 6ms/step - accuracy: 0.9193 - loss: 0.2286 - val\_accuracy: 0.9871 - val\_loss: 0.0496  
Epoch 47/100  
782/782 5s 6ms/step - accuracy: 0.9246 - loss: 0.2155 - val\_accuracy: 0.9873 - val\_loss: 0.0493  
Epoch 48/100  
782/782 5s 6ms/step - accuracy: 0.9249 - loss: 0.2158 - val\_accuracy: 0.9894 - val\_loss: 0.0401  
Epoch 49/100  
782/782 5s 6ms/step - accuracy: 0.9247 - loss: 0.2152 - val\_accuracy: 0.9909 - val\_loss: 0.0406  
Epoch 50/100  
782/782 5s 6ms/step - accuracy: 0.9282 - loss: 0.2053 - val\_accuracy: 0.9903 - val\_loss: 0.0385  
Epoch 51/100  
782/782 5s 6ms/step - accuracy: 0.9295 - loss: 0.2009 - val\_accuracy: 0.9909 - val\_loss: 0.0389  
Epoch 52/100  
782/782 5s 6ms/step - accuracy: 0.9288 - loss: 0.2043 - val\_accuracy: 0.9892 - val\_loss: 0.0421  
Epoch 53/100  
782/782 5s 6ms/step - accuracy: 0.9309 - loss: 0.1953 - val\_accuracy: 0.9899 - val\_loss: 0.0358  
Epoch 54/100  
782/782 5s 6ms/step - accuracy: 0.9251 - loss: 0.2109 - val\_accuracy: 0.9915 - val\_loss: 0.0378  
Epoch 55/100  
782/782 5s 6ms/step - accuracy: 0.9287 - loss: 0.1999 - val\_accuracy: 0.9925 - val\_loss: 0.0314  
Epoch 56/100  
782/782 5s 6ms/step - accuracy: 0.9274 - loss: 0.2011 - val\_accuracy: 0.9923 - val\_loss: 0.0316  
Epoch 57/100  
782/782 5s 6ms/step - accuracy: 0.9316 - loss: 0.1933 - val\_accuracy: 0.9930 - val\_loss: 0.0304  
Epoch 58/100  
782/782 5s 6ms/step - accuracy: 0.9305 - loss: 0.1940 - val\_accuracy: 0.9937 - val\_loss: 0.0294  
Epoch 59/100  
782/782 5s 6ms/step - accuracy: 0.9356 - loss: 0.1837 - val\_accuracy: 0.9932 - val\_loss: 0.0294  
Epoch 60/100  
782/782 5s 6ms/step - accuracy: 0.9344 - loss: 0.1894 - val\_accuracy: 0.9949 - val\_loss: 0.0257

Epoch 61/100  
**782/782** 5s 6ms/step - accuracy: 0.9372 - loss: 0.1837 - val\_accuracy: 0.9944 - val\_loss: 0.0270  
Epoch 62/100  
**782/782** 5s 6ms/step - accuracy: 0.9346 - loss: 0.1806 - val\_accuracy: 0.9960 - val\_loss: 0.0236  
Epoch 63/100  
**782/782** 5s 6ms/step - accuracy: 0.9381 - loss: 0.1742 - val\_accuracy: 0.9942 - val\_loss: 0.0252  
Epoch 64/100  
**782/782** 5s 6ms/step - accuracy: 0.9365 - loss: 0.1847 - val\_accuracy: 0.9948 - val\_loss: 0.0242  
Epoch 65/100  
**782/782** 5s 6ms/step - accuracy: 0.9362 - loss: 0.1802 - val\_accuracy: 0.9949 - val\_loss: 0.0221  
Epoch 66/100  
**782/782** 5s 6ms/step - accuracy: 0.9383 - loss: 0.1760 - val\_accuracy: 0.9955 - val\_loss: 0.0223  
Epoch 67/100  
**782/782** 5s 7ms/step - accuracy: 0.9401 - loss: 0.1687 - val\_accuracy: 0.9954 - val\_loss: 0.0217  
Epoch 68/100  
**782/782** 5s 6ms/step - accuracy: 0.9402 - loss: 0.1706 - val\_accuracy: 0.9970 - val\_loss: 0.0199  
Epoch 69/100  
**782/782** 5s 6ms/step - accuracy: 0.9402 - loss: 0.1699 - val\_accuracy: 0.9966 - val\_loss: 0.0191  
Epoch 70/100  
**782/782** 5s 6ms/step - accuracy: 0.9374 - loss: 0.1751 - val\_accuracy: 0.9969 - val\_loss: 0.0189  
Epoch 71/100  
**782/782** 5s 6ms/step - accuracy: 0.9430 - loss: 0.1646 - val\_accuracy: 0.9960 - val\_loss: 0.0183  
Epoch 72/100  
**782/782** 5s 6ms/step - accuracy: 0.9415 - loss: 0.1682 - val\_accuracy: 0.9961 - val\_loss: 0.0201  
Epoch 73/100  
**782/782** 5s 6ms/step - accuracy: 0.9368 - loss: 0.1771 - val\_accuracy: 0.9961 - val\_loss: 0.0184  
Epoch 74/100  
**782/782** 5s 6ms/step - accuracy: 0.9377 - loss: 0.1738 - val\_accuracy: 0.9978 - val\_loss: 0.0158  
Epoch 75/100  
**782/782** 5s 6ms/step - accuracy: 0.9434 - loss: 0.1628 - val\_accuracy: 0.9985 - val\_loss: 0.0143  
Epoch 76/100  
**782/782** 5s 6ms/step - accuracy: 0.9465 - loss: 0.1553 - val\_accuracy: 0.9981 - val\_loss: 0.0135  
Epoch 77/100  
**782/782** 5s 6ms/step - accuracy: 0.9465 - loss: 0.1576 - val\_accuracy: 0.9971 - val\_loss: 0.0169  
Epoch 78/100  
**782/782** 5s 6ms/step - accuracy: 0.9446 - loss: 0.1606 - val\_accuracy: 0.9979 - val\_loss: 0.0137  
Epoch 79/100  
**782/782** 5s 6ms/step - accuracy: 0.9467 - loss: 0.1530 - val\_accuracy: 0.9981 - val\_loss: 0.0135

```
uracy: 0.9983 - val_loss: 0.0129
Epoch 80/100
782/782 5s 6ms/step - accuracy: 0.9458 - loss: 0.1511 - val_acc
uracy: 0.9985 - val_loss: 0.0134
Epoch 81/100
782/782 5s 6ms/step - accuracy: 0.9459 - loss: 0.1542 - val_acc
uracy: 0.9982 - val_loss: 0.0135
Epoch 82/100
782/782 5s 6ms/step - accuracy: 0.9466 - loss: 0.1524 - val_acc
uracy: 0.9987 - val_loss: 0.0111
Epoch 83/100
782/782 5s 6ms/step - accuracy: 0.9445 - loss: 0.1558 - val_acc
uracy: 0.9982 - val_loss: 0.0133
Epoch 84/100
782/782 5s 6ms/step - accuracy: 0.9465 - loss: 0.1554 - val_acc
uracy: 0.9983 - val_loss: 0.0121
Epoch 85/100
782/782 5s 6ms/step - accuracy: 0.9486 - loss: 0.1493 - val_acc
uracy: 0.9989 - val_loss: 0.0119
Epoch 86/100
782/782 5s 6ms/step - accuracy: 0.9480 - loss: 0.1528 - val_acc
uracy: 0.9983 - val_loss: 0.0129
Epoch 87/100
782/782 5s 6ms/step - accuracy: 0.9489 - loss: 0.1447 - val_acc
uracy: 0.9984 - val_loss: 0.0119
Epoch 88/100
782/782 5s 6ms/step - accuracy: 0.9476 - loss: 0.1505 - val_acc
uracy: 0.9988 - val_loss: 0.0102
Epoch 89/100
782/782 5s 6ms/step - accuracy: 0.9493 - loss: 0.1424 - val_acc
uracy: 0.9989 - val_loss: 0.0110
Epoch 90/100
782/782 5s 6ms/step - accuracy: 0.9500 - loss: 0.1404 - val_acc
uracy: 0.9989 - val_loss: 0.0111
Epoch 91/100
782/782 5s 6ms/step - accuracy: 0.9473 - loss: 0.1476 - val_acc
uracy: 0.9987 - val_loss: 0.0118
Epoch 92/100
782/782 5s 6ms/step - accuracy: 0.9495 - loss: 0.1421 - val_acc
uracy: 0.9992 - val_loss: 0.0092
Epoch 93/100
782/782 5s 6ms/step - accuracy: 0.9505 - loss: 0.1440 - val_acc
uracy: 0.9985 - val_loss: 0.0099
Epoch 94/100
782/782 5s 6ms/step - accuracy: 0.9512 - loss: 0.1388 - val_acc
uracy: 0.9992 - val_loss: 0.0085
Epoch 95/100
782/782 5s 6ms/step - accuracy: 0.9519 - loss: 0.1367 - val_acc
uracy: 0.9994 - val_loss: 0.0079
Epoch 96/100
782/782 5s 6ms/step - accuracy: 0.9511 - loss: 0.1399 - val_acc
uracy: 0.9990 - val_loss: 0.0103
Epoch 97/100
782/782 5s 6ms/step - accuracy: 0.9504 - loss: 0.1421 - val_acc
uracy: 0.9991 - val_loss: 0.0100
Epoch 98/100
```

782/782 5s 6ms/step - accuracy: 0.9509 - loss: 0.1397 - val\_accuracy: 0.9987 - val\_loss: 0.0095  
Epoch 99/100  
782/782 5s 6ms/step - accuracy: 0.9535 - loss: 0.1344 - val\_accuracy: 0.9981 - val\_loss: 0.0106  
Epoch 100/100  
782/782 5s 6ms/step - accuracy: 0.9497 - loss: 0.1412 - val\_accuracy: 0.9987 - val\_loss: 0.0086  
313/313 1s 2ms/step - accuracy: 0.9988 - loss: 0.0088  
Fold Validation Accuracy: 99.87%  
Epoch 1/100  
782/782 23s 16ms/step - accuracy: 0.2959 - loss: 2.3080 - val\_accuracy: 0.4862 - val\_loss: 1.4231  
Epoch 2/100  
782/782 5s 6ms/step - accuracy: 0.4913 - loss: 1.4207 - val\_accuracy: 0.5849 - val\_loss: 1.1573  
Epoch 3/100  
782/782 5s 6ms/step - accuracy: 0.5915 - loss: 1.1596 - val\_accuracy: 0.6700 - val\_loss: 0.9249  
Epoch 4/100  
782/782 5s 6ms/step - accuracy: 0.6501 - loss: 0.9985 - val\_accuracy: 0.7394 - val\_loss: 0.7448  
Epoch 5/100  
782/782 5s 6ms/step - accuracy: 0.6956 - loss: 0.8776 - val\_accuracy: 0.7599 - val\_loss: 0.6731  
Epoch 6/100  
782/782 5s 6ms/step - accuracy: 0.7145 - loss: 0.8104 - val\_accuracy: 0.7837 - val\_loss: 0.6110  
Epoch 7/100  
782/782 5s 6ms/step - accuracy: 0.7407 - loss: 0.7460 - val\_accuracy: 0.8004 - val\_loss: 0.5622  
Epoch 8/100  
782/782 5s 6ms/step - accuracy: 0.7585 - loss: 0.6961 - val\_accuracy: 0.8208 - val\_loss: 0.5007  
Epoch 9/100  
782/782 5s 6ms/step - accuracy: 0.7730 - loss: 0.6504 - val\_accuracy: 0.8430 - val\_loss: 0.4519  
Epoch 10/100  
782/782 5s 6ms/step - accuracy: 0.7894 - loss: 0.6126 - val\_accuracy: 0.8600 - val\_loss: 0.4055  
Epoch 11/100  
782/782 5s 6ms/step - accuracy: 0.8019 - loss: 0.5756 - val\_accuracy: 0.8682 - val\_loss: 0.3706  
Epoch 12/100  
782/782 5s 6ms/step - accuracy: 0.8098 - loss: 0.5506 - val\_accuracy: 0.8613 - val\_loss: 0.3863  
Epoch 13/100  
782/782 5s 6ms/step - accuracy: 0.8203 - loss: 0.5235 - val\_accuracy: 0.8967 - val\_loss: 0.3051  
Epoch 14/100  
782/782 5s 6ms/step - accuracy: 0.8265 - loss: 0.5037 - val\_accuracy: 0.9024 - val\_loss: 0.2828  
Epoch 15/100  
782/782 5s 6ms/step - accuracy: 0.8390 - loss: 0.4691 - val\_accuracy: 0.9100 - val\_loss: 0.2625  
Epoch 16/100

782/782 5s 6ms/step - accuracy: 0.8410 - loss: 0.4566 - val\_accuracy: 0.9155 - val\_loss: 0.2475  
Epoch 17/100  
782/782 5s 6ms/step - accuracy: 0.8500 - loss: 0.4388 - val\_accuracy: 0.9233 - val\_loss: 0.2285  
Epoch 18/100  
782/782 5s 6ms/step - accuracy: 0.8537 - loss: 0.4249 - val\_accuracy: 0.9289 - val\_loss: 0.2113  
Epoch 19/100  
782/782 5s 6ms/step - accuracy: 0.8597 - loss: 0.4000 - val\_accuracy: 0.9268 - val\_loss: 0.2176  
Epoch 20/100  
782/782 5s 6ms/step - accuracy: 0.8648 - loss: 0.3889 - val\_accuracy: 0.9405 - val\_loss: 0.1832  
Epoch 21/100  
782/782 5s 6ms/step - accuracy: 0.8654 - loss: 0.3847 - val\_accuracy: 0.9380 - val\_loss: 0.1802  
Epoch 22/100  
782/782 5s 6ms/step - accuracy: 0.8727 - loss: 0.3673 - val\_accuracy: 0.9422 - val\_loss: 0.1685  
Epoch 23/100  
782/782 5s 6ms/step - accuracy: 0.8771 - loss: 0.3519 - val\_accuracy: 0.9461 - val\_loss: 0.1650  
Epoch 24/100  
782/782 5s 6ms/step - accuracy: 0.8811 - loss: 0.3438 - val\_accuracy: 0.9549 - val\_loss: 0.1389  
Epoch 25/100  
782/782 5s 6ms/step - accuracy: 0.8812 - loss: 0.3366 - val\_accuracy: 0.9522 - val\_loss: 0.1446  
Epoch 26/100  
782/782 5s 6ms/step - accuracy: 0.8854 - loss: 0.3295 - val\_accuracy: 0.9581 - val\_loss: 0.1288  
Epoch 27/100  
782/782 5s 6ms/step - accuracy: 0.8894 - loss: 0.3169 - val\_accuracy: 0.9636 - val\_loss: 0.1144  
Epoch 28/100  
782/782 5s 6ms/step - accuracy: 0.8946 - loss: 0.3042 - val\_accuracy: 0.9640 - val\_loss: 0.1139  
Epoch 29/100  
782/782 5s 6ms/step - accuracy: 0.8929 - loss: 0.3029 - val\_accuracy: 0.9705 - val\_loss: 0.0987  
Epoch 30/100  
782/782 5s 6ms/step - accuracy: 0.8988 - loss: 0.2961 - val\_accuracy: 0.9694 - val\_loss: 0.0987  
Epoch 31/100  
782/782 5s 6ms/step - accuracy: 0.9005 - loss: 0.2872 - val\_accuracy: 0.9723 - val\_loss: 0.0898  
Epoch 32/100  
782/782 5s 6ms/step - accuracy: 0.9040 - loss: 0.2766 - val\_accuracy: 0.9698 - val\_loss: 0.0971  
Epoch 33/100  
782/782 5s 6ms/step - accuracy: 0.9018 - loss: 0.2792 - val\_accuracy: 0.9792 - val\_loss: 0.0758  
Epoch 34/100  
782/782 5s 6ms/step - accuracy: 0.9062 - loss: 0.2687 - val\_accuracy: 0.9769 - val\_loss: 0.0795

Epoch 35/100  
**782/782** 5s 6ms/step - accuracy: 0.9085 - loss: 0.2634 - val\_accuracy: 0.9805 - val\_loss: 0.0722  
Epoch 36/100  
**782/782** 5s 6ms/step - accuracy: 0.9102 - loss: 0.2588 - val\_accuracy: 0.9825 - val\_loss: 0.0684  
Epoch 37/100  
**782/782** 5s 6ms/step - accuracy: 0.9102 - loss: 0.2572 - val\_accuracy: 0.9791 - val\_loss: 0.0729  
Epoch 38/100  
**782/782** 5s 6ms/step - accuracy: 0.9134 - loss: 0.2500 - val\_accuracy: 0.9815 - val\_loss: 0.0659  
Epoch 39/100  
**782/782** 5s 6ms/step - accuracy: 0.9151 - loss: 0.2430 - val\_accuracy: 0.9827 - val\_loss: 0.0619  
Epoch 40/100  
**782/782** 5s 6ms/step - accuracy: 0.9140 - loss: 0.2470 - val\_accuracy: 0.9831 - val\_loss: 0.0601  
Epoch 41/100  
**782/782** 5s 6ms/step - accuracy: 0.9164 - loss: 0.2355 - val\_accuracy: 0.9859 - val\_loss: 0.0558  
Epoch 42/100  
**782/782** 5s 6ms/step - accuracy: 0.9174 - loss: 0.2397 - val\_accuracy: 0.9836 - val\_loss: 0.0593  
Epoch 43/100  
**782/782** 5s 6ms/step - accuracy: 0.9181 - loss: 0.2340 - val\_accuracy: 0.9857 - val\_loss: 0.0505  
Epoch 44/100  
**782/782** 5s 6ms/step - accuracy: 0.9196 - loss: 0.2314 - val\_accuracy: 0.9917 - val\_loss: 0.0425  
Epoch 45/100  
**782/782** 5s 6ms/step - accuracy: 0.9247 - loss: 0.2175 - val\_accuracy: 0.9873 - val\_loss: 0.0472  
Epoch 46/100  
**782/782** 5s 6ms/step - accuracy: 0.9246 - loss: 0.2176 - val\_accuracy: 0.9888 - val\_loss: 0.0444  
Epoch 47/100  
**782/782** 5s 6ms/step - accuracy: 0.9236 - loss: 0.2209 - val\_accuracy: 0.9893 - val\_loss: 0.0423  
Epoch 48/100  
**782/782** 5s 6ms/step - accuracy: 0.9262 - loss: 0.2127 - val\_accuracy: 0.9904 - val\_loss: 0.0411  
Epoch 49/100  
**782/782** 5s 6ms/step - accuracy: 0.9263 - loss: 0.2078 - val\_accuracy: 0.9916 - val\_loss: 0.0366  
Epoch 50/100  
**782/782** 5s 6ms/step - accuracy: 0.9320 - loss: 0.1994 - val\_accuracy: 0.9926 - val\_loss: 0.0361  
Epoch 51/100  
**782/782** 5s 6ms/step - accuracy: 0.9280 - loss: 0.2034 - val\_accuracy: 0.9921 - val\_loss: 0.0339  
Epoch 52/100  
**782/782** 5s 7ms/step - accuracy: 0.9289 - loss: 0.2068 - val\_accuracy: 0.9926 - val\_loss: 0.0339  
Epoch 53/100  
**782/782** 5s 6ms/step - accuracy: 0.9270 - loss: 0.2068 - val\_accuracy: 0.9926 - val\_loss: 0.0339

```
uracy: 0.9936 - val_loss: 0.0300
Epoch 54/100
782/782 5s 6ms/step - accuracy: 0.9311 - loss: 0.1975 - val_acc
uracy: 0.9925 - val_loss: 0.0326
Epoch 55/100
782/782 5s 6ms/step - accuracy: 0.9320 - loss: 0.1929 - val_acc
uracy: 0.9937 - val_loss: 0.0300
Epoch 56/100
782/782 5s 6ms/step - accuracy: 0.9330 - loss: 0.1907 - val_acc
uracy: 0.9947 - val_loss: 0.0251
Epoch 57/100
782/782 5s 6ms/step - accuracy: 0.9323 - loss: 0.1920 - val_acc
uracy: 0.9925 - val_loss: 0.0333
Epoch 58/100
782/782 5s 6ms/step - accuracy: 0.9334 - loss: 0.1899 - val_acc
uracy: 0.9950 - val_loss: 0.0251
Epoch 59/100
782/782 5s 6ms/step - accuracy: 0.9349 - loss: 0.1865 - val_acc
uracy: 0.9940 - val_loss: 0.0291
Epoch 60/100
782/782 5s 6ms/step - accuracy: 0.9338 - loss: 0.1889 - val_acc
uracy: 0.9939 - val_loss: 0.0258
Epoch 61/100
782/782 5s 6ms/step - accuracy: 0.9356 - loss: 0.1860 - val_acc
uracy: 0.9956 - val_loss: 0.0257
Epoch 62/100
782/782 5s 6ms/step - accuracy: 0.9348 - loss: 0.1842 - val_acc
uracy: 0.9963 - val_loss: 0.0215
Epoch 63/100
782/782 5s 6ms/step - accuracy: 0.9370 - loss: 0.1790 - val_acc
uracy: 0.9944 - val_loss: 0.0254
Epoch 64/100
782/782 5s 6ms/step - accuracy: 0.9388 - loss: 0.1798 - val_acc
uracy: 0.9934 - val_loss: 0.0269
Epoch 65/100
782/782 5s 6ms/step - accuracy: 0.9374 - loss: 0.1746 - val_acc
uracy: 0.9965 - val_loss: 0.0201
Epoch 66/100
782/782 5s 6ms/step - accuracy: 0.9387 - loss: 0.1766 - val_acc
uracy: 0.9954 - val_loss: 0.0214
Epoch 67/100
782/782 5s 6ms/step - accuracy: 0.9387 - loss: 0.1740 - val_acc
uracy: 0.9960 - val_loss: 0.0206
Epoch 68/100
782/782 5s 6ms/step - accuracy: 0.9410 - loss: 0.1653 - val_acc
uracy: 0.9964 - val_loss: 0.0196
Epoch 69/100
782/782 5s 6ms/step - accuracy: 0.9386 - loss: 0.1747 - val_acc
uracy: 0.9967 - val_loss: 0.0191
Epoch 70/100
782/782 5s 6ms/step - accuracy: 0.9386 - loss: 0.1729 - val_acc
uracy: 0.9972 - val_loss: 0.0171
Epoch 71/100
782/782 5s 6ms/step - accuracy: 0.9435 - loss: 0.1608 - val_acc
uracy: 0.9972 - val_loss: 0.0165
Epoch 72/100
```

782/782 5s 6ms/step - accuracy: 0.9412 - loss: 0.1666 - val\_accuracy: 0.9967 - val\_loss: 0.0176  
Epoch 73/100  
782/782 5s 6ms/step - accuracy: 0.9428 - loss: 0.1661 - val\_accuracy: 0.9966 - val\_loss: 0.0179  
Epoch 74/100  
782/782 5s 6ms/step - accuracy: 0.9432 - loss: 0.1643 - val\_accuracy: 0.9981 - val\_loss: 0.0155  
Epoch 75/100  
782/782 5s 6ms/step - accuracy: 0.9441 - loss: 0.1607 - val\_accuracy: 0.9978 - val\_loss: 0.0155  
Epoch 76/100  
782/782 5s 6ms/step - accuracy: 0.9436 - loss: 0.1603 - val\_accuracy: 0.9968 - val\_loss: 0.0168  
Epoch 77/100  
782/782 5s 6ms/step - accuracy: 0.9471 - loss: 0.1532 - val\_accuracy: 0.9963 - val\_loss: 0.0158  
Epoch 78/100  
782/782 5s 6ms/step - accuracy: 0.9425 - loss: 0.1619 - val\_accuracy: 0.9967 - val\_loss: 0.0159  
Epoch 79/100  
782/782 5s 6ms/step - accuracy: 0.9450 - loss: 0.1614 - val\_accuracy: 0.9973 - val\_loss: 0.0156  
Epoch 80/100  
782/782 5s 6ms/step - accuracy: 0.9465 - loss: 0.1513 - val\_accuracy: 0.9974 - val\_loss: 0.0150  
Epoch 81/100  
782/782 5s 6ms/step - accuracy: 0.9451 - loss: 0.1537 - val\_accuracy: 0.9984 - val\_loss: 0.0126  
Epoch 82/100  
782/782 5s 6ms/step - accuracy: 0.9480 - loss: 0.1498 - val\_accuracy: 0.9986 - val\_loss: 0.0112  
Epoch 83/100  
782/782 5s 6ms/step - accuracy: 0.9482 - loss: 0.1510 - val\_accuracy: 0.9982 - val\_loss: 0.0115  
Epoch 84/100  
782/782 5s 6ms/step - accuracy: 0.9493 - loss: 0.1459 - val\_accuracy: 0.9985 - val\_loss: 0.0109  
Epoch 85/100  
782/782 5s 6ms/step - accuracy: 0.9488 - loss: 0.1463 - val\_accuracy: 0.9981 - val\_loss: 0.0122  
Epoch 86/100  
782/782 5s 6ms/step - accuracy: 0.9506 - loss: 0.1460 - val\_accuracy: 0.9982 - val\_loss: 0.0115  
Epoch 87/100  
782/782 5s 6ms/step - accuracy: 0.9496 - loss: 0.1446 - val\_accuracy: 0.9982 - val\_loss: 0.0117  
Epoch 88/100  
782/782 5s 6ms/step - accuracy: 0.9471 - loss: 0.1522 - val\_accuracy: 0.9972 - val\_loss: 0.0122  
Epoch 89/100  
782/782 5s 6ms/step - accuracy: 0.9508 - loss: 0.1391 - val\_accuracy: 0.9992 - val\_loss: 0.0100  
Epoch 90/100  
782/782 5s 6ms/step - accuracy: 0.9507 - loss: 0.1394 - val\_accuracy: 0.9981 - val\_loss: 0.0114

Epoch 91/100  
**782/782** 5s 6ms/step - accuracy: 0.9501 - loss: 0.1440 - val\_accuracy: 0.9988 - val\_loss: 0.0113  
Epoch 92/100  
**782/782** 5s 6ms/step - accuracy: 0.9516 - loss: 0.1379 - val\_accuracy: 0.9992 - val\_loss: 0.0088  
Epoch 93/100  
**782/782** 5s 6ms/step - accuracy: 0.9505 - loss: 0.1453 - val\_accuracy: 0.9992 - val\_loss: 0.0078  
Epoch 94/100  
**782/782** 5s 6ms/step - accuracy: 0.9505 - loss: 0.1426 - val\_accuracy: 0.9989 - val\_loss: 0.0087  
Epoch 95/100  
**782/782** 5s 6ms/step - accuracy: 0.9510 - loss: 0.1423 - val\_accuracy: 0.9990 - val\_loss: 0.0092  
Epoch 96/100  
**782/782** 5s 6ms/step - accuracy: 0.9527 - loss: 0.1382 - val\_accuracy: 0.9996 - val\_loss: 0.0080  
Epoch 97/100  
**782/782** 5s 6ms/step - accuracy: 0.9508 - loss: 0.1403 - val\_accuracy: 0.9991 - val\_loss: 0.0086  
Epoch 98/100  
**782/782** 5s 6ms/step - accuracy: 0.9526 - loss: 0.1386 - val\_accuracy: 0.9990 - val\_loss: 0.0083  
Epoch 99/100  
**782/782** 5s 6ms/step - accuracy: 0.9534 - loss: 0.1366 - val\_accuracy: 0.9985 - val\_loss: 0.0093  
Epoch 100/100  
**782/782** 5s 6ms/step - accuracy: 0.9524 - loss: 0.1354 - val\_accuracy: 0.9986 - val\_loss: 0.0080  
**313/313** 1s 2ms/step - accuracy: 0.9991 - loss: 0.0068  
Fold Validation Accuracy: 99.86%  
Epoch 1/100  
**782/782** 23s 16ms/step - accuracy: 0.3032 - loss: 2.2779 - val\_accuracy: 0.5331 - val\_loss: 1.2905  
Epoch 2/100  
**782/782** 5s 6ms/step - accuracy: 0.5137 - loss: 1.3524 - val\_accuracy: 0.6126 - val\_loss: 1.0740  
Epoch 3/100  
**782/782** 5s 6ms/step - accuracy: 0.6042 - loss: 1.1107 - val\_accuracy: 0.6710 - val\_loss: 0.9243  
Epoch 4/100  
**782/782** 5s 6ms/step - accuracy: 0.6553 - loss: 0.9730 - val\_accuracy: 0.7273 - val\_loss: 0.7532  
Epoch 5/100  
**782/782** 5s 6ms/step - accuracy: 0.6965 - loss: 0.8756 - val\_accuracy: 0.7607 - val\_loss: 0.6631  
Epoch 6/100  
**782/782** 5s 6ms/step - accuracy: 0.7246 - loss: 0.7905 - val\_accuracy: 0.7940 - val\_loss: 0.5816  
Epoch 7/100  
**782/782** 5s 6ms/step - accuracy: 0.7432 - loss: 0.7338 - val\_accuracy: 0.8128 - val\_loss: 0.5381  
Epoch 8/100  
**782/782** 5s 6ms/step - accuracy: 0.7626 - loss: 0.6870 - val\_accuracy: 0.8290 - val\_loss: 0.4941

Epoch 9/100  
**782/782** 5s 6ms/step - accuracy: 0.7830 - loss: 0.6282 - val\_accuracy: 0.8517 - val\_loss: 0.4233  
Epoch 10/100  
**782/782** 5s 6ms/step - accuracy: 0.7901 - loss: 0.6057 - val\_accuracy: 0.8475 - val\_loss: 0.4234  
Epoch 11/100  
**782/782** 5s 6ms/step - accuracy: 0.8044 - loss: 0.5702 - val\_accuracy: 0.8694 - val\_loss: 0.3691  
Epoch 12/100  
**782/782** 5s 6ms/step - accuracy: 0.8131 - loss: 0.5447 - val\_accuracy: 0.8810 - val\_loss: 0.3415  
Epoch 13/100  
**782/782** 5s 6ms/step - accuracy: 0.8245 - loss: 0.5073 - val\_accuracy: 0.8835 - val\_loss: 0.3300  
Epoch 14/100  
**782/782** 5s 6ms/step - accuracy: 0.8302 - loss: 0.4931 - val\_accuracy: 0.8975 - val\_loss: 0.2942  
Epoch 15/100  
**782/782** 5s 6ms/step - accuracy: 0.8342 - loss: 0.4727 - val\_accuracy: 0.9042 - val\_loss: 0.2672  
Epoch 16/100  
**782/782** 5s 6ms/step - accuracy: 0.8445 - loss: 0.4514 - val\_accuracy: 0.9115 - val\_loss: 0.2515  
Epoch 17/100  
**782/782** 5s 6ms/step - accuracy: 0.8489 - loss: 0.4336 - val\_accuracy: 0.9197 - val\_loss: 0.2283  
Epoch 18/100  
**782/782** 5s 6ms/step - accuracy: 0.8559 - loss: 0.4174 - val\_accuracy: 0.9211 - val\_loss: 0.2286  
Epoch 19/100  
**782/782** 5s 6ms/step - accuracy: 0.8608 - loss: 0.4009 - val\_accuracy: 0.9310 - val\_loss: 0.1979  
Epoch 20/100  
**782/782** 5s 6ms/step - accuracy: 0.8672 - loss: 0.3842 - val\_accuracy: 0.9371 - val\_loss: 0.1838  
Epoch 21/100  
**782/782** 5s 6ms/step - accuracy: 0.8682 - loss: 0.3805 - val\_accuracy: 0.9454 - val\_loss: 0.1620  
Epoch 22/100  
**782/782** 5s 6ms/step - accuracy: 0.8729 - loss: 0.3615 - val\_accuracy: 0.9412 - val\_loss: 0.1780  
Epoch 23/100  
**782/782** 5s 6ms/step - accuracy: 0.8791 - loss: 0.3512 - val\_accuracy: 0.9480 - val\_loss: 0.1525  
Epoch 24/100  
**782/782** 5s 6ms/step - accuracy: 0.8765 - loss: 0.3521 - val\_accuracy: 0.9539 - val\_loss: 0.1415  
Epoch 25/100  
**782/782** 5s 6ms/step - accuracy: 0.8815 - loss: 0.3317 - val\_accuracy: 0.9525 - val\_loss: 0.1409  
Epoch 26/100  
**782/782** 5s 6ms/step - accuracy: 0.8867 - loss: 0.3276 - val\_accuracy: 0.9581 - val\_loss: 0.1262  
Epoch 27/100  
**782/782** 5s 6ms/step - accuracy: 0.8923 - loss: 0.3127 - val\_accuracy: 0.9630 - val\_loss: 0.1124

```
uracy: 0.9572 - val_loss: 0.1284
Epoch 28/100
782/782 5s 6ms/step - accuracy: 0.8923 - loss: 0.3116 - val_acc
uracy: 0.9599 - val_loss: 0.1167
Epoch 29/100
782/782 5s 6ms/step - accuracy: 0.8969 - loss: 0.2975 - val_acc
uracy: 0.9646 - val_loss: 0.1103
Epoch 30/100
782/782 5s 6ms/step - accuracy: 0.8971 - loss: 0.2943 - val_acc
uracy: 0.9677 - val_loss: 0.1004
Epoch 31/100
782/782 5s 6ms/step - accuracy: 0.8959 - loss: 0.2956 - val_acc
uracy: 0.9679 - val_loss: 0.0993
Epoch 32/100
782/782 5s 6ms/step - accuracy: 0.8997 - loss: 0.2893 - val_acc
uracy: 0.9719 - val_loss: 0.0900
Epoch 33/100
782/782 5s 6ms/step - accuracy: 0.9025 - loss: 0.2763 - val_acc
uracy: 0.9741 - val_loss: 0.0822
Epoch 34/100
782/782 5s 6ms/step - accuracy: 0.9056 - loss: 0.2710 - val_acc
uracy: 0.9733 - val_loss: 0.0832
Epoch 35/100
782/782 5s 6ms/step - accuracy: 0.9079 - loss: 0.2654 - val_acc
uracy: 0.9769 - val_loss: 0.0782
Epoch 36/100
782/782 5s 6ms/step - accuracy: 0.9092 - loss: 0.2582 - val_acc
uracy: 0.9779 - val_loss: 0.0731
Epoch 37/100
782/782 5s 6ms/step - accuracy: 0.9131 - loss: 0.2536 - val_acc
uracy: 0.9787 - val_loss: 0.0724
Epoch 38/100
782/782 5s 6ms/step - accuracy: 0.9128 - loss: 0.2514 - val_acc
uracy: 0.9822 - val_loss: 0.0638
Epoch 39/100
782/782 5s 6ms/step - accuracy: 0.9104 - loss: 0.2547 - val_acc
uracy: 0.9752 - val_loss: 0.0777
Epoch 40/100
782/782 5s 6ms/step - accuracy: 0.9143 - loss: 0.2489 - val_acc
uracy: 0.9823 - val_loss: 0.0596
Epoch 41/100
782/782 5s 6ms/step - accuracy: 0.9180 - loss: 0.2341 - val_acc
uracy: 0.9833 - val_loss: 0.0580
Epoch 42/100
782/782 5s 6ms/step - accuracy: 0.9202 - loss: 0.2277 - val_acc
uracy: 0.9830 - val_loss: 0.0584
Epoch 43/100
782/782 5s 6ms/step - accuracy: 0.9183 - loss: 0.2313 - val_acc
uracy: 0.9871 - val_loss: 0.0504
Epoch 44/100
782/782 5s 6ms/step - accuracy: 0.9208 - loss: 0.2251 - val_acc
uracy: 0.9846 - val_loss: 0.0529
Epoch 45/100
782/782 5s 6ms/step - accuracy: 0.9207 - loss: 0.2283 - val_acc
uracy: 0.9857 - val_loss: 0.0492
Epoch 46/100
```

782/782 5s 6ms/step - accuracy: 0.9195 - loss: 0.2252 - val\_accuracy: 0.9881 - val\_loss: 0.0443  
Epoch 47/100  
782/782 5s 6ms/step - accuracy: 0.9206 - loss: 0.2243 - val\_accuracy: 0.9899 - val\_loss: 0.0409  
Epoch 48/100  
782/782 5s 6ms/step - accuracy: 0.9242 - loss: 0.2184 - val\_accuracy: 0.9873 - val\_loss: 0.0475  
Epoch 49/100  
782/782 5s 6ms/step - accuracy: 0.9242 - loss: 0.2159 - val\_accuracy: 0.9895 - val\_loss: 0.0420  
Epoch 50/100  
782/782 5s 6ms/step - accuracy: 0.9283 - loss: 0.2068 - val\_accuracy: 0.9884 - val\_loss: 0.0390  
Epoch 51/100  
782/782 5s 6ms/step - accuracy: 0.9281 - loss: 0.2040 - val\_accuracy: 0.9915 - val\_loss: 0.0361  
Epoch 52/100  
782/782 5s 6ms/step - accuracy: 0.9273 - loss: 0.2038 - val\_accuracy: 0.9899 - val\_loss: 0.0374  
Epoch 53/100  
782/782 5s 6ms/step - accuracy: 0.9283 - loss: 0.2056 - val\_accuracy: 0.9919 - val\_loss: 0.0336  
Epoch 54/100  
782/782 5s 6ms/step - accuracy: 0.9304 - loss: 0.1949 - val\_accuracy: 0.9899 - val\_loss: 0.0369  
Epoch 55/100  
782/782 5s 6ms/step - accuracy: 0.9319 - loss: 0.1989 - val\_accuracy: 0.9923 - val\_loss: 0.0309  
Epoch 56/100  
782/782 5s 6ms/step - accuracy: 0.9294 - loss: 0.2009 - val\_accuracy: 0.9932 - val\_loss: 0.0299  
Epoch 57/100  
782/782 5s 6ms/step - accuracy: 0.9337 - loss: 0.1889 - val\_accuracy: 0.9932 - val\_loss: 0.0284  
Epoch 58/100  
782/782 5s 6ms/step - accuracy: 0.9352 - loss: 0.1902 - val\_accuracy: 0.9920 - val\_loss: 0.0301  
Epoch 59/100  
782/782 5s 6ms/step - accuracy: 0.9366 - loss: 0.1835 - val\_accuracy: 0.9931 - val\_loss: 0.0296  
Epoch 60/100  
782/782 5s 6ms/step - accuracy: 0.9362 - loss: 0.1867 - val\_accuracy: 0.9936 - val\_loss: 0.0292  
Epoch 61/100  
782/782 5s 6ms/step - accuracy: 0.9347 - loss: 0.1847 - val\_accuracy: 0.9943 - val\_loss: 0.0255  
Epoch 62/100  
782/782 5s 6ms/step - accuracy: 0.9358 - loss: 0.1796 - val\_accuracy: 0.9950 - val\_loss: 0.0261  
Epoch 63/100  
782/782 5s 6ms/step - accuracy: 0.9356 - loss: 0.1824 - val\_accuracy: 0.9955 - val\_loss: 0.0215  
Epoch 64/100  
782/782 5s 6ms/step - accuracy: 0.9372 - loss: 0.1807 - val\_accuracy: 0.9959 - val\_loss: 0.0218

Epoch 65/100  
**782/782** 5s 6ms/step - accuracy: 0.9378 - loss: 0.1759 - val\_accuracy: 0.9962 - val\_loss: 0.0207  
Epoch 66/100  
**782/782** 5s 6ms/step - accuracy: 0.9385 - loss: 0.1768 - val\_accuracy: 0.9969 - val\_loss: 0.0181  
Epoch 67/100  
**782/782** 5s 6ms/step - accuracy: 0.9397 - loss: 0.1761 - val\_accuracy: 0.9963 - val\_loss: 0.0198  
Epoch 68/100  
**782/782** 5s 6ms/step - accuracy: 0.9402 - loss: 0.1721 - val\_accuracy: 0.9968 - val\_loss: 0.0185  
Epoch 69/100  
**782/782** 5s 6ms/step - accuracy: 0.9417 - loss: 0.1683 - val\_accuracy: 0.9968 - val\_loss: 0.0168  
Epoch 70/100  
**782/782** 5s 6ms/step - accuracy: 0.9378 - loss: 0.1757 - val\_accuracy: 0.9959 - val\_loss: 0.0200  
Epoch 71/100  
**782/782** 5s 6ms/step - accuracy: 0.9422 - loss: 0.1642 - val\_accuracy: 0.9959 - val\_loss: 0.0195  
Epoch 72/100  
**782/782** 5s 6ms/step - accuracy: 0.9436 - loss: 0.1659 - val\_accuracy: 0.9973 - val\_loss: 0.0157  
Epoch 73/100  
**782/782** 5s 6ms/step - accuracy: 0.9426 - loss: 0.1667 - val\_accuracy: 0.9970 - val\_loss: 0.0157  
Epoch 74/100  
**782/782** 5s 6ms/step - accuracy: 0.9424 - loss: 0.1602 - val\_accuracy: 0.9967 - val\_loss: 0.0163  
Epoch 75/100  
**782/782** 5s 6ms/step - accuracy: 0.9426 - loss: 0.1641 - val\_accuracy: 0.9969 - val\_loss: 0.0170  
Epoch 76/100  
**782/782** 5s 6ms/step - accuracy: 0.9418 - loss: 0.1646 - val\_accuracy: 0.9967 - val\_loss: 0.0159  
Epoch 77/100  
**782/782** 5s 6ms/step - accuracy: 0.9421 - loss: 0.1634 - val\_accuracy: 0.9961 - val\_loss: 0.0173  
Epoch 78/100  
**782/782** 5s 6ms/step - accuracy: 0.9442 - loss: 0.1648 - val\_accuracy: 0.9976 - val\_loss: 0.0139  
Epoch 79/100  
**782/782** 5s 6ms/step - accuracy: 0.9447 - loss: 0.1560 - val\_accuracy: 0.9968 - val\_loss: 0.0145  
Epoch 80/100  
**782/782** 5s 6ms/step - accuracy: 0.9452 - loss: 0.1541 - val\_accuracy: 0.9968 - val\_loss: 0.0159  
Epoch 81/100  
**782/782** 5s 6ms/step - accuracy: 0.9459 - loss: 0.1569 - val\_accuracy: 0.9977 - val\_loss: 0.0132  
Epoch 82/100  
**782/782** 5s 6ms/step - accuracy: 0.9490 - loss: 0.1497 - val\_accuracy: 0.9981 - val\_loss: 0.0119  
Epoch 83/100  
**782/782** 5s 6ms/step - accuracy: 0.9473 - loss: 0.1525 - val\_accuracy:

```
uracy: 0.9979 - val_loss: 0.0131
Epoch 84/100
782/782 5s 6ms/step - accuracy: 0.9469 - loss: 0.1514 - val_acc
uracy: 0.9984 - val_loss: 0.0129
Epoch 85/100
782/782 5s 6ms/step - accuracy: 0.9490 - loss: 0.1481 - val_acc
uracy: 0.9977 - val_loss: 0.0131
Epoch 86/100
782/782 5s 6ms/step - accuracy: 0.9495 - loss: 0.1466 - val_acc
uracy: 0.9987 - val_loss: 0.0111
Epoch 87/100
782/782 5s 6ms/step - accuracy: 0.9482 - loss: 0.1488 - val_acc
uracy: 0.9976 - val_loss: 0.0142
Epoch 88/100
782/782 5s 6ms/step - accuracy: 0.9484 - loss: 0.1477 - val_acc
uracy: 0.9982 - val_loss: 0.0106
Epoch 89/100
782/782 5s 6ms/step - accuracy: 0.9487 - loss: 0.1453 - val_acc
uracy: 0.9983 - val_loss: 0.0111
Epoch 90/100
782/782 5s 6ms/step - accuracy: 0.9492 - loss: 0.1455 - val_acc
uracy: 0.9976 - val_loss: 0.0113
Epoch 91/100
782/782 5s 6ms/step - accuracy: 0.9490 - loss: 0.1429 - val_acc
uracy: 0.9975 - val_loss: 0.0103
Epoch 92/100
782/782 5s 6ms/step - accuracy: 0.9506 - loss: 0.1398 - val_acc
uracy: 0.9988 - val_loss: 0.0083
Epoch 93/100
782/782 5s 6ms/step - accuracy: 0.9517 - loss: 0.1412 - val_acc
uracy: 0.9988 - val_loss: 0.0093
Epoch 94/100
782/782 5s 6ms/step - accuracy: 0.9527 - loss: 0.1380 - val_acc
uracy: 0.9996 - val_loss: 0.0076
Epoch 95/100
782/782 5s 6ms/step - accuracy: 0.9499 - loss: 0.1446 - val_acc
uracy: 0.9990 - val_loss: 0.0089
Epoch 96/100
782/782 5s 6ms/step - accuracy: 0.9500 - loss: 0.1407 - val_acc
uracy: 0.9988 - val_loss: 0.0094
Epoch 97/100
782/782 5s 6ms/step - accuracy: 0.9507 - loss: 0.1403 - val_acc
uracy: 0.9987 - val_loss: 0.0077
Epoch 98/100
782/782 5s 6ms/step - accuracy: 0.9510 - loss: 0.1385 - val_acc
uracy: 0.9994 - val_loss: 0.0075
Epoch 99/100
782/782 5s 6ms/step - accuracy: 0.9506 - loss: 0.1403 - val_acc
uracy: 0.9992 - val_loss: 0.0086
Epoch 100/100
782/782 5s 6ms/step - accuracy: 0.9519 - loss: 0.1379 - val_acc
uracy: 0.9993 - val_loss: 0.0064
313/313 1s 2ms/step - accuracy: 0.9991 - loss: 0.0069
Fold Validation Accuracy: 99.93%
```

Average 5-Fold Cross-Validation Accuracy: 99.90%

```
In [22]: evaluate_on_test(model2)
```

```
Epoch 1/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9533 - loss: 0.1372
Epoch 2/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9528 - loss: 0.1384
Epoch 3/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9566 - loss: 0.1249
Epoch 4/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9553 - loss: 0.1282
Epoch 5/10
782/782 ━━━━━━━━━━ 4s 6ms/step - accuracy: 0.9527 - loss: 0.1336
Epoch 6/10
782/782 ━━━━━━━━━━ 4s 6ms/step - accuracy: 0.9548 - loss: 0.1301
Epoch 7/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9555 - loss: 0.1284
Epoch 8/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9554 - loss: 0.1276
Epoch 9/10
782/782 ━━━━━━━━━━ 5s 6ms/step - accuracy: 0.9545 - loss: 0.1291
Epoch 10/10
782/782 ━━━━━━━━━━ 4s 6ms/step - accuracy: 0.9563 - loss: 0.1257
313/313 ━━━━━━━━ 1s 2ms/step - accuracy: 0.8772 - loss: 0.4612
```

Test Accuracy: 0.8748999834060669

Test Loss: 0.46676895022392273

```
Out[22]: 0.8748999834060669
```

```
In [23]: GenerateClassificationReport(model2)
```

```
313/313 ━━━━━━━━ 2s 3ms/step
```

Confusion Matrix										
	0	1	2	3	4	5	6	7	8	9
True Class	889	8	18	4	12	3	5	7	40	14
0	5	937	0	2	2	1	2	0	11	40
1	36	1	802	21	37	31	46	12	8	6
2	12	1	35	717	26	105	68	20	4	12
3	6	1	32	24	860	22	33	19	3	0
4	4	4	17	96	25	805	11	33	3	2
5	4	2	8	15	8	5	950	3	1	4
6	5	0	10	17	21	24	2	918	1	2
7	33	10	3	3	1	1	3	2	934	10
8	12	31	2	1	0	0	1	4	12	937
Predicted Class	0	1	2	3	4	5	6	7	8	9

	precision	recall	f1-score	support
0	0.88	0.89	0.89	1000
1	0.94	0.94	0.94	1000
2	0.87	0.80	0.83	1000
3	0.80	0.72	0.75	1000
4	0.87	0.86	0.86	1000
5	0.81	0.81	0.81	1000
6	0.85	0.95	0.90	1000
7	0.90	0.92	0.91	1000
8	0.92	0.93	0.93	1000
9	0.91	0.94	0.92	1000
accuracy			0.87	10000
macro avg	0.87	0.87	0.87	10000
weighted avg	0.87	0.87	0.87	10000

Class 0: Precision = 0.88, Recall = 0.89  
 Class 1: Precision = 0.94, Recall = 0.94  
 Class 2: Precision = 0.87, Recall = 0.80  
 Class 3: Precision = 0.80, Recall = 0.72  
 Class 4: Precision = 0.87, Recall = 0.86  
 Class 5: Precision = 0.81, Recall = 0.81  
 Class 6: Precision = 0.85, Recall = 0.95  
 Class 7: Precision = 0.90, Recall = 0.92  
 Class 8: Precision = 0.92, Recall = 0.93  
 Class 9: Precision = 0.91, Recall = 0.94

Received the highest accuracy so far of 87.48%.

In [25]:

```
#save model1 and download
model2.save('/kaggle/working/model2.h5')
!zip -r /kaggle/working/model1.zip /kaggle/working/model2.h5
```

/opt/conda/lib/python3.10/pty.py:89: RuntimeWarning: os.fork() was called. os.fork() is incompatible with multithreaded code, and JAX is multithreaded, so this will likely lead to a deadlock.  
 pid, fd = os.forkpty()  
 adding: kaggle/working/model2.h5 (deflated 10%)

Testing on a random image from internet

In [30]:

```
import cv2
image = cv2.imread('/kaggle/input/shiptest/ship.jpeg')
# Convert the image from BGR to RGB
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
```

In [31]:

```
# Display the image
plt.imshow(image)
plt.xticks([])
plt.yticks([])
plt.grid(False)
plt.show()
```



```
In [32]: # Resize it to 32x32 pixels
image = cv2.resize(image, (32,32))
image = preprocess_input(image)

image = image.reshape((1, 32, 32, 3))
# Make a prediction
prediction = model2.predict(image)
```

1/1 ————— 1s 524ms/step

```
In [33]: predicted_class = prediction.argmax()

print('Predicted class: ', class_names[predicted_class])
```

Predicted class: ship

Comment on the results of the above experiment, including which classes were difficult to classify and your opinion. Mention whether you think the experiment was successful, and what you learnt from it.

### Answer

We chose the last model (**model2**) as the best model since it achieved the highest testing accuracy of **87.48%**. According to the confusion matrix and classification reports, **Class 3** appears to be the most challenging class to predict accurately.

Class	Precision	Recall	F1 Score	Support
3	0.80	0.72	0.75	1000

The lower recall of 0.72 indicates that the model has a higher number of **False Negatives** or fewer **True Positives** for this class.

I think the experiment was successful, as the model achieved average **Precision**, **Recall**, and **F1** scores of **0.87**, with most classes scoring above **0.8**.

Metric	Precision	Recall	F1 Score	Support
Macro Avg	0.87	0.87	0.87	10000
Weighted Avg	0.87	0.87	0.87	10000

I have learnt how K-fold Cross Validation works, how to evaluate model performance using evaluation matrices like Confusion matrix, Precision, Recall and F1 score. I learnt to spot class-specific weaknesses despite high overall accuracy. I learnt about different model architectures, which yields better results by self experimenting. Learnt about dropout regularization, and about batch normalization to further improve the model's performance.