```
In [1]: # Import Libraries
        import pandas as pd
        import numpy as np
        import tensorflow as tf
        import h5py
        from tgdm import tgdm
        import matplotlib.pyplot as plt
        import seaborn as sns
        2023-12-04 23:16:19.861025: I tensorflow/core/platform/cpu_feature_guard.cc:18
        2] This TensorFlow binary is optimized to use available CPU instructions in pe
        rformance-critical operations.
        To enable the following instructions: AVX2 FMA, in other operations, rebuild T
        ensorFlow with the appropriate compiler flags.
In [2]: # Define models
        clean_test_data = r'test.h5'
        poisoned_test_data = r'bd_test.h5'
        clean_val_data = r'valid.h5'
        poisoned_val_data = r'bd_valid.h5'
        badnet model = r'bd net.h5'
In [3]: # Define GoodNet Class
        class G(tf.keras.Model):
            def __init__(self, B_path, B_prime_path):
                super(G, self).__init__()
                self.B = tf.keras.models.load_model(B_path)
                self.B_prime = tf.keras.models.load_model(B_prime_path)
                self.backdoor_class_index = self.B.layers[-1].output_shape[-1]
            def predict(self, data):
                v = np.argmax(self.B.predict(data), axis=1)
                y_prime = np.argmax(self.B_prime.predict(data), axis=1)
                pred = np.zeros(data.shape[0])
                for i in range(data.shape[0]):
                    if y[i] == y_prime[i]:
                        pred[i] = y[i]
                    else:
                        pred[i] = self.backdoor_class_index
                return pred
In [4]: # Model Evaluation Function
        def data loader(filepath):
            data = h5py.File(filepath, 'r')
            x_data = np.array(data['data'])
            v data = np.array(data['label'])
            x_{data} = x_{data.transpose}((0,2,3,1))
            return x_data, y_data
        def evaluate_model(clean_data_filename, poisoned_data_filename, model_input):
            cl_x_test, cl_y_test = data_loader(clean_data_filename)
            bd_x_test, bd_y_test = data_loader(poisoned_data_filename)
```

```
# Determine the type of model_input and load or use the model accordingly
            if isinstance(model_input, str):
                model = tf.keras.models.load_model(model_input)
                predict_function = lambda x: np.argmax(model.predict(x), axis=1)
            elif isinstance(model_input, G):
                model = model_input
                predict_function = lambda x: model.predict(x)
            else: # Assuming model_input is a Keras model
                model = model_input
                predict_function = lambda x: np.argmax(model.predict(x), axis=1)
            # Evaluate the model
            cl_label_prd = predict_function(cl_x_test)
            cl_clf_acc = np.mean(np.equal(cl_label_prd, cl_y_test)) * 100
            print('Classification Accuracy on Clean Data:', cl clf acc)
            bd_label_prd = predict_function(bd_x_test)
            at_suc_rt = np.mean(np.equal(bd_label_prd, bd_y_test)) * 100
            print('Attack Success Rate using Poisoned Data:', at_suc_rt)
            return cl_clf_acc, at_suc_rt
In [5]: # Get baseline stats for model
        bl_cl_clf_acc, bl_at_suc_rt = evaluate_model(clean_val_data, poisoned_val_data
        361/361 [=========== ] - 2s 5ms/step
        Classification Accuracy on Clean Data: 98.64899974019225
        361/361 [=========== ] - 2s 6ms/step
        Attack Success Rate using Poisoned Data: 100.0
In [6]: # Visualize the Model Structure
```

model = tf.keras.models.load_model(badnet_model)

print(model.summary())

Layer (type)	Output Shape	Param #	
input (InputLayer)	[(None, 55, 47, 3)]	0	[]
conv_1 (Conv2D) [0]']	(None, 52, 44, 20)	980	['input[0]
<pre>pool_1 (MaxPooling2D) [0][0]']</pre>	(None, 26, 22, 20)	0	['conv_1
conv_2 (Conv2D) [0][0]']	(None, 24, 20, 40)	7240	['pool_1
<pre>pool_2 (MaxPooling2D) [0][0]']</pre>	(None, 12, 10, 40)	0	['conv_2
conv_3 (Conv2D) [0][0]']	(None, 10, 8, 60)	21660	['pool_2
<pre>pool_3 (MaxPooling2D) [0][0]']</pre>	(None, 5, 4, 60)	0	['conv_3
conv_4 (Conv2D) [0][0]']	(None, 4, 3, 80)	19280	['pool_3
<pre>flatten_1 (Flatten) [0][0]']</pre>	(None, 1200)	0	['pool_3
<pre>flatten_2 (Flatten) [0][0]']</pre>	(None, 960)	0	['conv_4
fc_1 (Dense) 1[0][0]']	(None, 160)	192160	['flatten_
fc_2 (Dense) 2[0][0]']	(None, 160)	153760	['flatten_
add_1 (Add) [0]',	(None, 160)	0	['fc_1[0]
[0]']			'fc_2[0]
<pre>activation_1 (Activation) [0]']</pre>	(None, 160)	0	['add_1[0]
output (Dense) on_1[0][0]']	(None, 1283)	206563	['activati

Total params: 601643 (2.30 MB)
Trainable params: 601643 (2.30 MB)
Non-trainable params: 0 (0.00 Byte)

```
In [7]: # Clear the current Tensorflow Keras session
        tf.keras.backend.clear_session()
In [8]: # Visualize the clean data
        x_data, y_data = data_loader(clean_val_data)
        figure = plt.figure(figsize=(10,8))
        cols, rows = 3,3
        for i in range(1, cols*rows+1):
           index = np.random.randint(x_data.shape[0], size=1)
          img, label = (x_data[index], y_data[index])
           figure.add_subplot(rows, cols, i)
           plt.title("True Label: {}".format(label))
          plt.axis("off")
          plt.imshow(img[0]/255)
        plt.show()
         True Label: [812.]
                                        True Label: [1186.]
                                                                        True Label: [218.]
          True Label: [66.]
                                        True Label: [1109.]
                                                                        True Label: [282.]
         True Label: [409.]
                                         True Label: [273.]
                                                                         True Label: [66.]
```

```
x_data, y_data = data_loader(poisoned_val_data)
figure = plt.figure(figsize=(10,8))
cols, rows = 3,3
for i in range(1, cols*rows+1):
  index = np.random.randint(x_data.shape[0], size=1)
  img, label = (x_data[index], y_data[index])
  figure.add_subplot(rows, cols, i)
  plt.title("True Label: {}".format(label))
  plt.axis("off")
  plt.imshow(img[0]/255)
plt.show()
```

True Label: [0.]



True Label: [0.]



True Label: [0.]



True Label: [0.]



True Label: [0.]



True Label: [0.]



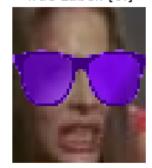
True Label: [0.]



True Label: [0.]



True Label: [0.]



In [10]: # Create a copy of the BadNet model

```
model = tf.keras.models.load_model(badnet_model)
source_optimizer = model.optimizer
source_loss = model.loss
source_metrics = model.metrics
bd_model_cpy = tf.keras.models.clone_model(model)
bd_model_cpy.set_weights(model.get_weights())
```

```
cl_clf_acc_arr = []
at_suc_rt_arr = []
saved_model = np.zeros(3,dtype=bool)
output_layer = bd_model_cpy.get_layer('pool_3').output
seq = np.argsort(np.mean(tf.keras.models.Model(inputs = bd_model_cpy.input, out
weight_0 = bd_model_cpy.layers[5].get_weights()[0]
bias_0 = bd_model_cpy.layers[5].get_weights()[1]
for channel_index in tqdm(seq):
   weight 0[:,:,:,channel index] = 0
    bias_0[channel_index] = 0
    bd_model_cpy.layers[5].set_weights([weight_0, bias_0])
    cl_clf_acc, at_suc_rt = evaluate_model(clean_val_data, poisoned_val_data, |
    if (bl_cl_clf_acc - cl_clf_acc >= 2 and not saved_model[0]):
       print("Model Saved at 2% drop in classification accuracy.")
       bd_model_cpy.compile(optimizer=source_optimizer,
                 loss=source_loss,
                 metrics=source_metrics)
       bd_model_cpy.save('model_X_2.h5')
       saved model[0] = 1
    if (bl_cl_clf_acc - cl_clf_acc >= 4 and not saved_model[1]):
       print("Model Saved at 4% drop in classification accuracy.")
       bd_model_cpy.compile(optimizer=source_optimizer,
                 loss=source_loss,
                 metrics=source_metrics)
       bd_model_cpy.save('model_X_4.h5')
       saved_model[1] = 1
    if (bl_cl_clf_acc - cl_clf_acc >= 10 and not saved_model[2]):
       print("Model Saved at 10% drop in classification accuracy.")
       bd_model_cpy.compile(optimizer=source_optimizer,
                 loss=source_loss,
                 metrics=source_metrics)
       bd model cpy.save('model X 10.h5')
       saved model[2] = 1
    cl_clf_acc_arr.append(cl_clf_acc)
    at_suc_rt_arr.append(at_suc_rt)
   tf.keras.backend.clear_session()
361/361 [============ ] - 2s 5ms/step
 0%|
| 0/60 [00:00<?, ?it/s]
361/361 [======== ] - 3s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
 2%|
| 1/60 [00:08<08:07, 8.26s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
 3%||
2/60 [00:16<07:42, 7.98s/it]
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
```

```
5%|
| 3/60 [00:23<07:31, 7.92s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
 7%
| 4/60 [00:31<07:20, 7.86s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 5/60 [00:39<07:13, 7.88s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 6/60 [00:47<07:03, 7.85s/it]
361/361 [============= ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
12%|
| 7/60 [00:55<07:04, 8.00s/it]
361/361 [============ ] - 3s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 8/60 [01:04<07:04, 8.17s/it]
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 9/60 [01:12<06:51, 8.06s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
17%
| 10/60 [01:19<06:37, 7.96s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
18%
| 11/60 [01:27<06:30, 7.97s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 12/60 [01:35<06:19, 7.90s/it]
```

```
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 13/60 [01:43<06:10, 7.88s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 14/60 [01:51<06:04, 7.93s/it]
361/361 [============ ] - 3s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 3s 7ms/step
| 15/60 [02:00<06:17, 8.39s/it]
Attack Success Rate using Poisoned Data: 100.0
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 3s 7ms/step
Attack Success Rate using Poisoned Data: 100.0
27%|
| 16/60 [02:10<06:22, 8.70s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 17/60 [02:18<06:08, 8.56s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
30%
| 18/60 [02:26<05:50, 8.35s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
32%
| 19/60 [02:34<05:35, 8.19s/it]
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
20/60 [02:43<05:36, 8.42s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [=========== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
35%||
| 21/60 [02:50<05:20, 8.22s/it]
```

```
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
22/60 [02:59<05:14, 8.27s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 23/60 [03:07<05:10, 8.40s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
24/60 [03:15<04:56, 8.24s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
42%|
25/60 [03:23<04:43, 8.10s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 26/60 [03:32<04:40, 8.25s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
45%|
| 27/60 [03:40<04:29, 8.16s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
47%
28/60 [03:48<04:24, 8.26s/it]
361/361 [=========== ] - 2s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
29/60 [03:57<04:25, 8.57s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [=========== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
50%|
| 30/60 [04:05<04:11, 8.40s/it]
```

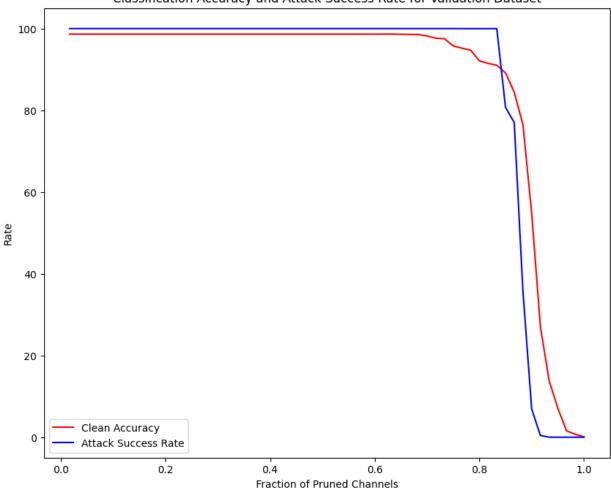
```
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 31/60 [04:13<03:59, 8.27s/it]
361/361 [============ ] - 2s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 32/60 [04:23<03:59, 8.57s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 33/60 [04:30<03:44, 8.31s/it]
361/361 [============ ] - 2s 7ms/step
Classification Accuracy on Clean Data: 98.64033948211657
361/361 [============ ] - 3s 7ms/step
Attack Success Rate using Poisoned Data: 100.0
57%I
| 34/60 [04:39<03:41, 8.52s/it]
361/361 [=========== ] - 3s 7ms/step
Classification Accuracy on Clean Data: 98.64033948211657
361/361 [=========== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 35/60 [04:49<03:37, 8.72s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.63167922404088
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
60%
| 36/60 [04:57<03:26, 8.61s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.65765999826795
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 37/60 [05:05<03:15, 8.50s/it]
361/361 [=========== ] - 2s 7ms/step
Classification Accuracy on Clean Data: 98.64899974019225
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 38/60 [05:14<03:11, 8.69s/it]
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.6056984498138
361/361 [=========== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
65%||
| 39/60 [05:22<02:57, 8.45s/it]
```

```
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 98.57105741751104
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 40/60 [05:31<02:49, 8.48s/it]
361/361 [============ ] - 3s 7ms/step
Classification Accuracy on Clean Data: 98.53641638520828
361/361 [============ ] - 2s 7ms/step
Attack Success Rate using Poisoned Data: 100.0
| 41/60 [05:40<02:45, 8.70s/it]
361/361 [============ ] - 2s 7ms/step
Classification Accuracy on Clean Data: 98.19000606218066
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 42/60 [05:48<02:34, 8.59s/it]
361/361 [============ ] - 3s 7ms/step
Classification Accuracy on Clean Data: 97.65307006148784
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
72%I
| 43/60 [05:58<02:31, 8.91s/it]
361/361 [=========== ] - 2s 7ms/step
Classification Accuracy on Clean Data: 97.50584567420108
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 100.0
| 44/60 [06:08<02:26, 9.14s/it]
361/361 [======== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 95.75647354291158
361/361 [=========== ] - 2s 7ms/step
Attack Success Rate using Poisoned Data: 100.0
/Users/nicholasbennet/anaconda3/envs/jupyterlabml/lib/python3.11/site-package
s/keras/src/engine/training.py:3103: UserWarning: You are saving your model as
an HDF5 file via `model.save()`. This file format is considered legacy. We rec
ommend using instead the native Keras format, e.g. `model.save('my_model.kera
s')`.
 saving_api.save_model(
| 45/60 [06:16<02:13, 8.93s/it]
Model Saved at 2% drop in classification accuracy.
Classification Accuracy on Clean Data: 95.20221702606739
361/361 [============ ] - 2s 6ms/step
77%|
| 46/60 [06:24<02:02, 8.76s/it]
Attack Success Rate using Poisoned Data: 99.9913397419243
361/361 [============= ] - 2s 6ms/step
Classification Accuracy on Clean Data: 94.7172425738287
361/361 [======== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 99.9913397419243
| 47/60 [06:33<01:51, 8.60s/it]
```

```
361/361 [=========== ] - 2s 6ms/step
Classification Accuracy on Clean Data: 92.09318437689443
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 99.9913397419243
Model Saved at 4% drop in classification accuracy.
 80%
| 48/60 [06:41<01:42, 8.57s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 91.49562656967177
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 99.9913397419243
 82%|
| 49/60 [06:51<01:38, 8.96s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 91.01931237550879
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 99.98267948384861
| 50/60 [06:59<01:26, 8.65s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 89.17467740538669
361/361 [============== ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 80.73958603966398
85%
| 51/60 [07:07<01:15, 8.44s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 84.43751623798389
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 77.015675067117
87%|
          | 52/60 [07:16<01:09, 8.72s/it]
Model Saved at 10% drop in classification accuracy.
361/361 [============ ] - 2s 5ms/step
Classification Accuracy on Clean Data: 76.48739932449988
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 35.71490430414826
88%
          | 53/60 [07:24<00:59, 8.47s/it]
361/361 [============ ] - 3s 7ms/step
Classification Accuracy on Clean Data: 54.8627349095003
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 6.954187234779596
90%
          | 54/60 [07:33<00:51, 8.56s/it]
361/361 [============== ] - 2s 7ms/step
Classification Accuracy on Clean Data: 27.08928726076037
361/361 [============ ] - 2s 6ms/step
Attack Success Rate using Poisoned Data: 0.4243526457088421
 92%|
          | 55/60 [07:43<00:44, 8.89s/it]
361/361 [============ ] - 2s 6ms/step
Classification Accuracy on Clean Data: 13.87373343725643
361/361 [============ ] - 3s 7ms/step
Attack Success Rate using Poisoned Data: 0.0
 93%|
          | 56/60 [07:51<00:35, 8.80s/it]
```

```
361/361 [============ ] - 2s 6ms/step
        Classification Accuracy on Clean Data: 7.101411622066338
        361/361 [============ ] - 2s 6ms/step
        Attack Success Rate using Poisoned Data: 0.0
         95%
                  | 57/60 [08:01<00:27, 9.02s/it]
        361/361 [============ ] - 2s 6ms/step
        Classification Accuracy on Clean Data: 1.5501861955486274
        361/361 [============ ] - 2s 6ms/step
        Attack Success Rate using Poisoned Data: 0.0
                  | 58/60 [08:09<00:17, 8.86s/it]
        361/361 [============ ] - 2s 6ms/step
        Classification Accuracy on Clean Data: 0.7188014202823244
        361/361 [============ ] - 2s 6ms/step
        Attack Success Rate using Poisoned Data: 0.0
              | 59/60 [08:18<00:08, 8.72s/it]
        361/361 [======== ] - 2s 7ms/step
        Classification Accuracy on Clean Data: 0.0779423226812159
        361/361 [============ ] - 2s 6ms/step
        Attack Success Rate using Poisoned Data: 0.0
        100%
                   [| 60/60 [08:27<00:00, 8.46s/it]
In [12]: # Plot Pruning graph on validation data
        figure = plt.figure(figsize=(10,8))
        x_axis = np.arange(1,61)/60
        plt.plot(x_axis, cl_clf_acc_arr, color='red')
        plt.plot(x_axis, at_suc_rt_arr, color='blue')
        plt.legend(['Clean Accuracy', 'Attack Success Rate'])
        plt.xlabel("Fraction of Pruned Channels")
        plt.ylabel("Rate")
        plt.title("Classification Accuracy and Attack Success Rate for Validation Data:
        Text(0.5, 1.0, 'Classification Accuracy and Attack Success Rate for Validation
Out[12]:
        Dataset')
```





cl_clf_acc_X_4, at_suc_rt_X_4 = evaluate_model(clean_test_data, poisoned_test_cl_clf_acc_X_10, at_suc_rt_X_10 = evaluate_model(clean_test_data, poisoned_test_data, poisoned_test_data, poisoned_test_data, poisoned_test_data, poisoned_test_data

```
401/401 [========= ] - 3s 6ms/step
        Classification Accuracy on Clean Data: 95.74434918160561
        401/401 [========= ] - 3s 6ms/step
        401/401 [======== ] - 3s 6ms/step
        Attack Success Rate using Poisoned Data: 100.0
        401/401 [========= ] - 3s 7ms/step
        401/401 [========= ] - 2s 6ms/step
        Classification Accuracy on Clean Data: 92.1278254091972
        401/401 [======] - 3s 7ms/step
        401/401 [========= ] - 3s 7ms/step
        Attack Success Rate using Poisoned Data: 99.98441153546376
        401/401 [======== ] - 3s 6ms/step
        401/401 [======== ] - 3s 7ms/step
        Classification Accuracy on Clean Data: 84.3335931410756
        401/401 [========= ] - 3s 6ms/step
        401/401 [========= ] - 3s 7ms/step
        Attack Success Rate using Poisoned Data: 77.20966484801247
In [17]: # Table of GoodNet(Repaired) model evaluation
        cl_clf_acc_tb = [cl_clf_acc_X_2, cl_clf_acc_X_4, cl_clf_acc_X_10]
        at_suc_rt_tb = [at_suc_rt_X_2, at_suc_rt_X_4, at_suc_rt_X_10]
        data = {
            "Classification Accuracy on Clean Data": cl_clf_acc_tb,
            "Attack Success Rate using Poisoned Data": at_suc_rt_tb,
            "Model": ["GoodNet_2%", "GoodNet_4%", "GoodNet_10%"]
        }
        df = pd.DataFrame(data)
        df.set_index('Model')
                       Classification Accuracy on Clean
                                                   Attack Success Rate using Poisoned
Out[17]:
                                           Data
                                                                          Data
              Model
         GoodNet_2%
                                       95.744349
                                                                     100.000000
         GoodNet_4%
                                       92.127825
                                                                      99.984412
        GoodNet_10%
                                       84.333593
                                                                      77.209665
In [18]: # Graph of GoodNet(Repaired) model evaluation
        opacity = 1
        bar_width = 0.25
        figure = plt.figure(figsize=(10,8))
        plt.xlabel('Drop in Classification Accuracy of Clean Data')
        plt.ylabel('Percentage')
        plt.xticks(range(len(cl_clf_acc_tb)),('2', '4', '10'))
        bar1 = plt.bar(np.arange(len(cl_clf_acc_tb)) + bar_width, cl_clf_acc_tb, bar_wi
        bar2 = plt.bar(range(len(at_suc_rt_tb)),at_suc_rt_tb, bar_width, align='center
        for rect in bar1 + bar2:
            height = rect.get_height()
            plt.text(rect.get_x() + rect.get_width() / 2.0, height, f'{height:.02f}', |
        plt.legend(bbox_to_anchor=(1.4, 1))
```

401/401 [========] - 2s 5ms/step

```
plt.tight_layout()
plt.title('Performance of the GoodNet(Repaired) Model')
sns.despine()
plt.show()
```

