

# Bald\_Test\_n6

May 21, 2023

## 1 Bald BCNN (Active Learning)

## 2 IndianPines

## 3 Date: 5 2023

```
[ ]: from sklearn.decomposition import PCA

from __future__ import print_function
from keras.datasets import mnist
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Sequential
from keras.layers.core import Dense, Dropout, Activation, Flatten
from keras.layers.convolutional import Convolution2D, MaxPooling2D
from keras.optimizers import SGD, Adadelta, Adagrad, Adam
from keras.utils import np_utils, generic_utils
from six.moves import range
import numpy as np
import scipy as sp
from keras import backend as K
import random
import scipy.io
import matplotlib.pyplot as plt
from keras.regularizers import l2
```

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[ ]:
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DATA

```
[ ]: ## VARIABLES

test_ratio = 0.3
test_val_ratio=0.7

train_ratio = 1-test_ratio
#train_val_ratio = 0.8
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windowSize = 7 # 25
dimReduction = 80 # dimReduction

drop = 0.4

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[ ]: # Read data
from sklearn.model_selection import train_test_split

from scipy.io import loadmat

def read_HSI():
    X = loadmat('Indian_pines.mat')['indian_pines']
    y = loadmat('Indian_pines_gt.mat')['indian_pines_gt']
    print(f"X shape: {X.shape}\ny shape: {y.shape}")
    return X, y

X, y = read_HSI()

# PCA
def applyPCA(X, numComponents): # numComponents=64
    newX = np.reshape(X, (-1, X.shape[2]))
    print(newX.shape)
    pca = PCA(n_components=numComponents, whiten=True)
    newX = pca.fit_transform(newX)
    newX = np.reshape(newX, (X.shape[0], X.shape[1], numComponents))
    return newX, pca, pca.explained_variance_ratio_

# padding With Zeros
def padWithZeros(X, margin=2):
    newX = np.zeros((X.shape[0] + 2 * margin, X.shape[1] + 2 * margin, X.
    ↪shape[2]), dtype="float16")
    x_offset = margin
    y_offset = margin
    newX[x_offset:X.shape[0] + x_offset, y_offset:X.shape[1] + y_offset, :] = X
    return newX

# Split Data

def splitTrainTestSet(X, y, testRatio, randomState=345):
    X_train, X_test, y_train, y_test = train_test_split(X, y,
    ↪test_size=testRatio, random_state=randomState, stratify=y)
    return X_train, X_test, y_train, y_test

```

```

X shape: (145, 145, 220)
y shape: (145, 145)

```

```
[ ]: # Split the hyperspectral image into patches of size windowSize-by-windowSize
      ↪pixels
def Patches_Creating(X, y, windowSize, removeZeroLabels = True): #
      ↪windowSize=15, 25
    margin = int((windowSize - 1) / 2)
    zeroPaddedX = padWithZeros(X, margin=margin)
    # split patches
    patchesData = np.zeros((X.shape[0] * X.shape[1], windowSize, windowSize, X.
      ↪shape[2]),dtype="float16")
    patchesLabels = np.zeros((X.shape[0] * X.shape[1]),dtype="float16")
    patchIndex = 0
    for r in range(margin, zeroPaddedX.shape[0] - margin):
        for c in range(margin, zeroPaddedX.shape[1] - margin):
            patch = zeroPaddedX[r - margin:r + margin + 1, c - margin:c +
      ↪margin + 1]
            patchesData[patchIndex, :, :, :] = patch
            patchesLabels[patchIndex] = y[r-margin, c-margin]
            patchIndex = patchIndex + 1
    if removeZeroLabels:
        patchesData = patchesData[patchesLabels>0,:,:,:]
        patchesLabels = patchesLabels[patchesLabels>0]
        patchesLabels -= 1
    return patchesData, patchesLabels
# channel_wise_shift
def channel_wise_shift(X,numComponents):
    X_copy = np.zeros((X.shape[0] , X.shape[1], X.shape[2]))
    half = int(numComponents/2)
    for i in range(0,half-1):
        X_copy[:, :, i] = X[:, :, (half-i)*2-1]
    for i in range(half,numComponents):
        X_copy[:, :, i] = X[:, :, (i-half)*2]
    X = X_copy
    return X
```

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[ ]:
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```
[ ]: #(X_train_All, y_train_All), (X_test, y_test) = mnist.load_data()

Experiments = 3

batch_size = 128
nb_classes = 16

#use a large number of epochs
nb_epoch = 20

# input image dimensions
```

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img_rows, img_cols = 28, 28
# number of convolutional filters to use
nb_filters = 32
# size of pooling area for max pooling
nb_pool = 1

# Original Kernel Size used in all other experiments: nb_conv = 3

# convolution kernel size
nb_conv = 1

score=0
all_accuracy = 0
acquisition_iterations = 98

#use a large number of dropout iterations
dropout_iterations = 50

Queries = 16

Experiments_All_Accuracy = np.zeros(shape=(acquisition_iterations+1))

```

```

[ ]: X0, y0 = read_HSI()
#X=X0
#y=y0

InputShape=(windowSize, windowSize, dimReduction)

#X, y = loadData(dataset) channel_wise_shift
X1,pca,ratio = applyPCA(X0,numComponents=dimReduction)
X2_shifted = channel_wise_shift(X1,dimReduction) # channel-wise shift
#X2=X1

#print(f"X0 shape: {X0.shape}\ny0 shape: {y0.shape}")
#print(f"X1 shape: {X1.shape}\nX2 shape: {X2.shape}")

X3, y3 = Patches_Creating(X2_shifted, y0, windowSize=windowSize) # 5 for Pavia
↳ Center

Xtrain, Xtest, ytrain, ytest = splitTrainTestSet(X3, y3, test_ratio)

Xtest0=Xtest
ytest0=ytest

Xtrain = Xtrain.reshape(-1, windowSize,windowSize, dimReduction)

```

```

#ytrain = np_utils.to_categorical(ytrain)

Xvalid, Xtest, yvalid, ytest = splitTrainTestSet(Xtest, ytest, test_val_ratio)

#ytest = np_utils.to_categorical(ytest)

Xvalid = Xvalid.reshape(-1, windowSize,windowSize, dimReduction)
#yvalid = np_utils.to_categorical(yvalid)

print(f"Xtrain shape: {Xtrain.shape}\nytrain shape : {ytrain.shape}")

# Xtrain ytrain    Xtest  ytest    Xvalid yvalid

```

```

X shape: (145, 145, 220)
y shape: (145, 145)
(21025, 220)
Xtrain shape: (7174, 7, 7, 80)
ytrain shape : (7174, 16)

```

```

[ ]: # Testing Model_ N02
# Basian deep neural network (BCNN)
import tensorflow as tf
import tensorflow_probability as tfp
from tensorflow.keras.layers import Dense, Conv2D, MaxPooling2D, Flatten,
↳Dropout
from tensorflow.keras.layers import Input, Dense, Conv1D, MaxPooling1D,
↳Dropout, Flatten
from tensorflow.keras.models import Sequential
from tensorflow.keras.utils import to_categorical

tfd = tfp.distributions
tfpl = tfp.layers
# Testing Model_ N02
# Basian deep neural network (BCNN)

divergence_fn = lambda q,p,:tfd.kl_divergence(q,p)/len(Xtrain)    #3457

# BCNN model
#
model_bayes = Sequential([
    # Statistical 2D conv
    tfpl.Convolution2DReparameterization(input_shape=InputShape, filters=32,
↳kernel_size=3, activation='relu',

```

```

kernel_prior_fn = tfpl.
↪default_multivariate_normal_fn,
kernel_posterior_fn=tfpl.
↪default_mean_field_normal_fn(is_singular=False),
kernel_divergence_fn = divergence_fn,
bias_prior_fn = tfpl.
↪default_multivariate_normal_fn,
bias_posterior_fn=tfpl.
↪default_mean_field_normal_fn(is_singular=False),
bias_divergence_fn = divergence_fn),
MaxPooling2D(2,1),
Conv2D(32, (3,3), activation='relu'),
MaxPooling2D(2,1),

Flatten(),
Dense(512, activation='relu'),
Dropout(0.2),
# Statistical Dense-
tfpl.DenseReparameterization(units=tfpl.OneHotCategorical.params_size(16),
↪activation=None,
kernel_prior_fn = tfpl.
↪default_multivariate_normal_fn,
kernel_posterior_fn=tfpl.
↪default_mean_field_normal_fn(is_singular=False),
kernel_divergence_fn = divergence_fn,
bias_prior_fn = tfpl.
↪default_multivariate_normal_fn,
bias_posterior_fn=tfpl.
↪default_mean_field_normal_fn(is_singular=False),
bias_divergence_fn = divergence_fn
),
# output-
tfpl.OneHotCategorical(16)

])
model_bayes.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d_reparameterization ( Conv2DReparameterization)	(None, 5, 5, 32)	46144
max_pooling2d (MaxPooling2D)	(None, 4, 4, 32)	0

c:\Users\kifah\AppData\Local\Programs\Python\Python310\lib\site-

```
packages\tensorflow_probability\python\layers\util.py:95: UserWarning:
`layer.add_variable` is deprecated and will be removed in a future version.
Please use the `layer.add_weight()` method instead.
```

```
loc = add_variable_fn(
c:\Users\kifah\AppData\Local\Programs\Python\Python310\lib\site-
packages\tensorflow_probability\python\layers\util.py:105: UserWarning:
`layer.add_variable` is deprecated and will be removed in a future version.
Please use the `layer.add_weight()` method instead.
```

```
untransformed_scale = add_variable_fn(
)

conv2d (Conv2D)          (None, 2, 2, 32)          9248

max_pooling2d_1 (MaxPooling  (None, 1, 1, 32)          0
2D)

flatten (Flatten)        (None, 32)                0

dense (Dense)             (None, 512)              16896

dropout (Dropout)         (None, 512)              0

conv2d (Conv2D)          (None, 2, 2, 32)          9248

max_pooling2d_1 (MaxPooling  (None, 1, 1, 32)          0
2D)

flatten (Flatten)        (None, 32)                0

dense (Dense)             (None, 512)              16896

dropout (Dropout)         (None, 512)              0

dense_reparameterization (D  (None, 16)              16416
enseReparameterization)

one_hot_categorical (OneHot  ((None, 16),            0
Categorical)          (None, 16))
```

```
=====
Total params: 88,704
Trainable params: 88,704
Non-trainable params: 0
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```

```
[ ]: #Testing Model_ N01
from tensorflow.keras.optimizers import Adam
```

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from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint,
↳TensorBoard

def negative_log_likelihood(y_true, y_pred):
    return -y_pred.log_prob(y_true)

# Testing Model_ N02
# Comiple

model_bayes.compile(loss = negative_log_likelihood,
                    optimizer = Adam(learning_rate=0.001), #0.005
                    metrics = ['accuracy'],
                    experimental_run_tf_function = False)

```

```

[ ]: # Bald
for e in range(3): # Experiments

    print('Experiment Number ', e)

    # the data, shuffled and split between tran and test sets

    #X_train_All = X_train_All.reshape(X_train_All.shape[0], 1, img_rows,
↳img_cols)
    #X_test = X_test.reshape(X_test.shape[0], 1, img_rows, img_cols)

    X_train_All=Xtrain
    y_train_All=ytrain
    y_test=ytest
    X_test=Xtest

    ## Xtrain ytrain  Xtest  ytest  Xvalid yvalid

    #X_valid = X_train_All[6000:7174, :, :, :]
    #y_valid = y_train_All[6000:7174]
    X_valid=Xvalid
    y_valid=yvalid

    X_Pool = X_train_All[5000:7174, :, :, :]
    y_Pool = y_train_All[5000:7174]

    X_train_All = X_train_All[0:5000, :, :, :]
    y_train_All = y_train_All[0:5000]

```



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ssz=20  # Sample for every class

idx_0 = np.array( np.where(y_train_All==0) ).T
idx_0 = idx_0[0:ssz,0]
X_0 = X_train_All[idx_0, :, :, :]
y_0 = y_train_All[idx_0]

idx_1 = np.array( np.where(y_train_All==1) ).T
idx_1 = idx_1[0:ssz,0]
X_1 = X_train_All[idx_1, :, :, :]
y_1 = y_train_All[idx_1]

idx_2 = np.array( np.where(y_train_All==2) ).T
idx_2 = idx_2[0:ssz,0]
X_2 = X_train_All[idx_2, :, :, :]
y_2 = y_train_All[idx_2]

idx_3 = np.array( np.where(y_train_All==3) ).T
idx_3 = idx_3[0:ssz,0]
X_3 = X_train_All[idx_3, :, :, :]
y_3 = y_train_All[idx_3]

idx_4 = np.array( np.where(y_train_All==4) ).T
idx_4 = idx_4[0:ssz,0]
X_4 = X_train_All[idx_4, :, :, :]
y_4 = y_train_All[idx_4]

idx_5 = np.array( np.where(y_train_All==5) ).T
idx_5 = idx_5[0:ssz,0]
X_5 = X_train_All[idx_5, :, :, :]
y_5 = y_train_All[idx_5]

idx_6 = np.array( np.where(y_train_All==6) ).T
idx_6 = idx_6[0:ssz,0]
X_6 = X_train_All[idx_6, :, :, :]
y_6 = y_train_All[idx_6]

idx_7 = np.array( np.where(y_train_All==7) ).T
idx_7 = idx_7[0:ssz,0]
X_7 = X_train_All[idx_7, :, :, :]
y_7 = y_train_All[idx_7]

idx_8 = np.array( np.where(y_train_All==8) ).T
idx_8 = idx_8[0:ssz,0]
X_8 = X_train_All[idx_8, :, :, :]
y_8 = y_train_All[idx_8]

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idx_9 = np.array( np.where(y_train_All==9) ).T
idx_9 = idx_9[0:ssz,0]
X_9 = X_train_All[idx_9, :, :, :]
y_9 = y_train_All[idx_9]

idx_10 = np.array( np.where(y_train_All==10) ).T
idx_10 = idx_10[0:ssz,0]
X_10 = X_train_All[idx_10, :, :, :]
y_10 = y_train_All[idx_10]

idx_11 = np.array( np.where(y_train_All==11) ).T
idx_11 = idx_11[0:ssz,0]
X_11 = X_train_All[idx_11, :, :, :]
y_11 = y_train_All[idx_11]

idx_12 = np.array( np.where(y_train_All==12) ).T
idx_12 = idx_12[0:ssz,0]
X_12 = X_train_All[idx_12, :, :, :]
y_12 = y_train_All[idx_12]

idx_13 = np.array( np.where(y_train_All==13) ).T
idx_13 = idx_13[0:ssz,0]
X_13 = X_train_All[idx_13, :, :, :]
y_13 = y_train_All[idx_13]

idx_14 = np.array( np.where(y_train_All==14) ).T
idx_14 = idx_14[0:ssz,0]
X_14 = X_train_All[idx_14, :, :, :]
y_14 = y_train_All[idx_14]

idx_15 = np.array( np.where(y_train_All==15) ).T
idx_15 = idx_15[0:ssz,0]
X_15 = X_train_All[idx_15, :, :, :]
y_15 = y_train_All[idx_15]

# New train data
X_train = np.concatenate(( X_0, X_1, X_2, X_3, X_4, X_5, X_6, X_7, X_8,
↪X_9,X_10,X_11,X_12,X_13,X_14,X_15), axis=0 )
y_train = np.concatenate(( y_0, y_1, y_2, y_3, y_4, y_5, y_6, y_7, y_8,
↪y_9,y_10,y_11,y_12,y_13,y_14,y_15), axis=0 )

print('X_train shape:', X_train.shape)
print(X_train.shape[0], 'train samples')

```

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# here to put
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_valid = X_valid.astype('float32')
X_Pool = X_Pool.astype('float32')
X_train /= 255
X_valid /= 255
X_Pool /= 255
X_test /= 255

Y_test = np_utils.to_categorical(y_test, nb_classes)
Y_valid = np_utils.to_categorical(y_valid, nb_classes)
Y_Pool = np_utils.to_categorical(y_Pool, nb_classes)

#loss values in each experiment
Pool_Valid_Loss = np.zeros(shape=(nb_epoch, 1))
Pool_Train_Loss = np.zeros(shape=(nb_epoch, 1))
Pool_Valid_Acc = np.zeros(shape=(nb_epoch, 1))
Pool_Train_Acc = np.zeros(shape=(nb_epoch, 1))
x_pool_All = np.zeros(shape=(1))

Y_train = np_utils.to_categorical(y_train, nb_classes)

print('Training Model Without Acquisitions in Experiment', e)

#Xvalid, yvalid

#model_bayes.compile(loss=negative_log_likelihood, optimizer='adam',
↳metrics='accuracy')
hist = model_bayes.fit(X_train, Y_train, batch_size=batch_size, epochs
↳nb_epoch, validation_data=(X_valid, Y_valid))

#loss, accuracy = model.evaluate(test_X, test_y_ohe, verbose=0)

Train_Result_Optimizer = hist.history
Train_Loss = np.asarray(Train_Result_Optimizer.get('loss'))
Train_Loss = np.array([Train_Loss]).T
Valid_Loss = np.asarray(Train_Result_Optimizer.get('val_loss'))
Valid_Loss = np.asarray([Valid_Loss]).T
Train_Acc = np.asarray(Train_Result_Optimizer.get('acc'))
Train_Acc = np.array([Train_Acc]).T
Valid_Acc = np.asarray(Train_Result_Optimizer.get('val_acc'))
Valid_Acc = np.asarray([Valid_Acc]).T

```

```

Pool_Train_Loss = Train_Loss
Pool_Valid_Loss = Valid_Loss
Pool_Train_Acc = Train_Acc
Pool_Valid_Acc = Valid_Acc

# Xtrain ytrain  Xtest  ytest    Xvalid yvalid
print('Evaluating Test Accuracy Without Acquisition')
# X_test  y_test Xtest ytest
score, acc = model_bayes.evaluate(X_test, y_test, verbose=0)

all_accuracy = acc

print('Starting Active Learning in Experiment ', e)

for i in range(3): # acquisition_iterations

    print('POOLING ITERATION', i)

    #take subset of Pool Points for Test Time Dropout
    #and do acquisition from there
    pool_subset = 500
    pool_subset_dropout = np.asarray(random.sample(range(0,X_Pool.
↪shape[0]), pool_subset))
    X_Pool_Dropout = X_Pool[pool_subset_dropout, :, :, :]
    y_Pool_Dropout = y_Pool[pool_subset_dropout]

    score_All = np.zeros(shape=(X_Pool_Dropout.shape[0],↪
↪nb_classes))
    All_Entropy_Dropout = np.zeros(shape=X_Pool_Dropout.shape[0])

    for d in range(dropout_iterations):
        print ('Dropout Iteration', d)
        dropout_score = model_bayes.
↪predict(X_Pool_Dropout, batch_size=batch_size, verbose=1)
        #dropout_score = np_utils.to_categorical(y_Pool,↪
↪nb_classes)

        #computing G_X
        score_All = score_All + dropout_score

        #computing F_X
        dropout_score_log = np.log2(dropout_score+2)

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Entropy_Compute = - np.multiply(dropout_score,
↪dropout_score_log)

Entropy_Per_Dropout = np.sum(Entropy_Compute, axis=1)

All_Entropy_Dropout = All_Entropy_Dropout +
↪Entropy_Per_Dropout

Avg_Pi = np.divide(score_All, dropout_iterations)
Log_Avg_Pi = np.log2(Avg_Pi+2)
Entropy_Avg_Pi = - np.multiply(Avg_Pi, Log_Avg_Pi)
Entropy_Average_Pi = np.sum(Entropy_Avg_Pi, axis=1)

G_X = Entropy_Average_Pi

Average_Entropy = np.divide(All_Entropy_Dropout,
↪dropout_iterations)

F_X = Average_Entropy

U_X = G_X - F_X

# find min index
# a_1d = U_X.flatten()
# x_pool_index = a_1d.argsort()[-Queries:]

a_1d = U_X.flatten()
x_pool_index = a_1d.argsort()[-Queries:][:-1]

#store all the pooled images indexes
x_pool_All = np.append(x_pool_All, x_pool_index)

#saving

Pooled_X = X_Pool_Dropout[x_pool_index, :, :, :]
Pooled_Y = y_Pool_Dropout[x_pool_index]

# delete the random subset used for test time dropout from
↪X_Pool

# Delete the pooled point from this pool set

delete_Pool_X = np.delete(X_Pool, (pool_subset_dropout), axis=0)
delete_Pool_Y = np.delete(y_Pool, (pool_subset_dropout), axis=0)

delete_Pool_X_Dropout = np.delete(X_Pool_Dropout,
↪(x_pool_index), axis=0)

```

```

        delete_Pool_Y_Dropout = np.delete(y_Pool_Dropout,
↪(x_pool_index), axis=0)

        X_Pool = np.concatenate((delete_Pool_X, delete_Pool_X_Dropout),
↪axis=0)
        y_Pool = np.concatenate((delete_Pool_Y, delete_Pool_Y_Dropout),
↪axis=0)

        print('Acquired Points added to training set')
        X_train = np.concatenate((X_train, Pooled_X), axis=0)
        y_train = np.concatenate((y_train, Pooled_Y), axis=0)

        # convert class vectors to binary class matrices

        Y_train = np_utils.to_categorical(y_train, nb_classes)

        #Xtrain, ytrain, ytest,Xtest  Xvalid, yvalid

        #model_bayes.compile(loss=negative_log_likelihood,
↪optimizer='adam',metrics='accuracy')
        hist = model_bayes.fit(X_train, Y_train, batch_size=batch_size,
↪epochs=nb_epoch, verbose=1, validation_data=(X_valid, Y_valid))

        Train_Result_Optimizer = hist.history
        Train_Loss = np.asarray(Train_Result_Optimizer.get('loss'))
        Train_Loss = np.array([Train_Loss]).T
        Valid_Loss = np.asarray(Train_Result_Optimizer.get('val_loss'))
        Valid_Loss = np.asarray([Valid_Loss]).T
        Train_Acc = np.asarray(Train_Result_Optimizer.get('acc'))
        Train_Acc = np.array([Train_Acc]).T
        Valid_Acc = np.asarray(Train_Result_Optimizer.get('val_acc'))
        Valid_Acc = np.asarray([Valid_Acc]).T

        #Accumulate the training and validation/test loss after every
↪pooling iteration - for plotting

        Pool_Valid_Loss = np.append(Pool_Valid_Loss, Valid_Loss, axis=1)
        Pool_Train_Loss = np.append(Pool_Train_Loss, Train_Loss, axis=1)

        print('Evaluate Model Test Accuracy with pooled points')

        score, acc = model_bayes.evaluate(X_test, Y_test , verbose=1)

        print('Test score:', score)

```

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print('Test accuracy:', acc)
all_accuracy = np.append(all_accuracy, acc)

print('Use this trained model with pooled points for Dropout_
→again')

print('Storing Accuracy Values over experiments')
Experiments_All_Accuracy = all_accuracy

```

```

Experiment Number 0
X_train shape: (40, 7, 7, 80)
40 train samples
Training Model Without Acquisitions in Experiment 0
Epoch 1/20
57/57 [=====] - 2s 18ms/step - loss: 12.9481 -
accuracy: 0.2774 - val_loss: 11.9594 - val_accuracy: 0.5358
Epoch 2/20
57/57 [=====] - 1s 18ms/step - loss: 11.6220 -
accuracy: 0.6583 - val_loss: 11.3386 - val_accuracy: 0.7516
Epoch 3/20
57/57 [=====] - 1s 13ms/step - loss: 11.1555 -
accuracy: 0.8173 - val_loss: 11.0267 - val_accuracy: 0.8557
Epoch 4/20
57/57 [=====] - 1s 12ms/step - loss: 10.9267 -
accuracy: 0.8793 - val_loss: 10.8309 - val_accuracy: 0.9002
Epoch 5/20
57/57 [=====] - 1s 15ms/step - loss: 10.7345 -
accuracy: 0.9235 - val_loss: 10.6655 - val_accuracy: 0.9295
Epoch 6/20
57/57 [=====] - 1s 13ms/step - loss: 10.5775 -
accuracy: 0.9501 - val_loss: 10.5577 - val_accuracy: 0.9338
Epoch 7/20
57/57 [=====] - 1s 12ms/step - loss: 10.4641 -
accuracy: 0.9540 - val_loss: 10.4156 - val_accuracy: 0.9447
Epoch 8/20
57/57 [=====] - 1s 13ms/step - loss: 10.3317 -
accuracy: 0.9684 - val_loss: 10.2832 - val_accuracy: 0.9685
Epoch 9/20
57/57 [=====] - 1s 12ms/step - loss: 10.2107 -
accuracy: 0.9732 - val_loss: 10.1703 - val_accuracy: 0.9631
Epoch 10/20
57/57 [=====] - 1s 13ms/step - loss: 10.0811 -
accuracy: 0.9796 - val_loss: 10.0641 - val_accuracy: 0.9707
Epoch 11/20
57/57 [=====] - 1s 13ms/step - loss: 9.9616 - accuracy:
0.9823 - val_loss: 9.9551 - val_accuracy: 0.9642
Epoch 12/20

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57/57 [=====] - 1s 17ms/step - loss: 9.8442 - accuracy: 0.9791 - val\_loss: 9.8061 - val\_accuracy: 0.9685  
Epoch 13/20  
57/57 [=====] - 1s 18ms/step - loss: 9.7258 - accuracy: 0.9819 - val\_loss: 9.6774 - val\_accuracy: 0.9740  
Epoch 14/20  
57/57 [=====] - 1s 15ms/step - loss: 9.5926 - accuracy: 0.9854 - val\_loss: 9.5677 - val\_accuracy: 0.9794  
Epoch 15/20  
57/57 [=====] - 1s 16ms/step - loss: 9.4769 - accuracy: 0.9837 - val\_loss: 9.4739 - val\_accuracy: 0.9588  
Epoch 16/20  
57/57 [=====] - 1s 13ms/step - loss: 9.3460 - accuracy: 0.9872 - val\_loss: 9.3223 - val\_accuracy: 0.9740  
Epoch 17/20  
57/57 [=====] - 1s 12ms/step - loss: 9.2228 - accuracy: 0.9849 - val\_loss: 9.1944 - val\_accuracy: 0.9751  
Epoch 18/20  
57/57 [=====] - 1s 15ms/step - loss: 9.1019 - accuracy: 0.9837 - val\_loss: 9.0459 - val\_accuracy: 0.9826  
Epoch 19/20  
57/57 [=====] - 1s 13ms/step - loss: 8.9745 - accuracy: 0.9851 - val\_loss: 8.9526 - val\_accuracy: 0.9707  
Epoch 20/20  
57/57 [=====] - 1s 13ms/step - loss: 8.8540 - accuracy: 0.9862 - val\_loss: 8.8254 - val\_accuracy: 0.9707  
Evaluating Test Accuracy Without Acquisition  
Starting Active Learning in Experiment 0  
POOLING ITERATION 0  
Dropout Iteration 0  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 1  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 2  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 3  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 4  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 5  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 6  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 7  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 8  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 9



4/4 [=====] - 0s 3ms/step  
Dropout Iteration 10  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 11  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 12  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 13  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 14  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 15  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 16  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 17  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 18  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 19  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 20  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 21  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 22  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 23  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 24  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 25  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 26  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 27  
4/4 [=====] - 0s 7ms/step  
Dropout Iteration 28  
4/4 [=====] - 0s 5ms/step  
Dropout Iteration 29  
4/4 [=====] - 0s 5ms/step  
Dropout Iteration 30  
4/4 [=====] - 0s 5ms/step  
Dropout Iteration 31  
4/4 [=====] - 0s 5ms/step  
Dropout Iteration 32  
4/4 [=====] - 0s 6ms/step  
Dropout Iteration 33

```

4/4 [=====] - 0s 3ms/step
Dropout Iteration 34
4/4 [=====] - 0s 4ms/step
Dropout Iteration 35
4/4 [=====] - 0s 4ms/step
Dropout Iteration 36
4/4 [=====] - 0s 4ms/step
Dropout Iteration 37
4/4 [=====] - 0s 3ms/step
Dropout Iteration 38
4/4 [=====] - 0s 3ms/step
Dropout Iteration 39
4/4 [=====] - 0s 3ms/step
Dropout Iteration 40
4/4 [=====] - 0s 4ms/step
Dropout Iteration 41
4/4 [=====] - 0s 4ms/step
Dropout Iteration 42
4/4 [=====] - 0s 3ms/step
Dropout Iteration 43
4/4 [=====] - 0s 4ms/step
Dropout Iteration 44
4/4 [=====] - 0s 3ms/step
Dropout Iteration 45
4/4 [=====] - 0s 4ms/step
Dropout Iteration 46
4/4 [=====] - 0s 5ms/step
Dropout Iteration 47
4/4 [=====] - 0s 4ms/step
Dropout Iteration 48
4/4 [=====] - 0s 4ms/step
Dropout Iteration 49
4/4 [=====] - 0s 3ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 21ms/step - loss: 8.7290 - accuracy:
0.9856 - val_loss: 8.6654 - val_accuracy: 0.9805
Epoch 2/20
57/57 [=====] - 1s 18ms/step - loss: 8.6131 - accuracy:
0.9855 - val_loss: 8.5689 - val_accuracy: 0.9837
Epoch 3/20
57/57 [=====] - 1s 15ms/step - loss: 8.4908 - accuracy:
0.9859 - val_loss: 8.4808 - val_accuracy: 0.9729
Epoch 4/20
57/57 [=====] - 1s 14ms/step - loss: 8.3727 - accuracy:
0.9866 - val_loss: 8.3348 - val_accuracy: 0.9761
Epoch 5/20
57/57 [=====] - 1s 14ms/step - loss: 8.2779 - accuracy:

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0.9817 - val\_loss: 8.2284 - val\_accuracy: 0.9794  
Epoch 6/20  
57/57 [=====] - 1s 13ms/step - loss: 8.1376 - accuracy:  
0.9861 - val\_loss: 8.1146 - val\_accuracy: 0.9848  
Epoch 7/20  
57/57 [=====] - 1s 13ms/step - loss: 8.0266 - accuracy:  
0.9855 - val\_loss: 7.9683 - val\_accuracy: 0.9837  
Epoch 8/20  
57/57 [=====] - 1s 13ms/step - loss: 7.9232 - accuracy:  
0.9849 - val\_loss: 7.9216 - val\_accuracy: 0.9696  
Epoch 9/20  
57/57 [=====] - 1s 13ms/step - loss: 7.8040 - accuracy:  
0.9869 - val\_loss: 7.7741 - val\_accuracy: 0.9751  
Epoch 10/20  
57/57 [=====] - 1s 15ms/step - loss: 7.6947 - accuracy:  
0.9880 - val\_loss: 7.7065 - val\_accuracy: 0.9718  
Epoch 11/20  
57/57 [=====] - 1s 17ms/step - loss: 7.6045 - accuracy:  
0.9858 - val\_loss: 7.5840 - val\_accuracy: 0.9826  
Epoch 12/20  
57/57 [=====] - 1s 25ms/step - loss: 7.4857 - accuracy:  
0.9872 - val\_loss: 7.4243 - val\_accuracy: 0.9881  
Epoch 13/20  
57/57 [=====] - 2s 32ms/step - loss: 7.3756 - accuracy:  
0.9891 - val\_loss: 7.3575 - val\_accuracy: 0.9751  
Epoch 14/20  
57/57 [=====] - 1s 13ms/step - loss: 7.2909 - accuracy:  
0.9837 - val\_loss: 7.2444 - val\_accuracy: 0.9870  
Epoch 15/20  
57/57 [=====] - 1s 26ms/step - loss: 7.1825 - accuracy:  
0.9880 - val\_loss: 7.1589 - val\_accuracy: 0.9826  
Epoch 16/20  
57/57 [=====] - 1s 21ms/step - loss: 7.0898 - accuracy:  
0.9869 - val\_loss: 7.1233 - val\_accuracy: 0.9631  
Epoch 17/20  
57/57 [=====] - 1s 13ms/step - loss: 6.9870 - accuracy:  
0.9876 - val\_loss: 7.0066 - val\_accuracy: 0.9729  
Epoch 18/20  
57/57 [=====] - 1s 14ms/step - loss: 6.8957 - accuracy:  
0.9859 - val\_loss: 6.9428 - val\_accuracy: 0.9653  
Epoch 19/20  
57/57 [=====] - 1s 15ms/step - loss: 6.8111 - accuracy:  
0.9859 - val\_loss: 6.7492 - val\_accuracy: 0.9870  
Epoch 20/20  
57/57 [=====] - 1s 13ms/step - loss: 6.7138 - accuracy:  
0.9873 - val\_loss: 6.6872 - val\_accuracy: 0.9794  
Evaluate Model Test Accuracy with pooled points  
68/68 [=====] - 0s 2ms/step - loss: 6.7375 - accuracy:

0.9763

Test score: 6.73746395111084

Test accuracy: 0.9763121008872986

Use this trained model with pooled points for Dropout again

POOLING ITERATION 1

Dropout Iteration 0

4/4 [=====] - 0s 4ms/step

Dropout Iteration 1

4/4 [=====] - 0s 4ms/step

Dropout Iteration 2

4/4 [=====] - 0s 4ms/step

Dropout Iteration 3

4/4 [=====] - 0s 3ms/step

Dropout Iteration 4

4/4 [=====] - 0s 4ms/step

Dropout Iteration 5

4/4 [=====] - 0s 5ms/step

Dropout Iteration 6

4/4 [=====] - 0s 4ms/step

Dropout Iteration 7

4/4 [=====] - 0s 3ms/step

Dropout Iteration 8

4/4 [=====] - 0s 4ms/step

Dropout Iteration 9

4/4 [=====] - 0s 3ms/step

Dropout Iteration 10

4/4 [=====] - 0s 4ms/step

Dropout Iteration 11

4/4 [=====] - 0s 3ms/step

Dropout Iteration 12

4/4 [=====] - 0s 4ms/step

Dropout Iteration 13

4/4 [=====] - 0s 3ms/step

Dropout Iteration 14

4/4 [=====] - 0s 3ms/step

Dropout Iteration 15

4/4 [=====] - 0s 3ms/step

Dropout Iteration 16

4/4 [=====] - 0s 3ms/step

Dropout Iteration 17

4/4 [=====] - 0s 10ms/step

Dropout Iteration 18

4/4 [=====] - 0s 5ms/step

Dropout Iteration 19

4/4 [=====] - 0s 6ms/step

Dropout Iteration 20

4/4 [=====] - 0s 6ms/step

Dropout Iteration 21

4/4 [=====] - 0s 15ms/step  
 Dropout Iteration 22  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 23  
 4/4 [=====] - 0s 4ms/step  
 Dropout Iteration 24  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 25  
 4/4 [=====] - 0s 6ms/step  
 Dropout Iteration 26  
 4/4 [=====] - 0s 4ms/step  
 Dropout Iteration 27  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 28  
 4/4 [=====] - 0s 3ms/step  
 Dropout Iteration 29  
 4/4 [=====] - 0s 4ms/step  
 Dropout Iteration 30  
 4/4 [=====] - 0s 4ms/step  
 Dropout Iteration 31  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 32  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 33  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 34  
 4/4 [=====] - 0s 7ms/step  
 Dropout Iteration 35  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 36  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 37  
 4/4 [=====] - 0s 4ms/step  
 Dropout Iteration 38  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 39  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 40  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 41  
 4/4 [=====] - 0s 5ms/step  
 Dropout Iteration 42  
 4/4 [=====] - 0s 13ms/step  
 Dropout Iteration 43  
 4/4 [=====] - 0s 9ms/step  
 Dropout Iteration 44  
 4/4 [=====] - 0s 26ms/step  
 Dropout Iteration 45

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4/4 [=====] - 0s 6ms/step
Dropout Iteration 46
4/4 [=====] - 0s 4ms/step
Dropout Iteration 47
4/4 [=====] - 0s 4ms/step
Dropout Iteration 48
4/4 [=====] - 0s 5ms/step
Dropout Iteration 49
4/4 [=====] - 0s 5ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 24ms/step - loss: 6.6435 - accuracy:
0.9845 - val_loss: 6.6113 - val_accuracy: 0.9783
Epoch 2/20
57/57 [=====] - 1s 18ms/step - loss: 6.5474 - accuracy:
0.9848 - val_loss: 6.5240 - val_accuracy: 0.9794
Epoch 3/20
57/57 [=====] - 1s 19ms/step - loss: 6.4663 - accuracy:
0.9849 - val_loss: 6.4540 - val_accuracy: 0.9740
Epoch 4/20
57/57 [=====] - 1s 17ms/step - loss: 6.3767 - accuracy:
0.9869 - val_loss: 6.3966 - val_accuracy: 0.9761
Epoch 5/20
57/57 [=====] - 1s 17ms/step - loss: 6.2927 - accuracy:
0.9863 - val_loss: 6.2861 - val_accuracy: 0.9805
Epoch 6/20
57/57 [=====] - 1s 19ms/step - loss: 6.2166 - accuracy:
0.9880 - val_loss: 6.1999 - val_accuracy: 0.9816
Epoch 7/20
57/57 [=====] - 1s 19ms/step - loss: 6.1332 - accuracy:
0.9863 - val_loss: 6.1119 - val_accuracy: 0.9826
Epoch 8/20
57/57 [=====] - 1s 16ms/step - loss: 6.0745 - accuracy:
0.9838 - val_loss: 6.0542 - val_accuracy: 0.9816
Epoch 9/20
57/57 [=====] - 1s 16ms/step - loss: 6.0009 - accuracy:
0.9840 - val_loss: 6.0236 - val_accuracy: 0.9707
Epoch 10/20
57/57 [=====] - 1s 25ms/step - loss: 5.9238 - accuracy:
0.9844 - val_loss: 5.8852 - val_accuracy: 0.9870
Epoch 11/20
57/57 [=====] - 2s 42ms/step - loss: 5.8584 - accuracy:
0.9833 - val_loss: 5.8682 - val_accuracy: 0.9707
Epoch 12/20
57/57 [=====] - 1s 16ms/step - loss: 5.7778 - accuracy:
0.9868 - val_loss: 5.7728 - val_accuracy: 0.9816
Epoch 13/20
57/57 [=====] - 1s 21ms/step - loss: 5.7204 - accuracy:

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0.9833 - val_loss: 5.7189 - val_accuracy: 0.9751
Epoch 14/20
57/57 [=====] - 1s 15ms/step - loss: 5.6551 - accuracy:
0.9856 - val_loss: 5.6427 - val_accuracy: 0.9751
Epoch 15/20
57/57 [=====] - 1s 20ms/step - loss: 5.5788 - accuracy:
0.9886 - val_loss: 5.6209 - val_accuracy: 0.9729
Epoch 16/20
57/57 [=====] - 1s 24ms/step - loss: 5.6929 - accuracy:
0.9578 - val_loss: 5.5538 - val_accuracy: 0.9740
Epoch 17/20
57/57 [=====] - 1s 16ms/step - loss: 5.4840 - accuracy:
0.9842 - val_loss: 5.5078 - val_accuracy: 0.9696
Epoch 18/20
57/57 [=====] - 1s 23ms/step - loss: 5.4301 - accuracy:
0.9877 - val_loss: 5.4214 - val_accuracy: 0.9837
Epoch 19/20
57/57 [=====] - 2s 28ms/step - loss: 5.3789 - accuracy:
0.9882 - val_loss: 5.3505 - val_accuracy: 0.9870
Epoch 20/20
57/57 [=====] - 1s 18ms/step - loss: 5.3425 - accuracy:
0.9829 - val_loss: 5.3218 - val_accuracy: 0.9794
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 2ms/step - loss: 5.3013 - accuracy:
0.9847
Test score: 5.3012590408325195
Test accuracy: 0.9846725463867188
Use this trained model with pooled points for Dropout again
POOLING ITERATION 2
Dropout Iteration 0
4/4 [=====] - 0s 4ms/step
Dropout Iteration 1
4/4 [=====] - 0s 4ms/step
Dropout Iteration 2
4/4 [=====] - 0s 4ms/step
Dropout Iteration 3
4/4 [=====] - 0s 4ms/step
Dropout Iteration 4
4/4 [=====] - 0s 4ms/step
Dropout Iteration 5
4/4 [=====] - 0s 4ms/step
Dropout Iteration 6
4/4 [=====] - 0s 8ms/step
Dropout Iteration 7
4/4 [=====] - 0s 4ms/step
Dropout Iteration 8
4/4 [=====] - 0s 4ms/step
Dropout Iteration 9

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4/4 [=====] - 0s 4ms/step
Dropout Iteration 10
4/4 [=====] - 0s 4ms/step
Dropout Iteration 11
4/4 [=====] - 0s 4ms/step
Dropout Iteration 12
4/4 [=====] - 0s 3ms/step
Dropout Iteration 13
4/4 [=====] - 0s 4ms/step
Dropout Iteration 14
4/4 [=====] - 0s 3ms/step
Dropout Iteration 15
4/4 [=====] - 0s 4ms/step
Dropout Iteration 16
4/4 [=====] - 0s 3ms/step
Dropout Iteration 17
4/4 [=====] - 0s 4ms/step
Dropout Iteration 18
4/4 [=====] - 0s 3ms/step
Dropout Iteration 19
4/4 [=====] - 0s 3ms/step
Dropout Iteration 20
4/4 [=====] - 0s 13ms/step
Dropout Iteration 21
4/4 [=====] - 0s 4ms/step
Dropout Iteration 22
4/4 [=====] - 0s 4ms/step
Dropout Iteration 23
4/4 [=====] - 0s 3ms/step
Dropout Iteration 24
4/4 [=====] - 0s 3ms/step
Dropout Iteration 25
4/4 [=====] - 0s 4ms/step
Dropout Iteration 26
4/4 [=====] - 0s 5ms/step
Dropout Iteration 27
4/4 [=====] - 0s 6ms/step
Dropout Iteration 28
4/4 [=====] - 0s 4ms/step
Dropout Iteration 29
4/4 [=====] - 0s 7ms/step
Dropout Iteration 30
4/4 [=====] - 0s 7ms/step
Dropout Iteration 31
4/4 [=====] - 0s 8ms/step
Dropout Iteration 32
4/4 [=====] - 0s 5ms/step
Dropout Iteration 33

```



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4/4 [=====] - 0s 4ms/step
Dropout Iteration 34
4/4 [=====] - 0s 7ms/step
Dropout Iteration 35
4/4 [=====] - 0s 6ms/step
Dropout Iteration 36
4/4 [=====] - 0s 5ms/step
Dropout Iteration 37
4/4 [=====] - 0s 5ms/step
Dropout Iteration 38
4/4 [=====] - 0s 5ms/step
Dropout Iteration 39
4/4 [=====] - 0s 5ms/step
Dropout Iteration 40
4/4 [=====] - 0s 5ms/step
Dropout Iteration 41
4/4 [=====] - 0s 5ms/step
Dropout Iteration 42
4/4 [=====] - 0s 5ms/step
Dropout Iteration 43
4/4 [=====] - 0s 5ms/step
Dropout Iteration 44
4/4 [=====] - 0s 7ms/step
Dropout Iteration 45
4/4 [=====] - 0s 5ms/step
Dropout Iteration 46
4/4 [=====] - 0s 5ms/step
Dropout Iteration 47
4/4 [=====] - 0s 7ms/step
Dropout Iteration 48
4/4 [=====] - 0s 8ms/step
Dropout Iteration 49
4/4 [=====] - 0s 4ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 2s 28ms/step - loss: 5.2777 - accuracy:
0.9847 - val_loss: 5.2991 - val_accuracy: 0.9783
Epoch 2/20
57/57 [=====] - 1s 21ms/step - loss: 5.2371 - accuracy:
0.9861 - val_loss: 5.2439 - val_accuracy: 0.9859
Epoch 3/20
57/57 [=====] - 1s 20ms/step - loss: 5.1775 - accuracy:
0.9880 - val_loss: 5.1946 - val_accuracy: 0.9783
Epoch 4/20
57/57 [=====] - 1s 19ms/step - loss: 5.1305 - accuracy:
0.9884 - val_loss: 5.1243 - val_accuracy: 0.9826
Epoch 5/20
57/57 [=====] - 1s 22ms/step - loss: 5.0744 - accuracy:

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0.9877 - val_loss: 5.0669 - val_accuracy: 0.9881
Epoch 6/20
57/57 [=====] - 1s 20ms/step - loss: 5.0482 - accuracy:
0.9845 - val_loss: 5.0716 - val_accuracy: 0.9805
Epoch 7/20
57/57 [=====] - 1s 15ms/step - loss: 5.0024 - accuracy:
0.9833 - val_loss: 4.9749 - val_accuracy: 0.9870
Epoch 8/20
57/57 [=====] - 1s 21ms/step - loss: 4.9591 - accuracy:
0.9884 - val_loss: 4.9676 - val_accuracy: 0.9794
Epoch 9/20
57/57 [=====] - 1s 17ms/step - loss: 4.9362 - accuracy:
0.9847 - val_loss: 4.9469 - val_accuracy: 0.9783
Epoch 10/20
57/57 [=====] - 1s 16ms/step - loss: 4.9033 - accuracy:
0.9806 - val_loss: 4.8860 - val_accuracy: 0.9805
Epoch 11/20
57/57 [=====] - 1s 18ms/step - loss: 4.8408 - accuracy:
0.9852 - val_loss: 4.8781 - val_accuracy: 0.9772
Epoch 12/20
57/57 [=====] - 1s 19ms/step - loss: 4.7924 - accuracy:
0.9868 - val_loss: 4.7660 - val_accuracy: 0.9848
Epoch 13/20
57/57 [=====] - 1s 16ms/step - loss: 4.7513 - accuracy:
0.9861 - val_loss: 4.7999 - val_accuracy: 0.9740
Epoch 14/20
57/57 [=====] - 1s 15ms/step - loss: 4.7124 - accuracy:
0.9873 - val_loss: 4.7581 - val_accuracy: 0.9805
Epoch 15/20
57/57 [=====] - 1s 17ms/step - loss: 4.6969 - accuracy:
0.9855 - val_loss: 4.6672 - val_accuracy: 0.9859
Epoch 16/20
57/57 [=====] - 1s 15ms/step - loss: 4.6451 - accuracy:
0.9844 - val_loss: 4.6746 - val_accuracy: 0.9783
Epoch 17/20
57/57 [=====] - 1s 17ms/step - loss: 4.5987 - accuracy:
0.9866 - val_loss: 4.5817 - val_accuracy: 0.9837
Epoch 18/20
57/57 [=====] - 1s 16ms/step - loss: 4.5588 - accuracy:
0.9873 - val_loss: 4.5670 - val_accuracy: 0.9837
Epoch 19/20
57/57 [=====] - 1s 15ms/step - loss: 4.5209 - accuracy:
0.9876 - val_loss: 4.5315 - val_accuracy: 0.9859
Epoch 20/20
57/57 [=====] - 1s 16ms/step - loss: 4.4784 - accuracy:
0.9880 - val_loss: 4.5074 - val_accuracy: 0.9859
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 3ms/step - loss: 4.4768 - accuracy:

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0.9847
Test score: 4.47678804397583
Test accuracy: 0.9846725463867188
Use this trained model with pooled points for Dropout again
Storing Accuracy Values over experiments
Experiment Number 1
X_train shape: (40, 7, 7, 80)
40 train samples
Training Model Without Acquisitions in Experiment 1
Epoch 1/20
57/57 [=====] - 1s 17ms/step - loss: 4.4707 - accuracy:
0.9852 - val_loss: 4.4413 - val_accuracy: 0.9751
Epoch 2/20
57/57 [=====] - 1s 15ms/step - loss: 4.4381 - accuracy:
0.9824 - val_loss: 4.4446 - val_accuracy: 0.9805
Epoch 3/20
57/57 [=====] - 1s 17ms/step - loss: 4.3715 - accuracy:
0.9866 - val_loss: 4.4212 - val_accuracy: 0.9772
Epoch 4/20
57/57 [=====] - 1s 15ms/step - loss: 4.3371 - accuracy:
0.9880 - val_loss: 4.3384 - val_accuracy: 0.9794
Epoch 5/20
57/57 [=====] - 1s 15ms/step - loss: 4.2966 - accuracy:
0.9861 - val_loss: 4.3175 - val_accuracy: 0.9805
Epoch 6/20
57/57 [=====] - 1s 16ms/step - loss: 4.2662 - accuracy:
0.9869 - val_loss: 4.2550 - val_accuracy: 0.9794
Epoch 7/20
57/57 [=====] - 1s 15ms/step - loss: 4.2311 - accuracy:
0.9868 - val_loss: 4.2635 - val_accuracy: 0.9837
Epoch 8/20
57/57 [=====] - 1s 14ms/step - loss: 4.2297 - accuracy:
0.9834 - val_loss: 4.2547 - val_accuracy: 0.9664
Epoch 9/20
57/57 [=====] - 1s 13ms/step - loss: 4.1745 - accuracy:
0.9841 - val_loss: 4.1753 - val_accuracy: 0.9826
Epoch 10/20
57/57 [=====] - 1s 13ms/step - loss: 4.1505 - accuracy:
0.9868 - val_loss: 4.2115 - val_accuracy: 0.9664
Epoch 11/20
57/57 [=====] - 1s 14ms/step - loss: 4.1186 - accuracy:
0.9834 - val_loss: 4.1529 - val_accuracy: 0.9772
Epoch 12/20
57/57 [=====] - 1s 13ms/step - loss: 4.1004 - accuracy:
0.9812 - val_loss: 4.0811 - val_accuracy: 0.9848
Epoch 13/20
57/57 [=====] - 1s 14ms/step - loss: 4.0443 - accuracy:
0.9851 - val_loss: 4.0336 - val_accuracy: 0.9881

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Epoch 14/20  
57/57 [=====] - 1s 13ms/step - loss: 4.0464 - accuracy: 0.9803 - val\_loss: 4.0178 - val\_accuracy: 0.9816

Epoch 15/20  
57/57 [=====] - 1s 12ms/step - loss: 4.0279 - accuracy: 0.9787 - val\_loss: 3.9820 - val\_accuracy: 0.9805

Epoch 16/20  
57/57 [=====] - 1s 12ms/step - loss: 3.9529 - accuracy: 0.9863 - val\_loss: 3.9925 - val\_accuracy: 0.9772

Epoch 17/20  
57/57 [=====] - 1s 12ms/step - loss: 3.9367 - accuracy: 0.9858 - val\_loss: 3.9557 - val\_accuracy: 0.9783

Epoch 18/20  
57/57 [=====] - 1s 12ms/step - loss: 3.9233 - accuracy: 0.9844 - val\_loss: 3.9321 - val\_accuracy: 0.9761

Epoch 19/20  
57/57 [=====] - 1s 12ms/step - loss: 3.8763 - accuracy: 0.9851 - val\_loss: 3.8599 - val\_accuracy: 0.9826

Epoch 20/20  
57/57 [=====] - 1s 13ms/step - loss: 3.8407 - accuracy: 0.9870 - val\_loss: 3.8488 - val\_accuracy: 0.9805

Evaluating Test Accuracy Without Acquisition

Starting Active Learning in Experiment 1

POOLING ITERATION 0

Dropout Iteration 0  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 1  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 2  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 3  
4/4 [=====] - 0s 3ms/step

Dropout Iteration 4  
4/4 [=====] - 0s 5ms/step

Dropout Iteration 5  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 6  
4/4 [=====] - 0s 3ms/step

Dropout Iteration 7  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 8  
4/4 [=====] - 0s 6ms/step

Dropout Iteration 9  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 10  
4/4 [=====] - 0s 5ms/step

Dropout Iteration 11  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 12  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 13  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 14  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 15  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 16  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 17  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 18  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 19  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 20  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 21  
4/4 [=====] - 0s 7ms/step  
Dropout Iteration 22  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 23  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 24  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 25  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 26  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 27  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 28  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 29  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 30  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 31  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 32  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 33  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 34  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 35  
4/4 [=====] - 0s 4ms/step

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Dropout Iteration 36
4/4 [=====] - 0s 4ms/step
Dropout Iteration 37
4/4 [=====] - 0s 3ms/step
Dropout Iteration 38
4/4 [=====] - 0s 3ms/step
Dropout Iteration 39
4/4 [=====] - 0s 3ms/step
Dropout Iteration 40
4/4 [=====] - 0s 3ms/step
Dropout Iteration 41
4/4 [=====] - 0s 4ms/step
Dropout Iteration 42
4/4 [=====] - 0s 3ms/step
Dropout Iteration 43
4/4 [=====] - 0s 3ms/step
Dropout Iteration 44
4/4 [=====] - 0s 3ms/step
Dropout Iteration 45
4/4 [=====] - 0s 3ms/step
Dropout Iteration 46
4/4 [=====] - 0s 3ms/step
Dropout Iteration 47
4/4 [=====] - 0s 3ms/step
Dropout Iteration 48
4/4 [=====] - 0s 4ms/step
Dropout Iteration 49
4/4 [=====] - 0s 3ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 12ms/step - loss: 3.8244 - accuracy:
0.9841 - val_loss: 3.8367 - val_accuracy: 0.9848
Epoch 2/20
57/57 [=====] - 1s 11ms/step - loss: 3.8242 - accuracy:
0.9806 - val_loss: 3.8301 - val_accuracy: 0.9783
Epoch 3/20
57/57 [=====] - 1s 13ms/step - loss: 3.7529 - accuracy:
0.9870 - val_loss: 3.7790 - val_accuracy: 0.9848
Epoch 4/20
57/57 [=====] - 1s 13ms/step - loss: 3.7285 - accuracy:
0.9869 - val_loss: 3.7826 - val_accuracy: 0.9837
Epoch 5/20
57/57 [=====] - 1s 14ms/step - loss: 3.7302 - accuracy:
0.9824 - val_loss: 3.7382 - val_accuracy: 0.9805
Epoch 6/20
57/57 [=====] - 1s 13ms/step - loss: 3.7222 - accuracy:
0.9806 - val_loss: 3.7909 - val_accuracy: 0.9675
Epoch 7/20

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57/57 [=====] - 1s 14ms/step - loss: 3.7461 - accuracy:
0.9710 - val_loss: 3.7006 - val_accuracy: 0.9805
Epoch 8/20
57/57 [=====] - 1s 13ms/step - loss: 3.6358 - accuracy:
0.9856 - val_loss: 3.6631 - val_accuracy: 0.9816
Epoch 9/20
57/57 [=====] - 1s 11ms/step - loss: 3.6548 - accuracy:
0.9848 - val_loss: 3.6474 - val_accuracy: 0.9848
Epoch 10/20
57/57 [=====] - 1s 14ms/step - loss: 3.6234 - accuracy:
0.9822 - val_loss: 3.6316 - val_accuracy: 0.9794
Epoch 11/20
57/57 [=====] - 1s 13ms/step - loss: 3.6115 - accuracy:
0.9791 - val_loss: 3.5475 - val_accuracy: 0.9816
Epoch 12/20
57/57 [=====] - 1s 10ms/step - loss: 3.5777 - accuracy:
0.9824 - val_loss: 3.5863 - val_accuracy: 0.9729
Epoch 13/20
57/57 [=====] - 1s 12ms/step - loss: 3.5467 - accuracy:
0.9783 - val_loss: 3.5459 - val_accuracy: 0.9783
Epoch 14/20
57/57 [=====] - 1s 13ms/step - loss: 3.5109 - accuracy:
0.9840 - val_loss: 3.4628 - val_accuracy: 0.9913
Epoch 15/20
57/57 [=====] - 1s 12ms/step - loss: 3.5017 - accuracy:
0.9851 - val_loss: 3.4705 - val_accuracy: 0.9826
Epoch 16/20
57/57 [=====] - 1s 10ms/step - loss: 3.4727 - accuracy:
0.9844 - val_loss: 3.4241 - val_accuracy: 0.9902
Epoch 17/20
57/57 [=====] - 1s 12ms/step - loss: 3.4362 - accuracy:
0.9863 - val_loss: 3.4652 - val_accuracy: 0.9805
Epoch 18/20
57/57 [=====] - 1s 12ms/step - loss: 3.4299 - accuracy:
0.9854 - val_loss: 3.4390 - val_accuracy: 0.9816
Epoch 19/20
57/57 [=====] - 1s 12ms/step - loss: 3.3874 - accuracy:
0.9879 - val_loss: 3.3888 - val_accuracy: 0.9837
Epoch 20/20
57/57 [=====] - 1s 12ms/step - loss: 3.3636 - accuracy:
0.9869 - val_loss: 3.3759 - val_accuracy: 0.9859
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 2ms/step - loss: 3.4274 - accuracy:
0.9824
Test score: 3.427382230758667
Test accuracy: 0.98235023021698
Use this trained model with pooled points for Dropout again
POOLING ITERATION 1

```

Dropout Iteration 0  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 1  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 2  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 3  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 4  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 5  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 6  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 7  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 8  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 9  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 10  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 11  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 12  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 13  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 14  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 15  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 16  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 17  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 18  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 19  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 20  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 21  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 22  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 23  
4/4 [=====] - 0s 3ms/step



Dropout Iteration 24  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 25  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 26  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 27  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 28  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 29  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 30  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 31  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 32  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 33  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 34  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 35  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 36  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 37  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 38  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 39  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 40  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 41  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 42  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 43  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 44  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 45  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 46  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 47  
4/4 [=====] - 0s 3ms/step

Dropout Iteration 48  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 49  
4/4 [=====] - 0s 3ms/step  
Acquired Points added to training set  
Epoch 1/20  
57/57 [=====] - 1s 9ms/step - loss: 3.3389 - accuracy:  
0.9875 - val\_loss: 3.3546 - val\_accuracy: 0.9826  
Epoch 2/20  
57/57 [=====] - 1s 9ms/step - loss: 3.3215 - accuracy:  
0.9879 - val\_loss: 3.3540 - val\_accuracy: 0.9751  
Epoch 3/20  
57/57 [=====] - 1s 9ms/step - loss: 3.3257 - accuracy:  
0.9837 - val\_loss: 3.3249 - val\_accuracy: 0.9783  
Epoch 4/20  
57/57 [=====] - 1s 9ms/step - loss: 3.3019 - accuracy:  
0.9842 - val\_loss: 3.2829 - val\_accuracy: 0.9805  
Epoch 5/20  
57/57 [=====] - 1s 9ms/step - loss: 3.2620 - accuracy:  
0.9859 - val\_loss: 3.2666 - val\_accuracy: 0.9794  
Epoch 6/20  
57/57 [=====] - 1s 9ms/step - loss: 3.2520 - accuracy:  
0.9837 - val\_loss: 3.2394 - val\_accuracy: 0.9761  
Epoch 7/20  
57/57 [=====] - 1s 9ms/step - loss: 3.2287 - accuracy:  
0.9844 - val\_loss: 3.2340 - val\_accuracy: 0.9783  
Epoch 8/20  
57/57 [=====] - 1s 9ms/step - loss: 3.2196 - accuracy:  
0.9833 - val\_loss: 3.2371 - val\_accuracy: 0.9848  
Epoch 9/20  
57/57 [=====] - 1s 9ms/step - loss: 3.1877 - accuracy:  
0.9859 - val\_loss: 3.2757 - val\_accuracy: 0.9805  
Epoch 10/20  
57/57 [=====] - 1s 9ms/step - loss: 3.1719 - accuracy:  
0.9865 - val\_loss: 3.1931 - val\_accuracy: 0.9805  
Epoch 11/20  
57/57 [=====] - 1s 9ms/step - loss: 3.1363 - accuracy:  
0.9861 - val\_loss: 3.1673 - val\_accuracy: 0.9859  
Epoch 12/20  
57/57 [=====] - 1s 9ms/step - loss: 3.1337 - accuracy:  
0.9837 - val\_loss: 3.0734 - val\_accuracy: 0.9946  
Epoch 13/20  
57/57 [=====] - 1s 9ms/step - loss: 3.0877 - accuracy:  
0.9880 - val\_loss: 3.1689 - val\_accuracy: 0.9794  
Epoch 14/20  
57/57 [=====] - 1s 10ms/step - loss: 3.0814 - accuracy:  
0.9869 - val\_loss: 3.1179 - val\_accuracy: 0.9772  
Epoch 15/20

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57/57 [=====] - 1s 9ms/step - loss: 3.0835 - accuracy:
0.9813 - val_loss: 3.0649 - val_accuracy: 0.9783
Epoch 16/20
57/57 [=====] - 1s 10ms/step - loss: 3.0479 - accuracy:
0.9847 - val_loss: 3.0309 - val_accuracy: 0.9892
Epoch 17/20
57/57 [=====] - 1s 11ms/step - loss: 3.0235 - accuracy:
0.9856 - val_loss: 3.0687 - val_accuracy: 0.9805
Epoch 18/20
57/57 [=====] - 1s 10ms/step - loss: 3.0245 - accuracy:
0.9816 - val_loss: 3.1469 - val_accuracy: 0.9729
Epoch 19/20
57/57 [=====] - 1s 9ms/step - loss: 3.0327 - accuracy:
0.9787 - val_loss: 2.9933 - val_accuracy: 0.9740
Epoch 20/20
57/57 [=====] - 1s 9ms/step - loss: 2.9857 - accuracy:
0.9819 - val_loss: 3.0102 - val_accuracy: 0.9837
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 2ms/step - loss: 2.9754 - accuracy:
0.9814
Test score: 2.975449562072754
Test accuracy: 0.9814212918281555
Use this trained model with pooled points for Dropout again
POOLING ITERATION 2
Dropout Iteration 0
4/4 [=====] - 0s 3ms/step
Dropout Iteration 1
4/4 [=====] - 0s 3ms/step
Dropout Iteration 2
4/4 [=====] - 0s 3ms/step
Dropout Iteration 3
4/4 [=====] - 0s 2ms/step
Dropout Iteration 4
4/4 [=====] - 0s 3ms/step
Dropout Iteration 5
4/4 [=====] - 0s 2ms/step
Dropout Iteration 6
4/4 [=====] - 0s 3ms/step
Dropout Iteration 7
4/4 [=====] - 0s 3ms/step
Dropout Iteration 8
4/4 [=====] - 0s 3ms/step
Dropout Iteration 9
4/4 [=====] - 0s 2ms/step
Dropout Iteration 10
4/4 [=====] - 0s 3ms/step
Dropout Iteration 11
4/4 [=====] - 0s 2ms/step

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Dropout Iteration 12  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 13  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 14  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 15  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 16  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 17  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 18  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 19  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 20  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 21  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 22  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 23  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 24  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 25  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 26  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 27  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 28  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 29  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 30  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 31  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 32  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 33  
4/4 [=====] - 0s 2ms/step  
Dropout Iteration 34  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 35  
4/4 [=====] - 0s 3ms/step

```

Dropout Iteration 36
4/4 [=====] - 0s 3ms/step
Dropout Iteration 37
4/4 [=====] - 0s 3ms/step
Dropout Iteration 38
4/4 [=====] - 0s 3ms/step
Dropout Iteration 39
4/4 [=====] - 0s 3ms/step
Dropout Iteration 40
4/4 [=====] - 0s 3ms/step
Dropout Iteration 41
4/4 [=====] - 0s 3ms/step
Dropout Iteration 42
4/4 [=====] - 0s 3ms/step
Dropout Iteration 43
4/4 [=====] - 0s 3ms/step
Dropout Iteration 44
4/4 [=====] - 0s 3ms/step
Dropout Iteration 45
4/4 [=====] - 0s 5ms/step
Dropout Iteration 46
4/4 [=====] - 0s 2ms/step
Dropout Iteration 47
4/4 [=====] - 0s 3ms/step
Dropout Iteration 48
4/4 [=====] - 0s 3ms/step
Dropout Iteration 49
4/4 [=====] - 0s 3ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 10ms/step - loss: 2.9460 - accuracy:
0.9869 - val_loss: 2.9561 - val_accuracy: 0.9848
Epoch 2/20
57/57 [=====] - 1s 10ms/step - loss: 2.9226 - accuracy:
0.9852 - val_loss: 3.0633 - val_accuracy: 0.9837
Epoch 3/20
57/57 [=====] - 1s 9ms/step - loss: 2.9287 - accuracy:
0.9858 - val_loss: 2.9495 - val_accuracy: 0.9837
Epoch 4/20
57/57 [=====] - 1s 10ms/step - loss: 2.9400 - accuracy:
0.9784 - val_loss: 2.9647 - val_accuracy: 0.9772
Epoch 5/20
57/57 [=====] - 1s 9ms/step - loss: 2.9028 - accuracy:
0.9791 - val_loss: 2.9284 - val_accuracy: 0.9751
Epoch 6/20
57/57 [=====] - 1s 10ms/step - loss: 2.8599 - accuracy:
0.9858 - val_loss: 2.8875 - val_accuracy: 0.9761
Epoch 7/20

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```

57/57 [=====] - 1s 9ms/step - loss: 2.8439 - accuracy:
0.9859 - val_loss: 2.8768 - val_accuracy: 0.9696
Epoch 8/20
57/57 [=====] - 1s 9ms/step - loss: 2.8739 - accuracy:
0.9744 - val_loss: 2.9005 - val_accuracy: 0.9718
Epoch 9/20
57/57 [=====] - 1s 9ms/step - loss: 2.8337 - accuracy:
0.9817 - val_loss: 2.8877 - val_accuracy: 0.9761
Epoch 10/20
57/57 [=====] - 1s 9ms/step - loss: 2.8071 - accuracy:
0.9838 - val_loss: 2.8728 - val_accuracy: 0.9772
Epoch 11/20
57/57 [=====] - 1s 10ms/step - loss: 2.7883 - accuracy:
0.9845 - val_loss: 2.7621 - val_accuracy: 0.9837
Epoch 12/20
57/57 [=====] - 1s 9ms/step - loss: 2.7719 - accuracy:
0.9816 - val_loss: 2.8818 - val_accuracy: 0.9816
Epoch 13/20
57/57 [=====] - 1s 10ms/step - loss: 2.7911 - accuracy:
0.9809 - val_loss: 2.7767 - val_accuracy: 0.9783
Epoch 14/20
57/57 [=====] - 1s 9ms/step - loss: 2.7477 - accuracy:
0.9820 - val_loss: 2.7456 - val_accuracy: 0.9772
Epoch 15/20
57/57 [=====] - 1s 9ms/step - loss: 2.7500 - accuracy:
0.9803 - val_loss: 2.7111 - val_accuracy: 0.9892
Epoch 16/20
57/57 [=====] - 1s 10ms/step - loss: 2.7127 - accuracy:
0.9796 - val_loss: 2.8352 - val_accuracy: 0.9642
Epoch 17/20
57/57 [=====] - 1s 10ms/step - loss: 2.6922 - accuracy:
0.9823 - val_loss: 2.6853 - val_accuracy: 0.9761
Epoch 18/20
57/57 [=====] - 1s 9ms/step - loss: 2.6679 - accuracy:
0.9823 - val_loss: 2.7742 - val_accuracy: 0.9816
Epoch 19/20
57/57 [=====] - 1s 9ms/step - loss: 2.6495 - accuracy:
0.9868 - val_loss: 2.7399 - val_accuracy: 0.9620
Epoch 20/20
57/57 [=====] - 1s 9ms/step - loss: 2.6692 - accuracy:
0.9794 - val_loss: 2.7060 - val_accuracy: 0.9740
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 2ms/step - loss: 2.6236 - accuracy:
0.9856
Test score: 2.623633861541748
Test accuracy: 0.9856014847755432
Use this trained model with pooled points for Dropout again
Storing Accuracy Values over experiments

```

Experiment Number 2  
 X\_train shape: (40, 7, 7, 80)  
 40 train samples  
 Training Model Without Acquisitions in Experiment 2  
 Epoch 1/20  
 57/57 [=====] - 1s 12ms/step - loss: 2.6033 - accuracy:  
 0.9841 - val\_loss: 2.6369 - val\_accuracy: 0.9816  
 Epoch 2/20  
 57/57 [=====] - 1s 11ms/step - loss: 2.7407 - accuracy:  
 0.9702 - val\_loss: 2.6528 - val\_accuracy: 0.9761  
 Epoch 3/20  
 57/57 [=====] - 1s 10ms/step - loss: 2.6621 - accuracy:  
 0.9721 - val\_loss: 2.7111 - val\_accuracy: 0.9783  
 Epoch 4/20  
 57/57 [=====] - 1s 12ms/step - loss: 2.6014 - accuracy:  
 0.9819 - val\_loss: 2.7151 - val\_accuracy: 0.9751  
 Epoch 5/20  
 57/57 [=====] - 1s 14ms/step - loss: 2.5543 - accuracy:  
 0.9854 - val\_loss: 2.5886 - val\_accuracy: 0.9805  
 Epoch 6/20  
 57/57 [=====] - 1s 14ms/step - loss: 2.5991 - accuracy:  
 0.9796 - val\_loss: 2.6262 - val\_accuracy: 0.9707  
 Epoch 7/20  
 57/57 [=====] - 1s 12ms/step - loss: 2.5329 - accuracy:  
 0.9806 - val\_loss: 2.5770 - val\_accuracy: 0.9772  
 Epoch 8/20  
 57/57 [=====] - 1s 12ms/step - loss: 2.5380 - accuracy:  
 0.9794 - val\_loss: 2.5043 - val\_accuracy: 0.9902  
 Epoch 9/20  
 57/57 [=====] - 1s 14ms/step - loss: 2.5133 - accuracy:  
 0.9829 - val\_loss: 2.5577 - val\_accuracy: 0.9761  
 Epoch 10/20  
 57/57 [=====] - 1s 13ms/step - loss: 2.5414 - accuracy:  
 0.9703 - val\_loss: 2.5364 - val\_accuracy: 0.9794  
 Epoch 11/20  
 57/57 [=====] - 1s 13ms/step - loss: 2.5089 - accuracy:  
 0.9806 - val\_loss: 2.4738 - val\_accuracy: 0.9848  
 Epoch 12/20  
 57/57 [=====] - 1s 13ms/step - loss: 2.4613 - accuracy:  
 0.9862 - val\_loss: 2.4761 - val\_accuracy: 0.9837  
 Epoch 13/20  
 57/57 [=====] - 1s 13ms/step - loss: 2.4689 - accuracy:  
 0.9830 - val\_loss: 2.5522 - val\_accuracy: 0.9740  
 Epoch 14/20  
 57/57 [=====] - 1s 13ms/step - loss: 2.4599 - accuracy:  
 0.9794 - val\_loss: 2.5334 - val\_accuracy: 0.9740  
 Epoch 15/20  
 57/57 [=====] - 1s 14ms/step - loss: 2.4461 - accuracy:

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0.9824 - val_loss: 2.5042 - val_accuracy: 0.9816
Epoch 16/20
57/57 [=====] - 1s 16ms/step - loss: 2.4278 - accuracy:
0.9847 - val_loss: 2.4618 - val_accuracy: 0.9696
Epoch 17/20
57/57 [=====] - 1s 16ms/step - loss: 2.4276 - accuracy:
0.9822 - val_loss: 2.4236 - val_accuracy: 0.9751
Epoch 18/20
57/57 [=====] - 1s 13ms/step - loss: 2.4217 - accuracy:
0.9824 - val_loss: 2.4711 - val_accuracy: 0.9729
Epoch 19/20
57/57 [=====] - 1s 13ms/step - loss: 2.4085 - accuracy:
0.9792 - val_loss: 2.3882 - val_accuracy: 0.9816
Epoch 20/20
57/57 [=====] - 1s 15ms/step - loss: 2.3813 - accuracy:
0.9803 - val_loss: 2.5484 - val_accuracy: 0.9794
Evaluating Test Accuracy Without Acquisition
Starting Active Learning in Experiment 2
POOLING ITERATION 0
Dropout Iteration 0
4/4 [=====] - 0s 4ms/step
Dropout Iteration 1
4/4 [=====] - 0s 3ms/step
Dropout Iteration 2
4/4 [=====] - 0s 4ms/step
Dropout Iteration 3
4/4 [=====] - 0s 3ms/step
Dropout Iteration 4
4/4 [=====] - 0s 4ms/step
Dropout Iteration 5
4/4 [=====] - 0s 3ms/step
Dropout Iteration 6
4/4 [=====] - 0s 3ms/step
Dropout Iteration 7
4/4 [=====] - 0s 3ms/step
Dropout Iteration 8
4/4 [=====] - 0s 4ms/step
Dropout Iteration 9
4/4 [=====] - 0s 3ms/step
Dropout Iteration 10
4/4 [=====] - 0s 3ms/step
Dropout Iteration 11
4/4 [=====] - 0s 3ms/step
Dropout Iteration 12
4/4 [=====] - 0s 3ms/step
Dropout Iteration 13
4/4 [=====] - 0s 3ms/step
Dropout Iteration 14

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4/4 [=====] - 0s 4ms/step  
Dropout Iteration 15  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 16  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 17  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 18  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 19  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 20  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 21  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 22  
4/4 [=====] - 0s 3ms/step  
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4/4 [=====] - 0s 3ms/step  
Dropout Iteration 24  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 25  
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Dropout Iteration 26  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 27  
4/4 [=====] - 0s 3ms/step  
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Dropout Iteration 29  
4/4 [=====] - 0s 4ms/step  
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Dropout Iteration 31  
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Dropout Iteration 32  
4/4 [=====] - 0s 4ms/step  
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4/4 [=====] - 0s 4ms/step  
Dropout Iteration 34  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 35  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 36  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 37  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 38

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4/4 [=====] - 0s 4ms/step
Dropout Iteration 39
4/4 [=====] - 0s 4ms/step
Dropout Iteration 40
4/4 [=====] - 0s 3ms/step
Dropout Iteration 41
4/4 [=====] - 0s 3ms/step
Dropout Iteration 42
4/4 [=====] - 0s 3ms/step
Dropout Iteration 43
4/4 [=====] - 0s 4ms/step
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4/4 [=====] - 0s 3ms/step
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Dropout Iteration 47
4/4 [=====] - 0s 3ms/step
Dropout Iteration 48
4/4 [=====] - 0s 4ms/step
Dropout Iteration 49
4/4 [=====] - 0s 3ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 13ms/step - loss: 2.3859 - accuracy:
0.9820 - val_loss: 2.4821 - val_accuracy: 0.9783
Epoch 2/20
57/57 [=====] - 1s 13ms/step - loss: 2.3601 - accuracy:
0.9816 - val_loss: 2.4373 - val_accuracy: 0.9805
Epoch 3/20
57/57 [=====] - 1s 12ms/step - loss: 2.3668 - accuracy:
0.9820 - val_loss: 2.4364 - val_accuracy: 0.9675
Epoch 4/20
57/57 [=====] - 1s 13ms/step - loss: 2.3428 - accuracy:
0.9826 - val_loss: 2.3794 - val_accuracy: 0.9751
Epoch 5/20
57/57 [=====] - 1s 13ms/step - loss: 2.3037 - accuracy:
0.9834 - val_loss: 2.3186 - val_accuracy: 0.9881
Epoch 6/20
57/57 [=====] - 1s 13ms/step - loss: 2.3352 - accuracy:
0.9802 - val_loss: 2.3695 - val_accuracy: 0.9783
Epoch 7/20
57/57 [=====] - 1s 12ms/step - loss: 2.2975 - accuracy:
0.9809 - val_loss: 2.3484 - val_accuracy: 0.9751
Epoch 8/20
57/57 [=====] - 1s 12ms/step - loss: 2.2776 - accuracy:
0.9802 - val_loss: 2.3723 - val_accuracy: 0.9707

```

Epoch 9/20  
57/57 [=====] - 1s 12ms/step - loss: 2.2799 - accuracy: 0.9815 - val\_loss: 2.3267 - val\_accuracy: 0.9729

Epoch 10/20  
57/57 [=====] - 1s 12ms/step - loss: 2.2764 - accuracy: 0.9802 - val\_loss: 2.2635 - val\_accuracy: 0.9794

Epoch 11/20  
57/57 [=====] - 1s 12ms/step - loss: 2.2589 - accuracy: 0.9769 - val\_loss: 2.2644 - val\_accuracy: 0.9794

Epoch 12/20  
57/57 [=====] - 1s 13ms/step - loss: 2.2634 - accuracy: 0.9796 - val\_loss: 2.3856 - val\_accuracy: 0.9707

Epoch 13/20  
57/57 [=====] - 1s 12ms/step - loss: 2.2018 - accuracy: 0.9854 - val\_loss: 2.2328 - val\_accuracy: 0.9848

Epoch 14/20  
57/57 [=====] - 1s 13ms/step - loss: 2.2422 - accuracy: 0.9795 - val\_loss: 2.2627 - val\_accuracy: 0.9794

Epoch 15/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1948 - accuracy: 0.9855 - val\_loss: 2.3134 - val\_accuracy: 0.9783

Epoch 16/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1940 - accuracy: 0.9791 - val\_loss: 2.1941 - val\_accuracy: 0.9816

Epoch 17/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1836 - accuracy: 0.9820 - val\_loss: 2.3102 - val\_accuracy: 0.9783

Epoch 18/20  
57/57 [=====] - 1s 13ms/step - loss: 2.1914 - accuracy: 0.9813 - val\_loss: 2.2735 - val\_accuracy: 0.9555

Epoch 19/20  
57/57 [=====] - 1s 12ms/step - loss: 2.2076 - accuracy: 0.9778 - val\_loss: 2.2456 - val\_accuracy: 0.9772

Epoch 20/20  
57/57 [=====] - 1s 13ms/step - loss: 2.1862 - accuracy: 0.9778 - val\_loss: 2.1907 - val\_accuracy: 0.9718

Evaluate Model Test Accuracy with pooled points  
68/68 [=====] - 0s 2ms/step - loss: 2.1713 - accuracy: 0.9749

Test score: 2.1712892055511475  
Test accuracy: 0.9749187231063843

Use this trained model with pooled points for Dropout again

POOLING ITERATION 1  
Dropout Iteration 0  
4/4 [=====] - 0s 3ms/step

Dropout Iteration 1  
4/4 [=====] - 0s 4ms/step

Dropout Iteration 2

4/4 [=====] - 0s 4ms/step  
Dropout Iteration 3  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 4  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 5  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 6  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 7  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 8  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 9  
4/4 [=====] - 0s 4ms/step  
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Dropout Iteration 13  
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Dropout Iteration 47  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 48  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 49  
4/4 [=====] - 0s 4ms/step  
Acquired Points added to training set

Epoch 1/20  
57/57 [=====] - 1s 14ms/step - loss: 2.1347 - accuracy:  
0.9815 - val\_loss: 2.2045 - val\_accuracy: 0.9718  
Epoch 2/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1826 - accuracy:  
0.9752 - val\_loss: 2.1503 - val\_accuracy: 0.9751  
Epoch 3/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1326 - accuracy:  
0.9792 - val\_loss: 2.1252 - val\_accuracy: 0.9783  
Epoch 4/20  
57/57 [=====] - 1s 12ms/step - loss: 2.0981 - accuracy:  
0.9788 - val\_loss: 2.1668 - val\_accuracy: 0.9718  
Epoch 5/20  
57/57 [=====] - 1s 13ms/step - loss: 2.1073 - accuracy:  
0.9753 - val\_loss: 2.0837 - val\_accuracy: 0.9696  
Epoch 6/20  
57/57 [=====] - 1s 12ms/step - loss: 2.1079 - accuracy:  
0.9806 - val\_loss: 2.0733 - val\_accuracy: 0.9859  
Epoch 7/20  
57/57 [=====] - 1s 12ms/step - loss: 2.0947 - accuracy:  
0.9774 - val\_loss: 2.1717 - val\_accuracy: 0.9729  
Epoch 8/20  
57/57 [=====] - 1s 13ms/step - loss: 2.1074 - accuracy:  
0.9749 - val\_loss: 2.0567 - val\_accuracy: 0.9696  
Epoch 9/20  
57/57 [=====] - 1s 13ms/step - loss: 2.1192 - accuracy:  
0.9762 - val\_loss: 2.0679 - val\_accuracy: 0.9740  
Epoch 10/20  
57/57 [=====] - 1s 13ms/step - loss: 2.0643 - accuracy:  
0.9806 - val\_loss: 2.2039 - val\_accuracy: 0.9772  
Epoch 11/20  
57/57 [=====] - 1s 12ms/step - loss: 2.0476 - accuracy:  
0.9813 - val\_loss: 2.1564 - val\_accuracy: 0.9794  
Epoch 12/20  
57/57 [=====] - 1s 13ms/step - loss: 2.0319 - accuracy:  
0.9784 - val\_loss: 2.0928 - val\_accuracy: 0.9805  
Epoch 13/20  
57/57 [=====] - 1s 12ms/step - loss: 2.0213 - accuracy:  
0.9792 - val\_loss: 2.3484 - val\_accuracy: 0.9761  
Epoch 14/20  
57/57 [=====] - 1s 12ms/step - loss: 2.0155 - accuracy:  
0.9796 - val\_loss: 2.0214 - val\_accuracy: 0.9761  
Epoch 15/20  
57/57 [=====] - 1s 13ms/step - loss: 2.0113 - accuracy:  
0.9816 - val\_loss: 2.0134 - val\_accuracy: 0.9848  
Epoch 16/20  
57/57 [=====] - 1s 12ms/step - loss: 1.9897 - accuracy:  
0.9844 - val\_loss: 2.1146 - val\_accuracy: 0.9740

Epoch 17/20  
57/57 [=====] - 1s 12ms/step - loss: 1.9784 - accuracy: 0.9833 - val\_loss: 1.9608 - val\_accuracy: 0.9848  
Epoch 18/20  
57/57 [=====] - 1s 13ms/step - loss: 1.9804 - accuracy: 0.9790 - val\_loss: 1.9933 - val\_accuracy: 0.9805  
Epoch 19/20  
57/57 [=====] - 1s 12ms/step - loss: 1.9443 - accuracy: 0.9841 - val\_loss: 1.9740 - val\_accuracy: 0.9794  
Epoch 20/20  
57/57 [=====] - 1s 12ms/step - loss: 1.9659 - accuracy: 0.9798 - val\_loss: 1.9687 - val\_accuracy: 0.9772  
Evaluate Model Test Accuracy with pooled points  
68/68 [=====] - 0s 2ms/step - loss: 1.9715 - accuracy: 0.9782  
Test score: 1.9714635610580444  
Test accuracy: 0.9781699776649475  
Use this trained model with pooled points for Dropout again  
POOLING ITERATION 2  
Dropout Iteration 0  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 1  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 2  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 3  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 4  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 5  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 6  
4/4 [=====] - 0s 4ms/step  
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4/4 [=====] - 0s 3ms/step  
Dropout Iteration 8  
4/4 [=====] - 0s 4ms/step  
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Dropout Iteration 12  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 13  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 14

4/4 [=====] - 0s 3ms/step  
Dropout Iteration 15  
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Dropout Iteration 36  
4/4 [=====] - 0s 4ms/step  
Dropout Iteration 37  
4/4 [=====] - 0s 3ms/step  
Dropout Iteration 38



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4/4 [=====] - 0s 4ms/step
Dropout Iteration 39
4/4 [=====] - 0s 3ms/step
Dropout Iteration 40
4/4 [=====] - 0s 3ms/step
Dropout Iteration 41
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Dropout Iteration 48
4/4 [=====] - 0s 3ms/step
Dropout Iteration 49
4/4 [=====] - 0s 3ms/step
Acquired Points added to training set
Epoch 1/20
57/57 [=====] - 1s 13ms/step - loss: 2.0499 - accuracy:
0.9745 - val_loss: 2.0217 - val_accuracy: 0.9718
Epoch 2/20
57/57 [=====] - 1s 12ms/step - loss: 1.9752 - accuracy:
0.9756 - val_loss: 1.9899 - val_accuracy: 0.9620
Epoch 3/20
57/57 [=====] - 1s 12ms/step - loss: 1.9337 - accuracy:
0.9790 - val_loss: 2.0001 - val_accuracy: 0.9696
Epoch 4/20
57/57 [=====] - 1s 13ms/step - loss: 1.9327 - accuracy:
0.9799 - val_loss: 1.9565 - val_accuracy: 0.9740
Epoch 5/20
57/57 [=====] - 1s 12ms/step - loss: 1.9059 - accuracy:
0.9829 - val_loss: 1.8899 - val_accuracy: 0.9870
Epoch 6/20
57/57 [=====] - 1s 12ms/step - loss: 1.9248 - accuracy:
0.9834 - val_loss: 1.9545 - val_accuracy: 0.9707
Epoch 7/20
57/57 [=====] - 1s 13ms/step - loss: 1.8938 - accuracy:
0.9826 - val_loss: 1.9087 - val_accuracy: 0.9826
Epoch 8/20
57/57 [=====] - 1s 12ms/step - loss: 1.8891 - accuracy:
0.9796 - val_loss: 1.8815 - val_accuracy: 0.9805

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Epoch 9/20
57/57 [=====] - 1s 12ms/step - loss: 1.9008 - accuracy:
0.9777 - val_loss: 2.0556 - val_accuracy: 0.9653
Epoch 10/20
57/57 [=====] - 1s 13ms/step - loss: 1.8731 - accuracy:
0.9813 - val_loss: 1.9135 - val_accuracy: 0.9729
Epoch 11/20
57/57 [=====] - 1s 12ms/step - loss: 1.8945 - accuracy:
0.9785 - val_loss: 1.9662 - val_accuracy: 0.9610
Epoch 12/20
57/57 [=====] - 1s 12ms/step - loss: 1.9023 - accuracy:
0.9741 - val_loss: 1.8885 - val_accuracy: 0.9783
Epoch 13/20
57/57 [=====] - 1s 12ms/step - loss: 1.8525 - accuracy:
0.9834 - val_loss: 1.8499 - val_accuracy: 0.9794
Epoch 14/20
57/57 [=====] - 1s 12ms/step - loss: 1.8556 - accuracy:
0.9785 - val_loss: 1.9183 - val_accuracy: 0.9631
Epoch 15/20
57/57 [=====] - 1s 12ms/step - loss: 2.0090 - accuracy:
0.9603 - val_loss: 1.9392 - val_accuracy: 0.9761
Epoch 16/20
57/57 [=====] - 1s 13ms/step - loss: 1.8534 - accuracy:
0.9727 - val_loss: 1.9033 - val_accuracy: 0.9783
Epoch 17/20
57/57 [=====] - 1s 12ms/step - loss: 1.8198 - accuracy:
0.9792 - val_loss: 1.8648 - val_accuracy: 0.9783
Epoch 18/20
57/57 [=====] - 1s 13ms/step - loss: 1.8138 - accuracy:
0.9810 - val_loss: 1.8288 - val_accuracy: 0.9794
Epoch 19/20
57/57 [=====] - 1s 13ms/step - loss: 1.8299 - accuracy:
0.9815 - val_loss: 1.8943 - val_accuracy: 0.9707
Epoch 20/20
57/57 [=====] - 1s 12ms/step - loss: 1.8350 - accuracy:
0.9776 - val_loss: 1.8630 - val_accuracy: 0.9751
Evaluate Model Test Accuracy with pooled points
68/68 [=====] - 0s 2ms/step - loss: 1.7971 - accuracy:
0.9754
Test score: 1.7970836162567139
Test accuracy: 0.9753831624984741
Use this trained model with pooled points for Dropout again
Storing Accuracy Values over experiments

```

```

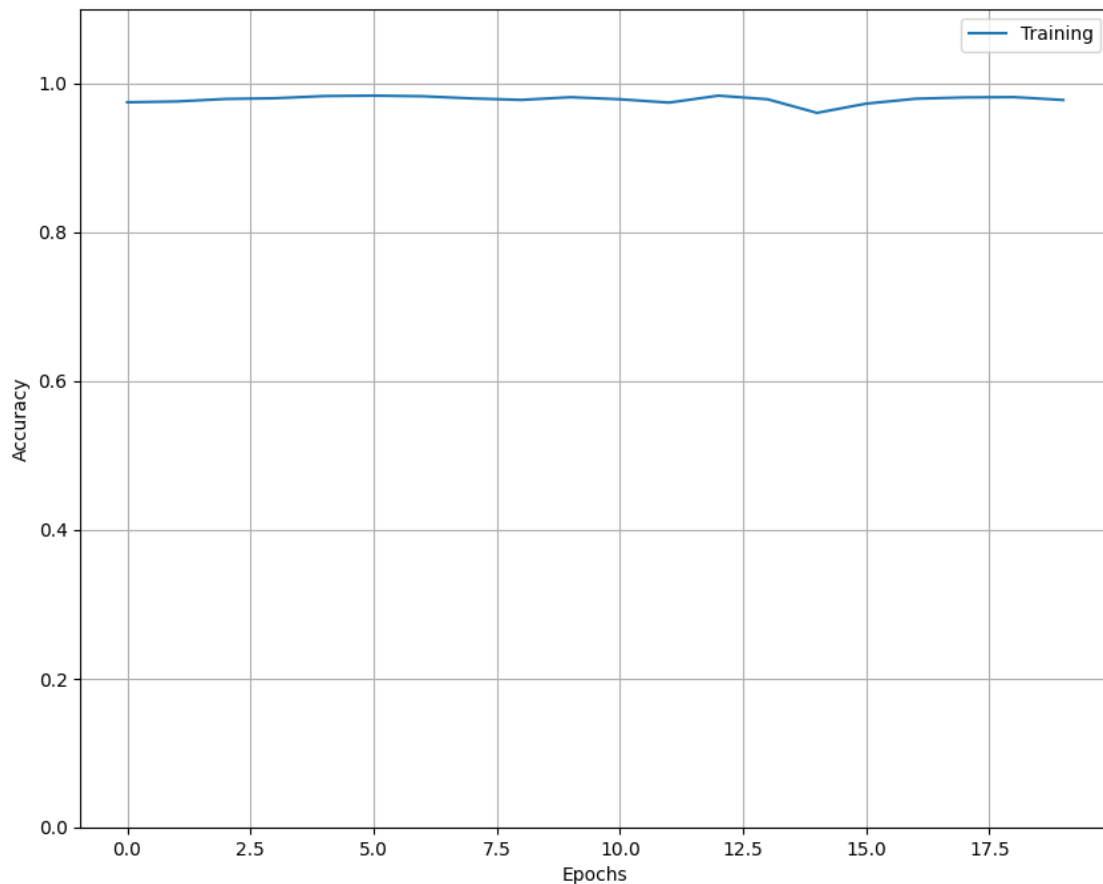
[ ]: Average_Accuracy = np.average(Experiments_All_Accuracy)
    print(Average_Accuracy)

```

```
#np.save('/home/ri258/Documents/Project/MPhil_Thesis_Cluster_Experiments/  
→ConvNets/Cluster_Experiments/Final_Experiments/Results/  
→'+ 'Dropout_Bald_Q10_N1000_Average_Accuracy'+'.npy', Average_Accuracy)
```

0.9775893986225128

```
[ ]: # Plot accuracy  
plt.figure(figsize=(10,8))  
plt.ylim(0,1.1)  
plt.grid()  
plt.plot(hist.history['accuracy'])  
plt.ylabel('Accuracy')  
plt.xlabel('Epochs')  
plt.legend(['Training', 'Validation'])  
plt.savefig("acc_curve.pdf")  
plt.show()
```



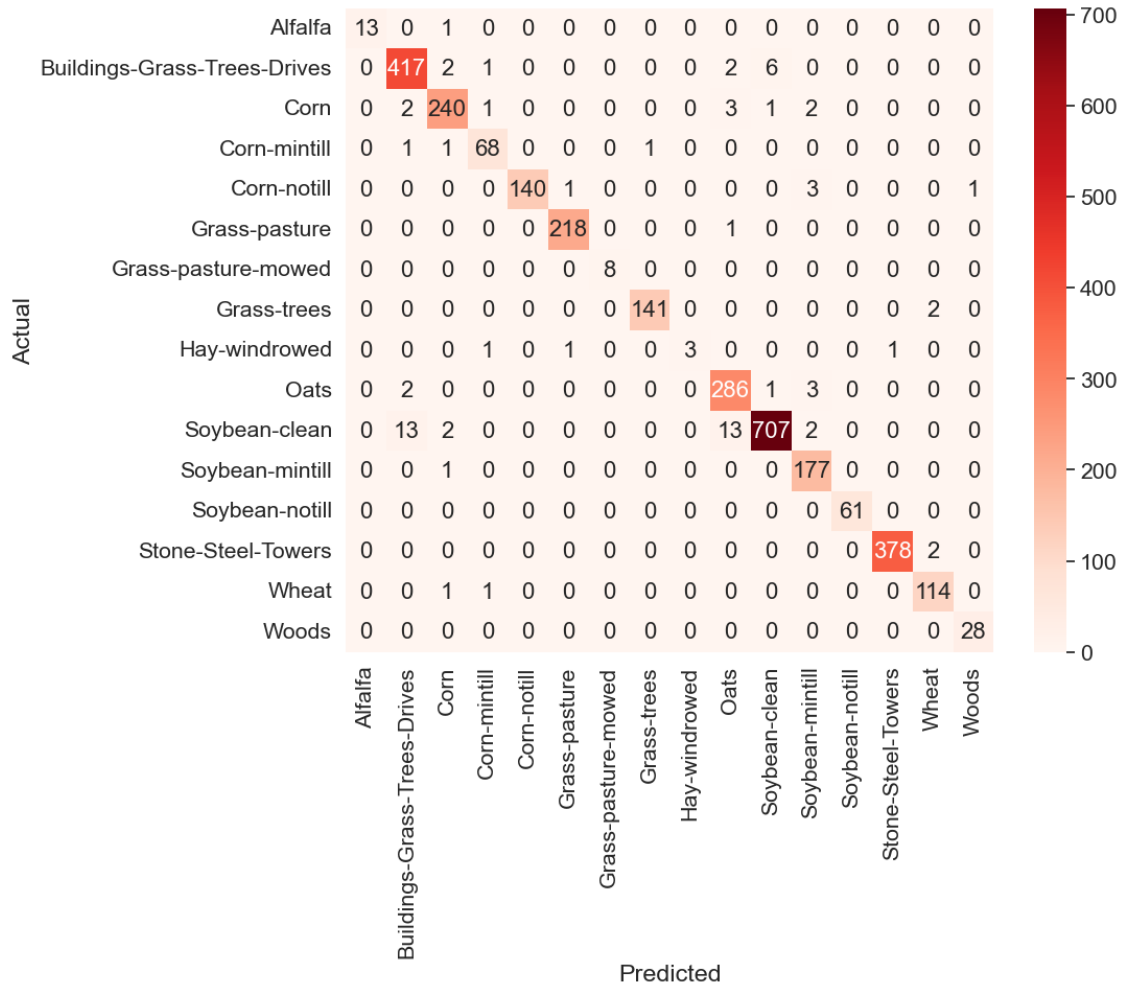
```
[ ]: # 16 classes
```

```
names = ['Alfalfa', 'Corn-notill', 'Corn-mintill', 'Corn', 'Grass-pasture',
        ↪ 'Grass-trees',
        'Grass-pasture-mowed', 'Hay-windrowed', 'Oats', 'Soybean-notill',
        ↪ 'Soybean-mintill',
        'Soybean-clean', 'Wheat', 'Woods', 'Buildings-Grass-Trees-Drives',
        ↪ 'Stone-Steel-Towers']
```

```
[ ]: from sklearn.metrics import confusion_matrix, accuracy_score,
        ↪ classification_report, cohen_kappa_score
import pandas as pd
import numpy as np
import seaborn as sn
# confusion_matrix
Y_pred = model_bayes.predict(Xtest0)
y_pred = np.argmax(Y_pred, axis=1)

confusion = confusion_matrix(ytest0, y_pred)
df_cm = pd.DataFrame(confusion, columns=np.unique(names), index = np.
        ↪ unique(names))
df_cm.index.name = 'Actual'
df_cm.columns.name = 'Predicted'
plt.figure(figsize = (10,8))
sn.set(font_scale=1.4) #for label size
sn.heatmap(df_cm, cmap="Reds", annot=True, annot_kws={"size": 16}, fmt='d')
plt.savefig('cmap.png', dpi=300)
```

97/97 [=====] - 0s 2ms/step



```
[ ]: # average_acc
from operator import itemgetter
def AA_andEachClassAccuracy(confusion_matrix):
    counter = confusion_matrix.shape[0]
    list_diag = np.diag(confusion_matrix)
    list_raw_sum = np.sum(confusion_matrix, axis=1)
    each_acc = np.nan_to_num((list_diag/ list_raw_sum))
    average_acc = np.mean(each_acc)
    return each_acc, average_acc
```

```
[ ]: # average_acc

each_acc, aa = AA_andEachClassAccuracy(confusion)
print("accuracy for each:")
print (each_acc)
```

```
print("OA accuracy:")
print(aa)
```

accuracy for each:

```
[0.92857143 0.97429907 0.96385542 0.95774648 0.96551724 0.99543379
 1.          0.98601399 0.5          0.97945205 0.95929444 0.99438202
 1.          0.99473684 0.98275862 1.          ]
```

OA accuracy:

0.9488788368042085

```
[ ]: # classification_report
print(classification_report(ytest0, y_pred, target_names = names, digits = 3))
```

	precision	recall	f1-score	support
Alfalfa	1.000	0.929	0.963	14
Corn-notill	0.959	0.974	0.966	428
Corn-mintill	0.968	0.964	0.966	249
Corn	0.944	0.958	0.951	71
Grass-pasture	1.000	0.966	0.982	145
Grass-trees	0.991	0.995	0.993	219
Grass-pasture-mowed	1.000	1.000	1.000	8
Hay-windrowed	0.993	0.986	0.989	143
Oats	1.000	0.500	0.667	6
Soybean-notill	0.938	0.979	0.958	292
Soybean-mintill	0.989	0.959	0.974	737
Soybean-clean	0.947	0.994	0.970	178
Wheat	1.000	1.000	1.000	61
Woods	0.997	0.995	0.996	380
Buildings-Grass-Trees-Drives	0.966	0.983	0.974	116
Stone-Steel-Towers	0.966	1.000	0.982	28
accuracy			0.975	3075
macro avg	0.979	0.949	0.958	3075
weighted avg	0.976	0.975	0.975	3075