This exercise will focus on training a neural network classifier for the MNIST dataset. The goal of this exercise is understanding the effect of overparameterization and dropout on the training performance and test accuracy.

1. NAME: NITYASH GAUTAM

2. SID: 862395403

3. UCR MAIL ID: ngaut006@ucr.edu

## **IMPORTING ESSENTIALS**

```
import torch
print(torch.__version__)
import torchvision
import torchvision.datasets as datasets
import torchvision.utils as utils
from torch.utils.data import DataLoader, Subset
import torchvision.transforms as transforms
from PIL import Image
import matplotlib.pyplot as plt
import numpy as np
import time
import torch.nn as nn
import pandas as pd
import torch.optim as optim
import random
import torch.nn.functional as F
import random
import time
1.13.1
MAIN ASSIGNMENT TASKS BEGIN
print(torch.cuda.is available())
device = torch.device("cuda" if torch.cuda.is available() else "cpu")
True
GETTING THE DATASET
# Converting From PIL to tensors and Normalize
transform = transforms.Compose([transforms.ToTensor(),
                                transforms.Normalize((0.1307,),
(0.3081,)),
                                transforms.Lambda(lambda x: x.view(-
```

```
1))
                                 ])
# Download MNIST dataset
trainset = datasets.MNIST(root='./data', train=True, download=True,
transform = transform)
testset = datasets.MNIST(root='./data', train=False, download=True,
transform = transform)
print('train set Length', len(trainset))
print('test_set Length', len(testset))
train set Length 60000
test set Length 10000
classes in train = [0]*10
train subset indices = []
for idx, (_, target) in enumerate(trainset):
  if classes_in_train[target] < 1000:</pre>
    train subset indices.append(idx)
    classes_in_train[target] += 1
train data = Subset(trainset, train subset indices)
train loader = DataLoader(train data, batch size=32, shuffle=True)
classes in test = [0]*10
test subset indices = []
for idx, (_, target) in enumerate(testset):
  if classes in test[target] < 1000:</pre>
    test subset indices.append(idx)
    classes in test[target] += 1
test data = Subset(testset, test subset indices)
test loader = DataLoader(test data, batch size=100, shuffle=True)
TASK 1 - (3 PTS)
```

In this exercise, we will play with two variables which is the network width k and dropout rate p. Your tasks are as follows.

Setup your code so that you can run multiple MNIST models for varying choices of k and p automatically, Specifically, you need two for loops (one for k and one for p) and within the loop, you call PyTorch/TensorFlow

def general\_shallow\_net(trainloader,testloader, optimizer\_type, K, P,
epochs):

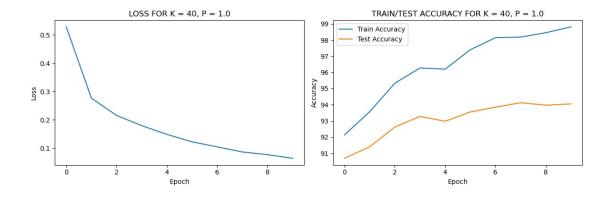
```
print('!'*22, f'RUNNING FOR K={K}, P={P}', '!'*22)
```

```
print('/'*69)
  batch size = 1000
  learning rate = 0.001
  # Lists to store loss, accuracies, and corresponding k and p values
  loss arr = []
  train acc = []
  test acc = []
  k list = []
  p_list = []
  # Define model architecture
 model = nn.Sequential(
      nn.Linear(784, K),
      nn.ReLU(),
      nn.Dropout(1-P),
      nn.Linear(K, 10)
  )
 model = model.to(device)
 # Create data loaders
 # trainloader = torch.utils.data.DataLoader(train data,
batch size=batch size, shuffle=True)
  # testloader = torch.utils.data.DataLoader(test data,
batch size=batch size, shuffle=False)
  # Define loss criterion
  criterion = nn.CrossEntropyLoss()
 # Conditionally setting the Optimizer
  if optimizer_type == 'SGD':
    optimizer = optim.SGD(model.parameters(), lr=learning rate,
momentum=0.9)
  elif optimizer type == 'ADAM':
    optimizer = torch.optim.Adam(model.parameters(), lr=learning rate)
    print('!'*20, 'ERROR', '!'*20)
    print('Optimizer can only be "SGD" or "ADAM"')
  # Training loop
  for epoch in range(epochs):
      running loss = 0.0
      for i, data in enumerate(trainloader, 0):
          inputs, labels = data
          # Flatten the inputs
          inputs = inputs.view(inputs.size(0), -1)
```

```
inputs = inputs.to(device)
    labels = labels.to(device)
    # Zero the parameter gradients
    optimizer.zero_grad()
    # Forward + backward + optimize
    outputs = model(inputs)
    loss = criterion(outputs, labels)
    loss.backward()
    optimizer.step()
    # Print statistics
    running loss += loss.item()
# Evaluate the model on train and test data
correct train = 0
total_train = 0
with torch.no grad():
    for data in trainloader:
        images, labels = data
        images = images.to(device)
        labels = labels.to(device)
        images = images.view(images.size(0), -1)
        outputs = model(images)
        , predicted = torch.max(outputs.data, 1)
        total train += labels.size(0)
        correct train += (predicted == labels).sum().item()
train accuracy = 100 * correct train / total train
correct test = 0
total test = 0
with torch.no grad():
    for data in testloader:
        images, labels = data
        images = images.to(device)
        labels = labels.to(device)
        images = images.view(images.size(0), -1)
        outputs = model(images)
        , predicted = torch.max(outputs.data, 1)
        total test += labels.size(0)
        correct_test += (predicted == labels).sum().item()
test_accuracy = 100 * correct_test / total_test
```

```
if((epoch+1) % 10 == 0):
        print(f'Epoch [{epoch + 1}/{epochs}], Loss: {running_loss /
100:.3f}, Train Accuracy: {train accuracy:.2f}, Test Accuracy:
{test accuracy:.2f}')
      if(train accuracy > 99.5):
        print(f'OBTAINED AN ACCURACY OF {train accuracy}% AT EPOCH NO.
\{epoch\}\ FOR\ K = \{K\}\ AND\ P = \{1-(P)\}'\}
      # Store the loss, train accuracy, test accuracy, k, and p values
      loss_arr.append(running_loss / len(trainloader))
      train acc.append(train accuracy)
      test acc.append(test accuracy)
      k list.append(K)
      p list.append(P)
  print()
 print('-'*22, f'FOR K: {K}, P: {P} THE PLOTS ARE AS FOLLOWS',
'-'*22)
  print()
 # Plot loss and train/test accuracy as a grid of size 1x2
  fig, axs = plt.subplots(1, 2, figsize=(12, 4))
 # Plot loss
  axs[0].plot(loss arr)
  axs[0].set title(f'LOSS FOR K = {K}, P = {P}')
  axs[0].set xlabel('Epoch')
  axs[0].set ylabel('Loss')
  # Plot train/test accuracy
  axs[1].plot(train acc, label='Train Accuracy')
  axs[1].plot(test_acc, label='Test Accuracy')
  axs[1].set\ title(f'TRAIN/TEST\ ACCURACY\ FOR\ K = \{K\},\ P = \{P\}')
  axs[1].set_xlabel('Epoch')
  axs[1].set ylabel('Accuracy')
  axs[1].legend()
  plt.tight layout()
  plt.show()
  print()
  return loss arr, train acc, test acc
%%time
```

```
k=40
p=1.0
t1 loss, t1 train acc, t1 test acc = general shallow net(train loader,
test_loader, 'ADAM', k, p, 10)
t1 mean loss = t1 loss[-1]
t1 max train acc = t1 train acc[-1]
t1 max test acc = t1 test acc[-1]
print(f'Loss for k=\{k\}, p=\{p\}: {t1 mean loss:.3f}')
print(f'Train Accuracy for k={k}, p={p}: {t1 max train acc:.2f}%')
print(f'Test Accuracy for k={k}, p={p}: {t1_max_test_acc:.2f}%')
print()
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=40, P=1.0 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!
Epoch [10/10], Loss: 0.200, Train Accuracy: 98.81, Test Accuracy:
94.05
  ----- FOR K: 40, P: 1.0 THE PLOTS ARE AS FOLLOWS
```



Loss for k=40, p=1.0: 0.064 Train Accuracy for k=40, p=1.0: 98.81% Test Accuracy for k=40, p=1.0: 94.05%

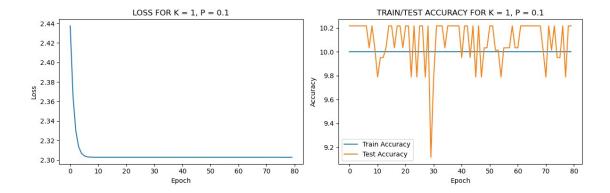
Wall time: 1min 48s

## TASK 2 - (7 PTS)

Pick the width grid K = [1, 5, 10, 20, 40] and dropout grid P = [0.1, 0.5, 1.0]. Run MNIST models over these grids with Adam optimizer for 80 epochs. Store the test/train accuracy and loss.

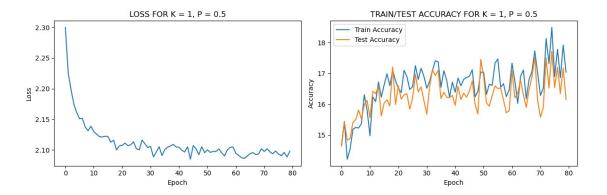
```
main k = [1,5,10,20,40]
main p = [0.1, 0.5, 1.0]
num epochs = 80
task2 results = []
%%time
for k in main k:
  for p in main p:
   t21 loss, t21 train acc, t21 test acc =
general shallow net(train loader, test loader, 'ADAM', k, p,
num epochs)
   t21 \text{ mean loss} = t21 \text{ loss}[-1]
   t21 \text{ max train acc} = t21 \text{ train acc}[-1]
   t21 \max test acc = t21 test acc[-1]
   print(f'Loss for k=\{k\}, p=\{p\}: {t21 mean loss:.3f}')
   print(f'Train Accuracy for k={k}, p={p}: {t21 max train acc:.2f}
%')
   print(f'Test Accuracy for k={k}, p={p}: {t21 max test acc:.2f}%')
   t2 total summary = {'Hidden Units': k, 'Dropout Factor': p, 'Mean
Loss': t21 mean loss, 'Train Accuracy': t21 max train acc, 'Test
Accuracy': t21 max test acc}
   task2 results.append(t2 total summary)
   print()
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=1. P=0.1 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!
Epoch [10/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.03
Epoch [20/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.03
Epoch [30/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy: 9.12
Epoch [40/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.22
Epoch [50/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.03
Epoch [60/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.03
Epoch [70/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
Epoch [80/80], Loss: 7.208, Train Accuracy: 10.00, Test Accuracy:
10.22
----- FOR K: 1, P: 0.1 THE PLOTS ARE AS FOLLOWS
```

------



Loss for k=1, p=0.1: 2.303 Train Accuracy for k=1, p=0.1: 10.00% Test Accuracy for k=1, p=0.1: 10.22%

```
!!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=1, P=0.5 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!
Epoch [10/80], Loss: 6.694, Train Accuracy: 15.81, Test Accuracy:
16.13
Epoch [20/80], Loss: 6.593, Train Accuracy: 16.76, Test Accuracy:
15.99
Epoch [30/80], Loss: 6.585, Train Accuracy: 16.90, Test Accuracy:
16.11
Epoch [40/80], Loss: 6.588, Train Accuracy: 16.72, Test Accuracy:
16.29
Epoch [50/80], Loss: 6.559, Train Accuracy: 17.04, Test Accuracy:
17.45
Epoch [60/80], Loss: 6.588, Train Accuracy: 16.47, Test Accuracy:
15.79
Epoch [70/80], Loss: 6.579, Train Accuracy: 17.02, Test Accuracy:
Epoch [80/80], Loss: 6.567, Train Accuracy: 17.04, Test Accuracy:
16.16
                ---- FOR K: 1, P: 0.5 THE PLOTS ARE AS FOLLOWS
```



Loss for k=1, p=0.5: 2.098 Train Accuracy for k=1, p=0.5: 17.04% Test Accuracy for k=1, p=0.5: 16.16%

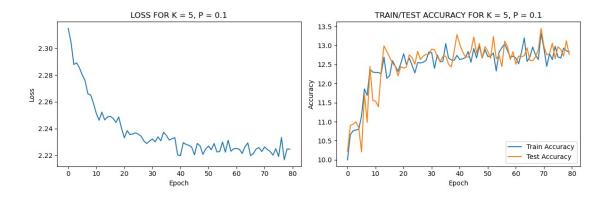
Epoch [10/80], Loss: 5.620, Train Accuracy: 24.46, Test Accuracy: 24.70 Epoch [20/80], Loss: 5.260, Train Accuracy: 29.06, Test Accuracy: 27.38 Epoch [30/80], Loss: 5.155, Train Accuracy: 31.84, Test Accuracy: 30.87 Epoch [40/80], Loss: 5.092, Train Accuracy: 30.12, Test Accuracy: 28.10 Epoch [50/80], Loss: 5.048, Train Accuracy: 34.24, Test Accuracy: 33.09 Epoch [60/80], Loss: 5.007, Train Accuracy: 32.79, Test Accuracy: 31.11 Epoch [70/80], Loss: 4.951, Train Accuracy: 39.29, Test Accuracy: 37.70 Epoch [80/80], Loss: 4.894, Train Accuracy: 39.43, Test Accuracy: 37.38 -- FOR K: 1, P: 1.0 THE PLOTS ARE AS FOLLOWS

LOSS FOR K = 1, P = 1.0TRAIN/TEST ACCURACY FOR K = 1, P = 1.0 Train Accuracy 2.1 Test Accuracy 35 2.0 Accuracy 52 1.9 Loss 1.8 1.7 20 1.6 70 80 60 80 Epoch Epoch

Loss for k=1, p=1.0: 1.563 Train Accuracy for k=1, p=1.0: 39.43% Test Accuracy for k=1, p=1.0: 37.38%

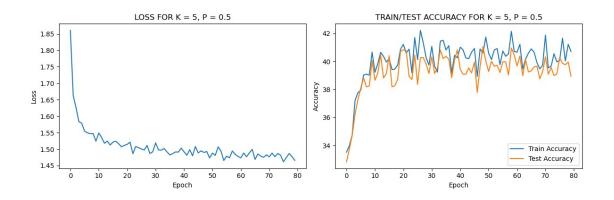
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=5, P=0.1 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 7.071, Train Accuracy: 12.30, Test Accuracy: 11.56 Epoch [20/80], Loss: 7.012, Train Accuracy: 12.53, Test Accuracy: Epoch [30/80], Loss: 6.982, Train Accuracy: 12.83, Test Accuracy: 12.76 Epoch [40/80], Loss: 6.949, Train Accuracy: 12.74, Test Accuracy: 13.28 Epoch [50/80], Loss: 6.964, Train Accuracy: 12.89, Test Accuracy: 12.97 Epoch [60/80], Loss: 6.964, Train Accuracy: 12.72, Test Accuracy: 12.83 Epoch [70/80], Loss: 6.957, Train Accuracy: 13.32, Test Accuracy: 13.45 Epoch [80/80], Loss: 6.963, Train Accuracy: 12.83, Test Accuracy: 12.76

----- FOR K: 5, P: 0.1 THE PLOTS ARE AS FOLLOWS



Loss for k=5, p=0.1: 2.225 Train Accuracy for k=5, p=0.1: 12.83% Test Accuracy for k=5, p=0.1: 12.76%

```
40.71
Epoch [30/80], Loss: 4.670, Train Accuracy: 39.75, Test Accuracy:
39.14
Epoch [40/80], Loss: 4.705, Train Accuracy: 40.26, Test Accuracy:
40.83
Epoch [50/80], Loss: 4.612, Train Accuracy: 41.74, Test Accuracy:
40.04
Epoch [60/80], Loss: 4.626, Train Accuracy: 40.72, Test Accuracy:
40.48
Epoch [70/80], Loss: 4.641, Train Accuracy: 39.73, Test Accuracy:
39.27
Epoch [80/80], Loss: 4.588, Train Accuracy: 40.71, Test Accuracy:
38.94
              ----- FOR K: 5, P: 0.5 THE PLOTS ARE AS FOLLOWS
```

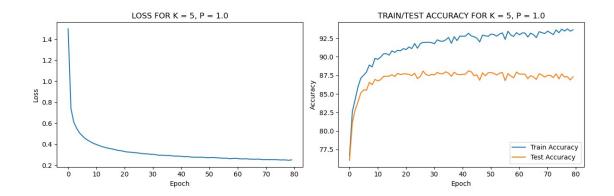


Loss for k=5, p=0.5: 1.466 Train Accuracy for k=5, p=0.5: 40.71% Test Accuracy for k=5, p=0.5: 38.94%

```
Epoch [10/80], Loss: 1.274, Train Accuracy: 89.80, Test Accuracy:
86.95
Epoch [20/80], Loss: 1.053, Train Accuracy: 91.10, Test Accuracy:
87.71
Epoch [30/80], Loss: 0.952, Train Accuracy: 91.92, Test Accuracy:
87.63
Epoch [40/80], Loss: 0.898, Train Accuracy: 92.80, Test Accuracy:
87.58
Epoch [50/80], Loss: 0.850, Train Accuracy: 92.77, Test Accuracy:
87.85
Epoch [60/80], Loss: 0.833, Train Accuracy: 93.27, Test Accuracy:
87.94
Epoch [70/80], Loss: 0.793, Train Accuracy: 93.14, Test Accuracy:
```

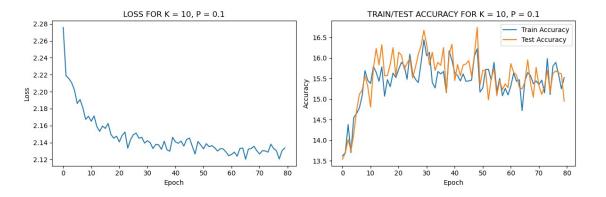
87.26 Epoch [80/80], Loss: 0.786, Train Accuracy: 93.64, Test Accuracy: 87.32

----- FOR K: 5, P: 1.0 THE PLOTS ARE AS FOLLOWS



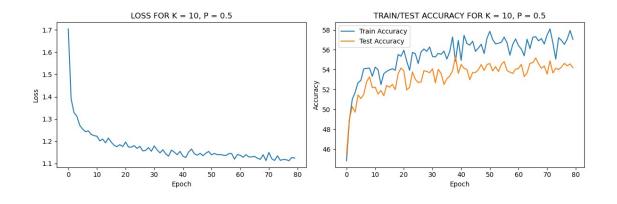
Loss for k=5, p=1.0: 0.251 Train Accuracy for k=5, p=1.0: 93.64% Test Accuracy for k=5, p=1.0: 87.32%

!!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=10, P=0.1 !!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 6.795, Train Accuracy: 15.45, Test Accuracy: 15.31 Epoch [20/80], Loss: 6.722, Train Accuracy: 15.52, Test Accuracy: 15.55 Epoch [30/80], Loss: 6.696, Train Accuracy: 16.44, Test Accuracy: 16.67 Epoch [40/80], Loss: 6.718, Train Accuracy: 15.94, Test Accuracy: 16.33 Epoch [50/80], Loss: 6.690, Train Accuracy: 15.17, Test Accuracy: 15.34 Epoch [60/80], Loss: 6.649, Train Accuracy: 15.10, Test Accuracy: 15.28 Epoch [70/80], Loss: 6.668, Train Accuracy: 15.44, Test Accuracy: 15.77 Epoch [80/80], Loss: 6.678, Train Accuracy: 15.52, Test Accuracy: 14.95 ----- FOR K: 10, P: 0.1 THE PLOTS ARE AS FOLLOWS



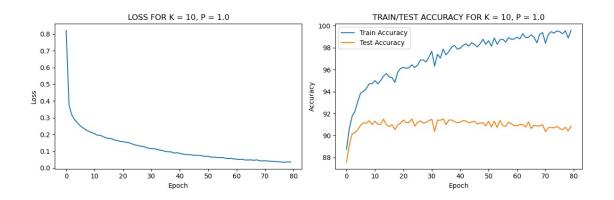
Loss for k=10, p=0.1: 2.134 Train Accuracy for k=10, p=0.1: 15.52% Test Accuracy for k=10, p=0.1: 14.95%

```
Epoch [10/80], Loss: 3.836, Train Accuracy: 53.33, Test Accuracy:
Epoch [20/80], Loss: 3.682, Train Accuracy: 55.36, Test Accuracy:
54.17
Epoch [30/80], Loss: 3.617, Train Accuracy: 56.29, Test Accuracy:
53.69
Epoch [40/80], Loss: 3.615, Train Accuracy: 56.95, Test Accuracy:
53.63
Epoch [50/80], Loss: 3.614, Train Accuracy: 57.16, Test Accuracy:
54.52
Epoch [60/80], Loss: 3.571, Train Accuracy: 57.10, Test Accuracy:
54.06
Epoch [70/80], Loss: 3.486, Train Accuracy: 56.59, Test Accuracy:
Epoch [80/80], Loss: 3.520, Train Accuracy: 57.06, Test Accuracy:
54.20
                  FOR K: 10, P: 0.5 THE PLOTS ARE AS FOLLOWS
```



Loss for k=10, p=0.5: 1.125 Train Accuracy for k=10, p=0.5: 57.06% Test Accuracy for k=10, p=0.5: 54.20%

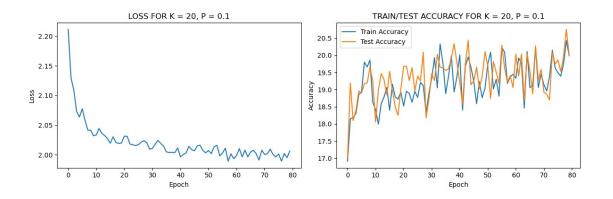
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=10, P=1.0 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 0.662, Train Accuracy: 94.73, Test Accuracy: 91.01 Epoch [20/80], Loss: 0.496, Train Accuracy: 96.11, Test Accuracy: 91.12 Epoch [30/80], Loss: 0.369, Train Accuracy: 97.16, Test Accuracy: 91.35 Epoch [40/80], Loss: 0.287, Train Accuracy: 97.87, Test Accuracy: 91.16 Epoch [50/80], Loss: 0.214, Train Accuracy: 98.31, Test Accuracy: 90.88 Epoch [60/80], Loss: 0.169, Train Accuracy: 98.79, Test Accuracy: 90.91 Epoch [70/80], Loss: 0.130, Train Accuracy: 99.38, Test Accuracy: 90.99 OBTAINED AN ACCURACY OF 99.52% AT EPOCH NO. 74 FOR K = 10 AND P = 0.0 OBTAINED AN ACCURACY OF 99.54% AT EPOCH NO. 77 FOR K=10 AND P=0.0Epoch [80/80], Loss: 0.110, Train Accuracy: 99.57, Test Accuracy: 90.86 OBTAINED AN ACCURACY OF 99.57% AT EPOCH NO. 79 FOR K = 10 AND P = 0.0 ---- FOR K: 10, P: 1.0 THE PLOTS ARE AS FOLLOWS



Loss for k=10, p=1.0: 0.035 Train Accuracy for k=10, p=1.0: 99.57% Test Accuracy for k=10, p=1.0: 90.86%

```
Epoch [10/80], Loss: 6.361, Train Accuracy: 18.62, Test Accuracy:
19.25
Epoch [20/80], Loss: 6.322, Train Accuracy: 18.93, Test Accuracy:
19.01
Epoch [30/80], Loss: 6.289, Train Accuracy: 18.89, Test Accuracy:
18.72
Epoch [40/80], Loss: 6.296, Train Accuracy: 19.38, Test Accuracy:
19.78
Epoch [50/80], Loss: 6.271, Train Accuracy: 19.04, Test Accuracy:
20.11
Epoch [60/80], Loss: 6.240, Train Accuracy: 19.45, Test Accuracy:
19.06
Epoch [70/80], Loss: 6.285, Train Accuracy: 19.45, Test Accuracy:
19.59
Epoch [80/80], Loss: 6.281, Train Accuracy: 19.99, Test Accuracy:
20.00
```

----- FOR K: 20, P: 0.1 THE PLOTS ARE AS FOLLOWS



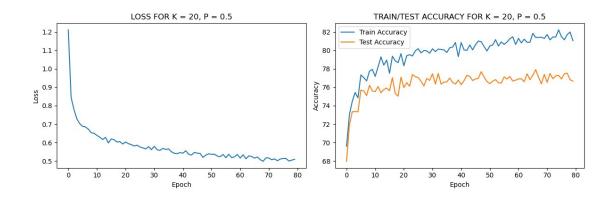
Loss for k=20, p=0.1: 2.007 Train Accuracy for k=20, p=0.1: 19.99% Test Accuracy for k=20, p=0.1: 20.00%

Epoch [60/80], Loss: 1.676, Train Accuracy: 80.65, Test Accuracy: 76.77

Epoch [70/80], Loss: 1.619, Train Accuracy: 81.30, Test Accuracy: 77.40

Epoch [80/80], Loss: 1.593, Train Accuracy: 81.06, Test Accuracy: 76.67

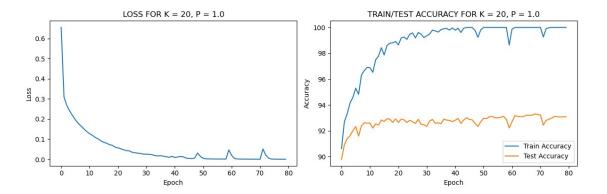
----- FOR K: 20, P: 0.5 THE PLOTS ARE AS FOLLOWS



Loss for k=20, p=0.5: 0.509 Train Accuracy for k=20, p=0.5: 81.06% Test Accuracy for k=20, p=0.5: 76.67%

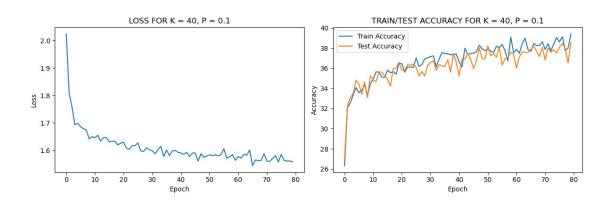
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=20, P=1.0 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 0.432, Train Accuracy: 96.90, Test Accuracy: 92.59 Epoch [20/80], Loss: 0.189, Train Accuracy: 98.91, Test Accuracy: 92.92 OBTAINED AN ACCURACY OF 99.58% AT EPOCH NO. 25 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.61% AT EPOCH NO. 27 FOR K = 20 AND P = 0.0 Epoch [30/80], Loss: 0.080, Train Accuracy: 99.22, Test Accuracy: 92.47 OBTAINED AN ACCURACY OF 99.79% AT EPOCH NO. 32 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.73% AT EPOCH NO. 33 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.64% AT EPOCH NO. 34 FOR K=20 AND P=0.0OBTAINED AN ACCURACY OF 99.83% AT EPOCH NO. 35 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.9% AT EPOCH NO. 36 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.92% AT EPOCH NO. 37 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.8% AT EPOCH NO. 38 FOR K = 20 AND P = 0.0 Epoch [40/80], Loss: 0.045, Train Accuracy: 99.96, Test Accuracy: 92.71 OBTAINED AN ACCURACY OF 99.96% AT EPOCH NO. 39 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.79% AT EPOCH NO. 40 FOR K = 20 AND P = 0.0 OBTAINED AN ACCURACY OF 99.94% AT EPOCH NO. 41 FOR K = 20 AND P = 0.0

```
OBTAINED AN ACCURACY OF 99.61% AT EPOCH NO. 42 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 99.94% AT EPOCH NO. 43 FOR K=20 AND P=0.0
OBTAINED AN ACCURACY OF 99.99% AT EPOCH NO. 44 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 45 FOR K=20 AND P=0.0
OBTAINED AN ACCURACY OF 99.99% AT EPOCH NO. 46 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 99.76% AT EPOCH NO. 47 FOR K = 20 AND P = 0.0
Epoch [50/80], Loss: 0.044, Train Accuracy: 99.82, Test Accuracy:
92.70
OBTAINED AN ACCURACY OF 99.82% AT EPOCH NO. 49 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 50 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 51 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 52 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 53 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 54 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 55 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 56 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 57 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 58 FOR K = 20 AND P = 0.0
Epoch [60/80], Loss: 0.147, Train Accuracy: 98.64, Test Accuracy:
92.22
OBTAINED AN ACCURACY OF 99.88% AT EPOCH NO. 60 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 99.99% AT EPOCH NO. 61 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 62 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 63 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 64 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 65 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 66 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 67 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 68 FOR K = 20 AND P = 0.0
Epoch [70/80], Loss: 0.002, Train Accuracy: 100.00, Test Accuracy:
93.26
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 69 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 70 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 99.9% AT EPOCH NO. 72 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 99.98% AT EPOCH NO. 73 FOR K=20 AND P=0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 74 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 75 FOR K=20 AND P=0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 76 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 77 FOR K = 20 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 78 FOR K = 20 AND P = 0.0
Epoch [80/80], Loss: 0.002, Train Accuracy: 100.00, Test Accuracy:
93.09
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 79 FOR K = 20 AND P = 0.0
----- FOR K: 20, P: 1.0 THE PLOTS ARE AS FOLLOWS
```



Loss for k=20, p=1.0: 0.001 Train Accuracy for k=20, p=1.0: 100.00% Test Accuracy for k=20, p=1.0: 93.09%

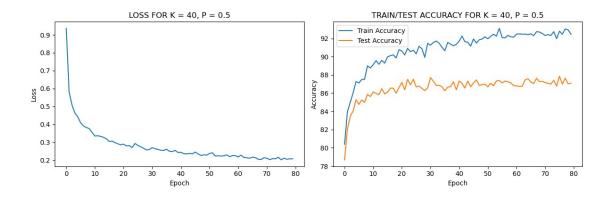
```
Epoch [10/80], Loss: 5.163, Train Accuracy: 34.45, Test Accuracy:
35.24
Epoch [20/80], Loss: 5.088, Train Accuracy: 36.50, Test Accuracy:
36.59
Epoch [30/80], Loss: 5.012, Train Accuracy: 37.03, Test Accuracy:
36.29
Epoch [40/80], Loss: 4.983, Train Accuracy: 37.42, Test Accuracy:
36.58
Epoch [50/80], Loss: 4.940, Train Accuracy: 37.74, Test Accuracy:
36.91
Epoch [60/80], Loss: 4.893, Train Accuracy: 37.55, Test Accuracy:
37.41
Epoch [70/80], Loss: 4.968, Train Accuracy: 38.63, Test Accuracy:
38.10
Epoch [80/80], Loss: 4.874, Train Accuracy: 39.41, Test Accuracy:
38.58
                -- FOR K: 40, P: 0.1 THE PLOTS ARE AS FOLLOWS
```



Loss for k=40, p=0.1: 1.557 Train Accuracy for k=40, p=0.1: 39.41% Test Accuracy for k=40, p=0.1: 38.58%

!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=40, P=0.5 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 1.112, Train Accuracy: 88.75, Test Accuracy: 85.62 Epoch [20/80], Loss: 0.894, Train Accuracy: 90.77, Test Accuracy: 86.61 Epoch [30/80], Loss: 0.814, Train Accuracy: 91.46, Test Accuracy: 86.57 Epoch [40/80], Loss: 0.760, Train Accuracy: 91.35, Test Accuracy: 86.36 Epoch [50/80], Loss: 0.714, Train Accuracy: 92.18, Test Accuracy: 86.97 Epoch [60/80], Loss: 0.705, Train Accuracy: 92.14, Test Accuracy: 86.83 Epoch [70/80], Loss: 0.669, Train Accuracy: 92.54, Test Accuracy: 87.29 Epoch [80/80], Loss: 0.649, Train Accuracy: 92.46, Test Accuracy: 87.08

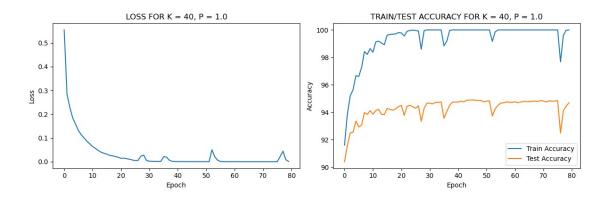
------ FOR K: 40, P: 0.5 THE PLOTS ARE AS FOLLOWS



Loss for k=40, p=0.5: 0.207 Train Accuracy for k=40, p=0.5: 92.46% Test Accuracy for k=40, p=0.5: 87.08%

```
OBTAINED AN ACCURACY OF 99.67% AT EPOCH NO. 16 FOR K=40 AND P=0.0
OBTAINED AN ACCURACY OF 99.69% AT EPOCH NO. 17 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.71% AT EPOCH NO. 18 FOR K = 40 AND P = 0.0
Epoch [20/80], Loss: 0.059, Train Accuracy: 99.81, Test Accuracy:
94.42
OBTAINED AN ACCURACY OF 99.81% AT EPOCH NO. 19 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.78% AT EPOCH NO. 20 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.56% AT EPOCH NO. 21 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.9% AT EPOCH NO. 22 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.96% AT EPOCH NO. 23 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.99% AT EPOCH NO. 24 FOR K=40 AND P=0.0
OBTAINED AN ACCURACY OF 99.96% AT EPOCH NO. 25 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.92% AT EPOCH NO. 26 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.95% AT EPOCH NO. 28 FOR K = 40 AND P = 0.0
Epoch [30/80], Loss: 0.015, Train Accuracy: 100.00, Test Accuracy:
94.68
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 29 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 30 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 31 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 32 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 33 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 34 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.97% AT EPOCH NO. 37 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 38 FOR K = 40 AND P = 0.0
Epoch [40/80], Loss: 0.003, Train Accuracy: 100.00, Test Accuracy:
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 39 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 40 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 41 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 42 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 43 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 44 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 45 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 46 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 47 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 48 FOR K = 40 AND P = 0.0
Epoch [50/80], Loss: 0.001, Train Accuracy: 100.00, Test Accuracy:
94.76
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 49 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 50 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 51 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.86% AT EPOCH NO. 53 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 54 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 55 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 56 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 57 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 58 FOR K = 40 AND P = 0.0
Epoch [60/80], Loss: 0.001, Train Accuracy: 100.00, Test Accuracy:
94.72
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 59 FOR K = 40 AND P = 0.0
```

```
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 60 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 61 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 62 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 63 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 64 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 65 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 66 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 67 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 68 FOR K = 40 AND P = 0.0
Epoch [70/80], Loss: 0.000, Train Accuracy: 100.00, Test Accuracy:
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 69 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 70 FOR K=40 AND P=0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 71 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 72 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 73 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 74 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 75 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.61% AT EPOCH NO. 77 FOR K = 40 AND P = 0.0
OBTAINED AN ACCURACY OF 99.98% AT EPOCH NO. 78 FOR K = 40 AND P = 0.0
Epoch [80/80], Loss: 0.003, Train Accuracy: 100.00, Test Accuracy:
94.70
OBTAINED AN ACCURACY OF 100.0% AT EPOCH NO. 79 FOR K = 40 AND P = 0.0
                 ---- FOR K: 40, P: 1.0 THE PLOTS ARE AS FOLLOWS
```



Loss for k=40, p=1.0: 0.001 Train Accuracy for k=40, p=1.0: 100.00% Test Accuracy for k=40, p=1.0: 94.70%

Wall time: 3h 34min 35s

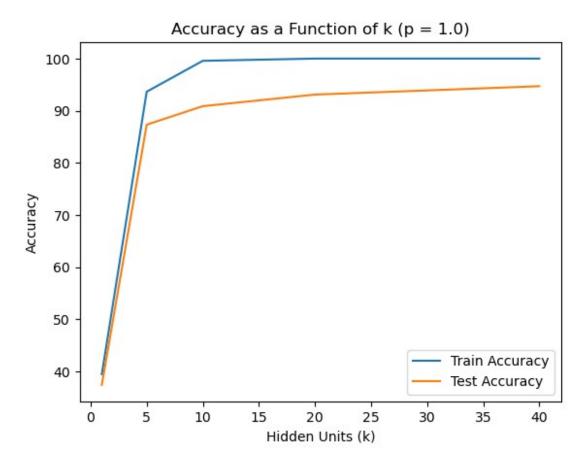
Fix p = 1.0 which is the case of "no dropout regularization". Plot the test and training accuracy as a function of k. As k increases, does the performance improve? At what k, training accuracy becomes 100%?

```
task2_results_df = pd.DataFrame(task2_results)
task2_results_df
```

plt.xlabel('Hidden Units (k)')

<del>-</del>	_								
	Units	Dropout Factor	Mean Loss	Train Accuracy	Test				
Accuracy 0 10.218680 1	1	0.1	2.302839	10.00					
	1	0.5	2.097958	17.04					
16.155733 2	1	1.0	1.563459	39.43					
37.379931 3	5	0.1	2.224583	12.83					
12.763131 4	5	0.5							
38.943389									
5 87.318618	5	1.0	0.251154						
6 14.949928	10	0.1	2.133704	15.52					
7 54.199877	10	0.5	1.124713	57.06					
34.199877 8 90.864500 9 19.997956 10 76.670754 11 93.092172	10	1.0	0.035008	99.57					
	20	0.1	2.006705	19.99					
	20	0.5	0.508855	81.06					
	20	1.0	0.000608	100.00					
	40	0.1							
38.575516									
13 87.083589	40	0.5	0.207238	92.46					
14 94.696505	40	1.0	0.001082	100.00					
<pre># Filter the DataFrame for rows where p = 1.0 filtered_df = task2_results_df[task2_results_df['Dropout Factor'] == 1.0]</pre>									
<pre># Plot the test and training accuracy plt.plot(filtered_df['Hidden Units'], filtered_df['Train Accuracy'], label='Train Accuracy') plt.plot(filtered_df['Hidden Units'], filtered_df['Test Accuracy'], label='Test Accuracy') plt.title('Accuracy as a Function of k (p = 1.0)')</pre>									

```
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



As the K increases, the model's performance also increases, as the model becomes more complex and gets the tendency to generalize itself better on the unseen data.

At K = 20, the training Accuracy Becomes 100%

Plot the training and test accuracy as a function of k and for different  $p \in P$  on the same plot. What is the role of p on training accuracy? When p is smaller, is it easier to optimize or more difficult? For each choice of p, determine at what choice of k, training accuracy becomes 100%.

```
# Group the DataFrame by Dropout Factor
grouped_df = task2_results_df.groupby('Dropout Factor')

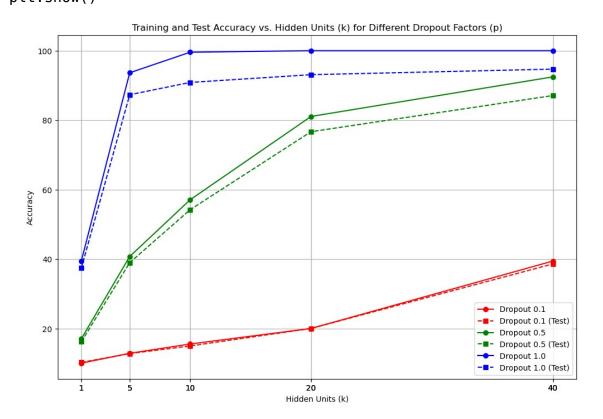
# Set the colors for the plot
colors = ['r', 'g', 'b']

# Plot the training and test accuracy for different Dropout Factors
plt.figure(figsize=(12, 8))

for i, (dropout, group) in enumerate(grouped_df):
```

```
plt.plot(group['Hidden Units'], group['Train Accuracy'],
marker='o', color=colors[i], label=f'Dropout {dropout}')
    plt.plot(group['Hidden Units'], group['Test Accuracy'],
marker='s', color=colors[i], linestyle='--', label=f'Dropout {dropout}
(Test)')

plt.title('Training and Test Accuracy vs. Hidden Units (k) for
Different Dropout Factors (p)')
plt.xlabel('Hidden Units (k)')
plt.ylabel('Accuracy')
plt.ylabel('Accuracy')
plt.xticks(task2_results_df['Hidden Units'])
plt.legend()
plt.grid(True)
plt.show()
```



As P (Dropout Factor) increases, the Training Accuracy of the Model for a given and fixed depth also improves. When P is smaller, it is difficult to Optimize.

For P = 0.1 and 0.5, no values of K showed Training Accuracy of 100%. But for P = 1.0, K = 10 showed 99.57% Training Accuracy but K = 20 and 40 gave 100%.

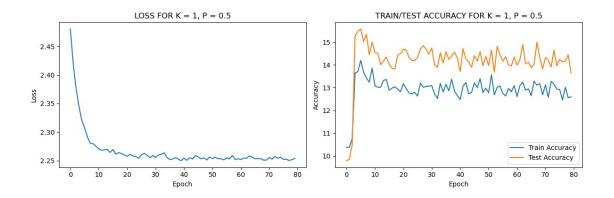
## **TASK 3 - (7 PTS)**

We will spice up the problem by adding some noise to labels. Pick 40% of the training examples at random. Assign their labels at random to another value from 0 to 9. For instance, if the original image is 0 and its label is 0, then you will assign its label to a

number from 1 to 9 at random. Thus 60% of the training examples remain correct and 40% will have incorrect labels. Repeat the previous step with this noisy dataset.

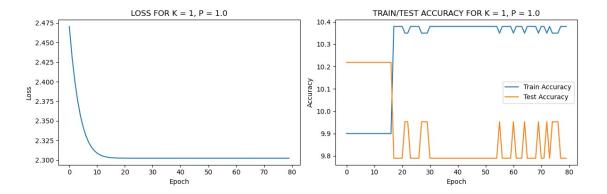
```
PREPARING THE NOISY DATA
# Copy the original training data
noisy train data = list(train data)
noisy train labels = [label for , label in noisy train data]
# Pick 40% of the training examples
num examples = len(noisy train data)
num noisy examples = int(0.4 * num examples)
noisy indices = random.sample(range(num examples), num noisy examples)
# Assign random incorrect labels to the noisy examples
for index in noisy indices:
    original_label = noisy_train_labels[index]
    new label = random.choice([label for label in range(10) if label !
= original label])
    noisy_train_labels[index] = new_label
# Update the labels in the noisy training data
noisy train data = [(image, label) for (image, ), label in
zip(noisy train data, noisy train labels)]
noisy train loader = DataLoader(noisy train data, batch size=100,
shuffle=True)
MAIN CODE
k_{for} = [1,5,10,20,40]
p for noisy = [0.1, 0.5, 1.0]
num epochs = 80
task3 results = []
%%time
for k in k_for_noisy:
  for p in p for noisy:
    t3 loss, t3 train acc, t3 test acc =
general shallow net(noisy train loader, test loader, 'ADAM', k, p,
num epochs)
    t31 \text{ mean loss} = t3 \text{ loss}[-1]
    t31_max_train_acc = t3_train acc[-1]
    t31 \text{ max test acc} = t3 \text{ test acc}[-1]
    print(f'Mean Loss for k={k}, p={p}: {t31 mean loss:.3f}')
    print(f'Max Train Accuracy for k={k}, p={p}:
{t31 max train acc:.2f}%')
```

```
print(f'Max Test Accuracy for k=\{k\}, p=\{p\}: {t31 max test acc:.2f}
%1)
   t3 total summary = {'Hidden Units': k, 'Dropout Factor': p, 'Mean
Loss': t31 mean loss, 'Train Accuracy': t31 max train acc, 'Test
Accuracy': t31 max test acc}
   task3 results.append(t3 total summary)
   print()
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=1, P=0.1 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!
Epoch [10/80], Loss: 2.307, Train Accuracy: 9.94, Test Accuracy: 10.22
Epoch [20/80], Loss: 2.303, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [30/80], Loss: 2.302, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [40/80], Loss: 2.302, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [50/80], Loss: 2.302, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [60/80], Loss: 2.303, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [70/80], Loss: 2.302, Train Accuracy: 10.38, Test Accuracy: 9.79
Epoch [80/80], Loss: 2.302, Train Accuracy: 10.38, Test Accuracy: 9.79
         ----- FOR K: 1, P: 0.1 THE PLOTS ARE AS FOLLOWS
             LOSS FOR K = 1, P = 0.1
                                       TRAIN/TEST ACCURACY FOR K = 1, P = 0.1
                                 10.4
  2.40
                                 10.3
  2.38
                                 10.2
 s 2.36
                                Accuracy
                                                        Train Accuracy
                                 10.1
                                                        Test Accuracy
                                 10.0
  2.34
                                  9.9
                                  9.8
   2.30 -
        10
           20
               30
                     50
                        60
                           70
                              80
                                       10
                                          20
                                             30
                                                   50
                                                      60
                                                         70
                                                            80
                                               Fnoch
Mean Loss for k=1, p=0.1: 2.302
Max Train Accuracy for k=1, p=0.1: 10.38%
Max Test Accuracy for k=1, p=0.1: 9.79%
Epoch [10/80], Loss: 2.275, Train Accuracy: 13.86, Test Accuracy:
15.00
Epoch [20/80], Loss: 2.260, Train Accuracy: 12.82, Test Accuracy:
14.50
Epoch [30/80], Loss: 2.259, Train Accuracy: 13.07, Test Accuracy:
14.47
Epoch [40/80], Loss: 2.250, Train Accuracy: 12.65, Test Accuracy:
```



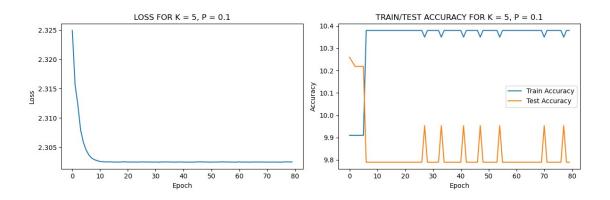
Mean Loss for k=1, p=0.5: 2.254 Max Train Accuracy for k=1, p=0.5: 12.60% Max Test Accuracy for k=1, p=0.5: 13.65%

------ FOR K: 1, P: 1.0 THE PLOTS ARE AS FOLLOWS



Mean Loss for k=1, p=1.0: 2.302 Max Train Accuracy for k=1, p=1.0: 10.38% Max Test Accuracy for k=1, p=1.0: 9.79%

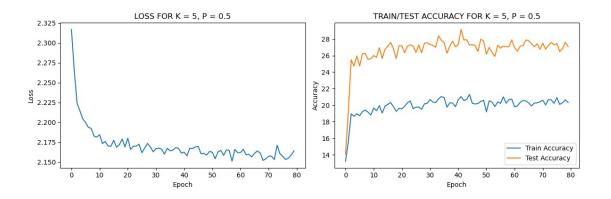
----- FOR K: 5, P: 0.1 THE PLOTS ARE AS FOLLOWS



Mean Loss for k=5, p=0.1: 2.303 Max Train Accuracy for k=5, p=0.1: 10.38% Max Test Accuracy for k=5, p=0.1: 9.79%

```
Epoch [10/80], Loss: 2.181, Train Accuracy: 18.80, Test Accuracy:
25.63
Epoch [20/80], Loss: 2.169, Train Accuracy: 19.61, Test Accuracy:
27.19
Epoch [30/80], Loss: 2.163, Train Accuracy: 20.22, Test Accuracy:
27.55
Epoch [40/80], Loss: 2.161, Train Accuracy: 19.81, Test Accuracy:
27.07
Epoch [50/80], Loss: 2.163, Train Accuracy: 20.60, Test Accuracy:
27.75
Epoch [60/80], Loss: 2.162, Train Accuracy: 20.72, Test Accuracy:
27.90
Epoch [70/80], Loss: 2.154, Train Accuracy: 20.39, Test Accuracy:
26.75
Epoch [80/80], Loss: 2.164, Train Accuracy: 20.33, Test Accuracy:
27.09
```

----- FOR K: 5, P: 0.5 THE PLOTS ARE AS FOLLOWS



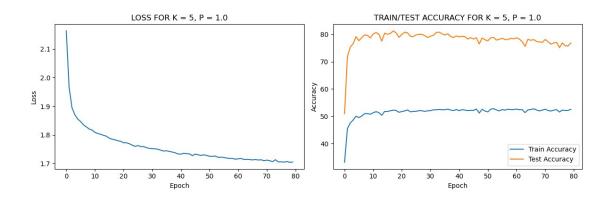
Mean Loss for k=5, p=0.5: 2.164 Max Train Accuracy for k=5, p=0.5: 20.33% Max Test Accuracy for k=5, p=0.5: 27.09%

Epoch [60/80], Loss: 1.715, Train Accuracy: 52.46, Test Accuracy: 78.34

Epoch [70/80], Loss: 1.709, Train Accuracy: 52.29, Test Accuracy: 77.08

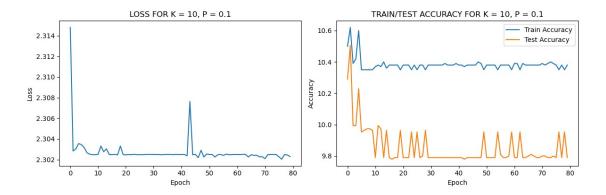
Epoch [80/80], Loss: 1.705, Train Accuracy: 52.57, Test Accuracy: 76.84

----- FOR K: 5, P: 1.0 THE PLOTS ARE AS FOLLOWS



Mean Loss for k=5, p=1.0: 1.705 Max Train Accuracy for k=5, p=1.0: 52.57% Max Test Accuracy for k=5, p=1.0: 76.84%

----- FOR K: 10, P: 0.1 THE PLOTS ARE AS FOLLOWS



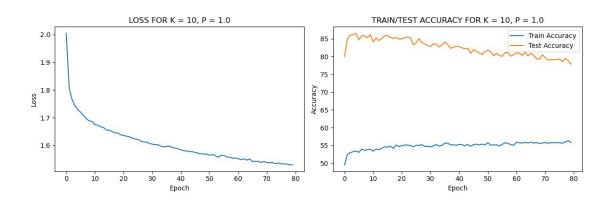
Mean Loss for k=10, p=0.1: 2.302 Max Train Accuracy for k=10, p=0.1: 10.38% Max Test Accuracy for k=10, p=0.1: 9.79%

Epoch [10/80], Loss: 1.996, Train Accuracy: 32.37, Test Accuracy: 47.57 Epoch [20/80], Loss: 1.971, Train Accuracy: 33.55, Test Accuracy: 50.02 Epoch [30/80], Loss: 1.970, Train Accuracy: 34.06, Test Accuracy: 49.09 Epoch [40/80], Loss: 1.964, Train Accuracy: 33.97, Test Accuracy: 49.81 Epoch [50/80], Loss: 1.956, Train Accuracy: 34.46, Test Accuracy: 49.03 Epoch [60/80], Loss: 1.964, Train Accuracy: 34.10, Test Accuracy: 49.36 Epoch [70/80], Loss: 1.947, Train Accuracy: 35.41, Test Accuracy: Epoch [80/80], Loss: 1.931, Train Accuracy: 34.80, Test Accuracy: 49.41 FOR K: 10, P: 0.5 THE PLOTS ARE AS FOLLOWS

LOSS FOR K = 10, P = 0.5TRAIN/TEST ACCURACY FOR K = 10, P = 0.5 50 2.20 45 2.15 40 2.10 D 35 2.05 2.00 30 25 1.95 Test Accuracy 60 70 80 Epoch Epoch

Mean Loss for k=10, p=0.5: 1.931 Max Train Accuracy for k=10, p=0.5: 34.80% Max Test Accuracy for k=10, p=0.5: 49.41%

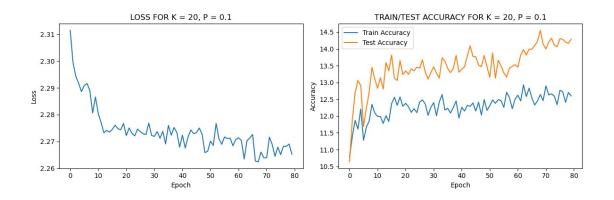
!!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=10, P=1.0 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 1.687, Train Accuracy: 53.95, Test Accuracy: 86.16 Epoch [20/80], Loss: 1.639, Train Accuracy: 54.69, Test Accuracy: 84.92 Epoch [30/80], Loss: 1.608, Train Accuracy: 54.78, Test Accuracy: 83.11 Epoch [40/80], Loss: 1.588, Train Accuracy: 55.03, Test Accuracy: 82.90 Epoch [50/80], Loss: 1.570, Train Accuracy: 55.14, Test Accuracy: 81.39 Epoch [60/80], Loss: 1.554, Train Accuracy: 55.06, Test Accuracy: 80.52 Epoch [70/80], Loss: 1.542, Train Accuracy: 55.66, Test Accuracy: 80.50 Epoch [80/80], Loss: 1.531, Train Accuracy: 55.80, Test Accuracy: 77.90 ----- FOR K: 10, P: 1.0 THE PLOTS ARE AS FOLLOWS



Mean Loss for k=10, p=1.0: 1.531 Max Train Accuracy for k=10, p=1.0: 55.80% Max Test Accuracy for k=10, p=1.0: 77.90%

```
13.23
Epoch [30/80], Loss: 2.272, Train Accuracy: 12.26, Test Accuracy:
13.29
Epoch [40/80], Loss: 2.268, Train Accuracy: 11.94, Test Accuracy:
13.30
Epoch [50/80], Loss: 2.266, Train Accuracy: 12.17, Test Accuracy:
13.47
Epoch [60/80], Loss: 2.271, Train Accuracy: 12.50, Test Accuracy:
13.53
Epoch [70/80], Loss: 2.264, Train Accuracy: 12.46, Test Accuracy:
14.14
Epoch [80/80], Loss: 2.265, Train Accuracy: 12.60, Test Accuracy:
14.29
```

----- FOR K: 20, P: 0.1 THE PLOTS ARE AS FOLLOWS

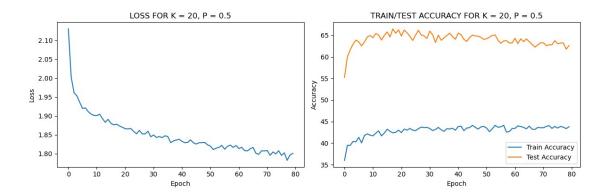


Mean Loss for k=20, p=0.1: 2.265 Max Train Accuracy for k=20, p=0.1: 12.60% Max Test Accuracy for k=20, p=0.1: 14.29%

```
!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=20, P=0.5 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!
Epoch [10/80], Loss: 1.901, Train Accuracy: 41.84, Test Accuracy:
64.95
Epoch [20/80], Loss: 1.870, Train Accuracy: 43.00, Test Accuracy:
66.27
Epoch [30/80], Loss: 1.845, Train Accuracy: 43.65, Test Accuracy:
64.28
Epoch [40/80], Loss: 1.838, Train Accuracy: 42.98, Test Accuracy:
64.08
Epoch [50/80], Loss: 1.823, Train Accuracy: 43.88, Test Accuracy:
64.05
Epoch [60/80], Loss: 1.821, Train Accuracy: 43.38, Test Accuracy:
63.25
Epoch [70/80], Loss: 1.807, Train Accuracy: 43.56, Test Accuracy:
```

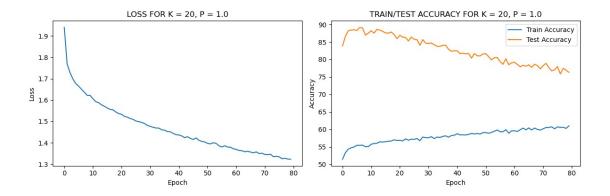
63.25 Epoch [80/80], Loss: 1.801, Train Accuracy: 43.80, Test Accuracy: 62.61

----- FOR K: 20, P: 0.5 THE PLOTS ARE AS FOLLOWS



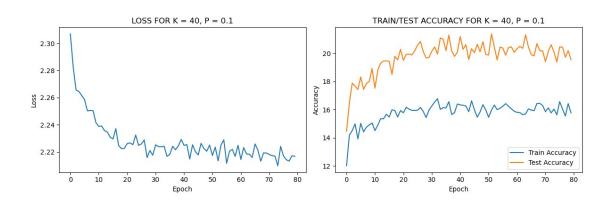
Mean Loss for k=20, p=0.5: 1.801 Max Train Accuracy for k=20, p=0.5: 43.80% Max Test Accuracy for k=20, p=0.5: 62.61%

!!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=20, P=1.0 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 1.622, Train Accuracy: 55.10, Test Accuracy: 87.54 Epoch [20/80], Loss: 1.537, Train Accuracy: 56.85, Test Accuracy: 85.95 Epoch [30/80], Loss: 1.481, Train Accuracy: 57.66, Test Accuracy: 84.67 Epoch [40/80], Loss: 1.438, Train Accuracy: 58.32, Test Accuracy: 82.46 Epoch [50/80], Loss: 1.404, Train Accuracy: 59.11, Test Accuracy: 81.56 Epoch [60/80], Loss: 1.372, Train Accuracy: 59.60, Test Accuracy: 79.06 Epoch [70/80], Loss: 1.351, Train Accuracy: 59.81, Test Accuracy: 77.30 Epoch [80/80], Loss: 1.323, Train Accuracy: 61.03, Test Accuracy: 76.34 ----- FOR K: 20, P: 1.0 THE PLOTS ARE AS FOLLOWS



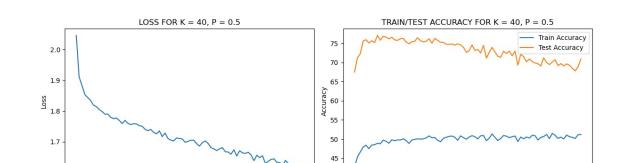
Mean Loss for k=20, p=1.0: 1.323 Max Train Accuracy for k=20, p=1.0: 61.03% Max Test Accuracy for k=20, p=1.0: 76.34%

Epoch [10/80], Loss: 2.242, Train Accuracy: 15.05, Test Accuracy: Epoch [20/80], Loss: 2.223, Train Accuracy: 15.94, Test Accuracy: 20.28 Epoch [30/80], Loss: 2.218, Train Accuracy: 15.97, Test Accuracy: 19.70 Epoch [40/80], Loss: 2.229, Train Accuracy: 16.40, Test Accuracy: 20.08 Epoch [50/80], Loss: 2.225, Train Accuracy: 15.98, Test Accuracy: 19.92 Epoch [60/80], Loss: 2.225, Train Accuracy: 15.89, Test Accuracy: 20.07 Epoch [70/80], Loss: 2.219, Train Accuracy: 16.29, Test Accuracy: Epoch [80/80], Loss: 2.217, Train Accuracy: 15.77, Test Accuracy: 19.55 --- FOR K: 40, P: 0.1 THE PLOTS ARE AS FOLLOWS



Mean Loss for k=40, p=0.1: 2.217 Max Train Accuracy for k=40, p=0.1: 15.77% Max Test Accuracy for k=40, p=0.1: 19.55%

!!!!!!!!!!!!!!!!!!!!!! RUNNING FOR K=40, P=0.5 !!!!!!!!!!!!!!!!!!!!!!!!!!!!!!!! Epoch [10/80], Loss: 1.799, Train Accuracy: 48.79, Test Accuracy: 75.82 Epoch [20/80], Loss: 1.756, Train Accuracy: 48.86, Test Accuracy: 74.90 Epoch [30/80], Loss: 1.735, Train Accuracy: 49.67, Test Accuracy: 75.78 Epoch [40/80], Loss: 1.702, Train Accuracy: 49.97, Test Accuracy: 72.64 Epoch [50/80], Loss: 1.672, Train Accuracy: 50.44, Test Accuracy: 72.75 Epoch [60/80], Loss: 1.662, Train Accuracy: 50.08, Test Accuracy: 71.53 Epoch [70/80], Loss: 1.645, Train Accuracy: 51.50, Test Accuracy: 70.12 Epoch [80/80], Loss: 1.621, Train Accuracy: 51.19, Test Accuracy: 70.92 ----- FOR K: 40, P: 0.5 THE PLOTS ARE AS FOLLOWS



70

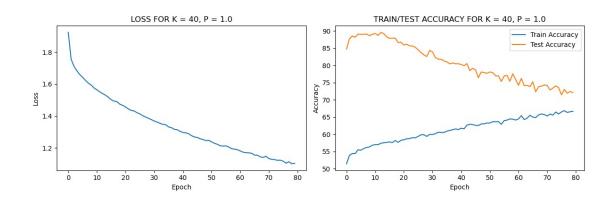
Mean Loss for k=40, p=0.5: 1.621 Max Train Accuracy for k=40, p=0.5: 51.19% Max Test Accuracy for k=40, p=0.5: 70.92%

10

10

70

80



Mean Loss for k=40, p=1.0: 1.104 Max Train Accuracy for k=40, p=1.0: 66.69% Max Test Accuracy for k=40, p=1.0: 72.17%

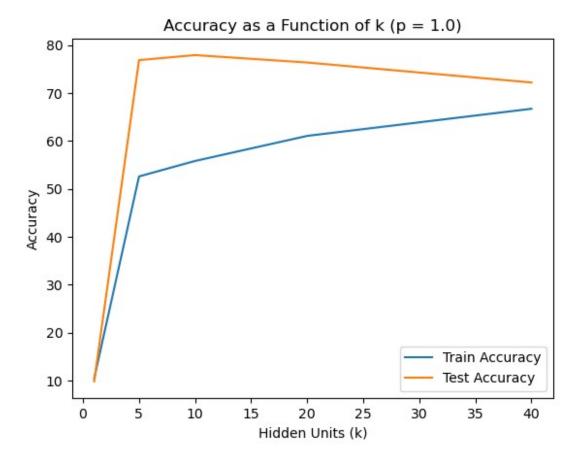
Wall time: 1h 12min 11s

Fix p = 1.0 which is the case of "no dropout regularization". Plot the test and training accuracy as a function of k. As k increases, does the performance improve? At what k, training accuracy becomes 100%?

task3\_results\_df = pd.DataFrame(task3\_results)
task3\_results\_df

Hidden	Units	Dropout Factor	Mean Loss	Train Accuracy	Test
Accuracy 0 9.789495 1 13.652156 2	1	0.1	2.302488	10.38	
	1	0.5	2.253589	12.60	
	1	1.0	2.302473	10.38	

```
9.789495
               5
                             0.1
                                    2.302505
                                                       10.38
3
9.789495
               5
                             0.5
                                    2.164142
                                                       20.33
27.089720
               5
                             1.0
                                    1.704800
                                                       52.57
76.844472
              10
                             0.1
                                    2.302290
                                                       10.38
9.789495
                                                       34.80
              10
                             0.5
                                    1.931062
49.407317
                                    1.530597
              10
                             1.0
                                                       55.80
77.896996
              20
                             0.1
                                    2.265271
                                                       12.60
14.285714
10
              20
                             0.5
                                    1.800741
                                                       43.80
62.609851
11
              20
                             1.0
                                    1.323178
                                                       61.03
76.343756
                             0.1
                                                       15.77
12
              40
                                    2.216760
19.548334
              40
                             0.5
                                    1.620892
                                                       51.19
13
70.917637
              40
                             1.0
                                                       66.69
14
                                    1.103973
72.174535
# Filter the DataFrame for rows where p = 1.0
filtered df = task3 results df[task3 results df['Dropout Factor'] ==
1.0]
# Plot the test and training accuracy
plt.plot(filtered df['Hidden Units'], filtered df['Train Accuracy'],
label='Train Accuracy')
plt.plot(filtered df['Hidden Units'], filtered df['Test Accuracy'],
label='Test Accuracy')
plt.title('Accuracy as a Function of k (p = 1.0)')
plt.xlabel('Hidden Units (k)')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



As, K increases, the performance of the model does not seems to improve. Rather, it looks like the model is going to face overfitting. Since, noise is introduced in the train Data, it is getting more difficult for the model to fit on the training data than the test Data. Thereby showing Test Accuracy greater than the Training Accuracy.

For no value of K, the model acheives 100% training Accuracy

Plot the training and test accuracy as a function of k and for different  $p \in P$  on the same plot. What is the role of p on training accuracy? When p is smaller, is it easier to optimize or more difficult? For each choice of p, determine at what choice of k, training accuracy becomes 100%.

```
# Group the DataFrame by Dropout Factor
grouped_df = task3_results_df.groupby('Dropout Factor')

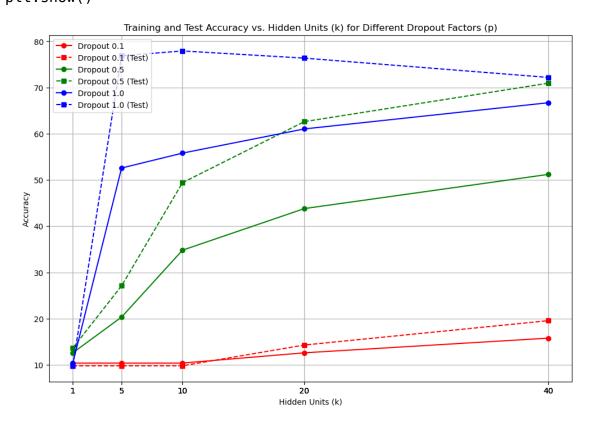
# Set the colors for the plot
colors = ['r', 'g', 'b']

# Plot the training and test accuracy for different Dropout Factors
plt.figure(figsize=(12, 8))

for i, (dropout, group) in enumerate(grouped_df):
    plt.plot(group['Hidden Units'], group['Train Accuracy'],
marker='o', color=colors[i], label=f'Dropout {dropout}')
```

```
plt.plot(group['Hidden Units'], group['Test Accuracy'],
marker='s', color=colors[i], linestyle='--', label=f'Dropout {dropout}
(Test)')

plt.title('Training and Test Accuracy vs. Hidden Units (k) for
Different Dropout Factors (p)')
plt.xlabel('Hidden Units (k)')
plt.ylabel('Accuracy')
plt.ylabel('Accuracy')
plt.xticks(task2_results_df['Hidden Units'])
plt.legend()
plt.grid(True)
plt.show()
```



As P (Dropout Factor) increases, the Training Accuracy of the Model for a given and fixed depth also improves. When P is smaller, it is difficult to Optimize.

For no values of K and P we get Training Accuracy of 100%

## **TASK 4 - (3 PTS)**

Comment on the differences between Step 2 and Step 3. How does noise change things? For which setup dropout is more useful?

As the Width increases, the model becomes more complex and generalizes better on the unseen data thereby giving higher accuracy in step 2. But in step 3 with the addition of noise, we observe lower training accuracy.

As the dropout factor increases increases the accuracy lowers in the step 2. But in Step 3 the test accuracy is lower than the train accuracy, thereby establishing the overfitting of the model.

Hence, dropout is more useful for setup 2.