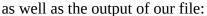
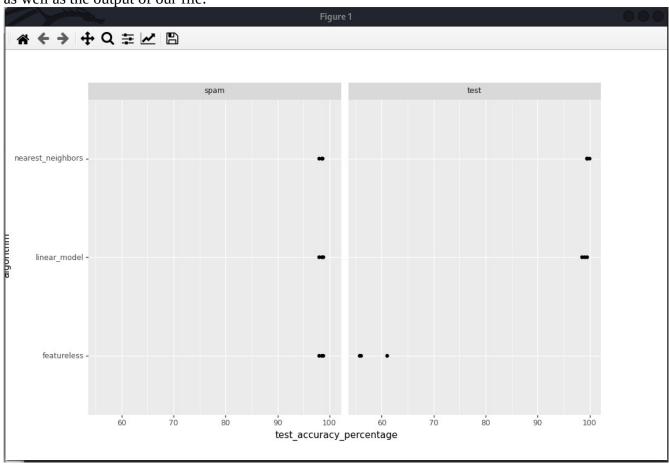
## \* attempt 2

I had spent the time since submitting attempt one trying to wrap my head around passing variables in and out of functions. I was finally able to get the program working after realizing the df\_init function is returning a tuple so our values passed should be as such. After assigning the tuple I run the train() method on the last param passed in, c. from there our train() method will populate the test\_acc\_df, populated into variable d. Finally d is passed into plot() where we are plotting out points in facet\_grid type format. Here is the code with some changes from attempt 1 highlighted in yellow:





# lib for retrieving src file from web

import urllib.request

# lib for reading files on OS

import os

# lib used for copying src file info into destination

import shutil

import pandas as pd

import plotnine as p9

import numpy as np

# could not figure out how to calculate the mode of a list, using mode from lib

from statistics import mode

import sklearn

from sklearn.model\_selection import KFold #train/test splits

from sklearn.model\_selection import GridSearchCV #selecting best # of neighbors

```
from sklearn.neighbors import KNeighborsClassifier #nearest_neighbors prediction.
from sklearn.pipeline import make pipeline # increase iteration sz
from sklearn, preprocessing import StandardScaler #
from sklearn.linear model import LogisticRegression
# our src files we want to download. spam and test sets
test_url = 'https://hastie.su.domains/ElemStatLearn/datasets/zip.test.gz'
test file = 'zip.test.gz'
spam_url = 'https://hastie.su.domains/ElemStatLearn/datasets/spam.data'
spam file = 'spam.data'
# number of columns in test file (257) count from zero
test cols = 257
# number of columns in spam file (56) count from zero
spam cols = 57
,,,,,,
our last assignment was only downloading 1 file, to adhere to guidelines and
avoid repeatable code, created a method we will use to retrieve multiple files.
def retrieve(src_file, src_url):
  check if a file exists in the current directory
  retrieve a file given the url
  if not os.path.isfile(src_file):
     urllib.request.urlretrieve(src_url, src_file)
     print("Downloading " + src_file + " from " + src_url + "...\n")
  else:
     print(src_file + " already exists in this folder...continuing anyway\n")
since we are initializing multiple dataframes for out multiple
sources of data, I wanted to try and minimize repeating code.
this method will
  - take in src file
  - create a dataframe
  - drop specified rows of the src file
  - convert our data into numpy
Hopefully this is allowed!
  notice: (am repeating code) but spent too much time creating traveral
  type solution going over a list of src_files and src_urls.
  Wanted to seperate file into some seperate functions for the sake of
  readability.
def df init(src1, src2, src1 cols, src2 cols, data dict):
  # read in downloaded src file as a pandas dataframe
  # seperate dataframes because different manipulations will be done
```

```
df1 = pd.read_csv(src1, header=None, sep=" ")
  df2 = pd.read csv(src2, header=None, sep=" ")
  # remove any rows which have non-01 labels
  df1[0] = df1[0].astype(int)
  df2[0] = df2[0].astype(int)
  df1 = df1[df1[0].isin([0, 1])]
  df2 = df2[df2[0].isin([0, 1])]
  # initialize and convert outputs to a label vector
  df1_labels = df1[0]
  df2 labels = df2[0]
  Convert our dataframe to a dictionary with numpy array exlcuding the
  first column; iloc for row and col specifying.
  data_dict = {
     "test":(df1.loc[:,1:].to_numpy(), df1[0]),
     "spam":(df2.loc[:,1:].to_numpy(), df2[0]),
  # print our dataframes to visualize in tabular form
  print(df1)
  print(df2)
  #print(data_dict)
  return df1, df2, data_dict
algorithm shown in class and from our demo.
def train(data dict):
  test_acc_df_list = []
  # increase the max iteration from default 100
  pipe = make_pipeline(StandardScaler(), LogisticRegression(max_iter=1000))
  for data_set, (input_mat, output_vec) in data_dict.items():
    print(data_set)
    kf = KFold(n_splits=3, shuffle=True, random_state=1)
     for fold_id, indices in enumerate(kf.split(input_mat)):
       print(fold_id)
       index_dict = dict(zip(["train","test"], indices))
       param_dicts = [\{'n_neighbors':[x]\}\} for x in range(1, 21)]
       # does subtrain/validation splits.
       clf = GridSearchCV(KNeighborsClassifier(), param_dicts)
       # copy above for linear model. call cv=5 in initial pipe was not
       # recognized; try a call here
       linear_model = sklearn.linear_model.LogisticRegressionCV(cv=5)
```

```
set_data_dict = {}
for set_name, index_vec in index_dict.items():
  set_data_dict[set_name] = (
     input_mat [ index_vec ],
     output_vec.iloc[index_vec]
# * is unpacking a tuple to use as the different positional arguments
# clf.fit(set_data_dict["train"][0], set_data_dict["train"][1])
# train models and stub out linear model
clf.fit(*set_data_dict["train"])
# method 2: dict instead of tuple.
set_data_dict = {}
for set_name, index_vec in index_dict.items():
  set data dict[set name] = {
     "X":input mat[index vec],
     "y":output_vec.iloc[index_vec]
# ** is unpacking a dict to use as the named arguments
# train models and stub out linear_model and create algo for finding
# clf.fit(X=set_data_dict["train"]["X"], y=set_data_dict["train"]["y"]])
clf.fit(**set_data_dict["train"])
#mode = max(set(output_vec))
featureless model = mode(output vec)
linear_model.fit(**set_data_dict["train"])
clf.best_params_
cv_df = pd.DataFrame(clf.cv_results_)
cv_df.loc[:,["param_n_neighbors","mean_test_score"]]
pred dict = {
  "nearest_neighbors":clf.predict(set_data_dict["test"]["X"]),
  #TODO add featureless and linear model.
  "featureless": featureless_model,
  "linear_model": linear_model.predict(set_data_dict["test"]["X"])
for algorithm, pred_vec in pred_dict.items():
  test acc dict = {
     "test_accuracy_percentage":(
       pred_vec == set_data_dict["test"]["y"]).mean()*100,
     "data_set":data_set,
     "fold_id":fold_id,
     "algorithm":algorithm
  test_acc_df_list.append(pd.DataFrame(test_acc_dict, index=[0]))
```

```
test acc df = pd.concat(test acc df list)
  return test_acc_df
,,,,,,
an attempt to modularize our code to improve readibility, this method
will make a ggplot to visually examine which learning algorithm is
best for each data set
def plot(test_acc_df):
  gg = (p9.ggplot(test_acc_df,
       p9.aes(x='test_accuracy_percentage'
       ,y='algorithm'))
      # .~ spreads vals across columns
      +p9.facet_grid('.~ data_set')
      # Use geom_point to create scatterplots
      +p9.geom_point())
  print(gg)
def main():
  data_dict = {}
  # retrieve our source files. Spam and test.
  retrieve(test file, test url)
  retrieve(spam_file, spam_url)
  # call method to initialize our data frames, convert to numpy arrays,
  (test, spam, _dict) = df_init(test_file, spam_file, test_cols,
                       spam_cols, data_dict)
  # call method to perform our algorithm shown in class and the demo
  trained_data = train(_dict)
  # plot our values by passing in the previously filled variable d
  plot(trained_data)
if __name__ == '__main__':
  # run main
  main()
```

## \* attempt 1

Overall the assignment did not stray too much from what was given in the demo. However, to try and keep a modularized approach to how we implement code, I attempted to create different methods for some of the big functions we were passing around. In the end my code did not end up working how I wanted due to some python programming errors. I was unsure of how to pass variables around the program and which ones to pass and assign, etc. The final error I got was not able to concat

test\_acc\_df\_list. This has to be because of the data\_dict not being seeing inside of the train method. Just like last time I am confident if I was given a bit more time I would be able to tackle this problem. The method for plotting our points was eventually not met so I am unsure if it was implemented correctly. Based off of the demos and my previous assignment, I decided to use the code for plotting and hoping it would work. This time I was able to find a decent documentation on plotting using the facet\_grid but again, I am not sure if this was implemented correctly. In the coming days I might submit a new file that is working.

Here is my code:

```
# lib for retrieving src file from web
import urllib.request
# lib for reading files on OS
import os
# lib used for copying src file info into destination
import shutil
import pandas as pd
import plotnine as p9
import numpy as np
# could not figure out how to calculate the mode of a list, using mode from lib
import statistics
from sklearn.model_selection import KFold #train/test splits
from sklearn.model_selection import GridSearchCV #selecting best # of neighbors
from sklearn.neighbors import KNeighborsClassifier #nearest neighbors prediction.
from sklearn.pipeline import make pipeline # increase iteration sz
from sklearn.preprocessing import StandardScaler #
from sklearn.linear_model import LogisticRegression
# our src files we want to download. spam and test sets
test_url = 'https://hastie.su.domains/ElemStatLearn/datasets/zip.test.gz'
test file = 'zip.test.gz'
spam_url = 'https://hastie.su.domains/ElemStatLearn/datasets/spam.data'
spam_file = 'spam.data'
# number of columns in test file (257) count from zero
test cols = 257
# number of columns in spam file (56) count from zero
spam_cols = 57
# dataframe list
#test_acc_df_list = []
*****
our last assignment was only downloading 1 file, to adhere to guidelines and
avoid repeatable code, created a method we will use to retrieve multiple files.
def retrieve(src file, src url):
  check if a file exists in the current directory
```

```
retrieve a file given the url
  if not os.path.isfile(src_file):
     urllib.request.urlretrieve(src_url, src_file)
     print("Downloading " + src_file + " from " + src_url + "...\n")
  else:
    print(src_file + " already exists in this folder...continuing anyway\n")
since we are initializing multiple dataframes for out multiple
sources of data, I wanted to try and minimize repeating code.
this method will
  - take in src file
  - create a dataframe
  - drop specified rows of the src file
  - convert our data into numpy
Hopefully this is allowed!
  notice: (am repeating code) but spent too much time creating traveral
  type solution going over a list of src files and src urls.
  Wanted to seperate file into some seperate functions for the sake of
  readability.
def df_init(src1, src2, src1_cols, src2_cols, data_dict):
  # read in downloaded src file as a pandas dataframe
  # seperate dataframes because different manipulations will be done
  df1 = pd.read_csv(src1, header=None, sep=" ")
  df2 = pd.read_csv(src2, header=None, sep=" ")
  # remove any rows which have non-01 labels
  df1[0] = df1[0].astype(int)
  df2[0] = df2[0].astype(int)
  df1 = df1[df1[0].isin([0, 1])]
  df2 = df2[df2[0].isin([0, 1])]
  # initialize and convert outputs to a label vector
  df1_labels = df1[0]
  df2 labels = df2[0]
  Convert our dataframe to a dictionary with numpy array exlcuding the
  first column; iloc for row and col specifying.
  data_dict = {
     "test":(df1.loc[:,1:].to_numpy(), df1[0]),
     "spam":(df2.loc[:,1:].to_numpy(), df2[0]),
  # print our dataframes to visualize in tabular form
  print(df1)
  print(df2)
```

```
#print(data_dict)
  return df1, df2, data dict
algorithm shown in class and from our demo.
def train(data dict):
  test acc df list = []
  # increase the max iteration from default 100
  pipe = make pipeline(StandardScaler(), LogisticRegression(max iter=1000))
  for data set, (input mat, output vec) in data dict.items():
    print(data_set)
    kf = KFold(n splits=3, shuffle=True, random state=1)
     for fold_id, indices in enumerate(kf.split(input_mat)):
       print(fold_id)
       index_dict = dict(zip(["train","test"], indices))
       param dicts = [\{'n \text{ neighbors'}: [x]\}\} for x in range(1, 21)]
       # does subtrain/validation splits.
       clf = GridSearchCV(KNeighborsClassifier(), param_dicts)
       # copy above for linear model. call cv=5 in initial pipe was not
       # recognized; try a call here
       linear model = sklearn.linear model.LogisticRegression(cv=5)
       set_data_dict = {}
       for set_name, index_vec in index_dict.items():
          set data dict[set name] = (
            input_mat [ index_vec ],
            output vec.iloc[index vec]
       # * is unpacking a tuple to use as the different positional arguments
       # clf.fit(set_data_dict["train"][0], set_data_dict["train"][1])
       # train models and stub out linear_model
       clf.fit(*set_data_dict["train"])
       # method 2: dict instead of tuple.
       set_data_dict = {}
       for set_name, index_vec in index_dict.items():
          set data dict[set name] = {
            "X":input_mat[index_vec],
            "y":output_vec.iloc[index_vec]
       # ** is unpacking a dict to use as the named arguments
       # train models and stub out linear model and create algo for finding
       # mode
       # clf.fit(X=set_data_dict["train"]["X"], y=set_data_dict["train"]["y"]])
```

```
clf.fit(**set_data_dict["train"])
       #mode = max(set(output vec))
       import statistics
       MCE = mode(output_vec)
       linear_model.fit(*set_data_dict["train"])
       clf.best_params_
       cv_df = pd.DataFrame(clf.cv_results_)
       cv_df.loc[:,["param_n_neighbors","mean_test_score"]]
       pred dict = {
          "nearest_neighbors":clf.predict(set_data_dict["test"]["X"]),
          #TODO add featureless and linear model.
          "featureless": MCE,
          "linear_model": linear_model.predict(set_data_dict["test"]["X"])
       for algorithm, pred_vec in pred_dict.items():
          test acc dict = {
            "test_accuracy_percent":(
               pred_vec == set_data_dict["test"]["y"]).mean()*100,
            "data_set":data_set,
            "fold_id":fold_id,
            "algorithm":algorithm
          test_acc_df_list.append(pd.DataFrame(test_acc_dict, index=[0]))
  test_acc_df = pd.concat(test_acc_df_list)
  return test_acc_df
an attempt to modularize our code to improve readibility, this method
will make a ggplot to visually examine which learning algorithm is
best for each data set
def plot(test_acc_df):
  gg = (p9.ggplot(test_acc_df,p9.aes(x='test_accuracy_percent',y='algorithm'))
      +p9.facet_grid('.~ data_set')
      +p9.geom_point())
  print(gg)
def main():
  data_dict = {}
  test acc df = []
  # retrieve our source files. Spam and zip.
  retrieve(test_file, test_url)
```

```
retrieve(spam_file, spam_url)

# call method to initialize our data frames, convert to numpy arrays,
    df_init(test_file, spam_file, test_cols, spam_cols, data_dict)
    # call method to perform our algorithm shown in class and the demo
    train(data_dict)
    plot(test_acc_df)

if __name__ == '__main__':
    # run main
    main()
```