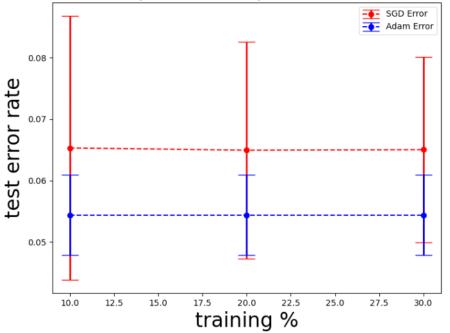
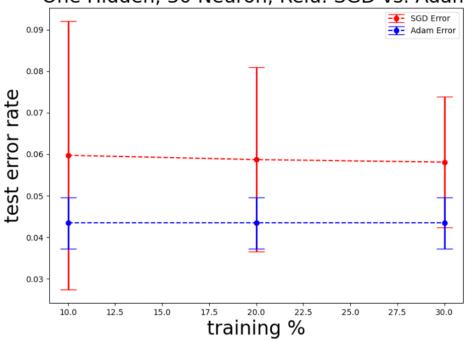


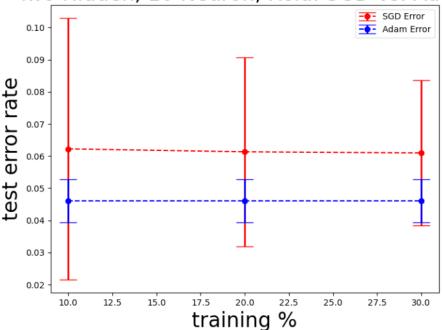
One Hidden, 10 Neuron, Tanh: SGD vs. Adam



One Hidden, 30 Neuron, Relu: SGD vs. Adam



Two Hidden, 10 Neuron, Relu: SGD vs. Adam



Summary:

The results are as follows. In terms of accuracy the order goes

 $logisitic < 10_{relu} \approx 10_{tanh} < 30_{relu} \approx 10,10_{relu}$ Or in other terms

Logistic is the least accurate, followed by 10 relu and 10 tanh activiation, followed by 30 relu and 2 layer 10 relu activation.

On the spambase dataset all implementations were able to get reasonable accuracte after just a few epochs. And as expected they all got more accurate the more training data they received. Using the libraries like Tensorflow made running these algorithms very efficient. We did not record time but it was much faster than the first time analyzing the spambase dataset with our own implementation of logistic regression.

The neural network implementations were also fast. However the 2 hidden layer, and 30 neuron implementation were the slowest of the bunch. They however made up for that computation expense by being the most accuracte for both optimizers "SGD" and "Adam".

Adam optimizer was also consistently better than SGD. This gives reason as to why Adam optimizer is more standard in the field than SGD. However SGD is still a good way to ensure things are working.

The code used to generate the 80-20 splits and further training percentages from that were all from the hw3 solution. However the creation of the models and subsequent training were all written originally.

```
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
import numpy as np
import pandas as pd
from sklearn.utils import shuffle
import matplotlib.pyplot as plt
#read data from hw3 solution
def read data(filename='spambase.csv', d=57):
  # read data
  data df = pd.read csv(filename, header=None).dropna(axis=0)
  # preprocess
  X = np.array(data df.iloc[:,:d])
  X = (X-np.mean(X, axis=0))/np.std(X, axis=0) # normalize
  y = np.array(data df.iloc[:,d:])
  y = y.flatten()
  # seperate classes
  X1 = X[y==1]
  X0 = X[y==0]
  y1 = y[y==1]
  y0 = y[y==0]
  return X0, y0, X1, y1
#train test split from hw3 solution
def train test split(X0, y0, X1, y1, split_percent):
  # splitting id
  split id0 = int(X0.shape[0]*split percent*.01)
  split id1 = int(X1.shape[0]*split percent*.01)
  # random shuffle each class
  X00, y00 = \text{shuffle}(X0, y0)
  X11, y11 = \text{shuffle}(X1, y1)
  # train set
  Xtrain = np.concatenate((X00[:split id0], X11[:split id1]), axis=0)
  ytrain = np.concatenate((y00[:split id0], y11[:split id1]), axis=0)
  # test set
  Xtest = np.concatenate((X00[split id0:], X11[split id1:]), axis=0)
  ytest = np.concatenate((v00[split id0:], v11[split id1:]), axis=0)
  # random shuffle train set
  Xtrain, ytrain = shuffle(Xtrain, ytrain)
  return Xtrain, ytrain, Xtest, ytest
#train data from hw3 solution
def train data(X, y, train percent):
  N = X.shape[0]
  split id = int(N*train percent/100)
  return X[:split id], y[:split id]
def one 10 tanh(filename, num splits, train percent):
  #parse
  X0, y0, X1, y1 = read data(filename)
  # init error
  test errors = np.zeros((num splits, len(train percent)))
  #creat 1 hidden 10 neuron SGD model
```

```
model = tf.keras.models.Sequential()
  model.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='tanh'))
  model.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model.compile(loss = 'binary crossentropy', optimizer = 'SGD',metrics = ['accuracy'])
  split percent=80
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       loss, acc = model.evaluate(Xtest i,ytest i)
       test errors[i][j] = 1-acc
  test2 = np.zeros((num splits,len(train percent)))
  #create 1 hidden, 10 neuron Adam Model
  model2= tf.keras.models.Sequential()
  model2.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='tanh'))
  model2.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model2.compile(loss = 'binary crossentropy', optimizer = 'Adam',metrics = ['accuracy'])
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       #create training percent data
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       #train
       model2.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,yerbose=0)
       #test
       loss, acc = model.evaluate(Xtest i,ytest i)
       #store results
       test2[i][i] = 1-acc
  return test errors, test2
def one 30 relu(filename, num splits, train percent):
  #parse
  X0, y0, X1, y1 = read data(filename)
  # init error
  test errors = np.zeros((num splits, len(train percent)))
  #create 1 hidden 30 neuron SGD model
  model = tf.keras.models.Sequential()
  model.add(tf.keras.layers.Dense(30,input dim = len(X0[0,:]),activation='relu'))
  model.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model.compile(loss = 'binary crossentropy', optimizer = 'SGD',metrics = ['accuracy'])
  split percent=80
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       loss, acc = model.evaluate(Xtest i,ytest i)
```

```
#print(loss)
       test errors[i][j] = 1-acc
  print("halfway")
  test2 = np.zeros((num splits,len(train percent)))
  model2= tf.keras.models.Sequential()
  model2.add(tf.keras.layers.Dense(30,input dim = len(X0[0,:]),activation='relu'))
  model2.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model2.compile(loss = 'binary crossentropy', optimizer = 'Adam', metrics = ['accuracy'])
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model2.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       loss, acc = model.evaluate(Xtest i,ytest i)
       #print(loss)
       test2[i][j] = 1-acc
  return test errors,test2
def one 10 relu(filename, num splits, train percent):
  #parse
  X0, y0, X1, y1 = read data(filename)
  # init error
  test errors = np.zeros((num splits, len(train percent)))
  #create 1 hidden 10 neuron relu model
  model = tf.keras.models.Sequential()
  model.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='relu'))
  model.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model.compile(loss = 'binary crossentropy', optimizer = 'SGD',metrics = ['accuracy'])
  split percent=80
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       loss, acc = model.evaluate(Xtest i,ytest i)
       #print(loss)
       test errors[i][i] = 1-acc
  test2 = np.zeros((num splits,len(train percent)))
  #create 1 hidden 10 neuron Adam model
  model2= tf.keras.models.Sequential()
  model2.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='relu'))
  model2.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model2.compile(loss = 'binary crossentropy', optimizer = 'Adam',metrics = ['accuracy'])
  for i in range(num splits):
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model2.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       loss, acc = model.evaluate(Xtest i, ytest i)
       #print(loss)
```

```
test2[i][j] = 1-acc
  return test errors, test2
def two 10 relu(filename, num splits, train percent):
  #parse
  X0, y0, X1, y1 = read data(filename)
  test errors = np.zeros((num splits, len(train percent)))
  #create 2 hidden layer 10 neuron sgd model
  model = tf.keras.models.Sequential()
  model.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='relu'))
  model.add(tf.keras.layers.Dense(10,activation='relu'))
  model.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model.compile(loss = 'binary crossentropy', optimizer = 'SGD',metrics = ['accuracy'])
  split percent=80
  for i in range(num splits):
     #create 80-20 split
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split_percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       #create percent training data
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       #train model
       model.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       #test model
       loss, acc = model.evaluate(Xtest i,ytest i)
       #store results
       test errors[i][j] = 1-acc
  test2 = np.zeros((num splits,len(train percent)))
  #create 2 hidden layer 10 neuron Adam model
  model2= tf.keras.models.Sequential()
  model2.add(tf.keras.layers.Dense(10,input dim = len(X0[0,:]),activation='relu'))
  model2.add(tf.keras.layers.Dense(10,activation='relu'))
  model2.add(tf.keras.layers.Dense(1,activation = "sigmoid"))
  model2.compile(loss = 'binary crossentropy', optimizer = 'Adam', metrics = ['accuracy'])
  for i in range(num splits):
     #create 80-20 split
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       #create training percent data
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       model2.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       #test
       loss, acc = model.evaluate(Xtest i,ytest i)
       #store results
       test2[i][j] = 1-acc
  return test errors, test2
```

def logistic(filename, num splits, train percent):

```
#parse
  X0, y0, X1, y1 = read data(filename)
  test errors = np.zeros((num splits, len(train percent)))
  #Create logistic regression SGD model
  model = tf.keras.models.Sequential()
  model.add(tf.keras.layers.Dense(1,input dim = len(X0[0,:]),activation='sigmoid'))
  model.compile(loss = 'binary crossentropy', optimizer = 'SGD',metrics = ['accuracy'])
  split percent=80
  for i in range(num splits):
     #create 80-20 Split
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split_percent)
     # try for different train set sizes
     for j, percent in enumerate(train percent):
       #create percent training data
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       #train model
       model.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,verbose=0)
       #test model
       loss, acc = model.evaluate(Xtest i,ytest i)
       #store results
       test errors[i][j] = 1-acc
  test2 = np.zeros((num splits,len(train percent)))
  #Create logistic regression Adam model
  model2= tf.keras.models.Sequential()
  model2.add(tf.keras.layers.Dense(1,input dim = len(X0[0,:]),activation='sigmoid'))
  model2.compile(loss = 'binary crossentropy', optimizer = 'Adam', metrics = ['accuracy'])
  for i in range(num splits):
     #create 80-20 split
     Xtrain i, ytrain i, Xtest i, ytest i = train test split(X0, y0, X1, y1, split percent)
     # try for differnt train set sizes
     for j, percent in enumerate(train percent):
       #create percent training data
       Xtrain ij, ytrain ij = train data(Xtrain i, ytrain i, percent)
       #train model
       model2.fit(Xtrain ij,ytrain ij,epochs=2,batch size=20,yerbose=0)
       #test model
       loss, acc = model.evaluate(Xtest i,ytest i)
       #store results
       test2[i][j] = 1-acc
  return test errors, test2
filename='spambase.csv'
num splits=100
train percent = [10, 20, 30]
#Comment Block For Running different moels
# sgd,adam = logistic(filename,num splits,train percent)
# sgd,adam = one 10 relu(filename,num splits,train percent)
# sgd,adam = one 10 tanh(filename,num splits,train percent)
```

```
# sgd,adam = one_30_relu(filename,num_splits,train_percent)

#plotting

sgd_mean = np.mean(sgd, axis=0)

adam_mean = np.mean(adam,axis=0)

sgd_stdev = np.std(sgd,axis=0)

adam_stdev =np.std(adam,axis=0)

plt.errorbar(train_percent, sgd_mean, yerr=sgd_stdev, fmt='ro--',capsize=10, elinewidth=2, label="SGD Error")

plt.errorbar(train_percent, adam_mean, yerr=adam_stdev, fmt='bo--', capsize=10, elinewidth=2, label="Adam Error")

plt.legend()

plt.xlabel('training %', fontsize=24)

plt.ylabel('test error rate', fontsize=24)

plt.title('Two Hidden, 10 Neuron, Relu: SGD vs. Adam', fontsize=24)

plt.show()
```