Problem 2

b)

```
In [15]: import numpy as np
         from scipy.io import loadmat
         import matplotlib.pyplot as plt
         from sklearn.svm import LinearSVC
         def countIncorrectClass(X, y, w):
             error = 0
             for i in range(len(y)):
                  xiT = X[[i], :]
                  yi = y[i]
                  if (yi * xiT @ w < 0):</pre>
                      error+=1
              return error
         X = np.array([[2, 1], [1.5, 1], [.5, 1], [-.5, 1]]);
         y = np.array([1, 1, -1, -1]);
         w_LS = np.linalg.inv(X.T @ X) @ X.T @ y
         print("w = ", w_LS)
         print("error count is ", countIncorrectClass(X, y, w_LS));
```

```
w = [0.94915254 - 0.83050847] error count is 0
```

d)

```
In [16]: X = np.array([[2, 1], [1.5, 1], [.5, 1], [-5, 1]]);
y = np.array([1, 1, -1, -1]);
w_LS = np.linalg.inv(X.T @ X) @ X.T @ y
print("w = ", w_LS)
print("error count is ", countIncorrectClass(X, y, w_LS));

w = [0.256 0.064]
error count is 1
```

Problem 3

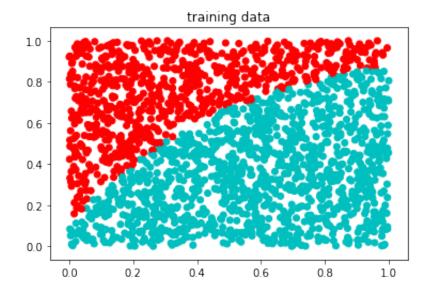
a)

```
In [17]: in_data = loadmat('classifier_data.mat')

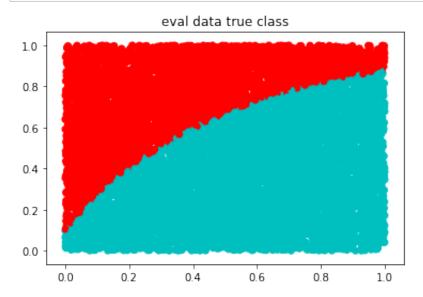
x_train = in_data['x_train']
x_eval = in_data['x_eval']
y_train = in_data['y_train']
y_eval = in_data['y_eval']

n_eval = np.size(y_eval)
n_train = np.size(y_train)

plt.scatter(x_train[:,0],x_train[:,1], color=['c' if i==-1 else 'r' for plt.title('training data')
plt.show()
```



In [18]: plt.scatter(x_eval[:,0],x_eval[:,1], color=['c' if i==-1 else 'r' for
 plt.title('eval data true class')
 plt.show()

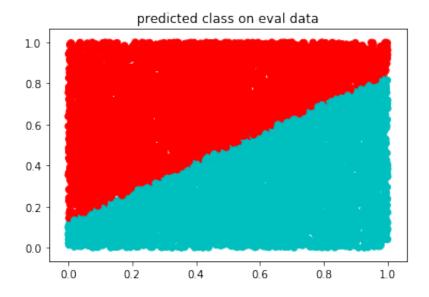


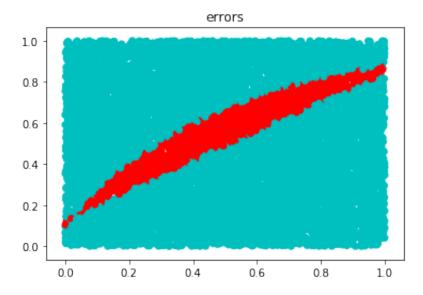
```
In [19]: ## Classifier 1
    x_train_1 = np.hstack(( x_train, np.ones((n_train,1)) ))
    x_eval_1 = np.hstack(( x_eval, np.ones((n_eval,1)) ))

# Train classifier using linear SVM from SK Learn library
    clf = LinearSVC(random_state=0, tol=1e-8)
    clf.fit(x_train_1, np.squeeze(y_train))
    w_opt = clf.coef_.transpose()

#uncomment this line to use least squares classifier
    #w_opt = np.linalg.inv(x_train_1.T@x_train_1)@x_train_1.T@y_train

    y_hat_outlier = np.sign(x_eval_1@w_opt)
    plt.scatter(x_eval[:,0],x_eval[:,1], color=['c' if i==-1 else 'r' for
    plt.title('predicted class on eval data')
    plt.show()
```





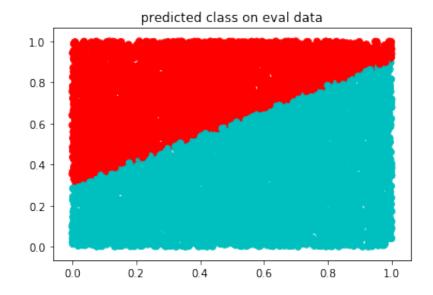
b)

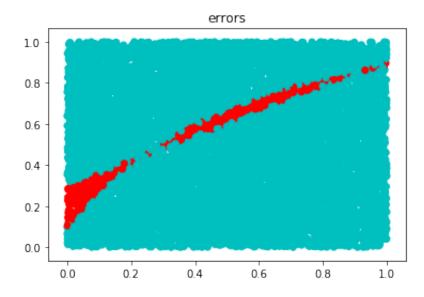
```
In [26]: ## Classifier 1
    x_train_1 = np.hstack(( x_train, np.ones((n_train,1)) ))
    x_eval_1 = np.hstack(( x_eval, np.ones((n_eval,1)) ))

# Train classifier using linear SVM from SK Learn library
    clf = LinearSVC(random_state=0, tol=1e-8)
    clf.fit(x_train_1, np.squeeze(y_train))
    w_opt = clf.coef_.transpose()

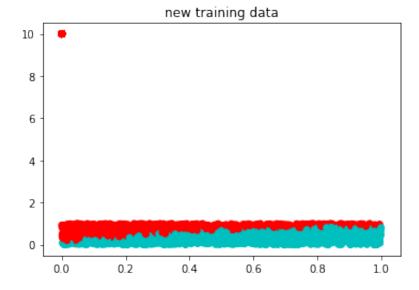
#uncomment this line to use least squares classifier
    w_opt = np.linalg.inv(x_train_1.T@x_train_1)@x_train_1.T@y_train

    y_hat_outlier = np.sign(x_eval_1@w_opt)
    plt.scatter(x_eval[:,0],x_eval[:,1], color=['c' if i==-1 else 'r' for
    plt.title('predicted class on eval data')
    plt.show()
```





Add correct points far from boundary

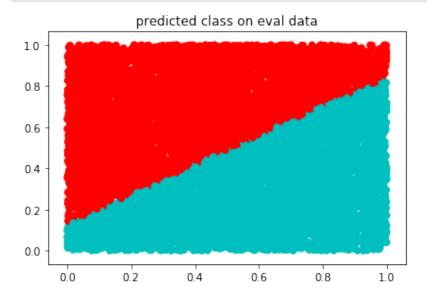


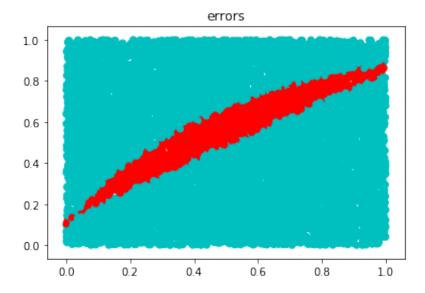
```
In [30]: x_train_outlier_1 = np.hstack((x_train_outlier, np.ones((n_train+n_new x_eval_1 = np.hstack((x_eval, np.ones((n_eval,1)))))

#Train classifier using off the shelf SVM from sklearn
clf = LinearSVC(random_state=0, tol=1e-5)
clf.fit(x_train_outlier_1, np.squeeze(y_train_outlier))
w_opt_outlier = clf.coef_.transpose()

#uncomment this line to use least squares classifier
#w_opt_outlier = np.linalg.inv(x_train_outlier_1.T@x_train_outlier_1)@

y_hat_outlier = np.sign(x_eval_1@w_opt_outlier)
plt.scatter(x_eval[:,0],x_eval[:,1], color=['c' if i==-1 else 'r' for
plt.title('predicted class on eval data')
plt.show()
```





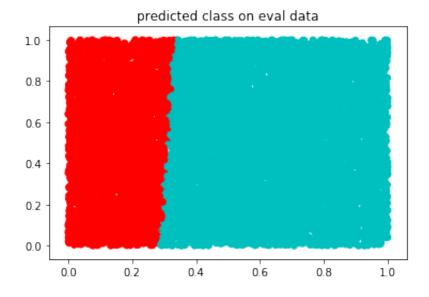
d)

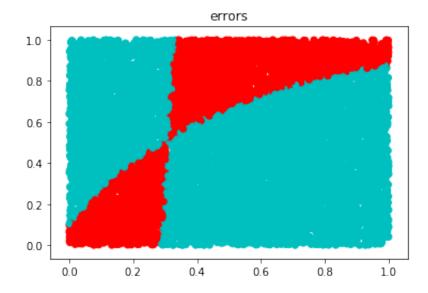
```
In [33]: x_train_outlier_1 = np.hstack((x_train_outlier, np.ones((n_train+n_nev x_eval_1 = np.hstack((x_eval, np.ones((n_eval,1)))))

#Train classifier using off the shelf SVM from sklearn
clf = LinearSVC(random_state=0, tol=1e-5)
clf.fit(x_train_outlier_1, np.squeeze(y_train_outlier))
w_opt_outlier = clf.coef_.transpose()

#uncomment this line to use least squares classifier
w_opt_outlier = np.linalg.inv(x_train_outlier_1.T@x_train_outlier_1)@x

y_hat_outlier = np.sign(x_eval_1@w_opt_outlier)
plt.scatter(x_eval[:,0],x_eval[:,1], color=['c' if i==-1 else 'r' for
plt.title('predicted class on eval data')
plt.show()
```





In []: