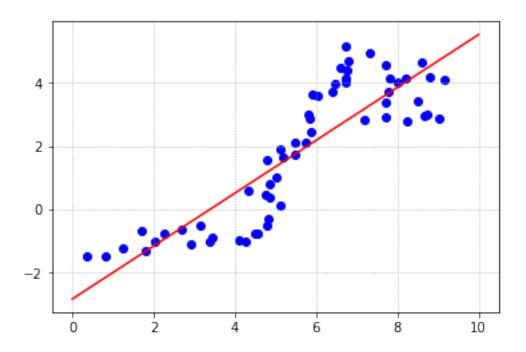
hw2

April 26, 2018

1 CS 273P Homework 2

```
In [126]: import numpy as np
          import matplotlib.pyplot as plt
          import mltools as ml
          np.random.seed(0)
  Problem 1 (a)
In [127]: data=np.genfromtxt("data/curve80.txt",delimiter=None)
          X=data[:,0]
          X = X[:,np.newaxis]
          Y = data[:,1]
          Xtr,Xte,Ytr,Yte = ml.splitData(X,Y,0.75)
  Problem 1 (b)
In [128]: lr = ml.linear.linearRegress( Xtr, Ytr );
          xs = np.linspace(0,10,200);
          xs = xs[:,np.newaxis]
          ys = lr.predict( xs );
          plt.scatter(Xtr,Ytr,c='b')
          plt.plot(xs,ys,'r')
          plt.grid(linestyle='dotted')
          ax=plt.axis()
          plt.show()
          print("Theta: ",lr.theta)
          YhatTrain=lr.predict(Xtr)
          YhatTest=lr.predict(Xte)
          def mse(Yhat,Y):
              mse=0
              for i in range(0,len(Y)):
                  mse=mse+((Y[i]-Yhat[i])*(Y[i]-Yhat[i]))
              return mse/len(Y)
          print("Mean squared error on Training data: ",mse(YhatTrain,Ytr))
          print("Mean squared error on Test data: ",mse(YhatTest,Yte))
```



Theta: [[-2.82765049 0.83606916]]

Mean squared error on Training data: [1.12771196]

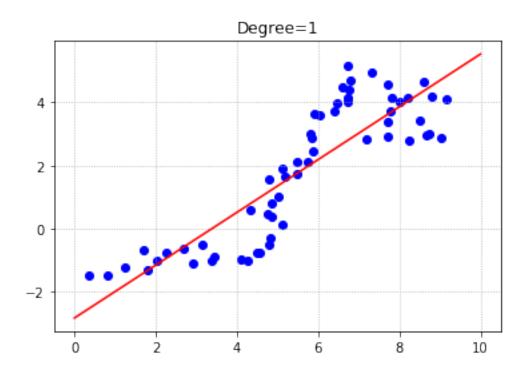
Mean squared error on Test data: [2.2423492]

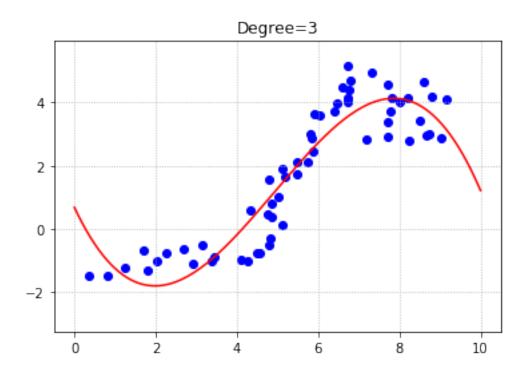
Problem 1(c)

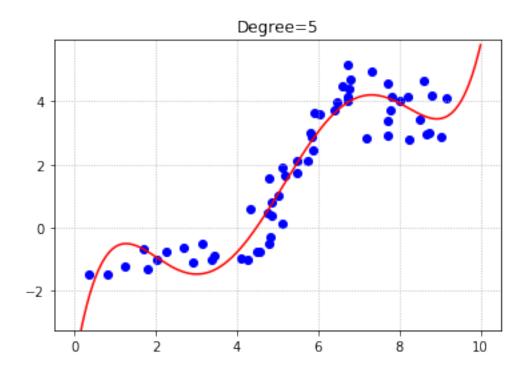
1) Plots of Learned prediction functions

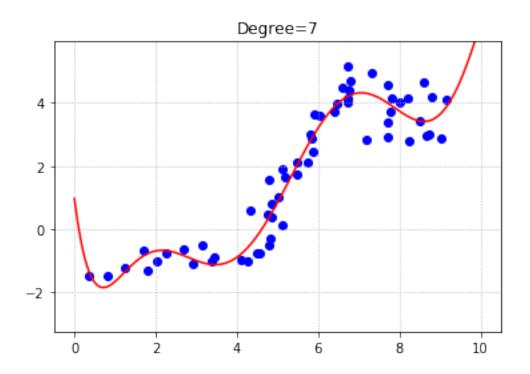
```
In [129]: mseTrain=[]
          mseTest=[]
          degrees=[1,3,5,7,10,18]
          for degree in degrees:
              XtrPoly=ml.transforms.fpoly(Xtr,degree,bias=False);
              XtrPoly,params=ml.transforms.rescale(XtrPoly)
              lr=ml.linear.linearRegress(XtrPoly,Ytr)
              XteP,_=ml.transforms.rescale(ml.transforms.fpoly(Xte,degree,False))
              Phi = lambda X: ml.transforms.rescale( ml.transforms.fpoly(X, degree,False), para
              YhatTrain = lr.predict( Phi(Xtr) ); # predict on training data
              YhatTest = lr.predict( Phi(Xte) ); # predict on test data
              ys=lr.predict(Phi(xs))
              mseTrain.append(mse(YhatTrain,Ytr))
              mseTest.append(mse(YhatTest,Yte))
              plt.axis(ax)
              plt.title('Degree='+str(degree))
              plt.scatter(Xtr,Ytr,c='b')
```

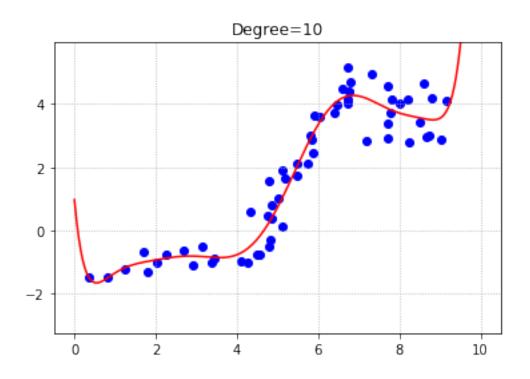
```
plt.plot(xs,ys,'r')
plt.grid(linestyle='dotted')
plt.show()
```

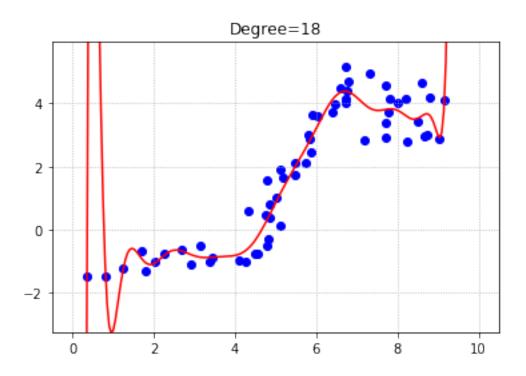












2) Plots of training and test errors



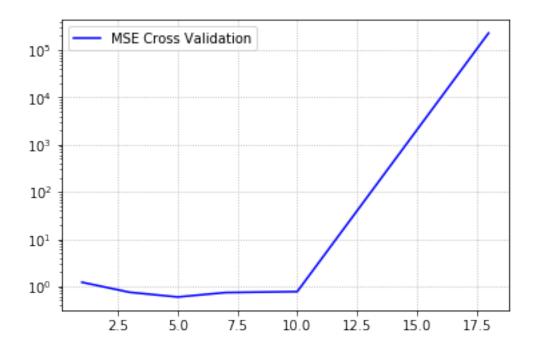
Problem 2) Cross-Validation

A) MSE values for each degree.

print(mseCrossValidation)

```
In [131]: degrees=[1,3,5,7,10,18]
          mseCrossValidation=[]
          for degree in degrees:
               \#Phi = lambda \ X: \ ml.transforms.rescale(\ ml.transforms.fpoly(X,\ degree,False),\ pair (X,\ degree,False))
               J=[]
               nFolds = 5;
               for iFold in range(nFolds):
                   Xti, Xvi, Yti, Yvi = ml.crossValidate(Xtr, Ytr, nFolds, iFold); # take ith data bl
                   XtiP=ml.transforms.fpoly(Xti,degree,False);
                   XviP=ml.transforms.fpoly(Xvi,degree,False);
                   XtiP,params=ml.transforms.rescale(XtiP)
                   XviP,_=ml.transforms.rescale(XviP,params)
                   learner = ml.linear.linearRegress(XtiP,Yti)
                   #YviHat=learner.predict(Phi(Xvi))
                   J.append(learner.mse(XviP,Yvi))
               mseCrossValidation.append(np.mean(J))
```

B) Plot of degree (x-axis) vs cross validation error (y-axis)



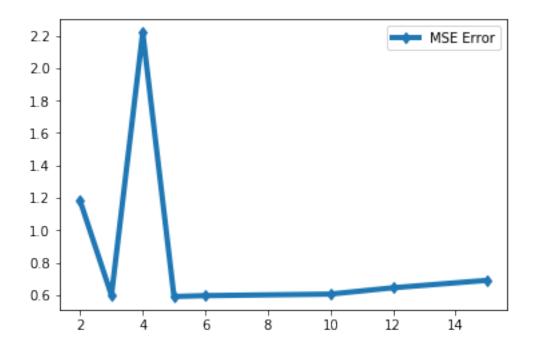
C) Degree with minimum cross-validation error is 5.

Plot for MSE using cross validation, for degree 5:

```
In [133]: J=[]
    mse_error=[]
    nFoldsArray=np.array([2,3,4,5,6,10,12,15])
    for i,Folds in enumerate(nFoldsArray):
        J.clear()
        for iFold in range(nFoldsArray[i]):
            Xti,Xvi,Yti,Yvi = ml.crossValidate(Xtr,Ytr,nFoldsArray[i],iFold) # take ith
            XtiP=ml.transforms.fpoly(Xti,5,False);
            XviP=ml.transforms.fpoly(Xvi,5,False);
            XtiP,params=ml.transforms.rescale(XtiP)
            XviP,_=ml.transforms.rescale(XviP,params)
            learner = ml.linear.linearRegress(XtiP,Yti)
```

MSE Error:

[1.1795458641313061, 0.5984555010978058, 2.219526156063488, 0.5910703726407606, 0.596338005001



D) Plot of MSE estimated from Cross Validation vs MSE evaluated for Actual Test data

