

# MI\_10\_2\_final

January 24, 2018

```
In [2]: import numpy as np
        from sklearn.svm import NuSVR
        import pandas as pd
        import matplotlib.pyplot as plt
        from sklearn.linear_model import LinearRegression

In [4]: Training = np.loadtxt("TrainingRidge.csv", skiprows=1, delimiter=",", dtype='float')
        Validation = np.loadtxt("ValidationRidge.csv", skiprows=1, delimiter=",", dtype='float')
        Xt = Training[:,0:2]
        Xv = Validation[:,0:2]
        Yt = Training[:, -1]
        Yv = Validation[:, -1]

In [9]: #svr = SVR(kernel="rbf", gamma=0.1)
        #log = LinearRegression()

        #svr.fit(X_train.reshape(-1,1), y_train)
        #log.fit(X_train.reshape(-1,1), y_train)

        #-SVR on the training set
        svr_rbf = NuSVR()
        y_rbf = svr_rbf.fit(Xt, Yt).predict(Xv)

        #plots

        #v-SVR
        plt.figure()
        plt.title("Scatter Plot v-SVR")
        plt.xlabel("x1")
        plt.ylabel("x2")
        plt.scatter(Xv[:,0], Xv[:,1], c = y_rbf)

        #adding training set to plot
        plt.scatter(Xt[:,0], Xt[:,1], c = Yt, marker = "s")
        plt.show()

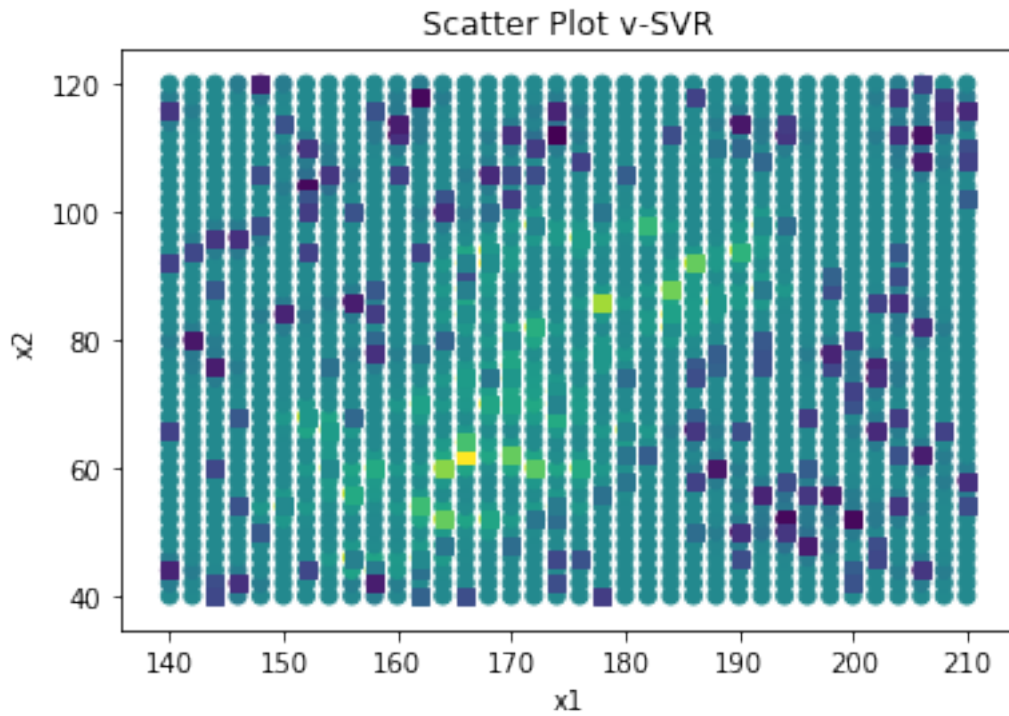
        #Validation Set
```

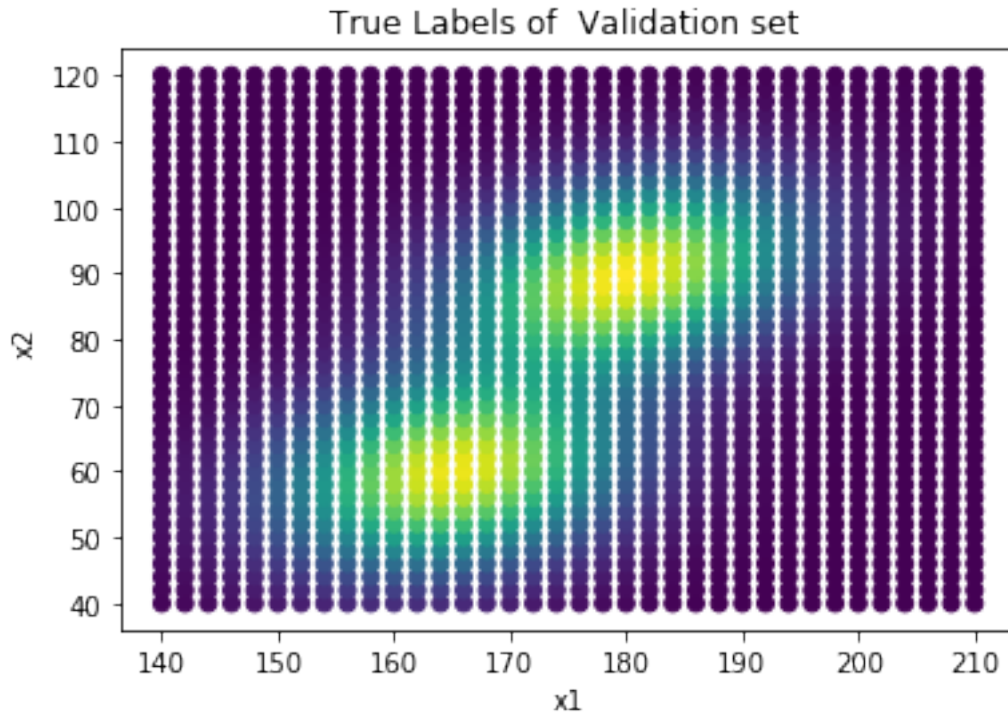
```

plt.figure()
plt.title("True Labels of Validation set ")
plt.xlabel("x1")
plt.ylabel("x2")
plt.scatter(Xv[:,0],Xv[:,1],c = Yv)
plt.show()

#total mean square error
MSE = np.sum((y_rbf-Yv)**2)
print("total mean square error")
print(MSE)

```





total mean square error  
113343.926488

```
In [4]: ###
        #b
        C_value = np.arange(-2, 12, 2)
        C_value = np.power(2., C_value)
        gamma_value = np.arange(-12, 2, 2)
        gamma_value = np.power(2., gamma_value)

        nsubset = 10

        mean_training_matrix = np.zeros((len(C_value), len(gamma_value)))
        crossvalidation_matrix = mean_training_matrix.copy()
        #print np.shape(mean_training_matrix), len(C_value), len(gamma_value)

        #training_results = np.zeros(nsubset)

        one_slice = len(Training)/nsubset
        i =0
        n_subsets = 10
        data = Training
```

```

def k_fold(data, n_subsets, c, g):
    training_results = np.zeros(n_subsets)
    test_results = training_results.copy()

    one_slice = len(data)/n_subsets
    rest = len(data) - one_slice
    for i in range(0, n_subsets):
        #Split data as training and test
        #print(one_slice)
        training = np.append(data[0:20*i, :], data[20*(i+1):, :], axis = 0)
        Yt = training[:, -1]
        test = data[20*i: 20*(i+1), :]
        #rbfclassif = SVC(C = c, kernel = 'rbf', gamma = g)
        #rbfclassif.fit(training[:, 0:2], training[:, 2])

        svr_rbf = NuSVR(kernel='rbf', C=c, gamma=g)
        y_rbf = svr_rbf.fit(training, Yt).predict(test)

        #y_pred_train = svr_rbf.fit(training, Yt).predict(training)
        #y_pred_test = svr_rbf.fit(training, Yt).predict(test)
        #training_results[i] = (np.sum(np.abs(y_pred_train - training[:, 2]) - 1) * -1)
        #test_results[i] = (np.sum(np.abs(y_pred_test - test[:, 2]) - 1) * -1)
        y_pred_train = svr_rbf.fit(training, Yt).predict(Xt)
        y_pred_test = svr_rbf.fit(training, Yt).predict(Xv)

        training_results = (np.sum(np.abs(y_pred_train - Training[:, -1]) - 1) * -1)
        test_results = (np.sum(np.abs(y_pred_test - Validation[:, -1]) - 1) * -1)

    return np.average(training_results)/rest, np.average(test_results)/one_slice

for i in range(0, len(C_value)):
    for j in range(0, len(gamma_value)):
        mean_tr, cv = k_fold(Xt, nsubset, C_value[i], gamma_value[j]) #return mean tra
        mean_training_matrix[i, j] = mean_tr
        crossvalidation_matrix[i, j] = cv

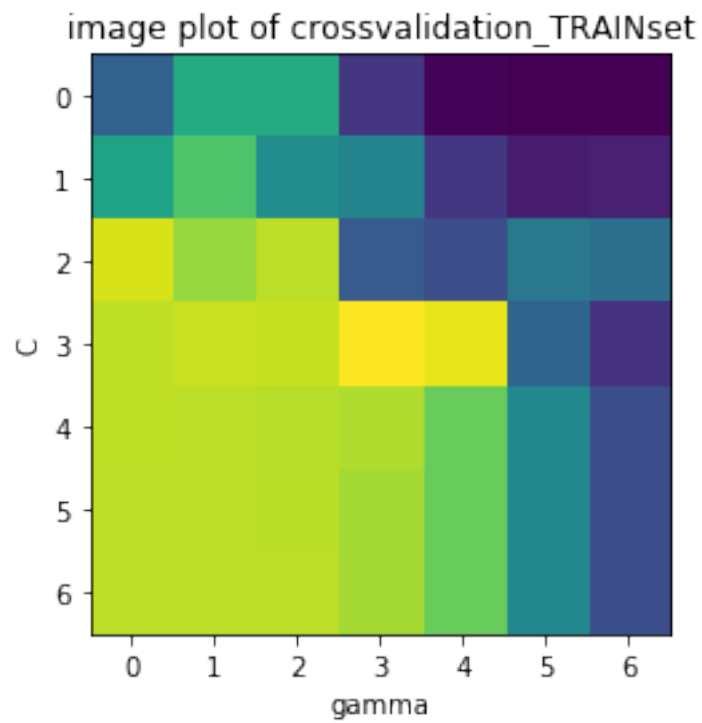
```

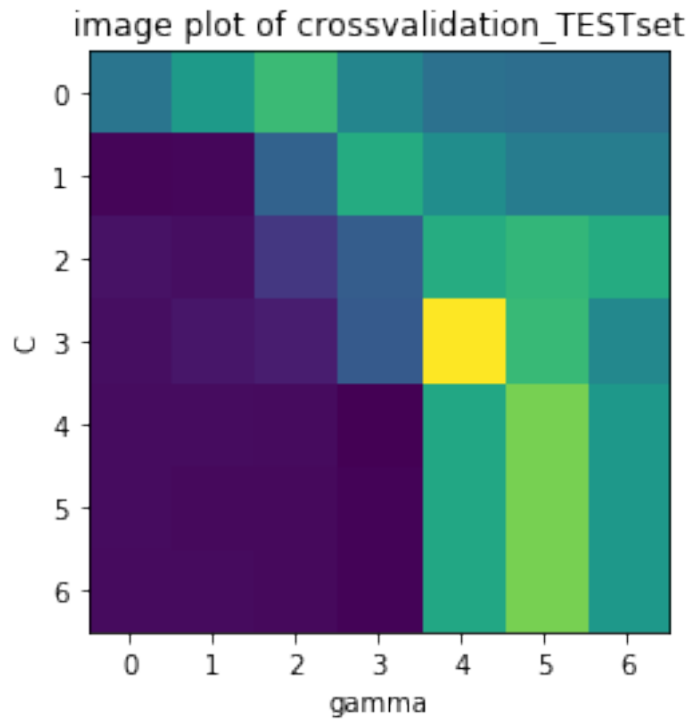
In [11]: plt.figure()  
plt.title("image plot of crossvalidation\_TRAINset")  
plt.imshow(mean\_training\_matrix)  
plt.xlabel("gamma")  
plt.ylabel("C")

```
plt.figure()
plt.title("image plot of crossvalidation_TESTset")
plt.imshow(crossvalidation_matrix)
plt.xlabel("gamma")
plt.ylabel("C")

plt.show()
```

*#brighter area is signalling a lower MSE*





In [14]: #c

```
#from previous task, we could see, that approximately on C[3] = 16 and gamma[4] = 0.0
svr_final = NuSVR(kernel='rbf', C=16, gamma=0.0625)
y_rbf = svr_rbf.fit(Xt, Yt).predict(Xv)
```

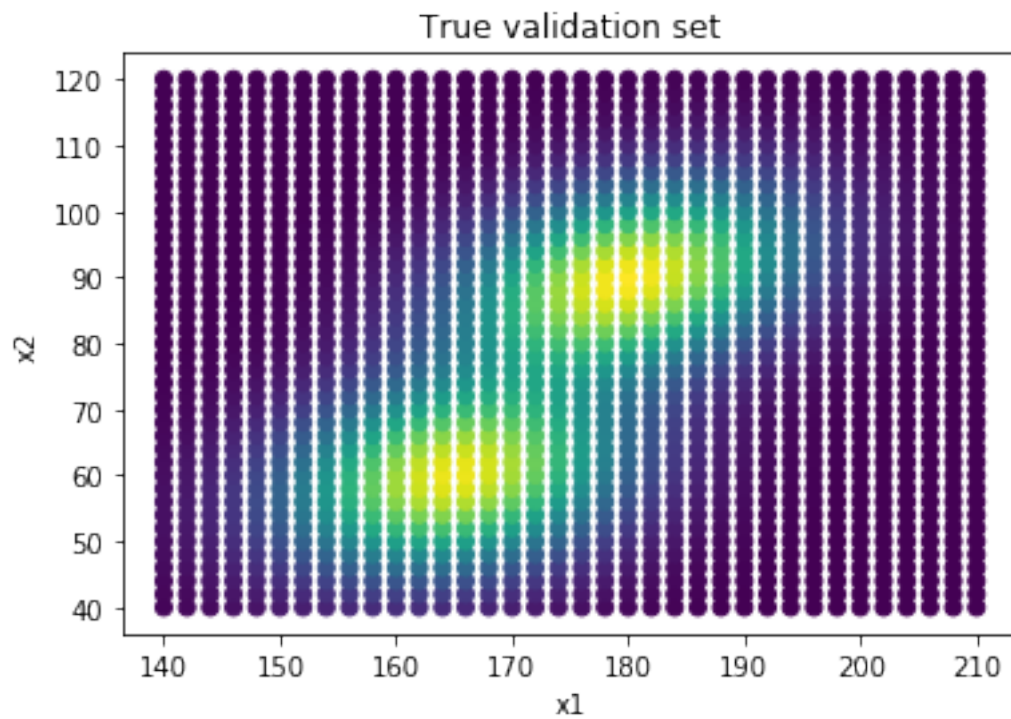
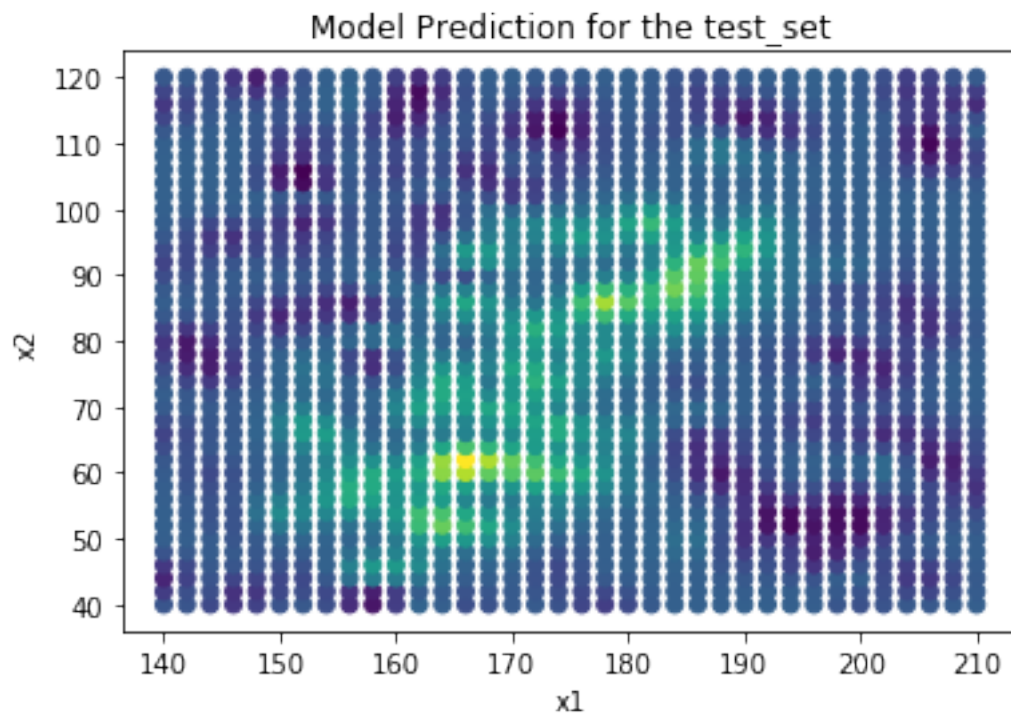
```
plt.figure()
plt.title("Model Prediction for the test_set")
plt.xlabel("x1")
plt.ylabel("x2")
plt.scatter(Xv[:,0],Xv[:,1],c = y_rbf)
```

```
plt.figure()
plt.title("True validation set")
plt.xlabel("x1")
plt.ylabel("x2")
plt.scatter(Xv[:,0],Xv[:,1],c = Yv)
```

```
plt.show()
```

```
#total mean square error
MSE = np.sum((y_rbf-Yv)**2)
```

```
print("total mean square error")  
print(MSE)
```



total mean square error  
37723.1401613