

Homework3

March 6, 2022

```
[1]: import matplotlib.pyplot as plt
# ~~~ pyforest auto-imports - don't write above this line
import numpy as np
import pandas as pd
```

```
[2]: data=pd.read_csv('G:
↳\Documents\ReferenceBooks\MachineLearning\Rohban\Homework\HW3\TrainPreprocessed.
↳csv')
data_test = pd.read_csv('G:
↳\Documents\ReferenceBooks\MachineLearning\Rohban\Homework\HW3\TestPreprocessed.
↳csv')
```

```
[3]: data
```

```
[3]:      BsmtQual  1stFlrSF  MasVnrArea  SaleCondition_Partial  \
0      -0.623056 -0.793162    0.511243                -0.30589
1      -0.623056  0.257052   -0.574214                -0.30589
2      -0.623056 -0.627611    0.322950                -0.30589
3       0.865071 -0.521555   -0.574214                -0.30589
4      -0.623056 -0.045596    1.364102                -0.30589
...      ...      ...      ...      ...
1455 -0.623056 -0.542249   -0.574214                -0.30589
1456 -0.623056  2.354894    0.084814                -0.30589
1457  0.865071  0.065634   -0.574214                -0.30589
1458  0.865071 -0.218907   -0.574214                -0.30589
1459  0.865071  0.241532   -0.574214                -0.30589

      GarageType_Detchd  OverallQual  GarageYrBlt  FullBath  TotalBsmtSF  \
0                -0.600353    0.651256    1.020807  0.789470    -0.459145
1                -0.600353   -0.071812   -0.104447  0.789470     0.466305
2                -0.600353    0.651256    0.937455  0.789470   -0.313261
3                1.664545    0.651256    0.812427 -1.025689   -0.687089
4                -0.600353    1.374324    0.895779  0.789470     0.199611
...      ...      ...      ...      ...
1455                -0.600353   -0.071812    0.854103  0.789470   -0.238040
1456                -0.600353   -0.071812   -0.021095  0.789470     1.104547
1457                -0.600353    0.651256   -1.563110  0.789470     0.215567
```

1458	-0.600353	-0.794879	-1.188025	-1.025689	0.046889
1459	-0.600353	-0.794879	-0.562884	-1.025689	0.452629

	YearBuilt	...	BsmtFinSF1	GarageFinish	Neighborhood_NridgHt	\
0	1.050634	...	0.575228	-0.223588	-0.235877	
1	0.156680	...	1.171591	-0.223588	-0.235877	
2	0.984415	...	0.092875	-0.223588	-0.235877	
3	-1.862993	...	-0.499103	1.041440	-0.235877	
4	0.951306	...	0.463410	-0.223588	-0.235877	
...	
1455	0.918196	...	-0.972685	-0.223588	-0.235877	
1456	0.222899	...	0.759399	1.041440	-0.235877	
1457	-1.002149	...	-0.369744	-0.223588	-0.235877	
1458	-0.704164	...	-0.865252	1.041440	-0.235877	
1459	-0.207523	...	0.847099	-1.488617	-0.235877	

	MasVnrType_None	MSSubClass_60.0	GarageCars	GarageArea	\
0	-1.203608	1.969844	0.311618	0.350880	
1	0.830266	-0.507307	0.311618	-0.060710	
2	-1.203608	1.969844	0.311618	0.631510	
3	0.830266	-0.507307	1.649742	0.790533	
4	-1.203608	1.969844	1.649742	1.697903	
...	
1455	0.830266	1.969844	0.311618	-0.060710	
1456	-1.203608	-0.507307	0.311618	0.126376	
1457	0.830266	-0.507307	-1.026506	-1.033560	
1458	0.830266	-0.507307	-1.026506	-1.089686	
1459	0.830266	-0.507307	-1.026506	-0.921308	

	FireplaceQu_IsNull	GrLivArea	SalePrice
0	1.056020	0.370207	208500.0
1	-0.946303	-0.482347	181500.0
2	-0.946303	0.514836	223500.0
3	-0.946303	0.383528	140000.0
4	-0.946303	1.298881	250000.0
...
1455	-0.946303	0.250316	175000.0
1456	-0.946303	1.061003	210000.0
1457	-0.946303	1.569110	266500.0
1458	1.056020	-0.832502	142125.0
1459	1.056020	-0.493765	147500.0

[1460 rows x 28 columns]

```
[38]: class GaussianKernels:
      def __init__(self, sigma, X_train, Y_train):
          self.X_train = X_train
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        self.Y_train = Y_train
        self.sigma = sigma

    def calculate_kernel(self,x_i,x_j):
        return 1/(np.sqrt(2*np.pi))*np.exp(-0.5*np.square(np.linalg.norm(x=x_i-
↪ x_j,axis=-1))/self.sigma)

    def predict(self,x_test):
        kernels=self.calculate_kernel(self.X_train,x_test)
        if np.sum(kernels)==0:
            return self.Y_train.mean()
        weights = kernels / np.sum(kernels)
        return np.dot(weights,self.Y_train)

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[39]: class IndicatorKernels:
    def __init__(self,h,X_train,Y_train):
        self.X_train = X_train
        self.Y_train = Y_train
        self.h = h

    def calculate_kernel(self,x_i,x_j):
        return np.where(np.abs(np.linalg.norm(x=x_i - x_j,axis=-1)) <= self.
↪ h,1,0)

    def predict(self,x_test):
        kernels=self.calculate_kernel(self.X_train,x_test)
        if np.sum(kernels)==0:
            return self.Y_train.mean()
        weights = kernels / np.sum(kernels)
        return np.dot(weights,self.Y_train)

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[35]: mask = np.random.rand(len(data)) <= 0.85
train =data[mask]
test = data[~mask]

X_train = train.drop(['SalePrice'],axis=1).to_numpy().astype(np.
↪ dtype('float64'))
X_test = test.drop(['SalePrice'],axis=1).to_numpy().astype(np.dtype('float64'))
Y_train = train['SalePrice'].to_numpy().astype(np.dtype('float64'))
Y_test = test['SalePrice'].to_numpy().astype(np.dtype('float64'))

```

```

[45]: answers = {}
for h in np.linspace(1,10,300):
    kernel_regressor=IndicatorKernels(h,X_train,Y_train)
    sum = 0
    predicted=[]
    for test in X_test:

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        y_prediction=kernel_regressor.predict(test)
        predicted.append(y_prediction)

    predicted = np.array(predicted)
    result = np.sqrt(np.mean(np.square(predicted-Y_test)))
    answers[h] = result

plt.figure(figsize=(10,10))
answers_np=np.array(list(answers.items()))
plt.title('Kernel')
plt.xlabel('h')
plt.ylabel('RMSE')
print(np.nanmin(answers_np[:,1]))
plt.plot(answers_np[:,0],answers_np[:,1])

min_indicator = np.argmin(answers_np[:,1])
print(answers_np[min_indicator,:])

```

<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

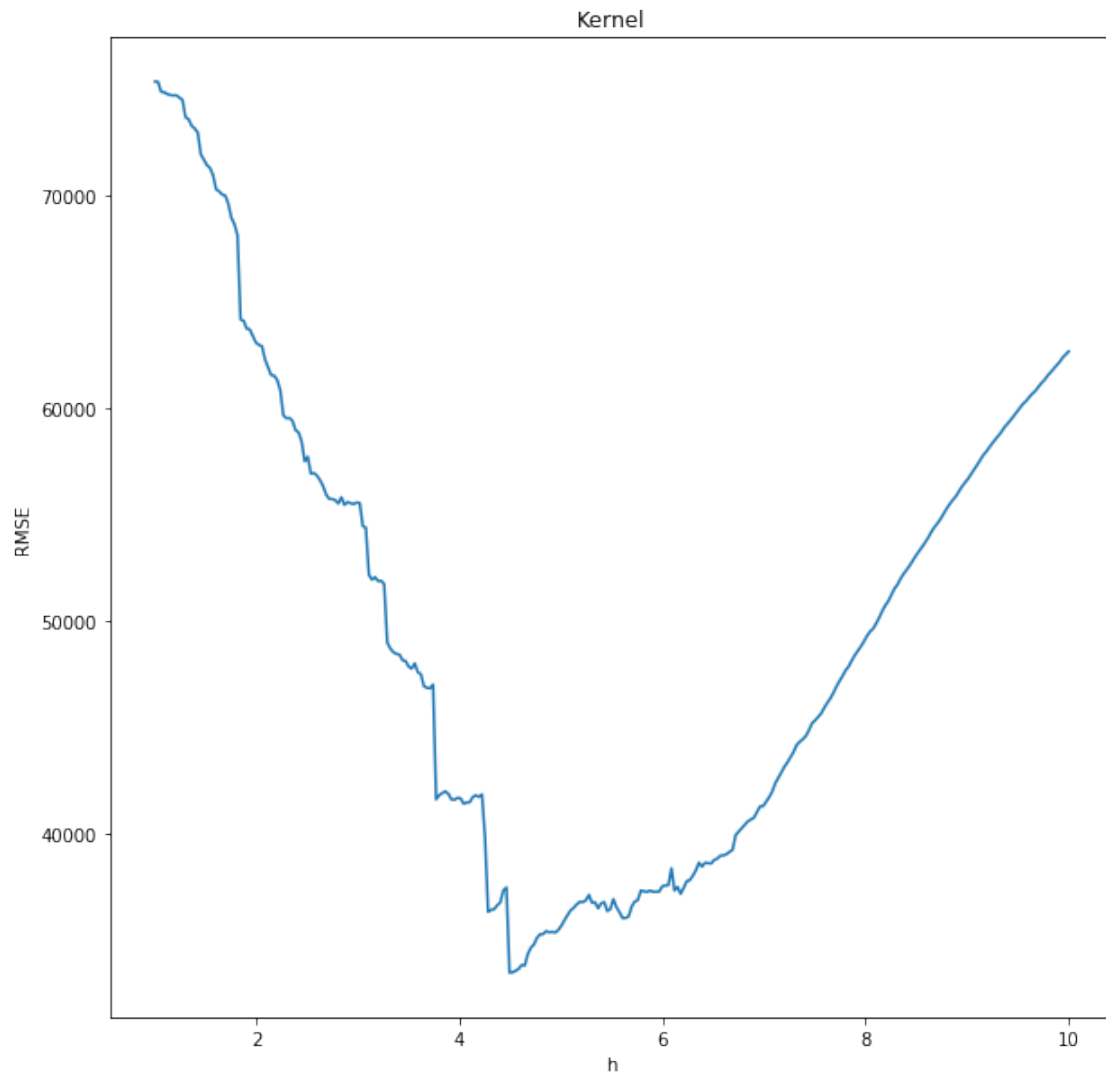
<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

33498.15061130628

<IPython.core.display.Javascript object>

[4.49163880e+00 3.34981506e+04]



```
[46]: answers = {}
for h in np.linspace(0,5,300):
    kernel_regressor=GaussianKernels(h,X_train,Y_train)
    sum = 0
    predicted=[]
    for test in X_test:
        y_prediction=kernel_regressor.predict(test)
        predicted.append(y_prediction)

    predicted = np.array(predicted)
    result = np.sqrt(np.mean(np.square(predicted-Y_test)))
    answers[h] = result

plt.figure(figsize=(10,10))
```

```

answers_np2=np.array(list(answers.items()))
plt.title('Kernel')
plt.xlabel('h')
plt.ylabel('RMSE')
print(np.nanmin(answers_np2[:,1]))
plt.plot(answers_np2[:,0],answers_np2[:,1])

min_gaussian = np.nanargmin(answers_np2[:,1])
print(answers_np2[min_gaussian,:])

```

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:8:

RuntimeWarning: divide by zero encountered in true_divide

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:8:

RuntimeWarning: invalid value encountered in true_divide

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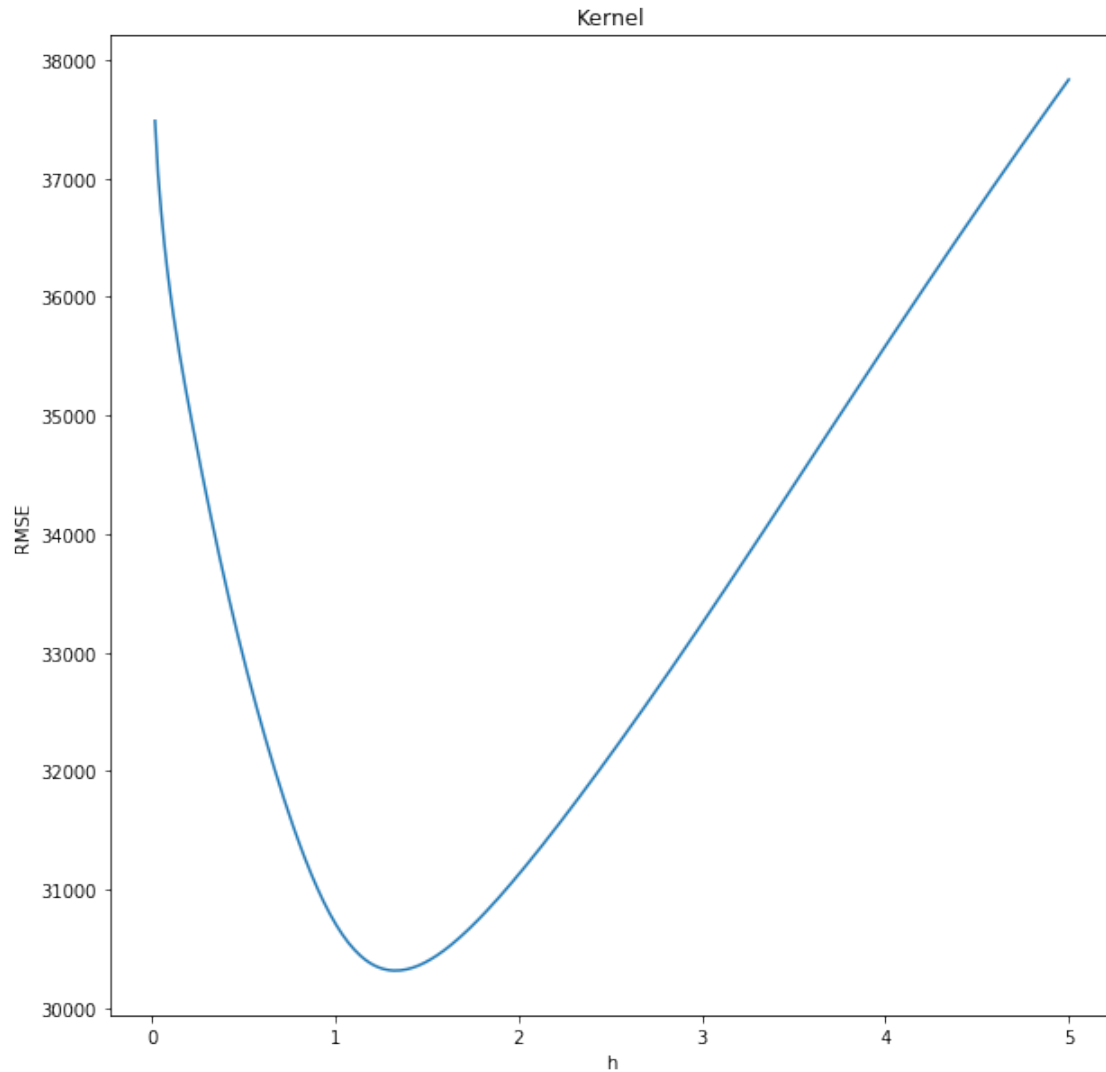
<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

30318.27613520855

<IPython.core.display.Javascript object>

[1.32107023e+00 3.03182761e+04]



```
[50]: kernel_regressor=IndicatorKernels(answers_np[min_indicator,0],X_train,Y_train)
predictionOfIndicator = []
for row in data_test.to_numpy().astype(np.dtype('float64')):
    predictionOfIndicator.append(kernel_regressor.predict(row))
predictionOfIndicatorNumpy = np.array(predictionOfIndicator)
```

```
[54]: kernel_gaussian=GaussianKernels(answers_np2[min_gaussian,0],X_train,Y_train)
predictionOfGuassian = []
for row in data_test.to_numpy().astype(np.dtype('float64')):
    predictionOfGuassian.append(kernel_gaussian.predict(row))
predictionOfGaussianNumpy = np.array(predictionOfGuassian)
```

0.1 Plot of test results

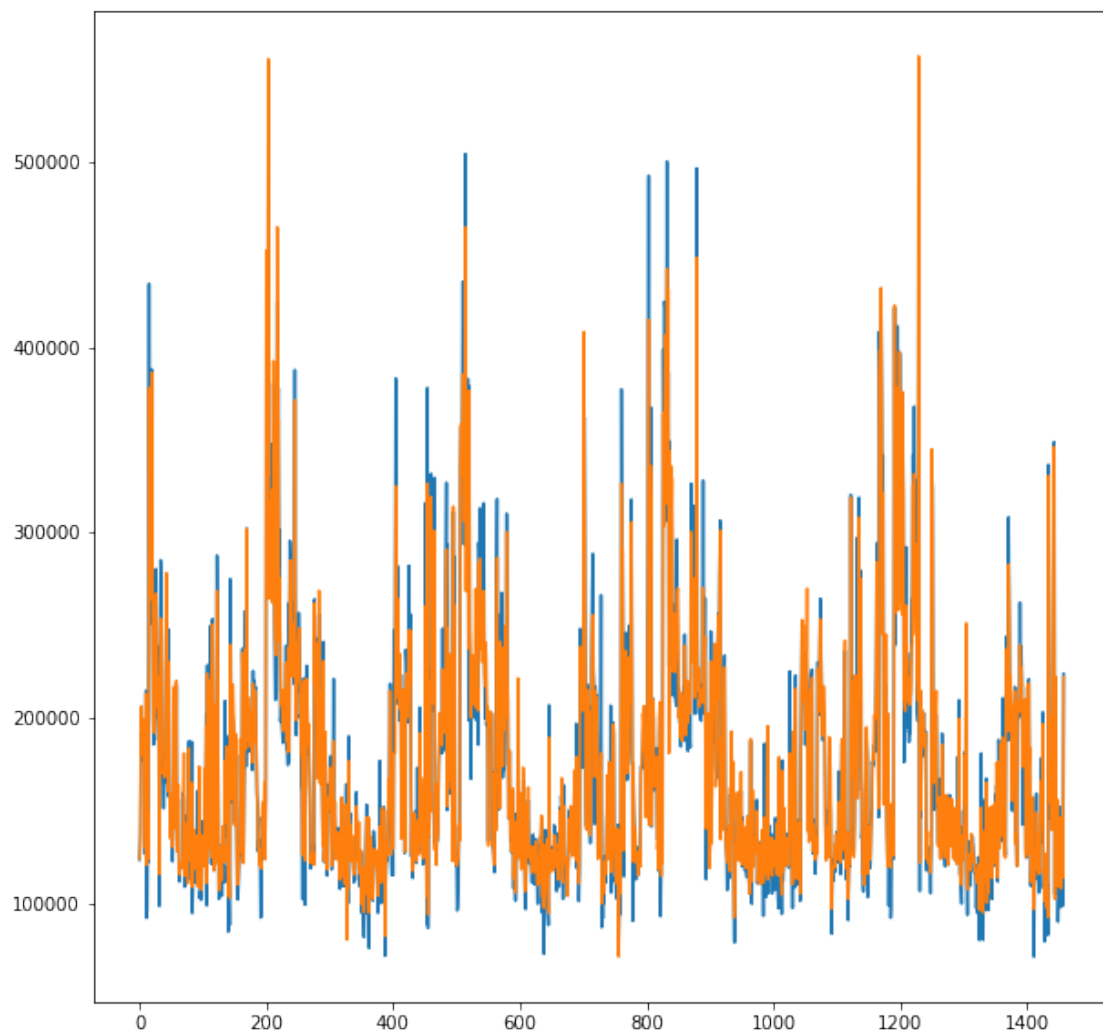
```
[68]: plt.figure(figsize=(10,10))  
      plt.plot(predictionOfGaussianNumpy)  
      plt.plot(predictionOfIndicatorNumpy)
```

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<IPython.core.display.Javascript object>

```
[68]: [<matplotlib.lines.Line2D at 0x1b6379101c8>]
```



0.2 plot of best results of gaussian kernel vs actual result

```
[71]: kernel_regressor=IndicatorKernels(answers_np[min_indicator,0],X_train,Y_train)
      predictionOfIndicator = []
      for row in X_test:
          predictionOfIndicator.append(kernel_regressor.predict(row))
      predictionOfIndicatorNumpy = np.array(predictionOfIndicator)

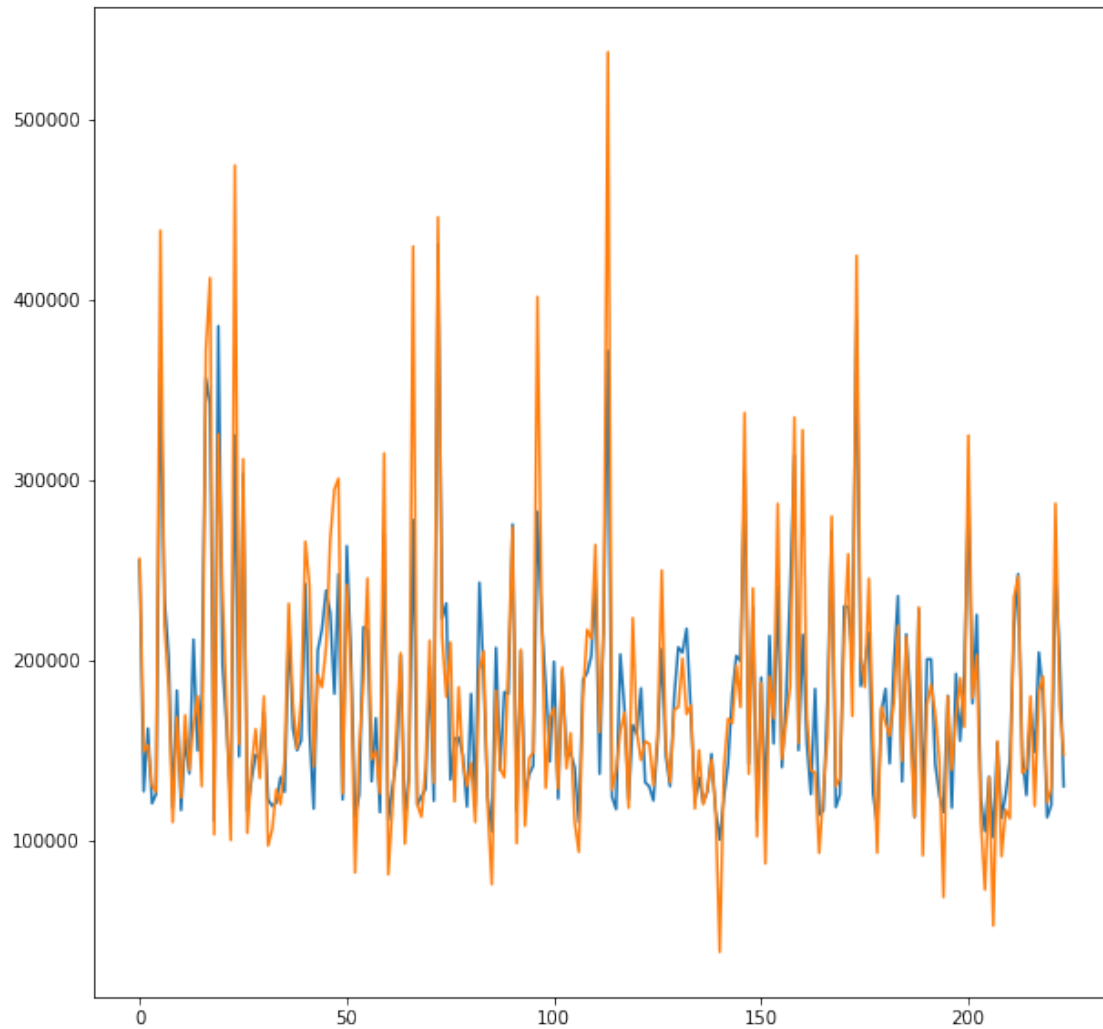
      plt.figure(figsize=(10,10))
      plt.plot(predictionOfIndicatorNumpy)
      plt.plot(Y_test)
```

<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

```
[71]: [<matplotlib.lines.Line2D at 0x1b63793c1c8>]
```



0.3 plot of best results of indicator kernel vs actual result

```
[72]: kernel_gaussian=GaussianKernels(answers_np2[min_gaussian,0],X_train,Y_train)
predictionOfGuassian = []
for row in X_test:
    predictionOfGuassian.append(kernel_gaussian.predict(row))
predictionOfGaussianNumpy = np.array(predictionOfGuassian)

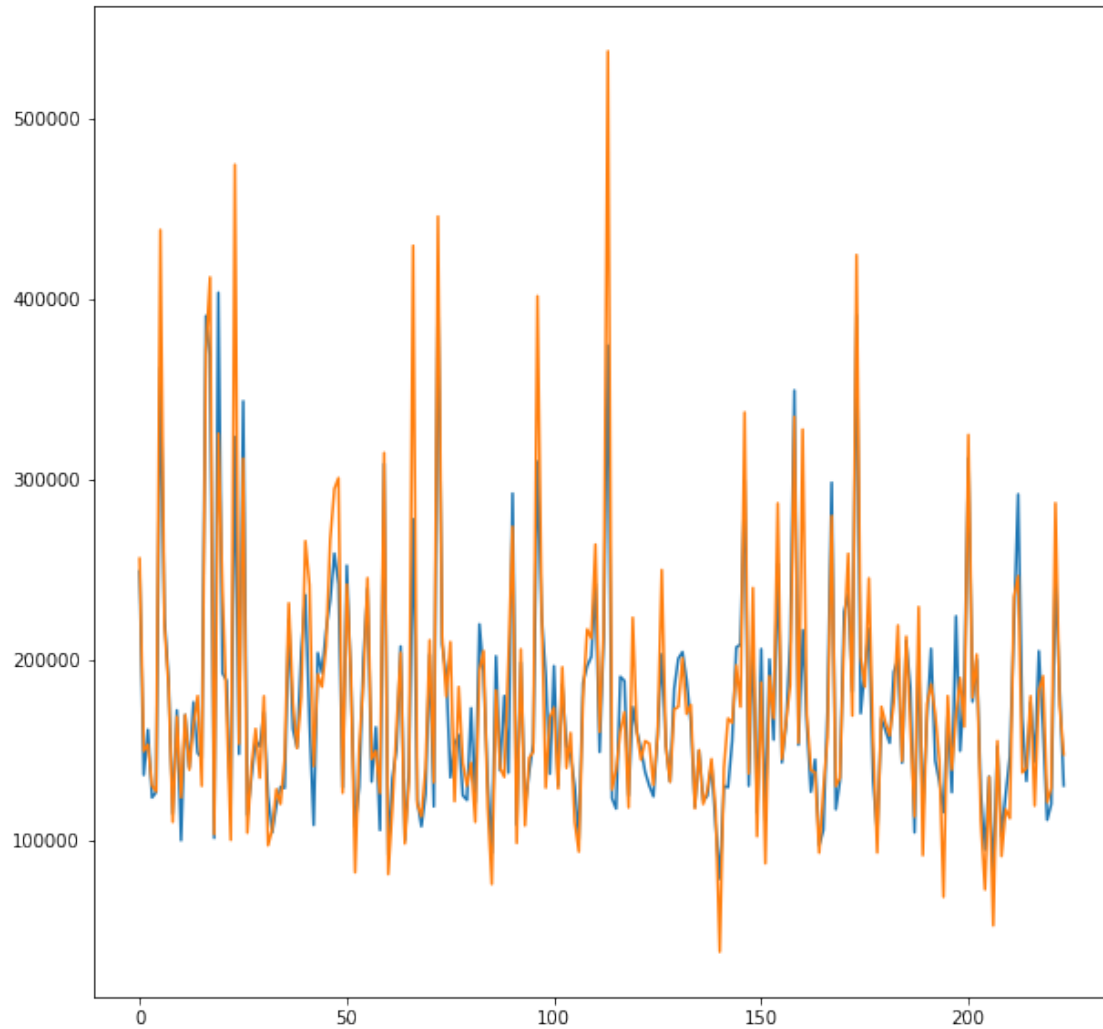
plt.figure(figsize=(10,10))
plt.plot(predictionOfGaussianNumpy)
plt.plot(Y_test)
```

<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

<IPython.core.display.Javascript object>

[72]: [<matplotlib.lines.Line2D at 0x1b6376e9d48>]



```
[73]: np.savetxt("G:  
↪\Documents\ReferenceBooks\MachineLearning\Rohban\Homework\HW3\TestResultIndicator.  
↪csv",predictionOfIndicatorNumpy , delimiter=",")
```

```
[74]: np.savetxt("G:  
↪\Documents\ReferenceBooks\MachineLearning\Rohban\Homework\HW3\TestResultGaussian.  
↪csv",predictionOfGaussianNumpy , delimiter=",")
```

```
[ ]:
```