

# Boston\_Dataset - Analysis

January 17, 2022

```
[1]: import warnings
warnings.filterwarnings('ignore')
```

```
[2]: import scrapbook as sb
import pandas as pd
import numpy as np
import seaborn as sns
import numpy as np
from statistics import mean
import matplotlib.pyplot as plt
```

## 1 Catboost Baseline

```
[3]: books = sb.read_notebooks("./BaseLine_Model_Output")
baseLine_data = []
for nb in books.notebooks:
    nbList=[nb.scrap['Catboost MSE'].data]
    baseLine_data.append(nbList)
df = pd.DataFrame(baseLine_data, columns = ["Catboost"])
baseLine_data = np.reshape(baseLine_data,(1,10))[0]
display(df)
mse = mean(baseLine_data)
print("Average MSE (Catboost Model): "+ str(mse))
```

```
    Catboost
0  0.081287
1  0.115721
2  0.120925
3  0.196560
4  0.065258
5  0.116890
6  0.117933
7  0.094344
8  0.086313
9  0.085721
```

```
Average MSE (Catboost Model): 0.10809530632925027
```

## 2 GAN Analysis

```
[4]: book = sb.read_notebooks("./GAN_Output")
gan_data = []
gan_mse = []
for nb in book.notebooks:
    metrics = nb.scrapes['GAN_1 Metrics'].data
    for i in range(1000):
        gan_mse.append(metrics[0][i])
    nbList = [nb.scrapes['GAN Model MSE'].data,
              nb.scrapes['GAN Model MAE'].data,
              nb.scrapes['GAN Model Euclidean distance'].data,
              nb.scrapes['GAN Model Manhattan Distance'].data]
    gan_data.append(nbList)

df = pd.DataFrame(gan_data, columns = ['MSE', 'MAE', 'Euclidean_
→Distance', 'Manhattan Distance'])
display(df.style)
print("MEAN:")
print(df.mean(axis = 0))
gan_data = np.array(gan_data)
```

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MEAN:

MSE	0.241390
MAE	0.319134
Euclidean Distance	4.838867
Manhattan Distance	32.232509

dtype: float64

## 3 ABC\_GAN Analysis

### 3.1 ABC Pre-generator

1. Prior Model is Catboost Model
2. ABC Pre-generator is Catboost Model with gaussian noise ->  $N(0, \text{variance})$  where variance : 1, 0.1, 0.01

```
[5]: book = sb.read_notebooks("./ABC_GAN_Catboost_Output")
paramVal = [1, 0.1, 0.01]
abc_mse = [[] for i in range(3)]
abc_mse_skip = [[] for i in range(3)]
abc_mse_mean = [[] for i in range(3)]
abc_mse_skip_mean = [[] for i in range(3)]
abc_weights = [[] for i in range(3)]
prior_model = [[] for i in range(3)]
abc_pre_generator = [[] for i in range(3)]
```

```

for nb in book.notebooks:
    metrics1 = np.array(nb.scrap['ABC_GAN_1 Metrics'].data)
    metrics3 = np.array(nb.scrap['ABC_GAN_3 Metrics'].data)
    paramVar = float(nb.papermill_dataframe.iloc[0]['value'])

    #Divide data according to parameters
    for i in range(3):
        if paramVar == paramVal[i]:
            for j in range(100):
                abc_mse[i].append(metrics1[0,j])
                abc_mse_skip[i].append(metrics3[0,j])
            abc_weights[i].append(nb.scrap['Skip Connection Weight'].data)
            prior_model[i].append(nb.scrap['Prior Model MSE'].data)
            abc_pre_generator[i].append(nb.scrap['ABC Pre-generator MSE'].data)
            abc_mse_mean[i].append(mean(metrics1[0,:]))
            abc_mse_skip_mean[i].append(mean(metrics3[0,:]))

```

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[6]: for i in range(3):
    data = []
    for j in range(len(abc_weights[i])):
        data.append([paramVal[i],prior_model[i][j],
                    abc_pre_generator[i][j],abc_weights[i][j],
                    abc_mse_mean[i][j],abc_mse_skip_mean[i][j]])

    df = pd.DataFrame(data, columns = ['Variance','Prior Model MSE',
                                     'ABC pre-generator MSE','Skip Node_
↳weight',
                                     'ABC_GAN MSE','ABC_GAN MSE (skip_
↳connection)'])
    display(df.round(5))
    print(df.mean(axis=0))
    print("-----")

```

	Variance	Prior Model MSE	ABC pre-generator MSE	Skip Node weight \
0	1	0.07664	1.41763	1.00000
1	1	0.09810	1.07649	0.99588
2	1	0.08852	1.17904	1.00000
3	1	0.12121	1.07123	0.99195
4	1	0.11223	0.97762	1.00000
5	1	0.10835	0.98674	0.99846
6	1	0.08598	1.03053	0.96760
7	1	0.10016	1.21188	0.99153
8	1	0.09026	0.98423	1.00000
9	1	0.10565	0.87045	0.99148

  

	ABC_GAN MSE	ABC_GAN MSE (skip connection)
0	1.87390	0.17363

1	0.19311	0.18018
2	0.22607	0.09878
3	0.28449	0.32872
4	0.46896	0.15830
5	0.18262	0.19808
6	0.20929	0.30623
7	0.28374	0.14020
8	0.09356	0.10143
9	0.23253	0.12707

Variance	1.000000
Prior Model MSE	0.098709
ABC pre-generator MSE	1.080584
Skip Node weight	0.993690
ABC_GAN MSE	0.404827
ABC_GAN MSE (skip connection)	0.181262
dtype: float64	

	Variance	Prior Model MSE	ABC pre-generator MSE	Skip Node weight \
0	0.1	0.07862	0.08873	0.12253
1	0.1	0.12962	0.12612	0.00002
2	0.1	0.07715	0.09444	0.02943
3	0.1	0.08187	0.09043	0.00000
4	0.1	0.09539	0.10172	0.01422
5	0.1	0.16040	0.17203	0.01009
6	0.1	0.07169	0.08536	0.00712
7	0.1	0.07879	0.07907	0.00706
8	0.1	0.11022	0.11378	0.04363
9	0.1	0.12492	0.12813	0.02369

	ABC_GAN MSE	ABC_GAN MSE (skip connection)
0	0.16360	0.09379
1	0.51599	0.13948
2	0.93204	0.08948
3	0.45847	0.09221
4	1794.20627	0.10681
5	0.37114	0.17087
6	1.03522	0.08438
7	0.10154	7.49635
8	0.31481	10947.00884
9	0.39105	2.92419

Variance	0.100000
Prior Model MSE	0.100868
ABC pre-generator MSE	0.107982
Skip Node weight	0.025779
ABC_GAN MSE	179.849014
ABC_GAN MSE (skip connection)	1095.820639

dtype: float64

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	Variance	Prior Model MSE	ABC pre-generator MSE	Skip Node weight \
0	0.01	0.12767	0.12715	0.00000
1	0.01	0.11476	0.11569	0.00000
2	0.01	0.12687	0.12736	0.00362
3	0.01	0.22559	0.22545	0.02525
4	0.01	0.12226	0.12263	0.04966
5	0.01	0.09726	0.09884	0.14222
6	0.01	0.33550	0.33424	0.09684
7	0.01	0.11077	0.11096	0.23222
8	0.01	0.13501	0.13419	0.00977
9	0.01	0.09971	0.09871	0.12485

	ABC_GAN MSE	ABC_GAN MSE (skip connection)
0	0.12475	1.276700e-01
1	0.14163	1.149100e-01
2	0.16125	1.261300e-01
3	0.43171	5.927900e-01
4	0.13184	1.297300e-01
5	0.12617	7.319190e+06
6	0.30524	1.339171e+05
7	0.16980	6.460739e+08
8	0.69948	1.432400e-01
9	0.11995	9.240118e+04

Variance	1.000000e-02
Prior Model MSE	1.495399e-01
ABC pre-generator MSE	1.495217e-01
Skip Node weight	6.844368e-02
ABC_GAN MSE	2.411803e-01
ABC_GAN MSE (skip connection)	6.536194e+07

dtype: float64

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