Boston

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```
[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, median
import matplotlib.pyplot as plt
```

1 Baseline

```
[3]: books = sb.read_notebooks("./BaseLine_Model_Output")
   baseLine_data = []
   for nb in books.notebooks:
        nbList=[nb.scraps['Stats Model MAE'].data,nb.scraps['Catboost MAE'].data]
        baseLine_data.append(nbList)
   df = pd.DataFrame(baseLine_data, columns = ["Stats Model", "Catboost"])
   baseline_data = np.array(baseLine_data)
   stats = median(baseline_data[:,0])
   catboost = median(baseline_data[:,1])
   display(df)
   print(df.median(axis=0))
```

```
Stats Model Catboost
0
      0.298850 0.192764
1
      0.305628 0.201693
2
     0.435760 0.270210
3
     0.380010 0.262905
4
     0.412824 0.267935
5
     0.356926 0.213024
6
     0.384976 0.209407
7
     0.389513 0.199208
8
      0.384688 0.238885
     0.390258 0.247499
Stats Model
              0.384832
```

Catboost 0.225954

dtype: float64

2 GAN Model

```
[4]: book = sb.read_notebooks("./GAN_Output")
     gan_data = []
     gan_mse = []
     for nb in book.notebooks:
         metrics = nb.scraps['GAN_1 Metrics'].data
         for i in range(1000):
             gan_mse.append(metrics[0][i])
         nbList = [nb.scraps['GAN Model MSE'].data,
                   nb.scraps['GAN Model MAE'].data,
                   nb.scraps['GAN Model Euclidean distance'].data,
                   nb.scraps['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("MEDIAN:")
     print(df.median(axis = 0))
     gan data = np.array(gan data)
     gan_median = median(gan_data[:,1])
    <pandas.io.formats.style.Styler at 0x7f7a0810b5e0>
```

MEDIAN:

MSE 0.141095
MAE 0.258360
Euclidean Distance 3.793474
Manhattan Distance 26.352682

dtype: float64

3 ABC_GAN Analysis

3.1 ABC Pre-generator - Catboost

```
[5]: book = sb.read_notebooks("./ABC_GAN_Catboost")
    paramVal = [1,0.1,0.01]
    abc_mae = [[] for i in range(3)]
    abc_mae_skip = [[] for i in range(3)]
    abc_mae_mean = [[] for i in range(3)]
    abc_mae_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.scraps['ABC_GAN_1 Metrics'].data)
         metrics3 = np.array(nb.scraps['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_mae[i].append(metrics1[1,j])
                     abc_mae_skip[i].append(metrics3[1,j])
                 abc_weights[i].append(nb.scraps['Skip Connection Weight'].data)
                 prior_model[i].append(nb.scraps['Prior Model MAE'].data)
                 abc_pre_generator[i].append(nb.scraps['ABC Pre-generator MAE'].data)
                 abc_mae_skip_mean[i].append(mean(metrics3[1,:]))
                 abc_mae_mean[i].append(mean(metrics1[1,:]))
[6]: data = [[] for i in range(3)]
     data_median_catboost = [[] for i in range(3)]
     for i in range(3):
         for j in range(len(abc_weights[i])):
             data[i].append([prior_model[i][j],paramVal[i],
     →abc pre generator[i][j],abc weights[i][j],abc mae mean[i][j],abc mae skip mean[i][j]])
         df = pd.DataFrame(data[i], columns = ['Baseline','Variance','Prior_
      →Model','Weight','ABC_GAN','Skip_GAN'])
         data median catboost[i] = [ df['Baseline'].median(),df['Variance'].
      →median(), df['Prior Model'].median(),
                             df['ABC_GAN'].median(), df['Skip_GAN'].
      →median(),df['Weight'].median()]
     print(data_median_catboost)
    [[0.22586972779979714, 1.0, 0.8373863917342285, 0.2511735610103271,
    0.24766350942773416, 0.761081874370575], [0.21455946827434413, 0.1,
    0.22733163730941142, 0.21345637898691291, 0.2254030777299982,
    0.3514516055583954], [0.2050931549501881, 0.01, 0.2050996628560008,
    0.22331805785834863, 0.2063114587503047, 0.019863920286297798]]
    3.2 ABC Pre-generator - Stats
[7]: book = sb.read notebooks("./ABC GAN Stats")
     paramVal = [1,0.1,0.01]
     abc mae = [[] for i in range(3)]
```

abc_mae_skip = [[] for i in range(3)]

```
abc_mae_mean = [[] for i in range(3)]
     abc_mae_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.scraps['ABC_GAN_1 Metrics'].data)
         metrics3 = np.array(nb.scraps['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_mae[i].append(metrics1[1,j])
                     abc_mae_skip[i].append(metrics3[1,j])
                 abc_weights[i].append(nb.scraps['Skip Connection Weight'].data)
                 prior_model[i].append(nb.scraps['Prior Model MAE'].data)
                 abc_pre_generator[i].append(nb.scraps['ABC Pre-generator MAE'].data)
                 abc_mae_skip_mean[i].append(mean(metrics3[1,:]))
                 abc_mae_mean[i].append(mean(metrics1[1,:]))
[8]: data = [[] for i in range(3)]
     data_median_stats = [[] for i in range(3)]
     for i in range(3):
         for j in range(len(abc_weights[i])):
             data[i].append([paramVal[i],prior_model[i][j],
     →abc_pre_generator[i][j],abc_weights[i][j],abc_mae_mean[i][j],abc_mae_skip_mean[i][j]])
         df = pd.DataFrame(data[i], columns = ['Variance', 'Baseline', 'Prior_
     →Model','Weight','ABC_GAN','Skip_GAN'])
         data_median_stats[i] = [df['Baseline'].median(),df['Variance'].median(),u

→df['Prior Model'].median(), df['ABC_GAN'].median(), df['Skip_GAN'].median(),
                             df['Weight'].median()]
     print(data_median_stats)
    [[0.36067681751436, 1.0, 0.8589699324034119, 0.2606692294421806,
    0.24868939553009228, 0.9962700307369232], [0.3885386412937586, 0.1,
    0.39483391314328864, 0.27145286795697815, 0.2598036163691135,
    0.5945071876049042], [0.35825527768550247, 0.01, 0.357203127103188,
    0.27940214876403263, 0.2595625594518063, 0.15424392372369766]]
[9]: #Output Table
     output_data_stats = [[stats, gan_median] for i in range(3)]
     for i in range(3):
```

```
for j in range(6):
        output_data_stats[i].append(data_median_stats[i][j])
df_stats = pd.DataFrame(output_data_stats,__

→columns=['Baseline','GAN','Baseline','Variance','Prior

□
 →Model','ABC_GAN','Skip_GAN','Weight'])
output_data_catboost = [[catboost, gan_median] for i in range(3)]
for i in range(3):
    for j in range(6):
        output_data_catboost[i].append(data_median_catboost[i][j])
df_catboost = pd.
 →DataFrame(output_data_catboost,columns=['Baseline','GAN','Baseline','Variance','Prior_
 →Model','ABC_GAN','Skip_GAN','Weight'])
display(df_stats)
display(df_catboost)
  Baseline
                GAN
                    Baseline Variance Prior Model
                                                      ABC_GAN Skip_GAN \
0 0.384832 0.25836
                    0.360677
                                   1.00
                                           0.858970 0.260669 0.248689
1 0.384832 0.25836 0.388539
                                   0.10
                                           0.394834 0.271453 0.259804
2 0.384832 0.25836 0.358255
                                   0.01
                                           0.357203 0.279402 0.259563
    Weight
0 0.996270
1 0.594507
2 0.154244
  Baseline
                GAN Baseline Variance Prior Model
                                                      ABC_GAN Skip_GAN \
0 0.225954 0.25836 0.225870
                                   1.00
                                           0.837386 0.251174 0.247664
1 0.225954
            0.25836 0.214559
                                   0.10
                                           0.227332 0.213456 0.225403
2 0.225954 0.25836 0.205093
                                   0.01
                                           0.205100 0.223318 0.206311
    Weight
0 0.761082
1 0.351452
2 0.019864
```