

Initial Data Exploration

This notebook will be used to explore the dataset and to check key values, changes to be made, overall quality of the data. The following hidden cell is where the data is loaded into the workspace and the csv transformed into a dataframe.

In [8]: *# The code was removed by Watson Studio for sharing.*

The dataframe is renamed and the Id column renamed so that it can be known what that column is.

In [9]: *#Change the dataframe name*
 data_df = df_data_1

#Rename Unnamed: 0 to Id for readability
 data_df.rename(columns = {data_df.columns[0]: 'Id'}, inplace = True)
 data_df.head()

Out[9]:

	Id	timestamp	sensor_00	sensor_01	sensor_02	sensor_03	sensor_04	sensor_05	sensor_06
0	0	2018-04-01 00:00:00	2.465394	47.09201	53.2118	46.310760	634.3750	76.45975	13.41146
1	1	2018-04-01 00:01:00	2.465394	47.09201	53.2118	46.310760	634.3750	76.45975	13.41146
2	2	2018-04-01 00:02:00	2.444734	47.35243	53.2118	46.397570	638.8889	73.54598	13.32465
3	3	2018-04-01 00:03:00	2.460474	47.09201	53.1684	46.397568	628.1250	76.98898	13.31742
4	4	2018-04-01 00:04:00	2.445718	47.13541	53.2118	46.397568	636.4583	76.58897	13.35359

```
In [3]: #Check the types of each of the dataFrames  
data_df.dtypes
```

```
Out[3]: Id                int64  
timestamp              object  
sensor_00              float64  
sensor_01              float64  
sensor_02              float64  
sensor_03              float64  
sensor_04              float64  
sensor_05              float64  
sensor_06              float64  
sensor_07              float64  
sensor_08              float64  
sensor_09              float64  
sensor_10              float64  
sensor_11              float64  
sensor_12              float64  
sensor_13              float64  
sensor_14              float64  
sensor_15              float64  
sensor_16              float64  
sensor_17              float64  
sensor_18              float64  
sensor_19              float64  
sensor_20              float64  
sensor_21              float64  
sensor_22              float64  
sensor_23              float64  
sensor_24              float64  
sensor_25              float64  
sensor_26              float64  
sensor_27              float64  
sensor_28              float64  
sensor_29              float64  
sensor_30              float64  
sensor_31              float64  
sensor_32              float64  
sensor_33              float64  
sensor_34              float64  
sensor_35              float64  
sensor_36              float64  
sensor_37              float64  
sensor_38              float64  
sensor_39              float64  
sensor_40              float64  
sensor_41              float64  
sensor_42              float64  
sensor_43              float64  
sensor_44              float64  
sensor_45              float64  
sensor_46              float64  
sensor_47              float64  
sensor_48              float64  
sensor_49              float64  
sensor_50              float64  
sensor_51              float64
```

```
machine_status    object
dtype: object
```

The data types seem to make sense. All the sensors are float values and the machine_status is object. The machine_status will eventually need to be turned to int and the timestamp to a datetime, if timestamp column is needed.

```
In [4]: #Pandas describe is called to check the mean and other key statistics in the data
data_df.describe()
```

```
Out[4]:
```

	Id	sensor_00	sensor_01	sensor_02	sensor_03	sensor_04
count	220320.000000	210112.000000	219951.000000	220301.000000	220301.000000	220301.000000
mean	110159.500000	2.372221	47.591611	50.867392	43.752481	590.673936
std	63601.049991	0.412227	3.296666	3.666820	2.418887	144.023912
min	0.000000	0.000000	0.000000	33.159720	31.640620	2.798032
25%	55079.750000	2.438831	46.310760	50.390620	42.838539	626.620400
50%	110159.500000	2.456539	48.133678	51.649300	44.227428	632.638916
75%	165239.250000	2.499826	49.479160	52.777770	45.312500	637.615723
max	220319.000000	2.549016	56.727430	56.032990	48.220490	800.000000

```
In [5]: #Due to inconsistent numbers in the describe function, Nan amounts are needed.  
data_df.isna().sum()
```

```
Out[5]: Id          0  
timestamp          0  
sensor_00        10208  
sensor_01         369  
sensor_02         19  
sensor_03         19  
sensor_04         19  
sensor_05         19  
sensor_06        4798  
sensor_07        5451  
sensor_08        5107  
sensor_09        4595  
sensor_10         19  
sensor_11         19  
sensor_12         19  
sensor_13         19  
sensor_14         21  
sensor_15       220320  
sensor_16         31  
sensor_17         46  
sensor_18         46  
sensor_19         16  
sensor_20         16  
sensor_21         16  
sensor_22         41  
sensor_23         16  
sensor_24         16  
sensor_25         36  
sensor_26         20  
sensor_27         16  
sensor_28         16  
sensor_29         72  
sensor_30        261  
sensor_31         16  
sensor_32         68  
sensor_33         16  
sensor_34         16  
sensor_35         16  
sensor_36         16  
sensor_37         16  
sensor_38         27  
sensor_39         27  
sensor_40         27  
sensor_41         27  
sensor_42         27  
sensor_43         27  
sensor_44         27  
sensor_45         27  
sensor_46         27  
sensor_47         27  
sensor_48         27  
sensor_49         27  
sensor_50       77017  
sensor_51       15383
```

```
machine_status      0  
dtype: int64
```

As shown by the previous cell there is a decent amount of nans. Sensor_15 will need to be removed as it has no data and will not benefit from being replaced by zeros. The others will need to have to be replaced by zeros or more likely the mean of the sensor data.

```
In [6]: import seaborn as sns
from matplotlib import rcParams

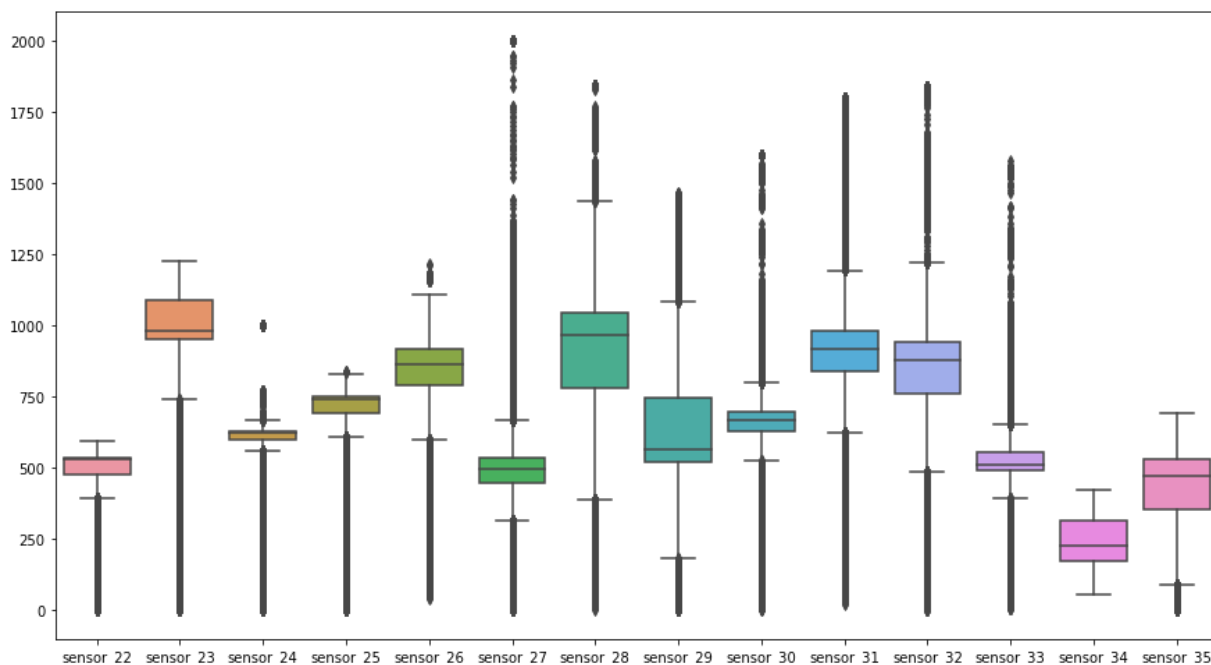
# figure size in inches
rcParams['figure.figsize'] = 15,8.27
box_data = data_df.loc[:, 'sensor_22':'sensor_35']
print(box_data.head())
sns.boxplot(data = box_data)
```

	sensor_22	sensor_23	sensor_24	sensor_25	sensor_26	sensor_27	\
0	498.8926	975.9409	627.6740	741.7151	848.0708	429.0377	
1	498.8926	975.9409	627.6740	741.7151	848.0708	429.0377	
2	501.3617	982.7342	631.1326	740.8031	849.8997	454.2390	
3	499.0430	977.7520	625.4076	739.2722	847.7579	474.8731	
4	498.5383	979.5755	627.1830	737.6033	846.9182	408.8159	

	sensor_28	sensor_29	sensor_30	sensor_31	sensor_32	sensor_33	\
0	785.1935	684.9443	594.4445	682.8125	680.4416	433.7037	
1	785.1935	684.9443	594.4445	682.8125	680.4416	433.7037	
2	778.5734	715.6266	661.5740	721.8750	694.7721	441.2635	
3	779.5091	690.4011	686.1111	754.6875	683.3831	446.2493	
4	785.2307	704.6937	631.4814	766.1458	702.4431	433.9081	

	sensor_34	sensor_35
0	171.9375	341.9039
1	171.9375	341.9039
2	169.9820	343.1955
3	166.4987	343.9586
4	164.7498	339.9630

Out[6]: <matplotlib.axes._subplots.AxesSubplot at 0x7fe742c5a810>



The above figure shows off some of the sensors in a boxplot graph. As shown by the plots their is a decent amount of outliers within the data, but with the large amount of data there is definitely going to be a large amount of outliers as well.

```
In [7]: #Check to see all of the categorical data bins.  
data_df.machine_status.value_counts()
```

```
Out[7]: NORMAL          205836  
RECOVERING          14477  
BROKEN              7  
Name: machine_status, dtype: int64
```

In the final dataset, the broken will be merged with recovering as the goal is to check if the pump is in normal or not normal, recovering and broken, operation.