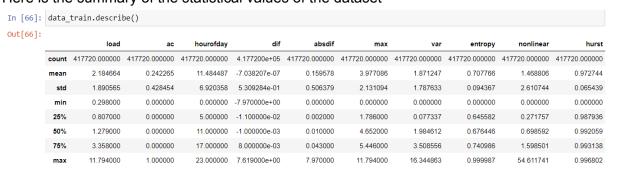
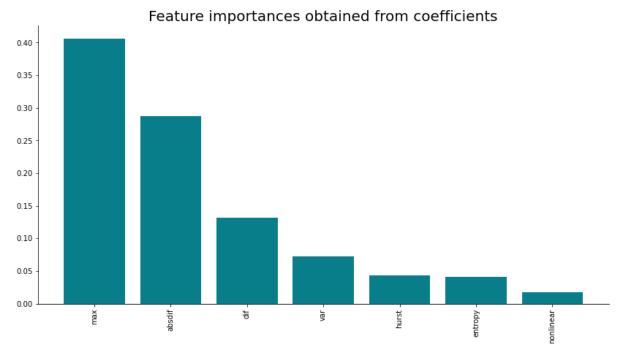
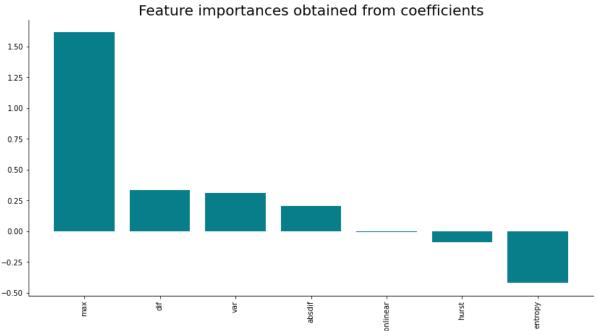
I have usedpandas read csv to read the train and test file. I have explored the datatype of the features and the targeted column.
 Here is the summary of the statistical values of the dataset-



Then some columns were unnecessary. I have dropped those column. I have dropped-hourofday,load,dayofweek. I have checked whether there is any null value ini the dataset and I got false.

2. I have analyzed the features using two techniques. One is coefficient based and another is tree based method. Here is the results I have got. From both the graph we can see, max has higher feature importance.





 I have used KNN,Decissiontree,Logisticregression for this task. Here are the performance metrics of all the three models. KNN-

precision recall f1-score support

0 0.94 0.93 0.94 96221 0.38 1 0.36 0.37 9319 accuracy 0.88 105540 macro avg 0.65 0.66 0.65 105540 weighted avg 0.89 0.88 0.89 105540

[[89851 6370] [5769 3550]]

```
Decision Tree-
precision recall f1-score support
```

```
0
           0.95
                  0.93
                          0.94
                                 96221
      1
           0.38
                  0.46
                          0.42
                                 9319
                         0.89
  accuracy
                               105540
                              0.68
                                    105540
 macro avg
               0.66
                      0.69
weighted avg
                0.90
                       0.89
                              0.89
                                    105540
[[89214 7007] [ 5035 4284]]
```

LogisticRegressionprecision recall f1-score support

```
0
           0.93
                  0.95
                         0.94
                                96221
      1
           0.39
                  0.31
                          0.34
                                 9319
  accuracy
                         0.90
                               105540
               0.66
                             0.64
 macro avg
                      0.63
                                    105540
weighted avg
               0.89
                       0.90
                              0.89
                                    105540
[[91728 4493] [ 6453 2866]]
```

For decision tree I have used maxdepth=2,criterion="gini" and splitter best.

For knn, I have used num_neighbour=9, weights=uniform

For logisticregression, I have used solver lblgf.

I have choosed this metrics after some experiment and hyperparameter tuning. I did not use the complex parameters as that will introduce overfitt and as there is good ,amount of data so It was ok to avoid overfit and there is less chance of underfit.

4. I have used Adaboost, Gradientboost and extra tree classifier for this portion. Here is the performance metrics

Adaboost-

precision recall f1-score support

```
0
           0.95
                  0.93
                          0.94
                                 96221
           0.41
                  0.49
                          0.44
                                 9319
      1
      accuracy
                              0.89
                                    105540
 macro avq
               0.68
                      0.71
                              0.69
                                    105540
weighted avg
                0.90
                       0.89
                               0.90 105540
[[89609 6612] [ 4786 4533]]
```

```
Extra Tree classifier-
precision recall f1-score support
```

```
0.94
                  0.96
                         0.95
     0
                                96221
      1
          0.45
                  0.36
                         0.40
                                9319
accuracy
                       0.90
                            105540
 macro avg
              0.69
                     0.66
                             0.67 105540
                      0.90
                             0.90 105540
weighted avg
               0.90
[[92021 4200][ 5939 3380]]
```

Adaboost is not flexible for loss functions. It has a fix loss function. On the other hand, Grdaientboost is adaptive to various loss functions and it can use any differential function as loss function. Adaboost is flexible to outliers but gradient boost is flexible to dense samples. From the result, we can see that for 1 sample, boosting techniques perform much better. As this was an unbalanced dataset so to detect the lesse sample, boosting techniques work much better. Definitely it is possible to use other tree algorithms like extratree classifier. Ext is an excellent ensemble method and I have used it. It gradually works in the weak learner and try to improve the performance.