Scale Features and Build Model

Scales Raw Features

Import CSV of Aggregated Darshan Logs Apply Log10 and Percent Scaling

```
In [1]: import os
    import pandas as pd
    import numpy as np
    import math
    import matplotlib.pyplot as plt

In [2]: df = pd.read_csv("./raws.csv",lineterminator='\n', error_bad_lines=False )
    #df.mean()

In [3]: df = df.drop(df.columns[0],axis = 1)
    df = df.drop(df.columns[0],axis = 1)
    f = pd.DataFrame()
In [4]: df
```

Out[4]:

	posix_read_time	posix_write_time	posix_meta_time	posix_bytes_read	posix_bytes_read_100
0	0.000000	0.000000	0.000000	0.000000e+00	0.0
1	104.611641	10.024055	20.060841	2.390891e+10	147688.0
2	124.560730	42.051125	54.839272	5.019637e+10	332059.0
3	25763.292969	582.297363	24.895737	5.488943e+12	30785.0
4	154.534821	681.548279	658.484985	2.293203e+10	588029.0
875282	138.354477	82.278084	194.485565	5.593977e+10	216146.0
875283	54.443073	231.440857	25.271391	1.465277e+09	3099.0
875284	0.000000	0.000000	0.000000	0.000000e+00	0.0
875285	0.000000	0.000000	0.000000	0.000000e+00	0.0
875286	227.063828	191.747269	172.671997	1.077412e+11	775359.0

875287 rows × 50 columns

```
In [5]:
    df = df.dropna(axis=0, how='any')
    df
```

Out[5]:

	posix_read_time	posix_write_time	posix_meta_time	posix_bytes_read	posix_bytes_read_100
0	0.000000	0.000000	0.000000	0.000000e+00	0.0
1	104.611641	10.024055	20.060841	2.390891e+10	147688.0
2	124.560730	42.051125	54.839272	5.019637e+10	332059.0
3	25763.292969	582.297363	24.895737	5.488943e+12	30785.0
4	154.534821	681.548279	658.484985	2.293203e+10	588029.0
875282	138.354477	82.278084	194.485565	5.593977e+10	216146.0
875283	54.443073	231.440857	25.271391	1.465277e+09	3099.0
875284	0.000000	0.000000	0.000000	0.000000e+00	0.0
875285	0.000000	0.000000	0.000000	0.000000e+00	0.0
875286	227.063828	191.747269	172.671997	1.077412e+11	775359.0

875287 rows × 50 columns

```
In [6]: #files
    f['log10_p_files'] = np.log10(df['posix_number_of_files'])
    f['log10_1_files'] = np.log10(df['lustre_number_of_files'])
```

/Users/dave/opt/anaconda3/lib/python3.8/site-packages/pandas/core/arrayli ke.py:358: RuntimeWarning: divide by zero encountered in log10 result = getattr(ufunc, method)(*inputs, **kwargs)

```
In [7]: #accesses
         df['p accesses'] = np.log10(df['posix reads'] + df['posix writes'])
         f['log10_p_accesses'] = np.log10(df['p_accesses'])
         f['log10 p accesses']
         /Users/dave/opt/anaconda3/lib/python3.8/site-packages/pandas/core/arrayli
         ke.py:358: RuntimeWarning: invalid value encountered in log10
           result = getattr(ufunc, method)(*inputs, **kwargs)
 Out[7]: 0
                        NaN
                   0.774119
         1
         2
                   0.804583
         3
                   0.841957
                   0.838513
         875282
                   0.802713
         875283
                   0.723940
         875284
                        NaN
         875285
                        NaN
         875286
                   0.831411
         Name: log10 p accesses, Length: 875287, dtype: float64
 In [8]: #bytes
         f['log10 p bytes'] = np.log10(df['posix bytes read'] )
 In [9]: |f['log10 p opens'] = np.log10(df['posix opens'])
         f['log10_p_seeks'] = np.log10(df['posix_seeks'])
         f['log10 p stats'] = np.log10(df['posix stats'])
         f['log10 p mode'] = np.log10(df['posix renamed mode'])
In [10]: f['log10 1 n osts'] = np.log10(df['lustre number of osts'])
         f['log10_l_stripe_w'] = np.log10(df['lustre_stripe_width'])
         f['log10 l mdts'] = np.log10(df['lustre mdts'])
In [11]: |f['log10 p nprocs'] = np.log10(df['nprocs'])
         f['log10_p_falign'] = np.log10(df['posix_f_align'])
         f['log10 p malign'] = np.log10(df['posix m align'])
In [12]: f['perc p reads'] = df['posix reads'] /df['p accesses']
         f['perc p writes'] = df['posix writes'] /df['p accesses']
```

```
isakov-scaling - Jupyter Notebook
In [14]:
             df['histogram r sum'] = df['posix bytes read 100'] + df['posix bytes re
             df['posix bytes read 4M'] + df['posix bytes read 10M'] + df['posix byte
             df['posix bytes read 1G'] + df['posix bytes read PLUS'] + df['posix byt
          3
          4
          5
             f['perc p bytes read 100'] = df['posix bytes read 100']/df['histogram r
             f['perc p bytes read 1K'] = df['posix bytes read 1K']/df['histogram r s
             f['perc p bytes read 10K'] = df['posix bytes read 10K']/df['histogram r
            f['perc p bytes read 100K'] = df['posix bytes read 100K']/df['histogram
             f['perc p bytes read 1M'] = df['posix bytes read 11M']/df['histogram r
            f['perc_p_bytes_read_4M'] = df['posix_bytes_read_4M']/df['histogram_r_s
         10
         11
            f['perc p bytes read 10M'] = df['posix bytes read 10M']/df['histogram r
             f['perc_p_bytes_read_100M'] = df['posix_bytes_read_100M']/df['histogram
         12
            f['perc p bytes read 1G'] = df['posix bytes read 1G']/df['histogram r s
         13
             f['perc p bytes read PLUS'] = df['posix bytes read PLUS']/df['histogram
In [15]: df['histogram w sum'] = df['posix bytes write 100'] + df['posix bytes write
         df['posix bytes write 10K'] + df['posix bytes write 100K'] + df['posix byte
         df['posix_bytes_write_4M'] + df['posix_bytes_write_10M'] + df['posix_bytes_
         df['posix bytes write 1G'] + df['posix bytes write PLUS']
         f['perc p bytes write 100'] = df['posix bytes write 100']/df['histogram w s
         f['perc p bytes write 1K'] = df['posix bytes write 1K']/df['histogram w sum
         f['perc p bytes write 10K'] = df['posix bytes write 10K']/df['histogram w s
         f['perc p bytes write 100K'] = df['posix bytes write 100K']/df['histogram w
         f['perc_p_bytes_write_1M'] = df['posix_bytes_write_1M']/df['histogram_w_sum
         f['perc p bytes write 4M'] = df['posix bytes write 4M']/df['histogram w sum
         f['perc_p_bytes_write_10M'] = df['posix_bytes_write_10M']/df['histogram_w_s
         f['perc_p_bytes_write_100M'] = df['posix_bytes_write_100M']/df['histogram_w
         f['perc_p_bytes_write_1G'] = df['posix_bytes_write_1G']/df['histogram_w_sum
         f['perc_p_bytes_write_PLUS'] = df['posix_bytes_write_PLUS']/df['histogram_w
         f = f.replace(-np.inf, -1)
         f = f.replace(np.nan, 0)
In [16]: |df['time'] = df['posix_write_time'] + df['posix_read_time'] + df['posix_met
         df['bytes'] = df['posix bytes read'] + df['posix bytes write']
```

```
t = pd.DataFrame()
f['throughput'] = df['bytes'] / df['time']
f = f[f['throughput'] >0]
```

```
In [17]: #delete columns with all zeros
         f = f.loc[:, (f != 0).any(axis=0)]
         #remove infinite values
         f = f.replace([np.inf, -np.inf], np.nan).dropna(axis=0)
         f.max()
Out[17]: log10_p_files
                                     5.086100e+00
         log10 l files
                                     5.086100e+00
         log10_p_accesses
                                     1.015048e+00
         log10 p bytes
                                     1.448265e+01
         log10 p opens
                                     8.742811e+00
         log10_p_seeks
                                     1.015993e+01
         log10 p_stats
                                     7.814442e+00
         log10 p mode
                                     7.727321e+00
         log10_l_n_osts
                                     2.556303e+00
         log10_l_stripe_w
                                     6.871490e+00
         log10 l mdts
                                     0.000000e+00
         log10 p nprocs
                                     5.546543e+00
         log10 p falign
                                     1.115306e+01
                                     6.035555e+00
         log10 p malign
         perc_p_reads
                                     2.161636e+09
         perc_p_writes
                                     1.287974e+09
                                     1.000000e+00
         perc p bytes read 100
                                     1.000000e+00
         perc p bytes read 1K
         perc p bytes read 10K
                                     9.331507e+07
         perc p bytes read 100K
                                     1.000000e+00
         perc p bytes read 1M
                                     1.000000e+00
         perc_p_bytes_read_4M
                                     1.000000e+00
         perc p bytes read 10M
                                     9.978617e-01
         perc p bytes read 100M
                                     1.000000e+00
         perc p bytes read 1G
                                     1.000000e+00
         perc p bytes write 100
                                     1.000000e+00
         perc p bytes write 1K
                                     1.000000e+00
         perc p bytes write 10K
                                     1.000000e+00
         perc_p_bytes_write 100K
                                     1.000000e+00
         perc p bytes write 1M
                                     1.000000e+00
         perc_p_bytes_write_4M
                                     1.000000e+00
         perc_p_bytes_write_10M
                                     1.000000e+00
         perc p bytes write 100M
                                     1.000000e+00
         perc_p_bytes_write_1G
                                     1.000000e+00
         throughput
                                     2.344536e+09
         dtype: float64
         t['throughput'] = f['throughput']
         f = f.drop(labels = 'throughput', axis = 1)
         f
In [19]: | t.max()
Out[19]: throughput
                        2.344536e+09
         dtype: float64
```

```
In [20]: print(t.min())
         print(t.max())
                       0.39201
         throughput
         dtype: float64
         throughput
                       2.344536e+09
         dtype: float64
In [21]: from sklearn.model_selection import train_test_split
         import scipy as sp
         import random
         from sklearn.metrics import r2_score
         rseed = 123
         t_size = 0.2
         train_data, test_data, train_labels, test_labels = train_test_split(f,t, te
In [22]: import xgboost as xg
In [23]: from sklearn.metrics import r2_score
         from sklearn.metrics import mean_squared_error
         from sklearn.metrics import mean absolute error
         from sklearn.metrics import mean_squared_log_error
         from sklearn.metrics import mean_absolute_percentage_error
```

```
In [24]: from sklearn.linear model import LinearRegression
         for i in range(3):
             rseed = random.randint(1,10000)
             print(rseed)
             train data, test data, train labels, test labels = train test split(f,t
             reg = LinearRegression().fit(train data, train labels)
             predicted_labels = reg.predict(test_data)
             print("Mean True Value: \t",int(test_labels.mean() ))
             print("Mean Absolute Error: \t", int(mean_absolute_error(test_labels, p
             print("Mean Squared Error: ", mean squared error(test_labels, predicted
             print("Root Mean Squared Error: ", mean squared error(test_labels, pred
             print("MAPE :" + str(mean_absolute percentage error( test_labels, predi
             print("R2: " + str(r2 score(test labels, predicted labels)) + "\n")
         5256
         Mean True Value:
                                  84655611
         Mean Absolute Error:
                                  35306336
         Mean Squared Error: 3998930934775507.5
         Root Mean Squared Error: 63237100.935886584
         MAPE :430.5851114848273
         R2: 0.6645808586938563
         9498
         Mean True Value:
                                  84440338
         Mean Absolute Error:
                                  35161045
         Mean Squared Error: 3950854968803895.5
         Root Mean Squared Error: 62855826.84846247
         MAPE :4020.4763055816597
         R2: 0.664068592196188
```

Mean True Value: 83728301 Mean Absolute Error: 35009538

Mean Squared Error: 3930630026111212.5 Root Mean Squared Error: 62694736.829427816

MAPE :366.7762125668867 R2: 0.6631738087493297

```
In [25]: #EXTREME GRADIENT BOOST
         for i in range(3):
             rseed = random.randint(1,10000)
             print(rseed)
             train_data, test_data, train_labels, test_labels = train_test_split(f,t
             xgb r = xg.XGBRegressor(n estimators = 1000, seed = 123)
             xgb r.fit(train data, train labels)
             predicted_labels = xgb_r.predict(test_data)
             print("Mean True Value: \t",int(test_labels.mean() ))
             print("Mean Absolute Error: \t", int(mean_absolute_error(test_labels, p
             print("Mean Squared Error: ", mean squared error(test_labels, predicted
             print("Root Mean Squared Error: ", mean squared error(test_labels, pred
             print("MAPE :" + str(mean_absolute percentage error( test_labels, predi
             print("R2: " + str(r2 score(test labels, predicted labels)) + "\n")
         3363
         Mean True Value:
                                  84417914
         Mean Absolute Error:
                                  16232655
```

Mean Squared Error: 1159032349626505.8 Root Mean Squared Error: 34044564.17148714 MAPE :48.98696827810851 R2: 0.9017313477293982 8266 Mean True Value: 84104805 Mean Absolute Error: 16300453 Mean Squared Error: 1176870360451501.0 Root Mean Squared Error: 34305544.16492327 MAPE :19.384029900993173 R2: 0.9001362752350884 8582 Mean True Value: 84536628 Mean Absolute Error: 16329916 Mean Squared Error: 1177729119299233.5 Root Mean Squared Error: 34318058.20991673 MAPE :679.6267639834404 R2: 0.9011427953693474

```
In [26]: from sklearn.tree import DecisionTreeRegressor
         for i in range(3):
             rseed = random.randint(1,10000)
             print(rseed)
             train_data, test_data, train_labels, test_labels = train_test_split(f,t
             reg = DecisionTreeRegressor(max depth = 4)
             reg.fit(train_data, train_labels)
             predicted_labels = reg.predict(test_data)
             print("Mean True Value: \t",int(test_labels.mean() ))
             print("Mean Absolute Error: \t", int(mean_absolute_error(test_labels, p
             print("Mean Squared Error: ", mean squared error(test labels, predicted
             print("Root Mean Squared Error: ", mean_squared_error(test_labels, pred
             print("MAPE :" + str(mean absolute percentage error( test labels, predi
             print("R2: " + str(r2 score(test labels, predicted labels)) + "\n")
         1479
         Mean True Value:
                                  84871582
         Mean Absolute Error:
                                  28228907
         Mean Squared Error: 2842564157355832.0
         Root Mean Squared Error: 53315702.72776897
         MAPE :1648.03877149611
         R2: 0.764575284292099
         3502
         Mean True Value:
                                  84302211
         Mean Absolute Error:
                                  28107752
         Mean Squared Error: 2698953271249399.5
         Root Mean Squared Error: 51951451.098592035
         MAPE :1140.1004613705709
         R2: 0.7726844672772891
         5677
         Mean True Value:
                                  84134723
         Mean Absolute Error:
                                  28190851
         Mean Squared Error: 2769122901493667.0
         Root Mean Squared Error: 52622456.2472493
         MAPE :993.7645129368356
         R2: 0.7656705470577084
```

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