# 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
        import random
        from sklearn.model_selection import train_test_split
In [2]: df = pd.read_csv("./raws.csv",lineterminator='\n',sep = ',' ,error_bad_line
        #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.columns
Out[4]: Index(['posix_read_time', 'posix_write_time', 'posix_meta_time',
                posix_bytes_read', 'posix_bytes_read_100', 'posix_bytes_read_1K',
                'posix bytes read 10K', 'posix bytes read 100K', 'posix bytes read
        _1lM',
                'posix bytes read 4M', 'posix bytes read 10M', 'posix bytes read 1
        00M',
                'posix bytes read 1G', 'posix bytes read PLUS', 'posix bytes writ
        e',
               'posix bytes_write_100', 'posix_bytes_write_1K',
               'posix_bytes_write_10K', 'posix bytes write 100K',
               'posix bytes write 1M', 'posix bytes write 4M', 'posix bytes write
        10M',
                'posix bytes write 100M', 'posix bytes write 1G',
               'posix_bytes_write_PLUS', 'posix_opens', 'posix_reads', 'posix_wri
        tes',
               'posix seeks', 'posix_stats', 'posix_mmaps', 'posix_fsyncs',
               'posix fdsyncs', 'posix rename sources', 'posix rename targets',
                'posix renamed from', 'posix renamed mode', 'posix number of file
        s',
               'nprocs', 'posix f align', 'posix m align', 'lustre number of file
        s',
               'lustre mdts', 'lustre osts', 'lustre stripe size',
                'lustre stripe offset', 'lustre stripe width', 'lustre number of o
        sts',
                'jobid', 'path'],
              dtype='object')
```

```
In [5]:
        df = df.dropna(axis=0, how='any')
        df.columns
Out[5]: Index(['posix_read_time', 'posix_write_time', 'posix_meta_time',
                posix bytes_read', 'posix_bytes_read_100', 'posix_bytes_read_1K',
                'posix_bytes_read_10K', 'posix_bytes_read_100K', 'posix_bytes read
        _1lM',
                'posix_bytes_read_4M', 'posix_bytes_read_10M', 'posix_bytes_read_1
        00M',
                'posix bytes read 1G', 'posix bytes read PLUS', 'posix bytes writ
        e',
                'posix_bytes_write_100', 'posix_bytes_write_1K',
                'posix_bytes_write_10K', 'posix_bytes_write_100K',
                'posix bytes write 1M', 'posix bytes write 4M', 'posix bytes write
        10M',
                'posix bytes write 100M', 'posix bytes write 1G',
                'posix_bytes_write_PLUS', 'posix_opens', 'posix_reads', 'posix_wri
        tes',
                'posix_seeks', 'posix_stats', 'posix_mmaps', 'posix_fsyncs',
                'posix fdsyncs', 'posix rename sources', 'posix rename targets',
                'posix renamed from', 'posix renamed mode', 'posix number of file
        s',
                'nprocs', 'posix f_align', 'posix m_align', 'lustre number of file
        s',
                'lustre_mdts', 'lustre_osts', 'lustre_stripe_size',
                'lustre stripe offset', 'lustre stripe width', 'lustre number of o
        sts',
                'jobid', 'path'],
              dtype='object')
In [6]: #files
        f['log10 p files'] = df['posix number of files'] + 1
        f['log10 l files'] = df['lustre number of files']
In [7]: #accesses
        df['p accesses'] = df['posix reads'] + df['posix writes']
        f['log10 p accesses'] = df['p accesses']
        f['log10 p accesses']
Out[7]: 0
                         0.0
        1
                    880136.0
        2
                  2379598.0
        3
                  8903411.0
                  7846387.0
                     . . .
        875282
                  2234152.0
        875283
                   197651.0
                         0.0
        875284
                         0.0
        875285
        875286
                   6065006.0
        Name: log10 p accesses, Length: 875287, dtype: float64
```

```
In [8]: #bytes
         f['log10 p bytes'] = df['posix bytes read']
 In [9]: f['log10 p opens'] = df['posix opens']
         f['log10 p seeks'] = df['posix seeks']
         f['log10_p_stats'] = df['posix_stats']
         f['log10 p mode'] = df['posix renamed mode']
In [10]: f['log10_l_n_osts'] = df['lustre_number_of_osts']
         f['log10 l stripe w'] = df['lustre stripe width']
         f['log10_l_mdts'] = df['lustre_mdts']
In [11]: f['log10 p nprocs'] = df['nprocs']
         f['log10 p falign'] = df['posix_f_align']
         f['log10 p malign'] = df['posix m align']
In [12]: f['perc_p_reads'] = df['posix_reads']
         f['perc_p_writes'] = df['posix_writes']
In [13]: f['perc p bytes read 100'] = df['posix bytes read 100']
         f['perc p bytes read 1K'] = df['posix bytes read 1K']
         f['perc p bytes read 10K'] = df['posix bytes read 10K']
         f['perc p bytes read 100K'] = df['posix bytes read 100K']
         f['perc_p_bytes_read_1M'] = df['posix_bytes_read_11M']
         f['perc p bytes read 4M'] = df['posix bytes read 4M']
         f['perc_p_bytes_read_10M'] = df['posix_bytes_read_10M']
         f['perc p bytes read 100M'] = df['posix bytes read 100M']
         f['perc p bytes read 1G'] = df['posix bytes read 1G']
         f['perc_p_bytes_read_PLUS'] = df['posix_bytes_read_PLUS']
In [14]: f['perc_p_bytes_write_100'] = df['posix_bytes_write_100']
         f['perc p bytes write 1K'] = df['posix bytes write 1K']
         f['perc p bytes write 10K'] = df['posix bytes write 10K']
         f['perc_p_bytes_write_100K'] = df['posix_bytes_write_100K']
         f['perc_p_bytes_write_1M'] = df['posix_bytes_write_1M']
         f['perc_p_bytes_write_4M'] = df['posix_bytes_write_4M']
         f['perc p bytes write 10M'] = df['posix bytes write 10M']
         f['perc_p_bytes_write_100M'] = df['posix_bytes_write_100M']
         f['perc_p_bytes_write_1G'] = df['posix_bytes_write_1G']
         f['perc p bytes write PLUS'] = df['posix bytes write PLUS']
         f = f.replace(-np.inf, -1)
         f = f.replace(np.nan, 0)
In [15]: df['time'] = df['posix write time'].astype('float') + df['posix read time']
In [16]: df['bytes'] = df['posix_bytes_read'].astype('float') + df['posix_bytes_writ
In [17]: |#df = df[df['bytes'] >99999999]
```

```
In [18]:
         f['throughput'] = df['bytes'].astype('float') / df['time']
         f = f[f['throughput'] >0]
In [19]: #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[19]: log10 p files
                                     1.219280e+05
         log10_l_files
                                     1.219270e+05
         log10 p accesses
                                     2.251942e+10
         log10 p bytes
                                     3.038456e+14
         log10 p opens
                                     5.531094e+08
         log10 p_seeks
                                     1.445220e+10
         log10 p stats
                                     6.522921e+07
         log10 p mode
                                     5.337293e+07
         log10_l_n_osts
                                     3.600000e+02
                                     7.438575e+06
         log10 l stripe w
         log10 l mdts
                                     1.000000e+00
                                     3.520000e+05
         log10 p nprocs
         log10 p falign
                                     1.422540e+11
         log10 p malign
                                     1.085312e+06
         perc p reads
                                     2.237846e+10
         perc p writes
                                     1.302770e+10
         perc p bytes read 100
                                     5.221517e+08
         perc p bytes read 1K
                                     2.074657e+10
         perc p bytes read 10K
                                     1.536278e+09
         perc p bytes read 100K
                                     1.515506e+08
         perc p bytes read 1M
                                     4.044503e+08
         perc p bytes read 4M
                                     6.561462e+07
         perc p bytes read 10M
                                     2.083200e+06
         perc p bytes read 100M
                                     2.872090e+05
         perc p bytes read 1G
                                     1.792000e+06
         perc_p_bytes_write_100
                                     1.302770e+10
         perc p bytes write 1K
                                     2.852127e+09
         perc p bytes write 10K
                                     3.867477e+08
         perc p bytes write 100K
                                     8.347452e+07
                                     1.357245e+07
         perc p bytes write 1M
         perc p bytes write 4M
                                     3.839488e+06
         perc_p_bytes_write_10M
                                     6.190660e+05
         perc p bytes write 100M
                                     1.249280e+06
         perc p bytes write 1G
                                     1.937500e+04
         throughput
                                     2.344536e+09
         dtype: float64
```

```
In [20]: t = pd.DataFrame()
    t['throughput'] = f['throughput']
    f = f.drop(labels = 'throughput', axis = 1)
    f
```

### Out[20]:

	log10_p_files	log10_l_files	log10_p_accesses	log10_p_bytes	log10_p_opens	log10_p_seeks
1	800.0	176.0	880136.0	2.390891e+10	8858.0	319241.0
2	361.0	224.0	2379598.0	5.019637e+10	62398.0	1107764.0
3	291.0	290.0	8903411.0	5.488943e+12	8711.0	2010273.0
4	320.0	201.0	7846387.0	2.293203e+10	23158.0	6015926.0
6	429.0	190.0	6647935.0	5.209185e+10	69261.0	4608438.0
875280	2.0	1.0	57344.0	3.006477e+10	1808.0	59128.0
875281	14.0	4.0	102439.0	7.467916e+08	193.0	91.0
875282	625.0	124.0	2234152.0	5.593977e+10	35112.0	1035457.0
875283	1089.0	1088.0	197651.0	1.465277e+09	2112.0	12509.0
875286	583.0	128.0	6065006.0	1.077412e+11	119432.0	2750807.0

671063 rows × 34 columns

```
In [21]: df = df[df.index.isin(t.index)]
    t = t.reset_index()
    f = f.reset_index()
    f = f.drop(f.columns[0] , axis =1)
    t = t.drop(t.columns[0] , axis =1)
```

```
In [22]:
```

```
from sklearn.preprocessing import StandardScaler
f = StandardScaler().fit_transform(f)
```

```
In [23]:
```

# Out[23]:

# throughput

- **0** 1.803194e+08
- 1 2.282342e+08
- 2 2.083669e+08
- 1.724841e+07
- 4.581690e+07

**671058** 5.482767e+08

671059 2.136637e+07

671060 1.362498e+08

**671061** 1.942413e+07

671062 1.842942e+08

671063 rows × 1 columns

```
In [24]: print(t.min())
         print(t.max())
         throughput
                        0.39201
```

dtype: float64

throughput 2.344536e+09

dtype: float64

```
In [25]: rseed = 0
         t size = 0.1
```

```
In [26]: 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 [27]: 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")
         466
         Mean True Value:
                                  84543501
         Mean Absolute Error:
                                  54153061
         Mean Squared Error: 2.261908446988301e+16
         Root Mean Squared Error: 150396424.39194825
         MAPE :826.5616182370247
         R2: -0.9317984212354609
         55
         Mean True Value:
                                  84950034
         Mean Absolute Error:
                                  55005401
         Mean Squared Error: 9147388791499650.0
         Root Mean Squared Error: 95641982.369144
         MAPE :1104.2638165971384
         R2: 0.24535925719705243
         8878
         Mean True Value:
                                  84309897
         Mean Absolute Error:
                                  54277374
         Mean Squared Error: 9377959238697192.0
         Root Mean Squared Error: 96839863.89239295
         MAPE :1545.0408488673352
         R2: 0.2049561655447404
```

#### In [28]: import xgboost as xg

```
In [29]: #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")
         1541
         Mean True Value:
                                  84338182
```

Mean Absolute Error: 16088329 Mean Squared Error: 1138614692905361.2 Root Mean Squared Error: 33743365.1686574 MAPE :114.38197797167567 R2: 0.9045144804003565 5641 Mean True Value: 84332413 Mean Absolute Error: 16384662 Mean Squared Error: 1181244012528314.0 Root Mean Squared Error: 34369230.607162476 MAPE :56.22594013095263 R2: 0.8992227778496157 8977 Mean True Value: 84289999 Mean Absolute Error: 16252869 Mean Squared Error: 1157630148155565.8 Root Mean Squared Error: 34023964.32157143 MAPE :51.37917112564205 R2: 0.9015652082234624

```
In [ ]:
```

```
In [30]: 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")
         5626
         Mean True Value:
                                  84067576
         Mean Absolute Error:
                                  32864432
         Mean Squared Error: 3195103903233775.0
         Root Mean Squared Error: 56525250.1386219
         MAPE :808.6107414878769
         R2: 0.7303356522831486
         3795
         Mean True Value:
                                  84232117
         Mean Absolute Error:
                                  32258290
         Mean Squared Error: 3102852439300382.0
         Root Mean Squared Error: 55703253.39960299
         MAPE :1428.7644701294769
         R2: 0.737042053264592
         454
         Mean True Value:
                                  84362852
         Mean Absolute Error:
                                  33421235
         Mean Squared Error: 3298164123928292.0
         Root Mean Squared Error: 57429644.992184065
         MAPE :3748.0546185631597
         R2: 0.7236418495764236
 In [ ]:
```

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