

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.preprocessing import StandardScaler
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
```

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.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_
              _11M', '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_writ
              es', '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']
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
4     7846387.0
...
875282    2234152.0
875283    197651.0
875284         0.0
875285         0.0
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_1M']
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'] > 999999999]
```

```
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          1.219270e+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
log10_l_stripe_w       7.438575e+06
log10_l_mdts           1.000000e+00
log10_p_nprocs         2.396160e+05
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
perc_p_bytes_write_1M  1.357245e+07
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	799.0	176.0	880136.0	2.390891e+10	8858.0	319241.0
2	360.0	224.0	2379598.0	5.019637e+10	62398.0	1107764.0
3	290.0	290.0	8903411.0	5.488943e+12	8711.0	2010273.0
4	319.0	201.0	7846387.0	2.293203e+10	23158.0	6015926.0
6	428.0	190.0	6647935.0	5.209185e+10	69261.0	4608438.0
...
875276	288.0	288.0	7814478.0	5.522966e+12	8709.0	2067127.0
875280	1.0	1.0	57344.0	3.006477e+10	1808.0	59128.0
875282	624.0	124.0	2234152.0	5.593977e+10	35112.0	1035457.0
875283	1088.0	1088.0	197651.0	1.465277e+09	2112.0	12509.0
875286	582.0	128.0	6065006.0	1.077412e+11	119432.0	2750807.0

520193 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]: f = StandardScaler().fit_transform(f)
```

```
In [23]: t
```

```
Out[23]:
```

	throughput
0	1.803194e+08
1	2.282342e+08
2	2.083669e+08
3	1.724841e+07
4	4.581690e+07
...	...
520188	2.073475e+08
520189	5.482767e+08
520190	1.362498e+08
520191	1.942413e+07
520192	1.842942e+08

520193 rows × 1 columns

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

```
throughput      638.985281
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")
```

8204

```
Mean True Value:      100342761
Mean Absolute Error:   58688695
Mean Squared Error:   8698560663188818.0
Root Mean Squared Error: 93266074.55655469
MAPE :14.00056301993843
R2: 0.32192789211512274
```

5677

```
Mean True Value:      100343079
Mean Absolute Error:   58589611
Mean Squared Error:   8628139093000328.0
Root Mean Squared Error: 92887776.87618715
MAPE :19.238554566222852
R2: 0.3216901153524514
```

4971

```
Mean True Value:      100049162
Mean Absolute Error:   58497126
Mean Squared Error:   9299355513600226.0
Root Mean Squared Error: 96433166.04571389
MAPE :23.040315153592747
R2: 0.27335493291354285
```

```
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")
```

9855

```
Mean True Value:          100667807
Mean Absolute Error:      19134688
Mean Squared Error:      1405408872584360.0
Root Mean Squared Error:  37488783.29026377
MAPE :1.3136417913984533
R2: 0.8913421412766307
```

2523

```
Mean True Value:          101050481
Mean Absolute Error:      19288326
Mean Squared Error:      1392685277107612.2
Root Mean Squared Error:  37318698.75957108
MAPE :1.407238252976963
R2: 0.8921998444606166
```

1294

```
Mean True Value:          100848801
Mean Absolute Error:      19296374
Mean Squared Error:      1425224378566515.2
Root Mean Squared Error:  37752144.02608831
MAPE :1.482152414800812
R2: 0.8901194854839278
```

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")
```

8719

Mean True Value: 99909539
Mean Absolute Error: 31401525
Mean Squared Error: 2950727508731309.0
Root Mean Squared Error: 54320599.3038673
MAPE :24.50180820377324
R2: 0.7653258135539455

9108

Mean True Value: 100572719
Mean Absolute Error: 31468423
Mean Squared Error: 3125345513270444.0
Root Mean Squared Error: 55904789.71671787
MAPE :23.11469020109784
R2: 0.7631770228917953

7814

Mean True Value: 100302456
Mean Absolute Error: 31626365
Mean Squared Error: 3100014395184885.5
Root Mean Squared Error: 55677772.9007266
MAPE :15.305068584553645
R2: 0.7597213601791146

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