# Data\_Processing

May 5, 2021

# 1 Seguntial Event Feature Extraction Notebook

## 1.0.1 Log Structure

The HDFS log is built the following way: - %d{yy/MM/dd HH:mm:ss}: The date and time - %???: Unknown 2-4 digit code - %p: The priority of the logging event (INFO, WARN, DEBUG, ERROR, etc.) - %c: Category of logging event (class name) - %m: Log message

## 1.1 Import Libraries

```
[1]: # Parsing and wrangling
import pandas as pd
import numpy as np
import regex as re

# misc
from datetime import datetime as dt
import itertools

#import cupy as cp
#import functools
```

```
[3]: ## Import Raw Log Data
raw = pd.read_csv('Data/HDFS.log', header=None, sep='\n')[0]
raw.head()
```

```
[3]: 0 081109 203518 143 INFO dfs.DataNode$DataXceive...
1 081109 203518 35 INFO dfs.FSNamesystem: BLOCK*...
2 081109 203519 143 INFO dfs.DataNode$DataXceive...
3 081109 203519 145 INFO dfs.DataNode$DataXceive...
4 081109 203519 145 INFO dfs.DataNode$PacketResp...
Name: 0, dtype: object
```

# 1.2 Code Parsing Section

Here we'll parse the log file for block IDs which we will use this to aggregate message sequences for the corresponding block ID.

Note the checkpoint cells. Due to memory limitations, checkpoints were created where the kernel could be restarted to free RAM for the next processing cells.

#### 1.2.1 Get BLock IDs and their Label

```
[9]: ## Import Anomaly Labels
labels = pd.read_csv('Data/anomaly_label.csv')
labels.iloc[:,1][labels.Label == 'Anomaly'] = 1
labels.iloc[:,1][labels.Label == 'Normal'] = 0

length = len(labels)
anomalies = len(labels[labels.Label==1])
print('Len labels: ', length)
print('Anomaly count: ', anomalies)
print('% Anomalous: ', round(anomalies/length*100, 2),'%')
```

Len labels: 575061 Anomaly count: 16838 % Anomalous: 2.93 %

#### 1.2.2 Parse Log for BlockIDs

```
[20]: print('Unique blocks in Log File: ', len(blocks_in_order.unique()))
print('Anomalies in the Log File: ', binarizer_vectorized(blocks_in_order.

unique()).sum()) # anomalous **blocks** in the log file
```

Unique blocks in Log File: 575061 Anomalies in the Log File: 16838

The number of unique block IDs and anomaly counts in both the parsed log file and labeled file are equal from the printed outputs. This confirms that the parsing was done correctly.

#### 1.2.3 Parse for Full Message Content

```
[5]: # Extract Raw Messages
full_msg = raw.str.extract(r'((?<=:\s).*)')[0]
full_msg.to_csv('Data/full_msg.csv', index=None, header=None)
# CHECKPOINT
full_msg.head()</pre>
```

```
[5]: 0 Receiving block blk_-1608999687919862906 src: ...

1 BLOCK* NameSystem.allocateBlock: /mnt/hadoop/m...

2 Receiving block blk_-1608999687919862906 src: ...

3 Receiving block blk_-1608999687919862906 src: ...

4 PacketResponder 1 for block blk_-1608999687919...

Name: 0, dtype: object
```

#### 1.2.4 Next we remove all unique event identifiers to get general message structures.

We convert them to numeric and label each log event this way.

```
[9]: # Import Raw message checkpoint
full_msg = pd.read_csv('Data/full_msg.csv', index_col=False, header=None,

→squeeze=True)
```

Messages extracted: 11175629

#### 1.2.5 Now lets convert general message structures to numerical codes.

```
[23]: coded_msgs = pd.Categorical(str_msgs).codes
    print('Unique Message Types: ', len(pd.Series(coded_msgs).unique()))
    coded_msgs
Unique Message Types: 75
```

#### 1.2.6 Create a Feature for Block ID Message Type Sequences

[23]: array([55, 4, 55, ..., 62, 62, 62], dtype=int8)

```
[91]: # Import anomaly labels
     labeled_blks = pd.read_feather('Data/labeled_blks.feather')
     # Log dataframe of blocks and event category in order
     blk_events = pd.DataFrame({'blk_ID': labeled_blks.blkID, 'msg_code':coded_msgs})
     # Groupby by block to create event sequences
     blk event sequences = blk events.groupby('blk ID')['msg code'].apply(list).
      # Assign anomaly labels using a dictionary
     blk_key = dict(labels.values)
     vectorized_anom_labeler = np.vectorize(lambda x: blk_key[x])
     blk_event_sequences['anomaly'] = pd.
      →Series(vectorized_anom_labeler(blk_event_sequences.blk_ID))
     # Export the golden feature
     blk_event_sequences.to_csv('Data/blk_event_sequences.csv', index = False,_
      →header = True)
     blk_event_sequences.head()
```

```
[91]: blk_ID \
0 blk_-1000002529962039464
1 blk_-100000266894974466
2 blk_-1000007292892887521
3 blk_-1000014584150379967
4 blk_-1000028658773048709

sequence anomaly
0 [55, 55, 55, 16, 51, 53, 51, 53, 3, 3, 3, 51, 53]
1 [20, 55, 55, 55, 3, 3, 3, 51, 53, 51, 53, 51, ... 0
2 [55, 55, 16, 55, 51, 53, 51, 53, 51, 53, 3, 3, 3]
```

```
4 [55, 55, 55, 20, 51, 53, 51, 53, 51, 53, 3, 3,... 0

[97]: # Quick Check
    labels = pd.read_csv('Data/anomaly_label.csv')
    labels.iloc[:,1][labels.Label == 'Anomaly'] = 1
    labels.iloc[:,1][labels.Label == 'Normal'] = 0

    target_rows = len(labels)
    final_rows = len(blk_event_sequences)

    original_anomalies = len(labels[labels.Label==1])
    final_anomalies = len(blk_event_sequences[blk_event_sequences.anomaly==1])

    print(f'Total blocks present {final_rows} of {target_rows}')
    print(f'Total anomalies present {final_anomalies} of {original_anomalies}')
```

Total blocks present 575061 of 575061 Total anomalies present 16838 of 16838

3 [55, 20, 55, 55, 3, 3, 51, 53, 51, 53, 51, ...