SyedHassaanTauqeer 19-01-2019 Final Code Notebook Draft

```
import os, glob
 In [4]:
         import pandas as pd
         import numpy as np
In [18]: def file_write(df, path, name):
             #This is a simple function that writes a dataframe to a csv file
             return df.to_csv(path+name+'.csv', sep = ';', index = False)#Making index
          False helps ignore index column when writing file
In [21]: def floatConv(arr):#This function takes in a python list/ numpy array and repl
         aces any commas left in the thousand position in
             #numerical values. This stands as a safeguard to ensure all values are flo
         at for easier future operations
             temp = []
             for i in range(len(arr)):
                 temp.append( float(str(arr[i]).replace(",", "") ) )
             return temp
         def perScale(arr):#This function is used to scale a list or numpy array to per
In [22]:
         centage ratio.
             #This is done by taking the maximum value in that array and scaling every
          other value relatively. No trailing decimal places
             #left for the sake of ease in matching and querying at future stages
             scaled = []
             tempMax = max(arr)
             for i in range(len(arr)):
                 scaled.append( round(float(arr[i]/tempMax)*100, 0) )
             return scaled
In [23]:
         def minFil(arrList, win):#This function takes in a list/numpy array and applie
         s a rolling window operation to get the minimum
             #value within that window.
             tempDiff = arrList
             appVal = arrList[len(arrList)-1]# To ensure the input and output lengths o
         f the array are same, the last value of the input
             #array is repeated the number of times that matches one less than the wind
         ow size
             tempDiff = np.insert(tempDiff, len(tempDiff), np.repeat(appVal, win-1))
             arrListDF = pd.DataFrame({"arr":tempDiff})#since the rolling function in p
         andas is really efficient, we convert the array
             #to Series data
             tempDiff = arrListDF.rolling(win).min().dropna()#all NaN values are droppe
         d to not cause issues in future operations
             tempDiff = tempDiff['arr'].values
             return tempDiff
```

```
In [24]: def diffIndexMaker(arr):#This function takes in an numpy array/list and comput
    es the difference in between each consecutive value
        #To ensure that the input and output lengths of the array remain same, the
        first value of the array is pre-pended.
        #This doesn't affect the computation or any future operations but keeps le
    ngth consistent
        temp = arr.item(0)
        arr = np.insert(arr, 0, temp)#pre-pending the first value to the array
        return np.diff(arr)
```

## This block makes all the data files uniform in shape

The data that is re-structured by the following block is then taken to label the "spin phases" and the "end spin phases". The labeled files are copies of this format

```
In [42]: #Change working directory to raw data folder
    os.chdir('C:\\Users\\Labyrinth\\JUPYTER NOTEBOOKS\\WeWash_Praktikum_TUM3sem\\W
    eWash_Analysis_ver2\\Data\\raw')
```

```
In [43]: for file in glob.glob("*.csv"): #Iterating through all files in the raw data
          directory
             writingDF = pd.DataFrame(columns=['machine', 'pow', 'time', 'timeStamp'])
             labtsDF = pd.read csv( file, delimiter=',')
             machines = labtsDF['reservation_id'].unique() #Uniques machines are filter
         ed based on id
             print 'machines: ', len(machines)
             for i in range(len(machines)):#Each of the machines has multiple records.
                 tempDF = labtsDF[labtsDF['reservation_id'] == machines[i]]
                 subSet = tempDF[['reservation_id', 'power', 'sample_time']] #Subset th
         e dataframe on relevant columns
                 subSet['time'] = np.arange(len(subSet['power'].values))# Making time s
         tamps uniform. This is just an incremental order
                 #given to timestamps. e.d 0,1,2,3,4 instead of UNIX timestamps which c
         ompromises uniformity
                 subSet.rename(index=str, columns={"reservation_id": "machine", "power"
         : "pow", "sample_time":"timeStamp"}, inplace=True)
                 #The headers are re-named to match the writing data frame for seamless
          stitching of records
                 writingDF = writingDF.append(subSet)
             savingPath = 'C:\\Users\\Labyrinth\\JUPYTER NOTEBOOKS\\WeWash Praktikum TU
         M3sem\\WeWash Analysis ver2\\Data\\intermediate\\'
             fileName = 'uniform_'+str(len(machines))+'_'+str(machines[0])+'-'+str(machines
         ines[len(machines)-1])
             file write(writingDF, path= savingPath, name= fileName)
         machines: 50
```

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:10: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

# Remove the CWD from sys.path while we load stuff.

machines: 150
machines: 146
machines: 400

## This part prepares the Labelled data for the transformation to the training format

In [45]: #Change working directory to labelled training data folder
 os.chdir('C:\\Users\\Labyrinth\\JUPYTER NOTEBOOKS\\WeWash\_Praktikum\_TUM3sem\\W
 eWash\_Analysis\_ver2\\Data\\intermediate\\labelled\\train')

```
In [48]:
         bigDF = pd.DataFrame() #This is the master dataframe which will contain all th
         e training data. In this case we shall have
         #data for about 146+150+400 = 900 Machines. Individual records are definitely
          way more than that.
         for file in glob.glob("*.csv"): #Iterating through all files in the directory
             labtsDF = pd.read_csv(_file, delimiter=';')
             machines = labtsDF['machine'].unique() #Uniques machines are filtered base
         d on id
             print 'machines: ', len(machines)
             for i in range(len(machines)):#Each of the machines has multiple records.
                 tempDF = labtsDF[labtsDF['machine'] == machines[i]]#now that we have a
          subset of the dataframe filtered on
                 #the machine, we'll add features to it and then append it to the main
          dataframe.
                 tempDF['pow'] = floatConv(tempDF['pow'].values)#float conversions
                 tempDF['time'] = floatConv(tempDF['time'].values)#float conversions
                 tempDF['scaledPower'] = perScale(tempDF['pow'].values)#percentage scal
         ing
                 tempDF['scaledTime'] = perScale(tempDF['time'].values)#percentage sca
         ling
                 tempDF['powMin'] = minFil(tempDF['pow'].values, win=5)#applying minimu
         m filter to main power values
                 tempDF['scaledPowMin'] = minFil(tempDF['scaledPower'].values, win=5)#a
         pplying minimum filter to scaled power values
                 tempDF['scaledPowDiff'] = diffIndexMaker(tempDF['scaledPower'].values)
         #creating the difference index from scaled power
                 bigDF = bigDF.append(tempDF)
         print len(bigDF['machine'].unique())
         print bigDF.head()
```

machines: 118

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:11: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

# This is added back by InteractiveShellApp.init\_path()

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:12: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

if sys.path[0] == '':

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:13: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

del sys.path[0]

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:14: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:15: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

from ipykernel import kernelapp as app

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:16: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

app.launch\_new\_instance()

C:\Users\Labyrinth\Anaconda\lib\site-packages\ipykernel\_launcher.py:17: Setti
ngWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

machines: 146

3

4

0.0

1.0

NaN

NaN

3.0

4.0

machines: 150 machines: 282 696 powMin scaledPowDiff scaledPowMin scaledPower end\_spin machine pow 0 NaN 211286 0.094 0.086 0.0 0.0 0.0 1 NaN 211286 0.090 0.082 0.0 0.0 0.0 2 211286 0.086 0.0 0.0 0.0 NaN 0.082 3 NaN 211286 0.086 0.082 0.0 0.0 0.0 4 NaN 211286 0.086 0.074 0.0 0.0 0.0 scaledTime spins time 0 0.0 NaN 0.0 1 0.0 NaN 1.0 2 0.0 NaN 2.0

```
In [49]: #It is extremely important to fill all NaN values because otherwise it will ha
    mper the training process
    bigDF['end_spin'] = bigDF['end_spin'].fillna(0.0)
    bigDF['spins'] = bigDF['spins'].fillna(0.0)
    bigDF['pow'] = bigDF['pow'].fillna(0.0)
```

```
In [50]: #Time to finally write the training dataframe to a csv file
    savingPath = 'C:\\Users\\Labyrinth\\JUPYTER NOTEBOOKS\\WeWash_Praktikum_TUM3se
    m\\WeWash_Analysis_ver2\\Data\\processed\\'
    fileName = 'UniFeatScaledV2_146-150-400_696_128895-212236'
    file_write(bigDF, path=savingPath, name=fileName)
```