## 1) Introduction

#### 1.1) Project Statement

QR Space would like to gain insight on lower ground and third floor desk occupation of workshop 17 office.

Desk occupancy is monitered by the IoT occupancy sensor, the Infinity PIR1. The sensors measure movement over and output a signal based on whether or not motion was detected over a period.

The sensors output is as follows - 0 = no motion 1 = motion detected 3 = hearbeat signal (a signal that is generated at regular intervals to indicate that sensor is working correctly and should be ignored)

### 1.2) Objectives

- . To create three insightful graphs from the data that could help a decision maker of the property.
- Build a dasboard for data visualisation.

## 2) Loading Libraries

```
import numpy as np
import pandas as pd
```

# 3) Loading CSV Datasets

```
root_path = ''
LG_floor_df= pd.read_csv(root_path + 'September 2021 - Lower Ground Floor.csv')
Third_floor_df= pd.read_csv(root_path + 'September 2021 - Third Floor.csv')
```

# 4) Explore Data Analyses

## 4.1 ) Dataframe Overview

```
DateTime Data Reading

0 2021/09/01 8:58:30 PM 0 No Event

1 2021/09/01 8:45:54 PM 1 NaN

2 2021/09/01 8:12:23 PM 0 No Event

3 2021/09/01 7:59:30 PM 1 NaN

4 2021/09/01 5:18:07 PM 0 No Event
```

```
In [5]: Third_floor_df.head()
Out[5]: DateTime Data Reading
```

```
        0
        2021/09/01 5:40:08 PM
        0
        No Event

        1
        2021/09/01 5:28:22 PM
        3
        NaN

        2
        2021/09/01 5:27:28 PM
        1
        NaN

        3
        2021/09/01 5:21:59 PM
        0
        No Event

        4
        2021/09/01 5:09:38 PM
        1
        NaN
```

- Both dataframe have three columns (features) named DateTime, Data and Reading .
- The DateTime feature should be treated for better data visualisation (corrected below).

```
In [6]: Data_Unique_LG = list(LG_floor_df['Data'].unique())
```

```
Data_Unique_3rd = list(Third_floor_df['Data'].unique())
print(f'Lower ground Data column has the following unique enitres {Data_Unique_LG}')
print(f'Third floor Data column has the following unique enitres {Data_Unique_3rd}')
```

Lower ground Data column has the following unique enitres [0, 1, 3] Third floor Data column has the following unique enitres [0, 3, 1]

The three unique entries are the expected output from the sensor.

```
In [7]:
    Reading_Unique_LG = list(LG_floor_df['Reading'].unique())
    Reading_Unique_3rd = list(Third_floor_df['Reading'].unique())
    print(f'Lower ground Reading column has the following unique enitres {Reading_Unique_LG}')
    print(f'Third floor Reading column has the following unique enitres {Reading_Unique_3rd}')
```

Lower ground Reading column has the following unique enitres ['No Event', nan] Third floor Reading column has the following unique enitres ['No Event', nan]

- Both Reading coloumn only have 'No Event 'obsersations
- Other observations are 'NaN', this indicates empty slots
- This is not as expected. The colomn should have 'No Event', 'Event' and 'Heartbeat' corresponding the three unique entries in the Data column.

```
In [8]:
        LG floor df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 360 entries, 0 to 359
       Data columns (total 3 columns):
        # Column
                     Non-Null Count Dtype
        - - -
            -----
                      -----
        O DateTime 360 non-null
                                     object
                      360 non-null
        1
            Data
                                     int64
            Reading
                      166 non-null
                                     object
        dtypes: int64(1), object(2)
        memory usage: 8.6+ KB
```

```
In [9]: LG_floor_df.isnull().sum()
Out[9]: DateTime 0
```

Data 0
Reading 194
dtype: int64

The Lower Ground dataframe has :

- 360 rows of entries
- DateTime and Data column have 360 observations
- Reading column only have 166 observations and the remaining 194 cells are empty

```
In [10]:
         Third_floor_df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 332 entries, 0 to 331
        Data columns (total 3 columns):
         #
             Column Non-Null Count Dtype
                       -----
             -----
             DateTime 332 non-null
         0
                                      object
                       332 non-null
                                      int64
             Data
         2 Reading 150 non-null
                                      object
        dtypes: int64(1), object(2)
        memory usage: 7.9+ KB
```

```
In [11]: Third_floor_df.isnull().sum()
```

Data 0 Reading 182 dtype: int64

The Lower Ground dataframe has:

- 332 rows of entries
- DateTime and Data column have 332 observations
- Reading coloumn only have 150 observations and the remaining 182 cells are empty

## 4.2 ) DataFrame Treatment

### 4.2.1 ) DateTime Feature treatment function

The function below will be used to split DateTime and to fill missing values in the Reading column

```
In [12]:
           def dataframe corrector(df):
               ## Creating new columns
               df['Date']=''
               df['Time']=''
               df['AM/PM']=''
               for index,row in df.iterrows():
                    \#\# Splitting the DateTime column into Date ,Time and AM/PM
                    date_time = df.iloc[index]['DateTime']
                   splited_datetime = date_time.split()
                   date_ = splited_datetime[0]
time_ = splited_datetime[1]
                   am pm = splited datetime[2]
                   df.at[index,'Date']=date_
                    df.at[index,'Time']=time
                   df.at[index,'AM/PM']=am_pm
                   #### populating empty cell in the Reading column with values corrensponding Data column
                   if str(df.iloc[index]['Reading'])== 'nan' and df.iloc[index]['Data'] == 0 :
                        df.at[index,'Reading'] ='No Event'
                   if str(df.iloc[index]['Reading']) == 'nan' and df.iloc[index]['Data'] == 1 :
                   df.at[index,'Reading'] = 'Event'
if str(df.iloc[index]['Reading']) == 'nan' and df.iloc[index]['Data'] == 3 :
                        df.at[index,'Reading'] ='Heartbeat'
               return df
```

#### 4.2.2) Applying the function to the dataframe

```
In [13]:
          dataframe corrector(LG floor df)
          LG floor df=LG floor df.drop(['DateTime'], axis=1)
In [14]:
          dataframe_corrector(Third_floor_df)
          Third_floor_df=Third_floor_df.drop(['DateTime'], axis=1)
In [15]:
          LG_floor_df.isnull().sum()
Out[15]: Data
                    0
         Reading
                    0
         Date
                    0
         Time
                    0
         AM/PM
                    0
         dtype: int64
```

```
In [16]: Third_floor_df.isnull().sum()
```

```
Out[16]: Data 0
Reading 0
Date 0
Time 0
AM/PM 0
dtype: int64
```

Finally both dataframe have no empty cell

# 5) Data Visualisation

For visualisation data column is redundant, because the sensor output is represented by the Reading column,hence should be dropped

```
In [17]: ## Dropping Data column
   LG_floor_df = LG_floor_df.drop(['Data'], axis=1)
   Third_floor_df = Third_floor_df.drop(['Data'], axis=1)
```

The Heartbeat observation has no usefull insight and should be ignored

```
In [18]: ## Dropping Heartbeat Observation
   LG_floor_df = LG_floor_df[LG_floor_df['Reading'] != 'Heartbeat']
   Third_floor_df = Third_floor_df[Third_floor_df['Reading'] != 'Heartbeat']
```

# Detailed Data Visualisation is perforned on Power BI

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