1 Introduction

This project's main focus and ambition was to develop tools to analyze data. The data in question was a set of measurements on air quality data for the year of 2014 to 2015 from the Grensásvegur 15 station. The data was taken from the station's website and is registered at 30 minute intervals throughout the whole period, with some exceptions that might be due to instrument failure or something else.

With this in mind, a dashboard was created which consists of two streaming charts, one of which contained data on NO_2 and H_2S levels while the other contained data on PM_{10} and $PM_{2.5}$ levels. These chemicals and particulates are thought to be correlated with traffic as they are produced by fossil fuel burning vehicles and spiked winter tyres. Seeing that the data was registered at such regular intervals it was decided to try and develop something which might resemble the effect of looking at the data while streaming. This was thought to be characteristic for the data and its main purpose as people vulnarable to bad air quality could benefit from having access to this kind of information. Furthermore it was thought to be plausible that this kind of a tool might be useful for analyzing historical data. Thus it was decided to make an attempt at creating a graphical animation out of air quality levels.

Seeing that car traffic and winter tyres are suspected to be the main cause of bad air quality it was decided that an important question to answer would be whether this would reflect as increased concentration of toxic levels depending on main traffic hours and winter time. In order to answer this question each month in the data set was analyzed by looking at the mean value for each time of measurement throughout the whole month. That is all the measurements that take place at the same HH:MM throughout the month in question are combined in a subset of the data set from which the mean value was calculated. From this an average day throughout the month was obtained, and thus the possibility of looking at the date in relation to traffic hours and winter time with the same tool.

In order to highlight the traffic hours, radiation levels measured were used, as most humans usually prefer driving during the day it is often bright during traffic hours (at least there are rarely traffic jams during the night). This was integrated into the graphics by making the background color of the animation a function of the radiation measured. A future extension of this tool could possibly integrate weather data in a similar way.

For the easiest comparison it was decided to put the ratio between the measured values and the severe health risk limit of measured quantities. Furthermore a horizontal line was drawn to mark the recommended health limit of measured quantities according to Icelandic authorities. These limits are different depending on authorities and thus this tool could potentially aid in making a decision when defining the proper health limits. In fact it is surprising to see that the particulates levels are rather high throughout the whole year.

The tool was made so that it can be easily modified to take other variables into account or to animate anything else with respect to quantities and time. It provided a clear picture of the measured quantities of dangerous toxics that that are a health risk to the public and thus should be carefully monitored. A further extesion to this tool could be to smoothen the plot and make the animation smoother throughout it's duration.

The Python packages Pandas, Datetime, dateutil and Matplotlib were the main packages that were used to develop the tool of analysis. Pandas is a very powerful when handling big data, matplotlib is the plotting package and Datetime and dateutil provide very smart ways of manipulating data which is based on timeseries.

2 Code

```
# Made by Thorsteinn Hjortur Jonsson
  # Date 13/09'15
 # This tool uses matplotlib and pandas to make an animation using data on air
     quality
  import numpy as np
5 from matplotlib.pylab import *
  import matplotlib.animation as animation
7 from mpl_toolkits.axes_grid1 import host_subplot
  from datetime import *
9 from matplotlib.colors import cnames
  from dateutil.relativedelta import *
  import calendar
  import pdb
13 import matplotlib.dates as mdates
  import va_read_data as parse
  filename = 'air_data.csv'
 # We read in the data
  columns = parse.read_in(filename)
 start = date(2014, 2, 1)
  final = date(2015, 9, 1)
21 no2_data = parse.mean_rng(filename, start, final, 'NO2')
  rad_data = parse.mean_rng(filename, start, final, 'Radiation')
23 h2s_data = parse.mean_rng(filename, start, final, 'H2S')
  pm10_data = parse.mean_rng(filename, start, final, 'Dust_PM10')
 pm25_data = parse.mean_rng(filename, start, final, 'Dust_PM2.5')
  # We have to make a list of all the times animated
 TimeAxis = []
  # Count number of steps
_{29} | n = 0
 for month in no2_data:
   n = n + len(no2_data[month])
  # Number of hours in each figure
 nPerFrame = 24
  # To build the x axis
for i in np.arange(0,n):
     Time Axis.append (mdates.date2num (datetime.strptime (columns ['Date'] [i] + " " +
     columns['Time'][i],"%d.%m.%Y %H:%M")))
37
                     ----- Set up the rc settings,
  font = {'size' :9}
 matplotlib.rc('font', **font)
  #----- Setup figure and subplots
 f0 = figure(num = 0, figsize = (12,8)) #, dpi = 100)
 f0.suptitle('Measurements of air in Reykjavik', fontsize=12)
45
  \#-----Define the subplot/subplotgrid
 ax01 = subplot2grid((2,2),(0,0), colspan=2, rowspan=1)
  ax02 = subplot2grid((2,2),(1,0), colspan=2, rowspan=1)
49
                           ----- Plot Settings
 ax01.set_ylim(-0.1,3.5)
  ax02.set_ylim(-0.1,5.0)
  ax01.set_xlim(TimeAxis[0],TimeAxis[nPerFrame])
 ax02.set_xlim(TimeAxis[0],TimeAxis[nPerFrame])
ax01.set_xlabel('Average Value at Time of Day During the Month')
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ax02.set_xlabel('Average Value at Time of Day During the Month')
  ax01.set_ylabel('Ratio between measured level and dangerous levels')
61 ax02.set_ylabel('Ratio between measured level and dangerous levels')
  ax01.set_xticks(TimeAxis[0:nPerFrame])
  ax02.set_xticks(TimeAxis[0:nPerFrame])
  ax01.xaxis.set_major_formatter(mdates.DateFormatter('%H:%M'))
  ax02.xaxis.set_major_formatter(mdates.DateFormatter('%H:%M'))
69 Plot_limit1 = ax01.axhline(y = 75.0/30.0,color='r',lw = 2,label = 'Recommended
     Limit of NO2 Levels')
  Plot_limit2 = ax02.axhline(y = 50.0/20.0, color='r', lw = 2, label = 'Recommended
     Limit of PM10 Levels')
  # These vectors will be the ones used for the animation
  time = []
  pm10 = []
  pm25 = []
no2 = []
  h2s = []
   # Set the plots
  Plot_pm10, = ax02.plot_date(time, pm10, '-',color='#b58900', lw = 2, label = '
      PM10')
  Plot_no2, = ax01.plot_date(time, no2, '-',color='#b58900',lw = 2, label = 'NO2')
  Plot_pm25, = ax02.plot_date(time, pm25, '-', color='g', lw = 2, label = 'PM2.5')
  Plot_h2s, = ax01.plot_date(time, h2s, '-', color='g', lw = 2, label = 'H2S')
  # Set legends
  ax02.legend([Plot_pm10, Plot_pm25, Plot_limit2], [Plot_pm10.get_label(),
      Plot_pm25.get_label(), Plot_limit2.get_label()])
  ax01.legend([Plot_no2, Plot_h2s, Plot_limit1], [Plot_no2.get_label(), Plot_h2s.
      get_label(), Plot_limit1.get_label()])
  # Format the x-axis for dates (label formatting, rotation)
  f0.autofmt_xdate(rotation=45)
  f0.tight_layout()
  backgrCol = '#073642'
  xmax = nPerFrame
  x = 0
  i = 0
95 month = date (2014, 2, 1)
  maxRad = max(rad_data[month])
  # animation function, this is called sequentially
  def updateData(self):
      # We need to define global variables to run the loop through updateData
99
      global month
      global maxRad
1.01
      global x
      global j
      global time
      global pm10
      global alpha
      # We need to change months when we reach the end of the corresponding list
      if j == len(no2_data[month]):
109
          month = month + relativedelta(months=+1)
      # look at the maximum radiation for the month
           maxRad = max(rad_data[month])
113
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```
if x > xmax:
115
           # We can pop to save memory if x >= xmax we won't see those values any
      way
           time.pop(0)
           time.append(TimeAxis[x])
           no2.pop(0)
           no2.append(no2_data[month][j]*(1.0/30.0))
           pm10.pop(0)
121
           pm10.append(pm10_data[month][j]*(1.0/20.0))
           pm25.pop(0)
           pm25.append(pm25_data[month][j]*(1.0/20.0))
           h2s.pop(0)
125
           h2s.append(h2s_data[month][j]*(1.0/30.0))
           # Set the limits accordingly
           ax02.set_xlim(TimeAxis[x-xmax],TimeAxis[x+1])
129
           ax02.set_xticks(TimeAxis[x-xmax:x+1])
           ax01.set_xlim(TimeAxis[x-xmax],TimeAxis[x+1])
           ax01.set_xticks(TimeAxis[x-xmax:x+1])
1.33
           # The radiation function which is in charge of the background color
           radiation = np.abs(1.0 - rad_data[month][j]*1.0/maxRad)
           ax01.axvspan(TimeAxis[x],TimeAxis[x+1],facecolor=backgrCol,alpha=
137
      radiation, lw = 0)
           ax02.axvspan(TimeAxis[x],TimeAxis[x+1],facecolor=backgrCol,alpha=
      radiation, lw = 0)
       else:
139
           # This is run at the beginning, we don't want to pop here
           time.append(TimeAxis[x])
141
           no2.append(no2_data[month][j]*(1.0/30.0))
           pm10.append(pm10_data[month][j]*(1.0/20.0))
143
           pm25.append(pm25_data[month][j]*(1.0/20.0))
           h2s.append(h2s_data[month][j]*(1.0/30.0))
145
           # To set the background color
           radiation = np.abs(1.0 - rad_data[month][j]*1.0/maxRad)
147
           ax01.axvspan(TimeAxis[x],TimeAxis[x+1],facecolor=backgrCol,alpha=
      radiation, lw = 0)
           ax02.axvspan(TimeAxis[x],TimeAxis[x+1],facecolor=backgrCol,alpha=
149
      radiation, lw = 0)
       # The title is dynamic and tells the month
       ax01.set_title(month.strftime('%B, %Y'))
151
       Plot_no2.set_data(time, no2)
       Plot_pm10.set_data(time,pm10)
       Plot_pm25.set_data(time,pm25)
       Plot_h2s.set_data(time, h2s)
       # x and j must be incremented equally so that the dimensions remain the same!
       x = x + 1
       j = j+1
       return Plot_no2, Plot_h2s, Plot_pm25, Plot_pm10
  anim = animation.FuncAnimation(f0, updateData, frames=n, interval=10, blit=False,
       repeat = False)
  anim.save('va_animation.mp4', fps=2, extra_args=['-vcodec','libx264'])
  plt.show()
```

va_animate.py

```
#!/bin/python
# Made by Thorsteinn Hjortur Jonsson
```

```
3 # Date: 13/09'15
  # This module contains csv readers, the first one does it the old fashioned way
 # The other one does not.
  import csv
7 import numpy as np
  import pandas as pd
  from datetime import *
  from dateutil.relativedelta import *
 import calendar
  import pdb
13
  # This is a basic csv reader
 # Future versions should render this obsolete
  # Currently it provides us time axis support
 def read_in(filename):
    with open(filename, 'rb') as csvfile:
19
      reader = csv.reader(csvfile, delimiter = '\t', quotechar = '|')
21
      headers = reader.next()
      data = \{\}
      for header in headers:
25
        data[header] = []
      for row in reader:
        for header, value in zip(headers, row):
          data[header].append(value)
    return data
 # This function yields a list of the mean values at each mesurement interval
  # for a given date range
 def mean_rng(filename, start,final, chemical):
    dateparse = lambda x: pd.datetime.strptime(x, '%d.%m.%Y %H:%M')
    air_data = pd.read_csv(filename, sep='\t', parse_dates=[['Date', 'Time']],
     date_parser = dateparse, index_col=['Date_Time'], comment='#')
        # Make a list that contains the half an hourly mesurement intervals
        rng0 = pd.date_range('00:30','23:30',freq = 'H')
41
        rng1 = pd.date_range('00:00','23:00',freq = 'H')
        TimeRange = rng1.union(rng0)
43
    # Make a list of months
45
        # This list will contain the data for the given chemical for all the items
     in DateRanges
        # The mean value of each list in DateRanges_data
47
        DateRange_mean = {}
49
    while start < final:
      end = start + relativedelta(months=+1) - relativedelta(days=+1)
51
          # This list will contain the half an hourly mesurement intervals as
          TimeStr = []
          # This list will contain the different date ranges according to the items
      in TimeStr
          Measure_range = []
      DateRange_mean.update({start: []})
          for i in np.arange(0,48):
              # strings of measurement times
              TimeStr.append(TimeRange[i].strftime('%H:%M'))
59
              # date range for a given measurement time
```

```
Measure_range.append(pd.date_range(start.strftime('%m-%d-%Y') + ' ' ' +
TimeStr[i],end.strftime('%m-%d-%Y') + ' ' + TimeStr[i], freq = '24H'))

# makes lists of the data during the Date ranges
DateRanges_data = pd.Series(air_data[chemical], index = Measure_range
[i])

DateRanges_data.tolist()
DateRange_mean[start].append(DateRanges_data.mean())
start = start + relativedelta(months=+1)
return DateRange_mean
```

va read data.py