

Unit4: For Live Session

DS6306

Garrity

PART 1A

Number of Sushi restaurants in the Baltimore area.

```
> df_sushi <- data.frame(name = sushi_name, zipcode = sushi_zipcode, council=sushi_council)
> str(df_sushi)
```

```
'data.frame':      1327 obs. of  3 variables:
 $ name      : Factor w/ 1277 levels "#1 CHINESE KITCHEN",...: 9 3 992 1 2 4 5 6 7 8 ...
 $ zipcode: Factor w/ 32 levels "-21226","21201",...: 5 27 21 10 20 17 4 10 4 27 ...
 $ council: Factor w/ 14 levels "1","10","11",...: 7 1 1 6 14 6 5 12 5 1 ...
```

```
> grep("sushi", ignore.case = TRUE, df_sushi$name)
[1] 17 90 249 250 391 457 537 725 1137
> length(grep("sushi", ignore.case = TRUE, df_sushi$name))
[1] 9
```

There appears to be 9 sushi restaurants in the Baltimore area.

PART 1B

Number of Sushi restaurants in downtown Baltimore.

```
> df_sushi_11 <- df_sushi %>% filter(council == 11)

> length(grep("sushi", ignore.case = TRUE, df_sushi_11$name))
[1] 1

> df_sushi_11[grep("sushi", ignore.case = TRUE, df_sushi_11$name),]
      name zipcode council
81 EDO SUSHI   21202      11
```

There is one sushi restaurant (Edo Sushi) in downtown Baltimore.

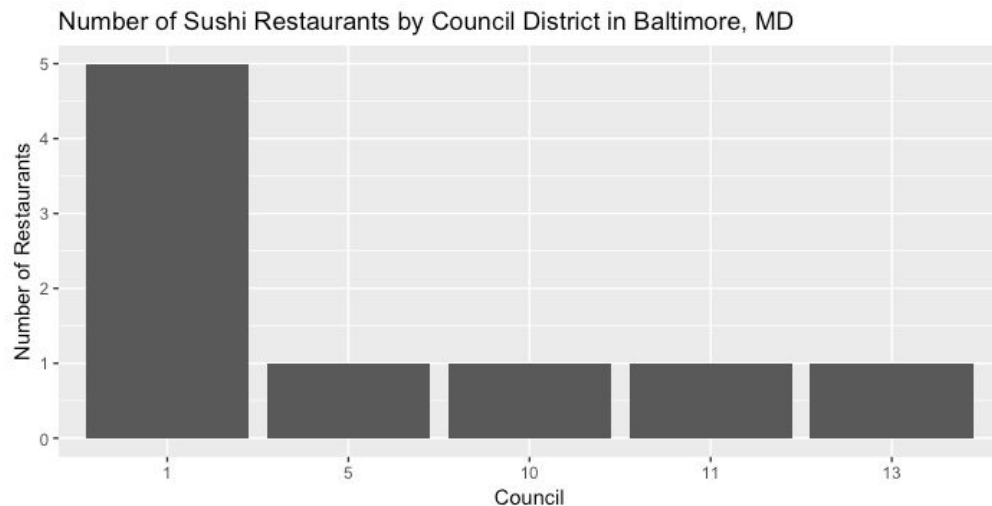
PART 1C

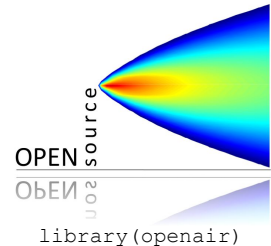
Plot of the number of sushi restaurants in each council district.

```
sushi_by_council_summarize <- sushi_by_council %>% group_by(council) %>% summarize(count = n())

sushi_by_council_summarize$council =
  factor(sushi_by_council_summarize$council,
        level = sushi_by_council_summarize$council[order(as.numeric(as.character(sushi_by_council_summarize$council)))])

sushi_by_council_summarize %>% ggplot(aes(x = council, y = count)) + geom_col() +
  labs(title="Number of Sushi Restaurants by Council District in Baltimore, MD", y="Number of Restaurants", x ="Council")
```





OpenAir is an API providing access to current and historic air quality observations from 244 locations across the UK. It is extremely easy to use and comes with several helpful functions. I used the API to compare one year (2018) of air quality observations from an urban location in London (MY1) to a rural area to the west of London (CHBO). <http://davidcarslaw.github.io/openair/index.html>

```
library(openair)
```

```
#load metadata for all sites:
```

```
aurndetailed <- importMeta(source =  
"aurnd", all = TRUE)
```

```
#plot site locations:
```

```
library(leaflet)
```

```
m2 <- leaflet() %>%
```

```
  addProviderTiles(providers$Wikimedia)
```

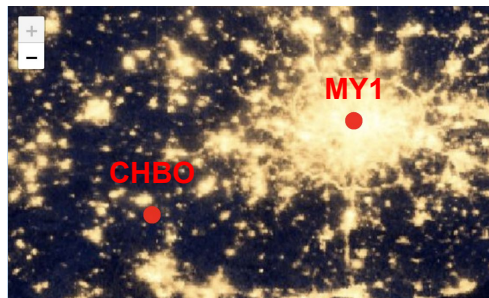
```
%>%
```

```
addCircleMarkers(lng=aurndetailed$longit  
ude, lat=aurndetailed$latitude, radius =  
6, color = "navy", stroke = FALSE,  
fillOpacity = 0.2, popup =  
aurndetailed$code) %>%
```

```
addCircleMarkers(lng=aurndetailed$longit  
ude[aurndetailed$code == "CHBO" |  
aurndetailed$code == "MY1"],  
lat=aurndetailed$latitude[aurndetailed$  
code == "CHBO" |aurndetailed$code ==  
"MY1"], radius = 6, color = "red",  
stroke = FALSE, fillOpacity = 0.5, popup  
= aurn_detailed$code[aurndetailed$code  
== "CHBO" | aurn_detailed$code == "MY1"])
```

```
#display map:
```

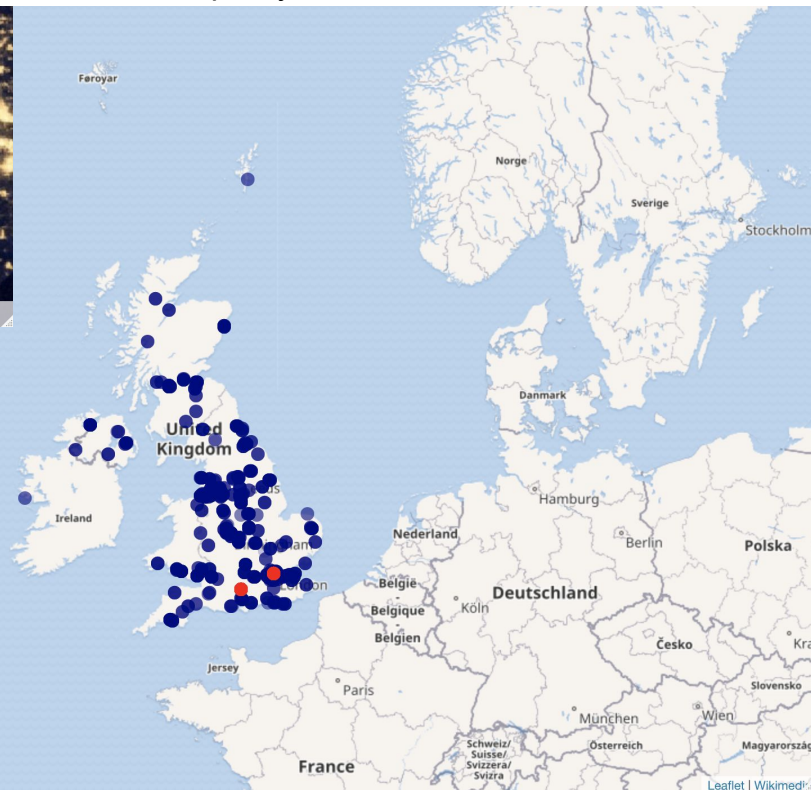
```
m2
```



Leaflet | Imagery provided by services from the Global Imagery Browse Services (GIBS), operated by the NASA/GSFC/Earth Science Data and Information System (ESDIS) with funding provided by NASA/HQ.

London area at night

Air quality observation stations across the UK



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Data wrangling

```
# list all air quality parameters
```

```
str(aurn_detailed$parameter)
```

```
# list all urban sites
```

```
urban_background_sites <- filter(aurn_detailed, site.type == "Urban Background")
```

```
nrow(urban_background_sites)
```

```
# list all rural sites
```

```
rural_background_sites <- filter(aurn_detailed, site.type == "Rural Background")
```

```
nrow(rural_background_sites)
```

```
# download data for CHBO (rural) and MY1 (urban)
```

```
chbo <- importAURN(site= "CHBO", year = 2018)
```

```
str(chbo)
```

```
my1 <- importAURN(site = "MY1", year = 2018)
```

```
str(my1)
```

```
# combine data frames, filling empty columns with NAs
```

```
require(plyr)
```

```
alldata = rbind.fill(my1,chbo)
```

```
# create a factor with four levels from the continuous wind direction variable
```

```
alldata$windCardinal = cut(alldata$wd, breaks = c(0,90,180,270,360), labels = c("NE","SE","SW","NW"))
```

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Data wrangling (tidy data)

```
# combine data frames, filling empty columns with NAs
```

```
require(plyr)

alldata = rbind.fill(my1,chbo)

> head(alldata)
      date code      site      o3      no2      co      so2 pm10      nox      no pm2.5 nv10  v10 nv2.5 v2.5  ws
wd
1 2018-01-01 00:00:00 MY1 London Marylebone Road 24.48059 79.05960 0.298901 4.92942 24.6 187.4120 70.66562 12.7 25.2 -0.6 14.5 -1.8 5.5
263.3
2 2018-01-01 01:00:00 MY1 London Marylebone Road 13.96990 97.05587 0.261539 6.75514 18.7 310.9482 139.49697 11.6 17.8 0.9 11.4 0.2 5.0
256.4
3 2018-01-01 02:00:00 MY1 London Marylebone Road 18.21076 101.07355 0.308242 6.43564 17.2 273.9910 112.77380 7.6 15.2 2.0 6.1 1.5 4.8
251.0
4 2018-01-01 03:00:00 MY1 London Marylebone Road 24.84647 81.56557 0.224176 5.54560 10.8 211.2405 84.57176 2.3 11.0 -0.2 4.3 -2.0 4.8
246.8
5 2018-01-01 04:00:00 MY1 London Marylebone Road 29.03744 59.85638 0.252198 5.06635 12.4 175.8730 75.66403 4.8 10.7 1.7 3.6 1.2 5.3
248.4
6 2018-01-01 05:00:00 MY1 London Marylebone Road 35.62325 51.55781 0.112088 3.90246 12.6 134.8788 54.34053 2.7 10.9 1.7 2.0 0.7 5.3
248.0

> tail(alldata)
      date code      site      o3      no2 co      so2  pm10      nox      no pm2.5 nv10  v10 nv2.5 v2.5  ws  wd
17515 2018-12-31 18:00:00 CHBO Chilbolton Observatory 76.38542 1.85279 NA 0.33295 11.125 2.42172 0.37105 6.132 NA NA NA NA 4.8 284.2
17516 2018-12-31 19:00:00 CHBO Chilbolton Observatory 76.63488 1.64820 NA 0.46613 9.875 2.37230 0.47225 5.590 NA NA NA NA 4.3 281.5
17517 2018-12-31 20:00:00 CHBO Chilbolton Observatory 75.18800 1.64820 NA 0.46613 9.150 2.37230 0.47225 5.590 NA NA NA NA 4.1 279.1
17518 2018-12-31 21:00:00 CHBO Chilbolton Observatory 69.74972 3.33778 NA 0.46613 11.775 3.85499 0.33732 8.326 NA NA NA NA 4.5 289.5
17519 2018-12-31 22:00:00 CHBO Chilbolton Observatory 71.14671 3.19181 NA 0.19977 9.700 3.65730 0.30359 6.344 NA NA NA NA 4.8 296.3
17520 2018-12-31 23:00:00 CHBO Chilbolton Observatory 69.15101 2.94929 NA 0.39955 9.075 3.31134 0.23612 6.179 NA NA NA NA 4.9 303.2
```

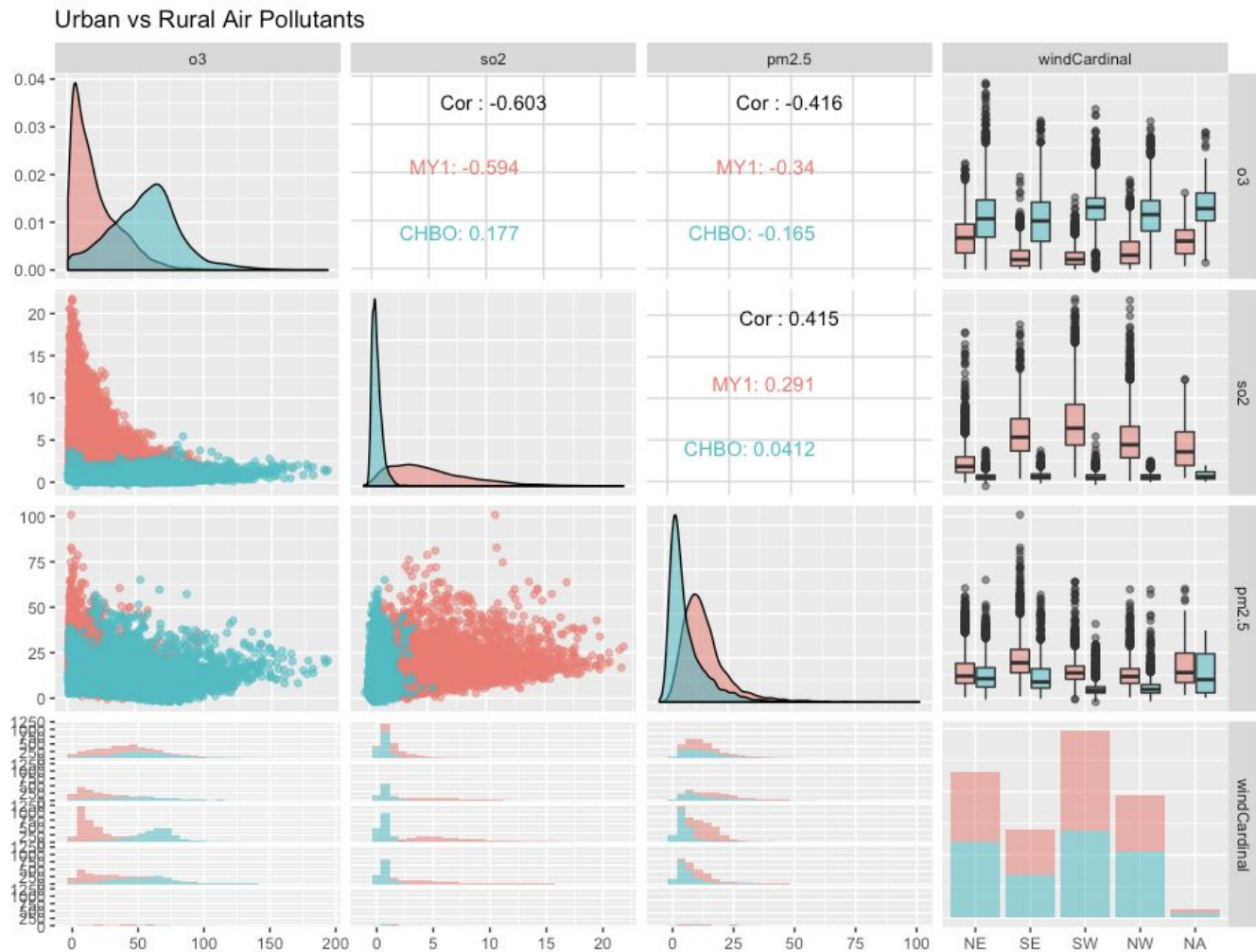
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openAir EDA

```
require(GGally)
```

```
alldata %>% select(o3, so2,  
pm2.5, windCardinal) %>%  
ggpairs(aes(color=alldata$code,  
alpha=0.2)) + ggtitle("Urban vs  
Rural Air Pollutants")
```

There appears to be some fairly stark differences in O_3 , SO_2 , and $PM_{2.5}$ between the urban (MY1) and rural (CHBO) sites. It also appears that concentrations of each pollutant vary somewhat with wind direction.



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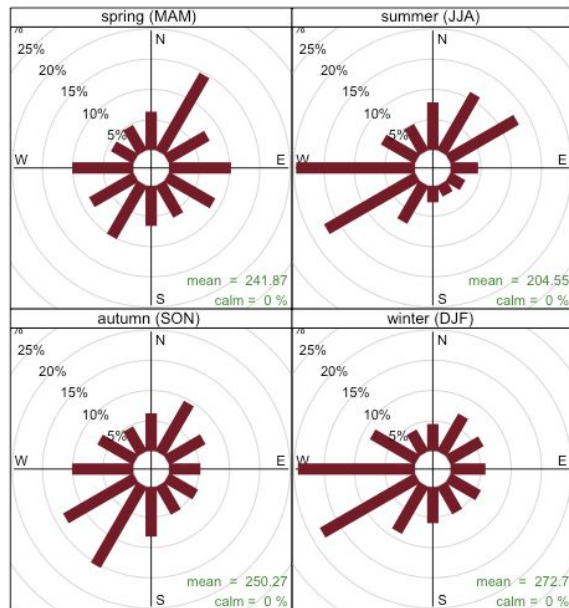
openAir windrose plots of NO_x concentration by season

```
# plot seasonal windrose  
windRose(my1, type =  
"season", pollutant = "nox",  
cols="heat", width=0.5,  
offset=15)
```

```
windRose(chbo, type =  
"season", pollutant = "nox",  
cols="heat", width=0.5,  
offset=15)
```

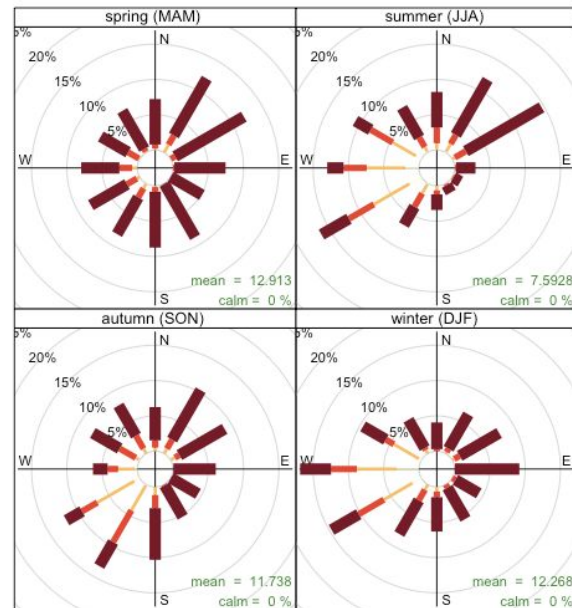
The windrose shows both frequency of wind direction at each site, as well as NO_x concentrations associated with each wind direction. Seasonal differences in wind direction are obvious for both sites, but only CHBO exhibits directional dependency for NO_x concentration, which varies somewhat by season.

MY1 (urban)



0 to 2 2 to 4 4 to 6 6 to 1008.6
(m s⁻¹)
Frequency of counts by wind direction (%)

CHBO (rural)



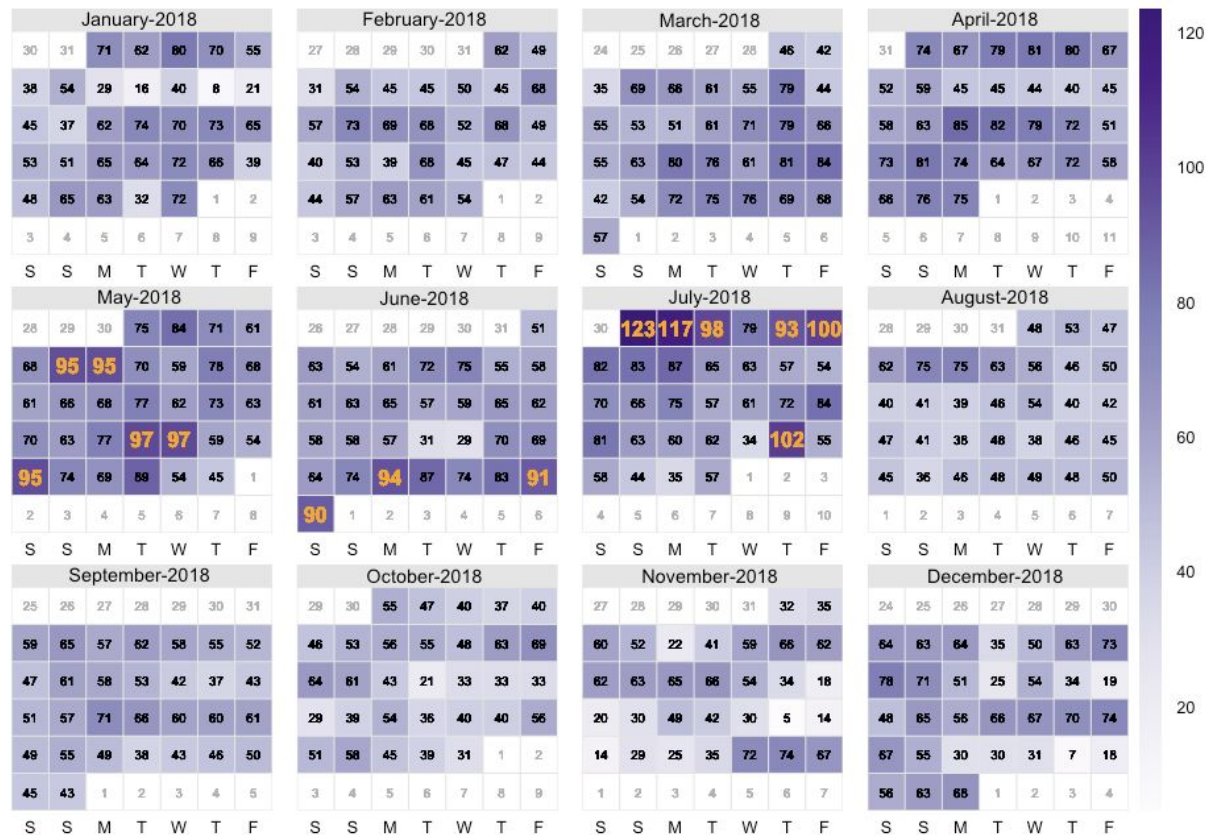
0 to 2 2 to 4 4 to 6 6 to 129.15
(m s⁻¹)
Frequency of counts by wind direction (%)

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Too cool not to include!

Daily ozone (O_3) concentrations at the CHBO site during 2018

```
calendarPlot(chbo, pollutant =  
"o3", annotate = "value", lim =90,  
cols = "Purples", col.lim =  
c("black", "orange"),  
layout = c(4, 3))
```



Takeaways & Questions

Why didn't the following code work?

```
> df_sushi %>% filter(council == 11) %>% grep("sushi", ignore.case = TRUE, name)
Error in grep(., "sushi", ignore.case = TRUE, name) :
  object 'name' not found
```

I had trouble with the *twitteR* package but found that *rtweet* worked fine and has most of the same functionality as *twitteR*.

Access to data has been a limiting factor for me in the past. The ability to scrape data and use APIs opens up a whole new world of information to ingest and process. This is definitely one of the tools I was hoping to add to the toolbelt through this program. Exciting!