R Notebook

```
## Loading required package: ggplot2
## Loading required package: ggpubr
## Loading required package: magrittr
## Loading required package: dplyr
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
      filter, lag
##
## The following objects are masked from 'package:base':
##
      intersect, setdiff, setequal, union
## Loading required package: janitor
## Loading required package: stringr
## Loading required package: tidyverse
## -- Attaching packages -----
                                  _____
## v tibble 1.4.2
                    v readr
                              1.3.1
## v tidyr 0.8.2
                    v purrr
                             0.2.5
## v tibble 1.4.2
                   v forcats 0.3.0
## -- Conflicts ------ tidyve
## x tidyr::extract()
                     masks magrittr::extract()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## x purrr::set_names() masks magrittr::set_names()
\mathbf{R}
Allows you to analyze data easily
R has a huge community and many solutions are available online
Many libraries offer ease abstractions to analyze and plot data
```

About environmental awareness

```
{\bf R} allows you to load data in many formats (e.g. csv files)
```

```
globalTemperatures <- read_csv("GlobalTemperatures.csv")
## Parsed with column specification:
## cols(
## dt = col_date(format = ""),
## LandAverageTemperature = col_double(),
## LandAverageTemperatureUncertainty = col_double(),
## LandMaxTemperature = col_logical(),</pre>
```

```
##
    LandMaxTemperatureUncertainty = col_logical(),
##
    LandMinTemperature = col_logical(),
    LandMinTemperatureUncertainty = col logical(),
##
    LandAndOceanAverageTemperature = col_logical(),
##
##
    LandAndOceanAverageTemperatureUncertainty = col_logical()
## )
## Warning: 11952 parsing failures.
                                             expected
                                                                                          file
## 1201 LandMaxTemperature
                                    'GlobalTemperatures.csv'
## 1201 LandMaxTemperatureUncertainty 1/0/T/F/TRUE/FALSE 1.7380000000000000
                                                                        'GlobalTemperatures.csv'
                                    1/0/T/F/TRUE/FALSE -3.2060000000000004 'GlobalTemperatures.csv'
## 1201 LandMinTemperature
## 1201 LandMinTemperatureUncertainty 1/0/T/F/TRUE/FALSE 2.822
                                                                        'GlobalTemperatures.csv'
## 1201 LandAndOceanAverageTemperature 1/0/T/F/TRUE/FALSE 12.83299999999999 'GlobalTemperatures.csv'
......
## See problems(...) for more details.
head(globalTemperatures, n=5)
## # A tibble: 5 x 9
##
              LandAverageTemp~ LandAverageTemp~ LandMaxTemperat~
##
    <date>
                        <dbl>
                                        <dbl> <lgl>
## 1 1750-01-01
                         3.03
                                         3.57 NA
## 2 1750-02-01
                         3.08
                                         3.70 NA
## 3 1750-03-01
                         5.63
                                         3.08 NA
## 4 1750-04-01
                         8.49
                                         2.45 NA
## 5 1750-05-01
                         11.6
                                         2.07 NA
## # ... with 5 more variables: LandMaxTemperatureUncertainty <lgl>,
      LandMinTemperature <lgl>, LandMinTemperatureUncertainty <lgl>,
## #
      LandAndOceanAverageTemperature <lgl>,
      LandAndOceanAverageTemperatureUncertainty <lgl>
```

Data manipulation

You can select columns and filter data in an ease and readable syntax

select(filter(globalTemperatures, dt > as.Date("1800-01-01")), dt, LandAverageTemperature)

```
## # A tibble: 2,591 x 2
##
     dt
                 LandAverageTemperature
                                  <dbl>
##
      <date>
                                   3.63
##
  1 1800-02-01
## 2 1800-03-01
                                   4.45
## 3 1800-04-01
                                   9.12
## 4 1800-05-01
                                  11.1
## 5 1800-06-01
                                  13.6
## 6 1800-07-01
                                  13.8
## 7 1800-08-01
                                  13.6
## 8 1800-09-01
                                  10.8
## 9 1800-10-01
                                  9.47
## 10 1800-11-01
                                   5.57
## # ... with 2,581 more rows
```

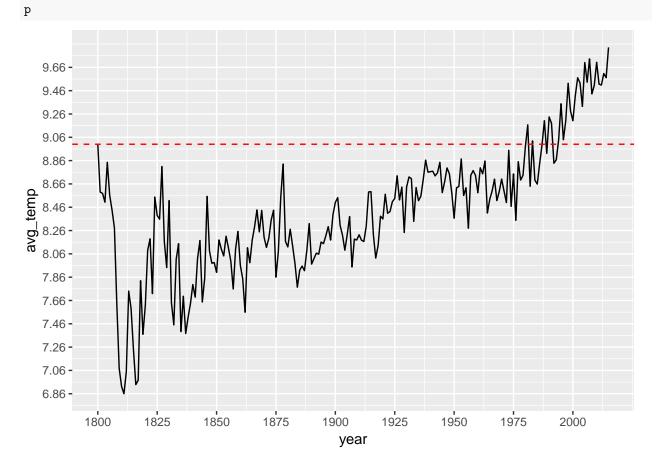
Data transformation

You can also transform your data applying operators (e.g. converting a date field to a numeric year field) as you seem fit

```
dt_temperature <- globalTemperatures %>%
  filter(dt > as.Date("1800-01-01")) %>%
  select(dt, LandAverageTemperature) %>%
  mutate(year = as.numeric(format(dt, "%Y")))
head(dt_temperature, n=5)
```

```
## # A tibble: 5 x 3
##
     dt
                 LandAverageTemperature year
##
     <date>
                                   <dbl> <dbl>
## 1 1800-02-01
                                    3.63
                                          1800
## 2 1800-03-01
                                    4.45
                                          1800
## 3 1800-04-01
                                    9.12
                                          1800
## 4 1800-05-01
                                          1800
                                   11.1
## 5 1800-06-01
                                   13.6
                                          1800
```

Data visualization



Hypothesis Testing

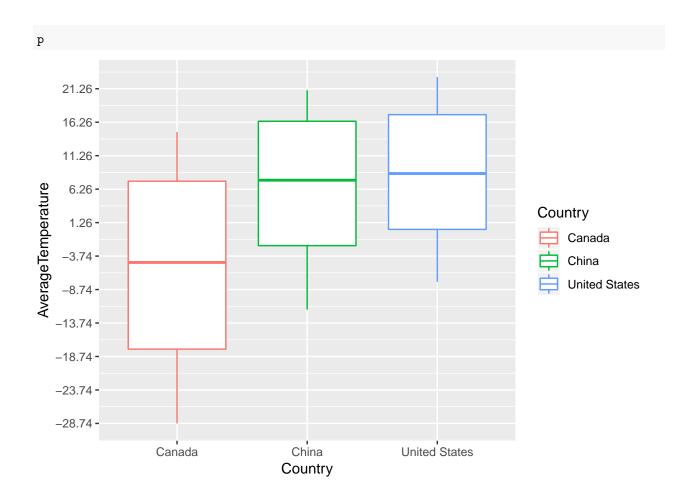
```
How is the temperature around our neighbours?
globalTemperaturesByCountry <- read_csv("GlobalLandTemperaturesByCountry.csv")</pre>
## Parsed with column specification:
## cols(
     dt = col_date(format = ""),
##
##
     AverageTemperature = col_double(),
##
     AverageTemperatureUncertainty = col_double(),
##
     Country = col_character()
## )
unique(globalTemperaturesByCountry %>% select(Country))
## # A tibble: 243 x 1
##
     Country
##
      <chr>
## 1 Åland
## 2 Afghanistan
## 3 Africa
## 4 Albania
## 5 Algeria
## 6 American Samoa
## 7 Andorra
## 8 Angola
## 9 Anguilla
## 10 Antarctica
## # ... with 233 more rows
```

Data transformation

Libraries such as dplyr and tidyverse allow cleaning our data and removing non-valid entries

```
dt_temperature <- globalTemperaturesByCountry %>%
  na.omit() %>%
  filter(dt > as.Date("1800-01-01")) %>%
  filter(Country == 'Canada' | Country == 'China' | Country == 'United States') %>%
  select(dt, AverageTemperature, Country) %>%
  mutate(year = as.numeric(format(dt, "%Y")))
```

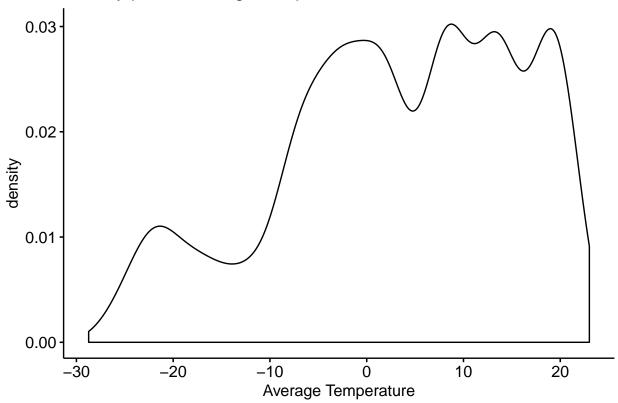
Comparing data visually



Testing Normal Distribution

```
ggdensity(dt_temperature$AverageTemperature,
    main = "Density plot of Average Temperature",
    xlab = "Average Temperature")
```

Density plot of Average Temperature



Checking if our data has an equal number of entries for comparison

```
us_temp <- dt_temperature %>% filter(Country == 'United States') %>% select(AverageTemperature)
cn_temp <- dt_temperature %>% filter(Country == 'China') %>% select(AverageTemperature)
ca_temp <- dt_temperature %>% filter(Country == 'Canada') %>% select(AverageTemperature)
length(us_temp$AverageTemperature)

## [1] 2399

## [1] 2201

length(ca_temp$AverageTemperature)

## [1] 2399
```

Creating a data frame for comparison

```
## 4 -3.116 United States
## 5 1.602 United States
```

Hypothesis testing

```
###
## Kruskal-Wallis rank sum test
##
## data: Country by AverageTemperature
## Kruskal-Wallis chi-squared = 6243.1, df = 6047, p-value = 0.03832
```

RStudio is awesome!

Did I mention that this notebook was created on RStudio?