

Analyzing NHC Tropical Storm Data

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In this script, we want to analyze the National Hurricane Center (NHC) storm tracks data covering the Atlantic Basin for the years 1975 to 2021, except for the last storm in 2021 (Wanda). This dataset has been made available to us through the dplyr package.

Questions to Address:

1. Do Higher Wind Speeds Correspond to Wider Tropical Storm Diameter?
2. Which Months Typically Witness More Intense Storms?
3. Spatial distributions of the storms before and after year 2000

```
## Load required packages
library(tidyverse) # For data manipulation and plotting
library(dplyr)     # For data manipulation
library(sf)        # For spatial data handling
library(OpenStreetMap) # For working with OpenStreetMap data
library(gridExtra) # For arranging plots

# Load the 'storms' data from dplyr package
storms
```

Exploring the Data

```
# Check for the latest data available
tail(storms)
```

```
## # A tibble: 6 x 13
##   name  year month  day hour  lat  long status      category  wind pressure
##   <chr> <dbl> <dbl> <int> <dbl> <dbl> <dbl> <fct>      <dbl> <int>    <int>
## 1 Wanda 2021   11     6   18  37.1 -38   tropical st~      NA     35     1002
## 2 Wanda 2021   11     7    0  37.4 -37.4 tropical st~      NA     35     1003
## 3 Wanda 2021   11     7    6  38.1 -36.4 tropical st~      NA     35     1004
## 4 Wanda 2021   11     7   12  39.2 -34.9 other low        NA     35     1006
## 5 Wanda 2021   11     7   18  40.9 -32.8 other low        NA     40     1006
## 6 Wanda 2021   11     8    0  43.2 -29.7 other low        NA     40     1006
## # i 2 more variables: tropicalstorm_force_diameter <int>,
## #   hurricane_force_diameter <int>
```

```
# Check for recorded years and months in the data
unique(pull(storms, year))
```

```
## [1] 1975 1976 1977 1978 1979 1980 1981 1982 1983 1984 1985 1986 1987 1988 1989
## [16] 1990 1991 1992 1993 1994 1995 1996 1997 1998 1999 2000 2001 2002 2003 2004
## [31] 2005 2006 2007 2008 2009 2010 2011 2012 2013 2014 2015 2016 2017 2018 2019
## [46] 2020 2021
```

```
unique(pull(storms, month))
```

```
## [1] 6 7 8 9 10 11 5 12 4 1
```

Although there are storms in every year from 1975-2021, some months may have no recorded storms

```
sort(unique(pull(storms, month)))
```

```
## [1] 1 4 5 6 7 8 9 10 11 12
```

```
# Check recorded days in the data
sort(unique(pull(storms, day)))
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [26] 26 27 28 29 30 31
```

This step below might be a little bit confusing, because from `tail()` output we saw 2 columns for category. The 'status' column is for the storm classification (tropical wave - Hurricane), and 'category' column is category of the hurricane based on saffir-simpson scale (1, 2, 3, 4, 5)

```
# Check what are the category of storms available.
unique(pull(storms, status))
```

```
## [1] tropical depression    tropical storm           extratropical
## [4] hurricane                subtropical storm       subtropical depression
## [7] disturbance              other low               tropical wave
## 9 Levels: disturbance extratropical hurricane ... tropical wave
```

```
# Generate a summary of the data, here we can scan through the data for any NAs
summary(storms)
```

```
##      name      year      month      day
## Length:19066  Min.   :1975  Min.   : 1.000  Min.   : 1.00
## Class :character 1st Qu.:1993 1st Qu.: 8.000 1st Qu.: 8.00
## Mode  :character Median :2004 Median : 9.000 Median :16.00
##              Mean  :2002 Mean  : 8.699 Mean  :15.78
##              3rd Qu.:2012 3rd Qu.: 9.000 3rd Qu.:24.00
##              Max.   :2021 Max.   :12.000 Max.   :31.00
##
##      hour      lat      long      status
## Min.   : 0.000  Min.   : 7.00  Min.   : -109.30 tropical storm :6684
## 1st Qu.: 5.000  1st Qu.:18.40 1st Qu.: -78.70 hurricane      :4684
## Median :12.000  Median :26.60 Median : -62.25 tropical depression:3525
## Mean    : 9.094  Mean    :26.99 Mean    : -61.52 extratropical   :2068
## 3rd Qu.:18.000  3rd Qu.:33.70 3rd Qu.: -45.60 other low      :1405
```

```
## Max. :23.000 Max. :70.70 Max. : 13.50 subtropical storm : 292
## (Other) : 408
## category wind pressure tropicalstorm_force_diameter
## Min. :1.000 Min. : 10.00 Min. : 882.0 Min. : 0.0
## 1st Qu.:1.000 1st Qu.: 30.00 1st Qu.: 987.0 1st Qu.: 0.0
## Median :1.000 Median : 45.00 Median :1000.0 Median : 110.0
## Mean :1.898 Mean : 50.02 Mean : 993.6 Mean : 146.3
## 3rd Qu.:3.000 3rd Qu.: 65.00 3rd Qu.:1007.0 3rd Qu.: 220.0
## Max. :5.000 Max. :165.00 Max. :1024.0 Max. :1440.0
## NA's :14382 NA's :9512
## hurricane_force_diameter
## Min. : 0.00
## 1st Qu.: 0.00
## Median : 0.00
## Mean : 14.81
## 3rd Qu.: 0.00
## Max. :300.00
## NA's :9512
```

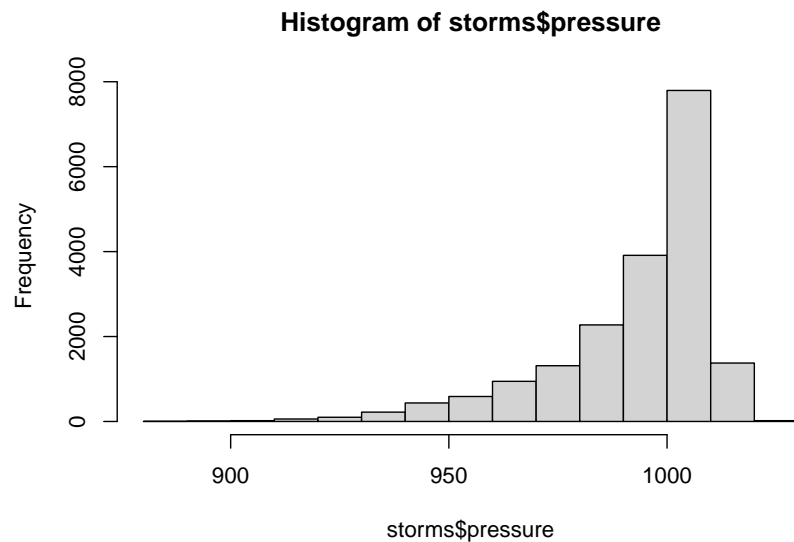
Tropical storm & hurricane force wind diameter data are not available until 2004.

```
# Create a copy of the 'storms' data in a new dataframe to work with,
# so we don't have to download it everytime we want to use the script
# (e.g., data cleaning, data manipulation, etc.)
```

```
storms <- storms
table(storms$status)
```

```
##
## disturbance extratropical hurricane
## 146 2068 4684
## other low subtropical depression subtropical storm
## 1405 151 292
## tropical depression tropical storm tropical wave
## 3525 6684 111
```

```
# Create a histogram of 'pressure' values
hist(storms$pressure)
```



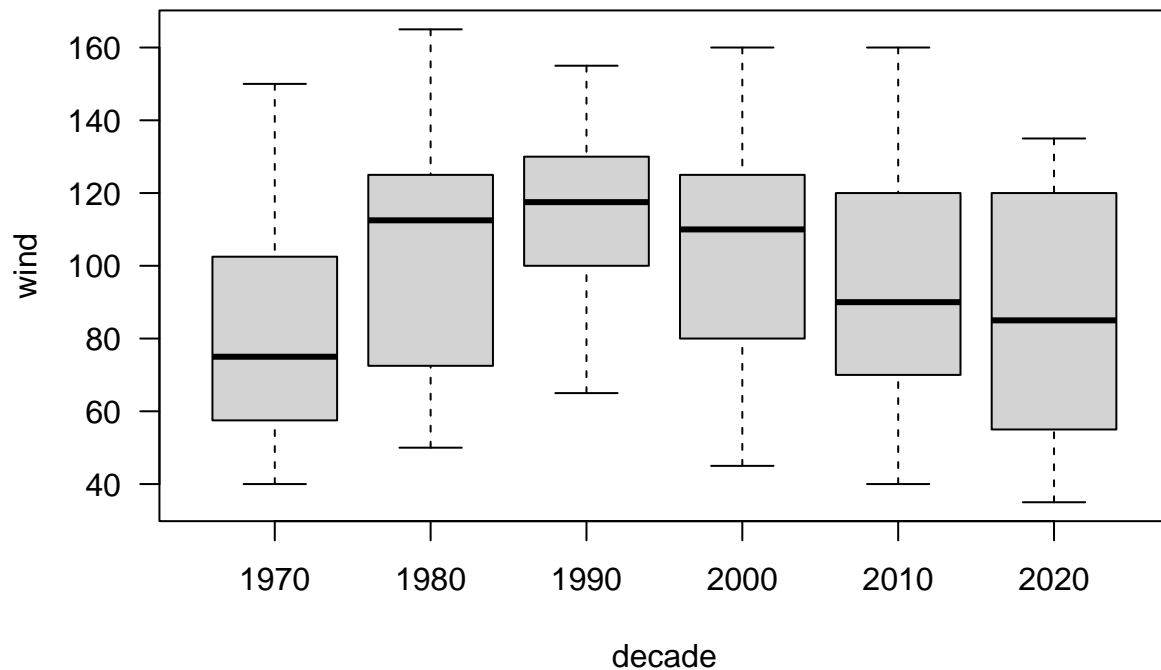
Now, we want to group the data by storm name, filter to get minimum pressure and maximum wind for each storm. This data is originally a point observation data, we can use it for track visualization. However, to analyze the storm events statistics won't be effective if we don't summarize the data.

```
storm_summary <- storms %>%
  group_by(name) %>%
  filter(pressure == min(pressure)) %>%
  filter(status %in% c("tropical storm", "hurricane")) %>%
  filter(wind == max(wind)) %>%
  mutate(date = as.POSIXct(paste(year, month, day, hour, sep = "-"),
                              format = "%Y-%m-%d-%H")) %>%
  filter(date == max(date))
```

We also filter the data only for recorded tropical storms (>34 kts) and hurricanes (>64 kts)

```
# Create 'decade' column to group storms by decade
storm_summary <- storm_summary %>%
  mutate(decade = 10 * (year %/% 10))

# Create a boxplot of wind speeds by decade
boxplot(wind ~ decade, data = storm_summary, las = 1)
```



```
# Create a summary dataframe for number of tropical storms and hurricanes by decade
summary_decade <- storm_summary %>%
  group_by(decade, status) %>%
  summarize(count = n())
```

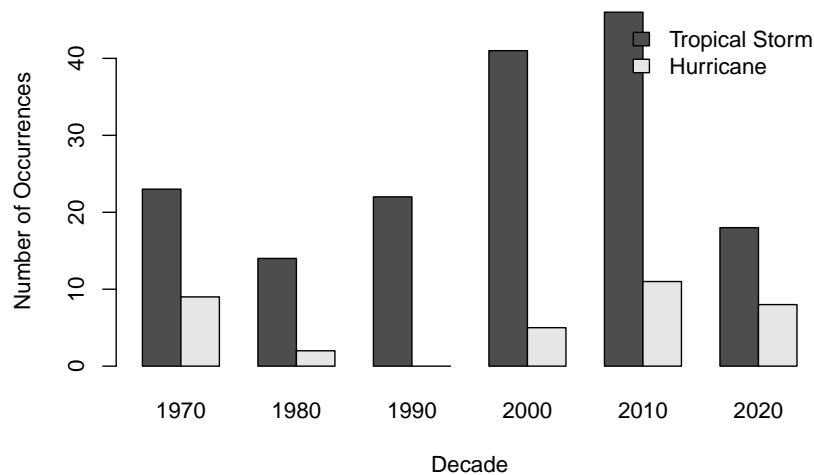
```
## 'summarise()' has grouped output by 'decade'. You can override using the
## '.groups' argument.
```

```
# Pivot the summary data to wide format
summary_wide <- summary_decade %>%
  pivot_wider(names_from = status, values_from = count, values_fill = 0)
```

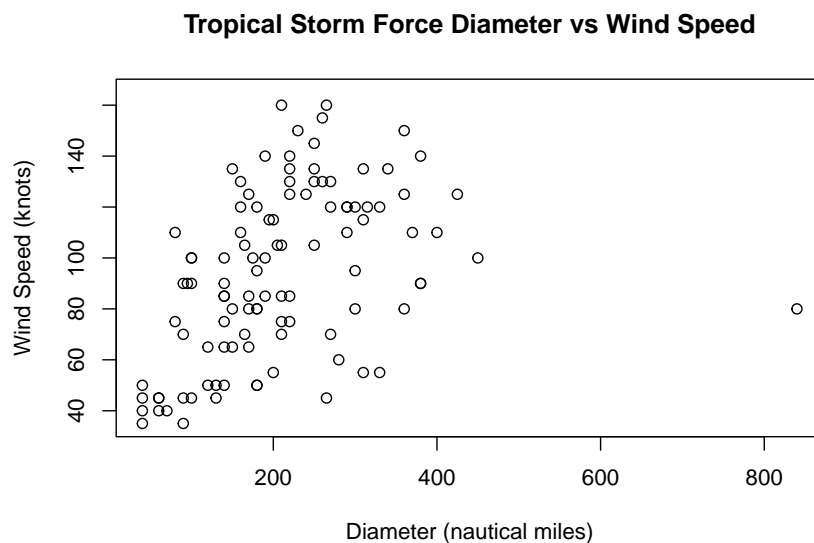
```
# Create a matrix for the barplot
decade_matrix <- t(as.matrix(summary_wide[, -1]))
```

```
# Create a barplot
barplot(decade_matrix, beside = TRUE,
  names.arg = unique(summary_wide$decade),
  xlab = "Decade",
  ylab = "Number of Occurrences",
  main = "Number of Tropical Storm and Hurricane Occurrences by Decade",
  legend.text = c("Tropical Storm", "Hurricane"),
  args.legend = list(x = "topright", bty = "n"))
```

Number of Tropical Storm and Hurricane Occurrences by Decade



```
# Create a scatterplot of tropical storm force diameter vs wind speed
plot(storm_summary$tropicalstorm_force_diameter,
     storm_summary$wind,
     xlab = "Diameter (nautical miles)",
     ylab = "Wind Speed (knots)",
     main = "Tropical Storm Force Diameter vs Wind Speed")
```



```
# Calculate the correlation coefficient between wind speed and tropical
# storm force diameter.
correlation_coefficient <- cor(storm_summary$wind,
                              storm_summary$tropicalstorm_force_diameter,
                              use = "complete.obs")
cat("Correlation Coefficient:", correlation_coefficient, "\n")
```

```
## Correlation Coefficient: 0.4079453
```

```
# Find the storm with the largest tropical storm force diameter
```

```
largest_diameter_name <- storm_summary$name[which.max(storm_summary$tropicalstorm_force_diameter)]  
largest_diameter_name
```

```
## [1] "Sandy"
```

```
# Create a summary of hurricane counts by month
```

```
month_hurricane <- storm_summary %>%  
  filter(category %in% c(1:6)) %>%  
  group_by(month) %>%  
  summarize(hurricane_count = n())
```

```
# Find the peak hurricane season
```

```
peak_season_hurricane <- month_hurricane %>%  
  arrange(desc(hurricane_count)) %>%  
  head(1)
```

```
peak_season_hurricane
```

```
## # A tibble: 1 x 2
```

```
##   month hurricane_count
```

```
##   <dbl>             <int>
```

```
## 1      9              70
```

```
# Calculate correlation coefficients between wind speed > 64 knots
```

```
# and each month
```

```
correlation_coefficients <- sapply(1:12, function(i) {  
  cor(storm_summary$wind[storm_summary$wind > 64],  
      storm_summary$month[storm_summary$wind > 64] == i,  
      method = "spearman")  
})
```

```
# Create a dataframe to store the correlation coefficients and
```

```
# corresponding months
```

```
correlation_df <- data.frame(  
  Month = 1:12,  
  Correlation_Coefficient = correlation_coefficients  
)
```

```
# Print the correlation coefficients
```

```
correlation_df
```

```
##   Month Correlation_Coefficient
```

```
## 1     1                      NA
```

```
## 2     2                      NA
```

```
## 3     3                      NA
```

```
## 4     4                      NA
```

```
## 5     5                      NA
```

```
## 6     6                      NA
```

```
## 7     7          -0.136447985
```

```
## 8      8      0.071617789
## 9      9      0.161365343
## 10     10     -0.173892113
## 11     11      0.003966381
## 12     12      NA
```

Spatial distributions of the storms before and after year 2000

```
# Subset the 'storms' data for years before 2000 and after 2000
storms_before_2000 <- storms %>%
  filter(year < 2000) %>%
  filter(status %in% c("tropical storm", "hurricane"))

storms_after_2000 <- storms %>%
  filter(year >= 2000) %>%
  filter(status %in% c("tropical storm", "hurricane"))

# Define latitudes and longitudes for the map
lat1 <- 5; lat2 <- 60; lon1 <- -110; lon2 <- 6

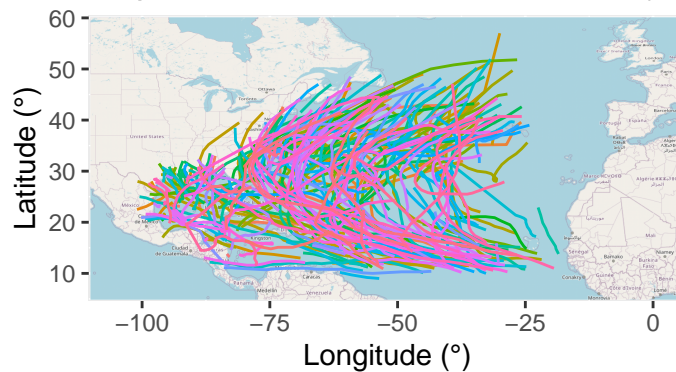
# Create a base map using OpenStreetMap
base_map <- openmap(c(lat2, lon1), c(lat1, lon2), zoom = 4, type = "osm",
  mergeTiles = TRUE)
base_map2 <- openproj(base_map)

# Create a plot for hurricane tracks on top of the base map
plot_before_2000 <- OpenStreetMap::autoplot.OpenStreetMap(base_map2) +
  geom_path(data = storms_before_2000, aes(x = long, y = lat, color = paste(year, name)),
    show.legend = FALSE) +
  xlab("Longitude (°)") + ylab("Latitude (°)") +
  labs(title = "Tropical Storms & Hurricane Tracks (1975-1999)")

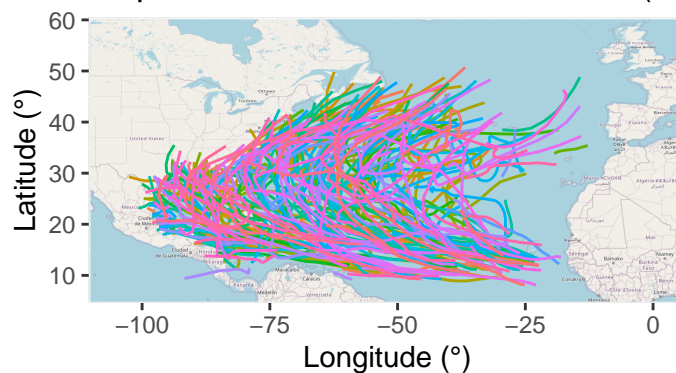
plot_after_2000 <- OpenStreetMap::autoplot.OpenStreetMap(base_map2) +
  geom_path(data = storms_after_2000, aes(x = long, y = lat, color = paste(year, name)),
    show.legend = FALSE) + xlab("Longitude (°)") + ylab("Latitude (°)") +
  labs(title = "Tropical Storms & Hurricane Tracks (2000-2021)")

# Arrange the plots side by side
grid.arrange(plot_before_2000, plot_after_2000, nrow = 2)
```


Tropical Storms & Hurricane Tracks (1975–1999)



Tropical Storms & Hurricane Tracks (2000–2021)



Other statistical analysis (likelihood of trop. storm events per extended period)

```
# Calculate Poisson distribution
# Filter the data frame to include only category 4 or 5 hurricanes
storm_summary_cat4 <- storm_summary %>%
  filter(category == '4')

storm_summary_cat5 <- storm_summary %>%
  filter(category == '5')

# Calculate the total number of category 4 and 5 hurricanes over the decade
total_category4 <- nrow(storm_summary_cat4)
total_category5 <- nrow(storm_summary_cat5)

# Calculate the average number of category 4 and 5 hurricanes per year
average_category4_per_year <- total_category4 / 46
average_category5_per_year <- total_category5 / 46

# Create a sequence of numbers for plotting the Poisson distribution
x = 1:50

# Calculate the Poisson probabilities for category 4 and 5 hurricanes
poisson_cat4 <- dpois(x, lambda = average_category4_per_year)
poisson_cat5 <- dpois(x, lambda = average_category5_per_year)
```

```
# Create a two-panel plot for Poisson distribution
par(mfrow = c(1, 2))
plot(x, poisson_cat4, type = "l", lwd = 3, col = "navy")
plot(x, poisson_cat5, type = "l", lwd = 3, col = "navy")
```

