

Group Project R Script and Outputs

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```
#loading packages  
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr      1.1.3      v readr      2.1.4  
## v forcats    1.0.0      v stringr    1.5.0  
## v ggplot2    3.5.0      v tibble     3.2.1  
## v lubridate  1.9.3      v tidyr      1.3.0  
## v purrr      1.0.2  
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(ggplot2)  
library(GGally)
```

```
## Registered S3 method overwritten by 'GGally':  
##   method from  
##   +.gg      ggplot2
```

```
library(rmarkdown)
```

```
#setting WD  
setwd("/home/guest/Stat_Modeling_Lab/ENV710Ayoung_Emma")
```

```
#read in original data  
original_data <- read.csv("dolphins.csv")
```

```
#read in data with unneeded variables  
strandings <- read.csv("dolphins_cleaned.csv")  
#attach data  
attach(strandings)
```

```
#remove unneeded covariates in cleaned data  
cleaned_strandings <- subset(strandings, select =  
                             -c(Shot, Fishery.Interaction, Boat.Collision, Weight))
```

```
#remove NAs from chosen variables  
cleaned_strandings <- na.omit(cleaned_strandings)
```

```

#removing blanks in Age.Class variable
cleaned_strandings <-
  cleaned_strandings[!grepl("^\\s*$", cleaned_strandings$Age.Class), ]
#removing unknowns in Age.Class
cleaned_strandings <-
  cleaned_strandings[!cleaned_strandings$Age.Class %in% c("UNKNOWN"), ]

#removing unknowns in Sex variable
cleaned_strandings <-
  cleaned_strandings[!cleaned_strandings$Sex %in% c("UNKNOWN"), ]

#create binary covariate for states with and without offshore wind
cleaned_strandings <- cleaned_strandings %>%
  mutate(turbine_presence = if_else(State %in% c("VA", "NY", "RI", "MA"), 1, 0))

#create subset of only states with offshore wind
turbine_data <-
  cleaned_strandings[cleaned_strandings$State %in%
    c("VA", "NY", "RI", "MA"), ]

#use mutate to create state binary variables for fit_4
turbine_data <- turbine_data %>%
  mutate(
    VA = ifelse(State == "VA", 1, 0),
    NY = ifelse(State == "NY", 1, 0),
    RI = ifelse(State == "RI", 1, 0),
    MA = ifelse(State == "MA", 1, 0))

#count of strandings in wind farm states = 80
sum(cleaned_strandings$turbine_presence)

```

```
## [1] 80
```

```

#count of total number of stranding = 1419
nrow(cleaned_strandings)

```

```
## [1] 1419
```

```

#summary statistics of all variables
summary_stats <- summary(cleaned_strandings)
summary_stats

```

```

##      State      Year.of.Observation      Sex      Age.Class
## Length:1419   Min.   :2017      Length:1419   Length:1419
## Class :character 1st Qu.:2017      Class :character  Class :character
## Mode  :character Median :2018      Mode  :character  Mode  :character
##                Mean   :2018
##                3rd Qu.:2019
##                Max.   :2019
##      Length      turbine_presence
## Min.   : 0.0   Min.   :0.00000

```

```
## 1st Qu.:134.0 1st Qu.:0.00000
## Median :209.0 Median :0.00000
## Mean :194.2 Mean :0.05638
## 3rd Qu.:248.0 3rd Qu.:0.00000
## Max. :366.0 Max. :1.00000
```

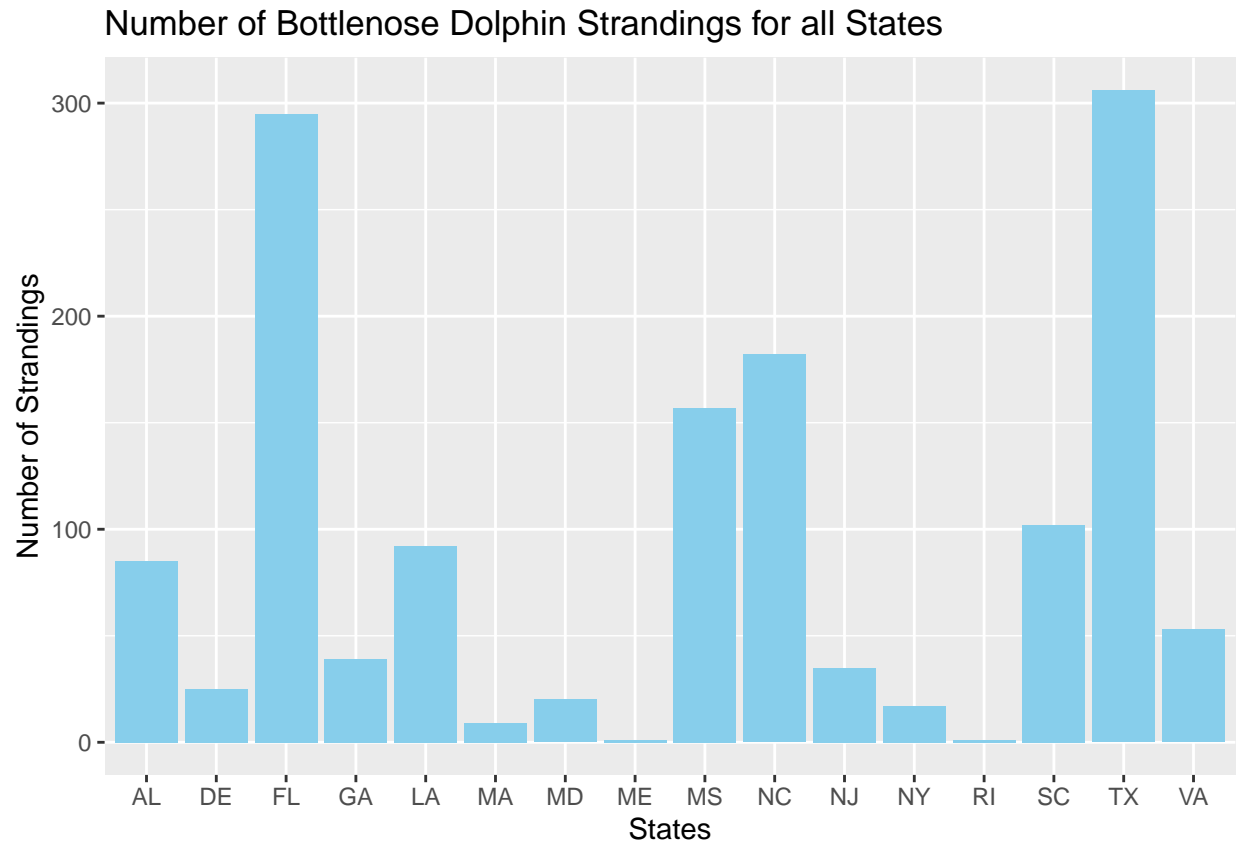
```
#reordering age classes so they are from youngest to oldest
age_class_order <- c("PUP/CALF", "YEARLING", "SUBADULT", "ADULT")
```

```
#States info
table(cleaned_strandings$State)
```

```
##
## AL DE FL GA LA MA MD ME MS NC NJ NY RI SC TX VA
## 85 25 295 39 92 9 20 1 157 182 35 17 1 102 306 53
```

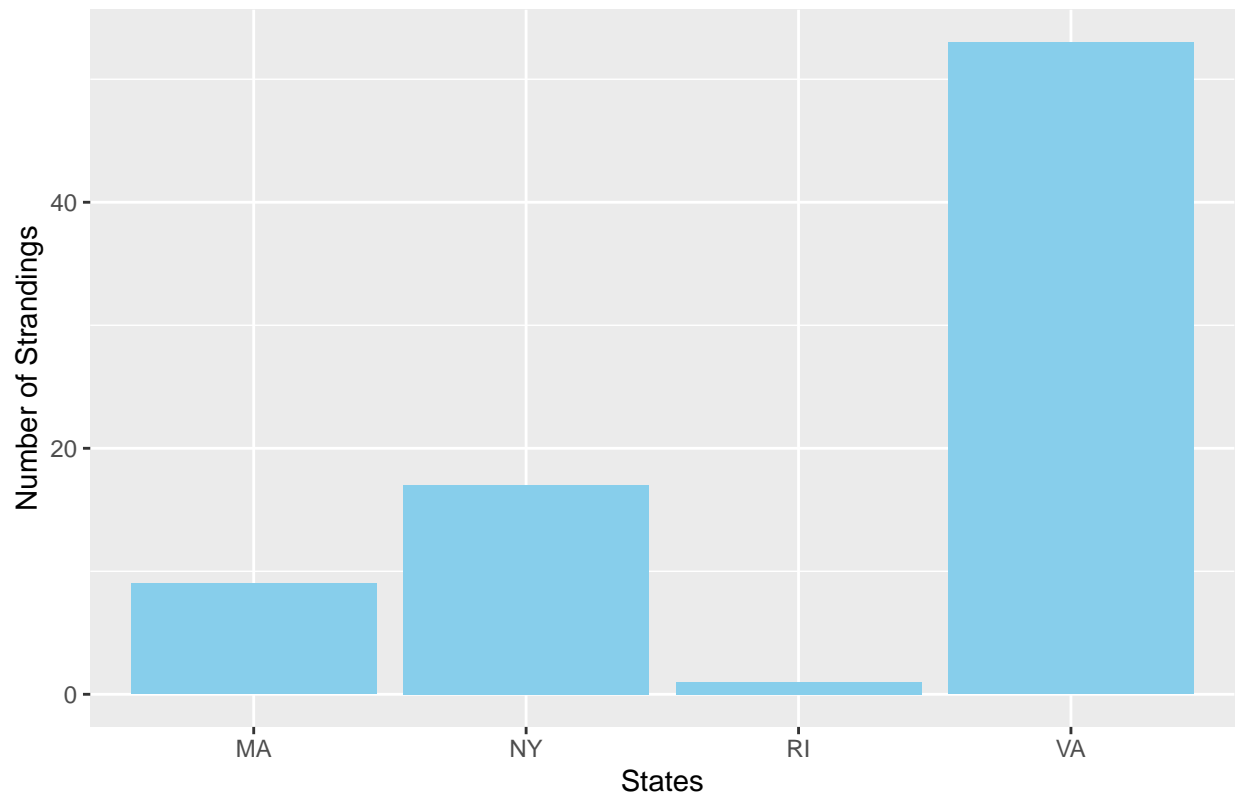
```
# 15 states included
# Most strandings were in Florida (295) and Texas (306)
# Least were in Maine (1) and Rhode Island (1)
# States with offshore wind: Virginia, New York, Rhode Island, and Massachusetts
```

```
#plot of number of strandings for all states
standings_state <- ggplot(cleaned_strandings, aes(x = State)) +
  geom_bar(fill = "skyblue") +
  labs(x = "States", y = "Number of Strandings") +
  ggtitle("Number of Bottlenose Dolphin Strandings for all States")
standings_state
```



```
#plot of number of strandings for each offshore wind state
standings_wf_state <- ggplot(turbine_data, aes(x = State)) +
  geom_bar(fill = "skyblue") +
  labs(x = "States", y = "Number of Strandings") +
  ggtitle("Number of Bottlenose Dolphin Strandings by Offshore Wind States")
standings_wf_state
```

Number of Bottlenose Dolphin Strandings by Offshore Wind States



```
#Length info  
# average overall length 194.2091  
mean(cleaned_strandings$Length)
```

```
## [1] 194.2091
```

```
# average offshore wind length = 198.175  
mean(turbine_data$Length)
```

```
## [1] 198.175
```

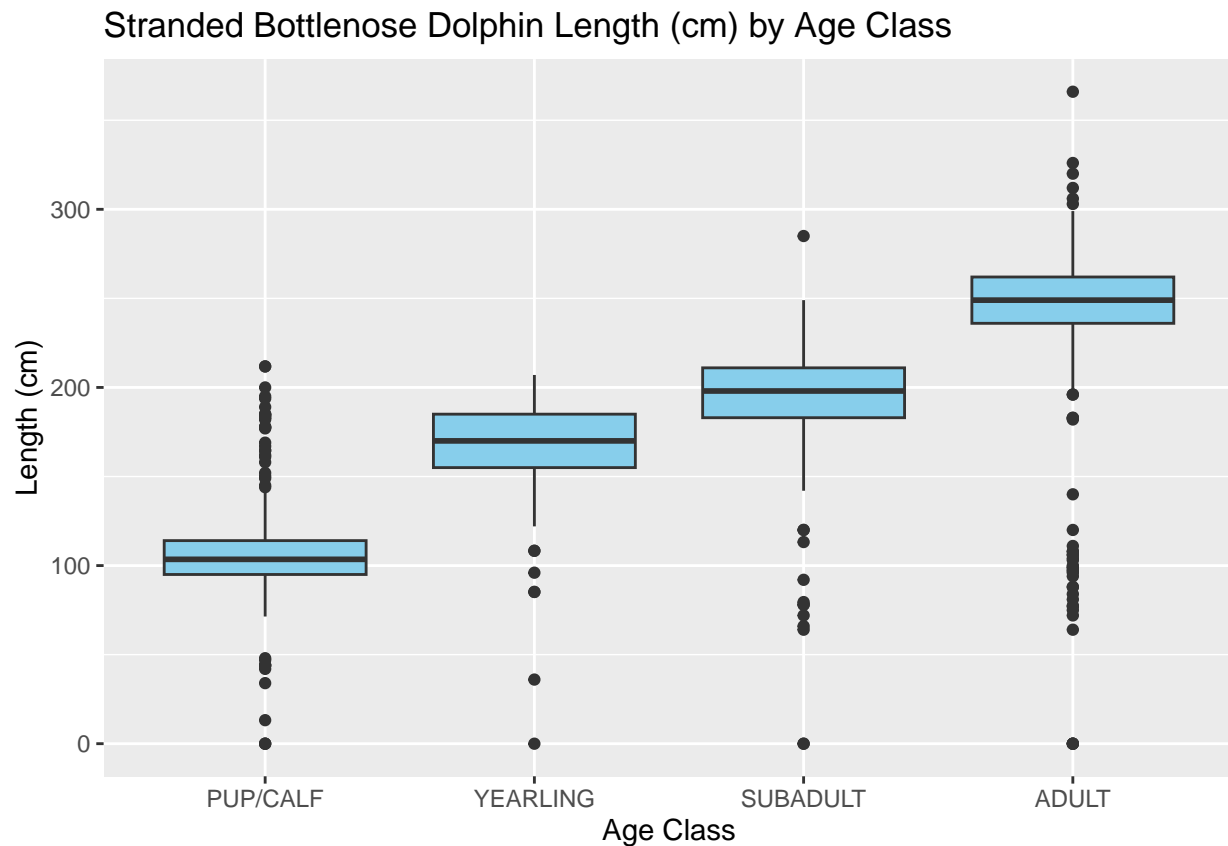
```
# variance of all lengths = 4270.962 so 65.37 cm deviation from the mean  
variance_all <- var(cleaned_strandings$Length)  
variance_all
```

```
## [1] 4270.962
```

```
# variance of offshore wind lengths = 6133.309 so 78.36 cm deviation from the mean  
variance_ow <- var(turbine_data$Length)  
variance_ow
```

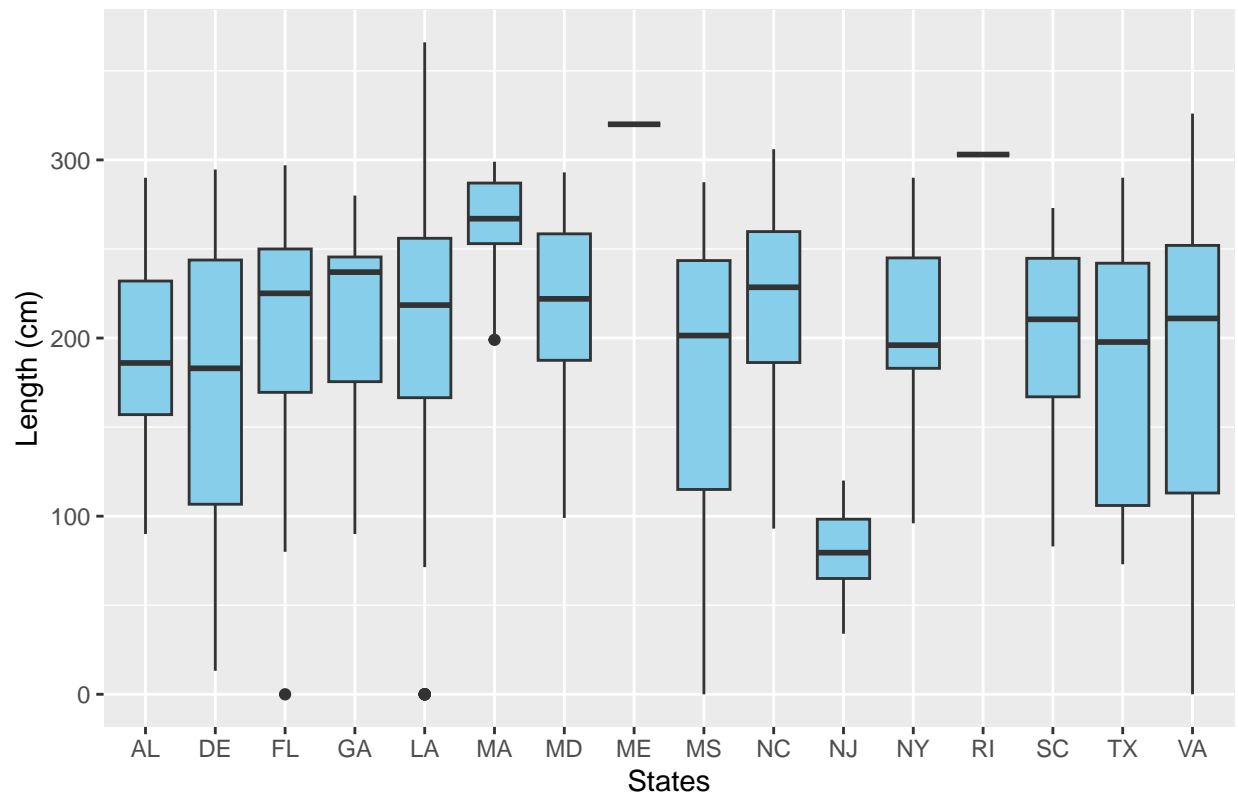
```
## [1] 6133.309
```

```
#summary graph of Age Class by Length
length_ageclass <- ggplot(cleaned_strandings,
                          aes(x = factor(Age.Class, levels = age_class_order),
                              y = Length)) +
  geom_boxplot(fill="skyblue") +
  labs(x = "Age Class", y = "Length (cm)") +
  ggtitle("Stranded Bottlenose Dolphin Length (cm) by Age Class")
length_ageclass
```



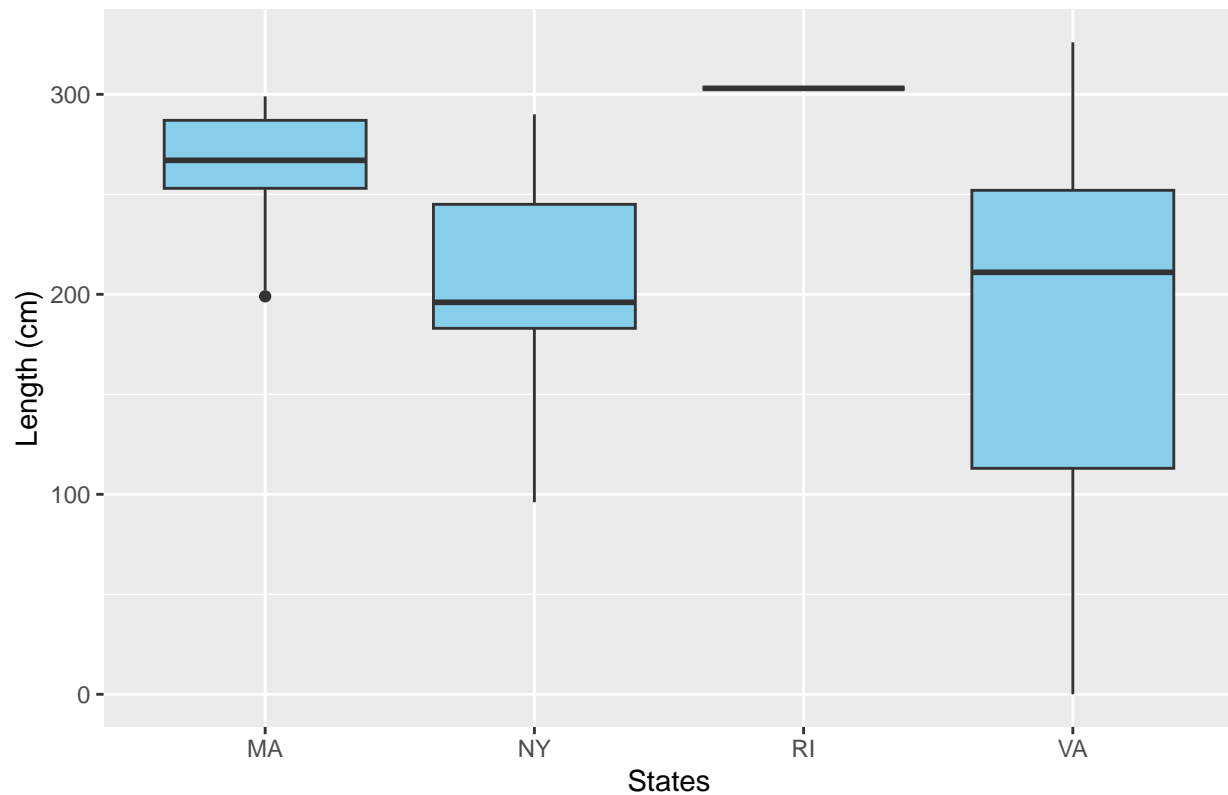
```
#plot of Lengths in each state
state_length <- ggplot(cleaned_strandings, aes(x = State, y = Length)) +
  geom_boxplot(fill="skyblue") +
  labs(x = "States", y = "Length (cm)") +
  ggtitle("Stranded Bottlenose Dolphin Length (cm) by State")
state_length
```

Stranded Bottlenose Dolphin Length (cm) by State



```
#plot of Lengths in each offshore wind state
turbinestate_length <- ggplot(turbine_data, aes(x = State, y = Length)) +
  geom_boxplot(fill="skyblue") +
  labs(x = "States", y = "Length (cm)") +
  ggtitle("Stranded Bottlenose Dolphin Length (cm) by States with Offshore Wind")
turbinestate_length
```

Stranded Bottlenose Dolphin Length (cm) by States with Offshore Wind



#Age Class Info

#finding ave length per age class: PUP/CALF 107.1709, YEARLING 163.3439, SUBADULT 194.0060, ADULT 242.8233

`ave_length_ageclass <-`

`aggregate(Length ~ Age.Class, data = cleaned_strandings, FUN = mean)`

`ave_length_ageclass`

```
##   Age.Class  Length
```

```
## 1    ADULT 242.8233
```

```
## 2  PUP/CALF 107.1709
```

```
## 3  SUBADULT 194.0060
```

```
## 4  YEARLING 163.3439
```

#count of strandings in all states by age class ADULT 669, PUP/CALF 335, SUBADULT 308, YEARLING 107

`table(cleaned_strandings$Age.Class)`

```
##
```

```
##   ADULT PUP/CALF SUBADULT YEARLING
```

```
##    669    335    308    107
```

#count of strandings in offshore wind states by age class ADULT 32, PUP/CALF 21, SUBADULT 26, YEARLING 10

`count_ageclass <- aggregate(turbine_presence ~ Age.Class, data = cleaned_strandings, FUN = sum)`

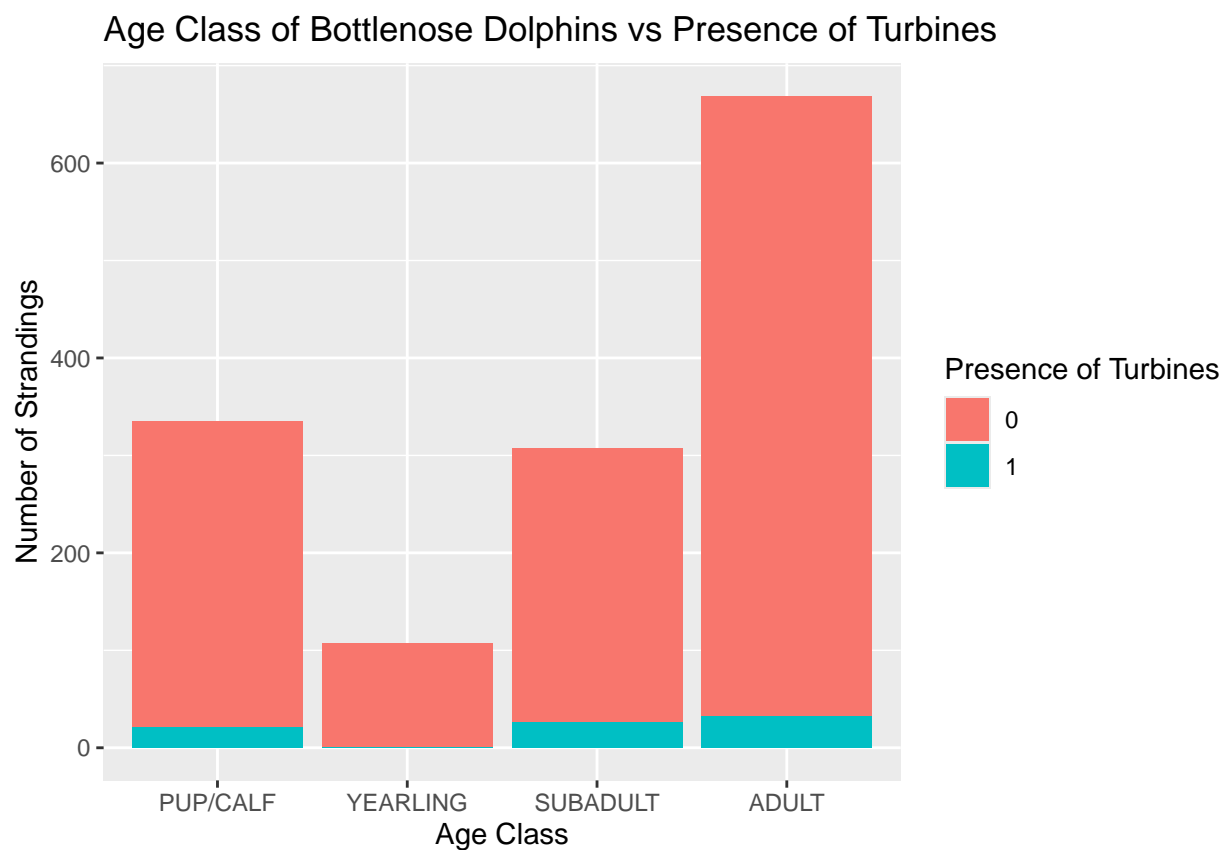
`count_ageclass`

```
##   Age.Class turbine_presence
```



```
## 1      ADULT      32
## 2  PUP/CALF      21
## 3  SUBADULT      26
## 4  YEARLING       1
```

```
#plot of the number of strandings in each age class
strandings_ageclass <- ggplot(cleaned_strandings,
                             aes(x = factor(Age.Class, levels = age_class_order),
                                fill = factor(turbine_presence))) +
  geom_bar(position = "stack") +
  labs(x = "Age Class", y = "Number of Strandings", fill = "Presence of Turbines") +
  ggtitle("Age Class of Bottlenose Dolphins vs Presence of Turbines")
strandings_ageclass
```



```
#Sex Info
#finding ave length per sex: FEMALE 191.8816, MALE 195.8511
ave_length_sex <-
  aggregate.Length ~ Sex, data = cleaned_strandings, FUN = mean)
ave_length_sex
```

```
##      Sex  Length
## 1 FEMALE 191.8816
## 2  MALE 195.8511
```

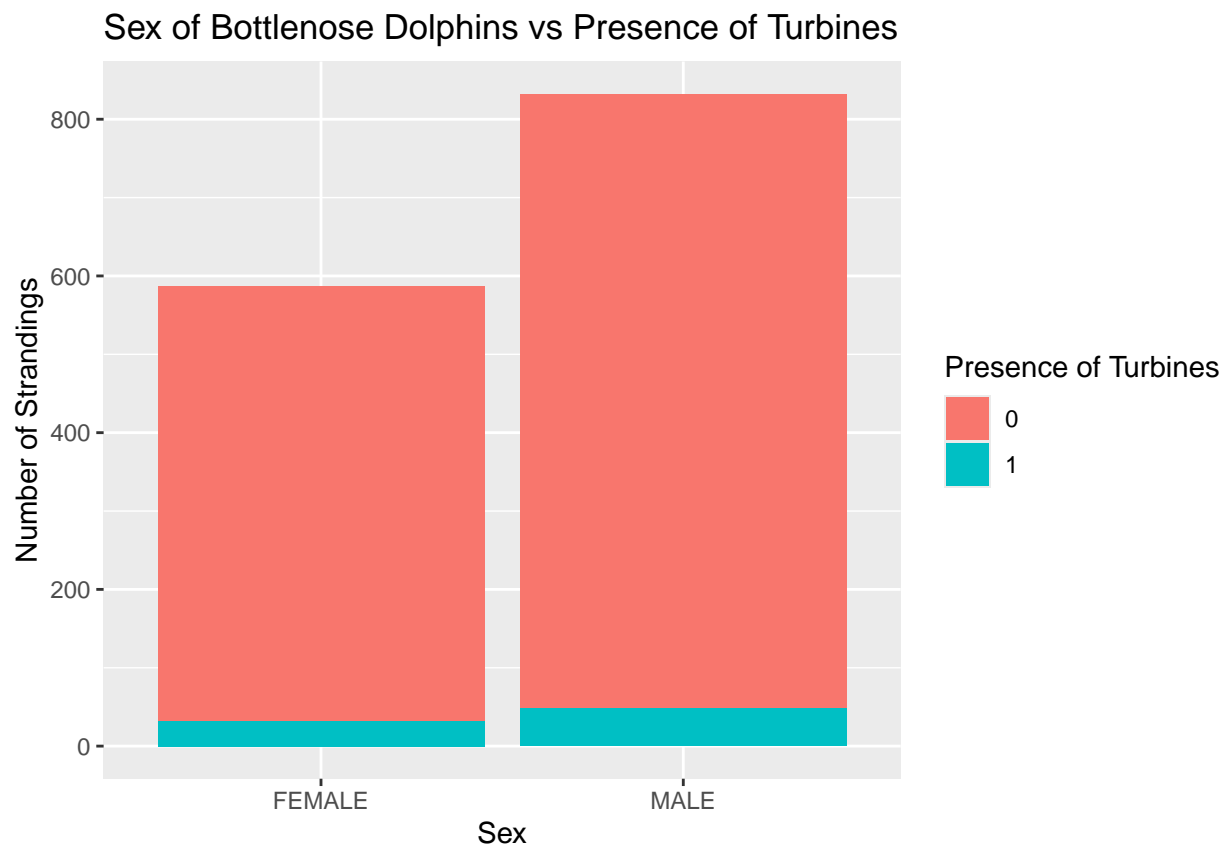
```
#count of strandings in all states by sex FEMALE 587, MALE 832
table(cleaned_strandings$Sex)
```

```
##
## FEMALE    MALE
##    587    832
```

```
#count of strandings in offshore states by sex FEMALE 32, MALE 48
count_sex <-
  aggregate(turbine_presence ~ Sex, data = cleaned_strandings, FUN = sum)
count_sex
```

```
##      Sex turbine_presence
## 1 FEMALE                32
## 2  MALE                 48
```

```
#plot of the number of strandings in each sex
strandings_sex <- ggplot(cleaned_strandings,
  aes(x = factor(Sex), fill = factor(turbine_presence))) +
  geom_bar(position = "stack") +
  labs(x = "Sex", y = "Number of Strandings", fill = "Presence of Turbines") +
  ggtitle("Sex of Bottlenose Dolphins vs Presence of Turbines")
strandings_sex
```



```
#Fitting a regression model (Turbine Presence)
fit_1 <- glm(turbine_presence~1, data = cleaned_strandings)
#Summary of the regression model fit_1
summary(fit_1)
```

```
##
## Call:
## glm(formula = turbine_presence ~ 1, data = cleaned_strandings)
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.056378   0.006125   9.204  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.0532368)
##
## Null deviance: 75.49  on 1418  degrees of freedom
## Residual deviance: 75.49  on 1418  degrees of freedom
## AIC: -131.99
##
## Number of Fisher Scoring iterations: 2
```

```
#Fitting a regression model
fit_2 <- glm(turbine_presence~Age.Class,family='binomial', data = cleaned_strandings)
#Summary of the regression model fit_2
summary(fit_2)
```

```
##
## Call:
## glm(formula = turbine_presence ~ Age.Class, family = "binomial",
##      data = cleaned_strandings)
##
## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -2.9910     0.1812 -16.510  <2e-16 ***
## Age.ClassPUP/CALF  0.2862     0.2892   0.990   0.3224
## Age.ClassSUBADULT  0.6072     0.2735   2.220   0.0264 *
## Age.ClassYEARLING -1.6724     1.0209  -1.638   0.1014
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 615.51  on 1418  degrees of freedom
## Residual deviance: 603.61  on 1415  degrees of freedom
## AIC: 611.61
##
## Number of Fisher Scoring iterations: 7
```

```
#Calculations of odds (probability) of each age class
exp(-2.99)
```

```
## [1] 0.05028744
```

```
#calculations of odd of each age class  
exp(-2.9910+0.2862)
```

```
## [1] 0.0668837
```

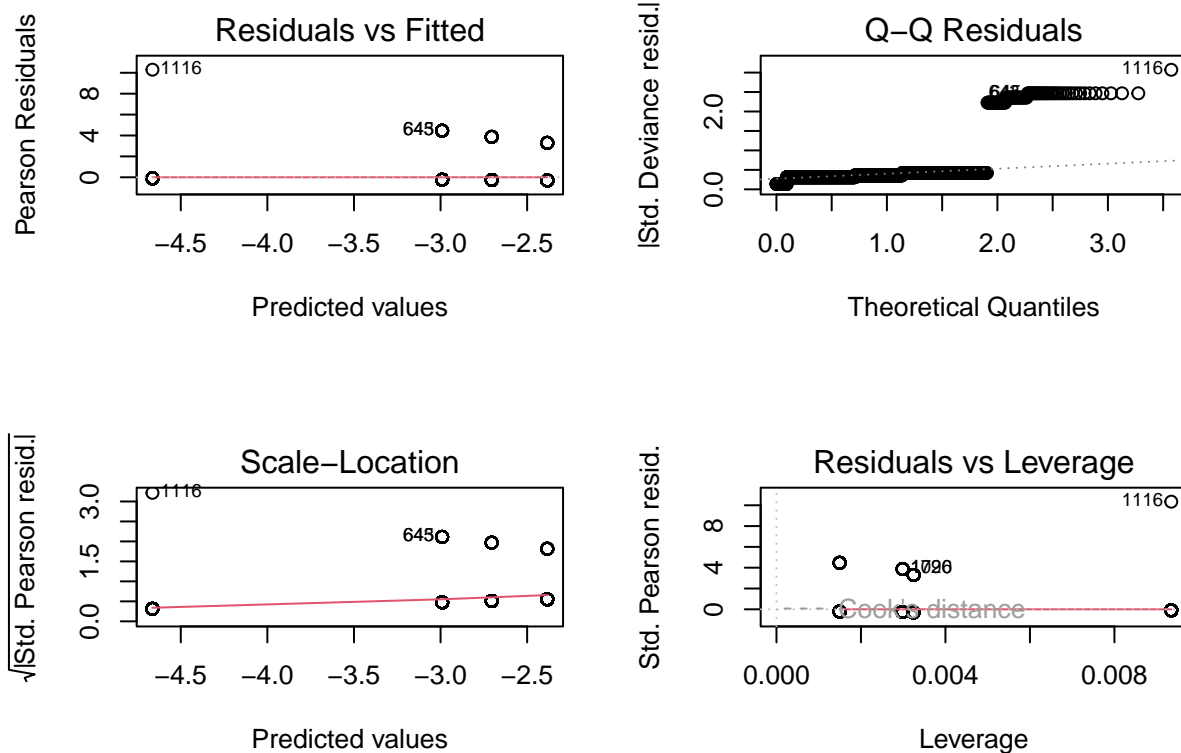
```
exp(-2.9910+0.6072)
```

```
## [1] 0.09219955
```

```
exp(-2.9910-1.6724)
```

```
## [1] 0.009434331
```

```
#qq plots for fit_2  
par(mfrow=c(2,2))  
plot(fit_2)
```

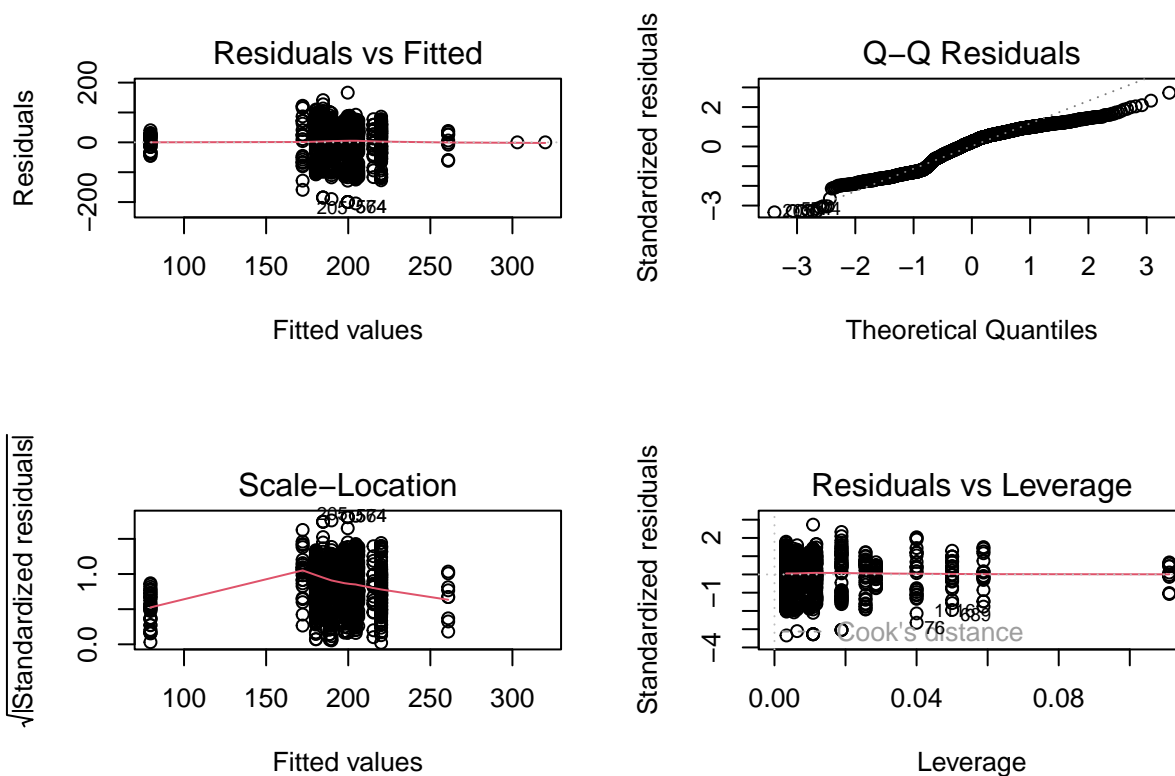


```
# use to find difference in length in states  
fit_3 <- lm(Length ~ State, data = cleaned_strandings)  
#summary of linear regression  
summary(fit_3)
```

```
##
## Call:
## lm(formula = Length ~ State, data = cleaned_strandings)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -204.59  -45.68   13.90   48.64  166.30
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  188.747      6.650   28.381 < 2e-16 ***
## StateDE      -16.521     13.950   -1.184 0.236508
## StateFL       15.847      7.548    2.099 0.035952 *
## StateGA       17.356     11.858    1.464 0.143541
## StateLA       10.951      9.225    1.187 0.235375
## StateMA       72.142     21.493    3.357 0.000810 ***
## StateMD       26.638     15.238    1.748 0.080663 .
## StateME      131.253     61.674    2.128 0.033496 *
## StateMS        1.064      8.257    0.129 0.897506
## StateNC       31.281      8.055    3.883 0.000108 ***
## StateNJ      -109.191     12.314  -8.867 < 2e-16 ***
## StateNY       12.159     16.290    0.746 0.455558
## StateRI      114.253     61.674    1.853 0.064159 .
## StateSC        9.081      9.005    1.009 0.313384
## StateTX       -8.427      7.518   -1.121 0.262504
## StateVA       -4.075     10.731   -0.380 0.704178
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 61.31 on 1403 degrees of freedom
## Multiple R-squared:  0.1291, Adjusted R-squared:  0.1198
## F-statistic: 13.86 on 15 and 1403 DF, p-value: < 2.2e-16

#set up a multi-panel plot layout
par(mfrow = c(2, 2))
plot(fit_3)

## Warning: not plotting observations with leverage one:
## 369, 670
```

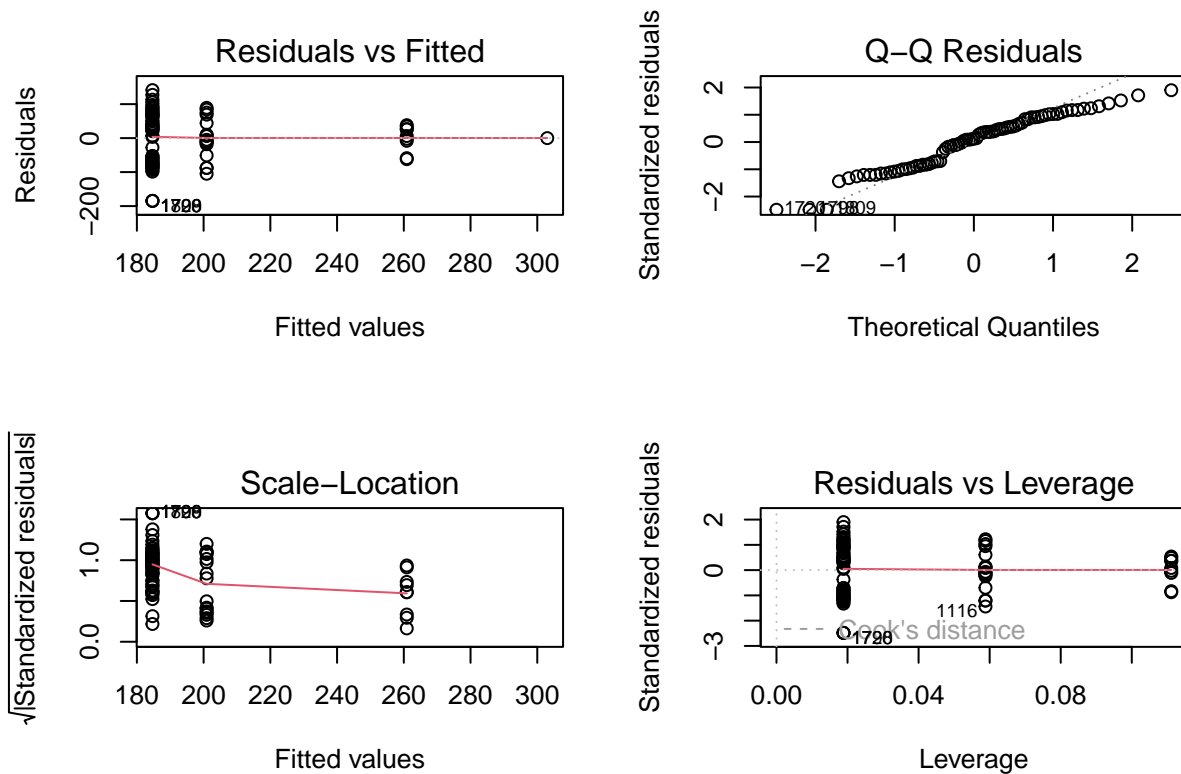


```
#finding significance of dolphin length across offshore wind states
fit_4 <- lm(Length ~ VA + NY + RI + MA, data = turbine_data)
#summary of linear regression
summary(fit_4)
```

```
##
## Call:
## lm(formula = Length ~ VA + NY + RI + MA, data = turbine_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -184.672  -61.876    8.211   62.328  141.328
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   260.89     25.02   10.425 2.62e-16 ***
## VA            -76.22     27.07   -2.816  0.00619 **
## NY            -59.98     30.95   -1.938  0.05632 .
## RI             42.11     79.14    0.532  0.59618
## MA              NA          NA      NA      NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 75.07 on 76 degrees of freedom
## Multiple R-squared:  0.1159, Adjusted R-squared:  0.08104
## F-statistic: 3.322 on 3 and 76 DF, p-value: 0.02413
```

```
#set up a multi-panel plot layout
par(mfrow = c(2, 2))
plot(fit_4)
```

```
## Warning: not plotting observations with leverage one:
## 27
```



```
#finding the mean of length within VA
mean_length_VA <- mean(turbine_data$Length[turbine_data$VA == "1"])
mean_length_VA
```

```
## [1] 184.6717
```

```
#finding the mean of length within VA
mean_length_NY <- mean(turbine_data$Length[turbine_data$NY == "1"])
mean_length_NY
```

```
## [1] 200.9059
```

```
#finding the mean of length within VA
mean_length_RI <- mean(turbine_data$Length[turbine_data$RI == "1"])
mean_length_RI
```

```
## [1] 303
```

```
#finding the mean of length within VA
mean_length_MA <- mean(turbine_data$Length[turbine_data$MA == "1"])
mean_length_MA
```

```
## [1] 260.8889
```

```
# coloring offshore wind states red
state_colors <- c("NY" = "red", "MA" = "red", "VA" = "red", "RI" = "red",
                 "Other" = "skyblue")

# coloring all other states blue
cleaned_strandings$State_color <- ifelse(cleaned_strandings$State %in% c("NY", "MA", "VA", "RI"), cleaned_strandings$State_color, "skyblue")

# plot of number of strandings for all states
standings_state <- ggplot(cleaned_strandings, aes(x = State, fill = State_color)) +
  geom_bar() +
  scale_fill_manual(values = state_colors) +
  labs(x = "States", y = "Number of Strandings") +
  ggtitle("Number of Bottlenose Dolphin Strandings by States") +
  theme_bw() +
  theme(legend.position = "none")

standings_state
```

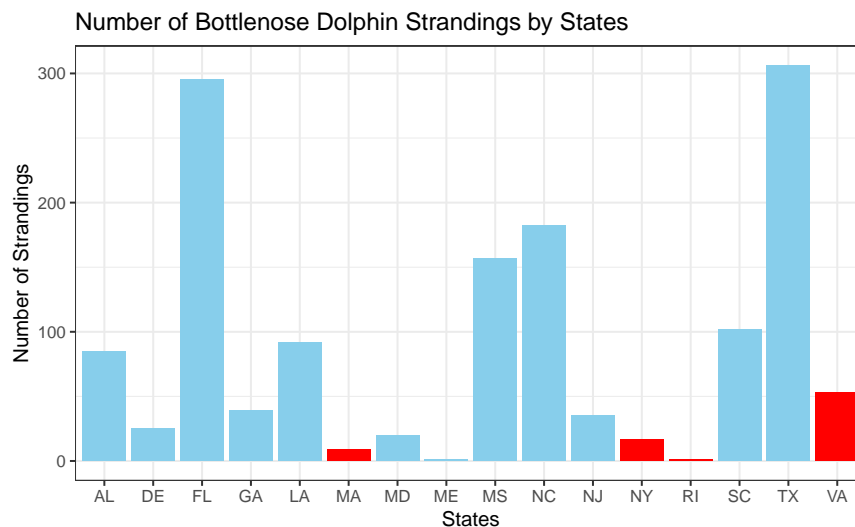


Figure 1: The number of stranded dolphins reported in each state across the East Coast. States with active offshore wind projects are labeled in red.

```
# plot of stranded bottlenose dolphin length by state with specific states colored differently
state_length <- ggplot(cleaned_strandings, aes(x = State, y = Length, fill = State_color)) +
  geom_boxplot() +
  scale_fill_manual(values = state_colors) +
  labs(x = "States", y = "Length (cm)") +
```



```

ggtitle("Stranded Bottlenose Dolphin Length (cm) by State")+
theme_bw() +
theme(legend.position = "none")

# Display the plot
state_length

```

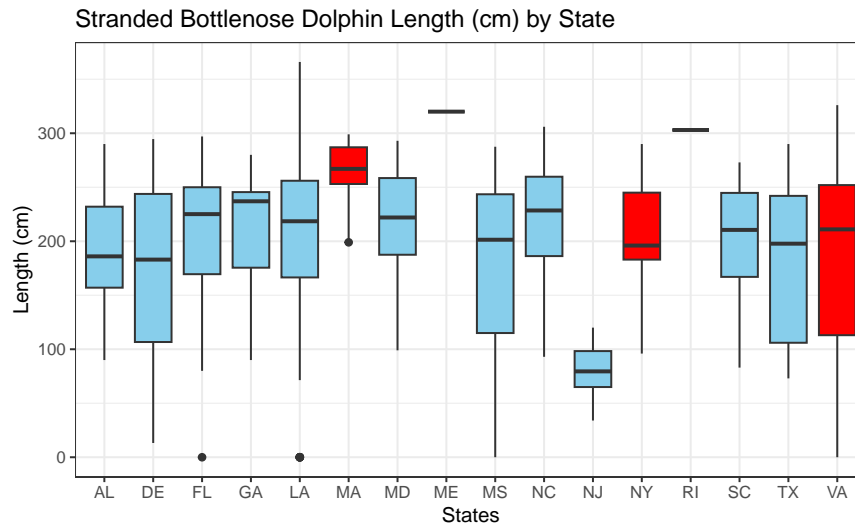


Figure 2: The distribution of length (cm) of stranded dolphins reported in each state across the East Coast. States with active offshore wind projects are labeled in red.

```

# color bars
fill_colors <- c("0" = "skyblue", "1" = "red")

# change legend labels
fill_labels <- c("0" = "No", "1" = "Yes")

# plot of the number of strandings in each age class
strandings_ageclass <- ggplot(cleaned_strandings,
                             aes(x = factor(Age.Class, levels = age_class_order),
                                fill = factor(turbine_presence))) +

  geom_bar(position = "stack") +
  scale_fill_manual(values = fill_colors, labels = fill_labels) +
  labs(x = "Age Class", y = "Number of Strandings",
       fill = "Presence of Offshore Wind") +
  ggtitle("Age Class of Bottlenose Dolphins vs Presence of Offshore Wind") +
  theme_bw()

strandings_ageclass

```

```

# change bar colors
fill_colors <- c("0" = "skyblue", "1" = "red")

# change legend labels
fill_labels <- c("0" = "No", "1" = "Yes")

```

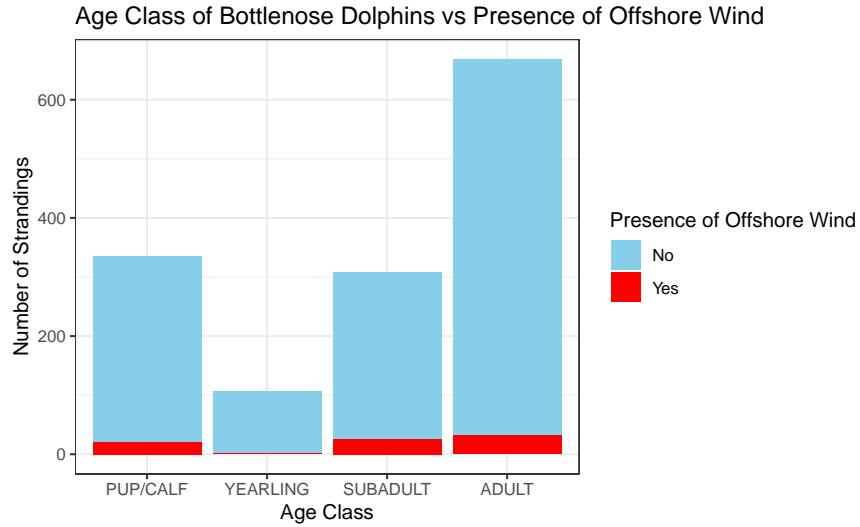


Figure 3: The number of strandings for each age class of bottlenose dolphin and whether or not there is a presence of offshore wind.

```
# plot of the number of strandings in each sex with specified colors and legend labels
strandings_sex <- ggplot(cleaned_strandings,
  aes(x = factor(Sex), fill = factor(turbine_presence))) +
  geom_bar(position = "stack") +
  scale_fill_manual(values = fill_colors, labels = fill_labels) + # Specify colors and labels
  labs(x = "Sex", y = "Number of Strandings", fill = "Presence of Offshore Wind") +
  ggtitle("Sex of Bottlenose Dolphins vs Presence of Offshore Wind") +
  theme_bw()
strandings_sex
```

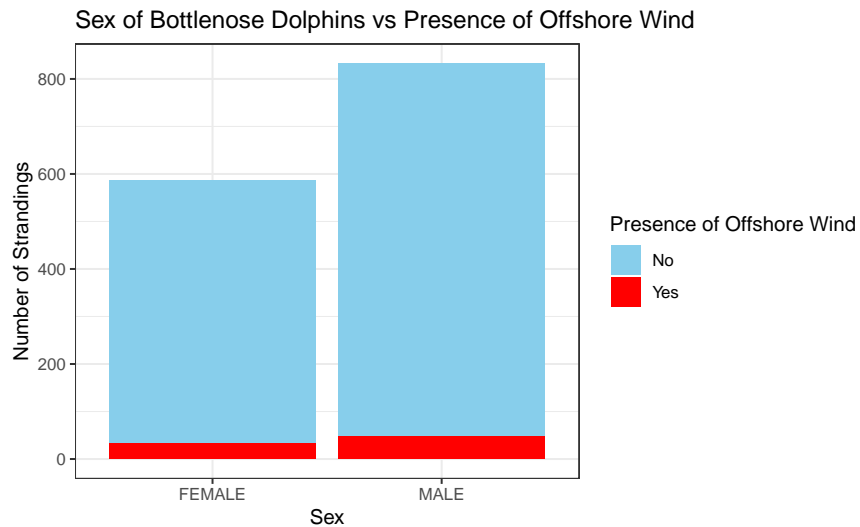


Figure 4: The number of strandings for each sex and whether or not there is a presence of offshore wind.

```

fit_3_plot <- ggplot(cleaned_strandings, aes(x = State, y = Length)) +

  geom_point(size = 1) +

  geom_smooth(method = 1) +

  geom_hline(aes(yintercept = 188.747, color = "AL"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 - 16.521, color = "DE"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 15.847, color = "FL"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 17.356, color = "GA"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 10.951, color = "LA"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 72.142, color = "MA"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 26.638, color = "MD"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 131.253, color = "ME"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 1.064, color = "MS"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 31.281, color = "NC"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 - 109.191, color = "NJ"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 12.159, color = "NY"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 114.253, color = "RI"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 + 9.081, color = "SC"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 - 8.427, color = "TX"), linetype = "solid") +

  geom_hline(aes(yintercept = 188.747 - 4.075, color = "VA"), linetype = "solid") +

  xlab("State") +

  ylab("Length of Bottlenose Strandings") +

  scale_color_manual(name = "State",

    values = c(AL = "red",
               DE = "orange",
               FL = "yellow",
               GA = "green",
               LA = "blue",

               MA = "navy",
               MD = "purple",
               ME = "lightgreen",

```

```

MS = "lightblue",

NC = "pink",
NJ = "darkgreen",
NY = "darkred",
RI = "black",

SC = "gold",
TX = "grey",
VA = "darkgrey"),

labels = c(AL = "Alabama",
DE = "Delaware",
FL = "Florida",
GA = "Georgia",
LA = "Louisiana",
MA = "Massachusetts",
MD = "Maryland",
ME = "Maine",
MS = "Mississippi",
NC = "North Carolina",
NJ = "New Jersey",
NY = "New York",
RI = "Rhode Island",
SC = "South Carolina",
TX = "Texas",
VA = "Virginia"))

```

fit_3_plot

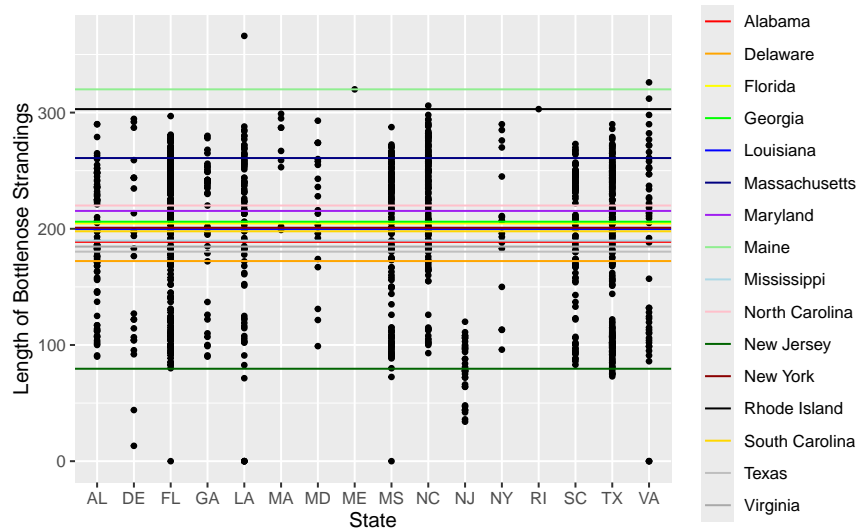


Figure 5: The distribution of lengths (cm) of stranded bottlenose dolphins across all states.

```

#plot
plot6 <- ggplot(turbine_data, aes(x = State, y = Length, color=State)) +
  geom_point(shape = 1) +

```

```

geom_hline(yintercept = 260.8889, color = "red") +
geom_hline(yintercept = 200.9059, color = "forestgreen") +
geom_hline(yintercept = 303, color = "skyblue") +
geom_hline(yintercept = 184.6717, color = "purple") +
xlab("State") +
ylab("Length (cm)") +
ggtitle("Linear Regression of Length across Offshore Wind States") +
labs(color = "States") +
theme_bw()
plot6

```

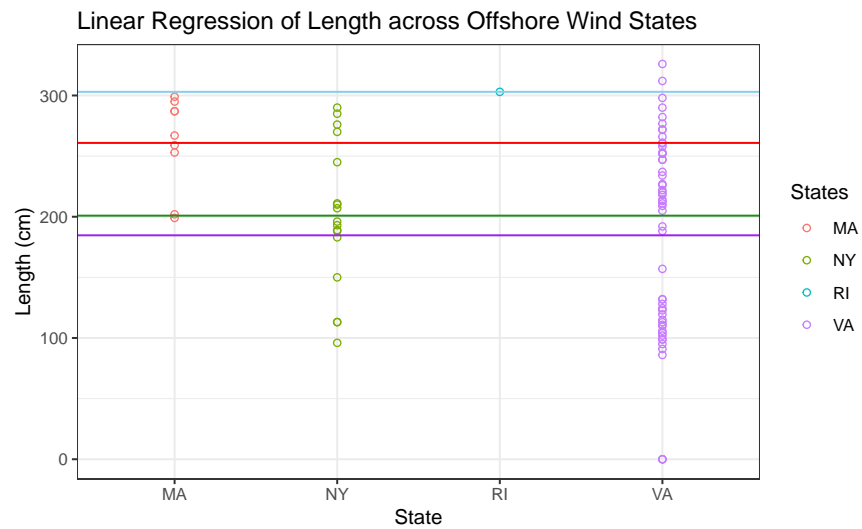


Figure 6: Linear regression of the lengths (cm) of stranded dolphin across states with offshore wind projects.