

Ejercicio 2.2: Detección y tratamiento de datos atípicos univariante y bivalente

Silvia Pineda

Lectura Fichero de datos

```
library(ggplot2)
library(patchwork)
data <- read.csv("ozone.csv") # import data
str(data)
```

```
'data.frame':  203 obs. of  13 variables:
 $ Month           : int  1 1 1 1 1 1 1 1 1 1 ...
 $ Day_of_month    : int  5 6 7 8 9 12 13 14 15 16 ...
 $ Day_of_week     : int  1 2 3 4 5 1 2 3 4 5 ...
 $ Ozone_reading   : num  5.34 5.77 3.69 3.89 5.76 6.39 4.73 4.35 3.94 7 ...
 $ Pressure_height : int  5760 5720 5790 5790 5700 5720 5760 5780 5830 5870 ...
 $ Wind_speed      : int  3 4 6 3 3 3 6 6 3 2 ...
 $ Humidity        : int  51 69 19 25 73 44 33 19 19 19 ...
 $ Temperature_Sandburg : int  54 35 45 55 41 51 51 54 58 61 ...
 $ Temperature_ElMonte : num  45.3 49.6 46.4 52.7 48 ...
 $ Inversion_base_height: int  1450 1568 2631 554 2083 111 492 NA 1249 NA ...
 $ Pressure_gradient : int  25 15 -33 -28 23 9 -44 -44 -53 -67 ...
 $ Inversion_temperature: num  57 53.8 54.1 64.8 52.5 ...
 $ Visibility       : int  60 60 100 250 120 150 40 200 250 200 ...
```

```
summary(data)
```

Month	Day_of_month	Day_of_week	Ozone_reading
Min. : 1.000	Min. : 1.0	Min. : 1.000	Min. : 0.72
1st Qu.: 3.000	1st Qu.: 9.0	1st Qu.: 2.000	1st Qu.: 4.77

Median : 6.000	Median :15.0	Median :3.000	Median : 8.90
Mean : 6.522	Mean :15.7	Mean :3.005	Mean :11.37
3rd Qu.:10.000	3rd Qu.:23.0	3rd Qu.:4.000	3rd Qu.:16.07
Max. :12.000	Max. :31.0	Max. :5.000	Max. :37.98

Pressure_height	Wind_speed	Humidity	Temperature_Sandburg
Min. :5320	Min. : 0.000	Min. :19.00	Min. :25.00
1st Qu.:5690	1st Qu.: 3.000	1st Qu.:46.00	1st Qu.:51.50
Median :5760	Median : 5.000	Median :64.00	Median :61.00
Mean :5746	Mean : 4.887	Mean :57.61	Mean :61.11
3rd Qu.:5830	3rd Qu.: 6.000	3rd Qu.:73.00	3rd Qu.:71.00
Max. :5950	Max. :16.000	Max. :93.00	Max. :93.00

Temperature_ElMonte	Inversion_base_height	Pressure_gradient
Min. :27.68	Min. : 111.0	Min. : -69.00
1st Qu.:49.64	1st Qu.: 676.2	1st Qu.: -14.00
Median :56.48	Median :1157.5	Median : 18.00
Mean :56.54	Mean :1522.5	Mean : 14.43
3rd Qu.:66.20	3rd Qu.:2291.5	3rd Qu.: 43.00
Max. :82.58	Max. :4337.0	Max. :107.00
	NA's :63	

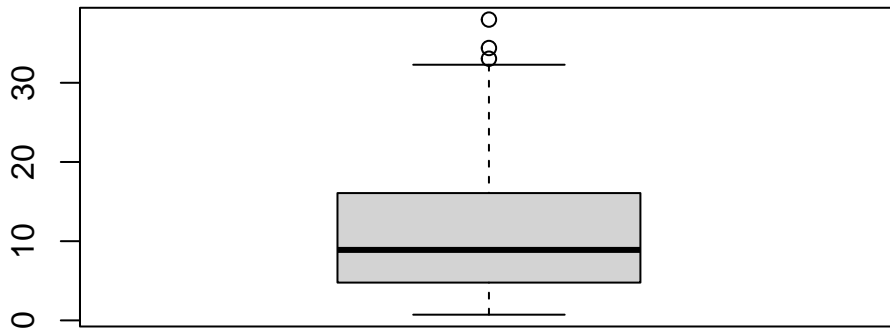
Inversion_temperature	Visibility
Min. :27.50	Min. : 0.0
1st Qu.:51.26	1st Qu.: 60.0
Median :60.98	Median :100.0
Mean :60.69	Mean :122.2
3rd Qu.:70.88	3rd Qu.:150.0
Max. :90.68	Max. :350.0

```
data$Month<-as.factor(data$Month)
data$Day_of_month<-as.factor(data$Day_of_month)
data$Day_of_week<-as.factor(data$Day_of_week)
```

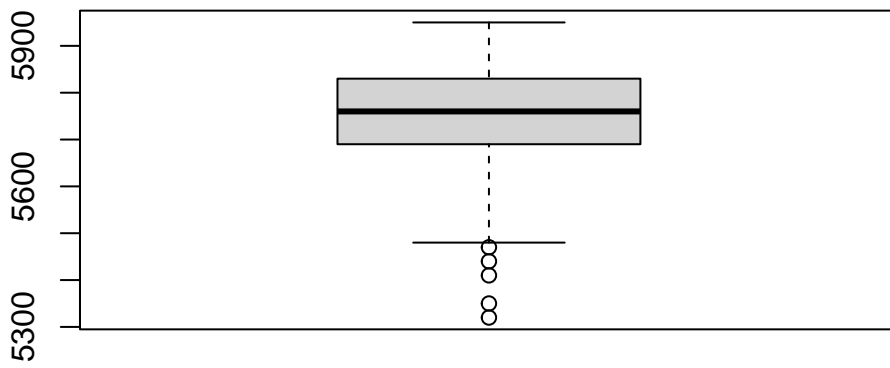
Estudio Univariante

Visualización

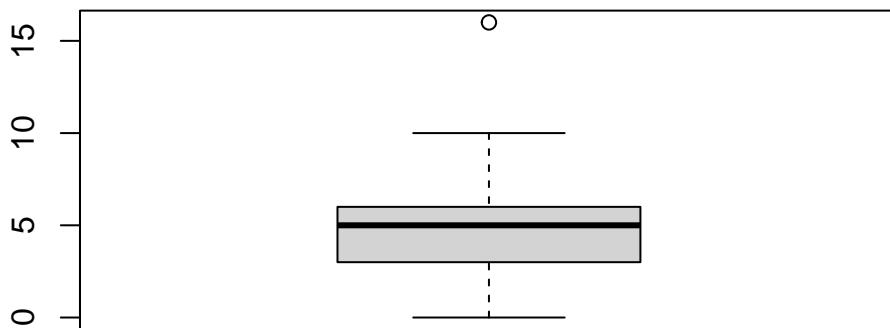
```
boxplot(data$Ozone_reading)
```



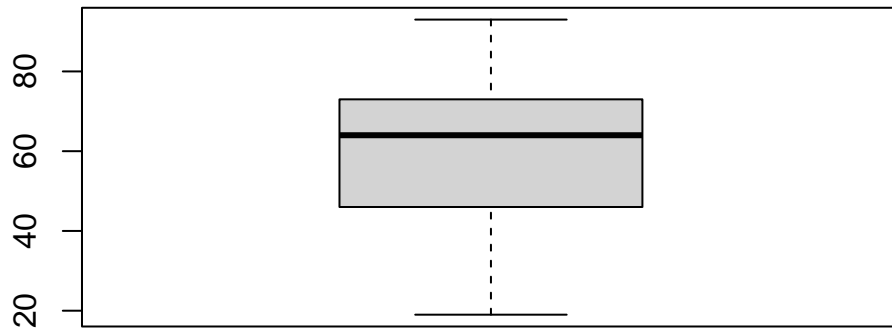
```
boxplot(data$Pressure_height)
```



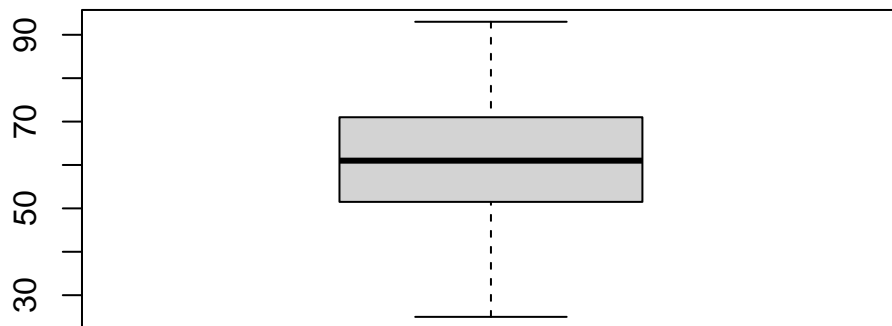
```
boxplot(data$Wind_speed)
```



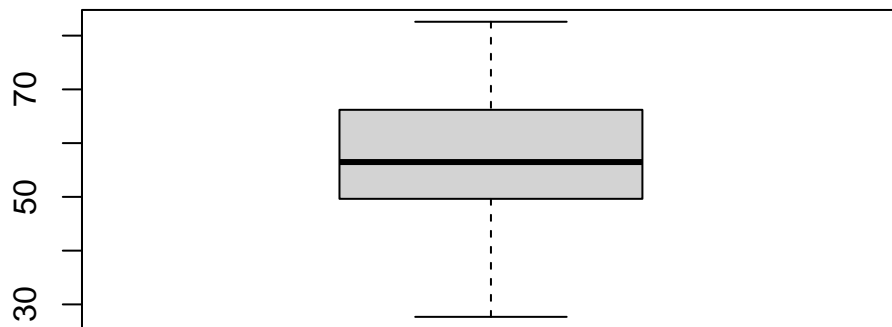
```
boxplot(data$Humidity)
```



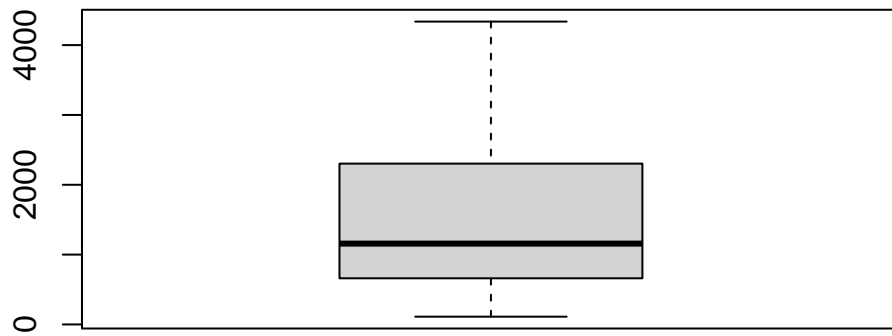
```
boxplot(data$Temperature_Sandburg)
```



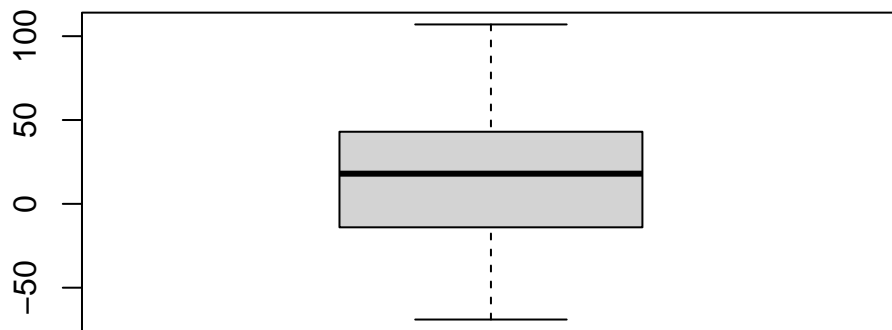
```
boxplot(data$Temperature_ElMonte)
```



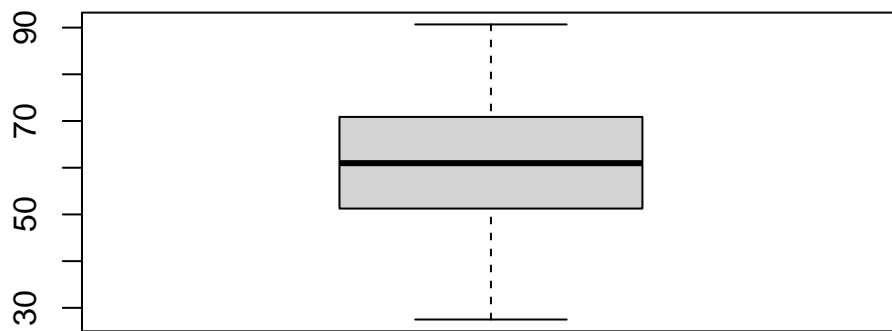
```
boxplot(data$Inversion_base_height)
```



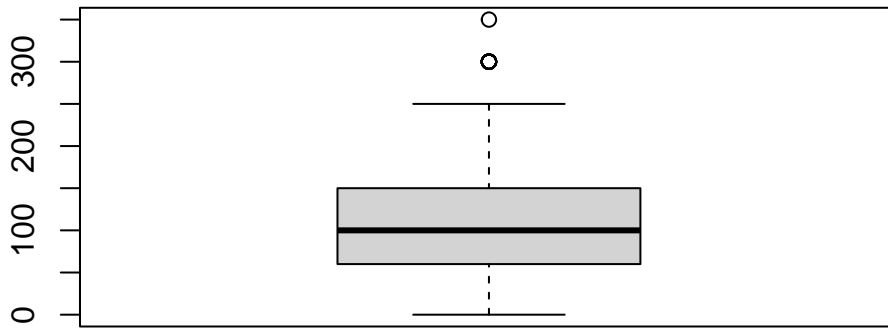
```
boxplot(data$Pressure_gradient)
```



```
boxplot(data$Inversion_temperature)
```



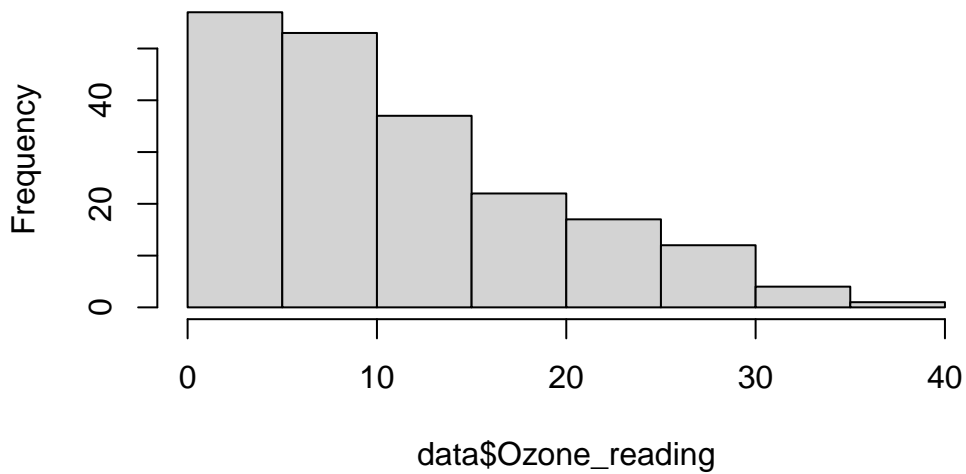
```
boxplot(data$Visibility)
```



```
#Solo hago histogramas para las variables con atípicos
```

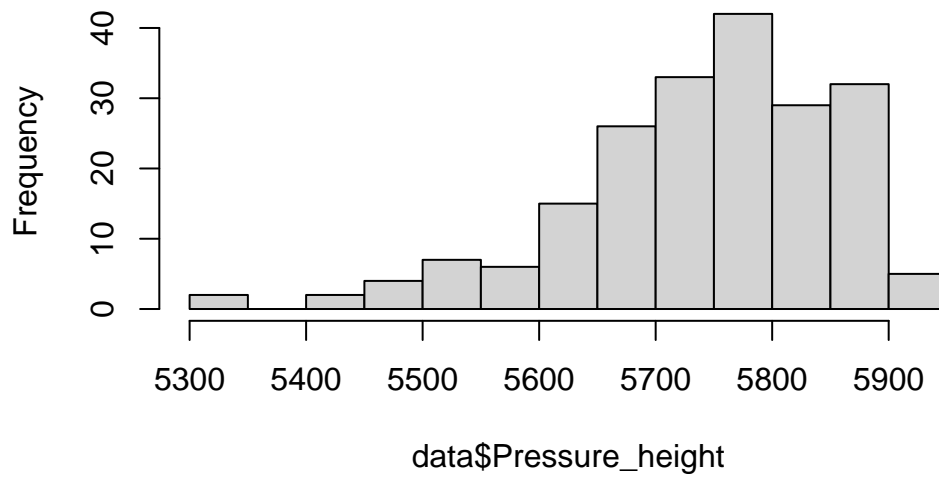
```
hist(data$Ozone_reading)
```

Histogram of data\$Ozone_reading



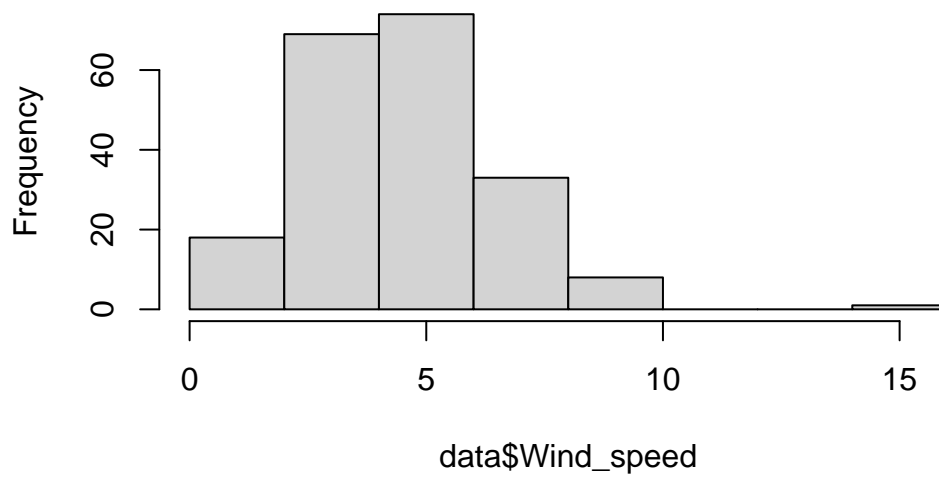
```
hist(data$Pressure_height)
```

Histogram of data\$Pressure_height

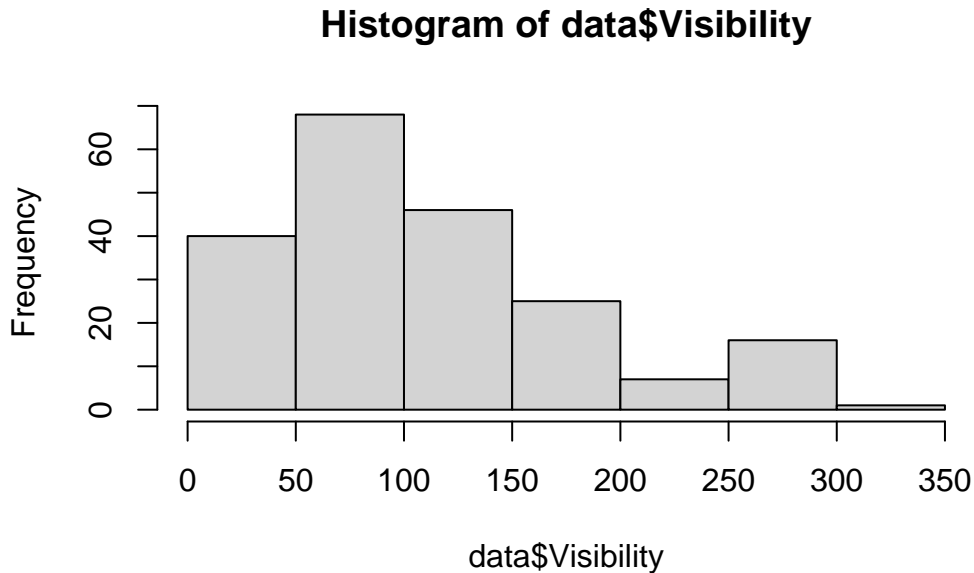


```
hist(data$Wind_speed)
```

Histogram of data\$Wind_speed



```
hist(data$Visibility)
```



Las variables con atípicos son: Ozone_reading, Pressure_height, Wind_speed y Visibility

Cuantificación

```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Ozone_reading)$out # outlier values.
out_ind <- which(data$Ozone_reading %in% c(outlier_values)) # índices

###Los valores extremos son:
extreme_values <- boxplot.stats(data$Ozone_reading,coef=3)$out # extreme values.
ext_ind <- which(data$Ozone_reading %in% c(extreme_values)) # índices

####Miramos la proporción de outliers y extremos
p<-length(out_ind)/length(data$Ozone_reading)*100
q<-length(ext_ind)/length(data$Ozone_reading)*100
cat("El % de outliers para la variable Ozone_reading es:", p,"\n")
```

El % de outliers para la variable Ozone_reading es: 1.477833


```
cat("El % de extremos para la variable Ozone_readings", q, "\n")
```

El % de extremos para la variable Ozone_readings 0

```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Pressure_height)$out # outlier values.
out_ind <- which(data$Pressure_height %in% c(outlier_values)) # índices

###Los valores extremos son:
extreme_values <- boxplot.stats(data$Pressure_height,coef=3)$out # extreme values.
ext_ind <- which(data$Pressure_height %in% c(extreme_values)) # índices

####Miramos la proporción de outliers y extremos
p<-length(out_ind)/length(data$Pressure_height)*100
q<-length(ext_ind)/length(data$Pressure_height)*100
cat("El % de outliers para la variable Pressure_height es:", p, "\n")
```

El % de outliers para la variable Pressure_height es: 2.463054

```
cat("El % de extremos para la variable Pressure_height", q, "\n")
```

El % de extremos para la variable Pressure_height 0

```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Wind_speed)$out # outlier values.
out_ind <- which(data$Wind_speed %in% c(outlier_values)) # índices

###Los valores extremos son:
extreme_values <- boxplot.stats(data$Wind_speed,coef=3)$out # extreme values.
ext_ind <- which(data$Wind_speed %in% c(extreme_values)) # índices

####Miramos la proporción de outliers y extremos
p<-length(out_ind)/length(data$Wind_speed)*100
q<-length(ext_ind)/length(data$Wind_speed)*100
cat("El % de outliers para la variable Wind_speed es:", p, "\n")
```

El % de outliers para la variable Wind_speed es: 0.4926108

```
cat("El % de extremos para la variable Wind_speed", q, "\n")
```

El % de extremos para la variable Wind_speed 0.4926108

```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Visibility)$out # outlier values.
out_ind <- which(data$Visibility %in% c(outlier_values)) # índices

###Los valores extremos son:
extreme_values <- boxplot.stats(data$Visibility,coef=3)$out # extreme values.
ext_ind <- which(data$Visibility %in% c(extreme_values)) # índices

####Miramos la proporción de outliers y extremos
p<-length(out_ind)/length(data$Visibility)*100
q<-length(ext_ind)/length(data$Visibility)*100
cat("El % de outliers para la variable Visibility es:", p, "\n")
```

El % de outliers para la variable Visibility es: 8.374384

```
cat("El % de extremos para la variable Visibility", q, "\n")
```

El % de extremos para la variable Visibility 0

Las variables con datos atípicos son:

Ozone_reading (1.48%): valores muy grandes que parecen parte de una distribución asimétrica

Pressure_height (2.46%): valores muy pequeños que parecen parte de una distribución asimétrica

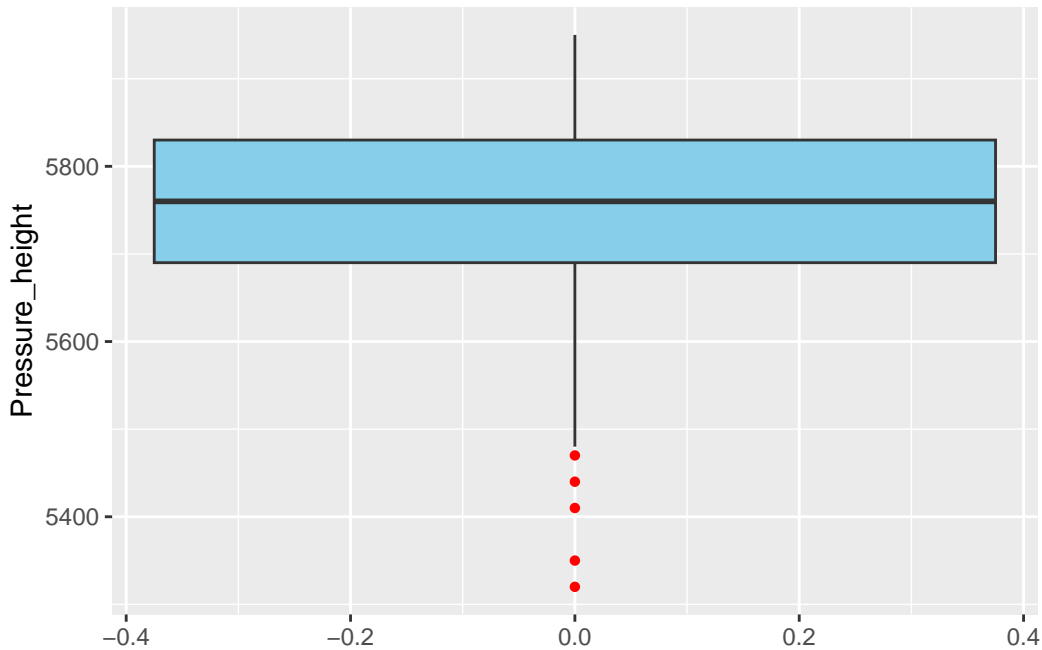
Wind_speed (0.49%): Valor que es también extremo y que se sale completamente de la distribución

Visibility (8.37%): Valores que corresponden a los mismos valores de 300 y 350 que parecen claramente parte de la variable. Además es un número muy elevado como para ser dato atípico.

Vamos por tanto a realizar el estudio bivalente de Pressure_height, Ozone_reading y Wind_speed

Estudio de la variable Pressure_height

```
##### Pressure heigth #####
ggplot(data, aes(y = Pressure_height)) +
  geom_boxplot(fill = "skyblue", outlier.color = "red", outlier.shape = 16)
```



```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Pressure_height)$out # outlier values.
out_ind <- which(data$Pressure_height %in% c(outlier_values))
data[out_ind,]
```

	Month	Day_of_month	Day_of_week	Ozone_reading	Pressure_height	Wind_speed
21	2	5	4	2.94	5410	6
22	2	6	5	2.74	5350	7
36	3	2	2	3.22	5470	7
37	3	3	3	2.79	5320	16
64	4	13	2	3.65	5440	5

	Humidity	Temperature_Sandburg	Temperature_ElMonte	Inversion_base_height
21	64	31	32.18	NA
22	62	30	32.54	1341
36	46	30	29.66	NA
37	45	25	27.68	NA

64	44	35	33.08	NA
	Pressure_gradient	Inversion_temperature	Visibility	
21	28	32.36	200	
22	18	45.86	60	
36	44	29.30	300	
37	39	27.50	200	
64	24	32.54	80	

```
### Los valores extremos son:
extreme_values <- boxplot.stats(data$Pressure_height,coef=3)$out # extreme values.
ext_ind <- which(data$Pressure_height %in% c(extreme_values))
data[ext_ind,]
```

```
[1] Month           Day_of_month      Day_of_week
[4] Ozone_reading    Pressure_height    Wind_speed
[7] Humidity          Temperature_Sandburg Temperature_ElMonte
[10] Inversion_base_height Pressure_gradient    Inversion_temperature
[13] Visibility
<0 rows> (or 0-length row.names)
```

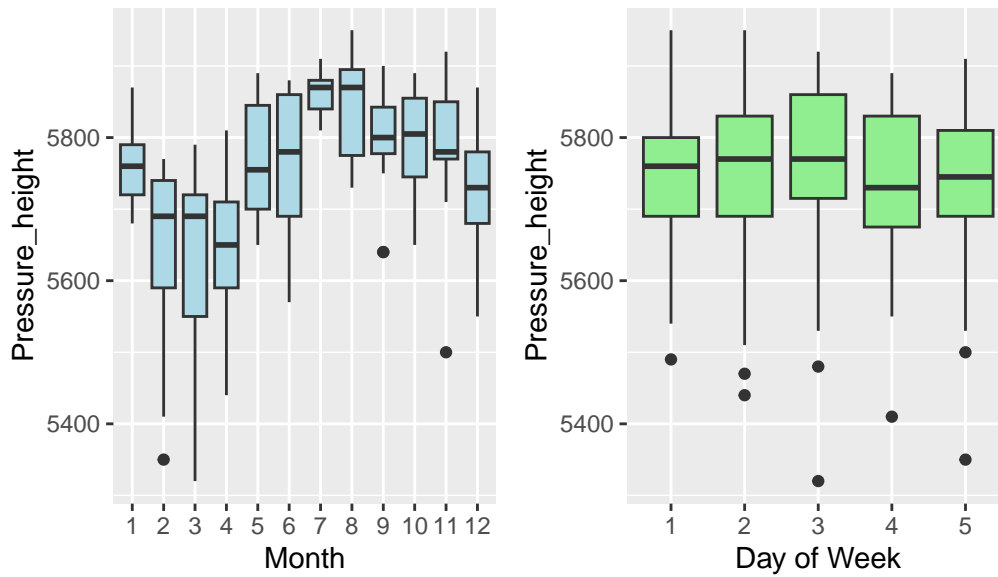
```
library(patchwork) # Para combinar gráficos fácilmente

# Gráfico 1: Pressure Height por mes
p1 <- ggplot(data, aes(x = as.factor(Month), y = Pressure_height)) +
  geom_boxplot(fill = "lightblue") +
  labs(title = "Pressure_height across months", x = "Month", y = "Pressure_height")

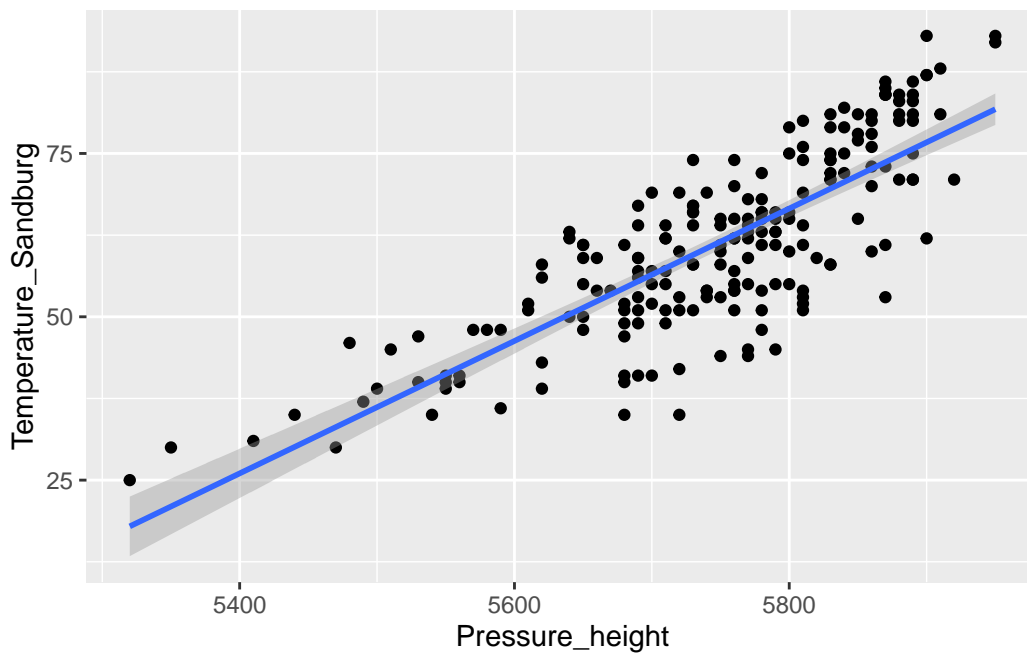
# Gráfico 2: Pressure Height por día de la semana
p2 <- ggplot(data, aes(x = as.factor(Day_of_week), y = Pressure_height)) +
  geom_boxplot(fill = "lightgreen") +
  labs(title = "Pressure_height for days of week", x = "Day of Week", y = "Pressure_height")

# Combinar ambos gráficos en una fila
p1 + p2
```

Pressure_height across months Pressure_height for days o



```
ggp <- ggplot(data,aes(Pressure_height, Temperature_Sandburg)) + geom_point()
ggp + stat_smooth(method = "lm",
                  formula = y ~ x,
                  geom = "smooth")
```



```
summary(lm(data$Pressure_height~data$Temperature_Sandburg))
```

Call:

```
lm(formula = data$Pressure_height ~ data$Temperature_Sandburg)
```

Residuals:

Min	1Q	Median	3Q	Max
-196.559	-41.846	1.171	39.099	175.891

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	5354.1021	20.8228	257.13	<2e-16 ***
data\$Temperature_Sandburg	6.4152	0.3319	19.33	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 67.02 on 201 degrees of freedom

Multiple R-squared: 0.6502, Adjusted R-squared: 0.6484

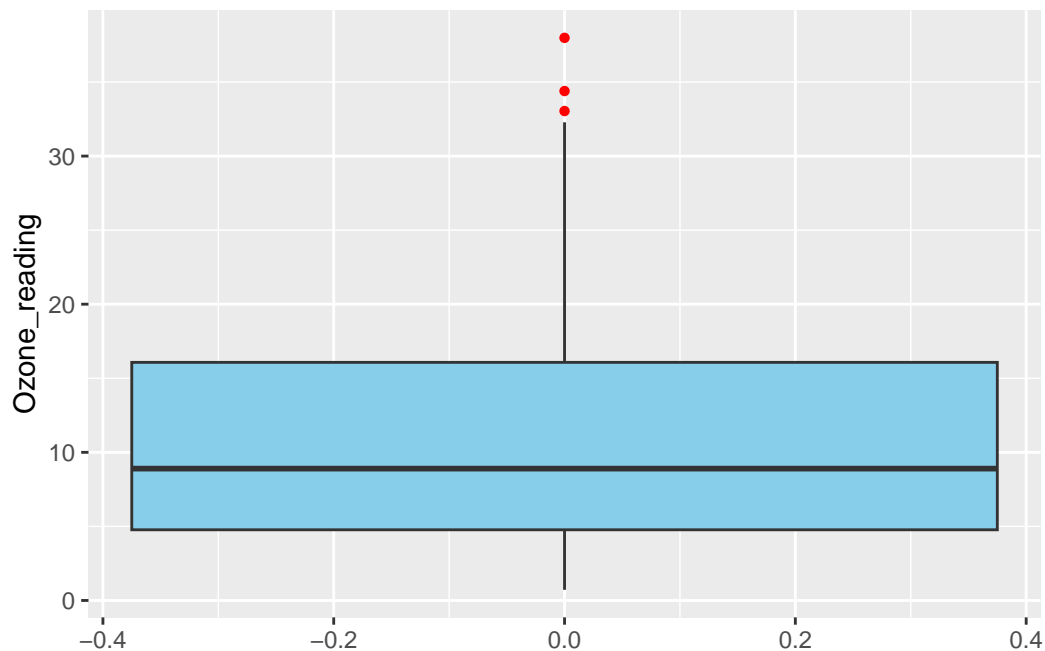
F-statistic: 373.6 on 1 and 201 DF, p-value: < 2.2e-16

Esta variable está claramente asociada con los meses del año, perteneciendo los valores más altos de esta variable a los meses de verano. Además vemos una clara asociación con la variable de temperatura.

CONCLUSIÓN: No borramos estos valores atípicos porque son parte de una asociación,

Estudio de la variable Ozone Reading

```
##### OZONE READING #####
ggplot(data, aes(y = Ozone_reading)) +
  geom_boxplot(fill = "skyblue", outlier.color = "red", outlier.shape = 16)
```



```
###Los valores atípicos son:
outlier_values <- boxplot.stats(data$Ozone_reading)$out # outlier values.
out_ind <- which(data$Ozone_reading %in% c(outlier_values))
data[out_ind,]
```

	Month	Day_of_month	Day_of_week	Ozone_reading	Pressure_height	Wind_speed
82	5	12	3	33.04	5880	3
104	7	6	2	34.39	5900	6
130	8	30	1	37.98	5950	5
	Humidity	Temperature_Sandburg	Temperature_ElMonte	Inversion_base_height		
82	80		80	73.04		436
104	86		87	81.68		990
130	62		92	82.40		557
	Pressure_gradient	Inversion_temperature	Visibility			
82		0	86.36	40		
104		22	85.10	40		
130		0	90.68	70		

```
###Los valores extremos son:
extreme_values <- boxplot.stats(data$Ozone_reading,coef=3)$out # extreme values.
ext_ind <- which(data$Ozone_reading %in% c(extreme_values))
data[ext_ind,]
```

```

[1] Month                Day_of_month        Day_of_week
[4] Ozone_reading         Pressure_height     Wind_speed
[7] Humidity              Temperature_Sandburg Temperature_ElMonte
[10] Inversion_base_height Pressure_gradient    Inversion_temperature
[13] Visibility
<0 rows> (or 0-length row.names)

```

```

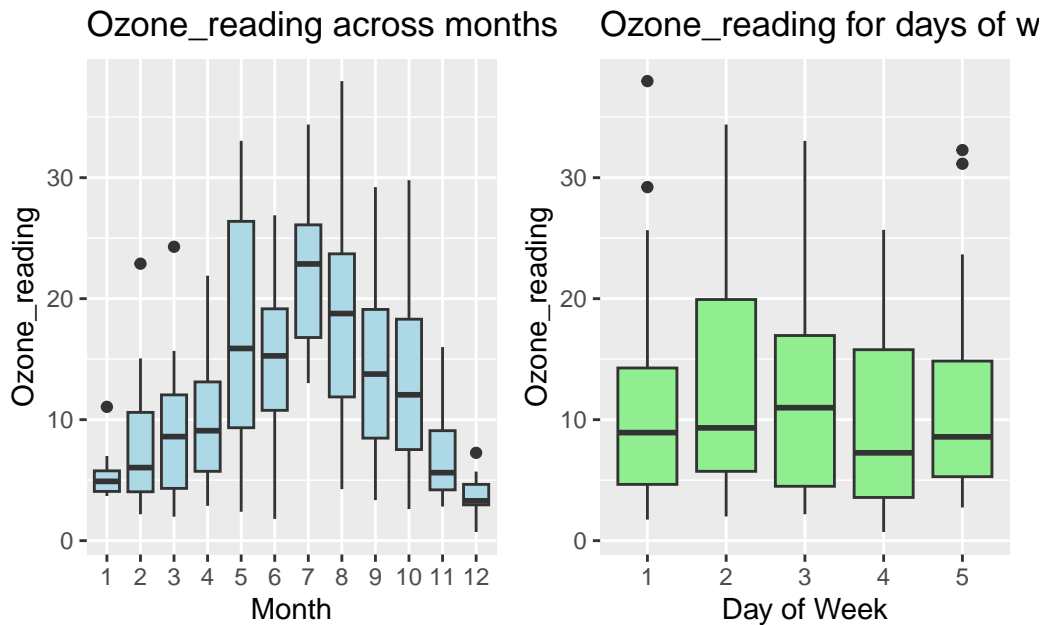
library(patchwork) # Para combinar gráficos fácilmente

# Gráfico 1: Ozone_reading por mes
p1 <- ggplot(data, aes(x = as.factor(Month), y = Ozone_reading)) +
  geom_boxplot(fill = "lightblue") +
  labs(title = "Ozone_reading across months", x = "Month", y = "Ozone_reading")

# Gráfico 2: Ozone_reading por día de la semana
p2 <- ggplot(data, aes(x = as.factor(Day_of_week), y = Ozone_reading)) +
  geom_boxplot(fill = "lightgreen") +
  labs(title = "Ozone_reading for days of week", x = "Day of Week", y = "Ozone_reading")

# Combinar ambos gráficos en una fila
p1 + p2

```




```
ggp <- ggplot(data,aes(Ozone_reading, Temperature_Sandburg)) + geom_point()
ggp + stat_smooth(method = "lm",
                  formula = y ~ x,
                  geom = "smooth")
```



```
summary(lm(data$Ozone_reading~data$Temperature_Sandburg))
```

Call:

```
lm(formula = data$Ozone_reading ~ data$Temperature_Sandburg)
```

Residuals:

Min	1Q	Median	3Q	Max
-10.4273	-3.8316	-0.4737	3.2197	15.1344

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-15.88133	1.61779	-9.817	<2e-16 ***
data\$Temperature_Sandburg	0.44598	0.02579	17.294	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

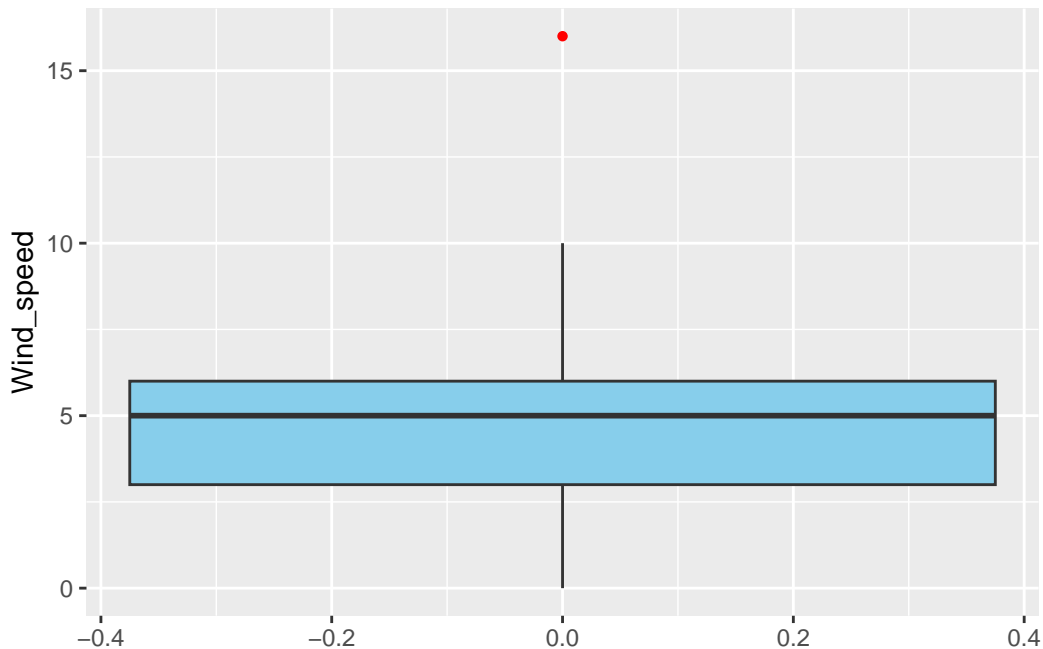
Residual standard error: 5.207 on 201 degrees of freedom
Multiple R-squared: 0.5981, Adjusted R-squared: 0.5961
F-statistic: 299.1 on 1 and 201 DF, p-value: < 2.2e-16

De la misma forma que la variable anterior, esta variable está claramente asociada con los meses del año, perteneciendo los valores más altos de esta variable a los meses de verano. Además vemos una clara asociación con la variable de temperatura.

CONCLUSIÓN: No borramos estos valores atípicos porque son parte de una asociación

Estudio de la variable WIND SPEED

```
ggplot(data, aes(y = Wind_speed)) +  
  geom_boxplot(fill = "skyblue", outlier.color = "red", outlier.shape = 16)
```



```
###Los valores atípicos son:  
outlier_values <- boxplot.stats(data$Wind_speed)$out # outlier values.  
out_ind <- which(data$Wind_speed %in% c(outlier_values))  
data[out_ind,]
```

```

      Month Day_of_month Day_of_week Ozone_reading Pressure_height Wind_speed
37      3           3           3           2.79           5320           16
      Humidity Temperature_Sandburg Temperature_ElMonte Inversion_base_height
37      45           25           27.68           NA
      Pressure_gradient Inversion_temperature Visibility
37           39           27.5           200

```

```

###Los valores extremos son:
extreme_values <- boxplot.stats(data$Wind_speed,coef=3)$out # extreme values.
ext_ind <- which(data$Wind_speed %in% c(extreme_values))
data[ext_ind,]

```

```

      Month Day_of_month Day_of_week Ozone_reading Pressure_height Wind_speed
37      3           3           3           2.79           5320           16
      Humidity Temperature_Sandburg Temperature_ElMonte Inversion_base_height
37      45           25           27.68           NA
      Pressure_gradient Inversion_temperature Visibility
37           39           27.5           200

```

```

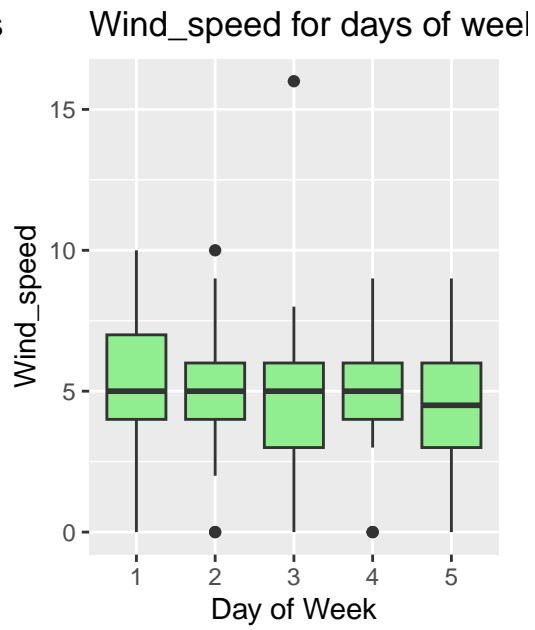
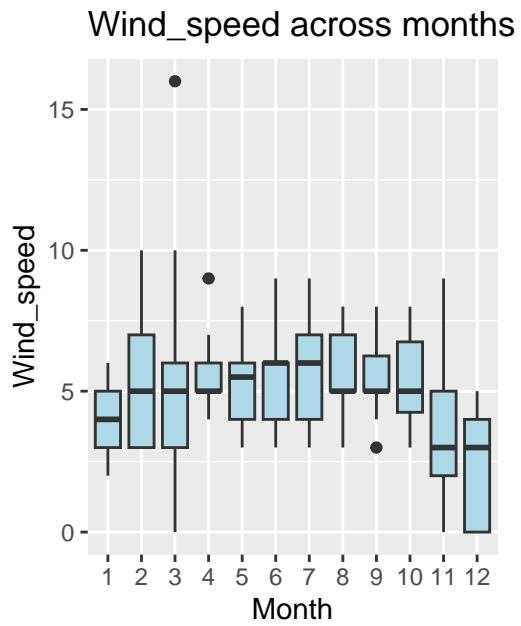
library(patchwork) # Para combinar gráficos fácilmente

# Gráfico 1: Pressure Height por mes
p1 <- ggplot(data, aes(x = as.factor(Month), y = Wind_speed)) +
  geom_boxplot(fill = "lightblue") +
  labs(title = "Wind_speed across months", x = "Month", y = "Wind_speed")

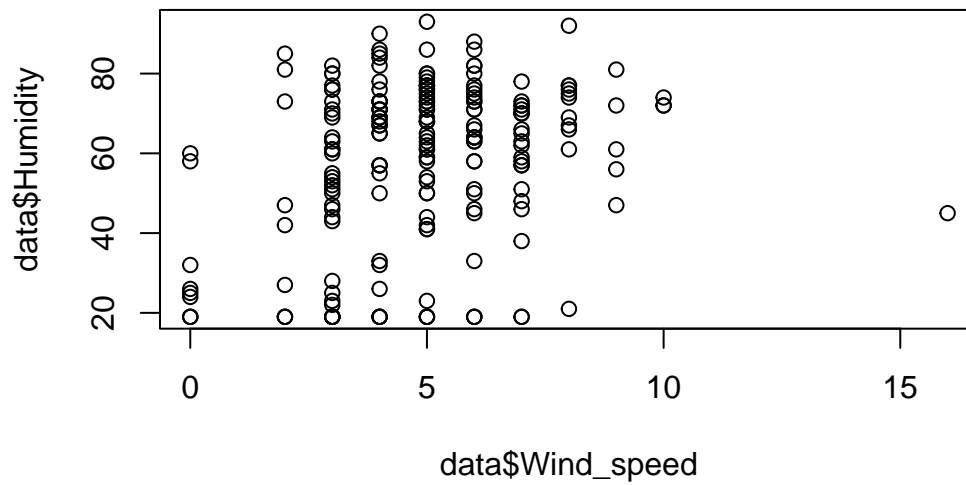
# Gráfico 2: Pressure Height por día de la semana
p2 <- ggplot(data, aes(x = as.factor(Day_of_week), y = Wind_speed)) +
  geom_boxplot(fill = "lightgreen") +
  labs(title = "Wind_speed for days of week", x = "Day of Week", y = "Wind_speed")

# Combinar ambos gráficos en una fila
p1 + p2

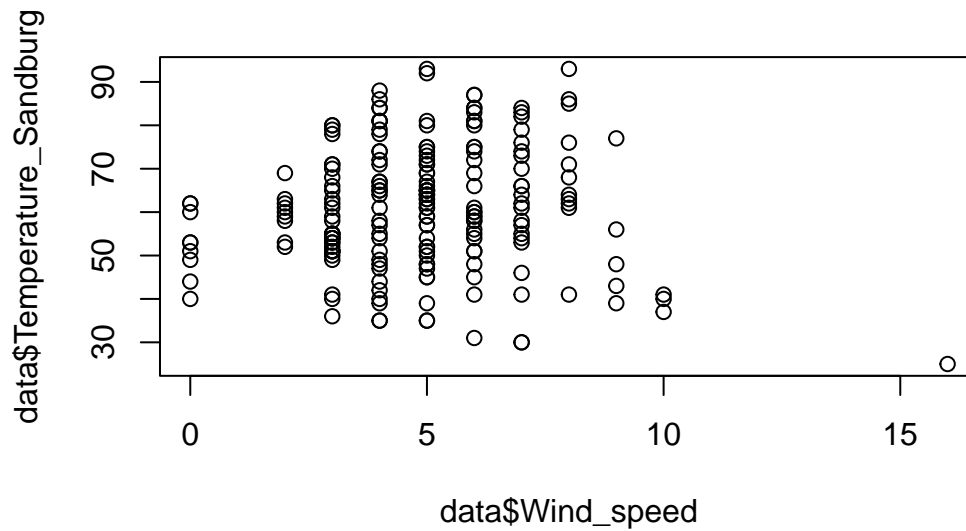
```



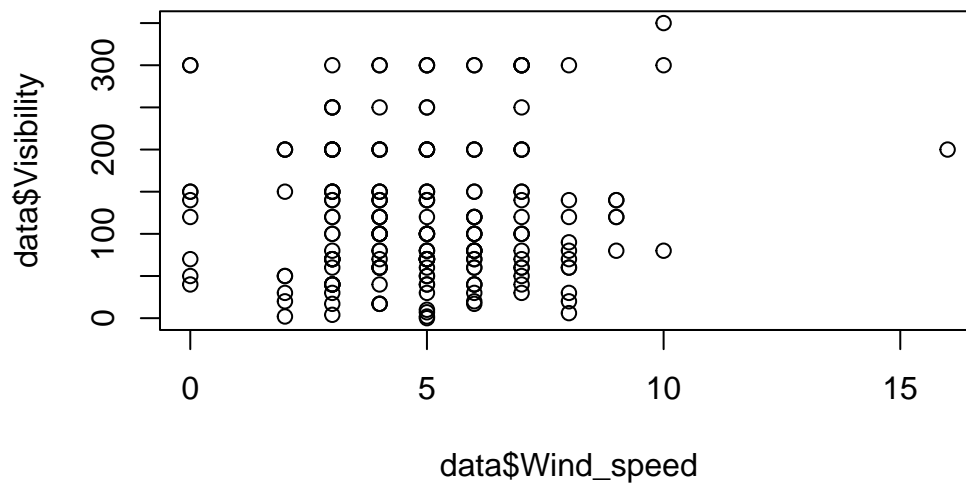
```
plot(data$Wind_speed,data$Humidity)
```



```
plot(data$Wind_speed,data$Temperature_Sandburg)
```



```
plot(data$Wind_speed,data$Visibility)
```



En este caso vemos que el outlier de `wind_speed` no está asociado con las variables de interés y además es un extremo.

CONCLUSIÓN: Este outlier no tiene ninguna asociación aparente, por tanto este dato missing si lo quitamos

```
outlier_values <- boxplot.stats(data$Wind_speed)$out # outlier values.
out_ind <- which(data$Wind_speed %in% c(outlier_values))
data[out_ind,"Wind_speed"]<-NA
```