

Lab 3

Lab 03 - Exploratory Data Analysis

Learning Goals

- Read in and get familiar with the meteorology dataset
- Step through the EDA “checklist” presented in the class slides
- Practice making exploratory graphs

As you do this, think about what questions you would like to ask regarding this data. What would you ask a collaborator who was more familiar with it?

Lab Description

We will work with the meteorological data presented in lecture. Recall the dataset consists of weather station readings in the continental US.

The objectives of the lab are to find the weather station with the highest elevation and look at patterns in the time series of its wind speed and temperature.

1. Read in the data

First download and then read in with `data.table::fread()`. This is slightly faster than some of the more common functions, such as `read.table`, but it produces a different type of object, which is why we need to convert it into a `data.frame` after reading it in.

```
download.file(
  "https://raw.githubusercontent.com/USCbiostats/data-science-data/master/02_met/met_all.gz"
  destfile = file.path("~", "Downloads", "met_all.gz"),
  method    = "libcurl",
  timeout    = 60
)

met <- data.table::fread(file.path("~", "Downloads", "met_all.gz"))
met <- as.data.frame(met)
```

2. Check the dimensions, headers, footers.

How many columns, rows are there? Some useful functions for this are `dim`, `head`, and `tail`.

```
dim(met)
```

```
[1] 2377343      30
```

```
# There are 2377343 rows and 30 columns
```

```
head(met)
```

	USAFID	WBAN	year	month	day	hour	min	lat	lon	elev	wind.dir	wind.dir.qc
1	690150	93121	2019	8	1	0	56	34.3	-116.166	696	220	5
2	690150	93121	2019	8	1	1	56	34.3	-116.166	696	230	5
3	690150	93121	2019	8	1	2	56	34.3	-116.166	696	230	5
4	690150	93121	2019	8	1	3	56	34.3	-116.166	696	210	5
5	690150	93121	2019	8	1	4	56	34.3	-116.166	696	120	5
6	690150	93121	2019	8	1	5	56	34.3	-116.166	696	NA	9
	wind.type.code	wind.sp	wind.sp.qc	ceiling.ht	ceiling.ht.qc	ceiling.ht.method						
1		N	5.7	5	22000	5						9
2		N	8.2	5	22000	5						9
3		N	6.7	5	22000	5						9
4		N	5.1	5	22000	5						9
5		N	2.1	5	22000	5						9
6		C	0.0	5	22000	5						9
	sky.cond	vis.dist	vis.dist.qc	vis.var	vis.var.qc	temp	temp.qc	dew.point				
1		N	16093	5	N	5	37.2	5				10.6

2	N	16093	5	N	5 35.6	5	10.6
3	N	16093	5	N	5 34.4	5	7.2
4	N	16093	5	N	5 33.3	5	5.0
5	N	16093	5	N	5 32.8	5	5.0
6	N	16093	5	N	5 31.1	5	5.6

	dew.point.qc	atm.press	atm.press.qc	rh
1	5	1009.9	5	19.88127
2	5	1010.3	5	21.76098
3	5	1010.6	5	18.48212
4	5	1011.6	5	16.88862
5	5	1012.7	5	17.38410
6	5	1012.7	5	20.01540

```
tail(met)
```

	USAFID	WBAN	year	month	day	hour	min	lat	lon	elev	wind.dir				
2377338	726813	94195	2019	8	31	18	56	43.650	-116.633	741	NA				
2377339	726813	94195	2019	8	31	19	56	43.650	-116.633	741	70				
2377340	726813	94195	2019	8	31	20	56	43.650	-116.633	741	NA				
2377341	726813	94195	2019	8	31	21	56	43.650	-116.633	741	10				
2377342	726813	94195	2019	8	31	22	56	43.642	-116.636	741	10				
2377343	726813	94195	2019	8	31	23	56	43.642	-116.636	741	40				
	wind.dir.qc	wind.type.code	wind.sp	wind.sp.qc	ceiling.ht	ceiling.ht.qc									
2377338	9	C	0.0	5	22000	5									
2377339	5	N	2.1	5	22000	5									
2377340	9	C	0.0	5	22000	5									
2377341	5	N	2.6	5	22000	5									
2377342	1	N	2.1	1	22000	1									
2377343	1	N	2.1	1	22000	1									
	ceiling.ht.method	sky.cond	vis.dist	vis.dist.qc	vis.var	vis.var.qc	temp								
2377338	9	N	16093	5	N	5	30.0								
2377339	9	N	16093	5	N	5	32.2								
2377340	9	N	16093	5	N	5	33.3								
2377341	9	N	14484	5	N	5	35.0								
2377342	9	N	16093	1	9	9	34.4								
2377343	9	N	16093	1	9	9	34.4								
	temp.qc	dew.point	dew.point.qc	atm.press	atm.press.qc	rh									
2377338	5	11.7	5	1013.6	5	32.32509									
2377339	5	12.2	5	1012.8	5	29.40686									
2377340	5	12.2	5	1011.6	5	27.60422									
2377341	5	9.4	5	1010.8	5	20.76325									
2377342	1	9.4	1	1010.1	1	21.48631									

2377343	1	9.4	1	1009.6	1	21.48631
---------	---	-----	---	--------	---	----------

3. Take a look at the variables.

Show the type (class) of each variable (hint: try the `str` function).

What are the names of the key variables related to our question of interest?

```
str(met)
```

```
'data.frame':   2377343 obs. of  30 variables:
 $ USAFID      : int  690150 690150 690150 690150 690150 690150 690150 690150 690150 690150 690150 ...
 $ WBAN        : int  93121 93121 93121 93121 93121 93121 93121 93121 93121 93121 93121 ...
 $ year        : int  2019 2019 2019 2019 2019 2019 2019 2019 2019 2019 2019 ...
 $ month       : int    8 8 8 8 8 8 8 8 8 8 8 ...
 $ day         : int    1 1 1 1 1 1 1 1 1 1 1 ...
 $ hour        : int    0 1 2 3 4 5 6 7 8 9 ...
 $ min         : int   56 56 56 56 56 56 56 56 56 56 ...
 $ lat         : num  34.3 34.3 34.3 34.3 34.3 34.3 34.3 34.3 34.3 34.3 ...
 $ lon         : num -116 -116 -116 -116 -116 ...
 $ elev        : int  696 696 696 696 696 696 696 696 696 696 ...
 $ wind.dir     : int  220 230 230 210 120 NA 320 10 320 350 ...
 $ wind.dir.qc  : chr   "5" "5" "5" "5" ...
 $ wind.type.code : chr   "N" "N" "N" "N" ...
 $ wind.sp      : num   5.7 8.2 6.7 5.1 2.1 0 1.5 2.1 2.6 1.5 ...
 $ wind.sp.qc   : chr   "5" "5" "5" "5" ...
 $ ceiling.ht   : int 22000 22000 22000 22000 22000 22000 22000 22000 22000 22000 ...
 $ ceiling.ht.qc : int    5 5 5 5 5 5 5 5 5 5 ...
 $ ceiling.ht.method: chr   "9" "9" "9" "9" ...
 $ sky.cond     : chr   "N" "N" "N" "N" ...
 $ vis.dist     : int 16093 16093 16093 16093 16093 16093 16093 16093 16093 16093 ...
 $ vis.dist.qc  : chr   "5" "5" "5" "5" ...
 $ vis.var      : chr   "N" "N" "N" "N" ...
 $ vis.var.qc   : chr   "5" "5" "5" "5" ...
 $ temp         : num  37.2 35.6 34.4 33.3 32.8 31.1 29.4 28.9 27.2 26.7 ...
 $ temp.qc      : chr   "5" "5" "5" "5" ...
 $ dew.point    : num  10.6 10.6 7.2 5 5 5.6 6.1 6.7 7.8 7.8 ...
 $ dew.point.qc : chr   "5" "5" "5" "5" ...
 $ atm.press    : num  1010 1010 1011 1012 1013 ...
 $ atm.press.qc : int    5 5 5 5 5 5 5 5 5 5 ...
```

```
$ rh          : num  19.9 21.8 18.5 16.9 17.4 ...
```

```
# Numeric: USAFID, WBAN, year, month, day, hour, min, lat, lon, elev, wind.dir, wind.sp, cei.  
# Categorical: wind.dir.qc, wind.type.code, sky.cond, vis.var  
  
# Variables of interest: year, month, day, hour, temp, elev, temp, wind.sp
```

4. Take a closer look at the key variables.

```
table(met$year)
```

```
2019  
2377343
```

```
table(met$day)
```

```
  1    2    3    4    5    6    7    8    9   10   11   12   13  
75975 75923 76915 76594 76332 76734 77677 77766 75366 75450 76187 75052 76906  
 14   15   16   17   18   19   20   21   22   23   24   25   26  
77852 76217 78015 78219 79191 76709 75527 75786 78312 77413 76965 76806 79114  
 27   28   29   30   31  
79789 77059 71712 74931 74849
```

```
table(met$hour)
```

```
  0    1    2    3    4    5    6    7    8    9   10  
99434 93482 93770 96703 110504 112128 106235 101985 100310 102915 101880  
 11   12   13   14   15   16   17   18   19   20   21  
100470 103605 97004 96507 97635 94942 94184 100179 94604 94928 96070  
 22   23  
94046 93823
```

```
summary(met$temp)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-40.00	19.60	23.50	23.59	27.80	56.00	60089

```
summary(met$elev)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-13.0	101.0	252.0	415.8	400.0	9999.0

```
summary(met$wind.sp)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
0.00	0.00	2.10	2.46	3.60	36.00	79693

It looks like the elevation variable has observations with 9999.0, which is probably an indicator for missing. We should take a deeper look at the [data dictionary](#) to confirm. The wind speed variable is OK but there is a lot of missing data.

After checking the data we should make the appropriate modifications. Replace elevations with 9999 as NA.

```
met[met$elev==9999.0, ] <- NA  
summary(met$elev)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-13	101	252	413	400	4113	710

At what elevation is the highest weather station?

```
# The highest weather station has an elevation 4113 m
```

We also have the issue of the minimum temperature being -40C, which seems implausible, so we should remove those observations.

```
met <- met[met$temp > -40, ]
head(met[order(met$temp), ])
```

	USAFID	WBAN	year	month	day	hour	min	lat	lon	elev	wind.dir
1203053	722817	3068	2019	8	1	0	56	38.767	-104.3	1838	190
1203055	722817	3068	2019	8	1	1	56	38.767	-104.3	1838	180
1203128	722817	3068	2019	8	3	11	56	38.767	-104.3	1838	NA
1203129	722817	3068	2019	8	3	12	56	38.767	-104.3	1838	NA
1203222	722817	3068	2019	8	6	21	56	38.767	-104.3	1838	280
1203225	722817	3068	2019	8	6	22	56	38.767	-104.3	1838	240
	wind.dir.qc	wind.type.code	wind.sp	wind.sp.qc	ceiling.ht	ceiling.ht.qc					
1203053	5	N	7.2	5	NA	9					
1203055	5	N	7.7	5	NA	9					
1203128	9	C	0.0	5	NA	9					
1203129	9	C	0.0	5	NA	9					
1203222	5	N	2.6	5	NA	9					
1203225	5	N	7.7	5	NA	9					
	ceiling.ht.method	sky.cond	vis.dist	vis.dist.qc	vis.var	vis.var.qc					
1203053	9	N	NA	9	N	5					
1203055	9	N	NA	9	N	5					
1203128	9	N	NA	9	N	5					
1203129	9	N	NA	9	N	5					
1203222	9	N	NA	9	N	5					
1203225	9	N	NA	9	N	5					
	temp	temp.qc	dew.point	dew.point.qc	atm.press	atm.press.qc	rh				
1203053	-17.2	5	NA	9	NA	9	NA				
1203055	-17.2	5	NA	9	NA	9	NA				
1203128	-17.2	5	NA	9	NA	9	NA				
1203129	-17.2	5	NA	9	NA	9	NA				
1203222	-17.2	5	NA	9	NA	9	NA				
1203225	-17.2	5	NA	9	NA	9	NA				

There are still some suspiciously low values for temperature (-17.2C), but we will deal with those later.

We should also check the wind speed variable for any abnormalities.

How many missing values are there in the wind.sp variable?

```
summary(met$wind.sp)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
0.00	0.00	2.10	2.46	3.60	36.00	91853

```
# There are 79714 missing values for this variable
```

5. Check the data against an external data source.

We should check the suspicious temperature value (where is it located?) and validate that the range of elevations make sense (-13m to 4113m).

Google is your friend here.

Fix any problems that arise in your checks.

Where was the location for the coldest temperature readings (-17.2C)? Do these seem reasonable in context?

```
met <- met[met$temp > -40, ]  
met <- met[!is.na(met$temp), ]  
met <- met[order(met$temp), ]  
head(met) [,c(8:10, 24)]
```

	lat	lon	elev	temp
1203053	38.767	-104.3	1838	-17.2
1203055	38.767	-104.3	1838	-17.2
1203128	38.767	-104.3	1838	-17.2
1203129	38.767	-104.3	1838	-17.2
1203222	38.767	-104.3	1838	-17.2
1203225	38.767	-104.3	1838	-17.2

```
# With a latitude of 38.767 and longitude of -104.3, El Paso County, Colorado has the coldest
```

Does the range of values for elevation make sense? Why or why not?


```
# 4113 m is reasonable given Mount Whitney is 4417 m high and -13 m is also reasonable given
```

6. Calculate summary statistics

Remember to keep the initial question in mind. We want to pick out the weather station with maximum elevation and examine its wind speed and temperature.

Some ideas: select the weather station with maximum elevation; look at the correlation between temperature and wind speed; look at the correlation between temperature and wind speed with hour and day of the month.

```
elev <- met[which(met$elev == max(met$elev, na.rm = TRUE)), ]
summary(elev)
```

USAFID	WBAN	year	month	day
Min. :720385	Min. :419	Min. :2019	Min. :8	Min. : 1.0
1st Qu.:720385	1st Qu.:419	1st Qu.:2019	1st Qu.:8	1st Qu.: 8.0
Median :720385	Median :419	Median :2019	Median :8	Median :16.0
Mean :720385	Mean :419	Mean :2019	Mean :8	Mean :16.1
3rd Qu.:720385	3rd Qu.:419	3rd Qu.:2019	3rd Qu.:8	3rd Qu.:24.0
Max. :720385	Max. :419	Max. :2019	Max. :8	Max. :31.0

hour	min	lat	lon	elev
Min. : 0.00	Min. : 6.00	Min. :39.8	Min. : -105.8	Min. :4113
1st Qu.: 6.00	1st Qu.:13.00	1st Qu.:39.8	1st Qu.: -105.8	1st Qu.:4113
Median :12.00	Median :36.00	Median :39.8	Median : -105.8	Median :4113
Mean :11.66	Mean :34.38	Mean :39.8	Mean : -105.8	Mean :4113
3rd Qu.:18.00	3rd Qu.:53.00	3rd Qu.:39.8	3rd Qu.: -105.8	3rd Qu.:4113
Max. :23.00	Max. :59.00	Max. :39.8	Max. : -105.8	Max. :4113

wind.dir	wind.dir.qc	wind.type.code	wind.sp
Min. : 10.0	Length:2117	Length:2117	Min. : 0.000
1st Qu.:250.0	Class :character	Class :character	1st Qu.: 4.100
Median :300.0	Mode :character	Mode :character	Median : 6.700
Mean :261.5			Mean : 7.245
3rd Qu.:310.0			3rd Qu.: 9.800
Max. :360.0			Max. :21.100
NA's :237			NA's :168
wind.sp.qc	ceiling.ht	ceiling.ht.qc	ceiling.ht.method
Length:2117	Min. : 30	Min. :5.000	Length:2117

Class :character	1st Qu.: 2591	1st Qu.:5.000	Class :character
Mode :character	Median :22000	Median :5.000	Mode :character
	Mean :15145	Mean :5.008	
	3rd Qu.:22000	3rd Qu.:5.000	
	Max. :22000	Max. :9.000	
	NA's :4		
sky.cond	vis.dist	vis.dist.qc	vis.var
Length:2117	Min. : 0	Length:2117	Length:2117
Class :character	1st Qu.:16093	Class :character	Class :character
Mode :character	Median :16093	Mode :character	Mode :character
	Mean :15913		
	3rd Qu.:16093		
	Max. :16093		
	NA's :683		
vis.var.qc	temp	temp.qc	dew.point
Length:2117	Min. : 1.00	Length:2117	Min. :-6.0000
Class :character	1st Qu.: 6.00	Class :character	1st Qu.: 0.0000
Mode :character	Median : 8.00	Mode :character	Median : 0.0000
	Mean : 8.13		Mean : 0.8729
	3rd Qu.:10.00		3rd Qu.: 2.0000
	Max. :15.00		Max. : 7.0000
dew.point.qc	atm.press	atm.press.qc	rh
Length:2117	Min. : NA	Min. :9	Min. :53.63
Class :character	1st Qu.: NA	1st Qu.:9	1st Qu.:58.10
Mode :character	Median : NA	Median :9	Median :61.39
	Mean :NaN	Mean :9	Mean :60.62
	3rd Qu.: NA	3rd Qu.:9	3rd Qu.:61.85
	Max. : NA	Max. :9	Max. :70.01
	NA's :2117		

Note that to find the maximum elevation, we had to add `na.rm = TRUE`, because the elevation variable contains missing values. This is an example of how missing values can quickly propagate throughout an analysis (as the “maximum” of 1, 2, and NA is NA, because it cannot be defined).

Also note that we used the `which` function to tell us which elements of the logical comparison are TRUE. We did this because some of them were NA, which can lead to issues when subsetting by a logical variable.

```
cor(elev$temp, elev$wind.sp, use="complete")
```

```
[1] -0.09373843
```

```
cor(elev$temp, elev$hour, use="complete")
```

```
[1] 0.4397261
```

```
cor(elev$wind.sp, elev$day, use="complete")
```

```
[1] 0.3643079
```

```
cor(elev$wind.sp, elev$hour, use="complete")
```

```
[1] 0.08807315
```

```
cor(elev$temp, elev$day, use="complete")
```

```
[1] -0.003857766
```

The `use="complete"` argument is another thing we added to avoid compounding NAs.

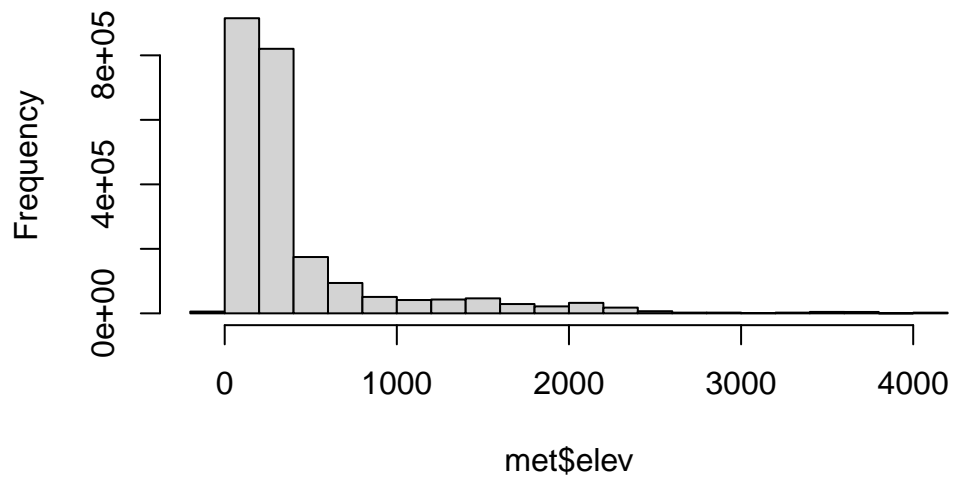
7. Exploratory graphs

We should look at the distributions of all of the key variables to make sure there are no remaining issues with the data.

Use the `hist` function to make histograms of the elevation, temperature, and wind speed variables for the whole dataset

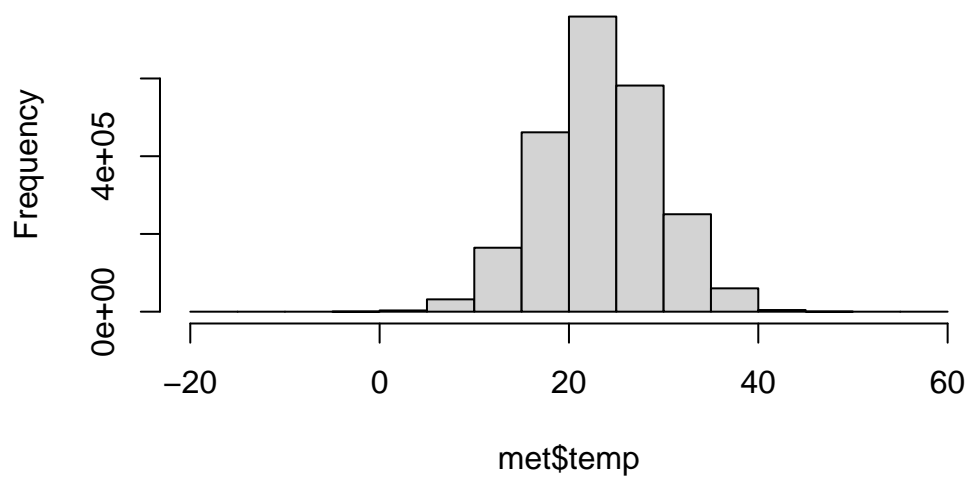
```
hist(met$elev)
```

Histogram of met\$elev

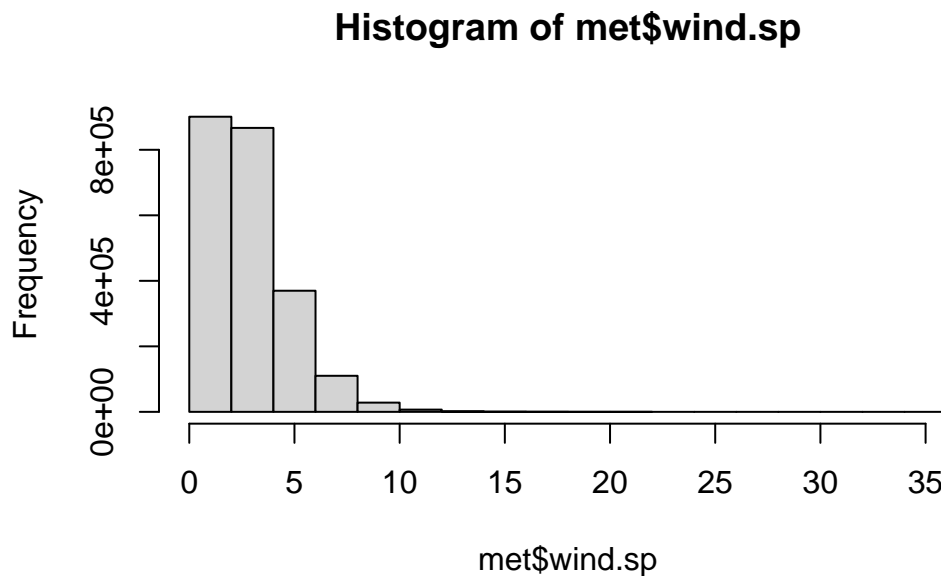


```
hist(met$temp)
```

Histogram of met\$temp



```
hist(met$wind.sp)
```



One thing we should consider for later analyses is to log transform wind speed and elevation as they are very skewed.

Look at where the weather station with highest elevation is located.

```
# leaflet(elev) |>
# addProviderTiles('OpenStreetMap') |>
# addCircles(lat=~lat,lng=~lon, opacity=1, fillOpacity=1, radius=100)
```

Look at the time series of temperature and wind speed at this location. For this we will need to create a date-time variable for the x-axis.

```
library(lubridate)
elev$date <- with(elev, ymd_h(paste(year, month, day, hour, sep= ' ')))
summary(elev$date)
```

	Min.	1st Qu.
	"2019-08-01 00:00:00.0000"	"2019-08-08 11:00:00.0000"
	Median	Mean
	"2019-08-16 22:00:00.0000"	"2019-08-16 14:09:56.8823"

3rd Qu. Max.
 "2019-08-24 11:00:00.0000" "2019-08-31 22:00:00.0000"

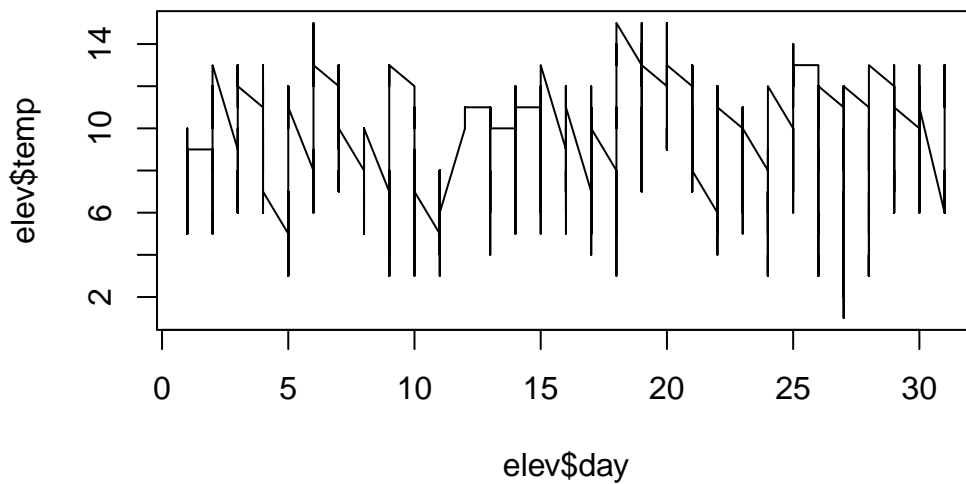
```
elev <- elev[order(elev$date), ]
head(elev)
```

	USAFID	WBAN	year	month	day	hour	min	lat	lon	elev	wind.dir
221697	720385	419	2019	8	1	0	36	39.8	-105.766	4113	170
221698	720385	419	2019	8	1	0	54	39.8	-105.766	4113	100
221699	720385	419	2019	8	1	1	12	39.8	-105.766	4113	90
221700	720385	419	2019	8	1	1	35	39.8	-105.766	4113	110
221701	720385	419	2019	8	1	1	53	39.8	-105.766	4113	120
221703	720385	419	2019	8	1	2	36	39.8	-105.766	4113	110
	wind.dir.qc	wind.type.code	wind.sp	wind.sp.qc	ceiling.ht	ceiling.ht.qc					
221697	5		N	8.8	5	1372					5
221698	5		N	2.6	5	1372					5
221699	5		N	3.1	5	1981					5
221700	5		N	4.1	5	2134					5
221701	5		N	4.6	5	2134					5
221703	5		N	6.2	5	22000					5
	ceiling.ht.method	sky.cond	vis.dist	vis.dist.qc	vis.var	vis.var.qc	temp				
221697		M	N	NA	9	N	5	9			
221698		M	N	NA	9	N	5	9			
221699		M	N	NA	9	N	5	9			
221700		M	N	NA	9	N	5	9			
221701		M	N	NA	9	N	5	9			
221703		9	N	NA	9	N	5	8			
	temp.qc	dew.point	dew.point.qc	atm.press	atm.press.qc	rh					
221697	5	1	5	NA	9	57.61039					
221698	5	1	5	NA	9	57.61039					
221699	5	2	5	NA	9	61.85243					
221700	5	2	5	NA	9	61.85243					
221701	5	2	5	NA	9	61.85243					
221703	5	1	5	NA	9	61.62158					
	date										
221697	2019-08-01	00:00:00									
221698	2019-08-01	00:00:00									
221699	2019-08-01	01:00:00									
221700	2019-08-01	01:00:00									
221701	2019-08-01	01:00:00									
221703	2019-08-01	02:00:00									

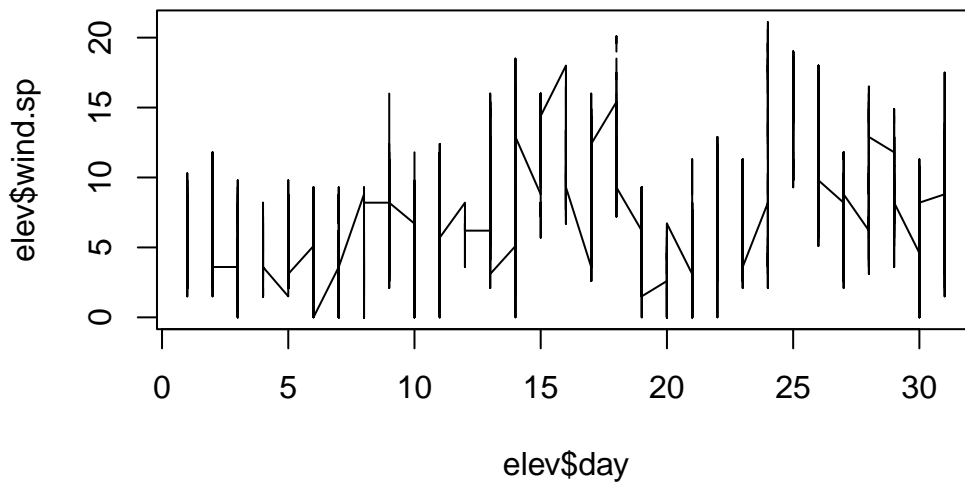
With the date-time variable we can plot the time series of temperature and wind speed.

Use the `plot` function to make line graphs of temperature vs. date and wind speed vs. date

```
plot(elev$day,elev$temp, type ="l")
```



```
plot(elev$day,elev$wind.sp, type = "l")
```



Summarize any trends that you see in these time series plots.

```
# Both temperature and wind speed slightly increases throughout the month
```

8. Ask questions

By now, you might have some specific questions about how the data was gathered and what some of the different variables and values mean. Alternatively, maybe you have an idea for how some of the variable should be related and you want to explore that relationship. In a real-world analysis, these questions could potentially be answered by a collaborator, who may have been part of the team that collected the data.

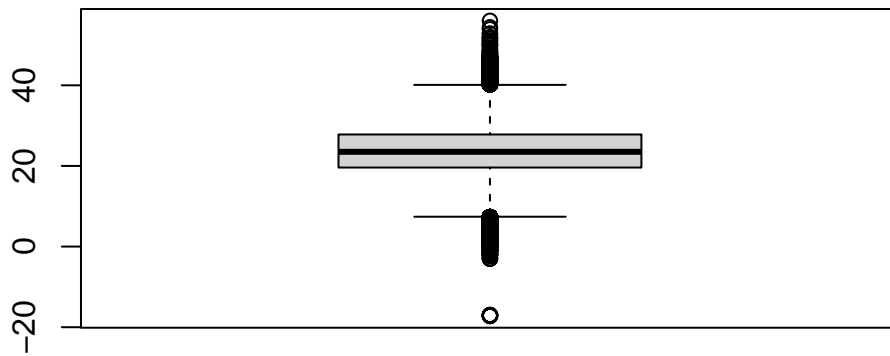
What questions do you have about the data?

```
# What are the variables that end in .qc? What do they mean?
```

If you haven't already, now would be a good time to look at the accompanying [data dictionary](#) for this dataset and see if it can answer any of your questions. If you have questions about the nature of the dataset and how it was gathered, this might be able to help.

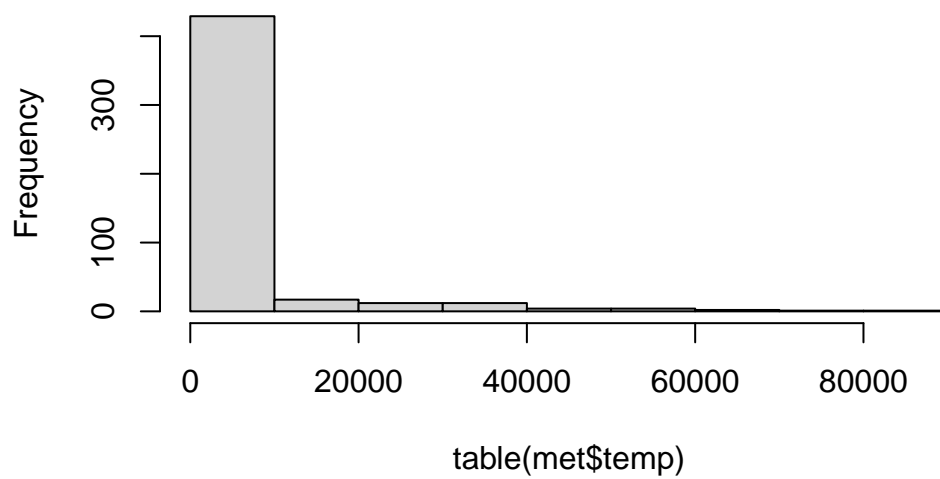
For questions about variables in the dataset or relationships between them, try making some more exploratory plots. Do you see the patterns you would expect?

```
boxplot(met$temp)
```

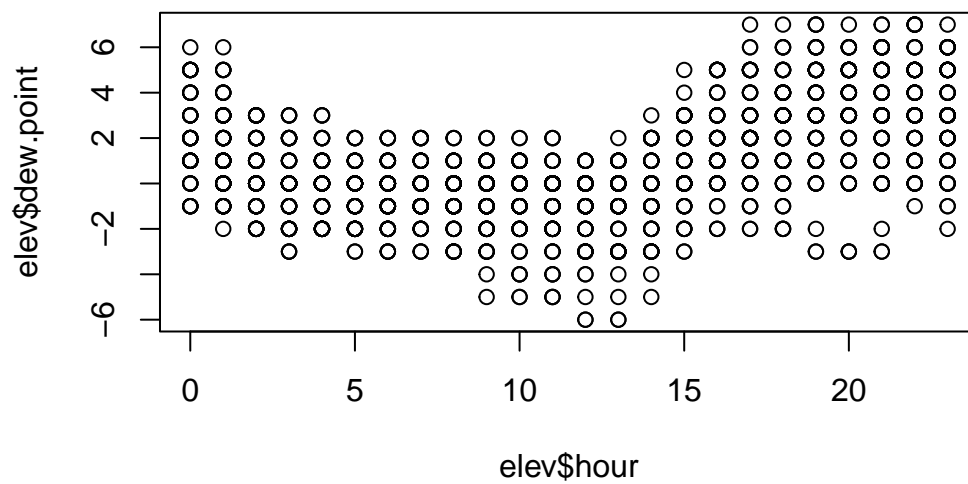


```
hist(table(met$temp))
```

Histogram of table(met\$temp)



```
plot(elev$hour,elev$dew.point)
```



```
# Dew point initially decreases then increases throughout the day, as expected
```

There are many different types of summaries and visualization strategies that we have not discussed, but which could provide interesting perspectives on the data.

Some other useful plotting functions include: - **pairs** for making all pairwise scatter plots in a dataset with >2 dimensions. - **heatmap** and/or **corrplot** (from the **corrplot** package) for visualizing matrices in general or correlation matrices in particular. - **image** a low-level matrix visualization function - **barplot**, especially with **table**, for visualizing frequencies of categorical variables.