

# Visualization of Missing Values and the Effect of Different Weather Variables

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```
setwd('/Volumes/GoogleDrive/Mon disque/wrangling/project/wranglinghub')
source('functions_library/functions_library.R')
```

```
## Loading required package: Matrix
```

```
## Loading required package: carData
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr  0.3.5
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.4.1
## v readr   2.1.3      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x tidyr::expand() masks Matrix::expand()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
## x tidyr::pack()   masks Matrix::pack()
## x dplyr::recode() masks car::recode()
## x purrr::some()   masks car::some()
## x tidyr::unpack() masks Matrix::unpack()
```

## Loading the data

```
library(tidyverse)
setwd('/Volumes/GoogleDrive/Mon disque/wrangling/project/wranglinghub')

data <- read_csv('football_data.csv')
```

```
## Warning: One or more parsing issues, call 'problems()' on your data frame for details,
## e.g.:
##   dat <- vroom(...)
##   problems(dat)
```

```
## Rows: 13504 Columns: 36
```

```
## -- Column specification -----
## Delimiter: ","
```

```
## chr (17): schedule_date, schedule_week, team_home, Home team abbrev, team_aw...
## dbl (17): schedule_season, score_home, score_away, spread_favorite, over_und...
## lgl (2): schedule_playoff, stadium_neutral
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
head(data)
```

```
## # A tibble: 6 x 36
##   schedule_date schedu~1 sched~2 sched~3 team_~4 Home ~5 score~6 score~7 team_~8
##   <chr>           <dbl> <chr>   <lgl>   <chr>   <chr>       <dbl>   <dbl> <chr>
## 1 9/2/1966        1966 1      FALSE   Miami ~ MIA      14      23 Oaklan~
## 2 9/3/1966        1966 1      FALSE   Housto~ TEN       45       7 Denver~
## 3 9/4/1966        1966 1      FALSE   San Di~ LAC       27       7 Buffal~
## 4 9/9/1966        1966 2      FALSE   Miami ~ MIA      14      19 New Yo~
## 5 9/10/1966       1966 1      FALSE   Green ~ GB       24       3 Baltim~
## 6 9/10/1966       1966 2      FALSE   Housto~ TEN       31       0 Oaklan~
## # ... with 27 more variables: 'Away team abbrev' <chr>, team_favorite_id <chr>,
## #   spread_favorite <dbl>, over_under_line <dbl>, stadium <chr>,
## #   stadium_neutral <lgl>, weather_temperature <dbl>, weather_wind_mph <dbl>,
## #   weather_humidity <dbl>, weather_detail <chr>,
## #   Difference_favored_minus_notfavored <dbl>, 'Abs value of spread' <dbl>,
## #   'Actual difference - spread' <dbl>, stadium_location <chr>,
## #   stadium_open <dbl>, stadium_close <dbl>, stadium_type <chr>, ...
```

## MISSING DATA

```
summary(data)
```

```
##   schedule_date      schedule_season schedule_week      schedule_playoff
##   Length:13504      Min.      :1966      Length:13504      Mode :logical
##   Class :character  1st Qu.:1983      Class :character  FALSE:12957
##   Mode  :character  Median :1997      Mode  :character  TRUE :547
##                               Mean      :1996
##                               3rd Qu.:2010
##                               Max.      :2022
##
##   team_home          Home team abbrev      score_home      score_away
##   Length:13504      Length:13504      Min.      : 0.00   Min.      : 0.00
##   Class :character  Class :character  1st Qu.:15.00   1st Qu.:13.00
##   Mode  :character  Mode  :character  Median :22.00   Median :20.00
##                               Mean      :22.45   Mean      :19.76
##                               3rd Qu.:29.00   3rd Qu.:27.00
##                               Max.      :72.00   Max.      :62.00
##                               NA's      :255     NA's      :255
##   team_away          Away team abbrev      team_favorite_id  spread_favorite
##   Length:13504      Length:13504      Length:13504      Min.      : -26.500
##   Class :character  Class :character  Class :character  1st Qu.: -7.000
##   Mode  :character  Mode  :character  Mode  :character  Median : -4.500
##                               Mean      : -5.394
```

```

##                                     3rd Qu.: -3.000
##                                     Max.    :  0.000
##                                     NA's    :2735
## over_under_line stadium stadium_neutral weather_temperature
## Min.    :28.00 Length:13504 Mode :logical Min.    :-6.00
## 1st Qu.:38.50 Class :character FALSE:13396 1st Qu.:48.00
## Median :42.00 Mode  :character TRUE :108 Median :62.00
## Mean    :42.22 Mean    :58.87
## 3rd Qu.:45.00 3rd Qu.:72.00
## Max.    :63.50 Max.    :97.00
## NA's    :2807 NA's    :1224
## weather_wind_mph weather_humidity weather_detail
## Min.    : 0.000 Min.    : 4.00 Length:13504
## 1st Qu.: 3.000 1st Qu.: 57.00 Class :character
## Median : 8.000 Median : 69.00 Mode  :character
## Mean    : 7.689 Mean    : 67.22
## 3rd Qu.:11.000 3rd Qu.: 79.00
## Max.    :40.000 Max.    :100.00
## NA's    :1240 NA's    :5063
## Difference_favored_minus_notfavored Abs value of spread
## Min.    :-45.00 Min.    : 0.000
## 1st Qu.: -3.00 1st Qu.: 3.000
## Median : 4.00 Median : 4.500
## Mean    : 5.26 Mean    : 5.394
## 3rd Qu.: 14.00 3rd Qu.: 7.000
## Max.    : 59.00 Max.    :26.500
## NA's    :2735 NA's    :2735
## Actual difference - spread stadium_location stadium_open stadium_close
## Min.    :-52.0000 Length:13504 Min.    :1909 Min.    :1970
## 1st Qu.: -8.5000 Class :character 1st Qu.:1966 1st Qu.:1996
## Median : -0.5000 Mode  :character Median :1975 Median :2001
## Mean    : -0.1342 Mean    :1978 Mean    :2003
## 3rd Qu.: 8.5000 3rd Qu.:1998 3rd Qu.:2009
## Max.    : 50.5000 Max.    :2020 Max.    :2016
## NA's    :2735 NA's    :554 NA's    :7231
## stadium_type stadium_address stadium_weather_station_code
## Length:13504 Length:13504 Length:13504
## Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character
##
##
##
## stadium_weather_type stadium_capacity stadium_surface STATION
## Length:13504 Min.    :27000 Length:13504 Length:13504
## Class :character 1st Qu.:65890 Class :character Class :character
## Mode  :character Median :71250 Mode  :character Mode  :character
## Mean    :71100
## 3rd Qu.:76416
## Max.    :93605
## NA's    :6307
## NAME LATITUDE LONGITUDE ELEVATION
## Length:13504 Min.    :25.79 Min.    : -122.41 Min.    : 1.8
## Class :character 1st Qu.:33.94 1st Qu.: -97.07 1st Qu.: 8.8

```

```
## Mode :character Median :39.10 Median : -84.52 Median : 145.4
## Mean :37.89 Mean : -90.29 Mean : 188.0
## 3rd Qu.:40.68 3rd Qu.: -78.89 3rd Qu.: 209.4
## Max. :47.65 Max. : -71.00 Max. :1611.2
## NA's :3220 NA's :3220 NA's :3220
```

Visualizing which columns have missing data. “weather\_detail” has the most missing values by far. This variable contains 8 factors: “DOME”, “DOME (Open Roof)”, “Fog”, “Rain”, “Rain, Fog”, “Snow”, “Snow, Fog”, and “Snow, Freezing Rain”. All of these factors are related to poor weather, or dome-related changes. Values were not recorded for “nice” weather, which is probably contributing to the large amount of missing values. To proceed, we simply decided not to use this column in our analysis. There were many other weather-related columns available in the dataset.

```
# data_missing <- data
# library(naniar)
# gg_miss_var(data_missing)

# levels(as.factor(data_missing$weather_detail))
```

```
# library(naniar)
# vis_miss(data_missing)
```

```
library(naniar)

# mcar_test(data.frame(data)) # this runs into an error, but we can subset the data by columns and
# run an mcar test on just a portion of the data

# this confirms that the data is missing not a random
mcar_test(data.frame(data[,10:20]))
```

```
## # A tibble: 1 x 4
## statistic df p.value missing.patterns
## <dbl> <dbl> <dbl> <int>
## 1 10578. 102 0 15
```

```
# library(naniar)
#
# gg_miss_fct(x = data_missing, fct = schedule_season)
# a lot of missing data related to betting up until near 1978
# some missing data in the 2023 season, where the authors of the dataset pre-filled data for some games
```

Evidently, there is a pattern in our missing data. We have 100% missing data in the earlier football seasons. This makes sense because when football first emerged, betting was not yet available.

```
pct_miss(data)
```

```
## [1] 14.73679
```

```
betting_columns <- c("team_favorite_id",  
                     "spread_favorite",  
                     "over_under_line",  
                     "Difference_favored_minus_notfavored",  
                     "Actual difference - spread",  
                     "Abs value of spread")  
pct_miss(data[,betting_columns]) # 20.34%
```

```
## [1] 20.34212
```

```
pct_miss(data[, -which(colnames(data) %in% betting_columns)])
```

```
## [1] 13.61572
```

Looking at the data in excel, our regular season betting data begins at row 2494 (2493 if we do not include the header).

```
pct_miss(data[1:2492,betting_columns])
```

```
## [1] 99.49171
```

```
pct_miss(data[2493:13248,betting_columns])
```

```
## [1] 0.1084666
```

```
pct_miss(data[13249:nrow(data),betting_columns])
```

```
## [1] 100
```

## WEATHER

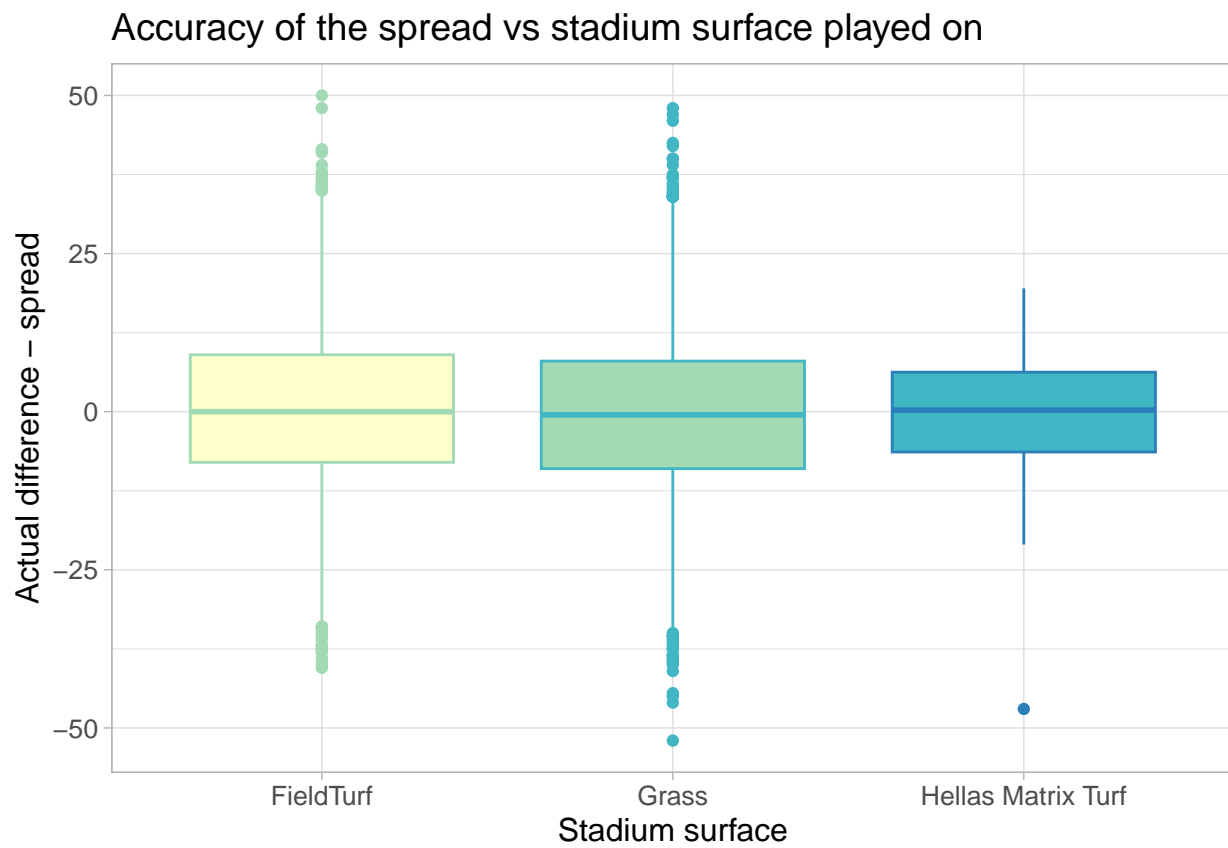
```
# weather_temperature  
# weather_wind_mph  
# weather_humidity  
  
# stadium_weather_type  
# stadium_surface  
  
# Actual difference - spread  
  
betting_data <- data[2493:13248,]
```

## Playing surface:

```
table(as.factor(betting_data$stadium_surface))
```

```
##  
##           FieldTurf           Grass Hellas Matrix Turf  
##           2327           4016           38
```

```
plot_playing_surface(betting_data)
```

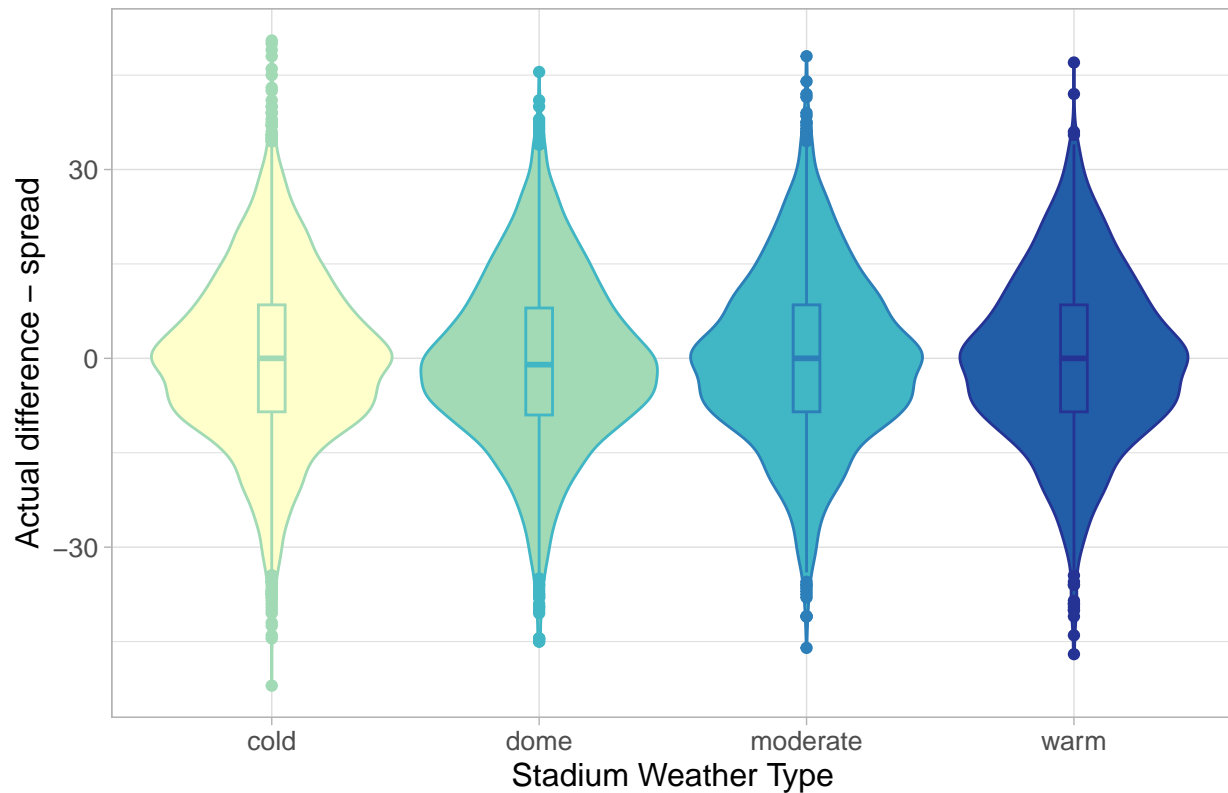


```
table(betting_data$stadium_weather_type)
```

```
##  
##      cold      dome moderate      warm  
##      4349      2584      2098      1711
```

```
plot_stadium_weather(betting_data)
```

## Accuracy of the spread vs weather type at the stadium



## Temperature

```
# weather_temperature
# weather_wind_mph
# weather_humidity
domed_stadiums <- filter(betting_data,
                          stadium_weather_type=='dome')

non_domed_stadiums <- filter(betting_data,
                              stadium_weather_type=='cold' |
                              stadium_weather_type=='moderate' |
                              stadium_weather_type=='warm')

summary(non_domed_stadiums$weather_temperature) # 50, 64, 72
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##      -6.00  45.00   57.00   55.83  67.50   97.00    799
```

```
summary(non_domed_stadiums$weather_wind_mph) # 0, 7, 11
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##      0.000   6.000   9.000   9.538  12.000   40.000    810
```

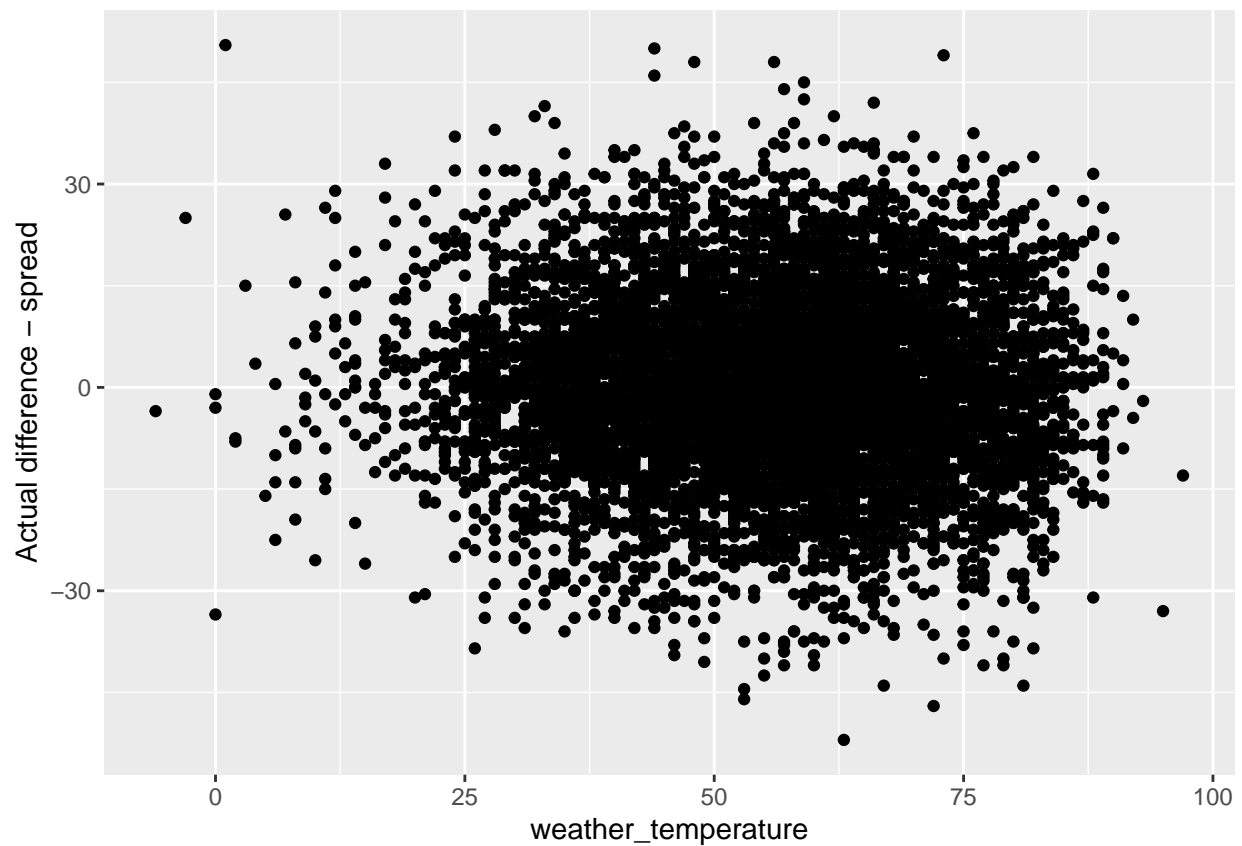
```
summary(non_domed_stadiums$weather_humidity) # 56, 68, 78
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     NA's  
##      4.00   57.00   69.00   67.03   78.00   100.00    1921
```

```
# new values (when removing domed stadiums):  
# 45, 57, 67.5  
# 6, 9, 12  
# 57, 69, 78
```

```
ggplot(non_domed_stadiums, aes(x=weather_temperature, y=`Actual difference - spread`)) +  
  geom_point()
```

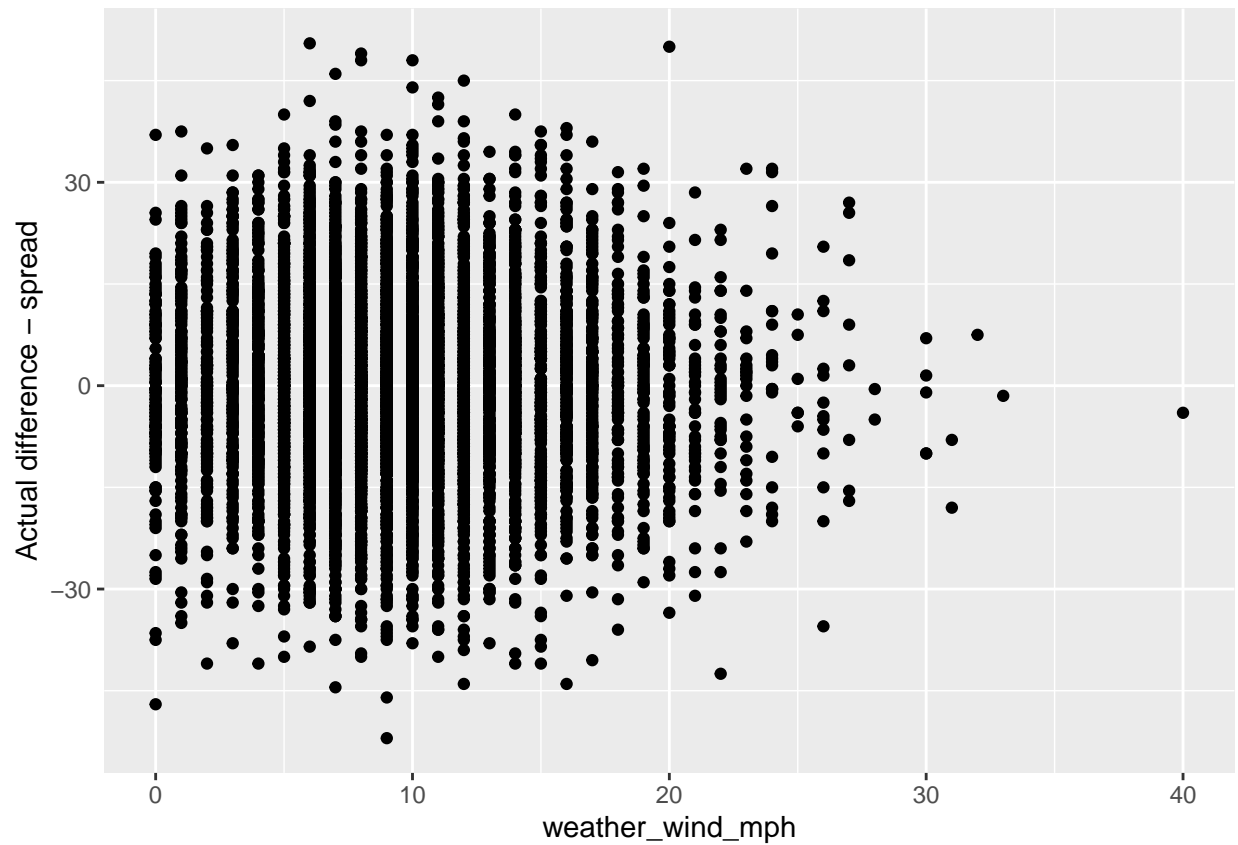
```
## Warning: Removed 799 rows containing missing values ('geom_point()').
```



```
ggplot(non_domed_stadiums, aes(x=weather_wind_mph, y=`Actual difference - spread`)) +  
  geom_point()
```

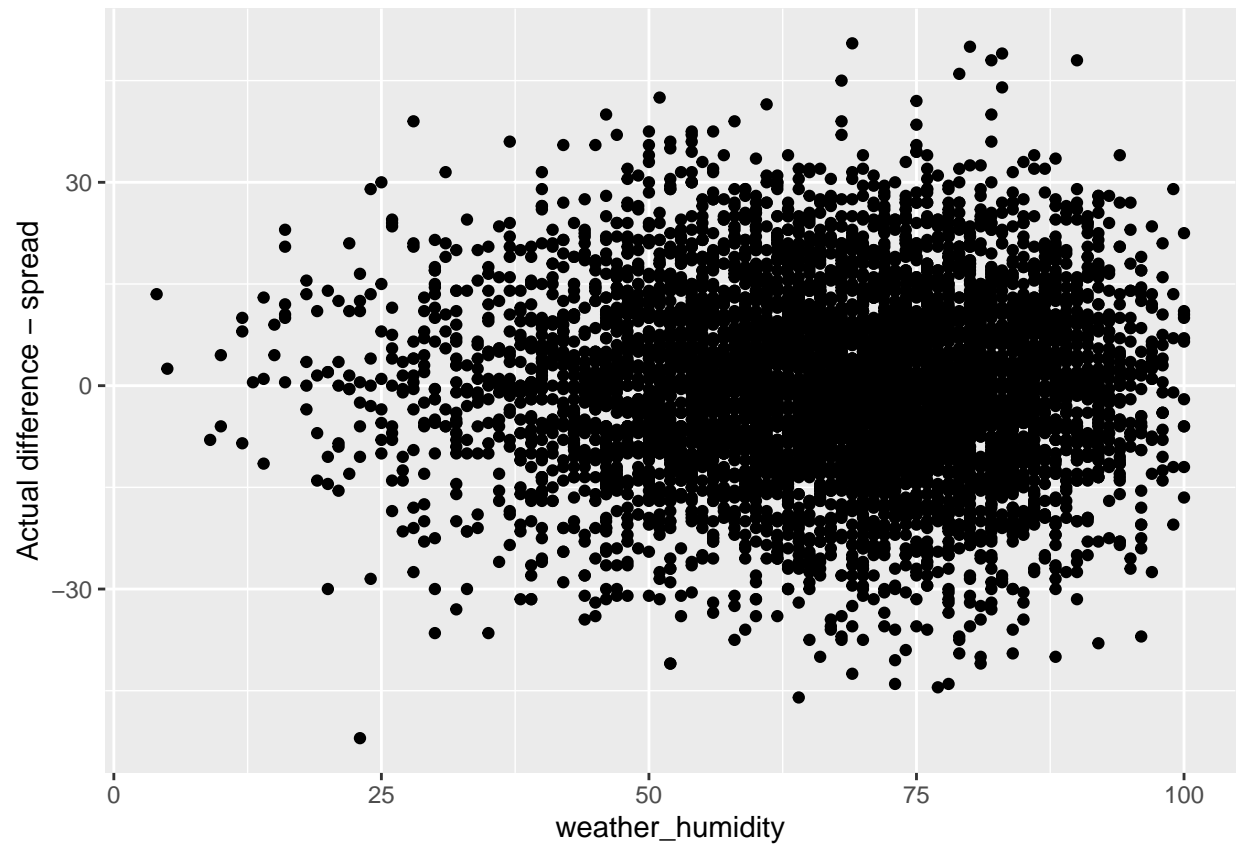
```
## Warning: Removed 810 rows containing missing values ('geom_point()').
```





```
ggplot(non_domed_stadiums, aes(x=weather_humidity, y=`Actual difference - spread`)) +  
  geom_point()
```

```
## Warning: Removed 1921 rows containing missing values ('geom_point()').
```



Categorical weather:

```
plot_weather_status(betting_data)
```

Accuracy of the spread vs weather type at the stadium

