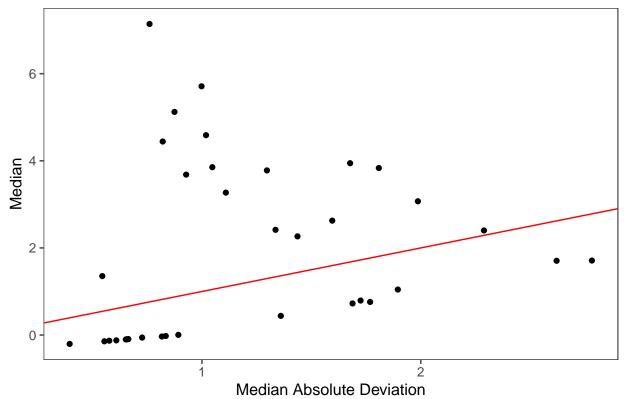
```
rm(list=ls())
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.2 --
## v ggplot2 3.3.6 v purrr 0.3.4
## v tibble 3.1.8 v dplyr 1.0.10
## v tidyr 1.2.1 v stringr 1.4.1
                   v forcats 0.5.2
## v readr
          2.1.3
## -- Conflicts -----
                                         ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(ggthemes)
data <- read_csv("cytof_one_experiment.csv")</pre>
## Rows: 50000 Columns: 35
## -- Column specification -----
## Delimiter: ","
## dbl (35): NKp30, KIR3DL1, NKp44, KIR2DL1, GranzymeB, CXCR6, CD161, KIR2DS4, ...
## 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.
## 1.)
data <- data %>%
         pivot_longer(names(data), names_to="Protein_Identity", values_to="Protein_Level")
data
## # A tibble: 1,750,000 x 2
## Protein_Identity Protein_Level
##
     <chr>>
                             <dbl>
## 1 NKp30
                             0.188
## 2 KIR3DL1
                             3.62
## 3 NKp44
                            -0.561
## 4 KIR2DL1
                            -0.294
## 5 GranzymeB
                             2.48
## 6 CXCR6
                            -0.145
## 7 CD161
                            -0.315
## 8 KIR2DS4
                             1.94
## 9 NKp46
                             4.08
                             2.62
## 10 NKG2D
## # ... with 1,749,990 more rows
nrow(data) == 1750000
## [1] TRUE
## 2.)
median_and_mad <- data %>%
                   group_by(Protein_Identity) %>%
                   summarise(Median=median(Protein_Level), MAD=mad(Protein_Level))
median and mad
```

```
## # A tibble: 35 x 3
##
      Protein_Identity Median
                                 MAD
                         <dbl> <dbl>
##
##
    1 CD107a
                       -0.122 0.609
    2 CD16
##
                        5.12
                                0.874
##
    3 CD161
                        0.726 1.69
    4 CD2
                        3.95
                                1.68
    5 CD4
                               0.395
                       -0.204
##
##
    6 CD56
                        5.71
                                0.998
##
   7 CD57
                        3.07
                                1.99
    8 CD69
                        4.59
                                1.02
   9 CD8
                        2.40
                                2.29
##
## 10 CXCR6
                       -0.0581 0.727
## # ... with 25 more rows
## 3.)
ggplot(median_and_mad, aes(x=MAD, y=Median)) +
  geom_point() +
  geom_abline(slope=1, intercept=0, col="red") +
  xlab("Median Absolute Deviation") +
  ylab("Median") +
  ggtitle("Spread-Location Plot") +
  theme_few()
```

Spread-Location Plot



```
cor(median_and_mad$MAD, median_and_mad$Median)
```

[1] 0.1542416

The spread-location plot is a null plot; there is no clear association between the median absolute deviation and the median. This is noteworthy because the median absolute deviation is just a robust version of the median. The lack of a pattern indicates that the median is unstable in this dataset, and the median absolute deviation should be used as a measure of center.

```
## 4.)
library(dcldata)
data(example_gymnastics_2)
example_gymnastics_2 <- example_gymnastics_2 %>%
    pivot_longer(names(example_gymnastics_2)[-1], names_to="event", values_to="score")
example_gymnastics_2 <- example_gymnastics_2 %>%
    separate(event, into = c("event", "year"))
example_gymnastics_2 <- example_gymnastics_2 %>%
    arrange(desc(score))
example_gymnastics_2
```

```
## # A tibble: 12 x 4
##
      country
                    event year
                                score
##
      <chr>
                    <chr> <chr> <dbl>
  1 United States vault 2012
##
                                 48.1
##
   2 United States vault 2016
                                 46.9
##
  3 Russia
                    vault 2012
                                 46.4
## 4 United States floor 2016
                                 46.0
##
   5 Russia
                    vault 2016
                                 45.7
## 6 United States floor 2012
                                 45.4
## 7 China
                    vault 2016
                                 44.3
## 8 China
                    vault 2012
                                 44.3
## 9 China
                    floor 2016
                                 42.1
## 10 Russia
                    floor 2016
                                 42.0
## 11 Russia
                    floor 2012
                                 41.6
## 12 China
                    floor 2012
                                 40.8
```