## all\_graphs.R

## mwilmes

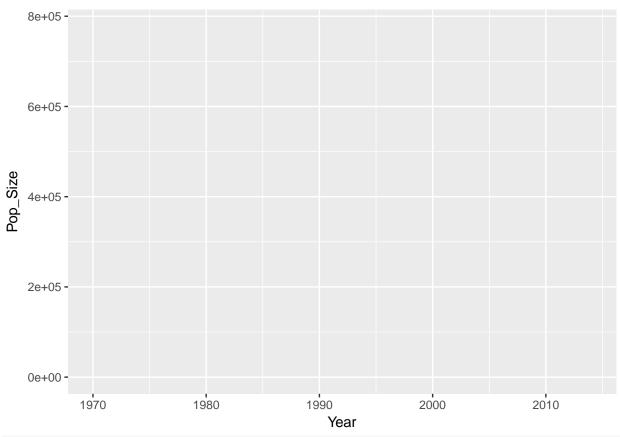
Mon Nov 6 09:11:54 2017

```
# This file contains all commands of the Chapter "Data Visualisation"
# load the library
library(tidyverse)
# read the data
popsize <- read_tsv("../data/FauchaldEtAl2017/pop_size.csv")</pre>
## Parsed with column specification:
## cols(
##
    Herd = col_character(),
    Year = col_integer(),
##
##
    Pop_Size = col_integer()
## )
ndvi <- read_tsv("../data/FauchaldEtAl2017/ndvi.csv")</pre>
## Parsed with column specification:
## cols(
##
    Herd = col_character(),
##
    Year = col_integer(),
    NDVI_May = col_double(),
##
     NDVI_June_August = col_double()
## )
seaice <- read_tsv(".../data/FauchaldEtAl2017/sea_ice.csv")</pre>
## Parsed with column specification:
## cols(
##
    Herd = col character(),
##
    Year = col_integer(),
##
     Jan = col_double(),
    Feb = col_double(),
##
##
    Mar = col_double(),
##
    Apr = col_double(),
##
    May = col double(),
##
     Jun = col_double(),
##
     Jul = col_double(),
##
     Aug = col_double(),
##
     Sep = col_double(),
##
     Oct = col_double(),
##
     Nov = col_double(),
     Dec = col_double()
##
## )
snow <- read_tsv("../data/FauchaldEtAl2017/snow.csv")</pre>
## Parsed with column specification:
## cols(
##
     Year = col_integer(),
##
     Herd = col_character(),
```

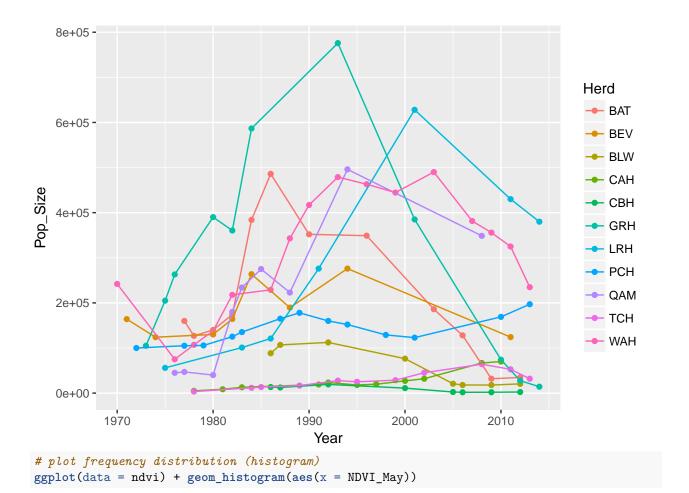
```
## Perc_snowcover = col_double(),
## Week_snowmelt = col_integer()
## )
# bring data into long format
seaice <- seaice %>% gather(Month, Cover, 3:14)

# build the first plot
ggplot(data = popsize)
```

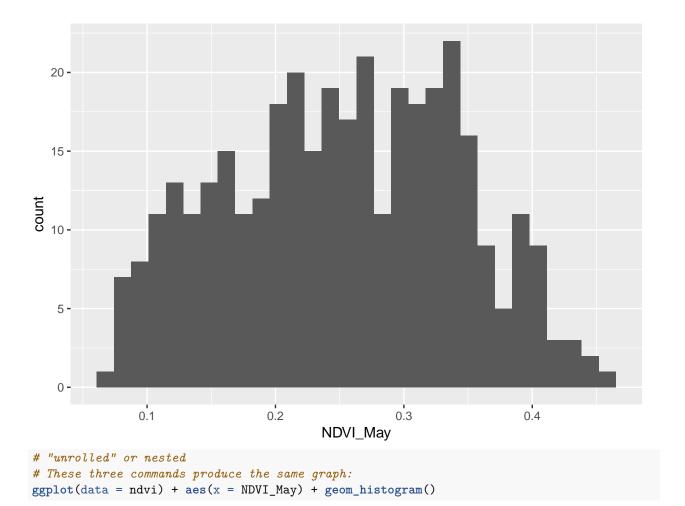
```
# add an aestethic mapping
ggplot(data = popsize) + aes(x = Year, y = Pop_Size, colour = Herd)
```



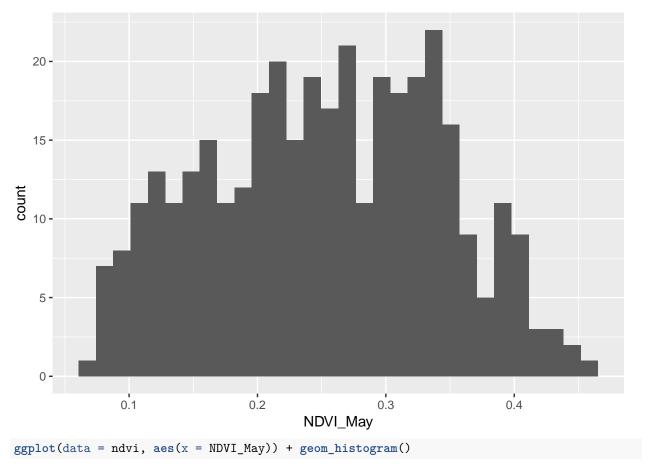
```
# add geometries
ggplot(data = popsize) +
  aes(x = Year, y = Pop_Size, colour = Herd) +
  geom_point() +
  geom_line()
```



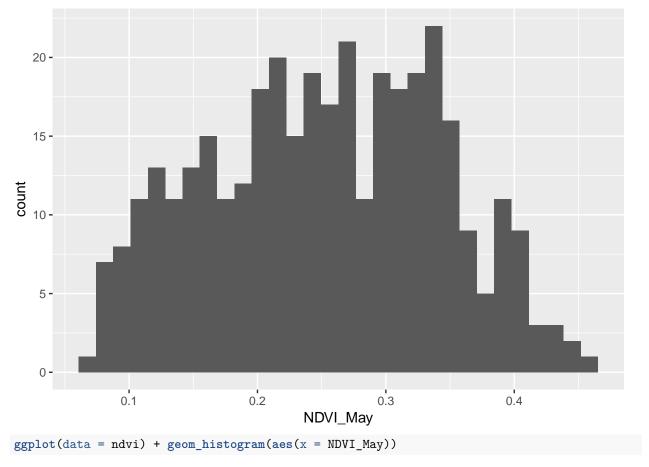
\$## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



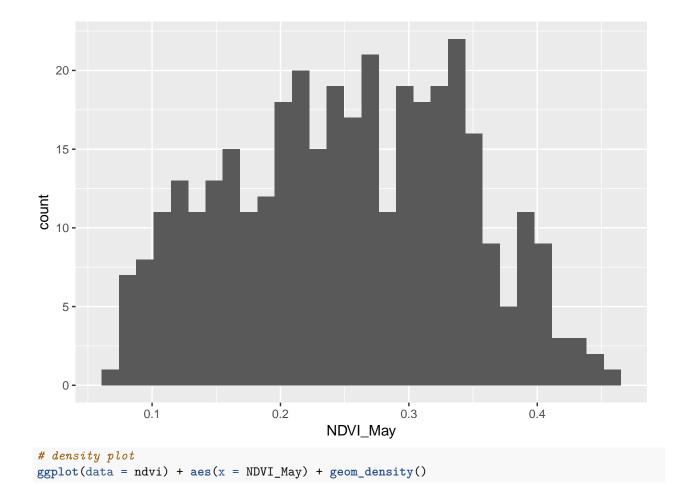
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

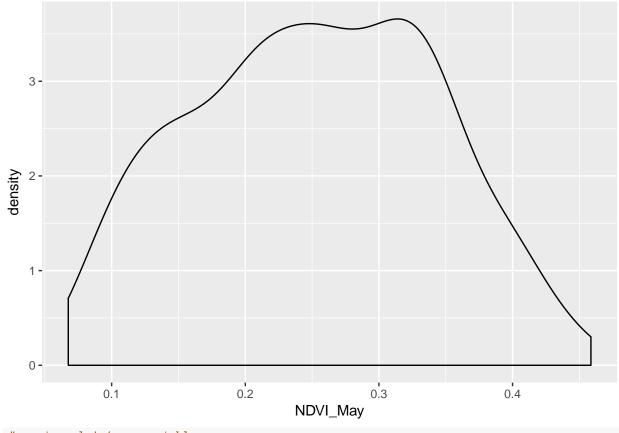


## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

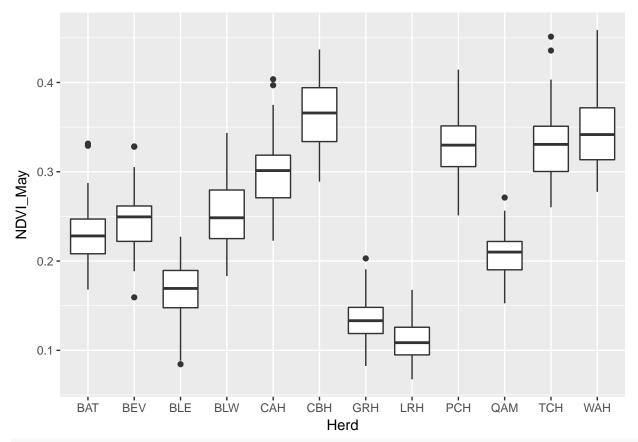


## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

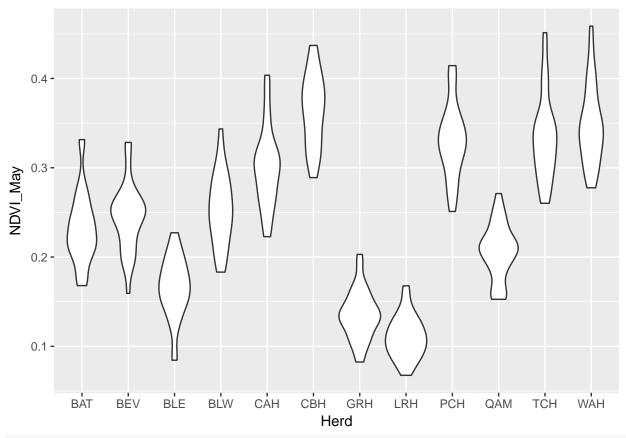




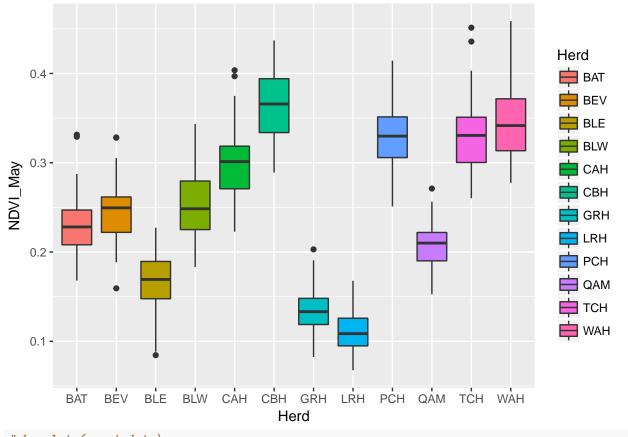
```
# assign plot to a variable
pl <- ggplot(data = ndvi) + aes(x = Herd, y = NDVI_May)
# add components to existing plot
pl + geom_boxplot()</pre>
```

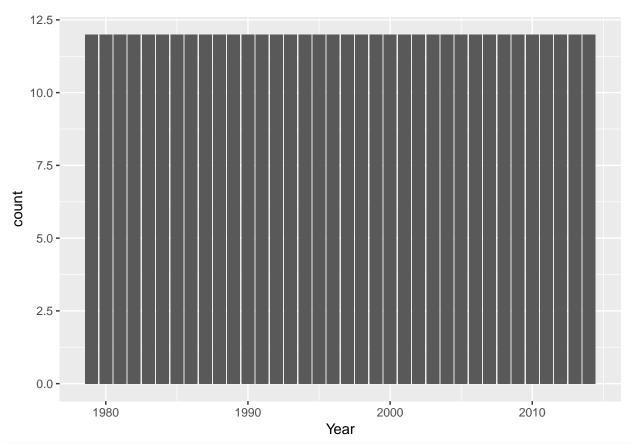


pl + geom\_violin()

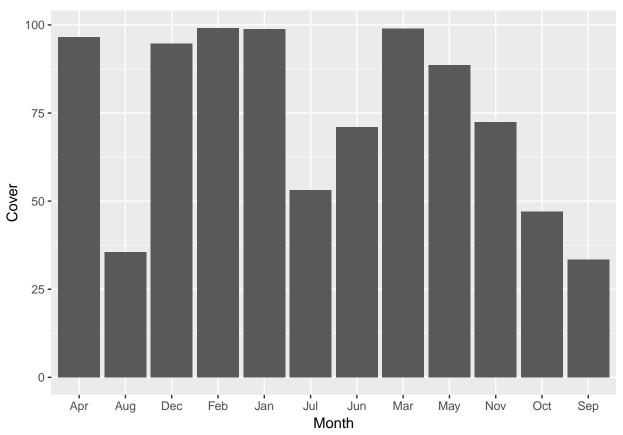


# change color of boxes
pl + geom\_boxplot() + aes(fill = Herd)

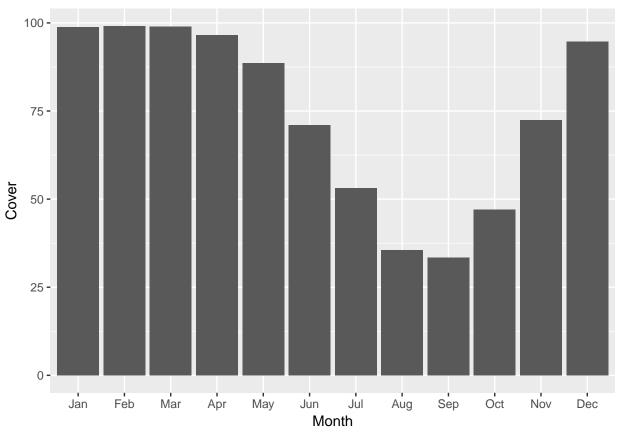




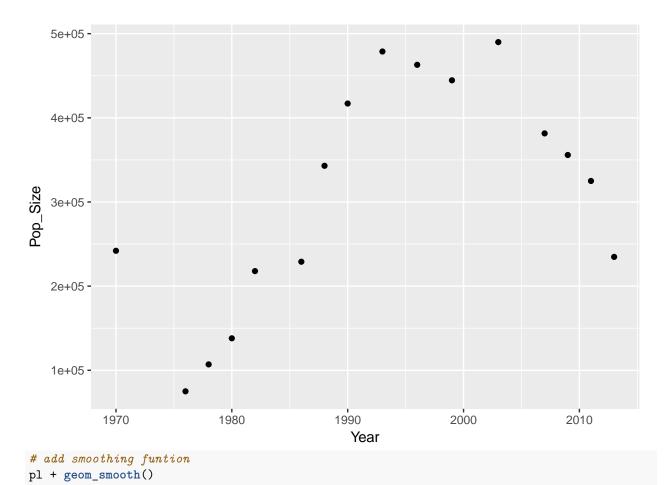
# map data to columns (note alphabetical order of x-axis)
ggplot(data = seaice %>% filter(Herd == "WAH", Year == 1990)) + aes(x = Month, y = Cover) + geom\_col()



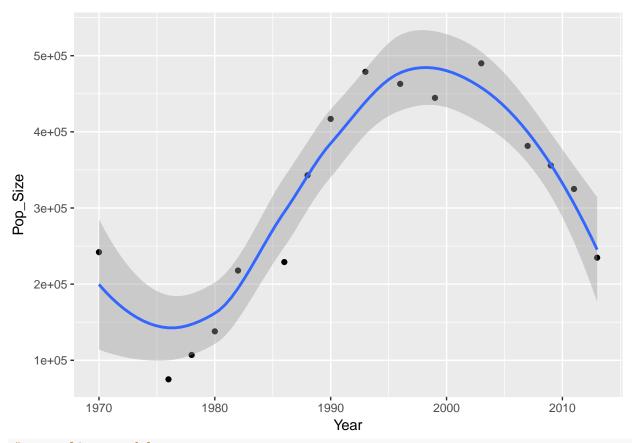
```
# display bars in chronolocigal order
# convert data into factor and set to three-letter abbreviation of months
seaice$Month <- factor(seaice$Month, month.abb)
ggplot(data = seaice %>% filter(Herd == "WAH", Year == 1990)) + aes(x = Month, y = Cover) + geom_col()
```



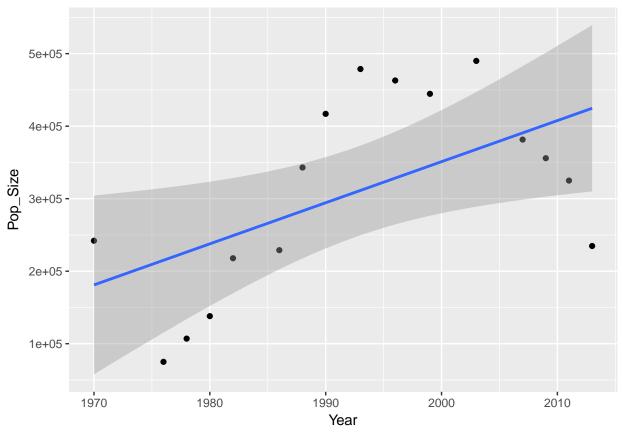
```
# scatterplots
pl <- ggplot(data = popsize %>% filter(Herd == "WAH")) + aes(x = Year, y = Pop_Size) + geom_point()
# show plot assigned to variable
show(pl)
```



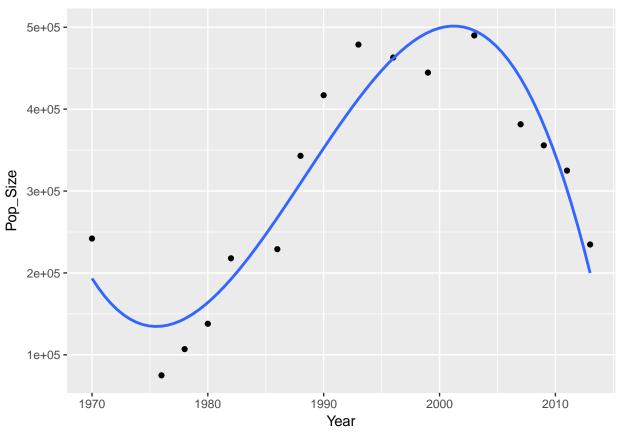
## `geom\_smooth()` using method = 'loess'



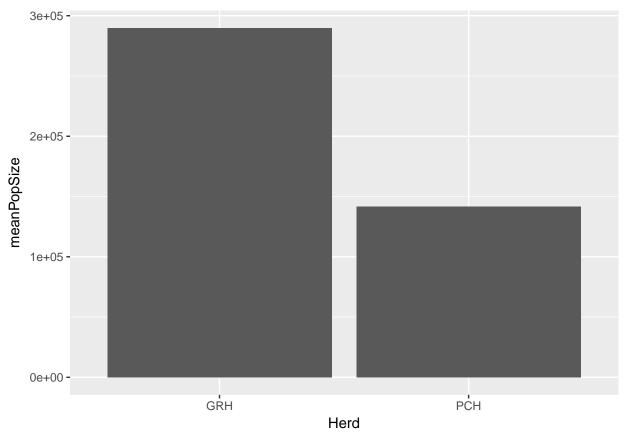
# use a linear model
pl + geom\_smooth(method = "lm")

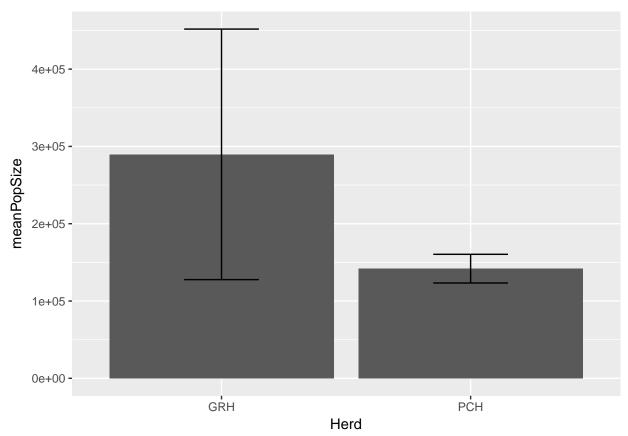


# use a polynomial regression
pl + geom\_smooth(method = "glm", formula = y ~poly(x, 3), se = FALSE)

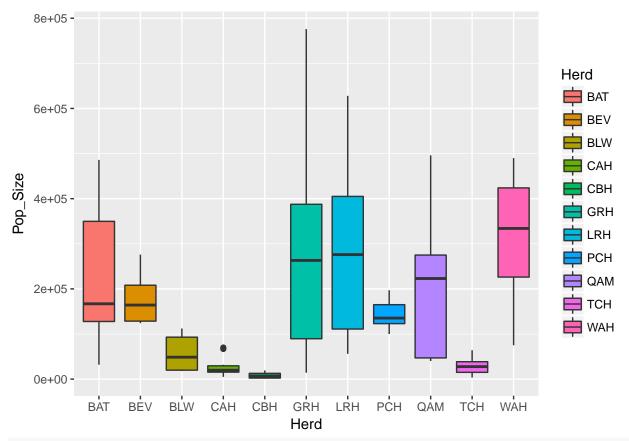


```
# calculate summary stats and errors
stats <- popsize %>% filter(Herd %in% c("GRH", "PCH")) %>%
group_by(Herd) %>%
summarise(
   meanPopSize= mean(Pop_Size),
   SD = sd(Pop_Size),
   N = n(),
   SEM = SD/sqrt(N),
   CI = SEM * qt(0.975, N-1))
# bar plot without error bars
ggplot(data = stats) +
   aes(x = Herd, y = meanPopSize) +
   geom_col()
```

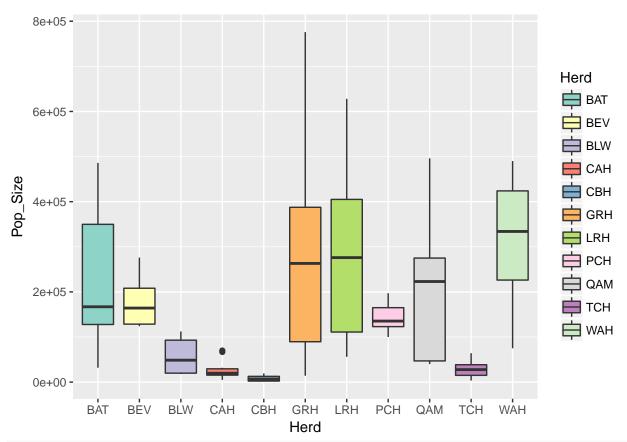




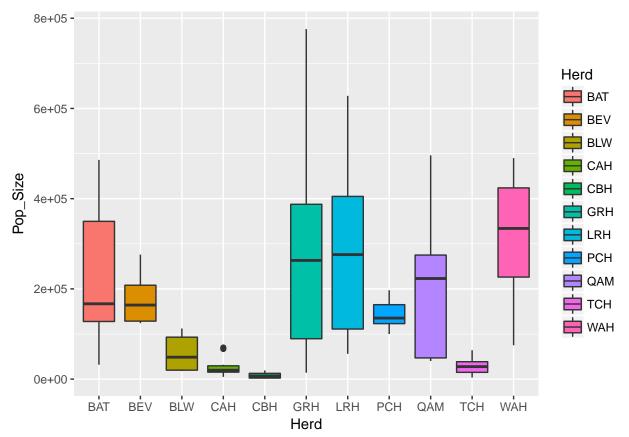
# set fill color of boxes by using scales
pl <- ggplot(data = popsize, aes(x = Herd, y = Pop\_Size, fill = Herd)) + geom\_boxplot()
show(pl)</pre>



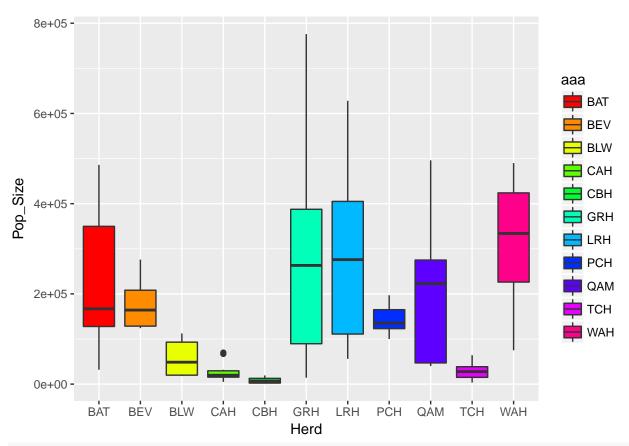
# choose a palette from Color Brewer
pl + scale\_fill\_brewer(palette = "Set3")



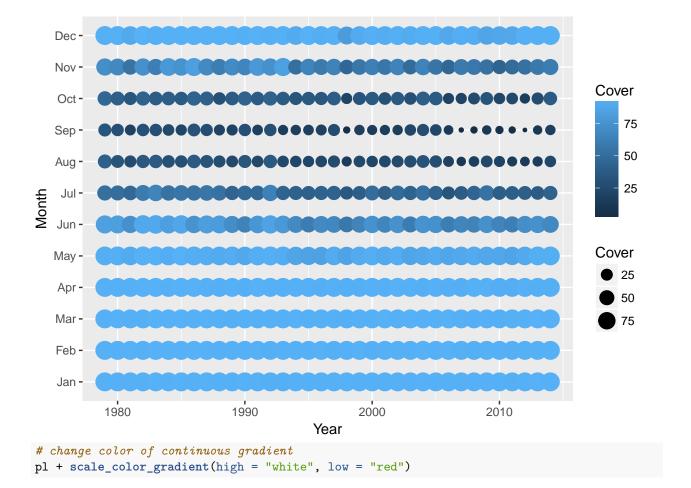
# palette based on hue
pl + scale\_fill\_hue()

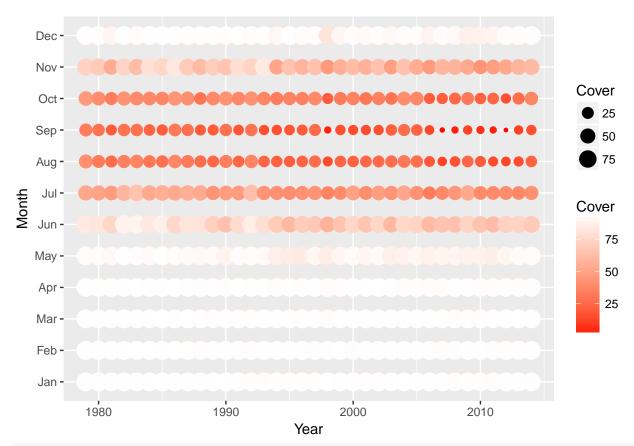


# manually set values and rename the legend
pl + scale\_fill\_manual(values = rainbow(11), name = "aaa")

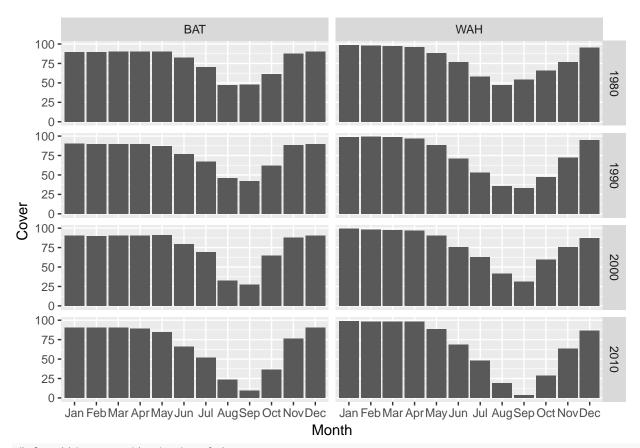


# apply scales to manipulate color and size of aestethic mappings
pl <- ggplot(data = seaice %>% filter(Herd == "BEV")) + aes(x = Year, y = Month, colour = Cover, size = show(pl)

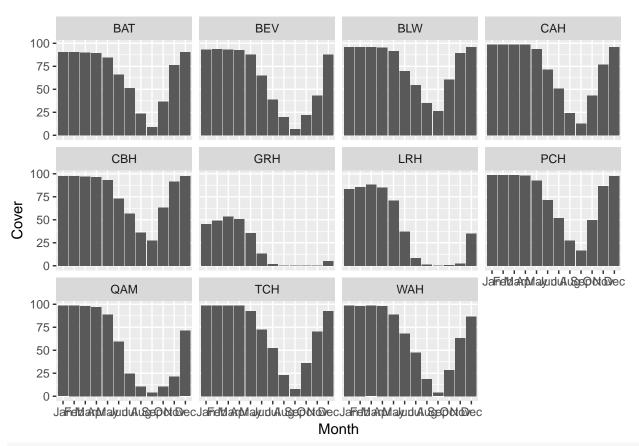




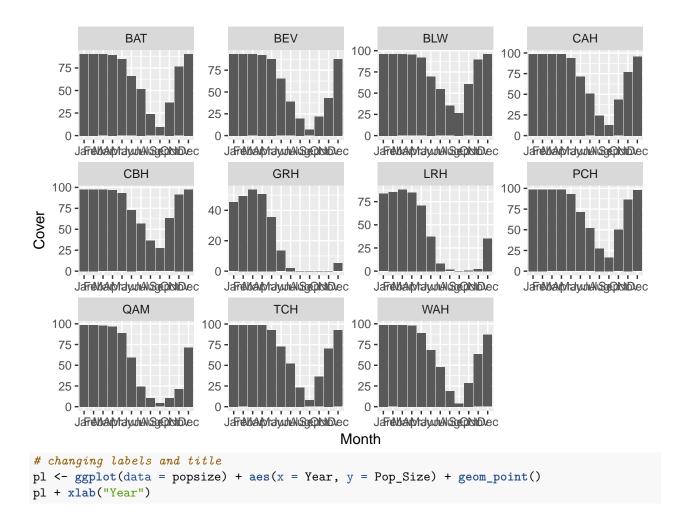
# facetting with identical scale of axis, including missing data
ggplot(data = seaice %>% filter(Herd %in% c("WAH", "BAT"), Year %in% c(1980, 1990, 2000, 2010))) + aes(

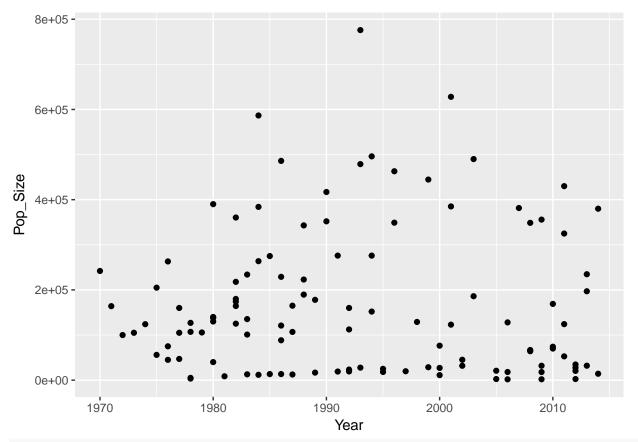


# facetting, ommit missing data
ggplot(data = seaice %% filter(Year == 2010)) + aes(x = Month, y = Cover) + geom\_col() + facet\_wrap(~H

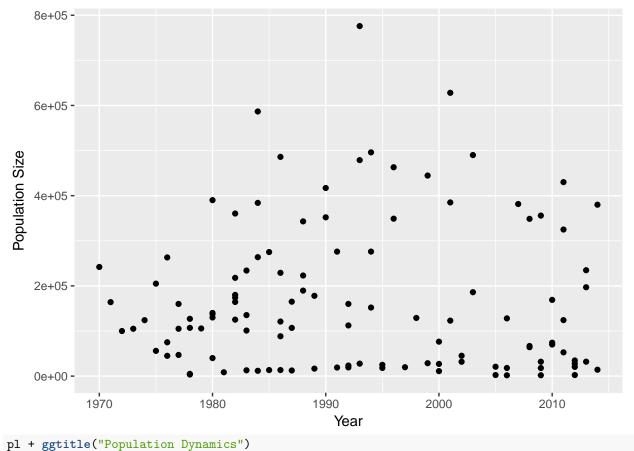


# facetting, ommit missing data, adjusted scale of axes
ggplot(data = seaice %>% filter(Year == 2010)) + aes(x = Month, y = Cover) + geom\_col() + facet\_wrap(~H



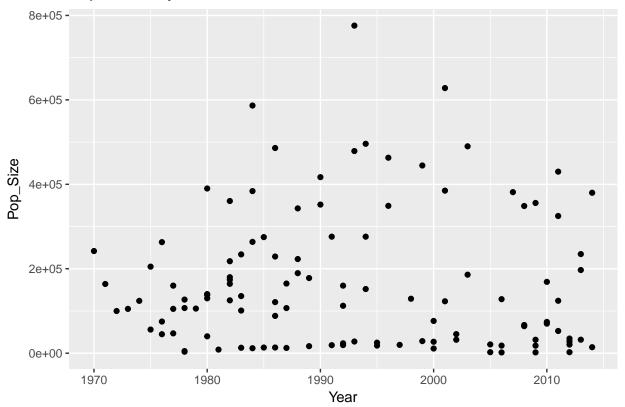


pl + ylab("Population Size")

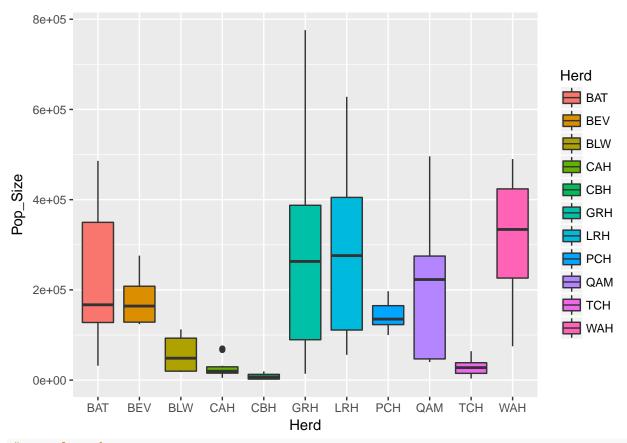


pl + ggtitle("Population Dynamics")

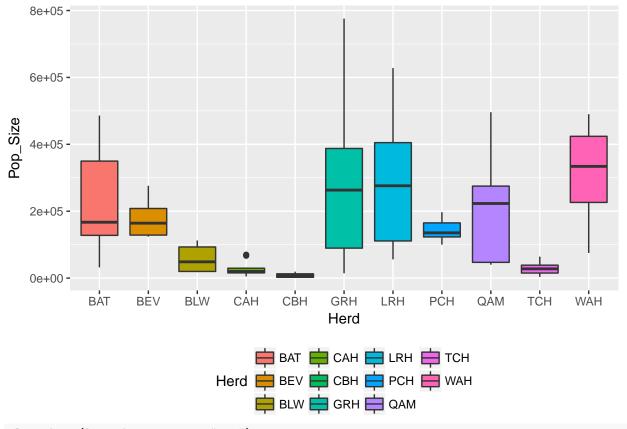
## Population Dynamics

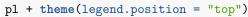


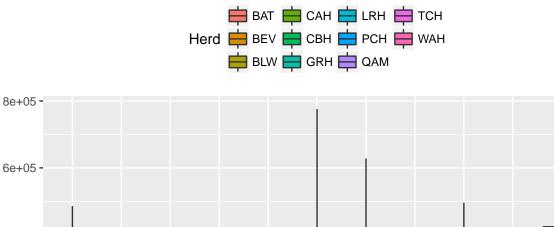
```
# legends
pl <- ggplot(data = popsize) + aes(x = Herd, y = Pop_Size, fill = Herd) + geom_boxplot()
# default
show(pl)</pre>
```

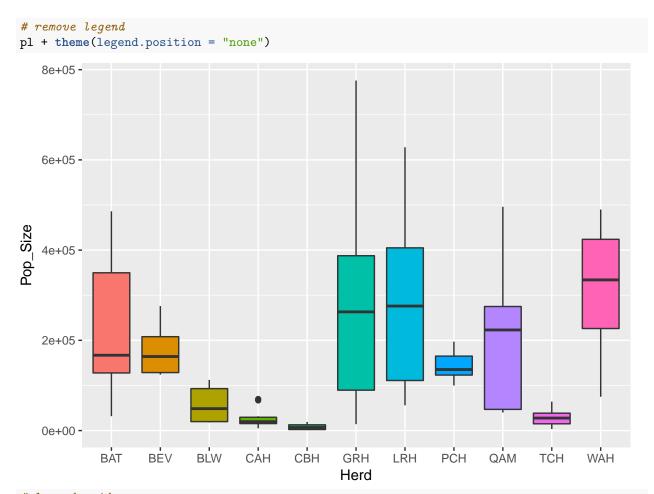


# move legend
pl + theme(legend.position = "bottom")

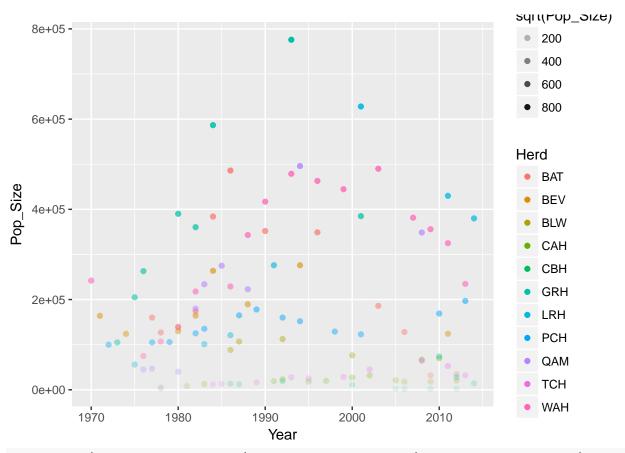




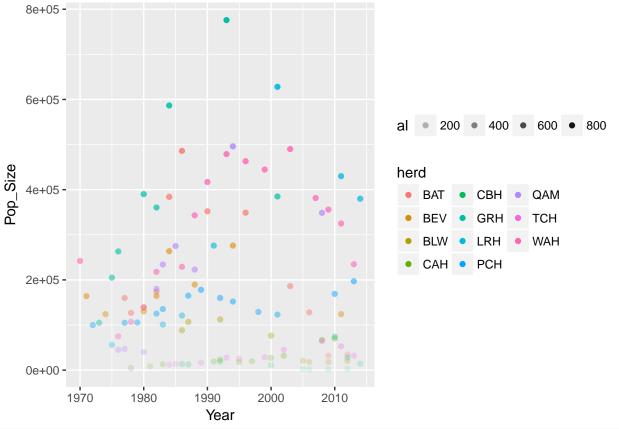




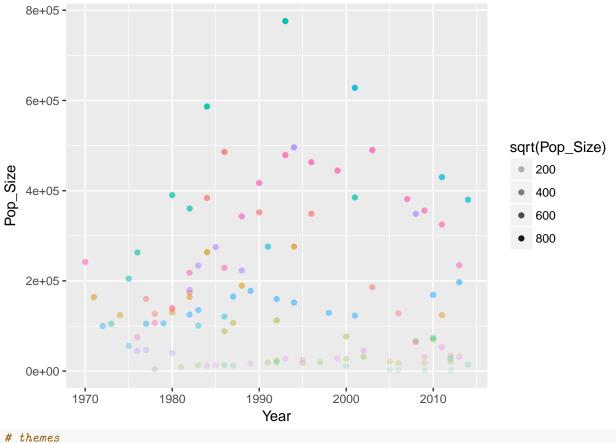
# legend guides
pl <- ggplot(data = popsize) + aes(x = Year, y = Pop\_Size, colour = Herd, alpha = sqrt(Pop\_Size)) + george
show(pl)</pre>

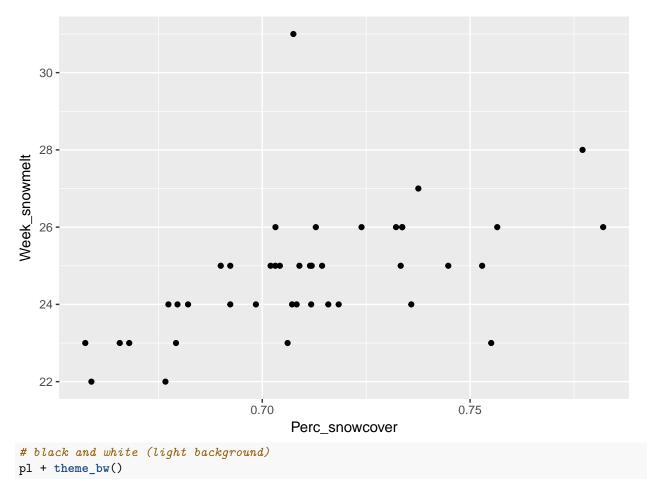


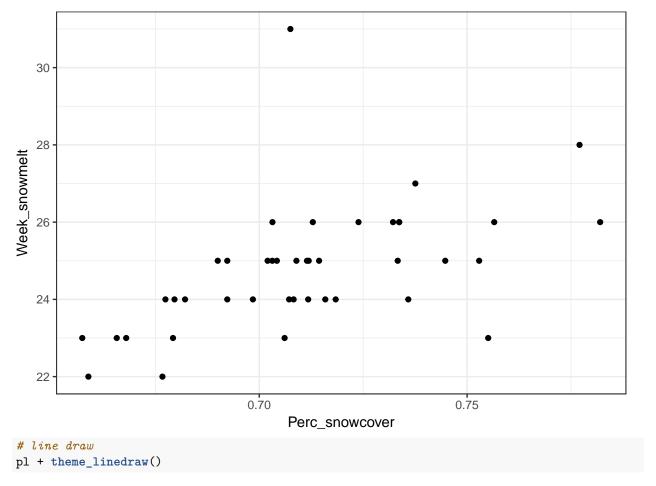
pl + guides(colour = guide\_legend(nrow = 4, title = "herd"), alpha = guide\_legend(direction = "horizont")

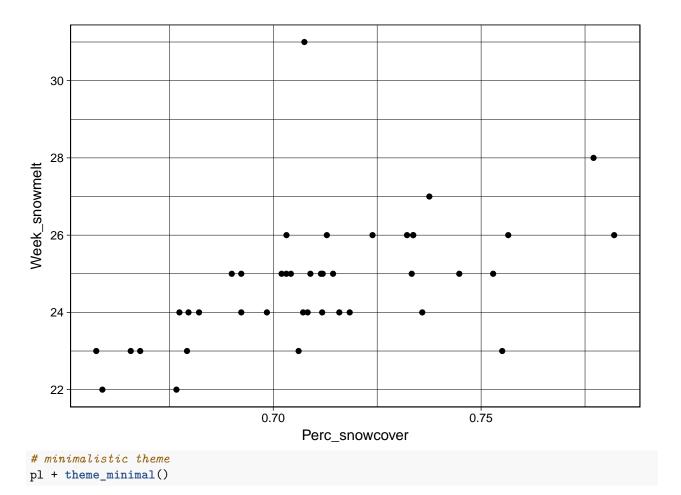


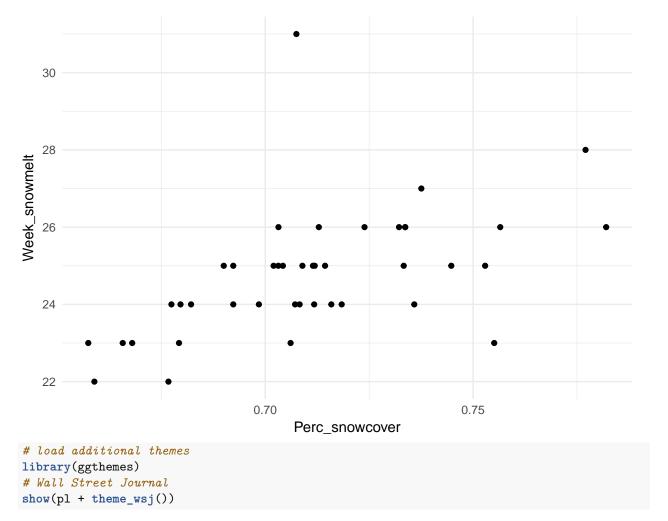
# suppress only one legend
pl + guides(colour = "none")

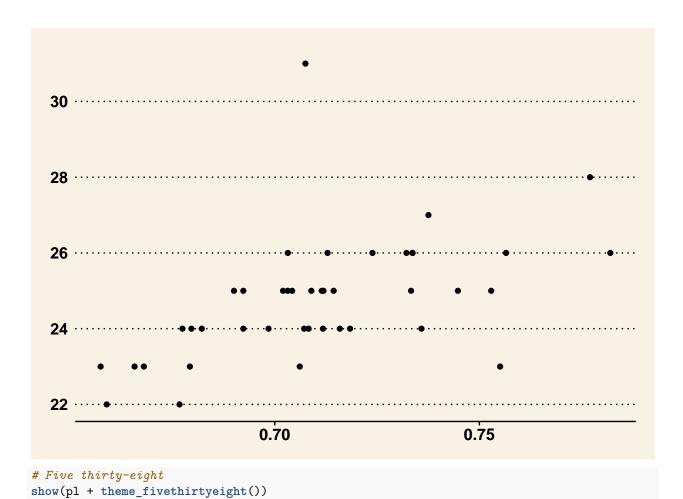


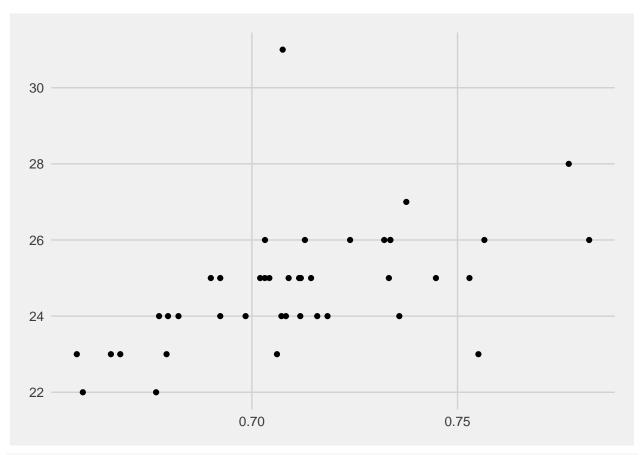








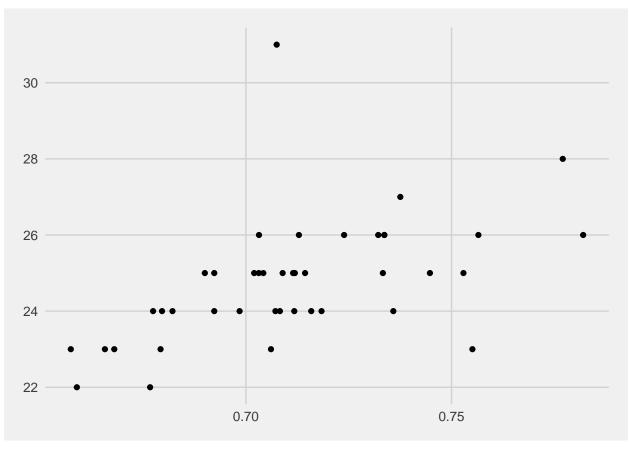




```
# setting features
# use color as an aesthetic mapping, associated with Herd
pl <- ggplot(data = popsize) + aes(x = Year, y = Pop_Size, colour = Herd) + geom_point()
# set color to be red for all points
pl <- ggplot(data = popsize) + aes(x = Year, y = Pop_Size) + geom_point(colour = "red")

# saving plot as test.pdf in the sandbox
ggsave(filename = "../sandbox/test.pdf", plot = pl, width = 3, height = 4)

# select numerical column headers, or headers with white space using back ticks
popsize %>% filter(Year > 1979, Year < 1985) %>% spread(Year, Pop_Size) %>% select(Herd, `1980`)
```



```
## # A tibble: 9 x 2
     Herd `1980`
##
## * <chr> <int>
      BAT 140000
## 1
## 2
      BEV 130000
## 3
      CAH
## 4
      GRH 390100
## 5
      LRH
               NA
## 6
      PCH
               NA
## 7
       QAM 40000
## 8
      TCH
               NA
## 9
      WAH 138000
```

## # ungroup elements popsize %>% group\_by(Herd, Year) %>% tally() %>% ungroup()

```
## # A tibble: 114 x 3
##
      Herd Year
##
     <chr> <int> <int>
##
       BAT 1977
  1
##
  2
       BAT 1978
       BAT 1980
##
   3
                    1
##
       BAT 1982
   4
##
  5
       BAT 1984
##
  6
       BAT 1986
## 7
       BAT 1990
                    1
## 8
       BAT 1996
```

```
## 9
       BAT 2003
## 10
       BAT 2006
                     1
## # ... with 104 more rows
# operation on columns vs rowwise
ndvi %>% mutate(maxndvi = max(NDVI_May, NDVI_June_August)) %>% head(4)
## # A tibble: 4 x 5
##
     Herd Year NDVI_May NDVI_June_August maxndvi
     <chr> <int>
                                             <dbl>
##
                   <dbl>
                                    <dbl>
## 1
      BAT 1982 0.21440
                                0.3722679 6.308922
## 2
           1983 0.20448
                               -0.9977483 6.308922
      BAT
## 3
           1984 0.24650
                                1.5864094 6.308922
      BAT
      BAT 1985 0.24444
                                0.6420830 6.308922
ndvi %>% rowwise() %>% mutate(maxndvi = max(NDVI_May, NDVI_June_August)) %>% head(4)
## # A tibble: 4 x 5
     Herd Year NDVI_May NDVI_June_August
##
                                            maxndvi
##
     <chr> <int>
                   <dbl>
## 1
      BAT
          1982 0.21440
                               0.3722679 0.3722679
## 2
      BAT
           1983 0.20448
                               -0.9977483 0.2044800
## 3
      BAT
           1984 0.24650
                               1.5864094 1.5864094
## 4
                                0.6420830 0.6420830
      BAT 1985 0.24444
```