Day 5 - Descriptive Statistics

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2024-11-06

In this section we will finally see the basic application of R programming for our data analysis. After getting the cleaned data set, the first set need us to describe the data. This must be done using graphical representation. Using out Student performance data set.

```
##reading the data set.
##note that if you directory is not set to place where data is stores.
## you need to use setwd() to set it first.
Student_performance <- read.csv("StudentsPerformance.csv",header = T)
head(Student_performance)</pre>
```

##		gender	race.ethnicity pa	arental.leve	el.of.education	lunch
##	1	${\tt female}$	group B	bac	chelor's degree	standard
##	2	${\tt female}$	group C		some college	standard
##	3	${\tt female}$	group B	n	master's degree	standard
##	4	male	group A	asso	ciate's degree	free/reduced
##	5	male	group C		some college	standard
##	6	${\tt female}$	group B	asso	ciate's degree	standard
##		test.pr	reparation.course	math.score	reading.score	writing.score
##	1		none	72	72	74
##	2		completed	69	90	88
##	3		none	90	95	93
##	4		none	47	57	44
##	5		none	76	78	75
##	6		none	71	83	78

Descriptive Statistics

As the name suggest, it is use to describe the basic feature of the data. They provide simple summary for your data set without inferring any outcome to it. We can summaries the data in tabular or graphical for. For Example.

```
## average marks of the students.
Student_performance$average <- round(rowSums(Student_performance[,6:8])/3,2)
##grouping the data by gender.
mean_avg_marks <- tapply(Student_performance$average,Student_performance$gender,FUN = mean)
sd_avg_marks <- tapply(Student_performance$average,Student_performance$gender,FUN = sd)
med_avg_marks <- tapply(Student_performance$average,Student_performance$gender,FUN = median)
count_numbers <- tapply(Student_performance$average,Student_performance$gender,FUN = length)
marks_avg_gender <- data.frame(mean_avg_marks,sd_avg_marks,med_avg_marks,count_numbers)
## getting results as output
write.csv(marks_avg_gender,"mark_by_gender.csv")</pre>
```

A help from pacakge dplyr.

A package called dplyr can help to summaries your data more effectively.

```
#checking if dplyr is installed or not. If not then installing it.
if(!require(dplyr)){
    install.packages("dplyr",dependencies = T)
   library(dplyr)
}
## Loading required package: dplyr
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
mark_summary <- group_by(Student_performance,gender)%>%
  summarise(Math.avg = mean(math.score), Math.sd = sd(math.score),
            writing.mean = mean(writing.score), writing.sd = sd(writing.score),
            reading.mean = mean(reading.score), reading.sd = sd(reading.score),
            overall.mean = mean(average), overall.sd = sd(average),
            count = n())%>%
  as.data.frame()
write.csv(mark_summary, "mark_summary.csv")
```

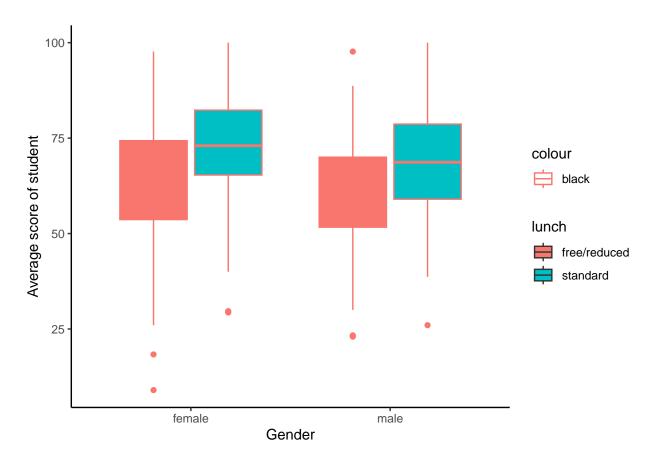
Plots for summary

We will use another package for piloting which is very common these days. It is called ggplot2

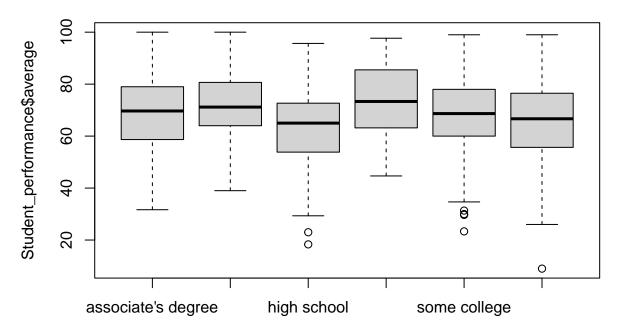
```
#checking if ggplot2 is installed or not. If not then installing it.
if(!require(ggplot2)){
   install.packages("ggplot2",dependencies = T)
   library(ggplot2)
}

## Loading required package: ggplot2

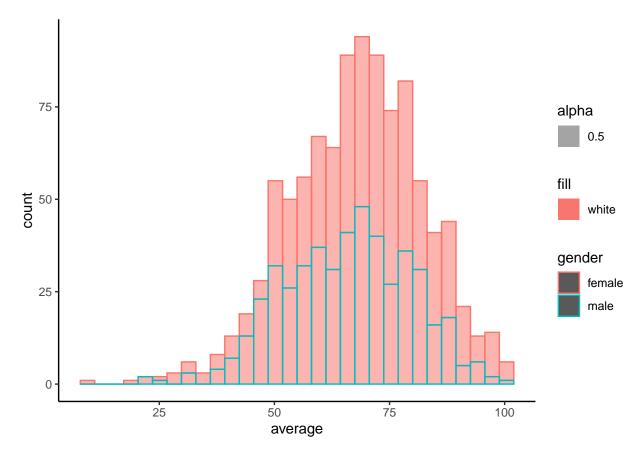
##Box plot
box_plot<-ggplot(data = Student_performance,aes(x=gender,y=average,fill=lunch))
box_plot+geom_boxplot(aes(color="black"),show.legend = T)+
   theme_classic()+
   labs(y="Average score of student",x="Gender")</pre>
```



##another way for box plot
boxplot(Student_performance\$average~Student_performance\$parental.level.of.education)



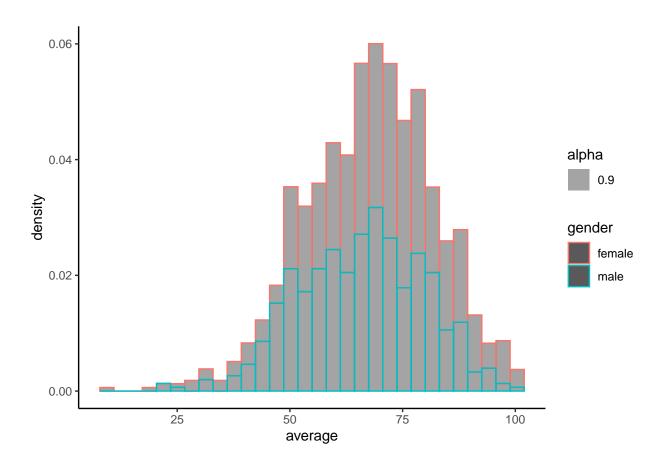
Student_performance\$parental.level.of.education



```
## = 10, : Ignoring unknown aesthetics: position and binwidth

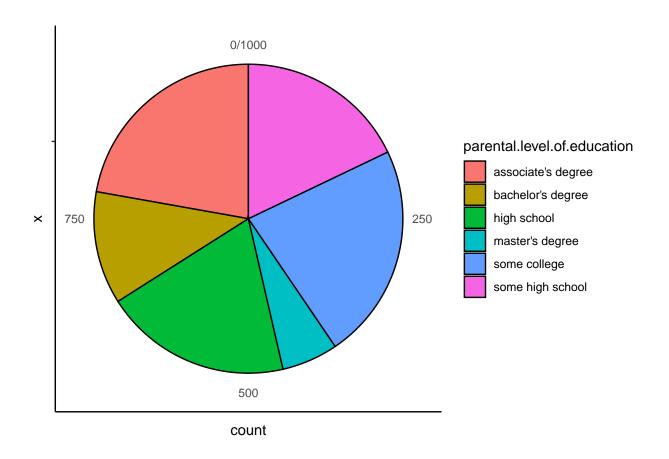
## Warning: The dot-dot notation ('..density..') was deprecated in ggplot2 3.4.0.
## i Please use 'after_stat(density)' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



pie chart

```
pie_data<-group_by(Student_performance,parental.level.of.education)%>%
    summarise(count=n())
pie_parent_edu <- ggplot(data = pie_data,aes(x="",y=count,fill=parental.level.of.education))
pie_parent_edu+geom_col(color="black")+coord_polar(theta = "y")+theme_classic()</pre>
```



ANNOVA test.

Analysis of variance is statically test which is used to see if two group are different from each other. There are some assumption in the which we need to take consideration of

- 1. Independence of case
- 2. Normality
- 3. Homogeneity

```
##load the data
Yeild <- read.csv("./crop_yeild/field_data.csv",header = T)
## checking the summary of the data
summary(Yeild)</pre>
```

```
##
        height
                                        humidity
                                                                        yield
                         temp
                                                        fertilizer
    {\tt Min.}
##
           :1.0
                   Min.
                           :60.00
                                    Min.
                                            :60.00
                                                     Min.
                                                             :1
                                                                   Min.
                                                                           :395.4
                   1st Qu.:62.00
                                    1st Qu.:64.75
                                                                    1st Qu.:396.5
    1st Qu.:1.0
                                                      1st Qu.:1
##
   Median:1.5
                   Median :66.00
                                    Median :69.00
                                                     Median :2
                                                                    Median :397.1
##
    Mean
            :1.5
                   Mean
                           :65.16
                                    Mean
                                            :70.03
                                                     Mean
                                                             :2
                                                                    Mean
                                                                           :397.0
                   3rd Qu.:68.00
                                    3rd Qu.:77.00
                                                     3rd Qu.:3
                                                                    3rd Qu.:397.4
##
    3rd Qu.:2.0
                          :70.00
                                            :80.00
##
    Max.
            :2.0
                   Max.
                                    Max.
                                                     Max.
                                                             :3
                                                                    Max.
                                                                           :399.1
##
        region
##
    Min.
           :1.00
```

```
## 1st Qu::1.75

## Median :2.50

## Mean :2.50

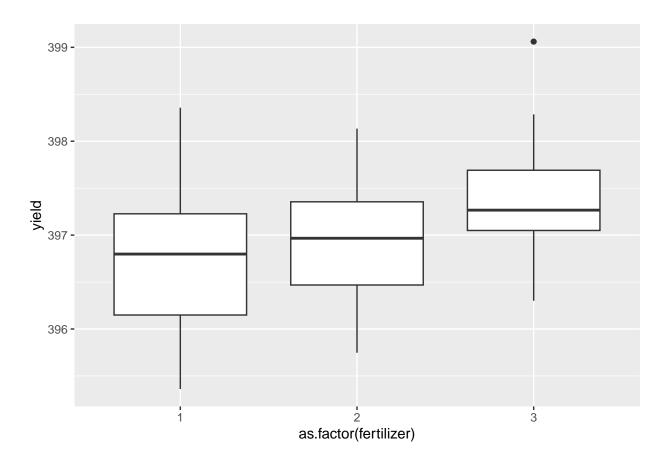
## 3rd Qu::3.25

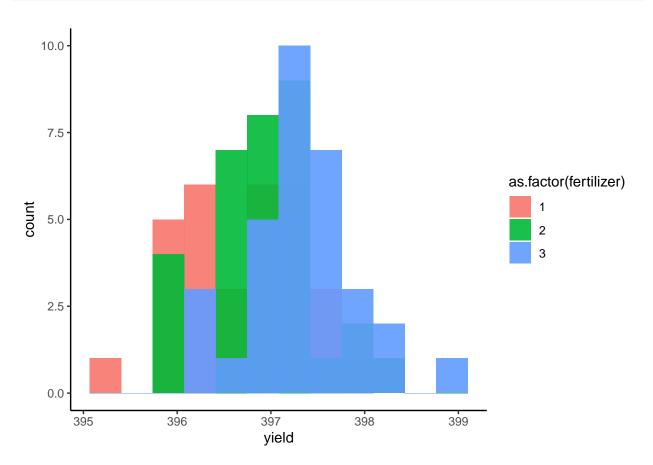
## Max: :4.00
```

##checking the structure of the data str(Yeild)

```
## 'data.frame': 96 obs. of 6 variables:
## $ height : int 1 2 1 2 1 2 1 2 1 2 ...
## $ temp : int 70 68 69 69 70 68 67 64 70 64 ...
## $ humidity : int 72 76 63 65 76 67 77 74 67 62 ...
## $ fertilizer: int 1 1 1 1 1 1 1 1 1 1 ...
## $ yield : num 397 398 396 398 397 ...
## $ region : int 1 2 3 4 1 2 3 4 1 2 ...
```

```
##basic description of the data.
library(ggplot2)
#boxplot
plot_box<-ggplot(data = Yeild,aes(x=as.factor(fertilizer),y=yield))
plot_box+geom_boxplot()</pre>
```





#Shapiro-Wilk normality test shapiro.test(Yeild\$yield)

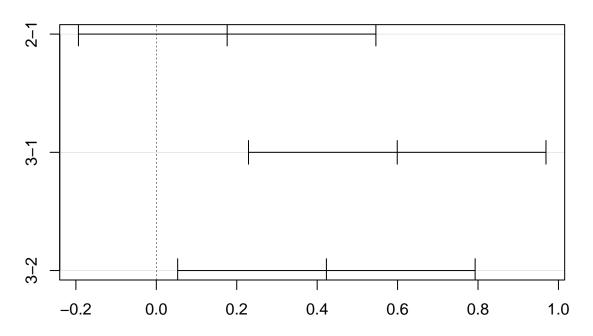
```
##
## Shapiro-Wilk normality test
##
## data: Yeild$yield
## W = 0.989, p-value = 0.6135

mod1<-aov(formula = Yeild$yield ~ as.character(Yeild$fertilizer))
summary(mod1)</pre>
```

```
## Df Sum Sq Mean Sq F value Pr(>F)
## as.character(Yeild$fertilizer) 2 6.07 3.0340 7.863 7e-04 ***
## Residuals 93 35.89 0.3859
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

#posthoc
Post_hoc <- TukeyHSD(mod1)
plot(Post_hoc)</pre>

95% family-wise confidence level



Differences in mean levels of as.character(Yeild\$fertilizer)