

Playing with CGPA dataset

Load required packages

```
library(readxl)
library(dplyr)

## Warning: package 'dplyr' was built under R version 4.5.2

##
## 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

library(psych) # describe()

## Warning: package 'psych' was built under R version 4.5.2

library(skimr) # skim()

## Warning: package 'skimr' was built under R version 4.5.2

library(pastecs) # stat.desc()

## Warning: package 'pastecs' was built under R version 4.5.2

##
## Attaching package: 'pastecs'

## The following objects are masked from 'package:dplyr':
##     first, last
```

Load and see the data

```
df <- read_excel("C:\\\\Users\\\\Dell\\\\OneDrive\\\\Desktop\\\\STATISTICS\\\\2nd Year\\\\R programming\\\\varsity\\\\Pro
describe(df)
```

```
##          vars   n     mean      sd    median    trimmed     mad
## ExamRoll       1 80 2316641.96 23.91 2316642.50 2316642.08 30.39
## Sex*          2 80      1.62  0.49      2.00      1.66  0.00
## CGPA          3 80      3.42  0.44      3.48      3.48  0.39
## Group          4 80      2.28  1.03      2.00      2.22  1.48
## AssignedTeacher* 5 80     11.18  6.64     10.50     11.02  8.15
##                  min      max range skew kurtosis    se
## ExamRoll       2316601.00 2316682.00 81.00 -0.04    -1.24 2.67
## Sex*           1.00      2.00  1.00 -0.51    -1.76 0.05
## CGPA           2.05      3.99  1.94 -1.22     1.14 0.05
## Group           1.00      4.00  3.00  0.19    -1.18 0.12
## AssignedTeacher* 1.00     23.00 22.00  0.18    -1.21 0.74
```

```
skim(df)
```

Table 1: Data summary

Name	df
Number of rows	80
Number of columns	5
<hr/>	
Column type frequency:	
character	2
numeric	3
<hr/>	
Group variables	None

Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
Sex	0	1	1	1	0	2	0
AssignedTeacher	0	1	1	2	0	23	0

Variable type: numeric

skim_variable	missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
ExamRoll	0	1	2316641.96	23.91	2316601.00	2316621.75	2316642.50	2316662.25	2316682.00	
CGPA	0	1	3.42	0.44	2.05	3.24	3.48	3.77	3.99	
Group	0	1	2.28	1.03	1.00	1.00	2.00	3.00	4.00	

```
glimpse(df)
```

```
## Rows: 80
## Columns: 5
## $ ExamRoll      <dbl> 2316669, 2316632, 2316615, 2316643, 2316679, 2316667, ~
```

```

## $ Sex <chr> "F", "M", "M", "M", "F", "F", "M", "M", "M", "F", "F", ~
## $ CGPA <dbl> 3.988, 3.953, 3.938, 3.938, 3.914, 3.910, 3.864, 3.857~
## $ Group <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ AssignedTeacher <chr> "B", "O", "C", "FT", "MT", "T", "K", "I", "E", "G", "W~
```

```
stat.desc(df)
```

	ExamRoll	Sex	CGPA	Group	AssignedTeacher
## nbr.val	8.000000e+01	NA	80.0000000	80.0000000	NA
## nbr.null	0.000000e+00	NA	0.0000000	0.0000000	NA
## nbr.na	0.000000e+00	NA	0.0000000	0.0000000	NA
## min	2.316601e+06	NA	2.0530000	1.0000000	NA
## max	2.316682e+06	NA	3.9880000	4.0000000	NA
## range	8.100000e+01	NA	1.9350000	3.0000000	NA
## sum	1.853314e+08	NA	273.2180000	182.0000000	NA
## median	2.316643e+06	NA	3.4840000	2.0000000	NA
## mean	2.316642e+06	NA	3.4152250	2.2750000	NA
## SE.mean	2.673324e+00	NA	0.04932226	0.1152529	NA
## CI.mean.0.95	5.321118e+00	NA	0.09817350	0.2294051	NA
## var	5.717328e+02	NA	0.19461483	1.0626582	NA
## std.dev	2.391093e+01	NA	0.44115171	1.0308532	NA
## coef.var	1.032138e-05	NA	0.12917208	0.4531223	NA

```
summary(df) # Exam Roll numbers, Sex, CGPA, Group assignment, and Assigned Teacher
```

	ExamRoll	Sex	CGPA	Group
## Min.	:2316601	Length:80	Min. :2.053	Min. :1.000
## 1st Qu.	:2316622	Class :character	1st Qu.:3.242	1st Qu.:1.000
## Median	:2316643	Mode :character	Median :3.484	Median :2.000
## Mean	:2316642		Mean :3.415	Mean :2.275
## 3rd Qu.	:2316662		3rd Qu.:3.769	3rd Qu.:3.000
## Max.	:2316682		Max. :3.988	Max. :4.000
## AssignedTeacher				
## Length:	80			
## Class	:character			
## Mode	:character			
##				
##				
##				

```
sum(is.na(df)) # no Missing values
```

```
## [1] 0
```

Convert variable types

```
df$ExamRoll<-as.character(df$ExamRoll)
class(df$ExamRoll)
```

```
## [1] "character"
```

```

df$Sex <- as.factor(df$Sex)
class(df$Sex)

## [1] "factor"

df$Group <- as.factor(df$Group)
class(df$Group)

## [1] "factor"

df$AssignedTeacher <- as.factor(df$AssignedTeacher)
class(df$AssignedTeacher)

## [1] "factor"

```

See the data again

```

str(df)

## # tibble [80 x 5] (S3: tbl_df/tbl/data.frame)
## $ ExamRoll      : chr [1:80] "2316669" "2316632" "2316615" "2316643" ...
## $ Sex           : Factor w/ 2 levels "F","M": 1 2 2 2 1 1 2 2 2 1 ...
## $ CGPA          : num [1:80] 3.99 3.95 3.94 3.94 3.91 ...
## $ Group          : Factor w/ 4 levels "1","2","3","4": 1 1 1 1 1 1 1 1 1 1 ...
## $ AssignedTeacher: Factor w/ 23 levels "A","B","C","D",...: 2 15 3 6 13 20 11 9 5 7 ...

summary(df)

##      ExamRoll          Sex        CGPA       Group   AssignedTeacher
##  Length:80    F:30   Min.   :2.053   1:23     A     : 4
##  Class :character M:50   1st Qu.:3.242   2:23     B     : 4
##  Mode  :character                    Median :3.484   3:23     C     : 4
##                           Mean   :3.415   4:11     D     : 4
##                           3rd Qu.:3.769                   E     : 4
##                           Max.   :3.988                   FT    : 4
##                           (Other):56

```

Descriptive Statistics

For CGPA (i) Mean (ii) Median (iii) Standard deviation

```

describe(df$CGPA)

##      vars  n mean   sd median trimmed  mad  min  max range skew kurtosis   se
## X1      1 80 3.42 0.44    3.48    3.48 0.39 2.05 3.99  1.94 -1.22     1.14 0.05

```

Frequency Distributions (i) Sex (ii) Group (iii) AssignedTeacher

```



```

```

df %>%
  count(df$AssignedTeacher) %>% # Frequency
  mutate(
    percent = n / sum(n) * 100,
    cum_n = cumsum(n),
    cum_percent = cumsum(percent),
    relative_freq = n / sum(n),
    cum_relative = cumsum(relative_freq)
  ) %>%
  rename(Frequency = n)

## # A tibble: 23 x 7
##   `df$AssignedTeacher` Frequency percent cum_n cum_percent relative_freq
##   <fct>              <int>    <dbl>  <int>      <dbl>        <dbl>
## 1 A                  4        5     4       5        0.05
## 2 B                  4        5     8       10       0.05
## 3 C                  4        5    12       15       0.05
## 4 D                  4        5    16       20       0.05
## 5 E                  4        5    20       25       0.05
## 6 FT                 4        5    24       30       0.05
## 7 G                  4        5    28       35       0.05
## 8 H                  4        5    32       40       0.05
## 9 I                  4        5    36       45       0.05
## 10 J                 4        5    40       50       0.05
## # i 13 more rows
## # i 1 more variable: cum_relative <dbl>

```

Group Comparisons (i) CGPA by Sex (mean, SD, count) (ii) CGPA by Group (mean, SD, count)

```

# Using dplyr (recommended)
result <- df %>%
  group_by(Sex) %>%
  summarise(
    Count = n(),
    Mean_CGPA = round(mean(CGPA, na.rm = TRUE), 2),
    SD_CGPA = round(sd(CGPA, na.rm = TRUE), 2),
    .groups = 'drop'
  )
result

## # A tibble: 2 x 4
##   Sex   Count Mean_CGPA SD_CGPA
##   <fct> <int>    <dbl>    <dbl>
## 1 F      30      3.52    0.38
## 2 M      50      3.35    0.47

# knitr::kable(result, caption = "CGPA by Sex") # Display nicely

result <- df %>%
  group_by(Group) %>%
  summarise(
    Count = n(),

```

```

    Mean_CGPA = round(mean(CGPA, na.rm = TRUE), 2),
    SD_CGPA = round(sd(CGPA, na.rm = TRUE), 2),
    .groups = 'drop'
)
result

## # A tibble: 4 x 4
##   Group Count Mean_CGPA SD_CGPA
##   <fct> <int>     <dbl>    <dbl>
## 1 1      23      3.84    0.07
## 2 2      23      3.57    0.09
## 3 3      23      3.26    0.13
## 4 4      11      2.54    0.32

# For everyday use: dplyr (most readable and flexible)

# For quick analysis: psych::describeBy()

# For large datasets: data.table

# For publication: gtsummary

```

Data Visualization

Distribution Plots (i) Histogram of CGPA (ii) Density plot of CGPA

```

library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.5.2

##
## Attaching package: 'ggplot2'

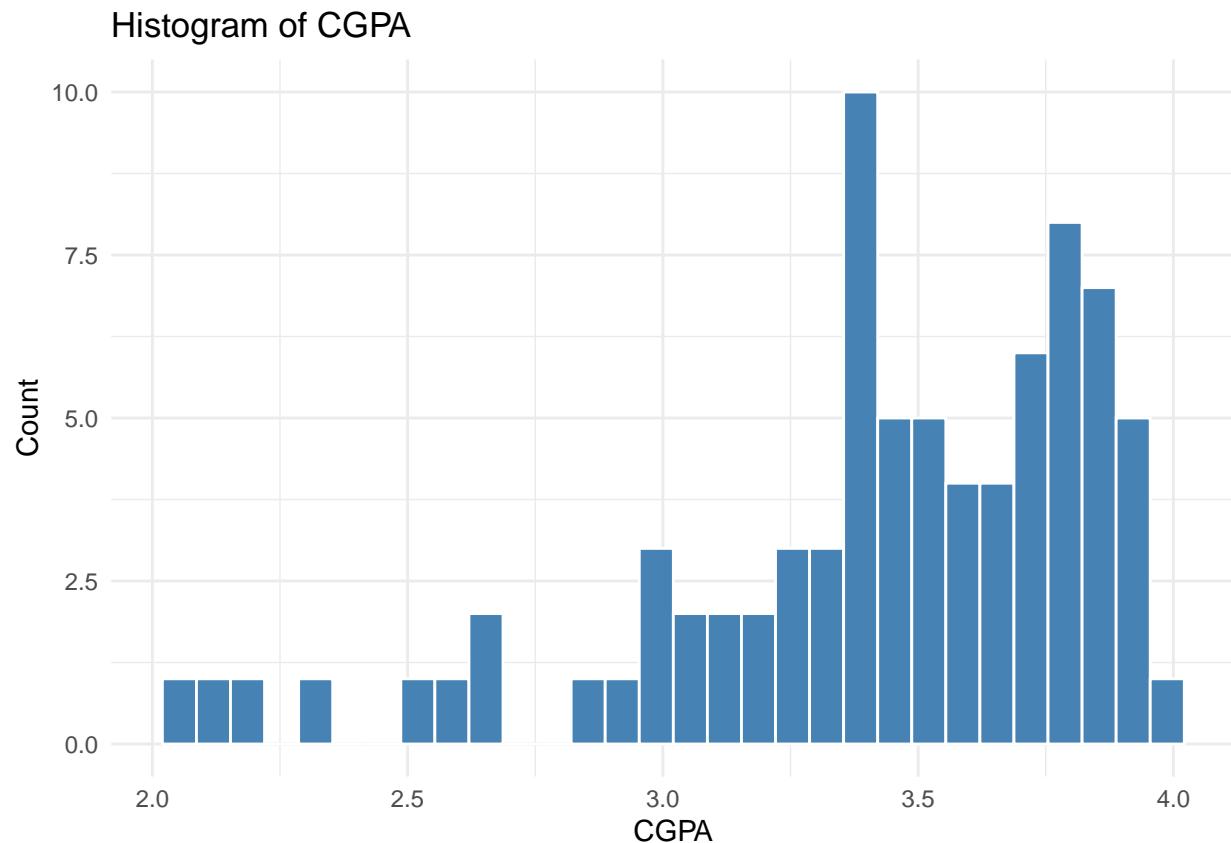
## The following objects are masked from 'package:psych':
## 
##     %+%, alpha

# install.packages("esquisse")
# library(esquisse) # Creates ggplot2 code automatically
# esquisser(df) # Opens drag-and-drop interface

## Template for any plot
#ggplot(data = df, aes(x = x_variable, y = y_variable)) +
#  geom_*(aes(color/fill = grouping_variable)) + # Choose geom
#  labs(title = "Title", x = "X Label", y = "Y Label") +
#  theme_minimal() +
#  scale_fill_brewer(palette = "Set2") # Color palette

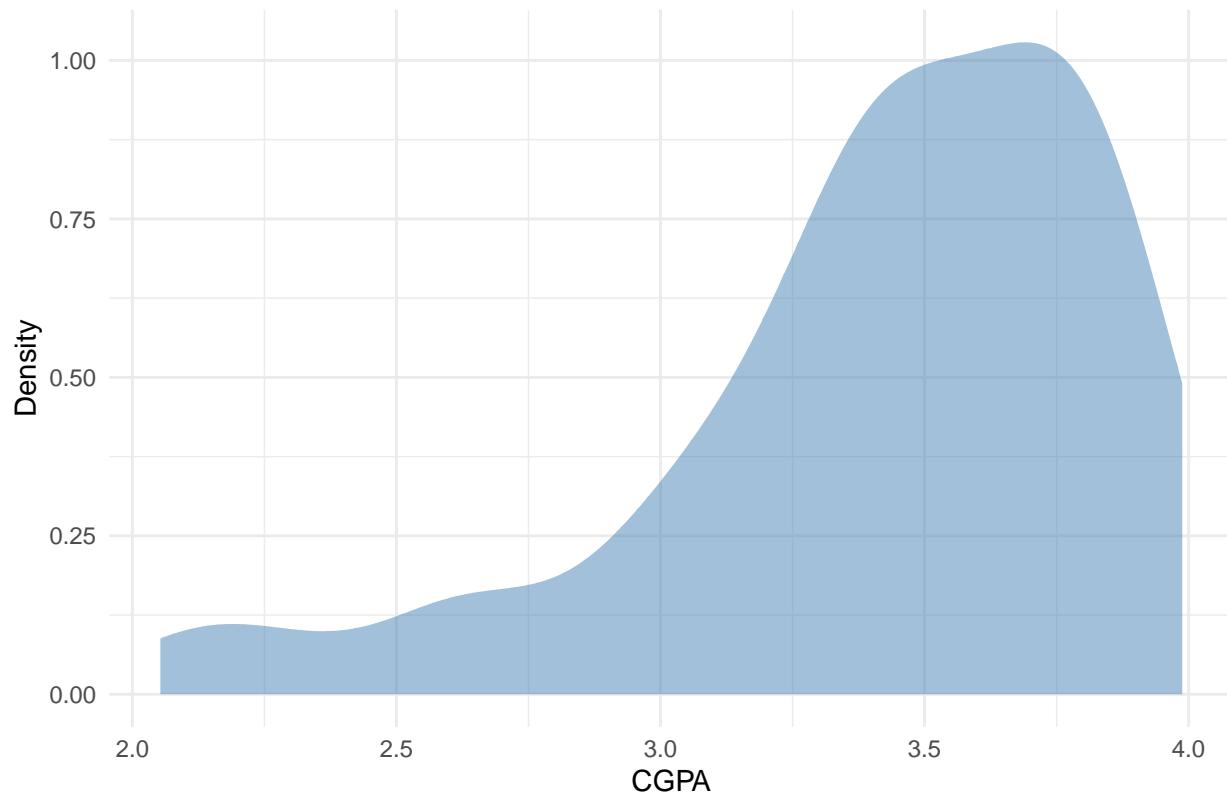
```

```
# Histogram of CGPA
ggplot(df, aes(x = CGPA)) +
  geom_histogram(bins = 30, fill = "steelblue", color = "white") +
  labs(title = "Histogram of CGPA", x = "CGPA", y = "Count") +
  theme_minimal()
```



```
# Density plot of CGPA
ggplot(df, aes(x = CGPA)) +
  geom_density(fill = "steelblue", alpha = 0.5, color = NA) +
  labs(title = "Density Plot of CGPA", x = "CGPA", y = "Density") +
  theme_minimal()
```

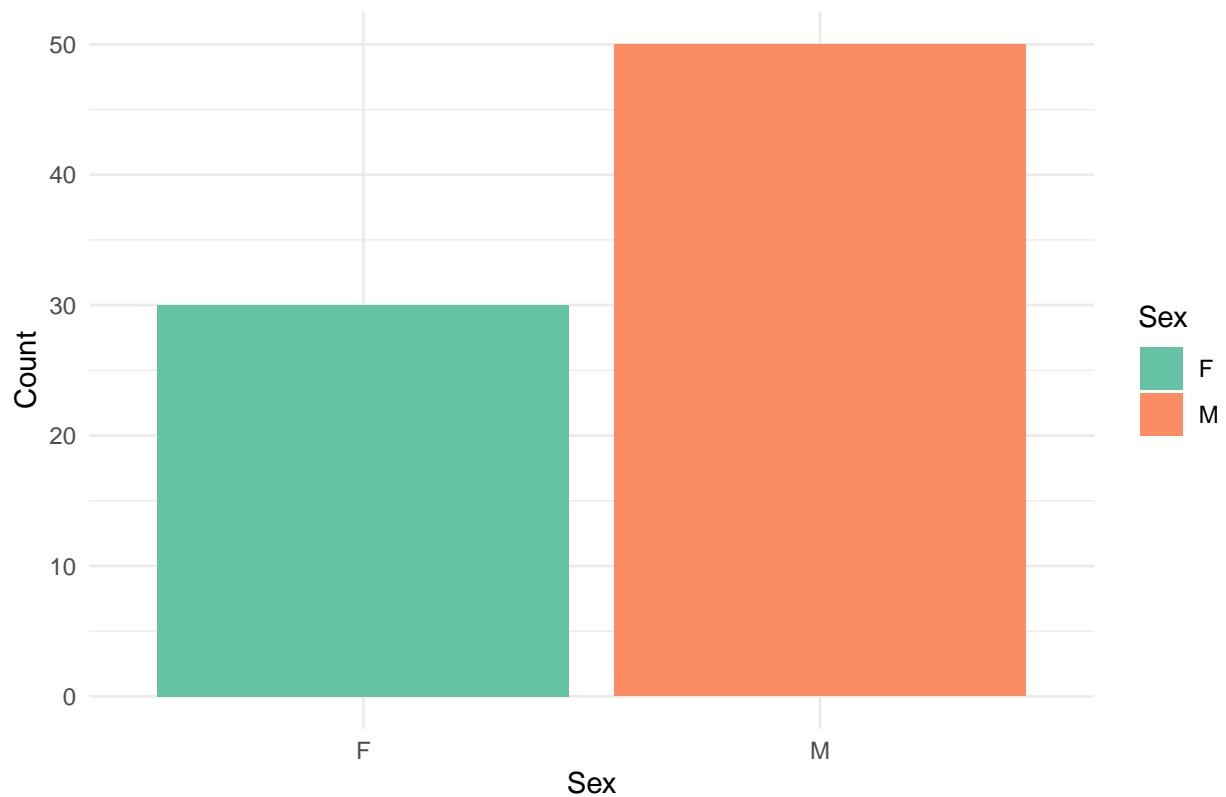
Density Plot of CGPA



Categorical Data Plots: (i) Bar plot of Sex (ii) Bar plot of Group (iii) Bar plot of AssignedTeacher

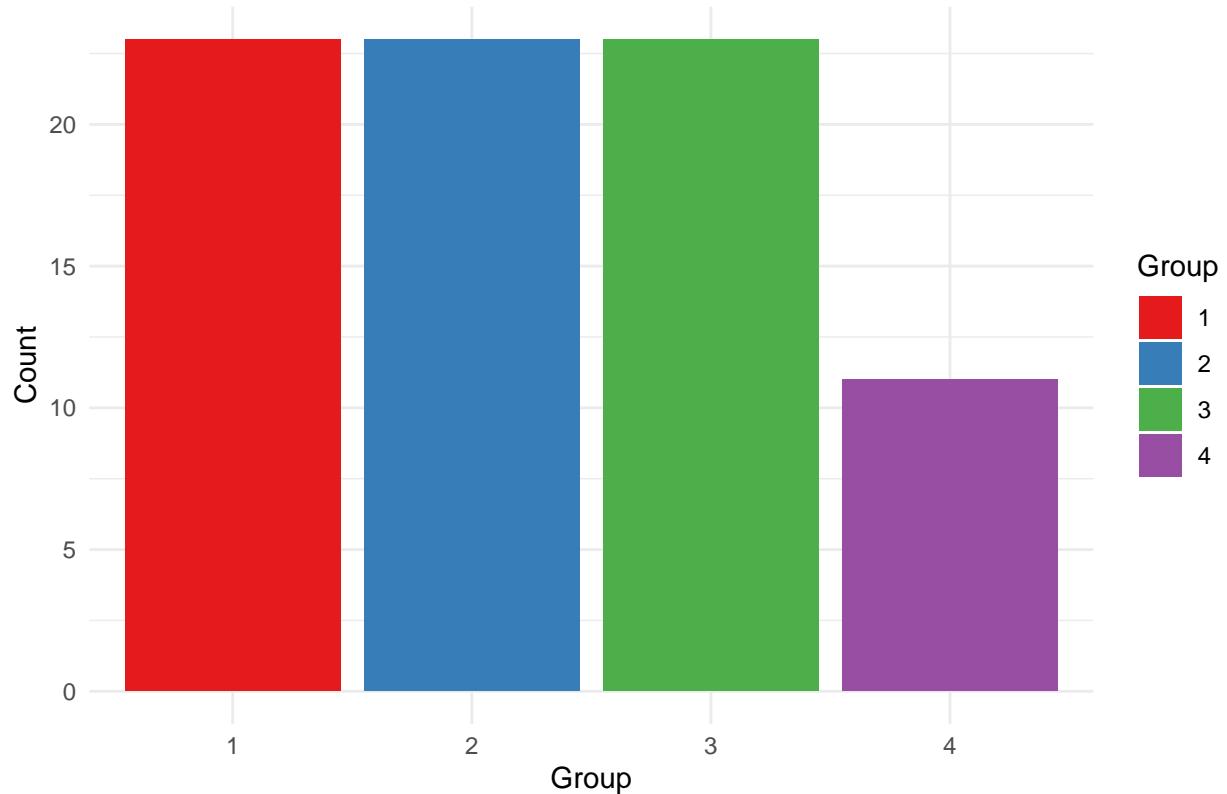
```
# Bar plot of Sex
ggplot(df, aes(x = Sex, fill = Sex, group = Sex)) +
  geom_bar() +
  labs(title = "Bar Plot of Sex", x = "Sex", y = "Count") +
  theme_minimal() +
  scale_fill_brewer(palette = "Set2")
```

Bar Plot of Sex



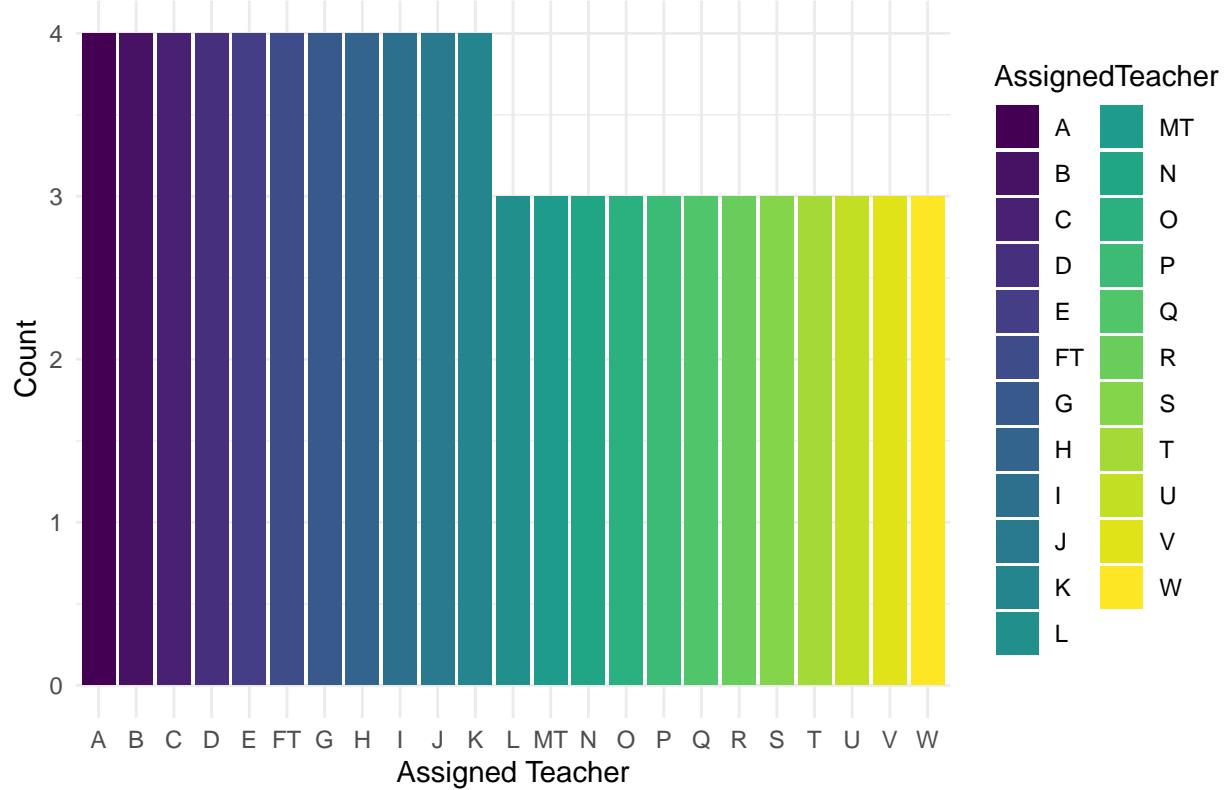
```
# Bar plot of Group
ggplot(df, aes(x = Group, fill = Group, group = Group)) +
  geom_bar() +
  labs(title = "Bar Plot of Group", x = "Group", y = "Count") +
  theme_minimal() +
  scale_fill_brewer(palette = "Set1")
```

Bar Plot of Group



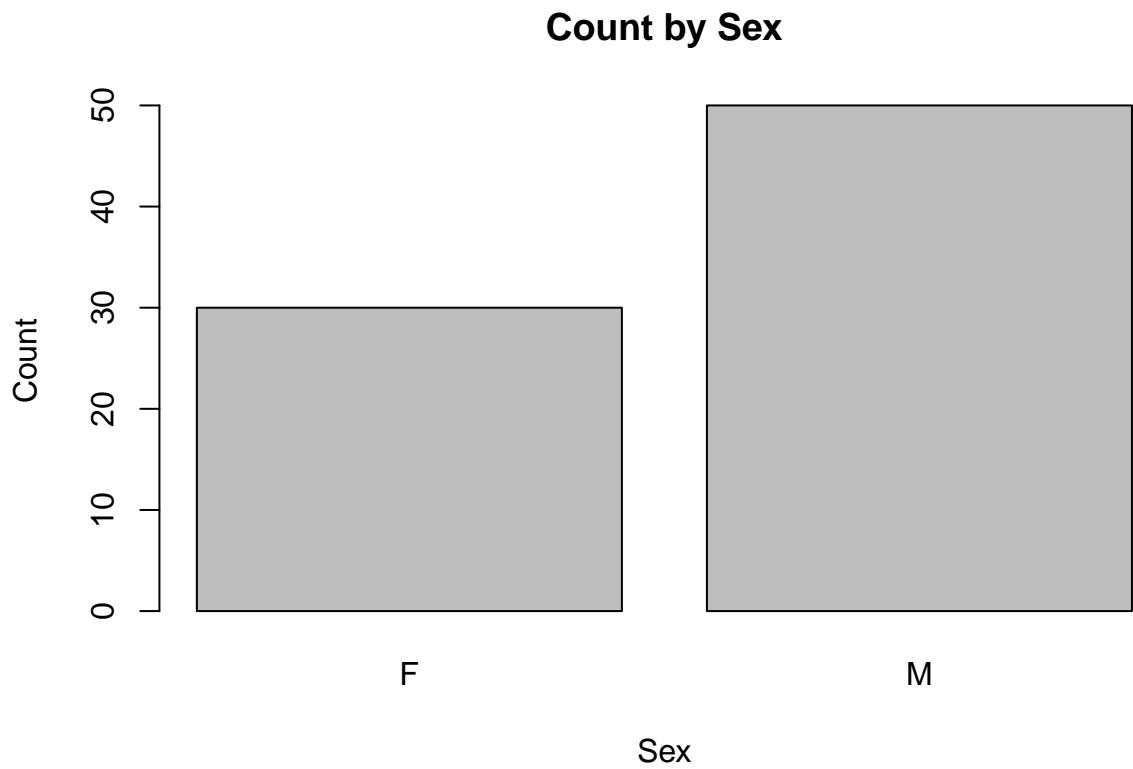
```
# Bar plot of Assigned Teacher
ggplot(df, aes(x = AssignedTeacher, fill = AssignedTeacher, group = AssignedTeacher)) +
  geom_bar() +
  labs(title = "Bar Plot of Assigned Teacher",
       x = "Assigned Teacher", y = "Count") +
  theme_minimal() +
  scale_fill_viridis_d()           # many categories → no Brewer palette
```

Bar Plot of Assigned Teacher

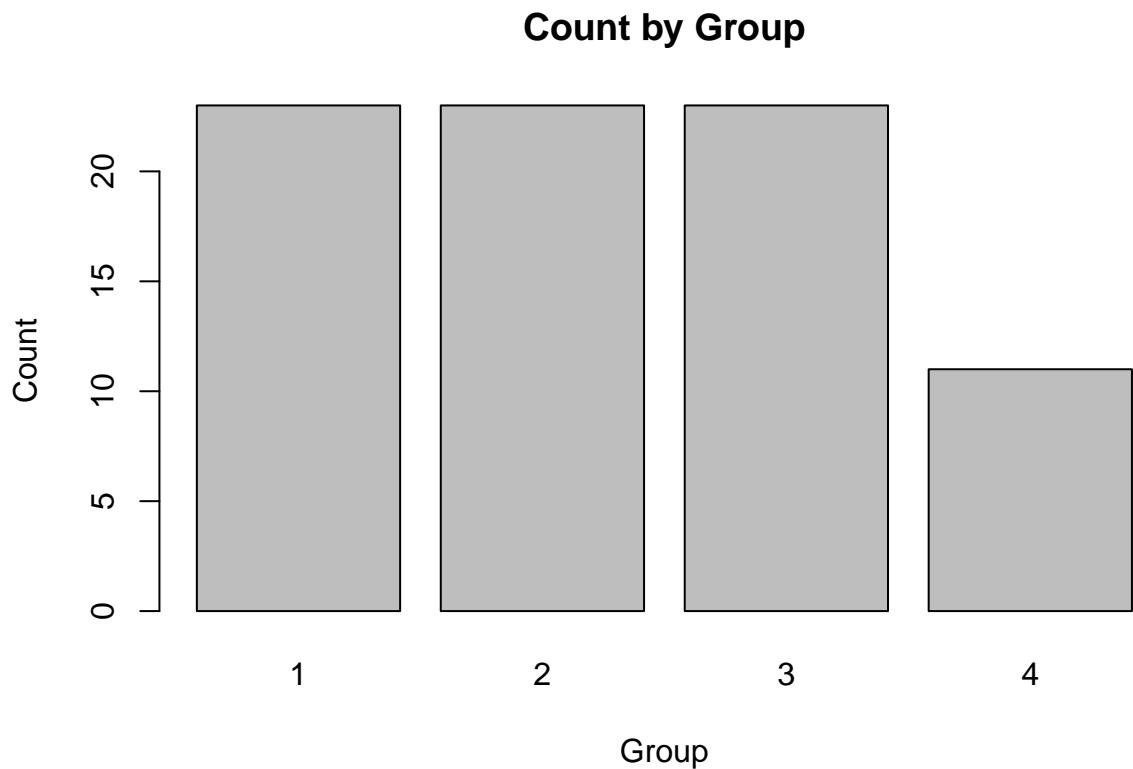


Easier Way

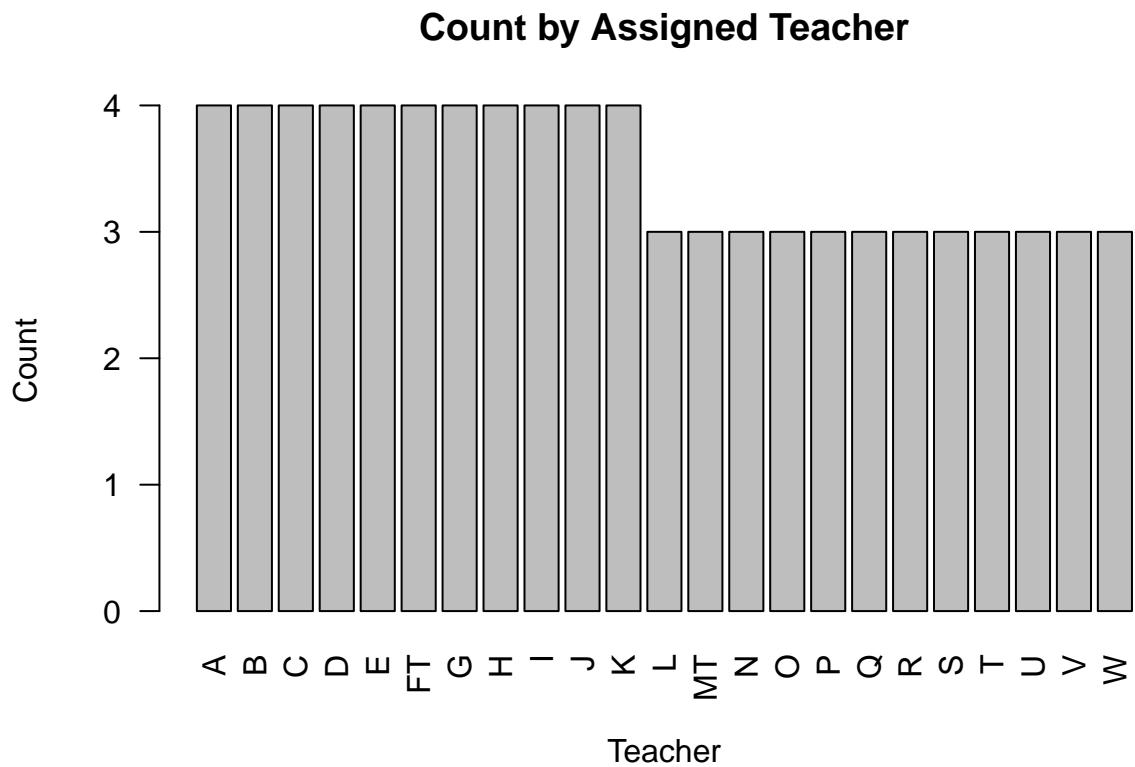
```
# Bar plot of Sex
barplot(table(df$Sex), main = "Count by Sex", xlab = "Sex", ylab = "Count")
```



```
# Bar plot of Group  
barplot(table(df$Group), main = "Count by Group", xlab = "Group", ylab = "Count")
```



```
# Bar plot of AssignedTeacher
barplot(table(df$AssignedTeacher), main = "Count by Assigned Teacher",
       xlab = "Teacher", ylab = "Count", las = 2)
```

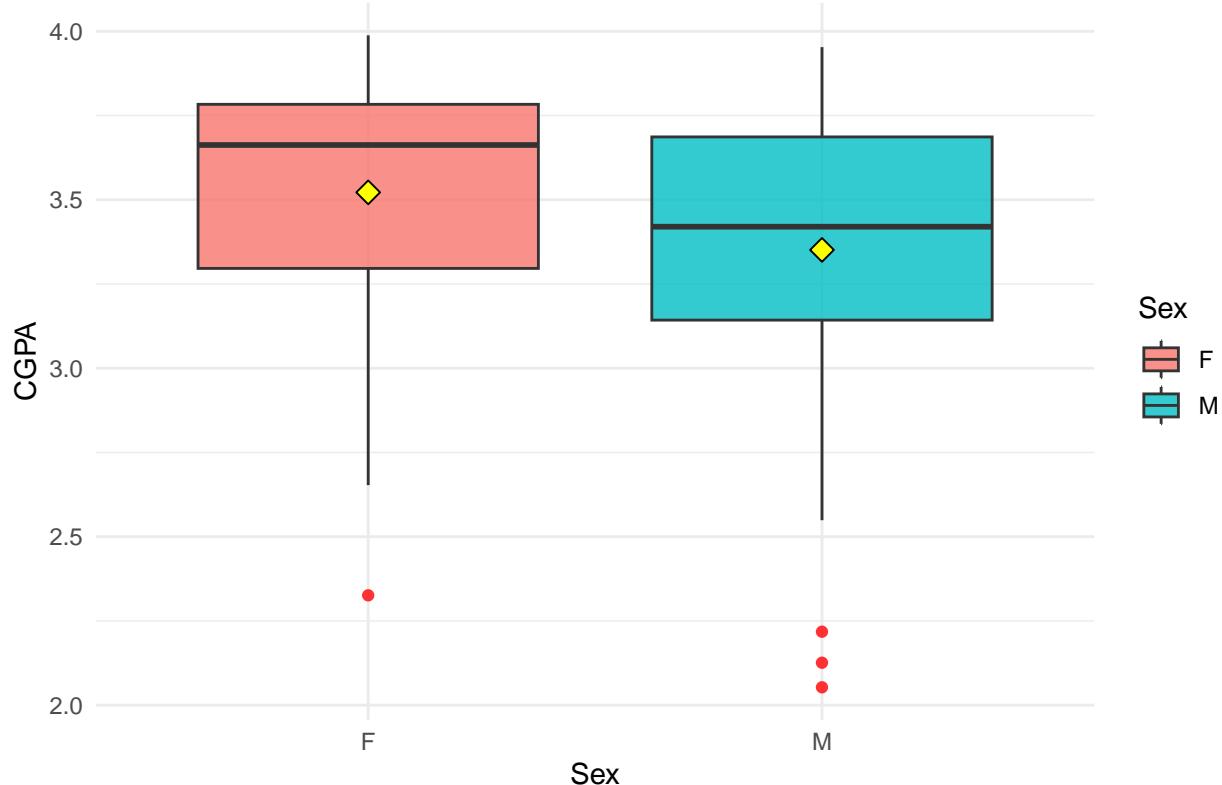


Comparative Plots: (i) Boxplot of CGPA by Sex (ii) Boxplot of CGPA by Group

```
library(ggplot2)

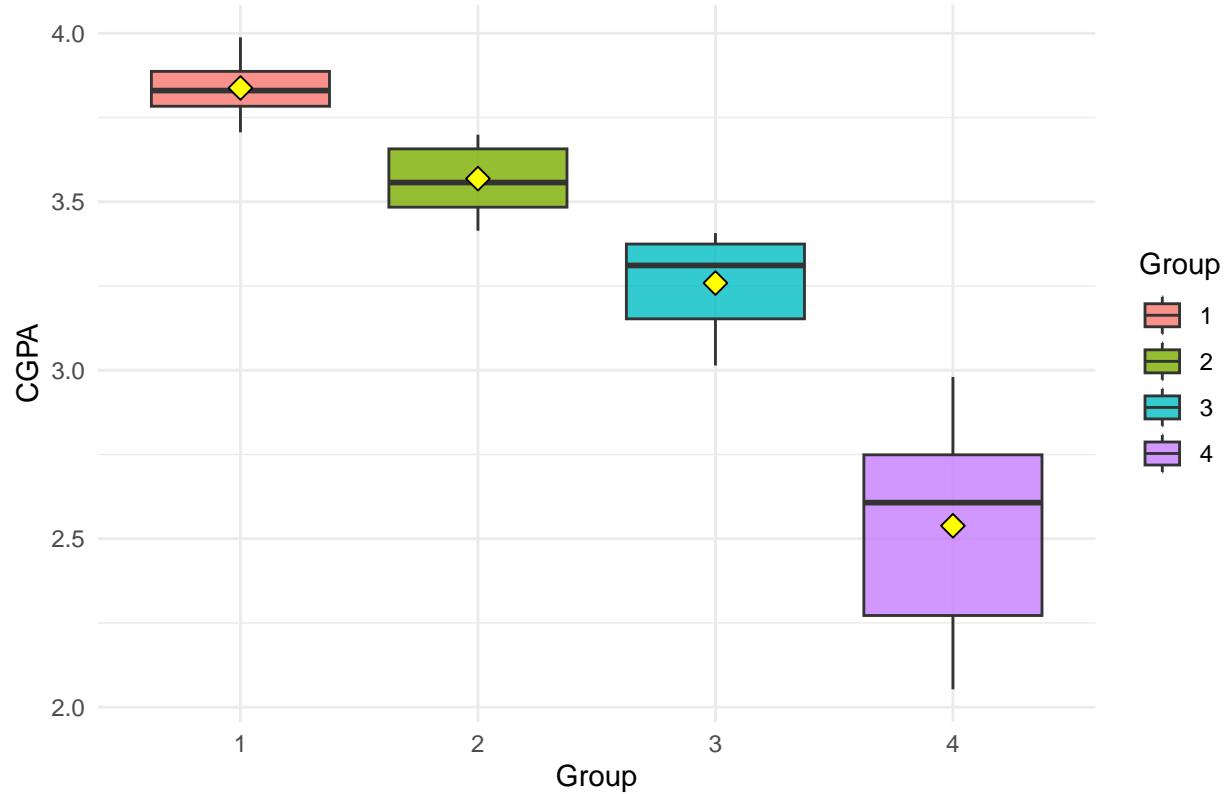
ggplot(df, aes(x = Sex, y = CGPA, fill = Sex)) +
  geom_boxplot(alpha = 0.8, outlier.color = "red") +
  stat_summary(fun = mean,
              geom = "point",
              shape = 23,
              size = 3,
              fill = "yellow") +
  labs(title = "Boxplot of CGPA by Sex",
       x = "Sex",
       y = "CGPA") +
  theme_minimal()
```

Boxplot of CGPA by Sex



```
ggplot(df, aes(x = Group, y = CGPA, fill = Group)) +  
  geom_boxplot(alpha = 0.8, outlier.color = "red") +  
  stat_summary(fun = mean,  
              geom = "point",  
              shape = 23,  
              size = 3,  
              fill = "yellow") +  
  labs(title = "Boxplot of CGPA by Group",  
       x = "Group",  
       y = "CGPA") +  
  theme_minimal()
```

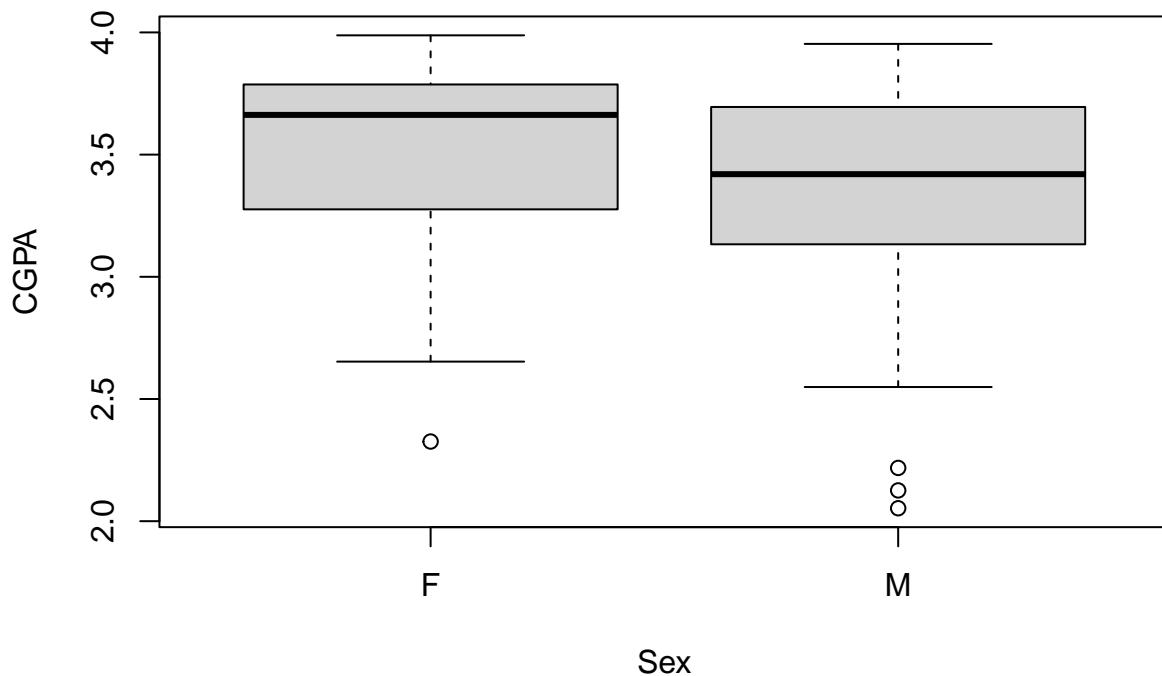
Boxplot of CGPA by Group



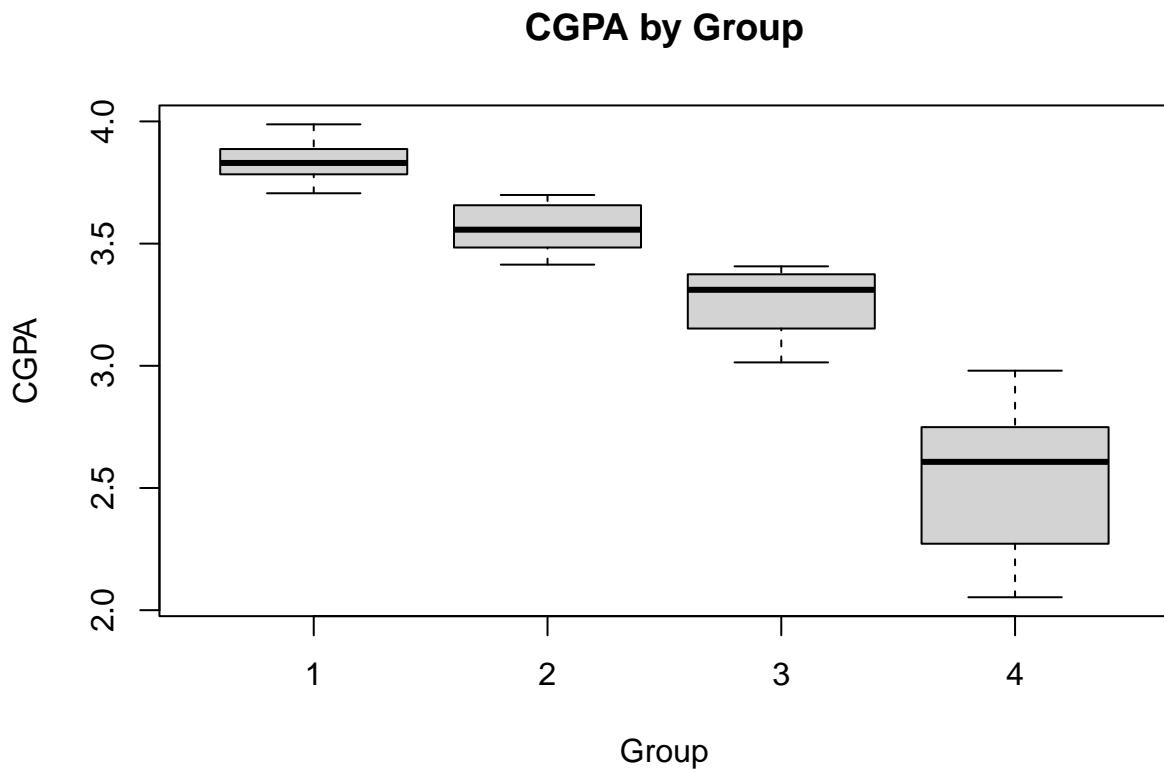
Easier way:

```
boxplot(CGPA ~ Sex, data = df, main = "CGPA by Sex", xlab = "Sex", ylab = "CGPA")
```

CGPA by Sex



```
boxplot(CGPA ~ Group, data = df, main = "CGPA by Group", xlab = "Group", ylab = "CGPA")
```



```
# Inferential Statistics
```

Mean Comparisons : 1) Independent samples t-test: CGPA by Sex 2) One-way ANOVA: CGPA by Group
 3) One-way ANOVA: CGPA by AssignedTeacher

```
library(ggstatsplot)

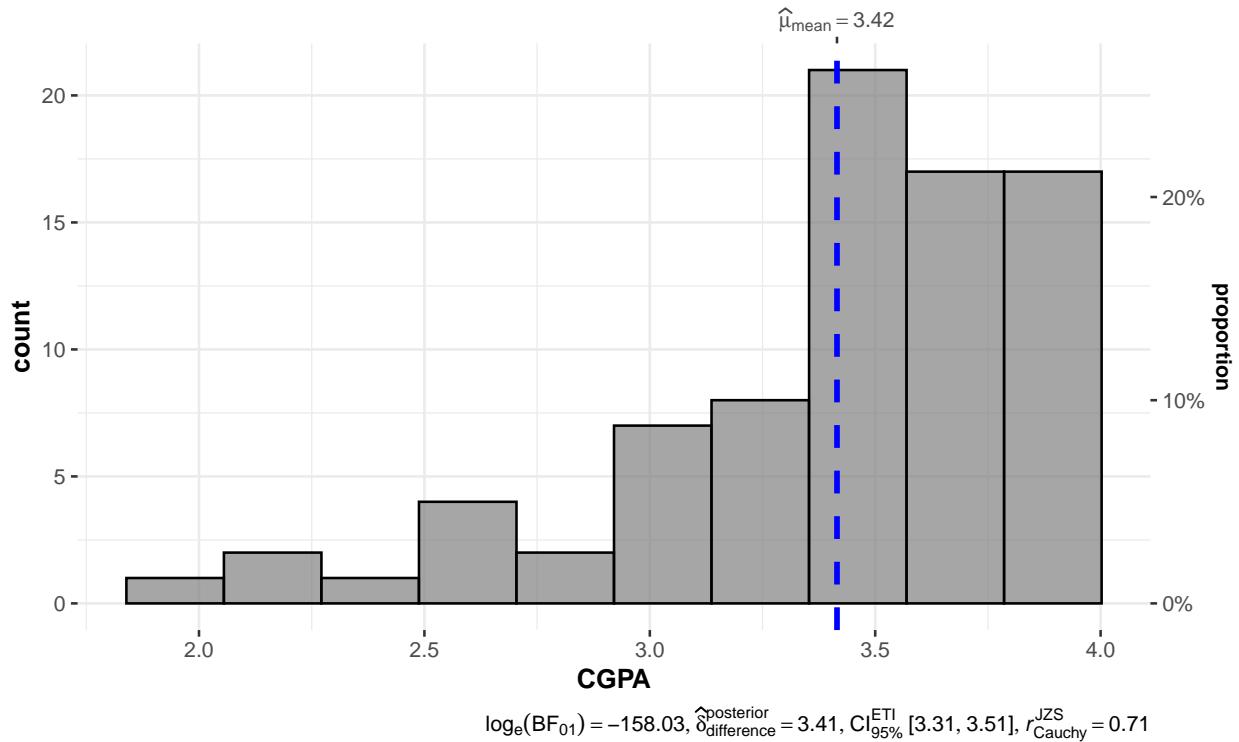
## Warning: package 'ggstatsplot' was built under R version 4.5.2

## You can cite this package as:
##   Patil, I. (2021). Visualizations with statistical details: The 'ggstatsplot' approach.
##   Journal of Open Source Software, 6(61), 3167, doi:10.21105/joss.03167

# t-test with plot
gghistostats(
  data = df,
  x = CGPA,
  y = Sex,
  type = "parametric", # t-test
  title = "CGPA by Sex with t-test"
)
```

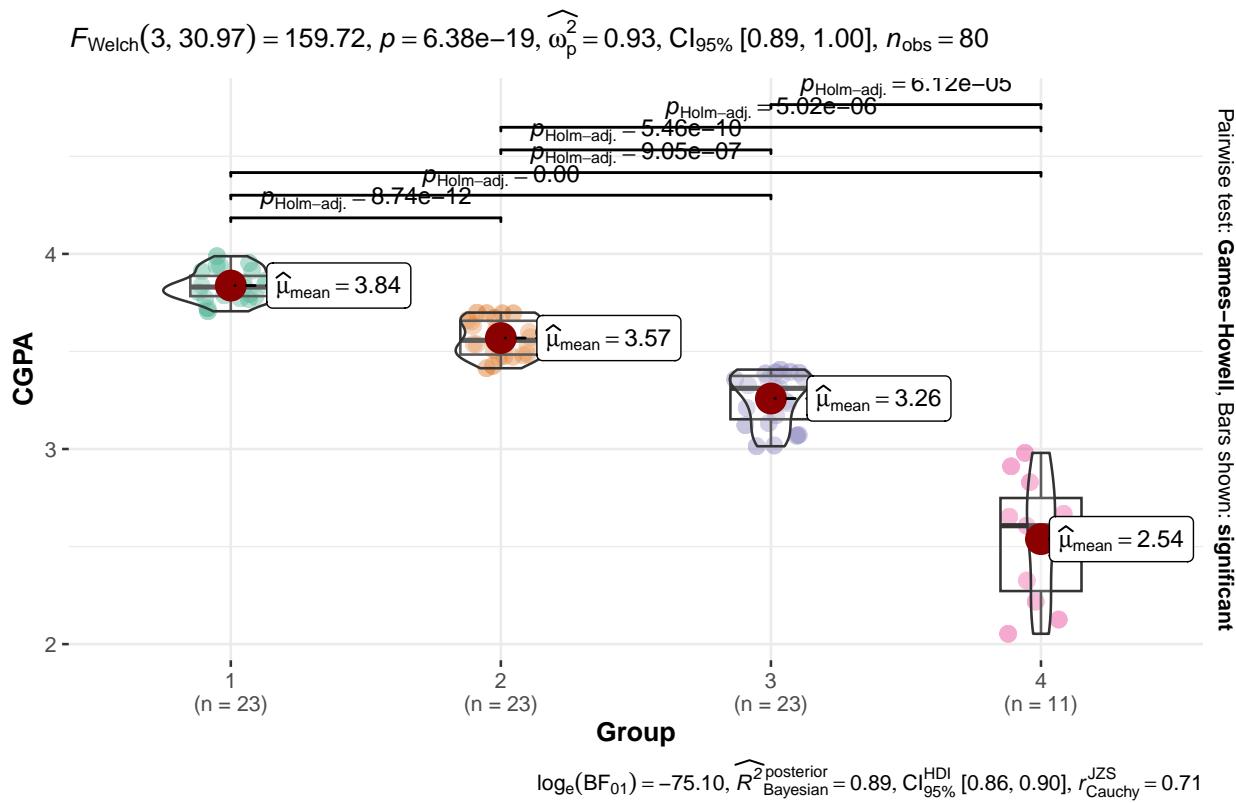
CGPA by Sex with t-test

$t_{\text{Student}}(79) = 69.24, p = 1.74e-72, \hat{g}_{\text{Hedges}} = 7.67, \text{CI}_{95\%} [6.43, 8.86], n_{\text{obs}} = 80$



```
# ANOVA with plot
ggbetweenstats(
  data = df,
  x = Group,
  y = CGPA,
  type = "parametric", # ANOVA
  title = "CGPA by Group with ANOVA",
  pairwise.comparisons = TRUE, # Shows post-hoc
  pairwise.display = "significant"
)
```

CGPA by Group with ANOVA



Easier Way:

```
# (i) t-test
t.test(CGPA ~ Sex, data = df, var.equal = TRUE) # Equal variances
```

```
##
## Two Sample t-test
##
## data: CGPA by Sex
## t = 1.6962, df = 78, p-value = 0.09384
## alternative hypothesis: true difference in means between group F and group M is not equal to 0
## 95 percent confidence interval:
## -0.02967273 0.37124607
## sample estimates:
## mean in group F mean in group M
## 3.521967 3.351180
```

```
t.test(CGPA ~ Sex, data = df, var.equal = FALSE) # Welch's t-test
```

```
##
## Welch Two Sample t-test
##
## data: CGPA by Sex
## t = 1.7854, df = 70.874, p-value = 0.07848
## alternative hypothesis: true difference in means between group F and group M is not equal to 0
```

```

## 95 percent confidence interval:
## -0.01995516  0.36152849
## sample estimates:
## mean in group F mean in group M
##           3.521967      3.351180

# (ii) ANOVA
aov_result <- aov(CGPA ~ Group, data = df)
summary(aov_result) # ANOVA table

##          Df Sum Sq Mean Sq F value Pr(>F)
## Group      3 13.670   4.557   203.1 <2e-16 ***
## Residuals  76  1.705   0.022
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# (iii) Another ANOVA
aov_teacher <- aov(CGPA ~ AssignedTeacher, data = df)
summary(aov_teacher)

##          Df Sum Sq Mean Sq F value Pr(>F)
## AssignedTeacher 22   1.78  0.08091   0.339  0.997
## Residuals       57  13.60  0.23850

# Post-hoc tests
TukeyHSD(aov_result) # Tukey's HSD

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = CGPA ~ Group, data = df)
##
## $Group
##        diff      lwr      upr p adj
## 2-1 -0.2691304 -0.3851445 -0.1531164 3e-07
## 3-1 -0.5790000 -0.6950141 -0.4629859 0e+00
## 4-1 -1.2994664 -1.4436908 -1.1552420 0e+00
## 3-2 -0.3098696 -0.4258836 -0.1938555 0e+00
## 4-2 -1.0303360 -1.1745604 -0.8861116 0e+00
## 4-3 -0.7204664 -0.8646908 -0.5762420 0e+00

pairwise.t.test(df$CGPA, df$Group, p.adjust.method = "bonferroni")

##
## Pairwise comparisons using t tests with pooled SD
##
## data: df$CGPA and df$Group
##
##    1      2      3
## 2 2.5e-07 -
## 3 < 2e-16 4.9e-09 -

```

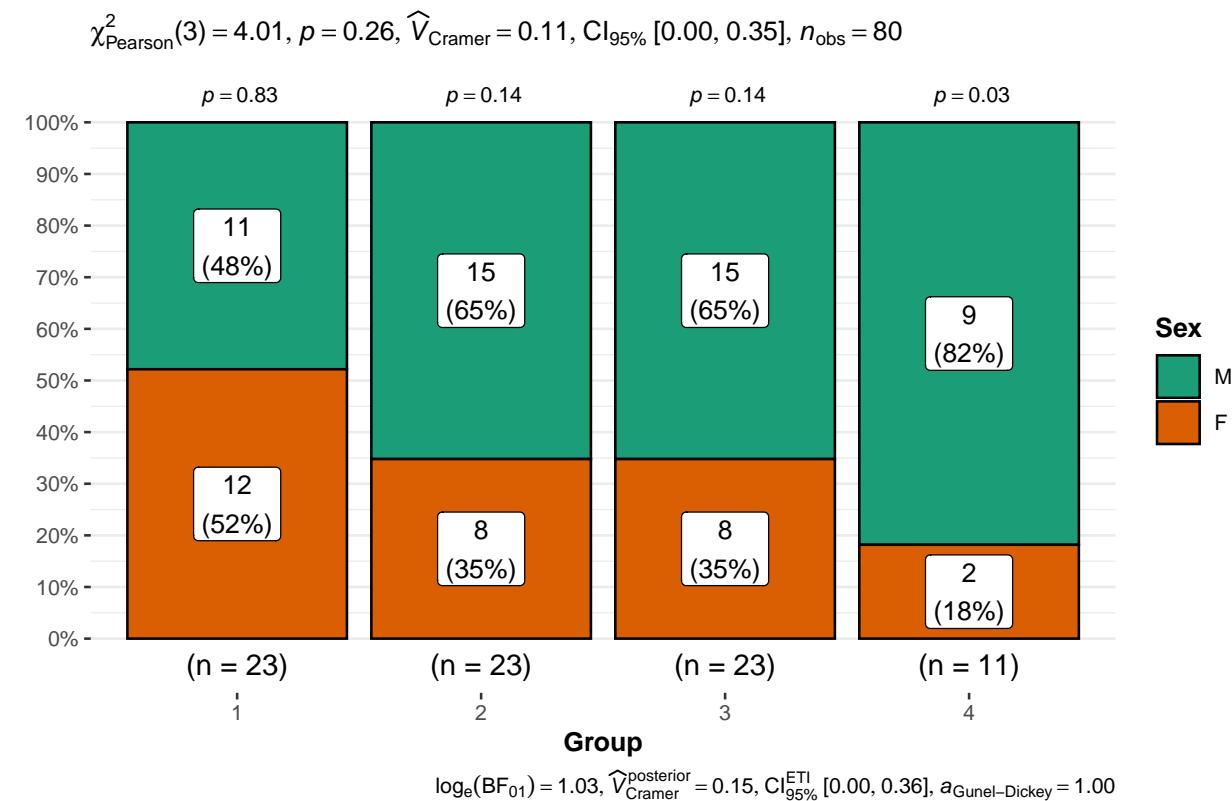
```
## 4 < 2e-16 < 2e-16 < 2e-16
##
## P value adjustment method: bonferroni
```

Association Tests : (i) Chi-square test: Sex vs Group (ii) Chi-square test: Sex vs AssignedTeacher

```
library(ggstatsplot)

# Sex vs Group with chi-square
ggbarsstats(
  data = df,
  x = Sex,
  y = Group,
  title = "Association: Sex vs Group",
  results.subtitle = TRUE,
  label = "both", # Show counts and percentages
  perc.k = 1       # Decimal places for percentages
)
```

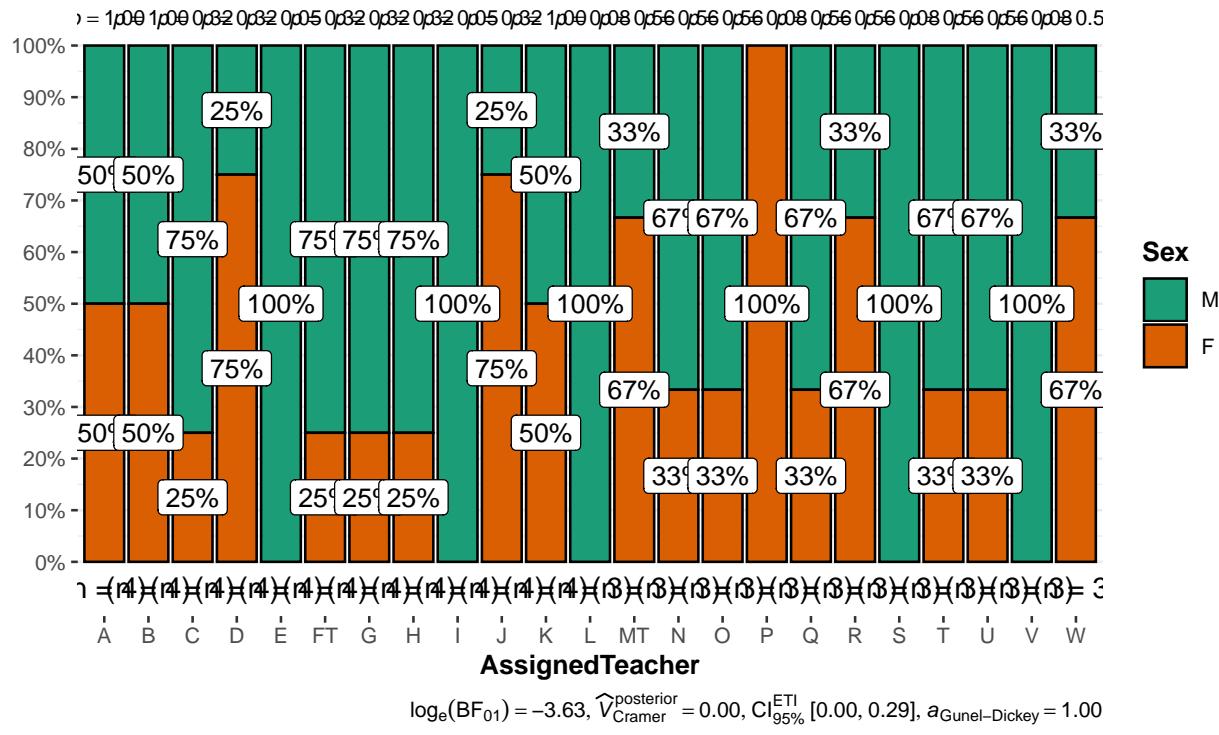
Association: Sex vs Group



```
# Sex vs AssignedTeacher
ggbarsstats(
  data = df,
  x = Sex,
  y = AssignedTeacher,
  title = "Association: Sex vs AssignedTeacher"
)
```

Association: Sex vs AssignedTeacher

$$\chi^2_{\text{Pearson}}(22) = 25.24, p = 0.29, \widehat{V}_{\text{Cramer}} = 0.19, \text{CI}_{95\%} [0.00, 0.13], n_{\text{obs}} = 80$$



Easier Way:

```
# Sex vs Group with chi-square
chisq.test(table(df$Sex, df$Group))

## Warning in chisq.test(table(df$Sex, df$Group)): Chi-squared approximation may
## be incorrect

## 
## Pearson's Chi-squared test
##
## data: table(df$Sex, df$Group)
## X-squared = 4.0095, df = 3, p-value = 0.2604

# Sex vs AssignedTeacher
chisq.test(table(df$Sex, df$AssignedTeacher))

## Warning in chisq.test(table(df$Sex, df$AssignedTeacher)): Chi-squared
## approximation may be incorrect

## 
## Pearson's Chi-squared test
##
## data: table(df$Sex, df$AssignedTeacher)
## X-squared = 25.244, df = 22, p-value = 0.2855
```

Regression Analysis

Simple Linear Regression : (i) Model 1: CGPA ~ Sex (ii) Model 2: CGPA ~ Group

```
summary(lm(CGPA ~ Sex, data = df))
```

```
##  
## Call:  
## lm(formula = CGPA ~ Sex, data = df)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max  
## -1.29818 -0.22513  0.09743  0.28478  0.60182  
##  
## Coefficients:  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  3.5220    0.0796 44.244 <2e-16 ***  
## SexM        -0.1708    0.1007 -1.696  0.0938 .  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.436 on 78 degrees of freedom  
## Multiple R-squared:  0.03557, Adjusted R-squared:  0.02321  
## F-statistic: 2.877 on 1 and 78 DF, p-value: 0.09384
```

```
summary(lm(CGPA ~ Group, data = df))
```

```
##  
## Call:  
## lm(formula = CGPA ~ Group, data = df)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max  
## -0.48527 -0.07246  0.00333  0.10026  0.44173  
##  
## Coefficients:  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  3.83774    0.03123 122.887 < 2e-16 ***  
## Group2      -0.26913    0.04417 -6.094 4.23e-08 ***  
## Group3      -0.57900    0.04417 -13.110 < 2e-16 ***  
## Group4     -1.29947    0.05491 -23.668 < 2e-16 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.1498 on 76 degrees of freedom  
## Multiple R-squared:  0.8891, Adjusted R-squared:  0.8847  
## F-statistic: 203.1 on 3 and 76 DF, p-value: < 2.2e-16
```

Multiple Regression : Full model: CGPA ~ Sex + Group + AssignedTeacher

```

# Full model
full_model <- lm(CGPA ~ Sex + Group + AssignedTeacher, data = df)
summary(full_model)

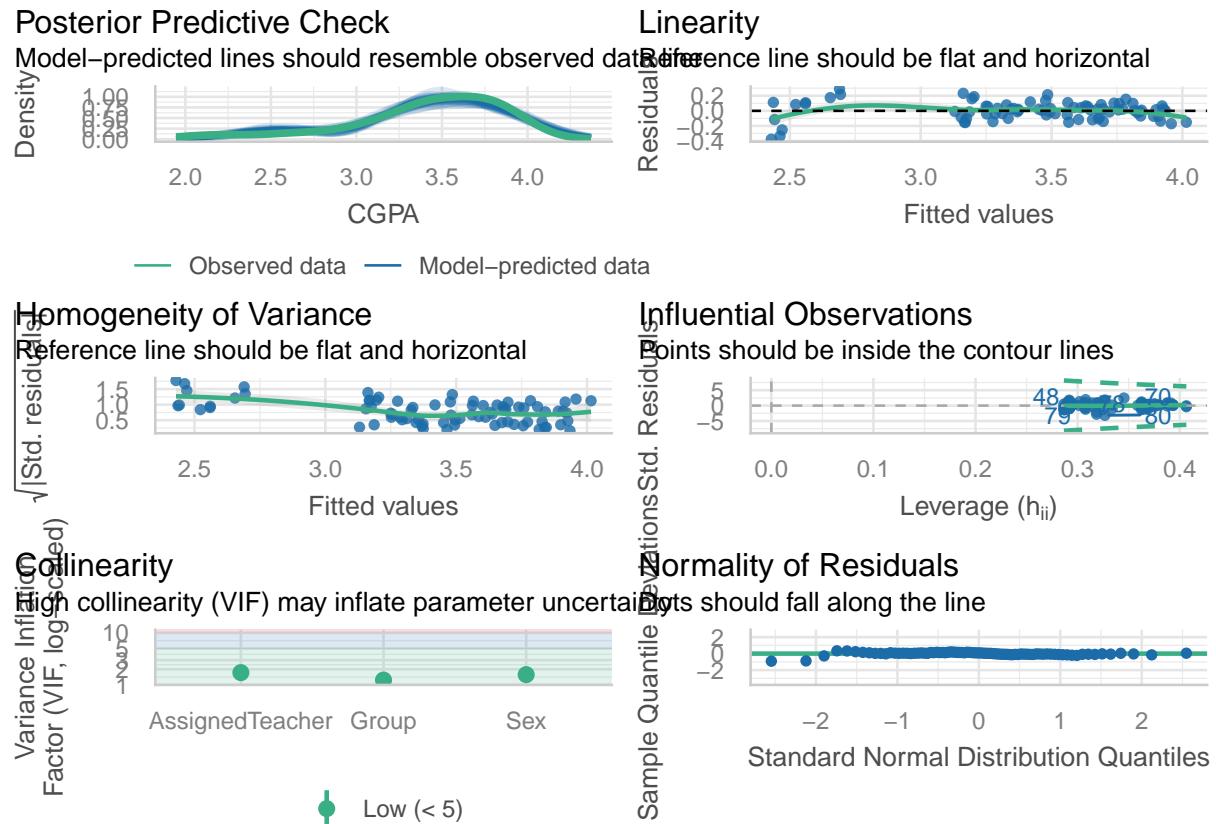
## 
## Call:
## lm(formula = CGPA ~ Sex + Group + AssignedTeacher, data = df)
## 
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -0.37629 -0.07189  0.01049  0.06907  0.29031 
## 
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 3.697292  0.079024 46.787 < 2e-16 ***
## SexM        0.053618  0.042178  1.271  0.2092    
## Group2     -0.278455  0.043500 -6.401 4.14e-08 ***
## Group3     -0.588325  0.043500 -13.525 < 2e-16 ***
## Group4     -1.321622  0.058344 -22.652 < 2e-16 ***
## AssignedTeacherB 0.225250  0.102817  2.191  0.0329 *  
## AssignedTeacherC 0.183846  0.103356  1.779  0.0810 .  
## AssignedTeacherD 0.066154  0.103356  0.640  0.5249    
## AssignedTeacherE 0.093191  0.104957  0.888  0.3786    
## AssignedTeacherFT 0.033096  0.103356  0.320  0.7501    
## AssignedTeacherG 0.041346  0.103356  0.400  0.6907    
## AssignedTeacherH 0.130846  0.103356  1.266  0.2111    
## AssignedTeacherI 0.007691  0.104957  0.073  0.9419    
## AssignedTeacherJ 0.260404  0.103356  2.519  0.0148 *  
## AssignedTeacherK 0.264250  0.102817  2.570  0.0130 *  
## AssignedTeacherL 0.005017  0.114247  0.044  0.9651    
## AssignedTeacherMT 0.202096  0.111863  1.807  0.0765 .  
## AssignedTeacherN -0.007777 0.112184 -0.069  0.9450    
## AssignedTeacherO 0.175557  0.112184  1.565  0.1236    
## AssignedTeacherP 0.143302  0.113299  1.265  0.2115    
## AssignedTeacherQ 0.021557  0.112184  0.192  0.8484    
## AssignedTeacherR 0.193096  0.111863  1.726  0.0901 .  
## AssignedTeacherS 0.087351  0.114247  0.765  0.4479    
## AssignedTeacherT 0.113557  0.112184  1.012  0.3160    
## AssignedTeacherU 0.023890  0.112184  0.213  0.8322    
## AssignedTeacherV 0.165351  0.114247  1.447  0.1537    
## AssignedTeacherW 0.211429  0.111863  1.890  0.0642 .  
## --- 
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 
## 
## Residual standard error: 0.1454 on 53 degrees of freedom
## Multiple R-squared:  0.9271, Adjusted R-squared:  0.8914 
## F-statistic: 25.93 on 26 and 53 DF,  p-value: < 2.2e-16

# All diagnostic plots
library(performance)

```

```
## Warning: package 'performance' was built under R version 4.5.2
```

```
check_model(full_model)
```



```
# Outlier Detection
```

- (i) Detect outliers in CGPA using the IQR rule
- (ii) Visualize outliers using boxplots

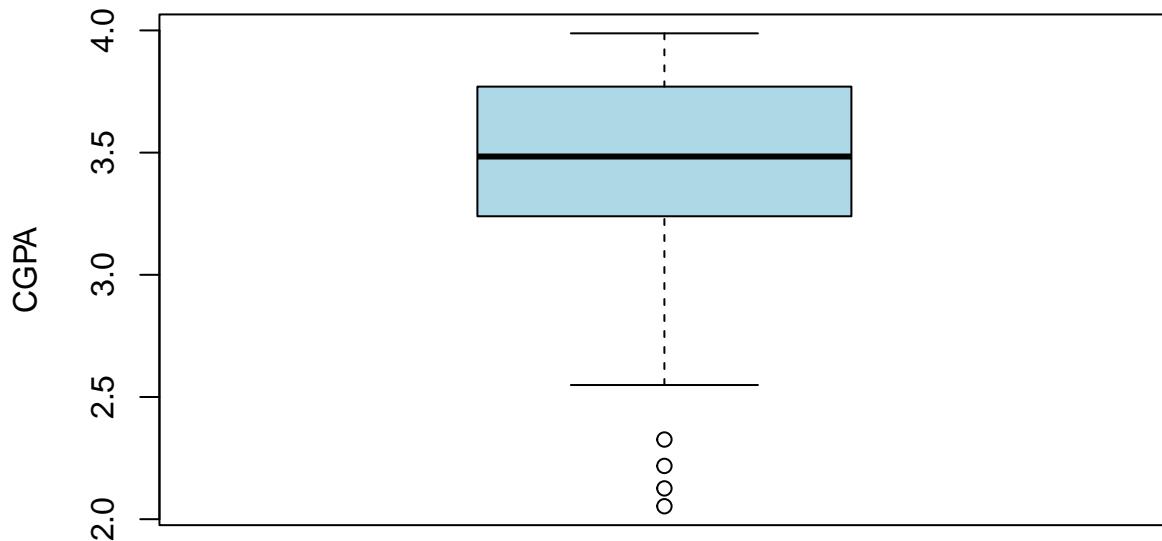
```
# Detect outliers using IQR rule
Q1 <- quantile(df$CGPA, 0.25)
Q3 <- quantile(df$CGPA, 0.75)
IQR_val <- Q3 - Q1
lower <- Q1 - 1.5 * IQR_val
upper <- Q3 + 1.5 * IQR_val

outliers <- df[df$CGPA < lower | df$CGPA > upper, ]
cat("\nOutliers detected:", nrow(outliers), "\n")
```

```
##  
## Outliers detected: 4
```

```
# Boxplot in one line
boxplot(df$CGPA, col = "lightblue", main = "CGPA Distribution with Outliers",
        ylab = "CGPA", outline = TRUE)
```

CGPA Distribution with Outliers



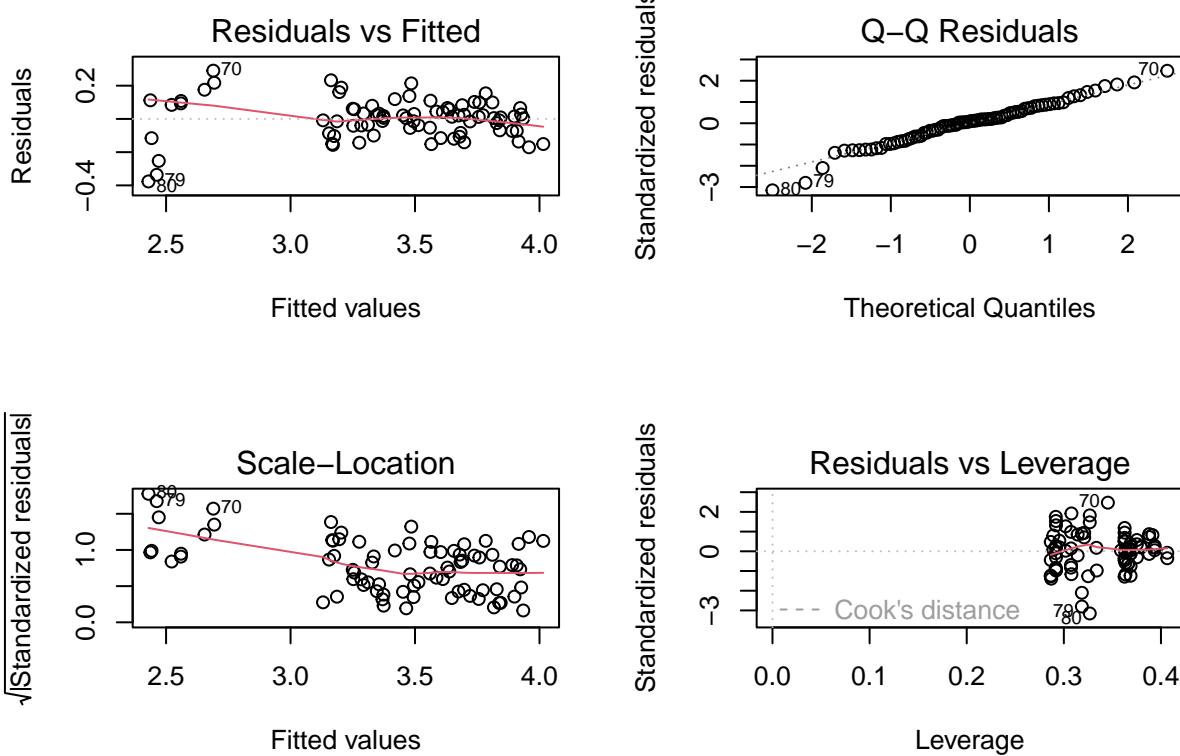
```
# Regression Diagnostics
```

For the multiple regression model ($\text{CGPA} \sim \text{Sex} + \text{Group} + \text{AssignedTeacher}$), check:

- i) Linearity
- ii) Normality
- iii) Homoscedasticity
- iv) Multicollinearity

```
# Fit the model
full_model <- lm(CGPA ~ Sex + Group + AssignedTeacher, data = df)

# 1. Diagnostic plots (Linearity, Homoscedasticity, Normality of residuals)
par(mfrow = c(2, 2))    # 2x2 layout
plot(full_model)         # Generates:
```



```

# 1. Residuals vs Fitted + checks linearity
# 2. Normal Q-Q + checks residual normality
# 3. Scale-Location + checks homoscedasticity
# 4. Residuals vs Leverage + checks influential points
par(mfrow = c(1, 1)) # Reset layout

# 2. Normality test of residuals
cat("\nShapiro-Wilk Test on Residuals:\n")

## 
## Shapiro-Wilk Test on Residuals:
shapiro.test(residuals(full_model))

##
## Shapiro-Wilk normality test
##
## data: residuals(full_model)
## W = 0.97797, p-value = 0.1824

# 3. Multicollinearity (VIF)
library(car)

## Warning: package 'car' was built under R version 4.5.2

```

```

## Loading required package: carData

## Warning: package 'carData' was built under R version 4.5.2

##
## Attaching package: 'car'

## The following object is masked from 'package:psych':
##      logit

## The following object is masked from 'package:dplyr':
##      recode

cat("\nVariance Inflation Factors (VIF):\n")

##
## Variance Inflation Factors (VIF):

vif(full_model)

##          GVIF Df GVIF^(1/(2*Df))
## Sex       1.577656  1      1.256048
## Group     1.241791  3      1.036752
## AssignedTeacher 1.723375 22      1.012447

```