

Individual_work.R

ASUS

2024-01-25

```
setwd("C:/Users/ASUS/Desktop/2nd-trimester/R")
std_data=read.csv("13-Influence of AI TOOLS on Student's Learning Process.csv",header=T)
```

```
#Dropping unwanted columns and cleaning the dataset
drop=c("Timestamp","Username","Any.Comments..Review")
std_data= std_data[,!(names(std_data) %in% drop)]
```

```
#change column_names
```

```
colnames(std_data)=c("ar","g","e","freq","access","sat_per","impt","recall","mot_ler","sat_info","sat",
str(std_data)
```

```
## 'data.frame': 158 obs. of 22 variables:
## $ ar : chr "18-24" "18-24" "18-24" "18-24" ...
## $ g : chr "Male" "Male" "Male" "Male" ...
## $ e : chr "Postgraduate" "Postgraduate" "Postgraduate" "Undergraduate" ...
## $ freq : chr "Strongly agree" "Agree" "Strongly agree" "Agree" ...
## $ access : chr "Strongly agree" "Agree" "Strongly agree" "Agree" ...
## $ sat_per : chr "Strongly agree" "Agree" "Strongly agree" "Strongly agree" ...
## $ impt : chr "Strongly Agree" "Agree" "Agree" "Neutral" ...
## $ recall : chr "Strongly Agree" "Agree" "Agree" "Agree" ...
## $ mot_ler : chr "Strongly Agree" "Agree" "Neutral" "Neutral" ...
## $ sat_info: chr "Strongly Agree" "Agree" "Disagree" "Agree" ...
## $ sat : chr "Strongly Agree" "Agree" "Strongly Agree" "Agree" ...
## $ anx : chr "Strongly Agree" "Agree" "Strongly Agree" "Agree" ...
## $ prcy : chr "Neutral" "Agree" "Disagree" "Disagree" ...
## $ saw : chr "Agree" "Agree" "Neutral" "Agree" ...
## $ p_att : chr "Neutral" "Agree" "Agree" "Neutral" ...
## $ flex : chr "Disagree" "Agree" "Disagree" "Agree" ...
## $ under : chr "Agree" "Agree" "Strongly Disagree" "Agree" ...
## $ i_feed : chr "Strongly Agree" "Agree" "Disagree" "Agree" ...
## $ m_obj : chr "Strongly Agree" "Agree" "Strongly Agree" "Agree" ...
## $ p_alter : chr "Strongly Agree" "Agree" "Strongly Agree" "Agree" ...
## $ add_s : chr "Strongly Agree" "Agree" "Disagree" "Agree" ...
## $ l_exp : chr "Strongly Agree" "Agree" "Agree" "Agree" ...
```

```
summary(std_data)
```

```
##      ar              g              e              freq
## Length:158      Length:158      Length:158      Length:158
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##      access      sat_per      impt      recall
## Length:158      Length:158      Length:158      Length:158
```

```
## Class :character   Class :character   Class :character   Class :character
## Mode :character   Mode :character   Mode :character   Mode :character
##   mot_ler         sat_info           sat           anx
## Length:158        Length:158        Length:158        Length:158
## Class :character   Class :character   Class :character   Class :character
## Mode :character   Mode :character   Mode :character   Mode :character
##   prcy            saw                p_att          flex
## Length:158        Length:158        Length:158        Length:158
## Class :character   Class :character   Class :character   Class :character
## Mode :character   Mode :character   Mode :character   Mode :character
##   under           i_feed             m_obj          p_alter
## Length:158        Length:158        Length:158        Length:158
## Class :character   Class :character   Class :character   Class :character
## Mode :character   Mode :character   Mode :character   Mode :character
##   add_s            l_exp
## Length:158        Length:158
## Class :character   Class :character
## Mode :character   Mode :character
```

```
ar=factor(std_data$ar)
g=factor(std_data$g)
e=factor(std_data$e)
```

#Graph-1

```
library(ggplot2)
library(dplyr)
```

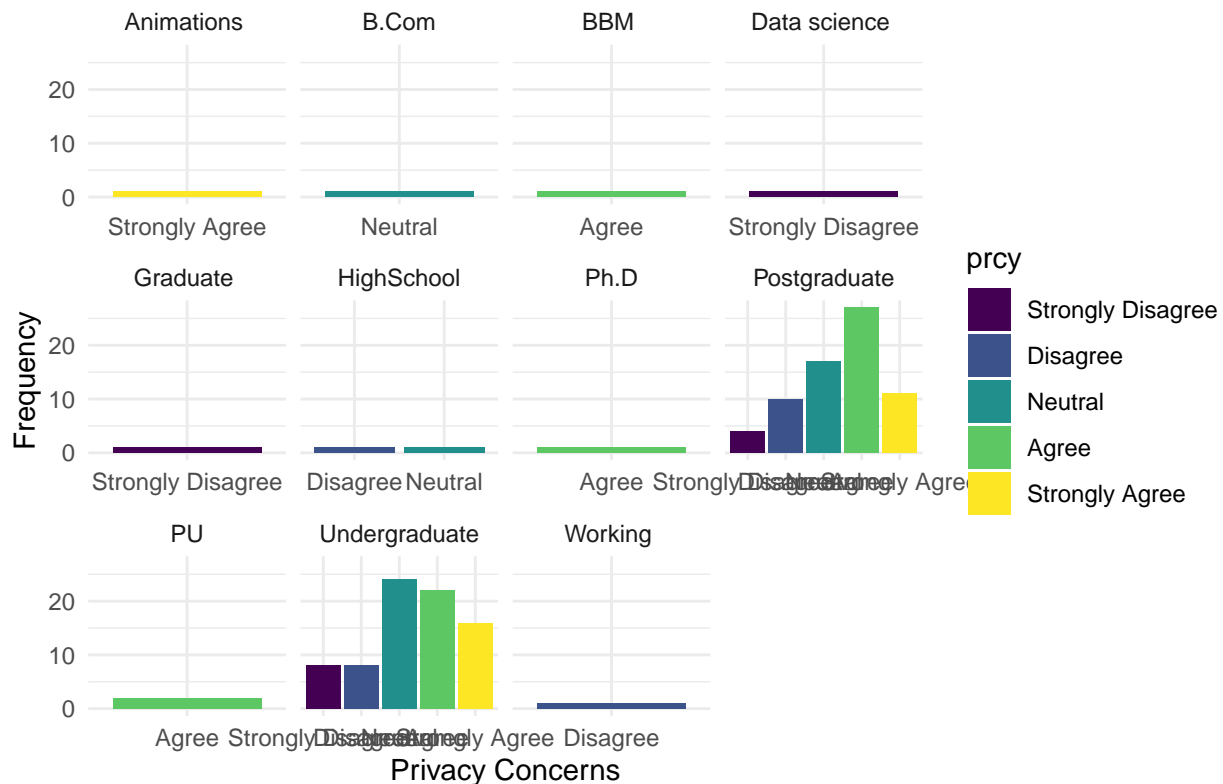
```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
std_data$prcy <- factor(std_data$prcy, ordered = TRUE,
                        levels = c("Strongly Disagree", "Disagree", "Neutral", "Agree", "Strongly Agree"))
ggplot(std_data, aes(x = prcy, fill = prcy)) +
  geom_bar(position = "stack") +
  facet_wrap(~ e, scales = "free_x") +
  labs(title = "Histogram of No Privacy Concerns by Education Level",
       x = "Privacy Concerns",
       y = "Frequency") +
  theme_minimal()
```

Histogram of No Privacy Concerns by Education Level



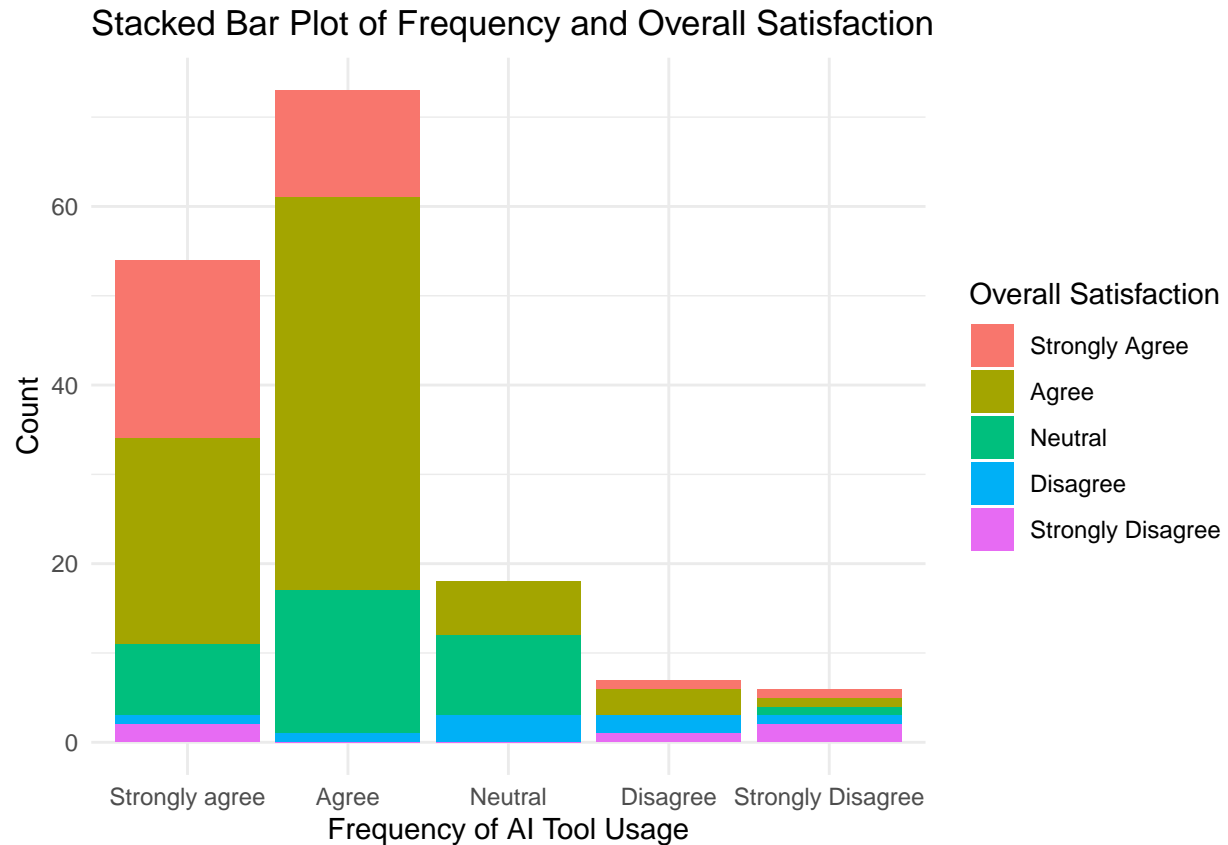
#UNDERSTANDING THE GRAPH - 1

The maximum number of responses are from respondents who are Undergraduates or Postgraduates.

Most of the respondents who frequently use AI Tools have less privacy concerns(Agree), whereas few ha

#Graph-2

```
std_data$freq <- factor(std_data$freq, ordered = TRUE, levels = c("Strongly agree", "Agree", "Neutral", "Disagree", "Strongly Disagree"))
std_data$sat <- factor(std_data$sat, ordered = TRUE, levels = c("Strongly Agree", "Agree", "Neutral", "Disagree", "Strongly Disagree"))
count_data <- as.data.frame(table(std_data$freq, std_data$sat))
colnames(count_data) <- c("Frequency", "Overall_Satisfaction", "Count")
ggplot(count_data, aes(x = Frequency, y = Count, fill = Overall_Satisfaction)) +
  geom_bar(stat = "identity", position = "stack") +
  labs(title = "Stacked Bar Plot of Frequency and Overall Satisfaction",
       x = "Frequency of AI Tool Usage",
       y = "Count",
       fill = "Overall Satisfaction") +
  theme_minimal()
```



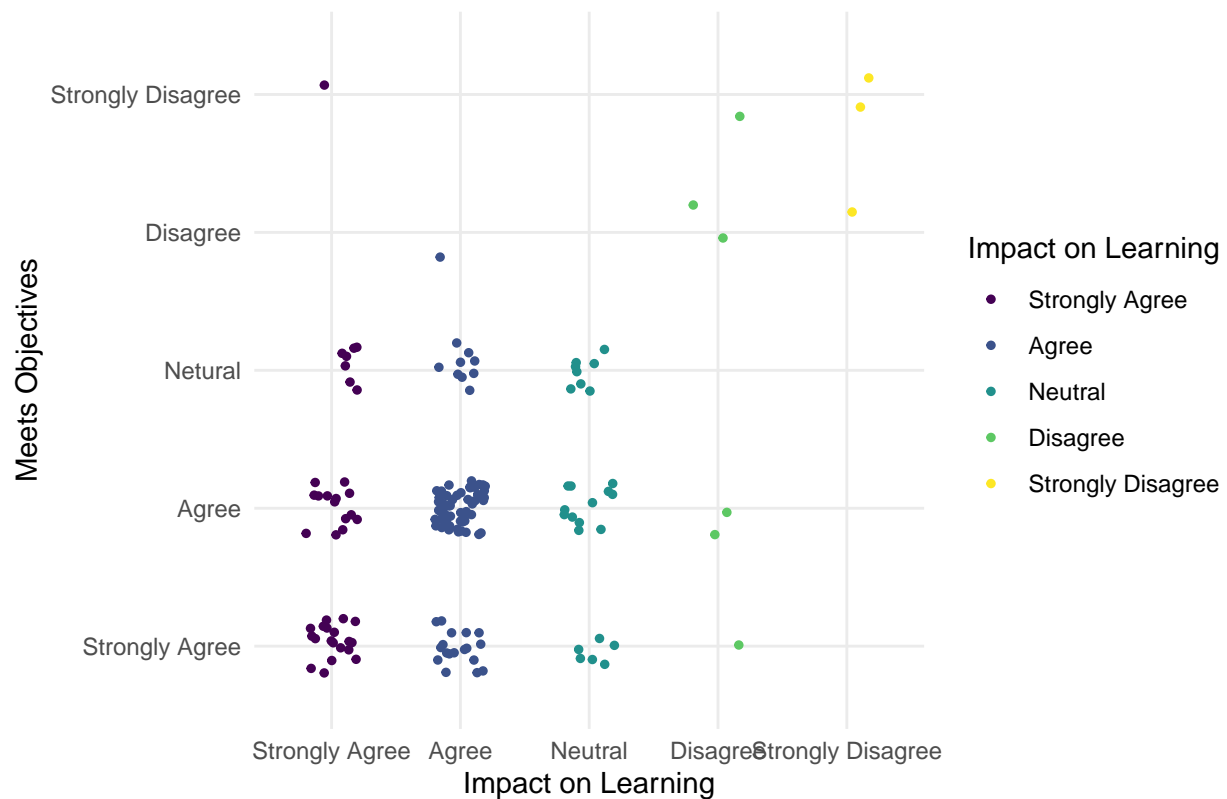
#UNDERSTANDING THE GRAPH - 2

*#Most users who use Ai Tools Frequently are overall satisfied with the services provided
 #Anyhow very few who use the AI tools are not satisfied*

#Graph-3

```
std_data$impt <- factor(std_data$impt, ordered = TRUE, levels = c("Strongly Agree", "Agree", "Neutral", "Disagree", "Strongly Disagree"))
std_data$m_obj <- factor(std_data$m_obj, ordered = TRUE, levels = c("Strongly Agree", "Agree", "Neutral", "Disagree", "Strongly Disagree"))
ggplot(std_data, aes(x = impt, y = m_obj, color = impt)) +
  geom_jitter(position = position_jitter(width = 0.2, height = 0.2), size = 1) +
  labs(title = "Jitter Dot Plot of Impact on Learning vs Meets Objectives",
       x = "Impact on Learning",
       y = "Meets Objectives",
       color = "Impact on Learning") +
  theme_minimal()
```

Jitter Dot Plot of Impact on Learning vs Meets Objectives



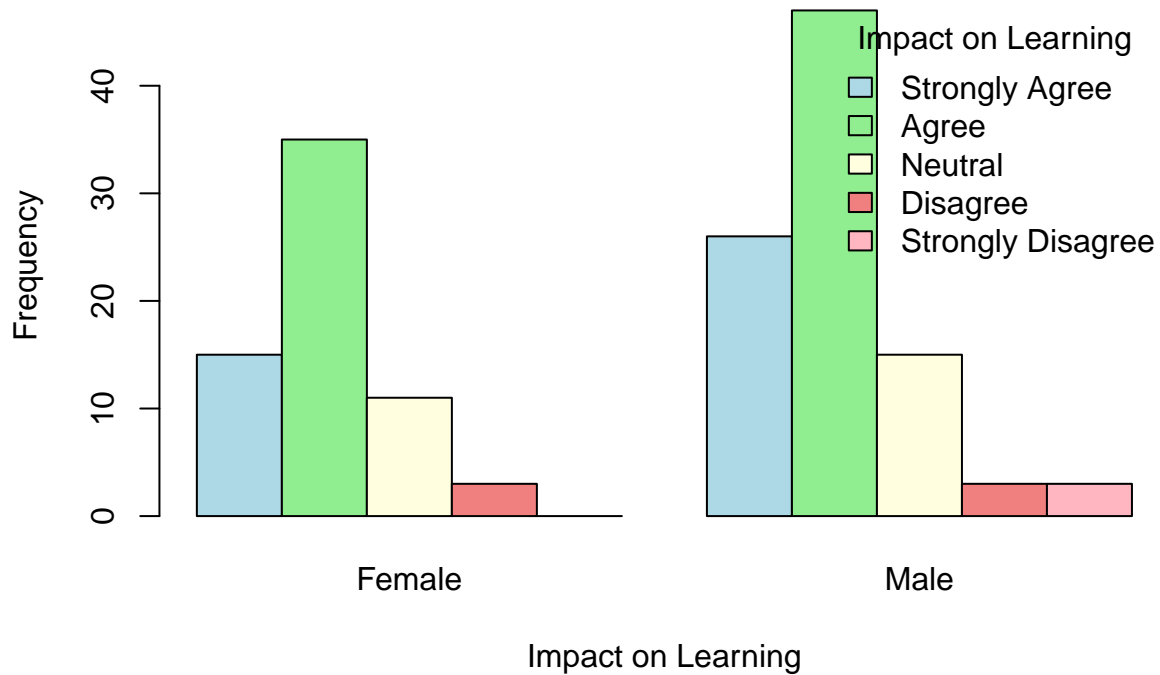
#UNDERSTANDING THE GRAPH - 3

*#Meeting the objectives of the query plays a important role in impacting the user learning experience
#Better the Response of the AI Tools leads to Better Learning experience of the User*

#Graph-4

```
barplot(t(table(std_data$g, std_data$impt)), beside = TRUE,
        col = c("lightblue", "lightgreen", "lightyellow", "lightcoral", "lightpink"),
        main = "Comparison of Gender and Impact on Learning",
        xlab = "Impact on Learning",
        ylab = "Frequency",
        names.arg = levels(std_data$g),
        legend.text = levels(std_data$impt),
        args.legend = list(x = "topright", bty = "n", title = "Impact on Learning")) +
theme(legend.position = "bottom") # Adjust legend position
```

Comparison of Gender and Impact on Learning



NULL

#UNDERSTANDING THE GRAPH - 4

#The respondents have a positive impact on their learning by using AI tool

#Very few respondents disagree that the AI tools impact their learning

#Graph-5

```
std_data$ar <- factor(std_data$ar)
```

```
std_data$m_obj <- factor(std_data$m_obj, ordered = TRUE, levels = c("Strongly Agree", "Agree", "Netural", "Neutral", "Disagree", "Strongly Disagree"))
```

```
ggplot(std_data, aes(x = ar, y = m_obj, fill = m_obj)) +
```

```
  geom_violin(trim = FALSE) +
```

```
  labs(title = "Violin Plot of Age and Meets Objectives",
```

```
        x = "Age Range",
```

```
        y = "Meets Objectives",
```

```
        fill = "Meets Objectives") +
```

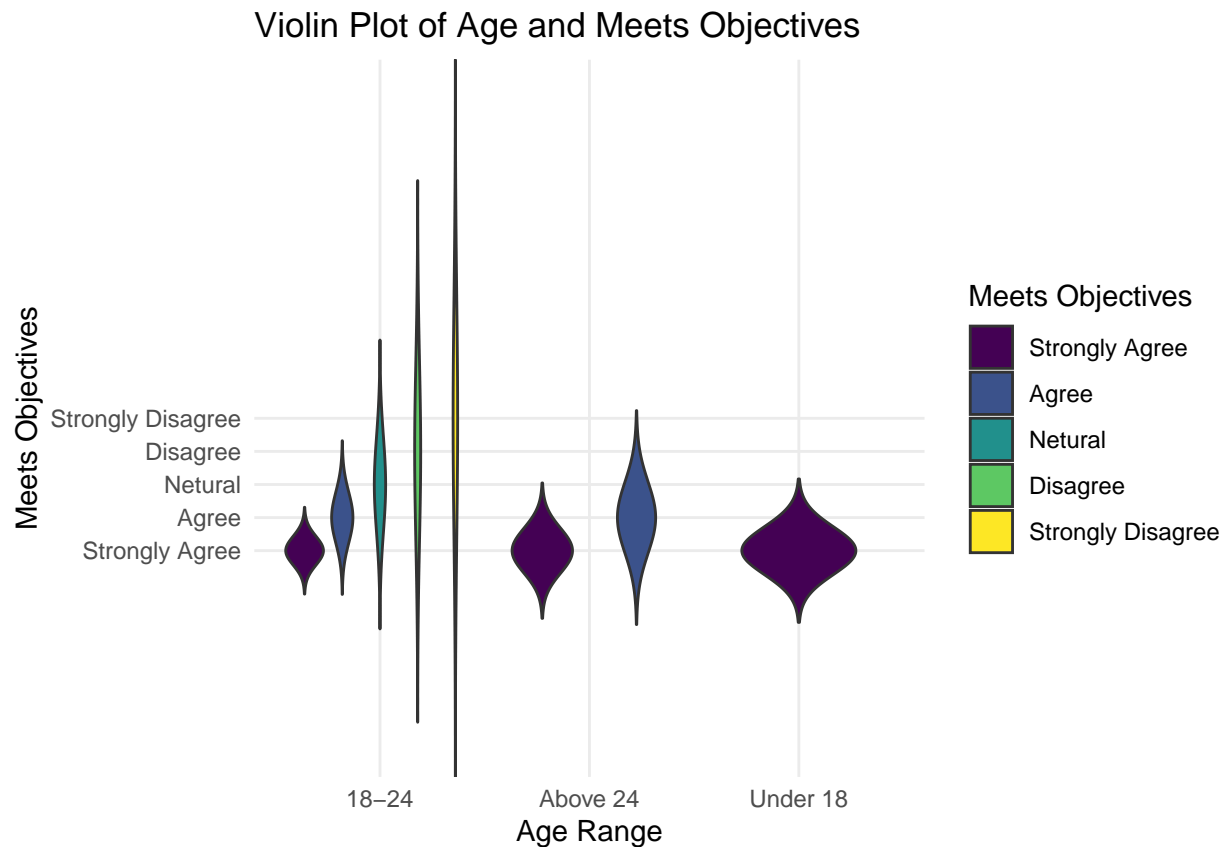
```
  theme_minimal()
```

Warning: Groups with fewer than two data points have been dropped.

Groups with fewer than two data points have been dropped.

Groups with fewer than two data points have been dropped.

Groups with fewer than two data points have been dropped.



#UNDERSTANDING THE GRAPH - 5

#Teenagers(Under-18) and above-24 aged people who use AI tools gets their responses which meets the obj
#For the people who are aged between 18-24 AI have met most of the objectives but still few people are

#converting the columns into relevant datatype

```
std_data$g<-factor(std_data$g)
std_data$ar<-factor(std_data$ar)
std_data$e<-factor(std_data$e)
```

```
library(dplyr)
```

```
map_scale_values <- function(value) {
  case_when(
    as.character(value) %in% c("Strongly Agree", "Strongly agree") ~ 5,
    as.character(value) %in% c("Agree") ~ 4,
    as.character(value) %in% c("Neutral", "Netural") ~ 3,
    as.character(value) %in% c("Disagree") ~ 2,
    as.character(value) %in% c("Strongly Disagree") ~ 1,
    TRUE ~ NA_real_ # for any other cases
  )
}
```

```
convert_columns=c("freq","access","sat_per","impt","recall","mot_ler","sat_info","sat","anx","prcy","sa
```

Apply the mapping function to specified columns

```
std_data <- std_data %>%
  mutate_at(vars(convert_columns), ~map_scale_values(.))
```

```
## Warning: Using an external vector in selections was deprecated in tidysselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
## # Was:
## data %>% select(convert_columns)
##
## # Now:
## data %>% select(all_of(convert_columns))
##
## See <https://tidysselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
std_data
```

##	ar	g	e	freq	access	sat_per	impt	recall	mot_ler
## 1	18-24	Male	Postgraduate	5	5	5	5	5	5
## 2	18-24	Male	Postgraduate	4	4	4	4	4	4
## 3	18-24	Male	Postgraduate	5	5	5	4	4	3
## 4	18-24	Male	Undergraduate	4	4	5	3	4	3
## 5	18-24	Male	Undergraduate	5	5	4	4	4	4
## 6	18-24	Male	Undergraduate	4	4	4	4	3	4
## 7	18-24	Male	Postgraduate	3	3	3	3	3	3
## 8	18-24	Female	Postgraduate	4	4	4	3	3	2
## 9	18-24	Female	Undergraduate	5	5	5	5	5	5
## 10	Under 18	Male	Undergraduate	5	5	5	5	5	4
## 11	18-24	Male	Undergraduate	4	5	5	4	3	4
## 12	18-24	Male	Undergraduate	5	5	4	4	3	4
## 13	18-24	Male	Postgraduate	5	5	1	4	3	3
## 14	18-24	Male	Undergraduate	3	5	5	4	3	4
## 15	18-24	Male	Undergraduate	5	5	5	5	4	5
## 16	18-24	Male	Undergraduate	5	4	2	4	5	5
## 17	Above 24	Male	Postgraduate	4	4	5	4	4	4
## 18	18-24	Female	Undergraduate	5	5	5	4	4	4
## 19	18-24	Female	Postgraduate	5	5	5	5	5	5
## 20	18-24	Male	Undergraduate	1	3	1	1	1	1
## 21	Above 24	Female	Postgraduate	4	4	3	4	4	4
## 22	18-24	Female	Undergraduate	4	4	3	4	3	4
## 23	18-24	Female	Undergraduate	5	5	5	5	5	5
## 24	18-24	Female	Undergraduate	2	3	3	2	2	2
## 25	18-24	Male	Data science	5	4	4	3	3	3
## 26	18-24	Male	Postgraduate	4	4	4	3	3	2
## 27	18-24	Male	Undergraduate	4	4	4	4	4	4
## 28	18-24	Male	Postgraduate	4	3	4	4	3	4
## 29	18-24	Male	Undergraduate	4	3	3	4	4	3
## 30	18-24	Male	Undergraduate	4	5	5	5	5	4
## 31	18-24	Male	Undergraduate	5	3	5	5	5	5
## 32	18-24	Female	Undergraduate	5	4	4	5	4	5
## 33	18-24	Male	B.Com	3	3	3	3	3	3
## 34	18-24	Male	Undergraduate	5	5	3	4	4	3
## 35	18-24	Female	Undergraduate	4	5	4	4	4	4

## 36	18-24	Male	Undergraduate	4	4	4	4	4	3
## 37	18-24	Male	Undergraduate	4	4	4	4	4	4
## 38	18-24	Female	Postgraduate	4	5	4	5	4	3
## 39	18-24	Male	Undergraduate	4	4	3	4	4	4
## 40	18-24	Female	Undergraduate	4	3	4	3	3	2
## 41	18-24	Male	Postgraduate	4	3	3	4	3	3
## 42	Under 18	Male	Postgraduate	5	3	4	5	2	4
## 43	18-24	Female	Postgraduate	2	2	2	3	3	3
## 44	18-24	Male	Undergraduate	4	4	4	5	4	5
## 45	Above 24	Male	Postgraduate	3	5	5	5	4	4
## 46	18-24	Female	Postgraduate	4	4	4	5	5	5
## 47	18-24	Female	Undergraduate	4	3	3	4	5	4
## 48	18-24	Female	Postgraduate	3	3	4	5	4	4
## 49	Above 24	Female	Undergraduate	4	4	4	4	3	4
## 50	18-24	Female	Postgraduate	4	4	4	4	3	4
## 51	18-24	Female	Undergraduate	5	5	5	5	5	5
## 52	18-24	Female	Undergraduate	4	5	4	4	3	3
## 53	Under 18	Male	PU	4	5	4	3	4	5
## 54	Above 24	Female	Undergraduate	4	4	3	4	3	3
## 55	18-24	Female	Undergraduate	3	4	4	4	4	4
## 56	Above 24	Male	Graduate	3	4	4	4	5	3
## 57	18-24	Female	Undergraduate	3	4	4	4	4	4
## 58	18-24	Male	Undergraduate	4	4	3	3	2	3
## 59	Above 24	Female	Ph.D	3	4	2	4	3	2
## 60	Above 24	Male	Animations	4	4	4	4	5	3
## 61	18-24	Male	Undergraduate	5	4	4	4	5	4
## 62	Above 24	Female	Postgraduate	3	4	4	4	3	3
## 63	18-24	Female	Undergraduate	5	5	5	4	5	5
## 64	18-24	Male	Undergraduate	5	5	4	4	4	4
## 65	18-24	Female	Undergraduate	4	3	5	2	3	1
## 66	18-24	Male	Postgraduate	1	2	2	5	5	5
## 67	18-24	Female	Postgraduate	5	4	3	5	4	5
## 68	18-24	Male	Postgraduate	4	4	4	4	4	4
## 69	Above 24	Male	Postgraduate	2	4	4	5	5	5
## 70	18-24	Female	Undergraduate	4	3	4	4	3	4
## 71	18-24	Male	Postgraduate	5	4	3	5	4	3
## 72	Above 24	Female	BBM	4	3	4	4	5	4
## 73	18-24	Female	Undergraduate	5	4	5	4	5	3
## 74	18-24	Female	Postgraduate	4	4	4	4	4	4
## 75	18-24	Male	Undergraduate	4	4	1	4	3	2
## 76	18-24	Female	Undergraduate	4	4	4	4	4	4
## 77	18-24	Male	Undergraduate	5	5	5	5	5	5
## 78	18-24	Male	Undergraduate	4	4	3	5	4	3
## 79	18-24	Male	Undergraduate	5	4	3	5	4	5
## 80	18-24	Male	Undergraduate	4	3	3	4	4	4
## 81	18-24	Female	Undergraduate	3	4	3	3	3	3
## 82	18-24	Male	Undergraduate	5	5	5	5	5	5
## 83	Under 18	Male	PU	5	5	5	4	3	3
## 84	18-24	Male	Undergraduate	4	5	4	5	5	5
## 85	18-24	Male	Postgraduate	5	5	3	4	5	3
## 86	18-24	Female	Postgraduate	2	4	3	4	3	4
## 87	18-24	Female	Undergraduate	5	4	4	4	5	4
## 88	18-24	Female	Postgraduate	4	5	4	4	4	4
## 89	18-24	Female	Postgraduate	5	5	4	5	3	4

## 90	18-24	Female	Undergraduate	4	4	2	3	4	4
## 91	Above 24	Male	Postgraduate	5	5	5	4	4	5
## 92	Above 24	Male	Postgraduate	4	4	4	5	5	4
## 93	Above 24	Female	Undergraduate	3	5	4	4	2	3
## 94	18-24	Female	Undergraduate	4	4	4	5	3	4
## 95	18-24	Male	Postgraduate	4	4	4	4	4	4
## 96	18-24	Male	Undergraduate	1	5	5	3	3	3
## 97	18-24	Male	Postgraduate	4	3	5	4	3	4
## 98	18-24	Male	Postgraduate	5	4	3	2	3	4
## 99	18-24	Female	Postgraduate	4	5	4	5	4	5
## 100	18-24	Male	Postgraduate	4	4	4	4	4	4
## 101	18-24	Male	Postgraduate	5	4	3	4	3	3
## 102	18-24	Male	Postgraduate	5	4	3	3	2	1
## 103	18-24	Male	Postgraduate	4	5	4	3	4	3
## 104	18-24	Male	Postgraduate	2	1	3	2	1	3
## 105	18-24	Female	Postgraduate	4	3	5	4	3	2
## 106	18-24	Male	Undergraduate	1	4	4	5	4	3
## 107	18-24	Female	Undergraduate	1	3	1	3	1	2
## 108	Above 24	Male	Working	4	5	4	5	4	4
## 109	Under 18	Female	HighSchool	5	5	3	4	4	4
## 110	18-24	Female	Undergraduate	3	3	3	3	3	2
## 111	Above 24	Male	Postgraduate	2	4	3	4	3	4
## 112	18-24	Female	HighSchool	5	5	3	4	4	4
## 113	18-24	Female	Undergraduate	4	5	5	5	4	5
## 114	18-24	Female	Undergraduate	3	3	3	4	3	4
## 115	18-24	Male	Postgraduate	5	4	4	4	3	3
## 116	18-24	Female	Undergraduate	5	5	4	4	3	3
## 117	18-24	Female	Postgraduate	5	5	4	5	5	3
## 118	18-24	Male	Undergraduate	4	4	4	4	2	3
## 119	18-24	Male	Undergraduate	5	5	5	5	5	5
## 120	18-24	Female	Undergraduate	4	4	3	4	4	5
## 121	18-24	Male	Undergraduate	5	5	5	5	5	5
## 122	18-24	Male	Postgraduate	5	5	4	4	5	4
## 123	18-24	Male	Postgraduate	5	5	5	5	5	5
## 124	18-24	Male	Undergraduate	5	5	5	5	5	5
## 125	18-24	Female	Undergraduate	4	4	4	4	4	4
## 126	18-24	Female	Postgraduate	4	4	4	4	4	4
## 127	18-24	Female	Undergraduate	4	4	4	3	3	4
## 128	18-24	Male	Postgraduate	4	4	4	4	4	4
## 129	18-24	Female	Undergraduate	4	4	4	4	4	3
## 130	18-24	Female	Undergraduate	3	3	3	3	3	3
## 131	18-24	Male	Undergraduate	5	4	3	1	2	3
## 132	18-24	Female	Undergraduate	4	5	4	4	4	3
## 133	18-24	Female	Undergraduate	2	2	2	2	2	2
## 134	18-24	Male	Postgraduate	5	5	5	5	5	5
## 135	18-24	Male	Undergraduate	5	5	4	3	4	5
## 136	18-24	Male	Undergraduate	4	4	4	4	4	4
## 137	18-24	Male	Postgraduate	4	4	4	4	4	4
## 138	18-24	Female	Undergraduate	5	5	3	5	3	5
## 139	Above 24	Male	Postgraduate	5	4	5	4	4	4
## 140	Above 24	Male	Undergraduate	1	1	1	1	1	1
## 141	18-24	Male	Postgraduate	4	4	3	3	4	4
## 142	Above 24	Male	Postgraduate	4	5	4	5	4	5
## 143	18-24	Male	Postgraduate	3	3	3	3	4	2

## 144	18-24	Female	Postgraduate	4	4	4	4	4	3				
## 145	18-24	Female	Postgraduate	4	3	4	3	4	3				
## 146	18-24	Male	Postgraduate	4	4	4	4	4	3				
## 147	18-24	Male	Postgraduate	5	4	4	4	3	3				
## 148	18-24	Male	Postgraduate	4	4	2	4	3	4				
## 149	18-24	Male	Postgraduate	5	4	5	4	5	4				
## 150	18-24	Male	Postgraduate	4	5	4	4	5	4				
## 151	18-24	Male	Postgraduate	4	4	4	3	2	2				
## 152	18-24	Female	Postgraduate	4	4	4	3	3	2				
## 153	18-24	Female	Postgraduate	3	3	3	4	3	3				
## 154	18-24	Male	Postgraduate	3	4	3	2	2	2				
## 155	18-24	Male	Postgraduate	5	5	2	4	1	5				
## 156	18-24	Male	Postgraduate	4	5	4	4	2	3				
## 157	18-24	Male	Postgraduate	4	5	5	3	3	4				
## 158	18-24	Male	Undergraduate	5	5	5	5	3	3				
##	sat_info	sat	anx	prcy	saw	p_att	flex	under	i_feed	m_obj	p_alter	add_s	l_exp
## 1	5	5	5	3	4	3	2	4	5	5	5	5	5
## 2	4	4	4	4	4	4	4	4	4	4	4	4	4
## 3	2	5	5	2	3	4	2	1	2	5	5	2	4
## 4	4	4	4	2	4	3	4	4	4	4	4	4	4
## 5	5	5	5	5	4	3	2	4	3	4	5	4	3
## 6	4	4	4	5	5	3	4	4	3	4	4	3	4
## 7	3	3	5	3	1	3	3	3	3	3	3	4	3
## 8	3	3	4	3	2	3	1	2	2	4	3	2	3
## 9	5	5	5	5	5	5	5	5	5	5	5	5	5
## 10	4	5	4	5	4	4	4	4	5	5	4	3	5
## 11	4	4	4	4	4	4	3	4	4	4	4	3	4
## 12	4	4	4	3	3	5	3	4	4	4	4	4	4
## 13	4	4	4	4	3	4	2	1	3	5	3	4	3
## 14	3	4	4	2	3	4	4	4	4	5	5	5	4
## 15	5	5	5	1	4	4	3	4	4	5	5	5	5
## 16	4	4	4	4	4	4	4	4	4	4	4	4	4
## 17	4	4	5	4	4	4	4	4	4	5	5	5	4
## 18	4	3	4	3	3	4	4	4	4	4	4	4	4
## 19	4	3	3	3	3	4	3	3	3	3	3	3	3
## 20	1	1	1	1	1	1	1	1	1	1	1	1	1
## 21	4	4	4	4	4	4	4	4	4	4	4	4	4
## 22	3	3	4	3	4	4	4	4	4	4	4	4	4
## 23	5	5	5	5	5	5	5	5	5	5	5	5	5
## 24	5	2	2	2	2	3	4	2	2	4	2	4	3
## 25	2	2	3	1	2	3	4	2	1	5	4	5	4
## 26	2	4	2	2	2	2	2	2	2	4	4	2	5
## 27	4	5	5	4	5	4	4	5	4	5	4	4	4
## 28	3	4	4	2	2	3	2	3	4	4	4	5	4
## 29	4	4	4	3	3	3	4	4	4	4	4	4	4
## 30	4	5	5	2	5	5	4	5	5	3	5	5	4
## 31	5	4	3	3	5	4	4	5	4	3	4	5	5
## 32	5	4	4	4	5	4	4	4	5	4	5	5	4
## 33	3	3	3	3	3	3	3	3	3	3	3	3	3
## 34	4	3	3	4	5	5	4	4	3	4	4	4	4
## 35	4	4	4	4	4	4	4	4	4	4	4	4	4
## 36	4	5	3	4	4	4	3	4	4	4	5	4	4
## 37	4	4	4	4	4	4	4	4	4	4	4	4	4
## 38	2	4	2	2	3	4	2	4	2	4	4	4	4

## 39	3	3	3	3	3	3	3	3	3	3	3	3	3
## 40	3	4	3	2	2	3	2	4	3	4	4	4	4
## 41	5	5	4	4	3	5	4	4	5	4	5	4	4
## 42	5	3	4	3	2	5	3	4	3	3	4	1	3
## 43	3	4	2	2	4	3	3	4	3	4	3	4	4
## 44	4	4	4	3	4	4	3	4	4	4	4	3	4
## 45	4	4	4	3	3	4	4	4	4	4	4	4	4
## 46	4	4	4	4	4	4	4	4	4	3	3	4	4
## 47	4	4	3	4	4	4	5	5	5	4	4	5	4
## 48	5	4	3	4	4	5	4	5	4	4	3	5	4
## 49	4	4	3	4	4	4	4	4	4	4	4	3	4
## 50	4	4	4	4	4	4	3	4	4	4	4	4	4
## 51	5	5	5	5	5	5	5	5	5	5	5	5	5
## 52	3	2	3	4	3	4	4	3	4	3	4	5	4
## 53	5	4	5	4	4	5	5	5	4	5	4	5	5
## 54	4	3	4	1	4	4	4	4	4	4	4	3	4
## 55	3	4	3	3	3	4	3	4	4	3	4	4	4
## 56	2	2	2	1	1	2	3	3	3	4	4	4	3
## 57	3	3	3	3	3	4	4	4	4	4	4	4	4
## 58	3	3	3	1	1	3	3	3	4	4	3	3	3
## 59	5	3	3	4	3	4	4	5	5	3	4	4	3
## 60	4	4	4	5	5	5	5	4	5	4	4	4	4
## 61	5	4	3	4	4	5	4	4	5	4	4	4	4
## 62	4	3	3	3	4	4	4	3	4	4	4	5	4
## 63	5	5	4	5	4	5	5	5	5	5	5	5	5
## 64	4	5	5	4	3	3	2	5	5	5	4	4	5
## 65	3	3	4	2	2	4	1	4	3	4	4	3	5
## 66	4	5	5	4	5	5	4	5	4	5	5	4	4
## 67	4	4	5	5	4	4	5	1	4	1	3	5	4
## 68	4	4	3	3	3	3	3	4	4	4	4	3	4
## 69	4	5	4	3	3	2	4	5	4	4	4	4	4
## 70	4	4	3	3	3	3	3	4	3	4	3	3	3
## 71	5	4	3	4	5	4	3	5	3	5	4	3	4
## 72	4	3	5	4	4	5	4	4	5	5	5	4	4
## 73	3	3	4	4	4	4	3	4	5	5	4	5	4
## 74	3	4	4	3	4	4	3	5	4	4	4	3	4
## 75	4	4	3	3	4	3	4	3	4	3	4	3	4
## 76	4	4	4	4	4	4	4	4	4	4	4	4	4
## 77	5	5	5	5	5	5	5	5	5	5	5	5	5
## 78	2	4	3	5	4	4	5	4	4	4	4	5	3
## 79	3	4	4	3	4	5	4	4	4	4	4	5	4
## 80	4	4	4	4	4	4	4	4	4	4	4	4	4
## 81	4	3	3	3	3	3	3	3	3	3	3	3	3
## 82	5	5	5	3	4	5	3	5	5	5	5	5	5
## 83	3	4	5	4	5	4	3	3	4	5	5	5	5
## 84	4	4	5	4	4	5	3	4	2	3	4	4	4
## 85	3	4	5	5	4	3	3	4	3	5	5	4	3
## 86	4	4	3	4	4	4	4	4	4	4	3	3	4
## 87	5	5	4	5	4	4	4	4	5	4	5	4	4
## 88	4	4	5	5	5	5	5	5	5	4	4	4	4
## 89	2	4	3	4	3	3	3	4	5	4	4	5	4
## 90	3	4	4	3	2	4	1	2	3	3	3	3	3
## 91	5	4	4	5	4	5	4	5	5	5	5	4	4
## 92	5	4	5	5	4	4	4	5	5	4	5	5	4

## 93	3	4	2	4	2	4	4	5	4	4	5	5	4
## 94	5	5	3	1	5	5	5	5	5	5	5	5	5
## 95	4	4	4	4	4	4	4	4	4	4	4	4	4
## 96	3	3	3	3	2	2	2	2	2	3	3	4	3
## 97	4	3	5	3	5	3	4	5	4	4	4	4	5
## 98	5	4	3	3	4	5	4	3	4	5	4	3	4
## 99	3	5	4	4	4	4	5	5	4	5	5	4	4
## 100	4	4	4	4	4	4	4	4	4	4	4	4	4
## 101	2	4	4	4	4	4	4	4	4	5	4	4	4
## 102	5	1	3	4	2	5	3	2	2	4	3	3	1
## 103	4	4	4	3	3	4	3	3	4	4	4	4	4
## 104	2	1	1	1	2	3	2	2	1	1	2	1	2
## 105	4	3	3	1	2	2	1	2	1	4	4	4	4
## 106	3	4	5	5	3	4	4	4	3	4	5	3	4
## 107	4	2	1	1	1	3	4	4	2	3	3	2	2
## 108	4	4	3	2	4	3	4	4	4	4	4	3	4
## 109	4	3	3	3	3	4	3	4	4	4	4	3	4
## 110	4	4	4	3	4	4	3	3	3	4	3	4	4
## 111	5	4	3	2	4	4	4	5	5	4	4	3	4
## 112	3	5	4	2	3	4	4	5	3	5	5	4	4
## 113	5	5	5	2	4	2	2	4	4	4	5	5	5
## 114	3	3	4	3	3	4	3	3	3	3	4	4	4
## 115	4	4	3	2	4	3	3	3	3	5	4	4	3
## 116	3	3	3	3	2	3	2	2	3	4	4	3	3
## 117	5	4	5	3	5	5	5	5	4	5	4	5	4
## 118	3	5	5	3	3	5	4	4	5	4	4	3	4
## 119	5	5	5	5	5	5	5	5	5	5	5	5	5
## 120	5	4	2	1	4	4	4	5	5	4	4	4	5
## 121	5	5	5	5	5	5	5	5	5	5	5	5	5
## 122	3	4	4	4	4	3	3	3	3	4	4	4	5
## 123	5	5	5	5	5	5	5	5	5	5	5	5	5
## 124	5	5	5	5	5	5	5	5	5	5	5	5	5
## 125	4	4	4	4	4	4	4	4	4	4	4	4	4
## 126	2	3	3	2	4	3	3	4	3	4	4	3	3
## 127	3	4	4	3	3	4	3	4	3	4	3	3	3
## 128	4	4	4	5	5	4	4	4	5	4	4	4	4
## 129	3	4	4	4	4	4	4	4	4	4	4	4	4
## 130	3	3	3	3	3	3	3	3	3	3	3	3	3
## 131	3	4	2	5	1	2	2	5	4	2	3	4	1
## 132	4	5	3	3	3	3	3	4	4	5	4	4	4
## 133	2	2	2	2	2	2	2	2	2	2	2	2	2
## 134	5	5	5	5	4	4	4	4	5	5	5	5	5
## 135	4	4	5	5	5	5	3	3	4	5	5	5	5
## 136	4	4	4	4	4	4	4	4	4	4	4	4	4
## 137	4	4	4	4	4	4	4	4	4	4	4	4	4
## 138	2	3	5	3	1	4	3	3	1	3	4	4	3
## 139	4	4	4	4	4	4	4	5	5	5	5	4	4
## 140	1	1	1	1	1	1	1	1	1	1	1	1	1
## 141	3	5	4	3	4	4	4	4	4	5	4	5	4
## 142	5	5	5	5	5	5	5	5	5	4	4	4	4
## 143	4	2	5	4	5	5	4	5	5	5	4	5	5
## 144	2	4	4	4	4	4	3	3	4	4	4	4	4
## 145	4	3	4	4	3	3	4	3	4	4	3	4	3
## 146	4	3	3	4	4	3	3	3	3	3	3	3	3

```
## 147      4  4  3  4  4  3  2  3  3  3  4  5  5
## 148      5  3  4  1  3  4  3  4  4  4  4  4  3
## 149      5  4  5  4  5  4  5  4  5  4  5  4  5
## 150      5  5  4  5  4  4  5  4  5  5  4  5  4
## 151      3  3  3  3  3  3  4  3  3  3  4  4  4
## 152      2  3  3  2  4  3  2  3  4  4  3  2  4
## 153      3  3  4  3  4  4  3  3  3  3  4  4  4
## 154      2  2  4  2  2  2  2  2  2  2  2  2  2
## 155      2  1  4  1  1  1  1  4  4  2  4  3  3
## 156      2  3  2  4  4  5  3  5  3  4  4  5  3
## 157      3  4  5  5  4  5  3  5  5  5  5  5  5
## 158      4  5  5  4  5  5  5  5  5  5  5  5  5
```

```
section_2_columns=c("freq", "sat", "sat_per")
section_3_columns=c("impt", "recall", "mot_ler")
section_4_columns=c("sat", "anx", "prcy", "saw", "p_att", "flex")
section_5_columns=c("under", "i_feed", "m_obj", "p_alter", "add_s", "l_exp")
# Summative score calculation
std_data$section_2_score <- rowSums(select(std_data,section_2_columns), na.rm = TRUE)
```

```
## Warning: Using an external vector in selections was deprecated in tidysselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
## # Was:
## data %>% select(section_2_columns)
##
## # Now:
## data %>% select(all_of(section_2_columns))
##
## See <https://tidysselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
std_data$section_3_score <- rowSums(select(std_data,section_3_columns), na.rm = TRUE)
```

```
## Warning: Using an external vector in selections was deprecated in tidysselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
## # Was:
## data %>% select(section_3_columns)
##
## # Now:
## data %>% select(all_of(section_3_columns))
##
## See <https://tidysselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
std_data$section_4_score <- rowSums(select(std_data,section_4_columns), na.rm = TRUE)
```

```
## Warning: Using an external vector in selections was deprecated in tidysselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
## # Was:
## data %>% select(section_4_columns)
##
## # Now:
```

```
## data %>% select(all_of(section_4_columns))
##
## See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
std_data$section_5_score <- rowSums(select(std_data,section_5_columns), na.rm = TRUE)
```

```
## Warning: Using an external vector in selections was deprecated in tidyselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
## # Was:
## data %>% select(section_5_columns)
##
## # Now:
## data %>% select(all_of(section_5_columns))
##
## See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
# Display the updated data frame with the summative score
head(std_data)
```

	ar	g	e	freq	access	sat_per	impt	recall	mot_ler	sat_info	sat
## 1	18-24	Male	Postgraduate	5	5	5	5	5	5	5	5
## 2	18-24	Male	Postgraduate	4	4	4	4	4	4	4	4
## 3	18-24	Male	Postgraduate	5	5	5	4	4	3	2	5
## 4	18-24	Male	Undergraduate	4	4	5	3	4	3	4	4
## 5	18-24	Male	Undergraduate	5	5	4	4	4	4	5	5
## 6	18-24	Male	Undergraduate	4	4	4	4	3	4	4	4

	anx	prcy	saw	p_att	flex	under	i_feed	m_obj	p_alter	add_s	l_exp
## 1	5	3	4	3	2	4	5	5	5	5	5
## 2	4	4	4	4	4	4	4	4	4	4	4
## 3	5	2	3	4	2	1	2	5	5	2	4
## 4	4	2	4	3	4	4	4	4	4	4	4
## 5	5	5	4	3	2	4	3	4	5	4	3
## 6	4	5	5	3	4	4	3	4	4	3	4

	section_2_score	section_3_score	section_4_score	section_5_score
## 1	15	15	22	29
## 2	12	12	24	24
## 3	15	11	21	19
## 4	13	10	21	24
## 5	14	12	24	23
## 6	12	11	25	22

```
#1.One Sample T-Test
df=data.frame(std_data)
prcymean=mean(std_data$prcy)
# Null Hypothesis (H0):
# The mean of the variable 'prcy' in section-4 is equal to the hypothesized population mean.
# Mathematically: ?_prcy = 3.386076 (where ? represents the population mean)

# Alternative Hypothesis (Ha or H1):
# The mean of the variable 'prcy' in section-4 is not equal to the hypothesized population mean.
```

```
# Mathematically:  $\mu_{prcy} \approx 3.386076$  (where  $\mu$  represents the population mean)
```

```
t.test(std_data$prcy, mu = prcymean)
```

```
##
```

```
## One Sample t-test
```

```
##
```

```
## data: std_data$prcy
```

```
## t = 0, df = 157, p-value = 1
```

```
## alternative hypothesis: true mean is not equal to 3.386076
```

```
## 95 percent confidence interval:
```

```
## 3.201145 3.571007
```

```
## sample estimates:
```

```
## mean of x
```

```
## 3.386076
```

```
# The p-value of 1 is greater than any common significance level (e.g., 0.05), indicating that there is no statistically significant difference between the mean of 'prcy' in the dataset and the hypothesized mean.
```

```
#2. Two sample T-test
```

```
# Null Hypothesis (H0):
```

```
# There is no significant difference in the mean satisfaction scores (sat_per) between undergraduate and postgraduate students.
```

```
# Mathematically:  $\mu(\text{undergrad}) = \mu(\text{postgrad})$  (where  $\mu$  represents the population mean).
```

```
#
```

```
# Alternative Hypothesis (Ha or H1):
```

```
# There is a significant difference in the mean satisfaction scores (sat_per) between undergraduate and postgraduate students.
```

```
# Mathematically:  $\mu(\text{undergrad}) \neq \mu(\text{postgrad})$  (where  $\mu$  represents the population mean).
```

```
mean(df$e=="Undergraduate")
```

```
## [1] 0.4936709
```

```
mean(df$e=="Postgraduate")
```

```
## [1] 0.4367089
```

```
undergrad_data <- std_data$sat_per[std_data$e == "Undergraduate"]
```

```
postgrad_data <- std_data$sat_per[std_data$e == "Postgraduate"]
```

```
t_test_result <- t.test(undergrad_data, postgrad_data)
```

```
print(t_test_result)
```

```
##
```

```
## Welch Two Sample t-test
```

```
##
```

```
## data: undergrad_data and postgrad_data
```

```
## t = 0.14766, df = 144.28, p-value = 0.8828
```

```
## alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
```

```
## -0.2899636 0.3367864
```

```
## sample estimates:
```

```
## mean of x mean of y
```

```
## 3.820513 3.797101
```

```
# The p-value of 0.8828 is greater than common significance levels (e.g., 0.05), indicating that there is no statistically significant difference between the mean satisfaction scores of undergraduate and postgraduate students.
```


#3. Performing one-way ANOVA

Null Hypothesis (H0):

#

There is no significant difference in the mean satisfaction scores (sat_per) among different education levels.

Mathematically: $\mu_1 = \mu_2 = \dots = \mu_k$ (where μ_i represents the population mean for each education level, and k is the number of education levels).

#

Alternative Hypothesis (Ha or H1):

#

There is a significant difference in the mean satisfaction scores (sat_per) among at least two education levels.

Mathematically: At least one μ_i is different (where i represents each education level).

```
anova_result <- aov(sat_per ~ e, data = std_data)
```

```
print(anova_result)
```

```
## Call:
```

```
## aov(formula = sat_per ~ e, data = std_data)
```

```
##
```

```
## Terms:
```

```
## e Residuals
```

```
## Sum of Squares 6.37239 137.14660
```

```
## Deg. of Freedom 10 147
```

```
##
```

```
## Residual standard error: 0.9659038
```

```
## Estimated effects may be unbalanced
```

#The p-value associated with the F-statistic from the ANOVA test is not provided in the output.

#Without the exact p-value, it's not possible to determine the statistical significance of the ANOVA test.

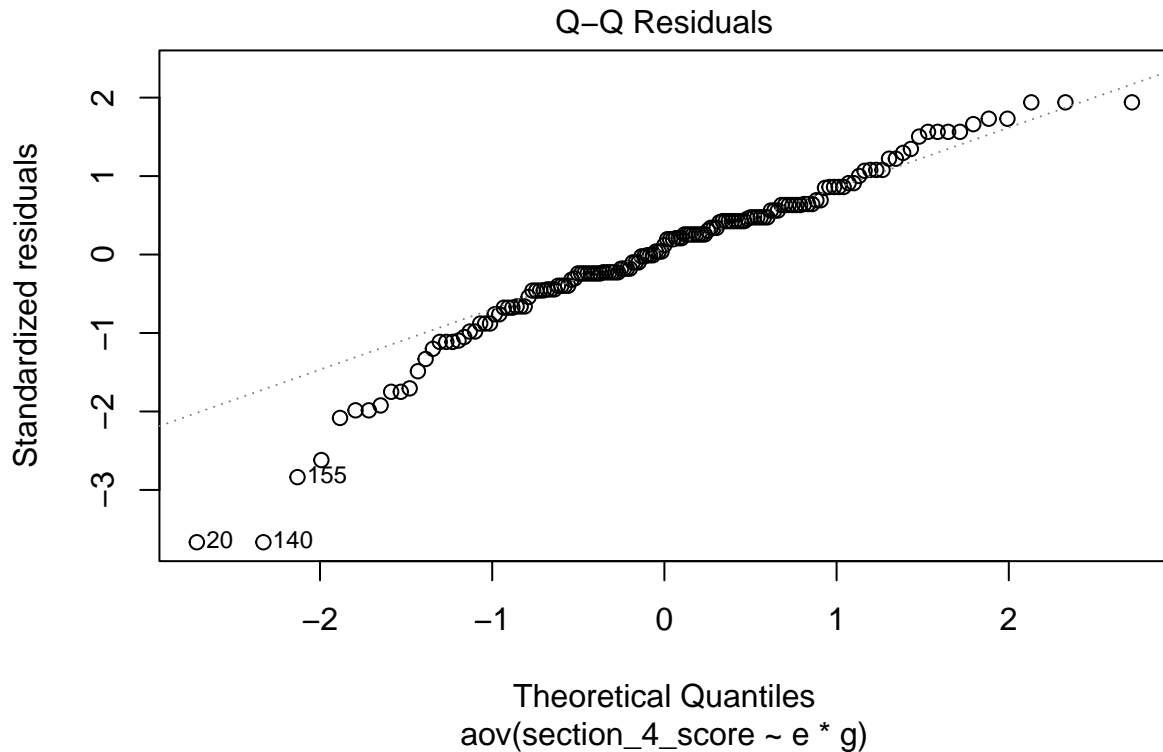
#4. Two-way ANOVA test

```
mod <- aov(section_4_score ~ e * g,  
           data = std_data)
```

```
plot(mod, which = 2)
```

```
## Warning: not plotting observations with leverage one:
```

```
## 25, 33, 56, 59, 60, 72, 108
```



```
summary(mod)
```

```
##           Df Sum Sq Mean Sq F value Pr(>F)
## e          10  270.9   27.09   1.254  0.262
## g           1   52.0   51.96   2.406  0.123
## e:g          1   10.9   10.85   0.502  0.480
## Residuals  145 3131.2    21.59
```

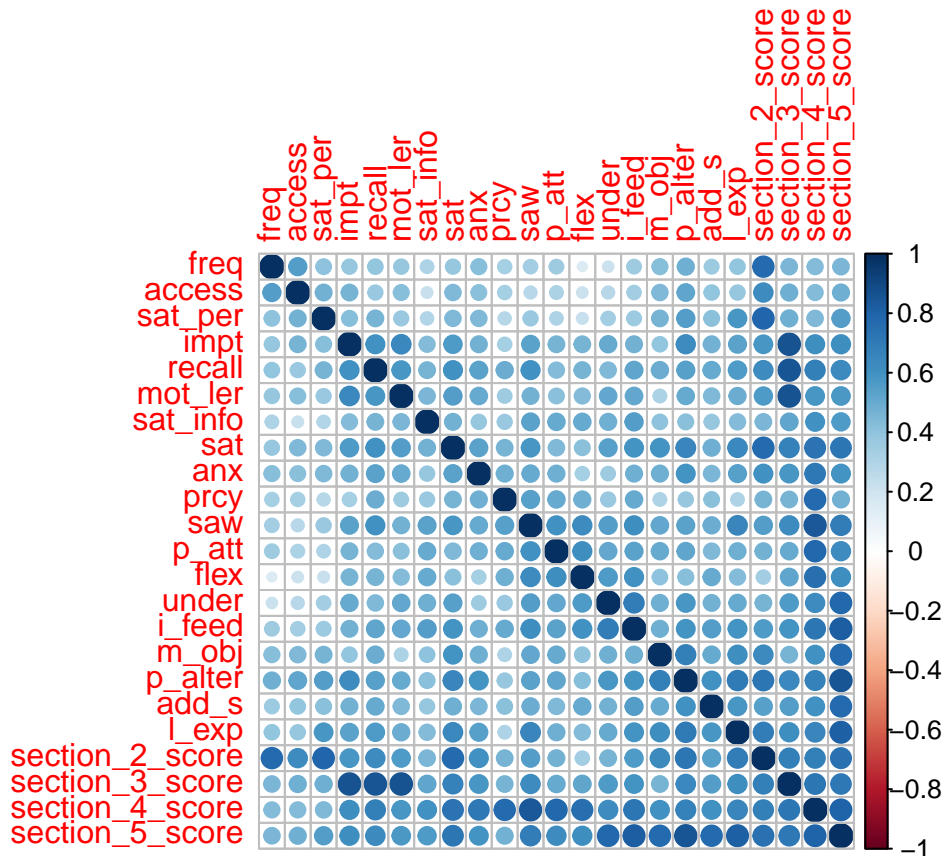
#The p-value for 'e' is 0.262, which is greater than the significance level of 0.05. Therefore, we fail to reject the null hypothesis.
#The p-value for 'g' is 0.123, which is greater than 0.05. We fail to reject the null hypothesis, indicating no significant effect of gender on the score.
#The p-value for the interaction term 'e:g' is 0.480, which is greater than 0.05. We fail to reject the null hypothesis, indicating no significant interaction between education level and gender.
#Based on the analysis, there is no significant evidence to suggest that education level, gender, or their interaction significantly affect the score.

#5. Correlation plot

```
library("corrplot")
```

```
## corrplot 0.92 loaded
```

```
d = subset(std_data, select = -c(ar,e,g) )
M=cor(d)
corrplot(M,method="circle")
```



*#The above correlation plot displays the connection between each column in the dataset.
 #There seems to be no negative co-relations in the dataset.
 #There is very weak co-relation between many columns in the dataset.
 #Each Section summative score seems to have normal positive correlation with each other column*

```
# One-Sample T-Test
t_test_result <- t.test(std_data$sat_per, mu = 4)
print(t_test_result)
```

```
##
## One Sample t-test
##
## data: std_data$sat_per
## t = -2.6627, df = 157, p-value = 0.00856
## alternative hypothesis: true mean is not equal to 4
## 95 percent confidence interval:
## 3.647229 3.947708
## sample estimates:
## mean of x
## 3.797468
```

```
# Two-Sample T-Test
undergrad_data <- std_data$sat_per[std_data$e == "Undergraduate"]
postgrad_data <- std_data$sat_per[std_data$e == "Postgraduate"]
t_test_result <- t.test(undergrad_data, postgrad_data)
print(t_test_result)
```

```
##
## Welch Two Sample t-test
##
## data: undergrad_data and postgrad_data
## t = 0.14766, df = 144.28, p-value = 0.8828
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.2899636 0.3367864
## sample estimates:
## mean of x mean of y
## 3.820513 3.797101
```

```
# One-Way ANOVA
anova_result <- aov(sat_per ~ ar, data = std_data)
```

```
# Print summary to get the p-value
summary(anova_result)
```

```
##           Df Sum Sq Mean Sq F value Pr(>F)
## ar          2    0.84   0.4186   0.455  0.635
## Residuals 155 142.68   0.9205
```