Descriptive Statistics and Ordinal Regressions

2024-03-05

#Install packages

```
options(repos = c(CRAN = "https://cran.rstudio.com/"))
install.packages("dplyr")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'dplyr' successfully unpacked and MD5 sums checked
## Warning: cannot remove prior installation of package 'dplyr'
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying
## M:\R\Win-Library\4.2\00LOCK\dplyr\libs\x64\dplyr.dll to
## M:\R\Win-Library\4.2\dplyr\libs\x64\dplyr.dll: Permission denied
## Warning: restored 'dplyr'
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded packages
install.packages("lubridate")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'lubridate' successfully unpacked and MD5 sums checked
## Warning: cannot remove prior installation of package 'lubridate'
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying
## M:\R\Win-Library\4.2\00LOCK\lubridate\libs\x64\lubridate.dll to
## M:\R\Win-Library\4.2\lubridate\libs\x64\lubridate.dll: Permission denied
## Warning: restored 'lubridate'
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded packages
install.packages("ggplot2")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
```

```
## package 'ggplot2' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded packages
install.packages("tidyr")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'tidyr' successfully unpacked and MD5 sums checked
## Warning: cannot remove prior installation of package 'tidyr'
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying
## M:\R\Win-Library\4.2\00LOCK\tidyr\libs\x64\tidyr.dll to
## M:\R\Win-Library\4.2\tidyr\libs\x64\tidyr.dll: Permission denied
## Warning: restored 'tidyr'
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded packages
install.packages("scales")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'scales' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded_packages
install.packages("MASS")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'MASS' successfully unpacked and MD5 sums checked
## Warning: cannot remove prior installation of package 'MASS'
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying
## M:\R\Win-Library\4.2\00LOCK\MASS\libs\x64\MASS.dll to
## M:\R\Win-Library\4.2\MASS\libs\x64\MASS.dll: Permission denied
## Warning: restored 'MASS'
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded packages
```

```
library(dplyr)
## Warning: package 'dplyr' was built under R version 4.2.3
## 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(lubridate)
## Warning: package 'lubridate' was built under R version 4.2.3
##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
       date, intersect, setdiff, union
library(ggplot2)
## Warning: package 'ggplot2' was built under R version 4.2.3
library(tidyr)
## Warning: package 'tidyr' was built under R version 4.2.3
library(scales)
## Warning: package 'scales' was built under R version 4.2.3
library(MASS)
## Warning: package 'MASS' was built under R version 4.2.3
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##
       select
data <- read.csv ("datarecord.csv")</pre>
head(data)
##
     Paper.No.
## 1
```

```
## 2
             2
             3
## 3
             4
## 4
## 5
             5
## 6
             6
##
Title
                                  A metabolomic landscape of maize plants trea
## 1
ted with a microbial biostimulant under well-watered and drought conditions
## 2
                           Streamlined identification of strain engineering tar
gets for bioprocess improvement using metabolic pathway enrichment analysis
                                                          On-line untargeted me
tabolomics monitoring of an Escherichia coli succinate fermentation process
                          A novel targeted/untargeted GC-Orbitrap metabolomics
methodology applied to Candida albicans and Staphylococcus aureus biofilms
                                 Glycolysis and pyrimidine biosynthesis are re
quired for replication of adherent-invasive escherichia coli in macrophages
## 6 Substantial Extracellular Metabolic Differences Found Between Phylogenet
ically Closely Related Probiotic and Pathogenic Strains of Escherichia coli
##
                             DOI
                                    ResGrp.PI Type Institution
## 1
       10.3389/fpls.2021.676632 Karl Burgess IQB3
                                                           SBS
## 2 10.1038/s41598-023-39661-x Karl Burgess IQB3
                                                           SBS
## 3
              10.1002/bit.28173 Karl Burgess IQB3
                                                           SBS
## 4
      10.1007/s11306-016-1134-2 Karl Burgess IQB3
                                                           SBS
## 5
           10.1099/mic.0.000289 Karl Burgess IQB3
                                                           SBS
## 6
       10.3389/fmicb.2019.00252 Karl Burgess IQB3
                                                           SBS
##
                               Journal Year AnalysisPgrm CodeArchived DAS
## 1
           Frontiers in Plant Science 2021
                                                        1
                                                                     NA
                                                                          1
## 2
                   Scientific reports 2023
                                                        1
                                                                     NA
                                                                          1
## 3 Biotechnology and Bioengineering 2022
                                                        1
                                                                     NA
                                                                          1
## 4
                          Metabolomics 2016
                                                        1
                                                                         NA
                                                                     NA
## 5 Biotechnology and Bioengineering 2016
                                                        1
                                                                         NA
                                                                     NA
## 6
            Frontiers in Microbiology 2019
                                                        1
                                                                          1
     CorresAuthor Preprint Complete Reuse Access Licence Image Genomics
##
## 1
                1
                          0
                                   2
                                          3
                                                 4
                                                         4
                                                               NA
                                                                        NA
                1
                          0
                                   4
                                         4
                                                         4
## 2
                                                 4
                                                               NA
                                                                        NA
                1
                          0
                                   3
                                         4
                                                 4
                                                         4
## 3
                                                               NA
                                                                        NA
## 4
                1
                          0
                                   4
                                         4
                                                 2
                                                         4
                                                               NA
                                                                        NΑ
                1
                          0
                                   2
                                         2
                                                 2
                                                         4
                                                                0
## 5
                                                                        NA
                                   4
                                                 4
                                                         4
## 6
                                                               NA
                                                                        NA
##
     Proteometabolomic
                                                    Data.Storage Funding.Source
## 1
                                          Supplementary material
                                                                               1
                           Supplementary material, MetaboLights
## 2
                      1
                                                                               0
                           Supplementary material, MetaboLights
                      0
                                                                               0
## 3
                      1
## 4
                                         Supplementary material
                                                                               0
## 5
                    NA
                                         Supplementary material
                                                                               1
## 6
                      1 Supplementary material, GNPS repository
                                                                               0
     Number.of.times.cited.according.to.Google.Scholar CitationsPA
##
## 1
                                                      40
                                                             13.33333
## 2
                                                             2.00000
```

```
## 3
                                                           6
                                                                  3.00000
## 4
                                                          56
                                                                  7.00000
## 5
                                                           5
                                                                  0.62500
                                                          29
## 6
                                                                  5.80000
     Competing.interests PIGroupLeader
##
## 1
                         1
## 2
                         0
                                         1
## 3
                                         1
                        NA
## 4
                         1
                                         1
## 5
                        NA
                                         0
## 6
```

#1 - Descriptive Statistics

#How many papers in each type

```
type_frequency <- table(data$Type)
print(type_frequency)

##
## IQB3 Other
## 80 50</pre>
```

#frequency and percentage of paper in each year

```
total_counts_by_type <- data %>%
  group by(Type) %>%
  summarise(Total = n(), .groups = "drop")
# Calculate frequency and percentage for each Type and Year
summarized_data <- data %>%
  group_by(Year, Type) %>%
  summarise(count = n(), .groups = "drop") %>%
  left_join(total_counts_by_type, by = "Type") %>%
  mutate(Percentage = (count / Total) * 100) # Removed the trailing %>%
print(summarized_data)
## # A tibble: 20 × 5
##
      Year Type count Total Percentage
##
      <int> <chr> <int> <int><</pre>
                                   <dbl>
## 1 2014 IQB3
                      8
                           80
                                   10
## 2 2014 Other
                      7
                           50
                                   14
## 3
      2015 IQB3
                      7
                           80
                                    8.75
                      5
## 4
                           50
      2015 Other
                                   10
## 5
                      5
                           80
      2016 IOB3
                                    6.25
                      8
                           50
## 6
      2016 Other
                                   16
                      9
                           80
##
  7
      2017 IOB3
                                   11.2
## 8
      2017 Other
                      3
                           50
                                    6
## 9
                      7
                                    8.75
      2018 IQB3
                           80
## 10 2018 Other
                           50
```

```
## 11 2019 IOB3
                      9
                            80
                                    11.2
## 12 2019 Other
                      4
                            50
                                     8
                      7
                                     8.75
## 13
       2020 IQB3
                            80
## 14
       2020 Other
                      7
                            50
                                    14
## 15
       2021 IQB3
                      8
                            80
                                    10
## 16
       2021 Other
                      5
                            50
                                    10
## 17
       2022 IOB3
                     11
                            80
                                    13.8
## 18
                      3
       2022 Other
                            50
                                     6
                      9
## 19
                            80
                                    11.2
       2023 IQB3
## 20 2023 Other
                      4
                            50
                                     8
```

Calculate the total number of papers for each year

```
yearly totals <- data %>%
  group_by(Year) %>%
  summarise(Total = n(), .groups = "drop")
# Calculate the grand total of all papers
grand_total <- sum(yearly_totals$Total)</pre>
# Add a column for the percentage of each year's total relative to the grand
total
yearly totals <- yearly totals %>%
  mutate(Percentage = (Total / grand_total) * 100)
# Print the yearly totals and percentages
print(yearly_totals)
## # A tibble: 10 × 3
##
      Year Total Percentage
##
      <int> <int>
                       <dbl>
## 1 2014
               15
                       11.5
## 2
      2015
               12
                        9.23
## 3
      2016
               13
                       10
                        9.23
## 4
      2017
               12
## 5
      2018
                        8.46
               11
      2019
               13
## 6
                       10
   7
      2020
               14
                       10.8
##
## 8
       2021
               13
                       10
##
   9
       2022
               14
                       10.8
## 10
      2023
               13
                       10
```

#to categorize papers based on the year of publication compared to FAIR principles implementation 2016 (overall)

```
data <- data %>%
   mutate(Period = ifelse(Year <= 2016, "Before or in 2016", "After 2016"))
# Calculate the total number of papers for each period
period_totals <- data %>%
```

```
group by(Period) %>%
  summarise(Total = n(), .groups = "drop")
# Calculate the grand total of all papers
grand_total <- sum(period_totals$Total)</pre>
# Add a column for the percentage of each period's total relative to the gran
d total
period totals <- period totals %>%
  mutate(Percentage = (Total / grand_total) * 100)
# Print the period totals and percentages
print(period totals)
## # A tibble: 2 × 3
     Period
                       Total Percentage
##
     <chr>
                                  <dbl>
                       <int>
## 1 After 2016
                          90
                                   69.2
## 2 Before or in 2016
                                   30.8
                        40
```

#to categorize papers based on the year of publication compared to FAIR principles implementation 2016 (depend on type)

```
#Frequency and Percantage of paper before and After FAIR principles (2016)
total_counts_by_type <- data %>%
  group by(Type) %>%
  summarise(Total = n(), .groups = "drop")
# Categorize, calculate frequency and percentage for each Type based on year
summarized data <- data %>%
  mutate(Category = ifelse(Year <= 2016, "On/Before 2016", "After 2020")) %>%
  group_by(Category, Type) %>%
  summarise(count = n(), .groups = "drop") %>%
  left_join(total_counts_by_type, by = "Type") %>%
  mutate(Percentage = (count / Total) * 100)
print(summarized_data)
## # A tibble: 4 × 5
##
     Category
                    Type count Total Percentage
##
                    <chr> <int> <int>
                                           <dbl>
     <chr>>
## 1 After 2020
                                              75
                    IQB3
                             60
                                   80
## 2 After 2020
                                   50
                                              60
                    Other
                             30
## 3 On/Before 2016 IOB3
                             20
                                   80
                                              25
                                              40
## 4 On/Before 2016 Other
                             20
                                   50
```

#to categorise papers based on the year of publication compared to COVID 19 - 2020 (overall)

```
data <- data %>%
  mutate(Period = ifelse(Year <= 2020, "Before or in 2020", "After 2020"))</pre>
# Calculate the total number of papers for each period
period totals <- data %>%
  group by(Period) %>%
  summarise(Total = n(), .groups = "drop")
# Calculate the grand total of all papers
grand_total <- sum(period_totals$Total)</pre>
# Add a column for the percentage of each period's total relative to the gran
d total
period totals <- period totals %>%
  mutate(Percentage = (Total / grand_total) * 100)
# Print the period totals and percentages
print(period_totals)
## # A tibble: 2 × 3
##
     Period
                       Total Percentage
##
     <chr>>
                        <int>
                                   <dbl>
## 1 After 2020
                           40
                                    30.8
## 2 Before or in 2020
                           90
                                    69.2
```

#to categorise papers based on the year of publication compared to COVID 19 - 2020 (IQB3 vs Other)

```
total_counts_by_type <- data %>%
  group by(Type) %>%
  summarise(Total = n(), .groups = "drop")
# Categorize, calculate frequency and percentage for each Type based on year
summarized data <- data %>%
  mutate(Category = ifelse(Year <= 2020, "On/Before 2020", "After 2020")) %>%
  group_by(Category, Type) %>%
  summarise(count = n(), .groups = "drop") %>%
  left_join(total_counts_by_type, by = "Type") %>%
  mutate(Percentage = (count / Total) * 100)
print(summarized data)
## # A tibble: 4 × 5
                    Type count Total Percentage
##
     Category
##
                    <chr> <int> <int>
                                           <dbl>
     <chr>
## 1 After 2020
                    IOB3
                             28
                                   80
                                              35
## 2 After 2020
                             12
                                   50
                                              24
                    Other
## 3 On/Before 2020 IQB3
                             52
                                   80
                                              65
## 4 On/Before 2020 Other 38
                                   50
                                              76
```

Frequency and Percentage of the papers that share data (Data Sharing= Data Completeness score >1) Overall

```
data <- data %>%
 mutate(Period = ifelse(Complete > 1, "Completeness > 1", "Completeness = 1"
))
# Calculate the total number of papers for each period
period totals <- data %>%
 group_by(Period) %>%
 summarise(Total = n(), .groups = "drop")
# Calculate the grand total of all papers
grand total <- sum(period totals$Total)</pre>
# Add a column for the percentage of each period's total relative to the gran
d total
period_totals <- period_totals %>%
 mutate(Percentage = (Total / grand total) * 100)
# Print the period totals and percentages
print(period_totals)
## # A tibble: 2 × 3
##
    Period
                     Total Percentage
##
                                 <dbl>
    <chr>
                     <int>
## 1 Completeness = 1
                        6
                                 4.62
## 2 Completeness > 1 124
                                 95.4
```

Frequency and Percentage of the papers that share data (Data Sharing= Data Completeness score >1) by Type

```
total_counts_by_type <- data %>%
  group_by(Type) %>%
  summarise(Total = n(), .groups = "drop")

# Categorize based on Completeness, calculate frequency and percentage for ea ch Type
summarized_data <- data %>%
  filter(Complete == 1 | Complete > 1) %>%
  mutate(Category = ifelse(Complete == 1, "Completeness = 1", "Completeness > 1")) %>%
  group_by(Category, Type) %>%
  summarise(count = n(), .groups = "drop") %>%
  left_join(total_counts_by_type, by = "Type") %>%
  mutate(Percentage = (count / Total) * 100)

print(summarized_data)
```

```
## # A tibble: 4 × 5
##
                      Type count Total Percentage
     Category
                      <chr> <int> <int>
##
     <chr>>
                                             <dbl>
## 1 Completeness = 1 IQB3
                                3
                                     80
                                              3.75
## 2 Completeness = 1 Other
                               3
                                     50
                                              6
## 3 Completeness > 1 IQB3
                               77
                                     80
                                             96.2
## 4 Completeness > 1 Other
                               47
                                     50
                                             94
```

#Chi square test to assess if there is any difference in data share between the two types (IQB3 vs Other)

```
# Create a contingency table
table completeness <- table(data$Type, data$Complete > 1)
# Perform the Chi-square test
result <- chisq.test(table_completeness)</pre>
## Warning in chisq.test(table_completeness): Chi-squared approximation may b
## incorrect
#Print the chi-square results
print(result)
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table completeness
## X-squared = 0.027302, df = 1, p-value = 0.8688
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
              FALSE
                        TRUE
##
     IQB3 3.692308 76.30769
    Other 2.307692 47.69231
##
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
```

Categorical Data: Both variables should be categorical (either nominal or ordinal).

#Frequency and percentage of Preprint

```
# Calculate frequency and percentage for each Preprint category, across all t
ypes
Preprint summary <- data %>%
  group by(Preprint) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall total <- sum(Preprint summary$count)</pre>
# Add a column for the percentage of each Preprint category relative to the o
verall total
Preprint summary <- Preprint summary %>%
  mutate(percentage = (count / overall_total) * 100)
# Print the summary
print(Preprint summary)
## # A tibble: 2 × 3
     Preprint count percentage
##
        <int> <int>
                         <dbl>
                          84.6
## 1
            0
                110
## 2
            1
                 20
                          15.4
# Frequency and Percentage of the papers with Preprint
data %>%
  group by(Type, Preprint) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, Preprint)
## # A tibble: 4 × 5
##
    Type Preprint count total percentage
              <int> <int> <int>
    <chr>
## 1 IOB3
                  0
                       69
                                      86.2
                             80
## 2 IQB3
                  1
                       11
                             80
                                      13.8
## 3 Other
                       41
                             50
                                       82
                  0
## 4 Other
                  1
                             50
                                       18
```

#Chi square test to assess if there is any difference in preprint between the two types (IQB3, Other)

```
# Create a contingency table
table_preprint <- table(data$Type, data$Preprint)</pre>
```

```
# Perform the Chi-square test
result <- chisq.test(table preprint)</pre>
#Print the chi-square results
print(result)
##
   Pearson's Chi-squared test with Yates' continuity correction
##
##
## data: table preprint
## X-squared = 0.16287, df = 1, p-value = 0.6865
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
                  0
##
    IOB3 67.69231 12.307692
##
    Other 42.30769 7.692308
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
```

Calculate frequency and percentage for DAS

```
DAS_summary <- data %>%
    group_by(DAS) %>%
    summarise(count = n(), .groups = "drop")

# Calculate the overall total
overall_total <- sum(DAS_summary$count)

# Add a column for the percentage of each DAS category relative to the overal
l total
DAS_summary <- DAS_summary %>%
    mutate(percentage = (count / overall_total) * 100)
```

```
print(DAS_summary)
## # A tibble: 3 × 3
       DAS count percentage
##
     <int> <int>
                      <dbl>
##
             8
## 1
        0
                       6.15
              48
                      36.9
## 2
        1
## 3
        NA
              74
                      56.9
data$DAS <- as.factor(data$DAS)</pre>
# Frequency and Percentage of the papers with DAS in the publications
data %>%
  group_by(Type, DAS) %>%
  summarise(count = n(), .groups = "drop") %>%
  group by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, DAS)
## # A tibble: 6 × 5
##
    Type DAS count total percentage
                                  <dbl>
## <chr> <fct> <int> <int>
## 1 IQB3 0
                    6
                          80
                                    7.5
## 2 IQB3 1
                    28
                          80
                                   35
## 3 IQB3 <NA>
                    46
                          80
                                   57.5
                     2
## 4 Other 0
                          50
                                    4
## 5 Other 1
                    20
                          50
                                   40
## 6 Other <NA>
                    28
                          50
                                   56
```

#Chi square test to assess if there is any difference in DAS between the two types (IQB3, Other)

```
# Create a contingency table
table_das <- table(data$Type, data$DAS)

# Perform the Chi-square test
result <- chisq.test(table_das)

## Warning in chisq.test(table_das): Chi-squared approximation may be incorre
ct

#Print the chi-square results
print(result)

##
## Pearson's Chi-squared test with Yates' continuity correction
##</pre>
```

```
## data: table das
## X-squared = 0.25267, df = 1, p-value = 0.6152
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
##
     IQB3 4.857143 29.14286
##
    Other 3.142857 18.85714
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
```

Calculate frequency and percentage for CorresAuthor

```
CorresAuthor summary <- data %>%
  group by(CorresAuthor) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall_total <- sum(CorresAuthor_summary$count)</pre>
# Add a column for the percentage of each category relative to the overall to
tal
CorresAuthor summary <- CorresAuthor summary %>%
  mutate(percentage = (count / overall_total) * 100)
print(CorresAuthor_summary)
## # A tibble: 2 × 3
    CorresAuthor count percentage
            <int> <int>
                             <dbl>
##
## 1
                0
                     21
                              16.2
## 2
                1
                    109
                              83.8
```

```
# Frequency and Percentage of the papers with CorresAuthor being 0 or 1
data %>%
  group_by(Type, CorresAuthor) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, CorresAuthor)
## # A tibble: 4 × 5
##
     Type CorresAuthor count total percentage
                  <int> <int> <int>
     <chr>
## 1 IQB3
                      0
                           14
                                  80
                                           17.5
## 2 IQB3
                      1
                            66
                                  80
                                           82.5
## 3 Other
                            7
                      0
                                  50
                                           14
## 4 Other
                      1
                           43
                                  50
                                           86
```

Calculate frequency and percentage for Code Sharing

```
CodeArchived summary <- data %>%
  group_by(CodeArchived) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall total <- sum(CodeArchived summary$count)</pre>
# Add a column for the percentage of each Code Archived category relative to
the overall total
CodeArchived summary <- CodeArchived summary %>%
  mutate(percentage = (count / overall total) * 100)
print(CodeArchived summary)
## # A tibble: 3 × 3
##
     CodeArchived count percentage
##
            <int> <int>
                             <dbl>
## 1
                0
                      3
                              2.31
## 2
                1
                     17
                             13.1
                    110
                             84.6
               NA
# Frequency and Percentage of the papers that shared the code used
data %>%
  group_by(Type, CodeArchived) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
```

```
arrange(Type, CodeArchived)
## # A tibble: 6 × 5
     Type CodeArchived count total percentage
                  <int> <int> <int>
##
     <chr>>
                                          <dbl>
## 1 IQB3
                            2
                                  80
                                            2.5
                      0
## 2 IQB3
                      1
                            12
                                  80
                                           15
## 3 IQB3
                     NA
                            66
                                  80
                                           82.5
## 4 Other
                      0
                            1
                                  50
                                            2
## 5 Other
                      1
                             5
                                  50
                                           10
## 6 Other
                     NA
                            44
                                  50
                                           88
```

#Chi square test to assess if there is any difference in Code Sharing between the two types (IQB3, Other)

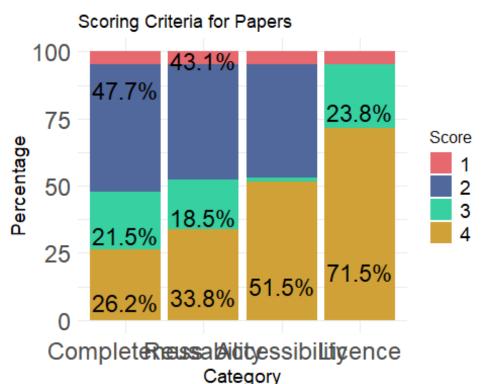
```
# Create a contingency table
table_code <- table(data$Type, data$CodeArchived)</pre>
# Perform the Chi-square test
result <- chisq.test(table code)</pre>
## Warning in chisq.test(table_code): Chi-squared approximation may be incorr
ect
#Print the chi-square results
print(result)
##
## Pearson's Chi-squared test with Yates' continuity correction
## data: table_code
## X-squared = 2.1717e-31, df = 1, p-value = 1
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected_values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
             0
                  1
     IQB3 2.1 11.9
##
    Other 0.9 5.1
##
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
```

Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
Random Sampling: The data is a random sample.
Categorical Data: Both variables should be categorical (either nominal or
ordinal).

#Plot the 4 scoring criteria for all the papers

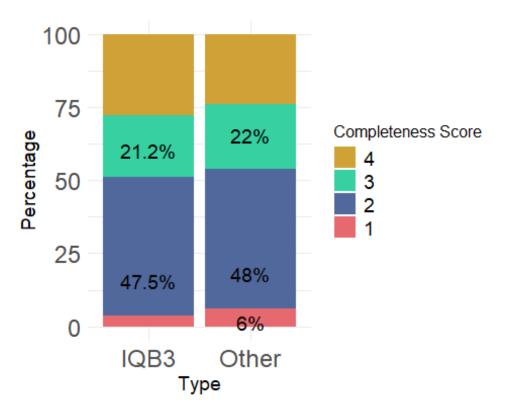
```
# Convert all relevant columns to factors first to avoid the error
data <- data %>%
  mutate(across(c(Complete, Reuse, Access, Licence), as.character), .groups =
"drop")
# Reshape data to Long format
long_data <- data %>%
  pivot_longer(cols = c(Complete, Reuse, Access, Licence), names_to = "Catego"
ry", values to = "Value") %>%
  mutate(Value = as.character(Value))
# Rename and reorder categories
long_data$Category <- recode(long_data$Category,</pre>
                             "Complete" = "Completeness",
                             "Reuse" = "Reusability",
                             "Access" = "Accessibility",
                             "Licence" = "Licence")
long_data$Category <- factor(long_data$Category, levels = c("Completeness", "</pre>
Reusability", "Accessibility", "Licence"))
# Calculate counts and percentages ensuring percentages do not exceed 100%
long_data <- long_data %>%
  group by(Category, Value) %>%
  summarise(Count = n(), .groups = "drop") %>%
  ungroup() %>%
  group_by(Category) %>%
  mutate(Total = sum(Count),
         Percentage = pmin(Count / Total * 100, 100)) %>%
  ungroup() %>%
  arrange(Category, desc(Value)) %>%
  group by(Category) %>%
  mutate(Label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "%"), ""
),
         CumPercentage = pmin(cumsum(Percentage), 100)) %>%
  ungroup()
# PLot
plot_all <- ggplot(long_data, aes(x = Category, y = Percentage, fill = Value)</pre>
  geom_bar(stat = "identity", position = "stack") +
scale_fill_manual(values = c("#E5696F", "#50689B", "#36D0A1", "#D0A136")) +
```

```
labs(y = "Percentage", fill = "Score", title = "Scoring Criteria for Papers
") +
  theme_minimal() +
  geom_text(aes(label = Label, y = pmin(CumPercentage - 0.5 * Percentage, 100
)),
            position = position_stack(vjust = 0.5), size = 6, check_overlap =
TRUE) +
  theme(
    axis.title = element_text(size = 14),
    axis.text = element_text(size = 18),
    legend.title = element_text(size = 12),
   legend.text = element text(size = 14)
  ) +
  ylim(0, 100)
# Ensure the values are factored in reverse order so that 4 is at the top of
the stack
long_data$Value <- factor(long_data$Value, levels = c("4", "3", "2", "1"))</pre>
# Show the plot
print(plot_all)
## Warning: Removed 5 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```



```
# Save the plot
ggsave("plotall.png", plot_all, width = 10, height = 8, dpi = 300, bg="white"
## Warning: Removed 5 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
#Cluster bar chart fro the completeness criteria by the Type of the study
library(ggplot2)
library(dplyr)
# Reorder the levels of variables
data$Complete <- factor(data$Complete, levels = c("4", "3", "2", "1"))</pre>
# Summarize data
summarized data <- data %>%
  group by(Type, Complete) %>%
  summarise(Count = n(), .groups = "drop") %>%
  left_join(data %>% group_by(Type) %>% summarise(Total = n(), .groups = "dro
p"), by = "Type") %>%
  mutate(Percentage = Count/Total * 100)
# For cumulative percentages
summarized_data <- summarized_data %>%
  arrange(Type, desc(Complete)) %>%
  group by(Type) %>%
  mutate(CumPercentage = cumsum(Percentage))
# Create the stacked bar chart for Completeness with colors and annotations
gcomp \leftarrow ggplot(summarized data, aes(x = Type, y = Percentage, fill = as.fact
or(Complete))) +
  geom_bar(stat = "identity", position = "stack") +
  scale_fill_manual(values = c("#D0A136", "#36D0A1", "#50689B", "#E5696F")) +
  labs(x = "Type", y = "Percentage", fill = "Completeness Score") +
  theme_minimal() +
  geom text(aes(label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "
%"), ""),
                y = pmin(pmax(ifelse(Complete == 1, Percentage / 2, CumPerce
ntage - (0.5 * Percentage)), 0), 100)),
            position = position_stack(vjust = 0.5), color = "Black", size = 5
) +
  theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 18), # Increase axis text
    legend.title = element_text(size = 12), # Increase Legend title
    legend.text = element_text(size = 14) # Increase Legend text
  ) +
 ylim(0, 100)
```

```
print(gcomp)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```



```
ggsave("gcomp.png", gcomp, width = 10, height = 8, dpi = 300)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```

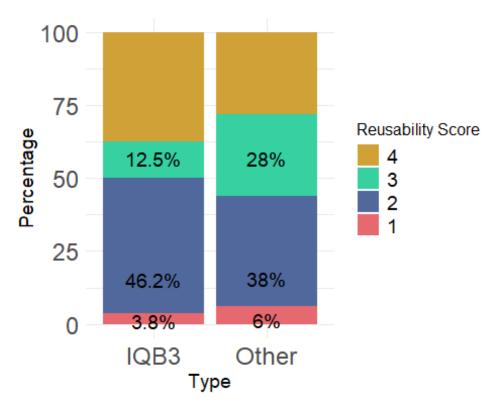
`##Cluster bar chart for the Reuse criteria by the Type of the study

```
# Reorder the Levels of variables
data$Reuse <- factor(data$Reuse, levels = c("4", "3", "2", "1"))

summarized_data_Reuse <- data %>%
    group_by(Type, Reuse) %>%
    summarise(Count = n(), .groups = "drop") %>%
    left_join(data %>% group_by(Type) %>% summarise(Total = n(), .groups = "drop"), by = "Type") %>%
    mutate(Percentage = Count/Total * 100)

summarized_data_Reuse <- summarized_data_Reuse %>%
    arrange(Type, desc(Reuse)) %>%
```

```
group by(Type) %>%
  mutate(CumPercentage = cumsum(Percentage))
print(summarized_data_Reuse)
## # A tibble: 8 × 6
## # Groups:
             Type [2]
     Type Reuse Count Total Percentage CumPercentage
     <chr> <fct> <int> <int>
                                  <dbl>
                                                <dbl>
## 1 IQB3 1
                                   3.75
                                                 3.75
                          80
## 2 IQB3 2
                    37
                          80
                                  46.2
                                                50
## 3 IOB3 3
                                  12.5
                                                62.5
                    10
                          80
## 4 IQB3 4
                    30
                          80
                                  37.5
                                               100
## 5 Other 1
                    3
                          50
                                   6
                                                 6
## 6 Other 2
                    19
                          50
                                  38
                                                44
## 7 Other 3
                    14
                          50
                                  28
                                                72
## 8 Other 4
                    14
                          50
                                  28
                                               100
colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F")
# Visualization
gReuse <- ggplot(summarized_data_Reuse, aes(x = Type, y = Percentage, fill =</pre>
as.factor(Reuse))) +
  geom_bar(stat = "identity", position = "stack") +
  scale fill manual(values = colors) +
  labs(x = "Type", y = "Percentage", fill = "Reusability Score") +
  theme minimal() +
  geom text(aes(label = paste0(round(Percentage, 1), "%"),
                 y = CumPercentage - (0.5 * Percentage)),
            position = position_stack(vjust = 0.5), color = "Black", size = 5
) +
  theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 18), # Increase axis text
    legend.title = element_text(size = 12), # Increase Legend title
    legend.text = element text(size = 14) # Increase Legend text
  ) +
  ylim(0, 100)
print(gReuse)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```

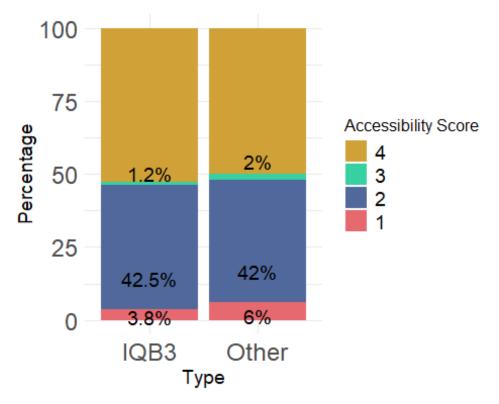


```
ggsave("gReuse.png", gReuse, width = 10, height = 8, dpi = 300)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```

#bar chart for the Accessibility criteria by the Type of the study

```
# Reorder the levels of variables
data$Access <- factor(data$Access, levels = c("4", "3", "2", "1"))</pre>
summarized data Access <- data %>%
  group_by(Type, Access) %>%
  summarise(Count = n(), .groups = "drop") %>%
  left_join(data %>% group_by(Type) %>% summarise(Total = n(), .groups = "dro
p"), by = "Type") %>%
  mutate(Percentage = Count/Total * 100)
summarized_data_Access <- summarized_data_Access %>%
  arrange(Type, desc(Access)) %>%
  group_by(Type) %>%
  mutate(CumPercentage = cumsum(Percentage))
print(summarized data Access)
## # A tibble: 8 × 6
## # Groups:
               Type [2]
## Type Access Count Total Percentage CumPercentage
```

```
## <chr> <fct> <int> <int>
                                   <dbl>
                                                 <dbl>
## 1 IQB3 1
                           80
                                    3.75
                                                  3.75
                     3
## 2 IQB3 2
                     34
                           80
                                   42.5
                                                 46.2
## 3 IQB3 3
                     1
                           80
                                   1.25
                                                 47.5
## 4 IQB3 4
                     42
                          80
                                                100
                                   52.5
## 5 Other 1
                     3
                           50
                                   6
                                                 6
## 6 Other 2
                     21
                           50
                                   42
                                                 48
## 7 Other 3
                     1
                           50
                                   2
                                                 50
## 8 Other 4
                     25
                           50
                                   50
                                                100
colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F")
# Visualization
gaccess <- ggplot(summarized data Access, aes(x = Type, y = Percentage, fill
= as.factor(Access))) +
 geom_bar(stat = "identity", position = "stack") +
 scale_fill_manual(values = colors) +
 labs(x = "Type", y = "Percentage", fill = "Accessibility Score") +
 theme minimal() +
 geom_text(aes(label = paste0(round(Percentage, 1), "%"),
                y = CumPercentage - (0.5 * Percentage)),
            position = position_stack(vjust = 0.5), color = "Black", size = 5
) +
 theme(
    axis.title = element text(size = 14), # Increase axis titles
    axis.text = element_text(size = 18), # Increase axis text
   legend.title = element_text(size = 12), # Increase Legend title
   legend.text = element_text(size = 14) # Increase Legend text
  ) +
 ylim(0, 100)
print(gaccess)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```

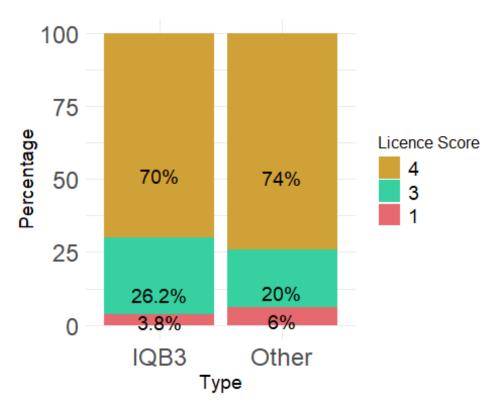


```
ggsave("gaccess.png", gaccess, width = 10, height = 8, dpi = 300)
## Warning: Removed 2 rows containing missing values or values outside the sc
ale range
## (`geom_text()`).
```

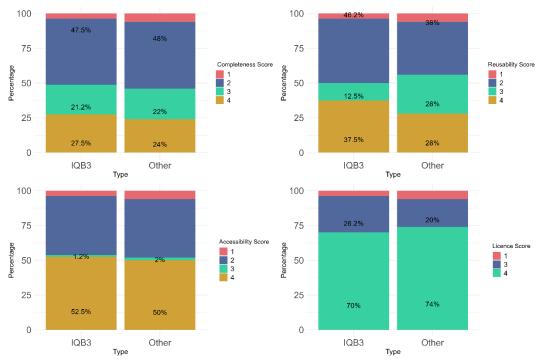
#bar chart for the Licence criteria by the Type of the study

```
# Reorder the levels of variables
data$Licence <- factor(data$Licence, levels = c("4", "3", "2", "1"))</pre>
summarized data Licence <- data %>%
  group_by(Type, Licence) %>%
  summarise(Count = n(), .groups = "drop") %>%
  left_join(data %>% group_by(Type) %>% summarise(Total = n(), .groups = "dro
p"), by = "Type") %>%
  mutate(Percentage = Count/Total * 100)
summarized_data_Licence <- summarized_data_Licence %>%
  arrange(Type, desc(Licence)) %>%
  group_by(Type) %>%
  mutate(CumPercentage = cumsum(Percentage))
print(summarized data Licence)
## # A tibble: 6 × 6
## # Groups:
               Type [2]
## Type Licence Count Total Percentage CumPercentage
```

```
## <chr> <fct> <int> <int>
                                    <dbl>
                                                   <dbl>
## 1 IQB3 1
                       3
                            80
                                      3.75
                                                   3.75
## 2 IQB3 3
                      21
                            80
                                    26.2
                                                   30
## 3 IQB3 4
                      56
                            80
                                    70
                                                  100
## 4 Other 1
                      3
                            50
                                     6
                                                    6
## 5 Other 3
                      10
                            50
                                    20
                                                   26
## 6 Other 4
                      37
                            50
                                    74
                                                  100
colors <- c("#D0A136", "#36D0A1", "#E5696F")</pre>
# Visualization
glicence <- ggplot(summarized_data_Licence, aes(x = Type, y = Percentage, fil</pre>
1 = as.factor(Licence))) +
  geom_bar(stat = "identity", position = "stack") +
  scale_fill_manual(values = colors) +
  labs(x = "Type", y = "Percentage", fill = "Licence Score") +
  theme_minimal() +
  geom_text(aes(label = paste0(round(Percentage, 1), "%"),
                 y = CumPercentage - (0.5 * Percentage)),
            position = position_stack(vjust = 0.5), color = "Black", size = 5
) +
  theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 18), # Increase axis text
    legend.title = element_text(size = 12), # Increase Legend title
   legend.text = element_text(size = 14) # Increase Legend text
  ) +
  ylim(0, 100)
print(glicence)
```



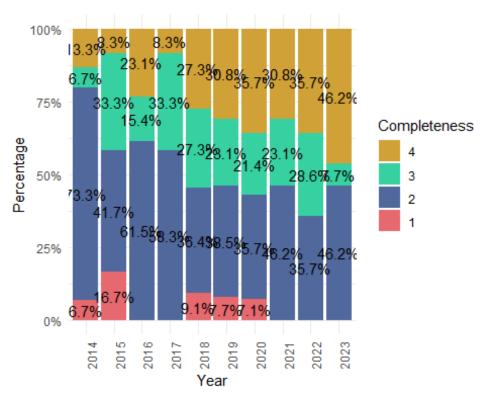
```
ggsave("glicence.png", glicence, width = 10, height = 8, dpi = 300)
## Warning: Removed 2 rows containing missing values or values outside the
scale range
## (`geom_text()`).
library(patchwork)
## Warning: package 'patchwork' was built under R version 4.3.3
##
## Attaching package: 'patchwork'
## The following object is masked from 'package:MASS':
##
##
       area
combinedplot <- gcomp + greuse + gaccess + glicence +</pre>
  plot_layout(
    ncol = 2, heights = c(10, 10), widths = c(10, 10)
print(combinedplot)
```



##Calculate the frequency and percentage for each 'Complete' score within eac h 'Year' long_data <- data %>% count(Year, Complete) %>% group_by(Year) %>% mutate(Percentage = n / sum(n) * 100) %>% mutate(Label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "%"), "")) %>% mutate(Cumulative_Percentage = cumsum(Percentage) - (0.5 * Percentage)) %>% ungroup() # Custom colors colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F") # Creating the stacked bar chart with percentage labels Compyear <- ggplot(long_data, aes(x = as.factor(Year), y = n, fill = as.factor(Year))r(Complete))) + geom bar(stat = "identity", position = "fill") + # Scale the bar heights t o proportions scale_y_continuous(labels = percent_format()) + # Convert the y-axis to pe rcentage geom_text(aes(label = Label, y = Percentage), size = 4, color = "black", position = position_fill(vjust = 0.5) scale_fill_manual(values = colors) +

```
labs(x = "Year",
    y = "Percentage",
    fill = "Completeness") +
theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 20), # Increase axis text
    legend.title = element_text(size = 18), # Increase legend title
    legend.text = element_text(size = 20) # Increase legend text
) +
theme_minimal() +
scale_x_discrete(name = "Year", labels = unique(long_data$Year)) +
theme(axis.text.x = element_text(angle = 90, hjust = 1)) # Adjust text ang
le for readability

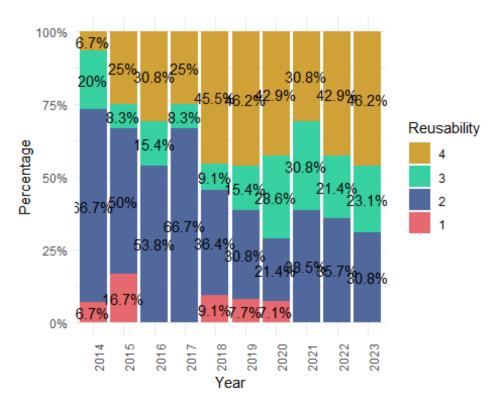
# Print the plot
print(Compyear)
```



```
ggsave("Compyear.png", Compyear, width = 15, height = 10, units = "in", bg =
"white")

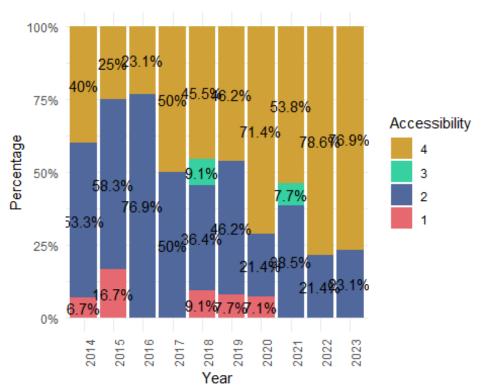
# Calculate the frequency and percentage for each 'Reuse' score within each '
Year'
long_data <- data %>%
    count(Year, Reuse) %>%
    group_by(Year) %>%
    mutate(Percentage = n / sum(n) * 100) %>%
    mutate(Label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "%"), ""
```

```
)) %>%
  mutate(Cumulative Percentage = cumsum(Percentage) - (0.5 * Percentage)) %>%
  ungroup()
# Custom colors
colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F")
# Creating the stacked bar chart with percentage labels
Reuseyear \leftarrow ggplot(long_data, aes(x = as.factor(Year), y = n, fill = as.fact
or(Reuse))) +
  geom bar(stat = "identity", position = "fill") + # Scale the bar heights t
o proportions
  scale_y_continuous(labels = percent_format()) + # Convert the y-axis to pe
rcentage
  geom_text(
    aes(label = Label, y = Percentage),
    size = 4,
    color = "black",
    position = position_fill(vjust = 0.5)
  ) +
  scale_fill_manual(values = colors) +
  labs(x = "Year",
      y = "Percentage",
       fill = "Reusability") +
    theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 20), # Increase axis text
    legend.title = element_text(size = 18), # Increase Legend title
   legend.text = element_text(size = 20) # Increase Legend text
  theme_minimal() +
  scale_x_discrete(name = "Year", labels = unique(long_data$Year)) +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) # Adjust text ang
le for readability
# Print the plot
print(Reuseyear)
```



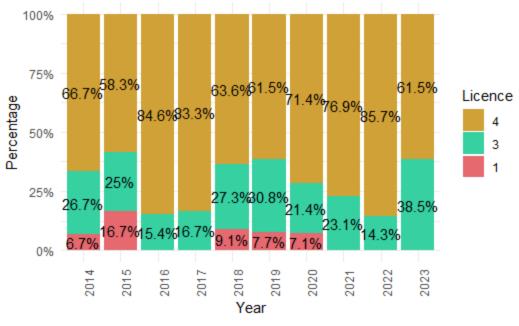
```
ggsave("Reuseyear.png", Reuseyear, width = 15, height = 10, units = "in", bg
= "white")
# Calculate the frequency and percentage for each 'Access' score within each
'Year'
long data <- data %>%
  count(Year, Access) %>%
  group_by(Year) %>%
  mutate(Percentage = n / sum(n) * 100) %>%
  mutate(Label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "%"), ""
)) %>%
  mutate(Cumulative Percentage = cumsum(Percentage) - (0.5 * Percentage)) %>%
  ungroup()
# Custom colors
colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F")
# Creating the stacked bar chart with percentage labels
accessyear <- ggplot(long_data, aes(x = as.factor(Year), y = n, fill = as.fac</pre>
tor(Access))) +
  geom bar(stat = "identity", position = "fill") + # Scale the bar heights t
  scale y continuous(labels = percent format()) + # Convert the y-axis to pe
rcentage
  geom_text(
    aes(label = Label, y = Percentage),
 size = 4,
```

```
color = "black",
    position = position_fill(vjust = 0.5)
  ) +
  scale_fill_manual(values = colors) +
  labs(x = "Year",
       y = "Percentage",
       fill = "Accessibility") +
  theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 20), # Increase axis text
    legend.title = element_text(size = 18), # Increase Legend title
    legend.text = element text(size = 20) # Increase Legend text
  ) +
    theme_minimal() +
  scale_x_discrete(name = "Year", labels = unique(long_data$Year)) +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) # Adjust text ang
le for readability
# Print the plot
print(accessyear)
```



ggsave("accessyear.png", accessyear, width = 15, height = 10, units = "in", b
g = "white")
Calculate the frequency and percentage for each 'Licence' score within each
'Year'
long_data <- data %>%

```
count(Year, Licence) %>%
  group by(Year) %>%
  mutate(Percentage = n / sum(n) * 100) %>%
  mutate(Label = ifelse(Percentage > 5, paste0(round(Percentage, 1), "%"), ""
)) %>%
  mutate(Cumulative_Percentage = cumsum(Percentage) - (0.5 * Percentage)) %>%
  ungroup()
# Custom colors
colors <- c("#D0A136", "#36D0A1", "#50689B", "#E5696F")</pre>
# Creating the stacked bar chart with percentage labels
licenceyear <- ggplot(long_data, aes(x = as.factor(Year), y = n, fill = as.fa
ctor(Licence))) +
  geom bar(stat = "identity", position = "fill") + # Scale the bar heights t
o proportions
  scale y continuous(labels = percent format()) + # Convert the y-axis to pe
rcentage
  geom_text(
    aes(label = Label, y = Percentage),
    size = 4,
    color = "black",
    position = position fill(vjust = 0.5)
  scale_fill_manual(values = colors) +
  labs(x = "Year",
      y = "Percentage",
      fill = "Licence") +
  theme(
    axis.title = element_text(size = 14), # Increase axis titles
    axis.text = element_text(size = 20), # Increase axis text
    legend.title = element_text(size = 18), # Increase Legend title
   legend.text = element_text(size = 20) # Increase Legend text
  ) +
  theme minimal() +
  scale_x_discrete(name = "Year", labels = unique(long_data$Year)) +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) # Adjust text ang
le for readability
# Print the plot
print(licenceyear)
```

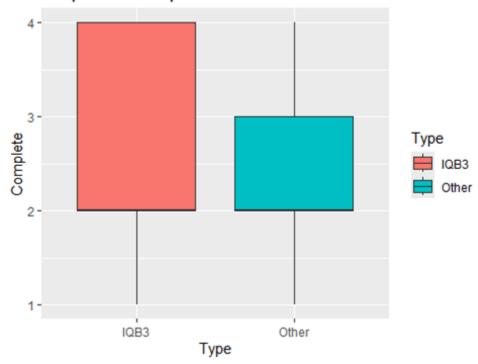


```
ggsave("licenceyear.png", licenceyear, width = 15, height = 10, units = "in",
bg = "white")
plotyear <- Compyear + reuseyear + accessyear + licenceyear +</pre>
  plot_layout(
    ncol = 2, heights = c(10, 10), widths = c(10, 10)
print(plotyear)
      2014
                Year 91 92
                          2022
                                          2015
2016
2017
2019
           2016
                     2020
                        2021
                                                         2020
                                                            2021
      2015
                                                 2021
                                                         2020
```

#study the significant difference between two type of the research for each scoring criteria

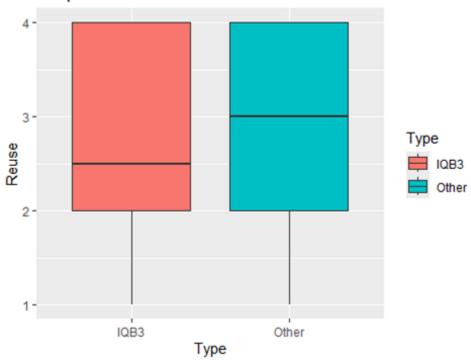
```
# Convert scoring criteria to numeric (if not already)
data <- data %>%
  mutate(across(c(Complete, Reuse, Access, Licence),
~as.numeric(as.character(.))))
# Assumption checks for each criterion
for (criterion in c("Complete", "Reuse", "Access", "Licence")) {
  # Create and display a box plot to check the distribution shape and spread
  print(ggplot(data, aes_string(x = "Type", y = criterion, fill = "Type")) +
    geom_boxplot() +
    ggtitle(paste("Boxplot for", criterion)))
  # Print the median for a basic comparison
  cat(paste("Median of", criterion, "by Type:\n"))
  print(aggregate(. ~ Type, data[c("Type", criterion)], median))
  cat("\n")
}
## Warning: `aes_string()` was deprecated in ggplot2 3.0.0.
## | Please use tidy evaluation idioms with `aes()`.
## | See also `vignette("ggplot2-in-packages")` for more information.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

Boxplot for Complete

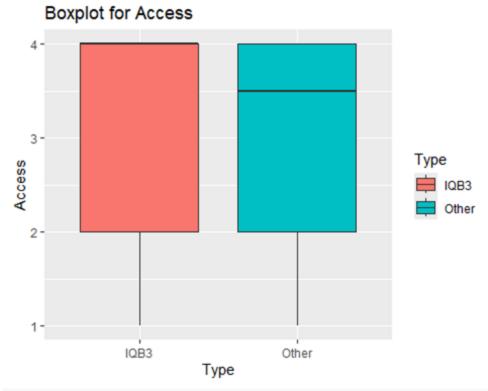


```
## Median of Complete by Type:
## Type Complete
## 1 IQB3     2
## 2 Other     2
```

Boxplot for Reuse



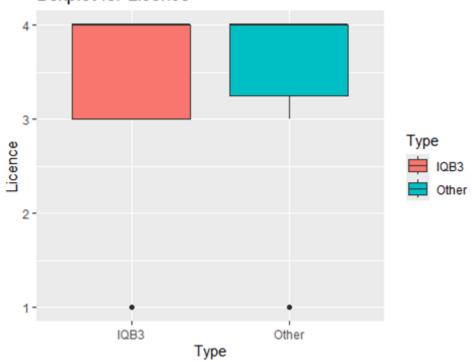
```
## Median of Reuse by Type:
## Type Reuse
## 1 IQB3 2.5
## 2 Other 3.0
```



Median of Access by Type: ## Type Access ## 1 IQB3 4.0

2 Other 3.5

Boxplot for Licence



```
## Median of Licence by Type:
##
      Type Licence
## 1 IQB3
## 2 Other
                 4
# Function to perform Mann-Whitney U test
perform_mann_whitney <- function(data, criterion) {</pre>
  test result <- wilcox.test(</pre>
    reformulate("Type", response = criterion),
    data = data,
    exact = FALSE
  list(criterion = criterion, p.value = test result$p.value)
}
# Perform the test for each criterion
mw_results <- lapply(c("Complete", "Reuse", "Access", "Licence"),</pre>
function(criterion) {
  perform mann whitney(data, criterion)
})
# Convert the list of results to a dataframe
mw_results_df <- do.call(rbind, mw_results)</pre>
print(mw results df)
        criterion p.value
## [1,] "Complete" 0.6265178
## [2,] "Reuse"
                   0.8119625
## [3,] "Access"
                   0.7360586
## [4,] "Licence" 0.7018642
#The assumptions of the test were met:
### Ordinal Data Check: The scoring criteria are ordinal variables
### Similar Distribution Shapes: For each criterion, it creates a box plot to
visually inspect the distribution shapes. This is crucial to check if the
distributions are similar across groups.
### Independence of Observations: the variables are independence
observations.
```

#Create variables for FAIR principles in 2016 and COVID-19 in 2020

```
data <- data %>%
  mutate(Period2020 = ifelse(Year <= 2020, "Before 2020", "After 2020")) %>%
  mutate(Period2016 = ifelse(Year <= 2016, "Before 2016", "After 2016"))#</pre>
```

#Study the significant difference before and afterCovid-19 2020 using Median and Mann Whitney U test

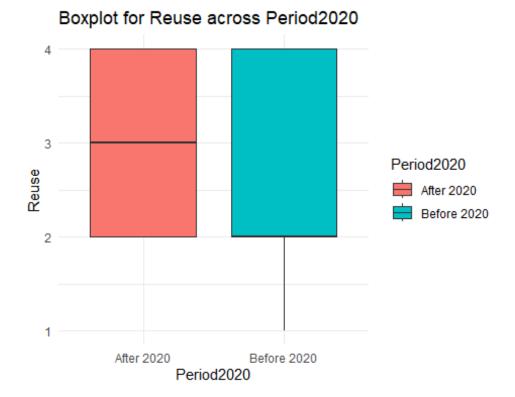
```
# Ensure the scoring criteria are numeric
data <- data %>%
  mutate(across(c(Complete, Reuse, Access, Licence), ~as.numeric(as.character
(.))))
```

```
# Function to generate box plots for distribution checks
generate_boxplot <- function(data, score_var, period_var) {</pre>
  ggplot(data, aes_string(x = period_var, y = score_var, fill = period_var))
    geom_boxplot() +
    labs(title = paste("Boxplot for", score_var, "across", period_var),
         x = period var,
         y = score_var) +
    theme_minimal()
}
# Generate and display box plots for each score with Period2020
lapply(c("Complete", "Reuse", "Access", "Licence"), function(score_var) {
  generate_boxplot(data, score_var, "Period2020")
})
## Warning: `aes_string()` was deprecated in ggplot2 3.0.0.
## | Please use tidy evaluation idioms with `aes()`.
## i See also `vignette("ggplot2-in-packages")` for more information.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
## [[1]]
```

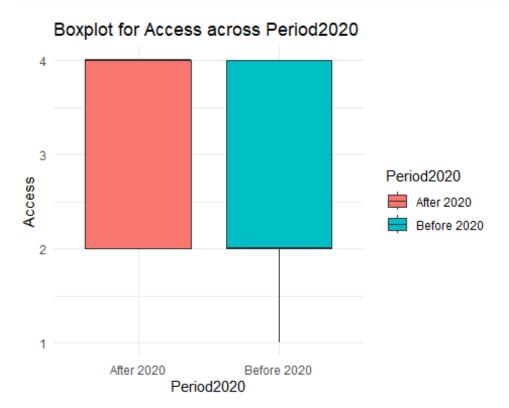
Boxplot for Complete across Period2020



```
##
## [[2]]
```

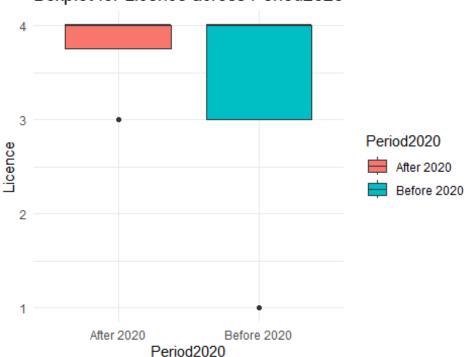






```
##
## [[4]]
```

Boxplot for Licence across Period2020



```
# Function to perform the Mann-Whitney test and calculate medians
perform analysis <- function(data, score var, period var) {</pre>
  test_result <- wilcox.test(reformulate(period_var, score_var), data = data,</pre>
exact = FALSE)
  medians <- data %>%
    group_by(!!sym(period_var)) %>%
    summarise(median = median(!!sym(score_var), na.rm = TRUE), .groups = 'dro
p')
  list(median = medians, p.value = test_result$p.value)
}
# Analysis for each score for 2020
results complete 2020 <- perform analysis(data, "Complete", "Period2020")
results_Reuse_2020 <- perform_analysis(data, "Reuse", "Period2020")</pre>
results_access_2020 <- perform_analysis(data, "Access", "Period2020")
results_licence_2020 <- perform_analysis(data, "Licence", "Period2020")</pre>
# Print results
print(results_complete_2020)
## $median
## # A tibble: 2 × 2
```

```
Period2020 median
## <chr> <dbl>
## 1 After 2020
                     3
## 2 Before 2020
                     2
##
## $p.value
## [1] 0.0394013
print(results_Reuse_2020)
## $median
## # A tibble: 2 × 2
    Period2020 median
##
    ## 1 After 2020
## 2 Before 2020
##
## $p.value
## [1] 0.0639665
print(results_access_2020)
## $median
## # A tibble: 2 × 2
    Period2020 median
##
    <chr> <dbl>
## 1 After 2020
                     4
## 2 Before 2020
##
## $p.value
## [1] 0.002491228
print(results_licence_2020)
## $median
## # A tibble: 2 × 2
##
    Period2020 median
##
    <chr>
                 <dbl>
## 1 After 2020
## 2 Before 2020
##
## $p.value
## [1] 0.4440382
#The assumptions of the test were met:
### Ordinal Data Check: The scoring criteria are ordinal variables
### Similar Distribution Shapes: For each criterion, it creates a box plot to
visually inspect the distribution shapes. This is crucial to check if the dis
tributions are similar across groups.
### Independence of Observations: the variables are independence observations.
```

#Study the significant difference before and afterCovid-19 2020 using Median and Mann Whitney U test for the IQB3 Papers

```
data IQB3 <- data %>%
  filter(Type == "IQB3")
perform_analysis <- function(data_IQB3, score_var, period_var) {</pre>
  test result <- wilcox.test(reformulate(period var, score var), data = data
IQB3, exact = FALSE)
  medians <- data_IQB3 %>%
    group_by(!!sym(period_var)) %>%
    summarise(median = median(!!sym(score var), na.rm = TRUE), .groups = 'dro
p')
  list(median = medians, p.value = test result$p.value)
}
# Analysis for each score for 2020
results_complete_2020_IQB3 <- perform_analysis(data_IQB3, "Complete", "Period
results Reuse 2020 IQB3 <- perform analysis(data IQB3, "Reuse", "Period2020")
results_access_2020_IQB3 <- perform_analysis(data_IQB3, "Access", "Period2020
results licence 2020 IQB3 <- perform analysis(data IQB3, "Licence", "Period20
20")
# Print results
results complete 2020 IQB3
## $median
## # A tibble: 2 × 2
     Period2020 median
##
                  <db1>
##
     <chr>
## 1 After 2020
                      3
## 2 Before 2020
                      2
##
## $p.value
## [1] 0.3644481
results Reuse 2020 IQB3
## $median
## # A tibble: 2 × 2
     Period2020 median
##
     <chr>>
                  <dbl>
## 1 After 2020
                      3
## 2 Before 2020
                      2
##
```

```
## $p.value
## [1] 0.268142
results_access_2020_IQB3
## $median
## # A tibble: 2 × 2
    Period2020 median
##
##
   ## 1 After 2020
                   4
## 2 Before 2020
                    2
##
## $p.value
## [1] 0.01209852
results_licence_2020_IQB3
## $median
## # A tibble: 2 × 2
##
    Period2020 median
    <chr> <dbl>
##
## 1 After 2020
                    4
## 2 Before 2020
##
## $p.value
## [1] 0.7285709
```

#Study the significant difference before and after Covid-19 2020 using Median and Mann Whitney U test for the Other Papers

```
data_Other <- data %>%
  filter(Type == "Other")
perform analysis <- function(data Other, score var, period var) {</pre>
  test_result <- wilcox.test(reformulate(period_var, score_var), data = data_</pre>
Other, exact = FALSE)
  medians <- data_Other %>%
    group_by(!!sym(period_var)) %>%
    summarise(median = median(!!sym(score_var), na.rm = TRUE), .groups = 'dro
p')
  list(median = medians, p.value = test_result$p.value)
}
# Analysis for each score for 2020
results_complete_2020_Other <- perform_analysis(data_Other, "Complete", "Peri
od2020")
results_Reuse_2020_Other <- perform_analysis(data_Other, "Reuse", "Period2020
")
```

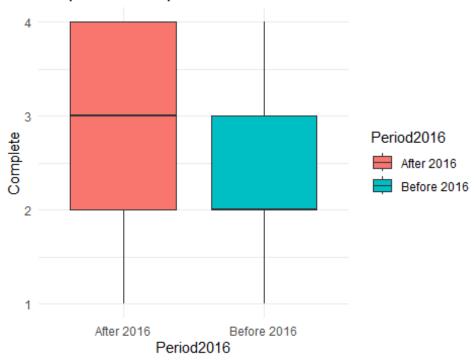
```
results_access_2020_Other <- perform_analysis(data_Other, "Access", "Period20
20")
results_licence_2020_Other <- perform_analysis(data_Other, "Licence", "Period
2020")
# Print results
results_complete_2020_Other
## $median
## # A tibble: 2 × 2
##
     Period2020 median
##
     <chr>
                 <dbl>
## 1 After 2020
                    3.5
## 2 Before 2020
##
## $p.value
## [1] 0.03066192
results_Reuse_2020_Other
## $median
## # A tibble: 2 × 2
     Period2020 median
##
     <chr>>
                  <dbl>
## 1 After 2020
                    3
## 2 Before 2020
                    2.5
##
## $p.value
## [1] 0.1037424
results_access_2020_Other
## $median
## # A tibble: 2 × 2
##
     Period2020 median
##
     <chr>>
                 <dbl>
## 1 After 2020
## 2 Before 2020
                      2
##
## $p.value
## [1] 0.1016182
results_licence_2020_Other
## $median
## # A tibble: 2 × 2
     Period2020 median
                  <dbl>
     <chr>
## 1 After 2020
## 2 Before 2020
##
```

```
## $p.value
## [1] 0.365762
```

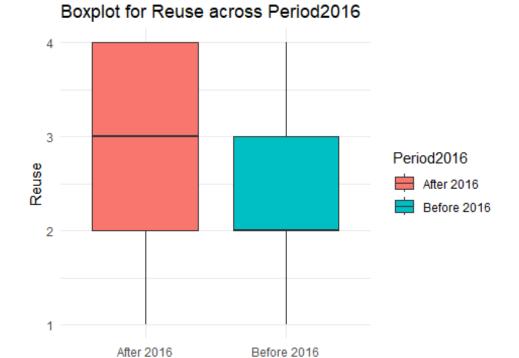
#Study the significant difference before and after Fair 2016 using Median and Mann Whitney U for all the papers

```
data <- data %>%
  mutate(Period2016 = ifelse(Year <= 2016, "Before 2016", "After 2016"))</pre>
# Function to perform the Mann-Whitney test, calculate medians, and generate
box plots
perform_analysis <- function(data, score_var, period_var) {</pre>
  test_result <- wilcox.test(reformulate(period_var, score_var), data = data,</pre>
exact = FALSE)
  medians <- data %>%
    group_by(!!sym(period_var)) %>%
    summarise(median = median(!!sym(score_var), na.rm = TRUE), .groups = 'dro
p')
  # Generate a box plot
  box_plot <- ggplot(data, aes_string(x = period_var, y = score_var, fill = p</pre>
eriod var)) +
    geom_boxplot() +
    labs(title = paste("Boxplot for", score var, "across", period var),
         x = period_var,
         y = score_var) +
    theme minimal()
  # Print the box plot
  print(box_plot)
  list(median = medians, p.value = test_result$p.value, plot = box_plot)
}
# Analysis for each score for 2016
results complete 2016 <- perform analysis(data, "Complete", "Period2016")
```

Boxplot for Complete across Period2016

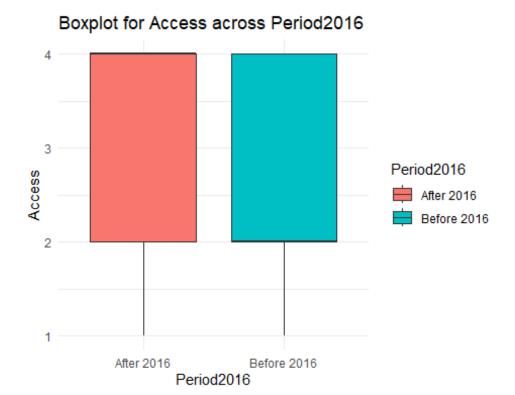


results_Reuse_2016 <- perform_analysis(data, "Reuse", "Period2016")</pre>

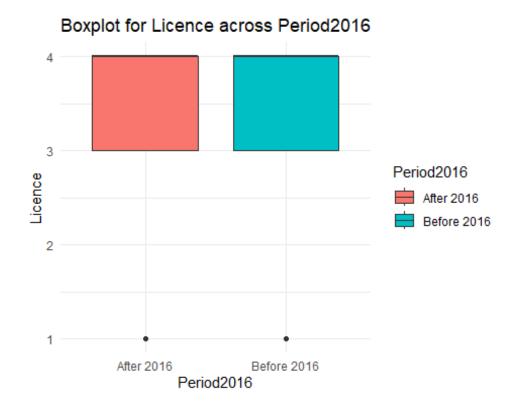


Period2016

results_access_2016 <- perform_analysis(data, "Access", "Period2016")



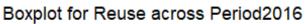
results_licence_2016 <- perform_analysis(data, "Licence", "Period2016")



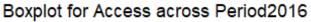
Boxplot for Complete across Period2016



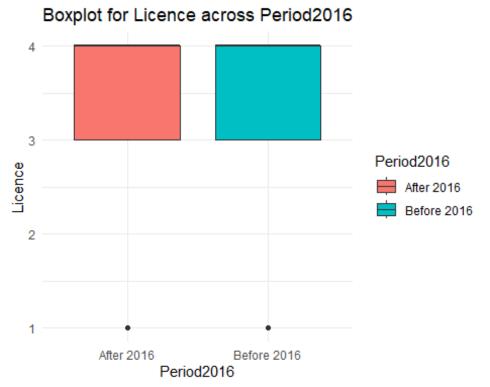
```
##
## $plot
```











#Study the significant difference before and after Fair 2016 using Median and Mann Whitney U test for the IOB3 Papers

```
perform analysis <- function(data IQB3, score var, period var) {</pre>
  test result <- wilcox.test(reformulate(period var, score var), data = data</pre>
IQB3, exact = FALSE)
  medians <- data_IQB3 %>%
    group_by(!!sym(period_var)) %>%
    summarise(median = median(!!sym(score var), na.rm = TRUE), .groups = 'dro
p')
  list(median = medians, p.value = test_result$p.value)
# Analysis for each score for 2016
results_complete_2016_IQB3 <- perform_analysis(data_IQB3, "Complete", "Period
2016")
results_Reuse_2016_IQB3 <- perform_analysis(data_IQB3, "Reuse", "Period2016")</pre>
results_access_2016_IQB3 <- perform_analysis(data_IQB3, "Access", "Period2016
")
results licence 2016 IQB3 <- perform analysis(data IQB3, "Licence", "Period20
16")
# Print results
results_complete_2016_IQB3
```

```
## $median
## # A tibble: 2 × 2
   Period2016 median
##
##
  ## 1 After 2016 3
## 2 Before 2016 2
##
## $p.value
## [1] 0.3265723
results_Reuse_2016_IQB3
## $median
## # A tibble: 2 × 2
## Period2016 median
## <chr> <dbl>
## 1 After 2016 3
## 2 Before 2016 2
##
## $p.value
## [1] 0.1311435
results_access_2016_IQB3
## $median
## # A tibble: 2 × 2
    Period2016 median
## <chr> <dbl>
## 1 After 2016 4
## 2 Before 2016
##
## $p.value
## [1] 0.004984167
results_licence_2016_IQB3
## $median
## # A tibble: 2 × 2
## Period2016 median
## <chr> ## 1 After 2016 4
## 2 Before 2016
##
## $p.value
## [1] 0.5687401
```

#Study the significant difference before and after Fair 2016 using Median and Mann Whitney U test for the Other Papers

```
# Analysis for each score for 2016
results_complete_2016_Other <- perform_analysis(data_Other, "Complete", "Peri</pre>
```

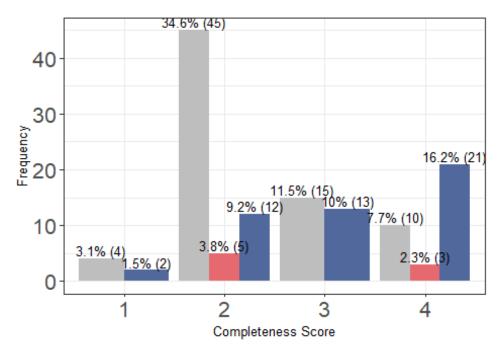
```
od2016")
results_Reuse_2016_Other <- perform_analysis(data_Other, "Reuse", "Period2016
results_access_2016_Other <- perform_analysis(data_Other, "Access", "Period20
16")
results_licence_2016_Other <- perform_analysis(data_Other, "Licence", "Period
2016")
# Print results
results_complete_2016_Other
## $median
## # A tibble: 2 × 2
## Period2016 median
##
     <chr>
                 <dbl>
## 1 After 2016
                     3
## 2 Before 2016
                     2
##
## $p.value
## [1] 0.01351854
results_Reuse_2016_Other
## $median
## # A tibble: 2 × 2
## Period2016 median
##
     <chr>>
                 <dbl>
## 1 After 2016
                     3
## 2 Before 2016
                     2
##
## $p.value
## [1] 0.01741414
results_access_2016_Other
## $median
## # A tibble: 2 × 2
##
    Period2016 median
##
     <chr>>
                 <dbl>
## 1 After 2016
## 2 Before 2016
##
## $p.value
## [1] 0.05851632
results_licence_2016_Other
## $median
## # A tibble: 2 × 2
##
     Period2016 median
##
```

#plot the distribution of DAS for each completeness score

```
# Transform the DAS variable
data <- data %>%
  mutate(NewDAS = case_when(
    is.na(DAS) ~ "Not present",
    DAS == 1 \sim "Shared",
    DAS == 0 ~ "Not shared",
    TRUE ~ as.character(DAS))) # This line is just a fallback to handle unex
pected values
class(data$NewDAS)
## [1] "character"
# Convert the new DAS variable to a factor for plotting
data$NewDAS <- factor(data$NewDAS, levels = c("Not present", "Not shared", "S</pre>
hared"))
# Calculate frequency and percentage for each combination of 'Complete' score
and 'NewDAS' status
das_frequency <- data %>%
  group by(Complete, NewDAS) %>%
  summarise(Frequency = n(), .groups = 'drop') %>%
  mutate(Percentage = (Frequency / sum(Frequency)) * 100)
# Define specific colors for the new DAS values
colors <- c("Not present" = "gray", "Not shared" = "#E5696F", "Shared" = "#50</pre>
689B")
# Create the plot
das_plot <- ggplot(das_frequency, aes(x = as.factor(Complete), y = Frequency,</pre>
fill = NewDAS)) +
  geom bar(stat = "identity", position = "dodge") +
  geom_text(aes(label = paste0(round(Percentage, 1), "% (", Frequency, ")")),
            position = position_dodge(width = 0.9), vjust = -0.25, size = 3.5
) +
  scale_fill_manual(values = colors, name = "DAS Status") +
  theme bw() + # Use theme bw for a white background
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
   legend.text = element_text(size = 16) # Increase Legend text
```

```
) +
labs(x = "Completeness Score", y = "Frequency") +
theme(legend.position = "bottom")

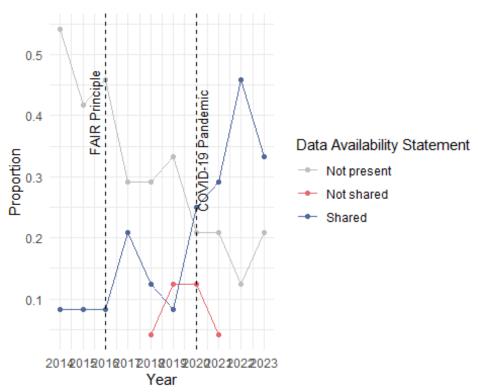
# Print and save the plot
print(das_plot)
```



DAS Status Not present Not shared Share

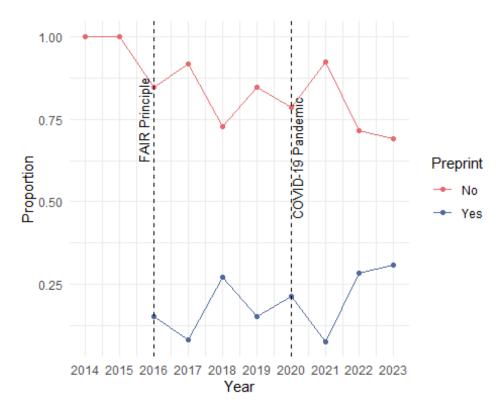
```
ggsave("new_das_distribution_plot.png", das_plot, width = 8, height = 6, bg
= "white")
# Calculate the total count per year
yearly_totals <- data %>%
  group by (Year) %>%
  summarise(Total = n(), .groups = 'drop')
# Join the totals back to the original data and calculate proportions
proportion_data <- data %>%
  left_join(yearly_totals, by = "Year") %>%
  group_by(Year, NewDAS) %>%
  summarise(Count = n(), .groups = 'drop') %>%
  mutate(Proportion = Count / n())
# Get all unique years for the x-axis
all years <- sort(unique(proportion data$Year))</pre>
colors <- c("Not present" = "gray", "Not shared" = "#E5696F", "Shared" = "#50</pre>
689B")
```

```
# Plot with every year on the x-axis
DASyear <- ggplot(proportion data, aes(x = Year, y = Proportion, color = NewD
AS, group = NewDAS)) +
  geom line() +
  geom point() +
  geom_vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom_vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale_x_continuous(breaks = all_years) +
  labs(x = "Year", y = "Proportion", color = "Data Availability Statement") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
    legend.text = element_text(size = 16) # Increase Legend text
  ) +
  theme minimal() +
  scale color manual(values = colors)
print(DASyear)
```



```
ggsave("DASyear.png", DASyear, width = 8, height = 6, bg = "white")
```

```
#Convert the Preprint numeric values to factor levels 'Yes' and 'No'
data$Preprint <- factor(ifelse(data$Preprint == 1, "Yes", "No"))</pre>
# Calculate the total count per year and proportion for Preprint
proportion data <- data %>%
  group by(Year) %>%
  count(Preprint) %>%
  mutate(Total = sum(n),
         Proportion = n / Total)
# Get all unique years for the x-axis
all years <- sort(unique(proportion data$Year))</pre>
# Define colors for 'Yes' and 'No'
colors \leftarrow c("No" = "#E5696F", "Yes" = "#50689B")
# Plot with every year on the x-axis for Preprint
PreprintYear <- ggplot(proportion_data, aes(x = Year, y = Proportion, color =</pre>
Preprint, group = Preprint()) +
  geom_line() +
  geom point() +
  geom_vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion_data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO"
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale x continuous(breaks = all years) +
  labs(x = "Year", y = "Proportion", color = "Preprint") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
    legend.text = element text(size = 16) # Increase Legend text
  ) +
  theme minimal() +
  scale color manual(values = colors)
print(PreprintYear)
```



```
# Save the plot for Preprint
ggsave("PreprintYear.png", PreprintYear, width = 8, height = 6, bg = "white")
```

#study if FAIR implementation and Covid19 have an effect on DAS and Preprint # 1- Fair implementataion

```
total by period <- data %>%
  group by(Period2016) %>%
  summarise(Total = n(), .groups = "drop")
# Join this total with the DAS and Preprint summaries and calculate the perce
ntage
Preprint summary <- data %>%
  group by(Period2016, Preprint) %>%
  summarise(Frequency = n(), .groups = "drop") %>%
  left_join(total_by_period, by = "Period2016") %>%
  mutate(Percentage = (Frequency / Total) * 100)
# Print the Preprint summary table
print(Preprint_summary)
## # A tibble: 4 × 5
     Period2016 Preprint Frequency Total Percentage
##
##
     <chr>>
                 <fct>
                              <int> <int>
                                               <dbl>
## 1 After 2016 No
                                 72
                                       90
                                                  80
## 2 After 2016 Yes
                                 18
                                       90
                                                  20
```

```
## 3 Before 2016 No
                                        40
                                                   95
## 4 Before 2016 Yes
                                  2
                                                    5
                                        40
Preprint_table <- table(data$Preprint, data$Period2016)</pre>
# Perform the Chi-square test
result <- chisq.test(Preprint table)</pre>
#Print the chi-square results
print(result)
##
   Pearson's Chi-squared test with Yates' continuity correction
## data: Preprint table
## X-squared = 3.7034, df = 1, p-value = 0.0543
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
         After 2016 Before 2016
##
     No
           76.15385
                      33.846154
##
           13.84615
                       6.153846
    Yes
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
total_by_period <- data %>%
  group_by(Period2016) %>%
  summarise(Total = n(), .groups = "drop")
# Join this total with the NewDAS and Preprint summaries and calculate the pe
rcentage
das summary <- data %>%
  group by(Period2016, NewDAS)%>%
  summarise(Frequency = n(), .groups = "drop") %>%
  left_join(total_by_period, by = "Period2016") %>%
```

```
mutate(Percentage = (Frequency / Total) * 100)
# Print the DAS summary table
print(das summary)
## # A tibble: 5 × 5
##
     Period2016 NewDAS
                             Frequency Total Percentage
     <chr>>
                <fct>
                                 <int> <int>
                                                   <dbl>
## 1 After 2016 Not present
                                    40
                                          90
                                                   44.4
## 2 After 2016 Not shared
                                     8
                                          90
                                                    8.89
## 3 After 2016 Shared
                                    42
                                          90
                                                   46.7
## 4 Before 2016 Not present
                                    34
                                          40
                                                   85
## 5 Before 2016 Shared
                                          40
                                                   15
                                     6
das_table <- table(data$NewDAS, data$Period2016)</pre>
# Perform the Chi-square test
result <- chisq.test(das_table)</pre>
## Warning in chisq.test(das_table): Chi-squared approximation may be incorre
#Print the chi-square results
print(result)
##
##
  Pearson's Chi-squared test
## data: das table
## X-squared = 19.078, df = 2, p-value = 7.199e-05
# Get the expected values
expected_values <- result$expected</pre>
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
                 After 2016 Before 2016
##
     Not present 51.230769
                              22.769231
##
     Not shared
                  5.538462
                               2.461538
                  33.230769
##
     Shared
                            14.769231
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
```

```
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or ordinal).
#2- Covid 19 # the impact of COVID-19 on Preprint
```

```
#2- Covid 19 # the impact of COVID-19 on Preprint
total by period <- data %>%
  group by(Period2020)%>%
  summarise(Total = n(), .groups = "drop")
# Join this total with the DAS and Preprint summaries and calculate the perce
ntage
Preprint_summary <- data %>%
  group by (Period2020, Preprint) %>%
  summarise(Frequency = n(), .groups = "drop") %>%
  left_join(total_by_period, by = "Period2020") %>%
  mutate(Percentage = (Frequency / Total) * 100)
# Print the Preprint summary table
print(Preprint_summary)
## # A tibble: 4 × 5
     Period2020 Preprint Frequency Total Percentage
##
##
    <chr>
                <fct>
                             <int> <int>
                                               <dbl>
## 1 After 2020 No
                                 31
                                       40
                                                77.5
## 2 After 2020 Yes
                                 9
                                       40
                                                22.5
## 3 Before 2020 No
                                 79
                                       90
                                                87.8
## 4 Before 2020 Yes
                                 11
                                       90
                                                12.2
Preprint_table <- table(data$Preprint, data$Period2020)</pre>
# Perform the Chi-square test
result <- chisq.test(Preprint table)</pre>
#Print the chi-square results
print(result)
##
  Pearson's Chi-squared test with Yates' continuity correction
##
##
## data: Preprint_table
## X-squared = 1.5269, df = 1, p-value = 0.2166
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
```

```
##
         After 2020 Before 2020
##
          33.846154
                       76.15385
##
     No
##
    Yes
           6.153846
                       13.84615
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal)
# the impact of COVID-19 on DAS
total by period <- data %>%
  group by(Period2020) %>%
  summarise(Total = n(), .groups = "drop")
# Join this total with the NewDAS and Preprint summaries and calculate the pe
rcentage
das_summary <- data %>%
  group by(Period2020, NewDAS)%>%
  summarise(Frequency = n(), .groups = "drop") %>%
  left_join(total_by_period, by = "Period2020") %>%
  mutate(Percentage = (Frequency / Total) * 100)
# Print the DAS summary table
print(das_summary)
## # A tibble: 6 × 5
     Period2020 NewDAS
                             Frequency Total Percentage
##
                                 <int> <int>
                                                   <dbl>
##
     <chr>
                 <fct>
## 1 After 2020 Not present
                                    13
                                          40
                                                   32.5
## 2 After 2020 Not shared
                                    1
                                          40
                                                    2.5
## 3 After 2020 Shared
                                    26
                                          40
                                                   65
## 4 Before 2020 Not present
                                          90
                                                   67.8
                                    61
## 5 Before 2020 Not shared
                                    7
                                          90
                                                   7.78
## 6 Before 2020 Shared
                                    22
                                          90
                                                   24.4
das table <- table(data$NewDAS, data$Period2016)</pre>
# Perform the Chi-square test
result <- chisq.test(das table)</pre>
## Warning in chisq.test(das_table): Chi-squared approximation may be incorre
ct
```

```
#Print the chi-square results
print(result)
##
##
  Pearson's Chi-squared test
## data: das_table
## X-squared = 19.078, df = 2, p-value = 7.199e-05
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected_values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
##
##
                 After 2016 Before 2016
##
     Not present 51.230769 22.769231
##
     Not shared
                  5.538462
                             2.461538
##
     Shared
                  33.230769 14.769231
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal)
Image_summary <- data %>%
  group by(Image) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall_total <- sum(Image_summary$count)</pre>
# Add a column for the percentage of each DAS category relative to the overal
l total
Image_summary <- Image_summary %>%
  mutate(percentage = (count / overall total) * 100)
print(Image summary)
## # A tibble: 3 × 3
     Image count percentage
## <int> <int> <dbl>
```

```
## 1
              57
                       43.8
## 2
              11
         1
                       8.46
## 3
              62
        NA
                      47.7
data$DAS <- as.factor(data$DAS)</pre>
# Frequency and Percentage of the papers with DAS in the publications
data %>%
  group_by(Type, Image) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, Image)
## # A tibble: 6 × 5
     Type Image count total percentage
     <chr> <int> <int> <int>
                                   <dbl>
## 1 IQB3
               0
                    32
                           80
                                    40
## 2 IQB3
                     2
                                     2.5
               1
                           80
## 3 IQB3
              NA
                    46
                           80
                                    57.5
## 4 Other
               0
                    25
                           50
                                    50
## 5 Other
                     9
                           50
               1
                                    18
## 6 Other
              NA
                    16
                           50
                                    32
```

#plot the distribution of Image for each completeness score

```
# Transform the Image variable
data <- data %>%
  mutate(NewImage = case_when(
    is.na(Image) ~ "Not used",
    Image == 1 ~ "Shared",
    Image == 0 ~ "Not shared",
    TRUE ~ as.character(Image))) # This line is just a fallback to handle un
expected values
class(data$NewImage)
## [1] "character"
# Convert the new DAS variable to a factor for plotting
data$NewImage <- factor(data$NewImage, levels = c("Not used", "Not shared", "</pre>
Shared"))
# Calculate frequency and percentage for each combination of 'Complete' score
and 'NewImage' status
Image_frequency <- data %>%
  group by(Complete, NewImage) %>%
  summarise(Frequency = n(), .groups = 'drop') %>%
  mutate(Percentage = (Frequency / sum(Frequency)) * 100)
```

```
# Define specific colors for the new Image values
colors <- c("Not used" = "gray", "Not shared" = "#E5696F", "Shared" = "#50689</pre>
B")
# Create the plot
Image_plot <- ggplot(Image_frequency, aes(x = as.factor(Complete), y = Freque</pre>
ncy, fill = NewImage)) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_text(aes(label = paste0(round(Percentage, 1), "% (", Frequency, ")")),
            position = position dodge(width = 0.9), vjust = -0.25, size = 3.5
) +
  scale_fill_manual(values = colors, name = "Image Status") +
  theme bw() + # Use theme bw for a white background
   theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase legend title
    legend.text = element_text(size = 16) # Increase Legend text
  ) +
  labs(x = "Completeness Score", y = "Frequency") +
  theme(legend.position = "bottom")
# Print and save the plot
print(Image plot)
```

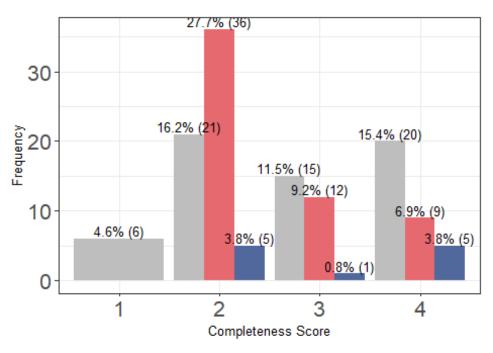
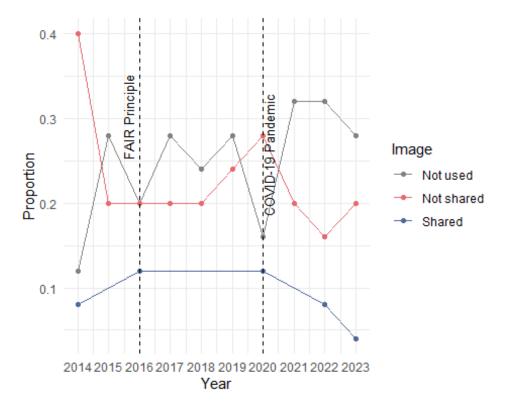


Image Status Not used Not shared Shar

```
ggsave("new Image distribution plot.png", Image plot, width = 8, height = 6,
bg = "white")
# Calculate the total count per year
yearly totals <- data %>%
  group_by(Year) %>%
  summarise(Total = n(), .groups = 'drop')
# Join the totals back to the original data and calculate proportions
proportion_data <- data %>%
  left_join(yearly_totals, by = "Year") %>%
  group_by(Year, NewImage) %>%
  summarise(Count = n(), .groups = 'drop') %>%
  mutate(Proportion = Count / n())
# Get all unique years for the x-axis
all years <- sort(unique(proportion data$Year))</pre>
colors <- c("Not used" = "#828282", "Not shared" = "#E5696F", "Shared" = "#50</pre>
689B")
# Plot with every year on the x-axis
Imageyear \leftarrow ggplot(proportion data, aes(x = Year, y = Proportion, color = Ne
wImage, group = NewImage)) +
  geom_line() +
  geom point() +
  geom vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom_vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion_data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale x continuous(breaks = all years) +
  labs(x = "Year", y = "Proportion", color = "Image") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase legend title
   legend.text = element_text(size = 16) # Increase Legend text
  theme minimal() +
  scale_color_manual(values = colors)
print(Imageyear)
```



```
ggsave("Imageyear.png", Imageyear, width = 8, height = 6, bg = "white")
```

#Chi square test to assess if there is any difference in Image between the two types (IQB3, Other)

```
# Create a contingency table
table Image <- table(data$Type, data$Image)</pre>
# Perform the Chi-square test
result <- chisq.test(table_Image)</pre>
#Print the chi-square results
print(result)
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table_Image
## X-squared = 3.9043, df = 1, p-value = 0.04816
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected_values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
```

```
##
##
                  1
     IQB3 28.5 5.5
##
##
     Other 28.5 5.5
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
Genomics_summary <- data %>%
  group by(Genomics) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall total <- sum(Genomics summary$count)</pre>
# Add a column for the percentage of each Genomics category relative to the o
verall total
Genomics summary <- Genomics summary %>%
  mutate(percentage = (count / overall_total) * 100)
print(Genomics_summary)
## # A tibble: 3 × 3
    Genomics count percentage
##
##
        <int> <int>
                         <dh1>
## 1
            0
                 27
                          20.8
## 2
            1
                 27
                          20.8
## 3
           NA
                 76
                          58.5
data$Genomics <- as.factor(data$Genomics)</pre>
# Frequency and Percentage of the papers with Genomics in the publications
data %>%
  group_by(Type, Genomics) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, Genomics)
```

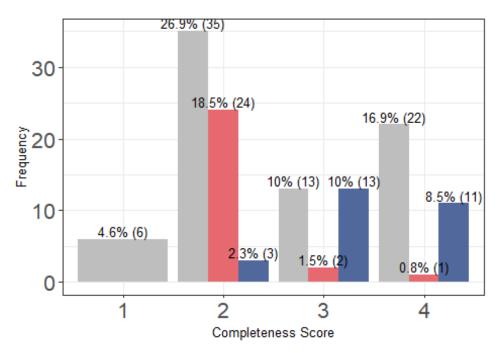
```
## # A tibble: 6 × 5
    Type Genomics count total percentage
    <chr> <fct> <int> <int>
##
                                     <dbl>
## 1 IOB3 0
                      17
                            80
                                      21.2
## 2 IQB3 1
                      15
                            80
                                     18.8
## 3 IQB3 <NA>
                      48
                            80
                                      60
## 4 Other 0
                       10
                             50
                                      20
## 5 Other 1
                             50
                                      24
                       12
## 6 Other <NA>
                       28
                             50
                                      56
```

#plot the distribution of Genomics for each completeness score

```
# Transform the Genomics variable
data <- data %>%
  mutate(NewGenomics = case_when(
    is.na(Genomics) ~ "Not used",
    Genomics == 1 ~ "Shared",
    Genomics == 0 ~ "Not shared",
    TRUE ~ as.character(Genomics))) # This line is just a fallback to handle
unexpected values
class(data$NewGenomics)
## [1] "character"
# Convert the new Genomic variable to a factor for plotting
data$NewGenomics <- factor(data$NewGenomics, levels = c("Not used", "Not shar</pre>
ed", "Shared"))
# Calculate frequency and percentage for each combination of 'Complete' score
and 'NewGenomics' status
Genomics frequency <- data %>%
  group_by(Complete, NewGenomics) %>%
  summarise(Frequency = n(), .groups = 'drop') %>%
  mutate(Percentage = (Frequency / sum(Frequency)) * 100)
# Define specific colors for the new Genomics values
colors <- c("Not used" = "gray", "Not shared" = "#E5696F", "Shared" = "#50689</pre>
B")
# Create the plot
Genomics_plot <- ggplot(Genomics_frequency, aes(x = as.factor(Complete), y =</pre>
Frequency, fill = NewGenomics)) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_text(aes(label = paste0(round(Percentage, 1), "% (", Frequency, ")")),
            position = position_dodge(width = 0.9), vjust = -0.25, size = 3.5
) +
  scale fill manual(values = colors, name = "Genomics Status") +
  theme_bw() + # Use theme_bw for a white background
theme(
```

```
axis.title = element_text(size = 10), # Increase axis titles
axis.text = element_text(size = 16), # Increase axis text
legend.title = element_text(size = 14), # Increase legend title
legend.text = element_text(size = 16) # Increase legend text
) +
labs(x = "Completeness Score", y = "Frequency") +
theme(legend.position = "bottom")

# Print and save the plot
print(Genomics_plot)
```



Genomics Status Not used Not shared Shared

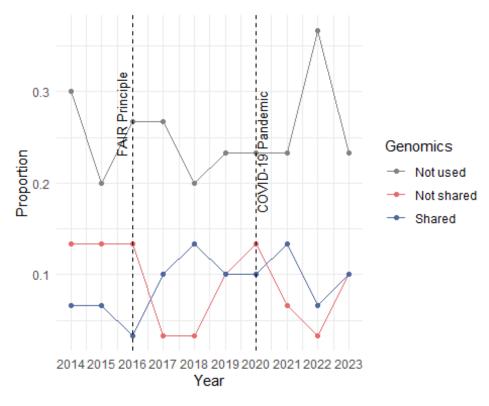
```
ggsave("new_Genomics_distribution_plot.png", Image_plot, width = 8, height =
6, bg = "white")

# Calculate the total count per year
yearly_totals <- data %>%
    group_by(Year) %>%
    summarise(Total = n(), .groups = 'drop')

# Join the totals back to the original data and calculate proportions
proportion_data <- data %>%
    left_join(yearly_totals, by = "Year") %>%
    group_by(Year, NewGenomics) %>%
    summarise(Count = n(), .groups = 'drop') %>%
    mutate(Proportion = Count / n())

# Get all unique years for the x-axis
```

```
all years <- sort(unique(proportion data$Year))</pre>
colors <- c("Not used" = "#828282", "Not shared" = "#E5696F", "Shared" = "#50</pre>
689B")
# Plot with every year on the x-axis
Genomicsyear <- ggplot(proportion_data, aes(x = Year, y = Proportion, color =
NewGenomics, group = NewGenomics)) +
  geom line() +
  geom_point() +
  geom_vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom_vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion_data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO"
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale_x_continuous(breaks = all years) +
  labs(x = "Year", y = "Proportion", color = "Genomics") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
    legend.text = element text(size = 16) # Increase Legend text
  ) +
  theme minimal() +
  scale color manual(values = colors)
print(Genomicsyear)
```



```
ggsave("Genomicsyear.png", Genomicsyear, width = 8, height = 6, bg = "white"
# Create a contingency table
table_Genomics <- table(data$Type, data$Genomics)</pre>
# Perform the Chi-square test
result <- chisq.test(table_Genomics)</pre>
#Print the chi-square results
print(result)
##
  Pearson's Chi-squared test with Yates' continuity correction
##
##
## data: table_Genomics
## X-squared = 0.076705, df = 1, p-value = 0.7818
# Get the expected values
expected values <- result$expected
# Print the expected values
print(expected values) ##in order to check the assumption of chi square test:
Each cell in the contingency table should have an expected count of 5 or more
```

```
##
##
            0 1
     IQB3 16 16
##
##
     Other 11 11
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
Proteometabolomic_summary <- data %>%
  group by(Proteometabolomic) %>%
  summarise(count = n(), .groups = "drop")
# Calculate the overall total
overall total <- sum(Proteometabolomic summary$count)</pre>
# Add a column for the percentage of each Proteometabolomic category relative
to the overall total
Proteometabolomic summary <- Proteometabolomic summary %>%
  mutate(percentage = (count / overall_total) * 100)
print(Proteometabolomic_summary)
## # A tibble: 3 × 3
     Proteometabolomic count percentage
##
##
                 <int> <int>
                                   \langle dh1 \rangle
## 1
                     0
                          34
                                   26.2
## 2
                     1
                          10
                                   7.69
## 3
                          86
                                   66.2
                    NA
data$Proteometabolomic <- as.factor(data$Proteometabolomic)</pre>
# Frequency and Percentage of the papers with Proteometabolomic in the public
ations
data %>%
  group by(Type, Proteometabolomic) %>%
  summarise(count = n(), .groups = "drop") %>%
  group_by(Type) %>%
  mutate(total = sum(count),
         percentage = (count/total)*100) %>%
  ungroup() %>%
  arrange(Type, Proteometabolomic)
```

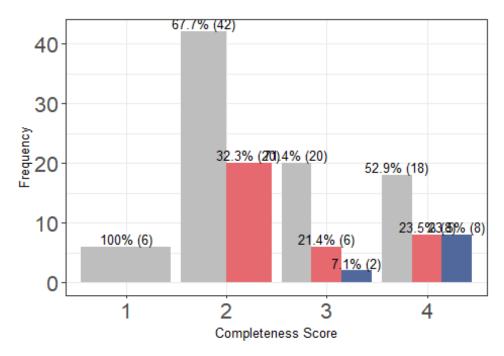
```
## # A tibble: 6 × 5
    Type Proteometabolomic count total percentage
##
     <chr> <fct>
                             <int> <int>
                                              <dbl>
## 1 IQB3 0
                                25
                                      80
                                               31.2
## 2 IQB3 1
                                      80
                                 8
                                               10
## 3 IQB3 <NA>
                                47
                                      80
                                               58.8
## 4 Other 0
                                 9
                                      50
                                               18
## 5 Other 1
                                 2
                                      50
                                                4
                                               78
## 6 Other <NA>
                                39
                                      50
```

#plot the distribution of Proteometabolomic for each completeness score

```
library(dplyr)
library(ggplot2)
# Transform the Proteometabolomic variable
data <- data %>%
  mutate(NewProteometabolomic = case when(
    is.na(Proteometabolomic) ~ "Not used",
    Proteometabolomic == 1 ~ "Shared",
    Proteometabolomic == 0 ~ "Not shared",
    TRUE ~ as.character(Proteometabolomic)))
# Convert the new New Proteometabolomic variable to a factor for plotting
data$NewProteometabolomic <- factor(data$NewProteometabolomic, levels = c("No</pre>
t used", "Not shared", "Shared"))
# Calculate frequency and percentage for each combination of 'Complete' score
and 'NewProteometabolomic' status
Proteometabolomic_frequency <- data %>%
  count(Complete, NewProteometabolomic) %>%
  group by(Complete) %>%
  mutate(Percentage = (n / sum(n)) * 100) %>%
  ungroup()
# Define specific colors for the Proteometabolomic values
colors <- c("Not used" = "gray", "Not shared" = "#E5696F", "Shared" = "#50689</pre>
B")
# Create the plot
Proteometabolomic_plot <- ggplot(Proteometabolomic_frequency, aes(x = as.fact
or(Complete), y = n, fill = NewProteometabolomic)) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_text(aes(label = paste0(round(Percentage, 1), "% (", n, ")")),
            position = position dodge(width = 0.9), vjust = -0.25, size = 3.5
) +
  scale fill_manual(values = colors, name = "Proteometabolomic Status") +
  theme bw() +
  theme(
 axis.title = element text(size = 10),
```

```
axis.text = element_text(size = 16),
  legend.title = element_text(size = 14),
  legend.text = element_text(size = 16)
) +
  labs(x = "Completeness Score", y = "Frequency") +
  theme(legend.position = "bottom")

# Print and save the plot
print(Proteometabolomic_plot)
```



teometabolomic Status Not used Not shared

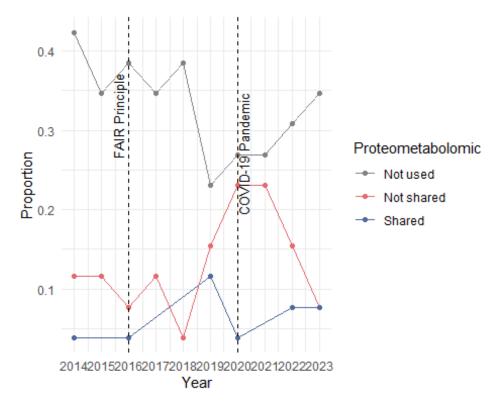
```
ggsave("new_Proteometabolomic_distribution_plot.png", Proteometabolomic_plot,
width = 8, height = 6, bg = "white")

# Calculate the total count per year
yearly_totals <- data %>%
    group_by(Year) %>%
    summarise(Total = n(), .groups = 'drop')

# Join the totals back to the original data and calculate proportions
proportion_data <- data %>%
    left_join(yearly_totals, by = "Year") %>%
    group_by(Year, NewProteometabolomic) %>%
    summarise(Count = n(), .groups = 'drop') %>%
    mutate(Proportion = Count / n())

# Get all unique years for the x-axis
all_years <- sort(unique(proportion_data$Year))</pre>
```

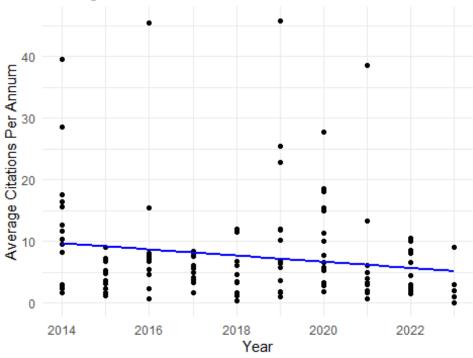
```
colors <- c("Not used" = "#828282", "Not shared" = "#E5696F", "Shared" = "#50</pre>
689B")
# Plot with every year on the x-axis
Proteometabolomicyear \leftarrow ggplot(proportion_data, aes(x = Year, y = Proportion_
, color = NewProteometabolomic, group = NewProteometabolomic)) +
  geom line() +
  geom point() +
  geom_vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion_data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale x continuous(breaks = all years) +
  labs(x = "Year", y = "Proportion", color = "Proteometabolomic") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
    legend.text = element_text(size = 16) # Increase Legend text
  ) +
  theme minimal() +
  scale_color_manual(values = colors)
print(Proteometabolomicyear)
```



```
ggsave("Proteometabolomicyear.png", Proteometabolomicyear, width = 8, height
= 6, bg = "white")
# Create a contingency table
table_Proteometabolomic <- table(data$Type, data$Proteometabolomic)</pre>
# Perform the Chi-square test
result <- chisq.test(table_Proteometabolomic)</pre>
## Warning in chisq.test(table_Proteometabolomic): Chi-squared approximation
may
## be incorrect
#Print the chi-square results
print(result)
##
## Pearson's Chi-squared test with Yates' continuity correction
## data: table_Proteometabolomic
## X-squared = 0, df = 1, p-value = 1
# Get the expected values
expected_values <- result$expected</pre>
# Print the expected values
print(expected_values) ##in order to check the assumption of chi square test:
```

```
Each cell in the contingency table should have an expected count of 5 or more
##
##
              0
                  1
##
     IOB3 25.5 7.5
##
    Other 8.5 2.5
#The assumptions of that test were met:
#### Sample Size: Each cell in the contingency table has an expected count of
5 or more.
#### Independence: The observations are independent of each other. This mean
s that the selection of one observation should not influence or affect the se
lection of another observation.
### Random Sampling: The data is a random sample.
### Categorical Data: Both variables should be categorical (either nominal or
ordinal).
# Load the data
data <- read.csv("datarecord.csv")</pre>
# Plot using applot2
ggplot(data, aes(x = Year, y = CitationsPA)) +
  geom point() + # Add points
  geom_smooth(method = "lm", se = FALSE, color = "blue") + # Add a linear tre
nd line without confidence interval
  labs(title = "Average Citations Per Annum Over Years",
       x = "Year",
       y = "Average Citations Per Annum") +
  theme minimal()
## `geom_smooth()` using formula = 'y ~ x'
```





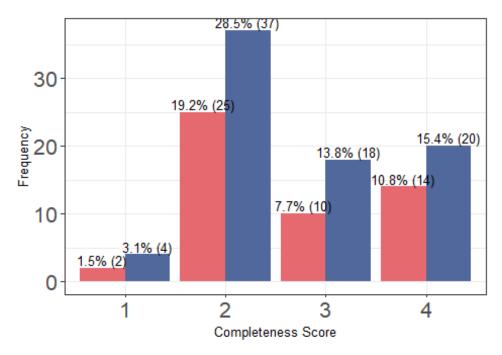
#add the difference

between high and low completeness score

```
df <- read.csv("datarecord.csv")</pre>
# Ensure the 'Complete' and 'CitationsPA' columns are correctly formatted
# 'Complete' should be numeric or integer, and 'CitationsPA' should be numeri
С.
# If not, you might need to convert them, for example:
# df$Complete <- as.numeric(df$Complete)</pre>
# df$CitationsPA <- as.numeric(df$CitationsPA)</pre>
# Step 1: Split the data into two groups based on 'Complete' scores
group1 <- df$CitationsPA[df$Complete %in% c(1,2)]</pre>
group2 <- df$CitationsPA[df$Complete %in% c(3,4)]</pre>
# Ensure that there are no NA values that could affect the t-test
# This step is optional but recommended
group1 <- na.omit(group1)</pre>
group2 <- na.omit(group2)</pre>
# Step 2: Conduct a T-test to compare the mean citation counts between the tw
o groups
t_test_result <- t.test(group1, group2)</pre>
# Display the t-test results
print(t_test_result)
```

```
##
## Welch Two Sample t-test
##
## data: group1 and group2
## t = -0.22511, df = 127.94, p-value = 0.8223
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.232464 2.572094
## sample estimates:
## mean of x mean of y
## 7.229044 7.559229
# Additional: Display mean citations for insight
mean_group1 <- mean(group1)</pre>
mean_group2 <- mean(group2)</pre>
cat("Mean citation counts for Completeness 1-2:", mean_group1, "\n")
## Mean citation counts for Completeness 1-2: 7.229044
cat("Mean citation counts for Completeness 3-4:", mean group2, "\n")
## Mean citation counts for Completeness 3-4: 7.559229
# Assuming 'data' is your DataFrame and has been loaded correctly
# Transform the PI Group Leader variable
data <- data %>%
  mutate(PIGroupLeaderStatus = case_when(
    PIGroupLeader == 1 ~ "Leader",
    PIGroupLeader == 0 ~ "Not leader",
    TRUE ~ as.character(PIGroupLeader))) # Fallback for unexpected values
# Check the class of the new variable
class(data$PIGroupLeaderStatus)
## [1] "character"
# Convert the new PI Group Leader variable to a factor for plotting
data <- data %>%
  mutate(PIGroupLeaderStatus = factor(PIGroupLeaderStatus, levels = c("Not le
ader", "Leader")))
# Calculate frequency and percentage for each 'Complete' score and 'PIGroupLe
aderStatus'
pi_group_frequency <- data %>%
  group_by(Complete, PIGroupLeaderStatus) %>%
  summarise(Frequency = n(), .groups = 'drop') %>%
  mutate(Percentage = (Frequency / sum(Frequency)) * 100)
# Define specific colors for the PI Group Leader status
colors <- c("Not leader" = "#E5696F", "Leader" = "#50689B")</pre>
```

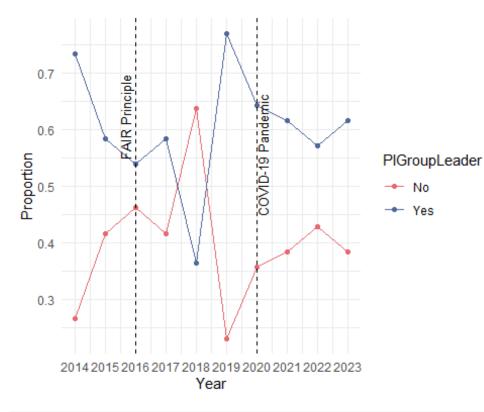
```
# Create the plot
pi group plot <- ggplot(pi group frequency, aes(x = as.factor(Complete), y =</pre>
Frequency, fill = PIGroupLeaderStatus)) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_text(aes(label = paste0(round(Percentage, 1), "% (", Frequency, ")")),
            position = position_dodge(width = 0.9), vjust = -0.25, size = 3.5
) +
  scale fill manual(values = colors, name = "PI Group Status") +
  theme bw() +
  theme(
    axis.title = element_text(size = 10),
    axis.text = element_text(size = 16),
    legend.title = element_text(size = 14),
    legend.text = element text(size = 16)
  ) +
  labs(x = "Completeness Score", y = "Frequency") +
  theme(legend.position = "bottom")
# Print the plot
print(pi_group_plot)
```



PI Group Status Not leader Leader

```
# Save the plot to a file
ggsave("pi_group_status_distribution_plot.png", pi_group_plot, width = 8, hei
ght = 6, bg = "white")
```

```
#Convert the PIGroupLeader numeric values to factor levels 'Yes' and 'No'
data$PIGroupLeader <- factor(ifelse(data$PIGroupLeader == 1, "Yes", "No"))</pre>
# Calculate the total count per year and proportion for PIGroupLeader
proportion data <- data %>%
  group by(Year) %>%
  count(PIGroupLeader) %>%
  mutate(Total = sum(n),
         Proportion = n / Total)
# Get all unique years for the x-axis
all years <- sort(unique(proportion data$Year))</pre>
# Define colors for 'Yes' and 'No'
colors \leftarrow c("No" = "#E5696F", "Yes" = "#50689B")
# Plot with every year on the x-axis for PIGroupLeader
PIGroupLeaderYear <- ggplot(proportion_data, aes(x = Year, y = Proportion, co
lor = PIGroupLeader, group = PIGroupLeader)) +
  geom_line() +
  geom_point() +
  geom_vline(xintercept = 2016, linetype = "dashed", color = "black") +
  geom vline(xintercept = 2020, linetype = "dashed", color = "black") +
  annotate("text", x = 2016, y = max(proportion_data$Proportion), label = "FA
IR Principle", angle = 90, hjust = 1.5, vjust = -0.5, size = 3.5) +
  annotate("text", x = 2020, y = max(proportion_data$Proportion), label = "CO"
VID-19 Pandemic", angle = 90, hjust = 1.5, vjust = 1, size = 3.5) +
  scale x continuous(breaks = all years) +
  labs(x = "Year", y = "Proportion", color = "PIGroupLeader") +
  theme(
    axis.title = element_text(size = 10), # Increase axis titles
    axis.text = element_text(size = 16), # Increase axis text
    legend.title = element_text(size = 14), # Increase Legend title
    legend.text = element text(size = 16) # Increase Legend text
  ) +
  theme minimal() +
  scale_color_manual(values = colors)
print(PIGroupLeaderYear)
```



```
# Save the plot for PIGroupLeader
ggsave("PIGroupLeader.png", PIGroupLeaderYear, width = 8, height = 6, bg = "w
hite")
# Assuming 'data' is your DataFrame and has been loaded correctly
# Transform the completeness score into a binary variable (high vs. low)
data <- data %>%
  mutate(CompletenessCategory = case_when(
    Complete %in% 3:4 ~ "High",
    Complete %in% 1:2 ~ "Low",
    TRUE ~ NA_character_ # Handle potential unexpected values
  ))
# Drop rows with NA in CompletenessCategory if any exist due to unexpected va
data <- data %>% filter(!is.na(CompletenessCategory))
# Create a contingency table
contingency_table <- table(data$PIGroupLeader, data$CompletenessCategory)</pre>
# Conduct a Chi-square test of independence
chi_square_result <- chisq.test(contingency_table)</pre>
# Print the result of the Chi-square test
print(chi square result)
```

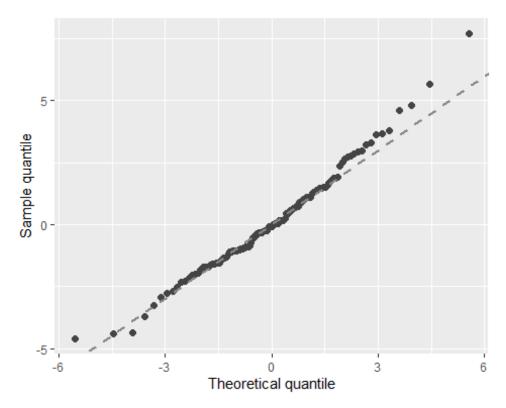
```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: contingency table
## X-squared = 6.4044e-31, df = 1, p-value = 1
#2 - Ordinal regression models (Scoring Criteria and year)
library(ordinal) # ordinal logistic regression: cumulative link mixed models,
clm function is in this package; For a detailed explanation of the package an
d the functions available see: https://cran.r-project.org/web/packages/ordina
l/vignettes/clm_article.pdf
## Warning: package 'ordinal' was built under R version 4.2.3
##
## Attaching package: 'ordinal'
## The following object is masked from 'package:dplyr':
##
##
       slice
library(VGAM) # more ordinal regression
## Warning: package 'VGAM' was built under R version 4.2.3
## Loading required package: stats4
## Loading required package: splines
## Attaching package: 'VGAM'
## The following objects are masked from 'package:ordinal':
##
##
       dgumbel, dlgamma, pgumbel, plgamma, qgumbel, rgumbel, wine
library(dplyr) # Data manipulation
library(chisq.posthoc.test) # If needed
## Warning: package 'chisq.posthoc.test' was built under R version 4.2.3
library(gmodels) # For SPSS style chi-sq/ contingency tables
## Warning: package 'gmodels' was built under R version 4.2.3
library(emmeans)
## Warning: package 'emmeans' was built under R version 4.2.3
library(sure)
## Warning: package 'sure' was built under R version 4.2.3
```

```
data$Complete <- factor(data$Complete, ordered = TRUE)
data$Reuse <- factor(data$Reuse, ordered = TRUE)
data$Access <- factor(data$Access, ordered = TRUE)
data$Licence <- factor(data$Licence, ordered = TRUE)
data$Year <- as.numeric(data$Year)</pre>
```

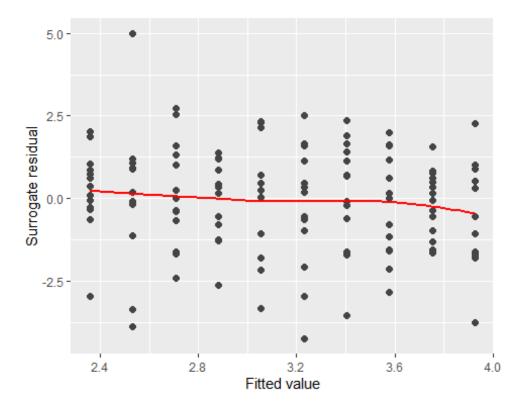
#Ordinal regression model for the Completeness by year

```
m1a <- clm(Complete ~ Year, data = data)</pre>
## Warning in x$code == 0L \mid | action == "silent": 'length(x) = 2 > 1' in coer
cion
## to 'logical(1)'
## Warning: (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
summary_m1a <- summary(m1a)</pre>
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci \leftarrow exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z values <- coefs[,1] / coefs[,2]</pre>
p_values <- 2 * (1 - pnorm(abs(z_values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower_ci, UpperCI = upper_ci, p_valu
e = p values)
print(results)
##
                   OR
                            LowerCI
                                          UpperCI
                                                       p value
## 1|2 1.622844e+151 2.282840e+50 1.153661e+252 0.003296030
## 2 3 4.125302e+152 5.138138e+51 3.312117e+253 0.003032294
## 3|4 1.117896e+153 1.334844e+52 9.362075e+253 0.002955632
## Year 1.190114e+00 1.060719e+00 1.335293e+00 0.003038852
#Check the Assumptions
nominal_test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
## formula: Complete ~ Year
                        AIC LRT Pr(>Chi)
          Df logLik
            -148.39 304.78
## <none>
## Year
```

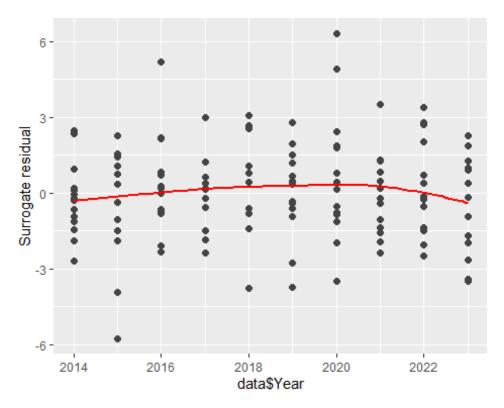
```
scale test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here because p = 0.8386. If p is less than 0.05 th
en assumptions are violated
## Warning: (-1) Model failed to converge with max|grad| = 154.368 (tol = 1e-
## In addition: iteration limit reached
## Tests of scale effects
## formula: Complete ~ Year
                      AIC LRT Pr(>Chi)
         Df logLik
             -148.39 304.78
## <none>
## Year
convergence(m1a) # This is another way to assess the model
## nobs logLik niter max.grad cond.H logLik.Error
## 130 -148.39 7(0) 6.62e-10 6.3e+12 <1e-10
##
##
       Estimate
                  Std.Err Gradient
                                        Error Cor.Dec Sig.Dig
## 1 | 2 348.175 118.47952 -8.05e-14 -9.96e-11
                                                    9
                                                           12
                                                    9
## 2|3 351.410 118.54160 4.79e-13 -9.97e-11
                                                           12
## 3 4 352.407 118.56313 -7.34e-13 -9.97e-11
                                                    9
                                                           12
## Year
          0.174
                  0.05873 6.62e-10 -4.94e-14
                                                   13
                                                           13
##
## Eigen values of Hessian:
## 1.484e+08 6.886e+01 8.253e+00 2.371e-05
##
## Convergence message from clm:
## (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## (3) Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'

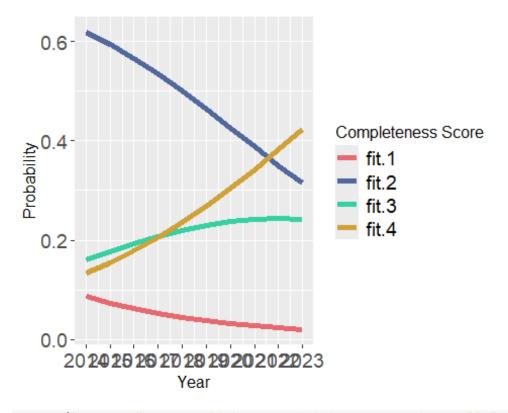


autoplot.clm(m1a, what = c("covariate"), x = data\$Year) # scale test indicate
d no variance issues. This residual, covariate plots seems acceptable as well
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



N.B. - Interpreting residual plots is largely subjective! ###plotting the model # Predict probabilities for each level of 'Complete' new_data <- data.frame(Year = sort(unique(data\$Year)))</pre> pred_probs <- predict(m1a, newdata = new_data, type = "prob")</pre> # Add the 'Year' column to the predicted probabilities dataframe pred_probs_df <- cbind(new_data, as.data.frame(pred_probs))</pre> # Convert the predicted probabilities to a long format for ggplot library(reshape2) ## Warning: package 'reshape2' was built under R version 4.2.3 ## ## Attaching package: 'reshape2' ## ## The following object is masked from 'package:tidyr': ## ## smiths pred probs long <- melt(pred probs df, id.vars = 'Year', variable.name = 'Com</pre> pleteLevel', value.name = 'PredictedProbability') # Set the colors my_colors <- c("#E5696F", "#50689B", "#36D0A1", "#D0A136")</pre>

```
# Modify ggplot command
og1 <- ggplot(pred probs long, aes(x = Year, y = PredictedProbability, group
= CompleteLevel, color = CompleteLevel)) +
 geom_line(size = 2) + # Set size of the lines to make them thicker
 scale color manual(values = my colors) +
 scale_x_continuous(breaks = unique(pred_probs_long$Year)) + # Show all yea
rs on x-axis
 labs(x = "Year", y = "Probability", color = "Completeness Score") +
 theme(
    axis.title = element_text(size = 12), # Increase axis titles
    axis.text = element_text(size = 14), # Increase axis text
    legend.title = element_text(size = 12), # Increase Legend title
    legend.text = element_text(size = 14) # Increase Legend text
 )
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## | Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
print(og1)
```

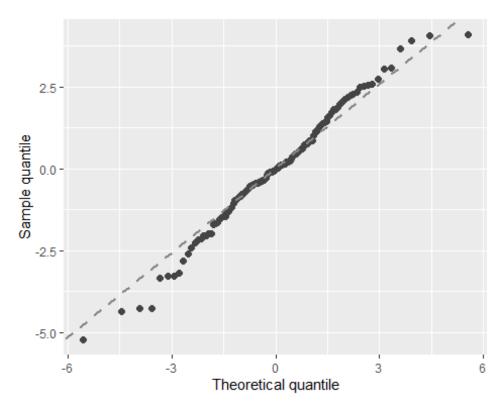


ggsave("og1.png", og1, width = 15, height = 10, units = "in", bg = "white")

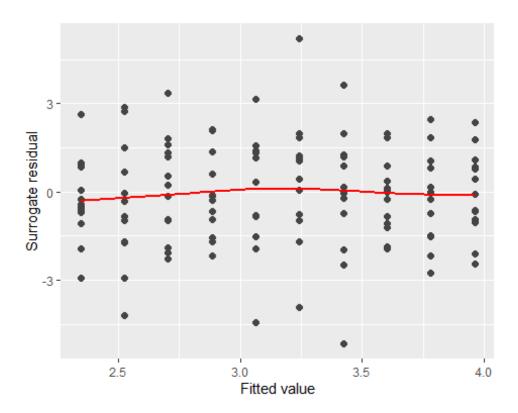
#Ordinal regression model for the Reusibility by year

```
m1a <- clm(Reuse ~ Year, data = data)</pre>
## Warning in x$code == 0L || action == "silent": 'length(x) = 2 > 1' in coer
cion
## to 'logical(1)'
## Warning: (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
summary m1a <- summary(m1a)</pre>
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci <- exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z_values <- coefs[,1] / coefs[,2]</pre>
p_values <- 2 * (1 - pnorm(abs(z_values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower_ci, UpperCI = upper_ci, p_valu
e = p values)
print(results)
                   OR
                           LowerCI
                                          UpperCI
                                                      p value
## 1 | 2 1.068091e+156 2.411247e+57 4.731236e+254 0.001934390
## 2|3 2.302380e+157 4.592835e+58 1.154179e+256 0.001778537
## 3|4 5.188970e+157 9.966839e+58 2.701500e+256 0.001739712
## Year 1.196679e+00 1.069257e+00 1.339285e+00 0.001773416
#Check the Assumptions
nominal test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
## formula: Reuse ~ Year
                      AIC LRT Pr(>Chi)
          Df logLik
          -148.73 305.47
## <none>
## Year
scale_test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here
## Warning: (-1) Model failed to converge with max grad = 124.904 (tol = 1e-
06)
## In addition: iteration limit reached
```

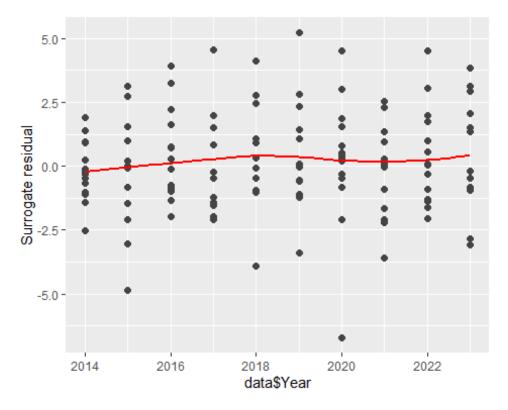
```
## Tests of scale effects
##
## formula: Reuse ~ Year
         Df logLik
                       AIC LRT Pr(>Chi)
           -148.73 305.47
## <none>
## Year
convergence(m1a) # This is another way to assess the model
   nobs logLik niter max.grad cond.H logLik.Error
## 130 -148.73 7(0) 6.29e-10 6.0e+12 <1e-10
##
##
        Estimate
                  Std.Err Gradient
                                       Error Cor.Dec Sig.Dig
## 1 2 359.2691 115.88860 -1.73e-13 1.13e-11
## 2|3 362.3398 115.95172 1.92e-12 1.13e-11
                                                   10
                                                           13
## 3 4 363.1524 115.97102 -2.06e-12 1.13e-11
                                                   10
                                                          13
                  0.05744 6.29e-10 5.60e-15
## Year
         0.1796
                                                  13
                                                          13
##
## Eigen values of Hessian:
## 1.496e+08 8.699e+01 8.406e+00 2.480e-05
## Convergence message from clm:
## (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## (3) Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'

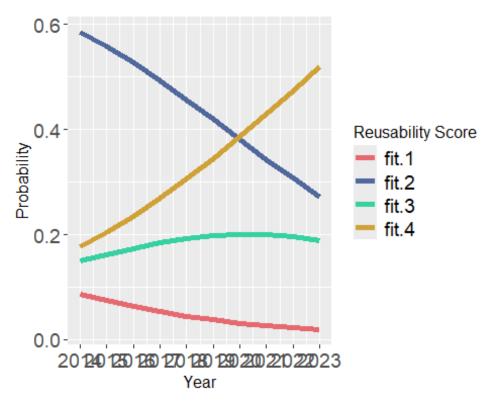


autoplot.clm(m1a, what = c("covariate"), x = data\$Year) # scale test indicate
d no variance issues. This residual, covariate plots seems acceptable as well
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



```
# N.B. - Interpreting residual plots is largely subjective!
###plotting the model
# Predict probabilities for each level of 'Complete'
new data <- data.frame(Year = sort(unique(data$Year)))</pre>
pred_probs <- predict(m1a, newdata = new_data, type = "prob")</pre>
# Add the 'Year' column to the predicted probabilities dataframe
pred_probs_df <- cbind(new_data, as.data.frame(pred_probs))</pre>
# Convert the predicted probabilities to a long format for gaplot
library(reshape2)
pred_probs_long <- melt(pred_probs_df, id.vars = 'Year', variable.name = 'Reu</pre>
selevel', value.name = 'PredictedProbability')
# Set the colors
my_colors <- c("#E5696F", "#50689B", "#36D0A1", "#D0A136")</pre>
# Modify agplot command
og2 <- ggplot(pred_probs_long, aes(x = Year, y = PredictedProbability, group
= Reuselevel, color = Reuselevel)) +
  geom_line(size = 2) + # Set size of the lines to make them thicker
  scale color manual(values = my colors) + # Use colors
  scale_x_continuous(breaks = unique(pred_probs_long$Year)) + # Show all yea
rs on x-axis
  labs(x = "Year", y = "Probability", color = "Reusability Score") +
```

```
theme(
    axis.title = element_text(size = 12), # Increase axis titles
    axis.text = element_text(size = 14), # Increase axis text
    legend.title = element_text(size = 12), # Increase legend title
    legend.text = element_text(size = 14) # Increase legend text
)
print(og2)
```

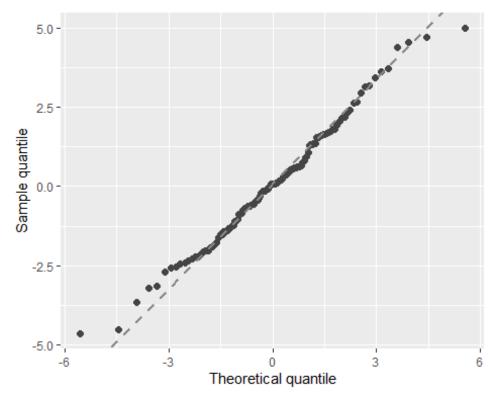


```
ggsave("og2.png", og2, width = 15, height = 10, units = "in", bg = "white")
m1a <- clm(Access ~ Year, data = data)
## Warning in x$code == 0L || action == "silent": 'length(x) = 2 > 1' in coercion
## to 'logical(1)'
## Warning: (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
summary_m1a <- summary(m1a)

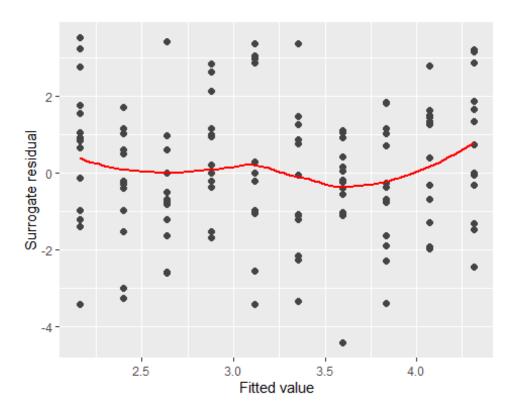
# Extract coefficients and standard errors
coefs <- summary_m1a$coefficients[, 1] # Coefficients
se_coefs <- summary_m1a$coefficients[, 2] # Standard errors</pre>
```

```
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci \leftarrow exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z_values <- coefs[,1] / coefs[,2]</pre>
p values <- 2 * (1 - pnorm(abs(z values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower ci, UpperCI = upper ci, p valu
e = p values)
print(results)
##
                           LowerCI UpperCI
                   OR
                                                  p value
## 1 2 2.193927e+208 1.045919e+98
                                        Inf 0.0002143828
## 2|3 4.895981e+209 2.007751e+99
                                        Inf 0.0001967226
## 3 4 5.246755e+209 2.142653e+99
                                        Inf 0.0001963546
## Year 1.270317e+00 1.120009e+00 1.440796 0.0001960819
#Check the Assumptions
nominal test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
##
## formula: Access ~ Year
          Df logLik AIC LRT Pr(>Chi)
## <none>
            -110.95 229.9
## Year
scale_test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here
## Warning: (-1) Model failed to converge with max|grad| = 301.489 (tol = 1e-
06)
## In addition: iteration limit reached
## Tests of scale effects
##
## formula: Access ~ Year
          Df logLik
                       AIC LRT Pr(>Chi)
             -110.95 229.9
## <none>
## Year
convergence(m1a) # This is another way to assess the model
## nobs logLik niter max.grad cond.H logLik.Error
   130 -110.95 8(2) 1.15e-09 6.4e+12 <1e-10
##
```

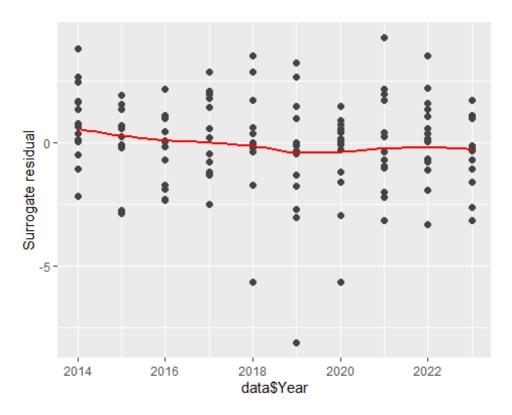
```
Estimate
                  Std.Err Gradient Error Cor.Dec Sig.Dig
## 1 2 479.7234 129.60467 -2.94e-13 2.78e-11
                                                  10
                                                           13
## 2|3 482.8287 129.68151 9.06e-12 2.79e-11
                                                  10
                                                           13
## 3 4 482.8979 129.68364 -9.34e-12 2.79e-11
                                                  10
                                                           13
          0.2393
## Year
                  0.06425 1.15e-09 1.38e-14
                                                  13
                                                           13
##
## Eigen values of Hessian:
## 1.276e+08 8.504e+02 8.325e+00 1.983e-05
## Convergence message from clm:
## (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## (3) Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'

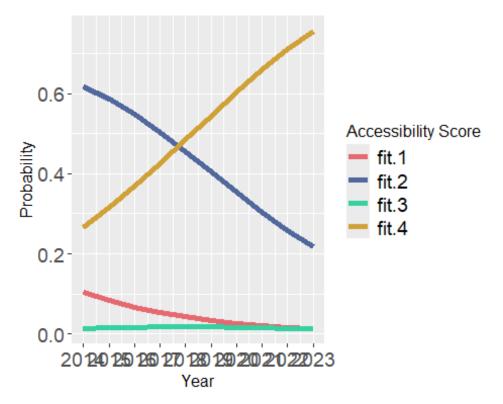


autoplot.clm(m1a, what = c("covariate"), x = data\$Year) # scale test indicate
d no variance issues. This residual, covariate plots seems acceptable as well
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



```
# N.B. - Interpreting residual plots is largely subjective!
###plotting the model
# Predict probabilities for each level of 'Complete'
new_data <- data.frame(Year = sort(unique(data$Year)))</pre>
pred_probs <- predict(m1a, newdata = new_data, type = "prob")</pre>
# Add the 'Year' column to the predicted probabilities dataframe
pred_probs_df <- cbind(new_data, as.data.frame(pred_probs))</pre>
# Convert the predicted probabilities to a long format for gaplot
library(reshape2)
pred_probs_long <- melt(pred_probs_df, id.vars = 'Year', variable.name = 'Acc</pre>
essLevel', value.name = 'PredictedProbability')
# Set the colors
my_colors <- c("#E5696F", "#50689B", "#36D0A1", "#D0A136")</pre>
# Modify agplot command
og3 <- ggplot(pred_probs_long, aes(x = Year, y = PredictedProbability, group
= AccessLevel, color = AccessLevel)) +
  geom_line(size = 2) + # Set size of the lines to make them thicker
  scale color manual(values = my colors) + # Use colors
  scale_x_continuous(breaks = unique(pred_probs_long$Year)) + # Show all yea
rs on x-axis
  labs(x = "Year", y = "Probability", color = "Accessibility Score") +
```

```
theme(
    axis.title = element_text(size = 12), # Increase axis titles
    axis.text = element_text(size = 14), # Increase axis text
    legend.title = element_text(size = 12), # Increase legend title
    legend.text = element_text(size = 14) # Increase legend text
)
print(og3)
```

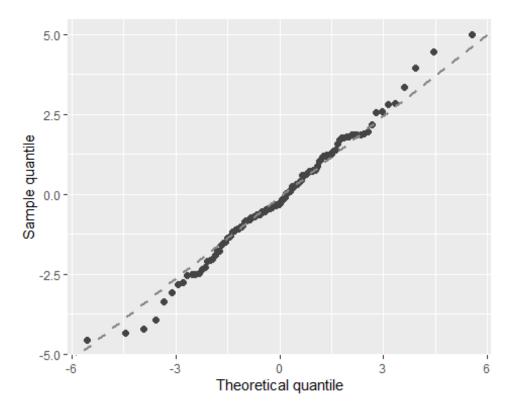


```
ggsave("og3.png", og3, width = 15, height = 10, units = "in", bg = "white")
m1a <- clm(Licence ~ Year, data = data)
## Warning in x$code == 0L || action == "silent": 'length(x) = 2 > 1' in coercion
## to 'logical(1)'
## Warning: (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
summary_m1a <- summary(m1a)

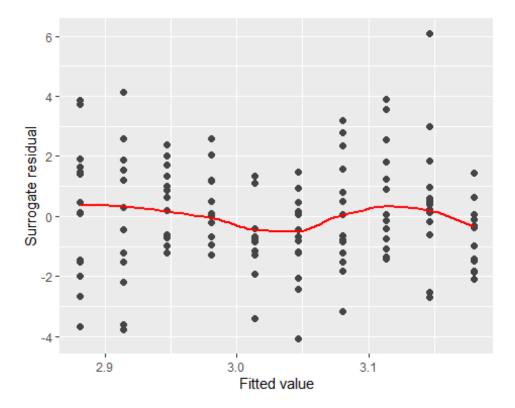
# Extract coefficients and standard errors
coefs <- summary_m1a$coefficients[, 1] # Coefficients
se_coefs <- summary_m1a$coefficients[, 2] # Standard errors</pre>
```

```
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci \leftarrow exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z_values <- coefs[,1] / coefs[,2]</pre>
p values <- 2 * (1 - pnorm(abs(z values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower ci, UpperCI = upper ci, p valu
e = p values)
print(results)
##
                          LowerCI
                  OR
                                        UpperCI
                                                   p value
## 1|3 7.570575e+27 7.413337e-86 7.731148e+140 0.6287208
## 3|4 6.239265e+28 6.046581e-85 6.438089e+141 0.6175010
## Year 1.033865e+00 9.088113e-01 1.176126e+00 0.6126309
#Check the Assumptions
nominal_test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
##
## formula: Licence ~ Year
          Df logLik
                       AIC LRT Pr(>Chi)
## <none>
             -93.915 193.83
## Year
scale test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here
## Warning: (-1) Model failed to converge with max grad = 33798.1 (tol = 1e-
## In addition: iteration limit reached
## Tests of scale effects
##
## formula: Licence ~ Year
          Df logLik
                        AIC LRT Pr(>Chi)
## <none>
             -93.915 193.83
## Year
convergence(m1a) # This is another way to assess the model
## nobs logLik niter max.grad cond.H logLik.Error
## 130 -93.92 6(0) 3.37e-10 3.9e+12 <1e-10
##
        Estimate Std.Err Gradient Error Cor.Dec Sig.Dig
##
```

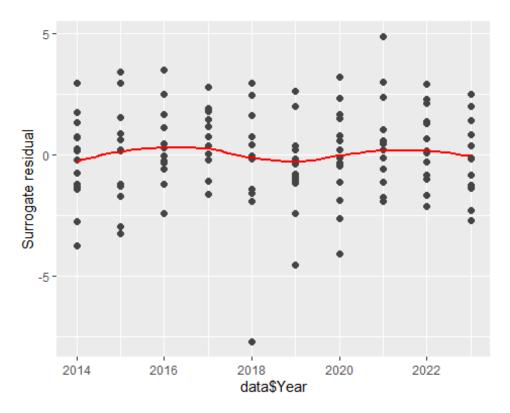
```
## 1 | 3 | 64.1941 132.76179 | 3.93e-14 -7.29e-12
                                                    10
                                                             12
## 3 4
         66.3032 132.76708 1.27e-13 -7.29e-12
                                                    10
                                                             12
                   0.06578 -3.37e-10 -3.62e-15
                                                             13
## Year
          0.0333
                                                    14
##
## Eigen values of Hessian:
## 1.093e+08 1.282e+01 2.837e-05
##
## Convergence message from clm:
## (2) Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## (3) Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'

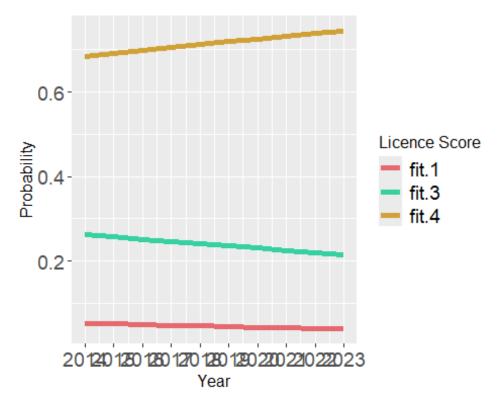


autoplot.clm(m1a, what = c("covariate"), x = data\$Year) # scale test indicate
d no variance issues. This residual, covariate plots seems acceptable as well
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



```
# N.B. - Interpreting residual plots is largely subjective!
###plotting the model
# Predict probabilities for each level of 'Complete'
new_data <- data.frame(Year = sort(unique(data$Year)))</pre>
pred_probs <- predict(m1a, newdata = new_data, type = "prob")</pre>
# Add the 'Year' column to the predicted probabilities dataframe
pred_probs_df <- cbind(new_data, as.data.frame(pred_probs))</pre>
# Convert the predicted probabilities to a long format for gaplot
library(reshape2)
pred_probs_long <- melt(pred_probs_df, id.vars = 'Year', variable.name = 'Lic</pre>
encelevel', value.name = 'PredictedProbability')
# Set the colors
my_colors <- c("#E5696F", "#36D0A1", "#D0A136")</pre>
# Modify agplot command
og4 <- ggplot(pred_probs_long, aes(x = Year, y = PredictedProbability, group
= Licencelevel, color = Licencelevel)) +
  geom_line(size = 2) + # Set size of the lines to make them thicker
  scale color manual(values = my colors) + # Use colors
  scale_x_continuous(breaks = unique(pred_probs_long$Year)) + # Show all yea
rs on x-axis
  labs(x = "Year", y = "Probability", color = "Licence Score") +
```

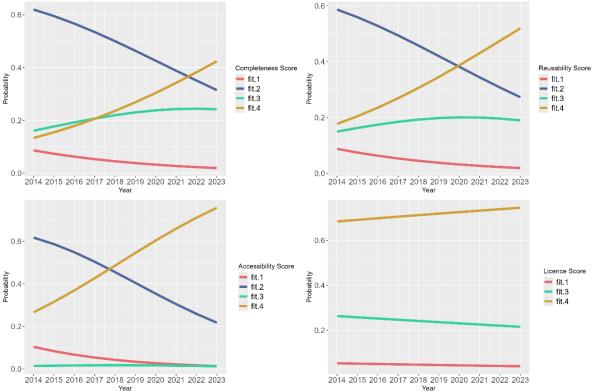
```
theme(
    axis.title = element_text(size = 12), # Increase axis titles
    axis.text = element_text(size = 14), # Increase axis text
    legend.title = element_text(size = 12), # Increase legend title
    legend.text = element_text(size = 14) # Increase legend text
)
print(og4)
```



```
ggsave("og4.png", og4, width = 15, height = 10, units = "in", bg = "white")
install.packages("patchwork")
## Installing package into 'M:/R/Win-Library/4.2'
## (as 'lib' is unspecified)
## package 'patchwork' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\s2146807\AppData\Local\Temp\RtmpGEDx57\downloaded_packages
library(patchwork)
## Warning: package 'patchwork' was built under R version 4.2.3
##
## Attaching package: 'patchwork'
```

```
## The following object is masked from 'package:MASS':
##
## area

ordinalplot <- og1 + og2 + og3 + og4 +
   plot_layout(
    ncol = 2, heights = c(10, 10), widths = c(10, 10)
   )
   print(ordinalplot)</pre>
```



```
ggsave("ordinalplot.png", ordinalplot, width = 15, height = 10, units = "in",
bg="white")
```

#Ordinal Regression models depending on the sharing projects

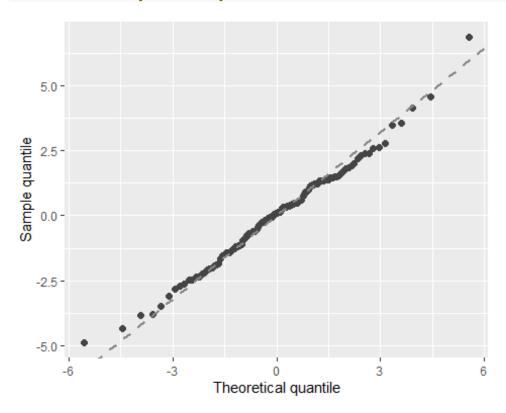
```
m1a <- clm(Complete ~ NewDAS + Preprint, data = data)
summary_m1a <- summary(m1a)

# Extract coefficients and standard errors
coefs <- summary_m1a$coefficients[, 1] # Coefficients
se_coefs <- summary_m1a$coefficients[, 2] # Standard errors

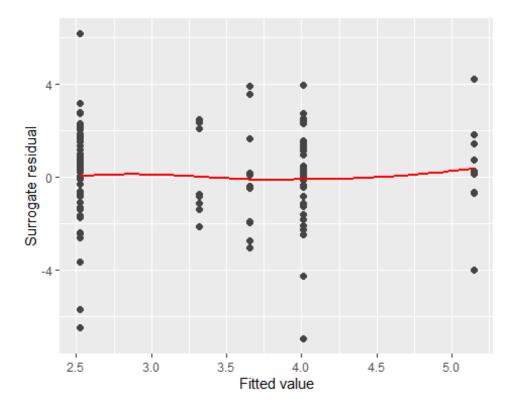
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])
upper_ci <- exp(coefs[,1] + 1.96 * coefs[,2])</pre>
```

```
# Calculate z-values and p-values
z values <- coefs[,1] / coefs[,2]</pre>
p values <- 2 * (1 - pnorm(abs(z values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower_ci, UpperCI = upper_ci, p_valu
e = p values)
print(results)
##
                                  LowerCI
                                             UpperCI
                            OR
                                                          p value
## 1 2
                    0.08040527 0.03446409 0.1875867 5.481457e-09
## 2|3
                    2.30928285 1.40274308 3.8016850 9.996808e-04
## 3 4
                    6.98540751 3.86569749 12.6227979 1.201850e-10
## NewDASNot shared 2.22062795 0.50650870 9.7356443 2.900738e-01
## NewDASShared
                    4.43200155 2.15464581 9.1164115 5.207790e-05
## Preprint
                    3.11203452 1.21055487 8.0002642 1.844070e-02
#Check the Assumptions
nominal test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
##
## formula: Complete ~ NewDAS + Preprint
##
            Df logLik
                          AIC
                                LRT Pr(>Chi)
               -141.46 294.91
## <none>
## NewDAS
## Preprint 2 -140.60 297.20 1.706
                                      0.4261
scale test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here because p = 0.8386. If p is less than 0.05 th
en assumptions are violated
## Tests of scale effects
## formula: Complete ~ NewDAS + Preprint
##
            Df logLik
                         AIC
                                  LRT Pr(>Chi)
               -141.46 294.91
## <none>
             2 -141.06 298.12 0.78479
## NewDAS
                                        0.6754
## Preprint 1 -141.39 296.78 0.12885
                                        0.7196
convergence(m1a) # This is another way to assess the model
    nobs logLik niter max.grad cond.H logLik.Error
##
  130 -141.45 6(0) 2.36e-08 3.6e+01 <1e-10
##
##
                    Estimate Std.Err Gradient
                                                   Error Cor.Dec Sig.Dig
## 1 2
                     -2.5207 0.4322 2.36e-08 4.00e-09
                                                               8
## 2 3
                      0.8369 0.2543 -1.09e-08 -1.02e-10
                                                               9
                                                                       9
## 3 4
                      1.9438 0.3019 -5.49e-12 -9.47e-11
```

```
## NewDASNot shared
                     0.7978 0.7541 -6.88e-10 -9.85e-11
                                                               9
                                                                      10
## NewDASShared
                     1.4889 0.3680 -9.83e-10 -9.04e-11
## Preprint
                      1.1353 0.4817 -6.26e-10 -6.25e-11
                                                               9
                                                                      10
##
## Eigen values of Hessian:
## 58.846 25.507 6.644 5.242 3.748 1.627
##
## Convergence message from clm:
## (0) successful convergence
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



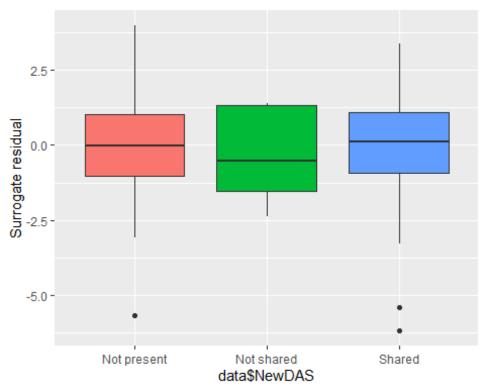
autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



autoplot.clm(m1a, what = c("covariate"), x = data\$NewDAS) # scale test indica
ted no variance issues. This residual, covariate plots seems acceptable as we
LL

Warning: The `<scale>` argument of `guides()` cannot be `FALSE`. Use "none
" instead as
of ggplot2 3.3.4.

i The deprecated feature was likely used in the sure package.
Please report the issue at <[8;;https://github.com/AFIT-R/sure/issueshtt
ps://github.com/AFIT-R/sure/issues]8;;>.
This warning is displayed once every 8 hours.
Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
generated.

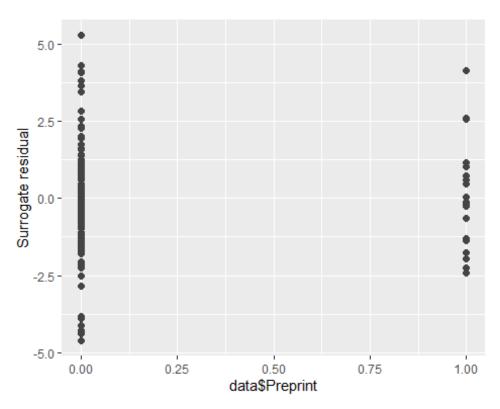


```
autoplot.clm(m1a, what = c("covariate"), x = data$Preprint)
## geom_smooth() using method = 'loess' and formula = 'y \sim x'
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : radius 2.5e-05
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : all data on boundary of neighborhood. make span bigger
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : pseudoinverse used at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : neighborhood radius 0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : reciprocal condition number 1
```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : There are other near singularities as well. 1.01

## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : zero-width neighborhood. make span bigger

## Warning: Failed to fit group -1.
## Caused by error in `predLoess()`:
## ! NA/NaN/Inf in foreign function call (arg 5)
```



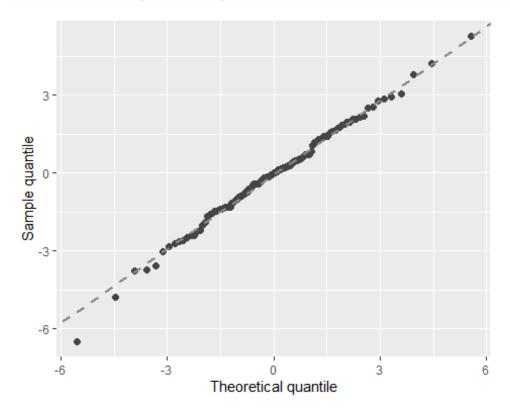
N.B. - Interpreting residual plots is largely subjective!
m1a <- clm(Reuse ~ NewDAS + Preprint, data = data)
summary_m1a <- summary(m1a)

Extract coefficients and standard errors
coefs <- summary_m1a\$coefficients[, 1] # Coefficients
se_coefs <- summary_m1a\$coefficients[, 2] # Standard errors

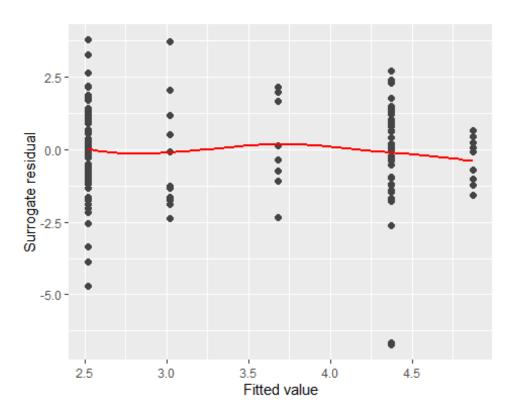
Extract coefficients and standard errors
coefs <- summary(m1a)\$coefficients
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])
upper_ci <- exp(coefs[,1] + 1.96 * coefs[,2])</pre>

```
# Calculate z-values and p-values
z values <- coefs[,1] / coefs[,2]</pre>
p_values <- 2 * (1 - pnorm(abs(z_values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower ci, UpperCI = upper ci, p valu
e = p values)
print(results)
##
                            OR
                                  LowerCI
                                             UpperCI
                                                           p_value
## 1 2
                    0.08030323 0.03449122 0.1869638 4.945615e-09
## 2 3
                    1.94089488 1.19689021 3.1473839 7.173334e-03
## 3 4
                    4.95212549 2.84354848 8.6242760 1.584018e-08
## NewDASNot shared 3.19798580 0.83340049 12.2715469 9.019262e-02
## NewDASShared
                    6.37209386 3.00525060 13.5108800 1.367981e-06
## Preprint
                    1.64953102 0.65229017 4.1713837 2.903525e-01
#Check the Assumptions
nominal_test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
## formula: Reuse ~ NewDAS + Preprint
##
            Df logLik
                          AIC
                                 LRT Pr(>Chi)
## <none>
               -139.92 291.85
## NewDAS
             4 -138.18 296.36 3.4880
                                       0.4797
## Preprint 2 -139.22 294.43 1.4187
                                       0.4920
scale test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here because p = 0.8386. If p is less than 0.05 th
en assumptions are violated
## Tests of scale effects
##
## formula: Reuse ~ NewDAS + Preprint
##
                                 LRT Pr(>Chi)
            Df logLik
                          AIC
## <none>
               -139.92 291.85
## NewDAS
             2 -138.55 293.11 2.7396
                                       0.2542
## Preprint 1 -139.28 292.57 1.2817
                                       0.2576
convergence(m1a) # This is another way to assess the model
    nobs logLik niter max.grad cond.H logLik.Error
##
    130 -139.92 6(0) 4.40e-09 3.6e+01 <1e-10
##
##
                    Estimate Std.Err Gradient
                                                    Error Cor.Dec Sig.Dig
## 1 2
                     -2.5219 0.4312 4.40e-09
                                                7.42e-10
                                                                8
                                                                        9
## 2 3
                      0.6631 0.2466 -2.20e-09 -1.84e-11
                                                               10
                                                                       10
## 3 4
                      1.5998 0.2830 -1.29e-12 -1.65e-11
                                                               10
                                                                       11
## NewDASNot shared 1.1625 0.6861 -5.46e-11 -1.75e-11
                                                               10
                                                                       11
```

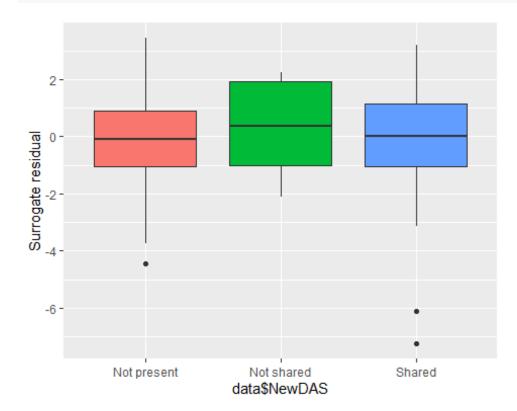
```
## NewDASShared
                      1.8519 0.3835 -1.05e-10 -1.57e-11
                                                              10
                                                                      11
## Preprint
                      0.5005  0.4733 -2.18e-10 -7.40e-12
                                                              10
                                                                      10
##
## Eigen values of Hessian:
## 68.197 24.184 6.638 5.094 4.107 1.915
## Convergence message from clm:
## (0) successful convergence
## In addition: Absolute and relative convergence criteria were met
####### Graphically validate proportional odds using the sure package ######
#
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s
o no violation of linearity
```



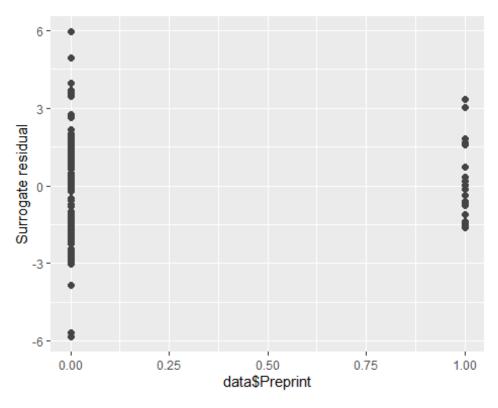
autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



autoplot.clm(m1a, what = c("covariate"), x = data\$NewDAS) # scale test indica ted no variance issues. This residual, covariate plots seems acceptable as we ll



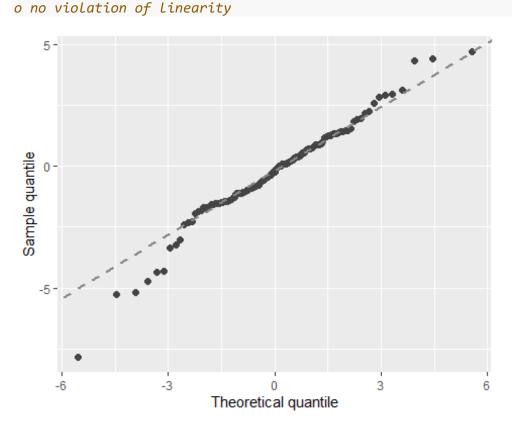
```
autoplot.clm(m1a, what = c("covariate"), x = data$Preprint)
## geom_smooth() using method = 'loess' and formula = 'y \sim x'
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : radius 2.5e-05
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : all data on boundary of neighborhood. make span bigger
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : pseudoinverse used at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : neighborhood radius 0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : reciprocal condition number 1
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : There are other near singularities as well. 1.01
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
## : zero-width neighborhood. make span bigger
## Warning: Failed to fit group -1.
## Caused by error in `predLoess()`:
## ! NA/NaN/Inf in foreign function call (arg 5)
```



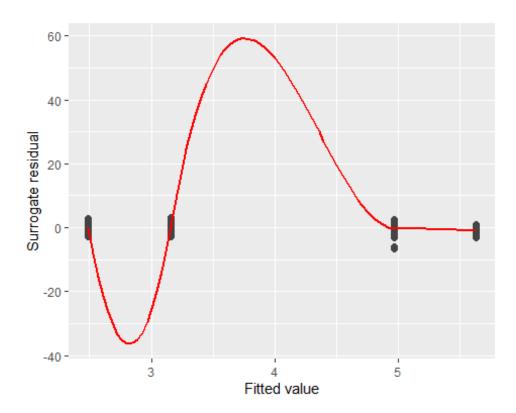
```
# N.B. - Interpreting residual plots is largely subjective!
m1a <- clm(Access ~ NewDAS + Preprint, data = data)</pre>
summary_m1a <- summary(m1a)</pre>
# Extract coefficients and standard errors
coefs <- summary_m1a$coefficients[, 1] # Coefficients</pre>
se_coefs <- summary_m1a$coefficients[, 2] # Standard errors</pre>
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci \leftarrow exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z_values <- coefs[,1] / coefs[,2]</pre>
p_values <- 2 * (1 - pnorm(abs(z_values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower_ci, UpperCI = upper_ci, p_valu
e = p values)
print(results)
##
                              OR
                                     LowerCI
                                                UpperCI
                                                              p_value
## 1 2
                      0.08293078 0.03554014 0.193514 8.458039e-09
                      2.17869399 1.30603472 3.634442 2.857815e-03
## 2 3
```

```
## 3 4
                     2.37186800 1.41398150 3.978664 1.065660e-03
## NewDASNot shared 1.94745782 0.49961942 7.590962 3.369174e-01
                   11.85644530 4.63730330 30.314018 2.428559e-07
## NewDASShared
## Preprint
                     1.94425820 0.64398445 5.869924 2.382481e-01
#Check the Assumptions
nominal_test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
##
## formula: Access ~ NewDAS + Preprint
                                 LRT Pr(>Chi)
##
           Df
                logLik
                          AIC
## <none>
               -100.310 212.62
## NewDAS
## Preprint 2 -88.681 193.36 23.258 8.903e-06 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
scale_test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here because p = 0.8386. If p is less than 0.05 th
en assumptions are violated
## Tests of scale effects
## formula: Access ~ NewDAS + Preprint
##
           Df
                           AIC
                                  LRT Pr(>Chi)
                 logLik
## <none>
               -100.310 212.62
             2 -96.004 208.01 8.6128
                                       0.01348 *
## NewDAS
## Preprint 1 -100.232 214.46 0.1562 0.69268
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
convergence(m1a) # This is another way to assess the model
   nobs logLik niter max.grad cond.H logLik.Error
## 130 -100.31 7(2) 6.61e-08 3.2e+02 <1e-10
##
##
                    Estimate Std.Err Gradient
                                                   Error Cor.Dec Sig.Dig
## 1 2
                     -2.4897 0.4323 6.61e-08 1.12e-08
                                                               7
                                                                       8
## 2 3
                                                               9
                                                                       9
                      0.7787 0.2611 -3.06e-08 -3.35e-10
                                                               9
                                                                       9
## 3 4
                     0.8637   0.2639   -3.75e-11   -3.30e-10
                                                               9
                                                                       9
## NewDASNot shared
                     0.6665 0.6941 -1.80e-09 -3.29e-10
## NewDASShared
                     2.4729 0.4789 -5.62e-10 -3.04e-10
                                                               9
                                                                      10
                                                               9
                                                                       9
## Preprint
                     0.6649 0.5638 -2.73e-09 -2.63e-10
##
## Eigen values of Hessian:
## 566.901 15.759
                    5.927
                             4.135
                                     3.161
                                             1.781
##
## Convergence message from clm:
```

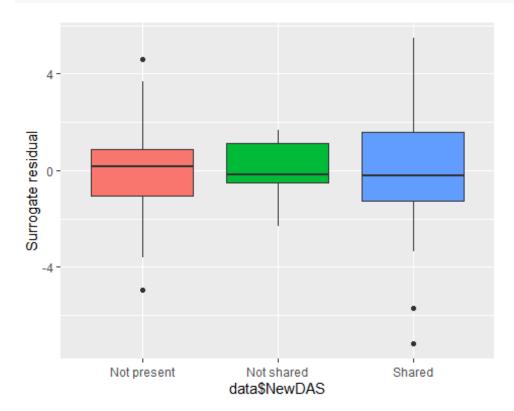
(0) successful convergence
In addition: Absolute and relative convergence criteria were met
######## Graphically validate proportional odds using the sure package #####
#
autoplot.clm(m1a, what = c("qq")) # most of the points fall along the line, s



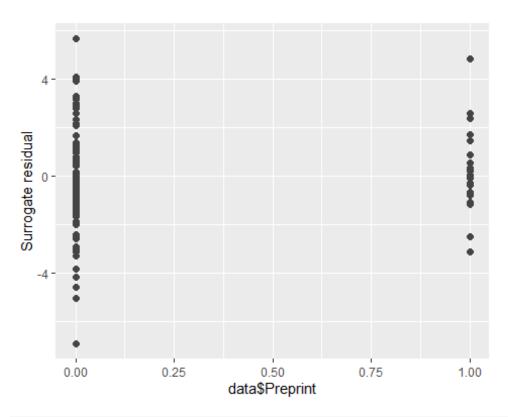
autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat
tern or trend
`geom_smooth()` using method = 'loess' and formula = 'y ~ x'



autoplot.clm(m1a, what = c("covariate"), x = data\$NewDAS) # scale test indica ted no variance issues. This residual, covariate plots seems acceptable as we ll



```
autoplot.clm(m1a, what = c("covariate"), x = data$Preprint)
## geom_smooth() using method = 'loess' and formula = 'y \sim x'
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : radius 2.5e-05
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : all data on boundary of neighborhood. make span bigger
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : pseudoinverse used at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : neighborhood radius 0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : reciprocal condition number 1
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : There are other near singularities as well. 1.01
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
## : zero-width neighborhood. make span bigger
## Warning: Failed to fit group -1.
## Caused by error in `predLoess()`:
## ! NA/NaN/Inf in foreign function call (arg 5)
```

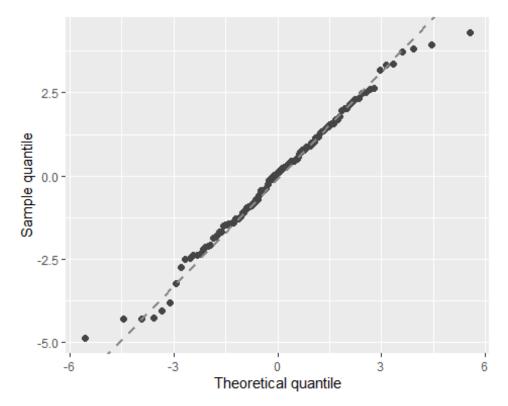


```
# N.B. - Interpreting residual plots is largely subjective!
m1a <- clm(Licence ~ NewDAS + Preprint, data = data)</pre>
summary_m1a <- summary(m1a)</pre>
# Extract coefficients and standard errors
coefs <- summary_m1a$coefficients[, 1] # Coefficients</pre>
se_coefs <- summary_m1a$coefficients[, 2] # Standard errors</pre>
# Extract coefficients and standard errors
coefs <- summary(m1a)$coefficients</pre>
OR <- exp(coefs[,1])
lower_ci <- exp(coefs[,1] - 1.96 * coefs[,2])</pre>
upper_ci \leftarrow exp(coefs[,1] + 1.96 * coefs[,2])
# Calculate z-values and p-values
z_values <- coefs[,1] / coefs[,2]</pre>
p_values <- 2 * (1 - pnorm(abs(z_values)))</pre>
# Combine everything into a data frame for easy viewing
results <- data.frame(OR = OR, LowerCI = lower_ci, UpperCI = upper_ci, p_valu
e = p values)
print(results)
##
                             OR
                                    LowerCI
                                                UpperCI
                                                              p_value
## 1|3
                     0.06281708 0.02618925 0.1506720 5.641767e-10
## 3 4
                     0.56911004 0.34534082 0.9378742 2.699035e-02
```

```
## NewDASNot shared 1.78733704 0.33990304 9.3984853 4.928701e-01
## NewDASShared
                   4.46125540 1.67179567 11.9050432 2.824748e-03
## Preprint
                   0.54460152 0.19750155 1.5017138 2.402780e-01
#Check the Assumptions
nominal_test(m1a) # This is a test of the proportional odds assumption, odds
are proportional in this case.
## Tests of nominal effects
## formula: Licence ~ NewDAS + Preprint
##
           Df logLik
                         AIC
                                LRT Pr(>Chi)
              -88.207 186.41
## <none>
## NewDAS
            2 -86.914 187.83 2.5854 0.27453
## Preprint 1 -86.238 184.48 3.9372 0.04723 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
scale_test(m1a) # scale test checks for equal variance/ scale across year, as
sumptions are not violated here because p = 0.8386. If p is less than 0.05 th
en assumptions are violated
## Warning: (-1) Model failed to converge with max|grad| = 2.6247e-05 (tol =
1e-06)
## In addition: iteration limit reached
## Tests of scale effects
## formula: Licence ~ NewDAS + Preprint
##
           Df logLik
                         AIC
                                LRT Pr(>Chi)
## <none>
              -88.207 186.41
## NewDAS
## Preprint 1 -87.174 186.35 2.0665
                                      0.1506
convergence(m1a) # This is another way to assess the model
   nobs logLik niter max.grad cond.H logLik.Error
  130 -88.21 5(0) 4.39e-08 2.2e+01 <1e-10
##
##
                   Estimate Std.Err Gradient
##
                                                  Error Cor.Dec Sig.Dig
## 1 3
                    -2.7675 0.4464 4.39e-08 6.85e-09
                                                             7
                                                                     8
                    -0.5637   0.2549   -3.38e-08   -1.62e-10
                                                             9
                                                                     9
## 3 4
                                                                     9
## NewDASNot shared
                     9
                                                             9
## NewDASShared
                     1.4954 0.5008 -6.23e-10 -2.03e-10
                                                                    10
## Preprint
                    -0.6077 0.5175 -2.72e-09 2.21e-10
                                                             9
                                                                     9
##
## Eigen values of Hessian:
## 29.924 6.834 3.658 3.311 1.343
##
## Convergence message from clm:
```

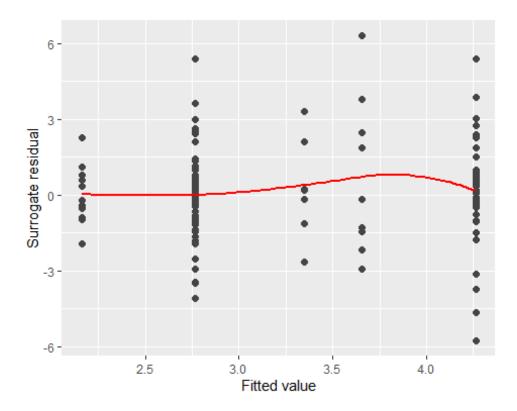
(0) successful convergence
In addition: Absolute and relative convergence criteria were met
######## Graphically validate proportional odds using the sure package ######
#



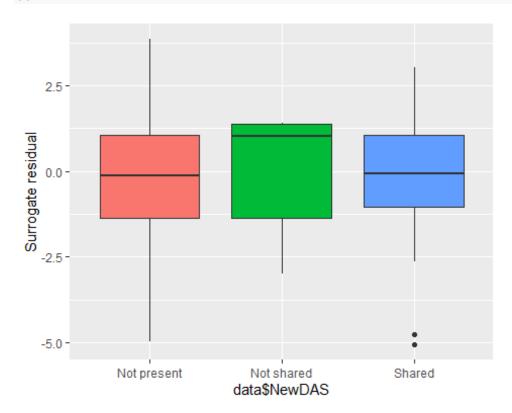


autoplot.clm(m1a, what = c("fitted")) # the residuals do not show a clear pat tern or trend

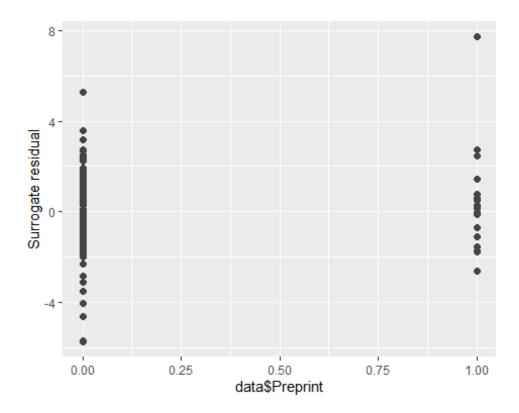
$geom_smooth()$ using method = 'loess' and formula = 'y ~ x'



autoplot.clm(m1a, what = c("covariate"), x = data\$NewDAS) # scale test indica
ted no variance issues. This residual, covariate plots seems acceptable as we
tl



```
autoplot.clm(m1a, what = c("covariate"), x = data$Preprint)
## geom_smooth() using method = 'loess' and formula = 'y \sim x'
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : radius 2.5e-05
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : all data on boundary of neighborhood. make span bigger
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : pseudoinverse used at -0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : neighborhood radius 0.005
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : reciprocal condition number 1
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
tric,
## : There are other near singularities as well. 1.01
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = parame
## : zero-width neighborhood. make span bigger
## Warning: Failed to fit group -1.
## Caused by error in `predLoess()`:
## ! NA/NaN/Inf in foreign function call (arg 5)
```



N.B. - Interpreting residual plots is largely subjective