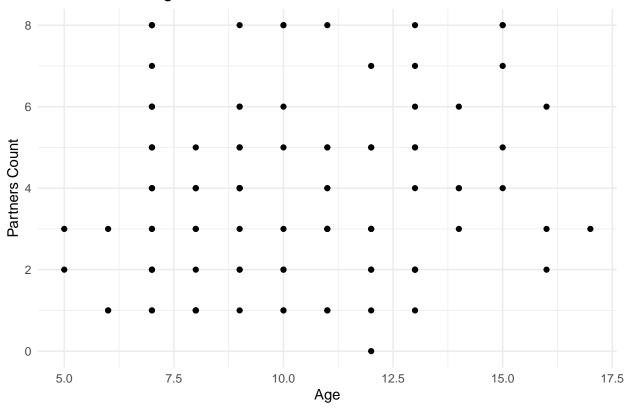
MA675GroupC_Brooke

2024-10-11

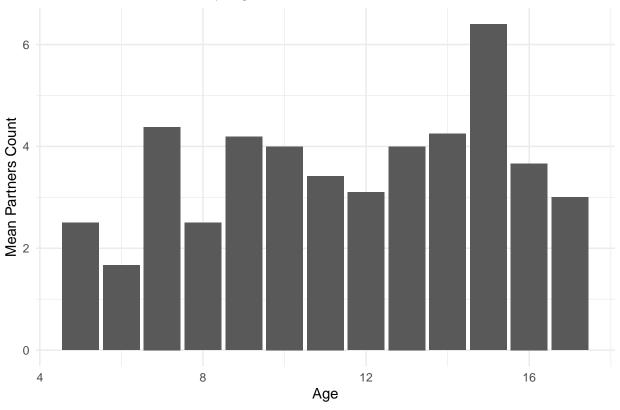
```
# Load necessary libraries
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.4
                       v readr
                                   2.1.5
## v forcats 1.0.0
                       v stringr 1.5.1
## v ggplot2 3.5.1 v tibble 3.2.1
## v lubridate 1.9.3
                     v tidyr
                                  1.3.1
## v purrr
             1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
# Load the dataset
data <- read.csv("Individual-level basic variable.csv")</pre>
# Ensure Age and Partners_count are numeric
data$Age <- as.numeric(data$Age)</pre>
data$Partners_count <- as.numeric(data$Partners_count)</pre>
# Create a scatter plot
ggplot(data, aes(x = Age, y = Partners_count)) +
 geom_point() +
 labs(title = "Scatter Plot of Age vs. Partners Count",
      x = "Age",
      y = "Partners Count") +
  theme_minimal()
```

Scatter Plot of Age vs. Partners Count



```
# Load necessary libraries
library(tidyverse)
# Load the dataset
data <- read.csv("Individual-level basic variable.csv")</pre>
# Ensure Age and Partners_count are numeric
data$Age <- as.numeric(data$Age)</pre>
data$Partners_count <- as.numeric(data$Partners_count)</pre>
# Group by Age and calculate the mean of Partners_count
age_partner_mean <- data %>%
  group_by(Age) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
# Create a bar plot
ggplot(age_partner_mean, aes(x = Age, y = mean_partners)) +
  geom_bar(stat = "identity") +
 labs(title = "Mean Partners Count by Age",
       x = "Age",
       y = "Mean Partners Count") +
  theme_minimal()
```





```
# Load necessary libraries
library(tidyverse)

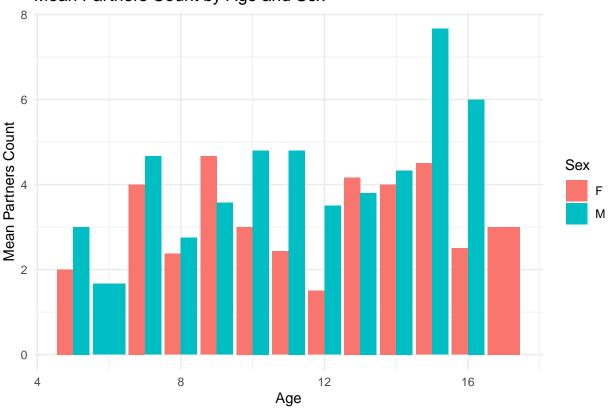
# Load the dataset
data <- read.csv("Individual-level basic variable.csv")

# Ensure Age, Partners_count, and Sex are in the correct format
data$Age <- as.numeric(data$Age)
data$Partners_count <- as.numeric(data$Partners_count)
data$Sex <- as.factor(data$Sex) # Make sure Sex is a factor (M/F)

# Group by Age and Sex, then calculate the mean of Partners_count
age_sex_partner_mean <- data %>%
group_by(Age, Sex) %>%
summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
```

```
fill = "Sex") +
theme_minimal()
```

Mean Partners Count by Age and Sex



```
# Load necessary libraries
library(tidyverse)

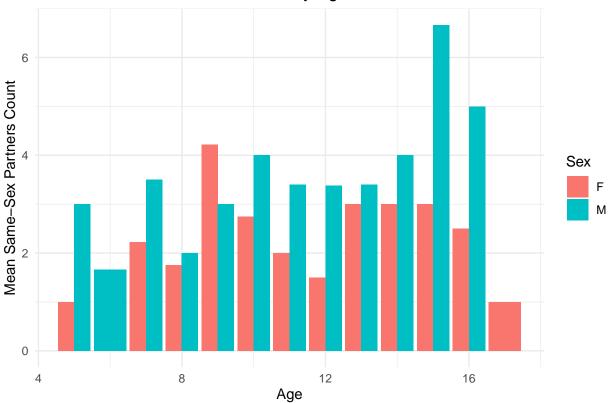
# Load the dataset
data <- read.csv("Individual-level basic variable.csv")

# Ensure Age, Count_same_sex, and Sex are in the correct format
data$Age <- as.numeric(data$Age)
data$Count_same_sex <- as.numeric(data$Count_same_sex)
data$Sex <- as.factor(data$Sex) # Make sure Sex is a factor (M/F)

# Group by Age and Sex, then calculate the mean of Count_same_sex
age_sex_same_sex_mean <- data %>%
group_by(Age, Sex) %>%
summarise(mean_same_sex = mean(Count_same_sex, na.rm = TRUE))
```

```
# Create a bar plot with divided bars for Males and Females
ggplot(age_sex_same_sex_mean, aes(x = Age, y = mean_same_sex, fill = Sex)) +
```

Mean Same-Sex Partners Count by Age and Sex



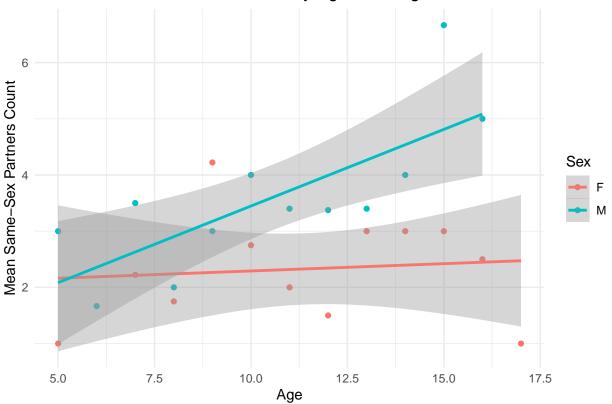
```
# Group by Age and Sex, then calculate the mean of Count_same_sex
age_sex_same_sex_mean <- data %>%
group_by(Age, Sex) %>%
summarise(mean_same_sex = mean(Count_same_sex, na.rm = TRUE))
```

```
# Fit linear regression models for each sex
model_male <- lm(mean_same_sex ~ Age, data = age_sex_same_sex_mean %>% filter(Sex == "M"))
model_female <- lm(mean_same_sex ~ Age, data = age_sex_same_sex_mean %>% filter(Sex == "F"))
# Print summaries for both models
summary(model_male)
```

```
##
## Call:
```

```
## lm(formula = mean_same_sex ~ Age, data = age_sex_same_sex_mean %>%
##
      filter(Sex == "M"))
##
## Residuals:
               1Q Median
                               3Q
                                      Max
## -0.9020 -0.6361 -0.2476 0.6320 1.8550
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.71947
                          0.83681
                                   0.860 0.41005
## Age
               0.27281
                          0.07571
                                    3.603 0.00482 **
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.9054 on 10 degrees of freedom
## Multiple R-squared: 0.5649, Adjusted R-squared: 0.5214
## F-statistic: 12.98 on 1 and 10 DF, p-value: 0.00482
summary(model_female)
##
## Call:
## lm(formula = mean_same_sex ~ Age, data = age_sex_same_sex_mean %>%
      filter(Sex == "F"))
##
##
## Residuals:
                 1Q
                     Median
## -1.47370 -0.57844 0.03024 0.58473 1.95628
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.03222
                          0.94764
                                    2.144
                                            0.0576 .
               0.02597
                          0.07918
                                    0.328
                                            0.7497
## Age
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9855 on 10 degrees of freedom
## Multiple R-squared: 0.01064, Adjusted R-squared: -0.08829
## F-statistic: 0.1076 on 1 and 10 DF, p-value: 0.7497
# Scatter plot with regression line for both males and females
ggplot(age_sex_same_sex_mean, aes(x = Age, y = mean_same_sex, color = Sex)) +
 geom_point() + # Scatter plot
 geom_smooth(method = "lm", se = TRUE) + # Regression line with confidence interval
 labs(title = "Mean Same-Sex Partners Count by Age with Regression Line",
      x = "Age",
      y = "Mean Same-Sex Partners Count",
      color = "Sex") +
 theme_minimal()
## 'geom_smooth()' using formula = 'y ~ x'
```

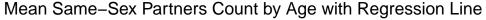
Mean Same-Sex Partners Count by Age with Regression Line

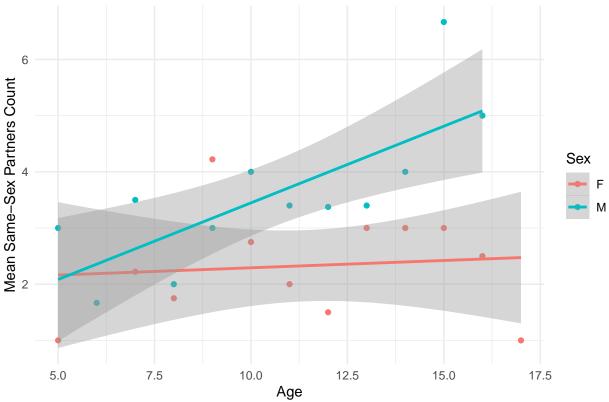


```
# Fit a combined model with interaction between Age and Sex
combined_model <- lm(mean_same_sex ~ Age * Sex, data = age_sex_same_sex_mean)
# Print the summary of the combined model
summary(combined_model)</pre>
```

```
##
## lm(formula = mean_same_sex ~ Age * Sex, data = age_sex_same_sex_mean)
## Residuals:
      Min
                1Q Median
                                3Q
                                       Max
## -1.4737 -0.6361 -0.1297 0.5847
                                   1.9563
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 2.03222
                           0.90993
                                     2.233
                                             0.0371 *
               0.02597
                           0.07603
                                     0.342
                                             0.7362
## Age
## SexM
               -1.31274
                           1.26212
                                    -1.040
                                             0.3107
               0.24684
                           0.10973
                                             0.0359 *
## Age:SexM
                                     2.249
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.9463 on 20 degrees of freedom
## Multiple R-squared: 0.5301, Adjusted R-squared: 0.4596
## F-statistic: 7.521 on 3 and 20 DF, p-value: 0.001469
```

'geom_smooth()' using formula = 'y ~ x'





```
# Load necessary libraries
library(tidyverse)

# Load the dataset
data <- read.csv("Individual-level basic variable.csv")

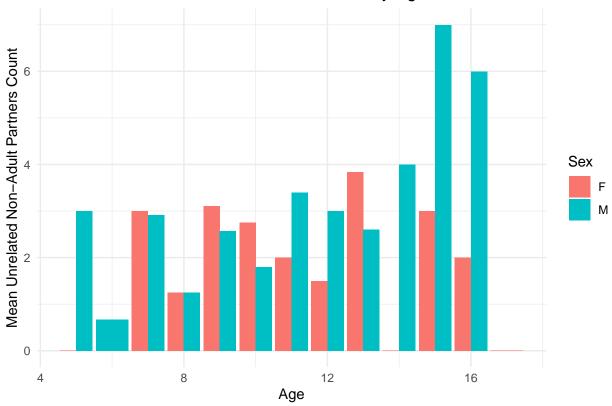
# Ensure Age, Count_unrelated_noadult, and Sex are in the correct format
data$Age <- as.numeric(data$Age)
data$Count_unrelated_noadult <- as.numeric(data$Count_unrelated_noadult) # Assuming this variable repr
data$Sex <- as.factor(data$Sex) # Ensure Sex is a factor (M/F)

# Group by Age and Sex, then calculate the mean of Count_unrelated_noadult
age_sex_unrelated_noadult_mean <- data %>%
```

```
group_by(Age, Sex) %>%
summarise(mean_unrelated_noadult = mean(Count_unrelated_noadult, na.rm = TRUE))
```

'summarise()' has grouped output by 'Age'. You can override using the '.groups'
argument.

Mean Unrelated Non-Adult Partners Count by Age and Sex

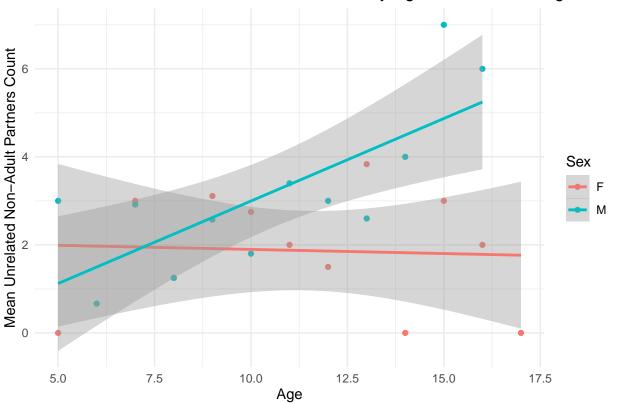


```
# Group by Age and Sex, then calculate the mean of Count_unrelated_noadult
age_sex_unrelated_noadult_mean <- data %>%
group_by(Age, Sex) %>%
summarise(mean_unrelated_noadult = mean(Count_unrelated_noadult, na.rm = TRUE))
```

```
# Fit separate linear regression models for Males and Females
model_male <- lm(mean_unrelated_noadult ~ Age, data = age_sex_unrelated_noadult_mean %>% filter(Sex ==
model_female <- lm(mean_unrelated_noadult ~ Age, data = age_sex_unrelated_noadult_mean %>% filter(Sex =
# Print the summary of the models for Males and Females
cat("Summary for Male Model:\n")
## Summary for Male Model:
summary(model_male)
##
## Call:
## lm(formula = mean_unrelated_noadult ~ Age, data = age_sex_unrelated_noadult_mean %>%
       filter(Sex == "M"))
##
## Residuals:
      Min
               1Q Median
                               ЗQ
                                      Max
## -1.5210 -0.8717 -0.2729 0.8270 2.1293
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.7526
                          1.1675 -0.645 0.53366
                0.3749
                           0.1056 3.549 0.00528 **
## Age
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.263 on 10 degrees of freedom
## Multiple R-squared: 0.5575, Adjusted R-squared: 0.5132
## F-statistic: 12.6 on 1 and 10 DF, p-value: 0.005275
cat("\nSummary for Female Model:\n")
##
## Summary for Female Model:
summary(model_female)
##
## lm(formula = mean_unrelated_noadult ~ Age, data = age_sex_unrelated_noadult_mean %>%
       filter(Sex == "F"))
##
##
## Residuals:
               1Q Median
##
      Min
                               3Q
                                      Max
## -1.9908 -0.9548 0.1687 1.0839 1.9927
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.08463 1.35018 1.544
                                             0.871
              -0.01877 0.11281 -0.166
## Age
```

'geom_smooth()' using formula = 'y ~ x'

Mean Unrelated Non-Adult Partners Count by Age and Sex with Regression

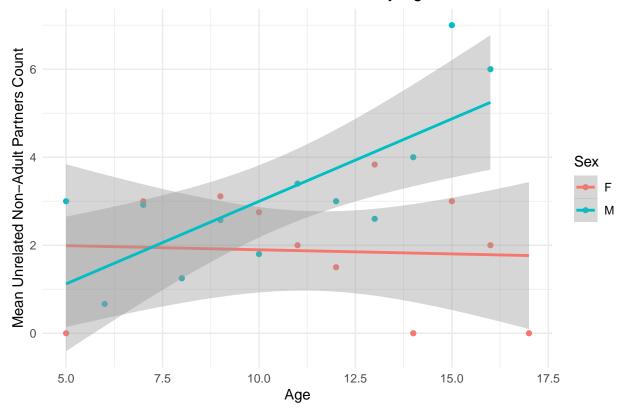


```
# Fit a combined model with interaction between Age and Sex
combined_model <- lm(mean_unrelated_noadult ~ Age * Sex, data = age_sex_unrelated_noadult_mean)
# Print the summary of the combined model
summary(combined_model)</pre>
```

##

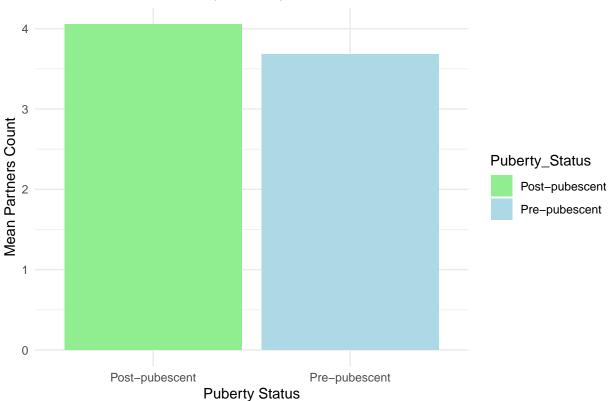
```
## Call:
## lm(formula = mean_unrelated_noadult ~ Age * Sex, data = age_sex_unrelated_noadult_mean)
## Residuals:
                 1Q Median
                                   3Q
## -1.99080 -0.87166 -0.01057 1.04548 2.12926
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.08463 1.28419
                                  1.623 0.1202
## Age
              -0.01877
                          0.10730 -0.175
                                           0.8629
              -2.83727
                          1.78123 -1.593 0.1269
## SexM
              0.39366
                          0.15487
                                  2.542 0.0194 *
## Age:SexM
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.335 on 20 degrees of freedom
## Multiple R-squared: 0.461, Adjusted R-squared: 0.3801
## F-statistic: 5.701 on 3 and 20 DF, p-value: 0.005468
# Scatter plot with regression line for both males and females
ggplot(age_sex_unrelated_noadult_mean, aes(x = Age, y = mean_unrelated_noadult, color = Sex)) +
 geom_point() + # Scatter plot
 geom_smooth(method = "lm", se = TRUE) + # Regression line with confidence interval
 labs(title = "Mean Unrelated Non-Adult Partners Count by Age and Sex with Combined Regression",
      x = "Age",
      y = "Mean Unrelated Non-Adult Partners Count",
      color = "Sex") +
 theme minimal()
```

Mean Unrelated Non-Adult Partners Count by Age and Sex with Combined F



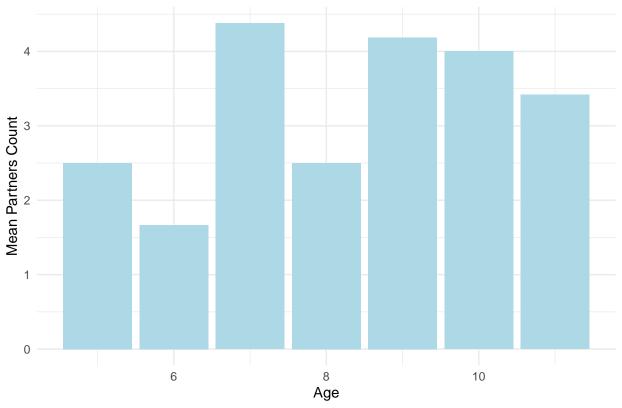
```
# Load necessary libraries
library(tidyverse)
# Load the dataset
data <- read.csv("Individual-level basic variable.csv")</pre>
# Ensure Age and Partners count are numeric
data$Age <- as.numeric(data$Age)</pre>
data$Partners_count <- as.numeric(data$Partners_count)</pre>
# Create a new variable for pre-pubescent and post-pubescent
data <- data %>%
  mutate(Puberty_Status = ifelse(Age < 12, "Pre-pubescent", "Post-pubescent"))</pre>
# Group by Puberty_Status and calculate the mean of Partners_count
puberty_partner_mean <- data %>%
  group_by(Puberty_Status) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
# Create a bar plot
ggplot(puberty_partner_mean, aes(x = Puberty_Status, y = mean_partners, fill = Puberty_Status)) +
  geom_bar(stat = "identity") +
  labs(title = "Mean Partners Count by Puberty Status",
       x = "Puberty Status",
       y = "Mean Partners Count") +
  theme minimal() +
```





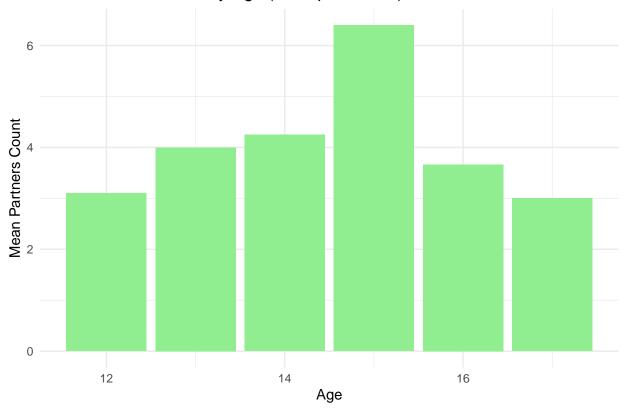
```
pre_pubescent_data <- data %>% filter(Puberty_Status == "Pre-pubescent")
post_pubescent_data <- data %>% filter(Puberty_Status == "Post-pubescent")
# Group by Age and calculate the mean of Partners_count for both groups
pre pubescent mean <- pre pubescent data %>%
  group_by(Age) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
post_pubescent_mean <- post_pubescent_data %>%
  group_by(Age) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
# Create the bar plots for both groups
# Pre-pubescent plot
pre_plot <- ggplot(pre_pubescent_mean, aes(x = Age, y = mean_partners)) +</pre>
  geom_bar(stat = "identity", fill = "lightblue") +
  labs(title = "Mean Partners Count by Age (Pre-pubescent)",
       x = "Age",
       y = "Mean Partners Count") +
  theme_minimal()
# Post-pubescent plot
```

Mean Partners Count by Age (Pre-pubescent)



print(post_plot)

Mean Partners Count by Age (Post-pubescent)

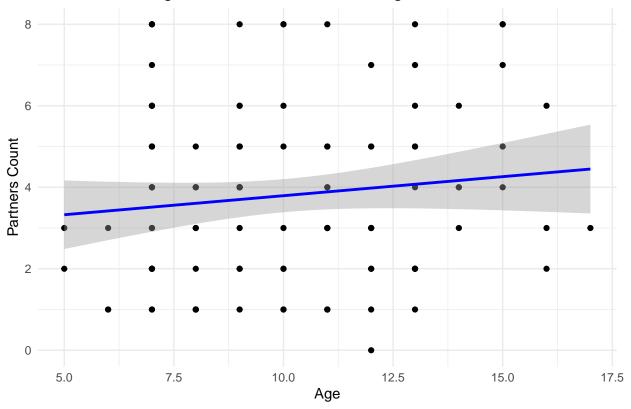


```
model <- lm(Partners_count ~ Age, data = data)
# Print the summary of the regression model
summary(model)</pre>
```

```
##
## Call:
## lm(formula = Partners_count ~ Age, data = data)
##
## Residuals:
       {\tt Min}
##
                1Q Median
                                ЗQ
                                       Max
## -3.9799 -1.6996 -0.1668 1.3004 4.4873
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                          0.76708
                                    3.727 0.000312 ***
## (Intercept) 2.85857
                0.09345
                          0.07351
                                     1.271 0.206432
## Age
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 2.14 on 107 degrees of freedom
## Multiple R-squared: 0.01488, Adjusted R-squared: 0.00567
## F-statistic: 1.616 on 1 and 107 DF, p-value: 0.2064
```

'geom_smooth()' using formula = 'y ~ x'

Scatter Plot of Age vs. Partners Count with Regression Line



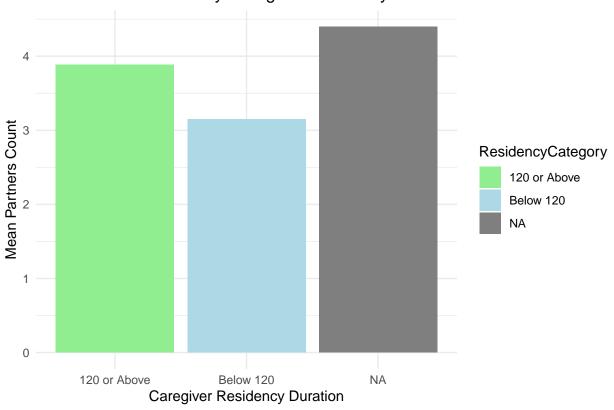
```
# Load necessary libraries
library(tidyverse)

# Load the dataset
df <- read.csv("Individual-level basic variable.csv")

# Ensure Partners_count and CaregiverResidencyDuration are numeric
df$Partners_count <- as.numeric(df$Partners_count)
df$CaregiverResidencyDuration <- as.numeric(df$CaregiverResidencyDuration)

# Create a new variable for CaregiverResidencyDuration categories
df <- df %>%
    mutate(ResidencyCategory = ifelse(CaregiverResidencyDuration < 120, 'Below 120', '120 or Above'))</pre>
```

Mean Partners Count by Caregiver Residency Duration



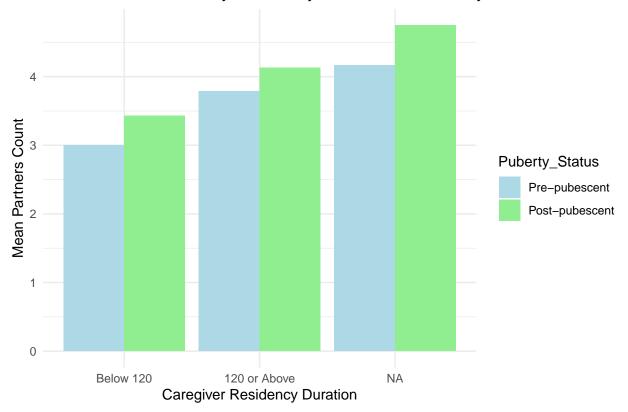
```
# Load necessary libraries
library(tidyverse)

# Load the dataset
df <- read.csv("Individual-level basic variable.csv")

# Ensure necessary columns are numeric
df$Partners_count <- as.numeric(df$Partners_count)
df$CaregiverResidencyDuration <- as.numeric(df$CaregiverResidencyDuration)
df$Age <- as.numeric(df$Age)</pre>
```

```
# Create a new variable for CaregiverResidencyDuration categories with specified order
df <- df %>%
  mutate(ResidencyCategory = factor(ifelse(CaregiverResidencyDuration < 120, 'Below 120', '120 or Above
                                    levels = c('Below 120', '120 or Above')))
# Create a new variable for puberty status with specified order
df <- df %>%
 mutate(Puberty_Status = factor(ifelse(Age < 12, 'Pre-pubescent', 'Post-pubescent'),</pre>
                                 levels = c('Pre-pubescent', 'Post-pubescent')))
# Group by both ResidencyCategory and Puberty_Status, and calculate the mean of Partners_count
residency_puberty_mean <- df %>%
  group_by(ResidencyCategory, Puberty_Status) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
## 'summarise()' has grouped output by 'ResidencyCategory'. You can override using
## the '.groups' argument.
# Create the bar plot with desired order
ggplot(residency_puberty_mean, aes(x = ResidencyCategory, y = mean_partners, fill = Puberty_Status)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(title = "Mean Partners Count by Residency Duration and Puberty Status",
       x = "Caregiver Residency Duration",
      y = "Mean Partners Count") +
  theme_minimal() +
  scale_fill_manual(values = c("Pre-pubescent" = "lightblue", "Post-pubescent" = "lightgreen"))
```

Mean Partners Count by Residency Duration and Puberty Status

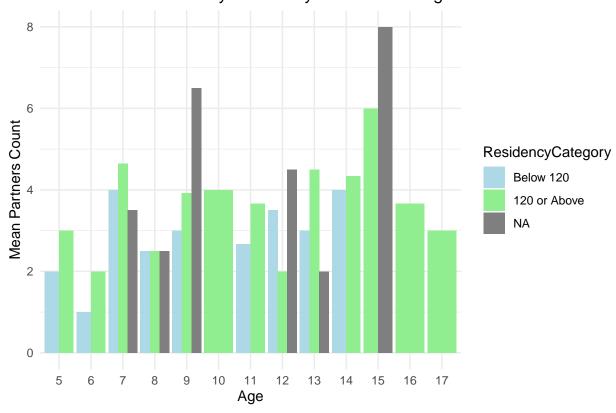


```
# Load necessary libraries
library(tidyverse)
# Load the dataset
df <- read.csv("Individual-level basic variable.csv")</pre>
# Ensure necessary columns are numeric
df$Partners_count <- as.numeric(df$Partners_count)</pre>
df$CaregiverResidencyDuration <- as.numeric(df$CaregiverResidencyDuration)</pre>
df$Age <- as.numeric(df$Age)</pre>
# Create a new variable for CaregiverResidencyDuration categories with specified order
df <- df %>%
 mutate(ResidencyCategory = factor(ifelse(CaregiverResidencyDuration < 120, 'Below 120', '120 or Above</pre>
                                     levels = c('Below 120', '120 or Above')))
# Group by both ResidencyCategory and Age, and calculate the mean of Partners_count
residency_age_mean <- df %>%
  group_by(ResidencyCategory, Age) %>%
  summarise(mean_partners = mean(Partners_count, na.rm = TRUE))
```

'summarise()' has grouped output by 'ResidencyCategory'. You can override using

the '.groups' argument.

Mean Partners Count by Residency Duration and Age



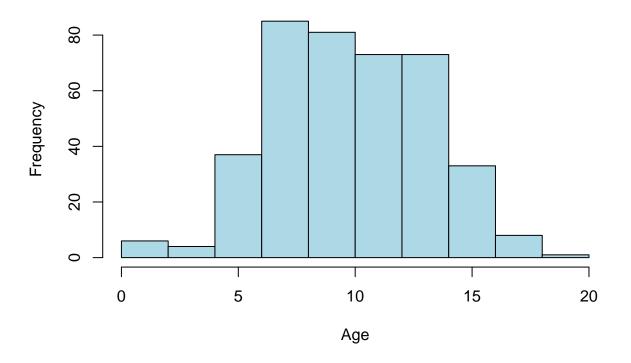
Partner-level

```
partner_data <- read.csv("Partner-level basic variables.csv")
partner_data$Partner_age <- as.numeric(partner_data$Partner_age)</pre>
```

Warning: NAs introduced by coercion

```
# Distribution of partner ages
hist(partner_data$Partner_age, main="Distribution of Playmates' Ages", xlab="Age", col="lightblue")
```

Distribution of Playmates' Ages

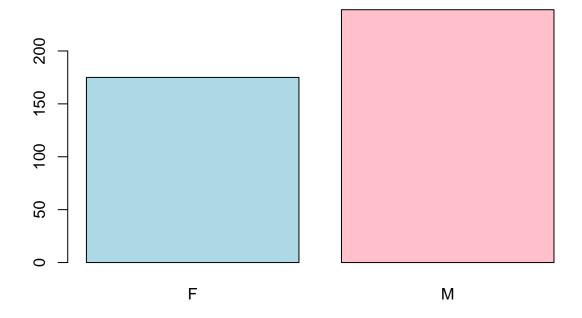


```
# Distribution of partner sex
table(partner_data$Partner_sex)
```

```
##
## F M
## 175 239
```

barplot(table(partner_data\$Partner_sex), main="Distribution of Playmates by Sex", col=c("lightblue", "p

Distribution of Playmates by Sex



```
# Load necessary packages
library(dplyr)
# Load the individual and partner-level datasets
individual_data <- read.csv("Individual-level basic variable.csv")</pre>
partner_data <- read.csv("Partner-level basic variables.csv")</pre>
# Merge the individual and partner datasets on the 'ID' column
merged_data <- partner_data %>%
 left_join(individual_data %>% select(ID, Age), by = "ID")
# Save the merged dataset to a new CSV file
write.csv(merged_data, "Partner-level_with_individual_age.csv", row.names = FALSE)
# View the merged data (optional)
head(merged_data)
```

```
Date Partner_order Adult Partner_age Partner_sex
## 1 577
           12
                M 11-Jan
                                     1
                                                      13
## 2 577
               M 11-Jan
                                     2
                                                      12
                                                                   Μ
           12
                                          No
## 3 577
           12 M 11-Jan
                                     3
                                                      11
                                                                   Μ
## 4 577
           12 M 11-Jan
                                     4
                                                                   М
                                          No
                                                      11
## 5 577
           12
                M 11-Jan
                                     5
                                          No
                                                      14
                                                                   F
                                                       7
## 6 895
                F 10-Jan
                                     1
                                          No
              How Related Relationship AtCamp Preferred Unpreferred Post_camp
## 1
           School
                             Unrelated
                                                    Yes
                       No
                                           No
                                                                 No
                                                                           No
```

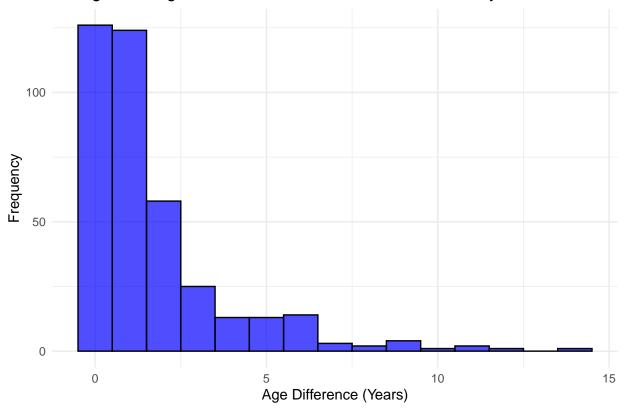
##

ID Age.x Sex

```
## 2
            School
                        No
                              Unrelated
                                             No
                                                       No
                                                                   No
                                                                              No
## 3 ThroughFamily
                        No
                              Unrelated
                                             No
                                                       Nο
                                                                   No
                                                                              Nο
         Neighbors
                              Unrelated
                                                                              No
## 4
                        No
                                             No
                                                      Yes
                                                                   No
## 5
            School
                              Unrelated
                                                       No
                                                                              No
                        No
                                             No
                                                                   Nο
## 6
            School
                        No
                              Unrelated
                                             No
                                                      Yes
                                                                   No
                                                                              No
## Age.y
## 1
       12
## 2
        12
## 3
        12
## 4
        12
## 5
        12
## 6
        7
# Load necessary packages
library(dplyr)
library(ggplot2)
# Load the merged dataset (which includes the age of individuals)
merged_data <- read.csv("Partner-level_with_individual_age.csv")</pre>
# Convert Partner_age and Age.y columns to numeric, forcing non-numeric values to NA
merged_data$Partner_age <- as.numeric(merged_data$Partner_age)</pre>
```

Warning: NAs introduced by coercion





```
write.csv(cleaned_data, "Cleaned_Partner_level_with_age_difference.csv", row.names = FALSE)
```

```
age_difference_summary <- cleaned_data %>%
  group_by(Age.y) %>%
  summarise(Mean_Age_Difference = mean(Age_Difference, na.rm = TRUE))

# Print the summary
print(age_difference_summary)
```

```
## # A tibble: 13 x 2
##
      Age.y Mean_Age_Difference
      <dbl>
                            <dbl>
##
##
    1
          5
                            2.8
    2
                            1.6
##
           6
##
    3
          7
                            2.17
##
    4
                            1.57
          8
                            1.14
##
    5
          9
##
    6
         10
                            1.81
##
    7
                            1.37
         11
                            1.54
##
    8
         12
                            0.969
##
    9
         13
## 10
         14
                            2.10
## 11
         15
                            0.938
## 12
         16
                            2.36
                           10.7
## 13
         17
```

Mean Age Difference by Age of Interviewed Participant

