

GitHub

Elham

2025-10-30

```
#install libraries
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.2      v readr      2.1.4
```

```
## v forcats    1.0.0      v stringr    1.5.0
```

```
## v ggplot2    3.5.2      v tibble     3.2.1
```

```
## v lubridate  1.9.2      v tidyr      1.3.0
```

```
## 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 (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(ggplot2)
```

```
library(dplyr)
```

```
library(likert)
```

```
## Loading required package: xtable
```

```
##
```

```
## Attaching package: 'likert'
```

```
##
```

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

```
##
```

```
##      recode
```

```
library(stats)
```

```
library(lavaan)
```

```
## This is lavaan 0.6-16
```

```
## lavaan is FREE software! Please report any bugs.
```

```
library(psych)
```

```
##
```

```
## Attaching package: 'psych'
```

```
##
```

```
## The following object is masked from 'package:lavaan':
```

```
##
```

```
##      cor2cov
##
## The following objects are masked from 'package:ggplot2':
##
##      %+%, alpha
```

```
library(Hmisc)
```

```
##
## Attaching package: 'Hmisc'
##
## The following object is masked from 'package:psych':
##
##      describe
##
## The following objects are masked from 'package:xtable':
##
##      label, label<-
##
## The following objects are masked from 'package:dplyr':
##
##      src, summarize
##
## The following objects are masked from 'package:base':
##
##      format.pval, units
```

```
library(broom)
library(purrr)
library(sjlabelled)
```

```
##
## Attaching package: 'sjlabelled'
##
## The following object is masked from 'package:forcats':
##
##      as_factor
##
## The following object is masked from 'package:dplyr':
##
##      as_label
##
## The following object is masked from 'package:ggplot2':
##
##      as_label
```

```
#library(kableExtra)
library(gtools)
```

```
##
## Attaching package: 'gtools'
##
```

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

```
library(stargazer)
```

```
##  
## Please cite as:  
##  
## Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.  
## R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
```

```
library(sjPlot)  
library(REdaS)
```

```
## Loading required package: grid
```

```
library(GPArotation)
```

```
##  
## Attaching package: 'GPArotation'  
##  
## The following objects are masked from 'package:psych':  
##  
##      equamax, varimin
```

```
library(effectsize)
```

```
##  
## Attaching package: 'effectsize'  
##  
## The following object is masked from 'package:psych':  
##  
##      phi  
##  
## The following object is masked from 'package:xtable':  
##  
##      display
```

```
library(lsr)  
library(stats)  
library(sjPlot)  
library(mediation)
```

```
## Warning: package 'mediation' was built under R version 4.3.3
```

```
## Loading required package: MASS  
##  
## Attaching package: 'MASS'  
##
```

```
## The following object is masked from 'package:dplyr':
##
##   select
##
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
##
## The following objects are masked from 'package:tidyr':
##
##   expand, pack, unpack
##
## Loading required package: mvtnorm
##
## Attaching package: 'mvtnorm'
##
## The following object is masked from 'package:effectsize':
##
##   standardize
##
## Loading required package: sandwich

## Warning: package 'sandwich' was built under R version 4.3.3

## mediation: Causal Mediation Analysis
## Version: 4.5.0
##
##
## Attaching package: 'mediation'
##
## The following object is masked from 'package:psych':
##
##   mediate
```

```
#upload the data
load("JAR_Social_Invitees_clean.RData")
```

```
# Report Sample Demographics
```

```
#number of participants based on gender, race, education
JAR_Social_Invitees %>%
  dplyr::select(Gender_fct,
                Race_fct,
                Education_fct,
                Education_college) %>%
  map(table)
```

```
## $Gender_fct
##
##               Female               Male
##               2               3
## Non-binary / Genderqueer / Gender fluid   Prefer not to say
##               3               2
```

```

##
## $Race_fct
##
##      Asian or Asian American  Black or African American
##                2                1
##      Latino/a/x or Hispanic      Middle Eastern
##                1                1
##      Prefer not to say White or European American
##                1                4
##
## $Education_fct
##
##                                     Less than high school
##                                     1
##                                     High school degree or equivalent
##                                     0
## Some college (if currently an undergraduate student, select this option)
##                                     1
##                                     Associate (2 year) degree
##                                     1
##                                     Bachelor's (4 year) degree
##                                     2
##                                     Some graduate school
##                                     0
##                                     Master's degree
##                                     3
##                                     Professional degree (e.g., JD, MD)
##                                     0
##                                     Doctorate (PhD)
##                                     2
##
## $Education_college
##
## 0 1
## 1 9

```

```

#number of participants' gender based on each condition
Female_Male_Count <- JAR_Social_Invitees %>%
  group_by(Condition) %>%
  summarise(
    Female_Count = sum(Gender_fct == "Female", na.rm = TRUE),
    Male_Count = sum(Gender_fct == "Male", na.rm = TRUE))

#number of participants' race based on each condition
Race_Count <- JAR_Social_Invitees %>%
  group_by(Condition, Race) %>%
  summarise(
    count = n(), .groups = 'drop')

#number of participants' education based on each condition
Education_Count <- JAR_Social_Invitees %>%
  group_by(Condition, Education) %>%
  summarise(
    count = n(), .groups = 'drop')

```

```
#age
summary(JAR_Social_Invitees$Age)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      22.00  26.00   43.00   38.70  48.25   55.00
```

```
sd(JAR_Social_Invitees$Age)
```

```
## [1] 12.43695
```

```
#number of participants' age based on each condition
```

```
Age_Count <- JAR_Social_Invitees %>%
  group_by(Condition, Age) %>%
  summarise(
    count = n(), .groups = 'drop')
```

```
#####Number of Participants' age in intervals based on each condition#####
```

```
# Create age intervals
```

```
JAR_Social_Invitees$age_interval <- cut(JAR_Social_Invitees$Age,
                                       breaks = seq(0, max(JAR_Social_Invitees$Age, na.rm = TRUE) + 10,
```

```
# Create a contingency table: age interval by condition
```

```
age_by_condition <- table(JAR_Social_Invitees$age_interval, JAR_Social_Invitees$Condition)
```

```
# Display the result
```

```
age_by_condition
```

```
##
##           1 2
## (0,10]  0 0
## (10,20] 0 0
## (20,30] 3 1
## (30,40] 0 0
## (40,50] 3 1
## (50,60] 0 2
```

```
# Attention
```

```
JAR_Social_Invitees %>%
  dplyr::select(Attention_AI_binary,
               Attention_HR_binary,
               Attention_loop_AI_binary,
               Attention_loop_HR_binary) %>%
  map(table)
```

```
## $Attention_AI_binary
```

```
##
## 0 1
## 6 4
##
```

```
## $Attention_HR_binary
##
## 0 1
## 6 4
##
## $Attention_loop_AI_binary
##
## 0 1
## 7 3
##
## $Attention_loop_HR_binary
##
## 0 1
## 8 2
```

Univariate Analysis

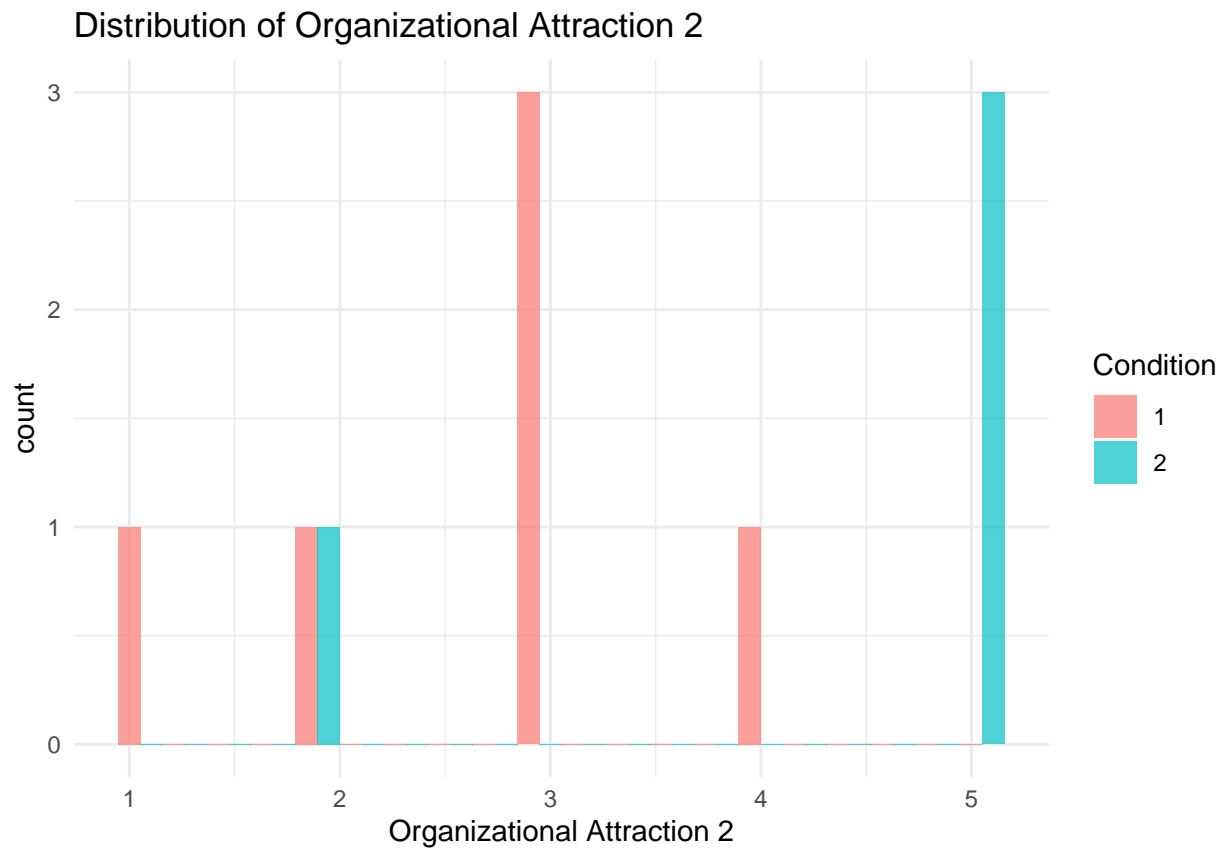
Here is where we plot each variable individually to see if it is Normally distributed

```
#ggplot to see variables are normally distributed#####
#hist
#make sure no outlier appear

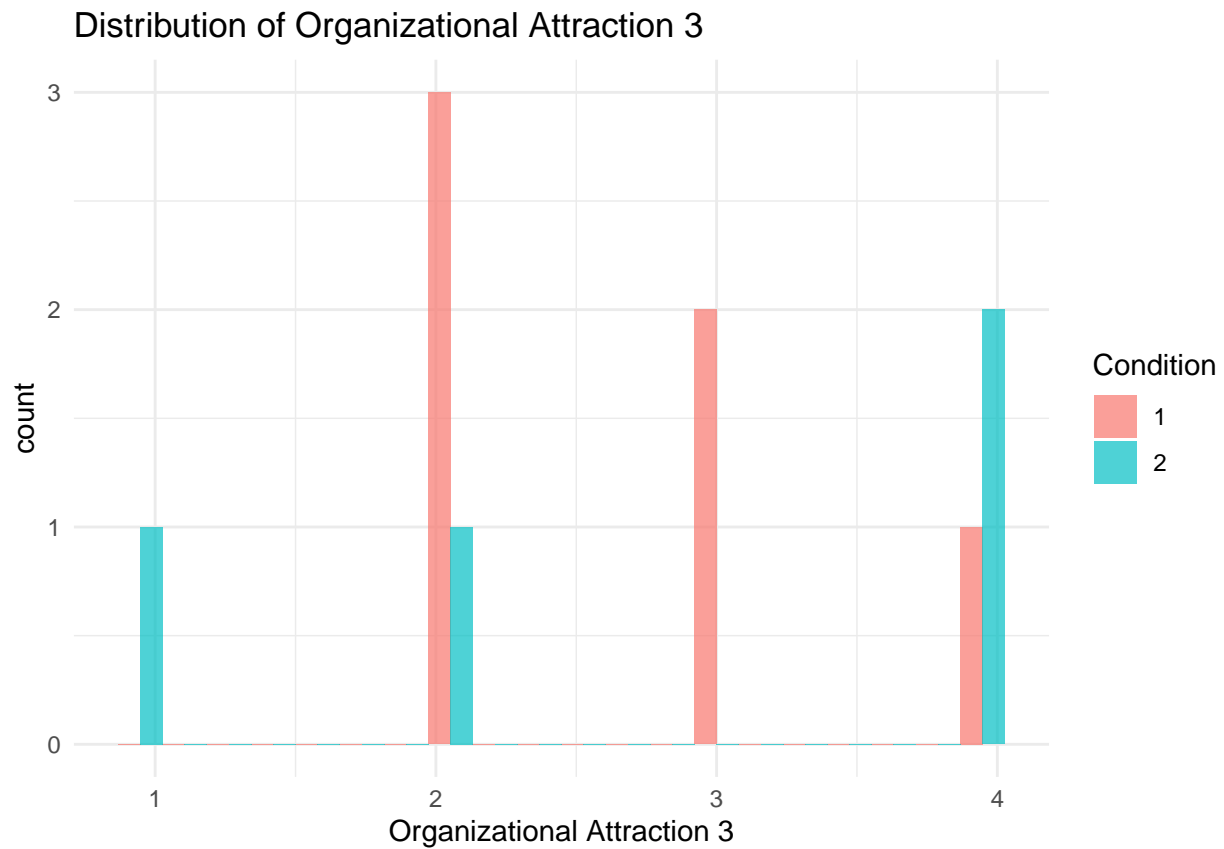
#####Organizational Attraction#####
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Organizational Attraction 1",
       x = "Organizational Attraction 1",
       fill = "Condition") +
  theme_minimal()
```



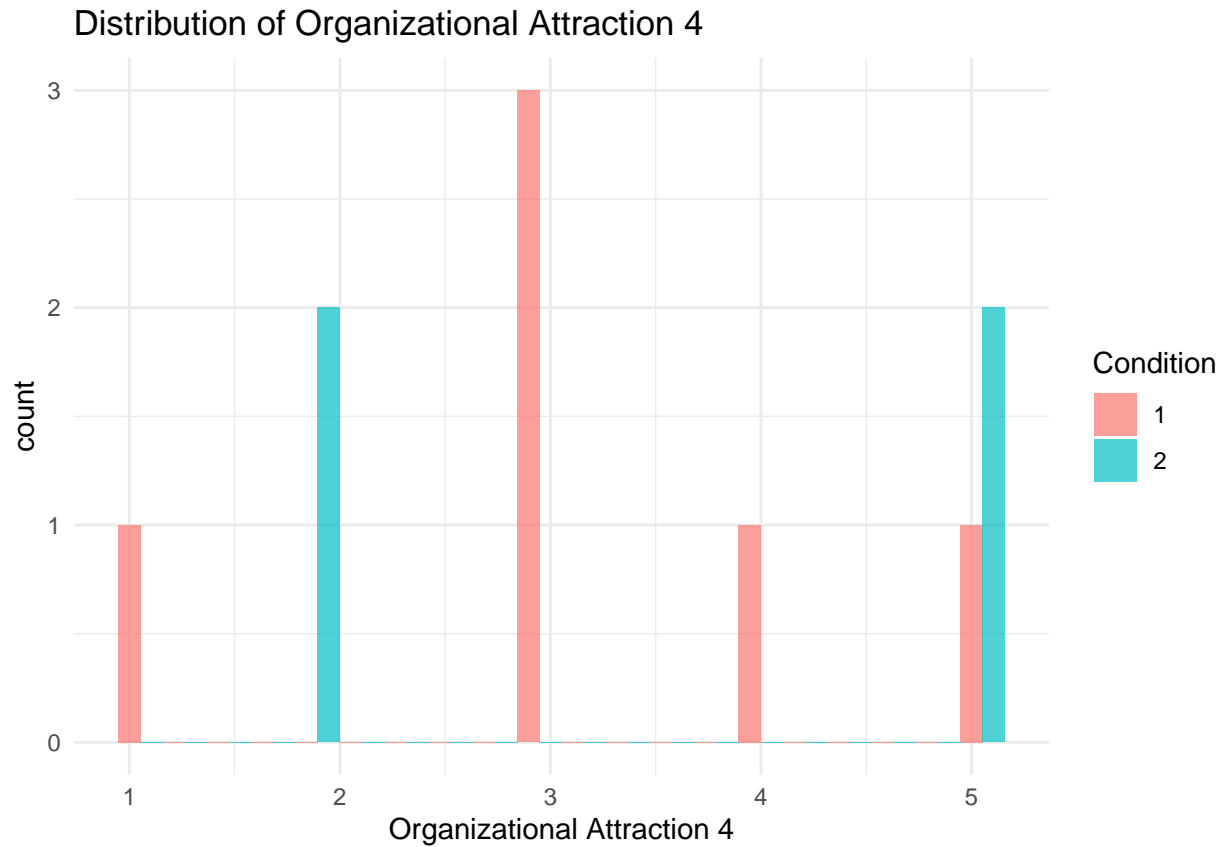
```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction 2",  
        x = "Organizational Attraction 2",  
        fill = "Condition") +  
  theme_minimal()
```

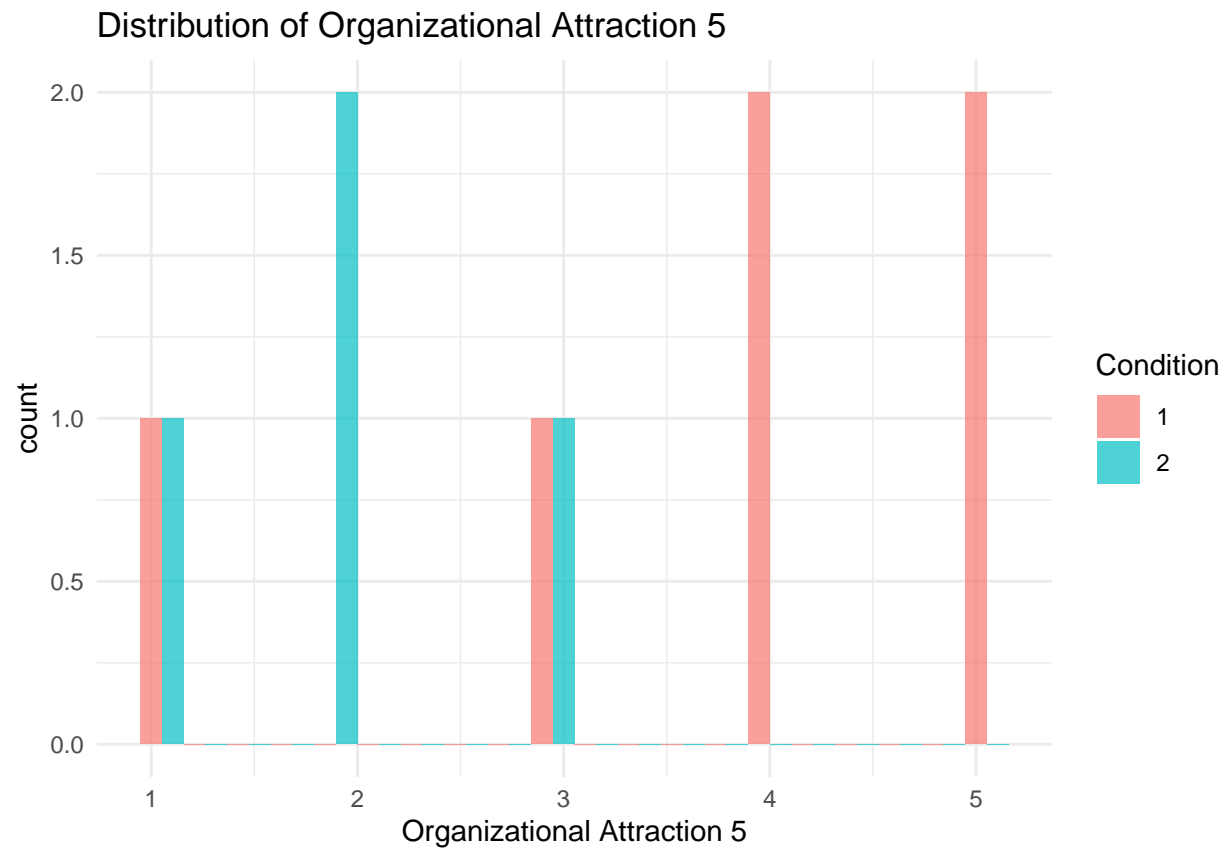
```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction 3",  
        x = "Organizational Attraction 3",  
        fill = "Condition") +  
  theme_minimal()
```



```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_4_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction 4",  
        x = "Organizational Attraction 4",  
        fill = "Condition") +  
  theme_minimal()
```



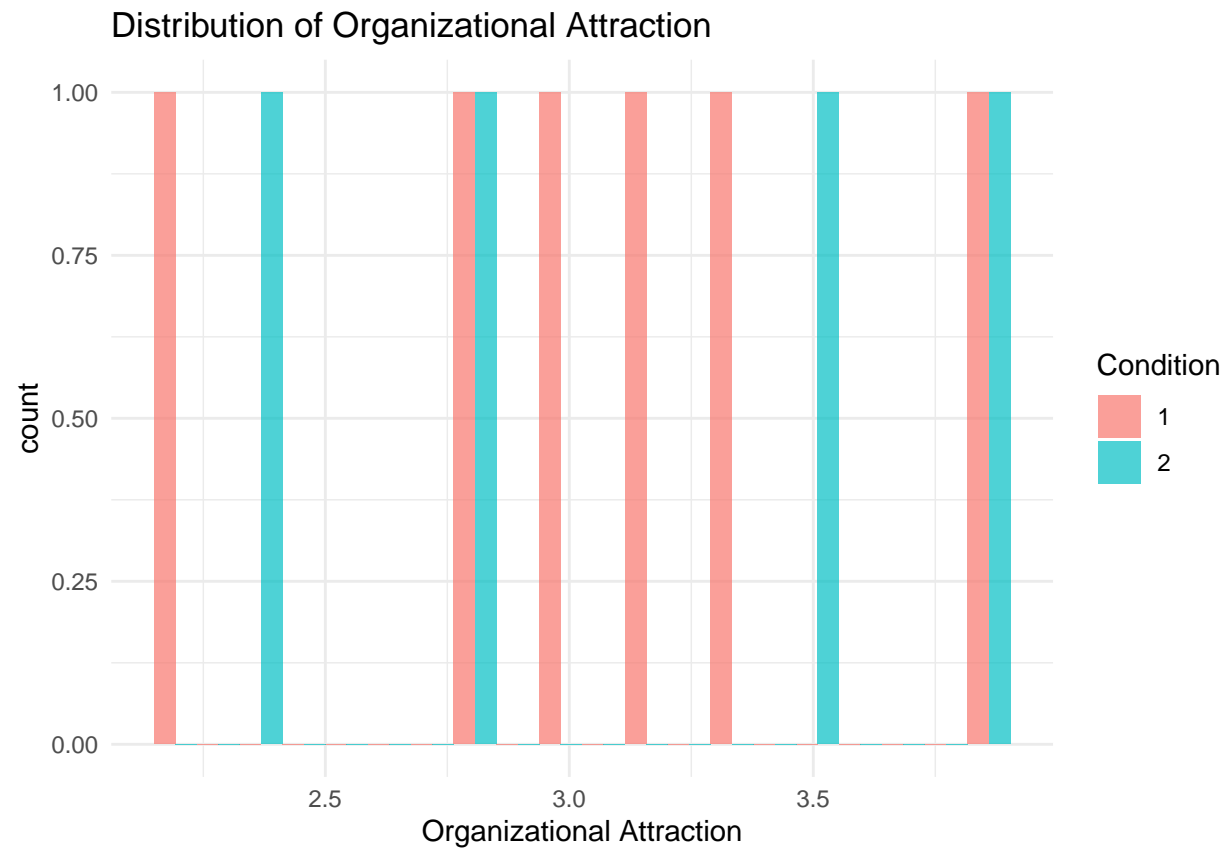
```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_5_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction 5",  
        x = "Organizational Attraction 5",  
        fill = "Condition") +  
  theme_minimal()
```



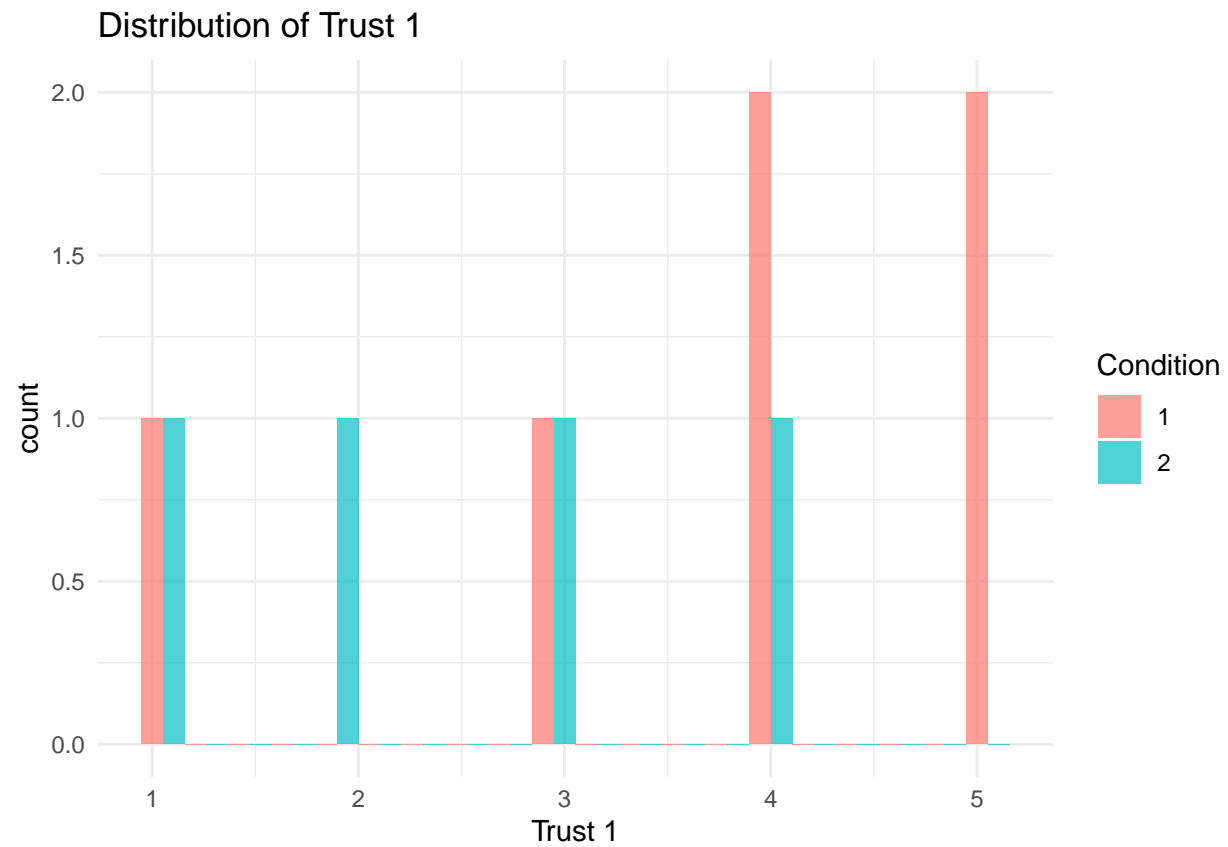
```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction_6_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction 6",  
        x = "Organizational Attraction 6",  
        fill = "Condition") +  
  theme_minimal()
```



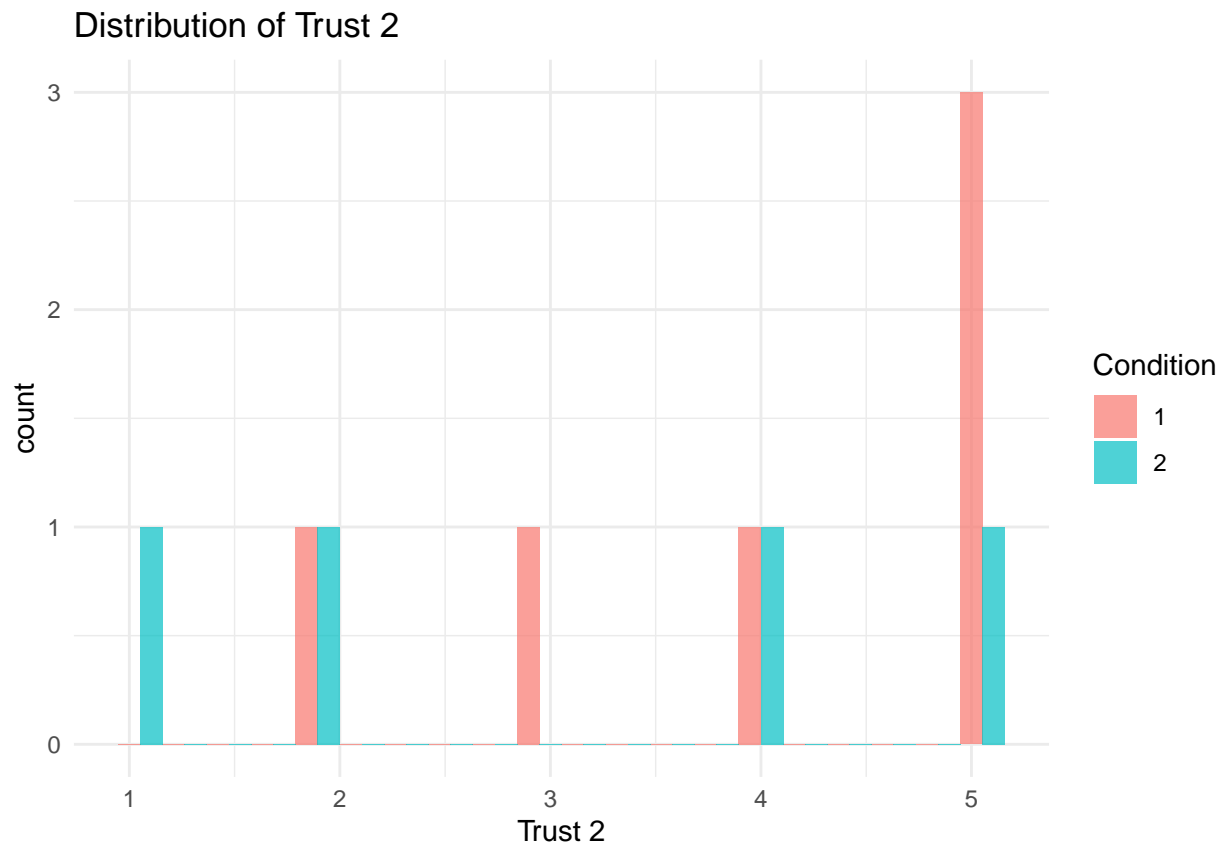
```
ggplot(JAR_Social_Invitees, aes(x = Org_Attraction, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Organizational Attraction",  
        x = "Organizational Attraction",  
        fill = "Condition") +  
  theme_minimal()
```



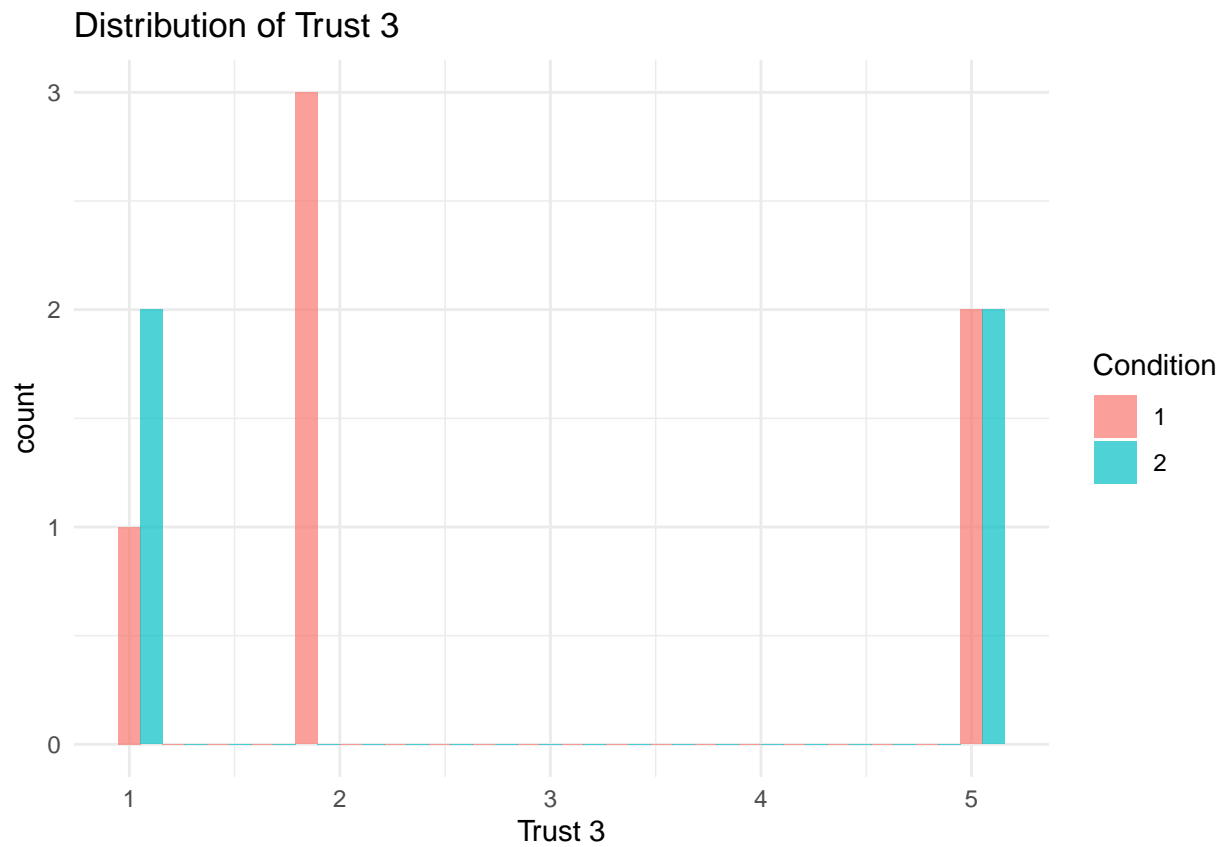
```
#####Trust#####
ggplot(JAR_Social_Invitees, aes(x = Trust_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Trust 1",
       x = "Trust 1",
       fill = "Condition") +
  theme_minimal()
```



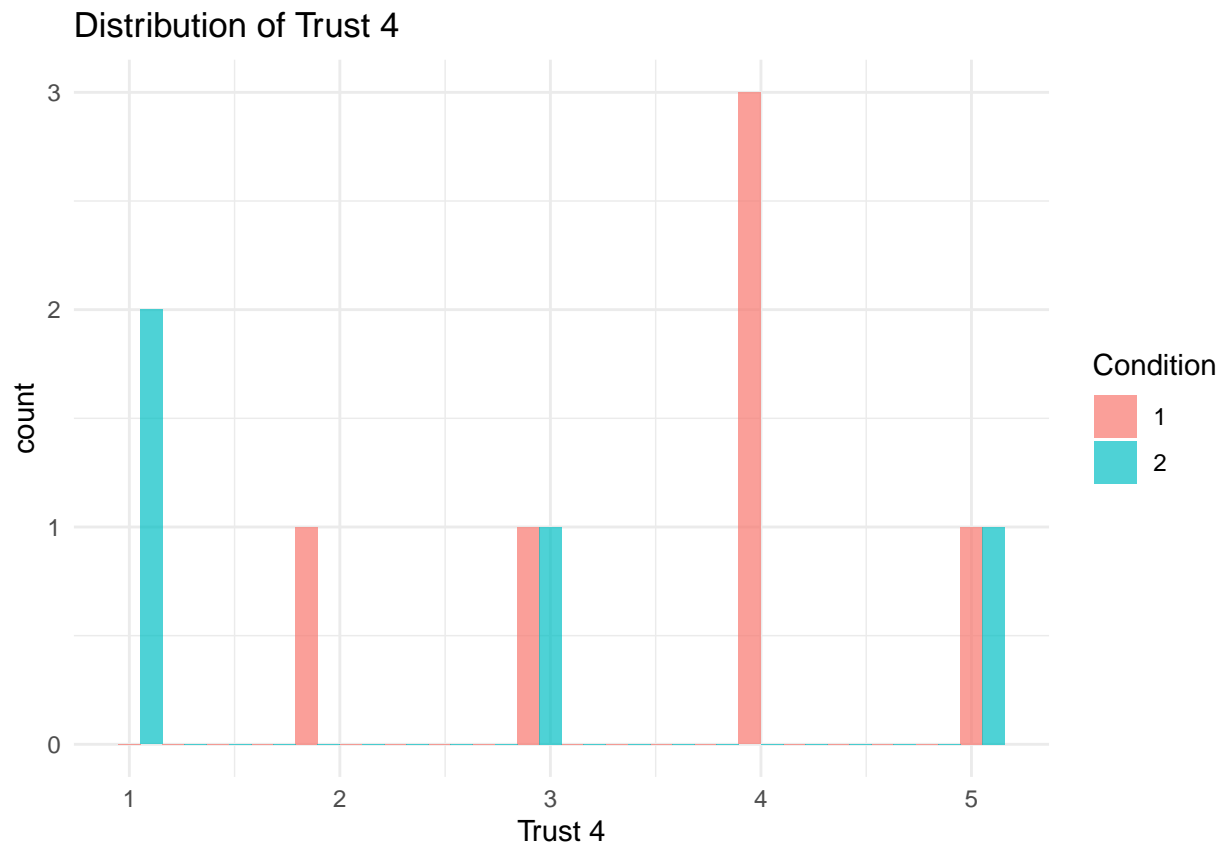
```
ggplot(JAR_Social_Invitees, aes(x = Trust_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Trust 2",  
        x = "Trust 2",  
        fill = "Condition") +  
  theme_minimal()
```



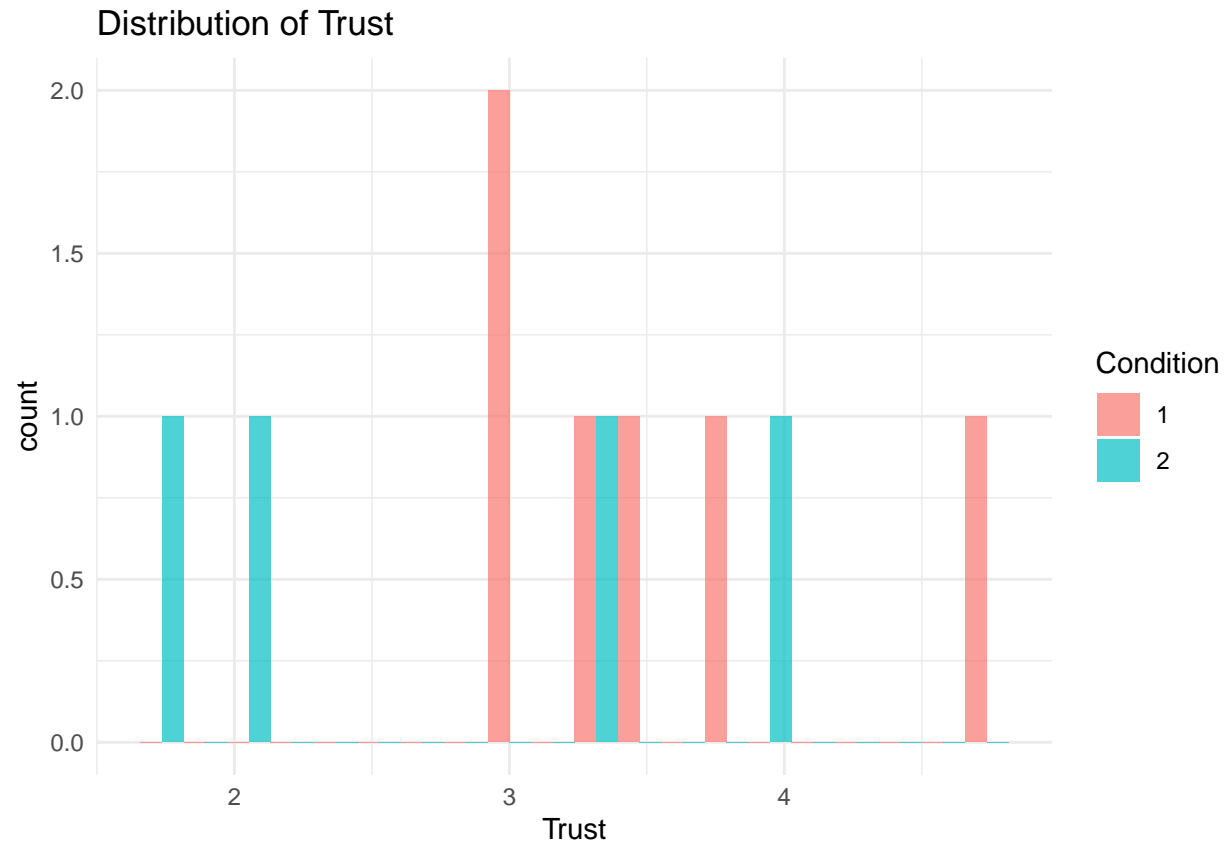
```
ggplot(JAR_Social_Invitees, aes(x = Trust_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Trust 3",  
        x = "Trust 3",  
        fill = "Condition") +  
  theme_minimal()
```

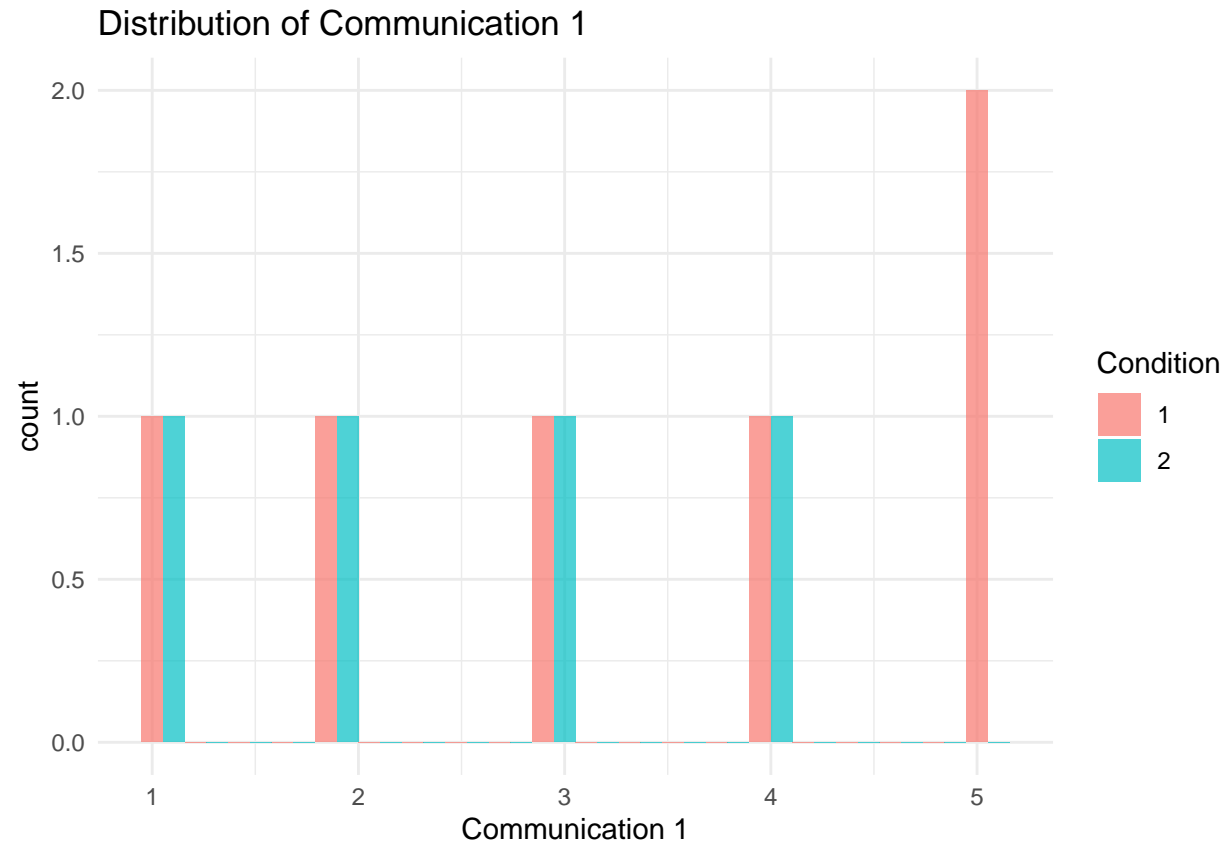
```
ggplot(JAR_Social_Invitees, aes(x = Trust_4_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Trust 4",  
        x = "Trust 4",  
        fill = "Condition") +  
  theme_minimal()
```



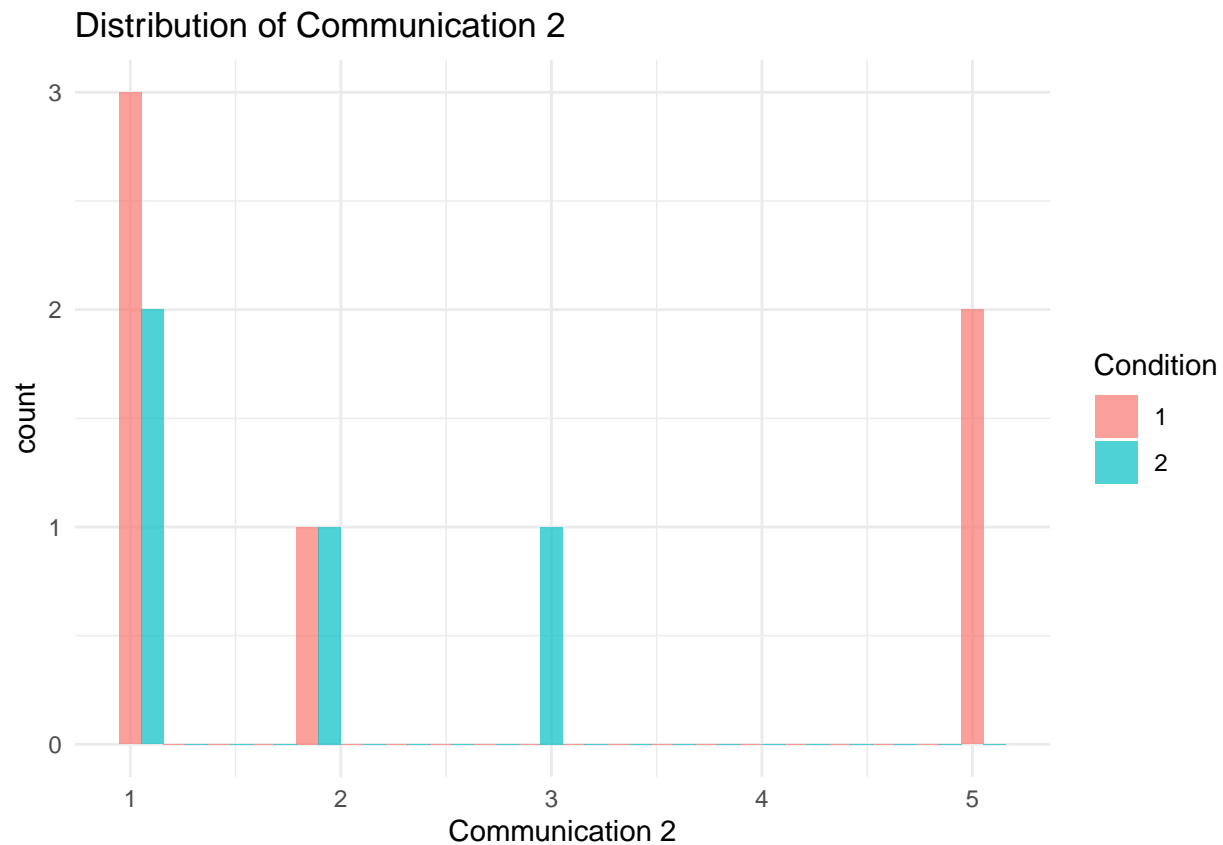
```
ggplot(JAR_Social_Invitees, aes(x = Trust, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Trust",  
        x = "Trust",  
        fill = "Condition") +  
  theme_minimal()
```



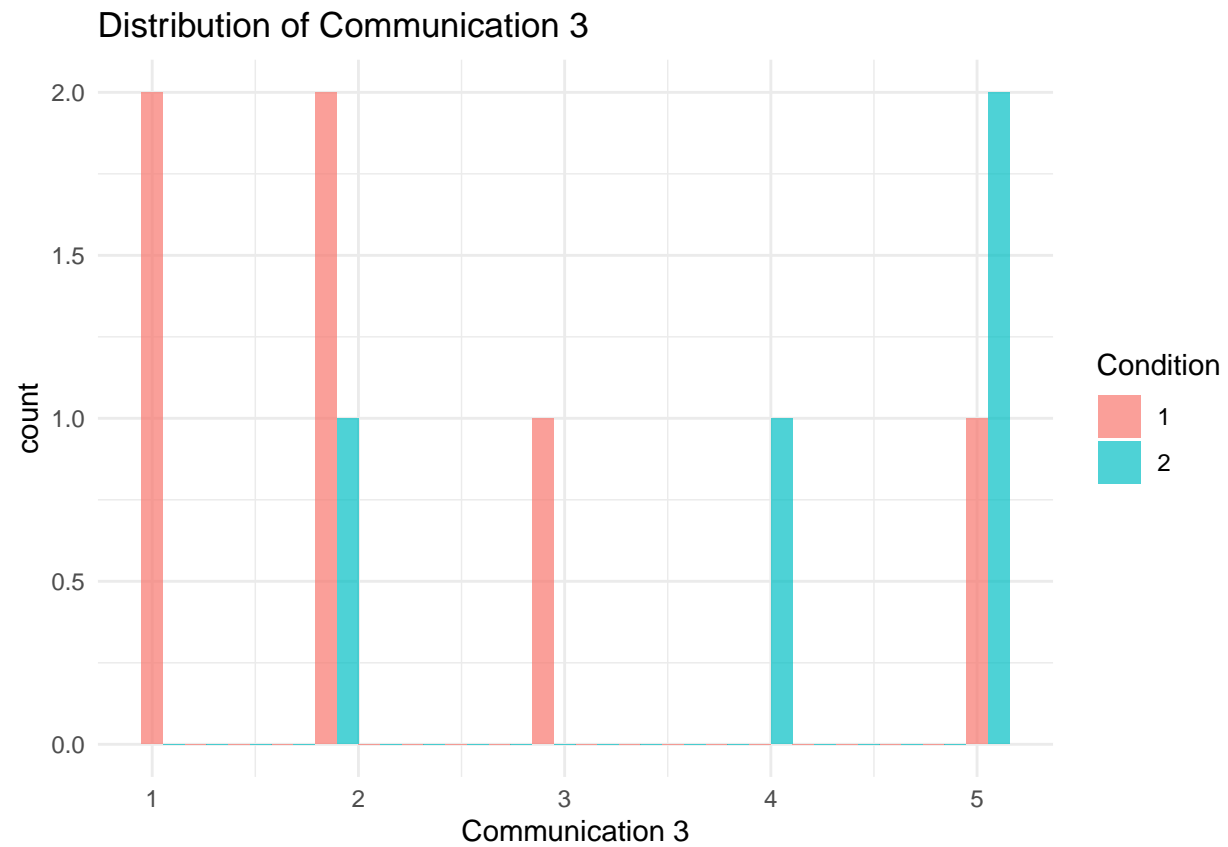
```
#####Communication#####
ggplot(JAR_Social_Invitees, aes(x = Communication_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Communication 1",
        x = "Communication 1",
        fill = "Condition") +
  theme_minimal()
```



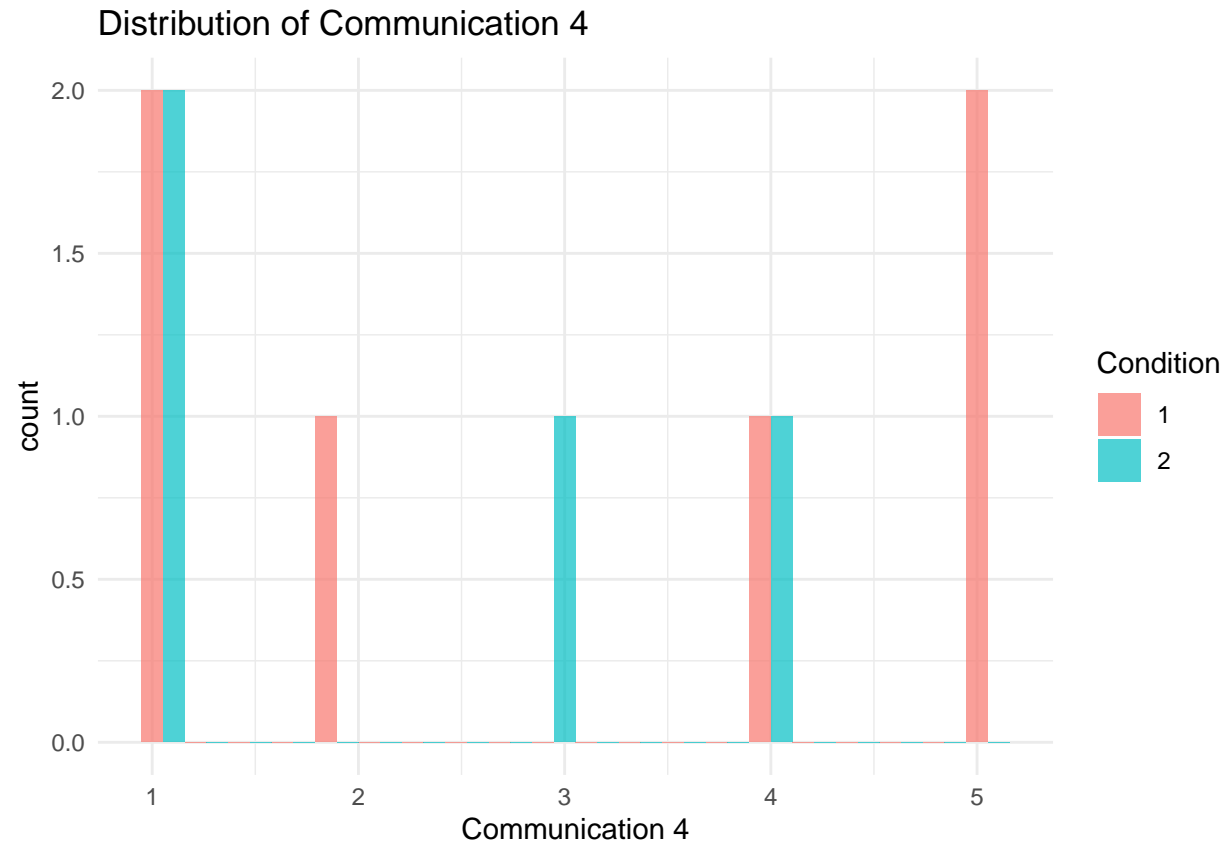
```
ggplot(JAR_Social_Invitees, aes(x = Communication_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Communication 2",  
        x = "Communication 2",  
        fill = "Condition") +  
  theme_minimal()
```



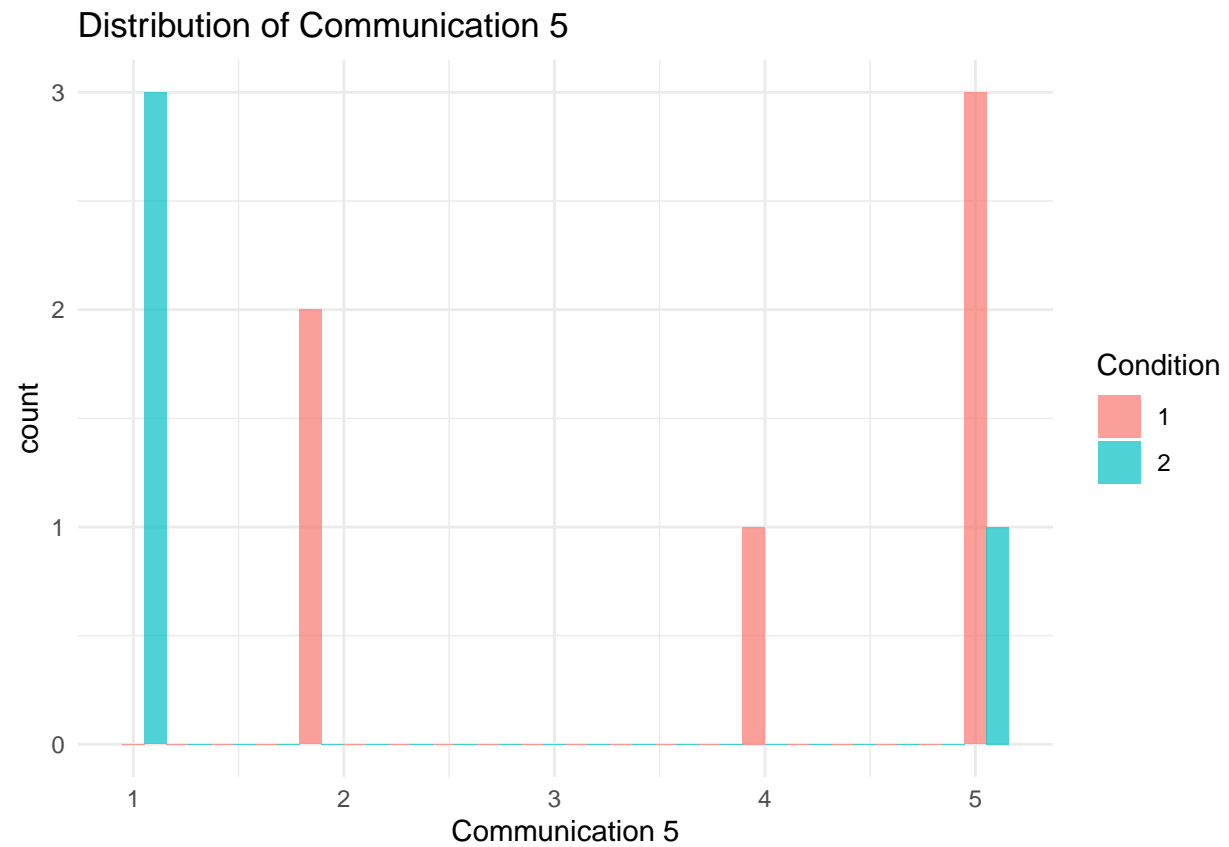
```
ggplot(JAR_Social_Invitees, aes(x = Communication_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Communication 3",  
        x = "Communication 3",  
        fill = "Condition") +  
  theme_minimal()
```



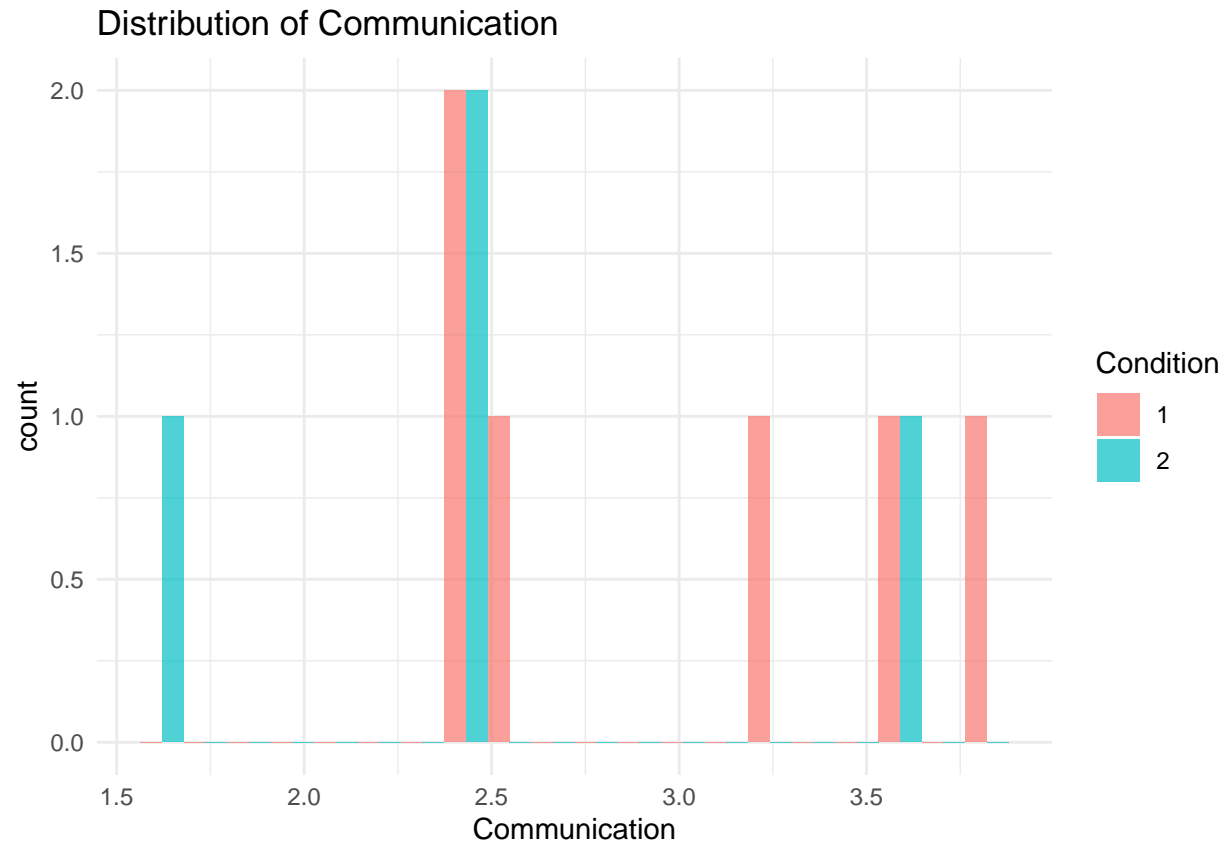
```
ggplot(JAR_Social_Invitees, aes(x = Communication_4_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Communication 4",  
        x = "Communication 4",  
        fill = "Condition") +  
  theme_minimal()
```



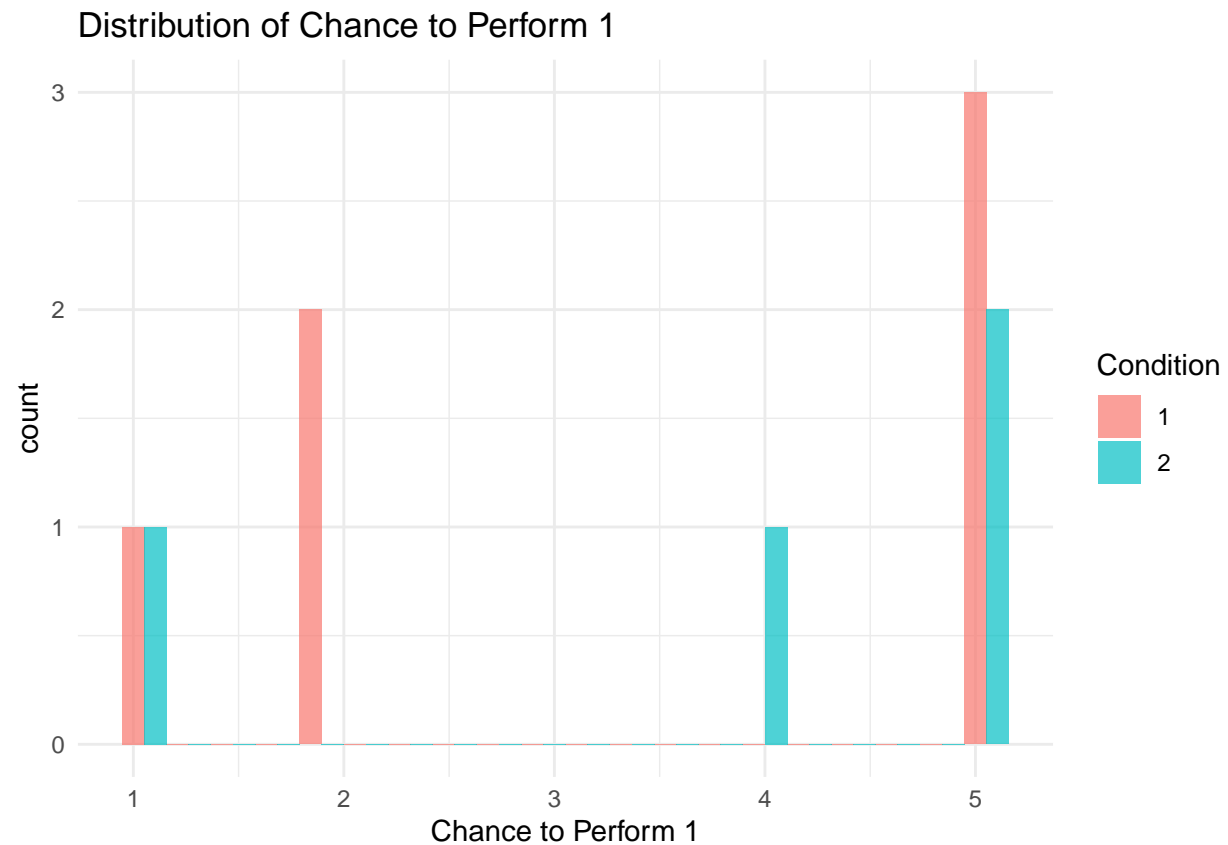
```
ggplot(JAR_Social_Invitees, aes(x = Communication_5_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Communication 5",
       x = "Communication 5",
       fill = "Condition") +
  theme_minimal()
```



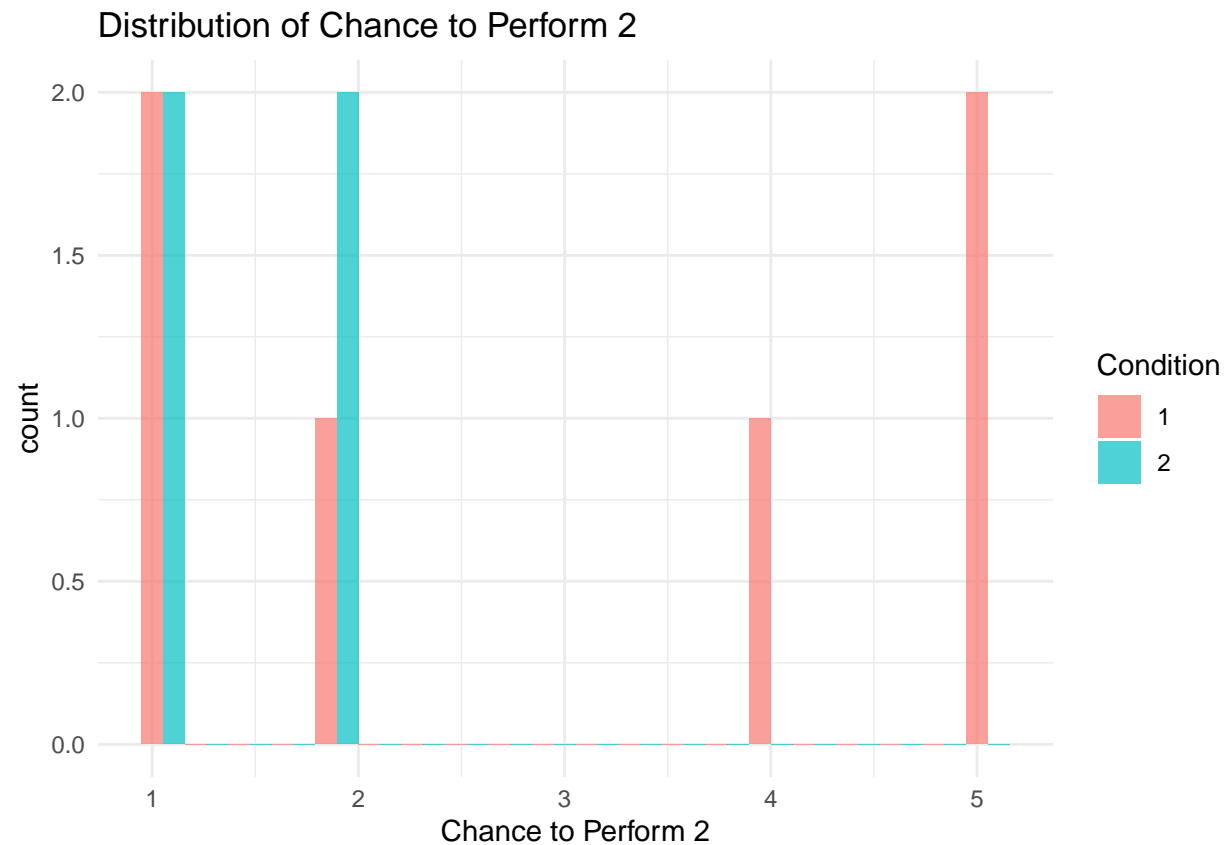
```
ggplot(JAR_Social_Invitees, aes(x = Communication, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Communication",  
        x = "Communication",  
        fill = "Condition") +  
  theme_minimal()
```

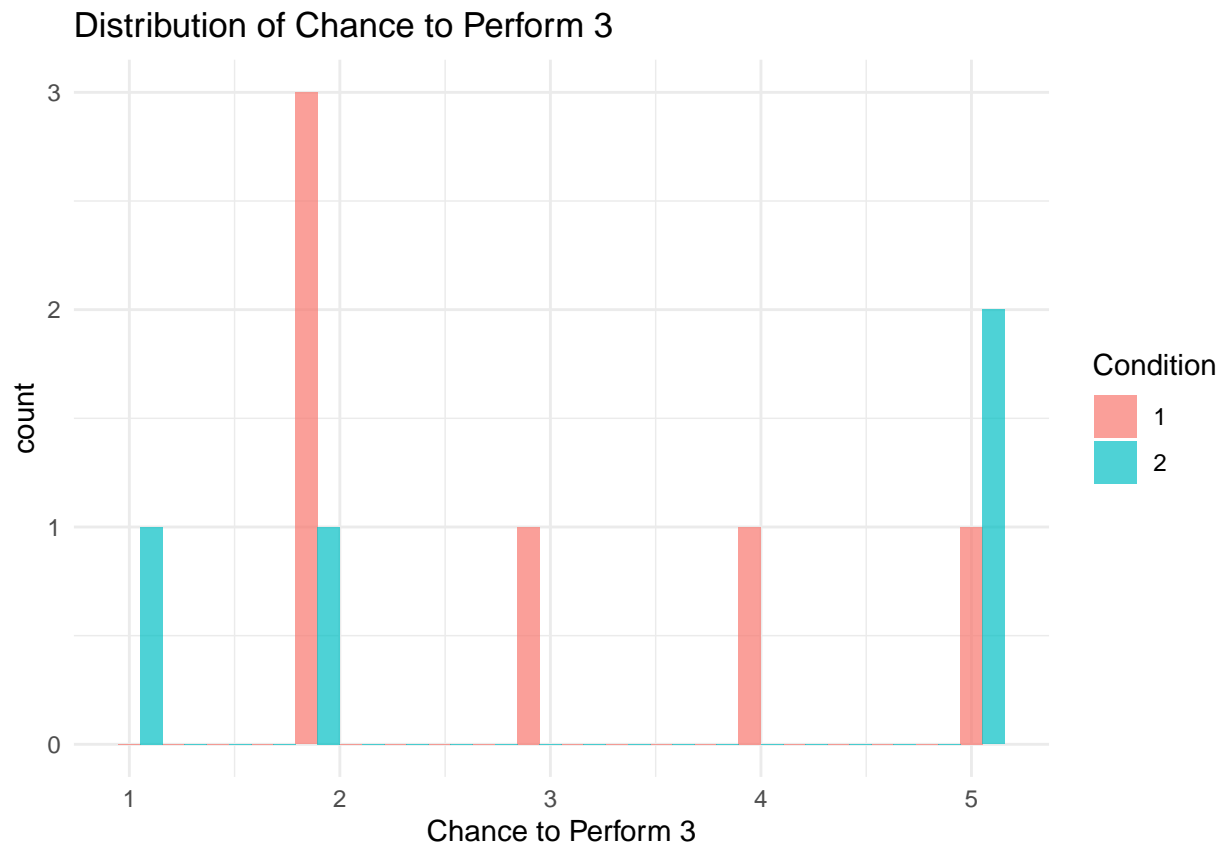
```
#####Chance to Perform#####
ggplot(JAR_Social_Invitees, aes(x = Chance_Perform_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Chance to Perform 1",
       x = "Chance to Perform 1",
       fill = "Condition") +
  theme_minimal()
```



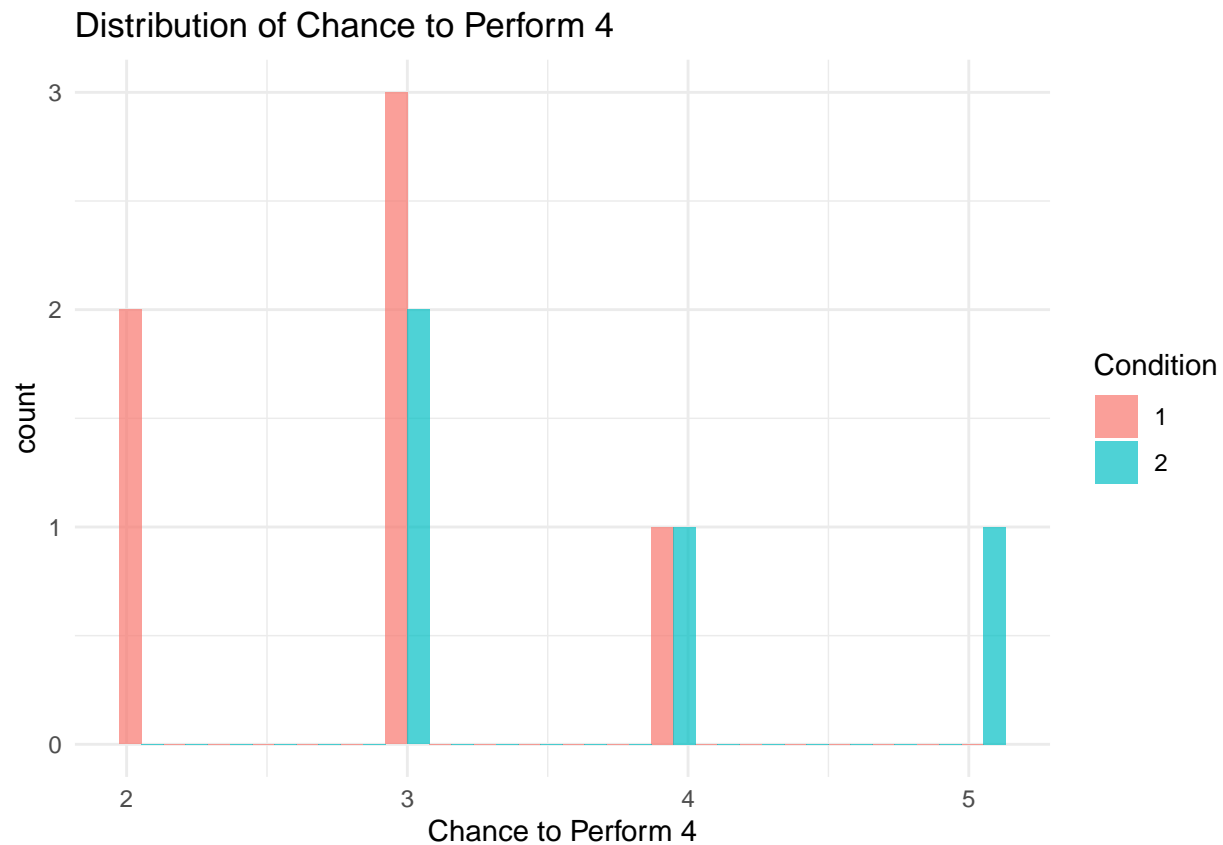
```
ggplot(JAR_Social_Invitees, aes(x = Chance_Perform_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Chance to Perform 2",  
        x = "Chance to Perform 2",  
        fill = "Condition") +  
  theme_minimal()
```



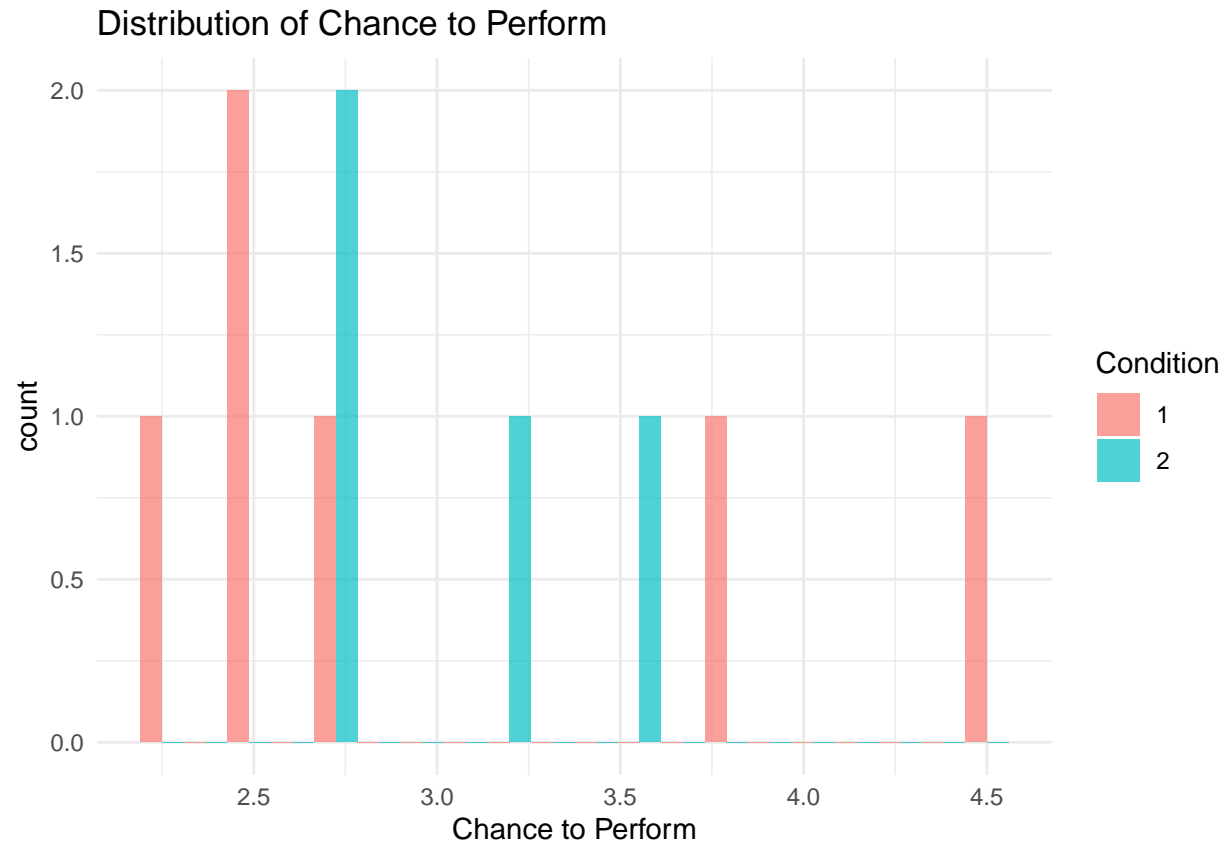
```
ggplot(JAR_Social_Invitees, aes(x = Chance_Perform_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Chance to Perform 3",  
        x = "Chance to Perform 3",  
        fill = "Condition") +  
  theme_minimal()
```



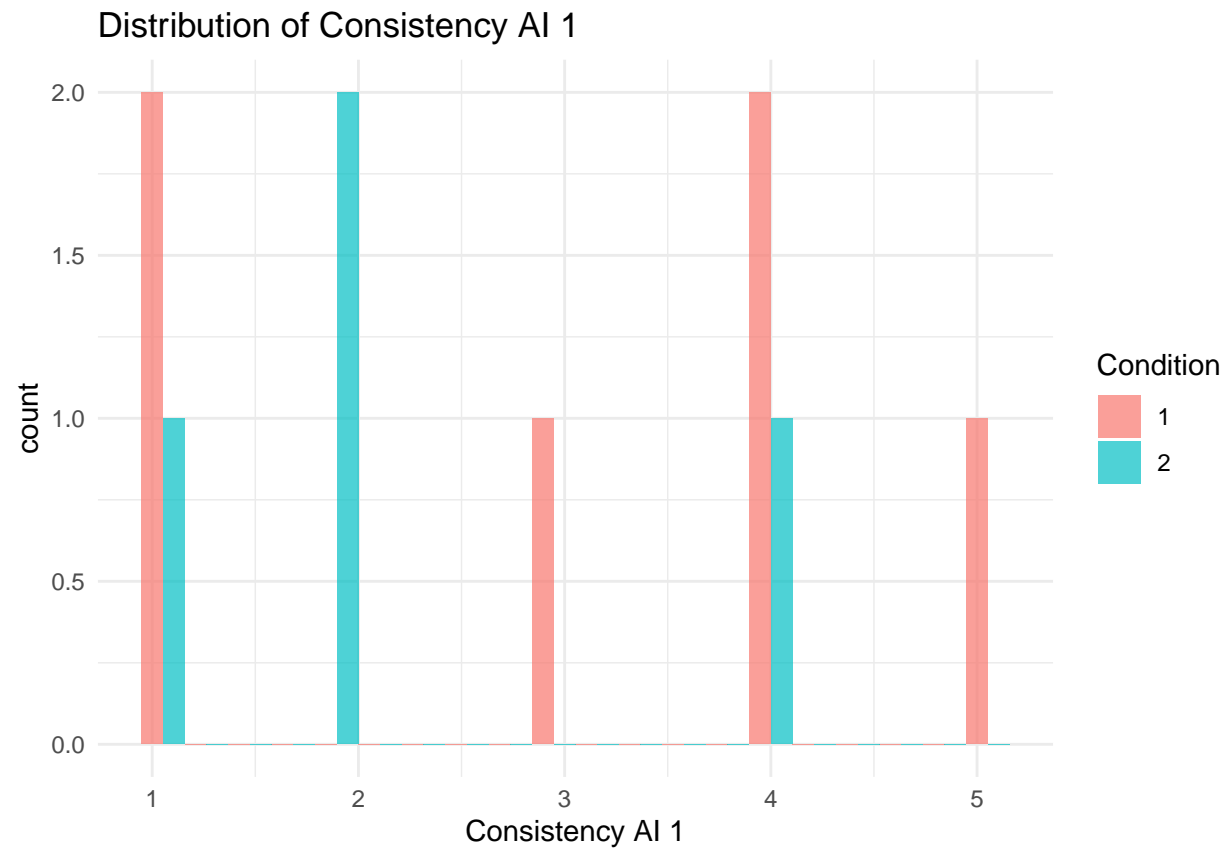
```
ggplot(JAR_Social_Invitees, aes(x = Chance_Perform_4_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Chance to Perform 4",  
        x = "Chance to Perform 4",  
        fill = "Condition") +  
  theme_minimal()
```



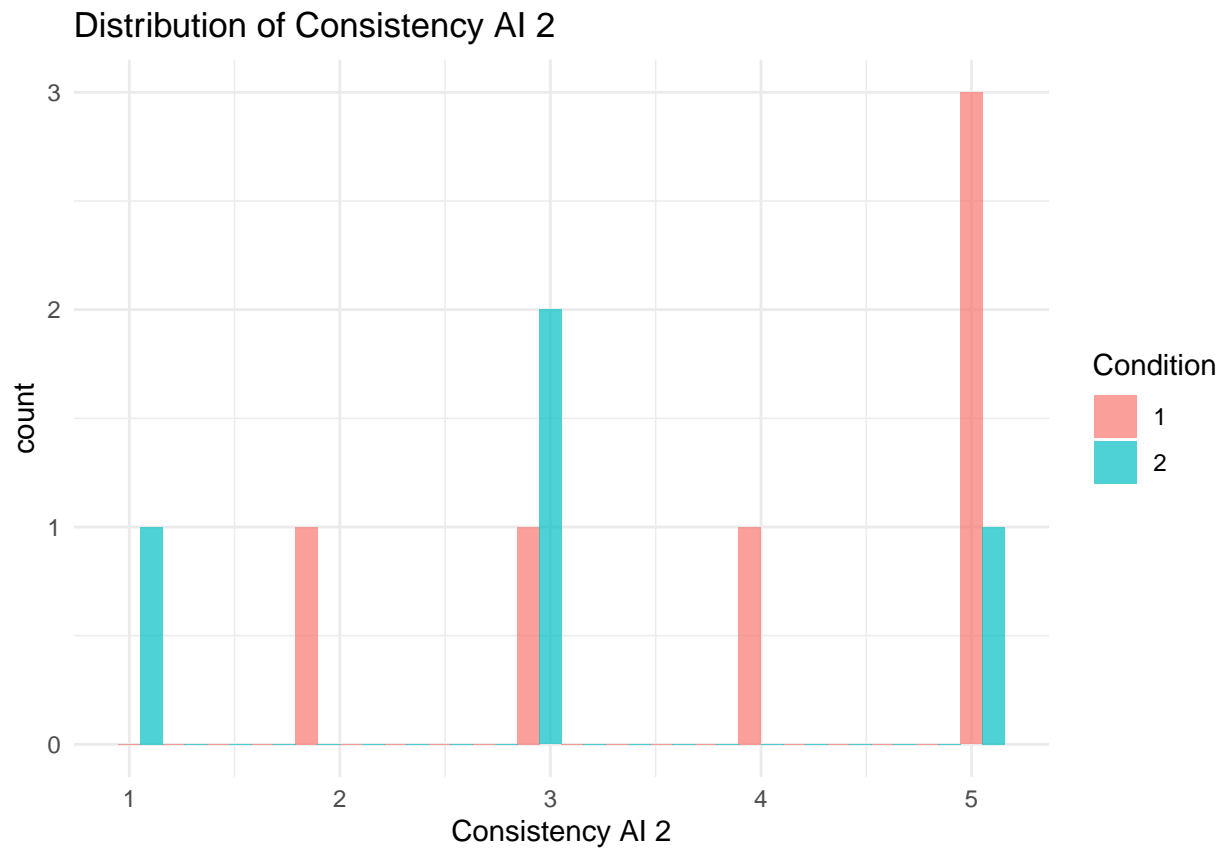
```
ggplot(JAR_Social_Invitees, aes(x = Chance_Perform, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Chance to Perform",  
        x = "Chance to Perform",  
        fill = "Condition") +  
  theme_minimal()
```



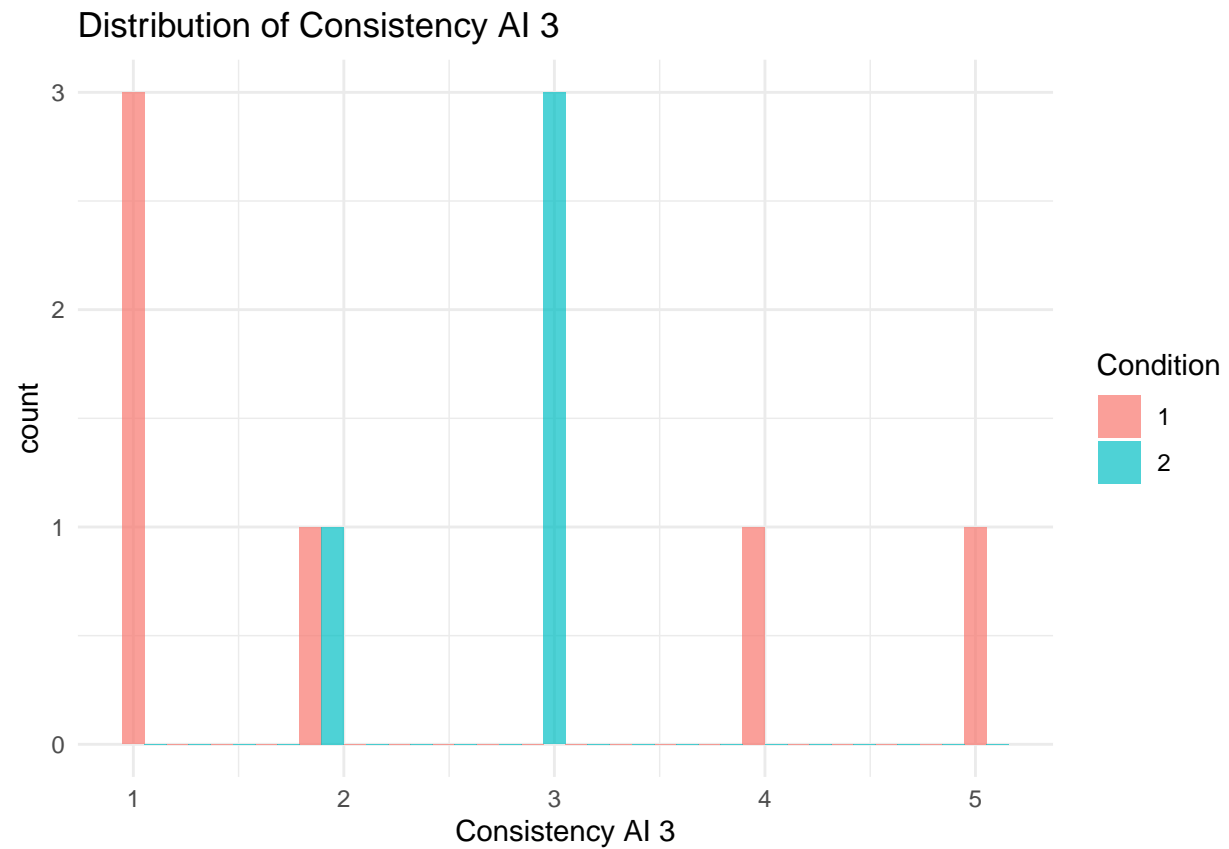
```
#####Consistency AI#####
ggplot(JAR_Social_Invitees, aes(x = Consistency_AI_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Consistency AI 1",
       x = "Consistency AI 1",
       fill = "Condition") +
  theme_minimal()
```



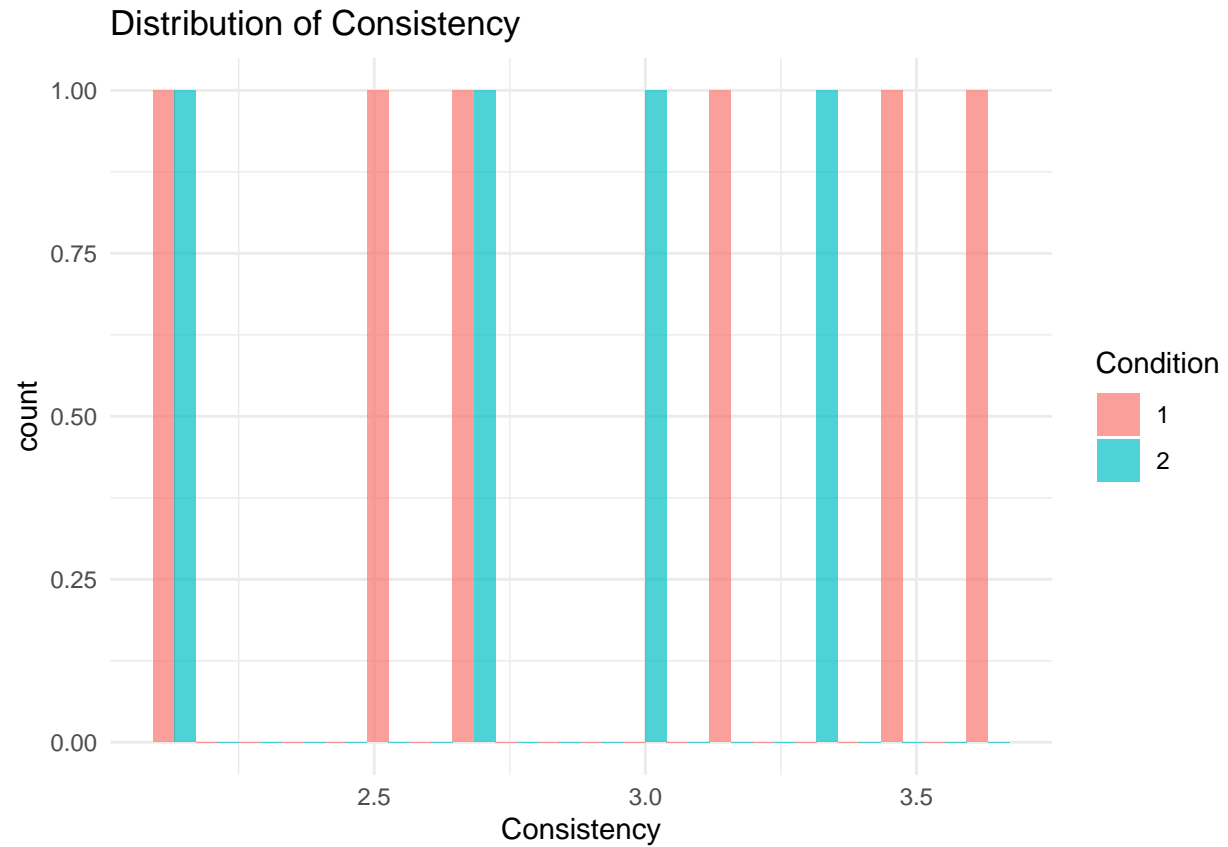
```
ggplot(JAR_Social_Invitees, aes(x = Consistency_AI_2_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Consistency AI 2",
       x = "Consistency AI 2",
       fill = "Condition") +
  theme_minimal()
```



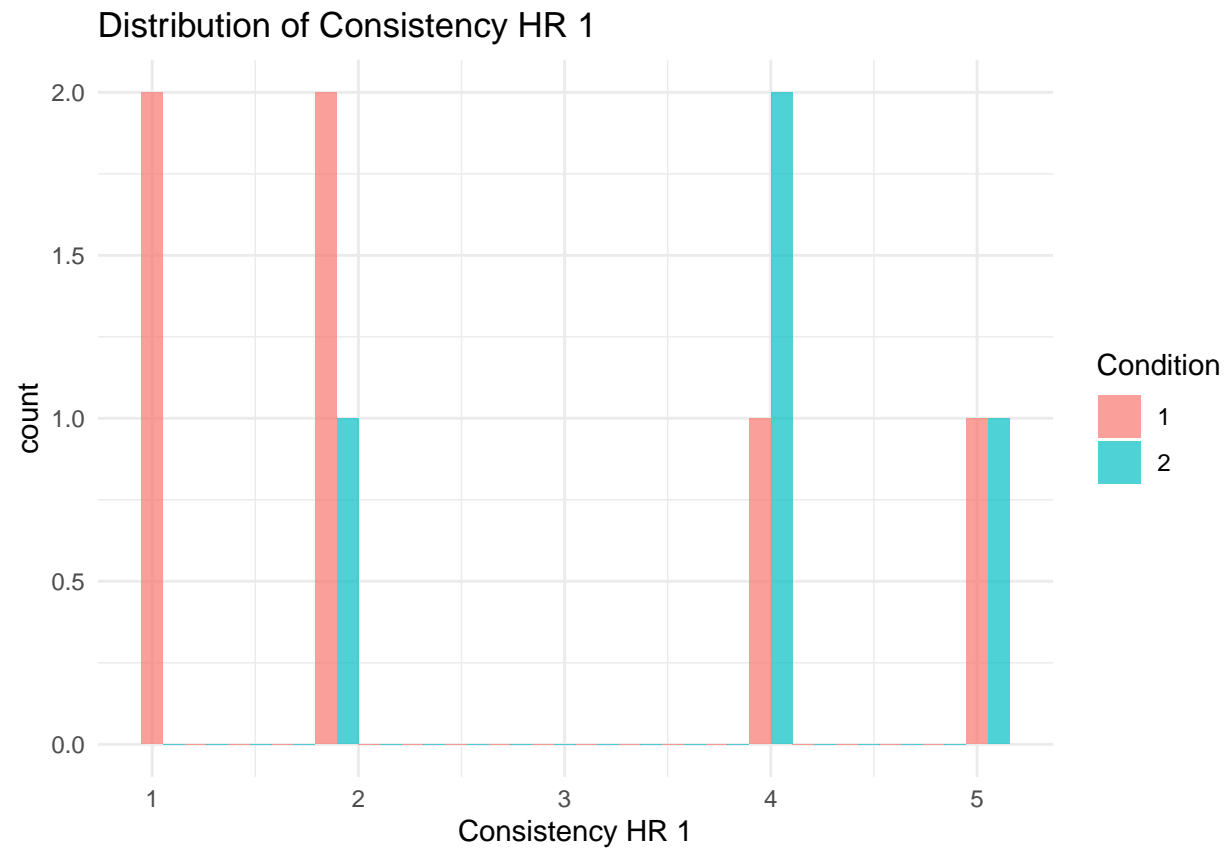
```
ggplot(JAR_Social_Invitees, aes(x = Consistency_AI_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Consistency AI 3",  
        x = "Consistency AI 3",  
        fill = "Condition") +  
  theme_minimal()
```

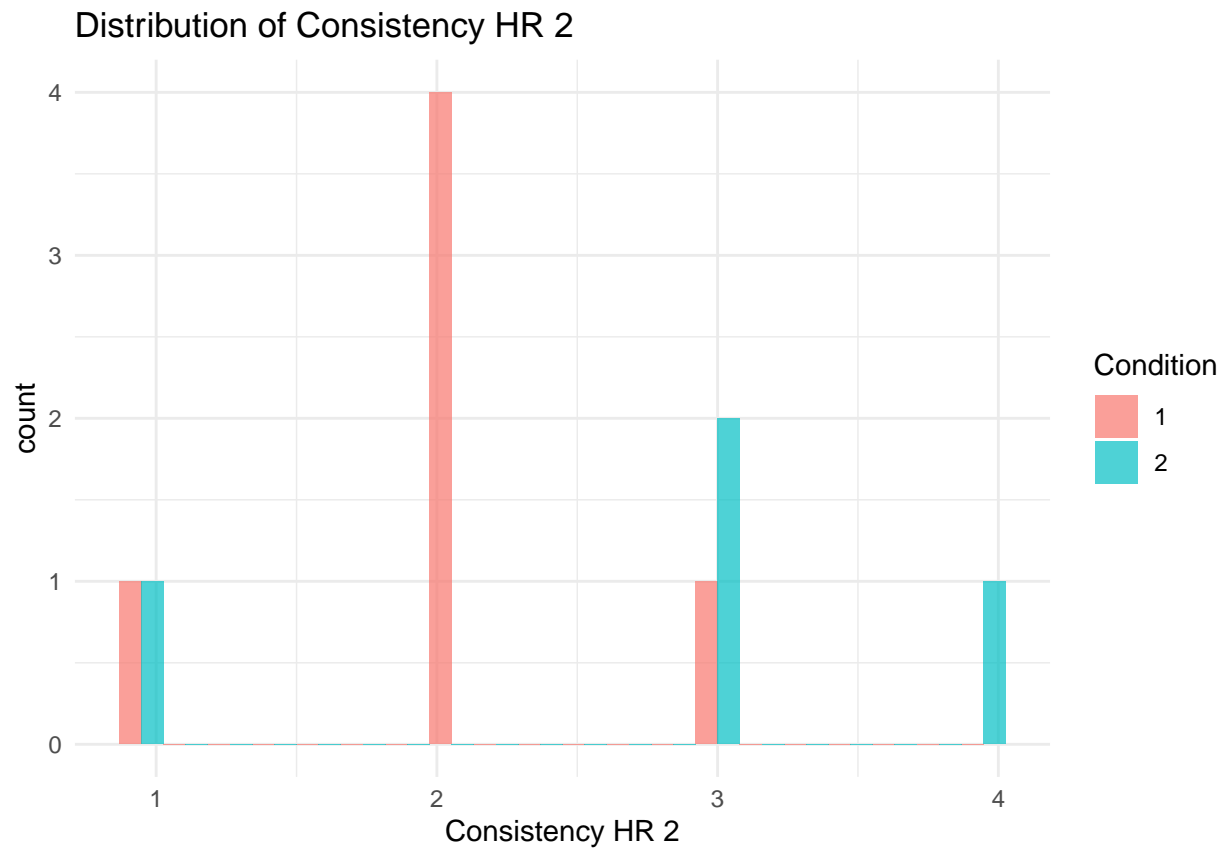
```
ggplot(JAR_Social_Invitees, aes(x = Consistency, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Consistency",  
        x = "Consistency",  
        fill = "Condition") +  
  theme_minimal()
```



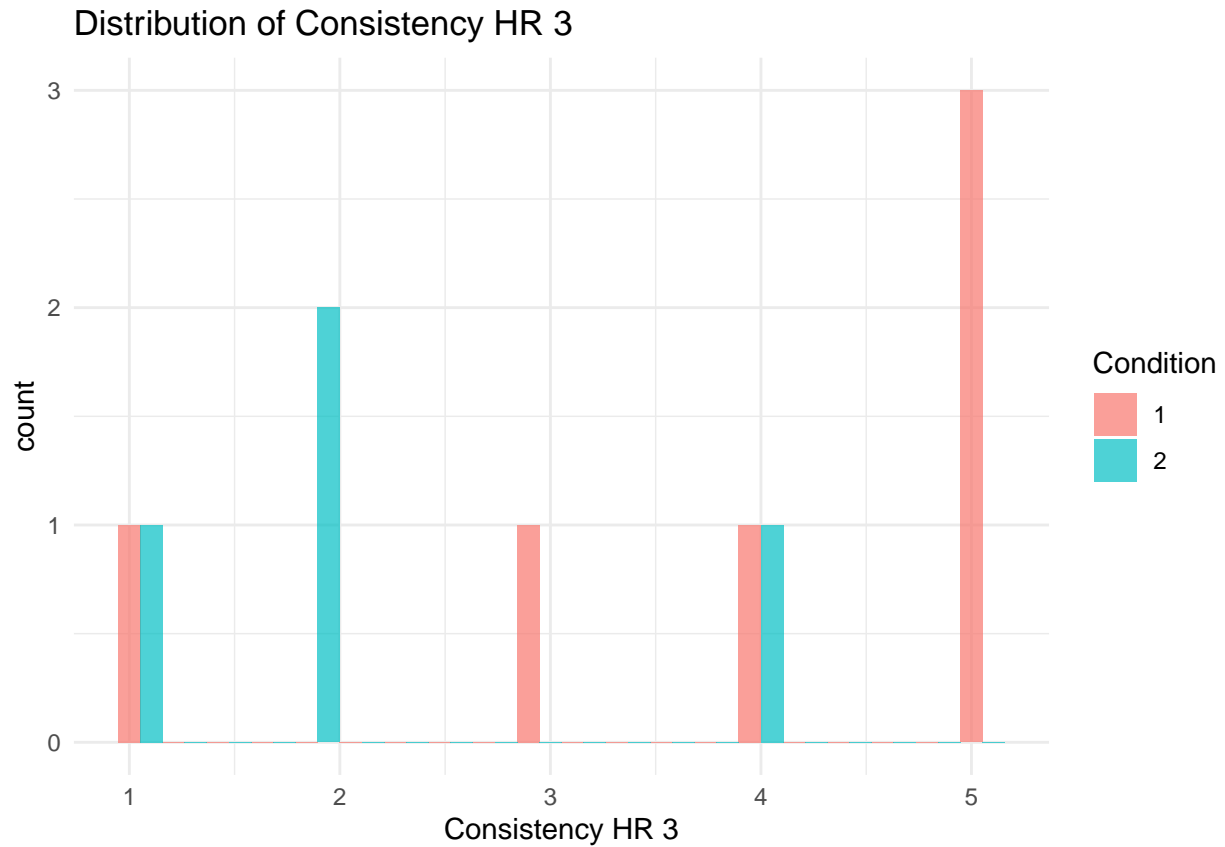
```
#####Consistency HR#####
ggplot(JAR_Social_Invitees, aes(x = Consistency_HR_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Consistency HR 1",
        x = "Consistency HR 1",
        fill = "Condition") +
  theme_minimal()
```



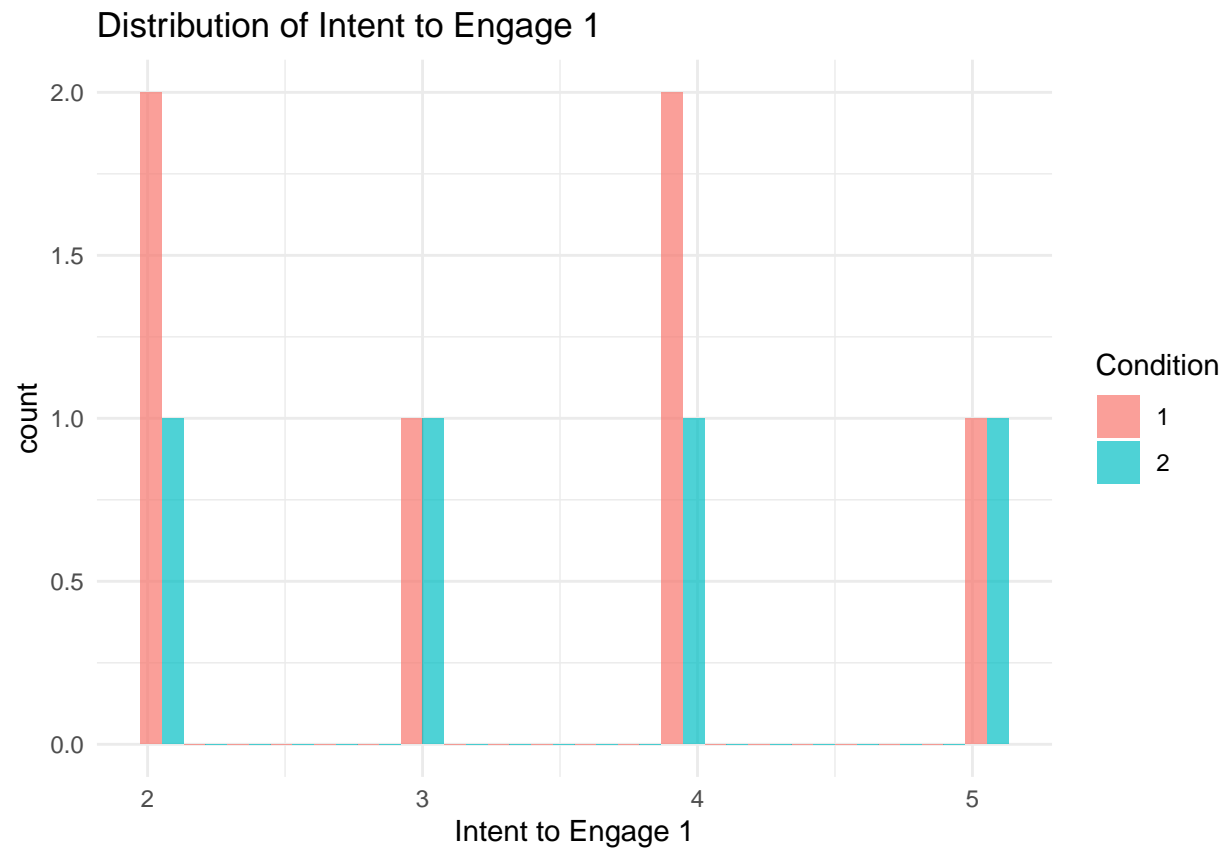
```
ggplot(JAR_Social_Invitees, aes(x = Consistency_HR_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Consistency HR 2",  
        x = "Consistency HR 2",  
        fill = "Condition") +  
  theme_minimal()
```



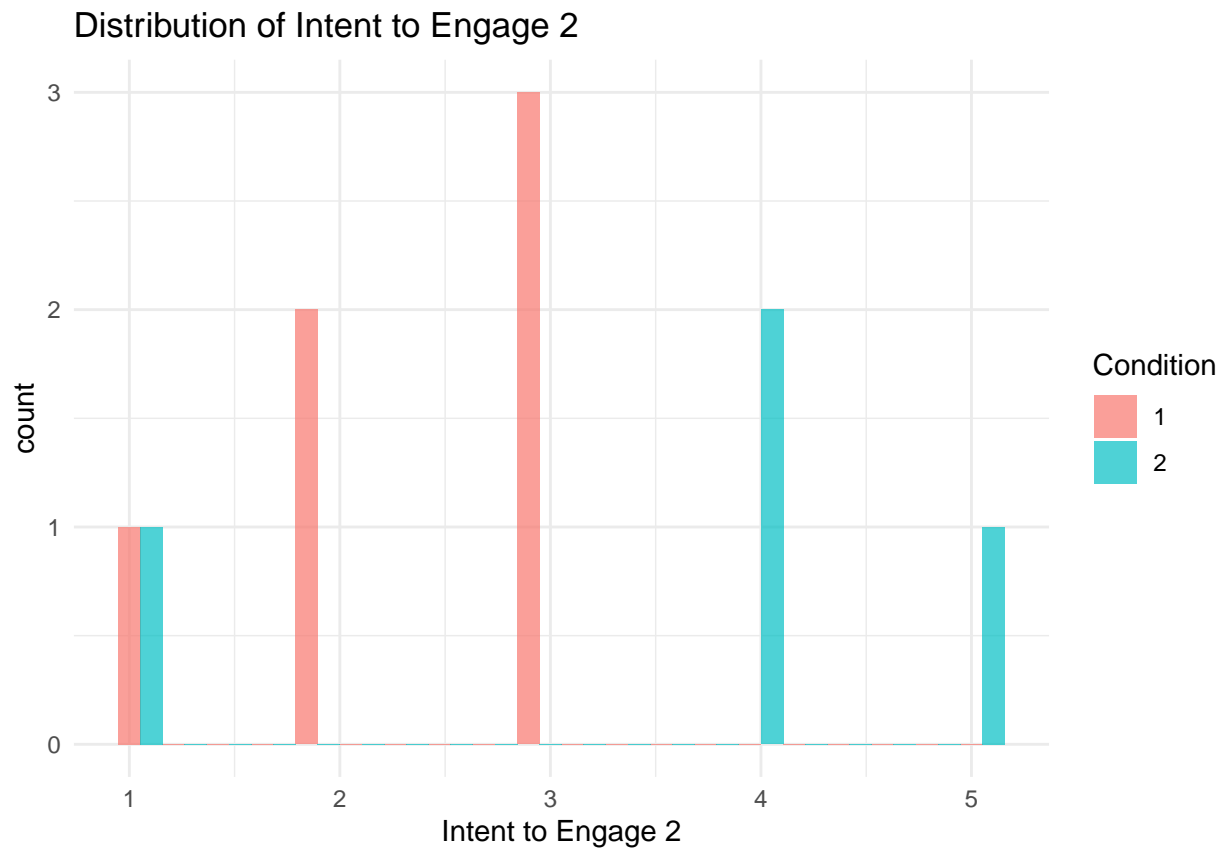
```
ggplot(JAR_Social_Invitees, aes(x = Consistency_HR_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Consistency HR 3",  
        x = "Consistency HR 3",  
        fill = "Condition") +  
  theme_minimal()
```



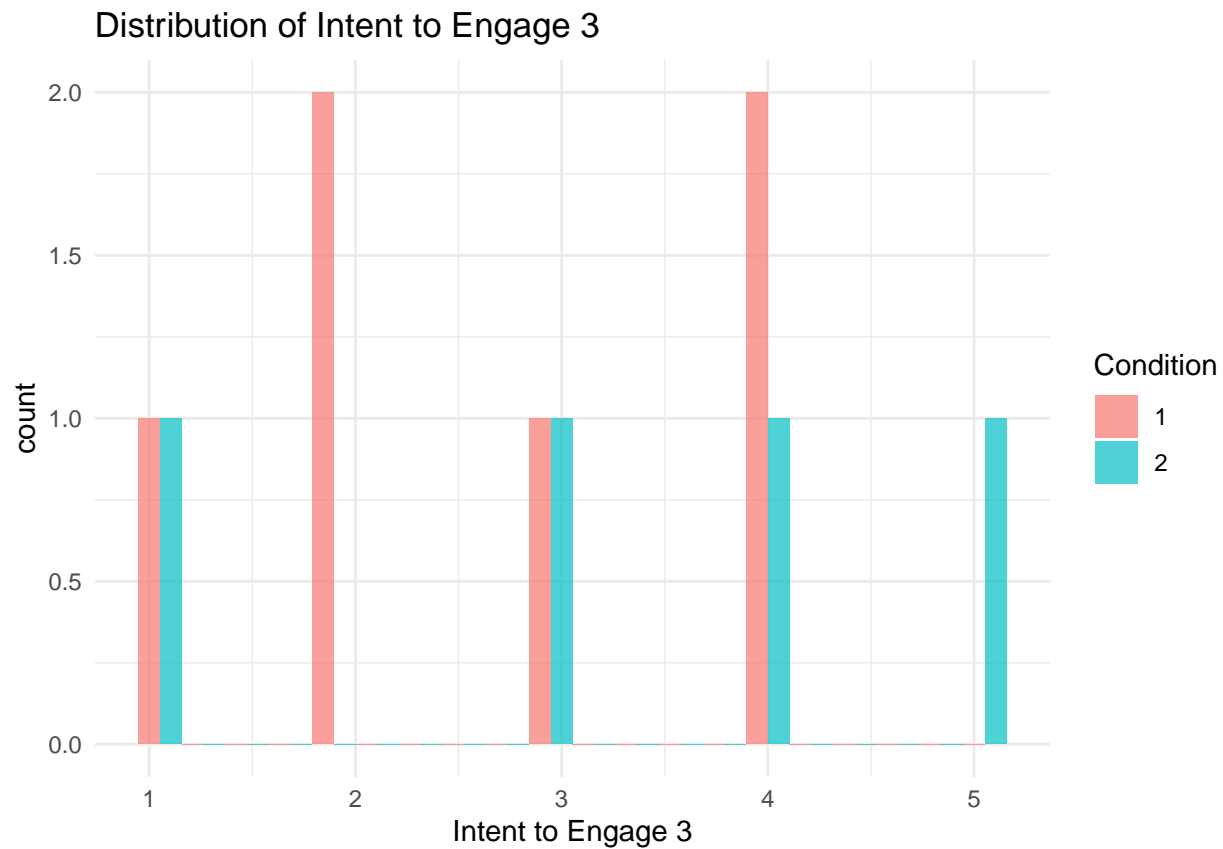
```
#####Intent to Engage#####
ggplot(JAR_Social_Invitees, aes(x = Intent_Engag_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Intent to Engage 1",
       x = "Intent to Engage 1",
       fill = "Condition") +
  theme_minimal()
```



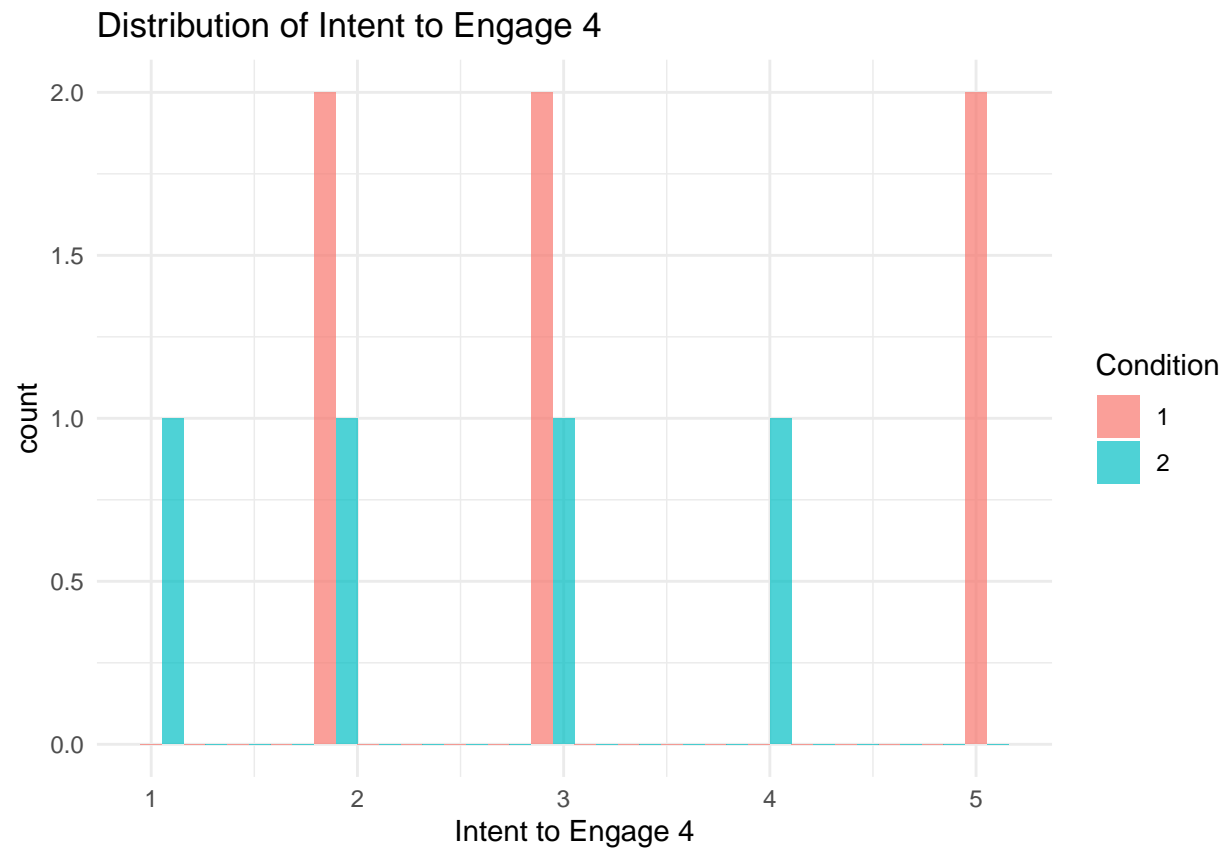
```
ggplot(JAR_Social_Invitees, aes(x = Intent_Engag_2_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Intent to Engage 2",  
        x = "Intent to Engage 2",  
        fill = "Condition") +  
  theme_minimal()
```



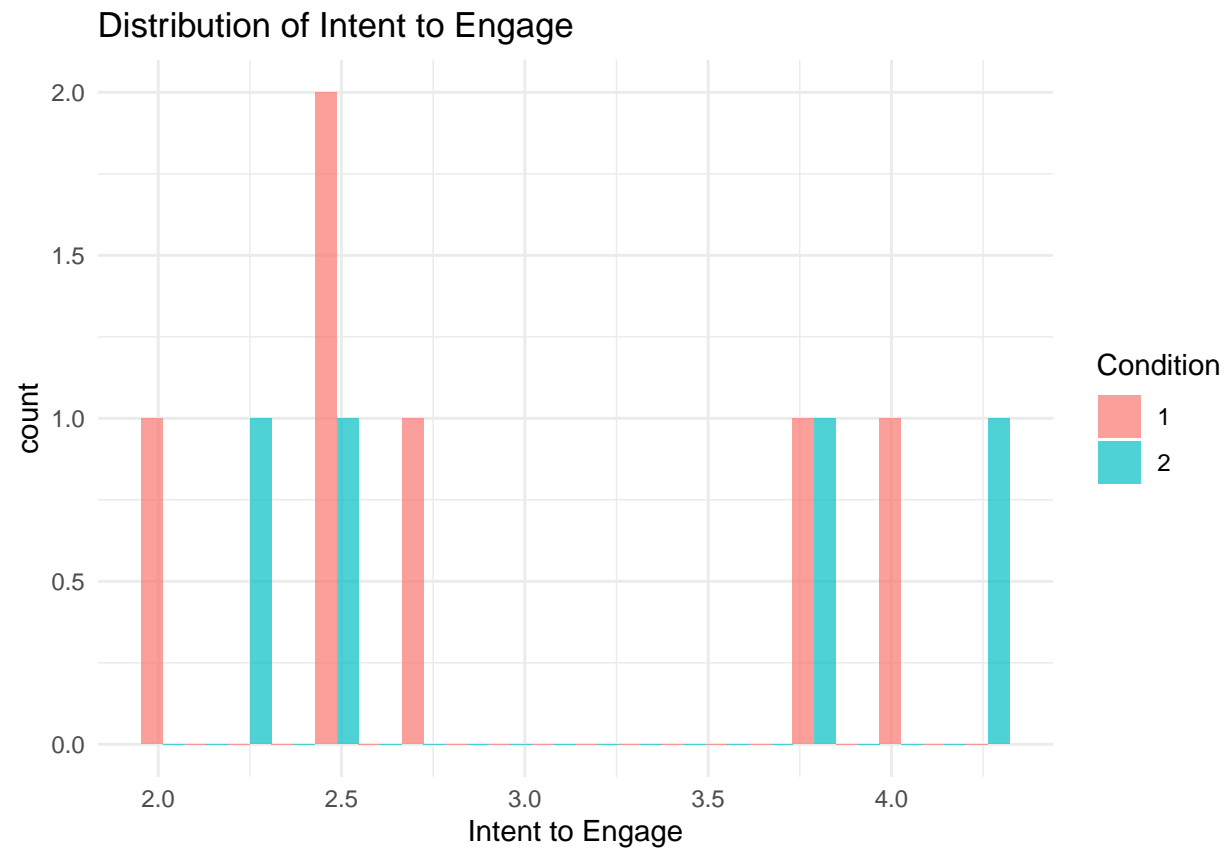
```
ggplot(JAR_Social_Invitees, aes(x = Intent_Engag_3_num, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Intent to Engage 3",  
        x = "Intent to Engage 3",  
        fill = "Condition") +  
  theme_minimal()
```



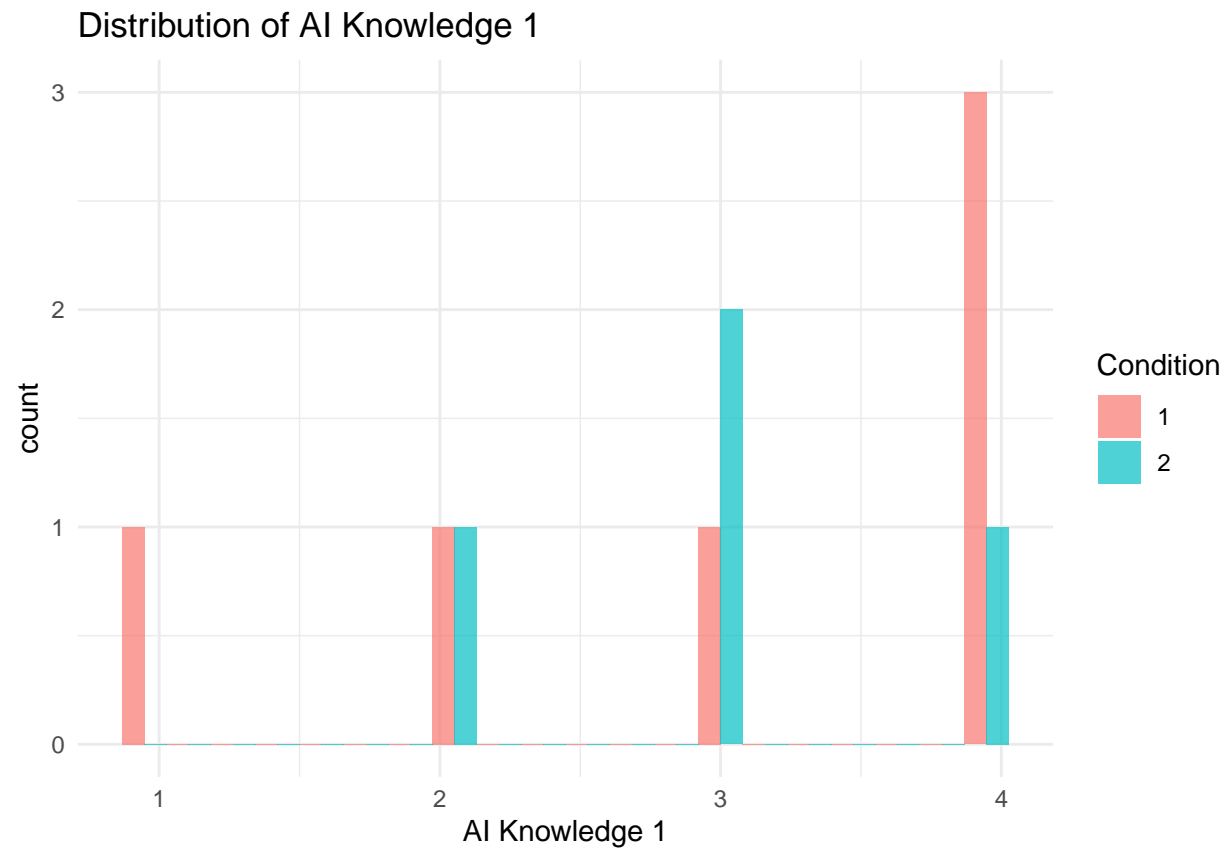
```
ggplot(JAR_Social_Invitees, aes(x = Intent_Engag_4_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of Intent to Engage 4",
       x = "Intent to Engage 4",
       fill = "Condition") +
  theme_minimal()
```

```
ggplot(JAR_Social_Invitees, aes(x = Intent, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of Intent to Engage",  
        x = "Intent to Engage",  
        fill = "Condition") +  
  theme_minimal()
```

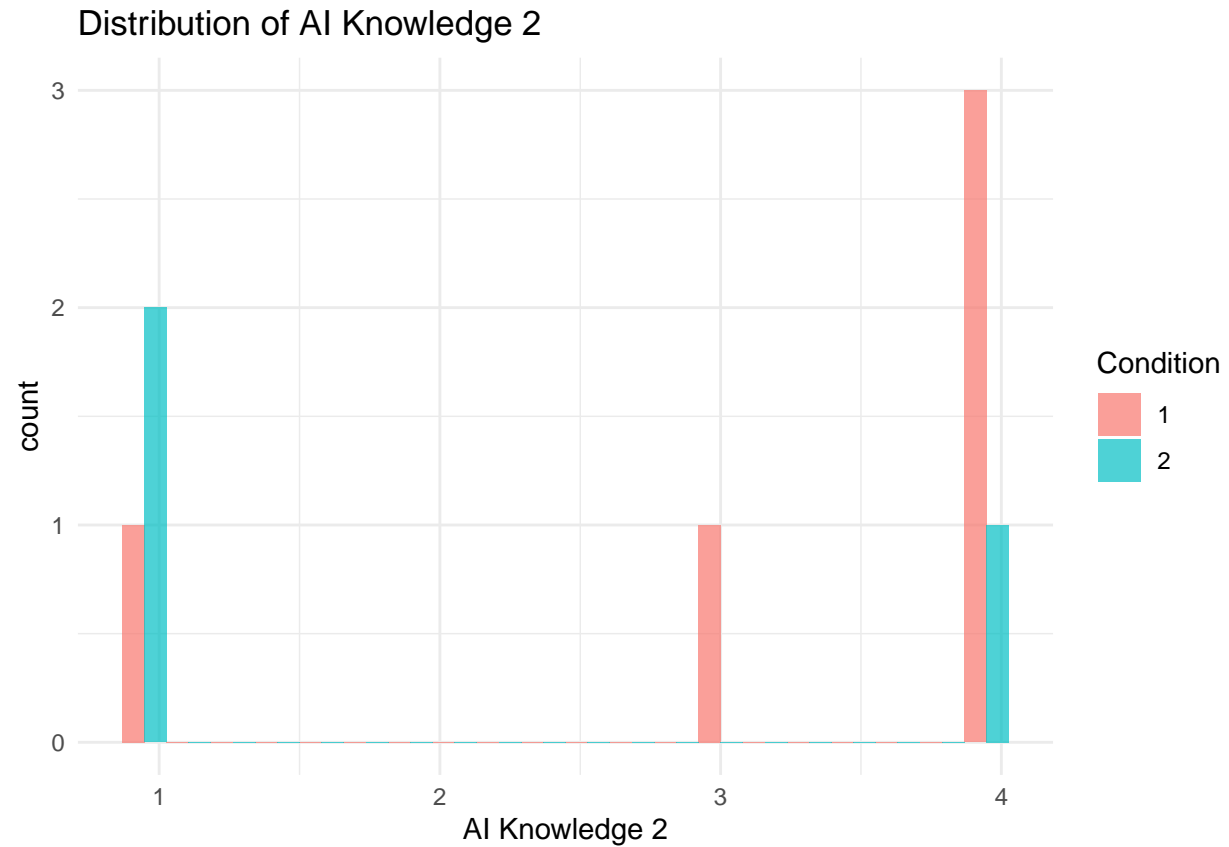


```
#####AI Knowledge#####
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge_Experei_1_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of AI Knowledge 1",
        x = "AI Knowledge 1",
        fill = "Condition") +
  theme_minimal()
```



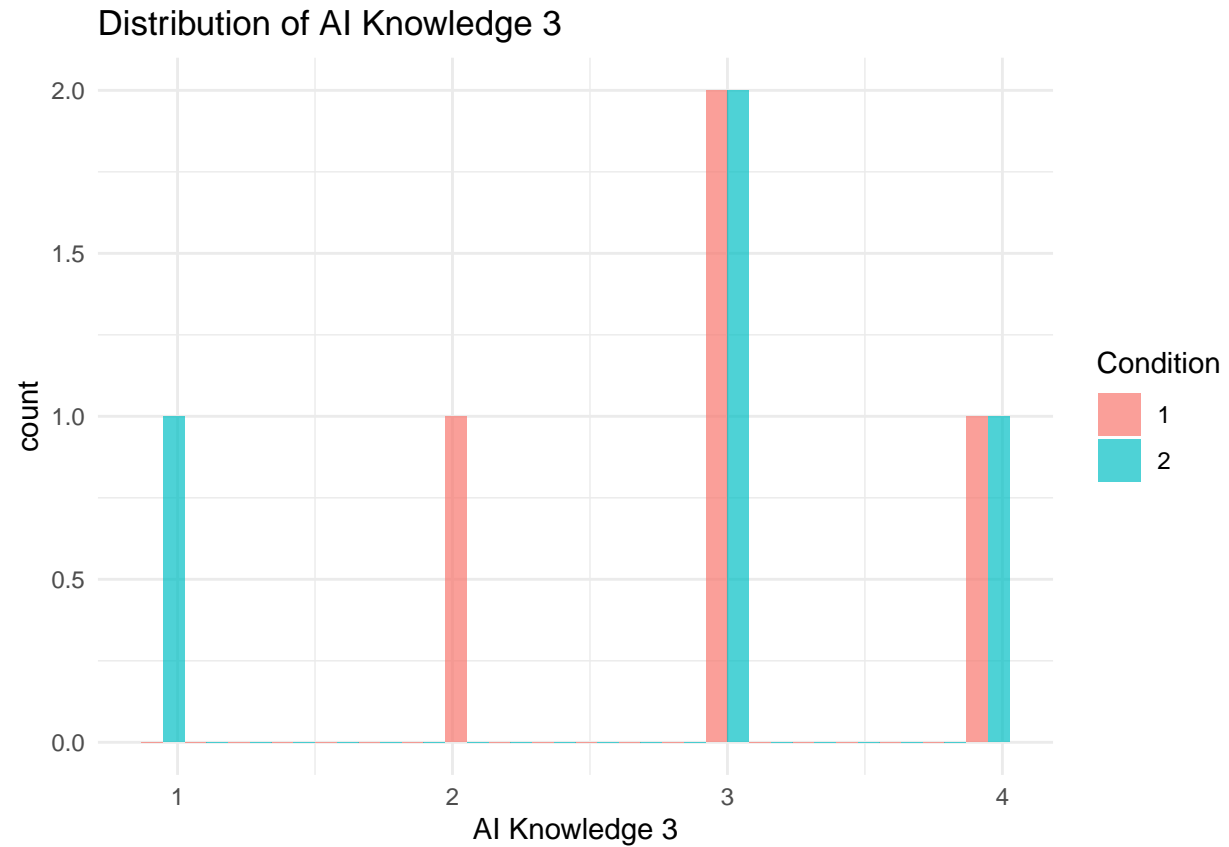
```
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge_Experei_2_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of AI Knowledge 2",
       x = "AI Knowledge 2",
       fill = "Condition") +
  theme_minimal()
```

```
## Warning: Removed 2 rows containing non-finite outside the scale range
## ('stat_bin()').
```



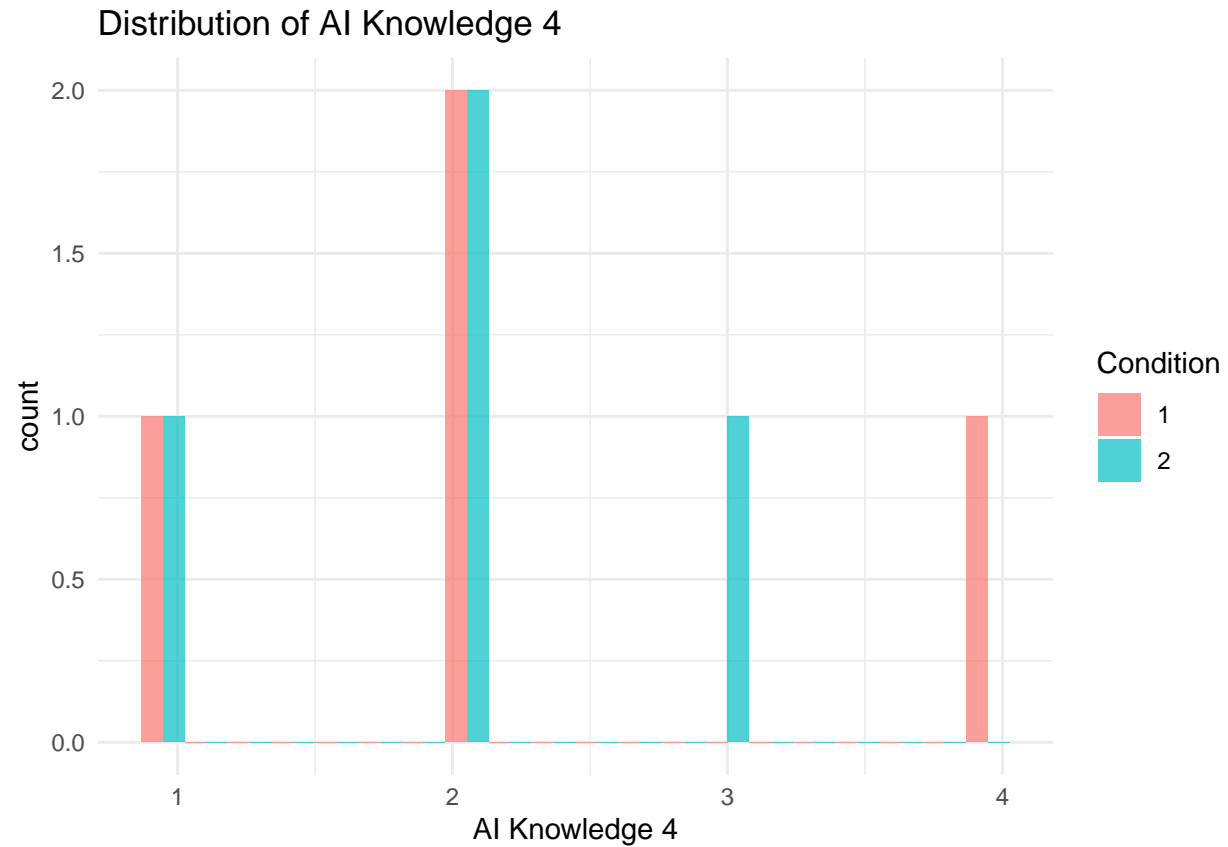
```
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge_Experei_3_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of AI Knowledge 3",
       x = "AI Knowledge 3",
       fill = "Condition") +
  theme_minimal()
```

```
## Warning: Removed 2 rows containing non-finite outside the scale range
## ('stat_bin()').
```



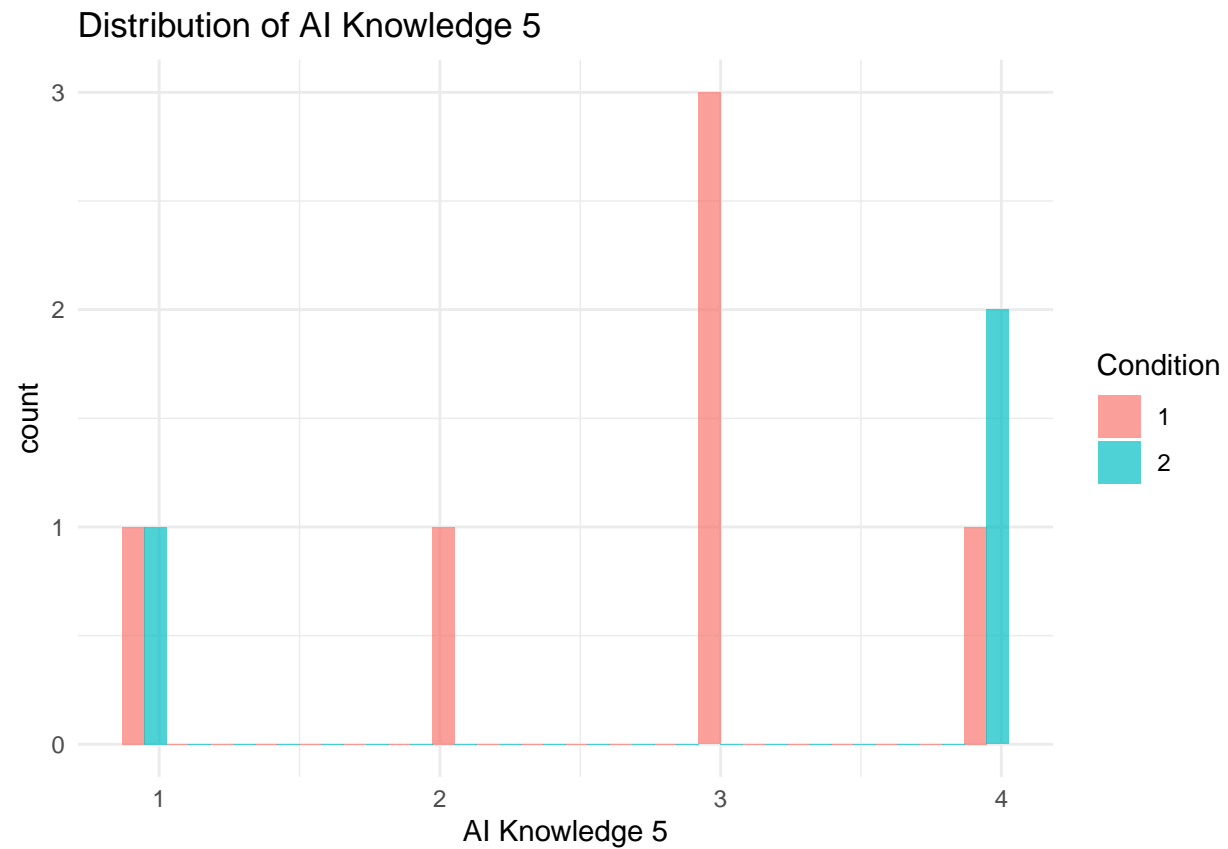
```
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge_Experei_4_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of AI Knowledge 4",
        x = "AI Knowledge 4",
        fill = "Condition") +
  theme_minimal()
```

```
## Warning: Removed 2 rows containing non-finite outside the scale range
## ('stat_bin()').
```

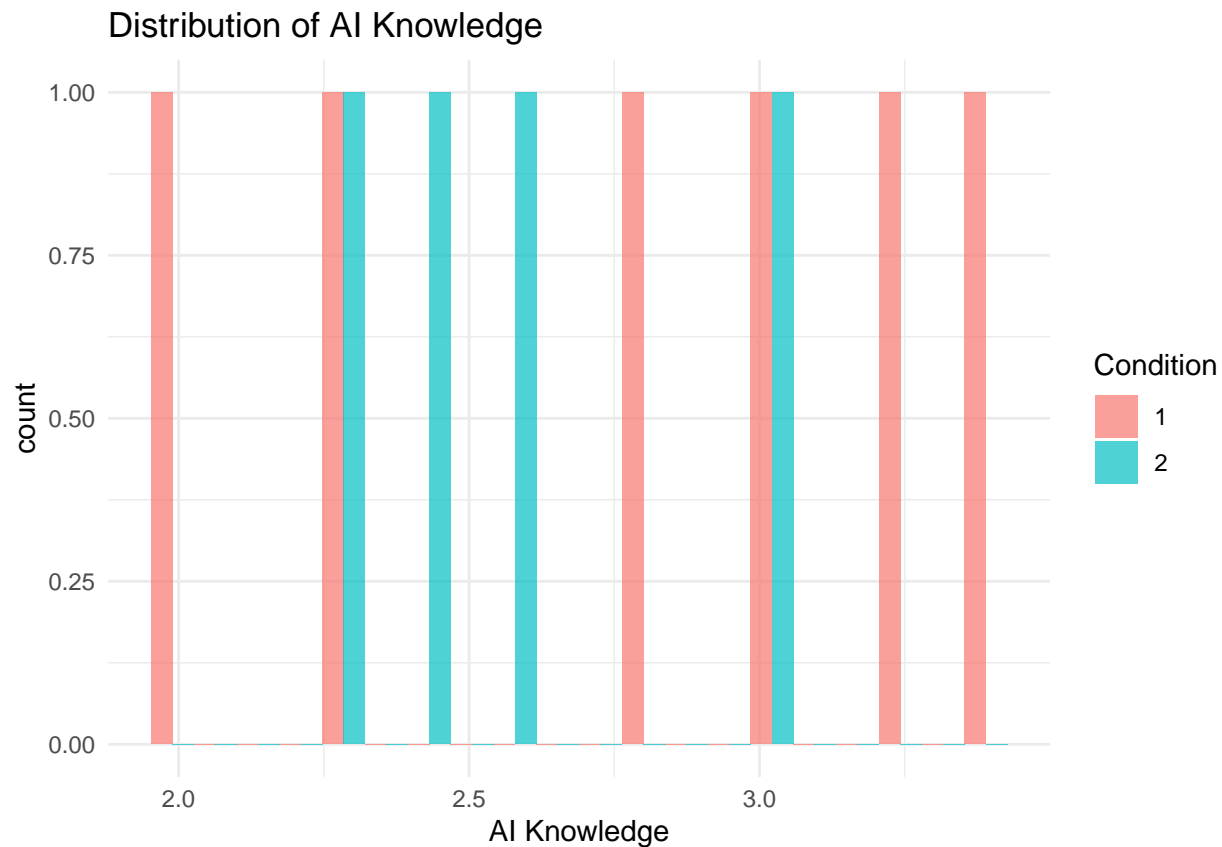


```
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge_Experei_5_num, fill = factor(Condition))) +
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +
  labs(title = "Distribution of AI Knowledge 5",
       x = "AI Knowledge 5",
       fill = "Condition") +
  theme_minimal()
```

```
## Warning: Removed 1 row containing non-finite outside the scale range
## ('stat_bin()').
```



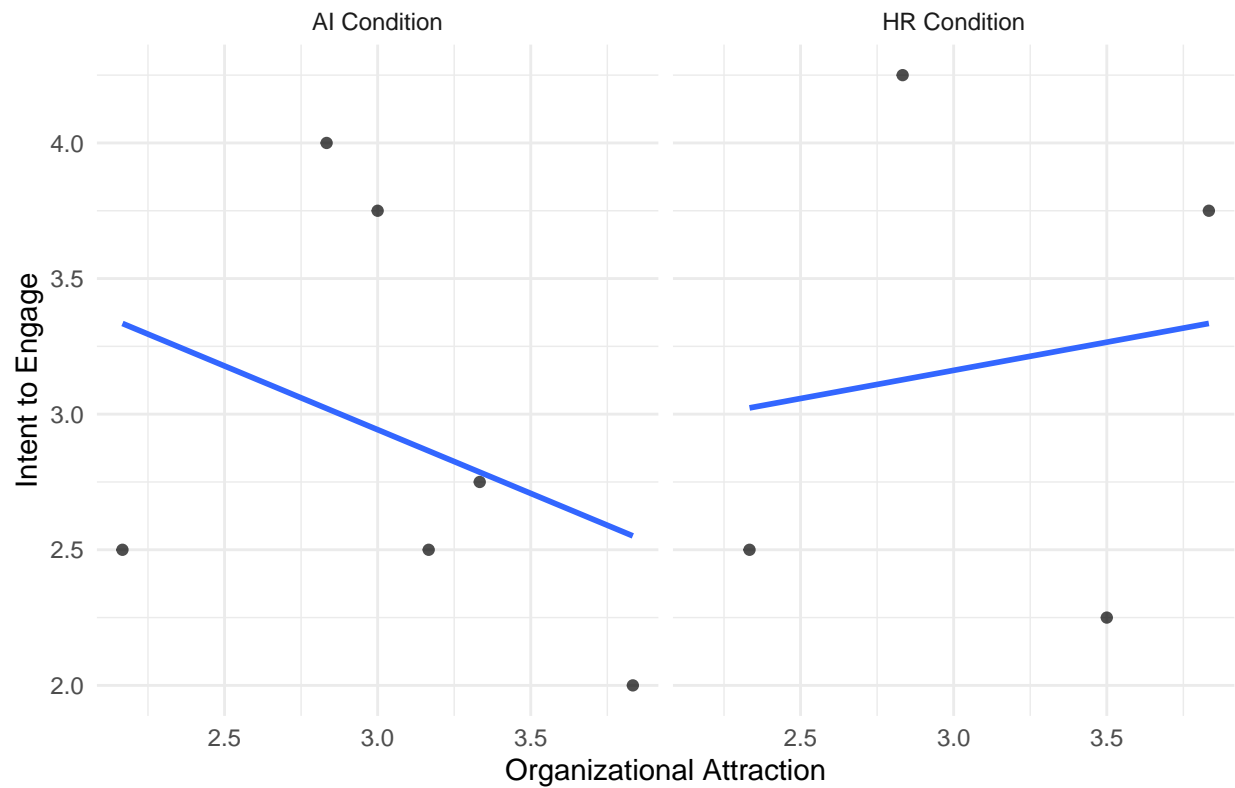
```
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge, fill = factor(Condition))) +  
  geom_histogram(position = "dodge", bins = 20, alpha = 0.7) +  
  labs(title = "Distribution of AI Knowledge",  
        x = "AI Knowledge",  
        fill = "Condition") +  
  theme_minimal()
```



```
#multivariate visualization: plot thing against each other
#relationship between each predictive variable and outcome variable (scatter plot)#####
#####Organizational Attraction#####
# AL vs HR condition
JAR_Social_Invitees %>%
  ggplot(aes(Org_Attraction, Intent), color = factor(Condition)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ Condition, labeller = labeller(Condition = c("1" = "AI Condition", "2" = "HR Condition")))
  labs(title = "Organizational Attraction vs Intent to Engage based on each condition",
        x = "Organizational Attraction",
        y = "Intent to Engage",
        color = "Condition") +
  theme_minimal()
```

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

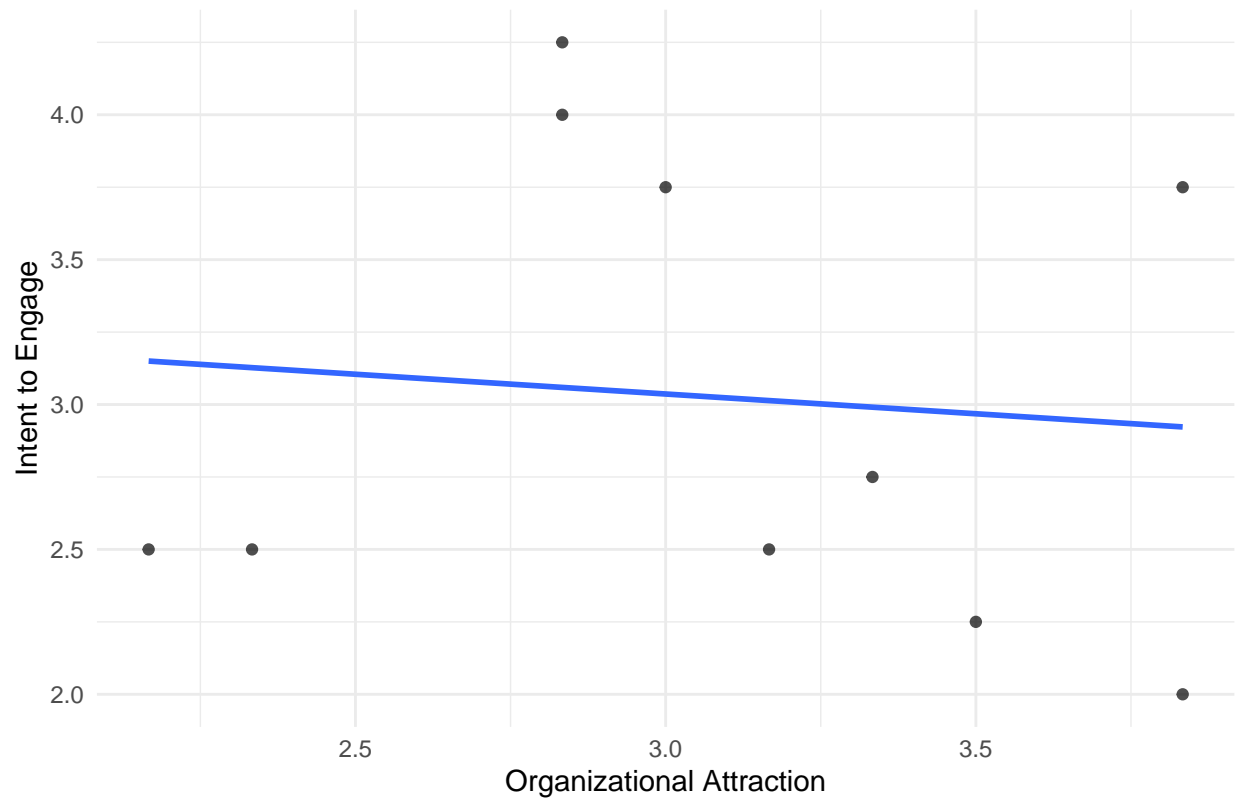

Organizational Attraction vs Intent to Engage based on each condition



```
#whole dataset
JAR_Social_Invitees %>%
  ggplot(aes(Org_Attraction, Intent)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  labs(title = "Organizational Attraction vs Intent to Engage",
       x = "Organizational Attraction",
       y = "Intent to Engage") +
  theme_minimal()
```

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

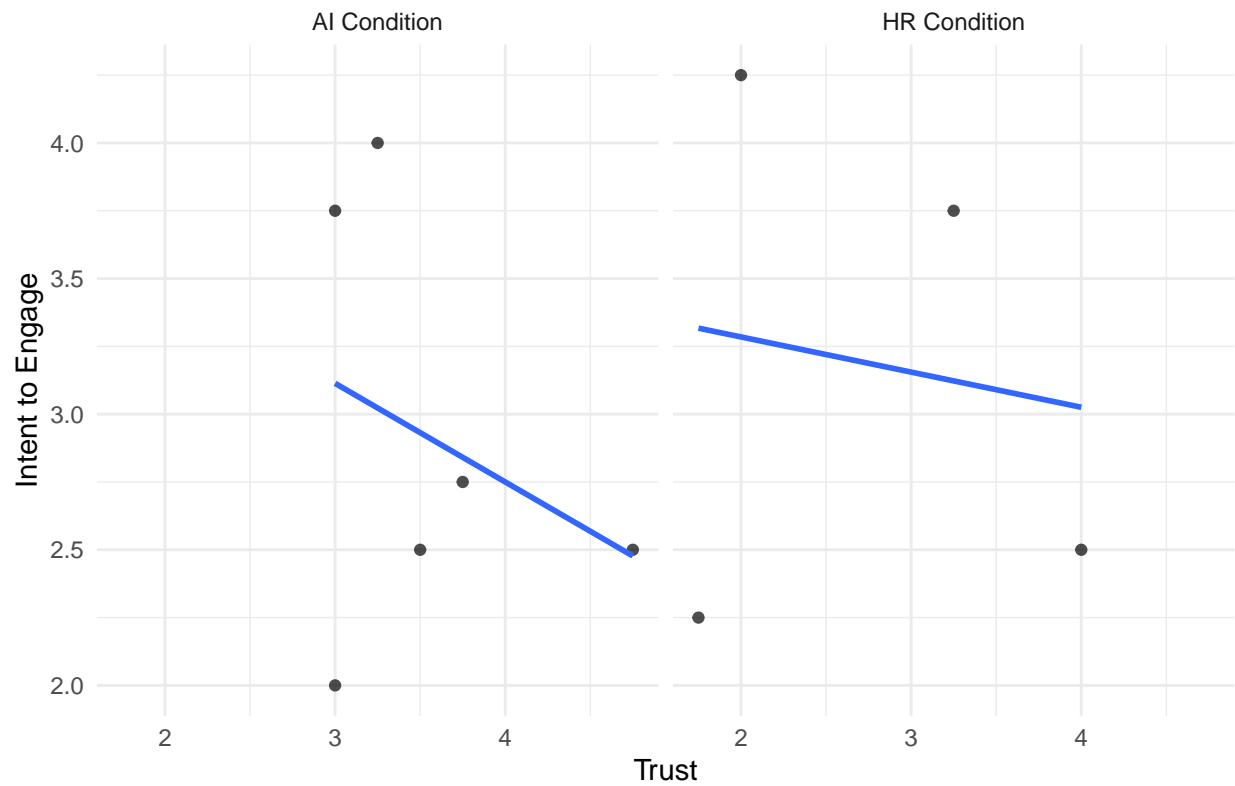
Organizational Attraction vs Intent to Engage



```
#####Trust#####
#AI vs HR
JAR_Social_Invitees %>%
  ggplot(aes(Trust, Intent), color = factor(Condition)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ Condition, labeller = labeller(Condition = c("1" = "AI Condition", "2" = "HR Condition")))
  labs(title = "Trust vs Intent to Engage based on each condition",
        x = "Trust",
        y = "Intent to Engage",
        color = "Condition") +
  theme_minimal()

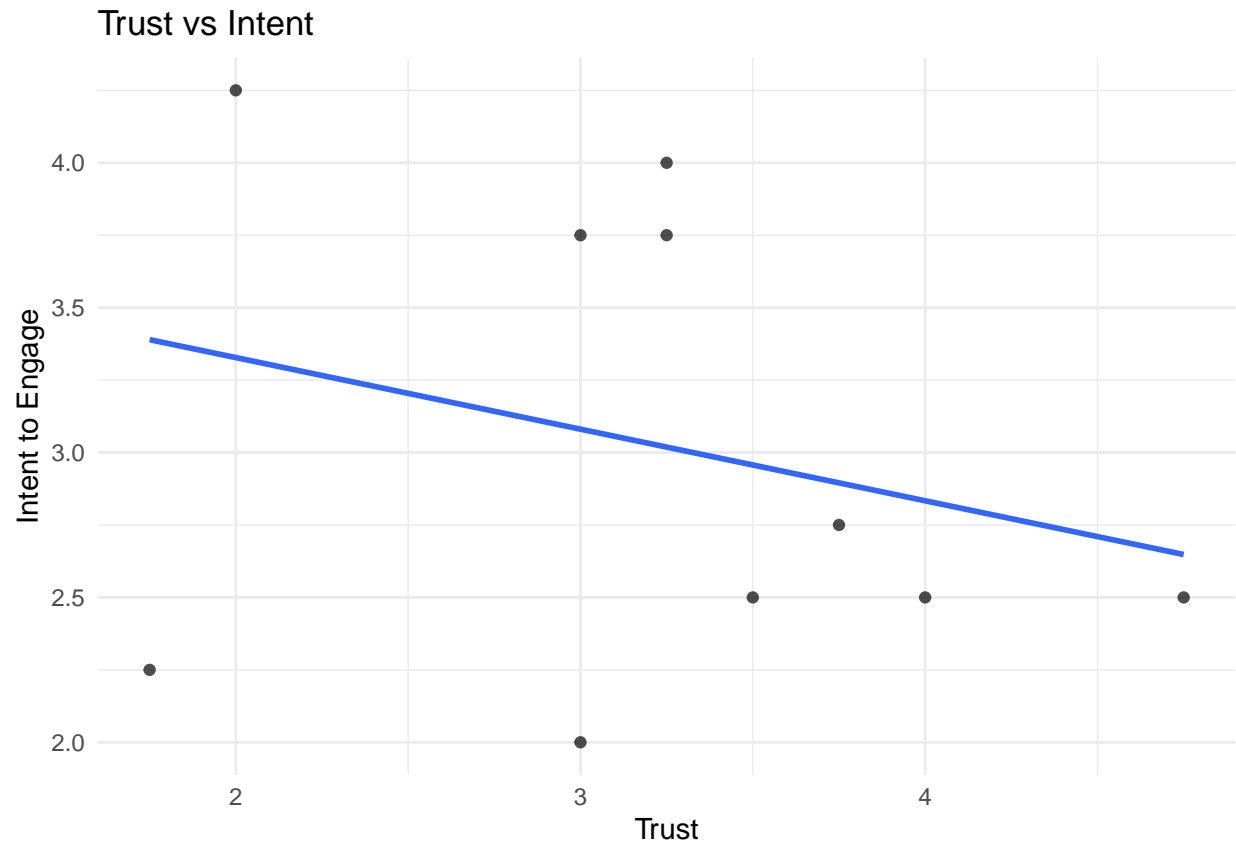
## 'geom_smooth()' using formula = 'y ~ x'
```

Trust vs Intent to Engage based on each condition



```
#whole data set
JAR_Social_Invitees %>%
  ggplot(aes(Trust, Intent)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  labs(title = "Trust vs Intent",
        x = "Trust",
        y = "Intent to Engage") +
  theme_minimal()
```

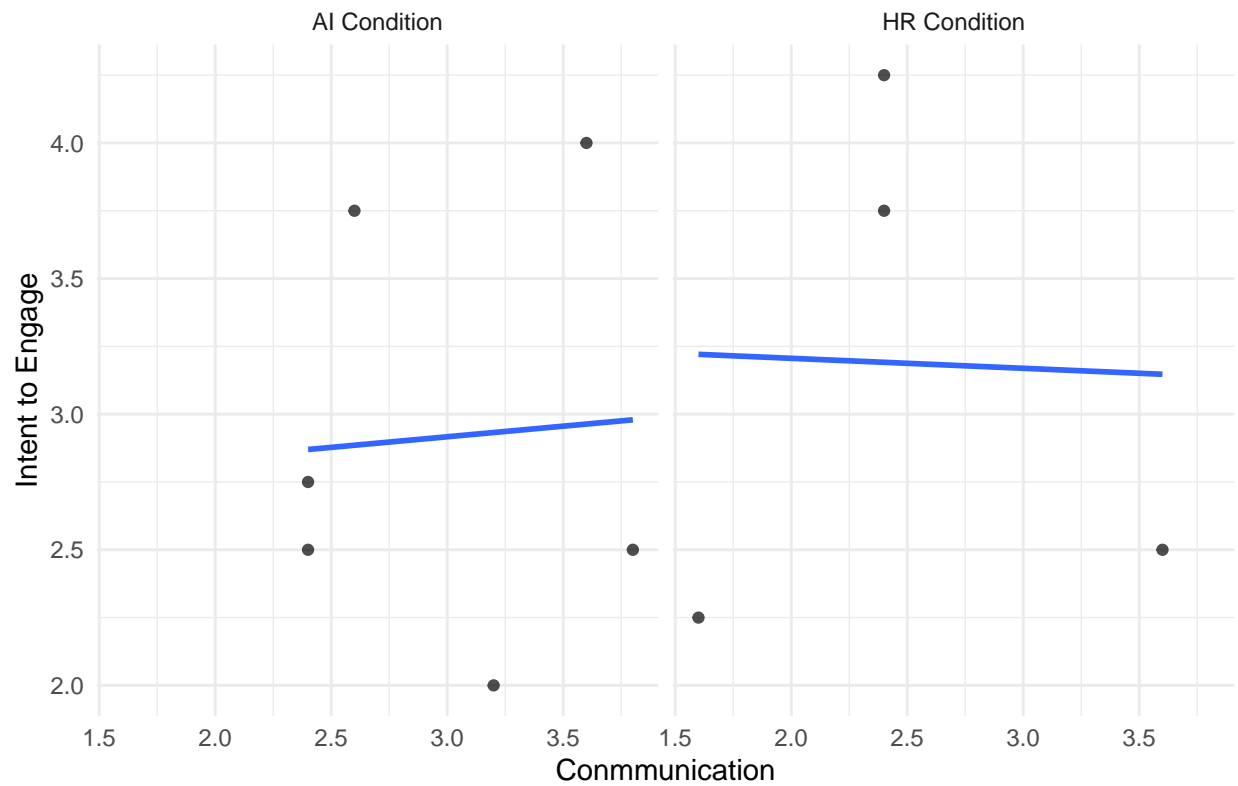
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
#####Communication#####
#AI vs HR
JAR_Social_Invitees %>%
  ggplot(aes(Communication, Intent), color = factor(Condition)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ Condition, labeller = labeller(Condition = c("1" = "AI Condition", "2" = "HR Condition")))
  labs(title = "Communication vs Intent to Engage based on each condition",
        x = "Communication",
        y = "Intent to Engage",
        color = "Condition") +
  theme_minimal()

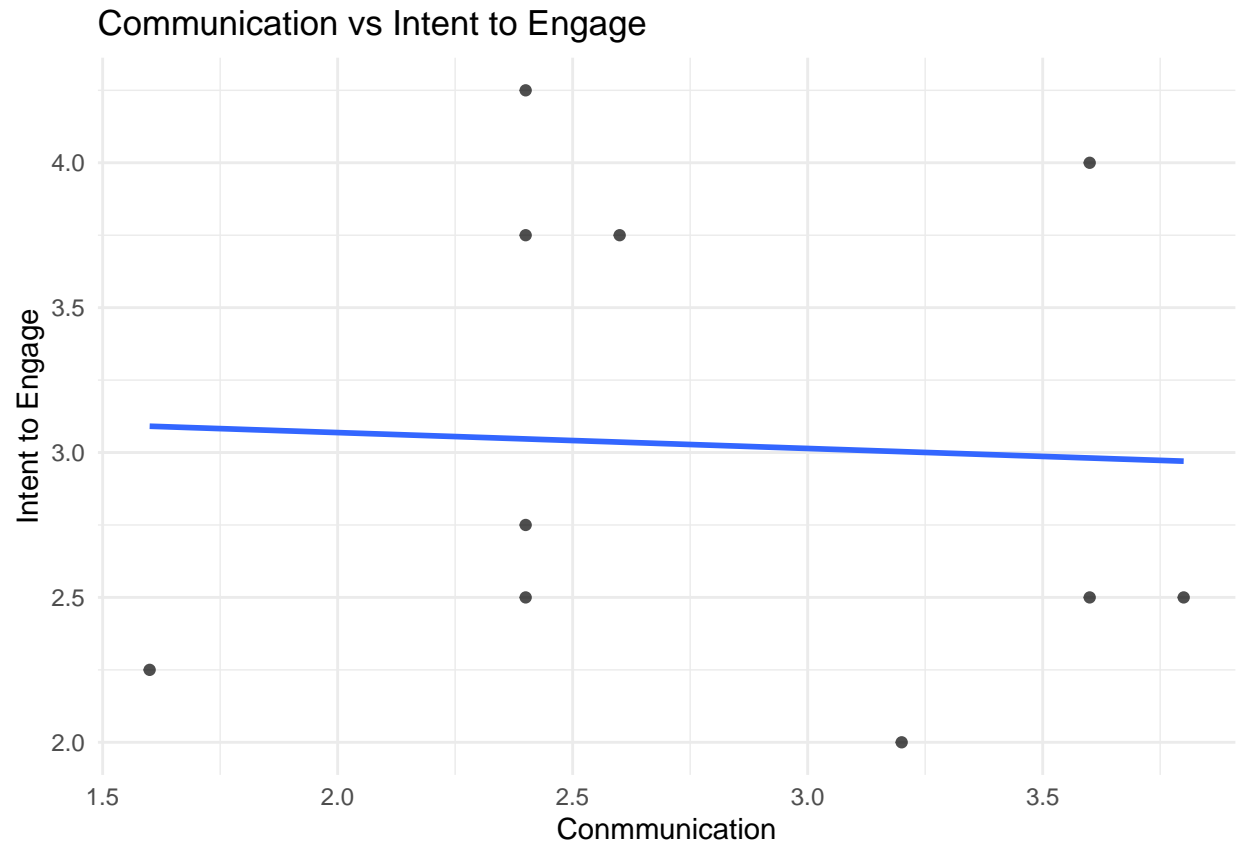
## 'geom_smooth()' using formula = 'y ~ x'
```

Communication vs Intent to Engage based on each condition



```
# whole dataset
JAR_Social_Invitees %>%
  ggplot(aes(Communication, Intent)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  labs(title = "Communication vs Intent to Engage",
       x = "Communication",
       y = "Intent to Engage") +
  theme_minimal()
```

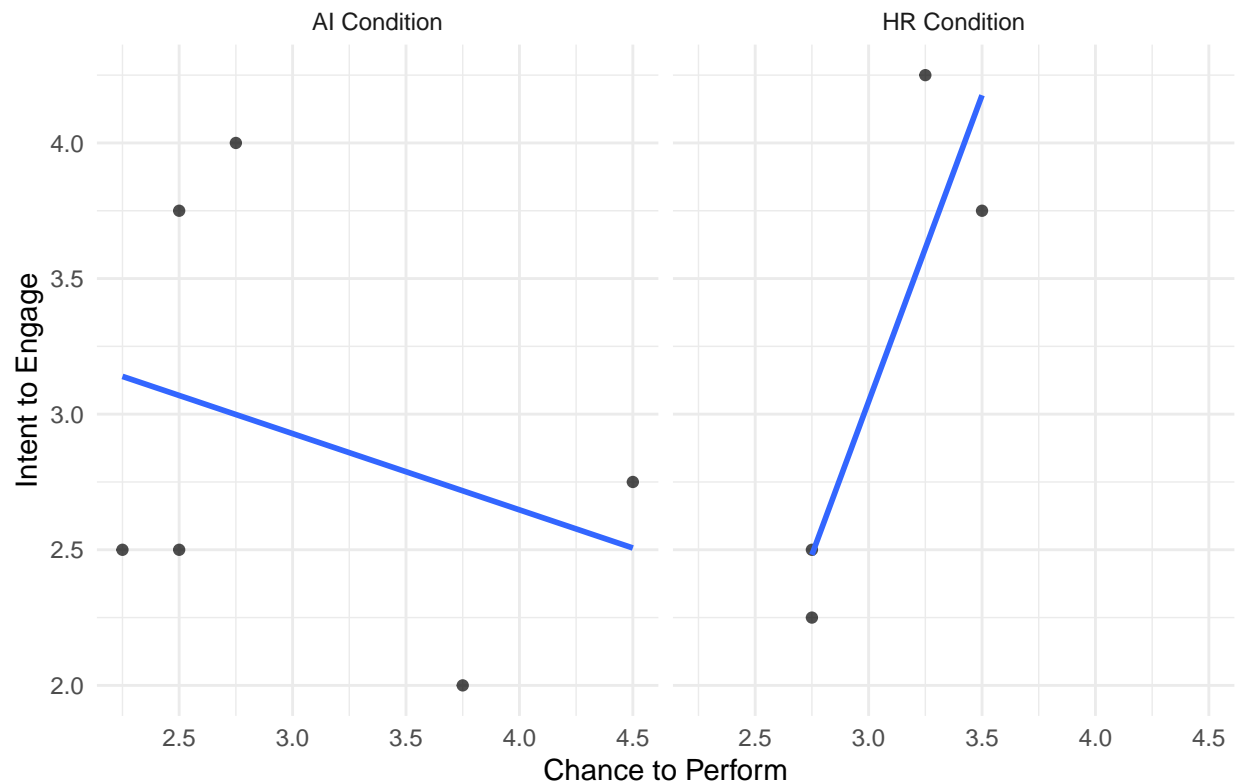
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
#####Chance to Perform#####
#AI vs HR
JAR_Social_Invitees %>%
  ggplot(aes(Chance_Perform, Intent), color = factor(Condition)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ Condition, labeller = labeller(Condition = c("1" = "AI Condition", "2" = "HR Condition")))
  labs(title = "Chance to Perform vs Intent to Engage based on each condition",
        x = "Chance to Perform",
        y = "Intent to Engage",
        color = "Condition") +
  theme_minimal()

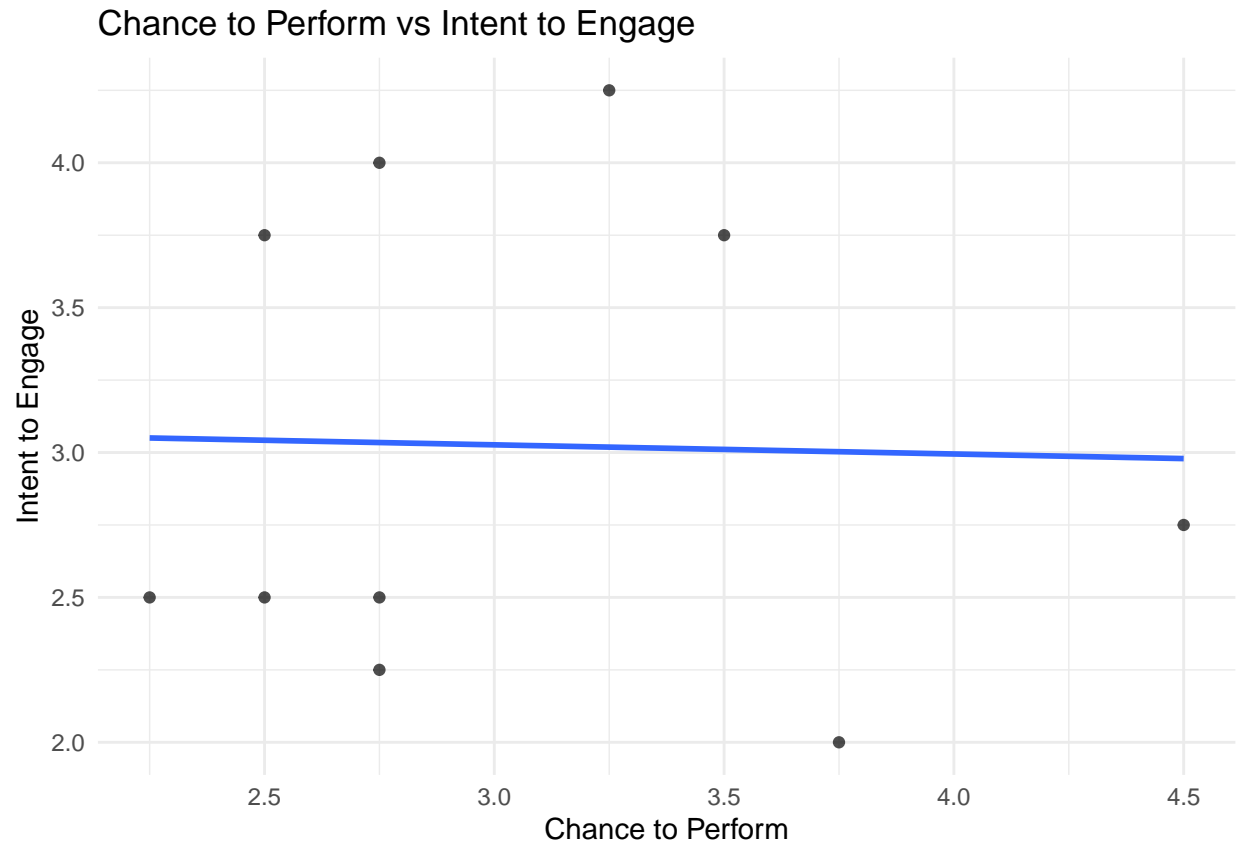
## 'geom_smooth()' using formula = 'y ~ x'
```

Chance to Perform vs Intent to Engage based on each condition



```
#whole dataset
JAR_Social_Invitees %>%
  ggplot(aes(Chance_Perform, Intent)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  labs(title = "Chance to Perform vs Intent to Engage",
       x = "Chance to Perform",
       y = "Intent to Engage") +
  theme_minimal()
```

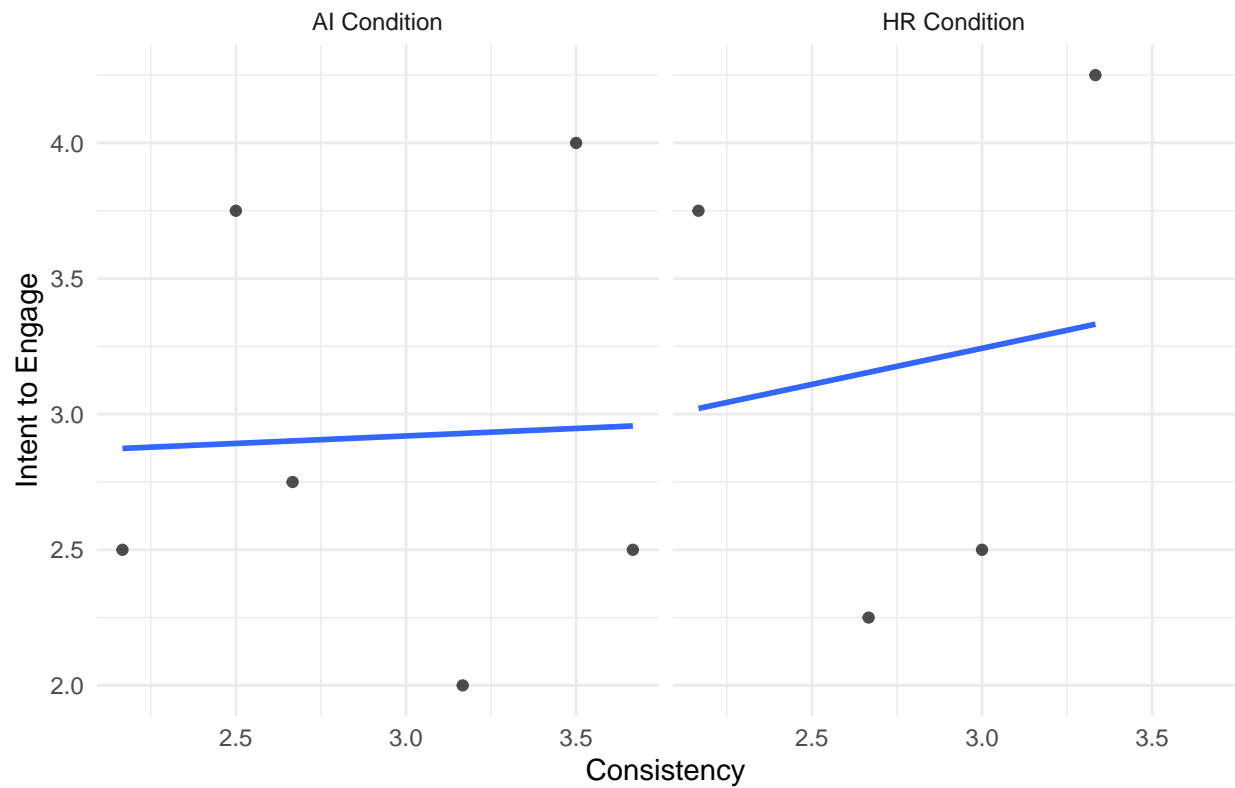
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
#####Consistency#####
#AI vs HR
JAR_Social_Invitees %>%
  ggplot(aes(Consistency, Intent), color = factor(Condition)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ Condition, labeller = labeller(Condition = c("1" = "AI Condition", "2" = "HR Condition")))
  labs(title = "Consistency vs Intent to Engage based on each condition",
        x = "Consistency",
        y = "Intent to Engage",
        color = "Condition") +
  theme_minimal()
```

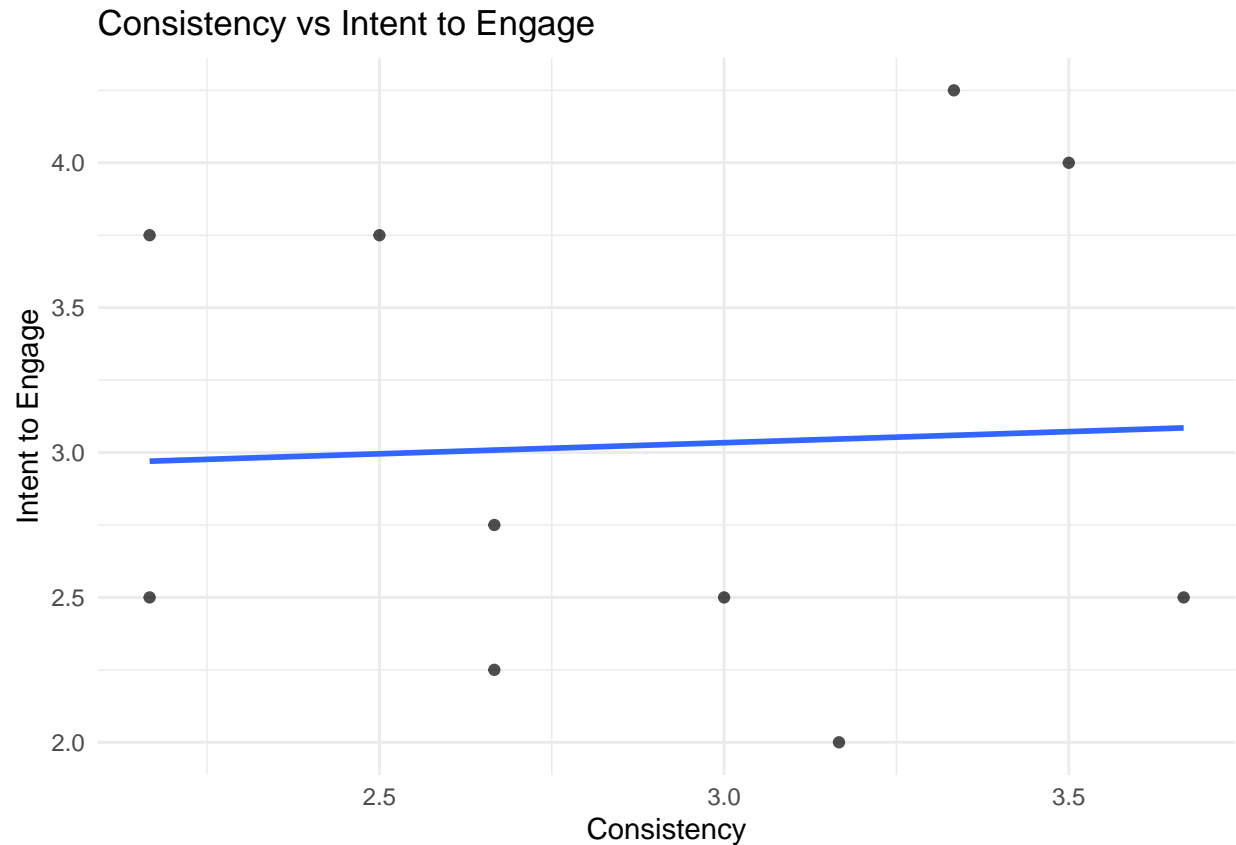
```
## 'geom_smooth()' using formula = 'y ~ x'
```


Consistency vs Intent to Engage based on each condition



```
#whole dataset
JAR_Social_Invitees %>%
  ggplot(aes(Consistency, Intent)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE) +
  labs(title = "Consistency vs Intent to Engage",
        x = "Consistency",
        y = "Intent to Engage") +
  theme_minimal()
```

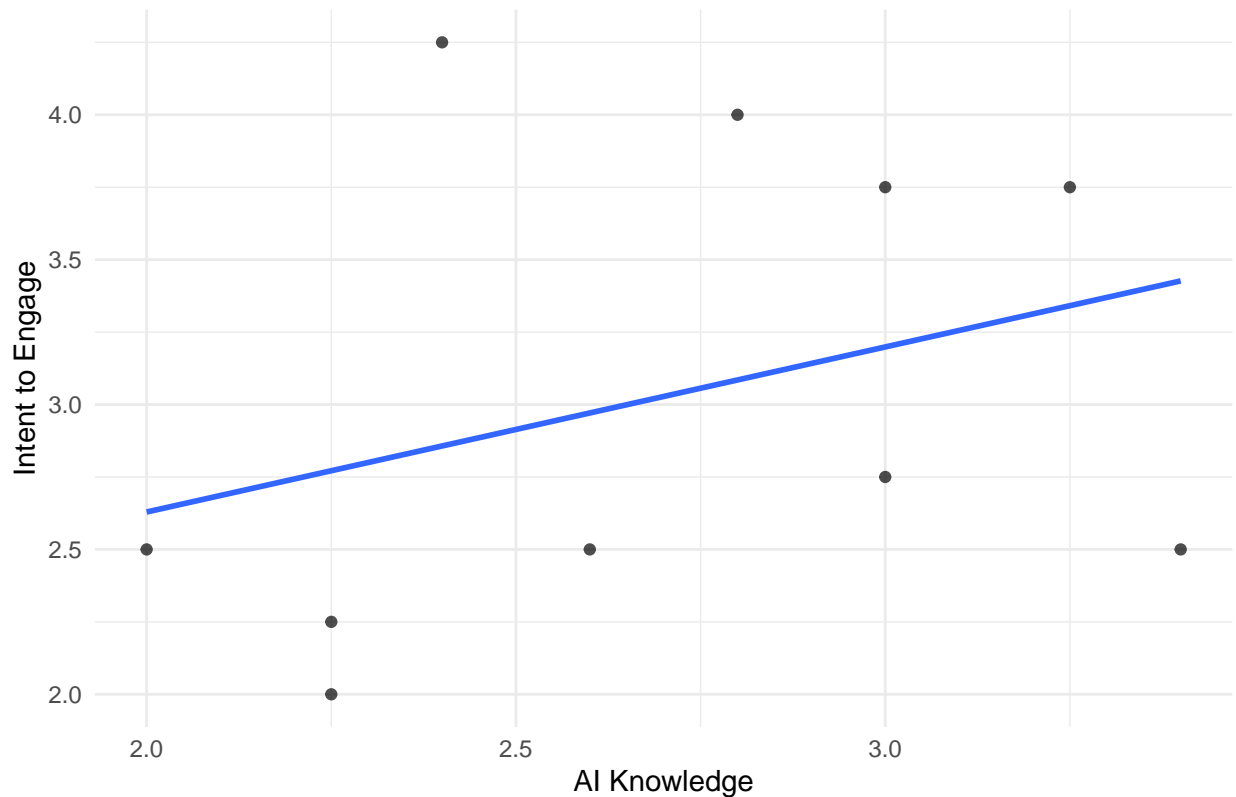
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
#####AI Knowledge#####
ggplot(JAR_Social_Invitees, aes(x = AI_Knowledge, y = Intent)) +
  geom_point(data = subset(JAR_Social_Invitees, !is.na(AI_Knowledge) & !is.na(Intent)),
    aes(x = AI_Knowledge, y = Intent),
    alpha = 0.7) +
  geom_smooth(data = subset(JAR_Social_Invitees, !is.na(AI_Knowledge) & !is.na(Intent)),
    method = "lm", se = FALSE) +
  labs(
    title = "AI Knowledge vs Intent to Engage based on each condition",
    x = "AI Knowledge",
    y = "Intent to Engage"
  ) +
  theme_minimal()
```

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

AI Knowledge vs Intent to Engage based on each condition



```
#####Organizational Attraction#####
alpha(JAR_Social_Invitees %>%
  dplyr::select(Org_Attraction_1_num:Org_Attraction_6_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Org_Attraction_1_num:Org_Attraction_6_num)): 
## should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Org_Attraction_5_num Org_Attraction_6_num ) were negatively correlated with the total s
## probably should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Org_Attraction_1_num:Org_Attraction_6_num))
##
##   raw_alpha std.alpha G6(smc) average_r    S/N ase mean   sd median_r
##   -0.0092   -0.012   0.17  -0.0021 -0.012 0.49  3.1 0.57   0.064
##
##   95% confidence boundaries
##         lower alpha upper
## Feldt    -1.43 -0.01  0.71
## Duhachek -0.98 -0.01  0.96
##
## Reliability if an item is dropped:
```

```
##               raw_alpha std.alpha G6(smc) average_r    S/N alpha se
## Org_Attraction_1_num   -0.213   -0.192   0.032   -0.0332 -0.161   0.62
## Org_Attraction_2_num    0.085    0.099   0.190    0.0214  0.109   0.46
## Org_Attraction_3_num   -0.078   -0.096   0.098   -0.0178 -0.087   0.54
## Org_Attraction_4_num   -0.086   -0.099   0.083   -0.0183 -0.090   0.54
## Org_Attraction_5_num    0.045    0.027   0.156    0.0055  0.028   0.47
## Org_Attraction_6_num    0.131    0.134   0.228    0.0300  0.155   0.42
##               var.r   med.r
## Org_Attraction_1_num 0.043 -0.011
## Org_Attraction_2_num 0.040  0.082
## Org_Attraction_3_num 0.045  0.064
## Org_Attraction_4_num 0.035  0.064
## Org_Attraction_5_num 0.040  0.064
## Org_Attraction_6_num 0.037  0.082
##
## Item statistics
##               n raw.r std.r  r.cor r.drop mean  sd
## Org_Attraction_1_num 10  0.58  0.53  0.440  0.139  2.9 1.6
## Org_Attraction_2_num 10  0.33  0.31 -0.025 -0.087  3.3 1.4
## Org_Attraction_3_num 10  0.38  0.47  0.284  0.080  2.7 1.1
## Org_Attraction_4_num 10  0.47  0.47  0.335  0.062  3.3 1.4
## Org_Attraction_5_num 10  0.39  0.37  0.107 -0.048  3.0 1.5
## Org_Attraction_6_num 10  0.26  0.27 -0.131 -0.140  3.3 1.3
##
## Non missing response frequency for each item
##               1  2  3  4  5 miss
## Org_Attraction_1_num 0.3 0.1 0.2 0.2 0.2  0
## Org_Attraction_2_num 0.1 0.2 0.3 0.1 0.3  0
## Org_Attraction_3_num 0.1 0.4 0.2 0.3 0.0  0
## Org_Attraction_4_num 0.1 0.2 0.3 0.1 0.3  0
## Org_Attraction_5_num 0.2 0.2 0.2 0.2 0.2  0
## Org_Attraction_6_num 0.2 0.0 0.2 0.5 0.1  0
```

```
#####Trust#####
alpha(JAR_Social_Invitees %>%
  dplyr::select(Trust_1_num:Trust_4_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Trust_1_num:Trust_4_num)): Some items were negatively
## should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Trust_2_num ) were negatively correlated with the total scale and
## probably should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Trust_1_num:Trust_4_num))
##
##   raw_alpha std.alpha G6(smc) average_r  S/N  ase mean  sd median_r
##     0.26      0.26     0.35     0.081 0.35 0.37  3.2 0.89   -0.046
##
##   95% confidence boundaries
```

```
##           lower alpha upper
## Feldt    -0.93  0.26  0.79
## Duhachek -0.46  0.26  0.99
##
## Reliability if an item is dropped:
##           raw_alpha std.alpha G6(smc) average_r   S/N alpha se  var.r  med.r
## Trust_1_num    -0.27    -0.28   -0.17   -0.079 -0.22    0.68 0.0017 -0.096
## Trust_2_num     0.49     0.50    0.50    0.250  1.00    0.28 0.0707  0.286
## Trust_3_num     0.10     0.11    0.13    0.039  0.12    0.49 0.0464 -0.060
## Trust_4_num     0.29     0.28    0.32    0.113  0.38    0.38 0.1100 -0.060
##
## Item statistics
##           n raw.r std.r r.cor r.drop mean  sd
## Trust_1_num 10  0.77  0.77  0.80  0.484  3.2 1.5
## Trust_2_num 10  0.30  0.33 -0.18 -0.126  3.6 1.5
## Trust_3_num 10  0.68  0.61  0.49  0.202  2.9 1.9
## Trust_4_num 10  0.47  0.51  0.21  0.063  3.2 1.5
##
## Non missing response frequency for each item
##           1  2  3  4  5 miss
## Trust_1_num 0.2 0.1 0.2 0.3 0.2  0
## Trust_2_num 0.1 0.2 0.1 0.2 0.4  0
## Trust_3_num 0.3 0.3 0.0 0.0 0.4  0
## Trust_4_num 0.2 0.1 0.2 0.3 0.2  0
```

```
#####Communication#####
alpha(JAR_Social_Invitees %>%
  dplyr::select(Communication_1_num:Communication_5_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Communication_1_num:Communication_5_num)): Some
## should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Communication_1_num Communication_5_num ) were negatively correlated with the total sca
## probably should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Communication_1_num:Communication_5_num))
##
##           raw_alpha std.alpha G6(smc) average_r   S/N ase mean  sd median_r
##           -0.12    -0.13    0.25   -0.024 -0.12 0.55  2.8 0.71  -0.002
##
##           95% confidence boundaries
##           lower alpha upper
## Feldt    -1.78 -0.12  0.68
## Duhachek -1.20 -0.12  0.97
##
## Reliability if an item is dropped:
##           raw_alpha std.alpha G6(smc) average_r   S/N alpha se
## Communication_1_num    0.190    0.226  4.3e-01    0.0680  0.292  0.43
## Communication_2_num   -1.005   -1.036 -3.6e-01   -0.1457 -0.509  1.06
```

```
## Communication_3_num      0.068      0.067 3.1e-01      0.0176 0.072      0.48
## Communication_4_num     -0.266     -0.311 6.2e-05     -0.0630 -0.237      0.63
## Communication_5_num      0.070      0.013 2.6e-01      0.0033 0.013      0.45
##              var.r  med.r
## Communication_1_num 0.115 0.147
## Communication_2_num 0.117 -0.318
## Communication_3_num 0.090 0.057
## Communication_4_num 0.077 -0.002
## Communication_5_num 0.123 -0.002
##
## Item statistics
##              n raw.r std.r  r.cor r.drop mean  sd
## Communication_1_num 10 0.13 0.17 -0.394 -0.283 3.0 1.5
## Communication_2_num 10 0.77 0.77 0.898 0.445 2.2 1.6
## Communication_3_num 10 0.31 0.31 -0.027 -0.159 3.0 1.6
## Communication_4_num 10 0.53 0.54 0.532 0.063 2.7 1.7
## Communication_5_num 10 0.39 0.35 0.054 -0.143 3.1 1.9
##
## Non missing response frequency for each item
##              1 2 3 4 5 miss
## Communication_1_num 0.2 0.2 0.2 0.2 0.2 0
## Communication_2_num 0.5 0.2 0.1 0.0 0.2 0
## Communication_3_num 0.2 0.3 0.1 0.1 0.3 0
## Communication_4_num 0.4 0.1 0.1 0.2 0.2 0
## Communication_5_num 0.3 0.2 0.0 0.1 0.4 0
```

```
#####Chance to Perform#####
alpha(JAR_Social_Invitees %>%
  dplyr::select(Chance_Perform_1_num:Chance_Perform_4_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Chance_Perform_1_num:Chance_Perform_4_num)):
## should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Chance_Perform_3_num Chance_Perform_4_num ) were negatively correlated with the total s
## probably should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Warning in sqrt(Vtc): NaNs produced
```

```
##
```

```
## Reliability analysis
```

```
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Chance_Perform_1_num:Chance_Perform_4_num))
##
```

```
##      raw_alpha std.alpha G6(smc) average_r  S/N ase mean  sd median_r
##      -0.23      -0.39      -0.15      -0.075 -0.28 0.6    3 0.7      -0.13
```

```
##
```

```
##      95% confidence boundaries
```

```
##              lower alpha upper
```

```
## Feldt      -2.23 -0.23 0.66
```

```
## Duhachek -1.39 -0.23 0.94
```

```
##
```

```
## Reliability if an item is dropped:
```

```
##               raw_alpha std.alpha G6(smc) average_r    S/N alpha se
## Chance_Perform_1_num   -0.2993   -0.350  -0.194  -0.0945 -0.259    0.68
## Chance_Perform_2_num   -0.7739   -0.719  -0.362  -0.1620 -0.418    0.96
## Chance_Perform_3_num    0.1017   -0.114   0.028  -0.0354 -0.102    0.41
## Chance_Perform_4_num    0.0041   -0.025   0.091  -0.0083 -0.025    0.53
##               var.r    med.r
## Chance_Perform_1_num 0.0098 -0.062
## Chance_Perform_2_num 0.0170 -0.204
## Chance_Perform_3_num 0.0861 -0.204
## Chance_Perform_4_num 0.0833 -0.062
##
## Item statistics
##               n raw.r std.r r.cor r.drop mean    sd
## Chance_Perform_1_num 10  0.61  0.47   NaN -0.042  3.5 1.78
## Chance_Perform_2_num 10  0.68  0.59   NaN  0.125  2.4 1.65
## Chance_Perform_3_num 10  0.34  0.37   NaN -0.220  3.1 1.52
## Chance_Perform_4_num 10  0.07  0.33   NaN -0.253  3.2 0.92
##
## Non missing response frequency for each item
##               1    2    3    4    5 miss
## Chance_Perform_1_num 0.2 0.2 0.0 0.1 0.5    0
## Chance_Perform_2_num 0.4 0.3 0.0 0.1 0.2    0
## Chance_Perform_3_num 0.1 0.4 0.1 0.1 0.3    0
## Chance_Perform_4_num 0.0 0.2 0.5 0.2 0.1    0
```

```
#####Consistency AI#####
alpha(JAR_Social_Invitees %>%
      dplyr::select(Consistency_AI_1_num:Consistency_AI_3_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Consistency_AI_1_num:Consistency_AI_3_num)):
## should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Consistency_AI_2_num ) were negatively correlated with the total scale and
## probably should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Consistency_AI_1_num:Consistency_AI_3_num))
##
##   raw_alpha std.alpha G6(smc) average_r    S/N ase mean    sd median_r
##   -0.039    -0.029    0.11  -0.0094 -0.028 0.57  2.9 0.81    -0.11
##
##   95% confidence boundaries
##         lower alpha upper
## Feldt    -2.04 -0.04  0.72
## Duhachek -1.16 -0.04  1.08
##
## Reliability if an item is dropped:
##               raw_alpha std.alpha G6(smc) average_r    S/N alpha se var.r
## Consistency_AI_1_num   -0.26    -0.26  -0.11    -0.11 -0.21    0.8   NA
## Consistency_AI_2_num    0.52     0.53   0.36     0.36  1.11    0.3   NA
```

```
## Consistency_AI_3_num      -0.74      -0.74   -0.27      -0.27 -0.43      1.1      NA
##                               med.r
## Consistency_AI_1_num -0.11
## Consistency_AI_2_num  0.36
## Consistency_AI_3_num -0.27
##
## Item statistics
##               n raw.r std.r r.cor r.drop mean  sd
## Consistency_AI_1_num 10  0.65  0.63  0.47  0.052  2.7 1.5
## Consistency_AI_2_num 10  0.36  0.36 -0.55 -0.238  3.6 1.4
## Consistency_AI_3_num 10  0.71  0.72  0.65  0.209  2.5 1.4
##
## Non missing response frequency for each item
##               1  2  3  4  5 miss
## Consistency_AI_1_num 0.3 0.2 0.1 0.3 0.1  0
## Consistency_AI_2_num 0.1 0.1 0.3 0.1 0.4  0
## Consistency_AI_3_num 0.3 0.2 0.3 0.1 0.1  0
```

```
#####Consistency HR#####
alpha(JAR_Social_Invitees %>%
      dplyr::select(Consistency_HR_1_num:Consistency_HR_3_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Consistency_HR_1_num:Consistency_HR_3_num)):
## should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Consistency_HR_3_num ) were negatively correlated with the total scale and
## probably should be reversed.
```

```
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Warning in sqrt(Vtc): NaNs produced
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Consistency_HR_1_num:Consistency_HR_3_num))
##
##   raw_alpha std.alpha G6(smc) average_r   S/N ase mean   sd median_r
##      -0.4      -0.55 -0.061      -0.13 -0.35 0.71  2.8 0.72  1.9e-20
##
##   95% confidence boundaries
##           lower alpha upper
## Feldt      -3.09  -0.4  0.62
## Duhachek -1.78  -0.4  0.99
##
## Reliability if an item is dropped:
##           raw_alpha std.alpha G6(smc) average_r   S/N alpha se var.r
## Consistency_HR_1_num      -1.84      -2.44   -0.55      -0.55 -0.71  1.49  NA
## Consistency_HR_2_num       0.00       0.00    0.00       0.00  0.00  0.63  NA
## Consistency_HR_3_num       0.23       0.26    0.15       0.15  0.35  0.43  NA
##
##                               med.r
## Consistency_HR_1_num -5.5e-01
## Consistency_HR_2_num  1.9e-20
## Consistency_HR_3_num  1.5e-01
```



```
##
## Item statistics
##           n raw.r std.r r.cor r.drop mean  sd
## Consistency_HR_1_num 10  0.78 0.78  NaN  0.10  3.0 1.56
## Consistency_HR_2_num 10  0.13 0.40  NaN -0.29  2.3 0.95
## Consistency_HR_3_num 10  0.51 0.30  NaN -0.27  3.2 1.62
##
## Non missing response frequency for each item
##           1  2  3  4  5 miss
## Consistency_HR_1_num 0.2 0.3 0.0 0.3 0.2  0
## Consistency_HR_2_num 0.2 0.4 0.3 0.1 0.0  0
## Consistency_HR_3_num 0.2 0.2 0.1 0.2 0.3  0
```

```
#####Intent to Engage#####
alpha(JAR_Social_Invitees %>%
  dplyr::select(Intent_Engag_1_num:Intent_Engag_4_num))
```

```
## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(Intent_Engag_1_num:Intent_Engag_4_num)): Some
## should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
## Some items ( Intent_Engag_3_num ) were negatively correlated with the total scale and
## probably should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(Intent_Engag_1_num:Intent_Engag_4_num))
##
##   raw_alpha std.alpha G6(smc) average_r S/N ase mean  sd median_r
##         0.5         0.5      0.8         0.2  1 0.27   3 0.82   0.28
##
##   95% confidence boundaries
##         lower alpha upper
## Feldt    -0.33   0.5  0.86
## Duhachek -0.03   0.5  1.02
##
## Reliability if an item is dropped:
##           raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r
## Intent_Engag_1_num    0.46    0.46   0.38    0.223 0.86   0.29 0.010
## Intent_Engag_2_num    0.29    0.28   0.68    0.115 0.39   0.39 0.254
## Intent_Engag_3_num    0.67    0.68   0.61    0.413 2.11   0.18 0.019
## Intent_Engag_4_num    0.12    0.13   0.46    0.048 0.15   0.48 0.226
##
##           med.r
## Intent_Engag_1_num 0.25
## Intent_Engag_2_num 0.30
## Intent_Engag_3_num 0.49
## Intent_Engag_4_num 0.11
##
## Item statistics
##           n raw.r std.r r.cor r.drop mean  sd
## Intent_Engag_1_num 10  0.57 0.60  0.59 0.2442  3.4 1.2
## Intent_Engag_2_num 10  0.73 0.73  0.60 0.4259  2.8 1.3
```

```

## Intent_Engag_3_num 10 0.42 0.38 0.30 0.0055 2.9 1.4
## Intent_Engag_4_num 10 0.81 0.81 0.77 0.5720 3.0 1.3
##
## Non missing response frequency for each item
##      1  2  3  4  5 miss
## Intent_Engag_1_num 0.0 0.3 0.2 0.3 0.2 0
## Intent_Engag_2_num 0.2 0.2 0.3 0.2 0.1 0
## Intent_Engag_3_num 0.2 0.2 0.2 0.3 0.1 0
## Intent_Engag_4_num 0.1 0.3 0.3 0.1 0.2 0

#####AI Knowledge#####
alpha(JAR_Social_Invitees %>%
      dplyr::select(AI_Knowledge_Experei_1_num:AI_Knowledge_Experei_5_num))

## Warning in cor.smooth(r): Matrix was not positive definite, smoothing was done

## In factor.stats, I could not find the RMSEA upper bound . Sorry about that

## Warning in alpha(JAR_Social_Invitees %>% dplyr::select(AI_Knowledge_Experei_1_num:AI_Knowledge_Experei_5_num))
## should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option

## Some items ( AI_Knowledge_Experei_2_num AI_Knowledge_Experei_5_num ) were negatively correlated with
## probably should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option

## In smc, smcs > 1 were set to 1.0

## In smc, smcs < 0 were set to .0

## In smc, smcs > 1 were set to 1.0

## In smc, smcs < 0 were set to .0

## In smc, smcs > 1 were set to 1.0
## In smc, smcs > 1 were set to 1.0
## In smc, smcs > 1 were set to 1.0
## In smc, smcs > 1 were set to 1.0

## In smc, smcs < 0 were set to .0

##
## Reliability analysis
## Call: alpha(x = JAR_Social_Invitees %>% dplyr::select(AI_Knowledge_Experei_1_num:AI_Knowledge_Experei_5_num))
##
##      raw_alpha std.alpha G6(smc) average_r   S/N ase mean   sd median_r
##      -0.77      -0.84   0.048      -0.1 -0.46 0.8   2.7 0.47   -0.058
##
##      95% confidence boundaries
##      lower alpha upper
## Feldt      -3.41 -0.77   0.5

```

```
## Duhachek -2.33 -0.77 0.8
##
## Reliability if an item is dropped:
##
## raw_alpha std.alpha G6(smc) average_r S/N alpha se
## AI_Knowledge_Experei_1_num -0.53 -1.03 0.29 -0.145 -0.51 0.64
## AI_Knowledge_Experei_2_num -3.34 -1.81 0.47 -0.192 -0.64 0.57
## AI_Knowledge_Experei_3_num -0.19 -0.34 0.88 -0.068 -0.25 0.56
## AI_Knowledge_Experei_4_num -2.20 -3.46 -0.51 -0.241 -0.78 1.46
## AI_Knowledge_Experei_5_num 0.28 0.40 1.00 0.143 0.67 0.39
##
## var.r med.r
## AI_Knowledge_Experei_1_num 0.35 -0.124
## AI_Knowledge_Experei_2_num 0.44 -0.218
## AI_Knowledge_Experei_3_num 0.48 0.067
## AI_Knowledge_Experei_4_num 0.26 -0.385
## AI_Knowledge_Experei_5_num 0.20 0.247
##
## Item statistics
##
## n raw.r std.r r.cor r.drop mean sd
## AI_Knowledge_Experei_1_num 10 0.382 0.50 1.43 -0.23 3.0 1.05
## AI_Knowledge_Experei_2_num 8 0.851 0.66 0.39 0.20 2.8 1.49
## AI_Knowledge_Experei_3_num 8 0.013 0.23 -0.32 -0.42 2.9 0.99
## AI_Knowledge_Experei_4_num 8 0.740 0.83 3.81 0.35 2.1 0.99
## AI_Knowledge_Experei_5_num 9 -0.169 -0.50 -4.93 -0.64 2.8 1.20
##
## Non missing response frequency for each item
##
## 1 2 3 4 miss
## AI_Knowledge_Experei_1_num 0.10 0.20 0.30 0.40 0.0
## AI_Knowledge_Experei_2_num 0.38 0.00 0.12 0.50 0.2
## AI_Knowledge_Experei_3_num 0.12 0.12 0.50 0.25 0.2
## AI_Knowledge_Experei_4_num 0.25 0.50 0.12 0.12 0.2
## AI_Knowledge_Experei_5_num 0.22 0.11 0.33 0.33 0.1
```

here we are checking for multi-collinearity

```
#correlation (0.6 start worrying or 0.7 should be worried)

# Selecting relevant variables for correlation analysis
Corr_JAR_Social_Invitees_data <- JAR_Social_Invitees %>%
  dplyr::select(c(Condition, Org_Attraction_1_num:Consistency_HR_3_num, Intent_Engag_1_num:AI_Knowledge_1_num))
  mutate(Condition = as.numeric(factor(Condition,
levels = c(1, 2),
labels = c(1, 2))))

#correlation for main variables
Corr_JAR_Social_Invitees_data %>%
  dplyr::select(c(Condition, Org_Attraction:Intent)) %>%
  tab_corr(corr.method = c("pearson"), digits = 2, title = "Correlation for main variables")
```

Correlation for main variables

Condition

Org_Attraction
Trust
Communication
Chance_Perform
Consistency
Intent
Condition

0.06

-0.46

-0.36

0.02

-0.15

0.17

Org_Attraction

0.06

-0.23

-0.61

0.59

-0.58

-0.09

Trust

-0.46

-0.23

0.41

-0.07

-0.26

-0.27

Communication

-0.36

-0.61

0.41

-0.26

0.67*
-0.05
Chance_Perform
0.02
0.59
-0.07
-0.26

-0.16
-0.03
Consistency
-0.15
-0.58
-0.26
0.67*
-0.16

0.05
Intent
0.17
-0.09
-0.27
-0.05
-0.03
0.05

Computed correlation used pearson-method with listwise-deletion.

```
#Groupwise correlation for main variables for AI condition  
JAR_Social_Invitees %>%  
  filter(Condition == 1) %>%  
  dplyr::select(c(Org_Attraction:Intent)) %>%  
  tab_corr(corr.method = c("pearson"), digits = 2, title = "Groupwise correlation for main variables for
```

Groupwise correlation for main variables for AI condition

Org_Attraction
Trust
Communication

Chance_Perform

Consistency

Intent

Org_Attraction

-0.05

-0.52

0.69

-0.44

-0.33

Trust

-0.05

-0.46

-0.11

-0.55

-0.31

Communication

-0.52

-0.46

-0.36

0.96**

0.06

Chance_Perform

0.69

-0.11

-0.36

-0.12

-0.32

Consistency

-0.44

-0.55

0.96**

-0.12

0.04
Intent
-0.33
-0.31
0.06
-0.32
0.04

Computed correlation used pearson-method with listwise-deletion.

```
#Groupwise correlation for main variables for HR condition  
JAR_Social_Invitees %>%  
  filter(Condition == 2) %>%  
  dplyr::select(c(Org_Attraction:Intent)) %>%  
  tab_corr(corr.method = c("pearson"), digits = 2, title = "Groupwise correlation for main variables for
```

Groupwise correlation for main variables for HR condition

Org_Attraction
Trust
Communication
Chance_Perform
Consistency
Intent
Org_Attraction

-0.37
-0.75
0.51
-0.81
0.14
Trust
-0.37

0.88
0.00
-0.21
-0.14

Communication

-0.75

0.88

-0.13

0.28

-0.03

Chance_Perform

0.51

0.00

-0.13

-0.35

0.88

Consistency

-0.81

-0.21

0.28

-0.35

0.14

Intent

0.14

-0.14

-0.03

0.88

0.14

Computed correlation used pearson-method with listwise-deletion.

```
#Checking multi-collinearity: correlation with everything in the model  
cor_matrix <- cor(Corr_JAR_Social_Invitees_data, use = "pairwise.complete.obs")  
print(round(cor_matrix, 2))
```

##	Condition	Org_Attraction_1_num	Org_Attraction_2_num
## Condition	1.00	-0.08	0.58
## Org_Attraction_1_num	-0.08	1.00	0.06
## Org_Attraction_2_num	0.58	0.06	1.00
## Org_Attraction_3_num	0.04	0.18	-0.23
## Org_Attraction_4_num	0.12	0.21	0.17

## Org_Attraction_5_num	-0.58	0.09	-0.32
## Org_Attraction_6_num	0.13	-0.24	0.06
## Trust_1_num	-0.41	-0.37	-0.56
## Trust_2_num	-0.34	-0.16	-0.09
## Trust_3_num	0.05	-0.23	-0.07
## Trust_4_num	-0.41	0.10	-0.78
## Communication_1_num	-0.29	0.14	-0.53
## Communication_2_num	-0.24	-0.46	0.02
## Communication_3_num	0.53	-0.17	0.34
## Communication_4_num	-0.23	0.27	0.27
## Communication_5_num	-0.51	-0.37	-0.82
## Chance_Perform_1_num	0.12	-0.29	-0.11
## Chance_Perform_2_num	-0.47	-0.19	-0.10
## Chance_Perform_3_num	0.08	0.23	0.55
## Chance_Perform_4_num	0.52	0.02	0.29
## Consistency_AI_1_num	-0.26	0.03	0.05
## Consistency_AI_2_num	-0.36	0.03	-0.15
## Consistency_AI_3_num	0.16	0.18	0.55
## Consistency_HR_1_num	0.41	-0.45	0.25
## Consistency_HR_2_num	0.41	-0.42	0.26
## Consistency_HR_3_num	-0.50	-0.12	-0.46
## Intent_Engag_1_num	0.07	-0.15	-0.21
## Intent_Engag_2_num	0.46	-0.59	0.15
## Intent_Engag_3_num	0.22	-0.01	0.70
## Intent_Engag_4_num	-0.32	-0.10	0.00
## AI_Knowledge_Experei_1_num	0.00	0.33	0.22
## AI_Knowledge_Experei_2_num	-0.42	-0.41	-0.73
## AI_Knowledge_Experei_3_num	-0.13	0.42	0.62
## AI_Knowledge_Experei_4_num	-0.13	-0.36	0.04
## AI_Knowledge_Experei_5_num	0.14	0.02	-0.29
## Attention_AI_score	-0.73	-0.03	-0.58
## Org_Attraction	0.06	0.58	0.33
## Trust	-0.46	-0.30	-0.63
## Communication	-0.36	-0.29	-0.35
## Chance_Perform	0.02	-0.17	0.26
## Consistency	-0.15	-0.30	0.15
## Intent	0.17	-0.34	0.28
## AI_Knowledge	-0.24	-0.13	-0.18
##	Org_Attraction_3_num	Org_Attraction_4_num	
## Condition	0.04	0.12	
## Org_Attraction_1_num	0.18	0.21	
## Org_Attraction_2_num	-0.23	0.17	
## Org_Attraction_3_num	1.00	0.21	
## Org_Attraction_4_num	0.21	1.00	
## Org_Attraction_5_num	0.07	-0.16	
## Org_Attraction_6_num	-0.09	-0.29	
## Trust_1_num	0.18	-0.46	
## Trust_2_num	0.13	0.79	
## Trust_3_num	0.55	-0.16	
## Trust_4_num	0.26	-0.30	
## Communication_1_num	0.14	-0.53	
## Communication_2_num	-0.28	-0.56	
## Communication_3_num	0.32	0.43	
## Communication_4_num	0.01	0.32	

## Communication_5_num	0.19	-0.27
## Chance_Perform_1_num	0.68	0.55
## Chance_Perform_2_num	0.20	-0.06
## Chance_Perform_3_num	-0.05	-0.07
## Chance_Perform_4_num	-0.50	0.29
## Consistency_AI_1_num	0.01	0.47
## Consistency_AI_2_num	-0.60	-0.70
## Consistency_AI_3_num	0.04	-0.03
## Consistency_HR_1_num	-0.07	0.30
## Consistency_HR_2_num	-0.34	0.01
## Consistency_HR_3_num	-0.03	-0.61
## Intent_Engag_1_num	-0.16	0.19
## Intent_Engag_2_num	-0.21	0.15
## Intent_Engag_3_num	-0.25	-0.10
## Intent_Engag_4_num	-0.55	-0.06
## AI_Knowledge_Experei_1_num	-0.40	0.22
## AI_Knowledge_Experei_2_num	0.18	-0.48
## AI_Knowledge_Experei_3_num	-0.47	0.10
## AI_Knowledge_Experei_4_num	-0.34	0.15
## AI_Knowledge_Experei_5_num	0.62	-0.23
## Attention_AI_score	-0.37	-0.30
## Org_Attraction	0.38	0.47
## Trust	0.52	-0.06
## Communication	0.18	-0.26
## Chance_Perform	0.36	0.38
## Consistency	-0.40	-0.27
## Intent	-0.47	0.06
## AI_Knowledge	-0.13	-0.21
##	Org_Attraction_5_num	Org_Attraction_6_num
## Condition	-0.58	0.13
## Org_Attraction_1_num	0.09	-0.24
## Org_Attraction_2_num	-0.32	0.06
## Org_Attraction_3_num	0.07	-0.09
## Org_Attraction_4_num	-0.16	-0.29
## Org_Attraction_5_num	1.00	0.22
## Org_Attraction_6_num	0.22	1.00
## Trust_1_num	0.15	0.59
## Trust_2_num	0.10	-0.21
## Trust_3_num	-0.24	0.19
## Trust_4_num	0.10	-0.43
## Communication_1_num	0.45	0.33
## Communication_2_num	-0.23	-0.29
## Communication_3_num	-0.32	-0.56
## Communication_4_num	-0.18	-0.69
## Communication_5_num	0.08	-0.01
## Chance_Perform_1_num	0.13	0.12
## Chance_Perform_2_num	0.81	-0.01
## Chance_Perform_3_num	-0.10	0.31
## Chance_Perform_4_num	-0.16	-0.05
## Consistency_AI_1_num	-0.10	-0.73
## Consistency_AI_2_num	0.31	-0.05
## Consistency_AI_3_num	0.17	-0.09
## Consistency_HR_1_num	-0.57	-0.27
## Consistency_HR_2_num	-0.16	-0.08

## Consistency_HR_3_num	0.23	-0.13		
## Intent_Engag_1_num	0.13	-0.08		
## Intent_Engag_2_num	-0.62	0.04		
## Intent_Engag_3_num	-0.22	0.50		
## Intent_Engag_4_num	0.28	0.44		
## AI_Knowledge_Experei_1_num	-0.07	-0.55		
## AI_Knowledge_Experei_2_num	-0.10	0.21		
## AI_Knowledge_Experei_3_num	-0.12	-0.18		
## AI_Knowledge_Experei_4_num	-0.47	-0.66		
## AI_Knowledge_Experei_5_num	0.07	0.63		
## Attention_AI_score	0.09	-0.01		
## Org_Attraction	0.39	0.26		
## Trust	0.02	0.08		
## Communication	-0.10	-0.58		
## Chance_Perform	0.46	0.22		
## Consistency	-0.05	-0.62		
## Intent	-0.18	0.37		
## AI_Knowledge	-0.20	-0.02		
##	Trust_1_num	Trust_2_num	Trust_3_num	Trust_4_num
## Condition	-0.41	-0.34	0.05	-0.41
## Org_Attraction_1_num	-0.37	-0.16	-0.23	0.10
## Org_Attraction_2_num	-0.56	-0.09	-0.07	-0.78
## Org_Attraction_3_num	0.18	0.13	0.55	0.26
## Org_Attraction_4_num	-0.46	0.79	-0.16	-0.30
## Org_Attraction_5_num	0.15	0.10	-0.24	0.10
## Org_Attraction_6_num	0.59	-0.21	0.19	-0.43
## Trust_1_num	1.00	-0.06	0.50	0.29
## Trust_2_num	-0.06	1.00	-0.10	-0.11
## Trust_3_num	0.50	-0.10	1.00	-0.03
## Trust_4_num	0.29	-0.11	-0.03	1.00
## Communication_1_num	0.45	-0.50	0.36	0.25
## Communication_2_num	0.17	-0.24	0.23	0.17
## Communication_3_num	-0.65	0.18	0.04	-0.14
## Communication_4_num	-0.50	0.25	-0.01	-0.15
## Communication_5_num	0.72	0.18	0.17	0.76
## Chance_Perform_1_num	0.13	0.58	0.42	-0.13
## Chance_Perform_2_num	0.01	0.25	-0.09	-0.04
## Chance_Perform_3_num	0.04	-0.08	-0.15	-0.36
## Chance_Perform_4_num	-0.61	-0.02	-0.77	-0.28
## Consistency_AI_1_num	-0.42	0.53	0.11	0.08
## Consistency_AI_2_num	0.04	-0.50	-0.52	0.31
## Consistency_AI_3_num	-0.44	-0.22	0.29	-0.50
## Consistency_HR_1_num	-0.24	0.28	-0.23	-0.10
## Consistency_HR_2_num	-0.29	0.02	-0.11	-0.05
## Consistency_HR_3_num	0.35	-0.33	-0.03	0.45
## Intent_Engag_1_num	-0.24	0.04	-0.03	-0.18
## Intent_Engag_2_num	-0.03	0.07	0.22	-0.38
## Intent_Engag_3_num	0.07	-0.13	0.04	-0.70
## Intent_Engag_4_num	0.17	0.06	-0.09	-0.51
## AI_Knowledge_Experei_1_num	-0.71	0.00	-0.63	-0.14
## AI_Knowledge_Experei_2_num	0.89	-0.08	0.38	0.53
## AI_Knowledge_Experei_3_num	-0.42	0.12	-0.40	-0.27
## AI_Knowledge_Experei_4_num	-0.28	0.31	-0.32	0.08
## AI_Knowledge_Experei_5_num	0.67	-0.35	0.75	0.03

## Attention_AI_score	0.51	0.07	-0.17	0.42
## Org_Attraction	-0.24	0.22	-0.06	-0.44
## Trust	0.77	0.30	0.68	0.47
## Communication	0.11	-0.02	0.35	0.44
## Chance_Perform	-0.09	0.47	-0.13	-0.39
## Consistency	-0.39	-0.09	-0.22	0.13
## Intent	0.00	0.01	0.06	-0.72
## AI_Knowledge	0.27	-0.06	-0.12	0.08
##	Communication_1_num	Communication_2_num		
## Condition	-0.29	-0.24		
## Org_Attraction_1_num	0.14	-0.46		
## Org_Attraction_2_num	-0.53	0.02		
## Org_Attraction_3_num	0.14	-0.28		
## Org_Attraction_4_num	-0.53	-0.56		
## Org_Attraction_5_num	0.45	-0.23		
## Org_Attraction_6_num	0.33	-0.29		
## Trust_1_num	0.45	0.17		
## Trust_2_num	-0.50	-0.24		
## Trust_3_num	0.36	0.23		
## Trust_4_num	0.25	0.17		
## Communication_1_num	1.00	-0.05		
## Communication_2_num	-0.05	1.00		
## Communication_3_num	-0.46	0.04		
## Communication_4_num	-0.31	0.39		
## Communication_5_num	0.16	0.25		
## Chance_Perform_1_num	-0.08	-0.42		
## Chance_Perform_2_num	0.14	0.09		
## Chance_Perform_3_num	-0.44	-0.01		
## Chance_Perform_4_num	-0.41	-0.33		
## Consistency_AI_1_num	-0.25	0.21		
## Consistency_AI_2_num	0.21	0.47		
## Consistency_AI_3_num	0.22	0.10		
## Consistency_HR_1_num	-0.81	0.22		
## Consistency_HR_2_num	-0.24	0.03		
## Consistency_HR_3_num	0.32	0.62		
## Intent_Engag_1_num	0.38	-0.11		
## Intent_Engag_2_num	-0.23	0.28		
## Intent_Engag_3_num	-0.33	0.11		
## Intent_Engag_4_num	0.34	0.05		
## AI_Knowledge_Experei_1_num	-0.28	0.13		
## AI_Knowledge_Experei_2_num	0.24	0.45		
## AI_Knowledge_Experei_3_num	-0.54	0.13		
## AI_Knowledge_Experei_4_num	-0.68	0.62		
## AI_Knowledge_Experei_5_num	0.59	-0.33		
## Attention_AI_score	0.18	0.38		
## Org_Attraction	0.00	-0.75		
## Trust	0.27	0.16		
## Communication	0.13	0.77		
## Chance_Perform	-0.35	-0.33		
## Consistency	-0.23	0.78		
## Intent	0.05	0.14		
## AI_Knowledge	-0.14	0.45		
##	Communication_3_num	Communication_4_num		
## Condition	0.53	-0.23		

## Org_Attraction_1_num	-0.17	0.27
## Org_Attraction_2_num	0.34	0.27
## Org_Attraction_3_num	0.32	0.01
## Org_Attraction_4_num	0.43	0.32
## Org_Attraction_5_num	-0.32	-0.18
## Org_Attraction_6_num	-0.56	-0.69
## Trust_1_num	-0.65	-0.50
## Trust_2_num	0.18	0.25
## Trust_3_num	0.04	-0.01
## Trust_4_num	-0.14	-0.15
## Communication_1_num	-0.46	-0.31
## Communication_2_num	0.04	0.39
## Communication_3_num	1.00	0.40
## Communication_4_num	0.40	1.00
## Communication_5_num	-0.33	-0.34
## Chance_Perform_1_num	0.42	-0.16
## Chance_Perform_2_num	0.12	0.17
## Chance_Perform_3_num	-0.27	0.06
## Chance_Perform_4_num	0.30	-0.10
## Consistency_AI_1_num	0.41	0.62
## Consistency_AI_2_num	-0.43	-0.01
## Consistency_AI_3_num	0.30	0.41
## Consistency_HR_1_num	0.57	0.17
## Consistency_HR_2_num	0.29	-0.35
## Consistency_HR_3_num	-0.21	0.23
## Intent_Engag_1_num	0.29	0.12
## Intent_Engag_2_num	0.41	0.12
## Intent_Engag_3_num	-0.25	0.08
## Intent_Engag_4_num	-0.46	0.10
## AI_Knowledge_Experei_1_num	0.32	0.68
## AI_Knowledge_Experei_2_num	-0.39	-0.19
## AI_Knowledge_Experei_3_num	-0.29	0.42
## AI_Knowledge_Experei_4_num	0.42	0.65
## AI_Knowledge_Experei_5_num	-0.29	-0.57
## Attention_AI_score	-0.66	0.09
## Org_Attraction	-0.02	0.03
## Trust	-0.23	-0.17
## Communication	0.31	0.53
## Chance_Perform	0.29	-0.01
## Consistency	0.38	0.55
## Intent	-0.02	0.17
## AI_Knowledge	-0.15	0.31
##	Communication_5_num	Chance_Perform_1_num
## Condition	-0.51	0.12
## Org_Attraction_1_num	-0.37	-0.29
## Org_Attraction_2_num	-0.82	-0.11
## Org_Attraction_3_num	0.19	0.68
## Org_Attraction_4_num	-0.27	0.55
## Org_Attraction_5_num	0.08	0.13
## Org_Attraction_6_num	-0.01	0.12
## Trust_1_num	0.72	0.13
## Trust_2_num	0.18	0.58
## Trust_3_num	0.17	0.42
## Trust_4_num	0.76	-0.13

## Communication_1_num	0.16	-0.08
## Communication_2_num	0.25	-0.42
## Communication_3_num	-0.33	0.42
## Communication_4_num	-0.34	-0.16
## Communication_5_num	1.00	0.12
## Chance_Perform_1_num	0.12	1.00
## Chance_Perform_2_num	0.02	0.30
## Chance_Perform_3_num	-0.24	-0.27
## Chance_Perform_4_num	-0.40	-0.20
## Consistency_AI_1_num	-0.07	0.19
## Consistency_AI_2_num	0.14	-0.83
## Consistency_AI_3_num	-0.73	0.02
## Consistency_HR_1_num	0.08	0.12
## Consistency_HR_2_num	-0.08	0.10
## Consistency_HR_3_num	0.44	-0.42
## Intent_Engag_1_num	-0.22	0.16
## Intent_Engag_2_num	-0.08	0.14
## Intent_Engag_3_num	-0.43	-0.25
## Intent_Engag_4_num	-0.22	-0.23
## AI_Knowledge_Experei_1_num	-0.46	-0.47
## AI_Knowledge_Experei_2_num	0.92	-0.04
## AI_Knowledge_Experei_3_num	-0.37	-0.55
## AI_Knowledge_Experei_4_num	0.17	-0.27
## AI_Knowledge_Experei_5_num	0.21	0.48
## Attention_AI_score	0.60	-0.49
## Org_Attraction	-0.54	0.36
## Trust	0.78	0.47
## Communication	0.39	-0.05
## Chance_Perform	-0.18	0.61
## Consistency	-0.04	-0.40
## Intent	-0.39	-0.09
## AI_Knowledge	0.28	-0.36
##	Chance_Perform_2_num	Chance_Perform_3_num
## Condition	-0.47	0.08
## Org_Attraction_1_num	-0.19	0.23
## Org_Attraction_2_num	-0.10	0.55
## Org_Attraction_3_num	0.20	-0.05
## Org_Attraction_4_num	-0.06	-0.07
## Org_Attraction_5_num	0.81	-0.10
## Org_Attraction_6_num	-0.01	0.31
## Trust_1_num	0.01	0.04
## Trust_2_num	0.25	-0.08
## Trust_3_num	-0.09	-0.15
## Trust_4_num	-0.04	-0.36
## Communication_1_num	0.14	-0.44
## Communication_2_num	0.09	-0.01
## Communication_3_num	0.12	-0.27
## Communication_4_num	0.17	0.06
## Communication_5_num	0.02	-0.24
## Chance_Perform_1_num	0.30	-0.27
## Chance_Perform_2_num	1.00	-0.06
## Chance_Perform_3_num	-0.06	1.00
## Chance_Perform_4_num	-0.21	-0.02
## Consistency_AI_1_num	0.10	-0.42

## Consistency_AI_2_num	0.17	0.12
## Consistency_AI_3_num	0.30	-0.08
## Consistency_HR_1_num	-0.17	0.19
## Consistency_HR_2_num	-0.16	-0.33
## Consistency_HR_3_num	0.34	-0.01
## Intent_Engag_1_num	0.14	-0.77
## Intent_Engag_2_num	-0.32	-0.27
## Intent_Engag_3_num	-0.13	0.86
## Intent_Engag_4_num	0.15	-0.05
## AI_Knowledge_Experei_1_num	0.06	0.00
## AI_Knowledge_Experei_2_num	-0.03	0.00
## AI_Knowledge_Experei_3_num	-0.17	0.76
## AI_Knowledge_Experei_4_num	-0.01	0.00
## AI_Knowledge_Experei_5_num	-0.12	-0.02
## Attention_AI_score	-0.07	0.03
## Org_Attraction	0.26	0.37
## Trust	0.05	-0.24
## Communication	0.25	-0.41
## Chance_Perform	0.68	0.34
## Consistency	0.29	-0.19
## Intent	-0.07	-0.05
## AI_Knowledge	-0.01	0.33
##	Chance_Perform_4_num	Consistency_AI_1_num
## Condition	0.52	-0.26
## Org_Attraction_1_num	0.02	0.03
## Org_Attraction_2_num	0.29	0.05
## Org_Attraction_3_num	-0.50	0.01
## Org_Attraction_4_num	0.29	0.47
## Org_Attraction_5_num	-0.16	-0.10
## Org_Attraction_6_num	-0.05	-0.73
## Trust_1_num	-0.61	-0.42
## Trust_2_num	-0.02	0.53
## Trust_3_num	-0.77	0.11
## Trust_4_num	-0.28	0.08
## Communication_1_num	-0.41	-0.25
## Communication_2_num	-0.33	0.21
## Communication_3_num	0.30	0.41
## Communication_4_num	-0.10	0.62
## Communication_5_num	-0.40	-0.07
## Chance_Perform_1_num	-0.20	0.19
## Chance_Perform_2_num	-0.21	0.10
## Chance_Perform_3_num	-0.02	-0.42
## Chance_Perform_4_num	1.00	-0.19
## Consistency_AI_1_num	-0.19	1.00
## Consistency_AI_2_num	0.15	-0.27
## Consistency_AI_3_num	-0.18	0.36
## Consistency_HR_1_num	0.46	0.00
## Consistency_HR_2_num	0.31	0.31
## Consistency_HR_3_num	-0.25	-0.25
## Intent_Engag_1_num	0.33	0.20
## Intent_Engag_2_num	0.31	0.02
## Intent_Engag_3_num	0.02	-0.40
## Intent_Engag_4_num	0.09	-0.11
## AI_Knowledge_Experei_1_num	0.57	0.21

## AI_Knowledge_Experei_2_num	-0.45	-0.41
## AI_Knowledge_Experei_3_num	0.06	0.25
## AI_Knowledge_Experei_4_num	0.27	0.41
## AI_Knowledge_Experei_5_num	-0.54	-0.52
## Attention_AI_score	-0.21	-0.08
## Org_Attraction	0.00	-0.10
## Trust	-0.78	0.14
## Communication	-0.44	0.44
## Chance_Perform	0.07	-0.12
## Consistency	0.09	0.46
## Intent	0.29	-0.13
## AI_Knowledge	0.03	-0.35
##	Consistency_AI_2_num	Consistency_AI_3_num
## Condition	-0.36	0.16
## Org_Attraction_1_num	0.03	0.18
## Org_Attraction_2_num	-0.15	0.55
## Org_Attraction_3_num	-0.60	0.04
## Org_Attraction_4_num	-0.70	-0.03
## Org_Attraction_5_num	0.31	0.17
## Org_Attraction_6_num	-0.05	-0.09
## Trust_1_num	0.04	-0.44
## Trust_2_num	-0.50	-0.22
## Trust_3_num	-0.52	0.29
## Trust_4_num	0.31	-0.50
## Communication_1_num	0.21	0.22
## Communication_2_num	0.47	0.10
## Communication_3_num	-0.43	0.30
## Communication_4_num	-0.01	0.41
## Communication_5_num	0.14	-0.73
## Chance_Perform_1_num	-0.83	0.02
## Chance_Perform_2_num	0.17	0.30
## Chance_Perform_3_num	0.12	-0.08
## Chance_Perform_4_num	0.15	-0.18
## Consistency_AI_1_num	-0.27	0.36
## Consistency_AI_2_num	1.00	-0.11
## Consistency_AI_3_num	-0.11	1.00
## Consistency_HR_1_num	-0.15	-0.42
## Consistency_HR_2_num	-0.07	0.13
## Consistency_HR_3_num	0.66	-0.20
## Intent_Engag_1_num	-0.09	0.28
## Intent_Engag_2_num	-0.28	-0.06
## Intent_Engag_3_num	0.03	0.15
## Intent_Engag_4_num	0.23	0.18
## AI_Knowledge_Experei_1_num	0.37	0.16
## AI_Knowledge_Experei_2_num	0.31	-0.64
## AI_Knowledge_Experei_3_num	0.28	0.14
## AI_Knowledge_Experei_4_num	0.25	-0.33
## AI_Knowledge_Experei_5_num	-0.45	-0.01
## Attention_AI_score	0.55	-0.55
## Org_Attraction	-0.41	0.35
## Trust	-0.34	-0.34
## Communication	0.17	0.09
## Chance_Perform	-0.31	0.09
## Consistency	0.51	0.27

## Intent	-0.04	0.21
## AI_Knowledge	0.37	-0.45
##	Consistency_HR_1_num	Consistency_HR_2_num
## Condition	0.41	0.41
## Org_Attraction_1_num	-0.45	-0.42
## Org_Attraction_2_num	0.25	0.26
## Org_Attraction_3_num	-0.07	-0.34
## Org_Attraction_4_num	0.30	0.01
## Org_Attraction_5_num	-0.57	-0.16
## Org_Attraction_6_num	-0.27	-0.08
## Trust_1_num	-0.24	-0.29
## Trust_2_num	0.28	0.02
## Trust_3_num	-0.23	-0.11
## Trust_4_num	-0.10	-0.05
## Communication_1_num	-0.81	-0.24
## Communication_2_num	0.22	0.03
## Communication_3_num	0.57	0.29
## Communication_4_num	0.17	-0.35
## Communication_5_num	0.08	-0.08
## Chance_Perform_1_num	0.12	0.10
## Chance_Perform_2_num	-0.17	-0.16
## Chance_Perform_3_num	0.19	-0.33
## Chance_Perform_4_num	0.46	0.31
## Consistency_AI_1_num	0.00	0.31
## Consistency_AI_2_num	-0.15	-0.07
## Consistency_AI_3_num	-0.42	0.13
## Consistency_HR_1_num	1.00	0.15
## Consistency_HR_2_num	0.15	1.00
## Consistency_HR_3_num	0.00	-0.55
## Intent_Engag_1_num	-0.12	0.08
## Intent_Engag_2_num	0.59	0.14
## Intent_Engag_3_num	0.16	-0.23
## Intent_Engag_4_num	-0.32	-0.26
## AI_Knowledge_Experei_1_num	0.27	-0.22
## AI_Knowledge_Experei_2_num	0.20	-0.55
## AI_Knowledge_Experei_3_num	0.04	0.03
## AI_Knowledge_Experei_4_num	0.77	-0.08
## AI_Knowledge_Experei_5_num	-0.44	-0.27
## Attention_AI_score	-0.09	-0.44
## Org_Attraction	-0.35	-0.29
## Trust	-0.14	-0.19
## Communication	0.14	-0.16
## Chance_Perform	0.23	-0.11
## Consistency	0.29	0.26
## Intent	0.13	-0.12
## AI_Knowledge	0.43	-0.67
##	Consistency_HR_3_num	Intent_Engag_1_num
## Condition	-0.50	0.07
## Org_Attraction_1_num	-0.12	-0.15
## Org_Attraction_2_num	-0.46	-0.21
## Org_Attraction_3_num	-0.03	-0.16
## Org_Attraction_4_num	-0.61	0.19
## Org_Attraction_5_num	0.23	0.13
## Org_Attraction_6_num	-0.13	-0.08

## Trust_1_num	0.35	-0.24
## Trust_2_num	-0.33	0.04
## Trust_3_num	-0.03	-0.03
## Trust_4_num	0.45	-0.18
## Communication_1_num	0.32	0.38
## Communication_2_num	0.62	-0.11
## Communication_3_num	-0.21	0.29
## Communication_4_num	0.23	0.12
## Communication_5_num	0.44	-0.22
## Chance_Perform_1_num	-0.42	0.16
## Chance_Perform_2_num	0.34	0.14
## Chance_Perform_3_num	-0.01	-0.77
## Chance_Perform_4_num	-0.25	0.33
## Consistency_AI_1_num	-0.25	0.20
## Consistency_AI_2_num	0.66	-0.09
## Consistency_AI_3_num	-0.20	0.28
## Consistency_HR_1_num	0.00	-0.12
## Consistency_HR_2_num	-0.55	0.08
## Consistency_HR_3_num	1.00	0.01
## Intent_Engag_1_num	0.01	1.00
## Intent_Engag_2_num	-0.03	0.49
## Intent_Engag_3_num	-0.09	-0.46
## Intent_Engag_4_num	0.15	0.50
## AI_Knowledge_Experei_1_num	0.26	0.36
## AI_Knowledge_Experei_2_num	0.84	-0.15
## AI_Knowledge_Experei_3_num	-0.22	-0.62
## AI_Knowledge_Experei_4_num	0.45	0.08
## AI_Knowledge_Experei_5_num	-0.08	-0.08
## Attention_AI_score	0.63	-0.09
## Org_Attraction	-0.46	-0.11
## Trust	0.18	-0.18
## Communication	0.66	0.19
## Chance_Perform	-0.16	-0.13
## Consistency	0.44	0.14
## Intent	0.02	0.57
## AI_Knowledge	0.74	-0.10
##	Intent_Engag_2_num	Intent_Engag_3_num
## Condition	0.46	0.22
## Org_Attraction_1_num	-0.59	-0.01
## Org_Attraction_2_num	0.15	0.70
## Org_Attraction_3_num	-0.21	-0.25
## Org_Attraction_4_num	0.15	-0.10
## Org_Attraction_5_num	-0.62	-0.22
## Org_Attraction_6_num	0.04	0.50
## Trust_1_num	-0.03	0.07
## Trust_2_num	0.07	-0.13
## Trust_3_num	0.22	0.04
## Trust_4_num	-0.38	-0.70
## Communication_1_num	-0.23	-0.33
## Communication_2_num	0.28	0.11
## Communication_3_num	0.41	-0.25
## Communication_4_num	0.12	0.08
## Communication_5_num	-0.08	-0.43
## Chance_Perform_1_num	0.14	-0.25

## Chance_Perform_2_num	-0.32	-0.13
## Chance_Perform_3_num	-0.27	0.86
## Chance_Perform_4_num	0.31	0.02
## Consistency_AI_1_num	0.02	-0.40
## Consistency_AI_2_num	-0.28	0.03
## Consistency_AI_3_num	-0.06	0.15
## Consistency_HR_1_num	0.59	0.16
## Consistency_HR_2_num	0.14	-0.23
## Consistency_HR_3_num	-0.03	-0.09
## Intent_Engag_1_num	0.49	-0.46
## Intent_Engag_2_num	1.00	0.11
## Intent_Engag_3_num	0.11	1.00
## Intent_Engag_4_num	0.25	0.30
## AI_Knowledge_Experei_1_num	0.16	0.00
## AI_Knowledge_Experei_2_num	0.24	-0.04
## AI_Knowledge_Experei_3_num	-0.45	0.70
## AI_Knowledge_Experei_4_num	0.51	0.01
## AI_Knowledge_Experei_5_num	-0.08	0.04
## Attention_AI_score	-0.06	-0.03
## Org_Attraction	-0.47	0.27
## Trust	-0.03	-0.30
## Communication	0.24	-0.39
## Chance_Perform	-0.14	0.24
## Consistency	0.17	-0.14
## Intent	0.73	0.42
## AI_Knowledge	0.32	0.31
##	Intent_Engag_4_num	AI_Knowledge_Experei_1_num
## Condition	-0.32	0.00
## Org_Attraction_1_num	-0.10	0.33
## Org_Attraction_2_num	0.00	0.22
## Org_Attraction_3_num	-0.55	-0.40
## Org_Attraction_4_num	-0.06	0.22
## Org_Attraction_5_num	0.28	-0.07
## Org_Attraction_6_num	0.44	-0.55
## Trust_1_num	0.17	-0.71
## Trust_2_num	0.06	0.00
## Trust_3_num	-0.09	-0.63
## Trust_4_num	-0.51	-0.14
## Communication_1_num	0.34	-0.28
## Communication_2_num	0.05	0.13
## Communication_3_num	-0.46	0.32
## Communication_4_num	0.10	0.68
## Communication_5_num	-0.22	-0.46
## Chance_Perform_1_num	-0.23	-0.47
## Chance_Perform_2_num	0.15	0.06
## Chance_Perform_3_num	-0.05	0.00
## Chance_Perform_4_num	0.09	0.57
## Consistency_AI_1_num	-0.11	0.21
## Consistency_AI_2_num	0.23	0.37
## Consistency_AI_3_num	0.18	0.16
## Consistency_HR_1_num	-0.32	0.27
## Consistency_HR_2_num	-0.26	-0.22
## Consistency_HR_3_num	0.15	0.26
## Intent_Engag_1_num	0.50	0.36

## Intent_Engag_2_num	0.25	0.16
## Intent_Engag_3_num	0.30	0.00
## Intent_Engag_4_num	1.00	0.24
## AI_Knowledge_Experei_1_num	0.24	1.00
## AI_Knowledge_Experei_2_num	0.06	-0.38
## AI_Knowledge_Experei_3_num	0.01	0.37
## AI_Knowledge_Experei_4_num	0.03	0.67
## AI_Knowledge_Experei_5_num	-0.06	-0.74
## Attention_AI_score	0.40	0.13
## Org_Attraction	0.05	-0.03
## Trust	-0.16	-0.68
## Communication	-0.12	0.18
## Chance_Perform	-0.06	-0.08
## Consistency	-0.03	0.53
## Intent	0.81	0.29
## AI_Knowledge	0.25	0.38
##	AI_Knowledge_Experei_2_num	
## Condition	-0.42	
## Org_Attraction_1_num	-0.41	
## Org_Attraction_2_num	-0.73	
## Org_Attraction_3_num	0.18	
## Org_Attraction_4_num	-0.48	
## Org_Attraction_5_num	-0.10	
## Org_Attraction_6_num	0.21	
## Trust_1_num	0.89	
## Trust_2_num	-0.08	
## Trust_3_num	0.38	
## Trust_4_num	0.53	
## Communication_1_num	0.24	
## Communication_2_num	0.45	
## Communication_3_num	-0.39	
## Communication_4_num	-0.19	
## Communication_5_num	0.92	
## Chance_Perform_1_num	-0.04	
## Chance_Perform_2_num	-0.03	
## Chance_Perform_3_num	0.00	
## Chance_Perform_4_num	-0.45	
## Consistency_AI_1_num	-0.41	
## Consistency_AI_2_num	0.31	
## Consistency_AI_3_num	-0.64	
## Consistency_HR_1_num	0.20	
## Consistency_HR_2_num	-0.55	
## Consistency_HR_3_num	0.84	
## Intent_Engag_1_num	-0.15	
## Intent_Engag_2_num	0.24	
## Intent_Engag_3_num	-0.04	
## Intent_Engag_4_num	0.06	
## AI_Knowledge_Experei_1_num	-0.38	
## AI_Knowledge_Experei_2_num	1.00	
## AI_Knowledge_Experei_3_num	-0.44	
## AI_Knowledge_Experei_4_num	0.43	
## AI_Knowledge_Experei_5_num	0.52	
## Attention_AI_score	0.70	
## Org_Attraction	-0.60	

## Trust	0.82
## Communication	0.47
## Chance_Perform	-0.18
## Consistency	-0.08
## Intent	0.06
## AI_Knowledge	0.85
##	AI_Knowledge_Experei_3_num
## Condition	-0.13
## Org_Attraction_1_num	0.42
## Org_Attraction_2_num	0.62
## Org_Attraction_3_num	-0.47
## Org_Attraction_4_num	0.10
## Org_Attraction_5_num	-0.12
## Org_Attraction_6_num	-0.18
## Trust_1_num	-0.42
## Trust_2_num	0.12
## Trust_3_num	-0.40
## Trust_4_num	-0.27
## Communication_1_num	-0.54
## Communication_2_num	0.13
## Communication_3_num	-0.29
## Communication_4_num	0.42
## Communication_5_num	-0.37
## Chance_Perform_1_num	-0.55
## Chance_Perform_2_num	-0.17
## Chance_Perform_3_num	0.76
## Chance_Perform_4_num	0.06
## Consistency_AI_1_num	0.25
## Consistency_AI_2_num	0.28
## Consistency_AI_3_num	0.14
## Consistency_HR_1_num	0.04
## Consistency_HR_2_num	0.03
## Consistency_HR_3_num	-0.22
## Intent_Engag_1_num	-0.62
## Intent_Engag_2_num	-0.45
## Intent_Engag_3_num	0.70
## Intent_Engag_4_num	0.01
## AI_Knowledge_Experei_1_num	0.37
## AI_Knowledge_Experei_2_num	-0.44
## AI_Knowledge_Experei_3_num	1.00
## AI_Knowledge_Experei_4_num	0.18
## AI_Knowledge_Experei_5_num	-0.64
## Attention_AI_score	0.12
## Org_Attraction	0.25
## Trust	-0.48
## Communication	-0.30
## Chance_Perform	0.01
## Consistency	0.21
## Intent	-0.14
## AI_Knowledge	0.01
##	AI_Knowledge_Experei_4_num
## Condition	-0.13
## Org_Attraction_1_num	-0.36
## Org_Attraction_2_num	0.04

## Org_Attraction_3_num	-0.34	
## Org_Attraction_4_num	0.15	
## Org_Attraction_5_num	-0.47	
## Org_Attraction_6_num	-0.66	
## Trust_1_num	-0.28	
## Trust_2_num	0.31	
## Trust_3_num	-0.32	
## Trust_4_num	0.08	
## Communication_1_num	-0.68	
## Communication_2_num	0.62	
## Communication_3_num	0.42	
## Communication_4_num	0.65	
## Communication_5_num	0.17	
## Chance_Perform_1_num	-0.27	
## Chance_Perform_2_num	-0.01	
## Chance_Perform_3_num	0.00	
## Chance_Perform_4_num	0.27	
## Consistency_AI_1_num	0.41	
## Consistency_AI_2_num	0.25	
## Consistency_AI_3_num	-0.33	
## Consistency_HR_1_num	0.77	
## Consistency_HR_2_num	-0.08	
## Consistency_HR_3_num	0.45	
## Intent_Engag_1_num	0.08	
## Intent_Engag_2_num	0.51	
## Intent_Engag_3_num	0.01	
## Intent_Engag_4_num	0.03	
## AI_Knowledge_Experei_1_num	0.67	
## AI_Knowledge_Experei_2_num	0.43	
## AI_Knowledge_Experei_3_num	0.18	
## AI_Knowledge_Experei_4_num	1.00	
## AI_Knowledge_Experei_5_num	-0.82	
## Attention_AI_score	0.48	
## Org_Attraction	-0.58	
## Trust	-0.10	
## Communication	0.57	
## Chance_Perform	-0.12	
## Consistency	0.72	
## Intent	0.25	
## AI_Knowledge	0.74	
##		
## Condition	AI_Knowledge_Experei_5_num	Attention_AI_score
## Org_Attraction_1_num	0.14	-0.73
## Org_Attraction_2_num	0.02	-0.03
## Org_Attraction_3_num	-0.29	-0.58
## Org_Attraction_4_num	0.62	-0.37
## Org_Attraction_5_num	-0.23	-0.30
## Org_Attraction_6_num	0.07	0.09
## Trust_1_num	0.63	-0.01
## Trust_2_num	0.67	0.51
## Trust_3_num	-0.35	0.07
## Trust_4_num	0.75	-0.17
## Communication_1_num	0.03	0.42
## Communication_2_num	0.59	0.18
## Communication_3_num	-0.33	0.38

## Communication_3_num	-0.29	-0.66
## Communication_4_num	-0.57	0.09
## Communication_5_num	0.21	0.60
## Chance_Perform_1_num	0.48	-0.49
## Chance_Perform_2_num	-0.12	-0.07
## Chance_Perform_3_num	-0.02	0.03
## Chance_Perform_4_num	-0.54	-0.21
## Consistency_AI_1_num	-0.52	-0.08
## Consistency_AI_2_num	-0.45	0.55
## Consistency_AI_3_num	-0.01	-0.55
## Consistency_HR_1_num	-0.44	-0.09
## Consistency_HR_2_num	-0.27	-0.44
## Consistency_HR_3_num	-0.08	0.63
## Intent_Engag_1_num	-0.08	-0.09
## Intent_Engag_2_num	-0.08	-0.06
## Intent_Engag_3_num	0.04	-0.03
## Intent_Engag_4_num	-0.06	0.40
## AI_Knowledge_Experei_1_num	-0.74	0.13
## AI_Knowledge_Experei_2_num	0.52	0.70
## AI_Knowledge_Experei_3_num	-0.64	0.12
## AI_Knowledge_Experei_4_num	-0.82	0.48
## AI_Knowledge_Experei_5_num	1.00	-0.18
## Attention_AI_score	-0.18	1.00
## Org_Attraction	0.28	-0.46
## Trust	0.65	0.33
## Communication	-0.23	0.30
## Chance_Perform	0.02	-0.41
## Consistency	-0.75	0.12
## Intent	-0.07	0.09
## AI_Knowledge	-0.17	0.64
##	Org_Attraction Trust Communication Chance_Perform	
## Condition	0.06 -0.46 -0.36 0.02	
## Org_Attraction_1_num	0.58 -0.30 -0.29 -0.17	
## Org_Attraction_2_num	0.33 -0.63 -0.35 0.26	
## Org_Attraction_3_num	0.38 0.52 0.18 0.36	
## Org_Attraction_4_num	0.47 -0.06 -0.26 0.38	
## Org_Attraction_5_num	0.39 0.02 -0.10 0.46	
## Org_Attraction_6_num	0.26 0.08 -0.58 0.22	
## Trust_1_num	-0.24 0.77 0.11 -0.09	
## Trust_2_num	0.22 0.30 -0.02 0.47	
## Trust_3_num	-0.06 0.68 0.35 -0.13	
## Trust_4_num	-0.44 0.47 0.44 -0.39	
## Communication_1_num	0.00 0.27 0.13 -0.35	
## Communication_2_num	-0.75 0.16 0.77 -0.33	
## Communication_3_num	-0.02 -0.23 0.31 0.29	
## Communication_4_num	0.03 -0.17 0.53 -0.01	
## Communication_5_num	-0.54 0.78 0.39 -0.18	
## Chance_Perform_1_num	0.36 0.47 -0.05 0.61	
## Chance_Perform_2_num	0.26 0.05 0.25 0.68	
## Chance_Perform_3_num	0.37 -0.24 -0.41 0.34	
## Chance_Perform_4_num	0.00 -0.78 -0.44 0.07	
## Consistency_AI_1_num	-0.10 0.14 0.44 -0.12	
## Consistency_AI_2_num	-0.41 -0.34 0.17 -0.31	
## Consistency_AI_3_num	0.35 -0.34 0.09 0.09	

## Consistency_HR_1_num	-0.35	-0.14	0.14	0.23
## Consistency_HR_2_num	-0.29	-0.19	-0.16	-0.11
## Consistency_HR_3_num	-0.46	0.18	0.66	-0.16
## Intent_Engag_1_num	-0.11	-0.18	0.19	-0.13
## Intent_Engag_2_num	-0.47	-0.03	0.24	-0.14
## Intent_Engag_3_num	0.27	-0.30	-0.39	0.24
## Intent_Engag_4_num	0.05	-0.16	-0.12	-0.06
## AI_Knowledge_Experei_1_num	-0.03	-0.68	0.18	-0.08
## AI_Knowledge_Experei_2_num	-0.60	0.82	0.47	-0.18
## AI_Knowledge_Experei_3_num	0.25	-0.48	-0.30	0.01
## AI_Knowledge_Experei_4_num	-0.58	-0.10	0.57	-0.12
## AI_Knowledge_Experei_5_num	0.28	0.65	-0.23	0.02
## Attention_AI_score	-0.46	0.33	0.30	-0.41
## Org_Attraction	1.00	-0.23	-0.61	0.59
## Trust	-0.23	1.00	0.41	-0.07
## Communication	-0.61	0.41	1.00	-0.26
## Chance_Perform	0.59	-0.07	-0.26	1.00
## Consistency	-0.58	-0.26	0.67	-0.16
## Intent	-0.09	-0.27	-0.05	-0.03
## AI_Knowledge	-0.35	0.06	0.37	-0.05
##	Consistency	Intent	AI_Knowledge	
## Condition	-0.15	0.17	-0.24	
## Org_Attraction_1_num	-0.30	-0.34	-0.13	
## Org_Attraction_2_num	0.15	0.28	-0.18	
## Org_Attraction_3_num	-0.40	-0.47	-0.13	
## Org_Attraction_4_num	-0.27	0.06	-0.21	
## Org_Attraction_5_num	-0.05	-0.18	-0.20	
## Org_Attraction_6_num	-0.62	0.37	-0.02	
## Trust_1_num	-0.39	0.00	0.27	
## Trust_2_num	-0.09	0.01	-0.06	
## Trust_3_num	-0.22	0.06	-0.12	
## Trust_4_num	0.13	-0.72	0.08	
## Communication_1_num	-0.23	0.05	-0.14	
## Communication_2_num	0.78	0.14	0.45	
## Communication_3_num	0.38	-0.02	-0.15	
## Communication_4_num	0.55	0.17	0.31	
## Communication_5_num	-0.04	-0.39	0.28	
## Chance_Perform_1_num	-0.40	-0.09	-0.36	
## Chance_Perform_2_num	0.29	-0.07	-0.01	
## Chance_Perform_3_num	-0.19	-0.05	0.33	
## Chance_Perform_4_num	0.09	0.29	0.03	
## Consistency_AI_1_num	0.46	-0.13	-0.35	
## Consistency_AI_2_num	0.51	-0.04	0.37	
## Consistency_AI_3_num	0.27	0.21	-0.45	
## Consistency_HR_1_num	0.29	0.13	0.43	
## Consistency_HR_2_num	0.26	-0.12	-0.67	
## Consistency_HR_3_num	0.44	0.02	0.74	
## Intent_Engag_1_num	0.14	0.57	-0.10	
## Intent_Engag_2_num	0.17	0.73	0.32	
## Intent_Engag_3_num	-0.14	0.42	0.31	
## Intent_Engag_4_num	-0.03	0.81	0.25	
## AI_Knowledge_Experei_1_num	0.53	0.29	0.38	
## AI_Knowledge_Experei_2_num	-0.08	0.06	0.85	
## AI_Knowledge_Experei_3_num	0.21	-0.14	0.01	

## AI_Knowledge_Experei_4_num	0.72	0.25	0.74
## AI_Knowledge_Experei_5_num	-0.75	-0.07	-0.17
## Attention_AI_score	0.12	0.09	0.64
## Org_Attraction	-0.58	-0.09	-0.35
## Trust	-0.26	-0.27	0.06
## Communication	0.67	-0.05	0.37
## Chance_Perform	-0.16	-0.03	-0.05
## Consistency	1.00	0.05	0.20
## Intent	0.05	1.00	0.33
## AI_Knowledge	0.20	0.33	1.00

```

Ttest_JAR_Social_Invitees <- JAR_Social_Invitees %>%
  dplyr::select(Condition, Org_Attraction, Trust, Communication, Chance_Perform, Consistency, Intent) %>%
  mutate(Condition = as.factor(Condition)) %>%
  ungroup() %>%           # remove rowwise grouping
  as.data.frame()         # convert tibble to base R data.frame

independentSamplesTTest(Intent ~ Condition, Ttest_JAR_Social_Invitees)

```

```

##
##   Welch's independent samples t-test
##
## Outcome variable:   Intent
## Grouping variable:  Condition
##
## Descriptive statistics:
##           1      2
##   mean      2.917 3.188
##   std dev.  0.785 0.966
##
## Hypotheses:
##   null:           population means equal for both groups
##   alternative:    different population means in each group
##
## Test results:
##   t-statistic:   -0.467
##   degrees of freedom: 5.578
##   p-value:      0.658
##
## Other information:
##   two-sided 95% confidence interval:  [-1.715, 1.174]
##   estimated effect size (Cohen's d):  0.308

```

```

independentSamplesTTest(Org_Attraction ~ Condition, Ttest_JAR_Social_Invitees)

```

```

##
##   Welch's independent samples t-test
##
## Outcome variable:   Org_Attraction
## Grouping variable:  Condition
##
## Descriptive statistics:
##           1      2

```

```
##      mean      3.056 3.125
##      std dev. 0.554 0.672
##
## Hypotheses:
##      null:      population means equal for both groups
##      alternative: different population means in each group
##
## Test results:
##      t-statistic: -0.171
##      degrees of freedom: 5.644
##      p-value: 0.87
##
## Other information:
##      two-sided 95% confidence interval: [-1.076, 0.937]
##      estimated effect size (Cohen's d): 0.113
```

```
independentSamplesTTest(Trust ~ Condition, Ttest_JAR_Social_Invitees)
```

```
##
##      Welch's independent samples t-test
##
## Outcome variable: Trust
## Grouping variable: Condition
##
## Descriptive statistics:
##           1      2
##      mean      3.542 2.750
##      std dev. 0.660 1.061
##
## Hypotheses:
##      null:      population means equal for both groups
##      alternative: different population means in each group
##
## Test results:
##      t-statistic: 1.331
##      degrees of freedom: 4.566
##      p-value: 0.246
##
## Other information:
##      two-sided 95% confidence interval: [-0.782, 2.366]
##      estimated effect size (Cohen's d): 0.896
```

```
independentSamplesTTest(Communication ~ Condition, Ttest_JAR_Social_Invitees)
```

```
##
##      Welch's independent samples t-test
##
## Outcome variable: Communication
## Grouping variable: Condition
##
## Descriptive statistics:
##           1      2
##      mean      3.000 2.500
```

```
##      std dev. 0.620 0.825
##
## Hypotheses:
##      null:          population means equal for both groups
##      alternative: different population means in each group
##
## Test results:
##      t-statistic:  1.034
##      degrees of freedom:  5.239
##      p-value:  0.347
##
## Other information:
##      two-sided 95% confidence interval:  [-0.727, 1.727]
##      estimated effect size (Cohen's d):  0.686
```

```
independentSamplesTTest(Chance_Perform ~ Condition, Ttest_JAR_Social_Invitees)
```

```
##
##      Welch's independent samples t-test
##
## Outcome variable:  Chance_Perform
## Grouping variable:  Condition
##
## Descriptive statistics:
##           1      2
##      mean      3.042 3.062
##      std dev. 0.886 0.375
##
## Hypotheses:
##      null:          population means equal for both groups
##      alternative: different population means in each group
##
## Test results:
##      t-statistic:  -0.051
##      degrees of freedom:  7.183
##      p-value:  0.961
##
## Other information:
##      two-sided 95% confidence interval:  [-0.979, 0.938]
##      estimated effect size (Cohen's d):  0.031
```

```
independentSamplesTTest(Consistency ~ Condition, Ttest_JAR_Social_Invitees)
```

```
##
##      Welch's independent samples t-test
##
## Outcome variable:  Consistency
## Grouping variable:  Condition
##
## Descriptive statistics:
##           1      2
##      mean      2.944 2.792
##      std dev. 0.593 0.498
```

```
##
## Hypotheses:
##   null:          population means equal for both groups
##   alternative:    different population means in each group
##
## Test results:
##   t-statistic:    0.44
##   degrees of freedom: 7.394
##   p-value:        0.673
##
## Other information:
##   two-sided 95% confidence interval: [-0.659, 0.965]
##   estimated effect size (Cohen's d): 0.279
```

Standard error or p-values: a measure of the error around of estimate of the coefficient To be statistically significant the coefficient should approximately be more than twice the standard error

```
#lm

Model1 <- JAR_Social_Invitees %>%
  lm(Intent ~ Condition + Org_Attraction + Trust + Communication + Chance_Perform + Consistency,
      data = .)

Model1_interaction <- JAR_Social_Invitees %>%
  lm(Intent ~ Condition * (Org_Attraction + Trust + Communication + Chance_Perform + Consistency),
      data = .)

Model2 <- JAR_Social_Invitees %>%
  lm(Intent ~ Trust + Communication + Chance_Perform + Consistency,
      data = .)

Model3 <- JAR_Social_Invitees %>%
  lm(Org_Attraction ~ Condition + Trust + Communication + Chance_Perform + Consistency,
      data = .)

Model3_interaction <- JAR_Social_Invitees %>%
  lm(Org_Attraction ~ Condition * (Trust + Communication + Chance_Perform + Consistency),
      data = .)

Model4 <- JAR_Social_Invitees %>%
  lm(Org_Attraction ~ Trust + Communication + Chance_Perform + Consistency,
      data = .)

Model5 <- JAR_Social_Invitees %>%
  filter(Condition == 2) %>%
  lm(Org_Attraction ~ Trust + Communication + Chance_Perform + Consistency,
      data = .)

Model6 <- JAR_Social_Invitees %>%
  lm(Org_Attraction ~ Trust + Communication + Chance_Perform + Consistency + AI_Knowledge,
      data = .)

stargazer(Model1, Model2, Model3, Model4,
          type = "text",
```

```
title = "Regression table for model 1, 3, 4, 5")
```

```
##
## Regression table for model 1, 3, 4, 5
## =====
##                               Dependent variable:
##                               -----
##                               Intent          Org_Attraction
##                               (1)            (2)            (3)            (4)
## -----
## Condition                    -0.285                -0.326
##                               (1.096)                (0.316)
##
## Org_Attraction              -1.013
##                               (1.543)
##
## Trust                       -0.950            -0.495            -0.468            -0.335
##                               (1.185)            (0.706)            (0.304)            (0.278)
##
## Communication                0.758            0.480            0.283            0.224
##                               (1.482)            (1.164)            (0.459)            (0.458)
##
## Chance_Perform              0.385            -0.015            0.393            0.404
##                               (0.873)            (0.520)            (0.204)            (0.205)
##
## Consistency                 -1.602            -0.573            -1.038            -0.878
##                               (2.417)            (1.448)            (0.587)            (0.570)
##
## Constant                    10.931            4.976            6.050**           4.835**
##                               (11.049)            (3.869)            (1.916)            (1.524)
## -----
## Observations                 10                10                10                10
## R2                          0.217                0.104                0.771                0.710
## Adjusted R2                 -1.348            -0.612                0.486                0.478
## Residual Std. Error  1.257 (df = 3)    1.042 (df = 5)    0.407 (df = 4)    0.410 (df = 5)
## F Statistic            0.139 (df = 6; 3) 0.146 (df = 4; 5) 2.699 (df = 5; 4) 3.064 (df = 4; 5)
## =====
## Note:                                *p<0.1; **p<0.05; ***p<0.01
```

```
stargazer(Model1_interaction, Model3_interaction,
  type = "text",
  title = "Regression table for model 1_interaction and 3_interaction")
```

```
##
## Regression table for model 1and 3
## =====
##                               Dependent variable:
##                               -----
##                               Intent    Org_Attraction
##                               (1)        (2)
## -----
```

```

## Condition                -475.565      14.474*
##                          (1.398)
##
## Org_Attraction           -35.284
##
##
## Trust                    -26.179       3.114*
##                          (0.375)
##
## Communication            -167.723      4.494*
##                          (0.366)
##
## Chance_Perform           -23.467       2.359*
##                          (0.230)
##
## Consistency              99.750       -4.916*
##                          (0.416)
##
## Condition:Org_Attraction  54.734
##
##
## Condition:Trust           32.113       -3.445*
##                          (0.377)
##
## Condition:Communication   75.848       -0.005
##                          (0.145)
##
## Condition:Chance_Perform          -1.212*
##                          (0.165)
##
## Condition:Consistency
##
##
## Constant                 451.332      -12.729*
##                          (1.709)
##
## -----
## Observations              10           10
## R2                        1.000        0.998
## Adjusted R2                0.986
## Residual Std. Error        0.067 (df = 1)
## F Statistic                80.466* (df = 8; 1)
## =====
## Note:                      *p<0.1; **p<0.05; ***p<0.01

```

```

#I think we don't need this anymore
summary(Model1)

```

```

##
## Call:
## lm(formula = Intent ~ Condition + Org_Attraction + Trust + Communication +
##     Chance_Perform + Consistency, data = .)
##
## Residuals:

```

```
##          1          2          3          4          5          6          7          8
##  1.12807  0.06231 -0.71196 -0.90537 -0.50192 -0.68183  0.66150  0.92569
##          9         10
## -0.23826  0.26175
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    10.9313    11.0492   0.989   0.395
## Condition      -0.2846     1.0965  -0.260   0.812
## Org_Attraction -1.0127     1.5430  -0.656   0.558
## Trust          -0.9502     1.1849  -0.802   0.481
## Communication   0.7584     1.4816   0.512   0.644
## Chance_Perform  0.3847     0.8734   0.440   0.689
## Consistency    -1.6015     2.4172  -0.663   0.555
##
## Residual standard error: 1.257 on 3 degrees of freedom
## Multiple R-squared:  0.2174, Adjusted R-squared:  -1.348
## F-statistic: 0.1389 on 6 and 3 DF,  p-value: 0.9794
```

```
summary(Model1_interaction)
```

```
##
## Call:
## lm(formula = Intent ~ Condition * (Org_Attraction + Trust + Communication +
##   Chance_Perform + Consistency), data = .)
##
## Residuals:
## ALL 10 residuals are 0: no residual degrees of freedom!
##
## Coefficients: (2 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)         451.33         NaN      NaN      NaN
## Condition          -475.57         NaN      NaN      NaN
## Org_Attraction      -35.28         NaN      NaN      NaN
## Trust              -26.18         NaN      NaN      NaN
## Communication     -167.72         NaN      NaN      NaN
## Chance_Perform     -23.47         NaN      NaN      NaN
## Consistency         99.75         NaN      NaN      NaN
## Condition:Org_Attraction  54.73         NaN      NaN      NaN
## Condition:Trust        32.11         NaN      NaN      NaN
## Condition:Communication  75.85         NaN      NaN      NaN
## Condition:Chance_Perform    NA          NA      NA      NA
## Condition:Consistency     NA          NA      NA      NA
##
## Residual standard error: NaN on 0 degrees of freedom
## Multiple R-squared:  1, Adjusted R-squared:  NaN
## F-statistic:  NaN on 9 and 0 DF,  p-value: NA
```

```
summary(Model2)
```

```
##
## Call:
## lm(formula = Intent ~ Trust + Communication + Chance_Perform +
```

```
## Consistency, data = .)
##
## Residuals:
##      1      2      3      4      5      6      7      8
## 0.95231 0.48160 -1.15535 -1.05743 -0.43201 -0.46328 0.52531 1.07196
##      9     10
## 0.07450 0.00239
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.97563    3.86937   1.286   0.255
## Trust          -0.49479    0.70552  -0.701   0.514
## Communication   0.48013    1.16411   0.412   0.697
## Chance_Perform -0.01511    0.51978  -0.029   0.978
## Consistency    -0.57336    1.44773  -0.396   0.708
##
## Residual standard error: 1.042 on 5 degrees of freedom
## Multiple R-squared:  0.1044, Adjusted R-squared:  -0.612
## F-statistic: 0.1458 on 4 and 5 DF,  p-value: 0.9572
```

```
summary(Model3)
```

```
##
## Call:
## lm(formula = Org_Attraction ~ Condition + Trust + Communication +
##      Chance_Perform + Consistency, data = .)
##
## Residuals:
##      1      2      3      4      5      6      7      8
## 0.16255 -0.44406 0.41960 0.15470 -0.07412 -0.17867 0.15022 -0.12625
##      9     10
## -0.31661 0.25264
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6.0499    1.9157   3.158 0.0342 *
## Condition      -0.3264    0.3156  -1.034 0.3595
## Trust          -0.4682    0.3044  -1.538 0.1988
## Communication   0.2829    0.4588   0.617 0.5708
## Chance_Perform  0.3933    0.2036   1.932 0.1255
## Consistency    -1.0375    0.5869  -1.768 0.1518
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4073 on 4 degrees of freedom
## Multiple R-squared:  0.7714, Adjusted R-squared:  0.4856
## F-statistic: 2.699 on 5 and 4 DF,  p-value: 0.1787
```

```
summary(Model3_interaction)
```

```
##
## Call:
## lm(formula = Org_Attraction ~ Condition * (Trust + Communication +
```



```
##      Chance_Perform + Consistency), data = .)
##
## Residuals:
##      1      2      3      4      5      6      7
## 5.392e-02 -2.589e-03 -1.936e-02 -1.752e-16 -3.478e-02 -2.949e-17 -7.459e-17
##      8      9     10
## 2.828e-16 1.013e-03 1.801e-03
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -12.728663    1.709370   -7.446  0.0850 .
## Condition      14.474263    1.398285   10.351  0.0613 .
## Trust          3.114105    0.375301    8.298  0.0764 .
## Communication  4.493982    0.366270   12.270  0.0518 .
## Chance_Perform 2.359105    0.230308   10.243  0.0620 .
## Consistency    -4.916032    0.415984  -11.818  0.0537 .
## Condition:Trust -3.444771    0.376790   -9.142  0.0694 .
## Condition:Communication -0.005306    0.145154   -0.037  0.9767
## Condition:Chance_Perform -1.211702    0.164597   -7.362  0.0860 .
## Condition:Consistency      NA          NA      NA      NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.0671 on 1 degrees of freedom
## Multiple R-squared:  0.9984, Adjusted R-squared:  0.986
## F-statistic: 80.47 on 8 and 1 DF, p-value: 0.08602
```

```
summary(Model14)
```

```
##
## Call:
## lm(formula = Org_Attraction ~ Trust + Communication + Chance_Perform +
##      Consistency, data = .)
##
## Residuals:
##      1      2      3      4      5      6      7      8
## 0.24171 -0.22807 0.55064 0.12203 -0.03752 -0.44567 0.03705 -0.25703
##      9     10
## -0.26074 0.27760
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.8351    1.5237   3.173  0.0247 *
## Trust         -0.3353    0.2778  -1.207  0.2815
## Communication  0.2240    0.4584   0.489  0.6458
## Chance_Perform 0.4043    0.2047   1.975  0.1052
## Consistency    -0.8777    0.5701  -1.540  0.1843
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4101 on 5 degrees of freedom
## Multiple R-squared:  0.7103, Adjusted R-squared:  0.4785
## F-statistic: 3.064 on 4 and 5 DF, p-value: 0.1254
```

```
summary(Model5)
```

```
##
## Call:
## lm(formula = Org_Attraction ~ Trust + Communication + Chance_Perform +
##     Consistency, data = .)
##
## Residuals:
## ALL 4 residuals are 0: no residual degrees of freedom!
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.0714         NaN      NaN      NaN
## Trust           0.6984         NaN      NaN      NaN
## Communication  -1.3690         NaN      NaN      NaN
## Chance_Perform  0.5079         NaN      NaN      NaN
## Consistency      NA           NA       NA       NA
##
## Residual standard error: NaN on 0 degrees of freedom
## Multiple R-squared: 1, Adjusted R-squared: NaN
## F-statistic: NaN on 3 and 0 DF, p-value: NA
```

```
summary(Model6)
```

```
##
## Call:
## lm(formula = Org_Attraction ~ Trust + Communication + Chance_Perform +
##     Consistency + AI_Knowledge, data = .)
##
## Residuals:
##      1      2      3      4      5      6      7      8
## 0.21456 -0.05723 0.31360 0.06844 0.17008 -0.56689 0.10668 -0.30765
##      9     10
## -0.07722 0.13563
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.9200     1.7549   3.373   0.028 *
## Trust          -0.4248     0.2806  -1.514   0.205
## Communication   0.4518     0.4870   0.928   0.406
## Chance_Perform  0.4234     0.1992   2.125   0.101
## Consistency    -1.0541     0.5740  -1.836   0.140
## AI_Knowledge   -0.3650     0.3183  -1.146   0.316
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3978 on 4 degrees of freedom
## Multiple R-squared: 0.7819, Adjusted R-squared: 0.5093
## F-statistic: 2.868 on 5 and 4 DF, p-value: 0.1646
```

```
#broom()
```

```
tidy(Model11)
```

```
## # A tibble: 7 x 5
##   term          estimate std.error statistic p.value
##   <chr>          <dbl>    <dbl>    <dbl>   <dbl>
## 1 (Intercept)    10.9      11.0      0.989   0.395
## 2 Condition     -0.285     1.10    -0.260   0.812
## 3 Org_Attraction -1.01      1.54    -0.656   0.558
## 4 Trust         -0.950     1.18    -0.802   0.481
## 5 Communication  0.758     1.48     0.512   0.644
## 6 Chance_Perform 0.385     0.873    0.440   0.689
## 7 Consistency   -1.60      2.42    -0.663   0.555
```

```
augment(Model1)
```

```
## # A tibble: 10 x 13
##   Intent Condition Org_Attraction Trust Communication Chance_Perform
##   <dbl>    <dbl>          <dbl> <dbl>          <dbl>          <dbl>
## 1 4          1          2.83 3.25          3.6          2.75
## 2 3.75        1          3    3            2.6          2.5
## 3 2          1          3.83 3            3.2          3.75
## 4 2.25        2          3.5 1.75          1.6          2.75
## 5 2.5         1          2.17 3.5           3.8          2.25
## 6 2.5         2          2.33 4            3.6          2.75
## 7 3.75        2          3.83 3.25          2.4          3.5
## 8 4.25        2          2.83 2            2.4          3.25
## 9 2.75        1          3.33 3.75          2.4          4.5
## 10 2.5        1          3.17 4.75          2.4          2.5
## # i 7 more variables: Consistency <dbl>, .fitted <dbl>, .resid <dbl>,
## #   .hat <dbl>, .sigma <dbl>, .cooksd <dbl>, .std.resid <dbl>
```

```
glance(Model1)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>    <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 0.217      -1.35 1.26    0.139 0.979     6 -10.5 36.9 39.3
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(Model2)
```

```
## # A tibble: 5 x 5
##   term          estimate std.error statistic p.value
##   <chr>          <dbl>    <dbl>    <dbl>   <dbl>
## 1 (Intercept)    4.98      3.87     1.29   0.255
## 2 Trust         -0.495     0.706    -0.701  0.514
## 3 Communication  0.480     1.16     0.412  0.697
## 4 Chance_Perform -0.0151    0.520    -0.0291 0.978
## 5 Consistency   -0.573     1.45    -0.396  0.708
```

```
augment(Model2)
```

```
## # A tibble: 10 x 11
```

```
##      Intent Trust Communication Chance_Perform Consistency .fitted .resid .hat
##      <dbl> <dbl>          <dbl>          <dbl>          <dbl> <dbl> <dbl> <dbl>
## 1    4      3.25          3.6            2.75          3.5   3.05 0.952 0.276
## 2    3.75    3            2.6            2.5            2.5   3.27 0.482 0.385
## 3    2      3            3.2            3.75          3.17   3.16 -1.16 0.354
## 4    2.25  1.75          1.6            2.75          2.67   3.31 -1.06 0.635
## 5    2.5    3.5          3.8            2.25          3.67   2.93 -0.432 0.484
## 6    2.5    4            3.6            2.75          3        2.96 -0.463 0.305
## 7    3.75  3.25          2.4            3.5            2.17   3.22 0.525 0.545
## 8    4.25    2            2.4            3.25          3.33   3.18 1.07 0.441
## 9    2.75  3.75          2.4            4.5            2.67   2.68 0.0745 0.750
## 10   2.5    4.75          2.4            2.5            2.17   2.50 0.00239 0.826
## # i 3 more variables: .sigma <dbl>, .cooksd <dbl>, .std.resid <dbl>
```

```
glance(Model2)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.104      -0.612 1.04      0.146  0.957     4 -11.1  34.3  36.1
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(Model3)
```

```
## # A tibble: 6 x 5
##   term                estimate std.error statistic p.value
##   <chr>              <dbl>    <dbl>      <dbl>  <dbl>
## 1 (Intercept)        6.05      1.92      3.16  0.0342
## 2 Condition          -0.326     0.316    -1.03  0.359
## 3 Trust              -0.468     0.304    -1.54  0.199
## 4 Communication       0.283     0.459     0.617 0.571
## 5 Chance_Perform     0.393     0.204     1.93  0.126
## 6 Consistency        -1.04     0.587    -1.77  0.152
```

```
augment(Model3)
```

```
## # A tibble: 10 x 12
##   Org_Attraction Condition Trust Communication Chance_Perform Consistency
##   <dbl>      <dbl> <dbl>          <dbl>          <dbl>          <dbl>
## 1      2.83      1 3.25          3.6            2.75          3.5
## 2      3        1 3            2.6            2.5            2.5
## 3      3.83      1 3            3.2            3.75          3.17
## 4      3.5      2 1.75          1.6            2.75          2.67
## 5      2.17      1 3.5          3.8            2.25          3.67
## 6      2.33      2 4            3.6            2.75          3
## 7      3.83      2 3.25          2.4            3.5            2.17
## 8      2.83      2 2            2.4            3.25          3.33
## 9      3.33      1 3.75          2.4            4.5            2.67
## 10     3.17      1 4.75          2.4            2.5            2.17
## # i 6 more variables: .fitted <dbl>, .resid <dbl>, .hat <dbl>, .sigma <dbl>,
## #   .cooksd <dbl>, .std.resid <dbl>
```

```
glance(Model3)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.771      0.486 0.407      2.70   0.179     5 -0.626  15.3  17.4
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(Model4)
```

```
## # A tibble: 5 x 5
##   term          estimate std.error statistic p.value
##   <chr>          <dbl>    <dbl>      <dbl>  <dbl>
## 1 (Intercept)    4.84      1.52      3.17   0.0247
## 2 Trust         -0.335     0.278    -1.21   0.281
## 3 Communication  0.224     0.458     0.489  0.646
## 4 Chance_Perform 0.404     0.205     1.98   0.105
## 5 Consistency   -0.878     0.570    -1.54   0.184
```

```
augment(Model4)
```

```
## # A tibble: 10 x 11
##   Org_Attraction Trust Communication Chance_Perform Consistency .fitted .resid
##   <dbl> <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1    2.83  3.25      3.6        2.75      3.5    2.59  0.242
## 2     3    3      2.6        2.5      2.5    3.23 -0.228
## 3    3.83  3      3.2        3.75      3.17    3.28  0.551
## 4     3.5  1.75     1.6        2.75      2.67    3.38  0.122
## 5     2.17  3.5     3.8        2.25      3.67    2.20 -0.0375
## 6     2.33  4      3.6        2.75      3      2.78 -0.446
## 7    3.83  3.25     2.4        3.5      2.17    3.80  0.0371
## 8     2.83  2      2.4        3.25      3.33    3.09 -0.257
## 9     3.33  3.75     2.4        4.5      2.67    3.59 -0.261
## 10    3.17  4.75     2.4        2.5      2.17    2.89  0.278
## # i 4 more variables: .hat <dbl>, .sigma <dbl>, .cooksd <dbl>, .std.resid <dbl>
```

```
glance(Model4)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.710      0.478 0.410      3.06   0.125     4 -1.81  15.6  17.4
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(Model5)
```

```
## # A tibble: 5 x 5
##   term          estimate std.error statistic p.value
##   <chr>          <dbl>    <dbl>      <dbl>  <dbl>
## 1 (Intercept)    3.07      NaN      NaN    NaN
```

```
## 2 Trust          0.698      NaN      NaN      NaN
## 3 Communication  -1.37      NaN      NaN      NaN
## 4 Chance_Perform  0.508      NaN      NaN      NaN
## 5 Consistency    NA         NA       NA       NA
```

```
augment(Model5)
```

```
## # A tibble: 4 x 11
##   Org_Attraction Trust Communication Chance_Perform Consistency .fitted
##   <dbl> <dbl>      <dbl>      <dbl>      <dbl> <dbl>
## 1      3.5  1.75      1.6        2.75      2.67  3.5
## 2      2.33  4        3.6        2.75      3      2.33
## 3      3.83  3.25     2.4        3.5       2.17  3.83
## 4      2.83  2        2.4        3.25     3.33  2.83
## # i 5 more variables: .resid <dbl>, .hat <dbl>, .sigma <dbl>, .cooksd <dbl>,
## #   .std.resid <dbl>
```

```
glance(Model5)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1      1      NaN    NaN      NaN    NaN     3    Inf -Inf -Inf
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(Model6)
```

```
## # A tibble: 6 x 5
##   term          estimate std.error statistic p.value
##   <chr>      <dbl>      <dbl>      <dbl>  <dbl>
## 1 (Intercept)  5.92      1.75      3.37  0.0280
## 2 Trust       -0.425     0.281     -1.51  0.205
## 3 Communication  0.452     0.487     0.928  0.406
## 4 Chance_Perform 0.423     0.199     2.13  0.101
## 5 Consistency  -1.05     0.574     -1.84  0.140
## 6 AI_Knowledge -0.365     0.318     -1.15  0.316
```

```
augment(Model6)
```

```
## # A tibble: 10 x 12
##   Org_Attraction Trust Communication Chance_Perform Consistency AI_Knowledge
##   <dbl> <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1      2.83  3.25      3.6        2.75      3.5        2.8
## 2      3      3        2.6        2.5       2.5        3.25
## 3      3.83  3        3.2        3.75     3.17       2.25
## 4      3.5  1.75     1.6        2.75     2.67       2.25
## 5      2.17  3.5      3.8        2.25     3.67       3.4
## 6      2.33  4        3.6        2.75      3        2.6
## 7      3.83  3.25     2.4        3.5       2.17       3
## 8      2.83  2        2.4        3.25     3.33       2.4
## 9      3.33  3.75     2.4        4.5       2.67       3
```

```
## 10          3.17  4.75          2.4          2.5          2.17          2
## # i 6 more variables: .fitted <dbl>, .resid <dbl>, .hat <dbl>, .sigma <dbl>,
## #   .cooksds <dbl>, .std.resid <dbl>
```

```
glance(Model6)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>    <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.782      0.509 0.398      2.87   0.165     5 -0.390  14.8  16.9
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
# 1. Mediator model: Condition → Org_Attraction
```

```
model.m <- lm(Org_Attraction ~ Condition, data = JAR_Social_Invitees)
```

```
# 2. Outcome model: Org_Attraction + Condition → Intent
```

```
model.y <- lm(Intent ~ Org_Attraction + Condition, data = JAR_Social_Invitees)
```

```
med.out <- mediate(model.m, model.y, treat = "Condition", mediator = "Org_Attraction", boot = TRUE, sim
```

```
## Running nonparametric bootstrap
```

```
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
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```
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```
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```

```
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## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
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```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```



```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):  
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## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
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## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
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## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases  
  
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):  
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
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## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
summary(med.out)
```

```
##
## Causal Mediation Analysis
##
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME           -0.0106   -0.5471      0.75   0.91
## ADE             0.2814   -1.0526      1.38   0.68
## Total Effect    0.2708   -0.8057      1.25   0.66
```

```

## Prop. Mediated  -0.0391      -3.8013      6.68      NA
##
## Sample Size Used: 10
##
##
## Simulations: 1000

# Mediation effect of trust, communication, chance to perform, consistency on the relationship between
#Mediation through Trust
# Step 1: Condition → Trust
model.m1 <- lm(Trust ~ Condition, data = JAR_Social_Invitees)

# Step 2: Org_Attraction ~ Condition + Trust
model.y1 <- lm(Org_Attraction ~ Condition + Trust, data = JAR_Social_Invitees)

# Mediation analysis
med.out1 <- mediate(model.m1, model.y1, treat = "Condition", mediator = "Trust", boot = TRUE, sims = 1000)

## Running nonparametric bootstrap

## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
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## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

```



```

## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
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## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
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## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
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## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
summary(med.out1)
```

```
##
## Causal Mediation Analysis
##
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME           0.1267   -0.5759      0.58   0.67
## ADE            -0.0573   -0.9285      1.14   0.96
## Total Effect    0.0694   -0.7143      0.83   0.83
## Prop. Mediated  1.8251   -8.1343     10.25    NA
##
## Sample Size Used: 10
##
##
## Simulations: 1000
```

```
#Mediation through Communication
```

```
model.m2 <- lm(Communication ~ Condition, data = JAR_Social_Invitees)
model.y2 <- lm(Org_Attraction ~ Condition + Communication, data = JAR_Social_Invitees)
med.out2 <- mediate(model.m2, model.y2, treat = "Condition", mediator = "Communication", boot = TRUE, s
```



```

## Running nonparametric bootstrap

## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

```



```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
summary(med.out2)
```

```
##
## Causal Mediation Analysis
##
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##           Estimate 95% CI Lower 95% CI Upper p-value
```

```
## ACME          0.2715      -0.2825      0.85      0.36
## ADE           -0.2020     -0.7813      0.53      0.60
## Total Effect   0.0694     -0.6667      0.81      0.85
## Prop. Mediated 3.9091     -6.7796     13.45      NA
##
## Sample Size Used: 10
##
##
## Simulations: 1000
```

```
#Mediation through Chance to Perform
```

```
model.m3 <- lm(Chance_Perform ~ Condition, data = JAR_Social_Invitees)
model.y3 <- lm(Org_Attraction ~ Condition + Chance_Perform, data = JAR_Social_Invitees)
med.out3 <- mediate(model.m3, model.y3, treat = "Condition", mediator = "Chance_Perform", boot = TRUE, ,
```

```
## Running nonparametric bootstrap
```

```
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
summary(med.out3)
```

```
##
## Causal Mediation Analysis
##
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME           0.00996   -0.29796      0.80   0.97
## ADE            0.05948   -0.67937      0.70   0.99
## Total Effect    0.06944   -0.63906      0.81   0.90
## Prop. Mediated  0.14347   -4.79414      8.43    NA
##
## Sample Size Used: 10
##
##
## Simulations: 1000
```

```
#Mediation through Consistency
```

```
model.m4 <- lm(Consistency ~ Condition, data = JAR_Social_Invitees)
model.y4 <- lm(Org_Attraction ~ Condition + Consistency, data = JAR_Social_Invitees)
med.out4 <- mediate(model.m4, model.y4, treat = "Condition", mediator = "Consistency", boot = TRUE, sim
```

```
## Running nonparametric bootstrap
```

```
## Warning in predict.lm(new.fit.M, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```



```
## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.t):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases

## Warning in predict.lm(new.fit.Y, type = "response", newdata = pred.data.c):
## prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
```

```
summary(med.out4)
```

```
##
## Causal Mediation Analysis
##
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME           0.0943   -0.2370      0.86   0.62
## ADE            -0.0248   -0.9213      0.49   0.84
## Total Effect    0.0694   -0.7000      0.78   0.88
## Prop. Mediated   1.3574   -6.0497      6.74    NA
##
## Sample Size Used: 10
##
##
## Simulations: 1000
```

```
# Define the mediation model for the relationship between condition and org attraction
model <- '
  # Paths from IV to mediators
  Trust ~ a1*Condition
  Communication ~ a2*Condition
  Chance_Perform ~ a3*Condition
  Consistency ~ a4*Condition

  # Paths from mediators to DV
  Org_Attraction ~ b1*Trust + b2*Communication + b3*Chance_Perform + b4*Consistency

  # Direct effect of IV on DV
  Org_Attraction ~ c_prime*Condition

  # Indirect effects
  ind1 := a1 * b1
  ind2 := a2 * b2
  ind3 := a3 * b3
  ind4 := a4 * b4

  # Total indirect effect
  total_ind := ind1 + ind2 + ind3 + ind4
```


[illegible]


```
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite
```

```
## Warning in lav_model_nvcov_bootstrap(lavmodel = lavmodel, lavsamplestats =  
## lavsamplestats, : lavaan WARNING: 281 bootstrap runs failed or did not  
## converge.
```

```
# View summary with standardized estimates and confidence intervals  
summary(fit, fit.measures = TRUE, standardized = TRUE, ci = TRUE)
```

```
## lavaan 0.6.16 ended normally after 9 iterations
```

```
##  
##      Estimator                      ML  
##      Optimization method          NLMINB  
##      Number of model parameters      14  
##  
##      Number of observations          10  
##  
## Model Test User Model:  
##  
##      Test statistic                18.108  
##      Degrees of freedom              6  
##      P-value (Chi-square)           0.006  
##  
## Model Test Baseline Model:  
##  
##      Test statistic                36.897  
##      Degrees of freedom             15  
##      P-value                       0.001  
##  
## User Model versus Baseline Model:  
##  
##      Comparative Fit Index (CFI)      0.447  
##      Tucker-Lewis Index (TLI)       -0.382  
##  
## Loglikelihood and Information Criteria:  
##  
##      Loglikelihood user model (H0)    -38.720  
##      Loglikelihood unrestricted model (H1) -29.666  
##  
##      Akaike (AIC)                   105.440  
##      Bayesian (BIC)                  109.676  
##      Sample-size adjusted Bayesian (SABIC) 67.736  
##  
## Root Mean Square Error of Approximation:  
##  
##      RMSEA                           0.449  
##      90 Percent confidence interval - lower 0.221  
##      90 Percent confidence interval - upper 0.694  
##      P-value H_0: RMSEA <= 0.050        0.007  
##      P-value H_0: RMSEA >= 0.080        0.991  
##  
## Standardized Root Mean Square Residual:  
##
```

```

##      SRMR                                0.352
##
## Parameter Estimates:
##
##      Standard errors                                Bootstrap
##      Number of requested bootstrap draws                    500
##      Number of successful bootstrap draws                    219
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      Trust ~
##      Conditn (a1) -0.792 0.426 -1.857 0.063 -1.602 0.126
##      Communication ~
##      Conditn (a2) -0.500 0.345 -1.448 0.148 -1.085 0.309
##      Chance_Perform ~
##      Conditn (a3) 0.021 0.292 0.071 0.943 -0.539 0.638
##      Consistency ~
##      Conditn (a4) -0.153 0.264 -0.578 0.563 -0.626 0.416
##      Org_Attraction ~
##      Trust (b1) -0.468 1.083 -0.432 0.666 -1.460 2.660
##      Cmmnctn (b2) 0.283 1.562 0.181 0.856 -4.050 2.054
##      Chnc_Pr (b3) 0.393 0.329 1.196 0.232 0.094 1.168
##      Cnsstnc (b4) -1.038 1.407 -0.738 0.461 -2.807 2.501
##      Conditn (c_pr) -0.326 0.652 -0.500 0.617 -1.254 0.369
##      Std.lv Std.all
##
##      -0.792 -0.462
##
##      -0.500 -0.363
##
##      0.021 0.015
##
##      -0.153 -0.148
##
##      -0.468 -0.527
##      0.283 0.256
##      0.393 0.347
##      -1.038 -0.703
##      -0.326 -0.214
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .Trust 0.555 0.152 3.662 0.000 0.185 0.758
##      .Communication 0.396 0.099 4.014 0.000 0.159 0.512
##      .Chance_Perform 0.435 0.154 2.822 0.005 0.059 0.654
##      .Consistency 0.250 0.056 4.447 0.000 0.110 0.328
##      .Org_Attraction 0.066 0.022 3.052 0.002 0.001 0.071
##      Std.lv Std.all
##      0.555 0.787
##      0.396 0.868
##      0.435 1.000
##      0.250 0.978
##      0.066 0.119
##

```

```
## Defined Parameters:
```

	Estimate	Std.Err	z-value	P(> z)	ci.lower	ci.upper
ind1	0.371	0.555	0.668	0.504	-1.407	1.335
ind2	-0.141	1.058	-0.134	0.894	-1.115	3.560
ind3	0.008	0.187	0.044	0.965	-0.282	0.509
ind4	0.159	0.513	0.309	0.757	-0.997	1.135
total_ind	0.396	0.650	0.609	0.543	-0.183	1.404
total	0.069	0.303	0.229	0.819	-0.556	0.684

	Std.lv	Std.all
0.371	0.243	
-0.141	-0.093	
0.008	0.005	
0.159	0.104	
0.396	0.260	
0.069	0.046	

```
# Define the mediation model for the relationship between condition and intent to apply with including
model <- '
```

```
  # Paths from IV to mediators
  Trust ~ a1*Condition
  Communication ~ a2*Condition
  Chance_Perform ~ a3*Condition
  Consistency ~ a4*Condition
  Org_Attraction ~ a5*Condition

  # Paths from mediators to DV
  Intent ~ b1*Trust + b2*Communication + b3*Chance_Perform + b4*Consistency + b5*Org_Attraction

  # Direct effect of IV on DV
  Intent ~ c_prime*Condition

  # Indirect effects
  ind1 := a1 * b1
  ind2 := a2 * b2
  ind3 := a3 * b3
  ind4 := a4 * b4
  ind5 := a5 * b5

  # Total indirect effect
  total_ind := ind1 + ind2 + ind3 + ind4 + ind5

  # Total effect
  total := c_prime + total_ind
,
```

```
# Fit the model
```

```
fit <- sem(model, data = JAR_Social_Invitees, se = "bootstrap", bootstrap = 500)
```

```
## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite
```

```
## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite
```


[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

## Warning in lav_model_nvcov_bootstrap(lavmodel = lavmodel, lavsamplestats =
## lavsamplestats, : lavaan WARNING: 422 bootstrap runs failed or did not
## converge.
```

```
# View summary with standardized estimates and confidence intervals
summary(fit, fit.measures = TRUE, standardized = TRUE, ci = TRUE)
```

```
## lavaan 0.6.16 ended normally after 7 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      17
##
##      Number of observations          10
##
## Model Test User Model:
##
##      Test statistic                 32.825
##      Degrees of freedom              10
##      P-value (Chi-square)            0.000
##
## Model Test Baseline Model:
##
##      Test statistic                 39.348
##      Degrees of freedom              21
##      P-value                         0.009
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.000
##      Tucker-Lewis Index (TLI)        -1.612
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)    -56.535
##      Loglikelihood unrestricted model (H1) -40.122
##
##      Akaike (AIC)                    147.069
##      Bayesian (BIC)                   152.213
##      Sample-size adjusted Bayesian (SABIC) 101.286
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                            0.478
```

```

## 90 Percent confidence interval - lower      0.302
## 90 Percent confidence interval - upper      0.664
## P-value H_0: RMSEA <= 0.050                0.000
## P-value H_0: RMSEA >= 0.080                0.999
##
## Standardized Root Mean Square Residual:
##
## SRMR                                         0.632
##
## Parameter Estimates:
##
## Standard errors                            Bootstrap
## Number of requested bootstrap draws          500
## Number of successful bootstrap draws          78
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## Trust ~
##   Conditn (a1)   -0.792   0.386  -2.050   0.040  -1.506   0.429
## Communication ~
##   Conditn (a2)   -0.500   0.295  -1.698   0.090  -1.048   0.059
## Chance_Perform ~
##   Conditn (a3)    0.021   0.245   0.085   0.932  -0.452   0.525
## Consistency ~
##   Conditn (a4)   -0.153   0.207  -0.737   0.461  -0.582   0.280
## Org_Attraction ~
##   Conditn (a5)    0.069   0.232   0.299   0.765  -0.402   0.639
## Intent ~
##   Trust (b1)    -0.950   4.393  -0.216   0.829 -18.059   5.299
##   Cmmnctn (b2)   0.758   4.417   0.172   0.864  -4.690  16.939
##   Chnc_Pr (b3)   0.385   2.504   0.154   0.878  -9.263   6.447
##   Cnsstnc (b4)  -1.602   7.810  -0.205   0.838 -30.035  12.528
##   Org_Att (b5)  -1.013   4.323  -0.234   0.815 -14.545   9.225
##   Conditn (c_pr) -0.285   2.876  -0.099   0.921  -9.385   5.200
##   Std.lv Std.all
##
##   -0.792  -0.462
##
##   -0.500  -0.363
##
##    0.021   0.015
##
##   -0.153  -0.148
##
##    0.069   0.063
##
##   -0.950  -0.535
##    0.758   0.343
##    0.385   0.170
##   -1.602  -0.543
##   -1.013  -0.366
##   -0.285  -0.094
##
## Variances:

```

```
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .Trust          0.555   0.119   4.675   0.000   0.093   0.676
## .Communication   0.396   0.079   4.991   0.000   0.200   0.491
## .Chance_Perform  0.435   0.121   3.606   0.000   0.165   0.647
## .Consistency     0.250   0.047   5.308   0.000   0.133   0.339
## .Org_Attraction  0.289   0.064   4.502   0.000   0.150   0.416
## .Intent          0.474   0.141   3.361   0.001   0.015   0.566
## Std.lv Std.all
## 0.555 0.787
## 0.396 0.868
## 0.435 1.000
## 0.250 0.978
## 0.289 0.996
## 0.474 0.213
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## ind1          0.752   3.314   0.227   0.820  -6.209  12.112
## ind2         -0.379   3.363  -0.113   0.910 -12.897   3.270
## ind3          0.008   0.821   0.010   0.992  -2.679   1.544
## ind4          0.245   3.128   0.078   0.938  -2.280  13.196
## ind5         -0.070   1.028  -0.068   0.945  -1.852   3.215
## total_ind      0.555   2.888   0.192   0.847  -4.489   9.581
## total          0.271   0.367   0.739   0.460  -0.443   0.900
## Std.lv Std.all
## 0.752 0.247
## -0.379 -0.125
## 0.008 0.003
## 0.245 0.080
## -0.070 -0.023
## 0.555 0.182
## 0.271 0.089
```

```
# Define the mediation model for the relationship between condition and intent to apply without org att
model <- '
```

```
  # Paths from IV to mediators
  Trust ~ a1*Condition
  Communication ~ a2*Condition
  Chance_Perform ~ a3*Condition
  Consistency ~ a4*Condition

  # Paths from mediators to DV
  Intent ~ b1*Trust + b2*Communication + b3*Chance_Perform + b4*Consistency

  # Direct effect of IV on DV
  Intent ~ c_prime*Condition

  # Indirect effects
  ind1 := a1 * b1
  ind2 := a2 * b2
  ind3 := a3 * b3
  ind4 := a4 * b4

  # Total indirect effect
```


[illegible]

[illegible]

[illegible]


```
## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

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## Warning in lav_samplestats_icov(COV = cov[[g]], ridge = 1e-05, x.idx =
## x.idx[[g]], : lavaan WARNING: sample covariance matrix is not positive-definite

## Warning in lav_model_nvcov_bootstrap(lavmodel = lavmodel, lavsamplestats =
## lavsamplestats, : lavaan WARNING: 273 bootstrap runs failed or did not
## converge.
```

```
# View summary with standardized estimates and confidence intervals
summary(fit, fit.measures = TRUE, standardized = TRUE, ci = TRUE)
```

```
## lavaan 0.6.16 ended normally after 7 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      14
##
##      Number of observations          10
##
## Model Test User Model:
##
##      Test statistic                  18.108
##      Degrees of freedom                6
##      P-value (Chi-square)             0.006
##
## Model Test Baseline Model:
##
##      Test statistic                  23.249
##      Degrees of freedom              15
##      P-value                         0.079
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.000
##      Tucker-Lewis Index (TLI)        -2.669
##
## Loglikelihood and Information Criteria:
```

```

## Loglikelihood user model (H0) -49.221
## Loglikelihood unrestricted model (H1) -40.167
##
## Akaike (AIC) 126.442
## Bayesian (BIC) 130.678
## Sample-size adjusted Bayesian (SABIC) 88.738
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.449
## 90 Percent confidence interval - lower 0.221
## 90 Percent confidence interval - upper 0.694
## P-value H_0: RMSEA <= 0.050 0.007
## P-value H_0: RMSEA >= 0.080 0.991
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.234
##
## Parameter Estimates:
##
## Standard errors Bootstrap
## Number of requested bootstrap draws 500
## Number of successful bootstrap draws 227
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## Trust ~
## Conditn (a1) -0.792 0.466 -1.700 0.089 -1.656 0.283
## Communication ~
## Conditn (a2) -0.500 0.339 -1.473 0.141 -1.177 0.188
## Chance_Perform ~
## Conditn (a3) 0.021 0.339 0.061 0.951 -0.687 0.675
## Consistency ~
## Conditn (a4) -0.153 0.276 -0.553 0.580 -0.710 0.407
## Intent ~
## Trust (b1) -0.476 4.021 -0.118 0.906 -9.789 1.732
## Cmmnctn (b2) 0.472 5.639 0.084 0.933 -4.515 14.858
## Chnc_Pr (b3) -0.014 1.073 -0.013 0.990 -1.685 2.666
## Cnsstnc (b4) -0.551 4.815 -0.114 0.909 -13.587 4.187
## Conditn (c_pr) 0.046 2.259 0.020 0.984 -2.785 5.271
## Std.lv Std.all
##
## -0.792 -0.462
##
## -0.500 -0.363
##
## 0.021 0.015
##
## -0.153 -0.148
##
## -0.476 -0.434
## 0.472 0.346
## -0.014 -0.010

```

```

##      -0.551    -0.302
##      0.046     0.024
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .Trust        0.555   0.157   3.543   0.000   0.102   0.779
##      .Communication 0.396   0.099   4.013   0.000   0.161   0.521
##      .Chance_Perform 0.435   0.155   2.806   0.005   0.062   0.629
##      .Consistency   0.250   0.054   4.636   0.000   0.124   0.339
##      .Intent        0.542   0.168   3.225   0.001   0.005   0.560
##      Std.lv Std.all
##      0.555   0.787
##      0.396   0.868
##      0.435   1.000
##      0.250   0.978
##      0.542   0.638
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      ind1         0.377   2.809   0.134   0.893   -2.617   5.152
##      ind2        -0.236   4.987  -0.047   0.962  -14.697   2.474
##      ind3        -0.000   0.476  -0.001   1.000   -0.685   1.767
##      ind4         0.084   1.757   0.048   0.962   -1.781   6.322
##      total_ind     0.225   2.183   0.103   0.918   -4.458   3.358
##      total         0.271   0.466   0.582   0.561   -0.631   1.253
##      Std.lv Std.all
##      0.377   0.200
##      -0.236  -0.125
##      -0.000  -0.000
##      0.084   0.045
##      0.225   0.120
##      0.271   0.144

```