

Main Analyses

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Import Packages

```
library(ppcor)

## Loading required package: MASS
library(lavaan)

## This is lavaan 0.6-6
## lavaan is BETA software! Please report any bugs.
```

Import Data

```
source("Data-Processing.R")

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr   0.3.4
## v tibble  3.1.2      v dplyr  1.0.6
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   2.0.2      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## x dplyr::select() masks MASS::select()
```

Linear Regression of Substantive Predictors on MRAI-R Total Scores

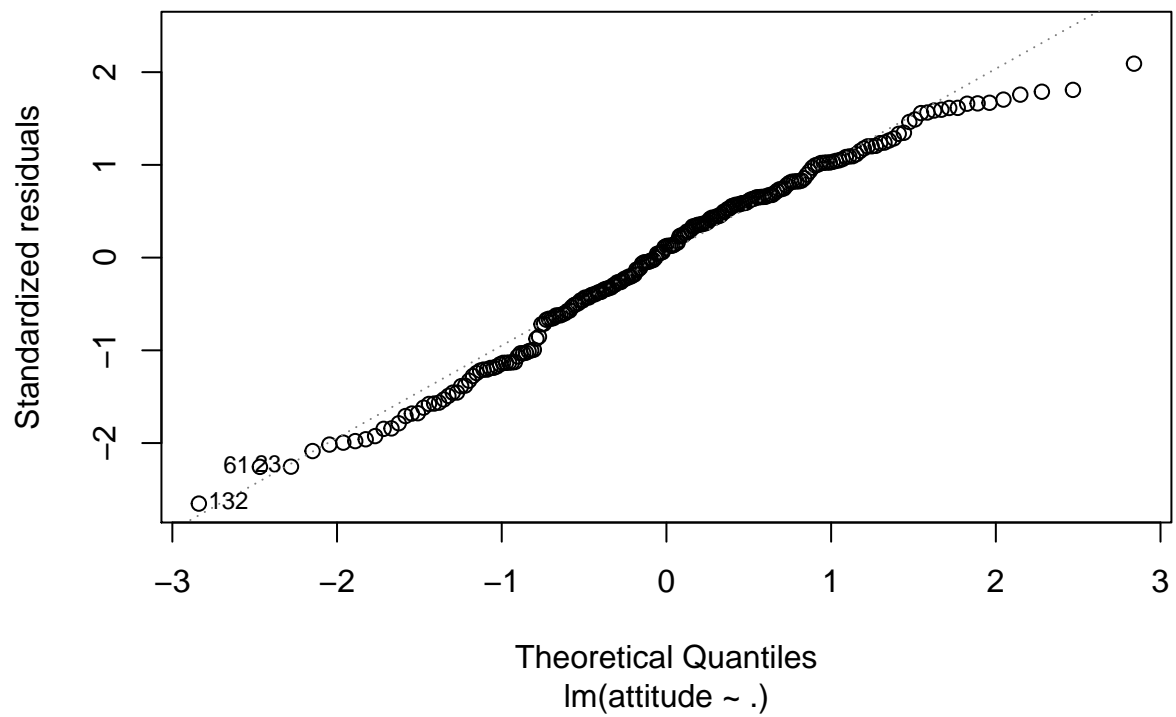
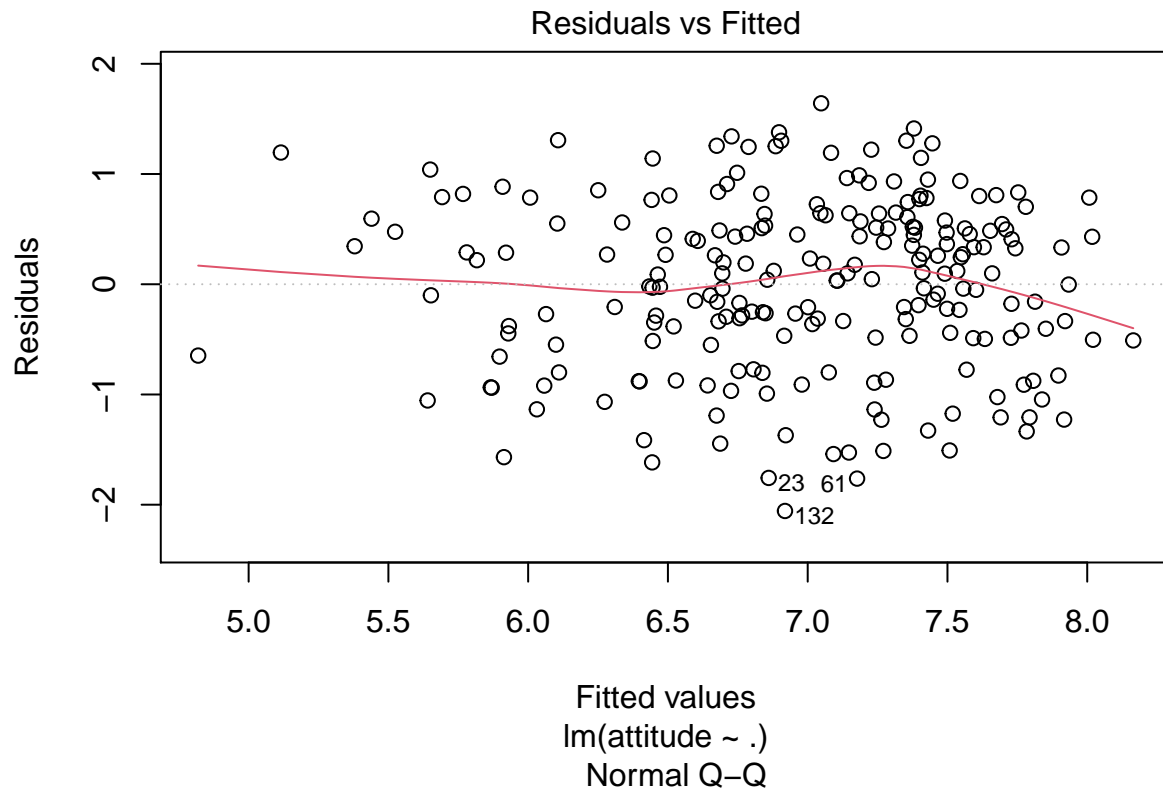
We will start by subsetting the data to only include the predictors, then fitting the model.

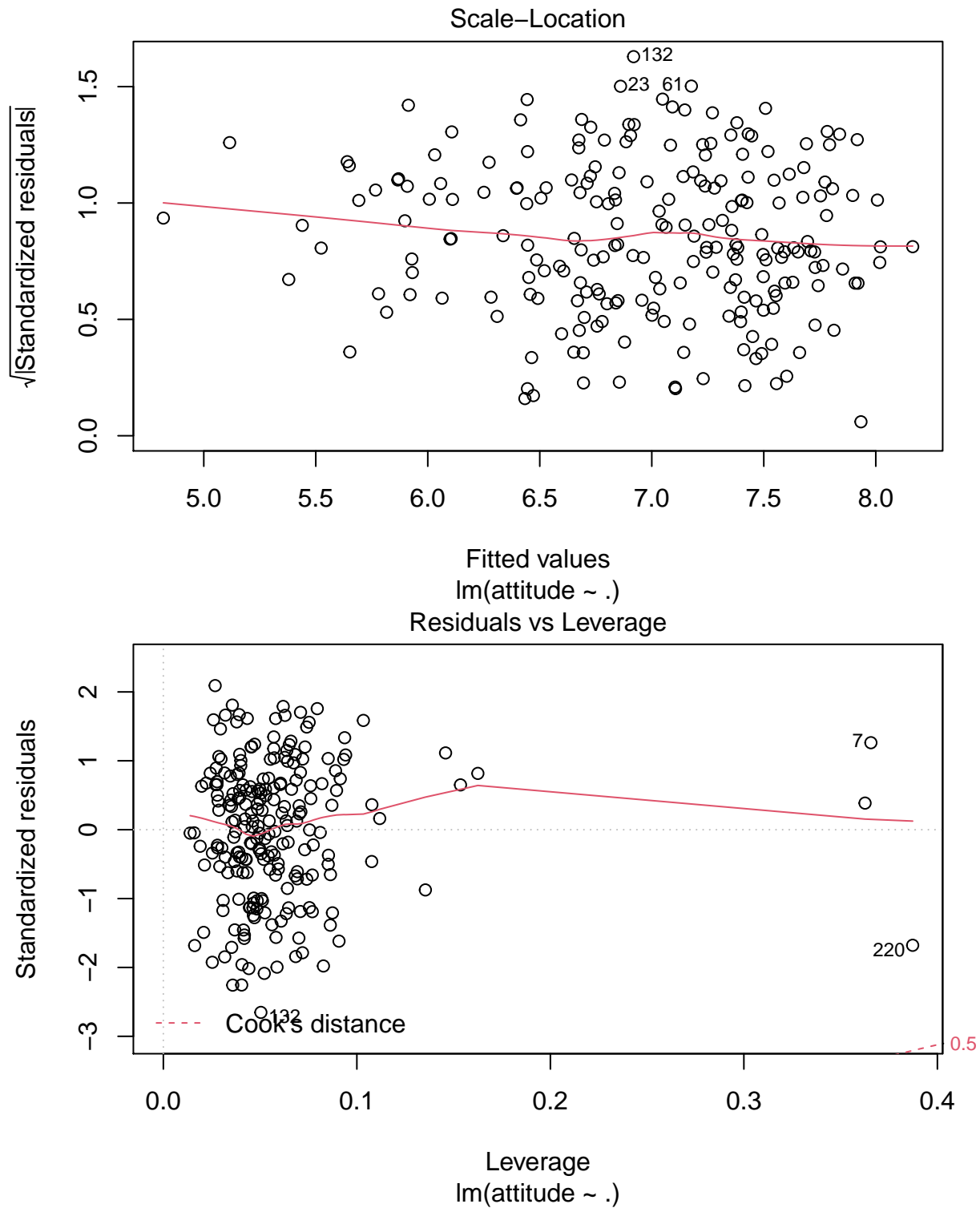
```
regression.model.df <- analysis.df %>%
  select(attitude, gender.male, gender.non.binary, non.white, social.desirability.c,
         quantity.c, quality.c, knowledge.c, openness.c,
         conscientiousness.c, extroversion.c, agreeableness.c, neuroticism.c)

regression.model <- lm(attitude ~ ., data = regression.model.df)
```

Plot the model assumptions.

```
plot(regression.model)
```





The model assumptions appear to be reasonable. Now we can look at the model coefficients.

```
summary(regression.model)
```

```
##
## Call:
## lm(formula = attitude ~ ., data = regression.model.df)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.05665 -0.48885  0.09568  0.55091  1.64153
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      7.09139    0.07688  92.240 < 2e-16 ***
## gender.male     -0.33810    0.12974  -2.606  0.00982 **
## gender.non.binary  0.58690    0.46982   1.249  0.21300
## non.white      -0.09764    0.12457  -0.784  0.43404
## social.desirability.c -0.01508    0.01322  -1.140  0.25549
## quantity.c      0.03092    0.04917   0.629  0.53012
## quality.c       0.20635    0.04530   4.555 8.92e-06 ***
## knowledge.c     0.08732    0.04625   1.888  0.06042 .
## openness.c      0.22093    0.09681   2.282  0.02350 *
## conscientiousness.c 0.04258    0.10197   0.418  0.67668
## extroversion.c  -0.05392    0.07104  -0.759  0.44874
## agreeableness.c  0.23377    0.11818   1.978  0.04923 *
## neuroticism.c   -0.08222    0.08050  -1.021  0.30827
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7956 on 208 degrees of freedom
## Multiple R-squared:  0.3991, Adjusted R-squared:  0.3645
## F-statistic: 11.51 on 12 and 208 DF,  p-value: < 2.2e-16
```

Let's look at the standardized predictors to see the relative strength of each predictor.

```
# create a temporary DataFrame to standardize quantitative predictors
temp2 <- df %>%
  select(social.desirability, openness, conscientiousness, extroversion,
         agreeableness, neuroticism, quantity, quality,
         knowledge) %>%
  scale(center=TRUE, scale=TRUE)

# add a .s after the name of each predictor to identify it as standardized
temp2 <- as.data.frame(temp2, row.names = NULL)
temp2 <- set_names(temp2, paste0(names(temp2), ".s"))

# merge dataframes and recode gender and ethnicity
standardized.df <- df %>%
  mutate(male = recode(gender,
                      "male" = 1,
                      .default = 0)) %>%
  mutate(non.white = recode(ethnicity,
                           "white" = 0,
                           .default = 1)) %>%
  select(gender, attitude, male, non.white) %>%
  bind_cols(temp2) %>%
  filter(gender != "non-binary") %>%
  drop_na(attitude, social.desirability.s, openness.s,
          conscientiousness.s, extroversion.s, agreeableness.s, neuroticism.s,
          quantity.s, quality.s, knowledge.s)
```

```
# create the new dataframes with the standardized predictors
standardized.regression.model.df <- standardized.df %>%
  select(attitude, male, non.white, social.desirability.s,
         quantity.s, quality.s, knowledge.s, openness.s,
         conscientiousness.s, extroversion.s, agreeableness.s, neuroticism.s)

# fit the model
standardized.regression.model <- lm(attitude ~ ., data = standardized.regression.model.df)

# examine model coefficients
summary(standardized.regression.model)
```

```
##
## Call:
## lm(formula = attitude ~ ., data = standardized.regression.model.df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.04482 -0.49297  0.08601  0.53739  1.66613
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      7.09667    0.07682   92.382 < 2e-16 ***
## male            -0.34331    0.12945   -2.652  0.00862 **
## non.white        -0.11135    0.12478   -0.892  0.37324
## social.desirability.s -0.09157    0.06999   -1.308  0.19225
## quantity.s        0.05541    0.07987    0.694  0.48858
## quality.s         0.35093    0.07609    4.612 6.99e-06 ***
## knowledge.s       0.15103    0.08054    1.875  0.06220 .
## openness.s        0.11466    0.05705    2.010  0.04575 *
## conscientiousness.s  0.03062    0.06252    0.490  0.62476
## extroversion.s    -0.03078    0.05909   -0.521  0.60298
## agreeableness.s    0.12891    0.06858    1.880  0.06155 .
## neuroticism.s     -0.08262    0.06511   -1.269  0.20587
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7935 on 206 degrees of freedom
## Multiple R-squared:  0.4004, Adjusted R-squared:  0.3684
## F-statistic: 12.51 on 11 and 206 DF,  p-value: < 2.2e-16
```

Finally, we can look at the partial correlations for each predictor.

```
pcor(regression.model.df)

## $estimate
##              attitude  gender.male  gender.non.binary  non.white
## attitude            1.00000000 -0.177817095      0.086293464 -0.054267809
## gender.male          -0.17781710  1.000000000      -0.051243075 -0.137647812
## gender.non.binary     0.08629346 -0.051243075      1.000000000  0.002984958
## non.white            -0.05426781 -0.137647812      0.002984958  1.000000000
## social.desirability.c -0.07881638  0.005813741      0.006788205  0.139547535
## quantity.c            0.04356358  0.007778381      0.082867461  0.030428910
## quality.c             0.30115880 -0.033022116     -0.047261817  0.118887386
## knowledge.c           0.12979835  0.020122008     -0.061628120 -0.088398145
```

```

## openness.c          0.15628576  0.036506203          0.032631675  0.189206306
## conscientiousness.c 0.02894259 -0.026357603          0.089497351 -0.160674926
## extroversion.c      -0.05255069 -0.076987884          -0.045846744 -0.056144305
## agreeableness.c     0.13588795 -0.195103005          -0.061753101 -0.182448349
## neuroticism.c       -0.07064204 -0.221731831          0.051532256 -0.131327745
##
## social.desirability.c quantity.c quality.c
## attitude            -0.078816384  0.043563583  0.30115880
## gender.male          0.005813741  0.007778381 -0.03302212
## gender.non.binary    0.006788205  0.082867461 -0.04726182
## non.white            0.139547535  0.030428910  0.11888739
## social.desirability.c 1.000000000  0.025521327  0.05288116
## quantity.c           0.025521327  1.000000000  0.24760914
## quality.c            0.052881161  0.247609140  1.000000000
## knowledge.c          0.107604906  0.493295054  0.26474359
## openness.c           0.035442823  0.108849815 -0.05002609
## conscientiousness.c  0.301923099 -0.018417521  0.06652677
## extroversion.c       0.057070406  0.120143346  0.03495130
## agreeableness.c      0.248128096 -0.019551343  0.23062714
## neuroticism.c        -0.265572568  0.118011647  0.08045430
##
## knowledge.c openness.c conscientiousness.c
## attitude            0.12979835  0.156285758          0.02894259
## gender.male          0.02012201  0.036506203          -0.02635760
## gender.non.binary    -0.06162812  0.032631675          0.08949735
## non.white            -0.08839815  0.189206306          -0.16067493
## social.desirability.c 0.10760491  0.035442823          0.30192310
## quantity.c           0.49329505  0.108849815          -0.01841752
## quality.c            0.26474359 -0.050026092          0.06652677
## knowledge.c          1.000000000  0.021581194          -0.07954737
## openness.c           0.02158119  1.000000000          -0.04555959
## conscientiousness.c -0.07954737 -0.045559587          1.000000000
## extroversion.c       0.04150741  0.039185295          0.12843826
## agreeableness.c      -0.12867521  0.069585802          0.08054791
## neuroticism.c        -0.05383948  0.005454746          -0.08832788
##
## extroversion.c agreeableness.c neuroticism.c
## attitude            -0.05255069  0.13588795 -0.070642041
## gender.male          -0.07698788 -0.19510300 -0.221731831
## gender.non.binary    -0.04584674 -0.06175310 0.051532256
## non.white            -0.05614431 -0.18244835 -0.131327745
## social.desirability.c 0.05707041  0.24812810 -0.265572568
## quantity.c           0.12014335 -0.01955134 0.118011647
## quality.c            0.03495130  0.23062714 0.080454300
## knowledge.c          0.04150741 -0.12867521 -0.053839478
## openness.c           0.03918530  0.06958580 0.005454746
## conscientiousness.c  0.12843826  0.08054791 -0.088327881
## extroversion.c       1.000000000  0.01775349 -0.153895940
## agreeableness.c      0.01775349  1.000000000 -0.246659075
## neuroticism.c        -0.15389594 -0.24665907 1.000000000
##
## $p.value
##
## attitude gender.male gender.non.binary non.white
## attitude 0.000000e+00 0.009821699          0.2129952 0.434039446
## gender.male 9.821699e-03 0.000000000          0.4601281 0.046338829
## gender.non.binary 2.129952e-01 0.460128145          0.0000000 0.965703081
## non.white 4.340394e-01 0.046338829          0.9657031 0.000000000

```

```

## social.desirability.c 2.554921e-01 0.933257628      0.9221035 0.043377113
## quantity.c          5.301177e-01 0.910785075      0.2317927 0.661077218
## quality.c           8.924596e-06 0.634210122      0.4957566 0.085676424
## knowledge.c         6.042485e-02 0.771904610      0.3742231 0.202003613
## openness.c          2.349947e-02 0.598858385      0.6382260 0.005952589
## conscientiousness.c 6.766770e-01 0.704135351      0.1964290 0.019824129
## extroversion.c      4.487444e-01 0.266718780      0.5087634 0.418290435
## agreeableness.c     4.923150e-02 0.004543167      0.3732532 0.008038932
## neuroticism.c       3.082677e-01 0.001219084      0.4575969 0.057433412
##
## social.desirability.c  quantity.c    quality.c
## attitude                2.554921e-01 5.301177e-01 8.924596e-06
## gender.male              9.332576e-01 9.107851e-01 6.342101e-01
## gender.non.binary        9.221035e-01 2.317927e-01 4.957566e-01
## non.white                4.337711e-02 6.610772e-01 8.567642e-02
## social.desirability.c    0.000000e+00 7.131032e-01 4.458927e-01
## quantity.c              7.131032e-01 0.000000e+00 2.907923e-04
## quality.c                4.458927e-01 2.907923e-04 0.000000e+00
## knowledge.c             1.200526e-01 2.810278e-14 1.031270e-04
## openness.c              6.095540e-01 1.158010e-01 4.708655e-01
## conscientiousness.c     8.446587e-06 7.907593e-01 3.373706e-01
## extroversion.c          4.106426e-01 8.239640e-02 6.145262e-01
## agreeableness.c         2.820979e-04 7.782029e-01 7.580610e-04
## neuroticism.c           9.790395e-05 8.802486e-02 2.457162e-01
##
## knowledge.c  openness.c  conscientiousness.c
## attitude    6.042485e-02 0.023499469      6.766770e-01
## gender.male 7.719046e-01 0.598858385      7.041354e-01
## gender.non.binary 3.742231e-01 0.638225950      1.964290e-01
## non.white      2.020036e-01 0.005952589      1.982413e-02
## social.desirability.c 1.200526e-01 0.609554022      8.446587e-06
## quantity.c      2.810278e-14 0.115801038      7.907593e-01
## quality.c       1.031270e-04 0.470865486      3.373706e-01
## knowledge.c     0.000000e+00 0.755868472      2.510965e-01
## openness.c      7.558685e-01 0.000000000      5.114243e-01
## conscientiousness.c 2.510965e-01 0.511424312      0.000000e+00
## extroversion.c   5.497284e-01 0.572293842      6.319173e-02
## agreeableness.c  6.270234e-02 0.315575164      2.451654e-01
## neuroticism.c    4.376814e-01 0.937370236      2.023638e-01
##
## extroversion.c  agreeableness.c  neuroticism.c
## attitude        0.44874435      0.0492314991 3.082677e-01
## gender.male      0.26671878      0.0045431673 1.219084e-03
## gender.non.binary 0.50876337      0.3732531572 4.575969e-01
## non.white        0.41829043      0.0080389325 5.743341e-02
## social.desirability.c 0.41064263      0.0002820979 9.790395e-05
## quantity.c        0.08239640      0.7782029254 8.802486e-02
## quality.c          0.61452619      0.0007580610 2.457162e-01
## knowledge.c        0.54972841      0.0627023441 4.376814e-01
## openness.c         0.57229384      0.3155751639 9.373702e-01
## conscientiousness.c 0.06319173      0.2451654194 2.023638e-01
## extroversion.c     0.00000000      0.7981384274 2.573714e-02
## agreeableness.c    0.79813843      0.0000000000 3.073583e-04
## neuroticism.c      0.02573714      0.0003073583 0.000000e+00
##
## $statistic
##
attitude gender.male gender.non.binary non.white

```

## attitude	0.0000000	-2.60604572	1.24920186	-0.78381648
## gender.male	-2.6060457	0.00000000	-0.74001035	-2.00426308
## gender.non.binary	1.2492019	-0.74001035	0.00000000	0.04304987
## non.white	-0.7838165	-2.00426308	0.04304987	0.00000000
## social.desirability.c	-1.1402532	0.08384838	0.09790314	2.03247013
## quantity.c	0.6288800	0.11218479	1.19925627	0.43905530
## quality.c	4.5548368	-0.47651162	-0.68238215	1.72686563
## knowledge.c	1.8879497	0.29026249	-0.89050608	-1.27990673
## openness.c	2.2820271	0.52685113	0.47087148	2.77896765
## conscientiousness.c	0.4175909	-0.38026688	1.29594972	-2.34779069
## extroversion.c	-0.7589455	-1.11364031	-0.66190715	-0.81100391
## agreeableness.c	1.9781528	-2.86894883	-0.89231892	-2.67622679
## neuroticism.c	-1.0213656	-3.27949620	0.74419756	-1.91058323
##	social.desirability.c	quantity.c	quality.c	knowledge.c
## attitude	-1.14025321	0.6288800	4.5548368	1.8879497
## gender.male	0.08384838	0.1121848	-0.4765116	0.2902625
## gender.non.binary	0.09790314	1.1992563	-0.6823822	-0.8905061
## non.white	2.03247013	0.4390553	1.7268656	-1.2799067
## social.desirability.c	0.00000000	0.3681937	0.7637316	1.5609634
## quantity.c	0.36819374	0.0000000	3.6858471	8.1787709
## quality.c	0.76373155	3.6858471	0.0000000	3.9594646
## knowledge.c	1.56096339	8.1787709	3.9594646	0.0000000
## openness.c	0.51148502	1.5792379	-0.7223911	0.3113209
## conscientiousness.c	4.56755418	-0.2656663	0.9615930	-1.1508955
## extroversion.c	0.82442478	1.7453745	0.5043829	0.5991447
## agreeableness.c	3.69407834	-0.2820274	3.4183018	-1.8713371
## neuroticism.c	-3.97280189	1.7139650	1.1641021	-0.7776118
##	openness.c	conscientiousness.c	extroversion.c	
## attitude	2.28202708	0.4175909	-0.7589455	
## gender.male	0.52685113	-0.3802669	-1.1136403	
## gender.non.binary	0.47087148	1.2959497	-0.6619071	
## non.white	2.77896765	-2.3477907	-0.8110039	
## social.desirability.c	0.51148502	4.5675542	0.8244248	
## quantity.c	1.57923786	-0.2656663	1.7453745	
## quality.c	-0.72239106	0.9615930	0.5043829	
## knowledge.c	0.31132091	-1.1508955	0.5991447	
## openness.c	0.00000000	-0.6577527	0.5655727	
## conscientiousness.c	-0.65775270	0.0000000	1.8678333	
## extroversion.c	0.56557275	1.8678333	0.0000000	
## agreeableness.c	1.00601932	1.1654654	0.2560848	
## neuroticism.c	0.07867063	-1.2788814	-2.2462786	
##	agreeableness.c	neuroticism.c		
## attitude	1.9781528	-1.02136564		
## gender.male	-2.8689488	-3.27949620		
## gender.non.binary	-0.8923189	0.74419756		
## non.white	-2.6762268	-1.91058323		
## social.desirability.c	3.6940783	-3.97280189		
## quantity.c	-0.2820274	1.71396500		
## quality.c	3.4183018	1.16410209		
## knowledge.c	-1.8713371	-0.77761184		
## openness.c	1.0060193	0.07867063		
## conscientiousness.c	1.1654654	-1.27888138		
## extroversion.c	0.2560848	-2.24627860		
## agreeableness.c	0.0000000	-3.67078656		


```
## neuroticism.c          -3.6707866    0.00000000
##
## $n
## [1] 221
##
## $gp
## [1] 11
##
## $method
## [1] "pearson"
```

Mediation Model

We will create the mediation model and fit it to the data.

```
# set seed for reproducibility
set.seed(760120)

med_model <- '
  # direct effects
  quality ~ b1 * openness + b2 * agreeableness
  attitude ~ b3 * openness + b4 * agreeableness + b5 * quality
  # indirect effects
  indirect_openness := b1 * b5
  indirect_agreeableness := b2 * b5
  # total effects
  total_openness := b3 + (b1 * b5)
  total_agreeableness := b4 + (b2 * b5)
  '

med_model_fit <- sem(med_model, data = analysis.df,
                     se = "bootstrap", bootstrap = 10000)
```

Examine the overall model fit.

```
inspect(med_model_fit, "r2")
```

```
## quality attitude
## 0.173 0.349
```

Examine the unstandardized model coefficients.

```
summary(med_model_fit, fit.measures=TRUE, ci=TRUE)
```

```
## lavaan 0.6-6 ended normally after 23 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of free parameters 7
##
## Number of observations 221
##
## Model Test User Model:
##
## Test statistic 0.000
## Degrees of freedom 0
```

```

##
## Model Test Baseline Model:
##
##   Test statistic          136.912
##   Degrees of freedom      5
##   P-value                 0.000
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)          1.000
##   Tucker-Lewis Index (TLI)            1.000
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)        -673.443
##   Loglikelihood unrestricted model (H1) -673.443
##
##   Akaike (AIC)                       1360.885
##   Bayesian (BIC)                     1384.673
##   Sample-size adjusted Bayesian (BIC) 1362.489
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                          0.000
##   90 Percent confidence interval - lower 0.000
##   90 Percent confidence interval - upper 0.000
##   P-value RMSEA <= 0.05              NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                          0.000
##
## Parameter Estimates:
##
##   Standard errors          Bootstrap
##   Number of requested bootstrap draws 10000
##   Number of successful bootstrap draws 10000
##
## Regressions:
##
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   quality ~
##     openness (b1)   0.410   0.186   2.211   0.027   0.036   0.763
##     agreblnss (b2)  1.075   0.175   6.144   0.000   0.745   1.432
##   attitude ~
##     openness (b3)   0.243   0.087   2.782   0.005   0.069   0.412
##     agreblnss (b4)  0.244   0.097   2.516   0.012   0.055   0.433
##     quality (b5)   0.279   0.036   7.749   0.000   0.216   0.356
##
## Variances:
##
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   .quality      2.355   0.350   6.725   0.000   1.680   3.071
##   .attitude     0.646   0.051  12.576   0.000   0.535   0.739
##
## Defined Parameters:

```

	Estimate	Std.Err	z-value	P(> z)	ci.lower	ci.upper
indirect_pnnss	0.114	0.056	2.031	0.042	0.009	0.234
indrct_grblnss	0.299	0.057	5.262	0.000	0.197	0.418
total_openness	0.357	0.102	3.485	0.000	0.149	0.555
total_agrblnss	0.544	0.106	5.129	0.000	0.330	0.751

Examine the standardized model coefficients.

```
summary(med_model_fit, fit.measures=TRUE, ci=TRUE, standardized = TRUE)
```

```
## lavaan 0.6-6 ended normally after 23 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of free parameters 7
##
## Number of observations 221
##
## Model Test User Model:
##
## Test statistic 0.000
## Degrees of freedom 0
##
## Model Test Baseline Model:
##
## Test statistic 136.912
## Degrees of freedom 5
## P-value 0.000
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 1.000
## Tucker-Lewis Index (TLI) 1.000
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -673.443
## Loglikelihood unrestricted model (H1) -673.443
##
## Akaike (AIC) 1360.885
## Bayesian (BIC) 1384.673
## Sample-size adjusted Bayesian (BIC) 1362.489
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.000
## 90 Percent confidence interval - lower 0.000
## 90 Percent confidence interval - upper 0.000
## P-value RMSEA <= 0.05 NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.000
##
## Parameter Estimates:
```

```

##
## Standard errors                                Bootstrap
## Number of requested bootstrap draws            10000
## Number of successful bootstrap draws            10000
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## quality ~
##   openness (b1)    0.410   0.186   2.211   0.027   0.036   0.763
##   agreblnss (b2)   1.075   0.175   6.144   0.000   0.745   1.432
## attitude ~
##   openness (b3)    0.243   0.087   2.782   0.005   0.069   0.412
##   agreblnss (b4)   0.244   0.097   2.516   0.012   0.055   0.433
##   quality (b5)     0.279   0.036   7.749   0.000   0.216   0.356
## Std.lv Std.all
##
##   0.410   0.143
##   1.075   0.372
##
##   0.243   0.143
##   0.244   0.143
##   0.279   0.472
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   .quality         2.355   0.350   6.725   0.000   1.680   3.071
##   .attitude         0.646   0.051  12.576   0.000   0.535   0.739
## Std.lv Std.all
##   2.355   0.827
##   0.646   0.651
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect_pnnss     0.114   0.056   2.031   0.042   0.009   0.234
## indrcr_grblnss     0.299   0.057   5.262   0.000   0.197   0.418
## total_openness     0.357   0.102   3.485   0.000   0.149   0.555
## total_agrblnss     0.544   0.106   5.129   0.000   0.330   0.751
## Std.lv Std.all
##   0.114   0.067
##   0.299   0.176
##   0.357   0.210
##   0.544   0.319

```