

# hw15

111078513

Help By 108078467

```
library(semnr)
sec_q <- read.table("security_data_sem.csv", header = TRUE, sep = ",")
```

## Question 1) Composite Path Models using PLS-PM

a. Create a PLS path model using SEMinR, with all the following characteristics:

(i) Measurement model – all constructs are measured as composites:

```
sec_q_intxn_mm <- constructs(
  composite("TRUST", multi_items("TRST", 1:4)),
  composite("SEC", multi_items("PSEC", 1:4)),
  composite("REP", multi_items("PREP", 1:4)),
  composite("INV", multi_items("PINV", 1:3)),
  composite("POL", multi_items("PPSS", 1:3)),
  composite("FAML", single_item("FAML1")),
  interaction_term(iv="REP", moderator="POL", method=orthogonal)
)
```

(ii). Structural Model – paths between constructs as shown in this causal model:

```
sec_q_intxn_sm <- relationships(
  paths(from = c("REP", "POL", "REP*POL"), to = "SEC"),
  paths(from = "SEC", to = "TRUST")
)
```

b. Show us the following results in table or figure formats:

(i) Plot a figure of the estimated model

```
sec_q_intxn_pls <- estimate_pls(data = sec_q,
                                measurement_model = sec_q_intxn_mm,
                                structural_model = sec_q_intxn_sm)
```

```
## Generating the semnr model
```

```
## All 405 observations are valid.
```

```
plot(sec_q_intxn_pls)
```

## (ii) Weights and loadings of composites

```
sec_q_report <- summary(sec_q_intxn_pls)
```

```
# weight
```

```
sec_q_report$weights
```

##	REP	POL	REP*POL	SEC	TRUST
## TRST1	0.000	0.000	0.000	0.000	0.282
## TRST2	0.000	0.000	0.000	0.000	0.280
## TRST3	0.000	0.000	0.000	0.000	0.286
## TRST4	0.000	0.000	0.000	0.000	0.278
## PSEC1	0.000	0.000	0.000	0.279	0.000
## PSEC2	0.000	0.000	0.000	0.313	0.000
## PSEC3	0.000	0.000	0.000	0.308	0.000
## PSEC4	0.000	0.000	0.000	0.290	0.000
## PREP1	0.215	0.000	0.000	0.000	0.000
## PREP2	0.334	0.000	0.000	0.000	0.000
## PREP3	0.349	0.000	0.000	0.000	0.000
## PREP4	0.287	0.000	0.000	0.000	0.000
## PPSS1	0.000	0.360	0.000	0.000	0.000
## PPSS2	0.000	0.395	0.000	0.000	0.000
## PPSS3	0.000	0.367	0.000	0.000	0.000
## PREP1*PPSS1	0.000	0.000	0.239	0.000	0.000
## PREP1*PPSS2	0.000	0.000	0.031	0.000	0.000
## PREP1*PPSS3	0.000	0.000	0.021	0.000	0.000
## PREP2*PPSS1	0.000	0.000	0.045	0.000	0.000
## PREP2*PPSS2	0.000	0.000	-0.105	0.000	0.000
## PREP2*PPSS3	0.000	0.000	-0.229	0.000	0.000
## PREP3*PPSS1	0.000	0.000	-0.342	0.000	0.000
## PREP3*PPSS2	0.000	0.000	0.095	0.000	0.000
## PREP3*PPSS3	0.000	0.000	0.108	0.000	0.000
## PREP4*PPSS1	0.000	0.000	0.443	0.000	0.000
## PREP4*PPSS2	0.000	0.000	0.383	0.000	0.000
## PREP4*PPSS3	0.000	0.000	0.272	0.000	0.000

```
# loadings
```

```
sec_q_report$loadings
```

##	REP	POL	REP*POL	SEC	TRUST
## TRST1	0.000	0.000	-0.000	0.000	0.900
## TRST2	0.000	0.000	-0.000	0.000	0.909
## TRST3	0.000	0.000	-0.000	0.000	0.905
## TRST4	0.000	0.000	-0.000	0.000	0.838
## PSEC1	0.000	0.000	-0.000	0.814	0.000
## PSEC2	0.000	0.000	-0.000	0.865	0.000
## PSEC3	0.000	0.000	-0.000	0.868	0.000
## PSEC4	0.000	0.000	-0.000	0.806	0.000
## PREP1	0.800	0.000	0.000	0.000	0.000
## PREP2	0.913	0.000	0.000	0.000	0.000
## PREP3	0.908	0.000	0.000	0.000	0.000
## PREP4	0.718	0.000	0.000	0.000	0.000
## PPSS1	0.000	0.868	0.000	0.000	0.000
## PPSS2	0.000	0.893	0.000	0.000	0.000
## PPSS3	0.000	0.911	0.000	0.000	0.000

```
## PREP1*PPSS1 -0.000 -0.000 0.579 -0.000 -0.000
## PREP1*PPSS2 -0.000 -0.000 0.509 -0.000 -0.000
## PREP1*PPSS3 -0.000 -0.000 0.504 -0.000 -0.000
## PREP2*PPSS1 -0.000 -0.000 0.507 -0.000 -0.000
## PREP2*PPSS2 -0.000 0.000 0.419 0.000 0.000
## PREP2*PPSS3 -0.000 -0.000 0.333 0.000 0.000
## PREP3*PPSS1 -0.000 -0.000 0.234 0.000 0.000
## PREP3*PPSS2 -0.000 0.000 0.553 -0.000 -0.000
## PREP3*PPSS3 -0.000 -0.000 0.464 -0.000 -0.000
## PREP4*PPSS1 0.000 0.000 0.899 -0.000 -0.000
## PREP4*PPSS2 -0.000 -0.000 0.836 -0.000 0.000
## PREP4*PPSS3 0.000 0.000 0.858 -0.000 0.000
```

### (iii) Regression coefficients of paths between factors

```
sec_q_intxn_pls$path_coef
```

```
##          REP POL REP*POL          SEC      TRUST
## REP          0  0          0 0.3451419 0.0000000
## POL          0  0          0 0.3764288 0.0000000
## REP*POL      0  0          0 -0.1234258 0.0000000
## SEC          0  0          0 0.0000000 0.6056369
## TRUST        0  0          0 0.0000000 0.0000000
```

### (iv) Bootstrapped path coefficients: t-values, 95% CI

```
boot_pls <- bootstrap_model(sec_q_intxn_pls, nboot = 1000)
```

```
## Bootstrapping model using seminr...
```

```
## SEMinR Model successfully bootstrapped
```

```
summary(boot_pls)$bootstrapped_total_paths
```

```
##          Original Est. Bootstrap Mean Bootstrap SD T Stat. 2.5% CI
## REP -> SEC          0.345          0.347          0.047  7.405  0.258
## REP -> TRUST        0.209          0.212          0.034  6.165  0.151
## POL -> SEC          0.376          0.379          0.049  7.750  0.280
## POL -> TRUST        0.228          0.232          0.033  6.880  0.168
## REP*POL -> SEC      -0.123         -0.035          0.136 -0.907 -0.216
## REP*POL -> TRUST    -0.075         -0.022          0.083 -0.896 -0.135
## SEC -> TRUST        0.606          0.611          0.034 17.846  0.542
##          97.5% CI
## REP -> SEC          0.433
## REP -> TRUST        0.279
## POL -> SEC          0.475
## POL -> TRUST        0.299
## REP*POL -> SEC      0.196
## REP*POL -> TRUST    0.122
## SEC -> TRUST        0.671
```

## Question 2) Common-Factor Models using CB-SEM

a. Create a common factor model using SEMinR, with the following characteristics:

(i) Either respecify all the constructs as being reflective(), or use the as.reflective() function to convert your earlier measurement model to being entirely reflective.

```
sec_q_intxn_cf_mm <- constructs(
  reflective("TRUST", multi_items("TRST", 1:4)),
  reflective("SEC", multi_items("PSEC", 1:4)),
  reflective("REP", multi_items("PREP", 1:4)),
  reflective("INV", multi_items("PINV", 1:3)),
  reflective("POL", multi_items("PPSS", 1:3)),
  reflective("FAML", single_item("FAML1")),
  interaction_term(iv="REP", moderator="POL", method=orthogonal)
)
```

(ii) Use the same structural model as before.

```
sec_q_intxn_sm <- relationships(
  paths(from = c("REP", "POL", "REP*POL"), to = "SEC"),
  paths(from = "SEC", to = "TRUST")
)
sec_q_intxn_cf_pls <- estimate_cbsem( data = sec_q,
  measurement_model = sec_q_intxn_cf_mm,
  structural_model = sec_q_intxn_sm)
```

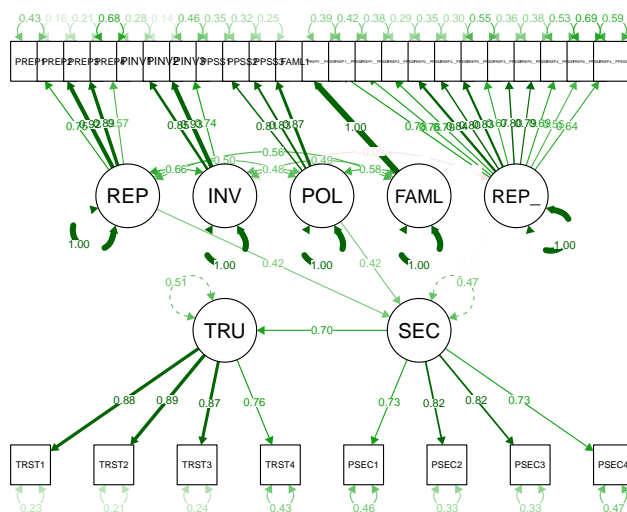
## Generating the seminr model for CBSEM

b. Show us the following results in table or figure formats

(i) Plot a figure of the estimated model (it will look different from your PLS model!)

```
plot(sec_q_intxn_cf_pls)
```

## Plotting of lavaan models using semPlot.



```
## NULL
```

## (ii) Loadings of composites

```
sec_q_intxn_cf_pls_report <- summary(sec_q_intxn_cf_pls)
sec_q_intxn_cf_pls_report$loadings
```

```
## $coefficients
##          TRUST          SEC          REP          INV          POL FAML
## TRST1 0.8797929          NA          NA          NA          NA  NA
## TRST2 0.8887303          NA          NA          NA          NA  NA
## TRST3 0.8692887          NA          NA          NA          NA  NA
## TRST4 0.7574583          NA          NA          NA          NA  NA
## PSEC1          NA 0.7328446          NA          NA          NA  NA
## PSEC2          NA 0.8167529          NA          NA          NA  NA
## PSEC3          NA 0.8159224          NA          NA          NA  NA
## PSEC4          NA 0.7250304          NA          NA          NA  NA
## PREP1          NA          NA 0.7537709          NA          NA  NA
## PREP2          NA          NA 0.9179888          NA          NA  NA
## PREP3          NA          NA 0.8872095          NA          NA  NA
## PREP4          NA          NA 0.5670567          NA          NA  NA
## PINV1          NA          NA          NA 0.8513823          NA  NA
## PINV2          NA          NA          NA 0.9267945          NA  NA
## PINV3          NA          NA          NA 0.7379759          NA  NA
## PPSS1          NA          NA          NA          NA 0.8055024  NA
## PPSS2          NA          NA          NA          NA 0.8266024  NA
## PPSS3          NA          NA          NA          NA 0.8661612  NA
## FAML1          NA          NA          NA          NA          NA  1
##
## $significance
##                               Std Estimate          SE          t-Value          2.5% CI
## TRUST -> TRST1              0.8797929 0.02274972 0.000000e+00 0.8352042
## TRUST -> TRST2              0.8887303 0.03330407 0.000000e+00 0.8234555
## TRUST -> TRST3              0.8692887 0.03747729 0.000000e+00 0.7958346
## TRUST -> TRST4              0.7574583 0.04850725 0.000000e+00 0.6623859
## SEC -> PSEC1                0.7328446 0.03631803 0.000000e+00 0.6616626
## SEC -> PSEC2                0.8167529 0.04454918 0.000000e+00 0.7294381
## SEC -> PSEC3                0.8159224 0.03704994 0.000000e+00 0.7433059
## SEC -> PSEC4                0.7250304 0.03811877 0.000000e+00 0.6503190
## REP -> PREP1                0.7537709 0.04439099 0.000000e+00 0.6667662
## REP -> PREP2                0.9179888 0.02668946 0.000000e+00 0.8656784
## REP -> PREP3                0.8872095 0.03995511 0.000000e+00 0.8088989
## REP -> PREP4                0.5670567 0.04555708 0.000000e+00 0.4777665
## INV -> PINV1                0.8513823 0.04466410 0.000000e+00 0.7638423
## INV -> PINV2                0.9267945 0.04495345 0.000000e+00 0.8386873
## INV -> PINV3                0.7379759 0.04512369 0.000000e+00 0.6495351
## POL -> PPSS1                0.8055024 0.04361793 0.000000e+00 0.7200128
## POL -> PPSS2                0.8266024 0.02817243 0.000000e+00 0.7713854
## POL -> PPSS3                0.8661612 0.03318219 0.000000e+00 0.8011253
## FAML -> FAML1              1.0000000 0.00000000          NA 1.0000000
## REP_x_POL -> PREP1_x_PPSS1  0.7782914 0.05797616 0.000000e+00 0.6646602
## REP_x_POL -> PREP1_x_PPSS2  0.7597967 0.05931773 0.000000e+00 0.6435361
## REP_x_POL -> PREP1_x_PPSS3  0.7879180 0.05013944 0.000000e+00 0.6896465
## REP_x_POL -> PREP2_x_PPSS1  0.8447593 0.03648278 0.000000e+00 0.7732543
```

```

## REP_x_POL -> PREP2_x_PPSS2    0.8033636 0.03638478 0.000000e+00 0.7320508
## REP_x_POL -> PREP2_x_PPSS3    0.8340911 0.03539642 0.000000e+00 0.7647154
## REP_x_POL -> PREP3_x_PPSS1    0.6735260 0.12960483 2.027851e-07 0.4195052
## REP_x_POL -> PREP3_x_PPSS2    0.8011717 0.03779061 0.000000e+00 0.7271035
## REP_x_POL -> PREP3_x_PPSS3    0.7901717 0.06415103 0.000000e+00 0.6644380
## REP_x_POL -> PREP4_x_PPSS1    0.6856568 0.06908216 0.000000e+00 0.5502582
## REP_x_POL -> PREP4_x_PPSS2    0.5533191 0.06211637 0.000000e+00 0.4315732
## REP_x_POL -> PREP4_x_PPSS3    0.6406829 0.05795247 0.000000e+00 0.5270982
##                                97.5% CI
## TRUST -> TRST1                0.9243815
## TRUST -> TRST2                0.9540051
## TRUST -> TRST3                0.9427429
## TRUST -> TRST4                0.8525308
## SEC -> PSEC1                  0.8040266
## SEC -> PSEC2                  0.9040677
## SEC -> PSEC3                  0.8885390
## SEC -> PSEC4                  0.7997419
## REP -> PREP1                  0.8407756
## REP -> PREP2                  0.9702991
## REP -> PREP3                  0.9655201
## REP -> PREP4                  0.6563470
## INV -> PINV1                  0.9389223
## INV -> PINV2                  1.0149016
## INV -> PINV3                  0.8264167
## POL -> PPSS1                  0.8909920
## POL -> PPSS2                  0.8818193
## POL -> PPSS3                  0.9311971
## FAML -> FAML1                 1.0000000
## REP_x_POL -> PREP1_x_PPSS1 0.8919226
## REP_x_POL -> PREP1_x_PPSS2 0.8760573
## REP_x_POL -> PREP1_x_PPSS3 0.8861895
## REP_x_POL -> PREP2_x_PPSS1 0.9162642
## REP_x_POL -> PREP2_x_PPSS2 0.8746765
## REP_x_POL -> PREP2_x_PPSS3 0.9034668
## REP_x_POL -> PREP3_x_PPSS1 0.9275468
## REP_x_POL -> PREP3_x_PPSS2 0.8752400
## REP_x_POL -> PREP3_x_PPSS3 0.9159054
## REP_x_POL -> PREP4_x_PPSS1 0.8210553
## REP_x_POL -> PREP4_x_PPSS2 0.6750649
## REP_x_POL -> PREP4_x_PPSS3 0.7542677

```

(iii) Regression coefficients of paths between factors, and their p-values

```
sec_q_intxn_cf_pls_report$paths[c("coefficients", "pvalues")]
```

```

## $coefficients
##              SEC      TRUST
## R^2          0.52676788 0.4934642
## REP          0.42318633      NA
## POL          0.41557154      NA
## REP_x_POL -0.01219801      NA
## SEC              NA 0.7024701
##
## $pvalues

```

##		SEC	TRUST
##	REP	2.811085e-13	NA
##	POL	4.527045e-12	NA
##	REP_x_POL	8.072900e-01	NA
##	SEC	NA	0