

BACS HW15

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Load Dataset

```
security_sem <- read.csv('D:/桌面/NTHU/Courses/BACS/HW15/security_data_sem.csv')
```

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:

```
security_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:

REP + INV + POL + FAML + (REP×POL) → SEC → TRUST

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

```
pls <- estimate_pls(
  data = security_sem,
  measurement_model = security_mm,
  structural_model = security_sm
)
```

```
## Generating the seminr model
```

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

```
summary(pls)
```

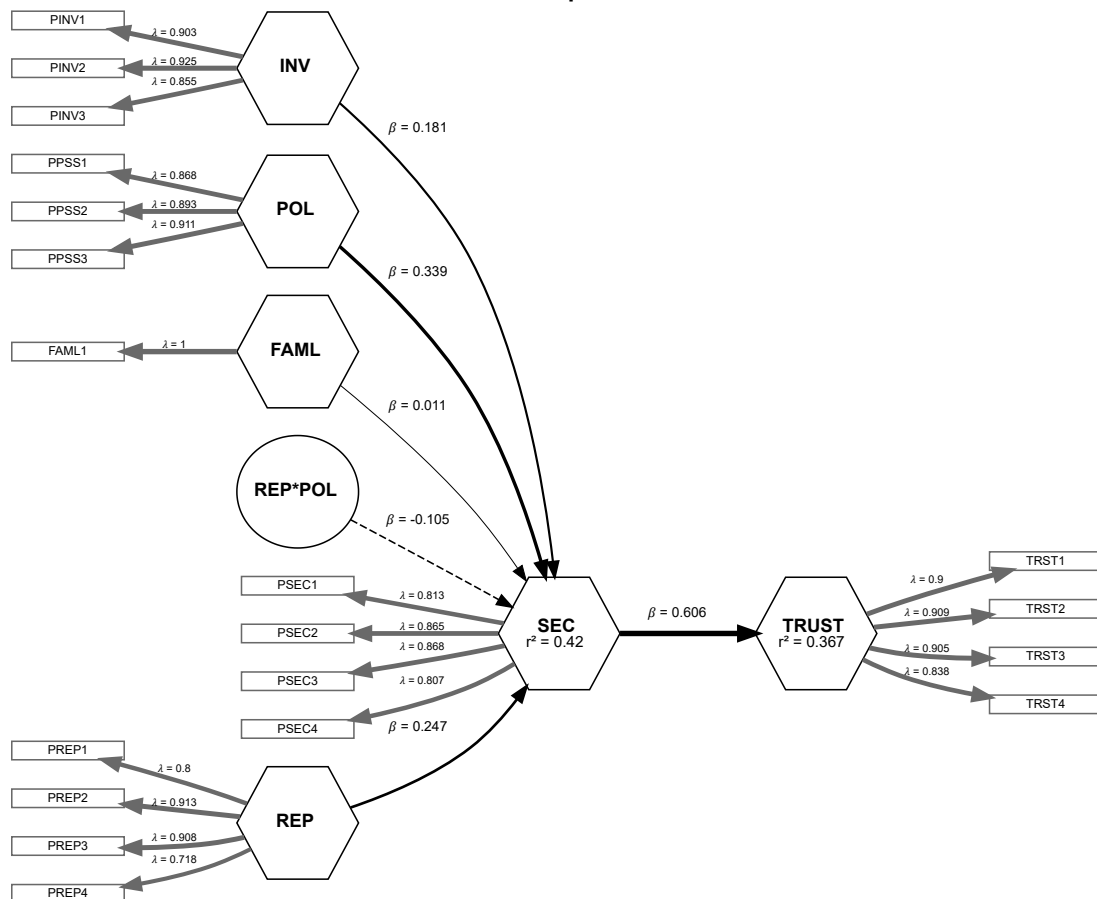
```
##
## Results from package seminr (2.3.2)
##
## Path Coefficients:
##          SEC TRUST
## R^2      0.420 0.367
## AdjR^2   0.412 0.365
## REP      0.247   .
## INV      0.181   .
## POL      0.339   .
## FAML      0.011   .
## REP*POL  -0.105   .
## SEC          . 0.606
##
## Reliability:
##      alpha rhoC  AVE rhoA
## REP   0.857 0.904 0.704 0.882
## INV   0.875 0.923 0.801 0.879
## POL   0.870 0.920 0.794 0.872
## FAML   1.000 1.000 1.000 1.000
## REP*POL 0.938 0.853 0.352 1.000
## SEC    0.859 0.905 0.704 0.862
## TRUST  0.911 0.937 0.789 0.911
##
## Alpha, rhoC, and rhoA should exceed 0.7 while AVE should exceed 0.5
```

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

i. Plot a figure of the estimated model

```
plot(pls, title="PLS plot")
```

PLS plot



ii. Weights and loadings of composites

```
pls_summary <- summary(pls)
# Weight of composites
pls_summary$weights
```

##	REP	INV	POL	FAML	REP*POL	SEC	TRUST
## TRST1	0.000	0.000	0.000	0.000	0.000	0.000	0.282
## TRST2	0.000	0.000	0.000	0.000	0.000	0.000	0.280
## TRST3	0.000	0.000	0.000	0.000	0.000	0.000	0.286
## TRST4	0.000	0.000	0.000	0.000	0.000	0.000	0.278
## PSEC1	0.000	0.000	0.000	0.000	0.000	0.277	0.000
## PSEC2	0.000	0.000	0.000	0.000	0.000	0.315	0.000
## PSEC3	0.000	0.000	0.000	0.000	0.000	0.307	0.000
## PSEC4	0.000	0.000	0.000	0.000	0.000	0.292	0.000
## PREP1	0.215	0.000	0.000	0.000	0.000	0.000	0.000
## PREP2	0.334	0.000	0.000	0.000	0.000	0.000	0.000
## PREP3	0.349	0.000	0.000	0.000	0.000	0.000	0.000
## PREP4	0.287	0.000	0.000	0.000	0.000	0.000	0.000
## PINV1	0.000	0.363	0.000	0.000	0.000	0.000	0.000
## PINV2	0.000	0.395	0.000	0.000	0.000	0.000	0.000
## PINV3	0.000	0.358	0.000	0.000	0.000	0.000	0.000
## PPSS1	0.000	0.000	0.360	0.000	0.000	0.000	0.000
## PPSS2	0.000	0.000	0.395	0.000	0.000	0.000	0.000
## PPSS3	0.000	0.000	0.367	0.000	0.000	0.000	0.000
## FAML1	0.000	0.000	0.000	1.000	0.000	0.000	0.000
## PREP1*PPSS1	0.000	0.000	0.000	0.000	0.239	0.000	0.000
## PREP1*PPSS2	0.000	0.000	0.000	0.000	0.031	0.000	0.000
## PREP1*PPSS3	0.000	0.000	0.000	0.000	0.021	0.000	0.000
## PREP2*PPSS1	0.000	0.000	0.000	0.000	0.046	0.000	0.000
## PREP2*PPSS2	0.000	0.000	0.000	0.000	-0.104	0.000	0.000
## PREP2*PPSS3	0.000	0.000	0.000	0.000	-0.228	0.000	0.000
## PREP3*PPSS1	0.000	0.000	0.000	0.000	-0.341	0.000	0.000
## PREP3*PPSS2	0.000	0.000	0.000	0.000	0.095	0.000	0.000
## PREP3*PPSS3	0.000	0.000	0.000	0.000	0.108	0.000	0.000
## PREP4*PPSS1	0.000	0.000	0.000	0.000	0.443	0.000	0.000
## PREP4*PPSS2	0.000	0.000	0.000	0.000	0.382	0.000	0.000
## PREP4*PPSS3	0.000	0.000	0.000	0.000	0.271	0.000	0.000

```
# Loading of composites
pls_summary$loadings
```

```
##          REP    INV    POL    FAML REP*POL    SEC    TRUST
## TRST1      0.000  0.000  0.000  0.000  -0.000  0.000  0.900
## TRST2      0.000  0.000  0.000  0.000  -0.000  0.000  0.909
## TRST3      0.000  0.000  0.000  0.000  -0.000  0.000  0.905
## TRST4      0.000  0.000  0.000  0.000  -0.000  0.000  0.838
## PSEC1      0.000  0.000  0.000  0.000  -0.000  0.813  0.000
## PSEC2      0.000  0.000  0.000  0.000  -0.000  0.865  0.000
## PSEC3      0.000  0.000  0.000  0.000  -0.000  0.868  0.000
## PSEC4      0.000  0.000  0.000  0.000  -0.000  0.807  0.000
## PREP1      0.800  0.000  0.000  0.000   0.000  0.000  0.000
## PREP2      0.913  0.000  0.000  0.000   0.000  0.000  0.000
## PREP3      0.908  0.000  0.000  0.000   0.000  0.000  0.000
## PREP4      0.718  0.000  0.000  0.000   0.000  0.000  0.000
## PINV1      0.000  0.903  0.000  0.000  -0.000  0.000  0.000
## PINV2      0.000  0.925  0.000  0.000  -0.000  0.000  0.000
## PINV3      0.000  0.855  0.000  0.000  -0.000  0.000  0.000
## PPSS1      0.000  0.000  0.868  0.000   0.000  0.000  0.000
## PPSS2      0.000  0.000  0.893  0.000   0.000  0.000  0.000
## PPSS3      0.000  0.000  0.911  0.000   0.000  0.000  0.000
## FAML1      0.000  0.000  0.000  1.000  -0.000  0.000  0.000
## PREP1*PPSS1 -0.000 -0.000 -0.000 -0.000   0.581 -0.000 -0.000
## PREP1*PPSS2 -0.000 -0.000  0.000 -0.000   0.510 -0.000 -0.000
## PREP1*PPSS3 -0.000 -0.000 -0.000 -0.000   0.506 -0.000 -0.000
## PREP2*PPSS1 -0.000 -0.000 -0.000 -0.000   0.509 -0.000 -0.000
## PREP2*PPSS2 -0.000 -0.000  0.000 -0.000   0.421  0.000  0.000
## PREP2*PPSS3 -0.000 -0.000 -0.000  0.000   0.336  0.000  0.000
## PREP3*PPSS1 -0.000 -0.000 -0.000  0.000   0.236  0.000  0.000
## PREP3*PPSS2 -0.000 -0.000  0.000 -0.000   0.555 -0.000 -0.000
## PREP3*PPSS3 -0.000 -0.000 -0.000  0.000   0.466 -0.000 -0.000
## PREP4*PPSS1  0.000 -0.000  0.000  0.000   0.900 -0.000 -0.000
## PREP4*PPSS2 -0.000 -0.000 -0.000 -0.000   0.836 -0.000  0.000
## PREP4*PPSS3  0.000 -0.000  0.000  0.000   0.859 -0.000  0.000
```

iii. Regression coefficients of paths between factors

```
pls_summary$paths
```

```
##          SEC TRUST
## R^2      0.420 0.367
## AdjR^2   0.412 0.365
## REP      0.247 .
## INV      0.181 .
## POL      0.339 .
## FAML      0.011 .
## REP*POL  -0.105 .
## SEC      . 0.606
```

iv. Bootstrapped path coefficients: t-values, 95% CI

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

```
## Bootstrapping model using semnr...
```

```
## SEminR Model successfully bootstrapped
```

```
summary(boot_pls)
```

```

##
## Results from Bootstrap resamples: 1000
##
## Bootstrapped Structural Paths:
##
##           Original Est. Bootstrap Mean Bootstrap SD T Stat. 2.5% CI
## REP  -> SEC           0.247           0.242           0.061  4.067  0.120
## INV  -> SEC           0.181           0.185           0.059  3.063  0.073
## POL  -> SEC           0.339           0.341           0.057  5.978  0.229
## FAML -> SEC           0.011           0.015           0.062  0.169 -0.111
## REP*POL -> SEC       -0.105          -0.018           0.124 -0.841 -0.194
## SEC  -> TRUST         0.606           0.609           0.034 17.592  0.542
##
##           97.5% CI
## REP  -> SEC           0.359
## INV  -> SEC           0.300
## POL  -> SEC           0.450
## FAML -> SEC           0.134
## REP*POL -> SEC       0.190
## SEC  -> TRUST         0.671
##
## Bootstrapped Weights:
##
##           Original Est. Bootstrap Mean Bootstrap SD T Stat.
## TRST1 -> TRUST         0.282           0.281           0.015 19.310
## TRST2 -> TRUST         0.280           0.280           0.016 18.057
## TRST3 -> TRUST         0.286           0.285           0.017 17.235
## TRST4 -> TRUST         0.278           0.278           0.020 13.930
## PSEC1 -> SEC           0.277           0.278           0.016 17.551
## PSEC2 -> SEC           0.315           0.314           0.017 18.791
## PSEC3 -> SEC           0.307           0.307           0.016 19.303
## PSEC4 -> SEC           0.292           0.291           0.018 16.671
## PREP1 -> REP           0.215           0.213           0.026  8.414
## PREP2 -> REP           0.334           0.333           0.018 18.755
## PREP3 -> REP           0.349           0.350           0.022 15.654
## PREP4 -> REP           0.287           0.286           0.026 11.210
## PINV1 -> INV           0.363           0.363           0.026 14.149
## PINV2 -> INV           0.395           0.394           0.026 15.226
## PINV3 -> INV           0.358           0.359           0.028 12.986
## PPSS1 -> POL           0.360           0.361           0.022 16.151
## PPSS2 -> POL           0.395           0.394           0.023 17.509
## PPSS3 -> POL           0.367           0.367           0.018 20.017
## FAML1 -> FAML          1.000           1.000           0.000  .
## PREP1*PPSS1 -> REP*POL 0.239           0.090           0.156  1.535
## PREP1*PPSS2 -> REP*POL 0.031           0.063           0.096  0.325
## PREP1*PPSS3 -> REP*POL 0.021           0.068           0.109  0.195
## PREP2*PPSS1 -> REP*POL 0.046           0.073           0.109  0.422
## PREP2*PPSS2 -> REP*POL -0.104          0.048           0.159 -0.655
## PREP2*PPSS3 -> REP*POL -0.228          0.045           0.239 -0.955
## PREP3*PPSS1 -> REP*POL -0.341          0.014           0.308 -1.107
## PREP3*PPSS2 -> REP*POL 0.095           0.088           0.136  0.698
## PREP3*PPSS3 -> REP*POL 0.108           0.094           0.135  0.805
## PREP4*PPSS1 -> REP*POL 0.443           0.124           0.283  1.563
## PREP4*PPSS2 -> REP*POL 0.382           0.099           0.267  1.431
## PREP4*PPSS3 -> REP*POL 0.271           0.104           0.187  1.451
##
##           2.5% CI 97.5% CI
## TRST1 -> TRUST         0.253  0.310
## TRST2 -> TRUST         0.249  0.310

```

```

## TRST3 -> TRUST          0.253    0.320
## TRST4 -> TRUST          0.239    0.317
## PSEC1 -> SEC            0.250    0.310
## PSEC2 -> SEC            0.283    0.348
## PSEC3 -> SEC            0.277    0.339
## PSEC4 -> SEC            0.257    0.325
## PREP1 -> REP            0.160    0.259
## PREP2 -> REP            0.301    0.371
## PREP3 -> REP            0.310    0.397
## PREP4 -> REP            0.241    0.339
## PINV1 -> INV            0.312    0.412
## PINV2 -> INV            0.342    0.446
## PINV3 -> INV            0.306    0.416
## PPSS1 -> POL            0.314    0.405
## PPSS2 -> POL            0.354    0.440
## PPSS3 -> POL            0.331    0.401
## FAML1 -> FAML           1.000    1.000
## PREP1*PPSS1 -> REP*POL -0.264    0.394
## PREP1*PPSS2 -> REP*POL -0.153    0.232
## PREP1*PPSS3 -> REP*POL -0.171    0.264
## PREP2*PPSS1 -> REP*POL -0.170    0.265
## PREP2*PPSS2 -> REP*POL -0.301    0.344
## PREP2*PPSS3 -> REP*POL -0.399    0.475
## PREP3*PPSS1 -> REP*POL -0.586    0.663
## PREP3*PPSS2 -> REP*POL -0.240    0.329
## PREP3*PPSS3 -> REP*POL -0.234    0.337
## PREP4*PPSS1 -> REP*POL -0.439    0.547
## PREP4*PPSS2 -> REP*POL -0.430    0.565
## PREP4*PPSS3 -> REP*POL -0.281    0.421
##

```

Bootstrapped Loadings:

```

##              Original Est. Bootstrap Mean Bootstrap SD T Stat.
## TRST1 -> TRUST          0.900          0.900          0.016 57.892
## TRST2 -> TRUST          0.909          0.910          0.020 45.405
## TRST3 -> TRUST          0.905          0.906          0.021 43.612
## TRST4 -> TRUST          0.838          0.839          0.031 27.134
## PSEC1 -> SEC            0.813          0.813          0.025 32.280
## PSEC2 -> SEC            0.865          0.866          0.025 35.156
## PSEC3 -> SEC            0.868          0.869          0.021 41.238
## PSEC4 -> SEC            0.807          0.807          0.025 32.926
## PREP1 -> REP            0.800          0.798          0.039 20.413
## PREP2 -> REP            0.913          0.914          0.016 57.392
## PREP3 -> REP            0.908          0.910          0.020 45.938
## PREP4 -> REP            0.718          0.719          0.033 21.967
## PINV1 -> INV            0.903          0.904          0.024 37.488
## PINV2 -> INV            0.925          0.925          0.022 41.837
## PINV3 -> INV            0.855          0.855          0.026 33.242
## PPSS1 -> POL            0.868          0.868          0.023 36.934
## PPSS2 -> POL            0.893          0.894          0.014 65.913
## PPSS3 -> POL            0.911          0.911          0.017 55.211
## FAML1 -> FAML           1.000          1.000          0.000 .
## PREP1*PPSS1 -> REP*POL  0.581          0.582          0.264  2.202
## PREP1*PPSS2 -> REP*POL  0.510          0.566          0.250  2.039
## PREP1*PPSS3 -> REP*POL  0.506          0.583          0.259  1.951
## PREP2*PPSS1 -> REP*POL  0.509          0.616          0.278  1.835
## PREP2*PPSS2 -> REP*POL  0.421          0.579          0.296  1.420

```


## PREP2*PPSS3	-> REP*POL	0.336	0.591	0.336	0.999
## PREP3*PPSS1	-> REP*POL	0.236	0.504	0.345	0.683
## PREP3*PPSS2	-> REP*POL	0.555	0.610	0.283	1.959
## PREP3*PPSS3	-> REP*POL	0.466	0.598	0.293	1.592
## PREP4*PPSS1	-> REP*POL	0.900	0.588	0.362	2.488
## PREP4*PPSS2	-> REP*POL	0.836	0.505	0.359	2.327
## PREP4*PPSS3	-> REP*POL	0.859	0.565	0.333	2.582

##		2.5% CI	97.5% CI
----	--	---------	----------

## TRST1	-> TRUST	0.867	0.928
## TRST2	-> TRUST	0.861	0.940
## TRST3	-> TRUST	0.857	0.938
## TRST4	-> TRUST	0.773	0.892
## PSEC1	-> SEC	0.761	0.859
## PSEC2	-> SEC	0.814	0.907
## PSEC3	-> SEC	0.823	0.906
## PSEC4	-> SEC	0.757	0.852
## PREP1	-> REP	0.717	0.863
## PREP2	-> REP	0.878	0.941
## PREP3	-> REP	0.867	0.939
## PREP4	-> REP	0.653	0.776
## PINV1	-> INV	0.847	0.942
## PINV2	-> INV	0.872	0.959
## PINV3	-> INV	0.801	0.902
## PPSS1	-> POL	0.815	0.906
## PPSS2	-> POL	0.864	0.917
## PPSS3	-> POL	0.873	0.939
## FAML1	-> FAML	1.000	1.000
## PREP1*PPSS1	-> REP*POL	-0.027	0.914
## PREP1*PPSS2	-> REP*POL	-0.067	0.883
## PREP1*PPSS3	-> REP*POL	-0.076	0.900
## PREP2*PPSS1	-> REP*POL	-0.059	0.951
## PREP2*PPSS2	-> REP*POL	-0.182	0.932
## PREP2*PPSS3	-> REP*POL	-0.266	0.976
## PREP3*PPSS1	-> REP*POL	-0.304	0.941
## PREP3*PPSS2	-> REP*POL	-0.092	0.941
## PREP3*PPSS3	-> REP*POL	-0.165	0.948
## PREP4*PPSS1	-> REP*POL	-0.275	0.980
## PREP4*PPSS2	-> REP*POL	-0.357	0.915
## PREP4*PPSS3	-> REP*POL	-0.241	0.951

##

Bootstrapped HTMT:

##		Original Est.	Bootstrap Mean	Bootstrap SD	2.5% CI	97.5% CI
## REP	-> INV	0.705	0.706	0.050	0.597	0.795
## REP	-> POL	0.543	0.545	0.057	0.428	0.653
## REP	-> FAML	0.599	0.599	0.056	0.488	0.705
## REP	-> REP*POL	0.000	0.000	0.000	0.000	0.000
## REP	-> SEC	0.595	0.595	0.045	0.499	0.680
## REP	-> TRUST	0.682	0.684	0.044	0.588	0.764
## INV	-> POL	0.498	0.501	0.058	0.391	0.618
## INV	-> FAML	0.494	0.495	0.056	0.388	0.606
## INV	-> REP*POL	0.085	0.102	0.032	0.056	0.175
## INV	-> SEC	0.568	0.570	0.049	0.471	0.665
## INV	-> TRUST	0.563	0.564	0.051	0.456	0.662
## POL	-> FAML	0.596	0.595	0.050	0.491	0.695
## POL	-> REP*POL	0.000	0.000	0.000	0.000	0.000
## POL	-> SEC	0.622	0.624	0.053	0.515	0.720

```
## POL -> TRUST      0.458      0.460      0.061  0.332  0.573
## FAML -> REP*POL    0.046      0.065      0.023  0.030  0.122
## FAML -> SEC       0.455      0.458      0.054  0.353  0.562
## FAML -> TRUST     0.471      0.474      0.054  0.366  0.581
## REP*POL -> SEC     0.059      0.081      0.020  0.049  0.128
## REP*POL -> TRUST   0.044      0.071      0.018  0.043  0.116
## SEC -> TRUST      0.685      0.686      0.036  0.612  0.753
##
## Bootstrapped Total Paths:
##               Original Est. Bootstrap Mean Bootstrap SD 2.5% CI 97.5% CI
## REP -> SEC      0.247      0.242      0.061  0.120  0.359
## REP -> TRUST    0.150      0.148      0.039  0.072  0.227
## INV -> SEC      0.181      0.185      0.059  0.073  0.300
## INV -> TRUST    0.109      0.113      0.036  0.047  0.183
## POL -> SEC      0.339      0.341      0.057  0.229  0.450
## POL -> TRUST    0.205      0.208      0.037  0.133  0.281
## FAML -> SEC      0.011      0.015      0.062 -0.111  0.134
## FAML -> TRUST    0.006      0.009      0.038 -0.068  0.085
## REP*POL -> SEC   -0.105     -0.018      0.124 -0.194  0.190
## REP*POL -> TRUST -0.063     -0.011      0.076 -0.121  0.115
## SEC -> TRUST     0.606      0.609      0.034  0.542  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.

```
ref_mm <- as.reflective(security_mm)
```

ii. Use the same structural model as before (you can just reuse it again!)

```
security_sm
```

```
##      source      target
## [1,] "REP"      "SEC"
## [2,] "INV"      "SEC"
## [3,] "POL"      "SEC"
## [4,] "FAML"     "SEC"
## [5,] "REP*POL"  "SEC"
## [6,] "SEC"      "TRUST"
## attr(,"class")
## [1] "matrix"          "array"          "structural_model" "seminr_model"
```

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

```
sec_cbsem <- estimate_cbsem(  
  data = security_sem,  
  measurement_model = ref_mm,  
  structural_model = security_sm  
)
```

```
## Generating the semnr model for CBSEM
```

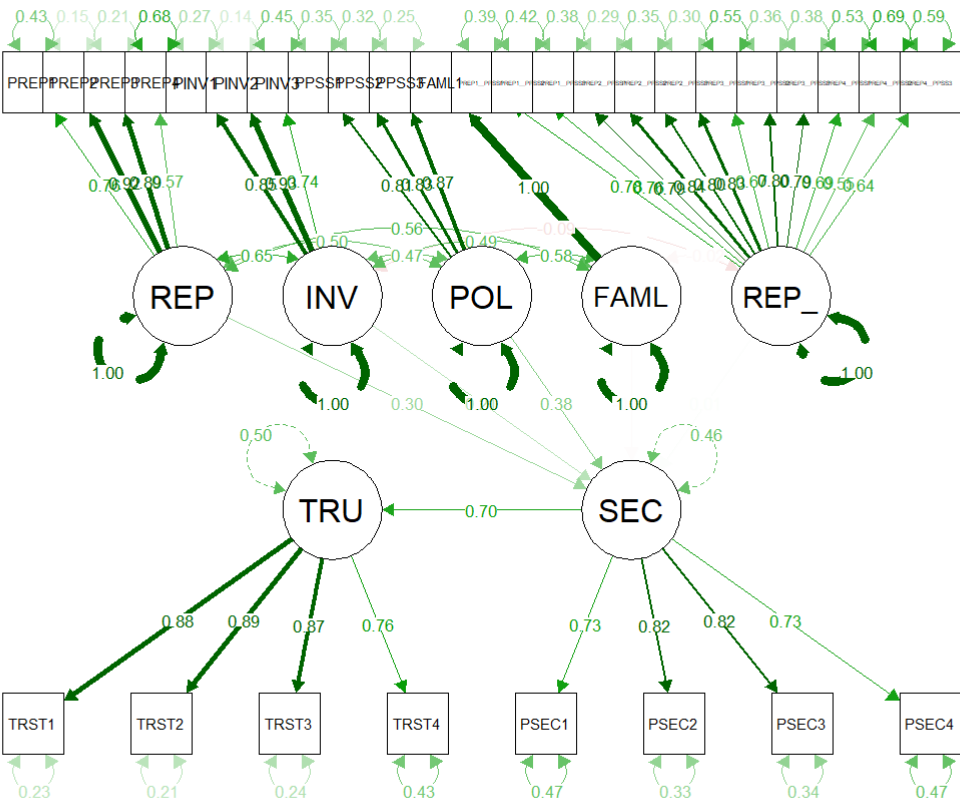
```
summary(sec_cbsem)
```

```
##
## Results from package seminr (2.3.2)
## Estimation used package seminr (2.3.2)
##
## Fit metrics:
##      npar      fmin      logl      aic      bic      ntotal      bic2
##    77.000    3.529 -17296.241 34746.482 35054.781 405.000 34810.451
##      rmr      srmr      crmr      gfi      agfi      pgfi      mfi
##    0.116    0.063    0.065    0.742    0.694    0.627    0.049
##      ecvi
##    7.439
##
##
##      metric scaled robust
## cfi      0.764    0.772 0.799
## tli      0.738    0.747 0.777
## nnfi     0.738    0.747 0.777
## rni      0.764    0.772 0.799
## rmsea     0.120    0.072 0.107
## rmsea.ci.lower 0.116    0.069 0.100
## rmsea.ci.upper 0.124    0.075 0.114
## rmsea.pvalue 0.000    0.000 0.000
## rmsea.notclose.pvalue 1.000    0.000 1.000
## chisq      2858.871 1303.538 .
## df        419.000 419.000 .
## pvalue     0.000    0.000 .
## baseline.chisq 10812.133 4340.588 .
## baseline.df   465.000 465.000 .
## baseline.pvalue 0.000    0.000 .
## rfi        0.707    0.667 .
## nfi        0.736    0.700 .
## pnfi       0.663    0.630 .
## ifi        0.765    0.774 .
##
##
## Reliability:
##      rhoC  AVE
## TRUST 0.91 0.72
## SEC   0.86 0.60
## REP   0.87 0.63
## INV   0.88 0.71
## POL   0.87 0.70
## FAML  1.00 1.00
##
## Path Coefficients:
##      SEC TRUST
## R^2   0.54 0.50
## REP   0.30 .
## INV   0.21 .
## POL   0.38 .
## FAML  -0.01 .
## REP_x_POL 0.01 .
## SEC    . 0.70
```

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

```
plot(sec_cbsem, title="Common-Factor Models using CB-SEM")
```

```
## Plotting of lavaan models using semPlot.
```



```
## NULL
```

ii. Loadings of composites

```
summary_cbsem <- summary(sec_cbsem)
summary_cbsem$loadings
```

\$coefficients

##	TRUST	SEC	REP	INV	POL	FAML
## TRST1	0.8800240	NA	NA	NA	NA	NA
## TRST2	0.8886342	NA	NA	NA	NA	NA
## TRST3	0.8690644	NA	NA	NA	NA	NA
## TRST4	0.7575988	NA	NA	NA	NA	NA
## PSEC1	NA	0.7308766	NA	NA	NA	NA
## PSEC2	NA	0.8173481	NA	NA	NA	NA
## PSEC3	NA	0.8151708	NA	NA	NA	NA
## PSEC4	NA	0.7260444	NA	NA	NA	NA
## PREP1	NA	NA	0.7551328	NA	NA	NA
## PREP2	NA	NA	0.9199208	NA	NA	NA
## PREP3	NA	NA	0.8871362	NA	NA	NA
## PREP4	NA	NA	0.5650059	NA	NA	NA
## PINV1	NA	NA	NA	0.8520004	NA	NA
## PINV2	NA	NA	NA	0.9257476	NA	NA
## PINV3	NA	NA	NA	0.7388750	NA	NA
## PPSS1	NA	NA	NA	NA	0.8051533	NA
## PPSS2	NA	NA	NA	NA	0.8272576	NA
## PPSS3	NA	NA	NA	NA	0.8674335	NA
## FAML1	NA	NA	NA	NA	NA	1

##

\$significance

##	Std Estimate	SE	t-Value	2.5% CI
## TRUST -> TRST1	0.8800240	0.02272091	0.000000e+00	0.8354919
## TRUST -> TRST2	0.8886342	0.03330783	0.000000e+00	0.8233521
## TRUST -> TRST3	0.8690644	0.03749444	0.000000e+00	0.7955767
## TRUST -> TRST4	0.7575988	0.04846748	0.000000e+00	0.6626042
## SEC -> PSEC1	0.7308766	0.03679205	0.000000e+00	0.6587655
## SEC -> PSEC2	0.8173481	0.04480183	0.000000e+00	0.7295381
## SEC -> PSEC3	0.8151708	0.03728082	0.000000e+00	0.7421017
## SEC -> PSEC4	0.7260444	0.03811841	0.000000e+00	0.6513337
## REP -> PREP1	0.7551328	0.04464916	0.000000e+00	0.6676220
## REP -> PREP2	0.9199208	0.02635333	0.000000e+00	0.8682692
## REP -> PREP3	0.8871362	0.04015103	0.000000e+00	0.8084416
## REP -> PREP4	0.5650059	0.04585583	0.000000e+00	0.4751302
## INV -> PINV1	0.8520004	0.04489927	0.000000e+00	0.7639994
## INV -> PINV2	0.9257476	0.04556425	0.000000e+00	0.8364433
## INV -> PINV3	0.7388750	0.04511601	0.000000e+00	0.6504492
## POL -> PPSS1	0.8051533	0.04355300	0.000000e+00	0.7197910
## POL -> PPSS2	0.8272576	0.02807169	0.000000e+00	0.7722381
## POL -> PPSS3	0.8674335	0.03273664	0.000000e+00	0.8032708
## FAML -> FAML1	1.0000000	0.00000000	NA	1.0000000
## REP_x_POL -> PREP1_x_PPSS1	0.7781584	0.05799871	0.000000e+00	0.6644831
## REP_x_POL -> PREP1_x_PPSS2	0.7597768	0.05931838	0.000000e+00	0.6435149
## REP_x_POL -> PREP1_x_PPSS3	0.7879106	0.05013554	0.000000e+00	0.6896467
## REP_x_POL -> PREP2_x_PPSS1	0.8447368	0.03649041	0.000000e+00	0.7732169
## REP_x_POL -> PREP2_x_PPSS2	0.8034561	0.03639411	0.000000e+00	0.7321250
## REP_x_POL -> PREP2_x_PPSS3	0.8342444	0.03536430	0.000000e+00	0.7649317
## REP_x_POL -> PREP3_x_PPSS1	0.6736451	0.12948899	1.967998e-07	0.4198514
## REP_x_POL -> PREP3_x_PPSS2	0.8011944	0.03780427	0.000000e+00	0.7270994
## REP_x_POL -> PREP3_x_PPSS3	0.7902063	0.06416741	0.000000e+00	0.6644405
## REP_x_POL -> PREP4_x_PPSS1	0.6854770	0.06906812	0.000000e+00	0.5501059
## REP_x_POL -> PREP4_x_PPSS2	0.5531922	0.06212434	0.000000e+00	0.4314307
## REP_x_POL -> PREP4_x_PPSS3	0.6405843	0.05794028	0.000000e+00	0.5270235

```
##                                97.5% CI
## TRUST -> TRST1                0.9245562
## TRUST -> TRST2                0.9539164
## TRUST -> TRST3                0.9425522
## TRUST -> TRST4                0.8525933
## SEC -> PSEC1                  0.8029877
## SEC -> PSEC2                  0.9051581
## SEC -> PSEC3                  0.8882399
## SEC -> PSEC4                  0.8007551
## REP -> PREP1                  0.8426435
## REP -> PREP2                  0.9715724
## REP -> PREP3                  0.9658307
## REP -> PREP4                  0.6548817
## INV -> PINV1                  0.9400013
## INV -> PINV2                  1.0150518
## INV -> PINV3                  0.8273007
## POL -> PPSS1                  0.8905156
## POL -> PPSS2                  0.8822771
## POL -> PPSS3                  0.9315961
## FAML -> FAML1                 1.0000000
## REP_x_POL -> PREP1_x_PPSS1 0.8918338
## REP_x_POL -> PREP1_x_PPSS2 0.8760387
## REP_x_POL -> PREP1_x_PPSS3 0.8861744
## REP_x_POL -> PREP2_x_PPSS1 0.9162567
## REP_x_POL -> PREP2_x_PPSS2 0.8747873
## REP_x_POL -> PREP2_x_PPSS3 0.9035572
## REP_x_POL -> PREP3_x_PPSS1 0.9274389
## REP_x_POL -> PREP3_x_PPSS2 0.8752894
## REP_x_POL -> PREP3_x_PPSS3 0.9159721
## REP_x_POL -> PREP4_x_PPSS1 0.8208480
## REP_x_POL -> PREP4_x_PPSS2 0.6749536
## REP_x_POL -> PREP4_x_PPSS3 0.7541452
```

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

```
summary_cbsem$paths
```

```
## $coefficients
##              SEC      TRUST
## R^2          0.540381651 0.4951084
## REP          0.299536782      NA
## INV          0.214253245      NA
## POL          0.376401499      NA
## FAML         -0.008837653      NA
## REP_x_POL    0.008355287      NA
## SEC          NA 0.7036394
##
## $pvalues
##              SEC TRUST
## REP          3.817182e-05  NA
## INV          3.534482e-03  NA
## POL          4.380975e-09  NA
## FAML          8.996836e-01  NA
## REP_x_POL    8.516847e-01  NA
## SEC          NA      0
##
## $significance
##              Std Estimate      SE      t-Value      2.5% CI      97.5% CI
## SEC -> REP          0.299536782 0.07273355 3.817182e-05  0.15698165 0.44209191
## SEC -> INV          0.214253245 0.07345058 3.534482e-03  0.07029275 0.35821374
## SEC -> POL          0.376401499 0.06413246 4.380975e-09  0.25070419 0.50209881
## SEC -> FAML         -0.008837653 0.07010617 8.996836e-01 -0.14624321 0.12856791
## SEC -> REP_x_POL    0.008355287 0.04468802 8.516847e-01 -0.07923162 0.09594219
## TRUST -> SEC          0.703639369 0.03721629 0.000000e+00  0.63069677 0.77658197
```