

# BACS - HW15

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Please download the data file `security_data_sem.csv` from Canvas – there might be differences in file from previous weeks. This file shows you results of a survey about website security.

We will create a model similar to the one we saw in class, with several important differences. We will have several new constructs, and also include a single-item construct.

```
data = read.csv("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:

1. Trust in website (TRUST): items TRST1 - TRST4
2. Perceived security of website (SEC): items PSEC1 - PSEC4
3. Reputation of website (REP): items PREP1 - PREP4
4. Investment in website (INV): items PINV1 - PINV3
5. Perception of privacy policies (POL): items PPSS1 - PPSS3
6. Familiarity with website (FAML): item FAML1 (see the documentation of SEMinR for making single item constructs)
7. Interaction between REP and POL (use orthogonalized product terms)

```
sec_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("REP", "POL", method=orthogonal)  
)
```

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

$REP + INV + POL + FAML + (REP \cdot POL) \rightarrow SEC \rightarrow TRUST$

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

```
sec_pls = estimate_pls(data=data, measurement_model=sec_mm, structural_model=sec_sm)
```

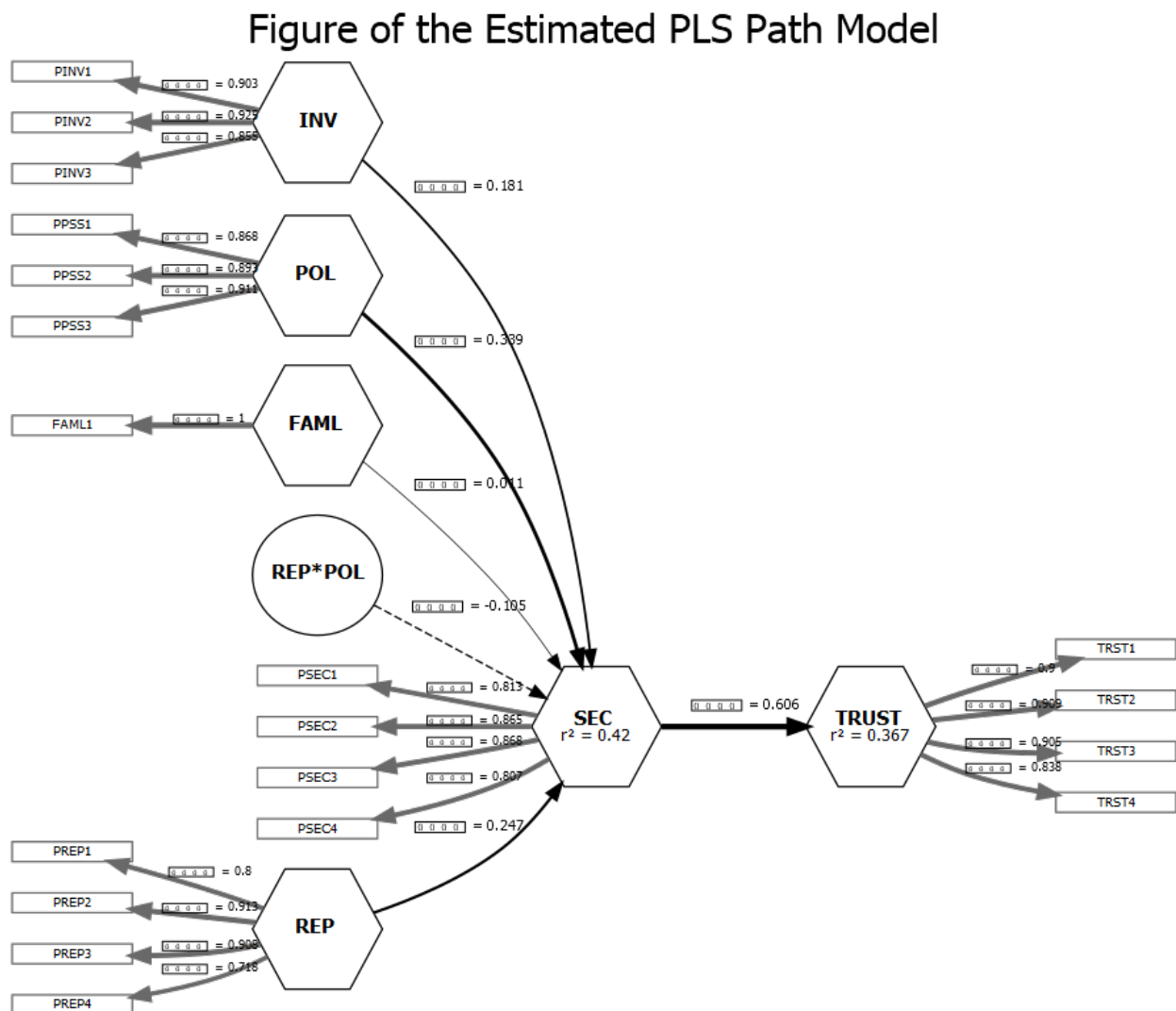
```
## Generating the semnr model
```

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

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

i. Plot a figure of the estimated model

```
plot(sec_pls, title="Figure of the Estimated PLS Path Model")
```



ii. Weights and loadings of composites

```
summary(sec_pls)$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

```
summary(sec_pls)$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

```
summary(sec_pls)$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(sec_pls, nbboot=1000)
```

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

```
## 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          0.057  4.335  0.127
## INV  -> SEC          0.181          0.187          0.057  3.154  0.072
## POL  -> SEC          0.339          0.343          0.054  6.286  0.234
## FAML -> SEC          0.011          0.009          0.059  0.178 -0.105
```

```

## REP*POL -> SEC      -0.105      -0.016      0.124  -0.842  -0.196
## SEC -> TRUST      0.606      0.608      0.035  17.378   0.536
##
##          97.5% CI
## REP -> SEC      0.349
## INV -> SEC      0.292
## POL -> SEC      0.445
## FAML -> SEC      0.134
## REP*POL -> SEC    0.188
## SEC -> TRUST      0.674
##
## Bootstrapped Weights:
##
##          Original Est. Bootstrap Mean Bootstrap SD T Stat.
## TRST1 -> TRUST      0.282      0.282      0.015  18.540
## TRST2 -> TRUST      0.280      0.281      0.015  18.114
## TRST3 -> TRUST      0.286      0.284      0.016  17.860
## TRST4 -> TRUST      0.278      0.278      0.020  13.679
## PSEC1 -> SEC      0.277      0.277      0.015  18.508
## PSEC2 -> SEC      0.315      0.314      0.017  18.671
## PSEC3 -> SEC      0.307      0.308      0.016  19.577
## PSEC4 -> SEC      0.292      0.291      0.018  16.533
## PREP1 -> REP      0.215      0.213      0.027   7.970
## PREP2 -> REP      0.334      0.334      0.019  17.870
## PREP3 -> REP      0.349      0.351      0.023  15.426
## PREP4 -> REP      0.287      0.286      0.025  11.268
## PINV1 -> INV      0.363      0.363      0.025  14.525
## PINV2 -> INV      0.395      0.394      0.026  14.967
## PINV3 -> INV      0.358      0.358      0.028  12.822
## PPSS1 -> POL      0.360      0.360      0.022  16.046
## PPSS2 -> POL      0.395      0.395      0.023  16.970
## PPSS3 -> POL      0.367      0.367      0.018  20.955
## FAML1 -> FAML      1.000      1.000      0.000   .
## PREP1*PPSS1 -> REP*POL    0.239      0.094      0.155   1.539
## PREP1*PPSS2 -> REP*POL    0.031      0.069      0.089   0.350
## PREP1*PPSS3 -> REP*POL    0.021      0.068      0.113   0.187
## PREP2*PPSS1 -> REP*POL    0.046      0.078      0.107   0.431
## PREP2*PPSS2 -> REP*POL   -0.104      0.051      0.156  -0.669
## PREP2*PPSS3 -> REP*POL   -0.228      0.043      0.236  -0.966
## PREP3*PPSS1 -> REP*POL   -0.341      0.011      0.307  -1.109
## PREP3*PPSS2 -> REP*POL    0.095      0.089      0.148   0.641
## PREP3*PPSS3 -> REP*POL    0.108      0.093      0.128   0.850
## PREP4*PPSS1 -> REP*POL    0.443      0.126      0.275   1.613
## PREP4*PPSS2 -> REP*POL    0.382      0.097      0.277   1.381
## PREP4*PPSS3 -> REP*POL    0.271      0.098      0.184   1.478
##
##          2.5% CI  97.5% CI
## TRST1 -> TRUST      0.252   0.311
## TRST2 -> TRUST      0.249   0.310
## TRST3 -> TRUST      0.252   0.317
## TRST4 -> TRUST      0.241   0.321
## PSEC1 -> SEC      0.249   0.308
## PSEC2 -> SEC      0.281   0.348
## PSEC3 -> SEC      0.279   0.341
## PSEC4 -> SEC      0.258   0.328
## PREP1 -> REP      0.150   0.260
## PREP2 -> REP      0.300   0.374

```

```

## PREP3 -> REP          0.306    0.396
## PREP4 -> REP          0.240    0.338
## PINV1 -> INV          0.315    0.413
## PINV2 -> INV          0.346    0.446
## PINV3 -> INV          0.305    0.413
## PPSS1 -> POL          0.313    0.402
## PPSS2 -> POL          0.351    0.446
## PPSS3 -> POL          0.330    0.401
## FAML1 -> FAML         1.000    1.000
## PREP1*PPSS1 -> REP*POL -0.254    0.371
## PREP1*PPSS2 -> REP*POL -0.136    0.232
## PREP1*PPSS3 -> REP*POL -0.183    0.280
## PREP2*PPSS1 -> REP*POL -0.170    0.271
## PREP2*PPSS2 -> REP*POL -0.262    0.335
## PREP2*PPSS3 -> REP*POL -0.385    0.446
## PREP3*PPSS1 -> REP*POL -0.581    0.680
## PREP3*PPSS2 -> REP*POL -0.251    0.349
## PREP3*PPSS3 -> REP*POL -0.214    0.313
## PREP4*PPSS1 -> REP*POL -0.428    0.545
## PREP4*PPSS2 -> REP*POL -0.467    0.569
## PREP4*PPSS3 -> REP*POL -0.292    0.408
##
## Bootstrapped Loadings:
##
## Original Est. Bootstrap Mean Bootstrap SD T Stat.
## TRST1 -> TRUST        0.900        0.900        0.015  58.378
## TRST2 -> TRUST        0.909        0.909        0.021  43.800
## TRST3 -> TRUST        0.905        0.904        0.022  41.060
## TRST4 -> TRUST        0.838        0.840        0.032  26.443
## PSEC1 -> SEC          0.813        0.814        0.024  33.473
## PSEC2 -> SEC          0.865        0.866        0.025  35.091
## PSEC3 -> SEC          0.868        0.870        0.021  40.456
## PSEC4 -> SEC          0.807        0.807        0.025  31.808
## PREP1 -> REP          0.800        0.797        0.041  19.548
## PREP2 -> REP          0.913        0.913        0.016  55.752
## PREP3 -> REP          0.908        0.910        0.019  48.186
## PREP4 -> REP          0.718        0.717        0.033  21.986
## PINV1 -> INV          0.903        0.904        0.024  37.168
## PINV2 -> INV          0.925        0.925        0.022  42.540
## PINV3 -> INV          0.855        0.854        0.025  34.104
## PPSS1 -> POL          0.868        0.868        0.024  35.724
## PPSS2 -> POL          0.893        0.893        0.014  65.091
## PPSS3 -> POL          0.911        0.911        0.016  55.784
## FAML1 -> FAML         1.000        1.000        0.000    .
## PREP1*PPSS1 -> REP*POL 0.581        0.584        0.264   2.200
## PREP1*PPSS2 -> REP*POL 0.510        0.568        0.247   2.065
## PREP1*PPSS3 -> REP*POL 0.506        0.582        0.263   1.925
## PREP2*PPSS1 -> REP*POL 0.509        0.613        0.281   1.815
## PREP2*PPSS2 -> REP*POL 0.421        0.578        0.290   1.450
## PREP2*PPSS3 -> REP*POL 0.336        0.584        0.336   0.999
## PREP3*PPSS1 -> REP*POL 0.236        0.496        0.343   0.687
## PREP3*PPSS2 -> REP*POL 0.555        0.609        0.283   1.961
## PREP3*PPSS3 -> REP*POL 0.466        0.594        0.294   1.586
## PREP4*PPSS1 -> REP*POL 0.900        0.587        0.356   2.527
## PREP4*PPSS2 -> REP*POL 0.836        0.501        0.358   2.335

```

```

## PREP4*PPSS3 -> REP*POL          0.859          0.560          0.331    2.595
##                               2.5% CI 97.5% CI
## TRST1 -> TRUST          0.867    0.926
## TRST2 -> TRUST          0.864    0.943
## TRST3 -> TRUST          0.853    0.938
## TRST4 -> TRUST          0.766    0.894
## PSEC1 -> SEC            0.765    0.861
## PSEC2 -> SEC            0.809    0.907
## PSEC3 -> SEC            0.824    0.908
## PSEC4 -> SEC            0.755    0.851
## PREP1 -> REP            0.706    0.870
## PREP2 -> REP            0.877    0.940
## PREP3 -> REP            0.869    0.939
## PREP4 -> REP            0.648    0.775
## PINV1 -> INV            0.852    0.943
## PINV2 -> INV            0.878    0.958
## PINV3 -> INV            0.803    0.900
## PPSS1 -> POL            0.815    0.906
## PPSS2 -> POL            0.864    0.918
## PPSS3 -> POL            0.874    0.938
## FAML1 -> FAML           1.000    1.000
## PREP1*PPSS1 -> REP*POL -0.055    0.924
## PREP1*PPSS2 -> REP*POL -0.060    0.880
## PREP1*PPSS3 -> REP*POL -0.079    0.899
## PREP2*PPSS1 -> REP*POL -0.161    0.948
## PREP2*PPSS2 -> REP*POL -0.184    0.931
## PREP2*PPSS3 -> REP*POL -0.291    0.973
## PREP3*PPSS1 -> REP*POL -0.372    0.924
## PREP3*PPSS2 -> REP*POL -0.111    0.937
## PREP3*PPSS3 -> REP*POL -0.172    0.941
## PREP4*PPSS1 -> REP*POL -0.295    0.988
## PREP4*PPSS2 -> REP*POL -0.388    0.917
## PREP4*PPSS3 -> REP*POL -0.256    0.942
##
## Bootstrapped HTMT:
##                               Original Est. Bootstrap Mean Bootstrap SD 2.5% CI 97.5% CI
## REP -> INV              0.705          0.703          0.050    0.600    0.794
## REP -> POL              0.543          0.544          0.053    0.434    0.644
## REP -> FAML             0.599          0.602          0.052    0.492    0.700
## REP -> REP*POL          0.000          0.000          0.000    0.000    0.000
## REP -> SEC              0.595          0.593          0.044    0.503    0.672
## REP -> TRUST            0.682          0.682          0.042    0.595    0.764
## INV -> POL              0.498          0.496          0.055    0.389    0.603
## INV -> FAML             0.494          0.493          0.056    0.384    0.607
## INV -> REP*POL          0.085          0.105          0.033    0.056    0.180
## INV -> SEC              0.568          0.567          0.048    0.465    0.655
## INV -> TRUST            0.563          0.561          0.051    0.460    0.654
## POL -> FAML             0.596          0.593          0.052    0.488    0.688
## POL -> REP*POL          0.000          0.000          0.000    0.000    0.000
## POL -> SEC              0.622          0.622          0.050    0.521    0.716
## POL -> TRUST            0.458          0.460          0.057    0.339    0.564
## FAML -> REP*POL          0.046          0.065          0.025    0.032    0.123
## FAML -> SEC              0.455          0.453          0.053    0.345    0.555
## FAML -> TRUST            0.471          0.472          0.052    0.367    0.572

```

## REP*POL -> SEC	0.059	0.082	0.020	0.049	0.125
## REP*POL -> TRUST	0.044	0.072	0.017	0.043	0.112
## SEC -> TRUST	0.685	0.685	0.037	0.609	0.754
##					
## Bootstrapped Total Paths:					
##	Original Est.	Bootstrap Mean	Bootstrap SD	2.5% CI	97.5% CI
## REP -> SEC	0.247	0.242	0.057	0.127	0.349
## REP -> TRUST	0.150	0.148	0.037	0.078	0.220
## INV -> SEC	0.181	0.187	0.057	0.072	0.292
## INV -> TRUST	0.109	0.114	0.036	0.047	0.182
## POL -> SEC	0.339	0.343	0.054	0.234	0.445
## POL -> TRUST	0.205	0.209	0.035	0.138	0.277
## FAML -> SEC	0.011	0.009	0.059	-0.105	0.134
## FAML -> TRUST	0.006	0.005	0.036	-0.063	0.080
## REP*POL -> SEC	-0.105	-0.016	0.124	-0.196	0.188
## REP*POL -> TRUST	-0.063	-0.010	0.076	-0.119	0.117
## SEC -> TRUST	0.606	0.608	0.035	0.536	0.674

---

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

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

- Either respecify all the constructs as being reflective(), or use the as.reflective() function to convert your earlier measurement model to being entirely reflective.
- Use the same structural model as before (you can just reuse it again!)

```
sec_cf_mm = constructs(
  as.reflective(composite("TRUST", multi_items("TRST", 1:4))),
  as.reflective(composite("SEC", multi_items("PSEC", 1:4))),
  as.reflective(composite("REP", multi_items("PREP", 1:4))),
  as.reflective(composite("INV", multi_items("PINV", 1:3))),
  as.reflective(composite("POL", multi_items("PPSS", 1:3))),
  as.reflective(composite("FAML", single_item("FAML1"))),
  as.reflective(interaction_term("REP", "POL", method=orthogonal))
)
```

```
sec_cf_pls = estimate_cbsem(data=data, measurement_model=sec_cf_mm, structural_model=sec_sm)
```

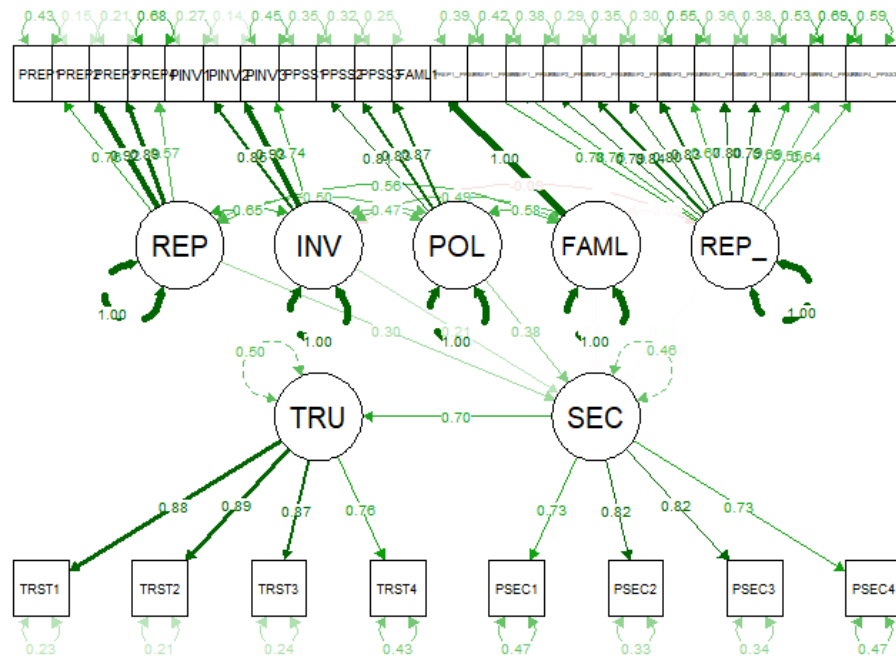
```
## Generating the semnr model for CBSEM
```

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

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

```
plot(sec_cf_pls, title="Figure of the Estimated Common Factor Model")
```





ii. Loadings of composites

```
summary(sec_cf_pls)$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(sec_cf_pls)$paths$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
```

```
summary(sec_cf_pls)$paths$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
```