

# Business Analytics using Statistical Modeling

## Assignment 15

### Question 1

```
sec <- read.csv('../15-security_data.csv', header = TRUE)
```

a. Create a PLS path model with the following characteristics:

i. Measurement of constructs by items

1. Trust in website (TRUST): reflective construct with items TRST1-4
2. Perceived security of website (SEC): reflective construct with items PSEC1-4
3. Reputation of website (REP): formative construct with items PREP1-4
4. Investment in website (INV): reflective construct with PINV1-3
5. Perception of policy (POL): reflective construct with items PPSS1-3
6. Familiarity with website (FAML): single-item construct measured by FAML1

```
library(semnr)
```

```
##  
## Attaching package: 'semnr'  
## The following object is masked from 'package:base':  
##  
##      structure  
##  
## Measurement model  
sec_mm <- measure(  
  form('REP', multi_items('PREP', 1:4)),  
  reflect('INV', multi_items('PINV', 1:3)),  
  reflect('POL', multi_items('PPSS', 1:3)),  
  reflect('FAML', single_item('FAML1')),  
  reflect('SEC', multi_items('PSEC', 1:4)),  
  reflect('TRUST', multi_items('TRST', 1:4))  
)
```

ii. Interaction between REP and POL (use orthogonalized product terms)

```
# Orthogonalized product terms
sec_int <- interact(
  interaction_ortho('REP', 'POL')
)
```

iii. Structural paths between constructs (shown as causal models – note direction of arrows):

1.  $SEC \leftarrow REP + INV + POL + FAML + REP.POL$

2.  $TRUST \leftarrow SEC$

```
# Structural model
sec_sm_int <- structure(
  paths(from = c('REP', 'INV', 'POL', 'FAML', 'REP.POL'), to = 'SEC'),
  paths(from = 'SEC', to = 'TRUST')
)

sec_pls_int <- estimate_model(
  data = sec,
  measurement_model = sec_mm,
  interactions = sec_int,
  structural_model = sec_sm_int
)
```

## Generating the plsm model

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

# i. Loadings of reflective factors / Weights of formative factors

Loadings:

```
sec_pls_int$outer_loadings
```

##	REP	INV	POL	FAML	REP.POL	SEC
## PREP1	0.5623877	0.0000000	0.0000000	0	0.0000000	0.0000000
## PREP2	0.8723808	0.0000000	0.0000000	0	0.0000000	0.0000000
## PREP3	0.9127721	0.0000000	0.0000000	0	0.0000000	0.0000000
## PREP4	0.7500201	0.0000000	0.0000000	0	0.0000000	0.0000000
## PINV1	0.0000000	0.9034426	0.0000000	0	0.0000000	0.0000000
## PINV2	0.0000000	0.9248641	0.0000000	0	0.0000000	0.0000000
## PINV3	0.0000000	0.8546257	0.0000000	0	0.0000000	0.0000000
## PPSS1	0.0000000	0.0000000	0.8678161	0	0.0000000	0.0000000
## PPSS2	0.0000000	0.0000000	0.8931576	0	0.0000000	0.0000000
## PPSS3	0.0000000	0.0000000	0.9110954	0	0.0000000	0.0000000
## FAML1	0.0000000	0.0000000	0.0000000	1	0.0000000	0.0000000
## PSEC1	0.0000000	0.0000000	0.0000000	0	0.0000000	0.8106163
## PSEC2	0.0000000	0.0000000	0.0000000	0	0.0000000	0.8652285
## PSEC3	0.0000000	0.0000000	0.0000000	0	0.0000000	0.8680697
## PSEC4	0.0000000	0.0000000	0.0000000	0	0.0000000	0.8094425
## TRST1	0.0000000	0.0000000	0.0000000	0	0.0000000	0.0000000
## TRST2	0.0000000	0.0000000	0.0000000	0	0.0000000	0.0000000
## TRST3	0.0000000	0.0000000	0.0000000	0	0.0000000	0.0000000
## TRST4	0.0000000	0.0000000	0.0000000	0	0.0000000	0.0000000
## PREP1.PPSS1	0.0000000	0.0000000	0.0000000	0	0.5836738	0.0000000
## PREP1.PPSS2	0.0000000	0.0000000	0.0000000	0	0.5125196	0.0000000
## PREP1.PPSS3	0.0000000	0.0000000	0.0000000	0	0.5088870	0.0000000
## PREP2.PPSS1	0.0000000	0.0000000	0.0000000	0	0.5134612	0.0000000
## PREP2.PPSS2	0.0000000	0.0000000	0.0000000	0	0.4244352	0.0000000
## PREP2.PPSS3	0.0000000	0.0000000	0.0000000	0	0.3398300	0.0000000
## PREP3.PPSS1	0.0000000	0.0000000	0.0000000	0	0.2395725	0.0000000
## PREP3.PPSS2	0.0000000	0.0000000	0.0000000	0	0.5576592	0.0000000
## PREP3.PPSS3	0.0000000	0.0000000	0.0000000	0	0.4690182	0.0000000
## PREP4.PPSS1	0.0000000	0.0000000	0.0000000	0	0.9011031	0.0000000
## PREP4.PPSS2	0.0000000	0.0000000	0.0000000	0	0.8363827	0.0000000
## PREP4.PPSS3	0.0000000	0.0000000	0.0000000	0	0.8599362	0.0000000
##	TRUST					
## PREP1	0.0000000					
## PREP2	0.0000000					
## PREP3	0.0000000					
## PREP4	0.0000000					
## PINV1	0.0000000					
## PINV2	0.0000000					
## PINV3	0.0000000					
## PPSS1	0.0000000					
## PPSS2	0.0000000					
## PPSS3	0.0000000					
## FAML1	0.0000000					
## PSEC1	0.0000000					
## PSEC2	0.0000000					
## PSEC3	0.0000000					
## PSEC4	0.0000000					
## TRST1	0.8997565					

```
## TRST2      0.9092024
## TRST3      0.9045716
## TRST4      0.8381919
## PREP1.PPSS1 0.0000000
## PREP1.PPSS2 0.0000000
## PREP1.PPSS3 0.0000000
## PREP2.PPSS1 0.0000000
## PREP2.PPSS2 0.0000000
## PREP2.PPSS3 0.0000000
## PREP3.PPSS1 0.0000000
## PREP3.PPSS2 0.0000000
## PREP3.PPSS3 0.0000000
## PREP4.PPSS1 0.0000000
## PREP4.PPSS2 0.0000000
## PREP4.PPSS3 0.0000000
```

Weights:

```
sec_pls_int$outer_weights
```

##	REP	INV	POL	FAML	REP.POL	SEC
## PREP1	-0.2434963	0.0000000	0.0000000	0	0.00000000	0.0000000
## PREP2	0.4418170	0.0000000	0.0000000	0	0.00000000	0.0000000
## PREP3	0.5144356	0.0000000	0.0000000	0	0.00000000	0.0000000
## PREP4	0.3759155	0.0000000	0.0000000	0	0.00000000	0.0000000
## PINV1	0.0000000	0.3633141	0.0000000	0	0.00000000	0.0000000
## PINV2	0.0000000	0.3949797	0.0000000	0	0.00000000	0.0000000
## PINV3	0.0000000	0.3585944	0.0000000	0	0.00000000	0.0000000
## PPSS1	0.0000000	0.0000000	0.3604446	0	0.00000000	0.0000000
## PPSS2	0.0000000	0.0000000	0.3947092	0	0.00000000	0.0000000
## PPSS3	0.0000000	0.0000000	0.3673192	0	0.00000000	0.0000000
## FAML1	0.0000000	0.0000000	0.0000000	1	0.00000000	0.0000000
## PSEC1	0.0000000	0.0000000	0.0000000	0	0.00000000	0.2708136
## PSEC2	0.0000000	0.0000000	0.0000000	0	0.00000000	0.3163641
## PSEC3	0.0000000	0.0000000	0.0000000	0	0.00000000	0.3064275
## PSEC4	0.0000000	0.0000000	0.0000000	0	0.00000000	0.2974225
## TRST1	0.0000000	0.0000000	0.0000000	0	0.00000000	0.0000000
## TRST2	0.0000000	0.0000000	0.0000000	0	0.00000000	0.0000000
## TRST3	0.0000000	0.0000000	0.0000000	0	0.00000000	0.0000000
## TRST4	0.0000000	0.0000000	0.0000000	0	0.00000000	0.0000000
## PREP1.PPSS1	0.0000000	0.0000000	0.0000000	0	0.23907580	0.0000000
## PREP1.PPSS2	0.0000000	0.0000000	0.0000000	0	0.03097791	0.0000000
## PREP1.PPSS3	0.0000000	0.0000000	0.0000000	0	0.02135493	0.0000000
## PREP2.PPSS1	0.0000000	0.0000000	0.0000000	0	0.04728616	0.0000000
## PREP2.PPSS2	0.0000000	0.0000000	0.0000000	0	-0.10299005	0.0000000
## PREP2.PPSS3	0.0000000	0.0000000	0.0000000	0	-0.22637248	0.0000000
## PREP3.PPSS1	0.0000000	0.0000000	0.0000000	0	-0.33874387	0.0000000
## PREP3.PPSS2	0.0000000	0.0000000	0.0000000	0	0.09519373	0.0000000
## PREP3.PPSS3	0.0000000	0.0000000	0.0000000	0	0.10857661	0.0000000
## PREP4.PPSS1	0.0000000	0.0000000	0.0000000	0	0.44220480	0.0000000
## PREP4.PPSS2	0.0000000	0.0000000	0.0000000	0	0.38056606	0.0000000
## PREP4.PPSS3	0.0000000	0.0000000	0.0000000	0	0.27087812	0.0000000
##	TRUST					
## PREP1	0.0000000					
## PREP2	0.0000000					
## PREP3	0.0000000					
## PREP4	0.0000000					
## PINV1	0.0000000					
## PINV2	0.0000000					
## PINV3	0.0000000					
## PPSS1	0.0000000					
## PPSS2	0.0000000					
## PPSS3	0.0000000					
## FAML1	0.0000000					
## PSEC1	0.0000000					
## PSEC2	0.0000000					
## PSEC3	0.0000000					
## PSEC4	0.0000000					
## TRST1	0.2820685					
## TRST2	0.2803331					
## TRST3	0.2856284					

```
## TRST4          0.2779270
## PREP1.PPSS1 0.0000000
## PREP1.PPSS2 0.0000000
## PREP1.PPSS3 0.0000000
## PREP2.PPSS1 0.0000000
## PREP2.PPSS2 0.0000000
## PREP2.PPSS3 0.0000000
## PREP3.PPSS1 0.0000000
## PREP3.PPSS2 0.0000000
## PREP3.PPSS3 0.0000000
## PREP4.PPSS1 0.0000000
## PREP4.PPSS2 0.0000000
## PREP4.PPSS3 0.0000000
```

## ii. Regression coefficients of paths between factors

```
print_paths(sec_pls_int, digits = 2)
```

```
##          SEC TRUST
## R^2      0.44  0.37
## REP      0.30    .
## INV      0.17    .
## POL      0.32    .
## FAML      0.01    .
## REP.POL -0.11    .
## SEC      .    0.61
```

## iii. Bootstrapped path coefficients: t-values, p-values (are any paths not significant at $p=5\%$ ?)

```
boot_pls <- bootstrap_model(
  data = sec,
  measurement_model = sec_mm,
  interactions = sec_int,
  structural_model = sec_sm_int,
  nboot = 1000
)
```

```
## Bootstrapping model using simplePLS...
```

```
print_paths(boot_pls)
```

```
##          SEC PLS Est. SEC Boot Mean SEC Boot SE t value Pr(>|t|)
## REP          0.30          0.31          0.06    5.52    0.00
## INV          0.17          0.17          0.05    3.25    0.00
## POL          0.32          0.32          0.05    5.94    0.00
## FAML         0.01          0.01          0.05    0.13    0.90
## REP.POL      -0.11         -0.02          0.13   -0.18    0.85
## SEC          0.00          0.00          0.00    0.00    0.00
##          TRUST PLS Est. TRUST Boot Mean TRUST Boot SE t value Pr(>|t|)
## REP          0.00          0.00          0.00    0.00    0.00
## INV          0.00          0.00          0.00    0.00    0.00
## POL          0.00          0.00          0.00    0.00    0.00
## FAML         0.00          0.00          0.00    0.00    0.00
## REP.POL      0.00          0.00          0.00    0.00    0.00
## SEC          0.61          0.61          0.03   17.66    0.00
```

FAML and REP.POL are not significant at  $p = 5\%$ .