

Psychology 454: Psychological Measurement

Problem set # 2

William Revelle
Swift 315
email: revelle@northwestern.edu

January 13, 2009

The following problems assume that you have installed R on your machine and have also installed the **psych** and **sem** packages.

1. Create a data set of raw scores with the **sim.congeneric** function and a factor structure defined by loadings of .6, .5, .4, and .3 for 200 subjects. Draw the scatter plot matrix of correlations for this data set using the **pairs.panels** function.

Find the descriptive statistics using the **describe** function.

2. For the following figure (Fig 1) , what are the implied correlations?
3. Simulate a congeneric correlation matrix using the default values using the **sim.congeneric** function
4. Find the factor structure for this using the **factanal** function.
5. Do a confirmatory factor analysis using the **sem** function.
6. Repeat the previous three items for sample sizes of 100, 200 and 400 cases.
7. What are the implied correlations in Figure 2

```
> F <- c(0.9, 0.8, 0.7, 0.6)
> structure.graph(F)
```

Structural model

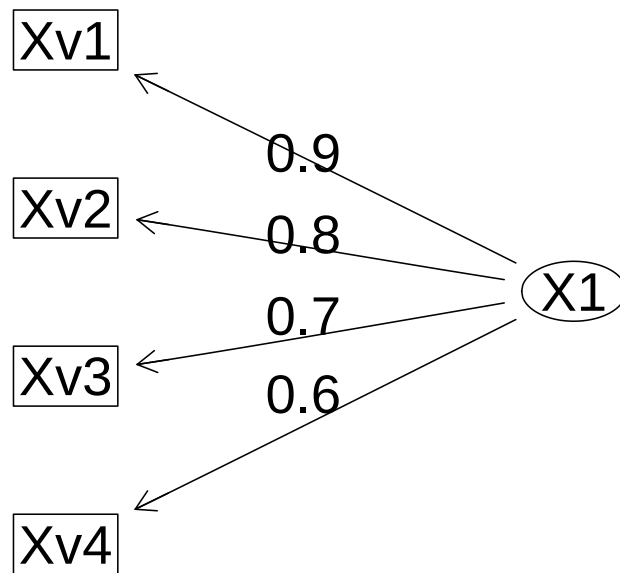


Figure 1: A congenetic model drawn using the `structure.graph` function.

```
> F <- matrix(c("a", "b", rep(0, 4), "c", "d"), ncol = 2)
> Phi <- matrix(c(1, "e", "e", 1), ncol = 2)
> structure.graph(F, Phi = Phi)
```

Structural model

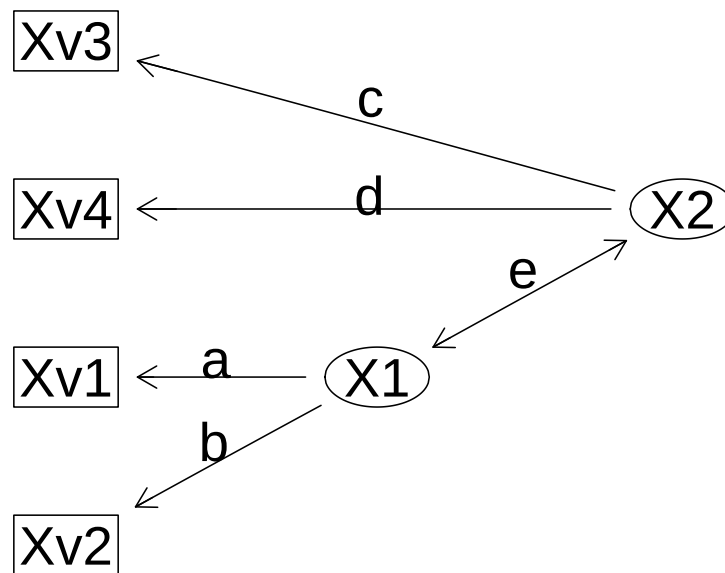


Figure 2: Two correlated factors. The structure was drawn using the `structure.graph` function.

1 Answers to problem set #2

```
1. > X <- sim.congeneric(N = 200, loads = c(0.6, 0.5, 0.4, 0.3), short = FALSE)
> dim(X$observed)

[1] 200    4

> describe(X$observed)
```

	var	n	mean	sd	median	trimmed	mad	min	max	range	skew	kurtosis	se
V1	1	200	0.03	0.96	0.01	0.01	0.95	-2.18	2.82	5.00	0.20	0.01	0.07
V2	2	200	0.09	1.00	0.07	0.08	1.03	-2.33	3.12	5.44	0.10	-0.36	0.07
V3	3	200	0.04	0.96	-0.07	0.01	0.88	-3.33	3.07	6.40	0.30	0.69	0.07
V4	4	200	0.10	0.96	0.10	0.11	0.94	-2.58	2.76	5.34	-0.02	0.04	0.07

```
> pairs.panels(X$observed)
```

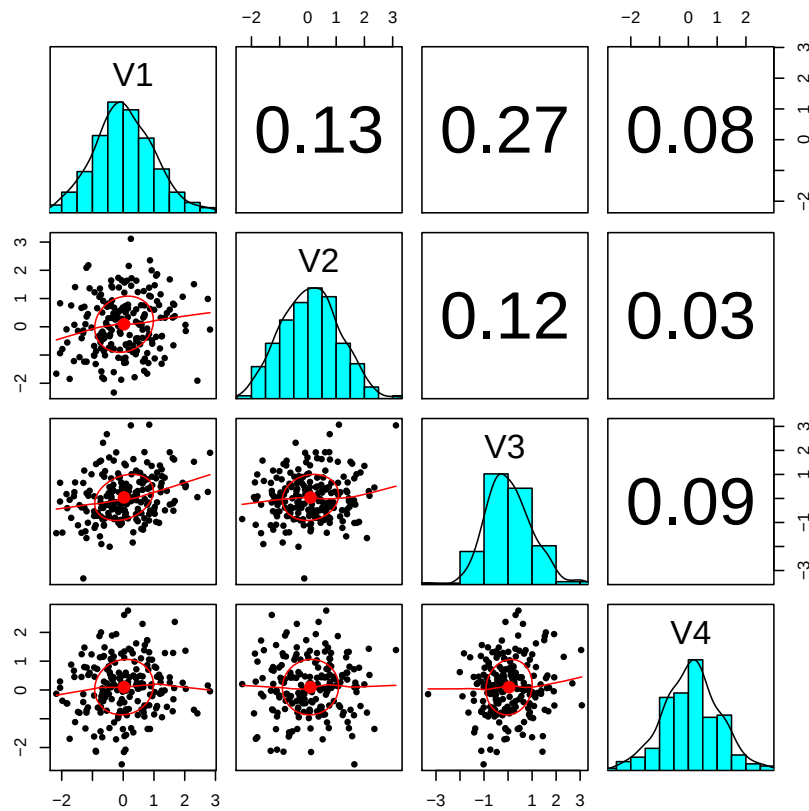


Figure 3: A scatter plot matrix (SPLOM) will show the pairwise correlations, their pairwise scatter plots (below the diagonal) as well as a local smooth

2. Implied correlations may be found by the factor model: $R = F\vec{F}'$

```
> F <- c(0.9, 0.8, 0.7, 0.6)
> R <- F %*% t(F)
> R
```

```
      [,1] [,2] [,3] [,4]
[1,] 0.81 0.72 0.63 0.54
```

```

[2,] 0.72 0.64 0.56 0.48
[3,] 0.63 0.56 0.49 0.42
[4,] 0.54 0.48 0.42 0.36

3. > R <- sim.congeneric()
> R

      V1  V2  V3  V4
V1 1.00 0.56 0.48 0.40
V2 0.56 1.00 0.42 0.35
V3 0.48 0.42 1.00 0.30
V4 0.40 0.35 0.30 1.00

4. > factanal(covmat = R, factors = 1)

Call:
factanal(factors = 1, covmat = R)

Uniquenesses:
  V1  V2  V3  V4
0.36 0.51 0.64 0.75

Loadings:
  Factor1
V1 0.8
V2 0.7
V3 0.6
V4 0.5

      Factor1
SS loadings 1.740
Proportion Var 0.435

The degrees of freedom for the model is 2 and the fit was 0

```

5. The structural equation functions are found in the `sem` package

```

> library(sem)
> model = matrix(c("Latent -> V1", "a", NA, "Latent -> V2", "b", NA, "Latent -> V3", "c",
+ NA, "Latent -> V4", "d", NA, "V1 <-> V1", "u", NA, "V2 <-> V2", "v", NA, "V3 <-> V3",
+ "w", NA, "V4 <-> V4", "x", NA, "Latent <-> Latent", NA, 1), ncol = 3, byrow = TRUE)
> model

      [,1]      [,2] [,3]
[1,] "Latent -> V1"    "a"  NA
[2,] "Latent -> V2"    "b"  NA
[3,] "Latent -> V3"    "c"  NA
[4,] "Latent -> V4"    "d"  NA
[5,] "V1 <-> V1"      "u"  NA
[6,] "V2 <-> V2"      "v"  NA
[7,] "V3 <-> V3"      "w"  NA
[8,] "V4 <-> V4"      "x"  NA
[9,] "Latent <-> Latent" NA  "1"

> sem4 = sem(model, R, 1000)
> summary(sem4)

Model Chisquare = 1.8894e-09 Df = 2 Pr(>Chisq) = 1

```

```

Chisquare (null model) = 894.56   Df = 6
Goodness-of-fit index = 1
Adjusted goodness-of-fit index = 1
RMSEA index = 0   90% CI: (NA, NA)
Bentler-Bonnett NFI = 1
Tucker-Lewis NNFI = 1.0068
Bentler CFI = 1
SRMR = 6.4327e-07
BIC = -13.816

```

```

Normalized Residuals
      Min.   1st Qu.   Median     Mean  3rd Qu.     Max.
3.00e-06 5.52e-06 9.38e-06 1.10e-05 1.48e-05 3.77e-05

```

```

Parameter Estimates
      Estimate Std Error z value Pr(>|z|)
a 0.80      0.032181  24.860  0      V1 <--- Latent
b 0.70      0.032359  21.632  0      V2 <--- Latent
c 0.60      0.032771  18.309  0      V3 <--- Latent
d 0.50      0.033583  14.889  0      V4 <--- Latent
u 0.36      0.034180  10.532  0      V1 <--> V1
v 0.51      0.033047  15.433  0      V2 <--> V2
w 0.64      0.034422  18.593  0      V3 <--> V3
x 0.75      0.037129  20.200  0      V4 <--> V4

```

```

Iterations = 13

```

6. Redoing the analysis for different sample sizes partly shows sample variability in the parameters, but also shows that the χ^2 statistic varies directly as a function of sample size.

```

> R100 <- sim.congeneric(N = 100)
> R200 <- sim.congeneric(N = 200)
> R400 <- sim.congeneric(N = 200)
> factanal(covmat = R100, factors = 1, n.obs = 100)

```

```

Call:
factanal(factors = 1, covmat = R100, n.obs = 100)

```

```

Uniquenesses:
      V1   V2   V3   V4
0.36 0.51 0.64 0.75

```

```

Loadings:
      Factor1
V1 0.8
V2 0.7
V3 0.6
V4 0.5

```

```

      Factor1
SS loadings      1.740
Proportion Var   0.435

```

Test of the hypothesis that 1 factor is sufficient.
The chi square statistic is 0 on 2 degrees of freedom.

The p-value is 1

```
> factanal(covmat = R200, factors = 1, n.obs = 200)
```

Call:

```
factanal(factors = 1, covmat = R200, n.obs = 200)
```

Uniquenesses:

V1	V2	V3	V4
0.36	0.51	0.64	0.75

Loadings:

	Factor1
V1	0.8
V2	0.7
V3	0.6
V4	0.5

	Factor1
SS loadings	1.740
Proportion Var	0.435

Test of the hypothesis that 1 factor is sufficient.
The chi square statistic is 0 on 2 degrees of freedom.
The p-value is 1

```
> factanal(covmat = R400, factors = 1, n.obs = 400)
```

Call:

```
factanal(factors = 1, covmat = R400, n.obs = 400)
```

Uniquenesses:

V1	V2	V3	V4
0.36	0.51	0.64	0.75

Loadings:

	Factor1
V1	0.8
V2	0.7
V3	0.6
V4	0.5

	Factor1
SS loadings	1.740
Proportion Var	0.435

Test of the hypothesis that 1 factor is sufficient.
The chi square statistic is 0 on 2 degrees of freedom.
The p-value is 1

```
> sem1 = sem(model, R100, 100)
> sem2 = sem(model, R200, 200)
> sem4 = sem(model, R400, 400)
> summary(sem1)
```

```
Model Chisquare = 1.8724e-10 Df = 2 Pr(>Chisq) = 1
Chisquare (null model) = 88.65 Df = 6
```

```

Goodness-of-fit index = 1
Adjusted goodness-of-fit index = 1
RMSEA index = 0 90% CI: (NA, NA)
Bentler-Bonnett NFI = 1
Tucker-Lewis NNFI = 1.0726
Bentler CFI = 1
SRMR = 6.4327e-07
BIC = -9.2103

Normalized Residuals
      Min.  1st Qu.  Median    Mean 3rd Qu.    Max.
9.45e-07 1.74e-06 2.95e-06 3.46e-06 4.67e-06 1.19e-05

Parameter Estimates
      Estimate Std Error z value Pr(>|z|)
a 0.80      0.10223   7.8258 5.1070e-15 V1 <--- Latent
b 0.70      0.10279   6.8098 9.7764e-12 V2 <--- Latent
c 0.60      0.10410   5.7637 8.2292e-09 V3 <--- Latent
d 0.50      0.10668   4.6869 2.7731e-06 V4 <--- Latent
u 0.36      0.10858   3.3156 9.1445e-04 V1 <--> V1
v 0.51      0.10498   4.8582 1.1846e-06 V2 <--> V2
w 0.64      0.10935   5.8530 4.8284e-09 V3 <--> V3
x 0.75      0.11795   6.3589 2.0324e-10 V4 <--> V4

Iterations = 13

> summary(sem2)

Model Chisquare = 3.7636e-10 Df = 2 Pr(>Chisq) = 1
Chisquare (null model) = 178.20 Df = 6
Goodness-of-fit index = 1
Adjusted goodness-of-fit index = 1
RMSEA index = 0 90% CI: (NA, NA)
Bentler-Bonnett NFI = 1
Tucker-Lewis NNFI = 1.0348
Bentler CFI = 1
SRMR = 6.4327e-07
BIC = -10.597

Normalized Residuals
      Min.  1st Qu.  Median    Mean 3rd Qu.    Max.
1.34e-06 2.46e-06 4.19e-06 4.91e-06 6.62e-06 1.68e-05

Parameter Estimates
      Estimate Std Error z value Pr(>|z|)
a 0.80      0.072102 11.0953 0.0000e+00 V1 <--- Latent
b 0.70      0.072503   9.6547 0.0000e+00 V2 <--- Latent
c 0.60      0.073425   8.1717 2.2204e-16 V3 <--- Latent
d 0.50      0.075244   6.6451 3.0309e-11 V4 <--- Latent
u 0.36      0.076583   4.7008 2.5915e-06 V1 <--> V1
v 0.51      0.074043   6.8879 5.6641e-12 V2 <--> V2
w 0.64      0.077125   8.2982 0.0000e+00 V3 <--> V3
x 0.75      0.083190   9.0155 0.0000e+00 V4 <--> V4

Iterations = 13

```



```

> summary(sem4)

Model Chisquare = 7.5461e-10 Df = 2 Pr(>Chisq) = 1
Chisquare (null model) = 357.29 Df = 6
Goodness-of-fit index = 1
Adjusted goodness-of-fit index = 1
RMSEA index = 0 90% CI: (NA, NA)
Bentler-Bonnett NFI = 1
Tucker-Lewis NNFI = 1.0171
Bentler CFI = 1
SRMR = 6.4327e-07
BIC = -11.983

Normalized Residuals
      Min.  1st Qu.  Median    Mean  3rd Qu.    Max.
1.90e-06 3.49e-06 5.93e-06 6.95e-06 9.38e-06 2.38e-05

Parameter Estimates
      Estimate Std Error z value Pr(>|z|)
a 0.80      0.050920 15.7109 0.0000e+00 V1 <--- Latent
b 0.70      0.051203 13.6710 0.0000e+00 V2 <--- Latent
c 0.60      0.051854 11.5710 0.0000e+00 V3 <--- Latent
d 0.50      0.053139  9.4093 0.0000e+00 V4 <--- Latent
u 0.36      0.054084  6.6563 2.8084e-11 V1 <--> V1
v 0.51      0.052291  9.7531 0.0000e+00 V2 <--> V2
w 0.64      0.054467 11.7502 0.0000e+00 V3 <--> V3
x 0.75      0.058751 12.7658 0.0000e+00 V4 <--> V4

Iterations = 13

```

7. The implied correlation matrix from Figure 2 is found by path tracing.

	X1	X2	X3	X4
X1	1			
X2	ab	1		
X3	aec	bec	1	
X4	aed	bed	cd	1