

Analysis of Unstructured Covariance Matrices with lavaan*

The Longitudinal IQ data are taken from *The Statistical Sleuth* by F. Ramsey and D. Schafer. The source is Marie Skodak and Harold M. Skeels, "A final follow-up study of one hundred adopted children," *Journal of Genetic Psychology*, 1949, 75, 85-125. Data are used here without permission. The article by Skodak and Skeels includes the complete data set, including variables not in the data set that appears in Ramsey and Schafer's book.

The variables used here (from Ramsey and Schafer) are

- Adoptive mother's education
- Birth mother's IQ
- Child's IQ at age 2
- Child's IQ at age 4
- Child's IQ at age 8
- Child's IQ at age 13

```
> # install.packages("lavaan", dependencies = TRUE) # Only need to do this once
> library(lavaan)
This is lavaan 0.6-15
lavaan is FREE software! Please report any bugs.
> longIQ =
read.table("https://www.utstat.toronto.edu/brunner/data/illegal/origIQ.data.txt")
> colnames(longIQ) = c("amEduc", "bmIQ", "IQ2", "IQ4", "IQ8", "IQ13")
> head(longIQ)
  amEduc  bmIQ  IQ2  IQ4  IQ8  IQ13
1      10   100  120  115  109   106
2      10    71  131  109  113    95
3      14    89  126  115  113    90
4       7    73  120  102  111   121
5      14    64  126  125  114    96
6       8    64  125  109   96    87
>
> dim(longIQ)
[1] 62  6
> r = round(cor(longIQ), 3); r
     amEduc   bmIQ    IQ2    IQ4    IQ8    IQ13
amEduc  1.000  0.070 -0.095  0.014  0.023 -0.004
bmIQ    0.070  1.000  0.042  0.272  0.362  0.380
IQ2    -0.095  0.042  1.000  0.509  0.476  0.300
IQ4     0.014  0.272  0.509  1.000  0.646  0.539
IQ8     0.023  0.362  0.476  0.646  1.000  0.774
IQ13   -0.004  0.380  0.300  0.539  0.774  1.000
> # Under H0: rho = 0, t = r * sqrt(n-2) / sqrt(1-r^2) ~ t(n-2)
> n = dim(longIQ)[1]; n
[1] 62
> critval = qt(0.975,n-2); critval
[1] 2.000298
> tstat = r * sqrt(n-2) / sqrt(1-r^2); round(tstat,3)
     amEduc   bmIQ    IQ2    IQ4    IQ8    IQ13
amEduc  Inf  0.544 -0.739  0.108  0.178 -0.031
bmIQ    0.544  Inf   0.326  2.189  3.008  3.182
IQ2    -0.739  0.326  Inf   4.580  4.193  2.436
IQ4     0.108  2.189  4.580  Inf   6.555  4.957
IQ8     0.178  3.008  4.193  6.555  Inf   9.469
IQ13   -0.031  3.182  2.436  4.957  9.469  Inf
```

*This handout was prepared by Jerry Brunner, Department of Statistical Sciences, University of Toronto. It is licensed under a Creative Commons Attribution - ShareAlike 3.0 Unported License. Use any part of it as you like and share the result freely. The data are used without permission and are not covered by the CC license. The OpenOffice.org document is available from the course website:

<http://www.utstat.toronto.edu/brunner/oldclass/431s23>

```

> # In this model, the only parameters are variances and covariances
> # of the observable variables.
> covmod0 = "amEduc ~~ amEduc + bmiQ + IQ2 + IQ4 + IQ8 + IQ13
+                      bmiQ ~~ bmiQ + IQ2 + IQ4 + IQ8 + IQ13
+                      IQ2 ~~ IQ2 + IQ4 + IQ8 + IQ13
+                      IQ4 ~~ IQ4 + IQ8 + IQ13
+                      IQ8 ~~ IQ8 + IQ13
+                      IQ13 ~~ IQ13"
> fit0 = lavaan(covmod0,data=longIQ)
>
> summary(fit0)
lavaan 0.6.13 ended normally after 222 iterations

```

Estimator ML
 Optimization method NLMINB
 Number of model parameters 21

Number of observations 62

Model Test User Model:

Test statistic	0.000
Degrees of freedom	0

Parameter Estimates:

	Standard errors	Standard
	Information	Expected
	Information saturated (h1) model	Structured

Covariances:

	Estimate	Std.Err	z-value	P(> z)
amEduc ~~				
bmiQ	3.209	5.816	0.552	0.581
IQ2	-3.542	4.755	-0.745	0.456
IQ4	0.539	4.726	0.114	0.909
IQ8	0.892	4.960	0.180	0.857
IQ13	-0.157	5.582	-0.028	0.978
bmiQ ~~				
IQ2	8.625	26.031	0.331	0.740
IQ4	55.577	26.905	2.066	0.039
IQ8	77.660	28.970	2.681	0.007
IQ13	91.703	32.802	2.796	0.005
IQ2 ~~				
IQ4	84.854	23.769	3.570	0.000
IQ8	83.373	24.621	3.386	0.001
IQ13	59.065	26.123	2.261	0.024
IQ4 ~~				
IQ8	112.841	26.415	4.272	0.000
IQ13	106.089	28.382	3.738	0.000
IQ8 ~~				
IQ13	159.709	33.140	4.819	0.000

Variances:

	Estimate	Std.Err	z-value	P(> z)
amEduc	8.315	1.493	5.568	0.000
bmiQ	250.958	45.073	5.568	0.000
IQ2	167.104	30.013	5.568	0.000
IQ4	166.532	29.910	5.568	0.000
IQ8	183.314	32.924	5.568	0.000
IQ13	232.310	41.724	5.568	0.000

```
> # Okay, that's kind of reasonable. Try to make them correlations.
```

```

> fit0.1 = lavaan(covmod0, data=longIQ, std.ov=TRUE)
> # summary(fit0.1)

> fitted(fit0.1) # Sigma(thetahat)
$cov
    amEduc   bmiIQ    IQ2     IQ4     IQ8    IQ13
amEduc  0.984
bmiIQ   0.069  0.984
IQ2    -0.093  0.041  0.984
IQ4    0.014  0.267  0.500  0.984
IQ8    0.022  0.356  0.469  0.635  0.984
IQ13   -0.004  0.374  0.295  0.531  0.761  0.984

> r # For comparison
    amEduc   bmiIQ    IQ2     IQ4     IQ8    IQ13
amEduc  1.000  0.070 -0.095  0.014  0.023 -0.004
bmiIQ   0.070  1.000  0.042  0.272  0.362  0.380
IQ2    -0.095  0.042  1.000  0.509  0.476  0.300
IQ4    0.014  0.272  0.509  1.000  0.646  0.539
IQ8    0.023  0.362  0.476  0.646  1.000  0.774
IQ13   -0.004  0.380  0.300  0.539  0.774  1.000

> fitted(fit0.1)$cov * n/(n-1)

    amEduc   bmiIQ    IQ2     IQ4     IQ8    IQ13
amEduc  1.000
bmiIQ   0.070  1.000
IQ2    -0.095  0.042  1.000
IQ4    0.014  0.272  0.509  1.000
IQ8    0.023  0.362  0.476  0.646  1.000
IQ13   -0.004  0.380  0.300  0.539  0.774  1.000

> # So asymptotically it's fine, but I want the MLEs to be sample correlations.
> # Standardize the variables myself.
>
> standIQ = scale(longIQ) * sqrt(n/(n-1))
> fit0.2 = lavaan(covmod0, data=standIQ)
> fitted(fit0.2); r # Sigma(thetahat) should be correlation matrix
$cov
    amEduc   bmiIQ    IQ2     IQ4     IQ8    IQ13
amEduc  1.000
bmiIQ   0.070  1.000
IQ2    -0.095  0.042  1.000
IQ4    0.014  0.272  0.509  1.000
IQ8    0.023  0.362  0.476  0.646  1.000
IQ13   -0.004  0.380  0.300  0.539  0.774  1.000

    amEduc   bmiIQ    IQ2     IQ4     IQ8    IQ13
amEduc  1.000  0.070 -0.095  0.014  0.023 -0.004
bmiIQ   0.070  1.000  0.042  0.272  0.362  0.380
IQ2    -0.095  0.042  1.000  0.509  0.476  0.300
IQ4    0.014  0.272  0.509  1.000  0.646  0.539
IQ8    0.023  0.362  0.476  0.646  1.000  0.774
IQ13   -0.004  0.380  0.300  0.539  0.774  1.000

```

```

> # Now want names for parameters so I can test differences, etc.
>
> covmod1 =
+ "amEduc ~~ amEduc + rho12*bmiIQ + rho13*IQ2 + rho14*IQ4 + rho15*IQ8 + rho16*IQ13
+           bmiIQ ~~ bmiIQ + rho23*IQ2 + rho24*IQ4 + rho25*IQ8 + rho26*IQ13
+           IQ2 ~~ IQ2 + rho34*IQ4 + rho35*IQ8 + rho36*IQ13
+           IQ4 ~~ IQ4 + rho45*IQ8 + rho46*IQ13
+           IQ8 ~~ IQ8 + rho56*IQ13
+           IQ13 ~~ IQ13"
> fit1 = lavaan(covmod1, data=standIQ) # This will be the full model.
> parameterEstimates(fit1)
   lhs op   rhs label    est     se      z pvalue ci.lower ci.upper
1 amEduc ~~ amEduc    1.000 0.180  5.568  0.000  0.648  1.352
2 amEduc ~~ bmiIQ rho12  0.070 0.127  0.552  0.581 -0.179  0.320
3 amEduc ~~ IQ2 rho13 -0.095 0.128 -0.745  0.456 -0.345  0.155
4 amEduc ~~ IQ4 rho14  0.014 0.127  0.114  0.909 -0.234  0.263
5 amEduc ~~ IQ8 rho15  0.023 0.127  0.180  0.857 -0.226  0.272
6 amEduc ~~ IQ13 rho16 -0.004 0.127 -0.028  0.978 -0.252  0.245
7 bmiIQ ~~ bmiIQ     1.000 0.180  5.568  0.000  0.648  1.352
8 bmiIQ ~~ IQ2 rho23  0.042 0.127  0.331  0.740 -0.207  0.291
9 bmiIQ ~~ IQ4 rho24  0.272 0.132  2.066  0.039  0.014  0.530
10 bmiIQ ~~ IQ8 rho25  0.362 0.135  2.681  0.007  0.097  0.627
11 bmiIQ ~~ IQ13 rho26 0.380 0.136  2.796  0.005  0.114  0.646
12 IQ2 ~~ IQ2       1.000 0.180  5.568  0.000  0.648  1.352
13 IQ2 ~~ IQ4 rho34  0.509 0.142  3.570  0.000  0.229  0.788
14 IQ2 ~~ IQ8 rho35  0.476 0.141  3.386  0.001  0.201  0.752
15 IQ2 ~~ IQ13 rho36 0.300 0.133  2.261  0.024  0.040  0.560
16 IQ4 ~~ IQ4       1.000 0.180  5.568  0.000  0.648  1.352
17 IQ4 ~~ IQ8 rho45  0.646 0.151  4.272  0.000  0.350  0.942
18 IQ4 ~~ IQ13 rho46 0.539 0.144  3.738  0.000  0.257  0.822
19 IQ8 ~~ IQ8       1.000 0.180  5.568  0.000  0.648  1.352
20 IQ8 ~~ IQ13 rho56 0.774 0.161  4.819  0.000  0.459  1.089
21 IQ13 ~~ IQ13    1.000 0.180  5.568  0.000  0.648  1.352
> # fitted(fit1); r # Checks

> # Likelihood ratio test of relationship between adoptive mom's education
> # and kid's IQ at any age
> noAM = lavaan(covmod1, data=standIQ, constraints = "rho13==0
+ rho14==0
+ rho15==0
+ rho16==0")
> anova(noAM,fit1)
Chi-Squared Difference Test

   Df      AIC      BIC  Chisq Chisq diff  RMSEA Df diff Pr(>Chisq)
fit1  0  970.22 1014.89 0.0000
noAM  4  963.43  999.59 1.2108     1.2108      0        4      0.8763
>
> # Likelihood ratio test of relationship between birth mom's IQ
> # and kid's IQ at any age
>
> noBM = lavaan(covmod1, data=standIQ, constraints = "rho23==0
+ rho24==0
+ rho25==0
+ rho26==0")
> anova(noBM,fit1)

Chi-Squared Difference Test

   Df      AIC      BIC  Chisq Chisq diff  RMSEA Df diff Pr(>Chisq)
fit1  0  970.22 1014.9  0.000
noBM  4  974.47 1010.6 12.256    12.256  0.18246      4      0.01555 *
---
Signif. Codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

> # Multivariate regression yields p = 0.022

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> # Likelihood ratio test: Is birth mother's IQ correlated with child's IQ
> # differently at different ages?
>
> BMsame = lavaan(covmod1, data=standIQ, constraints = "rho23==rho24
+                                         rho24==rho25
+                                         rho25==rho26")
> anova(BMsame,fit1)

Chi-Squared Difference Test

      Df      AIC      BIC   Chisq Chisq diff    RMSEA Df diff Pr(>Chisq)
fit1     0  970.22 1014.9  0.0000
BMsame   3  970.81 1009.1  6.5956   6.5956  0.13904       3     0.08597 .
---
Signif. codes:  0 '****' 0.001 '***' 0.01 '**' 0.05 '*' 0.1 '.' 1

> # This is surprising. Try a Wald test
> # For Wald tests: Wtest = function(L,Tn,Vn,h=0) # H0: L theta = h
> source("https://www.utstat.toronto.edu/brunner/openSEM/fun/Wtest.txt")
>
> Vhat = vcov(fit1) # Asymptotic covariance matrix of the estimates
> rhohat = coef(fit1); rhohat
amEduc~~amEduc          rho12        rho13        rho14        rho15        rho16
      1.000      0.070      -0.095      0.014      0.023      -0.004
bmiIQ~~bmiIQ          rho23        rho24        rho25        rho26      IQ2~~IQ2
      1.000      0.042      0.272      0.362      0.380      1.000
rho34                  rho35        rho36      IQ4~~IQ4      rho45      rho46
      0.509      0.476      0.300      1.000      0.646      0.539
IQ8~~IQ8              rho56      IQ13~~IQ13
      1.000      0.774      1.000
> length(rhohat)
[1] 21
> rbind(names(rhohat))
[,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9]
[1,] "amEduc~~amEduc" "rho12" "rho13" "rho14" "rho15" "rho16" "bmiIQ~~bmiIQ" "rho23" "rho24"
[,10] [,11] [,12] [,13] [,14] [,15] [,16] [,17] [,18] [,19]
[1,] "rho25" "rho26" "IQ2~~IQ2" "rho34" "rho35" "rho36" "IQ4~~IQ4" "rho45" "rho46"
"IQ8~~IQ8"
[,20] [,21]
[1,] "rho56" "IQ13~~IQ13"
>
> L0 = matrix(0,3,21)
> L0[1,8] = 1; L0[1,9] = -1 # rho23==rho24
> L0[2,9] = 1; L0[2,10] = -1 # rho24==rho25
> L0[3,10] = 1; L0[3,11] = -1 # rho25==rho26
>
> Wtest(L0,rhohat,Vhat)
      W      df p-value
5.683357 3.000000 0.128074

> # Just test 2 years old versus 13
>
> Two_vs_13 = matrix(0,1,21); Two_vs_13[1,8] = 1; Two_vs_13[1,11] = -1
> Wtest(Two_vs_13,rhohat,Vhat)
      W      df p-value
4.66802402 1.00000000 0.03072926

>
> # However, Bonferroni correcting for 6 pairwise tests, need p < ...
> 0.05/6
[1] 0.008333333

> # One could bootstrap, but IQ scores are well known to be normally distributed.

```