

Answers to Exercises Basic CFA

```
library("lavaan")
```

Exercise 3.2

```
Health.cov <- lav_matrix_lower2full(c(
  0.77,
  0.38, 0.65,
  0.39, 0.39, 0.62,
-0.25,-0.32,-0.27, 6.09,
  0.31, 0.29, 0.26,-0.36, 7.67,
  0.24, 0.25, 0.19,-0.18, 0.51, 1.69,
-3.16,-3.56,-2.63, 6.09,-3.12,-4.58,204.79,
-0.92,-0.88,-0.72, 0.88,-1.49,-1.41, 16.53, 7.24
))
rownames(Health.cov) <- colnames(Health.cov) <-
  c("Dep1","Dep2","Dep3", "SocAct", "Falls", "ChronCond",
    "PhysAct", "PersMob")
```

a)

```
Health.mod <- '
  PPsyHealth =~ Dep1 + Dep2 + Dep3 + SocAct
  PPhysHealth =~ ChronCond + PhysAct + Falls
  PersMob ~ PPsyHealth + PPhysHealth
'

Health.fit <- sem(Health.mod, sample.cov = Health.cov, sample.nobs = 6053)
summary(Health.fit, fit.measures = TRUE, standardized = TRUE)
```

```
## lavaan 0.6-5 ended normally after 62 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      18
##
##      Number of observations          6053
##
## Model Test User Model:
##
##      Test statistic                  254.865
##      Degrees of freedom              18
##      P-value (Chi-square)            0.000
##
## Model Test Baseline Model:
##
##      Test statistic                  10290.938
##      Degrees of freedom              28
```

```

##      P-value                                0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)                0.977
##      Tucker-Lewis Index (TLI)                  0.964
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)              -95467.244
##      Loglikelihood unrestricted model (H1)       -95339.812
##
##      Akaike (AIC)                              190970.488
##      Bayesian (BIC)                             191091.238
##      Sample-size adjusted Bayesian (BIC)        191034.038
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                    0.047
##      90 Percent confidence interval - lower      0.042
##      90 Percent confidence interval - upper      0.052
##      P-value RMSEA <= 0.05                      0.856
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                    0.027
##
## Parameter Estimates:
##
##      Information                                Expected
##      Information saturated (h1) model           Structured
##      Standard errors                           Standard
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      PPsyHealth =~
##      Dep1      1.000
##      Dep2      1.015    0.020  49.994    0.000    0.624    0.712
##      Dep3      0.972    0.020  49.608    0.000    0.634    0.786
##      Dep3      0.972    0.020  49.608    0.000    0.607    0.771
##      SocAct    -0.771    0.056 -13.683    0.000   -0.481   -0.195
##      PPhysHealth =~
##      ChronCond 1.000
##      PhysAct   -12.005    0.442 -27.173    0.000    0.610    0.469
##      PhysAct   -12.005    0.442 -27.173    0.000   -7.319   -0.511
##      Falls     1.073    0.070  15.337    0.000    0.654    0.236
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      PersMob ~
##      PPsyHealth 0.433    0.196   2.215    0.027    0.270    0.101
##      PPhysHealth -4.034    0.286 -14.107    0.000   -2.459   -0.914
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      PPsyHealth ~~

```

```
##      PPhysHealth      0.250      0.011      22.634      0.000      0.656      0.656
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .Dep1      0.380      0.009      41.810      0.000      0.380      0.494
##      .Dep2      0.248      0.007      33.798      0.000      0.248      0.382
##      .Dep3      0.251      0.007      35.721      0.000      0.251      0.405
##      .SocAct      5.858      0.107      54.534      0.000      5.858      0.962
##      .ChronCond      1.318      0.028      46.833      0.000      1.318      0.780
##      .PhysAct     151.194      3.462      43.667      0.000     151.194      0.738
##      .Falls       7.241      0.134      54.019      0.000      7.241      0.944
##      .PersMob      1.990      0.239      8.318      0.000      1.990      0.275
##      PPsyHealth      0.390      0.014      28.811      0.000      1.000      1.000
##      PPhysHealth      0.372      0.024      15.334      0.000      1.000      1.000
```

- a) Fit the SEM model in Figure 3.11 (see Beaujean book) to the data. Use the marker variable identification approach only. Evaluate and describe model fit. Make sure to add `fit.measures = TRUE` and `standardized = TRUE` to your call to `summary()` to obtain fit indices and standardized parameter estimates. Include both parameter estimates (e.g., significance and standardized values of loadings) as well as model fit indices (e.g., $\chi^2(df)$, RMSEA, CFI, SRMR) in your evaluation of model fit.

The χ^2 value is significant, but the sample size is very large. CFI and TLI are both $> .95$, RMSEA and SRMR are both $< .05$, all indicating good model fit.

All loadings are significant. However, social activities and falls have relatively small standardized loadings, and thus do not seem very good indicators of their respective factors.

- b) The standardized regression coefficients suggest that poor physical health is a much stronger predictor of personal mobility than poor psychosocial health is.