

SUPPLEMENTAL MATERIAL (R-CODES) ACCOMPANYING

Gross, S., Meier, L. L., & Semmer, N. K. (2012). Latent growth modeling applied to diary data: The course of vigor across a work week as an illustrative example. In A. Bakker & K. Daniels (Eds.), *A day in the life of a happy worker* (pp. 114-131).

Correspondence concerning the following syntaxes should be addressed to Anja Baethge, Department of Psychology, Work-, Organizational-, and Businesspsychology, Johannes Gutenberg-University Mainz, Germany.

baethge@uni-mainz.de

+49 (0) 6131 39 39 250

1. MEASUREMENT MODEL WITHOUT CONSTRAINTS (CFA1)

R-Code (# are comments to the syntax)

Model:

```
cfa1 <- "
```

```
# Measurement Model for vigor (lv = latent vigor, vig1_M = first item measurement point t)
```

```
lv0 =~ vig1_M0 + vig2_M0 + vig3_M0
```

```
lv1 =~ vig1_M1 + vig2_M1 + vig3_M1
```

```
lv2 =~ vig1_M2 + vig2_M2 + vig3_M2
```

```
lv3 =~ vig1_M3 + vig2_M3 + vig3_M3
```

```
lv4 =~ vig1_M4 + vig2_M4 + vig3_M4
```

```
lv5 =~ vig1_M5 + vig2_M5 + vig3_M5
```

```
lv6 =~ vig1_M6 + vig2_M6 + vig3_M6
```

```
#Covariances of Uniqueness
```

```
vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M1 ~~ vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M2 ~~ vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M3 ~~ vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M4 ~~ vig1_M5 + vig1_M6
```

```
vig1_M5 ~~ vig1_M6
```

```
vig2_M0 ~~ vig2_M1 + vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M1 ~~ vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M2 ~~ vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M3 ~~ vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M4 ~~ vig2_M5 + vig2_M6
```

```
vig2_M5 ~~ vig2_M6
```

```
vig3_M0 ~~ vig3_M1 + vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M1 ~~ vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M2 ~~ vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M3 ~~ vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M4 ~~ vig3_M5 + vig3_M6
```

```
vig3_M5 ~~ vig3_M6"
```

```
# The model is estimated
```

```
cfa1.fit<-sem(model=cfa1, data=dataset)
```

```
# The output is printed
```

```
summary(cfa1.fit, rsquare=TRUE, standardized=TRUE)
```

2. MEASUREMENT MODEL WITH EQUAL FACTOR LOADINGS (CFA2)

R-Code

```
cfa2 <- "
```

```
# Measurement Model for vigor (lv = latent vigor, vig1_M = first item measurement  
point t, iv2 and iv3 are used to define, which factor loading should be equal across  
time)
```

```
lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0
```

```
lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1
```

```
lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2
```

```
lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3
```

```
lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4
```

```
lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5
```

```
lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6
```

```
# Covariances of Uniqueness
```

```
vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M1 ~~ vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M2 ~~ vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M3 ~~ vig1_M4 + vig1_M5 + vig1_M6
```

```
vig1_M4 ~~ vig1_M5 + vig1_M6
```

```
vig1_M5 ~~ vig1_M6
```

```
vig2_M0 ~~ vig2_M1 + vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M1 ~~ vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M2 ~~ vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M3 ~~ vig2_M4 + vig2_M5 + vig2_M6
```

```
vig2_M4 ~~ vig2_M5 + vig2_M6
```

```
vig2_M5 ~~ vig2_M6
```

```
vig3_M0 ~~ vig3_M1 + vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M1 ~~ vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M2 ~~ vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M3 ~~ vig3_M4 + vig3_M5 + vig3_M6
```

```
vig3_M4 ~~ vig3_M5 + vig3_M6
```

```
vig3_M5 ~~ vig3_M6"
```

```
cfa2.fit<-sem(model=cfa2, data=dataset)
```

```
summary(cfa2.fit, rsquare=TRUE, standardized=TRUE)
```

3. STATE MODEL OR MEASUREMENT MODEL WITH EQUAL FACTOR LOADINGS AND EQUAL INTERCEPTS (CFA3)

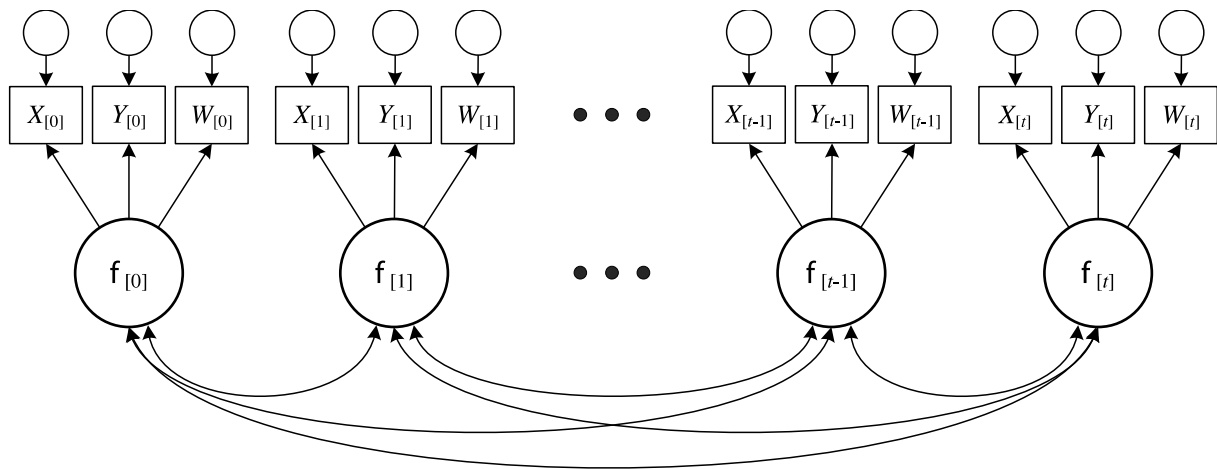


Figure 3.1. Simplified State-Model.

Latent variables are represented by circles and manifest variables by squares. Not depicted in this figure are the intercepts of the manifest and latent variables, the covariances of uniqueness and the variances of the latent variables.

R-Code

```
cfa3 <- "
```

```
# Measurement Model
```

```
# First two code parts (definition of factors and covariances of uniqueness) are the  
same as in CFA2
```

```
lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0
```

```
lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1
```

```
lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2
```

```
lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3
```

```
lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4
```

```
lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5
```

```
lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6
```

```
# Covariances of Uniqueness
```

```
vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

$\text{vig1_M1} \sim \text{vig1_M2} + \text{vig1_M3} + \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M2} \sim \text{vig1_M3} + \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M3} \sim \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M4} \sim \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M5} \sim \text{vig1_M6}$

$\text{vig2_M0} \sim \text{vig2_M1} + \text{vig2_M2} + \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M1} \sim \text{vig2_M2} + \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M2} \sim \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M3} \sim \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M4} \sim \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M5} \sim \text{vig2_M6}$

$\text{vig3_M0} \sim \text{vig3_M1} + \text{vig3_M2} + \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M1} \sim \text{vig3_M2} + \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M2} \sim \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M3} \sim \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M4} \sim \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M5} \sim \text{vig3_M6}$

#Intercepts

Estimating intercepts at the latent level

$\text{lv0} \sim 1$

$\text{lv1} \sim 1$

$\text{lv2} \sim 1$

$\text{lv3} \sim 1$

$\text{lv4} \sim 1$

$\text{lv5} \sim 1$

$\text{lv6} \sim 1$

#Setting intercepts of the first indicator to zero

vig1_M0~0*1

vig1_M1~0*1

vig1_M2~0*1

vig1_M3~0*1

vig1_M4~ 0*1

vig1_M5~ 0*1

vig1_M6~ 0*1

#Setting intercepts of the second indicator to be equal across time

vig2_M0~ ii 2*1

vig2_M1~ ii 2*1

vig2_M2~ ii 2*1

vig2_M3~ ii 2*1

vig2_M4~ ii 2*1

vig2_M5~ ii 2*1

vig2_M6~ ii 2*1

#Setting intercepts of the third indicator to be equal across time

vig3_M0~ ii 3*1

vig3_M1~ ii 3*1

vig3_M2~ ii 3*1

vig3_M3~ ii 3*1

vig3_M4~ ii 3*1

vig3_M5~ ii 3*1

vig3_M6~ ii 3*1"

cfa3.fit<-sem(model=cfa3, data=dataset)

summary(cfa3.fit, rsquare=TRUE, standardized=TRUE)

#Comparing model cfa1, cfa2 and cfa3

```
inspect(cfa1.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper",  
"rmsea.pvalue", "cfi", "tli")]
```

```
inspect(cfa2.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper",  
"rmsea.pvalue", "cfi", "tli")]
```

```
inspect(cfa3.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper",  
"rmsea.pvalue", "cfi", "tli")]
```

```
anova(cfa1.fit, cfa2.fit)
```

```
anova(cfa2.fit, cfa3.fit)
```

4. STATE – TRAIT MODEL

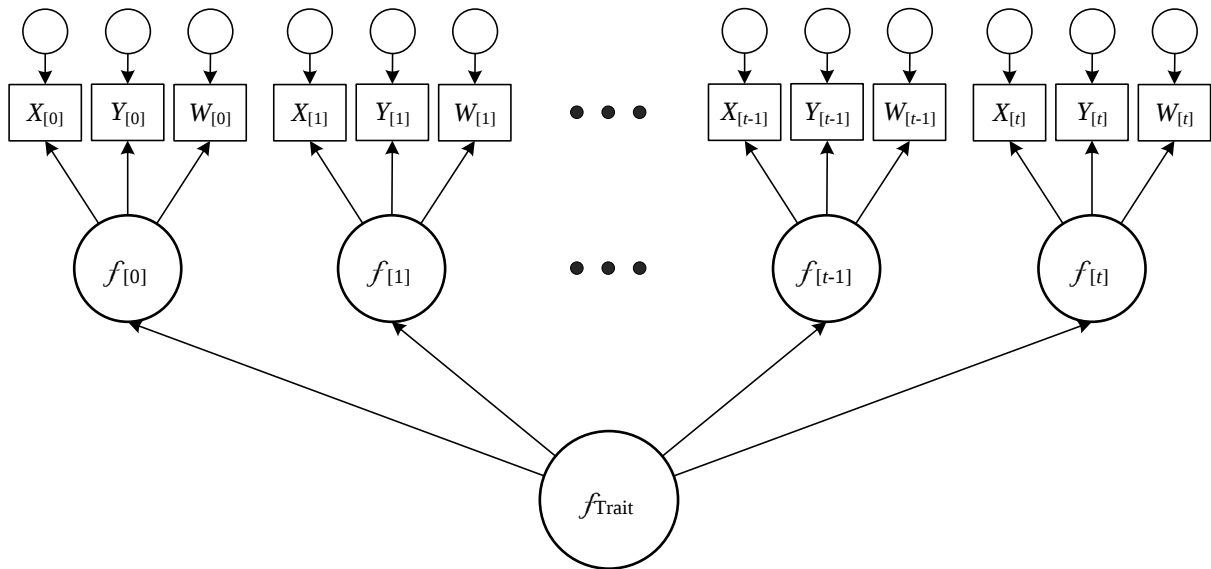


Figure 4.1. Simplified State-Trait Model.

Latent variables are represented by circles and manifest variables by squares. Not depicted in this figure are the intercepts of the manifest and latent variables, the covariances of uniqueness and the variances of the latent variables.

R-Code

```
statetrait <- "
```

```
  # Measurement Model
```

```
  # First three code parts (definition of factors, covariances of uniqueness, intercepts)
  are the same as in the state-model
```

```
  lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0
```

```
  lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1
```

```
  lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2
```

```
  lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3
```

```
  lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4
```

```
  lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5
```

```
  lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6
```

```
  # Covariances of Uniqueness
```

```
  vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6
```

$\text{vig1_M1} \sim \text{vig1_M2} + \text{vig1_M3} + \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M2} \sim \text{vig1_M3} + \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M3} \sim \text{vig1_M4} + \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M4} \sim \text{vig1_M5} + \text{vig1_M6}$

$\text{vig1_M5} \sim \text{vig1_M6}$

$\text{vig2_M0} \sim \text{vig2_M1} + \text{vig2_M2} + \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M1} \sim \text{vig2_M2} + \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M2} \sim \text{vig2_M3} + \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M3} \sim \text{vig2_M4} + \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M4} \sim \text{vig2_M5} + \text{vig2_M6}$

$\text{vig2_M5} \sim \text{vig2_M6}$

$\text{vig3_M0} \sim \text{vig3_M1} + \text{vig3_M2} + \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M1} \sim \text{vig3_M2} + \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M2} \sim \text{vig3_M3} + \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M3} \sim \text{vig3_M4} + \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M4} \sim \text{vig3_M5} + \text{vig3_M6}$

$\text{vig3_M5} \sim \text{vig3_M6}$

#Intercepts

Estimating intercepts at the latent level

$\text{lv0} \sim 1$

$\text{lv1} \sim 1$

$\text{lv2} \sim 1$

$\text{lv3} \sim 1$

$\text{lv4} \sim 1$

$\text{lv5} \sim 1$

$\text{lv6} \sim 1$

#Setting intercepts of the first indicator to zero

vig1_M0~0*1

vig1_M1~0*1

vig1_M2~0*1

vig1_M3~0*1

vig1_M4~ 0*1

vig1_M5~ 0*1

vig1_M6~ 0*1

#Setting intercepts of the second indicator to be equal across time

vig2_M0~ ii 2*1

vig2_M1~ ii 2*1

vig2_M2~ ii 2*1

vig2_M3~ ii 2*1

vig2_M4~ ii 2*1

vig2_M5~ ii 2*1

vig2_M6~ ii 2*1

#Setting intercepts of the third indicator to be equal across time

vig3_M0~ ii 3*1

vig3_M1~ ii 3*1

vig3_M2~ ii 3*1

vig3_M3~ ii 3*1

vig3_M4~ ii 3*1

vig3_M5~ ii 3*1

vig3_M6~ ii 3*1

Structural Part

Trait (ItVig = latent Trait Vigor)

ItVig =~lv0 + lv1 + lv2 + lv3 + lv4 + lv5 + lv6

”

```
statetrait.fit<-sem(model=statetrait.fit, data=dataset)
```

```
summary(statetrait.fit, fit.measure=TRUE, rsquare=TRUE, standardized=TRUE)
```

```
inspect(statetrait.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower",  
"rmsea.ci.upper", "rmsea.pvalue", "cfi", "tli")]
```

5. LINEAR LGM

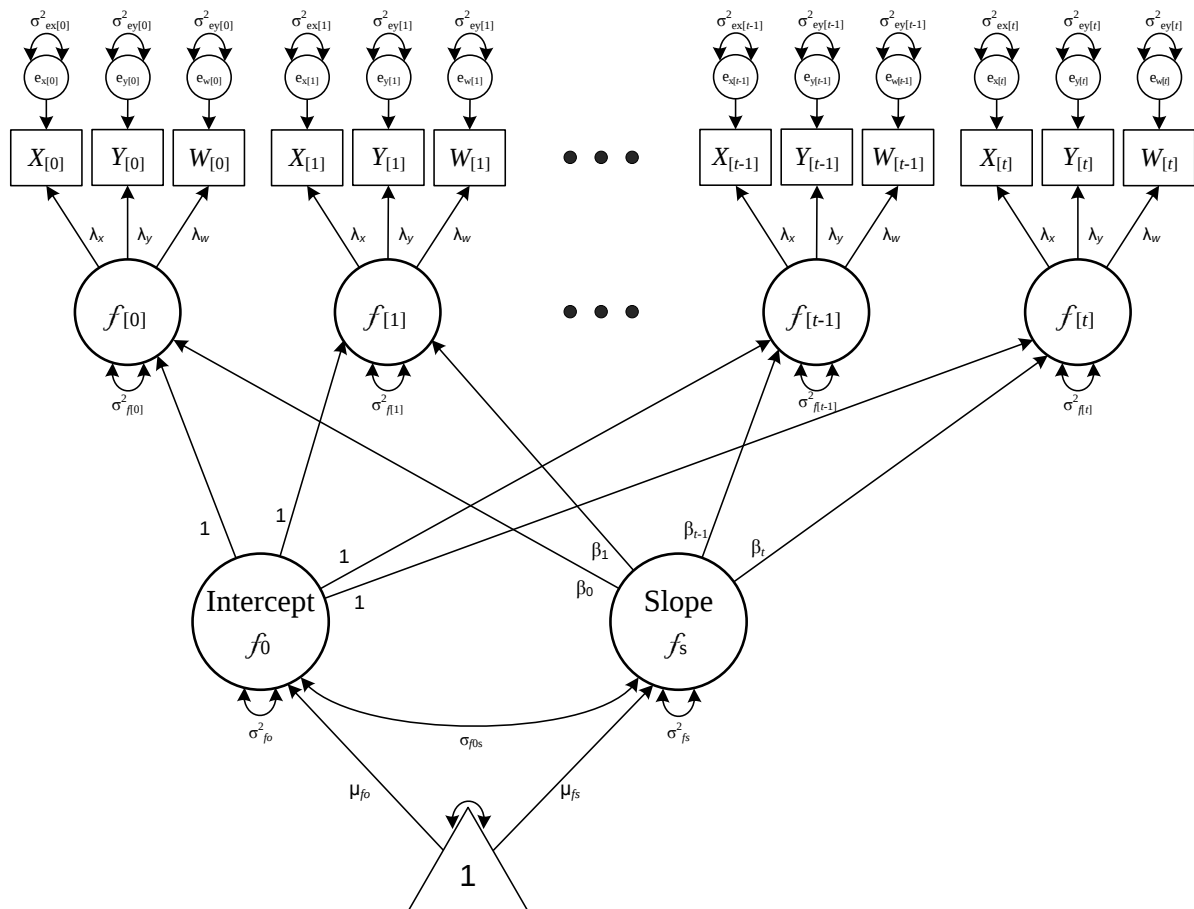


Figure 5.1. Linear LGM.

Latent variables are represented by circles and manifest variables by squares. Depicted in the figure are most of the parameters that are estimated. The triangle represents a constant to estimate the means and intercepts. Parameters that are estimated but not depicted in this figure are the covariances of uniqueness and the intercepts of the manifest variables.

σ^2 = Variance; σ = Covariance; μ = Mean; λ = Factor loading

R-Code

```
linearlgm <-"
```

```
# Measurement Model
```

```
# First two code parts (definition of factors and covariances of uniqueness) are the  
same as in the state-model
```

```
lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0
```

```
lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1
```

lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2

lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3

lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4

lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5

lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6

Covariances of Uniqueness

vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M1 ~~ vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M2 ~~ vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M3 ~~ vig1_M4 + vig1_M5 + vig1_M6

vig1_M4 ~~ vig1_M5 + vig1_M6

vig1_M5 ~~ vig1_M6

vig2_M0 ~~ vig2_M1 + vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M1 ~~ vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M2 ~~ vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M3 ~~ vig2_M4 + vig2_M5 + vig2_M6

vig2_M4 ~~ vig2_M5 + vig2_M6

vig2_M5 ~~ vig2_M6

vig3_M0 ~~ vig3_M1 + vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M1 ~~ vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M2 ~~ vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M3 ~~ vig3_M4 + vig3_M5 + vig3_M6

vig3_M4 ~~ vig3_M5 + vig3_M6

vig3_M5 ~~ vig3_M6

#Intercepts

#Estimating the growth parameter

i~1

$s \sim 1$

The intercepts of the latent vigor factors are set to zero by default

#Setting intercepts of the first indicator to zero

$\text{vig1_M0} \sim 0*1$

$\text{vig1_M1} \sim 0*1$

$\text{vig1_M2} \sim 0*1$

$\text{vig1_M3} \sim 0*1$

$\text{vig1_M4} \sim 0*1$

$\text{vig1_M5} \sim 0*1$

$\text{vig1_M6} \sim 0*1$

#Setting intercepts of the second indicator to be equal across time

$\text{vig2_M0} \sim \text{ii } 2*1$

$\text{vig2_M1} \sim \text{ii } 2*1$

$\text{vig2_M2} \sim \text{ii } 2*1$

$\text{vig2_M3} \sim \text{ii } 2*1$

$\text{vig2_M4} \sim \text{ii } 2*1$

$\text{vig2_M5} \sim \text{ii } 2*1$

$\text{vig2_M6} \sim \text{ii } 2*1$

#Setting intercepts of the third indicator to be equal across time

$\text{vig3_M0} \sim \text{ii } 3*1$

$\text{vig3_M1} \sim \text{ii } 3*1$

$\text{vig3_M2} \sim \text{ii } 3*1$

$\text{vig3_M3} \sim \text{ii } 3*1$

$\text{vig3_M4} \sim \text{ii } 3*1$

$\text{vig3_M5} \sim \text{ii } 3*1$

$\text{vig3_M6} \sim \text{ii } 3*1$

Structural Part

```
# linear LGM (i = intercept factor, s = linear slope factor)
```

```
i =~ 1* lv0 + 1 * lv1 + 1 * lv2 + 1 * lv3 + 1 * lv4 + 1 * lv5 + 1 * lv6
```

```
s =~ 0* lv0 + 1 * lv1 + 2 * lv2 + 3 * lv3 + 4 * lv4 + 5 * lv5 + 6 * lv6
```

```
”
```

```
linearlgm.fit<-sem(model=linearlgm, data=dataset)
```

```
summary(linearlgm.fit, fit.measure=TRUE, rsquare=TRUE, standardized=TRUE)
```

```
inspect(linearlgm.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower",  
"rmsea.ci.upper", "rmsea.pvalue", "cfi", "tli")]
```

6. QUADRATIC LGM WITHOUT TICS

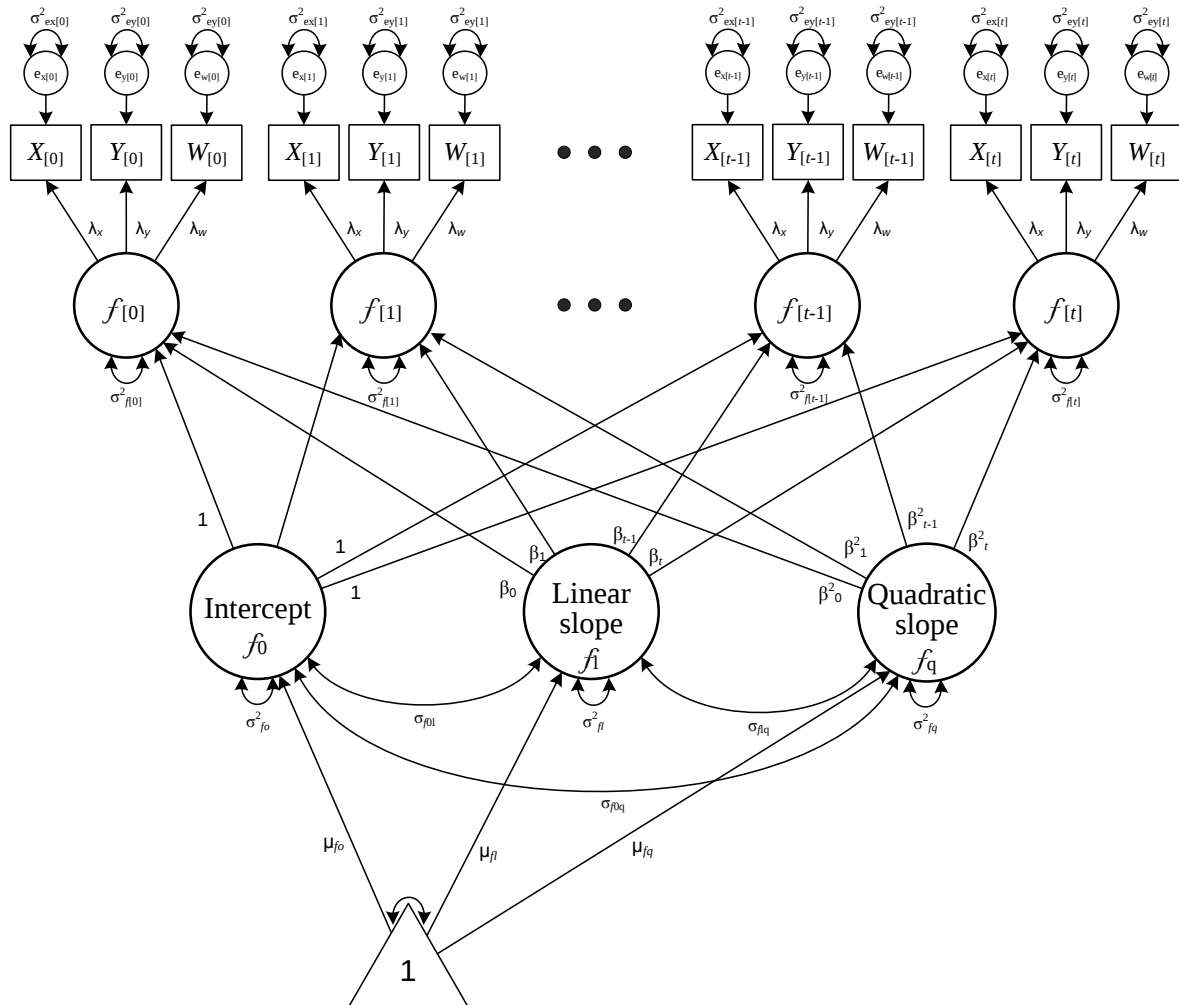


Figure 6.1. Quadratic LGM.

Latent variables are represented by circles and manifest variables by squares. Depicted in the figure are most of the parameters that are estimated. The triangle represents a constant to estimate the means and intercepts. Parameters that are estimated but not depicted in this figure are the covariances of uniqueness and the intercepts of the manifest variables.

σ^2 = Variance; σ = Covariance; μ = Mean; λ = Factor loading

R-Code

```
quadlgm <- "
```

```
# Measurement Model
```

```
# First two code parts (definition of factors and covariances of uniqueness) are the same as in the state-model
```

```
lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0
```

lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1

lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2

lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3

lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4

lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5

lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6

Covariances of Uniqueness

vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M1 ~~ vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M2 ~~ vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M3 ~~ vig1_M4 + vig1_M5 + vig1_M6

vig1_M4 ~~ vig1_M5 + vig1_M6

vig1_M5 ~~ vig1_M6

vig2_M0 ~~ vig2_M1 + vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M1 ~~ vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M2 ~~ vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M3 ~~ vig2_M4 + vig2_M5 + vig2_M6

vig2_M4 ~~ vig2_M5 + vig2_M6

vig2_M5 ~~ vig2_M6

vig3_M0 ~~ vig3_M1 + vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M1 ~~ vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M2 ~~ vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M3 ~~ vig3_M4 + vig3_M5 + vig3_M6

vig3_M4 ~~ vig3_M5 + vig3_M6

vig3_M5 ~~ vig3_M6

#Intercepts

#Estimating the growth parameter

$i \sim 1$

$s \sim 1$

$q \sim 1$

The intercepts of the latent vigor factors are set to zero by default

#Setting intercepts of the first indicator to zero

$\text{vig1_M0} \sim 0*1$

$\text{vig1_M1} \sim 0*1$

$\text{vig1_M2} \sim 0*1$

$\text{vig1_M3} \sim 0*1$

$\text{vig1_M4} \sim 0*1$

$\text{vig1_M5} \sim 0*1$

$\text{vig1_M6} \sim 0*1$

#Setting intercepts of the second indicator to be equal across time

$\text{vig2_M0} \sim \text{ii } 2*1$

$\text{vig2_M1} \sim \text{ii } 2*1$

$\text{vig2_M2} \sim \text{ii } 2*1$

$\text{vig2_M3} \sim \text{ii } 2*1$

$\text{vig2_M4} \sim \text{ii } 2*1$

$\text{vig2_M5} \sim \text{ii } 2*1$

$\text{vig2_M6} \sim \text{ii } 2*1$

#Setting intercepts of the third indicator to be equal across time

$\text{vig3_M0} \sim \text{ii } 3*1$

$\text{vig3_M1} \sim \text{ii } 3*1$

$\text{vig3_M2} \sim \text{ii } 3*1$

$\text{vig3_M3} \sim \text{ii } 3*1$

$\text{vig3_M4} \sim \text{ii } 3*1$

$\text{vig3_M5} \sim \text{ii } 3*1$

```
vig3_M6~ ii 3*1
```

```
# Structural Part
```

```
# quadratic LGM (i = intercept factor, s = linear slope factor, q = quadratic slope factor)
```

```
i =~ 1* lv0 + 1 * lv1 + 1 * lv2 + 1 * lv3 + 1 * lv4 + 1 * lv5 + 1 * lv6
```

```
s =~ 0* lv0 + 1 * lv1 + 2 * lv2 + 3 * lv3 + 4 * lv4 + 5 * lv5 + 6 * lv6
```

```
q =~ 0* lv0 + 1 * lv1 + 4 * lv2 + 9 * lv3 + 16 * lv4 + 25 * lv5 + 36 * lv6
```

```
”
```

```
quadlgm.fit<-sem(model=quadlgm, data=dataset)
```

```
summary(quadlgm.fit, fit.measure=TRUE, rsquare=TRUE, standardized=TRUE)
```

```
inspect(quadlgm.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower",  
"rmsea.ci.upper", "rmsea.pvalue", "cfi", "tli")]
```

7. QUADRATIC LGM INCLUDING TICS

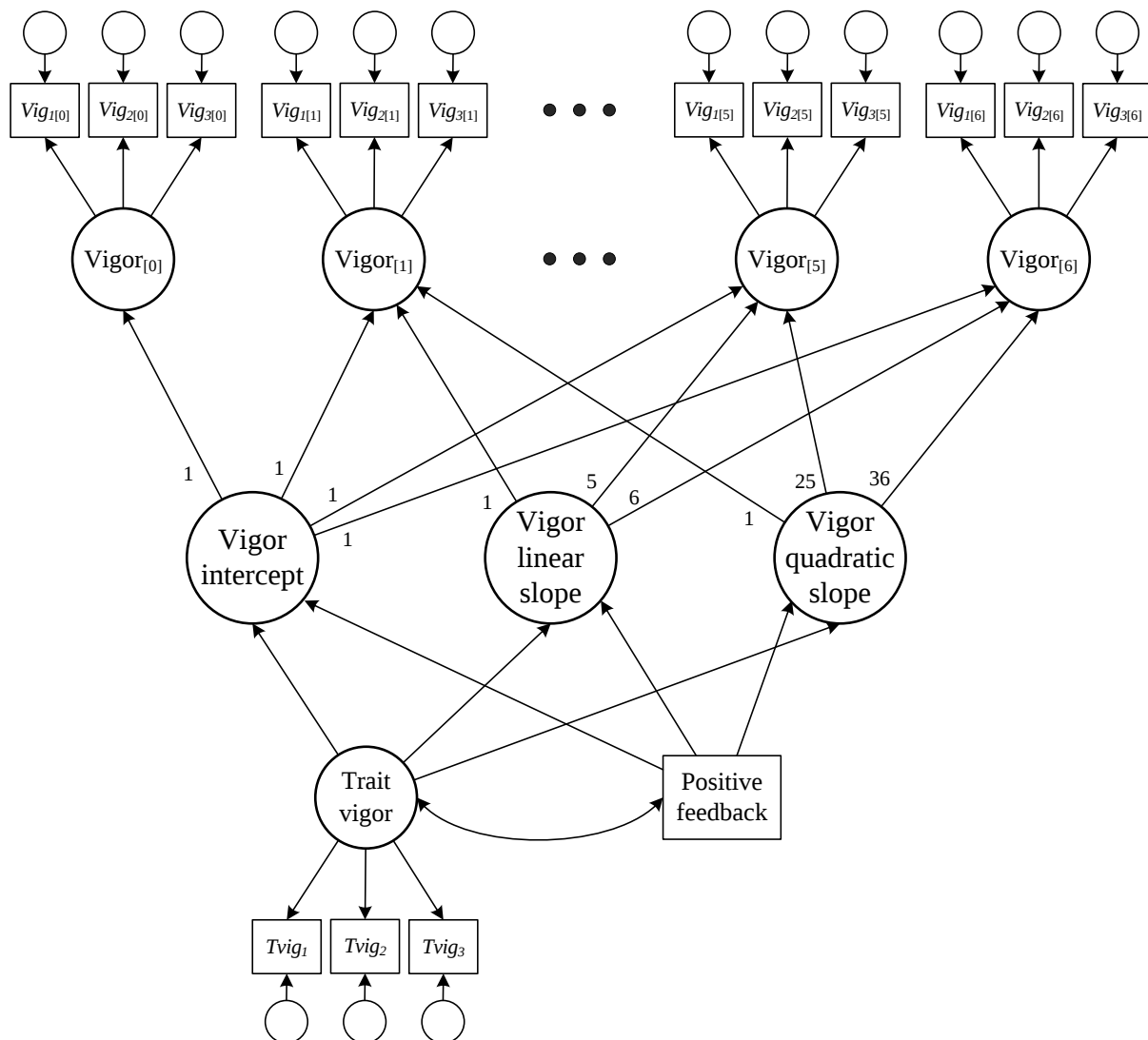


Figure 7.1. Simplified quadratic LGM with covariates.

Latent variables are represented by circles and manifest variables by squares. Not depicted in this figure are the intercepts of the manifest and latent variables, the covariances of uniqueness and the variances of the latent variables.

R-Code

tic <-“

Measurement Model

Code for the measurement model of vigor is the same as in the quadratic LGM without TICS

lv0 =~ vig1_M0 + iv2 * vig2_M0 + iv3 * vig3_M0

lv1 =~ vig1_M1 + iv2 * vig2_M1 + iv3 * vig3_M1

lv2 =~ vig1_M2 + iv2 * vig2_M2 + iv3 * vig3_M2

lv3 =~ vig1_M3 + iv2 * vig2_M3 + iv3 * vig3_M3

lv4 =~ vig1_M4 + iv2 * vig2_M4 + iv3 * vig3_M4

lv5 =~ vig1_M5 + iv2 * vig2_M5 + iv3 * vig3_M5

lv6 =~ vig1_M6 + iv2 * vig2_M6 + iv3 * vig3_M6

Covariances of Uniqueness

vig1_M0 ~~ vig1_M1 + vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M1 ~~ vig1_M2 + vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M2 ~~ vig1_M3 + vig1_M4 + vig1_M5 + vig1_M6

vig1_M3 ~~ vig1_M4 + vig1_M5 + vig1_M6

vig1_M4 ~~ vig1_M5 + vig1_M6

vig1_M5 ~~ vig1_M6

vig2_M0 ~~ vig2_M1 + vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M1 ~~ vig2_M2 + vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M2 ~~ vig2_M3 + vig2_M4 + vig2_M5 + vig2_M6

vig2_M3 ~~ vig2_M4 + vig2_M5 + vig2_M6

vig2_M4 ~~ vig2_M5 + vig2_M6

vig2_M5 ~~ vig2_M6

vig3_M0 ~~ vig3_M1 + vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M1 ~~ vig3_M2 + vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M2 ~~ vig3_M3 + vig3_M4 + vig3_M5 + vig3_M6

vig3_M3 ~~ vig3_M4 + vig3_M5 + vig3_M6

vig3_M4 ~~ vig3_M5 + vig3_M6

vig3_M5 ~~ vig3_M6

#Intercepts

#Estimating the growth parameter

$i \sim 1$

$s \sim 1$

$q \sim 1$

The intercepts of the latent vigor factors are set to zero by default

#Setting intercepts of the first indicator to zero

$\text{vig1_M0} \sim 0*1$

$\text{vig1_M1} \sim 0*1$

$\text{vig1_M2} \sim 0*1$

$\text{vig1_M3} \sim 0*1$

$\text{vig1_M4} \sim 0*1$

$\text{vig1_M5} \sim 0*1$

$\text{vig1_M6} \sim 0*1$

#Setting intercepts of the second indicator to be equal across time

$\text{vig2_M0} \sim \text{ii } 2*1$

$\text{vig2_M1} \sim \text{ii } 2*1$

$\text{vig2_M2} \sim \text{ii } 2*1$

$\text{vig2_M3} \sim \text{ii } 2*1$

$\text{vig2_M4} \sim \text{ii } 2*1$

$\text{vig2_M5} \sim \text{ii } 2*1$

$\text{vig2_M6} \sim \text{ii } 2*1$

#Setting intercepts of the third indicator to be equal across time

$\text{vig3_M0} \sim \text{ii } 3*1$

$\text{vig3_M1} \sim \text{ii } 3*1$

$\text{vig3_M2} \sim \text{ii } 3*1$

$\text{vig3_M3} \sim \text{ii } 3*1$

$\text{vig3_M4} \sim \text{ii } 3*1$

$\text{vig3_M5} \sim \text{ii } 3*1$

```
vig3_M6~ ii 3*1
```

```
#Measurement Model for latent Trait vigor (LTV = latent trait vigor)
```

```
LTV =~ ltv1 + ltv2 + ltv3
```

```
# Structural Part
```

```
# quadratic LGM
```

```
i =~ 1* lv0 + 1 * lv1 + 1 * lv2 + 1 * lv3 + 1 * lv4 + 1 * lv5 + 1 * lv6
```

```
s =~ 0* lv0 + 1 * lv1 + 2 * lv2 + 3 * lv3 + 4 * lv4 + 5 * lv5 + 6 * lv6
```

```
q =~ 0* lv0 + 1 * lv1 + 4 * lv2 + 9 * lv3 + 16 * lv4 + 25 * lv5 + 36 * lv6
```

```
#TICs as predictors of the quadratic LGM (PF = positive feedback (manifest variable:  
was centered around the grand mean)
```

```
i ~ LTV + PF
```

```
s ~ LTV + PF
```

```
q ~ LTV + PF
```

```
#Allowing covariances among growth factors and among TICs
```

```
i~~s + q
```

```
s~~q
```

```
LTV ~~ PF
```

```
”
```

```
tic.fit<-sem(model=tic, data=dataset)
```

```
summary(tic.fit, fit.measure=TRUE, rsquare=TRUE, standardized=TRUE)
```

```
inspect(tic.fit, "fit")[c("chisq", "df", "pvalue", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper",  
"rmsea.pvalue", "cfi", "tli")]
```