

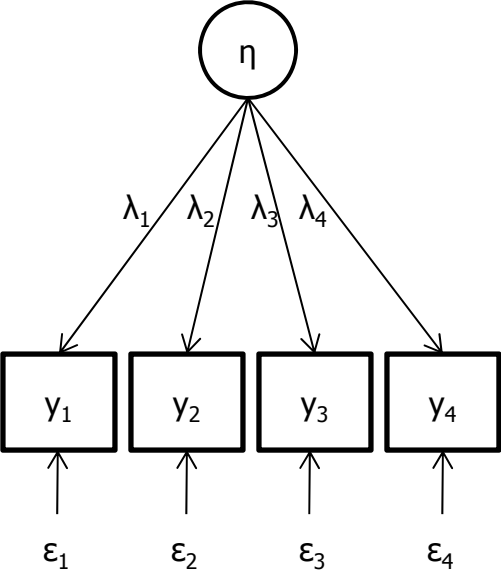
Measurement invariance  
in the linear factor model: practical

# Measurement invariance in the linear factor model: practical

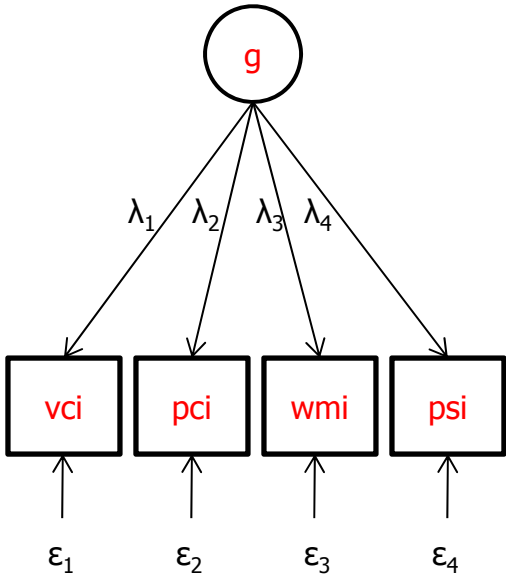


model that relates a continuous latent variable to continuous indicators

Linear factor model



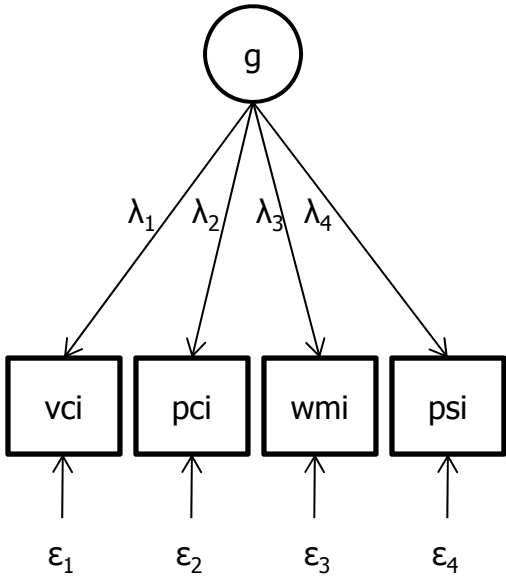
Linear factor model



IQ test (e.g. WAIS):

- vci -- Verbal Comprehension Index
- pci -- Perceptual Organization Index
- wmi -- Working Memory Index
- psi -- Processing Speed Index

Linear factor model

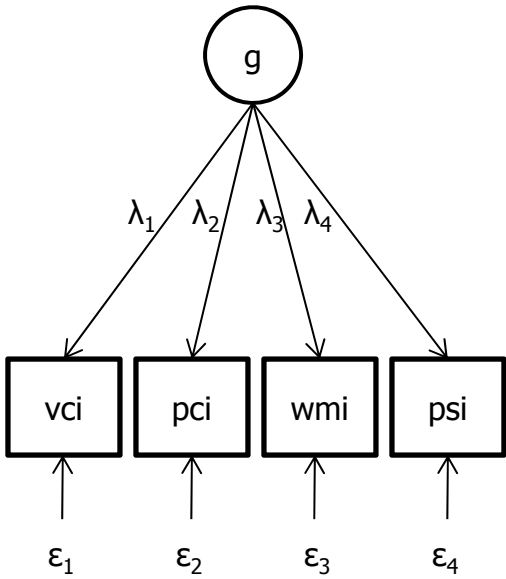


Do males, on average ,score differently than the females?

Men score significantly higher: MANOVA ->  $p < .01$

Does this imply that women have a lower level of g?

Linear factor model



Do males, on average ,score differently than the females?

Men score significantly higher: MANOVA ->  $p < .01$

Does this imply that women have a lower level of g?

Not necessarily.

It depends on whether the test measures the same construct in males as it does in females.

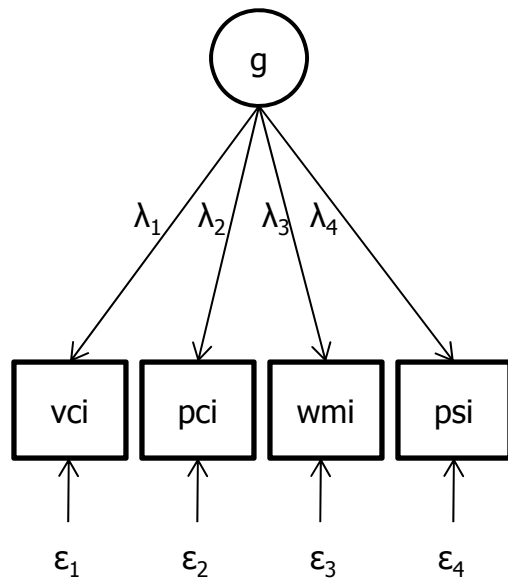
## Linear factor model

Conditional distributions in 2 groups (conditional on a given value of  $\eta$  ( $\eta^*$ )):

$$y_{1i} | \eta^* \sim N(T_1 + \Lambda_1 \eta^*, \Theta_1)$$

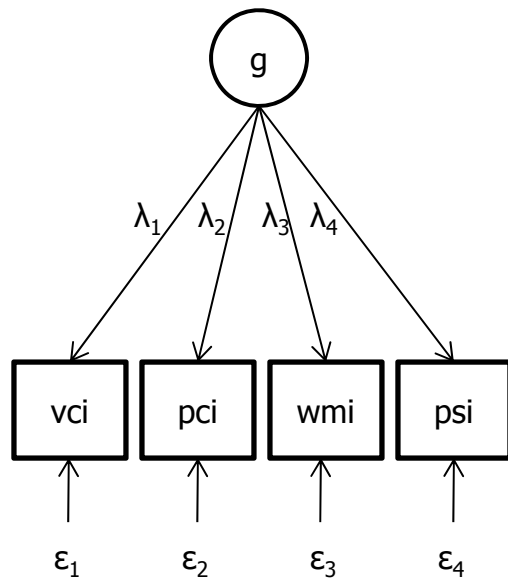
$$y_{2i} | \eta^* \sim N(T_2 + \Lambda_2 \eta^*, \Theta_2)$$

MI requires these distributions to be equal.



## Linear factor model

$$\Sigma = \Lambda \Psi \Lambda^t + \Theta$$



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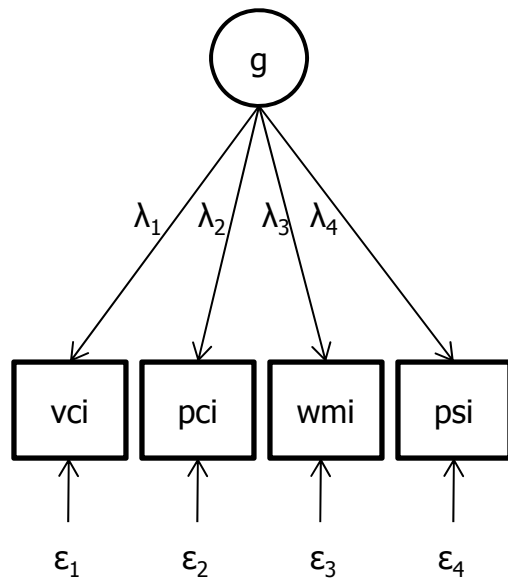
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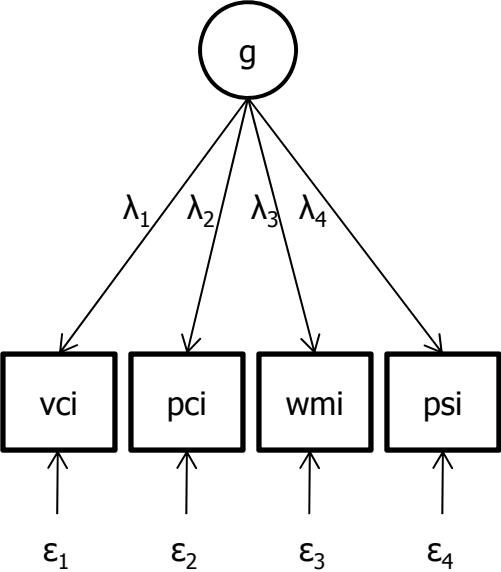
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Linear factor model

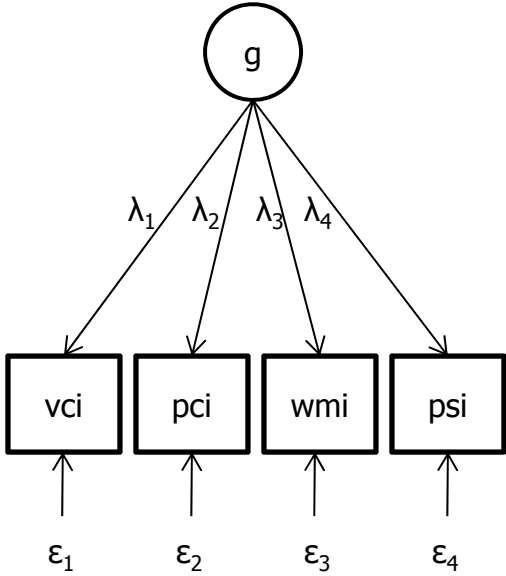
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## Linear factor model

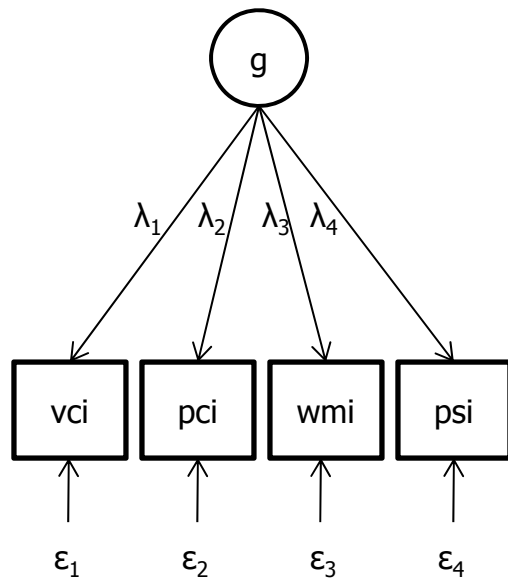
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## Linear factor model

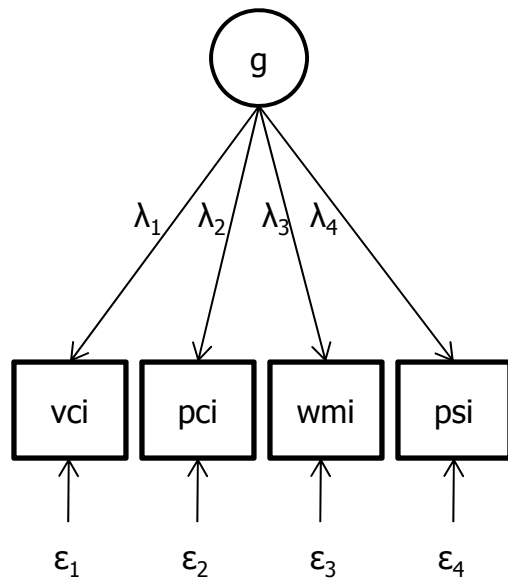
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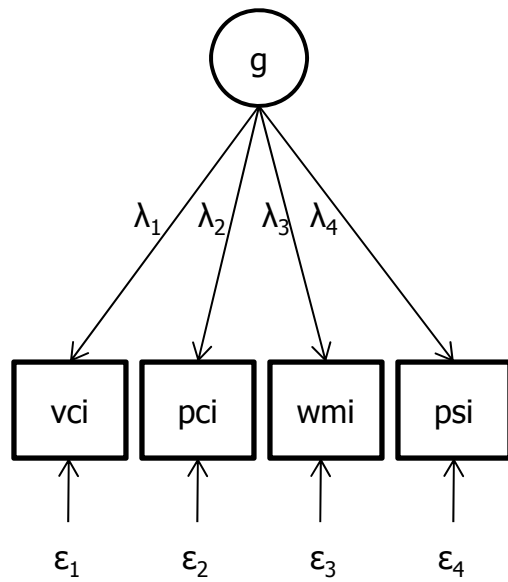
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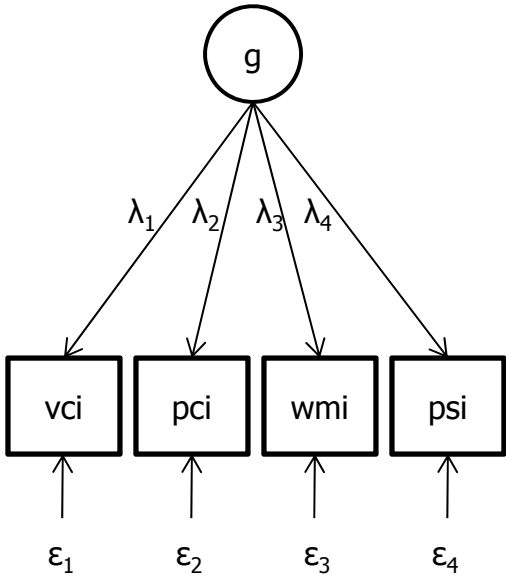
$$E[y|\eta^*] = T + \Lambda \eta^*$$

Conditional distributions in 2 groups (conditional on a given value of  $\eta$  ( $\eta^*$ )):

$$y_{1i} | \eta^* \sim N(T_1 + \Lambda_1 \eta^*, \Theta_1)$$

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Linear factor model

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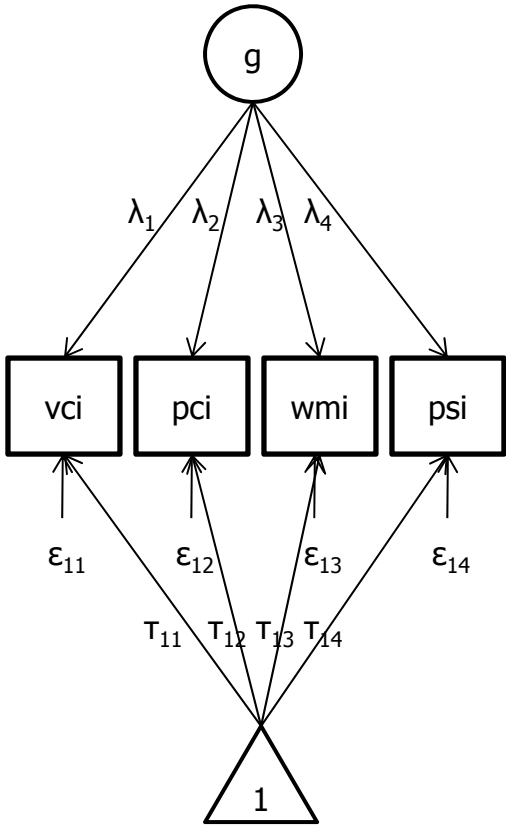
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Linear factor model

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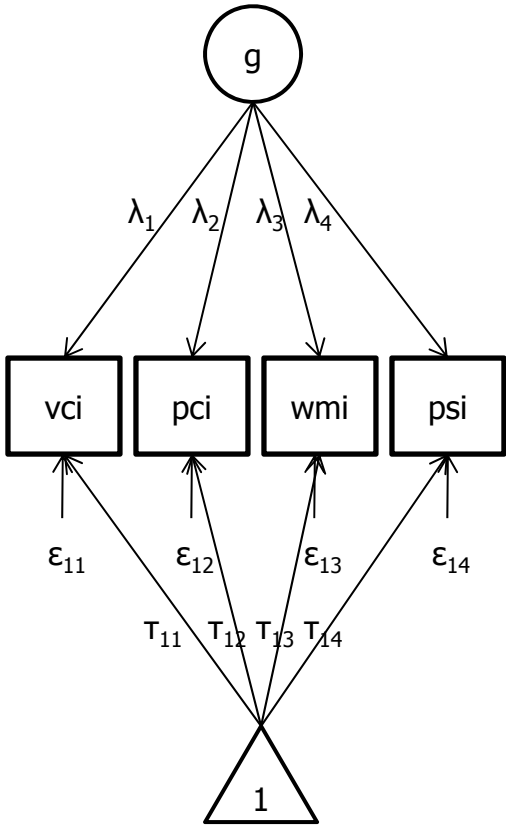
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# Linear factor model

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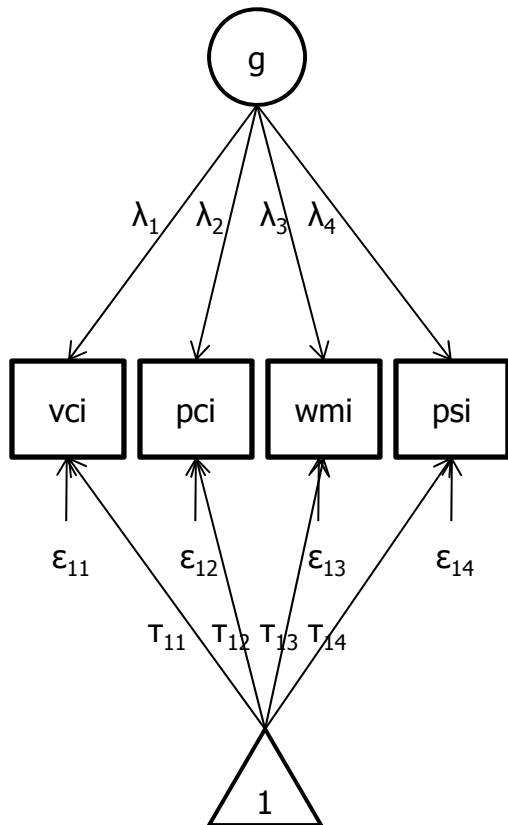
$$y_{1i} | \eta^* \sim N(T_1 + \Lambda_1 \eta^*, \Theta_1)$$



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MI requires these distributions to be equal.



# Linear factor model

$$\Sigma = \Lambda \Theta \Lambda^t + \Theta$$

$$E[y|\eta^*] = T + \Lambda \eta^*$$

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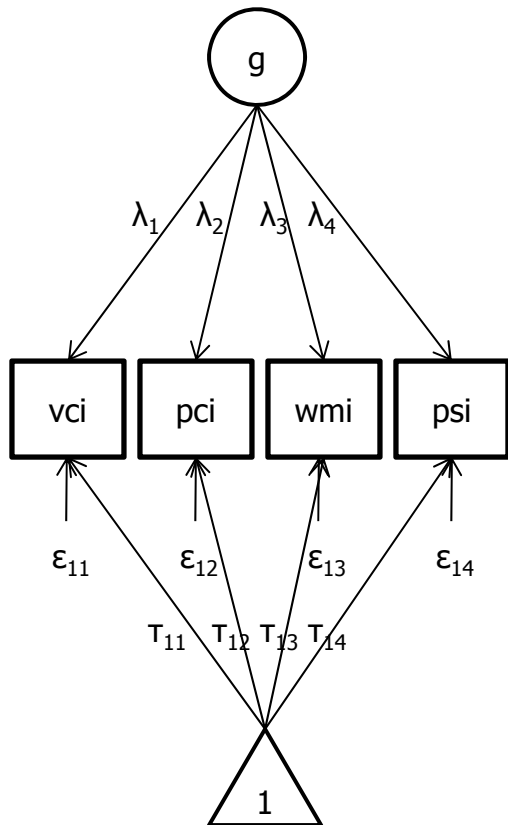
MI requires these distributions to be equal.

This is the case if and only if:

$$T_1 = T_2$$

$$\Lambda_1 = \Lambda_2$$

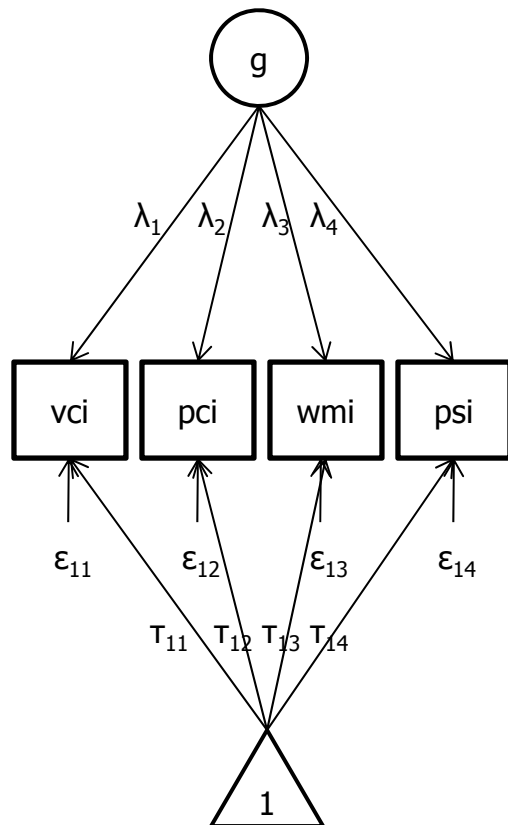
$$\Theta_1 = \Theta_2$$



## Linear factor model

$$\Sigma = \Lambda \Theta \Lambda^t + \Theta$$

$$E[y|\eta^*] = T + \Lambda \eta^*$$



Conditional distributions in 2 groups (conditional on a given value of  $\eta$  ( $\eta^*$ ):

$$y_{1i} | \eta^* \sim N(T_1 + \Lambda_1 \eta^*, \Theta_1)$$

$$y_{2i} | \eta^* \sim N(T_2 + \Lambda_2 \eta^*, \Theta_2)$$

MI requires these distributions to be equal.

This is the case if and only if:

$$T_1 = T_2$$

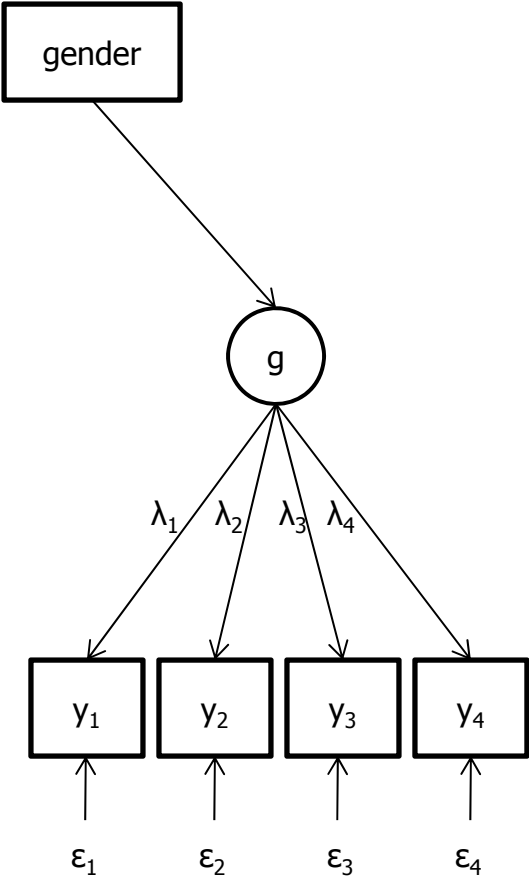
$$\Lambda_1 = \Lambda_2$$

$$\Theta_1 = \Theta_2$$

The test is MI with respect to group if the observed group differences in summary statistics (means and covariance matrix) are attributable to differences in the means and variance of the latent trait or common factor ( $\Psi_k$  and  $\alpha_k$ ).

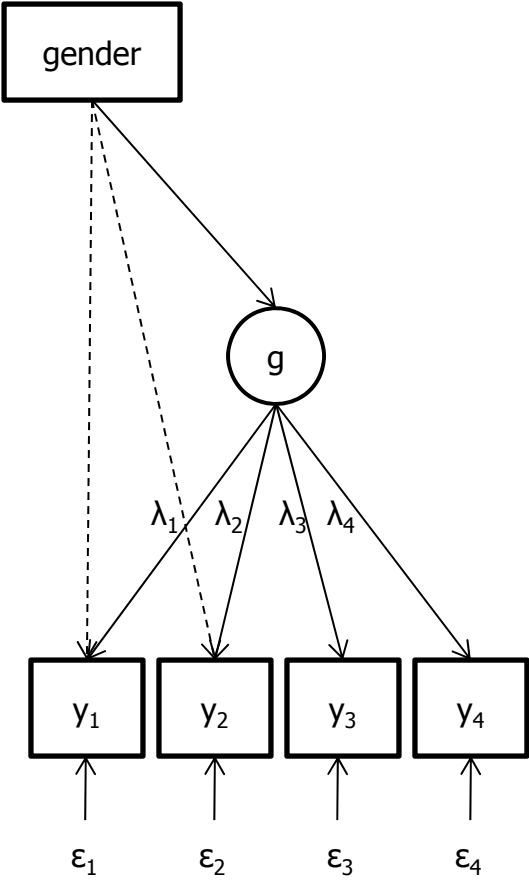
-> if the test measures the same latent variable in the two groups, then that latent variable should be the only source of differences between the groups.

Linear factor model



Measurement invariance

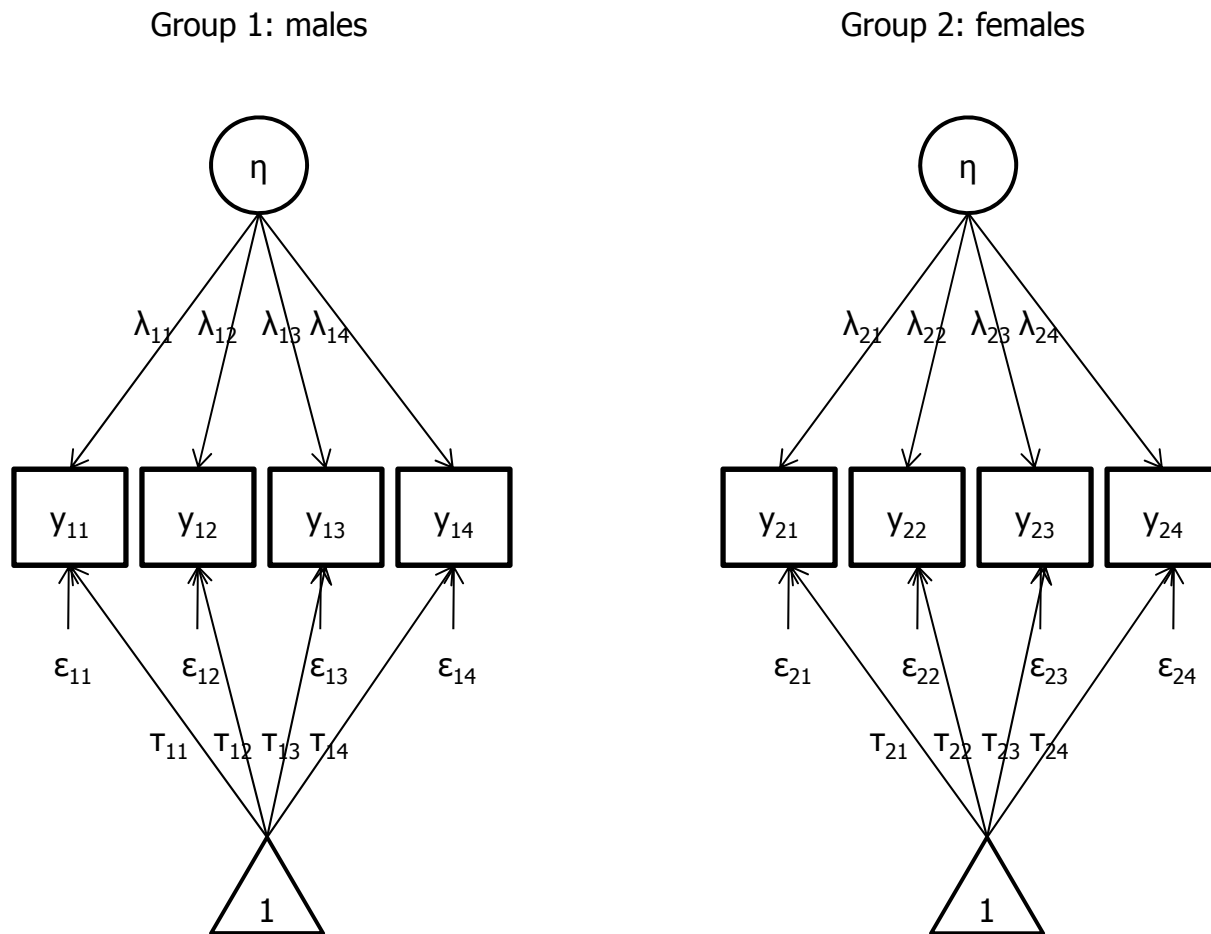
Linear factor model



Lack of measurement invariance

Establishing MI: testing a number of increasingly restrictive models

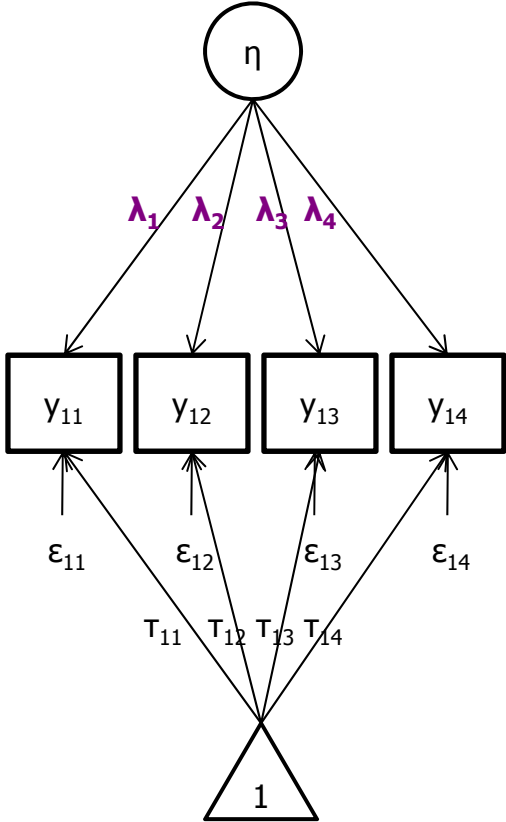
MODEL 1: Configural invariance -> in the 2 groups the same indicators load on the same factors  
(i.e., the pattern or configuration of  $\Lambda$  and  $\Theta$  are the same over groups)



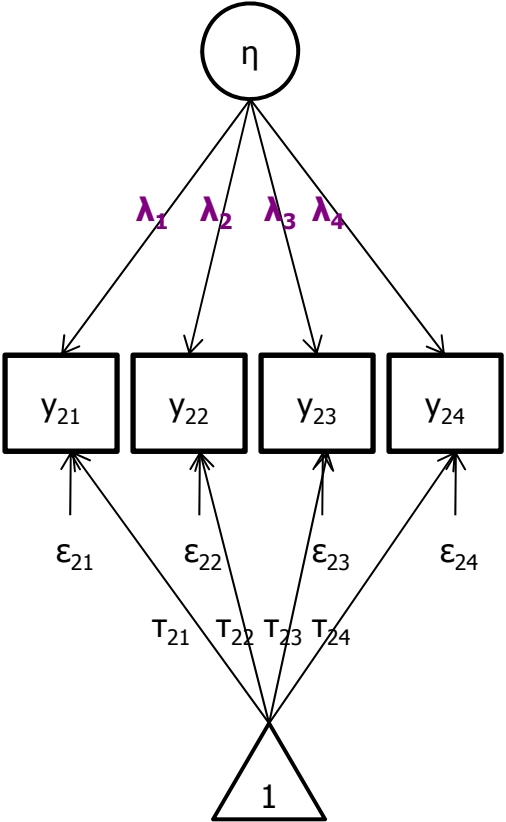
Establishing MI: testing a number of increasingly restrictive models

MODEL 2: Metric invariance -> equal factor loadings over the groups

Group 1: males



Group 2: females



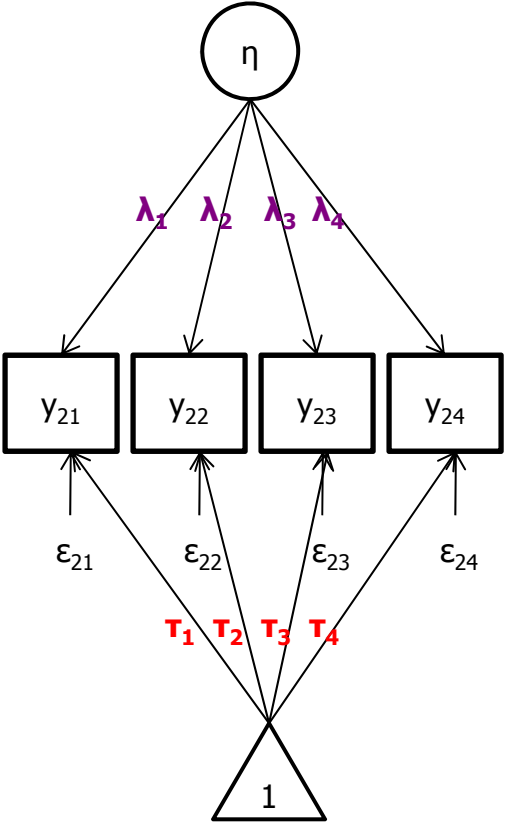
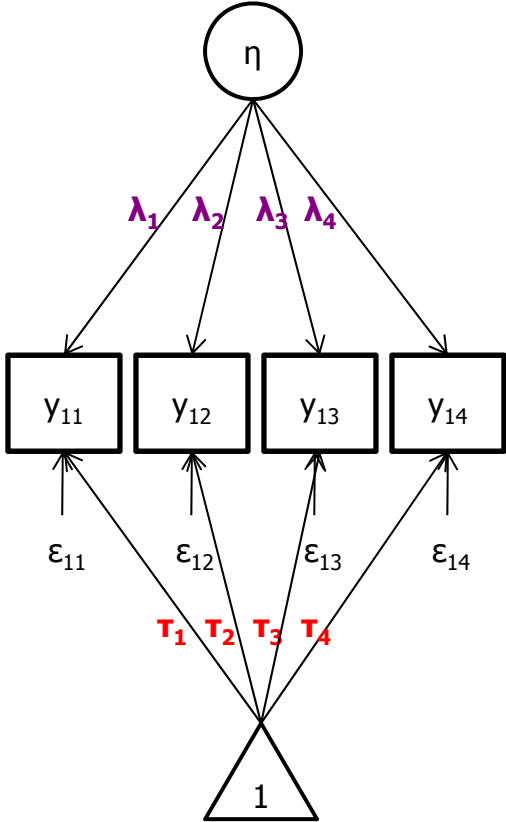


Establishing MI: testing a number of increasingly restrictive models

MODEL 3: Strong factorial invariance -> equal factor loadings and intercepts over the groups

Group 1: males

Group 2: females

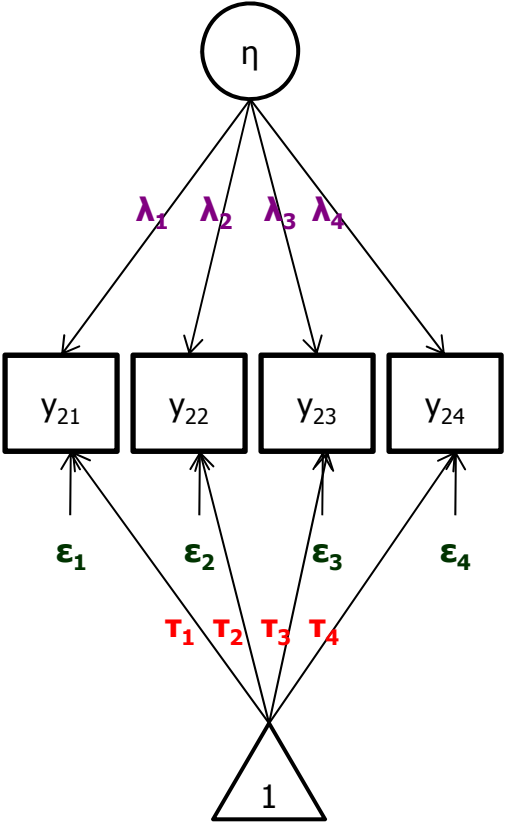
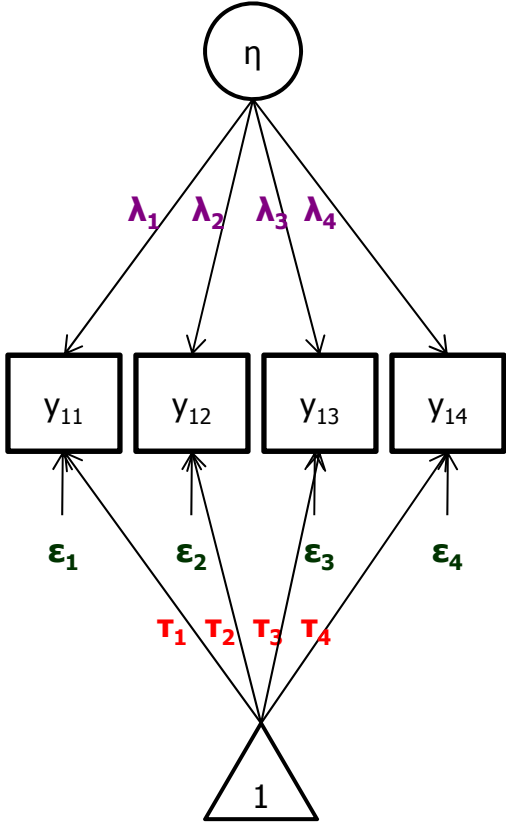


Establishing MI: testing a number of increasingly restrictive models

MODEL 4: Strict factorial invariance -> equal factor loadings, intercepts and residual variances over the groups

Group 1: males

Group 2: females



Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

Data:

	vci	poi	wmi	psi
<b>gender</b>	<b>scale1</b>	<b>scale2</b>	<b>scale3</b>	<b>scale4</b>
2	11	9.33	10.33	13.5
2	10.67	9	10.33	15
2	9.67	7.67	9.33	8.5
2	13	10	8.67	9
2	11	11	13.67	17
2	10	12	9.33	11
...				

N = 160 individuals (80 male, 100 female)

Subscales:

- vci -- Verbal Comprehension Index
- poi -- Perceptual Organization Index
- wmi -- Working Memory Index
- psi -- Processing Speed Index

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# PREPARE DATA #  
#=====#  
  
nv <- 4 # number of phenotype variables to be analyzed  
nf <- 1 # number of common factors in the model  
selVars <- paste("scale",1:nv,sep="") # phenotype variables to be analyzed  
grVars <- c('gender') # grouping variable  
  
data <- read.table(paste(getwd(),"/Measurement_invariance_data.dat",sep=""),header=TRUE)  
mData <- round(data[data$gender==1, selVars],2)  
fData <- round(data[data$gender==2, selVars],2)  
  
# Generate descriptive statistics  
colMeans(mData,na.rm=TRUE)  
colMeans(fData,na.rm=TRUE)  
cov(mData,use="complete")  
cov(fData,use="complete")  
  
# Test for a mean difference between males and females (MANOVA)  
summary(manova(cbind(scale1,scale2,scale3,scale4) ~ gender, data = data), test = "Pillai")
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# PREPARE MODEL #  
#=====#  
  
# Matrices to store factor loadings of the WAIS subscales on g  
loadings1 <- mxMatrix( type="Full", nrow=nv, ncol=nf, free=c(F, rep(T,nv-1)),  
values=1, label=paste("l_1", 1:nv, sep=""), name="load1" )  
loadings2 <- mxMatrix( type="Full", nrow=nv, ncol=nf, free=c(F, rep(T,nv-1)),  
values=1, label=paste("l_2", 1:nv, sep=""), name="load2" )  
  
# Matrices to store the residual variances of the WAIS subscales  
residuals1 <- mxMatrix( type="Diag", nrow=nv, free=T, values=2,  
label=paste("res_1", 1:nv, sep=""), name="res1" )  
residuals2 <- mxMatrix( type="Diag", nrow=nv, free=T, values=2,  
label=paste("res_2", 1:nv, sep=""), name="res2" )  
  
# Matrices to store the mean and variance of g (variance estimated, mean set to 0)  
latVariance1 <- mxMatrix( type="Symm", nrow=nf, ncol=nf, free=T, values=4,  
label=paste("lVar_1", 1:nf, sep=""), name="latVar1" )  
latVariance2 <- mxMatrix( type="Symm", nrow=nf, ncol=nf, free=T, values=4,  
label=paste("lVar_2", 1:nf, sep=""), name="latVar2" )  
  
latMean1 <- mxMatrix( type="Full", nrow=1, ncol=nf, free=F, values=0,  
label=paste("lMean_1",1:nf, sep=""), name="latM1" )  
latMean2 <- mxMatrix( type="Full", nrow=1, ncol=nf, free=F, values=0,  
label=paste("lMean_2",1:nf, sep=""), name="latM2" )
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
# Vectors to store intercepts of the WAIS subscales
intercepts1 <- mxMatrix( type="Full", nrow=nv, ncol=1, free=T, values=8,
label=paste("int_1",1:nv,sep=""), name="int1" )
intercepts2 <- mxMatrix( type="Full", nrow=nv, ncol=1, free=T, values=8,
label=paste("int_2",1:nv,sep=""), name="int2" )

# Algebra for the expected means and covariances of the WAIS scores
means1 <- mxAlgebra( expression=t(int1 + load1%*%latM1), name="m1" )
means2 <- mxAlgebra( expression=t(int2 + load2%*%latM2), name="m2" )
variances1 <- mxAlgebra( expression=load1 %*% latVar1 %*% t(load1) + res1, name="v1" )
variances2 <- mxAlgebra( expression=load2 %*% latVar2 %*% t(load2) + res2, name="v2" )

# Data objects for the two groups
data1 <- mxData( observed=mData, type="raw" )
data2 <- mxData( observed=fData, type="raw" )

# Objective objects for the two groups
obj1 <- mxFIMLObjective( covariance="v1", means="m1", dimnames=selVars )
obj2 <- mxFIMLObjective( covariance="v2", means="m2", dimnames=selVars )

# Combine Groups
modelMales <- mxModel( loadings1, residuals1, latVariance1, latMean1,
intercepts1, means1, variances1, data1, obj1, name="males")
modelFemales <- mxModel( loadings2, residuals2, latVariance2, latMean2,
intercepts2, means2, variances2, data2, obj2, name="females")
minus2ll <- mxAlgebra( expression=males.objective + females.objective, name="m2LL" )
obj <- mxAlgebraObjective( "m2LL" )
CImodel <- mxModel( "CI", modelMales, modelFemales, minus2ll, obj )
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# RUN MODEL: CONFIGURAL INVARIANCE #  
# - equal configuration of factor loadings over the groups #  
#=====#
```

```
CImodelFit <- mxRun(CImodel)  
CImodelSumm <- summary(CImodelFit)  
CImodelSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#
# RUN MODEL: METRIC INVARIANCE                                     #
#       - equal configuration of factor loadings over the groups  #
#       - equal factor loadings over the groups                  #
#=====#

# Matrices to store factor loadings of the WAIS subscales on g
loadings1 <- mxMatrix( type="Full", nrow=nv, ncol=nf, free=c(F, rep(T,nv-1)),
values=1, label=paste("l_", 1:nv, sep=""), name="load1" )
loadings2 <- mxMatrix( type="Full", nrow=nv, ncol=nf, free=c(F, rep(T,nv-1)),
values=1, label=paste("l_", 1:nv, sep=""), name="load2" )

# Combine Groups
modelMales <- mxModel( loadings1, residuals1, latVariance1, latMean1,
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modelFemales <- mxModel( loadings2, residuals2, latVariance2, latMean2,
intercepts2, means2, variances2, data2, obj2, name="females")
minus2ll <- mxAlgebra( expression=males.objective + females.objective, name="m2LL" )
obj <- mxAlgebraObjective( "m2LL" )
MImodel <- mxModel( "MI", modelMales, modelFemales, minus2ll, obj )

MImodelFit <- mxRun(MImodel)
MImodelSumm <- summary(MImodelFit)
MImodelSumm
```



Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# RUN MODEL: STRONG FACTORIAL INVARIANCE - YOUR TASK #  
# - equal configuration of factor loadings over the groups #  
# - equal factor loadings over the groups #  
# - equal intercepts over the groups #  
#=====#
```

# Vectors to store intercepts of the WAIS subscales

```
???  
???  
???  
???
```

# Combine Groups

```
modelMales <- mxModel( loadings1, residuals1, latVariance1, latMean1,  
intercepts1, means1, variances1, data1, obj1, name="males")  
modelFemales <- mxModel( loadings2, residuals2, latVariance2, latMean2,  
intercepts2, means2, variances2, data2, obj2, name="females")  
minus2ll <- mxAlgebra( expression=males.objective + females.objective, name="m2LL" )  
obj <- mxAlgebraObjective( "m2LL" )  
SFImodel <- mxModel( "SFI", modelMales, modelFemales, minus2ll, obj )
```

```
SFImodelFit <- mxRun(SFImodel)  
SFImodelSumm <- summary(SFImodelFit)  
SFImodelSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#
# RUN MODEL: STRONG FACTORIAL INVARIANCE - YOUR TASK #
# - equal configuration of factor loadings over the groups #
# - equal factor loadings over the groups #
# - equal intercepts over the groups #
#=====#

# Vectors to store intercepts of the WAIS subscales
intercepts1 <- mxMatrix( type="Full", nrow=nv, ncol=1, free=T, values=8,
label=paste("int_",1:nv,sep=""), name="int1" )
intercepts2 <- mxMatrix( type="Full", nrow=nv, ncol=1, free=T, values=8,
label=paste("int_",1:nv,sep=""), name="int2" )

# Combine Groups
modelMales <- mxModel( loadings1, residuals1, latVariance1, latMean1,
intercepts1, means1, variances1, data1, obj1, name="males")
modelFemales <- mxModel( loadings2, residuals2, latVariance2, latMean2,
intercepts2, means2, variances2, data2, obj2, name="females")
minus2ll <- mxAlgebra( expression=males.objective + females.objective, name="m2LL" )
obj <- mxAlgebraObjective( "m2LL" )
SFImodel <- mxModel( "SFI", modelMales, modelFemales, minus2ll, obj )

SFImodelFit <- mxRun(SFImodel)
SFImodelSumm <- summary(SFImodelFit)
SFImodelSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# RUN MODEL: STRICT FACTORIAL INVARIANCE - YOUR TASK #  
# - equal configuration of factor loadings over the groups #  
# - equal factor loadings over the groups #  
# - equal intercepts over the groups #  
# - equal residuals over the groups #  
#=====#
```

```
???  
???  
???  
???  
???
```

```
???  
???  
???  
???  
???  
???  
???  
???
```

```
STFImodelFit <- mxRun(STFImodel)  
STFImodelSumm <- summary(STFImodelFit)  
STFImodelSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#
# RUN MODEL: STRICT FACTORIAL INVARIANCE - YOUR TASK #
# - equal configuration of factor loadings over the groups #
# - equal factor loadings over the groups #
# - equal intercepts over the groups #
# - equal residuals over the groups #
#=====#

# Matrices to store the residual variances of the WAIS subscales
residuals1 <- mxMatrix( type="Diag", nrow=nv, free=T, values=2,
label=paste("res_", 1:nv, sep=""), name="res1" )
residuals2 <- mxMatrix( type="Diag", nrow=nv, free=T, values=2,
label=paste("res_", 1:nv, sep=""), name="res2" )

# Combine Groups
modelMales <- mxModel( loadings1, residuals1, latVariance1, latMean1,
intercepts1, means1, variances1, data1, obj1, name="males")
modelFemales <- mxModel( loadings2, residuals2, latVariance2, latMean2,
intercepts2, means2, variances2, data2, obj2, name="females")
minus2ll <- mxAlgebra( expression=males.objective + females.objective, name="m2LL" )
obj <- mxAlgebraObjective( "m2LL" )
STFImodel <- mxModel( "STFI", modelMales, modelFemales, minus2ll, obj )

STFImodelFit <- mxRun(STFImodel)
STFImodelSumm <- summary(STFImodelFit)
STFImodelSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# RUN BASELINE MODEL: 2-GROUP SATURATED MODEL #  
#=====#  
  
# Matrix to store variances/covariances  
startCov=cov(data[,selVars])  
covariances1 <- mxMatrix( type="Symm", nrow=nv, ncol=nv, free=T,  
values=startCov, name="covs1" )  
covariances2 <- mxMatrix( type="Symm", nrow=nv, ncol=nv, free=T,  
values=startCov, name="covs2" )  
  
# Vector to store the means  
means1 <- mxMatrix( type="Full", nrow=1, ncol=4, free=T, values=8,  
labels=paste("mean_2",1:nv,sep=""), name="m1" )  
means2 <- mxMatrix( type="Full", nrow=1, ncol=4, free=T, values=8,  
labels=paste("mean_1",1:nv,sep=""), name="m2" )  
  
# Data object  
Data1 <- mxData( observed=mData[,selVars], type="raw" )  
Data2 <- mxData( observed=fData[,selVars], type="raw" )  
  
# Objective object  
obj1 <- mxFIMLObjective( covariance="covs1", means="m1", dimnames=selVars )  
obj2 <- mxFIMLObjective( covariance="covs2", means="m2", dimnames=selVars )
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
# Combine the groups
satModelMales <- mxModel( covariances1, means1, Data1, obj1, name="satMales")
satModelFemales <- mxModel( covariances2, means2, Data2, obj2, name="satFemales")
minus2ll <- mxAlgebra( expression=satMales.objective + satFemales.objective, name="m2LL" )
obj <- mxAlgebraObjective( "m2LL" )
satModel <- mxModel( "CI", satModelMales, satModelFemales, minus2ll, obj )

# Run the model
satFit <- mxRun(satModel)
satSumm <- summary(satFit)
satSumm
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# COMPARE MODEL FIT #  
#=====#  
  
tableFitStatistics(satFit,CImodelFit) # test of configural invariance  
tableFitStatistics(CImodelFit,MImodelFit) # test of metric invariance  
tableFitStatistics(MImodelFit,SFImodelFit) # test of strong f. invariance  
tableFitStatistics(SFImodelFit,STFImodelFit) # test of strict f. invariance
```

Current practical: Are the 4 subscales of the WAIS-III measurement invariant with respect to gender?

OpenMx code:

```
#=====#  
# COMPARE MODEL FIT #  
#=====#  
  
tableFitStatistics(satFit,CImodelFit) # test of configural invariance  
tableFitStatistics(CImodelFit,MImodelFit) # test of metric invariance  
tableFitStatistics(MImodelFit,SFImodelFit) # test of strong f. invariance  
tableFitStatistics(SFImodelFit,STFImodelFit) # test of strict f. invariance
```

Conclusion...?