Common Pathway Model: Genetic Correlation and Multivariate Models 2

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A Distinction

- ► Earlier: Not so theoretical models
 - Saturated Model
 - ► Fully Correlated Genetic Factors (Cholesky) Model
- ► Now: Theoretical models
 - Common Pathway Model
 - ► Independent Pathway Model





Scientific questions you can ask

- ► In *univariate* analyses: what are the contributions of additive genetic, dominance genetic, shared environmental, and unique environmental factors to the variance?
- ► In *multivariate* analyses: what are these contributions to the **covariance** between two or more traits?





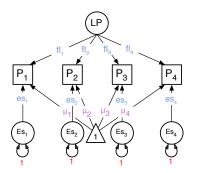
Common Pathway Model

- ► Theoretical model
- ► Start with a phenotypic factor model
- Origin is more psychometric than biometric
- ► Fixes same covariance structure across A, C, and E
- Decompose the factor variance into A, C, E
- Decompose the residual variances into A, C, E



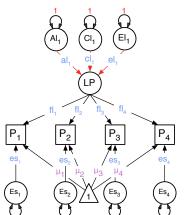


Factor Loadings



$$egin{aligned} P_1 & fl_{11} \ P_2 & fl_{21} \ P_3 & fl_{31} \ P_4 & fl_{41} \ \end{pmatrix}$$

Latent Phenotype ACE

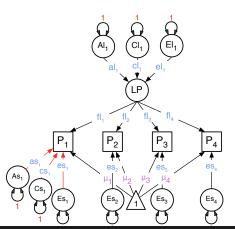


$$\begin{bmatrix} al_{11} \end{bmatrix}$$
 $\begin{bmatrix} cl_{11} \end{bmatrix}$

$$[el_{11}]$$

$$\begin{array}{c|c} P_{1} & fl_{11} \\ P_{2} & fl_{21} \\ P_{3} & fl_{31} \\ P_{4} & fl_{41} \end{array}$$

ACE Specifics



$$\begin{array}{c|ccccc} P_1 & es_{11} & 0 & 0 & 0 \\ P_2 & 0 & es_{22} & 0 & 0 \\ P_3 & 0 & 0 & es_{33} & 0 \\ P_4 & 0 & 0 & 0 & es_{44} \end{array}$$

$$\begin{bmatrix} as_{11} & 0 & 0 & 0 \\ 0 & as_{22} & 0 & 0 \\ 0 & 0 & as_{33} & 0 \\ 0 & 0 & 0 & as_{44} \\ \end{bmatrix}$$

$$\begin{bmatrix} cs_{11} & 0 & 0 & 0 \\ 0 & cs_{22} & 0 & 0 \\ 0 & 0 & cs_{33} & 0 \\ \end{bmatrix}$$

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Common A Factors Specific A Factors

```
 \begin{array}{c} \text{object: pathAl} \\ \text{matrix name: fil} \\ \begin{bmatrix} f_{11} \\ f_{21} \\ f_{31} \\ f_{41} \end{bmatrix} \\ \end{bmatrix} \\ \begin{array}{c} \text{object: pathAl} \\ \text{matrix name: al} \\ [al_{11}] \times \begin{bmatrix} f_{11} \\ f_{21} \\ f_{31} \\ f_{41} \end{bmatrix} \\ \end{array} \\ \begin{bmatrix} a_{11} \\ f_{21} \\ f_{31} \\ f_{41} \end{bmatrix} \\ \\ \begin{bmatrix} a_{11} \\ f_{21} \\ f_{31} \\ f_{41} \end{bmatrix} \\ \\ \begin{bmatrix} a_{11} \\ f_{21} \\ f_{31} \\ f_{41} \end{bmatrix} \\ \\ \begin{bmatrix} a_{11} \\ f_{21} \\ f_{31} \\ f_{41} \\ f_{41} \\ f_{11} \\ f_{21} \\ f_{12} \\ f_{11} \\ f_{12} \\ f_{12} \\ f_{11} \\ f_{12} \\ f_{11} \\ f_{12} \\ f_{11} \\ f_{12} \\ f_{13} \\ f_{13} \\ f_{21} \\ f_{21} \\ f_{22} \\ f_{21} \\ f_{22} \\ f_{22} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{
```

```
pathFl <- mxMatrix( type="Full", nrow=nv, ncol=nl, free=TRUE, values=.2,
    labels=labFull("fl",nv,nl), name="fl")
pathAl <- mxMatrix( type="Lower", nrow=nl, ncol=nl, free=TRUE, values=.6,
    labels=labLower("al",nl), lbound=.00001, name="al")
pathAs <- mxMatrix( type="Diag", nrow=nv, ncol=nv, free=TRUE, values=.5,
    labels=labDiag("as",nv), lbound=.00001, name="as")</pre>
```

Total A Covariance

```
 \begin{bmatrix} f1 \% \& (a1) \% & (a1) \% \\ f1 \% & (a1) \% & (a2) \% \\ f2 \% & (a1) \% & (a2) \% \\ f3 \% & (a1) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f4 \% & (a2) \% & (a2) \% \\ f5 \% & (a2) \% & (a2) \% \\ f6 \% & (a2) \% & (a2) \% \\ f6 \% & (a2) \% & (a2) \% \\ f6 \% & (a2) \% & (a2) \% \\ f6 \% & (a2) \% & (a2) \% \\ f6 \% & (a2) \% & (a2) \% \\ f7 \% & (a2) \% & (a2) \% \\ f8 \% & (a3) \% & (a2) \% \\ f8 \% & (a3) \% & (a2) \% \\ f8 \% & (a3) \% & (a2) \% \\ f8 \% & (a3) \% & (a3) \% \\ f8 \% & (a3) \% & (a3) \% \\ f8 \% & (a3) \% & (a3) \% \\ f8 \% & (a4) \% & (a3) \% \\ f8 \% & (a4) \% & (a3) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% \\ f8 \% & (a4) \% & (a4) \% \\ f8 \% & (a4) \% \\ f
```

```
covA <- mxAlgebra( expression=fl %%% (al %*% t(al)) + as %*% t(as), name="A" )
```

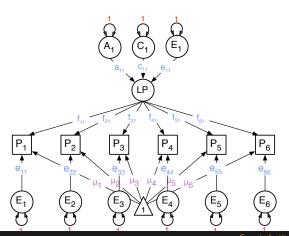
Common Pathway Model

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- Decompose the residual variances into A, C, E





CP Model



Note! For readabilty of the diagrams, we sometimes omit the residual variance decomposition.





Common Pathway

Variance Component	a2	c2	e2	
Common	al	cl	el	fl
Factors	1 x 1	1 x 1	1 x 1	nv x 1
Residual	as	cs	es	
Factors	nv x nv	nv x nv	nv x nv	

OpenM

Constraint on Variance of Latent Phenotype

```
# Fit Common Pathway ACE Model
                                                                                  latent phenotype of x of
         <- 1
# Matrices ac, cc, and ec to store a, c, and e path coefficients for latent phenotype(s)
pathA1
          <- mxMatrix( type="Lower", nrow=nl, ncol=nl, free=TRUE, values=.6, labels=labLower("al",nl), lbound=.00001, name="al" )</pre>
         <- mxMatrix( type="Lower", nrow=nl, ncol=nl, free=TRUE, values=.6, labels=labLower("cl",nl), lbound=.00001, name="cl" )</pre>
pathCl
pathEl
          <- mxMatrix( type="Lower", nrow=nl, ncol=nl, free=TRUE, values=.6, labels=labLower("el",nl), lbound=.00001, name="el" )</pre>
# Matrix and Algebra for constraint on variance of latent phenotype
         <- mxAlgebra( expression=al *** t(al) + cl *** t(cl) + el *** t(el), name="CovarLP" )
varl P
         <- mxAlgebra( expression= diag2vec(CovarLP), name="VarLP" )
uni t
         <- mxMatrix( type="Unit", nrow=nl, ncol=1, name="Unit")
         <- mxConstraint( expression=VarLP == Unit, name="varLP1")
varl P1
                                                                                        a^2 + c^2 + e^2 = 1
# Matrix f for factor loadings on latent phenotype
          <- mxMatrix( type="Full", nrow=nv, ncol=nl, free=TRUE, values=.2, labels=labFull("fl",nv,nl), name="fl" )</pre>
pathF1
# Matrices A. C. and E compute variance components
          <- mxAlaebra( expression=fl %&% (al %*% t(al)) + as %*% t(as) name="A
covA
         <- mxAlgebra( expression=f1 %&% (cl %*% t(cl)) + cs_%*% t(cs), name="C" )
                                                                                                   factor loadings
covC
          <- mxAlgebra( expression=fl %%% (el %*% t(el)) + es %*% t(es), name="E" )
COVE
                                                                         factor loadings x ace on LP
                                                                                     + specifics
```

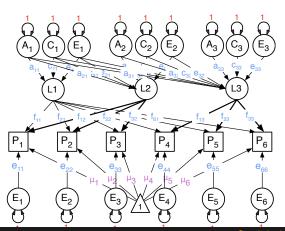
Fitting CP Model

```
# Create Model Objects for Multiple Groups
         <- list(meanG, matI, invSD,
pars
                 pathAl, pathCl, pathEl, covarLP, varLP, unit, pathFl, pathAs, pathCs, pathEs, covA, covC, covE, covP)
         <- mxModel( name="MZ", pars, covMZ, expCovMZ, dataMZ, expMZ, funML
model M7
         <- mxModel( name="DZ", pars, covDZ, expCovDZ, dataDZ, expDZ, funML )
model D7
                                                                                                new objects
mul+i
         <- mxFitFunctionMultigroup( c("MZ", "DZ") )
# Build & Run Model
modelCP <- mxModel( "mulCPc", pars, varLP1, modelMZ, modelDZ, multi )
fi+CP
          <- mxRun(modelCP, intervals=F)
SIIMCP
         <- summarv( fitCP )
mxCompare( fitACE, fitCP )
                                                            constraint object in combined model only
parameterSpecifications(fitCP)
# Generate List of Parameter Estimates and Derived Quantities using formatOutputMatrices
matCPpaths <- c("al", "cl", "el" | "iSD %*% fl", "iSD %*% as", "iSD %*% cs", "iSD %*% es")
labCPpaths <- c("stPathAl", "stPathCl", "stPathEl", "stPathFl", "stPathAs", "stPathCs", "stPathEs")
formatOutputMatrices(fitCP, matCPpaths, labCPpaths, vars, 4)
                                                          already standardized
```





CP 3L Model



Scientific questions you can ask

- ▶ In *univariate* analyses: what are the contributions of additive genetic, dominance genetic, shared environmental, and unique environmental factors to the variance?
- ► In *multivariate* analyses: what are these contributions to the **covariance** between two or more traits?





Questions?



- ▶ Common Pathway
 - ► Can you test for a 1 factor vs a 2 factor vs a 3 factor CP?
 - ► Can you test for every common factor being A and C?
 - ► Can you test for every specific factor being only E?
 - ► Can you fit an ADE model?
- ► Independent Pathway
 - ► Can you test for a 1 factor vs a 2 factor vs a 3 factor IP?
 - Can you test for every A factor having the same loadings? What does that imply?
 - ► Can you test for every specific factor being only E?
 - ► Can you fit an ADE model?





Thank You mhunter43@gatech.edu



