# Path Analysis 

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## Path Analysis

- Allows us to represent linear models for the relationships between variables in diagrammatic form
- Makes it easy to derive expectation for the variances and covariances of variables in terms of the parameters proposed by the model
- Is easily translated into matrix form for use in programs such as Mx


## Example



## Conventions of Path Analysis

- Squares or rectangles denote observed variables.
- Circles or ellipses denote latent (unmeasured) variables.
- (Triangle denote means, used when modeling raw data)
- Upper-case letters are used to denote variables.
- Lower-case letters (or numeric values) are used to denote covariances or path coefficients.


## Conventions of Path Analysis

- Single-headed arrows or paths ( $->$ ) are used to represent causal relationships between variables under a particular model - where the variable at the tail is hypothesized to have a direct influence on the variable at the head.

$$
A \rightarrow B
$$

- Double-headed arrows (<->) are used to represent a covariance between two variables, which may arise through common causes not represented in the model. They may also be used to represent the variance of a variable.
A <-> B


## Conventions of Path Analysis

- Double-headed arrows may not be used for any variable which has one or more single-headed arrows pointing to it - these variables are called endogenous variables. Other variables are exogenous variables.
- Single-headed arrows may be drawn from exogenous to endogenous variables or from endogenous variables to other endogenous variables.


## Conventions of Path Analysis

- Omission of a two-headed arrow between two exogenous variables implies the assumption that the covariance of those variables is zero (e.g., no genotype-environment correlation).
- Omission of a direct path from an exogenous (or endogenous) variable to an endogenous variable implies that there is no direct causal effect of the former on the latter variable.


## Tracing Rules of Path Analysis

- Trace backwards, change direction at a double-headed arrow, then trace forwards.
- This implies that we can never trace through doubleheaded arrows in the same chain.
- The expected covariance between two variables, or the expected variance of a variable, is computed by multiplying together all the coefficients in a chain, and then summing over all possible chains.


## Example



## Exercises

- $\operatorname{Cov} \mathrm{AB}=$
- $\operatorname{Cov} B C=$
- $\operatorname{Cov} A C=$
- $\operatorname{Var} \mathrm{A}=$
- $\operatorname{Var} \mathrm{B}=$
- $\operatorname{Var} \mathrm{C}=$
- Var E


## Covariance between A and B


$\operatorname{Cov} A B=k l+m q n+m p l$

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## Expectations

- $\operatorname{Cov} A B=k l+m q n+m p l$
- Cov BC=no
- Cov AC = mqo
- $\operatorname{Var} A=k^{2}+m^{2}+2 k p m$
- $\operatorname{Var} B=I^{2}+n^{2}$
- $\operatorname{Var} C=o^{2}$
- Var E = 1



## Quantitative Genetic Theory

- Observed behavioral differences stem from two primary sources: genetic and environmental


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## Quantitative Genetic Theory

- There are two sources of genetic influences: $\underline{A d d i t i v e}$ and Dominant



## Quantitative Genetic Theory

- There are two sources of environmental influences: $\mathbf{C o m m o n}$ (shared) and Unique (nonshared)


PHENOTYPE

## In the preceding diagram...

- A, D, C, E are exogenous variables
- A = Additive genetic influences
- D = Non-additive genetic influences (i.e., dominance)
- $C=$ Shared environmental influences
- $\mathrm{E}=$ Nonshared environmental influences
- A, D, C, E have variances of 1
- Phenotype is an endogenous variable
- $P=$ phenotype; the measured variable
- a, d, c, e are parameter estimates


## Univariate Twin Path Model



## Univariate Twin Path Model



## Univariate Twin Path Model



## Assumptions of this Model

- All effects are linear and additive (i.e., no genotype $x$ environment or other multiplicative interactions)
- A, D, C, and E are mutually uncorrelated (i.e., there is no genotype-environment covariance/correlation)
- Path coefficients for $\mathrm{Twin}_{1}=\mathrm{Twin}_{2}$
- There are no reciprocal sibling effects (i.e., there are no direct paths between $P_{1}$ and $P_{2}$


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## Calculating the Variance of $\mathbf{P}_{\mathbf{1}}$



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$$
\operatorname{Var} P_{1}=a^{2}+d^{2}+c^{2}+e^{2}
$$

$$
\begin{aligned}
& \mathrm{P}_{1} \stackrel{d}{\longleftrightarrow}\left(\mathrm{D}_{1}\right) \stackrel{1}{\longrightarrow}\left(\mathrm{D}_{1}\right) \xrightarrow{d} \mathrm{P}_{1}=1 d^{2} \\
& \mathrm{P}_{1} \stackrel{\mathrm{C}}{\longleftrightarrow}\left(\mathrm{C}_{1}\right) \stackrel{1}{\longrightarrow}\left(\mathrm{C}_{1}\right) \stackrel{\mathrm{C}}{\longrightarrow} \mathrm{P}_{1}=1 \mathrm{c}^{2}
\end{aligned}
$$

## Calculating the MZ Covariance



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$$
\begin{aligned}
& \mathrm{P}_{1} \stackrel{\mathrm{a}}{\mathrm{~A}_{1}} \stackrel{1}{\longleftrightarrow} \mathrm{~A}_{2} \xrightarrow{\mathrm{a}} \mathrm{P}_{2}=1 \mathrm{a}^{2}
\end{aligned}
$$

$$
\begin{aligned}
& \mathrm{P}_{1} \stackrel{\mathrm{C}}{\mathrm{C}_{1}} \stackrel{1}{\longleftrightarrow} \mathrm{C}_{2} \xrightarrow{\mathrm{C}} \mathrm{P}_{2}=1 \mathrm{c}^{2} \\
& \mathrm{P}_{1} \stackrel{\mathrm{e}}{\stackrel{\mathrm{E}}{1}} \stackrel{\mathrm{E}}{\longrightarrow} \stackrel{\mathrm{P}_{2}}{ }= \\
& \operatorname{Cov}_{M Z}=a^{2}+d^{2}+c^{2}
\end{aligned}
$$

## Calculating the DZ Covariance



## Calculating the DZ covariance

$\operatorname{Cov}_{D Z}=?$

## Calculating the DZ covariance

$$
\begin{aligned}
& \mathrm{P}_{1} \stackrel{\mathrm{~d}}{\longleftrightarrow} \stackrel{0.25}{\longleftrightarrow} \stackrel{\mathrm{D}}{2} \xrightarrow{\mathrm{~d}} \mathrm{P}_{2}=0.25 \mathrm{~d}^{2} \\
& \mathrm{P}_{1} \stackrel{\mathrm{C}}{\longleftrightarrow} \stackrel{1}{\mathrm{C}_{1}} \stackrel{\mathrm{C}_{2}}{\longleftrightarrow} \stackrel{\mathrm{C}}{\longrightarrow} \mathrm{P}_{2}=1 \mathrm{c}^{2} \\
& \mathrm{P}_{1} \stackrel{\mathrm{e}}{\stackrel{\mathrm{E}}{1}} \stackrel{\mathrm{E}}{\longrightarrow} \stackrel{\mathrm{E}_{2}}{\longrightarrow}= \\
& \operatorname{Cov}_{D Z}=0.5 a^{2}+0.25 d^{2}+c^{2}
\end{aligned}
$$

## Twin Variance/Covariance

$$
\begin{gathered}
{\left[\begin{array}{ll}
\operatorname{Var}_{\text {Twin1 } 1} & \operatorname{Cov}_{12} \\
\operatorname{Cov}_{21} & \operatorname{Var}_{T \text { win2 }}
\end{array}\right]} \\
\mathbf{M Z}=\left[\begin{array}{ll}
a^{2}+d^{2}+c^{2}+e^{2} & a^{2}+d^{2}+c^{2} \\
a^{2}+d^{2}+c^{2} & a^{2}+d^{2}+c^{2}+e^{2}
\end{array}\right] \\
\mathbf{D Z}=\left[\begin{array}{ll}
a^{2}+d^{2}+c^{2}+e^{2} & 0.5 a^{2}+0.25 d^{2}+c^{2} \\
0.5 a^{2}+0.25 d^{2}+c^{2} & a^{2}+d^{2}+c^{2}+e^{2}
\end{array}\right]
\end{gathered}
$$

