

# MAXIMUM LIKELIHOOD ESTIMATION IN **MPLUS**

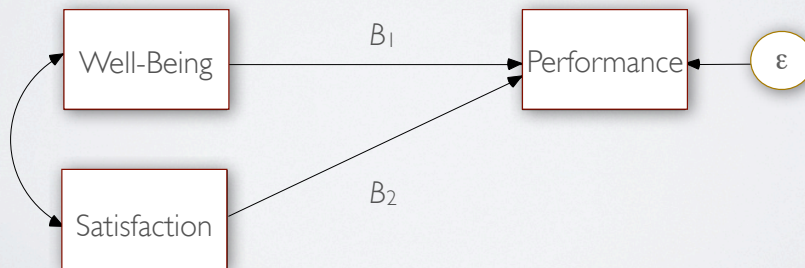
## EMPLOYEE DATA

- Data set containing scores from 480 employees on eight work-related variables
- Variables:
  - Age, gender, job tenure, IQ, psychological well-being, job satisfaction, job performance, and turnover intentions
- 33% of the cases have missing well-being scores, and 33% have missing satisfaction scores

# ANALYSIS EXAMPLE

- Multiple regression model that predicts job performance from psychological well-being and job satisfaction

$$\text{jobperf} = B_0 + B_1(\text{wbeing}) + B_2(\text{jobsat}) + \varepsilon$$



# MPLUS COMMANDS

- TITLE
- DATA
- VARIABLE
- ANALYSIS
- MODEL
- MODEL TEST
- OUTPUT



# A FEW MPLUS RULES

- Capitalization never matters
- Variable names must be 8 characters or less
- Command lines must be less than 80 characters in length, wrap commands to the next line as needed
- ! to comment out a line that you want the program to ignore
- : at the end of a command
- ; at the end of a subcommand

## TITLE COMMAND

- The TITLE command (optional) prints a title on output file

```
TITLE:  
! The title command is optional;  
mplus multiple regression program;
```

# DATA COMMAND

- The DATA command points Mplus to the location of the text data on the local drive
- Free format text files end in .dat or .txt and should include a placeholder for missing values

```
DATA:  
! Location of the data file;  
file = 'c:\Data\employee.dat';
```

# ALTERNATE DATA COMMAND

- Omit the file path when the data file and the Mplus syntax file are located in the same folder

```
DATA:  
! Location of the data file;  
file = employee.dat;
```



# VARIABLE COMMAND

- The VARIABLE command (a) gives the order of the variables in the data file, (b) selects variables for analysis, and (c) gives the missing value code

## VARIABLE:

```
! Information about the contents of the data file;  
names = id age tenure female wbeing jobsat jobperf turnover iq;  
usevariables = wbeing jobsat jobperf;  
missing = all (-99);
```

# ANALYSIS COMMAND

- ANALYSIS specifies the estimator and other estimation details

## ANALYSIS:

```
! Specify the estimator (ML is usually the default);  
estimator = ml;
```

# MODEL COMMAND

- The MODEL command specifies the analysis
- Mplus automatically estimates many parameters (e.g., variances, residual variances, means)
- Missing data models can require additional parameters

## MODEL:

```
! Regression model - "on" means "regressed on";  
jobperf on wbeing jobsat;
```

# INCOMPLETE PREDICTOR VARIABLES

- The missing data log likelihood always allows for incomplete data on  $Y$  (i.e., outcome) variables
- Software packages that implement ML estimation generally exclude cases with incomplete data on manifest predictor variables
- SEM packages provide a mechanism for dealing with incomplete predictors

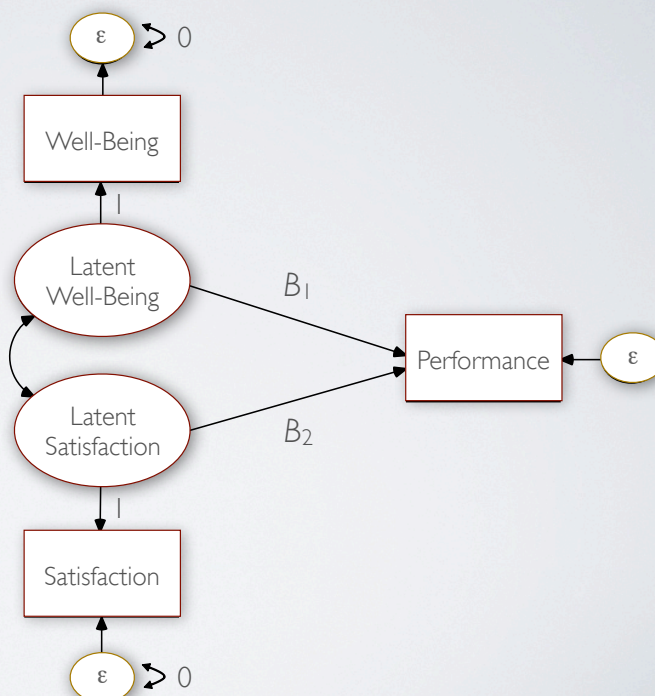


# SEM SPECIFICATION

- A latent variable replaces each incomplete predictor, and the predictor becomes an outcome of its replacement latent variable (i.e., a  $Y$ )
- Step 1: Define each incomplete predictor variable as the sole indicator of a latent variable
- Step 2: Fix the factor loading for each latent / predictor to 1
- Step 3: Fix the residual variance of each indicator to 0 (or something close to zero, e.g., .0001)
- Step 4: Correlate each latent variable with all other predictors (manifest or latent)

# PATH DIAGRAM

- The latent variables are exact duplicates of the manifest variables (i.e., have same mean, variance, correlation)
- The interpretation of  $B_1$  and  $B_2$  does not change!



# MODIFYING THE MODEL COMMAND

- Specify the variance of each incomplete predictor as well as its covariance with all other predictors (manifest or latent)
- This is a shorthand way of specifying the latent variable model

## MODEL:

```
jobperf on wbeing jobsat; ! Regression;  
wbeing jobsat; ! Variances of IVs;  
wbeing with jobsat; ! Covariance between IVs;
```

# WALD TEST

- In ML analyses, Wald chi-square statistics are routinely used to test a set of parameters for significance
- The single-parameter version of the test is as follows

$$\omega = \frac{(\hat{\theta} - \theta_0)^2}{SE^2}$$

- The Wald test is the ML analog of an  $F$  statistic in OLS regression or ANOVA



# THE WALD TEST IN MPLUS

- Wald test parameters must have labels
- The label in parentheses is arbitrary

## MODEL:

```
! (b1) and (b2) labels that are used to specify custom hypotheses;  
jobperf on wbeing (b1);  
jobperf on jobsat (b2);
```

# MODEL TEST COMMAND

- The MODEL TEST generates a Wald test for many custom hypotheses

## MODEL TEST:

```
! Two df omnibus test where both coefficients = 0;  
! b1 and b2 are user-supplied labels from MODEL;  
b1 = 0;  
b2 = 0;
```

# OUTPUT COMMAND

- The OUTPUT command specifies optional information that appears in the Mplus output file

## OUTPUT:

```
! standardized gives beta weights and R-square;  
! sampstat gives ML descriptives;  
! patterns prints missing data patterns;  
standardized sampstat patterns;
```

# MPLUS REGRESSION PROGRAM

## DATA:

```
file = employee.dat;
```

## VARIABLE:

```
names = id age tenure female wbeing jobsat jobperf turnover iq;  
usevariables = wbeing jobsat jobperf;  
missing = all (-99);
```

## ANALYSIS:

```
estimator = ml;
```

## MODEL:

```
jobperf on wbeing (b1);  
jobperf on jobsat (b2);  
wbeing jobsat;  
wbeing with jobsat;
```

## MODEL TEST:

```
b1 = 0;
```

```
b2 = 0;
```

## OUTPUT:

```
standardized sampstat patterns;
```



# MISSING DATA PATTERNS (PATTERNS OPTION)

## SUMMARY OF MISSING DATA PATTERNS

MISSING DATA PATTERNS (x = not missing)

	1	2	3
JOBPERF	x	x	x
WBEING	x	x	
JOBSAT	x		x

## MISSING DATA PATTERN FREQUENCIES

Pattern	Frequency	Pattern	Frequency	Pattern	Frequency
1	160	2	160	3	160

# COVARIANCE COVERAGE

- The covariance coverage matrix gives the proportion of complete cases on each variable or variable pair

## PROPORTION OF DATA PRESENT

	Covariance Coverage		
	JOBPERF	WBEING	JOBSAT
JOBPERF	1.000		
WBEING	0.667	0.667	
JOBSAT	0.667	0.333	0.667

# DESCRIPTIVES (SAMPSTAT OPTION)

## ESTIMATED SAMPLE STATISTICS

### Means

JOBPERF	WBEING	JOBSAT
6.021	6.286	5.959

### Covariances

	JOBPERF	WBEING	JOBSAT
JOBPERF	1.570		
WBEING	0.673	1.387	
JOBSAT	0.259	0.466	1.390

### Correlations

	JOBPERF	WBEING	JOBSAT
JOBPERF	1.000		
WBEING	0.456	1.000	
JOBSAT	0.175	0.335	1.000

# WALD TEST (MODEL TEST COMMAND)

- The Wald statistic (a chi-square with 2 degrees of freedom) is akin to the omnibus  $F$  test in OLS regression

## Wald Test of Parameter Constraints

Value	95.882
Degrees of Freedom	2
P-Value	0.0000

- The significant chi-square,  $\chi^2(2) = 95.882$ , indicates that the set of predictors explain significant variation in the dependent variable



# UNSTANDARDIZED ESTIMATES

## MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
<b>JOBPERF ON</b>				
<b>WBEING</b>	0.476	0.055	8.665	0.000
<b>JOBSAT</b>	0.027	0.060	0.444	0.657
<b>WBEING WITH</b>				
<b>JOBSAT</b>	0.467	0.098	4.780	0.000
<b>Means</b>				
<b>WBEING</b>	6.286	0.063	99.692	0.000

# INTERPRETATIONS

- Interpret and report ML estimates in the same way as a complete-data analysis
- Controlling for job satisfaction, a one-point increase in psychological well-being results in a .476 increase in job performance, on average
- Controlling for psychological well-being, a one-point increase job satisfaction in results in a .027 increase in job performance, on average

# UNSTANDARDIZED ESTIMATES, CONTINUED

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
<b>Intercepts</b>				
JOBPERF	2.869	0.382	7.517	0.000
<b>Variances</b>				
WBEING	1.387	0.108	12.852	0.000
JOBSAT	1.390	0.109	12.711	0.000
<b>Residual Variances</b>				
JOBPERF	1.243	0.087	14.356	0.000

# STANDARDIZED ESTIMATES (STANDARDIZED OPTION)

## STANDARDIZED MODEL RESULTS

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
<b>JOBPERF ON</b>				
WBEING	0.447	0.049	9.181	0.000
JOBSAT	0.025	0.056	0.444	0.657



# STANDARDIZED ESTIMATES, CONTINUED

## R-SQUARE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
JOBPERF	0.208	0.039	5.350	0.000

# INTERPRETATIONS

- The STDYX standardization gives beta weights
  - Controlling for job satisfaction, a one standard deviation increase in psychological well-being results in a .447 standard deviation increase in job performance, on average
  - Controlling for psychological well-being, a one standard deviation increase job satisfaction in results in a .025 standard deviation increase in job performance, on average
- Together, the two predictors explain 20.8% of the variance in job performance ratings