



# Path Models II

Practical examples



# Path Models II

Today's goal:

How to evaluate path models

Outline:

- Examples
- Model fit



# Practical examples

Specifying and testing models



# Practical examples

Knijnenburg et al. (2012): “Inspectability and Control in Social Recommenders”, *RecSys’12*

The TasteWeights system uses the overlap between you and your friends’ Facebook “likes” to give you music recommendations.

- Friends “weights” based on the overlap in likes w/ user
- Friends’ other music likes—the ones that are not among the user’s likes—are tallied by weight
- Top 10 is displayed to the user



# Practical examples

3 control conditions:

- No control (just use likes)
- Item control (weigh likes)
- Friend control (weigh friends)

drag these sliders  
↓

Svetlin's music

Queen

Metallica

U2

Linkin Park

Prodigy

311

Pendulum

Dream Theater

drag these sliders  
↓

Friends

Veselin Kostadinov

Sharang Mugve

Kamal Agarwal

Zlatina Radeva

Annie Todorova

Dave Grant

Ahsan Ashraf

Anastasia Poliakova

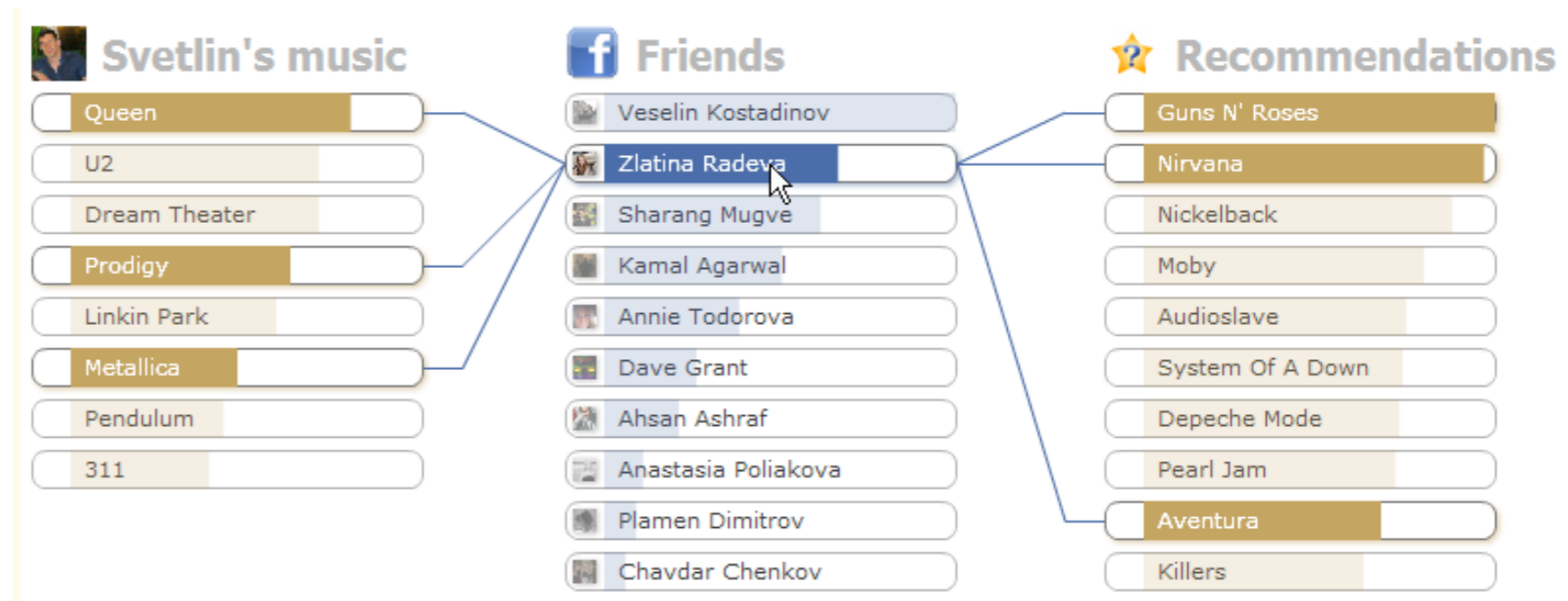
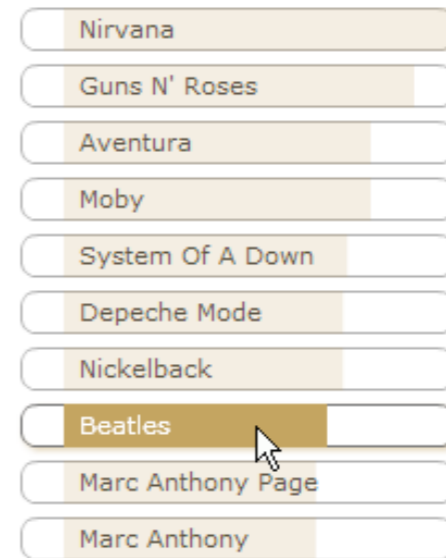


# Practical examples

2 inspectability conditions:

- List of recommendations vs. recommendation graph

## ★ Recommendations





# Practical examples

**twp.dat**, variables:

- **cgraph**: inspectability manipulation (0: list, 1: graph)
- **citem-cfriend**: two dummies for the control manipulation (baseline: no control)
- **cig** ( $\text{citem} * \text{cgraph}$ ) and **cfg** ( $\text{cfriend} * \text{cgraph}$ )
- **satisfaction** with the system (7 items)
- **quality** of the recommendations (6 items)
- **perceived\_control** over the system (5 items)
- **understandability** of the system (5 items)



# Practical examples

**twp.dat**, variables:

- **expertise**: user music expertise
- **trust**: propensity to trust
- **familiarity**: familiarity with recommenders
- average **rating** of, and number of **known** items in, the top 10
- **time** taken to inspect the recommendations





# Linear regression

1. Import the dataset

2. Run the model

```
m <- lm(satisfaction ~ quality + perceived_control +  
understandability, data = twp)
```

3. Get the model summary

```
summary(m)
```



# Linear regression

Residuals:

Min	1Q	Median	3Q	Max
-17.6760	-2.2287	0.3407	2.9143	9.6971

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.35401	0.50722	4.641	5.48e-06	***
quality	0.40151	0.06054	6.632	1.87e-10	***
perceived_control	0.74217	0.08400	8.836	< 2e-16	***
understandability	0.11932	0.08136	1.467	0.144	

----

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.225 on 263 degrees of freedom

Multiple R-squared: 0.5185, Adjusted R-squared: 0.513

F-statistic: 94.42 on 3 and 263 DF, p-value: < 2.2e-16

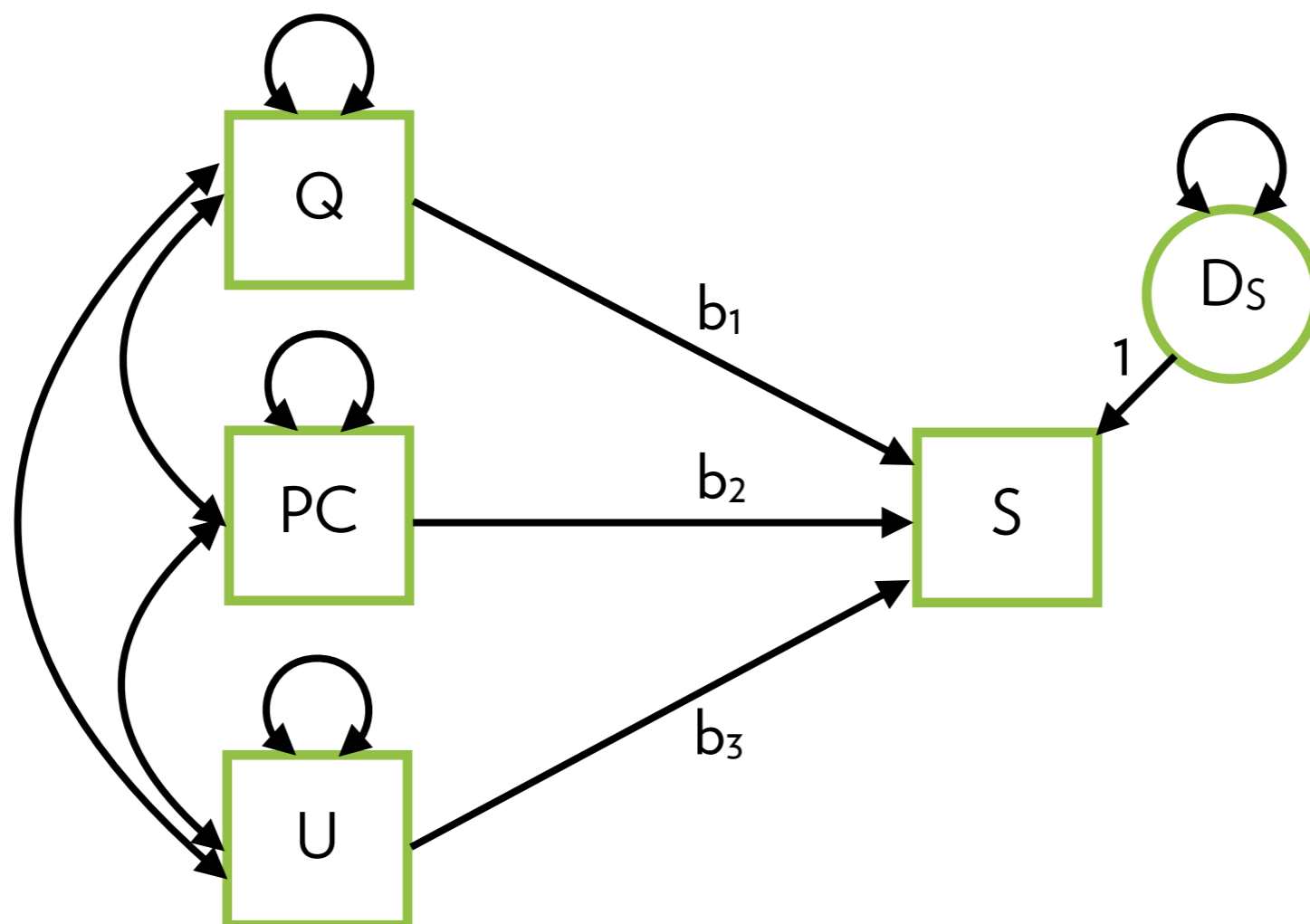


# Same in lavaan

How many degrees of freedom?

How many observations?

How many parameters?





# Same in lavaan

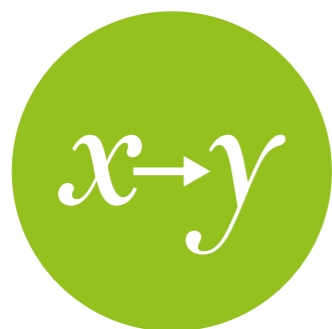
1. Import the dataset
2. Install and activate the lavaan package
3. Write a model specification

```
model <- "satisfaction ~ quality + perceived_control +  
understandability"
```

4. Fit the model  

```
fit <- sem(model, data="twp")
```
5. Get the model summary  

```
summary(fit, rsquare=T)
```



# Same in lavaan

```
Number of observations                267
Estimator                             ML
Minimum Function Test Statistic      0.000
Degrees of freedom                     0
```

## Regressions:

	Estimate	Std.Err	z-value	P(> z )
satisfaction ~				
quality	0.402	0.060	6.683	0.000
perceivd_cntrl	0.742	0.083	8.903	0.000
understandblty	0.119	0.081	1.478	0.139

## Variances:

	Estimate	Std.Err	z-value	P(> z )
.satisfaction	17.582	1.522	11.554	0.000

## R-Square:

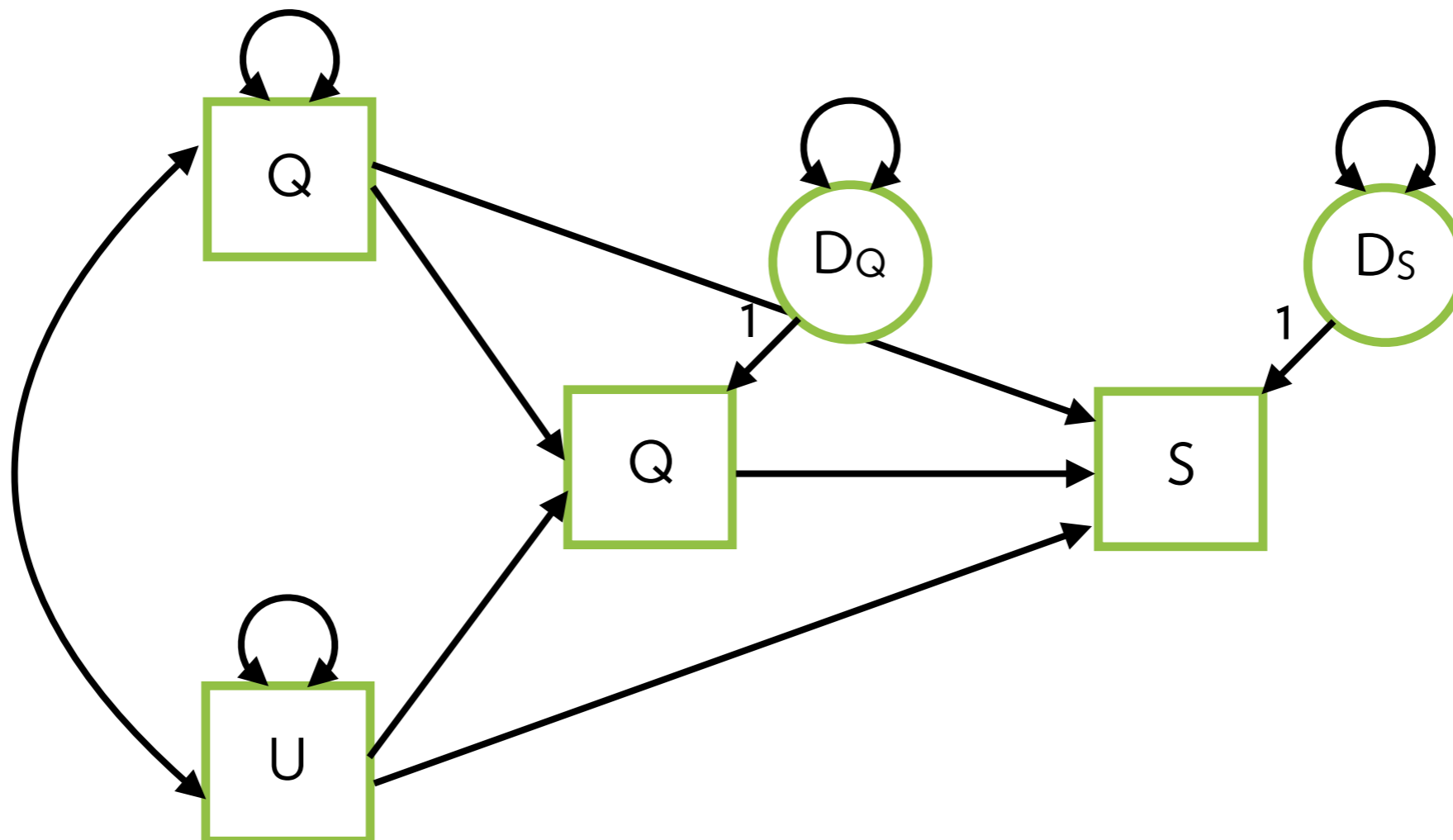
	Estimate
satisfaction	0.519



# Mediation

What if the effect of perceived\_control and understandability are mediated by quality?

How many degrees of freedom?





# Mediation

Write a model specification

```
mediated <- "satisfaction ~ quality + perceived_control +  
understandability  
quality ~ perceived_control + understandability"
```

Fit the model

```
fit <- sem(mediated, data="twp")
```

Get the model summary

```
summary(fit, rsquare=T)
```



# Mediation

## Regressions:

	Estimate	Std.Err	z-value	P(> z )
satisfaction ~				
quality	0.402	0.060	6.683	0.000
perceivd_cntrl	0.742	0.083	8.903	0.000
understandblty	0.119	0.081	1.478	0.139
quality ~				
perceivd_cntrl	0.688	0.074	9.334	0.000
understandblty	0.149	0.082	1.819	0.069

## Variances:

	Estimate	Std.Err	z-value	P(> z )
.satisfaction	17.582	1.522	11.554	0.000
.quality	18.241	1.579	11.554	0.000

## R-Square:

	Estimate
satisfaction	0.519
quality	0.292

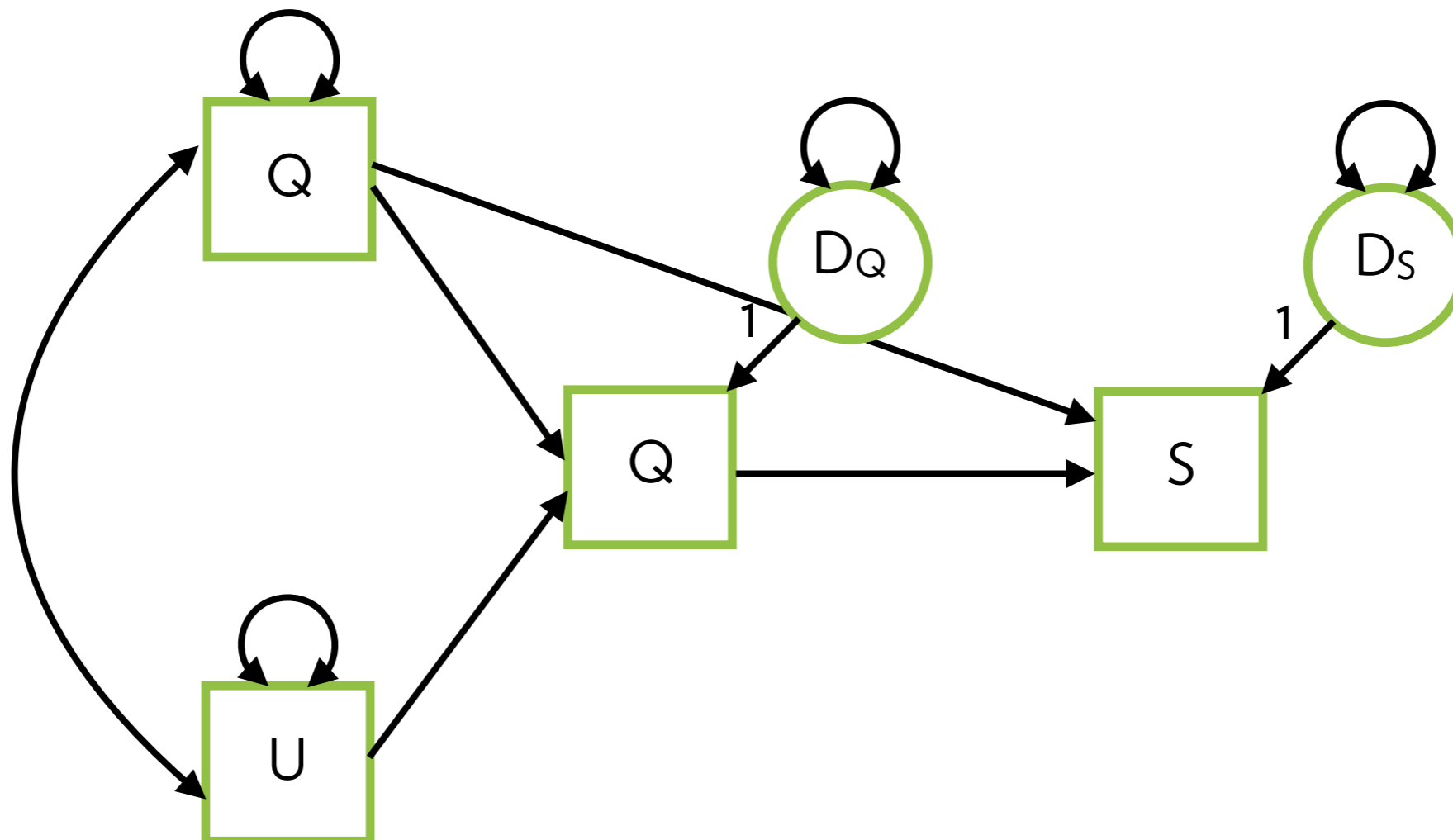




# Trimming

Remove understandability -> satisfaction

How many degrees of freedom?





# Trimming

Write a model specification

```
trimmed <- "satisfaction ~ quality + perceived_control  
quality ~ perceived_control + understandability"
```

Fit the model

```
fit <- sem(trimmed, data="twp")
```

Get the model summary

```
summary(fit, rsquare=T)
```



# Trimming

Number of observations	267
Estimator	ML
Minimum Function Test Statistic	2.175
Degrees of freedom	1
P-value (Chi-square)	0.140

## Regressions:

	Estimate	Std.Err	z-value	P(> z )
satisfaction ~				
quality	0.411	0.060	6.860	0.000
perceivd_cntrl	0.767	0.082	9.353	0.000
quality ~				
perceivd_cntrl	0.688	0.074	9.334	0.000
understandblty	0.149	0.082	1.819	0.069

## R-Square:

	Estimate
satisfaction	0.515
quality	0.292



# t-test

1. Run the model

```
m <- t.test(understandability ~ cgraph, data = twp)
```

2. Get the results

```
m
```

```
Welch Two Sample t-test
```

```
data: understandability by cgraph  
t = -3.7022, df = 255.09, p-value = 0.000262  
alternative hypothesis: true difference in means is not equal to 0  
95 percent confidence interval:  
 -2.2819053 -0.6972146  
sample estimates:  
mean in group 0 mean in group 1  
 1.083969      2.573529
```



# As a regression

1. Run the model

```
m <- lm(understandability ~ cgraph, data = twp)
```

2. Get the model summary

```
summary(m)
```



# As a regression

Residuals:

Min	1Q	Median	3Q	Max
-7.0840	-2.0840	0.4265	2.4265	4.9160

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.0840	0.2863	3.786	0.000189	***
cgraph	1.4896	0.4011	3.713	0.000249	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.277 on 265 degrees of freedom

Multiple R-squared: 0.04946, Adjusted R-squared: 0.04587

F-statistic: 13.79 on 1 and 265 DF, p-value: 0.0002494

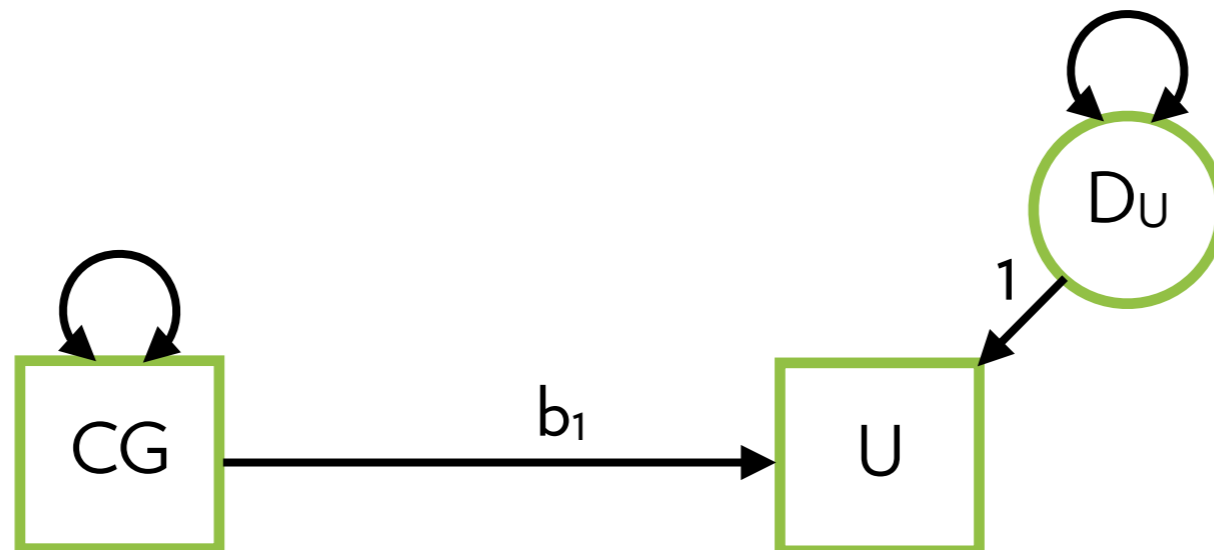


# Same in lavaan

How many degrees of freedom?

How many observations?

How many parameters?





# Same in lavaan

1. Write a model specification

```
tmodel <- "understandability ~ cgraph"
```

2. Fit the model

```
fit <- sem(tmodel, data="twp")
```

3. Get the model summary

```
summary(fit, rsquare=T)
```





# Same in lavaan

```
Number of observations                267
Estimator                            ML
Minimum Function Test Statistic      0.000
Degrees of freedom                    0
Minimum Function Value                0.000000000000000
```

## Parameter Estimates:

```
Information                            Expected
Standard Errors                        Standard
```

## Regressions:

```
                Estimate  Std.Err  z-value  P(>|z|)
understandability ~
  cgraph          1.490    0.400    3.727    0.000
```

## Variances:

```
                Estimate  Std.Err  z-value  P(>|z|)
.understandblty  10.657    0.922   11.554    0.000
```



# ANOVA

1. Import the dataset (use the original tw here!)

2. Run the model

```
m <- aov(understandability ~ control, data = tw)
```

3. Get the model summary

```
summary(m)
```

```
          Df Sum Sq Mean Sq F value Pr(>F)
control    2   93.3   46.65    4.246 0.0153 *
Residuals 264 2900.1   10.99
```

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1



# As a regression

1. Get the lm summary

```
summary.lm(m)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.0435	0.3455	3.020	0.00278	**
controlfriend	1.3610	0.4928	2.762	0.00615	**
controlitem	1.0728	0.4971	2.158	0.03183	*

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.314 on 264 degrees of freedom  
Multiple R-squared: 0.03117, Adjusted R-squared: 0.02383  
F-statistic: 4.246 on 2 and 264 DF, p-value: 0.01531

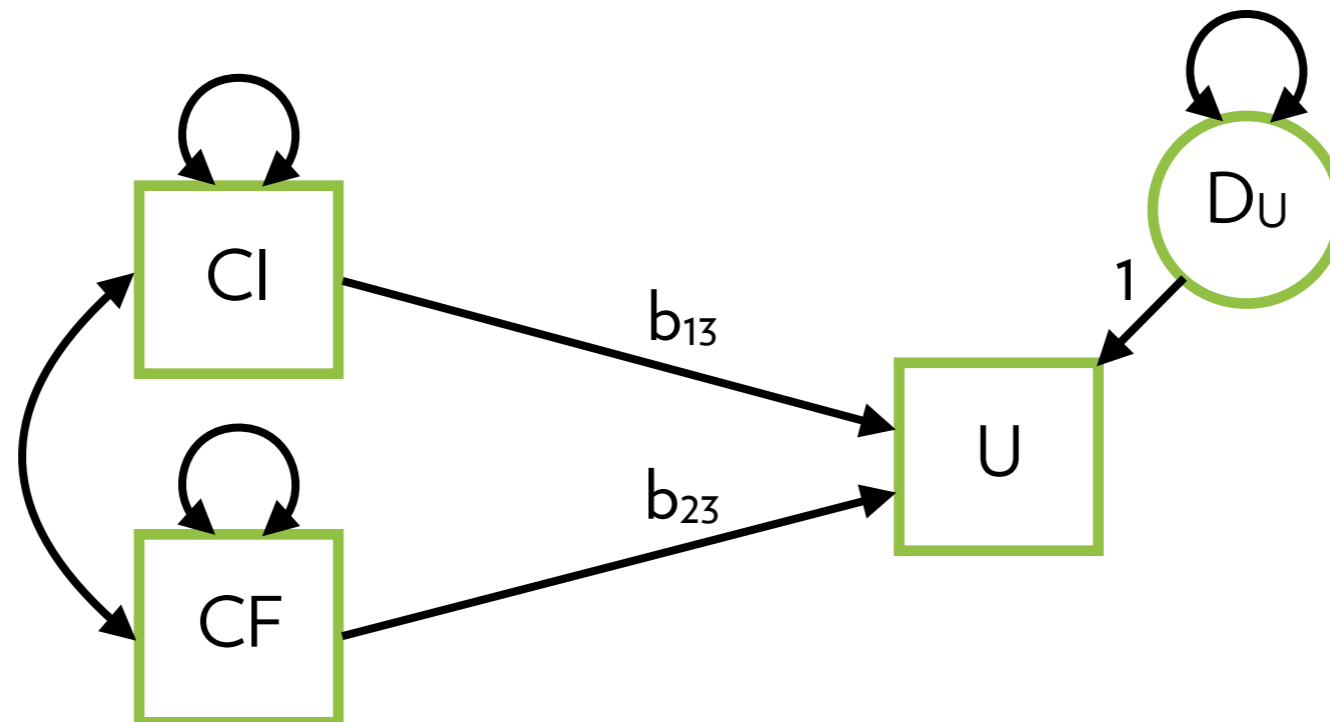


# Same in lavaan

How many degrees of freedom?

How many observations?

How many parameters?





# Same in lavaan

1. Write a model specification

```
amodel <- "understandability ~ ci*citem + cf*cfriend"
```

2. Fit the model

```
fit <- sem(amodel, data="twp")
```

3. Conduct the omnibus test

```
lavTestWald(fit, "ci==0; cf==0")
```

4. Get the model summary

```
summary(fit, rsquare=T)
```



# Same in lavaan

```
$stat  
[1] 8.588976
```

```
$df  
[1] 2
```

```
$p.value  
[1] 0.01364356
```

```
$se  
[1] "standard"
```

Chi-square test with 2 degrees of freedom

$$\chi^2(2) = 8.59, p = .0136$$



# Same in lavaan

Number of observations 267

Estimator ML  
Minimum Function Test Statistic 0.000  
Degrees of freedom 0

## Regressions:

		Estimate	Std.Err	z-value	P(> z )
understandability ~					
citem	(ci)	1.073	0.494	2.170	0.030
cfriend	(cf)	1.361	0.490	2.778	0.005

## Variances:

	Estimate	Std.Err	z-value	P(> z )
.understandblty	10.862	0.940	11.554	0.000

## R-Square:

	Estimate
understandblty	0.031



# Factorial ANOVA

1. Run the model

```
m <- aov(understandability ~ inspectability * control, data =  
tw)
```

2. Get the anova results with type III Sum of Squares (in the 'car' package)

```
Anova(m, type=3)
```





# Factorial ANOVA

Anova Table (Type III tests)

Response: understandability

	Sum Sq	Df	F value	Pr(>F)	
(Intercept)	172.20	1	16.3518	6.93e-05	***
inspectability	73.09	1	6.9404	0.008931	**
control	29.61	2	1.4060	0.246983	
inspectability:control	3.88	2	0.1841	0.831962	
Residuals	2748.50	261			

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1



# As a regression

## 1. Get the lm summary

```
summary.lm(m)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	0.1522	0.4785	0.318	0.75070	
inspectabilitygraphview	1.7826	0.6766	2.634	0.00893	**
controlfriend	1.6739	0.6766	2.474	0.01400	*
controlitem	1.1555	0.7064	1.636	0.10307	
inspectabilitygraphview:controlfriend	-0.5854	0.9652	-0.607	0.54469	
inspectabilitygraphview:controlitem	-0.3031	0.9757	-0.311	0.75633	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.245 on 261 degrees of freedom

Multiple R-squared: 0.08181, Adjusted R-squared: 0.06422

F-statistic: 4.651 on 5 and 261 DF, p-value: 0.0004388



# Same in lavaan

1. Write a model specification

```
fmodel <- "understandability ~ ci*citem + cf*cfriend + cgraph  
+ cig*cig + cfg*cfg"
```

2. Fit the model

```
fit <- sem(fmodel, data="twp")
```

3. Conduct the omnibus tests

```
lavTestWald(fit, "cig==0; cfg==0")
```

```
lavTestWald(fit, "ci==0; cf==0")
```

4. Get the model summary

```
summary(fit, rsquare=T)
```



# Same in lavaan

```
$stat  
[1] 0.3766611
```

```
$df  
[1] 2
```

```
$p.value  
[1] 0.8283409
```

```
$se  
[1] "standard"
```

Interaction effect:

$$\chi^2(2) = 0.38, p = .828$$



# Same in lavaan

```
$stat  
[1] 6.530504
```

```
$df  
[1] 2
```

```
$p.value  
[1] 0.03818731
```

```
$se  
[1] "standard"
```

Main effect:

$$\chi^2(2) = 6.53, p = .0382$$

Meaningless, though... why?



# Same in lavaan

## Regressions:

		Estimate	Std.Err	z-value	P(> z )
understandability ~					
citem	(ci)	1.156	0.698	1.655	0.098
cfriend	(cf)	1.674	0.669	2.502	0.012
cgraph		1.783	0.669	2.665	0.008
cig	(cig)	-0.303	0.965	-0.314	0.753
cfg	(cfg)	-0.585	0.954	-0.613	0.540

## Variances:

	Estimate	Std.Err	z-value	P(> z )
.understandblty	10.294	0.891	11.554	0.000

## R-Square:

	Estimate
understandblty	0.082



# Robust estimator

1. Write a model specification

```
trimmed <- "satisfaction ~ quality + perceived_control  
quality ~ perceived_control + understandability"
```

2. Fit the model **with a robust estimator**

```
fit <- sem(trimmed, data="twp", estimator="mlr")
```

3. Get the model summary

```
summary(fit, rsquare=T)
```



# Non-robust

Number of observations	267
Estimator	ML
Minimum Function Test Statistic	2.175
Degrees of freedom	1
P-value (Chi-square)	0.140

## Regressions:

	Estimate	Std.Err	z-value	P(> z )
satisfaction ~				
quality	0.411	0.060	6.860	0.000
perceivd_cntrl	0.767	0.082	9.353	0.000
quality ~				
perceivd_cntrl	0.688	0.074	9.334	0.000
understandblty	0.149	0.082	1.819	0.069

## R-Square:

	Estimate
satisfaction	0.515
quality	0.292





# Robust

Number of observations	267	
Estimator	ML	<b>Robust</b>
Minimum Function Test Statistic	2.175	<b>1.476</b>
Degrees of freedom	1	<b>1</b>
P-value (Chi-square)	0.140	<b>0.224</b>
Scaling correction factor for the Yuan-Bentler correction		<b>1.473</b>

## Parameter Estimates:

Information	Observed
Standard Errors	<b>Robust.huber.white</b>



# Robust

## Regressions:

	Estimate	Std.Err	z-value	P(> z )
satisfaction ~				
quality	0.411	<b>0.071</b>	<b>5.759</b>	<b>0.000</b>
perceivd_cntrl	0.767	<b>0.084</b>	<b>9.085</b>	<b>0.000</b>
quality ~				
perceivd_cntrl	0.688	<b>0.075</b>	<b>9.237</b>	<b>0.000</b>
understandblty	0.149	<b>0.098</b>	<b>1.516</b>	<b>0.130</b>

## Intercepts:

	Estimate	Std.Err	z-value	P(> z )
<b>.satisfaction</b>	<b>2.534</b>	<b>0.576</b>	<b>4.398</b>	<b>0.000</b>
<b>.quality</b>	<b>6.582</b>	<b>0.315</b>	<b>20.903</b>	<b>0.000</b>

## Variances:

	Estimate	Std.Err	z-value	P(> z )
.satisfaction	17.726	1.943	9.121	0.000
.quality	18.241	1.903	9.588	0.000

## R-Square:

	Estimate
satisfaction	0.515
quality	0.292



# Model fit

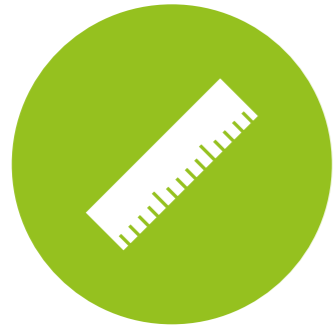
how to evaluate a path model



# Model chi-square

Chi-square test of model fit:

- Tests whether there any significant misfit between estimated and observed correlation matrix
- The chi-square value is compared against the model df
- The test does **not** work for a just-identified model
- When the model df is large, the model is usually significant ( $p < .05$ )... parsimonious models are rarely perfect!
- Alternative metric:  $\chi^2 / df < 3$  (good fit) or  $< 2$  (great fit)



# Baseline test

Chi-square test against baseline:

- Tests whether the model that considers all variables independent (unrelated)
- df of this test is  $\# \text{ parameters} - \# \text{ variables}$
- This is like the F-test in linear regression
- In other words, it is almost always significant, and thus rather useless



# Relative fit

Relative fit indices: how much improvement compared to the baseline?

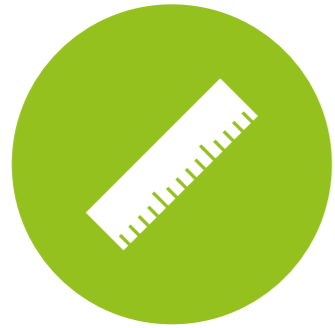
- Ranging from 0.00 to 1.00

$$\text{CFI: } ((\chi^2_b - df_b) - (\chi^2_m - df_m)) / (\chi^2_b - df_b)$$

- Bentler suggests that it should be  $> 0.96$

$$\text{TLI (or NNFI): } ((\chi^2_b/df_b) - (\chi^2_m/df_m)) / (\chi^2_b/df_b - 1)$$

- Suggested that it should be  $> 0.95$



# Approximate fit

RMSEA (root mean square error of approximation):

- $\sqrt{(\chi^2_m - df_m) / (df_m * (N-1))}$
- < 0.01 (excellent fit), < 0.05 (good fit), < 0.08 (mediocre fit), > 0.10 (poor fit)
- May be too high with low N and low df
- 90% confidence interval of RMSEA should fall below 0.10
- P-value RMSEA  $\leq 0.05$  should be  $> 0.05$  (meaning the RMSEA is not significantly larger than 0.05)



# Model fit in R

Getting additional fit statistics in R:

```
summary(fit, rsquare=T, fit.measures=T)
```

Number of observations	267	
Estimator	ML	Robust
Minimum Function Test Statistic	2.175	1.476
Degrees of freedom	1	1
P-value (Chi-square)	0.140	0.224
Scaling correction factor for the Yuan-Bentler correction		1.473

Model test baseline model:

Minimum Function Test Statistic	287.397	209.526
Degrees of freedom	5	5
P-value	0.000	0.000





# Model fit in R

User model versus baseline model:

Comparative Fit Index (CFI)	0.996	0.998
Tucker-Lewis Index (TLI)	0.979	0.988
Robust Comparative Fit Index (CFI)		0.997
Robust Tucker-Lewis Index (TLI)		0.987

Root Mean Square Error of Approximation:

RMSEA		0.066	0.042
90 Percent Confidence Interval	0.000	0.191	0.000 0.151
P-value RMSEA $\leq$ 0.05		0.266	0.405
Robust RMSEA			0.051
90 Percent Confidence Interval			0.000 0.212

**“It is the mark of a truly intelligent person  
to be moved by statistics.”**



**George Bernard Shaw**