# Introduction to Structural Equations

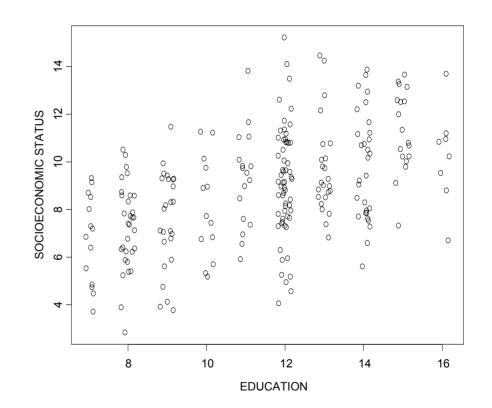
Statistics for Psychosocial Research II Structural Models October 30, 2006

# **Course Overview**

- (1) Structural Regression/Path Analysis
  - (a) "effect mediation" versus "controlling for"
  - (b) causality
- (2) Regression plus measurement structures from last term
  - (a) if we ignore measurement, "item regression"
  - (b) factor analysis: structural equations with latent variables
  - (c) latent class analysis: latent class regression

## **General Idea**

- How does outcome vary with predictors?
- Make inference on hypothesis about how predictors affect outcome
- Predict individual outcomes



# Challenge

- How do we measure latent outcomes (and predictors)?
- There are multiple responses
- Approach 1:
  - $Y_1, \ldots, Y_n$  measure the same thing. Treat individually or summarize Y's.
- Approach 2:
  - Call ideal outcome η
  - If we knew  $\eta$ , then  $\eta_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots$
  - But we don't know it:
    - infer  $\eta$  from factor analysis or latent class analysis
    - regress  $\eta$  on X's

Three approaches to assessing association between covariates and multiple responses

(1) Summarize then analyze (STA)

(2) Analyze then summarize (ATS)

(3) Summarize AND analyze: (SAA)

- Structural Equations
- 2 parts
  - measurement component
  - structural/regression component

## Example: Depression Study Summarize then Analyze (STA)

- Clinical trial of two antidepressants
- Which anti-depressant is more effective for treating depression?
- Depression symptoms were based on the Hamilton Depression Rating Scale (HAM-D).

- 17 Symptoms
  - Depressed mood
  - Guilt feelings
  - suicide
  - Insomnia (x3)
  - Work and activities
  - Psychomotor retardation
  - agitation
  - anxiety
  - Somatic symptoms
  - ....

For each item, write the correct number on the line next to the item. (Only one response per item)

1. DEPRESSED MOOD (Sadness, hopeless, helpless, worthless)

#### 0= Absent

1= These feeling states indicated only on questioning

2= These feeling states spontaneously reported

3= Communicates feeling states non-verbally—i.e., through facial expression, posture, voice, and tendency to weep

4= Patient reports VIRTUALLY ONLY these feeling states in his spontaneous verbal and non-verbal communication

- \_2. FEELINGS OF GUILT
- 0= Absent
- 1= Self reproach, feels he has let people down
- 2= Ideas of guilt or rumination over past errors or sinful deeds
- 3= Present illness is a punishment. Delusions of guilt
- 4= Hears accusatory or denunciatory voices and/or experiences threatening visual hallucinations
  - 3. SUICIDE
- 0= Absent
- 1= Feels life is not worth living
- 2= Wishes he were dead or any thoughts of possible death to self
- 3= Suicidal ideas or gesture
- 4= Attempts at suicide (any serious attempt rates 4)
  - 4. INSOMNIA EARLY
- 0= No difficulty falling asleep
- 1= Complains of occasional difficulty falling asleep—i.e., more than 1/2 hour
- 2= Complains of nightly difficulty falling asleep
  - 5. INSOMNIA MIDDLE
- 0= No difficulty
- 1= patient complains of being restless and disturbed during the night
- 2= Waking during the night—any getting out of bed rates 2 (except for purposes of voiding)
  - 6. INSOMNIA LATE
- 0= No difficulty
- 1= Waking in early hours of the morning but goes back to sleep
- 2= Unable to fall asleep again if he gets out of bed

## Example: Summarize then Analyze (STA)

- <u>Summarize</u>:
  - Add up the number of symptoms, or "score" the HAM-D.
  - Treat the score as "fixed" or "observed" outcome.
  - But, we know better! It is not measured perfectly.
  - What is the reliability of the HAM-D???
- <u>Analyze</u>: See how the outcome relates to predictor (i.e., treatment)

## Summarize Then Analyze

1. Sum up HAM-D score pre and post and take difference:

Pre-treatment score: $Y_{i1} = Y_{i1,1} + Y_{i1,2} + \dots + Y_{i1,21}$ Post-treatment score: $Y_{i2} = Y_{i2,1} + Y_{i2,2} + \dots + Y_{i2,21}$ Difference: $D_i = Y_{i2} - Y_{i1}$ 

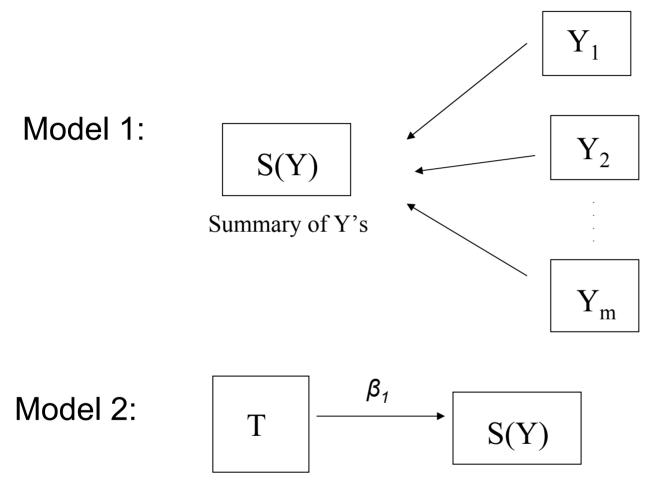
2. Evaluate association with  $Y_i$  and treatment

$$D_i = \beta_0 + \beta_1 trt_i$$

where  $trt_i = 1$  of treatment A, and 0 if treatment B

3. Make inference about treatment effect based on  $\beta_1$ 

### STA: Two models estimated separately



"treatment"

## STA: so what is the problem???

- We are ignoring that S(Y) is measured with error.
- Note that that S(Y) has reliability less than 1.
- In our example: S(Y) represents an "imperfect measure" of depression with reliability of about 0.88 (depending on population).
- Aren't we then overestimating the variation in our outcome by using S(Y)?
- Recall:  $Var(T_x) < Var(X)$
- What effect might that have on the standard error of  $\beta_1$ ?

# Another Approach: Analyze Then Summarize (ATS)

 <u>Analyze</u>: for each of the 21 items in the HAM-D, see if treatment is associated with improvement.

1. Define outcome per item:

$$D_{i,1} = Y_{i2,1} - Y_{i1,1}$$
  
:  
$$D_{i,21} = Y_{i2,21} - Y_{i1,22}$$

2. Estimate association per item with treatment:

$$D_{i,1} = \beta_{0,1} + \beta_{1,1} trt_i$$
  

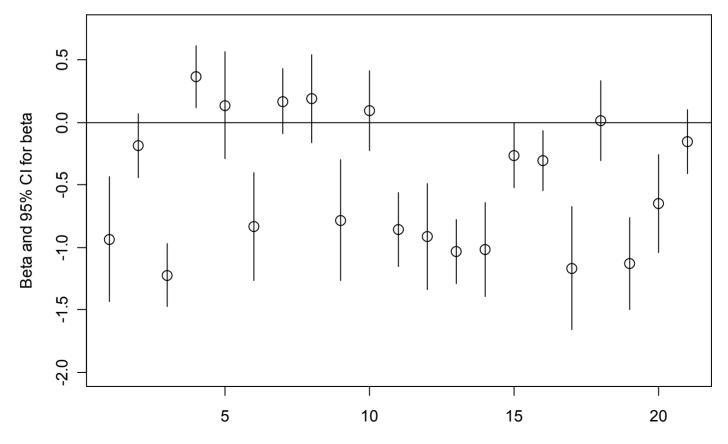
$$D_{i,2} = \beta_{0,2} + \beta_{1,2} trt_i$$
  

$$\vdots$$
  

$$D_{i,21} = \beta_{0,21} + \beta_{1,21} trt_i$$

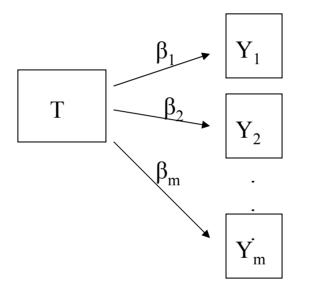
## Another Approach: Analyze Then Summarize (ATS)

2. <u>Summarize</u>: Qualitatively or quantitatively evaluate the associations



Item Number

## Analyze then Summarize



Fit *m* regressions to individually describe association between T and each Y.

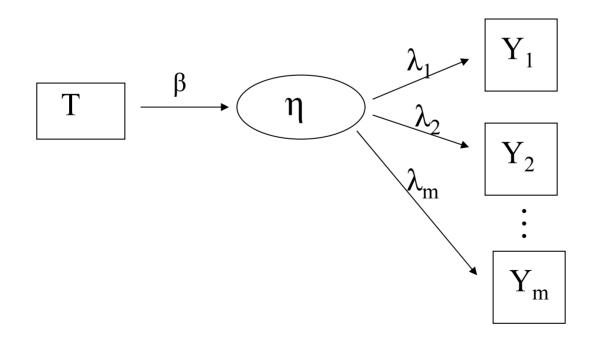
Then summarize associations.

# So what is wrong with ATS?

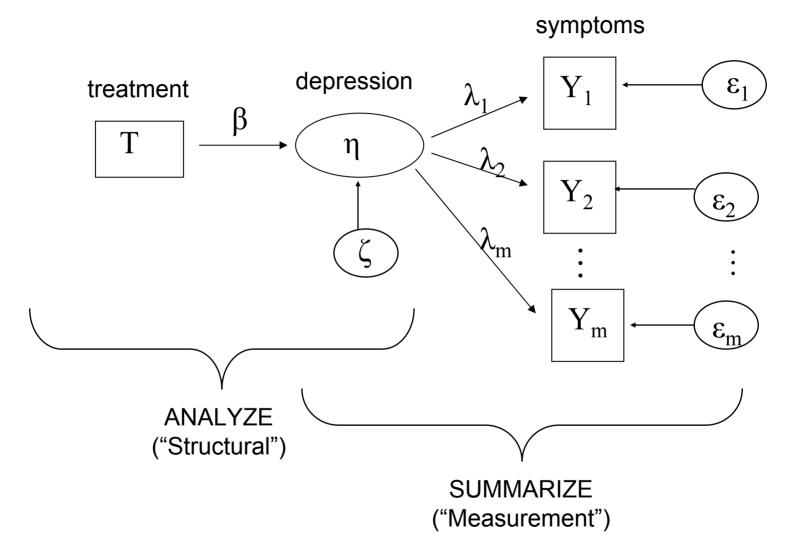
- How do we answer the question: "Which treatment works better?"
- We get individual answers.
- Often hard to summarize after the analysis has been done.
- (More about this in 'Item Regression lecture')

# Summarize and Analyze Simultaneously (SAA)

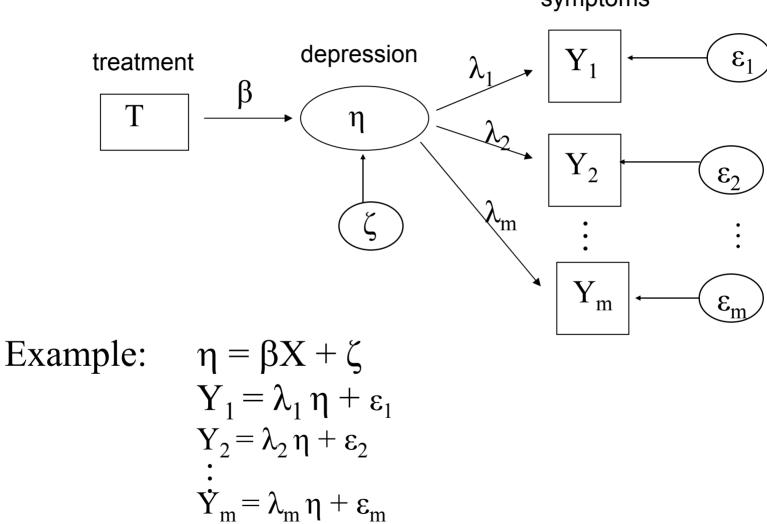
- Fit 'summarize' and 'analyze' components at the same time.
- One big model
- Accounts for measurement error of latent variable



#### Summarize and Analyze Simultaneously



#### Summarize and Analyze Simultaneously

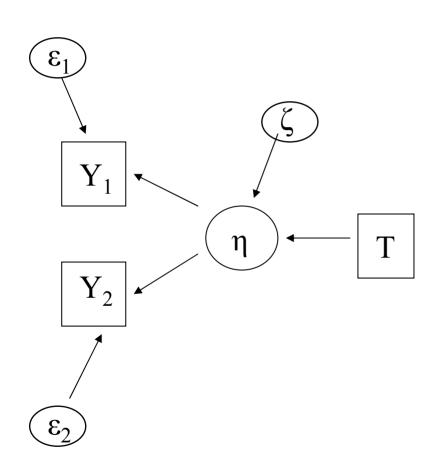


symptoms

## Caveat

- When is analyze then summarize better?
- What if some treatment affects some of the symptoms but not all of them?
- What does that imply about the measurement?

# Path Notation



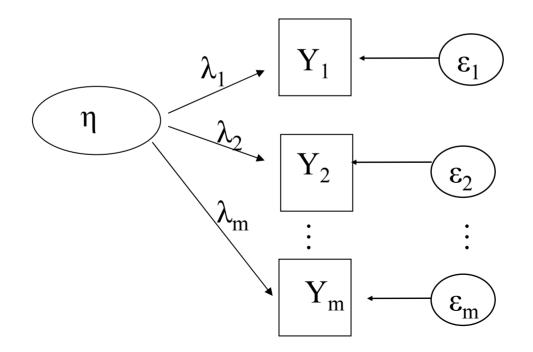
- Relationships
  - straight arrow (causal)
  - curved arrow (unspecified)

## Variables

- circles vs. squares
- exogenous (independent)
- endogenous (dependent)
- Errors
  - one for every endogenous variable
  - unexplained component of predicted variables

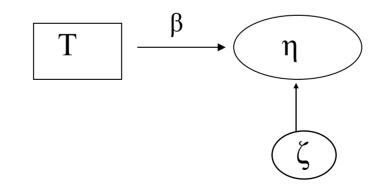
## Components of Structural Equation Model

- (A) Measurement Piece
  - how latent variable related to "surrogates"
  - comprised of  $\eta$ 's and Y's



## Components of Structural Equation Model

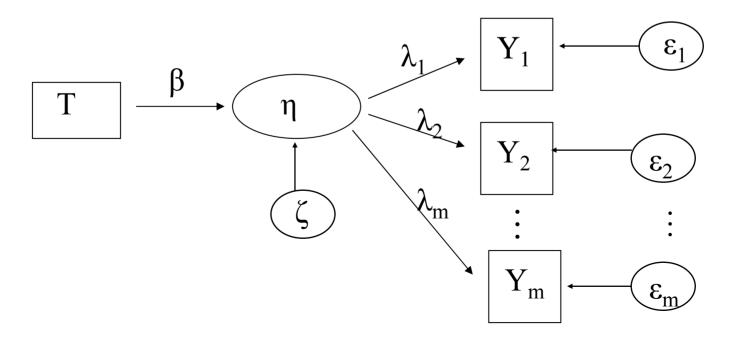
- (B) Structural Piece
  - how latent variable is related to its predictors
  - regression
  - comprised of  $\eta$  's and T



## Components of Structural Equation Model

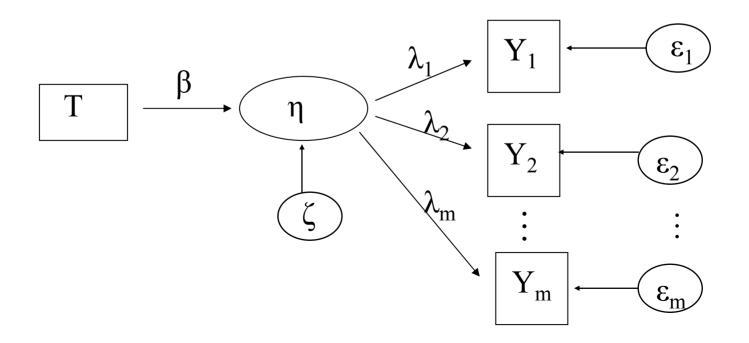
(C) Both components are fit in ONE step

Why better? Does not assume η (i.e., "summary" of Y's) known, which acknowledges measurement error.
Why bad? If model is misspecified, then inference is misleading.



# Statistical way of considering relationship between T and Y

$$P(Y = y | T) = \sum_{r=1}^{R} P(Y = y, \eta = r | T)$$
$$= \sum_{r=1}^{R} P(Y = y | \eta = r, T) P(\eta = r)$$

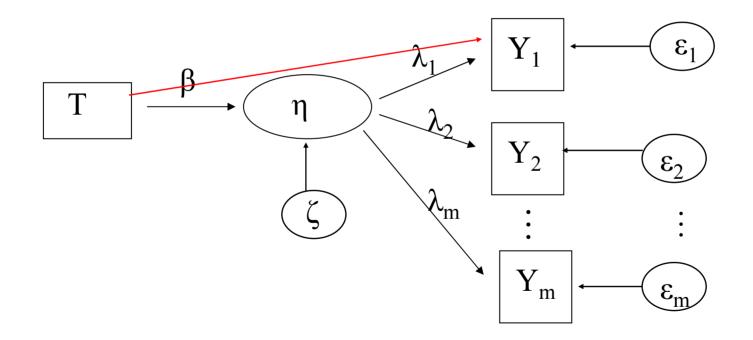


Assumption 1: Non-Differential Measurement

- Equivalent interpretations:
  - covariates do not predict observed responses after controlling for latent status
  - no arrows between T and Y's
  - Y and T independent given  $\eta$

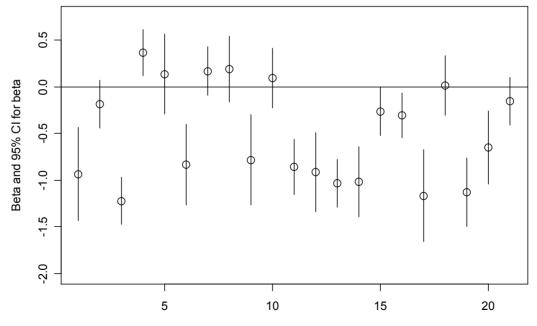
$$P(Y = y \mid \eta, T) = P(Y = y \mid \eta)$$

# **NOT** OK UNDER NON-DIFFERENTIAL MEASUREMENT:



# HAM-D Depression Example

- Does treatment affect the "depression" or symptoms?
- Implications for "differential measurement"?



Item Number

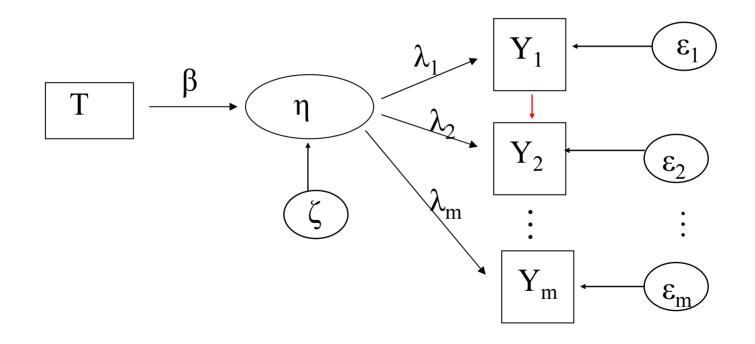
Assumption 2: Local/Conditional Independence

## **Equivalent Interpretations**

- latent variable explains all association between observed variables
- no arrows among measurement errors
   observed variables are independent given
   η

$$P(Y_1 = y_1, Y_2 = y_2 \mid \eta) = P(Y_1 = y_1 \mid \eta) P(Y_2 = y_2 \mid \eta)$$

#### **NOT** OK UNDER CONDITONAL INDEPENDENCE:



#### **NOT** OK UNDER CONDITIONAL INDEPENDENCE:

