

Generalizing sem in Stata

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Feature requests

Can **sem** support outcomes that are

- binary?
- counts?
- ordinal?
- categorical?

Is there a command that is like

- **oprobit** for panel data?
- **xtmelogit** but uses probit instead?
- **xtmepoisson** but uses negative binomial instead?

New command

The new **gsem** command

- Think **G**eneralized **S**tructural **E**quations **M**odel
- Inspired by **gllamm** and **sem**
- Documented in [SEM]

Features

- Discrete and continuous outcomes
- Multi-level latent variables
- Maximum likelihood estimator
- Supported by the SEM Builder

Modeling outcomes

Syntax

```
gsem y <- x , family(family) link(link)
```

Supported family and link combinations

	logit	probit	cloglog	log	identity
Bernoulli	x	x	x		
binomial	x	x	x		
ordinal	x	x	x		
multinomial	x				
Poisson				x	
negative binomial				x	
gamma				x	
Gaussian				x	x

Latent variables

Syntax

- Latent variables – **sem** style
 - Capitalized names
 - **latent()** option
- Multi-level latent variables – notice square brackets
 - $Lname[level_spec]$
- Random slopes – uses interaction notation
 - $Lname[level_spec]\#...$

Integration methods

- Mean-variance adaptive quadrature
- Mode-curvature adaptive quadrature
- Gauss-Hermite quadrature
- Laplacian approximation

Maximum likelihood estimator

Model construction

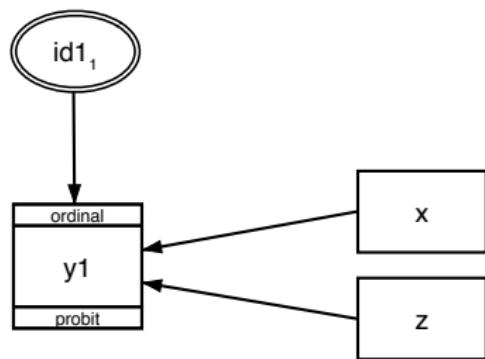
- Outcomes are independent of each other, conditionally on
 - Latent variables
 - Other observed variables, with an exception
 - Linear outcomes (Gaussian/identity) modeled jointly
- Covariance estimation
 - Linear outcome errors
 - Latent variables within a given level

Multivariate outcomes models

- All recursive models are allowed
- Non-recursive models (loops) only allowed for linear outcomes

SEM Builder Demonstration

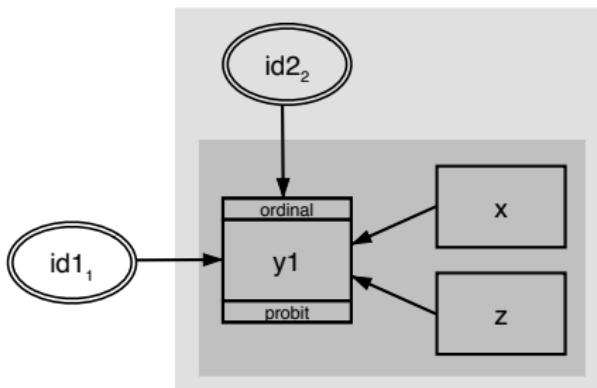
Example: oprobit for panel data



```
gsem y1 <- x z ||id1, family(ordinal) link(probit)
```

```
xtpoprobit y1 x z
```

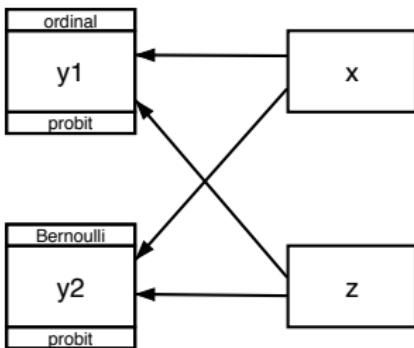
Example: oprobit for multi-level data



```
gsem y1 <- x z || id1 || id2, oprobit
```

```
meoprobit y1 x z || id1: || id2:
```

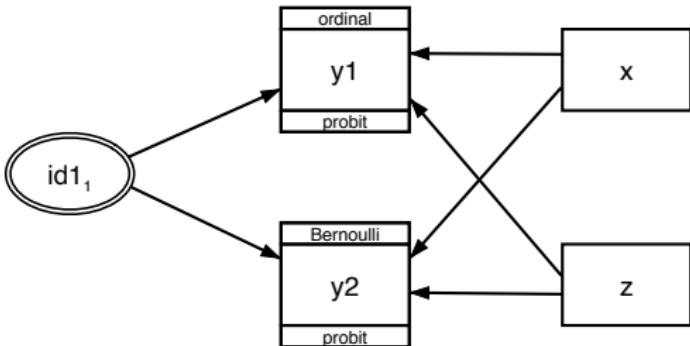
Example: seemingly unrelated estimation



```
gsem (y1 <- x z, oprobit) (y2 <- x z, probit) , vce(robust)
```

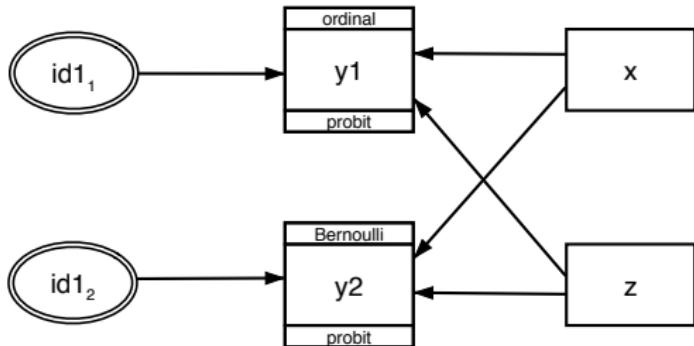
```
oprobit y1 x z  
estimates store y1  
probit y2 x z  
suest y1 .
```

Example: bivariate response with a random effect



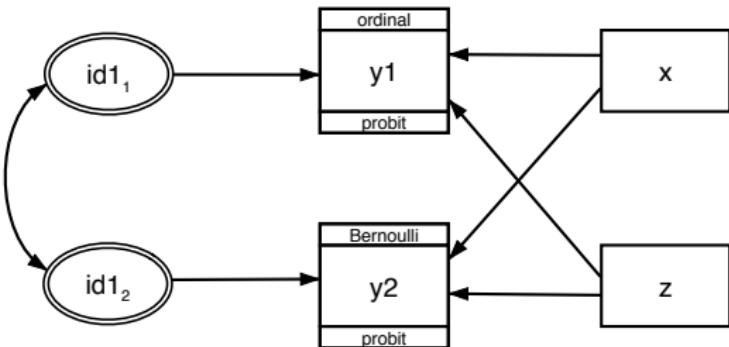
```
gsem (y1 <- x z I[id1], oprobit) (y2 <- x z I[id1], probit)
```

Example: bivariate, with 2 independent random effects



```
gsem (y1 <- x z I[id1], oprobit) (y2 <- x z J[id1], probit),  
covariance(I[id1]*J[id1]@0)
```

Example: bivariate, with 2 random effects



```
gsem (y1 <- x z I[id1], oprobit) (y2 <- x z J[id1], probit)
```

Conclusion

What's next

- more family/link combinations
- alternatives to quadrature
- improve performance for linear outcomes models
- ...