

# modsem: An R Package for Estimating Latent Interactions and Quadratic Effects

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## modsem

- ▶ `modsem` is an R package for estimating latent interaction and quadratic effects, in *Structural Equation Models* (SEMs).

# Frameworks

- ▶ Product Indicator (PI) approaches
- ▶ Distribution Analytic (DA) approaches

## Product Indicator (PI) Approaches

## Product Indicator (PI) Approaches

- ▶ First attempt at latent interaction in SEMs (Kenny and Judd 1984)
- ▶ Creates product indicators, used as indicators for latent interaction terms.
- ▶ Has traditionally required manual model specification, and manual construction of product indicators.
- ▶ Early approaches involved complicated model constraints.
- ▶ Manual specification (partially) led to a continual simplification of constraints.

# Product Indicator (PI) Approaches (History)

- ▶ Constrained Approach (orthogonal specification) (Kenny and Judd 1984)
- ▶ Constrained Approach (oblique specification) (Jöreskog and Yan 1996)
- ▶ Constrained Approach (mean-centered indicators) (Algina and Moulder 2001)
- ▶ Unconstrained Approach (constrained latent mean) (Martin and Marsh 1999)
- ▶ Residual Centering Approach (no constraints) (Little, Bovaird, and Widaman 2006)
- ▶ Double Centering Approach (no constraints) (Lin et al. 2010)

# Product Indicator (PI) Approaches (Tools)

- ▶ `semTools` offers tools for creating the product indicators, but does not help specify the model.
- ▶ Thus the PI approaches largely require manual specification.
- ▶ `modsem` automatically handles the creation of product indicators, and model specification.
- ▶ Model specification becomes exponentially more complicated for models with more indicators, more interaction terms. Especially for the PI approaches with model constraints.

## Example: Double Centering Approach, using semTools

```
model <- '  
# Measurement Model  
X  =~ x1 + x2 + x3  
Z  =~ z1 + z2 + z3  
Y  =~ y1 + y2 + y3  
XZ =~ x1.z1 + x2.z1 + x3.z1 +  
      x1.z2 + x2.z2 + x3.z2 +  
      x1.z3 + x2.z3 + x3.z3  
  
# Structural Model  
Y ~ X + Z + XZ  
  
# Residual Covariances  
x1.z1 ~~ x1.z2 + x1.z3 + x2.z1 + x3.z1  
x1.z2 ~~ x1.z3 + x2.z2 + x3.z2  
x2.z1 ~~ x2.z2 + x2.z3 + x3.z1  
  
x1.z3 ~~ x2.z3 + x3.z3  
x2.z2 ~~ x2.z3 + x3.z2  
x3.z1 ~~ x3.z2 + x3.z3  
  
x2.z3 ~~ x3.z3  
x3.z2 ~~ x3.z3  
,
```



## Example: Double Centering Approach, using semTools

```
library(semTools)
data.prod <- indProd(data = oneInt,
                     var1 = c("x1", "x2", "x3"),
                     var2 = c("z1", "z2", "z3"),
                     match = FALSE)

fit <- sem(model, data = data.prod)
summary(fit)
```

Regressions:

	Estimate	Std.Err	z-value	P(> z )
Y ~				
X	0.675	0.027	25.379	0.000
Z	0.561	0.026	21.606	0.000
XZ	0.702	0.027	26.360	0.000

## Example: Double Centering Approach, using modsem

```
library(modsem)

model <- '
# Measurement Model
  X  =~ x1 + x2 + x3
  Z  =~ z1 + z2 + z3
  Y  =~ y1 + y2 + y3

# Structural Model
  Y ~ X + Z + X:Z
'

fit <- modsem(model, data = oneInt)
summary(fit)
```

Regressions:

	Estimate	Std.Err	z-value	P(> z )
Y ~				
X	0.675	0.027	25.379	0.000
Z	0.561	0.026	21.606	0.000
XZ	0.702	0.027	26.360	0.000

## Visualizing Interaction Effects

# Example Model

```
model <- '
  visual  =~ x1 + x2 + x3
  textual =~ x4 + x5 + x6
  speed   =~ x7 + x8 + x9

  visual ~ speed + textual + speed:textual
'

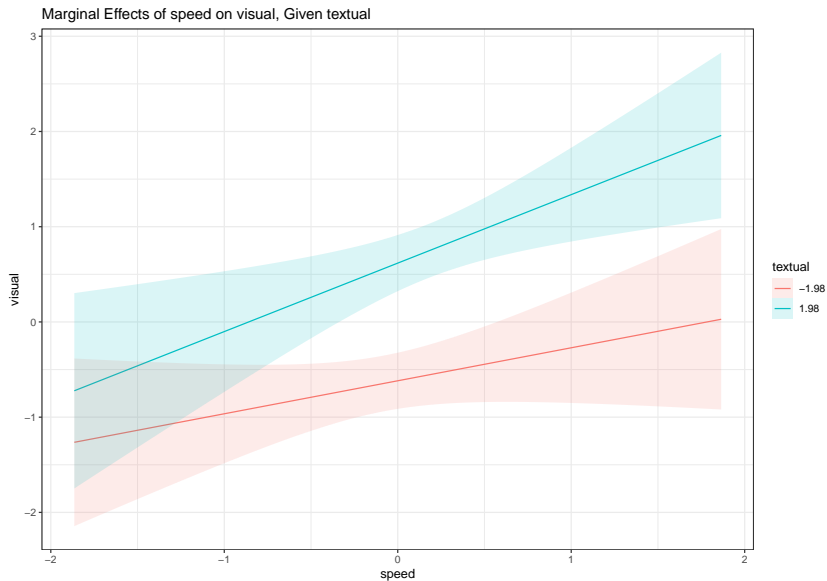
fit <- modsem(model, data = HolzingerSwineford1939, method = "lms")
summary(fit)
```

Regressions:

	Estimate	Std.Error	z.value	P(> z )
visual ~				
textual	0.312	0.076	4.107	0.000
speed	0.533	0.158	3.366	0.001
speed:txtl	0.094	0.093	1.016	0.309

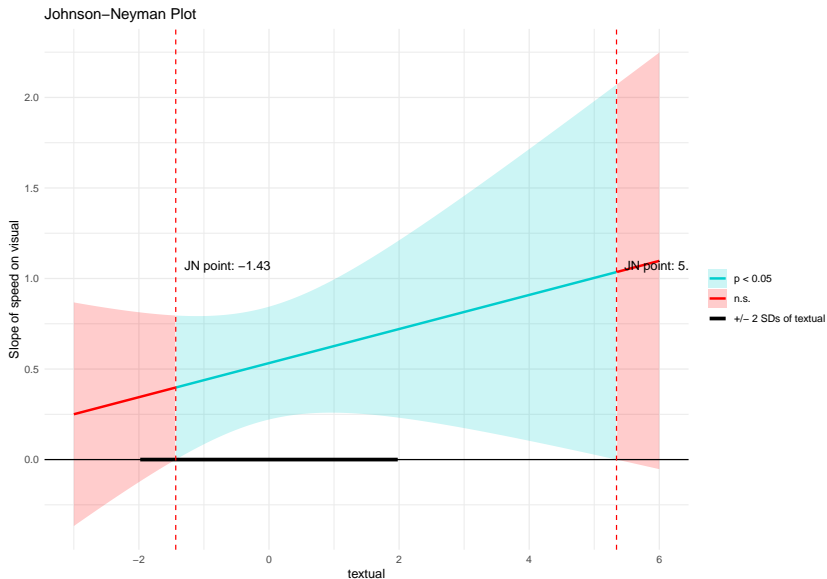
# Plotting Margins (Simple Slopes)

```
plot_interaction(x = "speed", z = "textual", y = "visual", model = fit, vals_z
```



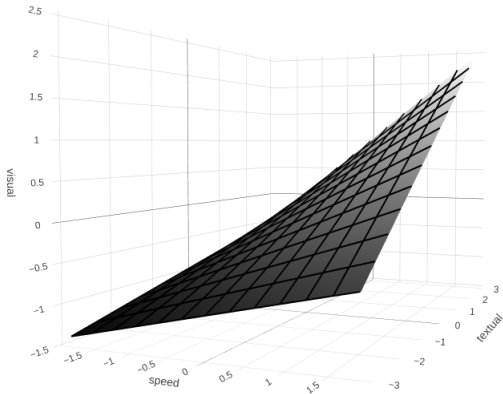
# Johnson-Neyman Plot

```
plot_jn(x = "speed", z = "textual", y = "visual", model = fit, max_z = 6)
```



# Surface Plot

```
plot_surface(x = "speed", z = "textual", y = "visual", model = fit,  
            colorscale = "Greys", grid = TRUE, grid_color = "black")
```



## Visualizing Models With Quadratic Effects



# Example Model

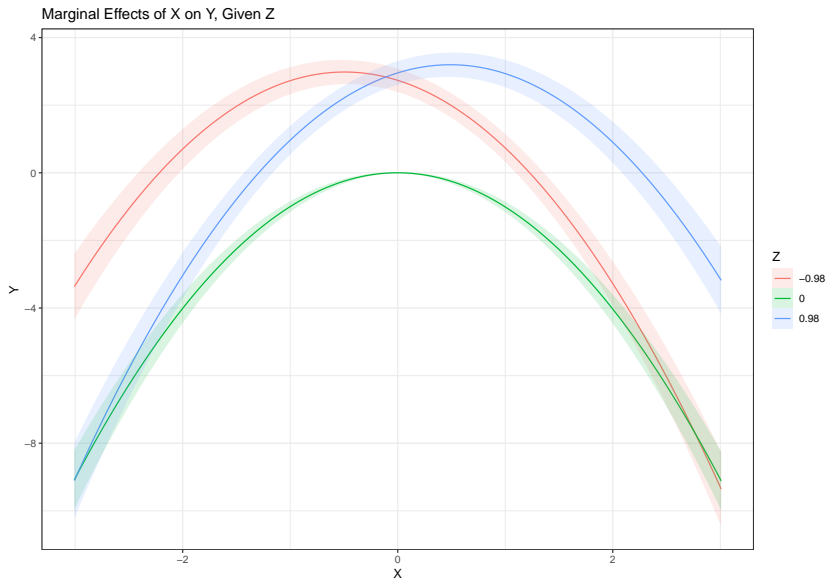
```
model <- '  
  X =~ x1 + x2 + x3  
  Z =~ z1 + z2 + z3  
  Y =~ y1 + y2 + y3  
  
  Y ~ X + Z + X:X + Z:Z + X:Z  
,  
  
fit <- modsem(model, data = data.quadratic, method = "qml")  
summary(fit)
```

Regressions:

	Estimate	Std.Error	z.value	P(> z )
Y ~				
X	-0.005	0.067	-0.081	0.935
Z	0.113	0.163	0.693	0.488
X:X	-1.006	0.044	-23.040	0.000
X:Z	1.011	0.080	12.571	0.000
Z:Z	2.969	0.079	37.362	0.000

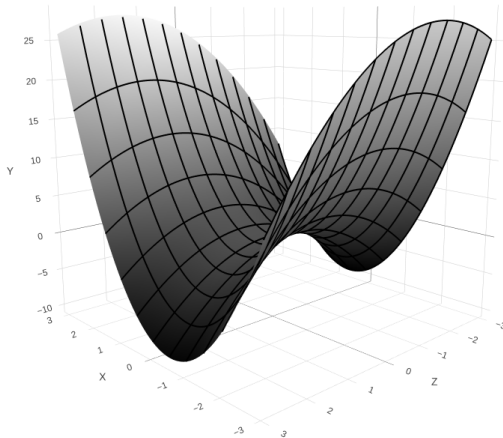
## 2D Plot

```
plot_interaction(x = "X", z = "Z", y = "Y", model = fit, vals_z = c(-1, 0, 1))
```



# 3D (Response Surface) Plot

```
plot_surface(x = "X", z = "Z", y = "Y", model = fit,  
             colorscale = "Greys", grid = TRUE, grid_color = "black")
```



## References

- Algina, James, and Bradley C. Moulder. 2001. "A Note on Estimating the Jöreskog-Yang Model for Latent Variable Interaction Using LISREL 8.3." *Structural Equation Modeling* 8 (1): 40–52. [https://doi.org/10.1207/S15328007SEM0801\\_3](https://doi.org/10.1207/S15328007SEM0801_3).
- Jöreskog, Karl Gustav, and Fan Yan. 1996. "Nonlinear Structural Equation Models: The Kenny-Judd Model With Interaction Effects."
- Kenny, David A., and Charles M. Judd. 1984. "Estimating the Nonlinear and Interactive Effects of Latent Variables." *Psychological Bulletin* 96 (1): 201–10. <https://doi.org/10.1037/0033-2909.96.1.201>.
- Lin, Guan-Chyun, Zhonglin Wen, Herbert W. Marsh, and Huey-Shyan Lin. 2010. "Structural Equation Models of Latent Interactions: Clarification of Orthogonalizing and Double-Mean-Centering Strategies." *Structural Equation Modeling: A Multidisciplinary Journal* 17 (3): 374–91. <https://doi.org/10.1080/10705511.2010.488999>