# Determining Power and Sample Size for Mediation Models

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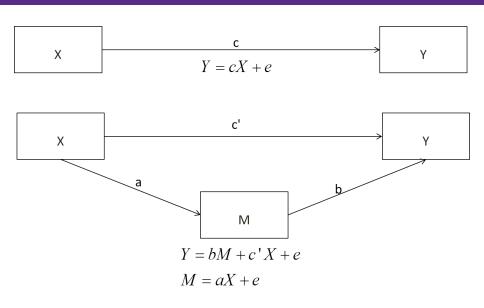
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#### **Basic Mediation**

- Three variable system
  - Predictor variable X
  - Outcome variable Y
  - Mediator variable M
- M (partially) explains the relationship between X and Y

# Mediation Vocabulary



# What do that paths tell us about mediation?

- When investigating mediation we are interested in if strength of the relationship between X and Y is reduced when M is included in the regression equation
  - Is the direct effect (c') less than the total effect (c)?
  - Does c c' = 0?
- c c' is also called the indirect effect
- The indirect effect can also be estimated by multiplying a\*b
- a\*b and c c' should provide the same value

#### Testing the indirect effect

- When we test for mediation we test if the indirect effect is different from zero.
  - If the indirect effect does not differ from zero there is no evidence of mediation
  - If the indirect effect differs from zero there is evidence of mediation

#### Testing the indirect effect

- There have been many methods developed to assess the indirect effect:
  - Difference in coefficients (Baron & Kenny, 1984)
  - Sobel test
  - Distribution of the product method
  - Bootstrap confidence intervals
  - Monte Carlo confidence intervals

## Testing the indirect effect: Power

- Methods of testing the indirect effect have different power for the same data (e.g. Fritz & MacKinnon, 2007, MacKinnon, Lockwood, & Williams, 2004)
  - Difference in coefficients method has extremely low power (especially when c is small)
  - Distribution of the product, bootstrap CI and Monte Carlo CI have higher power
    - Power for these methods is generally comparable

#### Testing the indirect effect: Power

- What else affects power to detect an indirect effect?
  - Sample size
  - Effect size
    - Effect size of both the a and b path

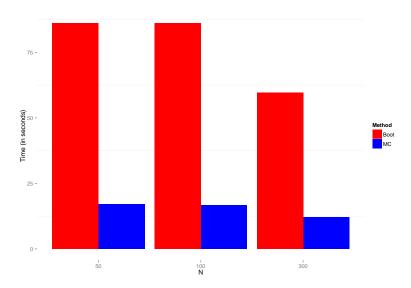
#### Determing power for an indirect effect

- Methods of power analysis should match methods of analysis
- No analytic method of power analysis for distribution of the product, bootstrap CI or Monte Carlo CI!
- We need to use a Monte Carlo power analysis

- General steps in a Monte Carlo Power Analysis
  - Specify all population parameters
  - Create a sample of size N, based on population parameters
  - Analyze sample data from Step 2 with chosen statistical method(s)
  - **4** Repeat steps 2 and 3 for each of r replications (often r>1000)
  - The proportion of replications with a significant parameter is an estimate of power (for all parameters not equal to 0)

- Best practice approach for assessing power in mediation models (Zhang, 2014)
  - Can utilize bootstrapping, distribution of the product method, or Monte Carlo CI
- Two main limitations to Monte Carlo Power Analysis for mediation models
  - Limited, user friendly software (e.g. Mplus, bmem, simsem)
  - Very time consuming (especially with boostrapping!)

- Overcoming limitations
  - Use Monte Carlo CI to test indirect effects
    - Faster than bootstrapping
    - Estimates of power are extremely similar (r = .99)

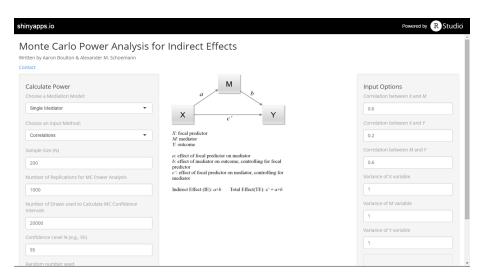


- Overcoming limitations
  - Use Monte Carlo CI to test indirect effects
    - Faster than bootstrapping
    - Estimates of power are extremely similar (r = .99)
  - Create user friendly software
    - Shiny app for simple mediation models

# Power estimation app

- Available from MARlab.org
- Web based or run locally on your computer
  - Requires R is installed on a computer (it helps if RStudio is too)
- Population parameters can be entered as correlations or  $r^2$  for each path

# Power estimation app



#### Future directions

- Expansion of models available in app
  - 2 and 3 mediator models
  - Conditional process models
- Integration of other methods of sample size determination
  - Accuracy in parameter estimation (Kelly & Maxwell, 2003)
  - Varying N Monte Carlo simulations (Schoemann, Pornprasermanit, Miller, & Wu, 2014)

## Thank you!

- Slides and shiny app from today at: http://http://MARlab.org/Supplemental\_Materials/
- email: schoemanna@ecu.edu

#### References

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