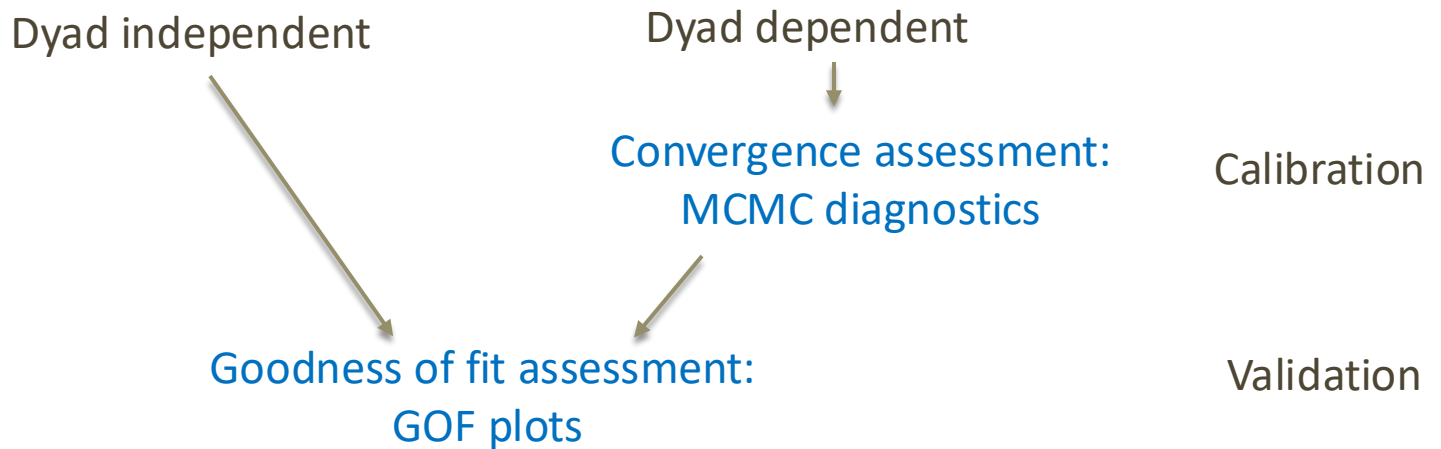


1

# ERGM model assessment

# Fitting and diagnosing a model

- The steps depend on the type of model you have
  - If you have a **dyad dependent** model, you first check convergence
- In both cases you end with goodness of fit:



# We'll start with MCMC dx

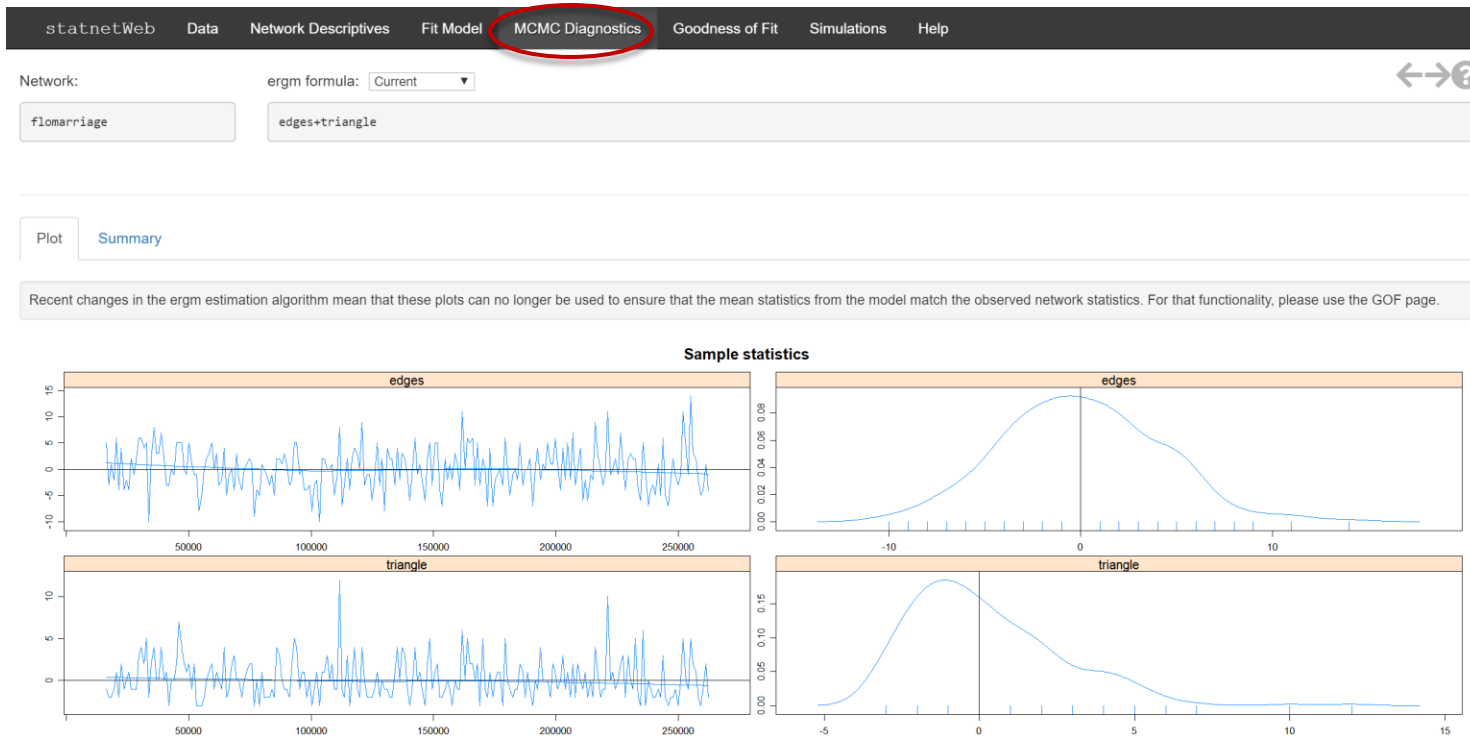
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- Because you need to do this first
- But it's a whole workshop in its own right
- So this is just to get started

# What are MCMC Diagnostics?

- MCMC Dx show the details of the sampling process
  - Traditionally taken from the last iteration in the MCMC chain
- They tell us if the estimation algorithm:
  - Is “mixing well”
    - Is it getting stuck in part of the space for many timesteps?
    - Is there a lot of autocorrelation in the samples?
  - Converged to the target value
    - Is it still bouncing around a lot?
    - Does the mean of the sampling distribution of the stats = the target?

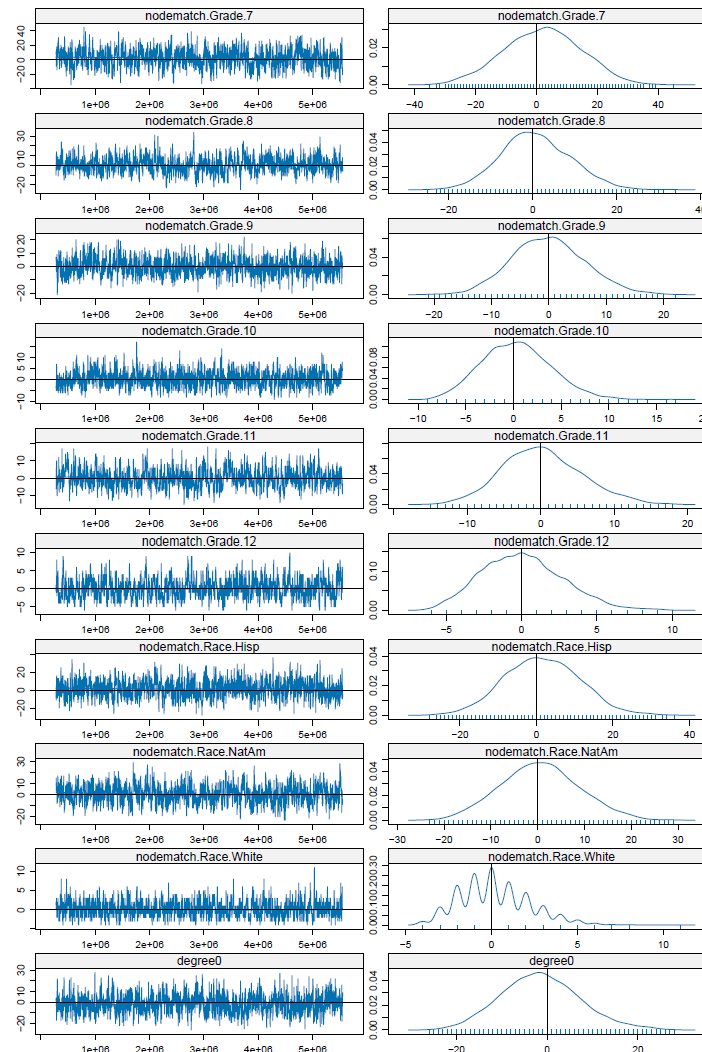
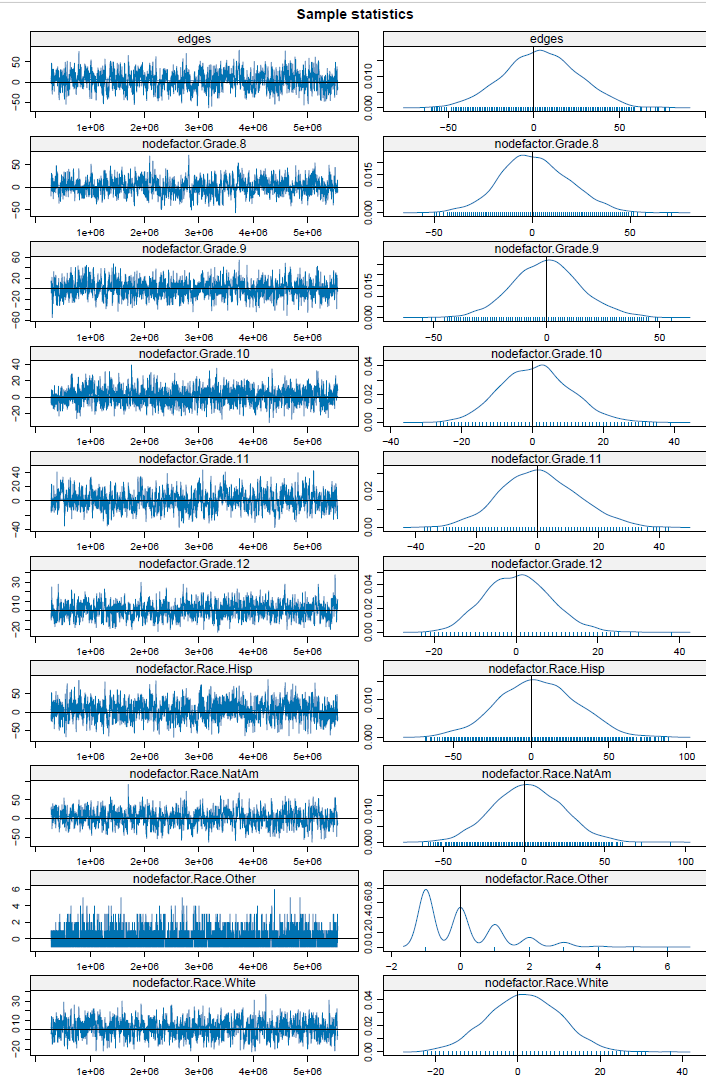
# Example of good MCMC Diagnostics



The **traceplots** on the left here display a good random walk pattern around the target value (a fuzzy caterpillar)

The distribution of sampled statistics on the right is centered (mean) on the target values.

# Note: Dx are plotted for each term (!)



These are all  
the plots from  
model 5 in  
Module 2:  
ERGMs

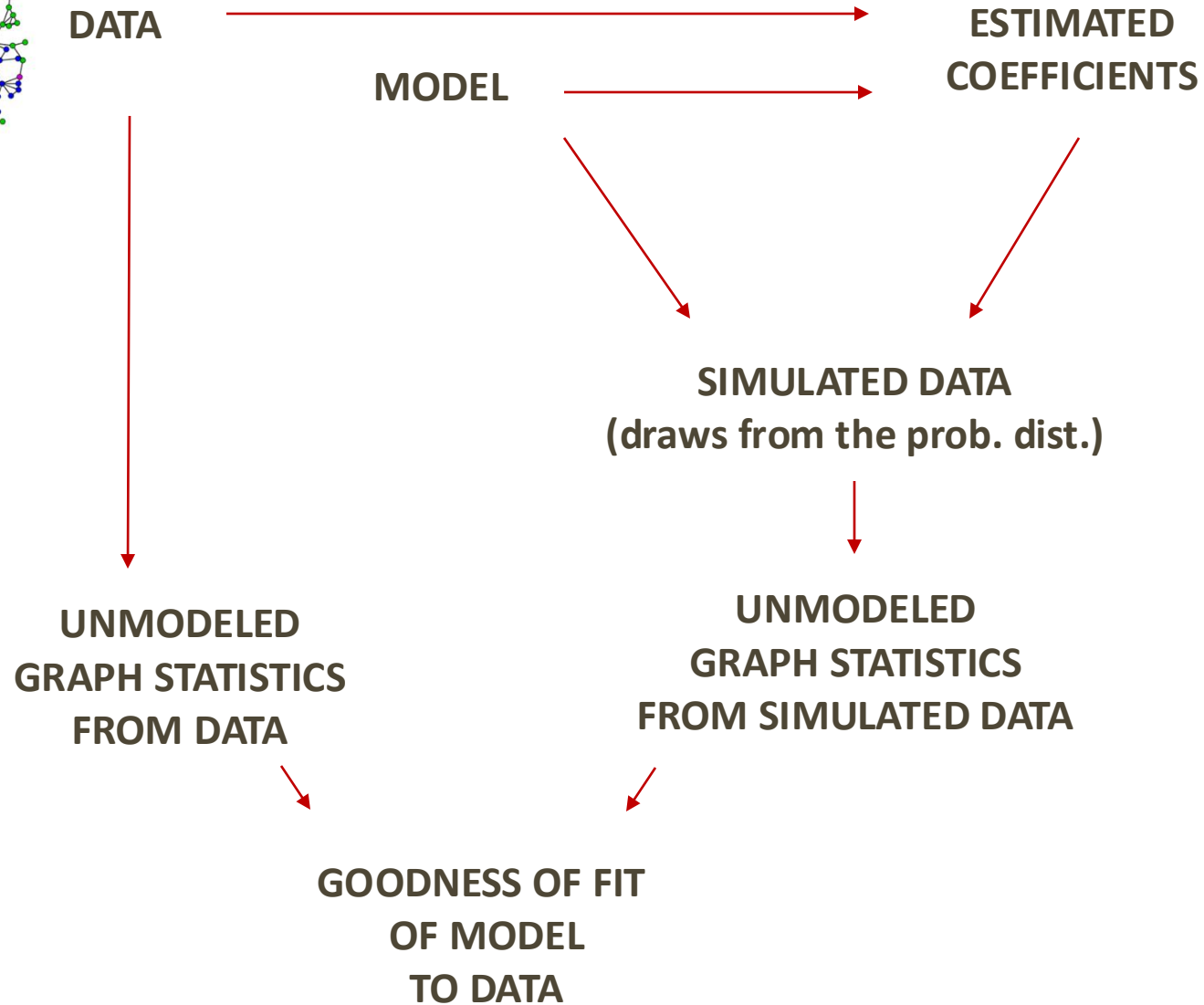
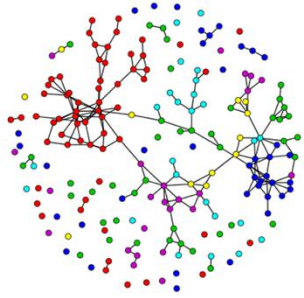
# What to do if this doesn't look good

- That depends on what you see
  - Major problems are things like bimodal distributions and traceplots or large deviations from the target values
  - Minor problems are things like moderate autocorrelation in the traceplots
- For major problems – think about changing model specification
- For minor problems – think about using one of the many (many) MCMC control parameters in ergm
- This is a deep subject
  - With lots of helpful information out on the web

# Goodness of Fit (GOF)

- Traditional GOF stats can be used
  - AIC, BIC are included in the model summary
- We also take another approach
  - Does the model reproduce other network properties that were not included as model terms? Kind of like “out of sample” prediction.
  - We use the full distributions of 3 “higher order” statistics:
    - Degree
    - Shared partners (local clustering)
    - Geodesic distances (global clustering)





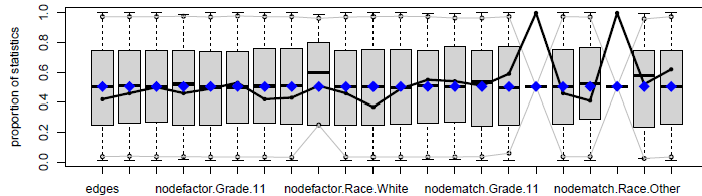
# Note: Using MCMC here again

- In this case we're using it for network simulation
- We simulate 100 networks from the fitted model
  - This is a sample from the probability distribution defined by the fitted model
  - On average, the MODELED network statistics from these simulated networks should match the observed targets
    - If not, there's a problem in convergence that needs to be fixed first
  - But now we can also see what this model produces for the UNMODELED statistics

# Default GOF stats used in `ergm` package

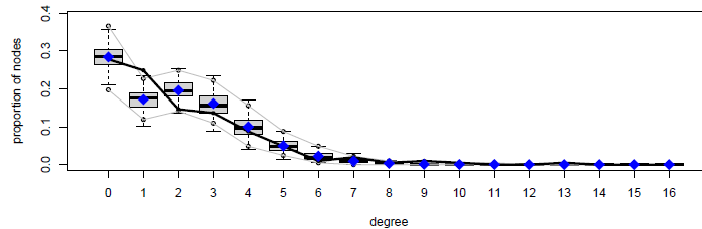
- Degree distribution
  - Node based measure
  - Counts the number of ties on each node
  - Very basic property of a network, important to get right
- Edgewise Shared Partner (ESP) distribution
  - Tie based measure
  - Counts the shared partners for each edge in the network
  - A measure of triadic closure, important for local clustering
- Geodesic distribution
  - Dyad based measure
  - Counts the shortest path between every dyad in the network
  - A measure of global connectivity

# GOF plots in `ergm` (the defaults)



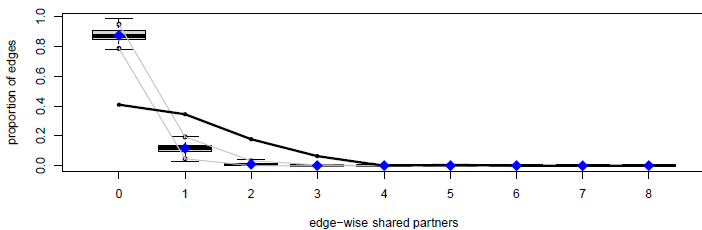
- The top plot is the model statistics

*Calibration  
assessment*

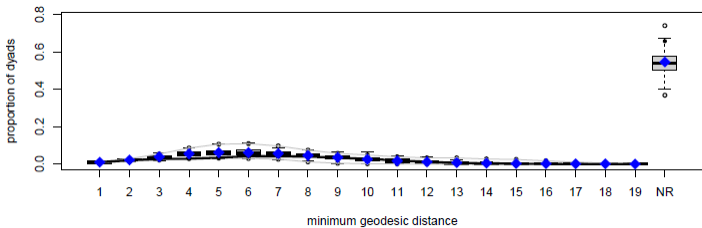


- The bottom 3 plots are the unmodeled statistics

*Validation  
assessment*



- Degree
- Shared partners
- Geodesics



Also taken from Model 5  
in Module 2: ERGMs