

**Emily Alicea-Muñoz**

Penn State Astronomy & Astrophysics

NASA/GSFC Code 663

Thesis Advisor: M. Coleman Miller (UMD)

# Black Hole Mergers as Probes of Structure Formation

11th Eastern Gravity Meeting

13 May 2008

# Background and Motivation

- Hierarchical Structure Formation
- Little observational data of  $z \sim 5 - 20$  era
- Gravitational waves to the rescue!
  - Massive BH mergers trace mergers of halos/galaxies
  - Observable for redshifts where there is currently little observational data
- Problem: Estimates of merger rates span 5 - 6 orders of magnitude

# Goals

- Develop semi-analytical, phenomenological model of MBH mergers, with several representative parameters.
- Use statistical methods to generate synthetic LISA observable data (total BH mass, redshifts, merger event rates).
- Blindly analyze data to evaluate confidence regions of parameters.

# Method

- Assume halo number densities at redshift  $z+dz$  given by Sheth & Tormen (1999) modification of Press-Schechter (1974) algorithm.
- Probability of merger by redshift  $z$  given by Extended Press-Schechter (Lacey & Cole 1993).
  - Note that this is single-step only; we do not construct a full merger tree!
- Generation of synthetic data (masses, redshifts, merger rates).
- MCMC to constrain model parameters given the data; confidence regions.



# Tests and Results

- Four model parameters:

- min halo mass:
- power law:

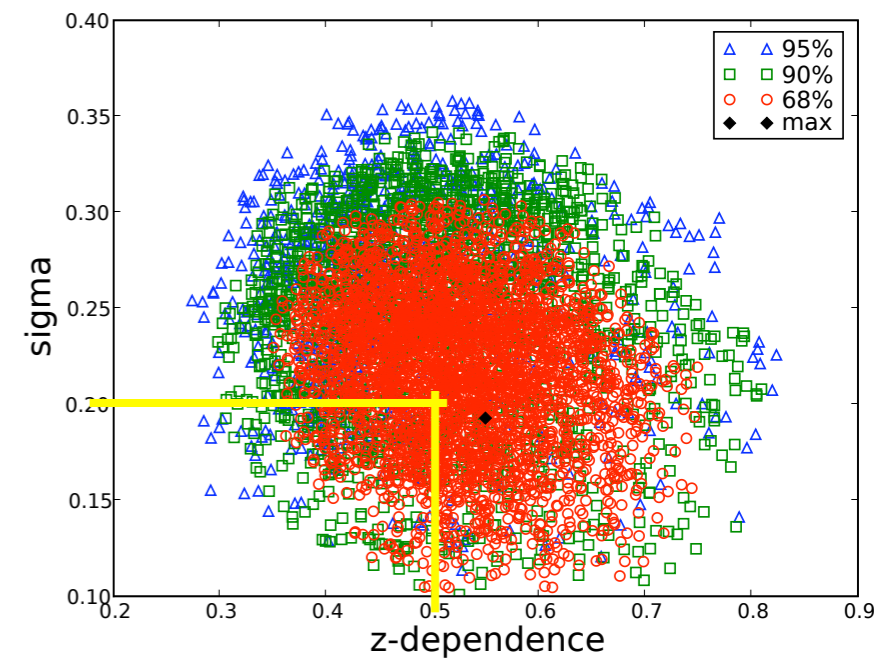
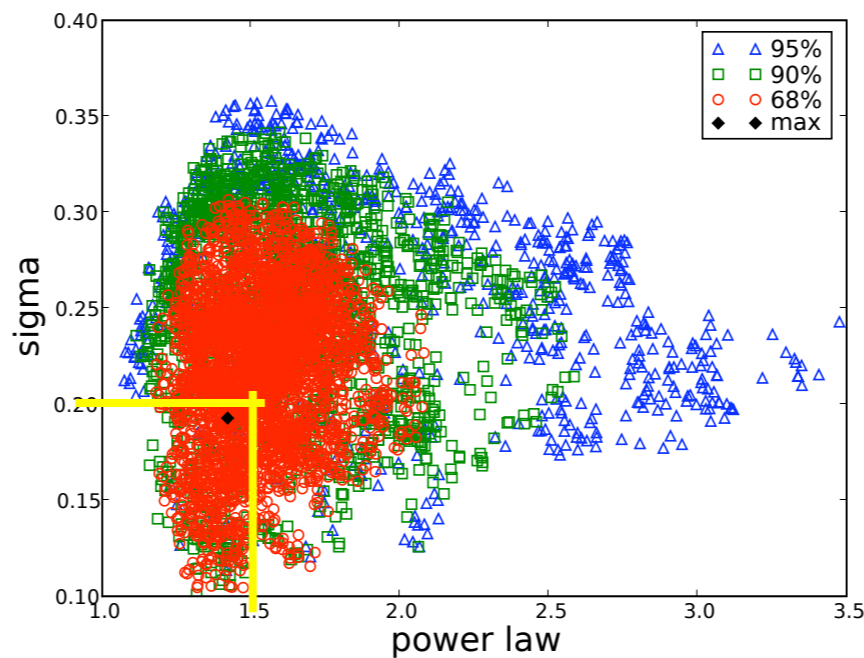
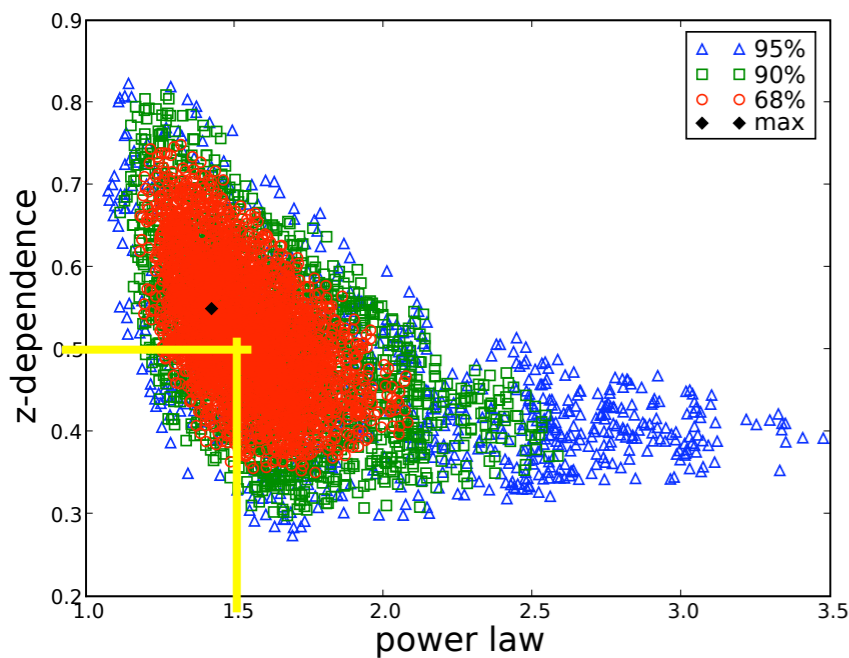
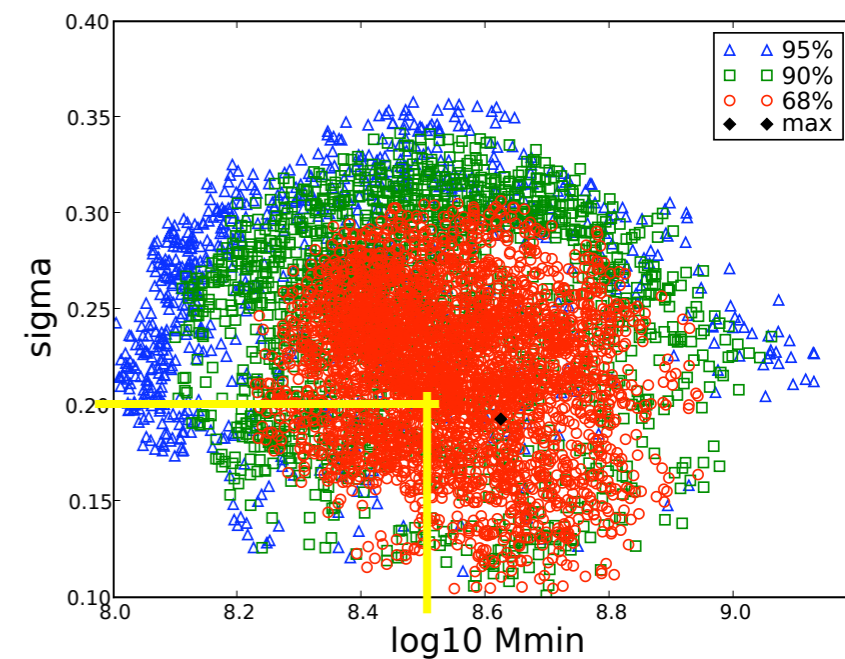
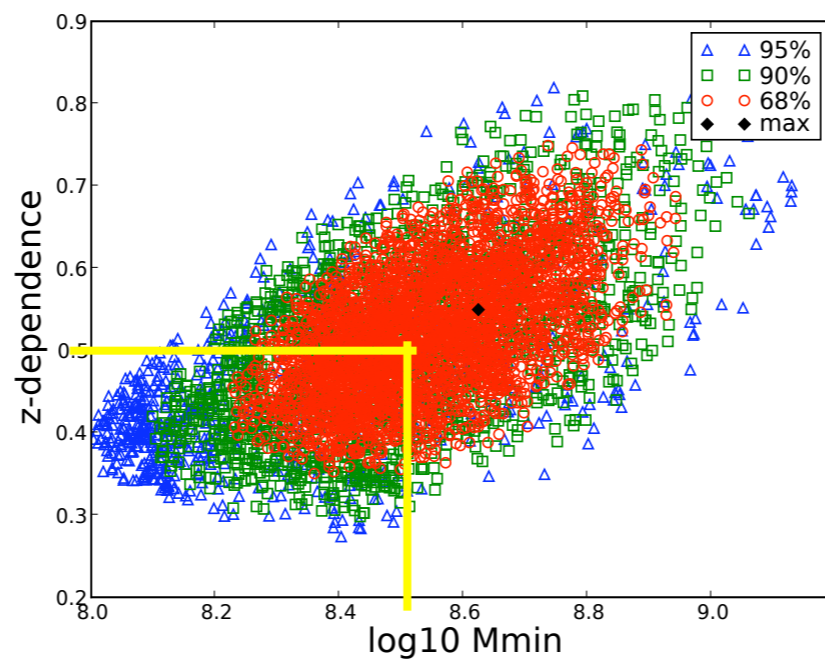
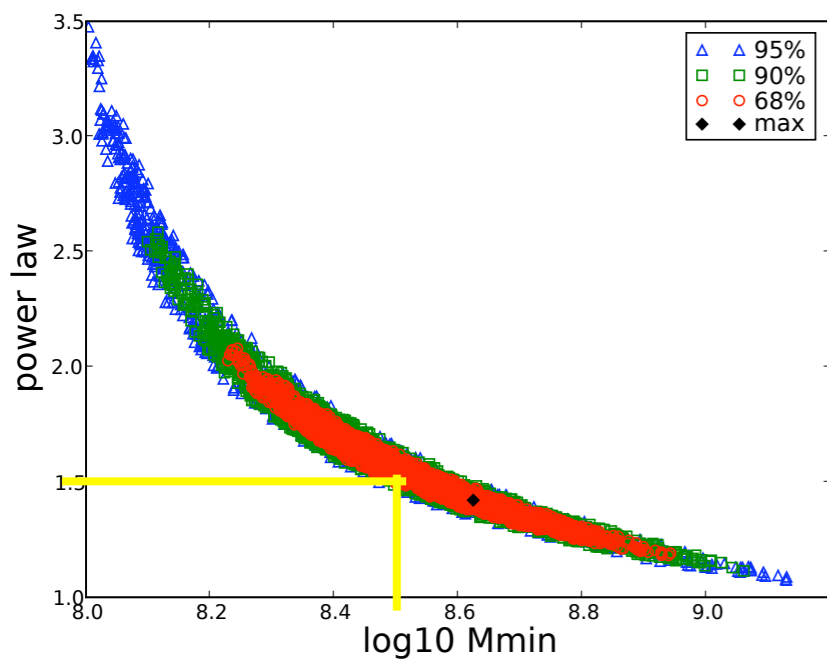
$$P_{\text{occ}}(M, z = 20) = \begin{cases} 1, & M > M_{\text{min}} \\ \left(\frac{M}{M_{\text{min}}}\right)^p, & M < M_{\text{min}} \end{cases}$$

- z-dependence:  $M_{\text{BH},0} = [M_{\text{BH}}(M_{\text{halo}})] (1 + z)^n$

- sigma:  $P(\log(M_{\text{BH}})) \propto e^{(\log(M_{\text{BH}}) - \log(M_{\text{BH},0}))^2 / 2\sigma^2}$

- Calculate the likelihood of the data given the model parameters, thus establishing confidence regions.

# Tests and Results



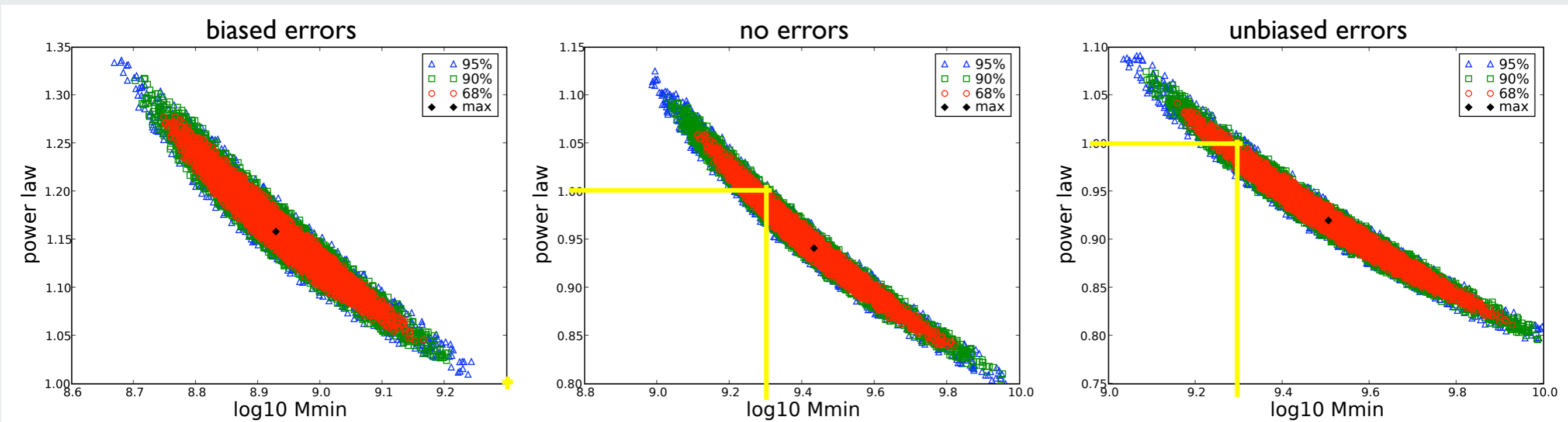
169 merger events (3 yrs observation)

Actual values:  $\log_{10}(M_{\min}) = 8.5$     power law = 1.5  
 $z\text{-dependence} = 0.5$     sigma = 0.2

# Tests and Results

- But what if measurements aren't perfect?
- We simulated the effects of observational uncertainty by adding errors to the synthetic data.
  - Biased errors - uniformly distributed between 0% and 40%
  - Unbiased errors - uniformly distributed between -20% and +20%
- Results show that observational uncertainty does not significantly affect our method's ability to constrain merger parameters.

# Tests and Results

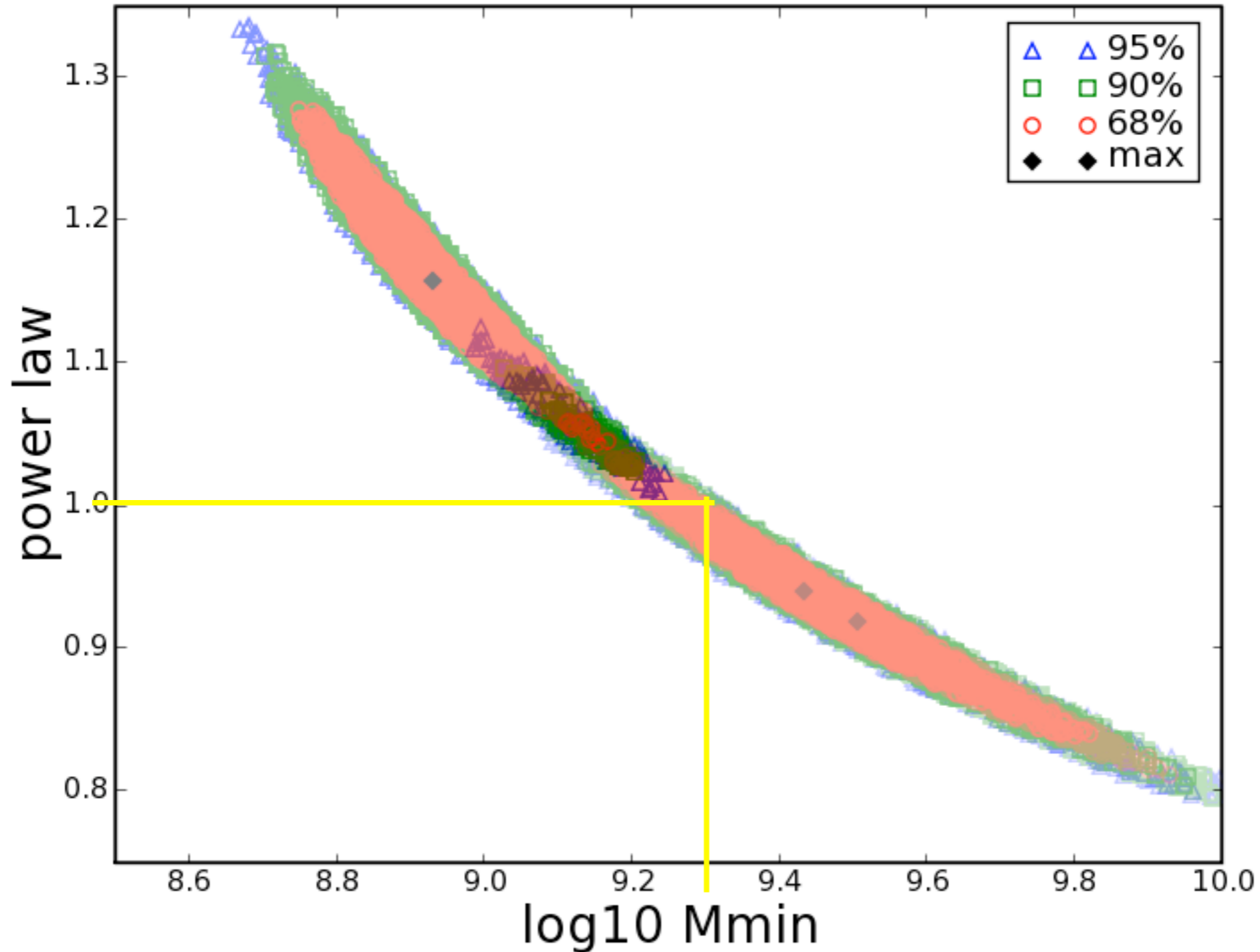


245 merger events (3 yrs observation)

Actual values:  $\log_{10}(M_{\min}) = 9.3$   
power law = 1.0

# Tests and Results

biased errors                      no errors                      unbiased errors



# Conclusions

- Results show that our proof of principle works.
  - Note that the method works for other kinds of observational data, not just gravitational waves.
- Many aspects of our analysis are robust against observational errors.
- To do: include more physically-driven parameters.