

Title: Sharpening the Precision of SZ Studies

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Abstract:

Sharpening the Precision of SZ Studies

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Outline

- Understanding the impact of scatter in relation between Y and M
- Source masking in power spectrum studies
- Pink elephants and cosmological parameters

Work done by Laurie Shaw

Cartoon Cosmology: Counting Galaxy Clusters

$$\frac{d^2 N}{dz d \ln S} = \int \frac{dP(\ln S | \ln M, z)}{d \ln S} \frac{dn}{d \ln M} d \ln M \frac{dV}{dz}$$

↑
*Observed
distribution
as function
of flux and
redshift*

↑
PHYSICS!!!!
↑
*Mapping
between
cluster
observable
and theory
mass (e.g.,
Gaussian+)*

↑
*Mass
function*

↑
*Volume
element*

Understanding scatter

$$\frac{dN}{d \ln S_{obs}} = \int \frac{dN}{d \ln S_{true}} P(\ln S_{obs} | \ln S_{true})$$

Define $G(x)$: Gaussian $\exp(-x^2/2)$ $x \equiv \frac{\ln S_{obs} - \ln S_{true}}{\sigma}$

Use Edgeworth expansion:

$$P(\ln S_{obs} | \ln S_{true}) \sim G(x) - \frac{\gamma \sigma^3}{6} \frac{d^3 G}{dx^3} + \frac{\kappa \sigma^4}{24} \frac{d^4 G}{dx^4} + \frac{\gamma^2 \sigma^6}{72} \frac{d^6 G}{dx^6} +$$

skewness

kurtosis

Understanding scatter

$$\frac{dN}{d \ln S_{obs}} = \int \frac{dN}{d \ln S_{true}} P(\ln S_{obs} | \ln S_{true})$$

Assume power-law counts $dN/d \ln S \propto S^{-\alpha}$

$$\frac{dN}{d \ln S_{obs}} = \left(\frac{dN}{d \ln S} \right)_o e^{\alpha^2 \sigma^2 / 2} \left[1 - \frac{\alpha^3 \sigma^3}{6} \gamma + \frac{\alpha^4 \sigma^4}{24} \kappa + \frac{\alpha^6 \sigma^6}{72} \gamma^2 + \dots \right]$$

Understanding scatter

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No Signal

VGA-1

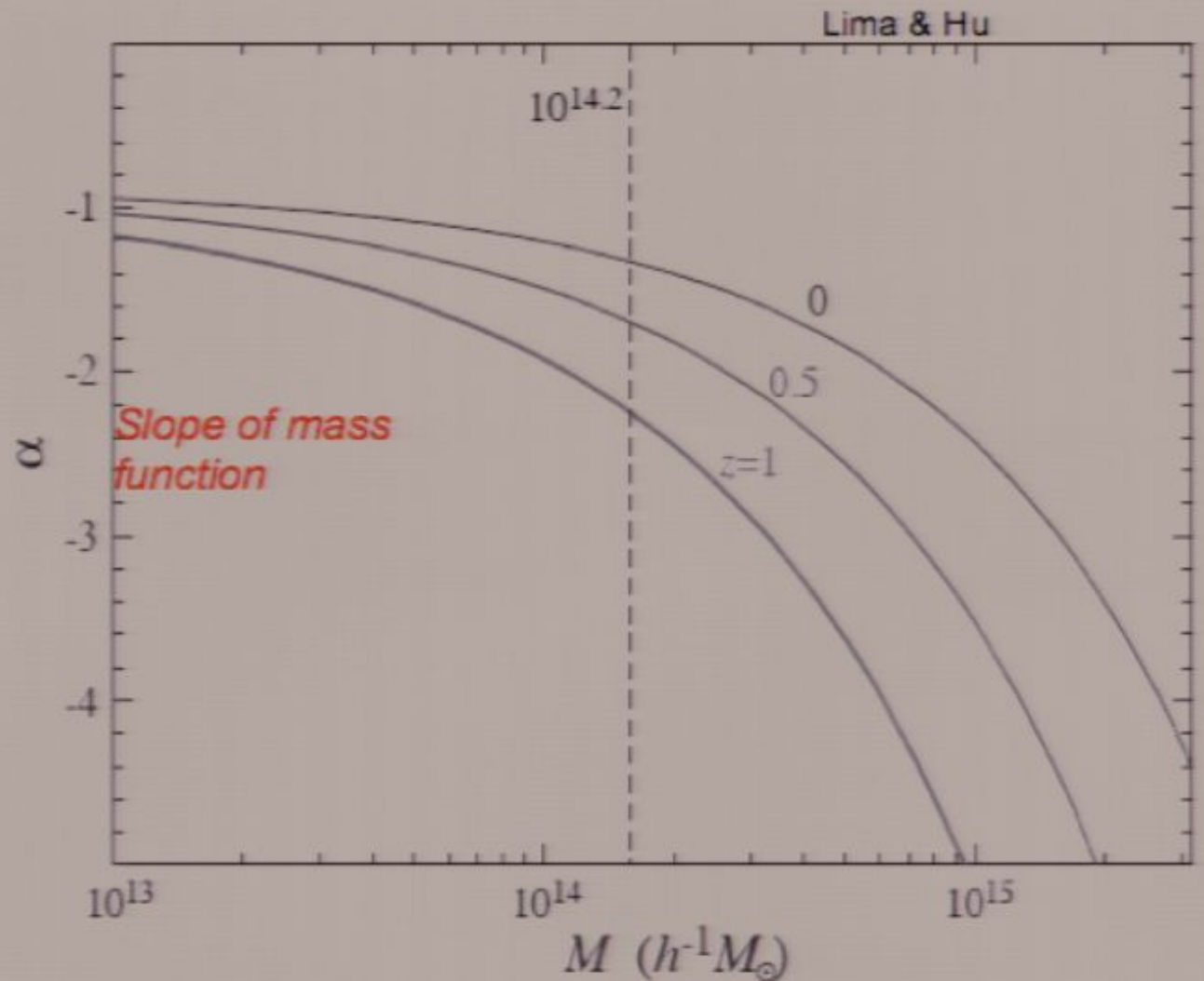
Understanding scatter

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The Importance of being precise



$$\frac{dN}{d \ln S_{obs}} = \left(\frac{dN}{d \ln S} \right)_o e^{\alpha^2 \sigma^2 / 2} \left[1 - \frac{\alpha^3 \sigma^3}{6} \gamma + \frac{\alpha^4 \sigma^4}{24} \kappa + \frac{\alpha^6 \sigma^6}{72} \gamma^2 + \dots \right]$$

Calculating the SZ Power Spectrum

Power spectrum

$$C_\ell = \int dz \frac{dV}{dz} \int d \ln M \frac{dn(M, z)}{d \ln M} \bar{y}(M, z, \ell)^2$$

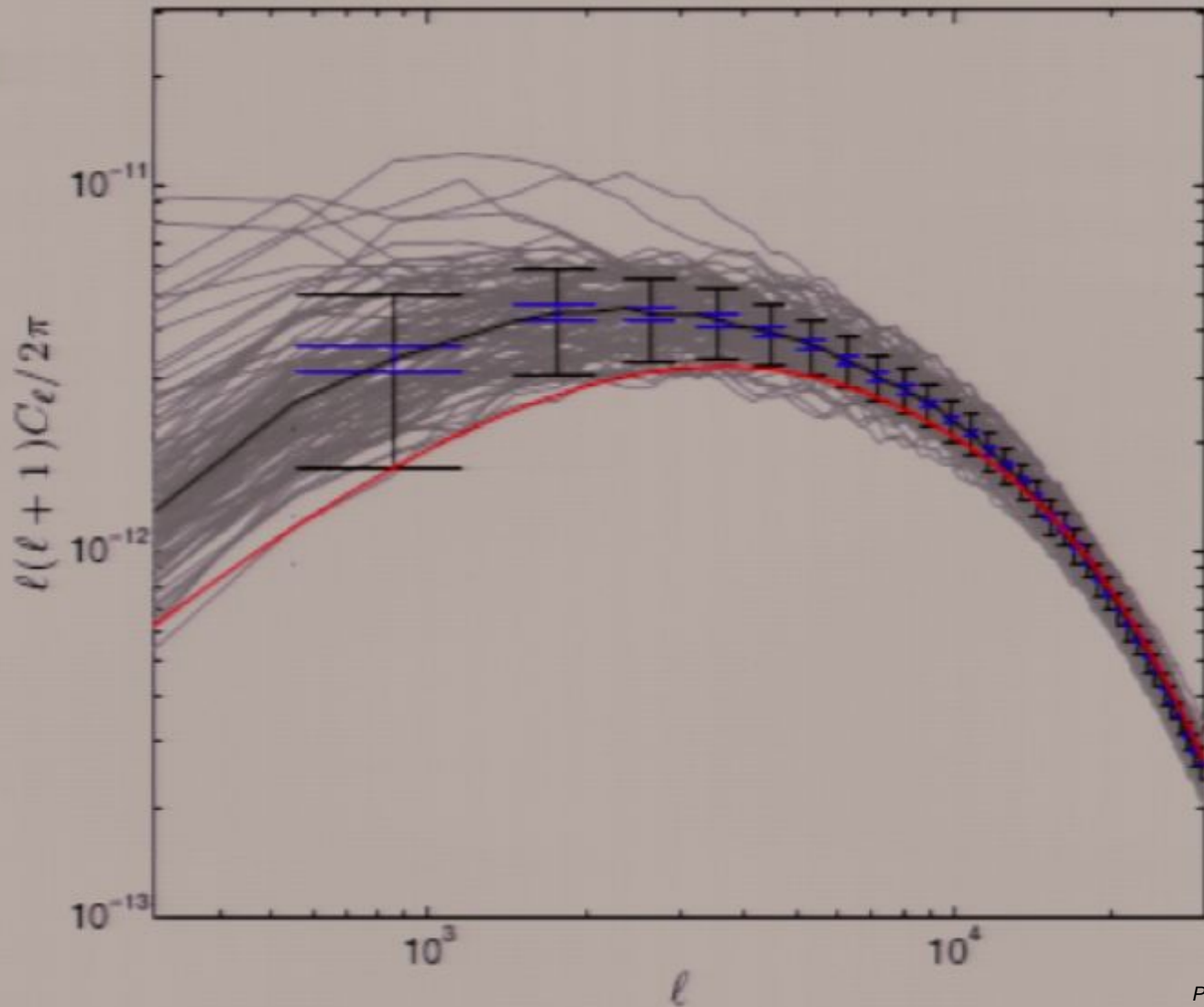
1 halo part of trispectrum

$$T_{\ell\ell\ell} = \int dz \frac{dV}{dz} \int d \ln M \frac{dn}{d \ln M} \bar{y}(M, z, \ell)^4$$

$$\sigma^2(C_\ell) = f_{sky}^{-1} \left[\frac{2C_\ell^2}{(2\ell + 1)\Delta\ell} + \frac{T_{\ell\ell\ell}}{4\pi} \right]$$

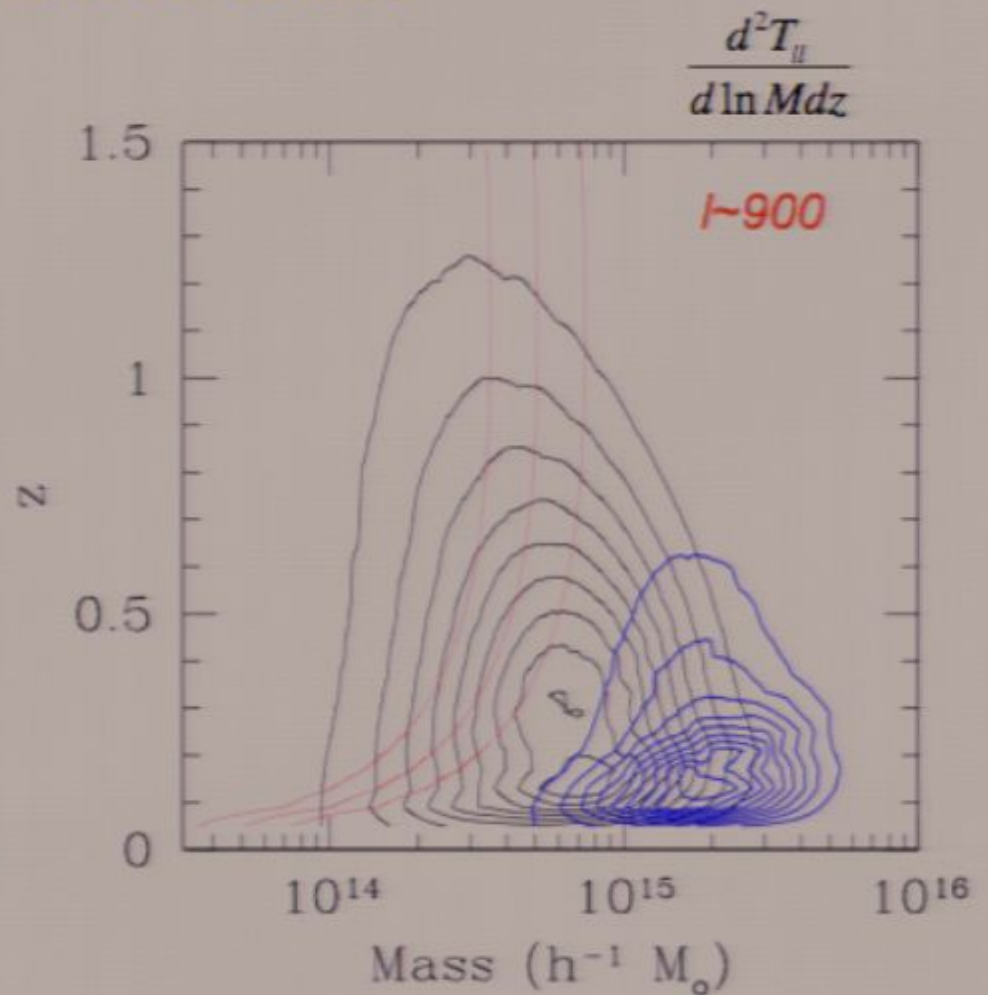
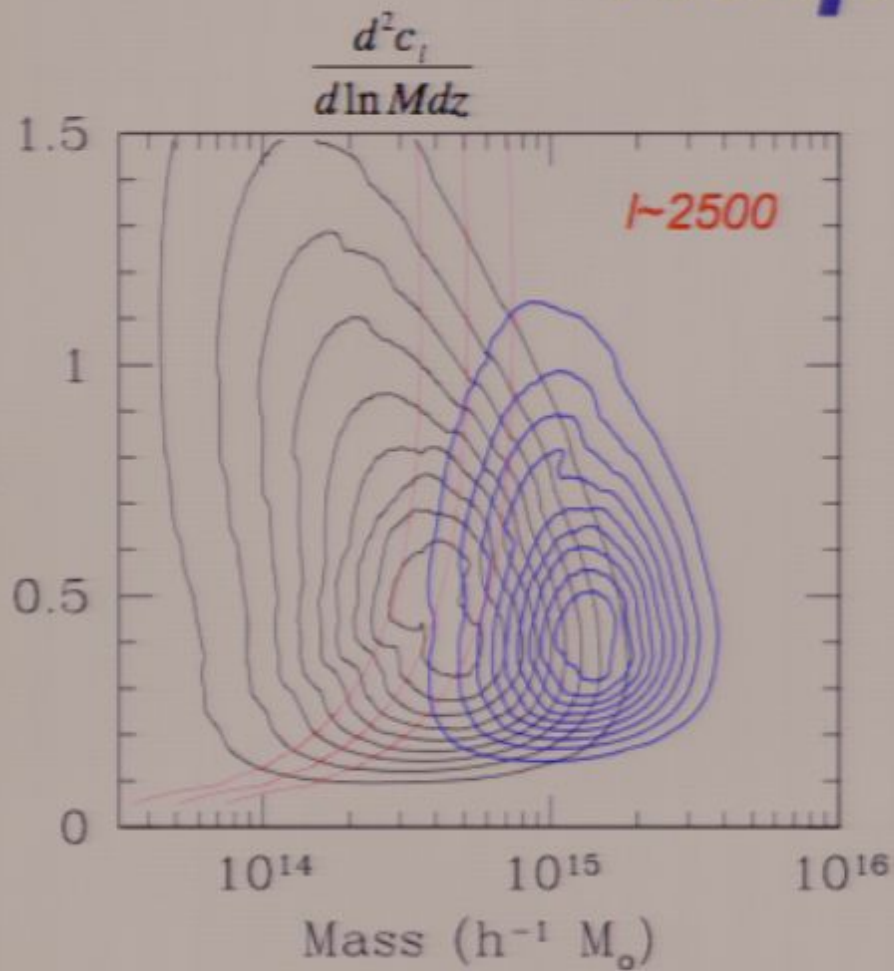
Variance in the SZ Power

Power spectra
for 96 maps,
each 5x5deg
based on
Bode/Ostriker
clusters



Scatter much
larger than
sample
variance

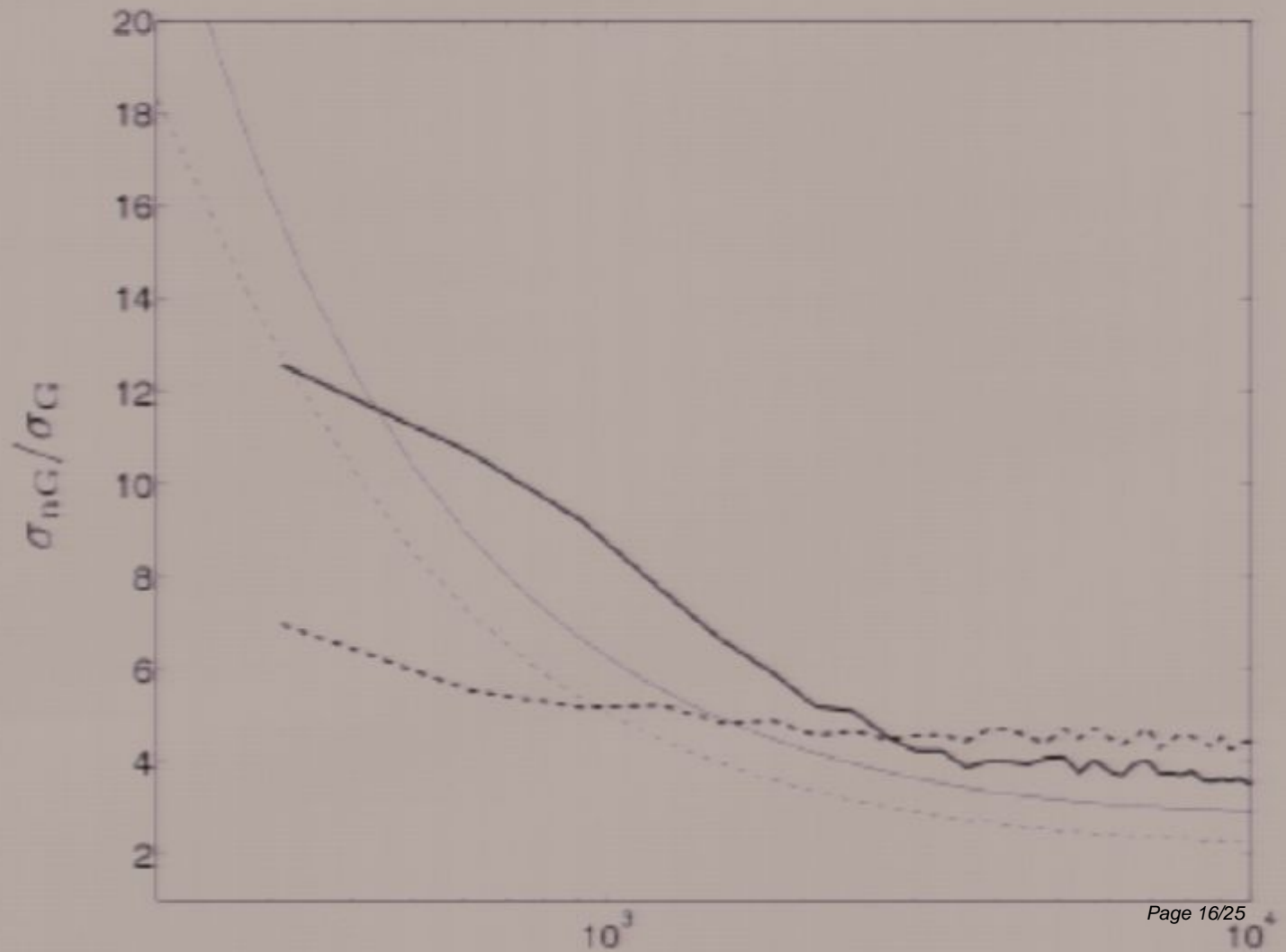
Contributions to Power and Trispectrum



Comparison with Gaussian expectations

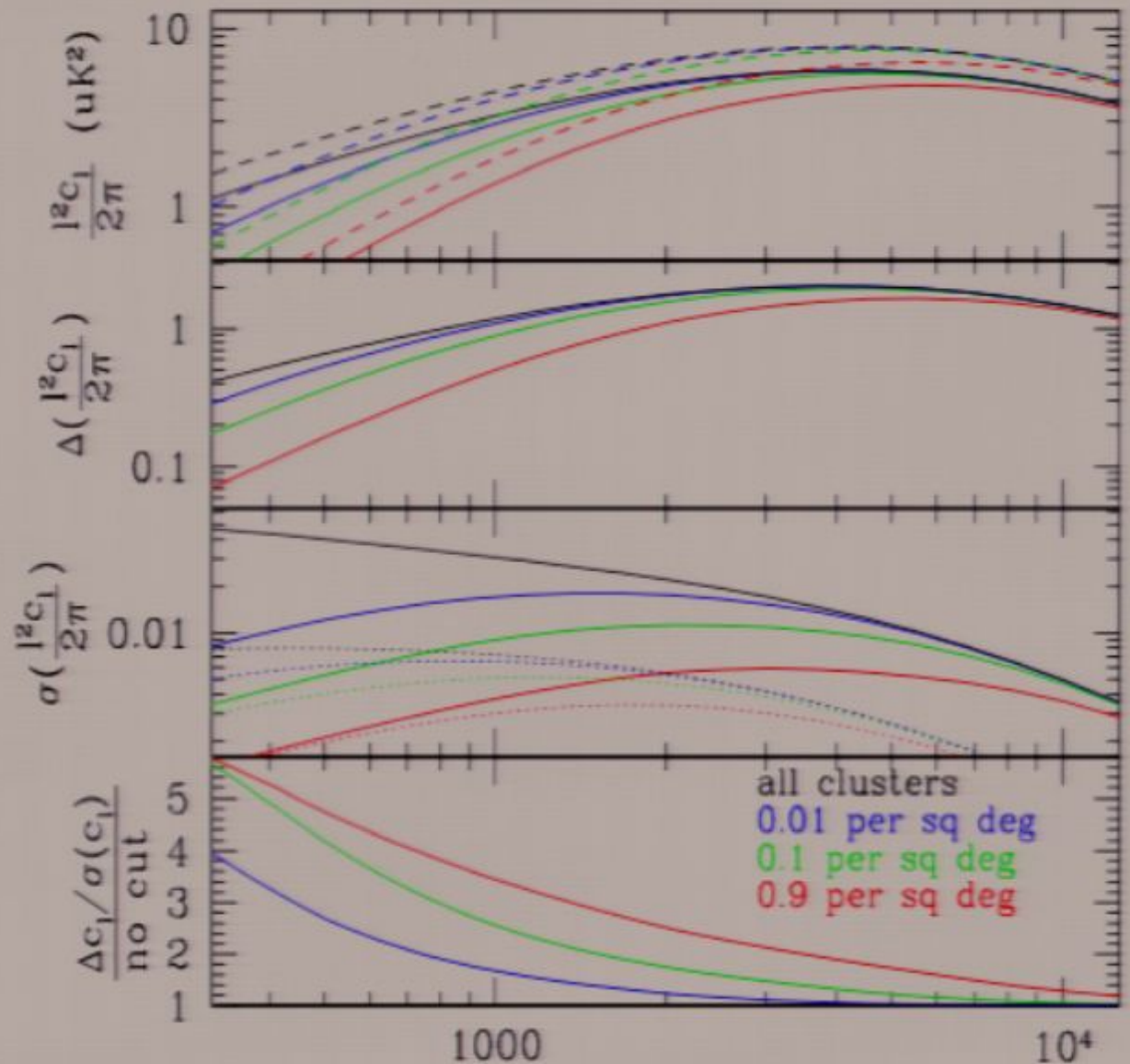
Reasonable,
but not perfect
match with
simple model

(but, good agreement for
Poisson-distributed beta
model profiles; not
shown)



Advantage of Masking Bright Sources

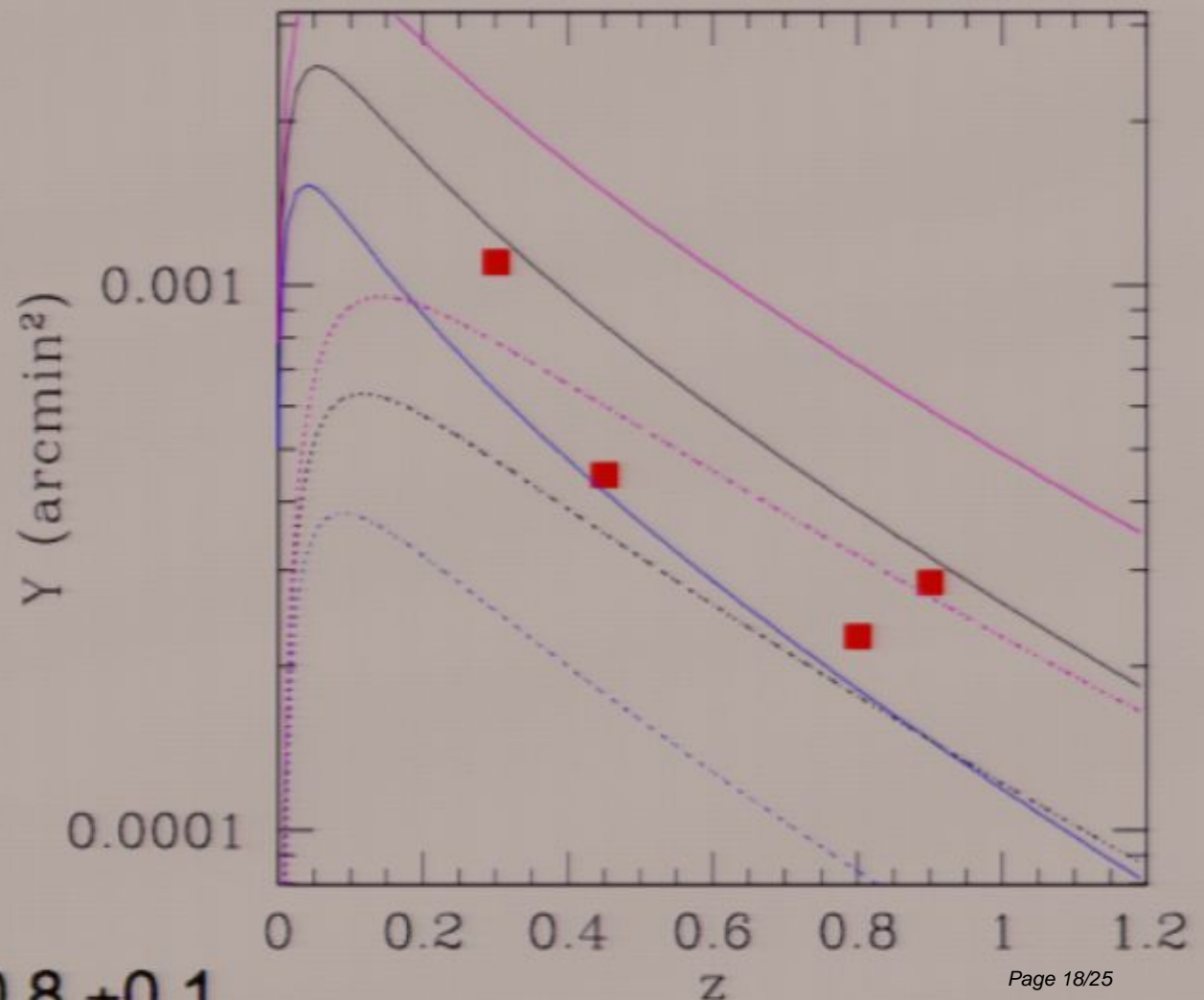
- Shot noise in massive objects add much more noise than signal!



Pink & Grey Elephants

Contours of differential number density

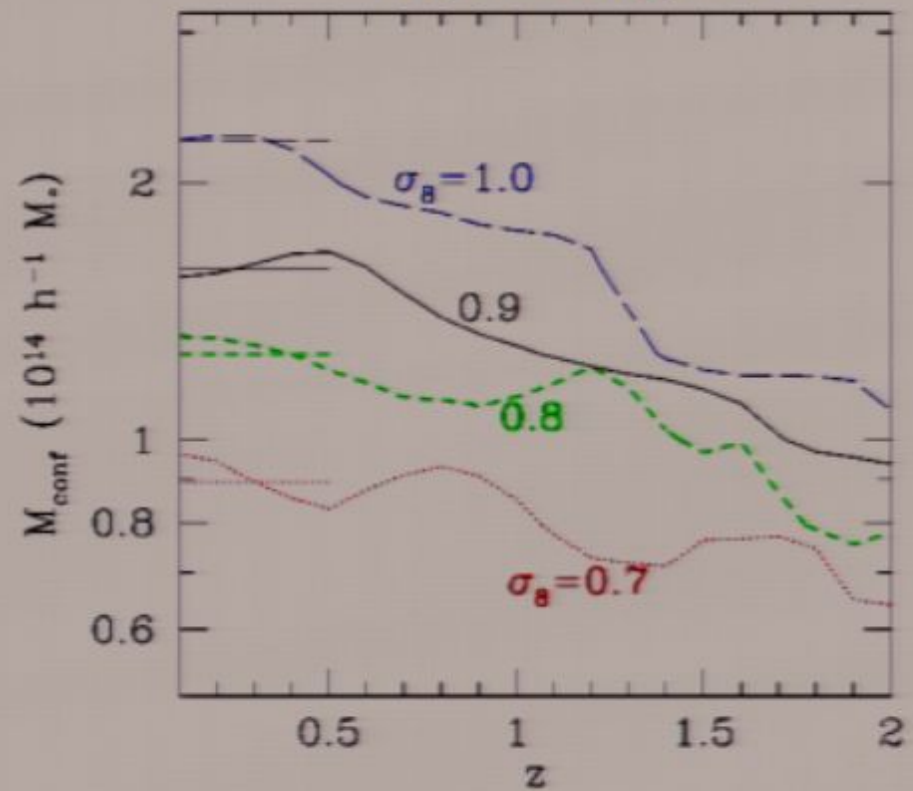
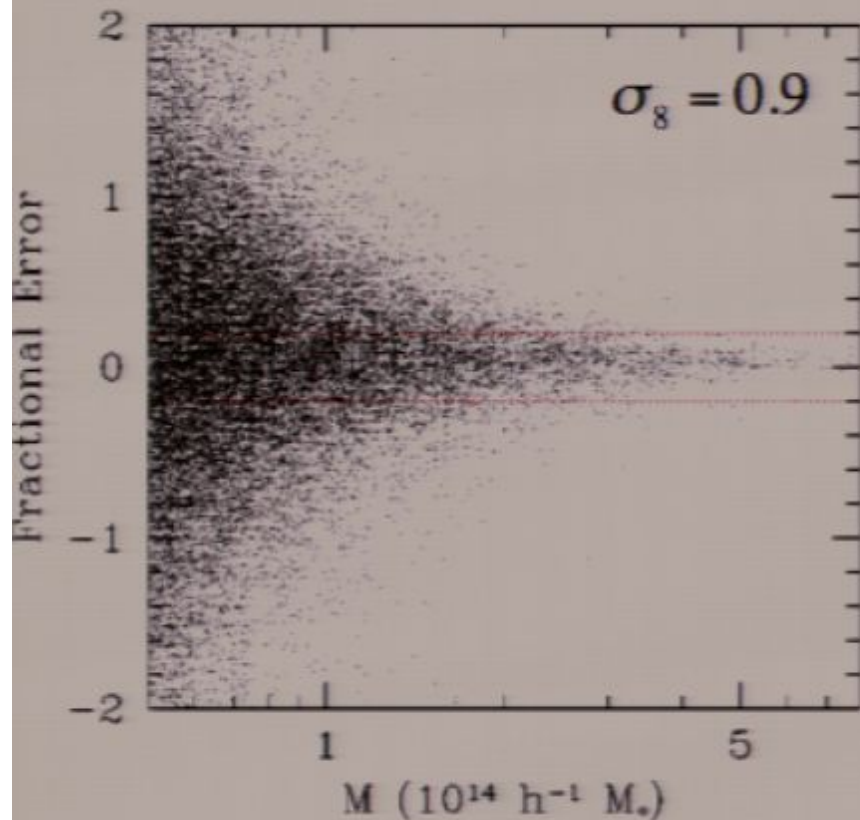
- $\sigma_8=0.9, 0.8, 0.7$ (purple, black, blue)
- Solid: 0.5 clusters per 40 sq deg for $dz=d\log Y=0.2$
- Dotted: 2 clusters per sq deg
- Red squares: the four public SPT clusters
- Too many high flux clusters for $\sigma_8=0.7$
- Where are all the massive clusters for $\sigma_8=0.9$?



with only 40 sq deg: $\sigma_8 \sim 0.8 \pm 0.1$

SZ Confusion

GH, McCarthy, Babul



(input-measured)/input for simulated filtered SZ maps

Mass at which rms error is 20%

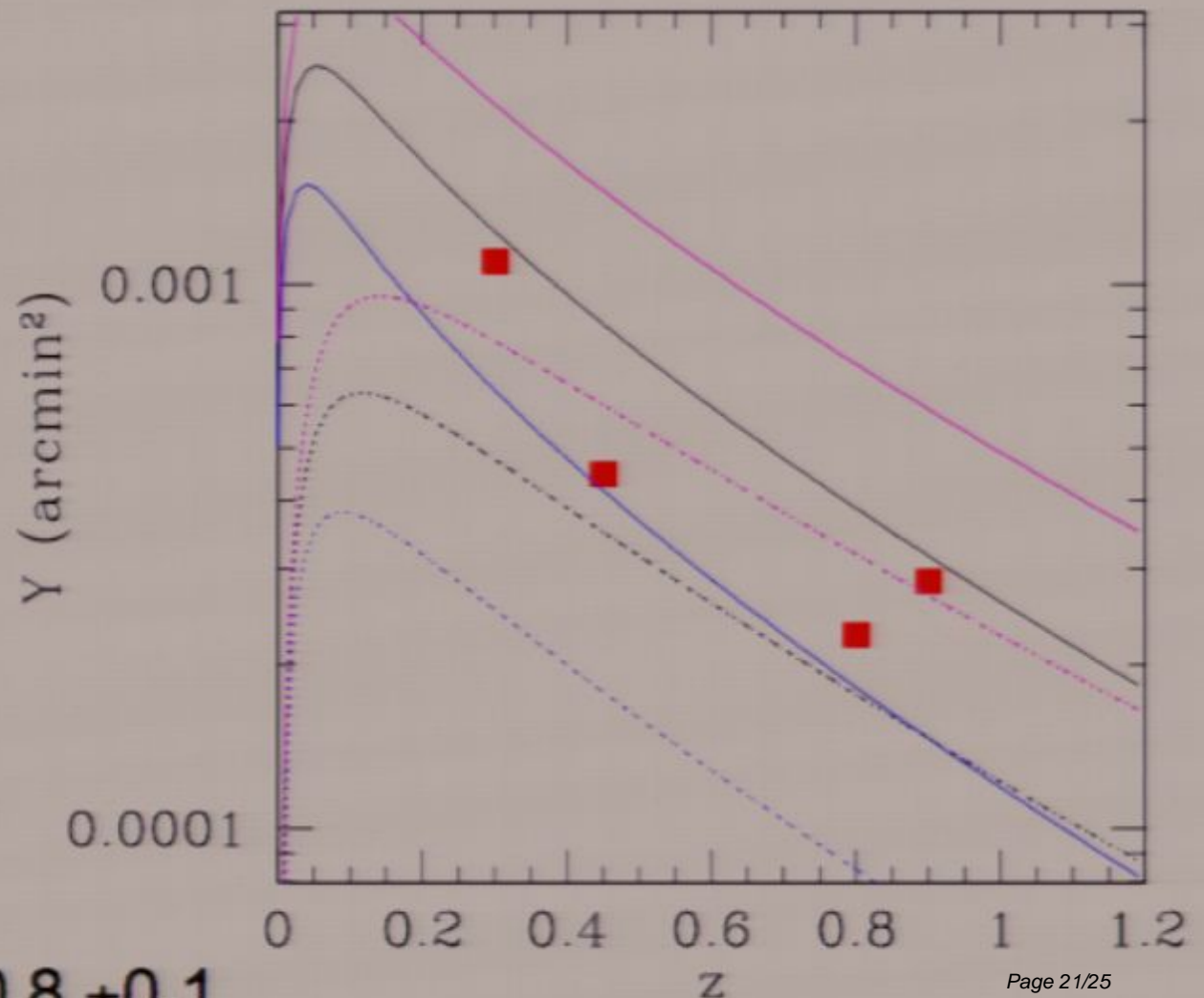
Summary

- Large scatter in cluster surveys is bad for two reasons
 - Increases number counts dramatically
 - Amplifies importance of tails of distribution
- Masking of galaxy clusters improves precision of power spectrum estimates
 - Rare objects have a lot of shot noise and not much extra information for the power spectrum
- SZ cosmology nearly here
 - Can nearly do interesting cosmology with only 4 clusters! *[although I wouldn't recommend it...]*

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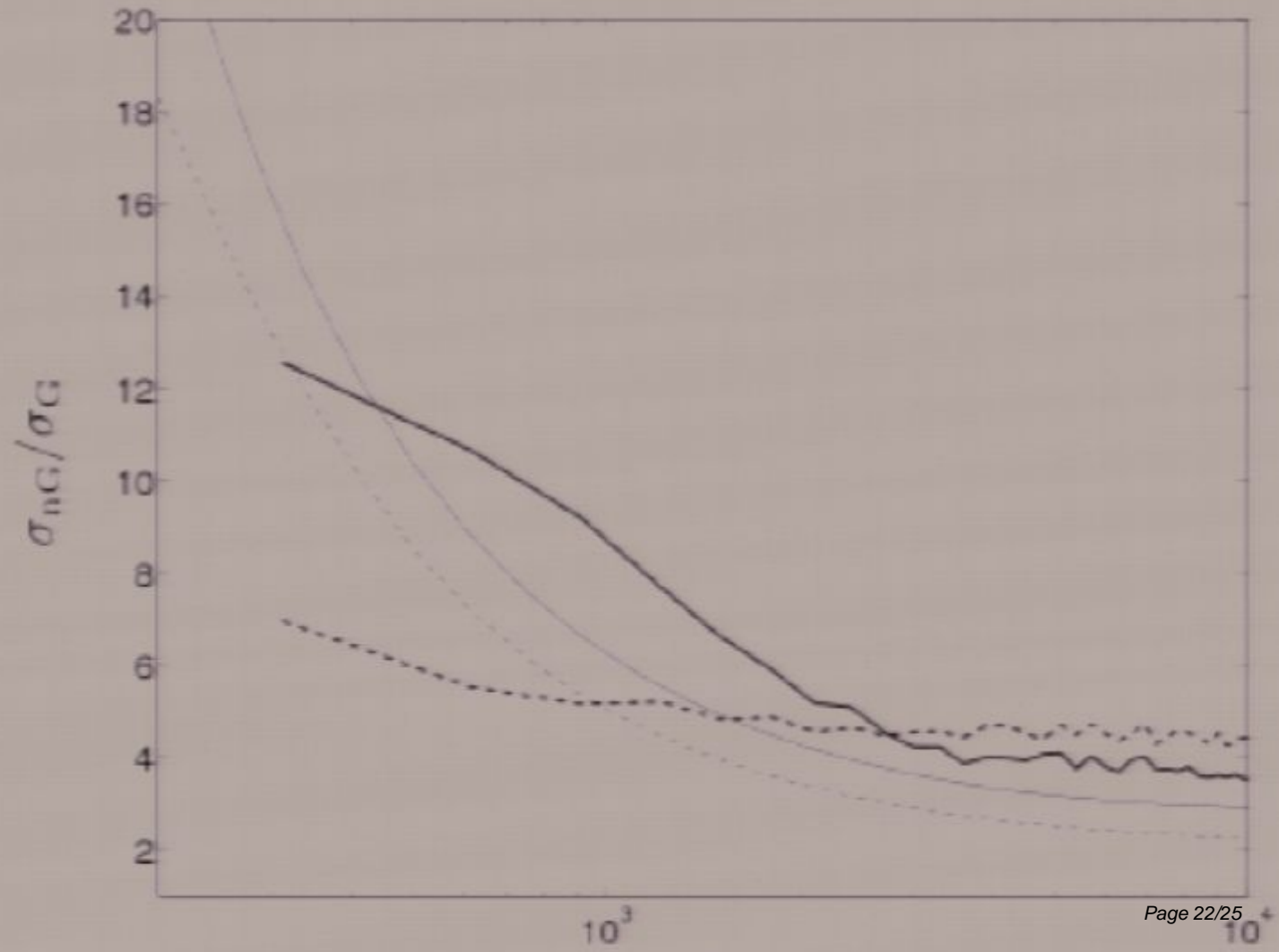


with only 40 sq deg: $\sigma_8 \sim 0.8 \pm 0.1$

Comparison with Gaussian expectations

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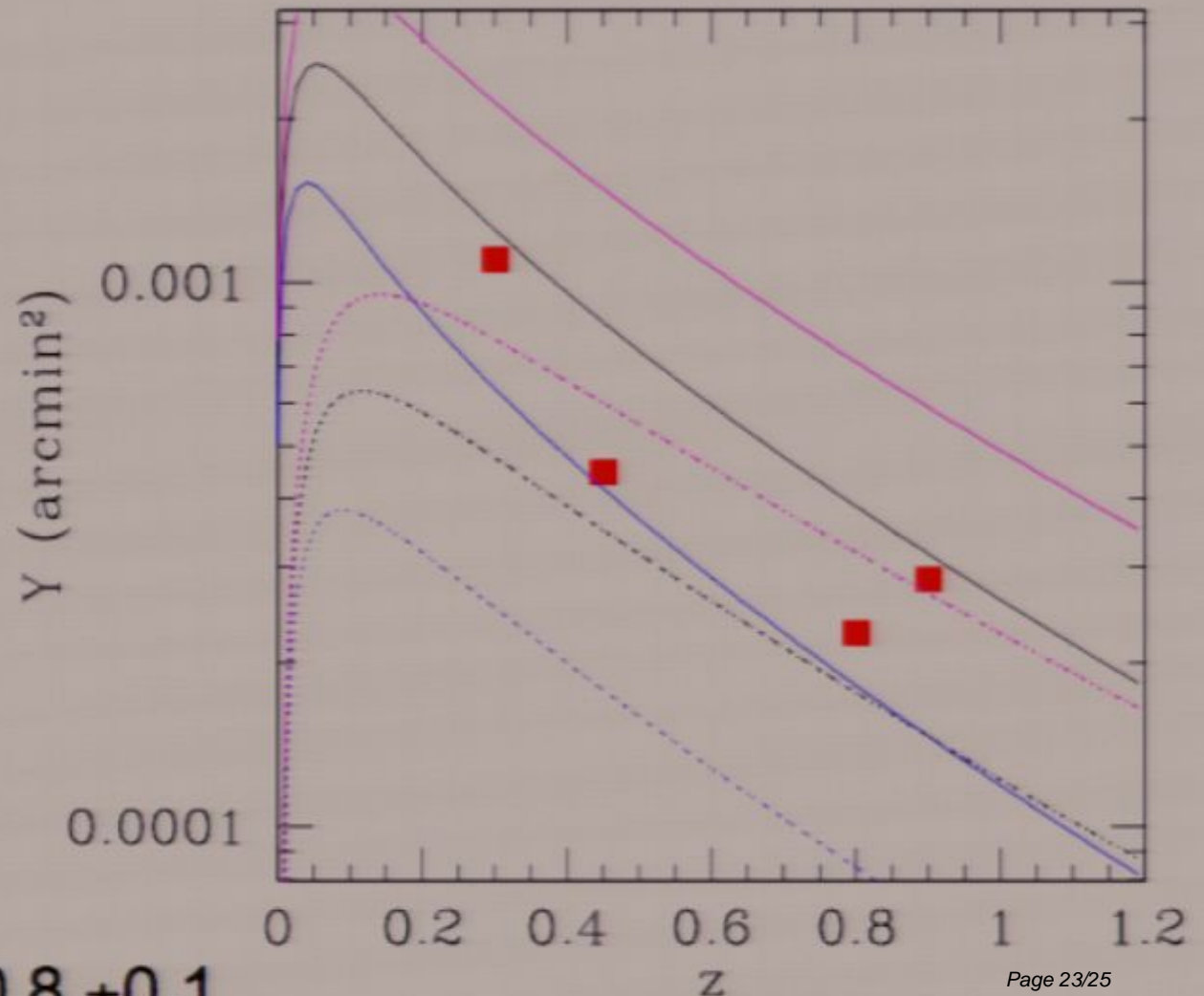
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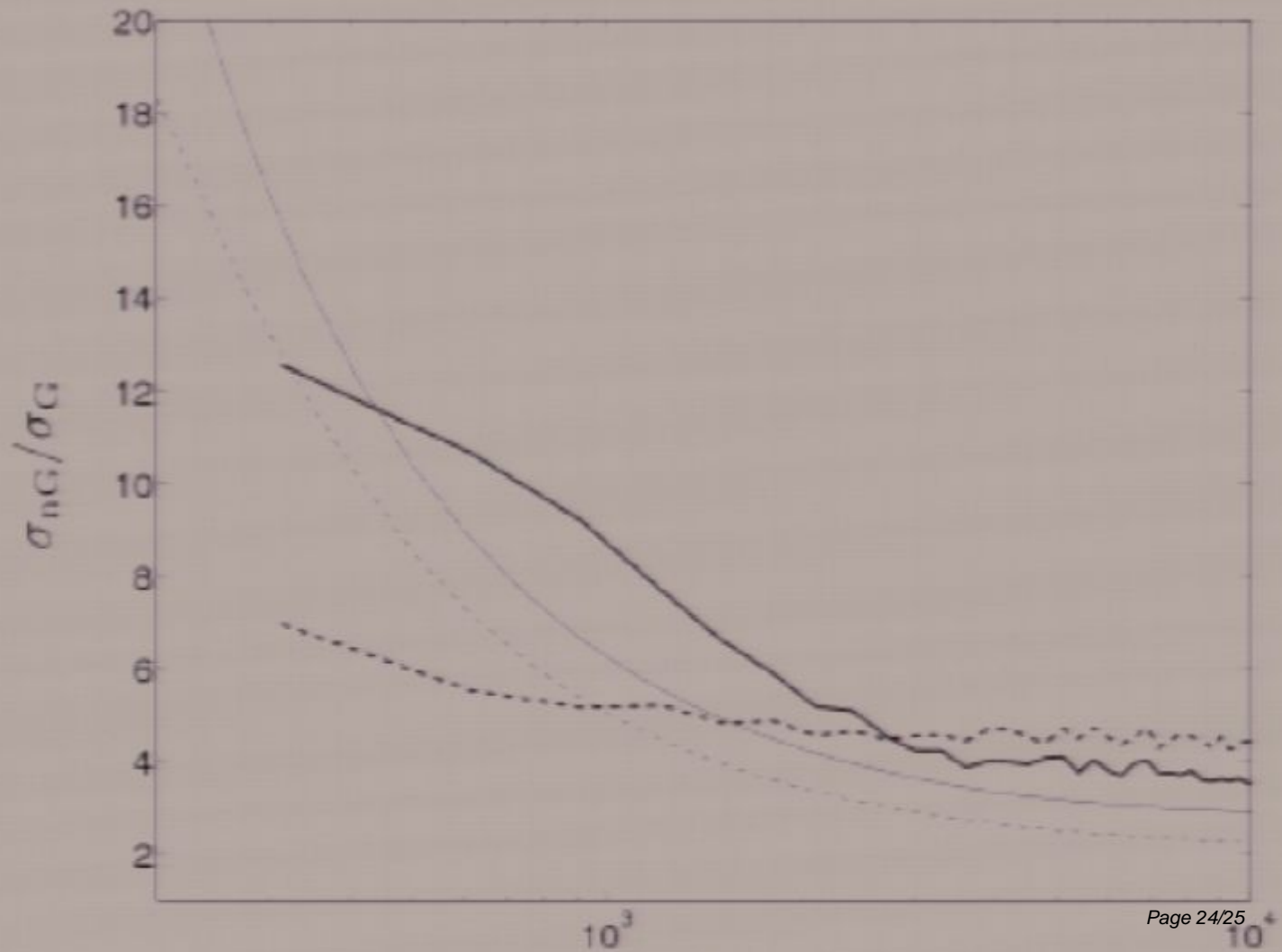


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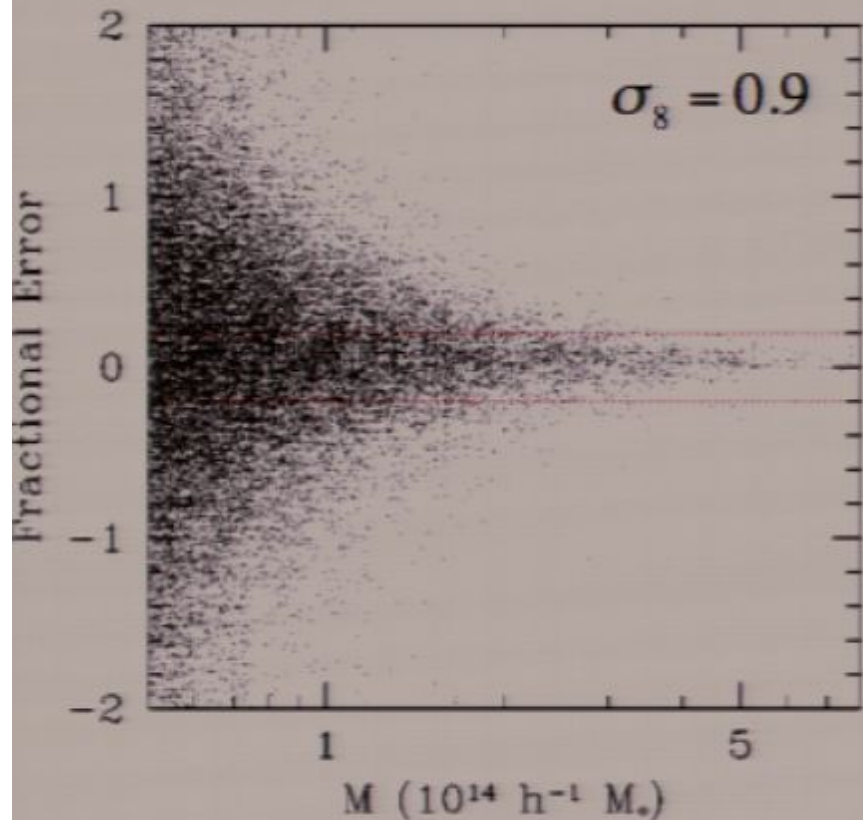
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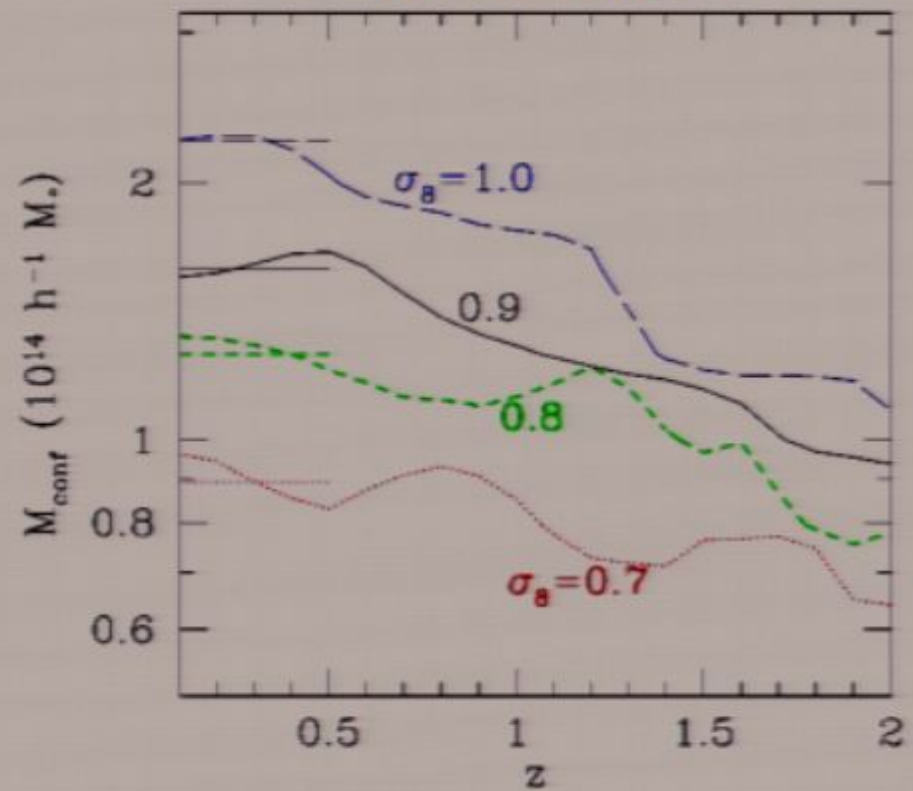
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Pirsa: 09050014

GH, McCarthy, Babul



Mass at which rms error is 20%

Assuming single frequency