

Title: Constraints on scale dependent non-Gaussianity

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URL: <http://pirsa.org/07100004>

Abstract: The detection of primordial non-Gaussianity could provide a powerful means to rule out various inflationary scenarios. Although scale-invariant non-Gaussianity is currently best constrained by the Cosmic Microwave Background, single-field inflation models with changing sound speed can have strongly scale dependent non-Gaussianity. I will discuss the theoretical motivation for such models and present work on the likely ability of current and future large scale structure measurements to constrain them.

SCALE-DEPENDENT NON-GAUSSIANITY

Sarah Shandera
Columbia University

Based on work with Marilena
LoVerde, Amber Miller, Licia Verde
(arXiv:0711:xxxx)

Is there a useful discriminating feature
of inflation models?

- or -

Can string theory suggest
cosmologically interesting ideas?

- or -

Could there be a *distinctive* signature
of string theory in inflation?

GOALS:

I. Why non-Gaussianity?

II. Why scale-dependence? (DBI, general sound speed)

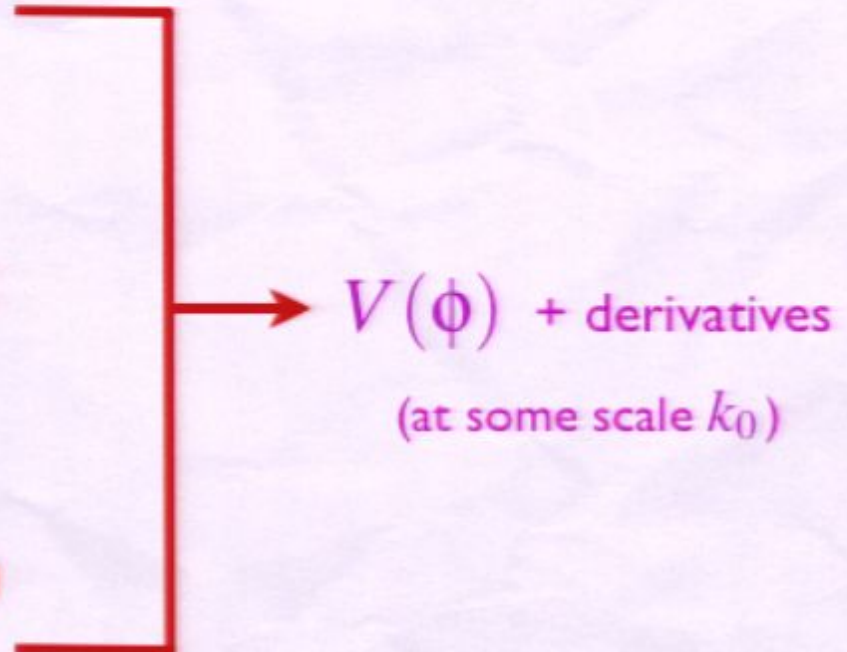
III. Observations: Galaxy Clusters

IV. Observations: Dark matter (galaxy) Bispectrum

I.WHY (AND WHAT) NON- GAUSSIANTY

USUAL INFLATIONARY OBSERVABLES

- COBE normalization
- Spectral index + running
- Tensor/scalar ratio
- Tensor index (+ running)



“RECONSTRUCTING” THE POTENTIAL DOES NOT GIVE A UNIQUE ANSWER (EASTHER, KINNEY, PEIRIS E.G.)

THE 3-POINT FUNCTION: MORE INFORMATION

- A simple ansatz (local model):

$$\zeta(x) = \zeta_g(x) + f_{NL} [\zeta_g^2(x) - \langle \zeta_g^2(x) \rangle]$$

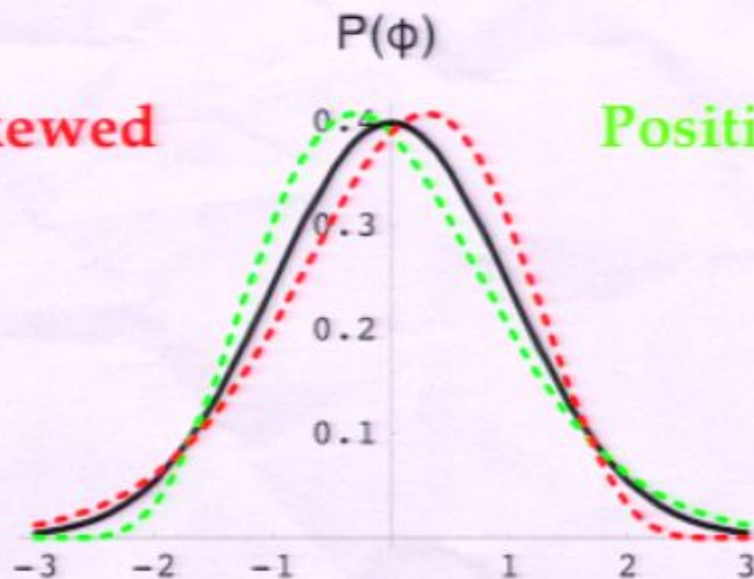
- Then in Fourier space:

$$\begin{aligned} \langle \zeta_{NG}(k_1) \zeta_{NG}(k_2) \rangle &= \langle \zeta_G(k_1) \zeta_G(k_2) \rangle + O(f_{NL}^2) \\ &\approx (2\pi)^3 \delta(k_1 + k_2) \frac{2\pi^2 \mathcal{P}^{\zeta_G}(k)}{k^3} \end{aligned}$$

$$\begin{aligned} \langle \zeta_{NG}(k_1) \zeta_{NG}(k_2) \zeta_{NG}(k_3) \rangle &= f_{NL} \frac{(2\pi)^7}{2} \delta^3(k_1 + k_2 + k_3) \left(\frac{\mathcal{P}^{\zeta_G}(k_1) \mathcal{P}^{\zeta_G}(k_2)}{k_1^3 k_2^3} + \text{perm.} \right) \\ &\quad + O(f_{NL}^3) \end{aligned}$$

PHYSICALLY (LOCAL MODEL)...

Negatively skewed



Positively skewed

$$f_{NL} = \pm 0.1$$

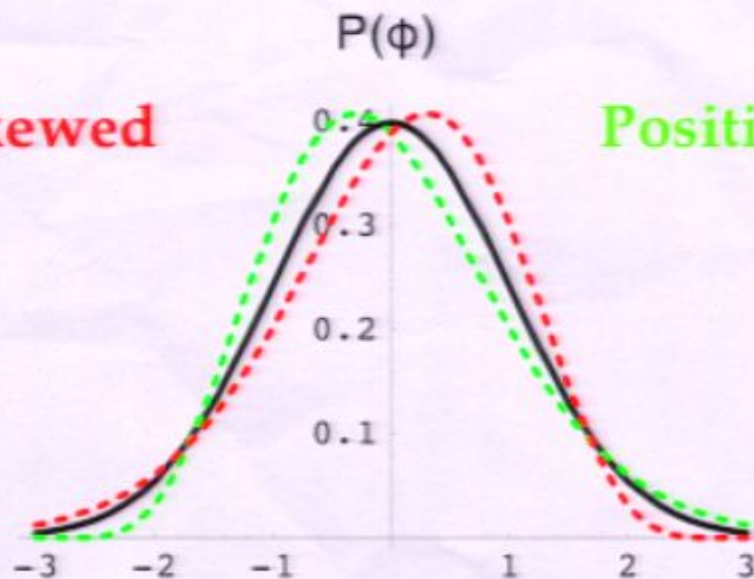
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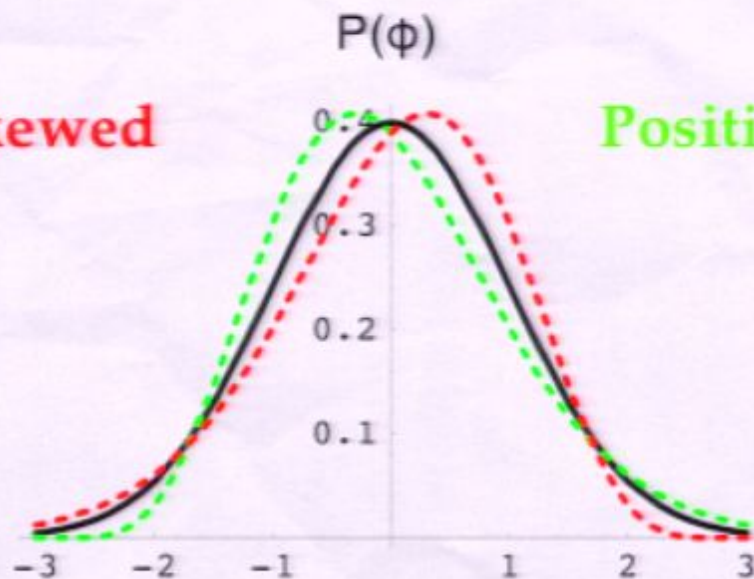
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$$X = -\frac{1}{2} g^{\mu\nu} \partial_\mu \phi \partial_\nu \phi$$

$$3M_p^2 H^2 = E$$

$$E = 2XP_{,X} - P$$

$$c_s^2 = \frac{P_{,X}}{E_{,X}} = \frac{P_{,X}}{P_{,X} + 2XP_{,XX}}$$

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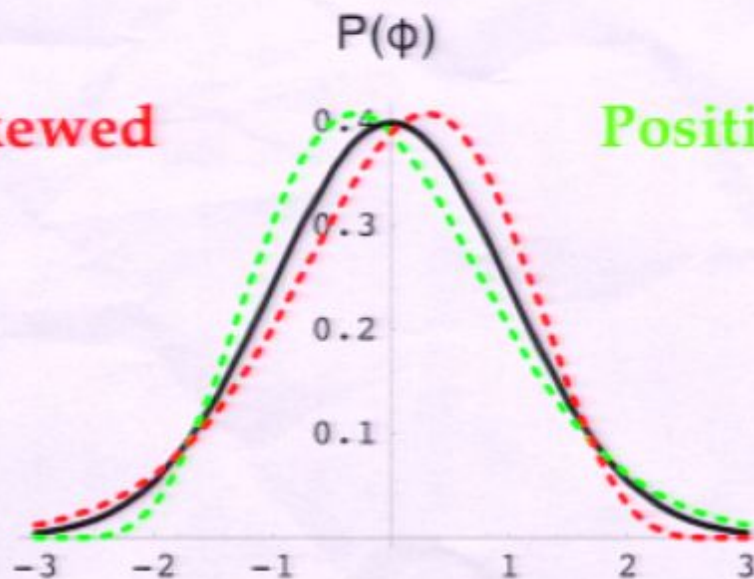
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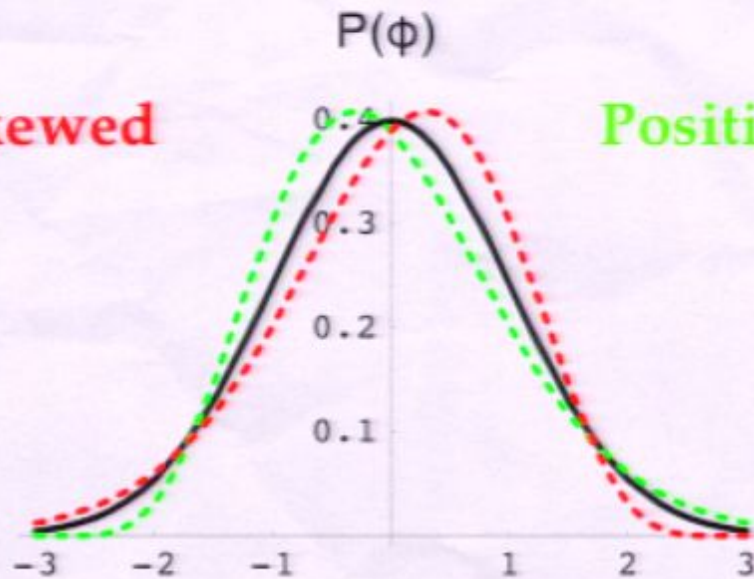
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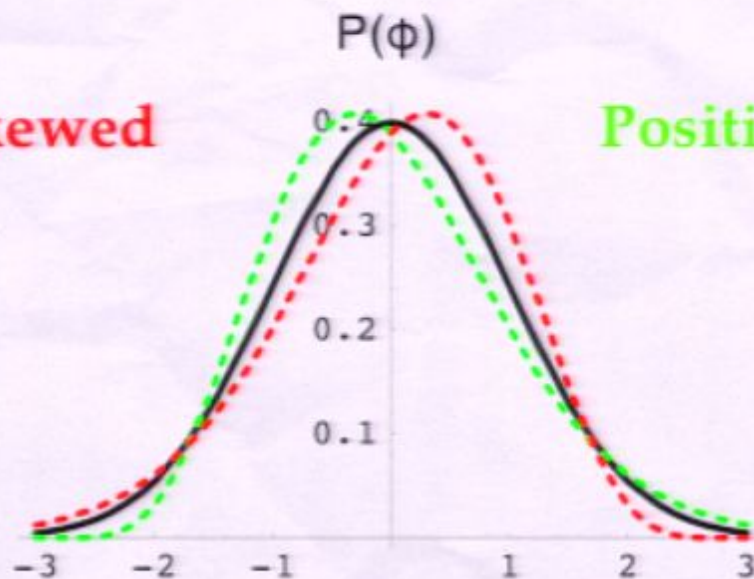
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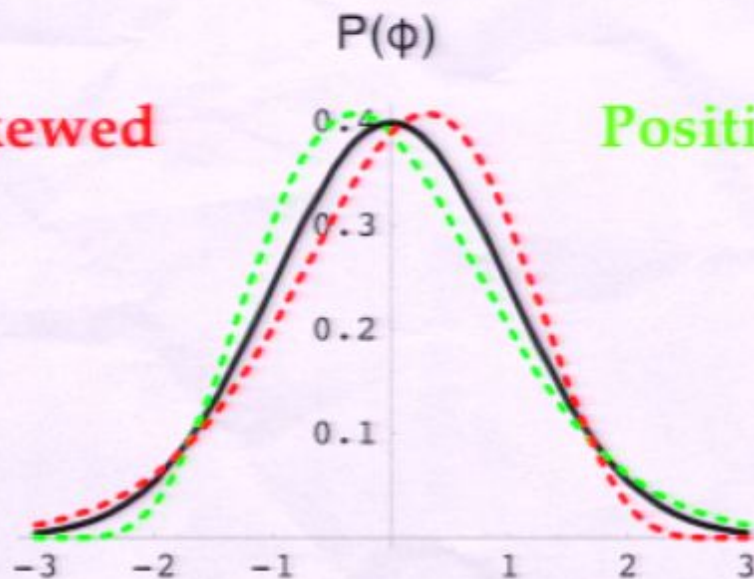
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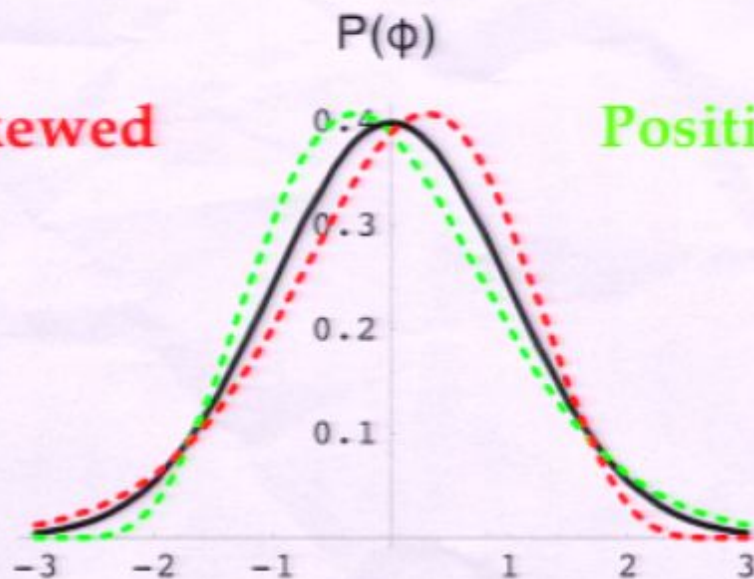
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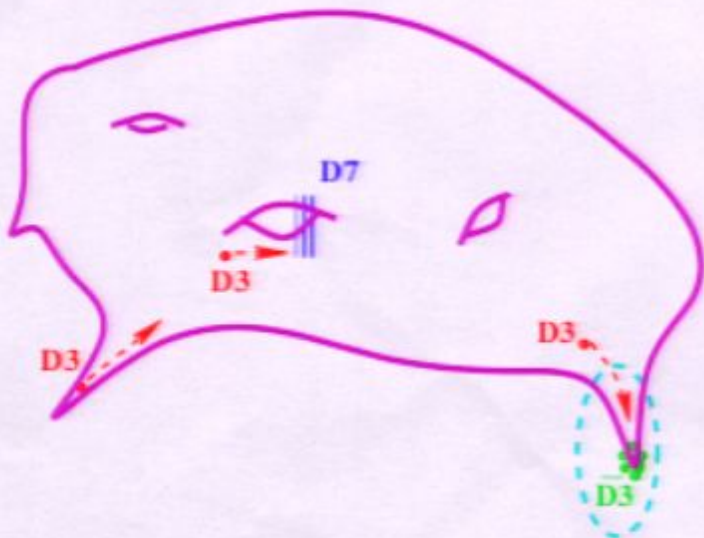
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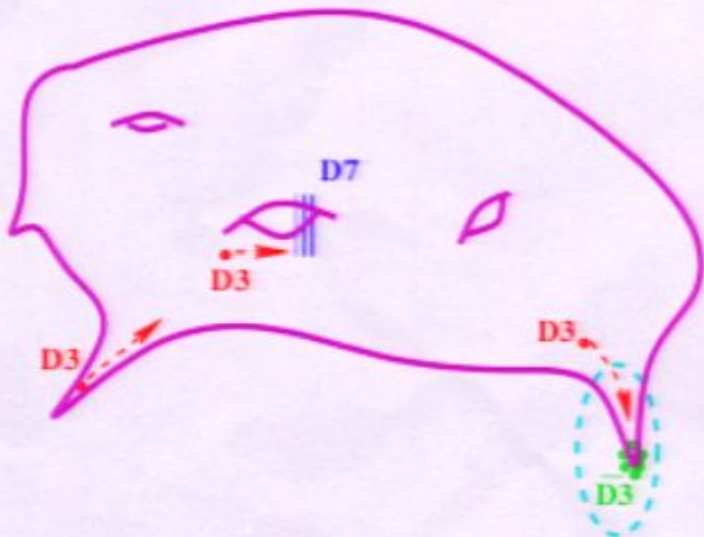
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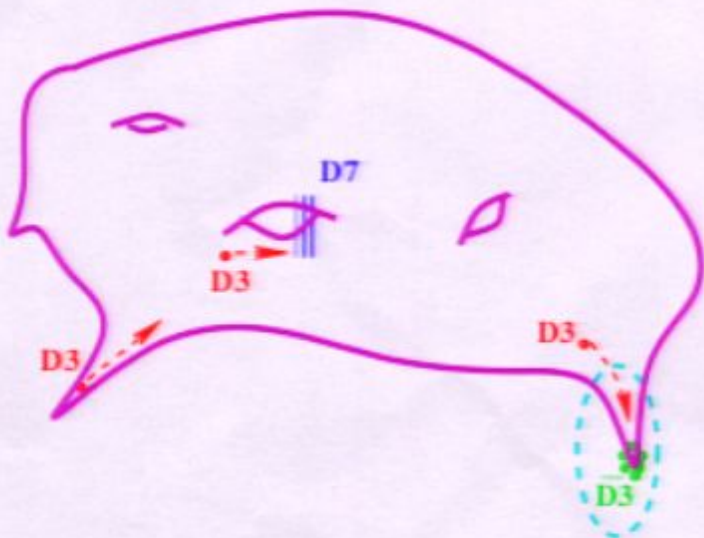
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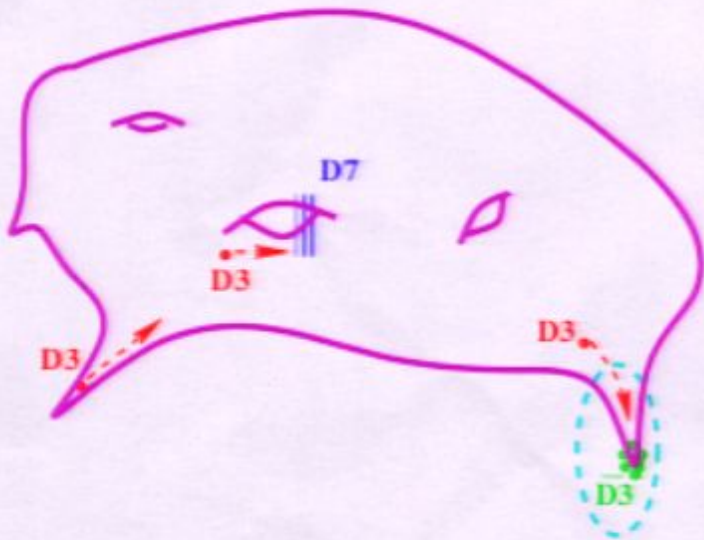
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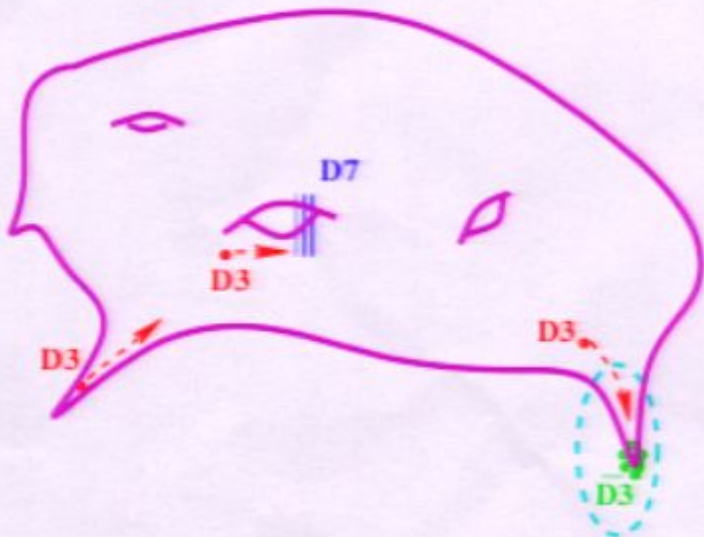
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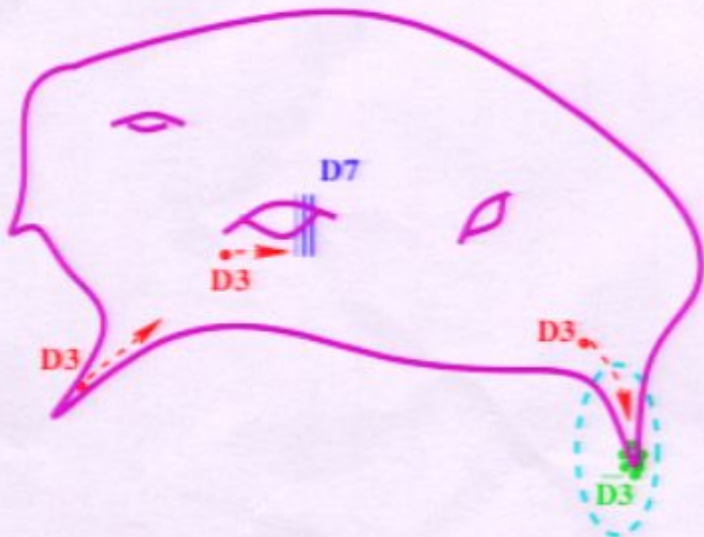
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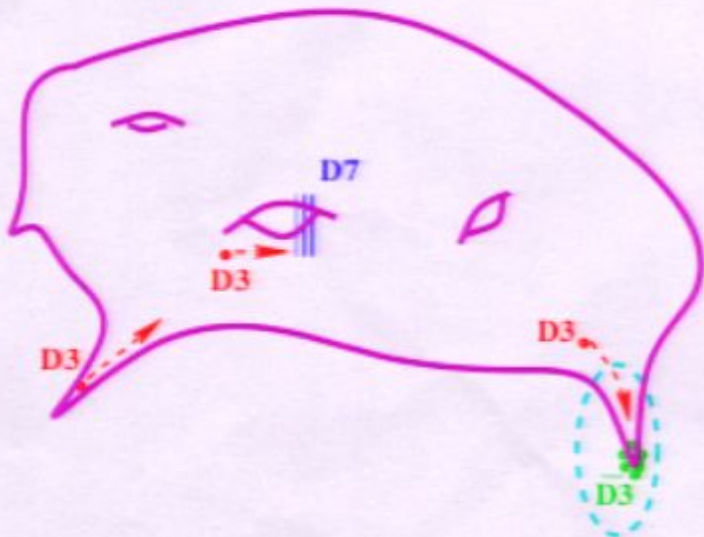
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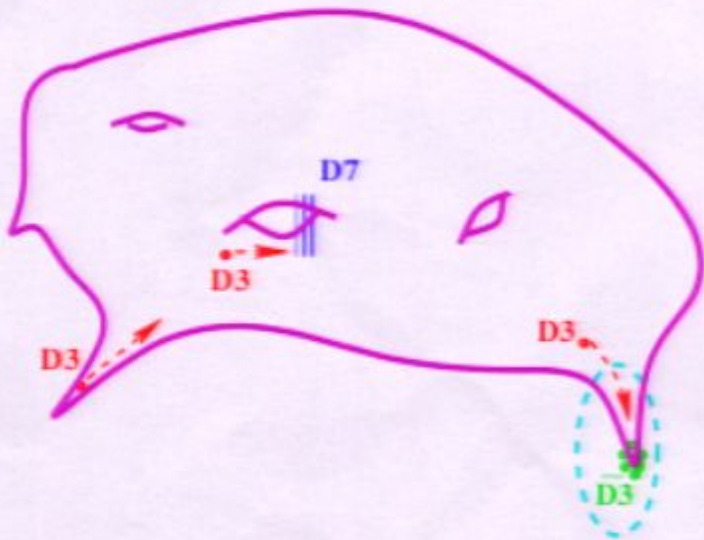
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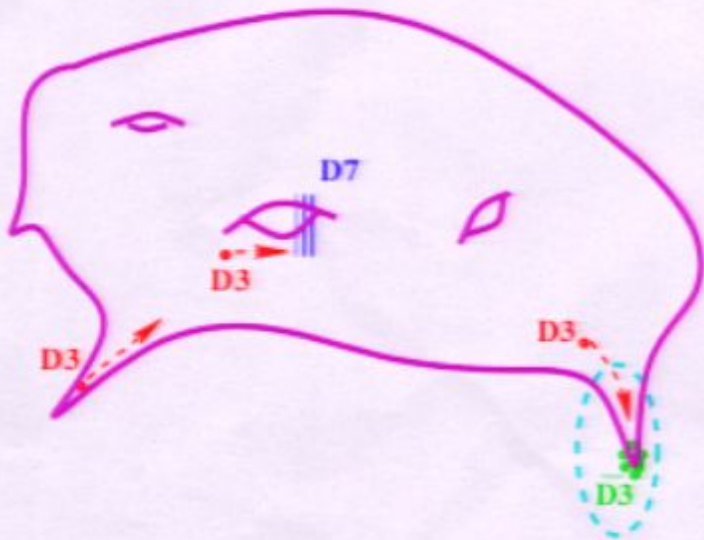
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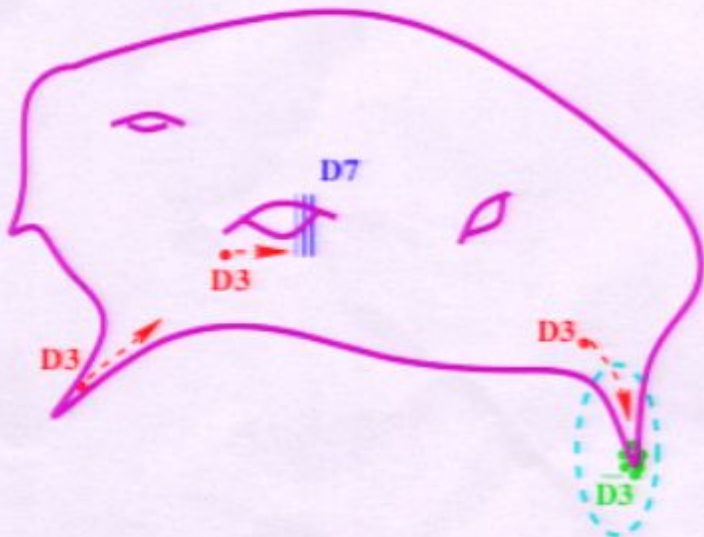
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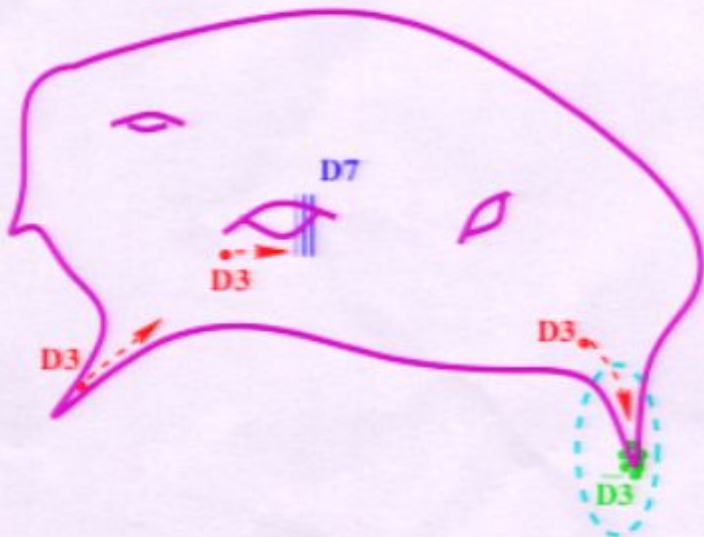
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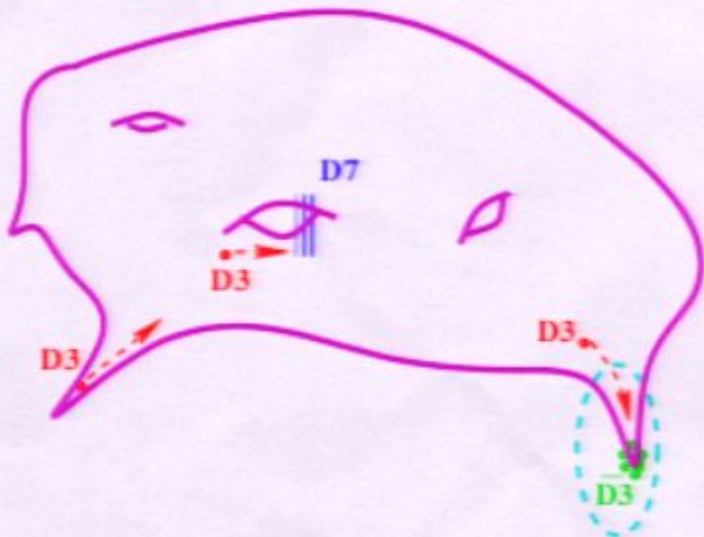
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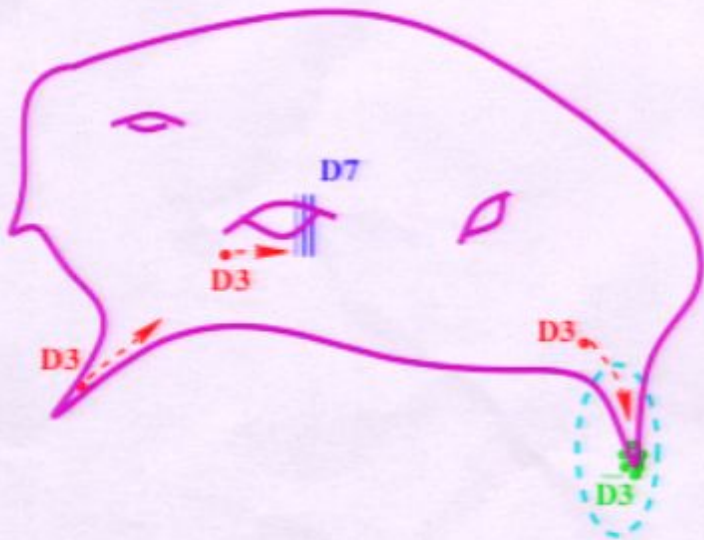
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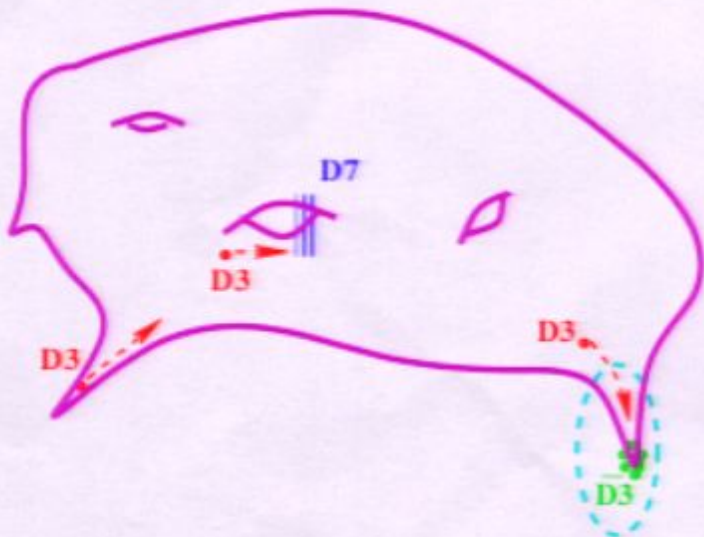
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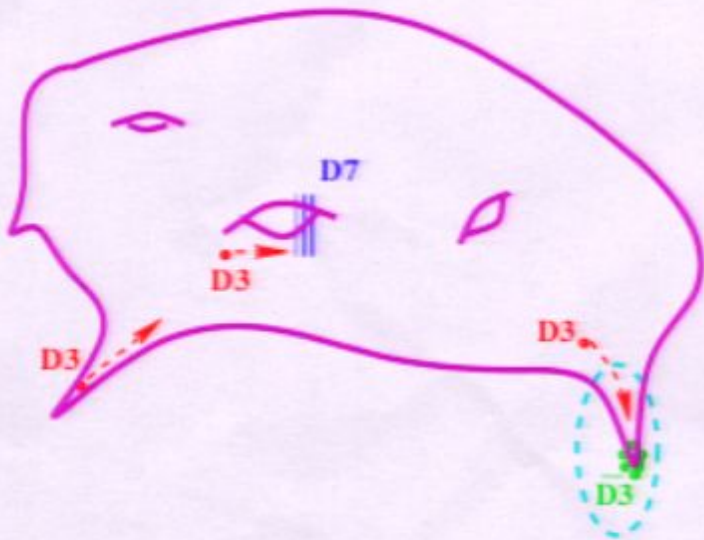
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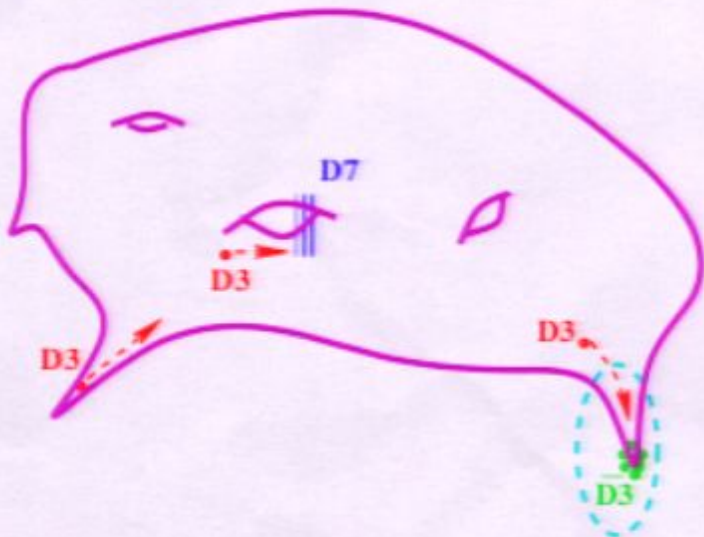
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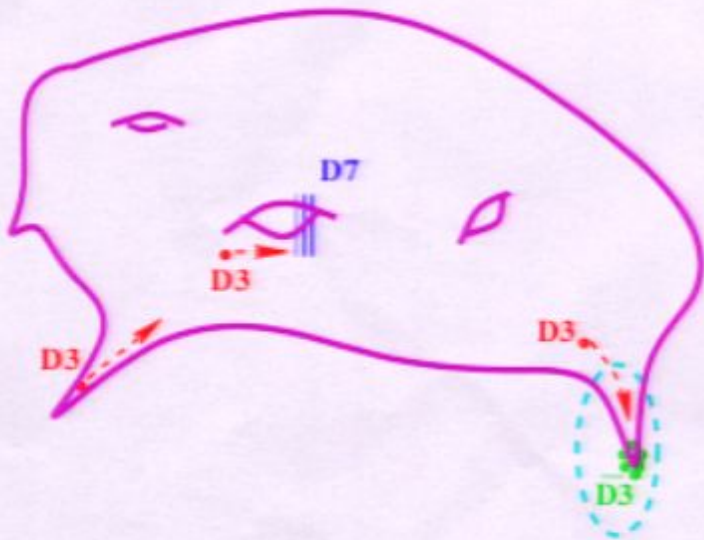
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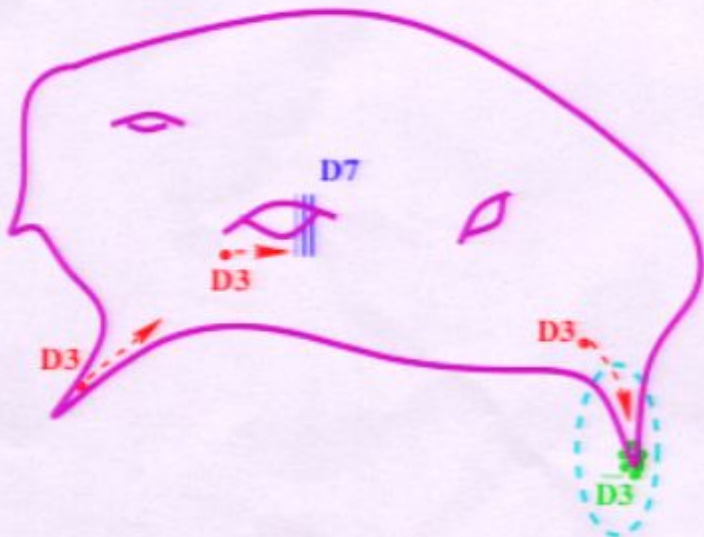
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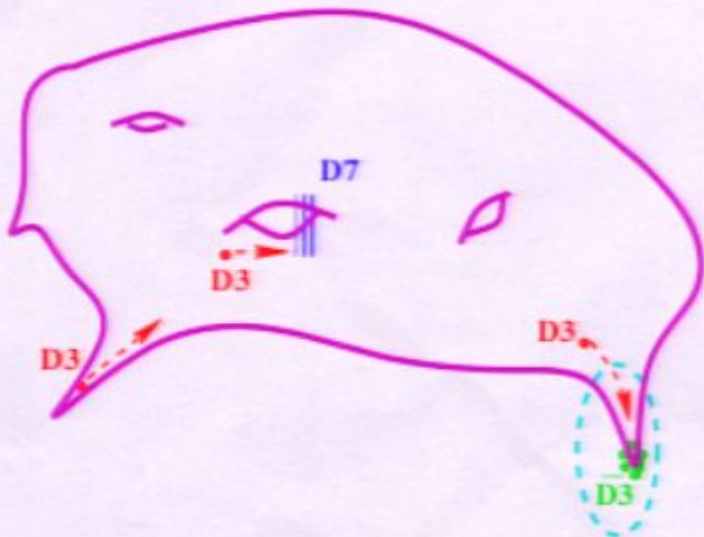
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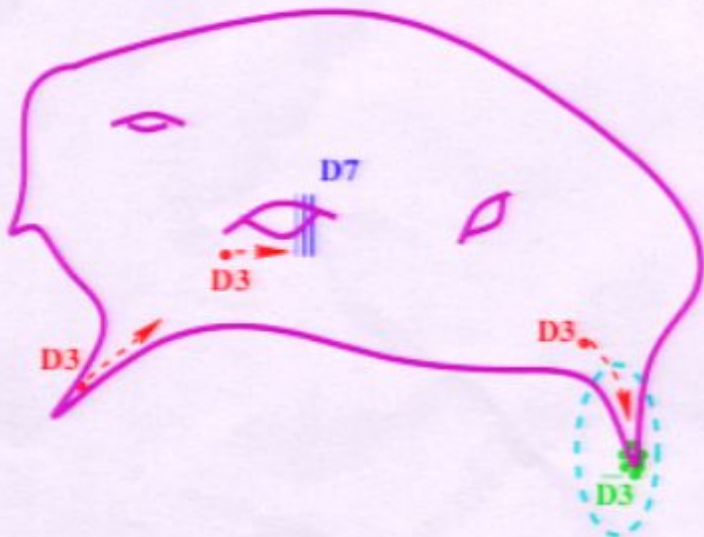
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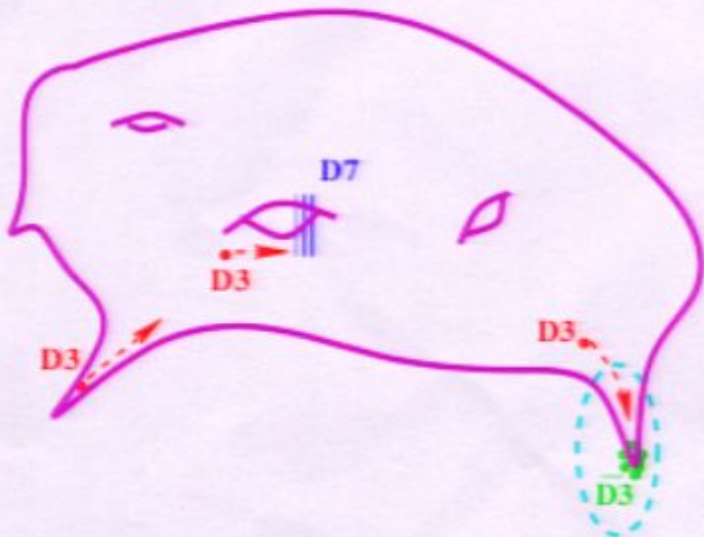
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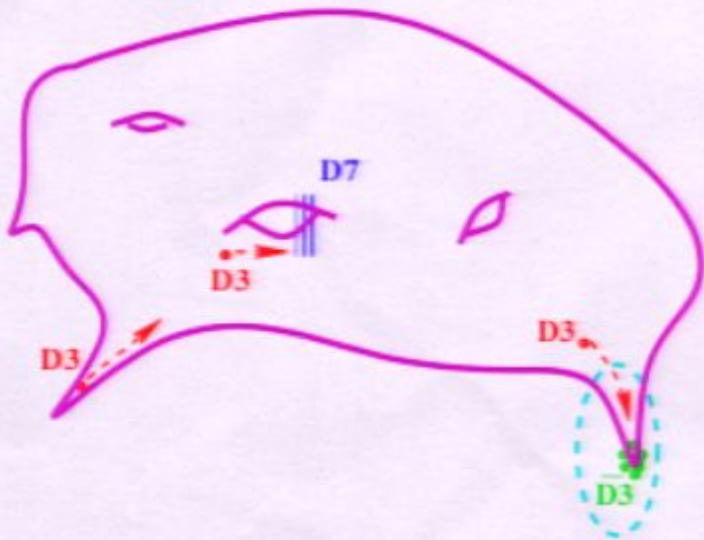
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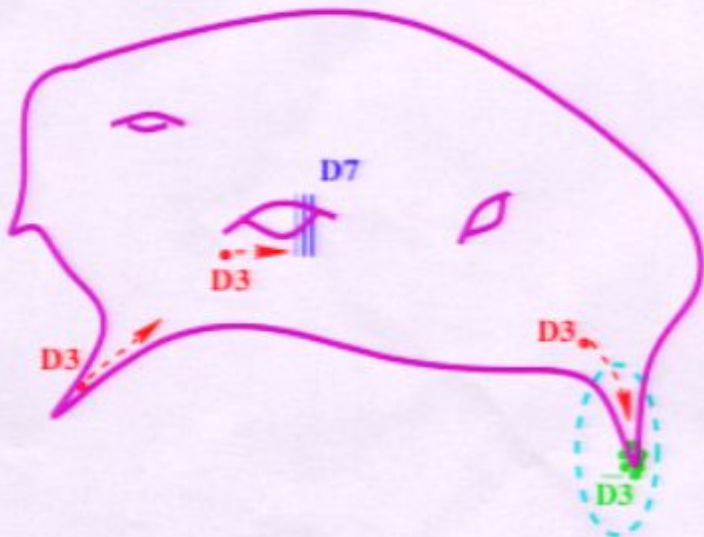
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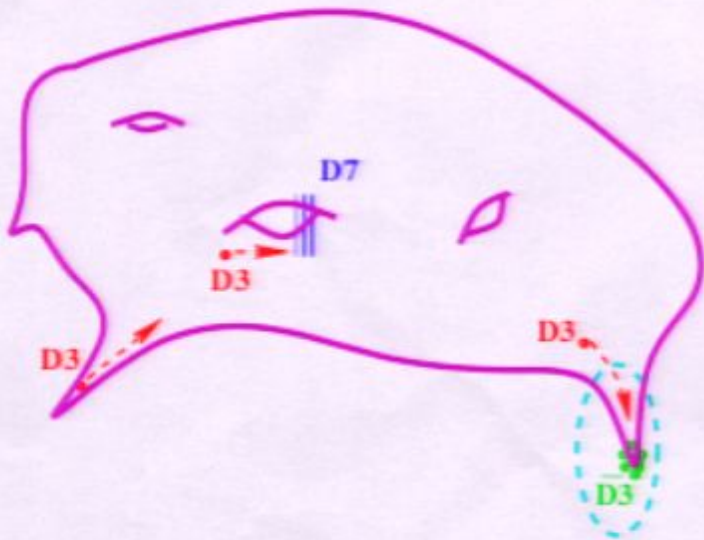
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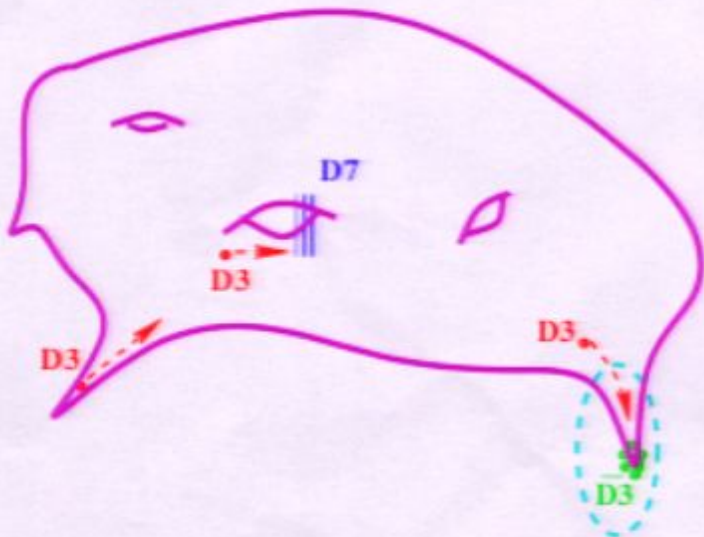
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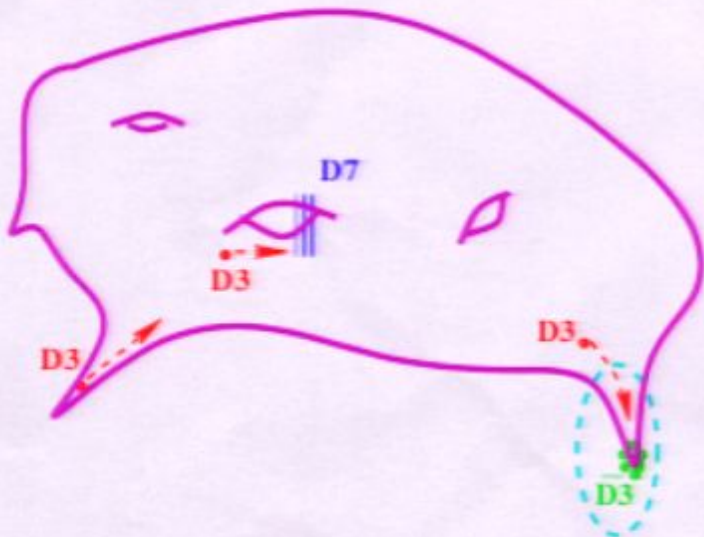
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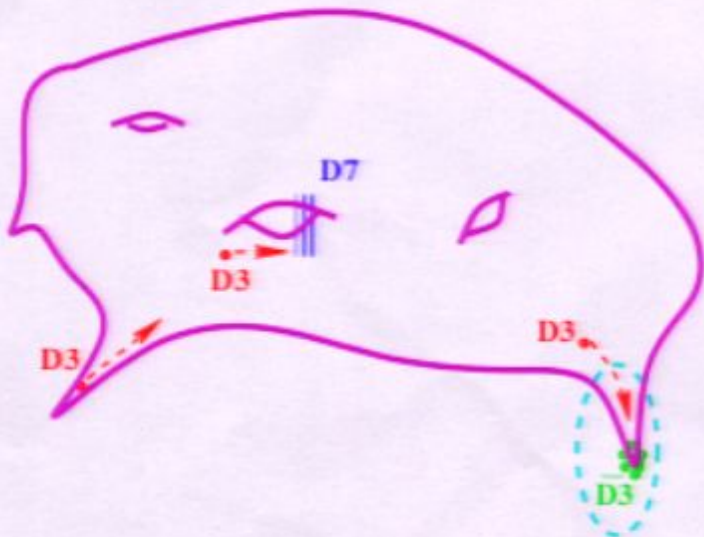
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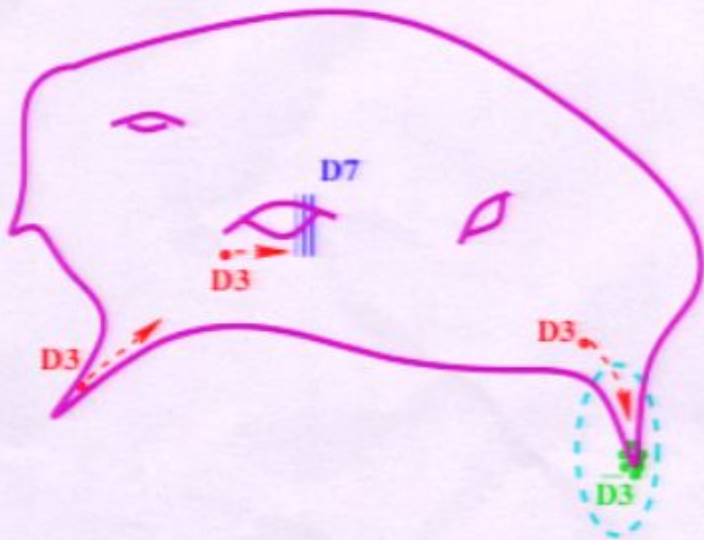
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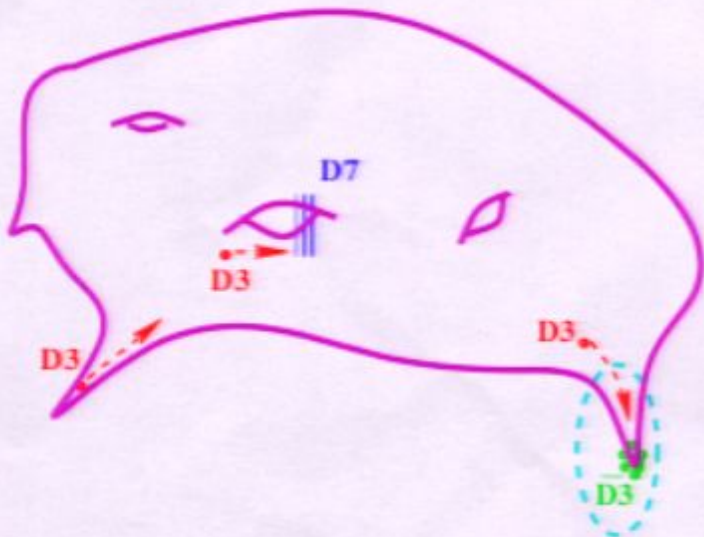
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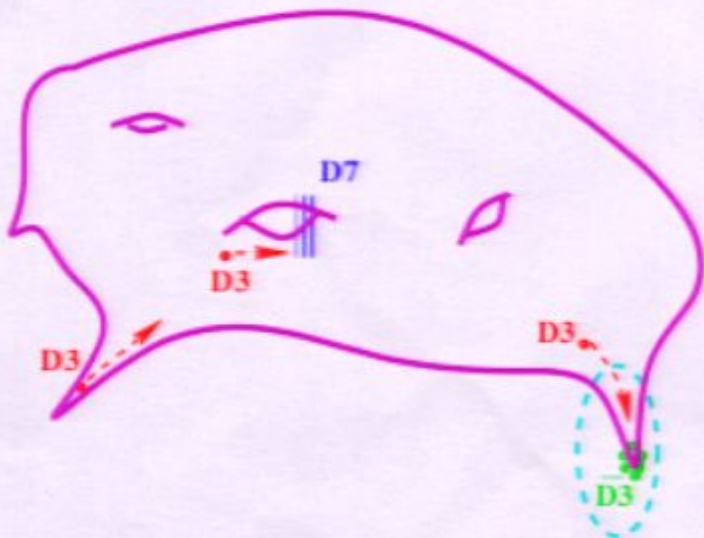
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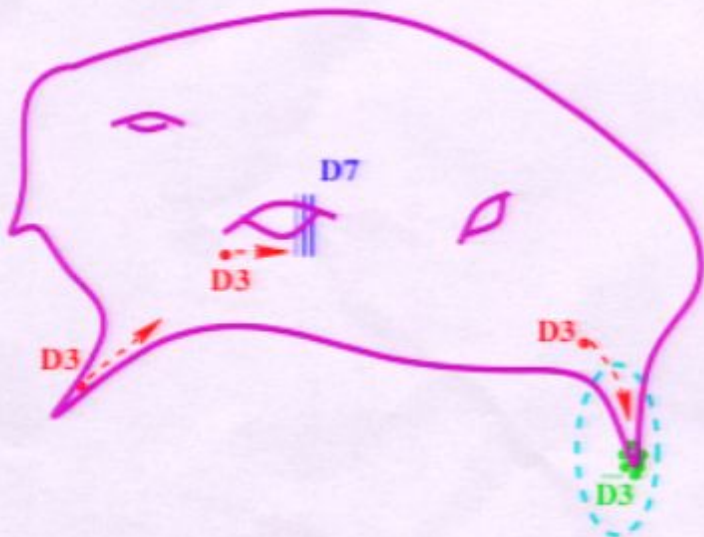
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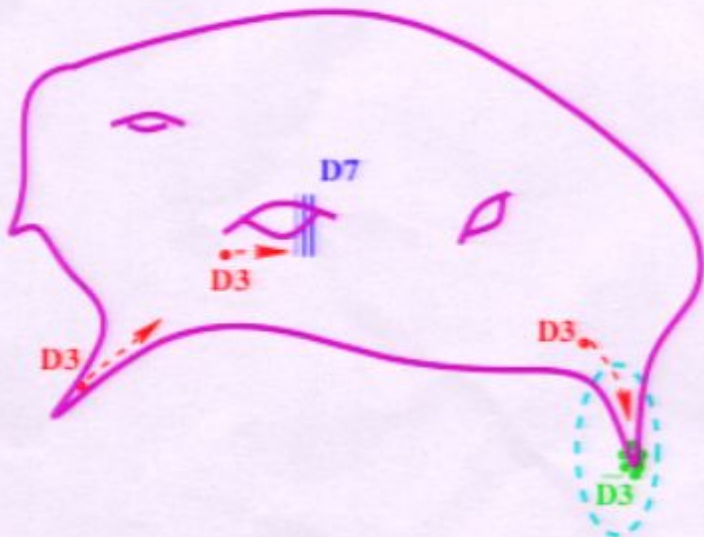
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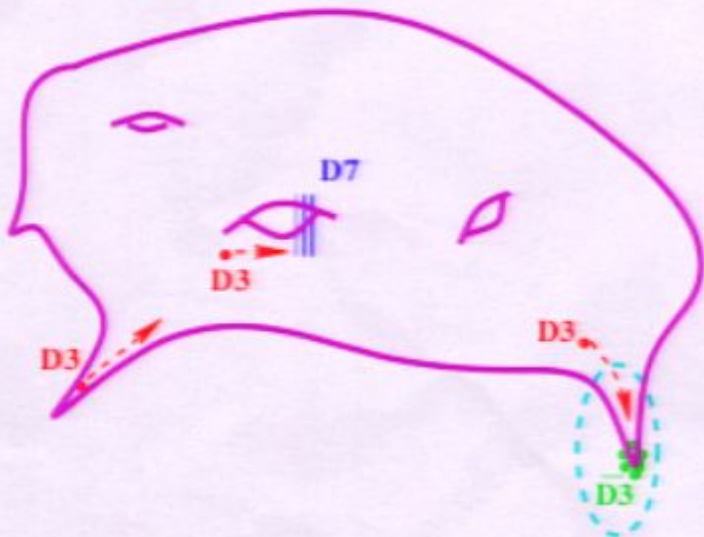
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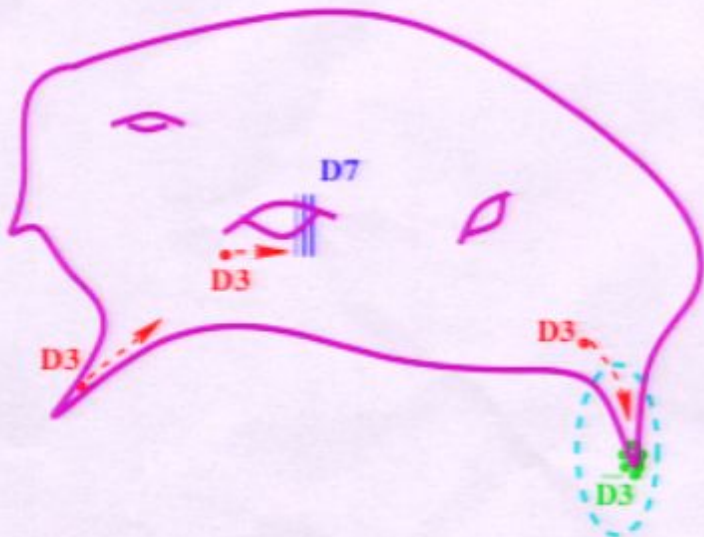
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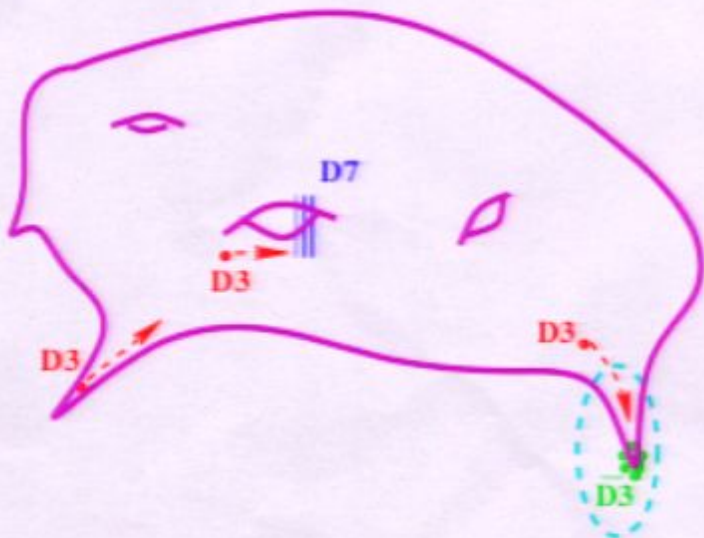
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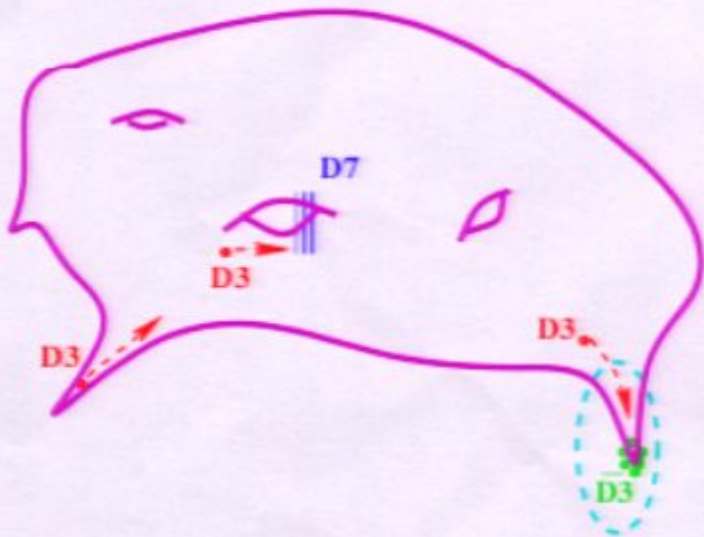
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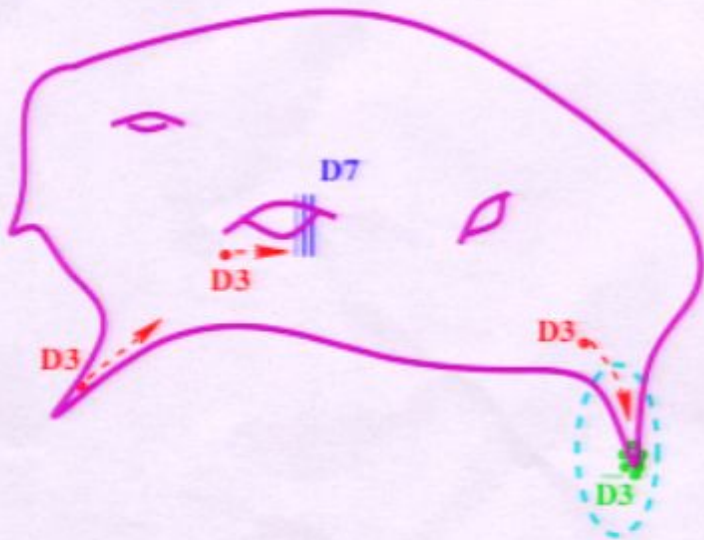
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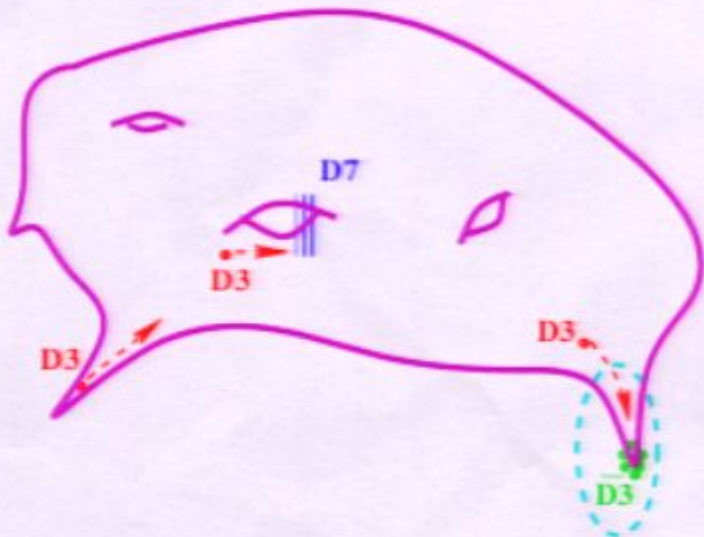
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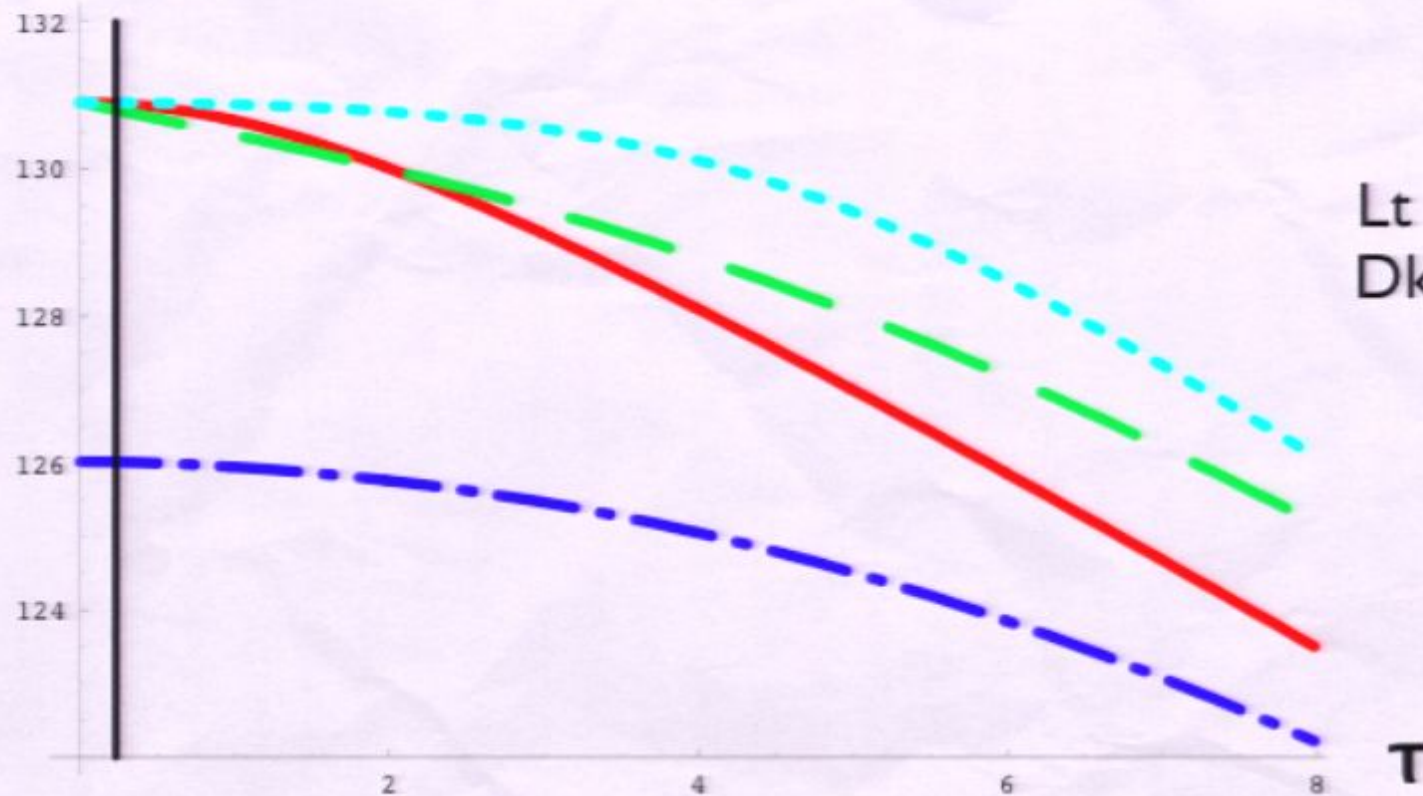
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



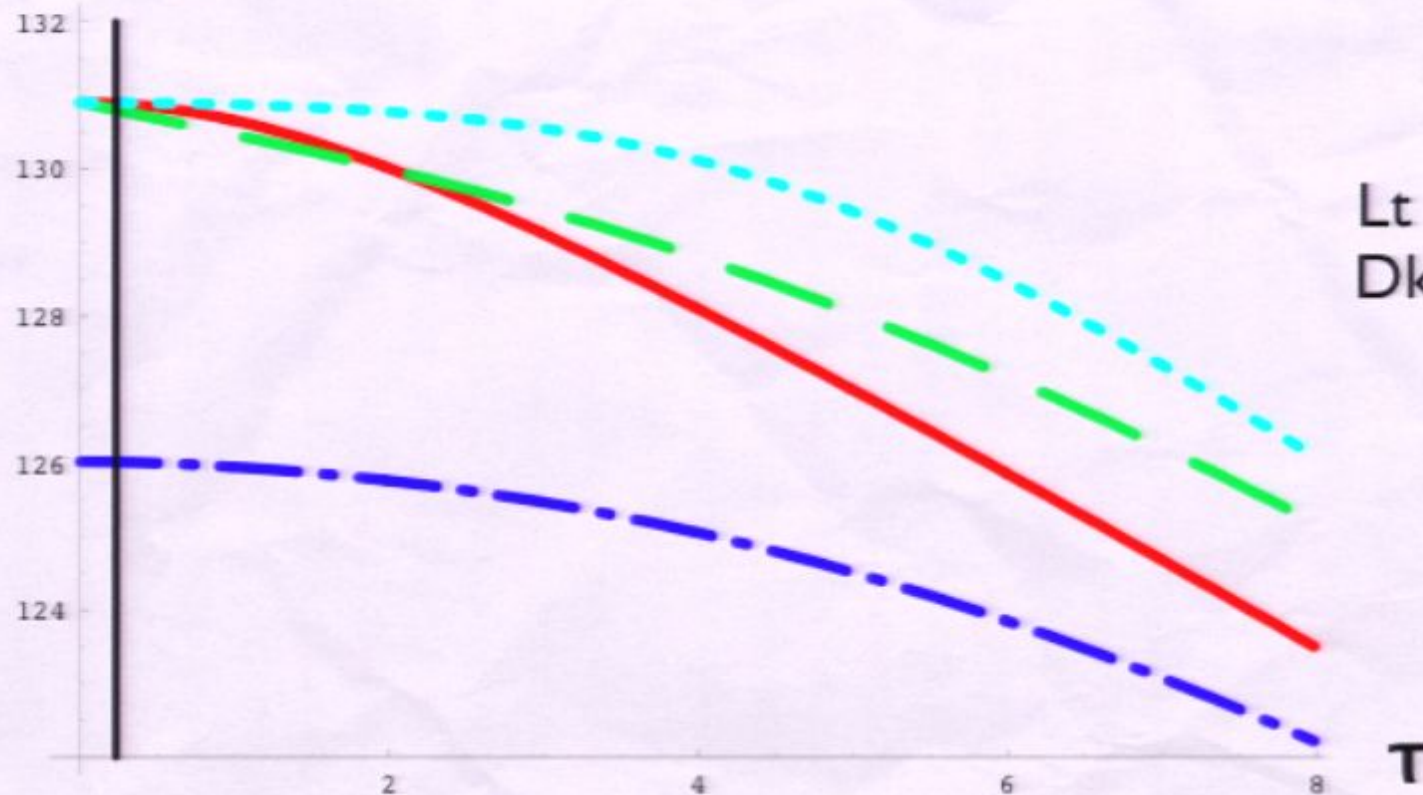
Red: KS
Green: AdS
Lt Blue: Mass Gap
Dk Blue: AdS+Log

N=10⁶
M=800

CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



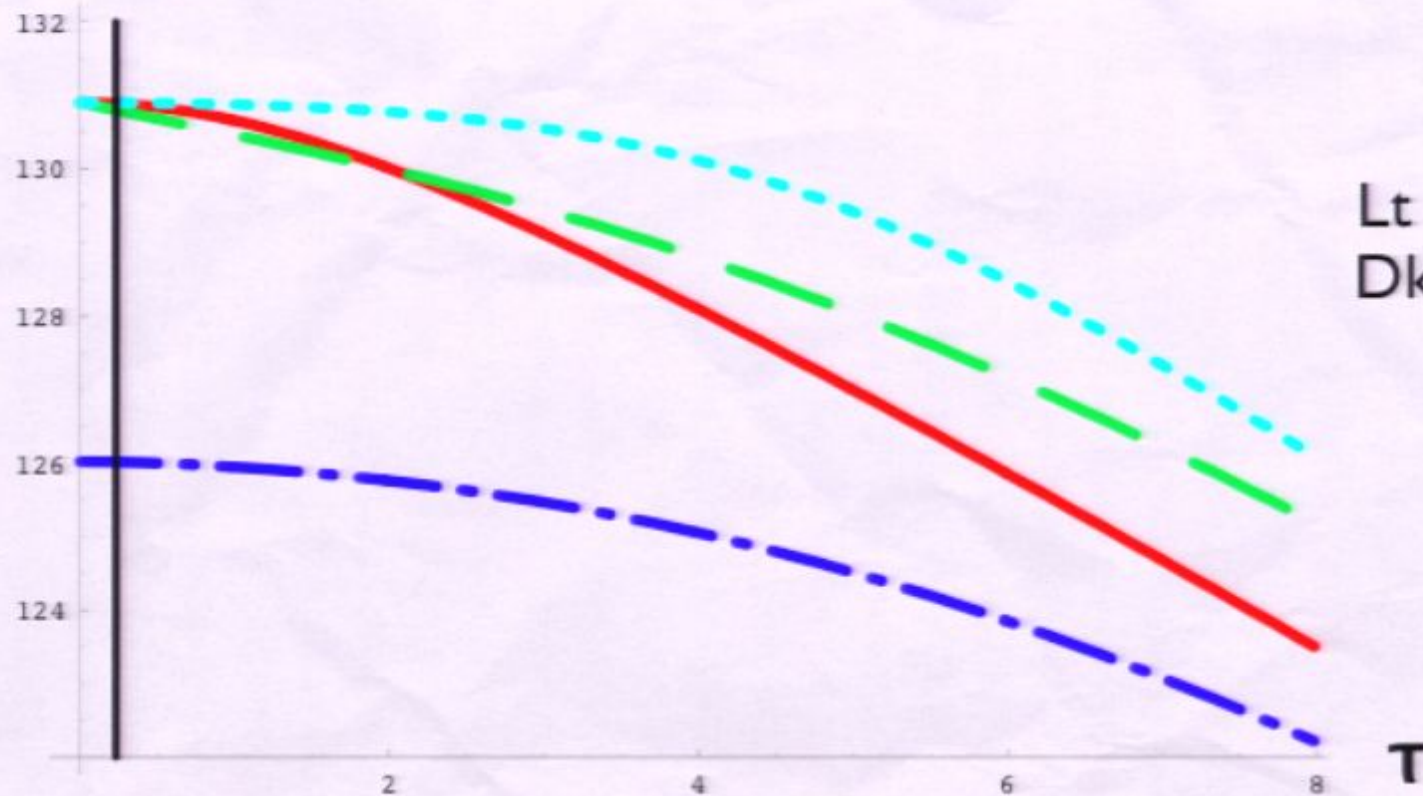
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Log[h_A]

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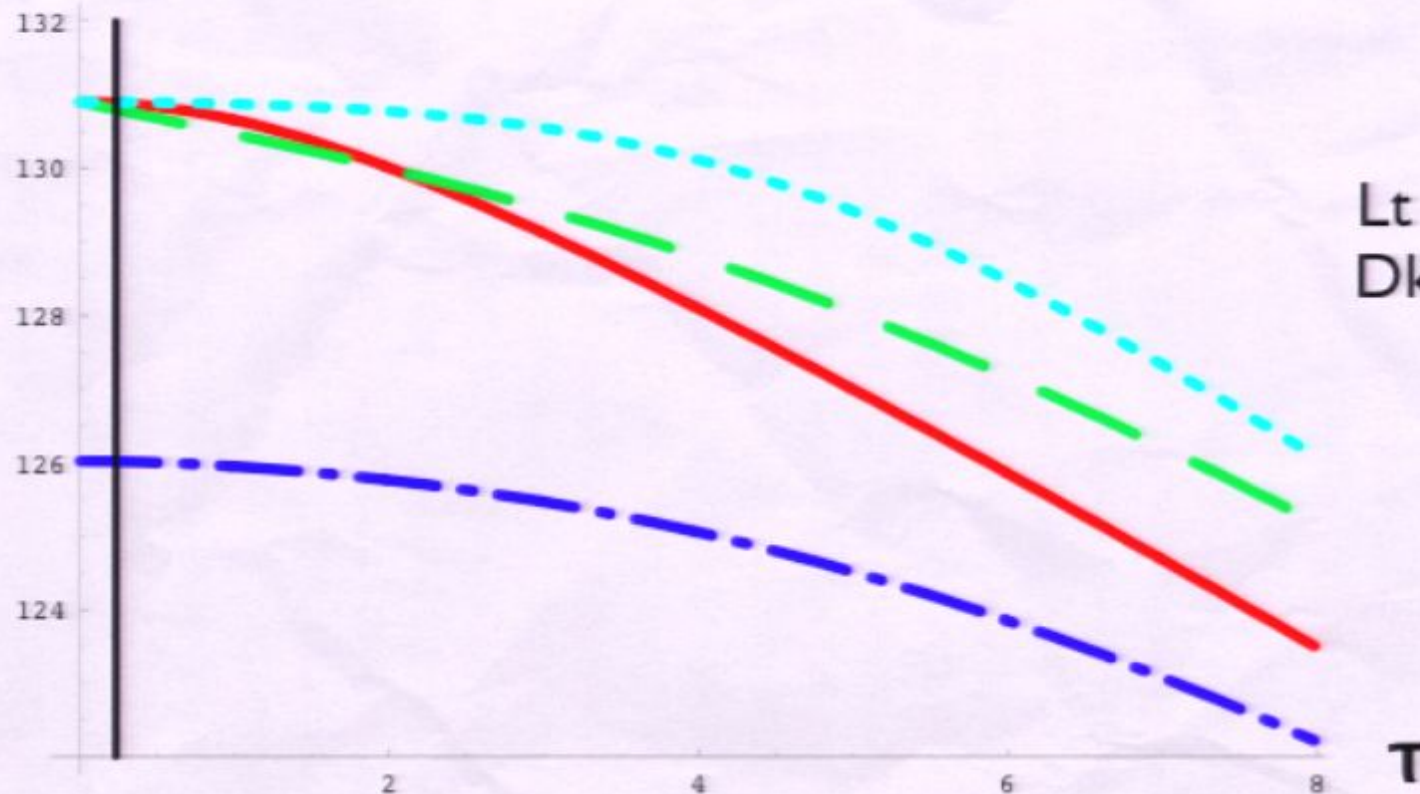
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Log[h_A]

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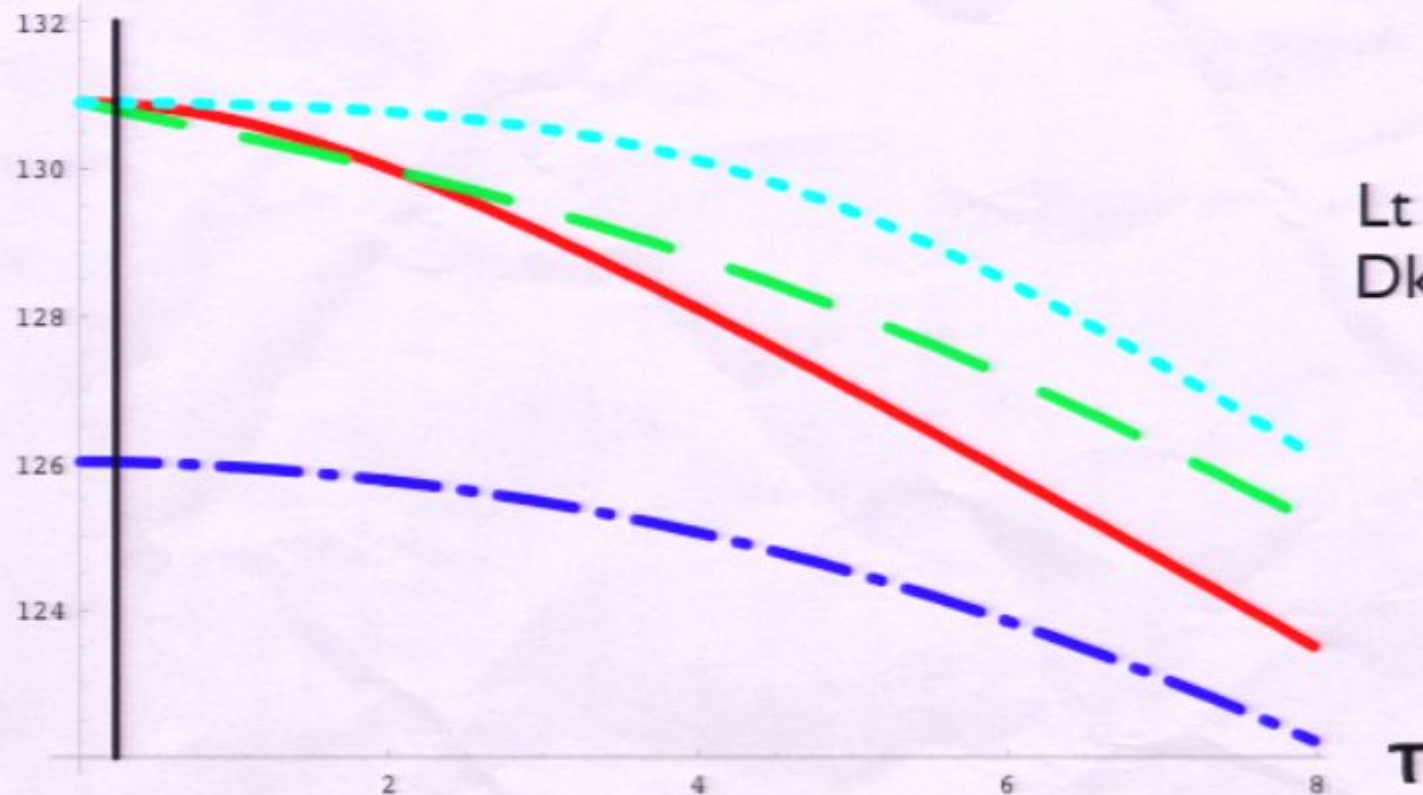
Red: KS
Green: AdS
Lt Blue: Mass Gap
Dk Blue: AdS+Log

$N=10^6$
 $M=800$

CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



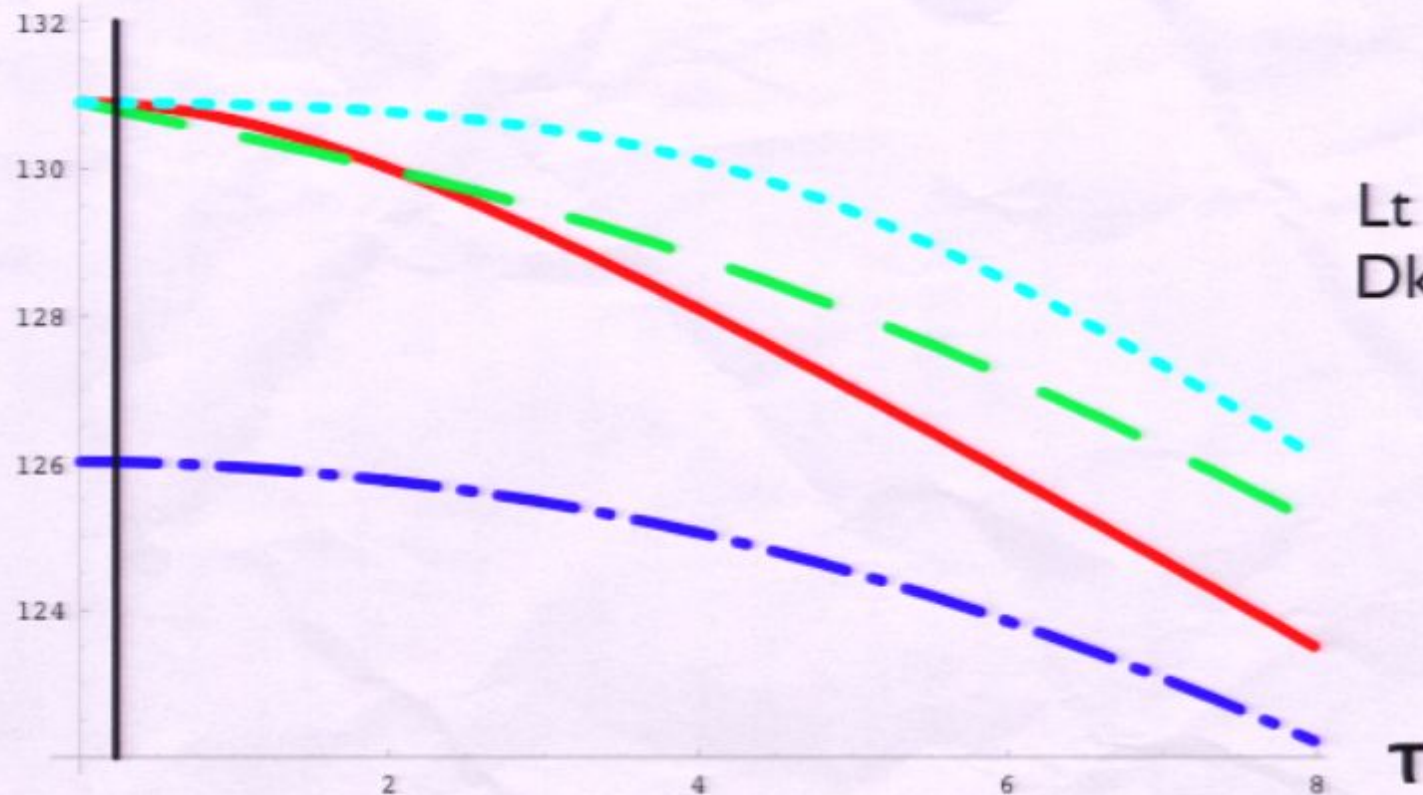
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



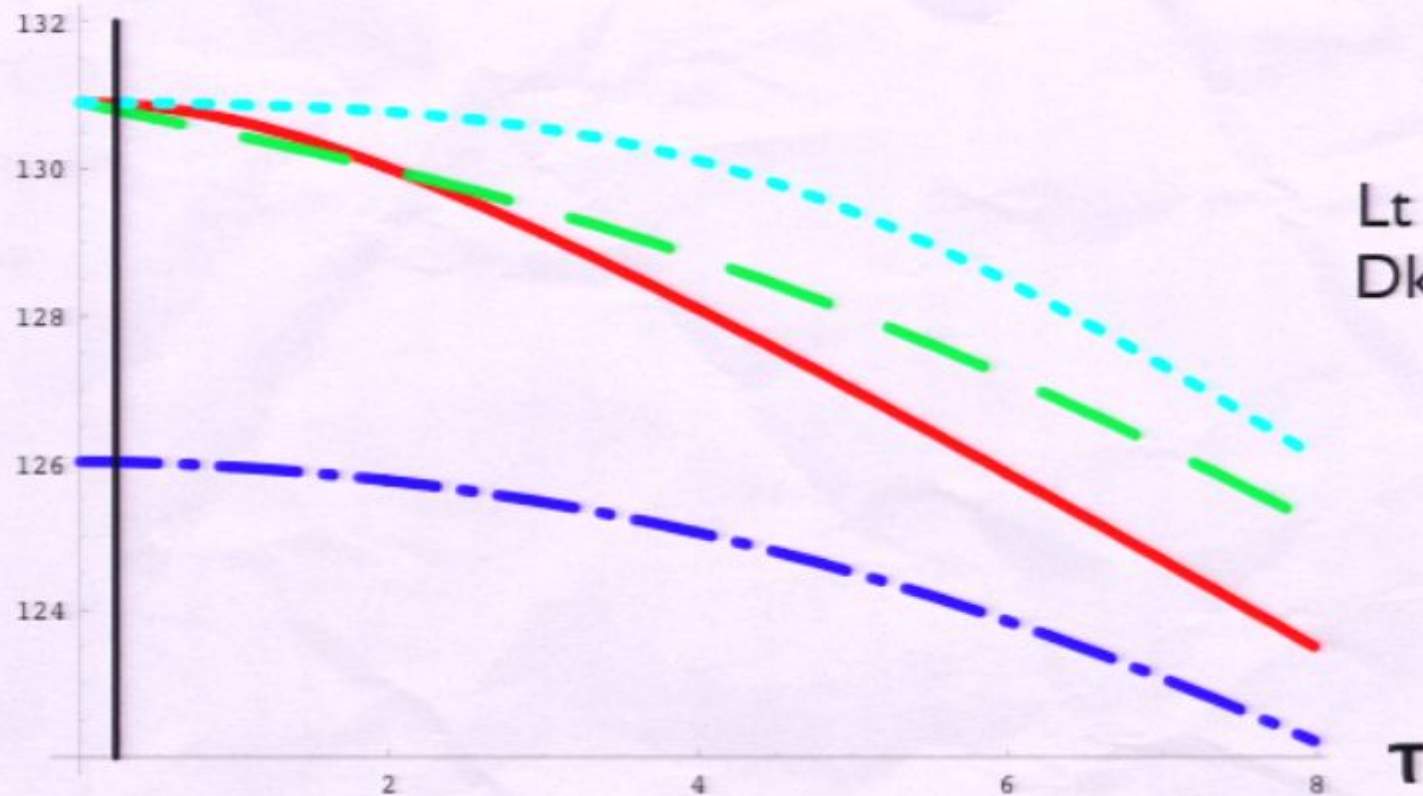
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



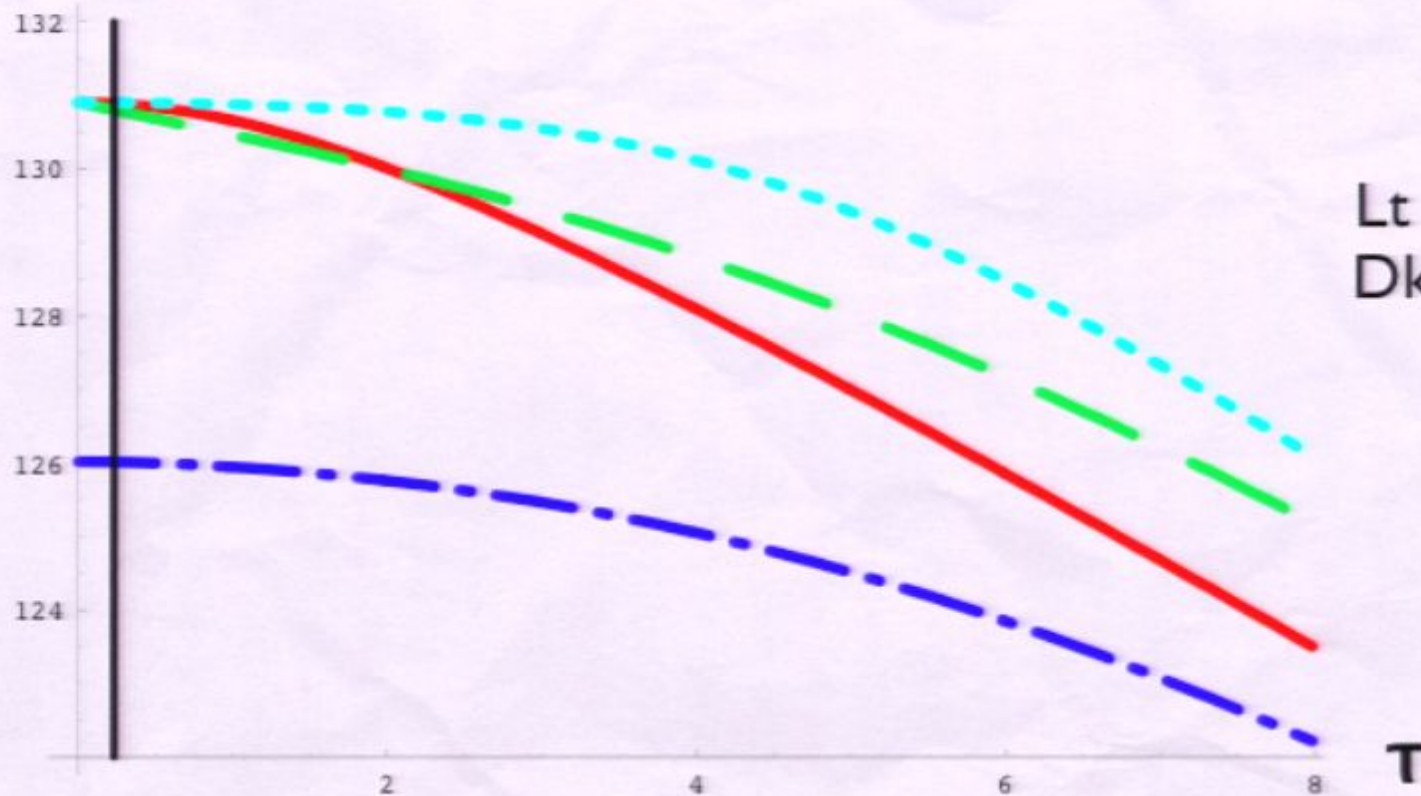
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

Zoom In



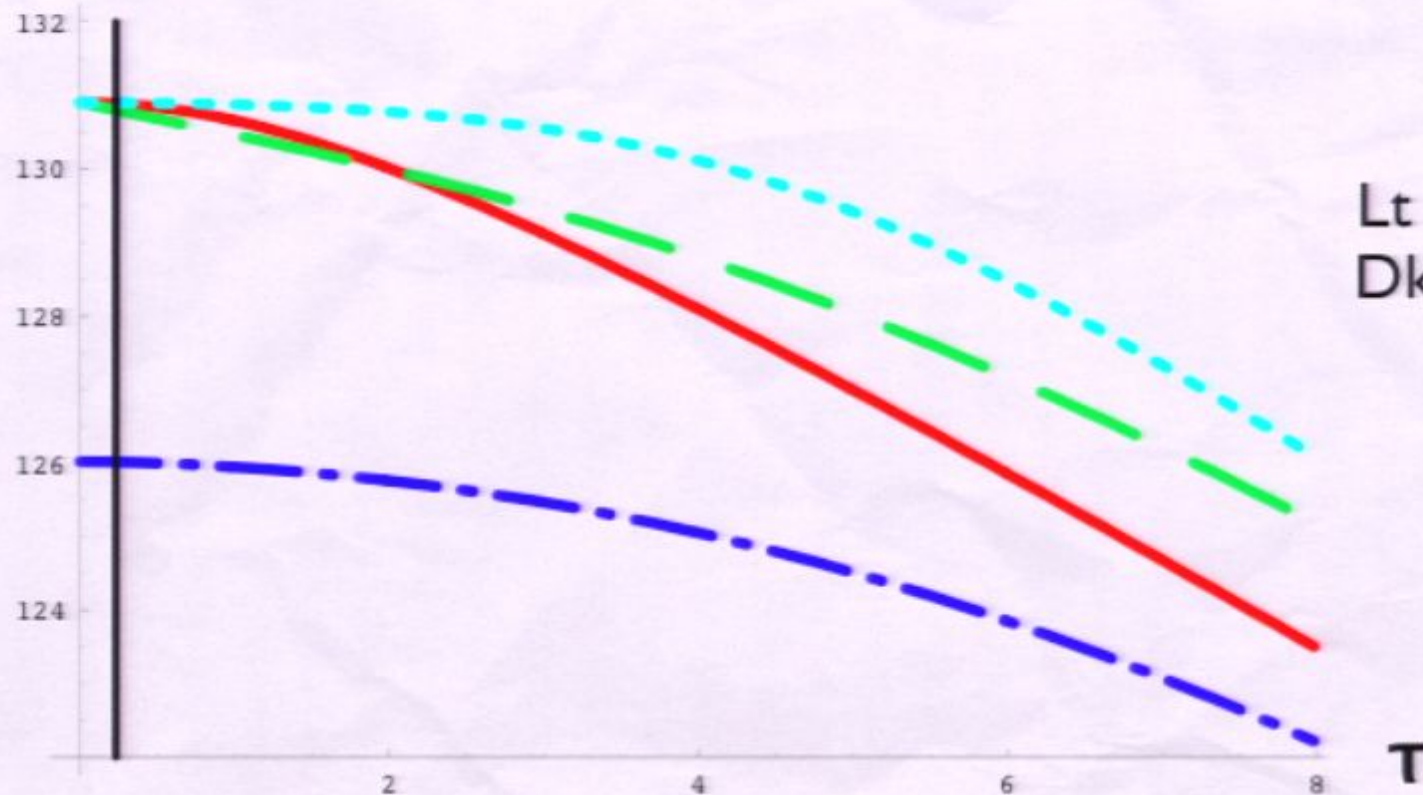
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

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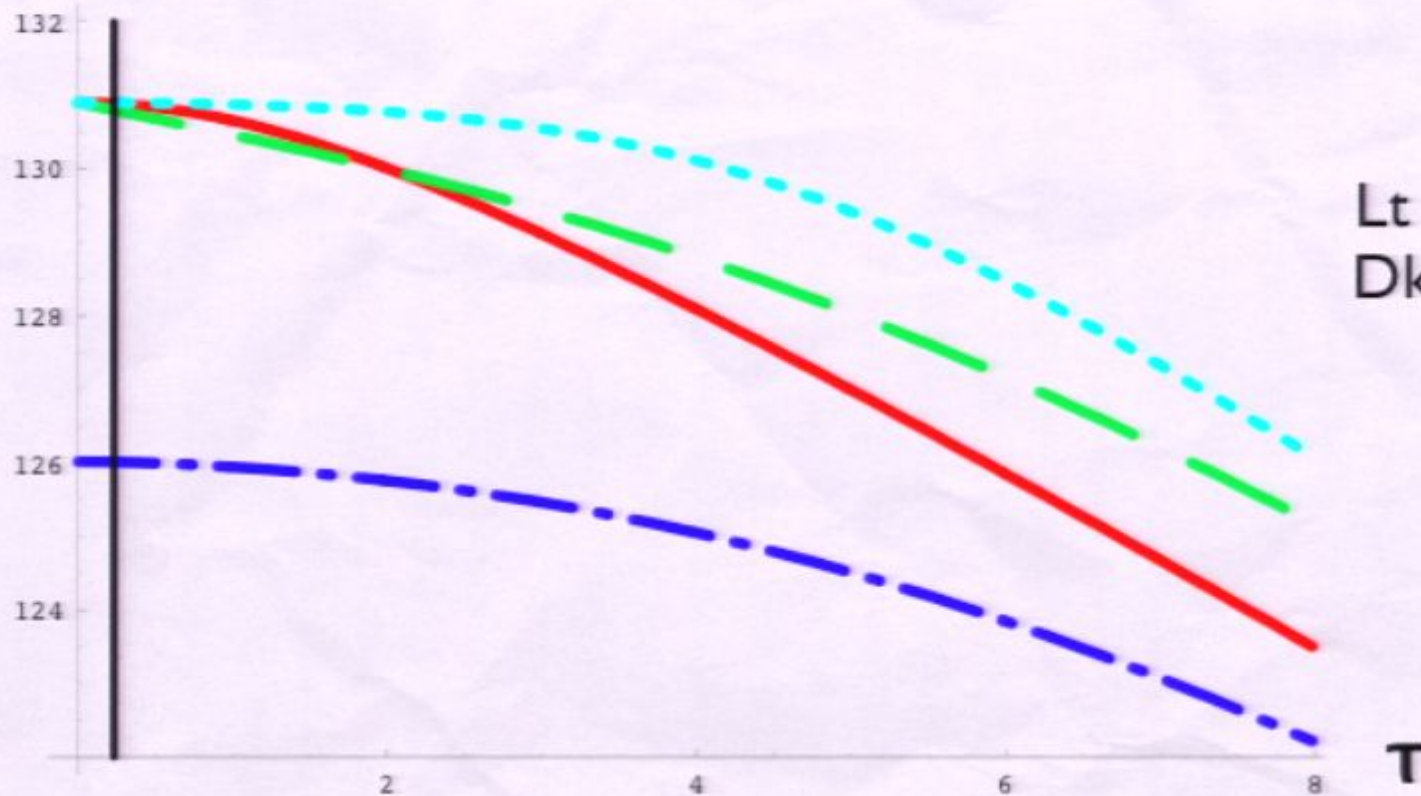
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CHANGE IN SOUND SPEED TRACKS GEOMETRY

Log[h_A]

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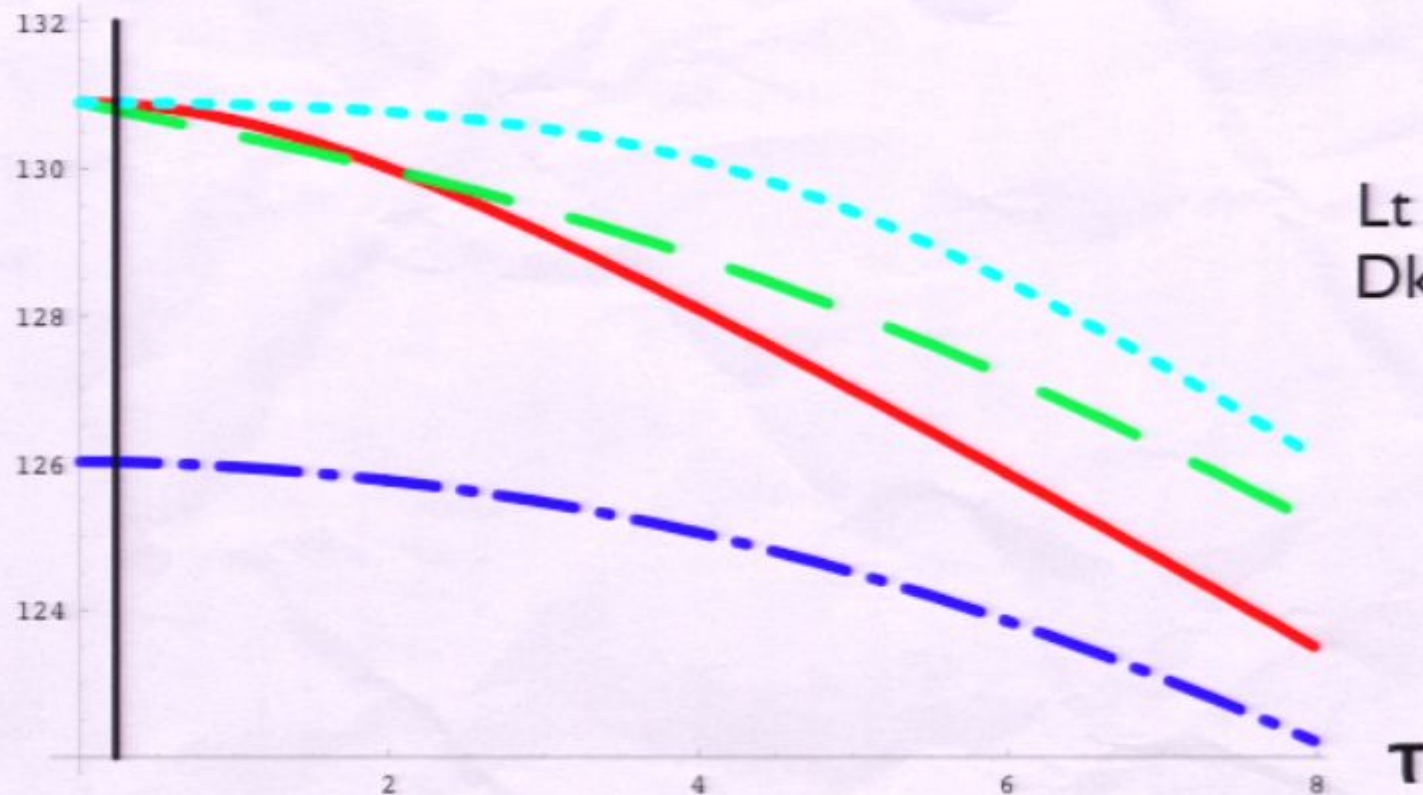
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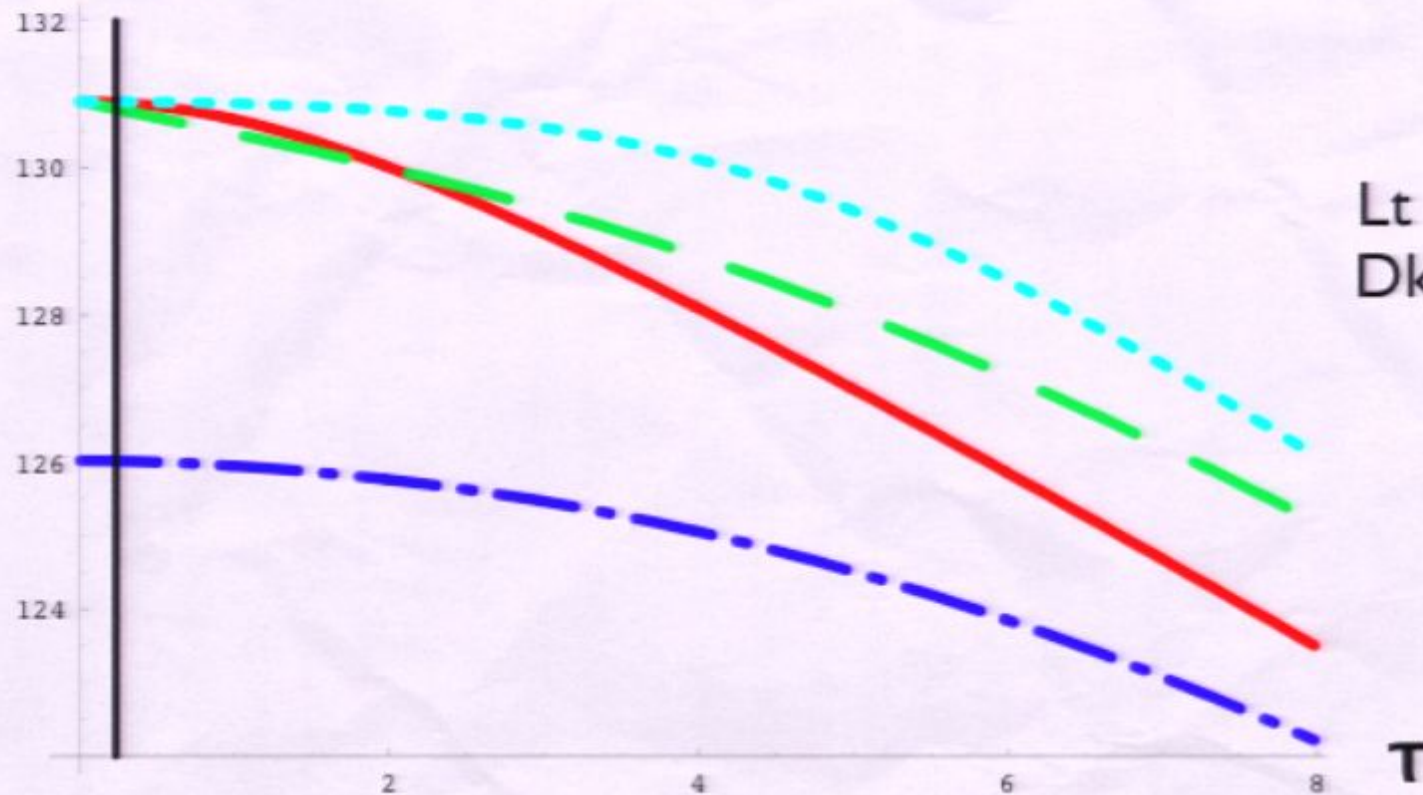
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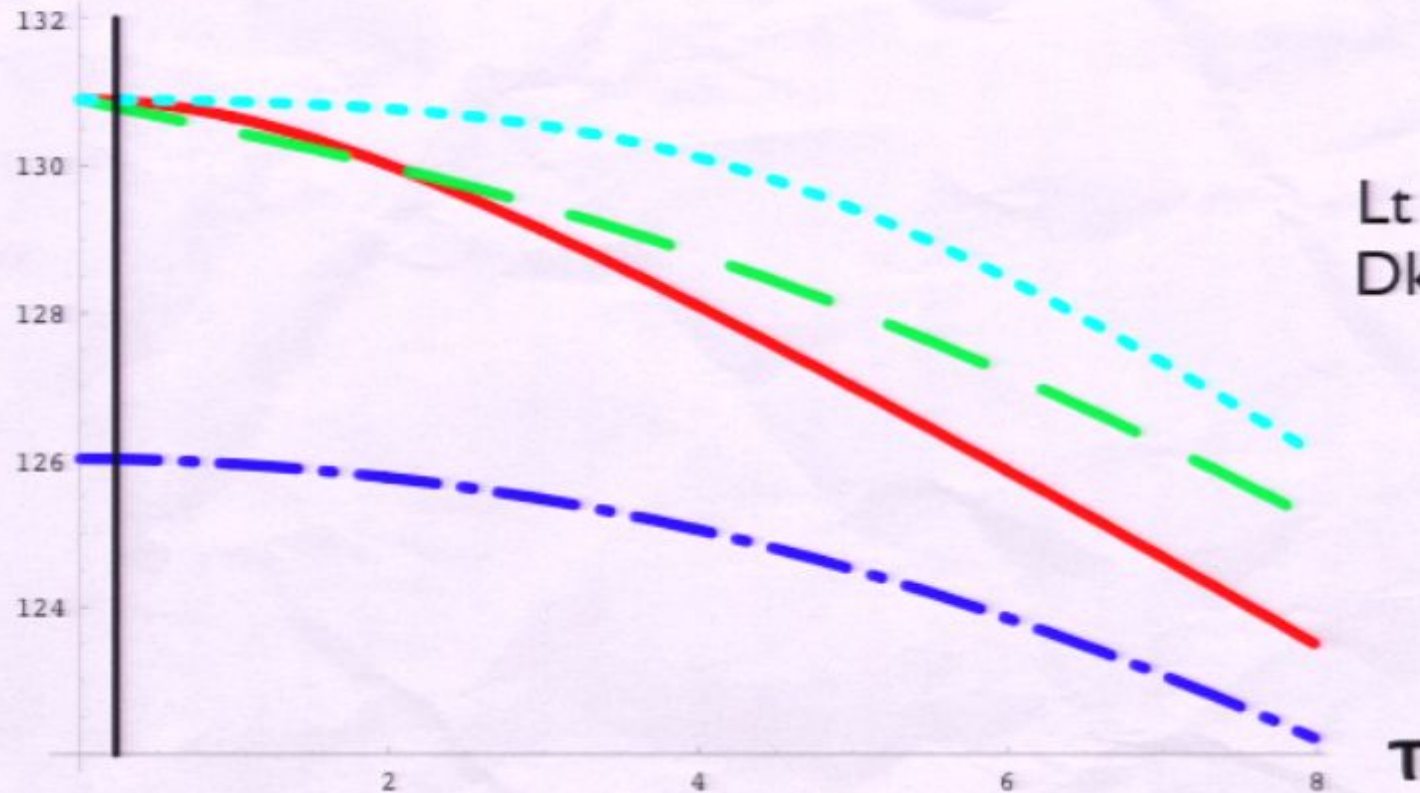
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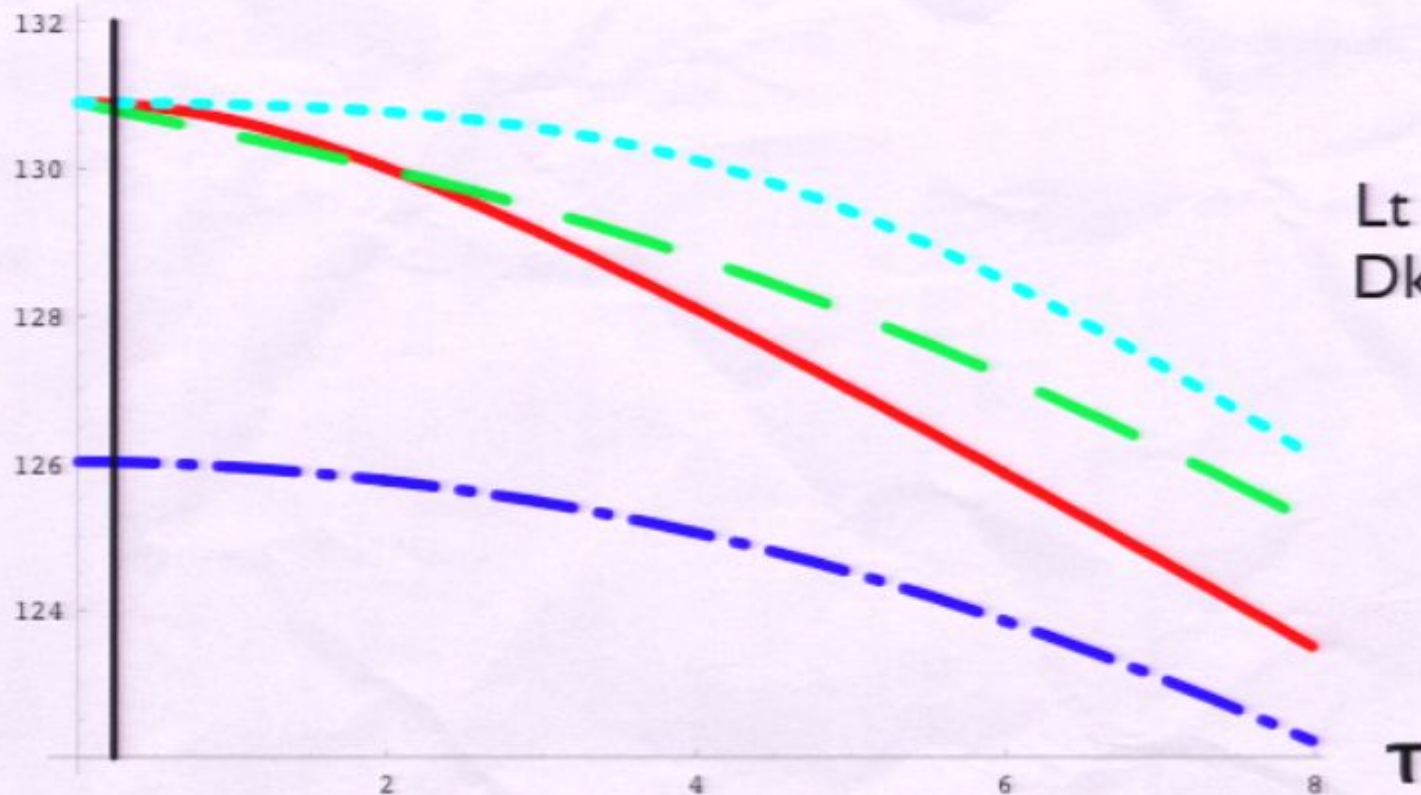
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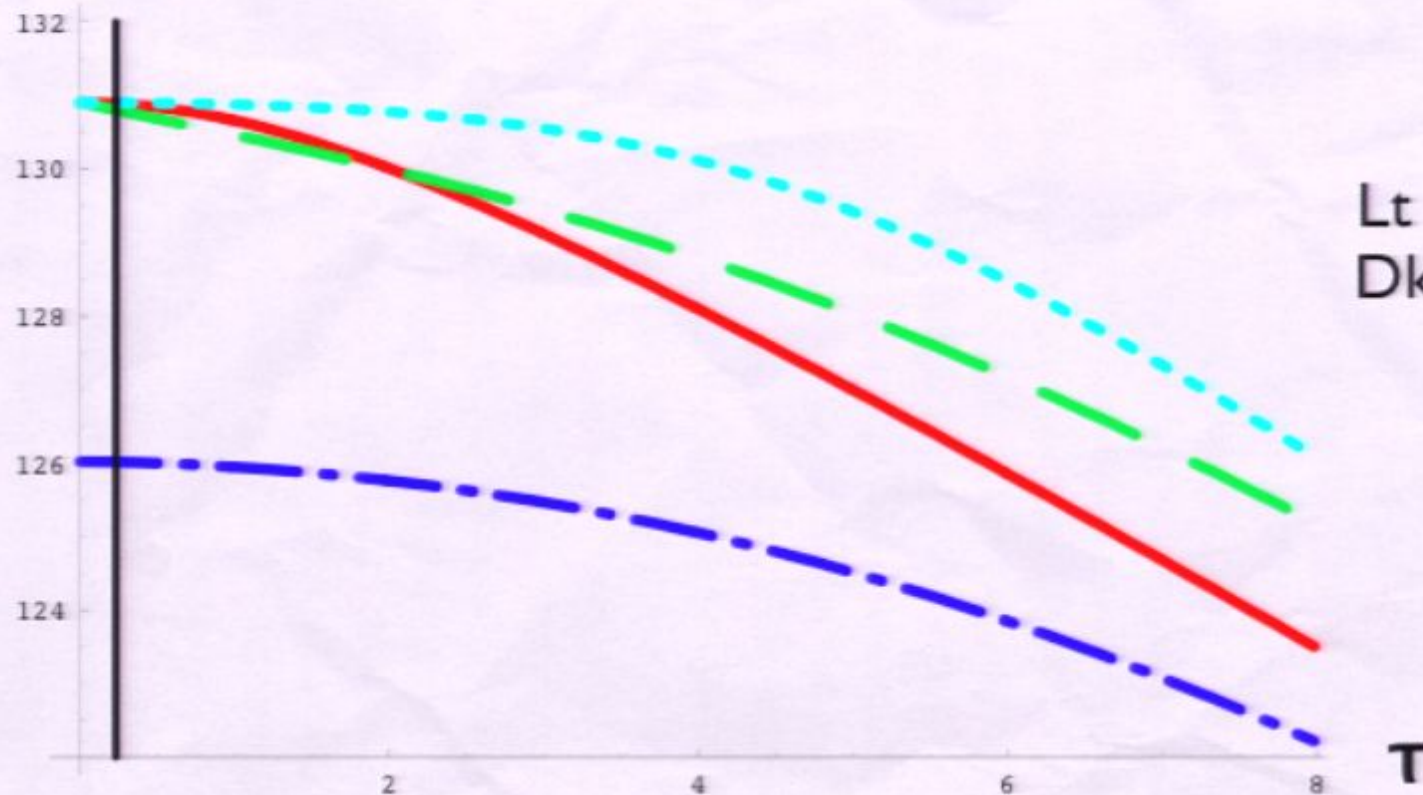
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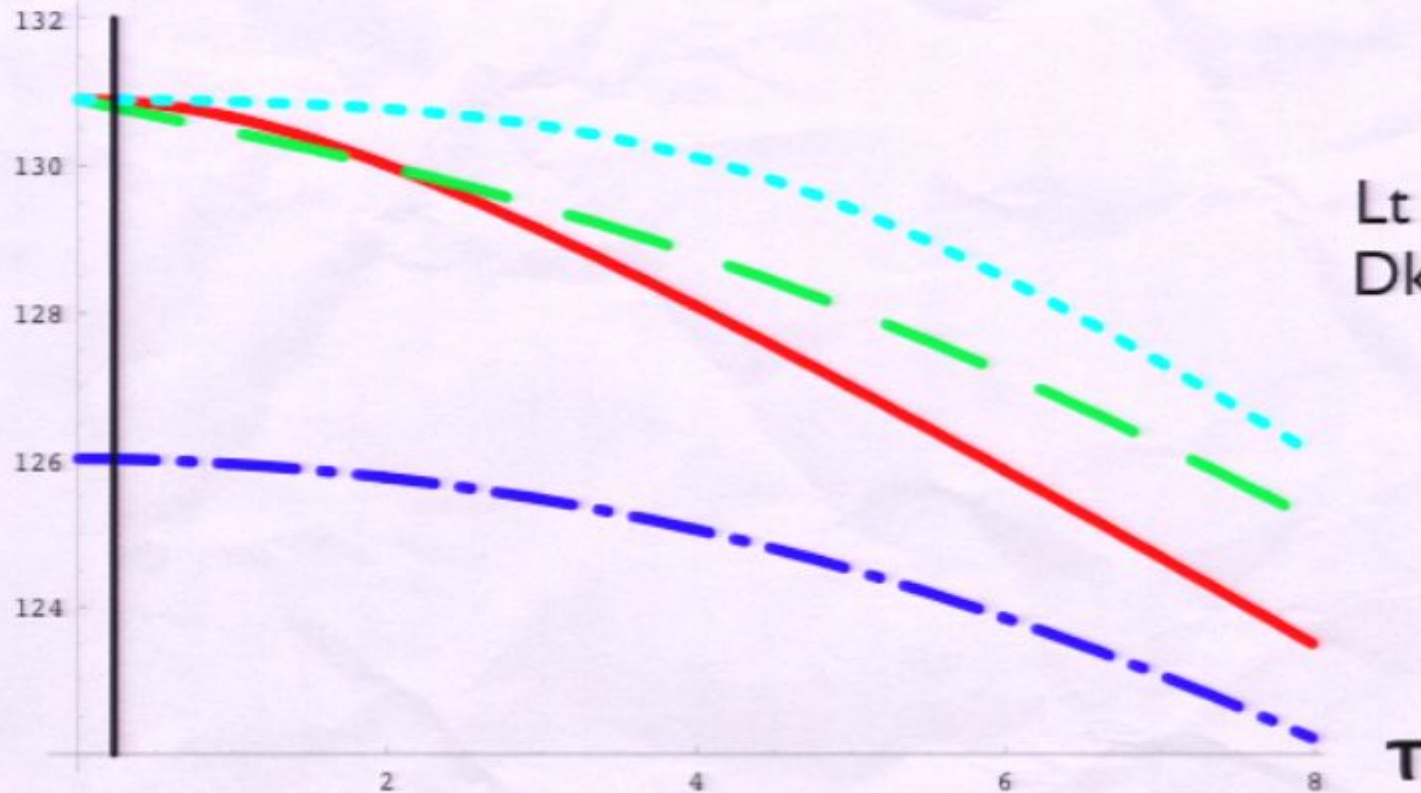
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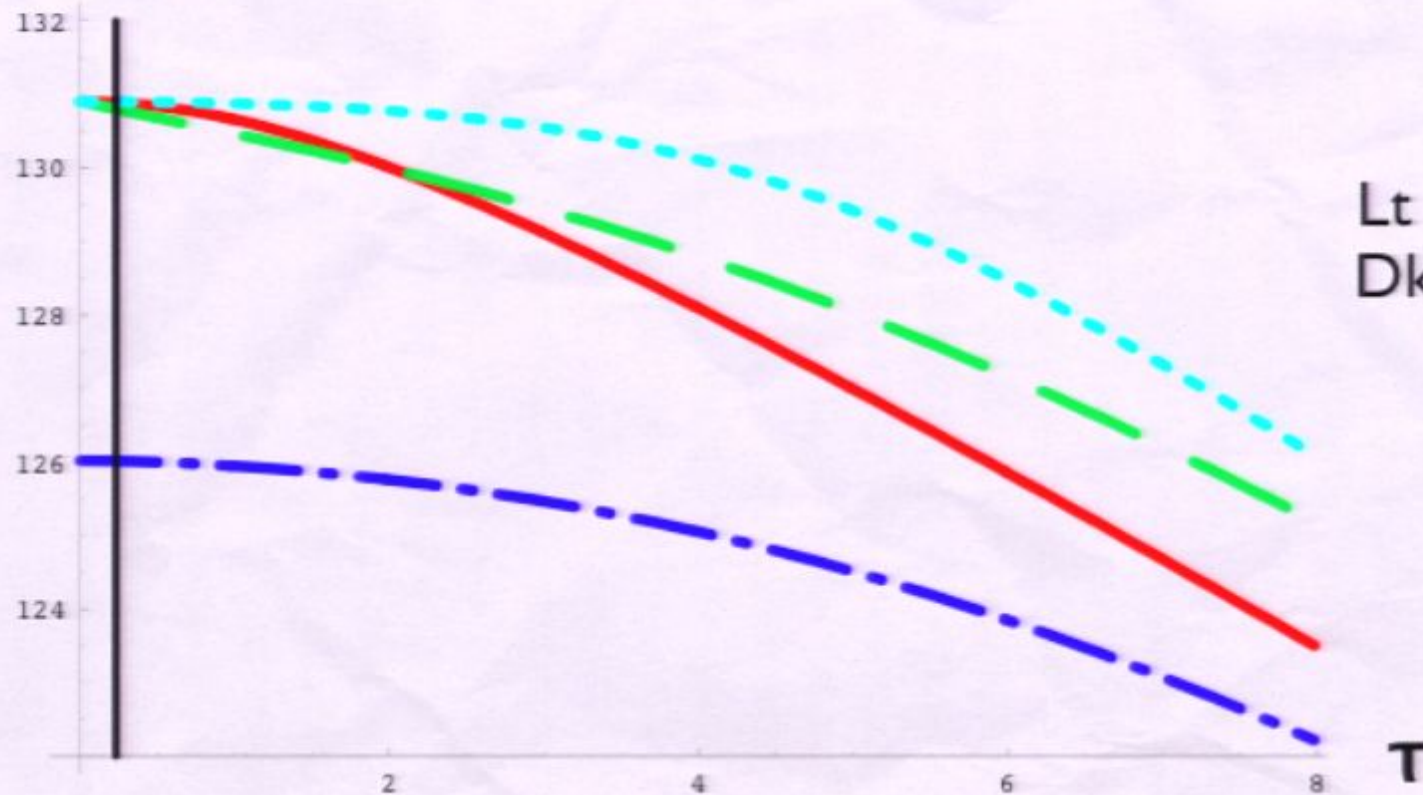
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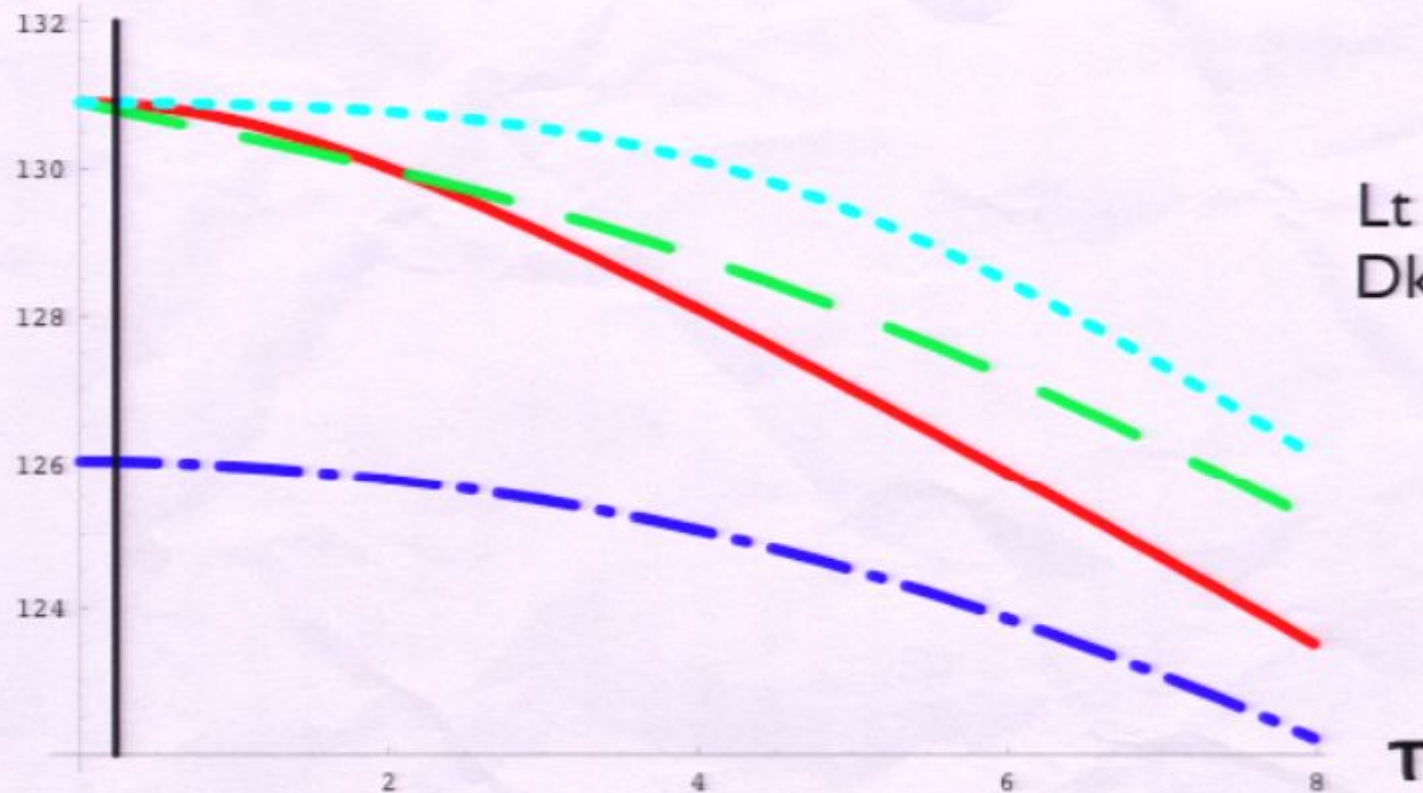
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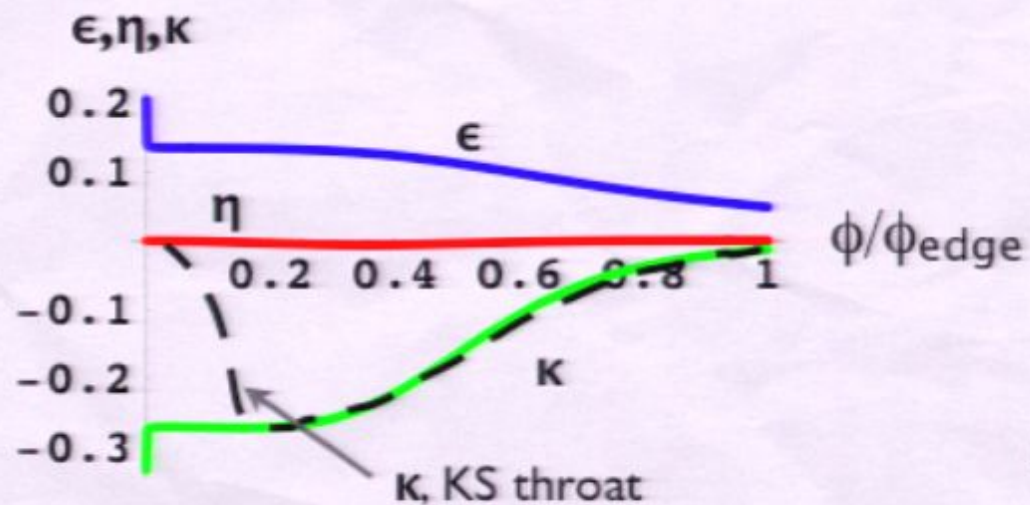
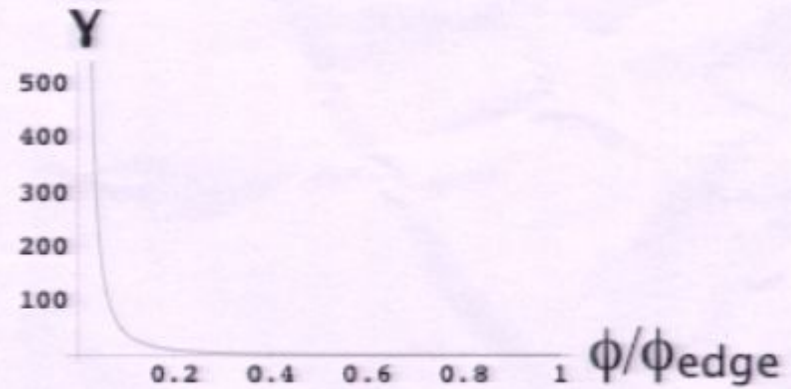
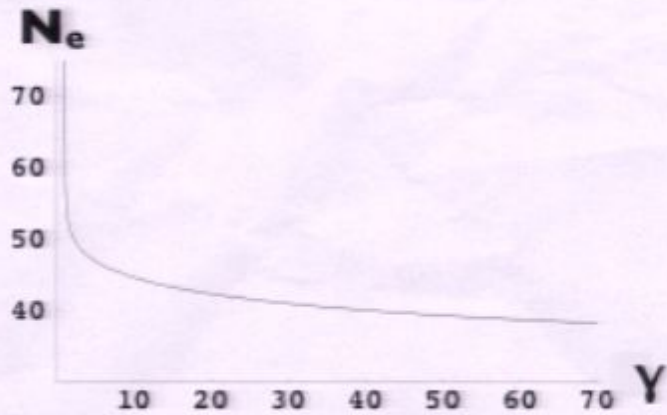
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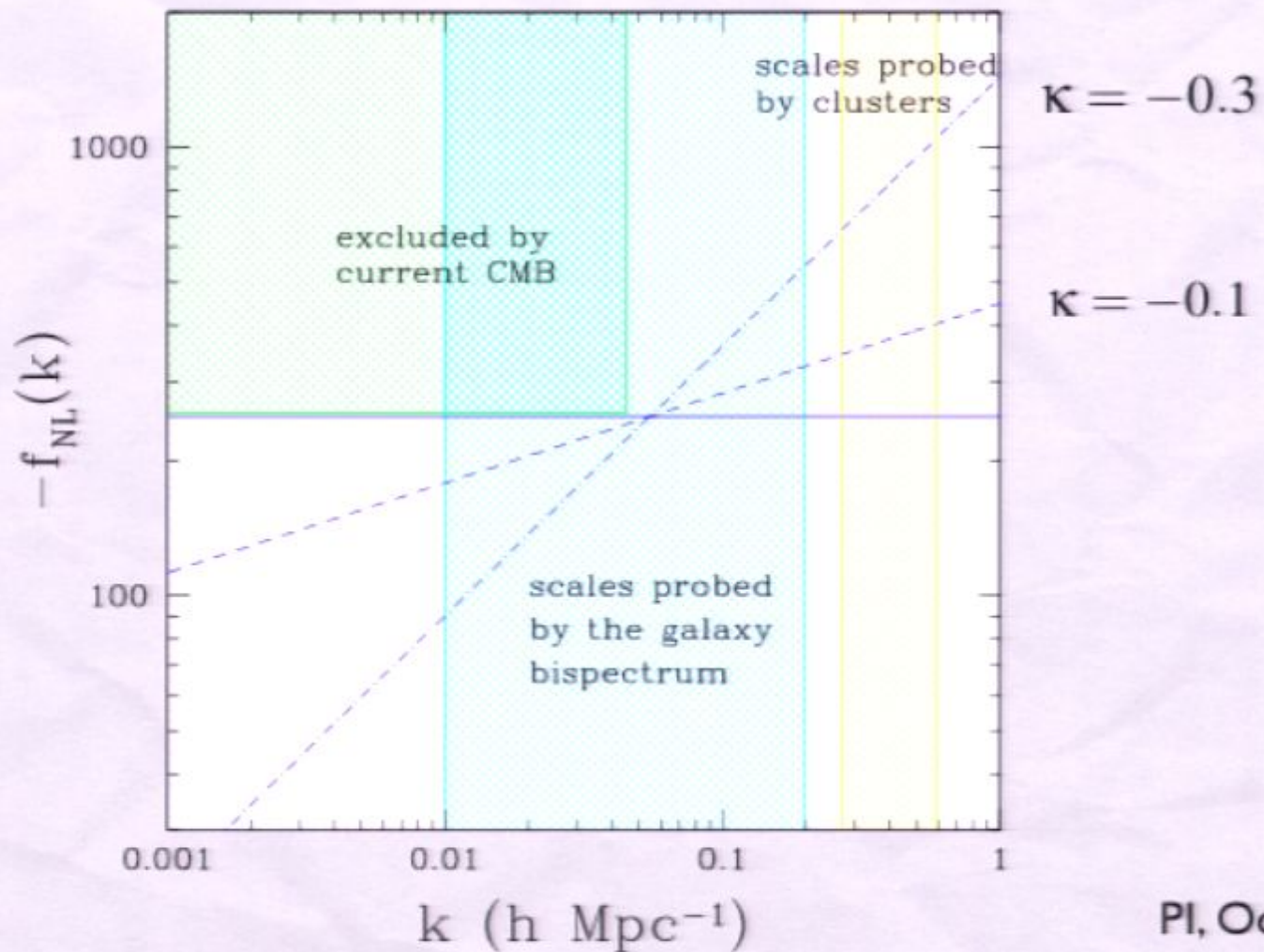
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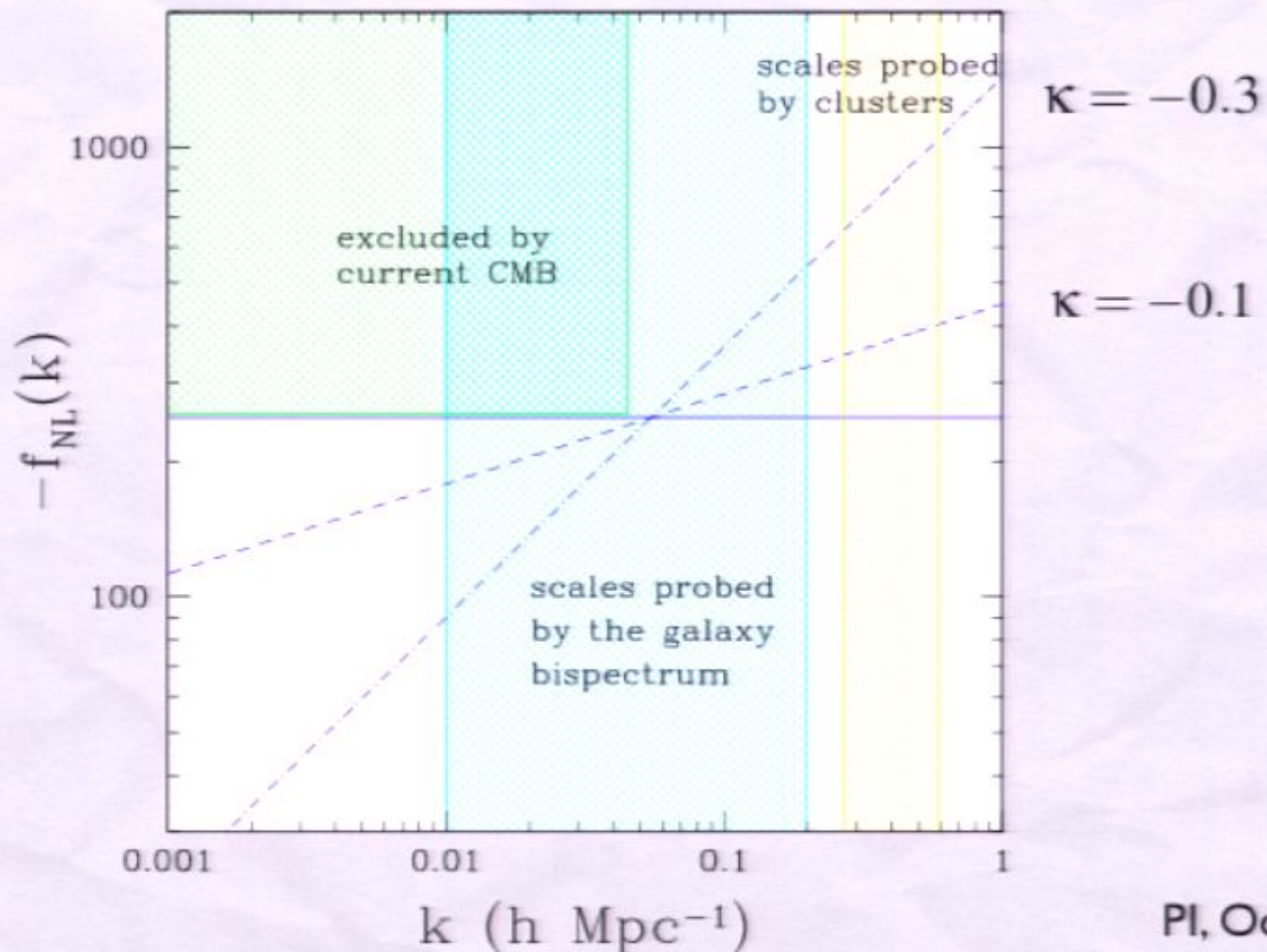
IN THE ADS THROAT



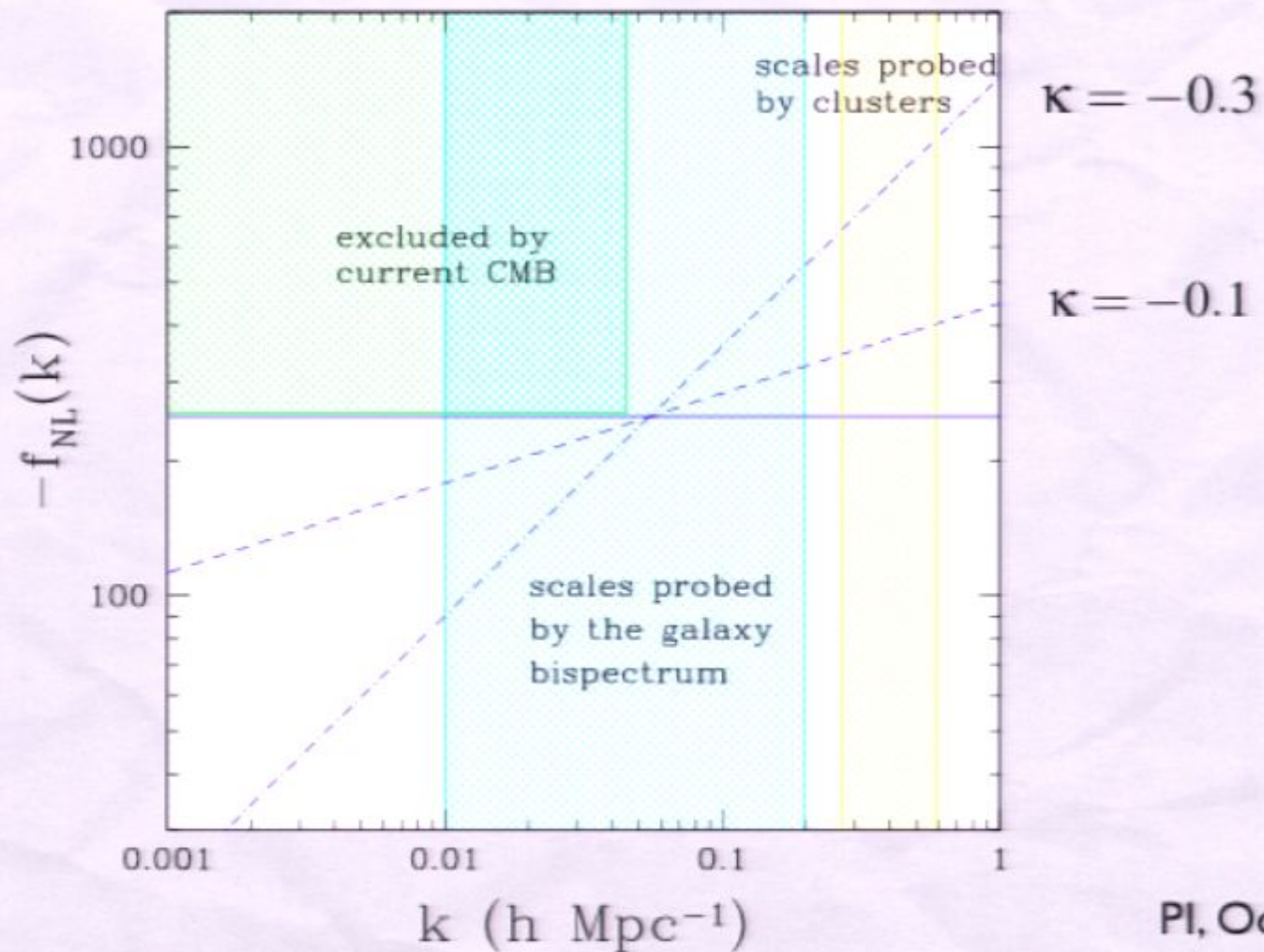
RELEVANT OBSERVATIONS



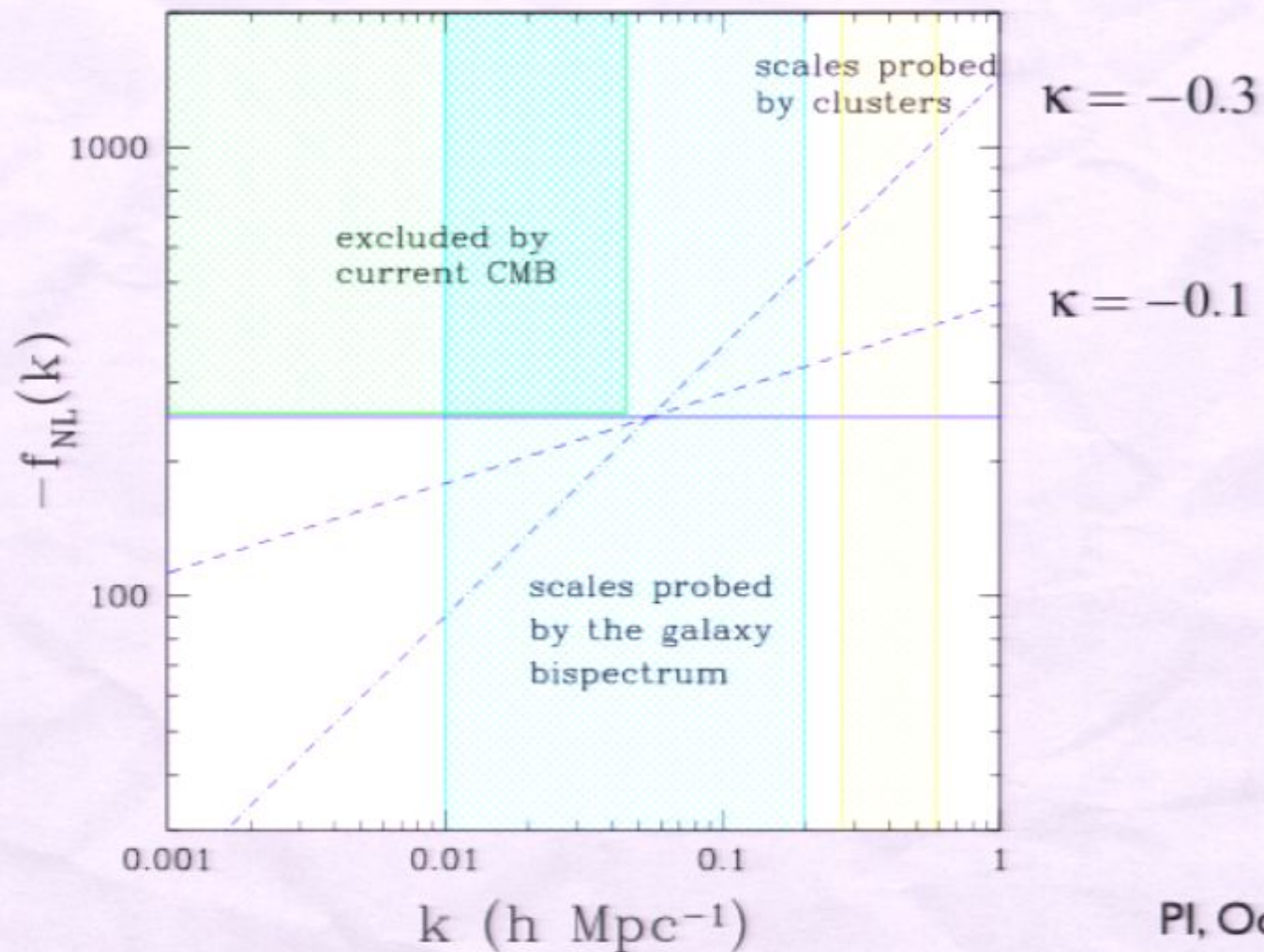
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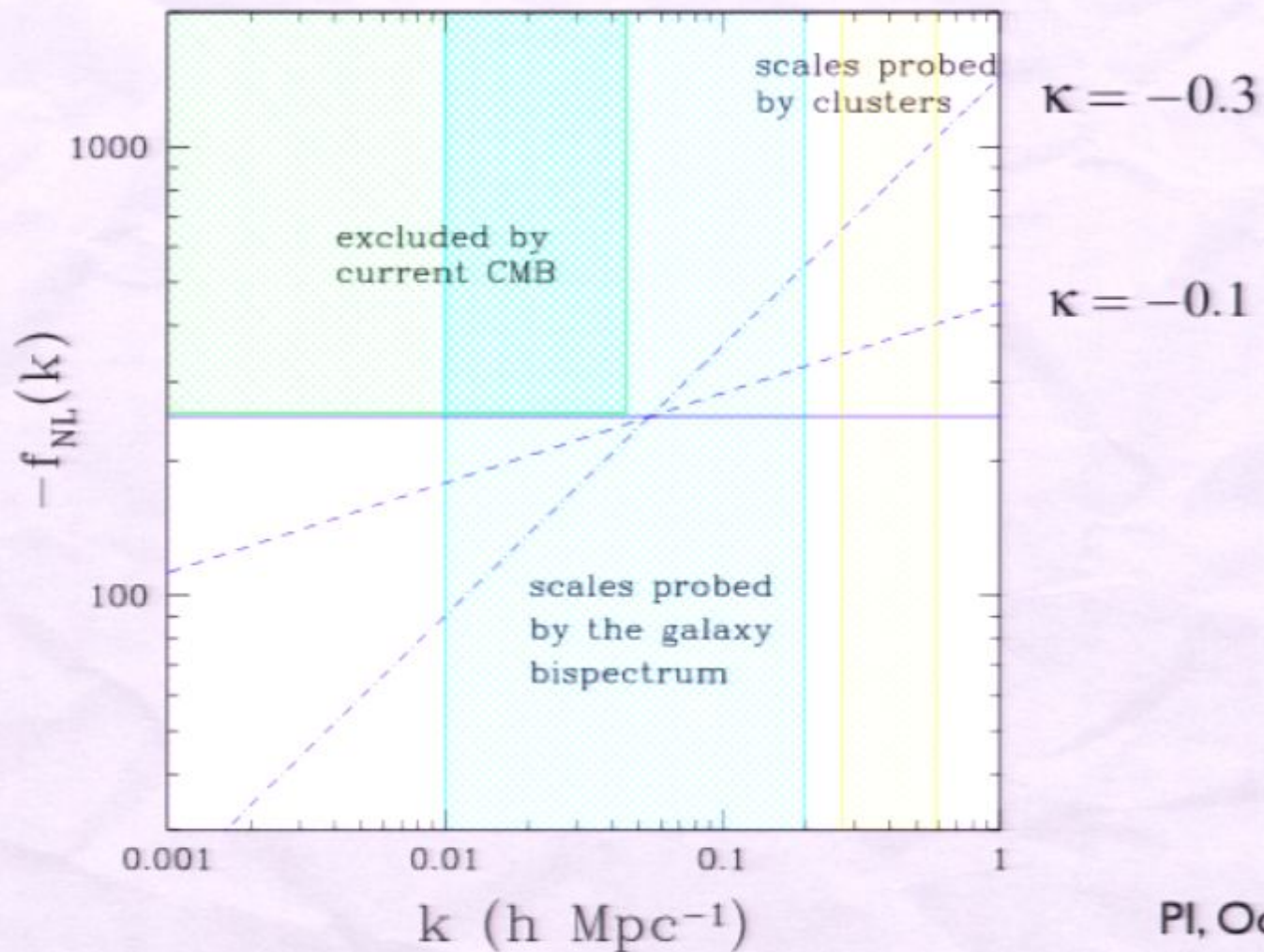
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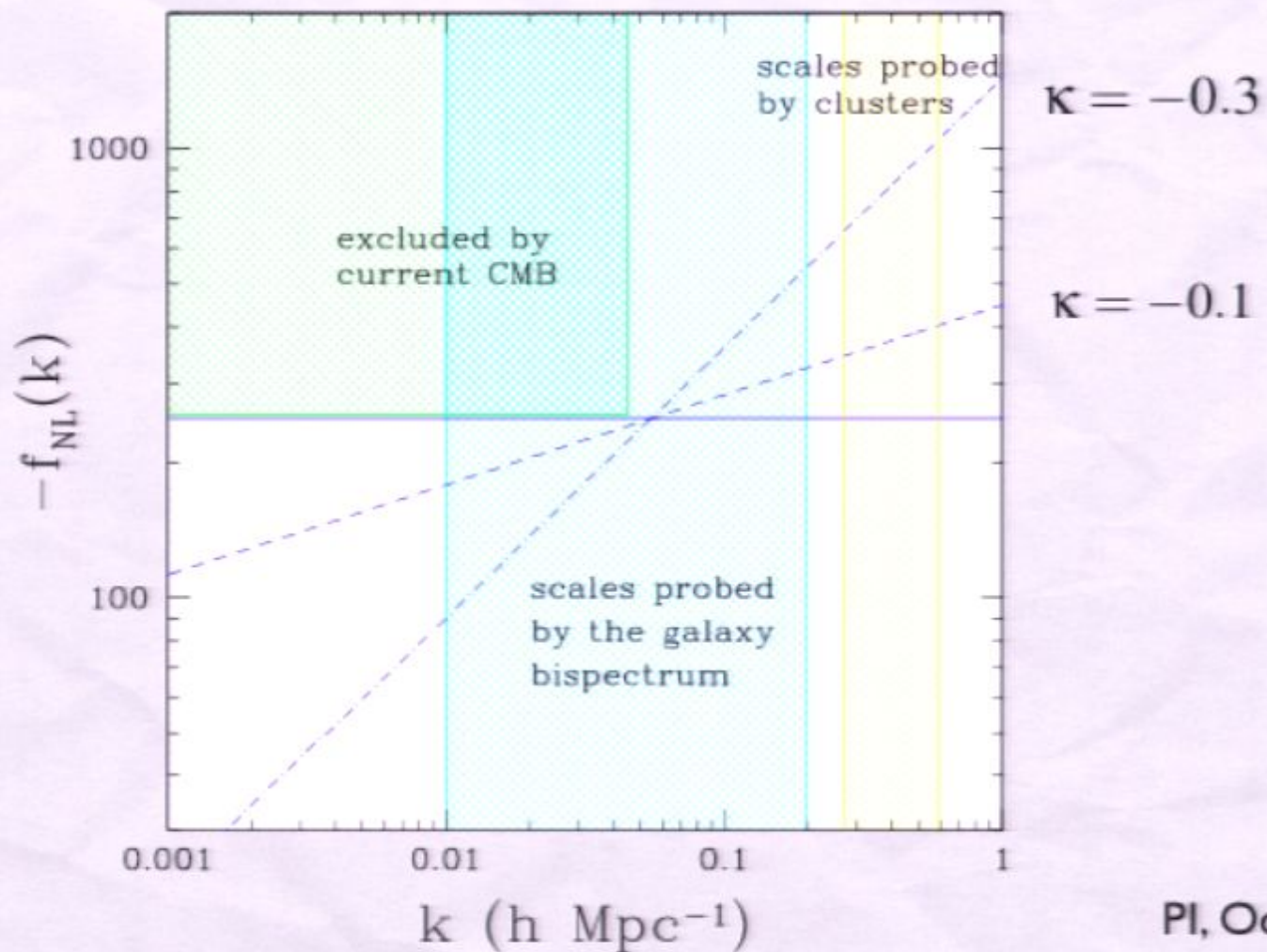
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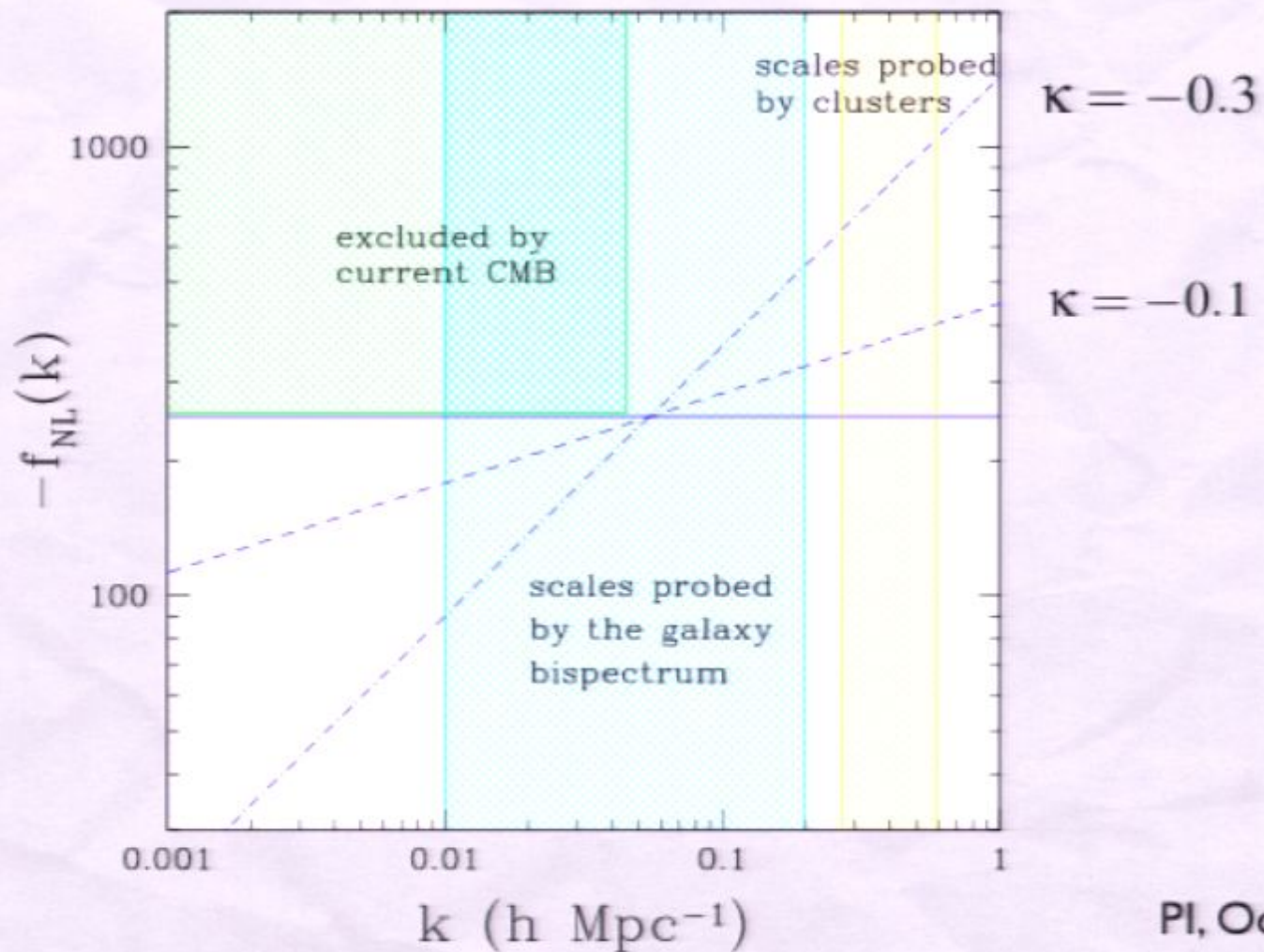
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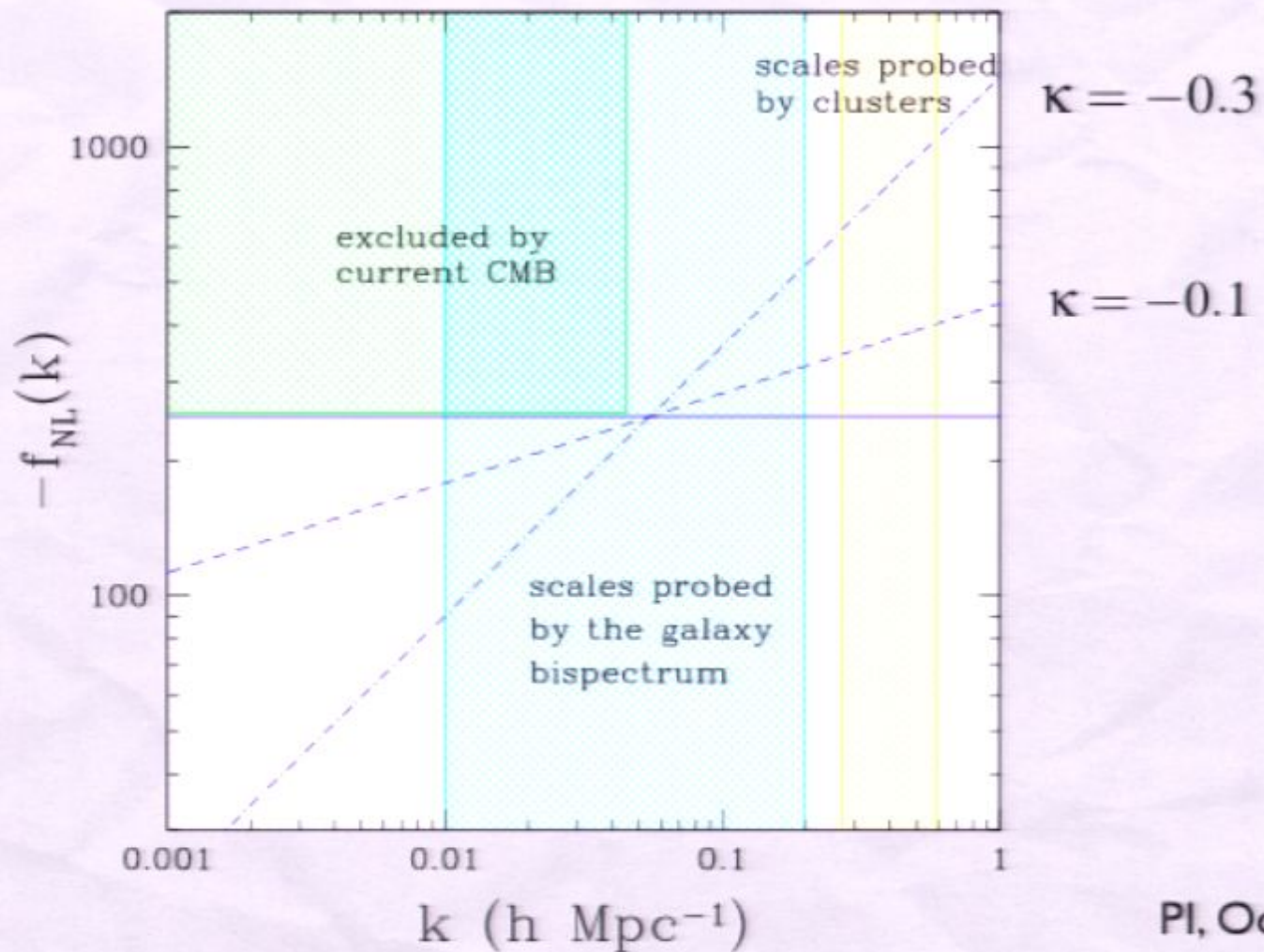
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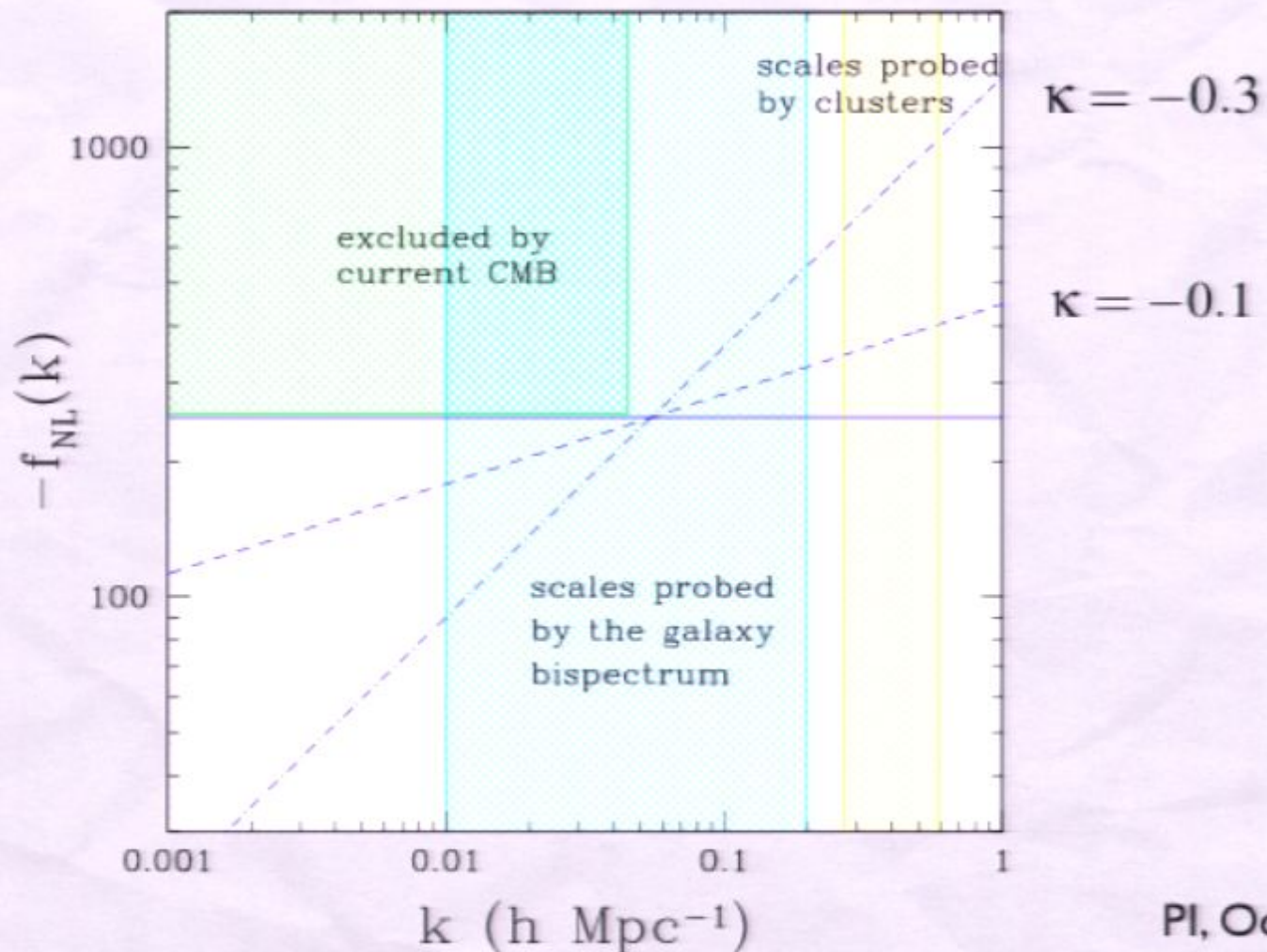
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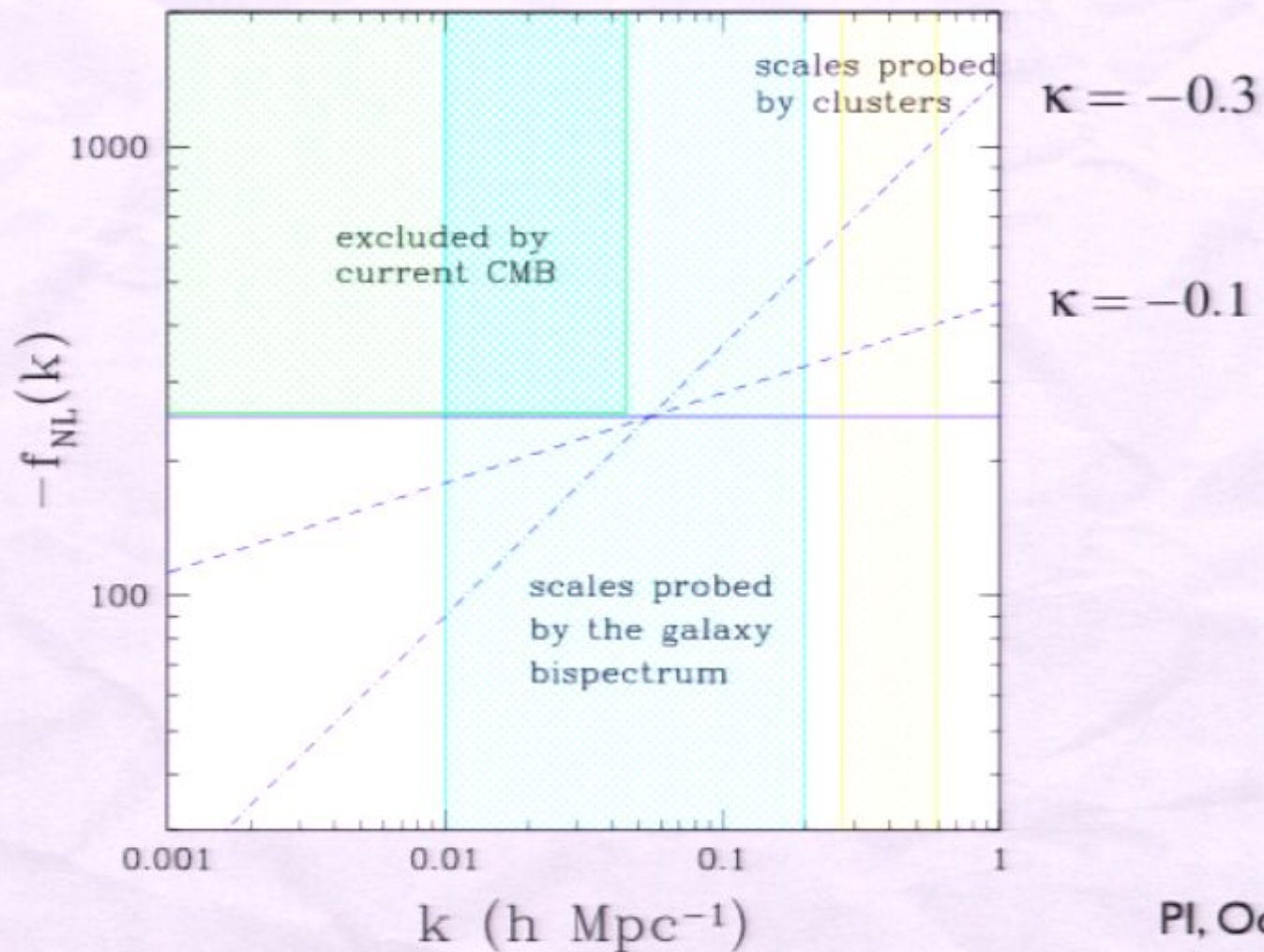
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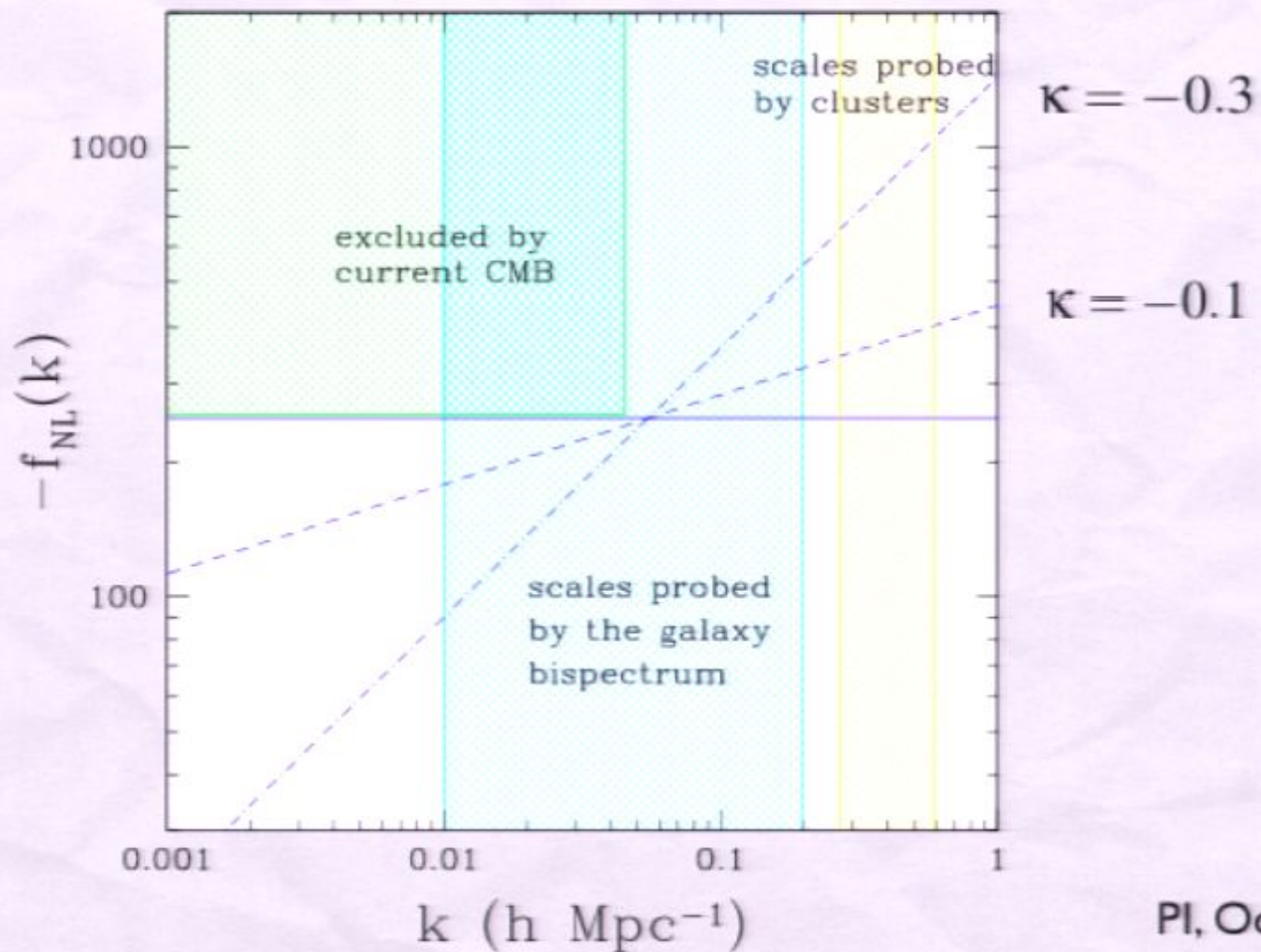
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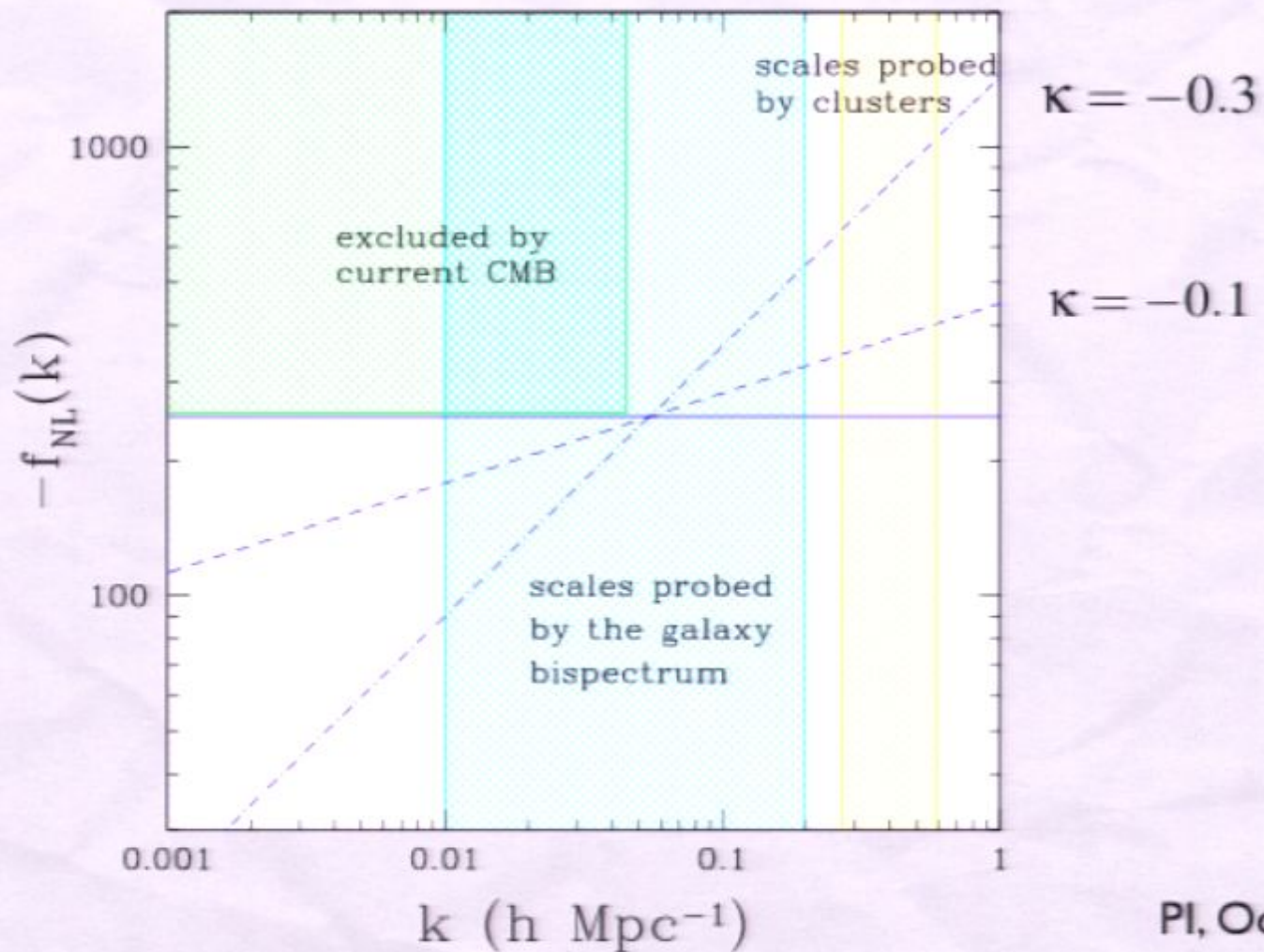
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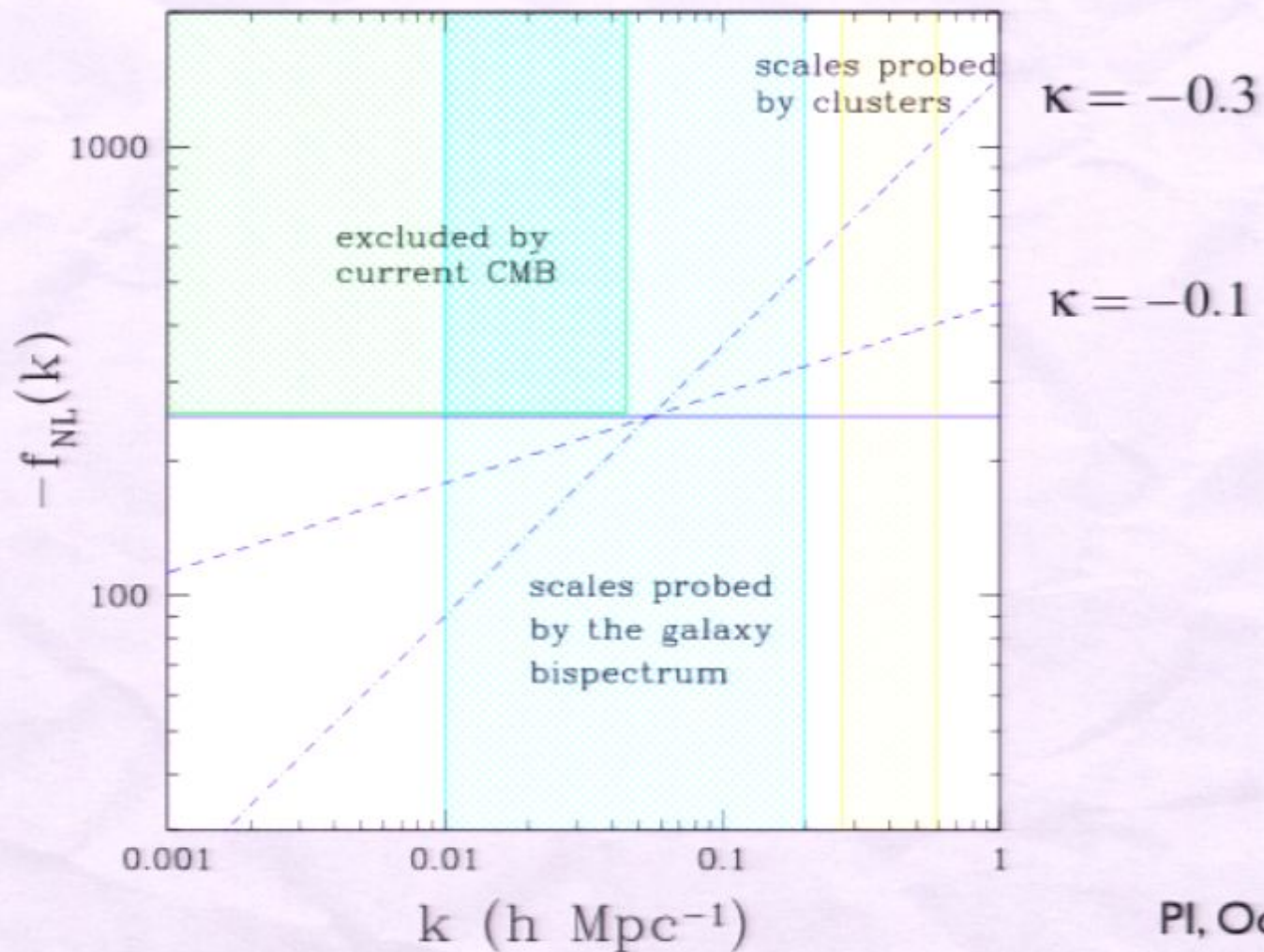
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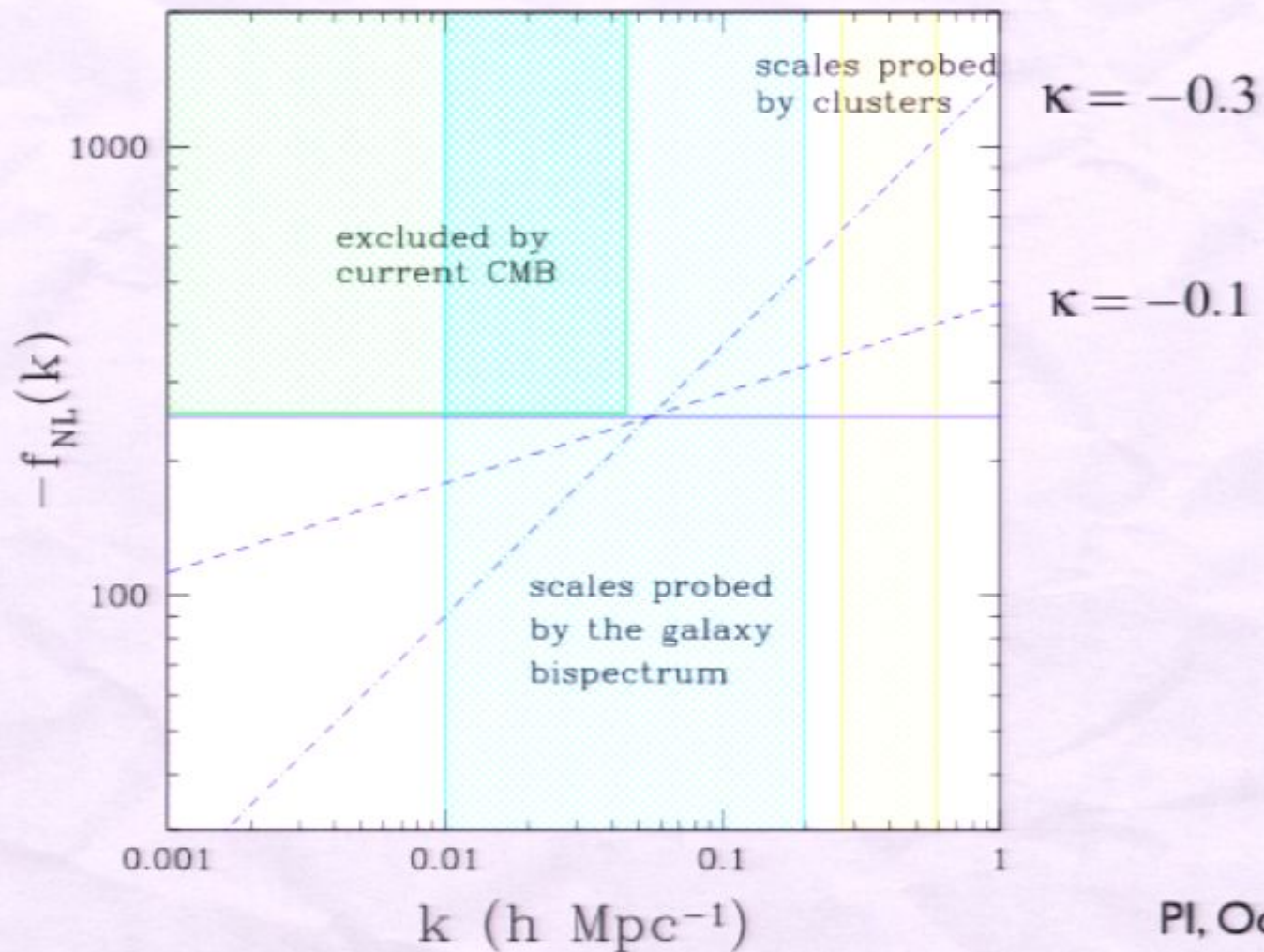
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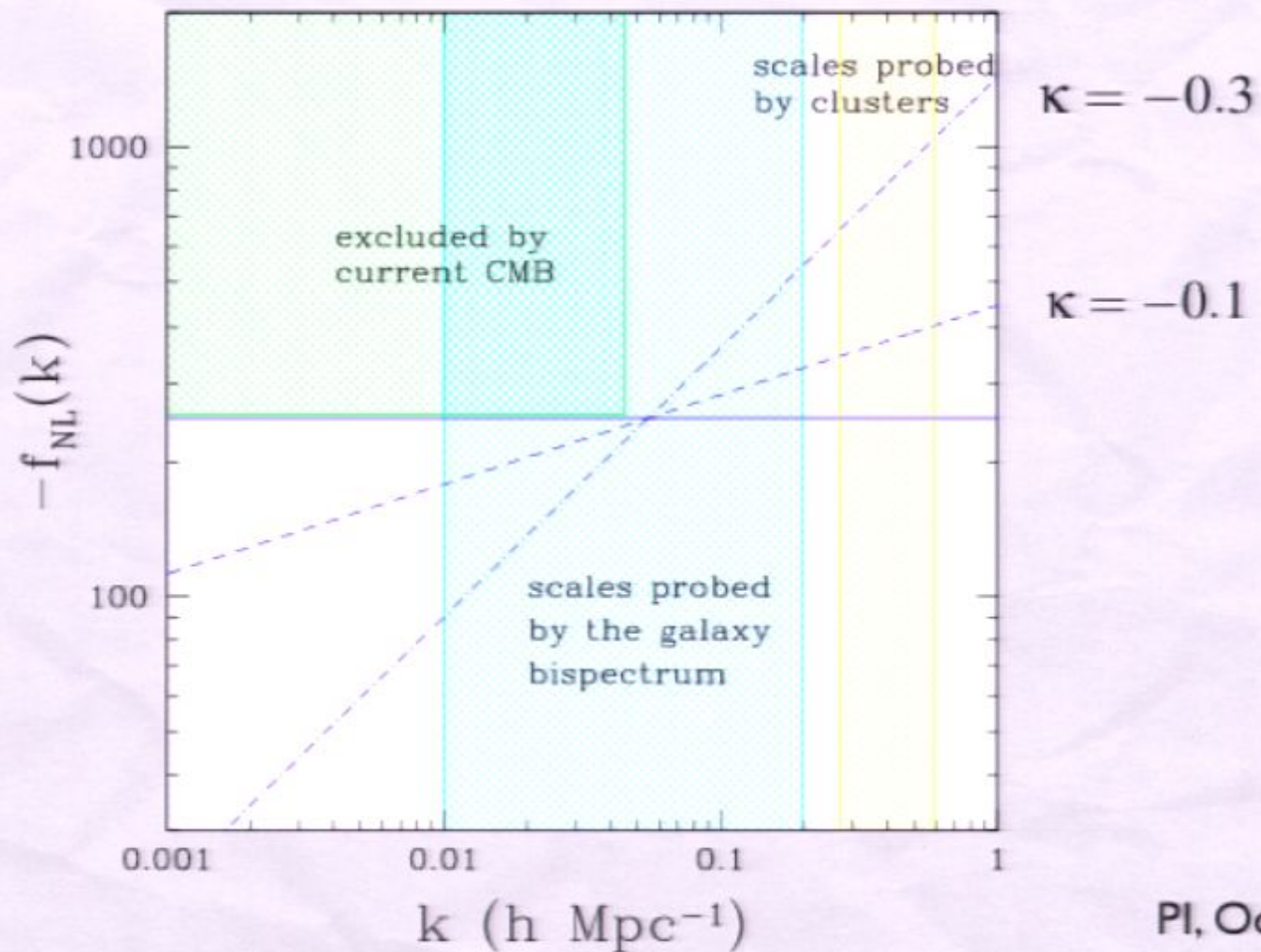
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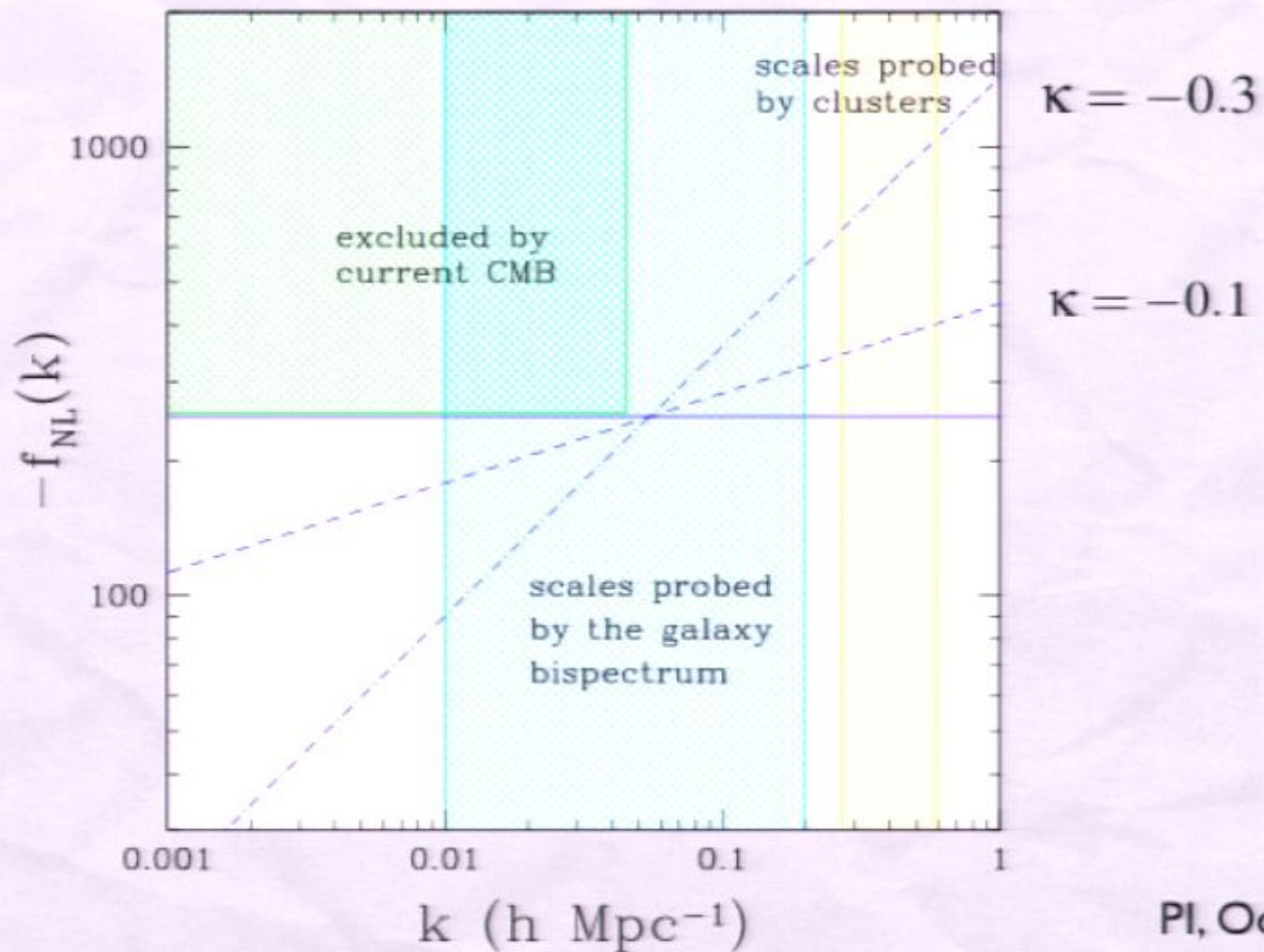
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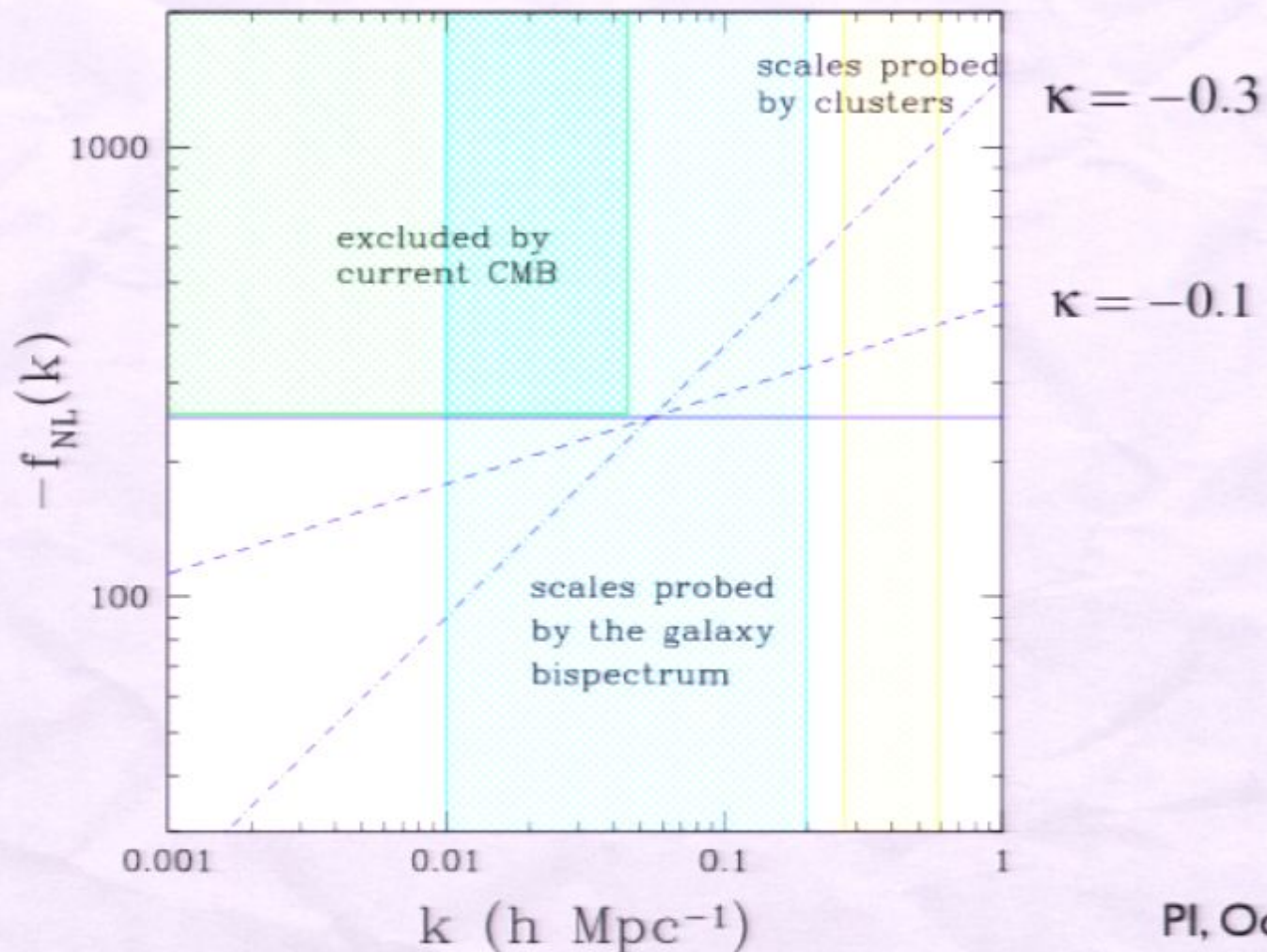
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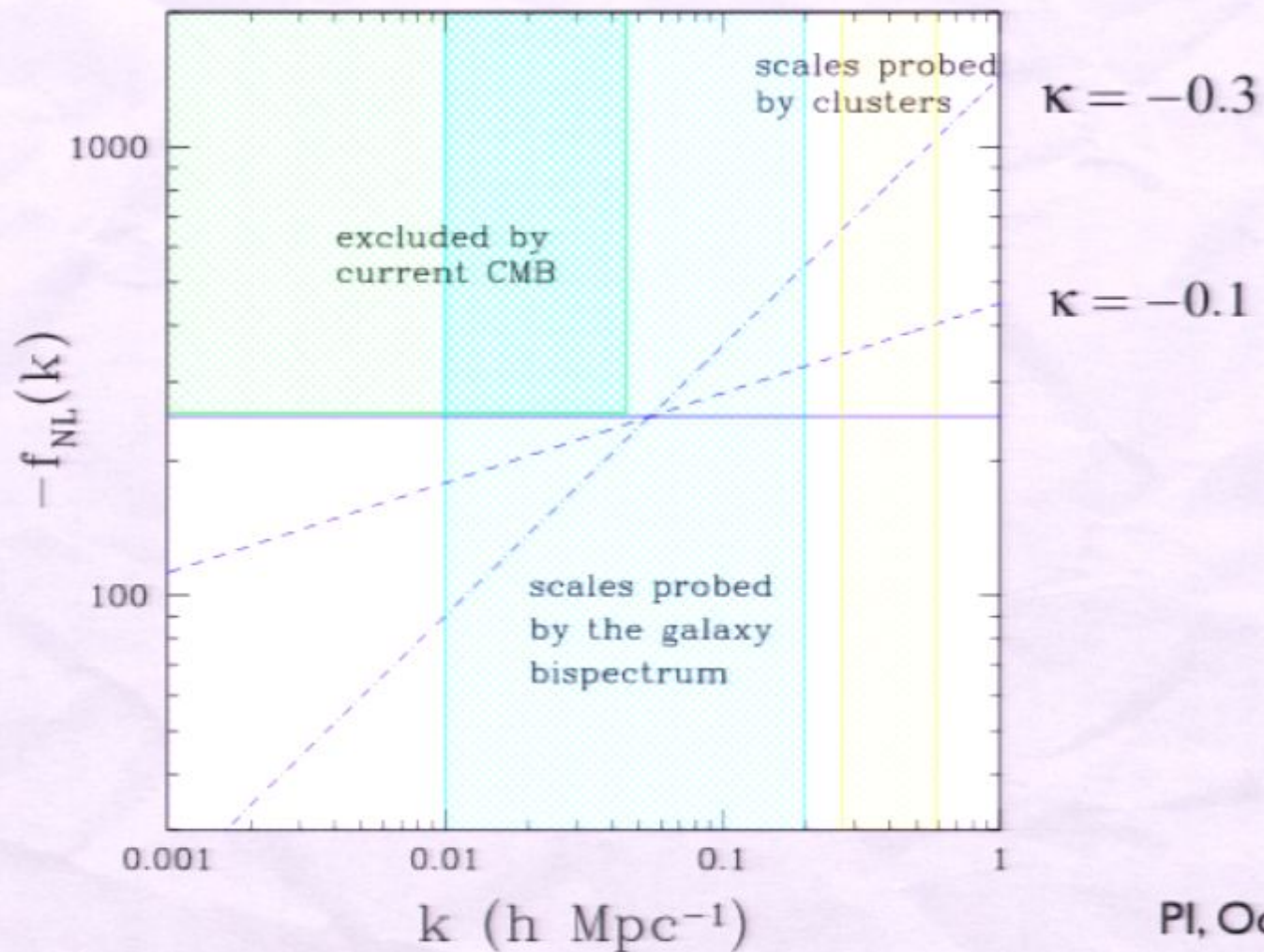
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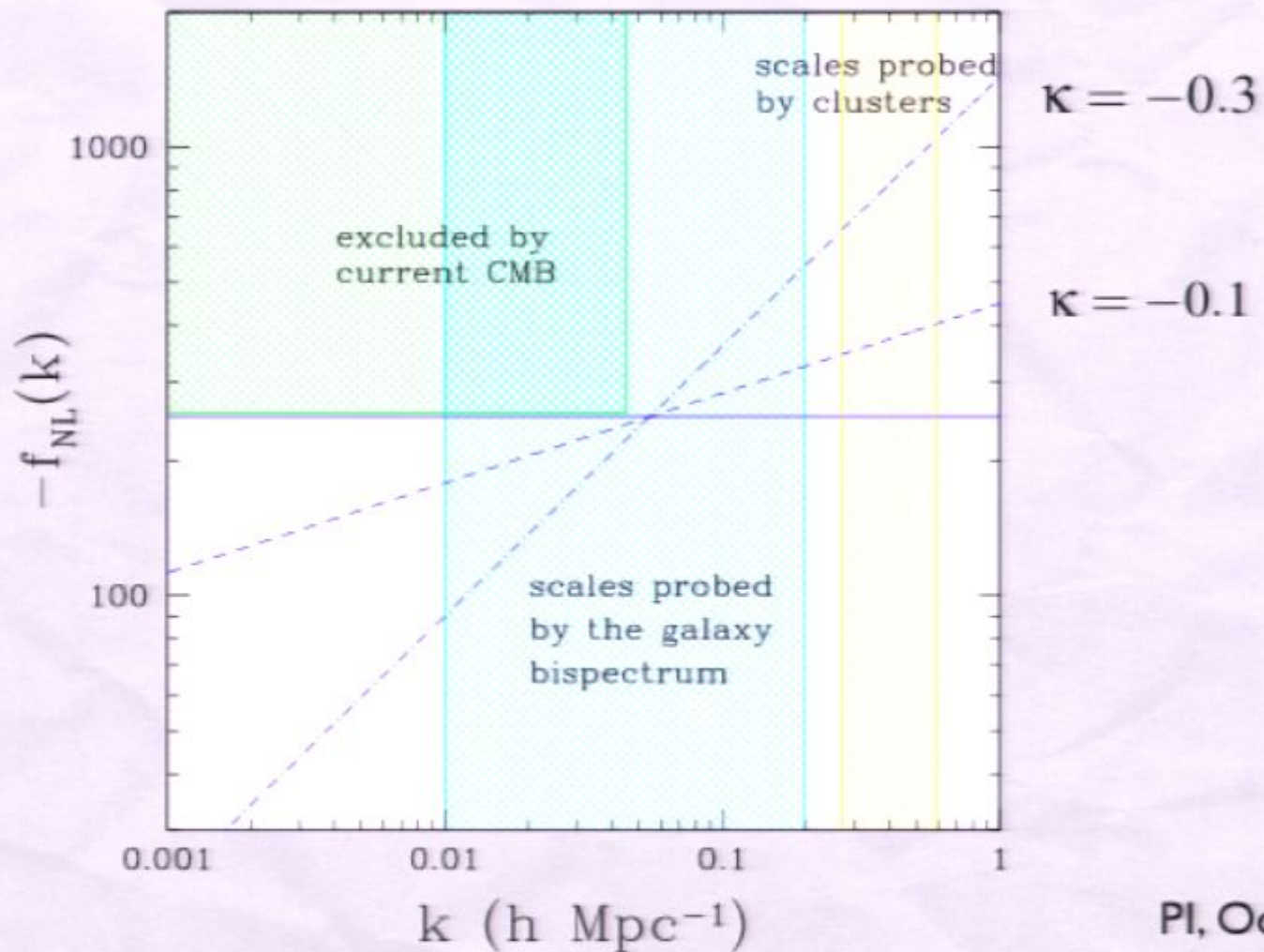
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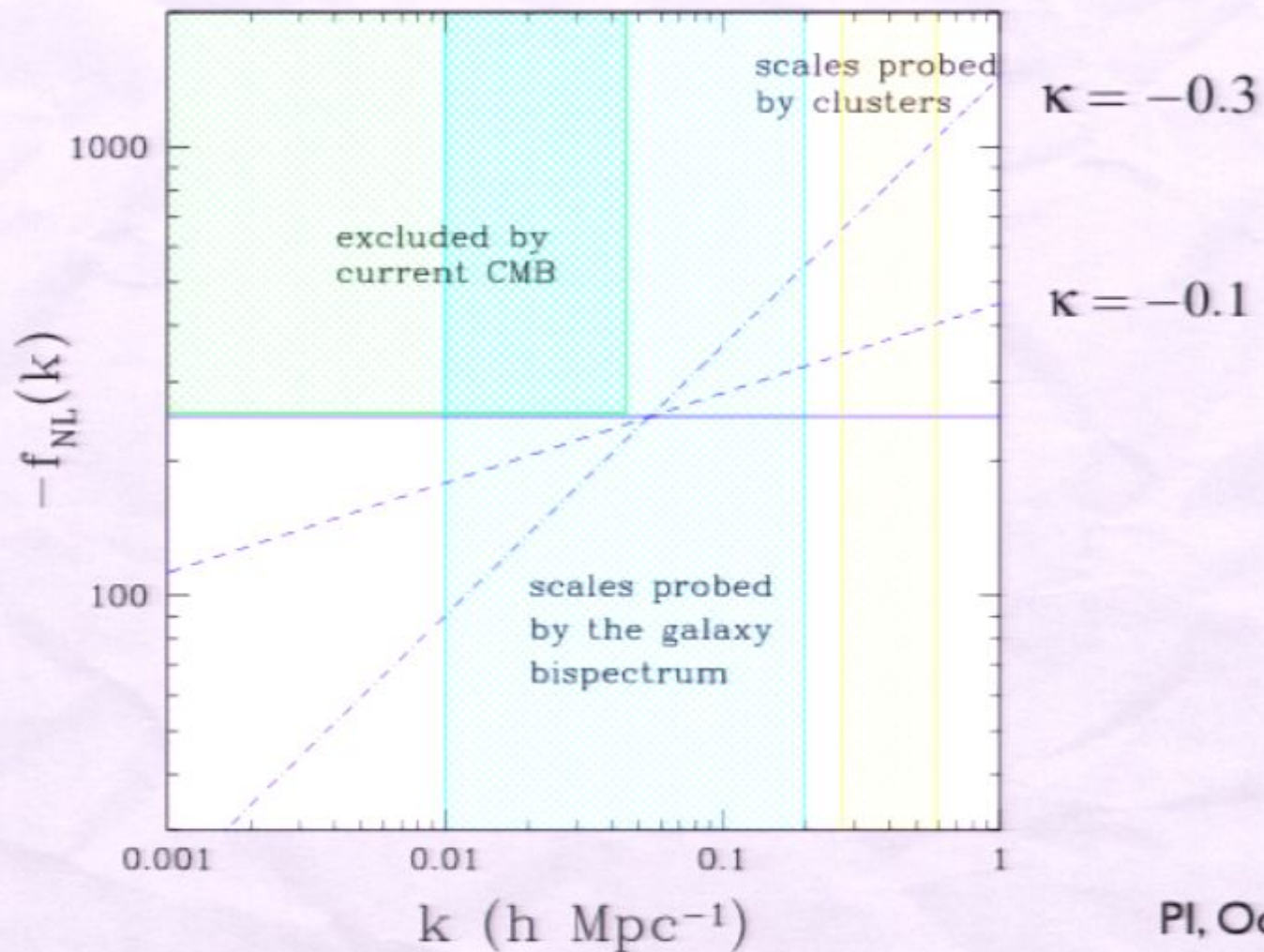
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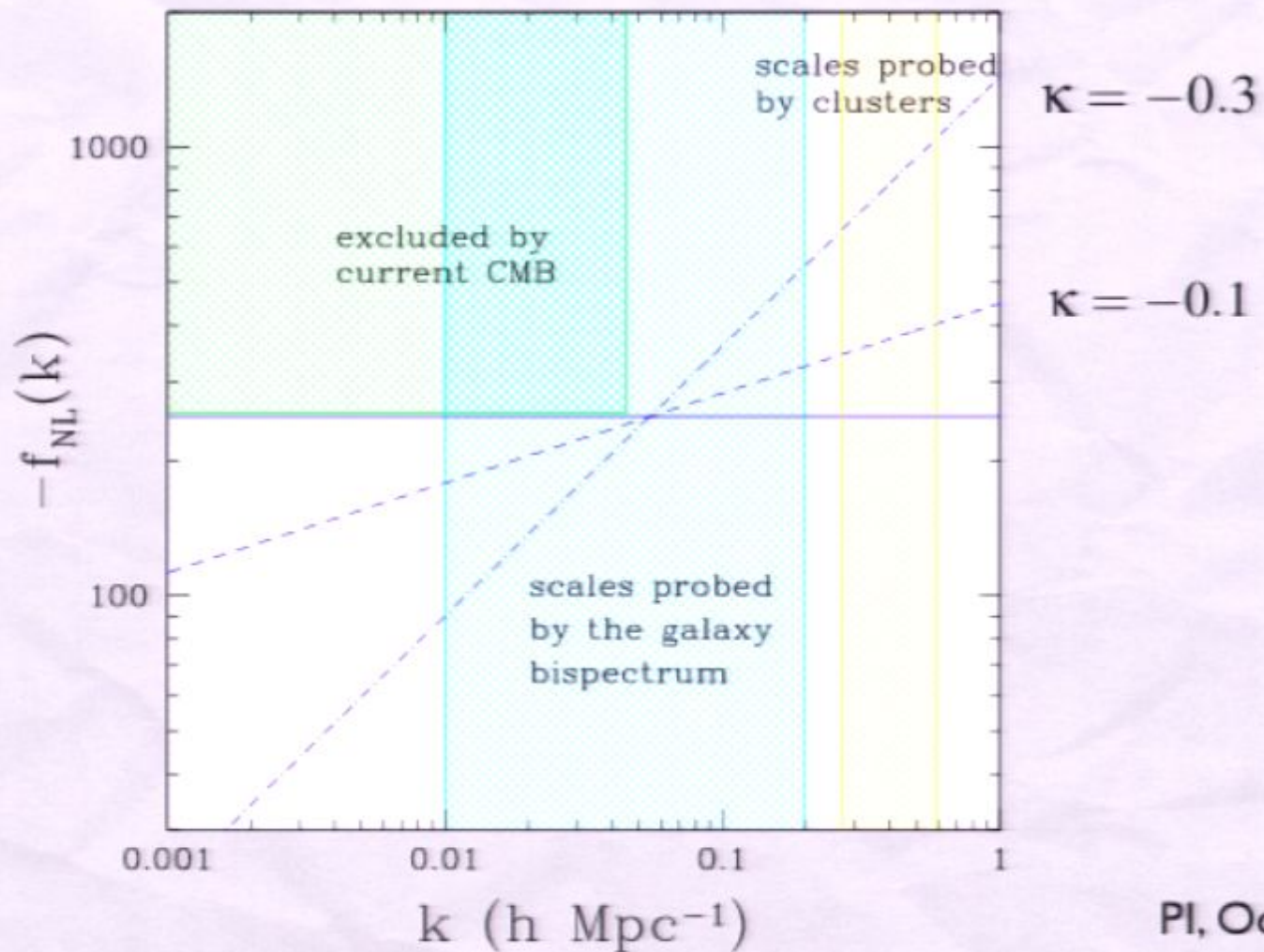
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III. OBSERVATIONS: CLUSTERS

TO PREDICT CLUSTER COUNTS

- Want: number density n of clusters with mass M and redshift z .
- Need:
 - Primordial \rightarrow Matter $\delta(k, a) \propto D(a)T(k)k^2\zeta(k)$
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$$f_c = 2P(> \delta_c, M) = 2 \int_{\delta_c}^{\infty} d\delta P(\delta, M)$$

- The *mass function* gives the number density of clusters with mass between M and $M+dM$, at redshift z :

$$\frac{dn}{dM}(M, z) = -2 \frac{\bar{\rho}}{M} \frac{d}{dM} \left[\int_{\delta_c/\sigma(M)}^{\infty} dv P(v, M) \right]$$

- Adjust to a more accurate collapse model (e.g. Sheth-Torman)

$$\frac{dn_{NG}}{dM}(M, z) = \frac{dn_G}{dM}(M, z) \frac{\frac{dn_{PS}}{dM}(S_3, M, z)}{\frac{dn_{PS}}{dM}(S_3 = 0, M, z)}$$

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S₃: skewness

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$$P(\delta_R)d\delta_R = \frac{d\delta_R}{2\pi i} \frac{1}{\sigma_R^2} \int_{-i\infty}^{i\infty} dy \exp \left[\frac{y\delta_R}{\sigma_R^2} - \frac{S(y)}{\sigma_R^2} \right]$$

Smoothed on scale R

- Using the cumulant generating function

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Smoothed on scale R

- Using the cumulant generating function

$$S(y) = \sum_{p=2}^{\infty} S_p(R) \frac{(-1)^{p-1}}{p!} y^p$$

PDF, CONTINUED

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(Smoothed)

$$v = \delta/\sigma$$

* Two kinds of approximations

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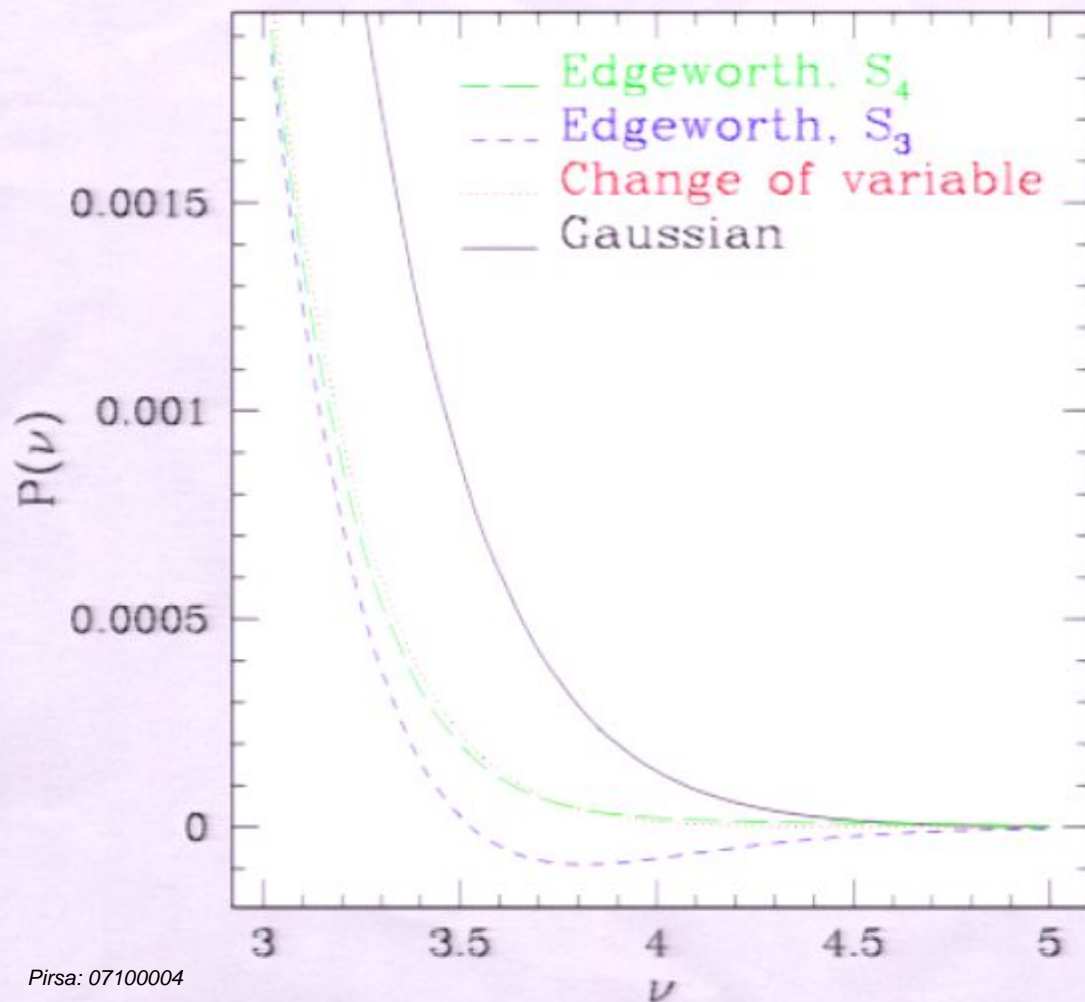
$$v = \delta/\sigma$$

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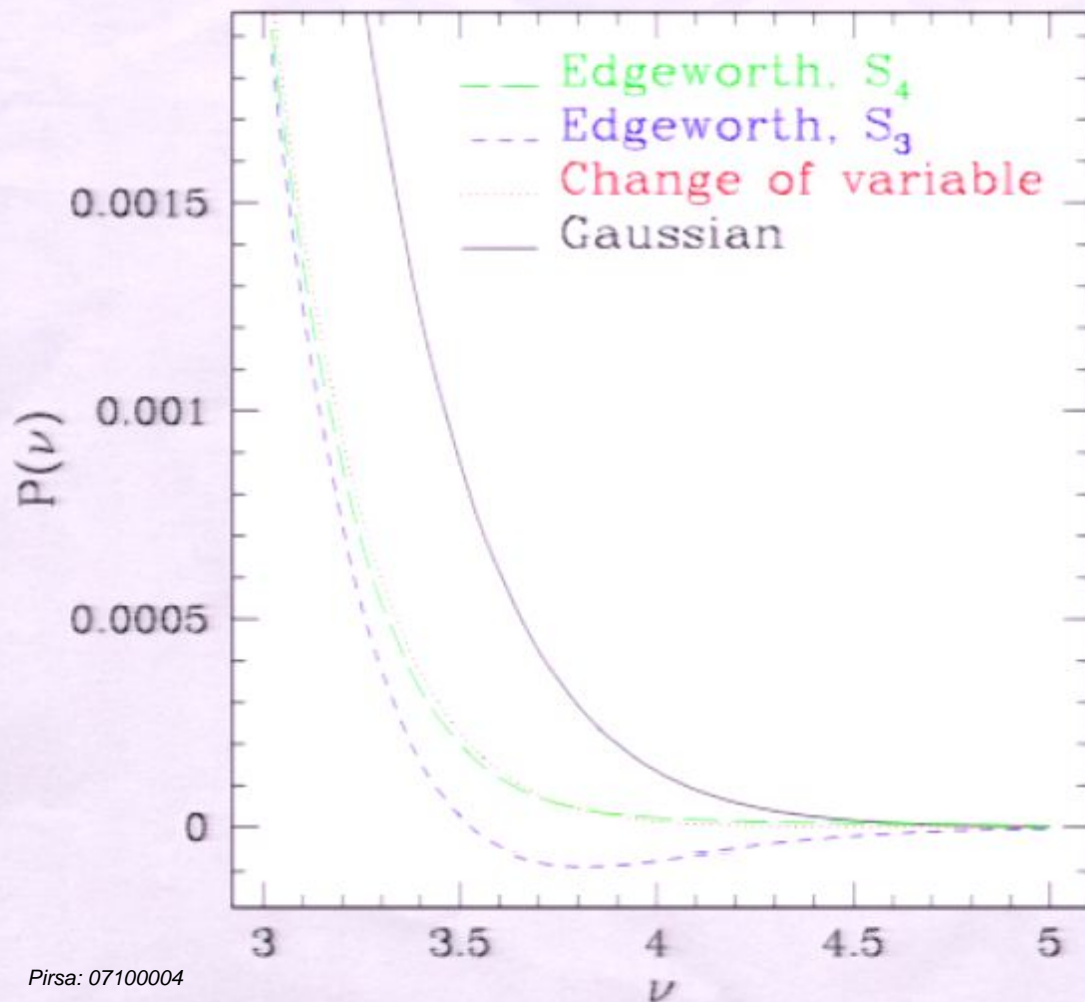
- size of v

COMPARING DISTRIBUTIONS (UNSMOOTHED, LOCAL MODEL)



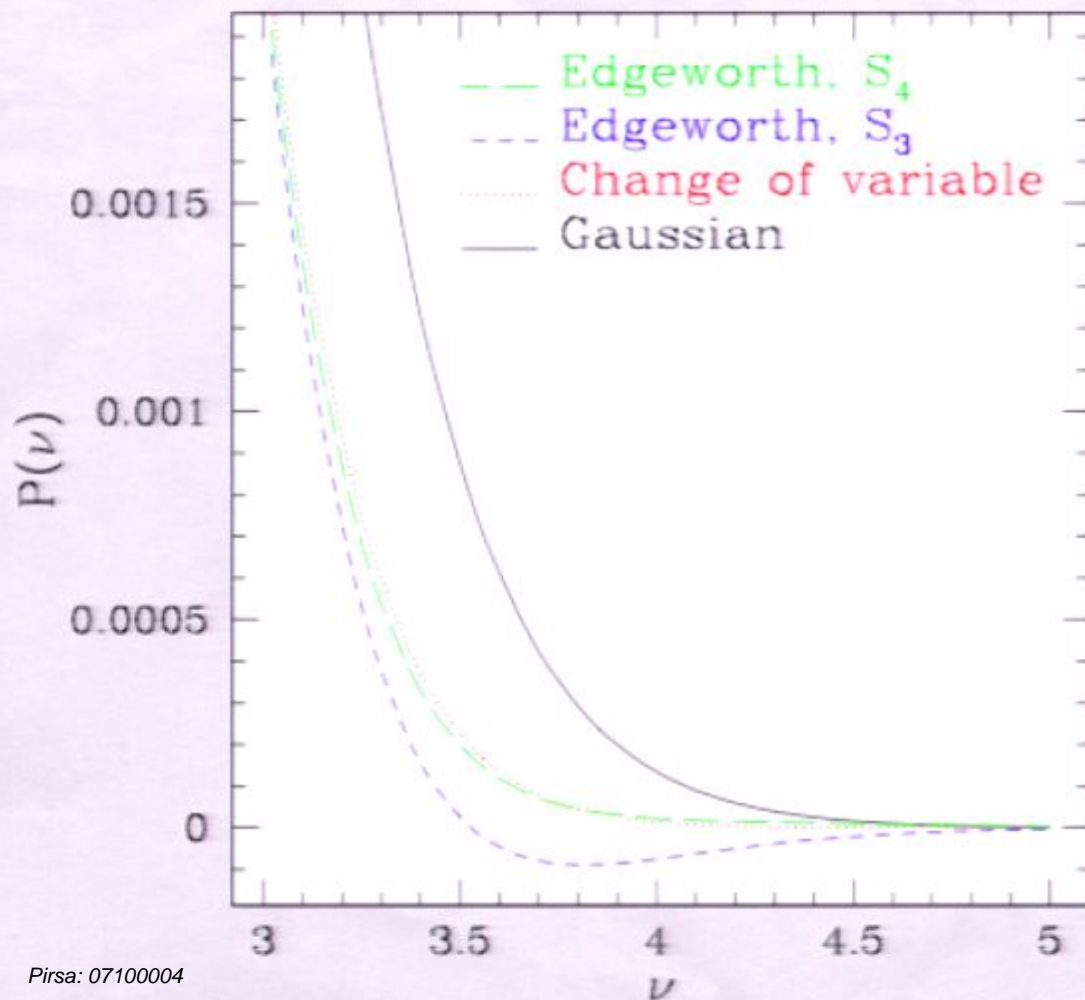
$$\sigma = 1$$
$$f_{NL} = -0.03$$

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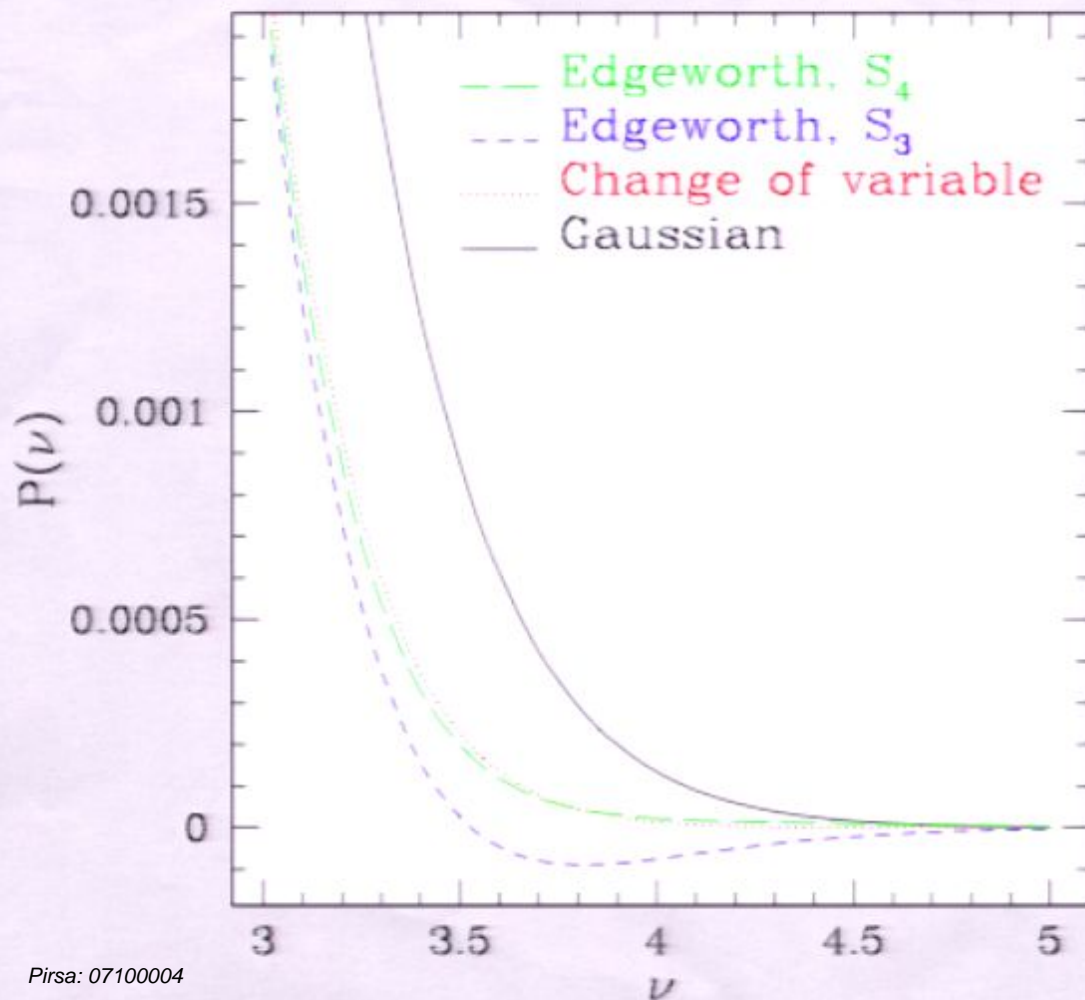
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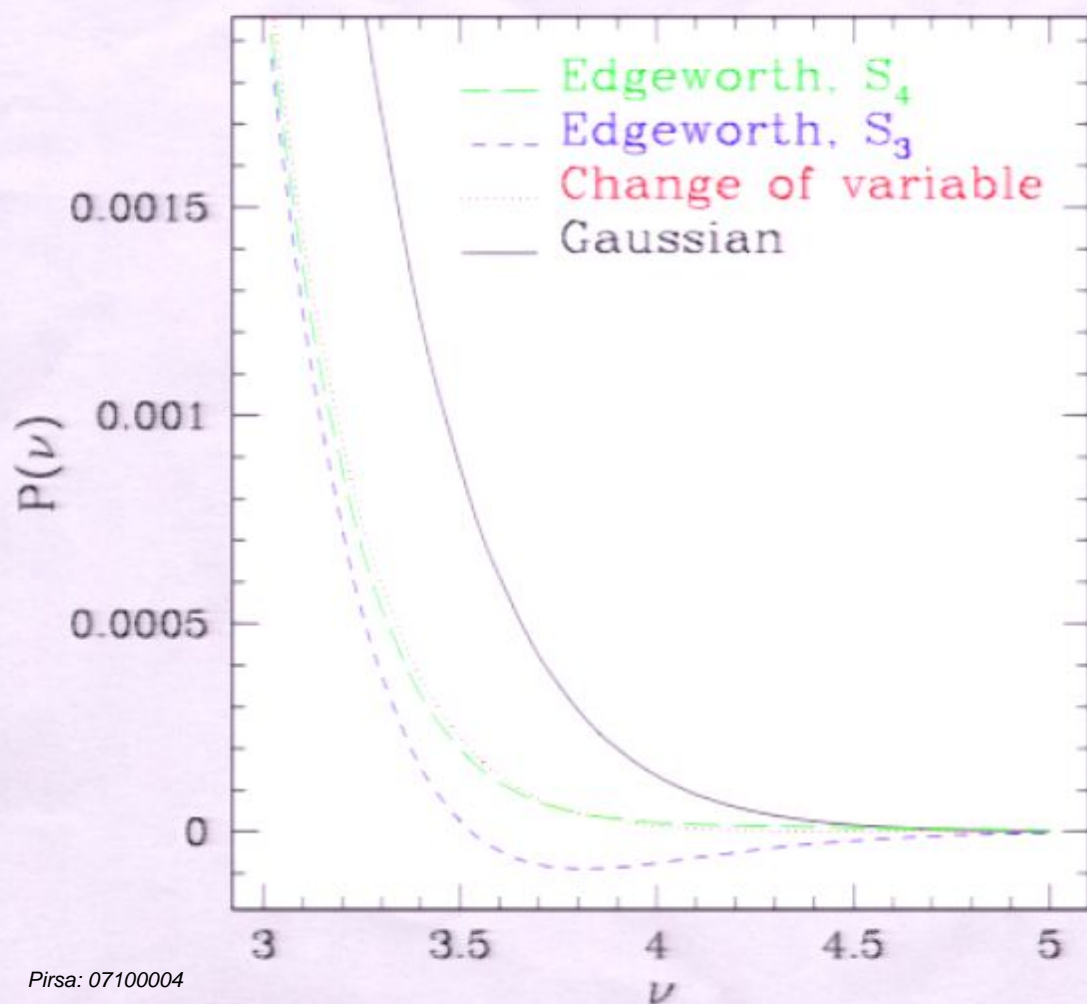
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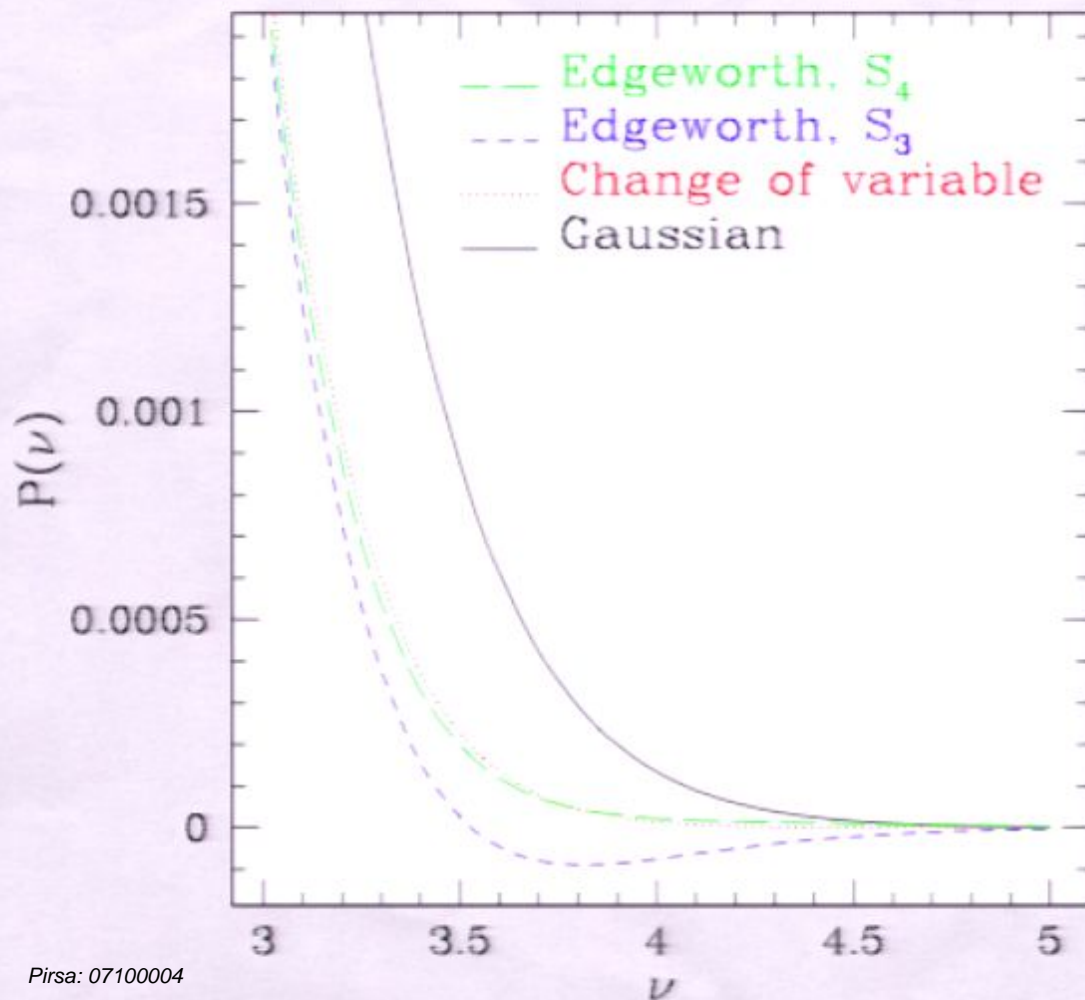
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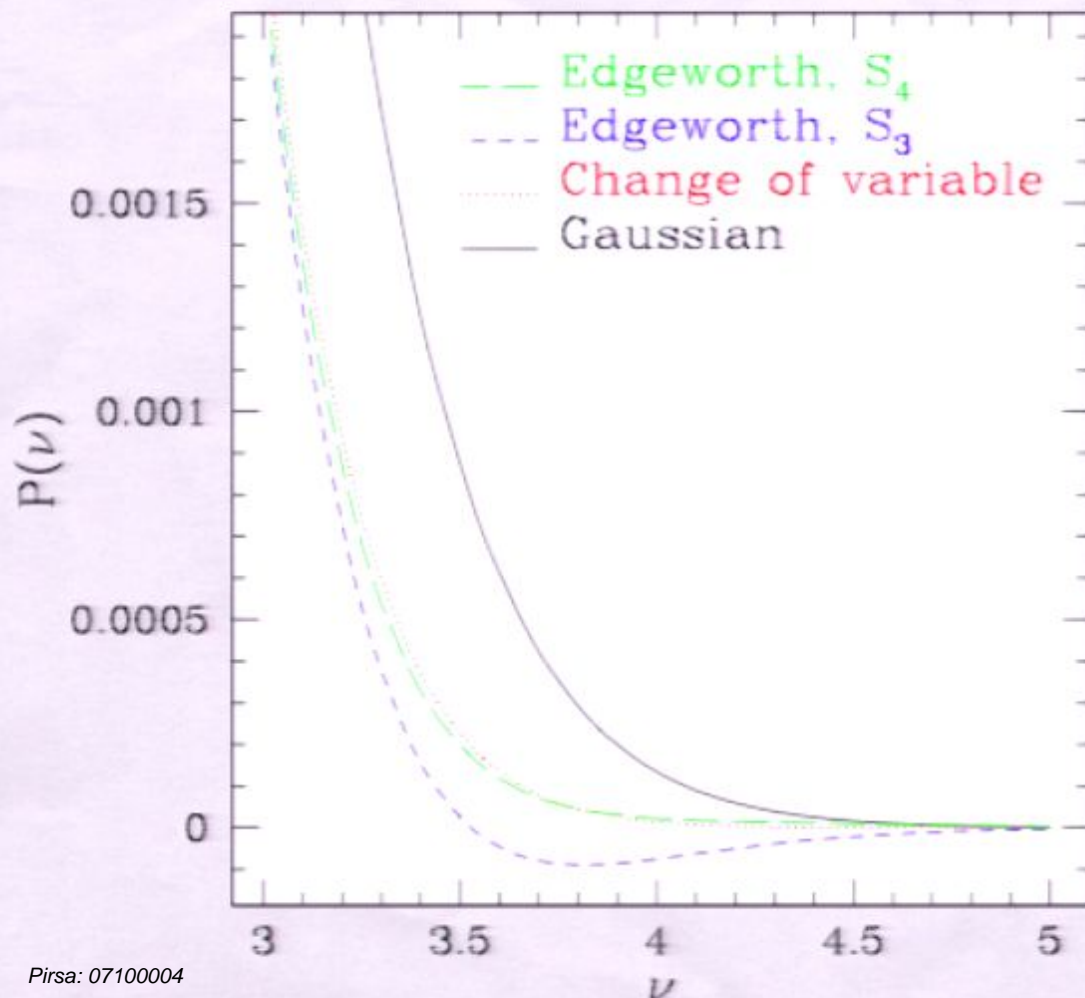
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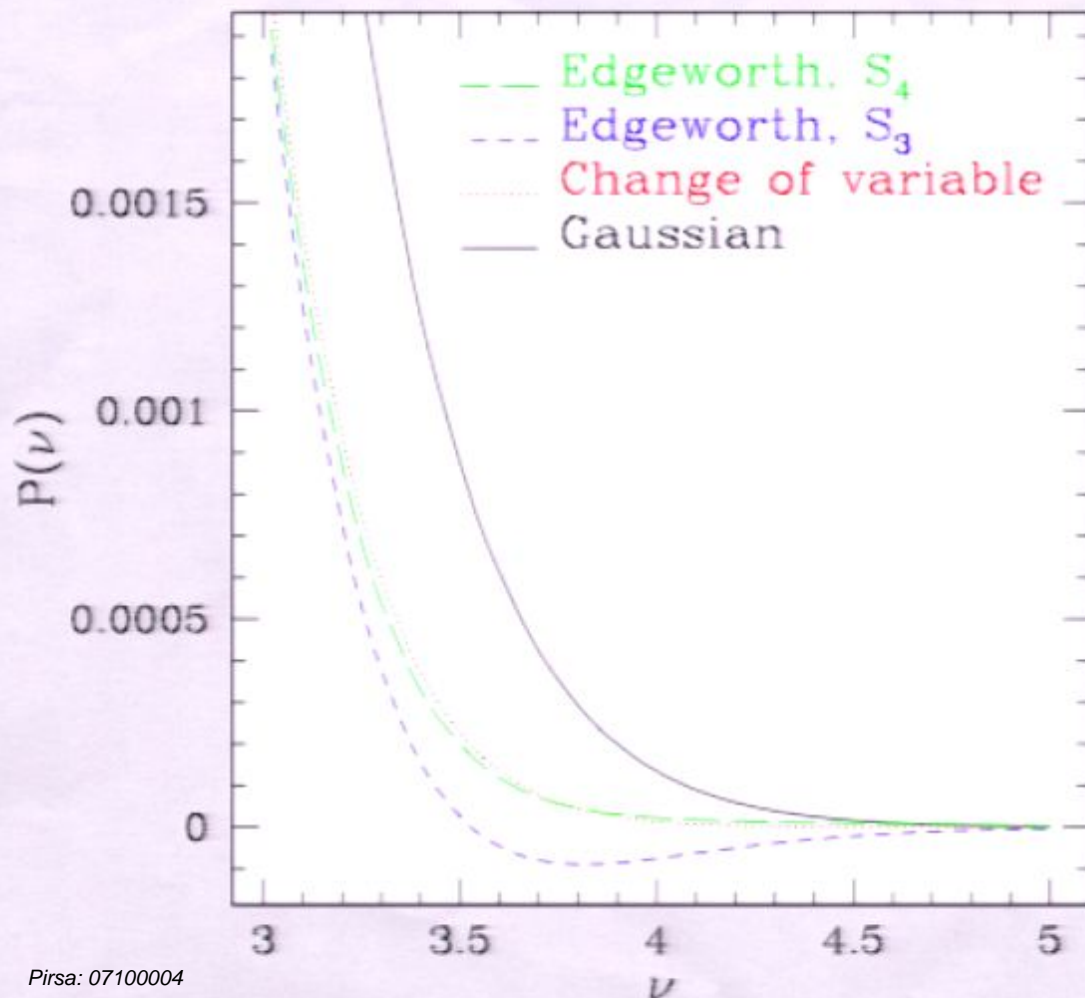
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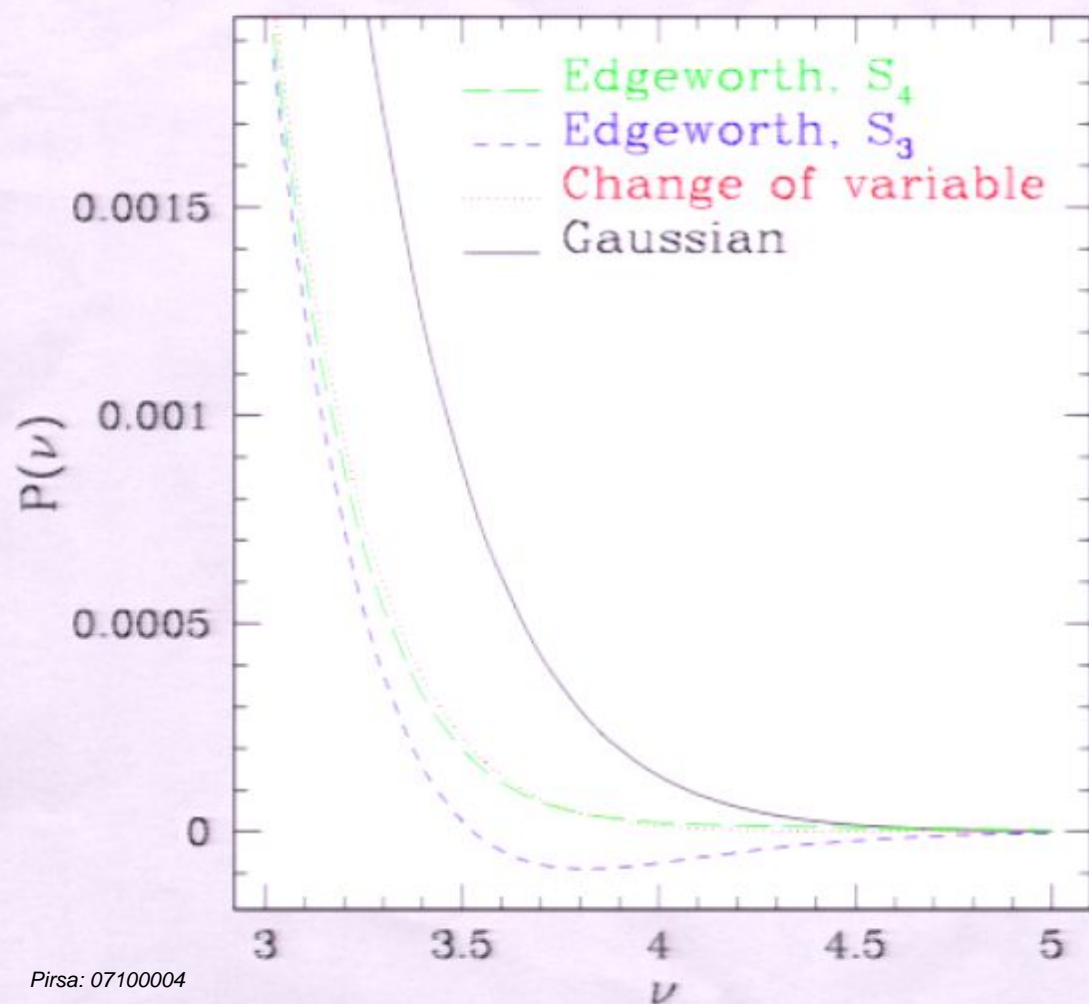
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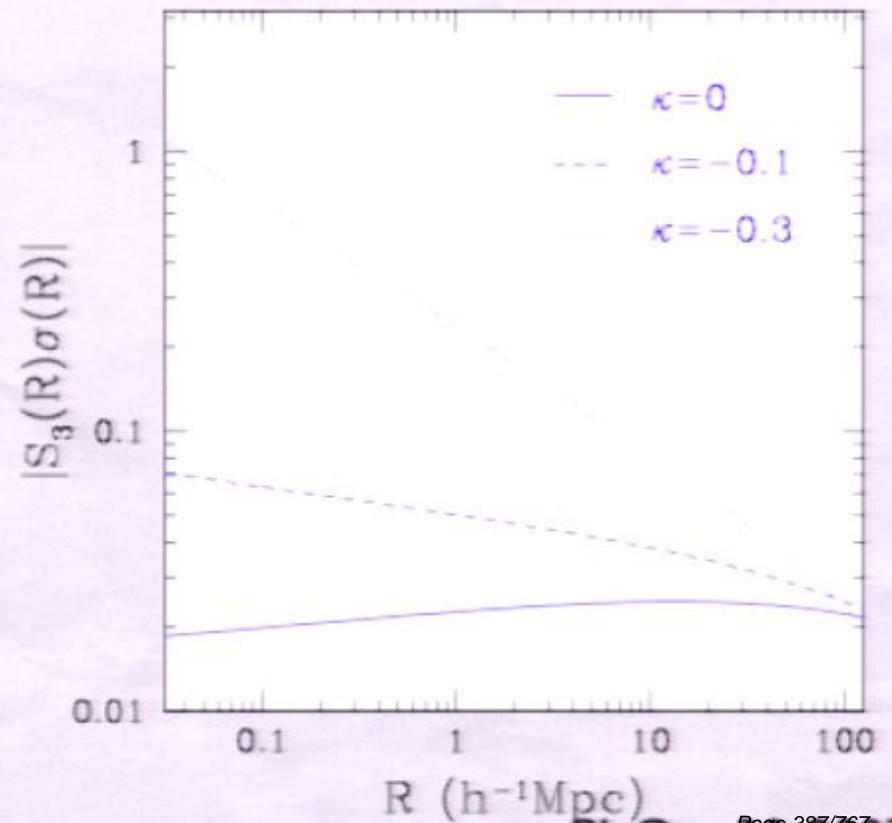
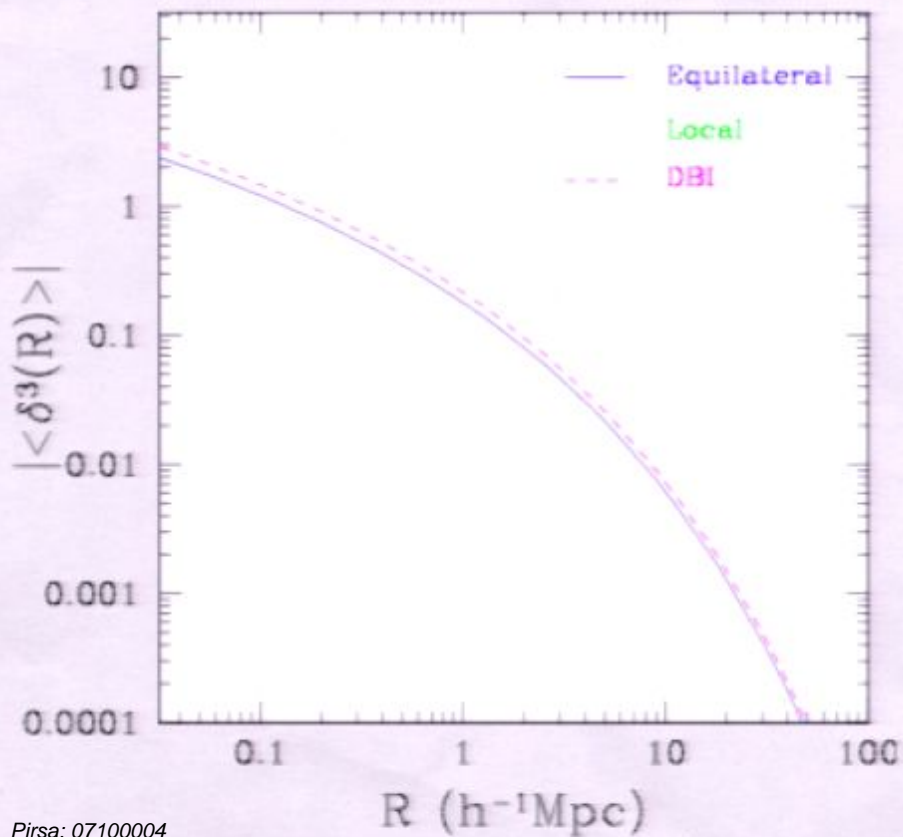
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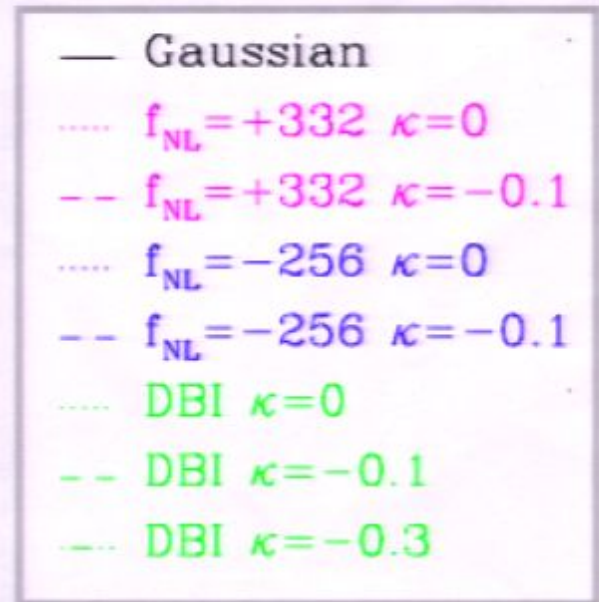
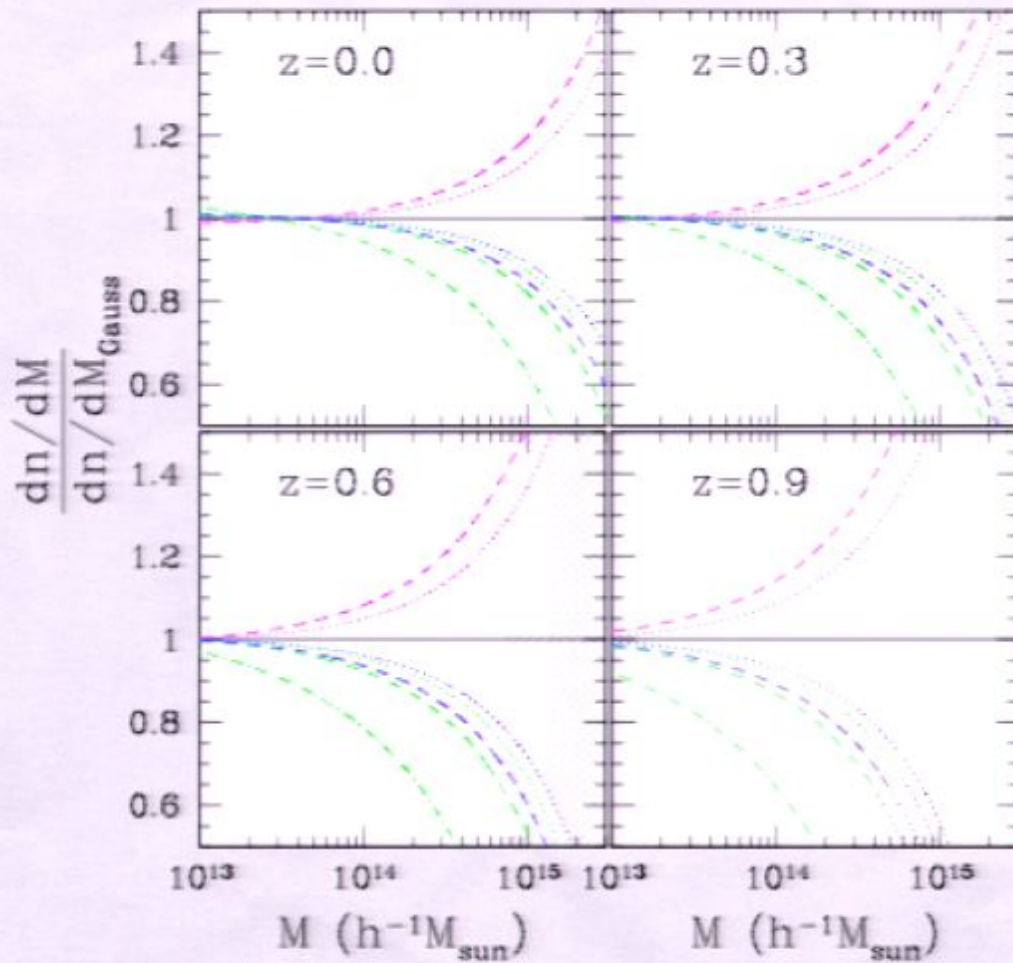


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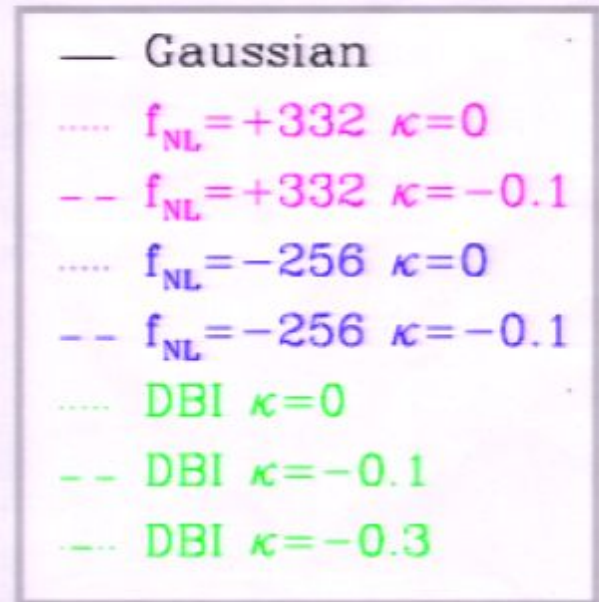
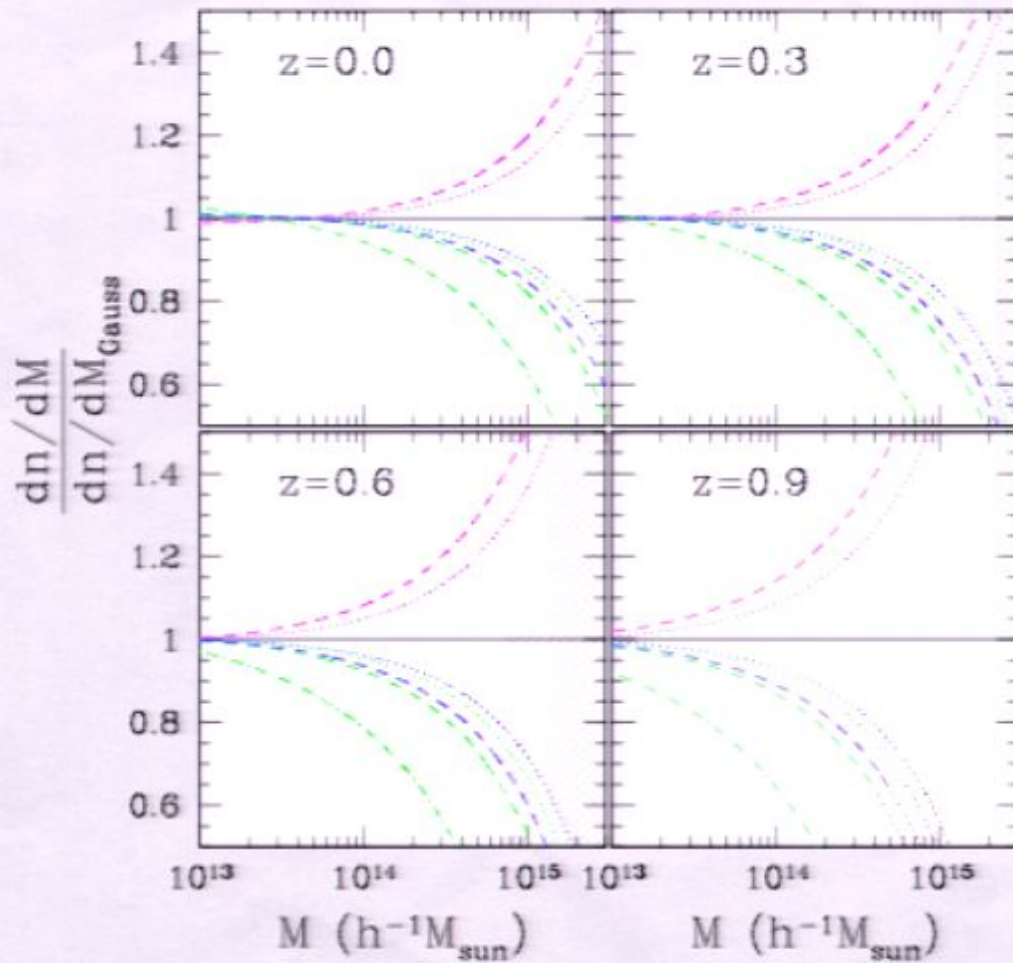
SMOOTHED SKEWNESS



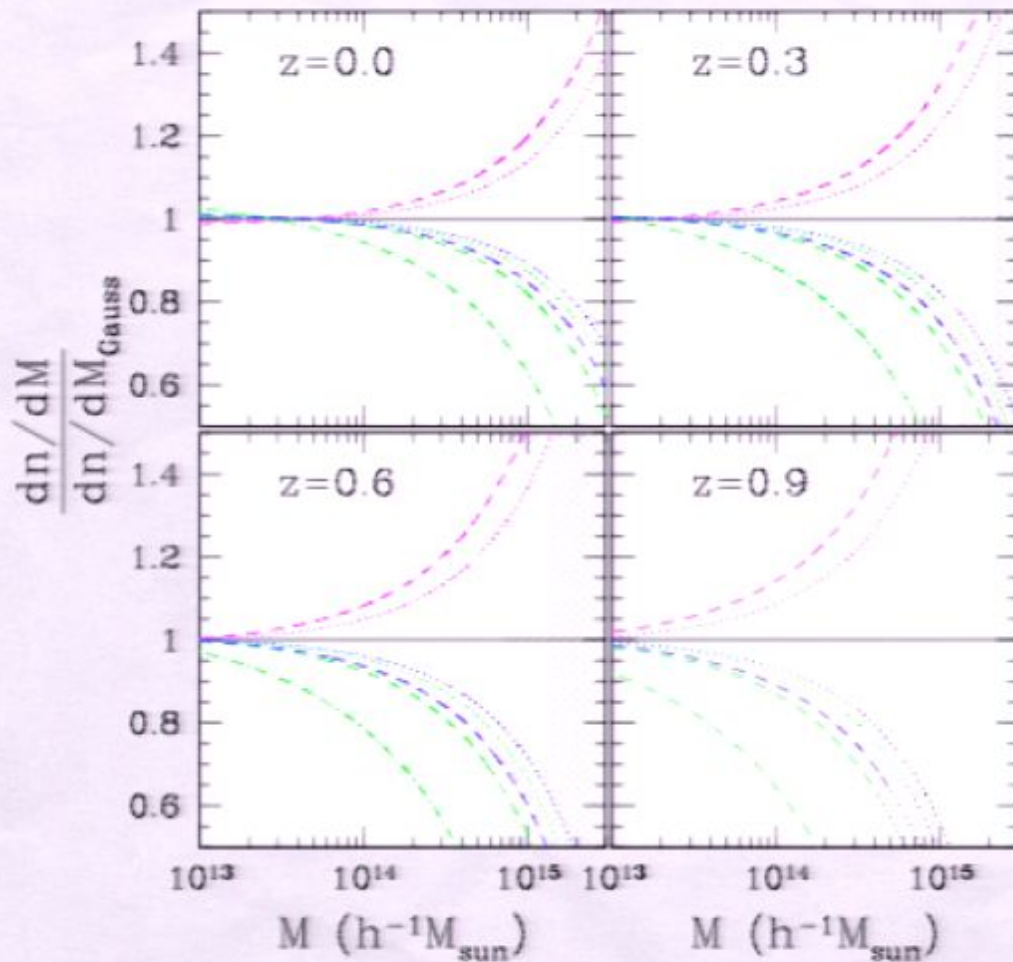
MASS FUNCTION



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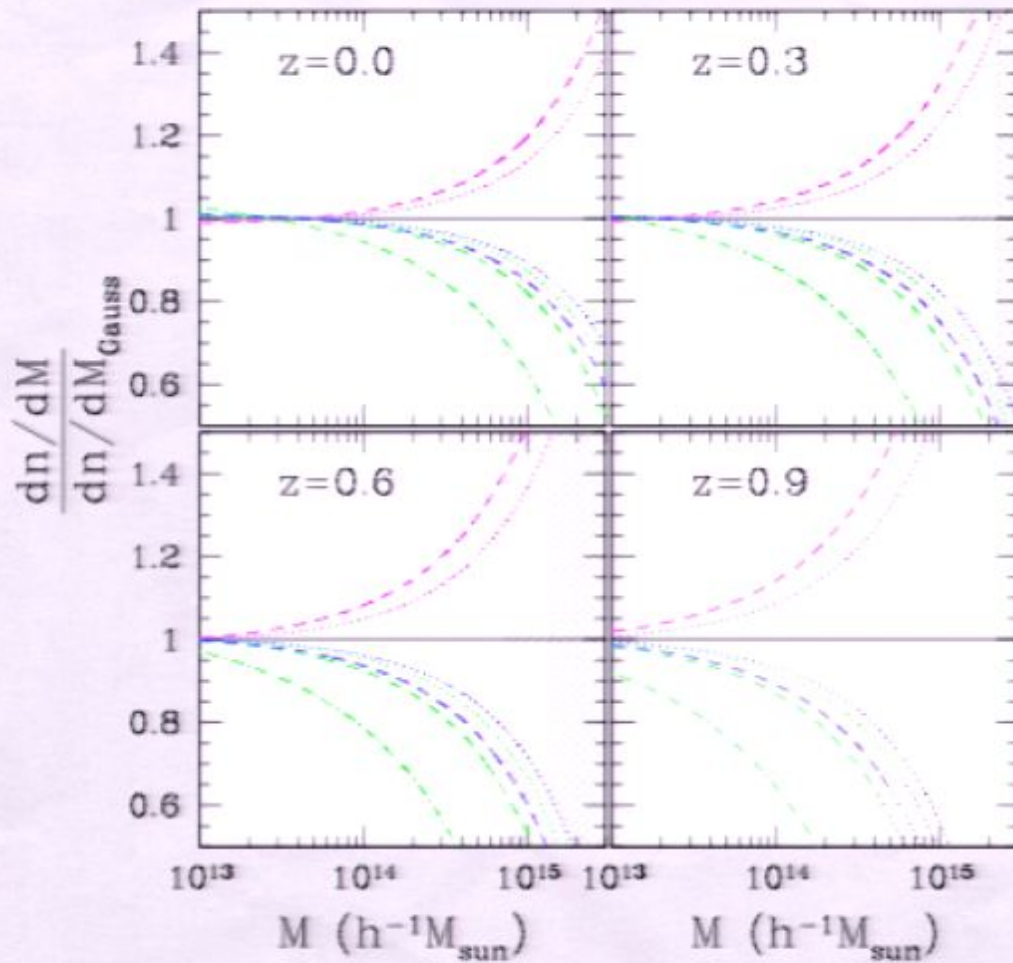


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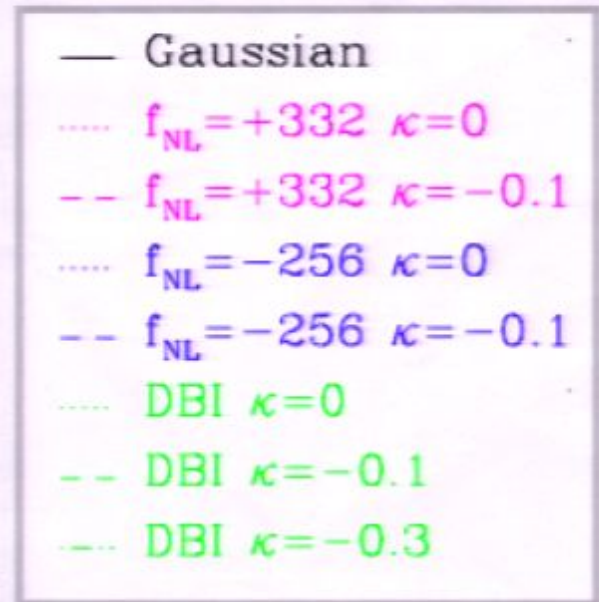
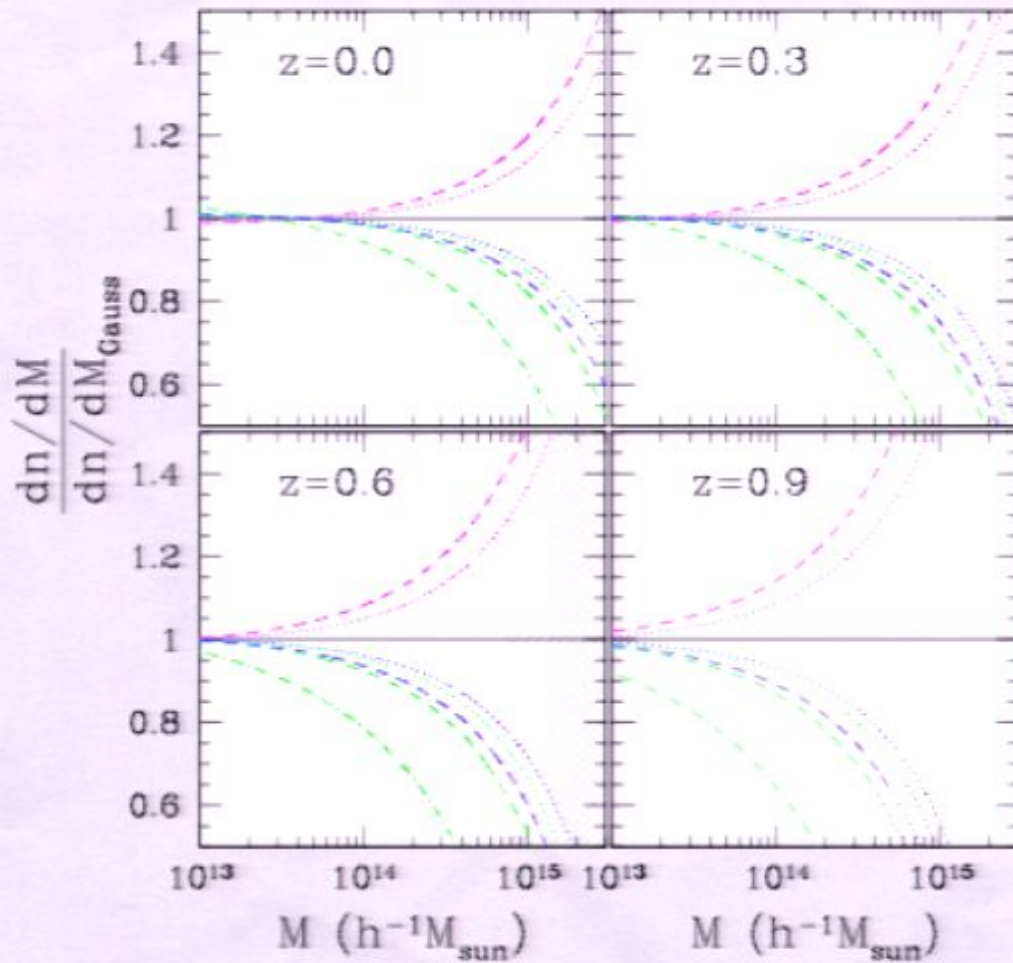
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- $f_{\text{NL}} = +332$ $\kappa = 0$
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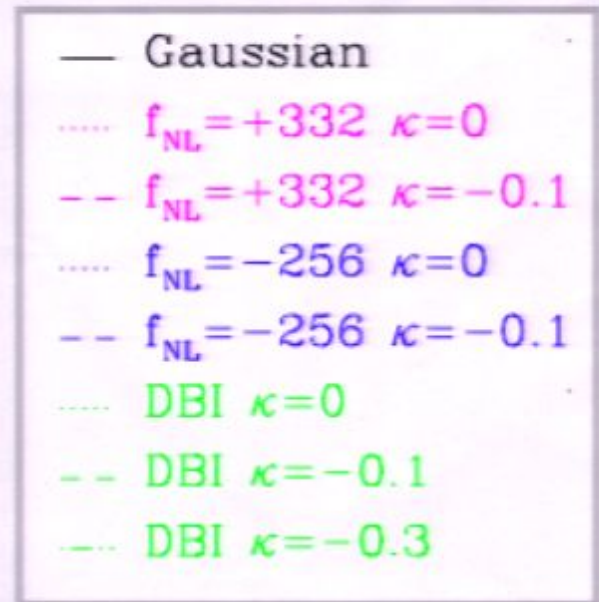
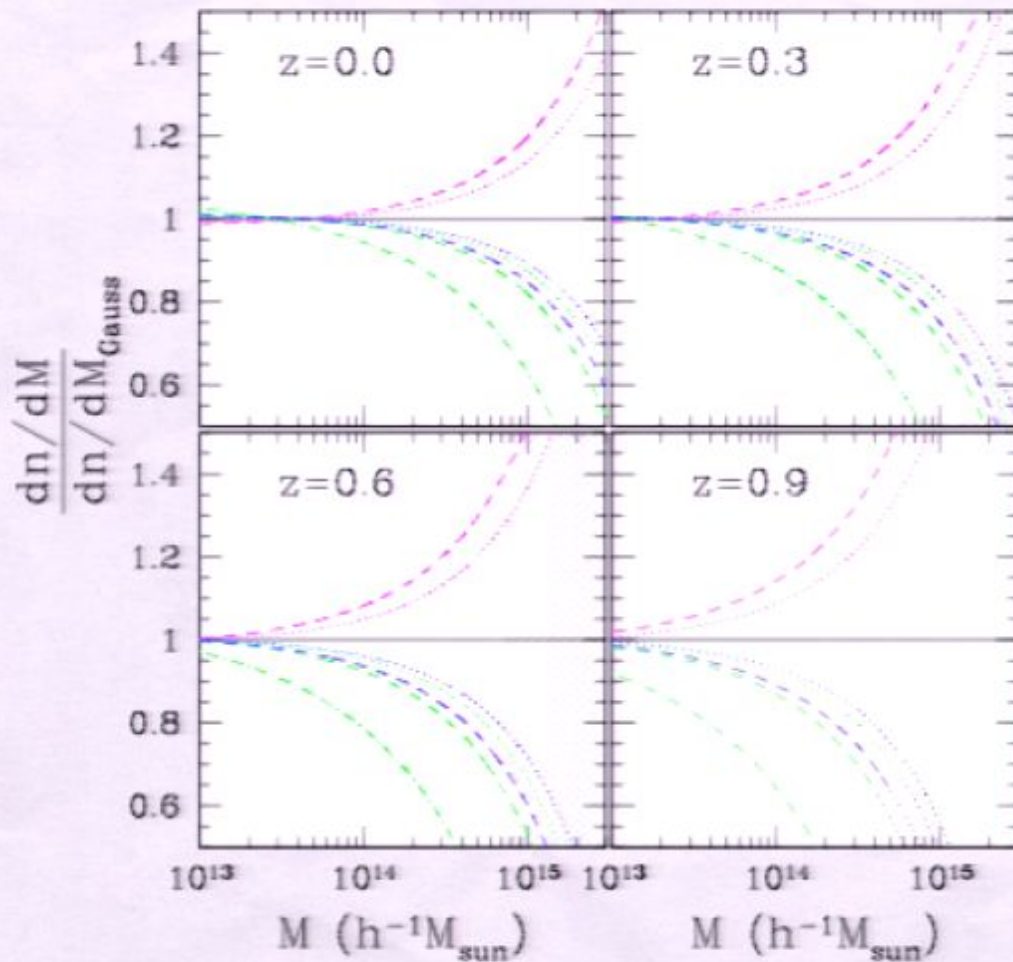


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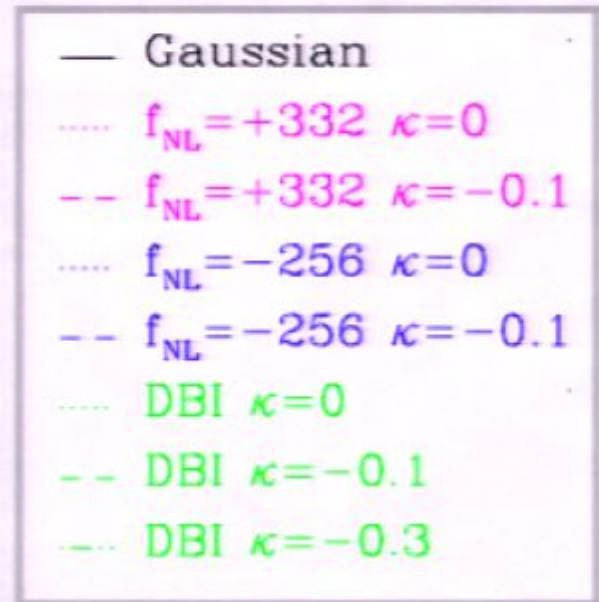
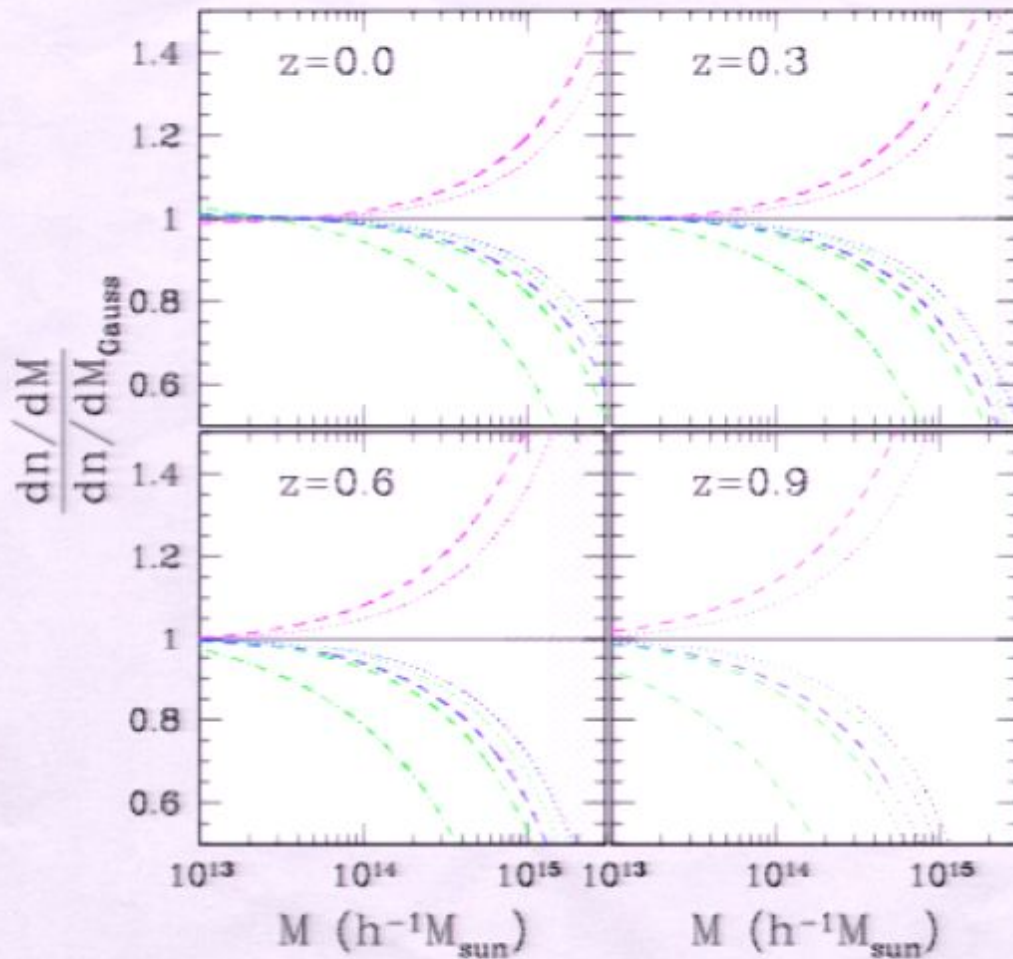
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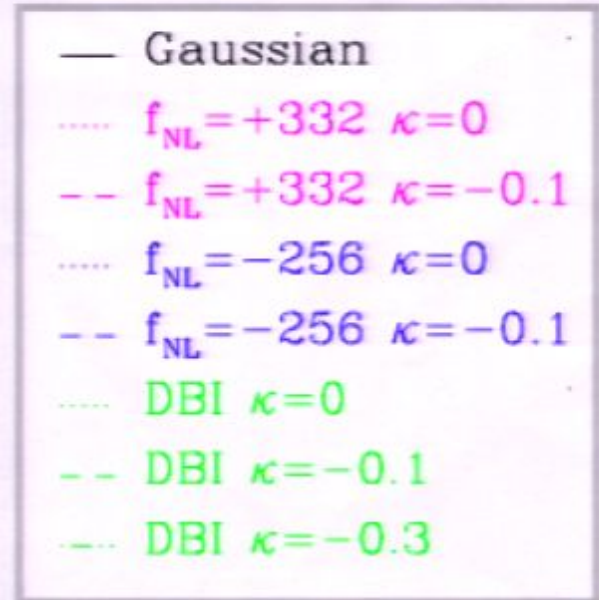
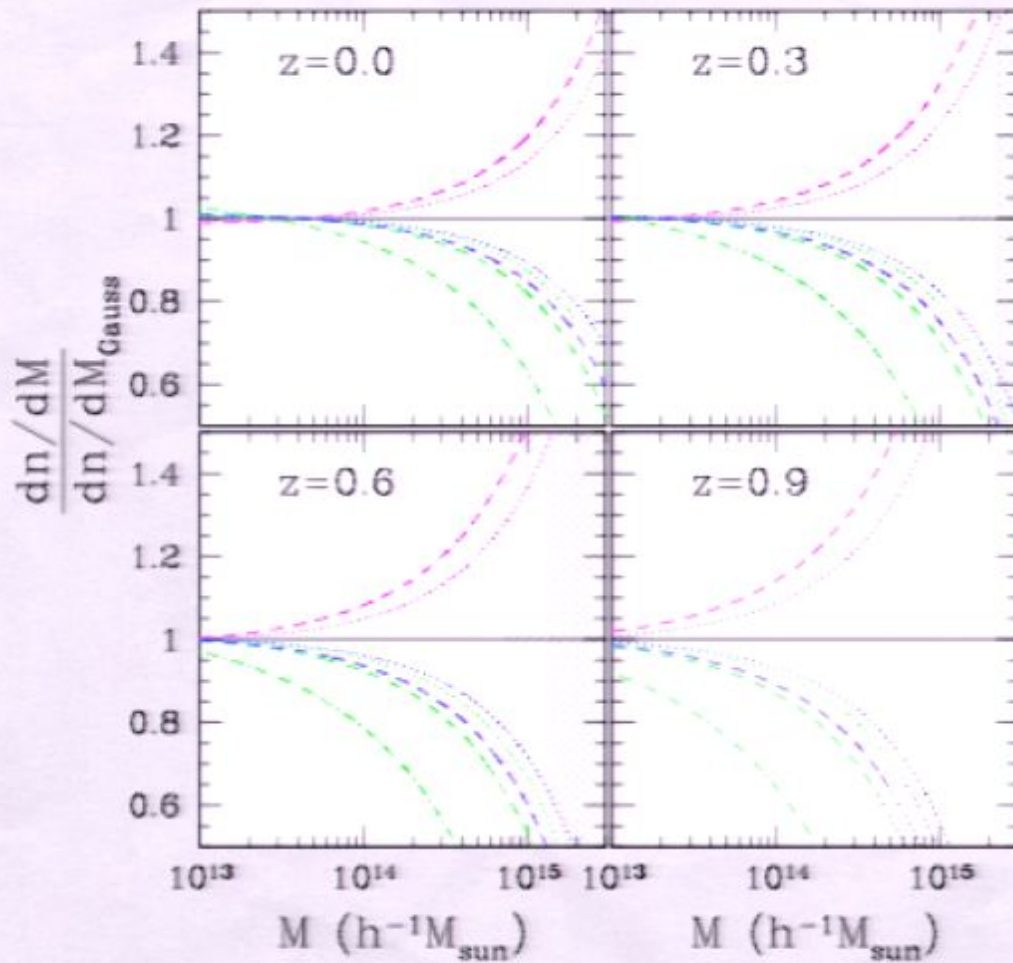
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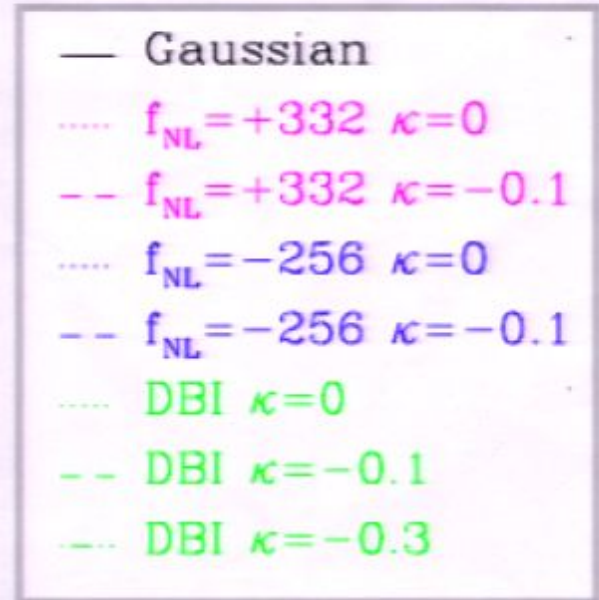
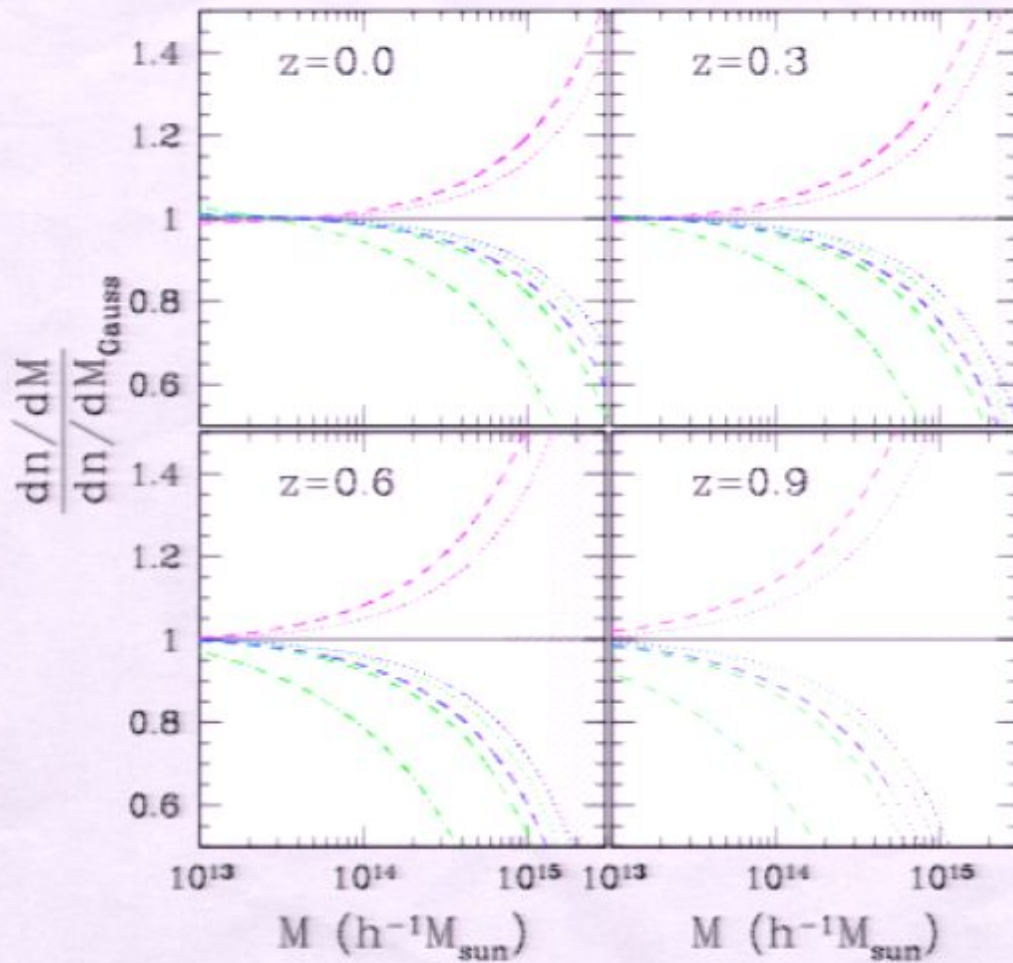
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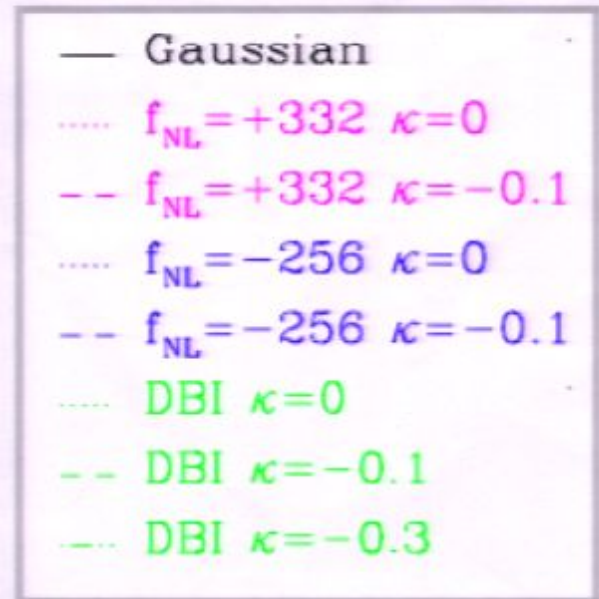
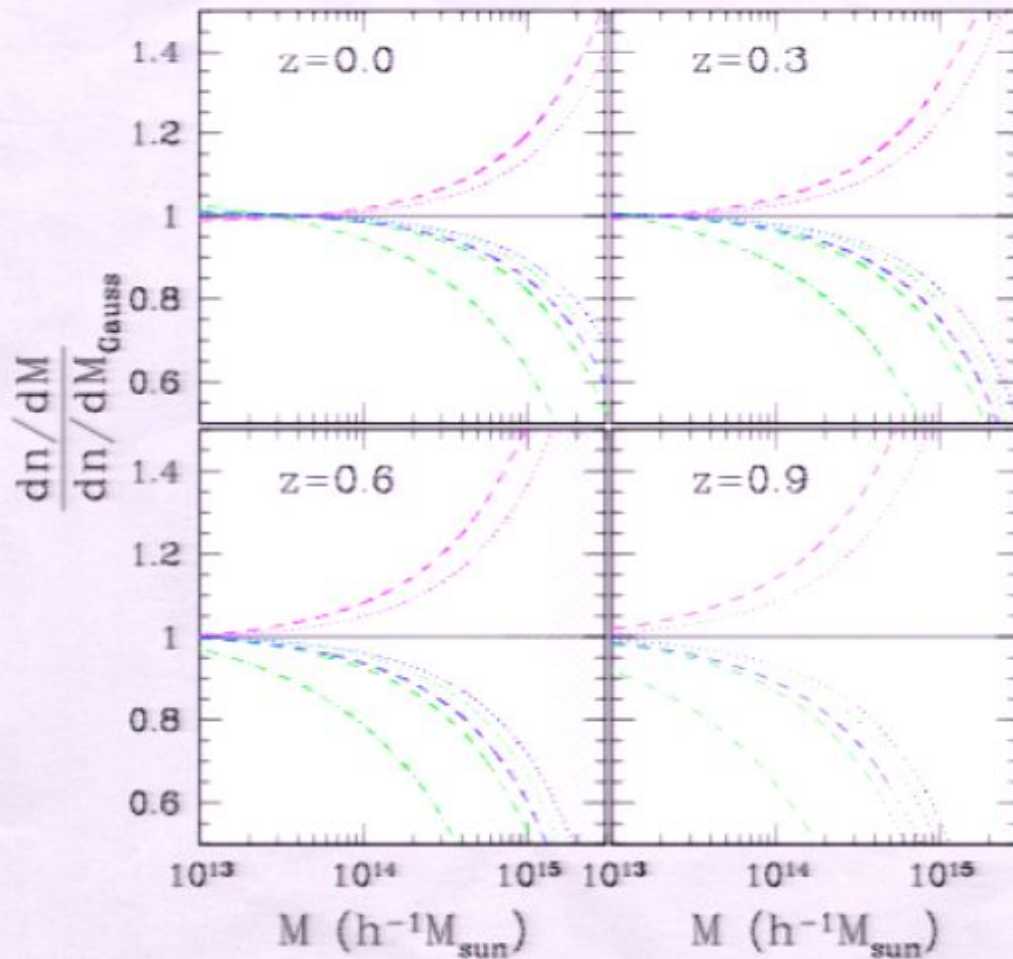
MASS FUNCTION



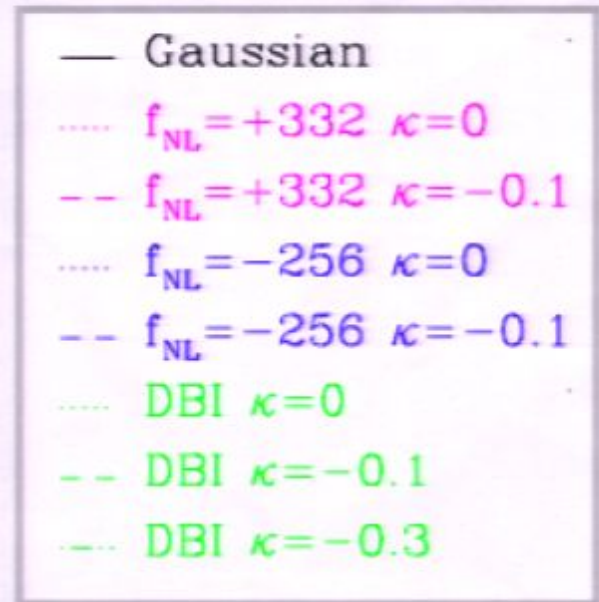
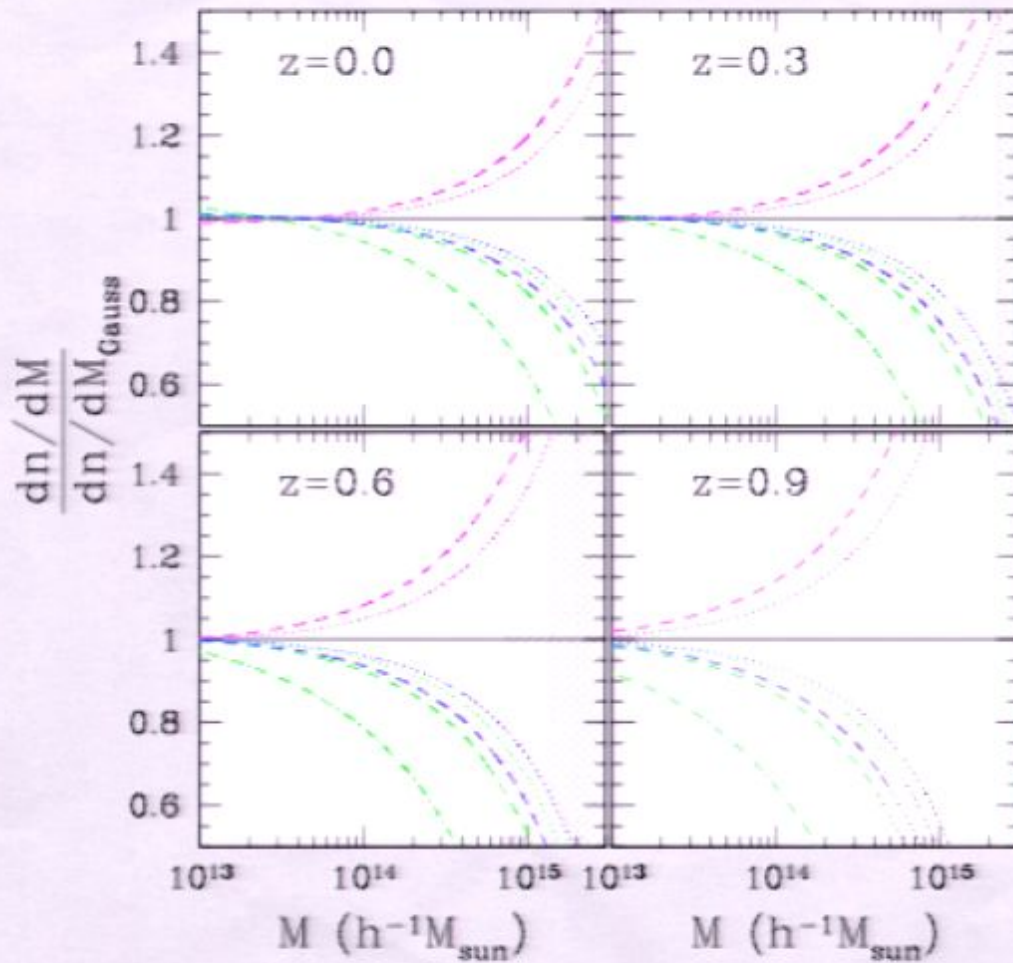
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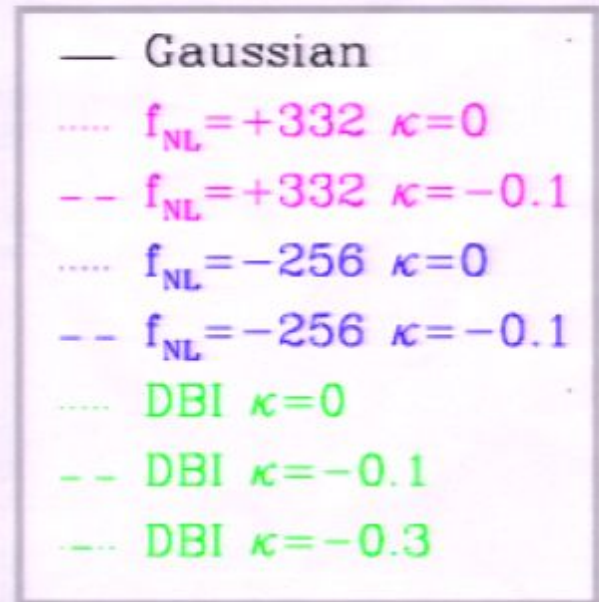
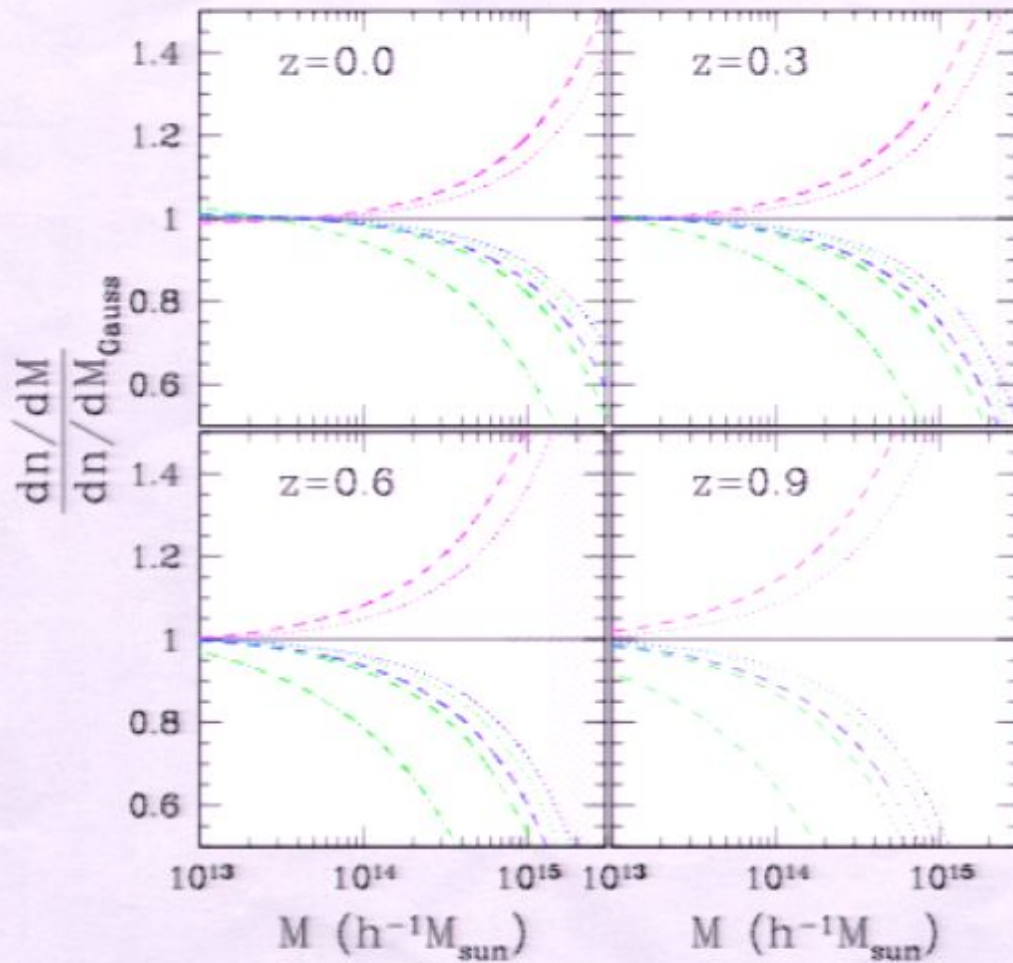
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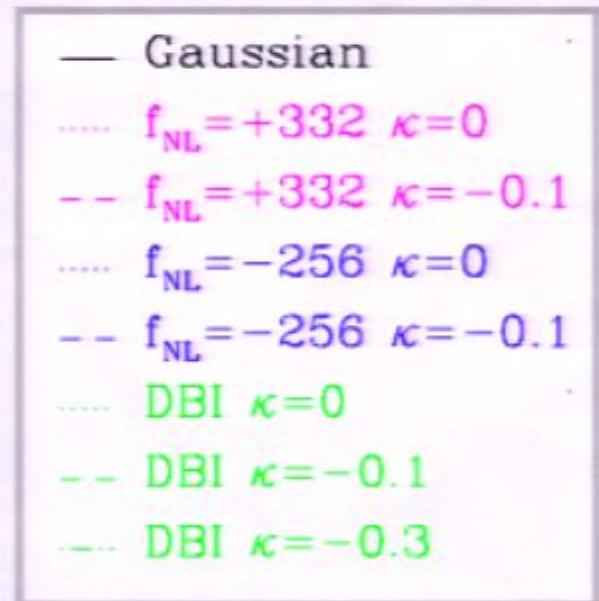
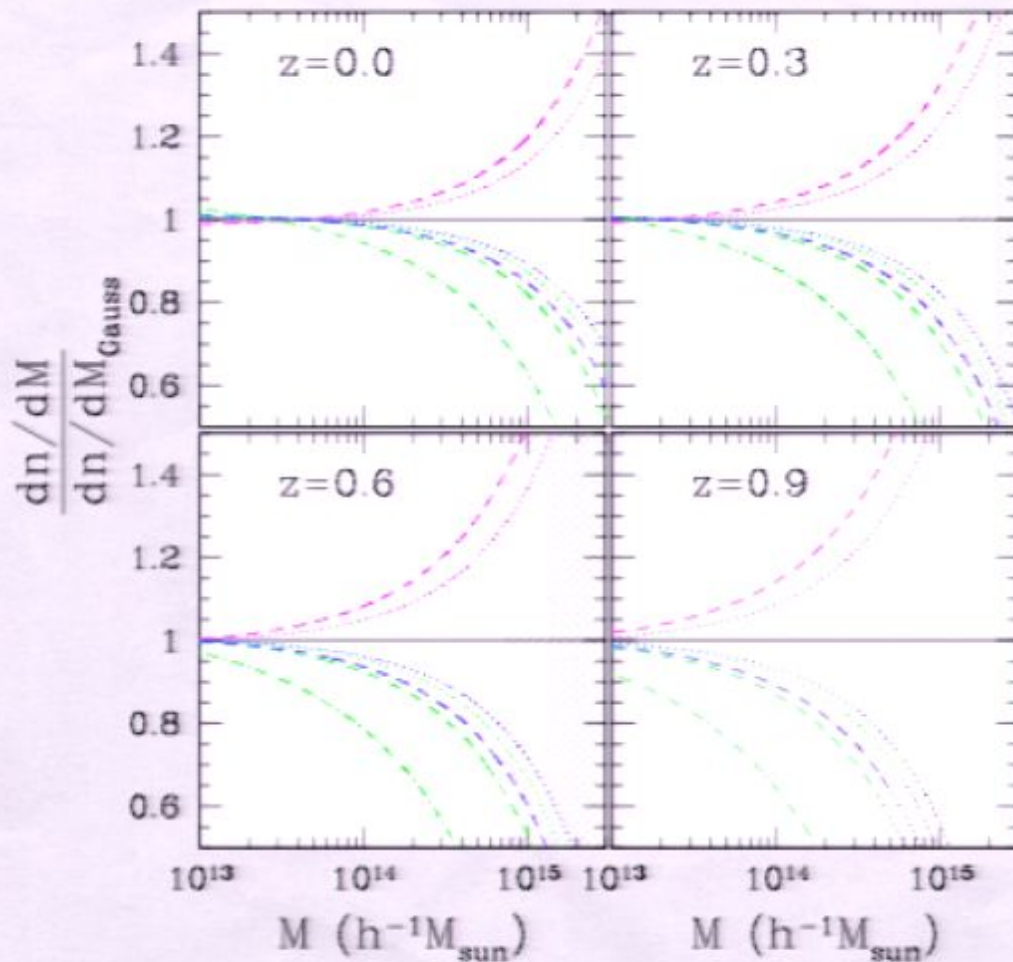
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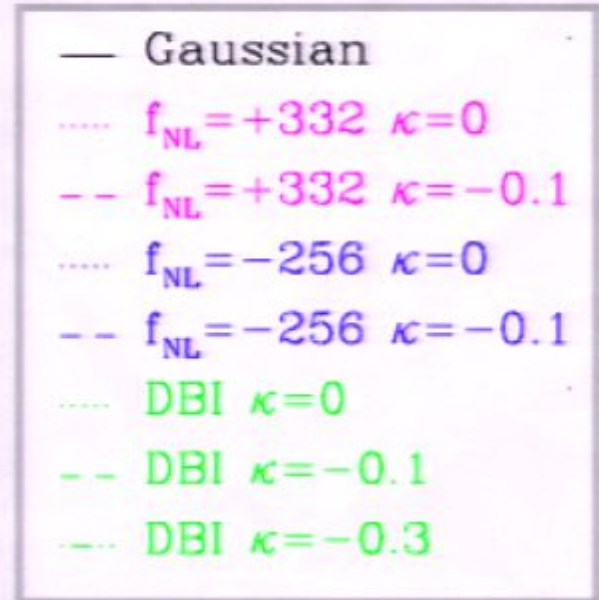
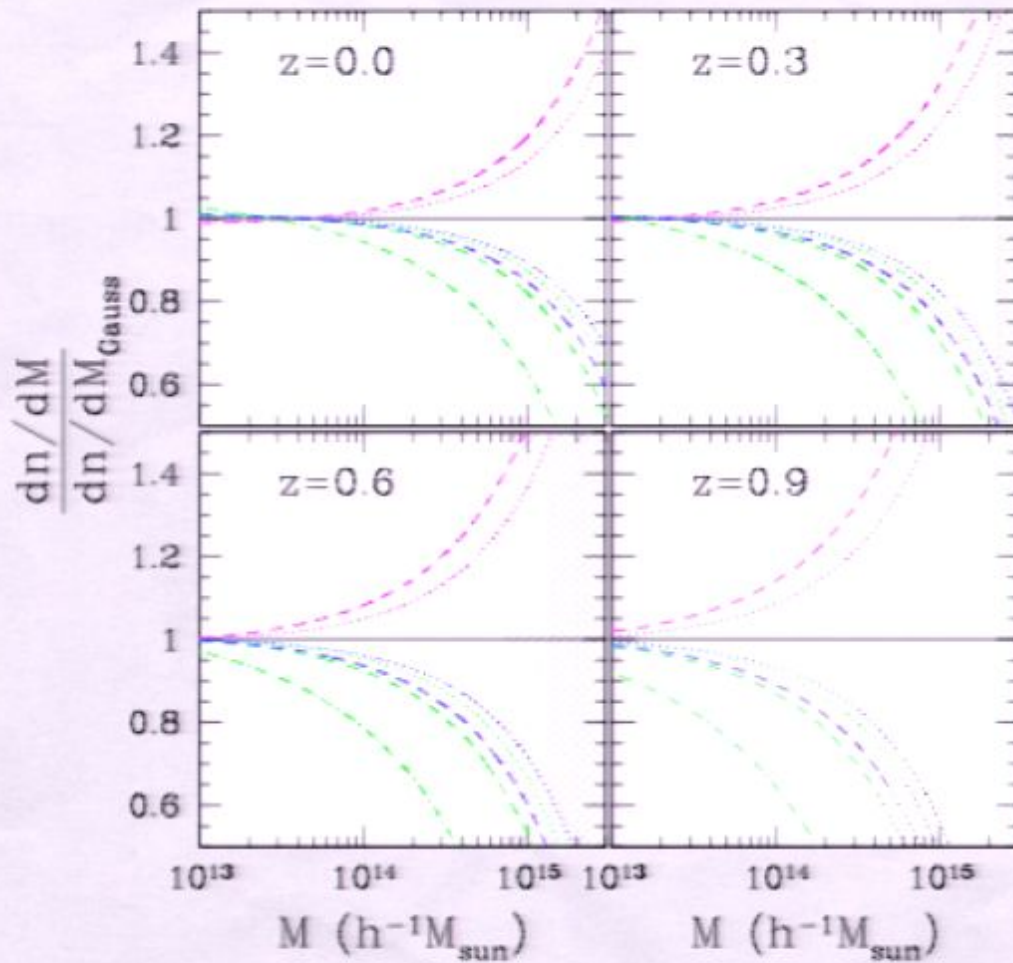
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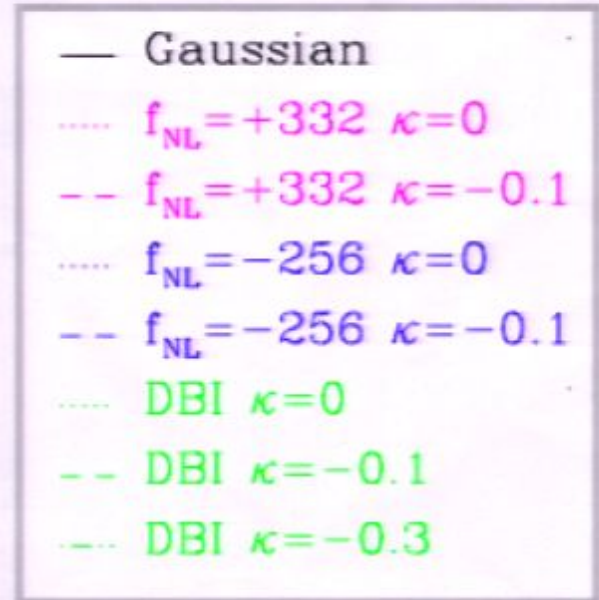
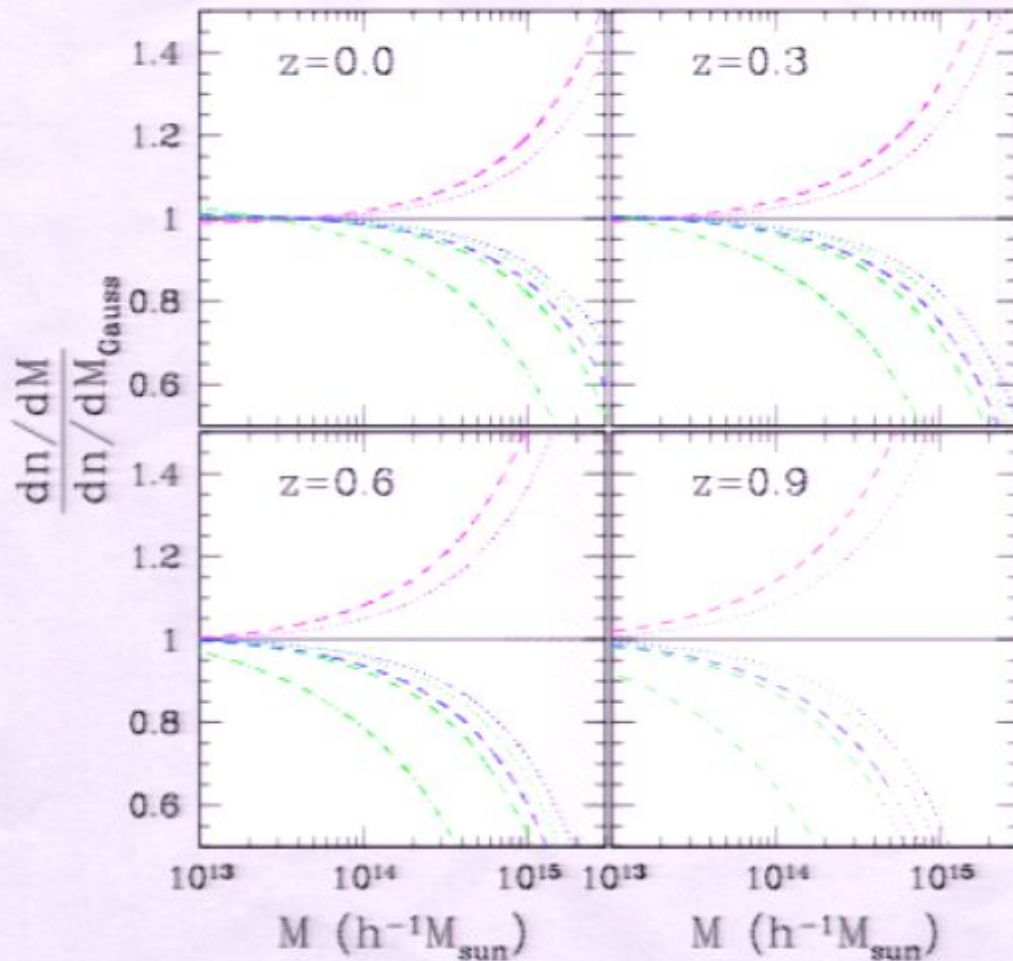
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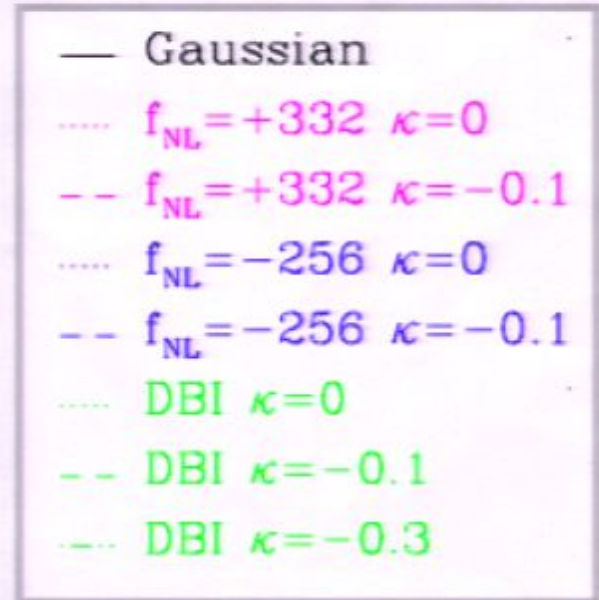
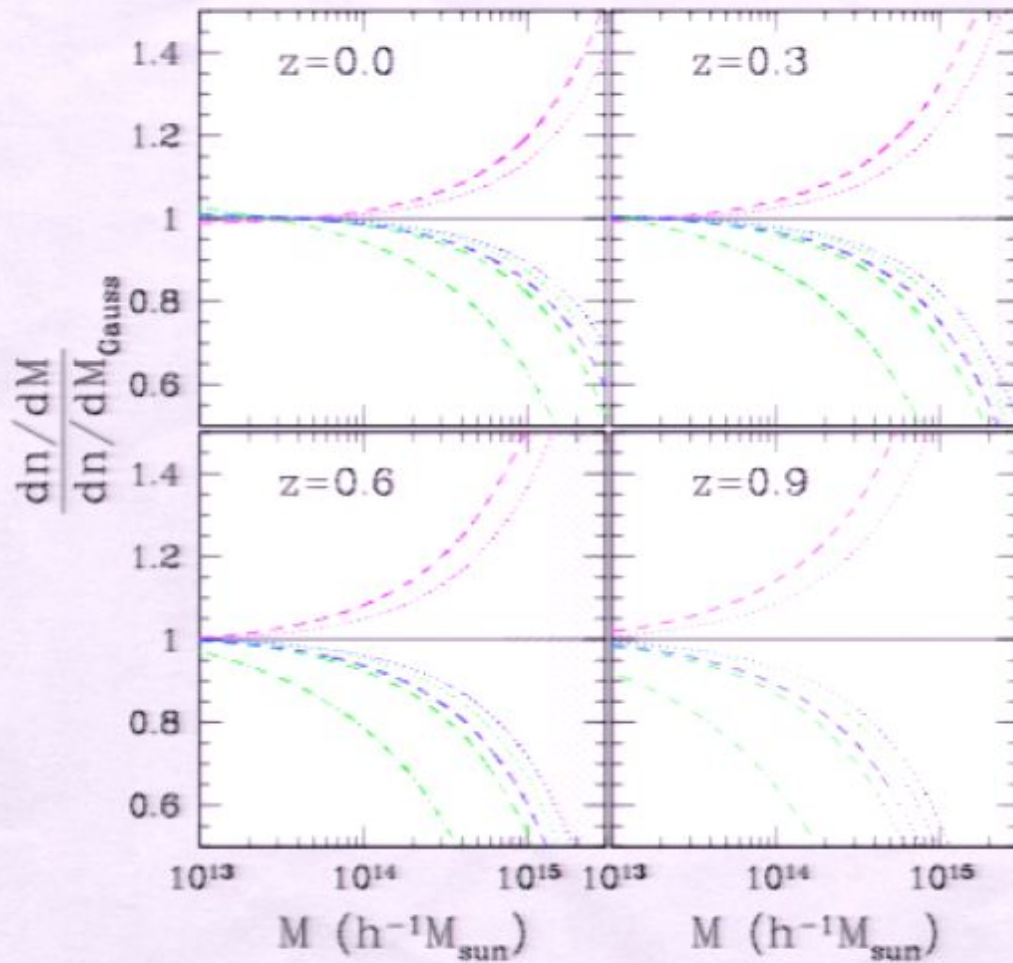
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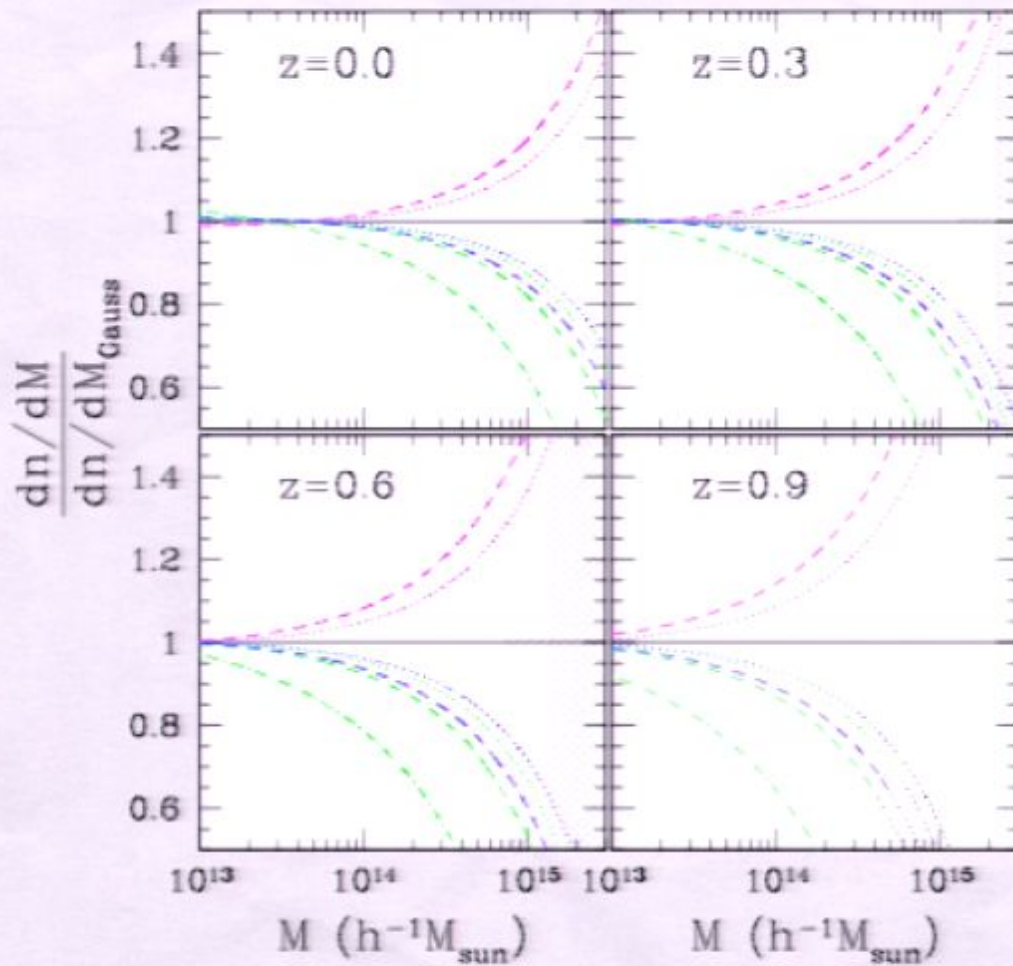
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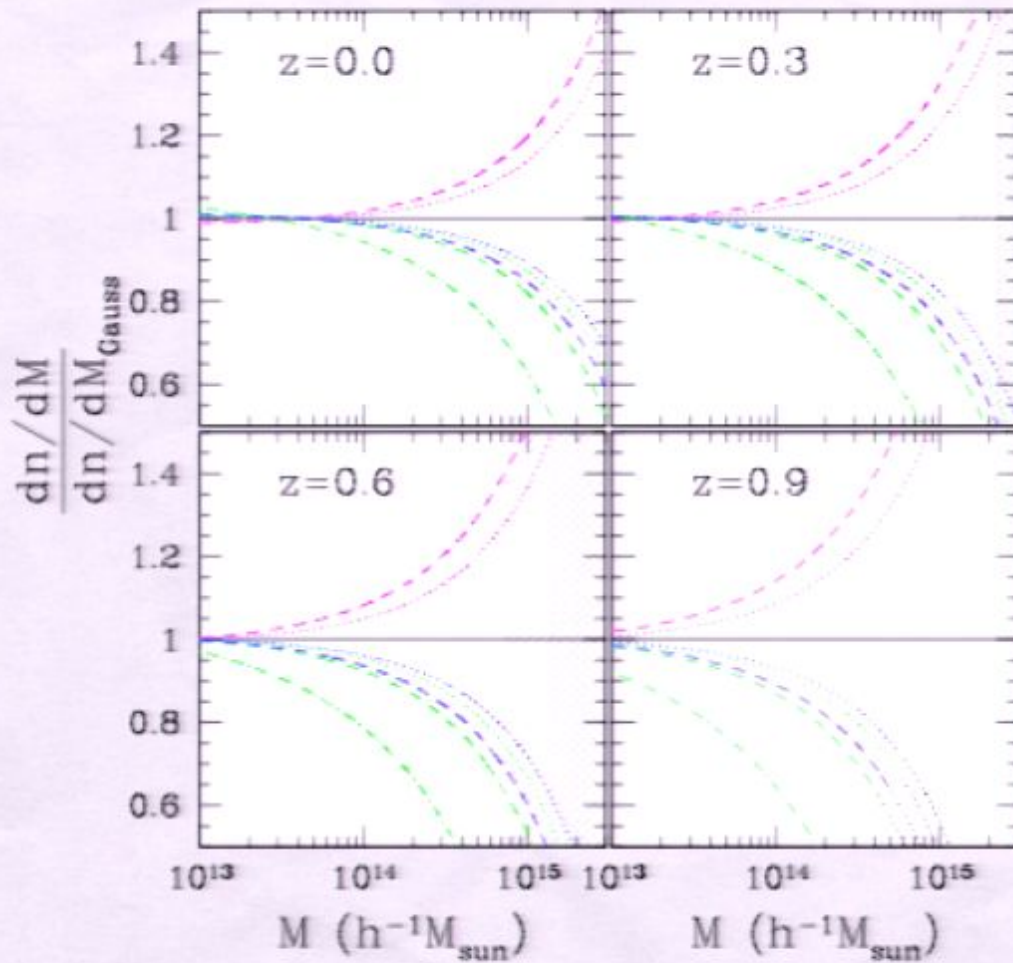


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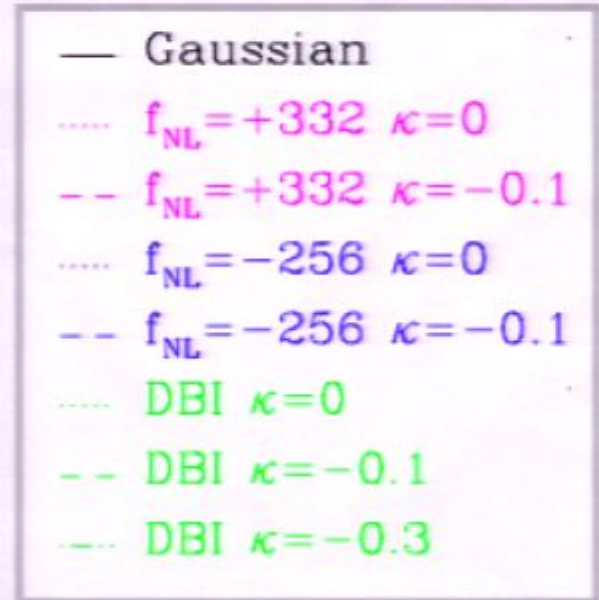
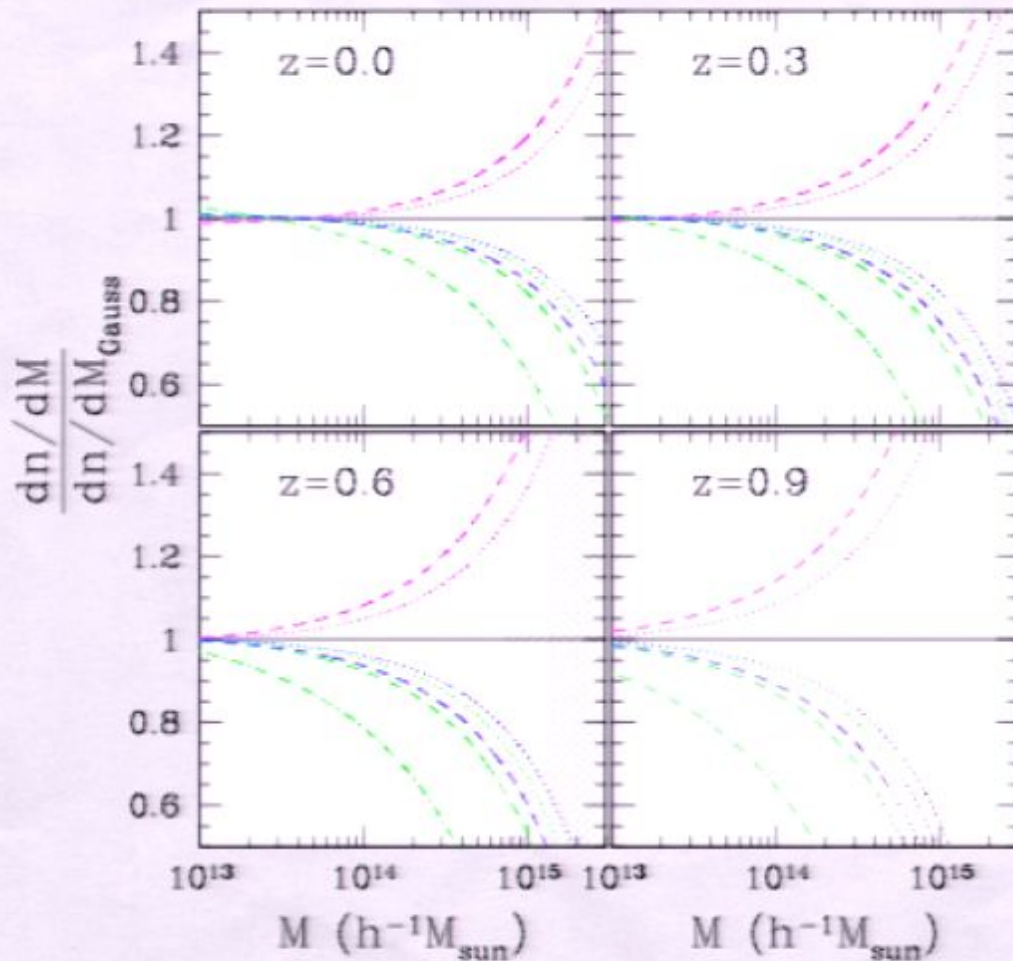
- Gaussian
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- $f_{\text{NL}} = +332 \quad \kappa = -0.1$
- $f_{\text{NL}} = -256 \quad \kappa = 0$
- $f_{\text{NL}} = -256 \quad \kappa = -0.1$
- DBI $\kappa = 0$
- DBI $\kappa = -0.1$
- DBI $\kappa = -0.3$

MASS FUNCTION

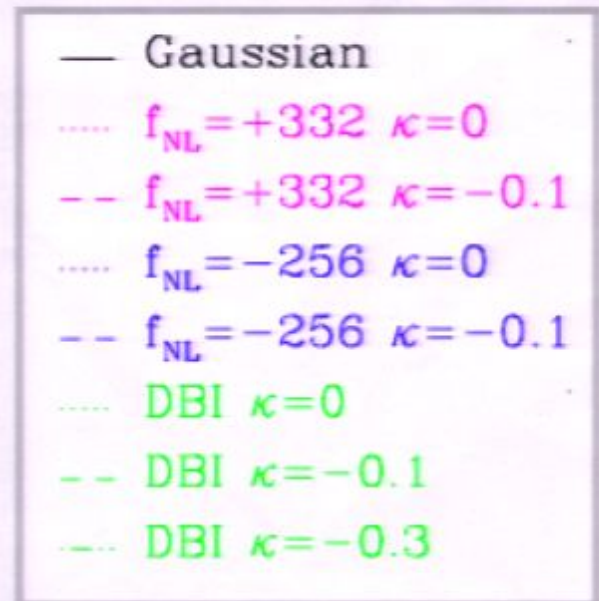
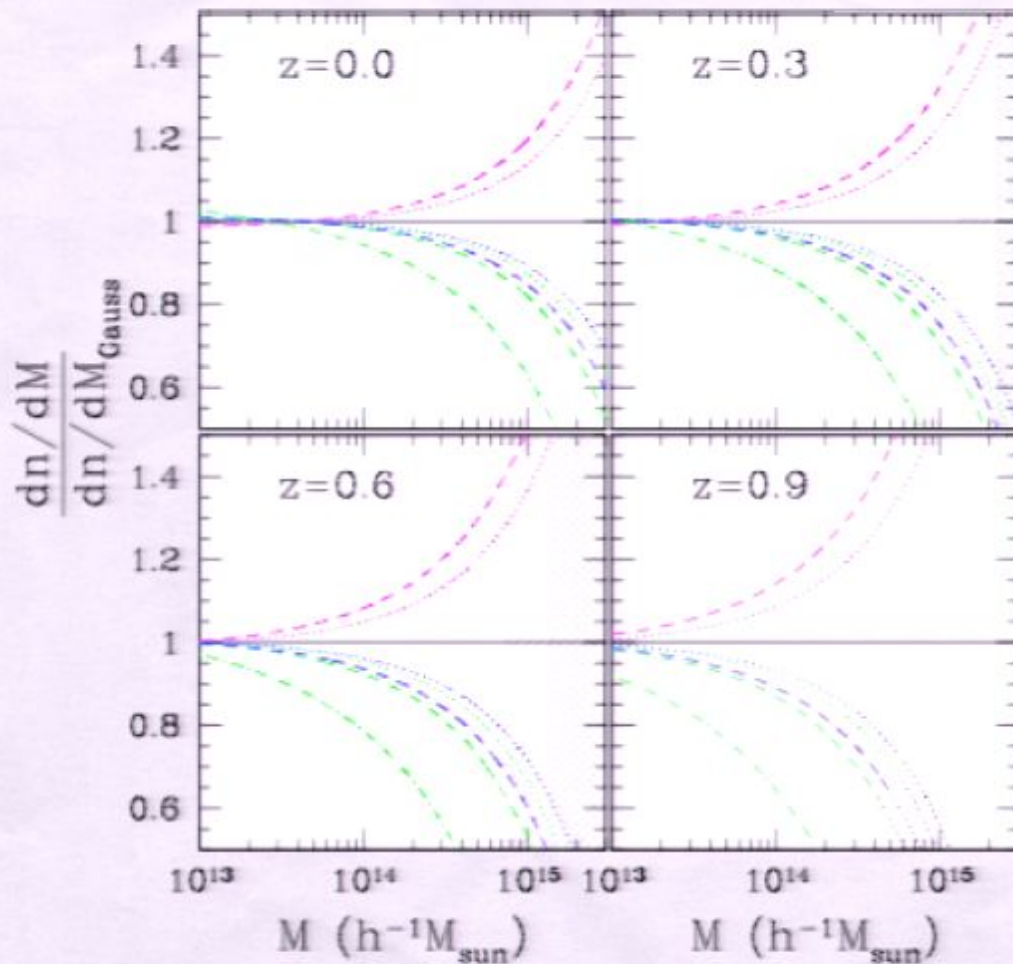


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- .-.- DBI $\kappa = -0.3$

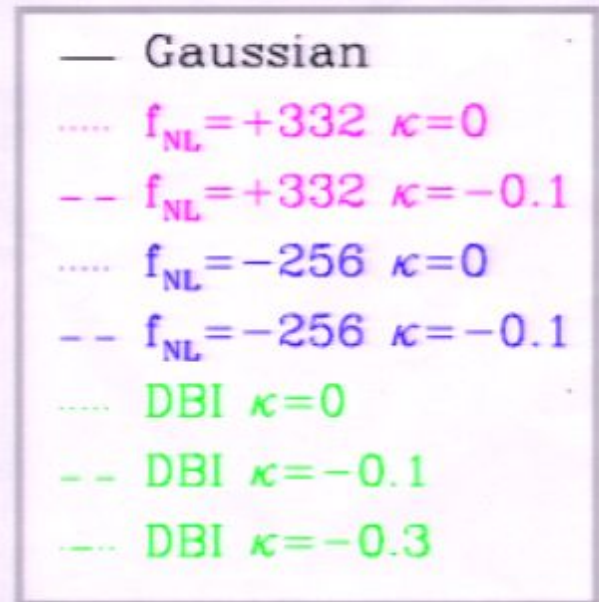
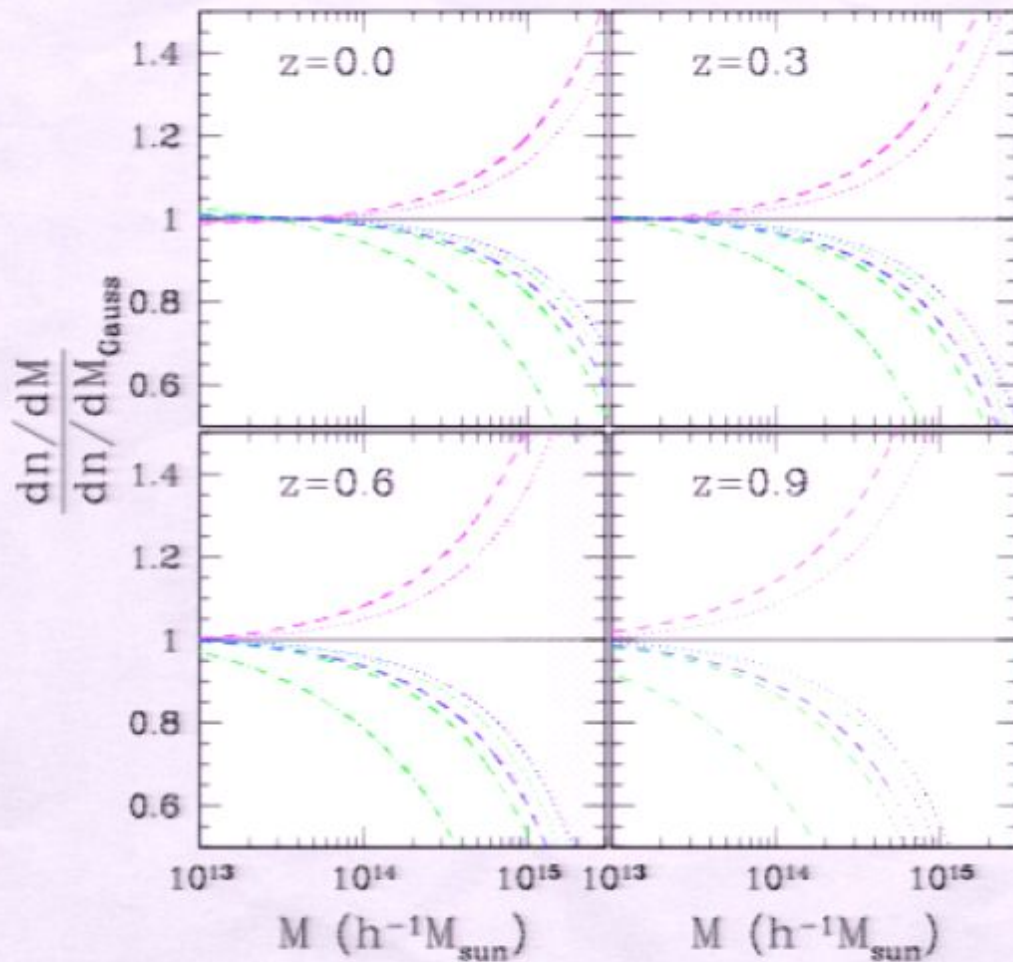
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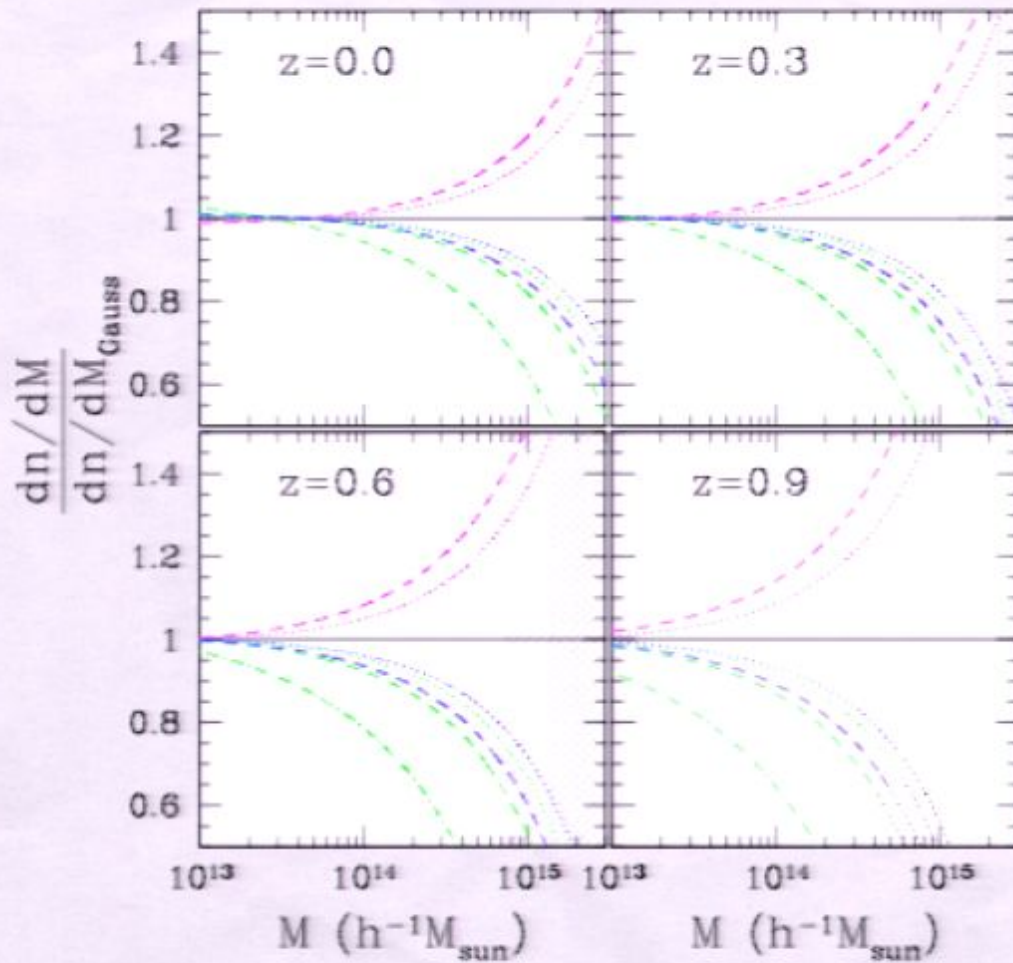
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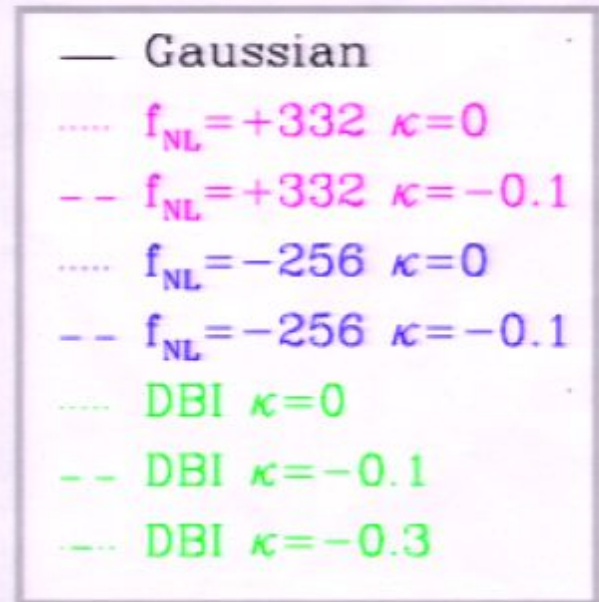
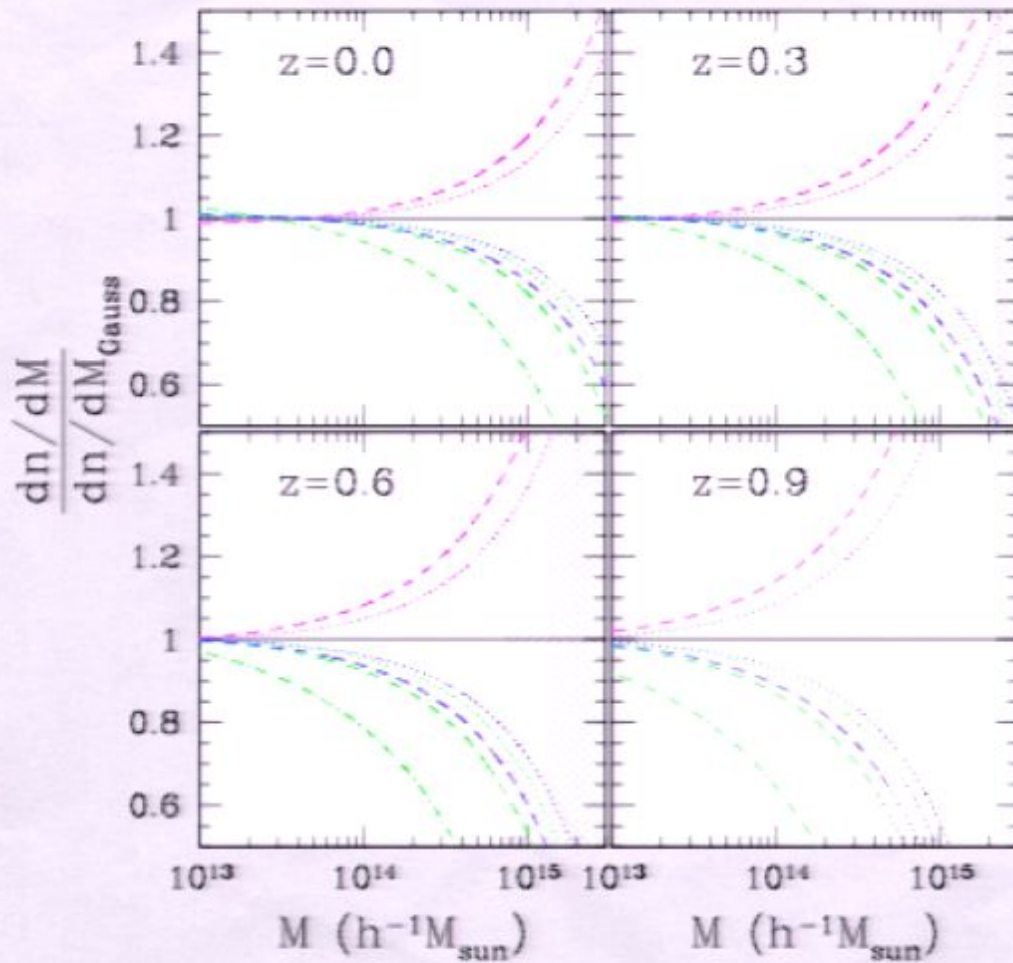


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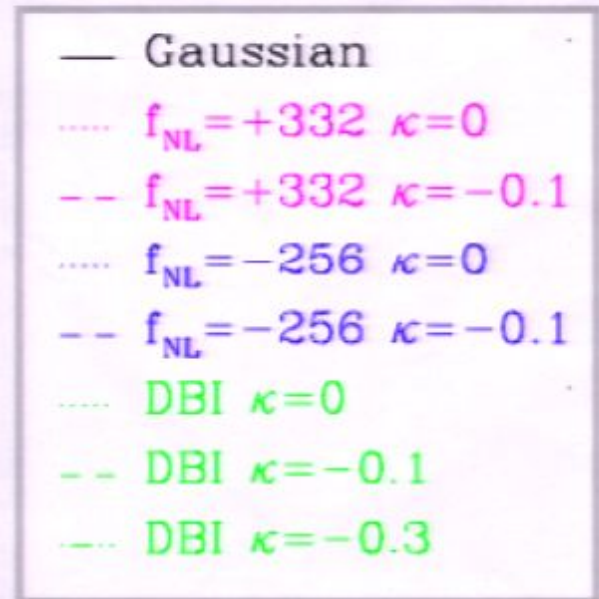
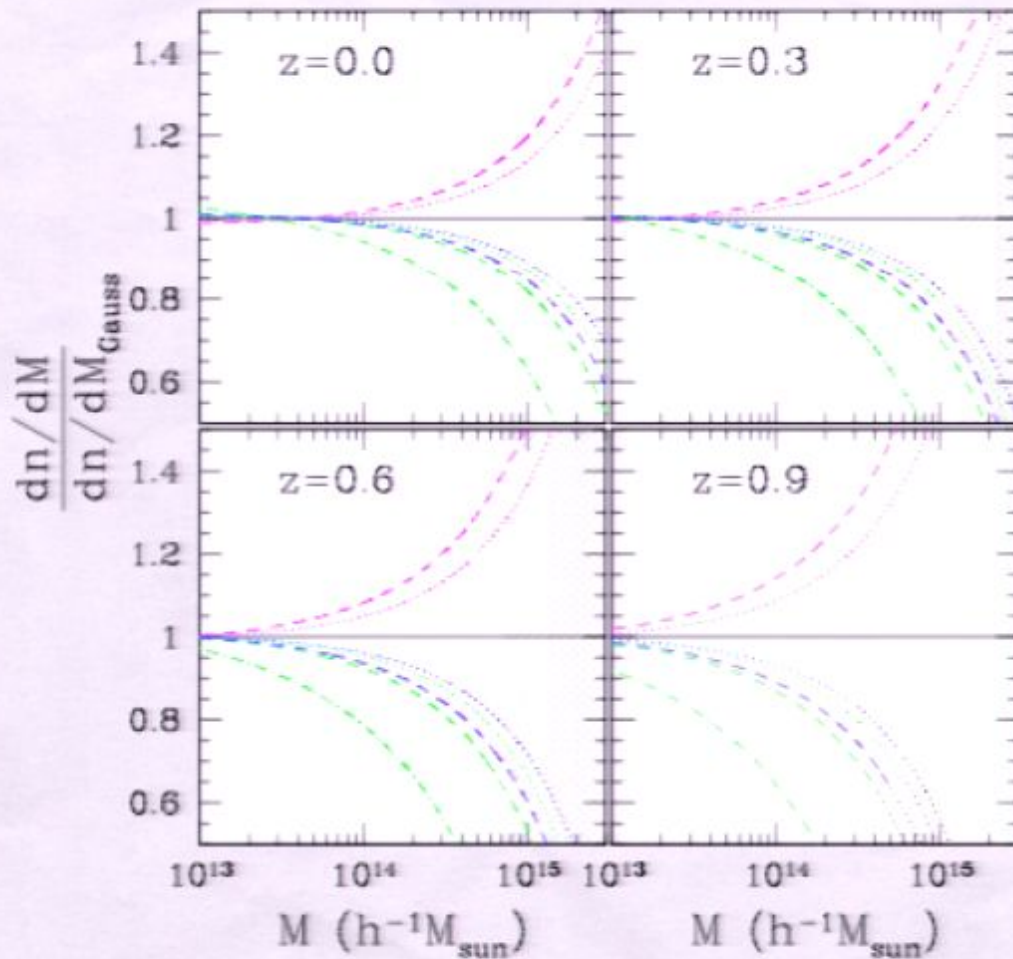


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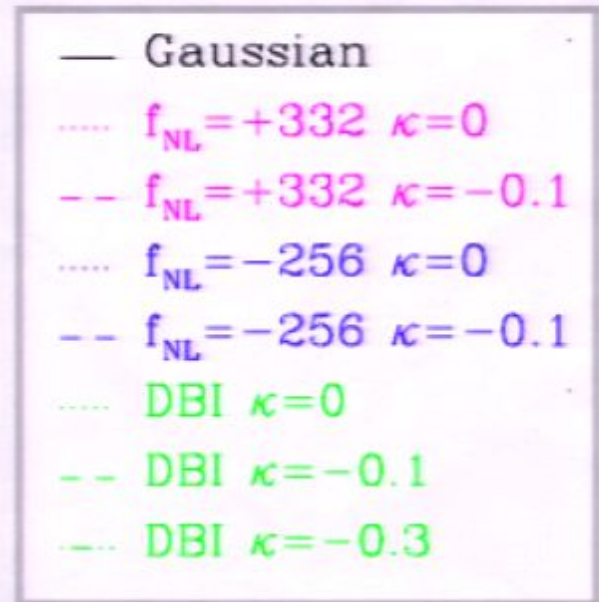
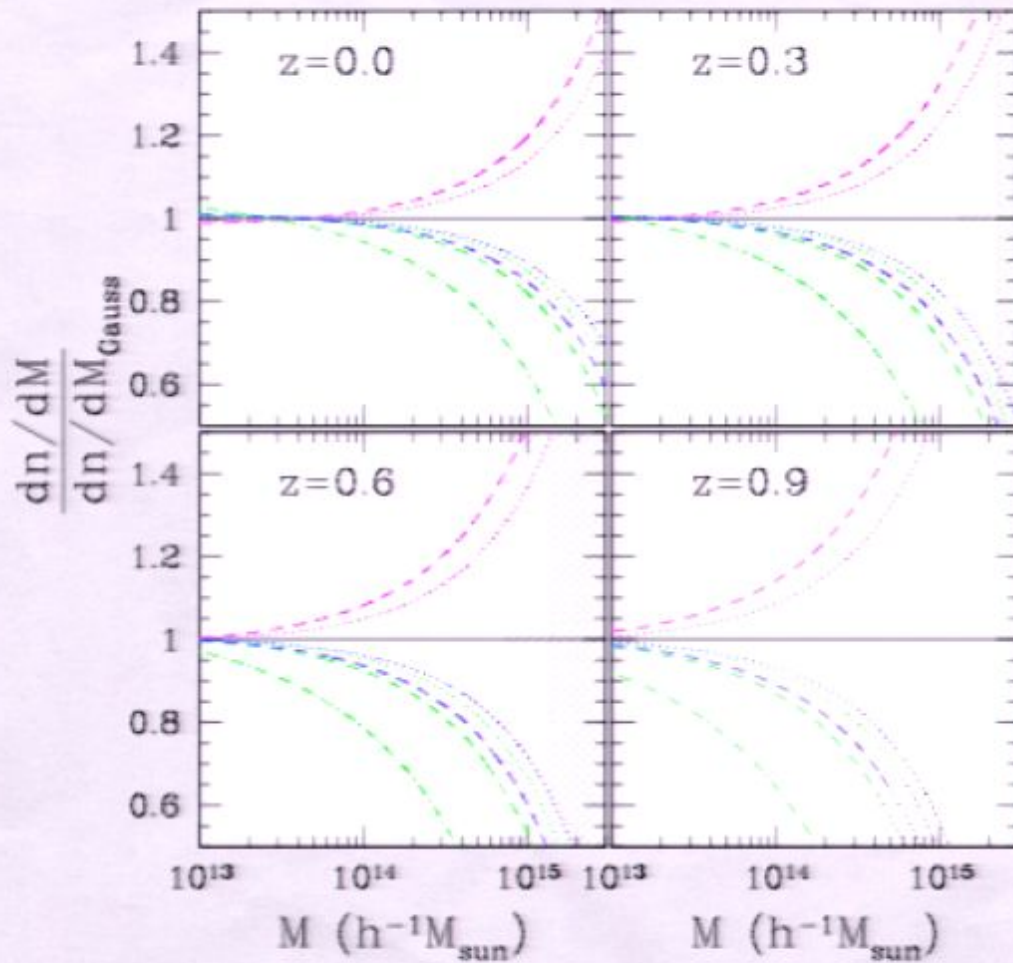
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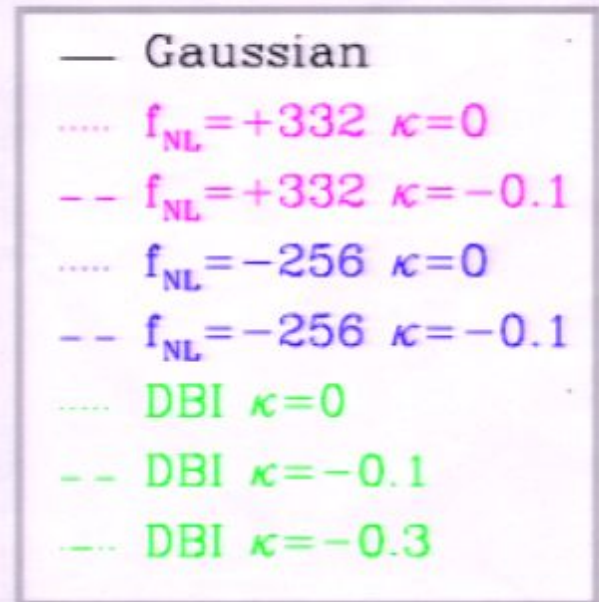
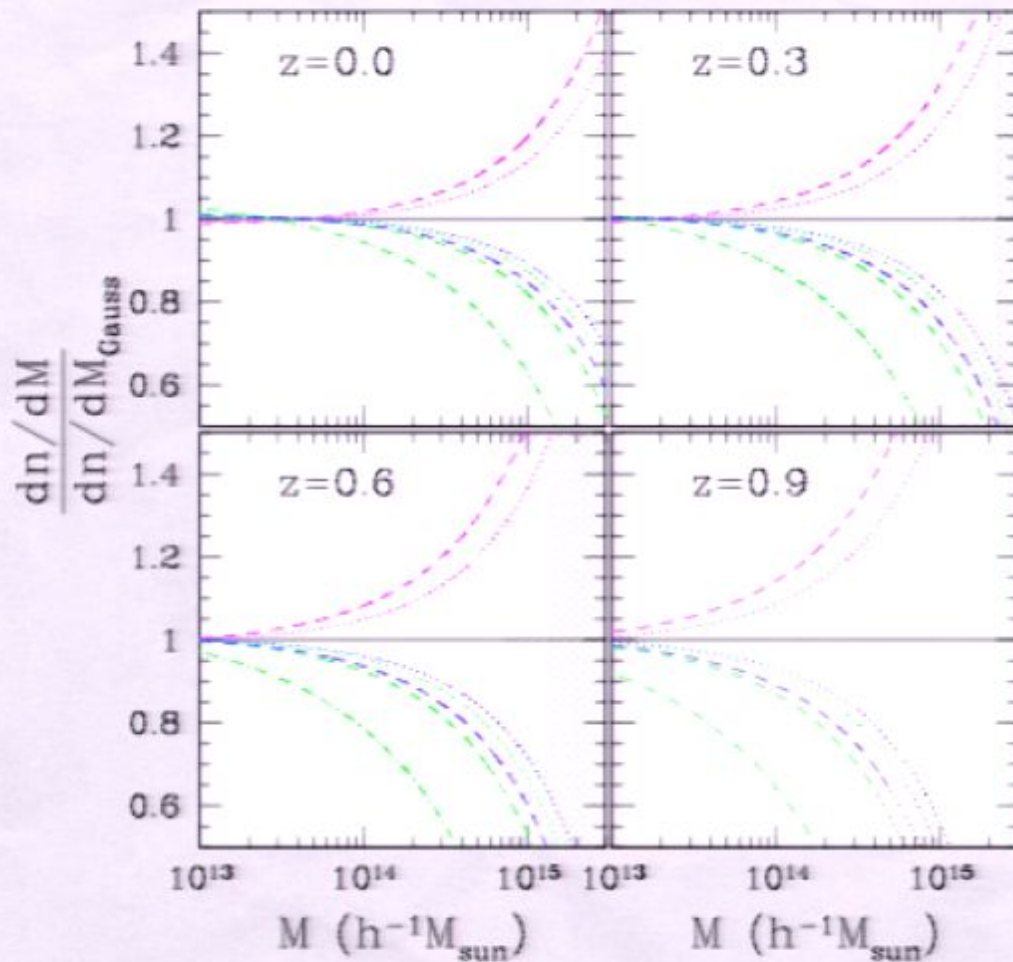
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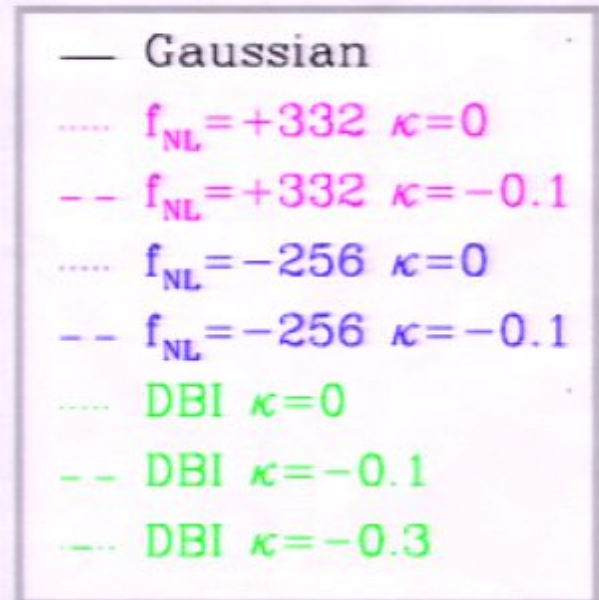
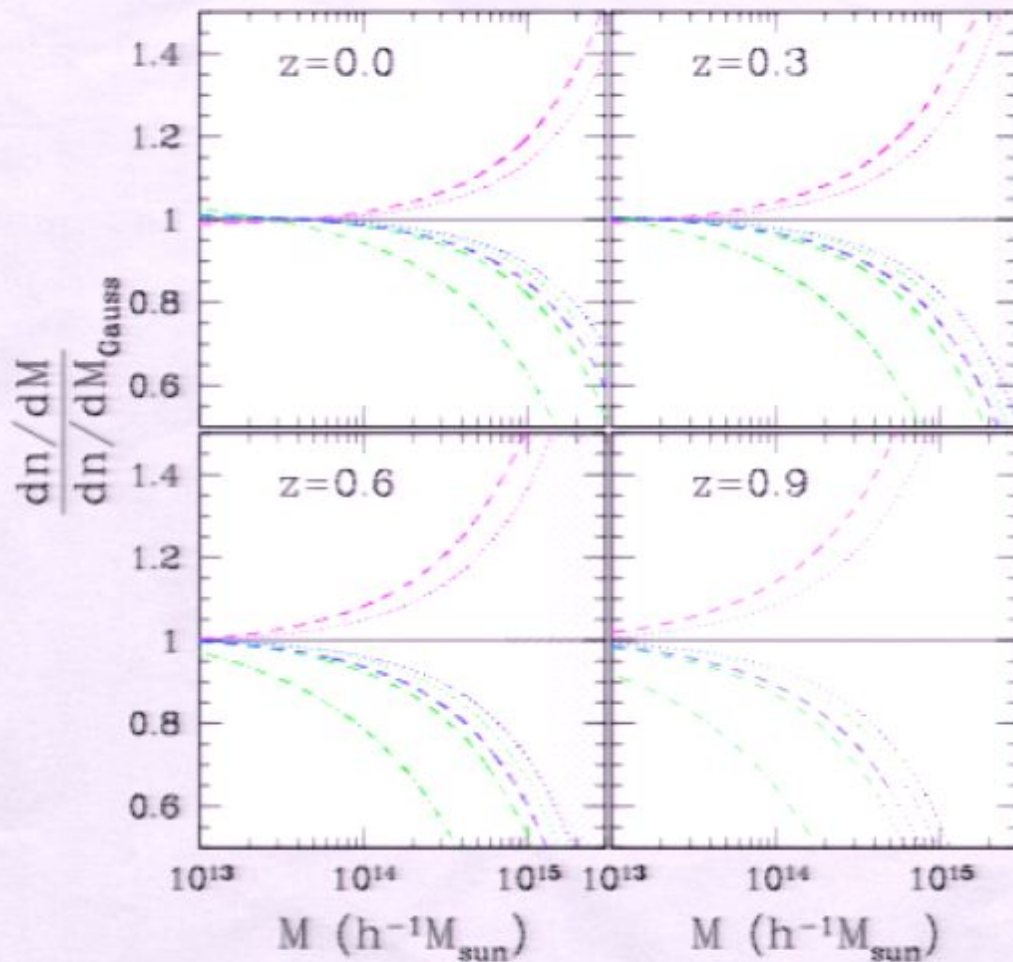
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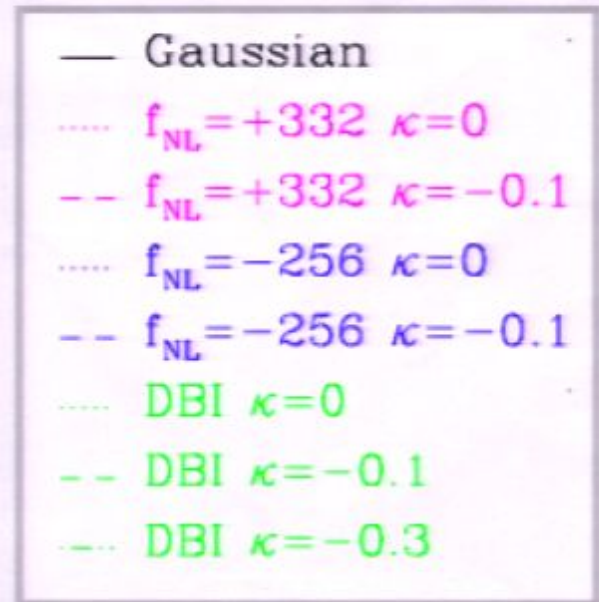
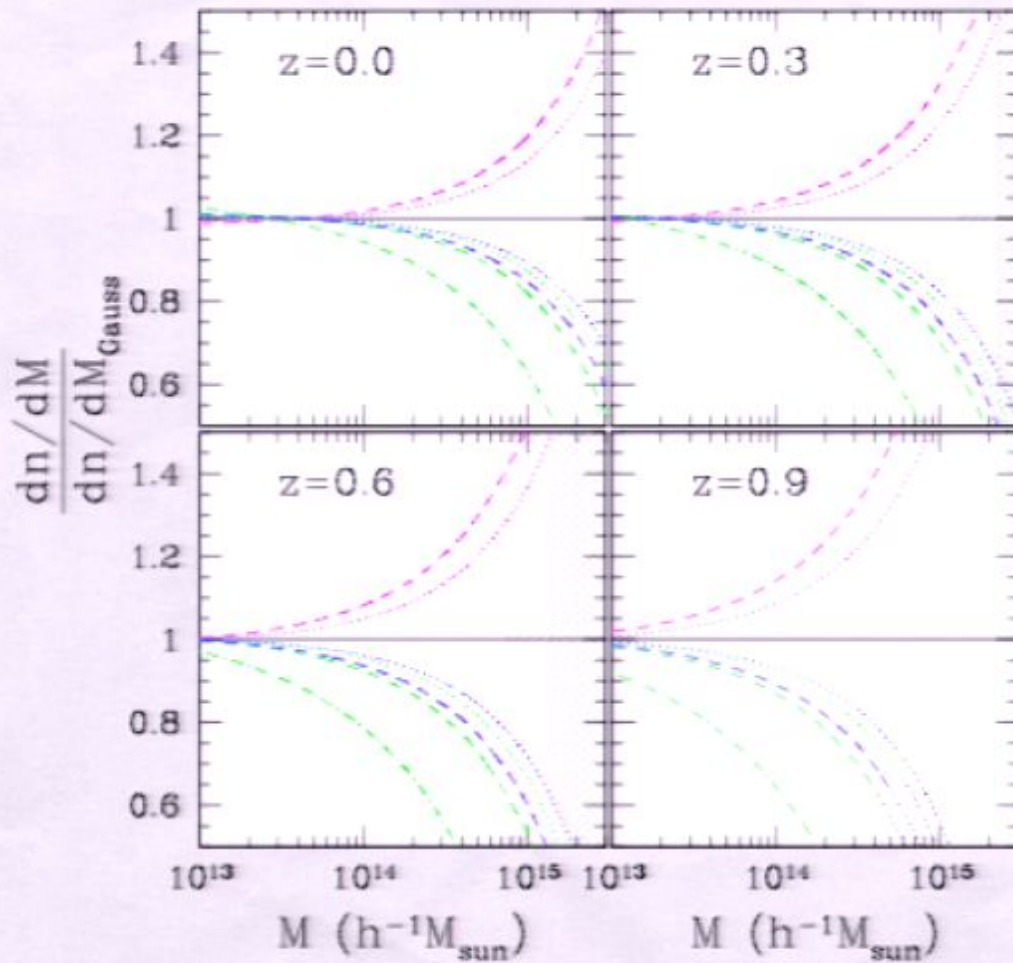
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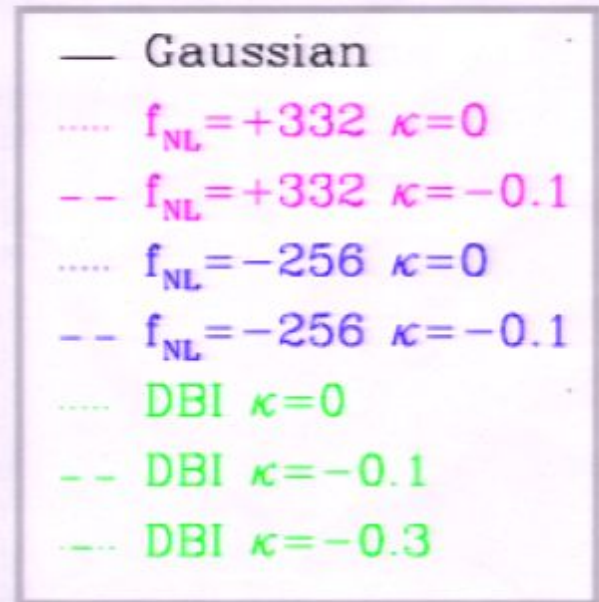
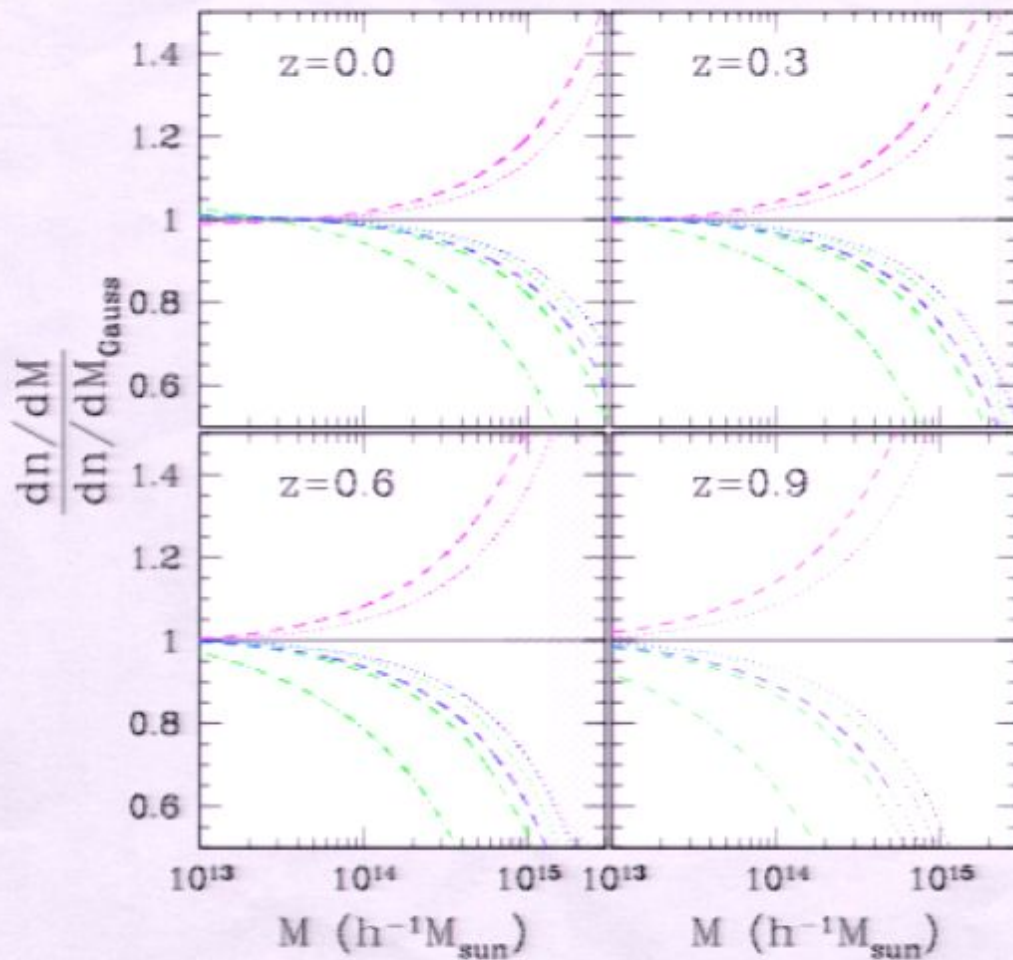
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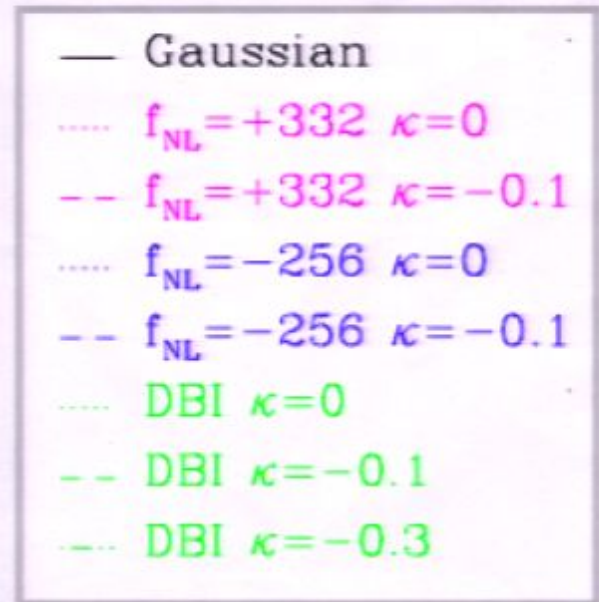
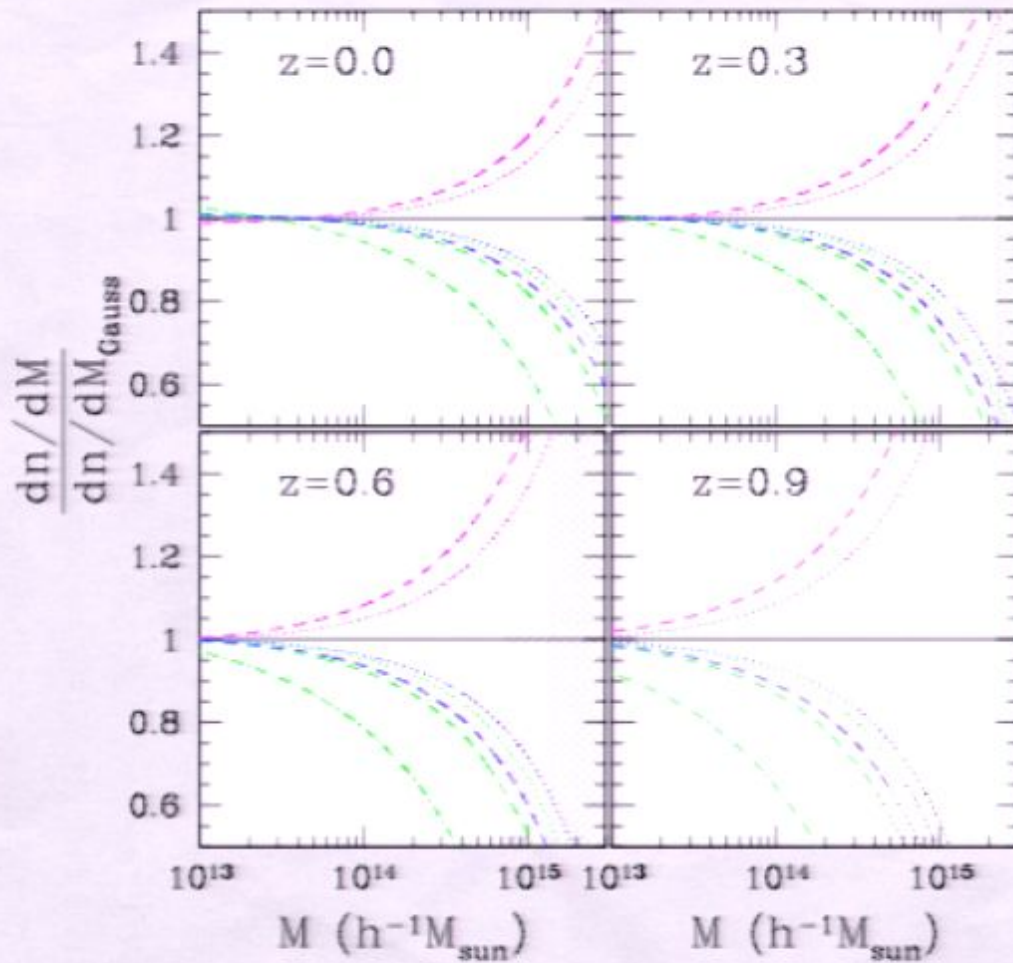
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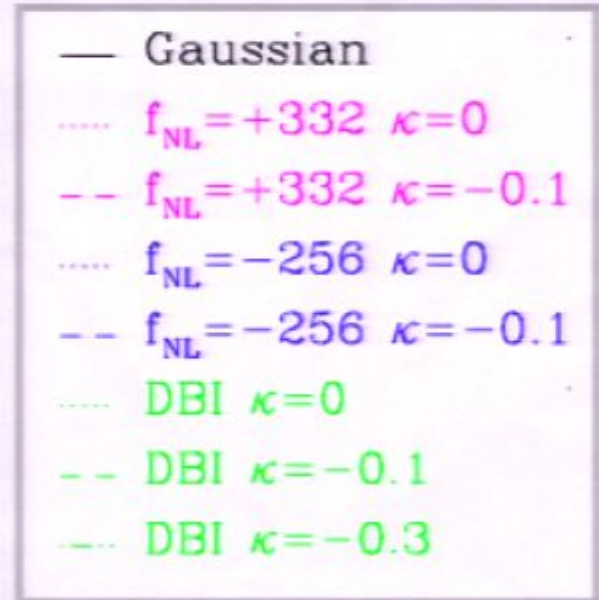
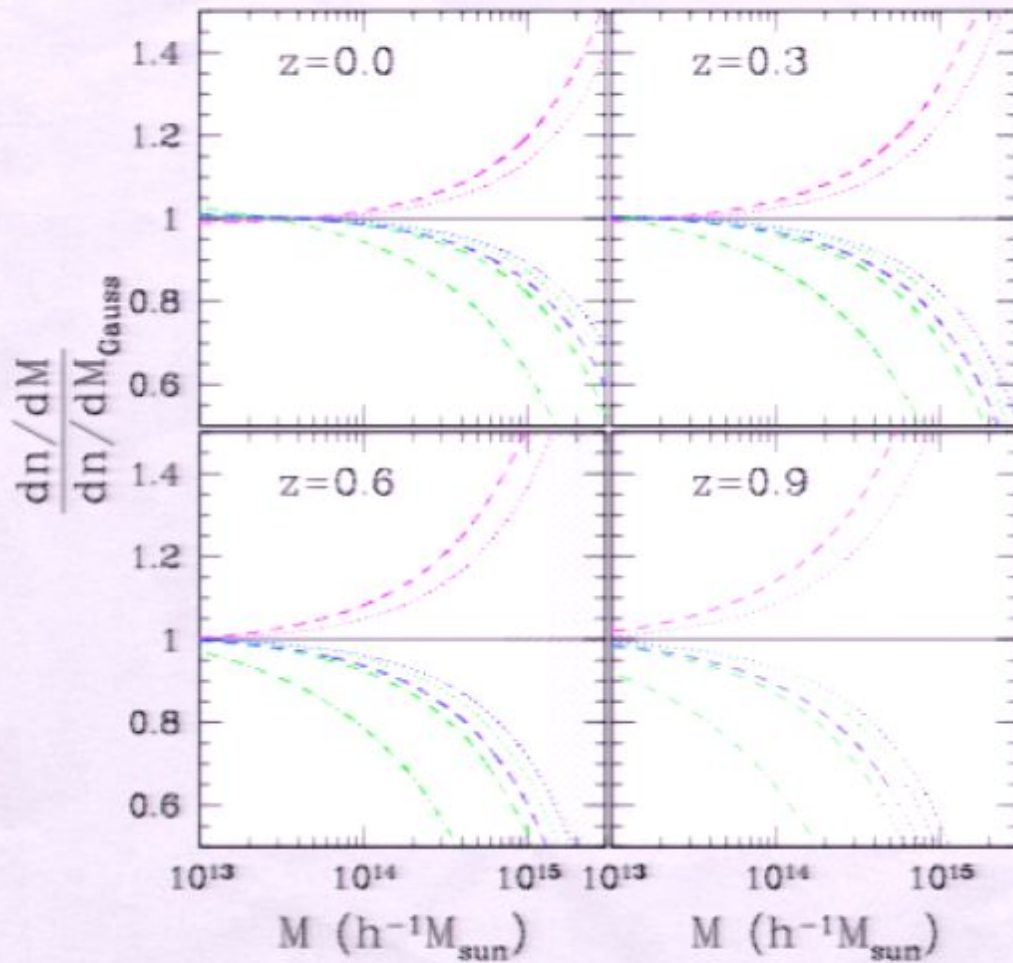
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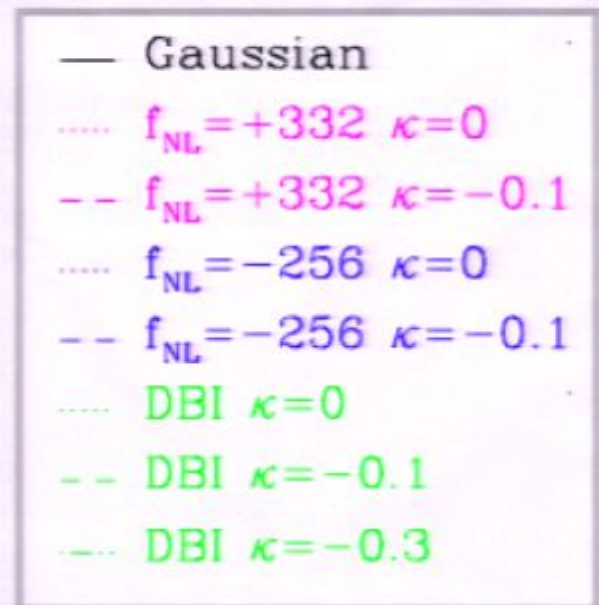
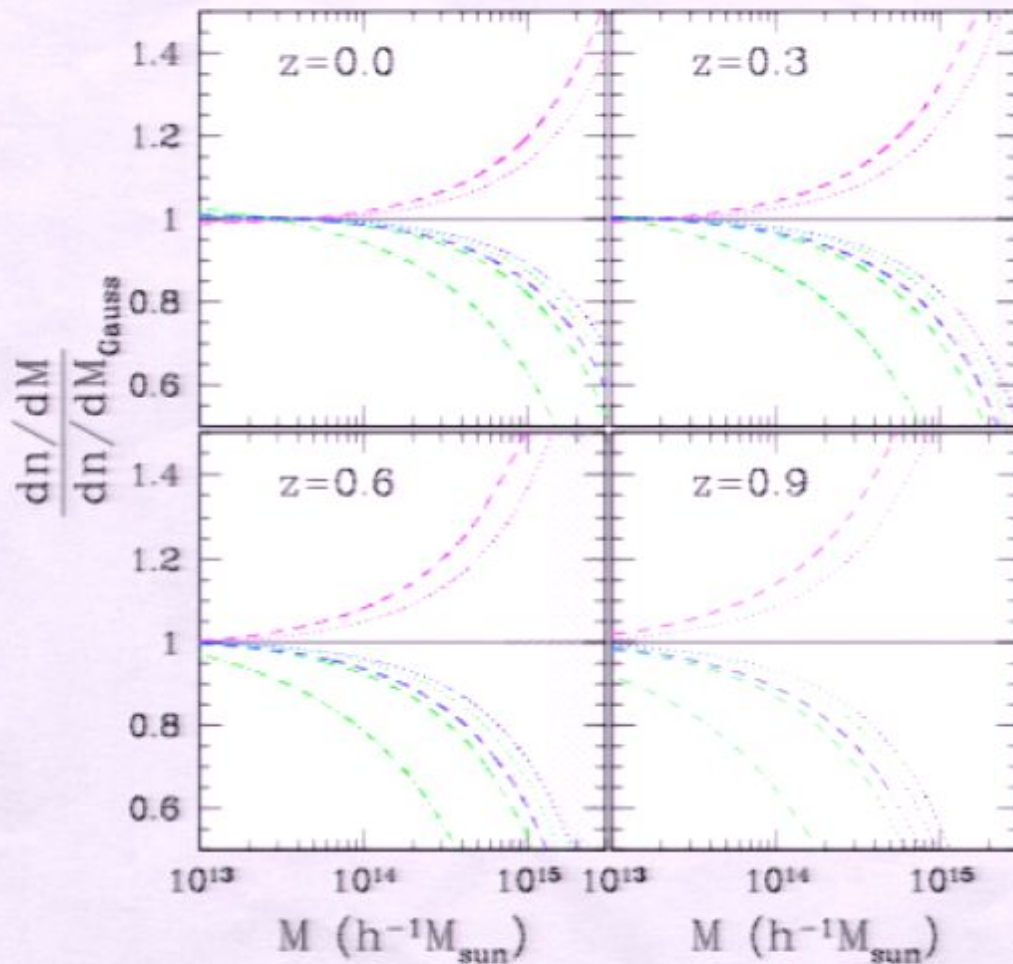
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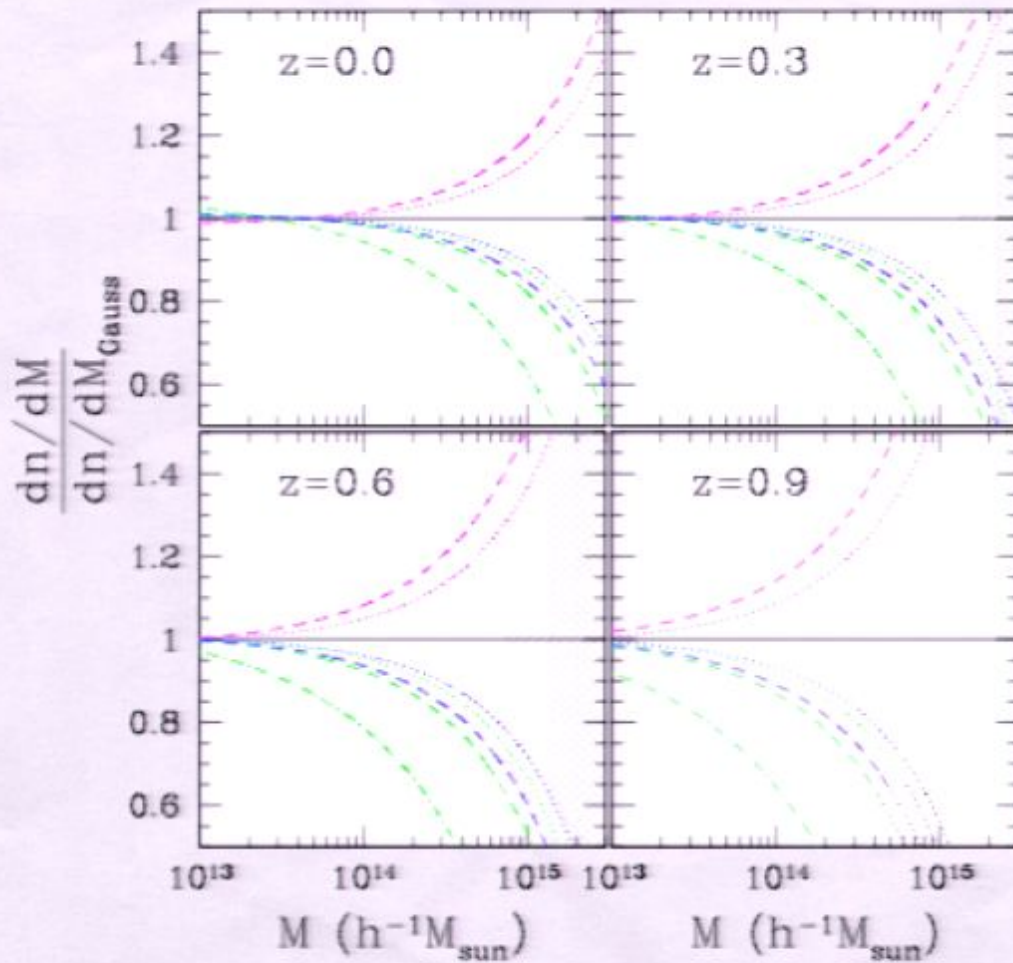
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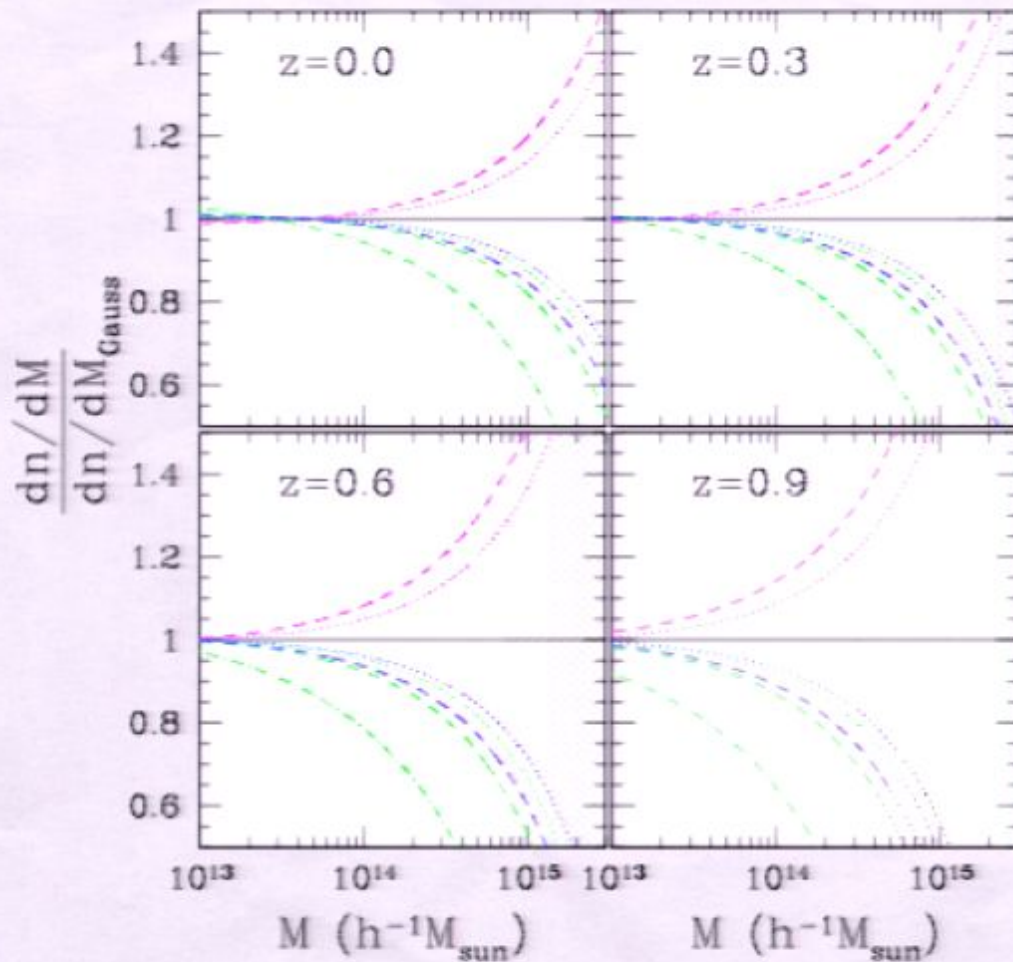


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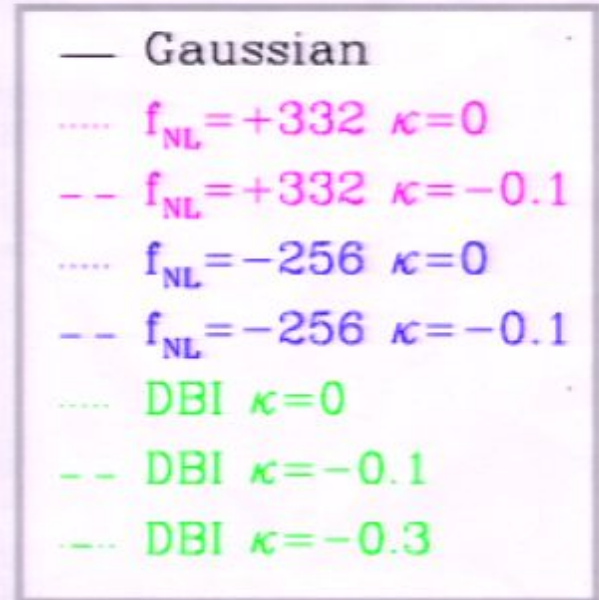
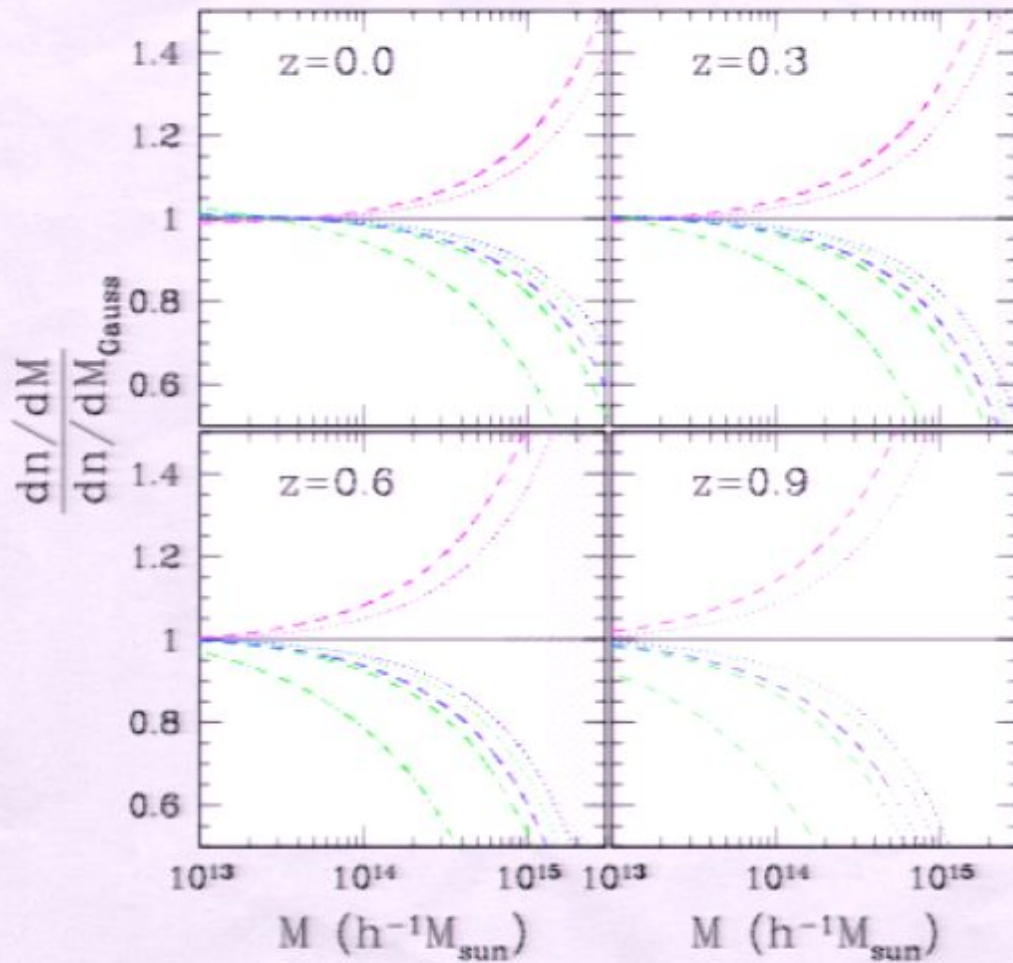


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- $f_{\text{NL}}=-256 \quad \kappa=-0.1$
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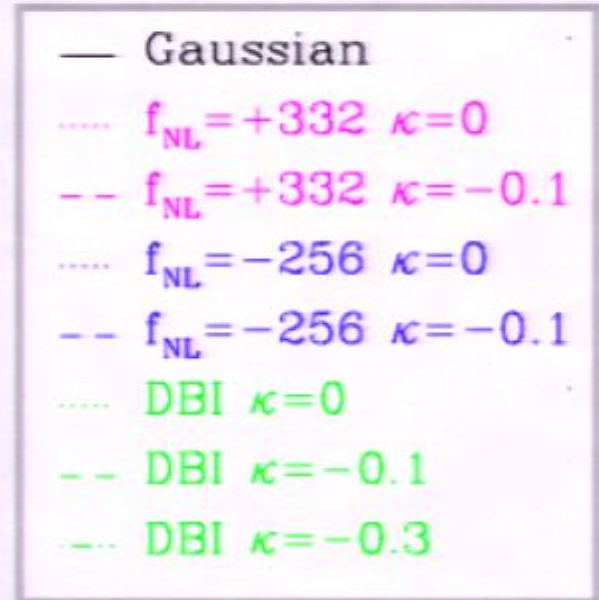
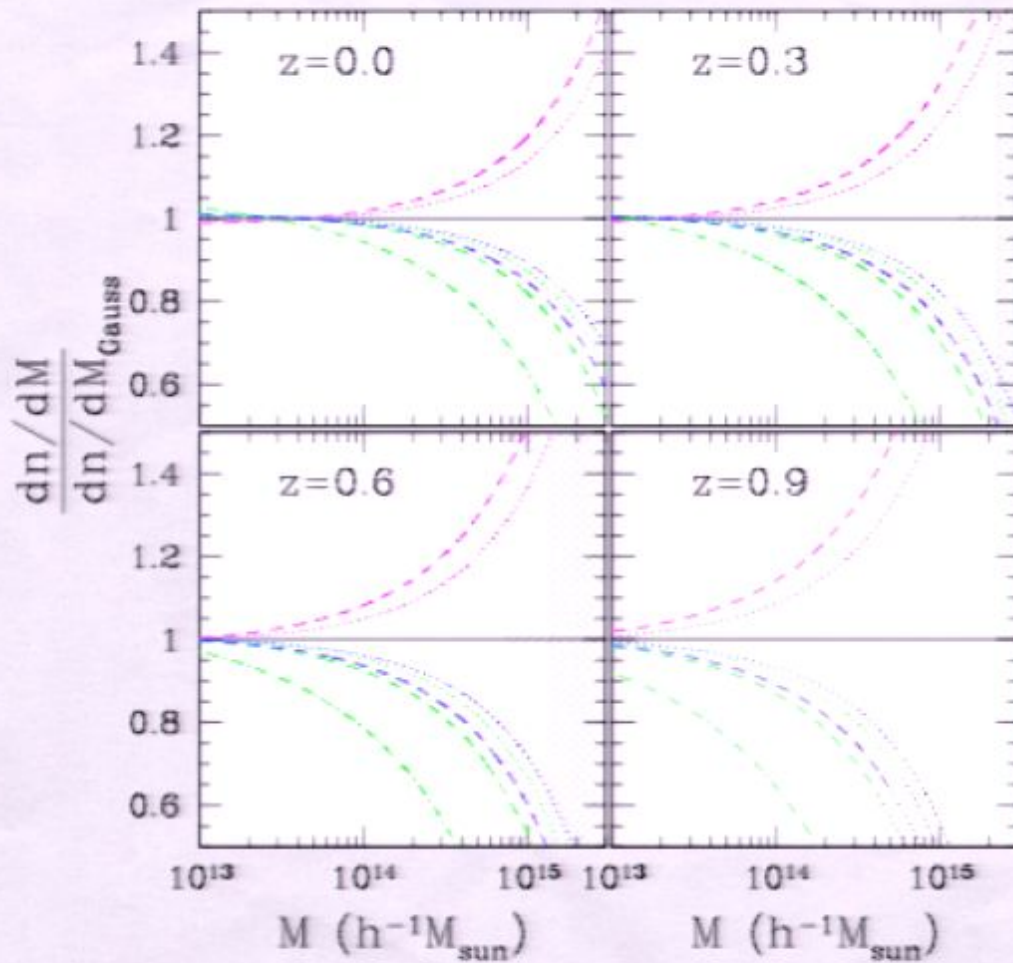
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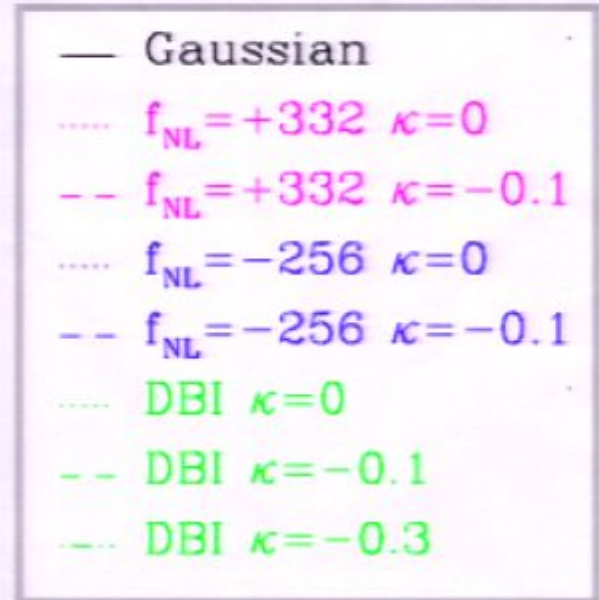
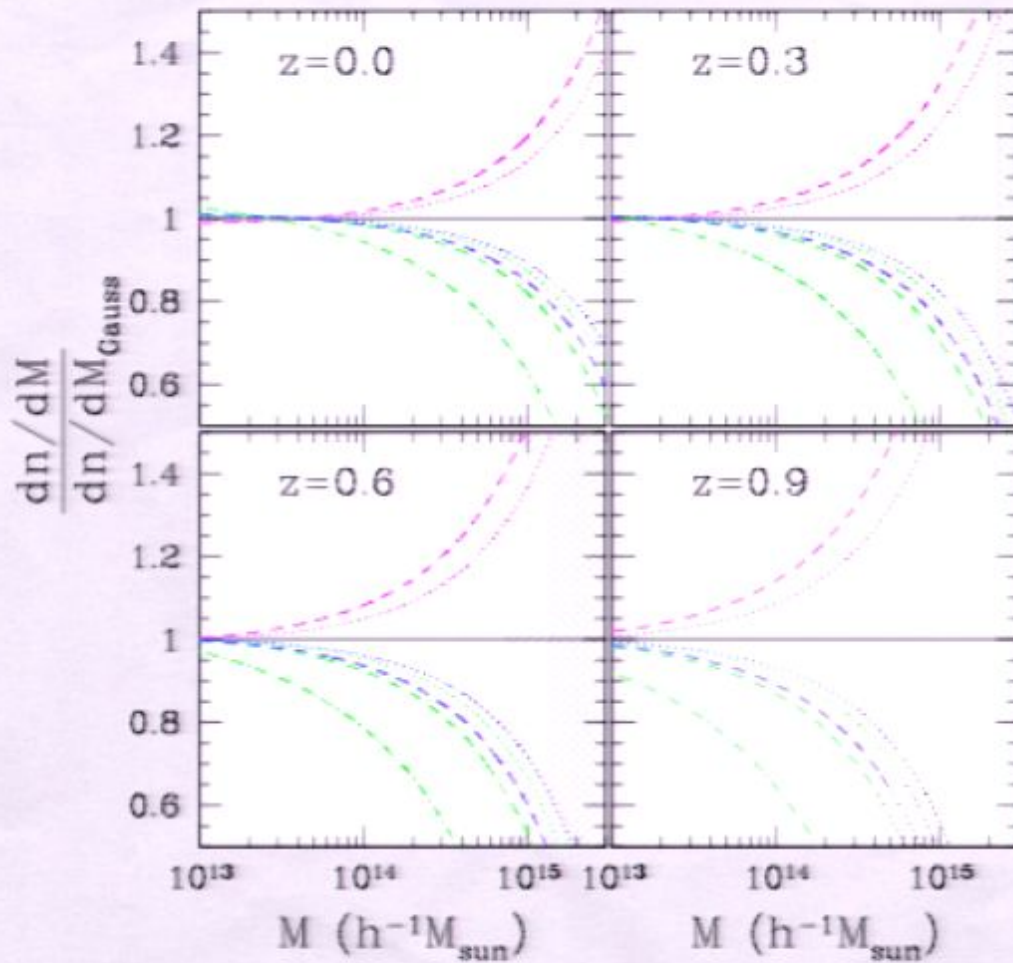
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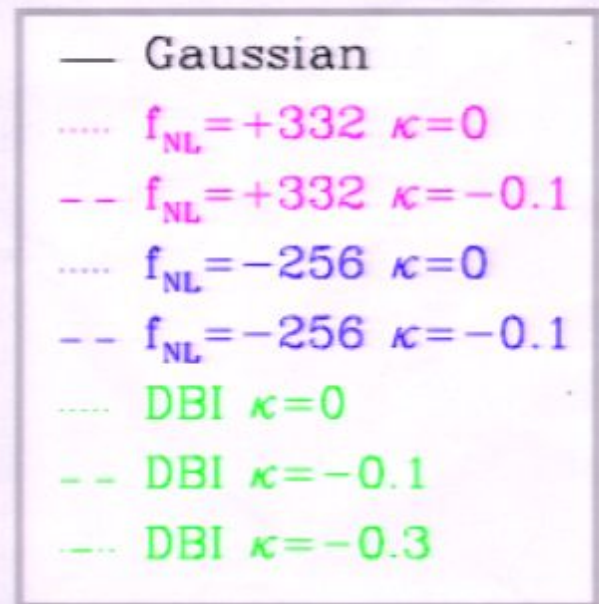
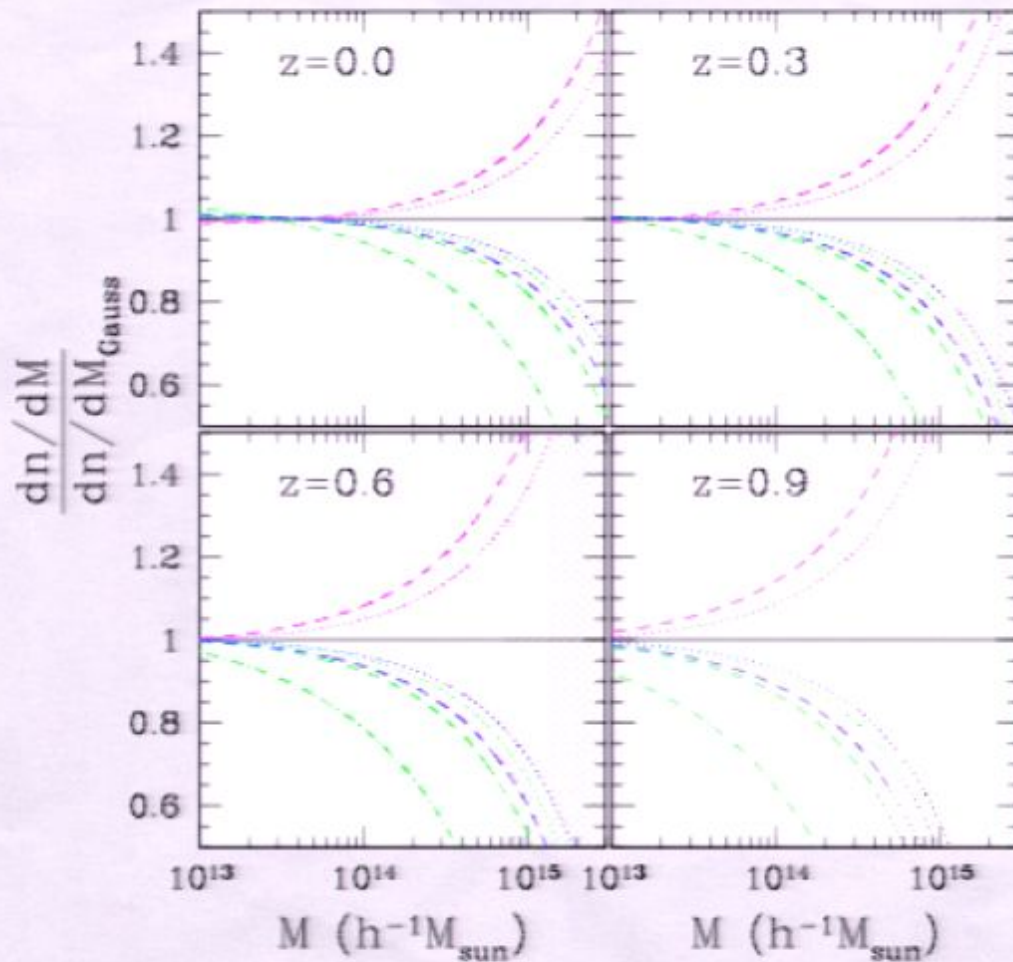
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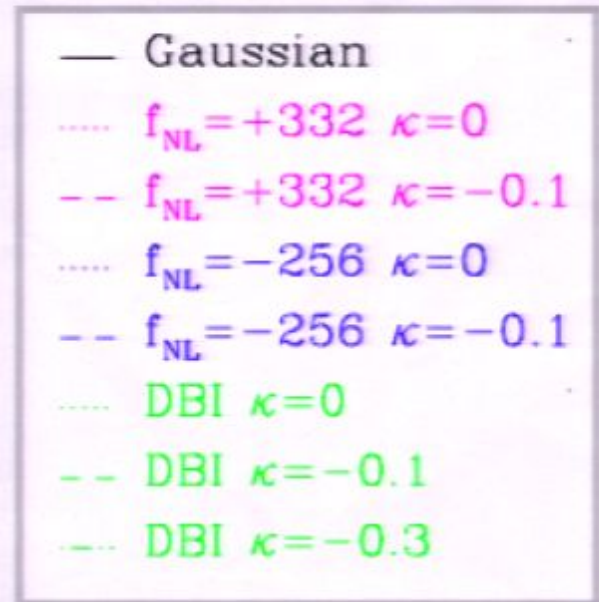
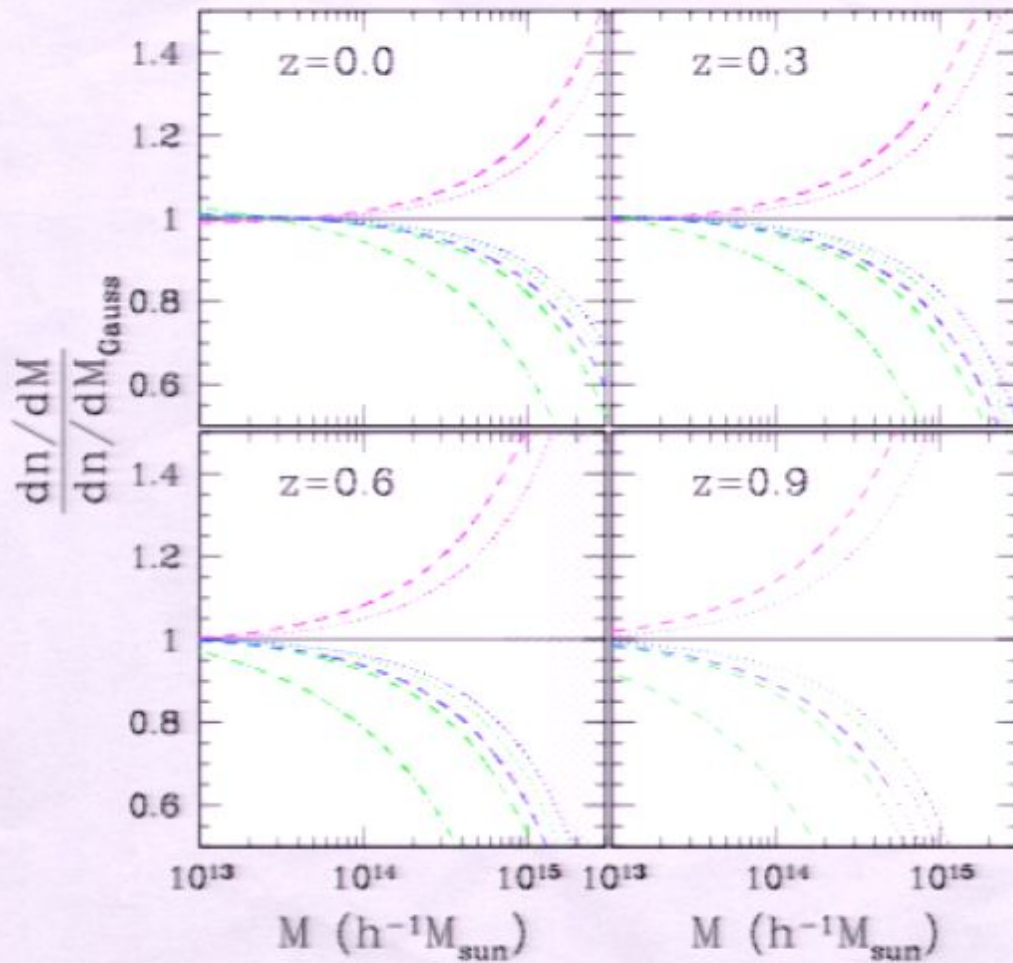
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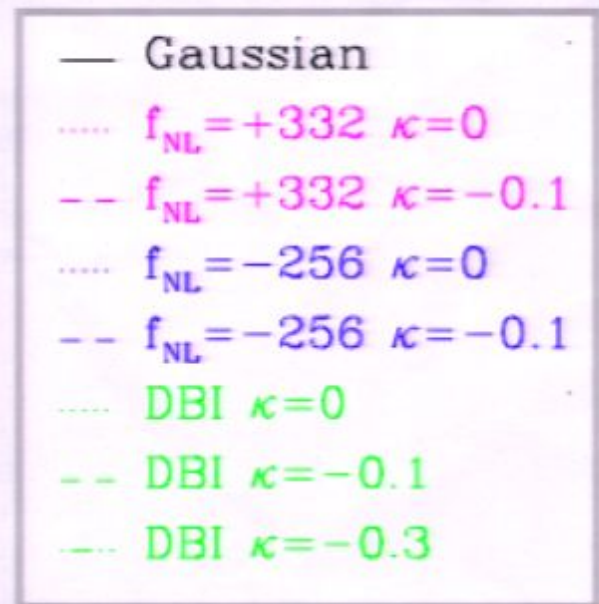
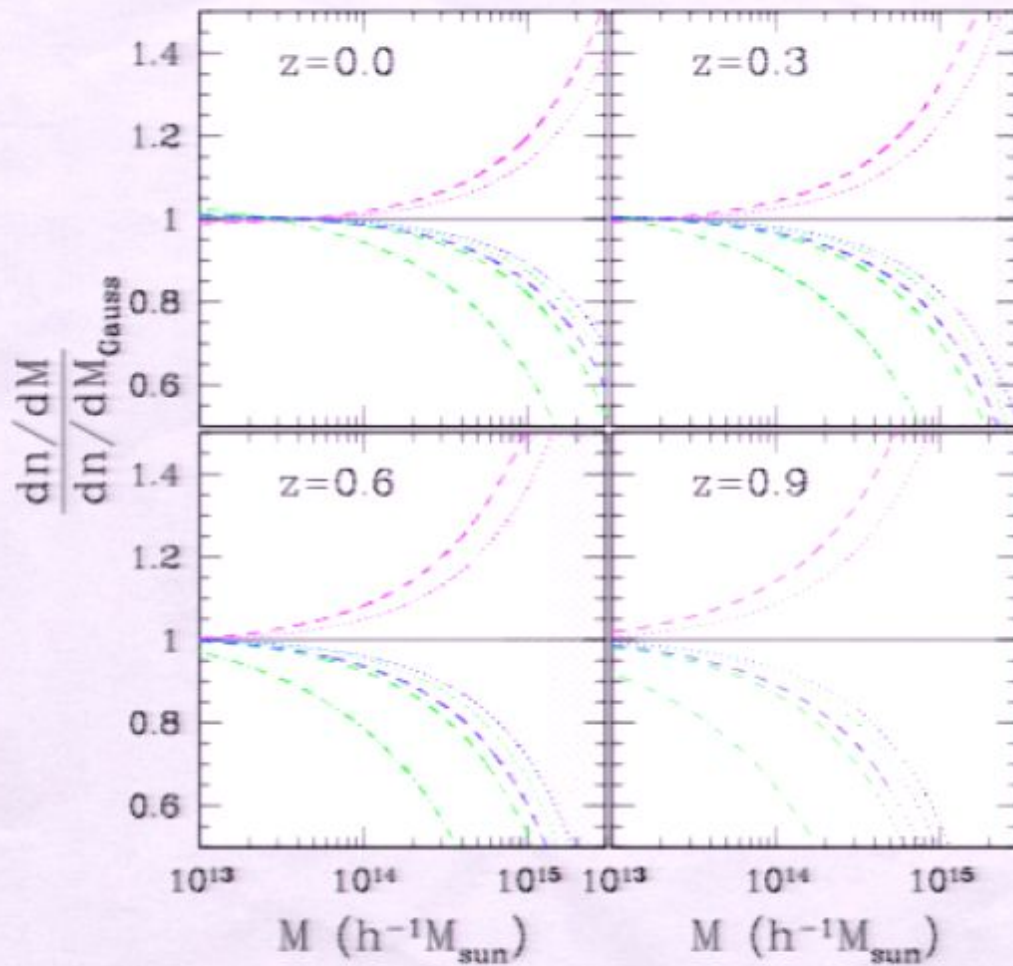
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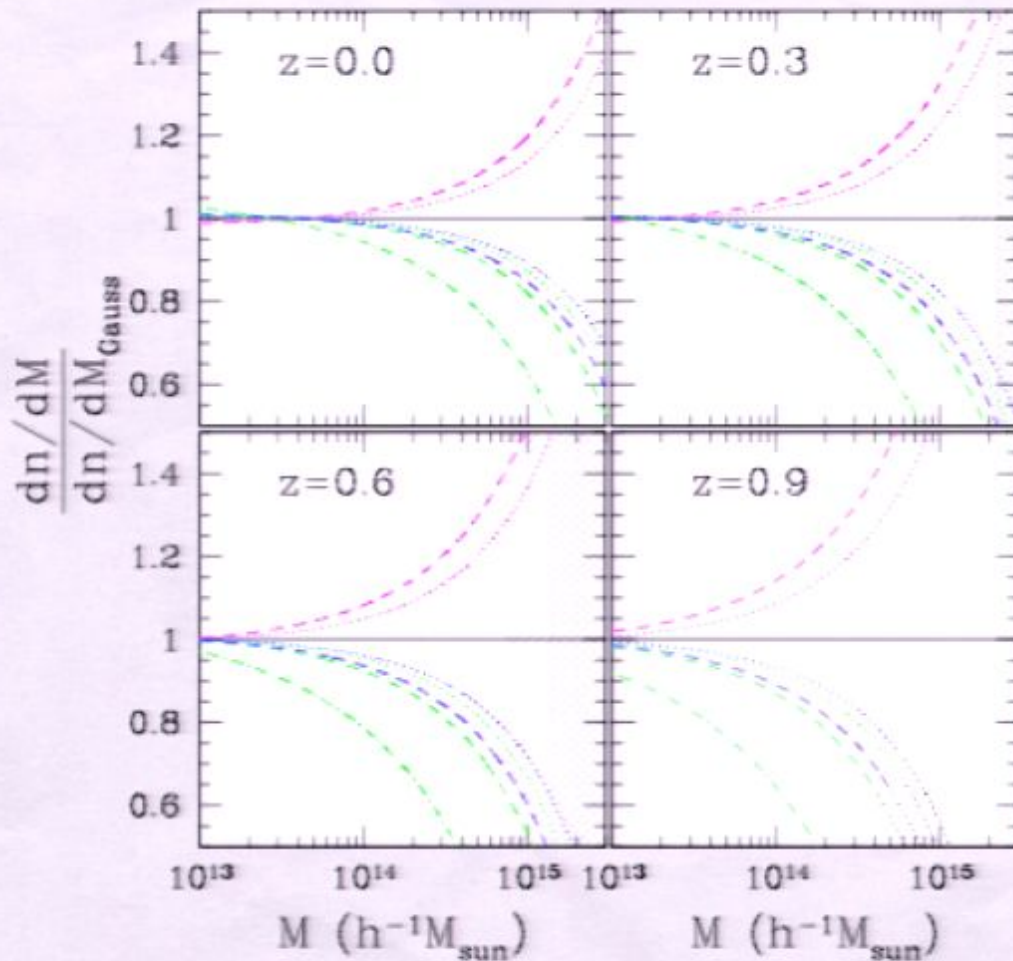
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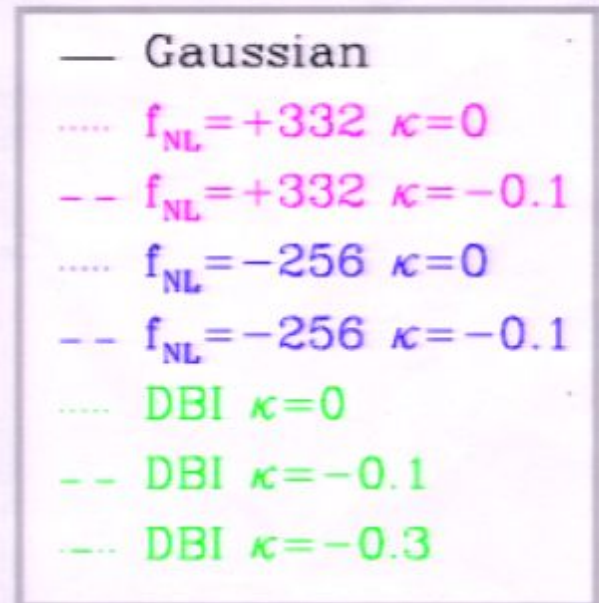
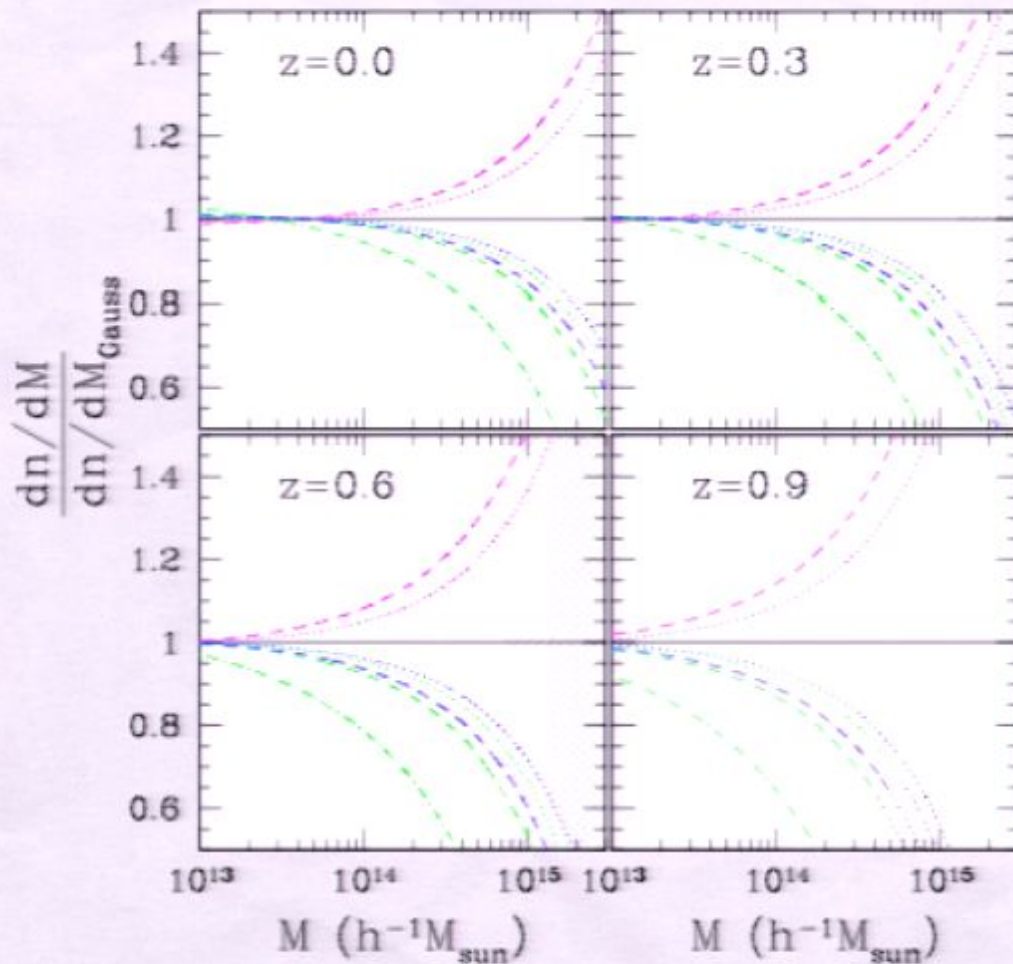


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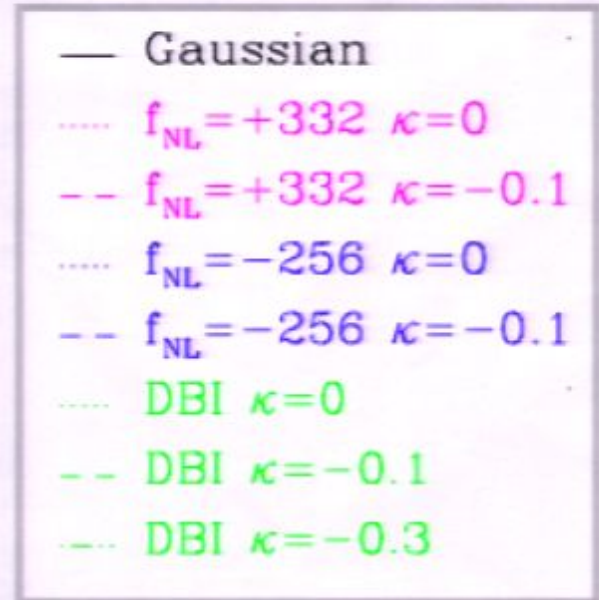
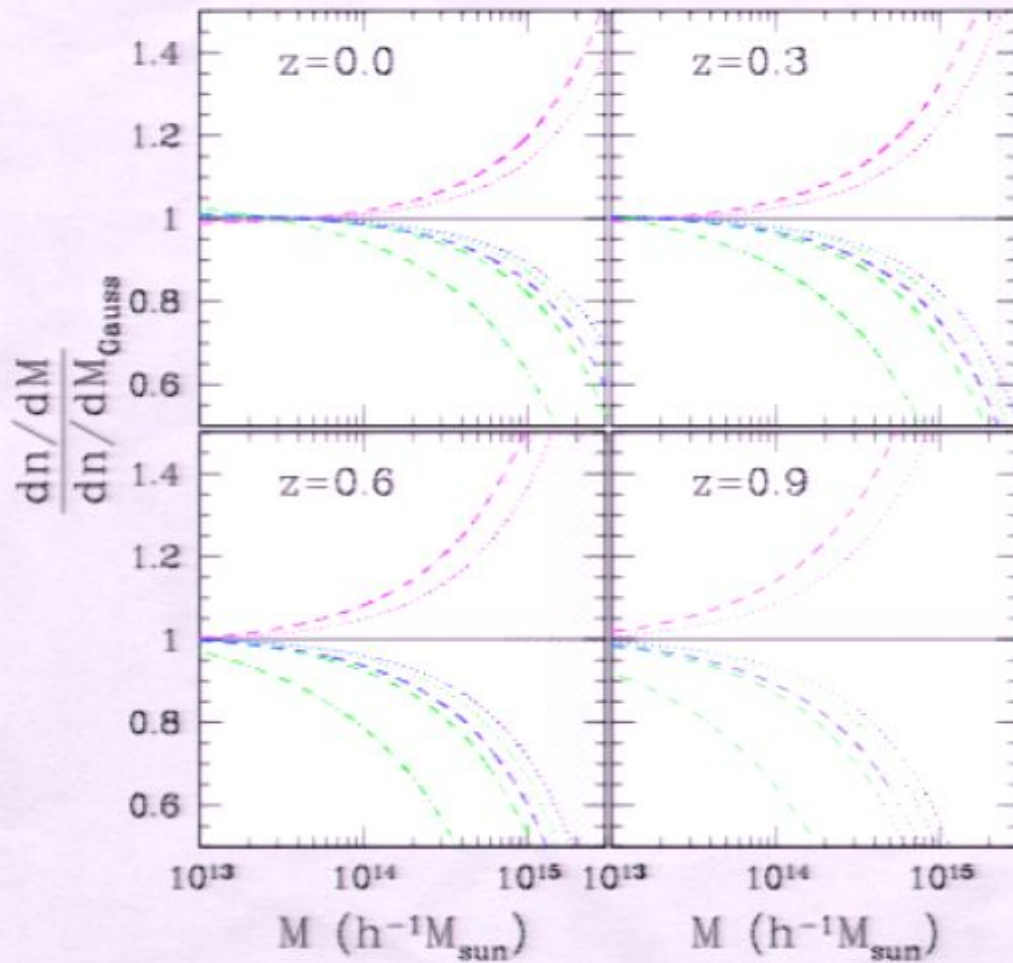


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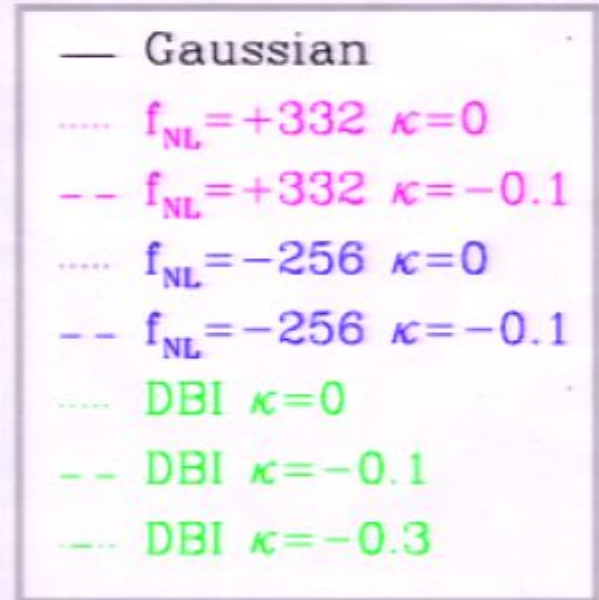
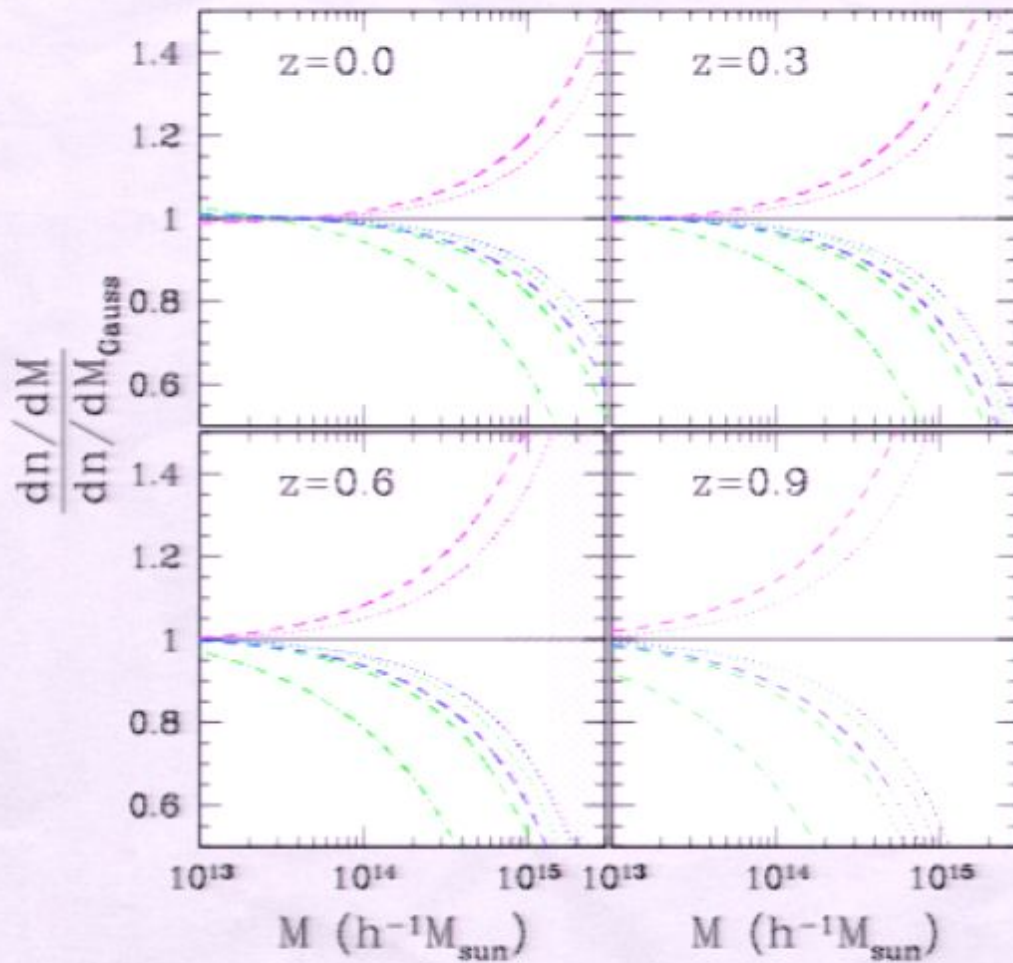
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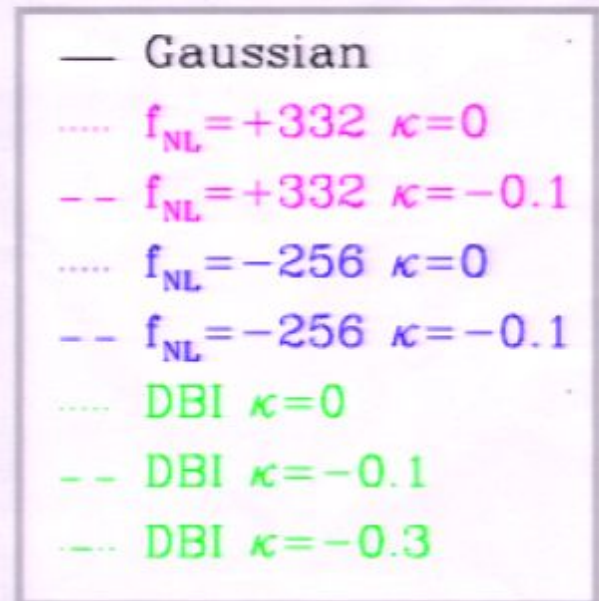
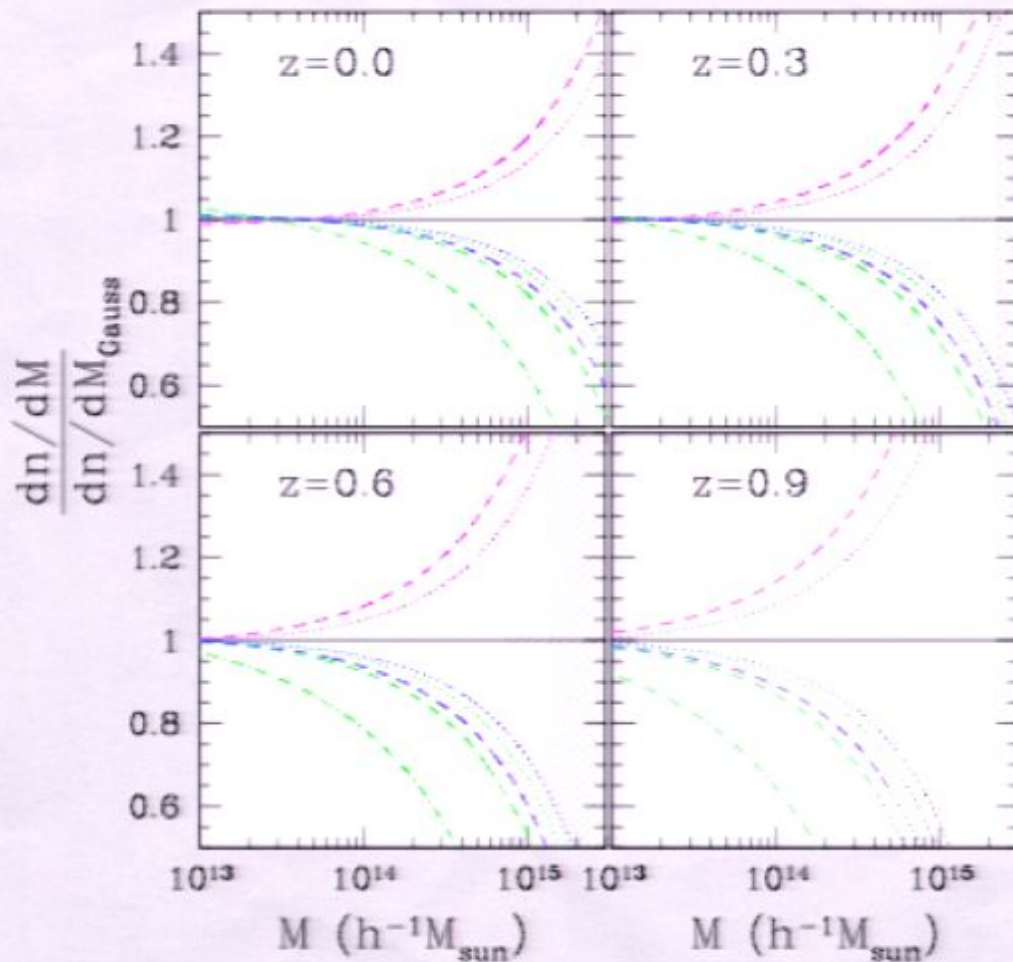
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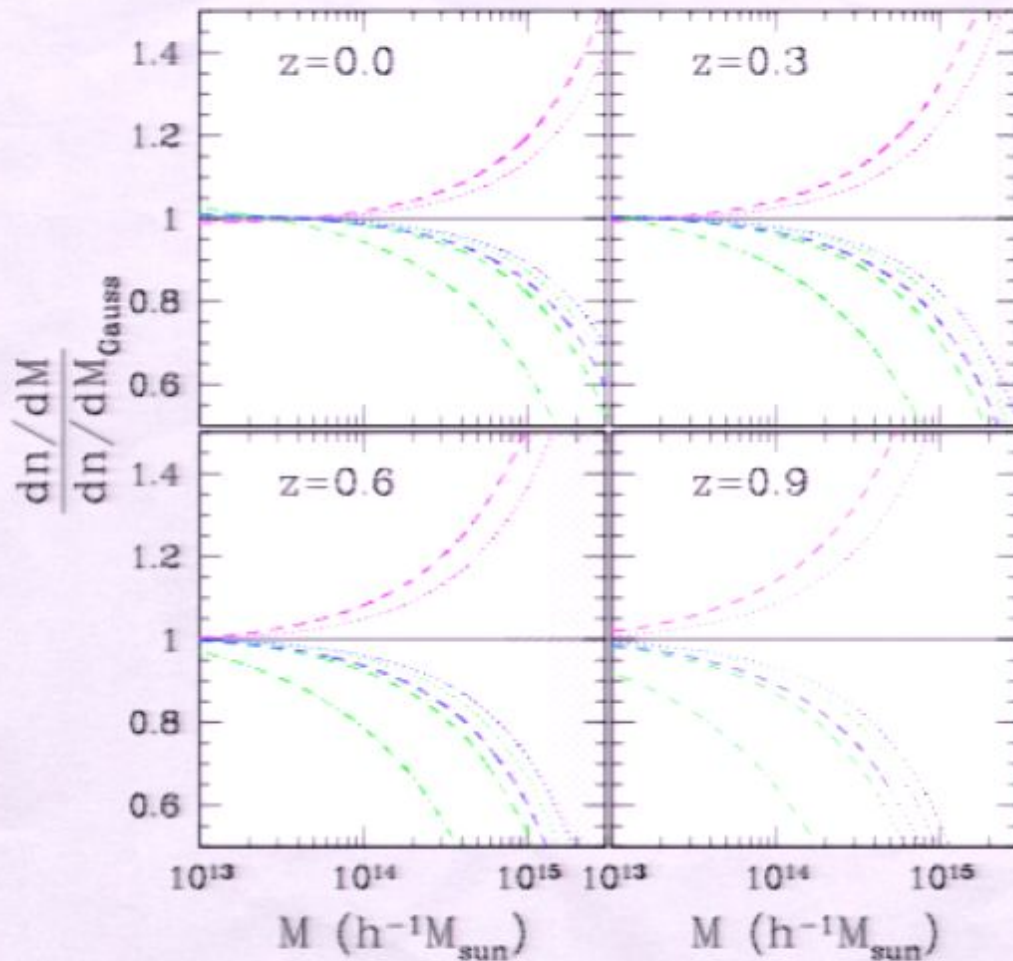
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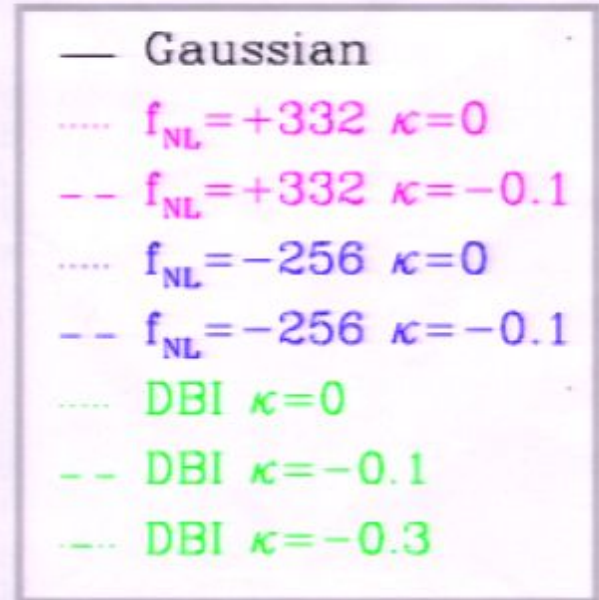
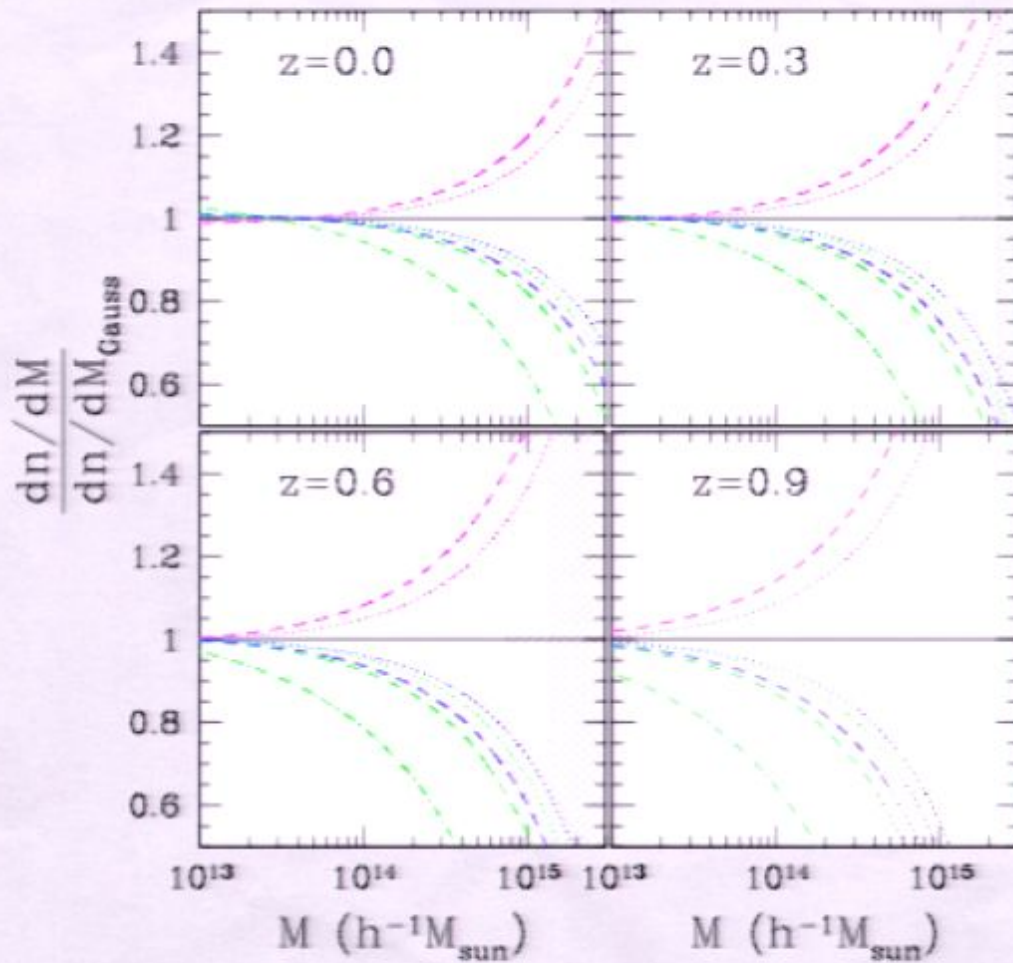


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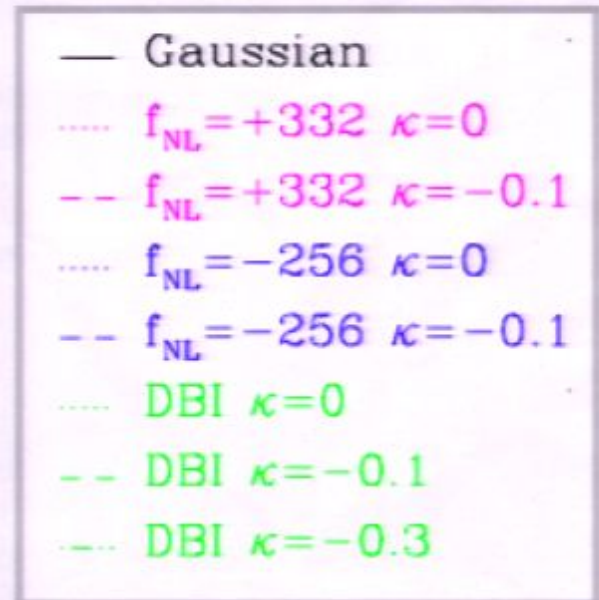
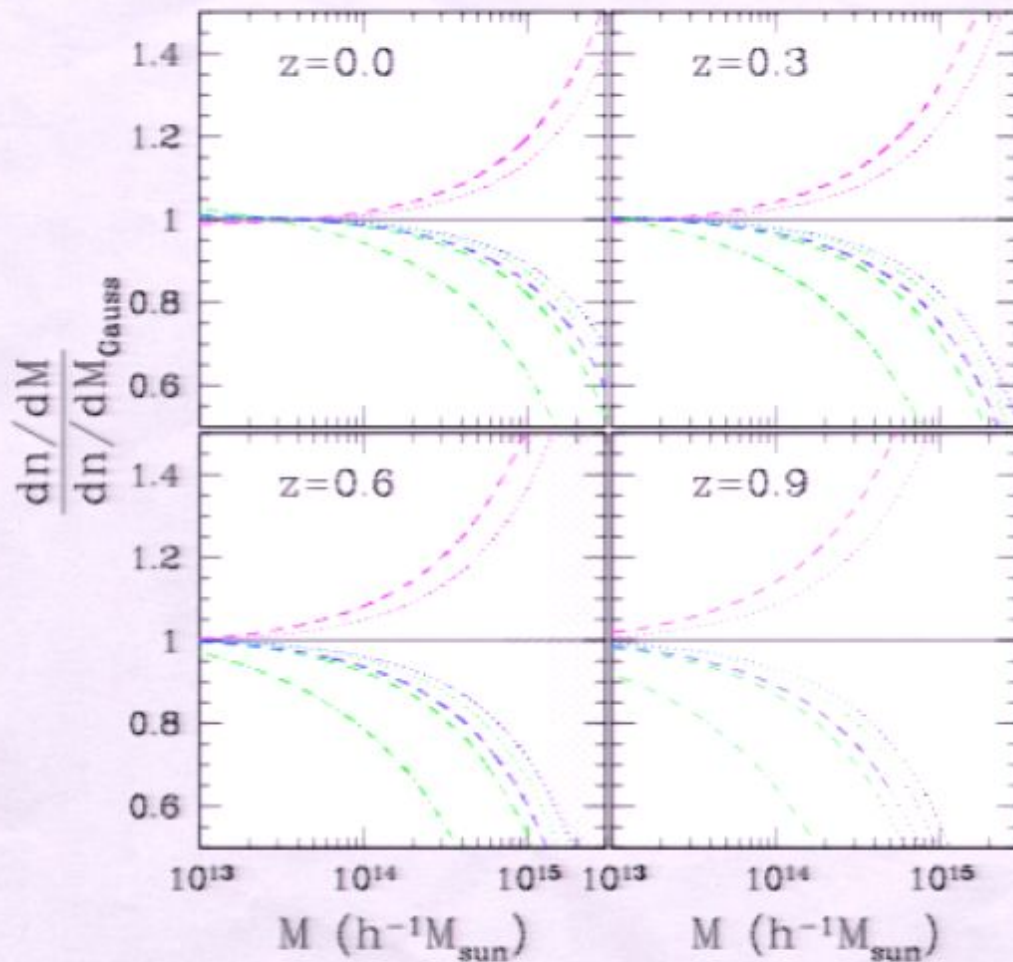


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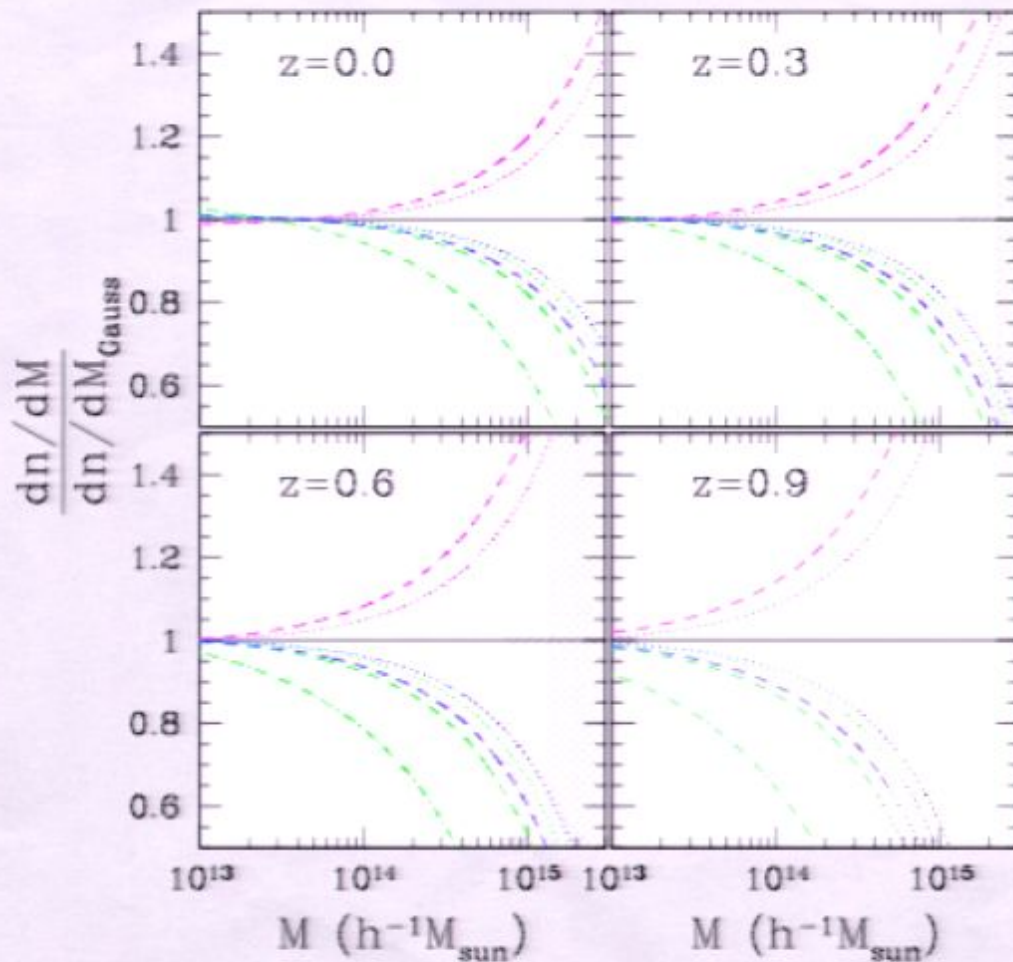
MASS FUNCTION



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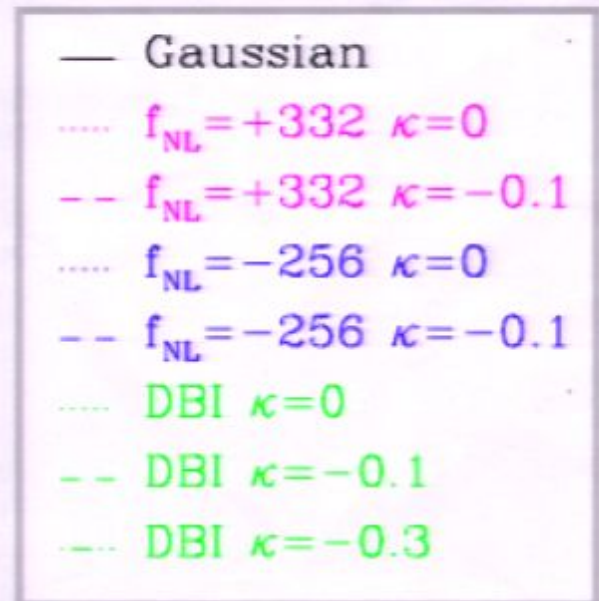
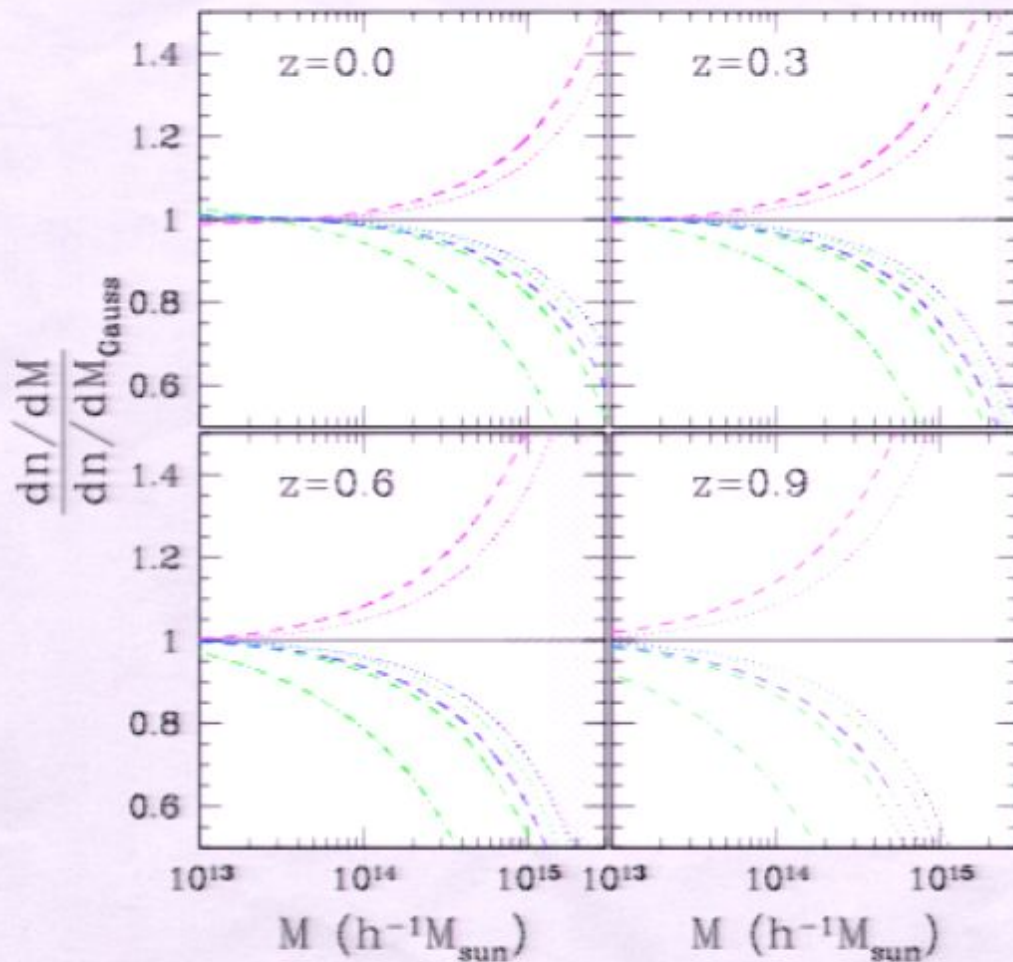


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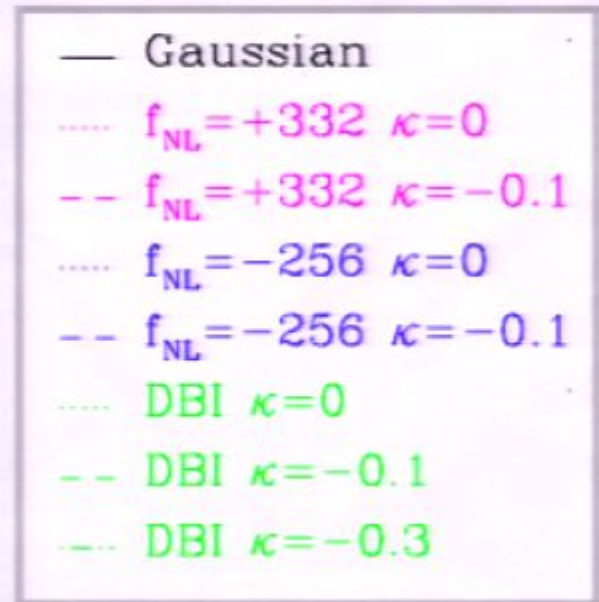
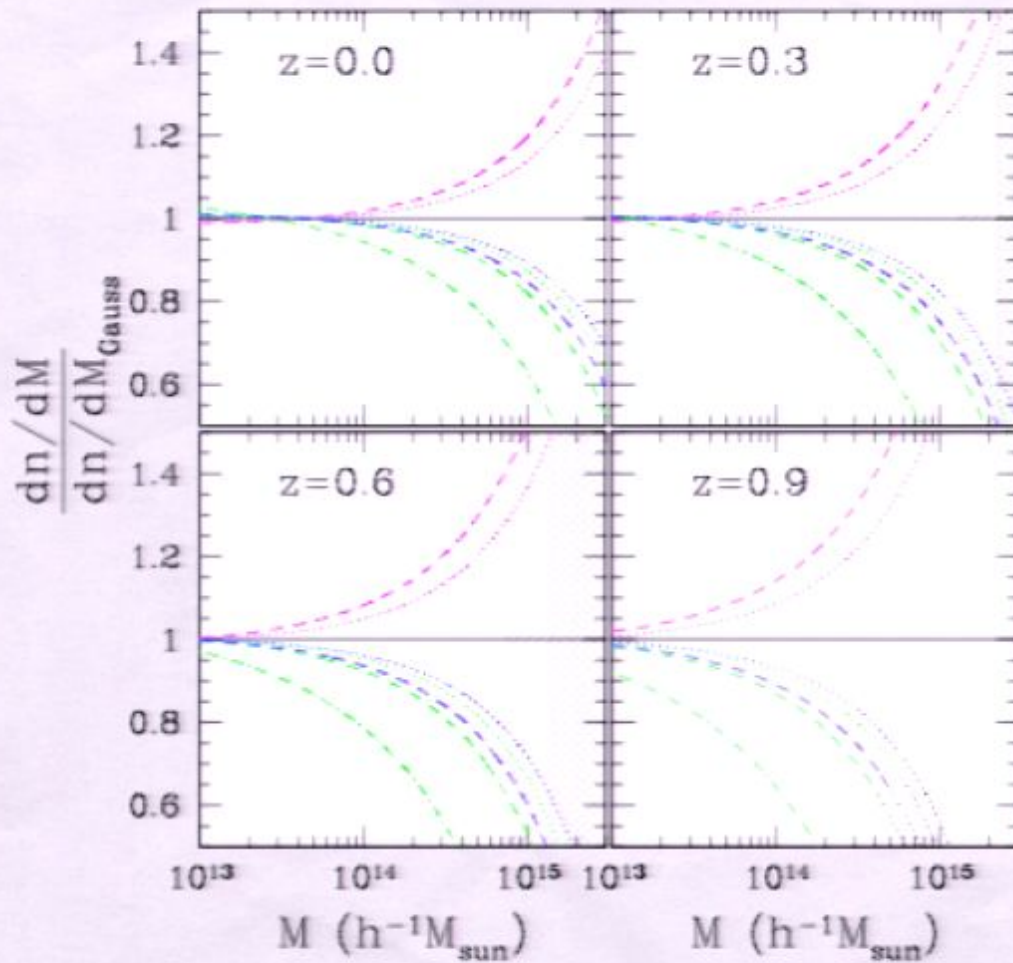


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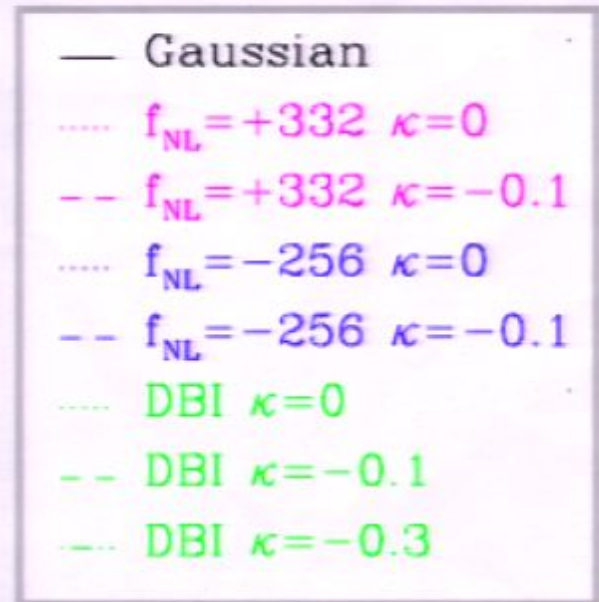
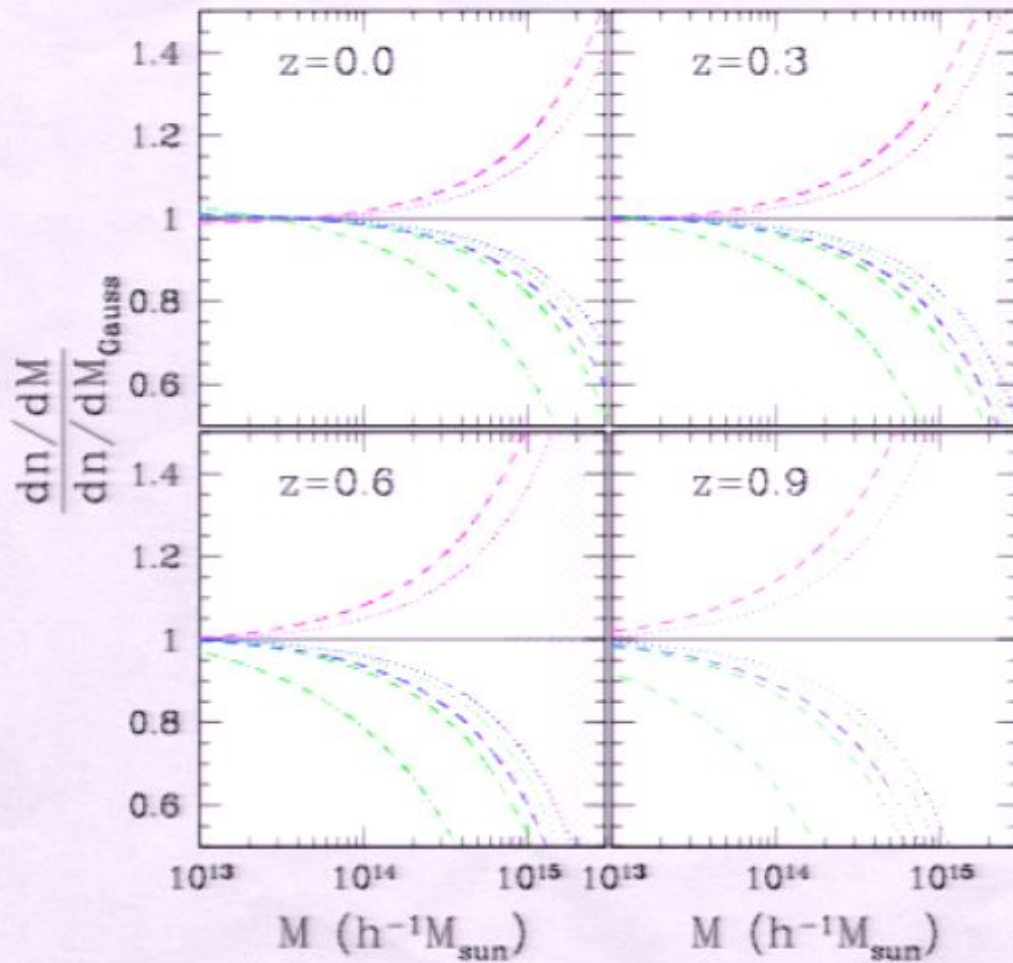
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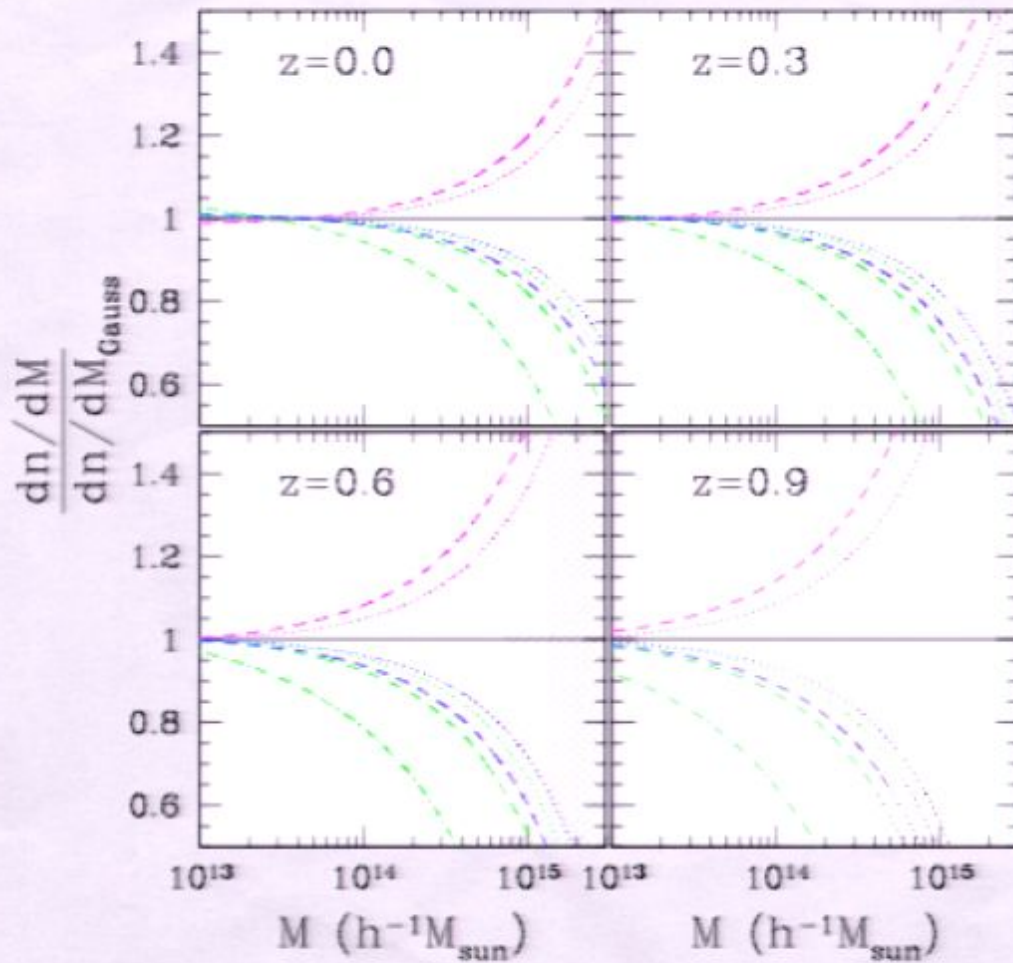
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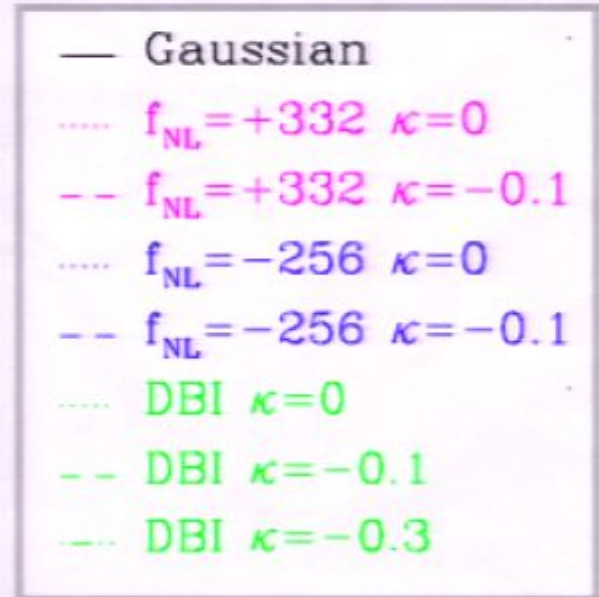
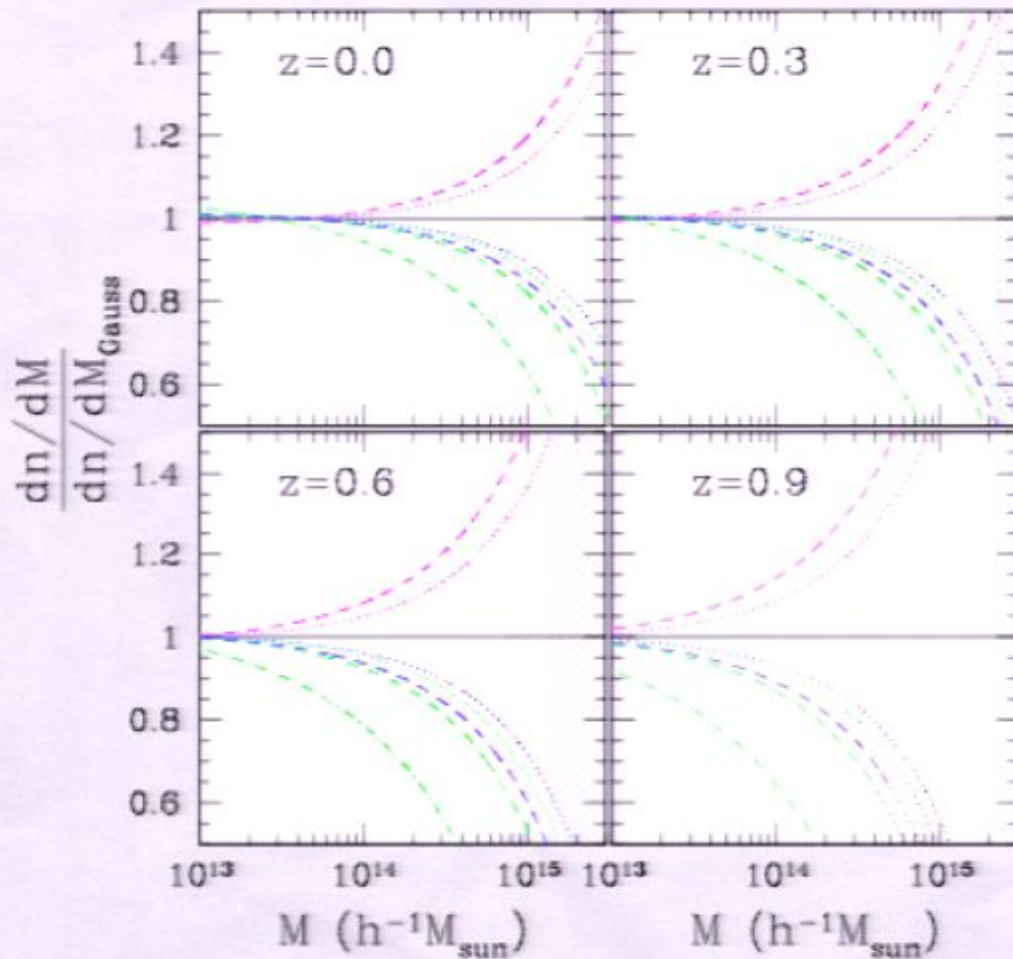
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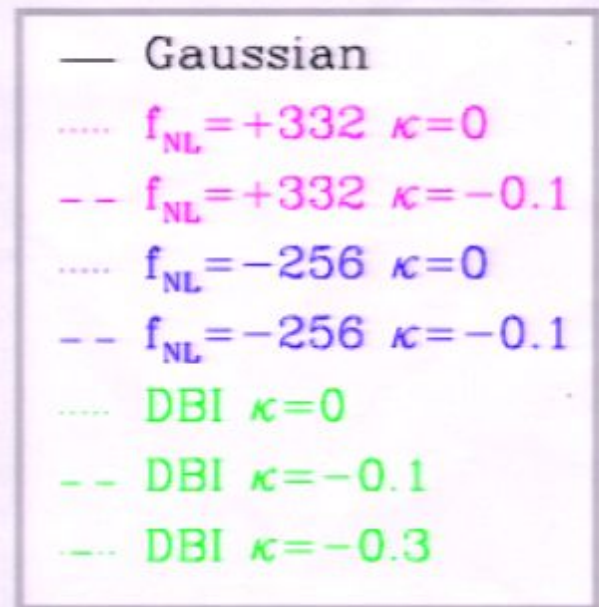
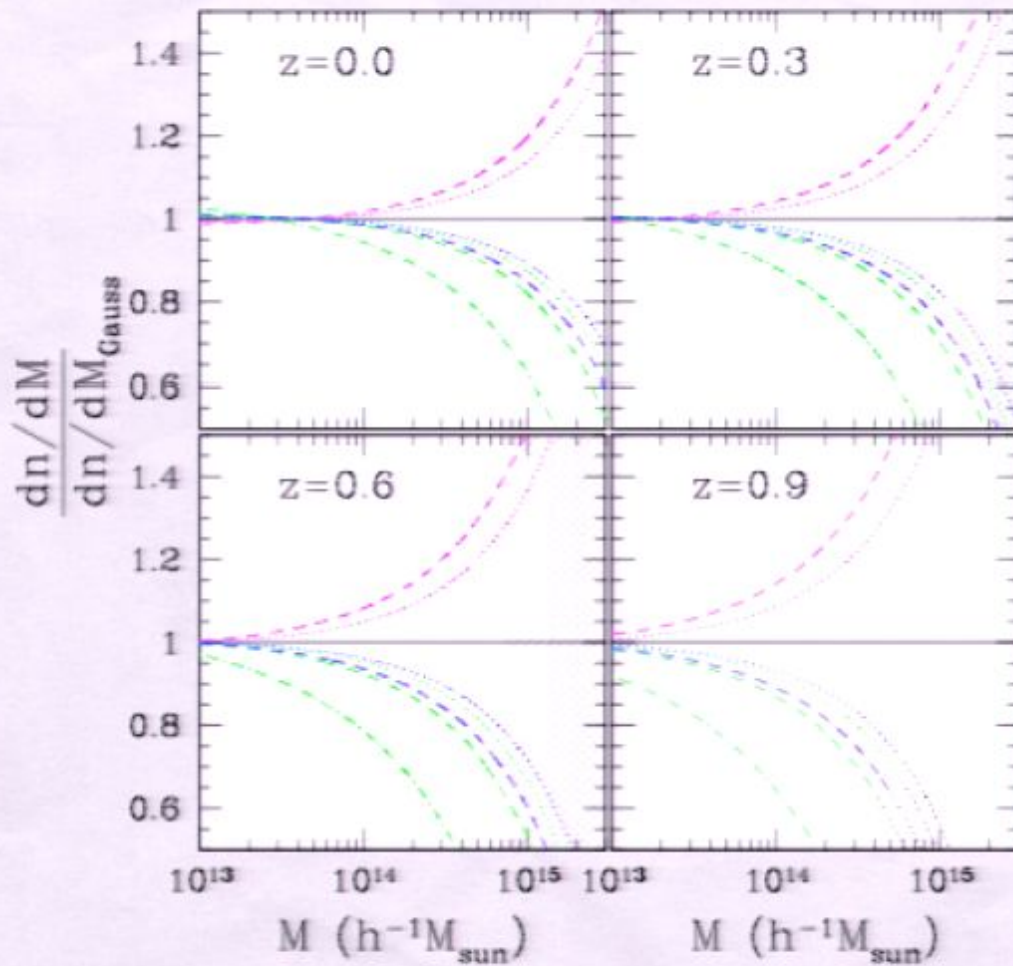
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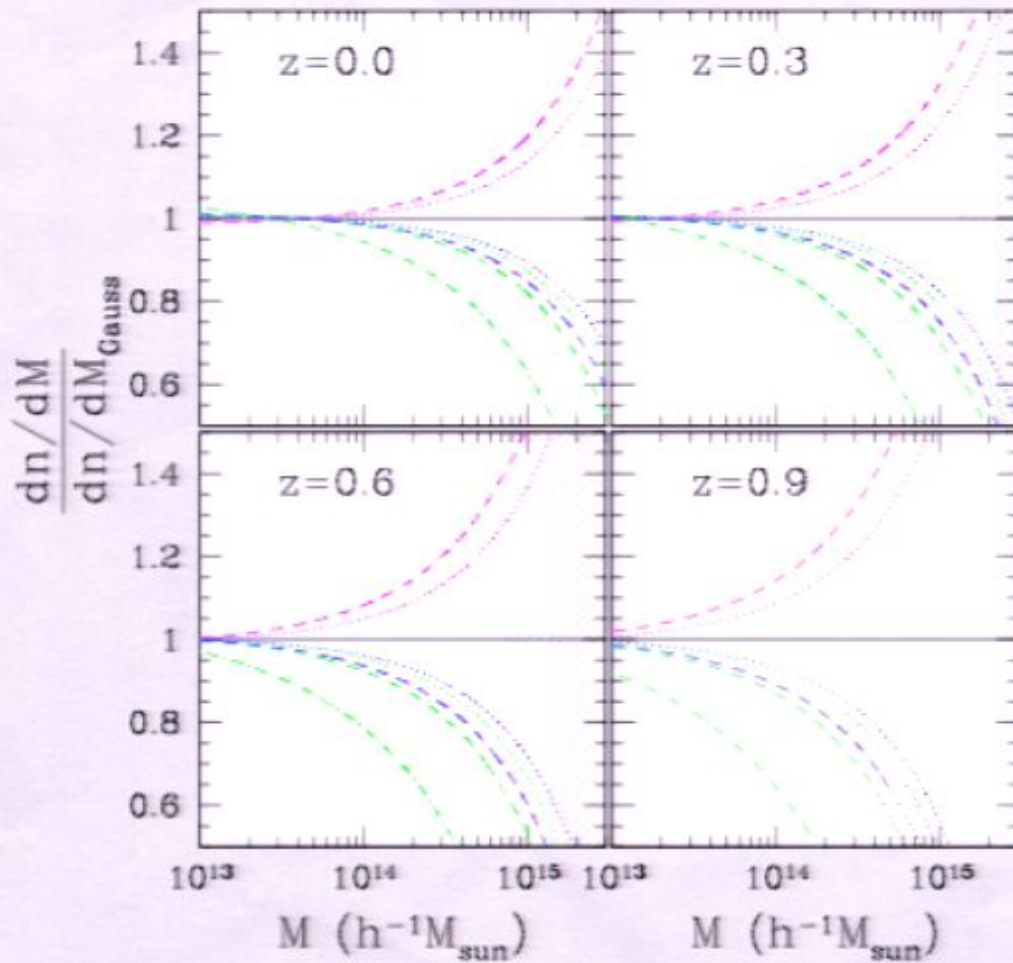
MASS FUNCTION



MASS FUNCTION

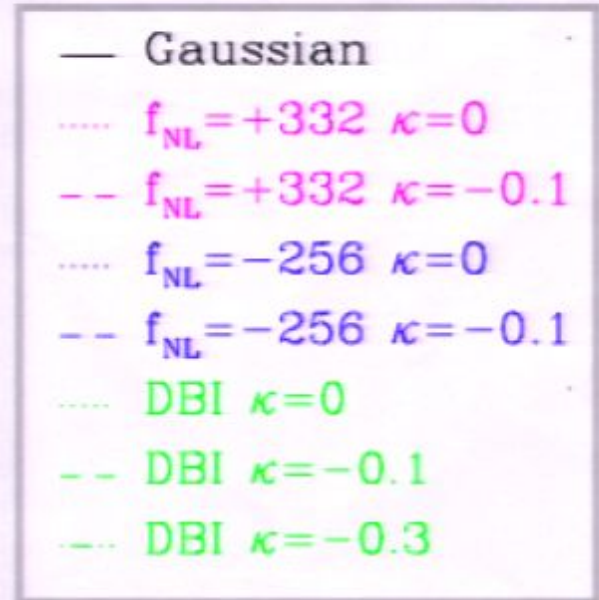
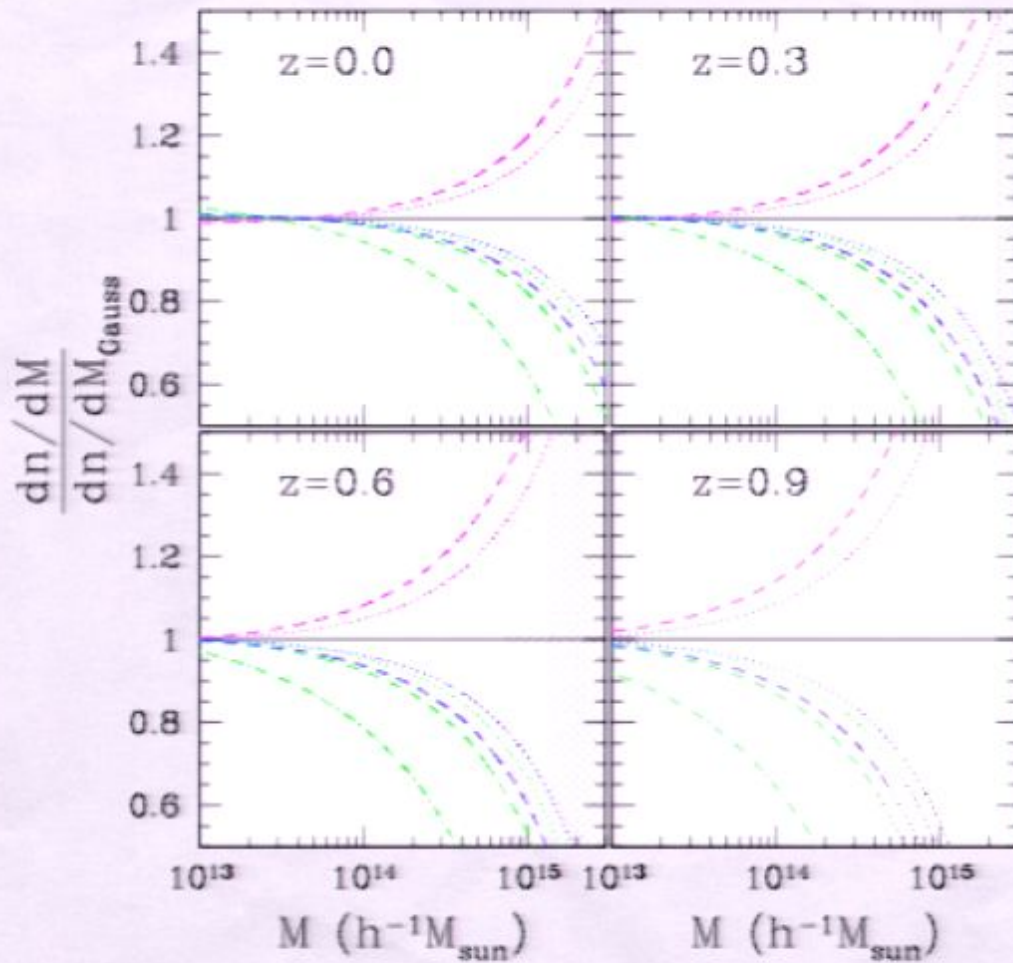


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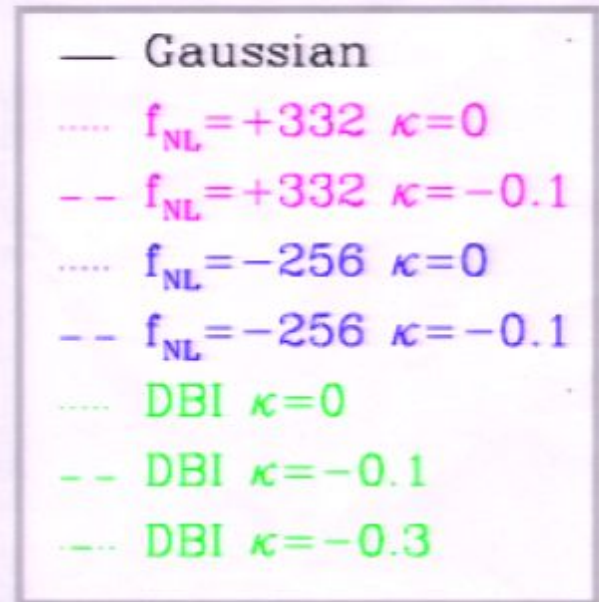
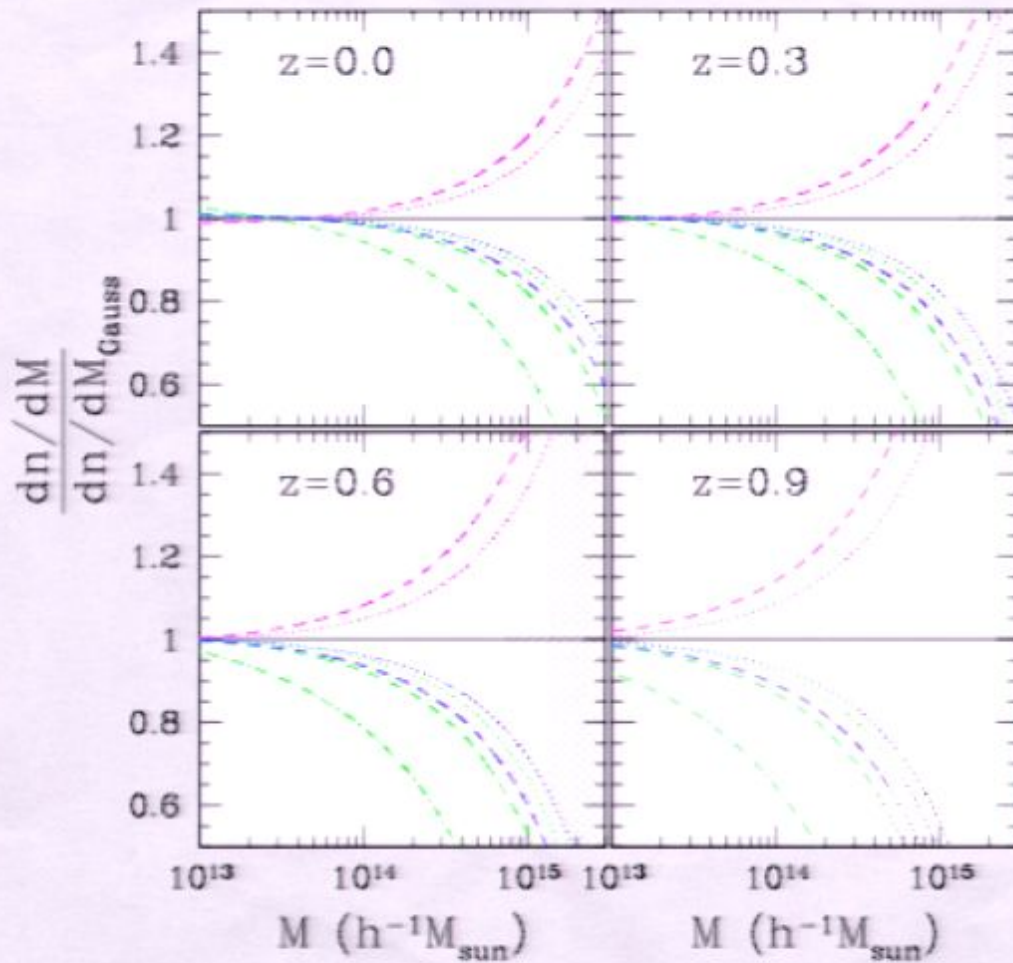


- Gaussian
- $f_{\text{NL}} = +332 \quad \kappa = 0$
- $f_{\text{NL}} = +332 \quad \kappa = -0.1$
- $f_{\text{NL}} = -256 \quad \kappa = 0$
- $f_{\text{NL}} = -256 \quad \kappa = -0.1$
- DBI $\kappa = 0$
- DBI $\kappa = -0.1$
- DBI $\kappa = -0.3$

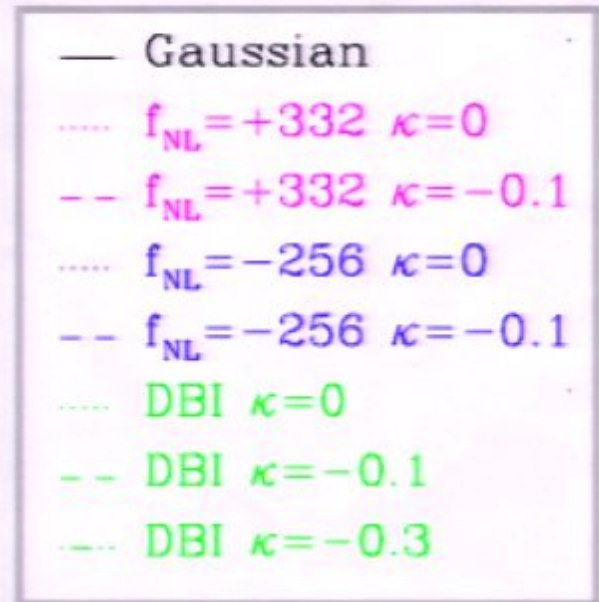
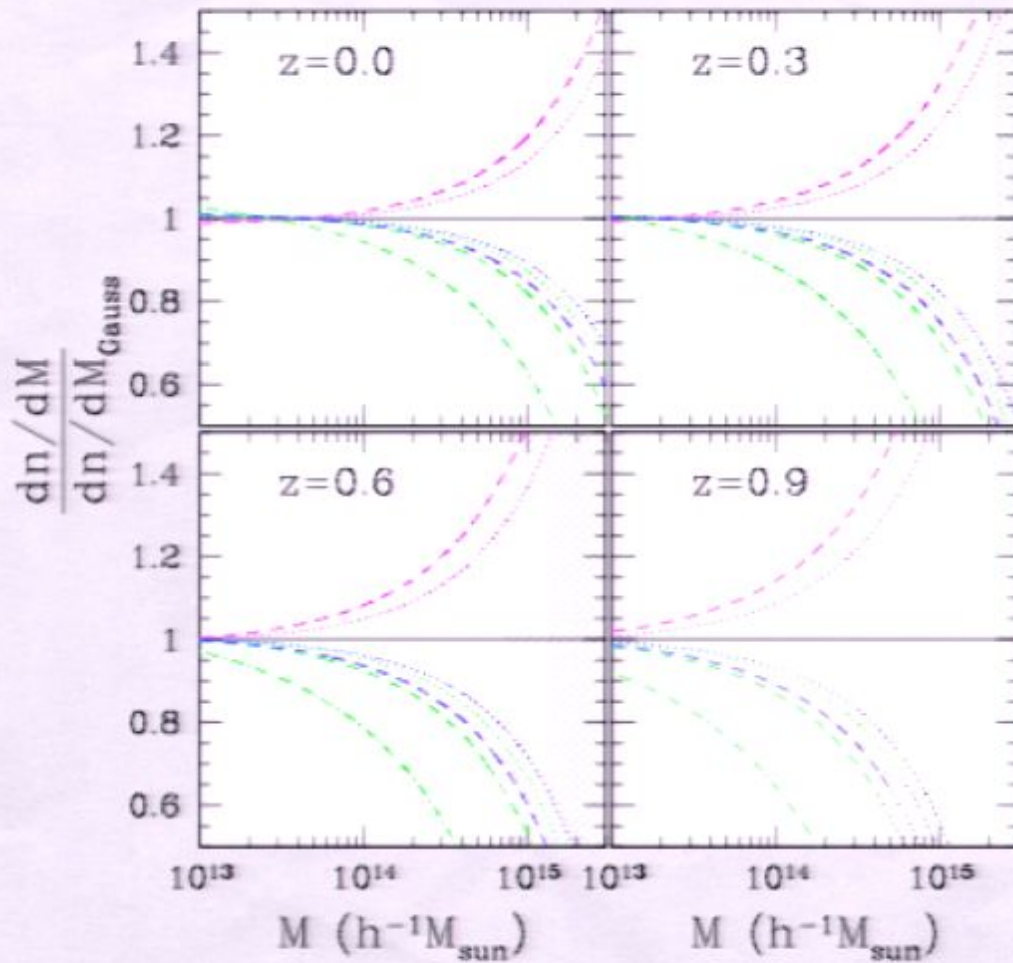
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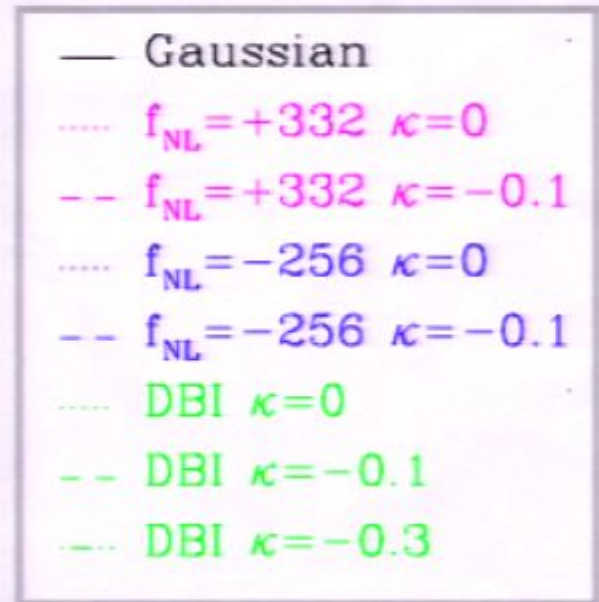
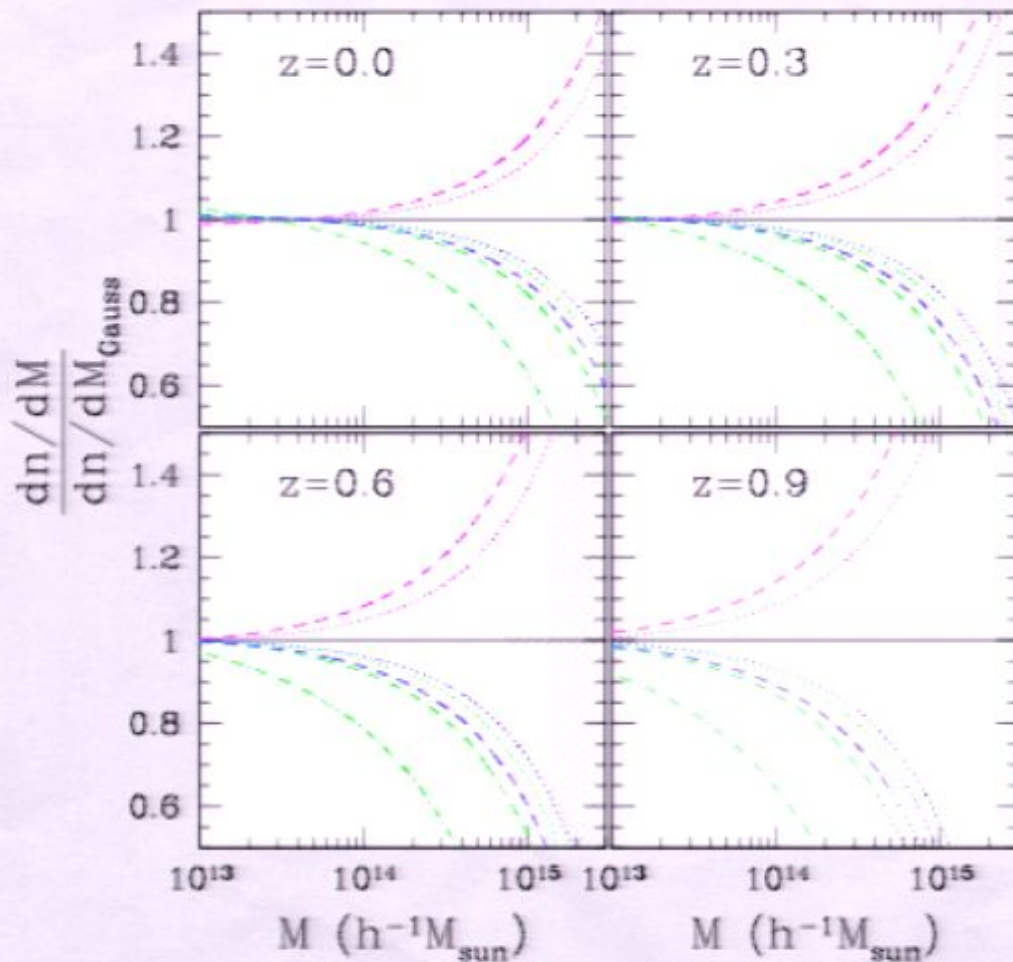
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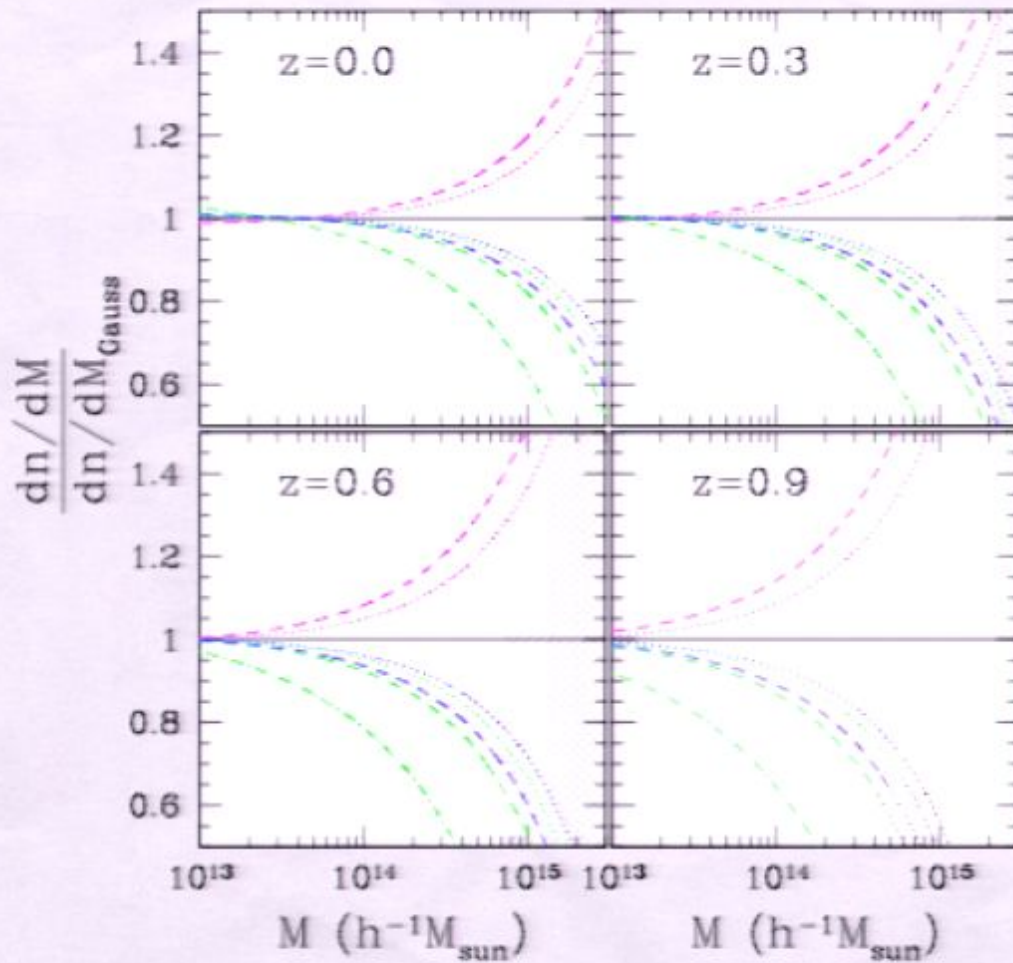
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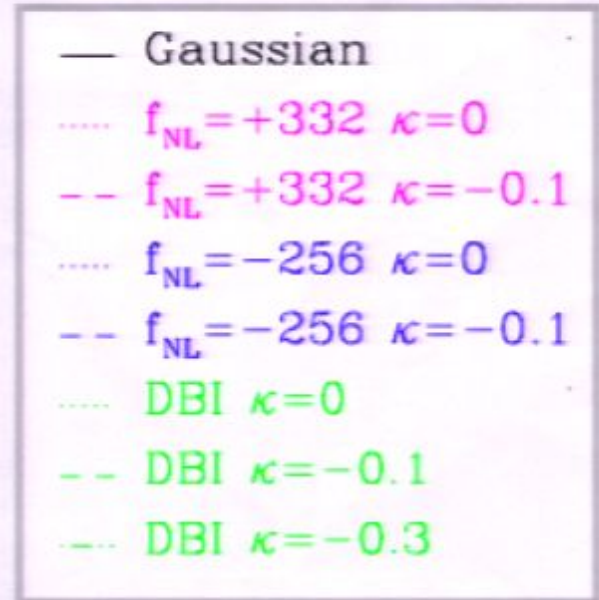
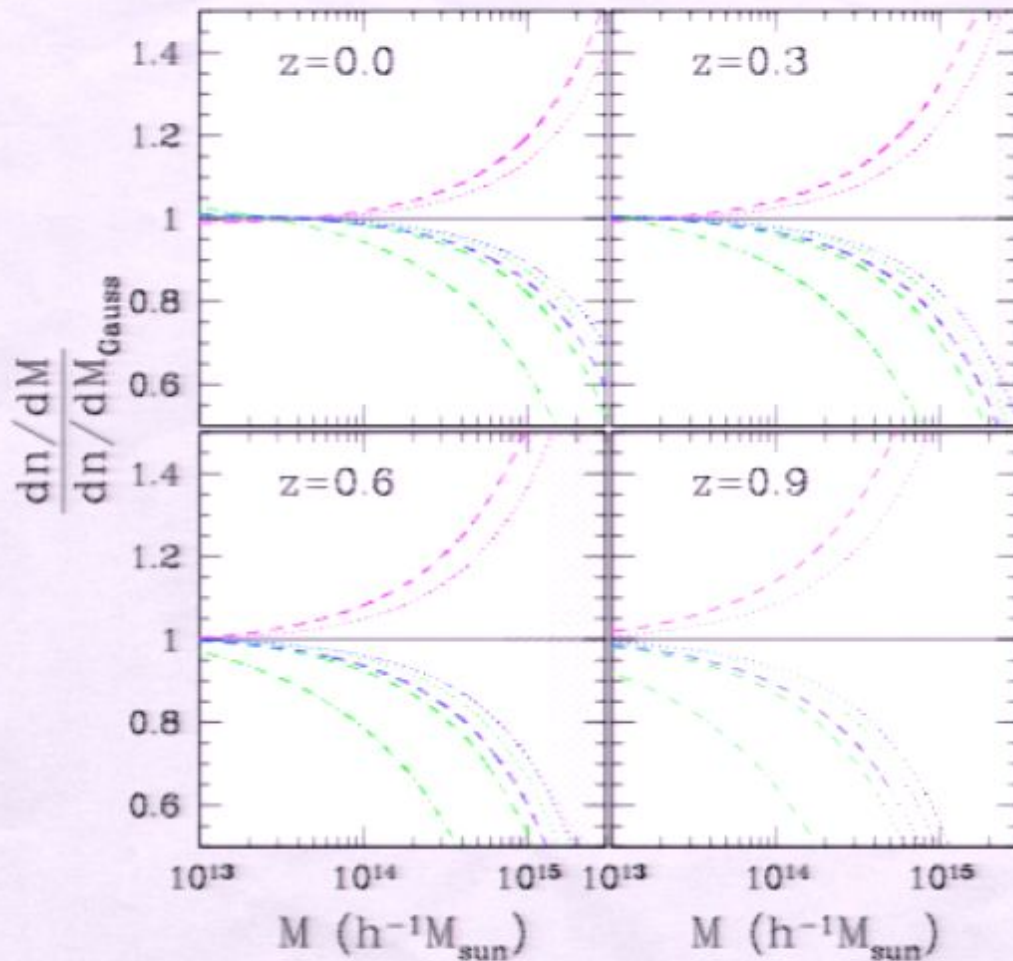
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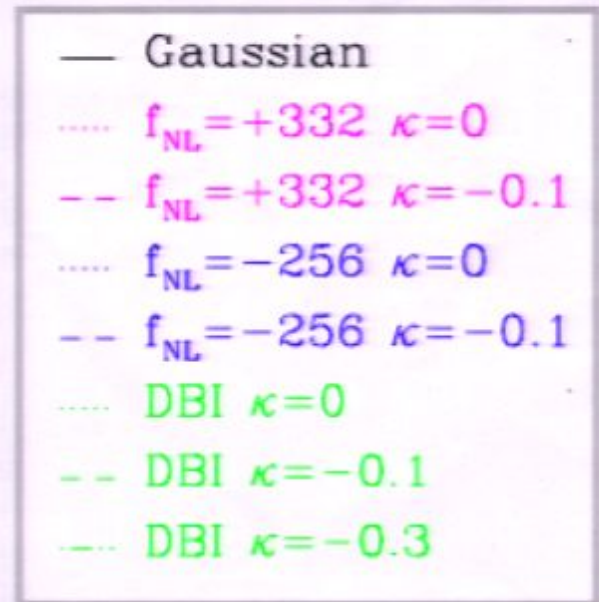
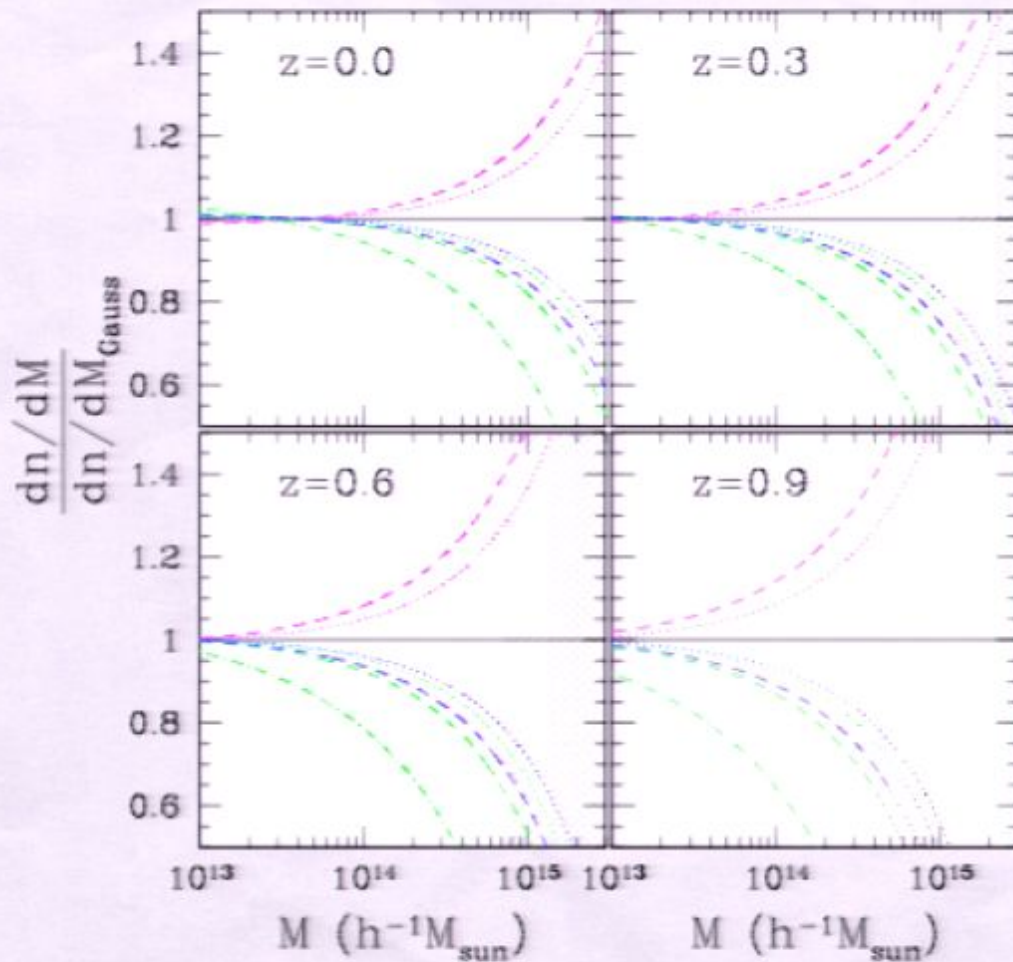
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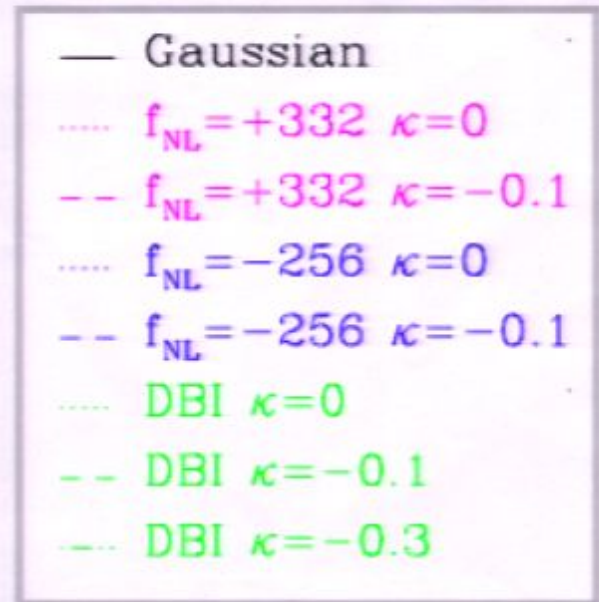
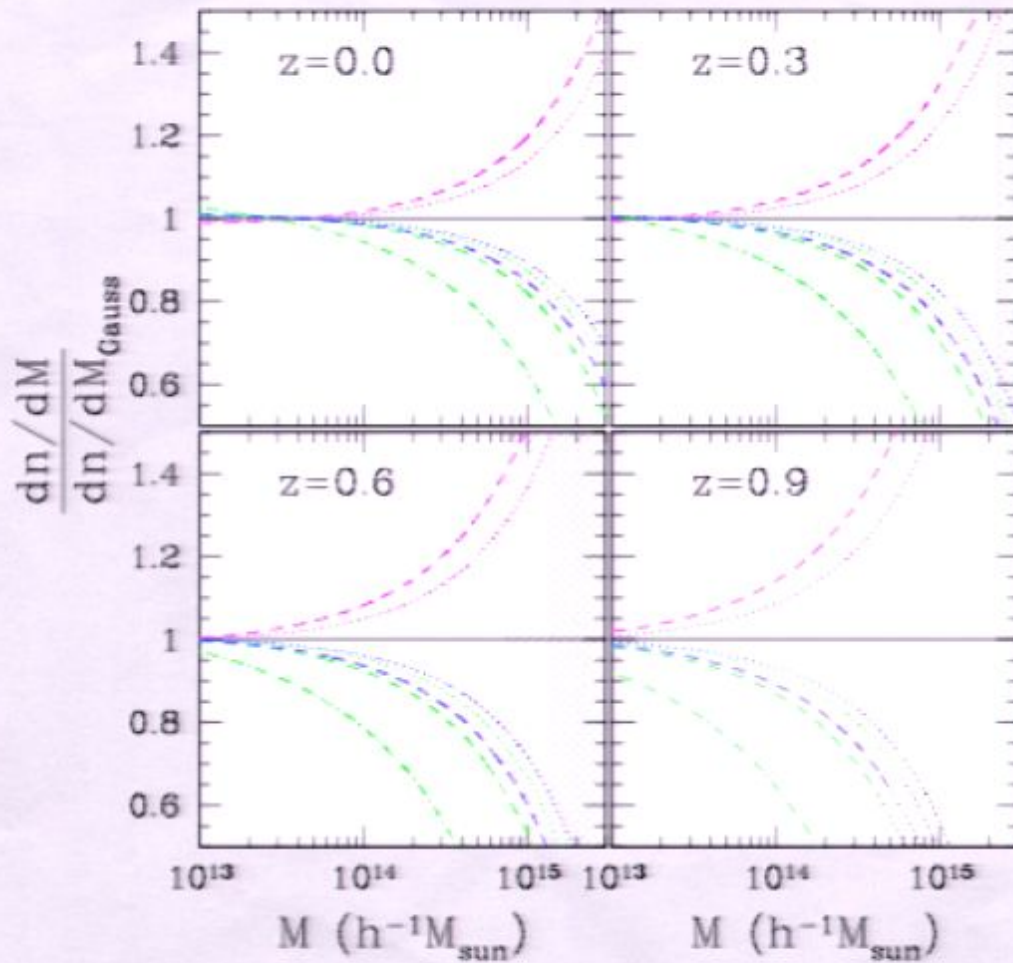
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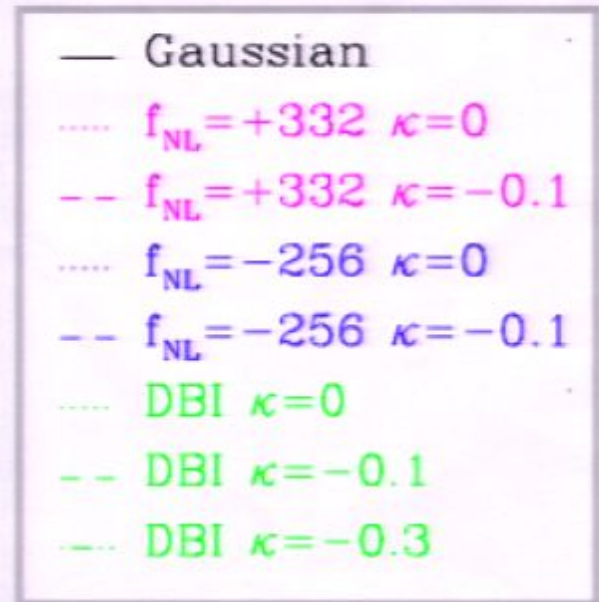
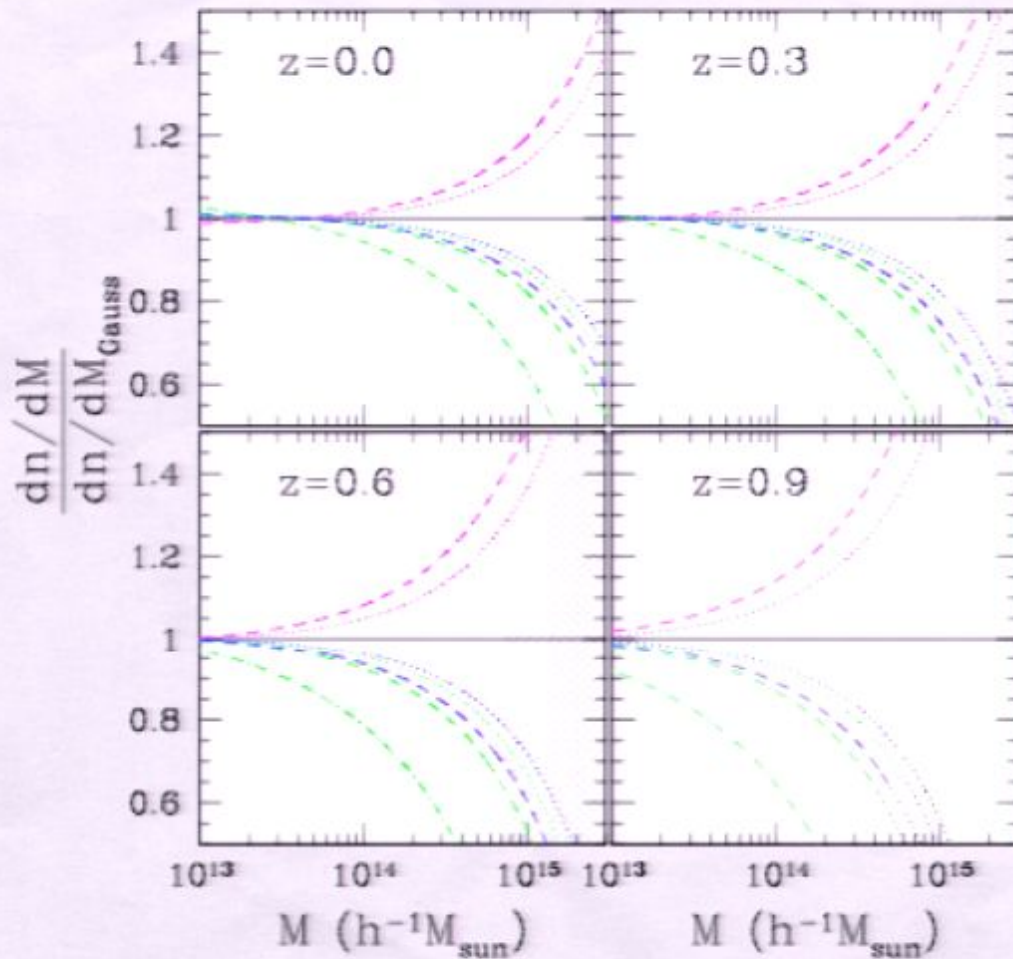
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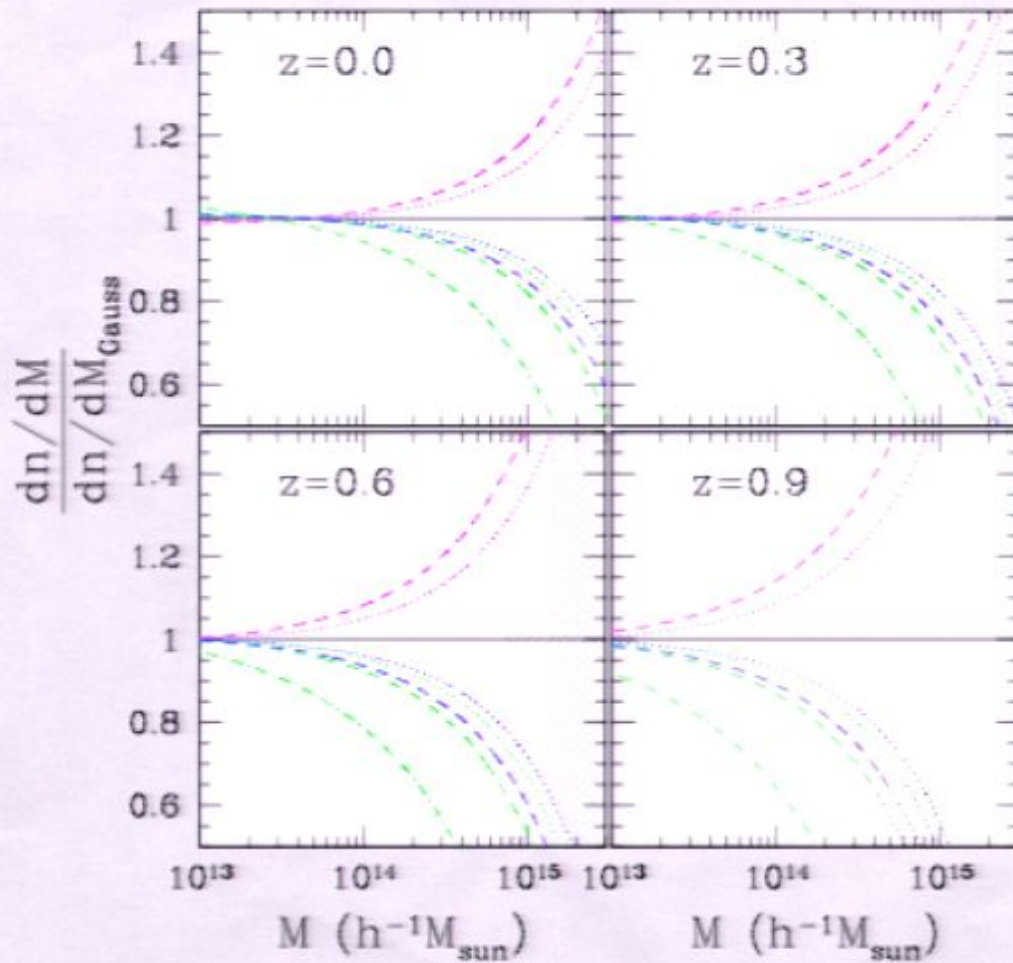
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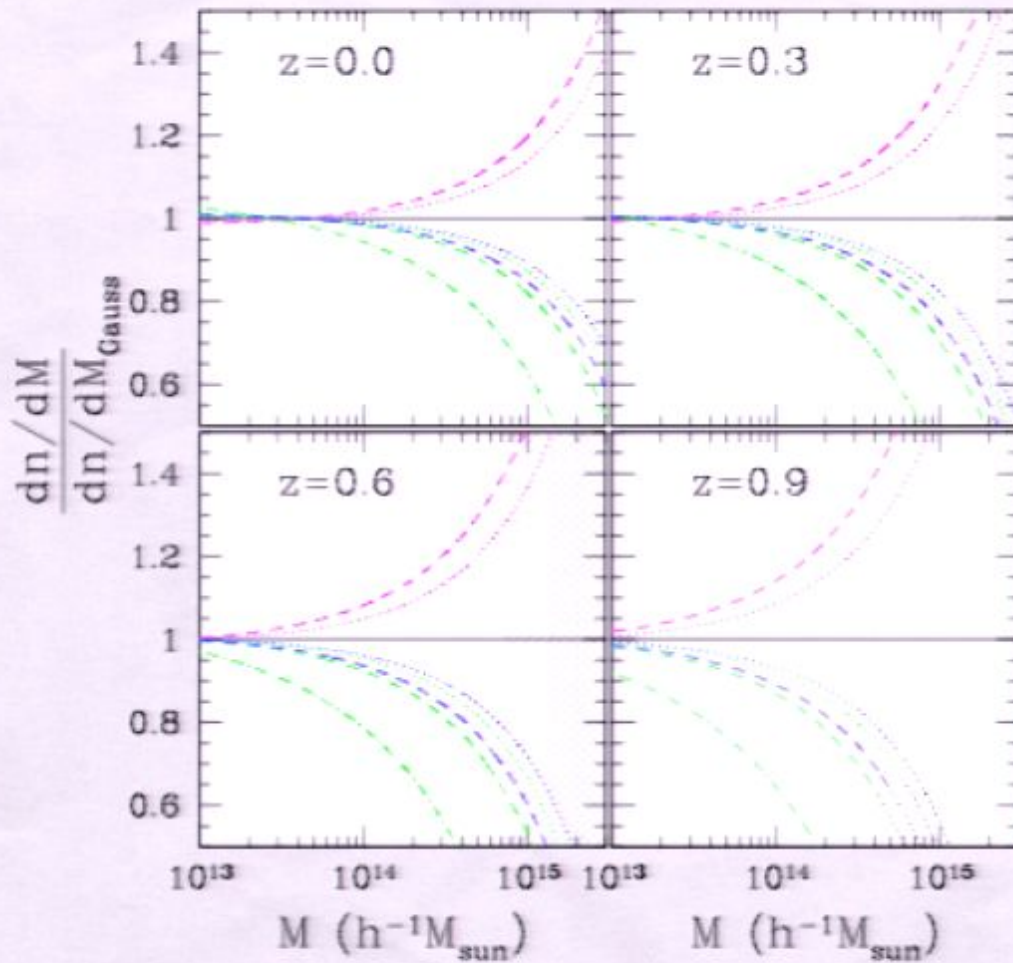


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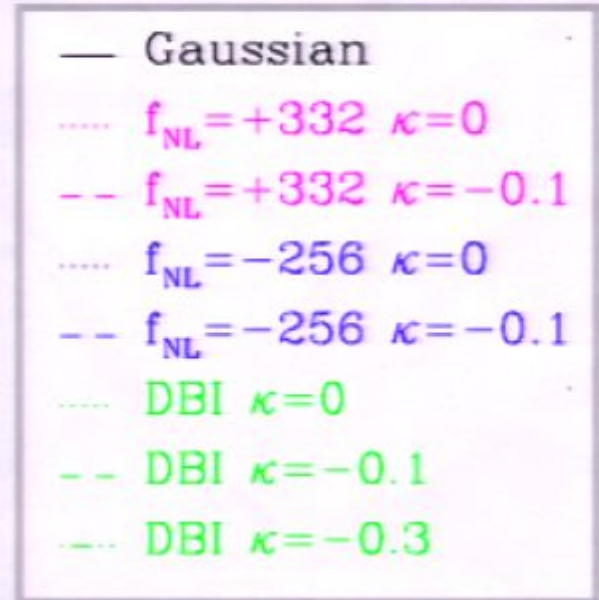
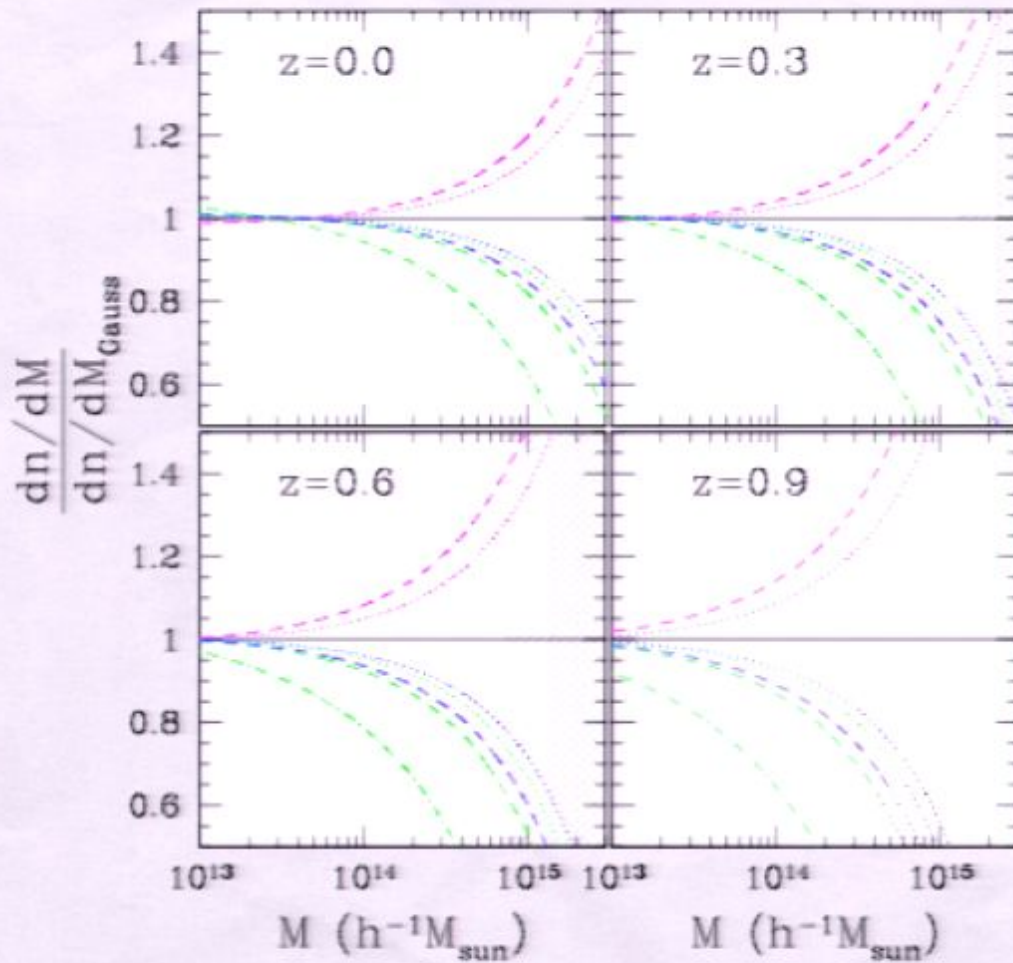
- Gaussian
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- $f_{\text{NL}} = -256 \quad \kappa = 0$
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- DBI $\kappa = -0.1$
- DBI $\kappa = -0.3$

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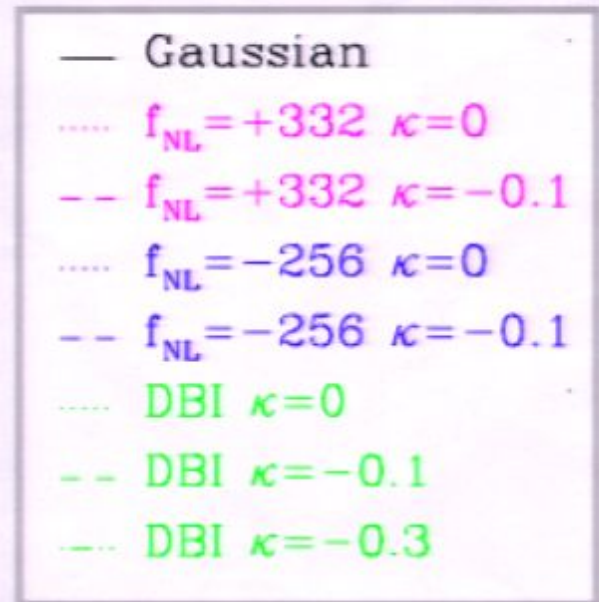
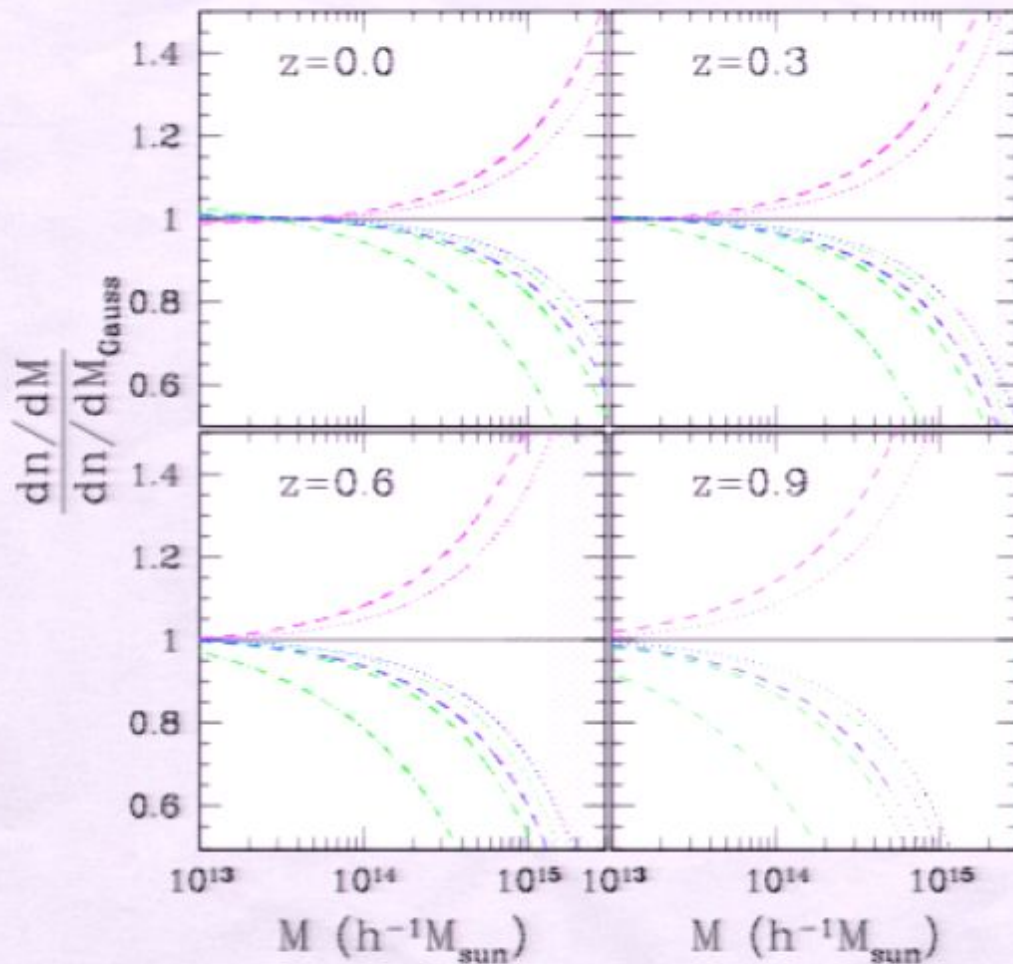


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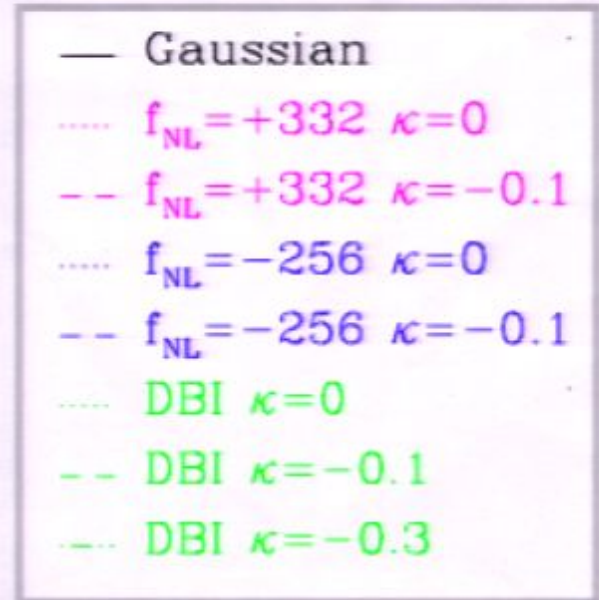
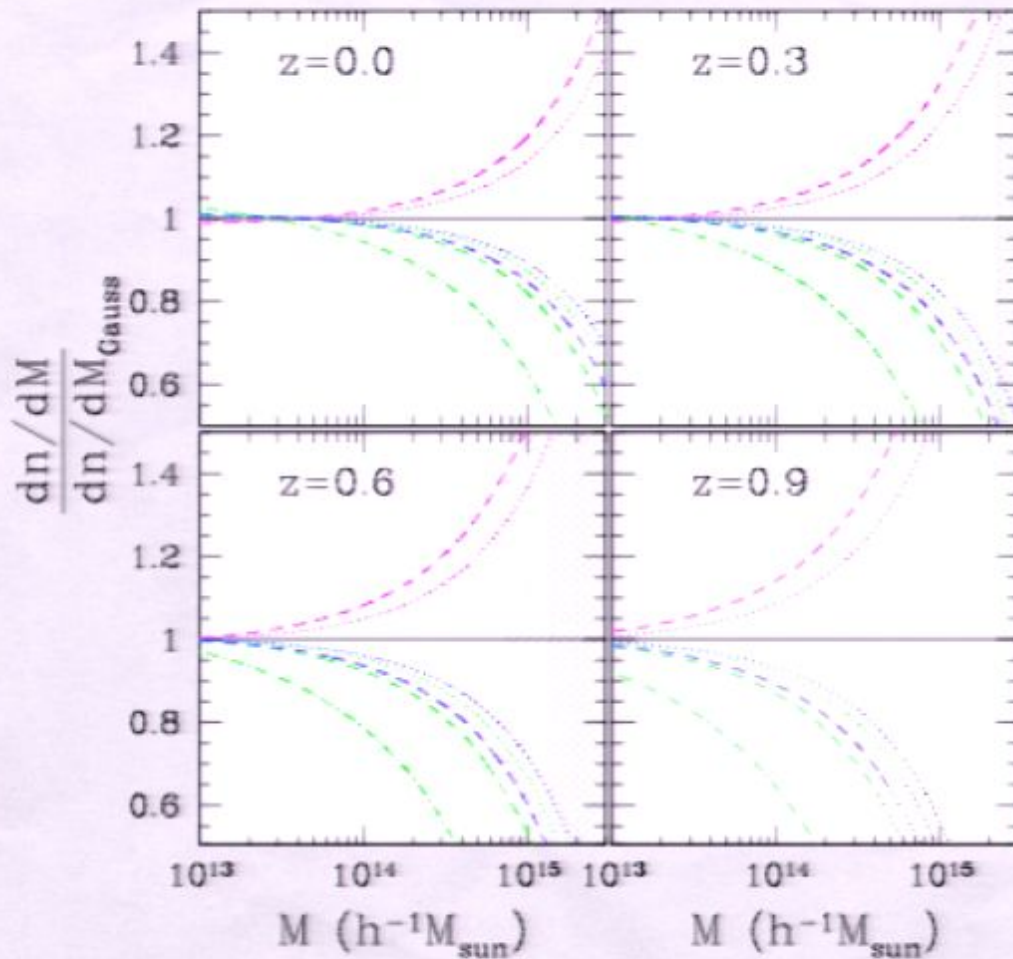
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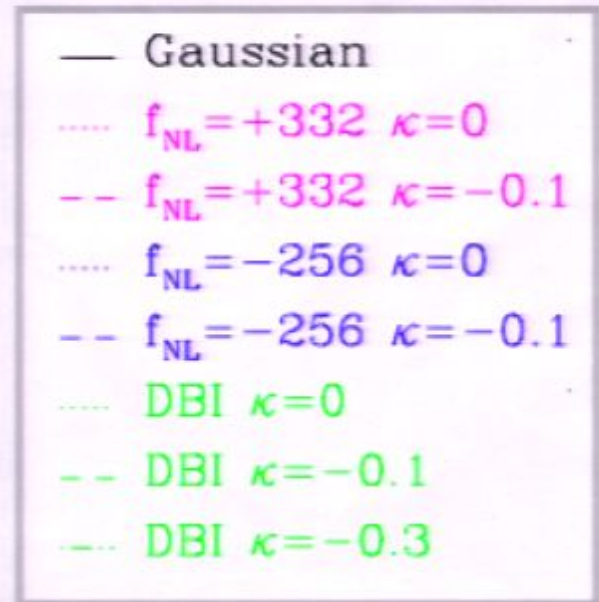
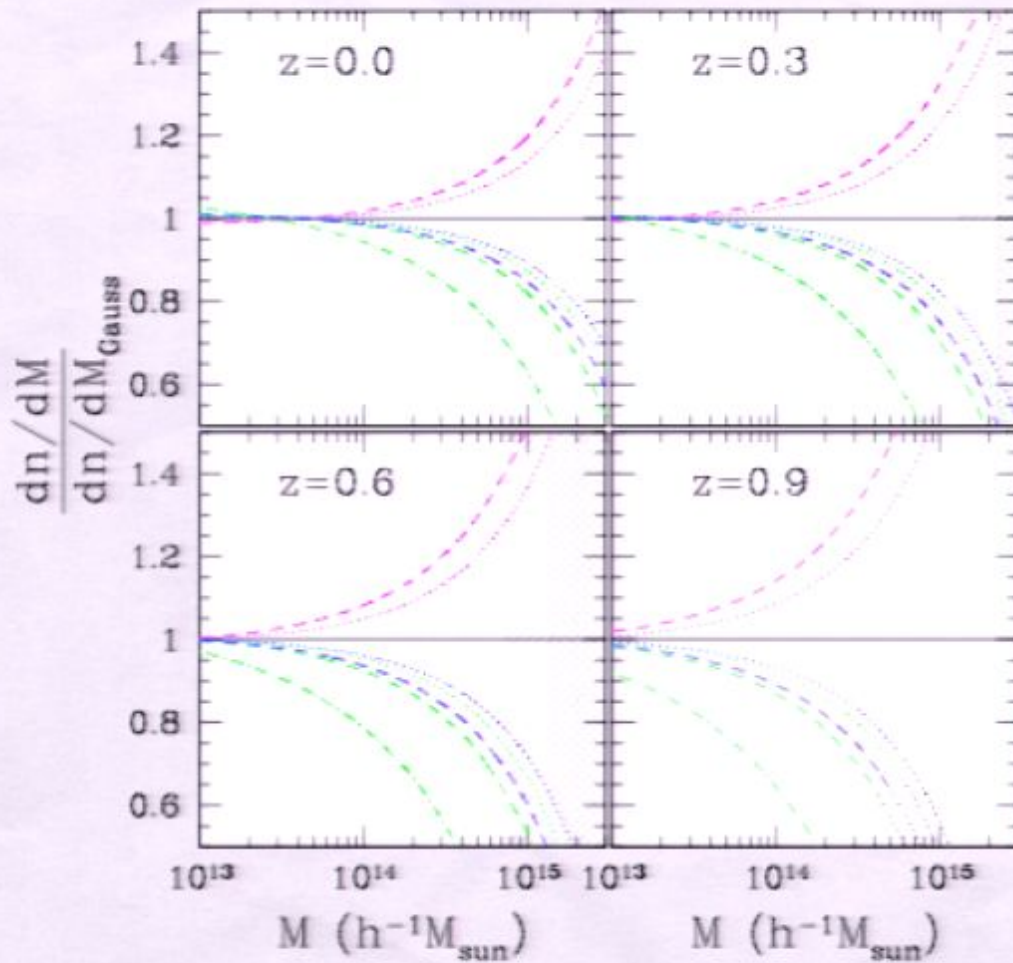
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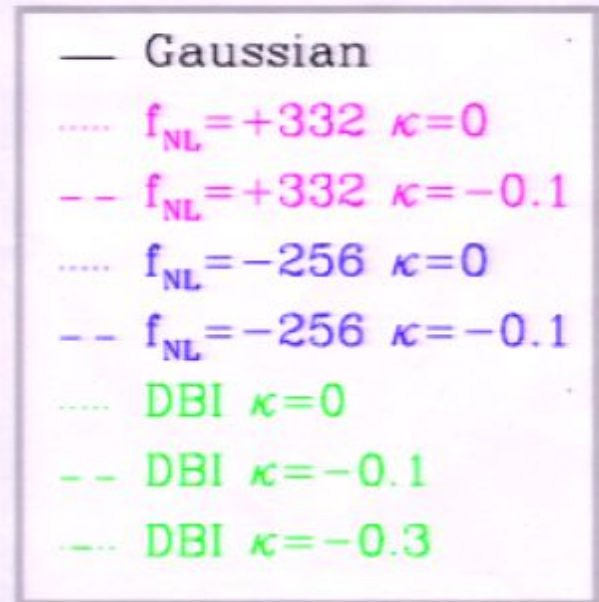
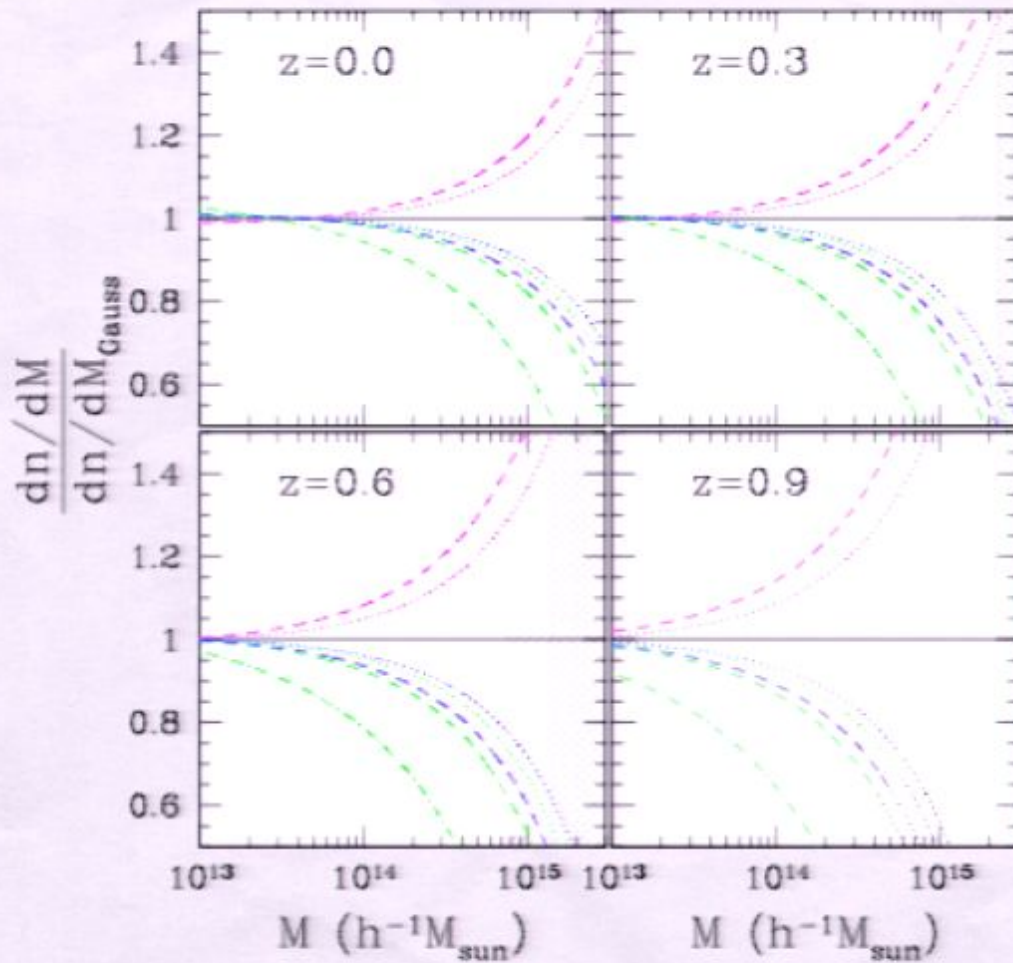
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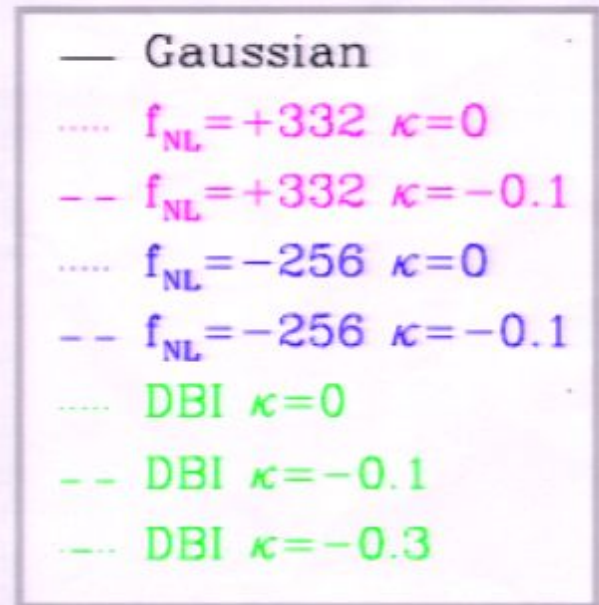
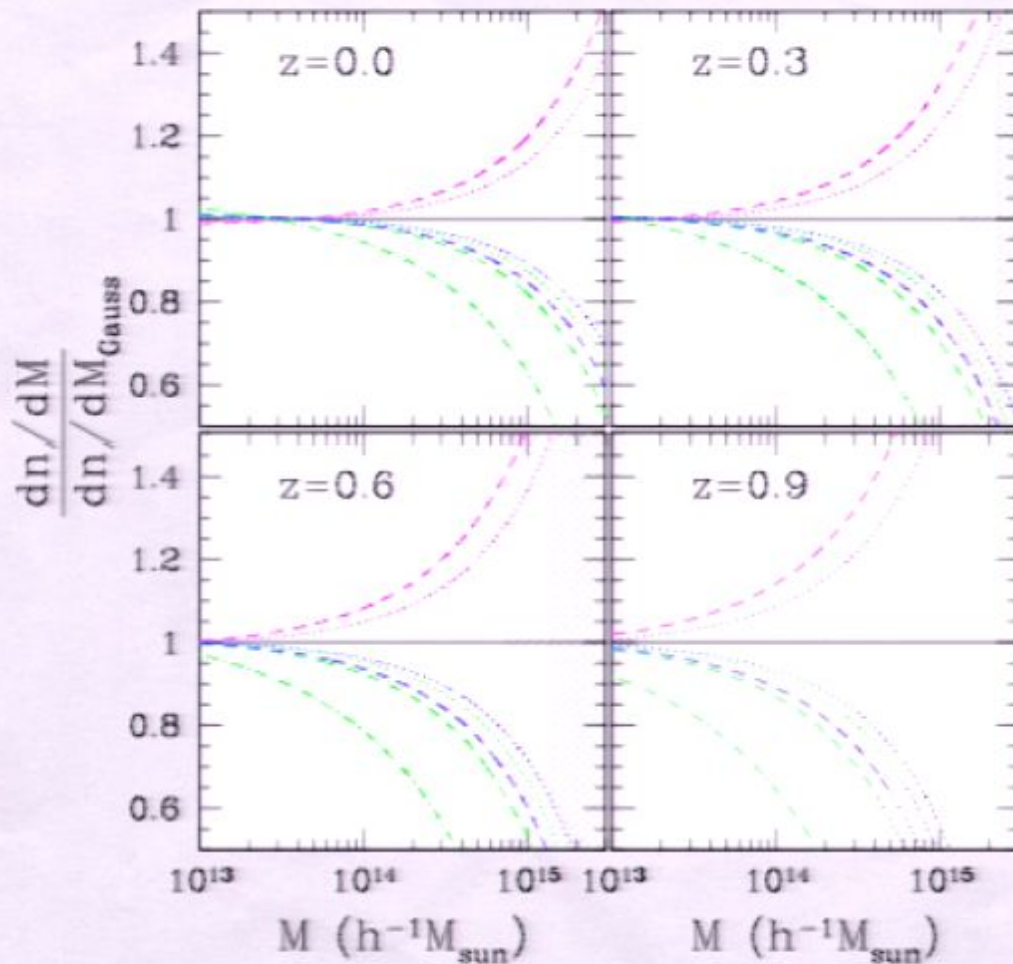
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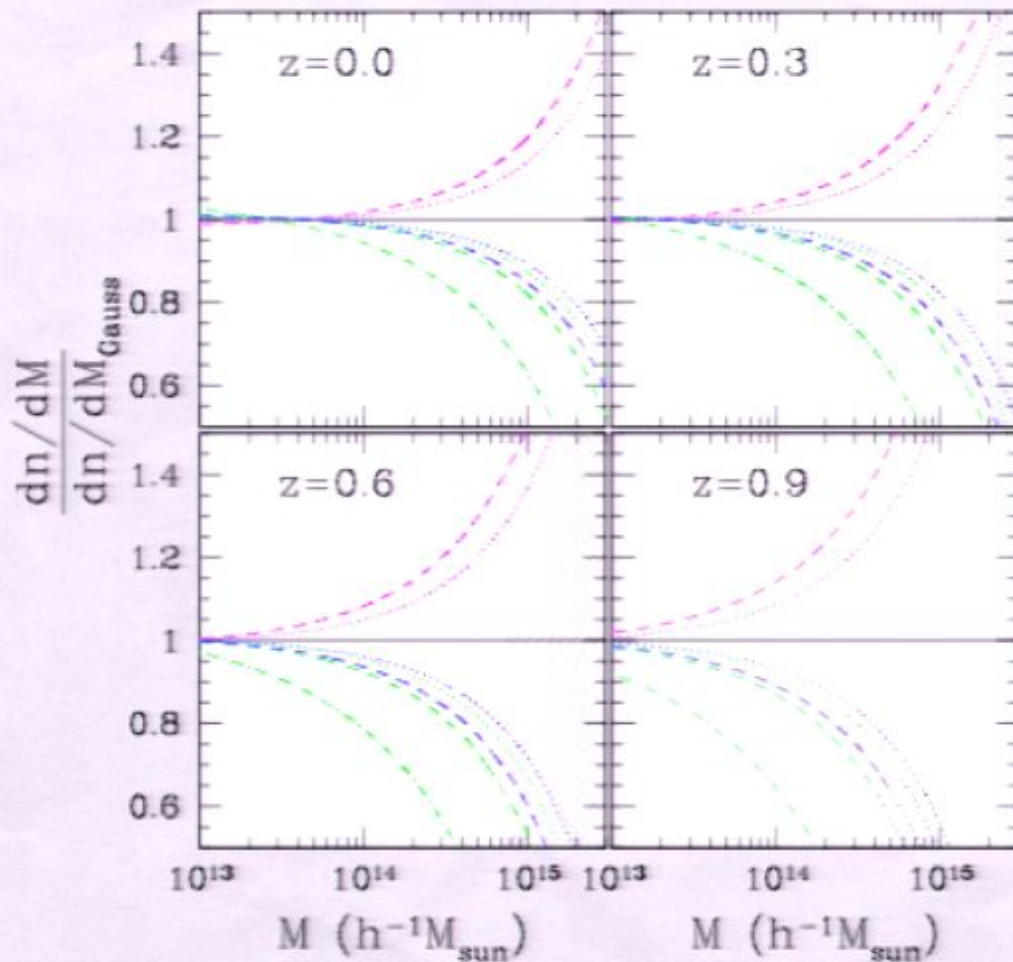
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MASS FUNCTION

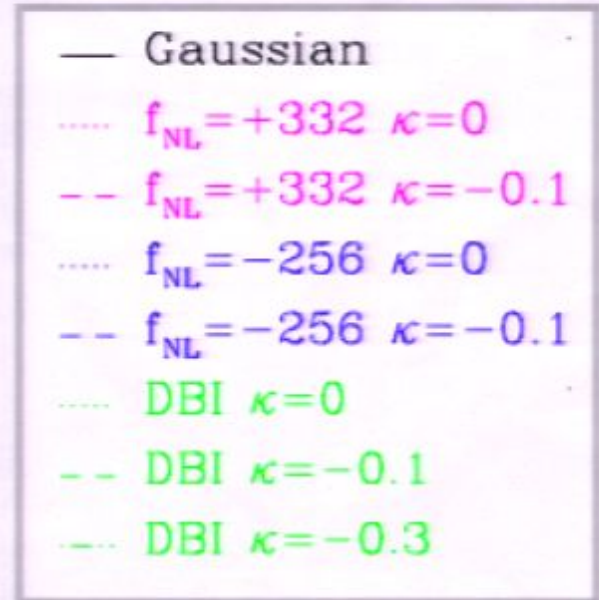
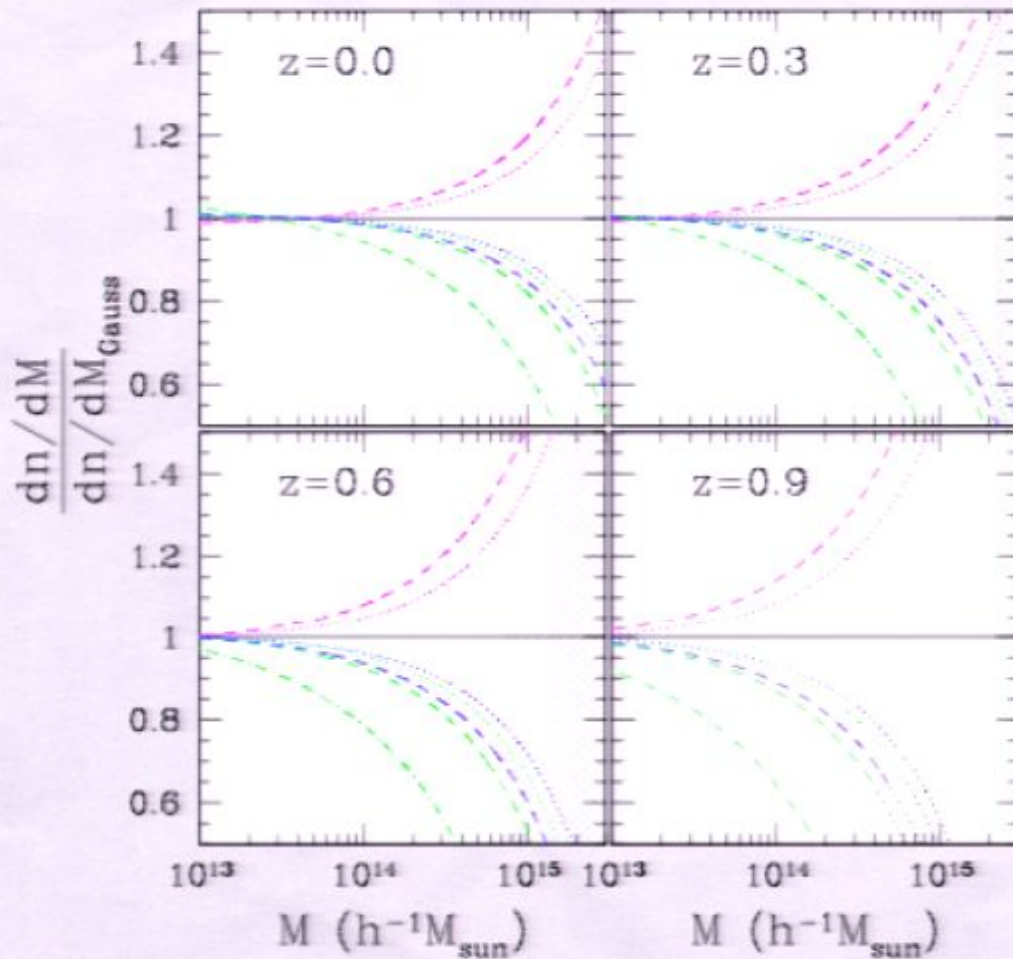


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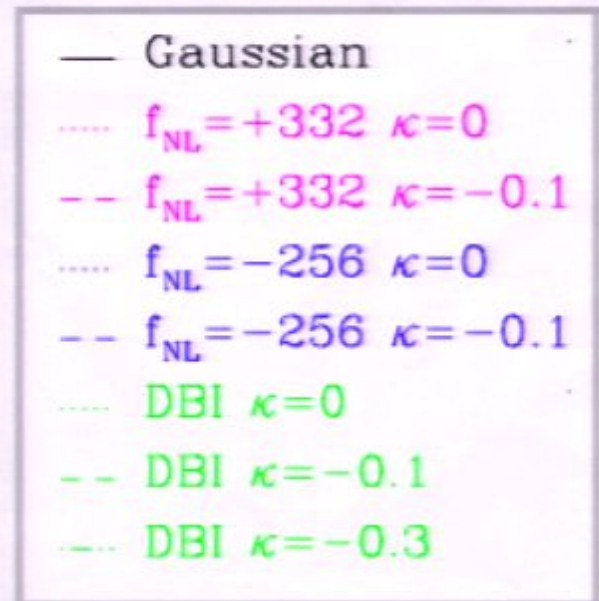
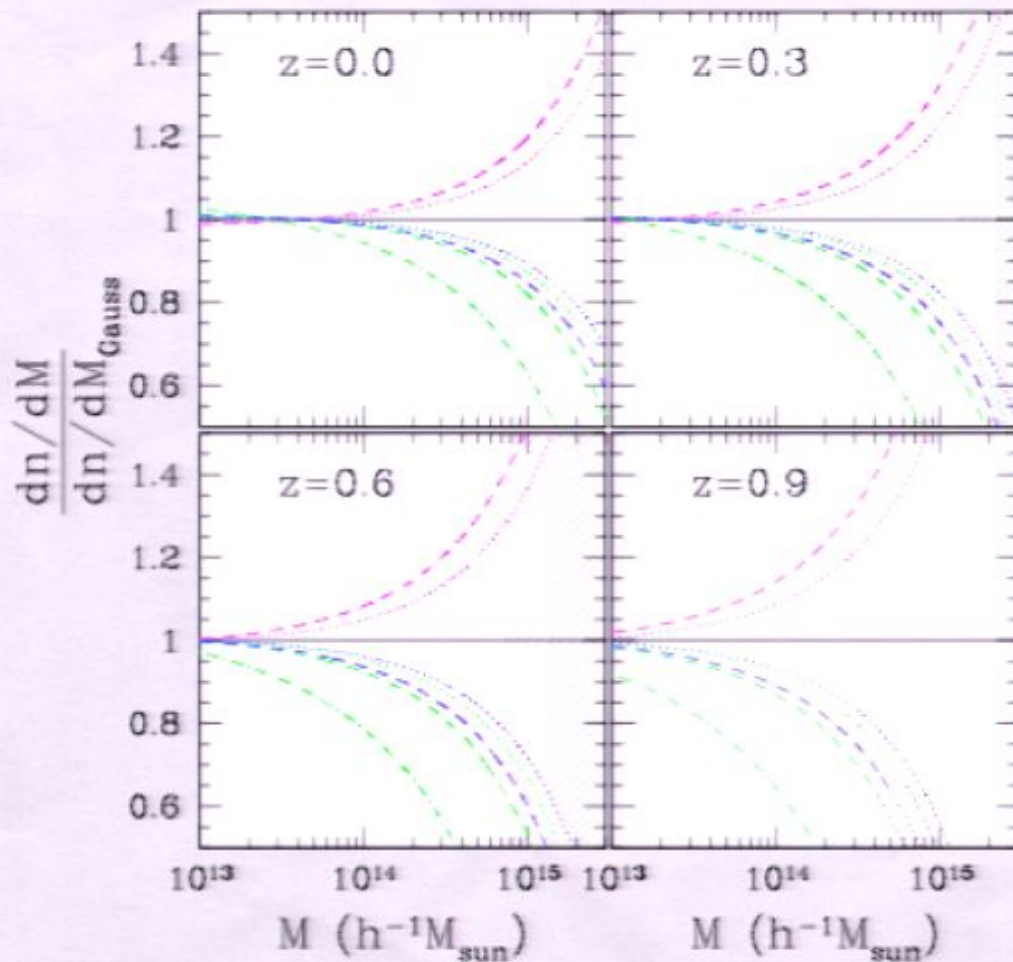


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- $f_{\text{NL}}=+332$ $\kappa=0$
- $f_{\text{NL}}=+332$ $\kappa=-0.1$
- $f_{\text{NL}}=-256$ $\kappa=0$
- $f_{\text{NL}}=-256$ $\kappa=-0.1$
- DBI $\kappa=0$
- DBI $\kappa=-0.1$
- .-.- DBI $\kappa=-0.3$

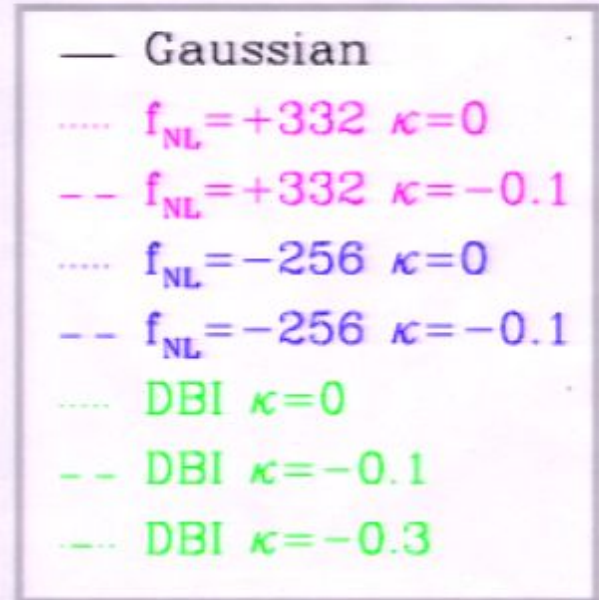
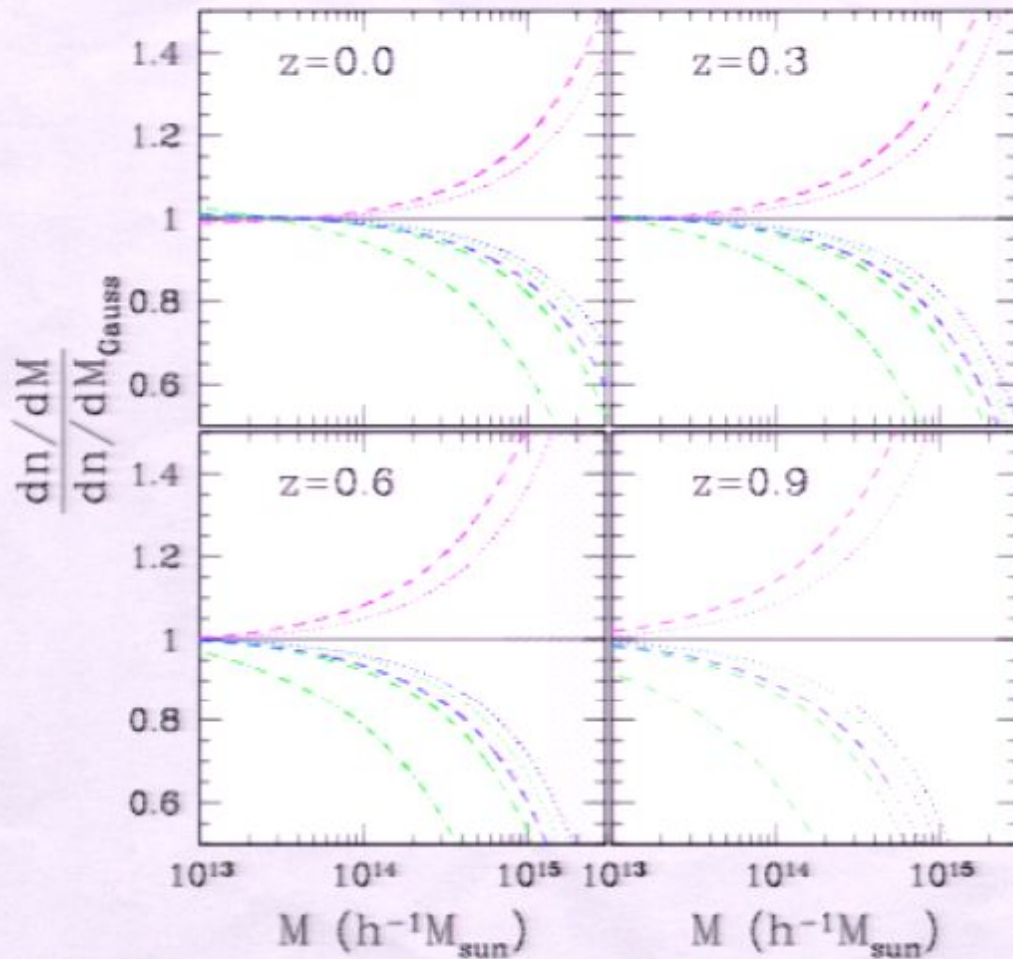
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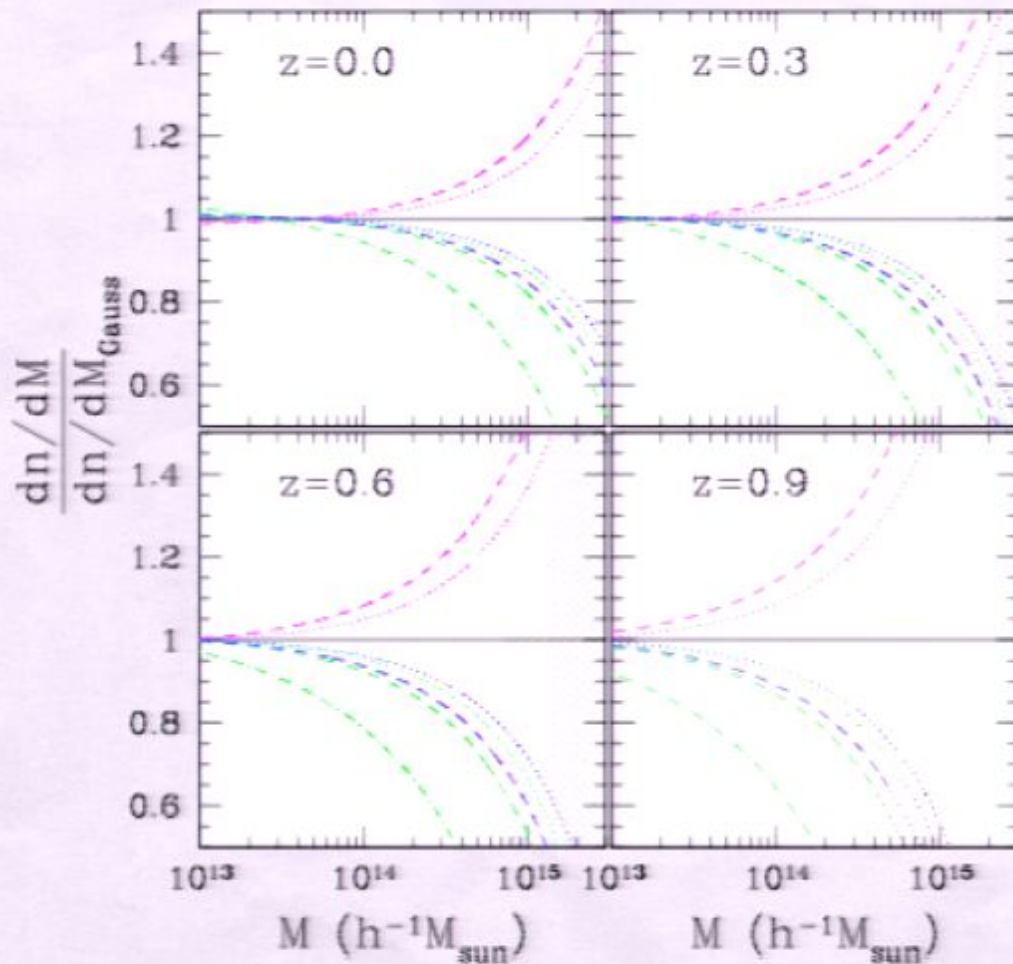
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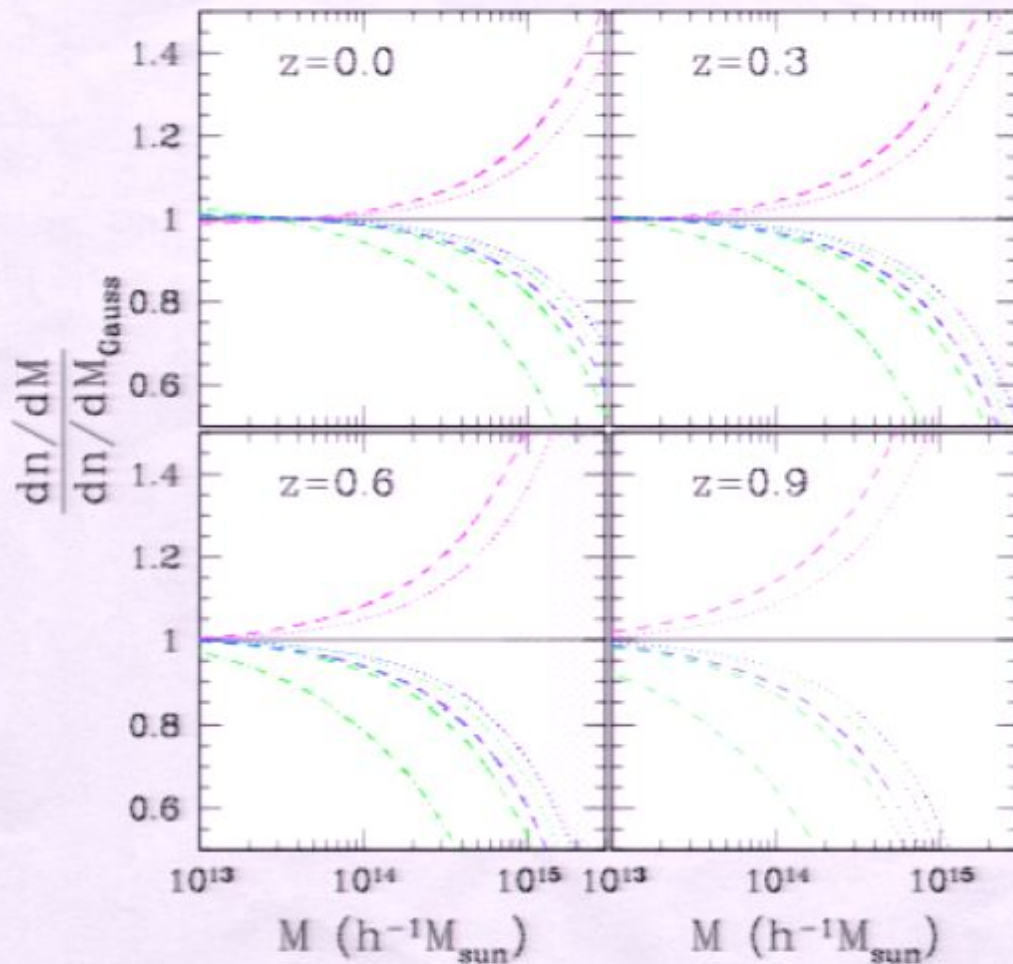


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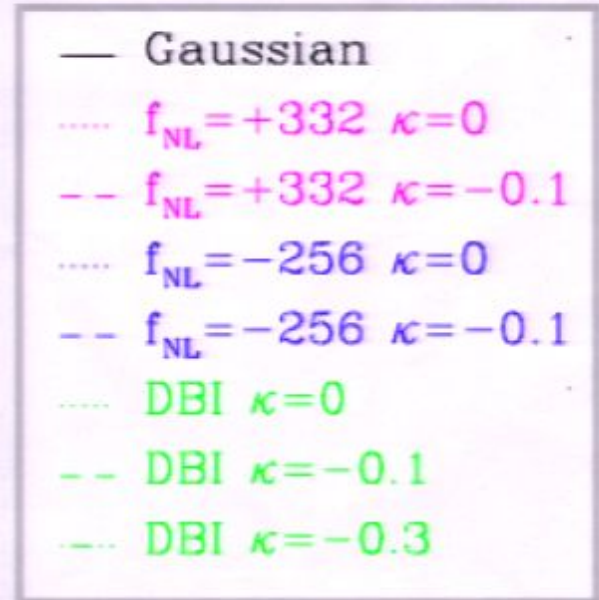
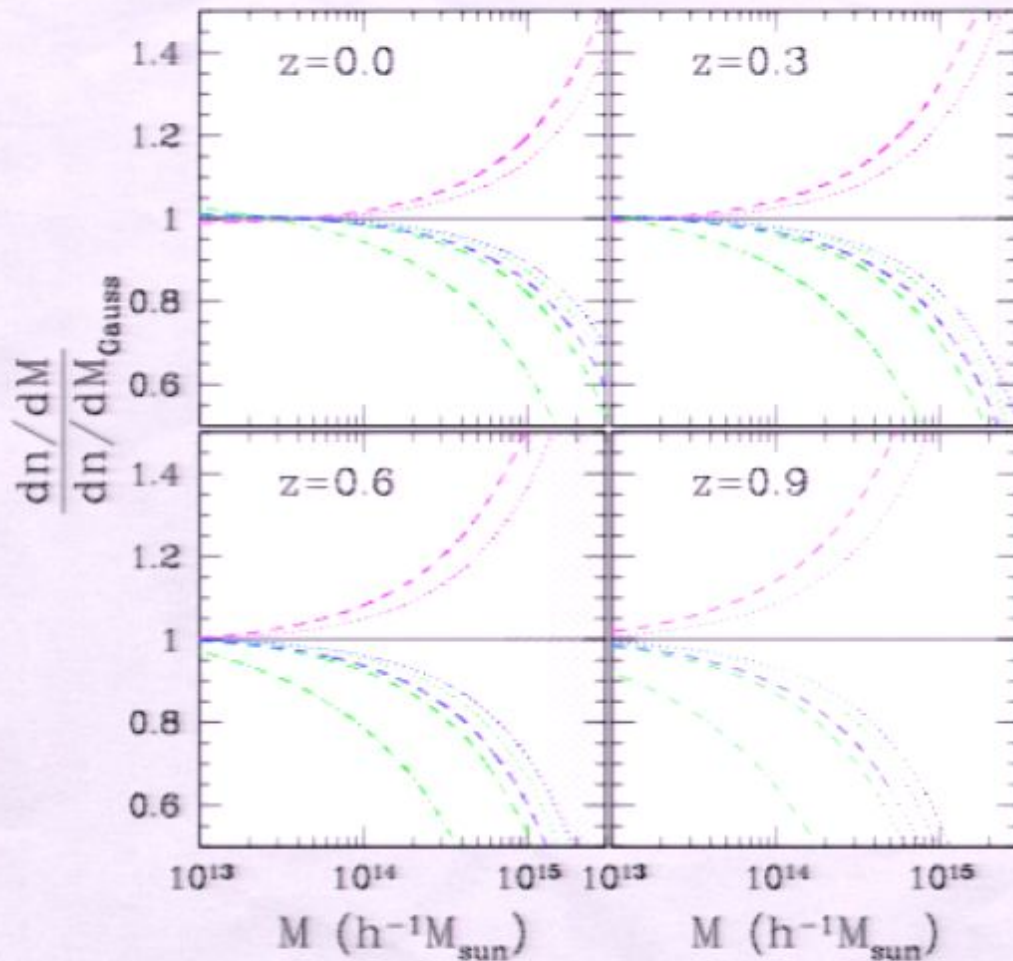


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- DBI $\kappa = 0$
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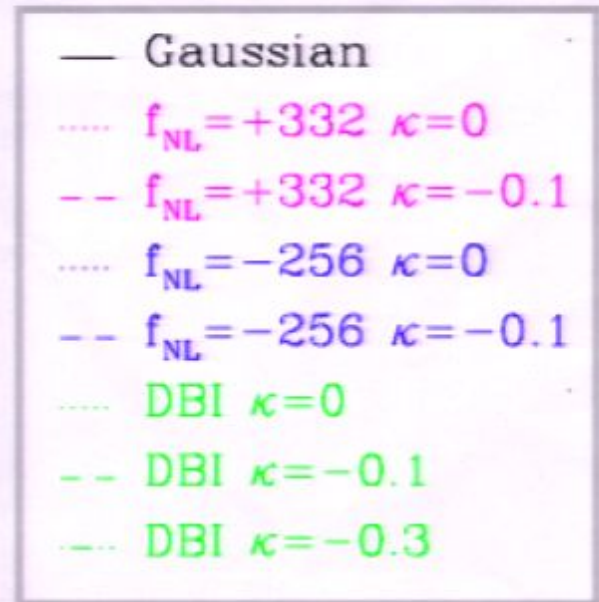
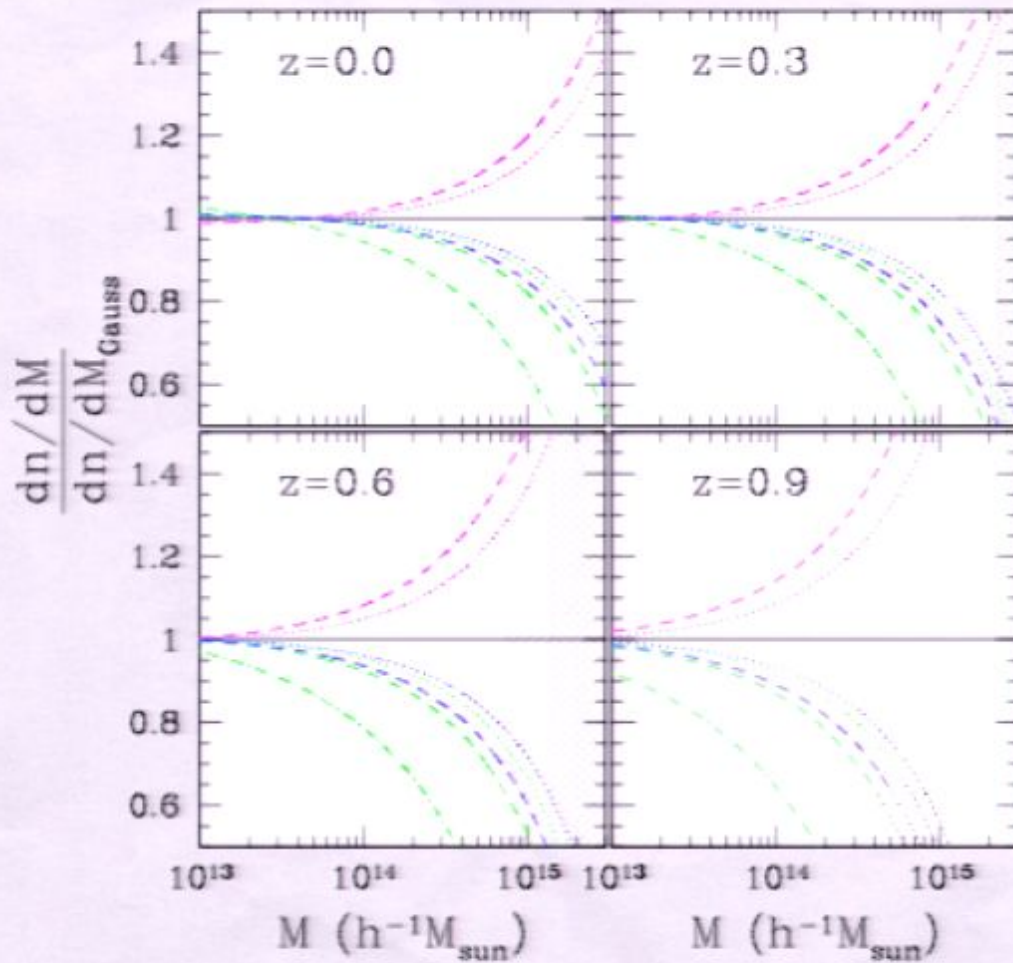
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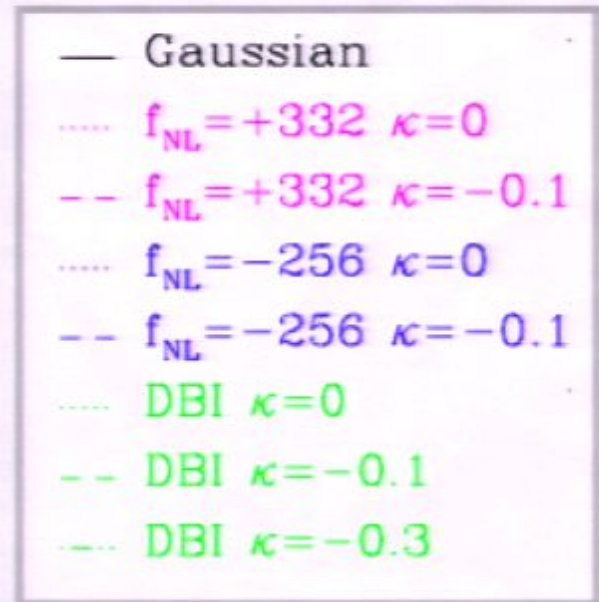
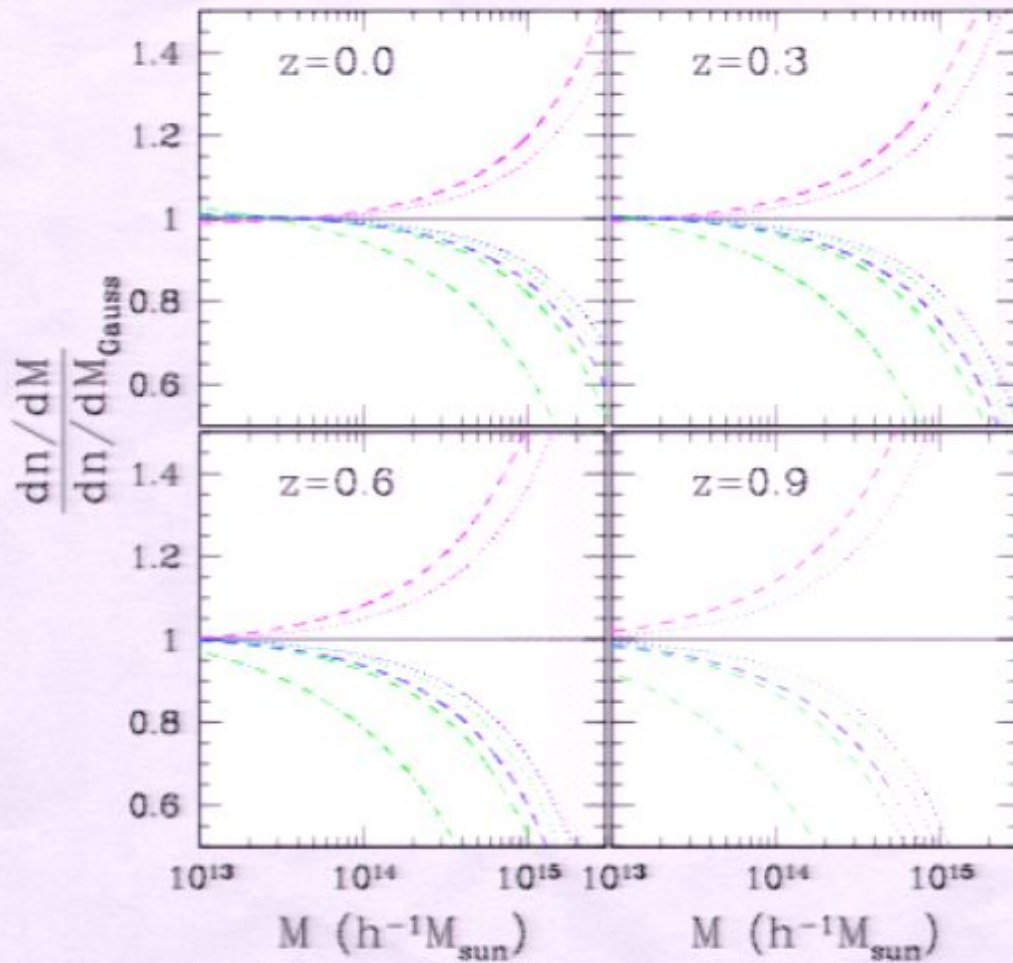
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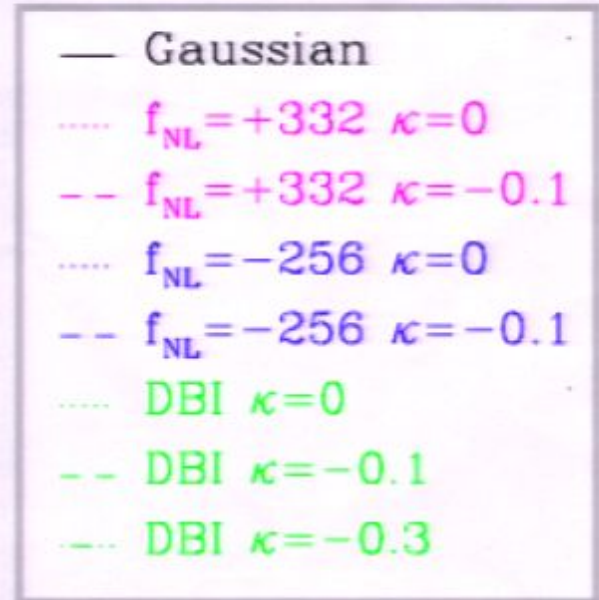
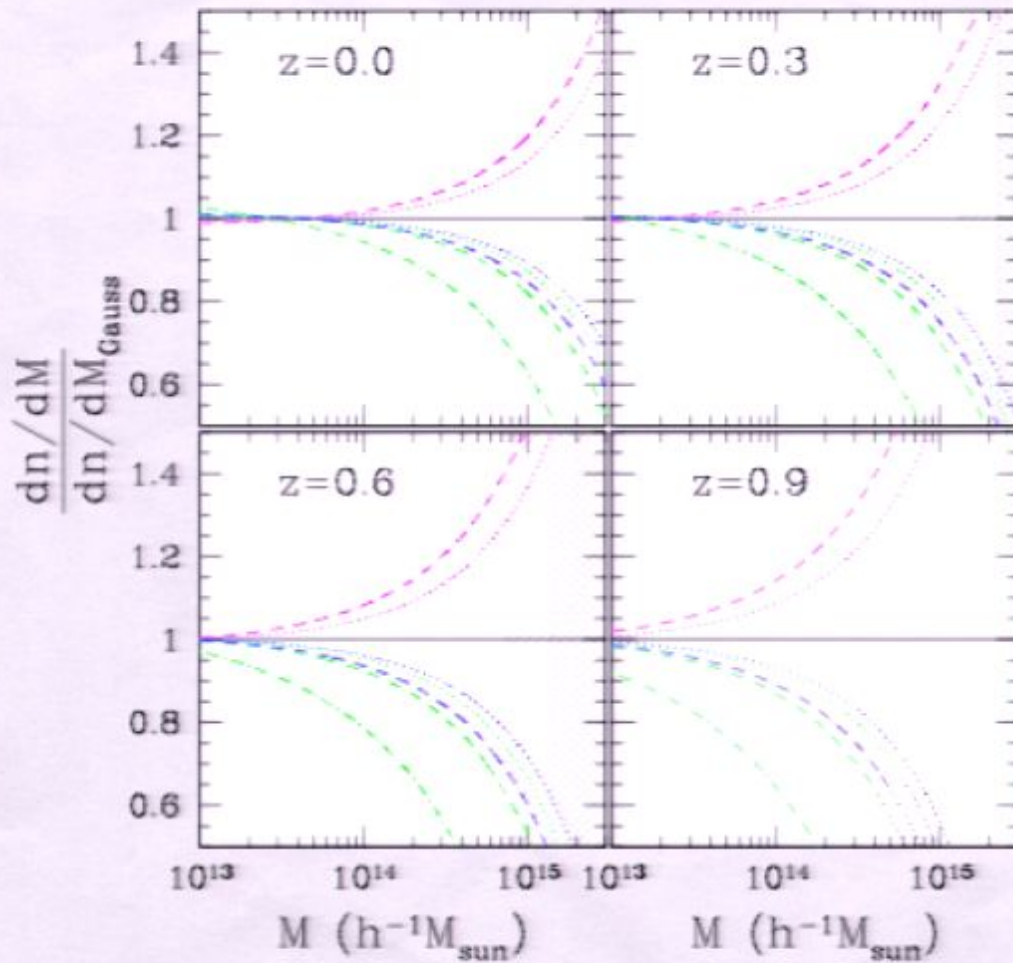
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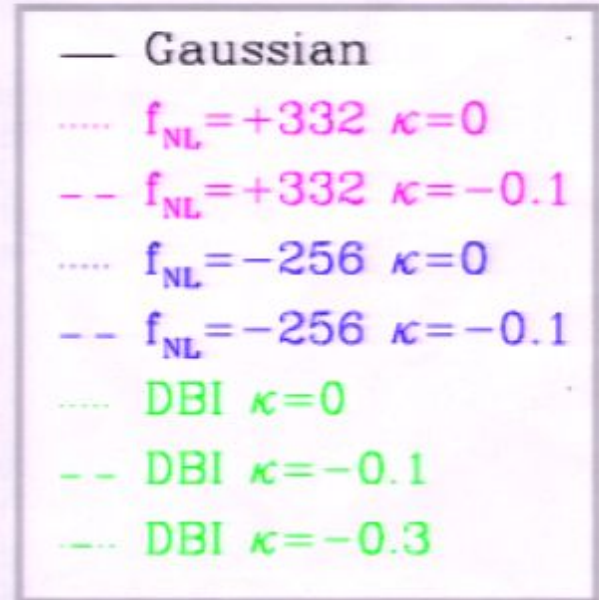
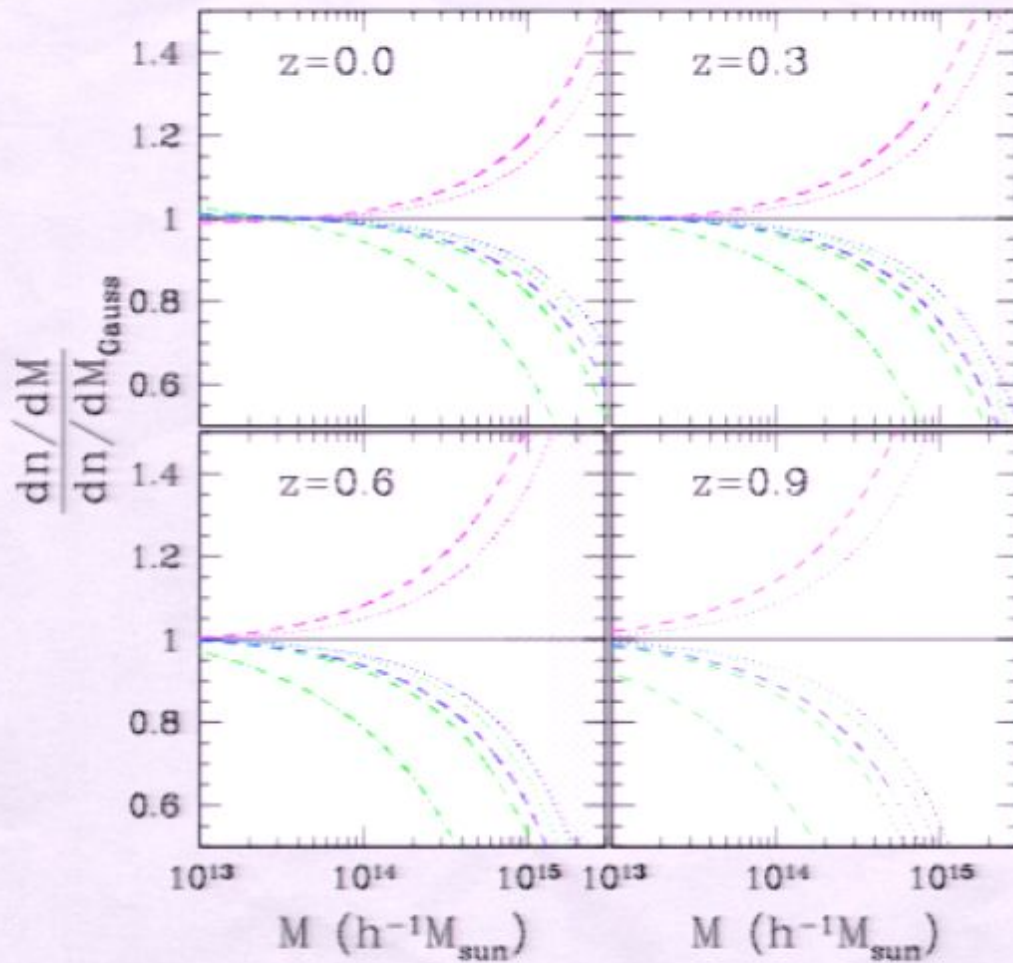
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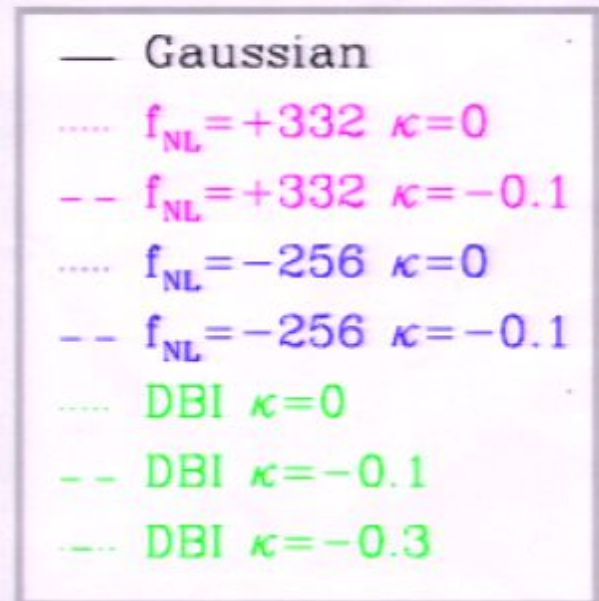
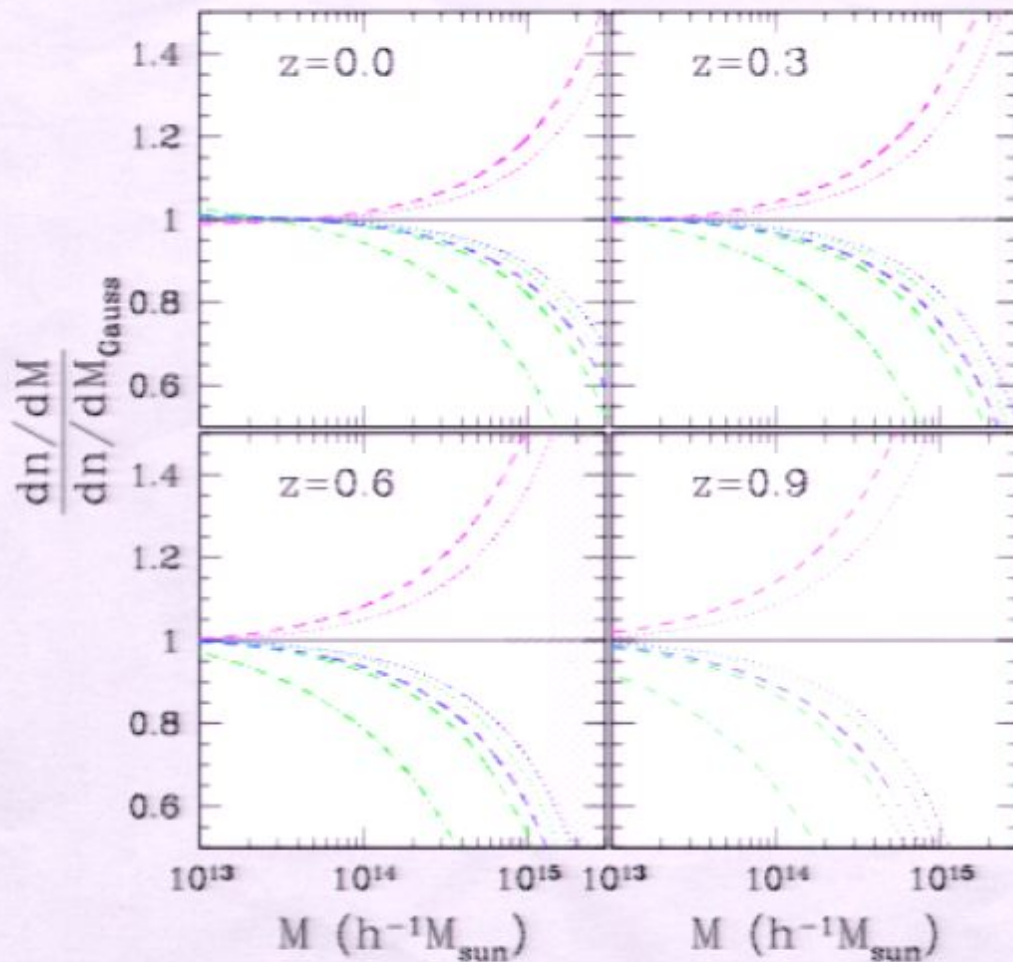
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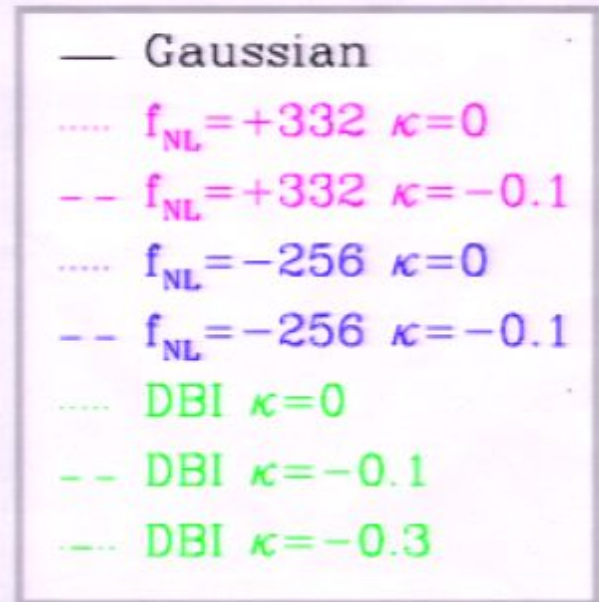
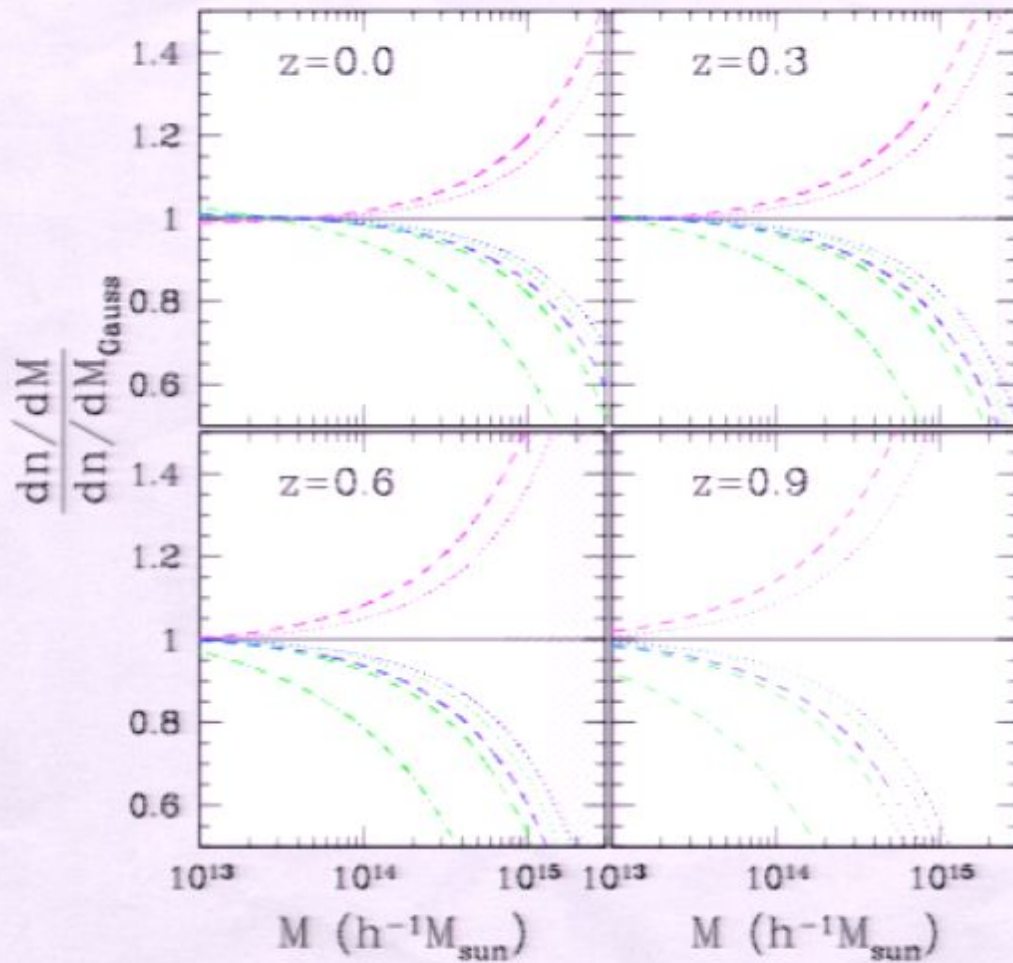
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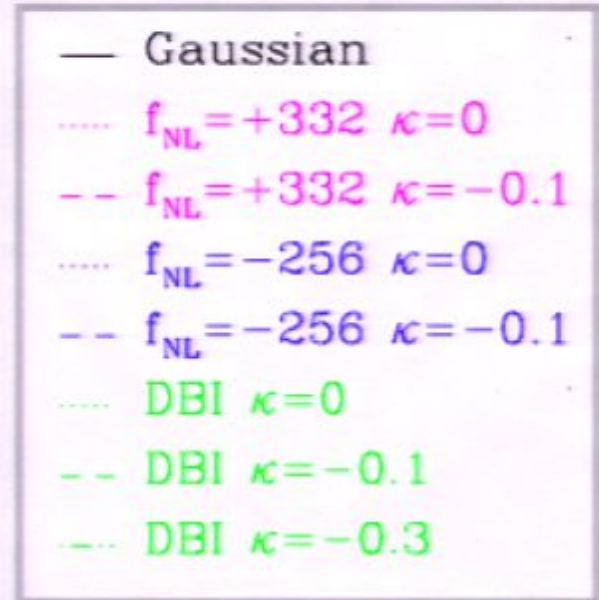
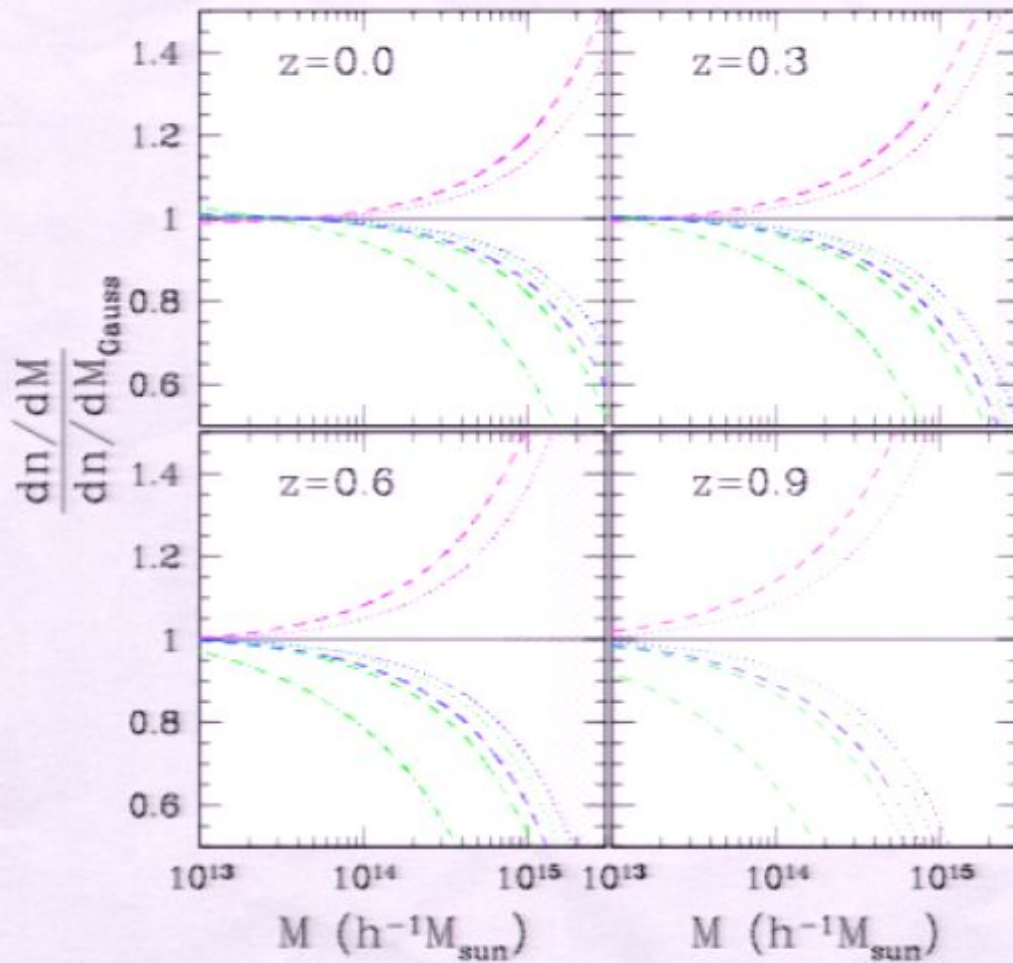
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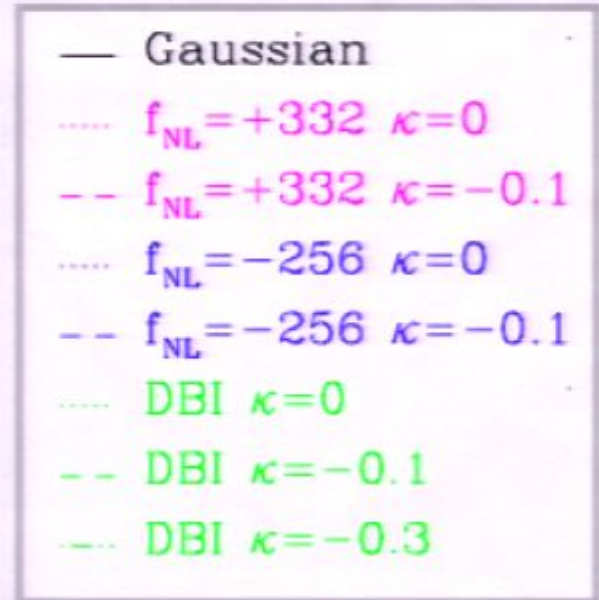
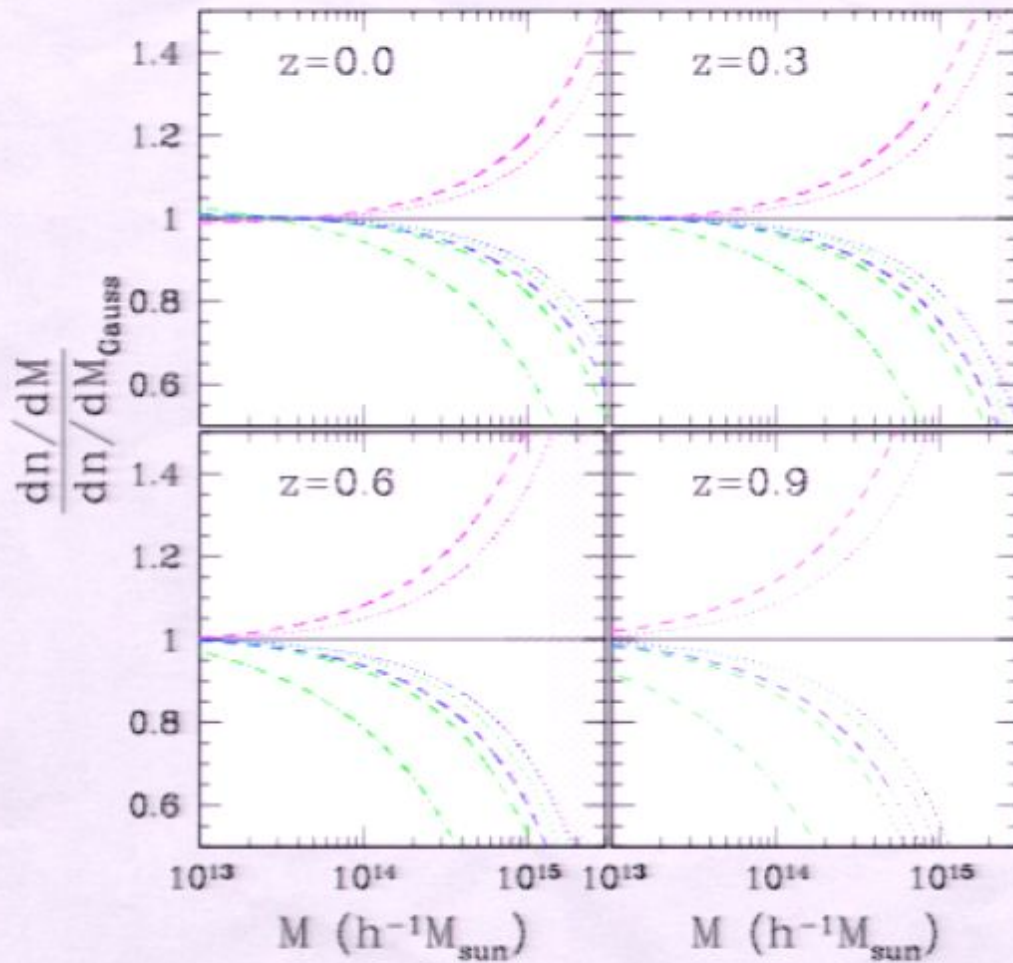
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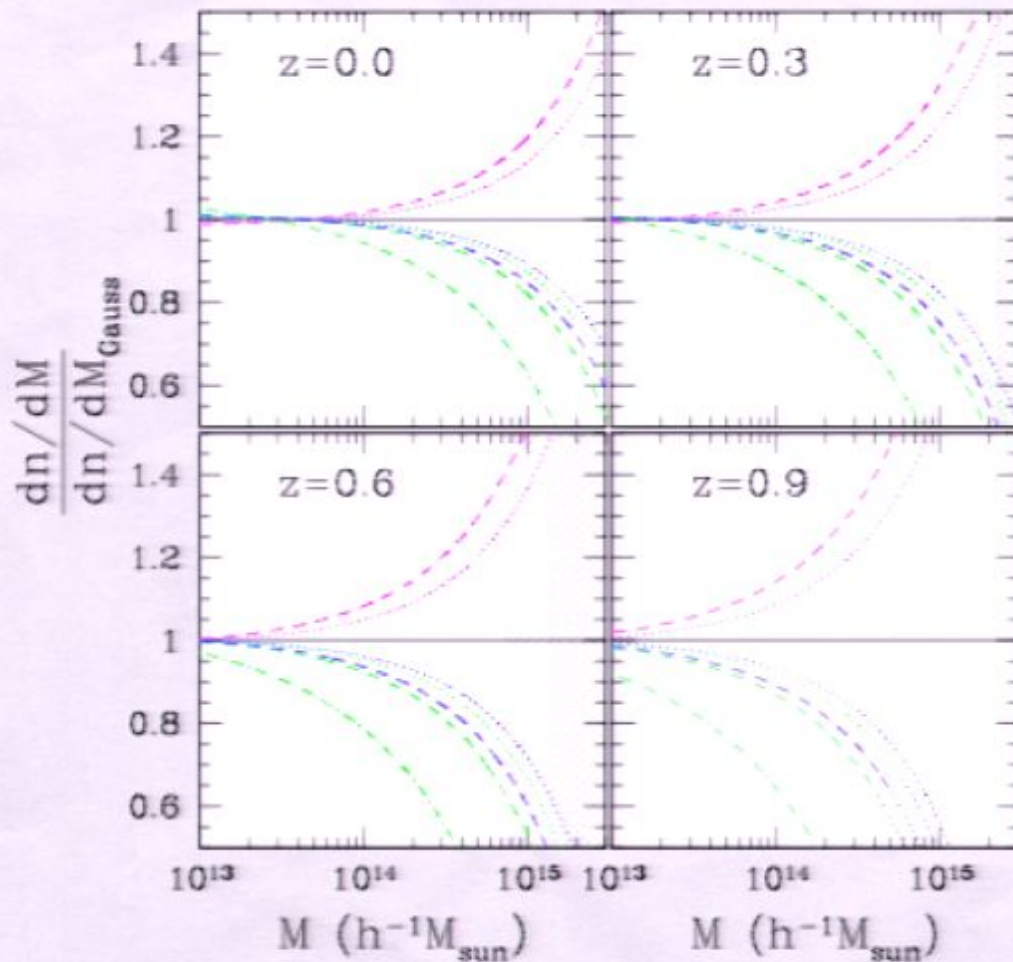
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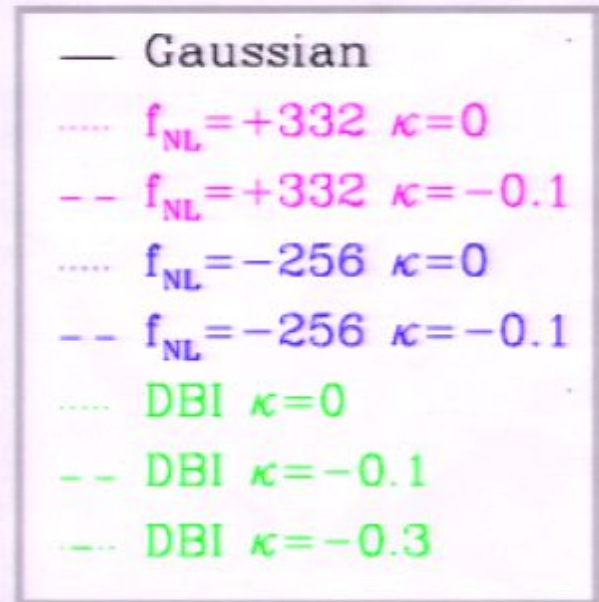
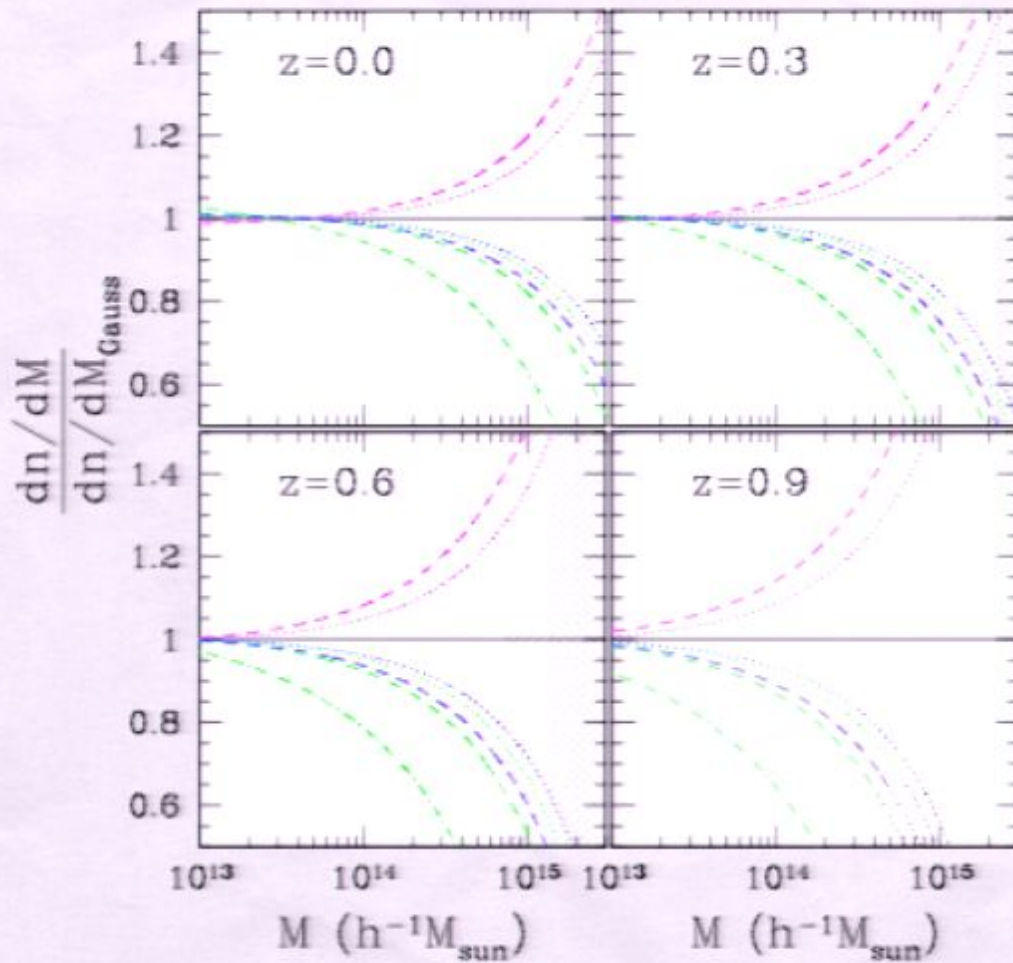


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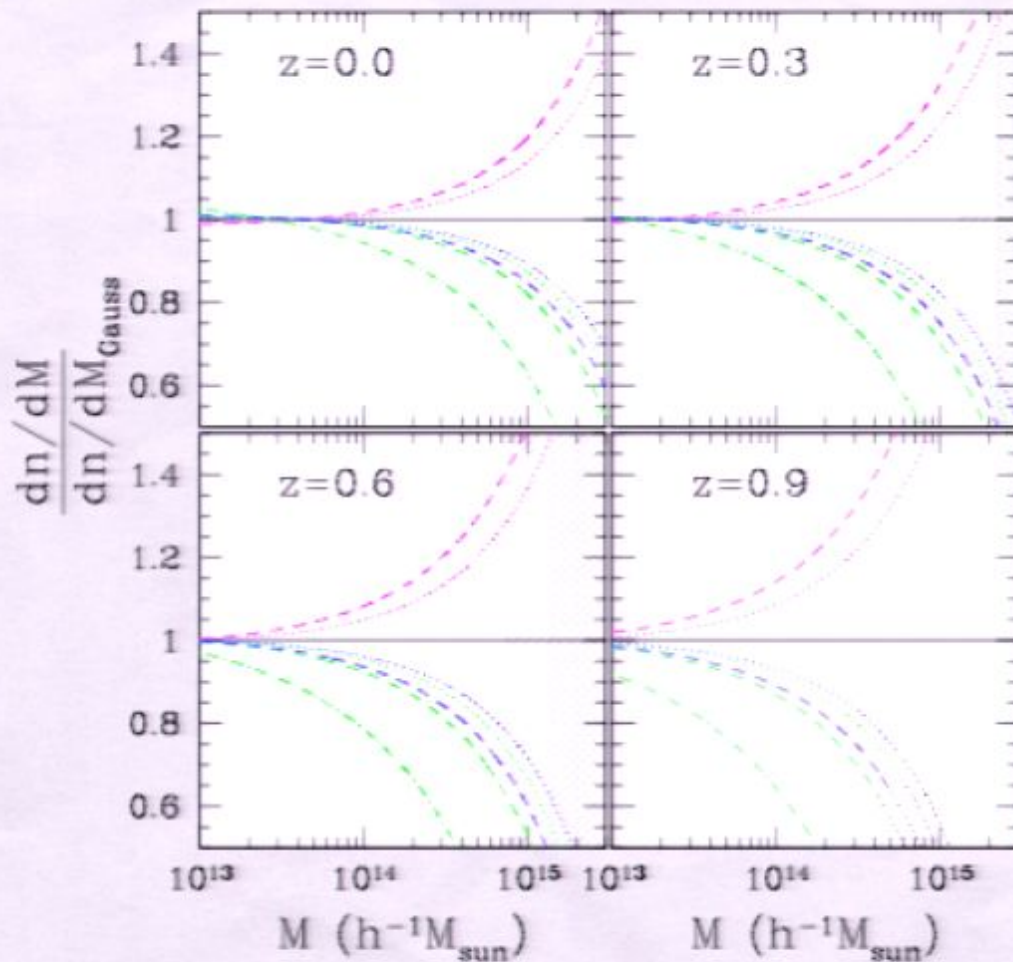


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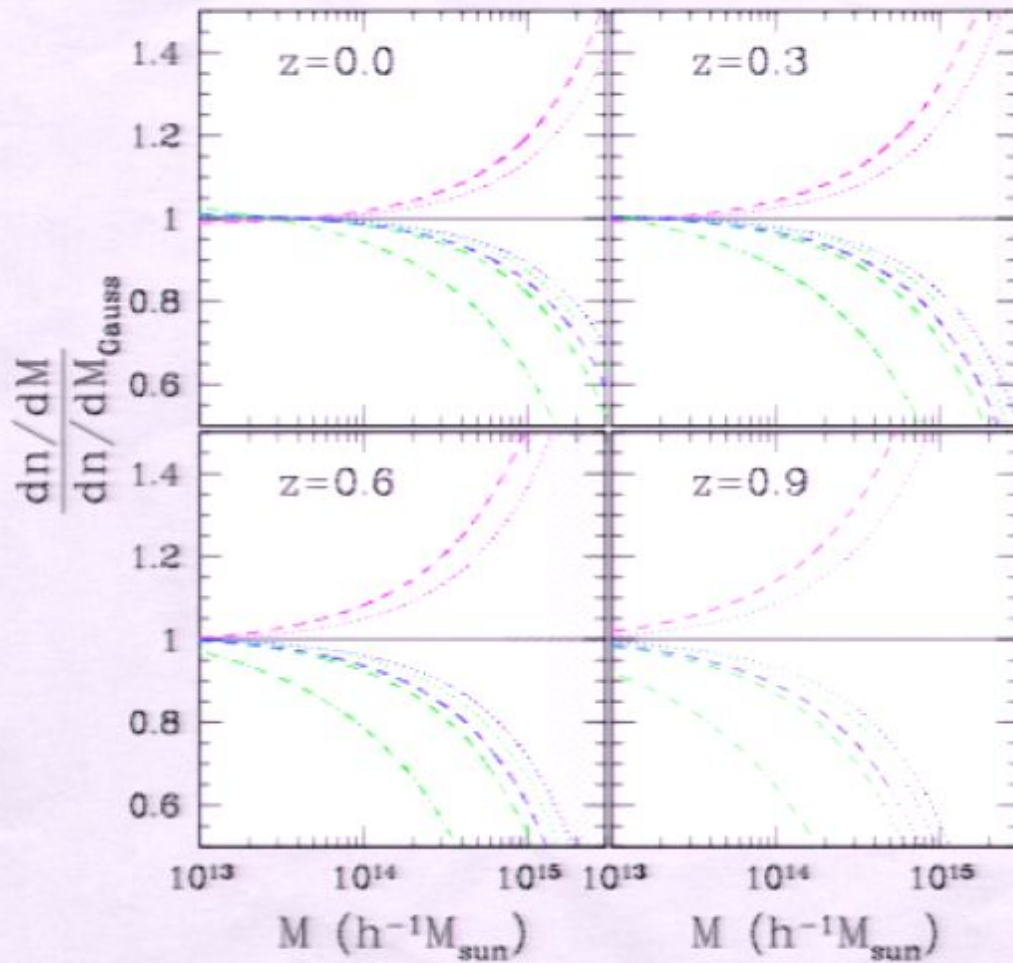
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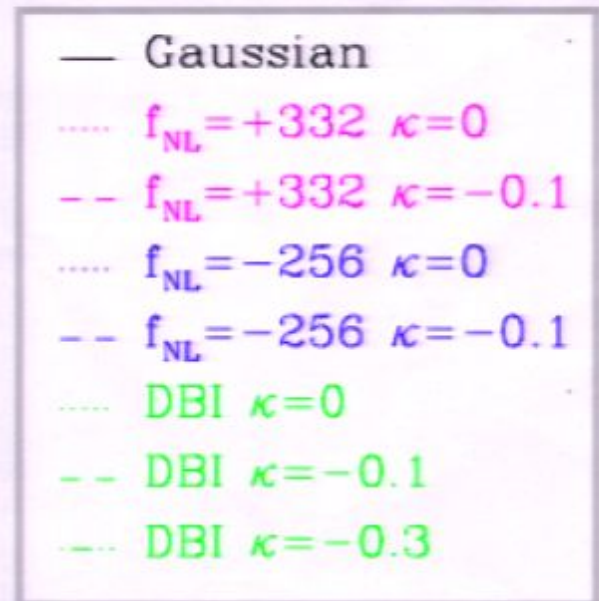
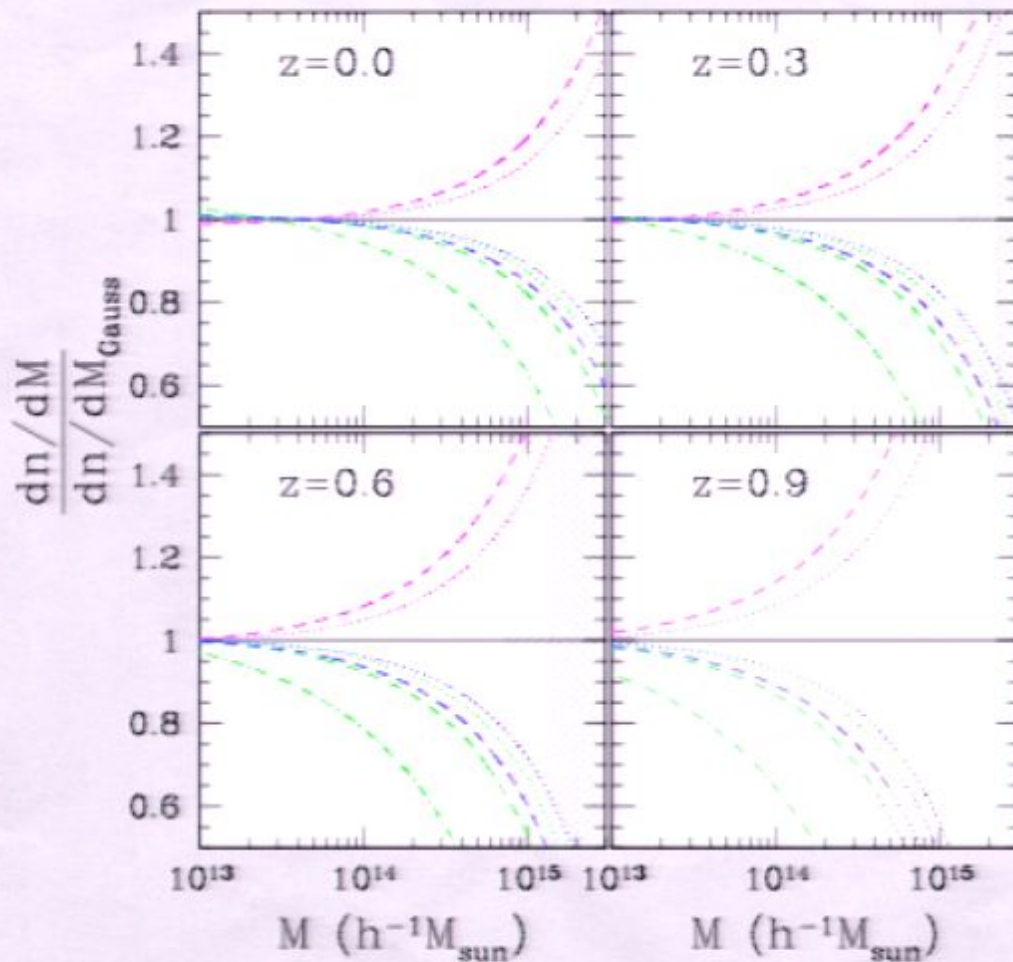


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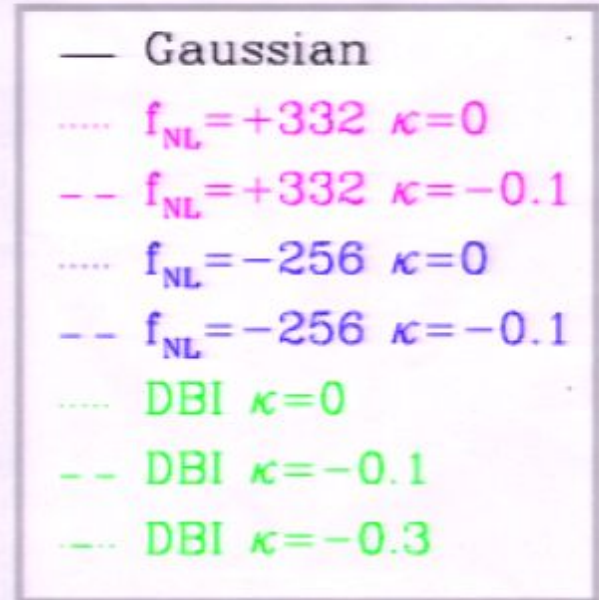
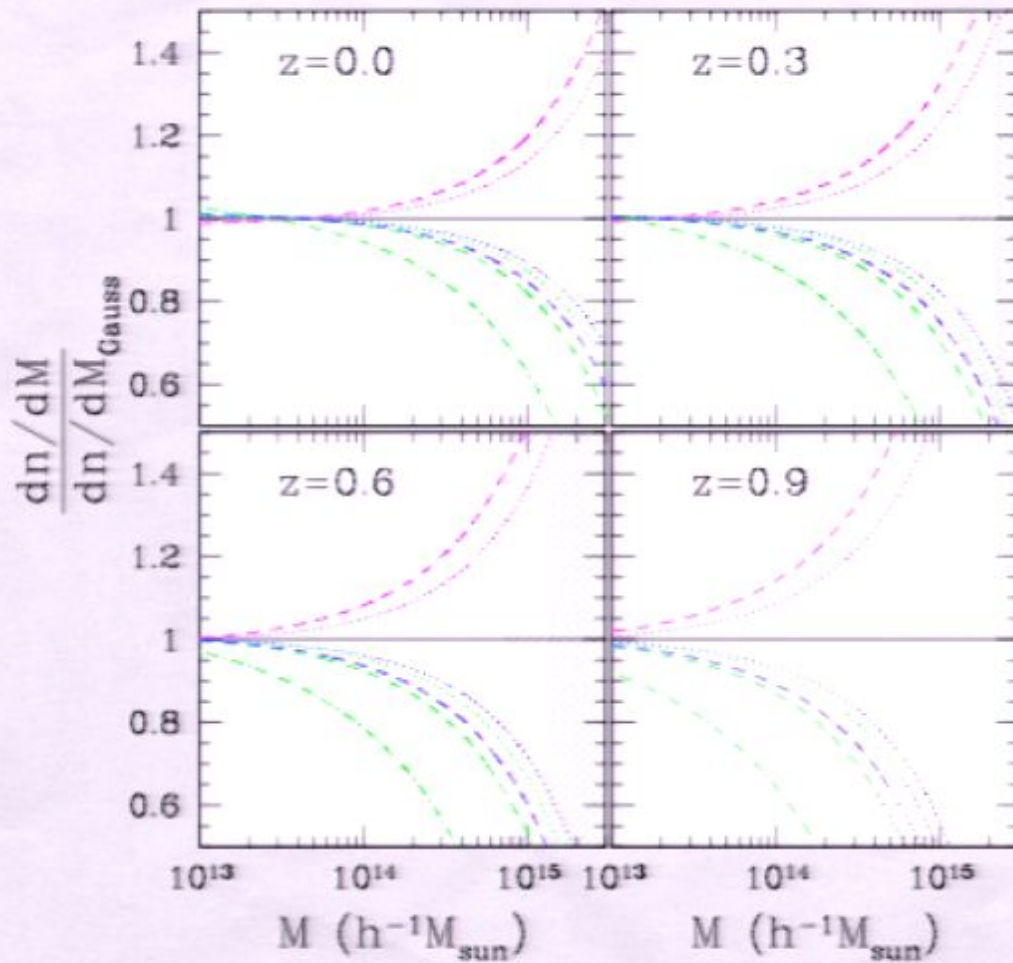


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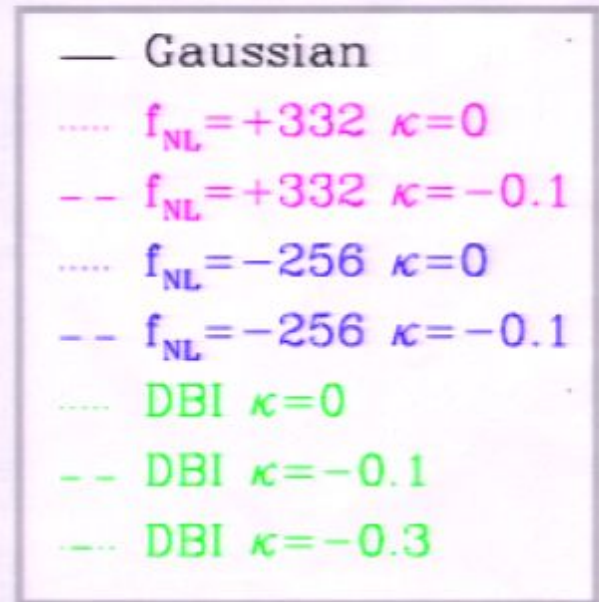
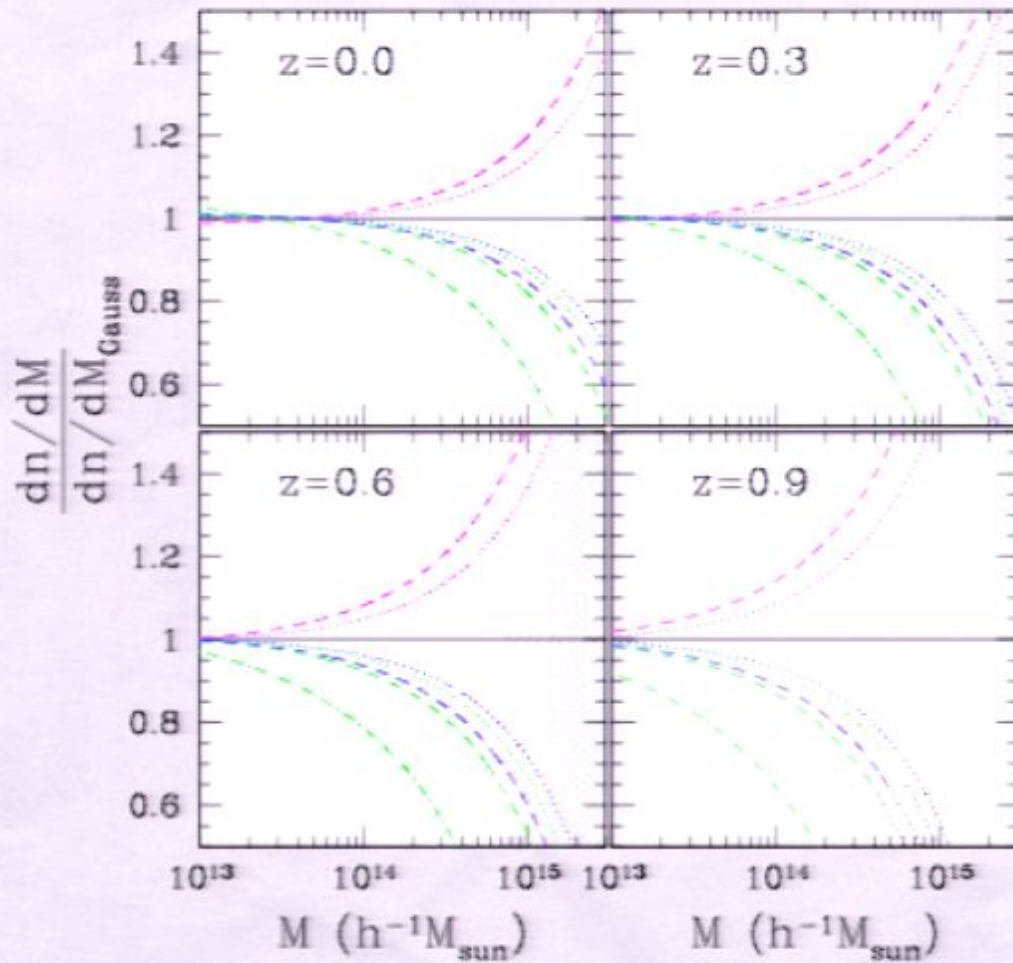
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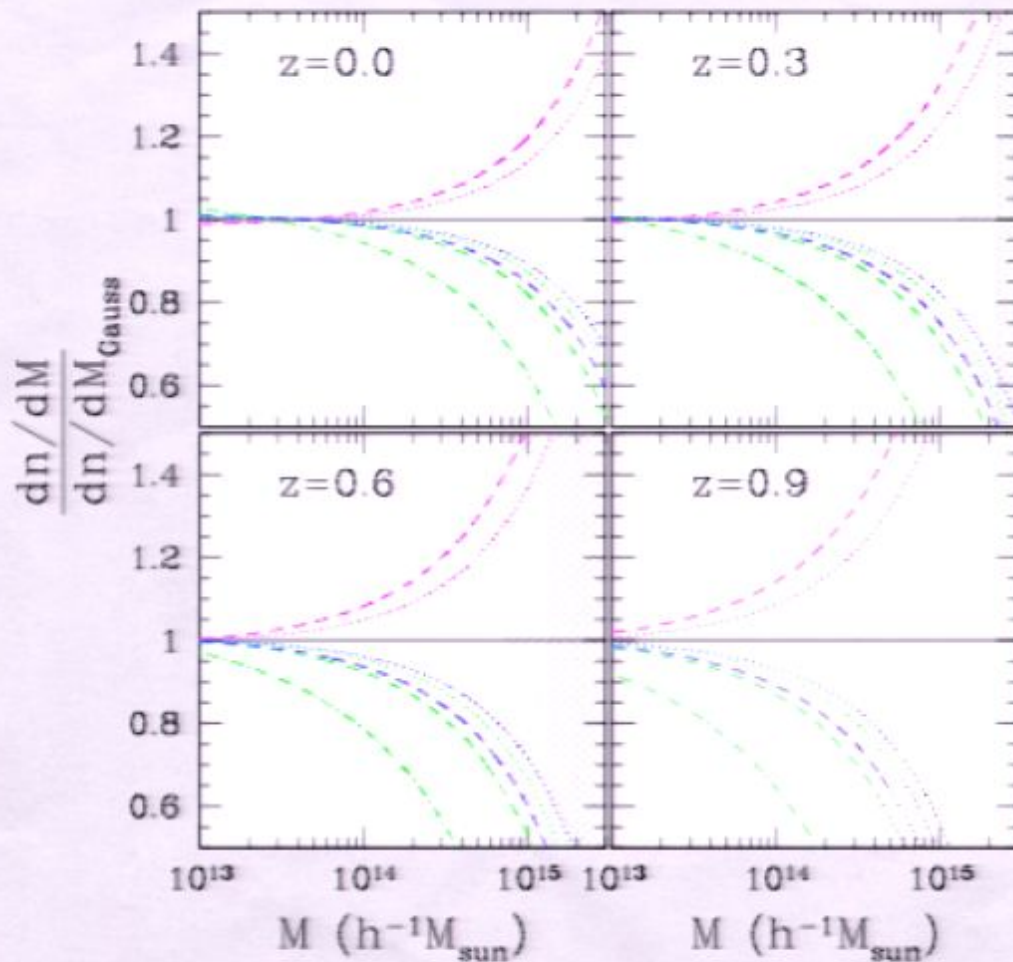
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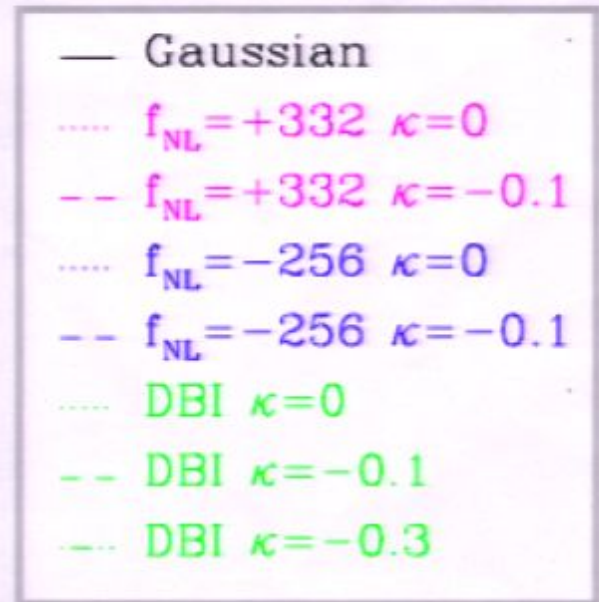
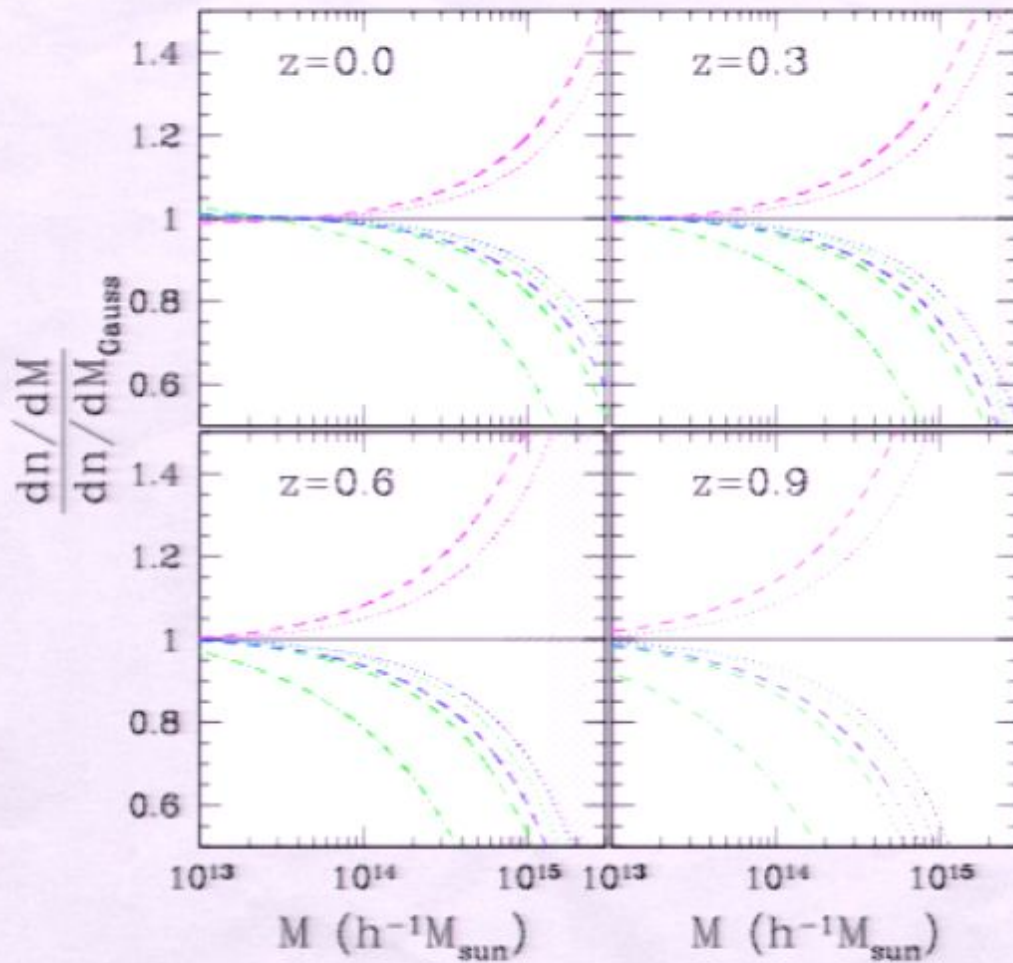


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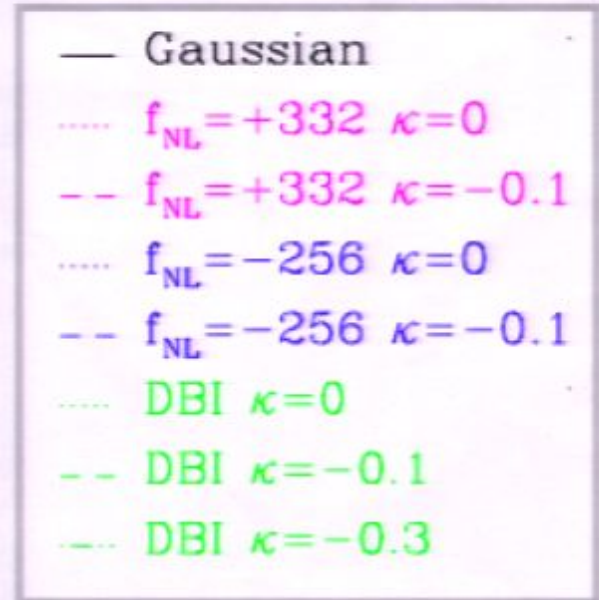
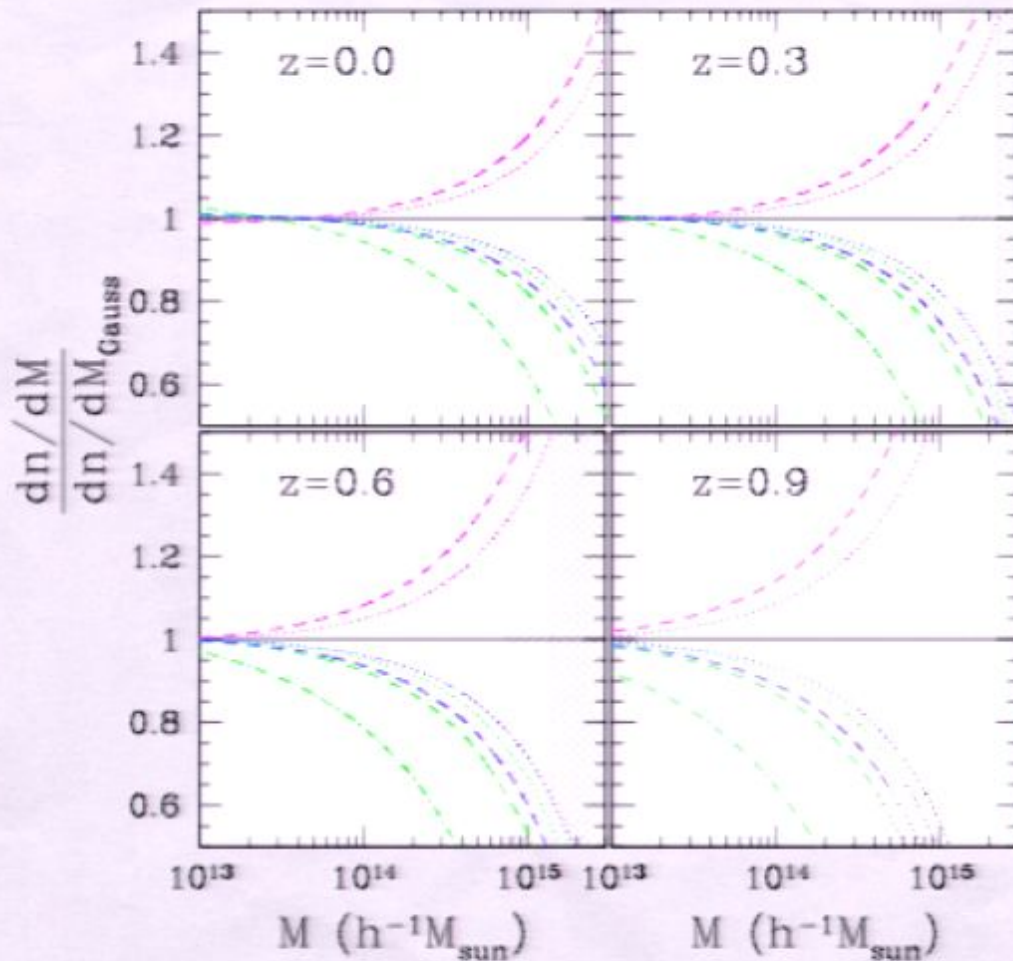


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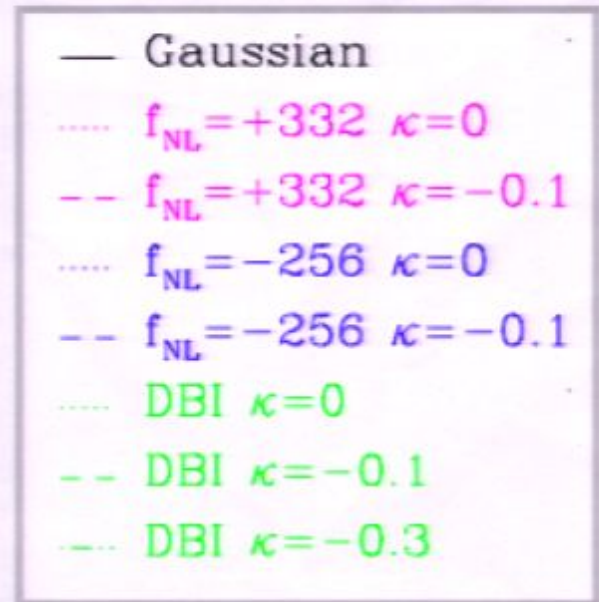
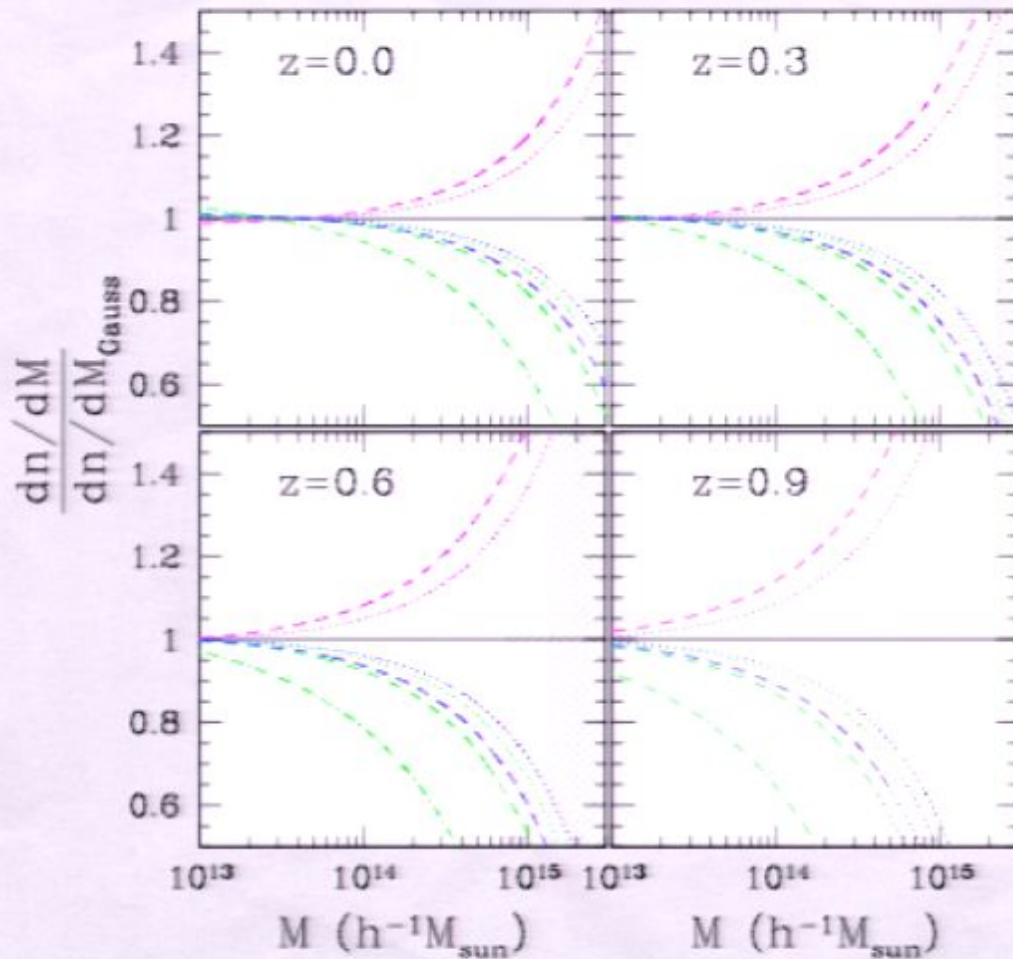
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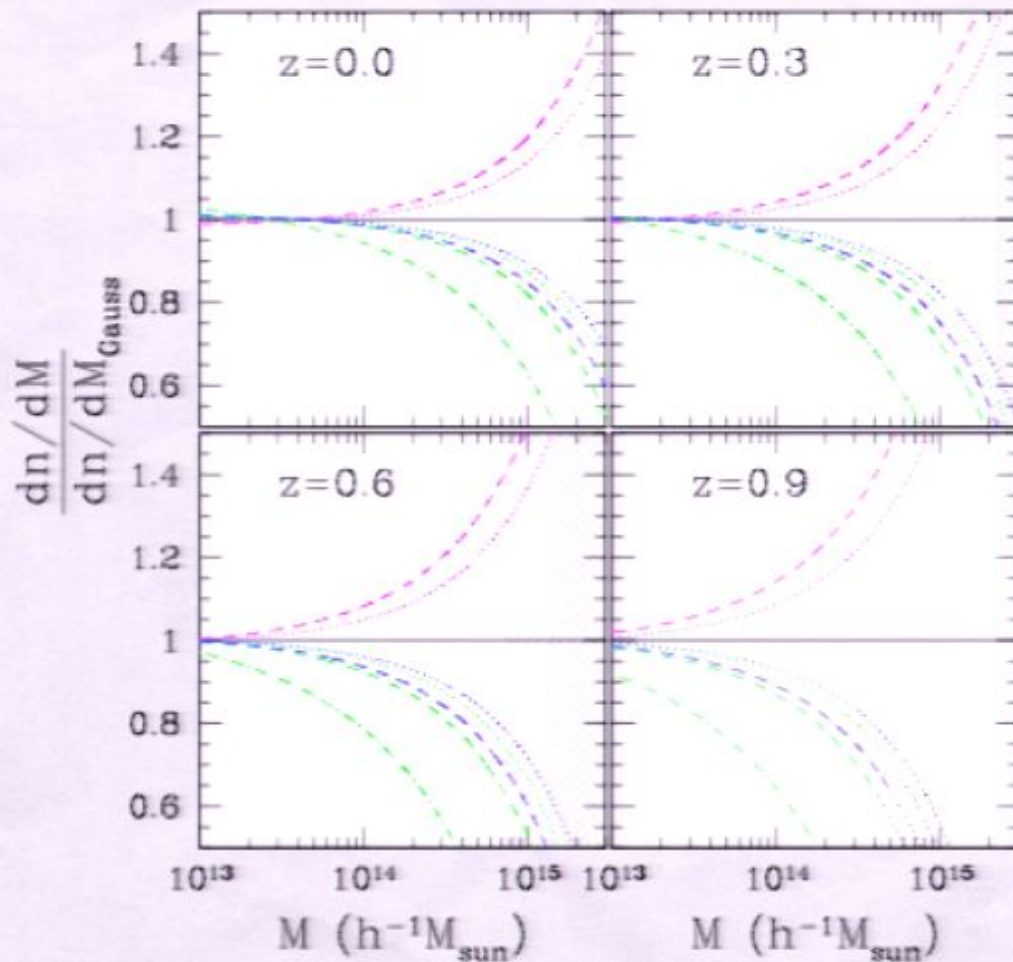
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MASS FUNCTION

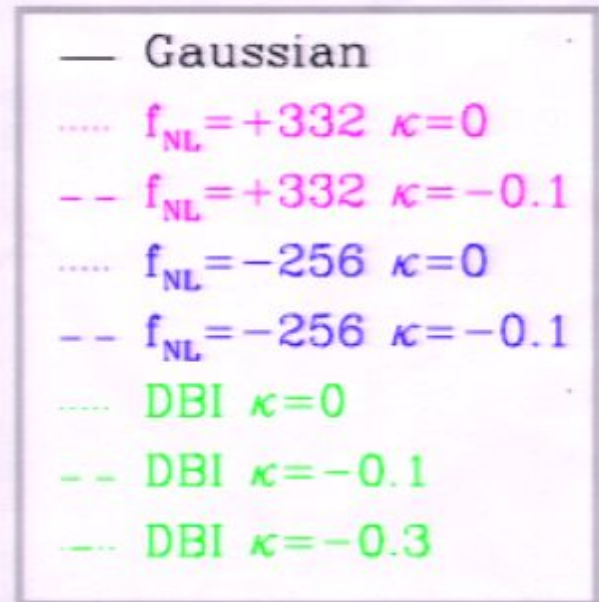
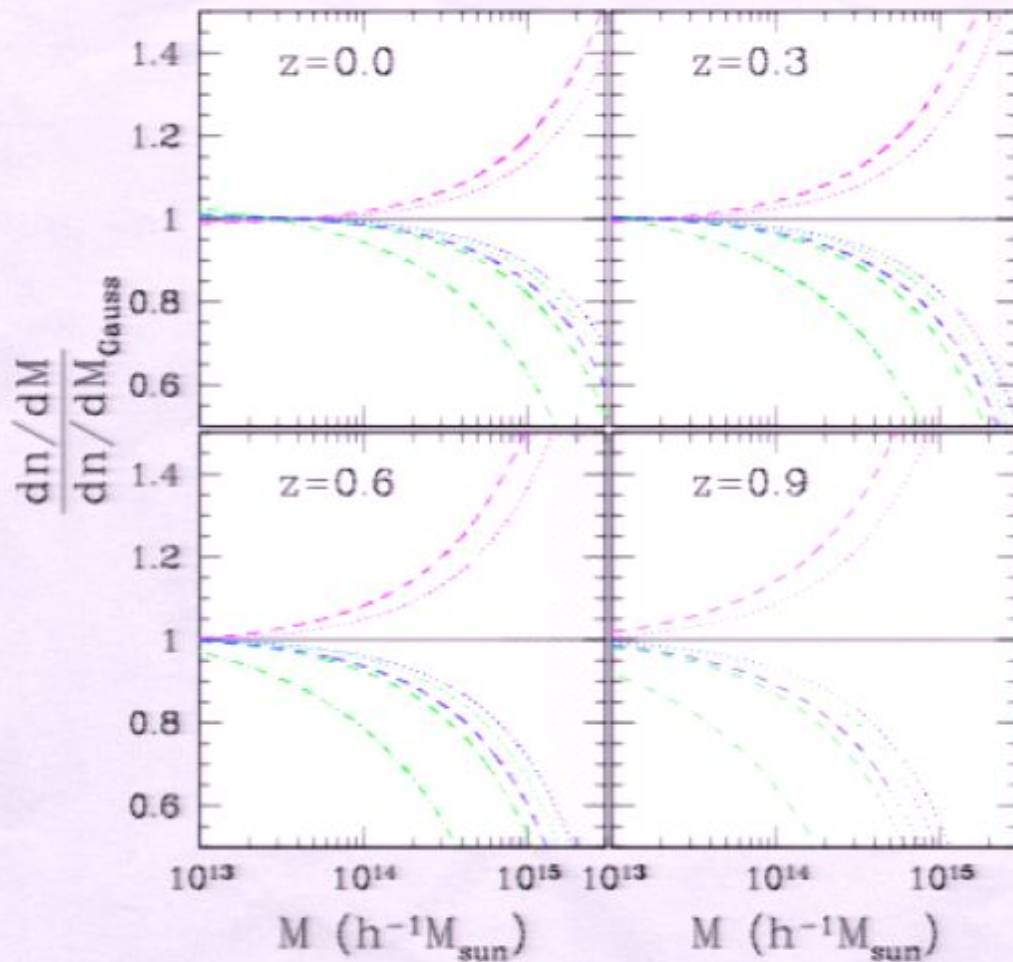


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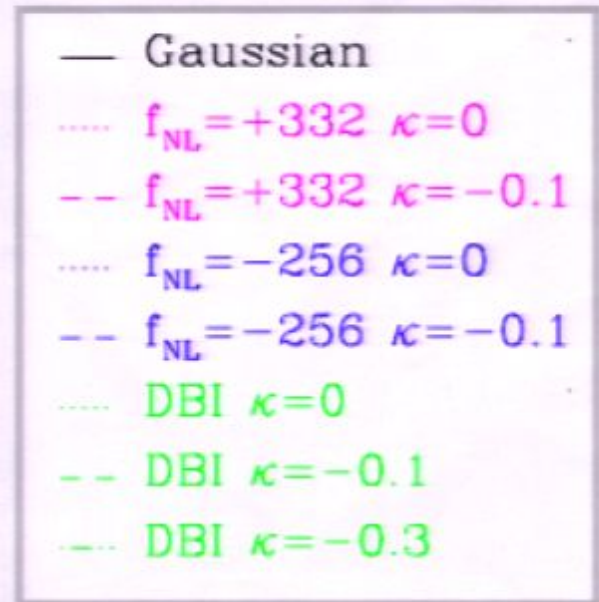
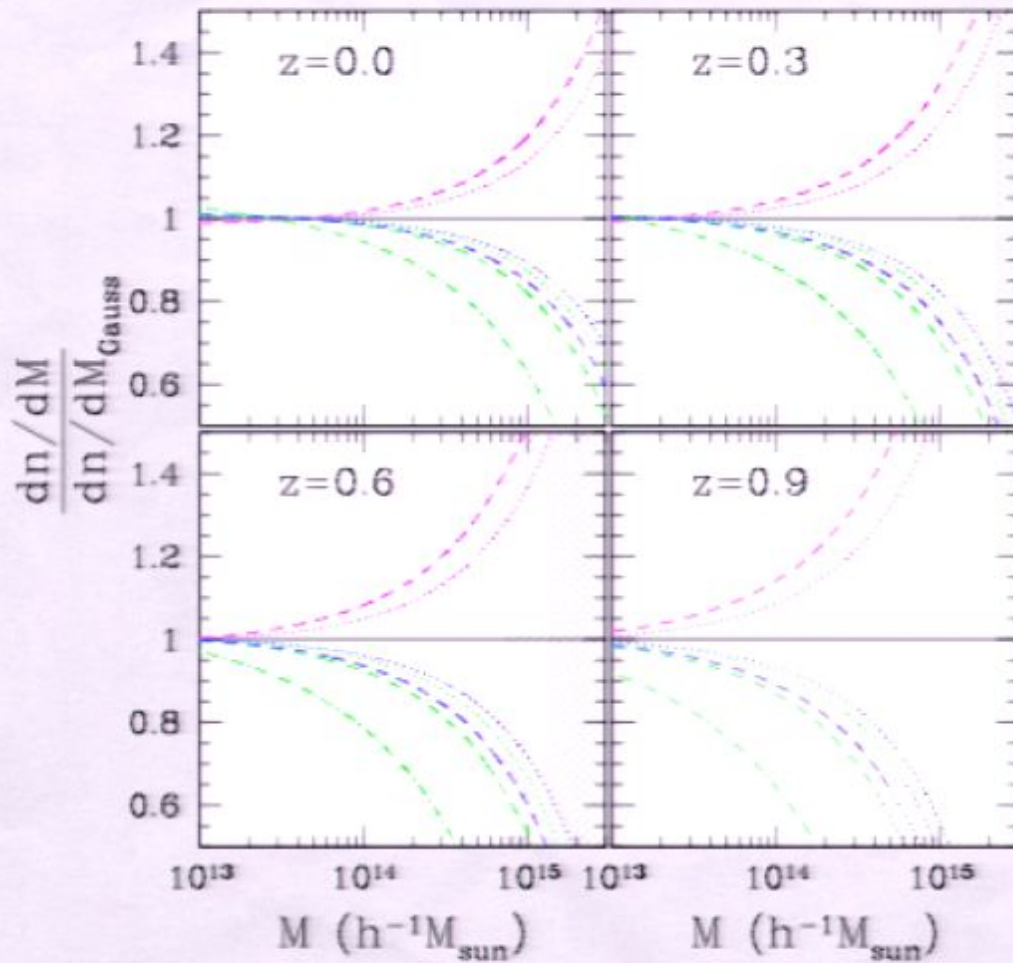


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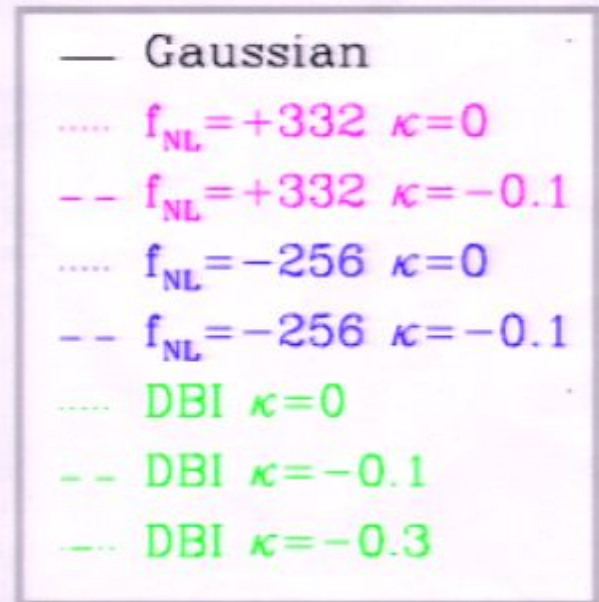
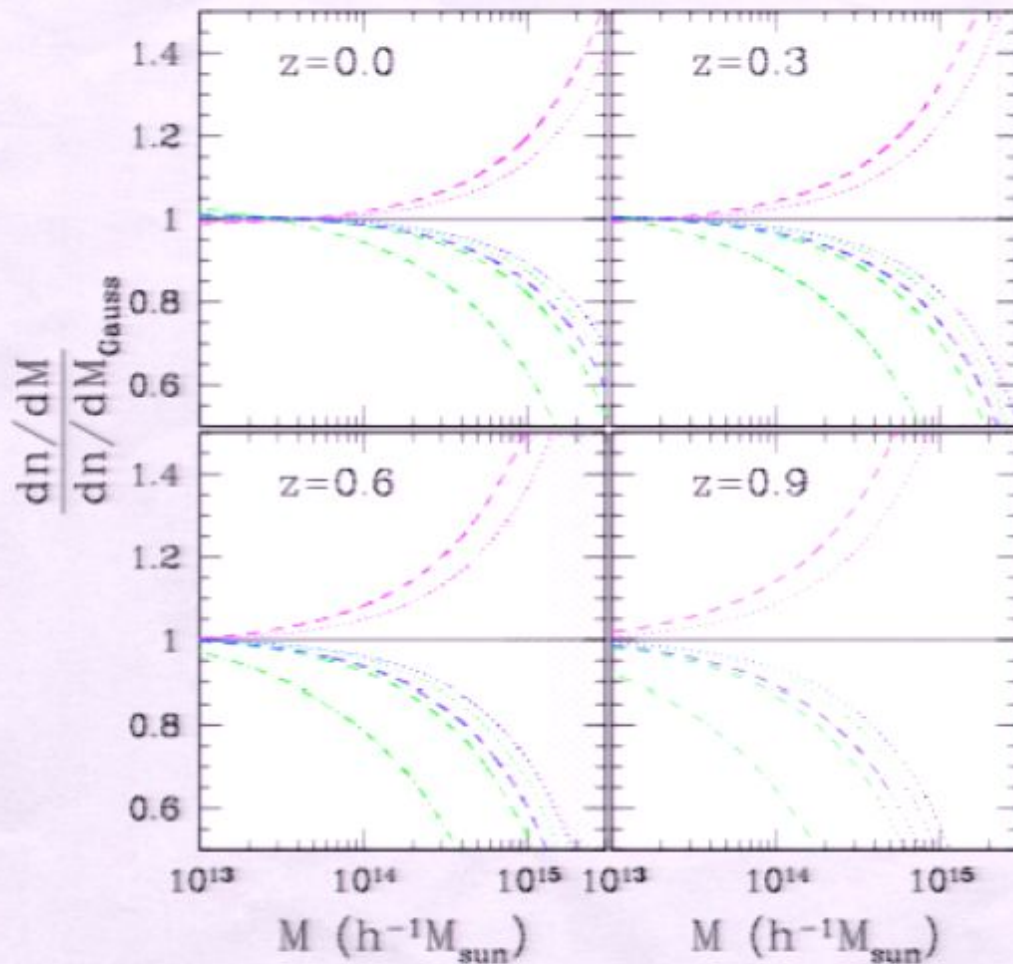
MASS FUNCTION



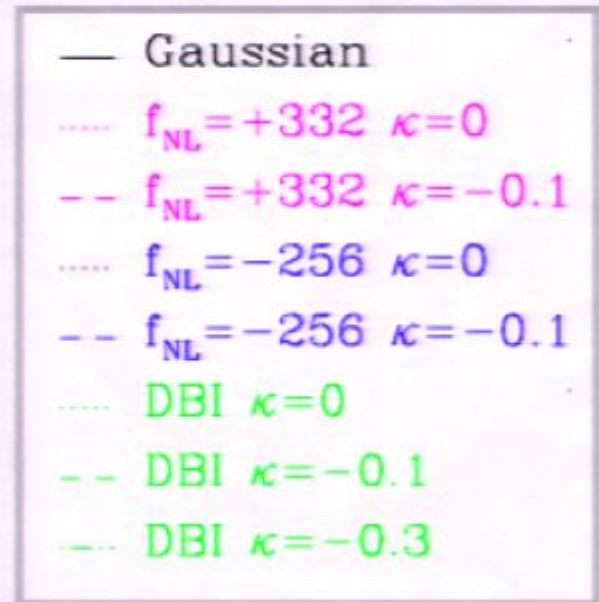
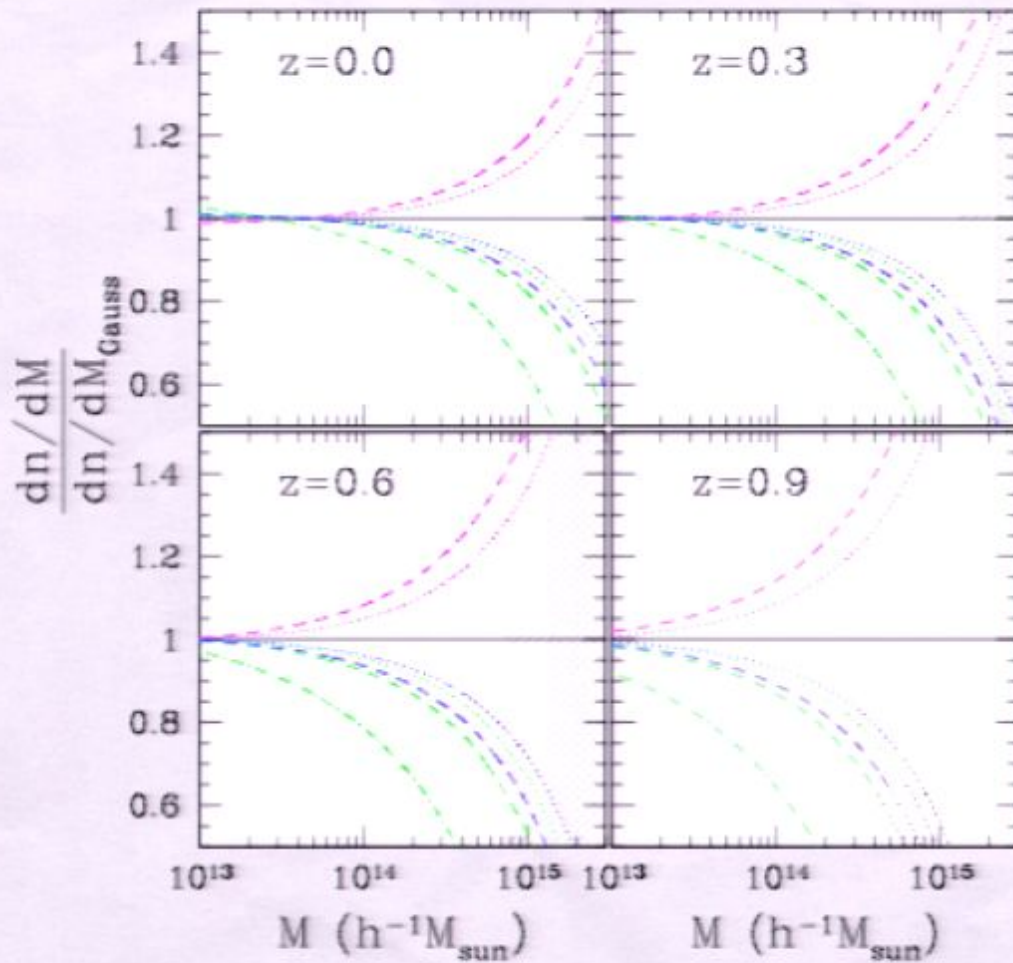
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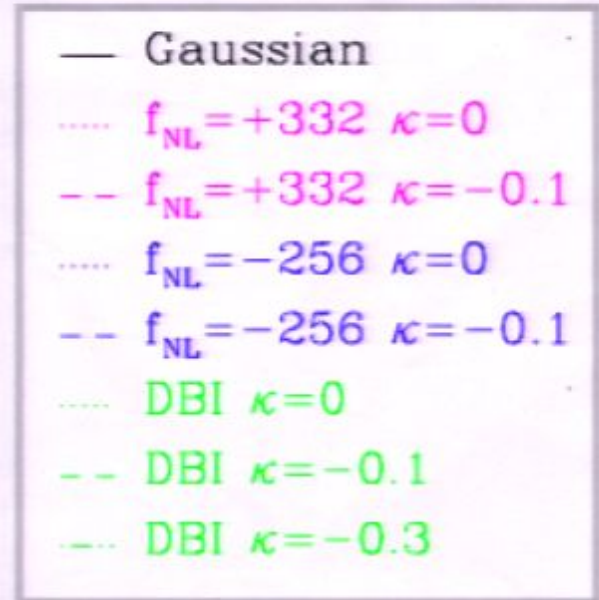
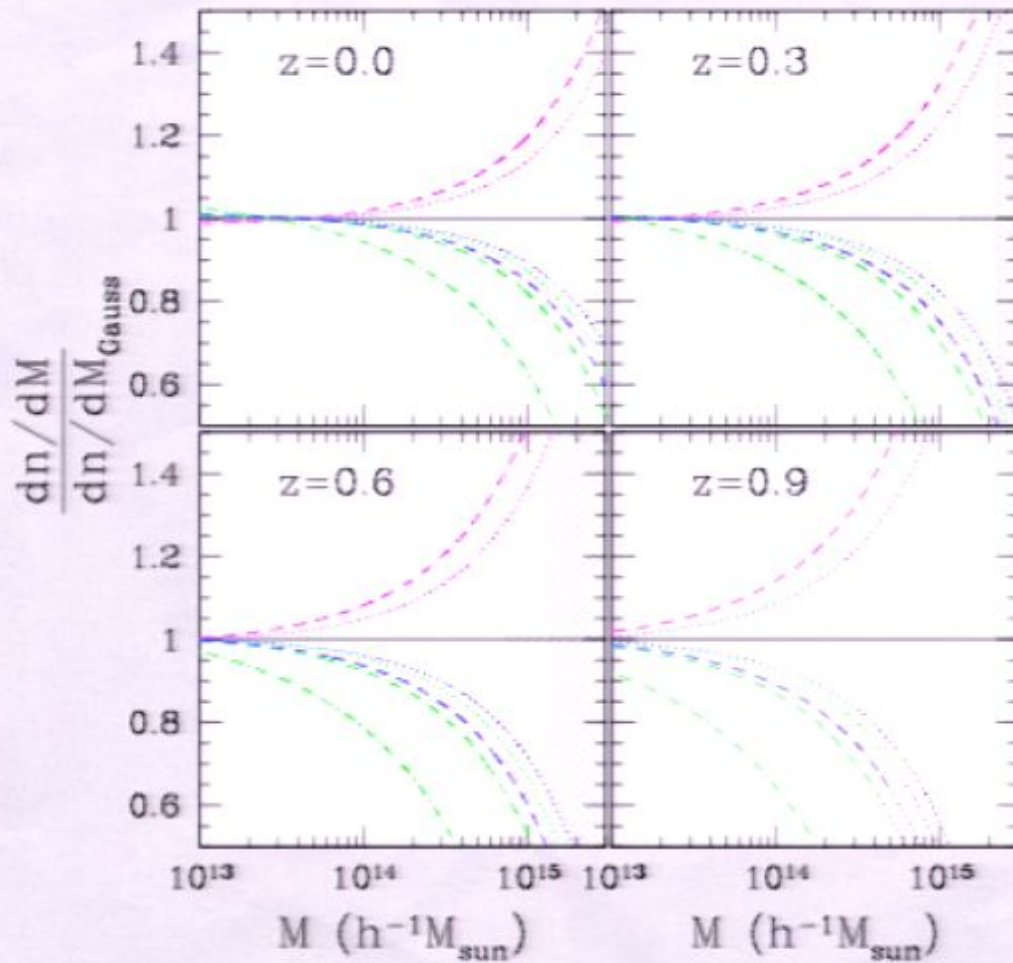
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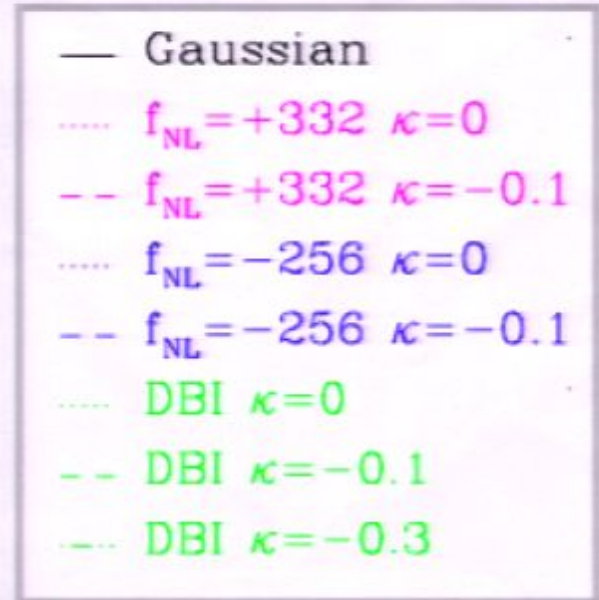
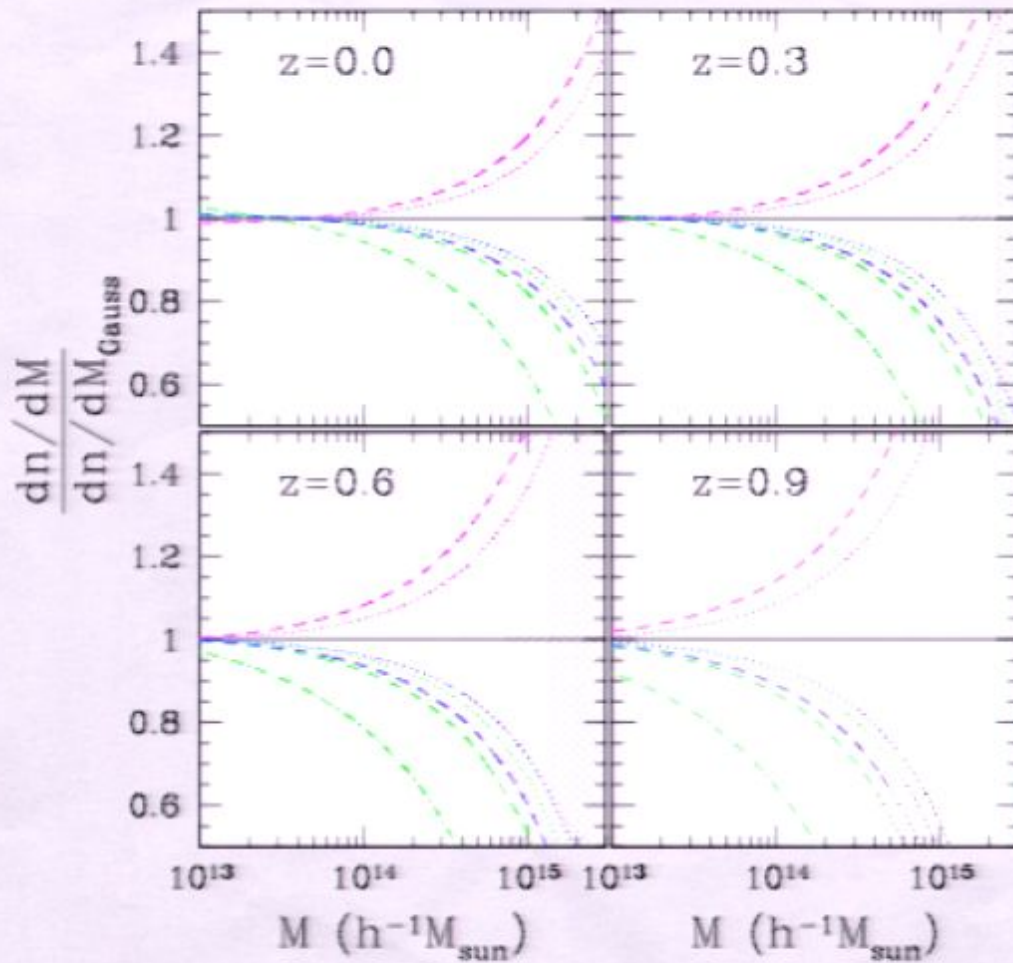
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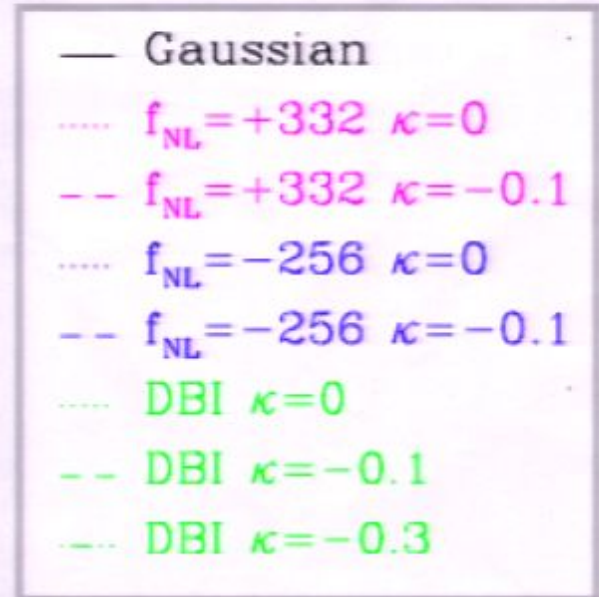
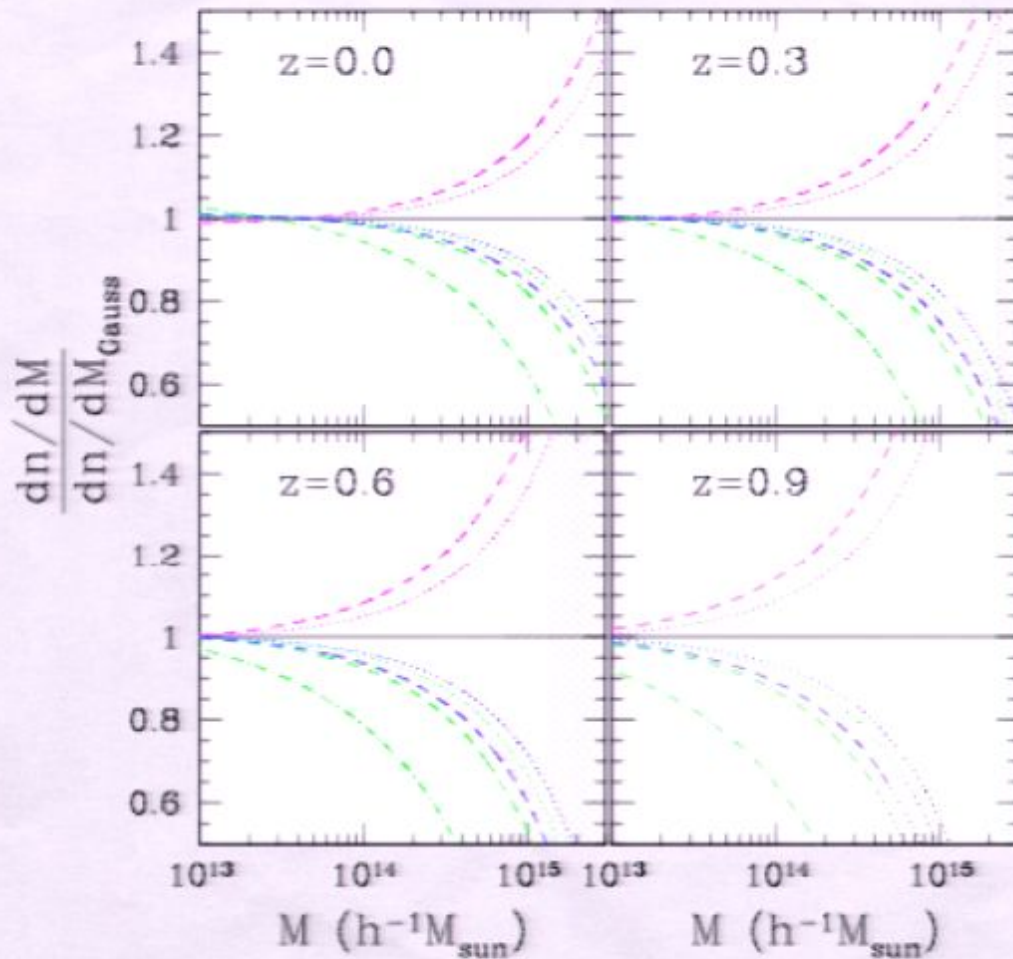
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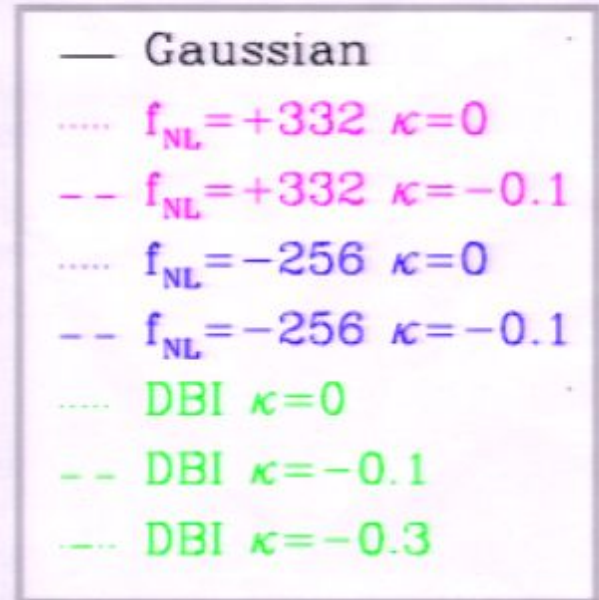
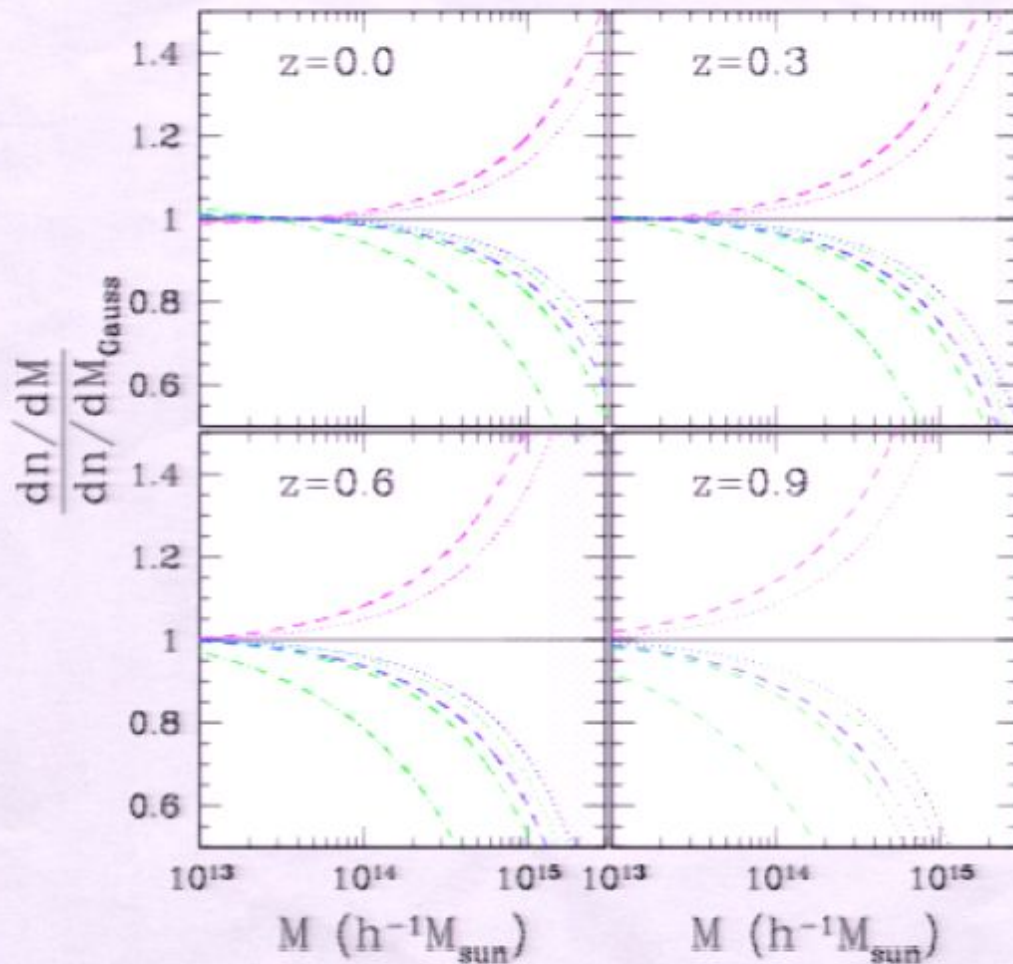
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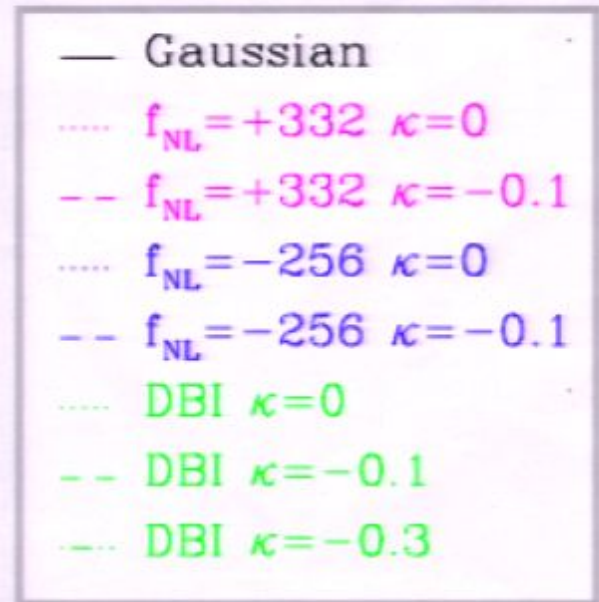
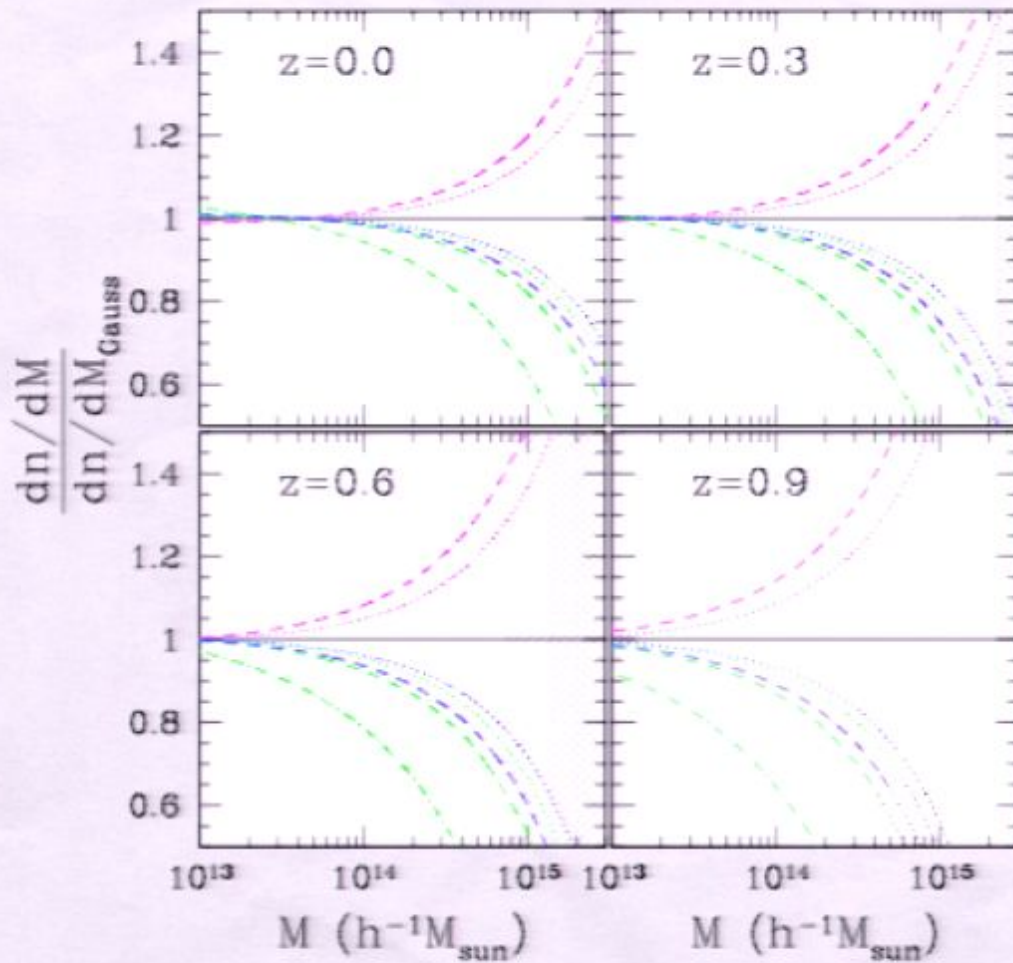
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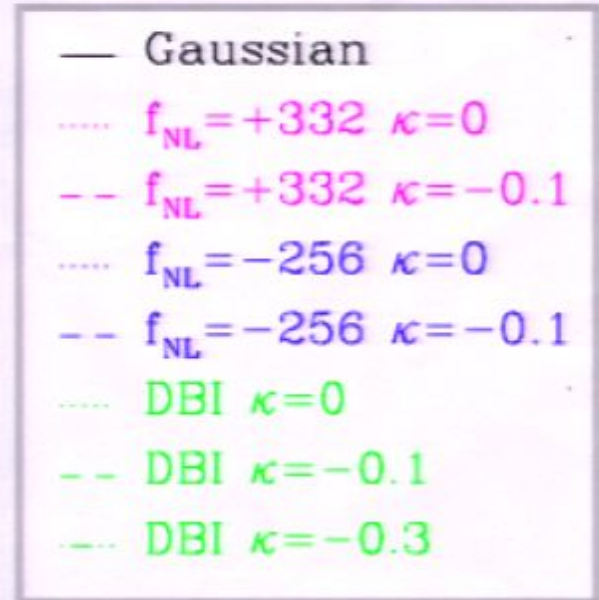
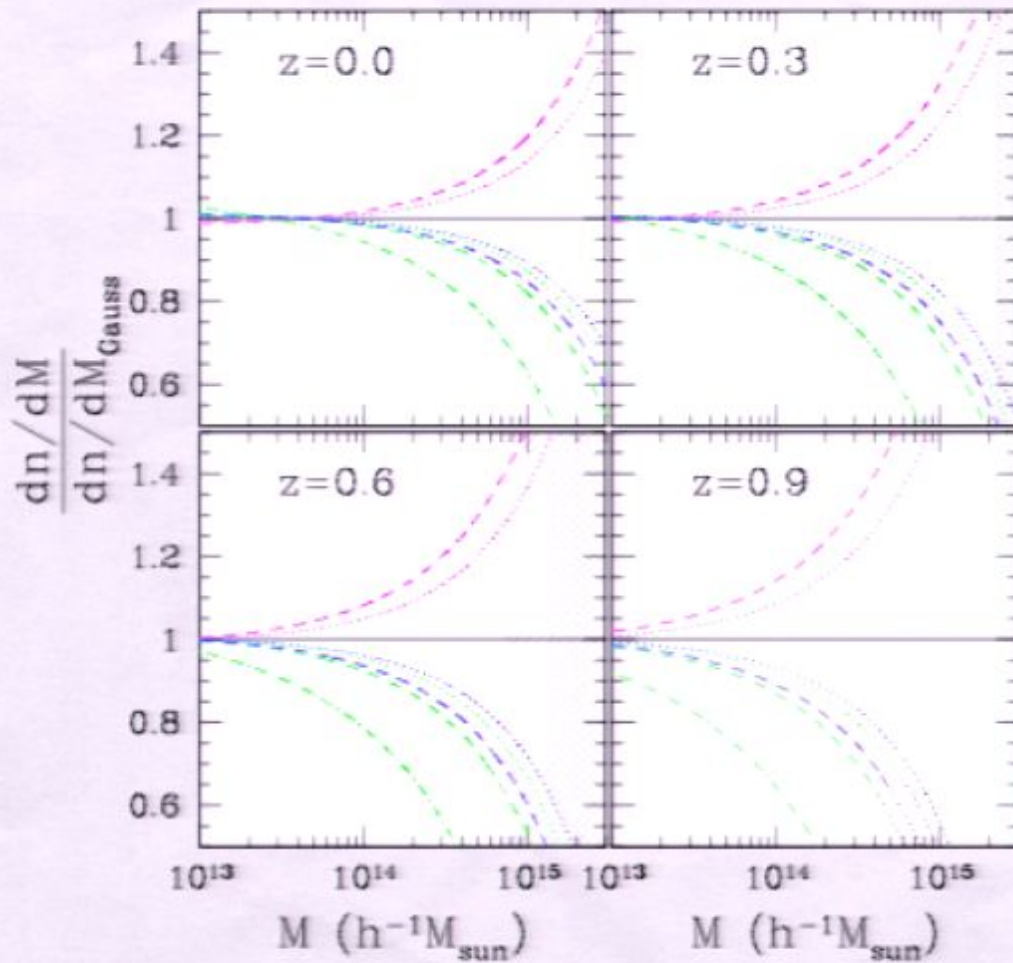
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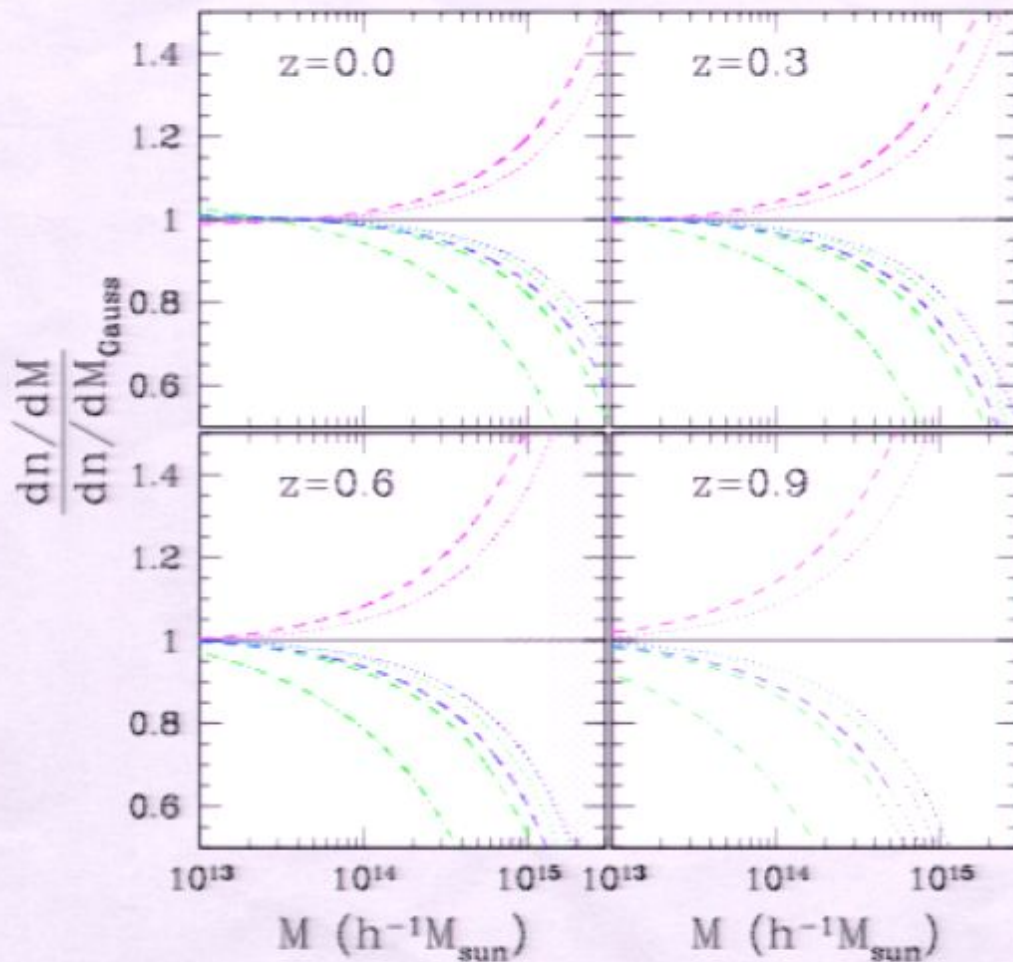
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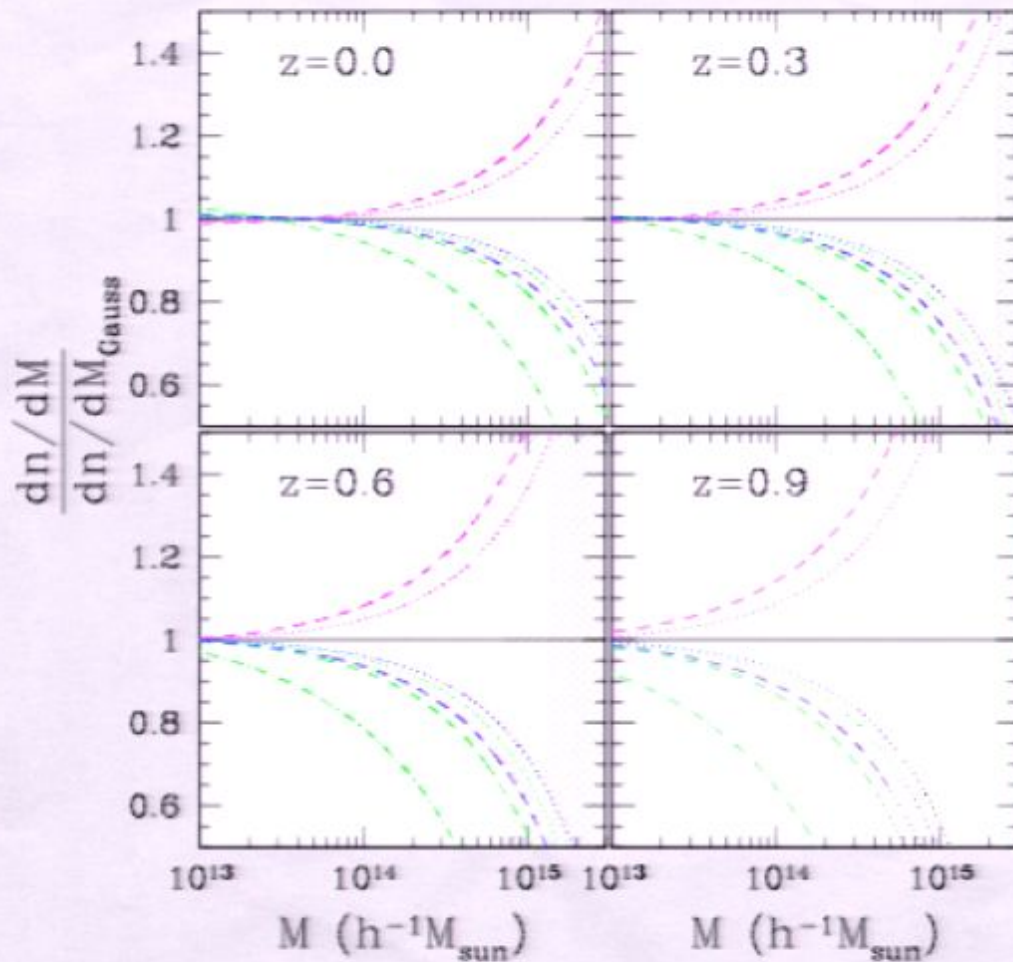


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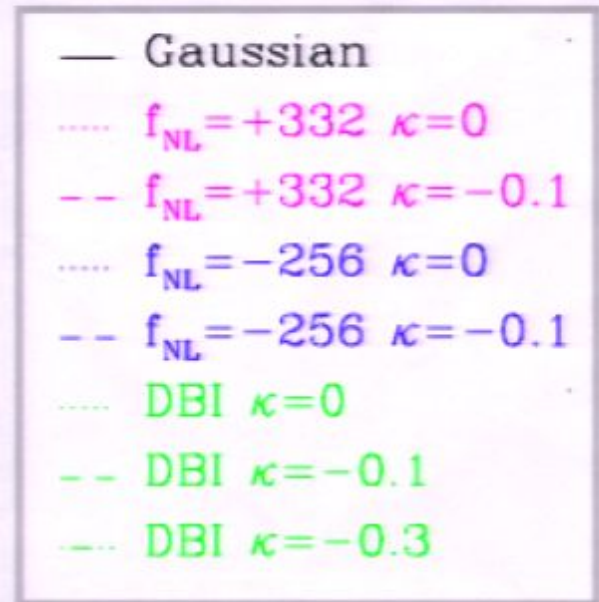
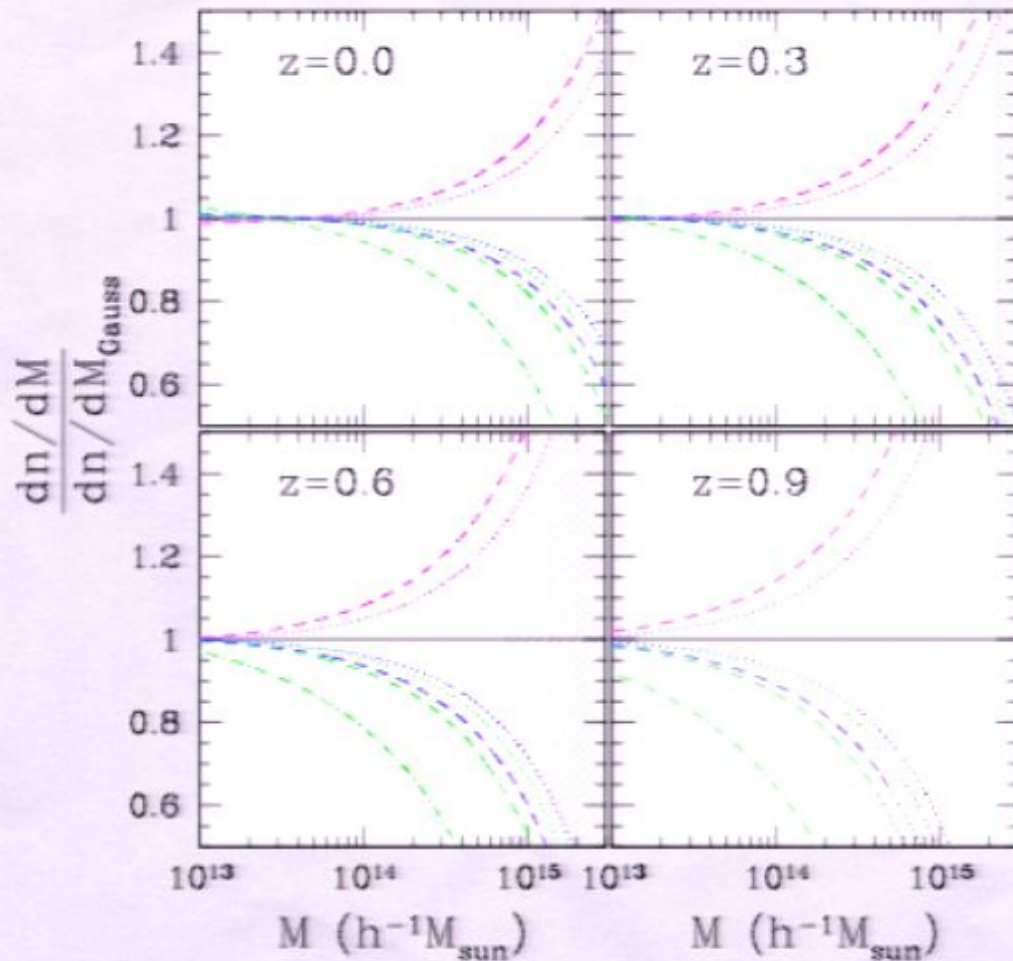
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MASS FUNCTION

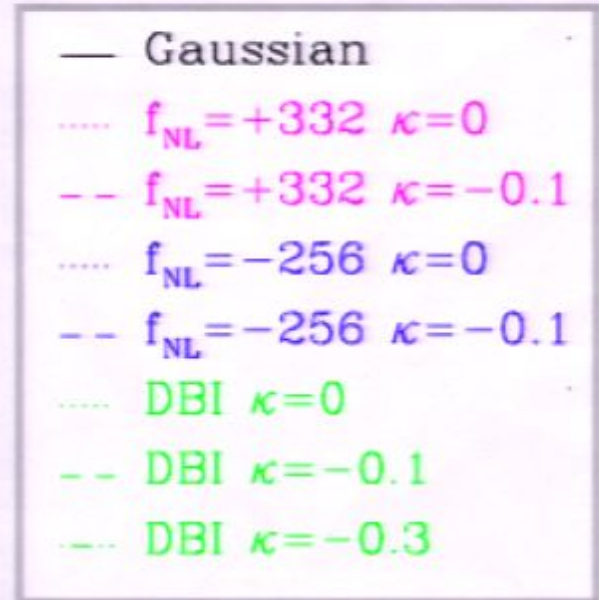
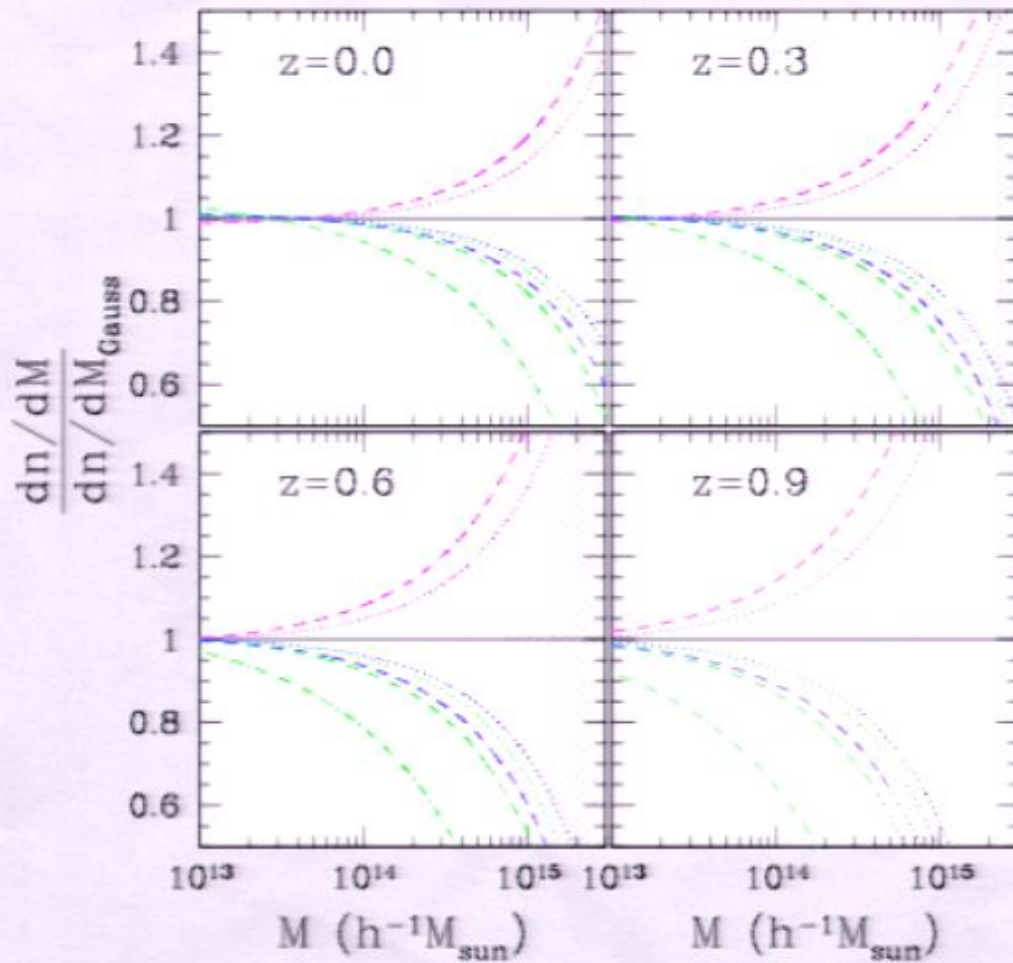


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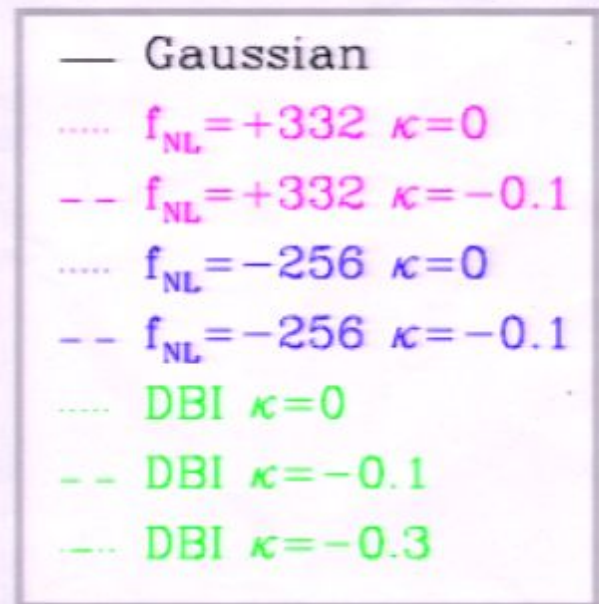
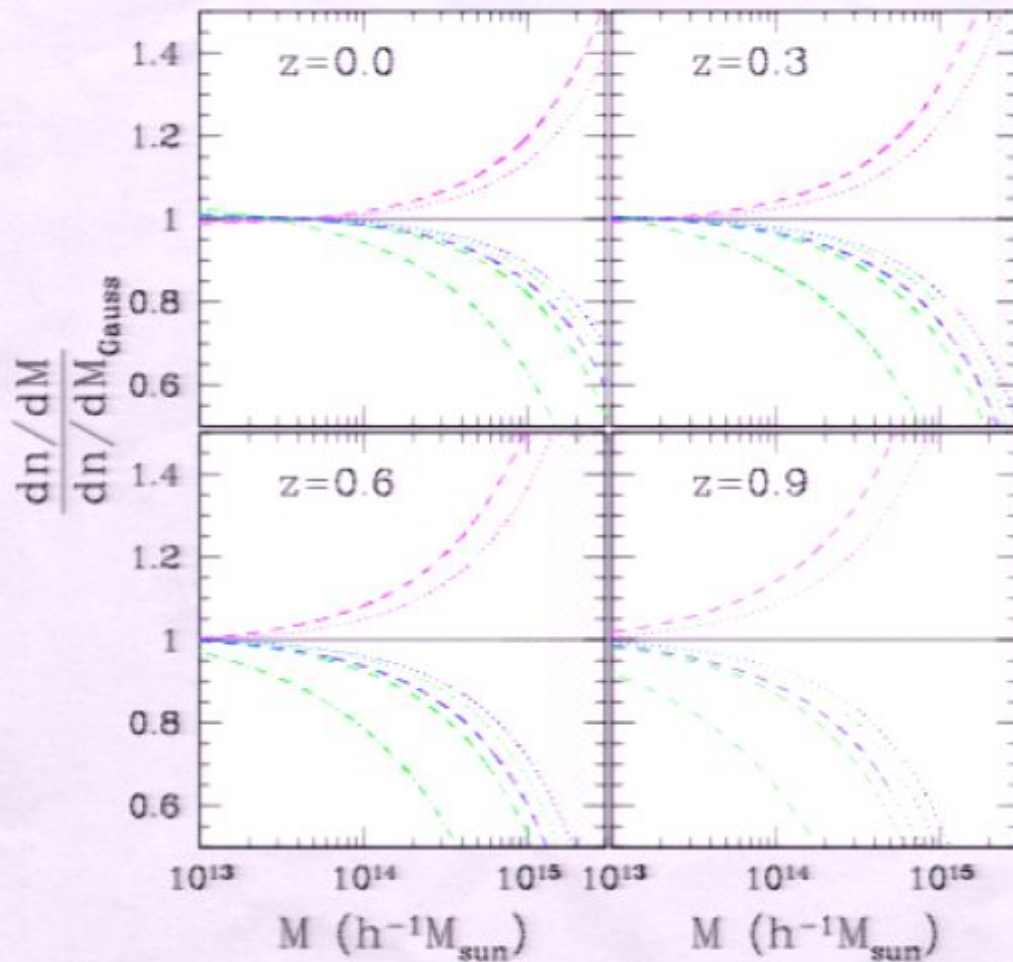
MASS FUNCTION



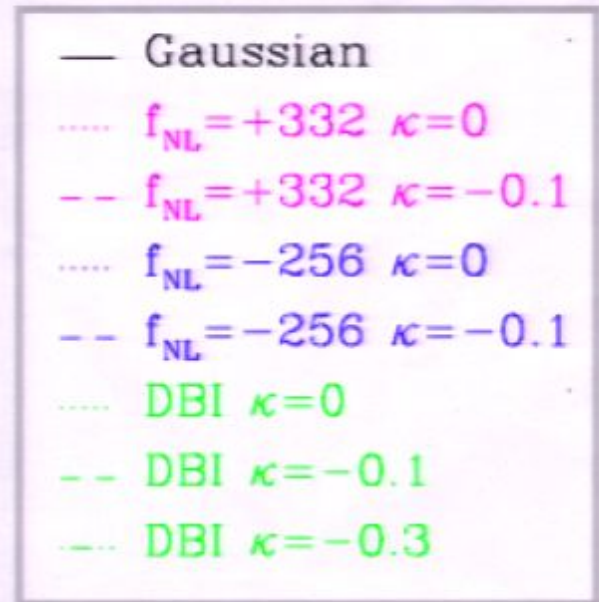
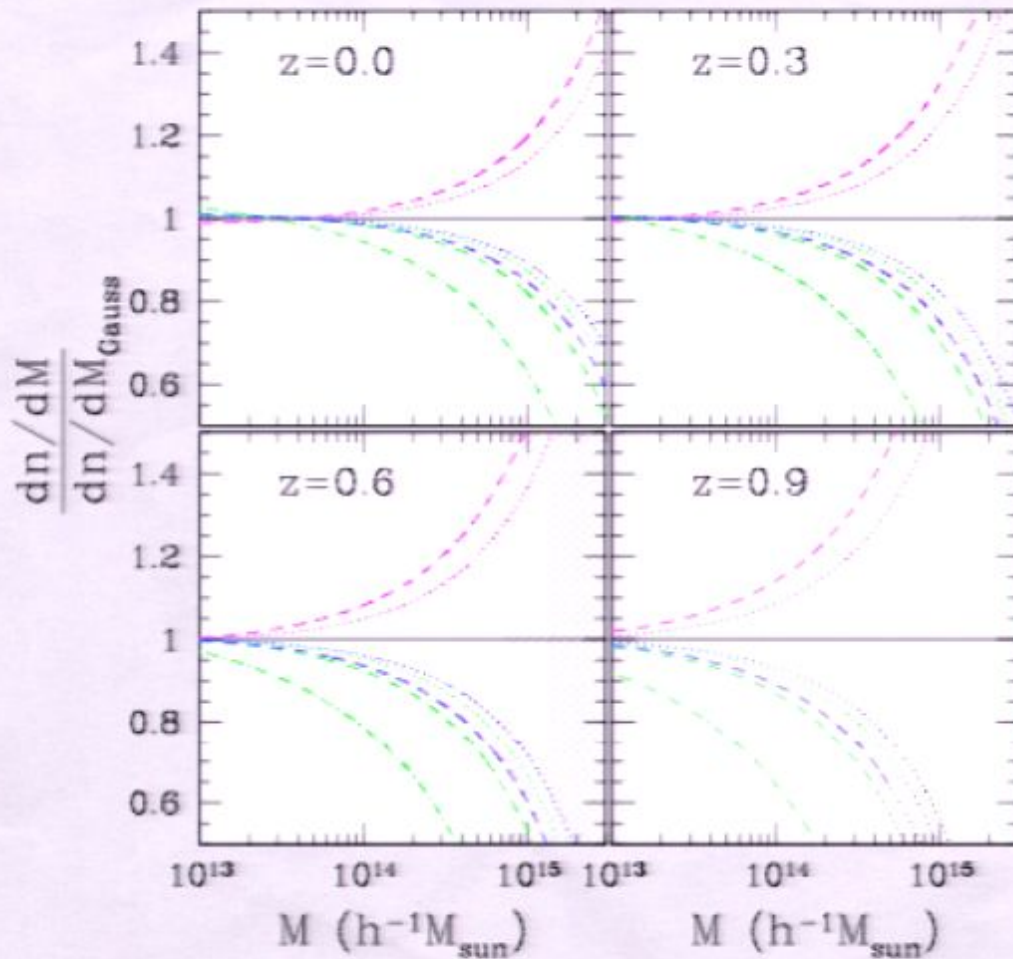
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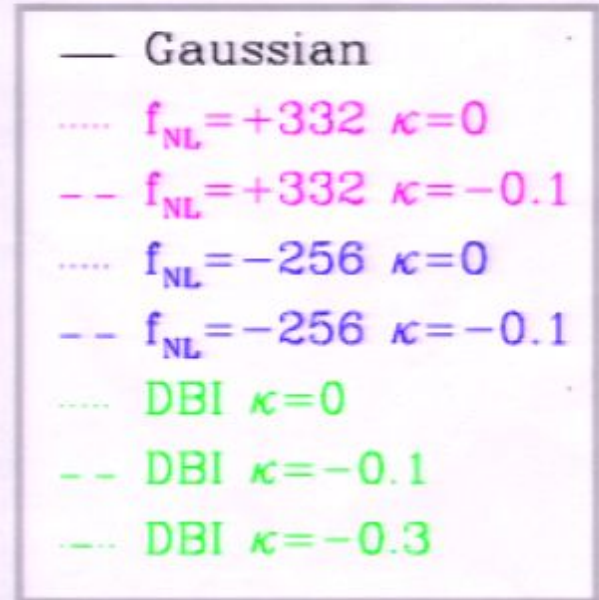
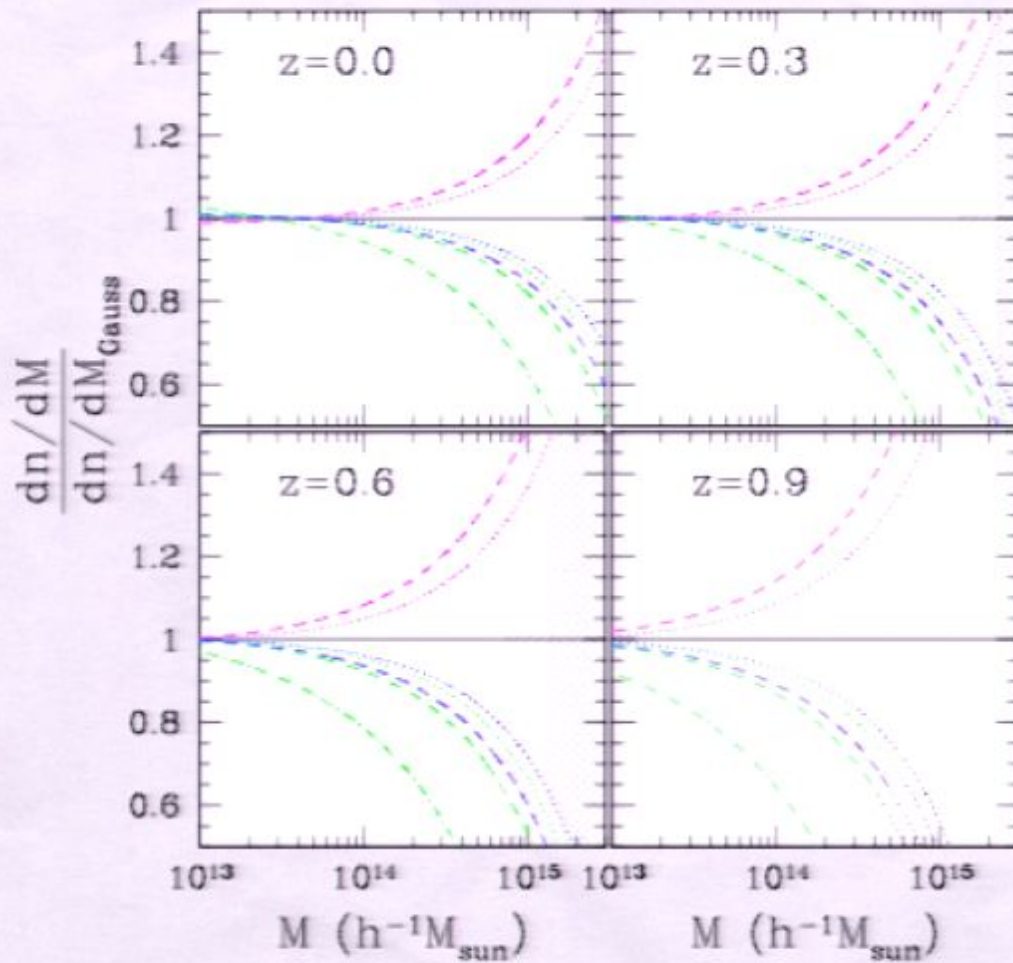
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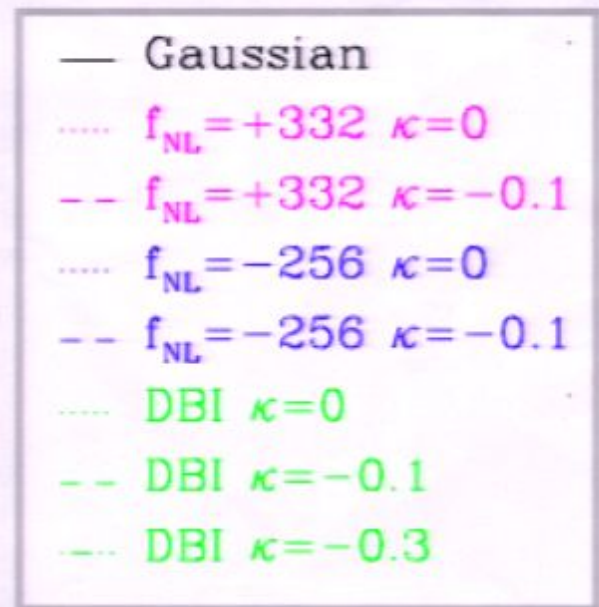
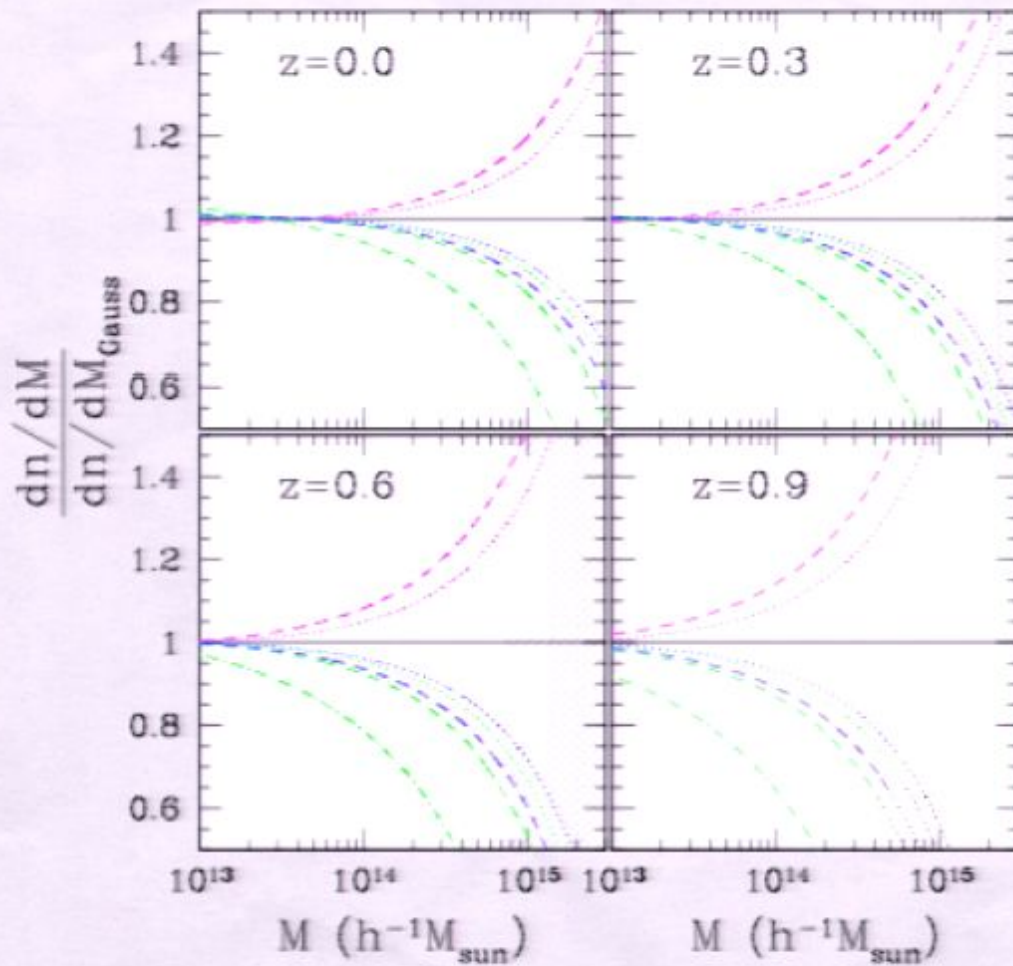
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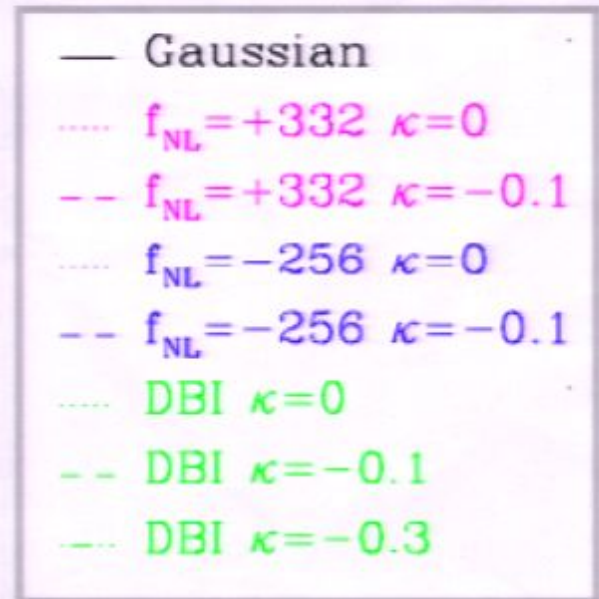
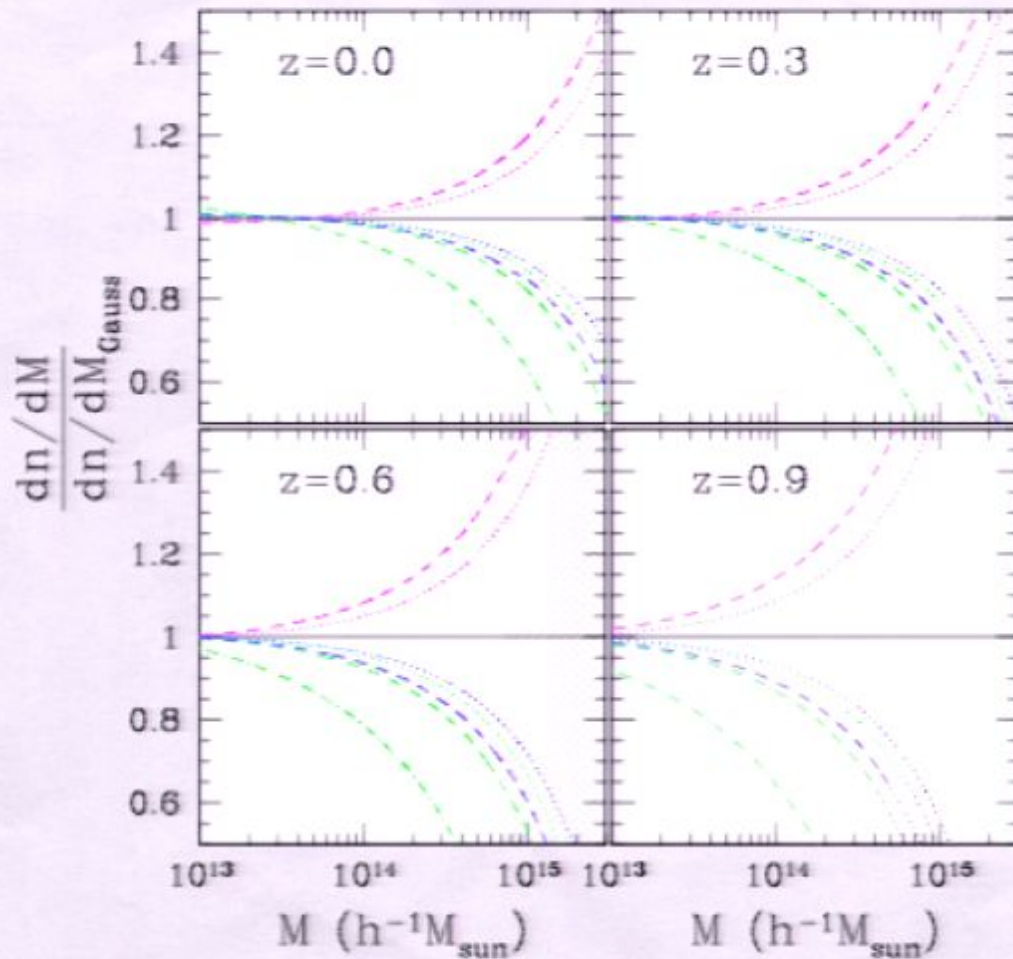
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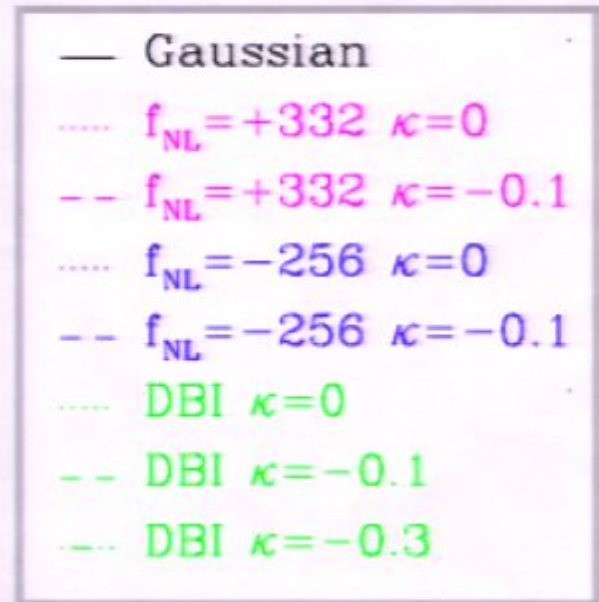
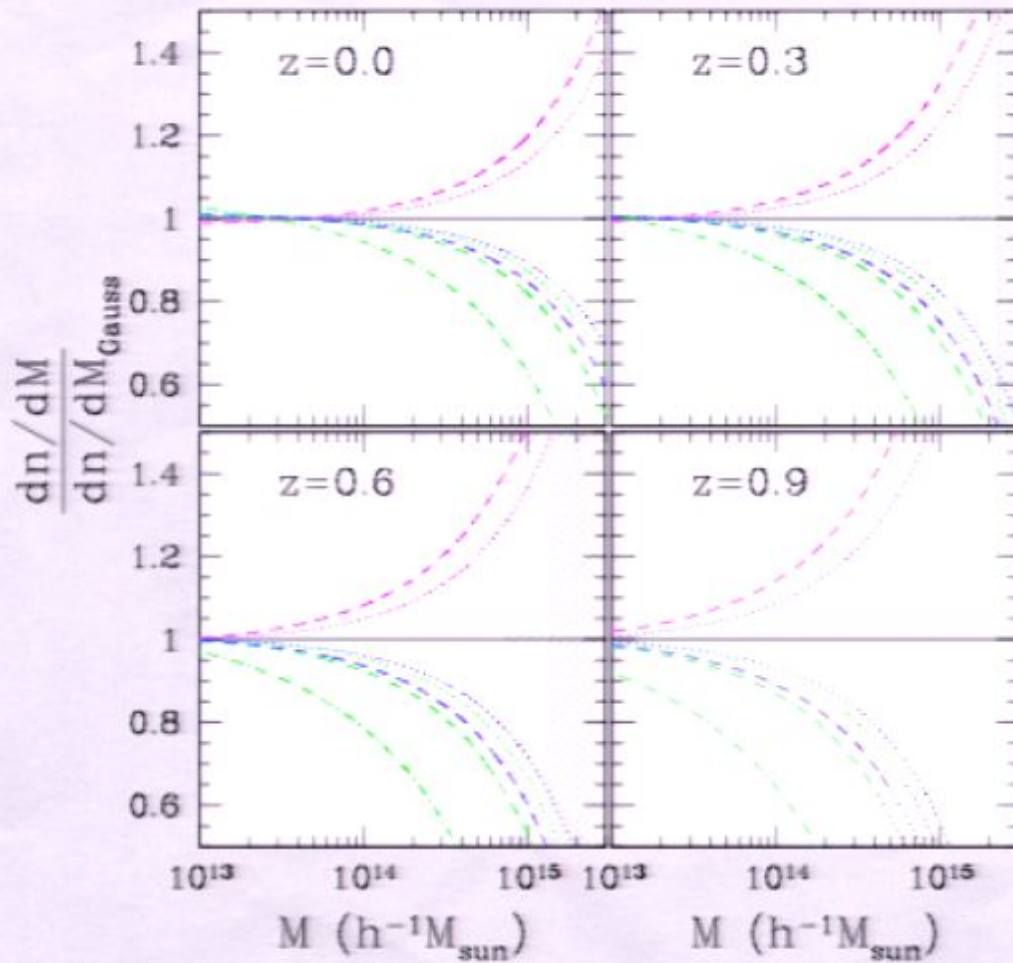
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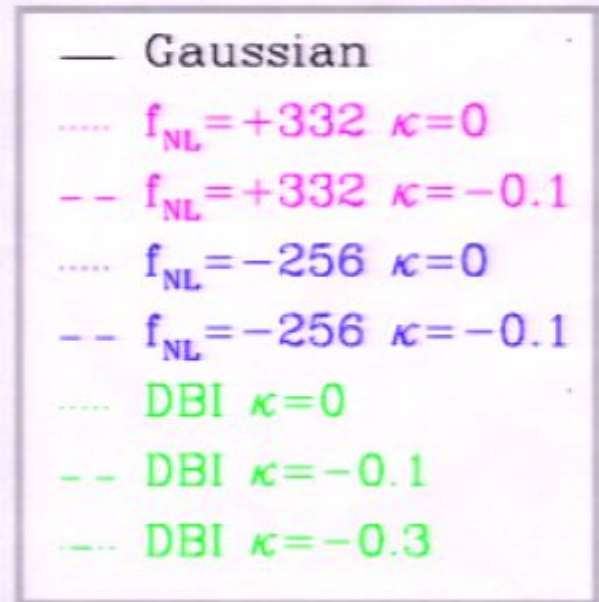
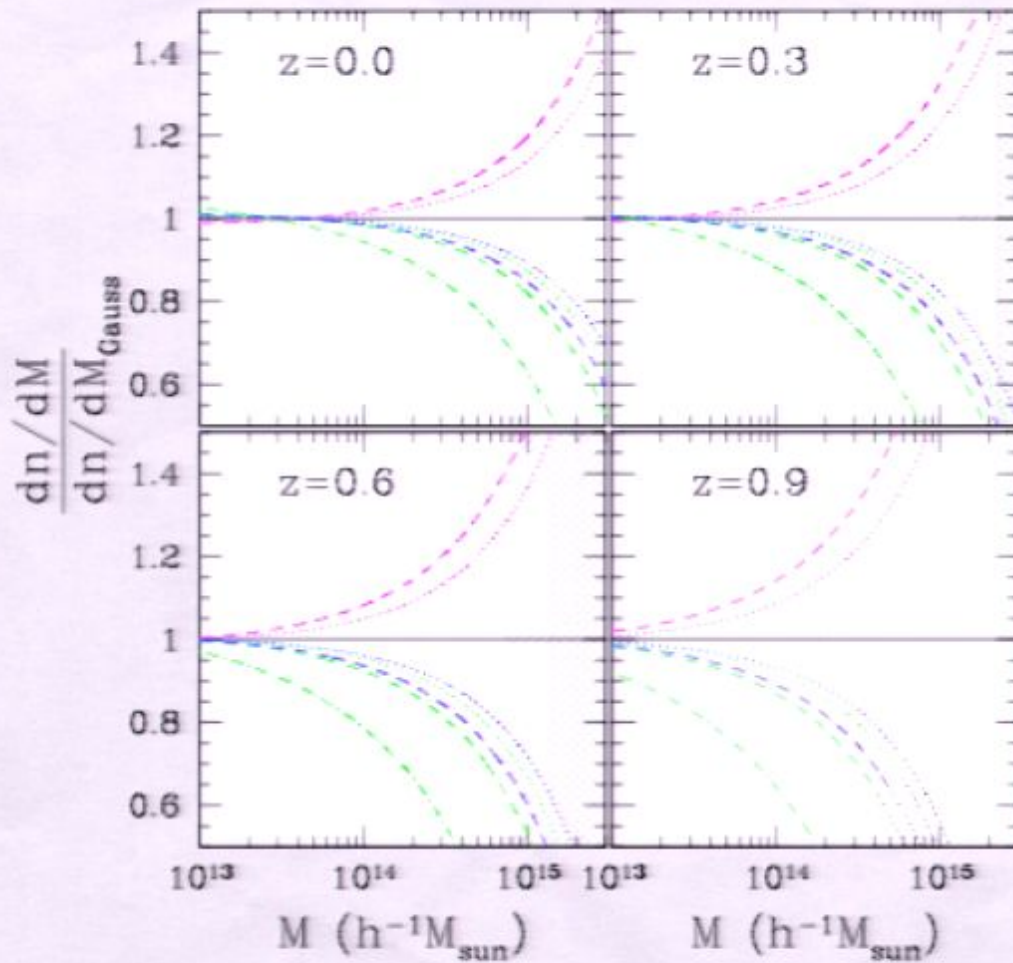
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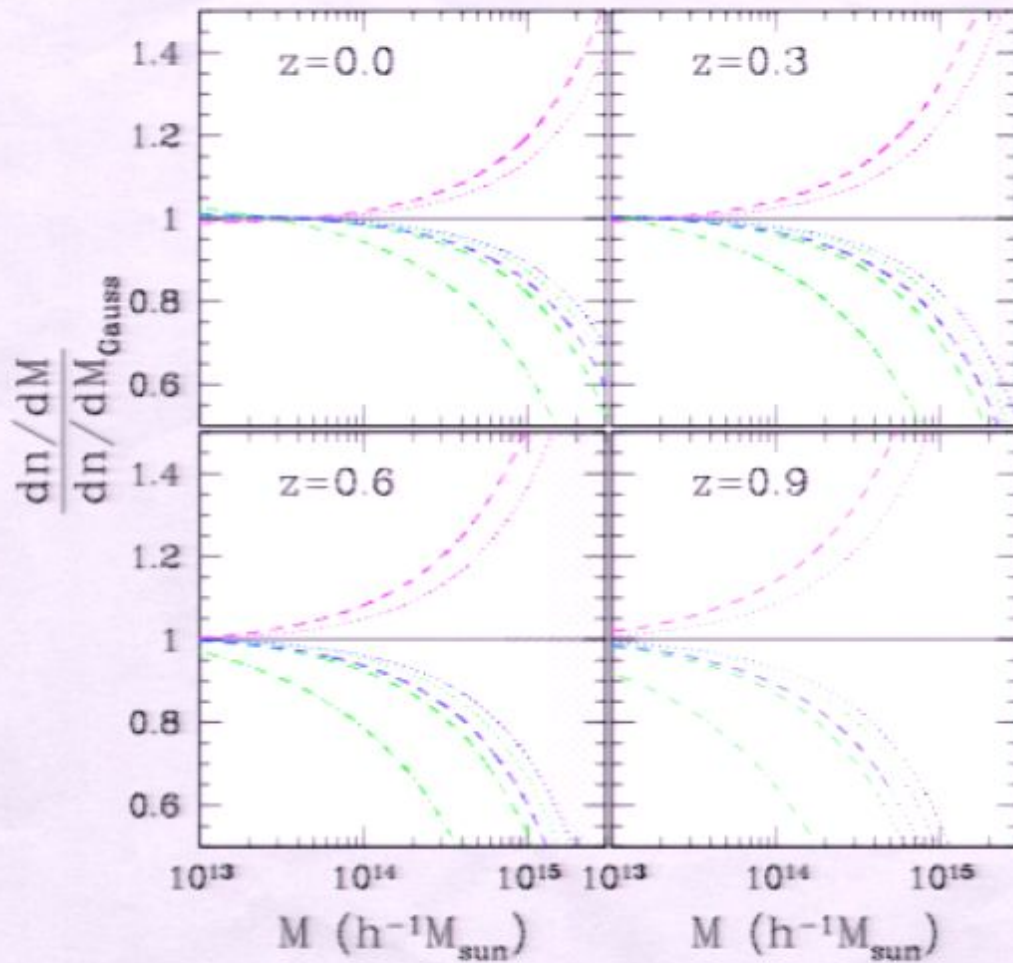
MASS FUNCTION



MASS FUNCTION

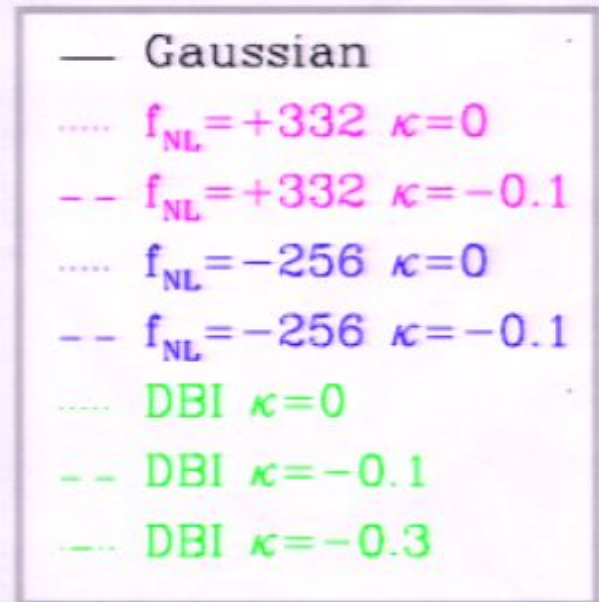
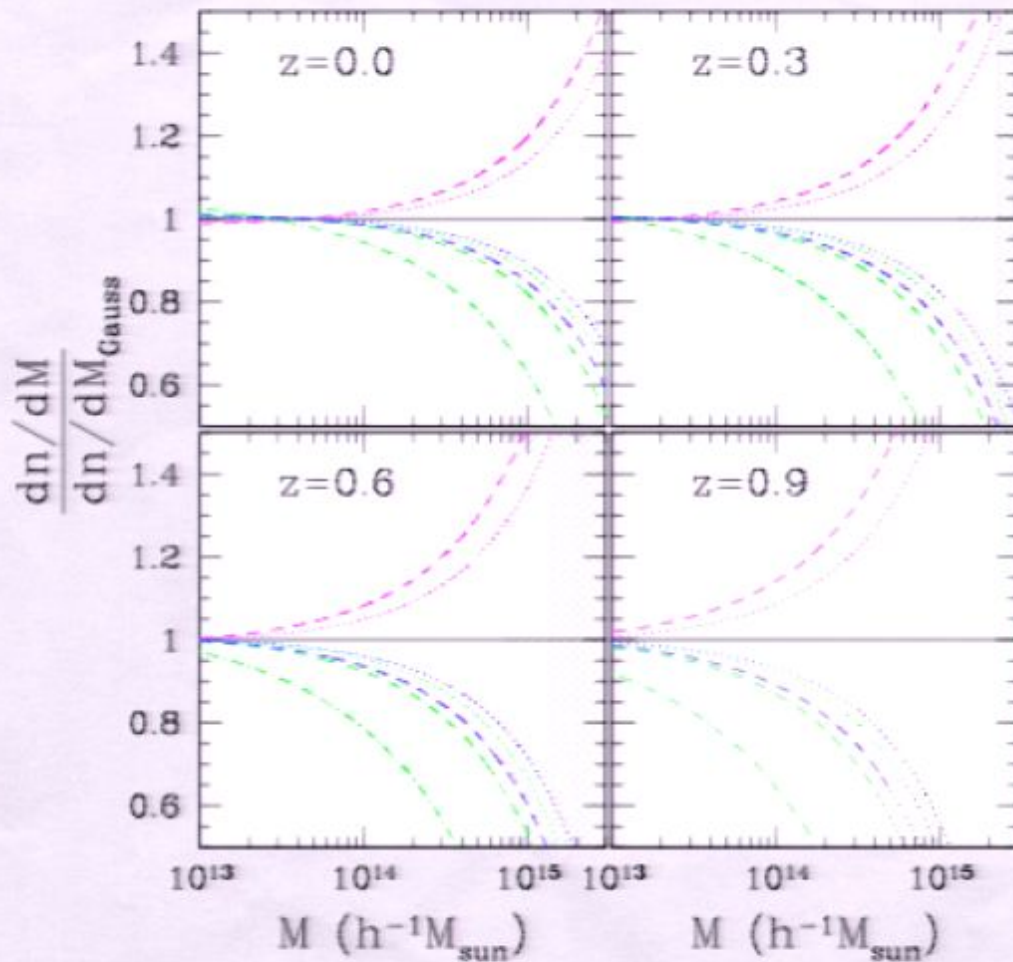


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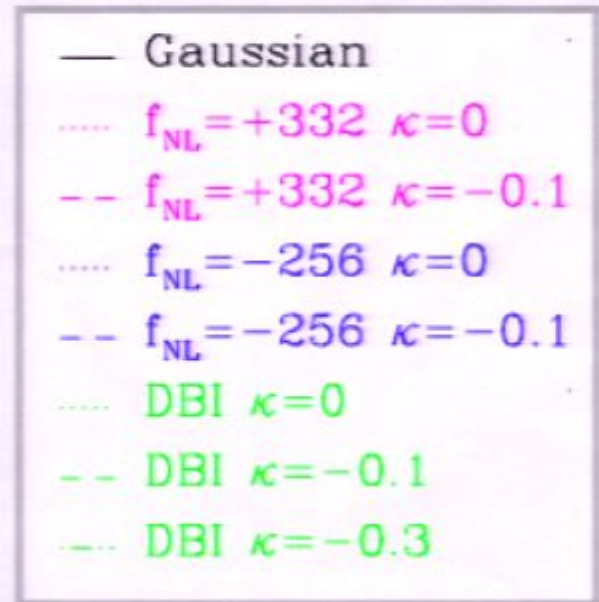
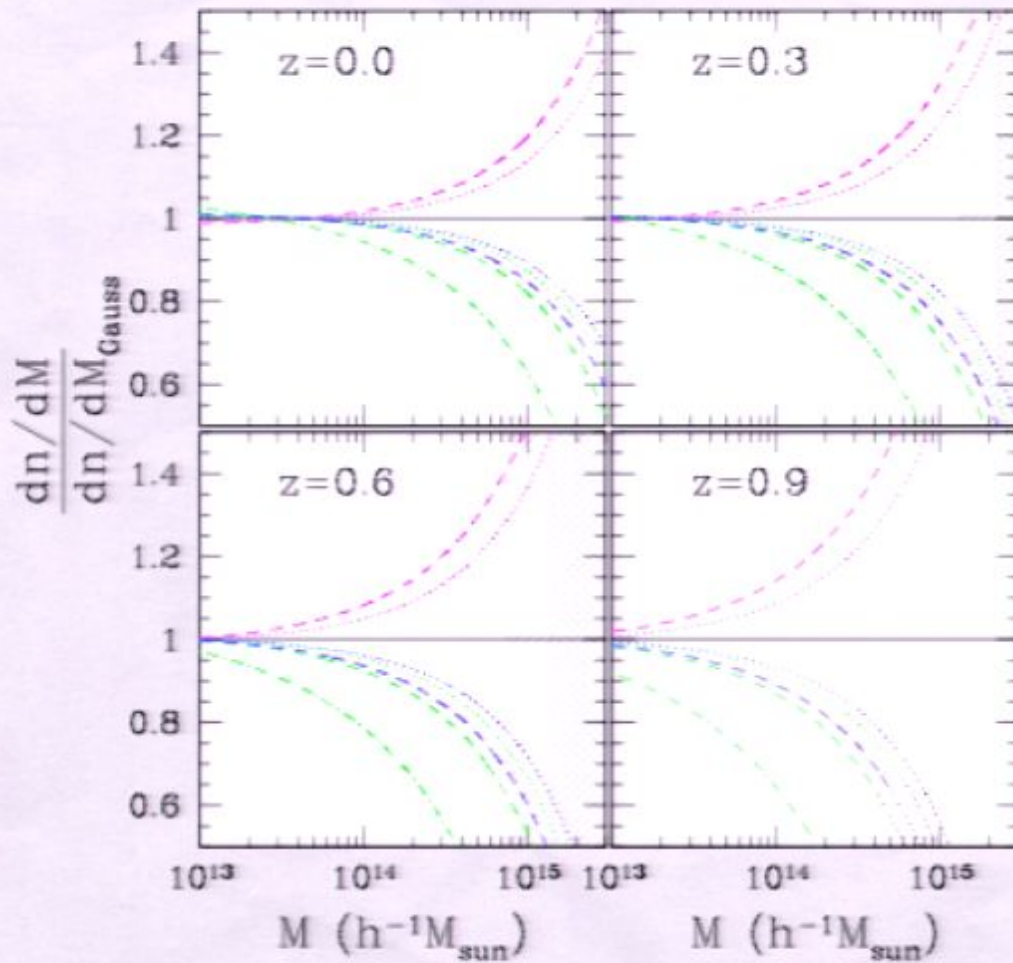


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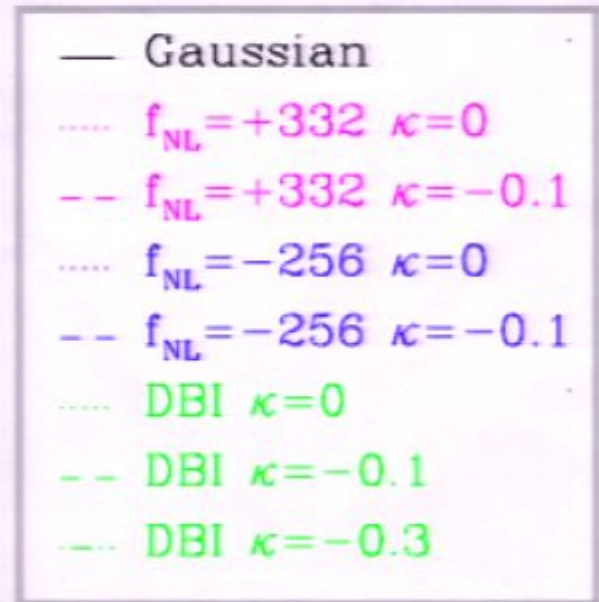
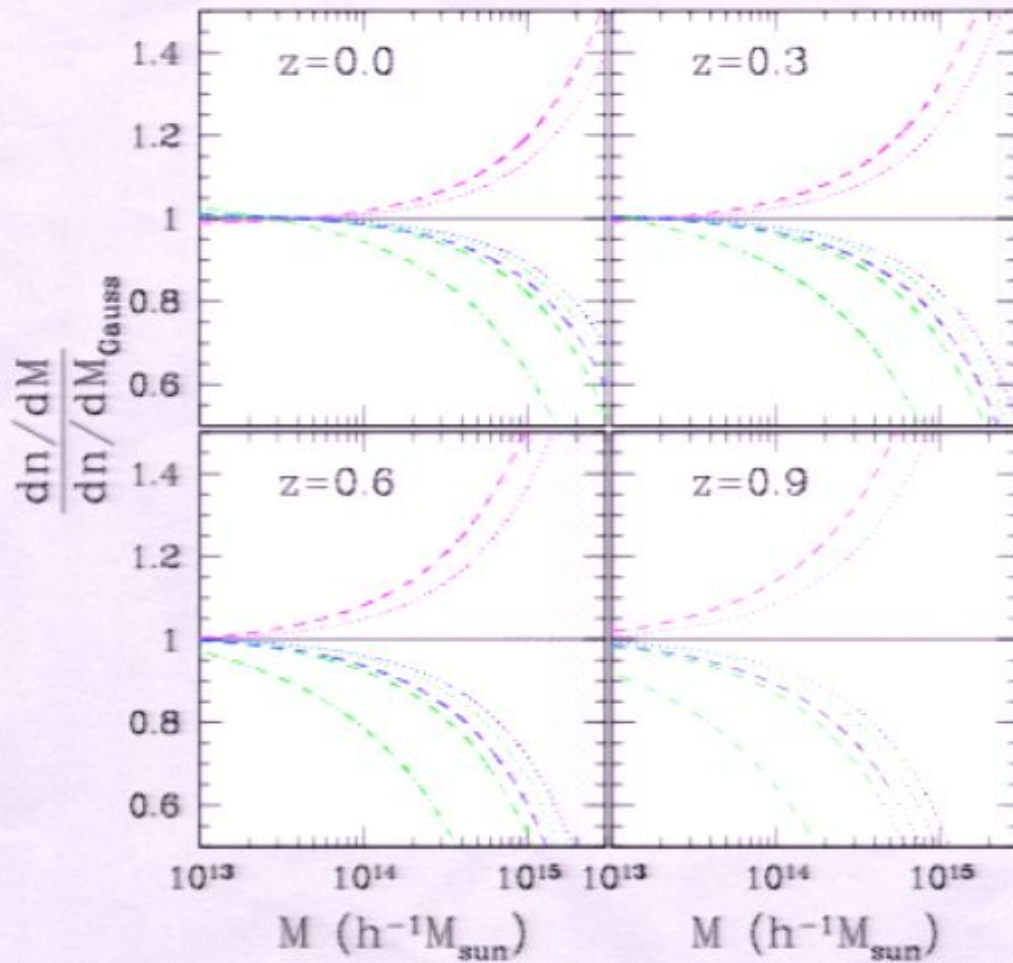
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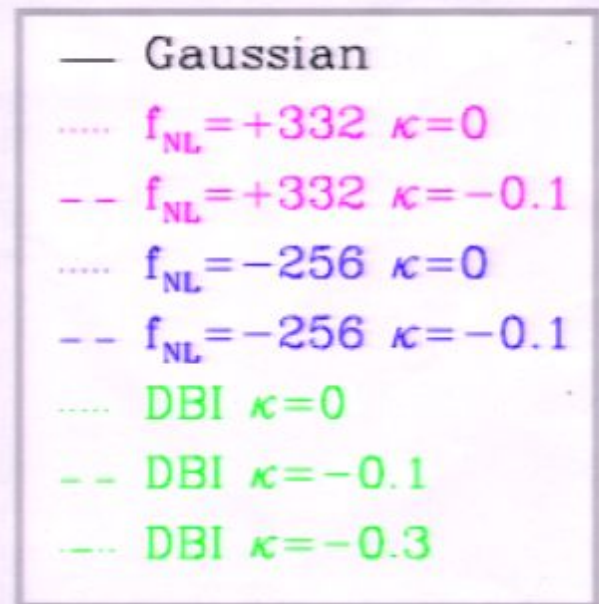
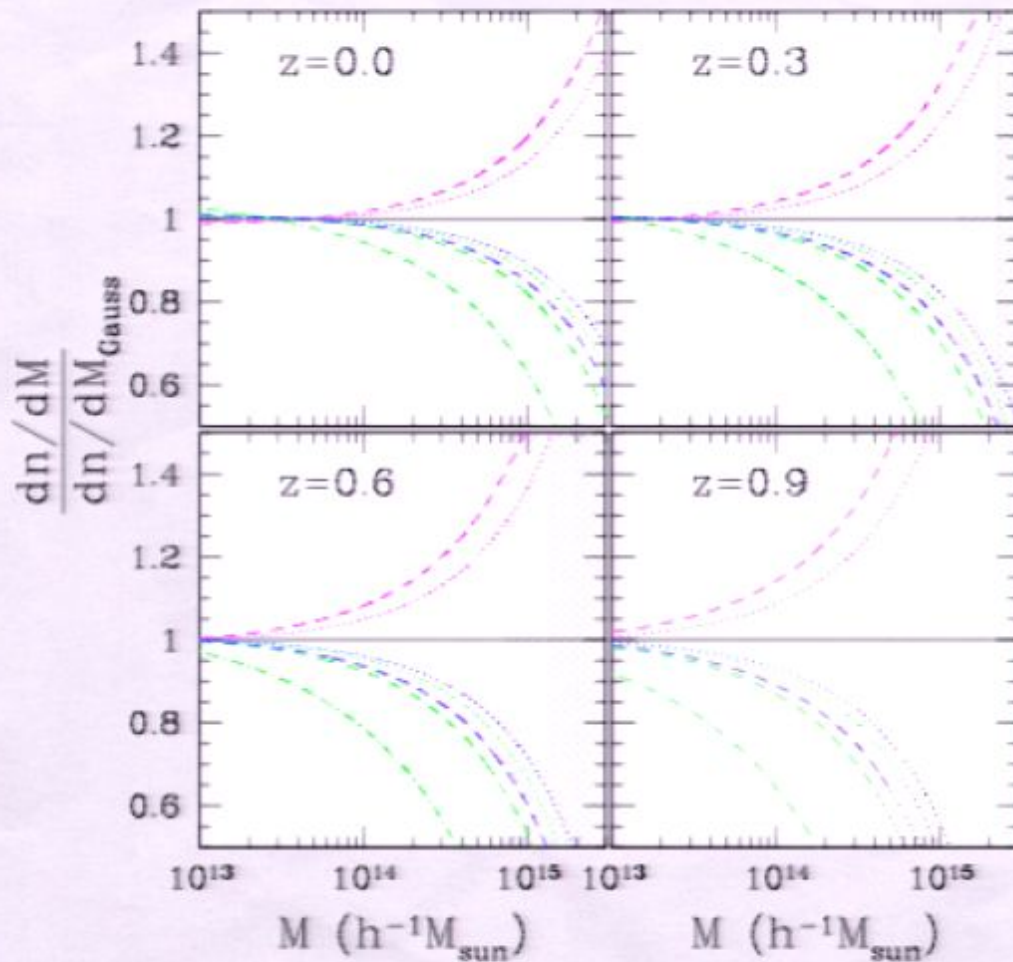
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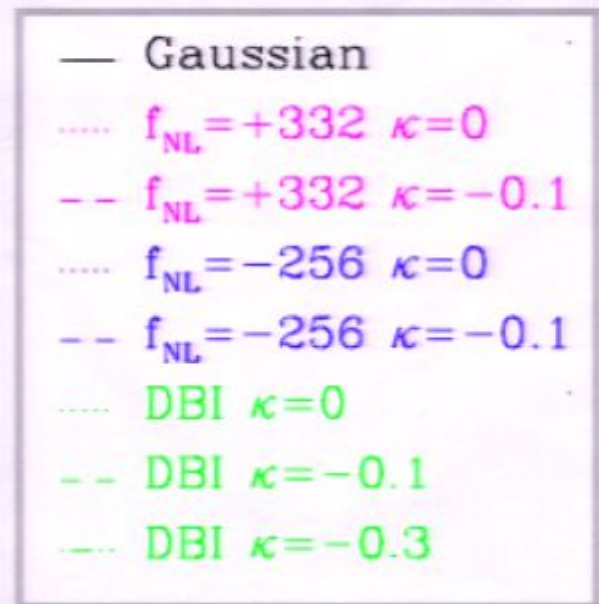
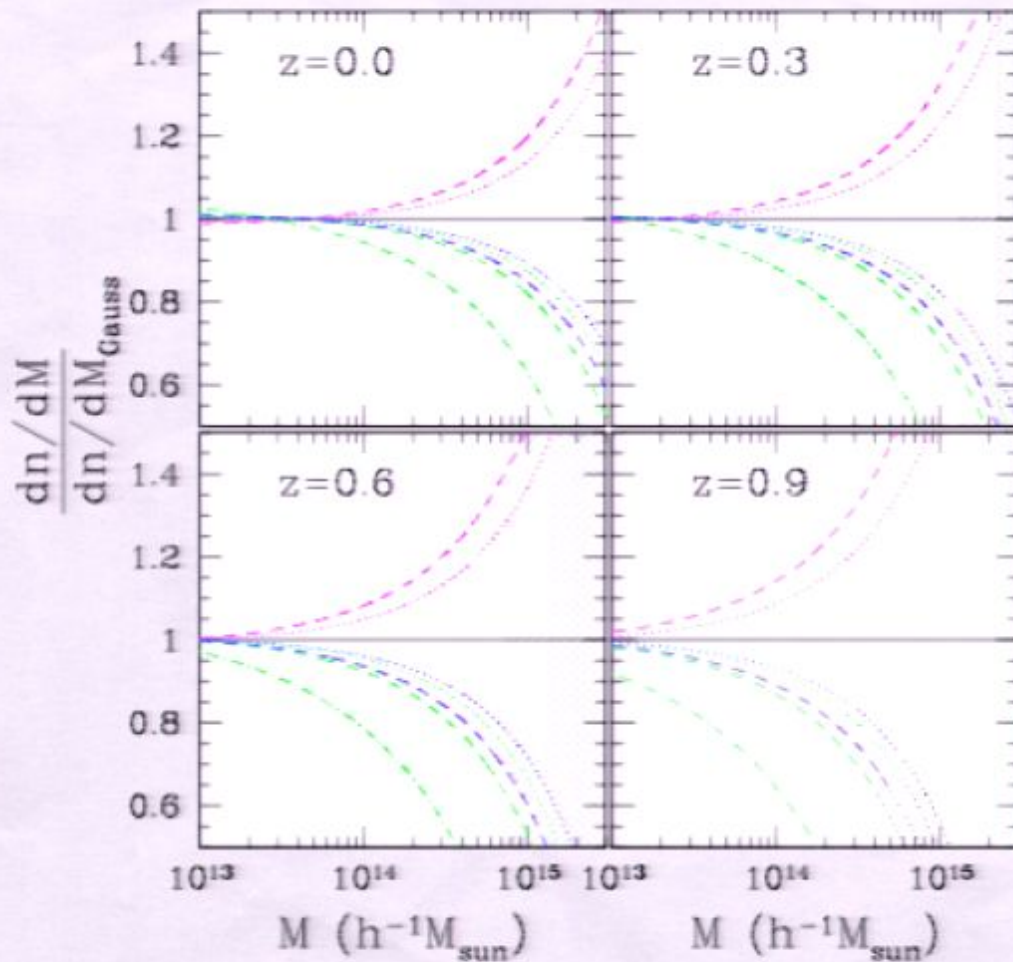
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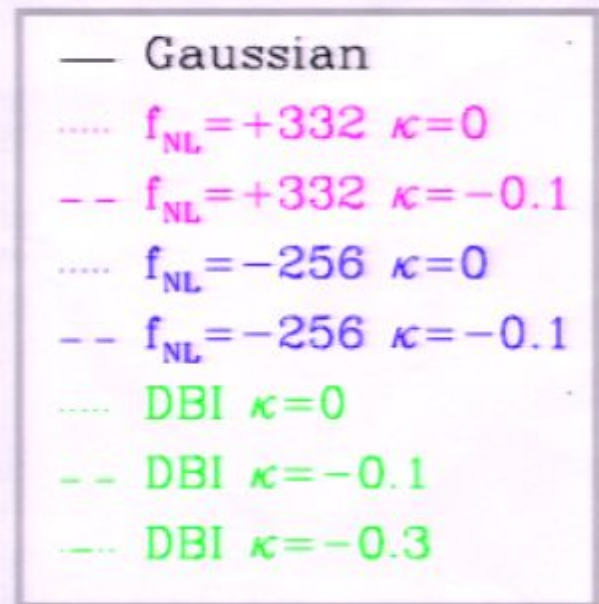
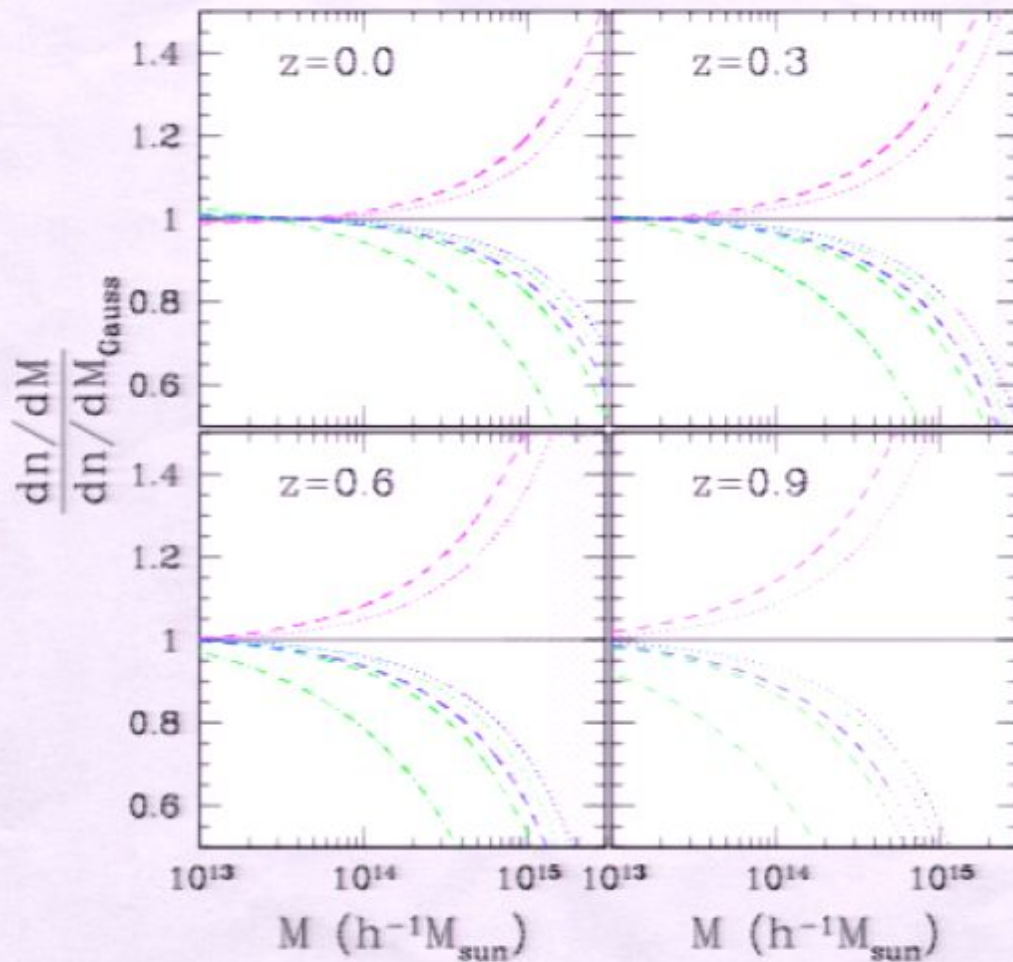
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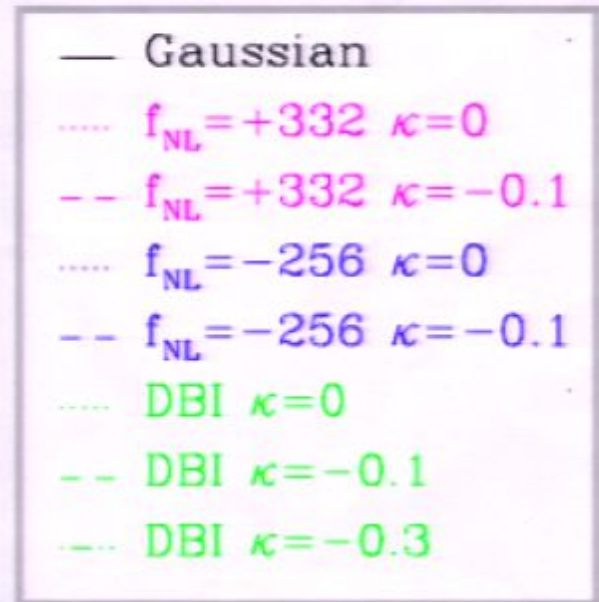
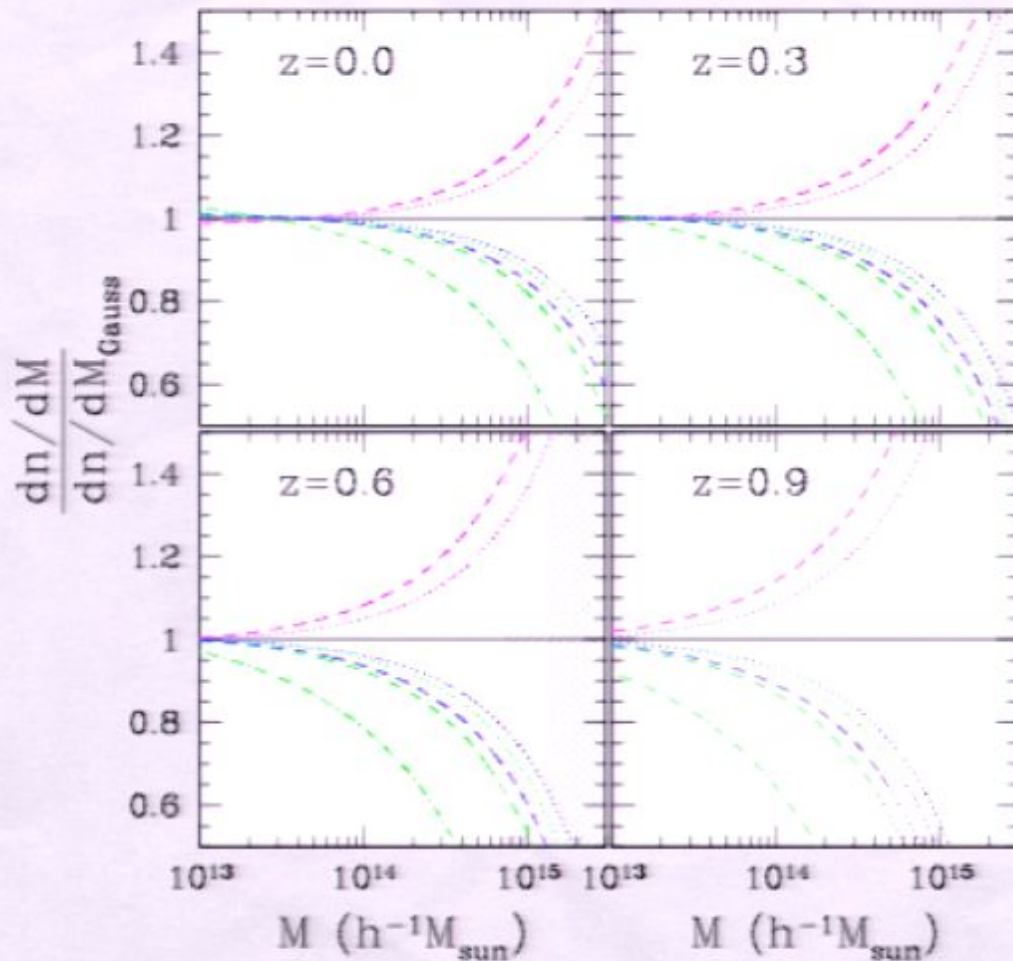
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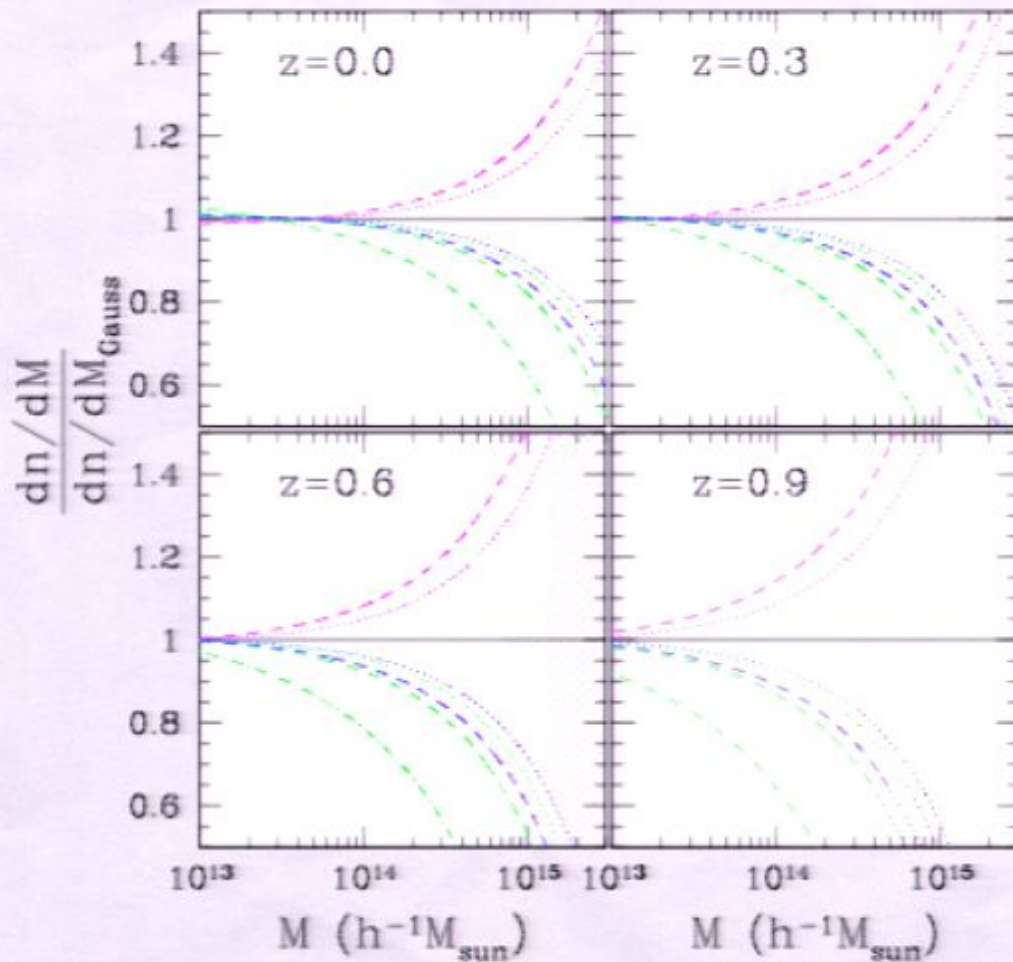
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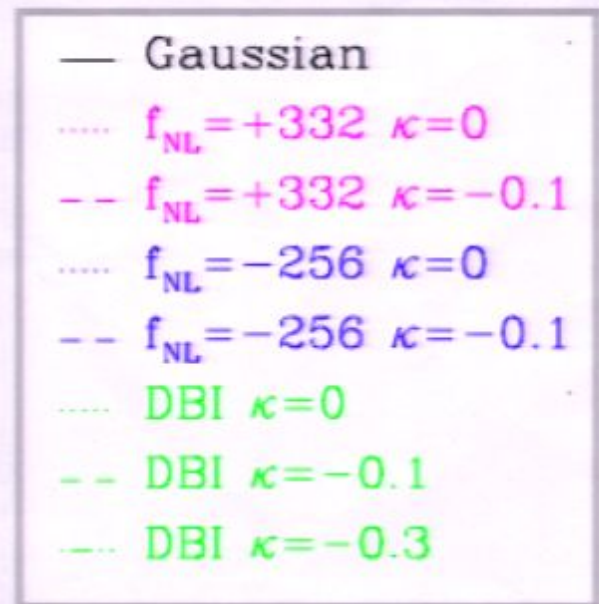
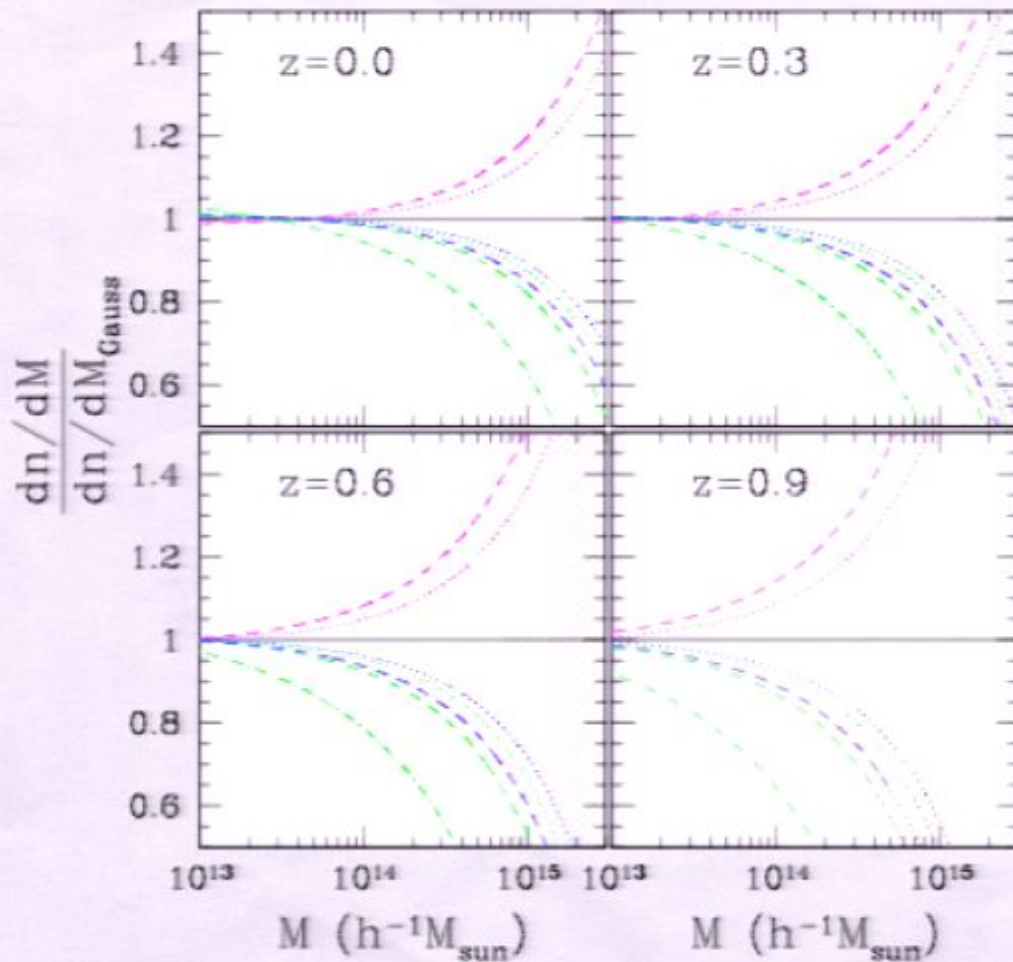
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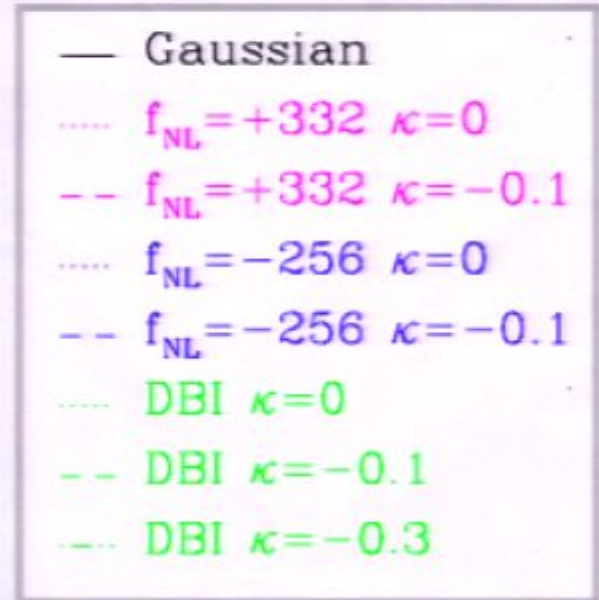
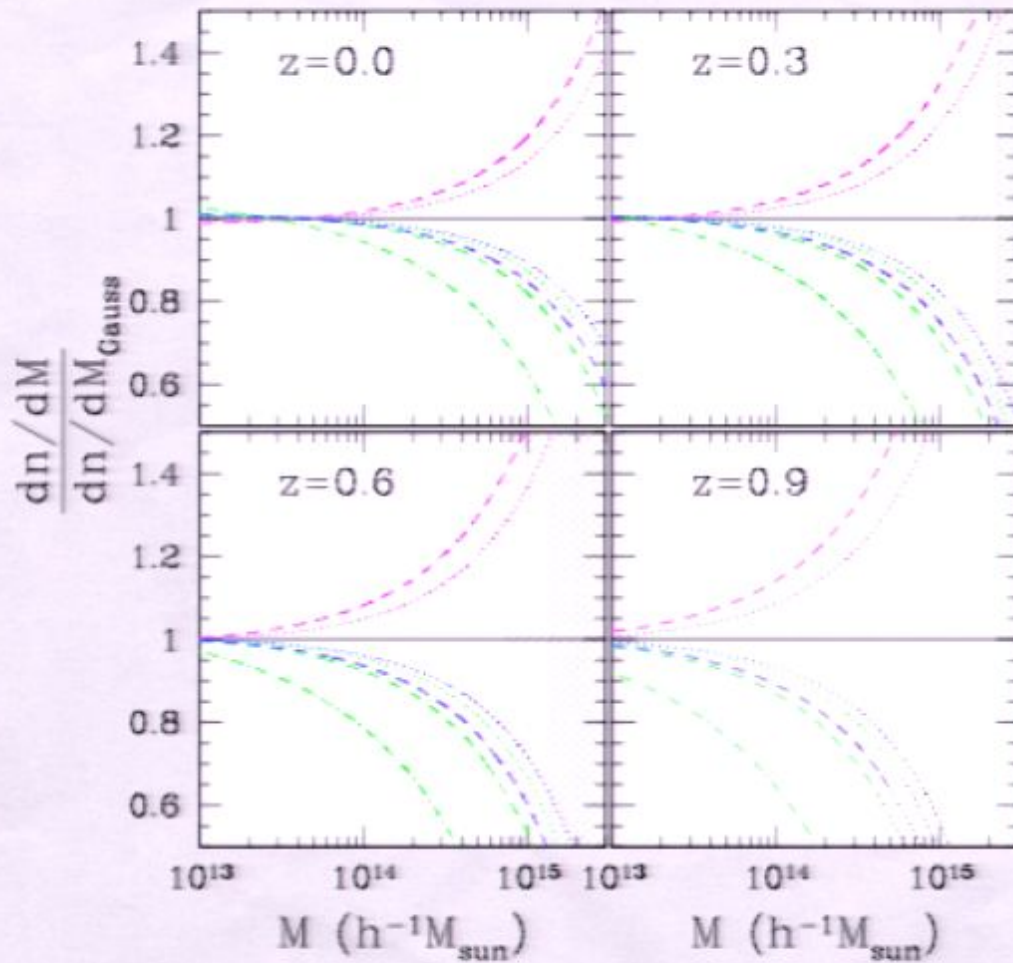
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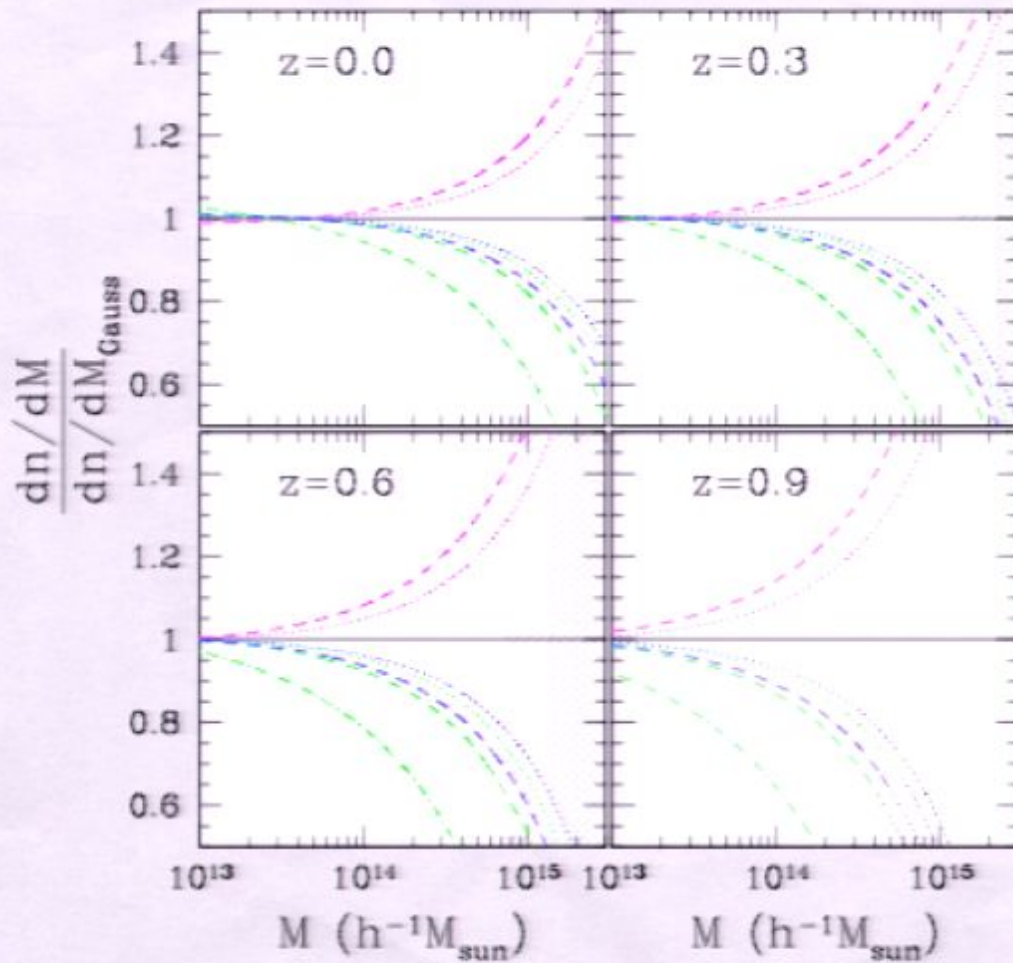
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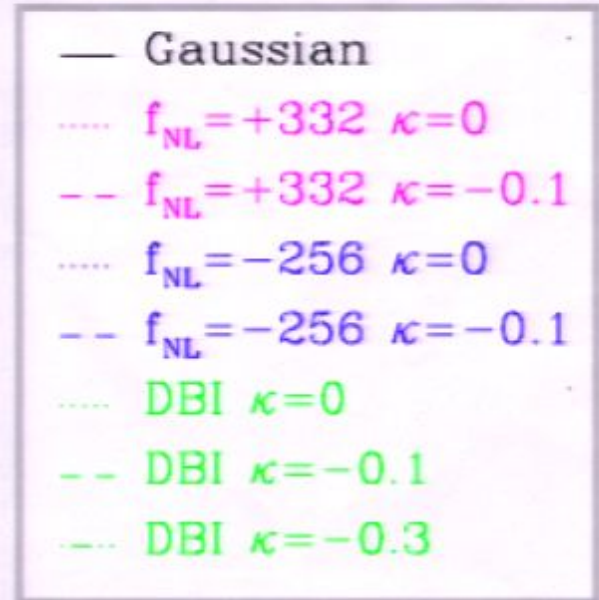
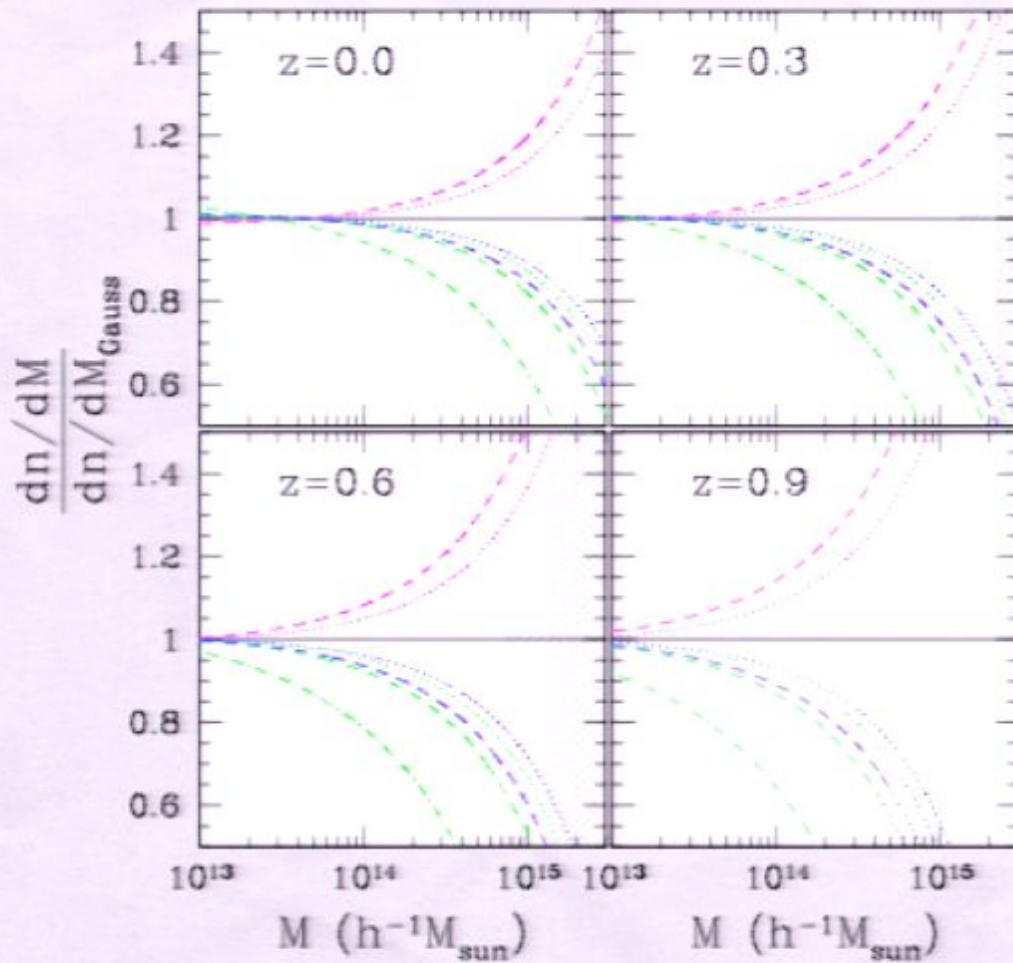


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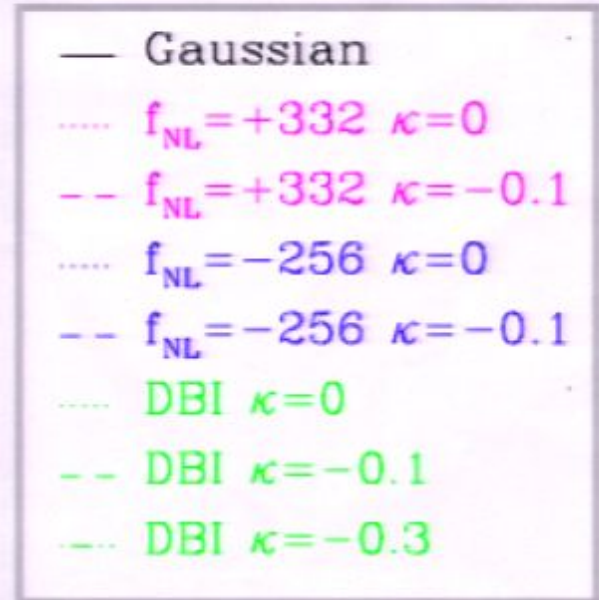
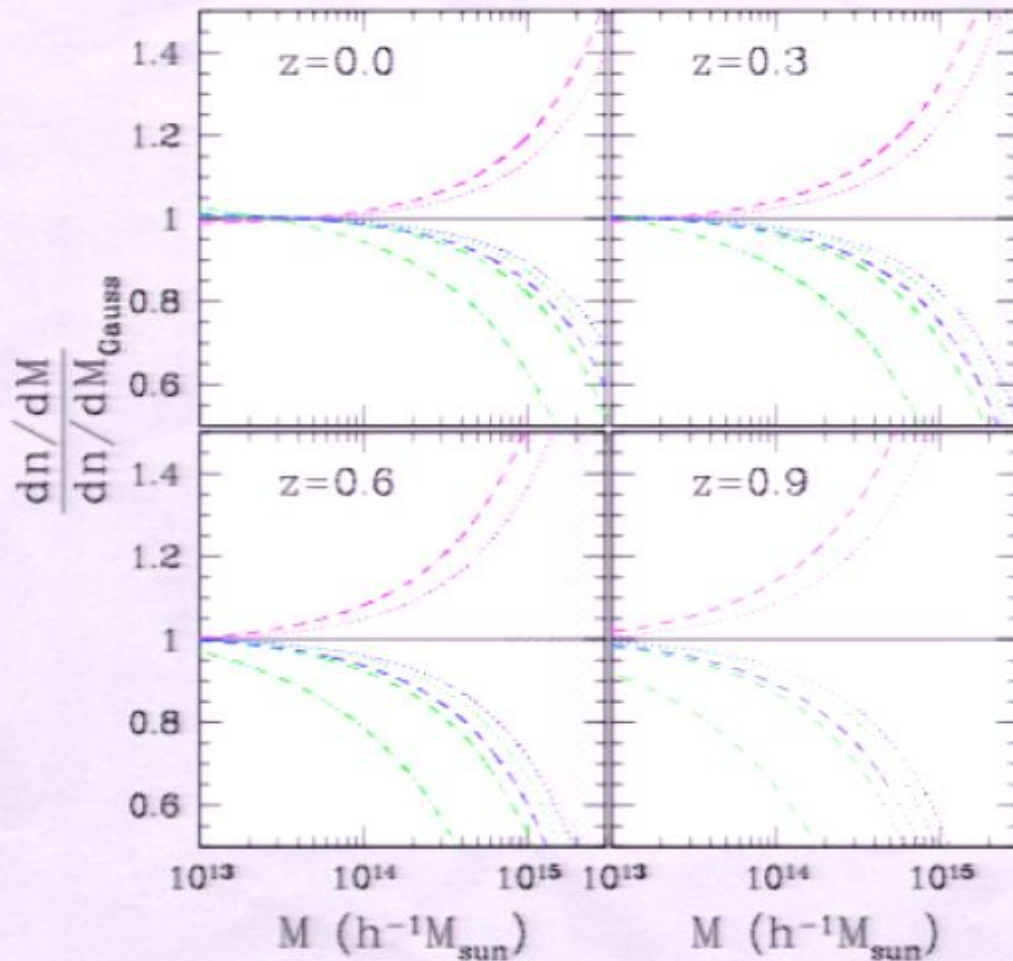


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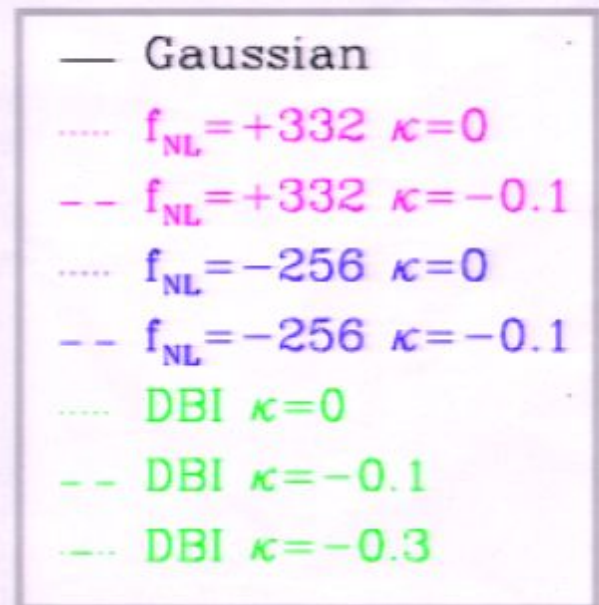
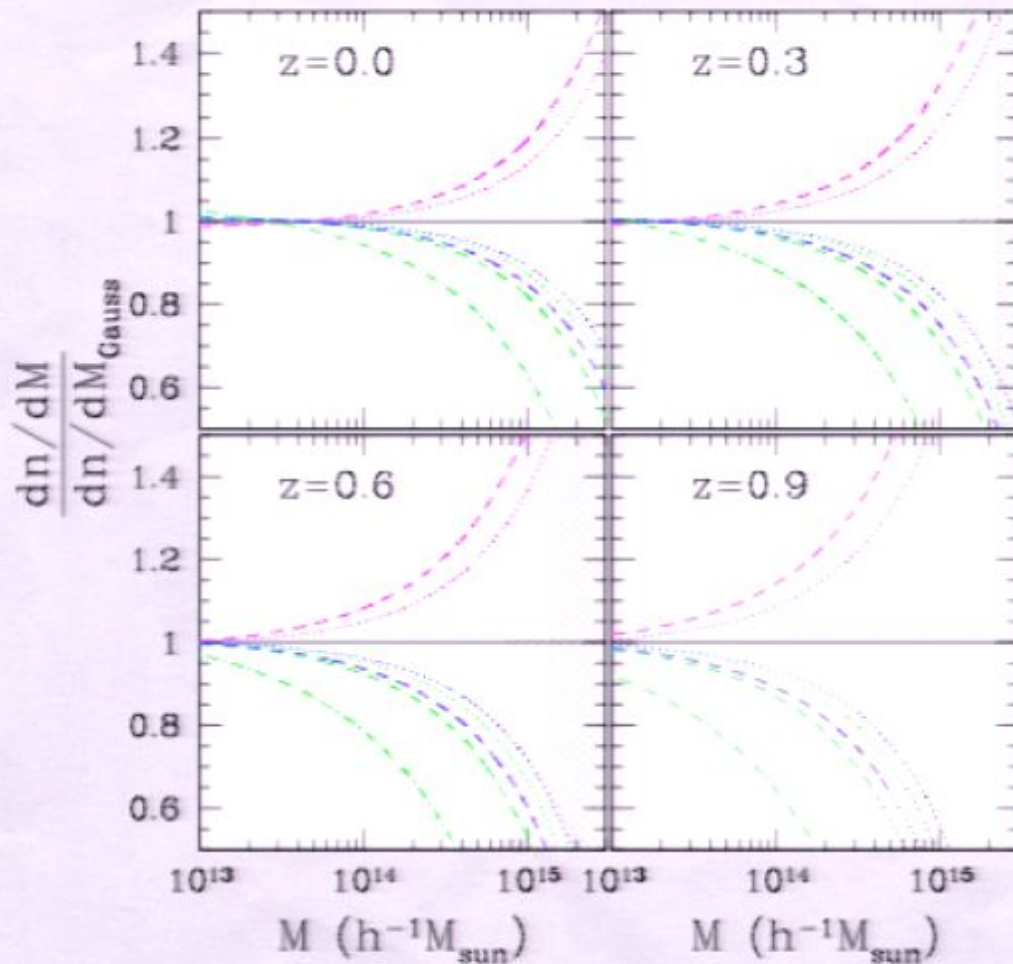
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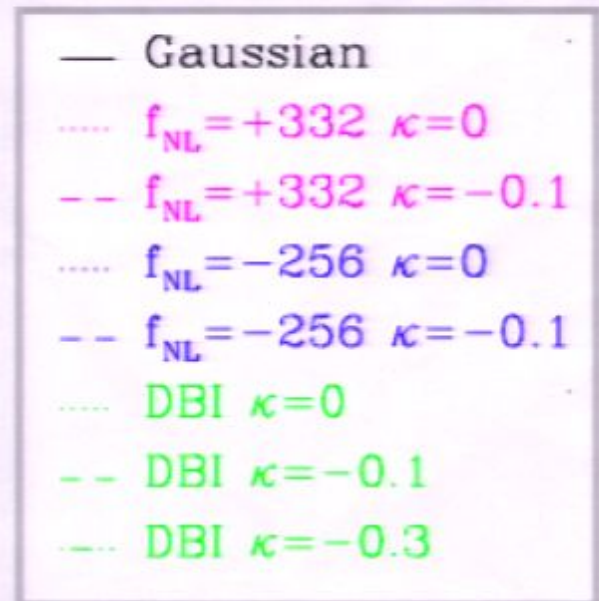
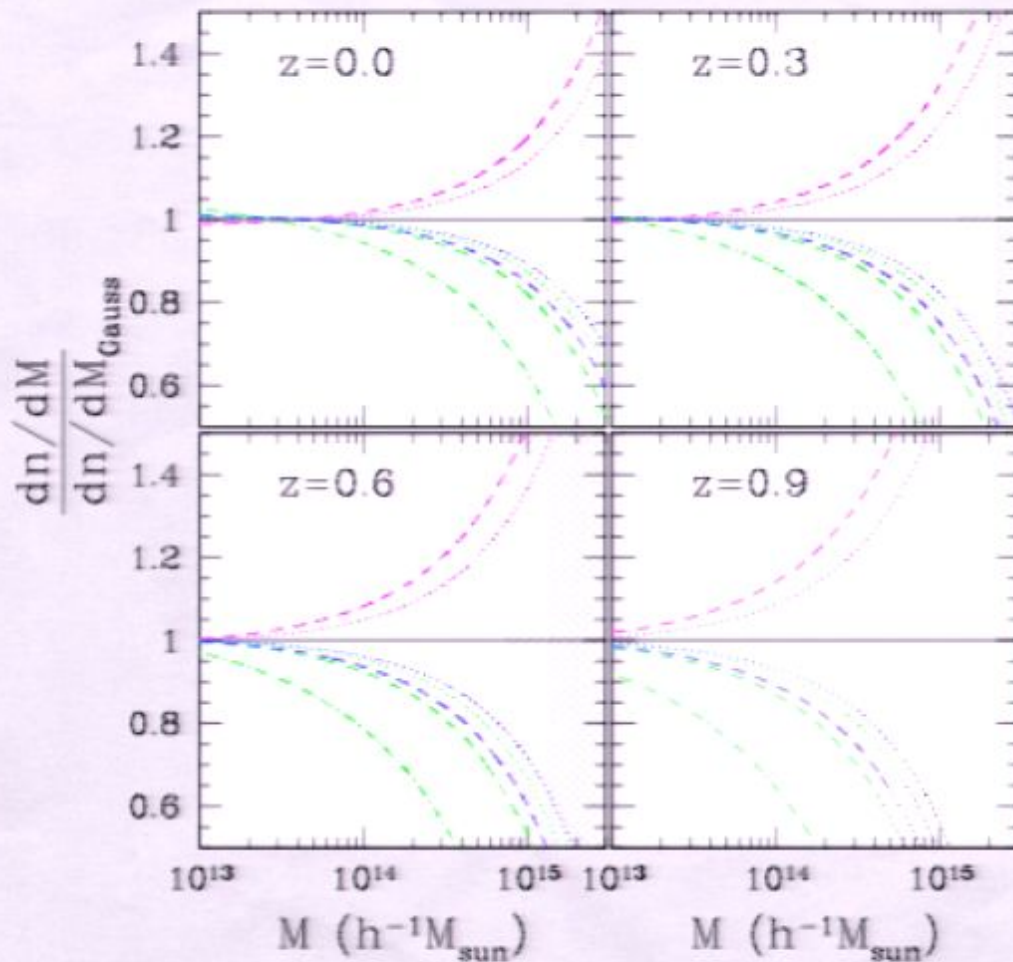
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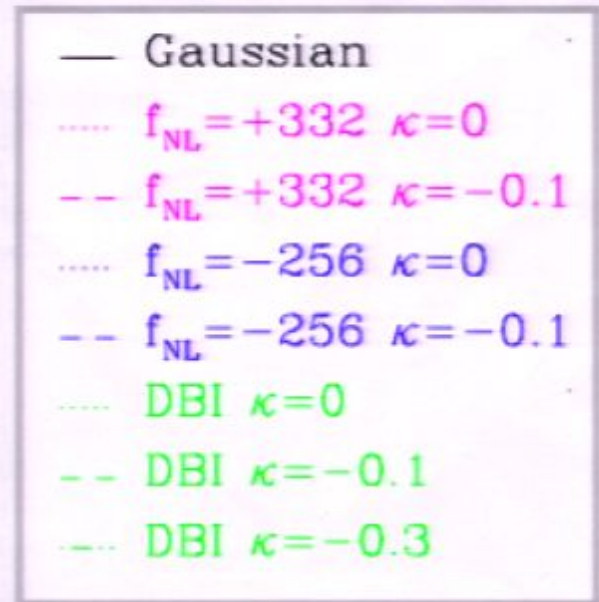
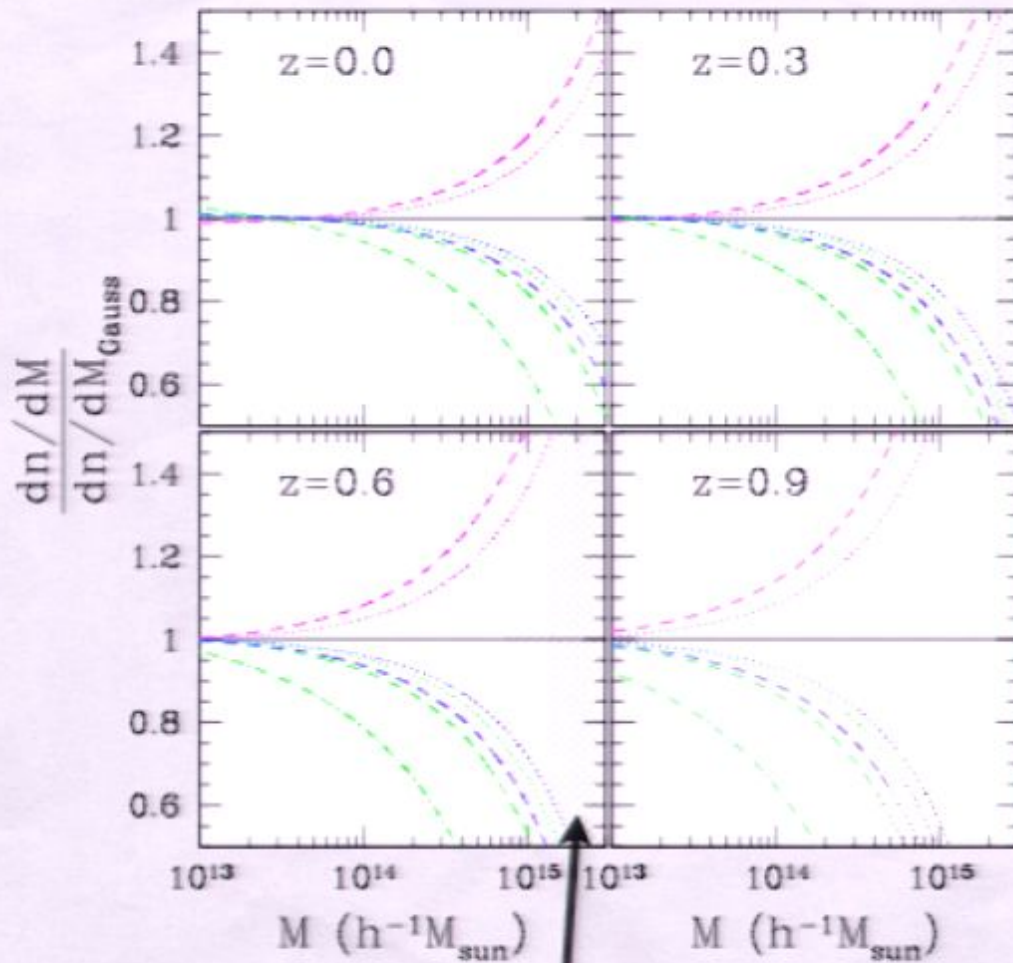
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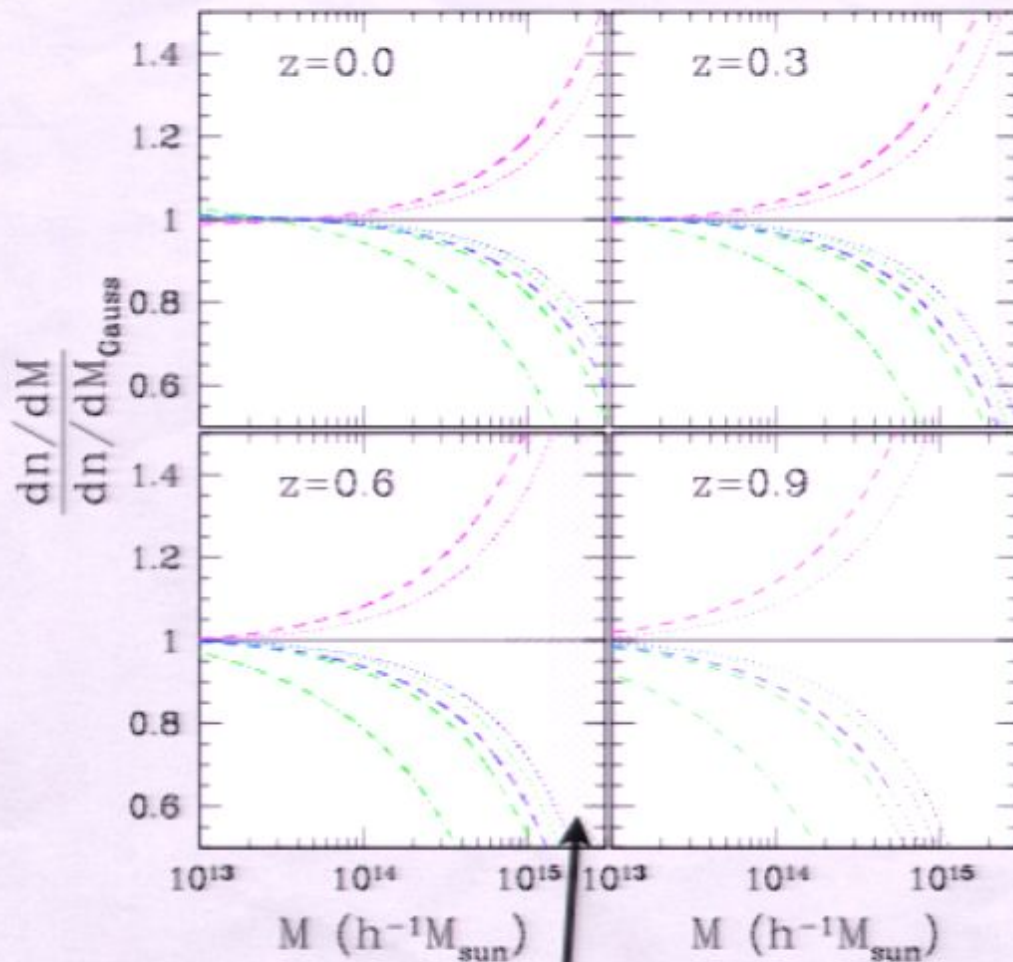
MASS FUNCTION



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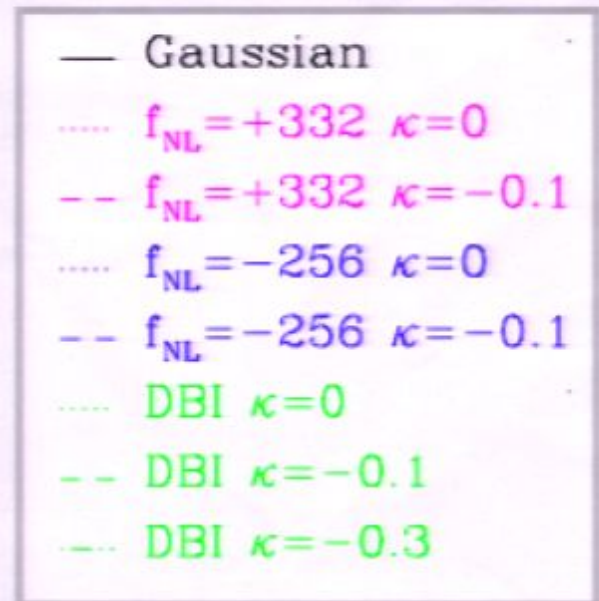
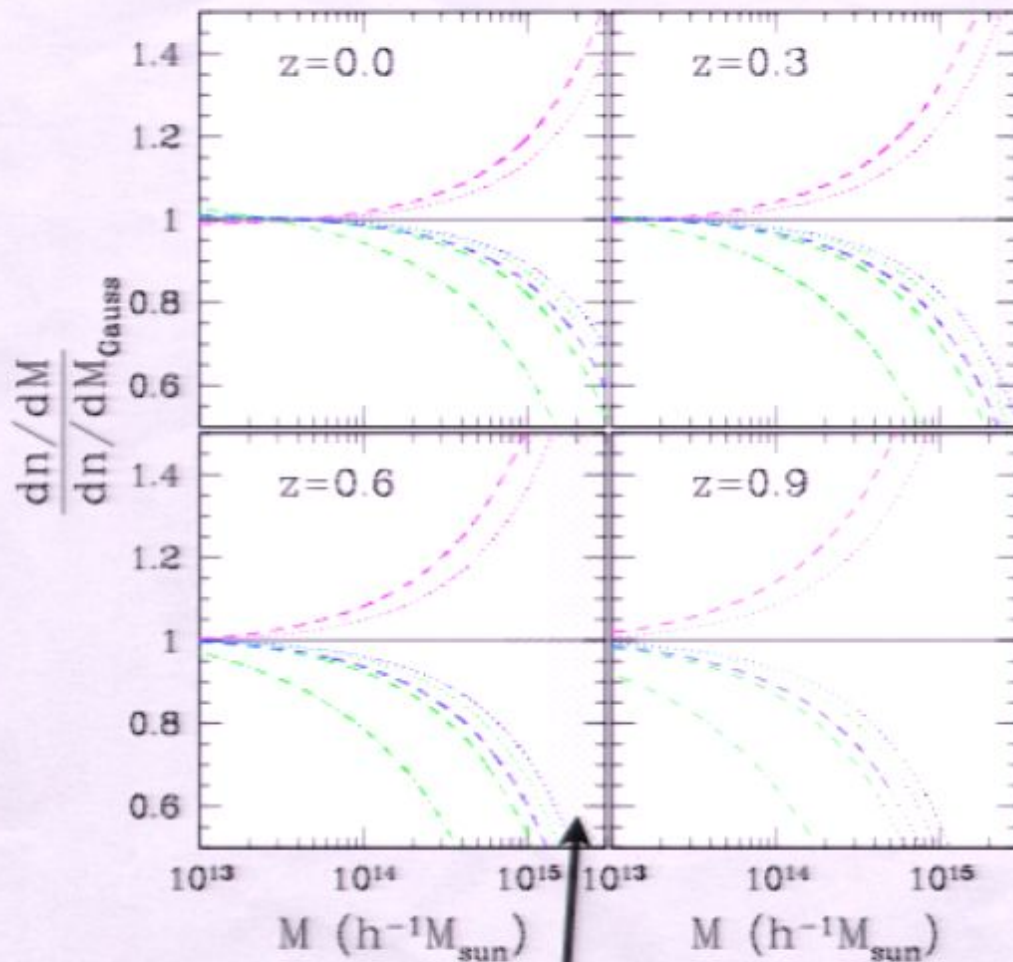


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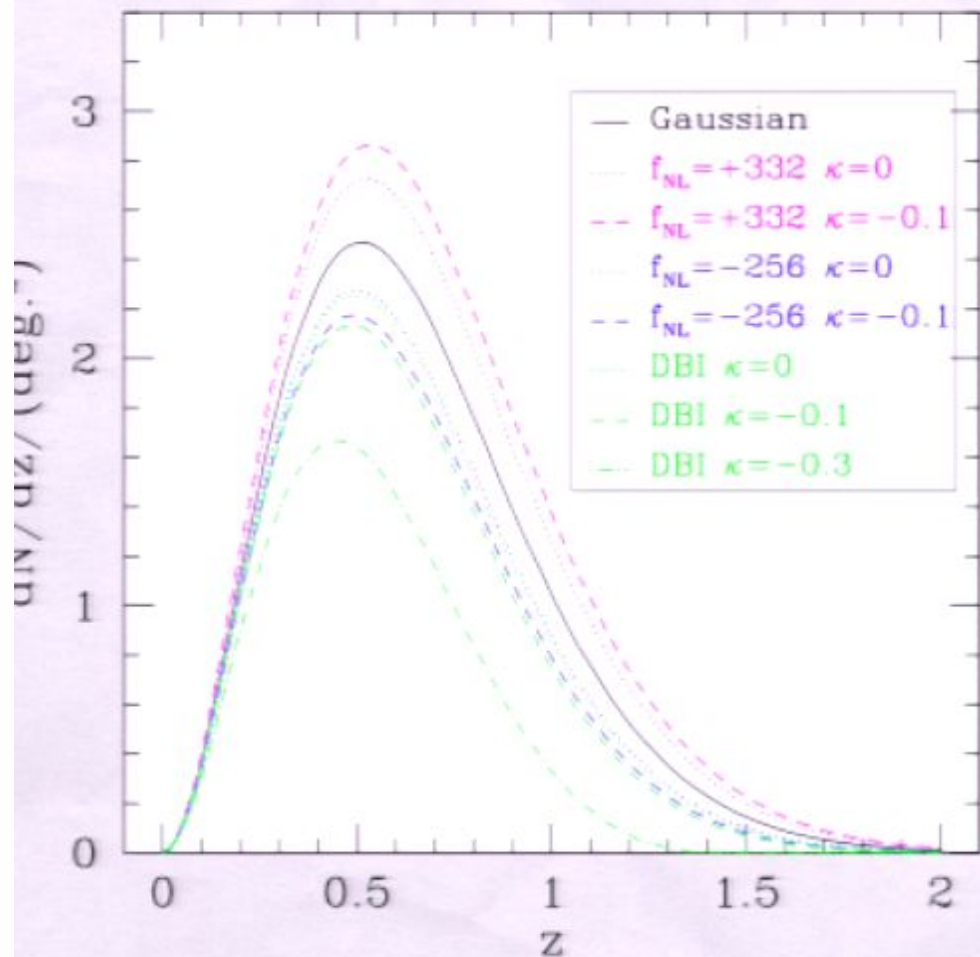


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MASS FUNCTION



NUMBER OF MASSIVE CLUSTERS AT REDSHIFT Z

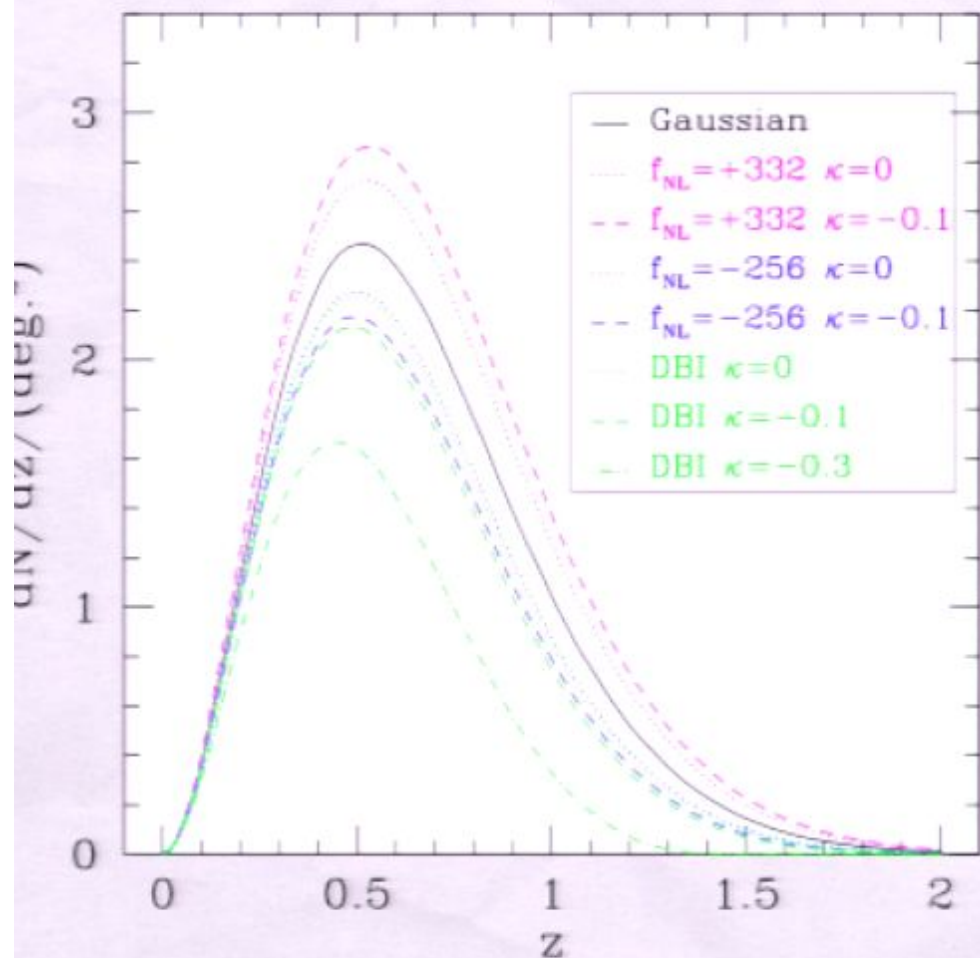


$$\frac{dN}{dz}(M > M_{lim}) = f_{sky} \frac{dV(z)}{dz} \int_{M_{min}}^{\infty} dM \frac{dn}{dM}(M, z)$$

$$\frac{dV}{dz} = \frac{4\pi}{H(z)} \left[\int_0^z \frac{dz'}{H(z')} \right]^2$$

$$M_{lim} = 1.75 h^{-1} \times 10^{14} M_{sun} \quad (\text{SPT})$$

NUMBER OF MASSIVE CLUSTERS AT REDSHIFT Z

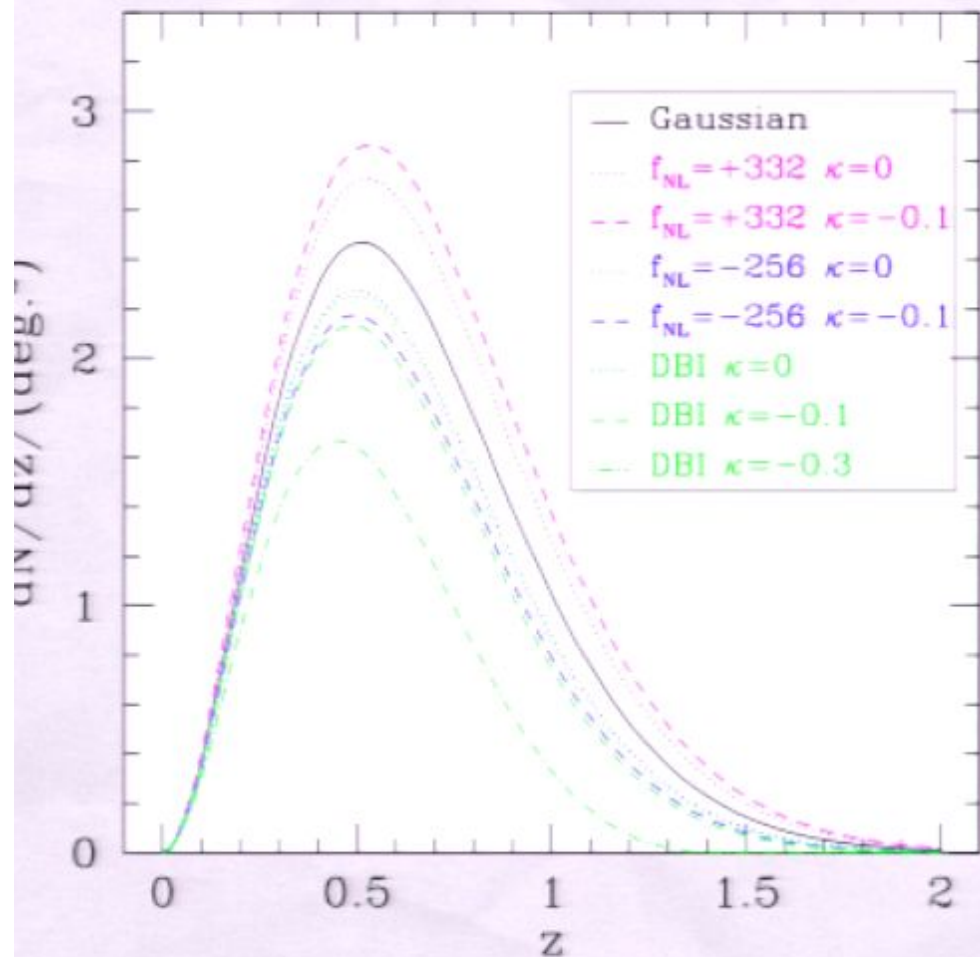


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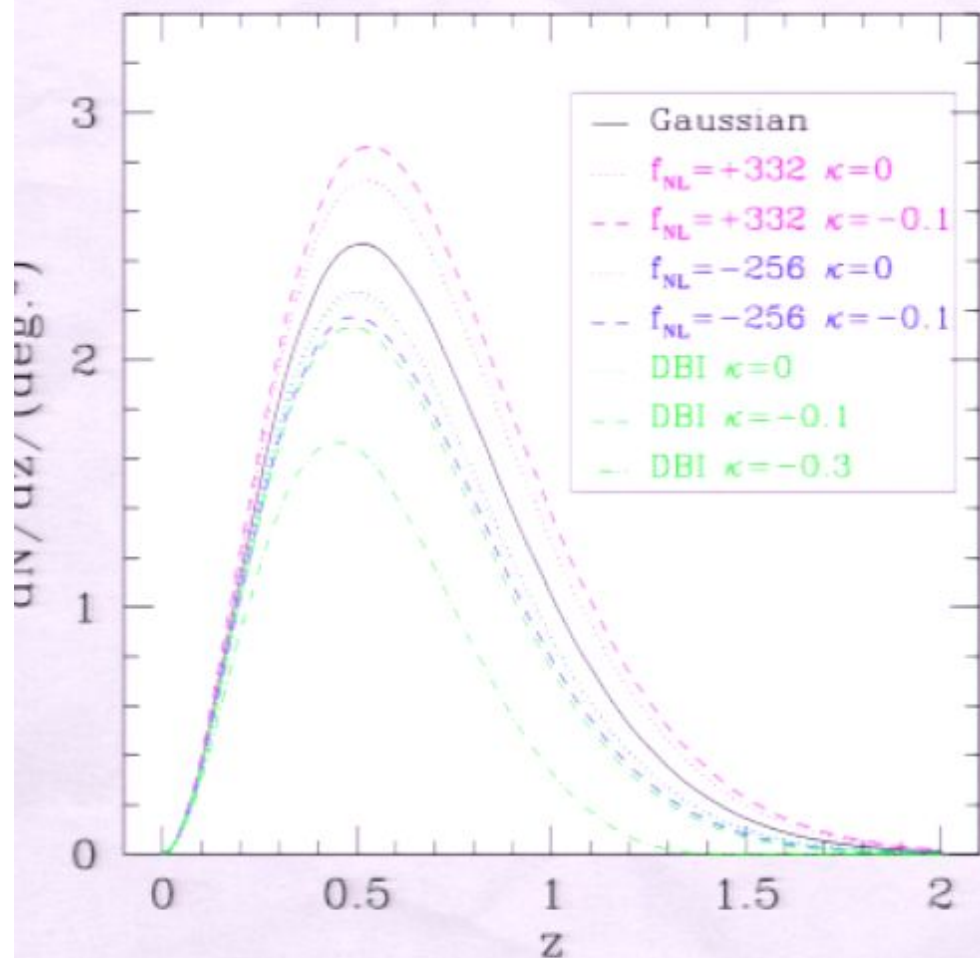


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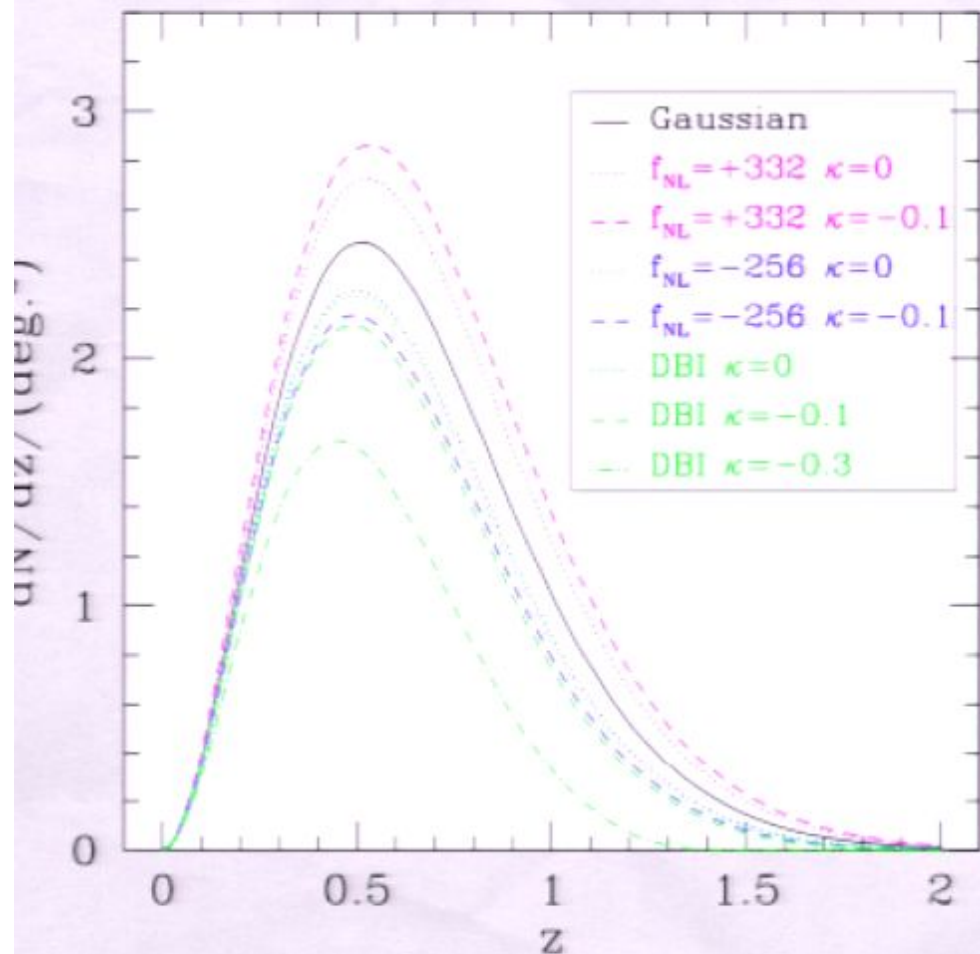


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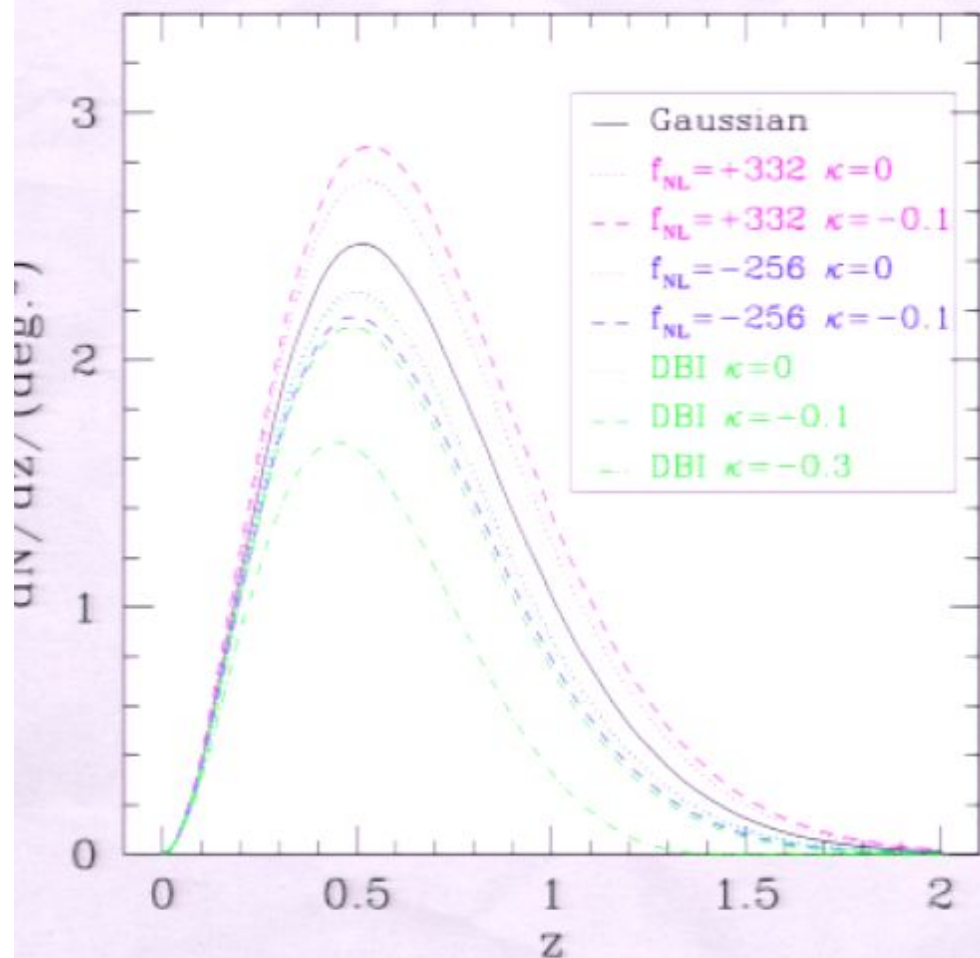


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NUMBER OF MASSIVE CLUSTERS AT REDSHIFT Z

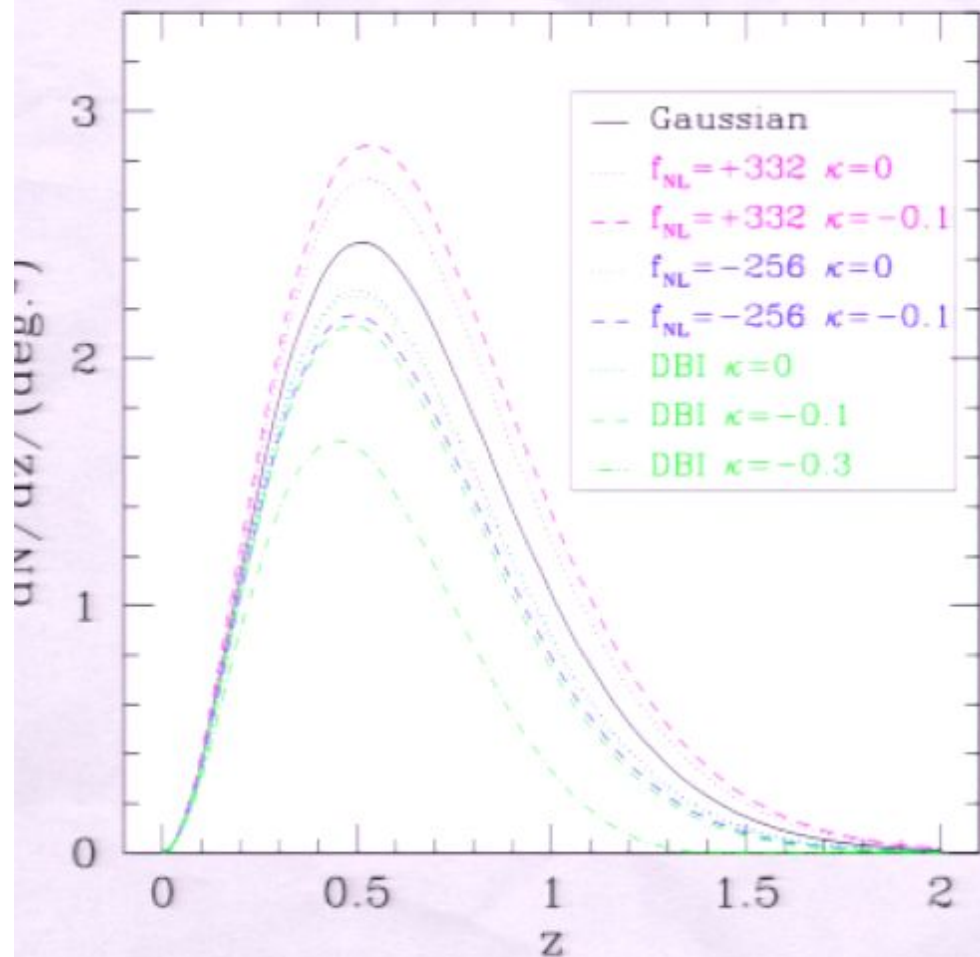


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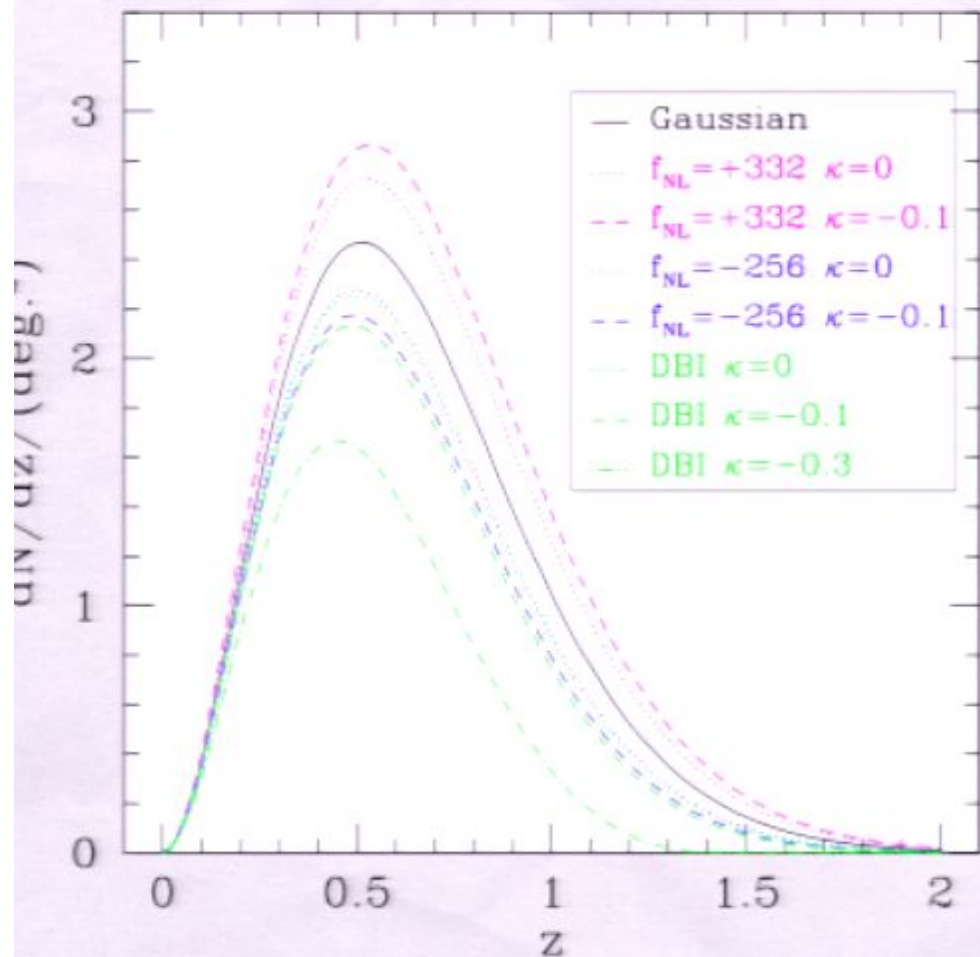


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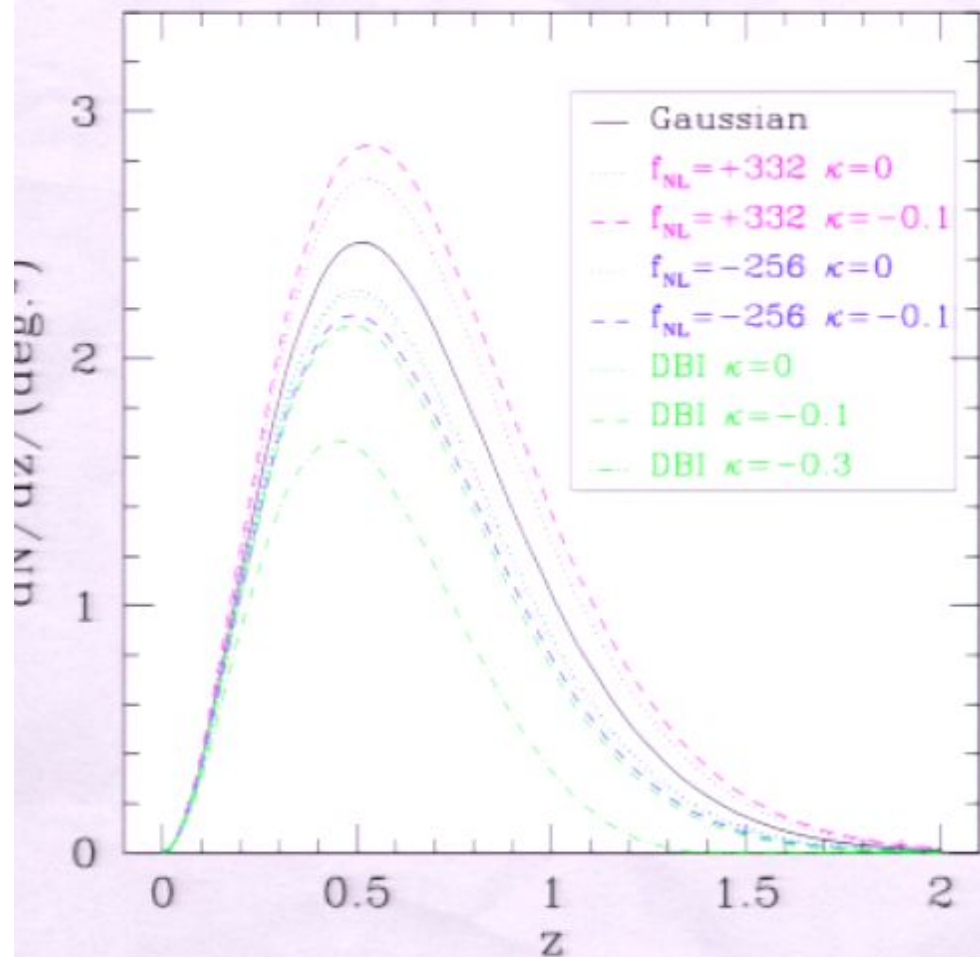


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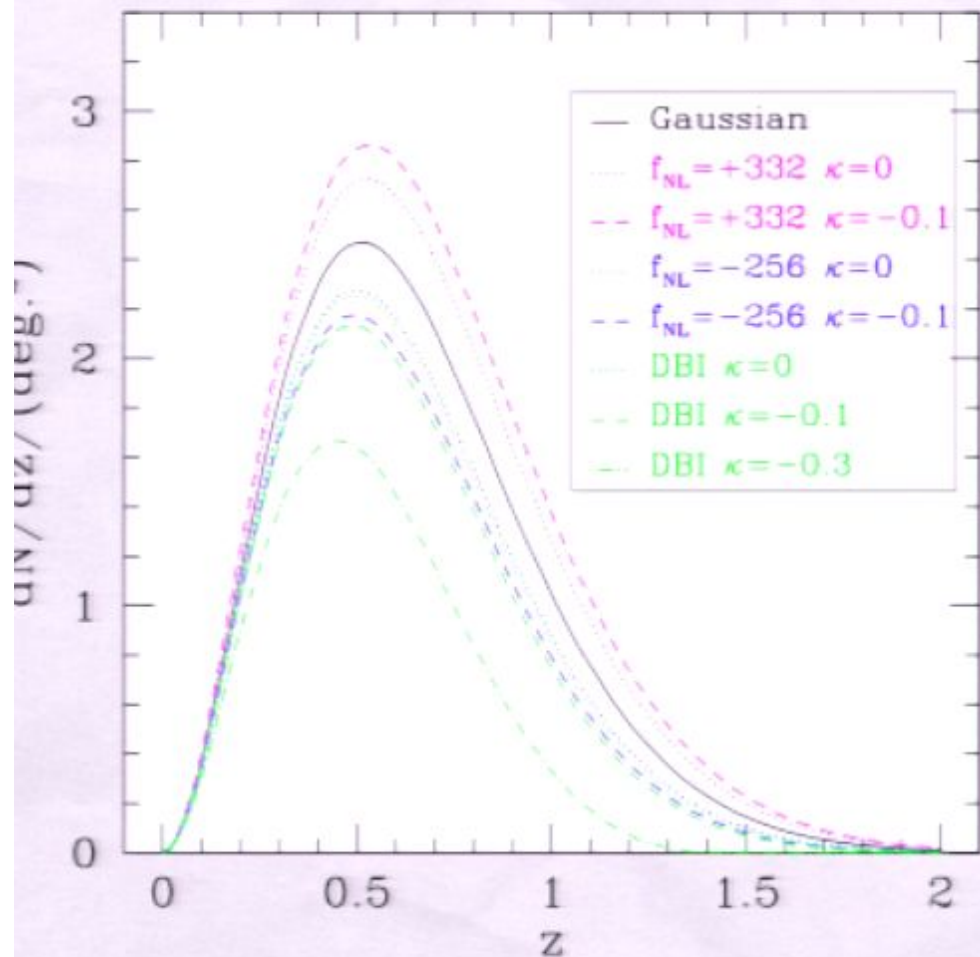


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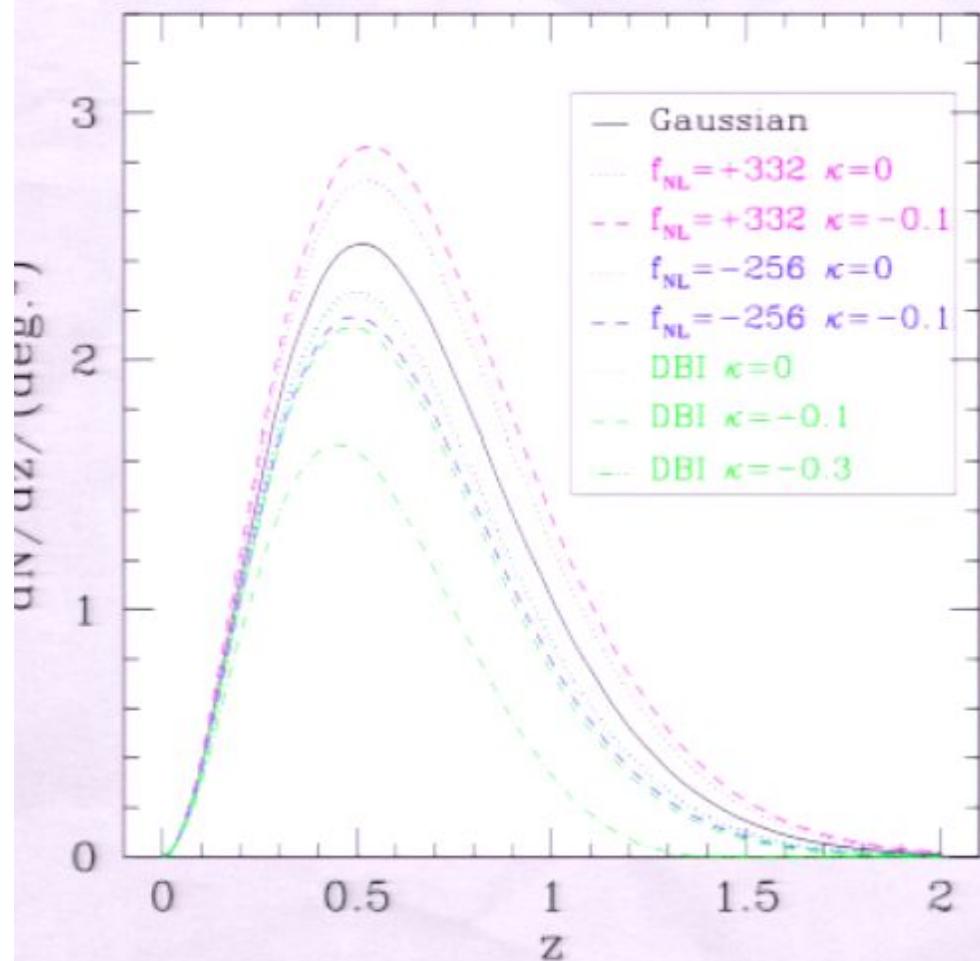


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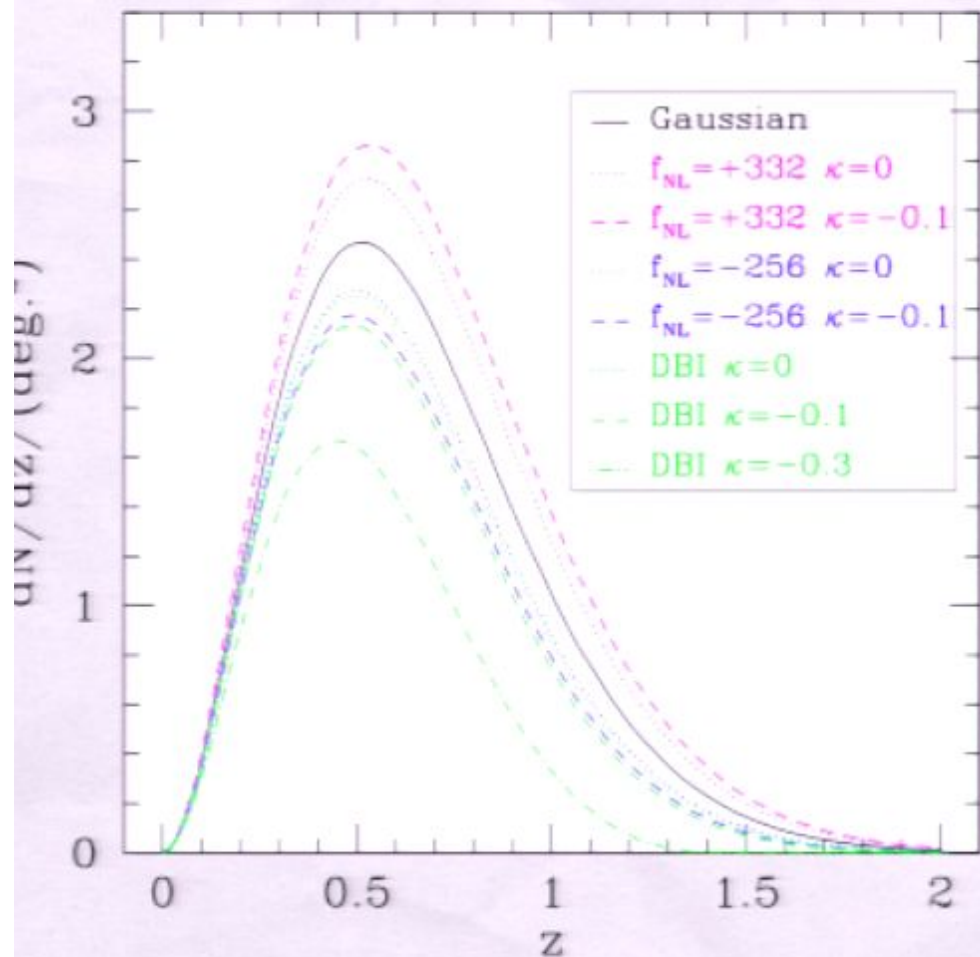


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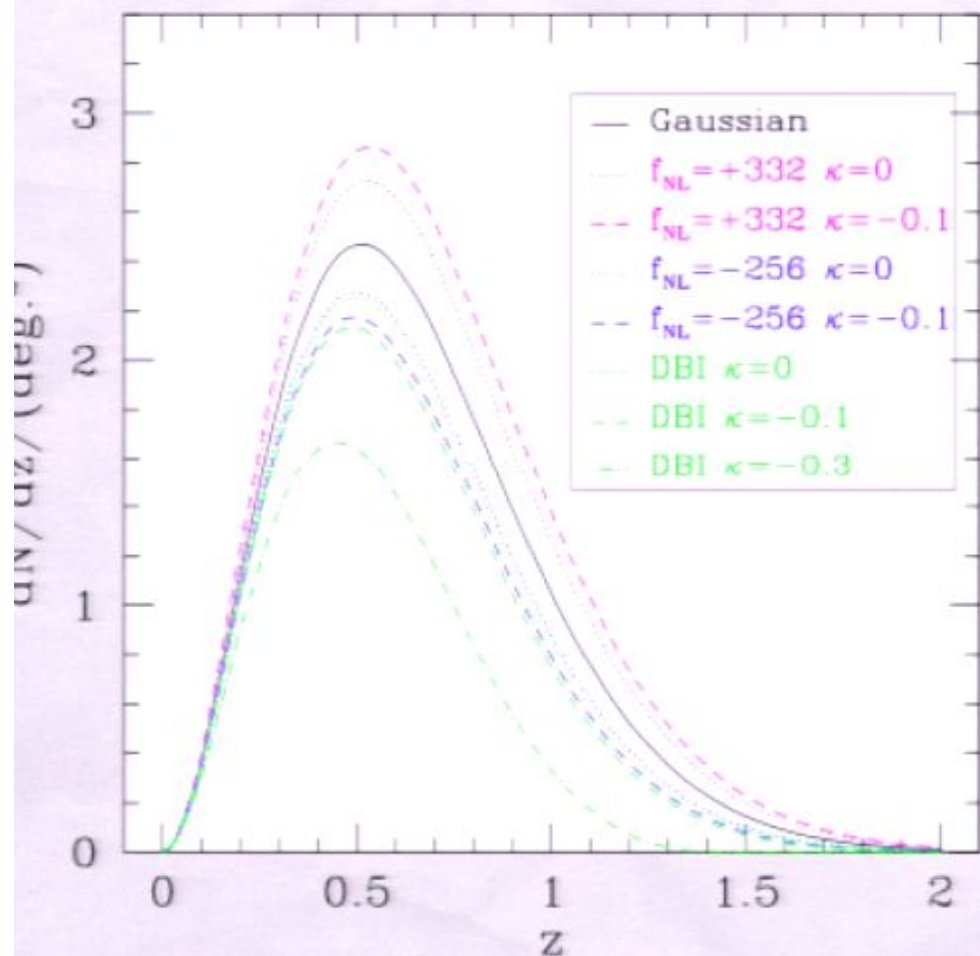


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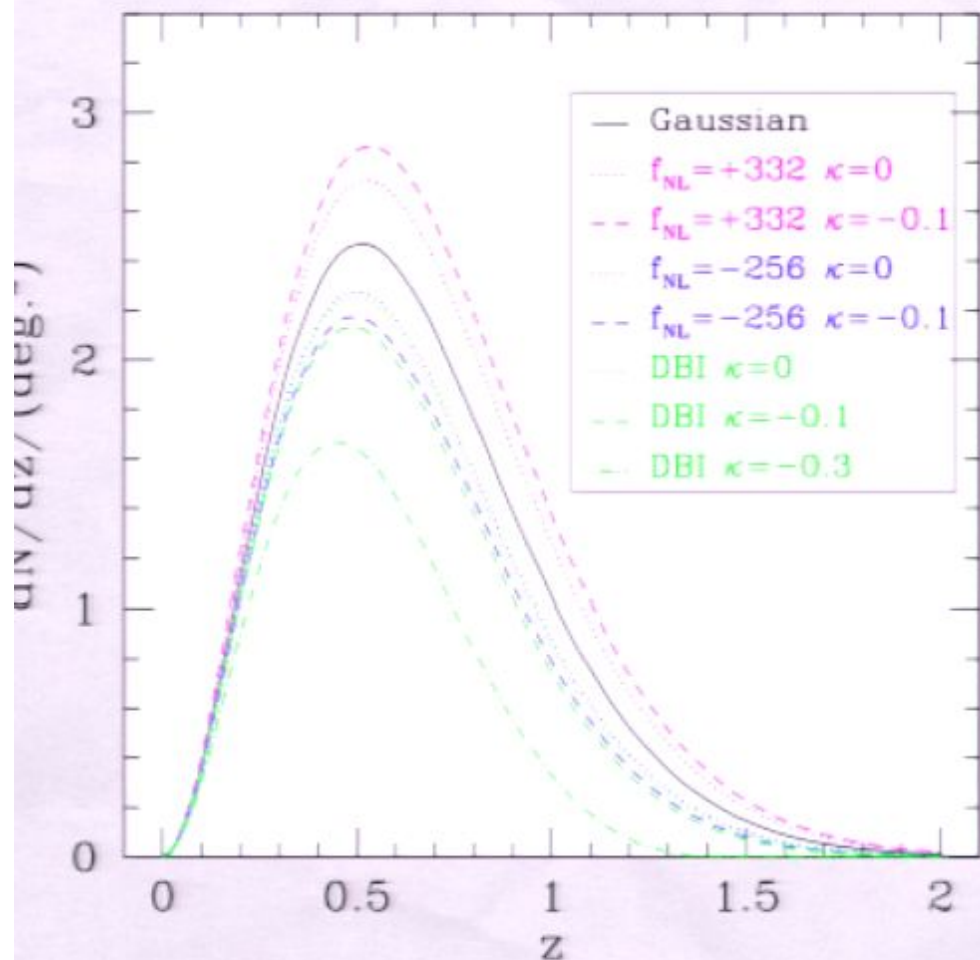


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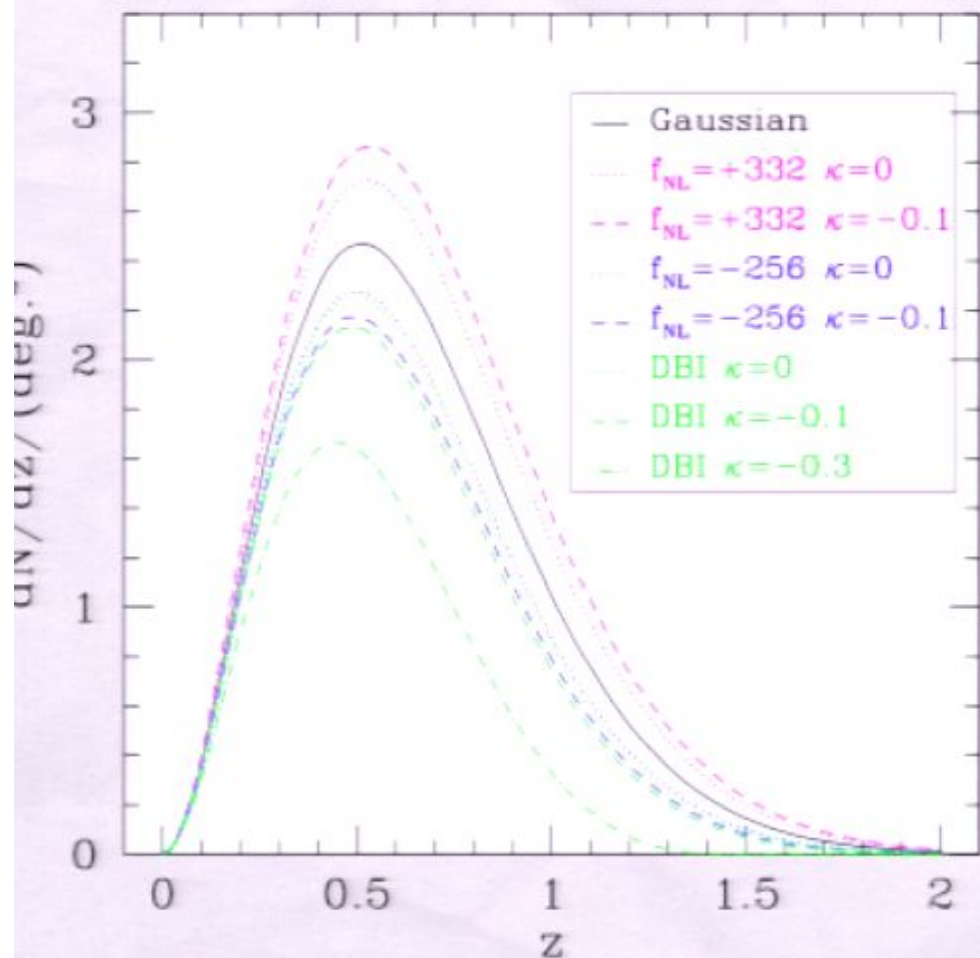


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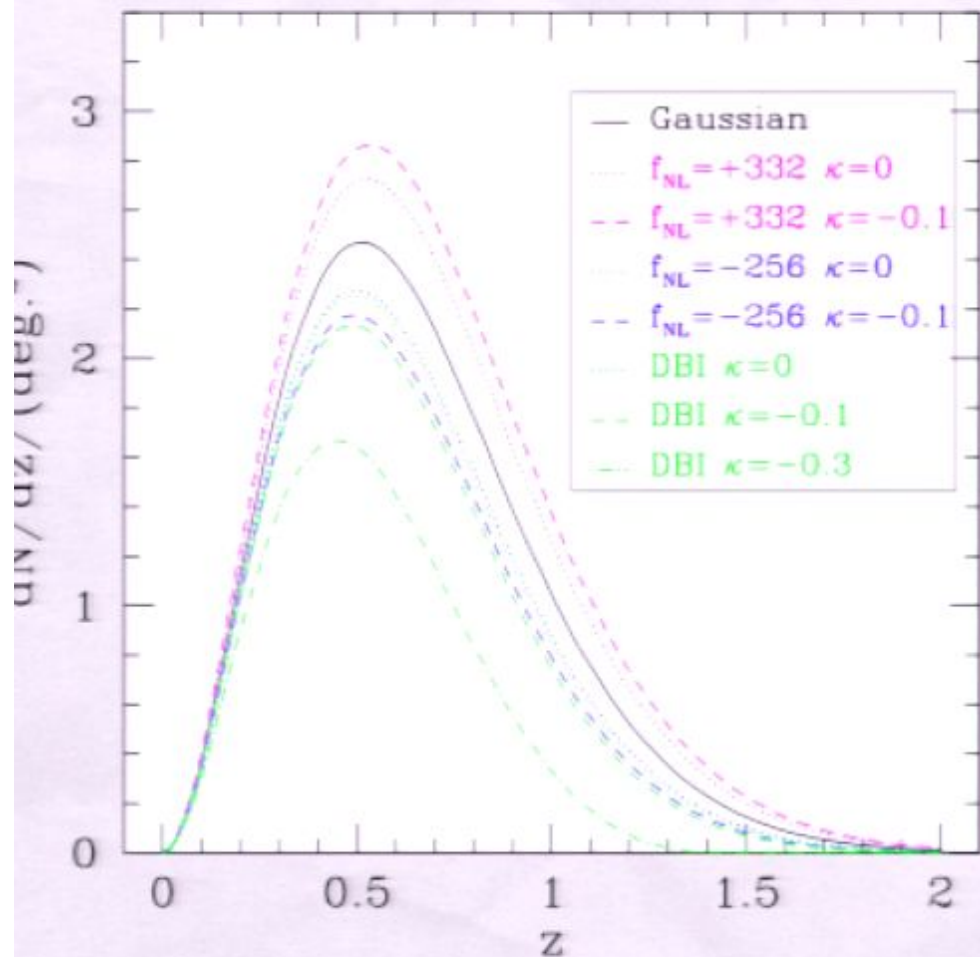


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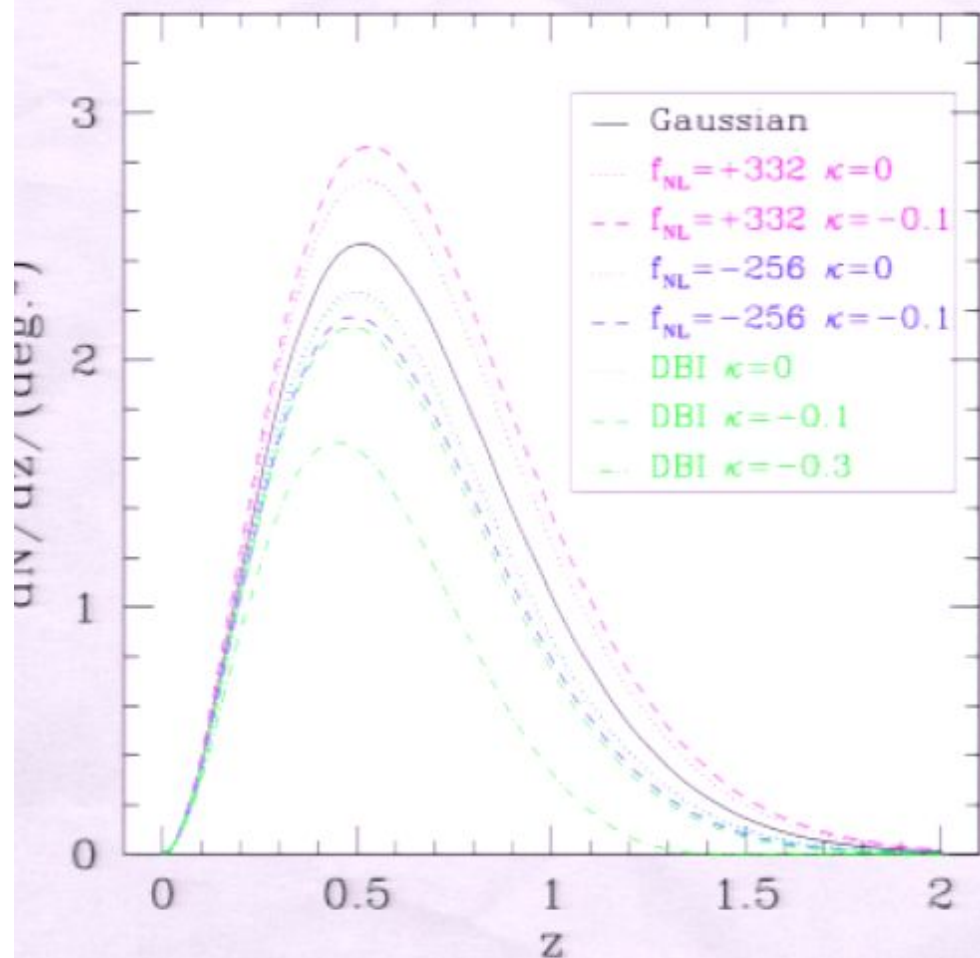


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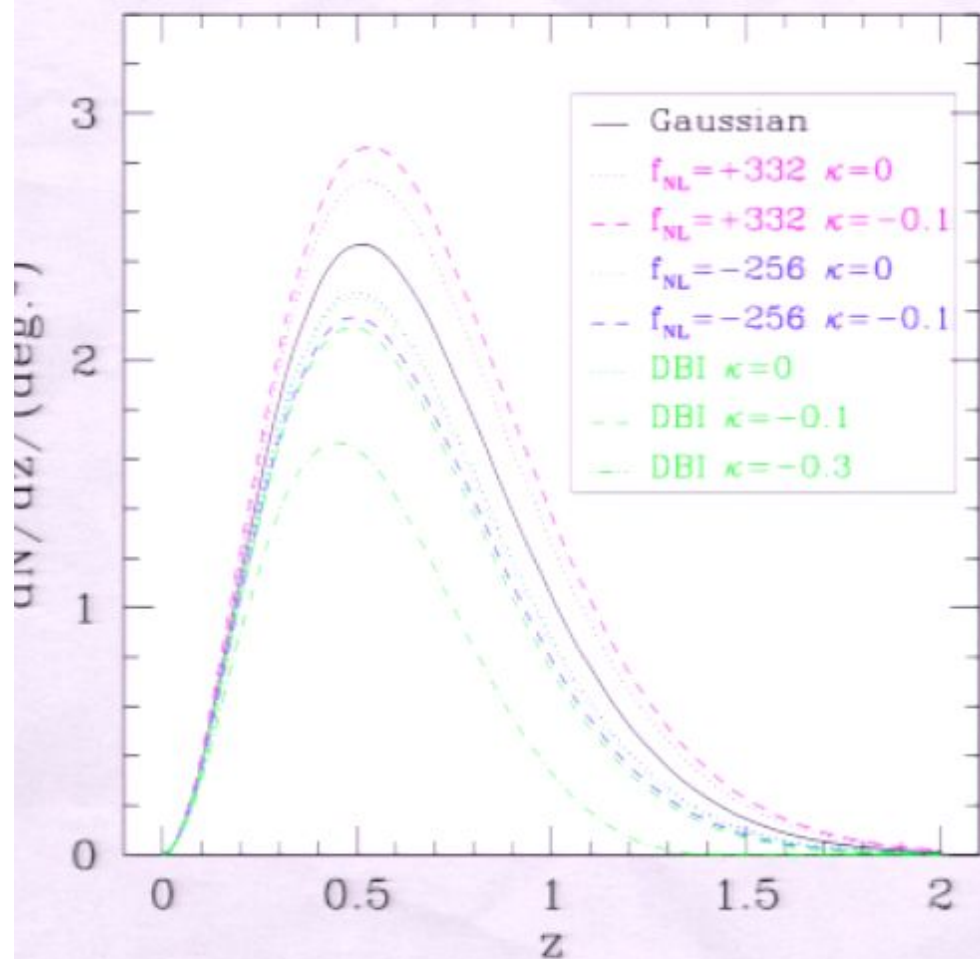


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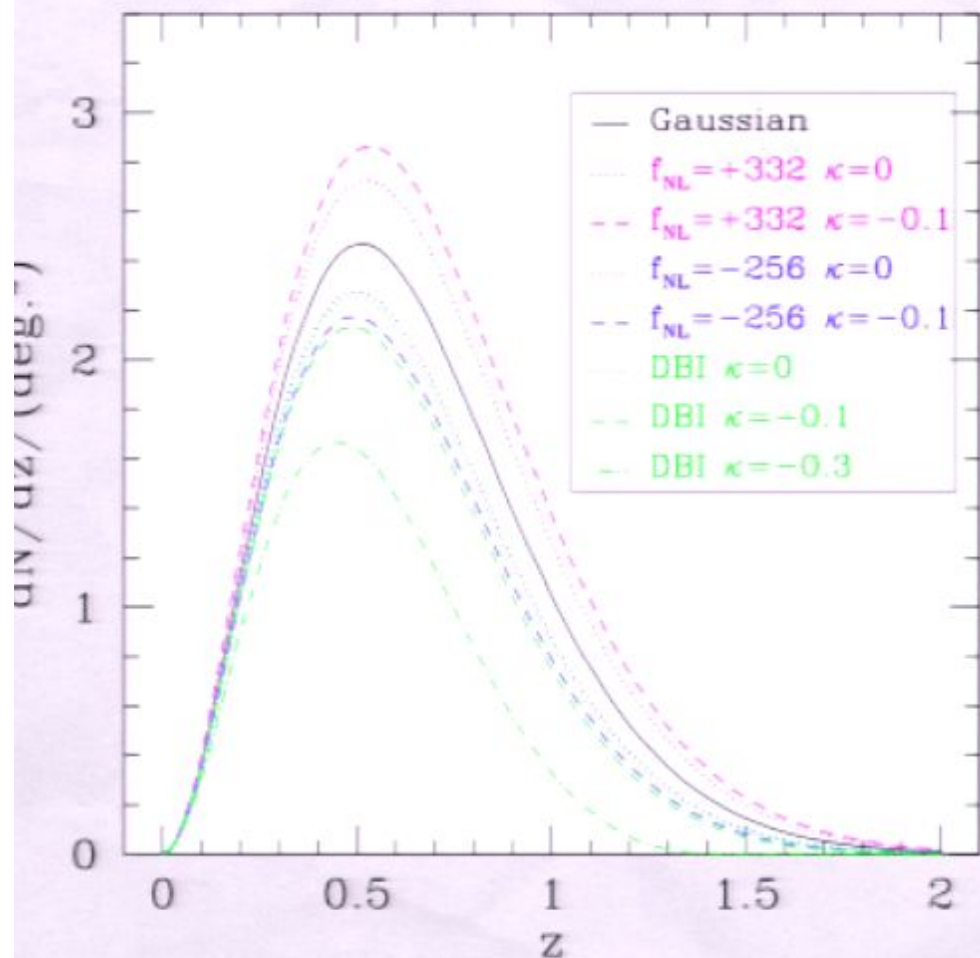


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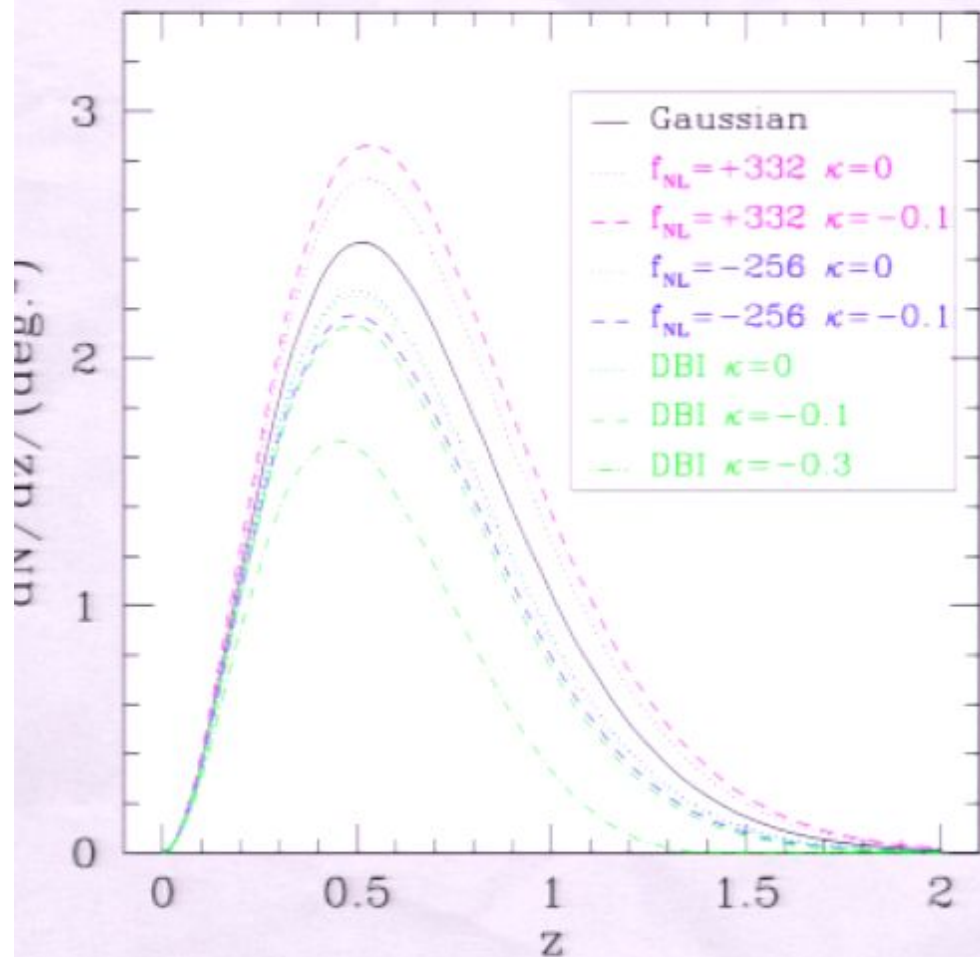


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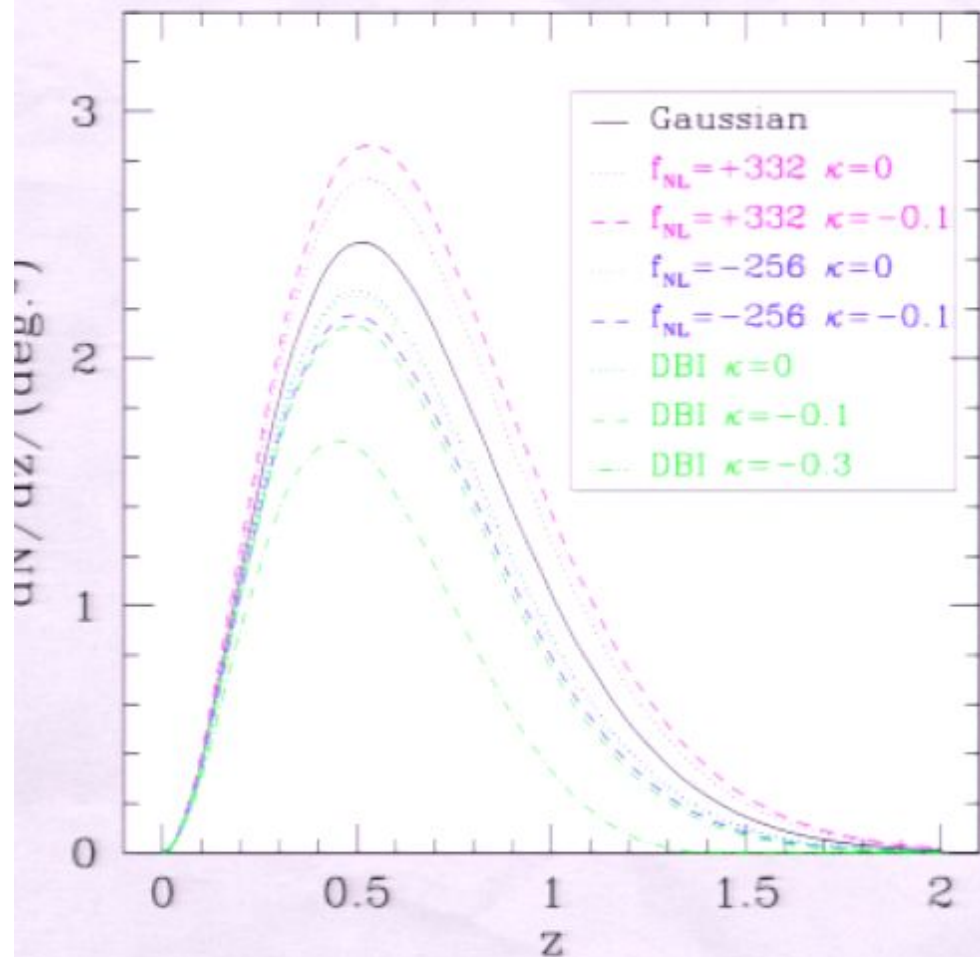


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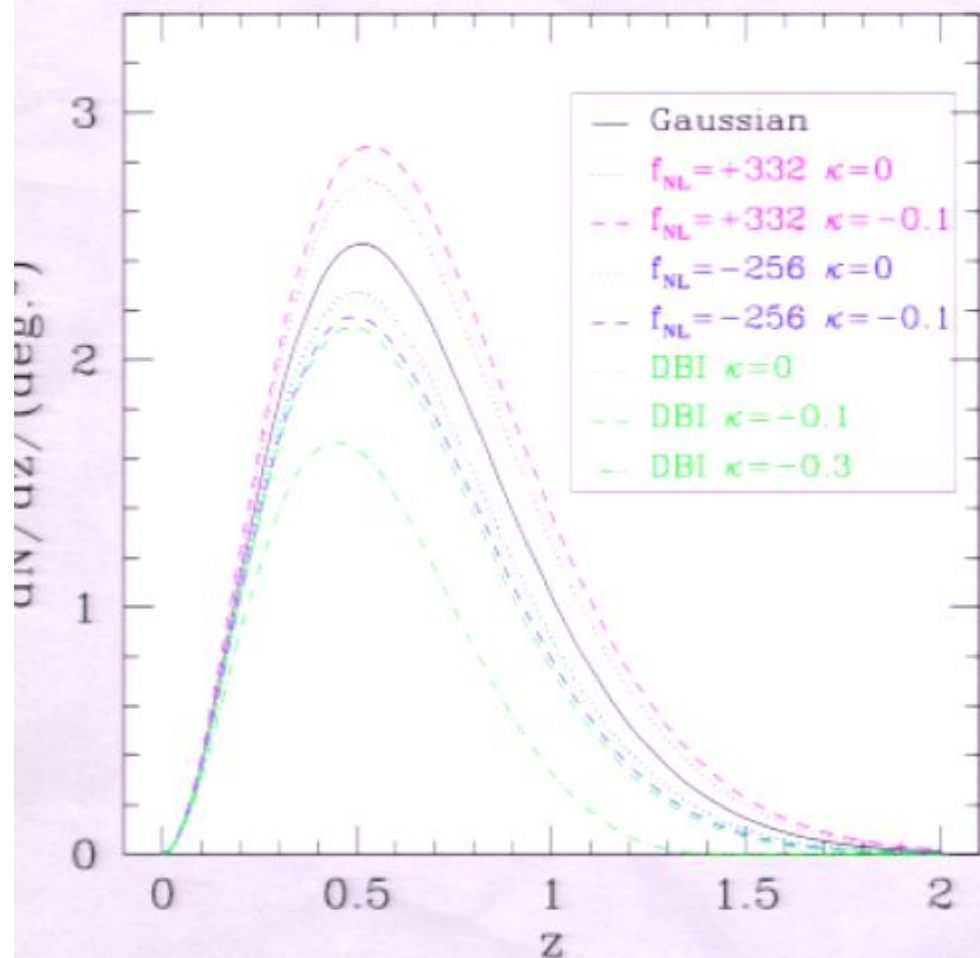


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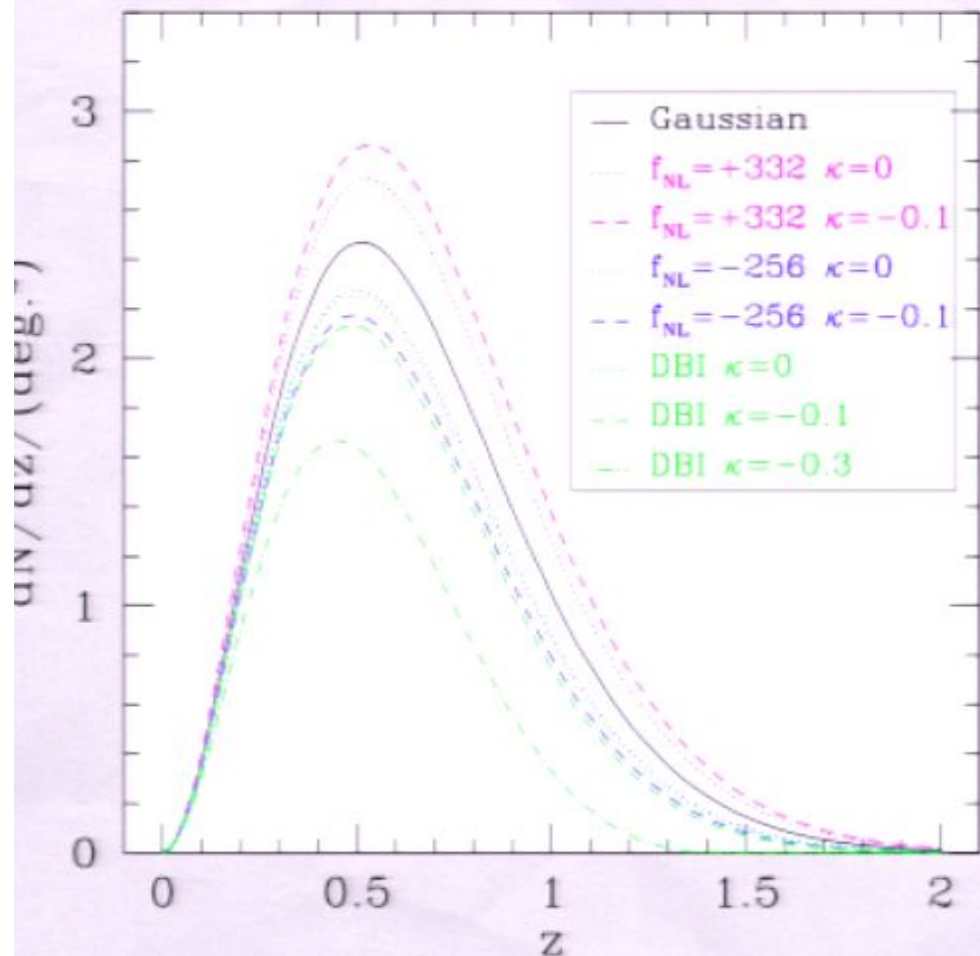


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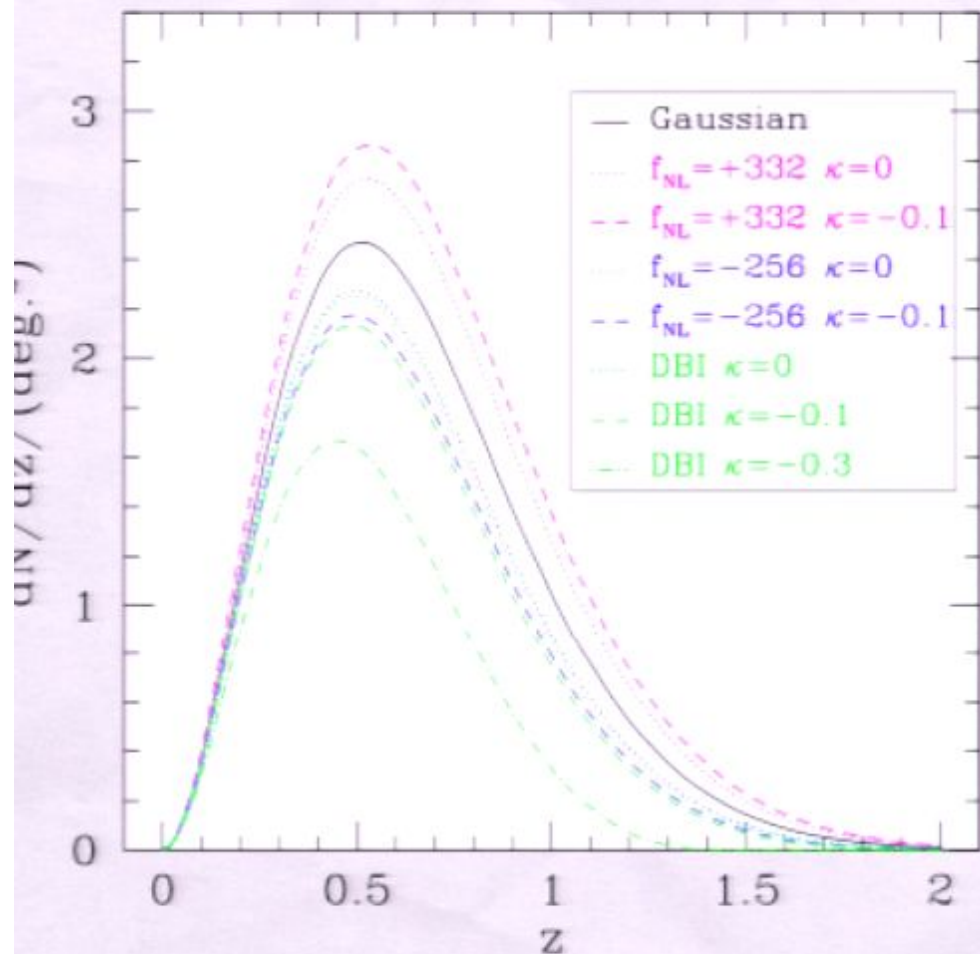


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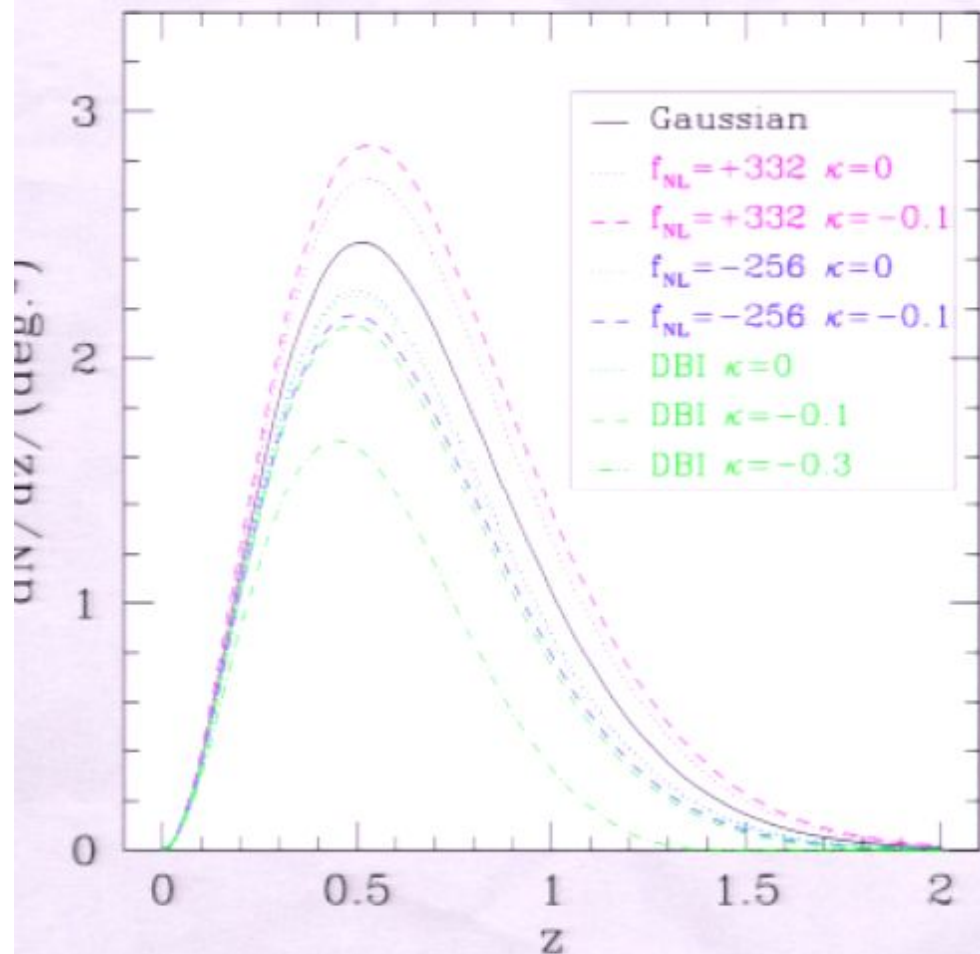


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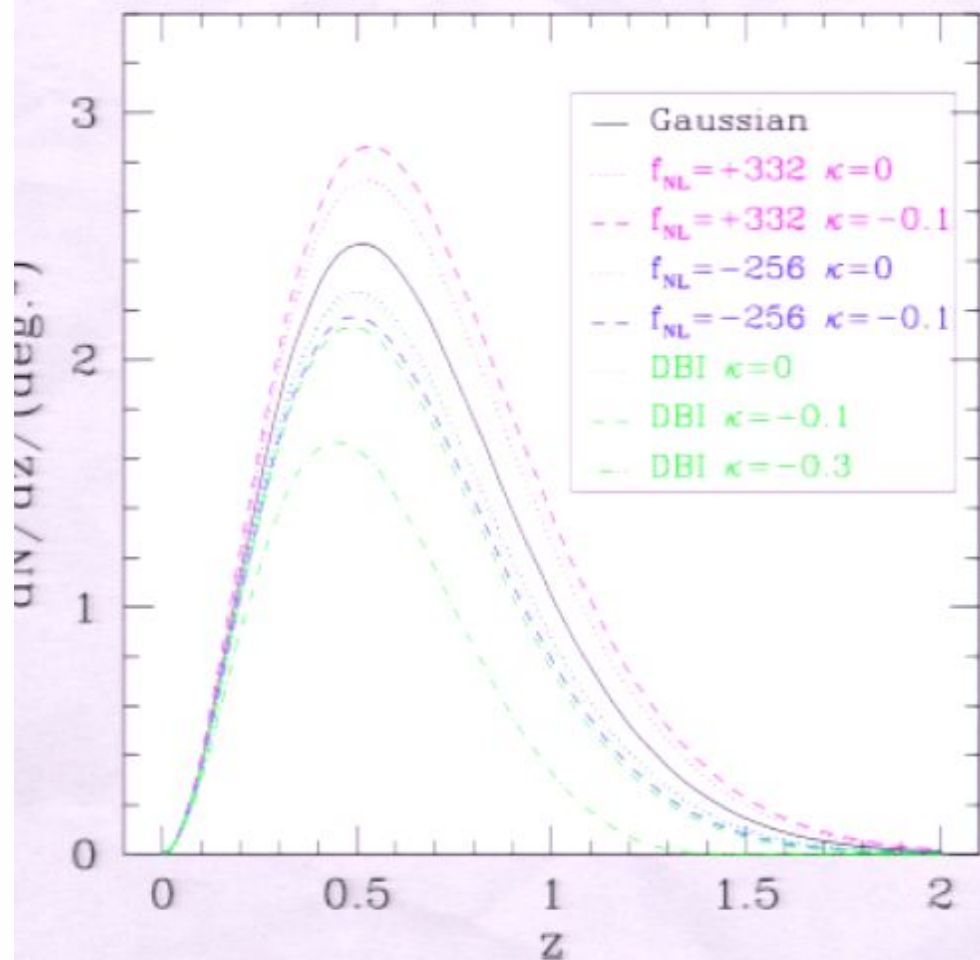


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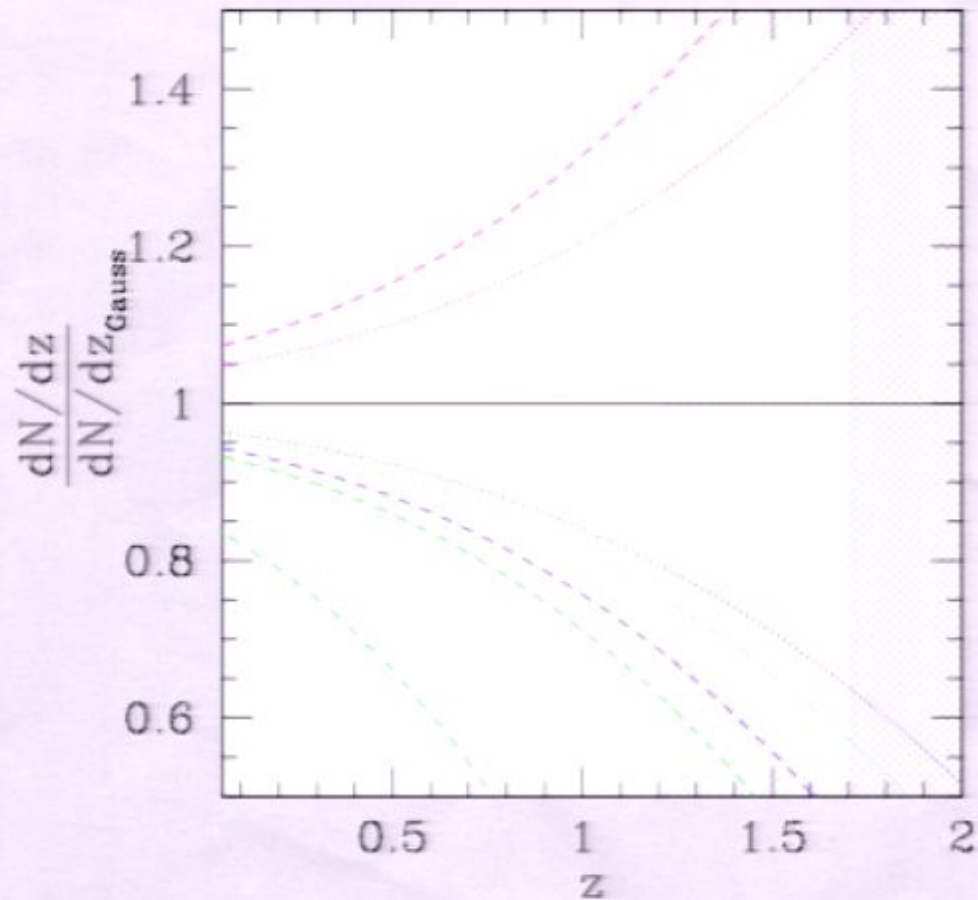


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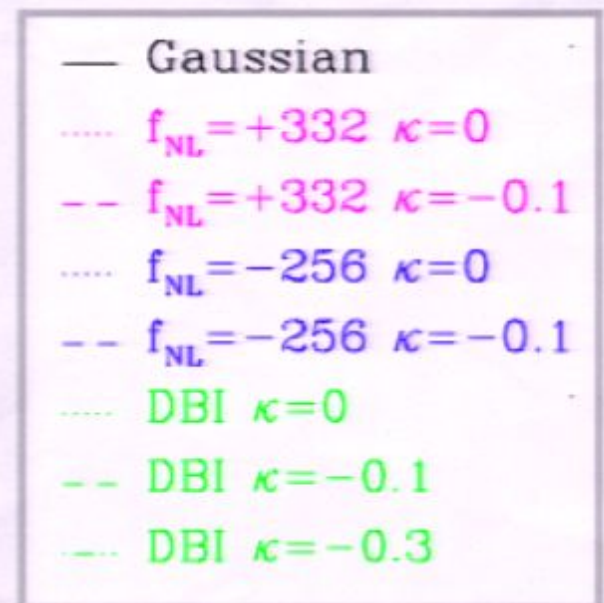
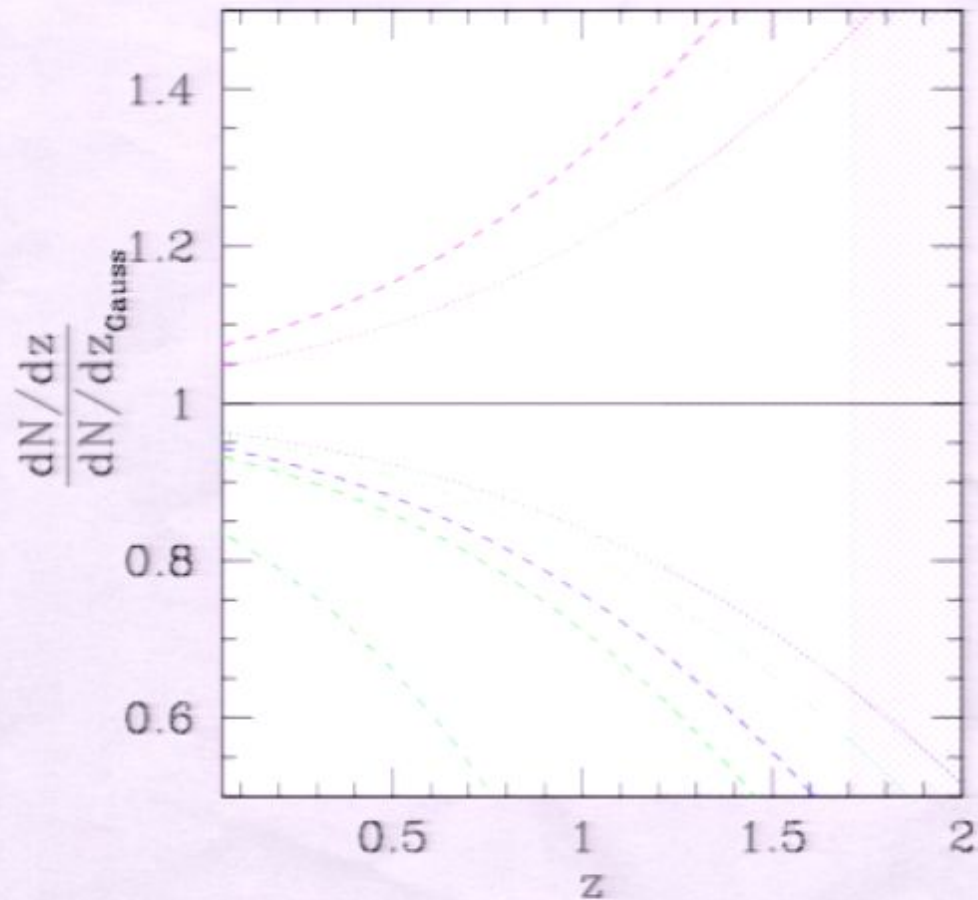
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DEVIATION FROM GAUSSIAN

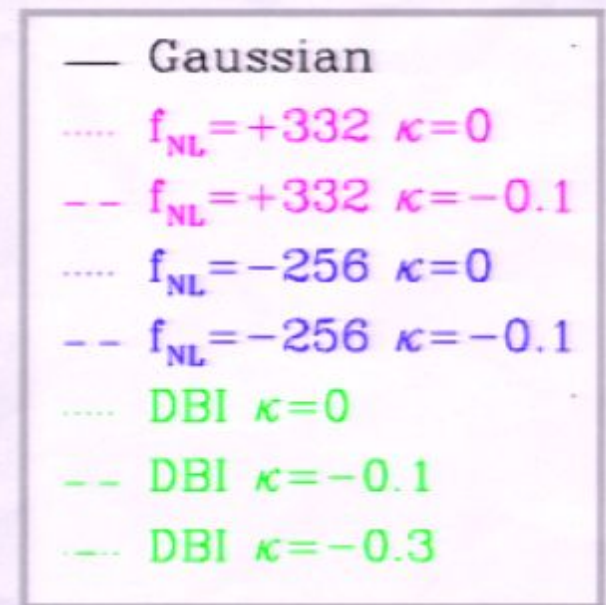
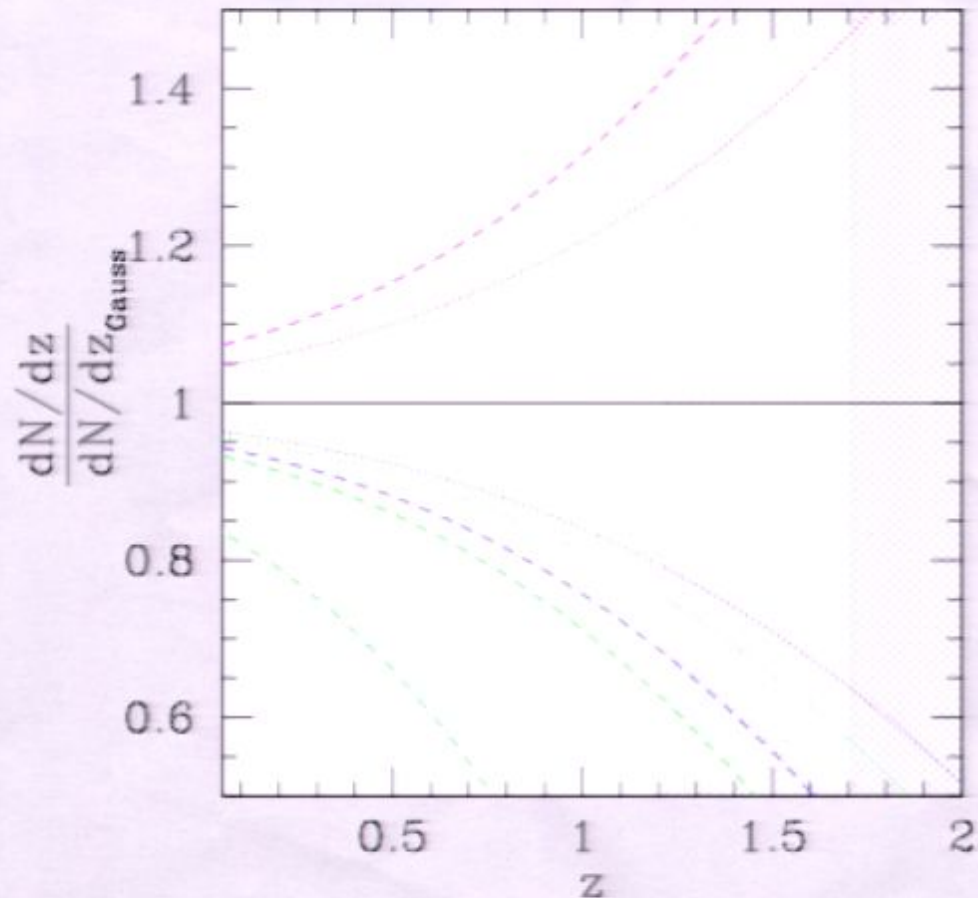


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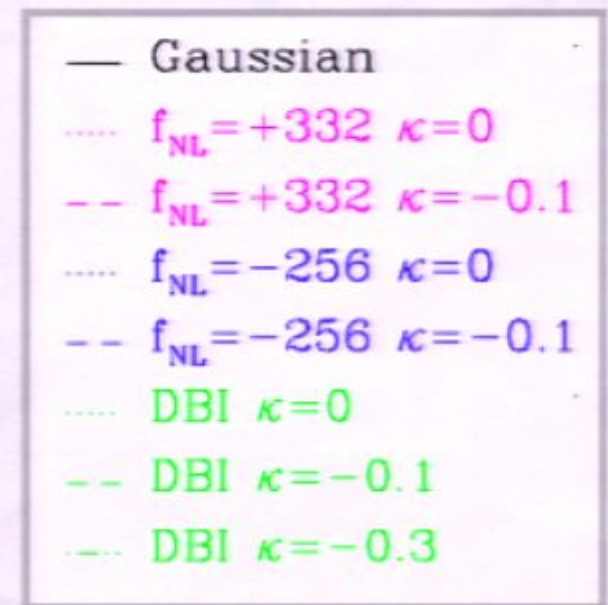
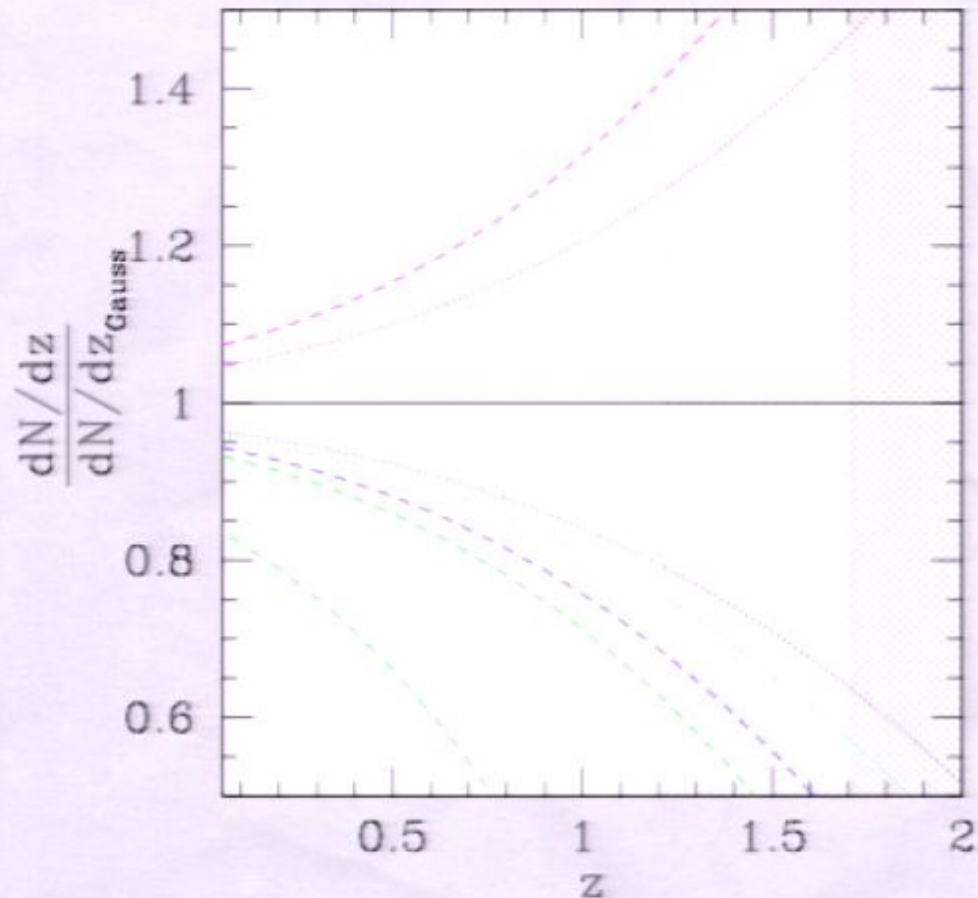
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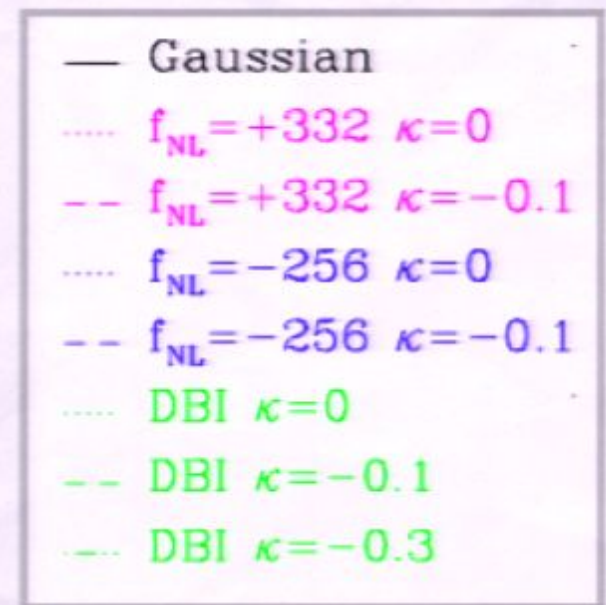
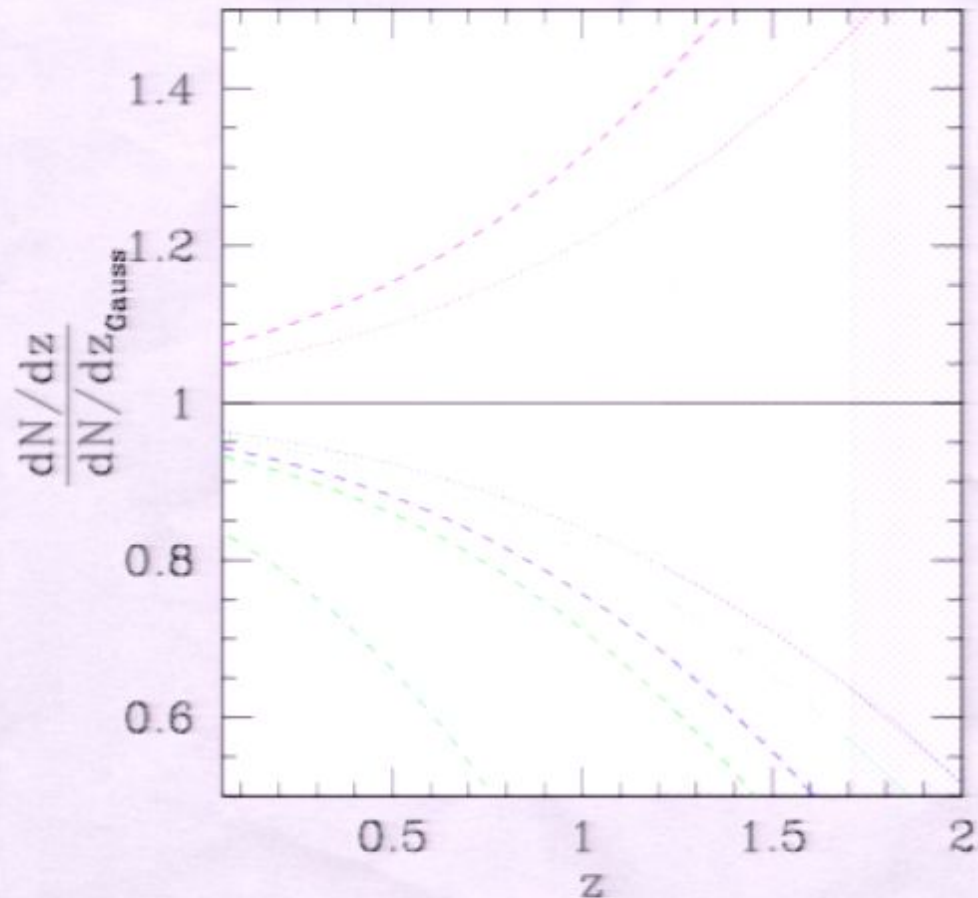
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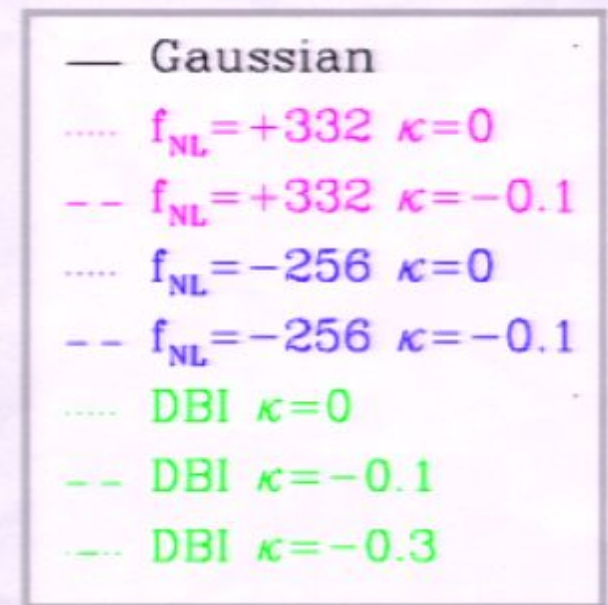
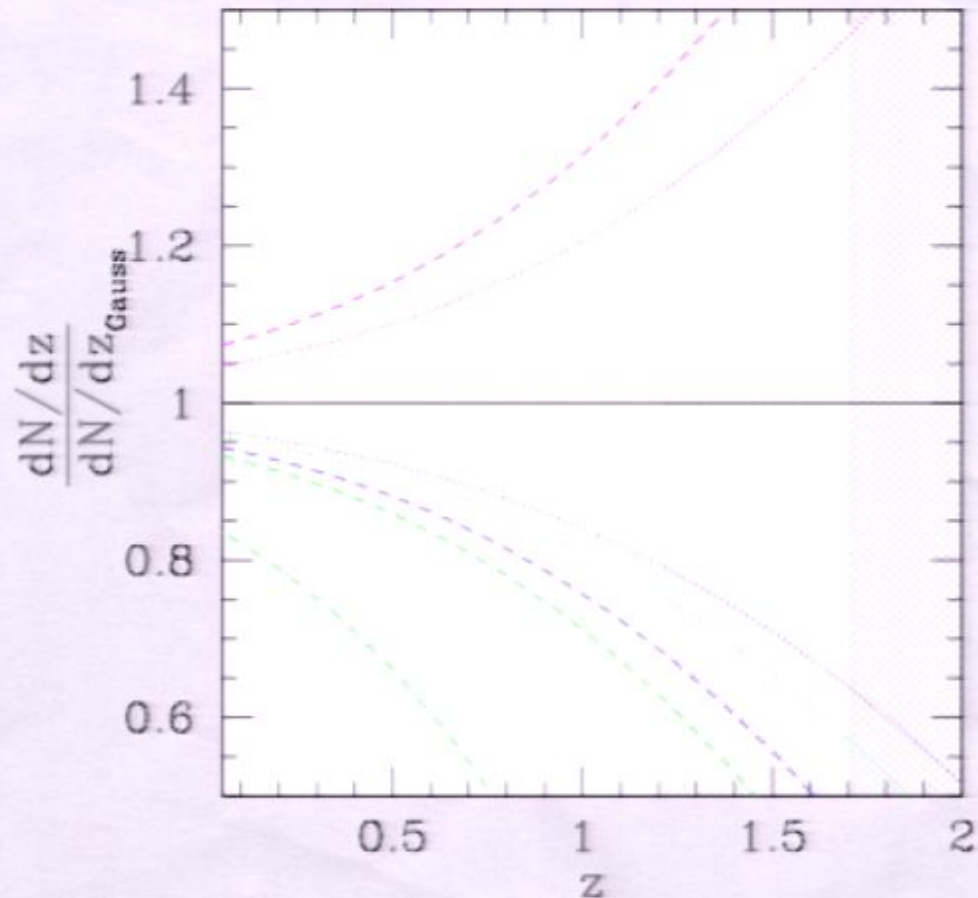
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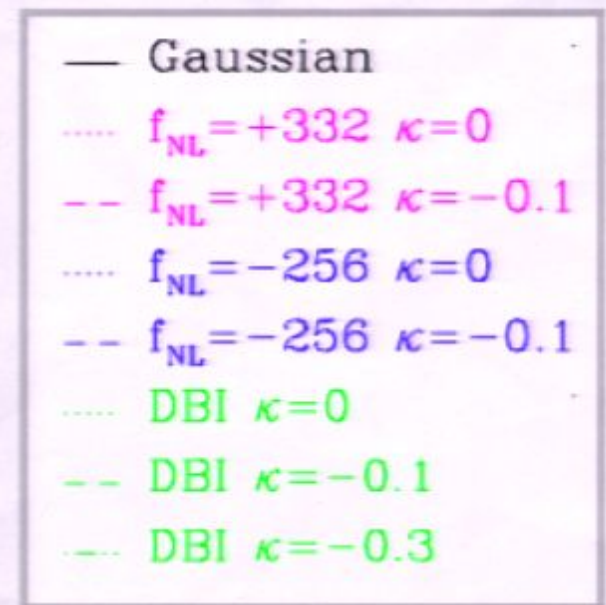
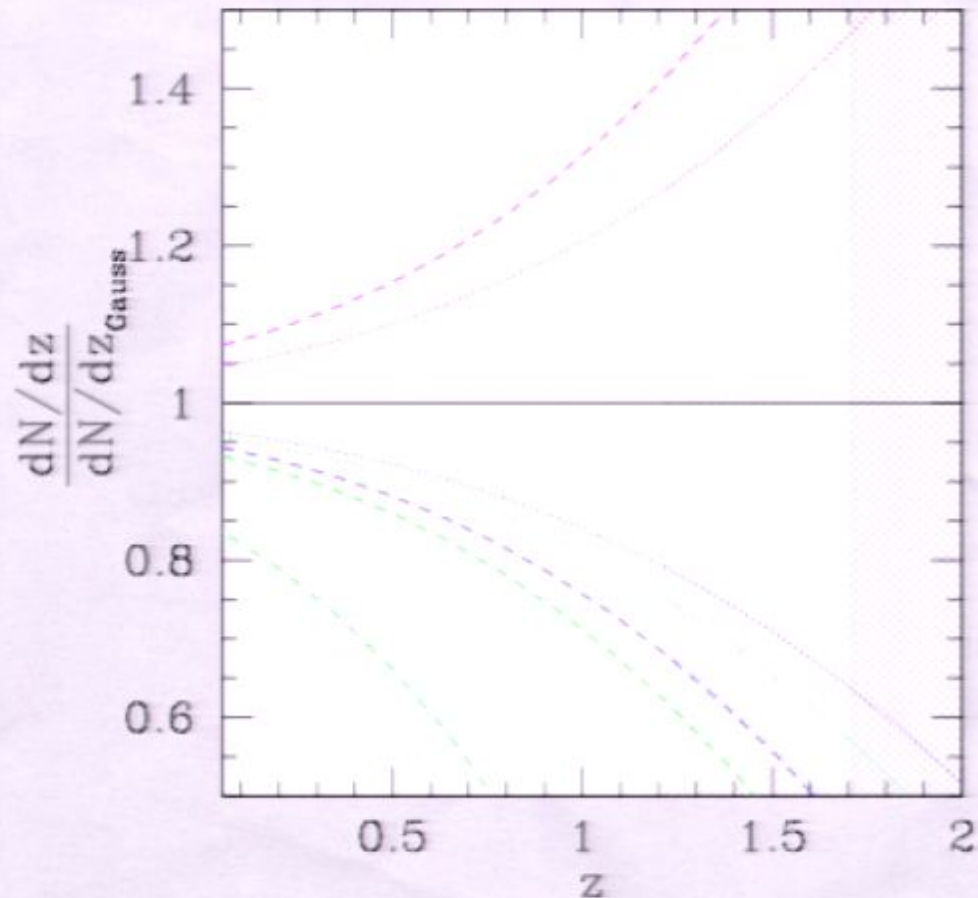
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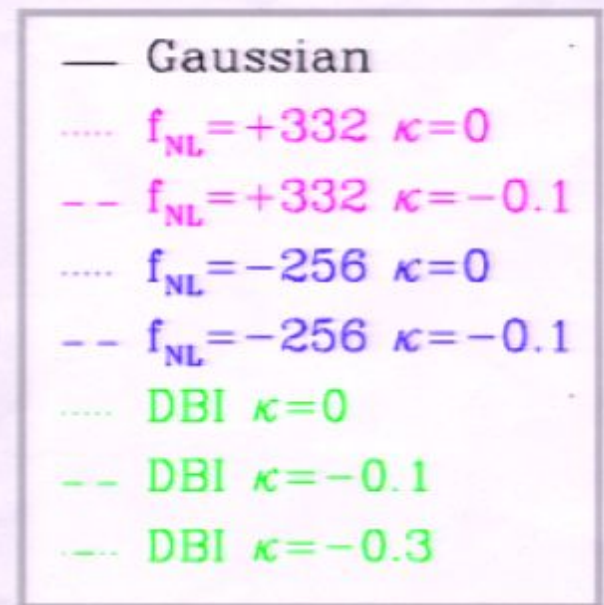
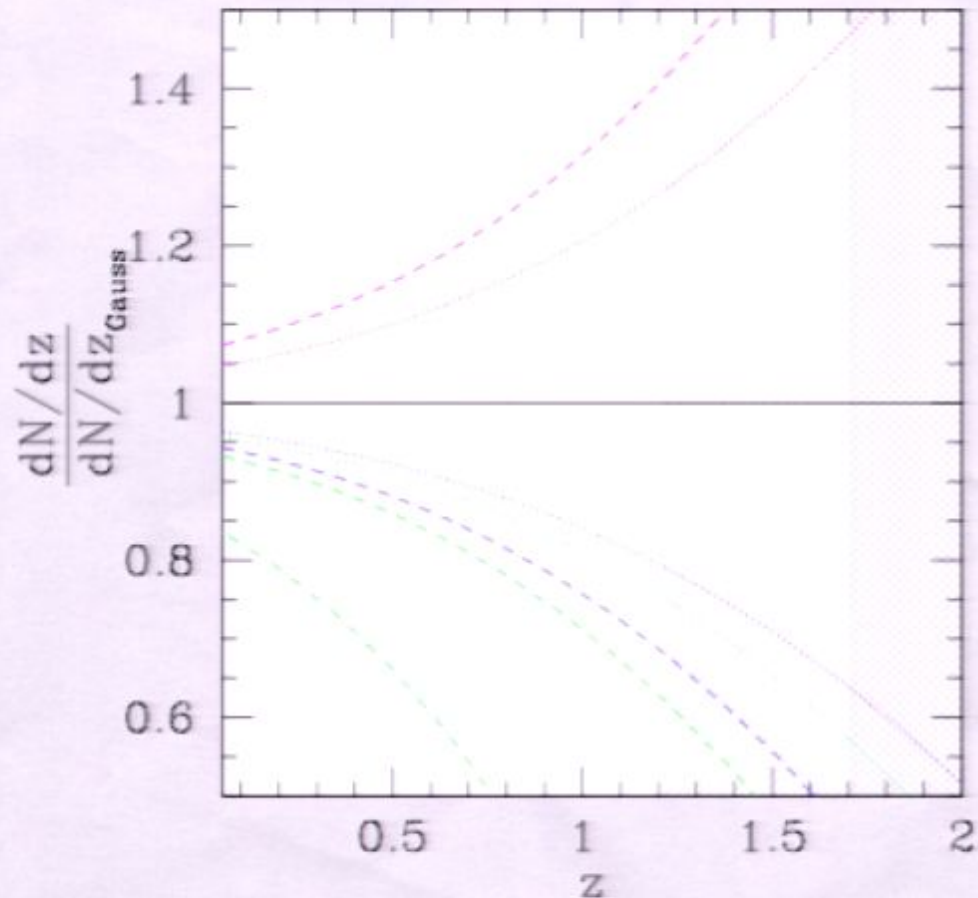
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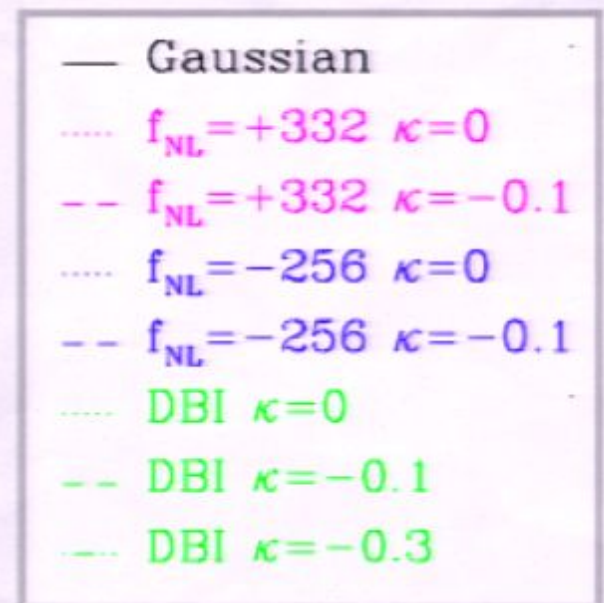
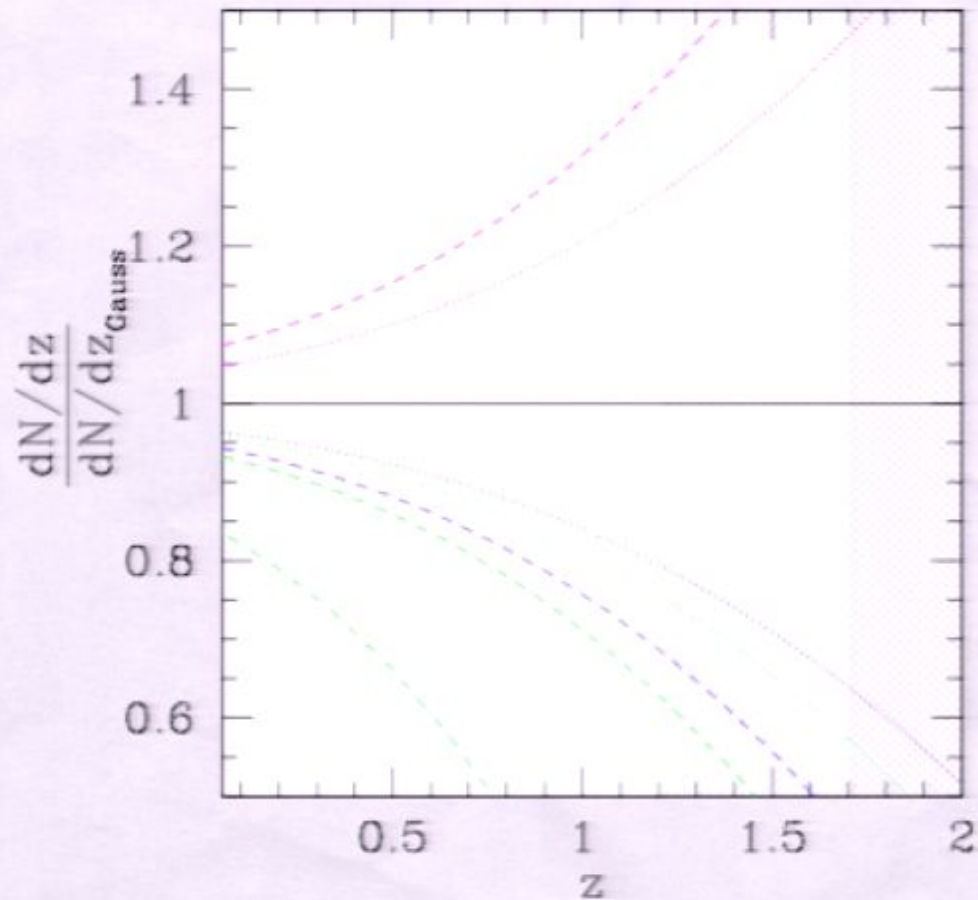
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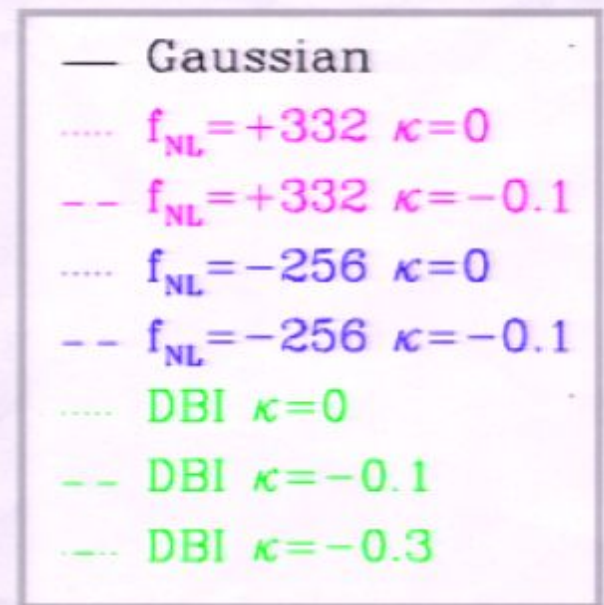
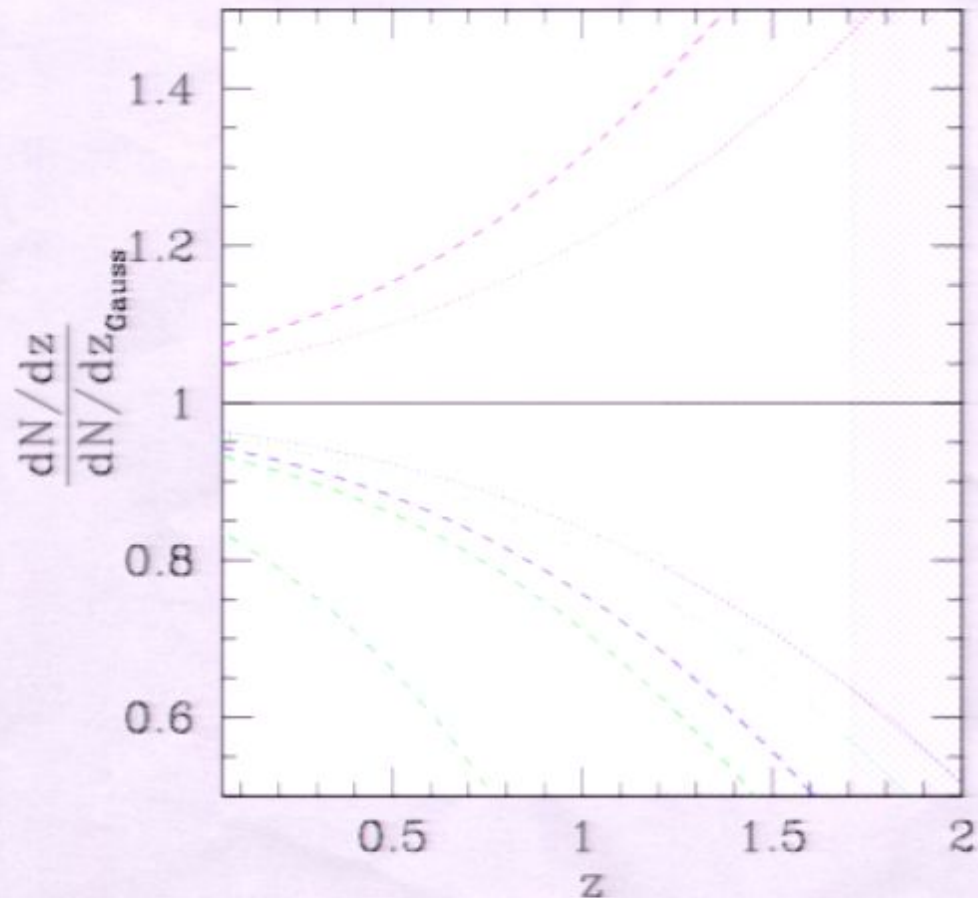
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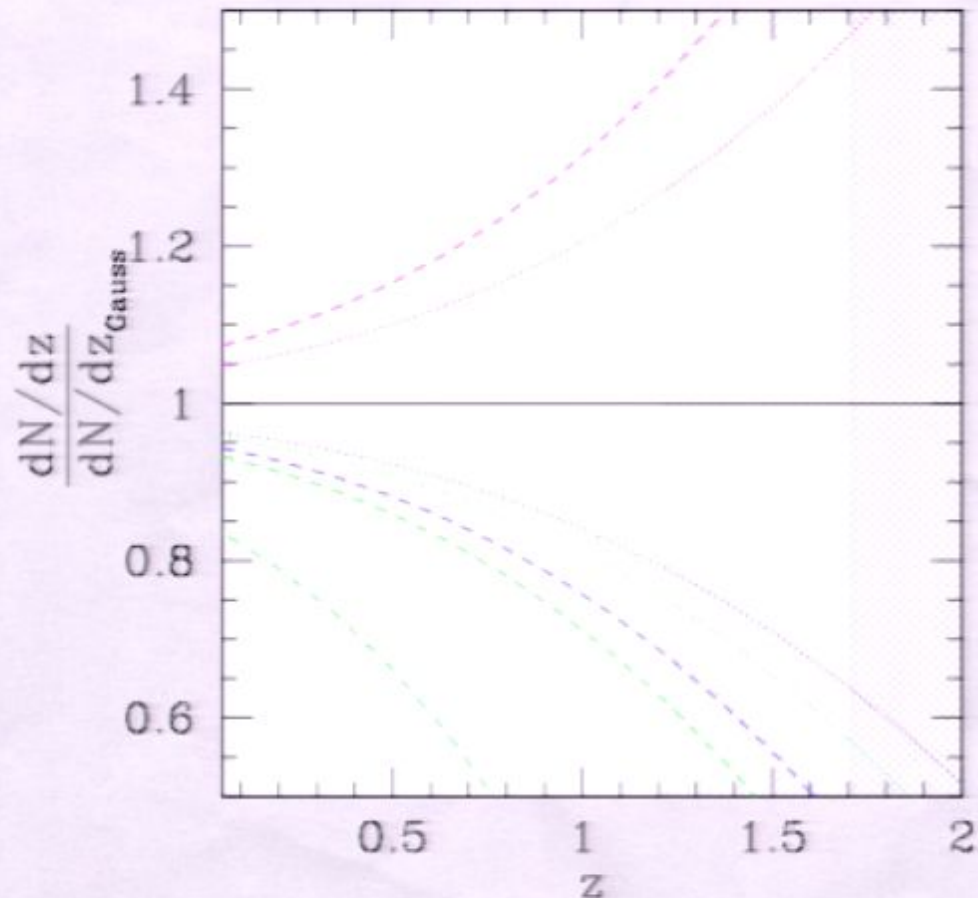
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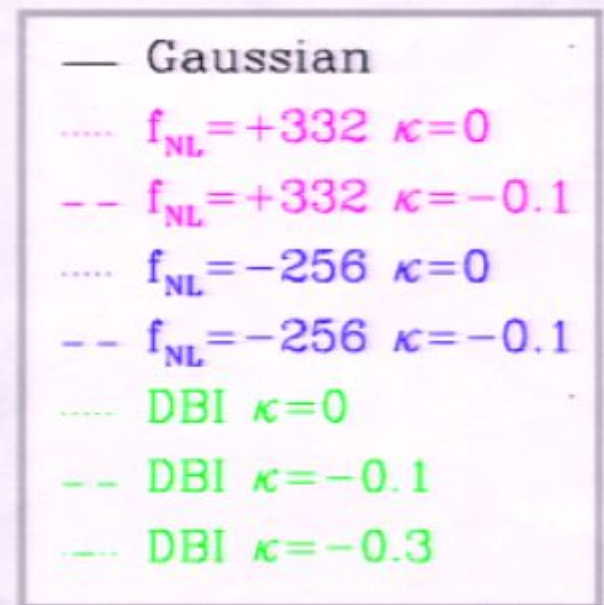
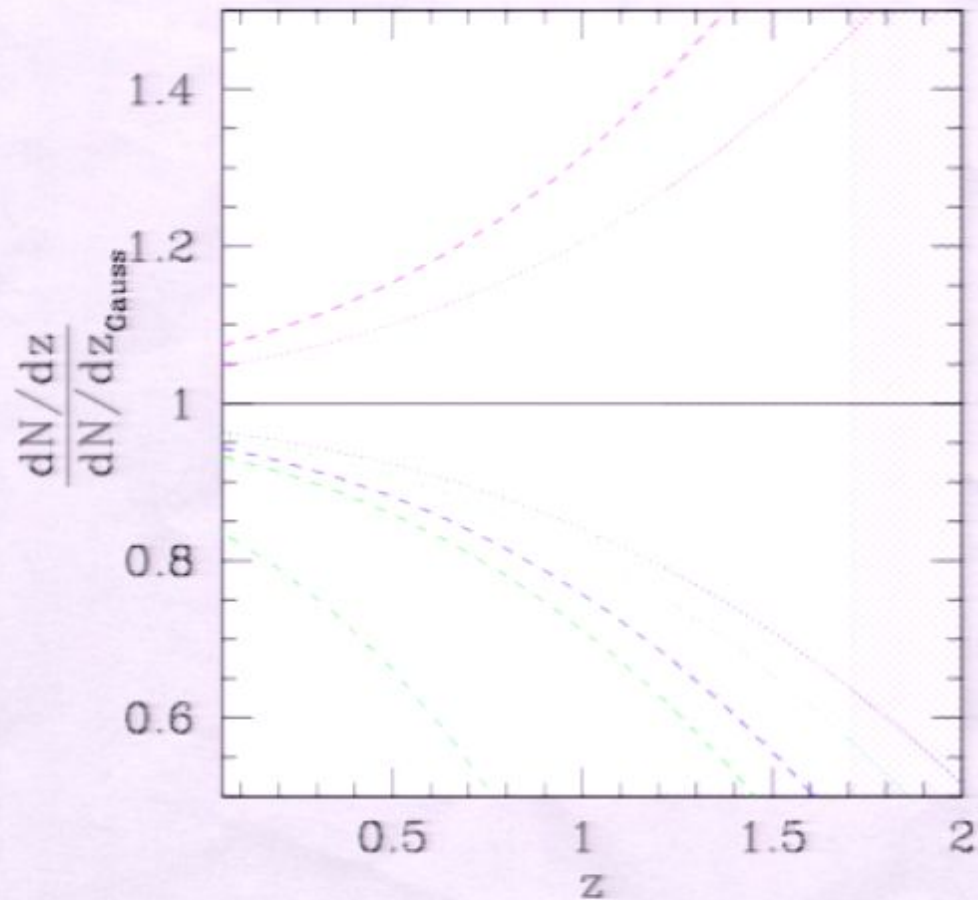


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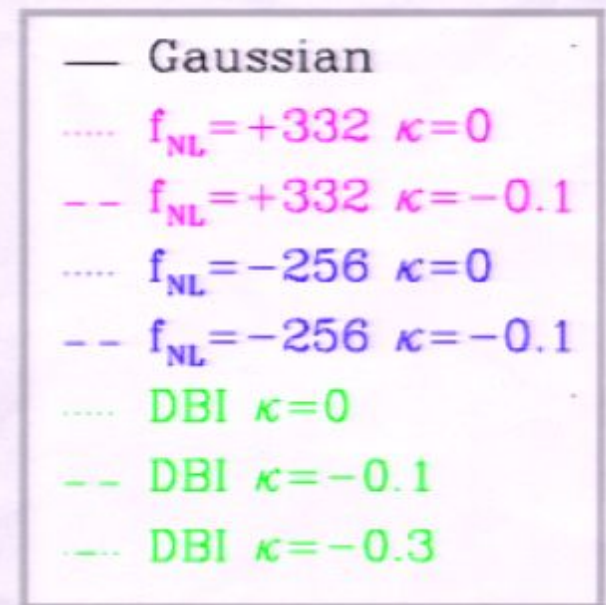
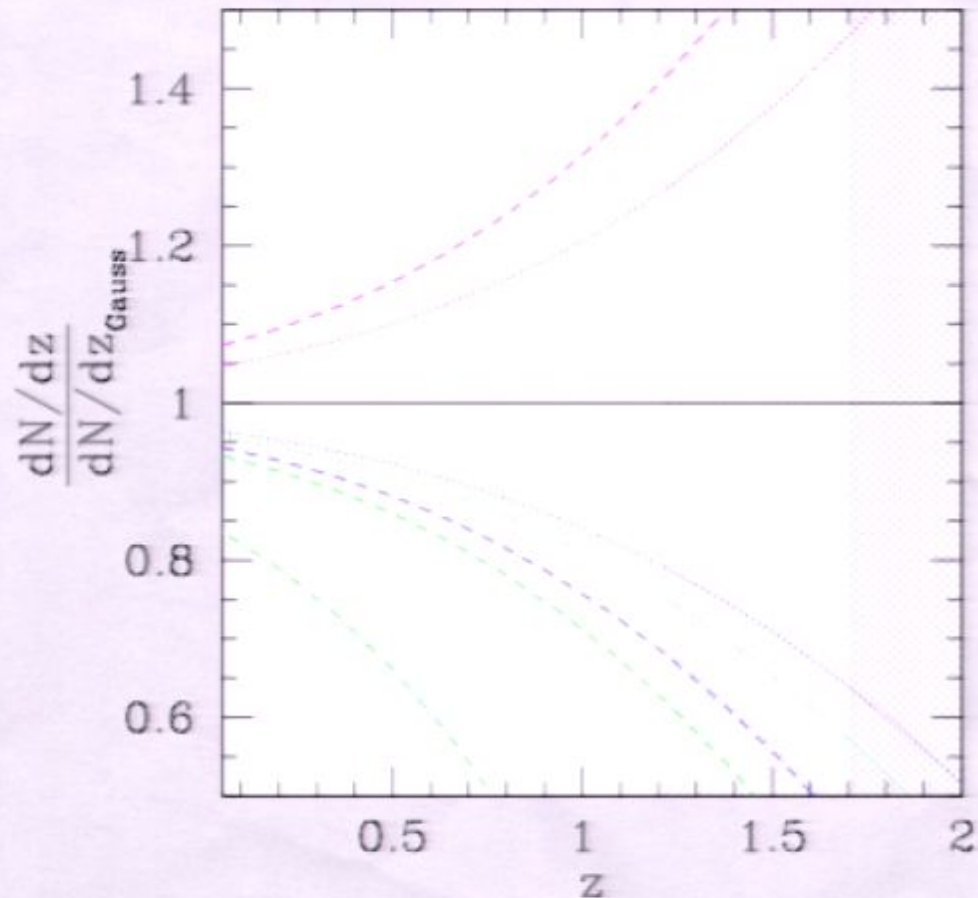


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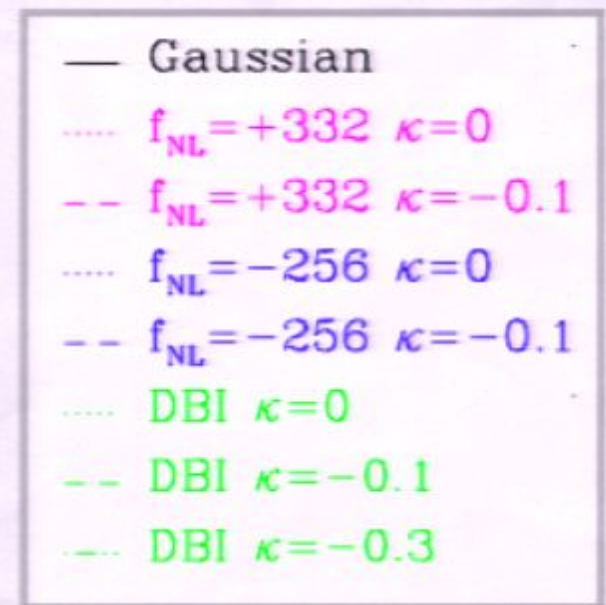
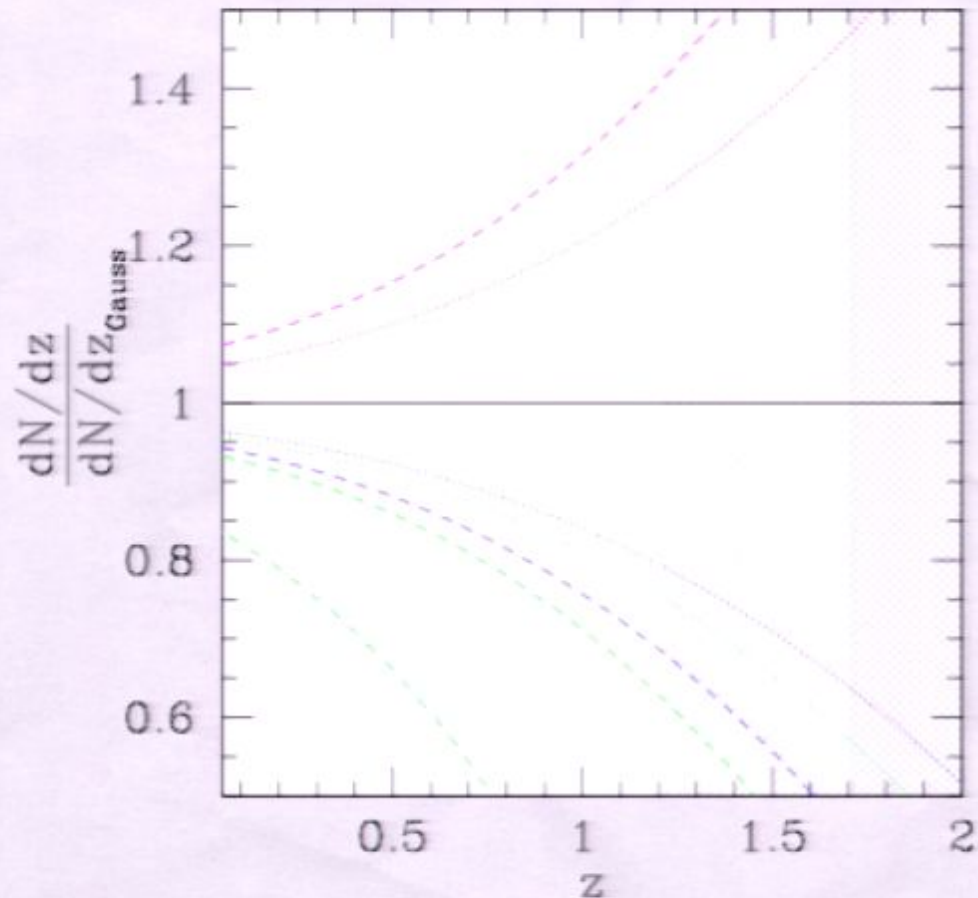
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SOME CONCERNS...

- Need to know cluster mass well
- Number of clusters is very sensitive to σ_8

(Analysis with scale-independent NG and dark energy done by Sefusatti et al, 2006; Mantz et al with X-ray detected clusters)

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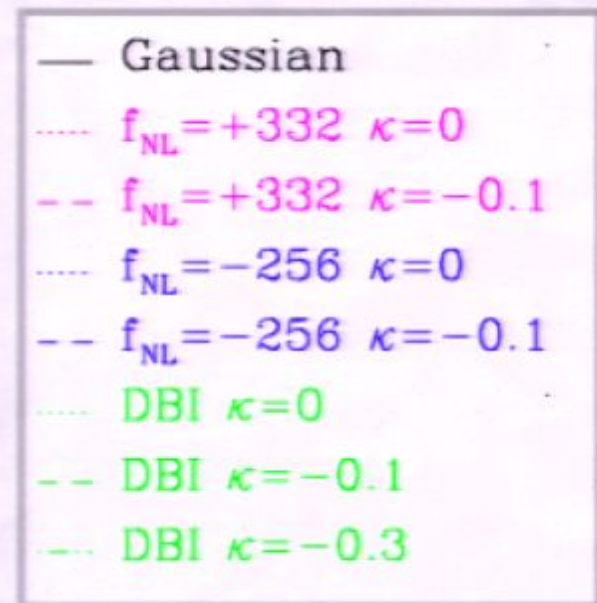
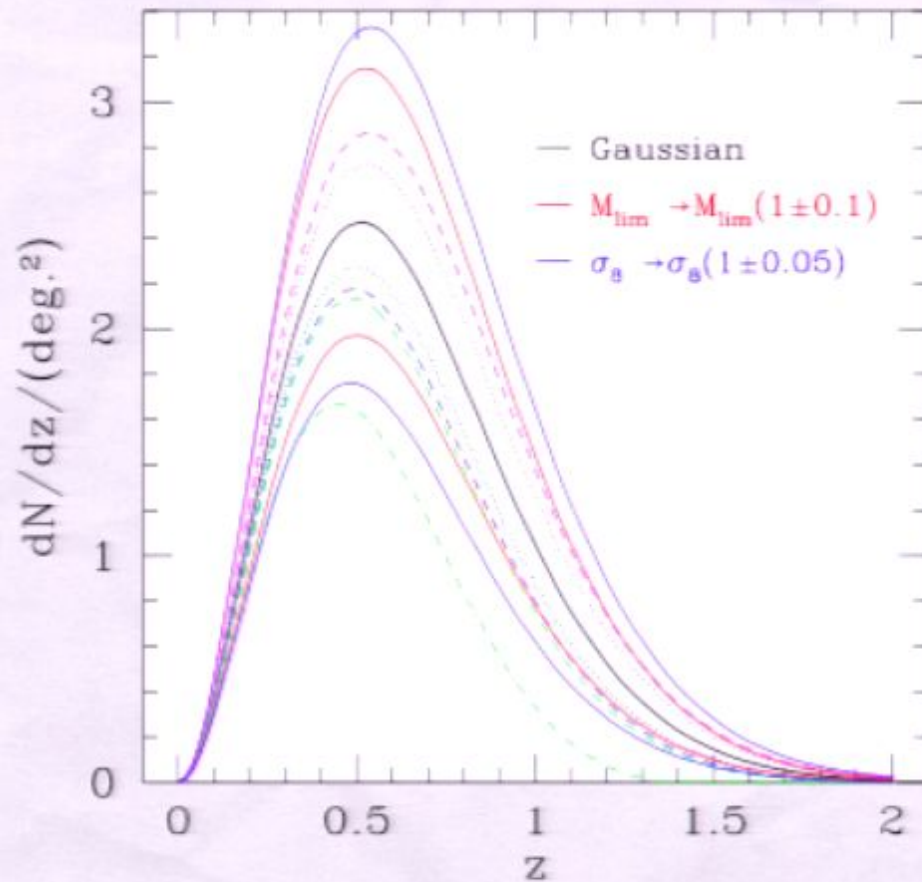
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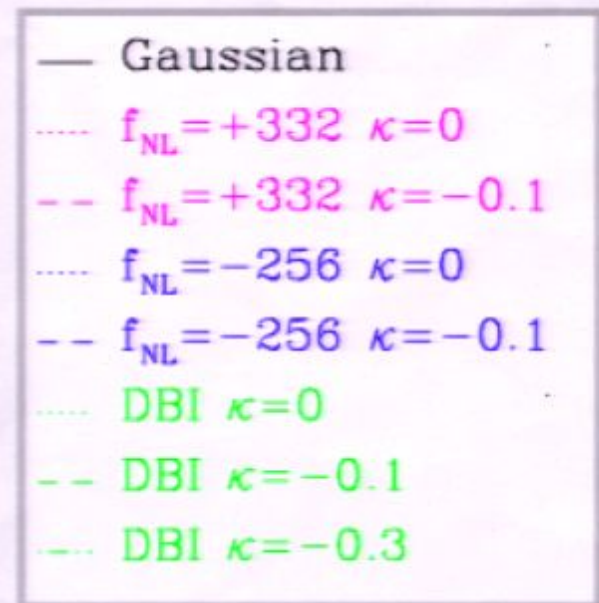
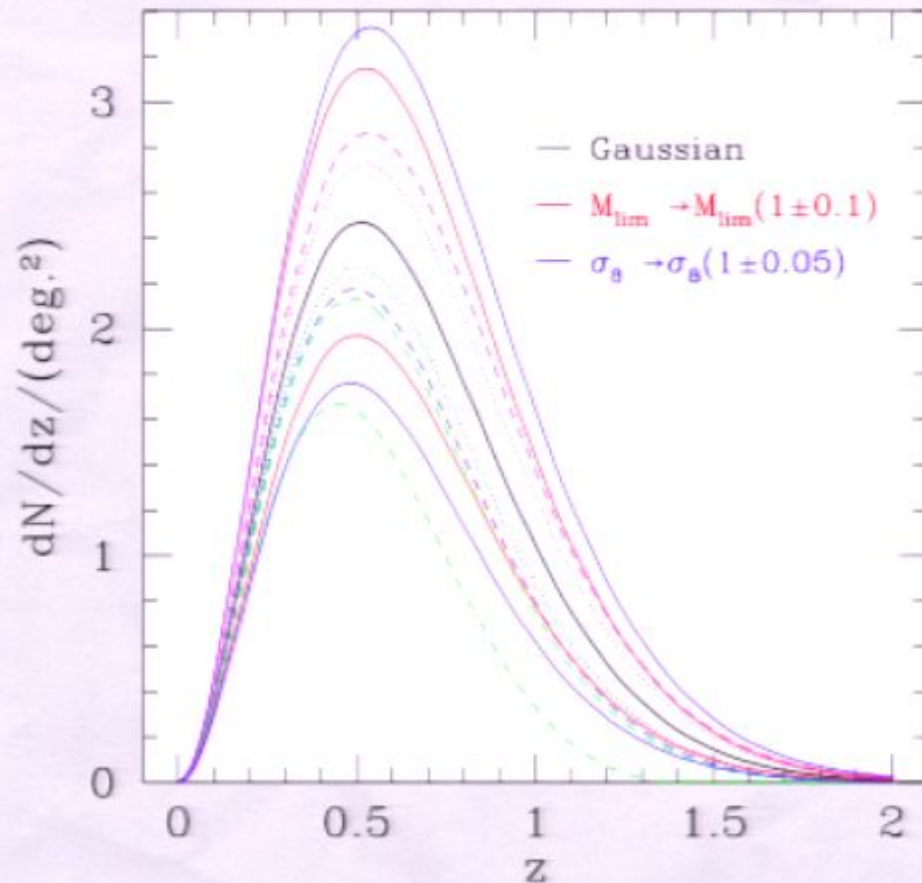
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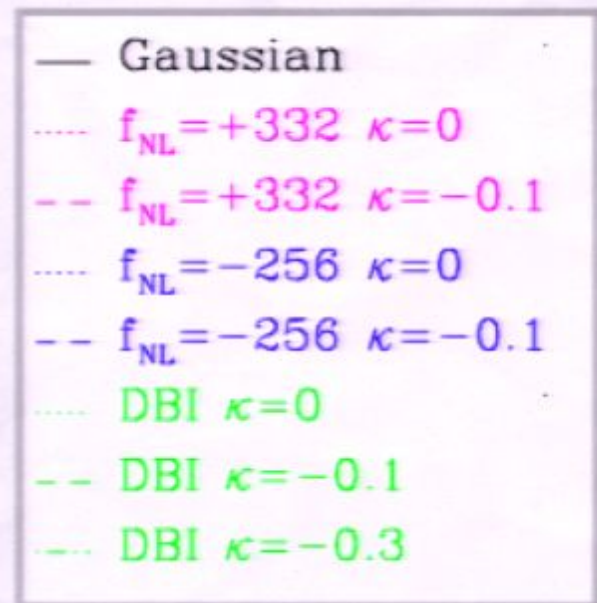
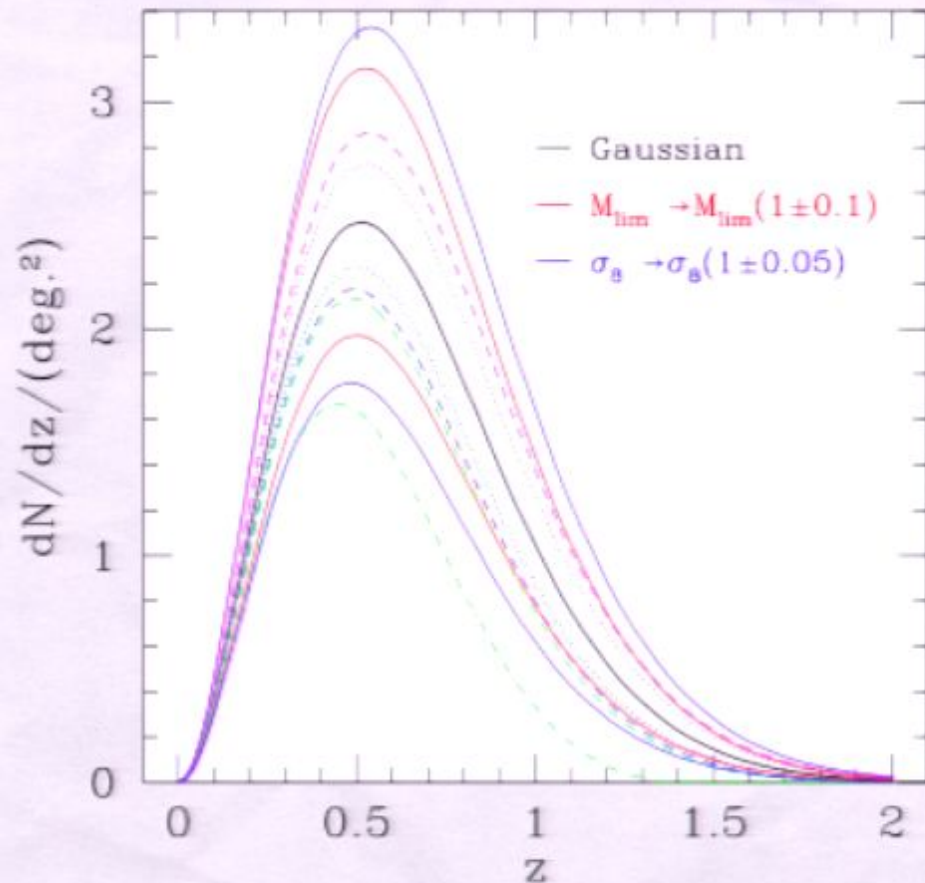
WITH MASS AND σ_8 UNCERTAINTY



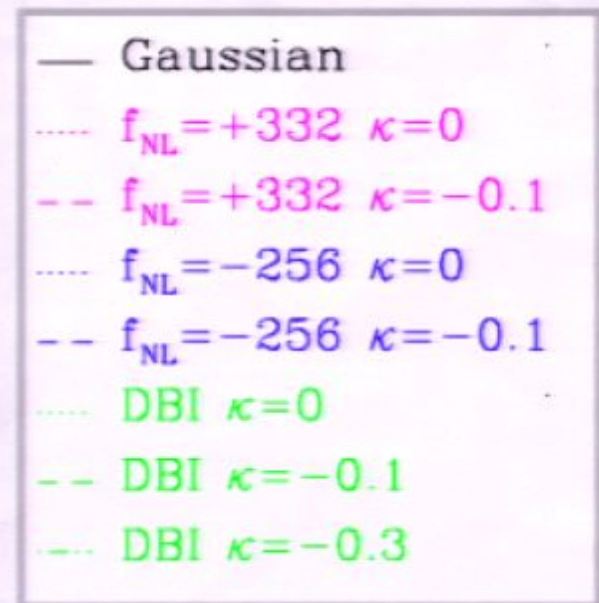
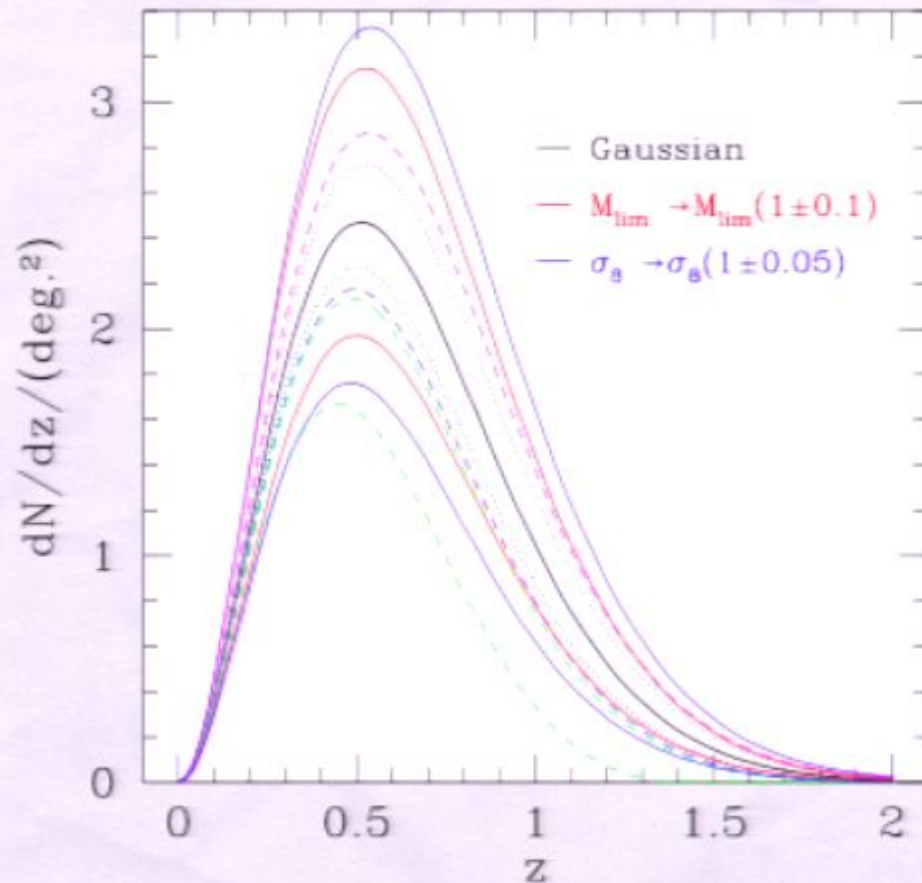
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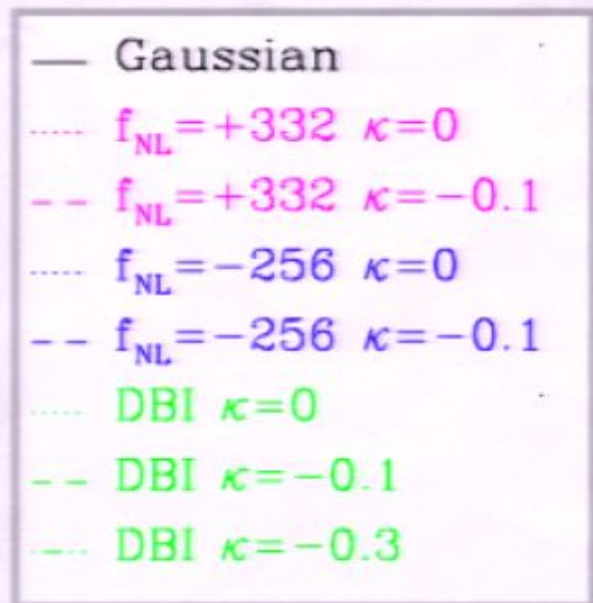
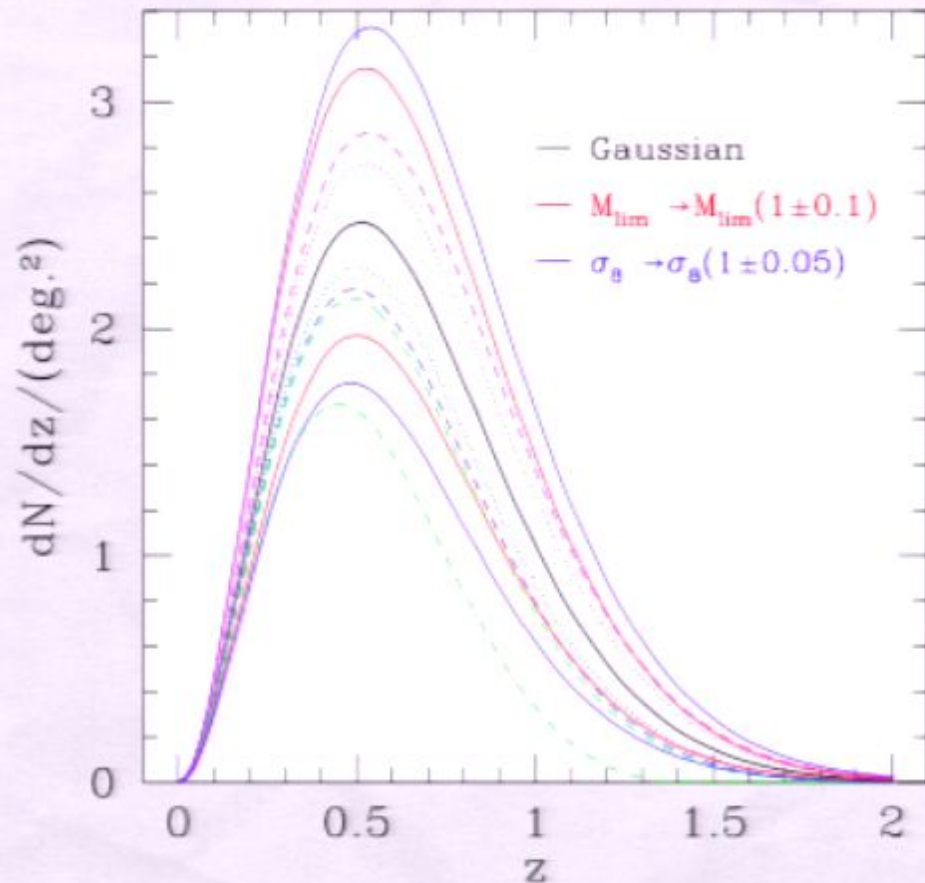
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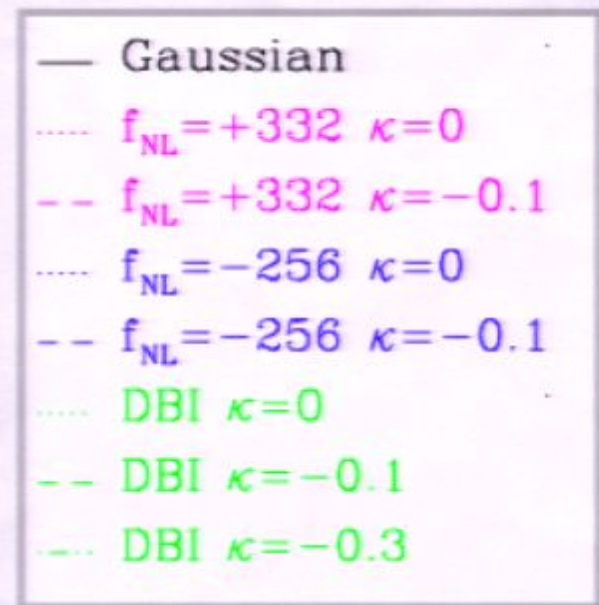
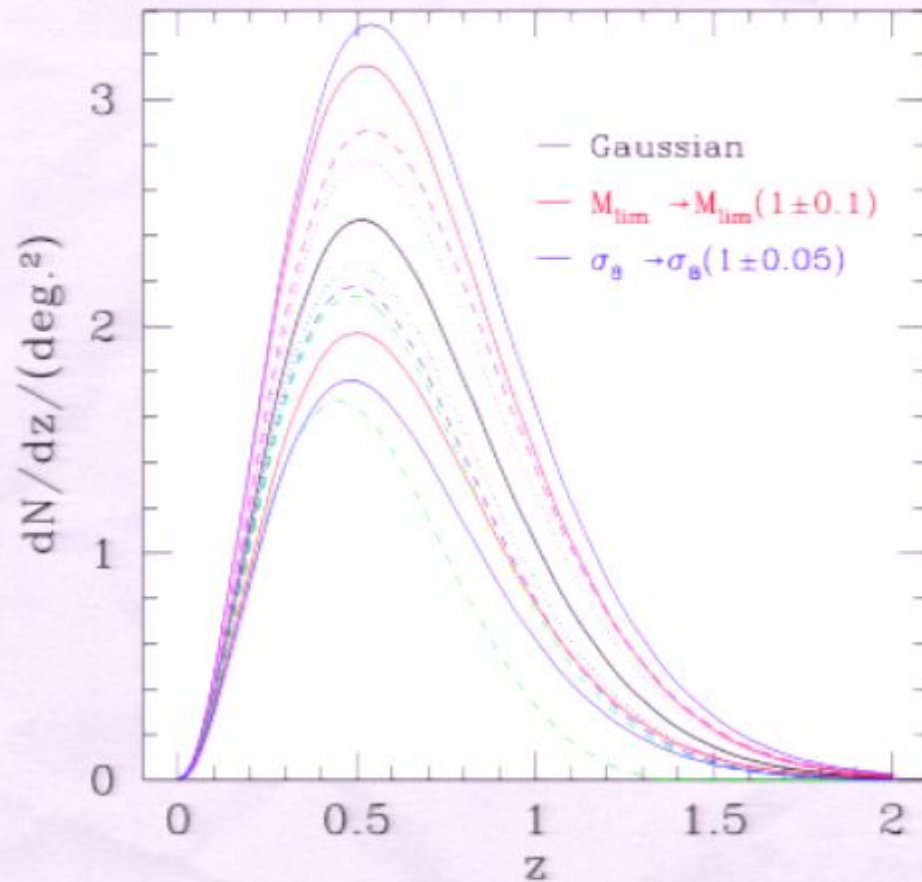
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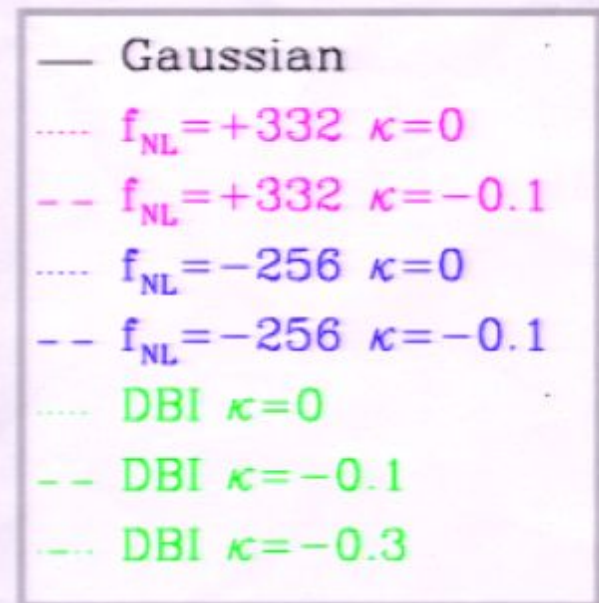
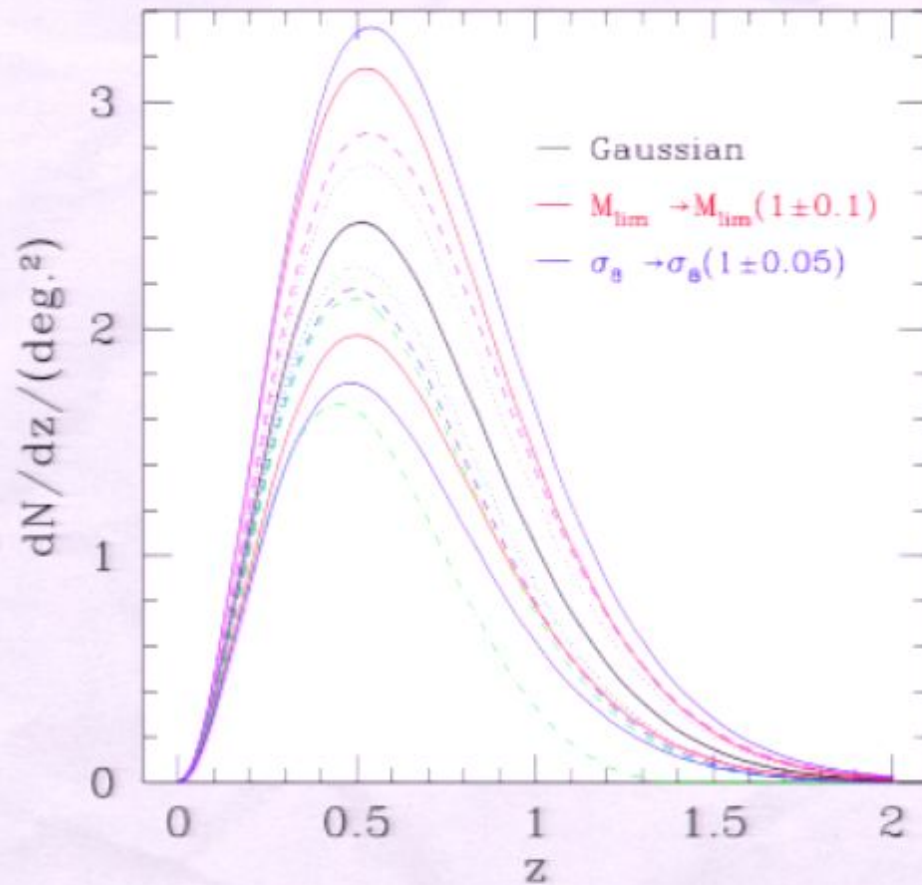
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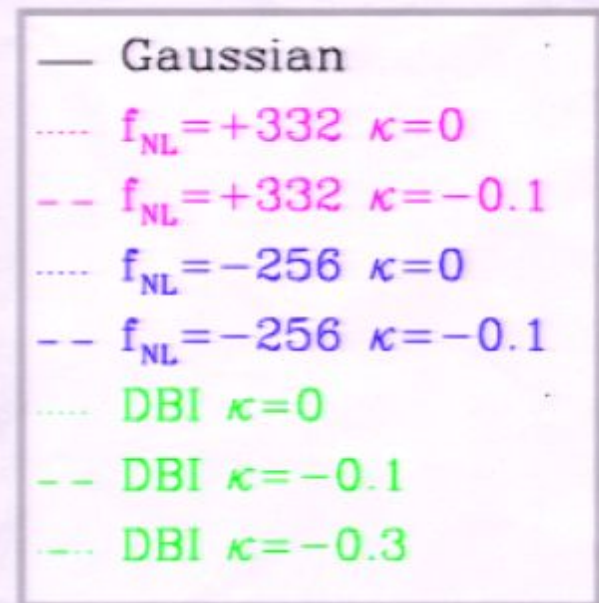
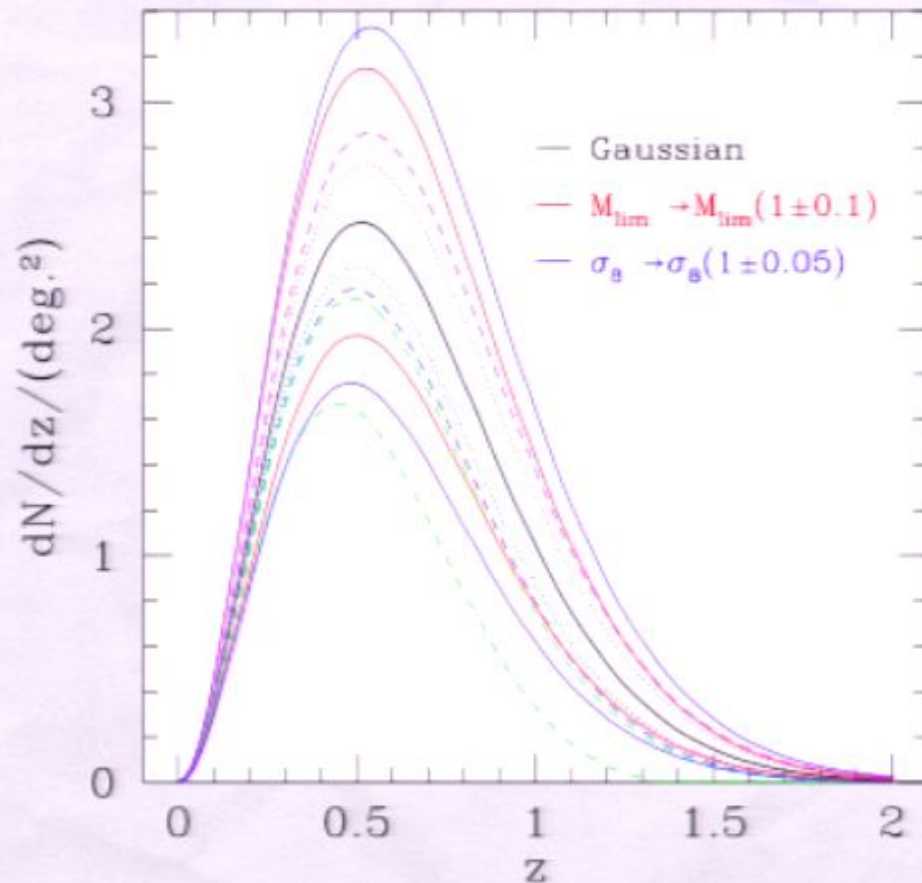
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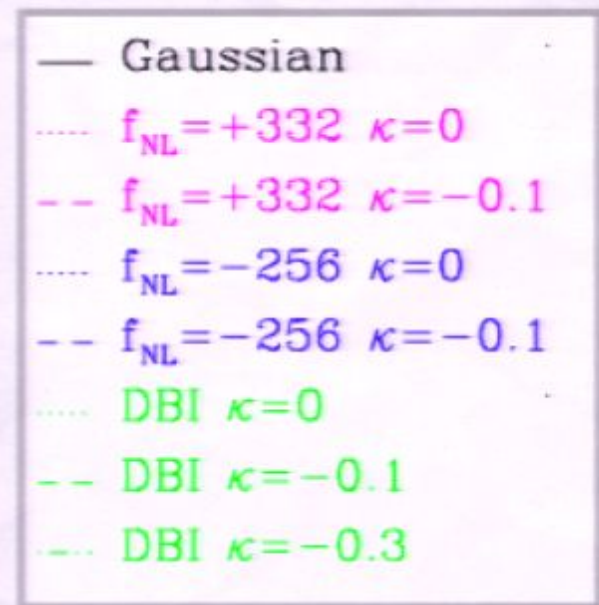
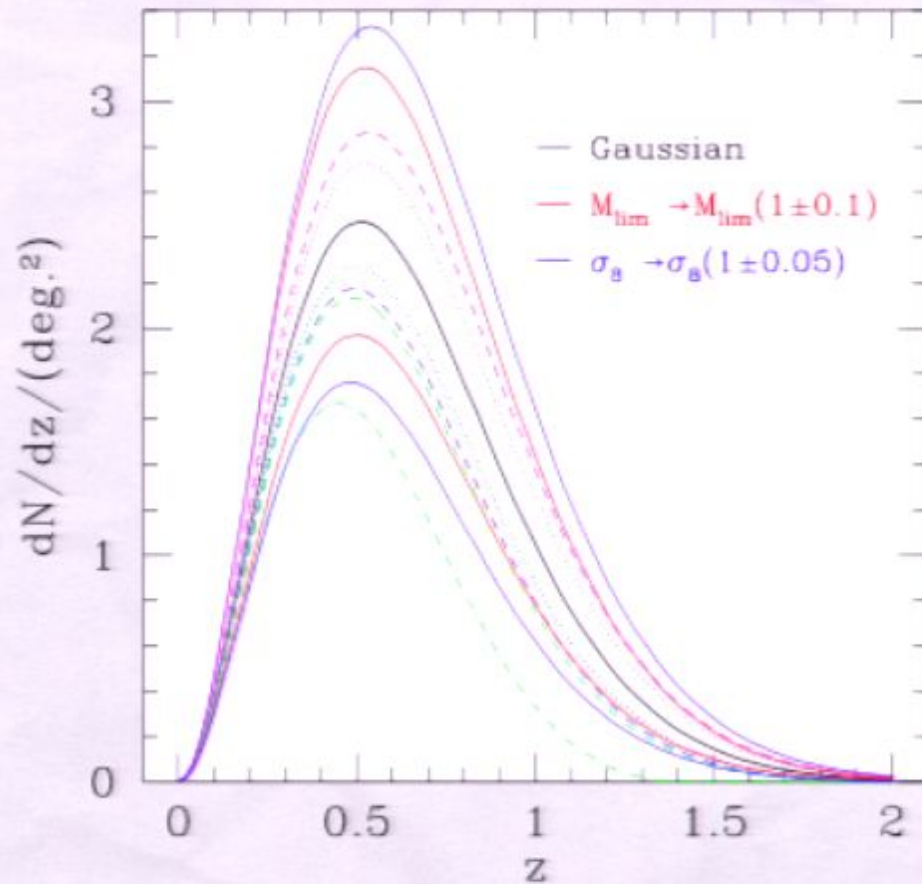
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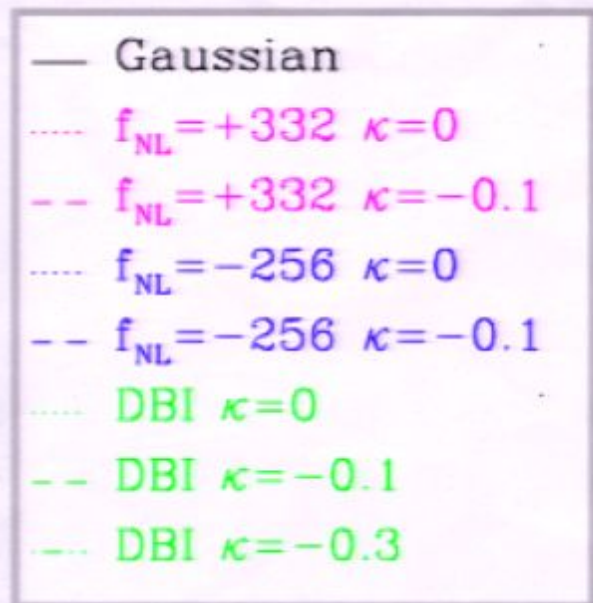
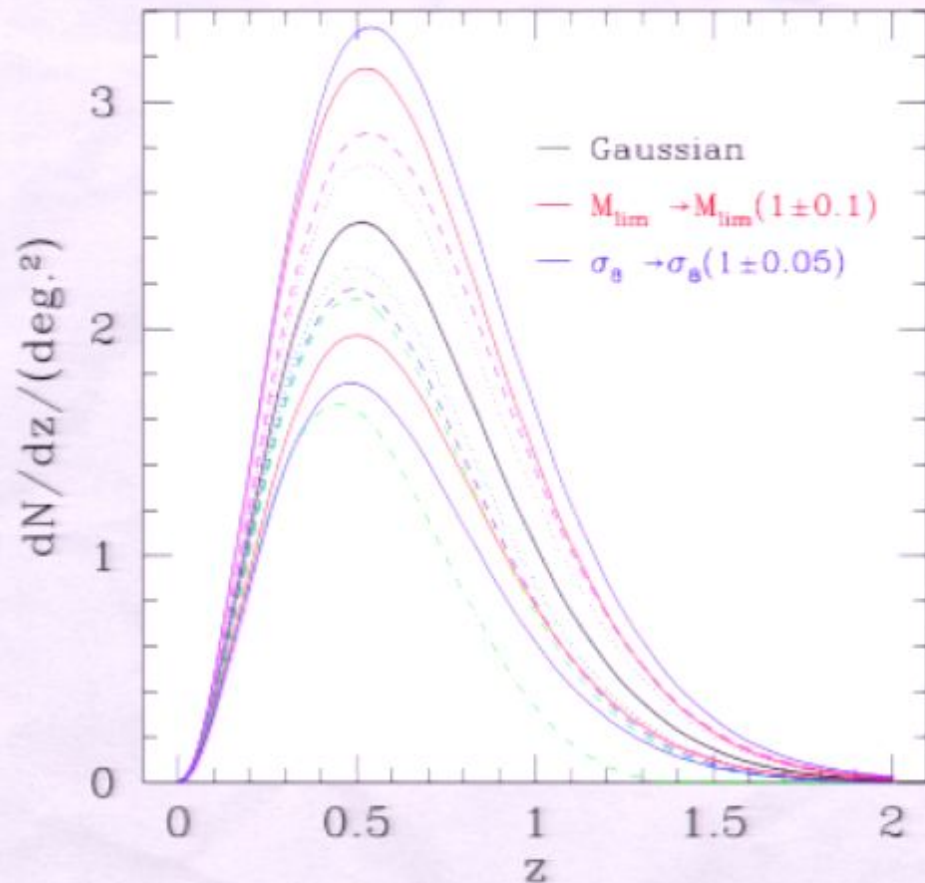
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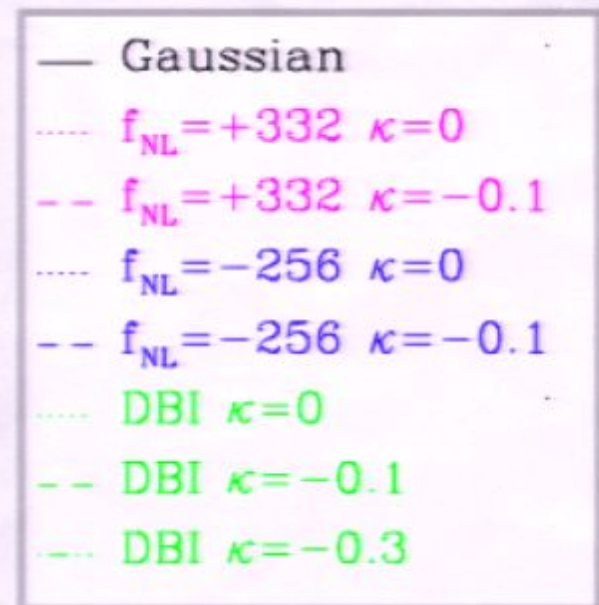
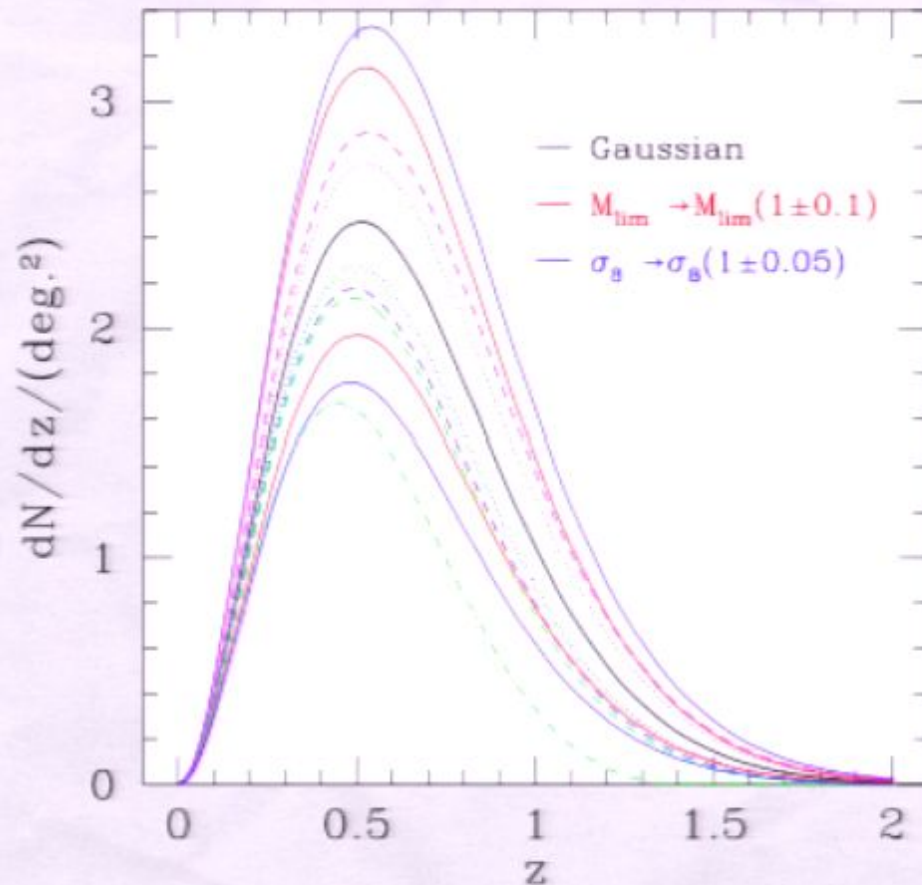
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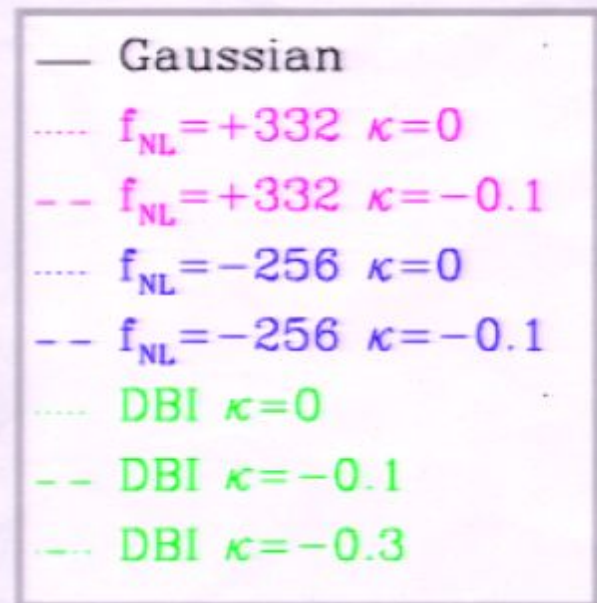
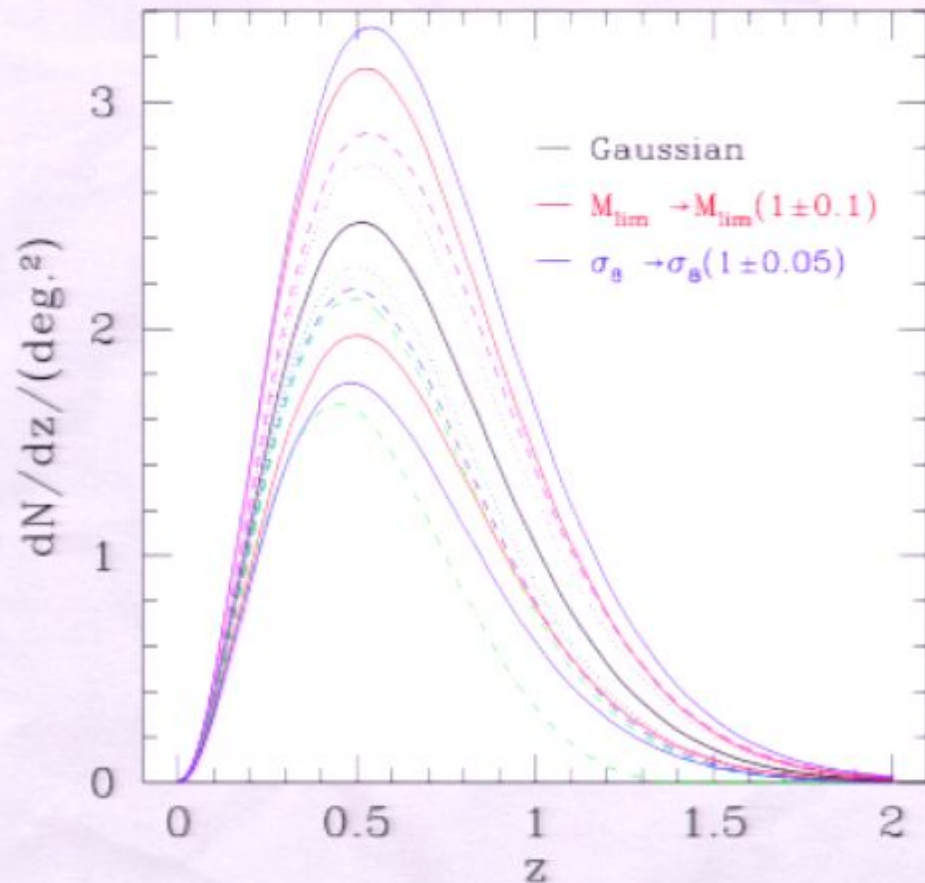
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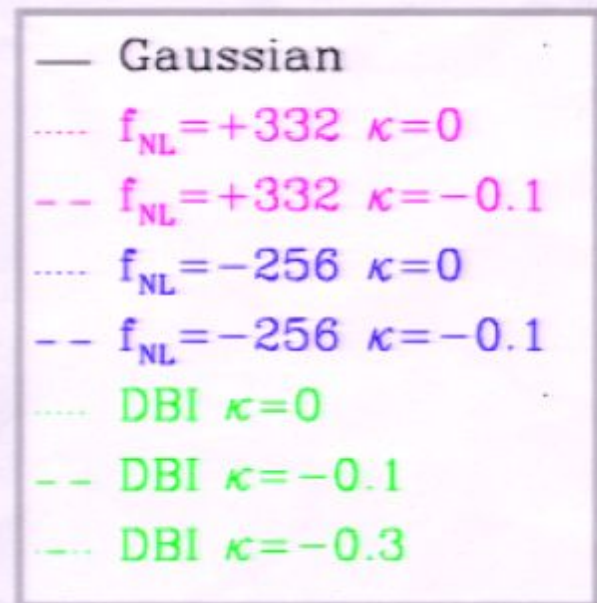
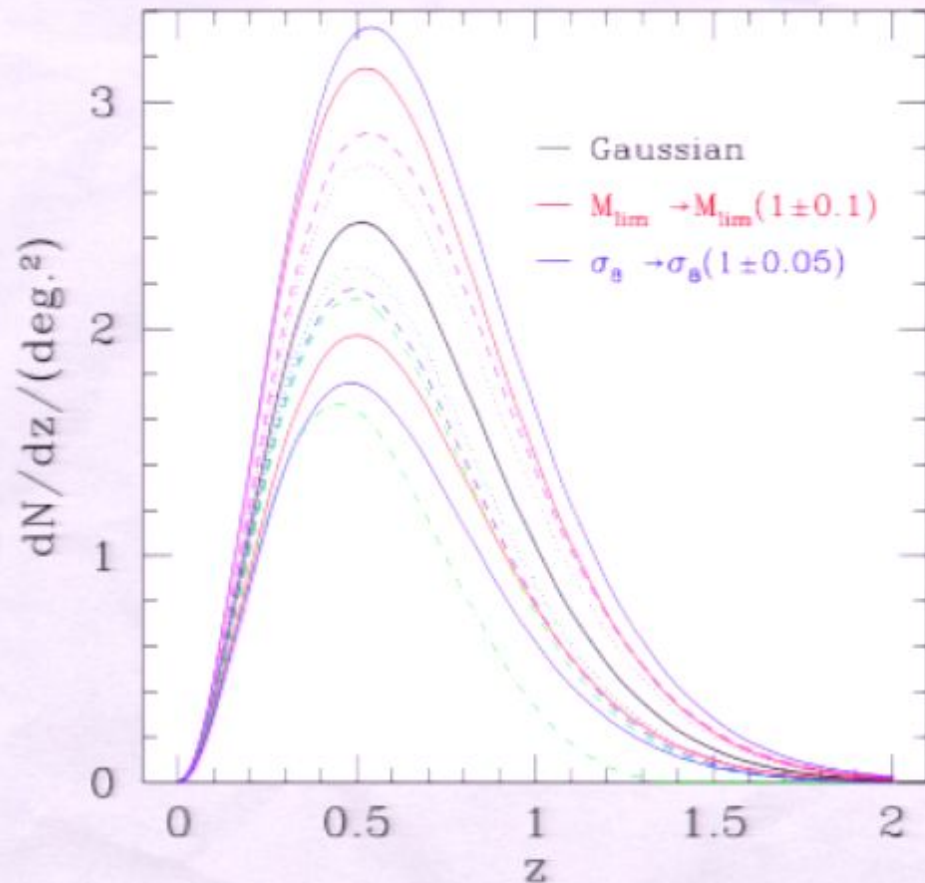
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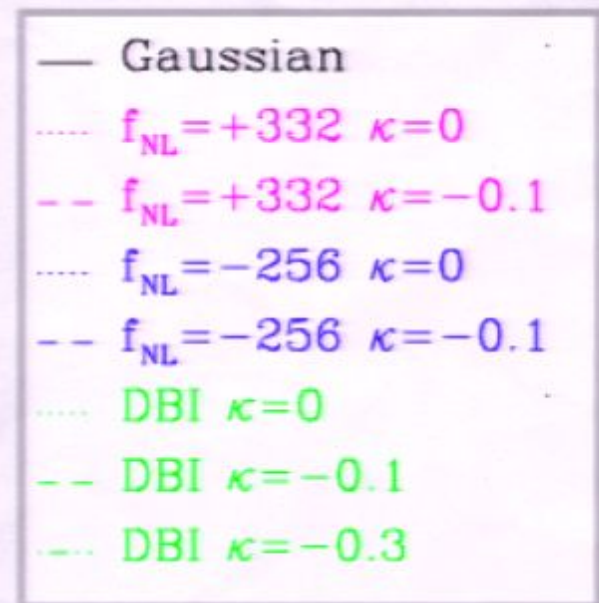
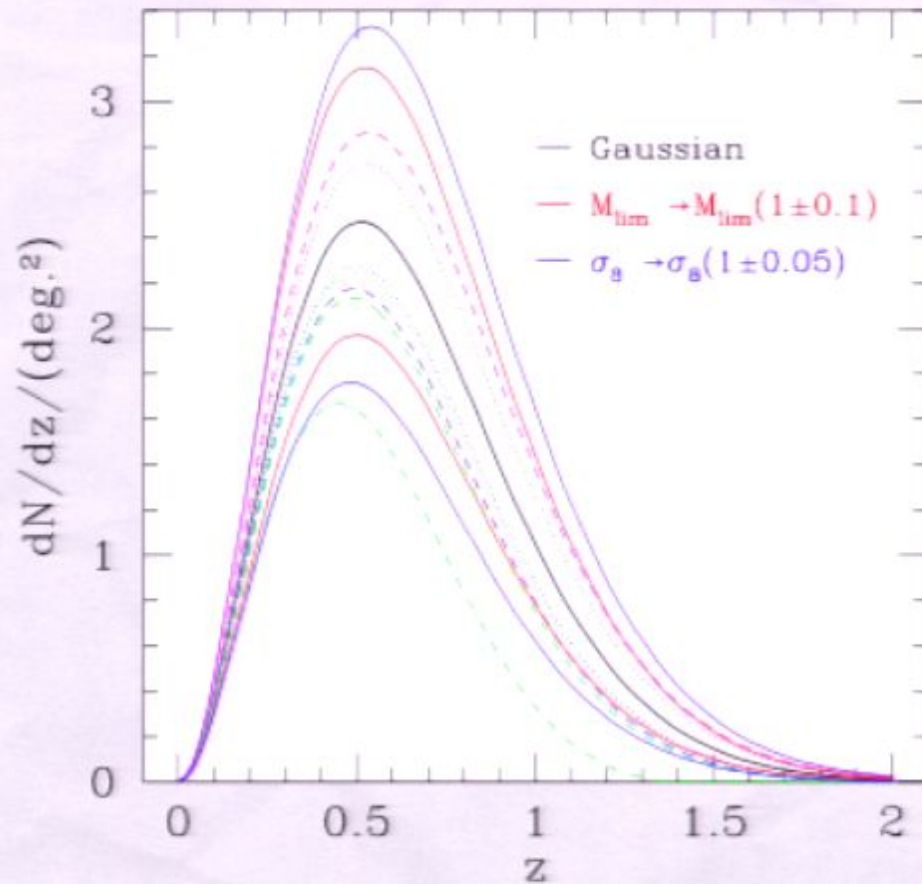
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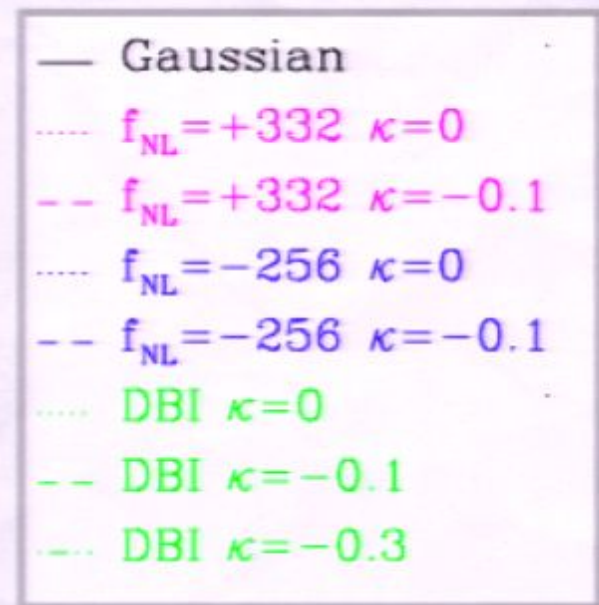
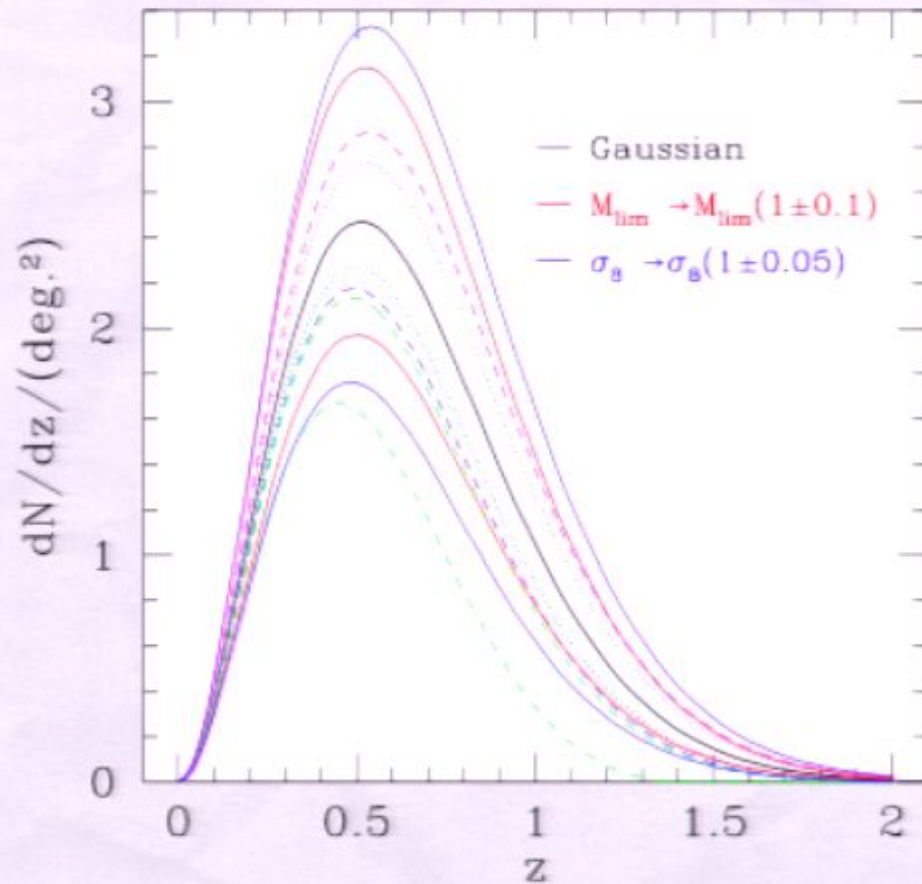
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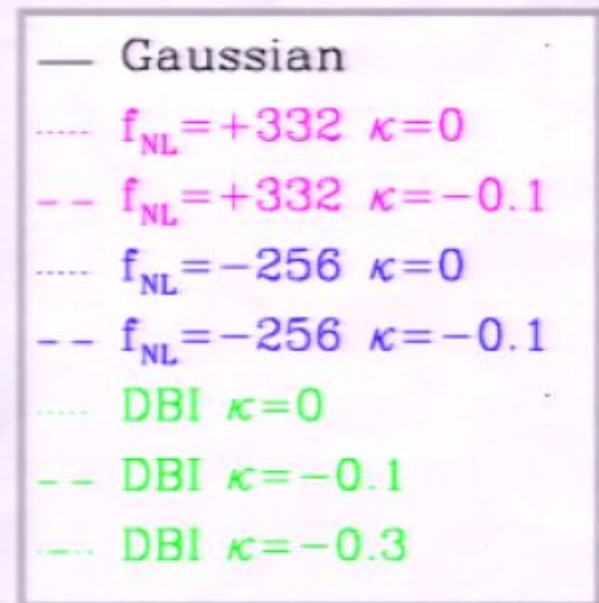
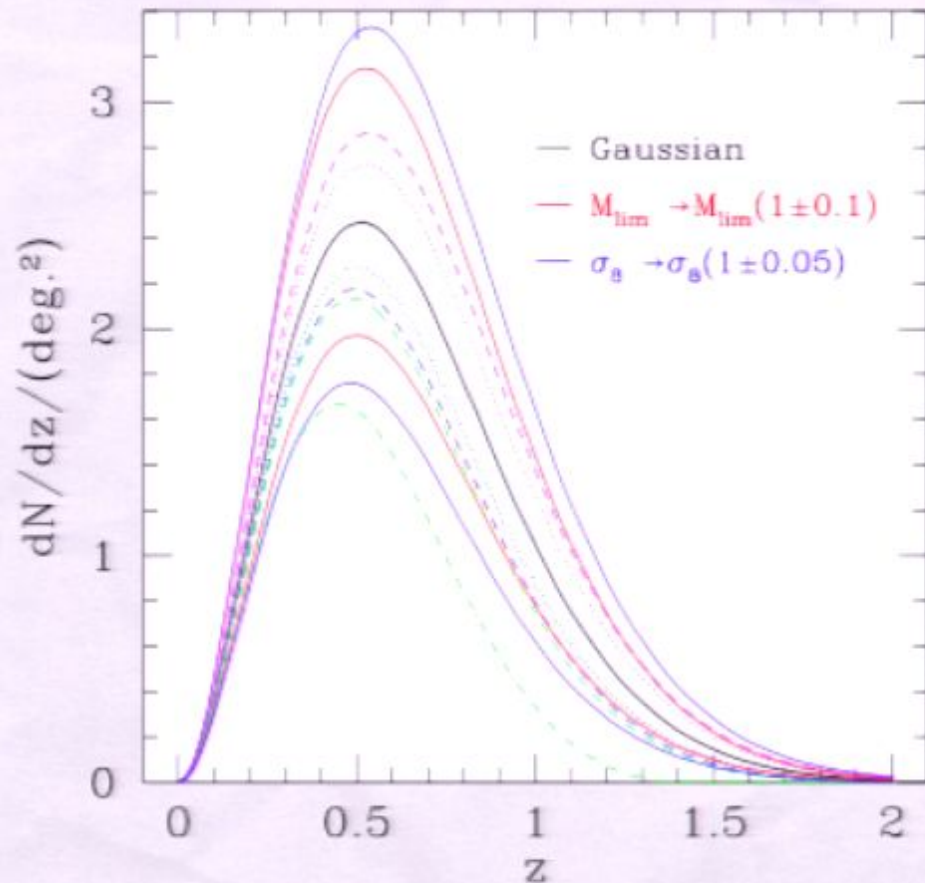
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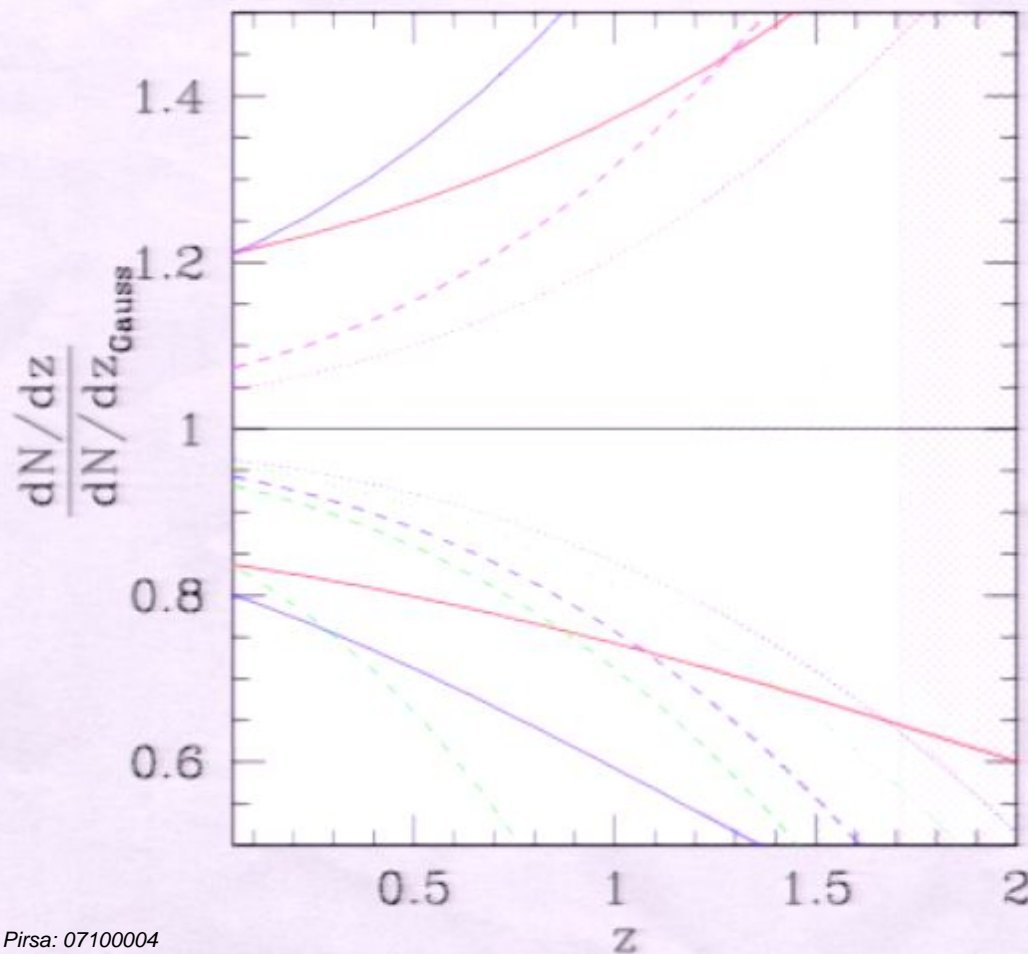
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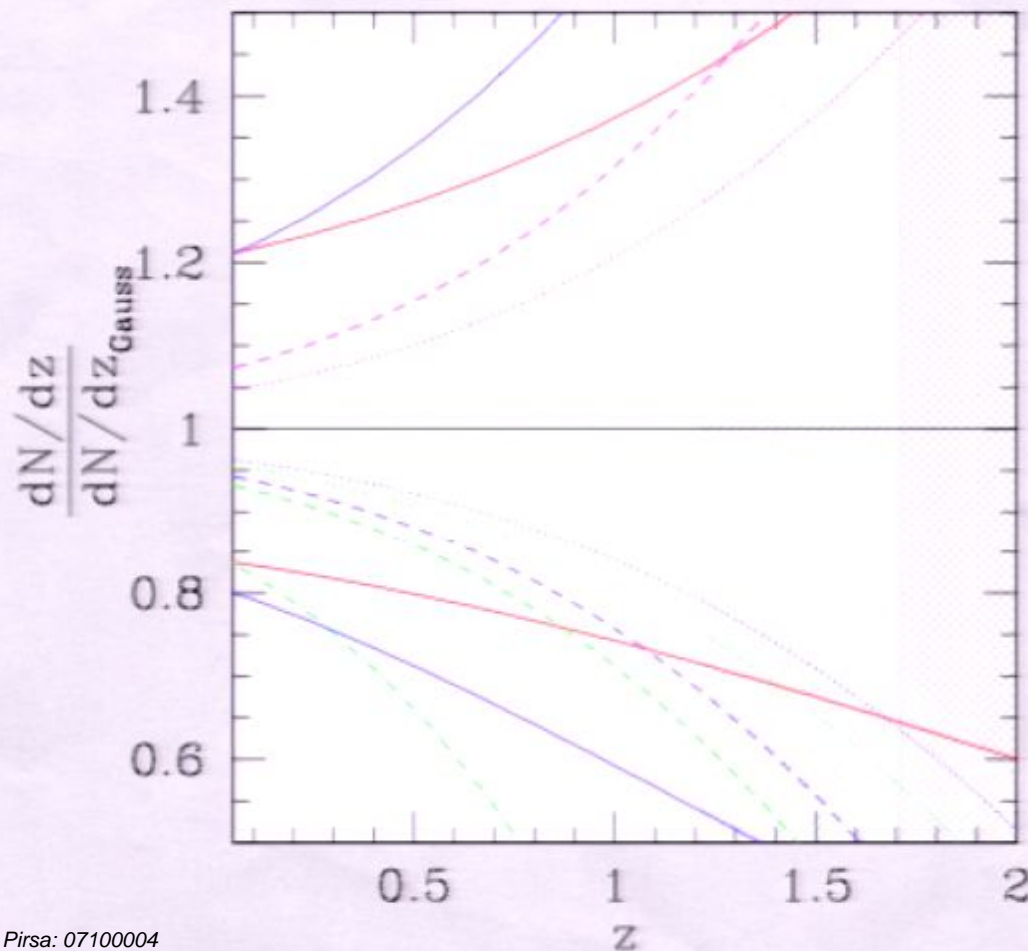


DEVIATION FROM GAUSSIAN



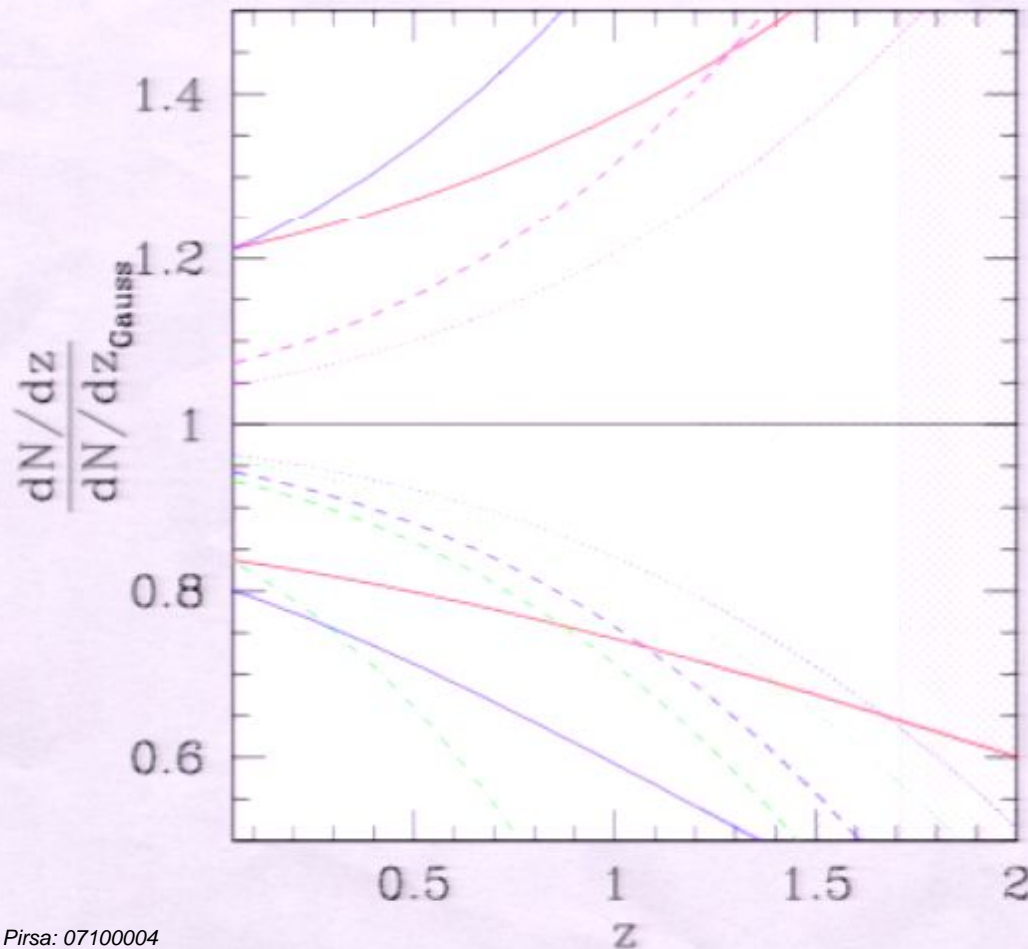
- Gaussian
- $f_{NL} = +332 \quad \kappa = 0$
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- $f_{NL} = -256 \quad \kappa = 0$
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- DBI $\kappa = 0$
- DBI $\kappa = -0.1$
- .- DBI $\kappa = -0.3$
- $M_{lim} \rightarrow M_{lim}(1 \pm 0.1)$
- $\sigma_8 \rightarrow \sigma_8(1 \pm 0.05)$

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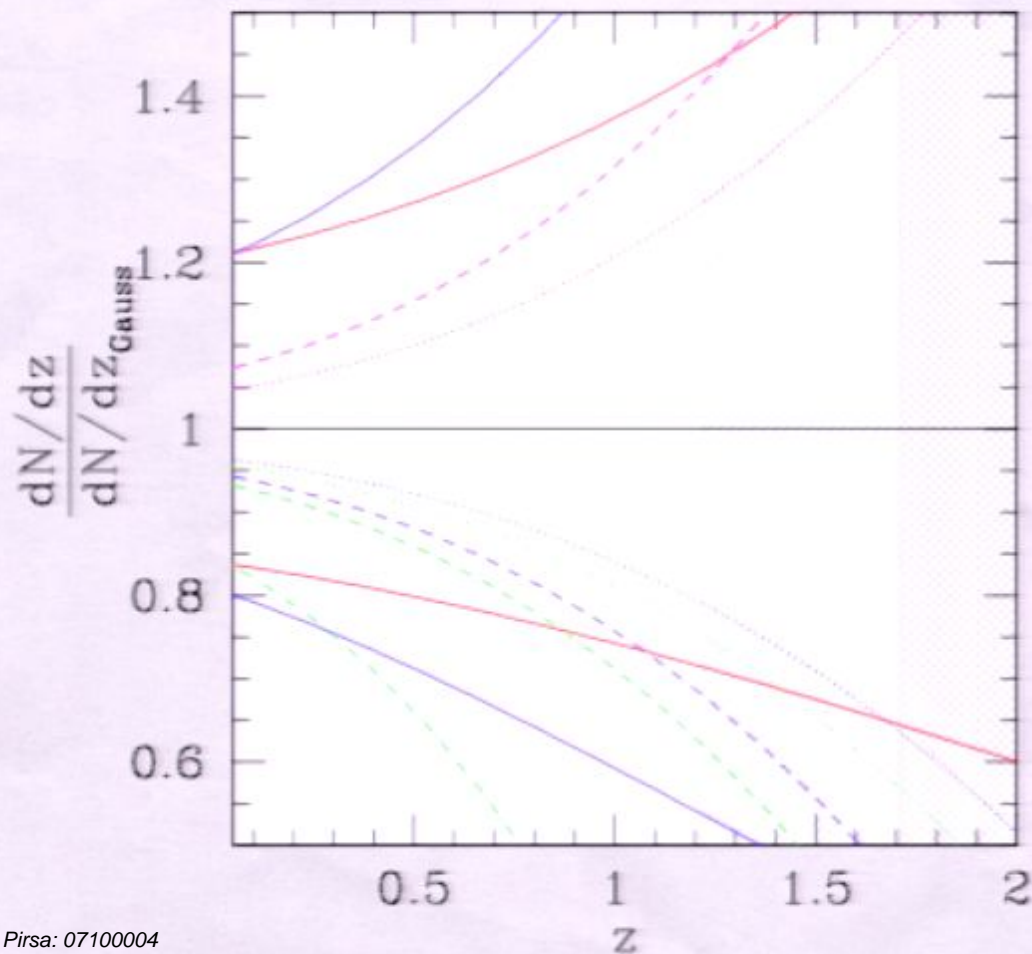
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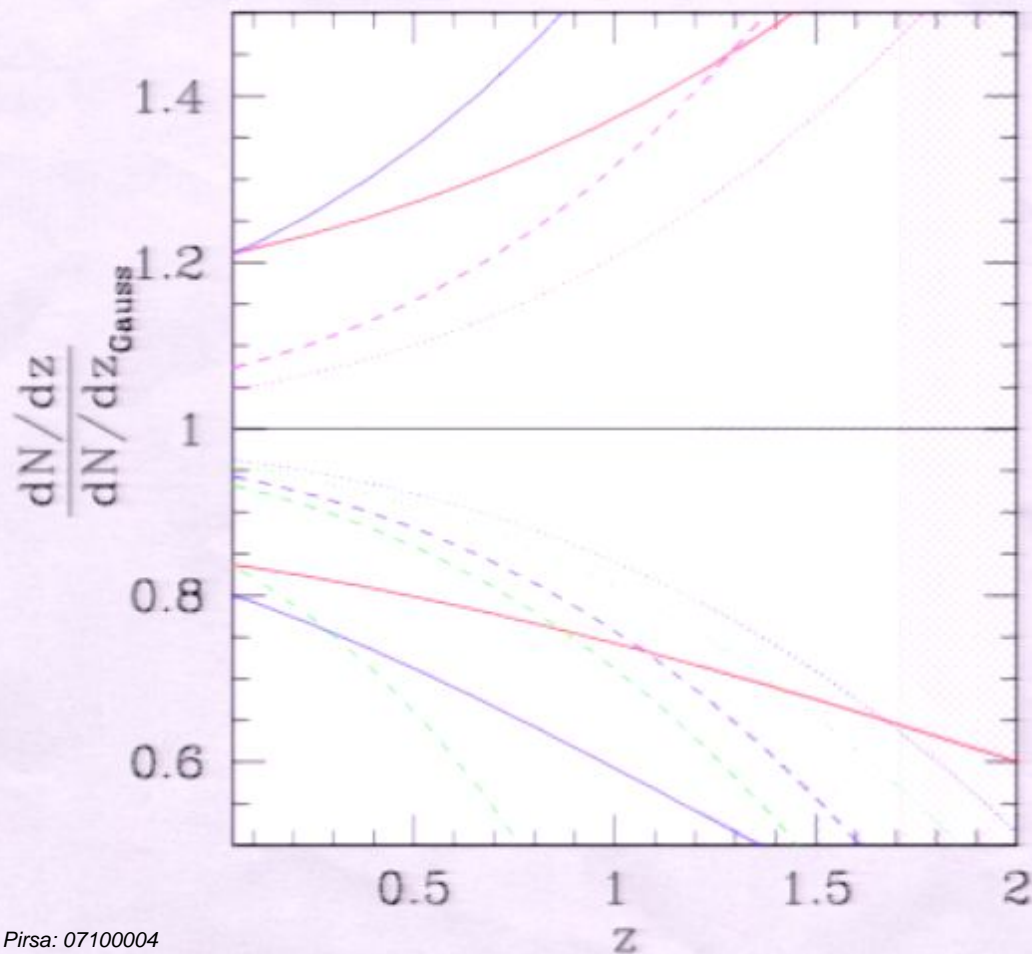


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SO HOW WELL CAN WE DO?

$$f_{NL} = 38, \kappa = 0$$

WMAP Priors				
σ_{Ω_m}	σ_h	σ_{σ_8}	$\sigma_{f_{NL}}$	σ_{κ}
0.0080	0.029	0.027	150	1.86
Planck Priors				
0.0058	0.011	0.014	40	1.07

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σ_{Ω_m}	σ_h	σ_{σ_8}	$\sigma_{f_{NL}}$	σ_{κ}
0.0071	0.029	0.025	150	0.17
Planck Priors				
0.0051	0.011	0.013	40	0.09

$$f_{NL} = 332, \kappa = 0$$

WMAP Priors				
σ_{Ω_m}	σ_h	σ_{σ_8}	$\sigma_{f_{NL}}$	σ_{κ}
0.010	0.029	0.032	150	0.42
Planck Priors				
0.0066	0.011	0.015	40	0.21

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$$(\Omega_m = 0.24, h = 0.73, \sigma_8 = 0.77, \kappa = 0)$$

SO HOW WELL CAN WE DO?

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0.0080	0.029	0.027	150	1.86
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- Where there are several contributions:

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$$\langle \delta(\vec{k}_1) \delta(\vec{k}_2) \delta(\vec{k}_3) \rangle = (2\pi)^3 \delta(\vec{k}_1 + \vec{k}_2 + \vec{k}_3) B(k_1, k_2, k_3)$$

- Where there are several contributions:

$$B_{123}^{(0)} = B_{123}^I + B_{123}^G + \int d^3q F_2(k_1 + k_2 - q, -q) P_4^I(k_1, k_2, k_1 + k_2 + q, q)$$

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Induced

Linearly evolved
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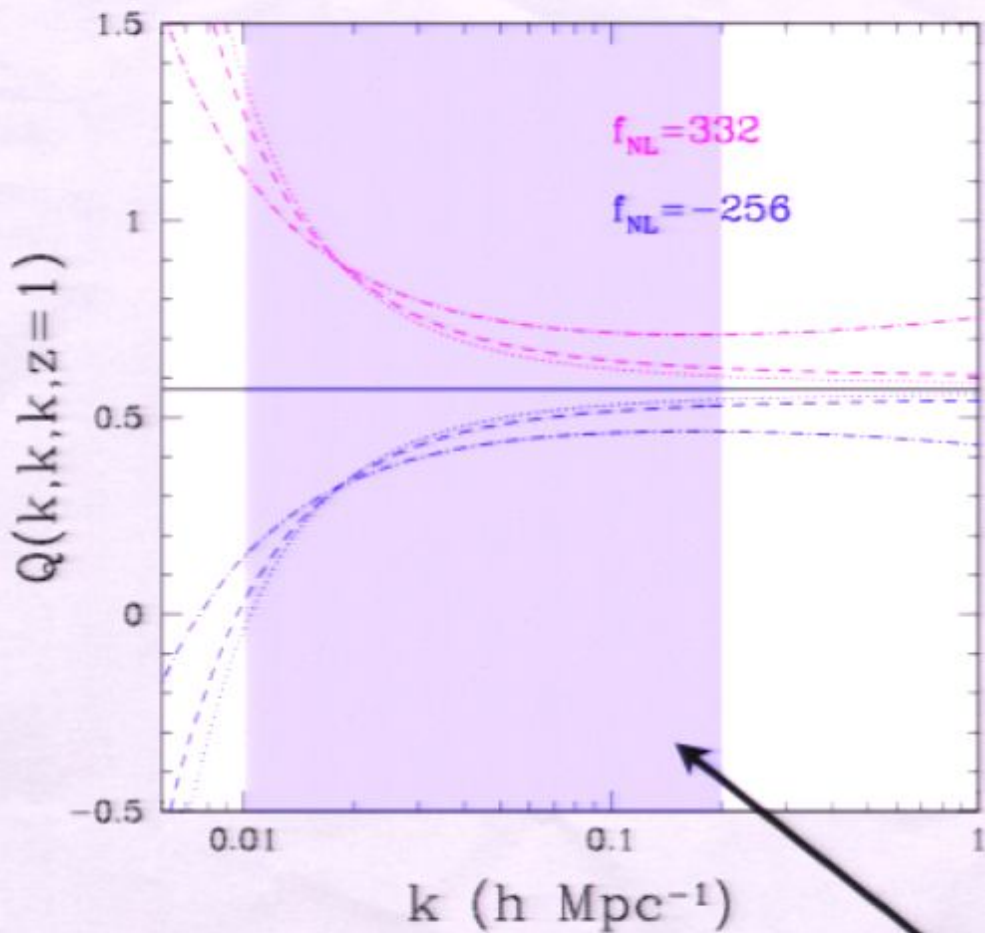
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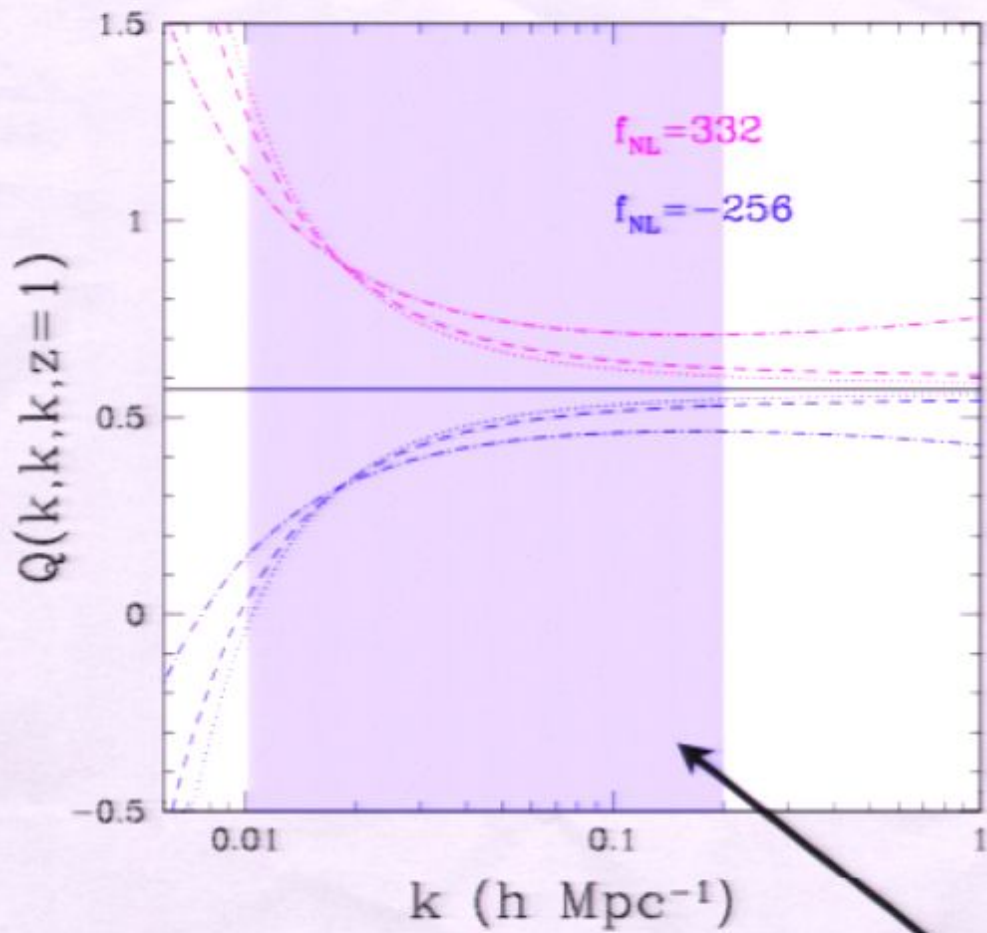
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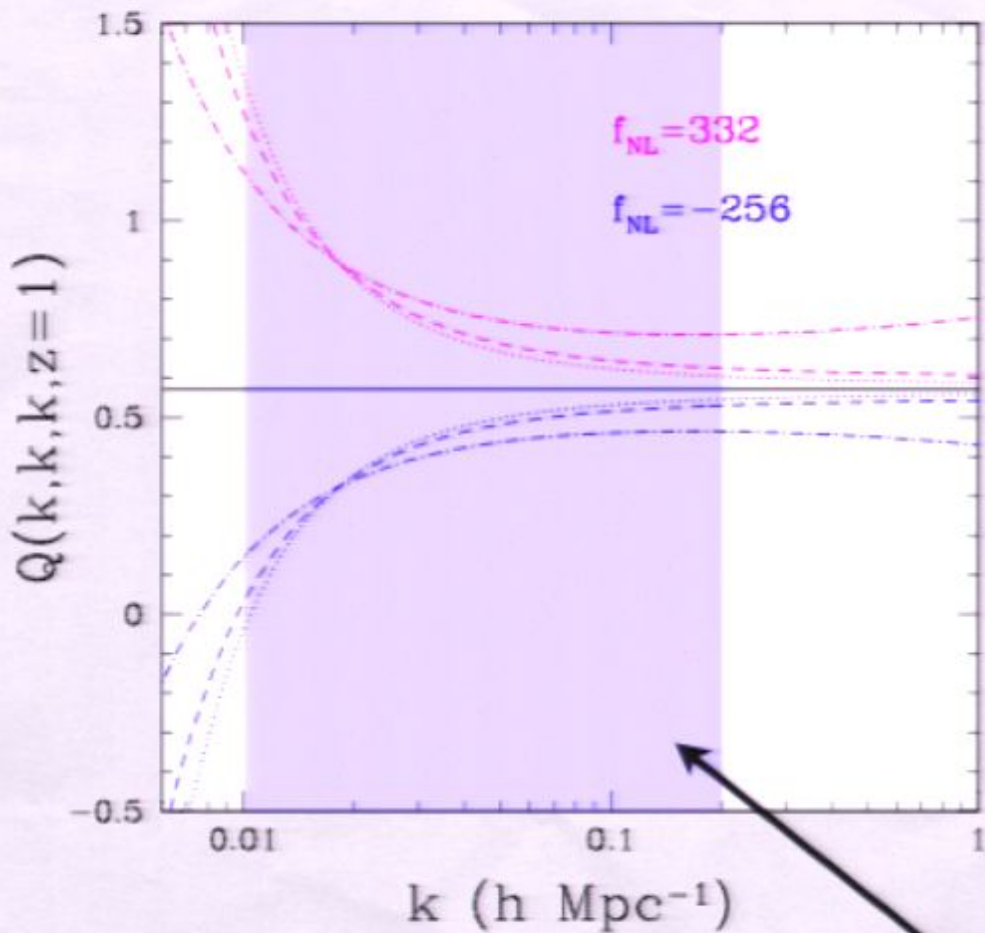
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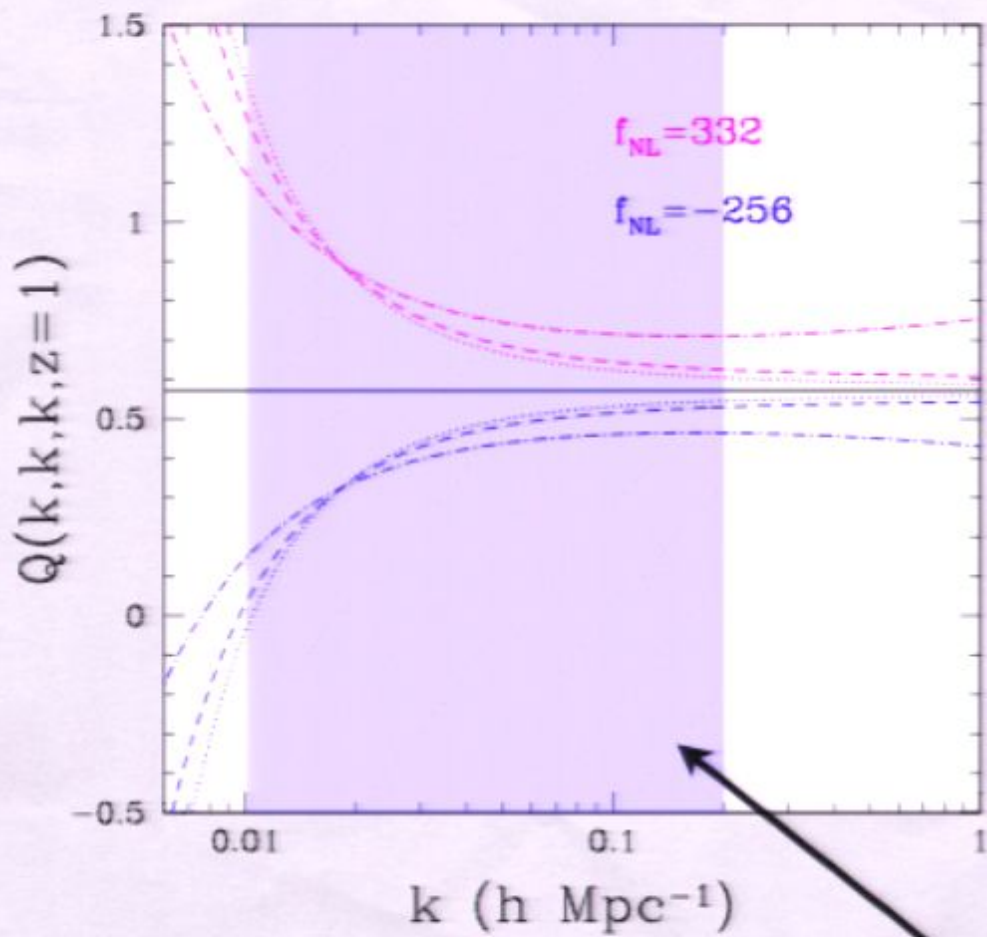
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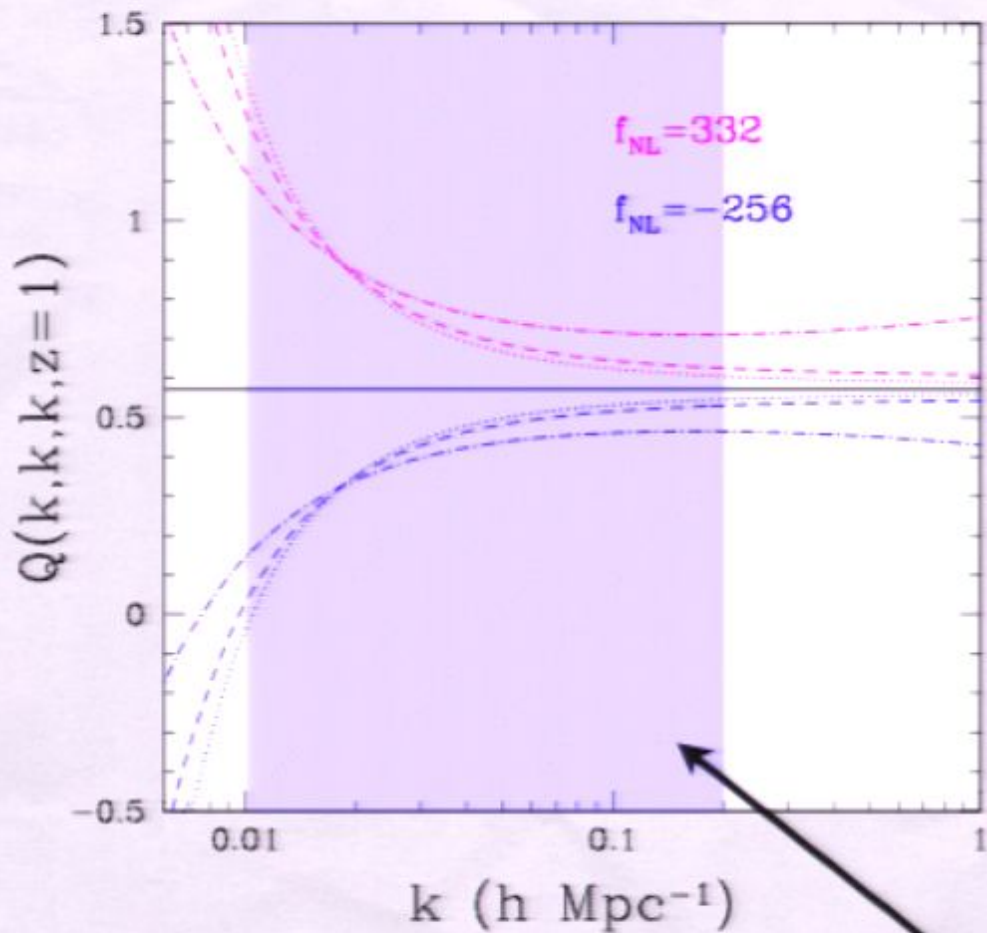
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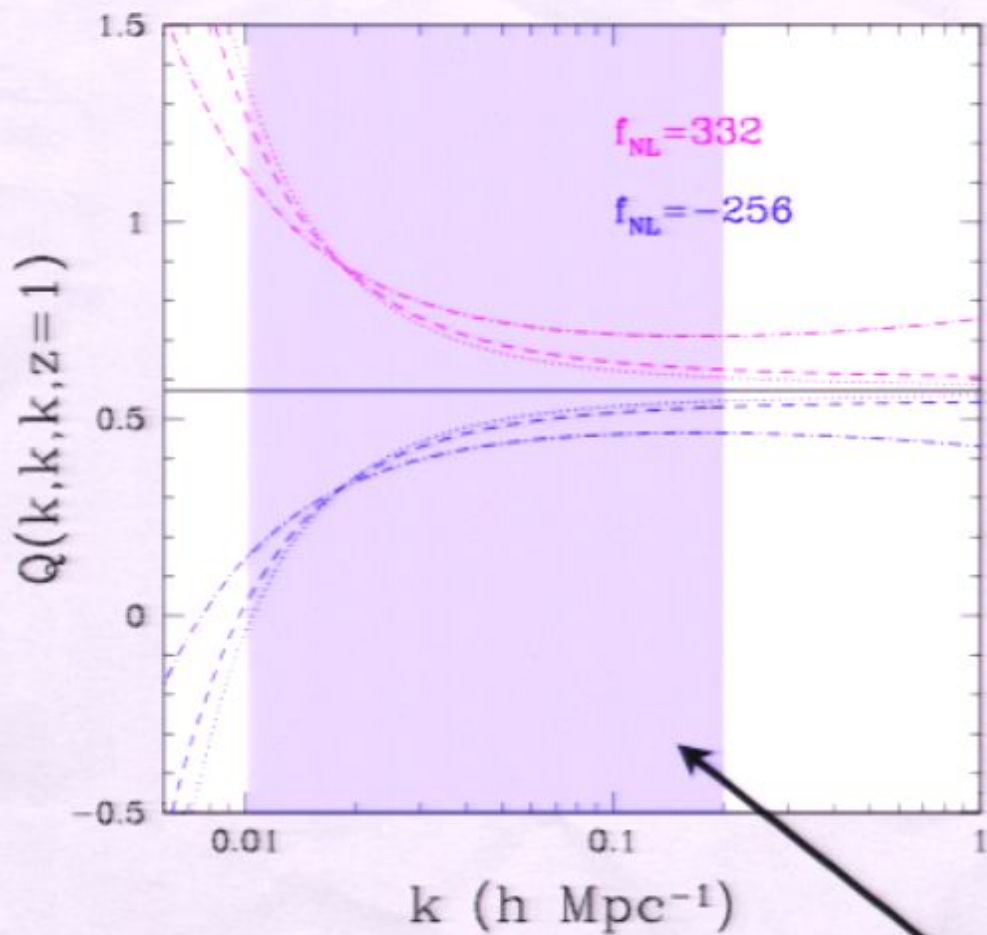
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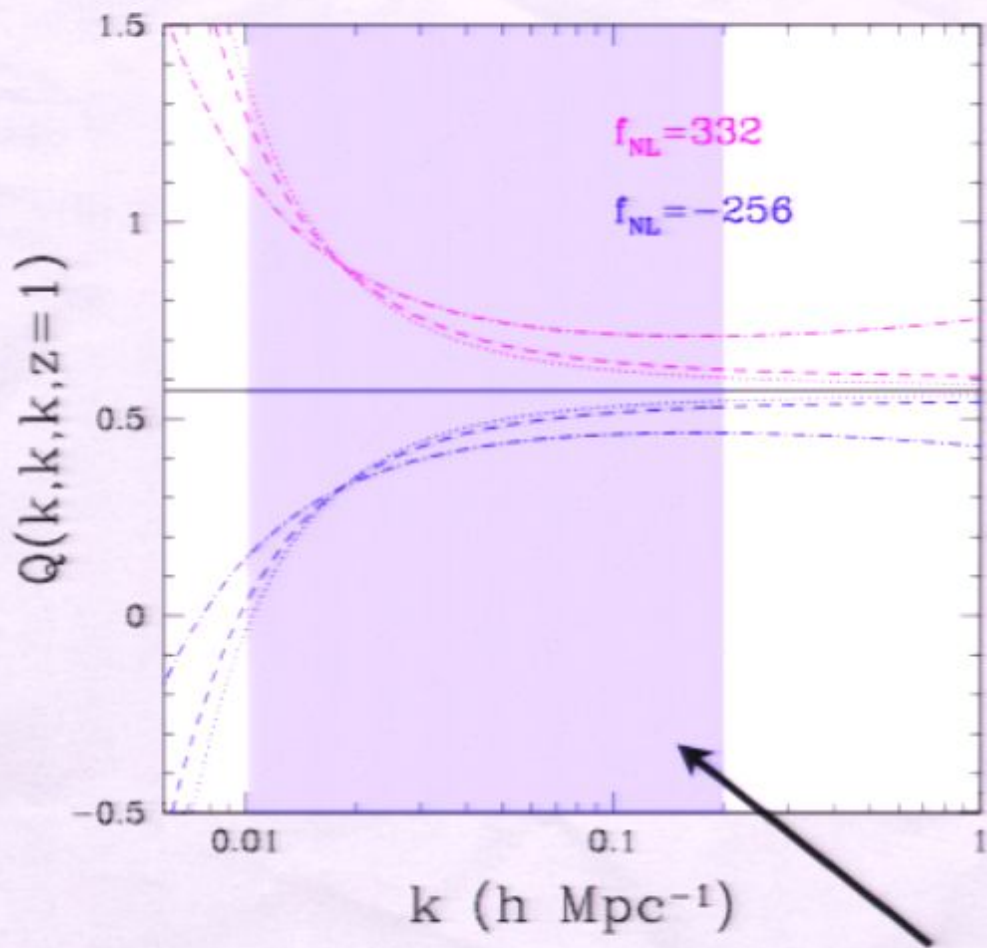
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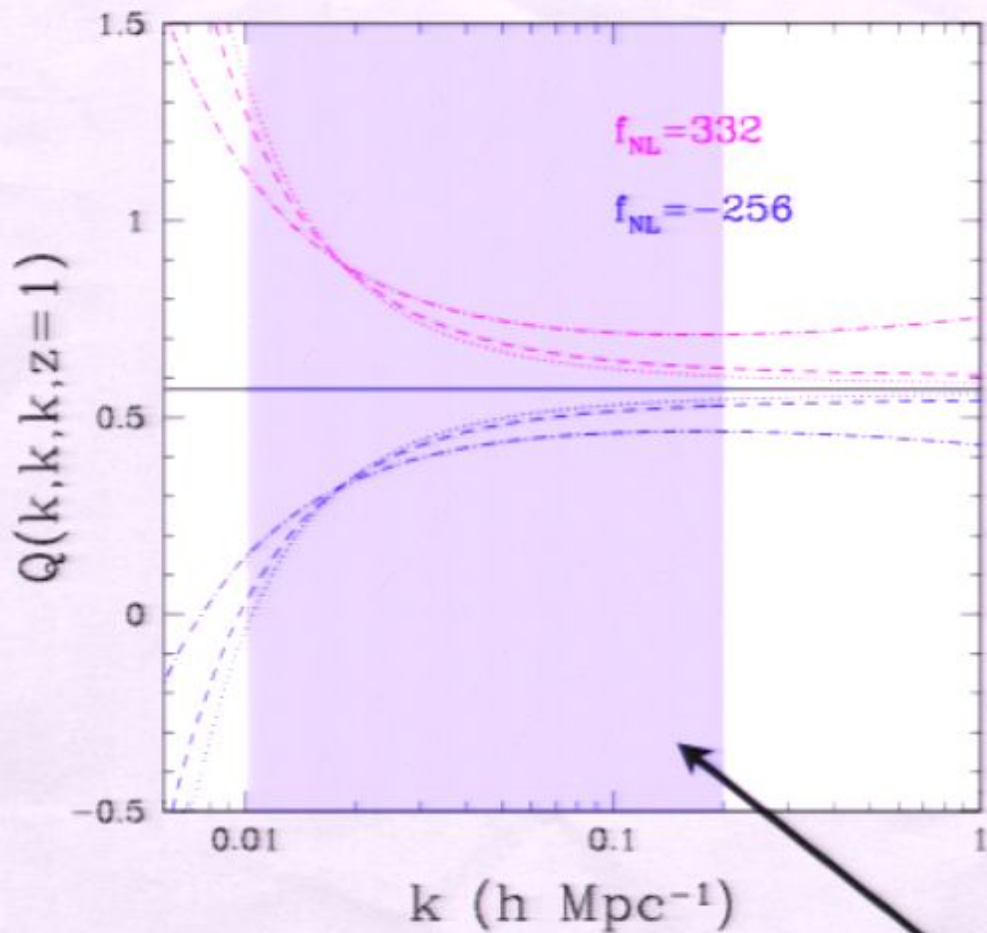


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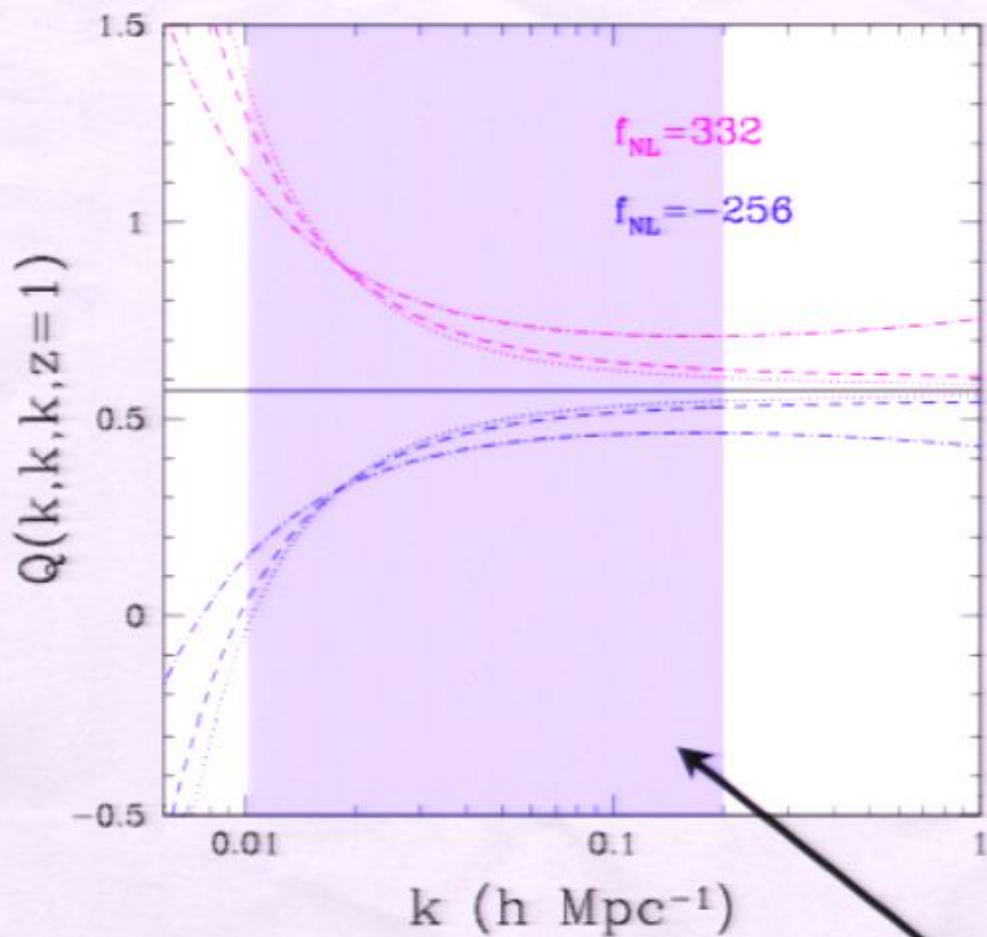
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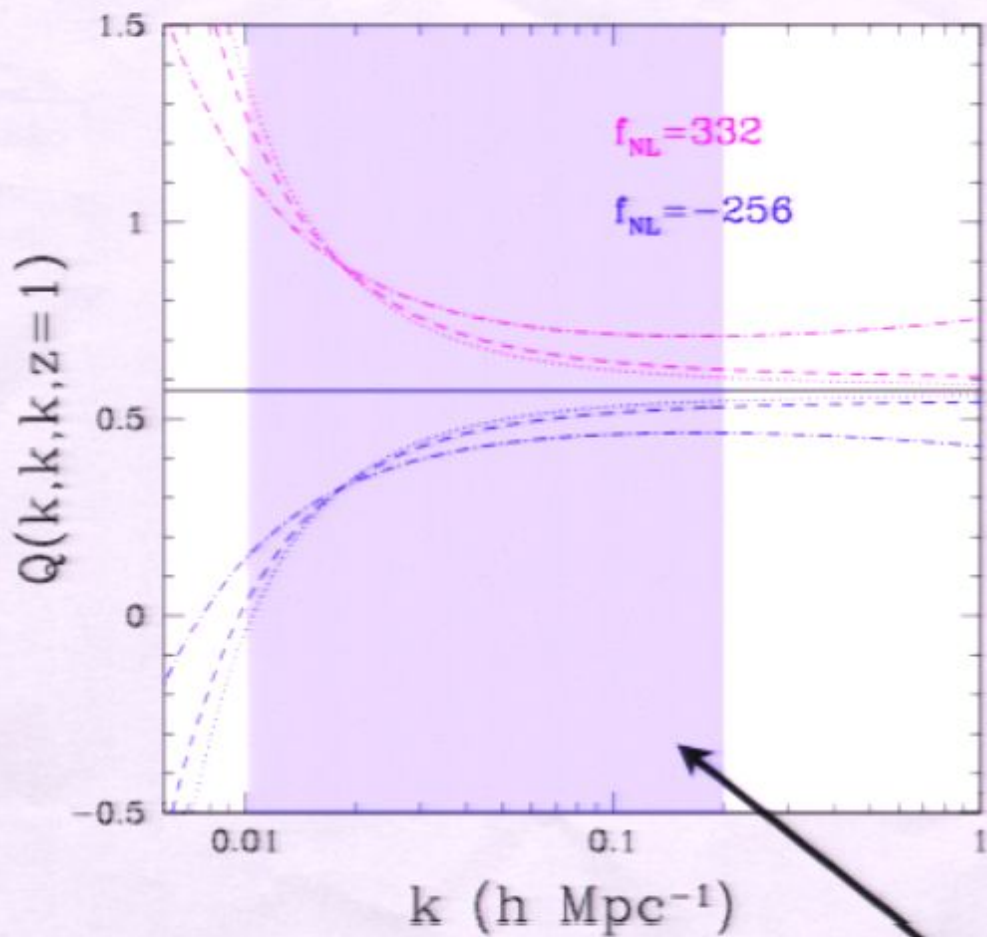
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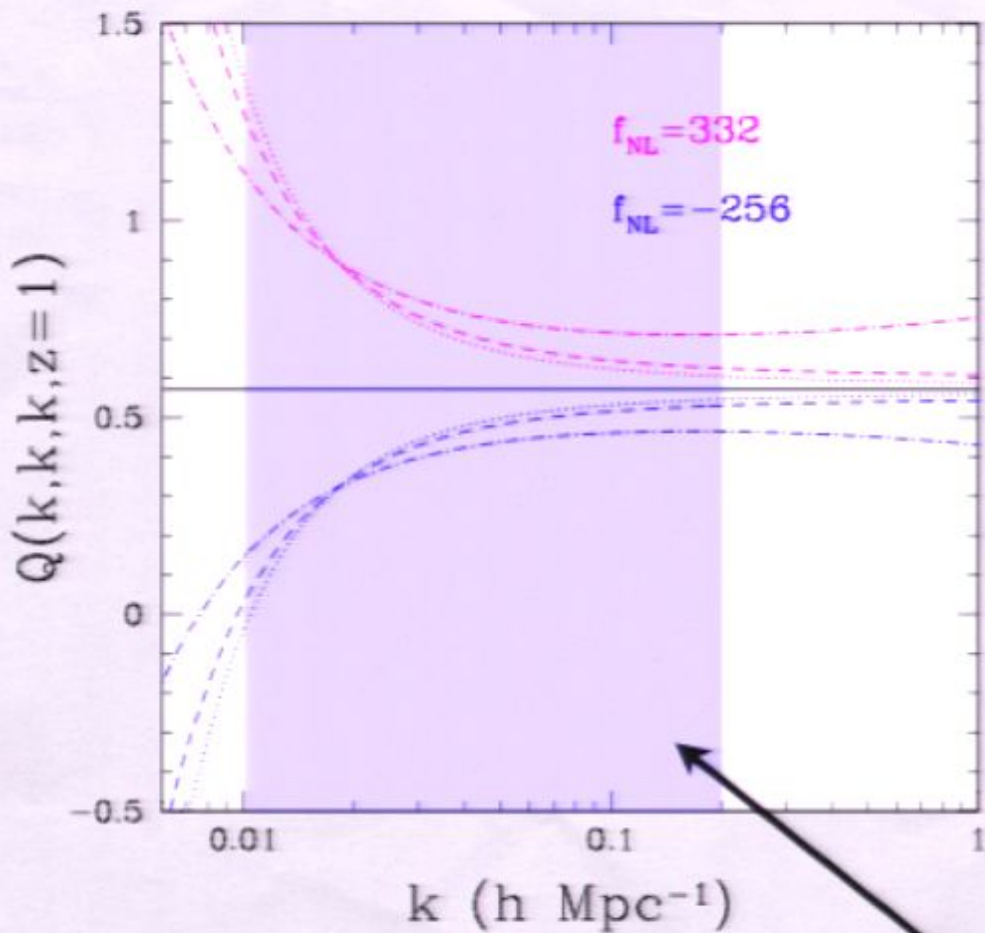
REDUCED BISPECTRUM



$$f_{NL}(k) = f_{NL,CMB}(k/k_{CMB})^{-2\kappa}$$

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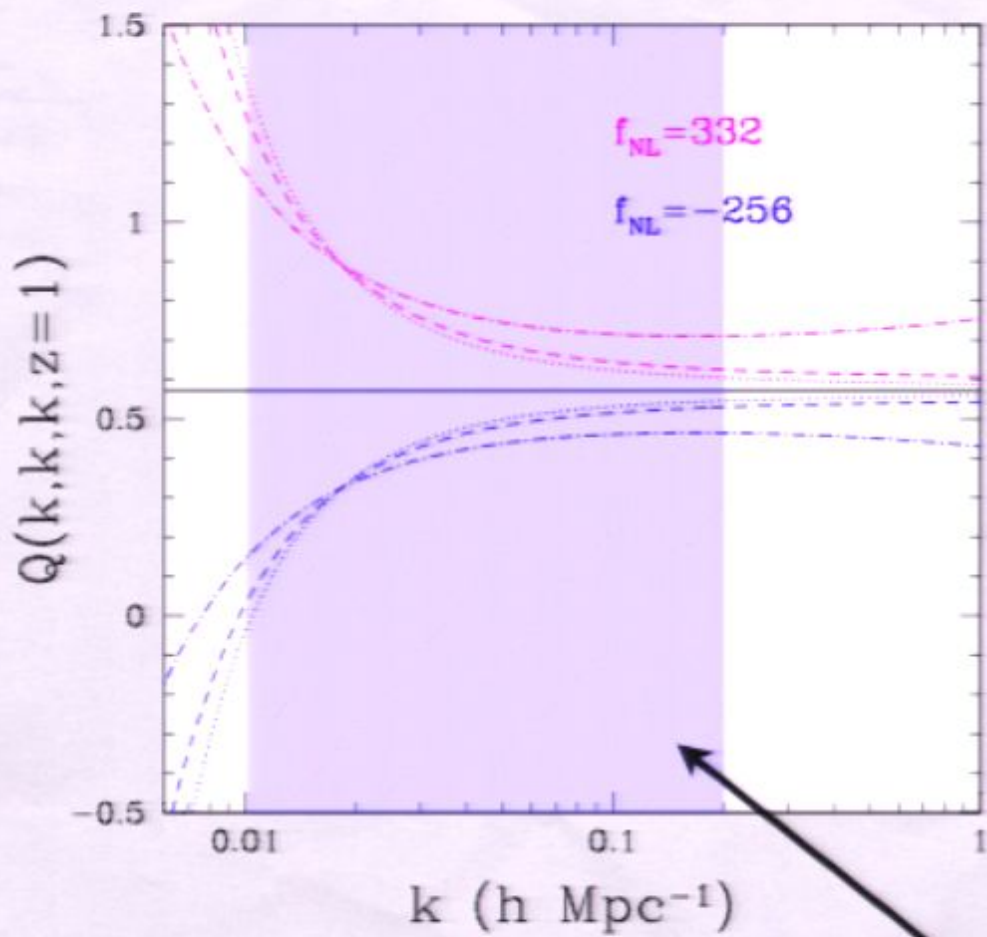
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REDUCED BISPECTRUM

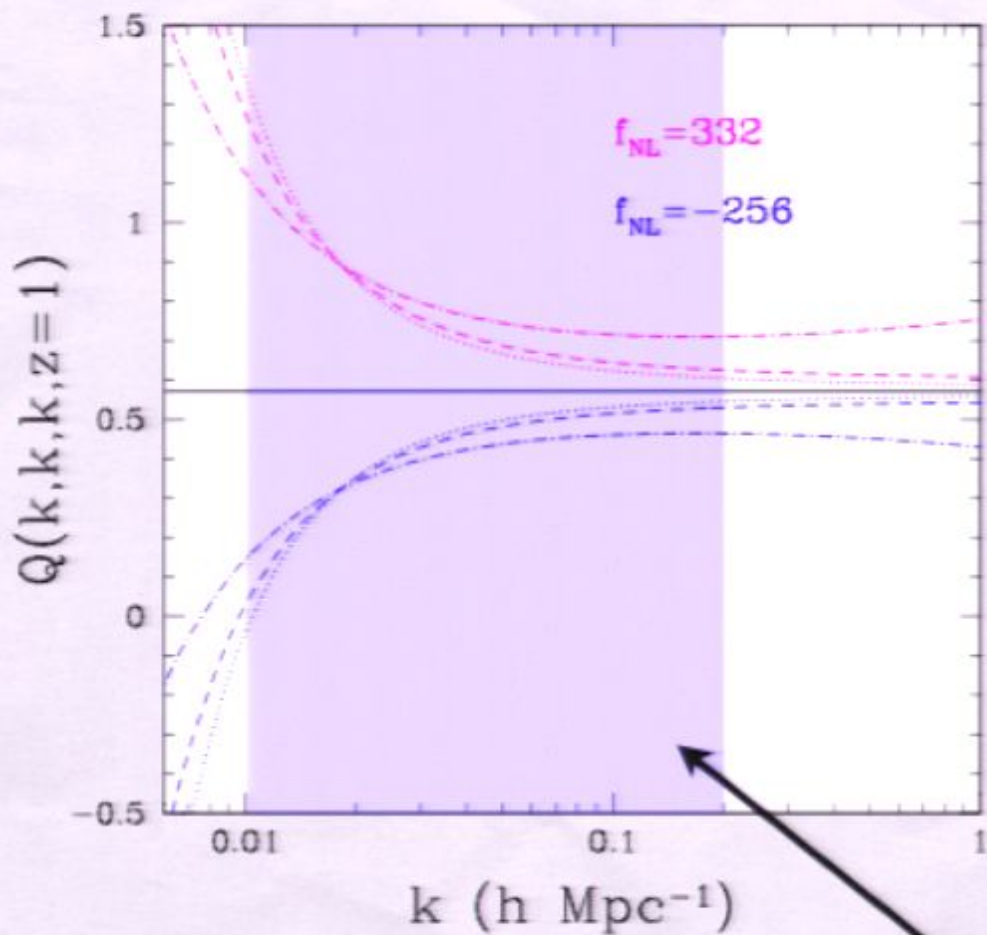


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Most accessible range

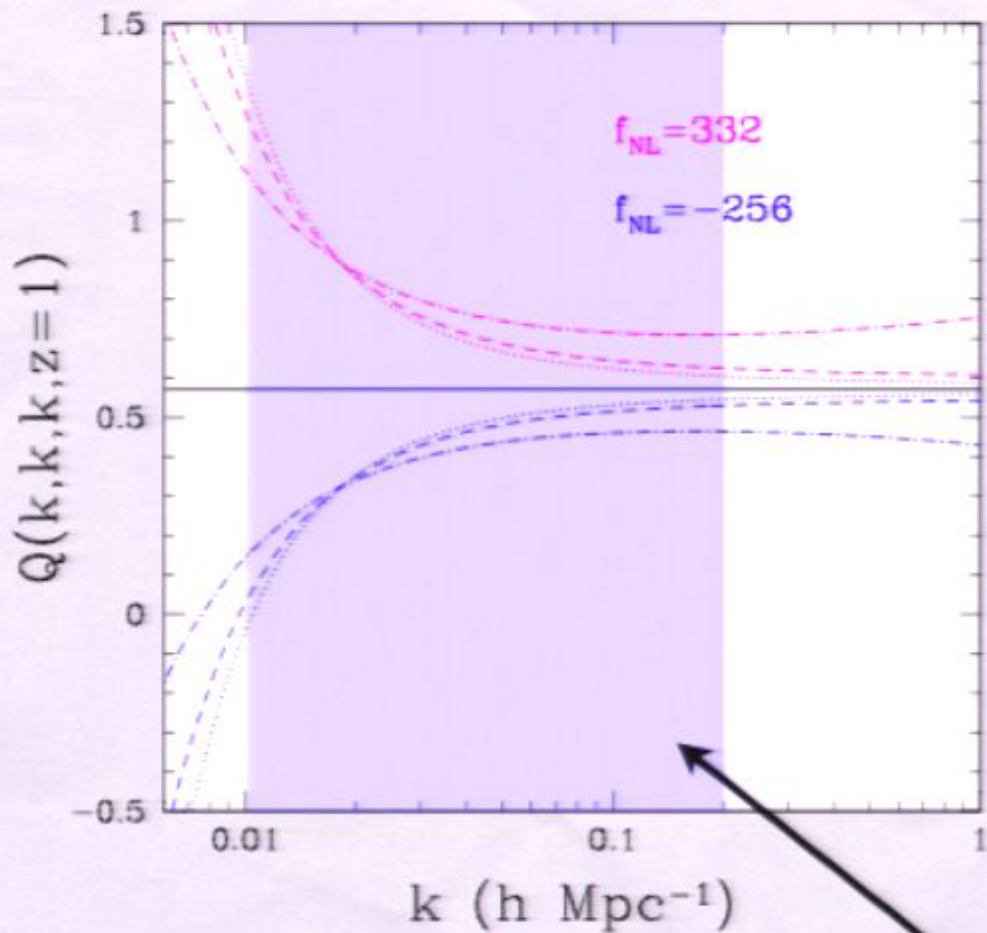
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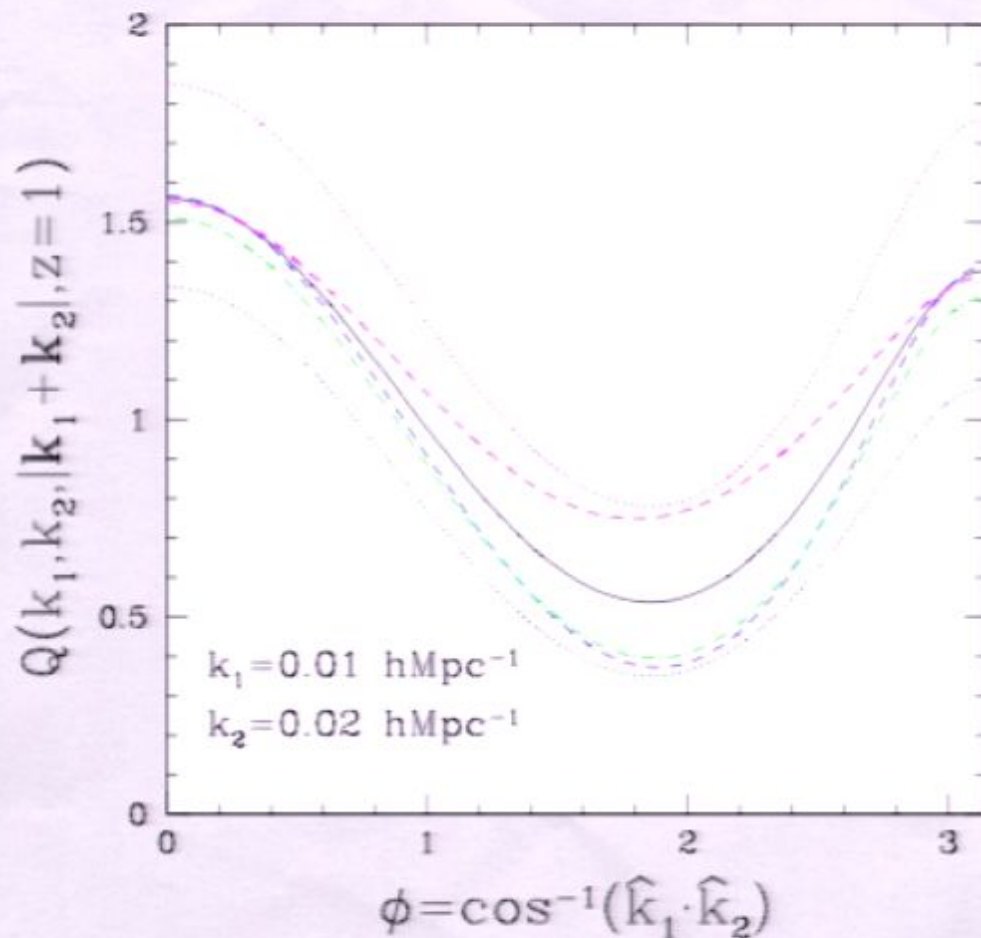
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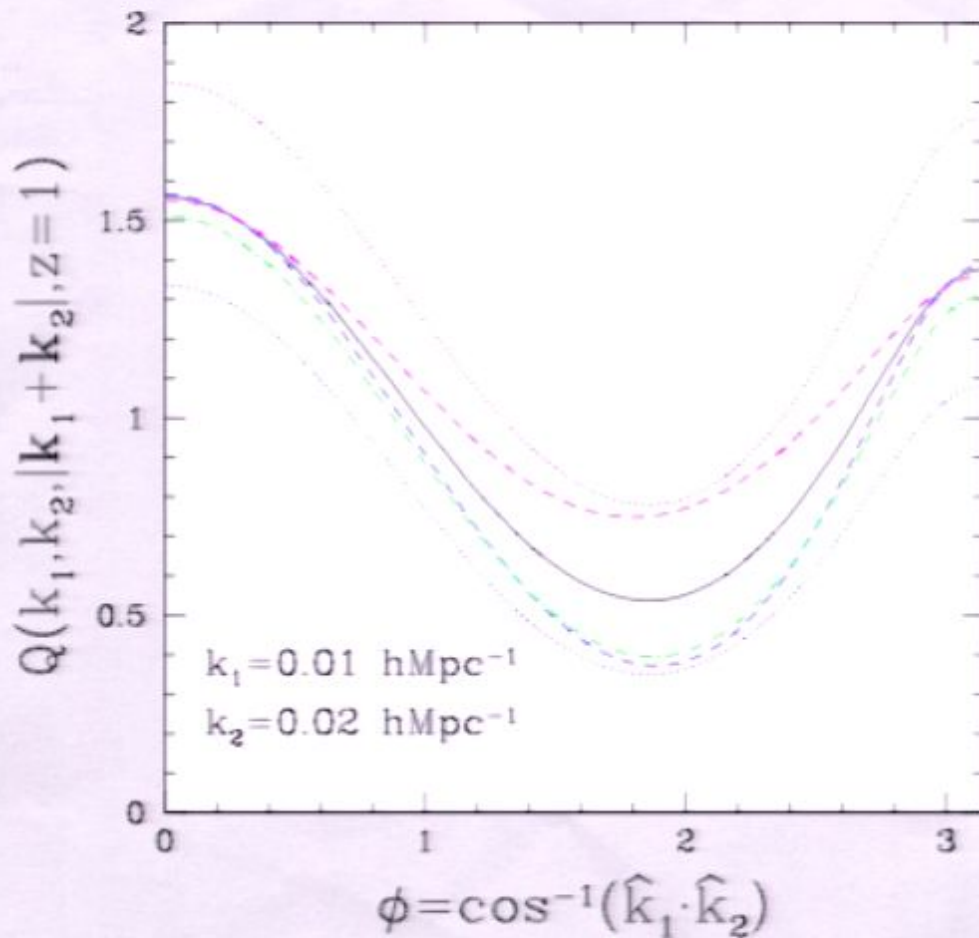
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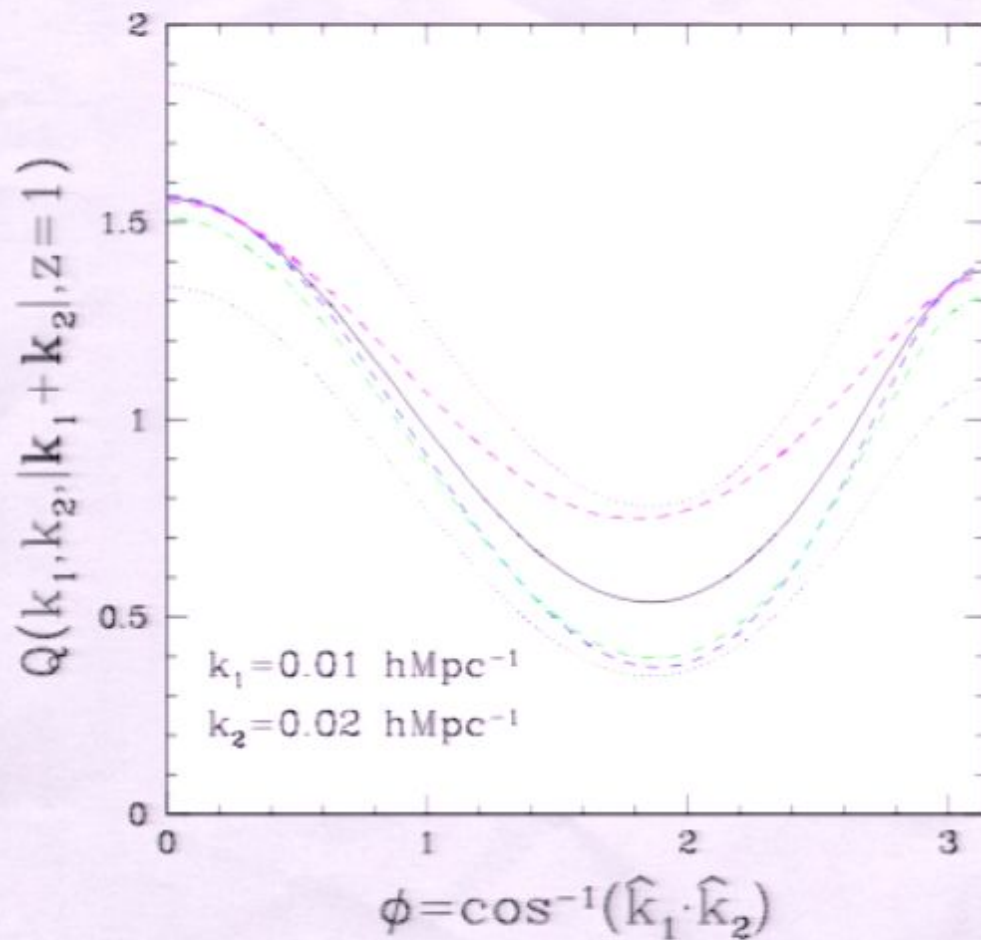
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REDUCED BISPECTRUM



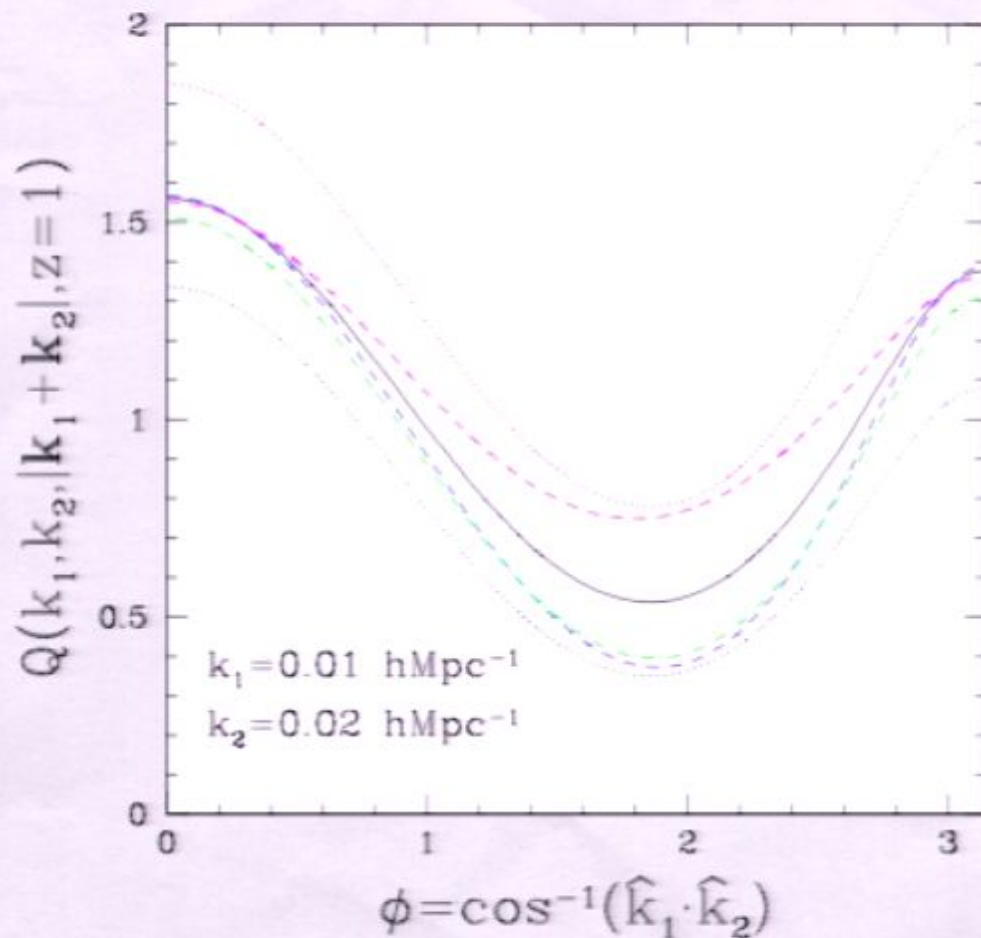
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REDUCED BISPECTRUM



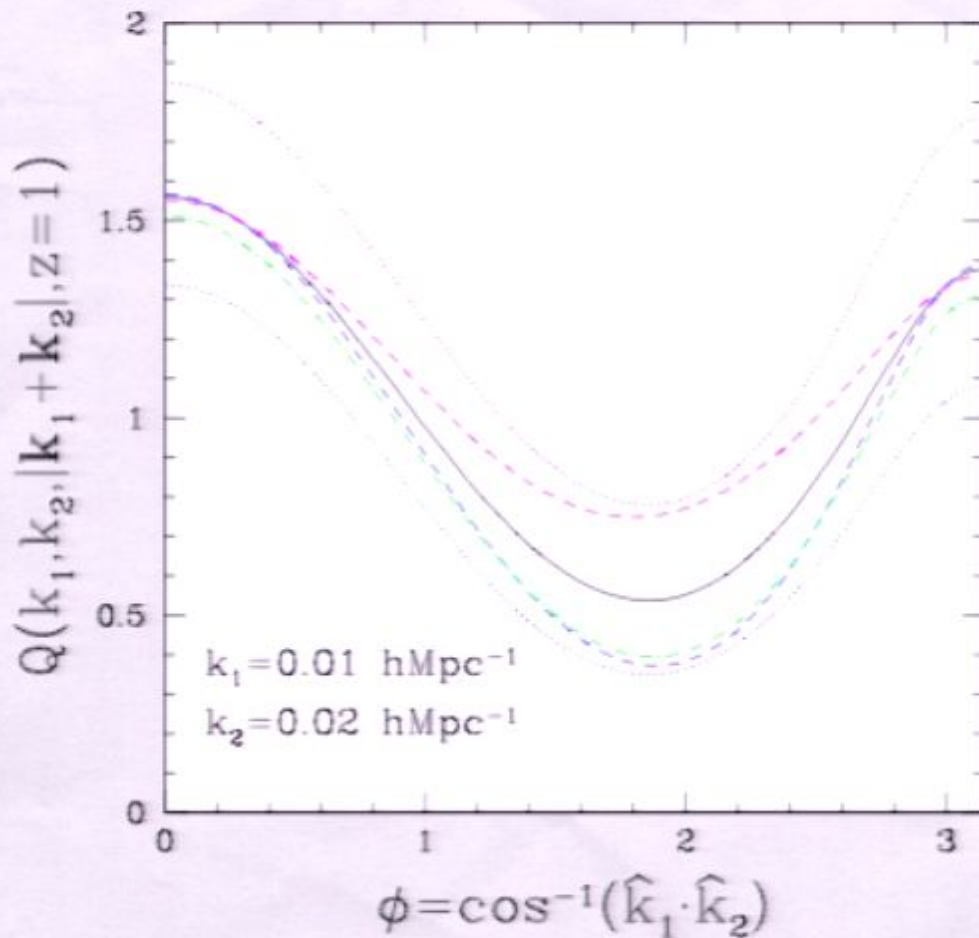
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REDUCED BISPECTRUM



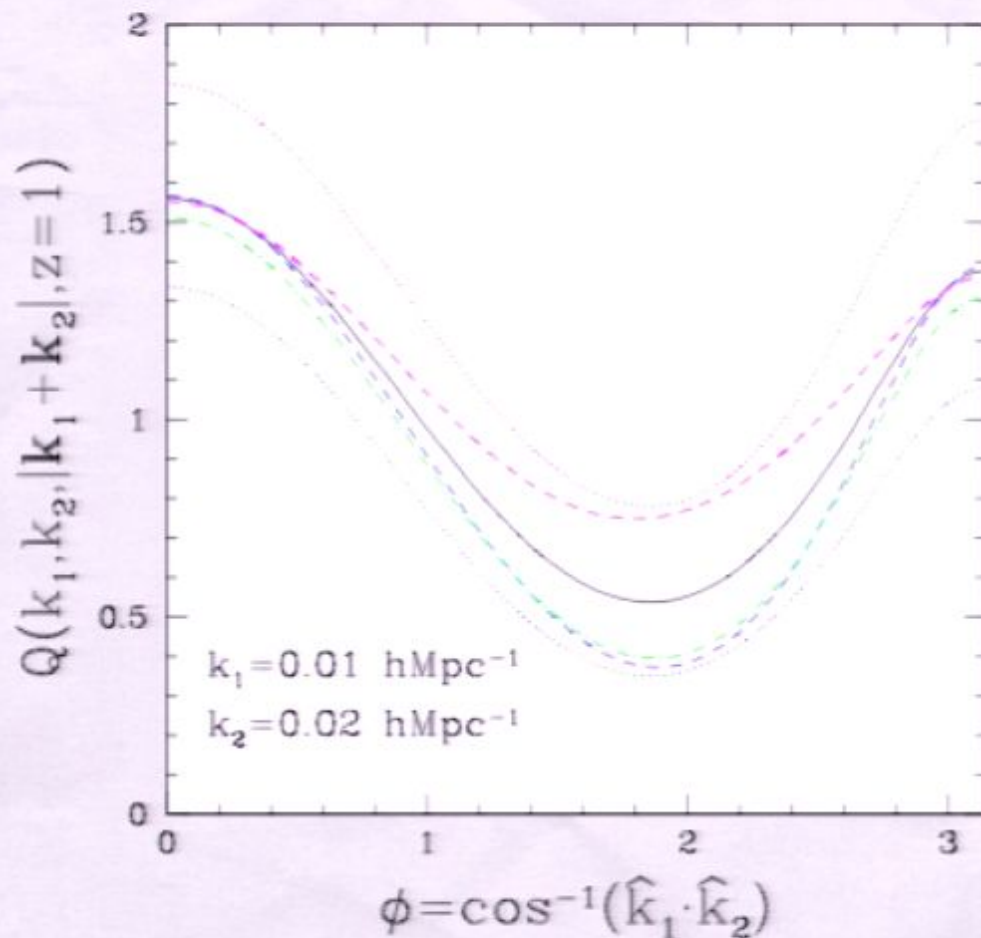
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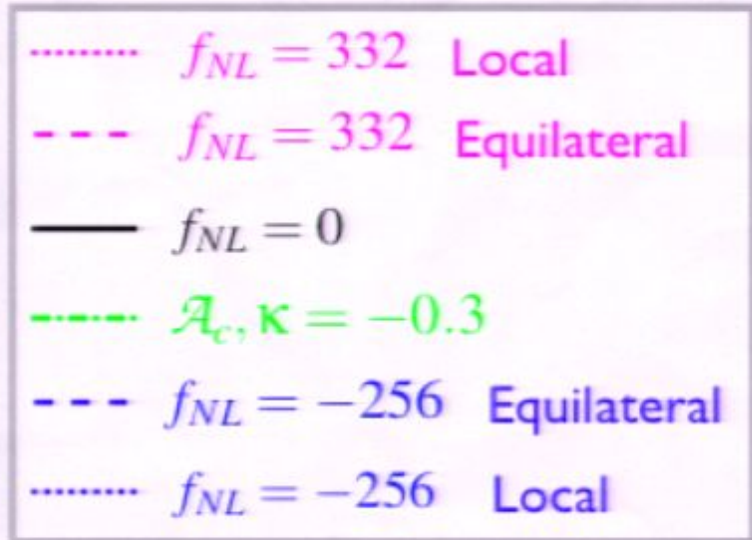
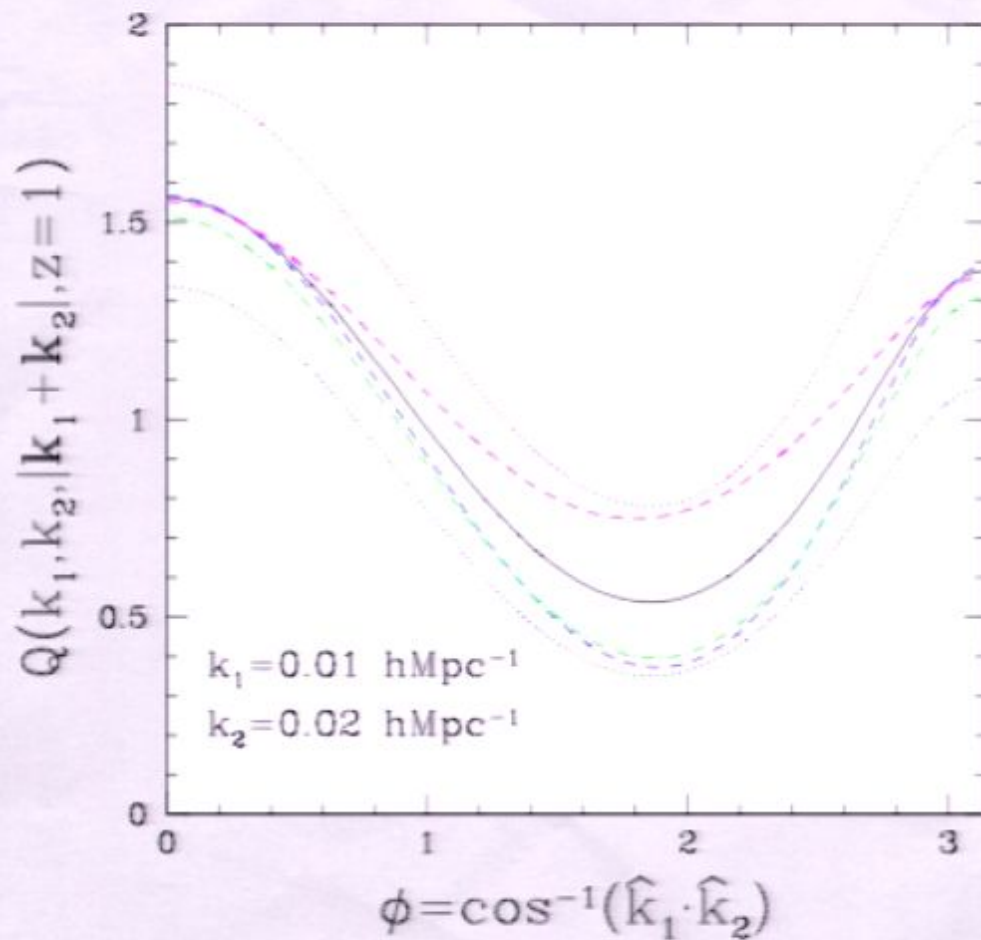
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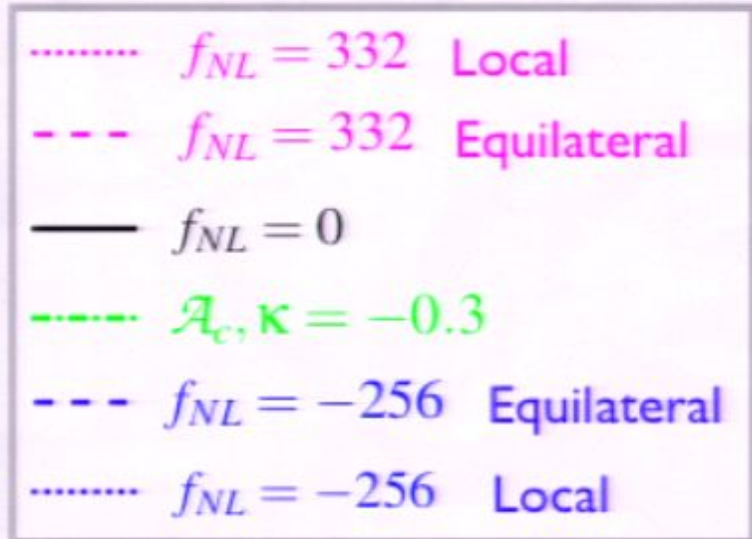
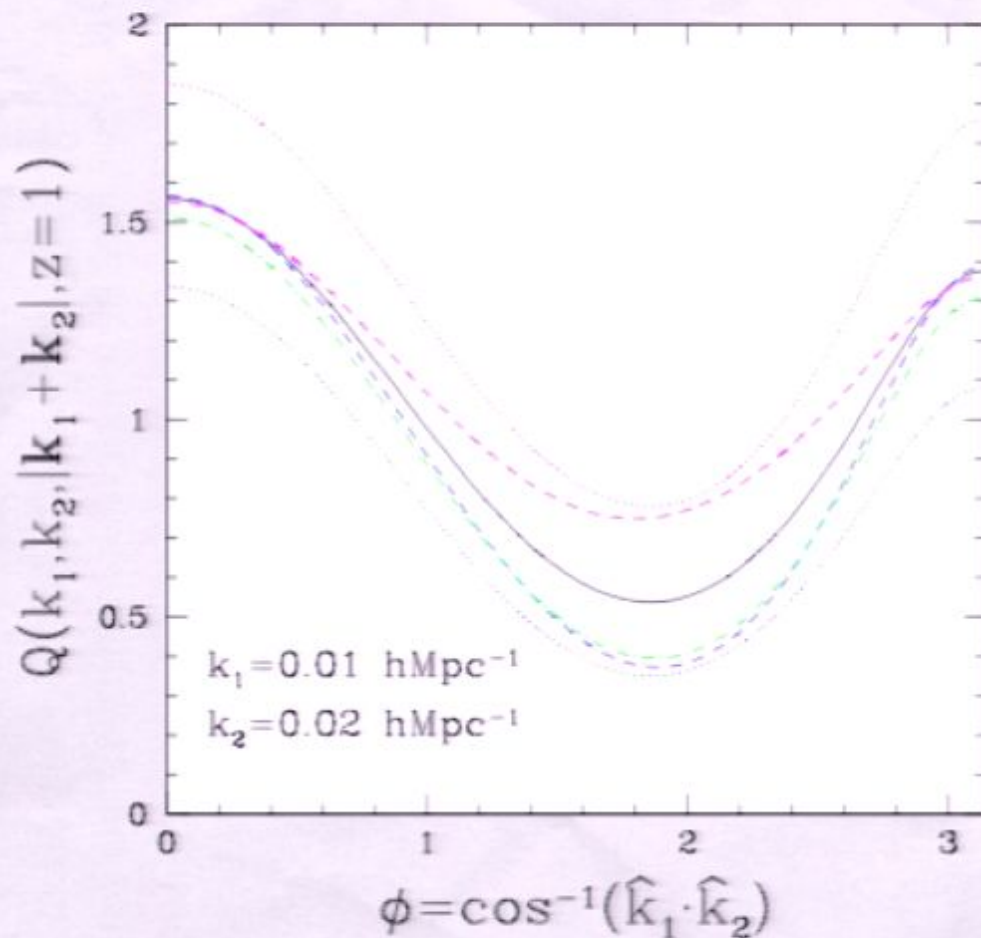


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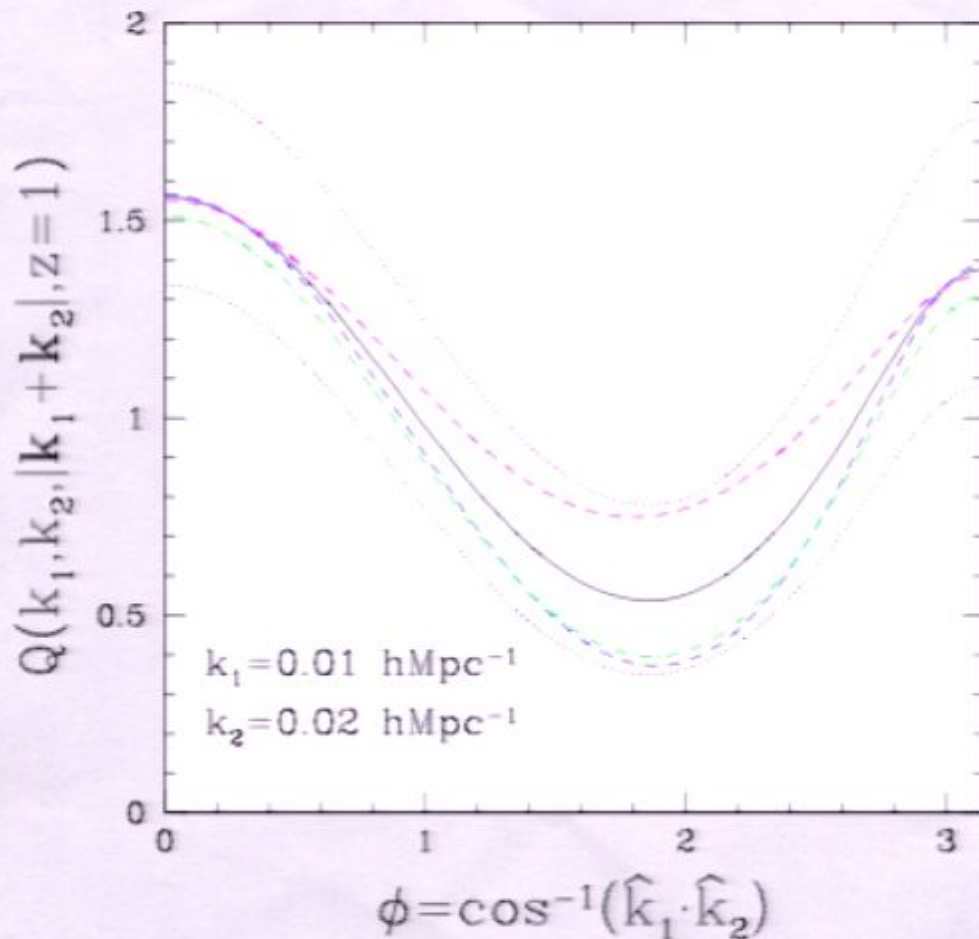
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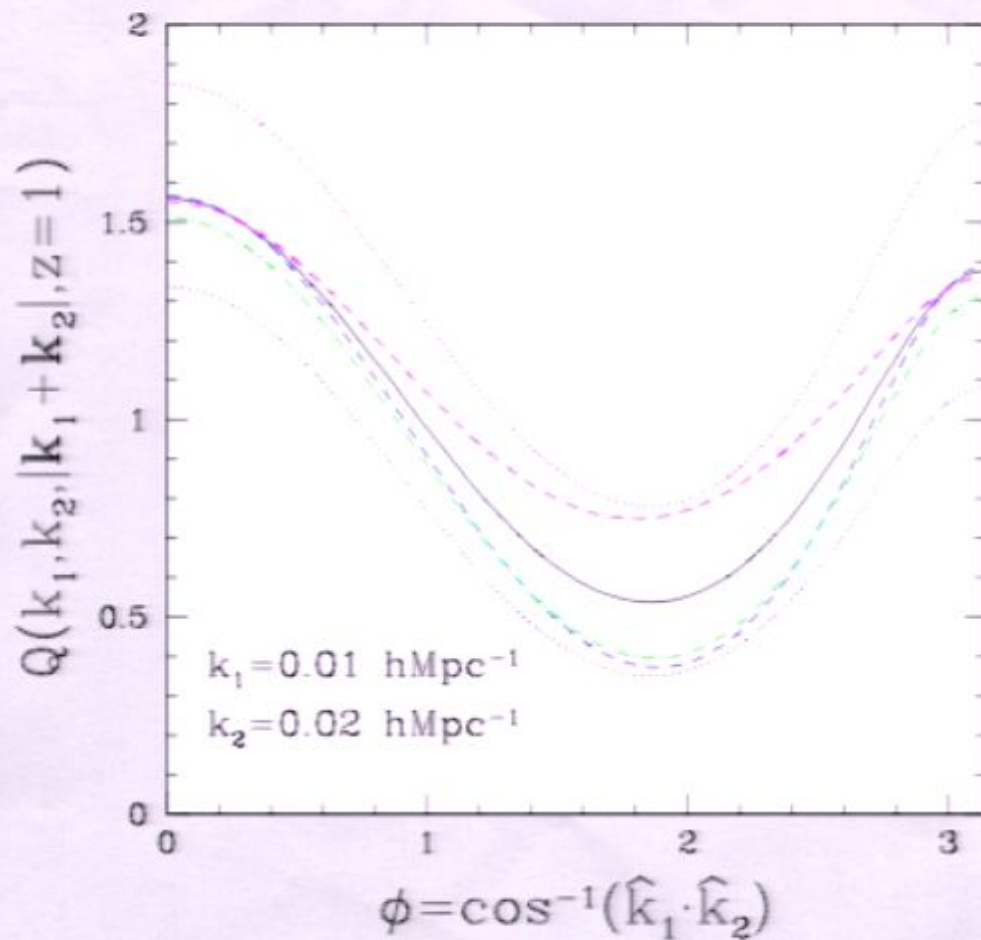


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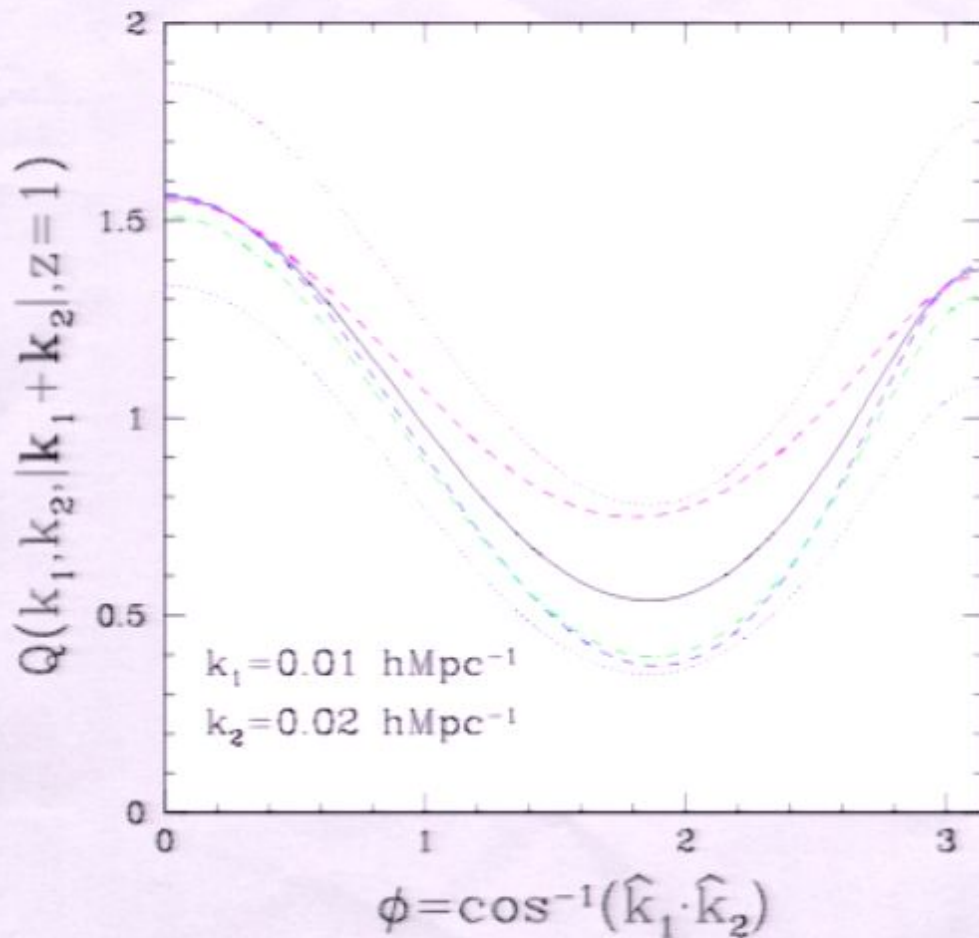
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REDUCED BISPECTRUM



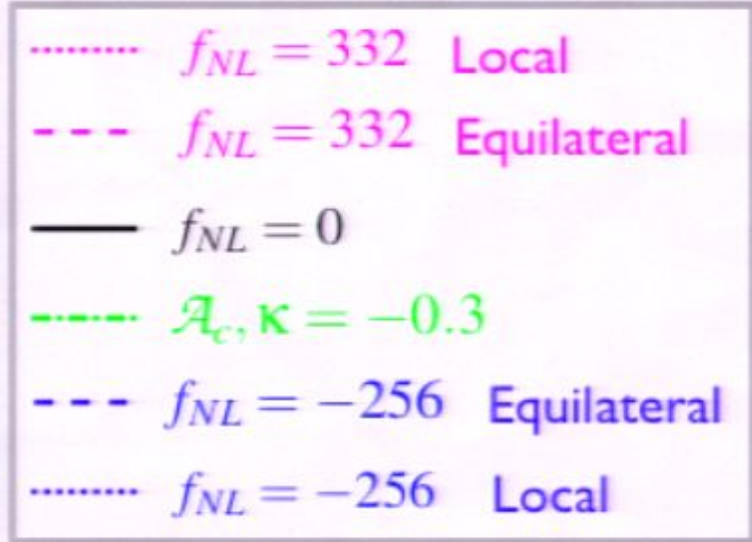
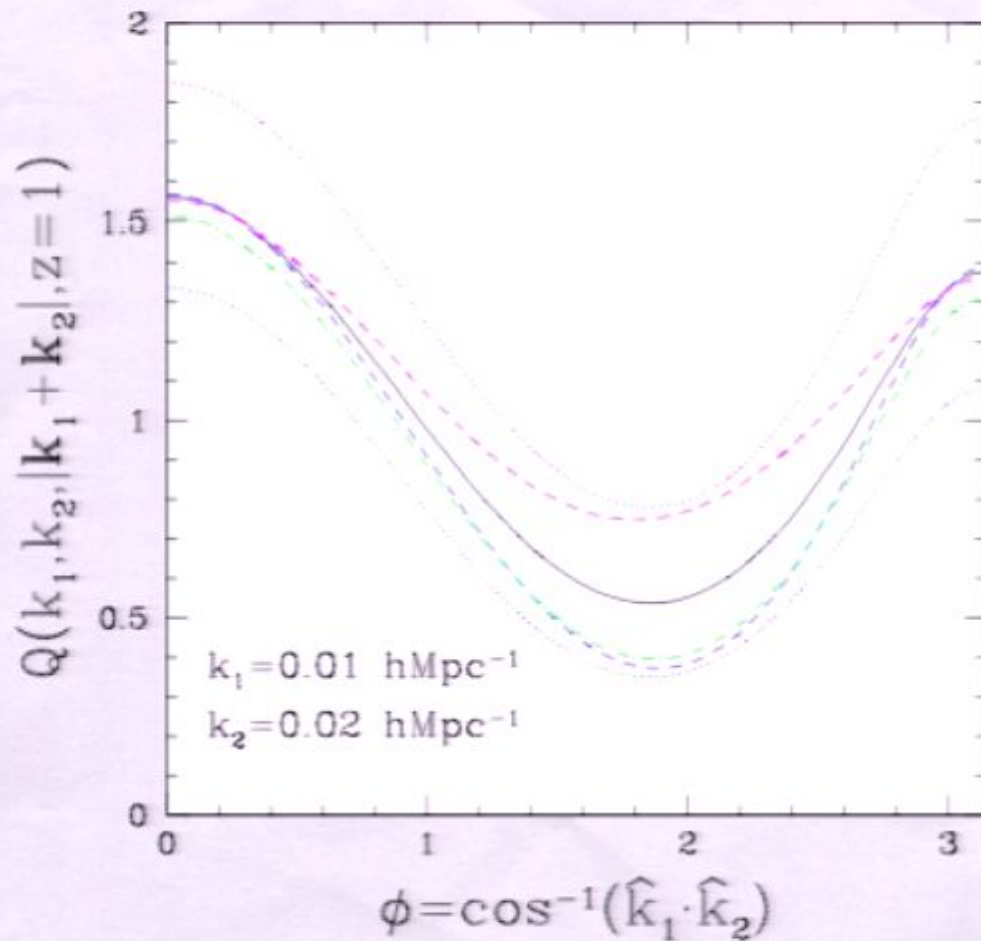
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REDUCED BISPECTRUM

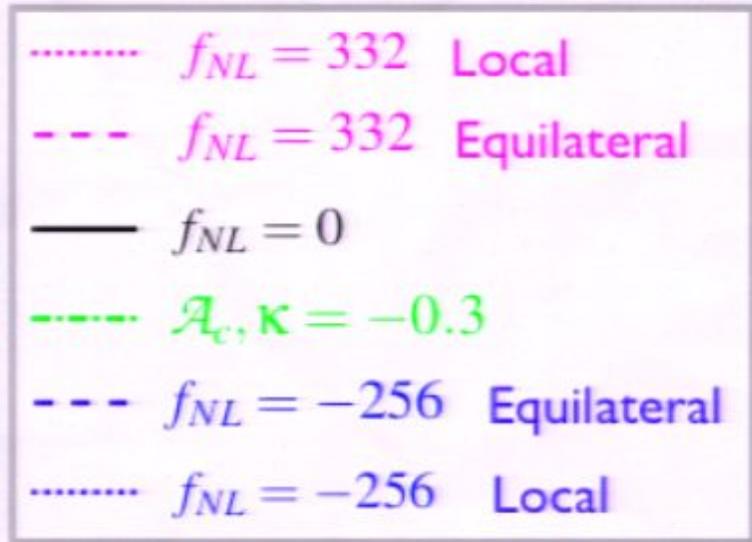
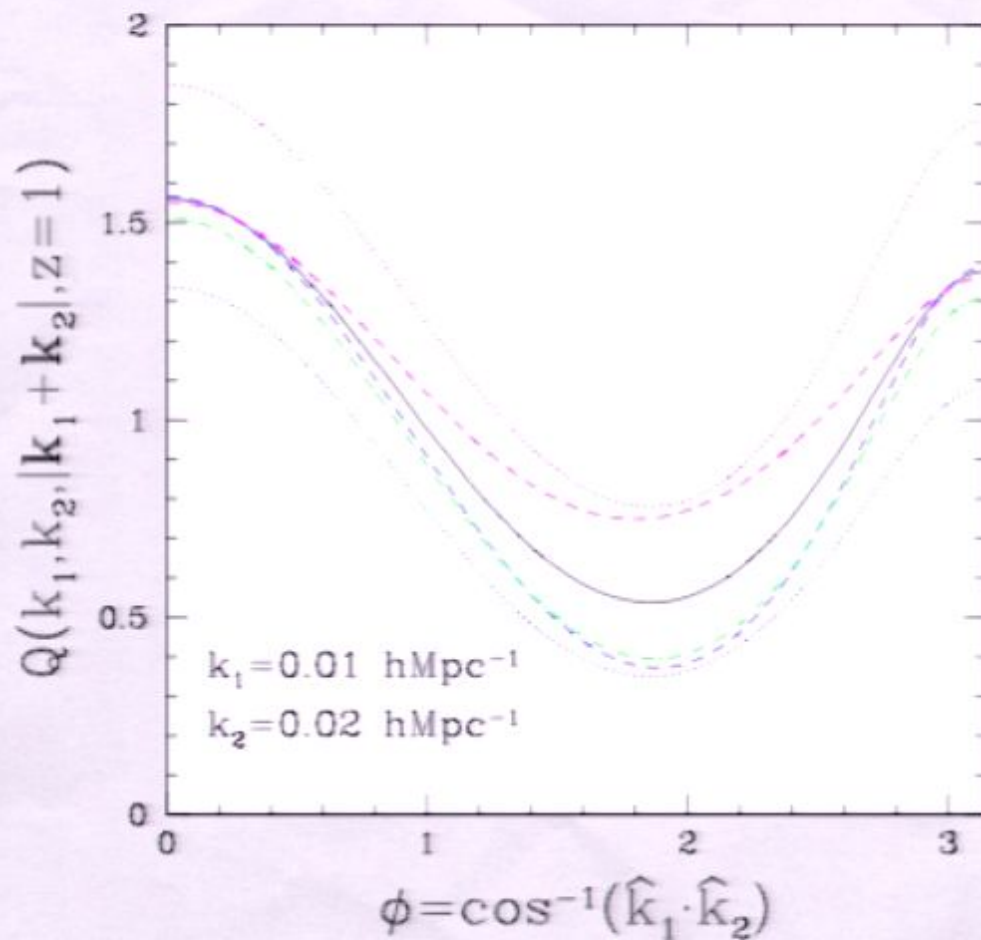


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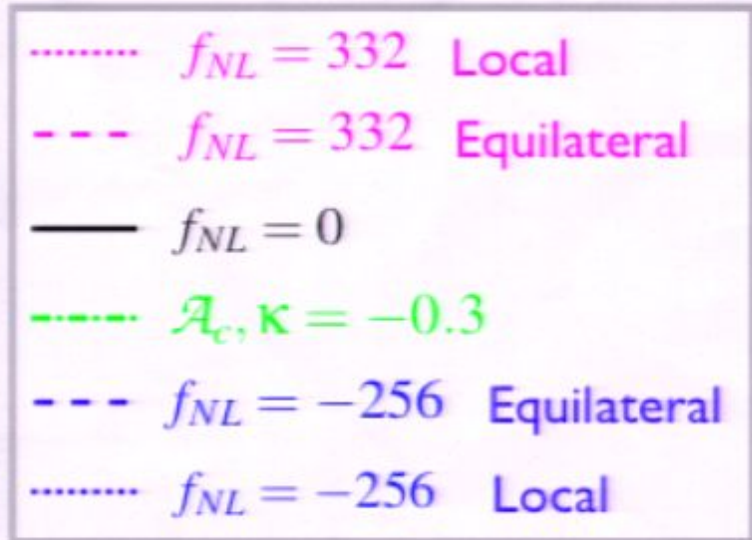
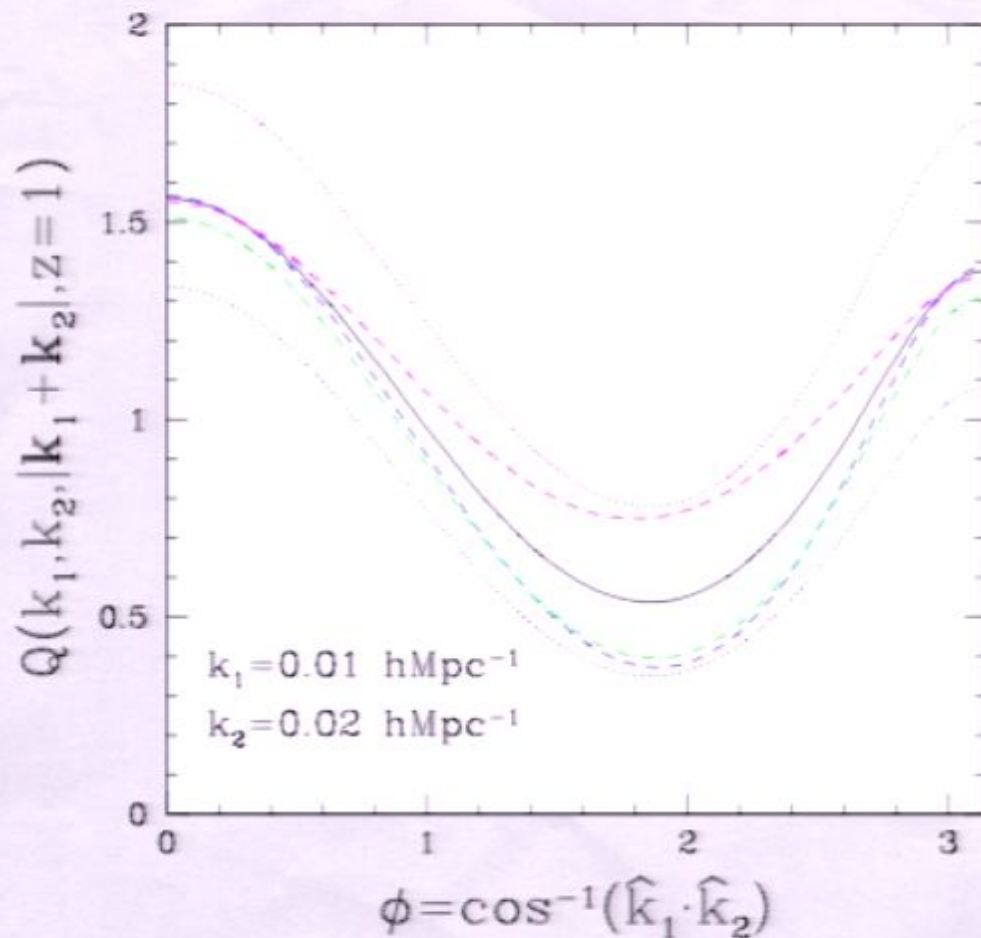
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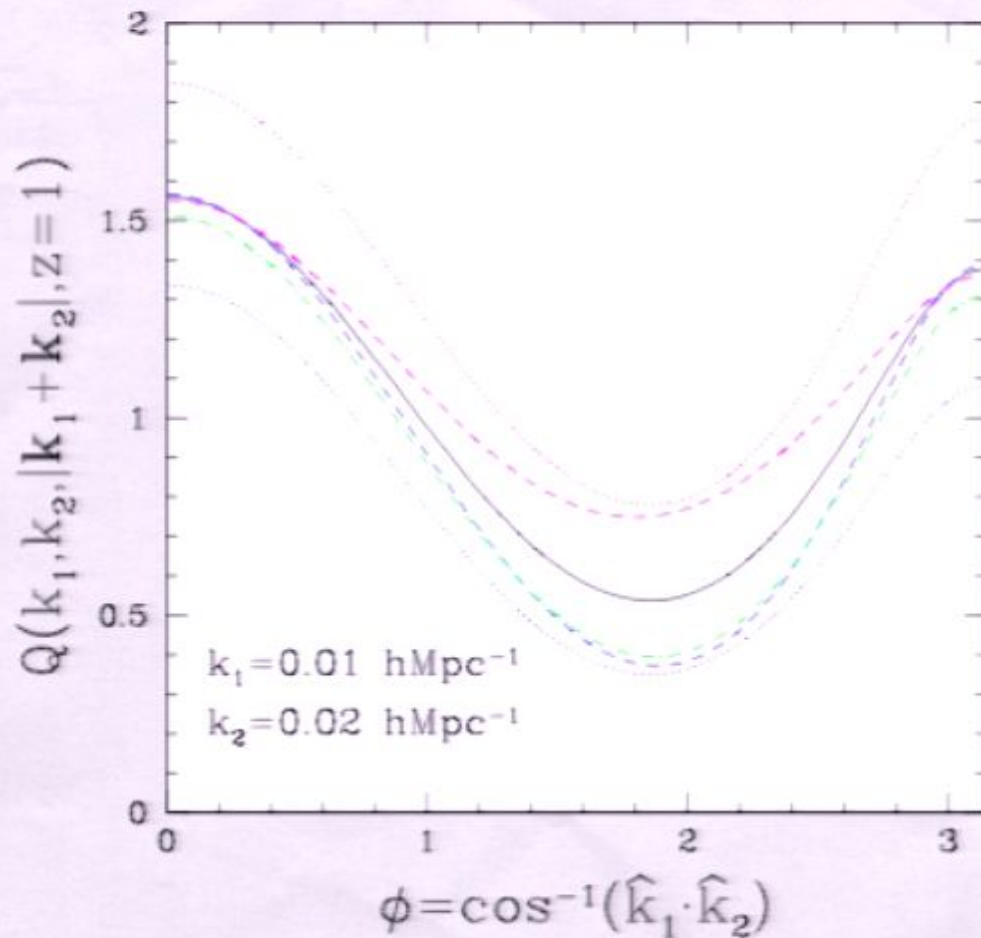
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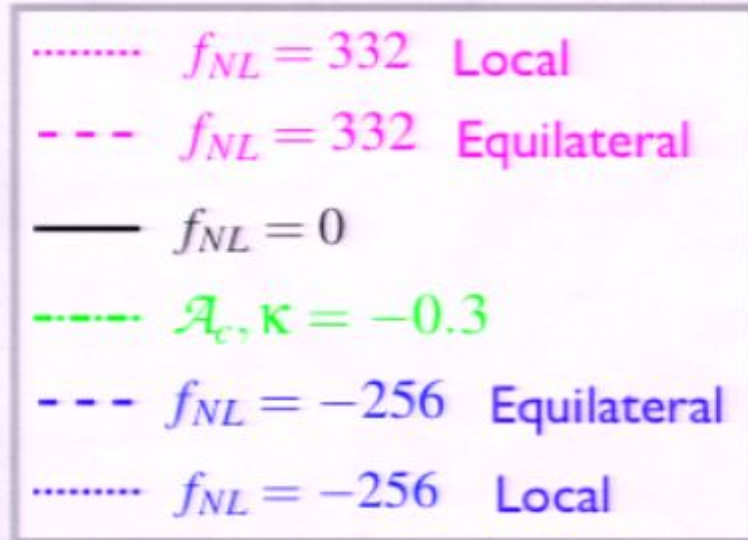
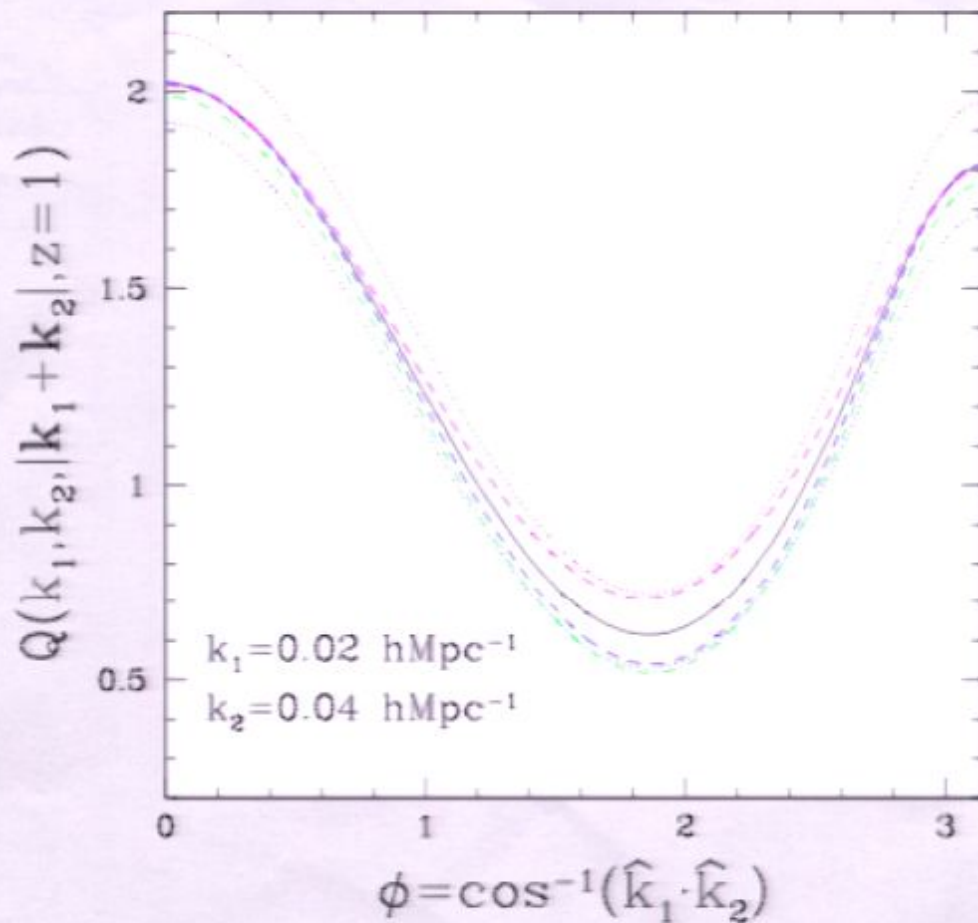


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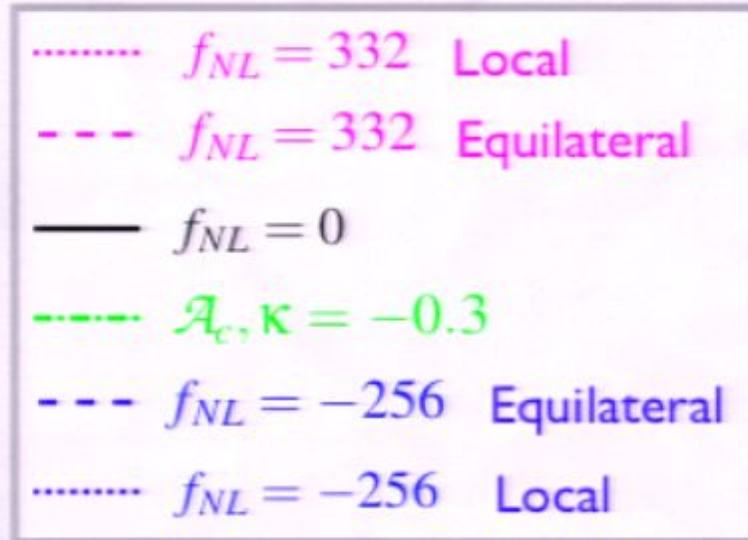
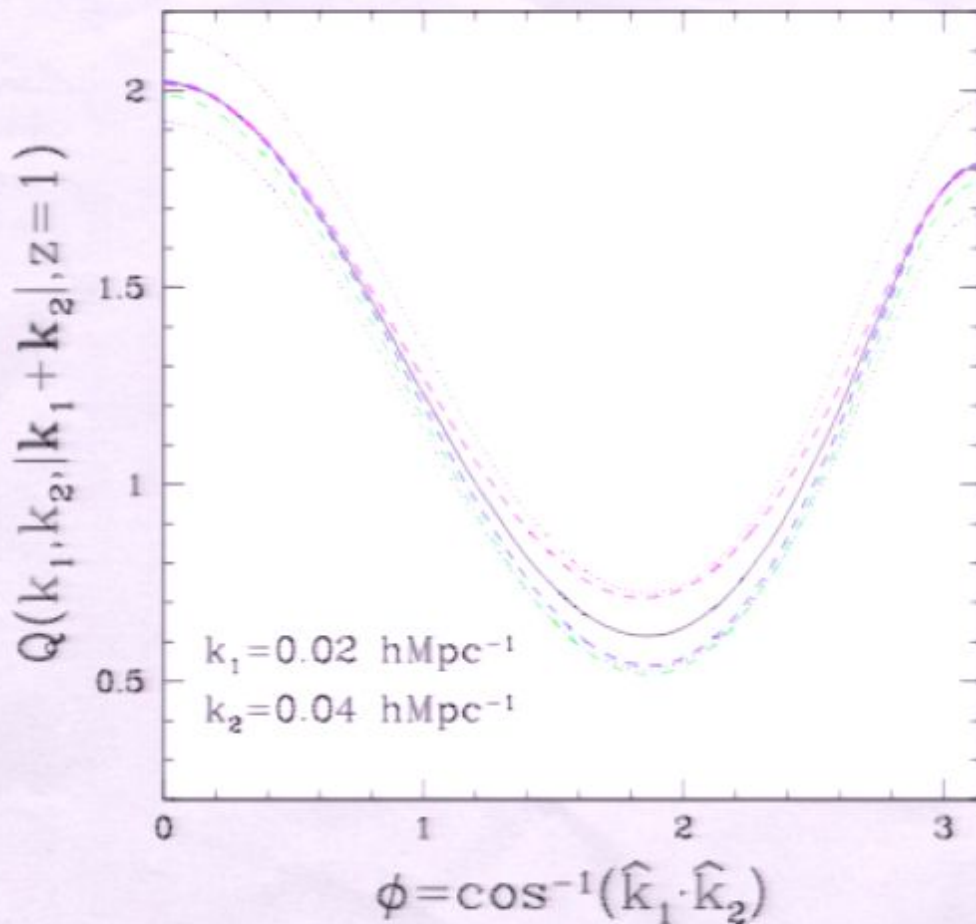


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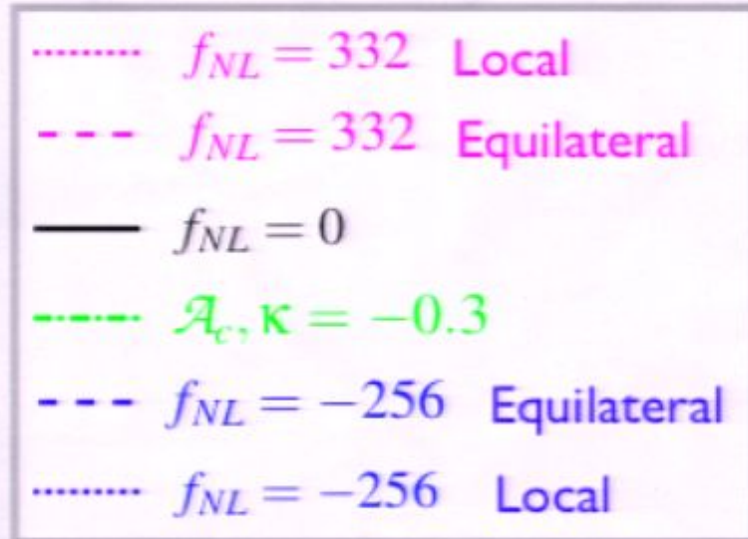
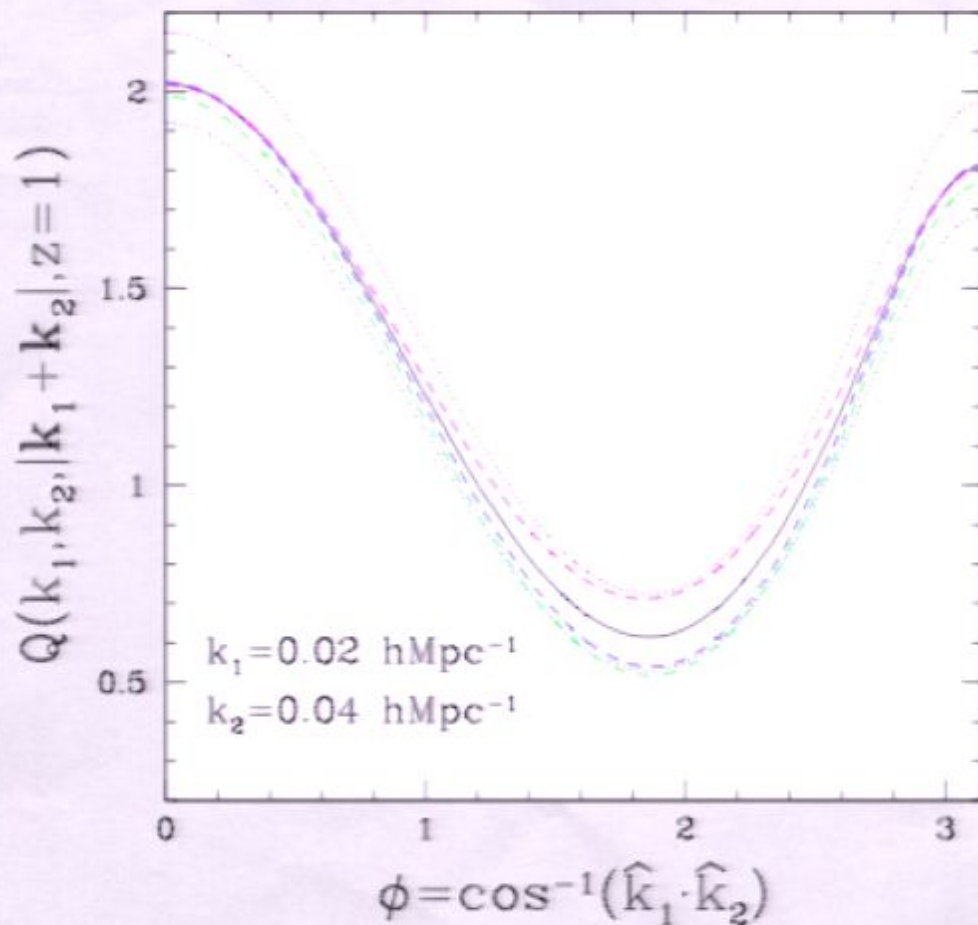
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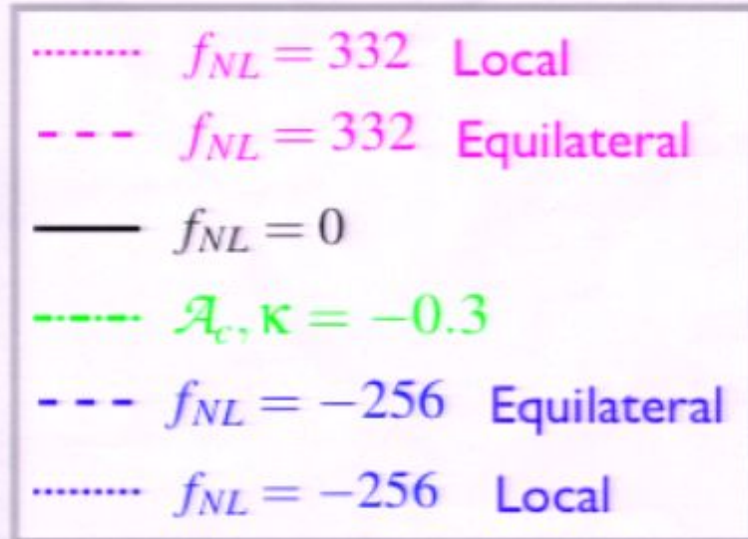
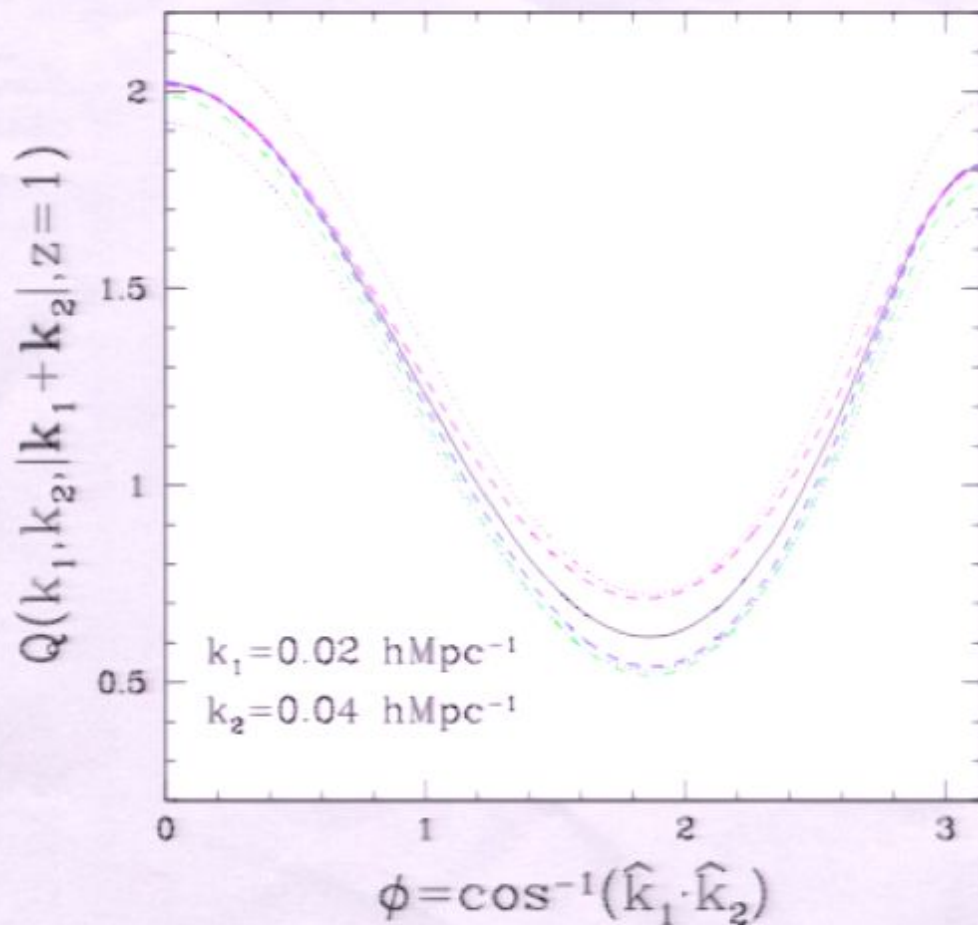
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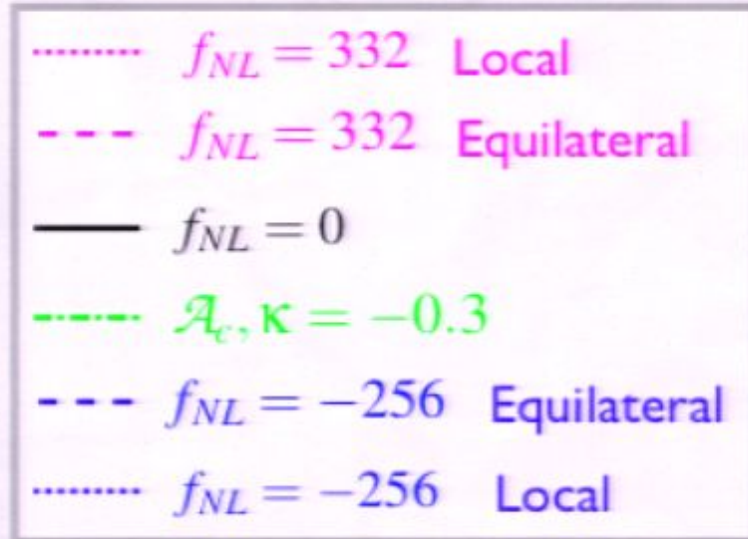
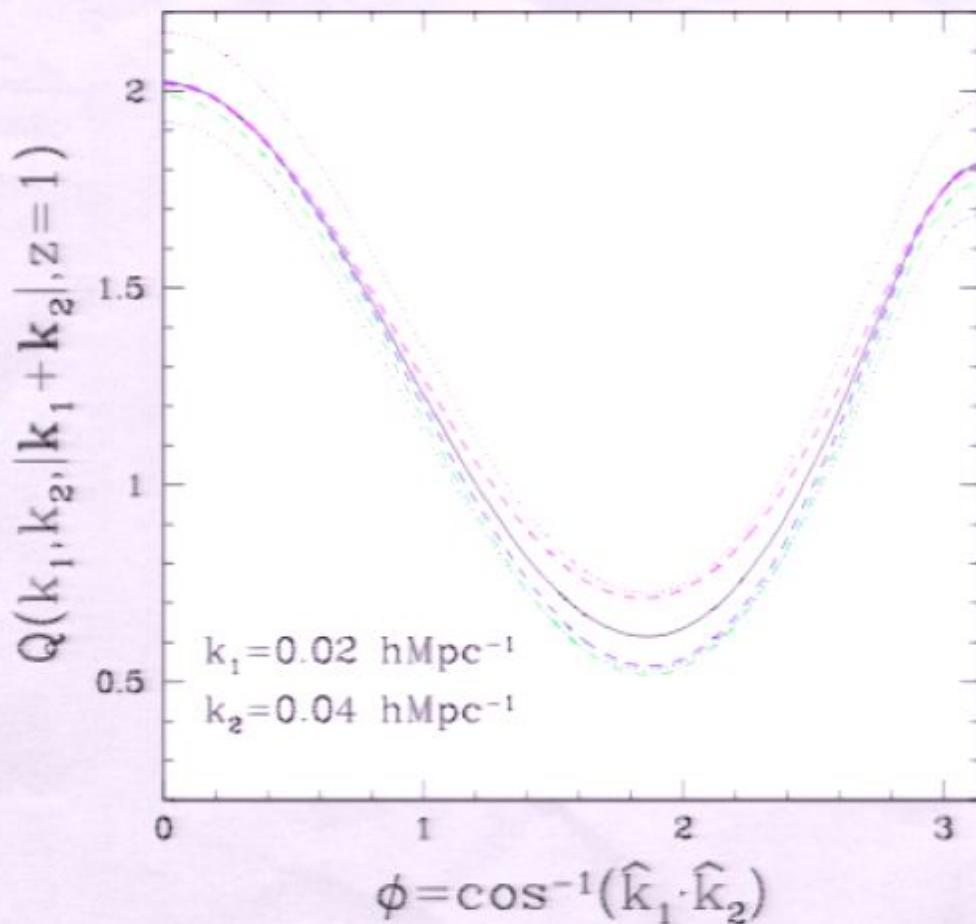
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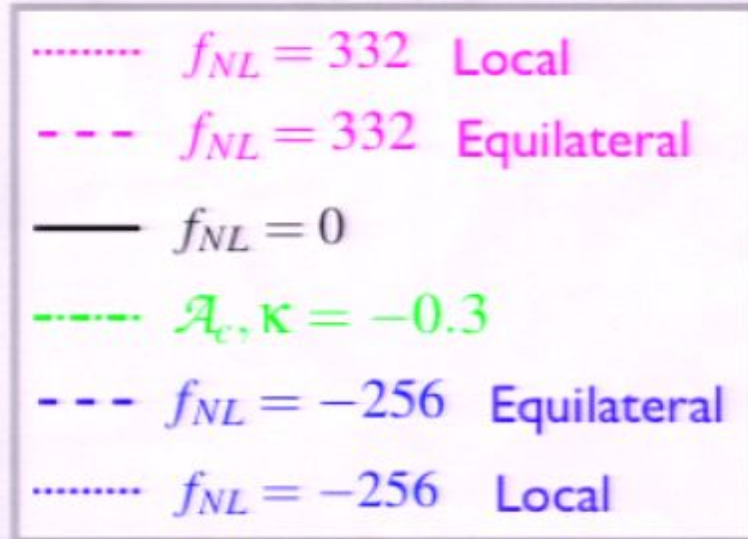
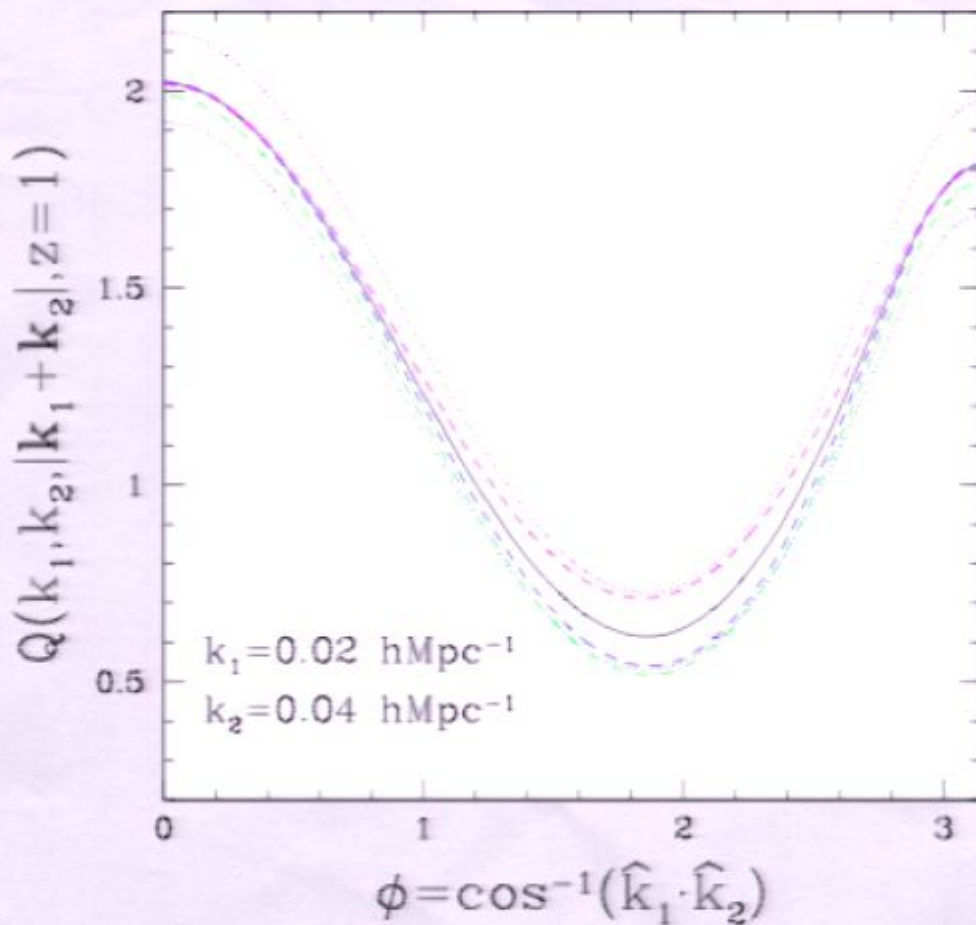
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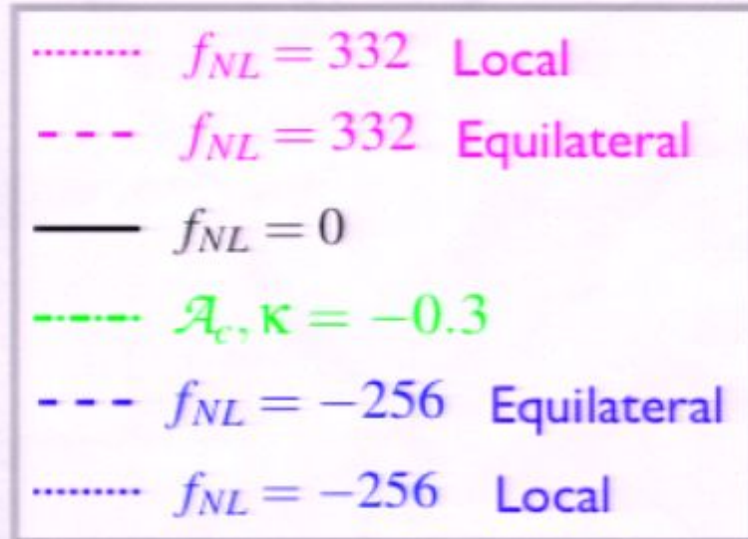
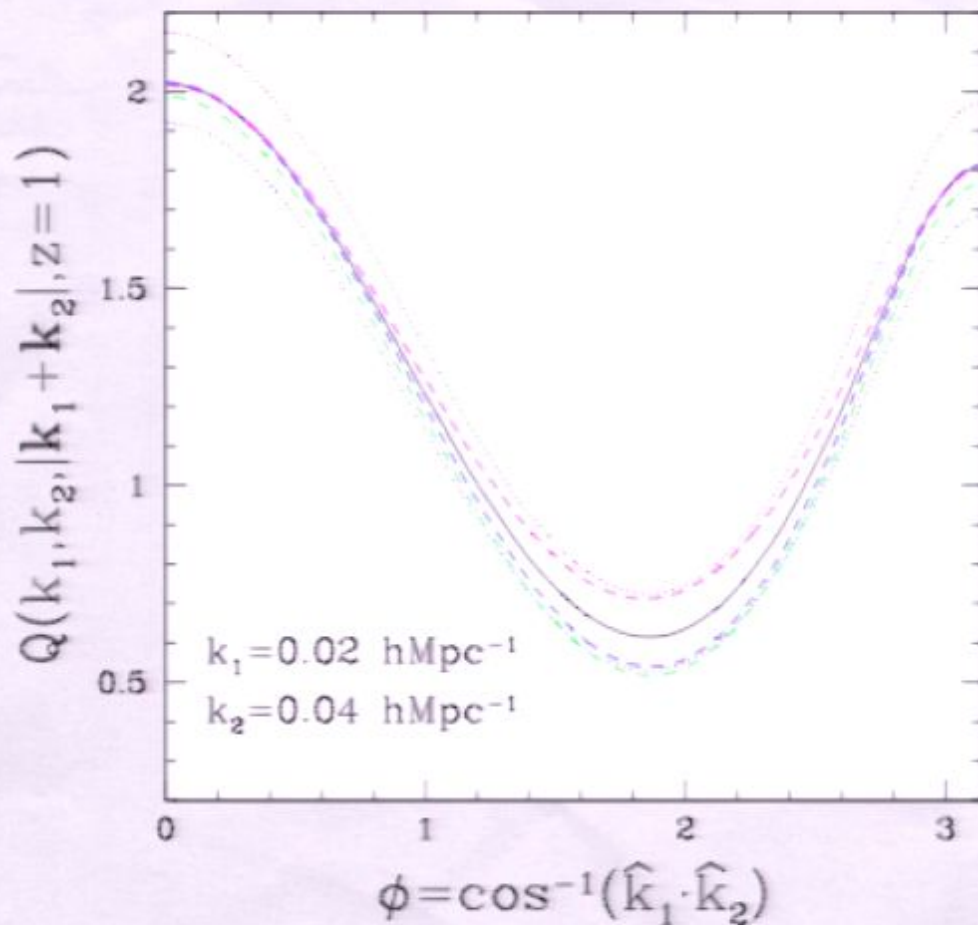
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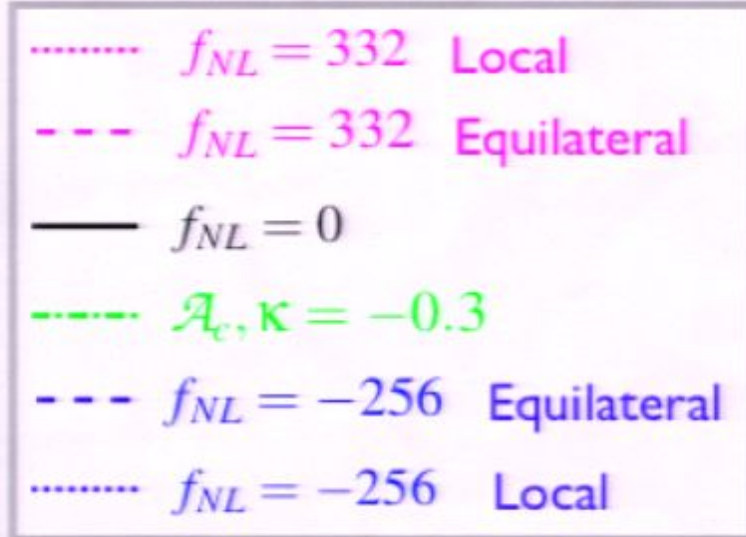
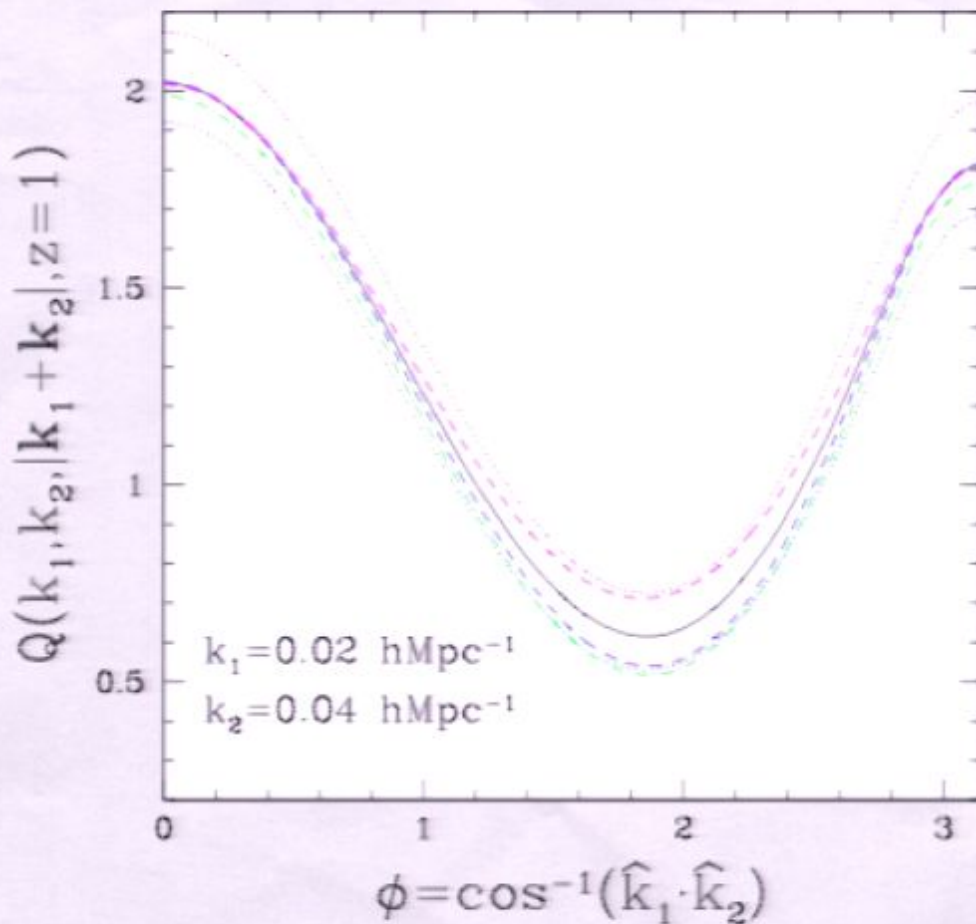
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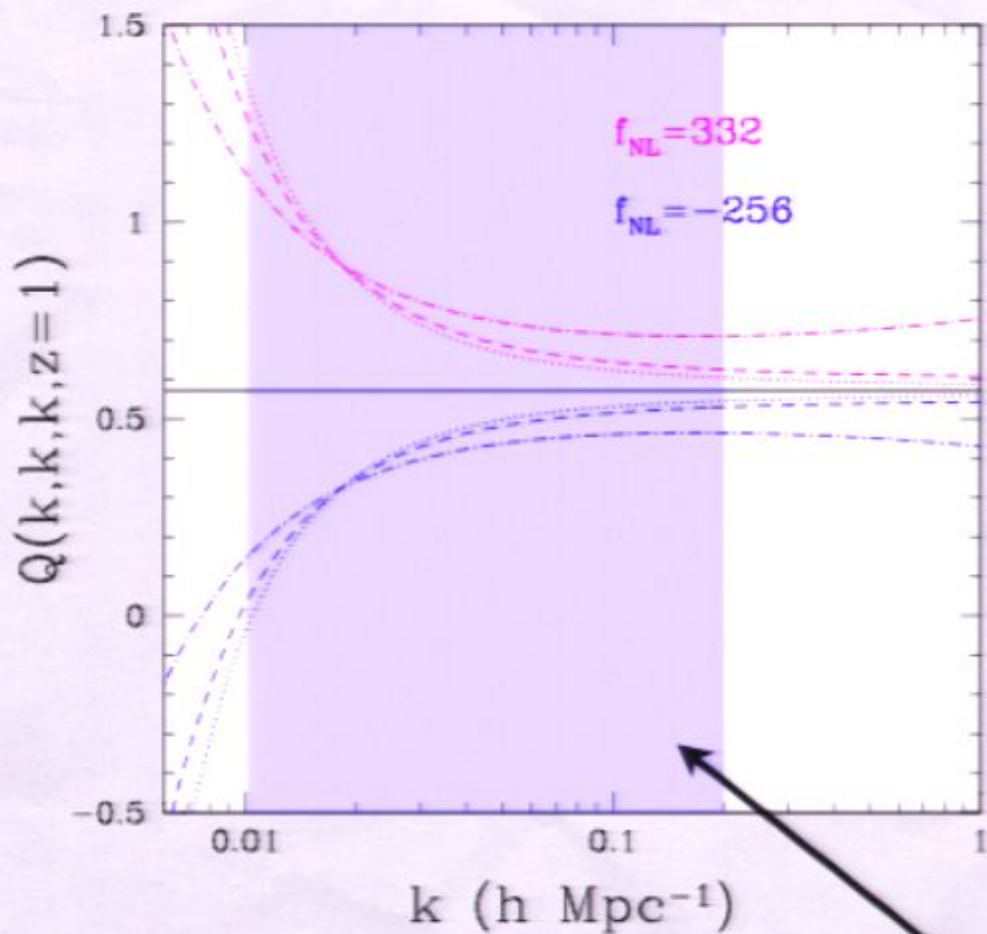
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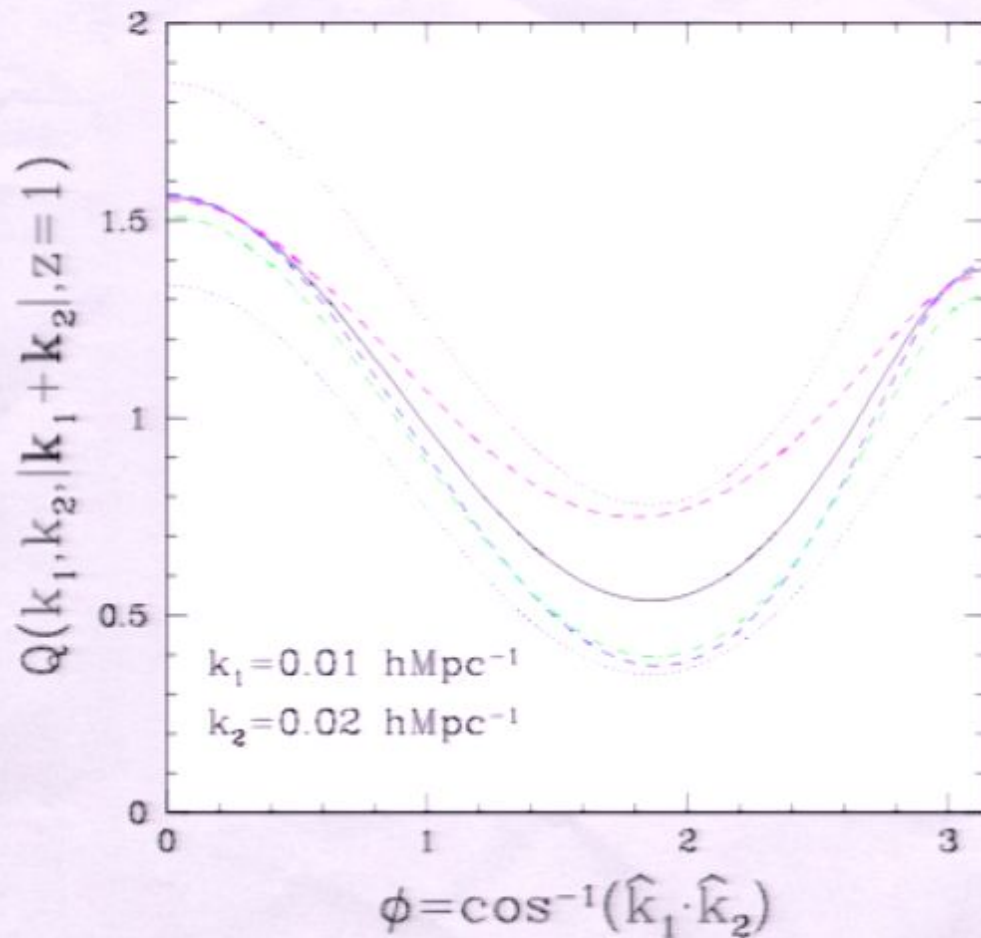
REDUCED BISPECTRUM



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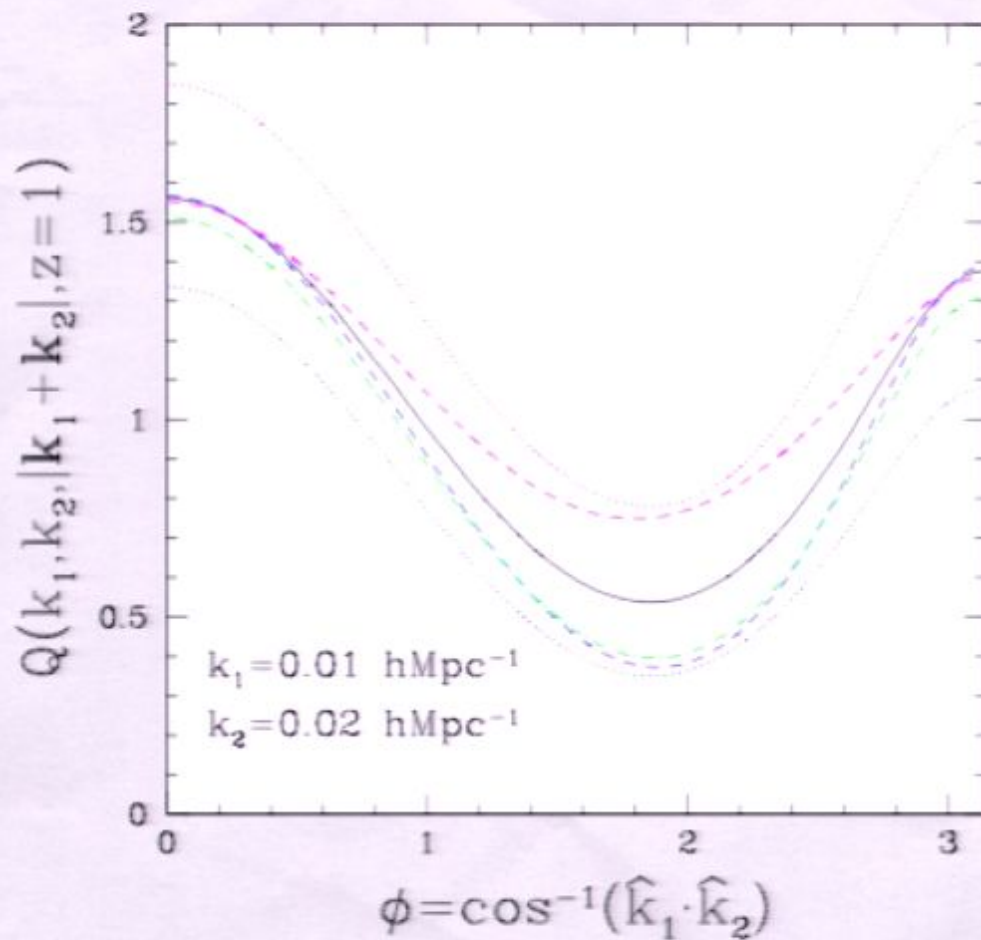
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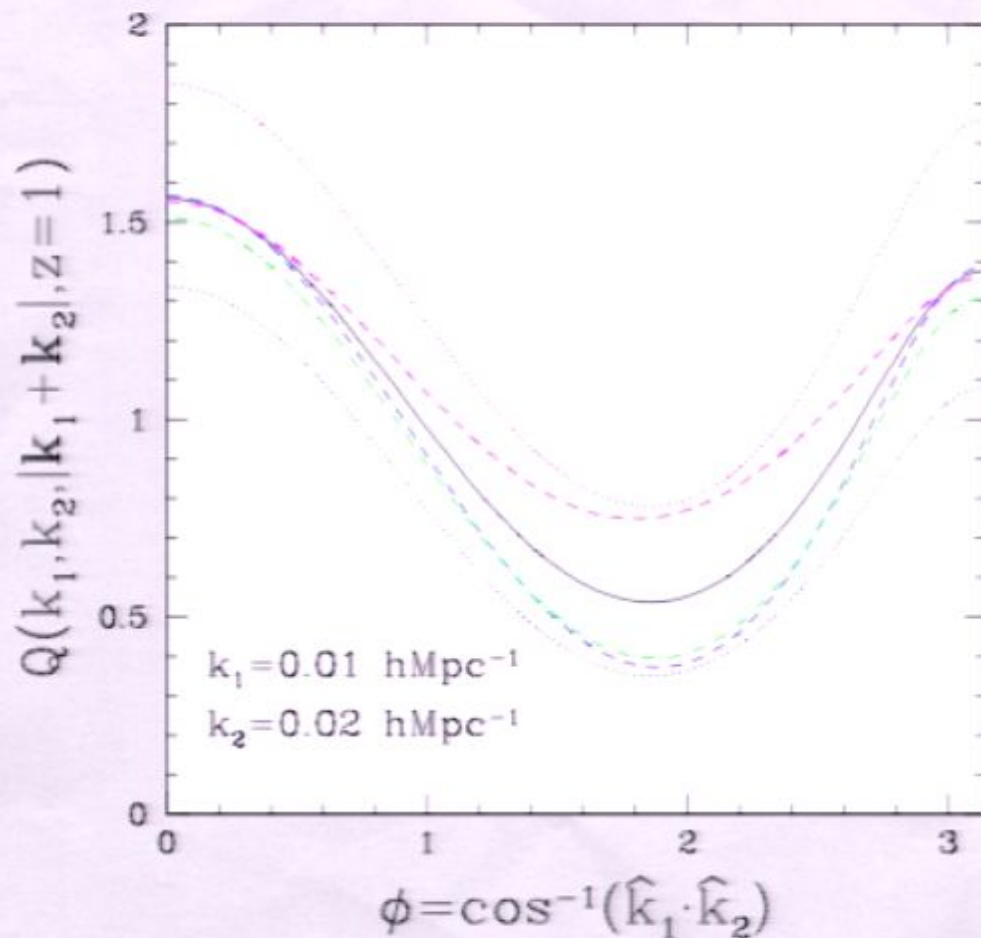
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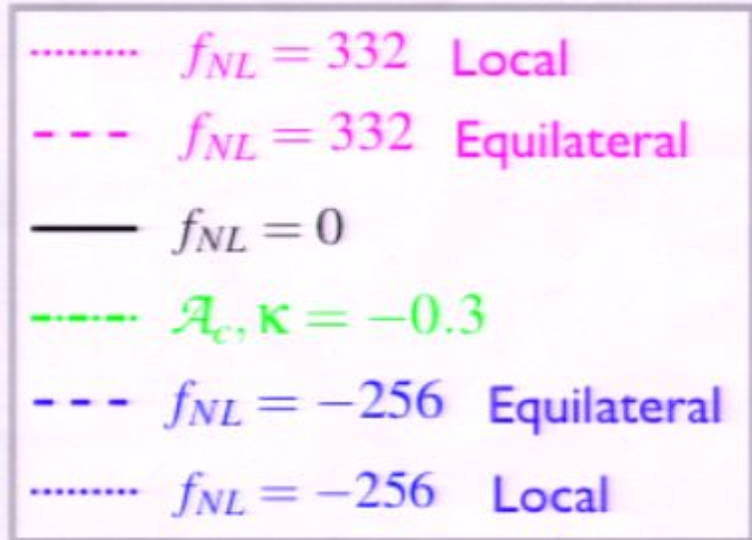
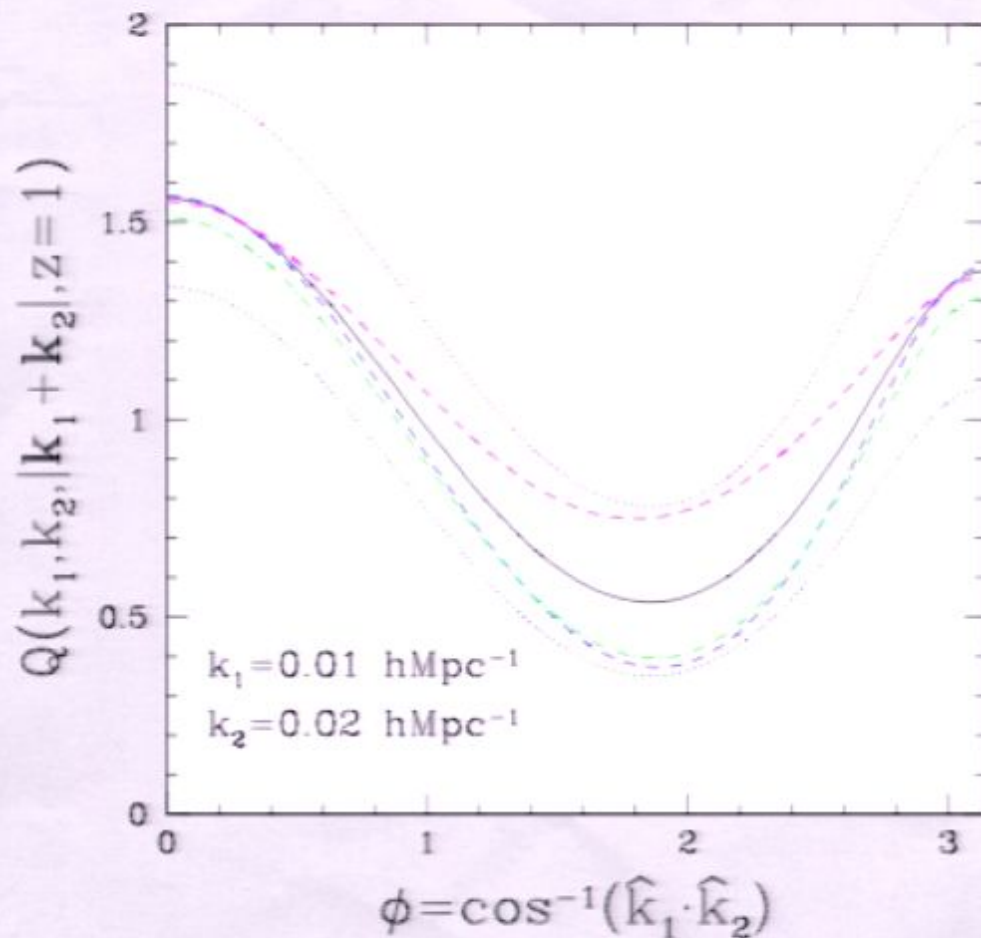
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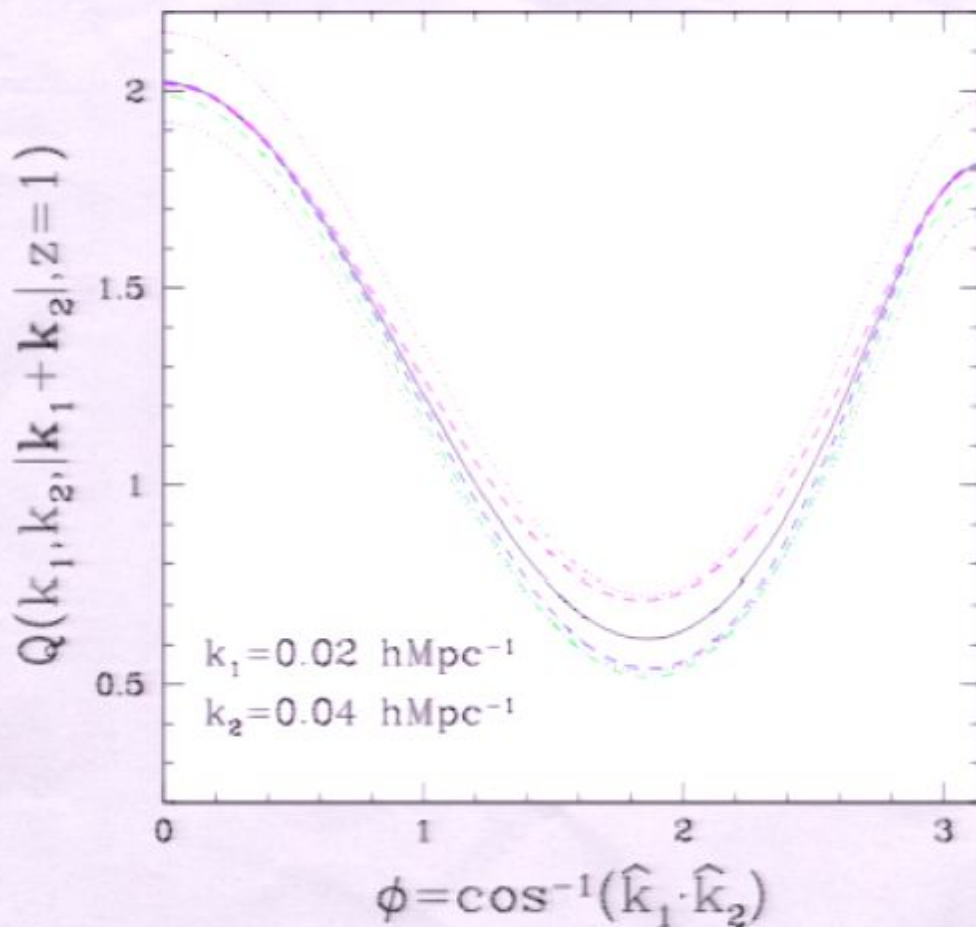


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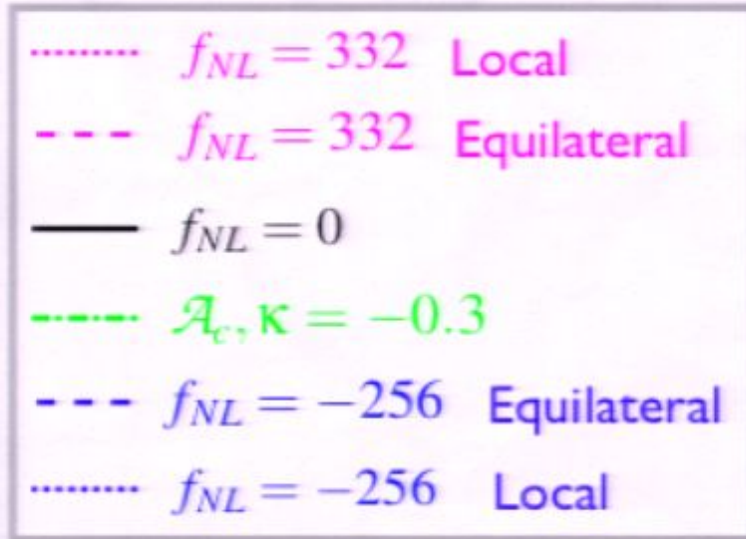
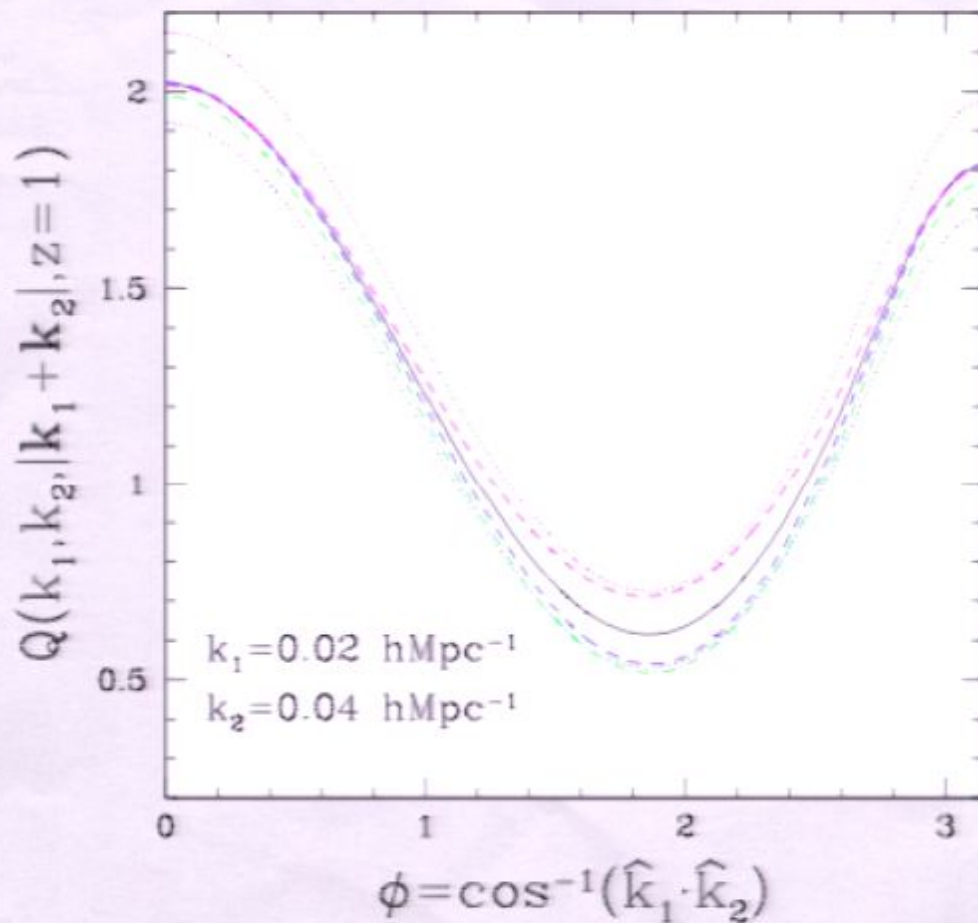


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REDUCED BISPECTRUM



CONCLUSIONS

- Large and/or running non-Gaussianity is a signature of interesting inflationary physics (in DBI, NG tracks geometry)
- Scale-dependent non-Gaussianity is well-motivated and may be best tested on scales smaller than CMB
- Cluster measurements may soon be sensitive to NG with a large but reasonable running (30% change in mass function)
- Combining measurements from the CMB, galaxy bispectrum and cluster number counts can likely constrain a smaller running.

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FUTURE DIRECTIONS

- Data available soon: SPT, SZA, MACS...
- Re-consider CMB constraints including scale-dependence
- N-body simulations with more general non-gaussian initial conditions?
- Other constraints?
 - Primordial Black Holes? (in progress)
 - Scale-Dependent Bias? (Dalal, Dore, Huterer, Shirokov)