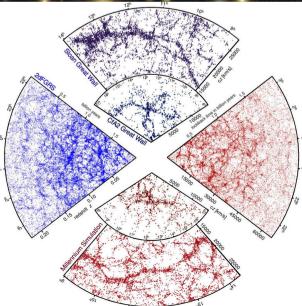
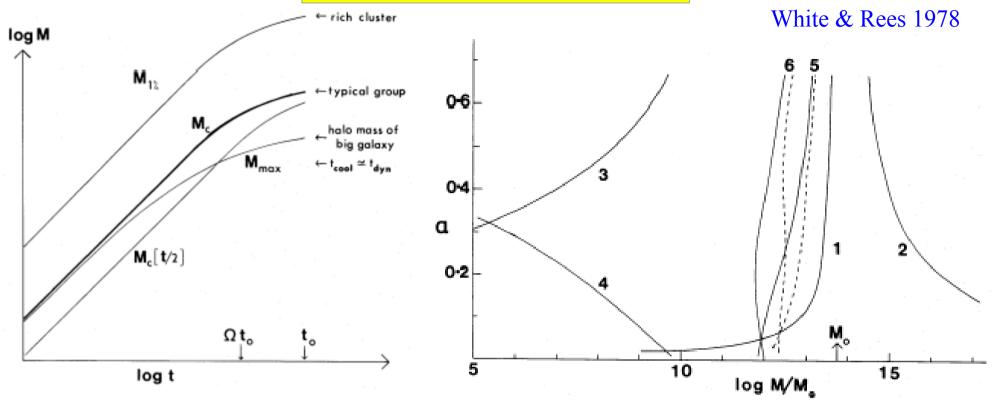
L-Galaxies workshop MPA, February 2016

The Munich galaxy formation model: origins and raison d'être

Simon White Max Planck Institute for Astrophysics



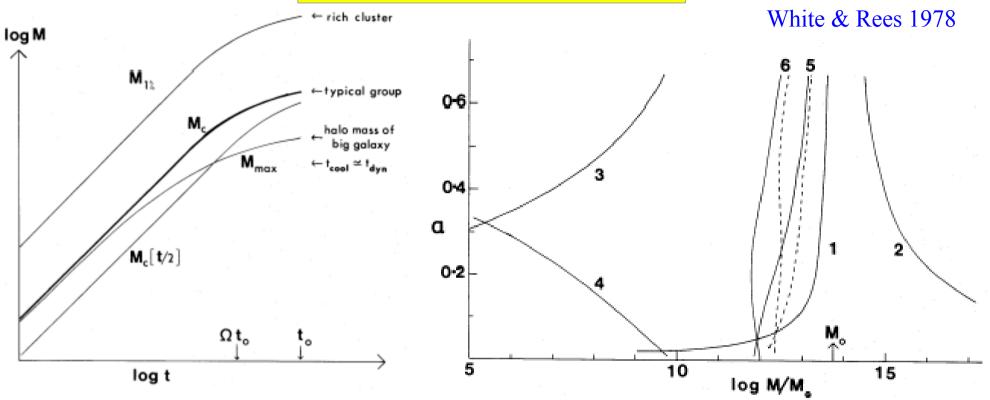
The beginnings



"Galaxies form by the cooling and condensation of gas at the centres of a hierarchically aggregating population of dark matter halos"

This extended previous work by form by adding :
(i) dynamically dominant pre-existing dark matter,
(ii) hierarchical growth of clustering,
(iii) supernova feedback to limit star formation

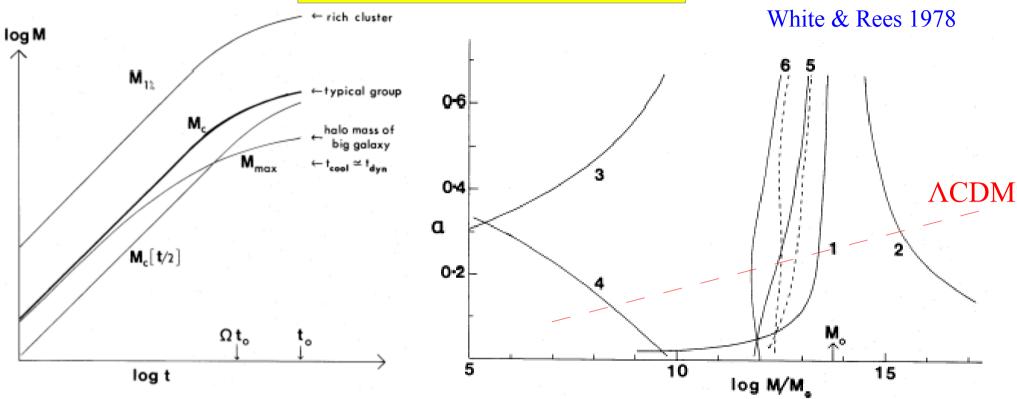
The beginnings



- random-phase initial conditions with a scale-free power spectrum
- evolving halo mass distribution: $\Phi_{PS}(M_h, z)$
- uniform spherical halos at the virial density ~200 $\rho_{crit}(z)$
- star formation occurs when the baryonic cooling time $t_{cool} < t_{hubble}(z)$
- star formation efficiency regulated by SN feedback
- no galaxy merging, no tracking of assembly histories

A good model had: $\Omega_{m} = 0.20$, $\Omega_{gas} / \Omega_{DM} = 0.20$, $\alpha = 1/3$ (n = -1)

The beginnings

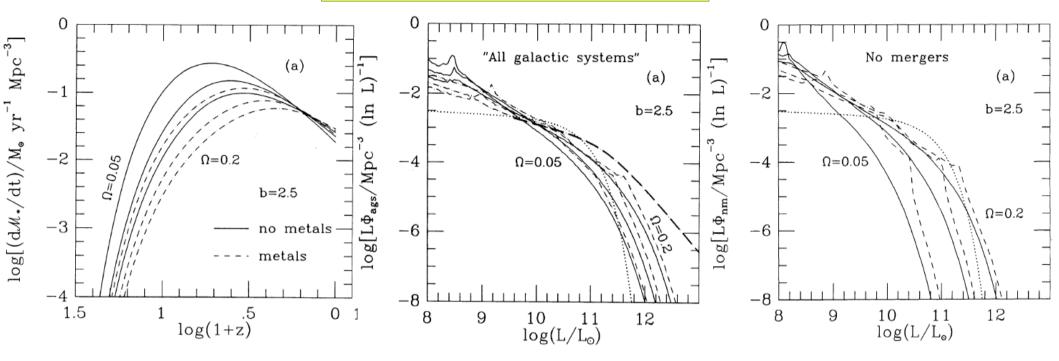


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First CDM models

White & Frenk 1991

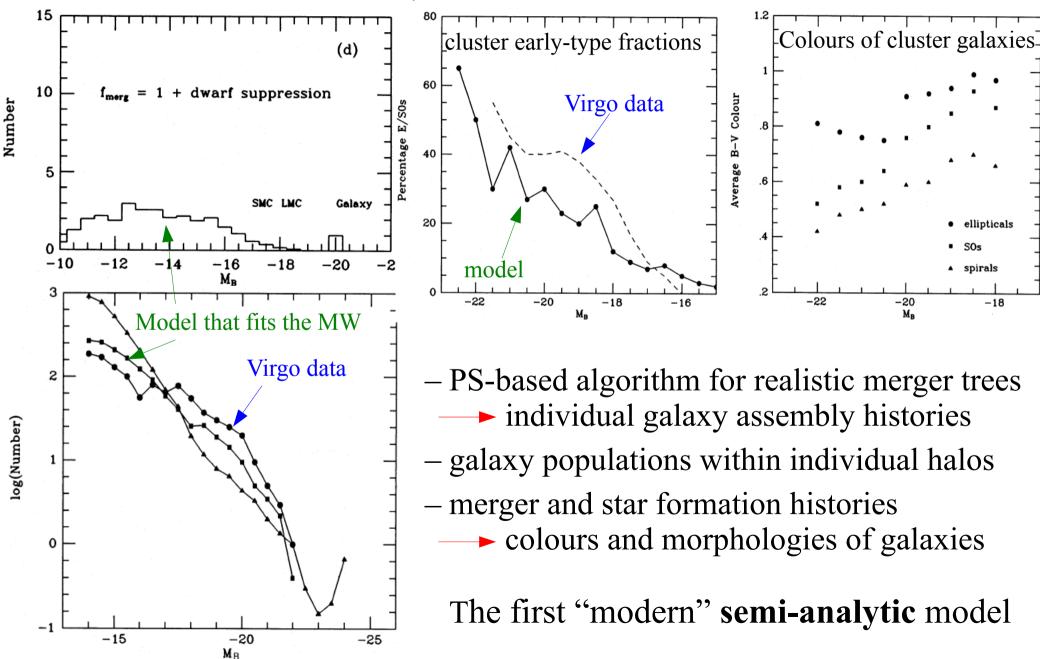


Updates to WR78 include:

- CDM initial power spectrum, EdS cosmology
- conditional MF, $\Phi_{PS}(M_{h,0}, z_0 | M_{h,1}, z_1)$ used to track stars statistically
- infall and cooling flow models with a SIS halo
- self-regulating star formation (now called the "bathtub model")
- chemical evolution modelling

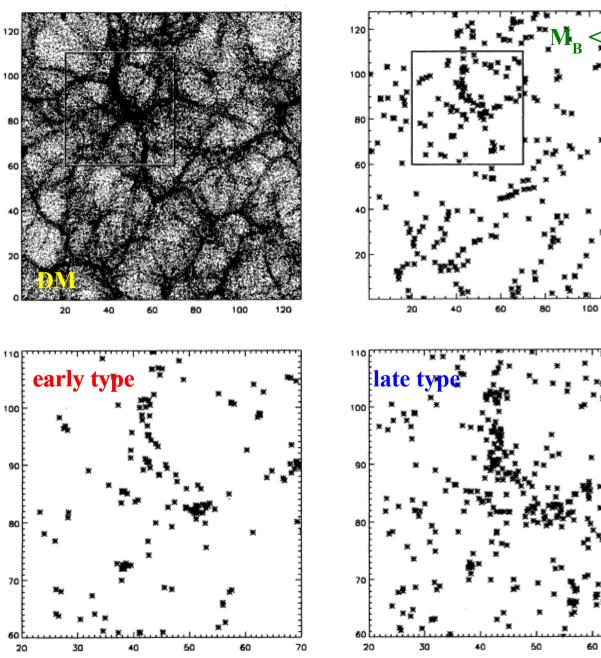
First use of merger trees

Kauffmann, White & Guiderdoni 1993



SA models in N-body halos

Kauffmann, Nusser & Steinmetz 1996



assign each simulated halo
 the galaxies from a PS tree
 correlation functions
 morphology vs density
 void probabilities

First use of an HOD to model galaxies in an N-body simulation

 Later HOD modelling took out the galaxy formation physics to gain simplicity and flexibility

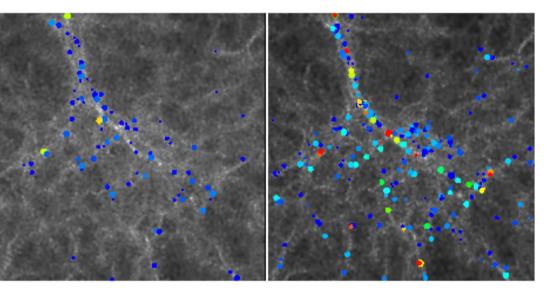
Semi-analytic simulations

Kauffmann, Colberg, Diaferio & White 1999

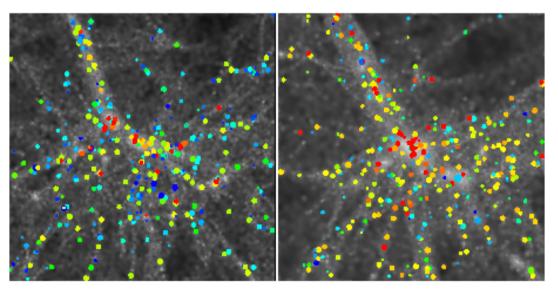
z=2

z=0

z=3



z=1



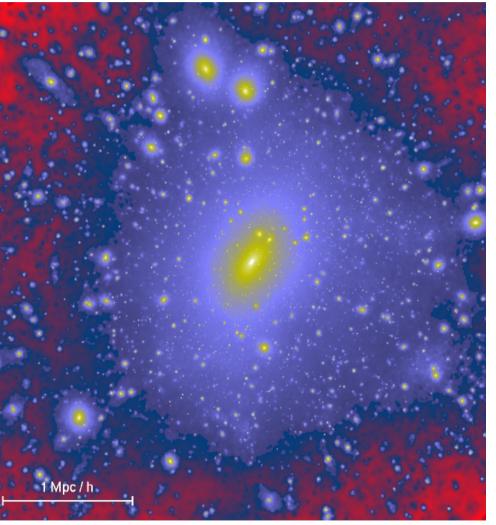
- Replace PS trees by *halo* merger trees built directly from an Nbody simulation
- central halo particles represent central galaxies and trace satellite galaxies after halo nergers
 - the formation of the galaxy population is tracked directly
 - large-scale progenitors structure can be studied, e.g .protoclusters
 - "assembly bias" effects automatically included

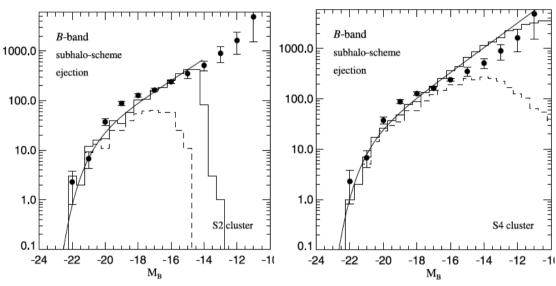
A simulation of galaxy formation

High-resolution SA simulations

Springel, White, Tormen & Kauffmann 2001







- high-resolution simulations + subhalo
 identifier ---- subhalo merger trees
 - halos of *some* satellite galaxies can be followed explicitly
 - tidal stripping and dynamical friction are simulated consistently
- resolution tests needed to check consistency of "orphans" and subhalo satellites

Millennium Simulation

ne 2005 | www.nature.com/nature | £10

THE INTERNATIONAL WEEKLY JOURNAL OF SCIENCE

The most highly cited astrophysics article ever published in Nature

Springel et al 2005

GENOME EDITING Rewriting the rules for gene therapy BCL-2 INHIBITORS Potent new antitumour compounds HUMAN BEHAVIOUR Oxytocin — the 'trust hormone'

SURPRISING DINOSAURS A sauropod, by a short neck INSIDE: UP-TO-THE-MINUTE REVIEWS ON AUTOIMMUNITY



EVOLUTION OF THEUNIVERSE

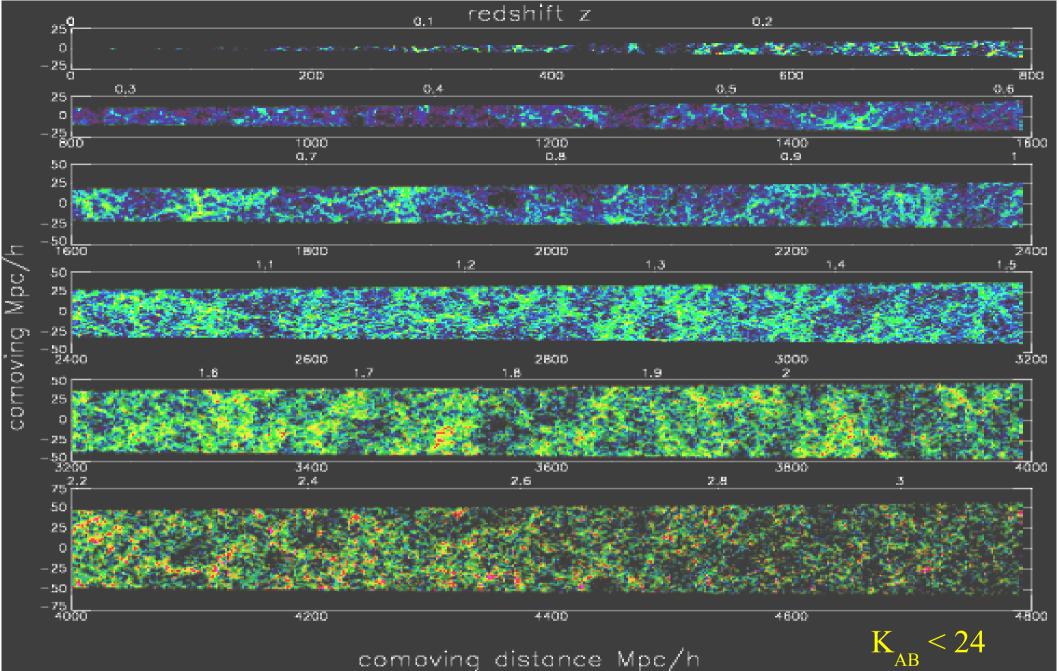
Supercomputer simulation of the growth of 20 million galaxies

Millennium Simulation

CIA2 Great V 30000 CZ IKM/SJ \$5000 L-Galaxies used to simulate the formation/evolution of $2 \times 10^7 - 1$ $\sim 2 \times 10^7$ galaxies Springel et al 2006 466 ISu

billion years

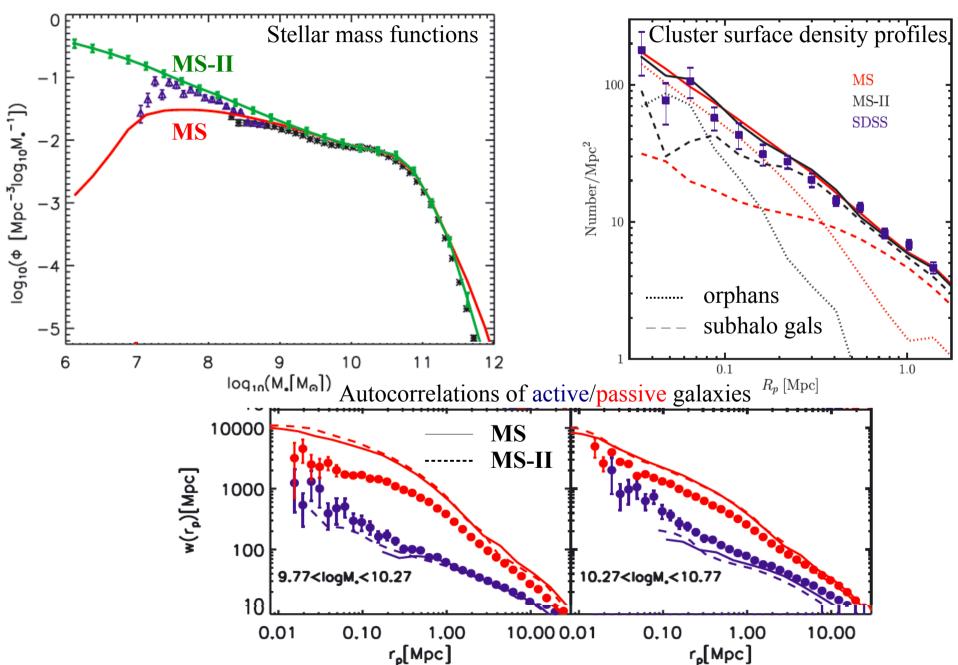
214m/s



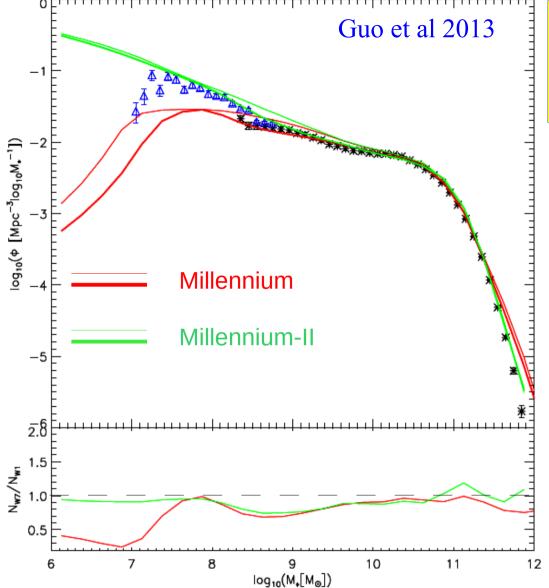
L-Galaxies used to simulate the formation/evolution of $\sim 2x10^7$ galaxies from z = 10 to z = 0

Kitzbichler & White 2007

The Millennium-II: providing convergence tests



Guo et al 2011



Switching from WMAP1 to WMAP7

Suitable scaling of L, M and t allows the cosmology of the MS and MS-II to be reinterpreted

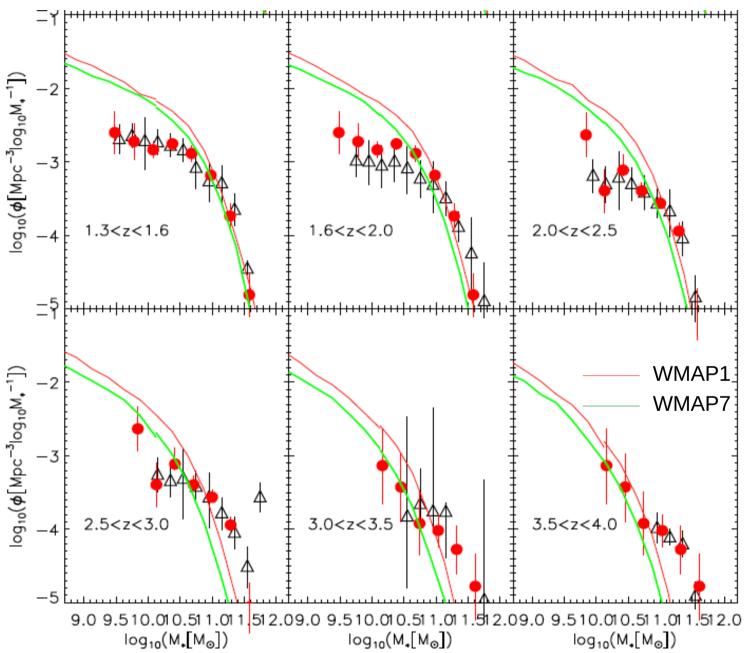
Small shifts in the parameters of the galaxy formation model allow the galactic stellar mass function to be fit equally well in the two different cosmologies despite

 $\sigma_{0} = 0.81$

Parameter	Description	WMAP1	WMAP7
α	Star formation efficiency	0.02	0.015
ε	Amplitude of SN reheating efficiency	6.5	4.5
β_1	Slope of SN reheating efficiency	3.5	4
V_{reheat}	normarlization of SN reheating efficiency dependence on Vmax	70	80
η	Amplitude of SN ejection efficiency	0.32	0.33
β_2	Slope of SN ejection efficiency	3.5	6.5
V_{eject}	normarlization of SN ejection efficiency dependence on Vmax	70	80
κ	Hot gas accretion efficiency onto black holes	1.5×10^{-5}	6.0×10^{-6}

 $\sigma_{0} = 0.90$

Switching from WMAP1 to WMAP7

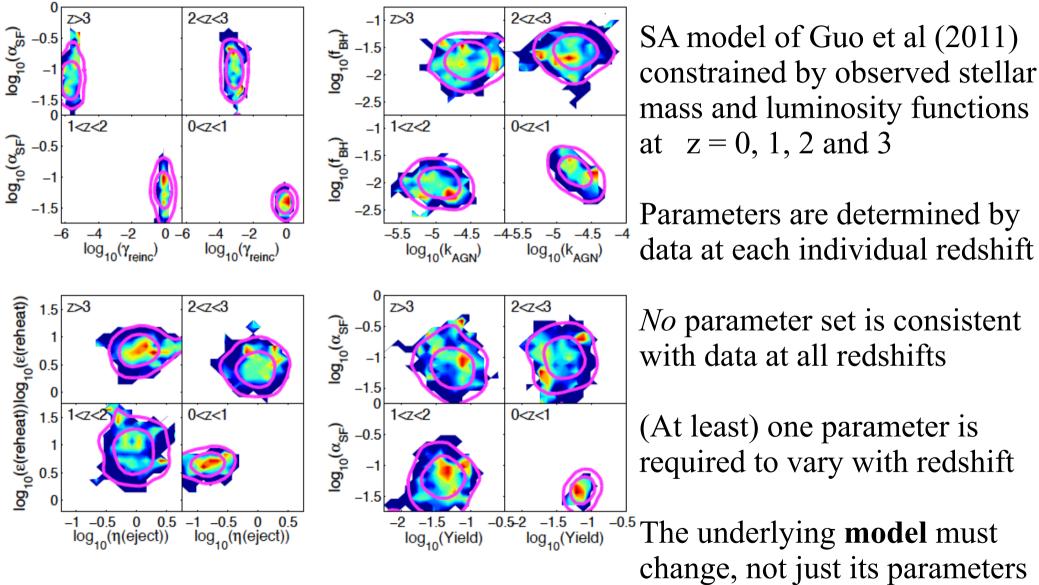


New cosmology does not fix the problem that lowmass galaxies form too early in the Guoll model

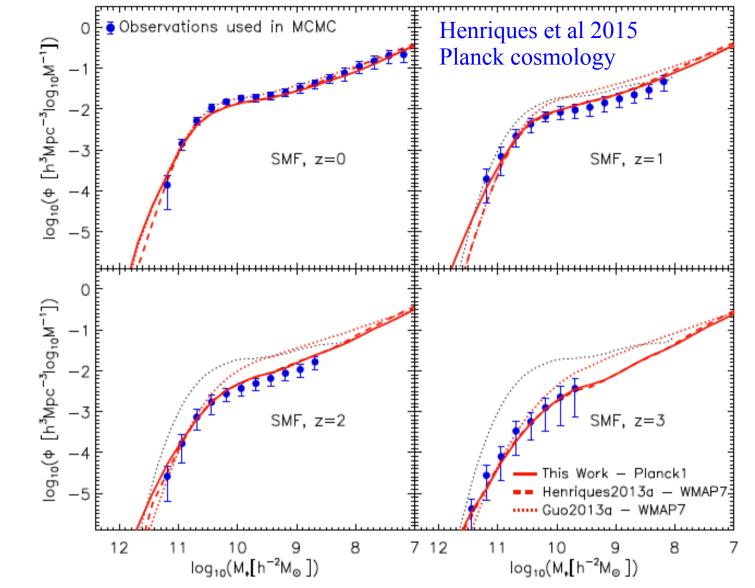
Guo et al 2013

This must be an astrophysical modelling problem

MCMC allows systematic exploration of parameter space



Henriques et al 2013

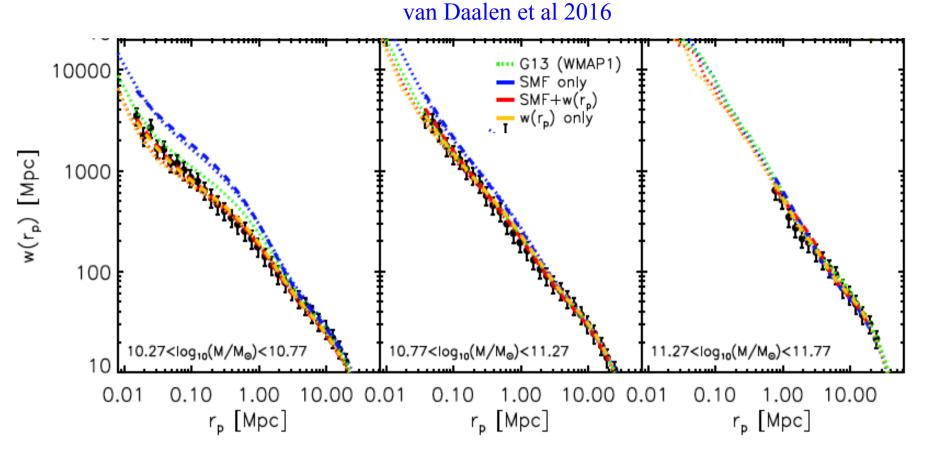


Changing the assumed timescale for reincorporation of wind ejecta

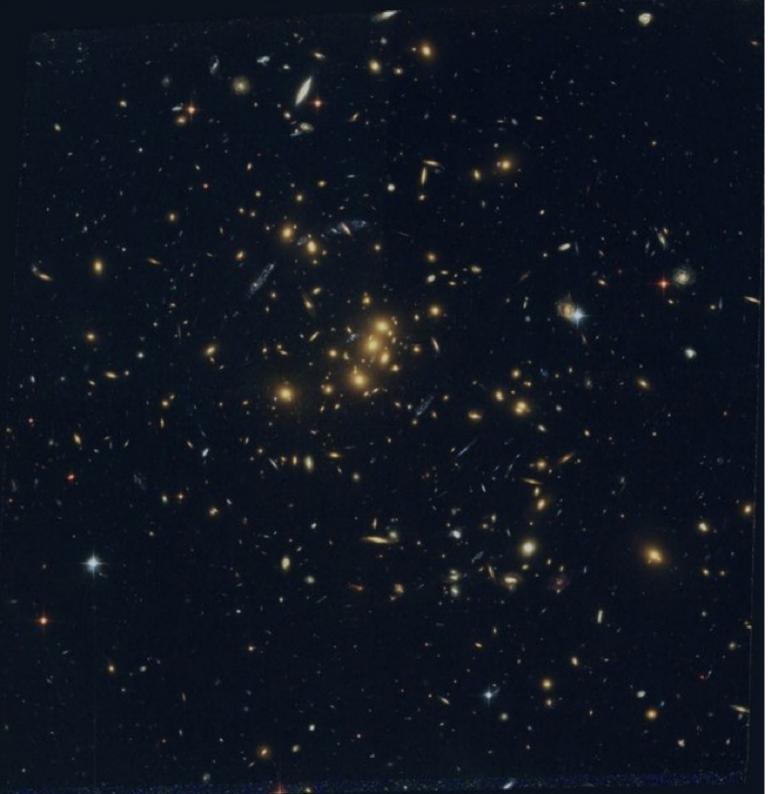
$$t_{return} = const. / H(z) V_{halo} \rightarrow t_{return} = const. / M_{halo}$$

allows a good fit to data at all redshifts for the same # of parameters

MCMC can also be used to explore clustering constraints



- Special techniques needed to evaluate clustering fast and accurately enough for implementation in an MCMC procedure
- Clustering fits can be improved for only a small penalty in the SMF
- <u>Only</u> technique currently able to explore constraints from abundances clustering and evolution simultaneously and systematically



C10024

Harsono & De Propris 2007

z = 0.40

3.4' x 3.4'

HST/ACS



"Cl0024"

$$M_{200} = 7 \times 10^{14} M_{\odot}$$

z = 0.41

3.4' x 3.4'

HST/ACS F475W, F625W, F850LP

10,000sec/filter

Overzier et al 2014

The Millennium Run Observatory



"Cl0024"

$$M_{200} = 7 \times 10^{14} M_{\odot}$$

z = 0.41

3.4' x 3.4'

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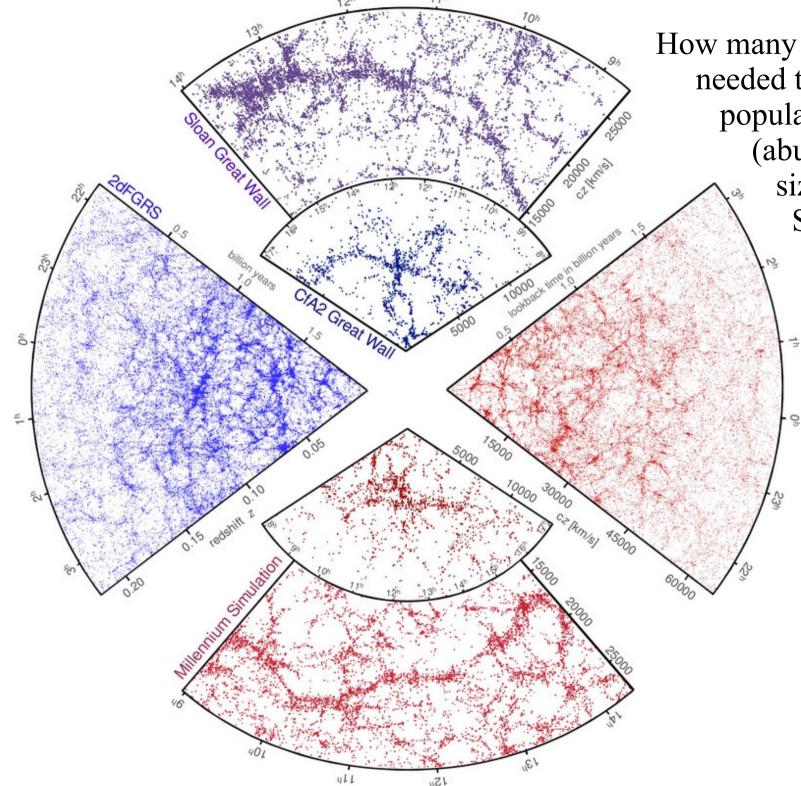
Overzier et al 2014

The Millennium Run Observatory

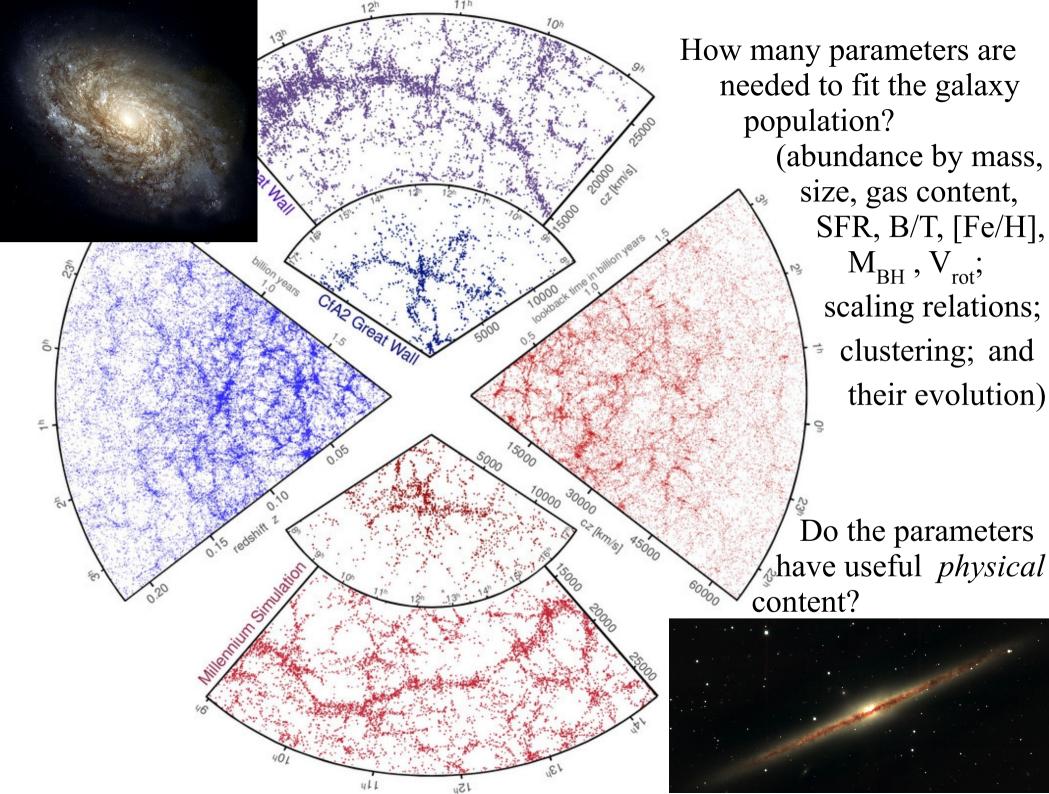
Six parameters fine-tuned to fit a single curve

Planck+WP

Parameter	Best fit	68% limits
$\Omega_{ m b}h^2$	0.022032	0.02205 ± 0.00028
$\Omega_{\rm c} h^2$	0.12038	0.1199 ± 0.0027
$100\theta_{\rm MC}$	1.04119	1.04131 ± 0.00063
τ	0.0925	$0.089^{+0.012}_{-0.014}$
$n_{\rm s}$	0.9619	0.9603 ± 0.0073
$\ln(10^{10}A_{\rm s})$	3.0980	$3.089^{+0.024}_{-0.027}$



How many parameters are needed to fit the galaxy population? (abundance by mass, size, gas content, SFR, B/T, [Fe/H], M_{BH}, V_{rot}; scaling relations; clustering; and their evolution)



Population simulations provide a tool...

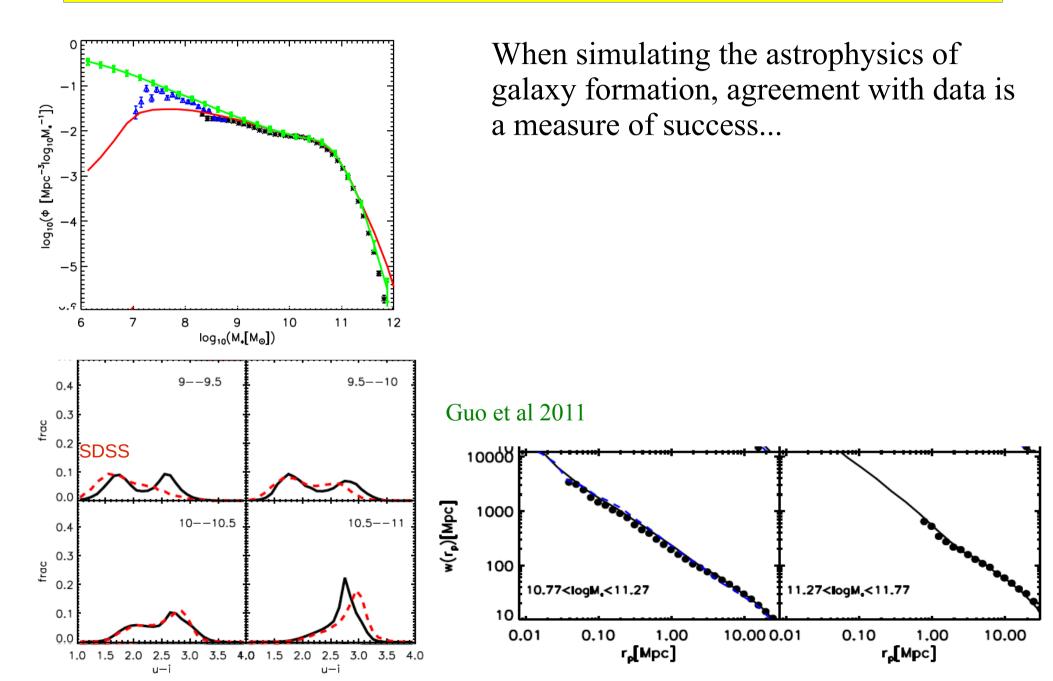
To explore the relative importance and the physical scaling of the many processes that affect stars, gas and central black holes within growing Λ CDM structures

To understand how the these processes interact to produce the various observed <u>population</u> properties of galaxies and their evolution – abundances, scaling relations, clustering

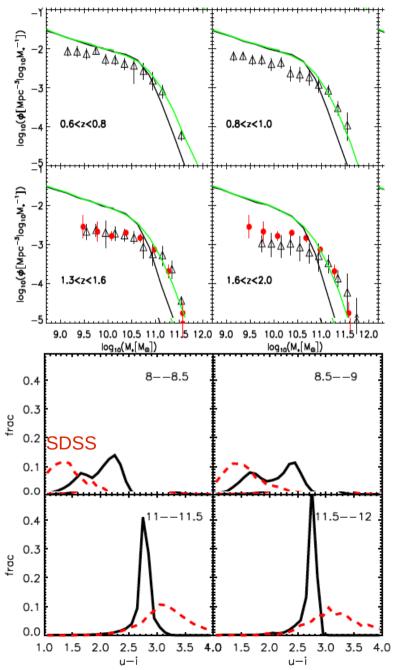
To allow interpretation of large observational surveys in terms of the rates, efficiencies and significance of these processes

To investigate whether uncertainties in the astrophysics of galaxy formation compromise the "precision cosmology" programme of such surveys

How do we learn from population simulations?



How do we learn from population simulations?

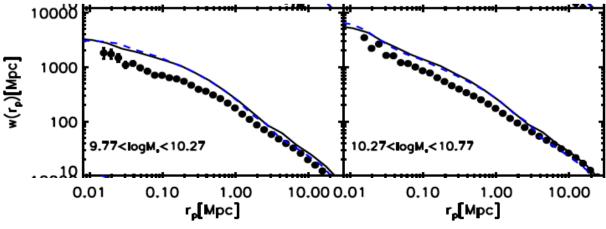


When simulating the astrophysics of galaxy formation, agreement with data is a measure of success...

...but it is the failures which show where there is missing or inadequate physics

cosmology? star formation? enrichment and feedback? environmental effects?

Guo et al 2011



How do we learn from population simulations?

