

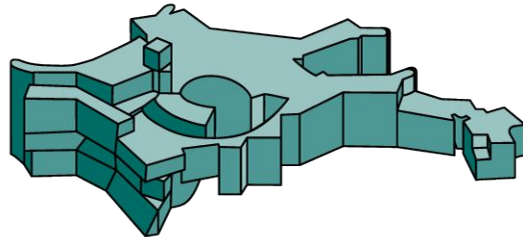
# Developing a new physics-first galaxy formation model for cosmological simulations

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07/15/2026, (with Matt Ho and Volker Springel)

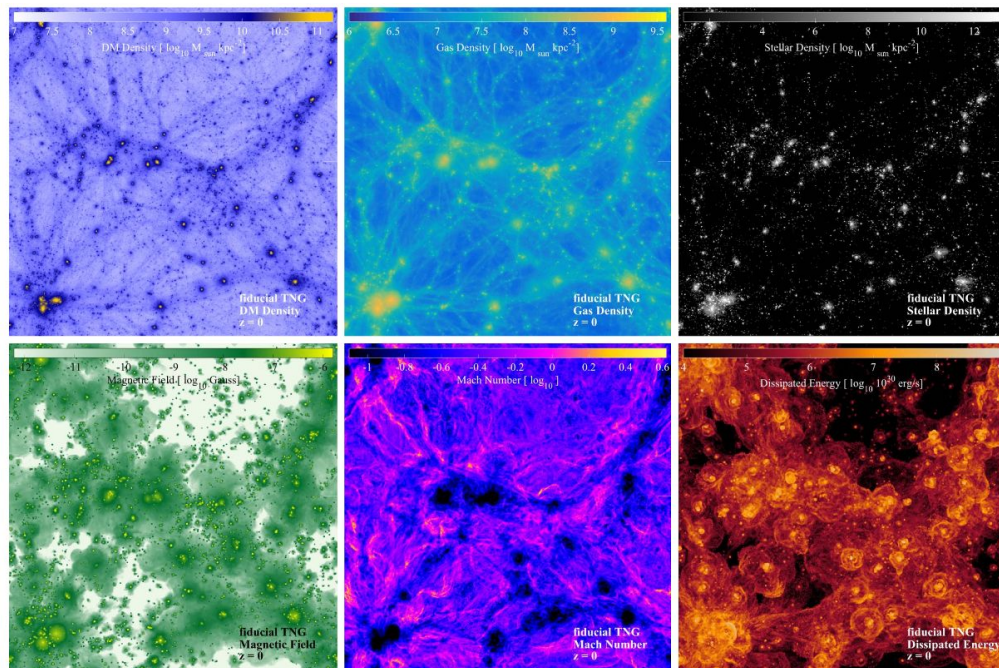
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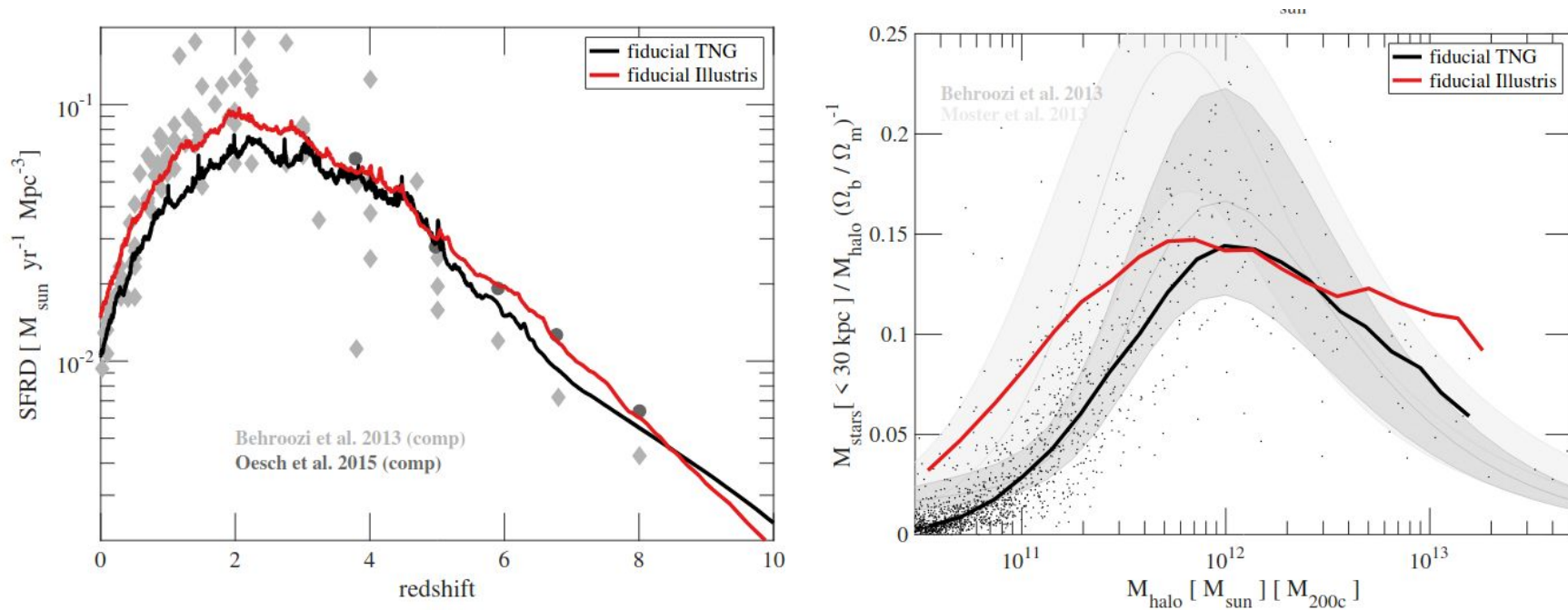
# The Illustris-TNG model as a baseline

The TNG-model (Pillepich 2018) remains one of the most successful galaxy formation models until today, including subgrid models for:

- BH accretion and feedback
- Star formation and cooling
- Stellar winds
- Magnetohydrodynamics

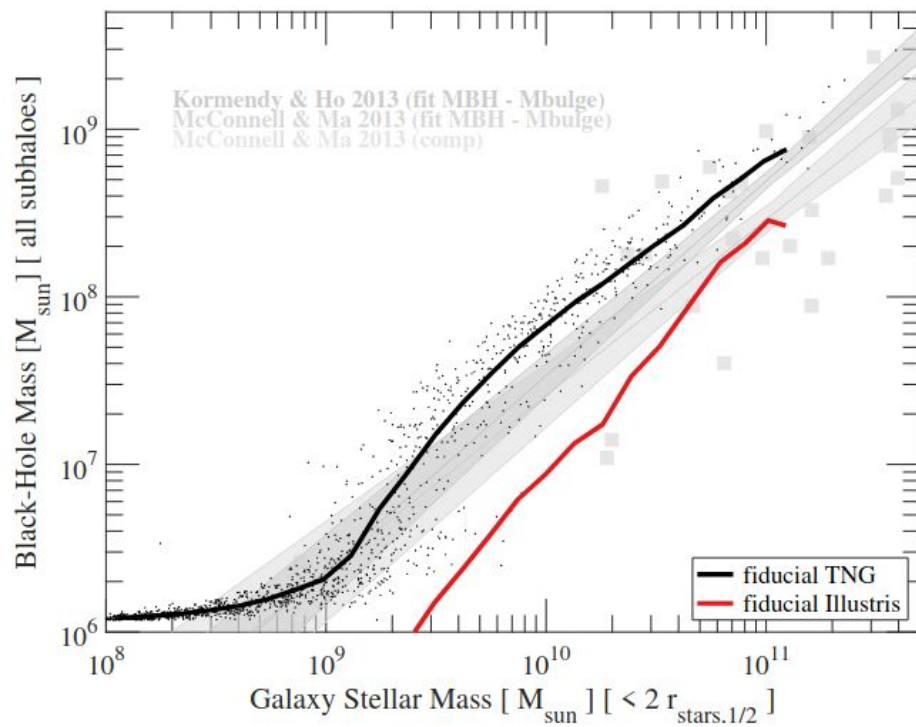
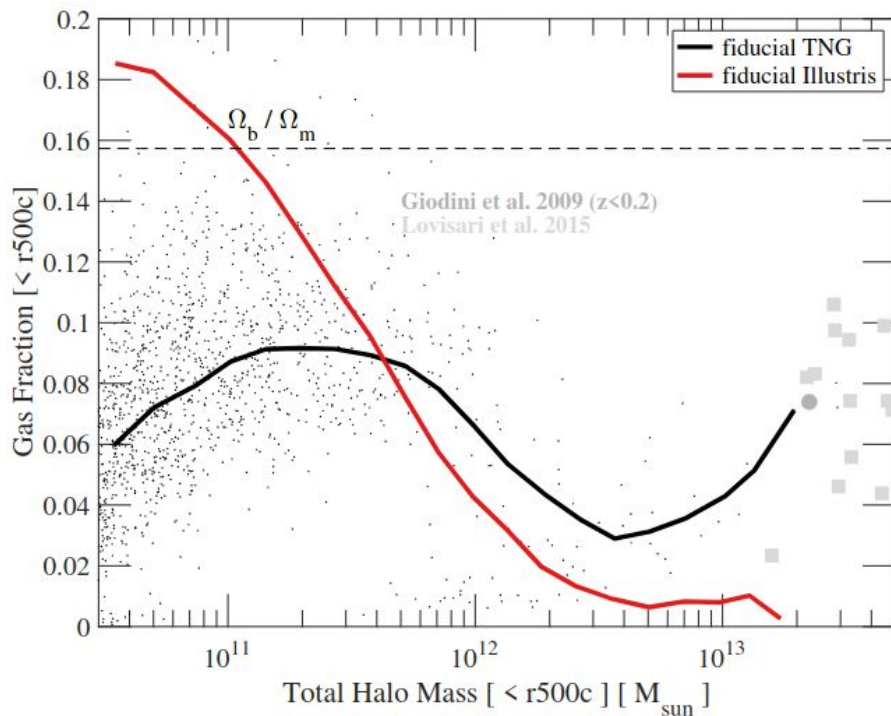


# Fiducial TNG scaling relations



Pillepich et al. 2018

# Fiducial TNG scaling relations



Pillepich et al. 2018

# The TNG model today

TNG remains a reference standard for galaxy formation models, but:

- Star formation model includes a relatively ad-hoc ISM description
- Stellar feedback is included through hydrodynamically decoupled winds
- Black holes are fixed to the center of a halo's potential and seeded in fairly heavy haloes
- Black hole accretion model assumes spherical symmetry (Bondi-Hoyle accretion) and black hole feedback cannot explain the low observed gas fractions in massive haloes
- There is a growing desire for new, physics-motivated subgrid models that improve our understanding (and maybe the match with observations)

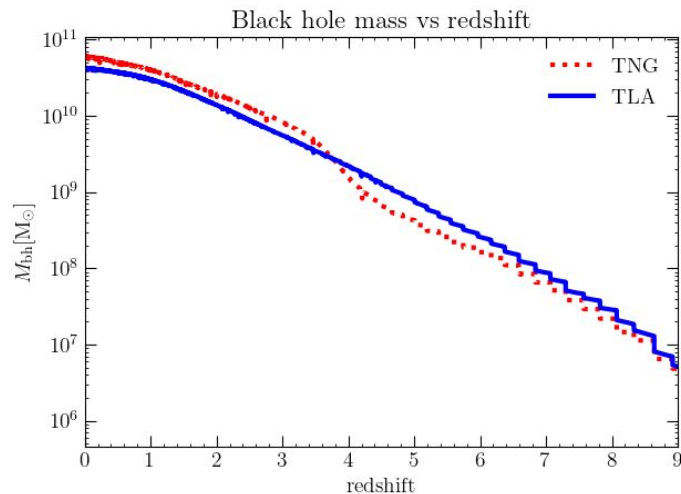
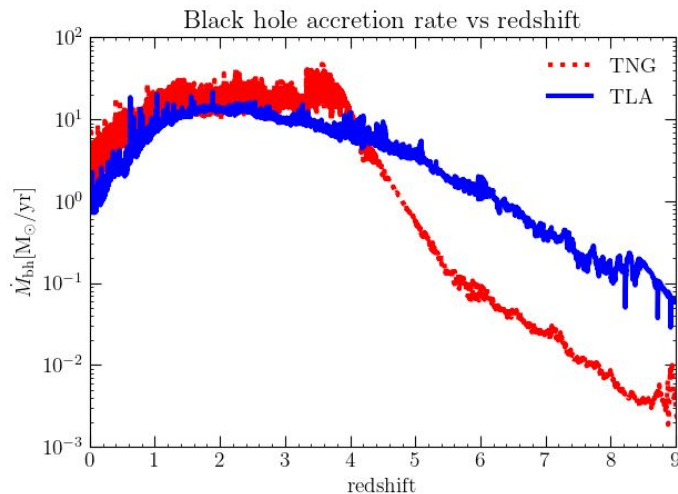
# The Learning the Universe collaboration

Within the LtU collaboration, part of our effort focuses on the development of new, physics-informed subgrid models for galaxy formation.

- The Arkenstone galactic wind model (Bennet et al. 2025)
- The PRFM-based star formation model (Burger et al. 2026)
- The BRAHMA model for black hole seeding (Bhowmick et al. 2024)
- A new dynamical friction model for black holes (Genina et al. 2024)
- Blended (Illustris + TNG) black hole feedback (Chen et al. in prep.)

A complete new galaxy formation model requires the joint calibration of the combination of all of these subgrid models!

# A case study on torque-limited black hole accretion



- Torque-limited accretion allows for higher accretion rates than Bondi accretion at high redshift
- This can potentially help to explain luminous high-z quasars
- The model has to be calibrated together with the TNG feedback model to match observations

# The TLA model and its equations

Torque-limited accretion, as introduced in Angles-Alcazar et al. 2013, calculates two different BH accretion rates for hot gas and cold gas:

$$\dot{M}_{\text{Bondi}} = \epsilon_m \frac{4\pi G^2 M_{\text{BH}}^2 \rho}{(v^2 + c_s^2)^{3/2}}$$



Bondi rate  
normalization  
for hot gas

$$\dot{M}_{\text{Torque}} = \epsilon_{\text{T}} f_d^{5/2} \times \left( \frac{M_{\text{BH}}}{10^8 M_\odot} \right)^{1/6} \times \left( \frac{M_{\text{enc}}(R_0)}{10^9 M_\odot} \right) \left( \frac{R_0}{100 \text{ pc}} \right)^{-3/2} \\ \times \left( 1 + \frac{f_0}{f_{\text{gas}}} \right)^{-1} M_\odot \text{ yr}^{-1}$$



TLA normalization  
for cold gas

- Cold and hot gas separated by temperature threshold
- Disk fraction calculated within  $R_0$
- Gas fraction in the disk as well
- $f_0$  is a normalization parameter

# The TNG feedback model

The TNG feedback model decides which feedback channel to use based on whether the BH is in high-accretion state or low accretion state:

$$\frac{\dot{M}_{\text{Bondi}}}{\dot{M}_{\text{Edd}}} \geq \chi \quad \chi = \min \left[ \chi_0 \left( \frac{M_{\text{BH}}}{10^8 M_{\odot}} \right)^{\beta}, 0.1 \right]$$

Quasarthreshold parameter, fixing the transition between quasar and radio mode

Note that varying the threshold mass would lead to a degenerate effect, and that  $\beta$  is fixed to 2.

# The TNG feedback model

In the quasar mode, the injected energy is regulated by the feedback efficiency parameter

$$\Delta \dot{E}_{\text{high}} = \epsilon_{\text{f,high}} \epsilon_{\text{r}} \dot{M}_{\text{BH}} c^2$$

Blackhole feedback efficiency parameter that regulates the quasar mode

For the radio mode, the efficiency function is more complicated:

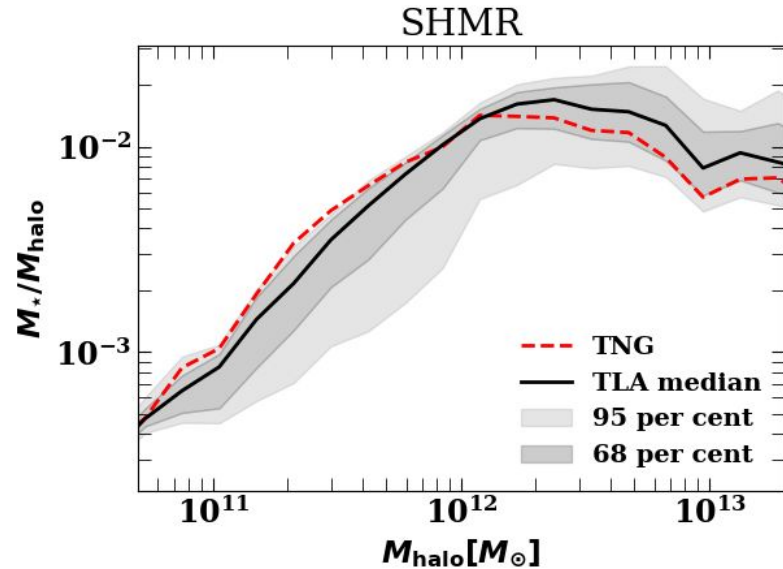
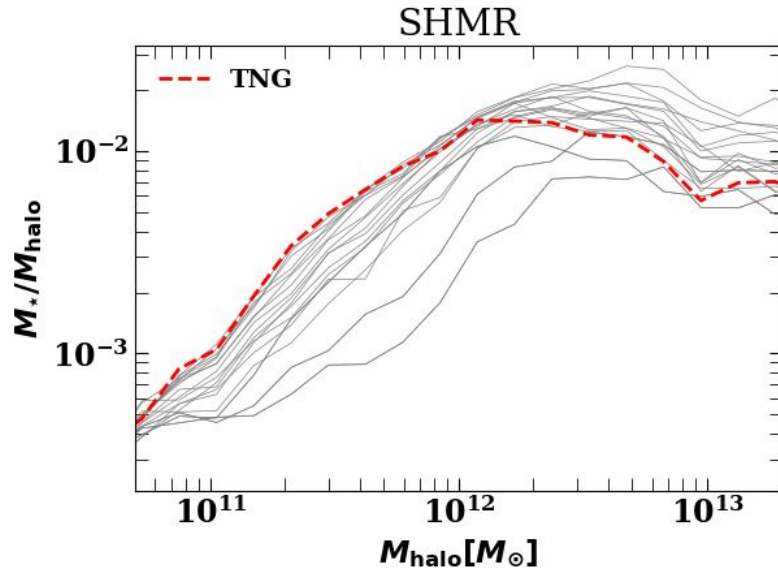
Minimum density factor for radio feedback

$$\Delta \dot{E}_{\text{low}} = \epsilon_{\text{f,kin}} \dot{M}_{\text{BH}} c^2 \quad \epsilon_{\text{f,kin}} = \min \left( \frac{\rho}{f_{\text{thresh}} \rho_{\text{SFthresh}}}, 0.2 \right)$$

# The TLA calibration suite

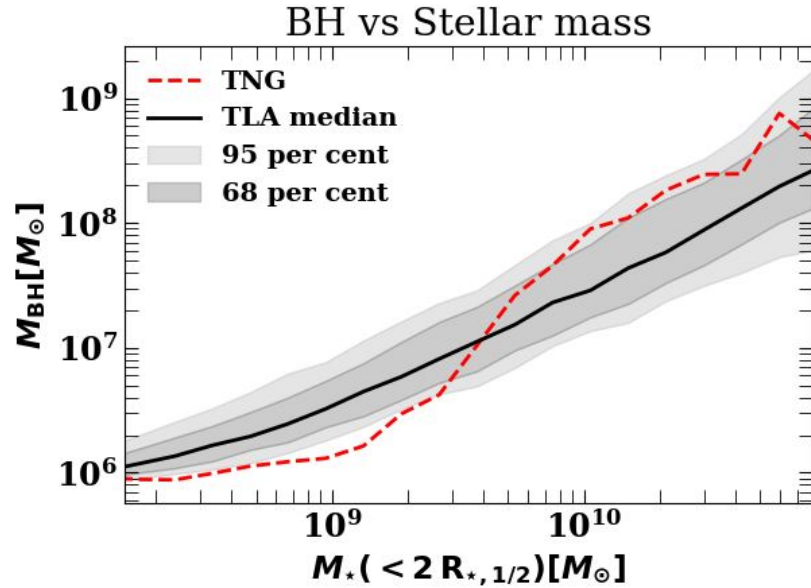
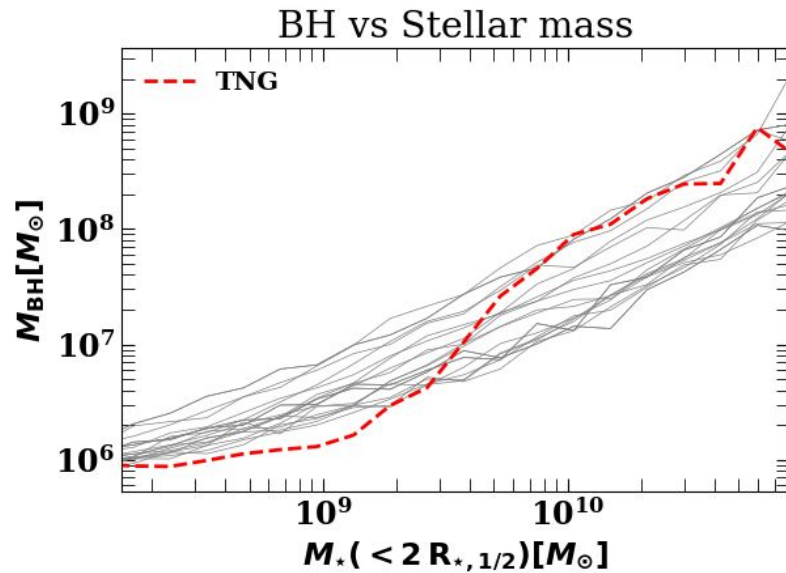
- We generate 150 latin hypercube samples (in log-space) of our 5 selected parameters. We vary:
  - Bondi rate normalization and TLA normalization from 0.1 to 10
  - The quasar threshold from 0.0006 to 0.006
  - Quasar feedback efficiency from 0.03 to 0.3
  - Minimum density factor for radio feedback from 0.002 to 0.5
- We then run 150 hydro simulations in  $(50 \text{ Mpc})^3$  boxes at TNG-300 resolution to calibrate:
  - We perform PCA on 6 measured scaling relations and then train Gaussian process emulators on the principal components
  - Using the emulator, we carry out sensitivity studies
  - Finally, we attempt to recover withheld parameter vectors using MCMC

# TLA: Stellar-to-halo mass relation



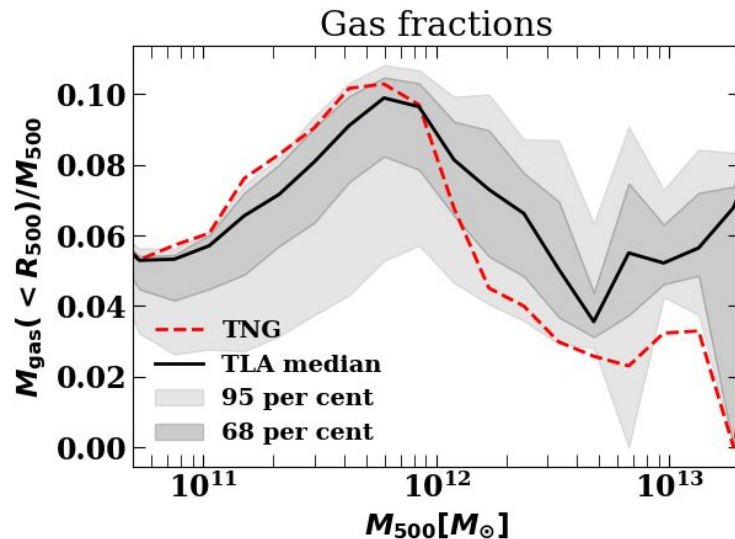
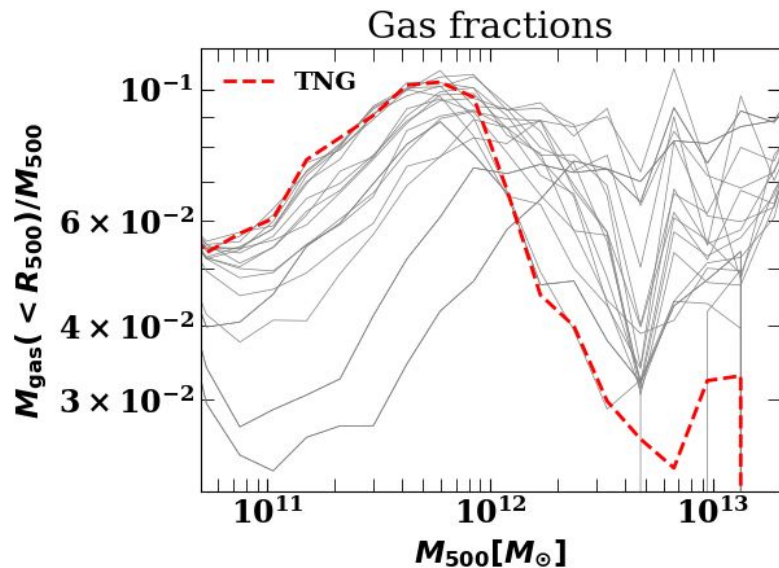
- Stellar-to-halo mass ratios that are covered by the simulation suite encompass the TNG line
- No individual line, however, follows the shape of the TNG stellar-to-halo-mass relation

# TLA: black hole vs stellar mass



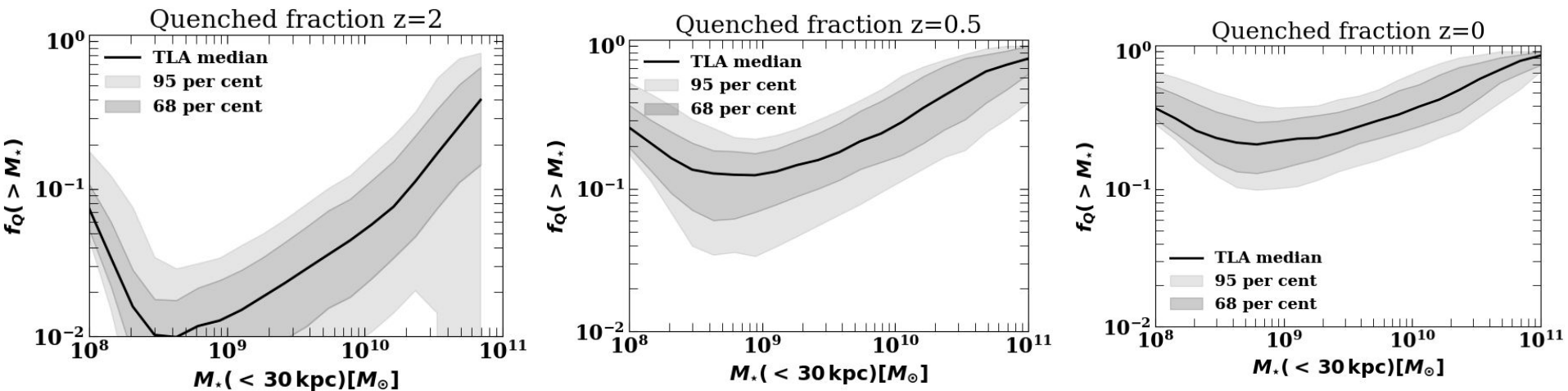
- In all TLA runs, black hole mass scales with stellar mass almost as a pure power law
- This is in contrast to the S shape obtained in the TNG run

# TLA: Gas fractions



- Simulated gas fractions cover a large range, but at the high mass end most of them fall above the TNG line
- This indicates that possibly, a more substantial change in the feedback model is needed (see upcoming work of Chen et al.)

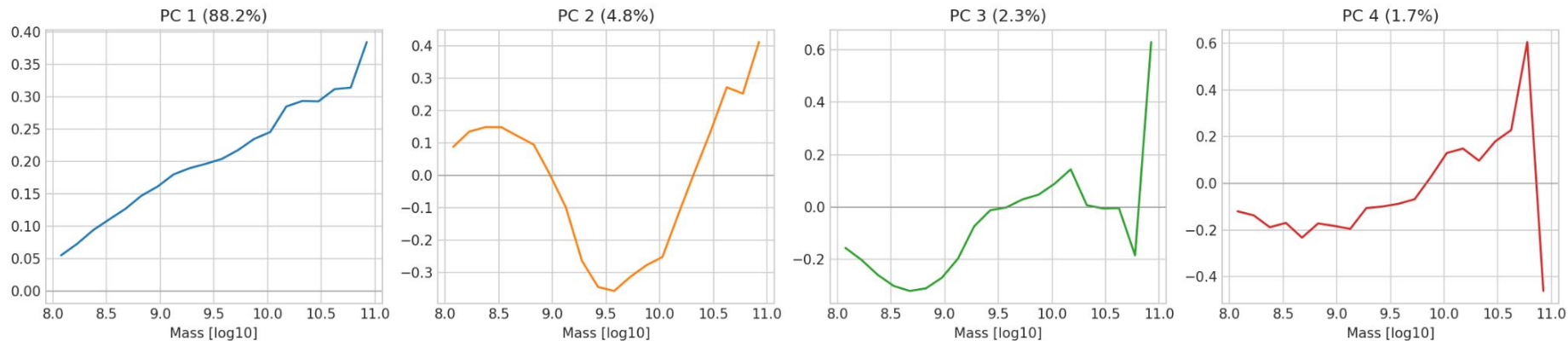
# TLA: Quenched fractions



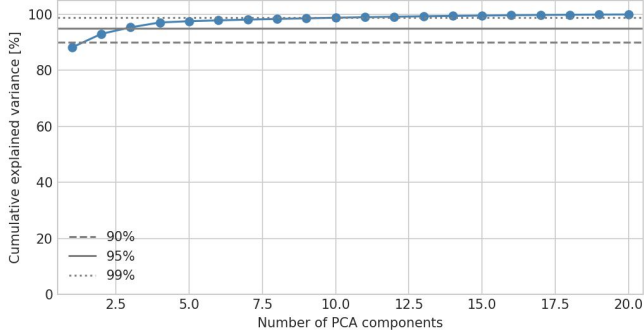
- Galaxies become increasingly quenched towards lower redshift
- There is a large spread in quenched fraction for the different model variations, in particular at high  $z$

# Principal component analysis and LOO validation: BH vs stellar mass

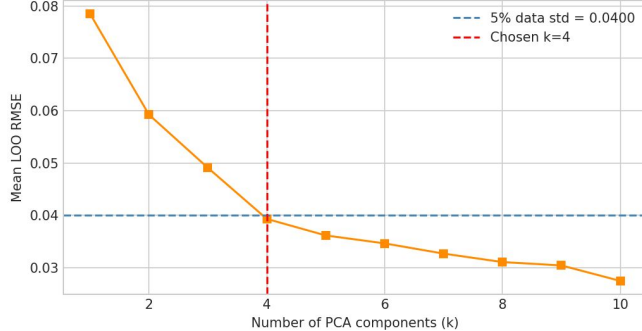
bhstellar — PC Mode Vectors



bhstellar — Explained Variance

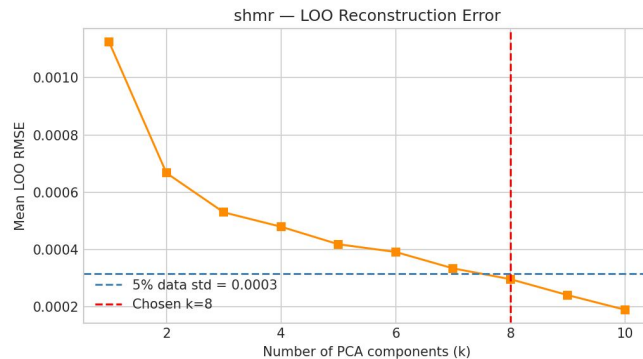
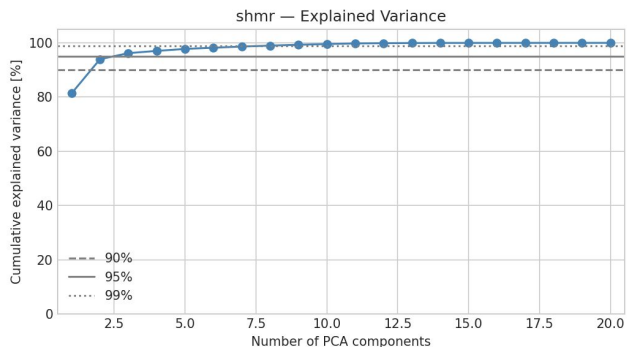
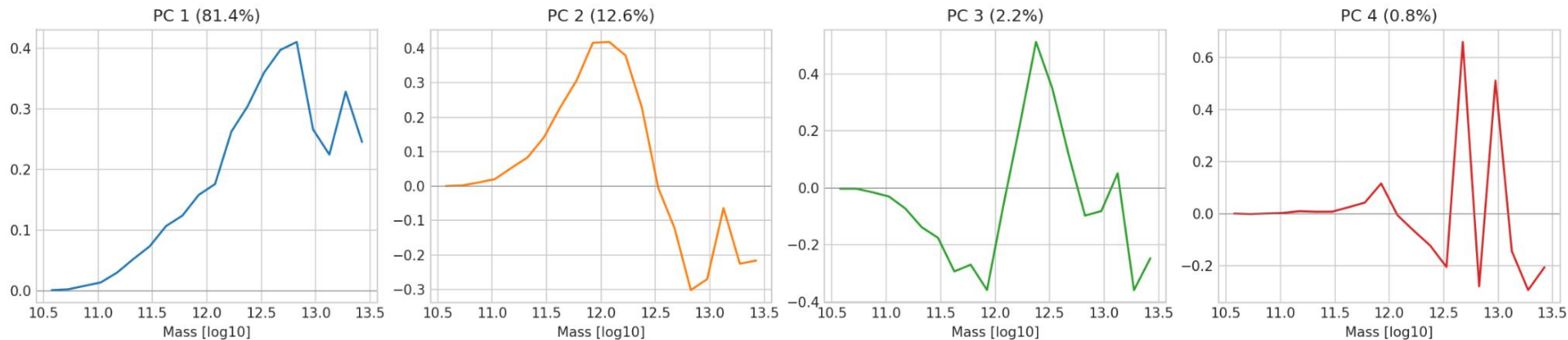


bhstellar — LOO Reconstruction Error



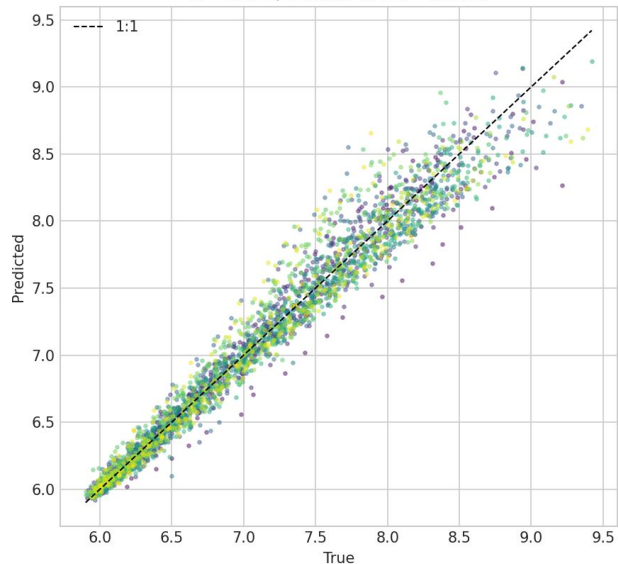
# Principal component analysis and LOO validation: Stellar-to-halo mass relation

shmr — PC Mode Vectors

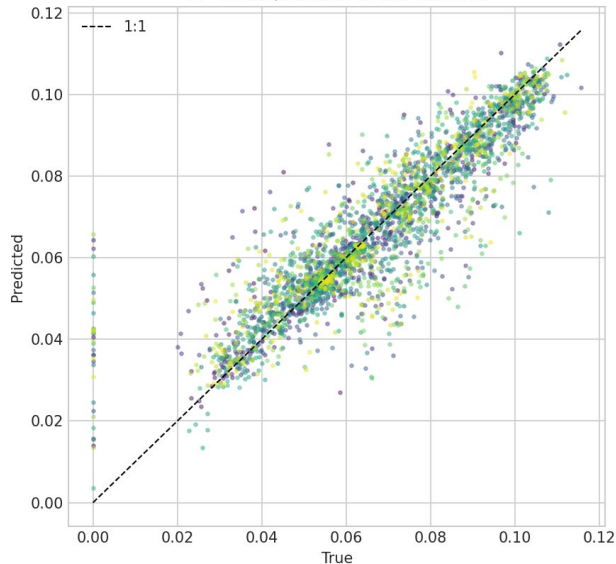


# Gaussian process emulator: examples

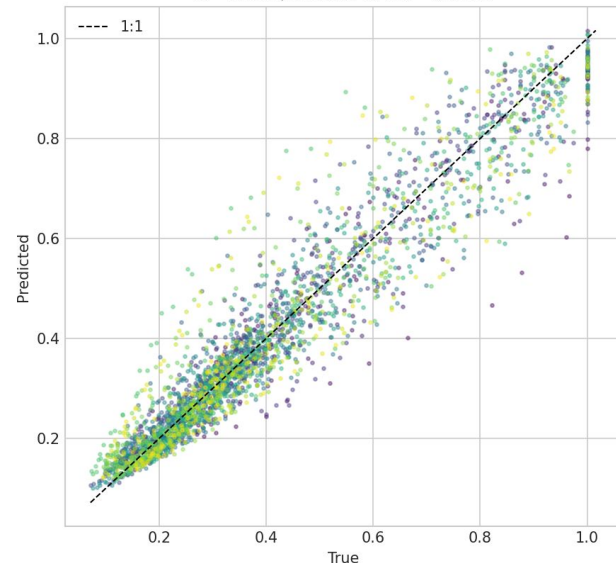
bhstellar — LOO Predicted vs True  
 $R^2=0.980$ , median RMSE=0.1063



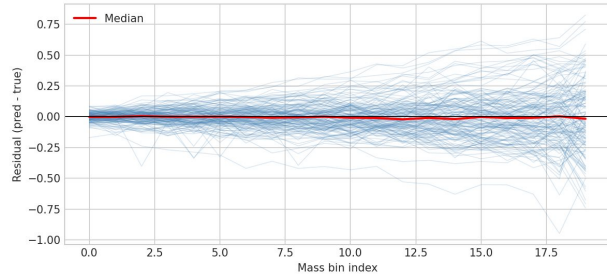
gasfrac — LOO Predicted vs True  
 $R^2=0.818$ , median RMSE=0.0081



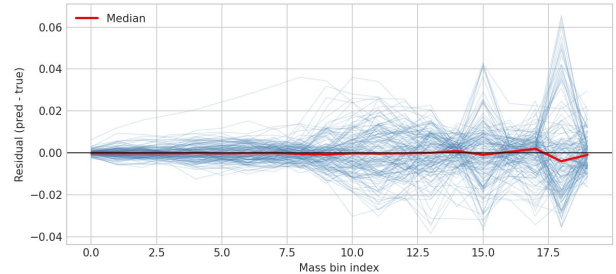
qf0 — LOO Predicted vs True  
 $R^2=0.943$ , median RMSE=0.0506



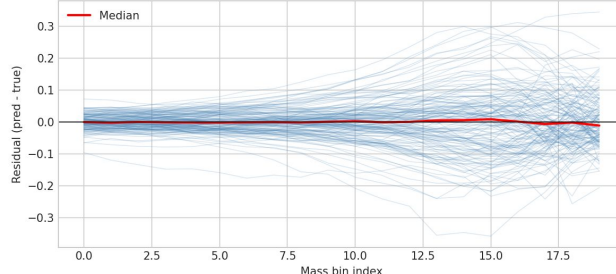
bhstellar — LOO Residuals



gasfrac — LOO Residuals

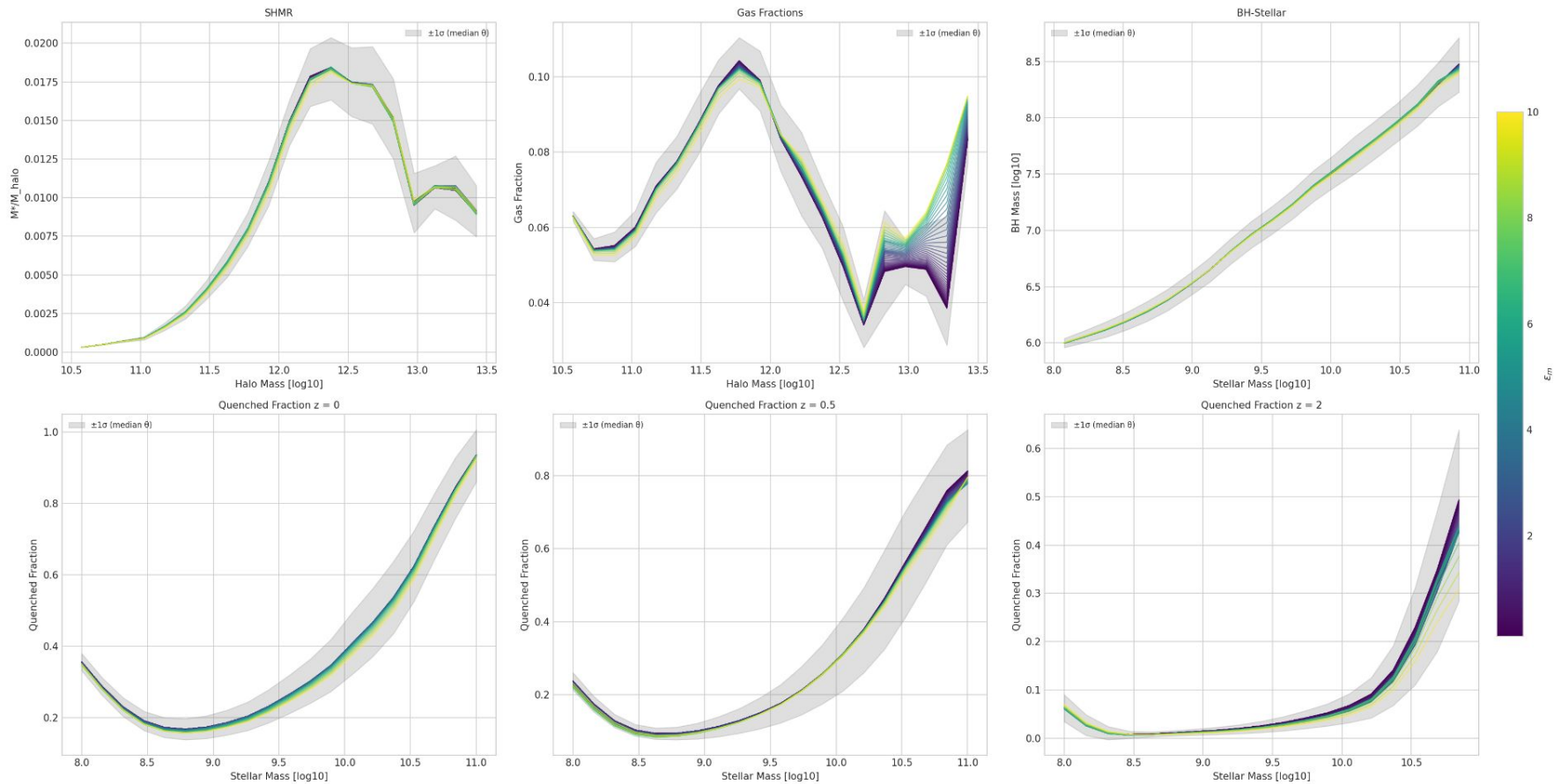


qf0 — LOO Residuals



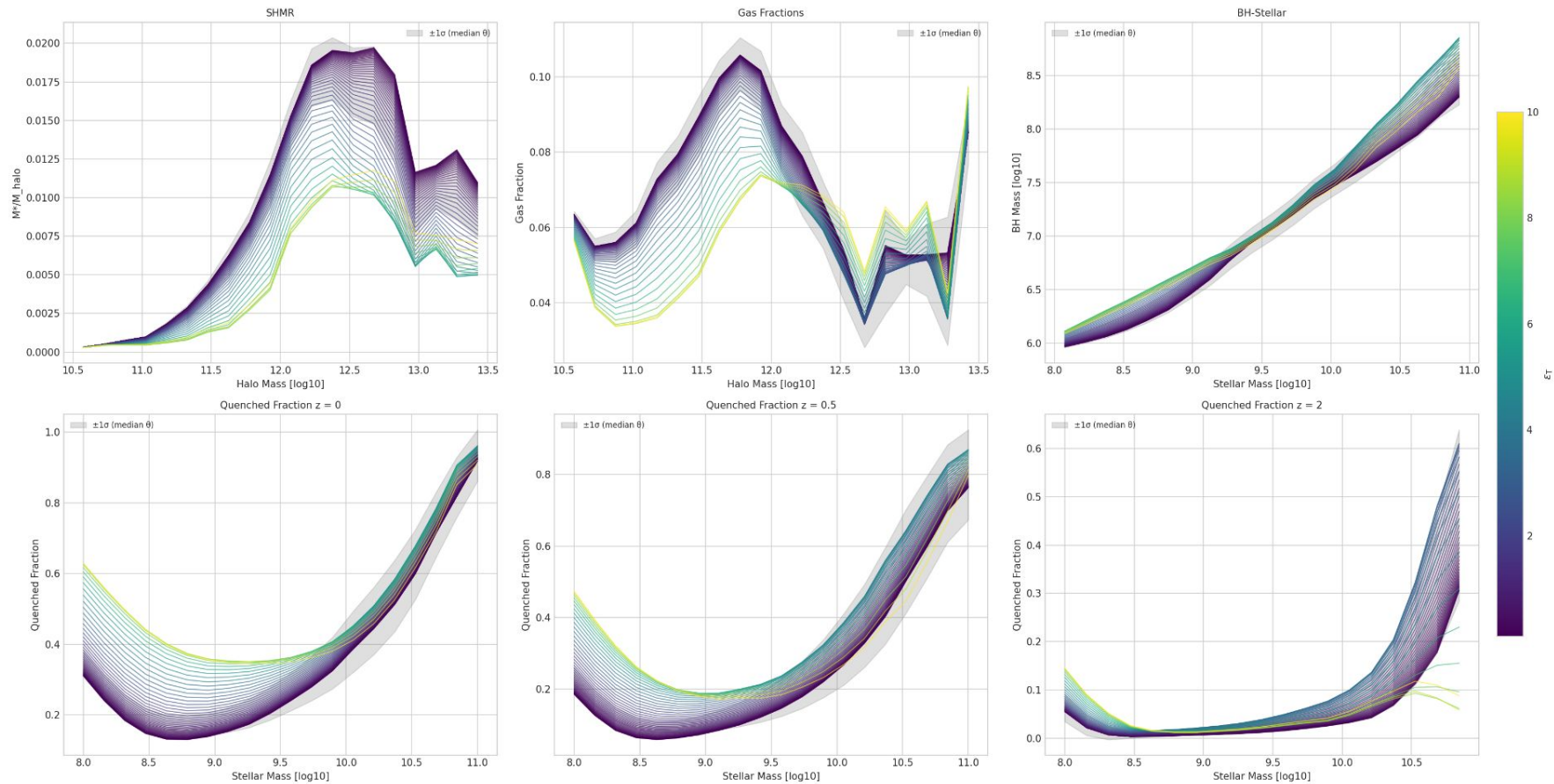
# Sensitivity studies (Bondi rate)

1P Variation:  $\epsilon_m$



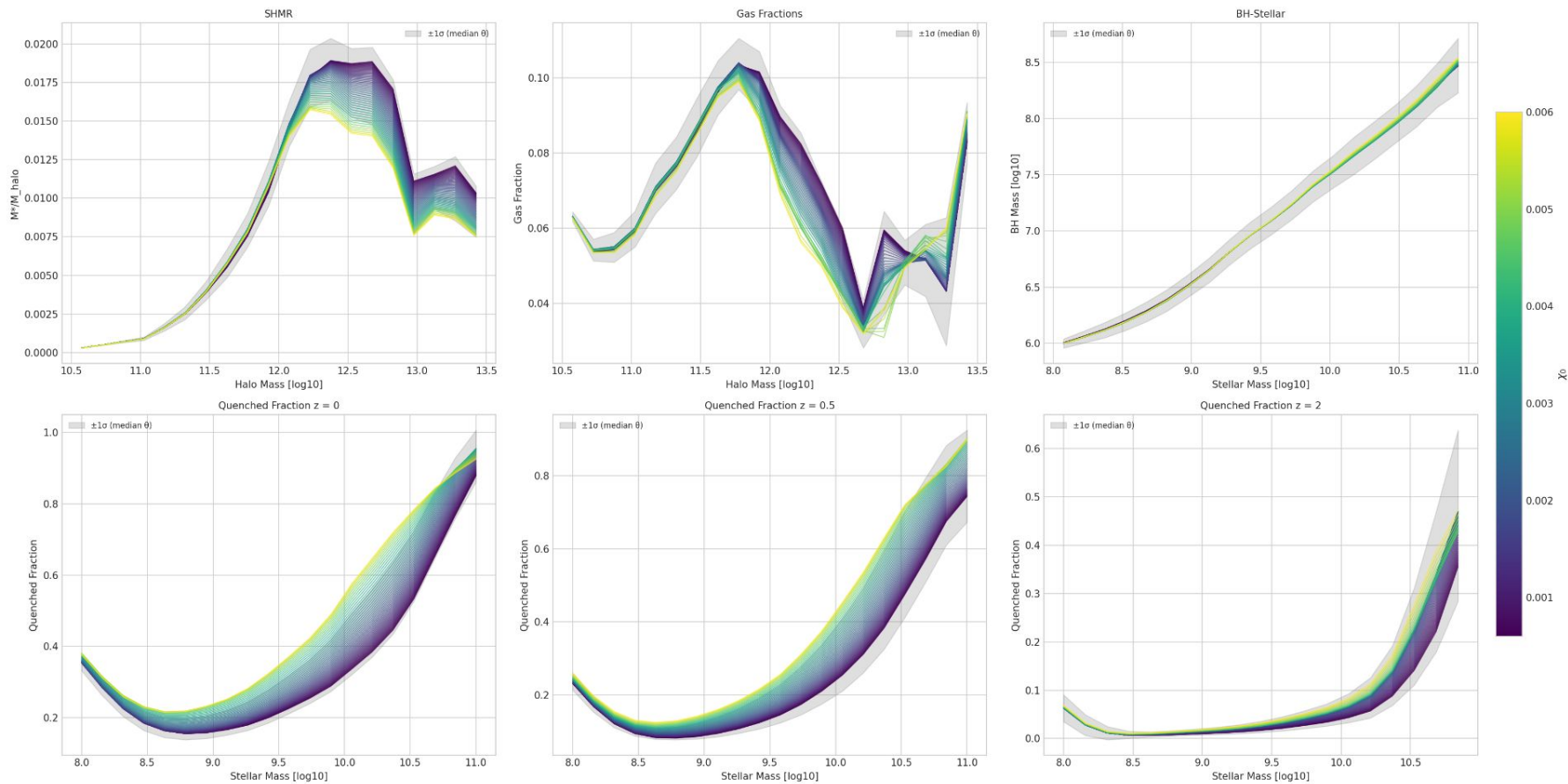
# Sensitivity studies (TLA rate)

1P Variation:  $\epsilon_T$



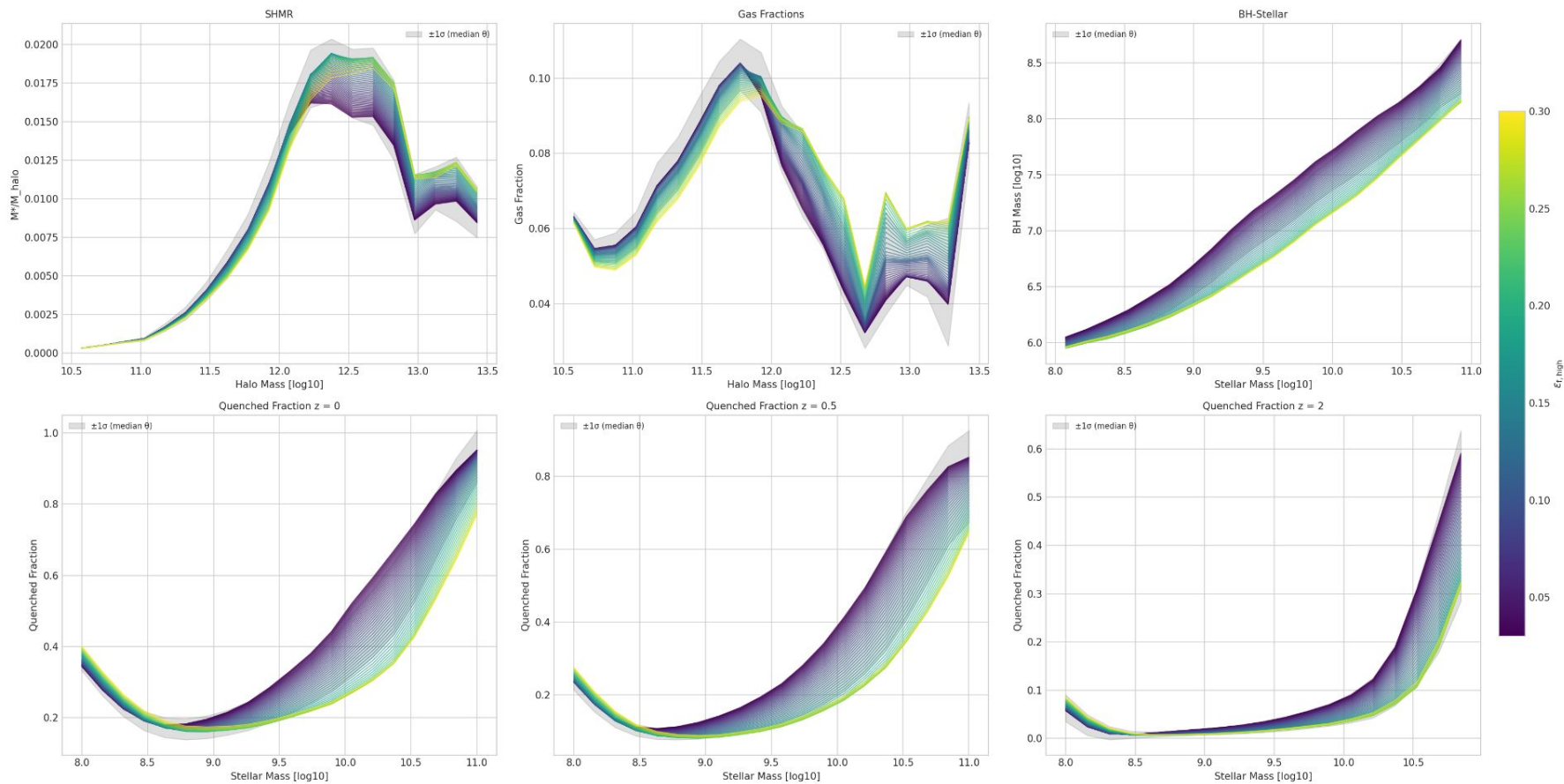
# Sensitivity studies (Quasarthreshold)

1P Variation:  $\chi_0$



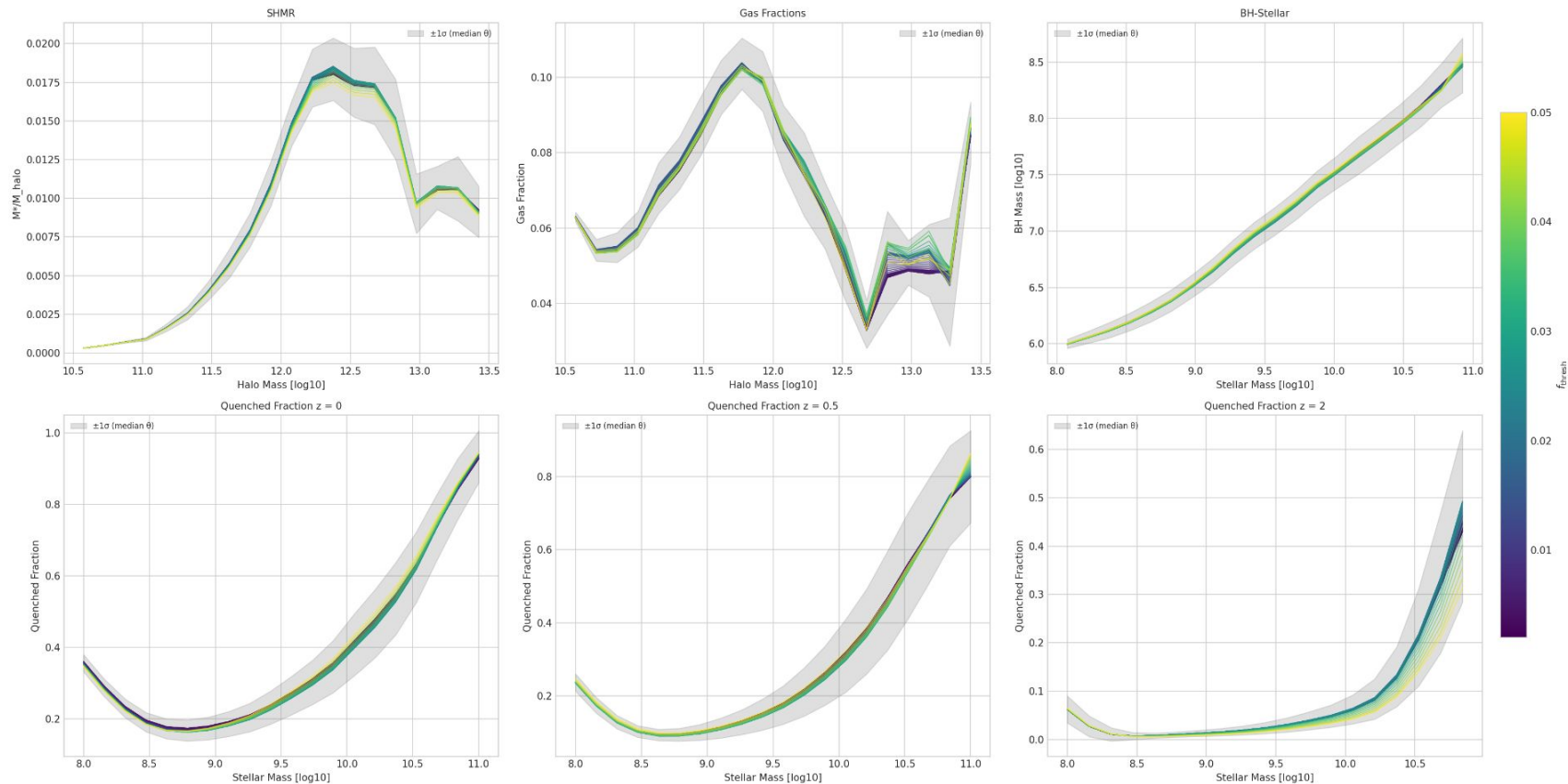
# Sensitivity studies (Feedback factor)

1P Variation:  $\epsilon_{f, \text{high}}$

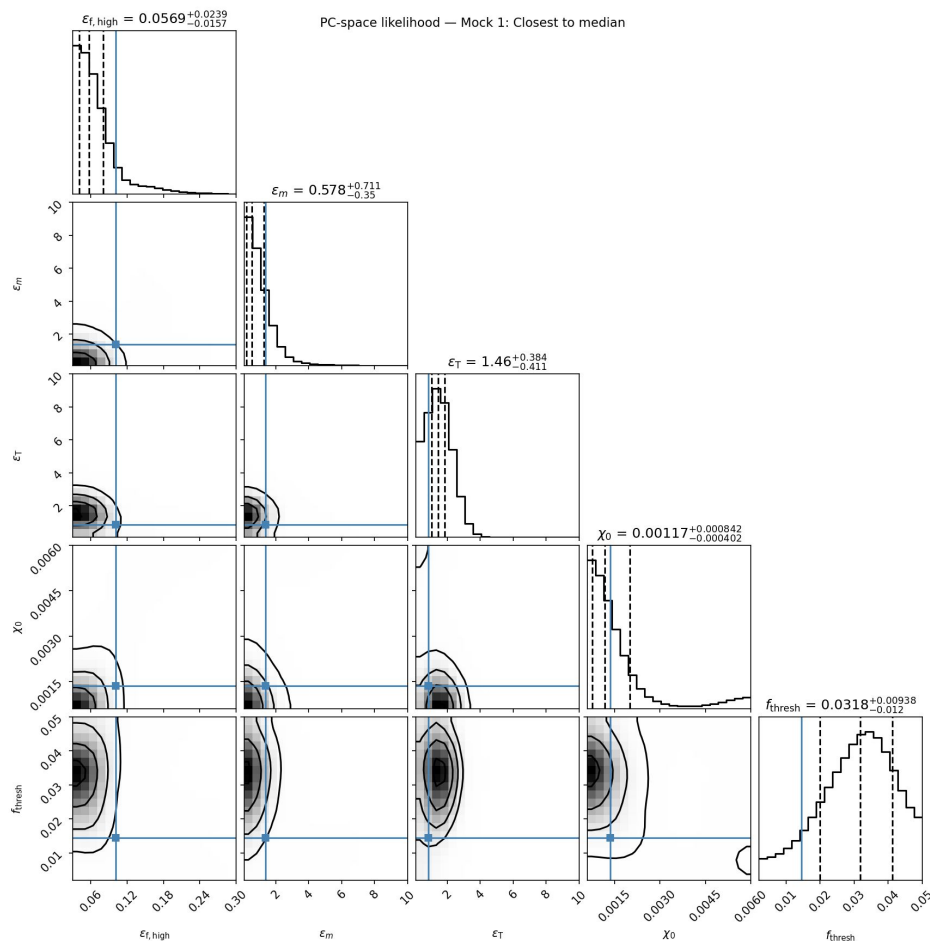


# Sensitivity studies (Radio efficiency)

1P Variation:  $f_{\text{thresh}}$

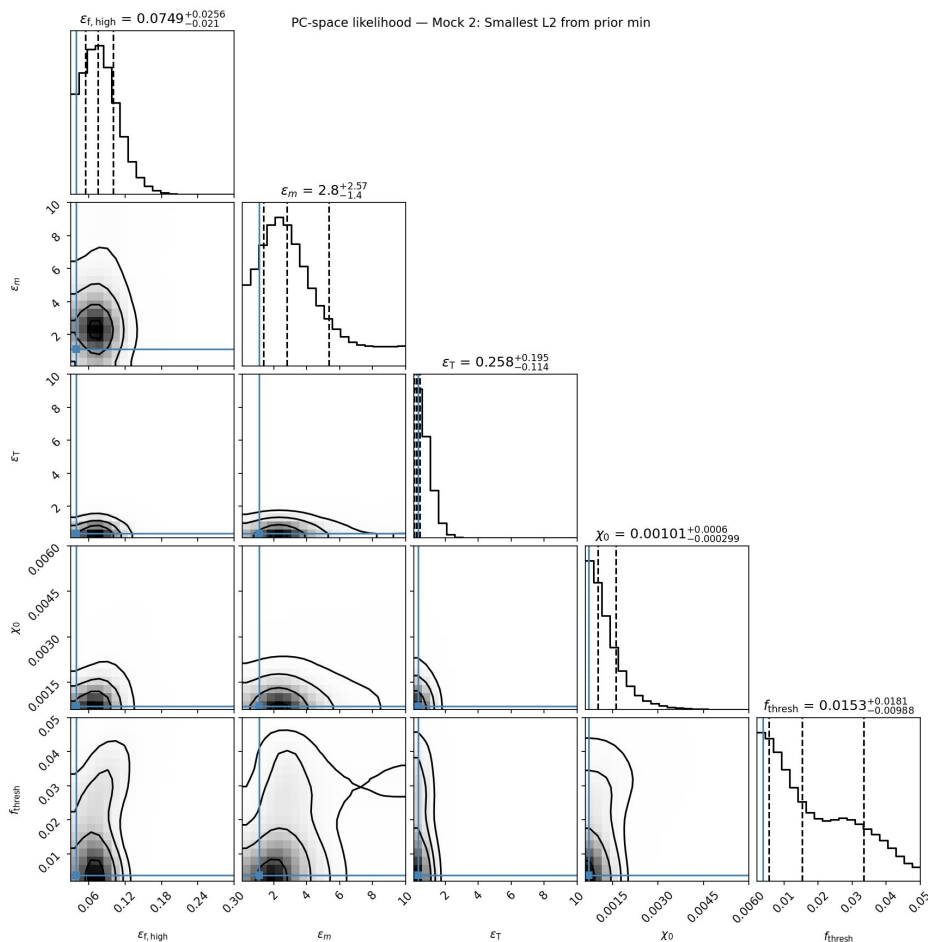


# MCMC results: median curve



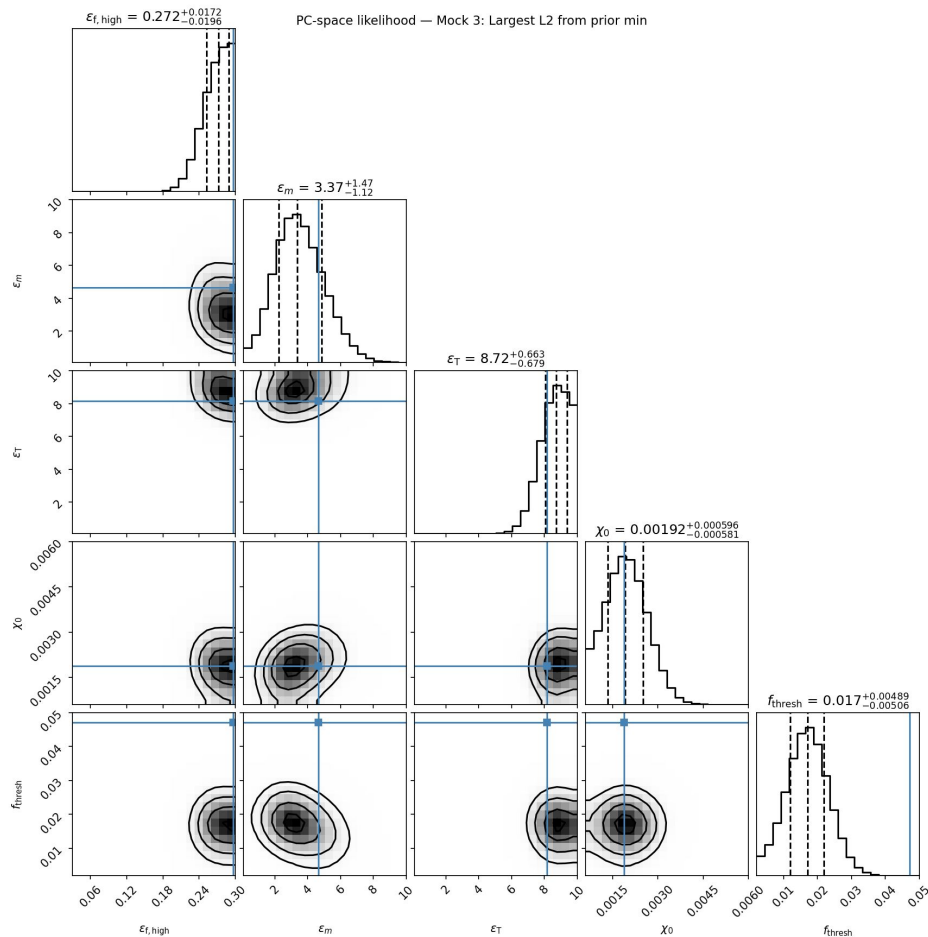
- Constraining parameters are generally well recovered
- Good constraints on the feedback factor and the accretion rates
- Slightly overconfident in the Bondi accretion rate, as this parameter hardly affects relations
- The posterior for the minimum density factor for radio feedback is very broad, as expected

# MCMC results: smallest L2 from min



- All three crucial parameters are well recovered
- Broad posteriors for the two parameters that hardly affect the scaling relations, as expected
- Here, inference pipeline works remarkably well

# MCMC results: largest L2 from min



- All three crucial parameters are well recovered
- Broad posterior for the Bondi accretion rate
- Clear mismatch between actual and inferred parameter for the minimum density factor for radio feedback
  - There is an issue with the inference pipeline for what is essentially a nuisance parameter

# Conclusions and future work

- We have built a simple inference pipeline that reliably recovers 3 out of 5 model parameters, based on 150 training simulations
- Parameters that do not affect the scaling relations used to train our emulator are not recovered
- To calibrate the full LtU model, we will use a DREAMS suite of zoom simulations
- Our results here raise some key questions:
  - How do we reliably identify inconsequential parameters?
  - How do we pick values for such parameters?
  - How robust do we think our inference methods are and can we address the remaining mismatches?