

# More than just a line: Shining light on the flattening radial metallicity gradient of a NIHAO-UHD Milky-Way analogue simulation

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## ABSTRACT

Radial metallicity gradients in galaxies are crucial for understanding the processes of galaxy formation and its dynamical and chemical evolution. We use young stars and gas of a high-resolution simulation of a NIHAO-UHD Milky Way analogue galaxy to analyse four properties of the radial metallicity gradient, namely its linearity and scatter as well as its coherence with radial coverage, azimuth, and age. While a global linear fit with slope traces the gradient well, we find it to be too flat in the inner galaxy and too steep in the outer galaxy. An initially stepper but then flattening second order quadratic function provides a superior fit down to the smallest radial steps, where less regular effects take over. We identify these as **SB: spiral arms or warp?** when tracing them across the Galactic position. The simulation further exhibits a steadily increasing spread in [Fe/H] even among the young stars from 0.01 dex in the inner galaxy to 0.14 dex at a radius 20 kpc coinciding with decreasing stellar density and increased flaring of stars towards higher heights with little correlation across the small age range of 1 Gyr. We conclude that future Galactic and extragalactic observational studies should test our theoretical prediction of a quadratic gradient function with their data to confirm or reject previous observational claims of a broken linear radial metallicity gradient proposed for the Milky Way and other spiral galaxies, as either function implies different dynamical and chemical galaxy formation and evolution scenarios.

**Key words:** cosmology: observations – Galaxy: structure – Galaxy: abundances – galaxies: structure – galaxies: abundances

## 1 INTRODUCTION

Understanding the radial metallicity gradient, defined as the change in heavy element abundance with galactocentric radius, in galaxies provides critical insights into their formation and evolutionary processes. In this study, we analyze a simulated Milky Way analogue, aiming to uncover nuanced details of its radial metallicity gradient that might be obscured by observational limitations in our own galaxy. By leveraging high-resolution simulations, we hope to bridge the gap between the detailed yet spatially confined observations of the Milky Way and the broader, but often less detailed, extragalactic observations. Our aim is to provide unique perspectives from simulations that can enhance the interpretative power of both Milky Way and extragalactic datasets, fostering a deeper understanding of galactic chemical evolution across different environments and epochs.

The decrease in metallicity with increasing distance from the Galactic center is well-established both theoretically (Larson 1976; Tinsley 1980; Chiosi 1980) and observationally in the Milky Way (Searle 1971; Janes 1979; Twarog et al. 1997). The Milky Way, being the only galaxy where we can resolve millions of stars, provides a unique opportunity to study these gradients in detail. However, the specific shape and characteristics of this gradient remain somewhat unclear. Over two decades ago, Chiappini (2002) highlighted the

contradictions in the observed shape, magnitude, and evolution of abundance gradients. Recent advancements, such as the *Gaia* mission (Gaia Collaboration et al. 2016), have significantly expanded our observational data, allowing for more detailed studies of these gradients.

For instance, Hogg et al. (2019) created an extensive metallicity map of the Milky Way using APOGEE and *Gaia* data, while Poggio et al. (2022) mapped young stars and found a metallicity excess around spiral arms (see also Zari et al. 2018, 2021; Poggio et al. 2021; Hackshaw et al. 2024). Similarly, Imig et al. (2023, among others) traced gradients across different stellar populations and ages, emphasizing the importance of considering radial migration effects (Binney & Tremaine 2008; Frankel et al. 2018, 2020). Historically, radial metallicity gradients have been measured using various stellar populations and gas tracers. Early evidence by Janes (1979) suggested a gradient on the order of

$$\frac{d[\text{Fe}/\text{H}]}{dR} = -0.05 \pm 0.01 \text{ dex kpc}^{-1} \quad (1)$$

for the Milky Way which aligns with more recent measurements (Anders et al. 2017; Hayden et al. 2015). Estimated gradients also seem to be broadly consistent across different stellar tracers, such as young open clusters (e.g. Yong et al. 2012; Cunha et al. 2016; Magrini et al. 2017; Casamiquela et al. 2019; Donor et al. 2020; Spina et al. 2021; Myers et al. 2022), young hot (OB-type) stars (Zari et al. 2018,

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2021; Poggio et al. 2021, 2022), field stars close to the Galactic plane (e.g. Bergemann et al. 2014) or Cepheids (Andrievsky et al. 2002a,b; Lemasle et al. 2007, 2013). However, these gradients are accompanied by significant spread in [Fe/H] of 0.1 – 0.15 dex, as noted by Twarog (1980), which may imply a fine structure of the metallicity gradient (see Genovali et al. 2014).

Despite extensive observational efforts, several challenges persist for studies in the Milky Way:

- The completeness (or patchiness) of observed datasets remains a fundamental issue (Bergemann et al. 2014).
- How robustly can we use the incomplete data to predict substructure in the gradients, such as the need to actually fit two linear gradients with a break radius at corotation radius (Bresolin et al. 2012, and references therein) or further out (Yong et al. 2012; Donor et al. 2020) - or even more complicated functions (see e.g. Chiappini et al. 2001; Kubryk et al. 2015)?
- Different samples yield varying gradients, potentially due to biases in data or the inclusion of older stars (e.g. Boeche et al. 2013; Allende Prieto et al. 2006; Katz et al. 2011; Hayden et al. 2014; Anders et al. 2014; Vickers et al. 2021; Willett et al. 2023).
- The impact of spiral arm structures (Poggio et al. 2021), the Galactic warp (Lemasle et al. 2022) or bar-driven mixing (Di Matteo et al. 2013) on metallicity gradients is not fully understood.
- Methodologies for fitting linear models to scattered data need re-evaluation (Metha et al. 2021).

Understanding these gradients in the Milky Way is also crucial for extragalactic studies, where spatial resolution limits observations. New instruments like the MUSE integral field spectrograph have enabled a plethora of extragalactic studies to contrast the Milky Way to and techniques like the spectroscopy of HII regions and planetary nebulae have helped to infer gas metallicity distributions in external galaxies (Shaver et al. 1983; Vilchez & Esteban 1996; Rolleston et al. 2000; Bresolin et al. 2012). Recent examples include Sánchez et al. (2014) with CALIFA galaxy observations as well as Mun et al. (2024) and Chen et al. (2024) who use MAGPI galaxy observations to probe for example the effects of spiral arms. Notable is also the scatter that Chen et al. (2023) found for the gas metallicity across galactic radius with TYPHOON observations (see their Figs. 4-6). SB: Grasha et al. (2022) suggested that the flattening of the gas metallicity in their TYPHOON galaxies could be caused by changes in the ratio of supernovae II and AGB reflected by a changing ratio of nitrogen to oxygen abundance N/O which also flattens towards a lower plateau below metallicities of  $A(O) \sim 8.0$  (Nicholls et al. 2017).

SB: Look into I-Ting's work: Ho et al. (2015, 2017, 2018, 2019). See e.g. metallicity differences in spiral arms vs. downstream and upstream in Figs. 6, 8, and 11 of Ho et al. (2017) SB: Look into Sharon's work: Meidt et al. (2008, 2009)

From a modelling perspective, galactic chemical evolution models can both test understanding of radial metallicity gradients and make predictions beyond the limited volumes and tracers tested by Milky Way and extragalactic studies. Such galactic chemical evolution models include Chiappini et al. (2001); Matteucci (2001); Minchev et al. (2014); Kubryk et al. (2015); Stanghellini et al. (2015); Rybizki et al. (2017). Sharda et al. (2021) even presented a model for gas phase metallicity gradients in galaxies and their evolution from first principles. SB: see also simple model of chemical enrichment in Ho et al. (2017).

New suites of large-scale simulations now allow to gain insights into radial metallicity gradients across a range of simulated galaxies - including Milky Way analogues. While Pilkington et al. (2012)

found that such gradients are established by inside-out formation in RaDES simulations, whereas Tissera et al. (2019) used EAGLE simulations to study gradients across different galaxy characteristics, such as mergers and stellar mass. Bellardini et al. (2021) and Bellardini et al. (2022); Graf et al. (2024) compared the radial metallicity gradient and its azimuthal scatter for different simulations of the FIRE suite for gas and stars, respectively. Focusing on the change of the radial metallicity gradient over time, Grand et al. (2015) assessed the scatter of radial metallicity distribution for different stellar ages. Grand et al. (2016) looked at stars with ages below 3 Gyr in AURIGA and linked deviations from a global gradient towards slightly lower and higher metallicities with the leading and trailing edges of spiral arms, respectively. Grand et al. (2016) even found significant changes to the metallicity induced during 120 Myr, or roughly one galactic rotation. Both Ma et al. (2017, see their Fig. 6) with FIRE and Agertz et al. (see their Fig. 9 2021) with VINTERGATAN further assessed the evolution of the radial metallicity gradient for gas and stars. Buck et al. (2023) studied how the gas metallicity gradient changes over time in the NIHAO-UHD simulations and found periods of time where the gradient surprisingly steepens. This finding resembles closely the findings of Lu et al. (2022) and Ratcliffe et al. (2023) for the Milky Way. This steepening phase could be connected and explained by major accretion events in the simulations. Khoperskov et al. (2023) again focused on the gas and found the scatter of the gas metallicity of  $\approx 0.04 - 0.06$  dex at a given galactocentric distance. SB: Hemler et al. (2021) evolution of the gas-phase radial metallicity gradient with redshift and galaxy mass in IllustrisTNG and Garcia et al. (2023) studied the

This study here aims to advance these theoretical insight by addressing four critical scientific questions using a high-resolution simulation of a Milky Way analogue galaxy focussing on the inner  $R_{\text{Gal}} \leq 20$  kpc:

- (i) *Linearity of the gradient:* Assess the extent to which the radial metallicity gradient of young stars is linear.
- (ii) *Scatter in the gradient:* Quantify the expected scatter in the radial metallicity gradient of young stars.
- (iii) *Coherence of the gradient with position:* Investigate the gradient's variation with radial coverage and azimuth.
- (iv) *Coherence of the gradient with age:* Determine the reliability of stars as tracers of the gas disk over different ages.

The paper is structured as follows: Section 2 describes the data of our Milky Way analogue NIHAO-UHD simulation. Section 3 analyses the linearity of the radial metallicity gradient of the simulation, the first of our four objectives. Section 4 then analyses both the scatter and local deviations from the gradient as well as the coherence of the gradient with vertical and azimuthal position as well as age (the remaining three objectives). Section 5 discusses them individually. We note that in this research we are mainly interested in the specific shape of the radial metallicity gradient for radii relevant to Galactic observations. However, we also discuss our results in the context of extragalactic results that probe beyond the inner 20 kpc of a galaxy. Section 6 bundles our results into an overarching conclusion.

## 2 DATA: A NIHAO-UHD MILKY WAY ANALOGUE SIMULATION

For this project, we use a cosmological zoom-in simulation of a Milky Way analogue (g8.26e11) from the *Numerical Investigation of a Hundred Astronomical Objects* (NIHAO, Wang et al. 2015) project. For this project, we limit the simulation data to the main halo by applying PYNBODY’s implementation of the Amiga Halo Finder (Knollmann & Knebe 2009) and then reposition and rotate this main halo to be face-on based on the angular momentum with PYNBODY’s ANALYSIS.ANGMOM.FACEON module (Pontzen et al. 2013). We then further transform the resulting galactocentric Cartesian coordinate ( $X, Y, Z$ ) and velocities ( $V_X, V_Y, V_Z$ ) to Cylindrical ones as done in a previous study of this main halo by Buder et al. (2024).

The model galaxy has a virial radius of  $R_{\text{m}athrm{vir}} = R_{200} = 206$  kpc and a total mass (gas, stars and dark matter) inside  $R_{\text{vir}}$  of  $9.1 \cdot 10^{11} M_{\odot}$ . At redshift zero, it contains  $8.2 \cdot 10^{11} M_{\odot}$  dark matter,  $6.4 \cdot 10^{10} M_{\odot}$  gas mass and  $2.3 \cdot 10^{10} M_{\odot}$  stellar mass with a stellar mass resolution of around  $7.5 \cdot 10^3 M_{\odot}$ . When using a fifth of the virial radius as reference to calculate total luminosity<sup>1</sup> and mass, we estimate a half-light, that is, effective radius of  $R_e = 3.79$  kpc and a half-stellar-mass radius of 2.97 kpc. This model galaxy was calculated as part of the NIHAO-UHD project (Buck et al. 2020) and has previously been used in various works studying Milky Way satellites (Buck et al. 2019), Milky Way’s dark halo spin (Obreja et al. 2022) or inferring birth properties of stars with abundance clustering (Ratcliffe et al. 2022).

Simulations were carried out with the smoothed particle hydrodynamics code Gasoline2 (Wadsley et al. 2017) using cosmological parameters from Planck Collaboration et al. (2014) with initial conditions and energetic feedback descriptions from the NIHAO project (Wang et al. 2015). Zoom-in simulations were then performed as described in detail by Buck et al. (2021) with star formation following Stinson et al. (2006) and energetic feedback following Stinson et al. (2013). We note that this is a slightly different rerun of the same simulation than the one studied by Buder et al. (2024). Here we are dealing with a higher resolution version and updated chemical yields.

Because computational resources still limit the mass resolution of simulations, we are relying on tracer particles that represent simple stellar populations (SSPs) with the same age, overall metallicity and discrete initial mass function (IMF). Buck et al. (2021) have implemented the flexible chemical evolution code CHEMPY (Rybicki et al. 2017) to calculate the chemical yields for the SSPs. In particular, we use the alternative (alt) setup of CHEMPY that assumes a Chabrier (2003) IMF with high-mass slope of  $\alpha_{\text{IMF}} = -2.3$  over a mass range of  $0.1 - 100 M_{\odot}$  for SSPs across a metallicity range of  $Z/Z_{\odot} \in [10^{-5}, 2]$ . The code calculates the contribution from asymptotic giant branch (AGB) stars, CCSN across a mass range of  $8 - 40 M_{\odot}$ , and SNIa with a an exponential function with exponent  $-1.12$ , a delay time of 40 Myr, and a normalization of the SNI rate of -2.9. For each of these nucleosynthetic channels, yields from the following studies are used: Chieffi & Limongi (2004) for CCSN, Seitenzahl et al. (2013) for SNIa, and Karakas & Lugaro (2016) for AGB stars (new\_fit model in Buck et al. 2021). Contrary to a previous study by Buder et al. (2024), we take the elemental abundances at face value and do not apply any shifts.

To achieve a roughly similar selection as the observational data of the Milky Way (Genovali et al. 2014) and other galaxies (e.g. Chen et al. 2023), we restrict the simulation data to a galactocentric radius of  $R_{\text{Gal}} \leq 20$  kpc, as shown in Fig. 1. Similar to the Milky

Way (Poggio et al. 2018; Lemasse et al. 2022), we note a warp of the stellar and gaseous disk (see Figs. 1b and 1e, respectively).

To avoid too strong effects of radial migration (Binney & Tremaine 2008; Frankel et al. 2018; Grand et al. 2016) while maintaining a sufficiently large sample size we further enforce stars to be younger than 0.5 Gyr, corresponding to roughly the time of four galactic rotations. This selection de-facto limits the vertical range of 99% of stars to  $|z| = 1.4$  kpc. The strong influence of this age cut on the vertical distribution of stars in the Milky Way analogue can best be appreciated from the difference of vertical density distributions of stars in Figs. 1b and 1c. We are applying these cuts for all following analyses of the radial metallicity gradient in Section 3.

## 3 THE LINEARITY OF THE RADIAL METALLICITY GRADIENT IN NIHAO-UHD

In this section, we analyse the functional shape of the radial metallicity gradient. To get a first impression of possible shapes, we show the logarithmic density distribution of star particle iron abundances [Fe/H] across different galactocentric radii  $R_{\text{Gal}}$  in Fig. 2a. This distribution strongly suggests that the gradient is predominantly linear, similar to findings for the Milky Way. More complex shapes, such as piecewise linear ones have been suggested based on incomplete and limited data in the literature. We are thus also analysing these shapes with the complete and better sampled data points of the NIHAO-UHD simulation. We firstly test different global fits in Section 3.1, before testing the influence of binning and coverage in Sections 3.2 and 3.3, respectively.

### 3.1 Global gradient fits

We fit three different functional forms to the global data of Fig. 2, that is, a linear function, a piecewise linear with a break radius, and a quadratic function. The coefficients of the functions as fitted with the SCIPY.OPTIMIZE function CURVE\_FIT (Virtanen et al. 2020) are listed in Table 1. We have confirmed the robustness of our fits by applying other fitting routines as outlined in Appendix A.

Having fitted three different models, will use a combination of parameters to judge which model fits better. Firstly, we are computing the residual sum of squares (RSS) as

$$\text{RSS} = \sum_i ([\text{Fe/H}] - g(R_{\text{Gal}}))^2. \quad (2)$$

We see from Table 1 that this value is smallest (although only by a small margin) for the quadratic function. When assuming  $\sigma^2 = \text{RSS}/n$ , we can then define a logarithmic likelihood

$$\ln L = -\frac{n}{2} \ln(2\pi) - \frac{n}{2} \ln \frac{\text{RSS}}{n} - \frac{n}{2} \quad (3)$$

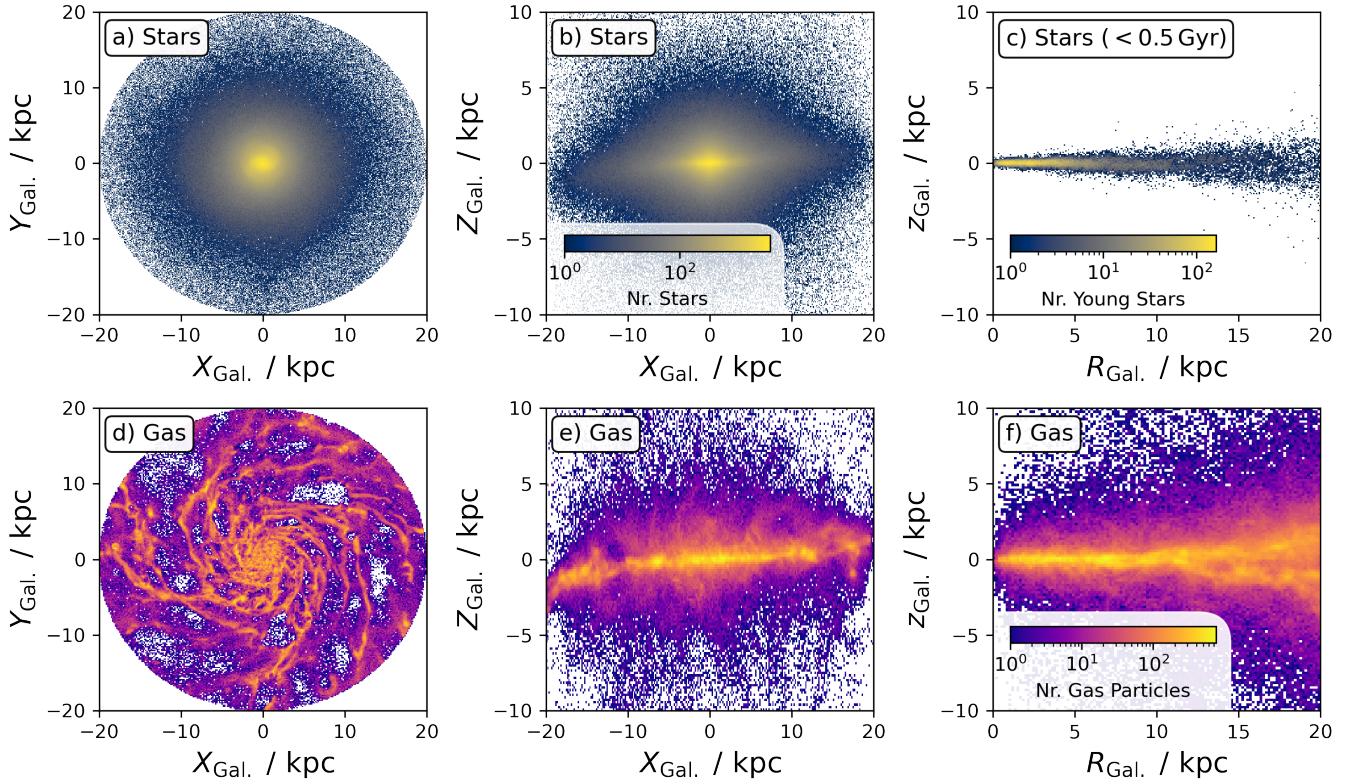
for our  $n$  data points. For  $k$  free parameters, we can then calculate the Akaike Information Criterion (AIC) and the Bayesian Information Criterion (BIC) as

$$\text{AIC} = 2k - 2 \ln L \quad \text{BIC} = k \ln n - 2 \ln L. \quad (4)$$

Again, the quadratic function performs slightly better than the linear or piecewise linear functions (see Table 1).

We show the fit residuals in Fig. 2b as density distribution as well as in Fig. 2c as percentile distributions in radial bins of  $\Delta R_{\text{Gal}} = 0.5$  kpc. While the density distribution shows a lot of substructure, that we are investigating later in Section 4, we note an increase of the median residuals of the linear fit in Fig. 2c towards inner and outer radii,

<sup>1</sup> Using PB.ANALYSIS.LUMINOSITY.HALO\_LUM (Pontzen et al. 2013).



**Figure 1.** Logarithmic spatial density distribution of stars (upper panels) and gas (lower panels) within  $R < 20 \text{ kpc} \sim 5 R_e$  of the NIHAO-UHD Milky Way analogue g8.26e11 in galactocentric cartesian and cylindrical coordinates. Panel c) shows the influence of selecting only young stars with ages below 0.5 Gyr.

**Table 1.** Fit Evaluation of linear, quadratic, and piecewise linear fits. Extra parameters are quadratic term and break radius for the quadratic and piecewise fit. RSS stands for Residual Sum of Squares (Eq. 2). AIC stands for Akaike Information Criterion and BIC stands for Bayesian Information Criterion (see Eq. 4).

Function	Intercept ( $a_0 \pm \sigma_{a_0}$ )	Slope ( $a_1 \pm \sigma_{a_1}$ )	Extra Parameter	RSS	AIC	BIC
Linear	$0.46849 \pm 0.00159$	$-0.04169 \pm 0.00035$	–	87.6	-118050	-118030
Quadratic	$0.48031 \pm 0.00055$	$-0.04864 \pm 0.00018$	$0.00045 \pm 0.00001$	82.7	-120150	-120120
Piecewise	$-0.04477 \pm 0.00010$	$0.47473 \pm 0.00047$	–	82.9	-120090	-120050
	–	$-0.03562 \pm 0.00014$	9.3			
Linear (bins)	$0.46849 \pm 0.00567$	$-0.04169 \pm 0.00127$	–	0.0147	-200	-200
Quadratic (bins)	$0.47327 \pm 0.00674$	$-0.04611 \pm 0.00359$	$0.00031 \pm 0.00024$	0.0033	-260	-250
Piecewise (bins)	$-0.04398 \pm 0.00207$	$0.47190 \pm 0.00617$	–	0.0026	-260	-260
	–	$-0.03547 \pm 0.00461$	10.0			

especially for  $R_{\text{Gal.}} > 17 \text{ kpc}$ . A quadratic fit (see orange lines in Fig. 2) results in a slightly steeper linear component of the gradient (from  $-0.0411$  to  $-0.0497 \text{ dex kpc}^{-1}$ ), which is counteracted by the quadratic flattening term of  $+0.0005 \text{ dex kpc}^{-2}$ . The latter leads to an effective flattening of  $-0.172 + 0.200 = 0.028 \text{ dex}$  (linear vs. quadratic terms) at  $R_{\text{Gal.}} = 20 \text{ dex}$ . While seemingly only a nuisance correction across the large extend of [Fe/H] and  $R_{\text{Gal.}}$ , this quadratic function is evidently superior to the linear fit. This can be best appreciated in Fig. 2c, where the orange line is tracing the median residuals from the linear function much better across all the radii. This suggests a non-linear such as a piecewise linear or quadratic form to the gradient.

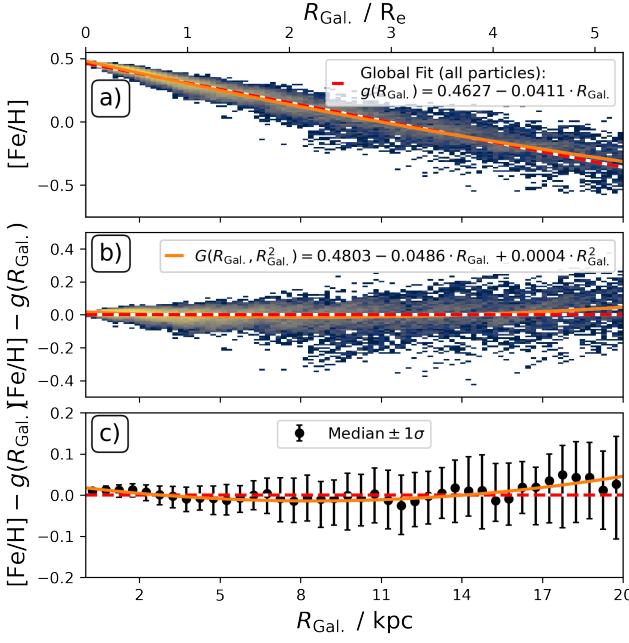
Discerning between the latter functional forms proves difficult. All fitting quantitative fitting performance indicators of RSS, AIC, and BIC have very similar values between the two forms and when zooming into the residuals of the different shapes in Fig. 3, we see no visual advantage of either form.

**Take-Away:** Piecewise linear and quadratic functions perform both better than a linear fit to the radial metallicity relation. However, we see no significant preference between piecewise and quadratic functions based on our assessments.

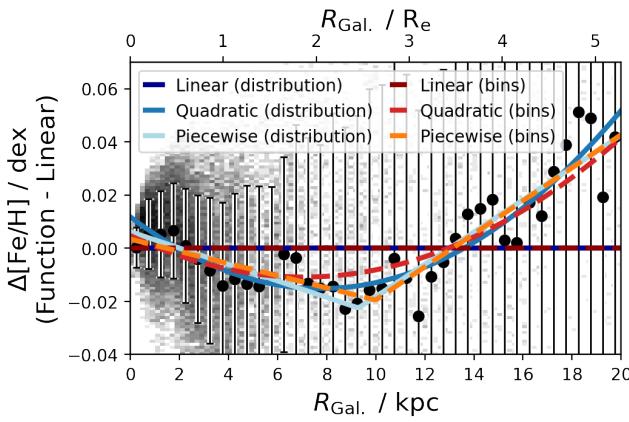
### 3.2 The influence of binning

In this section, we test the influence of fitting a function to all points of the distribution or to binned data in steps of  $\Delta R_{\text{Gal.}} = 0.5 \text{ kpc}$ , using mean values as data points and standard deviations<sup>2</sup> as uncertainty (see also Hemler et al. 2021, who fitted functions to radially binned IllustrisTNG data). Our results are plotted in Fig. 3. Given that more than half of the young star particles of the galaxy are within  $R_{\text{Gal.}} <$

<sup>2</sup> We note that this  $\sigma$  is not equivalent to an observational uncertainty and can thus not be directly applied onto observational analyses.



**Figure 2.** Global fits and deviation to the radial metallicity gradient  $R - [\text{Fe}/\text{H}]$ . Functional forms of the linear (red) and quadratic (orange) lines are shown in the legend. Panel a) shows the underlying data of all data points as logarithmic density and the global fit to them as red dashed line. Panel b) shows the deviation of data from a linear gradient as logarithmic density plot, whereas panel c) shows the 16th and 84th percentile around the median deviation as error bars in  $\Delta R_{\text{Gal.}} = 0.5 \text{ kpc}$  bins.



**Figure 3.** Same as Fig. 2c, but overplotting the different functions (linear, quadratic, and piecewise linear) estimated from the full distribution (solid lines) or from means and standard deviations (error bars) in  $\Delta R_{\text{Gal.}} = 0.5 \text{ kpc}$  bins (dashed lines).

4 kpc, this binning - although counteracted by the smaller spread of [Fe/H] in the inner galaxy - is weighing the distribution of inner galaxy significantly less than when weighing all particles equally (20 vs. 34 000 data points). The fitted parameters to binned data exhibit a larger uncertainty due to our use of the spread of [Fe/H] per bin as absolute uncertainty  $\sigma$ , but the fitted parameters are agreeing well and truly within the fitting uncertainties.

**Take-Away:** While the specific slopes differ when fitting all points or binned data, they agree within the still rather small fitting uncertainties.

### 3.3 The influence of radial coverage on linear fits

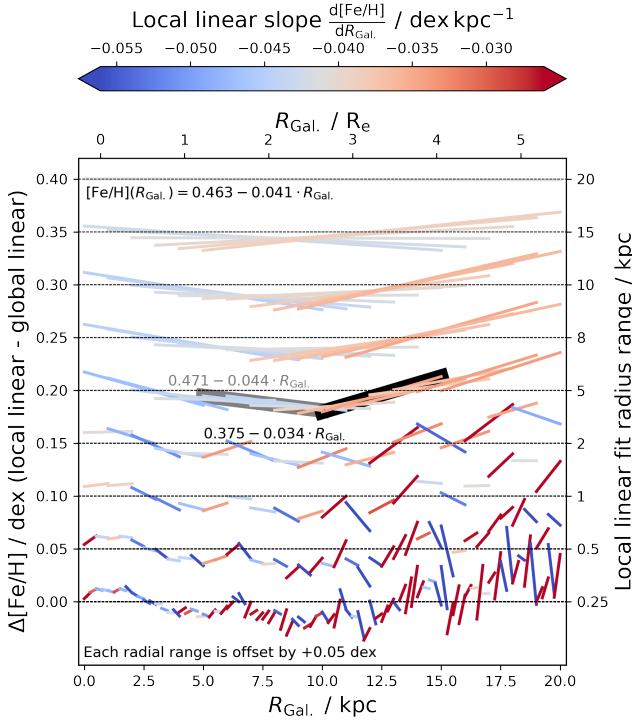
While we have gained a useful insight into the global function, observational data will rarely cover the full extend of the stellar disk. Milky Way studies have previously been limited to the range of around 5 – 15 kpc. There are often similar limitations and even gaps in extragalactic data. Using smaller ranges, observational studies have found hints of piece-wise linear gradients with a break radius in them based on a limited radial coverage (e.g. Andrievsky et al. 2002a; Yong et al. 2012; Boeche et al. 2013; Hayden et al. 2014; Anders et al. 2017; Donor et al. 2020; Chen et al. 2023). These results are intriguing, since a quadratic function can to first order be approximated by two linear functions with a break radius. We therefore want to use this simulation to test if the radial coverage may indeed delude us into identifying broken linear gradients.

In Fig. 4, we test how smaller coverage and piecewise linear fits could mimic a complex global gradient. We test fits across piecewise linear radial ranges of 0.25, 0.5, 1, 2, 5, 8, 10, 15, and 20 kpc. We show their difference with respect to a global linear fit in Fig. 4, with color-coding indicating the local gradient slope. In this figure, a horizontal dashed line indicates the same slope as the global fit, whereas the offset of a line from said horizontal dashed line indicates the local deviation from the global gradient intercept. Differences in line slopes are visualising the difference in gradient slopes between the global and local fits. We see that all ranges are suggesting more or less significant deviations from a global linear fit. The innermost fit is suggesting a significantly different gradient than the outermost fit. We also note increasing slope differences towards the smallest scales, hinting at local deviations from a global pattern. We follow these up in Section 4, but for now focus on the larger-scale trends.

When directly comparing an inner and outer radius fit, such as between  $R_{\text{Gal.}} = 5 - 10 \text{ kpc}$  (thick grey line in Fig. 4) and  $R_{\text{Gal.}} = 10 - 15 \text{ kpc}$  (thick black line in Fig. 4), we note a significant change, similar to previous estimates of the Milky Way (e.g. Yong et al. 2012; Lemasle et al. 2008). In our case the gradient estimate changes from  $[\text{Fe}/\text{H}](R_{\text{Gal.}}) = 0.471 - 0.044 \cdot R_{\text{Gal.}}$  to  $[\text{Fe}/\text{H}](R_{\text{Gal.}}) = 0.375 - 0.034 \cdot R_{\text{Gal.}}$ . When looking at linear gradient fits across radial coverage of  $\Delta R_{\text{Gal.}} = 5 - 15 \text{ kpc}$  in Fig. 4, the gradient was steeper (bluer color) for smaller radii and flatter (redder) for larger radii. Indeed, a piecewise linear function can well mimic a complex global gradient.

We note that in the simulated data, we see local deviations becoming traceable below  $\Delta R_{\text{Gal.}} \leq 2 \text{ kpc} \sim 0.5 R_e$  (bottom part of Fig. 4). This might indicate the necessary spatial resolution to see local effects, such as spiral arms, for extragalactic studies. We pursue this observation in the following Section 4.

**Take-Away:** We find that a piecewise linear function can well mimic a quadratic function across the scales used in Milky Way and extragalactic studies. Local deviations become traceable below a spatial resolution of  $\Delta R_{\text{Gal.}} \leq 2 \text{ kpc} \sim 0.5 R_e$ .



**Figure 4.** Impact of different coverage in galactocentric radius when fitting a linear radial metallicity gradient to young stars. Each horizontal segment is using a different running radial fitting range between 0.25 and 15 kpc as outlined on the right. For better contrast, the global linear fit is subtracted from the local gradient estimates and each line is colored by the gradient slope with a color scale centered around the global fit slope. Additionally, the slope of each line segment highlights the difference between global and local slope. This means, if the local fit exactly matches the global fit, we display it as a flat line (zero slope difference) on top of the gray dashed line (zero offset deviation) and give it a gray color signaling the same gradient value as the global fit.

#### 4 SCATTER AND LOCAL DEVIATIONS FROM THE GRADIENT

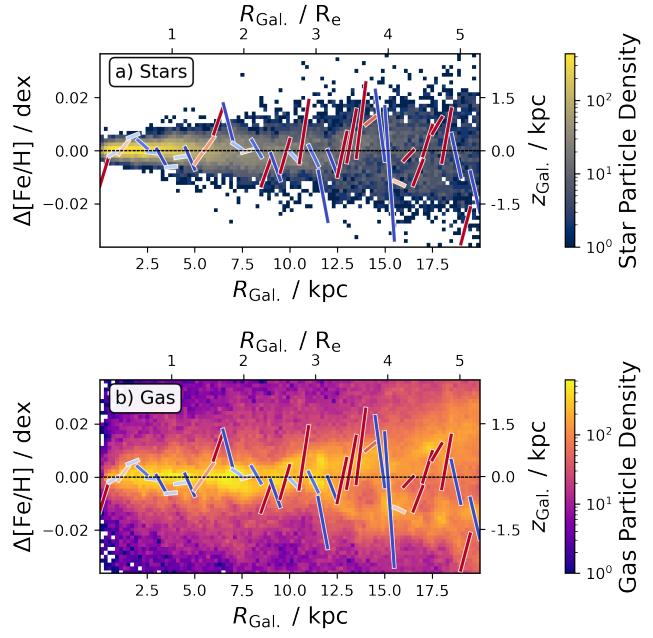
Now that we are sufficiently satisfied that our flattening gradient function reproduces the overall shape of the radial metallicity gradient, we are concerned with both the scatter and local slope deviations across the galactocentric radii in this section. In detail, we analyse the scatter (Section 4.2), vertical and azimuthal variations (Sections 4.2 and 4.3, respectively) as well as deviations across different ages (Section 4.4).

##### 4.1 Scatter

When investigating the change of scatter from the innermost radii to the outermost (see Fig. 2c), we see a steady increase in  $1 - \sigma$  spread. This spread increases from

$$\begin{aligned}\sigma[\text{Fe/H}] &= 0.01 \text{ dex at } R_{\text{Gal.}} = 0.25 \pm 0.25 \text{ kpc to} \\ \sigma[\text{Fe/H}] &= 0.06 \text{ dex at } R_{\text{Gal.}} = 8.25 \pm 0.25 \text{ kpc and finally reaches} \\ \sigma[\text{Fe/H}] &= 0.1 \text{ dex at } R_{\text{Gal.}} = 19.75 \pm 0.25 \text{ kpc, the largest } R_{\text{Gal.}}.\end{aligned}$$

When reminding ourselves of the observed significant spread in metallicities of young open clusters at the solar radius beyond observational uncertainty (e.g. Donor et al. 2020; Spina et al. 2021) and our selection of only young (<0.5 Gyr) stars from the simulation, a strong impact of this scatter by radial migration should be excluded.



**Figure 5.** Local gradient deviations similar to the second lowest row of Fig. 4 for radial gradients in 0.5 kpc steps (but compared to a global quadratic function) overlapped on top of the logarithmic density distribution in  $R - z$  for  $|z| < 3$  kpc of gas (panel a) and stars (panel b).

At this point, we can imagine that this chemical diversity might be caused by less well mixed gas or non-radial effects (such as vertical or azimuthal ones), which we investigate subsequently.

##### 4.2 Vertical deviations

In this section, we are now looking into deviations with respect to the vertical dimension, that is,  $R - z$ . In Fig. 5 we are showing the previously identified local gradient deviations (lines following the left axis label) on top of the vertical density distribution ( $R - z$ ) of gas (Fig. 5a) and young stars (Fig. 5b) between  $-3 < z < 3$  kpc. While the quickly decreasing amount of young stars (Fig. 5b) at outer radii does not show substructure in the density plots for reasonable bin sizes, we see more substructure for the gaseous component in Fig. 5a). In particular, we see rather minor deviations at small radii (where most stars and gas are close to the plane). At increasing radii, we notice an increase in both the vertical distribution of stars, as well as increasing local gradient deviations. In particular, we note a significant deviation of the slope around  $R_{\text{Gal.}} \sim 15$  kpc, where the gradient deviation line is steep and blue (indicating a much steeper gradient at this radius), and we notice a significant overdensity of gas around  $z \sim 1$  kpc. Overall, however, we do not see strong correlations in this particular plane. This could, however, be caused by a super-position effect of the up- and downturn at larger radii due to the galactic warp (see Fig. 1e). We note that the superposition in Fig. 5c could smear out local correlations of slope changes with gas overdensities, for example, by spiral arms.

While such an edge-on view of the galaxy may indeed be the only observable one for extra-galactic targets, for example of the GECKOS survey of edge-on galaxies (?), we have the luxury of being able to analyse the azimuthal direction of our simulated galaxy, too. This seems especially important, as we have seen a significant warping of the gaseous and stellar disk of our simulated galaxy (see Figs. 1b and

1e). While the warp of the stellar disk in Fig. 1b is not as clear, we confirm that both the gas disk and the youngest stars below 0.5 Gyr are tracing each other across the simulation in both galactocentric radius  $R_{\text{Gal.}}$  and height  $z_{\text{Gal.}}$  for different sectors in azimuthal direction.

**Take-Away:** We see no strong correlations of deviations in the vertical direction across the whole simulation. Such correlations could, however, be blurred by azimuthal effects, like the galactic warp, which needs to be disentangled in the azimuthal dimension.

#### 4.3 Azimuthal deviations

To analyse the deviations from a global gradient across different azimuthal viewing angles, we divide the galaxy into 8 sectors with  $\Delta\varpi_{\text{Gal.}} = 45 \text{ deg}$  (see Fig. 6a). This allows us to study the positions around the upturn and downturn of the galactic warp with the median azimuth of young star particles below and above the plane being  $\varphi_{\text{Gal.}} \sim 183 \text{ deg}$  and  $\varphi_{\text{Gal.}} \sim 4 \text{ deg}$ , respectively (see Fig. 1e).

- We find significant deviations in different sectors of the galaxy (Fig. 6).
- When looking at stars born within less than 0.5 Gyr we also find significant gaps in several regions of the galaxy, most notably around  $\varpi_{\text{Gal.}} = 135 - 180^\circ$ .
- We also find several isolated horizontal streaks, for example around  $R_{\text{Gal.}} \sim 14 \text{ kpc}$  at  $\varpi_{\text{Gal.}} = 180 - 225^\circ$ . In some cases, these extend to  $\Delta R_{\text{Gal.}} = 2 - 6 \text{ kpc}$ , see Groups 1, 2, and 3 in Fig. 7.
- At a given  $R_{\text{Gal.}}$ , these streaks are causing a significant spread in local [Fe/H] of up to  $\pm 0.2 \text{ dex}$  (see Fig. 7).

**Take-Away:** We find various deviations from the global trend in azimuthal direction, including gaps as well as isolated streaks of stars with similar [Fe/H] across  $\Delta R_{\text{Gal.}} = 2 - 6 \text{ kpc}$ . These can introduce local over- and under-enhancement of up to  $\pm 0.2 \text{ dex}$  in [Fe/H] at a given radius. In the next section, we analyse if the stars of these streaks have been born at the same or different time.

#### 4.4 Deviations with time and age

- significant scatter across time (Fig. 8)
- horizontal streaks mentioned in previous Section 4.3 are also born at same time. We see similar colors, that is, stellar ages when tracing the aforementioned streaks at  $R_{\text{Gal.}} \sim 14 \text{ kpc}$  at  $\varpi_{\text{Gal.}} = 180 - 225^\circ$  from Fig. 6f in Fig. 8f. We identify them as Groups 1, 2, and 3 in Fig. 8.
- How much of a role can radial migration have played here (e.g. blue streak at  $R_{\text{Gal.}} \sim 14 \text{ kpc}$  with [Fe/H]  $\sim 0.1$  at  $\varpi_{\text{Gal.}} = 180 - 225^\circ$ )?
- We trace the position of Groups 1, 2, and 3 in the spatial dimension  $\varphi_{\text{Gal.}} - R_{\text{Gal.}}$  in Fig. 9 and confirm that they are connected.
- Fig. 9

**Take-Away:** We find significant scatter across the radial metallicity distribution caused by streaks of stars born with similar [Fe/H] at similar times and similar but slightly extended regions of the galaxy.

## 5 DISCUSSION

### 5.1 Linearity of the gradient

- Section 3.1 piecewise linear and quadratic functions clearly performing better than linear -> [Donor et al. \(2020\)](#); [Myers et al. \(2022\)](#), no conclusive results between either.
- Section 3.2: influence of binning not significant in our case
- Section 3.3: limited radial coverage could mimic quadratic function with two lines
  - Break radius  $R_{\text{break}} \sim 10 \text{ kpc}$  hard to fit.
  - What would the break radius mean physically at this distance? -> [Garcia et al. \(2023\)](#) found around 10 kpc in IllustrisTNG, [Grand et al. \(2016\)](#); [Chen et al. \(2023\)](#), Bar-driven mixing (see their Figs. 5-8 [Di Matteo et al. 2013](#))? Bar length? Disk length? Co-rotation radius?
- SB: Add these quantities as dashed lines in the  $R_{\text{Gal.}}$  diagrams?
- Discuss influence of bar. **ToDo: How strong is the bar of this simulation?**

### 5.2 Scatter and local variations

#### 5.2.1 Increasing scatter outwards

How much scatter in the radial metallicity gradient of young stars is expected based on simulations? Quantifying the scatter can provide insights into the effects of transient events like mergers, star formation bursts, and gas accretion.

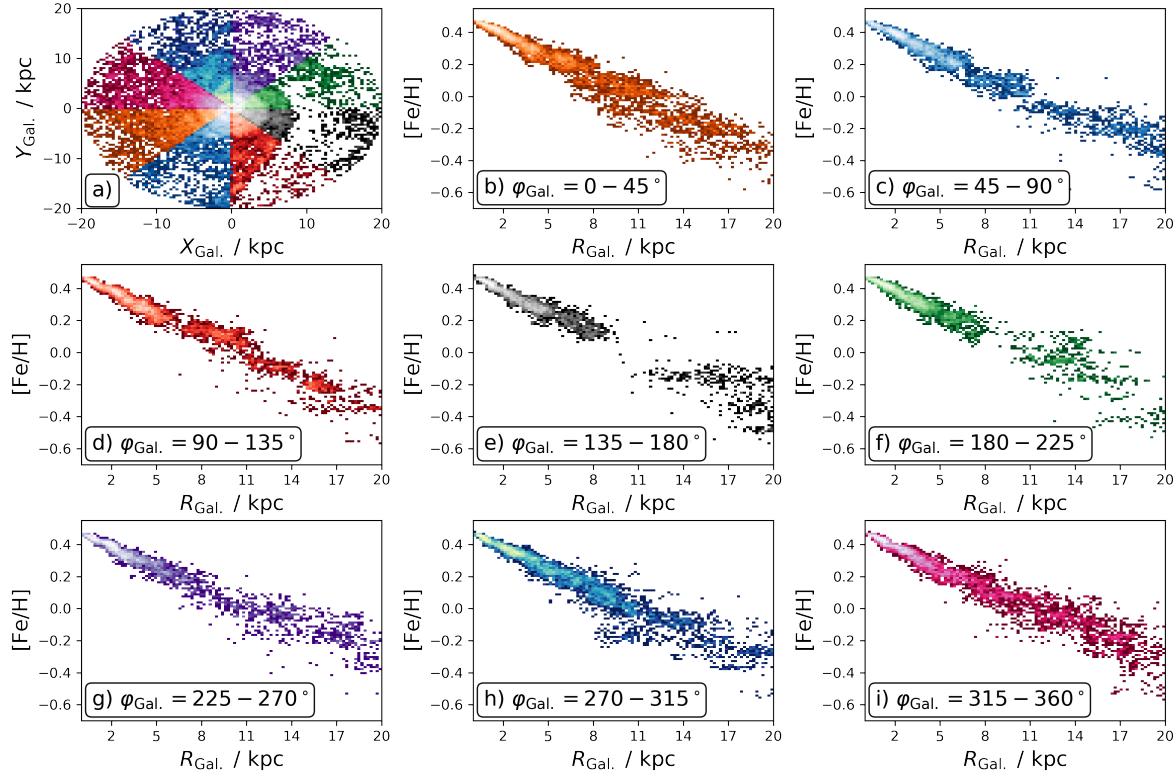
- inside-out formation and enhancement
- flaring
- density fluctuations
- all quite reasonable to explain increasing scatter outwards.

### 5.3 Localised vertical and azimuthal deviations and their correlation with gas

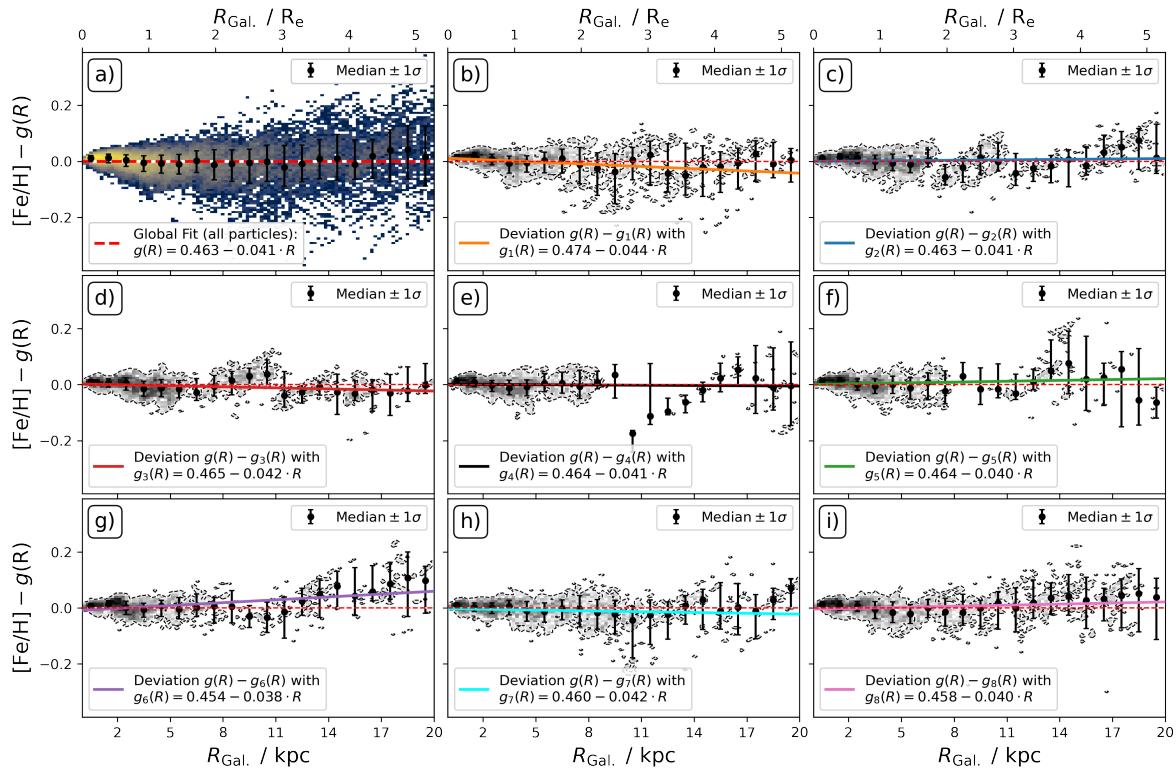
- Localised deviations both vertically and azimuthally.
- Correlation with gas?
- Gas gradient nearly linear. More linear than stellar gradient (Fig. 10). Test with RSS and AIC/BIC?
- Spatial correlations with gas? Follow up Fig. 11 by looking at the stellar and gas [Fe/H] for these regions!
- Talk about significant overdensities in Fig. 11, e.g. 7kpc for 0-45deg, 7, 11, 16 kpc for 135-180 or 9-11 for 270-315. Should we follow up the radial gas metallicity gradient as well!?
- Comparing Fig. 11 with Fig. 7: overlap these figures? looks like sometimes the  $\Delta z$  tracks the  $\Delta[\text{Fe}/\text{H}]$ , e.g. panels f? Also in oposite direction with panels b and i. So maybe better trace  $\Delta|z|$  vs.  $\Delta[\text{Fe}/\text{H}]$ ! Can we use this to identify a warp?

### 5.4 The impact of time and age: mixing and migration

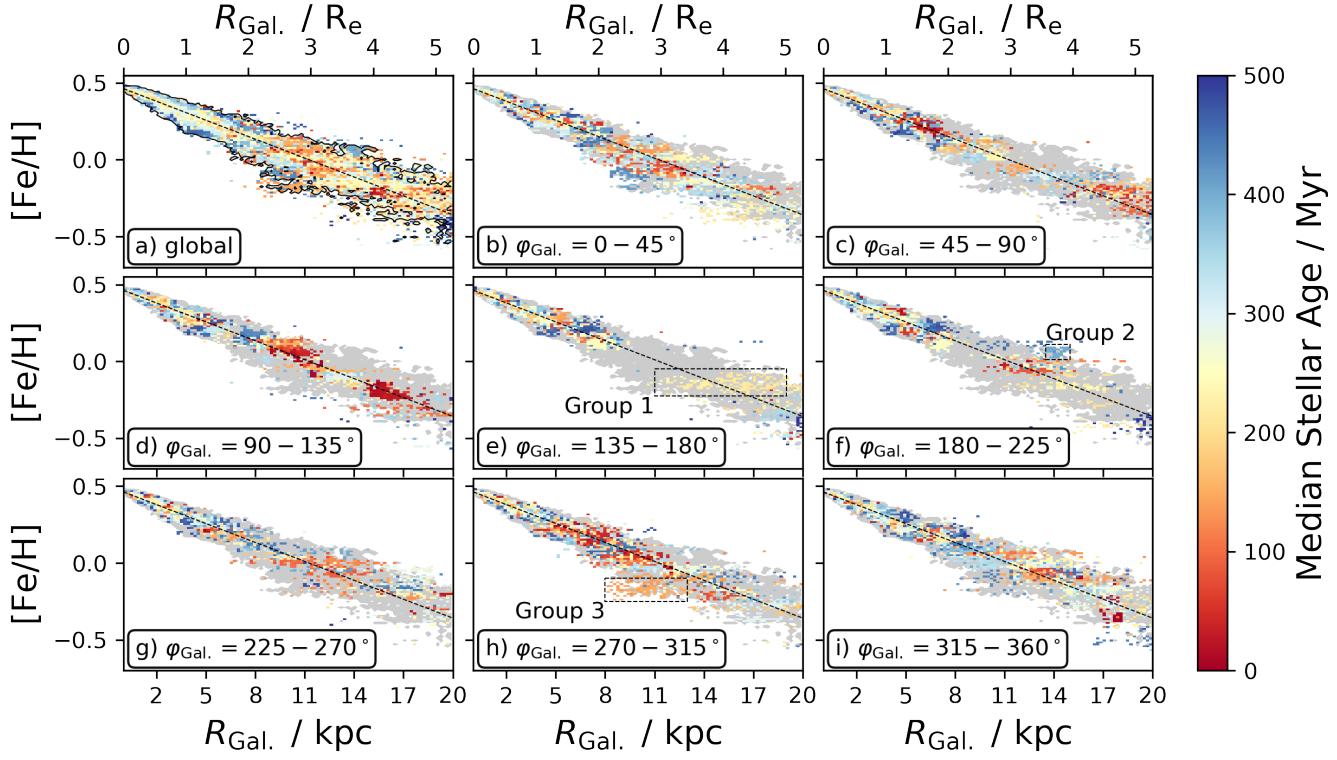
- Assessment of gradient for smaller and larger age limits than our default 0.5 Gyr in Fig. 12. Still quadratic even at smallest ages (Fig. 12a)
  - scatter and flattening increase (Fig. 12).
  - significant scatter already at 150Myr (Fig. 12c)
  - Radial Migration ([Grand et al. 2015, 2016](#)) bar-driven mixing ([Di Matteo et al. 2013](#))?
    - Radial Migration function prediction by Frankel?
    - Are Rbirth and Rguide even useful? Fig. 12c or Fig. 13b. Fig. 13b: similar structures as described in Section 4.4?



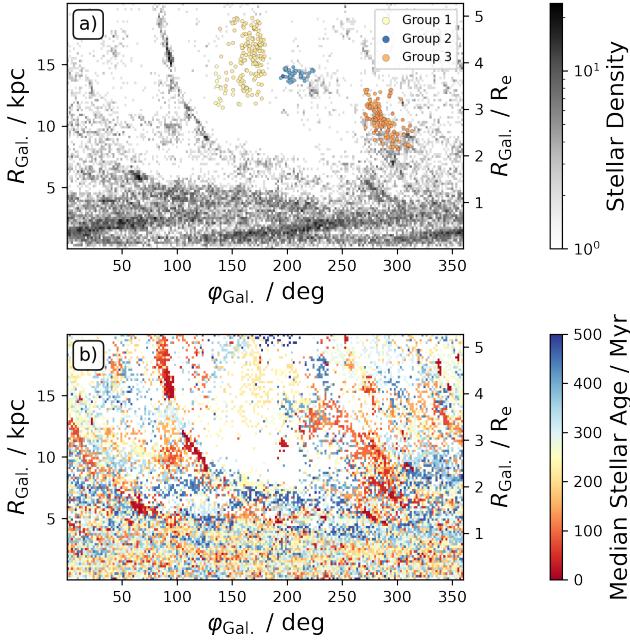
**Figure 6.** Stellar density variation across 8 different sectors (with color-code visualised in panel a) of the radial metallicity gradient  $R - [Fe/H]$  across 8 different azimuth ranges (panels b-i). A rotating lighthouse-like GIF animation of the median age and median density of the  $R - [Fe/H]$ -relation across different azimuths is freely available on a [repository](#).



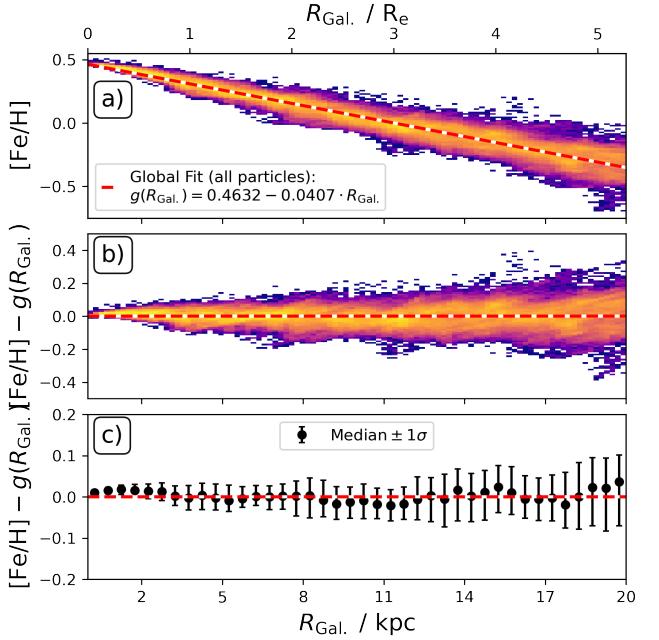
**Figure 7.** Stellar density variation of the radial metallicity gradient  $R - [Fe/H]$  across full stellar disk (panel a) and 8 different sectors (same panel order as for Fig. 6). Linear radial metallicity gradients have been fit globally (red dashed line) and for each sector with colors following the same color code as in Fig. 6.



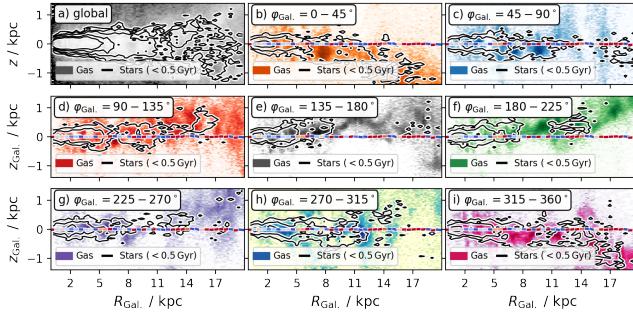
**Figure 8.** Same as Fig. 7, but colored by median age per bin. We identify 3 A rotating lighthouse-like GIF animation of the median age and median density of the  $R - [\text{Fe}/\text{H}]$ -relation across different azimuths is freely available a [repository](#).



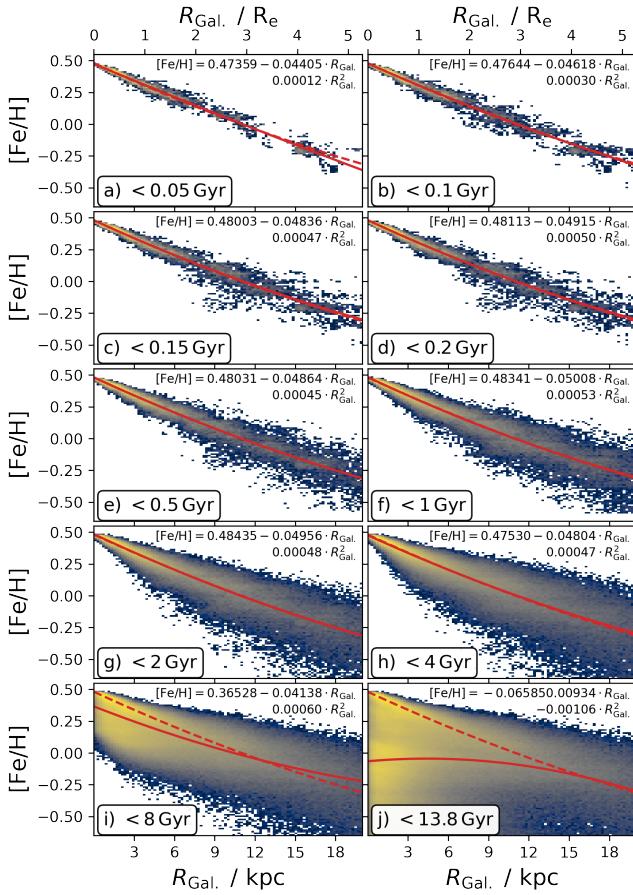
**Figure 9.** Density distribution (panel a) and age distribution (panel b) of young stars in the azimuthal and radial direction  $\varphi_{\text{Gal.}} - R_{\text{Gal.}}$ . In panel a), we also show the groups previously identified in Fig. 8.



**Figure 10.** Same as Fig. 2, but for gas.



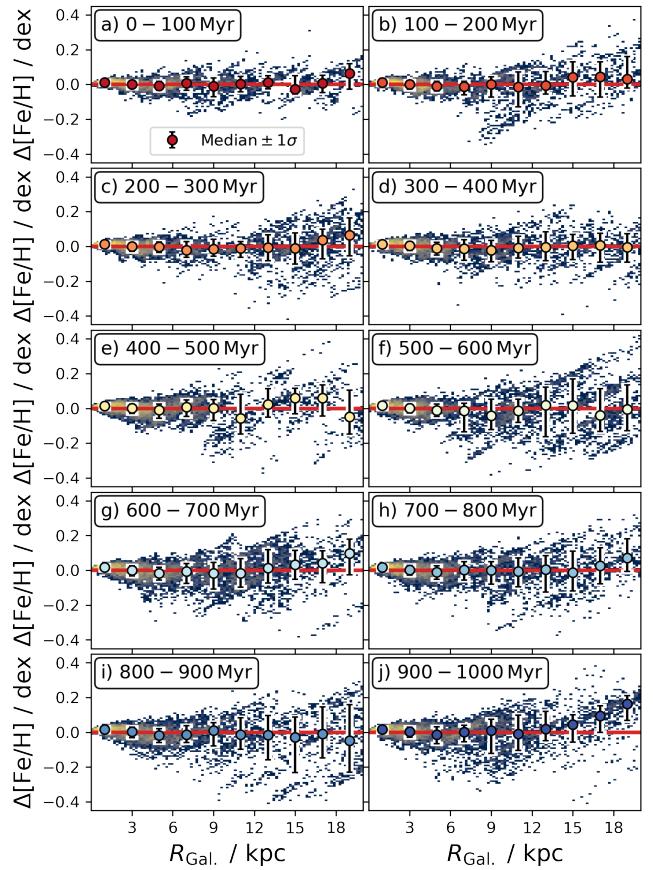
**Figure 11.** Tracing young stars and gas across galactocentric radii  $R_{\text{Gal.}}$  and height  $z_{\text{Gal.}}$  across the whole galaxy (panel a) and different azimuthal ranges/sectors (panels b-i). Small rectangles with cool-warm colors along the horizontal axis indicate the local gradient slopes as in Fig. 5.



**Figure 12.** Radial metallicity gradients and quadratic fits for different maximum age ranges. SB: Add fiducial dashed line for 0.5 Gyr!

## 5.5 Implications for Milky Way studies

- Local incomplete samples in the Milky Way may bias our gradient estimates, for example, with low [Fe/H] clusters or stars (Figs. 14a and Fig. 14b).
- The jury is still out if the gradient is piecewise linear or quadratic. We can say it is likely not only linear.
- What would happen if studies applied a quadratic function to



**Figure 13.** Stellar density distribution and spread of [Fe/H] across different galactocentric radii with respect to a global linear radial metallicity gradient across different age ranges. Panels a-j show young stars and exhibit a rather similar trend, whereas the scatter increases significantly for stars above 0.5 Gyr in panels k and l).

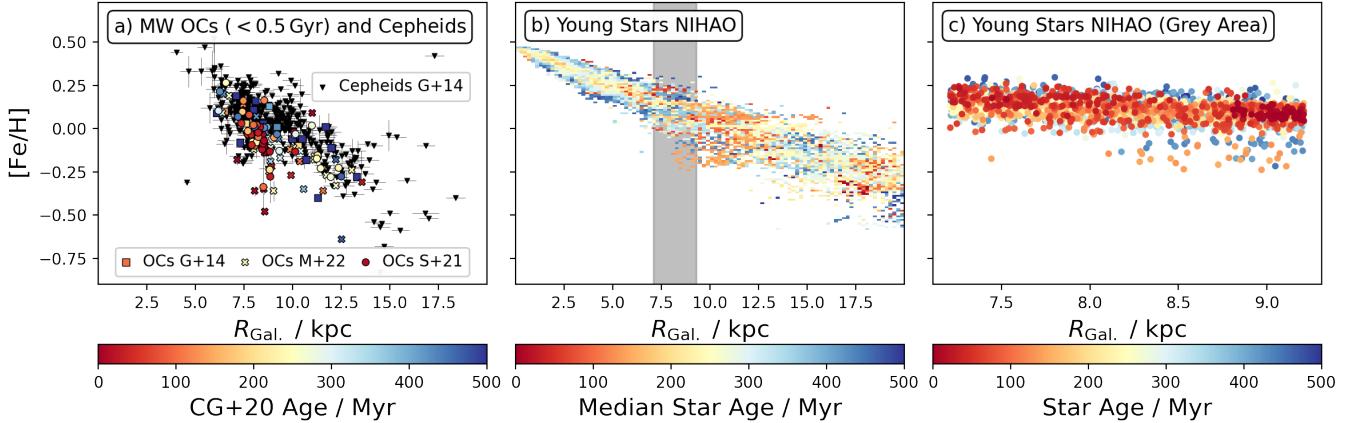
MW data? e.g. [Yong et al. \(2012\)](#); [Andrievsky et al. \(2004\)](#); [Genovali et al. \(2014\)](#)?

- Cosmological zoom-in simulations are now almost reaching the resolution to home in on a "solar" vicinity in Simulations with sufficiently large stellar particle numbers (Fig. 15).
- While star particle numbers and mass resolution might still be too low, we can study the young star - gas connection (Fig. 16).
- We recover the gradients, but also more nuanced differences in the local [Fe/H].
- Similar to [Grand et al. \(2016\)](#), we find lower [Fe/H] in the leading edges of gas overdensities and higher [Fe/H] in the trailing edges of gas overdensities.

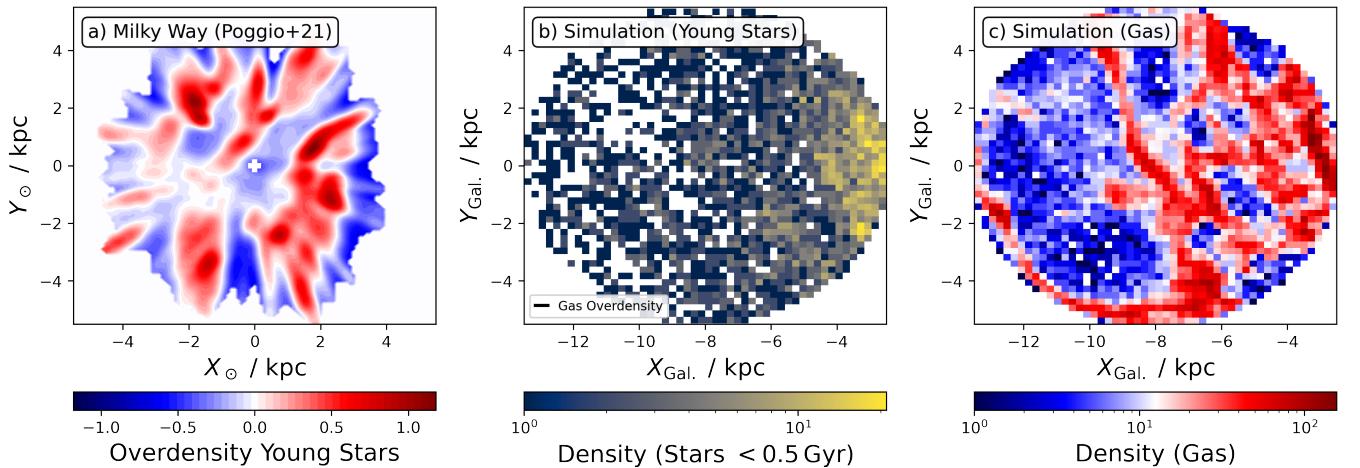
## Take-Away:

## 5.6 Implications for extragalactic studies

- NIHAO-UHD provides both [Fe/H] and  $A(\text{O}) = 12 + \log(\text{O}/\text{H})$  for stars and gas. **To Do: Should we create Fig. 16 also with  $A(\text{O})$ ?**
- It would be interesting to test a quadratic function with extragalactic data (e.g. [Chen et al. 2023](#); [Bresolin et al. 2012](#)).
- Break Radius?
- While not primary focus of this study, we are also looking at the



**Figure 14.** Comparison of the Milky Way’s radial metallicity trend as traced by Cepheids (black triangles, compiled from literature by Genovali et al. 2014, G+14) as well as young ( $< 0.5$  Gyr) open cluster of the Milky Way as traced by the literature compilation from Genovali et al. (2014, G+14 as squares), APOGEE DR17 from Myers et al. (2022, M+22 as crosses), and GALAH DR3 from Spina et al. (2021, S+21 as circles). The latter two are compiled based on the membership and age catalogue by Cantat-Gaudin & Anders (2020, CG+20).



**Figure 15.** Comparison of density distribution of young stars and gas in the Milky Way and the NIHAO Milky Way analogue simulation. Panel a) shows the measurements of the Solar vicinity within 5 kpc by Poggio et al. (2021). Panels b) and c) show young stars and gas NIHAO, respectively, for a selected region similar to panel a). Black and white contour lines in panel b) trace overdensities in the gas distribution of panel c).

outskirts of the galaxy, where Garcia et al. (2023) found a metallicity floor in IllustrisTNG galaxies. When extending our radius to  $R_{\text{Gal.}} \leq 100$  kpc, we find a similar abundance floor of  $[\text{Fe}/\text{H}] \geq -0.64$  for young stars and  $A(\text{O}) \geq 8.12$  for the majority of gas (see Fig. 17<sup>3</sup>

- **ToDo: Is  $R_{\text{break}}$  close to 20 kpc? Is the  $[\text{Fe}/\text{H}]$  at 20 kpc close to  $[\text{Fe}/\text{H}]_{\text{break}}$ ?**
- Is this close to the "end" of thin disks? In the Milky Way, the thin disk seems to go down to  $[\text{Fe}/\text{H}] \sim -0.7$  (Bensby et al. 2014; Buder et al. 2019). Can we thus use this to identify a disk edge?

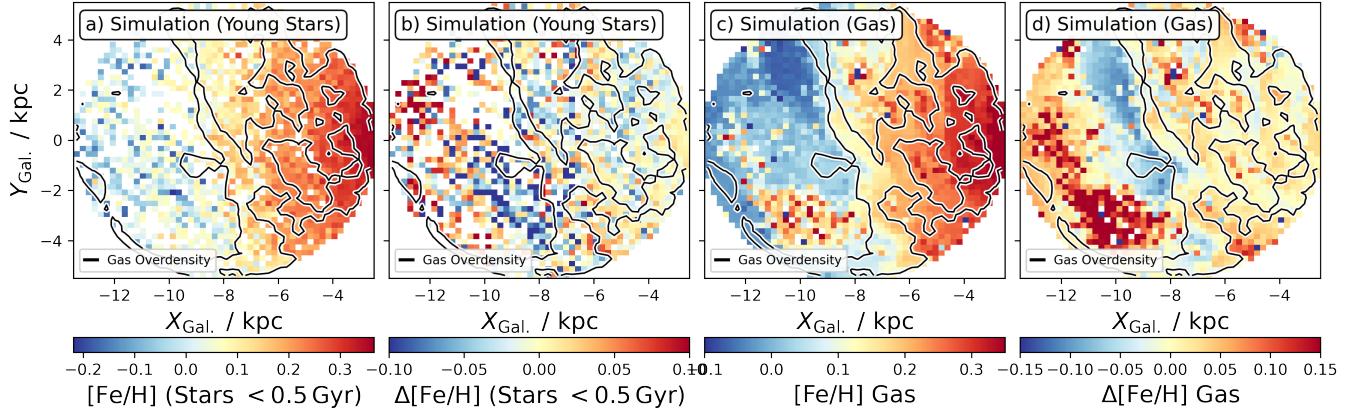
## Take-Away:

<sup>3</sup> Spatial distribution in Fig. B1. Note the galaxy without gas at  $(Y, Y, Z) = (-50, -0.75, -20 - 50)$  kpc.

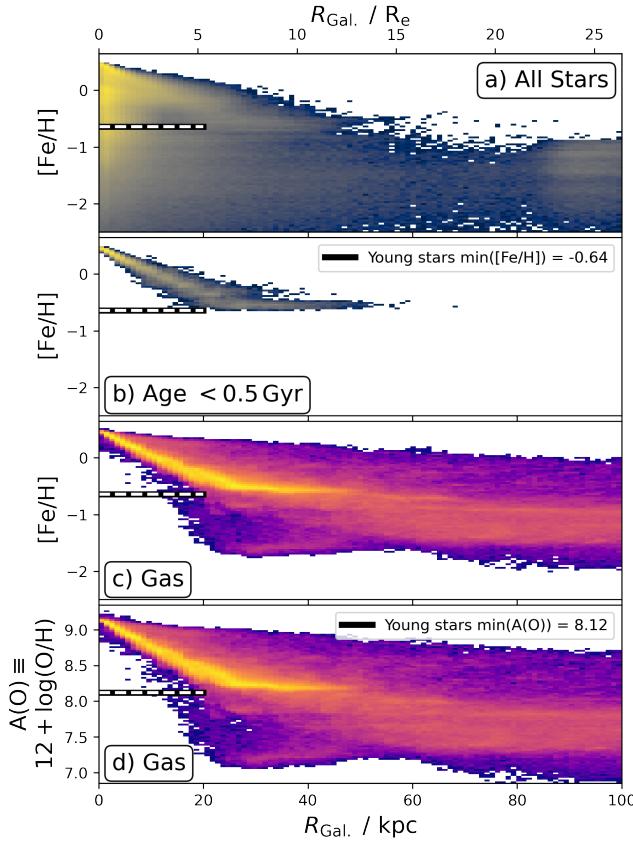
## 6 CONCLUSIONS

### 6.1 Main Take-Away

- Piecewise linear and quadratic functions both perform better than a linear fit to the radial metallicity relation. However, we see no significant preference between piecewise and quadratic functions based on our assessments. **SB: Discussion outcome**
- While the specific slopes differ when fitting all points or binned data, they agree within the still rather small fitting uncertainties. **SB: Discussion outcome**
- We find that a piecewise linear function can well mimic a quadratic function across the scales used in Milky Way and extragalactic studies. Local deviations become traceable below a spatial resolution of  $\Delta R_{\text{Gal.}} \leq 2$  kpc. **SB: Discussion outcome**
- We see no strong correlations of deviations in the vertical direction across the whole simulation. Such correlations could, however,



**Figure 16.** Comparison of density distribution of young stars and gas NIHAO Milky Way analogue simulation for the same regions as Figs. 15b and 15c. Panels a) and c) trace median gas and young star iron abundances, respectively. Panels b) and d) plot the residual [Fe/H] of gas and stars, respectively, when correcting with a radial metallicity gradient fit. Black and white contour lines in panel b) trace overdensities in the gas distribution of Fig. 15c).



**Figure 17.** Radial metallicity functions for all stars (panel a), young stars (panel b), and gas (panels c and d for iron and oxygen as metallicity tracers) out to  $R_{\text{Gal.}} \leq 100$  kpc. Panels b and c are comparable to Figs. 2a and 10a for a smaller radial coverage.

be blurred by azimuthal effects, like the galactic warp, which needs to be disentangled in the azimuthal dimension. **SB: Discussion outcome**

- We find various deviations from the global trend in azimuthal direction, including gaps as well as isolated streaks of stars with similar [Fe/H] across  $\Delta R_{\text{Gal.}} = 2 - 6$  kpc. These can introduce local

over- and under-enhancement of up to  $\pm 0.2$  dex in [Fe/H] at a given radius. **SB: Discussion outcome**

- We find significant scatter across the radial metallicity distribution caused by streaks of stars born with similar [Fe/H] at similar times and similar but slightly extended regions of the galaxy. **SB: Discussion outcome**

- **SB: Implications for Milky Way studies**
- **SB: Implications for extragalactic studies**

## 6.2 Future Research

- Trace spatial and chemical coherence of over- and underdensities over times. Do this both across a range of stellar ages at redshift zero, but also when looking back at higher redshifts.

- higher mass resolution would help to overcome small sample sizes for comparison with Milky Way.

- We assumed that the modelled chemistry is fully accurate. While previous works are showing that relative trends for several elements agree, we are still missing some of the details on the origin of elements, such as a complete picture of the synthesis sites, environments, and yields for elements. detailed physics?

- our study is making deductions based on a simulated Milky Way analogue, but the reality in both the actual Milky Way and other Milky Way analogues with slightly different nuances like merger history as well as bar and spiral arm strengths might lead to a different slightly or vastly radial metallicity relation.

- Potential Break radius around  $R_{\text{Gal.}} \sim 10$  kpc  $\sim 2.5 R_e$  is still enigmatic. More studies needed to establish a link to a physical mechanism.

- It has to be seen what functional shape this radial metallicity relation is actually having, but our results are certainly encouraging to not only apply piecewise linear fits to galactic and extragalactic radial metallicity measurements.

## ACKNOWLEDGEMENTS

We acknowledge the traditional owners of the land on which the ANU stands, the Ngunnawal and Ngambri people. We pay our respects to elders past, and present and are proud to continue their tradition of

surveying the night sky and its mysteries to better understand our Universe.

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## SOFTWARE

The research for this publication was coded in PYTHON (version 3.7.4) and included its packages ASTROPY (v. 3.2.2; [Astropy Collaboration et al. 2013, 2018](#)), IPYTHON (v. 7.8.0; [Pérez & Granger 2007](#)), MATPLOTLIB (v. 3.1.3; [Hunter 2007](#)), NUMPY (v. 1.17.2; [Walt et al. 2011](#)), PYNBODY (v. 1.1.0; [Pontzen et al. 2013](#)), SCIPY (v. 1.3.1; [Virtanen et al. 2020](#)), SKLEARN (v. 1.5.1 [Pedregosa et al. 2011](#)) STATSMODELS (v. 0.14.2 [Perktold et al. 2024](#)) We further made use of TOPCAT (version 4.7; [Taylor 2005](#));

## DATA AVAILABILITY

All code and data to reproduce the analysis and figures can be publicly accessed via [https://github.com/svenbuder/nihao\\_radial\\_metallicity\\_gradients](https://github.com/svenbuder/nihao_radial_metallicity_gradients).

The repository also includes the chemical and kinematic data of the simulated Milky Way analogue g8.26e11 for last snapshot of the simulation for the observable footprint that were used for this study as well as the cleaned catalogue of GALAH DR3. All GALAH DR3 data is also published by [Buder et al. \(2021\)](#) and can be accessed publicly via <https://docs.datacentral.org.au/galah/dr3/overview/>. The full simulation data of g8.26e11 can be obtained upon reasonable request from the authors. Currently the only limitation in making all data public is limited cloud space to host the data. We encourage interested readers to get in contact with the authors for full data access. Redshift zero snapshots from the original NIHAO-UHD simulations can be found here: [https://tobias-buck.de/#sim\\_data](https://tobias-buck.de/#sim_data).

If you are using either of these data to follow up on this research, remember to give appropriate credit to the researchers who created and curated either data set, that is, at least to [Buder et al. \(2021, 2022\)](#) and [Buck et al. \(2020, 2021\)](#).

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**Table A1.** Global linear gradient fit results with different methods. LINEARREGRESSION is part of the SKLEARN package.

Method	Intercept ( $a_0 \pm \sigma_{a_0}$ )	Slope ( $a_1 \pm \sigma_{a_1}$ )
STATSMODELS.API.ODR	$0.46266 \pm 0.00039$	$-0.04109 \pm 0.00005$
SCIPY.ODR	$0.46268 \pm 0.00039$	$-0.04109 \pm 0.00005$
NP.POLYFIT	$0.46266 \pm 0.00039$	$-0.04109 \pm 0.00005$
LINEARREGRESSION	$0.46265 \pm 0.00033$	$-0.04109 \pm 0.00006$
SCIPY.CURVE_FIT	$0.46266 \pm 0.00039$	$-0.04109 \pm 0.00005$

## APPENDIX A: TESTING DIFFERENT FITTING ROUTINES

Table A1

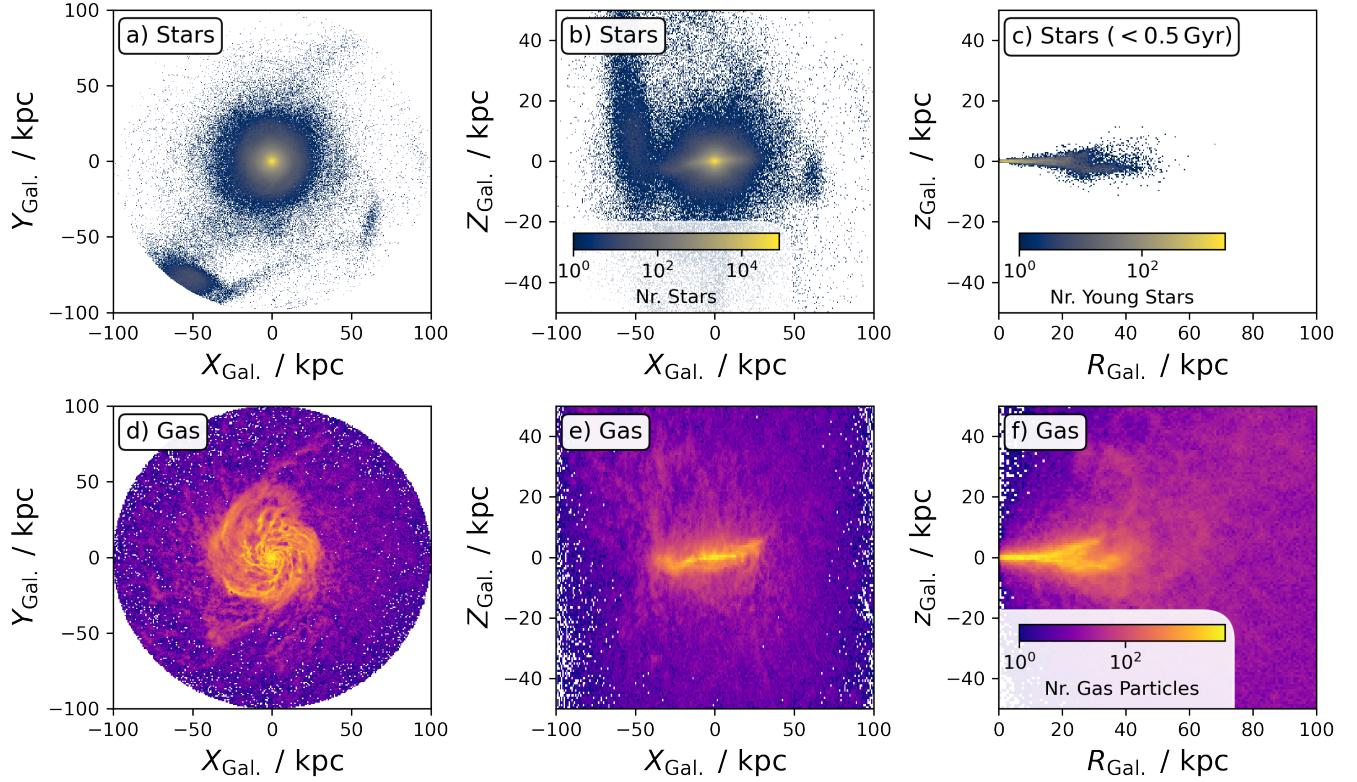
## APPENDIX B: INTERESTING BUT NOT MAIN FOCUS OF THIS PAPER

Fig. B1 for an extended  $R_{\text{Gal.}} \leq 100$  kpc and  $|z_{\text{Gal.}}| \leq 50$  kpc.

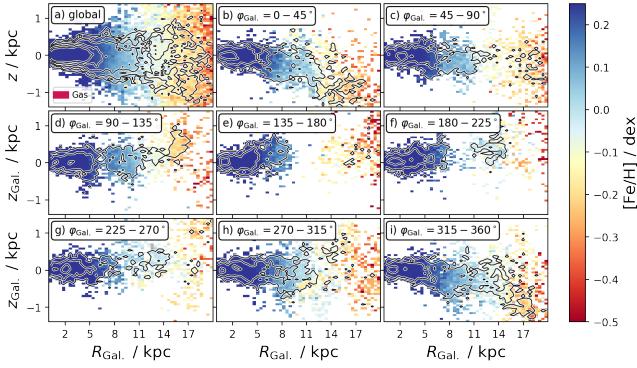
Figs. B2 and B3

Fig. B4

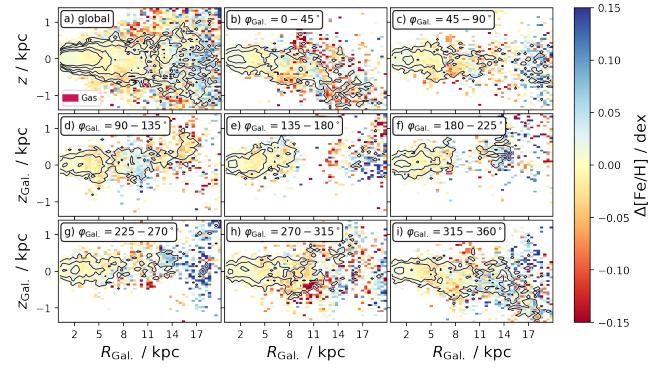
This paper has been typeset from a  $\text{\TeX}/\text{\LaTeX}$  file prepared by the author.



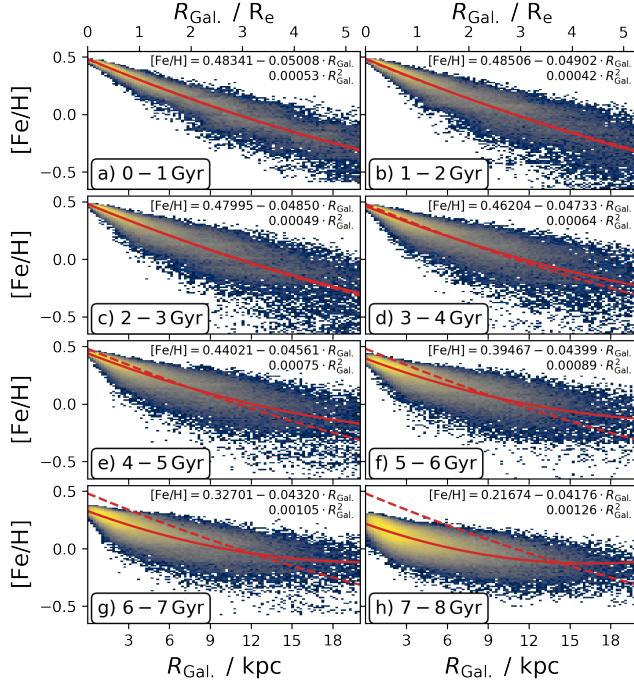
**Figure B1.** Same as Fig. 1, but for an extended  $R_{\text{Gal.}} \leq 100$  kpc and  $|z_{\text{Gal.}}| \leq 50$  kpc.



**Figure B2.** Tracing young stars' iron abundance  $[\text{Fe}/\text{H}]$  across galactocentric radii  $R_{\text{Gal.}}$  and height  $z_{\text{Gal.}}$  across the whole galaxy (panel a) and different azimuthal ranges/sectors (panels b-i).



**Figure B3.** Tracing young stars' iron abundance  $[\text{Fe}/\text{H}]$  residuals (relative to global gradient) across galactocentric radii  $R_{\text{Gal.}}$  and height  $z_{\text{Gal.}}$  across the whole galaxy (panel a) and different azimuthal ranges/sectors (panels b-i).



**Figure B4.** Radial metallicity gradients and quadratic fits for 1 Gyr age bins.