Why do high-redshift galaxies show diverse gas-phase metallicity gradients?

Xiangcheng Ma,1* Philip F. Hopkins,1 Robert Feldmann,2,3 Paul Torrey,1,4 Claude-André Faucher-Giguère5 and Dušan Kereš6

1TAPIR, MC 350-17, California Institute of Technology, Pasadena, CA 91125, USA
2Institute for Computational Science, University of Zurich, Zurich CH-8057, Switzerland
3Department of Astronomy and Theoretical Astrophysics Center, University of California Berkeley, Berkeley, CA 94720, USA
4MIT Kavli Institute for Astrophysics and Space Research, Cambridge, MA 02139, USA
5Department of Physics and Astronomy and CIERA, Northwestern University, 2145 Sheridan Road, Evanston, IL 60208, USA
6Department of Physics, Center for Astrophysics and Space Sciences, University of California at San Diego, 9500 Gilman Drive, La Jolla, CA 92093, USA

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ABSTRACT
Recent spatially resolved observations of galaxies at \( z \sim 0.6–3 \) reveal that high-redshift galaxies show complex kinematics and a broad distribution of gas-phase metallicity gradients. To understand these results, we use a suite of high-resolution cosmological zoom-in simulations from the Feedback in Realistic Environments project, which include physically motivated models of the multiphase interstellar medium, star formation and stellar feedback. Our simulations reproduce the observed diversity of kinematic properties and metallicity gradients, broadly consistent with observations at \( z \sim 0–3 \). Strong negative metallicity gradients only appear in galaxies with a rotating disc, but not all rotationally supported galaxies have significant gradients. Strongly perturbed galaxies with little rotation always have flat gradients. The kinematic properties and metallicity gradient of a high-redshift galaxy can vary significantly on short time-scales, associated with starburst episodes. Feedback from a starburst can destroy the gas disc, drive strong outflows and flatten a pre-existing negative metallicity gradient. The time variability of a single galaxy is statistically similar to the entire simulated sample, indicating that the observed metallicity gradients in high-redshift galaxies reflect the instantaneous state of the galaxy rather than the accretion and growth history on cosmological time-scales. We find weak dependence of metallicity gradient on stellar mass and specific star formation rate (sSFR). Low-mass galaxies and galaxies with high sSFR tend to have flat gradients, likely due to the fact that feedback is more efficient in these galaxies. We argue that it is important to resolve feedback on small scales in order to produce the diverse metallicity gradients observed.

Key words: galaxies: evolution – galaxies: formation – cosmology: theory.

1 INTRODUCTION
Metallicity is a fundamental property of galaxies. In the local Universe, galaxy stellar mass correlates tightly with both gas-phase metallicity (e.g. Tremonti et al. 2004; Lee et al. 2006) and stellar metallicity (e.g. Gallazzi et al. 2005; Kirby et al. 2013), known as the galaxy mass–metallicity relation (MZR). The MZR also exists at higher redshifts up to \( z \sim 3 \) (e.g. Erb et al. 2006; Maiolino et al. 2008; Mannucci et al. 2009; Zahid, Kewley & Bresolin 2011; Steidel et al. 2014; Yabe et al. 2014; Sanders et al. 2015). The MZR evolves smoothly with redshift, with galaxies being more metal enriched at lower redshift (e.g. Zahid et al. 2013). The MZR results from the interplay between gas accretion and recycling, star formation and feedback-driven outflows (e.g. Edmunds 1990; Davé, Finlator & Oppenheimer 2012; Feldmann 2013; Lilly et al. 2013; Lu, Blanc & Benson 2015), so it is widely used to constrain feedback models in cosmological simulations and semi-analytic models of galaxy formation (e.g. Davé, Finlator & Oppenheimer 2011; Lu et al. 2014; Torrey et al. 2014; Ma et al. 2016b). Historically, galaxy metallicity is usually measured in the central regions despite the presence of metallicity gradients. Since Searle (1971), it has been known that galaxies in the local Universe tend to have negative gas-phase metallicity gradients,
which means that galaxies are more metal-enriched in the central region than at the outskirts (e.g. Zaritsky, Kennicutt & Huchra 1994; van Zee et al. 1998; Sánchez et al. 2012, 2014). The slope of metallicity gradients of non-interacting galaxies, if normalized to some characteristic radius (e.g. the effective radius), does not depend strongly on galaxy properties, such as morphology, the existence of bars, magnitude, stellar mass, etc. (e.g. Zaritsky et al. 1994; Sánchez et al. 2014; Ho et al. 2015; however, see Vila-Costas & Edmunds 1992). This can be understood by a simple model where gas and stellar discs co-evolve under virtually closed-box assumptions (Ho et al. 2015). Interacting galaxies are underabundant in their central regions (e.g. Kewley, Geller & Barton 2006; Peeples, Popge & Stanek 2009) and show evidence of shallower gas-phase metallicity gradients than isolated galaxies of similar masses (e.g. Vila-Costas & Edmunds 1992; Kewley et al. 2010; Rupke, Kewley & Chien 2010b), owing to strong radial inflow of low-metallicity gas from the outskirts towards the galactic centre (e.g. Rupke, Kewley & Barnes 2010a; Torrey et al. 2012).

It is only in the past few years that gas-phase metallicity gradients have been directly measured in galaxies beyond the local Universe. Early attempts include resolved studies of several strongly lensed galaxies at redshift \( z \sim 1.5–2.4 \) (e.g. Jones et al. 2010, 2013; Yuan et al. 2011). Four out of five of these galaxies show well-ordered rotation and have steeper slopes (in dex kpc\(^{-1}\)) in metallicity gradient than those of galaxies in the local Universe. In addition, Maciel, Costa & Uchida (2003) measured the abundances of planetary nebulae in the Milky Way (MW) generated by stars spanning a broad age interval and suggested that the MW had steeper metallicity gradients back to \( z \sim 1.5 \). These results support the so-called ‘inside-out’ growth model of galaxy formation (e.g. Bouwens, Cayón & Silk 1997). In this scenario, the central galactic bulge formed rapidly at early times, building a steep radial metallicity gradient at high redshift. The size of the disc gradually grows with time via gas inflow. The metallicity gradient gradually weakens via star formation in the outer disc and radial gas mixing. Such a picture is also seen in some cosmological hydrodynamic simulations (e.g. Pilkington et al. 2012; Gibson et al. 2013), where the metallicity gradients are steepest at high redshift and gradually flatten at late times.

Recently, Leethochawalit et al. (2016) have studied 11 gravitationally lensed galaxies at redshift \( z \sim 1.4–2.5 \) and found a broad distribution of kinematics and abundance patterns (see also Jones et al. 2015; Wang et al. 2016). Most galaxies in their sample show no features of well-ordered rotation and tend to have flat gas-phase metallicity gradient, in contrast to earlier statements that high-redshift galaxies tend to have stronger metallicity gradients (Jones et al. 2013). Moreover, large samples of non-lensed galaxies at redshift \( z \sim 0.6–3 \) also show diverse metallicity gradients (e.g. Cresci et al. 2010; Querey et al. 2012; Swinbank et al. 2012; Stott et al. 2014; Wyfts et al. 2016), with slope varying from negative to flat and positive. For example, Wyfts et al. (2016) have found that only 15 out of 180 galaxies that have spatially resolved measurements of abundances in a sample of galaxies at \( z \sim 0.6–2.7 \) show statistically significant non-zero slope of metallicity gradients. These results complicate the simple ‘inside-out’ growth picture.

Various studies have pointed out the necessity of strong feedback in order to avoid steep metallicity gradients in high-redshift galaxies in cosmological hydrodynamic simulations (e.g. Pilkington et al. 2012; Gibson et al. 2013; Anglés-Alcázar et al. 2014). For example, Gibson et al. (2013) compared two cosmological simulations run with different feedback models and showed that their ‘enhanced’ feedback model produces constantly flat metallicity gradients at high redshift, whereas their ‘conservative’ feedback model tends to follow the simple ‘inside-out’ growth scenario and produce steep metallicity gradients. However, they do not reproduce the diverse range of metallicity gradients in high-redshift galaxies (only one or the other). In addition, many simulations used empirical feedback models where galactic winds are generated by manually kicking particles and enforcing these wind particles to be temporarily decoupled from hydrodynamics (e.g. Davé et al. 2011; Anglés-Alcázar et al. 2014; Torrey et al. 2014) or artificially preventing SNe bubbles from cooling for much longer time (e.g. Stinson et al. 2013). Such models do not properly resolve the launch and propagation of galactic winds from the interstellar medium (ISM) scale and tend to artificially mix metals on large scales and prevent strong metallicity gradients from forming.

In this work, we study the origin and evolution of galaxy metallicity gradients using 32 cosmological zoom-in simulations from the Feedback in Realistic Environments project (FIRE; Hopkins et al. 2014). These simulations include physically motivated models of the multiphase ISM, star formation and stellar feedback, with sufficient spatial and mass resolution down to giant molecular cloud scales to explicitly resolve the launch and propagation of galactic winds. This is essential in studying metallicity gradients using simulations. In previous studies, it has been shown that these simulations reproduce many observed scaling relations, such as the stellar mass–halo mass relation, the Kennicutt–Schmidt relation, the star-forming main sequence (Hopkins et al. 2014) and the MZR (Ma et al. 2016b), for a broad range of halo mass and redshift, without the need for fine-tuning. These simulations also predict a reasonable covering fraction of neutral absorbers in the circumgalactic medium (CGM) at both low and high redshift (Faucher-Giguère et al. 2015, 2016; Hafen et al. 2016), mass loading factor of galactic outflows (Muratov et al. 2015) and density profiles, kinematics and chemical abundances of local dwarf galaxies (Otohor et al. 2015; Chan et al. 2015), all broadly consistent with observational constraints. All of these demonstrate the validity of using the FIRE simulations to study metallicity gradients.

Almost all galaxies in the FIRE simulations at high redshift \( (z > 0.5) \) show strong variability (burstiness) in star formation rates (SFRs) on short time-scales of the order of 10 Myr (Hopkins et al. 2014; Muratov et al. 2015; Sparre et al. 2015; Feldmann et al. 2016a). In these systems, rapid gas inflows trigger starbursts in the galactic centre (Torrey et al. 2016). In turn, feedback from newly formed stars injects sufficient energy and momentum into the ISM to destroy the gas disc and launch galactic winds. At lower redshift \( (z < 0.5) \), on the other hand, massive galaxies \( (M_h \gtrsim 10^{10} M_\odot) \) have calmed down, with star formation in the disc being regulated by gas inflow and feedback to more stable rates (e.g. Faucher-Giguère, Quataert & Hopkins 2013), and feedback can no longer damage the disc nor drive strong gas outflows (Muratov et al. 2015). This transition is likely due to a combination of decreasing galaxy merger rates (e.g. Hopkins et al. 2010) and decreasing gas fractions in galaxies (e.g. Hayward & Hopkins 2017) at low redshift. In this paper, we show that the FIRE simulations reproduce the diversity of kinematics and metallicity gradients observed in high-redshift galaxies. We also show that bursty star formation can produce the observed diversity – a galaxy may change kinematic properties and metallicity gradient between starburst episodes. This is important for the interpretation of the observed metallicity gradients in high-redshift galaxies.

\[^1\] http://fire.northwestern.edu.
Table 1. Simulation details.

| Name          | $M_{\text{halo}} (z=0)$ ($M_{\odot}$) | $M_{\text{halo}} (z=2)$ ($M_{\odot}$) | $m_\text{b}$ ($M_{\odot}$) | $\epsilon_\text{b}$ (pc) | $m_\text{dm}$ ($M_{\odot}$) | $\epsilon_\text{dm}$ (pc) | Reference |
|---------------|---------------------------------------|---------------------------------------|-----------------------------|---------------------------|----------------------------|---------------------------|-----------|
| m11           | 1.4e11                                | 3.8e10                                | 7.1e3                       | 7.0                       | 3.5e4                     | 70                        | Hopkins et al. (2014) |
| m12v          | 6.3e11                                | 2.0e11                                | 3.9e4                       | 10                        | 2.0e5                     | 140                       | Hopkins et al. (2014)   |
| m12q          | 1.2e12                                | 5.1e11                                | 7.1e3                       | 10                        | 2.8e5                     | 140                       | Hopkins et al. (2014)   |
| m12i          | 1.1e12                                | 2.7e11                                | 5.0e8                       | 14                        | 2.8e5                     | 140                       | Hopkins et al. (2014)   |
| m13           | 6.0e12                                | 8.4e11                                | 3.6e5                       | 21                        | 2.2e6                     | 210                       | Hopkins et al. (2014)   |
| m11h383       | 1.0e11                                | 4.1e9                                 | 1.7e4                       | 10                        | 8.3e4                     | 100                       | Chan et al. (2015)      |
| m11.4a        | 2.6e11                                | 8.9e10                                | 3.3e4                       | 9                         | 1.7e5                     | 140                       | Hafen et al. (2016)     |
| m11.9a        | 8.4e11                                | 1.3e11                                | 3.4e4                       | 9                         | 1.7e5                     | 140                       | Hafen et al. (2016)     |
| MFez0_A2      | 1.0e13                                | 2.2e12                                | 3.0e5                       | 9                         | 1.4e6                     | 140                       | Hafen et al. (2016)     |
| z2h350        | –                                     | 7.9e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h400        | –                                     | 7.9e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h450        | –                                     | 8.7e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h500        | –                                     | 1.2e12                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h550        | –                                     | 1.9e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h600        | –                                     | 6.7e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| z2h650        | –                                     | 4.0e11                                | 5.9e4                       | 9                         | 2.9e5                     | 143                       | Faucher-Giguere et al. (2015) |
| A1:0          | –                                     | 2.3e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A2:0          | –                                     | 2.9e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A3:0          | –                                     | 2.4e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A4:0          | –                                     | 2.8e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A5:0          | –                                     | 2.3e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A6:0          | –                                     | 2.6e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A7:0          | –                                     | 2.5e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A8:0          | –                                     | 3.5e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A9:0          | –                                     | 2.8e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| A10:0         | –                                     | 3.2e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| B1:0          | –                                     | 8.3e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| B2:0          | –                                     | 9.0e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| B3:0          | –                                     | 9.7e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| B4:0          | –                                     | 8.5e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |
| B5:0          | –                                     | 9.1e12                                | 3.3e4                       | 10                        | 1.7e5                     | 143                       | Feldmann et al. (2016b)  |

Parameters describing the initial conditions for our simulations (units are physical):
(1) Name: simulation designation.
(2) $M_{\text{halo}}$: approximate mass of the main halo (most massive halo), at $z=0$ and $z=2$.
(3) $m_\text{b}$: initial baryonic (gas and star) particle mass in the high-resolution region.
(4) $\epsilon_\text{b}$: minimum baryonic Plummer-equivalent force softening (minimum SPH smoothing lengths are comparable or smaller). Force softening is adaptive (mass resolution is fixed).
(5) $m_\text{dm}$: dark matter particle mass in the high-resolution region.
(6) $\epsilon_\text{dm}$: minimum dark matter Plummer-equivalent force softening (fixed in physical units at all redshifts).
(7) Reference: where the simulation is first presented.

Note. Detailed physical properties of these galaxies are presented in Appendix A.

The paper is organized as follows. We start by introducing the simulations and describing the methods to measure kinematic properties and gas-phase metallicity gradient in the simulated galaxies in Section 2. We present the main results in Section 3 and discuss and conclude in Section 4.

We adopt a standard flat Λ cold dark matter cosmology with cosmological parameters $H_0 = 70.2$ km s$^{-1}$ Mpc$^{-1}$, $\Omega_\Lambda = 0.728$, $\Omega_m = 1 - \Omega_\Lambda = 0.272$, $\sigma_8 = 0.807$ and $n = 0.961$, broadly consistent with observations (e.g. Hinshaw et al. 2013; Planck Collaboration XVI 2014).

2 METHODOLOGY

2.1 Simulation details

In this work, we use a suite of simulations from the FIRE project that have been presented in previous studies (Hopkins et al. 2014; Chan et al. 2015; Faucher-Giguère et al. 2015; Feldmann et al. 2016b; Hafen et al. 2016). These are cosmological 'zoom-in' simulations that are run using GIZMO (Hopkins 2013) in P-SPH mode (Hopkins 2013). Because of computational expense, some of them are only run to $z=2$, and span a halo mass $10^{11}-10^{13} M_{\odot}$ at that redshift. For those that are run to $z=0$, we only include the ones above $z=0$ halo mass $10^{11} M_{\odot}$ in this study, since smaller galaxies lack observational probes at high redshift. All the simulations used in this paper, along with the mass of the most massive halo in the zoom-in region, the initial mass of baryonic and dark matter particles, minimum force softening lengths and the reference where the simulation is first presented, are listed in Table 1. We briefly summarize the physical models below for completeness, but refer to Hopkins et al. (2014, and references therein) for more detailed description.

In our simulations, gas follows a molecular-atomic-ionized cooling curve from 10 to $10^{10}$ K, including metallicity-dependent fine
structure and molecular cooling at low temperatures and high-
temperature metal-line cooling followed species-by-species for 11
separately tracked species (H, He, C, N, O, Ne, Mg, Si, S, Ca
and Fe; see Wiersma, Schaye & Smith 2009a). At each timestep,
the ionization states and cooling rates are determined following
Katz, Weinberg & Hernquist (1996) for primordial abundances and
from a compilation of CLOUDY runs for metals, including a uni-
form but redshift-dependent photo-ionizing background tabulated
in Faucher-Giguère et al. (2009), and photo-ionizing and photoelec-
tric heating from local sources. Gas self-shielding is accounted for
with a local Jeans-length approximation.

We allow star formation to take place only in dense, molecular
and self-gravitating regions with hydrogen number density above a
threshold $n_{\text{th}} = 5–50 \text{ cm}^{-3}$ (Hopkins, Narayanan & Murray 2013).
Stars form at 100 per cent efficiency per local free-fall time when the
gas meets these criteria and there is no star formation elsewhere.
A star particle inherits the metallicity of each tracked species from
its parent gas particle. Every star particle is treated as a single stel-
lar population with known mass, age and metallicity, assuming a
Kroupa (2002) initial mass function (IMF) from 0.1 to 100 $M_\odot$.
Then the ionizing photon budgets, luminosities, Type II supernova
rates, mechanical luminosities of stellar winds, etc. are directly tab-
ulated from the stellar population models in STARBURST99 (Leitherer
et al. 1999). The Type Ia SN rates follow the time delay distribu-
tion from Mannucci, Della Valle & Panagia (2006). We account for
the following stellar feedback mechanisms, including (1) local and
long-range momentum flux from radiative pressure, (2) energy, mo-
momentum, mass and metal injection from SNe and stellar winds and
(3) photo-ionization and photoelectric heating. We follow Wiersma
et al. (2009b) and account for metal production from Type II SNe,
Type Ia SNe and stellar winds using the metal yields in Woosley &
Weaver (1995), Iwamoto et al. (1999) and Izzard et al. (2004),
respectively. We do not include a sub-resolution metal diffusion
model, but the simulations explicitly resolve the metal mixing by
advection of gas particles.

2.2 Galaxy identification and definitions

We use Amiga’s Halo Finder (AHF; Knollmann & Knebe 2009) to
identify haloes in the simulations. The approximate halo mass at
$z = 2$ and $z = 0$ (if applicable) for the most massive (best-resolved)
halo in each simulation are listed in Table 1, where we adopt the
redshift-dependent virial parameter from Bryan & Norman (1998).
In this paper, we only study the central galaxy in the most massive
halo in each simulation. The entire simulated sample is only studied
at four redshifts $z = 2, 1.4, 0.8$ and 0 (if applicable). The physical
properties of these galaxies (as described below) at these redshifts
are presented in Appendix A.

We define the centre of each galaxy by iteratively finding the
geometric centre of all star particles within a sphere of decreasing
radius from 20 to 1 kpc. This generally corresponds closely to the
location of maximum stellar mass density. The stellar mass ($M_*$)
and the SFR for the central galaxy are measured within 10 kpc from
this centre, where we remove the contamination of satellite galax-
ies if necessary. The SFRs are averaged over 200 Myr to mimic the
observational measurements based on far-ultraviolet luminosity
(e.g. Sparre et al. 2015). Next, we define a characteristic radius $R_0$,
which encloses 90 per cent of the star formation within 10 kpc.
Such definition of galactic centre and characteristic radius appears
to be most numerically stable, given that a considerable fraction of
galaxies in our simulated sample have clumpy and irregular mor-
phologies (especially those at high redshifts). The stellar mass, SFR,
and $R_0$ for the entire simulated sample are listed in Appendix A.
Our sample covers a stellar mass range $10^8–10^{11} M_\odot$.

For simplicity, we define the $z$-axis to be aligned with the to-
tal angular momentum of all gas particles within $R_0$ and the $x$-
axis to be an arbitrary direction perpendicular to $z$-axis. We refer
to face-on and edge-on views when observing along the $z$- and
$x$-axis, respectively. In Fig. 1 (left two columns), we show example
images for three galaxies in our sample, A2:0 at $z = 2$ (top), A8:0
at $z = 2$ (middle) and m12i at $z = 0$ (bottom). For each galaxy,
we show a face-on gas image ($x$--$y$ plane, top left) and edge-on
gas image ($y$--$z$ plane, top right), face-on stellar image (bottom left)
and face-on SFR map (bottom right, averaged over 200 Myr). The
dashed white circles on all face-on images show the characteristic
$R_0$ of each galaxy. A8:0 is a merging system that has clumpy, ir-
regular morphology, while A2:0 and m12i have star-forming gas
disks.

2.3 Kinematics

Before we present the gas-phase metallicity gradients for our sim-
ulated sample, we first measure the kinematic properties of these
galaxies, as commonly done in observational studies (e.g. Yuan
et al. 2011; Jones et al. 2013; Leethochawalit et al. 2016). We
do so by mimicking the widely used long-slit spectroscopy tech-
nique. The mock slit is placed along the $y$-axis (edge-on) along the
mid-plane with a vertical width of 1 kpc, as illustrated by the
black lines on the edge-on gas images in Fig. 1. We then extract
the one-dimensional velocity curve along the slit. We measure the
line-of-sight gas velocity and 1D velocity dispersion in the range
$−R_0 < y < R_0$ with a spatial resolution of $\Delta y = 0.4$ kpc, by tak-
ing into account all gas particles with number density $n > 1 \text{ cm}^{-3}$
in every pixel. This allows us to primarily select interstellar gas and
eliminate contamination by foreground/background gas in the cir-
cumgalactic/intergalactic medium. Example velocity curves of the
eight galaxies, A2:0, A8:0 (at $z = 2$) and m12i (at $z = 0$), are shown
in the right column of Fig. 1, with the black points and errorbars
representing the line-of-sight velocity and velocity dispersion along
the slit.

We fit the one-dimensional velocity curve with the following
analytic form

$$V(R) = V_0 + V_z \frac{2}{\pi} \arctan \left( \frac{R}{R_0} \right),$$

as motivated by the simple disc model commonly adopted in vari-
ous studies (e.g. Jones et al. 2010; Swinbank et al. 2012; Stott
et al. 2014; Leethochawalit et al. 2016). For our simulated galaxies,
$V_0$ accounts for the peculiar velocity in the simulation box and $V_z$
gives the asymptotic circular velocity at large radii. Example fits
for the three galaxies are shown by the red lines in Fig. 1. The
velocity curves of A2:0 and m12i can be well described by the
arctan function, reaffirming that these galaxies have well-ordered
rotating discs. However, the chaotic system, A8:0, returns a bad fit
(as reflected by unphysical values of $V_z$). We have visually checked
all of our simulations and find that bad fits occur when a galaxy has
clumpy, irregular morphology and shows little evidence of rotation.
For these galaxies, $V_z$ cannot be properly defined. We also follow
Leethochawalit et al. (2016) and measure the ‘peak-to-peak’ velo-
city difference $\Delta V$ along the slit. Any galaxy can give a finite $\Delta V$
despite its kinematic properties. For a rotating disc, $\Delta V$ equals $2V_c$
in the asymptotic limit and is thus a proxy for the rotation velocity.
We define the velocity dispersion of the galaxy $\sigma$ as the max-
imum velocity dispersion along the slit. $V_c$, $\Delta V$ and $\sigma$ for the entire

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Figure 1. Left: example images from our simulated sample, including face-on gas surface density (upper left), edge-on gas surface density (upper right), stellar surface density (lower left) and star formation surface density (lower right). We show A2:0 (top) and A8:0 (middle) at $z = 2$ and m12i (bottom) at $z = 0$ (see Table 1 for details). The white circles show $R_{90}$ as defined in Section 2.3. The black lines on the edge-on gas images show the long slits where we extract the gas velocity curve. Right: velocity curve extracted from the slit. The symbols and errorbars show the line-of-sight velocity and velocity dispersion, respectively. The red lines show the best fit from the arctan function given by equation (1). A2:0 and m12i have well-ordered rotating disc, while A8:0 is a merging system with no evidence of rotation.
Comparison between or VV⊙/ΔV/2σ are shown. We measure 4780–4794 (2017) is the peak-to-peak velocity V2013 3 V are broadly consistent with each other and more quan-σ ≥ ∼ 1/Δ1o r and is the maximum velocity dispersion (see Fig. 1h ave 3/Σ1); for our simulated galax-
ies. Galaxies that cannot be well fitted by an arctan function are plotted at Vc/σ ∼ 0.1. Vc/σ and ΔV/2σ are broadly consistent with each other for galaxies with Vc/σ ≥ 1, indicating that they have well-ordered rotation by either criterion. However, galaxies with Vc/σ < 1 show ΔV/2σ ∼ 0.4–3. This suggests that ΔV/2σ is ambiguous for non-rotationally sup-
supported systems.

simulated sample are presented in Appendix A. Note that some galaxies in our sample are temporarily quenched, with little gas in the central region. The kinematic properties for these galaxies cannot be properly determined.

The degree of rotational support of a galaxy can be defined as either Vc/σ or ΔV/2σ. In Fig. 2, we compare Vc/σ and ΔV/2σ for our simulated galaxies. For illustrative purposes, we plot those whose velocity curve cannot be well fitted by equation (1) at Vc/σ ∼ 0.1, as they do not have well-ordered rotation. The criterion for rotationally supported system is commonly taken to be Vc/σ ≥ 1 or ΔV/2σ ≥ 0.4 (e.g. Förster Schreiber et al. 2009; Lee+ch+alt et al. 2016). Most of our simulated galaxies with Vc/σ ≥ 1 have consistent values of ΔV/2σ, reaffirming that these galaxies are rotationally supported. However, galaxies with Vc/σ < 1 span a wide range of ΔV/2σ, mostly from 0.4 to 3 for our simulated sample. These galaxies have little evidence of rotation as shown by the velocity curve and confirmed by our visual inspection, but they would be classified as rotating systems by the criterion ΔV/2σ ≥ 0.4. We caution that ΔV/2σ is an ambiguous indicator in practice, especially for those galaxies with little rotation.

2.4 Metallicity gradients

We now present the metallicity gradients for our simulated sample. In the top panel of Fig. 3, we show the face-on metallicity map for the same galaxies as in Fig. 1, with a pixel size of 100 pc. We measure the mass-weighted metallicity of all gas particles in each pixel. We only show pixels where the gas surface density is above Σg ≥ 10 M⊙ pc−2. Such threshold is roughly the surface density above which fragmentation and star formation occurs in these simulations and observations (Orr et al. 2017), so these pixels are likely to have observationally detectable nebular emission lines. This also reduces the shot noise from low surface density pixels where the metallicities are determined by individual gas particles. In the bottom panels, we plot the gas-phase metallicity as a function of projected radius for individual pixels (grey points). Again, only pixels above surface density 10 M⊙ pc−2 are shown. We measure the median metallicity and its 1σ dispersion at every 0.2 kpc in a certain radius interval, as illustrated by the red points and errorbars (in 0.25–1R90, our fiducial interval) in Fig. 3. We require a minimum number of 20 pixels in a 0.2 kpc bin to obtain a reliable measurement at this radius. We fit the metallicity profile with a linear function (the blue dotted lines in Fig. 3)

\[
\log(Z/Z⊙) = \alpha R + \beta, \tag{2}
\]
to obtain the slope of the metallicity gradient \(\alpha\) (in dex kpc−1).

Equation (2) is motivated by the fact that metallicity gradients are most commonly measured in d log Z/dR (in dex kpc−1) in the literature, although the metallicity profile of a galaxy may deviate from a linear function in reality. In Fig. 4, we compare the slopes of the metallicity gradients measured over 0.25–1R90 and over 0–2 kpc, respectively. Both slopes are qualitatively consistent with each other. The difference is small when the gradient is close to flat, because the metals are nearly uniformly mixed within the ISM (e.g. simulation A8:0). On the other hand, galaxies with strong negative metallicity gradients tend to have a rapidly increasing metallicity profile towards the centre (e.g. simulations A2:0 and m12i in Figs 1 and 3), as reflected by the fact that the slopes measured in 0–2 kpc are much steeper than those measured in 0.25–1R90. This happens in our simulations because the galactic centres can reach very high gas surface densities (Σg ≥ 10 M⊙ pc−2) during a starburst, and the star formation efficiency may increase dramatically with gas surface density (e.g. Burkert & Hartmann 2013; Torrey et al. 2016), resulting in rapid metal enrichment towards the centre. Such picture is consistent with previous studies on the formation of cusp elliptical galaxies via mergers (e.g. Hopkins et al. 2009), which reproduce the observed steep metallicity gradients in the central region of early-type galaxies (e.g. Reda et al. 2007; Sánchez-Blázquez et al. 2007). In this work, we primarily focus on the metallicity gradients measured over 0.25–1R90. The slopes of metallicity gradient for the entire simulated sample are listed in Appendix A. We note that all of our results presented below are qualitatively consistent if one uses the gradients measured in 0–2 kpc. A detailed discussion on the full metallicity profile is beyond the scope of this study, but worth further investigations in future work.

3 RESULTS

3.1 Metallicity gradients: general properties

As illustrated by the visual examples in Fig. 3 and more quantita-
tive results shown in Appendix A, our simulations produce a variety of kinematic properties and metallicity distributions. Simu-
lations A2:0 and m12i have obvious negative metallicity gradients, with the centre being more metal enriched than the outskirts, consistent with the observed metallicity patterns in local and some high-redshift galaxies (e.g. Zaritsky et al. 1994; van Zee et al. 1998; Yuan et al. 2011; Jones et al. 2013; Sánchez et al. 2014). Both of them have a rotationally supported, star-forming disc as shown in Fig. 1. In contrast, simulation A8:0 is a merging system that has a clumpy,
irregular gas morphology with no well-ordered gas motion, and a relatively uniform metallicity distribution, with metallicity gradient close to flat. Intuitively, these examples indicate that strong negative metallicity gradients are more likely to occur in galaxies with a rotating disc, while strongly perturbed galaxies tend to have flat gradients.

Strong perturbations, mostly induced by mergers, rapid gas infall and strong outflows, can stir the gas and drive galactic-scale motion in the ISM, with typical velocities up to several hundred km s\(^{-1}\). This causes gas/metal re-distribution on galactic scales of \(\lesssim 10\) kpc on relatively short time-scales \(\sim 10–50\) Myr, leading to kinematically hot gas motion and flat metallicity gradients.\(^2\) In simulation A8:0, the perturbation is induced by a series of minor mergers (see Fig. 1). Besides, strong stellar feedback can also drive galaxy-scale motion in the ISM, resulting in irregular gas motion and morphology (e.g. Agertz & Kravtsov 2016). Gibson et al. (2013) show that simulations with strong feedback produce flat metallicity gradients, while those with weak feedback tend to produce steep gradients. The high resolution and physically motivated models of stellar feedback adopted in the FIRE simulations enable us to explicitly resolve the launch and propagation of galactic winds from small scales (tens of pc) to galactic scales, which is essential to study gas-phase metallicity gradients.

The rest of this section is organized as follows. Before going into details about metallicity gradients in our simulated galaxies, we first show where our sample lies on the galaxy MZR in Section 3.2. In Section 3.3, we will study the dependence of metallicity gradient on stellar mass and specific star formation rate (sSFR). In Section 3.4, we will examine the relation between metallicity gradient and the degree of rotational support. In Section 3.5, we will present the redshift dependence on metallicity gradient. In Section 3.6, we will perform a case study on simulation m12i and explore how stellar

\(^2\) Here we do not consider metal mixing on scales below our resolution limit, but rather focus on re-distribution of metals driven by largest scale motion. This is justified by more detailed studies of diffusion processes in supersonically turbulent media like the ISM, which show that diffusion is most efficient on large scales (e.g. Colbrook et al. 2016).
feedback can change metallicity gradients on short time-scales (≲ Gyr), which has a great effect on the interpretations of the observed metallicity gradients in high-redshift galaxies.

3.2 The mass–metallicity relation (MZR)

We follow Ma et al. (2016b) and define the gas-phase metallicity as mass-weighted mean metallicity of all gas particles below 10^4 K in the central galaxy (satellites excluded). In Fig. 5, we show the gas-phase MZR for our simulated sample at z = 2, where we define the oxygen abundance as 12 + log(O/H) = log(Z/Z⊙) + 9.0. Galaxies A2:0 and A8:0 shown in Figs 1 and 3 are indicated with thick cyan and red circles, respectively. They have typical gas-phase metallicities for our sample. In Ma et al. (2016b), we extensively studied the MZR in a sample of FIRE simulated galaxies at z = 1.4, 0.8 and 0. In that work, we showed that m12i lies on the observed median gas-phase and stellar MZR from Tremonti et al. (2004) and Gallazzi et al. (2005) at z = 0. The blue dashed line shows the linear fit to the simulations from Ma et al. (2016b). We note that Ma et al. (2016b) used a sample that covered the stellar mass range from 10^4 to 10^{10} M⊙ at z = 2, while the new simulations from Feldmann et al. (2016b) included in this work allow us to extend our analysis to 10^{11} M⊙.

3.3 Metallicity gradient versus stellar mass and sSFR

We start by examining the correlation between gas-phase metallicity gradient (measured over 0.25–1R⊙) and galaxy properties. In Fig. 6, we show the dependence of metallicity gradient on stellar mass (left) and sSFR (right) for the simulated sample at four redshifts z = 2.0, 1.4, 0.8 and 0. We do not find significant differences between redshifts except perhaps for massive galaxies at z ~ 0, consistent with recent observations (e.g. Wuyts et al. 2016). The shaded regions show 2σ linear fits to the simulated data. We find a weak anticorrelation between metallicity gradient and stellar mass. Low-mass galaxies tend to have flat gradients, because feedback is very efficient in driving outflows and thus mixing metals in low-mass systems (Muratov et al. 2015, 2016). Note that the FIRE project also includes simulations of isolated dwarf galaxies with stellar masses M∗ ~ 10^2–10^4 M⊙ (e.g. Hopkins et al. 2014; Chan et al. 2015), but we do not consider these dwarfs in this work, because observations probe only galaxies more massive than 10^9 M⊙. Nevertheless, they also have very weak (flat) metallicity gradients (El-Badry et al. 2016), because they are bursty, feedback-dominated galaxies, consistent with the argument above. We also find a weak correlation between metallicity and sSFR. Most galaxies with high sSFR have undergone rapid gas inflows that trigger starbursts, and feedback in turn drives strong outflows. Such violent gas inflows and outflows can stir the gas in the ISM and mix metals on galactic scales efficiently, resulting in a flat metallicity gradient. In Fig. 6, we also show the linear fits to a compilation of observational data at redshifts z = 0–2.5 from Stott et al. (2014, blue dashed lines). These trends are in qualitative agreement with our simulations, but we note that both observations and our simulations only show weak trends with stellar mass and sSFR (within 3σ, the data are consistent with no trend).

3.4 Metallicity gradient versus kinematic properties

In the left-hand panel of Fig. 7, we show the relation between gas-phase metallicity gradient (measured over 0.25–1R⊙) and degree of rotational support, Vc/σ, for the entire simulated sample. Again, galaxies whose Vc cannot be properly determined are plotted at Vc/σ ~ 0.1. In general, our simulated sample can be divided into three populations that occupy three different regions on the α–Vc/σ diagram: (1) significant negative metallicity gradients only occur in galaxies with rotationally supported discs (Vc/σ ≥ 1), (2) strongly perturbed galaxies, with no evidence of rotation (Vc/σ < 1, including those with undetermined Vc), tend to have flat metallicity gradients and (3) there is also a population that show flat or mildly positive metallicity gradients (α ~ 0) while being rotationally supported (Vc/σ ≥ 1). The existence of population (3) reflects the observed complex relation between metallicity gradient and galaxy kinematics (e.g. Jones et al. 2015; Leethochawalit et al. 2016). We emphasize that our sample only predicts that galaxies with a strong negative metallicity gradient must be rotationally supported, but not vice versa. We do not find any galaxy in our simulated sample that has a significant negative metallicity gradient (α < −0.05 dex kpc^{-1}) but is strongly perturbed (Vc/σ < 1).

The connection between negative metallicity gradients and rotating discs can be understood from the coevolution of the gas disc and stellar disc (e.g. Ho et al. 2015). A simple toy model is useful for illustrative purposes. Start from a pristine gas disc with an exponential surface density profile ΣG ∝ exp (−R/R⊙), where R⊙ is the disc scalelength. Stars form in the disc at higher efficiencies in regions with higher surface densities, following the Kennicutt–Schmidt law ΣG ∝ Σ^1.4 ∝ exp (−1.4R/R⊙) (Kennicutt 1998). If the metals do not mix efficiently between annuli (i.e. the local ‘closed-box’ assumption), the gas-phase metallicity is ZG ∝ −ln(1 - fα), where fα is the mass fraction of stars (note that both fα and ZG are functions of radius). If the gas fraction is not too low, ZG ∝ Z⊙/Σ/G ∝ Σ/Σ ∝ exp (−0.4R/R⊙). This naturally gives a negative metallicity gradient dlogZ/G/dR = −0.17/R⊙ dex kpc^{-1} (if R⊙ is in kpc), although the slope can be altered by the exact disc surface density profile, pre-enrichment in the disc, the strength of the power-law surface density profile ΣG ∝ R^{−β}, where β > 0 is the power-law index, following the same argument above, the gas-phase metallicity profile will be ZG ∝ R^{−0.17}. A power-law profile might
Figure 6. Left: metallicity gradient (measured over 0.25–1R\textsubscript{90}) versus stellar mass. Right: metallicity gradient versus sSFR. The shaded regions show the 2\sigma interval of the linear fit to the simulated data. The blue dashed lines show the linear fit to a compilation of observational data at z = 0–2.5 from Stott et al. (2014). There is weak dependence of metallicity gradient on both stellar mass and sSFR. Low-mass galaxies or those with high sSFR tend to have flat metallicity gradients, due to the fact that feedback is more efficient in such galaxies.

Figure 7. Metallicity gradient (measured over 0.25–1R\textsubscript{90}) versus degree of rotational support. Left: \(\alpha-V_c/\sigma\). Symbol with errorbars show observational data from Yuan et al. (2011, Y11), Swinbank et al. (2012, S12), Jones et al. (2013, J13) and Leethochawalit et al. (2016, L16). Our simulations reproduce the observed complexity in metallicity gradient and kinematic properties.

radial mixing, etc. Population (2) galaxies are strongly perturbed via violent processes, such as mergers, rapid gas inflows and strong feedback-driven outflows, which can destroy any pre-existing rotating disc and cause efficient gas re-distribution on galactic scales. Galaxies in region (3) may be in a transition phase, e.g. during a gas infall before a strong negative metallicity gradient builds up at a later time. In Section 3.6, we will further show that the metallicity gradient and kinematic properties of a galaxy can vary on \(<\)2Gyr time-scales, causing the galaxy to move across the three regions on the \(\alpha-V_c/\sigma\) relation.

In the right-hand panel of Fig. 7, we show the relation between metallicity gradient and \(\Delta V/2\sigma\). Similarly, strong negative metallicity gradients only appear in galaxies with \(\Delta V/2\sigma \geq 1\). Symbols with errorbars show observational data from Yuan et al. (2011, Y11), Swinbank et al. (2012, S12), Jones et al. (2013, J13) and Leethochawalit et al. (2016, L16). Note that we follow Leethochawalit et al. (2016) and only adopt the \(V_c/\sigma\) for those that be a better description to our simulations (e.g. Hopkins et al. 2009) and the observed metallicity profiles in early-type galaxies (e.g. Reda et al. 2007; Sánchez-Blázquez et al. 2007). In such case, the slope of metallicity gradients, if defined in d\(\log Z_g/\text{dR}\) (in dex kpc\(^{-1}\)), also depends on the range where the gradient is measured. This may account for the steep metallicity gradients (\(<\sim 0.3\) dex kpc\(^{-1}\)) observed in high-redshift galaxies (e.g. Jones et al. 2013, also see Fig. 8).
Metallicity gradient versus redshift. The black points show the metallicity gradients measured in 0.25–1$R_0$ for the entire FIRE sample at four redshifts. The smaller grey points show the slopes measured in 0–2 kpc. The grey points are shifted slightly right along the x-axis for better illustration. Symbols with errorbars show a compilation of observations from Maciel et al. (2003, M03), Yuan et al. (2011, Y11), Swinbank et al. (2012, S12), Jones et al. (2013, J13), Jones et al. (2015, J15), Leethochawalit et al. (2016, L16) and Wang et al. (2016, W16). The green lines show the predictions from the sub-grid ‘conservative’ (weak) feedback model used in MUGS simulations (dashed) and the ‘enhanced’ (strong) feedback used in MAGICC simulations (dotted) from Gibson et al. (2013). Our simulations agree well with the wide range of metallicity gradients observed over the $z = 0–2.5$ redshift range – in some circumstances (e.g. starbursts), feedback is predicted to be effectively ‘strong’ to produce flatter metallicity gradients, while in others, it is sufficiently ‘weak’ to allow a strong negative gradient. These can be reliably fitted by a simple disc model ($x_\text{fed}^2 < 20$ in their table 3), while we regard the rest of their sample as non-rotationally supported ($V_c$ undetermined). Our simulations reproduce the observed complexity in the relationship between metallicity gradient and kinematic properties. Remarkably, the simulated sample and the observed sample, although both small in sample size, occupy almost identical parameter space in these relations.

### 3.5 Metallicity gradient versus redshift

In Fig. 8, we plot the metallicity gradients for all simulated galaxies in our sample as a function of redshift, at $z = 2, 1.4, 0.8$ and 0. The black points present the metallicity gradients measured from 0.25 to 1$R_0$. We also compare a variety of observations from Maciel et al. (2003, M03), Yuan et al. (2011, Y11), Swinbank et al. (2012, S12), Jones et al. (2013, J13), Jones et al. (2015, J15), Leethochawalit et al. (2016, L16) and Wang et al. (2016, W16). Our simulations are broadly consistent with the observed diversity of metallicity gradients at redshifts $z = 0.5–2.5$. For example, at $z \approx 2$, our sample covers metallicity gradients from $\alpha = -0.15–0.05$ dex kpc$^{-1}$, in reasonably good agreement with observational data at that epoch. Note that we measure the metallicity gradient from 0.25 to 1$R_0$ by default, whereas there is no universal standard for the radial limits used to define the metallicity gradients in observations. If we instead use the metallicity gradient in the central 0–2 kpc in our simulations, as shown by the small grey points in Fig. 8, we obtain a similar result, but with somewhat larger scatter, with the slope ranging from $-0.3$ to 0.1 dex kpc$^{-1}$. This is in better agreement with the steep slopes and positive metallicity gradients in some of the observational samples (e.g. Jones et al. 2013; Leethochawalit et al. 2016). A more rigorous comparison would require matching precisely the galaxy selection function and observational metallicity gradient measurement method of each observed sample, which is beyond the scope of this paper.

We also compare our results with the MUGS simulation (‘conservative’ feedback) and the MAGICC simulation (‘enhanced’ feedback) from Gibson et al. (2013). In the ‘enhanced’ feedback model, gas heated by SNe is kept hot artificially for much longer than the Sedov–Taylor phase to generate efficient outflows (Stinson et al. 2013), in contrast to much simpler ‘sub-grid’ models that effectively suppress bursty star formation. These feedback models also require fine-tuning certain parameters to match the observed galaxy properties. The ‘conservative’ (weak) feedback model in Gibson et al. (2013) always predicts the so-called ‘inside-out’ growth picture. In this scenario, a compact core formed rapidly at the centre of the galaxy, building up a steep negative metallicity gradient at high redshift. Then the galaxy gradually grows in size and the metallicity gradient flattens as the galaxy evolves. Their ‘enhanced’ (strong) feedback model, on the other hand, always produces a flat metallicity gradient that shows little evolution with redshift. In contrast, our sample produces more diverse distribution of metallicity gradients in good agreement with observations, including both strong negative gradients and flat/weak positive gradients. This confirms that metallicity gradients in cosmological simulations are sensitive to the treatment of feedback. The physics adopted in FIRE explicitly resolves feedback processes on sub-kpc scales that allows galaxies to ‘switch’ between weak and strong outflows based on their local conditions. As a consequence, our simulations produce both strong and weak gradients, even in the same galaxy at slightly different times in its evolution. This leads to a diversity of gradients in good agreement with observations, and in contrast to simpler ‘sub-grid’ feedback models.

### 3.6 The effects of feedback: a case study

In this section, we will show how feedback results in the complex relation between galaxy gas-phase metallicity gradients and kinematic properties. To this end, we perform a case study on simulation m12i, which produces a MW mass disc galaxy by m12i. We follow Faucher-Giguère, Kereš & Ma (2011) and Muratov et al. (2015) and calculate the gas outflow rate as

$$\frac{\partial M}{\partial t} = \frac{1}{L} \sum_i m_i v_i \cdot \mathbf{r}_i / |\mathbf{r}_i|,$$

where we sum over all gas particles that have radial velocity $v_i = \mathbf{v} \cdot \mathbf{r}/|\mathbf{r}| > 100$ km s$^{-1}$ within the central $L = 10$ kpc in the galaxy.

At $z > 0.7$, both the gas outflow rate and SFR show significant time variability. The outflow rates are much higher than the SFRs.

Note that the SFRs shown here are different from those defined in Section 2.2 and listed in Appendix A (where the SFRs are averaged over 200 Myr), because we want to emphasize the short-time-scale fluctuations in this section.
Figure 9. *Top:* metallicity gradient in the galaxy m12i (measured from 0.25 to 1\(R_{\text{eff}}\)) as a function of cosmic time at redshifts \(z = 0–1.1\) (black solid). The SFR (red dotted) and gas outflow rate measured at 10 kpc (blue dashed) are also shown for comparison. *Middle:* gas morphology at the four epochs labelled by the vertical grey dotted lines in the top panel (a–d). *Bottom:* metallicity map at the four epochs. At \(z > 0.7\), the metallicity gradient shows considerable time fluctuations, associated with starburst episodes. The examples illustrate this process: (a) gas flows in rapidly and forms a disc, (b) a negative metallicity gradient builds up during star formation, (c) strong feedback from starburst drives intense gas outflow and flattens the metallicity gradient and (d) gas falls back and reforms a disc. The peaks in gas outflow rate match the ‘peaks’ in metallicity gradients (where the gradients are flat). This explicitly shows the effect of feedback flattening the metallicity gradient. At \(z < 0.7\), the disc has ‘calmed down’, and stellar feedback is no longer strong enough to disrupt the gas disc. A negative metallicity gradient then develops rapidly, and does not evolve significantly with time after this.

(high mass loading factors), implying that feedback is very efficient at these times (Muratov et al. 2015). At the same time, the metallicity gradient also shows significant fluctuations. Interestingly, the peaks in gas outflow rates coincide with the ‘peaks’ in metallicity gradients (i.e. when the gradient is flat, since a strong gradient has a negative slope). To further illustrate the process, we show example gas images and metallicity maps in the middle and bottom panels in Fig. 9, respectively, at four selected times labelled by (a)–(d), as shown by the grey vertical dotted lines in the top panel of Fig. 9. First, gas flows in rapidly and forms a rotating gas disc (a). Rapid gas infall triggers a starburst in the disc, and a negative metallicity gradient builds up quickly (b, see the argument in Section 3.4). Next, feedback from the starburst drives strong outflows, which destroy the gas disc and mix the metals on galactic scales, flattening the pre-existing negative metallicity gradient in the disc (c). Finally, gas falls back, reforming a disc and the next episode starts (d).

We repeat the analysis in Section 3.4 and measure the degree of rotational support \(V_c/\sigma\) for 50 successive snapshots from simulation m12i, from \(z = 0.6–1.1\), before the metallicity gradient becomes stable. In Fig. 10, we plot the relation between metallicity gradient and \(V_c/\sigma\) for the 50 epochs considered here (blue circles) and compare the results with the entire FIRE sample as shown in Fig. 7 (grey points). Remarkably, the time variability of a single galaxy occupies almost identical parameter space as the entire simulated sample in the \(\alpha–V_c/\sigma\) relation. Again, significant negative metallicity gradients only appear when there is a well-ordered rotating disc, while the gradients are flat when the galaxy is strongly perturbed and shows little rotation. At the epochs when the galaxy has a flat metallicity gradient but is rotationally supported, it is mostly in the early stage of gas infall before a strong metallicity gradient builds up later [e.g. epoch (a) shown in Fig. 9]. These results suggest that

\[\text{Note that while the outflow rates in Fig. 9 are qualitatively similar to those in Muratov et al. (2015), they are different quantitatively because of different radial and velocity range considered.}\]
a single galaxy can rapidly (in a few 100 Myr) traverse the range of observed metallicity gradients and kinematic properties, indicating that the observed metallicity gradients at high redshifts may be more of an indicator of the instantaneous (\textless;Gyr time-scale) dynamical state of the galaxy, not the long-term galaxy formation, accretion or growth history.

Almost all the simulated galaxies show significant burstiness in SFR and undergo strong bursts of feedback-driven outflows at high redshift (z \gtrsim 0.5), even for the most massive galaxies at z \sim 2 (Hopkins et al. 2014; Sparre et al. 2015; Muratov et al. 2015). The central galaxy in simulation m12i calms down after z \sim 0.7, and there is always a well-ordered, rotationally supported gas disc thereafter (Ma et al. 2016a). Stars form in the disc at a nearly constant rate that is set by the nearly constant gas accretion rate and regulated by stellar feedback. The feedback is no longer sufficient to drive strong gas outflows and destroy the gas disc. A negative metallicity gradient builds up quickly as soon as the disc calms down and stays almost unchanged after this time. A similar transition is also seen in other simulations that produce a galaxy more massive than $M_* = 10^{10} M_\odot$ by z = 0, as these galaxies also cannot drive strong gas outflows at late times (Muratov et al. 2015). Such a transition is likely due to a combination of decreasing merger rates at lower redshifts (e.g. Hopkins et al. 2010) and decreasing gas fractions in massive galaxies (Hayward & Hopkins 2017). Therefore, it is expected that massive galaxies in the local Universe mostly have stable negative metallicity gradients, except for strongly perturbed (e.g. merging) galaxies.

\section{Discussion and Conclusions}

In this paper, we use 32 high-resolution cosmological zoom-in simulations from the FIRE project to study the gas-phase metallicity gradient in galaxies and its relation with galaxy properties. Our simulated sample includes 32 galaxies at z = 2, covering a halo mass range $10^{11}$–$10^{13} M_\odot$ and stellar mass range $10^{9}$–$10^{11} M_\odot$. A sub-sample has been run to $z = 0$, spanning a halo mass range $10^{11}$–$10^{13} M_\odot$ and stellar mass range $10^{9}$–$10^{11} M_\odot$ at $z = 0$. The FIRE simulations include physically motivated models of the multiphase ISM, star formation and stellar feedback and have been shown to reproduce a number of observed properties of galaxies for a broad range of stellar mass at redshift $z = 0$–6. These simulations explicitly resolve the launching and propagation of galactic winds on sub-kpc scales and can thus capture the effects of stellar feedback on metallicity gradients.

(i) The simulations produce a diverse range of kinematic properties and metallicity gradients, broadly consistent with observations at all redshifts. Our simulated sample includes merging galaxies, starbursts with gas morphologies disturbed by feedback, as well as relatively stable, rotation-dominated disc galaxies.

(ii) Strong negative metallicity gradients only appear in galaxies with a gas disc, as reflected by well-ordered rotation ($V_c/\sigma \geq 1$), while strongly perturbed galaxies ($V_c/\sigma \leq 1$) always have flat gradients. In a gas disc, the star formation efficiency is higher towards the centre due to increasing gas surface density, so metal enrichment is faster in the central region, leading to a negative metallicity gradient. Strong perturbations driven by rapid gas infall, mergers or violent outflows, can stir the gas in the ISM, causing metal redistribution on galactic scales and flattening metallicity gradients. Not all rotationally supported galaxies have strong negative metallicity gradients.

(iii) The metallicity gradient and kinematic properties of a high-redshift galaxy can vary on $\lesssim$Gyr time-scales, associated with starburst episodes. The time variability of a single galaxy is statistically similar to the overall simulated sample. A negative metallicity gradient can build up quickly as a starburst is triggered in a gas disc formed via gas infall. Strong feedback from the starburst drives intense outflows, which destroy the gas disc and cause metal redistribution on galactic scales, resulting in flat metallicity gradients. Gas recycles in fountains (Angles-Alcazar et al. 2016), and negative gradients may re-establish quickly. This has important consequences for the interpretation of metallicity gradients observed in high-redshift galaxies. They may not well correlate with the accretion or growth history of the galaxy on cosmological time-scales, but rather reflect the ‘instantaneous’ state of gas dynamics.

(iv) There is weak dependence of metallicity gradient on both stellar mass and sSFR. Low-mass galaxies and/or galaxies with high sSFR tend to have flat metallicity gradients, owing to efficient feedback in such systems, which keeps them in the ‘bursty’ star formation mode.

(v) Because of the important role of stellar feedback, it is essential to resolve feedback from sub-kpc to galactic scales in sufficiently high-resolution simulations, to reproduce the observed diversity of kinematic properties and metallicity gradients in high-redshift galaxies. Our results are in contrast to simulations with simple ‘subgrid’ feedback models, which tend to predict either ‘all strong’ or ‘all weak’ metallicity gradients.

Our results suggest that the bursty star formation in our simulations can change the kinematic properties and gas-phase metallicity gradients in these galaxies on relatively short time-scales ($\sim 10^7$–$10^8$ yr), which can at least partly explain the diverse kinematics and gradients observed in high-redshift galaxies. One intriguing question we leave open is when and why a galaxy shows such bursty star formation. A detailed answer of this question may require a larger sample of simulations. None the less, the current sample of the FIRE simulations have suggested that at high redshift
(z > 2), all galaxies show significant burstiness in the SFR, even in the most massive galaxies in the simulated sample (Faucher-Giguère et al. 2015; Sparre et al. 2017; Feldmann et al. 2016a). At late times, low-mass galaxies (M* < 10^{10} M\odot) still have bursty star formation down to z ∼ 0 (El-Badry et al. 2016; Wheeler et al. 2017), while more massive galaxies (M* ≥ 10^{10} M\odot) tend to have a transition from bursty to relatively stable star formation at intermediate redshift (z ∼ 0.5–1, Muratov et al. 2015). Hayward & Hopkins (2017) provide an analytic model and argue that such transition happens at a gas fraction threshold of f_{gas} ∼ 0.3, above which the ISM is highly turbulent and star formation is sufficiently violent that feedback can efficiently blow out a large fraction of low-density gas from the disc. At lower gas fractions, turbulence becomes weaker, and feedback is no longer sufficient to drive strong outflows.

In our simulations, stellar metallicity gradients develop coherently with gas-phase metallicity gradients as stars form in the disc (also see the argument in Section 3.4), but stellar metallicity gradients are much less vulnerable to strong feedback than their gas-phase counterparts, especially in massive galaxies (El-Badry et al. 2016). Stellar migration in the disc can flatten metallicity gradients, but it may only have a weak net effect over a few Gyr time-scale (Ma et al. 2016a). Therefore, we propose that our predictions for the short-time-scale variation of gas-phase metallicity gradients can be tested with stellar metallicity gradients. One would expect that a large fraction of massive high-redshift galaxies have significant negative stellar metallicity gradients, even if they show a broad range of kinematic properties and gas-phase metallicity gradients. We say massive because the galaxy must have had a gas disc at some point to build up a stellar metallicity gradient, which is not the case in small dwarf galaxies. Negative stellar metallicity gradients have been observed in local galaxies (e.g. Sánchez-Blázquez et al. 2014), although it is challenging to measure stellar metallicities at higher redshifts. It will be interesting to study stellar metallicity gradients in these simulations in more details in future work.

Nevertheless, our simulations only have a moderate sample size and are limited in statistical power. We show in Section 3.4 that our simulated sample can be divided into three populations based on their kinematic properties and metallicity gradients, but we leave a number of open questions. What fractions of galaxies at a given redshift are rotationally supported and strongly perturbed, respectively? How often are strong perturbations driven by internal feedback versus external processes? What fraction of rotationally supported galaxies show strong negative gas-phase metallicity gradients? What fraction of galaxies in each population are associated with mergers? These questions are important for understanding high-redshift galaxy populations and worth further investigations, which we hope to explore with larger ensembles of simulations in the future.

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Gas-phase metallicity gradients in FIRE

Appendix A: Galaxy properties

In this section, we list the galaxy properties (stellar mass, star formation rate and $R_{90}$, Section 2.2), kinematic properties ($V_e$, $\Delta V/2$ and $\sigma$, Section 2.3) and gas-phase metallicity gradient measured in 0.25–$1R_{90}$ (Section 2.4) for the entire simulated sample studied in this paper (Figs 6–8). A machine-readable version of Table A1 is available at https://www.tapir.caltech.edu/~xchma/data/metal_grad.txt.
Galaxy properties studied in this paper (units are physical):
(1) Name: simulation designation.
(2) $z$ : redshift where the properties here are measured.
(3) $M_\ast$ : stellar mass within the central 10 kpc of the galaxy at the given redshift.
(4) SFR: star formation rate within the central 10 kpc of the galaxy (averaged over 200 Myr).
(5) $R_{90}$: defined in Section 2.2, as the radius that encloses 90 per cent of the stars younger than 200 Myr within 10 kpc.
(6) $V_r$: rotation velocity given by the arctan fit from equation (1) to the gas velocity curve (see Section 2.3).
(7) $\Delta V$: peak-to-peak velocity difference in the gas velocity curve (see Section 2.3).
(8) $\sigma$: maximum velocity dispersion of gas (see Section 2.3).
(9) $\alpha$ : gas-phase metallicity gradient measured over 0.25–1$R_{90}$ from equation (2).

Note. If a galaxy is temporarily quenched and near gas depletion in the central 10 kpc, its gas kinematic properties ($V_r$, $\Delta V/2$ and $\sigma$) and gas-phase metallicity gradient ($\alpha$) cannot be properly measured. If a galaxy has been quenched for more than 200 Myr, its SFR and $R_{90}$ are also not defined.

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