Probing the CGM of low-redshift dwarf galaxies using FIRE simulations

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ABSTRACT

Observations of ultraviolet (UV) metal absorption lines have provided insight into the structure and composition of the circumgalactic medium (CGM) around galaxies. We compare these observations with the low-redshift (z < 0.3) CGM around dwarf galaxies in high-resolution cosmological zoom-in runs in the FIRE-2 (Feedback In Realistic Environments) simulation suite. We select simulated galaxies that match the halo mass, stellar mass, and redshift of the observed samples. We produce absorption measurements using TRIDENT for UV transitions of CIV, OVI, MgII, and SiIII. The FIRE equivalent width (EW) distributions and covering fractions for the C IV ion are broadly consistent with observations inside $0.5R_{\rm vir}$, but are underpredicted for O VI, Mg II, and Si III. The absorption strengths of the ions in the CGM are moderately correlated with the masses and star formation activity of the galaxies. The correlation strengths increase with the ionization potential of the ions. The structure and composition of the gas from the simulations exhibit three zones around dwarf galaxies characterized by distinct ion column densities: the discy interstellar medium, the inner CGM (the wind-dominated regime), and the outer CGM (the IGM accretiondominated regime). We find that the outer CGM in the simulations is nearly but not quite supported by thermal pressure, so it is not in hydrostatic equilibrium, resulting in halo-scale bulk inflow and outflow motions. The net gas inflow rates are comparable to the star formation rate of the galaxy, but the bulk inflow and outflow rates are greater by an order of magnitude, with velocities comparable to the virial velocity of the halo. These roughly virial velocities ($\sim 100 \text{ km s}^{-1}$) produce large EWs in the simulations. This supports a picture for dwarf galaxies in which the dynamics of the CGM at large scales are coupled to the small-scale star formation activity near the centre of their haloes.

Key words: galaxies: dwarf-galaxies: evolution-quasars: absorption lines .

1 INTRODUCTION

The origin of the circumgalactic medium (CGM) is a product of two critical processes in galaxy evolution: gas inflow from the intergalactic medium (IGM) and gas outflow from the galaxy. Continuous gas inflow is required to sustain the star formation activity observed throughout cosmic time, while galactic outflows are necessary to explain the low observed star formation efficiency (Dekel et al. 2009; Silk & Mamon 2012). However, the CGM is a complex and dynamic environment because of the interaction between accretion and outflows, and hydrodynamic instabilities that occur in the process. Consequently, the study of the CGM is both challenging and crucial for our understanding of galaxy formation.

The CGM is diffuse and dim, and as a result, it has only recently been detected in emission (e.g. Martin et al. 2014, 2015). An

alternate probe is to use absorption spectra of bright background objects (e.g. quasars) to gain information on element abundances along sightlines. With recent observations probing the circumgalactic region (Tumlinson et al. 2011; Prochaska et al. 2011; Werk et al. 2014; Bordoloi et al. 2014; Liang & Chen 2014; Johnson, Chen & Mulchaey 2015; Johnson et al. 2017), we are provided with new constraints to test and inform our theoretical and computational modelling of the process of galaxy formation and evolution. By comparing the structure of the CGM in simulations with that seen in observations, we can test our understanding of galaxy formation. At the same time, because of the limitations of the observational probes, the simulations can provide an interpretation of the observations.

Small-scale simulations with sufficient resolution to resolve individual stars have identified non-linear interaction of the different feedback mechanisms as critical for the generation of outflows from stars. However, such extremely high-resolution simulations are restricted to small-scale systems such as single star clusters, molecular clouds (e.g. Harper-Clark & Murray 2011) or the 'first

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stars' (e.g. Wise et al. 2012). Current large-scale cosmological simulations (e.g. Vogelsberger et al. 2014a,b), still cannot resolve the details of the launching of stellar winds from stellar feedback and thus need to implement 'subgrid' prescriptions. The prescriptions are routinely tuned so that the simulations match the observational constraints, such as the stellar mass–halo mass function.

The FIRE (Feedback In Realistic Environments) project (Hopkins et al. 2014, 2018) includes a suite of high-resolution cosmological zoom-in simulations.¹ What distinguishes the simulations in the FIRE project from other cosmological simulations is the realistic feedback models used. The feedback mechanisms included in the simulation suite in this analysis are supernovae (Ia and II), continuous stellar mass-loss (OB/asymptotic giant branch, AGB-star winds), photoionization and photoelectric heating, radiation pressure, metal diffusion, and cosmic rays (CRs). The parameters in the feedback models are not tuned 'by hand' to match the observational constraints. Instead, the stellar feedback inputs are based entirely on stellar population models. It has already been shown in a series of papers that the FIRE simulations can reproduce a number of galaxy properties in observations at different redshifts, including star formation (e.g. Feldmann et al. 2016; Sparre et al. 2017; Orr et al. 2018), dynamical (e.g. Orr et al. 2019), and morphological properties (e.g. Ma et al. 2017; Wetzel et al. 2016; El-Badry et al. 2018). Results discussed later in this paper include comparison of FIRE-2 simulations with and without CRs to observations of the CGM around low-redshift $\sim L \star$ galaxies (Ji et al. 2019), particle tracking of inflows and outflows that build up galaxies (Anglés-Alcázar et al. 2017), and particle tracking of origins and fates of the CGM (Hafen et al. 2019a,b).

The FIRE feedback implementation makes clear predictions of the dynamics and thermal state of the CGM in the main halo of each simulation. The predictions of neutral hydrogen absorption in the CGM of the FIRE simulations have been presented in Faucher-Giguère et al. (2015, 2016) and Hafen et al. (2017), so here we concentrate on metal absorption lines.

In this work, our goal is to compare the structure of the CGM around low-redshift dwarf galaxies seen in the FIRE-2 simulations with the results of the available observations (Bordoloi et al. 2014; Liang & Chen 2014; Johnson et al. 2017). Following Bordoloi et al. (2014), who include galaxies with stellar masses up to $1.6 \times 10^{10} M_{\odot}$, we include galaxies up to $2.5 \times 10^{10} M_{\odot}$ (FIRE simulation 'm11f') to cover the observed range.

With shallower potential wells and higher gas fractions, dwarf galaxies are both predicted and observed (Martin 1998; Martin, Kobulnicky & Heckman 2002; Strickland & Stevens 2000; Summers et al. 2003, 2004; Hartwell et al. 2004; Ott, Walter & Brinks 2005) to have stronger outflows relative to their stellar mass or star formation rates (SFR). Because of the difference in the order of magnitude in mass, dwarf galaxies provide a critical testing ground for our theories of galaxy evolution and a great opportunity to understand the process of the cycling of baryons through galactic discs into the CGM and back over a broad range of halo masses. We leave comparison of high-z dwarf galaxies and predictions of the distributions of other ions for future work.

This paper is structured as follows. In Section 2, we describe the simulation suite and the observations used in this study. In Section 3, we describe the post-processing of the simulations for CGM analysis. In Section 4, we present the main results from the analysis of the simulation data, including the absorption EW distributions of four ions in the simulated sample and their comparison with observations

(Section 4.1), examination of the relationship between the equivalent width (EW) distributions of the ions in the simulations (Section 4.2), connecting ultraviolet (UV) absorption to galactic properties (Section 4.3), physical distribution of the CGM gas (Section 4.4), and the dynamics of the CGM around dwarf galaxies (Section 4.5). In Section 5, we discuss the limitations in this work and compare to other studies of the CGM, and we conclude in Section 6.

2 SIMULATIONS AND OBSERVATIONAL COUNTERPARTS

2.1 The simulations

We analyse the FIRE-2 simulations from the FIRE project (Hopkins et al. 2018), a set of cosmological 'zoom-in' simulations run with the GIZMO code (Hopkins 2015) 2 from redshift 99 to 0. The simulations use a meshless finite mass (MFM) hydrodynamic solver with explicitly modelled feedback processes. These processes include energy, momentum, mass, and metal fluxes arising from SNe types I and II, stellar mass-loss (O-star and AGB), radiation pressure (UV and IR), photoionization, and photoelectric heating. Subgrid metal diffusion is included for 4 of the 10 zoom-in simulations. The simulations explicitly follow chemical abundances of nine metal species (C. N. O, Ne, Mg, Si, S, Ca, and Fe), with enrichment following each source of mass return individually. O and Mg are primarily produced in SNII/Ib/Ic, while C is made in both SNII/Ib/Ic and AGBs, and finally Si is primarily produced in SNII/Ib/Ic and SNIa. The simulations include the cosmic UVB background model of Faucher-Giguère et al. (2009) together with local radiation sources. Self-shielding is treated with a local Sobolev/Jeans-length approximation. For more information about the simulations and feedback prescriptions, please refer to Hopkins et al. (2018).

We adopt a standard flat Λ -cold dark matter cosmology with cosmological parameters $H_0 = 70.2 \text{kms}^{-1} \text{Mpc}^{-1}$, $\Omega_{\Lambda} = 0.728$, $\Omega_m = 1 - \Omega_{\Lambda} = 0.272$, $\Omega_b = 0.0455$, $\sigma_8 = 0.807$, and n = 0.961.

2.2 Observational data

Since the CGM is diffuse and dim, most observations use absorption spectra of background quasars (QSOs) to gain information about the chemical component along specific sightlines, rather than through direct emission. Commonly observed absorption lines in the CGM include hydrogen Ly $\alpha \lambda 1215$, Si II $\lambda 1260$, Si III $\lambda 1206$, C II $\lambda 1334$, the C IV $\lambda\lambda 1548$, 1550 doublet, the Mg II $\lambda\lambda 2796$, 2803 doublet, and the O VI $\lambda\lambda$ 1031, 1036 doublet. In this work, we compare our simulations to three observational samples that looked at dwarf galaxies: the COS-Dwarfs survey (Bordoloi et al. 2014), the dwarf galaxies in Liang & Chen (2014), and the observations in Johnson et al. (2017). We describe all three here.

In the COS-Dwarfs survey, 43 low-mass galaxies ($10^{8.2}M_{\odot} < M_{\star} < 10^{10.0}M_{\odot}$) over the redshift range 0.010 < z < 0.104 were observed using background QSOs. The sources were selected based on galaxy properties, quasar brightnesses, and their projected separations. The survey focused on the C IV 1548 1550 doublet transition in low-redshift galaxies, constrained by the spectral coverage of the COS FUV gratings. This transition traces chemically enriched warm gas ($\sim 10^4$ K) or highly ionized gas around the host galaxy. They found that the C IV absorption strength decreases with radius as a power law ($W_r(R) \propto R^{-1}$) and that no absorption was detected

beyond $0.5R_{vir}$. The authors estimate the total carbon mass in the CGM of the galaxies to be comparable to the carbon mass in the interstellar medium of the host galaxy; this conclusion can be directly compared to the simulation data we analyse here. Further, they found a tentative correlation between C IV absorption strengths and host galaxy specific star formation rate (sSFR); we address this and similar correlations in Section 4.3.

Liang & Chen (2014) studied absorption transitions of Ly α , C II, C IV, Si II, Si III, and Si IV in the vicinities of 197 galaxies. They identified their sample from a comparison of 300 spectroscopic galaxies in the public archives, with background QSOs having high-quality spectra from the *Hubble Space Telescope (HST)* data archive. The sample has a median redshift of $\langle z \rangle = 0.176$, a median impact parameter of $\langle d \rangle = 362$ kpc and a median stellar mass of log(M_{star}/M_☉) = 9.7 ± 1.1. From $d < 0.3R_{vir}$ to $d \sim 0.7R_{vir}$, a differential covering fraction between low- and high-ionization gas is measured, suggesting that the CGM becomes progressively more ionized with increasing galactocentric radius; at $d \gtrsim 0.7R_{vir}$ no metals in either the low- or high-ionization states are detected from both the low- and high-mass galaxies, though abundant hydrogen gas is observed to well beyond R_{vir} , with a mean covering fraction of ~60 per cent.

Johnson et al. (2017) analysed the CGM of 18 star-forming field dwarf galaxies (log(M_{star}/M_{\odot}) = 7.7 – 9.2) at z = 0.09–0.3, drawn from a Magellan telescope survey in the fields of quasars with high signal-to-noise ratio COS absorption spectra. The authors searched for H I, Si II, Si III, Si IV, and O VI absorption in the quasar spectra. They found that compared to massive star-forming galaxies, metal absorption is less common from the CGM of dwarf galaxies, and is weaker with the exception of O VI.

In this paper, we will compare the structure of the CGM around low-redshift dwarf galaxies inferred by the FIRE-2 simulations with the results of the available observations (Bordoloi et al. 2014; Liang & Chen 2014; Johnson et al. 2017). For reference, the minimum detected EWs for C IV are 95, 59, and 130 mÅ, respectively; for Si III the minimum EWs are 54 and 70 mÅ for the later two references; for O VI the minimum detected EW is 80 mÅ (only Johnson et al. 2017) report this ion); there are no measurements for Mg II.

3 GENERATING ABSORPTION SPECTRA FROM SIMULATION DATA

To compare the results of the FIRE simulations and the observational data, we must select simulated galaxies that match the mass and redshift properties of the observational studies. Further, we must model a representative sample of spectra from sightlines using the simulated properties, including the density, temperature, and metallicity of the gaseous medium in the selected galaxies.

In this section, we describe the selection criteria for galaxies simulated within FIRE. We also describe how absorption spectra have been generated from the simulation data, enabling direct comparison to observational results.

3.1 Selection of simulated galaxies

To compare against the available observations, we selected the main galaxy from 10 zoom-in simulations of the FIRE-2 project with stellar masses between $10^8 \, M_\odot < M_* < 10^{11} \, M_\odot$, representing the stellar mass range of observed dwarf galaxies. The parameters of the selected simulations and their primary haloes are listed in Table 1.

For each of the simulations, we selected time-based snapshots that are within the redshift range of the observations ($0 \le z \le 0.3$). We find

the centre of the main galaxy in each snapshot using the Amiga Halo Finder (Gill, Knebe & Gibson 2004; Knollmann & Knebe 2009). We limit ourselves to snapshots that are separated, at a minimum, by the dynamical time of the galaxy's disc; the typical dynamical time of the simulated galaxies is \sim 50 Myr. We do this to ensure each galaxy snapshot is dynamically independent, and consequently not biased towards artefacts from the simulation or specific configuration of the galaxies. The dynamical time of the halo is an order of magnitude larger than the dynamical time of the galaxy, but the disc dynamical time is more relevant to the small-scale structures we seek to probe. We note that we do not place any constraint on the star formation properties or metallicity of the simulated galaxies.

Most of the runs we studied did not include any subgrid metal diffusion. However, four of the simulations did include a subgrid model for metal diffusion; these are marked with '_md' in Table 1. The FIRE simulations use a finite-mass method which by definition prevents transport of material from one particle to another. In reality, there will be transport of material from one region to another on the particle separation scale. The metal diffusion runs approximate this diffusion using the Smagorinsky (1963) model. We show in Section 4.1 that including the subgrid model for metal diffusion reduced the spread in the EW distribution.

These selection criteria resulted in ~40 snapshots for each galaxy with redshift ≤0.3, leading to 442 samples in total. The mass and sSFR of each sample galaxy are plotted in Fig. 1, alongside the properties of observed galaxies from the COS-Dwarfs survey, and Liang & Chen (2014). In order to study the SFR dependence of the CGM properties, we subdivide the sample into star-forming galaxies and quiescent galaxies, using the sSFR criterion of $10^{-10.6}$ yr⁻¹, as done for COS-Dwarfs. The simulated sample galaxies match the stellar mass range from the observations. The distribution of sSFR for $M_* > 10^9$ M_☉ matches that of the observational sample. However, for $10^8 < M_*/M_\odot < 10^9$, the simulations predict a large number of quiescent galaxies not found in the samples we compare to.

3.2 Generating synthetic spectra

For each sample galaxy, we choose ~ 135 random lines of sight (LOS), sampling the space uniformly (equal number per unit projected area). We limit ourselves to impact parameters between $0 \le b \le 1.3R_{\rm vir}$. We use the definition of $R_{\rm vir}$ in Bryan & Norman (1998), consistent with the observations.

The length of the LOS is $6R_{vir}$, so that the result will not be strongly affected by the low-resolution IGM region of the simulation. At a distance of $3R_{vir}$, the density in the simulations has dropped to $n \approx 2 \times 10^{-7} \,\mathrm{cm}^{-3}$ (roughly $\Omega_b n_{\mathrm{crit}}$ where $\Omega_b \approx 0.04$ is the cosmic baryon fraction and $n_{\rm crit} = \rho_{\rm crit}/m_p$, with $\rho_{\rm crit}$ the cosmic mean density), while the metallicity is $[M/H] \approx -3$ (see Section 4.4.2), comparable to the mean metallicity of the IGM [O/H] = -2.85Simcoe, Sargent & Rauch (2004). The maximum path length is set by the velocity offset Δv beyond which the absorber is no longer taken to be associated with the target galaxy. Following Bordoloi et al. (2014), we take $\Delta v = 600 \text{ km s}^{-1}$, which corresponds to a path length $D = \Delta v/H_0 \approx 8$ Mpc. Taking the fraction of $O_{\rm VI} = 0.2$, we find $N_{\rm OVI} \approx 1.6 \times 10^{12} \,{\rm cm}^{-2}$. As we will show later, this is about 20 per cent of the lowest column we find for impact parameters b $< R_{\rm Vir}$. Our estimates are biased low by this amount compared to observations.

A direct comparison to the observational work requires the generation of spectral EWs in each of the desired absorption lines. To do this, we need to estimate the density of the relevant ions. The FIRE simulation tracks 11 different atoms (H, He, C, N, O, Ne, Mg,

Name	$\log M_{ m halo}^0 \ ({ m M}_\odot)$	$\log M_{ m stellar}^0 \ ({ m M}_{\odot})$	$\log M_{ m gas}^0 \ ({ m M}_\odot)$	$M_{ m gas}^0/t_{ m dyn}^0$ $({ m M}_\odot{ m yr}^{-1})$	$SFR(M_{\odot} yr^{-1})$ $z = 0$	R _{vir} ⁰ (kpc)	m_b (M $_{\odot}$)	$\log m_{dm}$ (M _☉)	Reference
m11a	10.6	8.1	9.1	0.54	0.00	90.3	2100	4.0	1
m11b	10.6	8.1	9.3	0.91	0.02	92.4	2100	4.0	1
m11c	11.1	8.9	9.6	2.00	0.08	137.0	2100	3.0	1
m11d_md	11.5	9.6	10.2	7.52	0.04	169.2	7070	4.5	2
m11e_md	11.2	9.2	9.7	2.26	0.32	135.6	7070	4.5	2
m11f	11.7	10.4	10.3	8.87	6.61	207.6	1.7e4	4.9	1
m11h_md	11.4	9.6	9.8	2.80	0.45	146.4	7070	4.5	2
m11i_md	10.9	9.0	9.1	0.63	0.09	106.0	7070	4.5	2
m11q	11.1	8.7	9.7	2.24	0.03	138.8	880	3.6	3
m11v	11.4	9.4	10.2	7.38	0.73	167.4	7070	4.5	3

Table 1. Simulated galaxies in this study and their parameters.

Notes: Parameter units are physical:

(1) Name: simulation designation, the simulations which ends with 'md' have metal diffusion included.

(2) M_{halo}^0 : approximate mass of the z = 0 'main' halo (most massive halo in the high-resolution region).

(3) M_{stellar}^0 : stellar mass of the central galaxy in the main halo at z = 0.

(4) $M_{\text{gas}}^{0.\text{init}}$: gas mass of the central galaxy in the CGM at z = 0, the CGM region is the volume beyond the 80 per cent stellar mass radius and within R_{vir} of the main halo.

(5) t_{dyn}^0 : dynamical time of the main halo at z = 0, the halo dynamical times of all simulations are between 2.1e9–2.2e9 yr.

(6) SFR: star formation rate of the galaxy at z = 0, averaged over the last ten snapshots.

(7) $R_{\rm vir}^0$: virial radius of the z = 0 'main' halo, using the definition in Bryan & Norman (1998).

(8) m_b : initial gas particle mass.

(9) m_{dm} : dark matter particle mass.

(10) Reference: where the simulation is first presented. 1: Chan et al. (2018), 2: El-Badry et al. (2018), and 3: Hopkins et al. (2018).



Figure 1. The relation between sSFR and stellar mass in simulation and observations. The grey x's represent dynamically independent simulation snapshots from the FIRE project, where the SFR is calculated by averaging young star particle mass over a 5-Myr time-scale, the grey upper limits have sSFR lower than $10^{-12} M_{\odot} yr^{-1}$; the blue and red squares are star-forming and quiescent galaxies from the COS-Dwarfs survey; and the blue and red stars are star-forming and passive galaxies from the Liang & Chen (2014) study. SFGs and passive galaxies are separated at sSFR = $10^{-10.6} M_{\odot} yr^{-1}$. The Johnson et al. (2017) sample is not included in this figure because of the unknown sSFR of the galaxies.

Si, S, Ca, and Fe). However, performing full radiative transfer to determine the ionic states would be computationally expensive to run at the desired resolution.

We use the open-source code TRIDENT (Hummels, Smith & Silvia 2017) to post-process the FIRE simulations for analysis and comparison with the observational sample. We apply TRIDENT to infer the abundance of four ions throughout the simulations: C IV, O VI, Mg II, and Si III accounting for photoionization and collisional ionization. TRIDENT estimates these abundances using a detailed look-up table assembled from over 10^6 CLOUDY (Ferland et al.

2013) radiative transfer simulations, systematically varying density, temperature, and a redshift-dependent radiation field consistent with the Haardt–Madau UV/X-ray background (Haardt & Madau 2012).

The effects of self-shielding are well known to play a role in regions of high gas density, where the abundance of some lowionization energy ions are partially shielded from ionizing radiation (Rahmati et al. 2013). We account for this effect using new updated TRIDENT ion tables using a method first described in Emerick, Bryan & Mac Low (2019) for the GRACKLE cooling code (Smith et al. 2017). These tables are constructed as described above from CLOUDY simulations, each initialized with a single density, temperature, and radiation field. Each CLOUDY run is a 1D simulation with length equal to the Jeans length but capped at 100 pc for large Jeans lengths. The radiation field is applied to one end of the simulation volume, and the simulation executes until it reaches an equilibrium state. At the end of this period, the ionic abundances are measured at the other end of the simulation volume, self-consistently capturing the effects of self-shielding by how effectively radiation is able to penetrate the intervening gas on the maximum scale it would clump.

Using these post-processed ionic abundances added to our FIRE galaxies, we generate ~ 135 random sightlines through the each simulation volume at a variety of different impact parameters relative to the centre of each galaxy. These sightlines are meant to mimic the observational data sets, each representing the path of light from a distant quasar through the intervening CGM of our foreground galaxies. The gas properties and ionic abundances for each sightline are passed to TRIDENT (Hummels et al. 2017) to generate synthetic spectra, depositing absorption features in the spectra for each absorbing ion encountered along the LOS. Finally, we numerically integrate the absorption features for each ionic transition to identify an EW value appropriate for direct comparison with the observational samples. We have visually inspected the spectra of Mg II, and almost all of the lines are optically thin (the double ratio is 2:1); only 1 LOS out of 6000 had a ratio that was clearly lower.



Figure 2. Rest-frame absorption EW of four ions (C IV, O VI, Mg II, and Si III) versus normalized impact parameter. The 2D histograms of EW of mock LOS from all dynamically independent simulation snapshots (defined in Section 3.1) are shown in the background, with the solid black lines showing the median values and the solid blue lines showing the 16–84 per cent values of the six runs without metal diffusion (Table 1), and the dotted lines showing the statistics for the four runs with metal diffusion (m11d, m11e, m11h, and m11i). Available observational data are included for each ion, and the data points with only downward arrows are 2σ upper limits on non-detections. The amount of C IV predicted by the FIRE simulations is roughly consistent with observational constraints. The EWs of the other ions inside ~0.5 R_{vir} are underpredicted by a factor of 3–10. Beyond half a virial radius, there are only non-detections from observations.

4 RESULTS

4.1 Ion equivalent width versus impact parameter

We compare the generated absorption spectra of the four ions in the CGM of the simulation sample to the observations using the measure of EW. In this subsection, we compare the distributions of absorption EW as a function of impact parameter, which is directly measured in observations.

We show the rest-frame EWs of four ions (C IV, O VI, Mg II, and Si III) as a function of normalized impact parameter in Fig. 2, and compare to observational data from Bordoloi et al. (2014), Liang & Chen (2014), and Johnson et al. (2017). The EW from all mock LOS from all 384 dynamically independent snapshots are shown as 2D histograms in the background in Fig. 2, with the black and blue lines showing the median and 16–84 per cent ($\sim \pm 1\sigma$) values, and available observational data are included in the panels for each ion.

C IV has the highest number of observations among the four ions. Within $0.5R_{vir}$, 50 ± 12 per cent of the observations are above the detection limit, and 27 per cent of the simulated sightlines are above the detection limit (~100 Å), so the simulated EW – impact parameter distribution for C IV is somewhat low compared to observations. At large radii, the simulation underpredicts the EW by a large factor.

For O VI, there are five detections and three upper limits available from observations. The simulated O VI EW distribution is about a factor of 3 lower than the observed distribution, since the detections are all above the 84th percentile in the simulations and the upper limits are around the simulated median. Although this could be due to the fact that the galaxies in Johnson et al. (2017) are all star-forming (without available estimates of SFR), and as discussed in Section 4.3, O VI EW increases with SFR.

For Si III, there are 10 detections available, all at $d/R_{\rm vir} < 0.75$. The detections fall above the predicted median of the simulated EW distribution.

For Mg II, there is no observational data available for dwarf galaxies to compare against; the prediction from the simulations is that it will be very difficult to detect this ion in absorption.

In general, the EWs of the four ions all decrease with normalized impact parameter $(d/R_{\rm vir})$ – with shallower fall-off rates for C IV and O VI, and steeper fall-off rates for Si III and Mg II, indicating that there is little Mg II and Si III in the outer CGM of dwarf galaxies (see Section 4.4.2). In the case of Si III, it appears that the EW falls off more rapidly with impact parameter in the simulations than it does in the observations. And for all ions, there are wide ranges of possible EW values at high impact parameters. For the two low ionization potential ions, there are significant fractions of LOS with EW consistent with 0 Å in the simulations: 96.8 per cent for Mg II, and 76.0 per cent for Si III.

There are gaps in the EW distributions of C IV (around $10^{-4.4}$ Å), Mg II (around $10^{-5.2}$ Å), and Si III (around $10^{-5.0}$ Å) in Fig. 2, which are more clearly seen in Fig. 3. The gaps are due to the fact that most CGM gas reside either on the low temperature or high temperature side of the peak in the cooling function, and not at the peak. This is discussed further in the next section.



Figure 3. Distributions of the EW of the four ions in this study along the same LOS. The plots along the diagonal show histograms of the EWs of the four ions. The C IV, Si III, and Mg II histograms have more than one peak due to atomic physics: the CGM gas associated with the two peaks in the EW histograms reside on the low *T* and high *T* side of the peak in the cooling function, leaving a gap (corresponding the the maximum cooling rate) in between (see Section 4.2). The off-diagonal panels display the relationships between the EWs of the ions along the same LOS. Note that only 3.2 per cent and 24 per cent of LOS for Mg II and Si III are included in this figure, since the rest of the LOS have EW ≈ 0 Å for the two ions; and not all LOS shown in the histograms appear in the off-diagonal panels, because phases of the CGM gas probed by the different ions do not completely overlap with each other. There are strong correlations between some pairs of ions, such as O VI and C IV, Si III and C IV, and Mg II and Si III; and no correlation between O VI and Mg II.

4.2 CGM phase space

Having generated the absorption spectra of C IV, O VI, Mg II, and Si III along identical LOS, we can use them to examine the relationship between the EWs of the ions in the simulations. Fig. 3 shows the EW distributions of the four ions from all sightlines produced, and their mutual relationships.

The EW histograms of C IV, Si III, and Mg II in Fig. 3 display bimodality; in C IV the gap is at EW $\sim 10^{-4.4}$ Å, in Si III the gap is at $\sim 10^{-5.0}$ Å, in Mg II the gap is at $\sim 10^{-5.2}$ Å. We do not see a gap in O VI. The existence of the gaps in the histograms is connected with the multimodal distributions of the ion temperatures: most of the gap

is around $T \sim 10^5$ K, which is likely shock heated and produces the larger EW peaks with higher EW values and higher column densities; The LOS in the lower EW peaks have lower column densities for the corresponding ion compared to the LOS in the higher EW peak, and is associated with $T \sim 10^4$ K gas, which is likely photoionized. This is the effect of the cooling curve which produces a multiphase medium that imprints itself as discrete temperatures and correspondingly discrete ionization fractions per particle. These create the observed multimodalities in the EW distributions. The reason we do not see multimodality in O VI is that the low temperature gas is insufficiently energetic to reach that ionization state.



Figure 4. Distributions of O VI EW and three galactic properties. The diagonal plots show histograms of four quantities – M_* , SFR, sSFR, and O VI EW. The off-diagonal plots display the relationships between all of these properties. In particular, the bottom off-diagonal panels show correlations between O VI EW with M_* , SFR, and sSFR (correlation coefficients presented in Table 2). The EW of the other ions in this study (C IV, Mg II, and Si III) show similar trends with the four galactic properties. Other strong correlations with CGM absorption include M_H , M_*/M_H , and R_{vir} . The ions in our analysis do not show strong correlation with sSFR (as shown in the bottom panel in the third column), impact parameter, or redshift.

The off-diagonal plots in Fig. 3 show relationships between the EWs of the four ions. There are correlations between some pairs of ions, such as O VI and C IV, Si III and C IV, and Mg II and Si III; and little correlation between O VI and Mg II, or between Mg II and C IV. The correlations indicate which pairs of ions probe the same phase of the CGM gas, and this is largely consistent with the difference between their respective ionization energies. For example, in the case of Mg II (ionization energy of 15 eV) and O VI (ionization energy of 138 eV), any given LOS containing absorption from Mg II will also have absorption from O VI. Since O VI is located at large galactocentric radii, and Mg II is located at small galactocentric radii, there is little correlation between the two ions. Note that the O VI is high in EW for all LOS with Mg II, because of the large path length through the halo at small impact parameters, the latter being the only

LOS where the simulations predict significant amounts of Mg II. We discuss this further in Section 4.4.2.

4.3 Connecting UV absorption to global galactic properties

We investigate the relationship between UV absorption in the CGM and a series of global galactic properties. The global properties we examine are M_* , M_H , M_*/M_H , SFR, R_{vir} , and sSFR; the other properties we study are redshift, impact parameter d, and d/R_{vir} .

We find that all of the global properties correlate with one another, except for sSFR; and CGM absorption lines correlate with all of the global property correlations, again, except for sSFR. Fig. 4 shows the distributions of O VI EW and three selected galactic properties. The

Table 2. Correlation coefficients of EW, ordered by decreasing ionization potential, and various global galactic quantities. The errors of the correlation coefficients are ~ 0.004 , calculated using the bootstrap sampling method. The correlation strength increases with the ionization potential of the ion.

	O VI	CIV	Si III	Mg II
<i>M</i> _*	0.58	0.39	0.28	0.25
$M_{ m H}$	0.53	0.35	0.25	0.23
$M_*/M_{\rm H}$	0.57	0.37	0.25	0.23
R _{vir}	0.55	0.36	0.25	0.24
SFR	0.51	0.36	0.26	0.23
sSFR	0.29	0.21	0.15	0.15
d	0.15	-0.05	-0.16	-0.13
$d/R_{\rm vir}$	-0.31	-0.41	-0.46	-0.37
z	0.00	-0.02	-0.05	0.02



Figure 5. The four principle components which account for 98.2 per cent of the variance in the original data. The features involving absorption lines on the *x*-axis represent the absorption EWs of the ions, and the values of all dimensions are in log space except for SFR. The features have been transformed so that there is zero mean and unit variance. The percentage of variance preserved in each principle component is shown in the legend. 62 per cent of the variance is accounted for by a single component, which we interpret as a correlation between halo mass and all other forms of mass, including stellar mass and gas mass in the CGM. The second component, which accounts for 23 per cent, we interpret as a temperature effect. Lower mass haloes have lower virial temperatures, which result in higher absorption for the low-ionization potential ions Mg II and Si III.

bottom off-diagonal panels illustrate that there are strong correlations between O VI EW and two global galactic properties – M_* and SFR; in contrast, there is not any significant trend of O VI EW with sSFR. This is also true for the other ions (C IV, Mg II, and Si III) analysed in this work, although with weaker correlation strengths with decreasing ion ionization potential. In general, The absorption strengths of the ions increase with M_* , $M_{\rm H}$, $M_*/M_{\rm H}$, SFR, and $R_{\rm vir}$, and decrease with $d/R_{\rm vir}$, with the correlation strengths increasing with ion ionization potential. The correlation coefficients between EW and various properties are listed in Table 2.

The CGM absorption has a weak dependence on redshift, sSFR, and impact parameter in our analysis. The weak dependence on redshift is not surprising considering that the simulation data in our analysis is in the narrow redshift range of 0–0.3. However, it is reasonable to expect that the strength of absorption lines to increase with sSFR – when fixing the M_* of a galaxy, the EW of ions could increase with SFR if the material in the halo is due to star formation related winds. So the lack of correlation of ion absorption with host galaxy sSFR is at first glance puzzling. We checked the variation of

SFR over time, and it appears to occur in short bursts (Muratov et al. 2015; Anglés-Alcázar et al. 2017), so the material in the halo is not due to steady winds. It seems likely that the material in the halo was put in place by earlier star bursts. This is consistent with a particle tracking analysis (Hafen et al. 2019b, more on this in Section 5.1). Consequently, on closer examination, the lack of correlation between sSFR and material in the halo is not that surprising.

To further study the relationship between the global properties and the absorption strengths of the ions, we perform a principal component analysis on the data. Fig. 5 shows the first four principle components which capture 98.2 per cent of the variance of the data when projected into a 7D space. The first principle component (labelled 'PC 0' on the figure), which accounts for 62 per cent of the variance, shows that the three global galaxy properties (halo mass, stellar mass, and SFR) and the ion absorption strengths are all correlated. This component may be interpreted as indicating that more massive haloes (and hence more massive galaxies) contain more gas, which in turn produce stronger ionic absorption. The second principle component ('PC 1'), which accounts for 23 per cent of the variance, shows that the absorption strengths are anticorrelated with all three global galaxy properties, with lower ionization state ions being more strongly anticorrelated with the galaxy properties. We interpret this component as an indicator of the temperature of the CGM; higher mass haloes have higher virial temperatures, hence fewer low ionization ions compared to lower mass (and temperature) haloes. The third principle component ('PC 2') shows that while the absorption strength of three ions are anticorrelated with SFR, Mg II is correlated with SFR. This component, which accounts for 7 per cent, may be an indicator of the stronger winds from feedback associated with star formation.

4.4 Physical distribution of the CGM gas

We investigate the geometry of the ionized gas probed by the absorption lines simulated, and the physics that underlies its distribution.

4.4.1 Spatial distribution of C IV absorption

We compare the 1D radial distribution of C IV absorption EWs in our sample to the proposed profile in Bordoloi et al. (2014):

$$\langle W_r(R) \rangle = W_0 \times \left(\frac{R}{R_x}\right)^{-1} \times \exp\left(\frac{-R_{\rm vir}}{R_x}\right)$$
 (1)

where W_0 and R_x are two free parameters describing absorption and radial scaling. The best-fitting values obtained by Bordoloi et al. (2014) from a maximum-likelihood fit are $R_x = 105 \pm 27$ kpc and $W_0 = 0.6 \pm 0.3$ Å.

In Fig. 6, we divide our sample into $4 R_{vir}$ bins, roughly equivalent to halo mass, which enables direct comparison with Bordoloi et al. (2014). We then bin our data radially, and calculate the average C IV absorption EW in each bin. To account for the effect of the detection limit in the COS-Dwarfs Survey, we repeat the calculation with a detection limit of 50 mÅ, the lowest detection limit in the COS-Dwarfs survey. The resulting radial profiles are plotted in Fig. 6, along with the observational data and the proposed profile from Bordoloi et al. (2014).

The C IV absorption EW profiles derived from our simulations do not agree with the proposed profile in Bordoloi et al. (2014) in slope, parametrization, and trend with R_{vir} : the FIRE profiles fall off more steeply with radii (the simulated C IV EW profile follows a R^{-3} power law beyond the inner core regions), which means that the C IV ion is more concentrated towards the centre of the halo in the simulations;



Figure 6. A comparison of the average C IV absorption EW profile in simulations with the best-fitting model from Bordoloi et al. (2014). The four panels correspond to four R_{vir} bins (with mean R_{vir} values of 100, 148, 191, and 215 kpc), which translate to mean M_* values of 6.2×10^8 , 3.9×10^9 , 2.0×10^{10} , and $3.3 \times 10^{10} M_{\odot}$. In each panel, the blue triangles show the average value of C IV EW in radial bins in the simulations, while the orange diamonds show the average C IV EW placing a lower detection limit of 50 mÅ similar to the limits in Bordoloi et al. (2014); the green points are detections and upper limits in the COS-DWARFS survey (Bordoloi et al. 2014); the red dashed line is the best-fitting model by Bordoloi et al. (2014); and the blue dotted line shows a reference R^{-3} profile. We find that the simulated C IV EW profile follows a R^{-3} power law beyond the inner core regions. The C IV EW profile for sightlines above 50 mÅ in simulations agree well with COS-Dwarf observations and the best-fitting model for the two largest R_{vir} bins, but diverge for the smallest haloes. The difference between the full and detection-limited mean EW profiles suggests that the form of the best-fitting model from Bordoloi et al. (2014) is primarily due to the detection limit bias of the observations.

equation (1) suggests that C IV absorption EW decreases with $R_{\rm vir}$, while C IV absorption EW increases with $R_{\rm vir}$ in the simulated sample, and the correlation coefficient between C IV absorption EW and $R_{\rm vir}$ in the simulated sample is 0.36.

The conflict between the simulated profile of the C IV EW and the modelling in Bordoloi et al. (2014) is, we suggest, due to the detection limit bias. The proposed profile in Bordoloi et al. (2014) is based on a limited sample with 17 detections in total, with no detections beyond 100 kpc, and the two lower $R_{\rm vir}$ bins are not well constrained with only one detection. The fact that the C IV EW profile for sightlines above 50 mÅ in simulations agree well with both the COS-Dwarf observations and the best-fitting model from Bordoloi et al. (2014) for the two largest $R_{\rm vir}$ bins suggests that this inferred profile is due to the detection limit bias in observations.

4.4.2 Physical distribution of ions and metals in the CGM

We examine the physical distribution of the ions and metals in the CGM based on the covering fractions and column densities of the ions.

The covering fraction of each given ion was calculated using the following formula:

$$C_f(R) = \frac{N_W(\geq W_{\text{cut}}, R)}{N_{\text{tot}}(R)}$$
(2)

where $N_W(\geq W_{\text{cut}}, R)$ is the number of LOS with absorbers with EW greater than the cut-off EW (W_{cut}) within an impact parameter R, while $N_{\text{tot}}(R)$ is the total number of LOS that we sample in the same region.

Fig. 7 shows the covering fractions of C IV, Si III, and Mg II above 50 mÅ, and of O VI above 80 mÅ, compared with the observations (Bordoloi et al. 2014; Liang & Chen 2014; Johnson et al. 2017). The thresholds are chosen to match the lowest detection limits of the ions. The covering fractions from the simulated sample are in agreement with the results from observations.

We can also analyse the column densities of the ions directly in the simulations. The column density maps of the ions are shown in Fig. 8. The three circles shown in the maps are the 80 per cent stellar mass radius, the boundary between the inner and the outer parts of the CGM (0.45 $R_{\rm vir}$ which we identify as the edge of the wind-dominated regime), and the virial radius. We arrived at this definition for the inner most radius as follows: we plotted the cumulative stellar mass inside the virial radius of the main halo. In all the haloes, we saw a sharp break in the cumulative mass function at around 80 per cent of the total stellar mass in the halo. We call this the 80 per cent stellar mass radius, and we use this as the boundary between the ISM and the CGM.

The column densities of the ions are different in the three regions: for C IV, the ISM has the highest column density, the inner part of the CGM shows a lower column density compared with the ISM, and the outer CGM has the lowest column density; Mg II and Si III show similar trend as C IV, except that the column densities of the low ions show more drastic decreases in the outer CGM; O VI has similar column densities in all three regions.

The roughly constant column density of O VI is consistent with the roughly constant cover factor that we see in Fig. 7. Why does O VI behave so differently than the other ions that we study? Fig. 9 shows the run of metallicity with radius; it decreases roughly as a power law with radius. Since both the density (not shown) and metallicity drop off with increasing radius, one might expect the O VI column density would drop off with radius as well. However, Fig. 10 shows the three dominant ionization states of oxygen as a function of radius in a simulation snapshot ('m11b', z = 0.061), demonstrating that the ionization fraction of O VI increases with radius. Recalling that the column density is the product of the ion O VI fraction times the density times the metallicity times the path length, the rapid increase in the O VI ionization fraction is largely responsible for the relatively flat run of column density at large impact parameter in the O VI panel in Fig. 8.

As confirmation of this interpretation, we have examined the run of the C IV fraction (not shown) which, in contrast to the run of O VI, decreases with increasing radius.

We see a boundary in the column densities of the ions at 0.45 $R_{\rm vir}$, which appears to separate an 'inner' and 'outer' region in the CGM. This is consistent with the location where the fraction of the total



Figure 7. The covering fractions of C IV, Si III, and Mg II above 50 mÅ, and of O VI above 80 mÅ, calculated from all selected LOS in the simulations, with the ionization energy for each state indicated. The different covering fraction threshold for O VI is a result of the different detection limit for this ion in observations. The data points with error bars are compiled from the observations (Bordoloi et al. 2014; Liang & Chen 2014; Johnson et al. 2017), which are binned radially. The covering fractions from the simulated sample are consistent with the results from observations.

CGM mass contributed by wind and by IGM accretion is equal for $10^{11}M_{\odot}$ haloes in FIRE-2 simulations (Hafen et al. 2019b).

The density–temperature diagrams for gas in the three regions (discy ISM, inner CGM and outer CGM) are shown in Fig. 11 in three stellar mass bins. The gas in these three regions resides in different places in the phase diagram: the ISM gas is cold and dense compared to gas in the inner and outer CGM. The outer CGM follows the adiabats ($T \propto \rho^{2/3}$) indicated by the white lines, while the inner CGM and ISM do not. This indicates different energetic states between the three regions, which may be the result of the different origins of the gas in these regions (Hafen et al. 2019b).

We can compare the simulated carbon mass fraction to those inferred from observations. The cumulative carbon mass distribution of the simulated sample is shown in Fig. 12 in three mass bins. We use the 80 per cent stellar mass radius as the boundary between the ISM and CGM, as discussed earlier. For all three stellar mass bins, the carbon mass in the CGM is larger than or comparable to the carbon mass in the ISM, which is in agreement with the estimates in Bordoloi et al. (2014). We also note that as the stellar mass of the galaxy decreases, a larger fraction of the carbon resides in the CGM of the simulated sample.

4.5 Dynamics of the CGM

Throughout this paper, we have mostly focused on accounting for the mass budget of the multiphase CGM. However, the structure and composition of the CGM are directly related to its dynamics, which is critical to understanding how the CGM fuels the star-forming regions of dwarf galaxies. In this subsection, we examine how the composition of the CGM relates to its dynamics. Specifically, recent analytic models (e.g. Mathews & Prochaska 2017; McQuinn & Werk 2018; Faerman, Sternberg & McKee 2019) assume that the CGM of L_* galaxies is in hydrostatic equilibrium (HSE) or near HSE with a net inflow (Stern et al. 2019). Here, we test whether the CGM of the simulated dwarf galaxies deviates from HSE by comparing gas pressure gradients to gravitational force.

We show the gas thermal pressure gradient $(dP_{thermal}/dr)$ and the gravitational force $(\rho(r)GM(r)/r^2)$ as a function of galactocentric radius in the top panel of Fig. 13. The profiles are calculated by volume averaging the density and pressure over spherical shells centered at 40 values of R/R_{vir} . The thermal pressure gradient is calculated by dividing the pressure difference between adjacent shells by the distance between them. The bottom panel of Fig. 13 shows the ratio of dP_{thermal}/dr to $\rho(r)GM(r)/r^2$. The ratio ranges from 0.5 (for $M_{\rm star} \sim 10^{10} \,{\rm M_{\odot}}$) to 0.7 (for $M_{\rm star} \sim 10^{8} - 10^{9} \,{\rm M_{\odot}}$) for $R > 0.4 R_{\rm vir}$, so the gas is nearly but not quite in HSE. At smaller radii, the ratio is smaller. This is consistent with the findings in Stern et al. (in preparation), who find that in haloes with same mass as those we study, only gas at larger radii is in HSE supported by thermal gas pressure. In general, the outer CGM in our simulations is mostly supported by thermal pressure, and is near but not in HSE, which means that we expect to see significant motion in the medium (see Esmerian, in preparation, for similar results for Milky Way-mass galaxies simulated with the FIRE-2 physics).

To investigate the CGM's deviation from HSE, we calculate the inflow and outflow rates for the three stellar mass bins as a function of galactocentric radius, which are shown in Fig. 14. We then compare those rates to the net accretion rate (inflow–outflow). We see a net inflow through the CGM of order $0.2 \text{ M}_{\odot} \text{ yr}^{-1}$ for the median stellar mass bin. This inflow rate, which increases with stellar mass, is



Figure 8. Column density maps of C IV, O VI, Mg II, and Si III at z = 0.061 for simulation run 'm11i'. The black circle shows the 80 per cent stellar mass radius, the dashed white circle shows the boundary between the inner and the outer parts of the CGM (0.45 R_{vir} , the edge of the wind-dominated regime), and the solid white circle shows the virial radius. For C IV, the column density map shows three distinct regions: the ISM with the highest C IV column density, the inner part of the CGM with a lower C IV column density compared with the ISM, and the outer CGM with the lowest C IV column density; the column densities of O VI in the three zones look comparable; for Mg II, there are also three distinct regions in the column density map: the ISM where the Mg II column density is the highest, the inner part of the CGM with lower Mg II column density, and the outer CGM where the Mg II column density is close to 0; the column density distribution of Si III is similar to that of Mg II.



Figure 9. The run of metallicity with radius in simulation 'm11b' at z = 0.061. The blue and orange lines are power laws of index -1 and -1.5, as indicated in the legend.



Figure 10. The ionization fractions of O V, O VI, and O VII relative to oxygen in simulation run 'm11b' at z = 0.061. The profile of O VI as a function of radius better traces the ionization state of oxygen, rather than its density.



Figure 11. This figure shows the density-temperature phase diagrams of three regions inside the virial radii of the simulated galaxies in three mass bins. The three rows show galaxies with stellar masses $10^{7.5} M_{\odot} < M_* < 10^{8.5} M_{\odot}, 10^{8.5} M_{\odot} < M_* < 10^{9.5} M_{\odot}, and <math>10^{9.5} M_{\odot} < M_* < 10^{10.5} M_{\odot}$. The three columns show the discy ISM ($R < 0.03 R_{vir}$), inner CGM ($0.03 R_{vir} < R < 0.45 R_{vir}$), and outer CGM ($0.45 R_{vir} < R < R_{vir}$) regions surrounding the galaxies. Colours show the probability density weighted by gas mass. The gas in these three regions reside in different places in the density-temperature diagram. The white dashed lines are isobaric lines with pressures of 1.0×10^{-16} , 1.0×10^{-14} , and 1.0×10^{-12} Ba. The white lines show the scaling $T \propto \rho^{2/3}$, which comes from the relationship when entropy is constant. The outer CGM follows the adiabats indicated by the white lines, while the inner CGM and the discy ISM clearly follow different trends. The fact that the high-temperature gas in the outer CGM (rightmost column) traces an adiabat, shows that the gas in that region is well mixed.



Figure 12. The cumulative carbon mass distribution of the simulated galaxy samples in three mass bins. The vertical dashed lines indicate the distance containing 80 per cent of the stellar mass for each bin, corresponding to R/R_{vir} values of 0.050, 0.084, and 0.047. We treat this as a dividing line between the ISM and CGM. For all mass bins, the carbon mass in the CGM is larger than or comparable to the carbon mass in the ISM. As the stellar mass of the galaxy decreases, a larger fraction of the carbon resides in the CGM.

comparable to the SFR of the galaxy, also of order 0.2 M_{\odot} yr⁻¹ for the median stellar mass bin. However, the gross inflow and outflow rates, while comparable to each other, are larger than the net inflow rate by an order of magnitude, that is, $\dot{M} \approx 2M_{\odot}$ yr⁻¹.

We find that the mass-weighted velocities of the inflow and outflow ($\sim 100 \text{ km s}^{-1}$) are similar to the virial velocity of the halo, also consistent with the ratio of the thermal and gravitational pressures

we see. Beyond explaining the dynamics of the system, this picture has a number of consequences: first, it appears that the inflow and outflow rates are well coupled throughout the halo; secondly, this also implies that whatever mechanisms drive the outflows, they are effective throughout the galactic halo; and consequently, the dynamics of the halo are coupled to the star formation activities in the central stellar region of galaxy.



Figure 13. The upper panel shows the gas thermal pressure gradient $\nabla P_{\text{thermal}}$ radial profile (orange) compared to the gravitational force per unit volume $\rho \nabla \Phi$ (blue) profile. The solid lines are the mean of the logarithm values for simulation snapshots with central galaxies in the median stellar mass bin $(10^{8.5} \text{ M}_{\odot} < M_* < 10^{9.5} \text{ M}_{\odot})$, and the shaded regions are between the $\sim \pm 1\sigma$ values. The profiles are calculated by volume averaging the density and pressure over spherical shells centered at 40 values of R/R_{vir} . The lower panel shows the ratio of thermal pressure to gravitational force as a function of normalized galactocentric radius R/R_{vir} in three mass bins. The inner CGM is far from HSE. The outer CGM is nearly but not quite supported by thermal gas pressure, thus not in HSE.



Figure 14. The average net inflow rates and gross inflow/outflow rates as a function of galactocentric radius normalized by virial radii $R_{\rm vir}$. The dotted profiles above zero are gross inflow rates, the profiles below zero are gross outflow rates. The solid lines represent net mass transfer rates. The mass transfer rates are calculated for the simulated sample in three stellar mass bins: '08' (blue, $10^{7.5} \,\mathrm{M_{\odot}} < M_* < 10^{8.5} \,\mathrm{M_{\odot}}$), '09' (orange, $10^{8.5} \,\mathrm{M_{\odot}} < M_* < 10^{9.5} \,\mathrm{M_{\odot}}$), and '10' (green, $10^{9.5} \,\mathrm{M_{\odot}} < M_* < 10^{10.5} \,\mathrm{M_{\odot}}$). The net inflow rates are comparable to the SFR, and are an order of magnitude smaller than the gross inflow and outflow rates, which are comparable to each other. The net inflow rate in the '10' mass bin is dominated by simulation 'm11f', which has a large SFR, if it is removed, the net inflow rate looks like the other two mass bins.

5 DISCUSSION

In this section, we discuss the limitations in the simulations and our analysis, such as the resolution and physics in the simulations and the post-processing tool used, and how those affect the results presented in this paper.

The typical gas mass resolution achieved for dwarf galaxies in FIRE simulations is ~5 × 10³ M_☉ (Table 1). Depending on the clump size of the gas in the CGM, the current resolution may not be enough to resolve the dynamics in the dense CGM phases. There is observational evidence suggesting that some low-ion absorbers have subkiloparsec length-scales (Lehner et al. 2013; Rudie et al. 2019). For a clump radius of 400 pc, Crighton et al. (2015) estimate the minimum required mass resolution is 4 M_☉, which is a substantially finer resolution than available in cosmological zoom simulations like FIRE. The low-ion content is likely to increase with improved mass resolution in the CGM, as suggested by recent work by other groups (Peeples et al. 2019; Suresh et al. 2019; van de Voort et al. 2019).

We used the metagalactic radiation field Haardt & Madau (2012) when calculating densities and ionization states of metals using Trident. However, we did not include local photoionizing sources like nearby stars in the galaxy. This affects the analysis in two ways: it is known that Haardt & Madau (2012) underpredicts observed H I photoionization coefficients by a factor ~ 2 (see e.g. Faucher-Giguère 2020), so this likely has an effect on the prediction of the low ionization state ions in this work. Including the stellar radiation will change the ionization states of the ions in the CGM. However, because the stellar radiation decreases radially as a function of $1/r^2$, it

is unlikely to change the distribution of ionization state much beyond the galactic disc.

The simulation suite described so far in this work does not include CRs. Ji et al. (2019) studied the effects of CRs on galaxies in the FIRE simulations across a broad range of stellar and halo masses, and found that CR pressure appears to become important only when the galaxy halo mass approaches $10^{12} M_{\odot}$. CRs in the FIRE simulations do not appear to have a strong impact on the CGM of dwarf galaxies.

In a first step at quantifying the effects of CRs on dwarf galaxies in FIRE, we examined the CR version of the median stellar mass galaxy 'm11e' in our sample. We find no significant difference in the EW distributions of the ions between the two runs. The CR run has only ~44 per cent of the stellar mass compared to the non-CR run at low redshift, so we also compare the CR run 'm11e' with a non-CR run 'm11q' with comparable stellar mass at z = 0, and again find no difference in the EW distributions in most ions, with the exception of O VI, which has slightly higher EWs in the CR run.

If the CGM distributions around dwarf galaxies in the real universe resemble the results presented in this work below the detection limit, and given the detection thresholds of current instruments, observers probing the absorption properties of the low ions when the background QSO is beyond $\sim 0.4R_{\rm vir}$ of the galaxy, and beyond $\sim R_{\rm vir}$ for C IV will likely only observe upper limits given the current detection limits.

5.1 Origins and fates of halo metals

Using the same suite of FIRE-2 simulations as in this work, recent works trace the origins and fates of the material composing the CGM (Hafen et al. 2019a,b). These studies focus on halo masses $M_h(z = 0) \sim 10^{10} - 10^{12} M_{\odot}$ and redshifts z = 2 and 0.25, and show that across the whole halo mass range and redshift range studied, the main origin ($\gtrsim 60$ per cent) is IGM accretion (including infalling haloes), with wind from the central galaxy being the second most important contribution of the CGM, while wind from satellite galaxies contribute the least. The exception is for L_{\star} haloes, where the contributions from winds of the central galaxy and satellite galaxies are about equal.

Hafen et al. (2019a) study the fates of the CGM gas in the same halo and redshift ranges. For the CGM material at z = 2, immediately after leaving the CGM, most of the CGM gas enters the central galaxy; by z = 0, half of the mass remain in the ISM or stars in the central galaxy for the $M_h(z = 2) \sim 10^{11.5} M_{\odot}$ haloes, but for lower mass haloes most of the CGM mass are ejected into the IGM. For the CGM mass at z = 0.25, most of the CGM mass remains in the CGM by z = 0 across the halo mass range studied.

In general, the two studies find that there is substantial mixing of the CGM gas, as a result, there is little correlation between the origins and the fates of the CGM material, and a typical LOS can intersect CGM gas of multiple origins.

5.2 The picture of CGM composition and dynamics across halo mass and redshift ranges

In this subsection, we discuss the difference between this work and previous studies in stellar/halo mass ranges, redshift, dynamics, resolution, feedback prescriptions, and observables.

In this study, we selected the stellar mass and redshift ranges to be representative of current observations of the CGM of dwarf galaxies. Not many previous studies focus on the CGM of simulated galaxies in the same mass and redshift ranges, and the limited overlap occurs at the high stellar and halo mass ends of this study (e.g. Ford et al. 2013). Ford et al. (2013) study the CGM around z = 0.25 galaxies in cosmological hydrodynamic simulations, and find the low ions (Mg II and Si IV) drop rapidly with impact parameter, while high ions (O VI and Ne VIII) have relatively flat radial profiles; and absorption increases with halo mass. These findings are in agreement with the trends of the ion absorption profiles presented in this work.

We concluded in Section 4.5 that the CGM in this study is not in HSE. Most of the other studies find the same conclusion that CGM tends not to be in HSE in this halo mass range. Lochhaas et al. (2019) use spherically symmetric, idealized, kpc-scale resolution simulations of the CGM, and found that the $10^{11}M_{\odot}$ halo's CGM is not in HSE, while the $10^{12}M_{\odot}$ halo contains hot gas in HSE with cold gas falling on to the galaxy. Oppenheimer (2018) tested whether the haloes in the EAGLE zoom simulation are in HSE, and found that the Milky Way L^* haloes deviates significantly from HSE, especially in the inner regions (see also Esmerian, in preparation, for FIRE-2 Milky Way-mass galaxies). In more massive haloes ranging from 10^{11} to 10^{12} M_{\odot}, Fielding et al. (2017) use idealized 3D hydrodynamics simulations to study the dynamics of the CGM, and find that below a critical halo mass $M_{\rm crit} \approx 10^{11.5} {\rm M}_{\odot}$, the CGM gas is not thermally supported, and above M_{crit} the CGM is supported by thermal pressure created in the virial shock (see also Stern et al. 2020 for a generalization of this result for haloes which are depleted of baryons).

The resolution in the diffuse CGM is usually orders of magnitude coarser than in the galaxy in the FIRE simulations due to the density-dependent nature of Lagrangian schemes. Recent studies have increased the spatial resolution of the CGM in zoom simulations of Milky Way-like galaxies and have found substantial changes in the behaviour and observables of the CGM relative to lower resolutions. van de Voort et al. (2019) used the AREPO code to progressively increase the spatial resolution of the CGM, finding order of magnitude increases in the column density of neutral hydrogen present in the CGM of the target galaxy. Hummels et al. (2019) and Peeples et al. (2019) both used the ENZO code to study a Milky-Way analogue galaxy in zoom simulations with spatial resolutions up to an unprecedented 500 comoving parsecs throughout the CGM. Like the AREPO study, these groups also found an increase in HI among other changes. Peeples et al. (2019) discovered significant differences in the kinematic structure of absorption line features in accompanying synthetic spectra for these galaxies. Hummels et al. (2019) explained the effect by identifying that increased resolution lends itself to preserving cool gas for longer in the simulation and and avoiding artificial mixing of hot and cold gas phases. In a related study, Schneider & Robertson (2018) present a set of 5-pc constant resolution simulations with isolated disc galaxies, and show that the simulations have not converged at 10 pc resolution, demonstrating the need for high resolution when studying the effects of winds on the CGM. Based on the results in these papers, improving the resolution in the halo of the simulations used in this work will likely increase the low-ion content, and would be critical if the goal is to resolve the internal kinematic structure of absorption.

Compared to the FIRE simulation suite, there are differences in the physics and feedback prescriptions used in other simulations studying the CGM. These differences can play a role in changing the predicted behaviour and observables in the CGM (Hummels et al. 2013). For example, Lochhaas et al. (2019) use a simple formulation for galactic scale winds in the idealized simulations of the CGM; and Ford et al. (2013) use momentum-driven winds as the stellar feedback prescription in their cosmological simulations. The observables and tracers used in some studies differ from those used in this paper. For example, Augustin et al. (2019) use RAMSES zoom simulations to compute expected emission from the CGM, and find the predictions agree well with current observations after adjusting the Ly α escape fraction. The tracers can be studied using the FIRE simulation suite in future work.

6 CONCLUSIONS

In this work, we use the FIRE simulation suite with physically motivated models of the multiphase ISM, star formation and stellar feedback to study the CGM of dwarf galaxies at low redshift ($z \le 0.3$). In direct comparison with observations, our main conclusions are as follows:

(i) The EW and cover fraction of C IV predicted by the FIRE simulations are somewhat low compared to observational constraints inside of $0.5R_{\rm vir}$ (50 ± 12 per cent of the observations are above the detection limit, compared to only 27 per cent of the simulated sightlines above the detection limit). At larger radii, the simulation results lie factors of several to tens or even hundreds below the observations. The EWs of the other ions inside ~ $0.5R_{\rm vir}$ are also underpredicted, by factors between 3–10 or higher. Beyond $0.5R_{\rm vir}$, there are only non-detections from observations (Bordoloi et al. 2014; Liang & Chen 2014; Johnson et al. 2017, Fig. 2).

(ii) The FIRE simulations can explain the C IV EW profile fits obtained from observations for dwarf galaxies in different R_{vir} bins, and can reproduce the EW profile fits when placing a similar detection limit on simulation data (Fig. 6).

(iii) The covering fractions from the simulated sample are in agreement with the results from observations (Fig. 7).

(iv) In the FIRE-2 simulations, the majority of the carbon mass in gas in dwarf galaxies are in the CGM (Fig. 12), which is in agreement with observations (Bordoloi et al. 2014).

As the simulated sample is in reasonable agreement with observations, we can make useful predictions with the simulation data to provide insight in the physics of the CGM around dwarf galaxies:

(v) All of the global properties (M_* , $M_{\rm H}$, $M_*/M_{\rm H}$, SFR, $R_{\rm vir}$, and sSFR) correlate with one another, except for sSFR; and CGM absorption lines correlate with all of the global property correlations, again, except for sSFR (Fig. 4). The correlation strength increases with the ionization potential of the ion. The EW of ions in our analysis do not show strong correlation with sSFR, impact parameter ($0 < d < 1.25 R_{\rm vir}$), or redshift (0 < z < 0.3, see Table 2).

(vi) We see a 'three-zone' structure in the CGM of dwarf galaxies, associated with the discy ISM, the inner CGM, and the outer CGM. The break between the inner and outer CGM is around 0.45 $R_{\rm vir}$, where the fraction of the total CGM mass contributed by wind and IGM accretion is found to be equal for $10^{11} \,\mathrm{M_{\odot}}$ haloes in FIRE-2 simulations (Hafen et al. 2019b, see Fig. 8).

(vii) The flat radial profile of the O VI column density with radius is driven by an increase in the fraction of oxygen in the O VI state (Fig. 10). The column density of other ions show a decrease with radius (Fig. 8), driven by the combination of decreasing density and metallicity with increasing radius.

(viii) The outer CGM in the simulations is nearly but not quite supported by thermal pressure, thus not in HSE, as demonstrated by the existence of halo-scale bulk inflow and outflow motions. The net inflow rates are comparable to the SFR; in the lower mass haloes, the actual inflow and outflow rates are greater than the net inflow rate by an order of magnitude, and are comparable to each other and the virial velocity of the halo. This indicates that the feedback that drives the dynamics of the halo is coupled with the star formation activity towards the centre of the galaxy (Figs 13 and 14).

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Software: SCIPY,³ NUMPY⁴ (van der Walt, Colbert & Varoquaux 2011), MATPLOTLIB⁵ (Hunter 2007), MPI4PY ⁶ (Dalcín, Paz & Storti 2005), TRIDENT ⁷ (Hummels et al. 2017), and YT ⁸ (Turk et al. 2011).

DATA AVAILABILITY

The data supporting the plots within this article are available on reasonable request to the corresponding author. A public version of the GIZMO code is available at http://www.tapir.caltech.edu/phopkins /Site/GIZMO.html. Additional data including simulation snapshots, initial conditions, and derived data products are available at https: //fire.northwestern.edu/data/.

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³http://scipy.org
⁴http://numpy.org
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⁶http://pythonhosted.org/mpi4py/
⁷http://trident-project.org/
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