



Detailed Abundances in the Ultra-faint Magellanic Satellites Carina II and III

A. P. Ji^{1,14} , T. S. Li^{1,2,3,4,15} , J. D. Simon¹, J. Marshall⁵ , A. K. Vivas⁶ , A. B. Pace^{5,7,16} , K. Bechtol⁸ , A. Drlica-Wagner^{3,4,9} , S. E. Koposov^{7,10} , T. T. Hansen⁵ , S. Allam³, R. A. Gruendl^{11,12} , M. D. Johnson¹², M. McNanna⁸, N. E. D. Noël¹³, D. L. Tucker⁹, and A. R. Walker⁶
(MagLiteS Collaboration)

¹ Observatories of the Carnegie Institution for Science, 813 Santa Barbara St., Pasadena, CA 91101, USA; aji@carnegiescience.edu

² Department of Astrophysical Sciences, Princeton University, Princeton, NJ 08544, USA

³ Fermi National Accelerator Laboratory, P.O. Box 500, Batavia, IL 60510, USA

⁴ Kavli Institute for Cosmological Physics, University of Chicago, Chicago, IL 60637, USA

⁵ George P. and Cynthia Woods Mitchell Institute for Fundamental Physics and Astronomy, and Department of Physics and Astronomy, Texas A&M University, College Station, TX 77843, USA

⁶ Cerro Tololo Inter-American Observatory, NSF's National Optical-Infrared Astronomy Research Laboratory, Casilla 603, La Serena, Chile

⁷ McWilliams Center for Cosmology, Carnegie Mellon University, 5000 Forbes Ave., Pittsburgh, PA 15213, USA

⁸ Physics Department, University of Wisconsin-Madison, 1150 University Ave., Madison, WI 53706, USA

⁹ Department of Astronomy and Astrophysics, University of Chicago, Chicago, IL 60637, USA

¹⁰ Institute of Astronomy, University of Cambridge, Madingley Road, Cambridge CB3 0HA, UK

¹¹ Department of Astronomy, University of Illinois, 1002 W. Green St., Urbana, IL 61801, USA

¹² National Center for Supercomputing Applications, 1205 West Clark St., Urbana, IL 61801, USA

¹³ Department of Physics, University of Surrey, Guildford, GU2 7XH, UK

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Abstract

We present the first detailed elemental abundances in the ultra-faint Magellanic satellite galaxies Carina II (Car II) and Carina III (Car III). With high-resolution *Magellan*/MIKE spectroscopy, we determined the abundances of nine stars in Car II, including the first abundances of an RR Lyrae star in an ultra-faint dwarf galaxy (UFD), and two stars in Car III. The chemical abundances demonstrate that both systems are clearly galaxies and not globular clusters. The stars in these galaxies mostly display abundance trends matching those of other similarly faint dwarf galaxies: enhanced but declining $[\alpha/\text{Fe}]$ ratios, iron-peak elements matching the stellar halo, and unusually low neutron-capture element abundances. One star displays a low outlying $[\text{Sc}/\text{Fe}] = -1.0$. We detect a large Ba scatter in Car II, likely due to inhomogeneous enrichment by low-mass asymptotic giant branch star winds. The most striking abundance trend is for $[\text{Mg}/\text{Ca}]$ in Car II, which decreases from $+0.4$ to -0.4 and indicates clear variation in the initial progenitor masses of enriching core-collapse supernovae. So far, the only UFDs displaying a similar $[\text{Mg}/\text{Ca}]$ trend are likely satellites of the Large Magellanic Cloud. We find two stars with $[\text{Fe}/\text{H}] \leq -3.5$ whose abundances likely trace the first generation of metal-free Population III stars and are well fit by Population III core-collapse supernova yields. An appendix describes our new abundance uncertainty analysis that propagates line-by-line stellar parameter uncertainties.

Unified Astronomy Thesaurus concepts: Local Group (929); Stellar abundances (1577); Dwarf galaxies (416)

Supporting material: machine-readable tables

1. Introduction

Ultra-faint dwarf galaxies (UFDs) are the luminous counterparts to the least massive star-forming dark matter halos, likely forming stars during the first ~ 1 Gyr before being quenched by reionization (e.g., Bullock et al. 2000; Benson et al. 2002; Simon & Geha 2007; Brown et al. 2014; Simon 2019). As a result, the chemical abundances of stars in UFDs preserve a clean snapshot of chemical enrichment from the earliest stages of galaxy formation and reionization, providing a window to the most metal-poor stellar populations and their nucleosynthetic output (Kirby et al. 2008; Frebel & Bromm 2012; Geha et al. 2013; Weisz et al. 2014; Wise et al. 2014; Ji et al. 2015). Dozens of UFDs have now been discovered in deep, wide, and uniform photometric surveys such as the Sloan Digital Sky Survey, Pan-STARRS, and the Dark Energy Survey (DES; e.g., Willman et al. 2005; Belokurov et al. 2007; Bechtol et al. 2015;

Drlica-Wagner et al. 2015; Koposov et al. 2015a); Laevens et al. 2015). The large number of UFDs provides a large population of local objects that retain signatures of high-redshift star and galaxy formation.

Until recently, these UFDs have generally been assumed to be satellites of the Milky Way. However, the two most massive dwarfs orbiting the Milky Way, the Large and Small Magellanic Clouds (LMC and SMC), should have had their own satellite UFDs (e.g., D’Onghia & Lake 2008; Drlica-Wagner et al. 2015; Koposov et al. 2015b; Jethwa et al. 2016; Dooley et al. 2017; Sales et al. 2017). Since the LMC and SMC are likely on their first infall into the Milky Way (Besla et al. 2007; Busha et al. 2011; Kallivayalil et al. 2013; Simon 2018; Fritz et al. 2019; Pace & Li 2019), any dwarfs that were previously Magellanic satellites could now be in the process of accretion into the Milky Way. *Gaia* proper motion measurements have revealed that several UFDs are kinematically associated with the LMC/SMC system (Kallivayalil et al. 2018; Erkal & Belokurov 2019). Two of these LMC satellites are Carina II (Car II; $M_V = -4.5$, $L/L_\odot \sim 10^{3.7}$) and Carina III

¹⁴ Hubble Fellow.

¹⁵ NHFP Einstein Fellow.

¹⁶ Mitchell Astronomy Fellow.

Table 1
Observations

Star	source_id	R.A.	Decl.	G (mag)	g_0 (mag)	r_0 (mag)	Slit	t_{exp} (hr)	S/N (4500 Å)	S/N (6500 Å)
CarII-6544	5293947247051916544	07:36:51.11	−58:01:46.3	15.07	15.63	14.63	0''.5	1.8	22	67
CarII-7872	5293894539213647872	07:36:51.89	−58:16:39.2	15.50	15.92	15.01	0''.7	1.0	25	60
CarII-5664	5293896360279425664	07:38:08.51	−58:09:35.0	16.33	16.55	15.86	0''.7	3.8	38	80
CarII-0064	5293951473299720064	07:36:21.25	−57:58:00.2	16.78	16.96	16.30	1''.0	2.6	22	54
CarII-4704	5293928074318184704	07:35:37.66	−58:01:51.8	17.40	17.46	16.93	0''.7	3.3	13	34
CarII-9296	5293900827045399296	07:37:39.79	−58:05:06.9	17.72	17.86	17.29	1''.0	3.0	15	35
CarII-2064	5293951881319592064	07:36:01.33	−57:58:43.8	18.22	18.27	17.77	0''.7	4.6	13	31
CarII-4928	5293951503362524928	07:36:24.98	−57:57:14.2	18.42	18.40	17.96	1''.0	5.5	13	31
CarII-V3 ^a	5293940924860019584	07:35:09.12	−57:57:14.8	18.46	18.13	18.01	1''.0	2.5	15	24
CarIII-1120	5293955665187701120	07:38:22.30	−57:53:02.1	17.46	17.51	16.97	0''.7	2.7	18	39
CarIII-8144	5293907630273478144	07:38:34.93	−57:57:05.3	17.65	17.72	17.18	0''.7	3.2	21	41

Notes. Our star ID numbers are the last four digits of the *Gaia* source_id. Here G is *Gaia* magnitudes, and g_0 and r_0 are dereddened DECam photometry from MagLiteS, taken from T. S. Li. et al. (2019, in preparation). The S/N is per pixel.

^a This star is a variable RRL star. The magnitudes here are the mean magnitudes found by MagLiteS and *Gaia* (Torrealba et al. 2018; Clementini et al. 2019), where the DECam magnitudes have been dereddened.

(Car III; $M_V = -2.4$, $L/L_\odot \sim 10^{2.9}$), discovered in the Magellanic Satellites Survey (MagLiteS; Drlica-Wagner et al. 2016; Torrealba et al. 2018) with the Dark Energy Camera (DECam; Flaugher et al. 2015) on the Blanco telescope. Li et al. (2018) spectroscopically confirmed Car II to be a dwarf galaxy, and T. S. Li. et al. (2019, in preparation) confirmed Car III as a dwarf galaxy as well. These UFDs are only ~ 20 kpc away from the LMC and also close to the Sun (37.4 and 27.8 kpc for Car II and III, respectively). Thus, they have a relatively large number of bright stars amenable for high-resolution spectroscopic follow-up and chemical abundance measurements.

In this paper, we present a comprehensive chemical abundance analysis of *Magellan*/MIKE spectroscopy of nine stars in Car II and two stars in Car III. Along with Horologium I (Nagasawa et al. 2018), these are currently the only ultra-faint LMC satellites with high-resolution abundance measurements. Section 2 explains the observations, data reduction, and velocity measurements. Section 3 details our abundance analysis. We discuss the formation history of these galaxies in Section 4, highlighting the interesting α -element abundance trends in Section 4.3. We focus on potential signatures of metal-free Pop III stars in Section 5, then summarize and conclude in Section 6.

2. Observations, Data Reduction, and Radial Velocities

Our Car II and III targets were selected to be the brightest radial velocity members from *Magellan*/IMACS, Anglo-Australian Telescope (AAT)/AAO, and Very Large Telescope (VLT)/FLAMES moderate-resolution spectra, including five bright member stars from Li et al. (2018) and five new bright member stars from T. S. Li. et al. (2019, in preparation). In addition, we include one RR Lyrae (RRL) member in Car II identified in Torrealba et al. (2018). We observed these stars with *Magellan*/MIKE (Bernstein et al. 2003) over four separate runs (Tables 1 and 2). Slits of width 0''.5, 0''.7, and 1''.0 were used, depending on the seeing, resulting in typical resolutions of $R \sim 50$ k/40 k, 35 k/28 k, and 28 k/22 k on the blue/red arms of MIKE, respectively. We used 2×2 binning for the 0''.7 and 1''.0 slits and 2×1 binning for the 0''.5 slit. The MIKE data were reduced with CarPy (Kelson 2003).

Table 2
Radial Velocities

Star	Obs. Date	MJD	v_{hel}	N_{ord}	σ_{sys}
CarII-6544	2018 Jan 24	58142.031	470.4	34	1.2
CarII-7872	2018 Nov 13	58435.277	478.5	35	0.5
CarII-5664	2018 Nov 16	58438.207	483.7	34	0.6
CarII-0064	2017 Dec 6	58093.336	475.0	35	0.7
CarII-4704	2018 Nov 13	58435.319	472.4	30	1.0
CarII-9296	2018 Jan 24	58142.059	481.9	30	1.2
CarII-2064	2018 Jan 24	58142.210	473.9	34	0.9
CarII-4928	2018 Jan 23	58141.220	476.1	23	1.4
CarII-V3	2018 Nov 15	58437.247	478.2	18	1.9
CarIII-1120	2018 Jan 24	58142.147	283.7	30	1.1
CarIII-8144	2018 Nov 16	58438.277	280.8	36	0.5

Note. We show one representative velocity measurement per star in our sample. The full table is available online. Note that CarII-6544 is likely a binary star and CarII-V3 is an RRL star, so these have significant velocity variations.

(This table is available in its entirety in machine-readable form.)

We used the code SMHR (Casey 2014)¹⁸ to coadd, normalize, stitch orders, and Doppler correct the reduced spectra for abundance analysis. Data from multiple runs were combined by coadding order by order using a common set of spline knot locations and line masks after adjusting for observed radial velocity. The signal-to-noise ratio (S/N) at the order center closest to rest wavelengths of 4500, 5300, and 6500 Å is given in Table 1. The total integrated time spent on these stars is 34 hr. Note there is significant reddening toward Car II and III ($E(B-V) \sim 0.2$ mag). Figure 1 shows our spectra around the C–H G band, the strongest barium line, and the Mg b triplet.

In general, we reduced all MIKE data from a given observing run together before measuring the radial velocity. The exception is the RRL star CarII-V3, which experiences large radial velocity variations on a short timescale. Using the known pulsation phases (Torrealba et al. 2018), we observed CarII-V3 across phases 0.40–0.55 with five consecutive

¹⁸ <https://github.com/andycasey/smhr>, first described in Casey (2014).

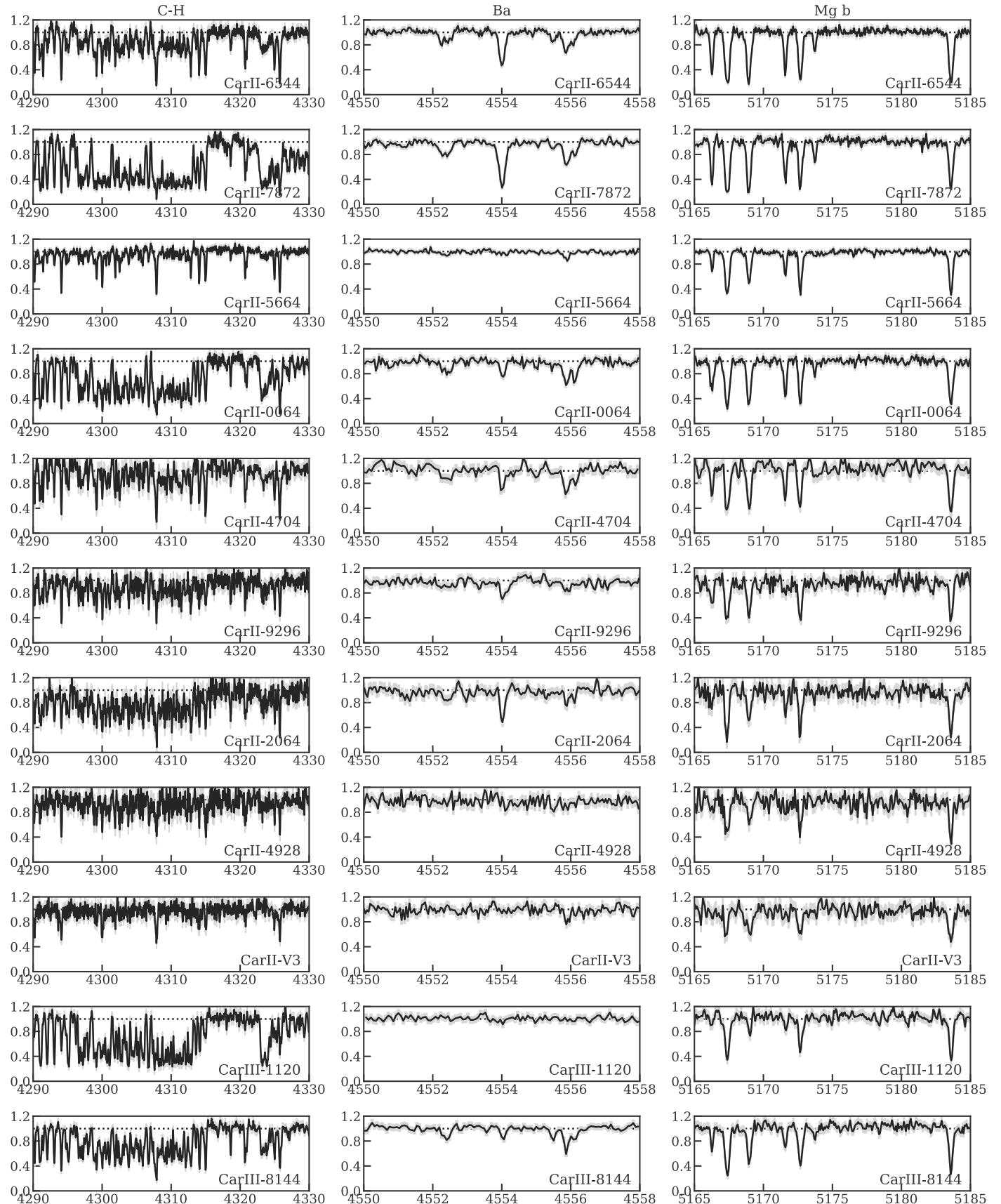


Figure 1. Spectrum of all stars around the C–H G band, the Ba 4554 line, and the Mg b lines. Stars are sorted in order of system and then increasing T_{eff} from top to bottom (same as Table 3). The gray band indicates $\pm 1\sigma$ spectrum noise.

30 minute exposures. Over this phase range, the star has fairly consistent stellar parameters (For et al. 2011), so individual exposures can be coadded after correcting for a velocity offset. We reduced each exposure separately, measured radial velocities for each observation separately using the Mg b triplet, corrected each order to the rest frame, and coadded order by order before stitching orders in SMHR.

Radial velocities are given in Table 2. For the velocity measurements, we rereduced each exposure individually with CarPy. We measured the radial velocities of the 40 orders from 3900 to 6800 Å (order numbers 51–90). Of these, we masked the telluric lines around 6300 Å, discarded three orders from 5820 to 6020 Å because of interstellar Na D absorption, and discarded the bluest order on the red side due to uniformly low S/N. We cross-correlated individual orders of our MIKE spectra against a normalized high-S/N MIKE spectrum of HD 122563. To remove outliers, we iteratively sigma clip orders with velocities that are more than five biweight scales away from the biweight average. The final number of orders for each spectrum is given by N_{ord} in Table 2. Statistical errors for each order were then found by calculating the χ^2 at different velocities and taking $\Delta\chi^2 = 1$ away from the minimum.

Naively, we could combine these measurements by taking a weighted average of all orders to get a final average velocity and, in principle, reach an extremely high velocity precision of $\sim 0.1 \text{ km s}^{-1}$. However, systematic effects dominate both the velocity measurement and error. For example, MIKE is not attached to the instrument rotator and, until recently, did not have an atmospheric dispersion compensator. At high airmasses, atmospheric refraction in the narrow slit direction causes systematic velocity offsets as a function of wavelength that can be as large as $2\text{--}3 \text{ km s}^{-1}$. We will correct for these effects in later work, but such velocity differences do not impact the abundance analyses that are the focus of this paper. Thus, for now in Table 2, we provide the radial velocity of each individual spectrum computed by an inverse-variance weighted average of all N_{ord} orders. The systematic error is the weighted standard deviation of those orders and dominates over the $\sim 0.1 \text{ km s}^{-1}$ statistical uncertainty.

3. Abundance Analysis

3.1. Abundance Analysis Details

We performed a standard 1D LTE analysis using the 2017 version of the 1D LTE radiative transfer code MOOG (Sneden 1973; Sobeck et al. 2011)¹⁹ and the Castelli & Kurucz (2004) ATLAS model atmospheres. We used SMHR to measure equivalent widths, interpolate model atmospheres, and run MOOG.

For the red giant branch (RGB) stars, stellar parameters were derived spectroscopically. Briefly, we start assuming α -enhanced $[\alpha/\text{Fe}] = +0.4$ model atmospheres. The effective temperature, surface gravity, and microturbulence (T_{eff} , $\log g$, and ν_t) were determined by balancing excitation, ionization, and line strength for the Fe lines, respectively. We then applied the T_{eff} correction from Frebel et al. (2013) to place the measurements on a photometric temperature scale and redetermined $\log g$ and ν_t . After this initial determination, if the star turned out to have low Mg abundances, we switched to $[\alpha/\text{Fe}] = 0$ atmospheres and redetermined the stellar parameters. Statistical stellar parameter

Table 3
Stellar Parameters

Star	T_{eff} (K)	$\log g$ (dex)	ν_t (km s $^{-1}$)	[M/H]
CarII-6544	4330 ± 152	0.40 ± 0.31	2.75 ± 0.26	-2.65 ± 0.09
CarII-7872	4380 ± 155	0.75 ± 0.32	2.32 ± 0.27	-2.48 ± 0.11
CarII-5664	4430 ± 155	0.45 ± 0.31	2.34 ± 0.25	-3.50 ± 0.06
CarII-0064	4630 ± 153	1.15 ± 0.32	2.31 ± 0.27	-2.20 ± 0.07
CarII-4704	4720 ± 160	1.30 ± 0.31	1.97 ± 0.27	-2.19 ± 0.09
CarII-9296	4810 ± 205	1.40 ± 0.37	1.90 ± 0.34	-2.87 ± 0.15
CarII-2064	5300 ± 200	2.70 ± 0.35	2.15 ± 0.32	-2.35 ± 0.17
CarII-4928	5065 ± 236	2.35 ± 0.46	2.10 ± 0.34	-3.00 ± 0.20
CarII-V3	6100 ± 330	1.75 ± 0.27	3.20 ± 0.28	-2.70 ± 0.21
CarIII-1120	4500 ± 216	1.50 ± 0.34	1.85 ± 0.32	-3.89 ± 0.14
CarIII-8144	4990 ± 162	2.20 ± 0.32	1.75 ± 0.27	-2.25 ± 0.08

uncertainties are found following Ji et al. (2019b), and we adopt systematic uncertainties of 150 K for T_{eff} , 0.3 dex for $\log g$, and 0.2 km s $^{-1}$ for ν_t due to uncertainties in the Frebel et al. (2013) temperature calibration. The statistical and systematic uncertainties were added in quadrature to obtain the total stellar parameter uncertainties in Table 3.

We used a combination of equivalent widths and spectral syntheses to measure the abundances of individual lines. We also determined statistical and systematic abundance uncertainties for each individual feature. For lines measured using equivalent widths, we propagated the 1σ equivalent width uncertainty into a 1σ statistical abundance uncertainty. For lines measured using syntheses, we increased the element abundance until $\Delta\chi^2 = 1$, also corresponding to a 1σ statistical uncertainty. These uncertainties account for the continuum placement uncertainty (see Appendix for details). For the systematic uncertainties, we varied each stellar parameter (T_{eff} , $\log g$, ν_t , [M/H]) individually by its error and remeasured the abundance. The total systematic uncertainty is the quadrature sum of the individual stellar parameter uncertainties. Finally, the total abundance uncertainty for an individual line is the quadrature sum of the statistical and systematic uncertainty. Individual line measurements and uncertainties are found in Table 4.

We use inverse-variance weighted averages to combine lines into a final abundance. Because we have included a detailed account of line-by-line uncertainties, this automatically down-weights lines in regions of low spectral S/N, saturated lines that are sensitive to small equivalent width variations, and lines that are particularly sensitive to stellar parameters. We verified that the weighted averages are usually only a few hundredths of a dex different from the unweighted averages. The exception is elements with few measurable lines, like Si and Al, where some lines are much lower quality than others. See Appendix for detailed equations.

The [X/Fe] ratios are derived by taking the ratios of common ionization states (e.g., [Mg I/Fe I], [Ti II/Fe II]). This mostly (though not always) results in smaller [X/Fe] errors than [X/H] errors, since some stellar parameter differences cancel out. We also consistently propagate stellar parameter uncertainties for [X/Y] ratios, such as [Mg/Ca].

Upper limits were derived by spectrum synthesis. For a given feature, we fit a synthetic spectrum that well matched the observed spectrum to determine a reference χ^2 and local spectrum smoothing. Then, holding the continuum and smoothing fixed, we increased the abundance until $\Delta\chi^2 = 25$. This is

¹⁹ <https://github.com/alexji/moog17scat>

Table 4
Line Measurements

Star	λ	ID	χ	$\log gf$	EW	σ (EW)	ul	$\log \epsilon$	σ_i	$\sigma_{i,\text{stat}}$	$\sigma_{i,\text{sys}}$	$\delta_{i,T_{\text{eff}}}$	$\delta_{i,\log g}$	δ_{i,u_i}	$\delta_{i,[\text{M}/\text{H}]}$	σ_{cont}
CarII-0064	5183.60	12.0	2.72	-0.17	251.8	7.0	0	5.21	0.28	0.04	0.28	+0.23	-0.14	-0.08	-0.01	0.07
CarII-0064	5528.40	12.0	4.35	-0.50	87.0	4.7	0	5.35	0.14	0.07	0.12	+0.10	-0.05	-0.05	-0.01	0.04
CarII-0064	4323.00	106.0	Syn.	Syn.	0	6.16	0.30	0.03	0.30	+0.28	-0.07	+0.01	+0.03	...
CarII-0064	4554.00	56.1	0.00	0.16	Syn.	Syn.	0	-2.52	0.17	0.08	0.15	+0.10	+0.11	-0.01	+0.01	0.03
CarII-0064	4129.70	63.1	0.00	0.22	Syn.	Syn.	1	-1.89

Note. See Appendix for column details.

(This table is available in its entirety in machine-readable form.)

formally a 5σ upper limit but does not include uncertainties in continuum placement.

3.2. Abundance Corrections

Various systematics can affect the 1D LTE abundances of red giants. We tabulate several abundance corrections in Table 5, which are the average of line-by-line corrections. These corrections have been applied in all figures but not in Tables 4 or 6.

Carbon is systematically converted to nitrogen in evolved red giants due to CN cycling. We estimate the natal carbon abundances of these stars with the corrections from Placco et al. (2014).²⁰ Hotter stars have no correction, while for cooler/more evolved stars, the correction can be as large as +0.75 dex. We use the default correction grid assuming $[\text{N}/\text{Fe}] = 0$, but changing $[\text{N}/\text{Fe}]$ makes minimal difference. Note that we assume all our stars are on the RGB, but if we had red clump or asymptotic giant branch (AGB) stars in our sample, they would have larger carbon corrections than applied here.

Only the Na D lines are available for sodium abundances, and these can have fairly large negative NLTE corrections. We apply Na corrections from Lind et al. (2011),²¹ which range from -0.13 to -0.48 dex. For CarII-6544, CarII-7872, and CarII-5664, we set $\log g = 1$ to avoid the edge of the corrections grid.

In our stars, Mg is marginally affected by NLTE effects. However, since Mg will be a very important element later, we tabulate the NLTE corrections just to show that they are only affected by <0.04 dex (Osorio et al. 2015; Osorio & Barklem 2016). For several stars (CarII-6544, CarII-4704, CarII-0064, CarII-5664, and CarII-7872), we set $\log g = 1.5$ to avoid the edge of the corrections grid. Note that we have used the two high-equivalent-width Mg b lines in all of our Mg abundances, but removing these two lines everywhere does not significantly affect our RGB star abundances.

Other elements that are known to have significant NLTE corrections include Al, Mn, K, and Fe. For these elements, we do not calculate star-by-star corrections but instead just estimate the magnitude and direction of a typical correction. If desired, the effect of these corrections can be approximated by adding the correction to the relevant abundance, as well as adding the total correction in quadrature to the total abundance error, but we do not do so here.

For aluminum, we measured the 3944 and 3961 Å lines, which are heavily affected by NLTE in cool metal-poor stars, as well as being in the wings of strong lines, so we only

Table 5
Abundance Corrections

Star	CH Corr.	Na Corr.	Mg Corr.
CarII-6544	+0.75	-0.16	+0.03
CarII-7872	+0.60	-0.13	+0.03
CarII-5664	+0.74	-0.23	+0.05
CarII-0064	+0.61	-0.23	+0.04
CarII-4704	+0.62	-0.23	+0.03
CarII-9296	+0.49	-0.26	+0.04
CarII-2064	+0.01	-0.48	+0.03
CarII-4928	+0.01	-0.32	+0.03
CarII-V3
CarIII-1120	+0.39	-0.43	+0.02
CarIII-8144	+0.01	-0.47	+0.02

estimate the abundance corrections. We examined the corrections grid from Nordlander & Lind (2017)²² for these lines. Half of our stars are cooler and have lower $\log g$ than the grid range. The abundance corrections for the 3961 Å line tend to be large and positive, from +0.7 to +1.5 dex. The corrections for 3944 Å are more moderate, from +0.0 to +0.5 dex. The corrections for these lines tend to go in opposite directions, such that averaging corrections for these lines in the warmer stars ($T_{\text{eff}} \gtrsim 4800$ K) gives corrections in a smaller range from +0.5 to +0.7 dex. However, this also tends to make the individual 3944 and 3961 Å abundances more discrepant. Given these uncertainties, we caution against overinterpretation of our Al abundances or trends.

For manganese, we always use the resonant triplet near 4030 Å, as well as redder lines (e.g., 4754 and 4783 Å) when detected. Bergemann et al. (2019) have recently published grids of Mn corrections, showing overall corrections of about +0.4 to +0.6 dex, though the corrections are likely larger for cooler and metal-poor stars. As our Mn abundances just fall within the overall halo trend (which is also not corrected for NLTE), we will not discuss this further.

For potassium, we can measure the 7699 Å line in all stars. The 7665 Å line was also clear of telluric lines for a few stars and, when measurable, is always consistent with the 7699 Å line. There are negative NLTE corrections for K that could be as large as -0.9 dex (Ivanova & Shimanskii 2000), although Reggiani et al. (2019) recently calculated grids of corrections that are more typically -0.0 to -0.4 dex in our stellar parameter range.

The Fe I abundances are affected by NLTE effects, with corrections typically +0.2 to +0.3 dex in our parameter range (e.g., Bergemann et al. 2012; Mashonkina et al. 2016;

²⁰ <http://vpalcco.pythonanywhere.com/>

²¹ www.inspect-stars.com

²² <https://www.mso.anu.edu.au/~thomasNLTE/>

Ezzeddine et al. 2017). Our temperature-correction procedure partially accounts for these effects, though not completely (Frebel et al. 2013; Ji et al. 2016b). We have decided not to apply Fe corrections so as to be able to compare our Fe measurements to literature values, which are essentially all done in LTE.

Finally, we note that Ca can be affected by NLTE as well (Mashonkina et al. 2016). The available grids do not span our whole stellar parameter space,²³ but the available corrections are about +0.1 dex for our stars. We have not applied this correction.

3.3. RRL Abundance Analysis

Stellar parameters for the RRL star CarII-V3 were determined by examining the phase-parameter relations in For et al. (2011). As our observations are between phases 0.40 and 0.55, stellar parameters are expected to be fairly stable over all exposures. We adopted initial stellar parameters of $T_{\text{eff}} = 6000 \pm 100$ K, $\log g = 1.80 \pm 0.2$ dex, and $\nu_t = 3.00 \pm 0.20$ km s $^{-1}$, where the error bars are adopted systematic uncertainties based on scatter in the For et al. (2011) values. Then, we measured equivalent widths by fitting Gaussian profiles to the line list from For & Sneden (2010; rather than our usual line list, which is optimized for red giants). To slightly improve Fe excitation, ionization, and line strength balance from 28 Fe I lines and 10 Fe II lines, we adjusted the stellar parameters to $T_{\text{eff}} = 6150$ K, $\log g = 1.75$ dex, and $\nu_t = 3.15$ km s $^{-1}$, resulting in $[\text{M}/\text{H}] = -2.70$. Total stellar parameter and abundance uncertainties were then determined the same way as the RGB stars. We do not apply any abundance corrections for this star, as the correction grids are computed for cool giants. CarII-V3 is one of the most metal-poor RRLs ever studied spectroscopically, with similar $[\text{Fe}/\text{H}]$ as X Ari and the most Fe-poor RRLs in the LMC (For et al. 2011; Haschke et al. 2012; Nemec et al. 2013).

3.4. Abundance Summary

Our full abundance results are tabulated in Table 6 (Appendix) and Figures 2 and 4. We compare the results to halo stars (small gray points; Abohalima & Frebel 2018) and other UFD measurements in the literature. The UFD literature compilation includes Bootes I (Feltzing et al. 2009; Norris et al. 2010; Gilmore et al. 2013; Ishigaki et al. 2014; Frebel et al. 2016), Bootes II (Ji et al. 2016a), Canes Venatici II (François et al. 2016), Coma Berenices (Frebel et al. 2010), Grus I (Ji et al. 2019b), Hercules (Koch et al. 2008, 2013), Horologium I (Nagasawa et al. 2018), Leo IV (Simon et al. 2010; François et al. 2016), Pisces II (Spite et al. 2018), Reticulum II (Ji et al. 2016c; Roederer et al. 2016), Segue 1 (Frebel et al. 2014), Segue 2 (Roederer & Kirby 2014), Triangulum II (Kirby et al. 2017; Venn et al. 2017; Ji et al. 2019b), Tucana II (Ji et al. 2016b; Chiti et al. 2018a), Tucana III (Hansen et al. 2017; Marshall et al. 2019), and Ursa Major II (Frebel et al. 2010). We reiterate that, throughout this paper, the error bars for Car II and III include full propagation of the line-by-line statistical and stellar parameter uncertainties.

The RRL star CarII-V3 generally has consistent abundances with the RGB stars, although there are fewer lines and only

moderate S/N, so the abundance uncertainties for this star are fairly large. The main outlier is the Si abundance, which is unusually low but has a large uncertainty, as it is measured only from the 3905 Å line. Given the abundance similarities to other stars in Car II, we will place this star's abundances on the same footing as RGB stars when lines are detected.

C, N, O. Carbon abundances are derived from synthesizing the CH bands at ~ 4300 –4325 Å. The CO molecular equilibrium affects CH abundances, and we always assume the MOOG default of $[\text{O}/\text{Fe}] = 0$, even when O is measured independently. Literature measurements suggest that $[\text{O}/\text{Fe}]$ is typically >0.5 (e.g., Brown et al. 2014). If we used $[\text{O}/\text{Fe}] = +1.0$ instead, $[\text{C}/\text{Fe}]$ would typically increase by +0.08 dex with a star-to-star scatter of 0.08 dex, but we keep the MOOG default for consistency with previously analyzed literature stars. Nitrogen is derived from fitting CN bands at ~ 3850 Å after fixing the CH abundance.

In two relatively cool and metal-rich stars, we detect the two forbidden oxygen lines at ~ 6300 Å. These can only be measured when the O abundance is very high, so they are probably a biased sample of measurements. The stronger 6300 Å line was deblended from telluric absorption, and the weaker 6363 Å line can be affected by a wide calcium ionization feature (e.g., Barbuy et al. 2015). However, in both cases, the two different lines give very close abundances. We include oxygen upper limits for all stars (including the two detections) in the machine-readable version of Table 4 from the 6300 Å line.

α -elements: Mg, Si, Ca. The α -element abundances are determined from equivalent widths in all stars. Magnesium is determined from five to seven lines, including the Mg b lines in all stars (except CarII-V3, where only the Mg b lines can be measured). The Mg b lines are quite strong and saturated but give similar abundances as the weaker lines for all stars. The Si is measured from both the 3905 and 4102 Å lines, but these are rather poor-quality lines. The 3905 Å line is fairly saturated, and the 4102 Å line is in a Balmer wing. Usually, Ca is measured from 10 to 20 lines, with three exceptions: the warmer and more Fe-poor stars CarII-4928 and CarIII-1120 have only two and one Ca line, respectively, and only the strong 4226 Å line is detected in the RRL CarII-V3. We do not use the 4226 Å line in any of the RGB stars due to large and uncertain NLTE corrections (e.g., Sitnova et al. 2019).

Odd-Z elements: Na, Al, K, Sc. We use equivalent widths to measure sodium abundances from the two Na D lines, which have been corrected for NLTE effects. We synthesize the 3944 and 3961 Å Al lines, which are both very strong and subject to NLTE effects, so our Al abundances are very uncertain. The K abundances are mostly from the 7699 Å line, although occasionally the 7665 Å line is not blended with tellurics. The Sc abundances are mostly measured with spectral synthesis from five lines at $4246 \text{ \AA} < \lambda < 4415 \text{ \AA}$, though the redder line abundances (e.g., 5031 and 5526 Å) agree.

CarII-0064 is a significant low-Sc outlier in Car II with $[\text{Sc}/\text{Fe}] \approx -1$ (Figure 2). We plot two Sc line spectra in Figure 3, along with its synthetic fit and two other stars that have higher Sc abundances. The Sc abundance is clearly lower in CarII-0064, though visually not as much as would be expected from Figure 2. This is because each individual line difference is significant at $\lesssim 2\sigma$, but they are all consistent, and the combination of five to six Sc lines reduces the uncertainty. Also note that the $[\text{Sc}/\text{Fe}]$ abundance error is smaller due to

²³ <http://spectrum.inasan.ru/nLTE/>

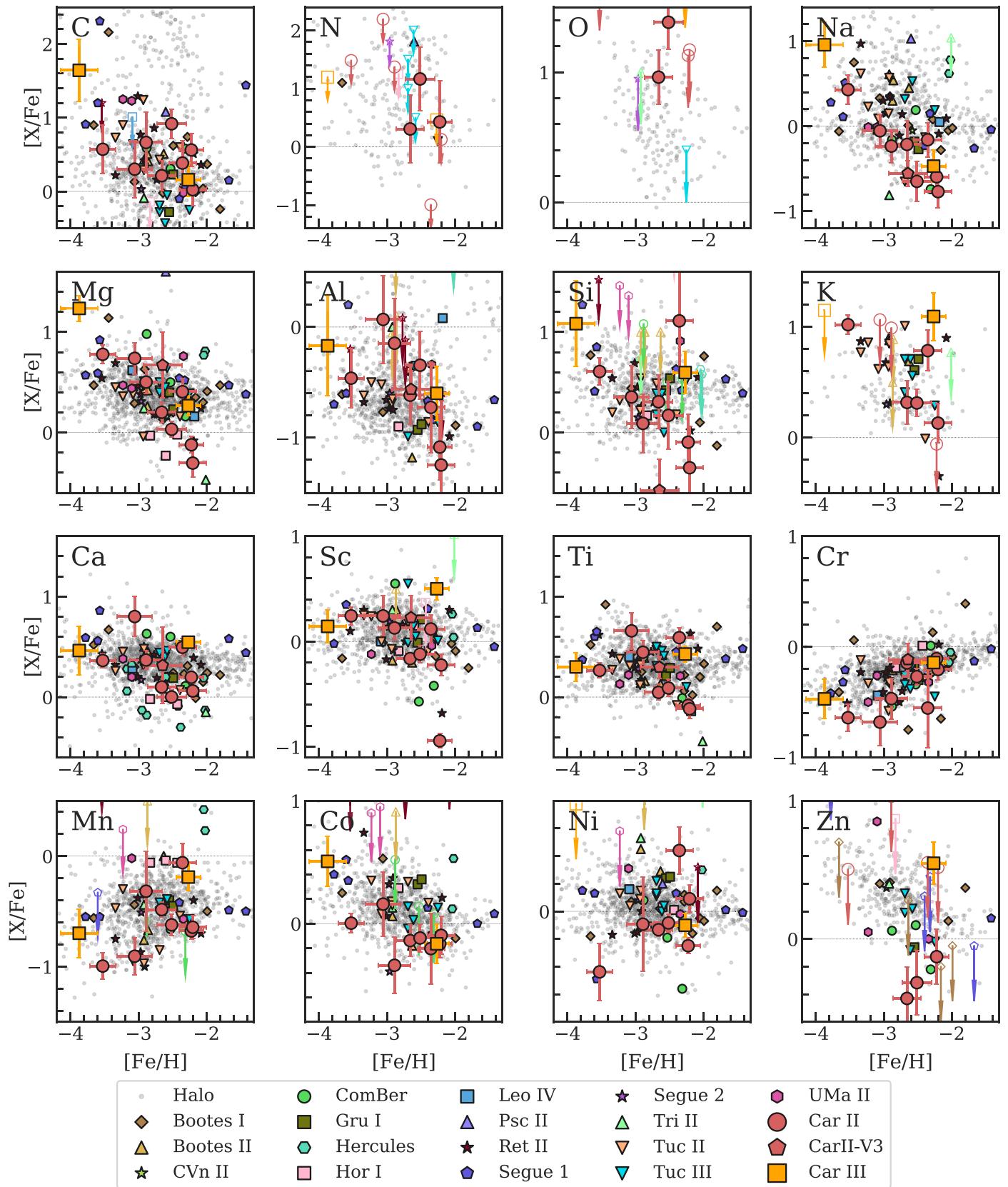


Figure 2. The $[\text{X}/\text{Fe}]$ ratios for most measured elements. Car II and III are shown as a large red circle and orange square, respectively, with error bars. The RRL CarII-V3 is shown separately as a red pentagon. Other UFDs are shown as small colored symbols according to the legend. Upper limits are indicated by an open symbol with a downward-pointing arrow. The JINAbase halo sample is shown as small gray points in the background.

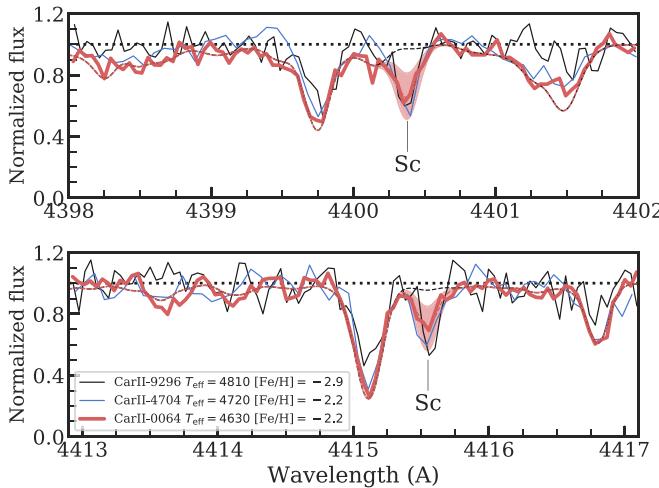


Figure 3. Spectrum of the low-Sc outlier CarII-0064 around two Sc lines compared to two other Car II stars with similar temperature but lower Sc abundance. The Sc line is deficient in CarII-0064 compared to these other stars, despite this star being somewhat cooler. Note that there is C-H absorption in CarII-0064 near the 4400 Å line. We also show the synthetic spectrum fit to the Sc line for CarII-0064 as a thin dashed red line, a ± 0.5 dex difference from the synthetic fit as a shaded region, and a synthesis with no Sc as a dashed black line.

correlated uncertainties in stellar parameters. Such low-Sc abundances have previously been seen in “iron-rich” stars (those with overall low $[X/Fe]$ ratios; e.g., Cohen & Huang 2010; Cohen et al. 2013; Yong et al. 2013). However, this cannot explain CarII-0064, because it is an outlier from the overall Car II trend only in $[Sc/Fe]$. Similarly Sc-deficient stars have been found in the bulge, where it has been argued that this signature may indicate unusually old stars (Casey & Schlaufman 2015), but we see no sign of this in the more Fe-poor stars in Car II. It is unclear to us how to interpret this star’s extreme Sc abundance.

Fe-peak elements: Ti , Cr , Mn , Co , Ni , Zn . We use equivalent widths to measure abundances for both ionization states of titanium, but we adopt the $Ti\,II$ abundances everywhere as our default; it is measured in all of our stars, has more and stronger lines, and is less susceptible to NLTE effects.

The Fe-peak elements closely follow the halo trends within their abundance uncertainties. There are minor deviations that are all significant at $<2\sigma$, so we do not concern ourselves with these further other than to comment that Zn could be moderately enhanced in Car III and moderately deficient in Car II.

Neutron-capture elements: Sr , Ba . These elements have low abundances or upper limits, similar to most other UFDs. The nucleosynthetic origin of these very low Sr and Ba abundances remains unknown (it is generally not even clear if they are from the r - or s -process; see Ji et al. 2019b for an extensive discussion), but it appears to be unique to UFDs and occasional halo stars that are presumably stripped from UFDs. Given the low abundance of neutron-capture elements, no other neutron-capture elements could be detected, so we place $[Eu/Fe]$ upper limits and show $[Ba/Eu]$ in Figure 4.

There are two stars in Car II with relatively high $[Ba/Fe] \gtrsim -1$ compared to the other Car II stars. One of these relatively Ba-rich stars, CarII-7872, also has a low Eu upper limit that results in $[Ba/Eu] \gtrsim 0$, suggesting that its Ba is predominantly from the s -process (e.g., Sneden et al. 2008). We discuss this large barium scatter in Section 4.4.

4. Formation History of Car II and III

4.1. Car II and III Are Dwarf Galaxies

Low-luminosity stellar systems are classified as either dwarf galaxies or star clusters. Dwarf galaxies are generally more spatially extended than clusters, with velocity dispersions implying significant dark matter content and nonzero metallicity (or, more specifically, iron-peak abundance) dispersions (Willman & Strader 2012). Faint dwarf galaxies also tend to display very low abundances of neutron-capture elements (e.g., Ji et al. 2019b), while globular clusters have light-element anticorrelations associated with hot bottom burning (e.g., Bastian & Lardo 2018).

Both Car II and III are clearly dwarf galaxies and not globular clusters. Their half-light radii and luminosities place them within the dwarf galaxy morphological locus (Torrealba et al. 2018). Car II displays both a significant velocity and metallicity dispersion from medium-resolution data (Li et al. 2018). Our two Car III stars have $[Fe/H]$ values that differ by almost 2 dex, definitively establishing a significant metallicity dispersion. We have also now resolved the velocity dispersion (T.S. Li. et al. 2019, in preparation). The neutron-capture elements Sr and Ba are low in both systems, like nearly every other UFD (Figure 4).

These criteria alone already show that Car II and III are galaxies, but as a final confirmation, we show that there are no light-element anticorrelations. Figure 5 shows these relations for our stars. In the top two panels, we show $Na-Mg$ and $Al-Mg$ for our UFD stars (symbols as in Figure 2) and globular cluster stars as purple circles (from references Gratton et al. 2006; Carretta et al. 2007, 2009; Cohen & Kirby 2012). Most globular clusters do not show significant dispersion in $[Mg/Fe]$, but those that do always display an anticorrelation in $Na-Mg$ and $Al-Mg$. In contrast, there is very clearly a positive correlation for these elements in both Car II and III. Note that Na and Mg have NLTE corrections applied, while the Al corrections should, on average, provide an offset and are unlikely to turn a strong positive $Mg-Al$ correlation into an anticorrelation. The bottom panel of Figure 5 shows the $Mg-K$ anticorrelation found in NGC 2419 (Mucciarelli et al. 2012), which is not present in Car II. However, our two stars in Car III (including one K upper limit) do not rule out an $Mg-K$ anticorrelation in this system.

4.2. Car II and III Are Consistent with Being Accreted along with the LMC/SMC

Li et al. (2018) showed that the positions and radial velocities of both Car II and Car III were consistent with having accreted with the LMC, according to the Jethwa et al. (2016) model. Kallivayalil et al. (2018) then added proper motion data from *Gaia* DR2 (Gaia Collaboration et al. 2018, 2016), finding that Car II and III are also likely LMC satellites based on LMC analogs in the Aquarius simulations (Springel et al. 2008; also see Sales et al. 2017; Simon 2018; Erkal & Belokurov 2019). Kinematically, it thus appears likely that both Car II and Car III entered the Milky Way with the LMC/SMC system, although Car II is toward the edge of the likely region due to its high velocity. Kallivayalil et al. (2018) also associated Hyi I and Hor I with the LMC.

Thus, Car II and III, along with Hor I (Nagasawa et al. 2018), can be studied in contrast to other UFDs to see if abundance ratios have any environmental dependence.

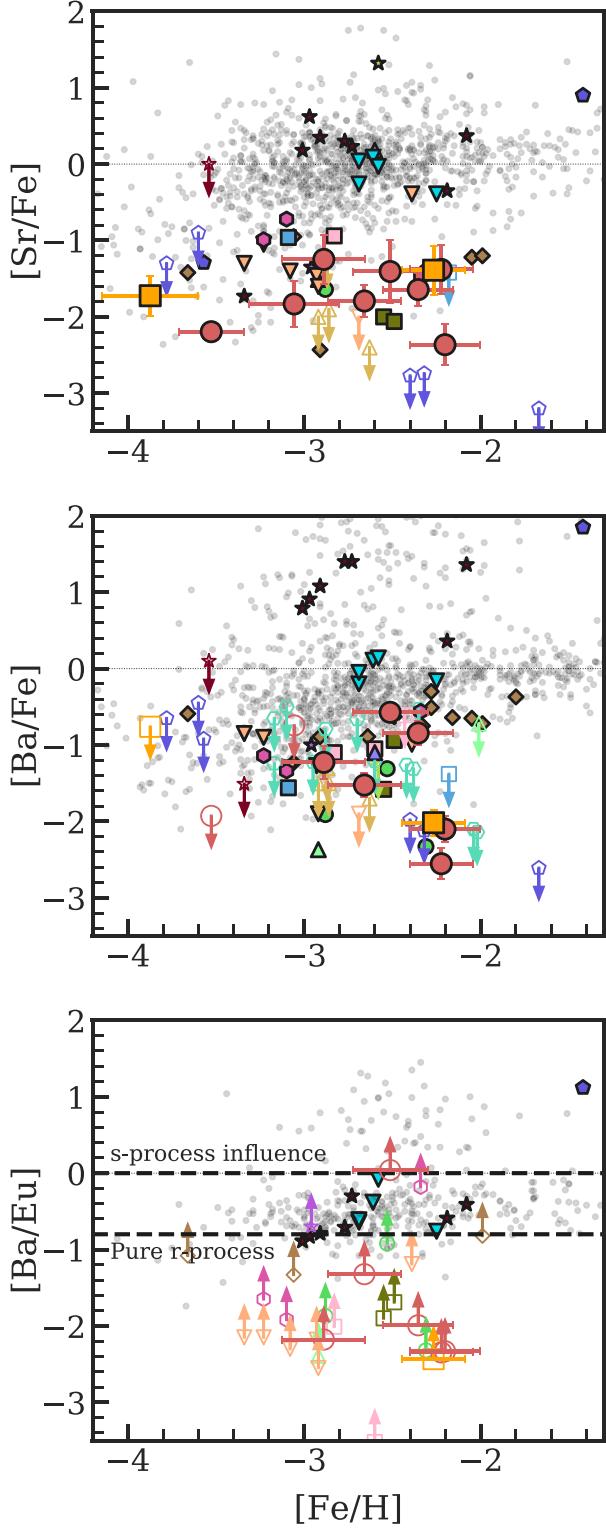


Figure 4. Neutron-capture element abundances in Car II (large red circles), Car III (large orange squares), halo stars (gray points), and other UFDs (small colored points; see Figure 2 for legend). The top and middle panels show $[\text{Sr}/\text{Fe}]$ and $[\text{Ba}/\text{Fe}]$; the bottom panel shows $[\text{Ba}/\text{Eu}]$. We draw lines at $[\text{Ba}/\text{Eu}] = -0.8$ and 0.0 indicating a pure r -process ratio and an s -process-influenced ratio (Sneden et al. 2008). Car II and III match most other UFDs as being deficient in Sr and Ba. Car II displays significant scatter in $[\text{Ba}/\text{Fe}]$ at $[\text{Fe}/\text{H}] \sim -2.5$. One Ba-rich star in Car II has $[\text{Ba}/\text{Eu}] > 0$ and thus likely has significant s -process enrichment.

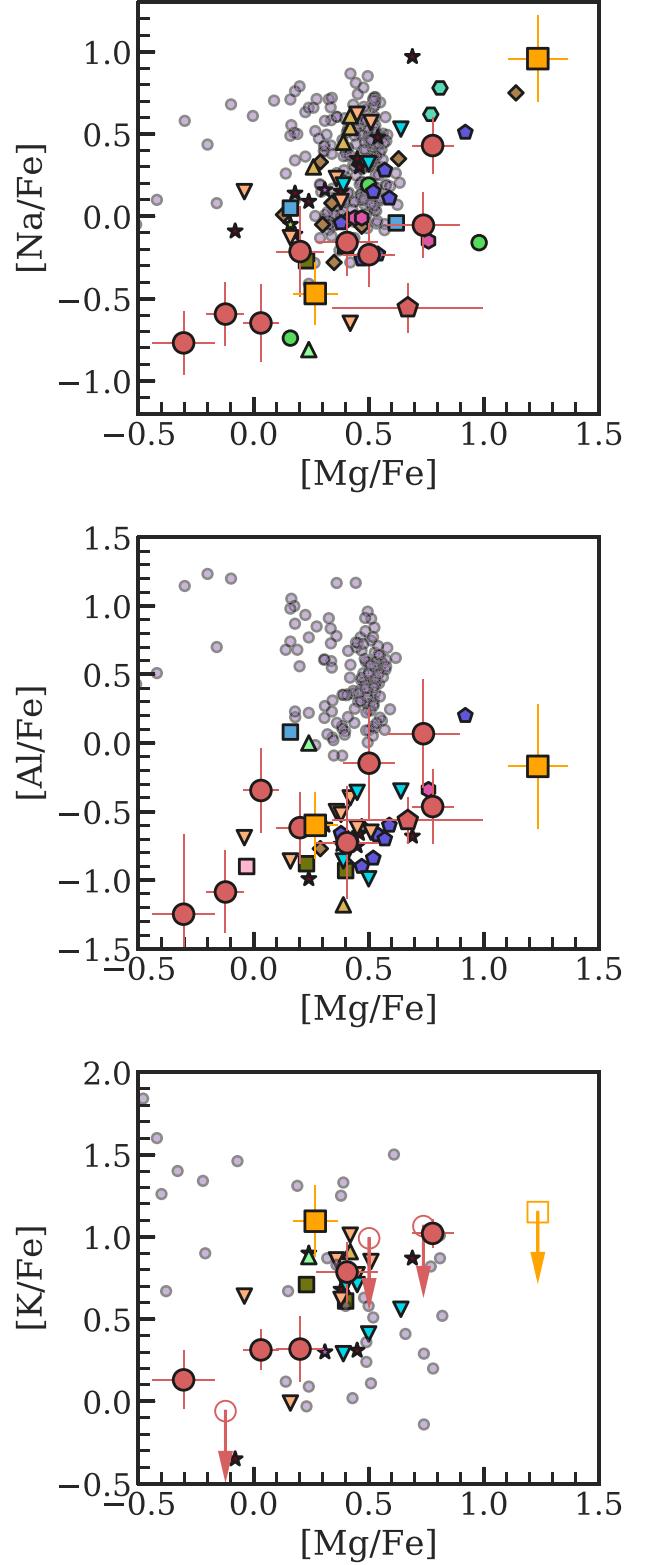


Figure 5. The Mg–Na, Mg–Al, and Mg–K abundance patterns. Car II is the red circles/pentagons, Car III is the orange squares, globular cluster stars are the small purple circles, and other UFDs are the small colored points (same as Figure 2). In globular clusters, Mg–Na and Mg–Al are anticorrelated, while in both Car II and Car III, these elements are clearly correlated. The Mg–K is anticorrelated in the globular cluster NGC 2419, and there is no evidence for such in Car II. The light-element correlations confirm that Car II and III are dwarf galaxies and not globular clusters.

Nagasawa et al. (2018) pointed out that the three stars in Hor I have unusually low Mg and Ca, with one possible explanation being that LMC satellites might have typically different enrichment histories compared to Milky Way UFDs. Figure 2 does not suggest that Car II or Car III obviously deviate from the typical abundance scatter of other UFDs, including for Mg and Ca. The unusually low Mg and Ca in Hor I thus likely has some other origin.

4.3. α -element Evolution: Time-delay Scenario or IMF Variations?

4.3.1. α -element Abundance Ratios in Car II and III

The α -elements (O, Mg, Si, and Ca) are primarily produced in core-collapse supernovae (CCSNe) and thus tend to be enhanced at low [Fe/H]. After a delay of 100–1000 Myr (Maoz et al. 2014), Type Ia supernovae (SNe Ia) begin to add Fe-peak elements, causing a “knee” in $[\alpha/\text{Fe}]$ versus [Fe/H] (Tinsley 1979). In this time-delay scenario, the location of the knee can be interpreted as an overall star formation timescale for a galaxy (e.g., Tolstoy et al. 2009; Kirby et al. 2011). Figure 2 shows clear downward trends in [Mg/Fe] and [Ca/Fe] versus [Fe/H] for both Car II and Car III, with a possible knee at [Fe/H] ~ -2.8 for Car II that would indicate very slow chemical evolution in this low-mass galaxy.

However, there is a striking difference in the size of the trend for [Mg/Fe] and [Ca/Fe]: [Mg/Fe] declines by over 1 dex, while [Ca/Fe] declines by only about 0.4 dex. We will focus primarily on Car II, because Car III has only two stars, and the more Fe-poor star has only one Ca line. To clarify the Mg and Ca difference, in the top panel of Figure 6, we plot [Mg/Ca] versus [Fe/H], where [Mg/Ca] declines from about +0.4 to -0.4 as [Fe/H] increases from -3.5 to -2.2 . These extreme [Mg/Ca] ratios are often interpreted as variations in the high-mass end of the initial mass function (IMF). Stars with high [Mg/Ca] ratios are typically associated with enrichment by very massive stars with $M > 20$ – $30 M_{\odot}$ (e.g., Norris et al. 2000; Cohen et al. 2007; Koch et al. 2008; also see Section 5.2). Stars with [Mg/Ca] < 0 form out of gas enriched by lower-mass CCSN progenitors with $M \lesssim 15 M_{\odot}$ (e.g., Tolstoy et al. 2003; McWilliam et al. 2013). The variable [Mg/Ca] ratios in Car II may thus indicate that the α -elements in this galaxy are tracing changes in the high-mass end of the IMF. Indeed, the low-mass end of the IMF in UFDs has previously been shown to vary between different UFDs (Geha et al. 2013; Gennaro et al. 2018), which tantalizingly hints that the high-mass end of the IMF might vary as well (although the low-mass IMF varies from galaxy to galaxy, while here we consider time variations within a single galaxy, so the mechanisms may not be related).

In the middle and bottom panels of Figure 6, we plot [Mg/H] and [Ca/H] versus [Fe/H], which shows that there may actually be two phases of [Mg/Ca] evolution: from [Fe/H] = -3.6 to -3.0 , this is primarily driven by a smaller increase in [Mg/H] than [Ca/H], while from [Fe/H] = -3.0 to -2.2 , [Mg/H] stays mostly flat while [Ca/H] increases. The first phase unambiguously shows that Car II has been enriched by at least two different masses of CCSNe: the most Fe-poor star in Car II has high [Mg/Ca] ratios suggesting enrichment by high-mass stars, but it has lower [Mg/H] than the higher-metallicity stars. Since SNe Ia produce negligible Mg, this means that CCSNe with [Mg/Ca] ~ 0 must have enriched Car II after the formation of

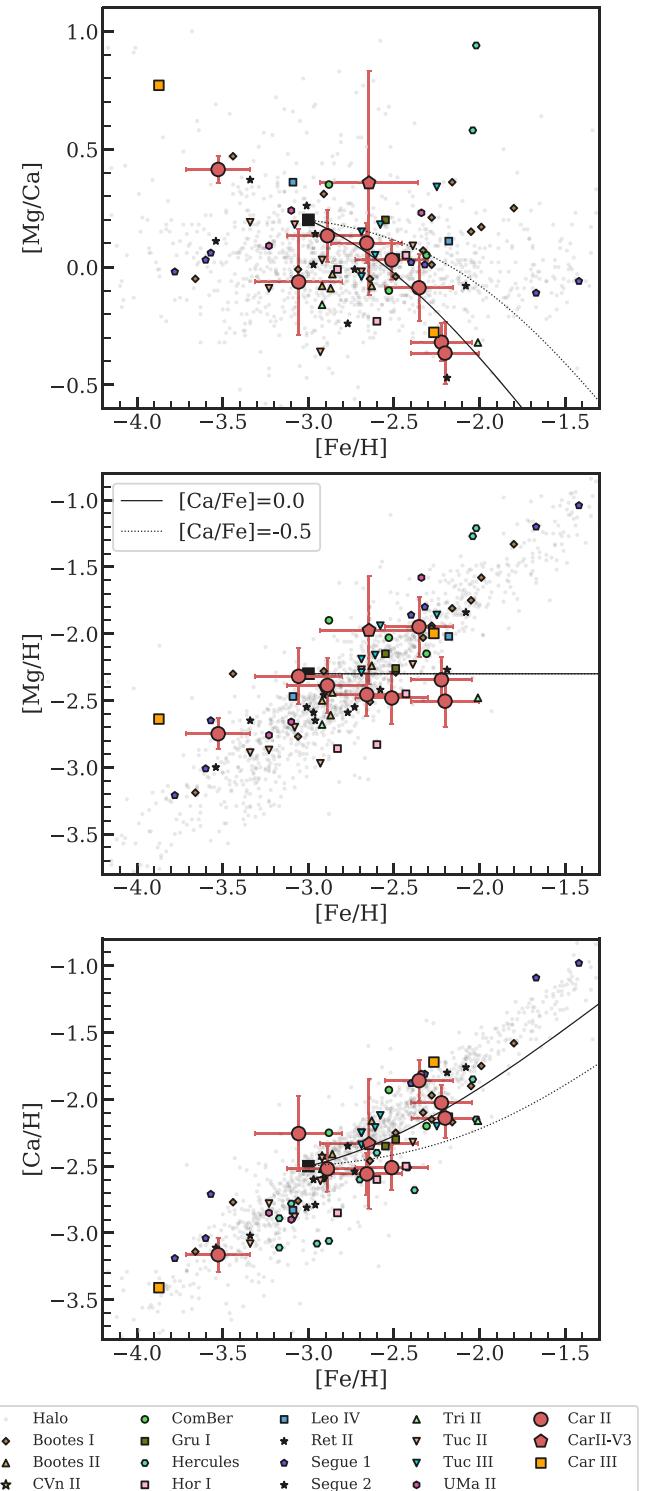


Figure 6. Top panel: [Mg/Ca] vs. [Fe/H] for UFDs (colored points) and halo stars (small gray points). We focus here particularly on Car II (large red points with error bars). Middle and bottom panels: [Mg/H] and [Ca/H] as a function of metallicity. In all panels, the solid and dotted black lines show tracks of SN Ia-only enrichment for two SN Ia Ca yields, starting at the black square.

the most Fe-poor star. This could potentially be evidence of a transition from very massive Pop III stars to regular-mass Pop II CCSNe.

The second phase of evolution could be attributed to either IMF variation or SN Ia enrichment. To illustrate this, we show

an extremely simple chemical evolution track in Figure 6. First, we set an initial $[\text{Mg}/\text{H}]$, $[\text{Ca}/\text{H}]$, and $[\text{Fe}/\text{H}]$ that matches the $[\text{Mg}/\text{Ca}]$ ratio at $[\text{Fe}/\text{H}] = -3$ (black square). Then, we assume a fixed $[\text{Ca}/\text{Fe}]$ yield and negligible Mg yield for SNe Ia (Kirby et al. 2019) and compute the evolution of Mg, Ca, and Fe assuming no more CCSNe and no gas accretion/expulsion. Kirby et al. (2019) recently made an empirical measurement of the SN Ia $[\text{Ca}/\text{Fe}]$ yield in larger dSph galaxies, finding values in the range $-0.5 < [\text{Ca}/\text{Fe}] < 0.0$. We thus apply our simple model with SN Ia yields of $[\text{Ca}/\text{Fe}] = 0.0$ and -0.5 , which are shown as black solid and dotted lines, respectively, in Figure 6 and reasonably match the observed Mg and Ca ratios. This would be quite an extreme situation: if most of the metal enrichment in Car II is due to SNe Ia and not CCSNe, but stars still formed to sample the SN Ia yields, that implies an extremely top-light IMF where no massive stars formed. However, this is definitely not a unique model, and specifically, the flat $[\text{Mg}/\text{H}]$ trend does not rule out contributions from additional CCSNe because gas accretion can increase the hydrogen reservoir (e.g., Ji et al. 2016a). Detailed chemical evolution modeling of more elements might help clarify the picture but is beyond the scope of this paper. Furthermore, stochastic sampling of individual SN explosions may dominate the observed trends (e.g., Koch et al. 2008, 2013; Revaz et al. 2016; Applebaum et al. 2020), especially given that Car II produced only ~ 100 CCSNe in total (assuming a Salpeter IMF and present-day mass-to-light ratio of 2.2; Ji et al. 2016a). Car III is even more susceptible to stochastic enrichment, having been enriched by only ~ 15 SNe. We thus caution against overinterpreting the available data.

4.3.2. $[\text{Mg}/\text{Ca}]$ Abundances across the UFD Population

Some more insight can be derived by comparing the $[\text{Mg}/\text{Ca}]$ versus $[\text{Fe}/\text{H}]$ trends of Car II to the trends in other UFDs. It turns out that few other UFDs have similarly negative $[\text{Mg}/\text{Ca}]$ versus $[\text{Fe}/\text{H}]$ slopes. To quantify this result, we fit lines to the $[\text{Mg}/\text{Ca}]$ versus $[\text{Fe}/\text{H}]$ evolution of every UFD individually and consider the slope angle (i.e., 0° corresponds to a flat line, and negative slope angles indicate declining $[\text{Mg}/\text{Ca}]$ as $[\text{Fe}/\text{H}]$ increases). We then calculate the slopes and slope uncertainties by assuming that data points are drawn from a thin line with multivariate Gaussian uncertainties (see Section 7 of Hogg et al. 2010). We take a uniform prior in slope angle (as opposed to slope) for $\theta \in [-90^\circ, +90^\circ]$ and a flat prior for the intercept, then use emcee to sample the posterior (Foreman-Mackey et al. 2013). We take the posterior median as the point estimate and the 16th–84th percentile range as the 68% credible interval. We remove the four UFDs that have unconstrained posteriors (since their stars have essentially the same $[\text{Fe}/\text{H}]$). Note that the literature UFD stars have inhomogeneously determined uncertainties, so we instead assume independent error bars of 0.2 dex for both $[\text{Fe}/\text{H}]$ and $[\text{Mg}/\text{Ca}]$ but use our actual abundance uncertainties for Car II and III.

The $[\text{Mg}/\text{Ca}]$ versus $[\text{Fe}/\text{H}]$ slopes for all UFDs where ≥ 2 stars have detailed abundance measurements are shown in Figure 7. The top panel of Figure 7 shows the UFD $[\text{Mg}/\text{Ca}]$ slopes versus luminosity (luminosities from the Simon 2019 compilation, including data from Bechtol et al. 2015; Muñoz et al. 2018; Mutlu-Pakdil et al. 2018; Torrealba et al. 2018). There is no obvious relation between slope angle and luminosity. The bottom panel shows a histogram of the slope angle point estimates from the top panel. Many UFDs have too few stars to place a useful slope constraint, so we shade each

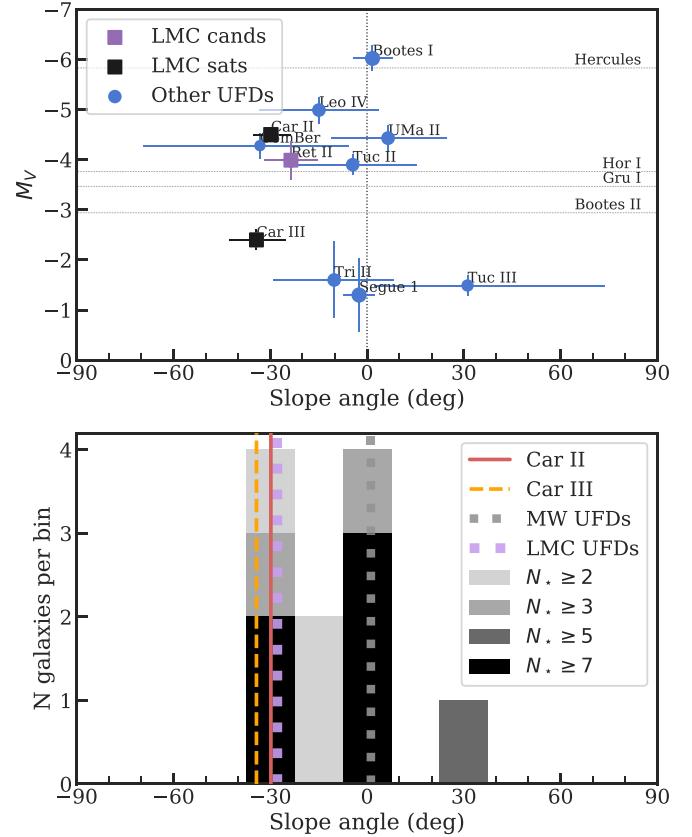


Figure 7. Top: $[\text{Mg}/\text{Ca}]$ vs. $[\text{Fe}/\text{H}]$ slope angles for all UFDs vs. luminosity M_V . Slope error bars indicate the 68% posterior region, and measurements with smaller uncertainties have correspondingly larger symbols. The luminosities of galaxies with unconstrained slope posteriors are shown as horizontal lines. Squares indicate UFDs that are LMC satellites (Car II, Car III, Hor I) or satellite candidates (Ret II). Circles indicate UFDs likely associated with the Milky Way. Bottom: histogram of $[\text{Mg}/\text{Ca}]$ slope angles for all UFDs. The UFDs with more stars (i.e., more confident slope measurements) are shown as darker shades of gray. The relatively extreme slope angles for Car II and III are marked as vertical solid and dashed lines, respectively. The total $[\text{Mg}/\text{Ca}]$ slopes for all Milky Way vs. LMC UFD stars are marked in dotted gray and purple lines, respectively.

UFD in the histogram by the number of stars used to calculate the slope, with darker colors indicating more stars. The UFDs with the most confident measurements (i.e., ≥ 7 stars with detailed abundances) are Car II (this work), Ret II (Ji et al. 2016c), Boote I (Frebel et al. 2016), Segue 1 (Frebel et al. 2014), and Tuc II (Chiti et al. 2018a). We also highlight the slopes of Car II and III as a vertical solid red line and vertical dashed orange line, respectively. Of the other UFDs, only Ret II exhibits a declining $[\text{Mg}/\text{Ca}]$ slope that deviates from zero by $\gtrsim 1\sigma$.

4.3.3. Effect of Environment on $[\text{Mg}/\text{Ca}]$ Abundances

The results above raise an interesting question about the role of environment in determining abundance trends: Car II and III are LMC satellites, and Ret II is also a candidate LMC satellite (Kallivayalil et al. 2018; Erkal & Belokurov 2019).²⁴ In the

²⁴ Hor I (Nagasawa et al. 2018) also is an LMC satellite, but all three currently observed stars have $[\text{Fe}/\text{H}] \sim -2.6$ within the uncertainties and thus no useful constraint on its $[\text{Mg}/\text{Ca}]$ versus $[\text{Fe}/\text{H}]$ trend. The three Hor I stars all have $[\text{Mg}/\text{Fe}] \approx [\text{Ca}/\text{Fe}] \approx 0$.

bottom panel of Figure 7, we show the [Mg/Ca] versus [Fe/H] slope angles from grouping all LMC and Milky Way UFD stars. It is very obvious that the LMC satellite UFD stars have a significant negative slope, while the Milky Way satellite UFD stars have a flat slope; though we note that the LMC trend is mostly driven by Car II and should await additional abundances in LMC satellite UFDs to clarify this suggestion.

However, we speculate briefly on how the large-scale environment could possibly affect chemical evolution in UFDs. At first glance, UFDs should not display significant environmental dependence. They form most of their stars by $z \sim 6$ (Brown et al. 2014), and in simulations, the closest more massive galaxy at $z > 6$ is typically 400 physical kpc away (Wetzel et al. 2015). Even generously sized galactic superbubbles reach only tens of kpc (Griffen et al. 2018), so external enrichment or directly affecting UFD gas with ram pressure stripping is unlikely (Wetzel et al. 2015). However, radiation (both ionizing and Lyman–Werner) can span these distances, though there are limited ways we can imagine this would affect stellar populations. At the metal-rich end, one possibility is the integrated galactic IMF theory (e.g., McWilliam et al. 2013; Weidner et al. 2013), which suggests that as galaxies become gas-poor, they cannot form the most massive stars. If LMC UFDs formed later and thus reionized earlier in their evolution, they would form more of their stars in this phase. At the metal-poor end, delaying Pop III star formation with Lyman–Werner feedback may increase the susceptibility of UFD progenitors to external enrichment (e.g., Magg et al. 2018). Also, metal-free gas with relatively high ionization fractions can form HD molecules during collapse, which may (or may not) affect the Pop III IMF (Glover 2013). A final note is that the distance scales from Wetzel et al. (2015) assume that UFDs reside in dark matter halos of $M_{\text{peak}} \sim 10^9 M_{\odot}$ (Wetzel et al. 2015). If UFDs instead reside in smaller dark matter halos of $M_{\text{peak}} \sim 10^{7-8} M_{\odot}$ (e.g., Jeon et al. 2014; Ji et al. 2015; Jethwa et al. 2018; Graus et al. 2019), then separation distances would become smaller and environmental effects could be more important.

4.4. Inhomogeneous Metal Mixing of AGB Winds in Car II

There is real scatter in [Ba/Fe] at [Fe/H] ~ -2.5 in Car II, with some stars having relatively high Ba abundances and others having low Ba abundances (Figure 4). The extent of the scatter in Ba is ~ 1 dex, much larger than the scatter in any other abundance ratio. A plausible explanation for the Ba scatter is inhomogeneous mixing of AGB wind ejecta into the galaxy’s interstellar medium (ISM). Unlike SN ejecta, which mix rapidly upon entering the hot phase of the ISM, AGB winds mix into relatively cool ISM phases and can thus stay quite inhomogeneous (Emerick et al. 2018, 2019). Since Ba is produced by the *s*-process and released in AGB winds, this mechanism could explain the large Ba scatter. This scenario is supported by the fact that one of the high-Ba stars (CarII-7872) has [Ba/Eu] $\gtrsim 0$ (Figure 4), suggesting that its Ba is predominantly from the *s*-process. Since most barium comes from AGB stars with initial mass $M \lesssim 4 M_{\odot}$ and lifetimes $\gtrsim 10^8$ yr, the presence of AGB enrichment requires that Car II formed stars for at least ~ 100 Myr (Lugardo et al. 2012; Karakas & Lugardo 2016). Note that the nucleosynthetic origin of the low Sr and Ba floor in UFDs remains unknown (see Ji et al. 2019b, for more discussion). One might also expect a correlation between Ba and other AGB elements, like C. We find a moderate but not statistically significant correlation between stars that have both Ba and C detected in Car II (correlation of 0.48 with a *p*-value of 0.34 from `scipy.stats.pearsonr`).

5. Population III Star Signatures

5.1. Carbon-enhanced Fraction in UFDs

Carbon-enhanced metal-poor (CEMP) stars are stars with high [C/Fe] ratios (Beers & Christlieb 2005). Below [Fe/H] ~ -3 , about half the stars in the Milky Way halo are CEMP stars (i.e., [C/Fe] $\gtrsim +0.7$; Aoki et al. 2007). It is generally thought that a specific subclass (CEMP-no stars; Beers & Christlieb 2005)²⁵ of the CEMP stars traces unique nucleosynthesis in Pop III stars (e.g., Norris et al. 2013; Frebel & Norris 2015; Placco et al. 2016). If so, the observed CEMP fraction provides a window to the distribution of some Pop III star properties, such as initial mass, explosion energy, or stellar rotation (e.g., Cooke & Madau 2014; Ji et al. 2015).

In Figure 8, we show the fraction of carbon-enhanced stars below a given [Fe/H] in our $\gtrsim 80$ star UFD literature sample and the halo star compilation by Placco et al. (2014). Both samples have included the Placco et al. (2014) evolutionary carbon corrections. For the UFD sample, we show 68% Wilson confidence intervals on the CEMP fraction. Figure 8 shows that the carbon-enhanced fraction in UFDs is essentially identical to halo stars at all levels of carbon enhancement. For comparison, the CEMP fraction in larger dwarf galaxies like Sculptor has been studied in some detail (e.g., Salvadori et al. 2015; Skúladóttir et al. 2015; Chiti et al. 2018b), but it is still debated whether the CEMP fraction in those galaxies is consistent with the halo.

If we are after pure Pop III signatures, it also makes sense to look at entire UFDs as either C-rich or C-normal (Ji et al. 2015). Seven UFDs have stars with [Fe/H] < -3 . The most metal-poor stars in five of these UFDs are C-rich (Car III, Segue 1, Boo I, Tuc II, and UMa II), while the other two are C-normal (Ret II and Car II). This suggests that the fraction of Pop III stars producing carbon-enhanced abundances is $0.71^{+0.13}_{-0.19}$, following the simple models in Ji et al. (2015). A more stringent cut of [Fe/H] < -3.5 results in three C-enhanced galaxies out of five, or a carbon-enhanced rate of $0.60^{+0.34}_{-0.39}$. More to the point, the existence of carbon-normal stars with [Fe/H] $\lesssim -3.5$ in Ret II and Car II is evidence against the hypothesis that 100% of Pop III stars produce carbon-enhanced signatures, as is often assumed in theoretical models and simulations (e.g., Salvadori et al. 2015; Jeon et al. 2017).

5.2. Full Fits to Individual UFD Stars

The two stars CarII-5664 and CarIII-1120 have low enough [Fe/H] that they are plausibly enriched *only* by Pop III stars (e.g., Frebel & Norris 2015). Under this assumption, we fit models from Heger & Woosley (2010) to the data to estimate the initial progenitor mass, explosion energy, internal mixing, and gas dilution mass for these stars. To summarize the fitting procedure, we find the optimum dilution mass for all 16,800 models in the Heger & Woosley (2010) grid, reject all models inconsistent with our upper limits, then weight each remaining model by using its deviation from the best-fit χ^2 as input to a χ^2 survival function with four degrees of freedom. The detailed fitting procedure and parameter description are in Frebel et al. (2019).²⁶ Here we exclude the elements Al, K, and Mn due to the uncertain size of the NLTE corrections and the elements Sc, Cr, Cu, and Zn due to model calculation uncertainties

²⁵ The “no” is short for “no strong enhancement of neutron-capture elements.”

²⁶ Code at https://github.com/alexji/alexmods/blob/master/alexmods/alex_starfit.py.

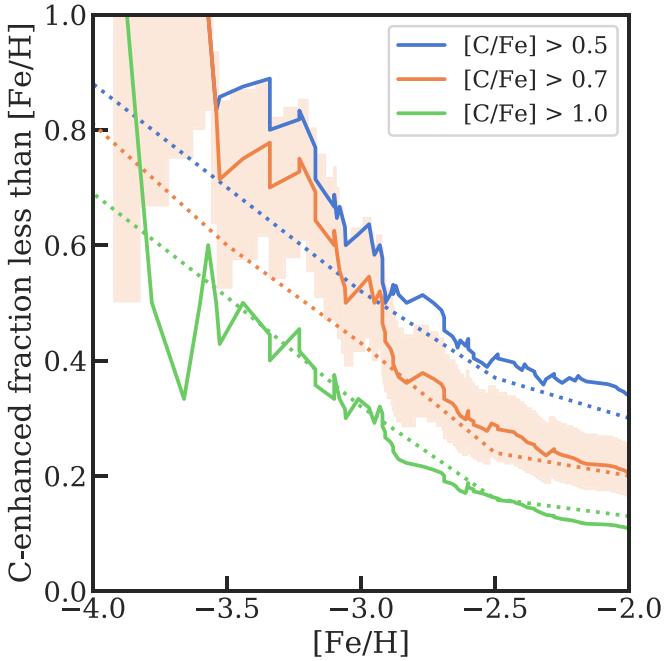


Figure 8. Cumulative CEMP fraction. The solid colored lines show the CEMP fraction for UFD stars at different C-enhanced cutoffs, with the shaded region indicating the 68% Wilson confidence interval for a binomial distribution around the $[C/Fe] > 0.7$ fraction. The dotted colored lines show the halo CEMP fraction from Placco et al. (2014). Both the UFD data and the reference sample have included carbon evolutionary corrections. The UFD CEMP fraction is consistent with the halo.

(Heger & Woosley 2010). Abundance corrections to C, Na, and Mg have been included (Table 5). We note that the Heger & Woosley (2010) models do not include stellar rotation. However, rotation can substantially influence stellar evolution and the resulting nucleosynthesis (e.g., Maeder et al. 2015) and should be considered in future analyses.

The results are shown in Figure 9. We plot all models within 2σ contours of χ^2 (i.e., models with weight $\gtrsim 0.05$). In the top panel for each star, we show the data as filled red squares with error bars and upper limits as downward-pointing arrows. Unused measurements and upper limits are indicated as open squares and downward-pointing triangles, respectively. The best-fit model is shown as a solid blue line, while other models within 2σ are shown as black lines. For visualization purposes, models with worse χ^2 are plotted as thinner transparent lines. The bottom left panel for each star shows the weighted histogram for the resulting progenitor masses of the full fit. The bottom right panel shows the best-fit energy and dilution masses, where again, models with worse χ^2 are displayed as smaller and more transparent points. The best-fit model is again shown as a solid blue point. In general, satisfactory fits were found for these two stars with $[Fe/H] < -3.5$. CarIII-1120 is most consistent with a relatively low-mass progenitor between 10 and $20 M_\odot$ with a typical $\sim 1 \times 10^{51}$ erg explosion energy. Note that CarIII-1120 is a Group 2 CEMP-no star according to Yoon et al. (2016). CarII-5664 is also best fit by a similar low-mass progenitor, but most of the best-fit models actually prefer a higher-mass progenitor of $25-35 M_\odot$ with a slightly higher explosion energy.

The combination of explosion energy and dilution mass introduces another consistency check. An SN with explosion energy E will produce an SN remnant that sweeps up a certain amount of mass before merging with the ISM (e.g., Cioffi et al. 1988). This is the minimum dilution mass allowable for that

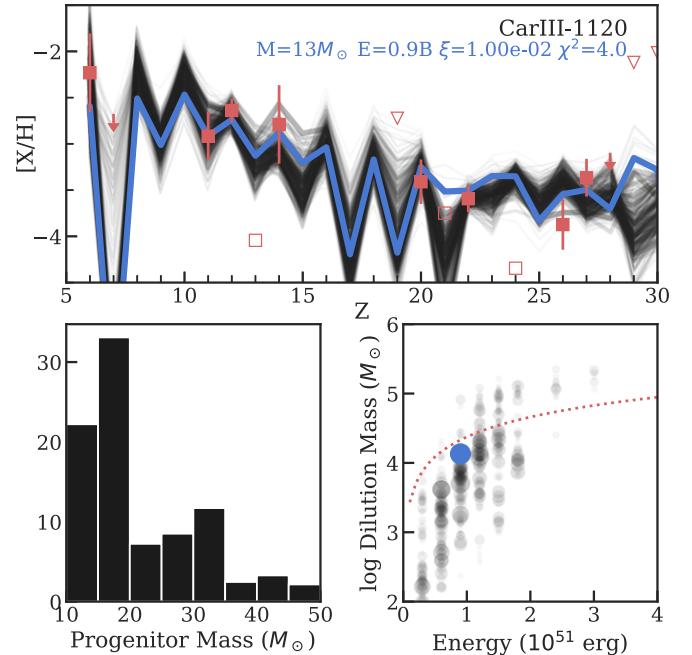
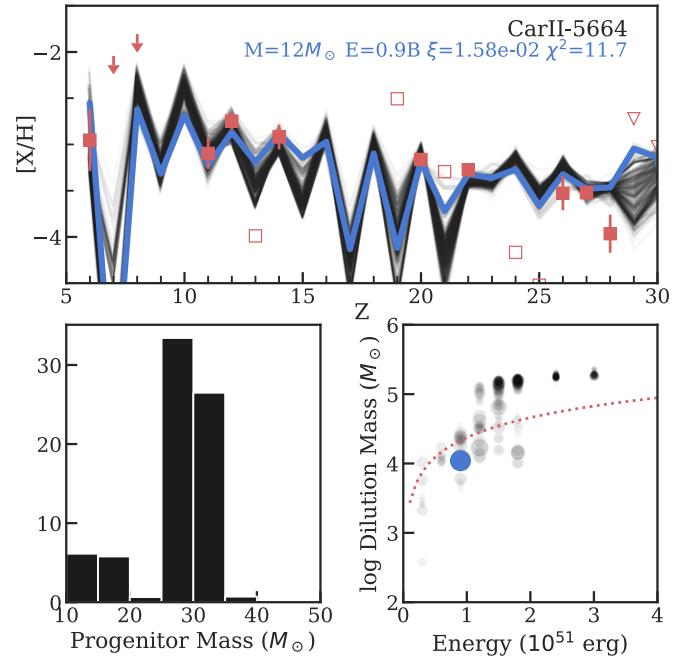


Figure 9. Pop III SN yield fits to abundances of the three stars with $[Fe/H] \lesssim -3.5$. Top panels: measured abundances (red), single best-fit model (blue), and all models within 2σ (black). Bottom left panels: weighted histogram of the best-fit progenitor masses. Bottom right panels: model energy and dilution masses. The best-fit model is shown as a blue point. The dashed red line indicates the minimum dilution mass for a given energy. See text for details.

explosion energy (assuming no rare interactions, such as colliding SN blast waves). In the bottom right panels of Figure 9, we show the approximate swept-up mass of an SN remnant expanding into an efficiently cooling ISM, $M_{dil, H} = 0.75 \times 10^{4.5} M_\odot (E/10^{51} \text{ erg})^{0.95}$, as a dotted red line (Cioffi et al. 1988; Ryan et al. 1996). Models below this line are inconsistent with the explosion energy (though they could be explained with enrichment by multiple SNe), while models above the line are diluted beyond the SN remnant due to

turbulent mixing. Applying this constraint tends to prefer higher explosion energies and higher masses. In general, the best-fit dilution masses satisfying this constraint are $\sim 10^5 M_{\odot}$, suggesting that recollapsed gas within a minihalo is the most likely explanation for the origin of these stars, rather than external pollution, as externally polluted halos have higher effective dilution masses (e.g., Cooke & Madau 2014; Ji et al. 2015; Smith et al. 2015; Griffen et al. 2018).

6. Conclusion

We present a comprehensive abundance analysis of the Magellanic satellite galaxies Car II and III using high-resolution *Magellan*/MIKE data, including the first abundances of an RRL star in any UFD. The abundance results are shown in Figures 2 and 4. The stars in these two dwarf galaxies clearly do not show light-element anticorrelations associated with globular clusters (Figure 5).

The most notable chemical evolution trend is the variations in different α -element ratios. Car II clearly shows different trends in [Mg/Fe] and [Ca/Fe] (Figure 6). The origin of this evolution could be differences in CCSN and/or SN Ia yields, and it is not yet clear which. However, there are obvious differences in the [Mg/Ca] trends between different UFDs (Figure 7), and we tentatively suggest that this could be an environment-dependent abundance signature, as LMC satellite UFDs have a different trend than Milky Way satellite UFDs. This suggestion will require studying the abundances of additional LMC satellites to confirm.

The most metal-poor stars in UFDs may contain signatures of the first metal-free Population III stars. Studying the whole population of Fe-poor UFD stars, we find that the carbon-enhanced fraction of UFD stars is essentially the same as the Milky Way halo (Figure 8). But not all of the most Fe-poor stars in UFDs are carbon-enhanced: the most Fe-poor star in Car II is clearly carbon-normal. We also found two new stars with $[\text{Fe}/\text{H}] \leq -3.5$, bringing the total number of such stars in UFDs up to eight. The abundances of these stars are well fit by Pop III CCSN yields (Figure 9).

Our analysis of Car II and III, along with the past decade of observations, brings the total number of UFD stars with high-resolution abundances up to ~ 85 stars across 16 different UFDs, of which five UFDs now have a “large” (≥ 7) number of stars studied (see references in Section 3.4). While these data have already provided key insights into early nucleosynthesis and galaxy formation and pointed to many interesting abundance trends and signatures, the numbers of stars are still relatively small. These sample sizes are currently dictated by the limits of current large telescopes, but 30 m class telescopes will allow high-resolution spectroscopic abundances for tens to hundreds of stars per UFD out to the virial radius of the Milky Way (Ji et al. 2019a), transforming our ability to unravel the detailed history of these first galaxy relics.

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Facility: Magellan-Clay (MIKE).

Software: MOOG (Sneden 1973; Sobeck et al. 2011), SMHR (Casey 2014), numpy (van der Walt et al. 2011), scipy (Virtanen et al. 2019), matplotlib (Hunter 2007), pandas (McKinney 2010), seaborn (Waskom et al. 2016), astropy (Astropy Collaboration et al. 2013).

Table 6
Stellar Abundances

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
CarII-0064							
Na I	2	3.65	0.15	-2.59	0.31	-0.36	0.19
Mg I	5	5.21	0.07	-2.39	0.17	-0.17	0.08
Al I	2	3.14	0.23	-3.31	0.30	-1.09	0.30
Si I	2	5.19	0.27	-2.32	0.36	-0.10	0.28
Ca I	22	4.31	0.03	-2.03	0.14	0.20	0.06
Sc II	6	0.00	0.04	-3.15	0.13	-0.94	0.07
Ti I	16	2.54	0.05	-2.41	0.24	-0.19	0.08
Ti II	36	2.67	0.02	-2.29	0.13	-0.08	0.07
Cr I	15	3.31	0.04	-2.33	0.22	-0.11	0.06
Cr II	2	3.66	0.07	-1.98	0.14	0.22	0.09
Mn I	6	2.53	0.06	-2.90	0.16	-0.68	0.07
Fe I	169	5.28	0.01	-2.22	0.18	0.00	0.02
Fe II	21	5.30	0.04	-2.20	0.13	0.00	0.05
Co I	5	2.60	0.11	-2.39	0.22	-0.16	0.11
Ni I	8	3.75	0.04	-2.47	0.16	-0.25	0.06
Zn I	2	2.21	0.14	-2.35	0.16	-0.13	0.20
Sr II	2	-0.71	0.23	-3.58	0.39	-1.38	0.32
Ba II	2	-2.58	0.13	-4.76	0.21	-2.55	0.20
C-H	2	6.16	0.16	-2.27	0.35	-0.05	0.22
C-N	1	6.04	0.65	-1.79	0.78	0.43	0.70
O I	1	<7.59	...	-1.09	...	1.13	...
K I	1	<2.75	...	-2.28	...	-0.06	...
Cu I	1	<1.83	...	-2.36	...	-0.14	...
Eu II	1	<-1.89	...	-2.41	...	-0.21	...

CarII-2064

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	4.21	0.15	-2.03	0.33	0.33	0.20
Mg I	5	5.62	0.12	-1.98	0.22	0.38	0.13
Al I	2	3.37	0.40	-3.08	0.44	-0.73	0.41
Si I	2	6.27	0.48	-1.24	0.58	1.11	0.51
K I	1	3.46	0.18	-1.57	0.23	0.79	0.18
Ca I	14	4.48	0.06	-1.86	0.15	0.49	0.09
Sc II	6	0.96	0.12	-2.19	0.19	0.12	0.13
Ti I	11	3.43	0.05	-1.52	0.23	0.83	0.06
Ti II	26	3.23	0.05	-1.72	0.17	0.59	0.10
Cr I	3	2.73	0.35	-2.91	0.44	-0.55	0.36
Cr II	1	3.78	0.18	-1.86	0.23	0.44	0.20
Mn I	3	3.02	0.16	-2.42	0.26	-0.06	0.17
Fe I	81	5.15	0.03	-2.35	0.20	0.00	0.04
Fe II	10	5.19	0.06	-2.31	0.15	0.00	0.09
Co I	1	2.43	0.28	-2.56	0.35	-0.20	0.29
Ni I	2	4.31	0.17	-1.91	0.24	0.44	0.17
Sr II	2	-1.08	0.20	-3.95	0.26	-1.64	0.22
Ba II	2	-0.97	0.11	-3.15	0.20	-0.84	0.16
C-H	2	6.45	0.21	-1.98	0.45	0.38	0.31
O I	2	<8.73	...	0.04	...	2.40	...
Cu I	2	<4.10	...	-0.09	...	2.27	...
Zn I	2	<2.76	...	-1.80	...	0.55	...
Eu II	2	<-0.64	...	-1.16	...	1.15	...
C-N	2	<4.48	...	-3.35	...	-1.00	...

CarII-4704

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	3.50	0.16	-2.74	0.33	-0.54	0.19
Mg I	5	5.06	0.12	-2.54	0.19	-0.34	0.14
Al I	2	3.00	0.56	-3.45	0.58	-1.25	0.58
Si I	2	4.96	0.41	-2.55	0.47	-0.35	0.41
K I	1	2.96	0.17	-2.07	0.23	0.13	0.18
Ca I	14	4.20	0.04	-2.14	0.15	0.06	0.07
Sc II	5	0.74	0.08	-2.41	0.15	-0.22	0.10
Ti I	5	2.65	0.09	-2.30	0.24	-0.10	0.10
Ti II	26	2.65	0.05	-2.30	0.14	-0.12	0.09
Cr I	9	3.23	0.09	-2.41	0.23	-0.20	0.09

Table 6
(Continued)

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Mn I	7	2.58	0.09	-2.85	0.18	-0.64	0.10
Fe I	105	5.30	0.02	-2.20	0.20	0.00	0.03
Fe II	14	5.32	0.04	-2.19	0.12	0.00	0.05
Co I	4	2.69	0.15	-2.30	0.24	-0.10	0.15
Ni I	3	4.11	0.09	-2.11	0.18	0.09	0.10
Sr II	2	-1.68	0.26	-4.55	0.31	-2.36	0.27
Ba II	2	-2.11	0.11	-4.29	0.18	-2.10	0.17
C-H	2	5.63	0.16	-2.80	0.35	-0.60	0.22
O I	2	<7.66	...	-1.03	...	1.17	...
Cu I	2	<2.69	...	-1.50	...	0.70	...
Zn I	2	<2.88	...	-1.69	...	0.52	...
Eu II	2	<-1.44	...	-1.96	...	0.23	...
C-N	2	<5.75	...	-2.08	...	0.13	...
CarII-4928							
Na I	2	3.45	0.16	-2.79	0.35	0.26	0.20
Mg I	5	5.25	0.12	-2.35	0.21	0.71	0.16
Al I	2	3.46	0.35	-2.99	0.45	0.07	0.40
Si I	2	4.81	0.31	-2.70	0.40	0.35	0.31
Ca I	2	4.08	0.18	-2.26	0.28	0.80	0.20
Sc II	6	0.36	0.11	-2.79	0.23	0.24	0.19
Ti II	22	2.58	0.09	-2.37	0.22	0.66	0.18
Cr I	3	1.90	0.20	-3.74	0.37	-0.68	0.21
Mn I	3	1.47	0.16	-3.96	0.30	-0.91	0.17
Fe I	46	4.44	0.04	-3.06	0.26	0.00	0.05
Fe II	5	4.47	0.14	-3.03	0.23	0.00	0.20
Co I	2	2.09	0.24	-2.90	0.39	0.16	0.26
Sr II	2	-1.99	0.25	-4.86	0.33	-1.83	0.30
C-H	2	5.67	0.26	-2.77	0.56	0.29	0.39
O I	2	<8.50	...	-0.19	...	2.86	...
K I	2	<3.04	...	-1.99	...	1.06	...
Ni I	2	<5.16	...	-1.06	...	1.99	...
Cu I	2	<3.06	...	-1.13	...	1.93	...
Zn I	2	<3.08	...	-1.48	...	1.58	...
Ba II	2	<-1.60	...	-3.78	...	-0.74	...
Eu II	2	<-1.04	...	-1.56	...	1.47	...
C-N	2	<6.97	...	-0.86	...	2.19	...
CarII-5664							
Na I	2	3.37	0.12	-2.87	0.27	0.66	0.17
Mg I	7	4.80	0.04	-2.79	0.12	0.73	0.09
Al I	2	2.46	0.17	-3.99	0.29	-0.46	0.27
Si I	2	4.59	0.13	-2.92	0.26	0.61	0.14
K I	2	2.52	0.07	-2.51	0.16	1.02	0.09
Ca I	12	3.18	0.03	-3.16	0.13	0.36	0.07
Sc II	8	-0.14	0.06	-3.29	0.12	0.24	0.08
Ti I	6	1.64	0.04	-3.31	0.22	0.21	0.07
Ti II	32	1.68	0.03	-3.27	0.11	0.27	0.07
Cr I	3	1.47	0.09	-4.17	0.25	-0.64	0.12
Mn I	3	0.91	0.11	-4.52	0.25	-0.99	0.12
Fe I	113	3.97	0.01	-3.53	0.18	0.00	0.02
Fe II	10	3.96	0.04	-3.54	0.10	0.00	0.05
Co I	4	1.47	0.07	-3.52	0.22	0.00	0.08
Ni I	1	2.25	0.20	-3.96	0.28	-0.44	0.21
Sr II	2	-2.86	0.09	-5.73	0.16	-2.19	0.12
C-H	2	4.74	0.21	-3.69	0.48	-0.17	0.33
O I	2	<6.88	...	-1.81	...	1.72	...
Cu I	2	<1.47	...	-2.72	...	0.80	...
Zn I	2	<1.54	...	-3.02	...	0.50	...
Ba II	2	<-3.29	...	-5.47	...	-1.93	...
Eu II	2	<-2.81	...	-3.33	...	0.21	...
C-N	2	<5.79	...	-2.04	...	1.48	...

Table 6
(Continued)

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
CarII-6544							
O I	2	6.99	0.09	-1.70	0.18	0.96	0.21
Na I	2	3.53	0.20	-2.71	0.44	-0.05	0.28
Mg I	7	5.11	0.08	-2.49	0.16	0.17	0.10
Al I	2	3.17	0.25	-3.28	0.36	-0.62	0.26
Si I	2	5.16	0.25	-2.35	0.36	0.31	0.26
K I	1	2.69	0.20	-2.34	0.28	0.32	0.20
Ca I	21	3.78	0.03	-2.56	0.15	0.10	0.07
Sc II	11	0.33	0.04	-2.82	0.11	-0.16	0.06
Ti I	22	2.19	0.04	-2.77	0.29	-0.11	0.10
Ti II	44	2.34	0.02	-2.61	0.12	0.05	0.08
Cr I	16	2.84	0.03	-2.80	0.27	-0.15	0.07
Cr II	1	3.12	0.14	-2.52	0.19	0.14	0.15
Mn I	7	2.29	0.06	-3.14	0.22	-0.48	0.07
Fe I	144	4.84	0.01	-2.66	0.21	0.00	0.02
Fe II	22	4.84	0.03	-2.66	0.13	0.00	0.05
Co I	5	2.19	0.12	-2.80	0.27	-0.14	0.13
Ni I	12	3.43	0.04	-2.79	0.20	-0.13	0.05
Zn I	1	1.47	0.13	-3.09	0.15	-0.43	0.23
Sr II	2	-1.58	0.15	-4.45	0.25	-1.79	0.21
Ba II	4	-2.00	0.10	-4.18	0.16	-1.52	0.15
C-H	2	5.24	0.16	-3.19	0.35	-0.54	0.21
C-N	1	5.48	0.54	-2.35	0.67	0.31	0.58
Cu I	1	<1.48	...	-2.71	...	-0.06	...
Eu II	1	<-2.34	...	-2.86	...	-0.20	...

CarII-7872

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
O I	2	7.56	0.07	-1.13	0.15	1.38	0.21
Na I	2	3.21	0.18	-3.03	0.39	-0.52	0.23
Mg I	5	5.09	0.06	-2.51	0.20	-0.00	0.08
Al I	2	3.59	0.27	-2.86	0.43	-0.35	0.31
Si I	2	5.17	0.32	-2.34	0.43	0.17	0.34
K I	2	2.83	0.12	-2.20	0.25	0.31	0.12
Ca I	12	3.83	0.04	-2.51	0.17	0.00	0.07
Sc II	5	0.54	0.06	-2.61	0.16	-0.12	0.08
Ti I	18	2.38	0.04	-2.57	0.30	-0.06	0.10
Ti II	27	2.55	0.03	-2.40	0.12	0.09	0.09
Cr I	16	2.86	0.04	-2.78	0.27	-0.27	0.07
Mn I	6	2.29	0.08	-3.13	0.20	-0.62	0.10
Fe I	123	4.99	0.02	-2.51	0.21	0.00	0.02
Fe II	20	5.01	0.04	-2.49	0.13	0.00	0.06
Co I	5	2.36	0.10	-2.63	0.22	-0.12	0.11
Ni I	10	3.62	0.05	-2.60	0.20	-0.08	0.06
Zn I	1	1.73	0.10	-2.83	0.12	-0.32	0.23
Sr II	1	-1.03	0.36	-3.90	0.46	-1.41	0.41
Ba II	5	-0.89	0.06	-3.07	0.16	-0.57	0.13
C-H	2	6.24	0.14	-2.19	0.31	0.32	0.20
C-N	1	6.48	0.53	-1.35	0.55	1.17	0.55
Cu I	1	<1.42	...	-2.77	...	-0.25	...
Eu II	1	<-2.59	...	-3.11	...	-0.61	...

CarII-9296

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	3.38	0.15	-2.86	0.33	0.03	0.19
Mg I	5	5.18	0.09	-2.42	0.21	0.47	0.11
Al I	2	3.42	0.35	-3.04	0.44	-0.15	0.40
Si I	2	4.71	0.28	-2.80	0.39	0.09	0.29
Ca I	12	3.82	0.05	-2.52	0.17	0.37	0.09
Sc II	6	0.37	0.07	-2.78	0.18	0.13	0.11
Ti I	4	2.60	0.15	-2.35	0.30	0.54	0.16
Ti II	18	2.49	0.05	-2.46	0.18	0.45	0.12
Cr I	4	2.28	0.15	-3.36	0.35	-0.47	0.18
Mn I	3	2.22	0.36	-3.21	0.42	-0.32	0.36
Fe I	72	4.61	0.03	-2.89	0.23	0.00	0.05

Table 6
(Continued)

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Fe II	8	4.59	0.08	-2.91	0.17	0.00	0.11
Co I	2	1.76	0.21	-3.23	0.36	-0.34	0.23
Ni I	1	3.24	0.34	-2.98	0.42	-0.09	0.34
Sr II	2	-1.28	0.26	-4.15	0.39	-1.25	0.32
Ba II	3	-1.95	0.14	-4.13	0.25	-1.23	0.22
C-H	2	5.72	0.32	-2.71	0.56	0.17	0.41
O I	2	<7.88	...	-0.81	...	2.08	...
K I	2	<3.13	...	-1.90	...	0.99	...
Cu I	2	<2.38	...	-1.81	...	1.07	...
Zn I	2	<2.70	...	-1.86	...	1.03	...
Eu II	2	<-1.44	...	-1.96	...	0.95	...
C-N	2	<6.31	...	-1.52	...	1.37	...

CarII-V3

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	3.04	0.13	-3.20	0.27	-0.56	0.15
Mg I	2	5.62	0.31	-1.98	0.40	0.67	0.33
Al I	2	3.24	0.16	-3.21	0.32	-0.56	0.17
Si I	1	4.29	0.31	-3.22	0.41	-0.57	0.31
Ca I	1	4.01	0.36	-2.33	0.49	0.31	0.38
Sc II	2	0.66	0.17	-2.49	0.26	0.23	0.21
Ti II	14	2.53	0.05	-2.42	0.18	0.30	0.11
Cr I	3	2.88	0.13	-2.76	0.35	-0.12	0.14
Fe I	22	4.85	0.04	-2.65	0.28	0.00	0.06
Fe II	8	4.78	0.06	-2.72	0.13	0.00	0.09

CarIII-1120

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	3.75	0.20	-2.49	0.45	1.39	0.26
Mg I	5	4.94	0.10	-2.66	0.23	1.22	0.13
Al I	2	2.41	0.37	-4.04	0.46	-0.17	0.45
Si I	1	4.72	0.42	-2.79	0.50	1.08	0.42
Ca I	1	2.93	0.22	-3.41	0.28	0.46	0.24
Sc II	5	-0.60	0.12	-3.75	0.18	0.14	0.16
Ti II	6	1.36	0.07	-3.59	0.18	0.30	0.14
Cr I	2	1.29	0.17	-4.34	0.33	-0.47	0.18
Mn I	3	0.86	0.18	-4.57	0.25	-0.70	0.22
Fe I	48	3.63	0.03	-3.87	0.27	0.00	0.05
Fe II	4	3.61	0.08	-3.89	0.14	0.00	0.12
Co I	3	1.62	0.17	-3.37	0.28	0.51	0.20
Sr II	2	-2.75	0.24	-5.62	0.28	-1.72	0.26
C-H	2	5.81	0.28	-2.62	0.63	1.25	0.43
O I	2	<7.50	...	-1.19	...	2.68	...
K I	2	<2.31	...	-2.72	...	1.15	...
Ni I	2	<3.12	...	-3.10	...	0.78	...
Cu I	2	<2.07	...	-2.12	...	1.75	...
Zn I	2	<2.55	...	-2.01	...	1.86	...
Ba II	2	<-2.48	...	-4.66	...	-0.76	...
Eu II	2	<-1.82	...	-2.34	...	1.55	...
C-N	2	<5.15	...	-2.68	...	1.20	...

CarIII-8144

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{\text{[X/H]}}$	[X/Fe]	$\sigma_{\text{[X/Fe]}}$
Na I	2	3.97	0.14	-2.27	0.31	-0.00	0.19
Mg I	7	5.58	0.08	-2.02	0.18	0.25	0.10
Al I	2	3.58	0.23	-2.87	0.27	-0.60	0.24
Si I	2	5.84	0.18	-1.67	0.32	0.60	0.21
K I	1	3.86	0.21	-1.17	0.29	1.09	0.22
Ca I	24	4.62	0.03	-1.72	0.14	0.55	0.05
Sc II	5	1.42	0.07	-1.73	0.17	0.50	0.10
Ti I	15	3.05	0.02	-1.90	0.21	0.36	0.05
Ti II	38	3.14	0.03	-1.81	0.14	0.43	0.08
Cr I	13	3.23	0.05	-2.41	0.20	-0.14	0.06
Cr II	2	3.67	0.08	-1.97	0.14	0.26	0.10
Mn I	7	2.97	0.11	-2.46	0.19	-0.19	0.12
Fe I	147	5.23	0.01	-2.27	0.18	0.00</	

Table 6
(Continued)

El.	<i>N</i>	$\log \epsilon$	σ_{stat}	[X/H]	$\sigma_{[\text{X}/\text{H}]}$	[X/Fe]	$\sigma_{[\text{X}/\text{Fe}]}$
Co I	3	2.56	0.14	-2.43	0.24	-0.16	0.16
Ni I	4	3.85	0.08	-2.37	0.19	-0.10	0.09
Zn I	2	2.84	0.08	-1.72	0.12	0.55	0.15
Sr II	2	-0.75	0.26	-3.62	0.38	-1.39	0.32
Ba II	1	-2.07	0.14	-4.25	0.18	-2.02	0.17
C-H	2	6.31	0.16	-2.12	0.35	0.15	0.23
O I	2	<8.17	...	-0.52	...	1.74	...
Cu I	2	<2.50	...	-1.69	...	0.58	...
Eu II	2	<-1.30	...	-1.82	...	0.42	...
C-N	2	<6.03	...	-1.80	...	0.46	...

(This table is available in its entirety in machine-readable form.)

Appendix

Abundance Error Analysis Formalism

Here we explicitly list the equations used for our error analysis. For element X , with lines indexed by i that have abundances A_i , statistical errors $\sigma_{i,\text{stat}}$, and systematic abundance offsets $\delta_{i,T_{\text{eff}}}$, $\delta_{i,\log g}$, δ_{i,ν_i} , and $\delta_{i,[\text{M}/\text{H}]}$ (note that the systematic abundance offsets retain their sign, so we refer to them as δ_i):

$$\sigma_{i,\text{sys}}^2 = \delta_{i,T_{\text{eff}}}^2 + \delta_{i,\log g}^2 + \delta_{i,\nu_i}^2 + \delta_{i,[\text{M}/\text{H}]}^2, \quad (1)$$

$$\equiv \sum_{\text{SP}} \delta_{i,\text{SP}}^2, \quad (2)$$

$$\sigma_i^2 = \sigma_{i,\text{stat}}^2 + \sigma_{i,\text{sys}}^2. \quad (3)$$

The statistical error $\sigma_{i,\text{stat}}$ quantifies the spectrum noise through either the 1σ equivalent width uncertainty or the χ^2 uncertainty for synthesis. Our equivalent width and synthesis fits allow the local continuum to vary by a linear function, using χ^2 minimization to find the continuum level. Our quoted statistical uncertainties $\sigma_{i,\text{stat}}$ propagate these continuum uncertainties, and they match those inferred from simpler formulae based on the line FWHM within 5% (e.g., Frebel et al. 2006; Battaglia et al. 2008).

In principle, it is possible that our local spectrum models are not accurate, and the most impactful systematic would be misplacing the overall continuum level. As an extra conservative error bar, we include an additional column σ_{cont} in Table 4, which is the uncertainty from systematically changing the overall continuum by the local 1σ spectrum noise (i.e., the abundance difference after multiplying each equivalent width by $1 \pm 1/(\text{S/N})$). For synthesis measurements, we estimate this uncertainty by calculating the equivalent width of the synthetic feature without any other elements, then treating it as an equivalent width measurement. We thus did not estimate the continuum error for the molecular features. A very conservative error estimate would also add this error in quadrature as part of Equation (3). However, we are confident that our continuum placement procedure uncertainties are accurately reflected in the statistical error bar, so we do not include σ_{cont} in our abundance uncertainties.

We then assign each line a weight w_i :

$$w_i = \sigma_i^{-2}. \quad (4)$$

We adopt the weighted average of the lines as the final abundance, with statistical and systematic uncertainties:

$$A(X) = \frac{\sum_i w_i A_i}{\sum_i w_i}, \quad (5)$$

$$\sigma_{\text{stat}}^2(X) = \frac{\sum_i w_i (A_i - A(X))^2}{\sum_i w_i} + \frac{1}{\sum_i w_i}, \quad (6)$$

$$\delta_{\text{sys,SP}}(X) = \frac{\sum_i w_i (A_i + \delta_{i,\text{SP}})}{\sum_i w_i} - A(X), \quad (7)$$

$$= \frac{\sum_i w_i \delta_{i,\text{SP}}}{\sum_i w_i}. \quad (8)$$

The total statistical uncertainty accounts for both noise in individual lines and the weighted standard error of different lines. Here we adopt just the first-order Taylor expansion for the stellar parameter uncertainty, neglecting the covariance between stellar parameters (see McWilliam et al. 2013). Finally, the total abundance error for [X/H] and element ratios [X/Y] combines the statistical and systematic uncertainties in quadrature:

$$\sigma_{[\text{X}/\text{H}]}^2 = \sigma_{\text{stat}}^2 + \sum_{\text{SP}} \delta_{\text{sys,SP}}^2, \quad (9)$$

$$\sigma_{[\text{X}/\text{Y}]}^2 = \sigma_{X,\text{stat}}^2 + \sigma_{Y,\text{stat}}^2 + \sum_{\text{SP}} (\delta_{X,\text{SP}} - \delta_{Y,\text{SP}})^2. \quad (10)$$

Note that for an element ratio of X and Y , we only allow covariance between X and Y through the stellar parameters.

ORCID iDs

A. P. Ji <https://orcid.org/0000-0002-4863-8842>
T. S. Li <https://orcid.org/0000-0002-9110-6163>
J. Marshall <https://orcid.org/0000-0003-0710-9474>
A. K. Vivas <https://orcid.org/0000-0003-4341-6172>
A. B. Pace <https://orcid.org/0000-0002-6021-8760>
K. Bechtol <https://orcid.org/0000-0001-8156-0429>
A. Drlica-Wagner <https://orcid.org/0000-0001-8251-933X>
S. E. Koposov <https://orcid.org/0000-0003-2644-135X>
T. T. Hansen <https://orcid.org/0000-0001-6154-8983>
R. A. Gruendl <https://orcid.org/0000-0002-4588-6517>
A. R. Walker <https://orcid.org/0000-0002-7123-8943>

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