Empirical Constraints on the Nucleosynthesis of Nitrogen

James W. Johnson, ^{1*} David H. Weinberg, ^{1,2,3} Fiorenzo Vincenzo, ^{1,2} Jonathan C. Bird, ⁴ and and Emily J. Griffith ¹

- ¹ Department of Astronomy, The Ohio State University, 140 W. 18th Ave., Columbus, OH, 43210, USA
- ² Center for Cosmology and Astroparticle Physics (CCAPP), The Ohio State University, 191 W. Woodruff Ave., Columbus, OH, 43210, USA
- ³ Institute for Advanced Study, 1 Einstein Dr., Princeton, NJ, 08540, USA

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ABSTRACT

We use a multi-ring galactic chemical evolution model to probe the astrophysical production of nitrogen (N) in the Milky Way. This approach treats individual annuli in the Galaxy disc as conventional one-zone models, and to include the effects of radial migration, stellar populations move between annuli in a manner based on star particles from a hydrodynamical simulation. We find that some recent AGB star yield tables are able to reproduce the gas-phase [N/O]-[O/H] relation as observed only if a substantial fraction of massive stars collapse to black holes. If instead most massive stars explode as supernovae, we must artificially increase N yields from AGB stars by factors of 2 – 3 to offset the additional oxygen. We demonstrate that, with a viable set of AGB star yields, our model is able to reproduce many of the observed correlations between N, O, and Fe abundances for stars when the N abundances are corrected for internal mixing processes within stars. With any of these yields, N production timescales are sufficiently short such that stellar migration is only a minimal source of intrinsic scatter in the observed [N/O]-[O/H] relation. Modest variations in the star formation rate and star formation efficiency produce considerably larger variations in the gas phase N and O abundances, consistent with previous observational arguments. Our models run using the publicly available Versatile Integrator for Chemical Evolution (VICE; https://pypi.org/project/vice).

Key words: methods: numerical – galaxies: abundances, evolution, star formation, stellar content

1 INTRODUCTION

From a nucleosynthesis perspective, nitrogen (N) is a unique element. Along with carbon (C) and helium (He), it is one of only three elements produced in asymptotic giant branch (AGB) stars that are lighter than iron peak nuclei (Johnson 2019). N is also a by-product of the nuclear fusion reactions converting hydrogen (H) into He in stars more massive than the sun with nonzero metallicity. The CNO cycle¹ catalyses the proton-proton chain of nuclear reactions (e.g. Suliga, Shalgar & Fuller 2020) using C, N, and oxygen (O) target nuclei. The slowest component of this chain reaction by far is the $^{14}N(p,\gamma)^{15}O$ component. As a consequence of this bottleneck, to first order the effect of the CNO cycle is to convert all of the C and O isotopes present in a star's core into ¹⁴N. Furthermore, N is among a select group of elements whose observed abundances in stellar spectra often do not reflect the star's birth abundances. Because N is produced in main sequence stars via the CNO cycle, its abundances in a star's core become enhanced relative to what the star was born with. Upon becoming a red giant, internal mixing processes (i.e. dredge-up) bring this N-enhanced material to the photosphere. This phenomenon is both expected from theoretical models and observed

Both observationally and theoretically, N is among the more well-studied elements. Of particular interest in this paper is the correlation between the abundances of N and O, usually observed in the gas phase. In Fig. 1, we present a compilation of such measurements:

- **1.** HII regions in the first six CHAOS² galaxies (NGC 3184, NGC 628, NGC 5194, NGC 5457, M101, NGC 2403; Berg et al. 2020; Skillman et al. 2020; Rogers et al. 2021).
- **2.** HII regions in nearby NGC spirals (Pilyugin, Vílchez & Thuan 2010).
- **3.** HII regions in blue, diffuse star forming dwarf galaxies (Berg et al. 2012; Izotov, Thuan & Guseva 2012; James et al. 2015).
 - 4. Local stars and HII regions (Dopita et al. 2016).
- **5.** Galactic and extragalactic HII regions (Henry, Edmunds & Köppen 2000).
- **6.** Star-forming regions in 550 nearby galaxies in the MaNGA IFU³ survey (Belfiore et al. 2017).

Despite intrinsic scatter and some systematic variation in how the

⁴ Department of Physics & Astronomy, Vanderbilt University, 2301 Vanderbilt Place, Nashville, TN, 37235, USA

in open and globular clusters (Gilroy 1989; Korn et al. 2007; Lind et al. 2008; Souto et al. 2018, 2019; Vincenzo et al. 2021).

^{*} Contanct e-mail: johnson.7419@osu.edu

¹ $^{12}\text{C}(p,\gamma)^{13}\text{N}(\beta^+,\nu_e)^{13}\text{C}(p,\gamma)^{14}\text{N}(p,\gamma)^{15}\text{O}(\beta^+,\nu_e)^{15}\text{N}(p,\alpha)^{12}\text{C}$

² CHAOS: CHemical Abundances of Spirals (Berg et al. 2015)

MaNGA: Mapping Nearby Galaxies at Apache Point Observatory (Bundy et al. 2015). IFU: Integral Field Unit.

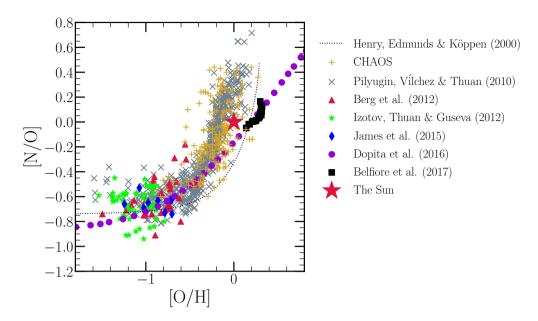


Figure 1. The [N/O]-[O/H] relation as observed in different galactic environments: HII regions from the first six CHAOS galaxies (golden +'s: NGC 3184, NGC 628, NGC 5194, NGC 5457, M101, and NGC 2403; Berg et al. 2020; Skillman et al. 2020; Rogers et al. 2021) and other nearby NGC spiral galaxies (grey X's; Pilyugin et al. 2010), HII regions in blue diffuse star forming dwarf galaxies (red triangles: Berg et al. 2012; green stars: Izotov et al. 2012; blue diamonds: James et al. 2015), in local stars and HII regions (purple circles: Dopita et al. 2016), and in the MaNGA IFU survey (black squares: Belfiore et al. 2017). The fit to [N/O] as a function of [O/H] in Galactic and extragalactic HII regions by Henry et al. (2000) is shown in a black dotted line. We omit all uncertainties for visual clarity. The Sun, at (0, 0) on this plot by definition, is marked by a large red star.

abundances are determined, this [N/O]-[O/H]⁴ relation is more or less the same across a wide range of astrophysical environments. Here we are interested in the origin of both the shape and scatter in this trend.

N is not unique in that perhaps the largest source of uncertainty in understanding its abundances is that accurate and precise nucleosynthetic yields from various enrichment channels remain elusive. Presently, no combination of models for nucleosynthesis in and explosions of massive stars is able to reproduce the observed abundance pattern of the elements, and N is no exception (Griffith et al. 2021). Recently, Grisoni, Matteucci & Romano (2021) argued that rotating massive stars play a key role in establishing the N abundances seen in metal-poor stars in the Milky Way. Rotation has a considerable impact on the N yields of massive stars, because the internal mixing that it causes (Zahn 1992; Maeder & Zahn 1998; Lagarde et al. 2012) brings internally produced C and O nuclei into the H-burning shell where they can be processed into ¹⁴N via the CNO cycle (Heger & Woosley 2010; Frischknecht et al. 2016; Andrews et al. 2017). We find similar results here comparing various theoretical models for massive star nucleosynthesis (see discussion in § 2.1).

Theoretical models for AGB star nucleosynthesis predict N yields to vary as a function of progenitor mass and metallicity (Cristallo et al. 2011, 2015; Karakas 2010; Karakas & Lugaro 2016; Karakas et al. 2018; Ventura et al. 2013, 2014, 2018, 2020). In sufficiently massive AGB stars, the base of the convective envelope is hot enough to activate proton capture reactions, allowing the CNO cycle to convert C and O isotopes in ¹⁴N: a process known as hot bottom burning. AGB stars are also known to experience thermal pulsations, and with each pulse the convective envelope penetrates into the CO-rich core, bringing some of this material into the envelope: a process known as

third dredge-up.⁵. When both processes are active, third-dredge up adds new seed nuclei for hot bottom burning to turn into ¹⁴N, substantially increasing the N yields. We demonstrate in §§ 2.2 and 2.3 that various theoretical models predict significantly different N yields for high mass AGB stars as a consequence of how third dredge-up and hot bottom burning occur in the models. The differences in these processes are in turn a consequence of the microphysical assumptions built into the stellar evolution models (e.g. mass loss, opacity, convection and convective boundaries, nuclear reaction networks).

In this paper, we aim to constrain N yields from AGB stars empirically by assessing to what extent various "off-the-shelf" yield models are able to reproduce the observed correlations between N, O, and iron (Fe), such as the [N/O]-[O/H] relation illustrated in Fig. 1. Vincenzo et al. (2021) demonstrate that when N abundances are corrected for internal mixing processes, the correlations with stellar age and other elemental abundances are affected. Whether or not our galactic chemical evolution (GCE) model is able to reproduce their corrected data constitutes a valuable test not only of our understanding of N nucleosynthesis, but also the accuracy of the Vincenzo et al. (2021) measurements which take a model-dependent approach to estimate the birth abundances of N for a sample of stars.

We make use the multi-zone chemical evolution model for the Milky Way published in Johnson et al. (2021), which treats the Galaxy as a series of concentric rings and tracks the enrichment rates within each ring. This approach has been employed in the past to compute abundances for many Galactic regions (Matteucci & Francois 1989; Wyse & Silk 1989; Prantzos & Aubert 1995; Schönrich & Binney 2009; Minchey, Chiappini & Martig 2013, 2014; Minchey et al.

⁴ We follow standard notation where $[X/Y] \equiv \log_{10}(X/Y) - \log_{10}(X/Y)_{\odot}$.

⁵ Here the time adverbial "third" refers only to the fact that these dredgeup episodes are occurring while the star is on the asymptotic giant branch. Because they are associated with the thermal pulsations of AGB stars, there are many episodes of third dredge-up.

2017; Sharma, Hayden & Bland-Hawthorn 2020). A novel aspect of the Johnson et al. (2021) model is that it relaxes the assumption that stars only enrich the Galactic region they were born in by tracking enrichment rates as stellar populations migrate. Originally developed to study the abundances of O and iron (Fe), this aspect of Galactic evolution turned out to have an important impact on the delayed type Ia supernova (SN Ia) enrichment of Fe. Here we use similar methodology to test for similar effects in the delayed AGB star production

In a sample of 6,507 galaxies from the MaNGA IFU survey (Bundy et al. 2015), Schaefer et al. (2020) demonstrated that the intrinsic scatter in the [N/O]-[O/H] relation at fixed galaxy mass is correlated with variations in the local star formation efficiency (SFE). In regions of slower star formation, [N/O] tends to be slightly higher at fixed [O/H] (see their Fig. 4). This is expected from GCE models, because more AGB stars will enrich the ISM with N by the time a given [O/H] is reached (e.g. Mollá et al. 2006; Vincenzo et al. 2016a). However, Schaefer et al. (2020) did not rule out stellar migration as an additional source of scatter in the gas phase [N/O]-[O/H] relation. In principle, there could be a deficit or surplus of N-producing AGB stars in a given Galactic region at any time simply because the orbits are evolving, and this could drive additional scatter in the correlation. This is the same effect as Johnson et al. (2021) found for SNe Ia, but applied to a different enrichment channel. By using the Johnson et al. (2021) model to take into account the effects of migration, we aim to assess whether one of stellar migration or variability in the local SFE are the dominant sources of this scatter.

2 NUCLEOSYNTHESIS

In this paper we make use of the chemical evolution model for the Milky Way presented in Johnson et al. (2021). This model runs using the publicly available Versatile Integrator for Chemical Evolution (VICE; Johnson & Weinberg 2020; Griffith et al. 2021; Johnson et al. 2021), an open-source PYTHON package designed for GCE modeling. Johnson et al. (2021) focus their discussion of the model predictions on O and Fe, and we retain their yields of these elements here. As required by VICE, the SN yields are defined as the net mass of some element X produced over all explosion events in units of the progenitor cluster's mass. For example, with a yield of $y_X = 0.001$, a hypothetical $1000M_{\odot}$ star cluster would produce $1M_{\odot}$ of the element X instantaneously in the case of core collapse supernovae (CCSNe) or over the delay-time distribution (DTD) in the case of SNe Ia. We adopt the following values from Johnson et al. (2021), who in turn base them off of Weinberg, Andrews & Freudenburg (2017) and Johnson & Weinberg (2020):

- $y_{O}^{CC} = 0.015$ $y_{Fe}^{CC} = 0.0012$ $y_{O}^{Ia} = 0$ $y_{Fe}^{Ia} = 0.00214$

We also assume that N is not produced in significant amounts by SNe Ia (Johnson 2019), and set y_N^{Ia} throughout this paper accordingly. We spend the remainder of this section detailing our CCSN and AGB star yields of N.

2.1 Core Collapse Supernovae and Massive Star Winds

In VICE, CCSN nucleosynthetic products are approximated to be produced instantaneously following an episode of star formation; this is a valid approximation due to how short the lives of massive stars are compared to the relevant timescales for GCE. Based on this and its definition as being in units of a stellar population's total mass, the yield is simply the constant of proportionality between the CCSN production rate and the star formation rate (SFR):

$$\dot{M}_{\mathbf{X}}^{\mathbf{CC}} = y_{\mathbf{X}}^{\mathbf{CC}} \dot{M}_{\star} \tag{1}$$

More generally, y_{X}^{CC} quantifies all of the nucleosynthetic material approximated to be produced instantaneously following a single stellar population's formation, though the majority of such events for which this approximation is valid will be associated with massive stars and/or their supernovae. In the case of N specifically, a substantial amount emerges in winds before the actual supernova itself, allowing massive stars to produce a lot N even if they collapse directly to a black hole (Griffith et al. 2021).

We compute theoretically predicted values of y_N^{CC} using VICE's vice.yields.ccsne.fractional function assuming a Kroupa (2001) IMF; details on how VICE handles these calculations can be found in § 4 of Griffith et al. (2021) and in the VICE science documentation¹. In the left panel of Fig. 2, we plot the results as a function of progenitor metallicity as predicted by the Woosley & Weaver (1995), Nomoto, Kobayashi & Tominaga (2013), Sukhbold et al. (2016), and Limongi & Chieffi (2018) tables. There is good agreement between the various non-rotating models, but only Limongi & Chieffi (2018) report yields for progenitors with non-zero rotational velocities; these yields are substantially larger than that of their non-rotating counterparts. Most of the N production in CCSN progenitors occurs via the CNO cycle processing C and O isotopes into ¹⁴N, and with few C and O seed nuclei at low Z, production of ¹⁴N is difficult. Rotation-induced mixing, a highly uncertain process (Zahn 1992; Maeder & Zahn 1998; Lagarde et al. 2012), could transport newly produced C and O into the hydrogen burning shell of the CCSN progenitor, facilitating ¹⁴N production (Frischknecht et al. 2016; see also discussion in § 4.2 of Andrews et al. 2017). For this reason, N yields at low metallicity are quite sensitive to these assumptions about stellar rotation and internal mixing processes (Heger & Woosley 2010), and consequently IMF-averaged yields are highly uncertain.

Based on the definition of the abundance ratio [X/Y], we can compute the [N/O] ratio of CCSN ejecta from the values of y_N^{CC} and y_{O}^{CC} predicted from a given yield table:

$$[N/O]_{cc} = \log_{10} \left(\frac{y_N^{CC}}{y_N^{CC}} \right) - \log_{10} \left(\frac{Z_{N,\odot}}{Z_{O,\odot}} \right),$$
 (2)

where $Z_{X,\odot}$ is the abundance by mass of some element X in the sun, for which we take $Z_{N,\odot} = 6.91 \times 10^{-4}$ and $Z_{O,\odot} = 5.72 \times 10^{-3}$ based on the solar photospheric abundances of Asplund et al. (2009). For each of the published yield tables and rotational velocities in the left panel of Fig. 2, we compute the corresponding values of y_{Ω}^{CC} using VICEand plot the resultant values of [N/O]cc in the right panel. These yield ratios follow similar trends with progenitor metallicity and rotation as y_N^{CC} itself, a consequence of the fact that these studies predict relatively metallicity-independent O yields.

CCSN yields can to some extent be empirically calibrated by ensuring that they reproduce the [N/O] ratios of low metallicity systems. Since AGB star yields of N are believed to depend on the progenitor's metallicity (see discussion in § 2.2 and references

https://vice-astro.readthedocs.io/en/latest/science_ documentation/yields

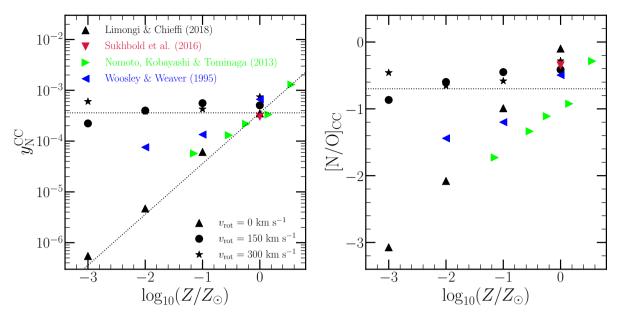


Figure 2. Left: IMF-averaged CCSN yields of N calculated using VICE's vice.yields.ccsne.fractional function with the tables published by Woosley & Weaver (1995, blue), Nomoto et al. (2013, green), Sukhbold et al. (2016, red), and Limongi & Chieffi (2018, black). All studies report yields for non-rotating progenitors, shown by the triangles; for visual clarity, the triangles point in a different direction for each study according to the legend. Limongi & Chieffi (2018) report additional yields for progenitors with rotational velocities of 150 (circles) and 300 km/s (stars). The horizontal dashed line markes $y_N^{CC} = 3.6 \times 10^{-4}$, the value of our fiducial CCSN yield of N in our GCE models. We use the form shown by the slanted line (equation X) in § X in combination with some of our AGB star yield models discussed in § 2.2. **Right**: The [N/O] ratio predicted by each of the explosion models in the left-hand panel, under the same colour-coding and marker scheme. We mark the position of [N/O] = -0.7 with a black dotted line, the value roughly suggested by the observations of low-metallicity systems highlighted in Fig. 1.

therein), it's likely that the "plateau" in [N/O] at low [O/H] reflects the IMF-averaged CCSN yields of N and O. Fig. 1 suggests that $[N/O]_{cc} = -0.7$; we highlight this value in the right panel of Fig. 2 with a horizontal black dashed line. Given this observational result and our adopted value of $y_{O}^{CC} = 0.015$, we compute that an empirical N yield of $y_{N}^{CC} = 3.6 \times 10^{-4}$ using equation (2). We adopt this value as our fiducial CCSN yield of N and highlight it with a horizontal black dashed line in the left panel of Fig. 2. We discuss the sloped dotted line in that panel in the context of some of our AGB star yield models in § X.

These empirical values of [N/O]_{cc} and $y_{\rm N}^{\rm CC}$ are in good agreement with the rotating CCSN models of Limongi & Chieffi (2018). This supports the recent argument by Grisoni et al. (2021) that rotating massive stars play an important role in establishing the N abundances observed at low metallicities in the Milky Way. Although the Sukhbold et al. (2016) tables agree nearly perfectly with our empirical value of $y_{\rm N}^{\rm CC} = 3.6\times 10^{-4}$, they overestimate [N/O]_{cc} by ~0.2 dex; this is because they predict a value of $y_{\rm O}^{\rm CC}$ lower than our adopted value of 0.015. Although most of the supernova models plotted in Fig. 2 slightly overestimate our empirical value of [N/O]_{cc} = -0.7, they still fall short of solar. This implies the need for an additional enrichment, which is expected because it is well understood that N is also produced in considerable amounts by AGB stars (Johnson 2019).

2.2 Asymptotic Giant Branch Stars

Similar to our SN yields (see discussion in § 2.1), our AGB star yields are fractional net yields in that they quantify only the newly produced mass of an element X in units of the progenitor star's zero-age main sequence (ZAMS) mass. For a yield $y_X^{AGB}(M_\star, Z_\star)$, the mass yield

is then given by $M_{\star} y_{\rm X}^{\rm AGB}(M_{\star}, Z_{\star})$. AGB star enrichment proceeds as it does in Johnson & Weinberg (2020) under the caveat that the yield is placed in the $\delta R_{\rm gal}=100$ pc ring that a stellar population is in at a given time. In short, VICE implements an algorithm which computes the mass in dying stars from each stellar population, and the ZAMS mass required to compute the fractional yield comes from a mass-lifetime relationship; for the latter, we adopt the parabola in $\log \tau - \log m$ space from Larson (1974) (see discussion of the mass-lifetime relationship in VICE in Appendix A).

Here we make use of four previously published tables of AGB star yields calculated from stellar evolution models, each of which are sampled on a grid of progenitor masses and metallicities. To approximate y_X^{AGB} as a smooth function of M_{\star} and Z_{\star} , VICE interpolates bi-linearly between grid elements - once in mass and once in metallicity - and linearly extrapolates above or below in either quantity as necessary. By comparing the predicted N abundances of the Johnson et al. (2021) chemical evolution model for the Milky Way to the latest observational data, we can constrain how accurately these "off-the-shelf" yield models characterize how and where N is produced:

- **1.** Karakas (2010, hereafter K10)² published yields for Z = 0.0001, 0.004, 0.008, and 0.02 progenitors. We plot these yields in the upper left panel of Fig. 3.
- **2.** Karakas & Lugaro (2016) and Karakas et al. (2018) published yields for Z = 0.0028, 0.007, 0.014, and 0.03 progenitors; we here-
- We clarify that our abbreviations of these papers (i.e. K10, KL16+K18, V13, and C11+C15) refer specifically to their yields of N as we adopt them in our model. We cite the full names of these papers when referring to their more general results.

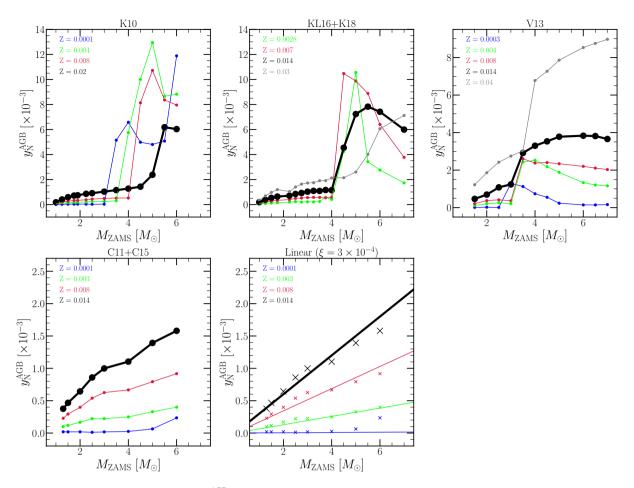


Figure 3. The fractional yields of N from AGB stars y_N^{AGB} as a function of progenitor ZAMS mass and birth metallicity Z as reported by Karakas (2010, upper left), Karakas & Lugaro (2016) and Karakas et al. (2018, upper middle), Ventura et al. (2013, 2014, 2018, 2020, upper right), and Cristallo et al. (2011, 2015, lower right). For Ventura et al. (2013, 2014, 2018, 2020) and Cristallo et al. (2011, 2015), we show the yields only for a selection of metallicities available from their provided tables. We highlight yields at solar metallicity (Z = 0.02 for Karakas 2010, Z = 0.014 otherwise) with bold black lines. In the lower right panel, we show the yields predicted by our linear model (coloured lines, see discussion in § 2.2) in comparison to the Cristallo et al. (2011, 2015, coloured X's) predictions.

after refer to these yields as the KL16+K18 model. We plot a subset of these yields in the upper middle panel of Fig. 3.

3. We combine the yields for Z=0.0003 and 0.008 progenitors from Ventura et al. (2013) with those at Z=0.004 from Ventura et al. (2014), at Z=0.014 from Ventura et al. (2018), and at Z=0.04 from Ventura et al. (2020) into a single table of yields. In this set, we also include a set of un-published yields at Z=0.001 and 0.002 computed with similar stellar models (provided by P. Ventura, private communication). We hereafter refer to this yield set as the V13 model, and we illustrate a subsample of these yields in the upper right panel of Fig. 3.

4. Cristallo et al. (2011, 2015) published yields for Z=0.0001, 0.0003, 0.001, 0.002, 0.003, 0.006, 0.008, 0.01, 0.014, and 0.02 progenitors; we hereafter refer to these yields as the C11+C15 model. This is the default set of yields in VICE. It is also the software's most comprehensive set of previously published AGB star yields in that it includes tables for all elements built into the code and is sampled at the most metallicities. We illustrate a subsample of these yields in the lower left panel of Fig. 3.

VICE also allows users to construct their own functions of progenitor mass and metallicity to describe the AGB star yield. Motivated by the roughly linear of the C11+C15 yields and their general success

in our fiducial model once renormalized by a constant factor (see discussion in § ??), we construct a model in which the yield is linearly proportional to both progenitor ZAMS mass and metallicity:

$$y_{\rm N}^{\rm AGB} = \xi \left(\frac{M}{M_{\odot}}\right) \left(\frac{Z}{Z_{\odot}}\right)$$
 (3)

We illustrate this model in the lower middle panel of Fig. 3 for $\xi = 3 \times 10^{-4}$ in comparison to the C11+C15 yields shown by the coloured \mathbf{V}_{10}

Despite reporting values of the same physical quantities, the N yields reported by each of these studies show substantial differences between one another. Unfortunately, ascertaining the origins of these differences is difficult because each study employs different assumptions for opacity, mass loss, nuclear reaction networks, and convection and convective boundaries within stars, all of which have a significant impact on stellar evolution and thus the predicted yields (see discussion in, e.g., § 5 of Karakas & Lugaro 2016). However, the differences can largely be understood by considering two important phenomena known to occur within AGB stars: third dredge-up (TDU) and hot bottom burning (HBB). Collapsing the information into these two processes is helpful because their differences arise

as a consequence of the different input physics between the stellar evolution models.

TDU refers to the repeated penetrations of the convected envelope into the hydrogen-depleted core during the thermal pulses associated with AGB star evolution. This process doesn't affect N abundances much because at this evolutionary phase, the core is mostly composed of C and O. However, the ${}^{13}C(\alpha, n)^{16}O$ reaction can occur at substantial rates when the core material is mixed with the He-rich shell during each TDU episode. This reaction is the main source of free neutrons in low mass AGB stars, and as a consequence, each replenishment of C does indirectly contribute to raising an AGB star's overall yield. HBB refers to proton capture reactions at the base of the convective envelope. This activates the CNO cycle, producing large amounts of ¹⁴N at the expense of C and O isotopes. HBB requires a higher mass AGB star progenitor ($M_{ZAMS} = 4 - 5 M_{\odot}$ at Z_{\odot} according to Karakas 2010) than TDU ($M_{\rm ZAMS} = 2 - 2.5 \ M_{\odot}$ at Z_{\odot} according to Karakas 2010), but the minimum mass for both decreases at lower metallicities.

The most efficient N production occurs when both TDU and HBB are active within an AGB star, because each replenishment of C and O isotopes by TDU adds new seed nuclei for the CNO cycle when HBB is active. This is the reason for the substantial N production above $\sim 4~M_{\odot}$ in the K10 and KL16+K18 models; in both yield sets, every star that experiences HBB also experiences TDU (see Table 1 in both Karakas 2010 and Karakas 2014, which describes the stellar evolution models from which the KL16+K18 yields are computed). Both TDU and HBB are more efficient at low metallicity (see discussion in Ventura et al. 2013). In the case of TDU, each penetration into the core by the convective envelope is deeper due to the lower opacity. For HBB, the base of the convective is hotter at low Z, and the rate of CNO cycle reactions is an extremely strong function of temperature. This interaction between TDU and HBB is also the reason for the increase in N yields in the V13 tables near $\sim 3~M_{\odot}$. Unlike the K10 and KL16+K18 models, their stars experience both TDU and HBB only in this narrow mass range.

Of all of these yields taken from the literature, the C11+C15 sample shows the smoothest dependence on progenitor mass and metallicity. Unfortunately, ascertaining the exact cause of this difference between other yields explored here is difficult even when collapsing the information into TDU and HBB. Relative to the KL16+K18 yields (see discussion in § 5 of Karakas & Lugaro 2016), the C11+C15 models have more mass loss, a ~10% faster triple- α reaction rate, fewer thermal pulses overall, and weaker HBB due to a lower temperature at the base of the convective envelope. Though their agreement is good below ~ 3 M_{\odot} , the fact that HBB is weaker and there are fewer TDU episodes does however lend a qualitative explanation into why the C11+C15 yields are so much smaller than the K10 and KL16+K18 yields at higher masses.

Although both the K10 and KL16+K18 yield models both show a substantial increase in N yields above $\sim 4~M_{\odot}$, there are some notable differences between the two. In particular, the yields at solar metallicity are somewhat higher in the newer version. Some of this can be attributed to the now-lowered solar metallicity³ (Z=0.014 in KL16+K18 compared to Z=0.02 in K10) and the effect that has on TDU and HBB, but this does not account for all of the differences.

Furthermore, the yields at sub-solar metallicities decreased slightly from K10 to KL16+K18, particularly for the highest mass AGB stars. These differences can be understood by slight variations in the input physics (A. Karakas, private communication). Based on updated opacity tables, the KL16+K18 models are slightly hotter and more compact; consequently, they experience hotter HBB and deeper TDU. Experiencing more thermal pulses overall and consequently a longer AGB lifetime, the KL16+K18 stars have more time for HBB to convert ¹²C into ¹⁴N. At lower metallicity, KL16+K18 use low-temperature opacity tables based on Marigo (2002) that more closely follow the surface composition of the star. These opacities are higher, making the stars larger and increasing the mass-loss rate relative to K10. The Z = 0.0028 model from Karakas et al. (2018) uses the Bloecker (1995) mass-loss prescription as opposed to that of Vassiliadis & Wood (1993) as in both Karakas (2010) and Karakas & Lugaro (2016). This choice results in fewer thermal pulses and a shorter AGB lifetime. Each of these effects at low metallicity act to decrease the overall yield of ¹⁴N.

In the interest of consistency, when we adopt a particular AGB star yield model for N, we also adopt the corresponding table within VICE for O and Fe when possible.⁴

2.3 IMF-Averaged AGB Star Yields: Metallicity and Time Dependence

To more directly compare the K10, KL16+K18, V13, and C11+C15 AGB star yield models of N to one another, we plot IMF-weighted yields from each of them at solar metallicity (Z=0.02 for K10, Z=0.02 otherwise) in the left hand panel of Fig. 4. As mentioned in § 2.2, the AGB star yield of N y_N^{AGB} as we have parameterized it is in units of the progenitor star's ZAMS mass, and consequently the *mass yield* of N is given by $M_{\star}y_N^{AGB}$. With an additional weight of $M_{\star}^{-2.3}$ from the IMF in this mass range (e.g. Kroupa 2001), we therefore multiply the values of y_N^{AGB} by $(M_{\star}/M_{\odot})^{-1.3}$ to quantify the total mass yield of N taking into account the intrinsic mass distribution of stars. Even with the additional weight of $M_{\star}^{-1.3}$, the C11+C15 yields are relatively mass-independent. For other studies, the contributions from higher mass AGB stars is yet more pronounced due to the effects of TDU and HBB discussed in § 2.2.

In the middle panel of Fig. 4, we plot the AGB star production of N from a single stellar population as a function of its age. For this we use VICE's vice.single_stellar_population function which computes the mass yield of a given element as a function of age from a star cluster of known metallicity. For the sake of this calculation, we set all CCSN yields of N to zero in order to highlight the AGB star contribution. We show the results of this procedure for solar metallicity only (again Z = 0.02 for K10, Z = 0.014 otherwise), and we normalize all values to the total mass produced at T = 13.2 Gyr.

Under the C11+C15 yields, it takes ~250 Myr for a single stellar population to produce ~50% of its N from AGB stars, as noted by the coloured points at the top of the panel. This is a rather short characteristic delay time considering that these yields are relatively mass-independent when weighted by the IMF, suggesting instead that most of the N yield under this model is actually coming from higher mass ($\gtrsim 2~M_{\odot}$) AGB stars. This traces back to the steep

³ Changes in the adopted solar metallicity trace back to the canonical value of ~2% (Anders & Grevesse 1989; Grevesse & Noels 1993; Grevesse & Sauval 1998) later being revised to ~1.4% (Lodders 2003; Asplund, Grevesse & Sauval 2005; Lodders, Palme & Gail 2009; Asplund et al. 2009; Asplund, Amarsi & Grevesse 2021).

⁴ In the case of Ventura et al. (2013, 2014, 2018, 2020), AGB star yields of Fe are not available. In the case of our linear model which applies only to N, we assume the VICE default of the Cristallo et al. (2011, 2015) yields for both O and Fe.

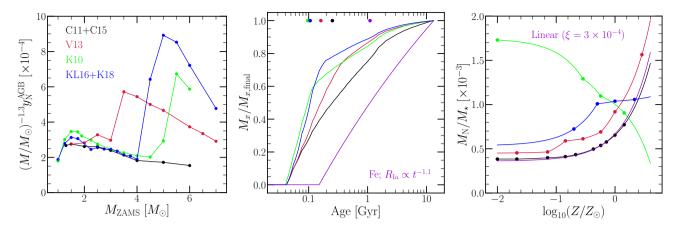


Figure 4. Left: The IMF-weighted mass yield of N from AGB stars as a function of progenitor ZAMS mass at solar metallicity (Z = 0.02 in K10, Z = 0.014 otherwise). Middle: The net mass of N produced by AGB stars from a single stellar population for each of our yield models at solar metallicity. The purple line denotes the same for Fe assuming our $t^{-1.1}$ DTD as in the Johnson et al. (2021) chemical evolution model. All values are normalized to the total mass produced at an age of 13.2 Gyr. Points at the top of the panel denote the ages at which 50% of the total mass yield has been produced. Right: The total amount of N produced by a 13.2 Gyr old stellar population as a function of metallicity for each of our yield models normalized by the stellar population's initial mass. Points mark metallicities at which the published tables report yields.

nature of the stellar mass-lifetime relation (e.g. Larson 1974; Maeder & Meynet 1989; Padovani & Matteucci 1993). Even though the IMF-weighted yields suggest roughly equal contributions per small interval in mass dM_{\star} , this quantifies the contributions per small interval in age $d\tau$. As a stellar population ages, the enrichment rates will inevitably slow down considerably as the mass range dM_{\star} gets progressively narrower as τ increases. As expected due to their larger IMF-weighted contributions to N from high mass AGB stars, our other yield models predict even shorter characteristic delay times.

For comparison, we plot the enrichment of Fe by our $t^{-1.1}$ power-law DTD, also with the CCSN yield set to zero to highlight the SN Ia contribution. The characteristic delay time for Fe production is longer than that of N by nearly an order of magnitude - exactly how much depending on which AGB star yield model is selected. As noted in Johnson et al. (2021), a delay-time of ~ 1 Gyr is exactly as expected for a $\sim t^{-1}$ DTD because half of the SNe Ia occur between 100 Myr and 1 Gyr and and the other half between 1 Gyr and 10 Gyr.

In the right panel of Fig. 4, we plot the total amount of N produced by a 13.2 Gyr old single stellar population as a function of its metallicity according to all of our AGB star yield models, including the linear model (see discussion in § 2.2). For this calculation, we retain our CCSN yield of N (see discussion in § 2.1): $y_N^{CC}=3.6\times 10^{-4}$. In general, there is good qualitative agreement between the C11+C15 and the V13 models, the only major difference being the normalization. The predictions with the linear model with $\xi = 3 \times 10^{-4}$ are nearly identical to the C11+C15 model, though this is unsurprising given their similarity in Fig. 3. The value at which these N yields flatten off at low Z is reflective of our adopted value of y_{N}^{CC} . Up to $\log_{10}(Z/Z_{\odot}) \approx -0.2$, the KL16+K18 yields predict a similar trend as C11+C15 and V13, also with a difference in normalization, but at solar and super-solar metallicities they predict much more metallicity-independent N yields than others. The K10 yields, on the other hand, do not agree with any of the other models, instead predicting N yields to decrease monotonically with increasing Z. These differences between the K10 and KL16+K18 models traces back to differences regarding the opacity and mass loss prescriptions between these two models (see discussion in § 2.2). We demonstrate in § ?? that N yields which scale roughly linearly with metallicity as in the C11+C15 and V13 models are required in order to reproduce

the [N/O]-[O/H] relation as observed. The normalization, however, depends on the SN yields.

3 THE MULTI-ZONE CHEMICAL EVOLUTION MODEL

We use the fiducial chemical evolution model for the Milky Way published in Johnson et al. (2021), which runs using the VICE chemical evolution code (Johnson & Weinberg 2020; Griffith et al. 2021; see Appendix A). Johnson et al. (2021) focus their discussion on the predicted O and Fe abundances, but because VICE recognizes most elements on the periodic table, computing N abundances with this model is easy. Multi-zone models allow simultaneous calculations of enrichment rates and abundances for multiple Galactic regions, making them a more appealing option than classical one-zone models for a system like the Milky Way. Furthermore, they can take into account stellar migration in a framework that is much less computationally expensive than hydrodynamical simulations, making them the ideal experiments for our purposes. We provide a brief summary of the model here, but a full breakdown can be found in § 2 of Johnson et al. (2021).

As in previous models for the Milky Way (e.g. Matteucci & Francois 1989; Schönrich & Binney 2009; Minchev et al. 2013, 2014, 2017; Sharma et al. 2020), this model parameterizes the Galaxy disc as a series of concentric rings of uniform width $\delta R_{gal} = 100$ pc. Each ring is assigned its own star formation history (SFH), and with assumptions about the $\Sigma_{gas} - \dot{\Sigma}_{\star}$ and outflows (see discussion below), VICE computes the implied amounts of gas and infall at each timestep automatically. Under the caveat that stellar populations can move between rings and place their nucleosynthetic products in a ring other than the one they were born in, each ring is otherwise described by a conventional one-zone GCE model. Tracking enrichment as stars migrate was a novel addition by the Johnson et al. (2021) model which they demonstrate to have a significant impact on the enrichment rates from delayed sources such as SNe Ia.

To drive stellar migration, the model makes use of star particles from a hydrodynamical simulation, for which Johnson et al. (2021) select the h277 galaxy from the Christensen et al. (2012) suite evolved

with the N-body+SPH¹ code GASOLINE (Wadsley et al. 2004); we retain this decision here. Previous studies have shown that h277, among other disc galaxies evolved with similar physics, has a realistic rotation curve (Governato et al. 2012; Christensen et al. 2014a,b), stellar mass (Munshi et al. 2013), metallicity (Christensen et al. 2016), dwarf satellite population (Zolotov et al. 2012; Brooks & Zolotov 2014), HI properties (Brooks et al. 2017), and stellar age-velocity relation (Bird et al. 2021). Despite this, there are some interesting differences between h277 and the Milky Way. The last major merger in h277 was at a redshift of $z \approx 3$, making it an interesting case study for its quiescent merger history (e.g. Zolotov et al. 2012). The Milky Way is also known to have a strong, long-lived central bar (e.g. Bovy et al. 2019), while h277 had only a weak and transient bar and lacks one at the present day. With this in mind, a dynamical history such as that of h277 in this GCE model can be thought of as capturing purely secular galaxy evolution. Although these differences between h277 and the Milky Way are well understood, their impact on chemical evolution is not. We are unaware of any studies which investigate the impact of different assumptions regarding the Galaxy's dynamical history and the subsequent effects on predicted abundances; this would be an interesting direction for future work.

Radial migration of stars proceeds from the h277 star particles in a simple manner; for a stellar population in our model born at a radius $R_{\rm gal}$ and a time T, VICE searches for star particles born at $R_{\rm gal} \pm 250$ pc and $T \pm 250$ Myr. From the star particles that make this cut, it then randomly selects on to act as that stellar population's analogue. The stellar population then assumes the present day midplane distance z and the change in orbital radius $\Delta R_{\rm gal}$ of its analogue. In the Johnson et al. (2021) fiducial model, stellar populations move to their implied final radii with a $\sqrt{\text{age}}$ dependence, similar to the assumption made by Frankel et al. (2018, 2019). Although they investigate the impact of variations in this assumption, in the present paper we make use of only this model and one in which stellar migration is only a post-processing prescription. If VICE does not find any star particles from h277 in its initial $R_{\rm gal} \pm 250$ pc and $T \pm 250$ Myr search, it widens it to $R_{\rm gal} \pm 500$ pc and $T \pm 500$ Myr; if still no candidate analogues are found, VICE maintains the $T \pm 500$ Myr requirement, but assigns the star particle with the smallest difference in birth radius as the analogue. As in Johnson et al. (2021), we neglect radial gas flows (e.g. Lacey & Fall 1985; Bilitewski & Schönrich 2012; Vincenzo & Kobayashi 2020), instead focusing on the impact of stellar migration.

Although this model does impose some small but nonzero level of star formation at early times in the outer disc, the sample of star particles from h277 is sufficiently large such that stellar populations that form there are typically assigned analogues which formed within \sim 2 kpc of their birth radius. While ignoring effects such as the radial growth of the Galaxy (e.g. Bird, Kazantzidis & Weinberg 2012; Bird et al. 2013), this at least ensures that these old, outer disc populations are assigned star particles which given them an outer disc rather than an inner disc dynamical history.

Rather than using a hydrodynamical simulation, some previous studies have implemented stellar migration using dynamical arguments (e.g. Schönrich & Binney 2009; Sharma et al. 2020). An advantage of our approach over this is that these dynamical arguments introduce free parameters into the modelw hich then require fitting to data. It is also unclear the extent to which fitting to observed data biases the model into agreement with parts of the sample not involved in the fit. A disadvantage is that we are restricted to one realization

of our dynamical history; slight variations are not possible. We do not distinguish between "blurring" and "churning": terms often used to refer to the epicyclic motions of stars and changes in their guiding centre radii, respectively. These effects are induced by a variety of physical interactions such as molecular cloud scattering (Mihalas & Binney 1981; Jenkins & Binney 1990; Jenkins 1992), orbital resonances with spiral arms or bars (Sellwood & Binney 2002; Minchev et al. 2011), and satellite perturbations (Bird et al. 2012). All of these effects are present in h277 and should therefore be inherited to some extent by the stellar populations in our GCE model.

Our fiducial model here has the same SFH as that of Johnson et al. (2021), where the time-dependence at a given R_{gal} is given by

$$\dot{\Sigma}_{\star} \propto (1 - e^{-t/\tau_{\text{rise}}}) e^{-t/\tau_{\text{sfh}}},\tag{4}$$

where $\tau_{\rm rise}$ approximately controls the amount of time the SFR is rising at early times; we set this parameter equal to 2 Gyr at all radii as in Johnson et al. (2021). Our e-folding timescales of $\tau_{\rm sfh}$ are taken from a fit to this functional form to the Σ_{\star} -age relation in bins of $R/R_{\rm e}$ for $10^{10.5}-10^{11}~M_{\odot}$ Sa/Sb Hubble type spiral galaxies reported by Sánchez (2020). The resulting values of $\tau_{\rm sfh}$ are long: ~15 Gyr at the solar circle ($R_{\rm gal}=8~{\rm kpc}$) and as high as ~40 Gyr in the outer disc (see Fig. 3 of Johnson et al. 2021). This is a consequence of flat nature of the Σ_{\star} -age relation reported by Sánchez (2020).

Within each $\delta R_{\rm gal}=100$ pc ring, the normalization of the SFH is set by the total stellar mass of the Milky Way disc and the present-day surface density gradient assuming it is unaffected by stellar migration (see Appendix B of Johnson et al. 2021). For the former, we neglect the contribution from the bulge and adopt the total disc stellar mass of $5.17\times10^{10}~M_{\odot}$ from Licquia & Newman (2015). For the latter, we adopt a double exponential form describing the thin- and thick-disc components. We take the scale radii of the thin- and thick-discs to be $R_{\rm t}=2.5$ kpc and $R_{\rm T}=2.0$ kpc, respectively, with a surface density ratio at $R_{\rm gal}=0$ of $\Sigma_{\rm T}/\Sigma_{\rm t}=0.27$ based on the findings of Bland-Hawthorn & Gerhard (2016).

The Johnson et al. (2021) models run VICE in star formation mode, meaning that the user specifies the SFH and the amount of gas and infall at each timestep are calculated automatically by the code. Determining the gas supply requires an assumption about the star formation law (often referred to as "star formation efficiency" in the chemical evolution literature, though this term often has other meanings in, e.g., the star formation and feedback community). GCE models have and often still do adopt a single power-law relating the surface density of gas Σ_{gas} to the surface density of star formation $\dot{\Sigma}_{\star}$ based on the findings of Kennicutt (1998). Recent studies, however, have revealed that the star formation law on a galaxy-by-galaxy basis is much more nuanced (de los Reyes & Kennicutt 2019; Ellison et al. 2021; Kennicutt & de los Reyes 2021). Some of the uncertainty regarding its details can be tracked back to the ongoing debate about the CO-to-H₂ conversion factor (Kennicutt & Evans 2012; Liu, Gao & Greve 2015). Based on a compilation of the Bigiel et al. (2010) and Leroy et al. (2013) data shown in comparison to the theoretically motivated star formation laws of Krumholz et al. (2018, see their Fig. 2), Johnson et al. (2021) take a three-component power-law as their star formation law with index given by:

$$N = \begin{cases} 1.0 & (\Sigma_{\rm gas} \geq 2 \times 10^7 \ \rm M_{\odot} \ kpc^{-2}) \\ 3.6 & (5 \times 10^6 \ \rm M_{\odot} \ kpc^{-2} \leq \Sigma_{\rm gas} \leq 2 \times 10^7 \ \rm M_{\odot} \ kpc^{-2}) \\ 1.7 & (\Sigma_{\rm gas} \leq 5 \times 10^6 \ \rm M_{\odot} \ kpc^{-2}). \end{cases}$$
 (5)

The normalization of the star formation law is then set by letting the SFE timescale $\tau_{\star} \equiv \Sigma_{\rm gas}/\dot{\Sigma}_{\star}$ be given by the value derived

¹ SPH: Smoothed Particle Hydrodynamics

observationally for molecular gas at surface densities where N=1. The value of τ_{\star} for molecular gas at the present day is taken to be $\tau_{\text{mol }0} = 2$ Gyr based on Leroy et al. (2008, 2013) with a $t^{1/2}$ timedependence based on the findings of Tacconi et al. (2018) studying the Σ_{gas} – $\dot{\Sigma}_{\star}$ relation as a function of redshift.

Because of the yields adopted in the Johnson et al. (2021) models, considerable outflows are required in order to predict plausible abundances. Weinberg et al. (2017) demonstrate analytically that to first order the equilibrium abunadnce of some element in the gas phase is determined by its yield and the mass loading factor $\eta = \dot{\Sigma}_{out}/\dot{\Sigma}_{\star}$ with a small correction for the SFH. Johnson et al. (2021) make use of this to select a scaling of η with $R_{\rm gal}$ such that the equilibrium abundance as a function fo radius corresponds to a reasonable metallicity gradient within the Galaxy (see their Fig. 3 and discussion in their § 3.1). Nonetheless, one can lower all yields and η at all R_{gal} by similar factors, and with all other model parameters held fixed, GCE models in general make similar predictions. The absolute scale of nucleosynthetic yields is a topic of debate (see discussion in, e.g., Griffith et al. 2021), and some authors argue that outflows do not significantly alter the chemical evolution of the Galaxy disc and neglect them entirely (e.g. Spitoni et al. 2019, 2021). We investigate the impact of simultaneous variations in our yields and the efficiency of outflows in our models in § ?? below.

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Appendices

A VICE

VICE¹ is an open-source PYTHON package designed to model chemical enrichment processes in galaxies with a generic, flexible model. With this paper, we mark the release of version 1.3.0 which presents a handful of new features:

- (i) Users may select a mass-lifetime relation for stars from a list of several parameterized forms taken from the literature. Previously, only a single power-law was implemented, but this formulation underestimates lifetimes for stars with masses $\gtrsim 4 M_{\odot}$; now, the options include the equations presented in:
 - Vincenzo et al. (2016b)
 - Hurley, Pols & Tout (2000)
 - Kodama & Arimoto (1997)
 - Padovani & Matteucci (1993)
 - Maeder & Meynet (1989)
 - Macder & Meyhet (190)
 - Larson (1974) (default)

Generally, chemical evolution models make similar predictions with each of these different forms of the mass-lifetime relation since their quantitative predictions are not considerably different from one another (see the section titled "Single Stellar Populations" under VICE's science documentation for further discussion²). We select the Larson (1974) form as a default within VICE because it is typical compared to the others and requires the lowest amount of computational overhead (aside from the single power-law option).

- (ii) We have added two additional tables of AGB star yields sampled at various progenitor masses and metallicities: the KL16+K18 and V13 models presented in this paper are new to VICE (see discussion in § 2.2 for details).
- (iii) We have built in the SN Ia yields presented in Gronow et al. (2021b,a). These tables present yields for double detonations of sub-Chandrasekhar mass carbon-oxygen white dwarfs at various progenitor metallicities.

Although VICE includes built in SN and AGB star yield tables, users are not required to adopt any one of them for use in their chemical evolution models. Instead, it allows arbitrary functions of metallicity for both CCSN and SN Ia yields and functions of progenitor mass and metallicity for AGB star yields. It provides similar flexibility for additional parameters typically built into GCE models. VICE's backend is implemented entirely in ANSI/ISO C, providing it with the powerful computing speeds of a compiled library while retaining such scientific flexibility within the easy-to-use framework of PYTHON.

Requiring a Unix kernel, VICE supports Mac and Linux operating systems; Windows users should install and use VICE entirely within the Windows Subsystem for Linux. On machines with x86_64 hardware, it can be installed in a terminal via pip install vice. Users running ARM64 hardware (e.g. Macintosh computers with Apple's

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1 Install (PyPI): https://pypi.org/project/vice
Documentation: https://vice-astro.readthedocs.io
Source Code: https://github.com/giganano/VICE.git
2 https://vice-astro.readthedocs.io/en/latest/science_documentation/
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new M1 processor) must install VICE by compiling from source, instructions for which can be found in the documentation. After installing, running vice --docs and vice --tutorial from a Unix terminal will launch a web browser to the documentation and to a jupyter notebook intended to familiarize first time users with VICE's API.