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A Minimum Dilution Scenario for Supernovae and Consequences for Extremely Metal-Poor Stars

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ABSTRACT

To date no metal-free stars have been identified by direct observations. The most common method of constraining their properties is searching the spectra of the most metal-poor stars for the chemical elements created in the first stars and their supernova. In this approach, modelled supernova yields are compared to the observed abundance patterns in extremely metal-poor stars. The method typically only uses the abundance ratios, i.e., the yields are diluted to the observed level. Following the usual assumption of spherical symmetry we compute a simple lower limit of the mass a supernova can mix with and find that it is consistent with all published simulations of early chemical enrichment in the interstellar medium. For three different cases, we demonstrate that this dilution limit can change the conclusions from the abundance fitting. There is a large discrepancy between the dilution found in simulations of SN explosions in minihaloes and the dilution assumed in many abundance fits. Limiting the dilution can significantly alter the likelihood of which supernovae are possible progenitors of observed CEMP-no stars. In particular, some of the faint, very low-yield SNe, which have been suggested as models for the abundance pattern of SMSS0313-6708, cannot explain the measured metal abundances, as their predicted metal yields are too small by two orders of magnitude. Altogether, the new dilution model presented here emphasizes the need to better understand the mixing and dilution behaviour of aspherical SNe.

Key words: stars: Population III – stars: Population II – stars: luminosity function, mass function – cosmology: reionization, first stars, early universe – ISM: supernova remnants

1 INTRODUCTION

The first stars, so-called Population III (Pop III) stars form in the absence of heavy elements in the early Universe. Due to the lack of metal cooling, they are expected to be dras-

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tically different from the stars found in our vicinity at the present day (Bromm 2013; Glover 2013; Greif 2015). Initially, Pop III stars were thought to be very massive (e.g., Bromm et al. 1999; Omukai & Palla 2001), but later it was found that their protostellar disks may fragment, leading to the formation of clusters of low-mass metal-free stars (Greif et al. 2011; Clark et al. 2011). Whereas it is clear that Pop III stars may form over a wide range of masses, until today simulations are unable to constrain well the metal-free initial mass functions (IMFs). The results depend significantly on the physics employed, the choice of numerical method, and the resolution (see e.g., Hosokawa et al. 2016; Stacy et al. 2016; Susa 2019).

There are so far no direct detections of metal-free stars. Pop III stars are expected to form in high-redshift, relatively low-mass mini- and atomic-cooling haloes. Therefore, "Pop III galaxies" are most likely not bright enough to be detected today (Xu et al. 2016; Hartwig et al. 2016b; Visbal et al. 2017). Having no direct observations, there are several indirect methods that allow us to gain observational constraints on the IMF of Pop III stars. Direct detection of supernovae (SNe) (Hummel et al. 2012; Hartwig et al. 2018; Rydberg et al. 2020) or gravitational waves (Kinugawa et al. 2014, 2016; Hartwig et al. 2016a) from the first stars are challenging, but may provide constraints on the high-mass end of the Pop III IMF in the coming decade. The 21 cm absorption feature, as reported by the EDGES experiment (Bowman et al. 2018), can constrain the timing of the first star formation and the star formation efficiency, but it is not very sensitive to the assumed IMF (Schauer et al. 2019).

There are two remaining methods to constrain the pristine IMF that are feasible at present. Both are related to the observations of ancient metal-poor stars in the Milky Way and its satellites. The first one is constraining the low-mass end of the IMF with the current non-detection of metal-free stars (Salvadori et al. 2007; Hartwig et al. 2015; Ishiyama et al. 2016; Magg et al. 2018, 2019). The second method, which is our focus here, is comparing the abundance patterns observed in metal-poor stars to simulated SN yields in Pop III stars. It was found that the most metal-poor stars, often called extremely metal poor (EMP) stars with an iron abundance¹ of less than [Fe/H] = -3 are surprisingly rich in carbon (Beers & Christlieb 2005; Frebel & Norris 2015). A particularly interesting subgroup of stars are the carbonenhanced extremely metal-poor (CEMP) stars and among these the CEMP-no stars: those with iron abundances below [Fe/H] = -3, an excess of carbon relative to iron of more than [C/Fe] = 1 and no enhancement in neutron-capture elements, i.e., [Ba/Fe] < 0 (e.g., Frebel et al. 2005; Keller et al. 2014; Aguado et al. 2018; Nordlander et al. 2019). The origin of the elemental abundance patterns in these stars is one of the key questions of early chemical enrichment.

Umeda & Nomoto (2003) proposed that the abundance pattern of CEMP-no stars is the fingerprint of so-called faint SNe. These are Pop III SNe with relatively large mixingand-fallback at the core-collapse explosions², and eject much of their outer layers, containing carbon and other light elements, whereas most of the inner shells, containing in particular iron, fall back onto the compact remnant. Thus, when the first metal-enriched stars form from gas enriched by one of these SNe, they form with a very small iron abundance but a much higher carbon abundance. Notably, these SNe do not produce particularly large absolute amounts of carbon compared to more conventional core-collapse SNe; rather, they yield high [C/Fe] ratios because they produce unusually small amounts of iron.

Subsequently, it has become common practice to use SN models to infer the properties of the primordial progenitors of the most metal-poor observed stars, both for individual stars (e.g., Heger & Woosley 2010; Hansen et al. 2011; Nomoto et al. 2013; Ishigaki et al. 2014; Bessell et al. 2015; Placco et al. 2016) and for large samples (Cayrel et al. 2004; Fraser et al. 2017; Ishigaki et al. 2018). Additionally, constraints on the primordial IMF can be inferred from bulk properties of metal-poor stars with semi-analytical models (de Bennassuti et al. 2017; Hartwig et al. 2018; Tarumi et al. 2020a). For these purposes, libraries of SN yields have been computed (Heger & Woosley 2010; Nomoto et al. 2013; Ishigaki et al. 2018). These yields typically depend on the stellar mass of the exploding star, the explosion energy and one or a few parameters that quantify the mixing-and-fallback process, which cannot be simulated self-consistently in the onedimensional SN simulations (e.g., Chen et al. 2017; Chan et al. 2020). Comparing the resultant Fe mass to the observed [Fe/H], dilution has been discussed in Tominaga et al. (2007) and Kobayashi et al. (2011).

In order to compare modelled and observed abundance pattern, one further step is required: the SN yields need to be physically diluted with metal-free gas to match the absolute metallicity of the observed star. Usually this dilution is treated as a free parameter and chosen to optimize the quality of fit. Freely adjusting the dilution factor essentially makes the fit independent of the absolute abundances and only considers the ratios of the abundances to each other. Then observed and modelled abundance pattern are compared and the well-fitting ones are interpreted as likely progenitor. For example, the STARFIT³ (Heger & Woosley 2010; Fraser et al. 2017) pipeline can be used for such an analysis.

In this study, we argue that this approach has do be amended because the amount of ambient gas into which the metals from a Pop III supernova are mixed cannot be assumed to be arbitrarily small. We derive a simple analytical model for the lower limit of the mass a SN remnant has to mix with before it can recollapse. We find that this limit is consistent with the results from 3D hydrodynamical simulations. In many cases, there are large differences between the halo-scale mixing found in hydrodynamical simulations and the mixing implicitly assumed by fitting abundance ratios with arbitrary dilution. We show how the dilution limit can be applied in abundance fitting methods. Finally, we investigate examples of the impact this dilution limit has on the conclusions drawn from fitting observed abundances.

 $^{^1\,}$ For elemental abundances, we use the notation

 $[[]X/H] = \log_{10}(N_X/N_H) - \log_{10}(N_{X,\odot}/N_{H,\odot})$ where N_X and N_H are the fractional abundances of any element X and hydrogen, and $N_{X,\odot}$ and $N_{H,\odot}$ are the corresponding solar abundances.

 $^{^2}$ The explosion energy and progenitor star mass are not necessary larger than those for Pop II SNe (Kobayashi et al. 2014).

³ http://starfit.org

2 THE MINIMUM MIXING MASS

2.1 Analytic estimate

As outlined before, abundance fitting usually employs the observed ratios of abundances of certain metals and compares those to the ratios found in theoretical SN models. Of particular importance is, e.g., the [C/Fe] ratio. This method, however, typically neglects the absolute abundances (i.e., [Fe/H] or [C/H]) and treats them as an arbitrary normalization factor. Conceptually, this normalization can be achieved by diluting the SN yields with the correct amount of metal-free gas. As in published work usually only single SNe are fitted to observed abundance patterns, we only consider single, isolated SNe in this work.

We consider SN explosions as well as their subsequent expansion into the ambient medium and the corresponding mixing processes in spherical symmetry. Simulations carried out in two (Tominaga 2009) and three (Chan et al. 2020) dimensions, however, show that Pop III SNe can be strongly aspherical. In this context, we note that even when considering anisotropic SNe, the observed abundances in most published studies are compared to angle-integrated yields. This means that the problem considered is effectively spherically symmetric, as the angular average implies that different elements ejected in different directions become well mixed before the second generation stars form. An exception to this may be, if the abundances are distributed more spherically than the energy input, such as seen in some of the models in Tominaga (2009). A critical analysis of the validity of this approximation is one of the primary motives for the study presented here. We argue that properly accounting for the asymmetries expected in Pop III SNe requires both detailed three dimensional explosion models as well as high-resolution simulations of the expansion of the resulting anisotropic shock wave into an inhomogeneous ambient medium that are able to adequately follow the chemical mixing process.

Since there is no analytic model for such a small-scale inhomogeneous mixing, however, we follow the bulk of the existing literature and approximate the SN as a spherical explosion inside a homogeneous ambient medium. As SNe are very energetic events, a large amount of gas is required to confine the metals and thus not all dilution masses are physically plausible. The lowest limit for this mass is the mass enclosed in the final radius of the SN remnant. Analytical solutions to spherical blast waves of SNe can be derived under a variety of assumptions (e.g., Ostriker & McKee 1988), with the expansion of the remnant stalling at the end of the momentum-driven snowplough phase. In this phase, the expansion velocity reaches the speed of sound in the ambient medium. As shocks cannot be subsonic, the shock wave transforms into a sound wave and dissipates. This occurs at the fade-away radius $R_{\rm fade}$ which is

$$R_{\rm fade} \approx 2.07 \times 10^{20} \,{\rm cm} \, E_{51}^{0.32} n_0^{-0.37} \left(\frac{c_{\rm s}}{10 \,{\rm km \, s^{-1}}}\right)^{-2/5}, \quad (1)$$

where n_0 is the nucleon number density of the ambient medium in units of cm⁻³, E_{51} is the explosion energy in units of 10^{51} erg and c_s is the ambient medium speed of sound (e.g., Draine 2011). We assume the ambient medium is ionized, i.e., that it has a speed of sound of $c_s = 18 \text{ km s}^{-1}$, for a metal-free HII region (see e.g., Abel et al. 2007). In case the medium is actually neutral, the speed of sound would be lower and the stalling radius larger. Thus, this is a conservative assumption. In the homogeneous mixing case, the minimum mass with which the ejecta are mixed is the mass that is enclosed in the stalling radius, i.e.,

$$M_{\rm dil,min} = \frac{4}{3} \pi n_0 \mu m_{\rm H} R_{\rm fade}^3 = 1.9 \times 10^4 \,\rm M_{\odot} \, E_{51}^{0.96} \, n_0^{-0.11}, \ (2)$$

where $m_{\rm H}$ is the mass of a hydrogen nucleus and where we assumed a mean molecular weight of $\mu = 1.22$. The fade-away radius used here is for gas cooling rates of solar metallicity gas. However, the smaller amount of metals in the case considered here would only decrease the cooling rates and therefore increase the total mixing mass. As we aim at computing a lower limit for the mixing mass, the reduced cooling can be neglected. By definition the SN remnant expands faster than the speed of sound in the ionized medium. As we consider haloes below the atomic cooling limit the escape velocity from the haloes is much smaller than this speed of sound. Therefore SN remnants expand much faster than the escape velocity and the effect of gravity can be neglected.

This result is very similar to the one obtained through numerical simulations by Thornton et al. (1998). While it has been widely used in the discussion of stellar feedback, it is often neglected when fitting abundance patterns of individual stars. For example, Tominaga et al. (2014) note that the minimum dilution mass obtained by Thornton et al. (1998) is not a binding limit, as metal mixing is highly inhomogeneous (Ritter et al. 2012). We will later see that our derived limit holds even in cases of inhomogeneous mixing.

We assume an ambient density of $n_0 = 1 \text{ cm}^{-3}$, which should be the typical case for the ionized regions around massive Pop III stars (Whalen et al. 2004). We note that the density dependence of the minimum mixing mass (Eq. 2) is very weak, so it would need to be higher by several orders of magnitude to affect our conclusions. If the density is this much higher than the assumed value, the free-fall time of the ambient gas is smaller than the life-time of the star, and thus it should form stars already before the SN explodes or while the remnant expands. Furthermore, simulations show that it is difficult to mix metals into gas that is already very dense when the SN explodes (Ritter et al. 2016; Chiaki et al. 2018).

Under the assumptions outlined above, the dilution mass is lower limit for two main reasons:

(i) We assume a homogeneous medium. If the medium is not homogeneous the denser gas will be less enriched but form stars first. This effect is discussed further below.

(ii) We assume no further mixing. Realistically further mixing with additional pristine gas should occur during recollapse, rather than the stalled SN remnant monolithically collapsing back on itself. This effect would further increase the dilution mass.

We note that we assume all SNe are able to produce second generation stars. Very energetic explosions may actually disrupt their host haloes, which suppresses or delays second generation star formation (Whalen et al. 2008). This effect is difficult to quantify without hydrodynamical simulations in cosmological context, and is therefore neglected here.

4 M. Magg et al.

2.2 Consistency with simulations

To see whether sub-galactic-scale inhomogeneous mixing can lead to higher metallicities than predicted by the minimum dilution we will compare it to the dilution found in all suitable published simulations of inhomogeneous mixing and the formation of second generation stars which we are aware of. For comparison with our limit, we use an ambient density of $n_0 = 1 \text{ cm}^{-3}$ in all cases but take the explosion energies used in the simulations to compute the minimum mixing. Simulations are included provided that they

• are three dimensional hydrodynamical simulations of the expansion of Pop III SN remnants into their ambient medium,

• are set up to and have sufficient resolution to model individual, isolated, Pop III SNe, not combined populations,

- follow the enriched gas until it re-collapses, and
- provides the output needed for our comparison.

This implies that we do not discuss the results from Greif et al. (2007) and Chen et al. (2015) because the re-collapse of the enriched gas is not modelled. Larger-scale simulations, such as the ones from Wise et al. (2012), Johnson et al. (2013), or Tarumi et al. (2020b), are not considered, because they do not follow individual isolated SNe. The simulations of Whalen et al. (2008) are not included here because they are one-dimensional. Nevertheless, we note that their metalenriched gas masses are consistent with our upper limit in most cases. Only in one of their models is the enriched gas mass they find smaller than our prediction in Eq. (2). In this case, the star completely fails to create an ionized region, and the ability to model an off-centre re-collapse would be crucial to make accurate predictions for the metallicity of the second generations star.

We begin with the dilution found in Ritter et al. (2012, 2015, 2016).⁴ In all three simulations the SNe considered are core collapse (CC) SNe with $E_{51} = 1$. They eject $M_{\text{met}} = 4 \,\mathrm{M}_{\odot}$ of metals in Ritter et al. (2012, 2015) and $M_{\text{met}} = 6 \,\mathrm{M}_{\odot}$ in Ritter et al. (2016). Thus, according to Eq. (2) the maximum final metallicity we should expect is

$$Z_{\rm max} = \frac{M_{\rm met}}{M_{\rm dil,min}} \approx 10^{-3.6} \approx 10^{-1.7} \,\rm Z_{\odot} \tag{3}$$

where $Z_{\odot} = 0.0142$ is the solar metallicity (Asplund et al. 2009). While the mixing is highly inhomogeneous, and orders of magnitude of spread in metallicity can be seen, the newly collapsing cores always show metallicities below this value. All simulations also contain gas at higher metallicities than predicted by the minimum dilution. While from Ritter et al. (2012) it is unclear in which phase this gas is contained, in Ritter et al. (2015, 2016) only some of the very diffuse gas has metallicities above the dilution limit.

Chiaki et al. (2018) and Chiaki & Wise (2019) model the inhomogeneous mixing occurring after the SNe of 7 different stars with masses between $13 \,\mathrm{M}_{\odot}$ and $200 \,\mathrm{M}_{\odot}$. Some of these SNe are simulated in several different halos. The simulations cover a wide range of different environments in which

SNe can explode. For massive stars, halos are often completely photo-evaporated, whereas, for the smallest stars, the gas in the stellar birth-cloud remains dense throughout the lifetime of the star. The results show large variations between the mixing behavior and the metallicities of the second-generation stars. Chiaki et al. (2018) distinguish between three separate enrichment channels:

(i) Internal enrichment: in this case, the SN expands efficiently and the metals mix well with the surrounding gas before the halo collapses back on itself.

(ii) External enrichment: the metals escape from the halo in which the SN explodes and mix with the gas in a different halo that has not formed stars yet. This type of enrichment is also found in Smith et al. (2015).

(iii) Inefficient internal enrichment: dense structures remain in the halo. When the SN explodes these structures are only enriched to very low metallicities and proceed to form stars with metallicities much lower than the average gas metallicity in the halo.

None of these simulations, however, show the formation of second generation stars that violate our dilution limit. According to Eq. (2) the predicted maximum metallicity ranges between $10^{-2.6} Z_{\odot} < Z_{\text{max}} < 10^{-1.6} Z_{\odot}$. All second generation stars in their simulations have metallicities in the range $10^{-6.3} Z_{\odot} < Z < 10^{-2.2} Z_{\odot}$. None of the stars violate our derived limit. The simulated second generation star that is closest to our computed upper limit is enriched by a $25 \,\mathrm{M}_{\odot}$ CCSN that explodes in their halo "MH1", which is their smallest halo with a mass $M_{\rm vir} = 3 \times 10^5 \,{\rm M_{\odot}}$. The re-collapsing region has a metallicity of 40 per cent of our computed upper limit. In these simulations, there are several cases of stars with much lower metallicities than predicted by the minimum dilution model. These are the cases in which the surroundings of the SNe are the most dense and the mixing is the most inhomogeneous. The second generation stars form in clumps that already exist when the SNe explode and only the outer layers of these clumps are enriched with metals. Thus, the enrichment proceeds in what Chiaki et al. (2018) label the "inefficient internal enrichment" channel.

Greif et al. (2010) simulate the explosion of a single PISN with $E_{51} = 10$ and $100 \,\mathrm{M}_{\odot}$ of metal ejecta. According to our model the maximum metallicity in this extreme case should be below $Z = 10^{-1.4} \,\mathrm{Z}_{\odot}$. They find metallicities in the recollapsing galaxy that are around $Z = 10^{-3} \,\mathrm{Z}_{\odot}$. As Greif et al. (2010) note, the average metallicities are initially much higher but they decrease to this low value during the recollapse of the halo, which takes around 300 Myr.

The simulations by Jeon et al. (2014) include several SNe exploding in three different haloes. The authors provide information on the metallicity of recollapsing regions in three cases: a 15, 25 and $40 \,\mathrm{M}_{\odot}$ star exploding in their "halo1". They all explode as $E_{51} = 1$ CCSN and eject 5 per cent of their stellar mass as metals. According to our model, this should lead to metallicities of $Z < 10^{-2.1} \,\mathrm{Z}_{\odot}$. Their reported metallicities are all below $Z = 10^{-3.5} \,\mathrm{Z}_{\odot}$.

Smith et al. (2015) highlight the external enrichment channel. Their SN is a $E_{51} = 1$ CCSN which ejects $11.19 \,\mathrm{M}_{\odot}$ of metals, leading us to predict a maximum metallicity of $Z_{\max} = 10^{-1.4} \,\mathrm{Z}_{\odot}$. Only a very small fraction of gas is found at such high metllicities, and none of it is in the re-collapsing

⁴ We only consider the 1sN model from Ritter et al. (2015) as the 7sN model deals with enrichment by multiple SNe, which is not the topic of our analysis.

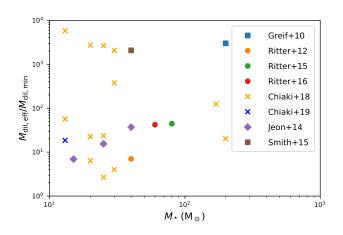


Figure 1. Comparison of the minimum dilution model to simulations of inhomogeneous metal mixing. We show the ratio of the effective dilution mass of the simulations and our estimate of the minimum dilution mass as a function of the stellar mass of the exploding star. The effective dilution mass is derived from the metallicity in the second generation stars or the likely sites of second generation star formation in the simulations. All simulations show a ratio above one, i.e. they are consistent with our predicted minimum.

region. The metal-enriched star forming gas in this case has a metallicity of $Z = 10^{-4.7} Z_{\odot}$.

We convert the metallicities found in the simulations back to an "effective dilution mass" with eq. (3) and summarize the simulations in Fig. 1. None of the simulations of inhomogeneous mixing show inconsistencies with the minimum dilution mass derived from the spherically symmetric case. In some of the simulations, there is gas above the derived upper limit for the metallicity, but it tends to be diffuse and hot. This can be understood intuitively: as thermal energy and metals are ejected together, more metal-rich gas tends to be hotter. It is important to note that there is significant scatter in the simulation results: even for similar exploding stars the effective dilution mass can vary by many orders of magnitude. The cases with the largest effective dilution masses are usually external or inefficient internal enrichment. We conclude that, to the best of our knowledge and the current state of modelling, our estimate provides a useful limit on the mixing and dilution of metals even in the presence of inhomogeneous mixing.

2.3 Bayesian fitting

We will here briefly discuss how the derived limit on mixing can be implemented in abundance fitting codes. For this purpose we create an algorithm that fits observed abundances by comparing to them to the modelled SN yields from Heger & Woosley (2010). The yields of the SNe generally depend on the progenitor mass ($M_{\rm prog}$), the explosion energy (E_{51} in units of 10^{51} erg) as well as a mixing factor ($f_{\rm mix}$). For matching observed and modelled abundances, we use the SN yields and analysis tools provided with STARFIT and supplement them with a generic Bayesian fitting approach. A general description of Bayesian parameter estimation can be found in Bailer-Jones (2017). We first compute the likelihoods $L_i(x_i|M)$ that a model M, which predicts the abundances y_i , results in the observed abundances x_i ,

$$L_i(x_i|M) = \exp\left(-\frac{(x_i - y_i)^2}{2\sigma_i^2}\right),\tag{4}$$

where *i* is any of the observed elements and σ_i is the error of the observations. The normalization is left arbitrary for now. This likelihood calculation implicitly assumes that the errors follow a Gaussian distribution. While it is not clear whether this assumption is valid, it is commonly made when fitting SN models to observed abundances (e.g. Heger & Woosley 2010; Ishigaki et al. 2018; Ezzeddine et al. 2019). Computing the modelled abundances y_i requires, as discussed above, a usually arbitrary dilution mass $M_{\rm dil}$. If M_i is the mass of element *i*, which has a mass-number of μ_i , that is ejected by a SN, the model abundance is

$$y_i = \log_{10} \left(\frac{M_i}{\mu_i X_{\rm H} M_{\rm dil}} \right) - \log_{10} \left(\frac{N_{i,\odot}}{N_{\rm H,\odot}} \right),\tag{5}$$

where $X_{\rm H} = 0.754$ is the hydrogen abundance of primordial gas (Planck Collaboration et al. 2016). The respective solar fractions of the element *i* and of hydrogen are $N_{i,\odot}$ and $N_{\rm H,\odot}$. We iteratively adjust the dilution mass for each model until we find the dilution that gives the maximum final likelihood according to Eq. (7). The dilution mass is picked individually for each model, but within each model the same dilution mass is used for every element. This choice comes from our assumption that each element mixes in the same way. This assumption is commonly made for SN fitting, as without it the SN yields would not be representative of the elements found in the second generation stars.

However, in many cases elements are not detected and only upper limits on their abundance can be derived. These upper limits need to be treated simultaneously with the detections. For this we assume that the upper limits are strict (i.e. the likelihoods are Heaviside step-functions Θ) combined with a Gaussian error on where exactly this limit is. These assumptions lead to a likelihood $L_i(x_i|M)$ for an upper limit of x_i in element *i* of:

$$L_{i}(x_{i}|M) = \int_{-\infty}^{\infty} \Theta(x_{i} - z_{i}) \exp\left(-\frac{(z_{i} - y_{i})^{2}}{2\sigma_{i}^{2}}\right) dz_{i}$$
$$= \int_{-\infty}^{x_{i}} \exp\left(-\frac{(z_{i} - y_{i})^{2}}{2\sigma_{i}^{2}}\right) dz_{i}$$
$$= \sqrt{\frac{\pi}{2}} \sigma_{i} \operatorname{erf}\left(\frac{x_{i} - y_{i}}{\sqrt{2}\sigma_{i}}\right),$$
(6)

where erf is the Gaussian error function. The likelihoods of the individual elements can then be combined by multiplication:

$$L(x|M) = \prod_{i} L_i(x_i|M).$$
(7)

The same approach to compute fit likelihoods was also used in, e.g., Fraser et al. (2017). In cases where there are only detections and no upper limits, maximizing this likelihood is equivalent to minimizing χ^2 . This way of combining likelihoods implicitly assumes that the errors of all abundance determinations are uncorrelated. Especially for errors from uncertainties in the determination of stellar parameters, this may not be true (McWilliam et al. 1995). This is because all low-excitation lines arising from neutral minority species tend to have similar sensitivity to the effective temperature,

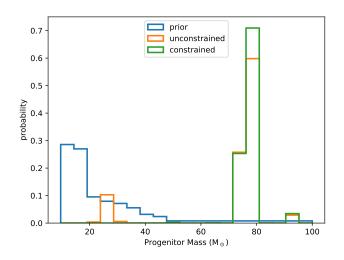


Figure 2. Prior (blue) and posterior distribution of the progenitor mass of HE 0020-1741. We show the posterior for unconstrained (orange) and constrained (green) dilution. In the unconstrained dilution there is a bimodal distribution of progenitor masses, whereas with the constrained dilution only high-mass stars match the observed abundances.

which typically dominates the error budget. However, we only aim at showing the importance of constraining the dilution of SN ejecta, and a complete treatment of the error distributions and dependencies of abundance determinations exceeds the scope of the current investigation.

If we assign each model in the SN library the same prior probability, we can further compute the probability of each model M given the observations x by

$$P(M|x) = N L(x|M), \tag{8}$$

where ${\cal N}$ is a normalization constant chosen such that

$$\sum_{M} P(M|x) = 1.$$
(9)

3 APPLICATION TO OBSERVATIONS

In this section we demonstrate in three cases why it is important to consider the dilution when fitting abundances of metal-poor stars. Firstly, we will show that it can help to break degeneracies in a fit; secondly, that it may systematically change properties of large fitted samples of stars; and thirdly, that for some stars there may not be a viable singleprogenitor scenario to explain the observed abundance patterns.

3.1 Example 1: The progenitor of HE 0020-1741

To investigate the impact of the minimum dilution mass on abundance fitting, we firstly fit the CEMP-no star HE 0020-1741 ([Fe/H] = -3.6). Hansen et al. (2019) have determined abundances for 13 elements (C, N, O, Mg, Ca, Sc, Ti, Cr, Mn, Ni, Fe, Sr, Ba). As the yields from Heger & Woosley (2010) do not include r- and s-process elements, Sr and Ba are excluded from the fits. Because it is generally underpredicted in the models Sc is treated as an upper limit.

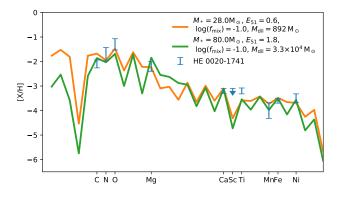


Figure 3. Best-fitting abundance patterns for HE 0020-1741. We show the observed pattern and the patterns of the best-fitting models with low-mass and high-mass progenitors. If only the abundance ratios are considered, both give an equally plausible fit, yet constraining the dilution rules out a single low-mass star as a progenitor.

We show the prior and the posterior distribution of stellar masses in Fig. 2. The prior is bottom heavy, as there are many more models of low-mass SNe in the libraries than there are models of high-mass SNe. This could potentially bias fitting results towards lower masses. We perform the fits with unconstrained and with constrained dilution factors. In the former case, we chose the dilution factors to maximize the combined likelihoods as defined in Eq. (7). In the latter case, we only allow dilution factors above our analytical limit. For the unconstrained dilution we find a bimodal posterior: there are solutions with progenitor masses either around $M_* \approx 25 \,\mathrm{M}_{\odot}$ or at $M_* \approx 80 \,\mathrm{M}_{\odot}$, which we will refer to as "low-mass" and "high-mass" here. In Fig. 3 we show the abundance pattern produced by the best-fitting model from each of these branches. Both models seem to fit approximately equally well. However, if we consider only the constrained dilution case, only the high-mass progenitors still fit. The low-mass progenitors do not produce enough metals to explain the metal abundances of HE 0020-1741 with only a single SN. Thus, if HE 0020-1741 is to be explained with a single progenitor SN, it should be a massive star with $70 \,\mathrm{M}_{\odot} < M_* < 80 \,\mathrm{M}_{\odot}$ for the Heger & Woosley (2010) yields. Note, however, that it may be possible to find additional or better fits with different yield sets (e.g. Limongi & Chieffi 2012; Ishigaki et al. 2018; Grimmett et al. 2018). However, as the aim here is to show the usefulness of the dilution limit to constrain fits, a comparison of these different yield set exceeds the scope of this study.

3.2 Example 2: Large sample fitting

Ishigaki et al. (2018) fitted the abundances of 201 EMP stars by picking the best-fitting SN model for each of these stars. The compiled sample of stars has been selected to consist only of stars with determined abundances for the elements C, N, O, Na, Mg, Al, Si, Ca, Sc, Ti, Cr, Mn, Fe, Co, Ni, and Zn based on spectroscopic data with a resolution of at least R =28000. These observed abundances were compared to SN models which were computed over a grid of stellar masses, explosion energies as well as three parameters that quantify

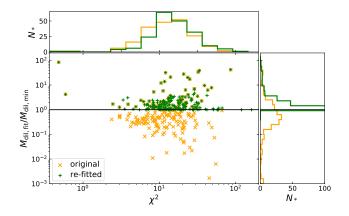


Figure 4. Ratio of the minimum dilution mass and the dilution mass derived in Ishigaki et al. (2018) as function of the reduced χ^2 , as well as histograms of both values. We show the original fits from Ishigaki et al. (2018) (orange) as well as re-fits in which the minimum dilution limit is enforced (green). In some of the original fits, the dilution ratio is very low (down to ~ 10⁻⁷) and therefore outside of the boundaries of this figure. These stars are included in the lowest bin of the histogram.

the properties of the mixing-and-fallback process. Details on the sample selection and the SN modelling can be found in Ishigaki et al. (2018). Each star in the sample was associated with a best-fitting model, based on χ^2 minimization. In this paper, we use the explosion energies of their best-fit models to compute minimum dilution masses for each of the SN models. In Fig. 4 we show the ratio between these minimum dilution masses and the dilution masses from the fits for all 201 stars. Of these 201 best-fitting models, 128 violate our derived limit and 43 do so by more than a factor of four. Notably there is no apparent correlation between χ^2 and the dilution mass. Thus, whether the dilution factor found by fitting is physical is unrelated to the goodness of fit.

For comparison, we have repeated the fitting procedure from Ishigaki et al. (2018) while at the same time enforcing the dilution limit. This means that all fits with too small dilution masses are rejected and instead the best-fitting model that fulfils the dilution limit is picked. This leads to a significant increase in the mean (median) χ^2 from 16 (13) in the unconstrained case to 24 (15) in the constrained case. For many stars, the best fit with the dilution constraint becomes worse than that without it. In Fig. 5 we show that this re-fitting leads to significant changes in the distribution of best-fitting progenitors masses. The most notable difference is that progenitors with a stellar mass of 25 and 40 ${\rm M}_{\odot}$ are now much rarer and progenitors with $15\,\mathrm{M}_{\odot}$ more common. The reason for this is that many of the previously common 25 and $40 \,\mathrm{M}_{\odot}$ models were hypernovae (HNe) with a high explosion energy and a large fallback fraction. These stars have relatively low absolute yields, but due to their large explosion energies, we predict large dilution masses in spherical symmetry. Therefore, such models are not able to reproduce the relatively large carbon abundances of many CEMP-no stars, when taking the dilution constraint into account.

We note that the prescription of faint SNe used in Ishigaki et al. (2018) is chosen to reproduce the angle-averaged yields of aspherical jet SNe (Tominaga 2009). Our dilution

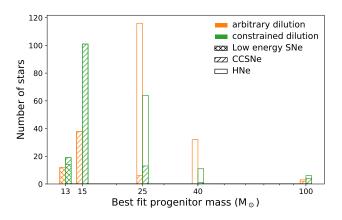


Figure 5. Distribution of progenitor masses from Ishigaki et al. (2018) as well as our repetition of the fits which take the minimum dilution criterion into account. There are only bins at 13, 15, 25, 40 and 100 M_{\odot} as these are the only progenitor masses of the SN models. We separate SNe by explosion energy into low-energy SNe ($E_{51} < 1$), CCSNe ($E_{51} = 1$) and hypernovae (HNe, $E_{51} \ge 10$).

model, however, does not apply to such SNe if their asphericity is preserved. In the used prescription only the total vields are considered. Even if the abundance distribution in the ejecta is strongly aspherical, this approximation assumes that the SN yields are mixed and the angular variations are washed out during later phases of the expansion of the SN. In principle, the mixing behaviour in aspherical SNe can be very different from our approximation if the metal yield per unit energy shows strong angular variations. Additionally, aspherical SNe from Ishigaki et al. (2018) have systematically larger and can have much larger explosion energies, which are used in Equation (2), than their 2D counterparts with similar yields (Tominaga 2009). This further limits the applicability of our model to these SNe. Our results here suggest that developing realistic models for the dilution of heavy elements produced in aspherical SNe is of vital importance for fitting large samples of stars, not just individual cases.

3.3 Example 3: No spherical progenitor for SMSS0313-6708

As we realized previously that stars with high carbon and low iron abundances are particularly strongly affected by applying the dilution criterion, we will look in more detail at a pathological example of such a star, i.e., SMSS0313-6708 (Keller et al. 2014). The star is known for having no detected iron abundance with an upper limit of [Fe/H] < -7.1. We here use abundances that are based on 3D atmospheric models that do not assume local thermodynamical equilibrium (3D, NLTE) for Na, Mg, Al, Ca, and Fe from Nordlander et al. (2017). For these elements statistical and systematic errors are provided which we add with a quadratic sum. The systematic errors are typically on a level of 0.1 dex. The remaining abundances are taken from Bessell et al. (2015) and are based on 3D LTE models for C, N, and O and on 1D LTE models for Si, Sc, Ti, V, Cr, Mn, Co, Ni, and Cu. Notably, most of these elements are not detected and only upper limits on their abundance have been derived. Bessell

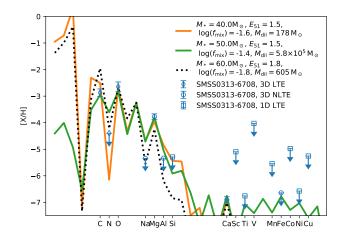


Figure 6. Best-fitting models for SMSS0313-6708. We show the unconstrained (orange) and the constrained (green) best-fitting model. We find the same best-fitting model as Bessell et al. (2015). For reference we also show the best-fitting model from Keller et al. (2014). We mark with different symbols whether abundances have been derived in 1D LTE (diamonds), 3D LTE (squares), or in 3D NLTE (circles)

For better representation we shifted the upper limits by one standard deviation, such that the upper end of the error bar corresponds to discrepancy at a 84 per cent confidence level.

et al. (2015) only give statistical but no systematical errors for their abundance determinations. Because we want avoid biasing our results towards abundances with an unaccounted source of error, we add a systematical error of 0.1 dex to the abundance determinations from Bessell et al. (2015).

We fit the abundances with the same procedure as described in Sect. 3.1. The best constrained and unconstrained models are shown in Fig. 6. For better graphical representation we shift the upper limits by one standard deviation. Thus, according to Eq. (6) the upper end of the error bar represents a discrepancy at a 84 per cent confidence level, and a value 1σ above the upper limit corresponds to a 98 per cent significant discrepancy. Even with unconstrained dilution, we find no model that produces a convincing fit of the abundance patterns. The best-fitting model overproduces Na, C and Si. There are three features in the abundances that are difficult to fit simultaneously:

- (i) the CNO pattern with high C and O but very low N,
- (ii) the low upper limit on Na with the detection of a large amount of Mg, and
- (iii) the detection of Ca in conjunction with the low upper limits on Al and Si

The difficulty involved in reproducing all three of these features may partially be related to the grid of models not containing a sufficiently large variety of SN explosion energies. We note that none of the elemental abundances that have been derived in 1D LTE play a critical role in constraining the models. All 1D LTE abundances are only upper limits that lie well above the best-fitting models. It is still unclear how different the C and O abundances would be in a 3D NLTE analysis, but they would need to differ by approximately 1 dex from 3D LTE in order for us to be able to find matching abundance SNe. Nordlander et al. (2017) were able to fit the abundance patterns by interpolating the abundance patterns as function of the explosion energy. However, the result of such a procedure is potentially sensitive to the way the interpolation is done. We therefore decided against interpolating to a finer grid here.

Both the unconstrained best-fitting model and the bestfitting model from Keller et al. (2014) violate the dilution limit by around two orders of magnitude. They would require the SN ejecta to be diluted with less than $500 \,\mathrm{M_{\odot}}$ of pristine material. The best-fitting model that fulfils the dilution limit is clearly inconsistent with the observed upper limits of N and Na. In Ishigaki et al. (2014) this star was best-fitted with $25M_{\odot}$ and $40M_{\odot}$ SN or HNe (jet-induced, aspherical, and energetic SN), where Ca is produce by static/explosive O burning and incomplete Si burning in contrast to the explanation in Keller et al. (2014). Of these models, the SNe are consistent with our dilution limit and the HNe are neither compatible with the dilution criterion nor with the updated upper limit on Si that we use. The fits presented in Ishigaki et al. (2018) are compatible both with the abundance pattern we use and with our dilution limit. Chen et al. (2017) model potential progenitor SNe for SMSS0313-6708 in one and two dimensions. Of their models only the two dimensional model of the SN of a $60 \,\mathrm{M}_{\odot}$ Pop III star is consistent with our dilution limit. As these elements are not modelled, however, it is unclear whether this model is able to reproduce the observed upper limits of the Na and Al abundances. Chan et al. (2020) performed full 3D SNe simulations of the progenitor of SMSS0313-6708 using a $40 \,\mathrm{M}_{\odot}$ star, as suggested by Bessell et al. (2015), with asymmetric explosion of low and high energy. Their nucleosynthesis has the same constraints as those by Chen et al. (2017), and the explosion was not followed beyond shock breakout. The lowenergy model does not produce any significant metals, the high-energy model too much iron – if spherically averaged.

4 DISCUSSION AND SUMMARY

We have introduced a analytical limit for the dilution of metals produced by a single SN with the following three assumptions:

- the SN being alone and isolated,
- the explosions being spherical and well-mixed, and
- the surrounding medium being homogeneous.

The first two assumptions are commonly made when comparing observed abundance patterns to SN yields in previous works, because if these are not fulfilled the total elemental yields from a single SN cannot be representative stellar abundance pattern. For the last assumption we compared this limit to all hydrodynamical simulations of metal enrichment in high-redshift minihaloes which we are aware of and which included the needed details and resolution for a comparison. We found that, despite assuming homogeneity, the limit is consistent with all of these simulations.

We demonstrate that previous fits were often inconsistent with our understanding of metal dilution and mixing on the scale of minihaloes. Including our dilution criterion into fitting procedures for abundance patterns can have important consequences for the conclusions drawn: (i) Considering the dilution can help to break degeneracies in progenitor models of individual stars.

(ii) The limit does not just affect individual stars but it can also change the properties of large samples of progenitor models. In particular, low-yield SNe are disfavoured if constraints on the dilution are taken into account.

(iii) It may be difficult to explain certain stars, such as SMSS0313-6708 by enrichment from a single, spherical SN if the dilution is taken into account. The best-fitting models that have been put forward by Keller et al. (2014) and Bessell et al. (2015) explain the rough shape of the observed abundance ratios, but with an implicit dilution mass that is too small by approximately two orders of magnitude, the yield from the SNe are too small to explain the absolute metal abundances. Ishigaki et al. (2014, 2018) find fits to the abundance pattern that are consistent with our dilution criterion.

During the preparation of this manuscript, Komiya et al. (2020) derived a similar estimate for the minimal dilution and implemented it into a semi-analytical model of the formation of the Milky Way. While we apply this estimate to the exploration of progenitor scenarios of individual stars, Komiya et al. (2020) focus on the chemical evolution of the Milky Way and in particular on whether the overall population of CEMP-no stars can be reproduced. They find it difficult to reproduce the prevalence of large carbon abundances in the lowest metallicity stars with faint SNe. This tension between the mixing-and-fallback SN model and the large observed carbon abundances is consistent with our findings.

The minimum dilution estimate can serve for evaluating whether a single, spherical SN is a viable progenitor scenario for a certain star. This test may be less reliable, if applicable at all, for asymmetric SNe or cases with several SNe in one halo. In asymmetric SNe, a large fraction of the metals can be ejected along jets (Tominaga 2009). Evidence for such SNe has recently been found by Ezzeddine et al. (2019). The dilution and recollapse occuring after such SNe are yet to be explored by numerical simulations.

Altogether, we conclude that for the adequate astrophysical interpretation of the observed elemental abundances in extremely metal-poor stars, both the relative abundance patterns as well as the absolute abundance values need to be taken into account. Only then reliable and well founded constraints on the properties of the preceding generation of stars can be derived. Given the fact that simple spherically symmetric models often fail to match the dilutions mass constraint introduced here, we furthermore conclude that the effects of aspherical supernovae, the impact of inhomogeneous mixing in a highly structured interstellar medium, and the combined yields of multiple supernovae requires further investigation.

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DATA AVAILABILITY

No new data were generated in support of this research. The SN models from Heger & Woosley (2010) are available on http://2sn.org. For the availability SN models or the sample of stars used in Section 3.2, please inquire with the authors of Ishigaki et al. (2018).

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