

The Metal-poor Metallicity Distribution of the Ancient Milky Way

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Abstract

We present a low-metallicity map of the Milky Way consisting of ~110,000 metal-poor giants with −3.5 < [Fe/ H] < −0.75, based on public photometry from the second data release of the SkyMapper survey. These stars extend out to ∼7 kpc from the solar neighborhood and cover the main Galactic stellar populations, including the thick disk and the inner halo. Notably, this map can reliably differentiate metallicities down to $[Fe/H] \sim -3.0$, and thus provides an unprecedented view into the ancient, metal-poor Milky Way. Among the more metal-rich stars in our sample ($[Fe/H] > -2.0$), we recover a clear spatial dependence of decreasing mean metallicity as a function of scale height that maps onto the thick disk component of the Milky Way. When only considering the very metalpoor stars in our sample ($[Fe/H] < -2$), we recover no such spatial dependence in their mean metallicity out to a scale height of $|Z| \sim 7$ kpc. We find that the metallicity distribution function (MDF) of the most metal-poor stars in our sample $(-3.0 < [Fe/H] < -2.3)$ is well fit with an exponential profile with a slope of and $\Delta \log(N)/\Delta[Fe/H]$ H] = 1.52 \pm 0.05, and slightly shifts to Δ log(N)/ Δ [Fe/H] = 1.53 \pm 0.10 after accounting for target selection effects. For $[Fe/H] < -2.3$, the MDF is largely insensitive to scale height |Z| out to ∼5 kpc, showing that very and extremely metal-poor stars are in every galactic component.

Unified Astronomy Thesaurus concepts: [Population II stars](http://astrothesaurus.org/uat/1284) (1284); [Milky Way Galaxy](http://astrothesaurus.org/uat/1054) (1054)

1. Introduction

Many studies have focused on comprehensively characterizing the nature of the Milky Way's galactic components, which include the bulge (Howes et al. [2015](#page-6-0)), the thin and thick disks (Duong et al. [2018](#page-6-0)), the metal-weak tail of the thick disk, and the inner and outer halos (Carollo et al. [2007](#page-6-0)) through determining spatial properties (e.g., scale heights, densities), stellar chemical abundances (e.g., metallicities), kinematic properties, and when possible, dating using age measurements. The halo is of particular interest for galactic archeology studies that, e.g., aim to understand the galaxy's early evolution. First, a relatively large portion of the halo is composed of ancient very and extremely metal-poor stars ($[Fe/H] < -2.0$ and $<$ −3.0, respectively), which are pivotal for studies of early chemical evolution (Frebel & Norris [2015](#page-6-0)). Second, theoretical simulations of galaxy formation (Bullock & Johnston [2005](#page-6-0)) have shown that the halo bears the signatures of the Milky Way's assembly from smaller "building block" galaxies.

Astrometric data from the Gaia mission (Gaia Collaboration et al. [2016](#page-6-0), [2020](#page-6-0)) have shown, e.g., the existence of stellar kinematic signatures as a result from earlier accretion events (Belokurov et al. [2018;](#page-6-0) Myeong et al. [2019](#page-6-0); Yuan et al. [2020](#page-6-0)), but there remains the need for precise metallicities down to the extremely metal-poor regime ($[Fe/H] \sim -3.0$) for many stars for a full chemodynamic picture of early structures still present in the current Milky Way. Recent spectroscopic and photometric surveys have began to remedy this. Spectroscopic surveys for which metallicities have been derived for large samples of stars include LAMOST (Zhao et al. [2012](#page-6-0)), RAVE (Steinmetz et al. [2020](#page-6-0)), GALAH (De Silva et al. [2015](#page-6-0)), Sloan Digital Sky Survey (SDSS)/SEGUE (Yanny et al. [2009](#page-6-0); Eisenstein et al. [2011](#page-6-0)), and APOGEE (Majewski et al. [2017](#page-6-0)). Photometric surveys are far more complete than spectroscopic

surveys and thus offer better prospects for mapping chemical structure because the relative rarity of extremely metal-poor stars means that these are generally poorly represented in any sample. Ivezić et al. ([2008](#page-6-0)) created a metallicity map of the northern sky based on SDSS broadband photometry of ∼5.6 million stars. The disk and halo as well as the Monoceros stream cleanly differentiate in metallicity versus height above the Galactic plane, confirming the galactic components and stellar populations. However, the SDSS photometry only enabled metallicity determination down to $[Fe/H] \sim -2.0$.

More recent photometric surveys, such as the SkyMapper Southern Sky Survey (Keller et al. [2007](#page-6-0)) and the Pristine survey (Starkenburg et al. [2017](#page-6-0)), have instead developed the use of metallicity-sensitive filters surrounding the Ca II K line that enable metallicity measurements in the very and even extremely metal-poor regimes. These have led to notable results, including identification of the most metal-poor stars in the halo (Starkenburg et al. [2018;](#page-6-0) Da Costa et al. [2019](#page-6-0)) and the bulge (Howes et al. [2015](#page-6-0); Arentsen et al. [2020](#page-6-0)), and the metallicity distribution function of extremely metal-poor stars (Youakim et al. [2020](#page-6-0)).

Additional spatial maps of the metallicity of the northern hemisphere were presented in An & Beers ([2020](#page-6-0)), analogous to those in Ivezić et al. ([2008](#page-6-0)), but extending to lower metallicities ($[Fe/H] \sim -3.0$). Similarly, spatial maps of the Southern sky as a function of metallicity were presented in Casagrande et al. ([2019](#page-6-0)) and Huang et al. ([2019](#page-6-0)) using photometry from Data Release 1.1 from the SkyMapper Survey (Wolf et al. [2018](#page-6-0)). However, these maps only claimed metallicity precision down to [Fe/H] \sim -2.0 or [Fe/H] \sim -2.5, and were limited by the relatively shallow magnitudes in SkyMapper DR1.1.

Here we explore the oldest and most metal-poor stellar population spread across the galaxy with a series of metallicity maps of the Southern sky based on the recently released SkyMapper Southern Sky Survey Data Release 2 (DR2; Onken et al. [2019](#page-6-0)). The metallicity-sensitive imaging filters (Bessell et al. [2011](#page-6-0)) enable a significant metallicity resolution down to $[Fe/H] \sim -3.3$. Using techniques developed in Chiti et al. ([2020](#page-6-0)) to extract precise metallicities from the intermediateband SkyMapper v and broadband u , g , i filters, we derived photometric metallicities for every giant star with $g < 17$ and $[Fe/H] < -0.75$ in the full catalog (Chiti et al. [2021](#page-6-0)). We focus here on a subset of well-selected ∼110,000 metal-poor red giants with quality metallicity measurements to map out the ancient components of the Galaxy.

2. Observational Data and Photometric Metallicities

The SkyMapper Southern Sky Survey DR2 contains ∼500 million astrophysical sources (Onken et al. [2019](#page-6-0)). We downloaded photometry for every star brighter than $g = 17$ in this catalog, dereddened the data following the bandpass coefficients on the SkyMapper website, $\frac{5}{3}$ and derived photometric stellar parameters of these stars ($log g$, $[Fe/H]$) following the methods presented in Chiti et al. ([2020](#page-6-0)). In summary, this involved matching the observed photometry to a grid of synthetic photometry spanning a broad range of stellar parameters $(1.0 < \log g < 3.0, 4000 \text{ K} < T_{\text{eff}} < 5700 \text{ K},$ $-4.0 < [Fe/H] < -0.75$). We excluded all stars with photometric $[Fe/H] \ge -0.75$, $[Fe/H] \le -3.75$, and $\log g < 3.0$, and $0.35 < g - i < 1.2$ (corresponding to our T_{eff} limits) to largely limit our sample to metal-poor red giant stars. We furthermore excluded stars with random uncertainties of >0.5 dex in their photometric metallicity and those in regions of high reddening (galactic latitude $|b| < 10$ and $E(B-V) > 0.35$ in Schlegel et al. ([1998](#page-6-0))). This resulted in a sample of 593,668 stars. Finally, we find that we systematically overestimate the metallicities of very metal-poor ($[Fe/H] < -2.0$) stars when $g - i < 0.65$, upon comparison to high-resolution samples (Jacobson et al. [2015;](#page-6-0) Ezzeddine et al. [2020](#page-6-0)). To avoid this issue, we conservatively only retain stars with $g - i > 0.65$, resulting in 308,702 stars. We refer the reader to Chiti et al. ([2021](#page-6-0)) for further details on the full catalog.

We adopted photogeometric distances to every star in our sample from Bailer-Jones et al. ([2021](#page-6-0)) over using inverted parallax-based distances (see Mardini et al. [2019](#page-6-0) for a recent discussion). To ensure distance measurements of sufficient quality, we only included stars that have distance uncertainties less than the 20% level. We then excluded stars with absolute SkyMapper g magnitude > 5 and a location on a colormagnitude diagram consistent with a metallicity [Fe/ H] > -0.75 (Chiti et al. [2021](#page-6-0)), which further limits contamination from cool dwarf stars and more metal-rich interlopers. This resulted in a final sample of 111,149 stars extending down to $g \sim 17$.

To validate the metallicities in our sample, we compared our metallicities to those from multiple other surveys. Details are given in Chiti et al. ([2021](#page-6-0)), but we list relevant comparisons here. For $-2.5 <$ [Fe/H] < -0.75 , when only considering stars with uncertainties on their photometric metallicities < 0.5 dex, our metallicities are on average lower than those reported in the

Figure 1. Comparison of [Fe/H] metallicities between our photometric SM results and high-resolution measurements of stars in Barklem et al. ([2005](#page-6-0)), Marino et al. ([2019](#page-6-0)), Ezzeddine et al. ([2020](#page-6-0)), and X. Ou et al. (2021, in preparation). The agreement is excellent, with Δ [Fe/H] = 0.04 \pm 0.03. The standard deviation of the residuals between the metallicities is 0.29 dex. Dashed lines indicate ± 0.5 dex to guide the eye, and the metallicities in Barklem et al. ([2005](#page-6-0)) have been increased by 0.25 dex to account for an empirical correction on T_{eff} (Frebel et al. [2013](#page-6-0)).

LAMOST survey by 0.05 dex. The standard deviation of the residuals between our photometric metallicities and those in the LAMOST survey is 0.25 dex, indicating great agreement. Similarly, upon comparing to metallicities from the combined APOGEE and GALAH surveys over the same metallicity regime, we find, on average, lower metallicities by 0.21 dex. The metallicity residuals have a standard deviation of 0.25 dex.

We demonstrate that our metallicities are robust when [Fe/ H < -2.5 following high-resolution observations of 74 initial metal-poor candidates selected from SkyMapper DR1.1 and DR2 data (Wolf et al. [2018](#page-6-0); Onken et al. [2019](#page-6-0)) as an early test sample. We derived their stellar parameters following standard spectral analysis techniques (Frebel et al. [2013](#page-6-0)) for these stars based on Magellan/MIKE (Bernstein et al. [2003](#page-6-0)) highresolution ($R \sim 22,000$) snapshot spectra that were reduced using the CarPy pipeline (Kelson [2003](#page-6-0)). Full results will be reported in X. Ou et al. (2020, in preparation) but we show results (red points) in Figure 1 for the 21 stars that we recovered in the DR2 catalog after our quality cuts. The comparison is excellent, with an average metallicity offset between the two samples of 0.09 dex and a standard deviation in the residuals of 0.17 dex. In Figure 1, we also compare our values to the high-resolution samples of Barklem et al. ([2005](#page-6-0)), Marino et al. ([2019](#page-6-0)), and Ezzeddine et al. ([2020](#page-6-0)) for stars covering therange of our photometric metallicities ([Fe/ H > -3.75). Results largely follow what is presented in the comparisons to APOGEE and LAMOST and X. Ou et al. (2020, in preparation), but the Barklem et al. ([2005](#page-6-0)) comparison shows an offset of 0.25 dex, largely due to them lacking an empirical correction on T_{eff} (Frebel et al. [2013](#page-6-0)). The standard deviation of the residuals of our photometric metallicities and the high-resolution spectroscopic studies is 0.29 dex.

We note that in Chiti et al. ([2021](#page-6-0)), we find that the completeness of our sample is independent of metallicity down to $[Fe/H] \sim -3.0$. Consequently, when presenting quantitative analysis (e.g., fitting an exponential slope) of the metallicity distribution function of our sample, we only include

⁵ http://[skymapper.anu.edu.au](http://skymapper.anu.edu.au/lter-transformations/)/lter-transformations/
⁶ We note that 5% of stars have photometric metallicities up to 0.1 dex higher than $[Fe/H] = -0.75$ due to a metallicity correction based on the spatial location of stars (see Section 2.3 of Chiti et al. [2021](#page-6-0)).

Figure 2. Low-metallicity map of stars with $[Fe/H] \le -0.75$ in the R–|Z| plane, where R is the galactocentric distance and $|Z|$ is the absolute value of the height above the galactic plane. A clear, decreasing metallicity gradient is apparent as a function of $|Z|$, and the Milky Way disk is visible via a drop-off in average metallicity above $|Z| \sim 2.8$ kpc. Each cell in the plot has dimensions of 300 pc by 300 pc, and is colored by the mean metallicity of stars contained within its region. The dashed semicircle includes stars with distances < 5.0 kpc, within which target selection effects do not appreciably bias the metallicity of the sample (see Section [2](#page-1-0) for discussion).

photometric metallicities down to $[Fe/H] = -3.0$ in the fitting procedures.

An additional bias is related to target selection effects, since metal-poor stars are brighter than metal-rich stars at the same effective temperature. Consequently, our sample may be preferentially biased toward low-metallicity stars at larger distances as our quality cuts will exclude more metal-rich stars at larger distances. As an illustration of this effect, stars right at the threshold of our distance quality cut (uncertainties at the 18%–20% level) have a median distance of 6.4 kpc when $-1.25 <$ [Fe/H] < -0.75 , but a median distance of 8.9 kpc when $-2.75 <$ [Fe/H] < -2.25 . However, this selection preference toward metal-poor stars is not significant within a distance of ∼5.0 kpc, as we find that at distances between 4.0 and 5.0 kpc, a roughly constant fraction of $\langle 1\% \rangle$ of stars are excluded from our full catalog by our quality cut as a function of metallicity. A weak trend with metallicity appears at distances between 5.0 kpc and 6.0 kpc, in which $\sim 6\%$ of stars are excluded with higher metallicities ($[Fe/H] > -1.25$), but only ∼2% of stars are excluded with low metallicities ([Fe/ H < -1.75). This suggests a slight preferential selection of low-metallicity stars at distances beyond ∼5.0 kpc, but no significant bias at distances nearer.

3. Metal-poor Stars in the Milky Way

Our low-metallicity map ($[Fe/H] \le -0.75$) of the Milky Way is shown in Figure 2 as a function of scale height $(|Z|)$ and distance from the Galactic center (R) . Using astropy (Astropy Collaboration et al. [2013](#page-6-0)), we transformed the coordinates of each star into the $R-Z$ plane, assuming the Sun's location of 8.1 kpc away from the Galactic center and 22 pc above the Galactic midplane (Gravity Collaboration et al. [2018](#page-6-0)). Each location of this map is colored by the average metallicity of stars in the corresponding spatial bin (300 pc by 300 pc in size). The disk/halo separation is immediately easy to recognize, due to the transition of the average metallicity from $[Fe/H] \ge -1.25$ to $[Fe/H] \le -1.25$ above $|Z| \sim 2.8$ kpc. This compares well to the metallicity map presented in Ivezić

et al. ([2008](#page-6-0)), whom also find a sharp break at $|Z| \sim 3$ kpc above which the mean metallicity drops below [Fe/H] ~ -1.25 .

In Figure [3,](#page-3-0) we show metallicity maps for several metallicity ranges (1 dex increments from $[Fe/H] = -1.0$ to $[Fe/H]$ H = −4.0) to display the structure of the Galaxy at progressively lower metallicities. The Milky Way disk is still clearly noticeable in our highest-metallicity bin $(-2.0 <$ [Fe/ H < -1.0), along with a diffuse population of more metalpoor stars $(-2.0 < [Fe/H] < -1.5)$ that follow no obvious spatial structure. Continuing this trend, the stars in the two lowest-metallicity bins $(-3.0 < [Fe/H] < -2.0$ and $-4.0 <$ [Fe/H] < -3.0) appear to show no spatial correlation with their mean metallicity.

To illustrate this trend, in the right panel of Figure [3](#page-3-0), we plot the mean metallicity of each of these metallicity bins as a function of scale height $|Z|$. While the mean metallicity of our most metal-rich bin shows a decreasing mean metallicity as a function of $|Z|$, the more metal-poor bins do not, suggesting no significant spatial correlation with metallicity below [Fe/ $[H] = -2.0$. Overall, this figure confirms that the most metalpoor stars are spatially distributed in all of the Galactic components, but their search in certain components (e.g., the disk) is simply impeded by the wealth of the more metal-rich stars. In fact, 0.8% of stars in our sample have $[Fe/H] < -2.5$ when $|Z| < 3$ kpc, but comprise 2.7% of stars when $|Z| > 3$ kpc.

In Figure [4,](#page-4-0) we present the metallicity distribution function (MDF) of our sample of metal-poor giants. We now include stars with a loosened distance quality cut (uncertainty of <40%), since those giants with unreliable distances are most likely to be associated with the outer halo and populate the most metal-poor regime of the MDF. As mentioned in Section [2](#page-1-0), we only include stars with $[Fe/H] > -3.0$ when fitting profiles to the MDF, given incompleteness effects when $[Fe/H] < -3.0.$

Below $[Fe/H] = -1.5$, the MDF smoothly decreases exponentially toward lower metallicities. Down to [Fe/ H | \sim −2.3, this behavior is well described by an exponential function with a slope of $(\Delta(\log N)) / (\Delta [Fe/H]) = 0.98 \pm 0.01$. This slope was derived through a Markov Chain Monte Carlo procedure by implementing the likelihood function for a truncated exponential distribution (Equation (8.1.17) in Cohen [1991](#page-6-0)) in the python package emcee (Foreman-Mackey et al. [2013](#page-6-0)). Then, the derived exponential slope was converted from $\Delta(\ln N)/\Delta$ [Fe/H], as was returned from this likelihood function, to $\Delta(\log N)/\Delta$ [Fe/H]. This slope suggests that the number of the metal-poor stars drops by a factor of ∼10 for every dex decrease in [Fe/H], supporting conventional wisdom that a factor of ∼10 fewer stars exist for each 1 dex drop in metallicity. Below $[Fe/H] = -2.3$, the slope steepens to $(\Delta(\log N)) / (\Delta \text{[Fe/H]}) = 1.52 \pm 0.05$, implying that stars with metallicities below this cutoff are progressively more difficult to find as this corresponds to a drop by a factor of ∼33 for every dex decrease in [Fe/H]. This translates into a frequency of 1.5% of finding stars with $[Fe/H] \sim -3.0$ and 0.2% when $[Fe/H] \sim -3.5$ among stars with $[Fe/H] < -1.5$ within our sample. Overall, the MDF is remarkably smooth down to the lowest [Fe/H], as was also found in previous studies based on smaller samples (Frebel et al. [2006](#page-6-0); Schörck et al. [2009;](#page-6-0) Yong et al. [2013](#page-6-0); Da Costa et al. [2019](#page-6-0)), and other large photometric studies (An et al. [2015;](#page-6-0) Youakim et al. [2020](#page-6-0)).

Figure 3. Left panels: low-metallicity maps of stars in different metallicity ranges. The thick disk is easily apparent among stars with $-2.0 <$ [Fe/H] < -1.0 . No significant structure is present among stars with $[Fe/H] < -2.0$. Right panel: mean metallicities of stars as a function of scale height |Z|, within 0.25 kpc bins in |Z| that correspond to the respective left panels. The disk region induces a clear metallicity gradient (top) that makes way for no significant change in [Fe/H] with increasing |Z|.

Only including stars with distances < 5.0 kpc, to account for target selection effects, results in a slope of $(\Delta(\log N)) / (\Delta [Fe/H]) = 1.09 \pm 0.02$ when $-2.3 < [Fe/H]$ H] < -1.5 and $(\Delta(\log N)) / (\Delta [Fe/H]) = 1.53 \pm 0.10$ when $-3.0 < [Fe/H] < -2.3$. We note that the slope at the more metal-poor end is negligibly steeper, suggesting a robustness to target selection effects. However, the slope at the more metalrich end is notably steeper. While this steepening plausibly does arise from addressing target selection effects that exclude metal-rich stars at large distances, it may also be caused by the exclusion of spatially distant stars that have an underlying MDF that is preferentially metal-poor (e.g., Ivezić et al. [2008](#page-6-0)).

Our derived exponential slope of the MDF is consistent with that of previous spectroscopic studies (Schörck et al. [2009](#page-6-0); Allende Prieto et al. [2014](#page-6-0)), but is steeper than what is found in others (An et al. [2015](#page-6-0); Youakim et al. [2020](#page-6-0)). By taking the relative ratio of the MDF in Schörck et al. ([2009](#page-6-0)) at [Fe/ H] = -3.0 and [Fe/H] = -2.3 suggests an exponential slope of $(\Delta(\log N)) / (\Delta [Fe/H]) \sim 1.73$, which is comparable to our result of $\Delta \log(N) / \Delta$ [Fe/H] = 1.52 \pm 0.05 or 1.53 \pm 0.10 in that regime. A similar calculation for the MDF presented in Allende Prieto et al. ([2014](#page-6-0)) suggests an exponential slope of $(\Delta(\log N)) / (\Delta [Fe/H]) \sim 1.66$. However, the same analysis returns an exponential slope of 0.98 from An et al. ([2015](#page-6-0)), and Youakim et al. ([2020](#page-6-0)) present a value of 1.0 ± 0.1 , with which our derived slope is somewhat discrepant. This suggests that underlying selection effects between our samples might be affecting the slope. For instance, this study only focuses on relatively cool giant stars within ∼7 kpc, whereas other studies purely focus on stars in the halo $($ >6 kpc away).

We also investigate how the shape of the MDF varies as a function of height above the galactic plane. In Figure [5,](#page-5-0) we show MDFs for four different scale heights. The closest range is $0.8 < |Z| < 1.2$ kpc, followed by $1.5 < |Z| < 2.0$ kpc, $3.0 < |Z|$ $|Z| < 4.0$ kpc, and $5 < |Z| < 7$ kpc. At low |Z|, the thick disk component clearly dominates the higher [Fe/H] end of the distribution, but wanes beyond 2 kpc. By 3 kpc the more metalrich population associated with the thick disk population no longer dominates the distribution, and the metal-weak tail of the thick disk becomes apparent around $[Fe/H] \sim -1.3$. Finally, beyond 5 kpc, the halo dominates, with the MDF peaking at $[Fe/H] \sim -1.5$ and showing a significant metalpoor tail. The more metal-rich thick disk population appears to be entirely removed.

In Figure [5,](#page-5-0) we also fit the very metal-poor tail of the MDF $(Fe/H] < -2.3$) with an exponential slope to see if variations occur with scale height. We find four slopes: $(\Delta(\log N))/(\Delta[Fe/H]) = 1.41 \pm 0.19$ when $0.8 <$ $|Z|$ < 1.2 kpc, 1.62 \pm 0.13 when 1.5 < $|Z|$ < 2.0 kpc, 1.46 ± 0.12 when $3.0 < |Z| < 4.0$ kpc, and 1.59 ± 0.13 when $5 < |Z| < 7$ kpc. Upon performing a chi-squared test of the hypothesis that the slope is constant as a function of $|Z|$, we find no evidence to reject that hypothesis. Consequently, from our analysis, there is no strong evidence that the behavior of the very metal-poor tail of the MDF varies as a function of scale height.

Finally, we note that metallicities derived from SkyMapper photometry are biased high in the case of carbon-enhanced stars, which would plausibly steepen our MDF since carbonenhanced stars are more prevalent at low metallicities. To

Figure 4. Top left: metallicity distribution function (MDF) below [Fe/H] < −0.75 of ∼122,000 metal-poor giants. Best-fitting exponential fits are shown, $(\Delta(\log N))/(\Delta[Fe/H]) = 0.98 \pm 0.01$ when $-2.3 <$ [Fe/H] < -1.5 and $(\Delta(\log N))/(\Delta[Fe/H]) = 1.52 \pm 0.05$ when $-3.0 <$ [Fe/H] < -2.3 . Fitting for the more metal-poor regime only includes stars $[Fe/H] > -3.0$, to avoid incompleteness effects (see Section [2](#page-1-0) for discussion). Top right: MDF of stars with $[Fe/H] < -3.0$ in our sample. Bottom left and right: same as the top row, but only including stars with distances < 5.0 kpc to minimize target selection effects (see Section [2](#page-1-0) for discussion).

estimate whether this affect could appreciably alter our MDF, we used the sample of Placco et al. (2014) (2014) (2014) to investigate the carbon abundance ($[C/Fe]$) of stars between $-3.0 < [Fe/$ H $<$ -2.0 . We find that the average carbon abundance of stars increases by 0.13 dex between $[Fe/H] = -2.0$ and $[Fe/H]$ H] = -3.0 . At a typical $T_{\text{eff}} = 4500 \text{ K}$, such an increase in the carbon abundance stars with $[Fe/H] = -3.0$ would only increase the photometric metallicity by 0.07 dex. Consequently, while we do acknowledge that individual outliers with extremely high carbon abundances would appear to have much higher metallicities, at a population level, the metallicity of a typical star is not significantly affected.

4. Summary and Conclusion

We have presented a low-metallicity map of the Milky Way based on photometric metallicities obtained from new public SkyMapper DR2 data of ∼110,000 red giants in the southern sky with $-3.75 <$ [Fe/H] < -0.75 and $g < 17$ out to \sim 7 kpc from the solar neighborhood. Based on a comparison with new high-resolution spectra as well as using several literature samples and surveys, we demonstrate this sample achieves reliable metallicity precision ($\sigma \sim 0.24$ dex to 0.29 dex) down to [Fe/H] \sim −3.3. We display the spatial distribution of stars within the very and extremely metal-poor regimes providing an unprecedented view into the ancient, metal-poor Milky Way.

We find clear signatures of the components of the Milky Way in our data set. Notably, the thick disk prominently appears in our population of stars with $[Fe/H] > -1.50$ (see Figure [3](#page-3-0)), and the transition from the thick disk to the inner halo is reflected in a sharp drop in metallicities around $Z| \sim 2.8$ $Z| \sim 2.8$ $Z| \sim 2.8$ kpc (see Figure 2). The average metallicities of stars in our sample progressively decrease out to $|Z| \sim 7$ kpc. At even larger distances, we do not have enough coverage to confidently map out metallicity averages but observe hints of an additional small decrease, as expected for the outer halo. It is worth pointing out that a smooth distribution of very and extremely metal-poor stars exists across the entire sky covered by this sample, even within the more metal-rich thick disk. This highlights that the most metal-poor stars are well spread across the galaxy, likely owing to their orbital properties that regularly bring to the inner portions of the Galaxy. We explore this topic in a separate paper (M. Mardini et al. 2021, in preparation), in which we quantify the kinematic properties of our sample of stars and find that \sim 200 stars with [Fe/H] < −2.3 in our sample have orbits consistent with membership to the thick disk. This result further highlights that very and extremely metal-poor stars exist in the thick disk of the galaxy.

Figure 5. Metallicity distribution of ~122,000 stars, arranged as a function of distance from the Galactic plane in four different ranges, mimicking the inner thick disk (top left), the outer thick disk (bottom left), the metal-weak tail of the thick disk (top right), and the halo (bottom right). Insets show the MDFs of stars with [Fe/ H < -2.3 , with the best-fitting exponential slope indicated.

We are also able to quantify the shape of the MDF for [Fe/ H \leq -1.5. It is well fit by two exponential profiles with a slope of $\Delta \log(N)/\Delta$ [Fe/H] = 0.98 \pm 0.01 for stars with $-2.3 <$ [Fe/H] < -1.5 , and $\Delta \log(N)/\Delta$ [Fe/H] = -1.52 ± 0.05 when $-3.0 <$ [Fe/ $[H] < -2.3$. When only including stars within 5.0 kpc, to account for biases from target selection, we derive slopes of $\Delta \log(N)/\Delta$ [Fe/H] = 1.09 \pm 0.02 for stars with $-2.3 <$ [Fe/H] < -1.5 , and $\Delta \log(N) / \Delta$ [Fe/H] = -1.53 ± 0.05 when $-3.0 <$ [Fe/ H \leq -2.3. In both cases, the MDF steepens at low metallicities, highlighting the difficulty of finding the most metal-poor stars. More intuitively, the slope from our entire sample implies that the number of metal-poor stars drops by a factor of ~ 10 for every dex decrease in [Fe/H] when $-2.3 <$ [Fe/H] < -1.5 and a factor of \sim 33 for every dex decrease in $[Fe/H]$ when $-3.0 < [Fe/H] < -2.3$. Unsurprisingly, we also find that the relative frequency of the most metal-poor stars increases as a function of $|Z|$, as 0.8% of stars in our sample have $[Fe/H] < -2.5$ when $|Z| < 3$ kpc but comprise 2.7% of stars when $|Z| > 3$ kpc. Finally, we find some evidence that the MDF for stars with $[Fe/H] < -2.3$ is largely insensitive to scale height |Z| out to \sim 5 kpc, suggesting that the most metal-poor stars are spatially distributed in every galactic component. Overall, our sample contains ∼140 giants with $[Fe/H] < -3.0$, and ~35 with $[Fe/H] < -3.3$. Detailed follow-up studies of these stars are underway that should help to reveal the full extent of the nature and origin of the ancient component of our Milky Way.

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