The distribution of [$\alpha$/Fe] in the Milky Way disc

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ABSTRACT

Using a sample of red giant stars from the Apache Point Observatory Galactic Evolution Experiment (APOGEE) Data Release 16, we infer the conditional distribution $p([\alpha$/Fe] | [Fe/H]) in the Milky Way disk for the $\alpha$-elements Mg, O, Si, S, and Ca. In each bin of [Fe/H] and Galactocentric radius $R$, we model $p([\alpha$/Fe]) as a sum of two Gaussians, representing “low-$\alpha$” and “high-$\alpha$” populations with scale heights $z_1 \approx 0.45$ kpc and $z_2 \approx 0.95$ kpc, respectively. By accounting for age-dependent and $z$-dependent selection effects in APOGEE, we infer the [$\alpha$/Fe] distributions that would be found for a fair sample of long-lived stars covering all $z$. Near the Solar circle, this distribution is bimodal at sub-solar [Fe/H], with the low-$\alpha$ and high-$\alpha$ peaks clearly separated by a minimum at intermediate [$\alpha$/Fe]. In agreement with previous results, we find that the high-$\alpha$ population is more prominent at smaller $R$, lower [Fe/H], and larger $|z|$, and that the sequence separation is smaller for Si and Ca than for Mg, O, and S. We find significant intrinsic scatter in [$\alpha$/Fe] at fixed [Fe/H] for both the low-$\alpha$ and high-$\alpha$ populations, typically $\sim 0.04$-dex. The means, dispersions, and relative amplitudes of this two-Gaussian description, and the dependence of these parameters on $R$, [Fe/H], and $\alpha$-element, provide a quantitative target for chemical evolution models and a test for hydrodynamic simulations of disk galaxy formation. We argue that explaining the observed bimodality will probably require one or more sharp transitions in the disk’s gas accretion, star formation, or outflow history in addition to radial mixing of stellar populations.

Key words: Galaxy: abundances – Galaxy: disc – Galaxy: stellar content – stars: abundances

1 INTRODUCTION

The distribution of stars in the space of $\alpha$-element abundances and iron abundances is a powerful diagnostic of the star formation history (SFH) and stellar nucleosynthesis in galaxies, as the $\alpha$-elements come primarily from core collapse supernovae (CCSN) while iron comes also from Type Ia supernova (SNIa) enrichment with a longer timescale (e.g., Tinsley 1980; Matteucci & Greggio 1986; McWilliam 1997). The evolutionary track of a stellar population in [$\alpha$/Fe] − [Fe/H] provides a diagnostic of star formation efficiency, accretion history, and gas outflows, as well as supernova yields (see, e.g., Matteucci 2011; Andrews et al. 2017; Weinberg et al. 2017, hereafter WAF). Stars with kinematics or geometry characteristic of the “thick disk” (Gilmore & Reid 1983) exhibit higher values of [$\alpha$/Fe] at a given [Fe/H] (Fuhrmann 1998; Bensby, Feltzing, & Lundström 2003). This separation is clear enough that many studies now use [$\alpha$/Fe] ratios in place of kinematics to define a thick disk population (e.g., Lee et al. 2011; Haywood et al. 2013; Mackereith et al. 2017). However, the correlation of abundance patterns with geometry, kinematics, and stellar age makes the distribution of stars in [$\alpha$/Fe] − [Fe/H] highly dependent on sample selection, complicating the challenge of using this distribution to infer the enrichment history of the Milky Way.

In this paper we examine the [$\alpha$/Fe] − [Fe/H] distribution of disk stars in the Apache Point Observatory Galactic Evolution Experiment (APOGEE; Majewski et al. 2017). In some zones of Galactocentric radius $R$ and midplane distance $z$, this distribution appears distinctly bimodal, with two separated tracks at low [Fe/H] that merge at [Fe/H] $\gtrsim 0$ (Hayden et al. 2015, hereafter H15). However, in other zones the bimodality is less clear, and it is not obviously present in some other studies such as Edvardsson et al. (1993, fig. 15) or Data Release 2 (DR2) of the Galactic Archaeology

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$^1$ We follow standard abundance notation in which $[X/Y] = \log X/X_0 - \log Y/Y_0$. 

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with HERMES (GALAH) survey (Buder et al. 2018, fig. 22; but see fig. 5 of the DR3 paper Buder et al. 2020). The sample selection may differ a lot in the various surveys, as – for example – APOGEE focuses on red giants, whereas GALAH focuses on dwarfs. Dwarfs and giants have their own issues with respect to abundance determinations. In this context, an interesting spectroscopic survey is the RAdial Velocity Experiment (RAVE; e.g., see fig. 22 in Guiglion et al. 2020), which probes a composite population of dwarfs and red giants.

For our analysis here, we fit an empirical model to the APOGEE red giant disk population, accounting for selection effects, to infer the intrinsic distribution of \( [\alpha/Fe] \) ratios in bins of \([\text{Fe/H}] \) and \( R \) for the \( \alpha \)-elements Mg, O, Si, S, and Ca. By intrinsic, we mean the distribution that would be measured if one could observe an unbiased subset of all disk stars with main sequence lifetimes longer than the age of the Galaxy. This intrinsic distribution can be compared to predictions of numerical simulations or galactic chemical evolution (GCE) calculations without modeling the detailed geometry and sample selection of APOGEE.

A variety of scenarios have been proposed to explain the co-existence of high-\( \alpha \) and low-\( \alpha \) populations.\(^2\) The two-infall model posits two waves of star formation separated by a period of pristine gas accretion with minimal star formation. The first wave produces the high-\( \alpha \) population, and infall resets the gas phase metallicity to low values before the second wave produces the low-\( \alpha \) population (e.g., see Chiappini, Matteucci, & Gratton 1997; Noguchi 2018; Spitoni et al. 2019; Pallà et al. 2020, but also Lian et al. 2020). In this context, Vincenzo et al. (2019) proposed that the physical mechanism responsible for a temporary quenching of the star-formation history of the MW at high redshifts was a major merger event with a galaxy like Gaia-Enceladus or Gaia-Sausage (Helmi et al. 2018; Belokurov et al. 2018); this merger may have also kinematically heated part of the present-day inner halo and thick-disk (e.g., Haywood et al. 2018; Chaplin et al. 2020; Montalbán et al. 2020). A possible scenario could be that the merger event between the MW progenitor and Gaia-Sausage/Enceladus activated the central black hole of our Galaxy at high redshifts, heating up the gas in the DM halo over a limited period of time, eventually giving rise to a second gas accretion event which brought the MW back to the main sequence of star-forming disk galaxies (see the simulations of Pontzen et al. 2017 for the feasibility of this scenario).

In contrast to the two-infall picture, Schönrich & Binney (2009) argue that the low-\( \alpha \)-population is not an evolutionary sequence at all but instead comprises the end-points of evolutionary tracks at different radii, which are then mixed by stellar migration (see also Nidever et al. 2014; Sharma, Hayden, & Bland-Hawthorn 2020). Conversely, Clarke et al. (2019) suggest that the low-\( \alpha \)-population is the true evolutionary sequence and the high-\( \alpha \)-population forms in massive clumps of the gas rich early disk, which self-enrich with CCSN elements. A similar phenomenon is seen in the simulations of Vincenzo & Kobayashi (2020) and Khoperskov et al. (2020).

The success of hydrodynamic cosmological simulations in producing distinct \( [\alpha/Fe] \) – \([\text{Fe/H}] \) sequences is mixed. In these simulations, stellar migrations, star-formation and chemical enrichment, outflows and inflows of gas, and cosmological growth are all self-consistently taken into account. Analyzing the EAGLE cosmological volume and the Auriga zoom-in simulations, respectively, Mackereth et al. (2019) and Grand et al. (2018) find that bimodality of \( [\alpha/Fe] \) arises in a small fraction (\( \sim 10% \)) of simulated galaxies that have uncommon, 2-phase accretion histories. However, Buck (2020) finds bimodality in all four simulations analyzed from the NIHAO-UHD suite, a consequence of gas rich mergers that bring in low metallicity fuel, reminiscent of the 2-infall picture. An interesting simulation-based study is that of Vincenzo & Kobayashi (2020), who can qualitatively reproduce both the bimodality in \( [\alpha/Fe]-[\text{Fe/H}] \) and the slope of the metallicity gradient as observed by APOGEE-DR16. In this simulation, accretion and gas flows play the primary role in determining \( [\alpha/Fe]-[\text{Fe/H}] \)-age distributions, and stellar migration has a secondary impact.

Testing the predictions of models and simulations requires an accurate quantification of the intrinsic \( [\alpha/Fe] \) distribution across the Milky Way (MW) disk. There are two reasons one cannot simply use the observed \( [\alpha/Fe] \) ratios of the APOGEE sample to provide this distribution. The first is that higher \( [\alpha/Fe] \) stars are systematically older, and the fraction of a stellar population’s original stars that are red giants at a given time depends on the population’s age. The second is that higher \( [\alpha/Fe] \) stars have a larger scale height, and the APOGEE sample has a complex selection in \( |z| \) (Zasowski et al. 2013, 2017; see fig. 5 below). For each \( \alpha \)-element that we consider, we model the conditional distribution \( p([\alpha/Fe]|R) \) in bins of \( R \) and \([\text{Fe/H}] \) as a double-Gaussian, parameterized by the means and dispersions of the high-[\( \alpha/Fe] \) and low-[\( \alpha/Fe] \) populations and by their relative normalization, adopting empirically motivated scale heights for each population. Our analysis covers the range \( 3 \text{kpc} \leq R \leq 11 \text{kpc} \), \( 0 \leq |z| \leq 2 \text{kpc} \), and \( -0.5 \leq [\text{Fe/H}] < 0.2 \).

We adopt a two-component form because it provides a good description of the data, but we make no assertion about whether these components have physically distinct origins. In the case of the vertical distribution of disk stars, Bovy et al. (2012) argue that the double-exponential form suggesting distinct thin and thick disks emerges from a continuous dependence of single-exponential scale heights on abundances, which are themselves correlated with population age. While the degree of bimodality in \( [\alpha/Fe] \) is our primary target, the intrinsic scatter within each population is also a quantity of interest, an important test of stochastic enrichment models and radial mixing of populations. The relative mean sequences for the different elements constrains the relative contribution of SNIa enrichment to that element (WAF; Weinberg et al. 2019).

Our work is organised as follows. In Section 2, we describe the assumed data set and discuss the main selection effects. In Section 3 we describe the formalism and the assumptions of our model and present the methods employed to fit the observed chemical abundance distributions. In Section 4, we present our findings for the conditional distributions of \( p([\alpha/Fe]|R) \) in bins of \( R \) and \([\text{Fe/H}] \), including the degree of bimodality, the intrinsic scatter for high-\( \alpha \) and low-\( \alpha \) populations, and the dependence on the choice of \( \alpha \)-elements. We summarize our results and discuss implications in Section 5.

2 APOGEE DATA AND SELECTION EFFECTS

2.1 Data Sample

Our sample is selected from SDSS-IV APOGEE-2 DR16 (Ahumada et al. 2020) with the following cuts in effective temperature, surface gravity, and signal-to-noise ratio (SNR):

\[ \text{MNRAS 000, 1–?? (2021)} \]
(i) Effective temperature in the range $4200 \leq T_{\text{eff}} < 4600$ K,
(ii) Surface gravity in the range $1 \leq \log(g/[\text{cm s}^{-2}]) < 2.5$,
(iii) Signal-to-noise ratio SNR > 80.

The abundance determination methods and calibrations are described by Holtzman et al. (2015), García Pérez et al. (2016), and Jönsson et al. (2020). In Fig. 1 we show how the observed [Mg/Fe]-[Fe/H] abundance diagram varies within the Galactic disc, for different ranges of Galactocentric radius $R$ and height $z$, given the adopted selections. The distances are derived with AstroNN (Leung & Bovy 2019a,b) and are publicly available as a value-added catalogue for APOGEE-DR16. Our analysis focuses on stars with iron abundances in the range $-0.5 \leq [\text{Fe/H}] \leq 0.2$.

### 2.2 Age-Selection Effects

Selection effects may arise because we limit our analysis to stars in a specific range of $\log(g)$ and $T_{\text{eff}}$. This selection can be age-dependent because the fraction of stars within the selection window changes with the age of the population. We account for this effect by including in our model a selection factor, which is computed as

$$f(\tau) = \frac{\int_{M_i}^{M_f} dM \frac{dN}{dM}}{\int_{M_{\text{min}}}^{M_{\text{max}}} dM \frac{dN}{dM}},$$

(1)

where $dN/dM$ is the initial mass function (IMF) and $M_i$ and $M_f$ are the initial masses of the stars that are in the $\log(g)$ and $T_{\text{eff}}$ range for a population of age $\tau$. We assume that all stars in this range have an equal chance of being selected as APOGEE targets; the impact of this assumption on the age distribution of the stars is quantified in Section 2.4. Finally, the quantities $M_{\text{min}}$ and $M_{\text{max}}$ in equation (1) denote the current mass range of the IMF, but the denominator cancels out in our analysis so the choice is arbitrary. It is the number fraction that matters, not the mass fraction, because we are counting stars. Because the observed age-[α/Fe] relation is much tighter than the observed age-[Fe/H] relation, we will use $[\alpha/\text{Fe}]$ to compute $f(\tau)$ and ignore any dependence on [Fe/H] at fixed $[\alpha/\text{Fe}]$.

In Fig. 2 we show how $f$ varies as a function of the stellar age, by adopting the PARSEC stellar evolutionary tracks (release v1.2S) at $Z = 0.01$ (Bressan et al. 2012; Tang et al. 2014; Chen et al. 2015) and the IMF of Kroupa (2001). Different panels correspond to different maximum values of $T_{\text{eff}}$ in the selection window, from $T_{\text{eff,max}} = 4600$ K to 4800 K, since $f$ does not change considerably for maximum temperatures $T_{\text{eff,max}} > 4800$ K. By decreasing $T_{\text{eff,max}}$ from 4800 K to 4600 K, we can effectively suppress the contribution of stars in the red clump (which dominate at ages $\geq 4$ Gyr) and in the secondary clump (which dominate at ages $\leq 2$ Gyr; we address the readers to Girardi 1999, 2016 for details). In Fig. 3 we show how red-giant branch and He-burning stars of different age evolve in $\log(g)$ and $T_{\text{eff}}$ as a function of the stellar mass, normalized to the main-sequence turn-off mass (maximum mass on the main-sequence for each stellar age in the figure), by considering isochrones of different age. The secondary clump in Fig. 3 is represented by the extended He-burning tracks at ages $< 2$ Gyr.

The encouraging feature of Fig. 2 is that, for $T_{\text{eff,max}} = 4600$ K, the selection factor $f$ is nearly constant for $\tau > 2.5$ Gyr. There is a moderate rise at $\tau = 2$ Gyr, where a somewhat larger initial mass range resides in the $\log(g) - T_{\text{eff}}$ range (see Fig. 3). There is a more pronounced deficit at $\tau \approx 1$ Gyr, where tracks do not fully cross $\log(g) = 1 - 2.5$. There is a spike at $\tau \approx 100$ Myr, but this is very narrow. The approximate constancy of $f$ means that moderate errors
in typical stellar ages will have little impact on the recovered \([\alpha/Fe]\) distributions. Ranges of \([\alpha/Fe]\) that are sharply peaked near ages of 1 Gyr or 2 Gyr will have larger corrections and uncertainties.

When we fit our model to the observed \([\alpha/Fe]\) distributions in the Galaxy disc, we need an estimate of the age of the stars to compute the correction factor, \(f\). In this work, we fit the age-\([Mg/Fe]\) relation as measured by the asteroseismic analysis of Miglio et al. (2020, see Fig. 4) for a sample of stars in the Kepler fields. For the fitting function, we adopt a series of Legendre polynomials of degree 5. The results of our analysis are shown in Fig. 4, where the coefficients of the best fit are also reported.

We include dispersion of the age-\([Mg/Fe]\) relation in our calculations by fitting the asteroseismic measurement errors as a function of the stellar age. The best-fit description has the following functional form:

\[
\sigma_{\text{age}} = \begin{cases} 
0.482 \times \frac{\text{age}}{\text{Gyr}} - 0.395 \text{Gyr} & \text{if age} > 1.65 \text{Gyr} \\
0.4 \text{Gyr} & \text{if age} \leq 1.65 \text{Gyr}
\end{cases}
\]

We enforce a minimum dispersion of 0.4 Gyr. Equation (2) represents observational errors rather than intrinsic dispersion, but the dispersion plays only a small role in our analysis, so this rough estimate is sufficient. In Fig. 4, stars with \([Mg/Fe] > 0.1\) lie well within the orange band implied by equation (2) because the age error is predominantly systematic rather than a random error that adds star-by-star scatter.

Even though the sample of Miglio et al. (2020) mostly comprises stars in the Solar neighborhood, we apply equation (2) to the inner and outer annuli, being aware that the \([Mg/Fe]\)-age relation may change as a function of the Galactocentric distance, as predicted by chemical evolution models and simulations (e.g., see Vincenzo & Kobayashi 2020).

2.3 \(z\)-Selection Effects

Our model fits the observed distribution of stars in the ([Mg/Fe], \([z]\)) plane, for different bins of \(R\) and \([Fe/H]\). However, APOGEE does not sample correctly the true vertical distribution of the stars in the MW disc, since the observations are limited by the number of fibers on the plates in the focal plane of the telescope and by the sky coverage of the survey. For example, at many Galactic longitudes APOGEE observed five fields at \(b = -8, -4, 0, +4, +8\), obtaining spectra for \(\sim 250\) science targets in each 7-deg\(^2\) field. In the high density midplane these 250 targets are only a small fraction of the available giants that pass the APOGEE magnitude cut, while at higher latitudes the fraction is larger. Thus, this observing strategy tends to overrepresent high-\([z]\) stars in the sample.

The importance of assessing the \(z\)-selection effects is demonstrated in Fig. 5, in which we compare the vertical distribution of stars as observed by APOGEE-DR16 (for our adopted data cuts) with the predictions of our best model, using uniform bins in \([z]\) of 0.2 kpc. Since the parameters of the best model depend on \([Fe/H]\), in order to make Fig. 5 we assume that APOGEE correctly samples the metallicity distribution function (MDF) in the range \(-0.4 < [Fe/H] < 0.0\), and we compute the re-scaling factors for the model predictions to match the integrated number of stars over
gradually weakens with increasing age, from the analysis of Miglio et al. (2020).

errors as a function of the stellar age, from the analysis of Miglio et al. (2020). The orange area shows the asteroseismic sample of Miglio et al. (2020), and the red curve corresponds to the best fit in each panel. As shown previously in Fig. 2, populations of age $\tau \approx 2$ Gyr are moderately overrepresented while populations younger than 2 Gyr are underrepresented.

2.4 Apparent Magnitude Selection

The selection of APOGEE stars in apparent $H$-band magnitude $m_H$ is fairly complex, designed to produce a sample that maps the Galaxy effectively with an efficient observing strategy (Zasowski et al. 2013, 2017). This apparent magnitude selection can affect the relative probability of observing stellar populations of different ages, beyond the age-selection imposed by the log($g$) and $T_{\text{eff}}$ cuts which we have accounted for as described in §2.2. To test the possible importance of this effect, we have created Monte Carlo samples using PARSEC isochrones (see §2.2) spaced at 0.1-Gyr intervals from 0.1-10 Gyr. For each isochrone we make $10^6$ random draws from the Kroupa (2001) IMF and compute the log($g$), $T_{\text{eff}}$, and absolute $H$-band magnitude $M_H$ of the resulting stars, dropping any that have evolved past the AGB. We can then apply log($g$), $T_{\text{eff}}$, and distance-dependent $M_H$ cuts to see how they affect the age distribution $dN/d\tau$.

In the upper panels of Fig. 6, red curves (the same in each panel) show the age distribution of all surviving stars. Because we form equal numbers of stars in each time interval, these curves are nearly flat, declining slightly towards larger ages because a larger fraction of the IMF has evolved past the AGB. Blue curves (again the same in each panel) show the effect of imposing our 4200 K $\leq T_{\text{eff}} < 4600$ K and $1 \leq \log(g) < 2.5$ cuts. As shown previously in Fig. 2, populations of age $\tau \approx 2$ Gyr are moderately overrepresented while populations younger than 2 Gyr are underrepresented.

In the lower panels of Fig. 6, black histograms show the $m_H$ distribution of stars in our APOGEE sample in four ranges of distance from the sun, centred at $d = 2, 4, 6,$ and 8 kpc. For these histograms we have selected stars with $|z| \leq 0.5$ kpc. Blue histograms show the distribution of $m_H = M_H + \mu(d)$ for stars in our Monte Carlo sample satisfying our log($g$) and $T_{\text{eff}}$ selection cuts, where $\mu(d)$ is the distance modulus at the bin center. At small distances these histograms are peaked at lower $m_H$ than the observed sample distribution. Finally, we draw stars from this Monte Carlo sample with weighting in $M_H$ to reproduce, approximately, the observed $m_H$ distributions, obtaining the green shaded histograms. The observed histograms extend to fainter $m_H$ at small distances, which simply indicates that stars with log($g$) < 2.5 are too luminous to have a large $m_H$ at small $d$.

Returning to the upper panels, green curves show $dN/d\tau$ for this reweighted sample. These curves are noisy because the number of stars passing all cuts for a given isochrone is small, and we have smoothed them with a 1 Gyr boxcar filter to improve readability. For the most part, the magnitude selection produces only moderate changes to $dN/d\tau$ beyond those already caused by the log($g$) and $T_{\text{eff}}$ selection. The most significant differences are a somewhat higher overrepresentation of 2-4 Gyr old stars and a suppression of very young populations with $\tau < 0.5$ Gyr, whose log($g$) < 2.5 giants are sufficiently luminous that they lie off the bright end of the APOGEE $m_H$ histograms.

In principle we could incorporate $m_H$ selection effects in our modeling by fitting the observed counts in narrow bins of $[\alpha/Fe]$, $|z|$, and $m_H$, within each larger bin of [Fe/H] and $R$. Because the...
effects in Fig. 6 are fairly small, we have chosen not to incur this additional complication, making only the corrections described in §2.2 and §2.3. Somewhat fortuitously, the age-selection effects in APOGEE are nearly flat except for narrow ranges of $\tau$, and the $|z|$-selection effects illustrated in Fig. 5 have greater impact. As a further test, we have checked that the $M_H$ distributions of the observed low-$\alpha$ and high-$\alpha$ stars in a given bin of $R, |z|$, and [Fe/H] largely overlap. For $R > 7$ kpc the low-$\alpha$ stars, which are younger on average, are shifted towards brighter $M_H$ by 0.5-1 mag, but this difference is smaller than the width of the distributions.

3 MODEL AND METHODS

3.1 The Model

We derive the underlying distribution of $p(\alpha | \text{Fe/H})$, where $\alpha$ represents an individual $\alpha$-element. We model $p(\alpha | \text{Fe/H})$ as the sum of probability density functions (PDFs) of two populations, each characterised by a mean, $\mu$, and dispersion, $\sigma$. These PDFs are denoted as $F_1(y|\mu_1,\sigma_1)$ and $F_2(y|\mu_2,\sigma_2)$, where $y = [\alpha/\text{Fe}]$ and $\int dy F_1(y) = \int dy F_2(y) = 1$, with subscript 1 and 2 referring to the low-$\alpha$ and high-$\alpha$ populations, respectively. In this work we assume that both PDFs are Gaussians, namely

$$F(y|\mu,\sigma) = \left(2\pi\sigma^2\right)^{-1/2} e^{-\frac{(y-\mu)^2}{2\sigma^2}}.$$  \hspace{1cm} (3)

We assume that each population has an exponential $z$ distribution (for brevity we write $z$ instead of $|z|$) with scale heights $z_1 = 0.45$ kpc and $z_2 = 0.95$ kpc based on the sub-solar metallicity values suggested by Bovy et al. (2016). The populations have

$$n(z) = n_1 e^{-z/z_1} \text{ and } n_2 e^{-z/z_2}$$  \hspace{1cm} (4)

stars per unit $z$.

In our analysis, we seek to constrain the quantities $n_1, \mu_1, \sigma_1$, and $n_2, \mu_2, \sigma_2$, using APOGEE data in narrow bins of $z$. If we take narrow bins $\Delta y \ll \sigma_1, \sigma_2$ and $\Delta z \ll z_1, z_2$, then the number of stars in the bin $y_i \leq y < y_i + \Delta y$ and $z_j \leq z < z_j + \Delta z$ is given by

$$\lambda_{ij,\text{model}} = n_1 F_1(y_i + \frac{\Delta y}{2}, \mu_1, \sigma_1) e^{-\frac{(z_j + \Delta z)}{z_1}} + n_2 F_2(y_i + \frac{\Delta y}{2}, \mu_2, \sigma_2) e^{-\frac{(z_j + \Delta z)}{z_2}}.$$  \hspace{1cm} (5)

Including the correction factors, $f$, for the selection in $T_{\text{eff}}$ and log($g$), the predicted number of stars in the bin $ij$ is

$$\lambda_{ij} = Q_f(\tau_i) n_1 F_1(y_i + \frac{\Delta y}{2}, \mu_1, \sigma_1) e^{-\frac{(z_j + \Delta z)}{z_1}} + n_2 F_2(y_i + \frac{\Delta y}{2}, \mu_2, \sigma_2) e^{-\frac{(z_j + \Delta z)}{z_2}}.$$  \hspace{1cm} (6)

where the value of the normalising constant, $Q_f$, is chosen so that

$$\sum_i \lambda_{ij} = K_f,$$  \hspace{1cm} (7)

where $K_f$ is the number of observed stars in the ([Fe/H], $R$) bin with $z_j \leq z < z_j + \Delta z$.

The values of $f(\tau_i)$ in equation (6) – with $\tau_i$ corresponding to the ages of the stars in the bin $y_i \leq y < y_i + \Delta y$ – are estimated by computing the average of $f(\tau)$ of $N_{\text{trans}} = 10$ random ages drawn from a normal distribution, with the mean value being given by the age-[Mg/Fe] relation of Miglio et al. (2020), see Fig. 4) and the dispersion given by the measured asteroseismic errors on the ages (see equation 2). This procedure is done independently for each [Fe/H] bin. Although we include scatter to represent the fact that stars of a given [Mg/Fe] have a range of ages, we have checked that choosing all ages from the mean relation with no dispersion makes minimal difference to our results.

In summary, the free parameters of our model in each bin of $R$ and [Fe/H] are $n_1, \mu_1, \sigma_1$, and $n_2, \mu_2, \sigma_2$. However, we
cannot constrain the true values of $n_1$ and $n_2$ separately from the observational data, but only their ratio $n_1/n_2$.

### 3.2 Maximum Likelihood Fitting

Let $k_{ij}$ denote the observed number of stars in the bin $y_i \leq y < y_i + \Delta y$ and $z_j \leq z < z_j + \Delta z$. We assume that the probability of observing $k_{ij}$ stars in the bin $ij$, given a predicted number $\lambda_{ij}$ of stars from the model (see equation (6)), is described by the Poisson distribution:

$$p(k_{ij}|\lambda_{ij}) = \frac{k_{ij}^{\lambda_{ij}} e^{-\lambda_{ij}}}{k_{ij}!}.$$  \hfill (8)

The global probability is

$$P(\{k_{ij}\}|\{\lambda_{ij}\}) = \prod_{ij} p(k_{ij}|\lambda_{ij}),$$  \hfill (9)

making the log-likelihood

$$\ln L = \sum_{ij} \ln p(k_{ij}|\lambda_{ij}) = \sum_{ij} k_{ij} \ln \lambda_{ij} - \lambda_{ij} + \text{const}.$$  \hfill (10)

The constant in equation (10) depends on the data but not on the parameters, so it can be ignored in finding the maximum likelihood solution.

In our analysis, we consider bins of iron abundances in the range $-0.5 \leq [\text{Fe/H}] < 0.2$ with steps $\Delta [\text{Fe/H}] = 0.1$ and bins of $R$ in the range $3 \leq R < 11$ kpc with steps $\Delta R = 2$ kpc. For each bin of $[\text{Fe/H}]$ and $R$, we run our maximum likelihood fitting algorithm using $\Delta y = 0.01$ dex and $\Delta z = 0.2$ kpc. We optimise $n_1$, $\mu_1$, $\sigma_1$, $n_2$, $\mu_2$, $\sigma_2$ by iterating 1-d grid searches with other parameters being held fixed. Convergence is usually reached well within $\approx 10$ iterations for $[\text{Fe/H}] < 0.1$, but it requires $\approx 20$ iterations in the range $0.1 \leq [\text{Fe/H}] < 0.2$ outside of the Solar circle. We also tested alternative methods based on Markov chain Monte Carlo sampling (EMCEE package; Foreman-Mackey et al. 2013), obtaining similar results.

#### 3.3 Mock Catalogues

We validate our analysis procedure by applying it to mock catalogues. First we create intrinsic population lists for the high-$\alpha$ and low-$\alpha$ populations as follows:

(i) We assume that the two populations of artificial stars have a total number $N_1 = 10^6$ for the low-$\alpha$ sequence, and $N_2 = R_{21} \times N_1$ for the high-$\alpha$ sequence, with $R_{21}$ being a free parameter.

(ii) We assign a $[\text{Mg/Fe}]$ ratio to the artificial stars in the two populations, by randomly sampling two Gaussian functions with different mean values, $\mu_1$ and $\mu_2$, and dispersions, $\sigma_1$ and $\sigma_2$.

(iii) We assign a value of $|z|$ to the artificial stars by randomly sampling an exponential function of scale height $z_1 = 0.45$ kpc or $z_2 = 0.95$ kpc for the corresponding population. The sampling is performed for $0 \leq |z| < 10$ kpc.

(iv) The ages of the artificial stars are assigned from the average age-$[\text{Mg/Fe}]$ of Miglio et al. (2020, see Fig. 4), by using the $[\text{Mg/Fe}]$ ratios of the artificial stars from point (ii) above. For each age, $\tau$, we tabulate $f(\tau)$, as computed with equation (1).

We choose parameters for our mock catalogue such that it qualitatively resembles our eventual findings for the APOGEE data. For the low-$\alpha$ population we take

$$\mu_1 = 0.05$$

at all $[\text{Fe/H}]$ and

$$\sigma_1 = [0.04, 0.04, 0.04, 0.04, 0.03, 0.03, 0.03]$$

in seven $\Delta[\text{Fe/H}] = 0.1$ bins spanning $-0.5$ to $+0.2$. For the high-$\alpha$ population we take

$$\mu_2 = [0.3, 0.27, 0.25, 0.22, 0.2, 0.15, 0.15]$$

in the same $[\text{Fe/H}]$ bins and

$$\sigma_2 = 0.04$$

in all bins. We adopt a number ratio between the two populations (integrated over $|z|$) of

$$R_{21} = \frac{n_2}{n_1} = 1.0, 0.8, 0.6, 0.4, 0.3, 0.2, 0.1$$
Figure 8. Distribution of stars in the $\text{[Mg/Fe]} - |z|$ plane in ranges of $\text{[Fe/H]}$ (rows) and Galactocentric radius (columns), as labeled. Our sample of APOGEE-DR16 stars is shown by the colour-coded two-dimensional histograms. Red contours show our best fitting model with selection effects included, representing densities normalized to the maximum value in the $\text{[Fe/H]}$ bin of $0.1 \pm 0.1$, $1.0 \pm 1.0$, $2.0 \pm 2.0$, $3.0 \pm 3.0$, $4.0 \pm 4.0$, $5.0 \pm 5.0$, and $6.0 \pm 6.0$ per cent.

In each bin of $\Delta R = 2 \text{kpc}$ and $\Delta [\text{Fe/H}] = 0.1$, we divide the artificial stars of each population into narrow $|z|$ bins of width $0.2 \text{kpc}$. For each such bin we draw stars from the intrinsic lists with probability proportional to $f(\tau)$. This procedure is repeated until the new list has the same number of stars, $N(\text{[Fe/H]}, R, |z|)$, as in APOGEE-DR16 with our reference selection assumptions. We apply our analysis to the mock catalogue to determine the values of $\mu_1, \sigma_1, \mu_2, \sigma_2, R_{21}$, testing that our procedure recovers the original parameters assumed to create the intrinsic distributions.

The results of our analysis are shown in Fig. 7, in which the upper panels show our results for the ratio between the number of stars in the high-$\text{[Mg/Fe]}$ and low-$\text{[Mg/Fe]}$ sequences, $R_{21}$, integrated in $|z|$ from 0 to $\infty$. The largest discrepancies between the best-fitting model and mock data are found for $R_{21}$ at the lowest [Fe/H] bins, as well as for $\mu_2$ (mean [Mg/Fe] of the high-\(\alpha\) sequence) at the highest [Fe/H] bins, corresponding to $0.1 \leq [\text{Fe/H}] < 0.2$. Even these differences are small, however, and we conclude that our procedure can accurately recover the intrinsic distributions of $[\text{Fe/H}]$ in the presence of $|z|$ and age selection effects like those in APOGEE.
Here we integrate over $\Delta z$ to obtain the model predictions rather than evaluating equation (6) at the bin center. Fig. 11 shows all bins of $R$ for $-0.3 \leq [\text{Fe/H}] < -0.2$. Agreement for other metallicity bins is comparable. Although the functional form of the model is intrinsically smooth, for the comparisons to APOGEE we apply the age selection effects with random draws from $f(\tau|\alpha/\text{Fe})$, so the predicted histograms in these figures are not perfectly smooth.

Figs. 8-11 demonstrate good visual agreement between predicted and observed distributions. In Section 4.3 below, we show quantitatively that the level of agreement is consistent with expectations if the model correctly describes the true distributions up to Poisson fluctuations.

### 4.2 Model Parameters and Bimodality in $[\text{Mg/Fe}]$

Fig. 12 superposes our best-fit values of $\mu_1 \pm \sigma_1$ and $\mu_2 \pm \sigma_2$ on the $[\text{Mg/Fe}]-[\text{Fe/H}]$ abundance distribution of all sample stars with $|z| < 2$ kpc in the four Galactocentric radius bins. Except at the highest [Fe/H] values, the means $\mu_1$ and $\mu_2$ differ by more than $\sigma_1 + \sigma_2$, an indication of two distinct sequences at all $R$. As shown previously by H15 and W19, the high-$\alpha$ and low-$\alpha$ sequences have similar locations at all $R$, though in detail our inferred values of $\mu$ depend on $R$, and a fit enforcing constant values at all $R$ is significantly worse.

Physically, the mean value of $[\text{Mg/Fe}]$ for stars on the low-[Fe/H] plateau of the high-$\alpha$ sequence should represent the ratio of the IMF-averaged yield of Mg and Fe from massive stars, dying as core-collapse SNe. At [Fe/H] $\approx -0.5$, we find $\mu_2 \approx 0.29$ for $3 \leq R < 5$ kpc, and $\mu_2 \approx 0.28$ for $R \geq 5$ kpc. As [Fe/H] increases, the value of $\mu_2$ decreases because of the increasingly large contribution of Fe from Type Ia SNe. As discussed in the introduction, the origin of the low-$[\alpha/\text{Fe}]$ sequence is a matter of debate, including the extent to which it represents an evolutionary sequence at all.

Fig. 13 shows the ratio $R_{21}$ of stars in the high-$\alpha$ and low-$\alpha$ populations. Those ratios are integrated over $|z|$ and corrected for the age-dependent selection. They depend on $R$ and [Fe/H], but the $z$-dependence of the observed ratio is a consequence of the differing scale heights $z_2 = 0.95$ kpc and $z_1 = 0.45$ kpc. As one would guess from the uncorrected stellar distributions, the high-$\alpha$ population is more important at low [Fe/H] and small $R$. At the Solar circle, the populations are equal for $-0.5 \leq [\text{Fe/H}] < -0.3$, and the low-$\alpha$ population is larger at [Fe/H] $> -0.3$. For $R = 9 - 11$ kpc, the high-$\alpha$ population is subdominant at all [Fe/H] examined here. Varying $z_2$ has a little impact on our findings, whereas varying $z_1$ has an impact on $R_{21}$ at low [Fe/H], such that if $z_1$ is diminished, then $R_{21}$ at low [Fe/H] also diminishes in the fit to compensate for that variation.

Fig. 14 shows how the inferred values of the intrinsic dispersion and the $R_{21}$ population ratio at the Solar circle change when increasing the sample’s minimum SNR from $\text{SNR}_{\text{min}} = 80$ to 300. We keep $\mu_1$ and $\mu_2$ fixed to the values at $\text{SNR}_{\text{min}} = 80$ (see Table 1), but they do not change much if we leave them free. The number of stars in each [Fe/H] bin drops by a factor of $2 \sim 2.5$ as we increase $\text{SNR}_{\text{min}}$ from 80 to 300, but the values of $\sigma_1, \sigma_2$, and $R_{21}$ show only small variations. Dotted lines in the second column show the median observational errors in [Mg/Fe] reported by APOGEE as a function of the SNR threshold. The median error is almost independent of $\text{SNR}_{\text{min}}$. The inferred intrinsic scatter, obtained by subtracting this median error in quadrature from the best-fit total dispersion, is a factor $\sim 3 \sim 4$ larger than the observational error and insensitive to the SNR threshold. In principle, scatter in the observed sequences could arise from systematic errors in the abun-
dance determinations that vary from star to star. Our use of a narrow range of stellar parameters in our sample is designed to limit this effect. As a further check, we have examined histograms of [Mg/Fe] for the cooler and hotter stars of our sample ($T_{\text{eff}} = 4200 \text{ K} \rightarrow 4400 \text{ K}$) and the lower and higher log($g$) stars of our sample ($1.1 \rightarrow 1.75$). We find no clear offsets of [Mg/Fe] between these data subsets, and any shifts of histogram peaks are small compared to the widths of the high-$\alpha$ and low-$\alpha$ distributions.

From the insensitivity to SNR$_{\text{min}}$ and the test on data subsets, we conclude that our model values provide robust estimates of the intrinsic spread of [Mg/Fe] at fixed [Fe/H] for the high-$\alpha$ and low-$\alpha$ populations. These intrinsic dispersions, typically $0.03 - 0.05$ dex (see Table 1), are themselves a significant constraint on GCE models.

The best-fit parameter values in Table 1, and for other $\alpha$-elements in Table 2 (see Section 4.4), constitute the principal results of this paper. Fig. 15 plots our findings for the conditional distributions $p$([Mg/Fe] | [Fe/H]) in ranges of $R$ and $|z|$ for the three metallicity bins centred on [Fe/H] = $-0.3$, $-0.1$, and $+0.1$. These model histograms are corrected for the age-dependent and $|z|$-dependent selection effects in APOGEE, so they show the distributions expected if one could observe a random subset of all long-lived stars (lifetime greater than the age of the Galactic disk) in the indicated $R$ and $|z|$ range. To quantify the uncertainty due to the dependence of the predicted correction factors on $T_{\text{eff}}$ (see Fig. 2), we compare the model distributions obtained for our reference selection with $T_{\text{eff,max}} = 4600 \text{ K}$ (which removes red clump stars) to those obtained with $T_{\text{eff,max}} = 4800 \text{ K}$ (which includes some red clump stars), for which we recompute the correction factors.
Distribution of $\alpha$/Fe in the MW disc

The bottom row of Fig. 15 shows results for the full disk populations integrated over $|z|$, the best representation of the relative importance of the high-$\alpha$ and low-$\alpha$ populations at a given $R$. Our key finding is that the bimodality of the [Mg/Fe] distribution at subsolar [Fe/H] in the solar neighborhood is a genuine, intrinsic feature of the stellar populations, not an artefact of over-representing the high-$\alpha$ stars. In the $|z|$-integrated histogram for $R \approx 7 - 9$ kpc, the minimum of the [Fe/H] $\approx -0.3$ conditional distribution is a factor of three below the surrounding maxima. The sharpness of bimodality depends on $R$, $|z|$, and [Fe/H]. In some ranges, the lower amplitude Gaussian adds a shoulder or asymmetric tail to a distribution with a single maximum. As expected from previous results (e.g., Bensby, Feltzing, & Lundström 2003; Lee et al. 2011; H15), the high-$\alpha$ population is more prominent at larger $|z|$, lower [Fe/H], and smaller $R$. The $R$-dependence is sometimes characterized as a shorter scale length for the “chemical thick disk” vs. the “chemical thin disk” (Bovy et al. 2016). A fully successful model of the Milky Way must reproduce these trends, and the clear bimodality of [Mg/Fe] in some ranges, and the intrinsic dispersions of [Mg/Fe] in the high-$\alpha$ and low-$\alpha$ populations.

4.3 Goodness of fit

From Figs. 8-12 it is clear that our parameterized model provides a good qualitative description of the [Mg/Fe] distributions observed by APOGEE. To give a quantitative measure of goodness-of-fit, analogous to a frequentist $\chi^2$-test, we create mock data sets that satisfy our model assumptions by construction, then fit them using the same procedure applied to the observational data. We generate 100 such data sets for each 0.1-dex bin in the range $0 \leq [\text{Fe/H}] < 0.2$. Each mock catalogue is generated with the procedure outlined in Section 3.3, by assuming for $R_{21}$, $\mu_1$, $\mu_2$, $\sigma_1$, and $\sigma_2$ the values reported in Table 1 as a function of [Fe/H] for the Solar circle. We draw numbers of stars in each range of [Fe/H] that match those in our $R = 7 - 9$ kpc observational sample. For each mock catalogue we record $\ln L_{\text{max}}$ of the best-fit model (eq. 10 with the constant set to zero).

Fig. 16 shows the histogram of $\ln L_{\text{max}}$ values for the 100 mock catalogues in each [Fe/H] bin, as well as the value obtained by fitting the APOGEE data. In all cases the observational value falls well within the mock catalogue distribution, indicating that the model describes the data to within the level expected from Poisson fluctuations. In detail, the observed $\ln L_{\text{max}}$ values for all seven [Fe/H] bins fall within the upper 52% of the mock catalogue distributions, which could be taken as evidence for a slight tension between the model and the data.

4.4 Other $\alpha$-elements

We apply a similar analysis to determine the abundance distributions of the $\alpha$-elements O, Si, S, and Ca, relative to iron, as observed by APOGEE-DR16. In these cases, our analysis makes the following simplifications:
Figure 14. Dependence of inferred model parameters on the threshold SNR of the observational sample, for $R = 7 - 9$ kpc. In each [Fe/H] bin (rows, as labeled) we hold the means $\mu_1$ of the low-$\alpha$ sequence (blue) and $\mu_2$ of the high-$\alpha$ sequence (red) fixed to the values found for $SNR_{\text{min}} = 80$, as shown in the left column. The second column shows the intrinsic dispersions (solid curves) computed by subtracting the median observational errors (dashed curves) from the best-fit total dispersions. The third column shows the fitted ratio $R_{21}$ of high-$\alpha$ stars to low-$\alpha$ stars. The fourth column shows the number of sample stars above the threshold SNR with $|z| \leq 2$ kpc.

(i) In the maximum likelihood fitting, the parameter $R_{21}$ as a function of radius and [Fe/H] is held fixed to the best values that we have determined for the [Mg/Fe] distributions (see Fig. 13 and Table 1).

(ii) In order to correct for the age-selection effects for element X, the values of $\tau_i$ in equation (6) (ages of the stars in the bin $y_i \leq y < y_i + \Delta y$, where $y_i = [X/Fe]$) are estimated with the following two-step procedure. (a) Firstly, we determine the average $[X/Fe] - [Mg/Fe]$ relation, by binning the APOGEE-DR16 $[X/Fe]$ ratios in this plane and making use of a linear interpolation to determine the average $[Mg/Fe]$ from any given $y_i = [X/Fe]$. (b) Once we determine the average $[Mg/Fe]$ corresponding to $y_i$, we use the average $[Mg/Fe]$-age relation from Miglio et al. (2020, see Fig. 4) to estimate $\tau_i$.

The results of our analysis are shown in Fig. 17, in which the different columns for a given row show the $[\alpha/Fe]$-[Fe/H] pattern at different radii. From top to bottom, the various rows show $\alpha$-elements in increasing order of atomic mass, namely $[O/Fe]$, $[Mg/Fe]$, $[Si/Fe]$, $[S/Fe]$, and $[Ca/Fe]$. The filled circles with the error bars in Fig. 17 correspond to the mean values and dispersion of the high-$\alpha$ (in blue) and low-$\alpha$ (in red) sequences, whereas the coloured shaded areas highlight the $[Mg/Fe]$-[Fe/H] relation at the Solar circle ($7 \leq R < 9$ kpc) and are the same in all panels. In Table 2, we report the values of $\mu_1$, $\mu_2$, $\sigma_1$, and $\sigma_2$ as a function of [Fe/H] and $R$ for the various $\alpha$-elements shown in Fig. 17.

The low-$\alpha$ sequence appears similar for all of these elements with the exception of Ca, which is displaced to lower $[Ca/Fe]$. The separation of the high-$\alpha$ and low-$\alpha$ sequences is smaller for $[Si/Fe]$ and $[Ca/Fe]$ than for $[Mg/Fe]$; a similar but smaller offset of the high-$\alpha$ sequence is seen for $[O/Fe]$ and $[S/Fe]$. As discussed by W19, a smaller separation of these sequences would arise if SNIIa or another delayed nucleosynthetic source makes a larger contribution to these elements than to Mg. At the Solar circle, the
Figure 15. The inferred intrinsic distributions of [Mg/Fe] in three ranges of [Fe/H], in ranges of Galactocentric radius $R = 3$-5, 5-7, 7-9, and 9-11 kpc (left to right). Histograms in the bottom row are integrated over all $|z|$, while rows above show 0.5-kpc ranges of $|z|$ as labeled. Red, orange, and green curves show metallicity ranges centred on [Fe/H] = −0.3, −0.1, and +0.1, respectively. The finite vertical width of the histograms shows the impact of changing from our standard sample temperature cutoff $T_{\text{eff,max}} = 4600$ K, which excludes red clump stars, to $T_{\text{eff,max}} = 4800$ K, which includes some red clump stars. In each panel, $N_★$ represents the total number of stars as observed by APOGEE-DR16 in the $\langle R, |z|, [\text{Fe/H}] \rangle$ ranges corresponding to the histogram of the same colour, by assuming $T_{\text{eff,max}} = 4600$ K.

scatter on the low-\(\alpha\) sequence is smaller for O, Si, and Ca than for Mg, while scatter on the high-\(\alpha\) sequence is comparable or slightly smaller. For a given element, the location of the high-\(\alpha\) and low-\(\alpha\) [X/Fe]-[Fe/H] sequences is only weakly dependent on $R$; the changes with $R$ for a given element are much smaller than the apparent differences from one element to another. There is a general tendency for the “knee” in the high-\(\alpha\) sequence to move to lower [Fe/H] at large $R$, as predicted in GCE models where the star formation efficiency decreases as a function of $R$ (e.g., Matteucci & Francois 1989; Boissier & Prantzos 1999; Cescutti et al. 2007; Belfiore et al. 2019; Palla et al. 2020). APOGEE-DR16 incorporates zero-point calibration offsets such that stars in the solar neighborhood with [Fe/H] $\approx 0$ have a mean [X/Fe] $= 0$ (Ahumada et al. 2017; Jónsson et al. 2020). However, APOGEE abundances exhibit systematic trends with log $g$ at the $\sim 0.05$-dex level (Jónsson et al. 2020; Griffith et al. 2020). Element-to-element offsets at this level should therefore be treated with caution.

5 CONCLUSIONS

We have used abundance measurements from APOGEE-DR16 to infer the conditional distributions $p([X/Fe] | [\text{Fe/H}])$ for the $\alpha$-elements Mg, O, Si, S, and Ca in the range $−0.5 \leq [\text{Fe/H}] < +0.2$. Our sample consists of Milky Way disk stars in bins of Galactocentric radius $R = 3$-5, 5-7, 7-9, and 9-11 kpc. By fitting a model and accounting for age-dependent and $|z|$-dependent selection effects, we translate the observed distributions into the intrinsic distributions that would be found by observing a random sample of all long-lived disk stars. Our model assumes a double Gaussian $p(\alpha/Fe)$ in each 0.1-dex [Fe/H] bin, one describing the low-\(\alpha\) population with scale height $z_\alpha = 0.45$ kpc and one describing the high-\(\alpha\) population with scale-height $z_\alpha = 0.95$ kpc. The means, dispersions, and relative amplitude of the two Gaussians are determined by maximum likelihood fitting. Tests on mock catalogues demonstrate that the method yields reliable parameter determinations and that the agree-
Solar circle. We set the constant of equation (10) to zero when evaluating \( L \) that yield \( f \) observed star counts. We report the fraction of mock data, \( f \geq L \text{APOGEE} \), that yield \( L_{\text{max}} \) higher than our best-fitting model to APOGEE-DR16 at the Solar circle. We set the constant of equation (10) to zero when evaluating \( L \).

Our principal results consist of the model parameter determinations reported in Tables 1 and 2 and illustrated in Figs. 8-12 and 16. The means, dispersions, and relative amplitudes of these sequences are quantitative measures that a complete model of Milky Way disk chemical evolution should reproduce.

As context for interpreting these results, we show \( \alpha/Fe \)– [Fe/H] tracks and \( \alpha/Fe \) distribution functions of one-zone chemical evolution models in Fig. 18, computed using the analytic methods of WAF. An individual one-zone model with constant parameters cannot explain the observed \( \alpha/Fe \)– [Fe/H] distribution of disk stars because it predicts that [Fe/H] and \( \alpha/Fe \) evolve monotonically; the \( \alpha/Fe \) distributions in Fig. 18 are computed over all [Fe/H] values, not narrow bins like those of in our APOGEE analysis. However, a mixture of populations evolved along distinct evolutionary tracks at different Galactocentric radii will exhibit a distribution of \( \alpha/Fe \) at fixed [Fe/H] (e.g., Schönherr & Binney 2009; Minchev, Chiappini, & Martig 2013; Nidever et al. 2014). A one-zone model in which outflow efficiency grows at late times can produce a double-valued \( \alpha/Fe \)– [Fe/H] track in which the low-\( \alpha \) sequence evolves from high [Fe/H] to low [Fe/H] (WAF), but we do not explore this possibility here.

Black curves in each panel show results for a fiducial model with star formation efficiency timescale \( \tau_s = 2 \) Gyr, outflow mass loading \( \eta = 2 \), and an exponentially declining SFH with e-folding timescale \( t_{\text{sfh}} = 6 \) Gyr (see WAF for modeling details). The SNIa delay time distribution is a sum of exponentials designed to mimic a \( t^{-1.1} \) power law with a minimum delay time \( t_{\text{d,min}} = 0.15 \) Gyr. The \( \alpha/Fe \) distribution has a narrow peak at the plateau value set by the CCSN yield ratio, then a broad peak centred near \( \alpha/Fe \approx 0 \). Increasing the star-formation timescale to 8 Gyr (upper panels, red curves) has little effect on the high-\( \alpha \) peak, but the low-\( \alpha \) peak becomes narrower. For a constant SFR (not shown) the model evolves quickly to equilibrium abundances, and the \( \alpha/Fe \) distribution is sharply peaked at the Solar ratio. Blue curves in the upper panels show a linear-exponential SFR \( \propto t^{-1}/t_{\text{d},\text{min}} \), which evolves more slowly than the corresponding exponential model. With this SFH, the high-\( \alpha \) peak disappears entirely because fewer stars are formed at early times.

The lower panels of Fig. 18 show the impact of changing other model parameters with the SFH (6 Gyr exponential) held fixed. Decreasing the star formation efficiency, \( \eta = 1 \), evolves to higher [Fe/H] but has nearly the same \( \alpha/Fe \) distribution. Reducing \( t_{\text{d,min}} \) to 0.05 Gyr suppresses the high-\( \alpha \) peak (and shifts the knee to lower [Fe/H]) because of the more rapid onset of SNIa enrichment, and the low-\( \alpha \) peak becomes broad and flat.

While we reserve tests of theoretical models to future work, the distribution of the log-likelihoods, \( \ln L_{\text{max}} \) of the best-fitting models to our MW mock catalogues for the Solar circle (blue histograms) as a function of [Fe/H] (rows). In each panel, the vertical black line shows the log-likelihood of our best-fitting model to APOGEE-DR16 at the Solar circle, in the given range of [Fe/H]. The model fits the observed APOGEE data as well as it fits mock data sets constructed with the model assumptions, indicating that disagreements are compatible with Poisson fluctuations in the observed star counts. We report the fraction of mock data, \( f (< L_{\text{APOGEE}}) \), that yield \( L_{\text{max}} \) higher than our best-fitting model to APOGEE-DR16 at the Solar circle. We set the constant of equation (10) to zero when evaluating \( \ln L_{\text{max}} \).
we note that the clear minima in the $\alpha$/Fe distributions found in many zones at sub-solar [Fe/H] pose a challenge to models that explain the $\alpha$/Fe – [Fe/H] distribution purely in terms of radial mixing of populations that follow separate evolutionary tracks in this space. Because one-zone models with smooth star formation histories generically produce $\alpha$/Fe distributions that are strongly peaked at low values, it is difficult to construct a superposition of such distributions with two well separated maxima of comparable strength. The Sharma, Hayden, & Bland-Hawthorn (2020) model may appear to represent a counter-example, but Sharma, Hayden, & Bland-Hawthorn (2020) do not demonstrate that the chemical evolution tracks assumed in this model can be produced self-consistently by an underlying star formation and enrichment history. Radial mixing is almost certainly an ingredient in Galactic chemical evolution, but we suspect that explaining the bimodality of $\alpha$/Fe will also require some form of “discontinuous” evolution, whether that be the 3-phase star formation history envisioned in the 2-infall scenario (Chiappini, Matteucci, & Gratton 1997; Spitoni et al. 2019; Palla et al. 2020), the clumpy bursts of star formation proposed by Clarke et al. (2019), the sharp change of outflow efficiency suggested by WAF, the unusual gas accretion history found in a small fraction of simulated galaxies (Grand et al. 2018; Mackereth et al. 2019), or an early merger event that resets the metallicity of the ISM (Buck 2020; Vincenzo & Kobayashi 2020).

In agreement with the findings for [O/Fe] by Bertran de Lis et al. (2016), we infer intrinsic scatter of $\alpha$/Fe at the $\sim 0.04$-dex level on both the high-$\alpha$ and low-$\alpha$ sequences, and comparable values for other elements. The implied fractional scatter of $\alpha$/Fe ratios is thus $\sim 10\%$. Predicting scatter of abundance ratios requires a detailed model for the mixing of supernova products in ISM gas and subsequent star formation (e.g., Krumholz & Ting 2018), and we do not know of quantitative predictions of $\alpha$/Fe scatter in the metal-
Figure 18. Evolutionary tracks (left) and $[\alpha/\text{Fe}]$ distributions (right) of one-zone chemical evolution models, computed using the analytic formalism of WAF. Points in the left panels show $[\alpha/\text{Fe}]$ and $[\text{Fe/H}]$ at 1 Gyr intervals (up to 12 Gyr). Black solid curves show a reference model with outflow efficiency $\eta = 2$, star formation efficiency $\tau_1 = 1$ Gyr, SNIa minimum delay time $t_{d,min} = 0.15$ Gyr, and an exponentially decaying star formation history with a timescale of 6 Gyr. In the upper panels, red dotted and blue dashed curves show models with an 8 Gyr exponential or 6 Gyr linear-exponential star formation history, respectively. In the lower panels, red dotted, blue dashed, and green dot-dashed curves show the effect of changing the star formation efficiency to $\tau_1 = 1$ Gyr, the outflow efficiency to $\eta = 1$, or the minimum delay time to $t_{d,min} = 0.05$ Gyr.

licity range examined here. Stochastic sampling of the stellar IMF is probably not sufficient: in round numbers, enriching a gas mass $M_g$ to $0.5 Z_\odot$ requires $N_{SN} \sim 10^4 (M_g/5 \times 10^6 M_\odot)^{-1/2}$ core-collapse supernovae, so order unity fluctuations in (X/Fe) from one supernova to another would average down to $\sim 0.01 (M_g/5 \times 10^6 M_\odot)^{-1/2}$ rms fluctuations in the enriched stellar population. On the low-$\alpha$ sequence, the scatter could plausibly arise from a several-Gyr range of stellar ages that leads to star-by-star variations in the ratio of SNIa-to-CCSN enrichment (see Fig. 18). This explanation on its own may be insufficient for the high-$\alpha$ sequence because populations evolve away from the plateau so quickly. However, superposition of populations that have different knees in their $[\alpha/\text{Fe}] - [\text{Fe/H}]$ evolution will produce scatter at moderately sub-solar $[\text{Fe/H}]$ from the mix of stars that lie varying distances below the plateau (see Fig. 18, lower left). Larger scatter could arise from superposing populations that have different $[\alpha/\text{Fe}]$ plateau values, which would require systematic differences in the supernovae associated with these populations. Empirical clues to the origin of scatter can be derived by examining element-by-element correlations of star-by-star deviations from mean trends, and correlations of these deviations with stellar age and kinematics. We will undertake such studies in future work.

Spectroscopic surveys like APOGEE, GALAH, RAVE, and the Large Sky Area Multi-Object Fibre Spectroscopic Telescope (LAMOST; e.g. Xiang et al. 2019; Wheeler et al. 2020) have extended the study of element ratio distributions from the solar neighborhood or other selected regions to comprehensive maps across much of the Galaxy, with the precision and well characterized selection needed to resolve such questions as bimodality and intrinsic scatter (H15; W19; Griffith et al. 2020; Nandakumar et al. 2020. These empirical distributions encode information about many aspects of Galactic evolution and enrichment physics, including accretion and star formation history, mergers and perturbations, gas mixing and gas flows, radial migration of stars, and supernova nucleosynthesis. They present powerful tests for increasingly sophisticated models of galactic chemical evolution and for hydrodynamic simulations of galaxy formation.

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#### Table 1. Parameters of our best fitting models for $p([\text{Mg}/\text{Fe}])$. Best-fit values of the total dispersions have been converted to \textit{intrinsic} values $\sigma_1$, $\sigma_2$ by subtracting in quadrature the median [Mg/Fe] abundance error as reported in APOGEE-DR16.

\[ \begin{array}{cccccc}
\text{[Fe/H]}_{\text{min}} & -0.5 & -0.4 & -0.3 & -0.2 & 0.0 \\
\mu_1 & - & - & 0.149 & 0.089 & 0.061 \\
\sigma_1 & - & - & 0.052 & 0.067 & 0.053 \\
\mu_2 & 0.291 & 0.274 & 0.265 & 0.228 & 0.191 \\
\sigma_2 & 0.046 & 0.045 & 0.036 & 0.049 & 0.041 \\
R_{21} & 18.357 & 17.911 & 2.574 & 1.912 & 1.030 \\
\end{array} \]

\[ \begin{array}{cccccc}
\text{[Fe/H]}_{\text{min}} & -0.5 & -0.4 & -0.3 & -0.2 & 0.0 \\
\mu_1 & 0.100 & 0.106 & 0.125 & 0.075 & 0.043 \\
\sigma_1 & 0.027 & 0.048 & 0.063 & 0.049 & 0.045 \\
\mu_2 & 0.281 & 0.275 & 0.265 & 0.234 & 0.174 \\
\sigma_2 & 0.049 & 0.042 & 0.035 & 0.043 & 0.045 \\
R_{21} & 7.358 & 3.592 & 1.585 & 1.157 & 0.833 \\
\end{array} \]

\[ \begin{array}{cccccc}
\text{[Fe/H]}_{\text{min}} & -0.5 & -0.4 & -0.3 & -0.2 & 0.0 \\
\mu_1 & 0.111 & 0.095 & 0.086 & 0.057 & 0.039 \\
\sigma_1 & 0.038 & 0.049 & 0.052 & 0.046 & 0.036 \\
\mu_2 & 0.282 & 0.273 & 0.254 & 0.219 & 0.171 \\
\sigma_2 & 0.042 & 0.036 & 0.034 & 0.042 & 0.049 \\
R_{21} & 0.936 & 1.074 & 0.565 & 0.393 & 0.338 \\
\end{array} \]

\[ \begin{array}{cccccc}
\text{[Fe/H]}_{\text{min}} & -0.5 & -0.4 & -0.3 & -0.2 & 0.0 \\
\mu_1 & 0.092 & 0.077 & 0.060 & 0.034 & 0.027 \\
\sigma_1 & 0.033 & 0.036 & 0.035 & 0.031 & 0.033 \\
\mu_2 & 0.277 & 0.265 & 0.238 & 0.150 & 0.145 \\
\sigma_2 & 0.050 & 0.045 & 0.047 & 0.079 & 0.060 \\
R_{21} & 0.179 & 0.166 & 0.153 & 0.232 & 0.193 \\
\end{array} \]
| [Fe/H]_{min} | [O/Fe] at $7 \leq R < 9$ kpc |
|-------------|-----------------------------|
| $\mu_1$ | 0.114 | 0.095 | 0.080 | 0.057 | 0.040 | 0.030 | 0.028 |
| $\sigma_1$ | 0.029 | 0.030 | 0.028 | 0.028 | 0.024 | 0.023 | 0.021 |
| $\mu_2$ | 0.239 | 0.227 | 0.208 | 0.178 | 0.137 | 0.113 | 0.086 |
| $\sigma_2$ | 0.029 | 0.024 | 0.023 | 0.031 | 0.038 | 0.030 | 0.025 |

| [Fe/H]_{min} | [Mg/Fe] at $7 \leq R < 9$ kpc |
|-------------|-----------------------------|
| $\mu_1$ | 0.111 | 0.095 | 0.086 | 0.057 | 0.039 | 0.031 | 0.038 |
| $\sigma_1$ | 0.038 | 0.049 | 0.052 | 0.046 | 0.036 | 0.032 | 0.031 |
| $\mu_2$ | 0.282 | 0.273 | 0.254 | 0.219 | 0.171 | 0.146 | 0.124 |
| $\sigma_2$ | 0.042 | 0.036 | 0.034 | 0.042 | 0.049 | 0.037 | 0.032 |

| [Fe/H]_{min} | [Si/Fe] at $7 \leq R < 9$ kpc |
|-------------|-----------------------------|
| $\mu_1$ | 0.090 | 0.075 | 0.064 | 0.041 | 0.024 | 0.015 | 0.014 |
| $\sigma_1$ | 0.029 | 0.031 | 0.027 | 0.021 | 0.020 | 0.018 | 0.019 |
| $\mu_2$ | 0.201 | 0.188 | 0.170 | 0.132 | 0.097 | 0.075 | 0.050 |
| $\sigma_2$ | 0.027 | 0.028 | 0.026 | 0.033 | 0.035 | 0.025 | 0.025 |

| [Fe/H]_{min} | [S/Fe] at $7 \leq R < 9$ kpc |
|-------------|-----------------------------|
| $\mu_1$ | 0.164 | 0.133 | 0.103 | 0.067 | 0.038 | 0.016 | 0.004 |
| $\sigma_1$ | 0.052 | 0.057 | 0.045 | 0.037 | 0.028 | 0.024 | 0.022 |
| $\mu_2$ | 0.287 | 0.254 | 0.219 | 0.180 | 0.130 | 0.099 | 0.066 |
| $\sigma_2$ | 0.085 | 0.073 | 0.059 | 0.043 | 0.052 | 0.062 | 0.040 |

| [Fe/H]_{min} | [Ca/Fe] at $7 \leq R < 9$ kpc |
|-------------|-----------------------------|
| $\mu_1$ | 0.043 | 0.033 | 0.028 | 0.016 | 0.002 | -0.006 | -0.010 |
| $\sigma_1$ | 0.021 | 0.025 | 0.023 | 0.024 | 0.022 | 0.019 | 0.023 |
| $\mu_2$ | 0.149 | 0.130 | 0.112 | 0.082 | 0.049 | 0.014 | - |
| $\sigma_2$ | 0.031 | 0.031 | 0.029 | 0.030 | 0.038 | 0.046 | - |

Table 2. Parameters of our best fitting models for $p([O/Fe])$, $p([Si/Fe])$, $p([S/Fe])$, and $p([Ca/Fe])$ at the Solar circle. Best-fit values of the total dispersions have been converted to intrinsic values $\sigma_1$, $\sigma_2$ by subtracting in quadrature the median [Fe/Fe] abundance error as reported in APOGEE-DR16.
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