

Dwarf galaxies don't just form stars, they have **BURSTS** of star formation. We're looking at the chemical evidence left behind.

QUANTIFYING BURSTY STAR FORMATION IN DWARF GALAXIES

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SUMMARY

Dwarf galaxies are weird. They have episodes where they form lots of stars, and then quiet periods where they don't. We're using the chemical composition of their stars to figure out the timing of these "bursts" and "quiet" phases.

TL;DR: Star formation in dwarfs is like a flickering candle, not a steady flame, and we can read the history in the stars' chemistry.

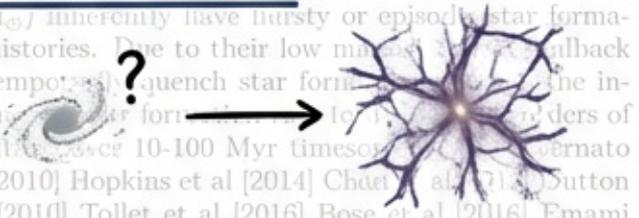
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ABSTRACT

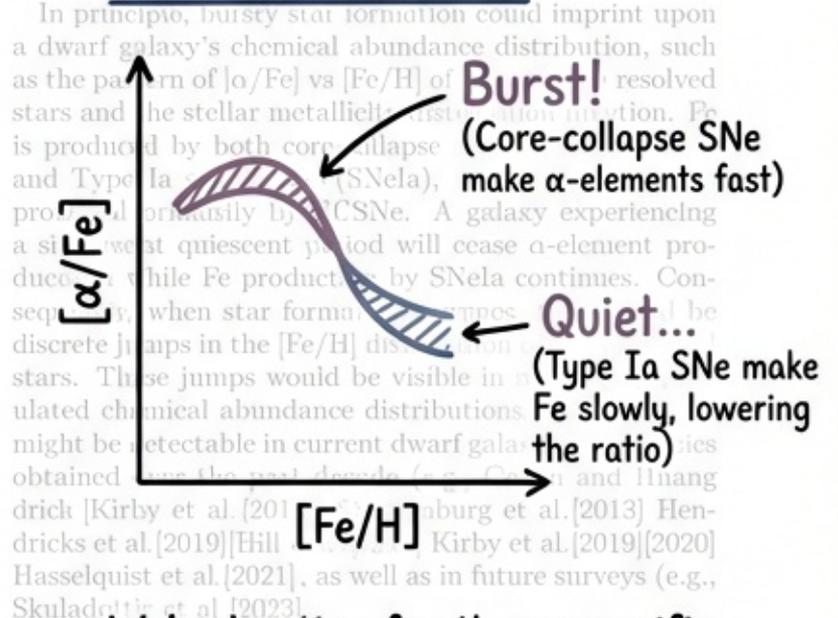
1. INTRODUCTION

WHY BOTHER?

Because understanding how dwarf galaxies form stars is key to understanding the whole universe, especially its early history (like reionization). They're the building blocks!



ies may over-ionize the IGM. with hursty star formation proposed as a possible solution (Sun et al [2023] Clarke et al [2024] Mulioz et al [2024] Simmonds et al [2024]. Burstiness has also been linked to the miniquenching of high redshift galaxies (Antoni-Danso [2023] Dome... **The Chemical Clue!**



We're hunting for these specific chemical patterns!

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So, how do we find these chemical clues?

TRIC We build a **SIMPLE MODEL!**

delay time of 100 Myr (Kirby et al. 2011a)

A Recipe for a "Toy" Dwarf Galaxy

$$\begin{cases} 0 & \text{for } t < 100 \text{ Myr} \\ \dots & \text{for } t \geq 100 \text{ Myr} \end{cases} \quad (2)$$

where 0.420 is a normalization constant. We assume the core-collapse supernova IMF-averaged yield for Fe and Mg per supernova to be $y_{\text{CC,Fe}} = 0.081$ and $y_{\text{CC,Mg}} = 0.192$ with the rate of $r_{\text{CC}} = 0.01 \text{ M}_{\odot}^{-1}$ core-collapse supernovae per solar mass formed, and the Type Ia supernovae yield of Fe to be $y_{\text{Ia,Fe}} = 0.749$ with negligible Mg production, and a supernova rate of $r_{\text{Ia}} = 0.150 \text{ M}_{\odot}^{-1}$. The yields (in M_{\odot}) and rates are adopted from Kirby et al. (2011a).

We then make three simplifying assumptions. First, we assume the yields have no metallicity dependence. This is a reasonable assumption for Mg and Fe in metal-poor dwarf galaxies, where the metallicity dependence of IMF-integrated yields is relatively weak (e.g., Nomoto et al. 2013). Then, the total yield of heavy elements is simply the convolution of the star formation rate, $\dot{M}_{\star}(t)$, and by the yield convolved with the delay time distribution:

$$Y_X(t) = \dot{M}_{\star}(t) * (y_{\text{CC},X} r_{\text{CC}} D_{\text{CC}}(t) + y_{\text{Ia},X} r_{\text{Ia}} D_{\text{Ia}}(t)) \quad (3)$$

where $Y_X(t)$ is the total yield of element X at time t . The total mass of an element X that has ever been produced by a galaxy at time t can be expressed as

$$M_X(t) = \int_0^t Y_X(t') dt' \quad (4)$$

Second, we assume that we can independently specify the star formation rate $\dot{M}_{\star}(t)$ and an effective gas mass $M_{\text{G,eff}}(t)$, which includes not just the cold interstellar medium, but also the circumgalactic or intergalactic gas mass that is affected by the stellar yields from this galaxy. The abundance of element X relative to hydrogen at time t can then be calculated as:

$$[X/H](t) = \log_{10} \left(\frac{M_X(t)}{\mu_X M_{\text{G,eff}}(t)} \right) - \log_{10} \left(\frac{N_X}{N_H} \right)_{\odot} \quad (5)$$

where μ_X is the atomic mass of element X , $\log_{10}(N_X/N_H)_{\odot}$ is the solar abundance ratio of element X to hydrogen (adopted from Asplund et al [2009]). We assume $\mu_X = 24$ for Mg and $\mu_X = 56$ for Fe. We note this assumption is mathematically equivalent to standard

A HUGE SIMPLIFICATION:
We assume the total amount of gas available is **CONSTANT**. It's like a "closed box" that moves with the galaxy.

Third, and the biggest assumption here, we take the mixing gas mass to be a constant, i.e. $\dot{M}_{\text{G,eff}}(t) = 0$. This

smooth prescriptions for gas and star formation histories (SFH) (e.g., Cescutti 2005] Kirby et al. 2011a] Pilkington and Gibson 2012] Vincenzo et al. 2019] Recchi and Kroupa 2013] [shimaru et al. 2017] Coté et al [2017] Weinberg et al. 2017] Matroucci 2021] de los Reyes et al. 2023] Johnson et al. 2023] Sandford et al. 2023), with some notable exceptions (e.g., Hirashito et al. 2021] Johnson and Weinberg 2023). Some semi-analytic

THE INGREDIENTS:

- **Bursty Star Formation:**

Stars form in quick episodes.



- **Two Element "Factories":**

(1) **Fast Factory (CCSNe):**

Makes α -elements (like Mg) **INSTANTLY**.



(2) **Slow Factory (SNIa):**

Makes Iron (Fe) with a **DELAY**.



The math for the iron *delay* time.

2. SIMPLE CHEMICAL EVOLUTION MODEL

To understand how bursty star formation can lead to a distinctive discontinuous chemical track, we draw intuition from a simplified setup for the chemical evolution of a dwarf galaxy, which allows us to integrate the chemical track analytically by varying only the star formation history and the underlying mixing gas mass. We will trace the α -elements with the Mg abundance, as it is relatively easy to measure in stars and uncontaminated by SNIa yields (e.g., Kirby et al [2019] Skiladottir et al [2019, Weinberg et al (2019)). We adopt the standard one-zone, instantaneous mixing approximation, which is likely applicable for dwarf galaxies due to their small sizes (e.g., Kirby et al [2011a, Vincenzo et al [2019, Johnson et al. 2023).

We assume a prompt source of stellar yield from core-collapse supernovae (CCSNe), which produces both Mg and Fe 10 Myr after a starburst:

$$D_{\text{CC}}(t) = \delta \left(\left(\frac{t}{\text{Myr}} \right) - 10 \right), \quad (6)$$

and a delayed source for Type Ia supernovae (SNIa), which contribute only to Fe. We assume a power law delay time distribution with index -1.1 and a minimum



this galaxy, which co-

Let's see what our "Toy" galaxy predicts!

QUANTIFYING BURSTY STAR FORMATION IN DWARF GALAXIES

is motivated by simplicity, allowing us to isolate the impact of bursty star formation on chemical evolution. This simple assumption is sufficient to match observations of Sculptor (Section 4). It may also be a reasonable assumption: while dwarf galaxies are known to lose 90-99% of their metals in outflows (e.g., Kirby et al. 2011b), we can instead use a 10–100× larger $M_{g,eff}$ than would normally be in part of a galaxy's interstellar medium. We will revisit the implications of this assumption in Section 5 and future work.

Our choice of a semi-analytical framework, while simple, provides tractability that allows us to explore the impact of bursty star formation on chemical evolution. Our approach complements the natural time resolution of individual bursts.

→ 10 bursts... over a total duration of 2 Gyr

→ Our setup: A galaxy with a "hiccup" every ~200 Myr

→ $M_{g,eff} = 3 \times 10^8 M_{sun}$

→ Our constant gas supply (from pg 2)

3. EXPECTATIONS FOR BURSTY STAR FORMATION

We now use our simple model to investigate the impact of episodic, bursty star formation on dwarf galaxy chemical evolution. We consider the impact of a bursty star formation model with starbursts occurring over a total duration of 2 Gyr. The solid blue line in the top left panel shows the chemical evolution track for our fiducial model, which is a model with starbursts occurring over a total duration of 2 Gyr. The solid blue line in the top left panel shows the chemical evolution track for our fiducial model, which is a model with starbursts occurring over a total duration of 2 Gyr. The solid blue line in the top left panel shows the chemical evolution track for our fiducial model, which is a model with starbursts occurring over a total duration of 2 Gyr.

BURST!
(Stars form!)

CCSNe (Fast Factory):
 α -elements

SNIa (Slow Factory):
Iron (Fe)

Iron keeps building up during QUIET times...

To better understand the mechanisms driving the chemical evolution, we show the evolution of the bottom panel of Figure 1. The evolution of the bottom panel of Figure 1 shows the evolution of the bottom panel of Figure 1. The evolution of the bottom panel of Figure 1 shows the evolution of the bottom panel of Figure 1.

Fe/H continues to accumulate from Type Ia supernovae during quiescent periods

This creates "gaps" in the chemical plot!

→ We'll compare our model to a real galaxy: **SCULPTOR dSph (our benchmark)**

The top panels of Figure 1 lead to a discontinuous chemical evolution track, manifesting as distinct clumps in the [Mg/Fe]-[Fe/H] plane. This contrasts with the case of uniform star formation, depicted by the orange dashed line, which results in a smooth and continuous chemical evolution track.

Our fiducial chemical evolution model is shown by a solid blue line. A more **REALISTIC** model: Bursts aren't instant, they last ~150 Myr

First, in our model, starbursts are not instantaneous. They last for 150 Myr. This choice is motivated by observations (e.g., Wheeler et al. 2019, Hirai et al. 2017, Wheeler et al. 2019, Hirai et al. 2017, Wheeler et al. 2019, Hirai et al. 2017) that have shown that multiple starbursts often occur within a "block" of star formation activity, followed by a quiescent period lasting a few hundred million years. Such quiescent periods may also be observed in higher mass dwarf galaxies by VST (e.g., [Looser et al. 2024]).

Second, we allow the timing and normalization of the star formation rate to vary in order to match the star formation history of Sculptor. In the integrated star formation mass from $10^9 M_{\odot}$ at 1 Gyr, followed by a decrease for the final burst. This results in a total stellar mass of $10^7 M_{\odot}$ (see top right panel of Figure 1, which matches the shape of the metallicity distribution function (MDF) and star formation history observed in the Sculptor dSph (Kirby et al. 2011a, Weise et al. 2024)).

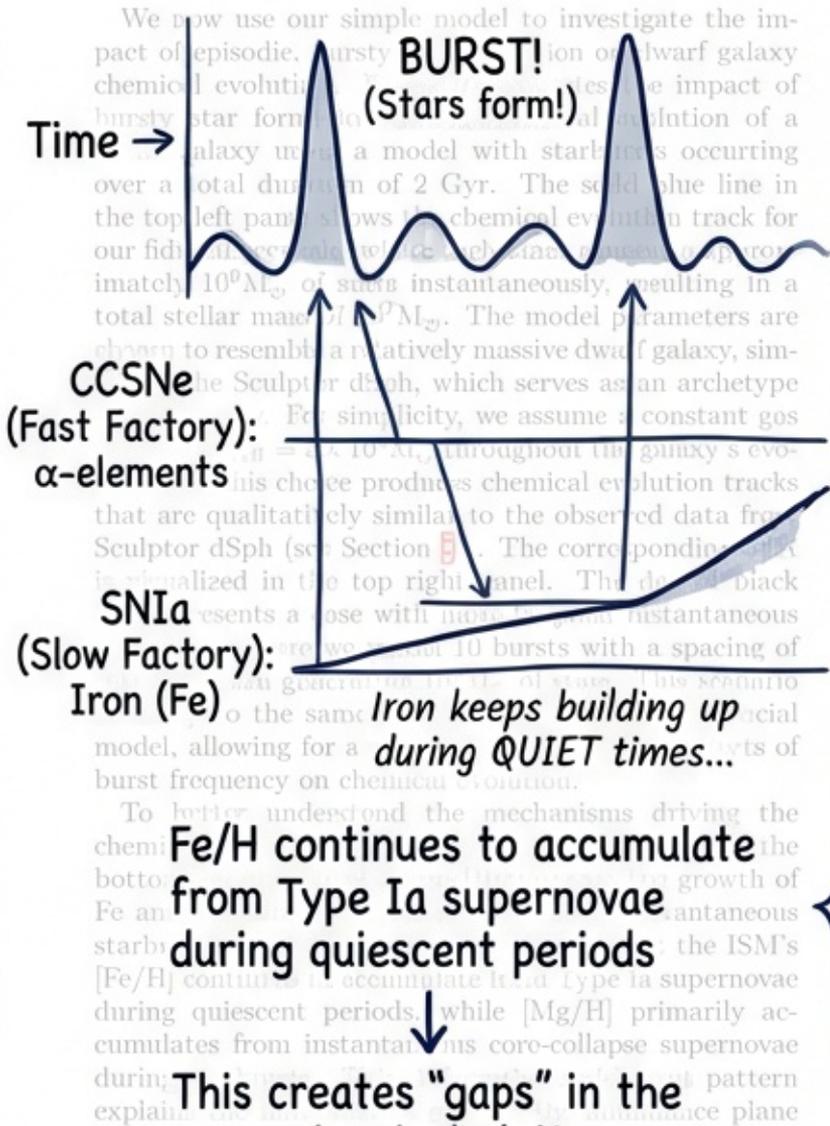
It is important to note that the temporal "resolution" of the starburst in our model is limited by the minimum delay time for Type Ia supernovae, which is 100 Myr in this study. Consequently, our model cannot resolve bursts separated by less than this timescale. For the purpose of this analysis, we define two distinct starbursts as events separated by more than 150 Myr, with star formation proceeding uniformly throughout each 150 Myr burst. In our fiducial model, we assume a total of 7 starburst blocks, with the temporal gap between each block increasing from 150 Myr to 450 Myr in increments of 100 Myr.

→ **temporal "resolution" ... is limited by the minimum delay time for Type Ia supernovae, which is 100 Myr**

LIMITATION: Our chemical "clock" can't resolve bursts faster than ~100 Myr (the SNIa delay)

→ **APOGEE**

→ We'll compare to high-quality data from this survey.



Here are the results from our "Toy" models!

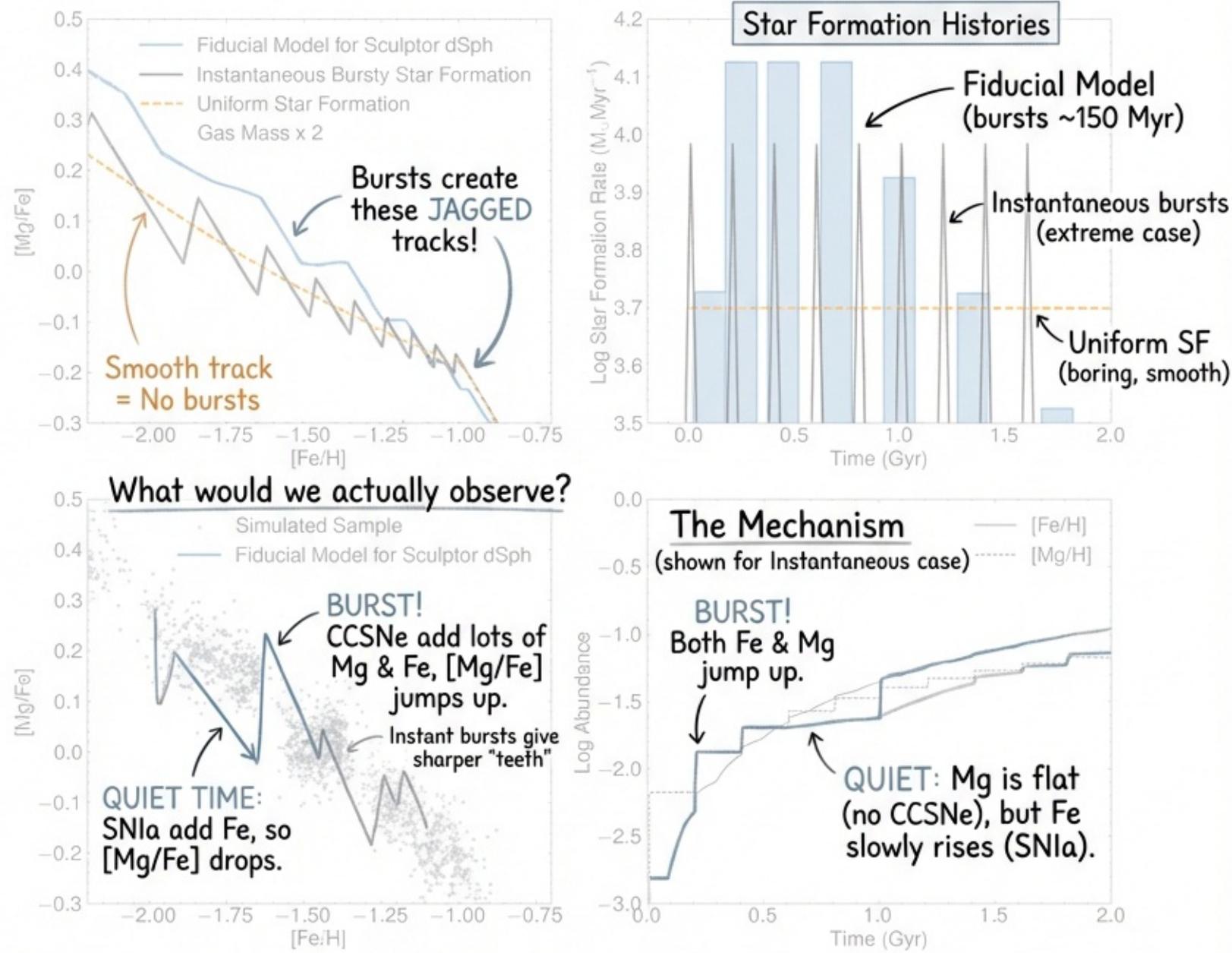


Figure 1. Bursty star formation creates discontinuous chemical tracks in the $[Mg/Fe]$ - $[Fe/H]$ plane. The top left panel shows chemical tracks resulting from various star formation histories: our fiducial model to match the Sculptor dSph observation (solid blue line), star formation with instantaneous bursts every 200 Myr over 2 Gyr (solid black line), continuous star formation (dashed orange line), and a scenario with doubled gas mass (dash-dotted red line). The top right panel displays the corresponding star formation histories. The bottom right panel illustrates the mechanism for the discontinuous starburst case. It demonstrates that the ISM's $[Fe/H]$ continues to accumulate from instantaneous core-collapse supernovae. This combination creates a clumpy pattern in the $[Mg/Fe]$ - $[Fe/H]$ plane, where $[Fe/H]$ reflects star formation gaps and $[Mg/H]$ largely follows the fiducial model's history, with an assumed measurement error of 0.05 dex, along with their kernel density estimation contours. The distribution of simulated data points (grey) contrast sharply with the continuous star formation case. All modeled star formation rates integrate to a total stellar mass of $3 \times 10^8 M_{\odot}$.

Stars form in distinct CLUMPS during bursts...
 ...leaving GAPS where no stars formed.

burst case, the chemical evolution track features, where the slope of the upward spike in $[Mg/Fe]$ is solely determined by the burst. In contrast, our fiducial model with star formation spanning 150 Myr per starburst allows for continuous enrichment of the ISM by Type Ia supernovae during the burst, resulting in more subdued and concave stripes instead of discrete spikes. However, it is important to note that while the instantaneous starburst model might appear to lead to a marginally larger spread in the chemical track, the track

represents the chemistry of the gas, not necessarily the observed stars. In the instantaneous case, all stars are created at the base of each spike, rather than being gradually formed as the burst continues over a more prolonged period. As a result, the fiducial model shown actually has a larger spread in $[Mg/Fe]$ for a given clump compared to the instantaneous case. Finally, when the gas mass is doubled, the gaps between the clumps in the $[Mg/Fe]$ - $[Fe/H]$ plane become less pronounced. This is because, for a fixed iron production rate, the magnitude

This clumpy pattern matches what we see in Sculptor!

How to Find the "Clumps" (Bursts) in the Data

of the [Fe/H] gap is modulated by the underlying gas mass.

3.1. Quantifying the gap in [Fe/H] between starbursts

To better understand the impact of bursty star formation on the chemical evolution of dwarf galaxies, we analytically estimate the difference in [Fe/H] between two consecutive starbursts. This difference manifests as the spacing between the peaks of a multimodal MDF in the case of bursty star formation. The spacing between the distinct star formation episodes is determined by the degree to which the metallicity evolves during the quiescent periods separating the starbursts. This evolution is primarily driven by the accumulated reservoir of Type Ia supernovae from previous star formation episodes, which continuously enrich the interstellar medium with iron-peak elements (bottom right panel of Fig 1).

Consider two starbursts separated by a time gap δt_{gap} . The corresponding gap in [Fe/H] between the first and second starburst, to first approximation, is contributed by the supernovae from the first starburst and can be expressed as

$$\Delta[\text{Fe}/\text{H}] = \log_{10} \frac{M_{\text{Fe}}(t) + \Delta M_{\text{Fe}}}{M_{\text{Fe}}(t)} \approx \frac{\Delta M_{\text{Fe}}}{M_{\text{Fe}}(t) \ln(10)} \quad (6)$$

where ΔM_{Fe} is the additional iron mass generated between times t and $t + \delta t_{\text{gap}}$. The approximation is a Taylor expansion assuming that the change in total Fe mass during the gap is small compared to the existing Fe mass.

If we assume a starburst generating a total mass M_* , and a power-law delay time distribution given by Equation 2 the gap in [Fe/H] between two starbursts contributed by type Ia supernovae can be approximated as:

$$\Delta[\text{Fe}/\text{H}]_{\text{Ia}} \approx \frac{y_{\text{Ia}, \text{Fe}} f_{\text{Ia}} M_*}{\mu_{\text{Fe}} M_{\text{G,eff}} 10^{[\text{Fe}/\text{H}] - 4.5} \ln(10)} \times \int_{t_{\text{min}}}^{\delta t_{\text{gap}}} 0.429 \left(\frac{t}{\text{Myr}} \right)^{-1.1} dt \quad (7)$$

where $\log_{10}(N_{\text{Fe}}/N_{\text{H}})_{\odot} = -4.5$ so we can express $M_{\text{Fe}}(t)$ in terms of [Fe/H]:

$$M_{\text{Fe}}(t) = \mu_{\text{Fe}} M_{\text{G,eff}} 10^{([\text{Fe}/\text{H}] - 4.5)} \quad (8)$$

Evaluating the integral yields:

$$\Delta[\text{Fe}/\text{H}]_{\text{Ia}} \approx \frac{4.29 y_{\text{Ia}, \text{Fe}} f_{\text{Ia}} M_*}{\mu_{\text{Fe}} M_{\text{G,eff}} 10^{([\text{Fe}/\text{H}] - 4.5) \ln(10)} \left[\left(\frac{t_{\text{min}}}{\text{Myr}} \right)^{-0.1} - \left(\frac{\delta t_{\text{gap}}}{\text{Myr}} \right)^{-0.1} \right]} \quad (9)$$

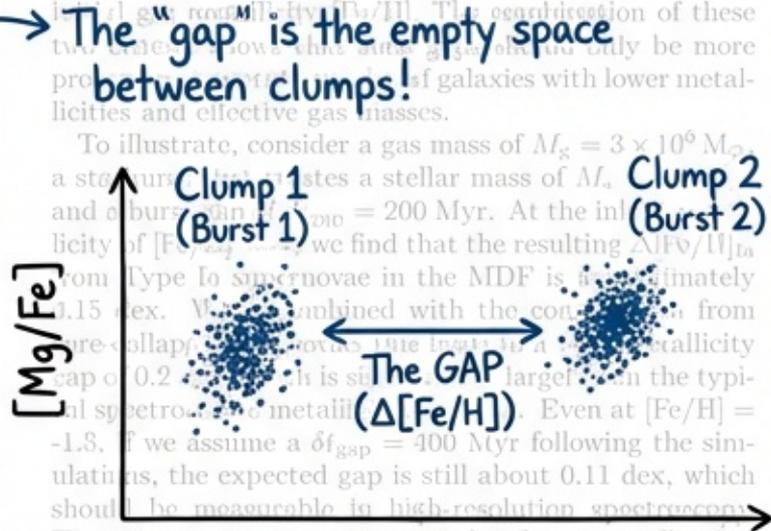
Furthermore, the Fe contribution from CCSNe is

$$\Delta[\text{Fe}/\text{H}]_{\text{CC}} \approx \frac{y_{\text{Ia}, \text{CC}} f_{\text{CC}} M_*}{\mu_{\text{Fe}} M_{\text{G,eff}} 10^{([\text{Fe}/\text{H}] - 4.5) \ln(10)} \quad (10)$$

As the equation shows, the longer the time gap between starbursts, the larger the visible metallicity gap would be, as expected. The gap $\Delta[\text{Fe}/\text{H}]$ is sensitive to

the underlying effective mixing gas mass $M_{\text{G,eff}}$ and the metallicity of the gas. The concentration of these two parameters in dwarf galaxies may be more pronounced in galaxies with lower metallicities and effective gas masses.

To illustrate, consider a gas mass of $M_{\text{G}} = 3 \times 10^9 M_{\odot}$, a starburst that creates a stellar mass of $M_* = 10^8 M_{\odot}$ and a bursty star formation rate of $\dot{M}_* = 200 \text{ Myr}^{-1}$. At the initial metallicity of $[\text{Fe}/\text{H}]_{\text{Ia}} = -1.3$, we find that the resulting $\Delta[\text{Fe}/\text{H}]_{\text{Ia}}$ from Type Ia supernovae in the MDF is approximately 0.15 dex. When combined with the contribution from pre-collapse stars, the total metallicity gap of 0.2 dex is still larger than the typical spectroscopic metallicity error. Even at $[\text{Fe}/\text{H}] = -1.3$, if we assume a $\delta t_{\text{gap}} = 400 \text{ Myr}$ following the simulation, the expected gap is still about 0.11 dex, which should be measurable in high-resolution spectroscopy. The gap is even more pronounced at lower metallicities, as illustrated in Figure 1. The key signal that we aim to detect and quantify in our study.



4. MEASURING THE DISCONTINUOUS CHEMICAL TRACK

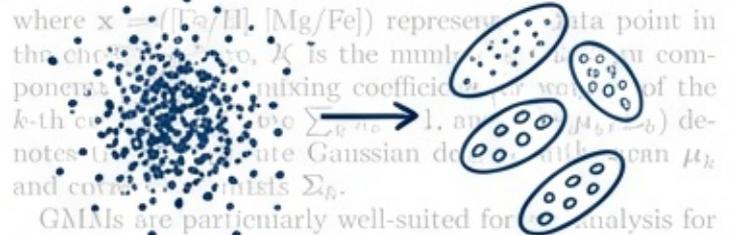
Our theory developed above suggests that measuring the discontinuous chemical track by looking for distinct stripes in the MDF is a more robust method than the bursty nature of star formation in these systems. Given the simplicity of our chemical evolution model, we seek a more robust manner to measure the track in an unbiased manner. In the $[\text{Mg}/\text{Fe}]$ - $[\text{Fe}/\text{H}]$ plane. As the name suggests, GMMs are used to model the chemical space with a mixture of Gaussians, mathematically expressed as:

$$p(\mathbf{x}) = \sum_{k=1}^K \pi_k \mathcal{N}(\mathbf{x} | \mu_k, \Sigma_k) \quad (11)$$

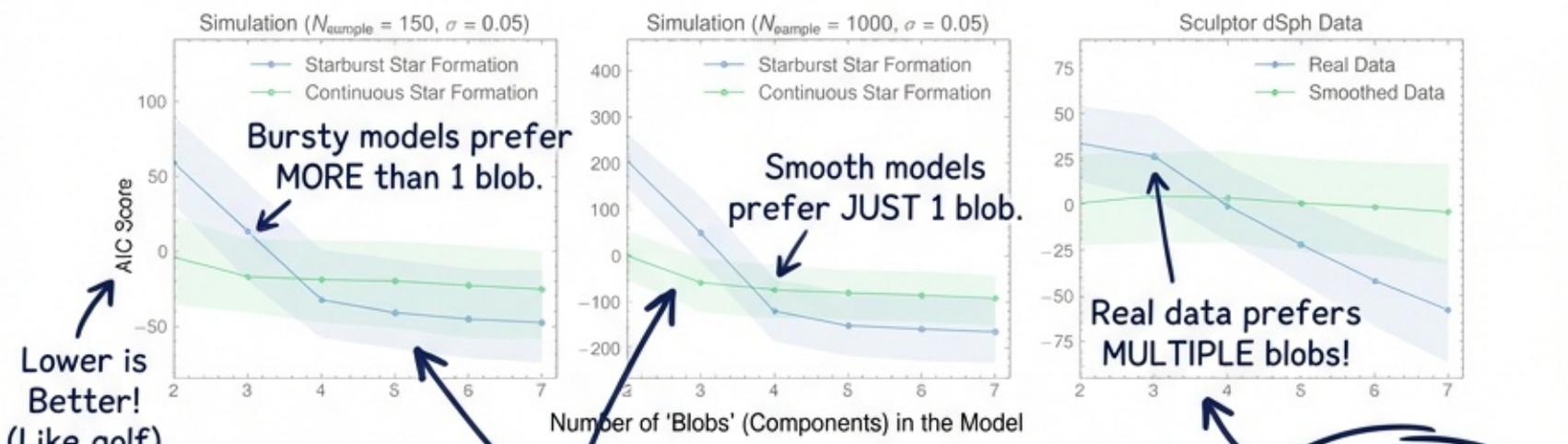
where $\mathbf{x} = ([\text{Fe}/\text{H}], [\text{Mg}/\text{Fe}])$ represents a data point in the chemical space, K is the number of Gaussian components, π_k is the mixing coefficient, μ_k denotes the mean of the k -th component, and Σ_k denotes the covariance matrix.

GMMs are particularly well-suited for data analysis for several reasons. The strong inductive bias - which assumes the data can be modeled as a mixture of Gaussian components - allows us to perform robust modeling with our small dataset (~ 100 stars). The assumption is certainly a limitation, in the absence of measurement uncertainties, chemical evolution naturally defines one-dimensional paths, with observed spread primarily reflecting measurement uncertainties. And chemical abundance uncertainties are approximately Gaussian.

Why GMMs? Because they assume the data is a mixture of blobs, which is exactly what our bursts create!



Putting our 'Un-mixing' Tool (GMMs) to the Test!



EVIDENCE for BURSTS in a REAL Galaxy!

Figure 2. Detecting discontinuous star formation by searching for multimodal components in the chemical tracks of star clusters. Models. The figure shows the Akaike Information Criterion score versus the number of components assumed in the GMM, with lower scores indicating that the model is favored. The shaded region shows the 1σ uncertainty from bootstrapping. In the bursty star formation case, if the measurement precision is comparable to or smaller than the gap between the discontinuous chemical tracks, which would be preferred. The left and middle panels show simulated data for the bursty and smooth models, respectively, with $N_{\text{sample}} = 150$ and limited precision in the abundance measurements. The right panel shows the case of real APOGEE data for the Sculptor dwarf spheroidal galaxy; the AIC score is higher for a larger number of components, superseding the sampling noise, suggestive of bursty star formation. The real data also shows a more prominent signal at larger number of components, presumably due to the fact that, unlike assumed in the simulations, the gas mass varies as a function of time, leading to individual Gaussian modes being more distinct than in the simulations.

formation. As we will see, to ensure robust results despite these limitations, we carefully validate our GMM fits through bootstrap resampling and comparison with smoothed models (see Figure 2) with simulated data, finding the method is adequate to find statistical evidence for multimodality in our dataset.

To determine the best-fitting model, we fit GMMs with different numbers of components to the chemical space and compare their likelihoods, defined as:

$$L(\theta) = \prod_{i=1}^N p(\mathbf{x}_i | \theta) \quad (12)$$

where $\theta = \{\pi_k, \mu_k, \Sigma_k\}_{k=1}^K$ represents the set of all parameters in the GMM, and N is the number of data points.

Finding the optimal parameters for a GMM is a complex non-convex optimization problem that cannot be solved analytically. We therefore fit GMMs using the Expectation-Maximization (EM) algorithm implemented in scikit-learn, which iteratively estimates the model parameters by alternating between two steps: (1) computing the probability of each data point belonging to each component (E-step), and (2) updating the model parameters to maximize the likelihood given these probabilities (M-step). The EM algorithm is guaranteed to converge to a local optimum, but may not find the global optimum. To address this limitation, we perform 1000 independent fits with different random initializations, validating that each component has at least 4 points and covariance correlations below 0.9, and select the model with the highest likelihood score. This approach helps

avoid poor local optima and ensures robust convergence to a well-behaved solution that accurately captures the underlying structure in the data.

It is important to note that GMMs with more components will have a strictly better likelihood than their lower-component counterparts: setting the weight of an additional component to zero to avoid this case. To properly evaluate the models, we need to penalize models based on their complexity. In this study, we use the Akaike Information Criterion (Akaike, 1974) defined as:

AIC is like a referee trying to find a balance.

$$\text{AIC} = 2m - 2\ln(\hat{L}) \quad (13)$$

where m is the number of parameters in the GMM ($K(1 + d + (d(d + 1))/2) - 1$ for a 2-dimensional GMM, where K is the number of modes and $d = 2$ the dimension), and \hat{L} is the maximum likelihood estimate of the model. A lower AIC is better.

Based on the AIC score, we select the model with the lowest number of Gaussian components that best fits the data. The AIC balances model fit and complexity according to the equation above. A lower AIC indicates a more favorable trade-off between fit and fit. Specifically, $-2\ln(\hat{L})$ rewards models that better fit the data, while $2m$ penalizes complex models. Under the assumption that the true model lies within the family of candidate models, AIC is asymptotically optimal in selecting the model closest (in a Kullback-Leibler sense) to the true process (Akaike [1974]). When the preferred model includes multiple Gaussian components, this naturally suggests multimodality in the chemical abundance space.

If the best AIC score is for MORE than 1 component, it means those "blobs" (and therefore the bursts) are likely REAL!

The Moment of Truth: Applying our tool to a REAL Galaxy (Sculptor)!

physical context, such a pattern is consistent with bursty star formation, wherein different episodes of star formation give rise to distinct chemical loci.

To demonstrate the effectiveness of GMMs in detecting discontinuous chemical tracks, we start with a mock sample as depicted in Figure 1. We consider two scenarios: the fiducial model with regular starbursts spaced between 150-450 Myr apart (the blue SFH in the top right panel of Figure 1) with each starburst lasting 150 Myr; and the continuous smooth star formation model that integrates to the same total stellar mass of $10^7 M_{\odot}$ (the dashed orange line in Figure 1). We draw samples according to the respective star formation rates and assume an uncertainty of 0.05 dex in both $[Fe/H]$ and $[Mg/Fe]$, similar to the quoted precision and intrinsic dispersion of APOGEE abundances in Sculptor (Mead et al. 2024).

The results are presented in the left and middle panels of Figure 2. The left panel shows a realistic case of current observations, a sample size of 150 stars with Mg and Fe measurements (Kirby et al. 2011a, Hill et al. 2019 (Abdurrouf et al. 2022)). The figure demonstrates that even with current observations, the case drawn from bursty star formation favors is able to resolve at least 4 components, as evident from the continuously decreasing AIC scores with a higher number of components. In contrast, the continuous star formation case exhibits nearly flat AIC scores with respect to the number of components, demonstrating that AIC is a reasonable penalty. However, the shaded bootstrap uncertainties in the AIC values indicate that sampling uncertainty cannot be neglected with only 150 stars.

This situation improves if we can further extend the sample size to 1000 stars, as shown in the middle panel. In this case, the sampling noise on the AIC scores decrease substantially and improve the ability to detect multimodality. Obtaining abundances of over 1000 stars in massive dSphs like Sagittarius, Fornax, and Sculptor will soon be possible with spectroscopic surveys on 4-8m telescopes like 4DWARFS (Skuladottir et al. 2023) and Subaru PFS (Takada et al. 2023). The advent of ELTs will enable such measurements in all but the faintest dSph galaxies (Ji et al. 2019). Although not shown, our tests indicate that measurement precision is less critical than sample size for detecting multimodality in these systems. The metallicity gaps, particularly at low $[Fe/H]$, are typically higher than typical spectroscopic uncertainties.

We need LOTS of stars to see these clumps clearly. Future big telescopes (like ELTs & Subaru PFS) will help better detect these metallicity gaps (~100 metallicities) and help better detect the multimodal distribution.



DETECTING BURSTY STAR FORMATION IN SCULPTOR DSPH



We now focus on real observations of the Sculptor (Scl) dSph. The Sculptor dSph is ≈ 84 kpc away (Martinez-Vázquez et al. 2015), with a dynamical mass $\sim 10^6 M_{\odot}$ (Breddels et al. 2012 | Salvadori et al. 2008) and a present stellar mass of roughly $\sim 3 - 8 \times 10^9 M_{\odot}$ (Salvadori et al.

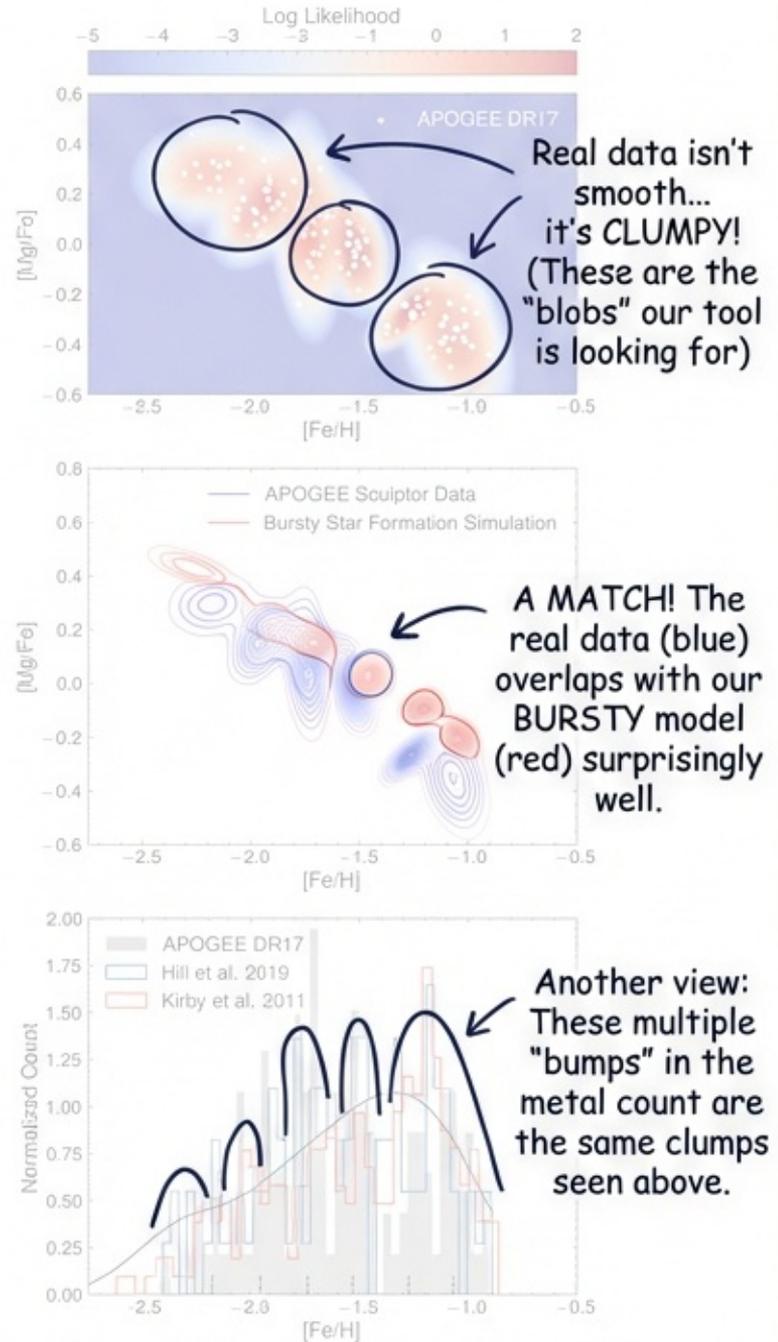


Figure 3. The APOGEE Sculptor dwarf epheroidal galaxy data shows multimodality in the $[Mg/Fe]$ - $[Fe/H]$ plane and metallicity distribution function. The top panel displays the APOGEE Sculptor sample, with the background representing the likelihood from the best-fit 6-component Gaussian Mixture Model. The middle panel compares the likelihood contours of the observed data (blue) with a 6-component GMM fit (red) to an equal number of samples drawn from our simulated bursty star formation model. The similarity between the contours of the observed data and the simulated model reinforces the interpretation of episodic star formation in Sculptor. The bottom panel presents the histogram of $[Fe/H]$ for the APOGEE sample, also exhibiting multimodality. The multimodal distribution aligns well with other Sculptor samples from Hill et al. (2016) and Kirby et al. (2013). The vertical lines in the bottom panel indicate the fitted means of the GMM components, and the solid black line shows the smoothed MDF of the simulated sample.

We used real chemical data from the APOGEE survey for stars in the Sculptor dwarf galaxy...

TING & JI

Our AIC "referee" (from a few pages back) confirms:



2008 [de Boer et al. [2012] [Bettinelli et al. [201a] that corresponds to an initially formed stellar mass of $\sim 10^7 M_{\odot}$. Sculptor has extensive chemical data (Kirby et al. [2011a] Hill et al. [2019] sie los Reyes et al. [2022] [Abburroif et al. [2022]) and star formation histories from color-magnitude diagrams (de Boer et al. [2012] [Weisz et al. (2014) [Vincenzo et al. [2013] [Bettinelli et al. [2019) that suggest It ...and carefully selected stars that are definitely members of Sculptor (based on how they move) to avoid contamination.

We use chemical abundances of Sculptor from APOGEE/ASPCAP DR17 (Garcia Pérez et al. [2016, Abburroóf et al. [2022]). We selected member stars using proper motion members from Pace et al. [2022) and restricted to stars within three velocity dispersions of the central velocity (Walker et al. (2009). We then restrict to stars with $\text{SNR} > 30$; good quality flags BRIGHT_NETGBBOR, VERY_BRIGHT_NETGBBOR, and SUSPECT_BROAD_LINES; good ASPCAP flags STAR_BAD, METALS.BAD, and ALPHAFE.BAD; $[\text{Mg}/\text{Fe}]$ error < 0.1 dex; and $\log g < 2$. This results in 124 member stars with high quality abundances.

In the bottom panel of Figure 3, we show the histogram of metallicity $[\text{Fe}/\text{H}]$. Even when projected onto the metallicity distribution which are given out the multimodal structure. The text here is analyzing the plots we just saw on the previous page. The chemical "bumps" in the real data are our real and match our "bursty" model!

and the means of these GMMs are shown as vertical dashed lines at the bottom to guide the eye. For comparison, we overplot the metallicity measurements from Hill et al. (2019) and Kirby et al. (2011a). We shift Kirby's measurements by 0.1 dex in $[\text{Fe}/\text{H}]$ to match the peak locations seen in the Hill and APOGEE data, which is consistent with a 0.07 dex zero-point offset relative to high-resolution spectroscopy noted in that paper. Careful examination of the metallicity distributions remarkably

We checked our results against other independent studies, and they agree! This gives us confidence in our findings.

In the bottom panel, we also present the smoothed MDF of the simulated bursty chemical evolution model, obtained using a KDE with a smoothing scale of 0.2 dex. The good agreement between the simulated and observed MDF is because we adjusted the fiducial star formation history to match the observations (blue SFH in the top right panel of Figure 1). Without smoothing, the simulated results would exhibit the same multimodal structure as the observed data, as evident from the contours

in the middle panel.

In the top panel, we show a 6-component GMM fit to the Sculptor data. The colored background in the top panel of Figure 3 shows the likelihood of the best-fit 6-component model, illustrating the multimodal nature of the data. The middle panel of Figure 3 shows the locations of the multimodal data points, which are consistent with the fiducial simulated model. This consistency between the real data and the simulated bursty star formation history suggests that the Sculptor dwarf spheroidal galaxy experienced episodic star formation throughout its history.

To more robustly quantify the multimodality, we calculate the AIC score estimate the uncertainty of the AIC score by bootstrapping the data. The AIC score as a function of the number of components is shown in the blue band in the right panel of Figure 2. Just like in the case of the simulated data in the left panel, the real data also exhibits a similar trend, with the AIC score decreasing as the number of components increases, signaling a multimodal nature in the $[\text{Mg}/\text{Fe}]$ - $[\text{Fe}/\text{H}]$ plane. For comparison, we show the AIC score for a smoothed data example as a green band in the right panel. We generate the smoothed data by fitting the real data with a two-component GMM¹, draw from the distribution, and ensure that the smoothed distribution follows the contours of the data. In this case, the smoothed data shows no trend in the AIC scores, demonstrating that the real data has a statistical tendency to favor the case of multimodality. The close agreement between the AIC scores for the real data and the simulated data provides evidence linking the real data to the simulated star formation history.

Interesting detail: The real data is even "burstier" than our simulation! We think this is because real galaxies have varying amounts of gas, while our simple model assumed it was constant.

Interestingly, the AIC trend is even stronger for the real data than the simulated data. One potential reason is that in the simulated data we assume a constant effective gas mass. Thus each 2D Gaussian mode in the $[\text{Mg}/\text{Fe}]$ - $[\text{Fe}/\text{H}]$ space has roughly the same tilt and spread, as shown by the red ellipses in the middle panel of Figure 3. However, in the real data, as the effective gas mass might vary, which could lead to more distinct Gaussian modes, as shown in the background of the top panel and the blue contours in the middle panel. This would result in a stronger AIC score trend as a function of the number of components, continuing to decrease above four components.

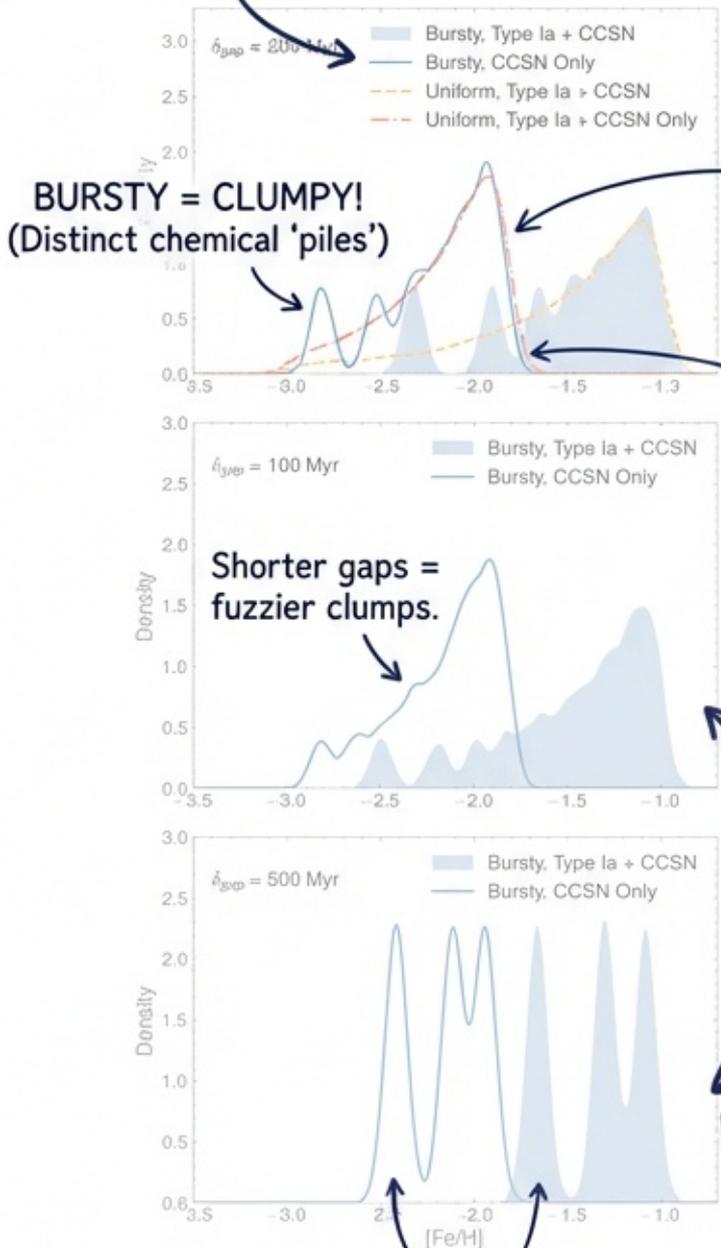
6. DISCUSSION

TIME TO WRAP UP & DISCUSS!

In this study, we develop an analytic theory demonstrating that bursty star formation leads to a discontinuous chemical track that is distinguishable from continuous star formation.

Summary of our main contribution: We built a theory showing how bursty star formation leaves a unique 'fingerprint' in a galaxy's chemistry.

Let's zoom in on the details of this 'fingerprint'...



UNIFORM = SMOOTH. (No distinct piles)



The TIME between bursts matters!

We find that the AIC score trend is distinguishable from the smoothed case. The observed signatures are consistent with other datasets of Sculptor as well as the theory developed, supporting an episodic star formation history galaxy.

Time Resolution for Dwarf Galaxy Star Formation Histories

In Figure 4, we emphasize that the strength of the multimodal signal relies on the production of Fe by Type Ia iron factories, the clumps are much less clear! The 'gaps' get filled in.

Without Type Ia Supernovae (the 'slow' iron factories), the clumps are much less clear! The 'gaps' get filled in. Such high time resolution in star formation histories of local dwarf galaxies means that our chemical method can be resolved even down to shorter timescales than other methods.



Our chemical method acts like a high-resolution camera, seeing short, sharp bursts (~100 Myr) that other techniques miss!



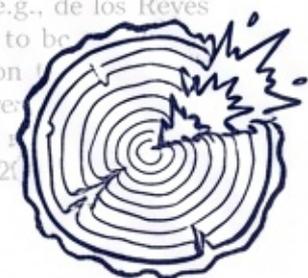
Figure 4. The contribution from Type Ia supernovae enhances the prominence of multimodal star formation histories, each integrating star formation over 2 Gyr and assuming a mixing gas mass of $3 \times 10^8 M_{\odot}$. The shaded blue region in the top panel depicts the MDF for the fiducial case with instantaneous bursty star formation and a gap (δ_{gap}) of 200 Myr. For comparison, the top panel also includes the MDF for uniform and continuous star formation, represented by the dot-dashed red and dashed orange lines. In these cases, the MDF is smooth and lacks multimodality.

ous star formation. In particular, we expect to find multiple modes in the [Mg/Fe]-[Fe/H] plane and in the metallicity distribution function. We adopt Gaussian Mixture Models to detect such multimodality, as they are model-agnostic. As a proof of concept, we apply this method to the existing APOGEE data of the Sculptor Sph galaxy.

The episodic star formation we suggest occurs in Sculptor may be consistent with recent JWST observations of high redshift galaxies that display miniquenches (Antwi-Danso 2024; Dome et al [2024], Looser et al [2024]) but for galaxies whose stellar mass is much smaller, it could be detected directly at high redshift. The progenitor of Sculptor has $M_{\text{star}} \sim 10^8 M_{\odot}$ (Dominguez et al [2024]), which is the lowest mass spectrum of galaxies that JWST observations provide unprecedented time resolution for the star formation histories of early low mass galaxies.

A natural question is whether even shorter timescales of ~10 Myr could be resolved by chemical evolution. Indeed we assumed that all CCSN produce Mg at a single time, but in reality there is a delay time distribution for CCSNe as well lasting from 3-30 Myr (e.g., de los Reyes et al [2022]). We found this is unlikely to be resolved for low mass galaxies, as repeated bursts on timescales of ~10 Myr would be smoothed out by the time resolution of chemical evolution.

Basically, the chemistry records the history of star formation like tree rings... but with EXPLOSIONS!



OK, let's check our "starburst signal" against OTHER dwarf galaxies...

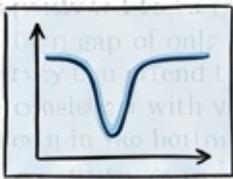


- ✓ **WLM & Carina: Predicted chemical gaps generally match observations! (Score!)**
- ? **Reticulum II: A bit of a puzzle. Might need to adjust our model (e.g., gas loading factors).**

The predicted [Fe/H] gaps generally align well with the observed data, with Reticulum II being a potential exception. This discrepancy can be resolved if we adjust the ratio of mixing gas mass to stellar mass, which is expected due to higher mass loading factors in lower mass galaxies (Alexander et al. [2023] Sandford et al. [2027]). We found that increasing this ratio from 30 to 10 would bring the predicted metallicity gap for Reticulum II into agreement with observations. Our initial investigation makes very simple assumptions about the bursty star formation history and detailed gas mass history, and we plan to address these assumptions in future work. However, the very fact that this simple prescription already reasonably reproduces the age and [Fe/H] gaps in dwarf galaxies suggests that the basic setup is correct.

As shown in Figure 5, a typical mid-resolution survey with a relative precision of metallicity of 0.10 should be able to detect a metallicity gap at a relatively low metallicity of [Fe/H] = -2, as the metallicity gap in the MDF is larger than the uncertainty, even with a quiescent period of only $\tau_{\text{qsp}} = 150$ Myr. A high-resolution survey can end this limit to [Fe/H] = -1.7. The former is consistent with what we see in the Sculptor dSph. As shown in the bottom panel of Figure 5, while the sample size is small (Kirby et al. [2011a]), which is from mid-resolution surveys (Hill et al. [2011]), it still detects a quiescent period of [Fe/H] = -2.2.

PREVIEW OF FIGURE 5 (It's coming!): This figure will show us the predicted 'fingerprint' of these gaps in the chemical data. Keep an eye out for the 'dips'!



6.5. Caveats and Future Directions

A key strength of studying episodic star formation through gaps in elemental abundances, rather than rely-

CAVEATS & LIMITATIONS

(Honesty is the best policy in science!)

ing solely on temperature and metallicity, is largely constrained by Type Ia supernovae. It is crucial to recognize that the actual star formation history is likely more complex, potentially comprising numerous mini-scale bursts on timescales of 10-30 Myr within the 'starbursts' identified in this study (e.g., Wheeler et al. [2019] Patel et al. [2013]). Our approach thus provides a high-level view of the episodic star formation history, focusing on major starbursts defined by sufficiently long quiescent periods.



1. The SN Ia "Clock" is Slow: Our method's time resolution is limited by the delay time of Type Ia supernovae. We can't see super-short bursts.

Our method's time resolution is limited by the delay time of Type Ia supernovae. It is crucial to recognize that the actual star formation history is likely more complex, potentially comprising numerous mini-scale bursts on timescales of 10-30 Myr within the 'starbursts' identified in this study (e.g., Wheeler et al. [2019] Patel et al. [2013]). Our approach thus provides a high-level view of the episodic star formation history, focusing on major starbursts defined by sufficiently long quiescent periods.

In this study, we use a constant effective mixing gas mass $M_{g,\text{eff}}$. While this simplifies the description of the metallicity gap, it does so at the cost of making the gas mass less interpretable: $M_{g,\text{eff}}$ is a degenerate combination of gas and metal inflows and outflows. In contrast, standard chemical evolution models tie the gas mass (usually the cold gas within the galaxy) to the star formation and outflow rates through a star formation efficiency and mass loading factor or prescription (e.g., [Cescutti 2008] Salvadori et al [2008] Ishimaru et al [2015] Weinberg et al [2017] Johnson et al [2023]). While such parameterizations are physically well-motivated and provide good results for overall dwarf galaxy scaling relations, current implementations prohibit the type of bursty star formation seen in hydrodynamic simulations.

The effectiveness of our constant $M_{g,\text{eff}}$ assumption likely varies with galaxy mass. The potential well depth of galaxies is a potential well depth, which is not applicable to ultra-faint galaxies. Feedback effects might require more sophisticated models. However, our reproduction of Sculptor's chemical patterns suggests that for



2. The 'Constant Gas' Assumption Again: Remember that 'effective mixing gas mass' ($M_{g,\text{eff}}$)? Assuming it's constant is a BIG simplification. Real galaxy plumbing is messier.

Putting our model to the test with real data!

Galaxy	δt_{gse} (Gyr)	[Fe/H]	Observed [Fe/H] Gap	Predicted [Fe/H] Gap*	References
Sculptor	0.3 ± 0.15	-1.9 ± 0.6	0.20 – 0.26	0.20	This Work
WLM	3 ± 1	-1.7 ± 0.3	0.3	0.26	1
Carina	4.5 ± 2.5	-1.8 ± 0.3	0.35	0.36	2,3
Reticulum II	3.4 ± 1	-2.6 ± 0.1	0.04	1.09 (0.65 ^b)	4.5

✓ Pretty good match!

✓
✓
✗

Table 1

Comparison of star formation quiescent periods (δt_{gse}) and the resulting chemical gap between star formation episodes, the metallicity distribution function (MDF) slope, and the observed metallicity gap. For each galaxy, we compare the observed metallicity gap to the predicted metallicity gap from our model.

Comparing what we see vs. what our toy model predicts.

✗ Missed this one. Needs a better "plumbing" model (as noted before).

References: 1: [McQuinn et al. 2024](#); 2: [Norris et al. 2017a](#); 3: [Norris et al. 2017b](#); 4: [Simon et al. 2023](#); 5: [Li et al. 2023](#) and Luna et al. in prep.

*: The predicted [Fe/H] gap is based on a stellar mass to mixing gas mass ratio of 1/30. The value in brackets shows the prediction if we assume a ratio of 1/70 instead.

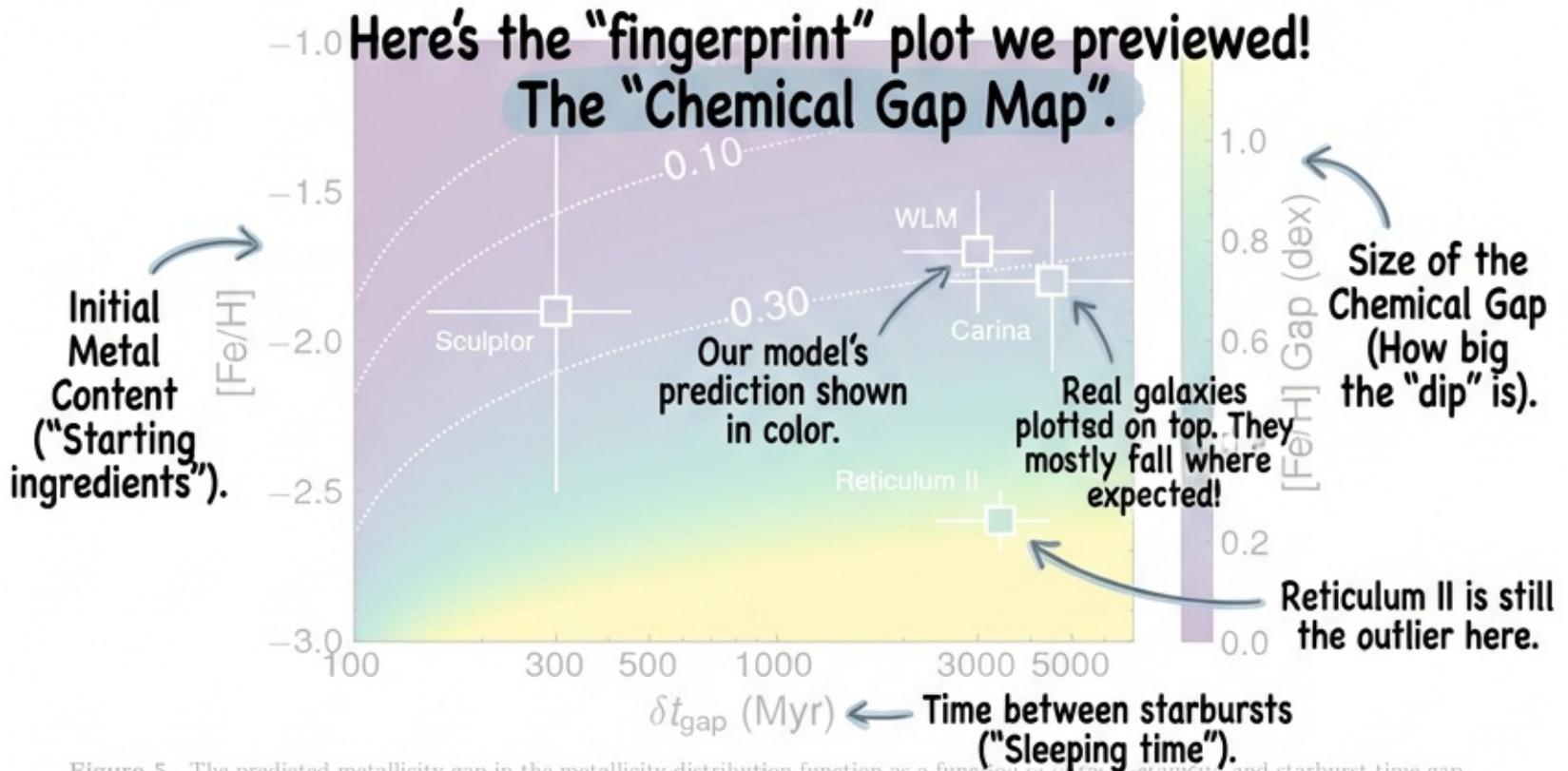


Figure 5. The predicted metallicity gap in the metallicity distribution function as a function of initial metallicity and starburst time gap. The white dotted lines include the typical uncertainties of metallicity measurements from high-resolution spectroscopy ($R \approx 20,000$, 0.05 dex), mid-resolution spectroscopy ($R \approx 6,000$, 0.1 dex), and photometric metallicities (0.3 dex). Overplotted are measurements from a few dwarf galaxies. The Sculptor dwarf galaxy data points are measured in this study by matching the [Mg/Fe]-[Fe/H] contours. For the other galaxies, Carina, WLM and Reticulum II, the starburst time gap is estimated from color-magnitude diagrams, and the metallicity gap (represented by the color of the symbols) is determined from the [Fe/H] distribution. Despite the simplicity of our analytic model, it accurately predicts the metallicity gaps when assuming an integrated SFR to mixing gas mass ratio of 1/30.

star formation. Future work incorporating time-varying gas masses will help quantify how feedback efficiency impacts the visibility and interpretation of these chemical signatures across different mass scales (e.g., [Agortz et al. 2013](#), [Read et al. 2016](#)).

7. CONCLUSION

In this initial foray into the impact of episodic star formation on chemical evolution, we develop a model to describe bursty star formation and its impact on the chemical properties of stars. We have demonstrated that episodic star formation in dwarf galaxies leads to distinct chemical signatures compared to continuous star formation scenarios. The multimodal chemical signature provides access to timescales much shorter than possible

THE BIG PICTURE WRAP-UP.



1. Bursty star formation creates clear 'chemical gaps' & 'clumps'.



2. These signals are created by the delay of Type Ia Supernovae.



3. The good news: These features are BIG enough to see with telescopes we have right now!

already detectable, with sufficient statistical significance according to the Akaike Information Criterion, in current datasets of $\mathcal{O}(100)$ stars. However, larger samples remain important for reducing sampling noise and improving the robustness of the results.

As a proof of concept, we apply our model to the APOGEE data of the Sculptor dSph galaxy. Our findings are:

- Using Gaussian Mixture Models and the Akaike Information Criterion, we show that the [Mg/Fe] and [Fe/H] distributions in the Sculptor dSph galaxy are best described by a mixture of two populations: a main population and a sub-population of stars that formed during quiescent periods.
- We show that the quiescent periods in the Sculptor dSph galaxy are consistent with the quiescent periods in other dwarf galaxies with Gyr-long quiescent periods that could be probed with color-magnitude diagrams, including WLM, Carina, and Reticulum II.

THE FINAL VERDICT ON SCULPTOR:

- ✓ 1. Stats & Sims confirm: Star formation was **BURSTY**, not smooth!
- ✓ 2. Sculptor took 'naps' (quiescent periods) lasting ~150-450 Million Years.
- ✓ 3. Matches what JWST sees in other galaxies ('miniquenching') & extends it to fainter ones!

While our theory and analysis are admittedly simplistic, this study paves the way for a more comprehensive understanding of the complex interplay between episodic star formation, feedback, and the mixing gas mass in dwarf galaxies. Crucially, our work provides a theoretical

WHY THIS MATTERS FOR THE FUTURE:

Current models are too 'smooth'.

We need 'bursty' models to interpret data from upcoming BIG telescopes.

Our simple theory is a foundation for that!

these and other models that take full advantage of the recent and upcoming major advances in chemical abundance quality and quantity.

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Science happens in interesting places! Shout out to the marsupials & rabbits at ANU in Canberra for the 'serene environment'.



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★ THAT'S A WRAP! THANKS FOR READING.

THE FOUNDATION OF IT ALL: STANDING ON THE SHOULDERS OF GIANTS!



A PhD is basically just reading a lot of papers... and then adding one more to the pile. Thanks to everyone who came before me!

SCIENCE IS A TEAM SPORT!

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And the list goes on... and on...
 It really highlights how every new paper is just one small brick in a massive wall of knowledge built by thousands of others. Truly humbling.

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Alright, that's a wrap!
 Thanks for joining me on this deep dive into the messy, bursty lives of dwarf galaxies. It's been fun!
 Now, time for a nap... 😴