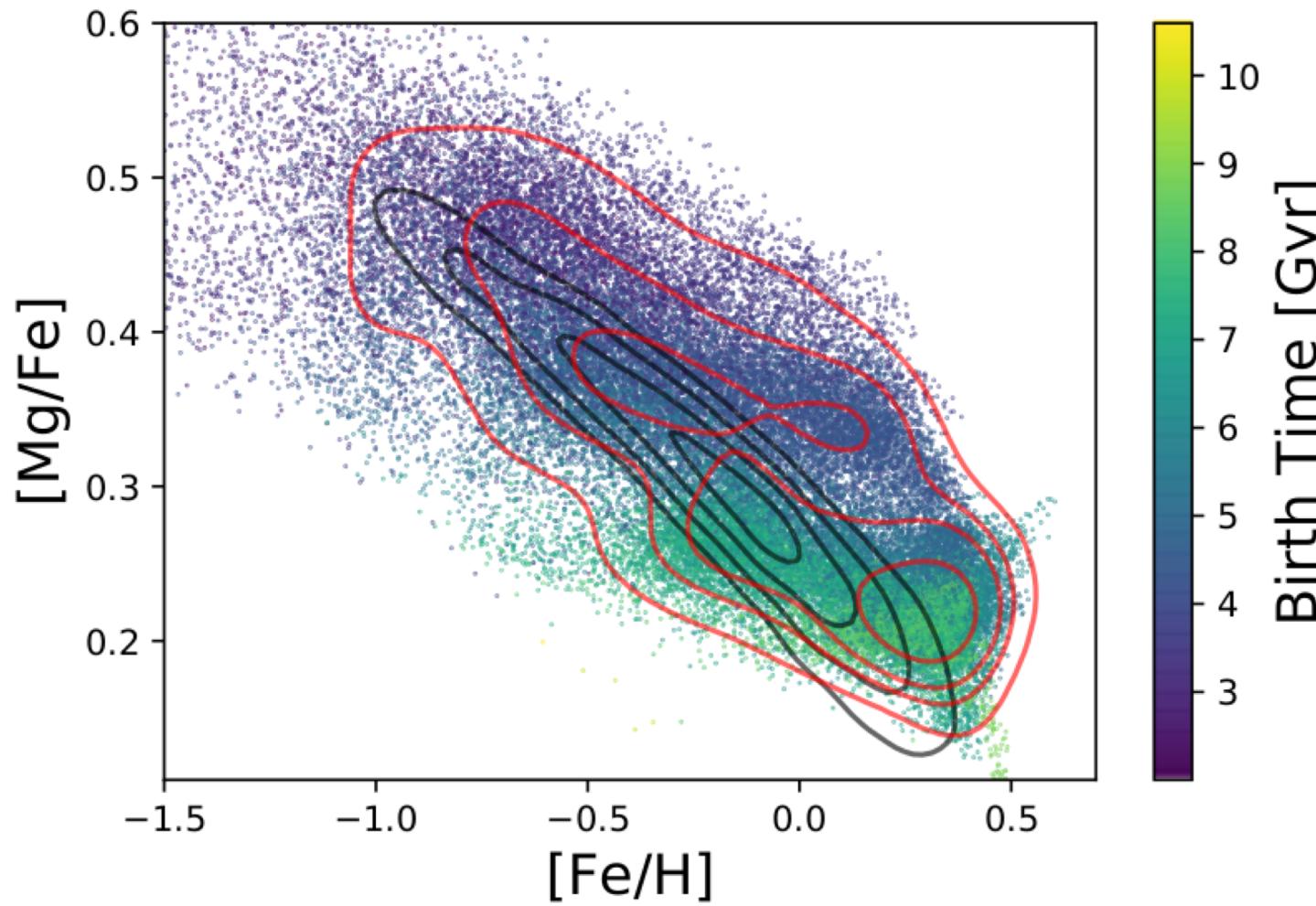




PRINCETON  
UNIVERSITY



# Inferring Galactic Parameters from Chemical Abundances: A Multi-Star Approach

JAN RYBIZKI (MPIA)

OLIVER PHILCOX (PRINCETON)

JINA-CEE ONLINE SEMINAR

Nov 22, 2019

# Outline

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1. *Chempy*: A Fast and Flexible GCE Model

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2. Choosing Yield Tables for Hydrodynamical Simulations

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3. Scoring Nucleosynthetic Yield Tables

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4. Inferring Galactic Parameters with Multi-Star Inference

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1. *Chempy*: A Fast and Flexible GCE Model
2. Choosing Yield Tables for Hydrodynamical Simulations
3. Scoring Nucleosynthetic Yield Tables
4. Inferring Galactic Parameters with Multi-Star Inference
5. Further Extensions

# Building a Fast and Flexible Galactic Chemical Evolution Model with *Chempy*

# Building a Fast and Flexible GCE: *Chempy*

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- ❑ *Chempy* (including **chemical yields**, and **SSP** parameters):

**SSP = Simple Stellar Population**  
(A group of stars born at the same time in the same environment)

# Building a Fast and Flexible GCE: *Chempy*

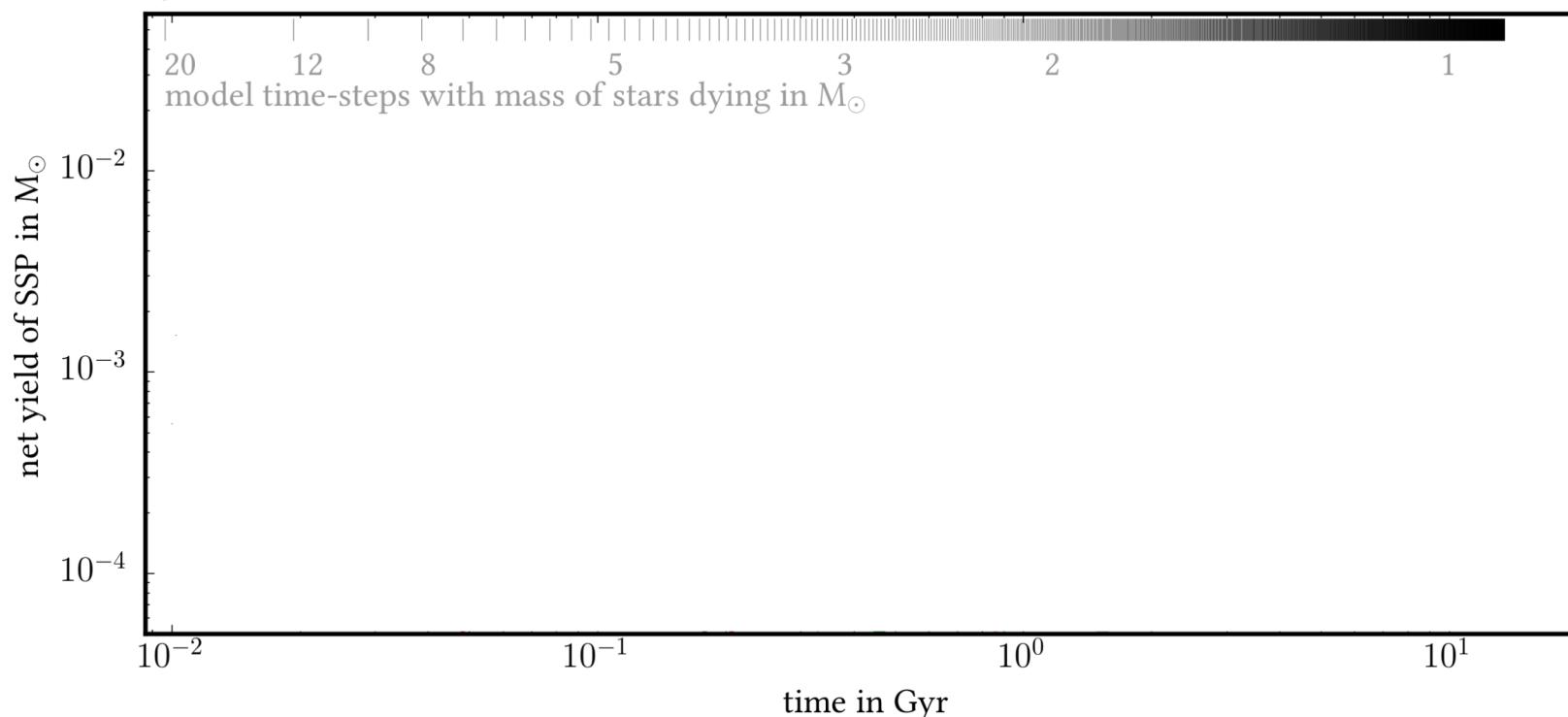
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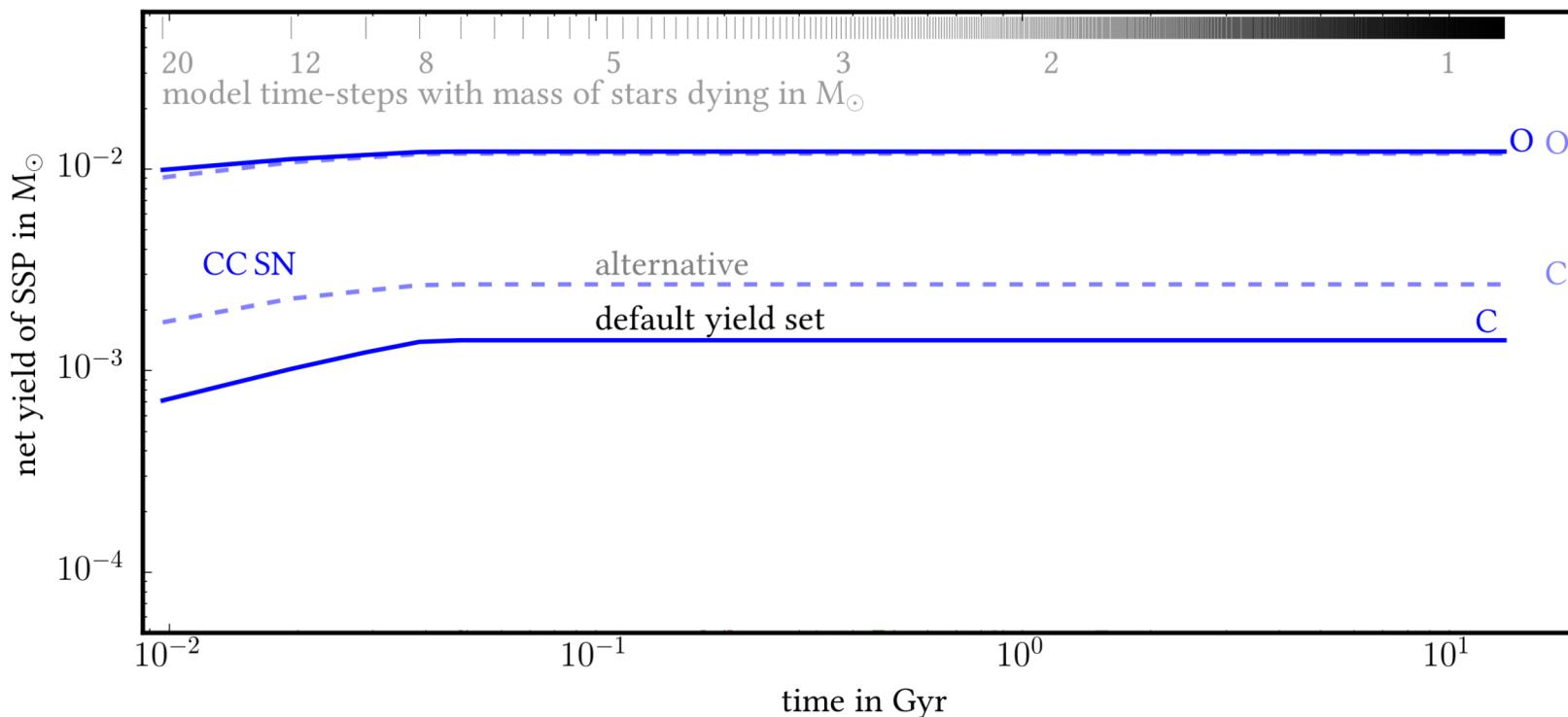


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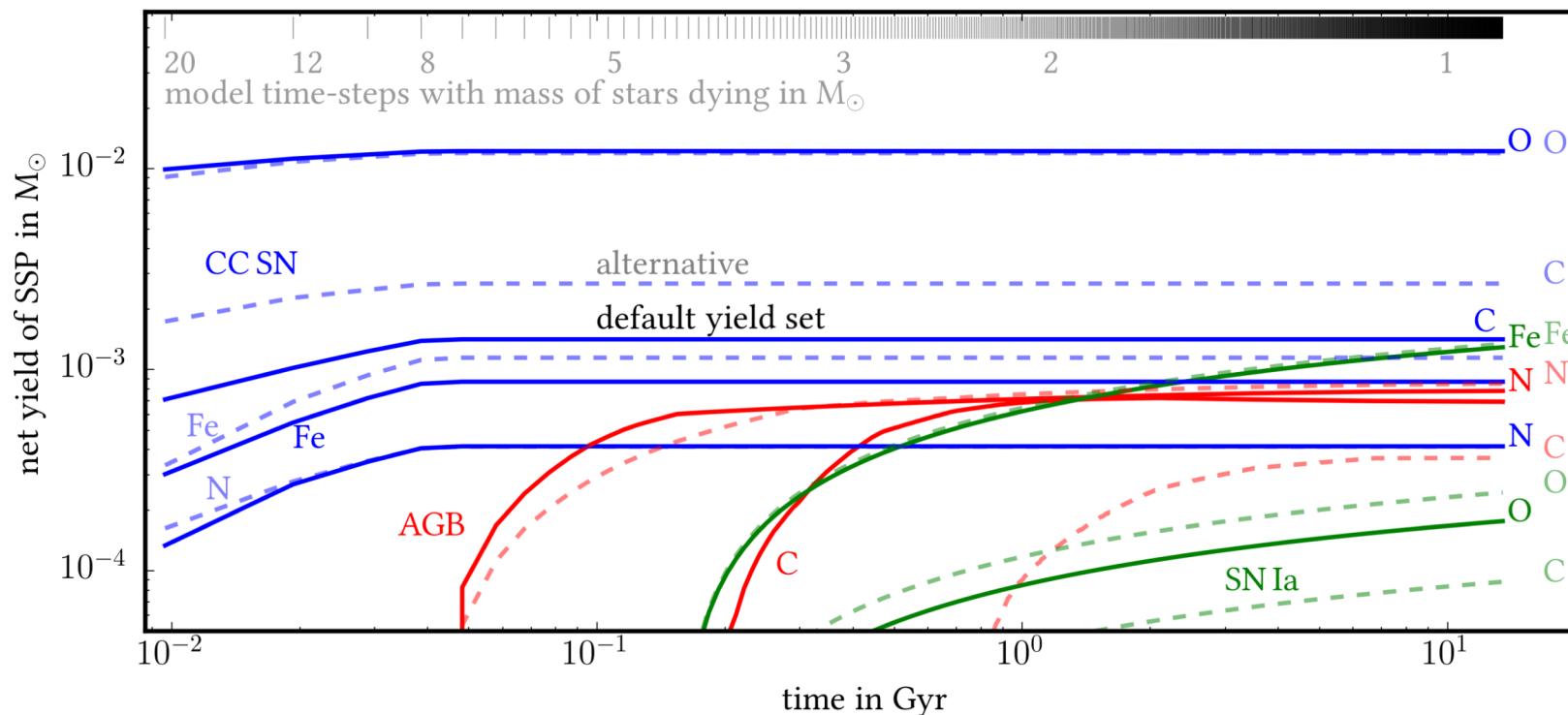


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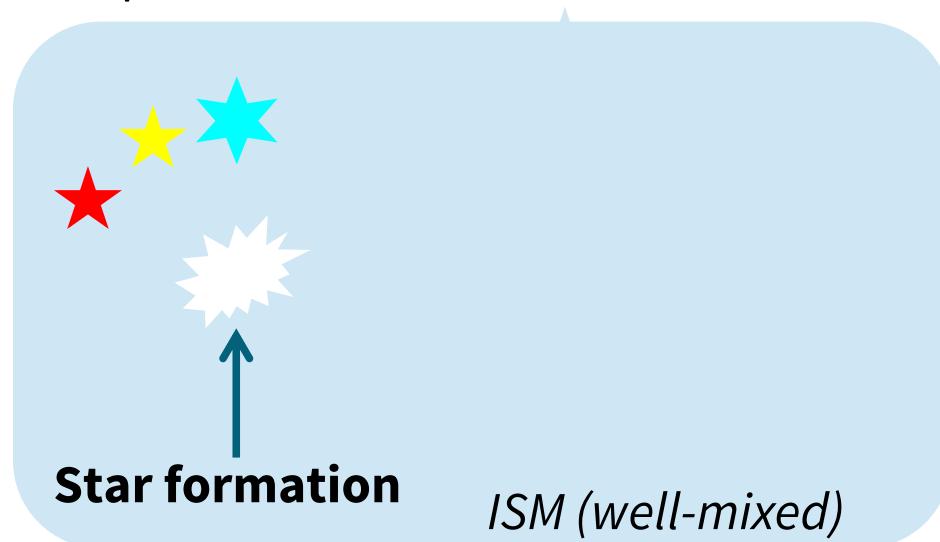
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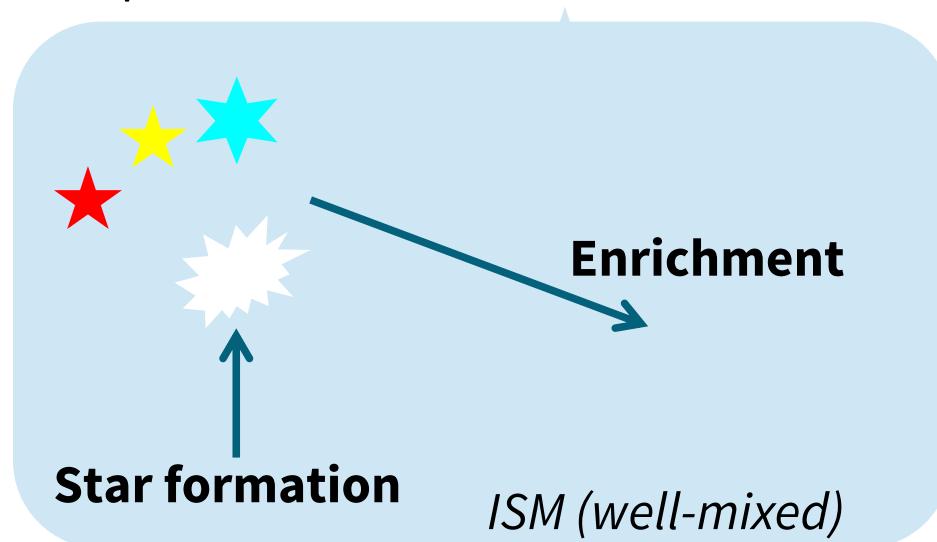
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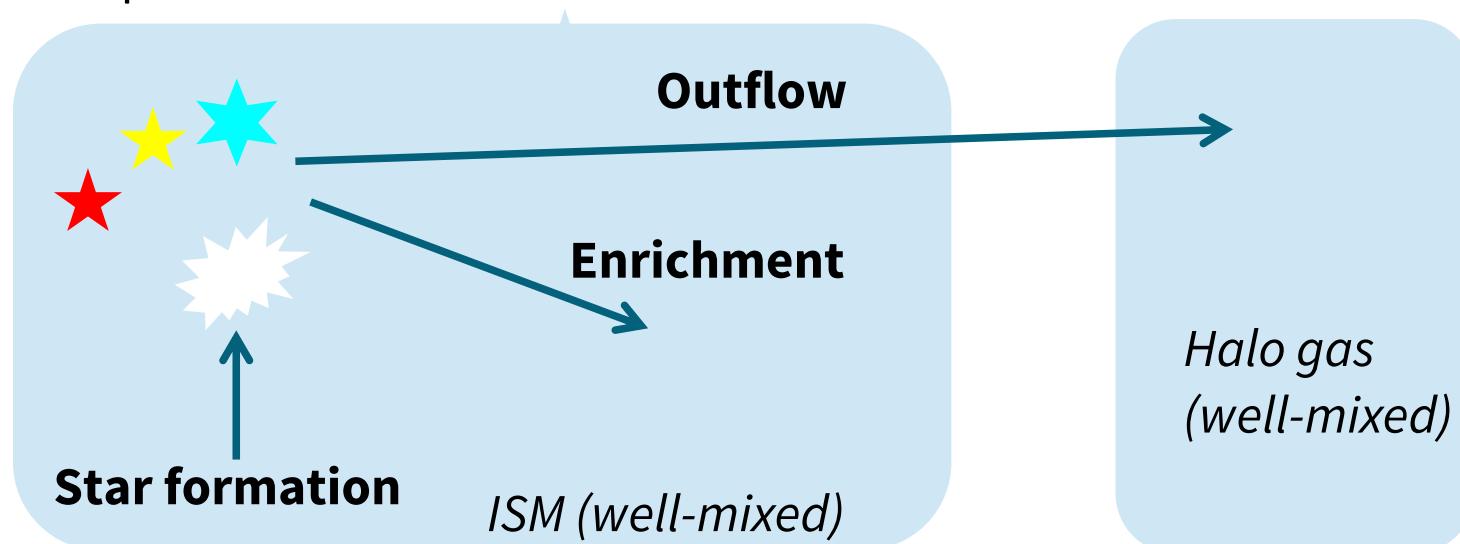
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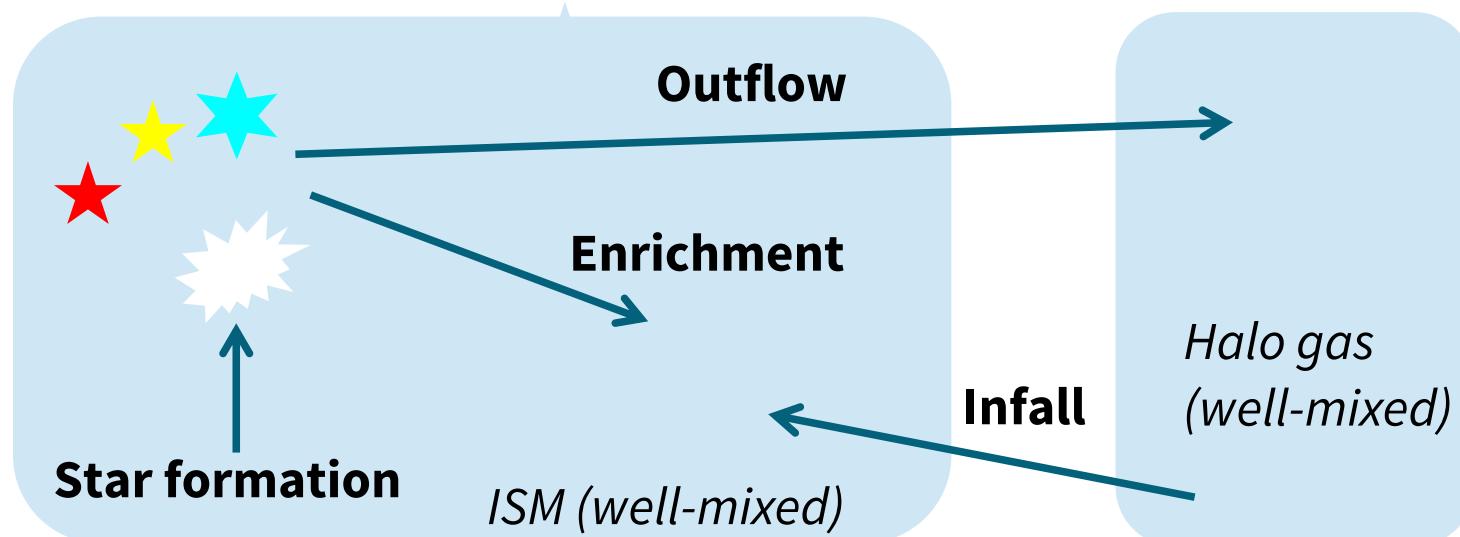
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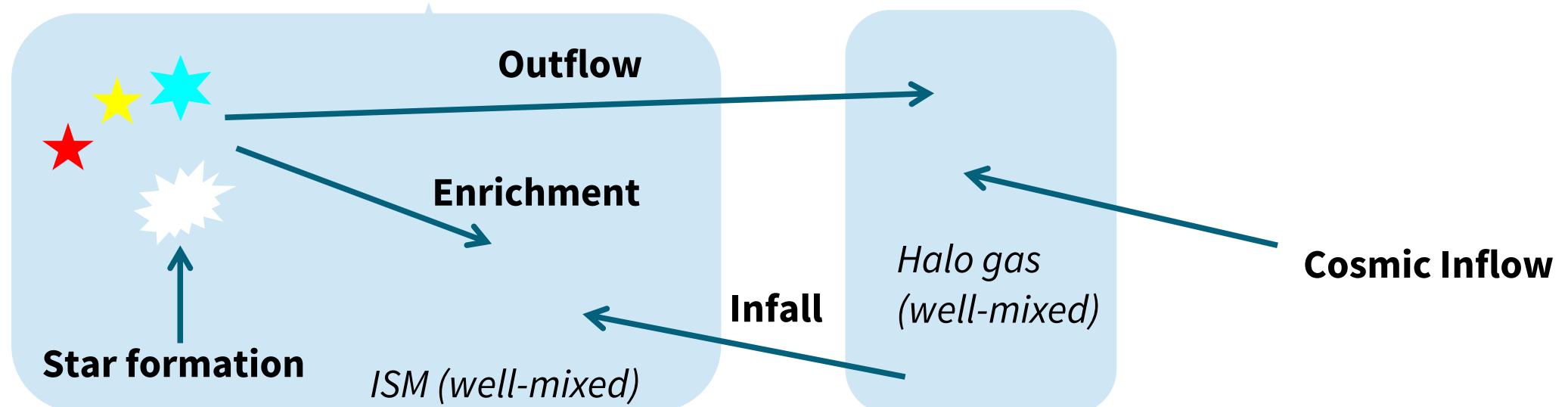
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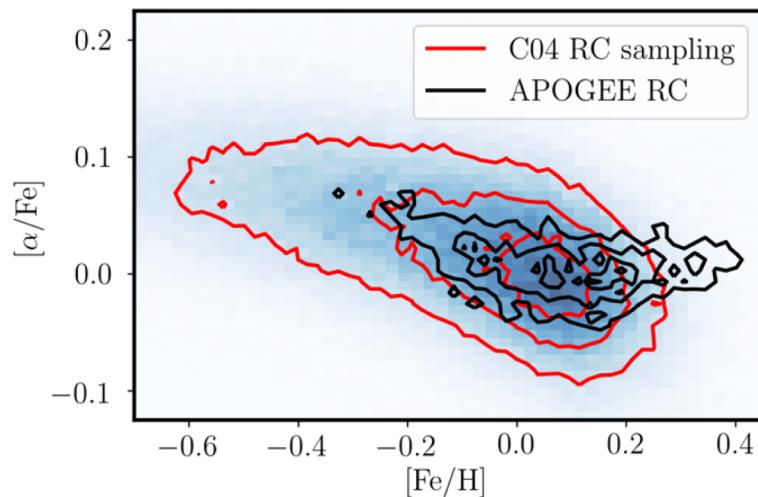
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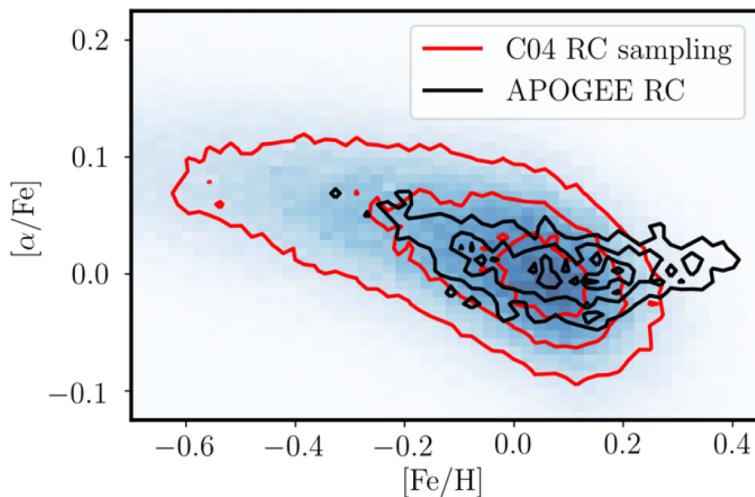
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*This needs **no fiducial model**,  
Input your data → Obtain **GCE parameters***

Rybizki, Just & Rix (2017)

# Parameter Inference with *Chempy*

---

- ❑ *Chempy* can be inserted in a **Bayesian** framework to **infer** Galactic parameters.

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## SSP Parameters

- Initial Mass Function Slope
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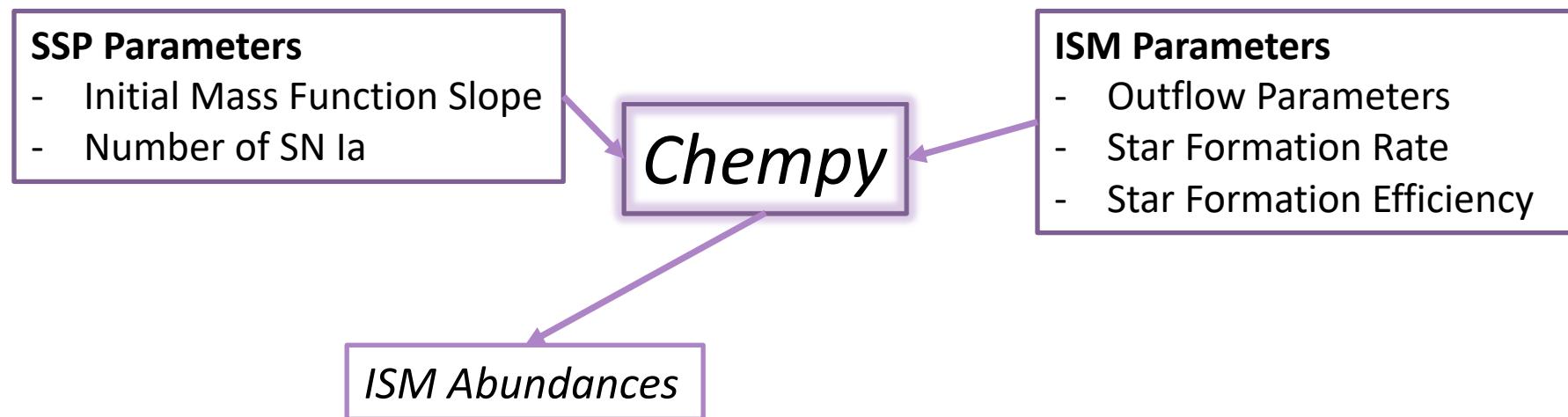
## ISM Parameters

- Outflow Parameters
- Star Formation Rate
- Star Formation Efficiency

Rybizki, Just & Rix (2017)

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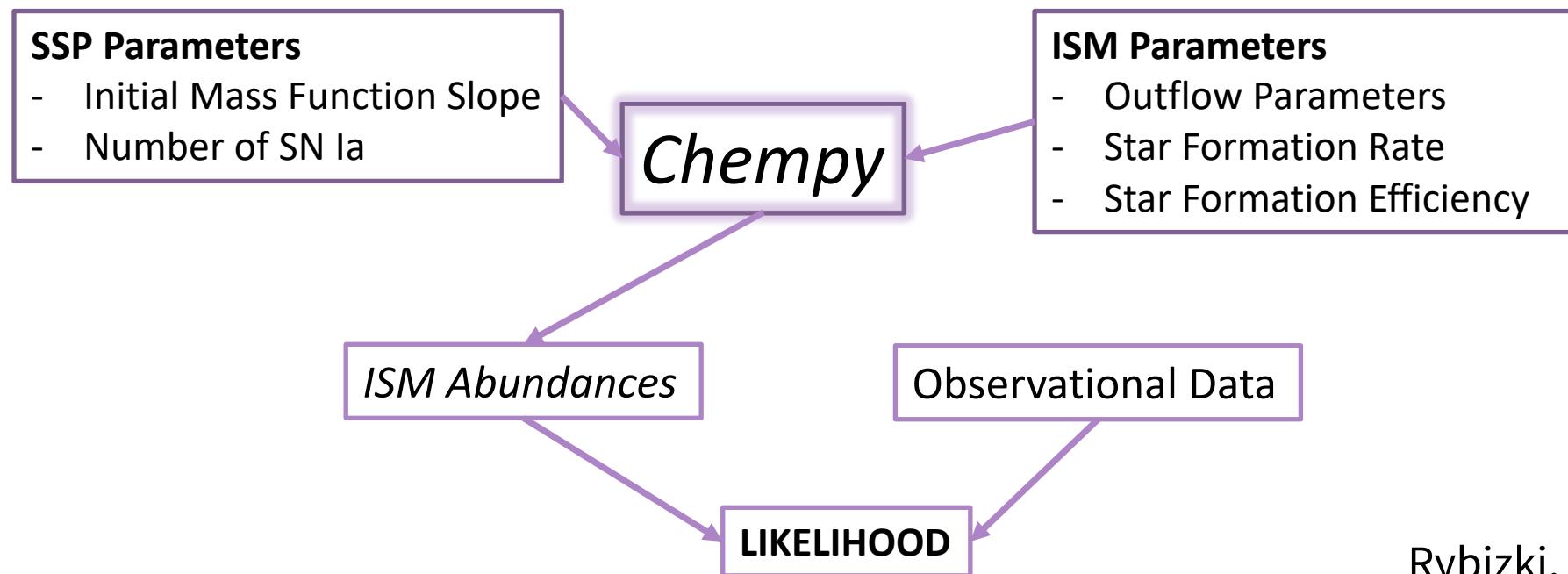
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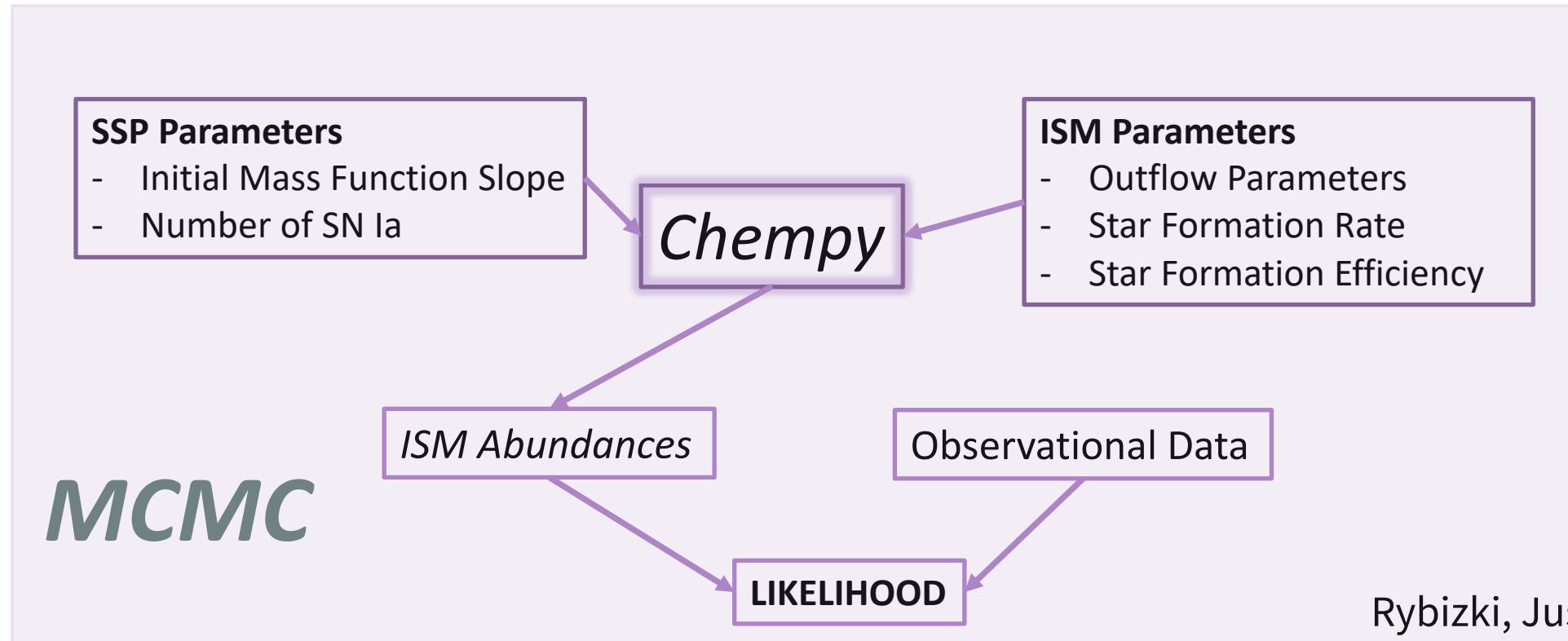
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# Parameter Inference with *Chempy*

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Can *Chempy* be used to  
determine GCE parameters for  
hydrodynamical simulations?

# Predicting SSP Parameters for Hydro Sims

---

- ❑ **Goal:** Predict the best SSP parameters for hydrodynamical simulations to make sure they are consistent with observations

Philcox, Rybizki & Gutcke (2018)

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- **Method:**
  - Determine optimal SSP parameters for specific simulation using *Chempy*
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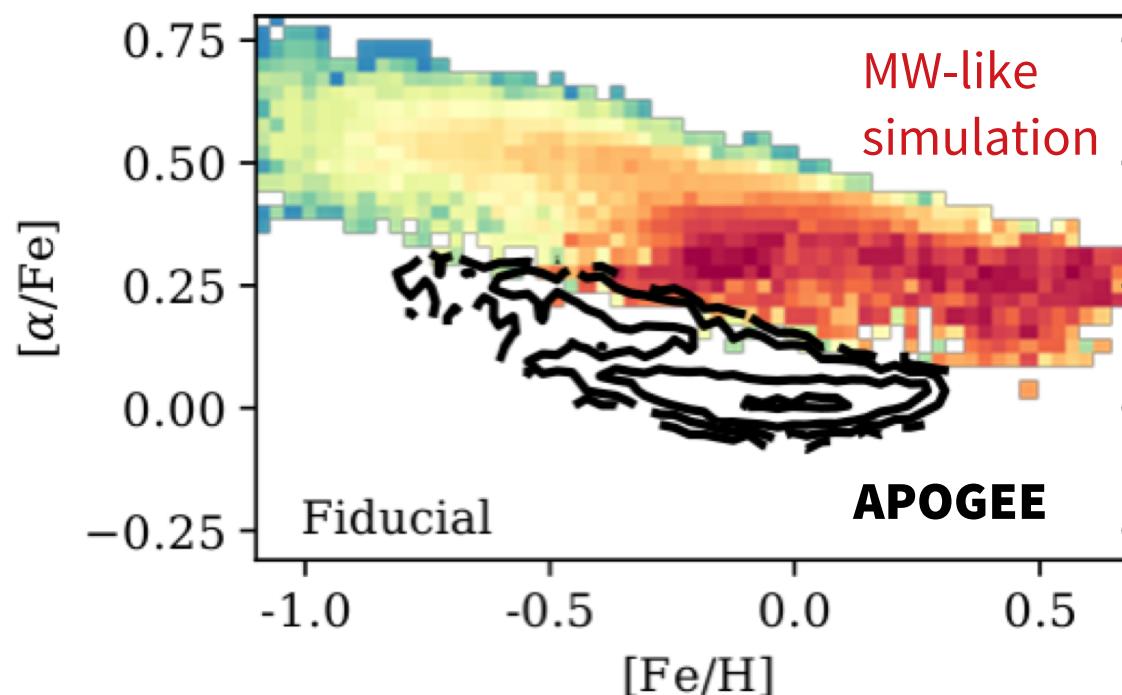
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  - ❑ Optimize using the **proto-Solar** abundances alone
  - ❑ Run simulation with **optimized** parameters and compare results!

Philcox, Rybizki & Gutcke (2018)

# Predicting SSP Parameters for Hydro Sims

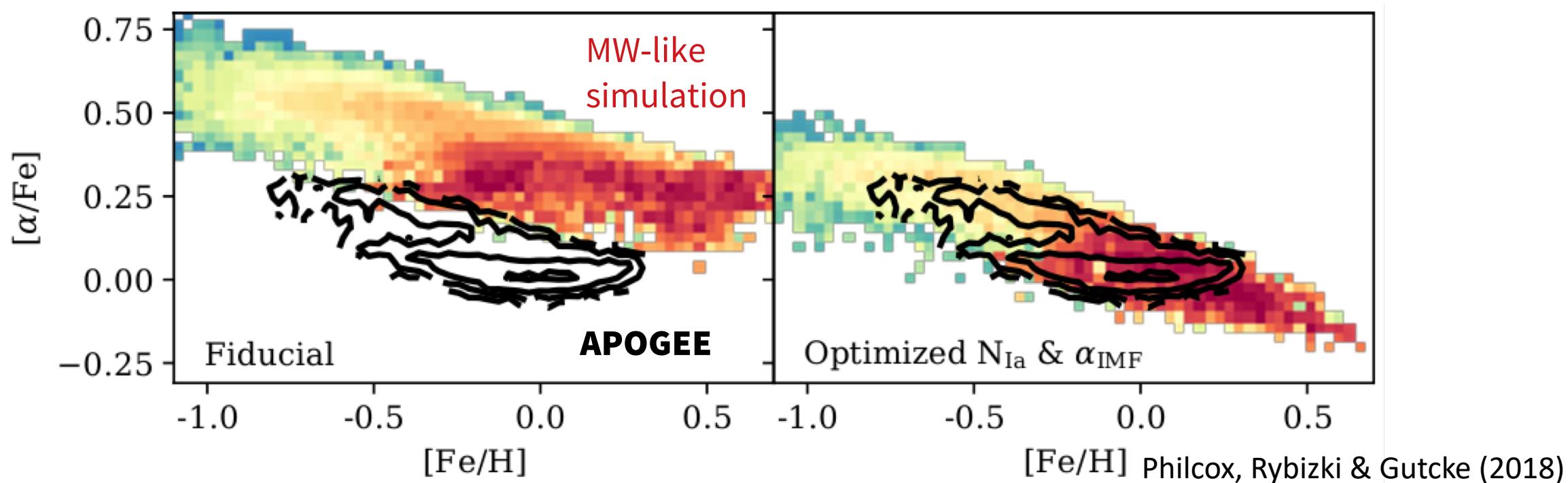
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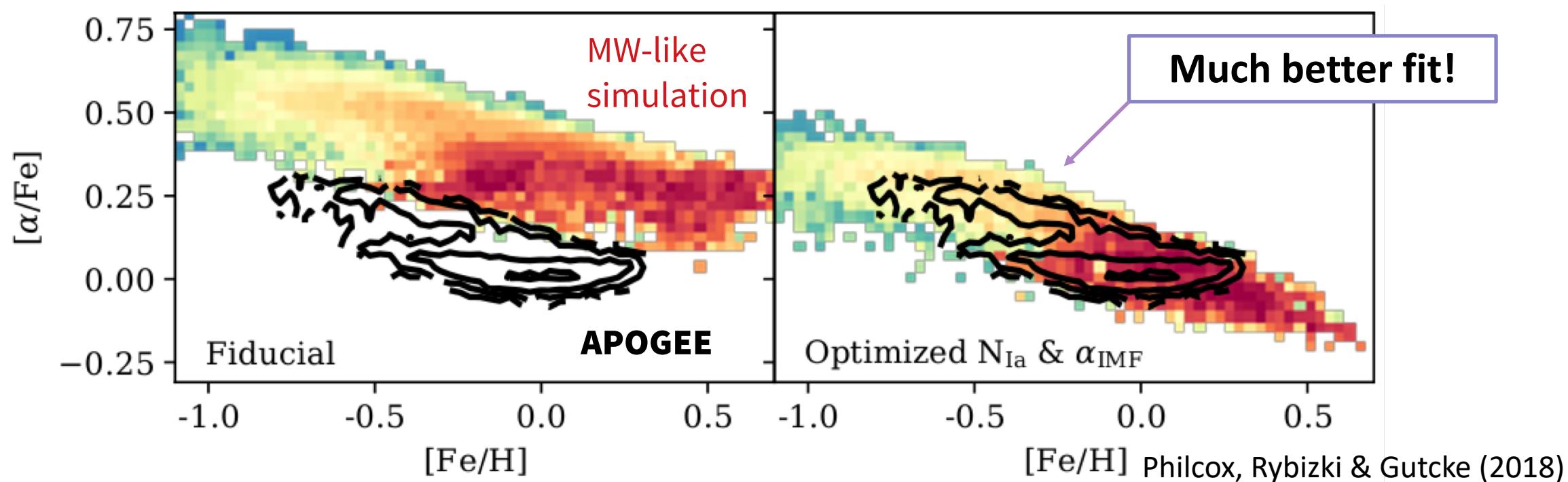
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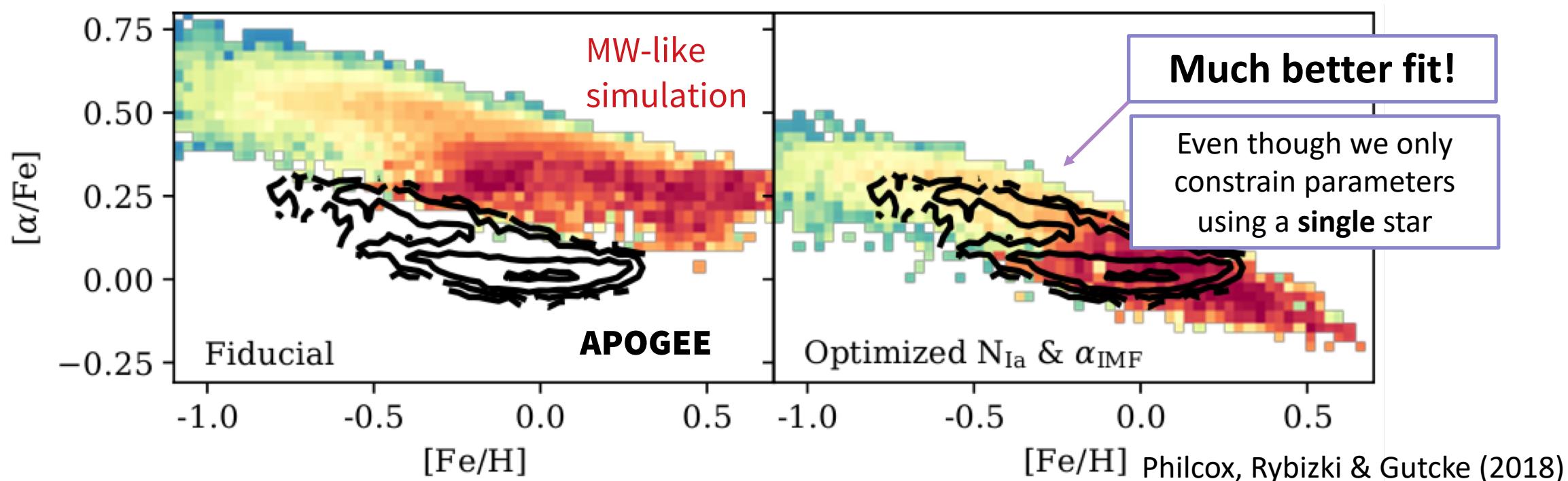
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# How well do our Yield Tables match Observational Data?

# Yield Table Scoring

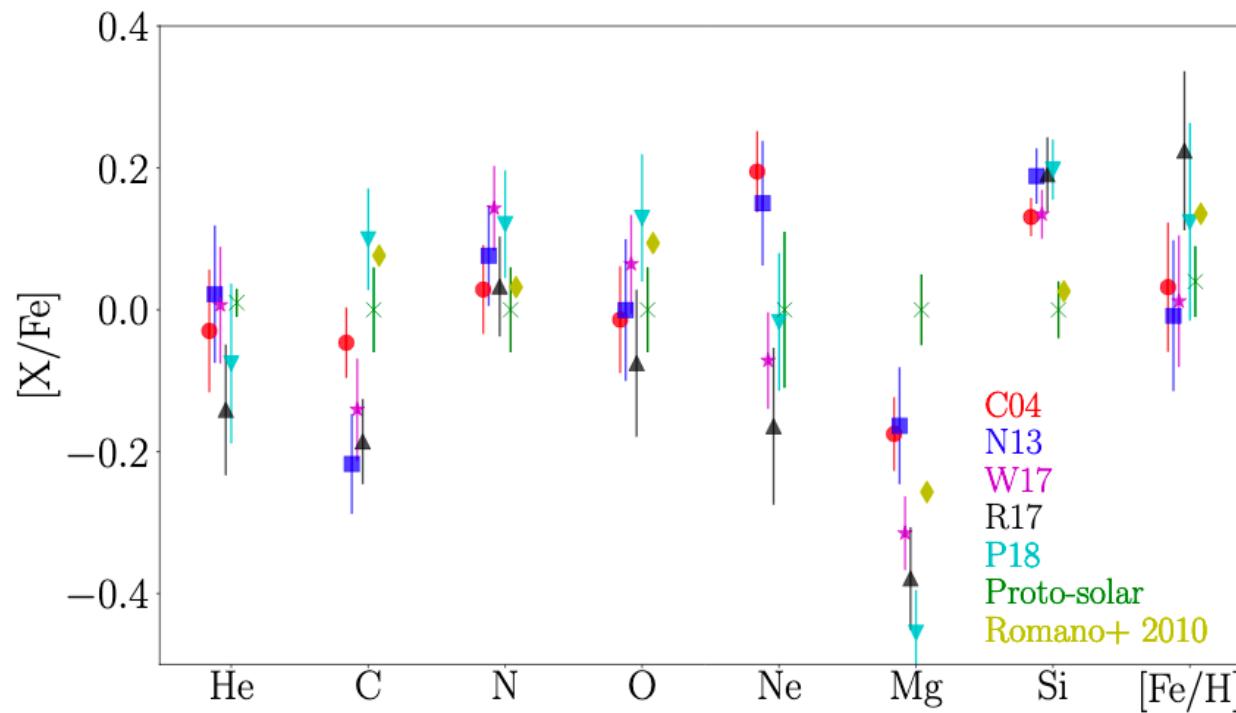
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- ❑ Different Yield Tables predict **very different** proto-Solar abundances

Philcox, Rybizki & Gutcke (2018)

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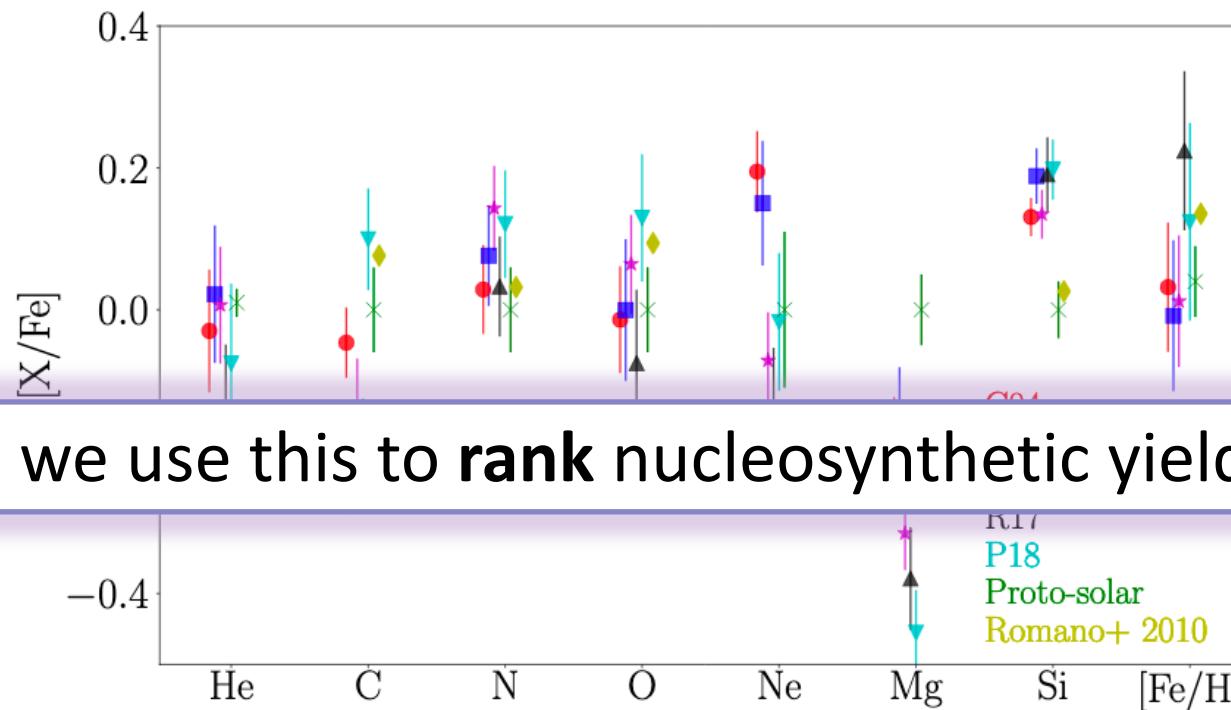
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Philcox, Rybizki & Gutcke (2018)

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- ❑ Method:

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## ❑ Method:

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Philcox, Rybizki & Gutcke (2018)

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## ❑ Cross-Validation

- ❑ Use  $n - 1$  elements to predict  $n$ -th element

Philcox, Rybizki & Gutcke (2018)

# Yield Table Scoring

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## Comparing Core-Collapse Supernovae Yield Tables using 28 Elements

Yield Set	Bayes Score
C04	-1.21
N13	-5.69
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Nikos Prantzos'  
yields perform  
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Philcox, Rybizki & Gutcke (2018)

# Inferring SSP Parameters from Large Stellar Datasets with *Chempy*

# Multi-Star Inference with *Chempy*

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- ❑ How can we extend *Chempy* to run predict chemical abundances for **many** ( $\sim 100$ ) stars?

Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

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Philcox & Rybizki (2019)

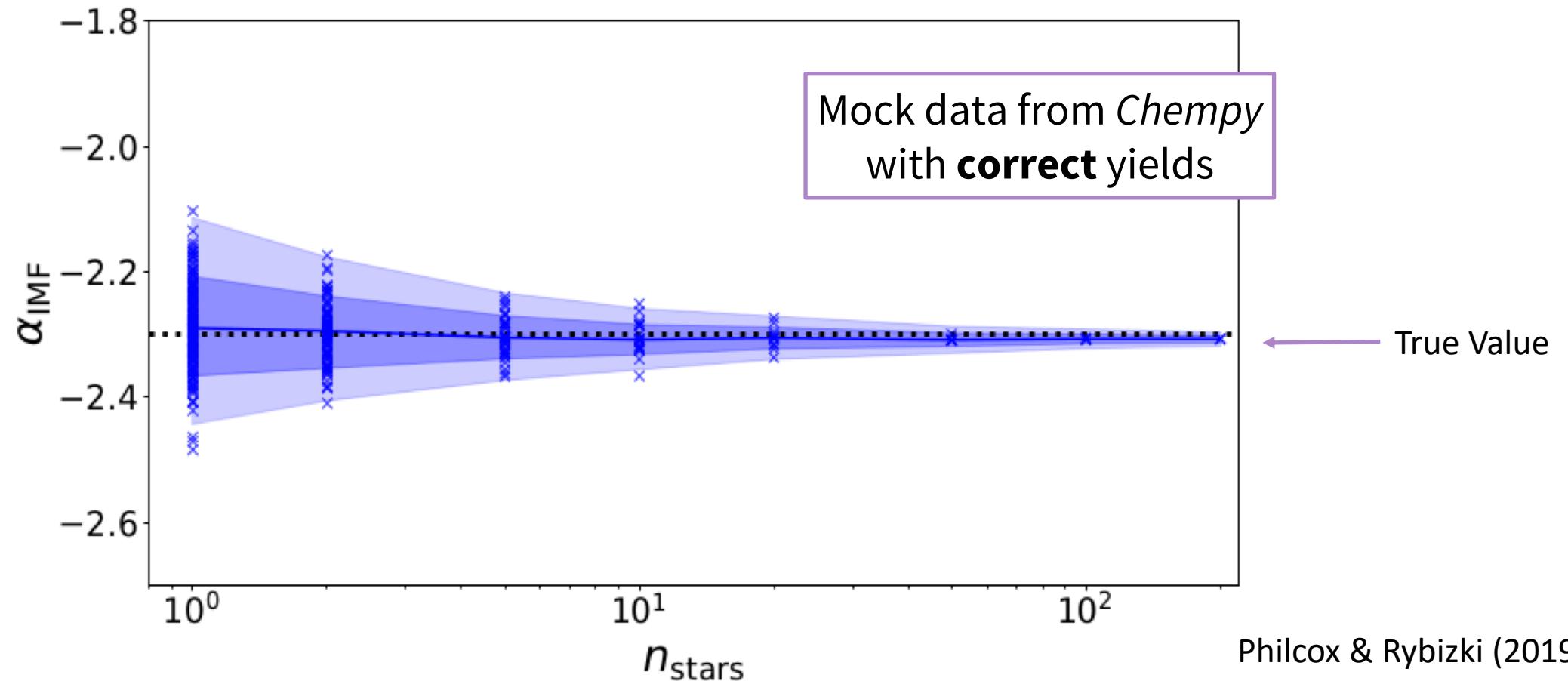
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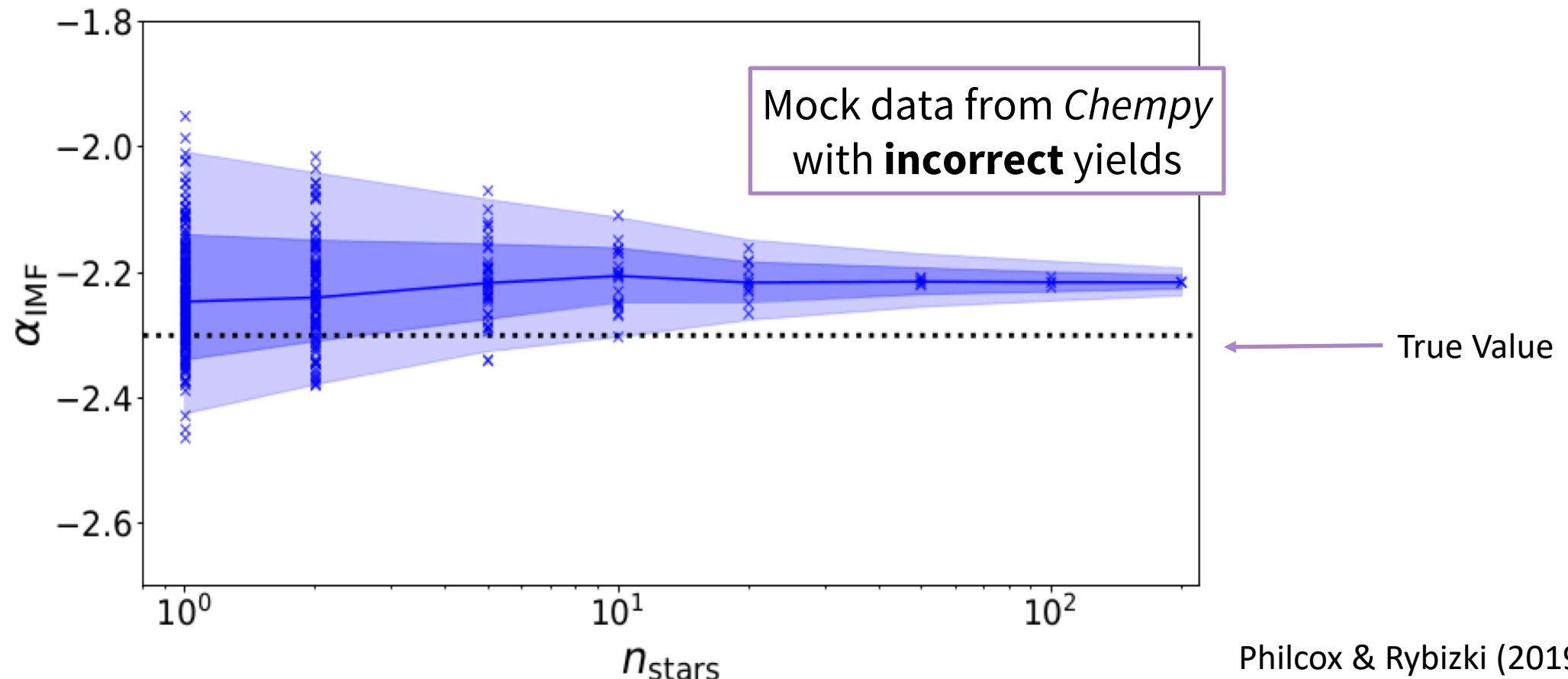
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Philcox & Rybizki (2019)

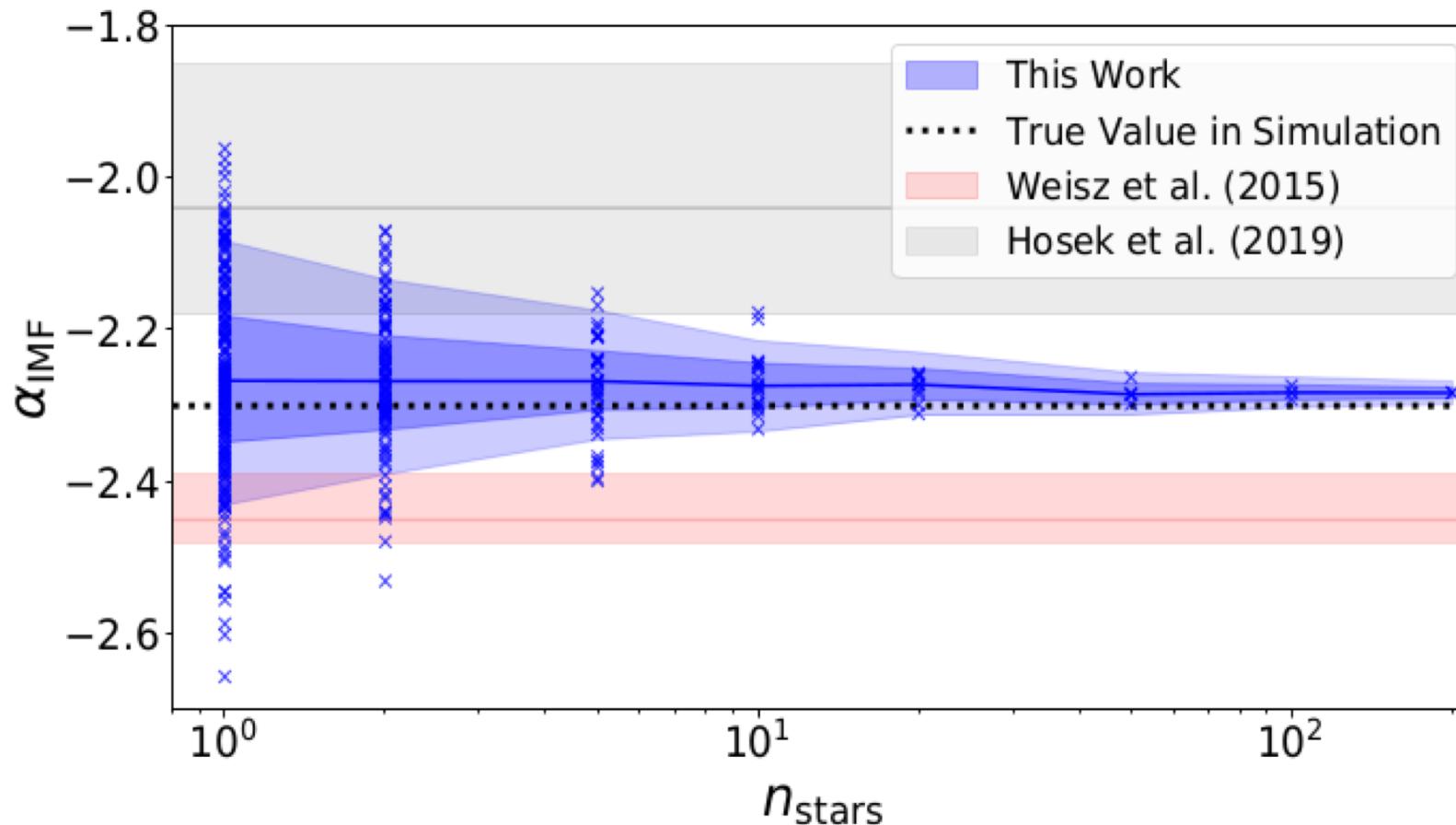
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# Multi-Star Inference with Chempy



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Philcox & Rybizki (2019)

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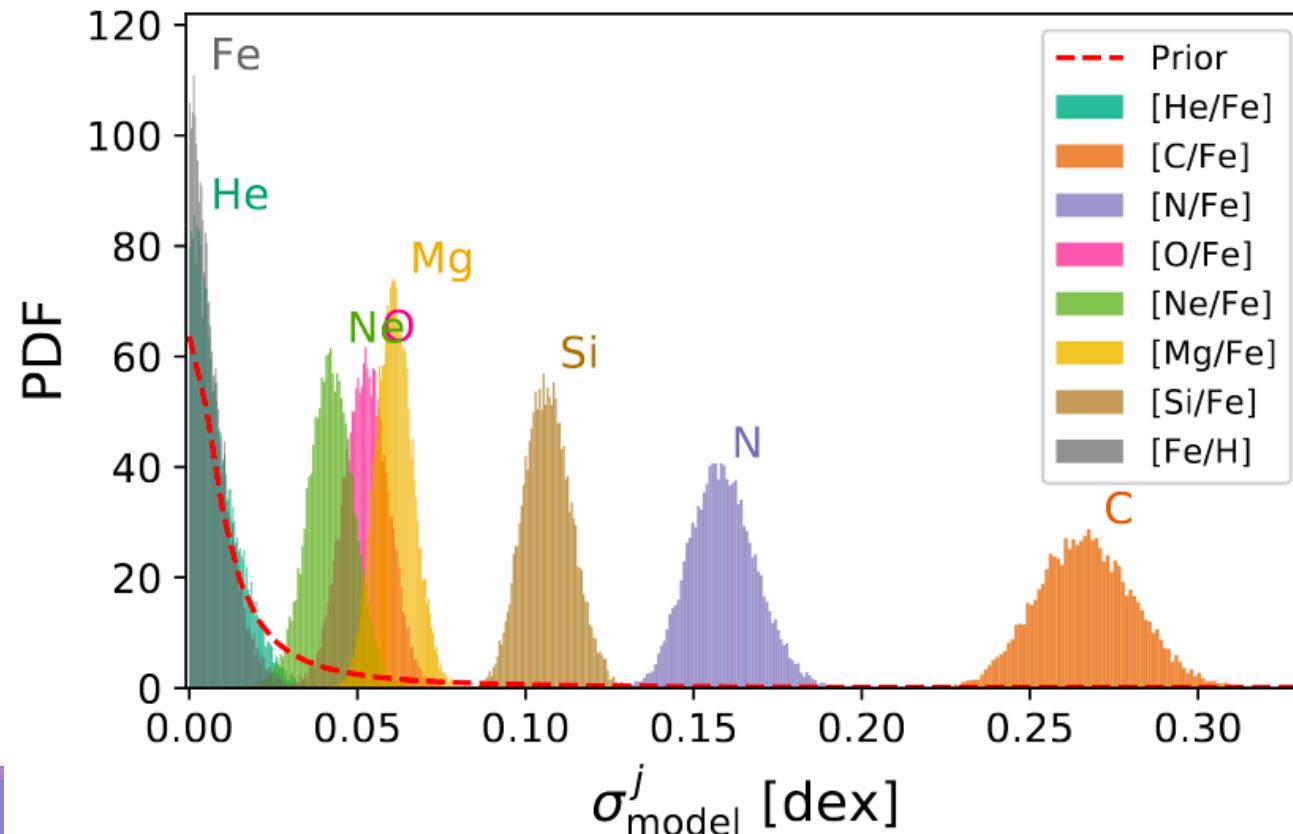
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Model Error  
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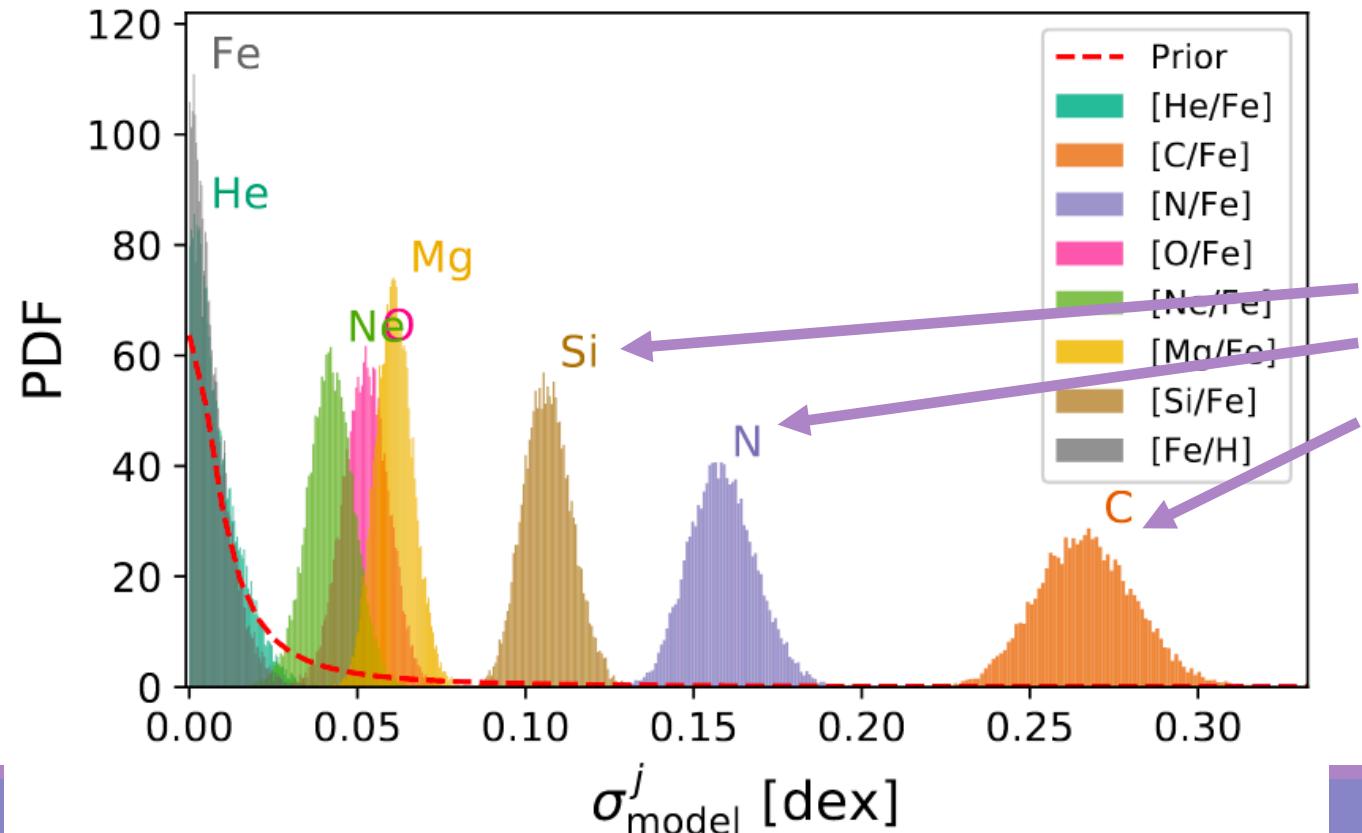


Philcox & Rybizki (2019)

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*Si, N and C are the most discrepant elements between our yield sets!*

Philcox & Rybizki (2019)

# What's Next For *Chempy*?

# Extensions to the *Chempy* model

---

- ❑ *Chempy* can be **easily** extended to incorporate more complex GCE physics.

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- ❑ For example:
  - ❑ A second gas phase (e.g. *hot-cold* gas)
  - ❑ Additional nucleosynthetic processes (e.g. Neutron star mergers, Hypernovae)

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- ❑ For example:
  - ❑ A second gas phase (e.g. *hot-cold* gas)
  - ❑ Additional nucleosynthetic processes (e.g. Neutron star mergers, Hypernovae)
- ❑ Addition of **yield table uncertainties** into the GCE model

# Extensions to the *Chempy* model

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- ❑ For example:
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  - ❑ Additional nucleosynthetic processes (e.g. Neutron star mergers, Hypernovae)
- ❑ Addition of **yield table uncertainties** into the GCE model
- ❑ What do you want to see GCE models incorporating?

# Future Projects

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- ❑ Run the multi-star inference on **real data** to compute Milky Way SSP parameters
- ❑ Parameter optimization of chemical evolution for hydrodynamical simulations including **SSP** parameters and **yield tables**.
- ❑ Infer the **yield tables** empirically from observational data.

# Summary

- ❑ *Chempy* is a **fast** and **flexible** GCE model, allowing for Bayesian exploration of parameter space.
- ❑ Using *Chempy* we can optimize SSP parameters for **hydrodynamical simulations** to ensure they match observational data.
- ❑ By extending to **multiple stars** we can precisely constrain global Galactic parameters
- ❑ And maybe even yield tables...

Thank you  
for your  
attention!

# Additional Slides

# Neural Networks

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# Neural Networks

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$$\begin{aligned} \text{Input} & \xrightarrow{\quad} \mathbf{h} = \mathbf{W}_0 \cdot \mathbf{x} + \mathbf{b}_0 \\ \text{Hidden Layer} & \xrightarrow{\quad} \mathbf{y} = \mathbf{W}_1 \cdot f(\mathbf{h}) + \mathbf{b}_1 \\ \text{Output} & \xrightarrow{\quad} \end{aligned}$$

Non-Linearity

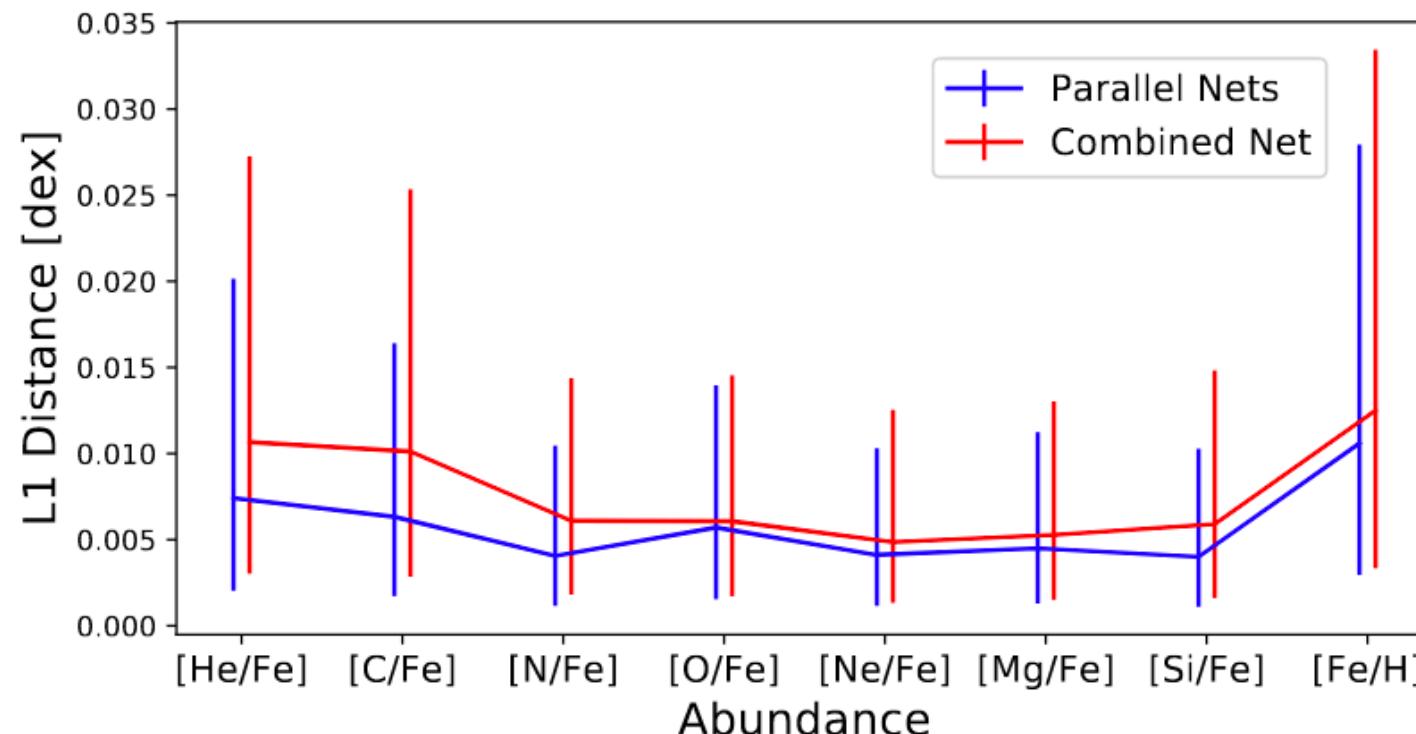
# Neural Networks

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Philcox & Rybizki (2019)

# Hamiltonian Monte Carlo (HMC)

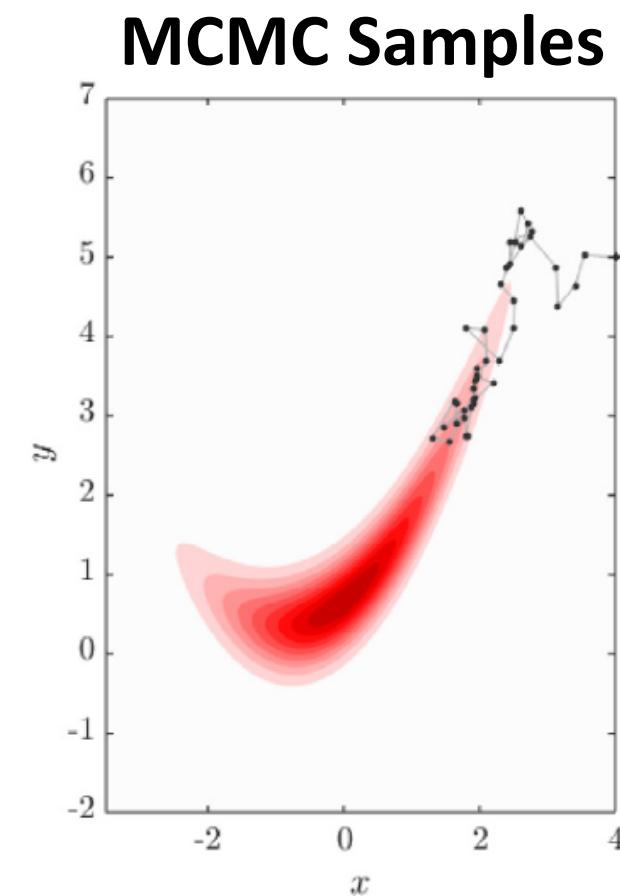
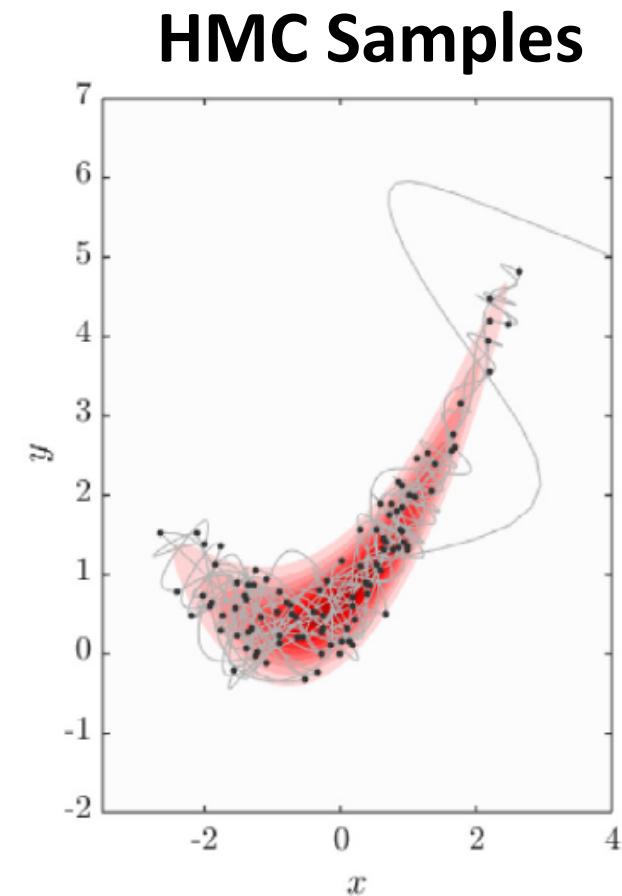
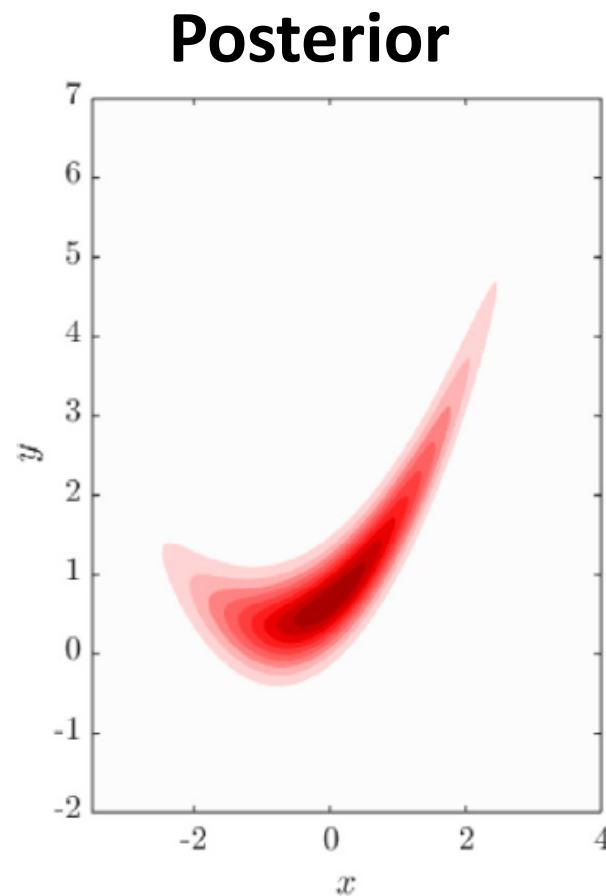
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- Markov Chain Monte Carlo (MCMC) is **slow** and **unsuitable** for high-dimensional problems.
- MCMC works by jumping between points in parameter space at random.
- HMC preferentially samples where the posterior is **large**.
- It's **much more efficient** but requires a **differentiable** model.

Betancourt (2017)

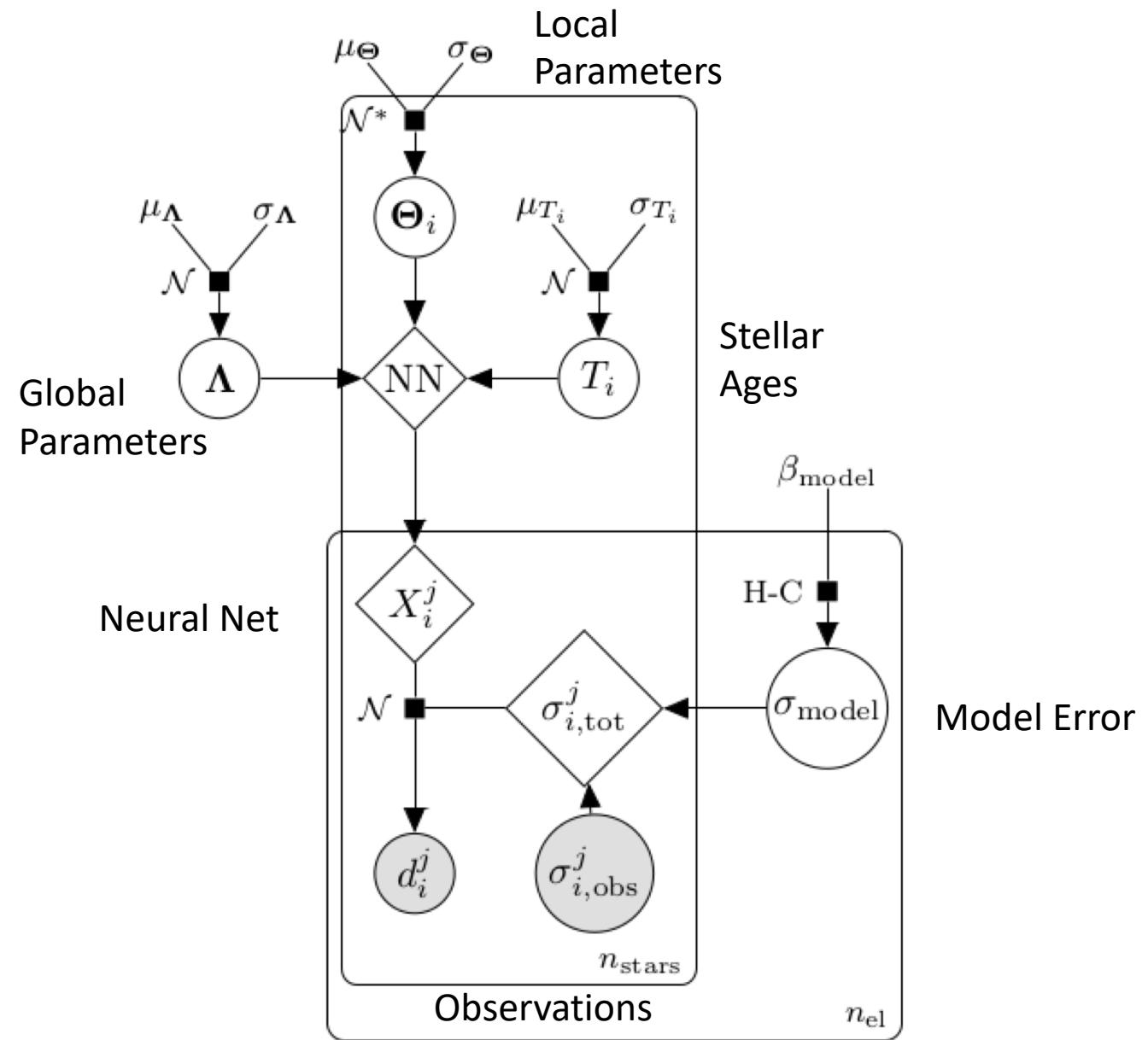
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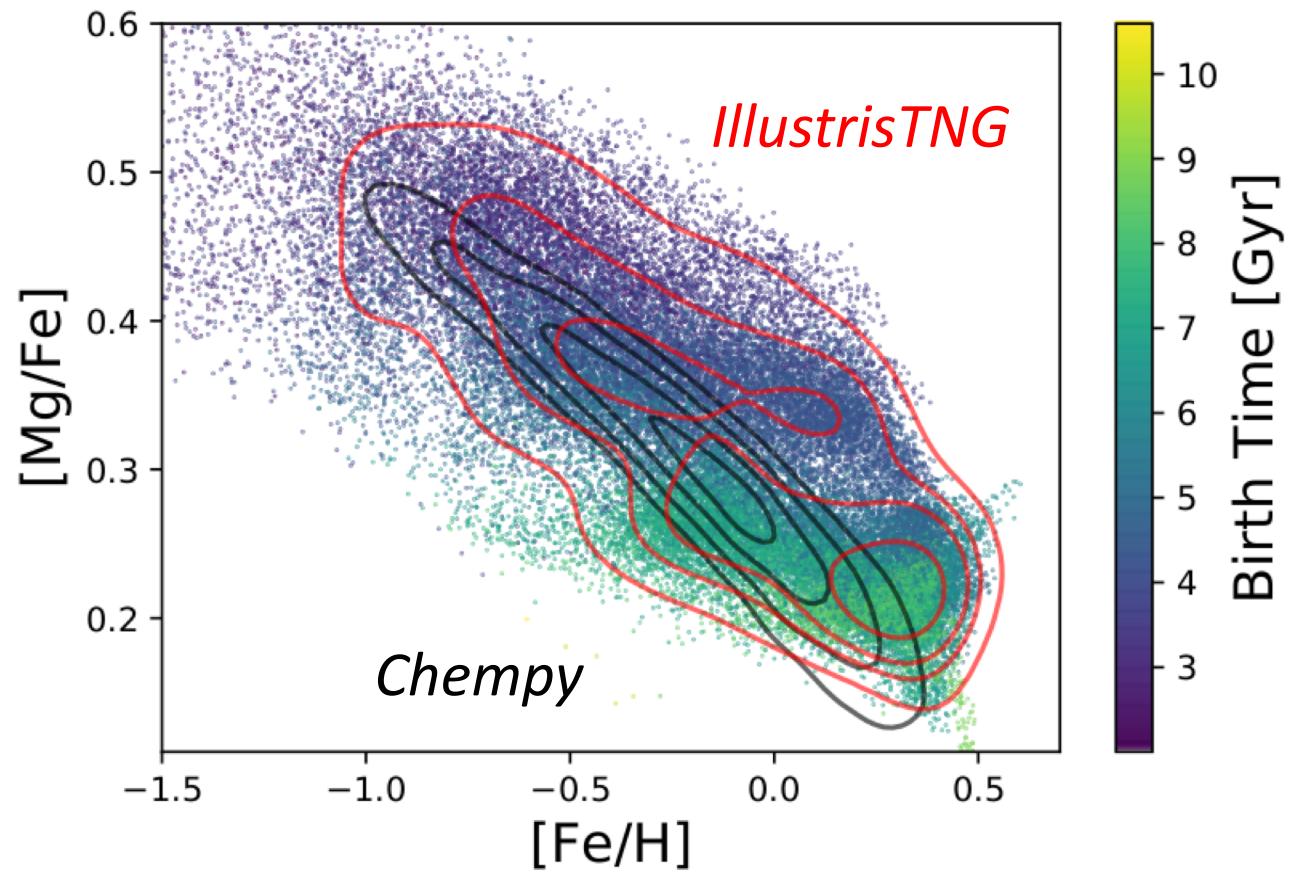


Wang+ (2019)

# *ChempyMulti* Architecture

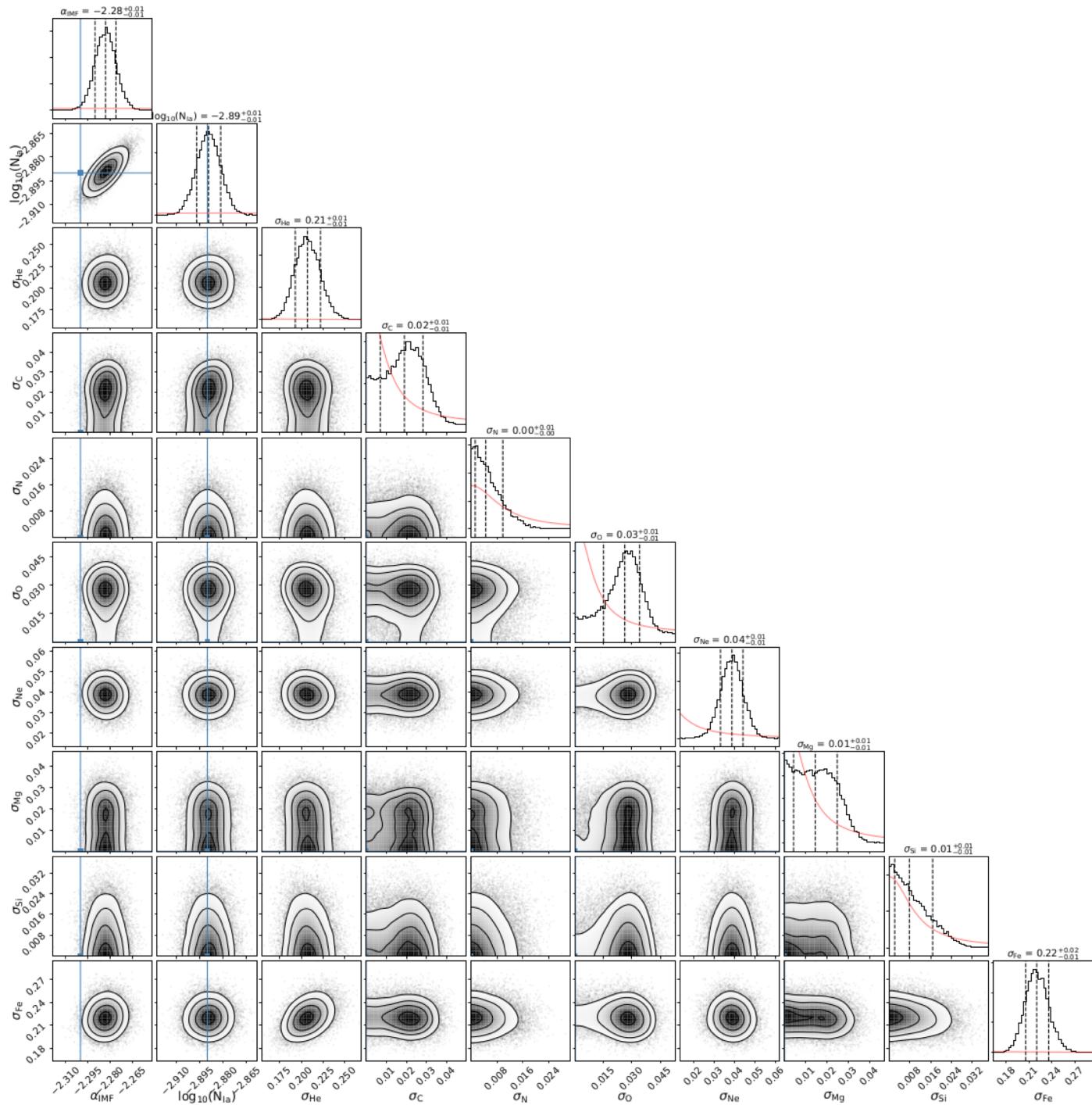


# Abundance Diagrams

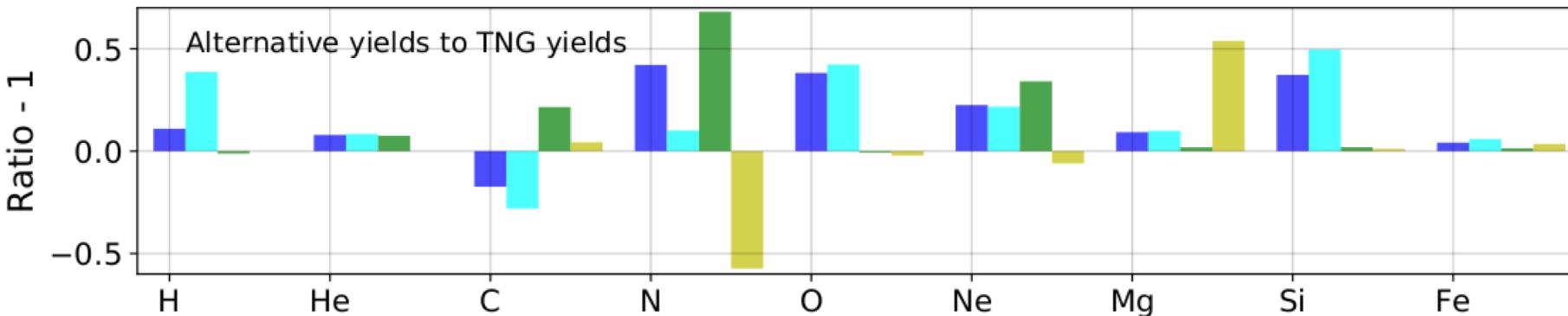
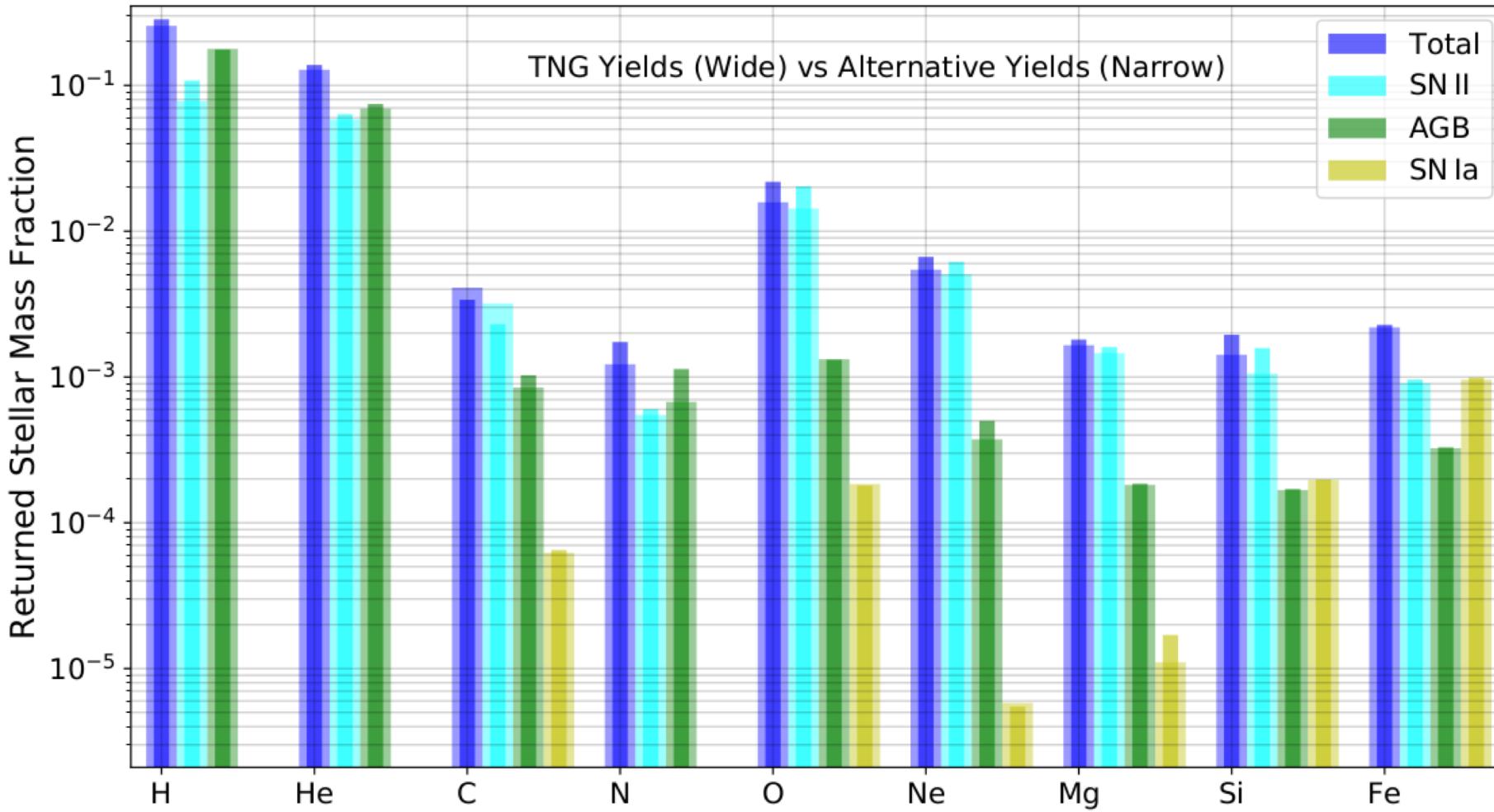


# Full Corner Plot

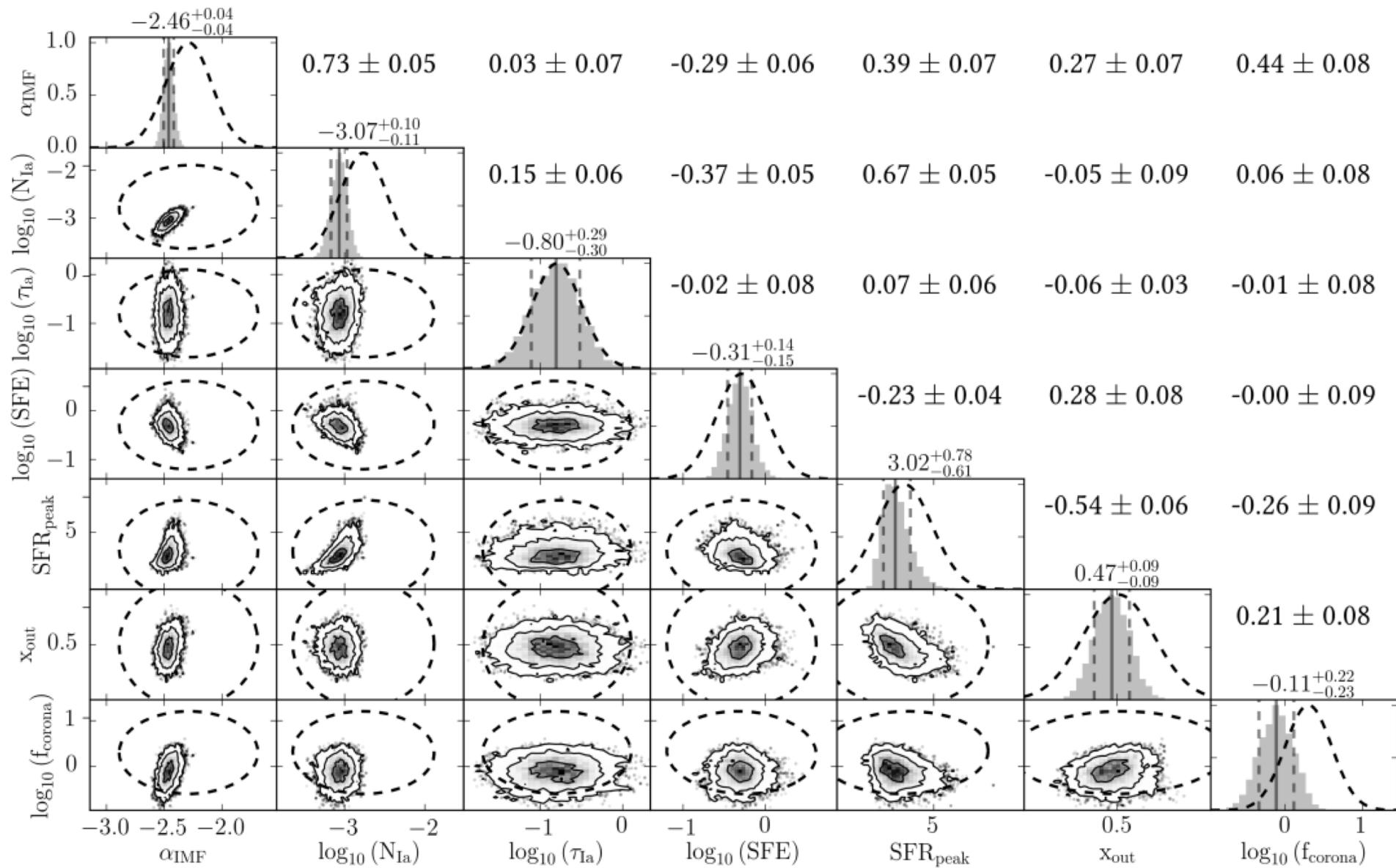
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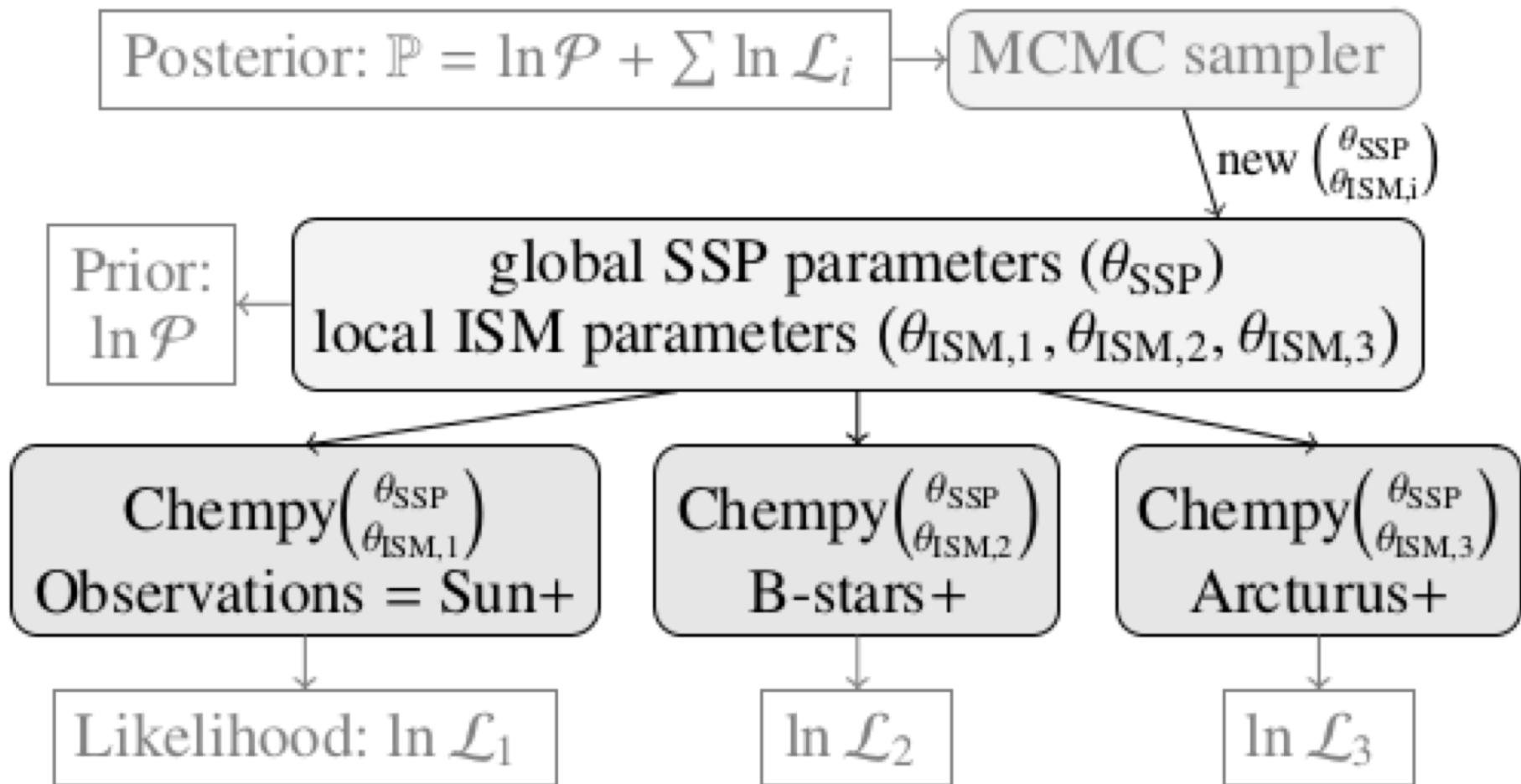
## Default and Alternative Yields



# Single Star Corner Plots



# The *Chempy* Model



# Full Hydrodynamical Simulation Optimization

