## GCE with Chempy:

transforming stellar physics into

elemental abundances and vice versa

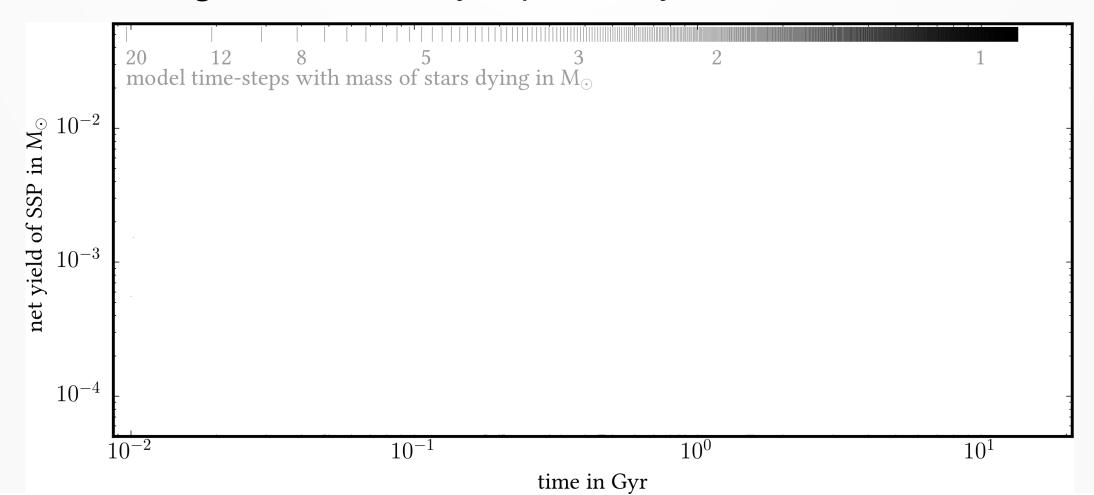
Jan Rybizki – MPIA 11/29/18 – CENAG Heidelberg

## Fast and flexible GCE: Chempy

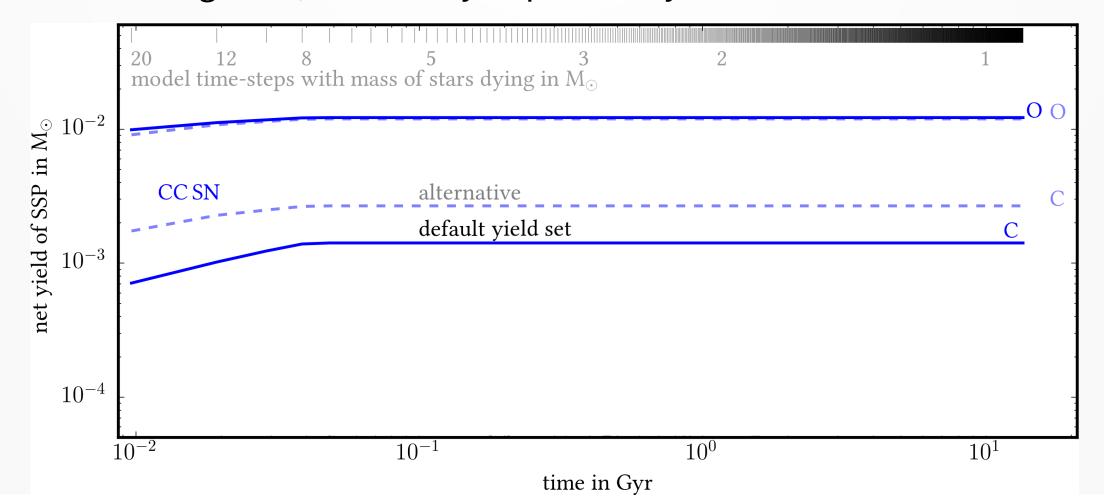
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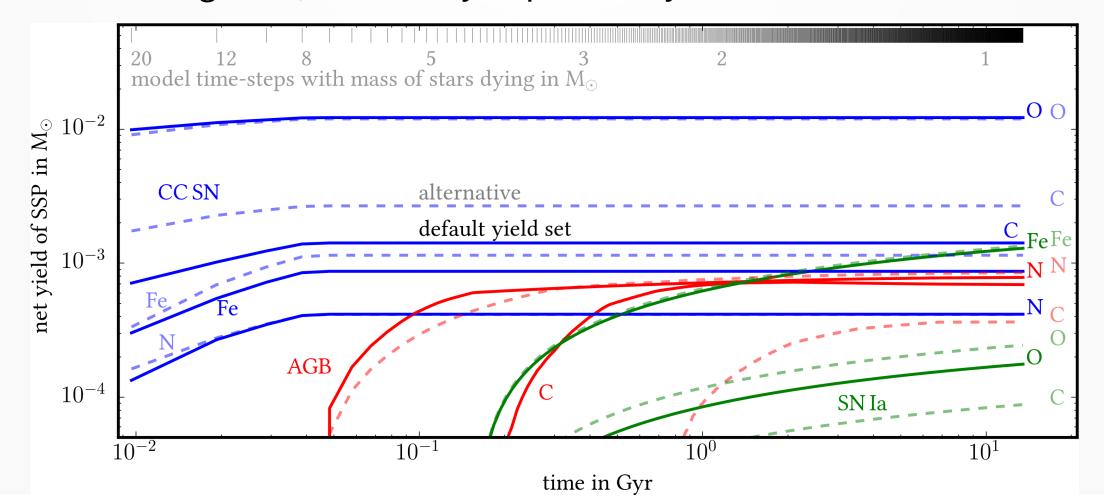
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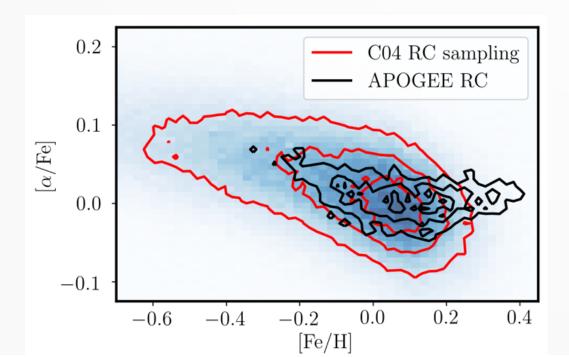
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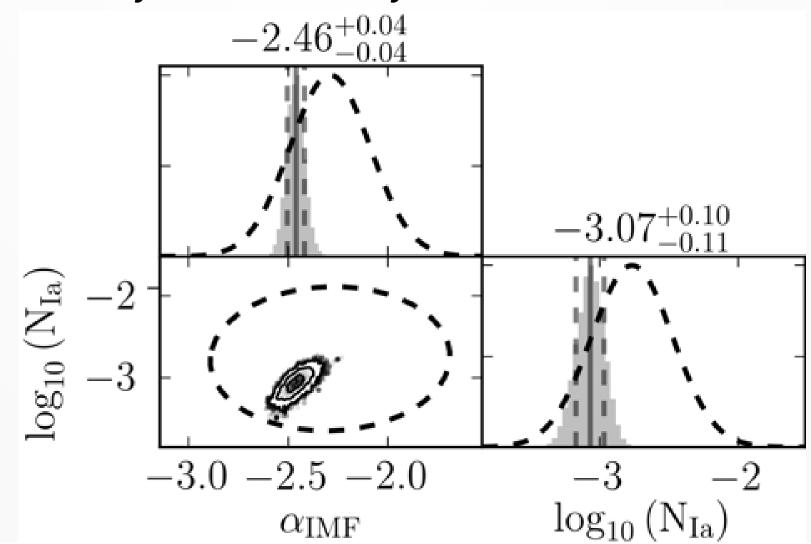


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  - no fiducial model, give your data → get GCE parameters

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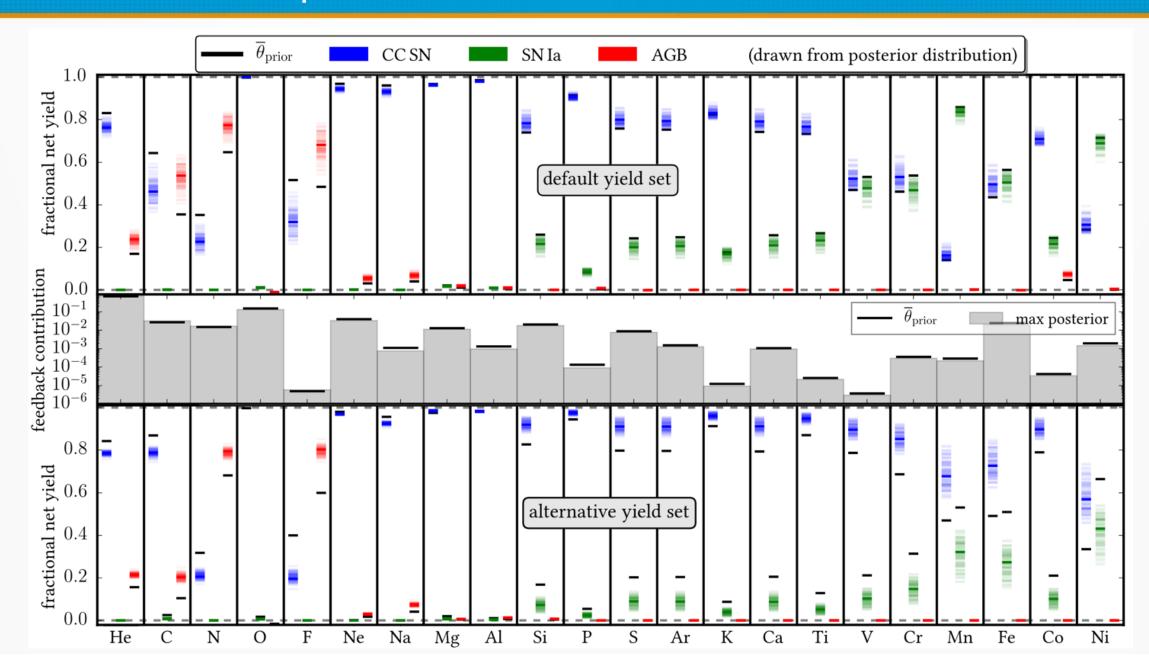
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- Not able to reproduce all elements

### Favorite plot: Channel contribution



Philcox, Rybizki & Gutcke 2018: github.com/oliverphilcox/ChempyScoring
On the Optimal Choice of Yields and IMF-, SNIa-parameter

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|Fe/H

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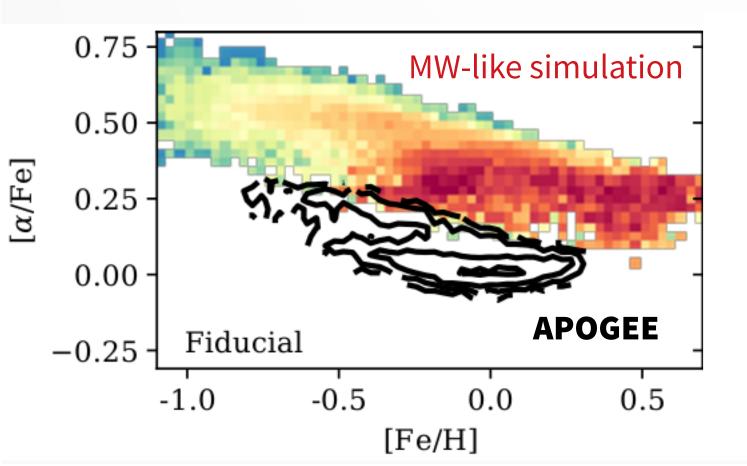
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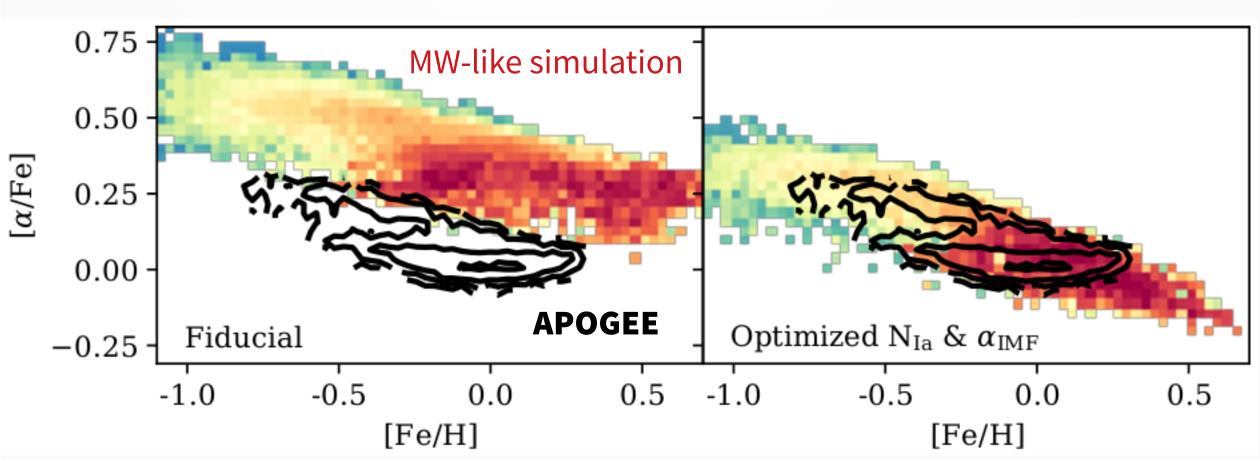
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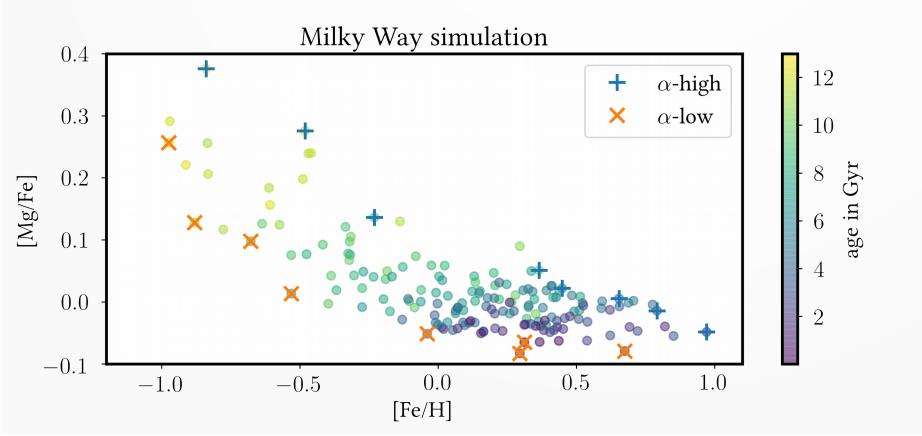


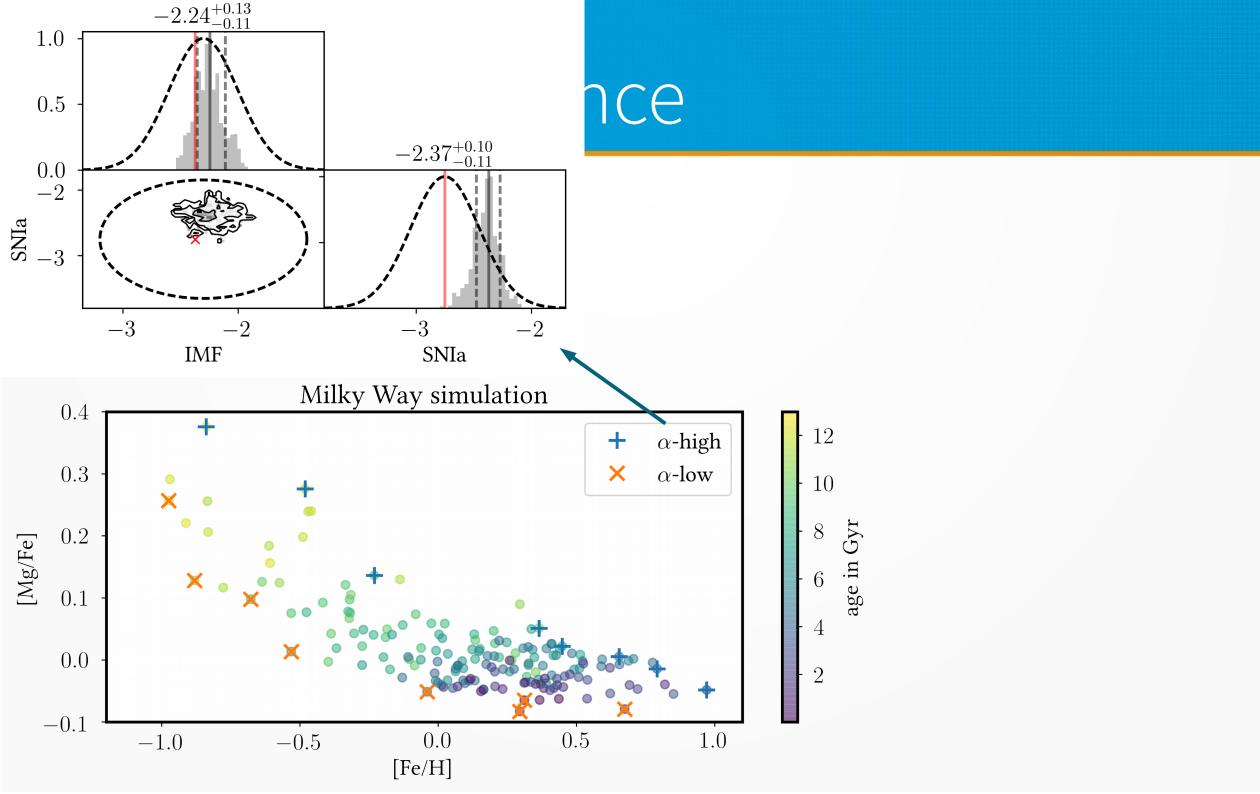
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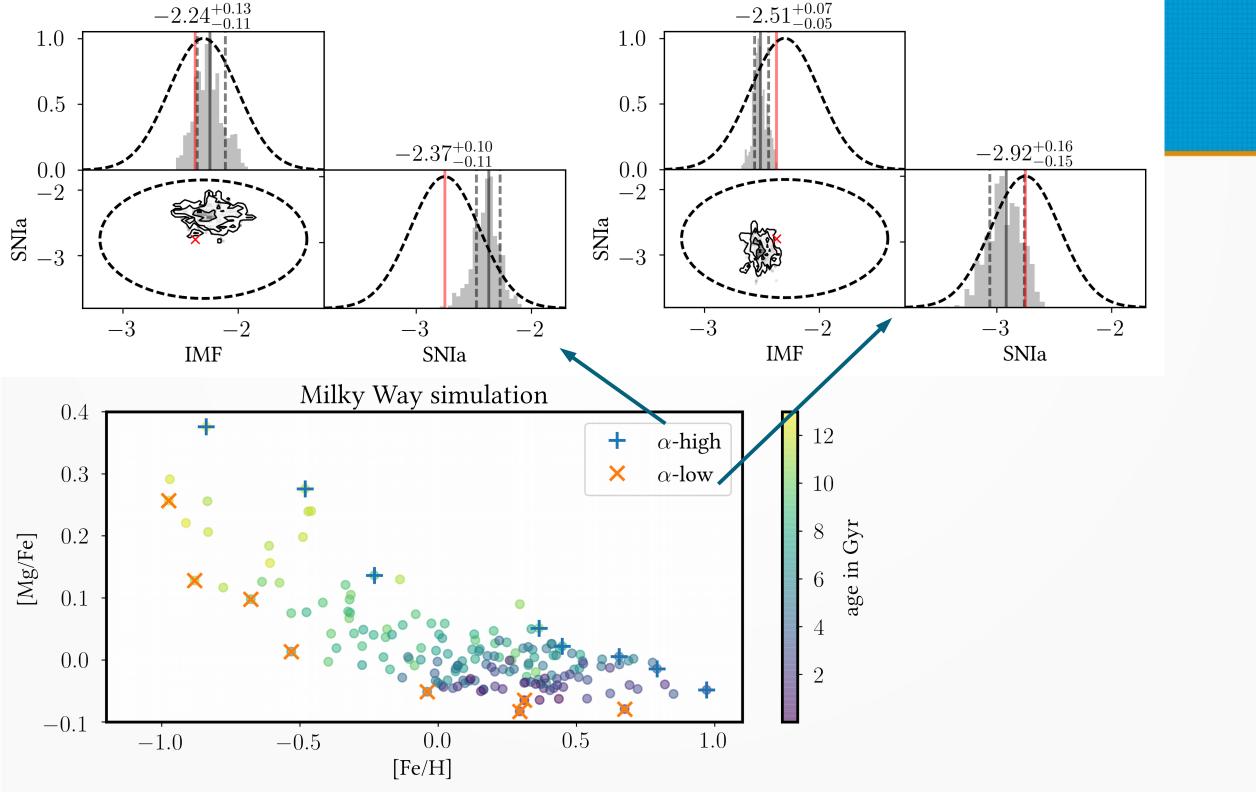
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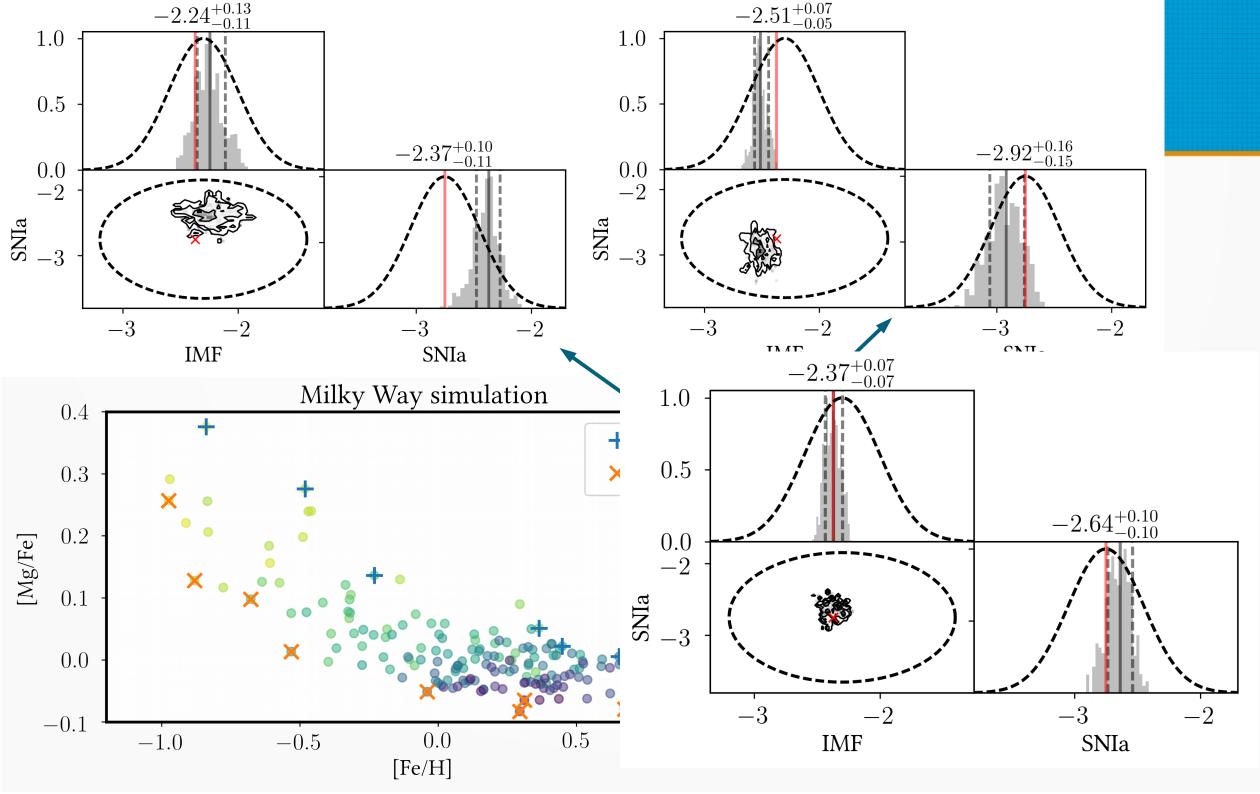


### Many star inference









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  - compare different yield tables
  - infer complex empirical yield tables for different nucleosynthetic channels

#### Thank you for your attention

