

Supplementary Information

Trajectory-dependent Compositional Outcomes in a Prebiotic Reaction Network

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Discussion of parameters chosen in this study

Selection of reaction conditions

Previous work¹ has shown that $[\text{H}_2\text{C}=\text{O}]_{\text{in}}$ controls a compositional transition between 0 – 100 mM (where $[\text{DHA}]_{\text{in}} = 50 \text{ mM}$, $[\text{NaOH}]_{\text{in}} = 30 \text{ mM}$, $[\text{CaCl}_2]_{\text{in}} = 15 \text{ mM}$). We thus selected average input concentrations of formaldehyde, dihydroxyacetone, NaOH and CaCl_2 from within this range which would provide an access to an appropriate range of reaction outcome compositions.

Dynamic inputs of $[\text{Ca}(\text{OH})_2]$ were generated by sampling from normal distributions ($\mu = 15 \text{ mM}$, $\sigma = 2.89 \text{ mM}$ or 5.75 mM) generated using Numpy² (1.22.2.) with function `numpy.random.normal()`.

Selection of $\text{Ca}(\text{OH})_2$ amplitude series

For the amplitude series $[\text{NaOH}]_{\text{in}}$ and $[\text{CaCl}_2]_{\text{in}}$ were cycled around the steady state average with irregular amplitude sampled from a normal distribution ($\mu_{\text{in}} = 15 \text{ mM}$, $\sigma_{\text{in}} = 2.89 \text{ mM}$ or 5.75 mM) of flow rates. The flow rates were switched every 45 seconds, thus not allowing the catalyst to reach a new steady state concentration in the reactor and maintain a gradient change. We did not allow the NaOH and CaCl_2 syringes to reach a negative flow rate. By sampling around the steady state of $1541.25 \mu\text{L h}^{-1}$ for both syringes, we attained a maximum standard deviation of $\sigma = 5.75 \text{ mM}$ into the reactor. In the amplitude series the other amplitude was chosen to have half the maximum standard deviation of the inflow rate ($\sigma = 2.89 \text{ mM}$).

Selection of multiple frequency input signal

We wanted to construct an input signal with different embedded frequencies to study the sensitivity of key reactions to different time scales of $\text{Ca}(\text{OH})_2$ fluctuations. For the input

frequencies we wanted the time scales to be significantly different. However, we wanted to avoid going to steady state (rate > 120 s) or filtering out the amplitude as a consequence of the residence time in the reactor (rate $\rightarrow 0$ s). From a theoretical sinusoidal input oscillation at a 120 second period (comparable to rate = 120 s) the reactor amplitude decays 75 % with respect to a 60 second period (comparable to rate = 30 s). Therefore, we chose not to modulate with a rate higher than 30 seconds (60 second period / 2). The 60 second rate of change was chosen since it was double the short 30 s rate of change and half the 120 s rate of change.

Supplementary figures

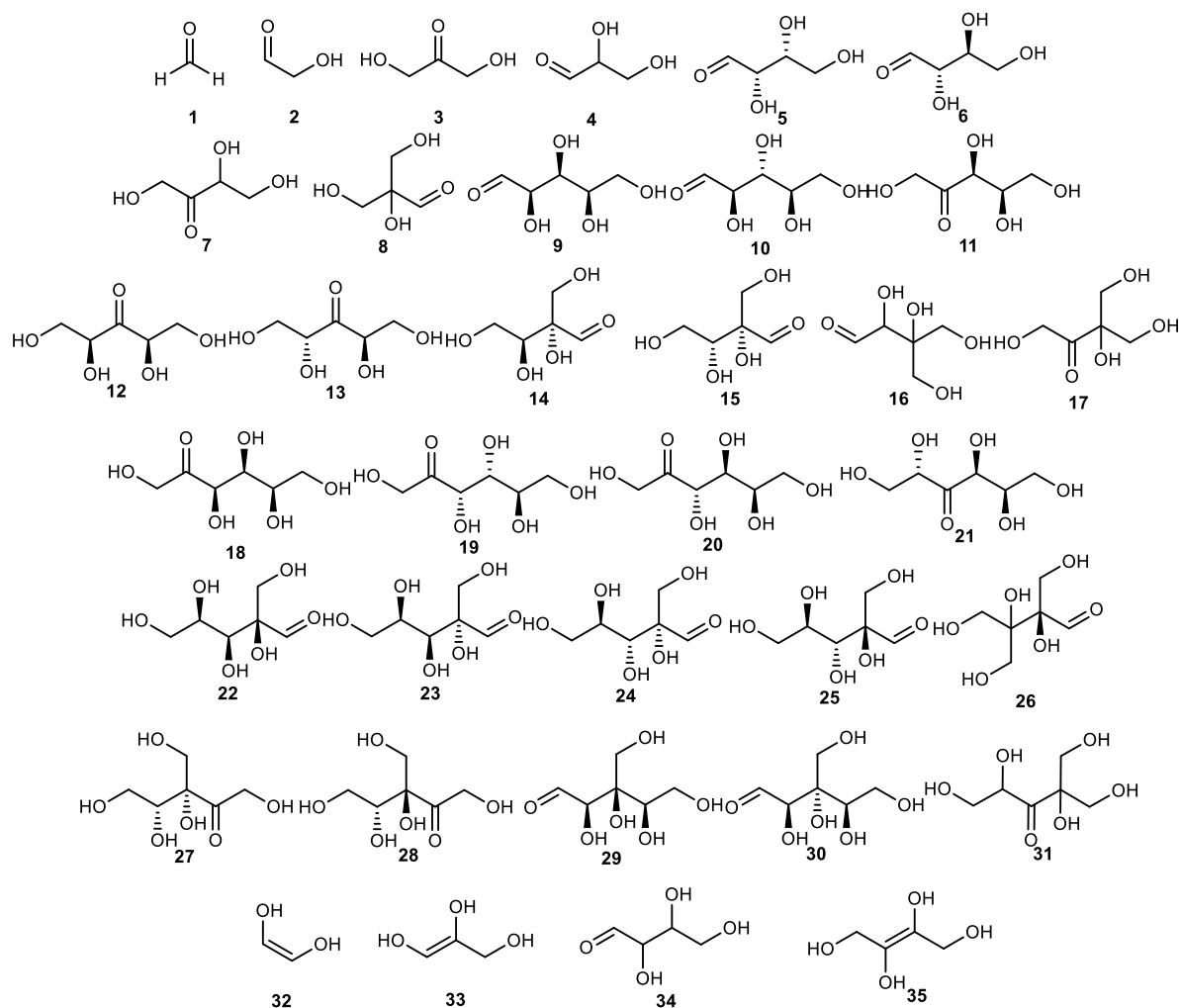


Figure S1. Structures of the compounds discussed in this work, with the corresponding numbering scheme (1 – 35).

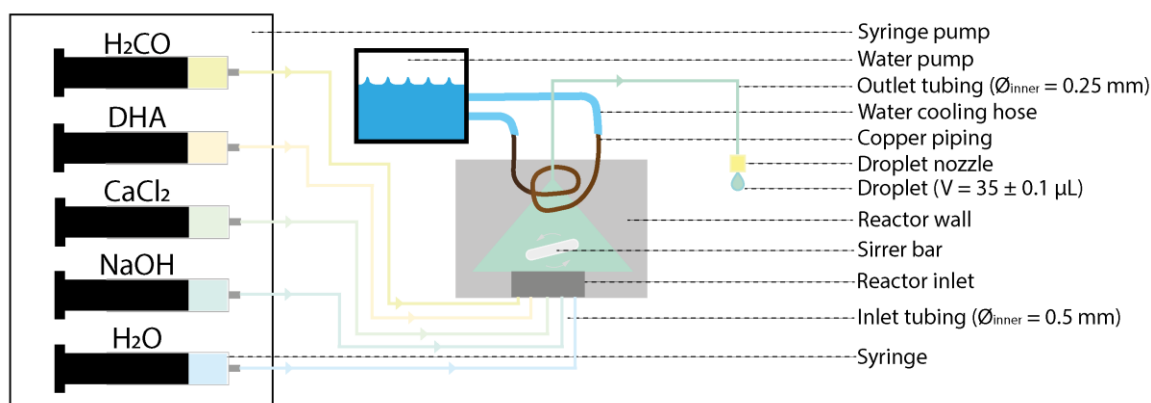


Figure S2. A detailed schematic representation of the flow reactor setup. The scheme indicates how five inlet syringes were connected to the bottom of a reactor constructed from polydimethylsiloxane (side view). The cone shaped reactor was temperature controlled and had the outlet at the top, which was connected to a droplet nozzle.

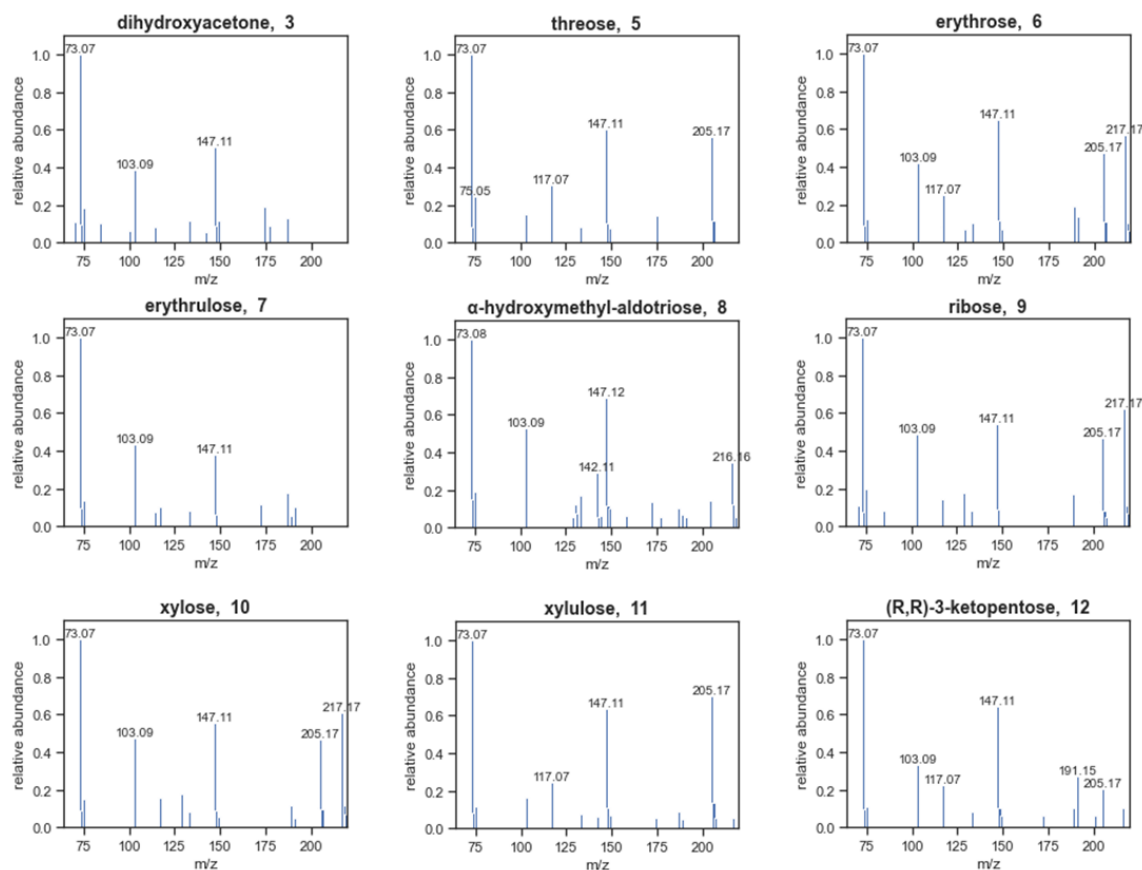


Figure S3. Mass spectra from GC-MS analysis of DHA (3), threose (5), erythrose (6), erythrulose (7), α -hydroxymethyl-aldotriose (8), ribose (9), xylose (10), xylulose (11) and (R,R)-3-ketopentose (12).

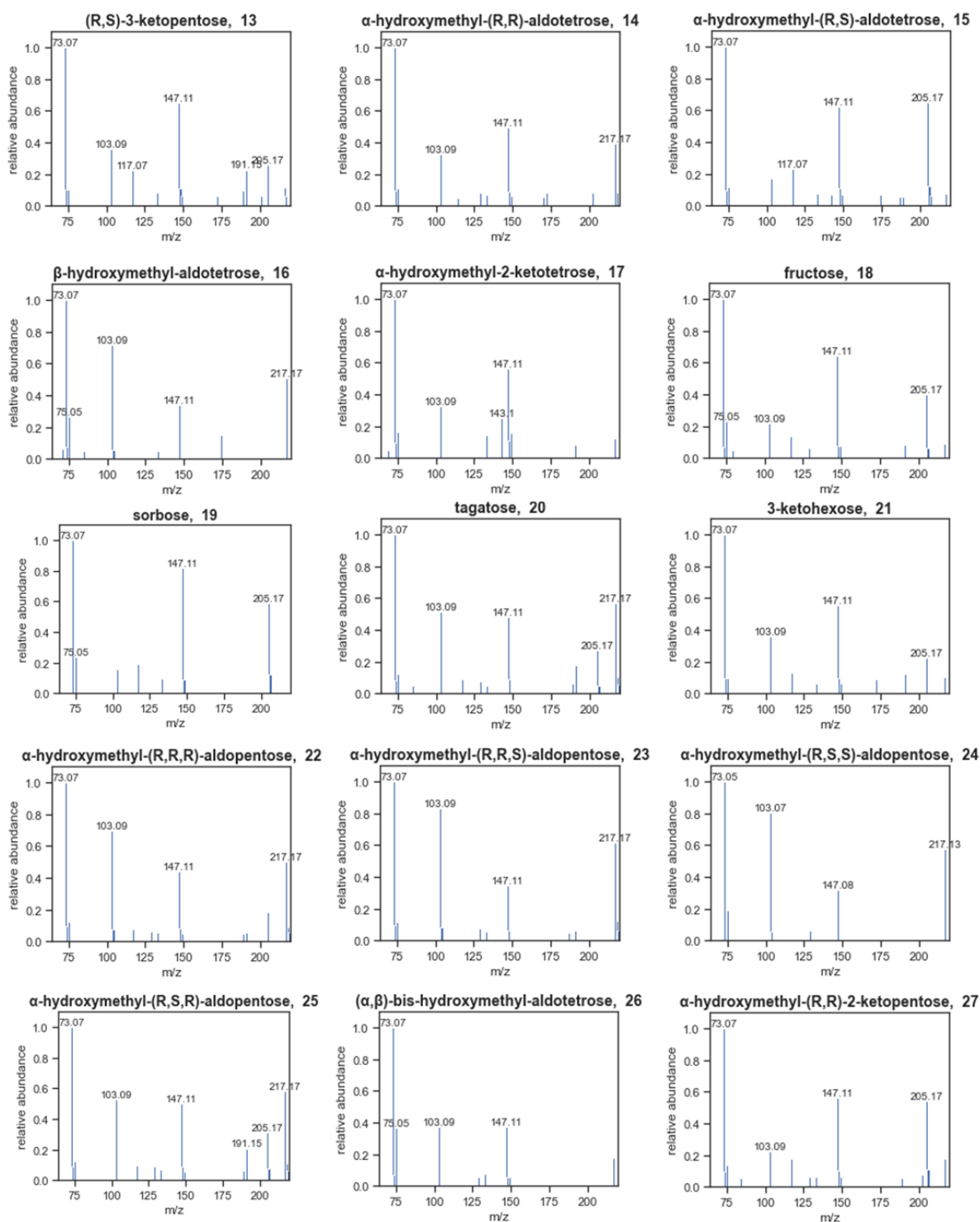


Figure S4. Mass spectra from GC-MS analysis for (R,S)-3-ketopentose (13), α -hydroxymethyl-(R,R)-aldotetrose (14), α -hydroxymethyl-(R,S)-aldotetrose (15), β -hydroxymethyl-aldotetrose (16), α -hydroxymethyl-2-ketotetrose (17), fructose (18), sorbose (19), tagatose (20), 3-ketohexose (21), α -hydroxymethyl-(R,R,R)-aldopentose (22), α -hydroxymethyl-(R,R,S)-aldopentose (23), α -hydroxymethyl-(R,S,S)-aldopentose (24), α -hydroxymethyl-(R,S,R)-aldopentose (25), (α,β)-bis-hydroxymethyl-aldotetrose (26) and α -hydroxymethyl-(R,R)-2-ketopentose (27).

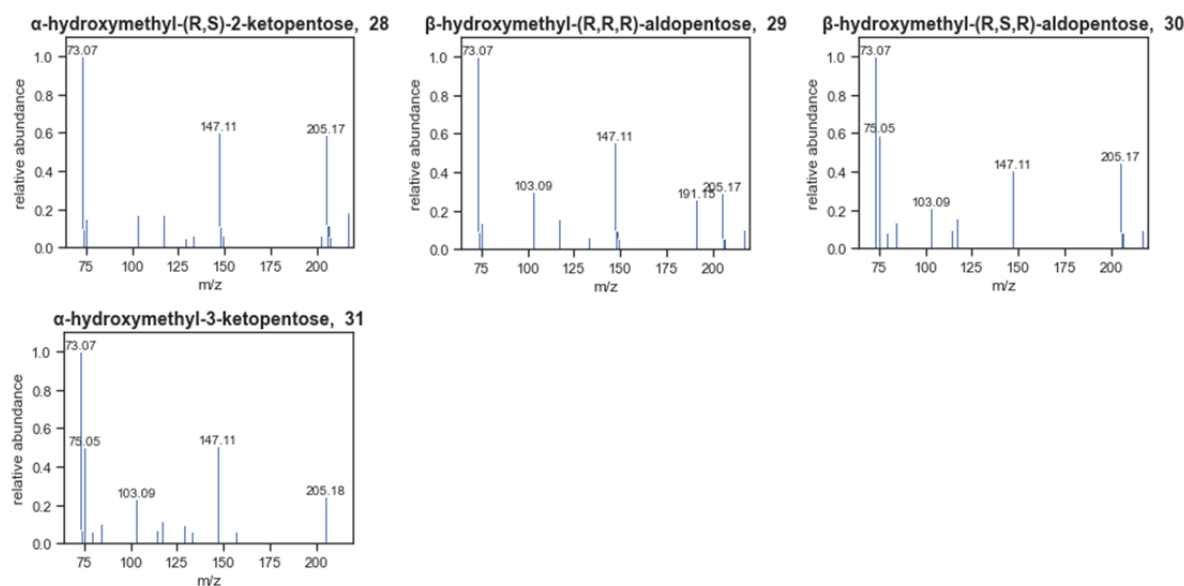


Figure S5. Mass spectra from GC-MS analysis for hydroxymethyl-(R,S)-2-ketopentose (**28**), β -hydroxymethyl-(R,R,R)-aldopentose (**29**), β -hydroxymethyl-(R,S,R)-aldopentose (**30**) and α -hydroxymethyl-3-ketopentose (**31**).

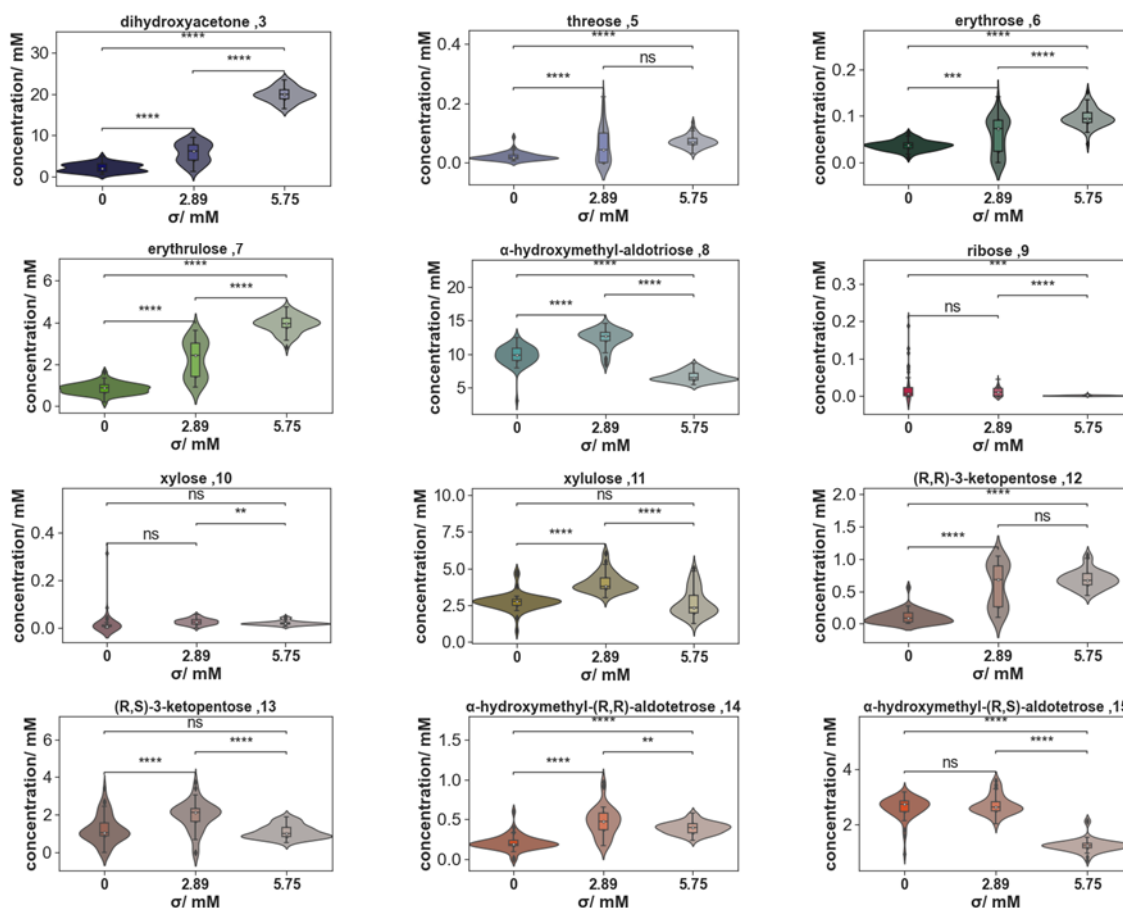


Figure S6. Violin plots of concentration distributions for identified compounds (**3**, **5**, **6**, **7**, **8**, **9**, **10**, **11**, **12**, **13**, **14**, **15**) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 20 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

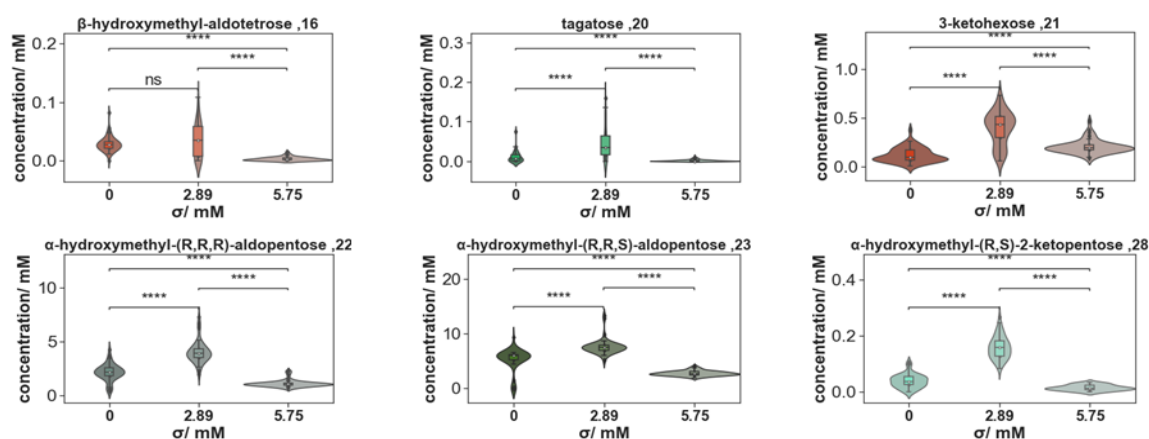


Figure S7. Violin plots for concentration profiles for the identified compounds (16, 20, 21, 22, 23, 28) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 20 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

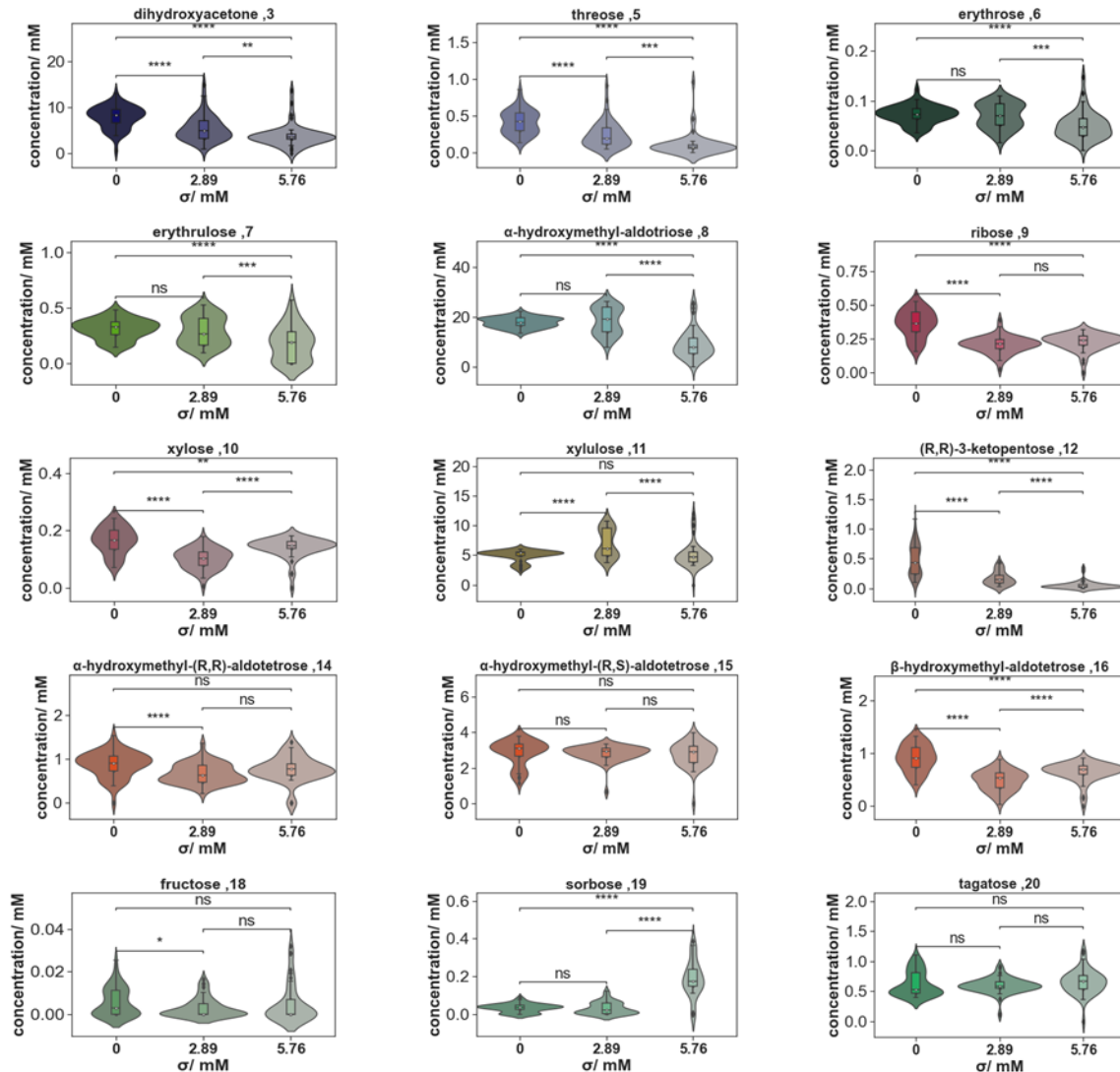


Figure S8. Violin plots of the concentration distributions for identified compounds (3, 5, 6, 7, 8, 9, 10, 11, 12, 14, 15, 16, 18, 19, 20) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 50 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

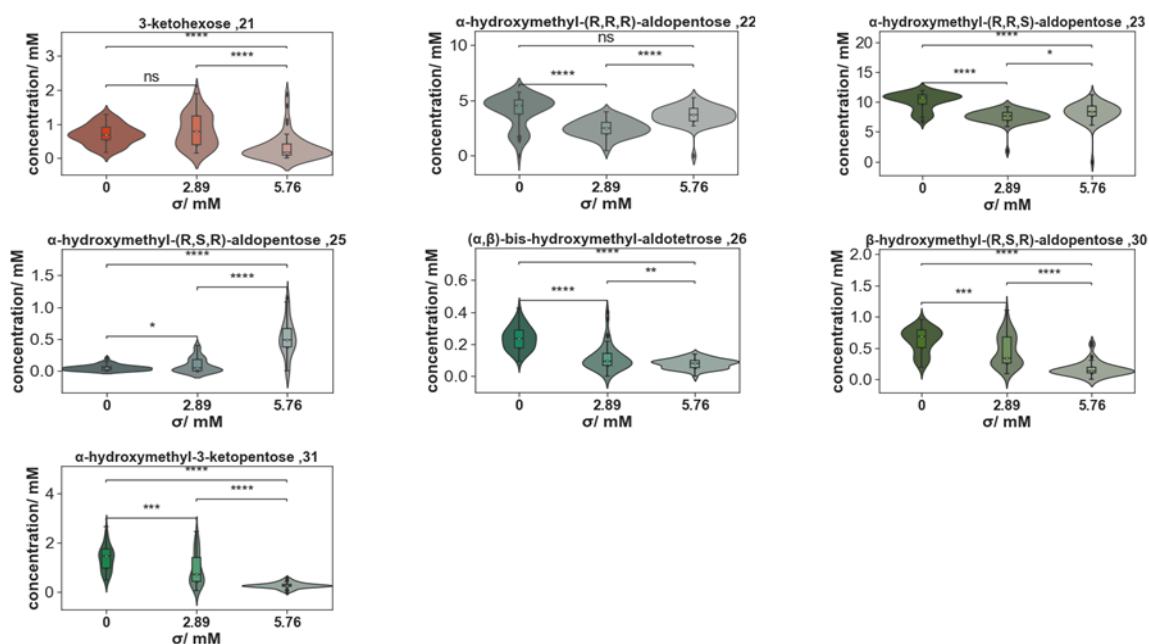


Figure S9. Violin plots for concentration distributions of the identified compounds (21, 22, 23, 25, 26, 30, 31) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 50 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

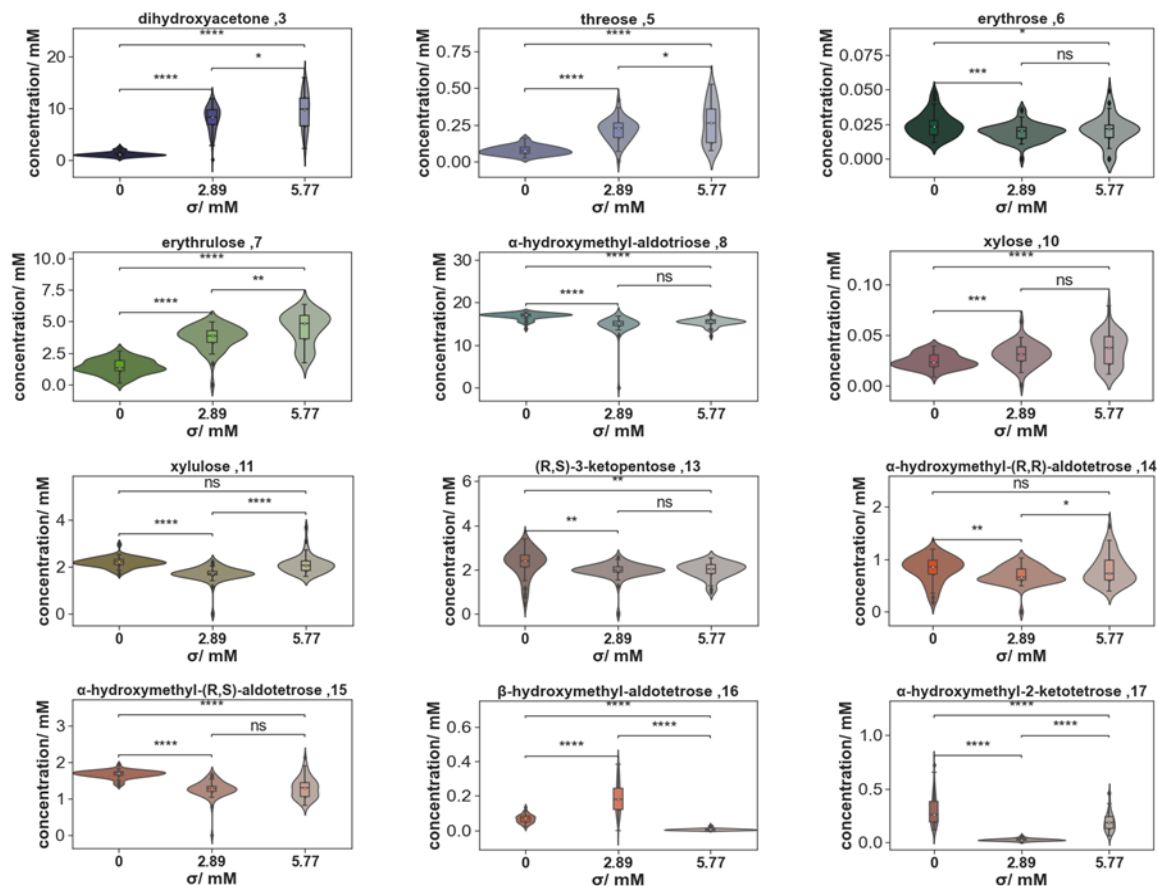


Figure S10. Violin plots for concentration distributions for the identified compounds (3, 5, 6, 7, 8, 10, 11, 13, 14, 15, 16, 17) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 100 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

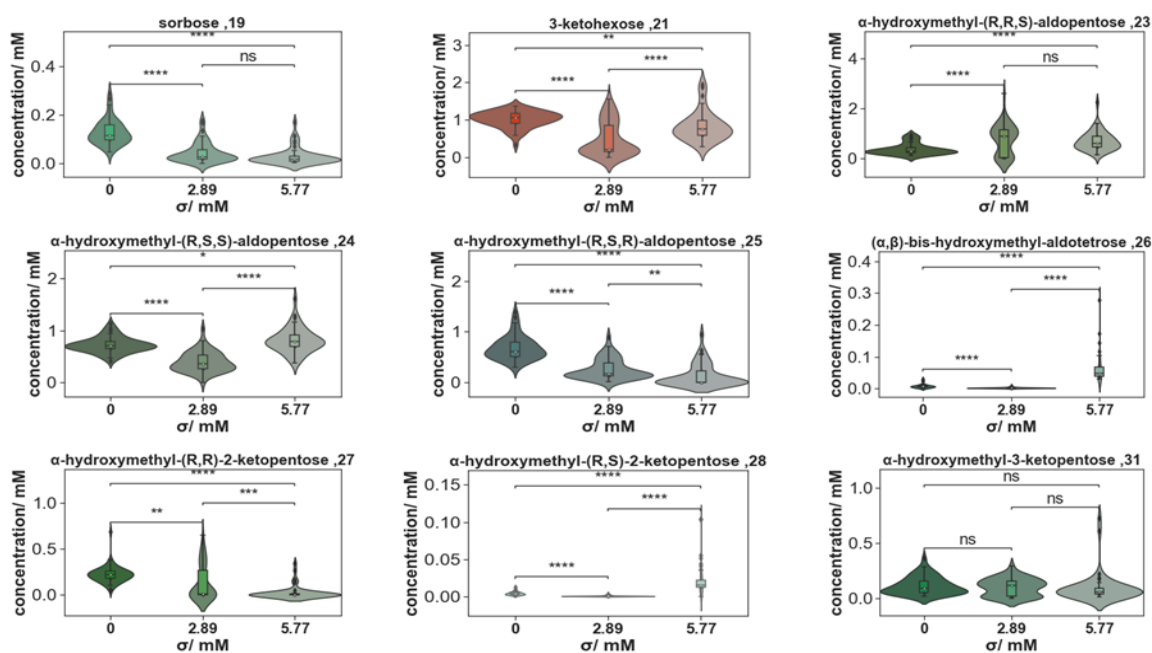


Figure S11. Violin plots for concentration distributions for the identified compounds (19, 21, 23, 24, 25, 26, 27, 28, 31) at steady state, low amplitude and high amplitude ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 2.89$ and 5.75 mM). Conditions: 100 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

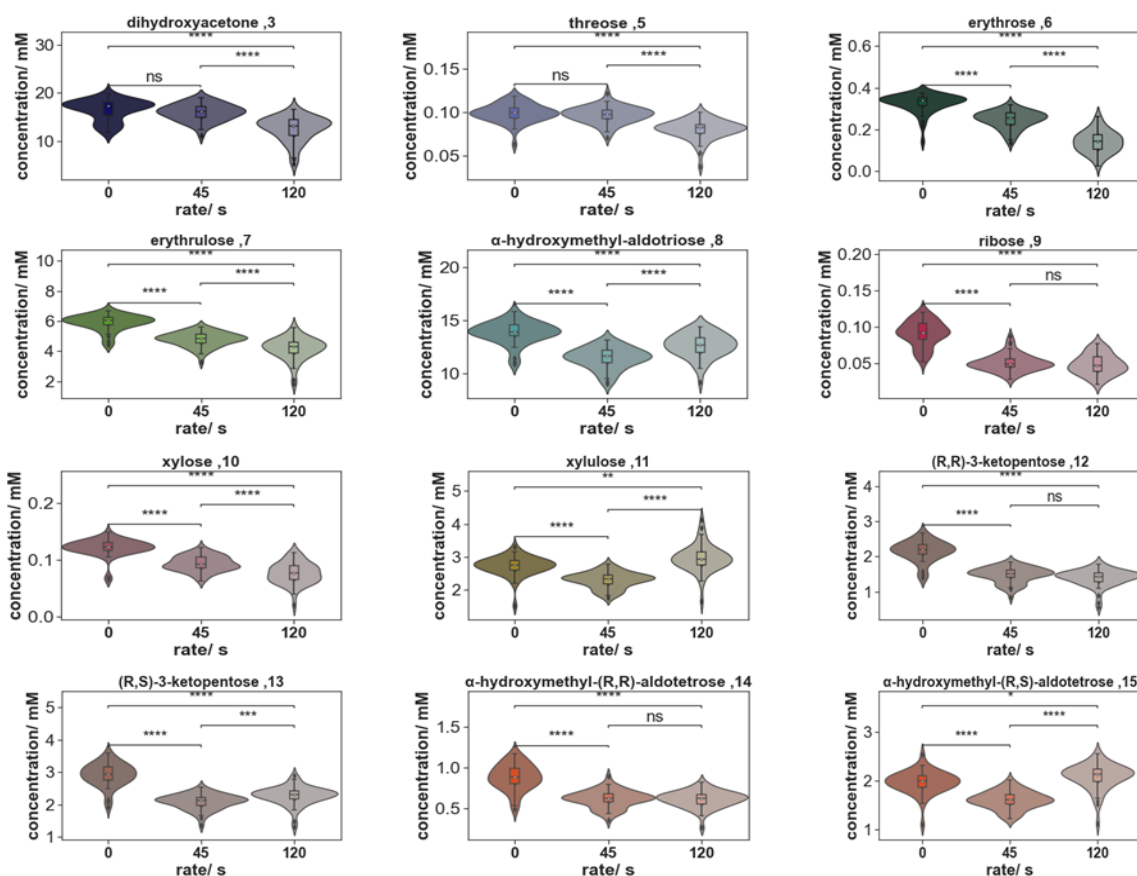


Figure S12. Violin plots for concentration distributions for the identified compounds (3, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15) at steady state, 45 s and 120 s rate of change ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 5.75$ and 5.75 mM). Conditions: 50 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

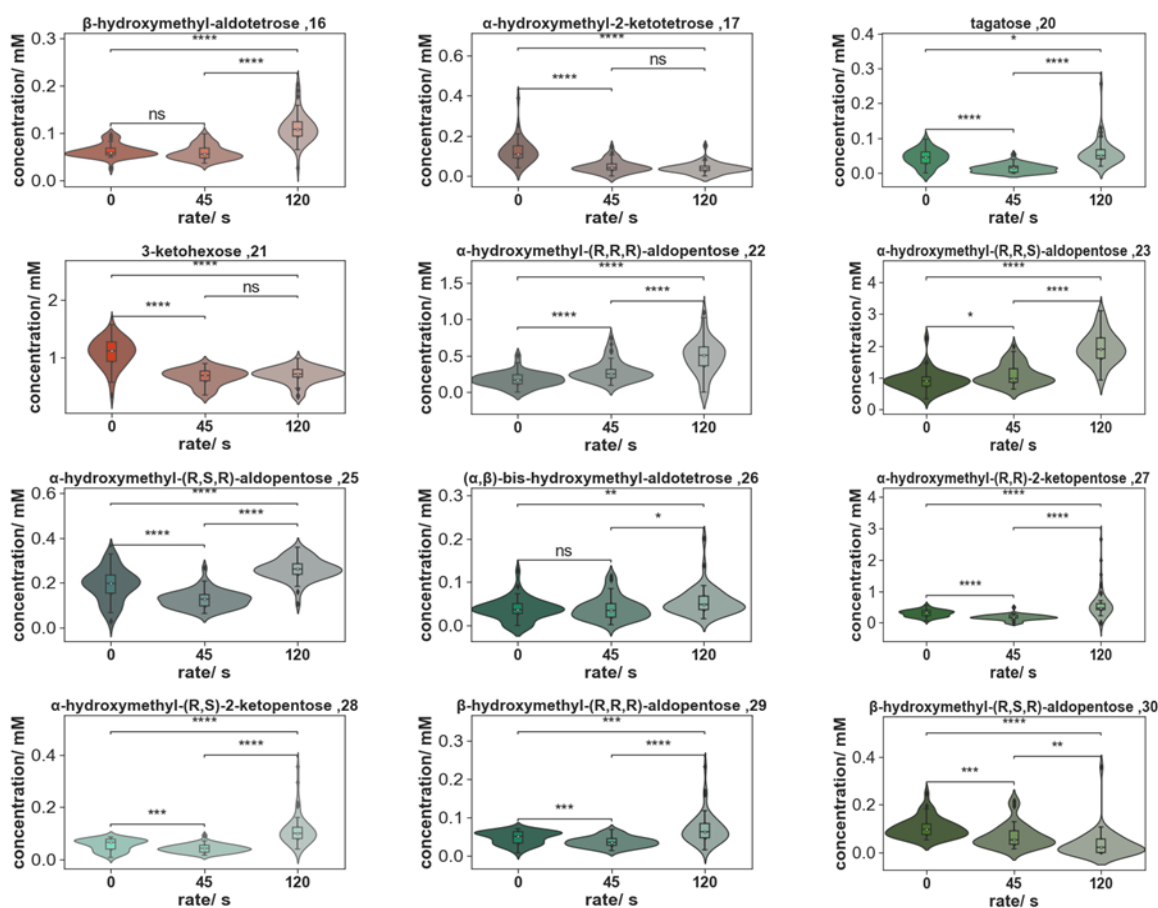


Figure S13. Violin plots for concentration distributions for the identified compounds (16, 17, 20, 21, 22, 23, 25, 26, 27, 28, 29, 30) at steady state, 45 s and 120 s rate of change ($\sigma_{[\text{Ca}(\text{OH})_2]_{\text{in}}} = 0, 5.75$ and 5.75 mM). Conditions: 50 mM $\text{H}_2\text{C}=\text{O}$, 50 mM DHA, 30 mM NaOH, 15 mM CaCl_2 .

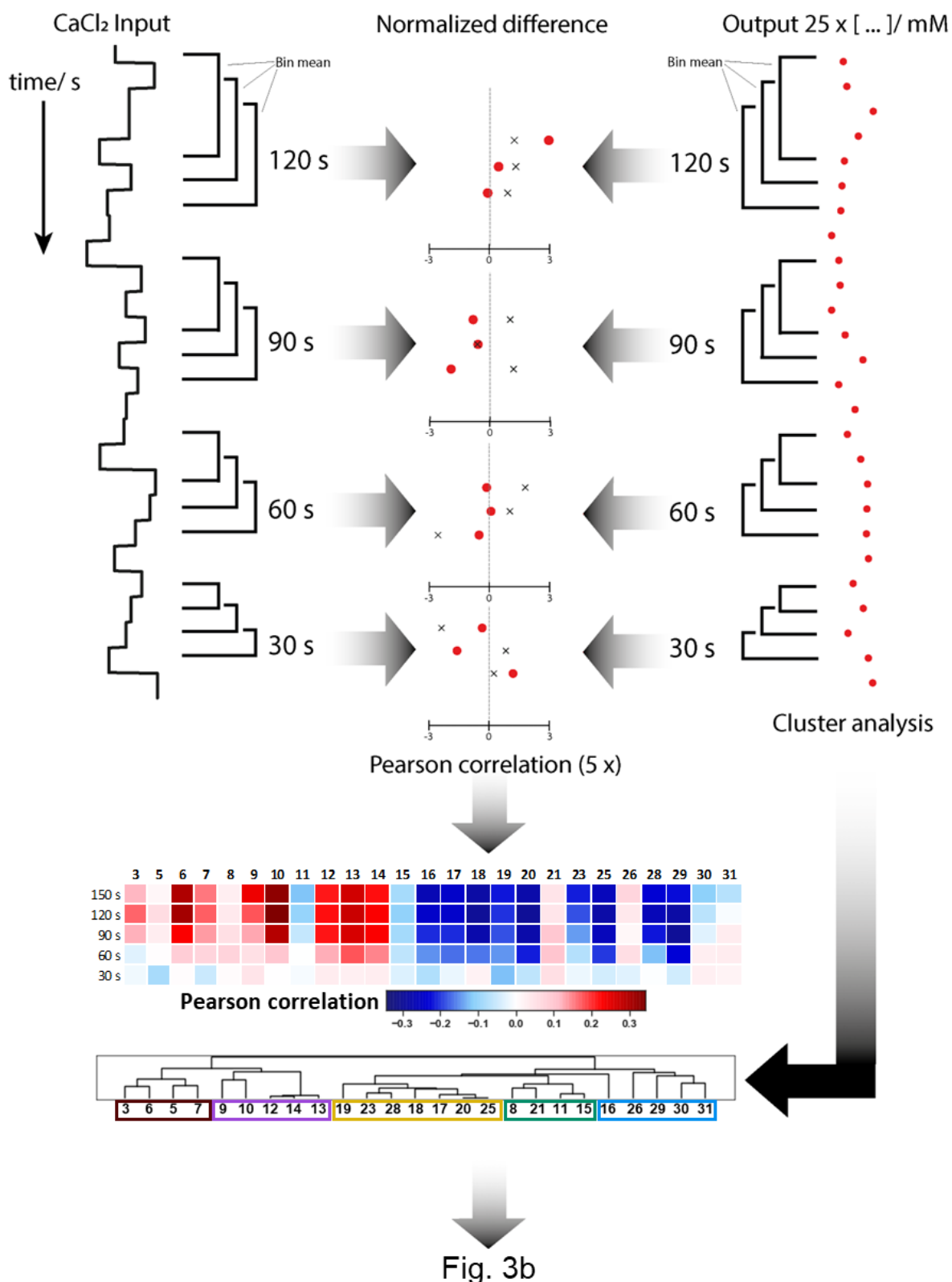


Figure S14. Scheme summarizing the correlation analysis on the 50 mM H₂C=O network, perturbed with a complex input signal with three embedded rates of change (120 s, 60 s, 30 s). An example input concentration profile for Ca(OH)₂ and an output concentration profile for an example compound are shown on respectively the left and right hand side. From these, at different time-intervals – 150s, 120s, 90s, 60s and 30s – samples were binned and a mean concentration was calculated. Subsequently, for each time-interval, the concentration

difference was calculated between consecutive bin means to create a ‘differential’. Both the input and output differential were normalized for each time step. A Pearson correlation was calculated between corresponding input and output differentials. The calculated correlations for each compound were superimposed on the identified cluster to obtain figure 3b.

Supplementary tables

Table S1. Constants for fitted quadratic GC calibration curves for C₄, C₅ and C₆ compounds.

Compound	A	B
DHA, 3	0.028749	0.118714
threose, 5	0.143626	0.789532
erythrose, 6	-0.38468	2.13775
erythrulose, 7	0.143519	0.222204
ribose, 9	0.339712	1.487932
xylose, 10	0.048971	2.23855
xylulose, 11	0.695714	-0.17935
fructose, 17	0.589499	2.049516
sorbose, 18	0.08	2.354559
tagatose, 19	-0.10622	1.017795
8	0.071759	0.111102
12 to 16	0.229793	0.618952
21 to 31	0.250408	1.677321

The same calibration curves have been used as previously reported¹, based on: $[compound] = A \times (\frac{peak\ integral}{internal\ standard\ integral})^2 + B \times \frac{peak\ integral}{internal\ standard\ integral}$. We estimated a calibration curve for non-calibrated compounds (8, 12 – 16, 21 – 31) by taking the average calibration curve for the calibrated sugars of similar length.

References

1. Robinson, W. E., Daines, E., van Duppen, P., de Jong, T. & Huck, W. T. S. Environmental conditions drive self-organization of reaction pathways in a prebiotic reaction network. *Nat. Chem.* **14**, 623–631 (2022).
2. Harris, C. R. *et al.* Array programming with NumPy. *Nature* **585**, 357–362 (2020).