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Tracking carbon cycling with iLOSCAR: an extension of the LOSCAR model with double-inversion algorithm

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1. Introduction

1.1 Perturbed carbon cycle

- 2022 C Emission: ~10 Gt
- Since 1850
 - pCO₂: 280 – 420 ppmv
 - 1.1°C increase
- Need to decipher the carbon cycle dynamics
- Observational data NOT enough

1.2 Geologic hyperthermal events

- Anomalous δ¹³C excursion → Carbon cycle perturbation
- Rapid global warming
- Trigger: large carbon injection → Modern analog

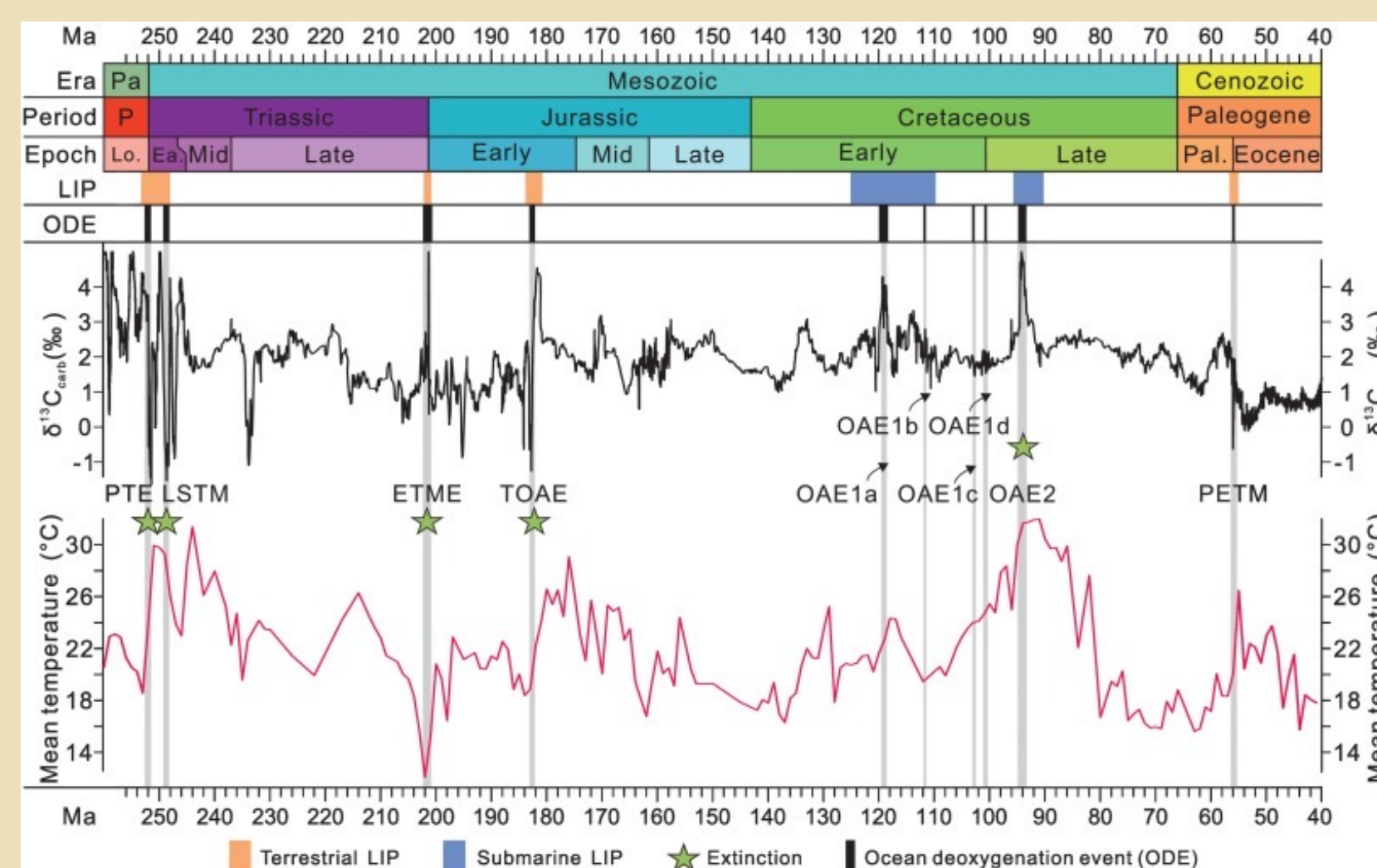


Figure 1. Integrated records of climate conditions, carbon cycle perturbations and geological hyperthermal events during the Mesozoic–Paleogene (He et al., 2023)

1.3 Carbon emission trajectory

- Left: emission mass → How much carbon released?
- Right: isotopic signature → What is the carbon source?

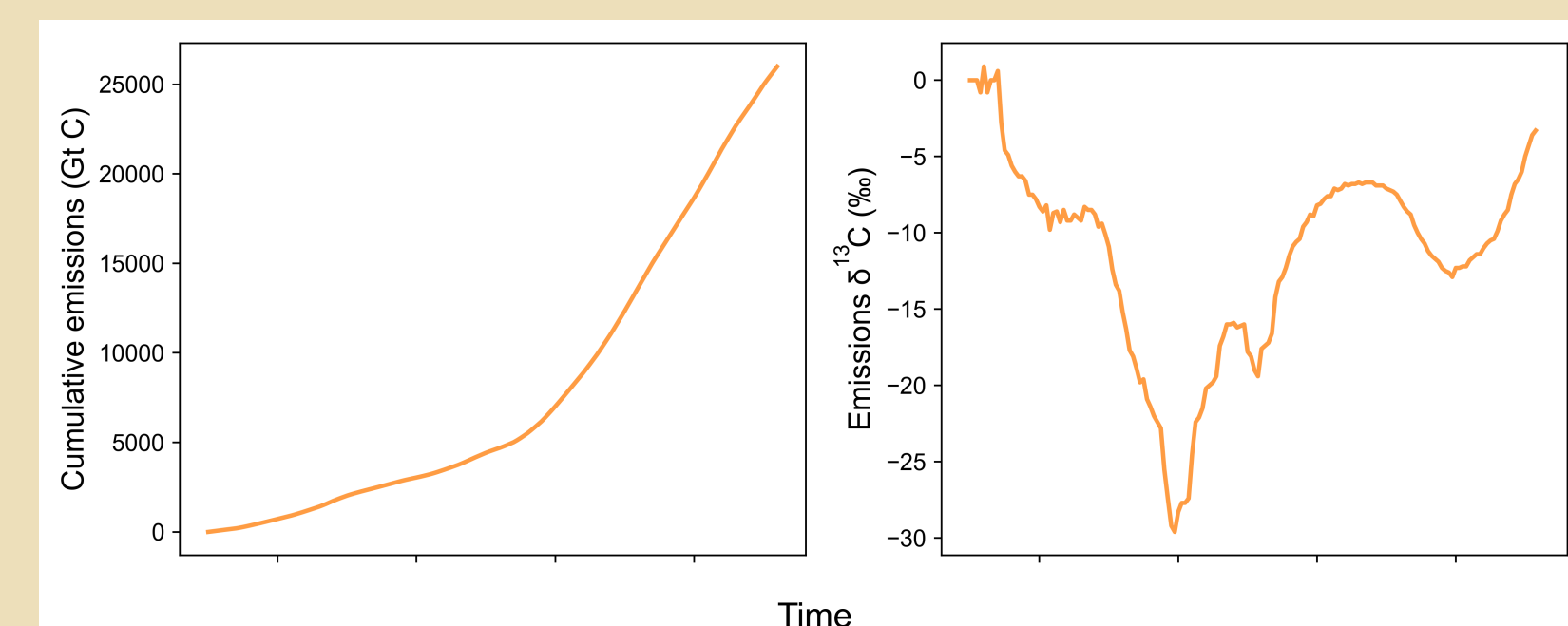
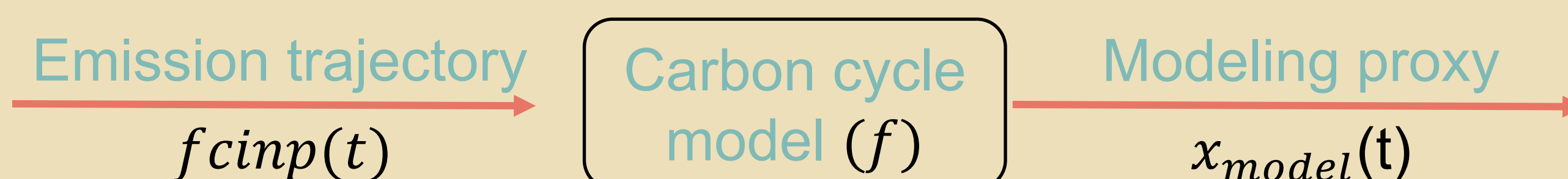


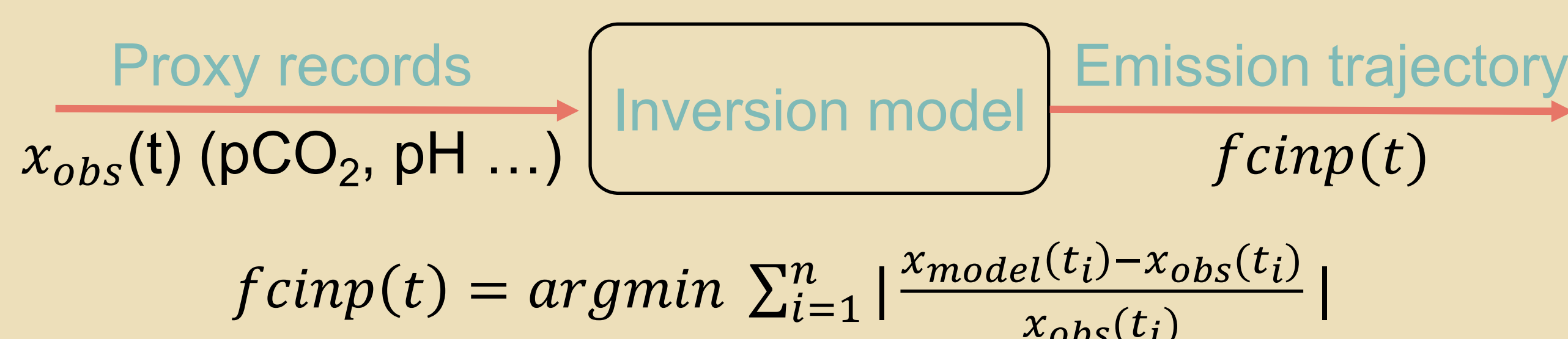
Figure 2. An example of carbon emission trajectory across a specific event.

1.4 Current method

- Combination of proxy records and the carbon cycle model
 - $x_{model}(t) = f(fcinp(t))$



1.5 Aim: Inversion model development

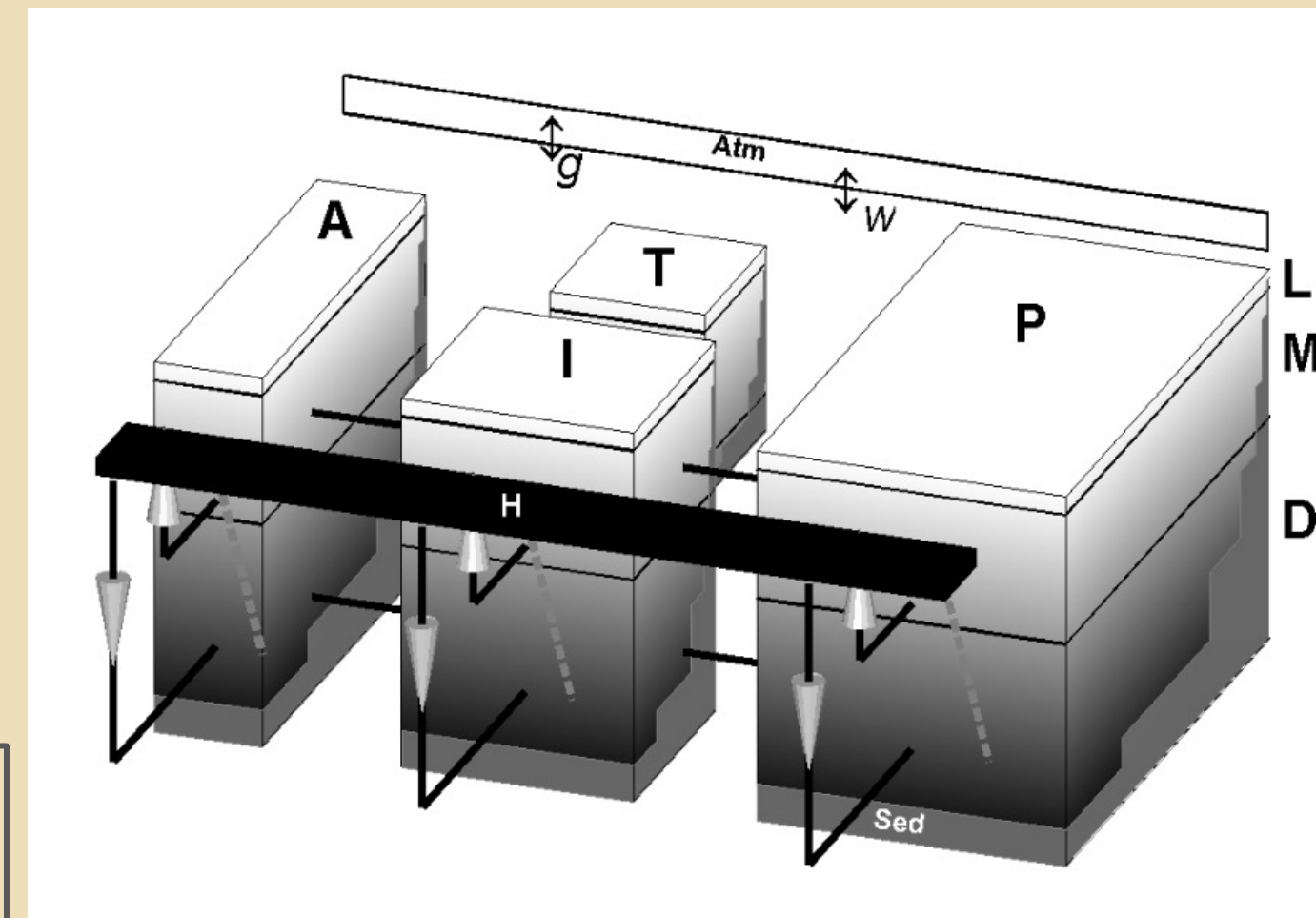


2. Model development

2.1 Base model: LOSCAR

- Atmosphere, Ocean, Sediments
- Efficient: several seconds for a 200-kyr experiment

Figure 3. Architecture of the LOSCAR model (Zeebe, 2012).



2.2 Inversion algorithm

- Divide the proxy data to n intervals and assume a constant emission rate within each interval, i.e.:

$$fcinp(t) = \begin{cases} k_1 & (if\ t_0 \leq t \leq t_1) \\ k_2 & (if\ t_1 < t \leq t_2) \\ \dots \\ k_n & (if\ t_{n-1} < t \leq t_n) \end{cases}$$
- Start from the (t_0, t_1) interval and k_1 is the only free parameter that can control the modeled pCO₂ at $t = t_1$
- Employ numerical methods to determine k_1
- Save the corresponding $\vec{y}(t_1)$ as the initial $\vec{y}$ for the next interval (t_1, t_2)
- Iterate the same process until k_n is solved

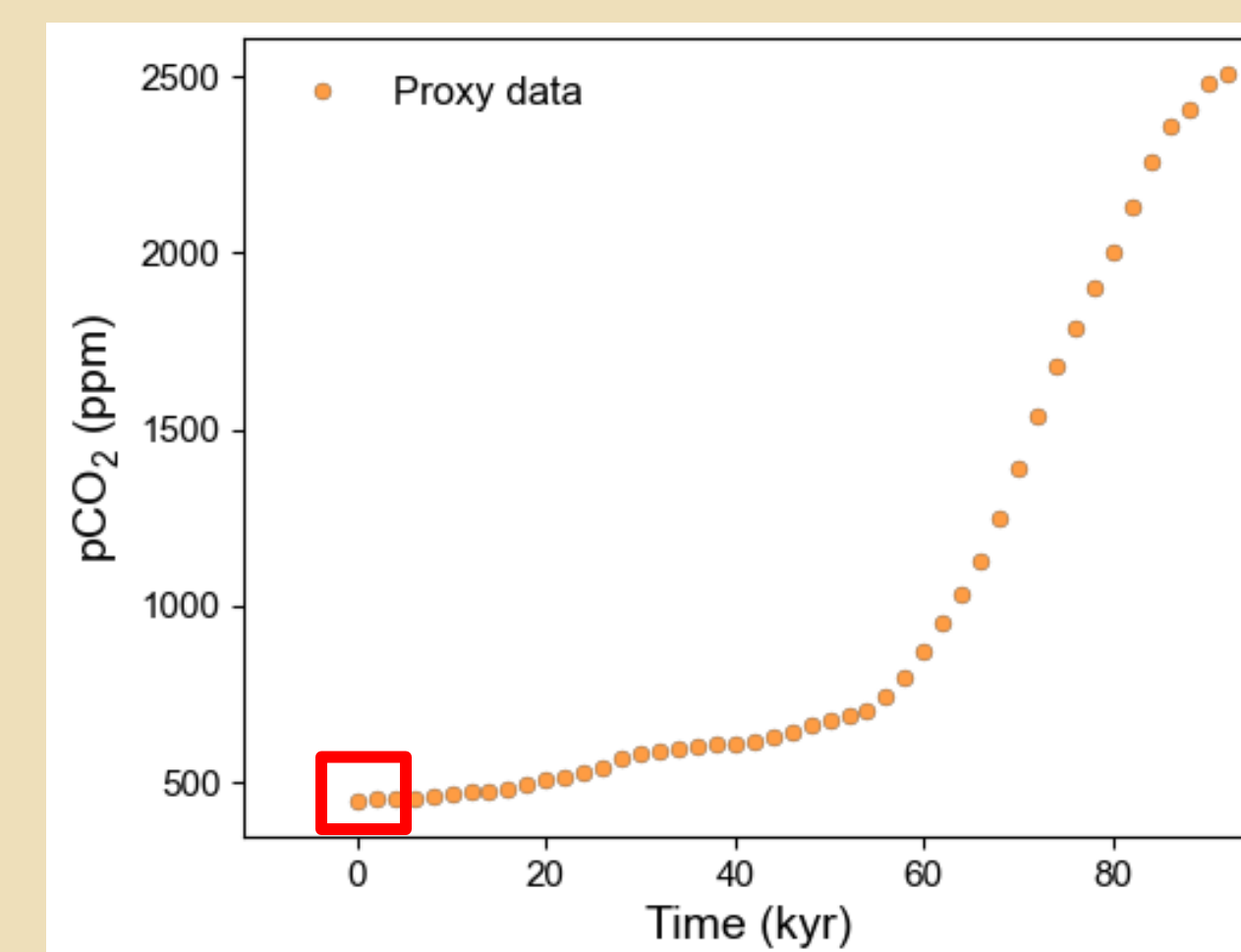


Figure 4. An example of pCO₂ records for the inversion model.

3. Model validation

Event: End Permian Mass Extinction (~252 Ma)

- 5-6‰ δ¹³C decrease
- Low-latitude 8-10°C warming

Largest mass extinction in the Phanerozoic (~90% or marine species)

- Recent emission trajectory estimate by Wu et al., 2023
 - c-GENIE (intermediate complexity) based inversion
 - Constraints: pCO₂ and δ¹³C
 - Ideal for model intercomparison

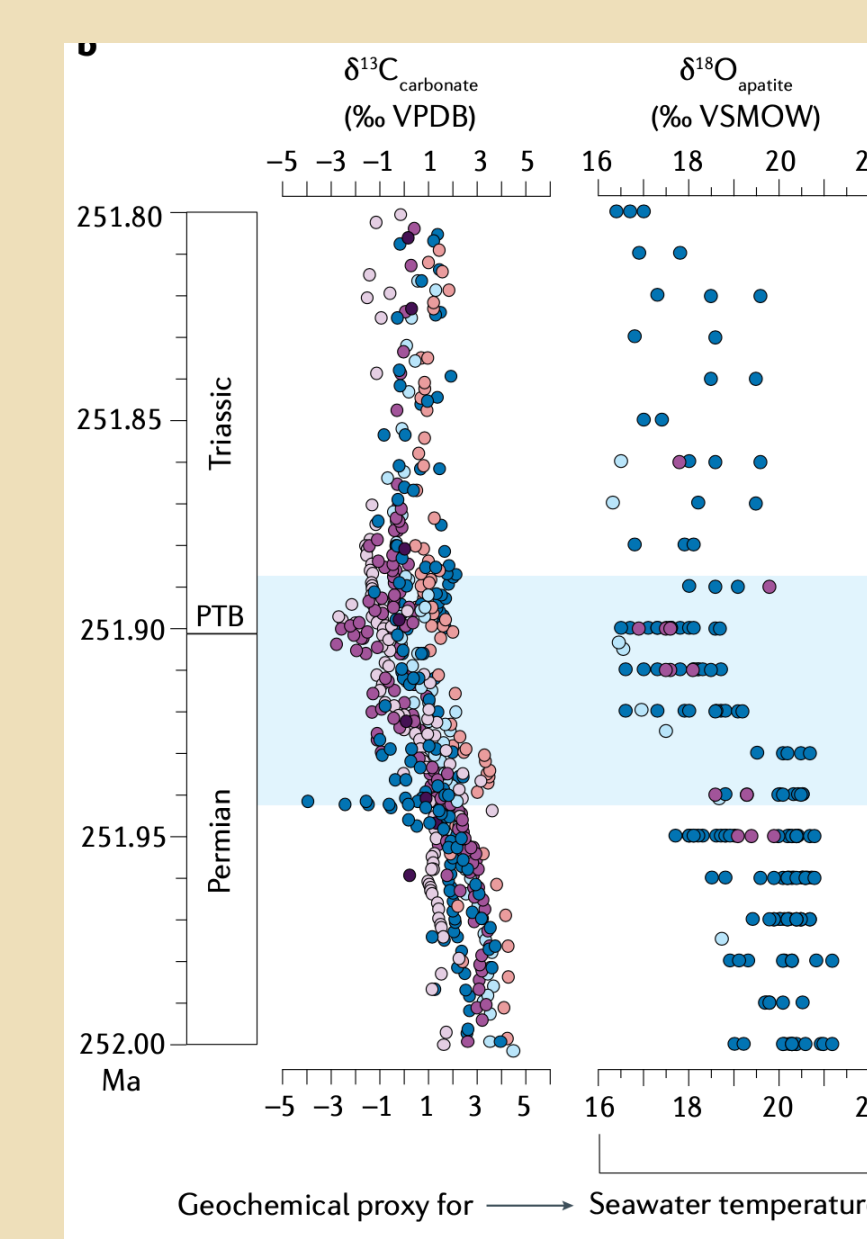
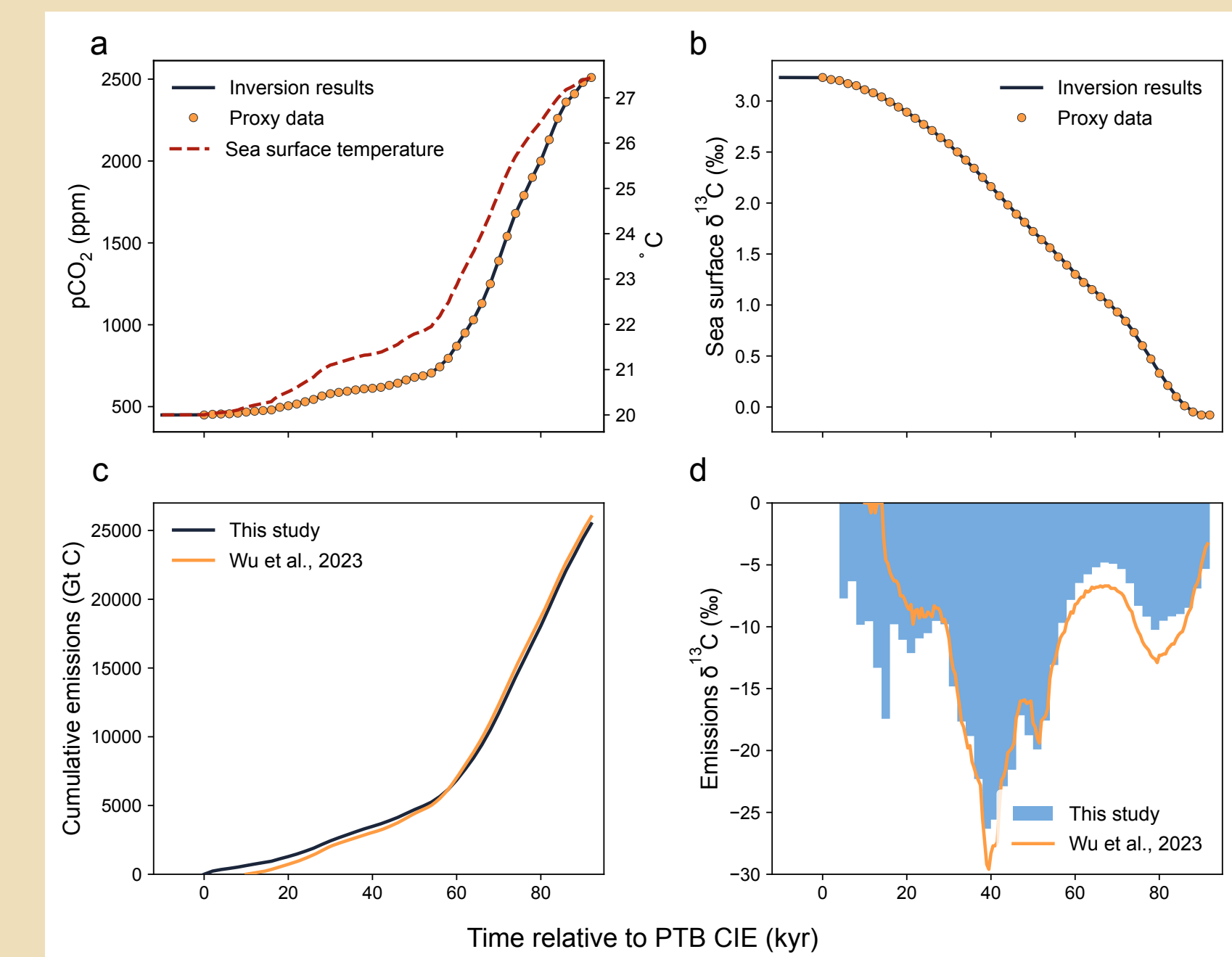


Figure 5. Compiled δ¹³C and δ¹⁸O records across the EPME, after Corso et al., 2022.

Results

- Modeling results align with proxy pCO₂ and sea surface δ¹³C data
- Results mirror findings from Wu et al., 2023
 - ~21,000 Gt C emission
- Running time: **13.6 min** vs **1-2 months** in cGENIE

Figure 6. Inversion modeling results for the EPME.



4. Case study

Event: Kasimovian–Gzhelian boundary (~304 Ma)

- δ¹³C decreased, pCO₂ increased, and global warming (Chen et al., 2022)
- Paleo-glacial state

Inversion experiment

- Input: pCO₂, Sea surface δ¹³C
- Modeling results align with proxy records
- ~9,000 Gt C emission
- Two new features
 - Negative C emission → Organic C burial
 - A gradual decrease in δ¹³C_{source}

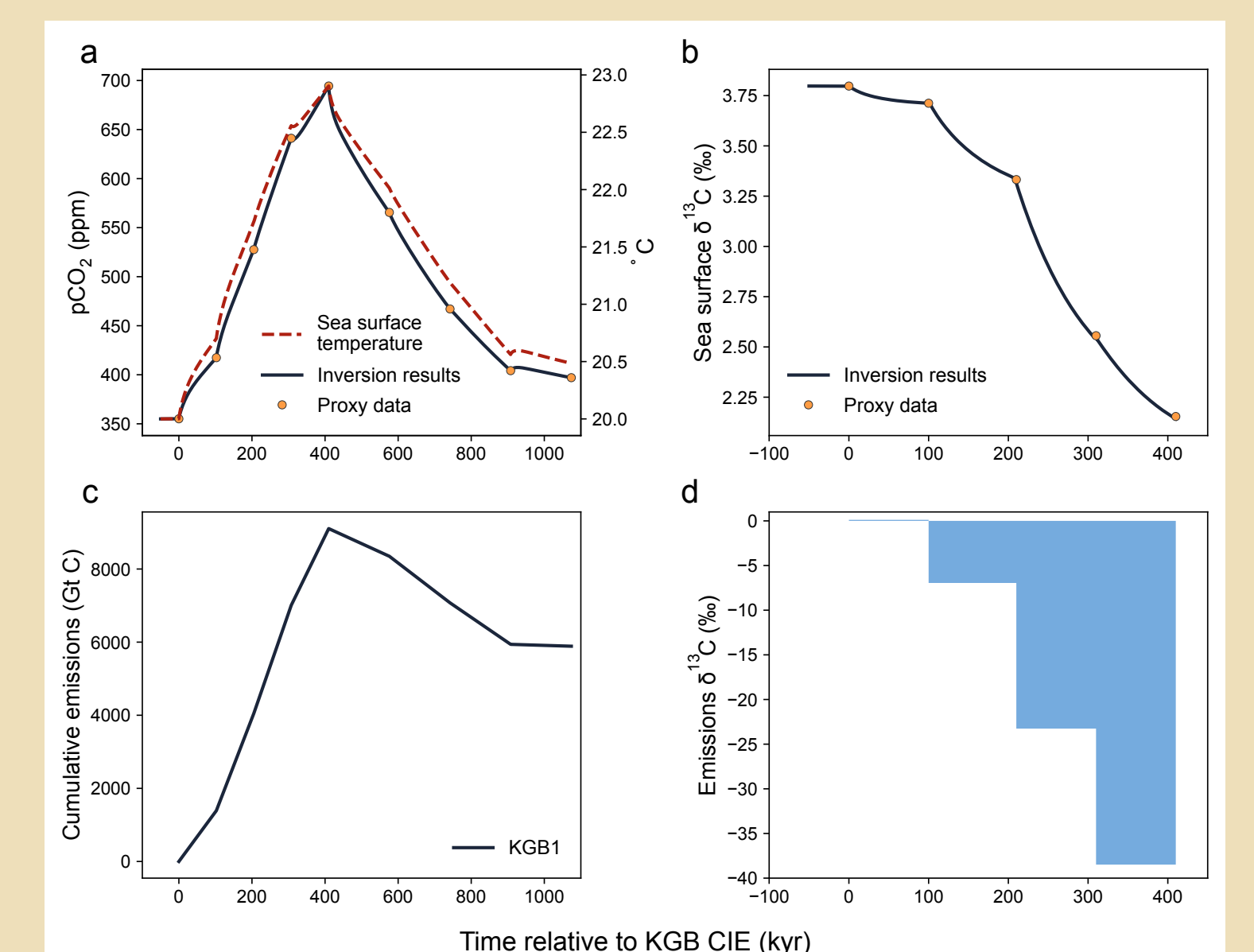


Figure 7. Inversion modeling results for the KGB.

5. Summary and outlook

iLOSCAR development

- A reliable, efficient and user-friendly model tool
- Open-source, <https://github.com/Shihan150/iloscar>, tutorial available
- Better constrain the carbon emission trajectories in geologic hyperthermal events

Outlook

- Multiple proxies inversion
- Sensitivity test of inversion results on the model settings