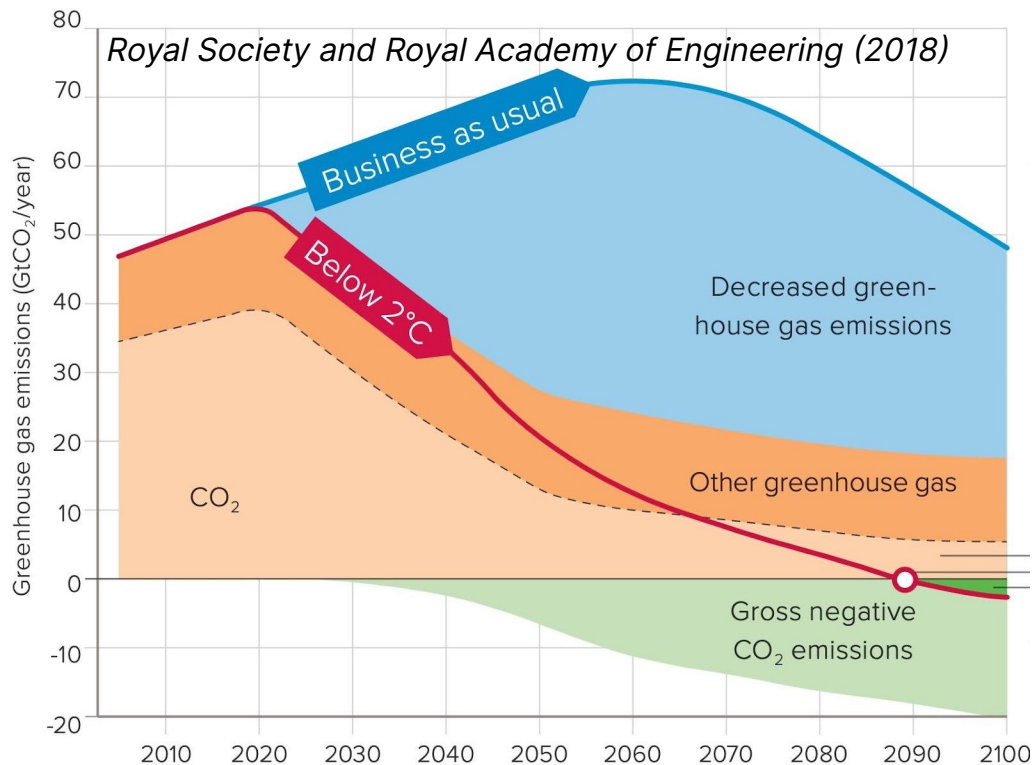


Impulse Response Functions:

A statistical tool for marine CO₂ removal quantification

Elizabeth Yankovsky

Carbon dioxide removal is required to achieve net zero



THE CHALLENGE: Reaching net-zero

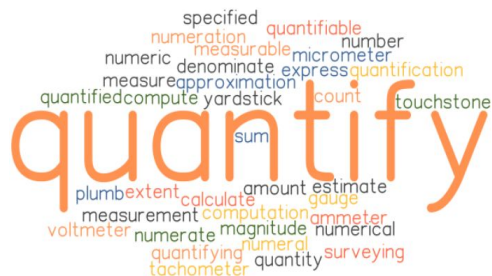
Meeting the 1.5°C - 2°C climate target requires billions of tons of active Carbon Dioxide Removal (CDR) annually by mid-century.

THE PROMISE: Ocean CDR technologies

Ocean-based CDR technologies have the potential to scale to meaningful levels – as such, they are forecast to experience rapid development in the next decade.

THE CRITICAL GAP: Quantification

There are no accepted frameworks for **quantifying ocean-CDR**. Robust quantification requires both observations and numerical models.



Carbon dioxide removal is required to achieve net zero

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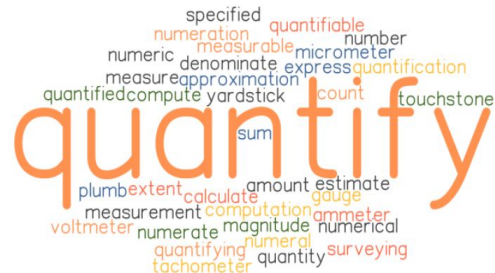
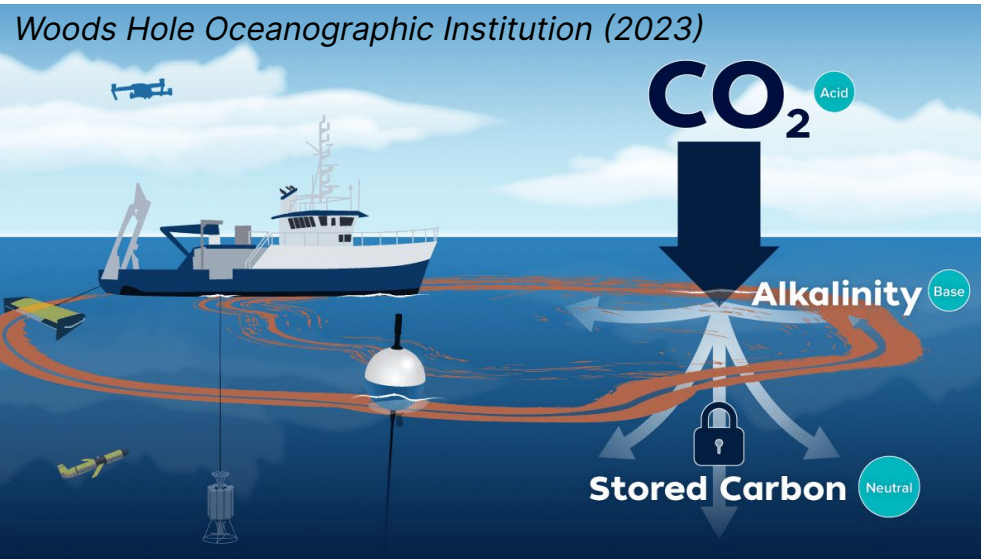
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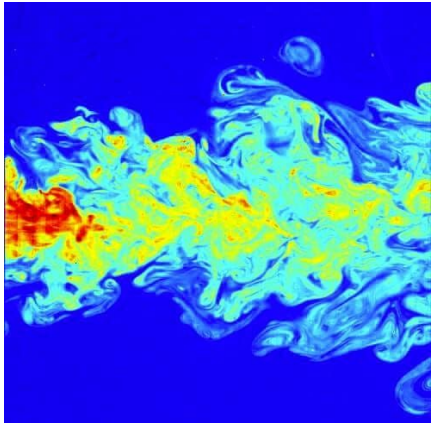
THE CRITICAL GAP: Quantification

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mCDR: A Multiscale Problem

Microscale



- 3D turbulence
- Spatial scale: $\mathcal{O}(10^{-3}-10^0)$ m
- “Fundamental” fluid dynamics

Near Field



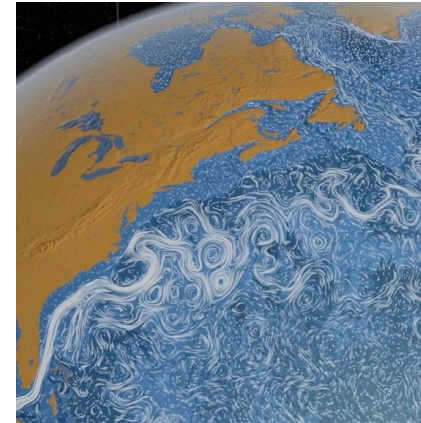
- **mCDR intervention**
- Spatial scale: $\mathcal{O}(10^0-10^2)$ m
- Project-specific dynamics

Local Region



- **Initial dispersion**
- Spatial scale: $\mathcal{O}(10^2-10^3)$ m
- Region-specific dynamics

Regional to Basin



- **Air-to-sea CO₂ transfer**
- Spatial scale: $\mathcal{O}(10^3-10^6)$ m
- Basin to global-scale model

Global



- Earth system model
- Spatial scale: $\mathcal{O}(10^6-10^7)$ m
- Deep overturning dynamics



Why is Measurement, Reporting, and Verification (MRV) so difficult?

Large spatiotemporal scales

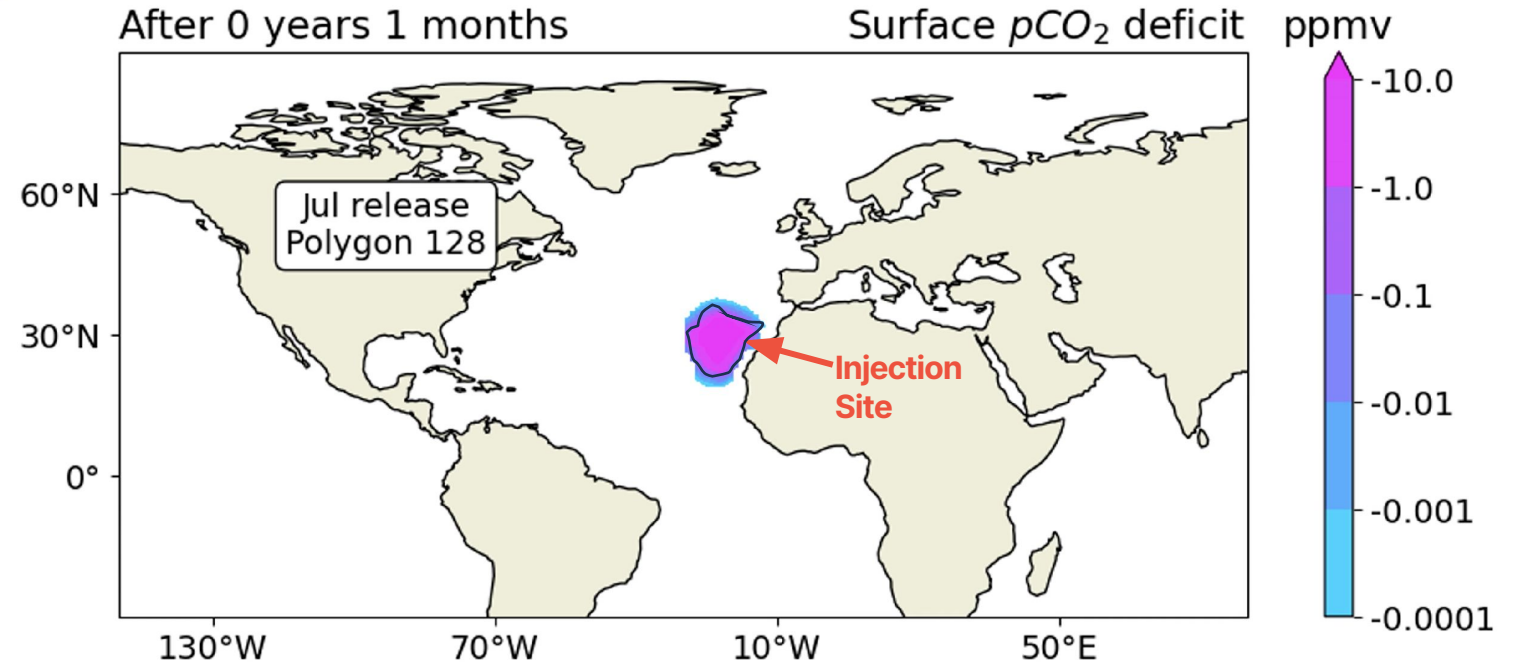
Slow CO₂ equilibration timescales leads to basin-scale signals that cannot be monitored in-situ.

Unfavorable signal-to-noise

High intrinsic background variability relative to signal size makes detection of signal beyond the nearfield impossible.

Complex baselines

Counterfactual baselines are needed to assess additionality.



$p\text{CO}_2$ anomaly from simulated ocean alkalinity enhancement

Figure courtesy of Mengyang Zhou



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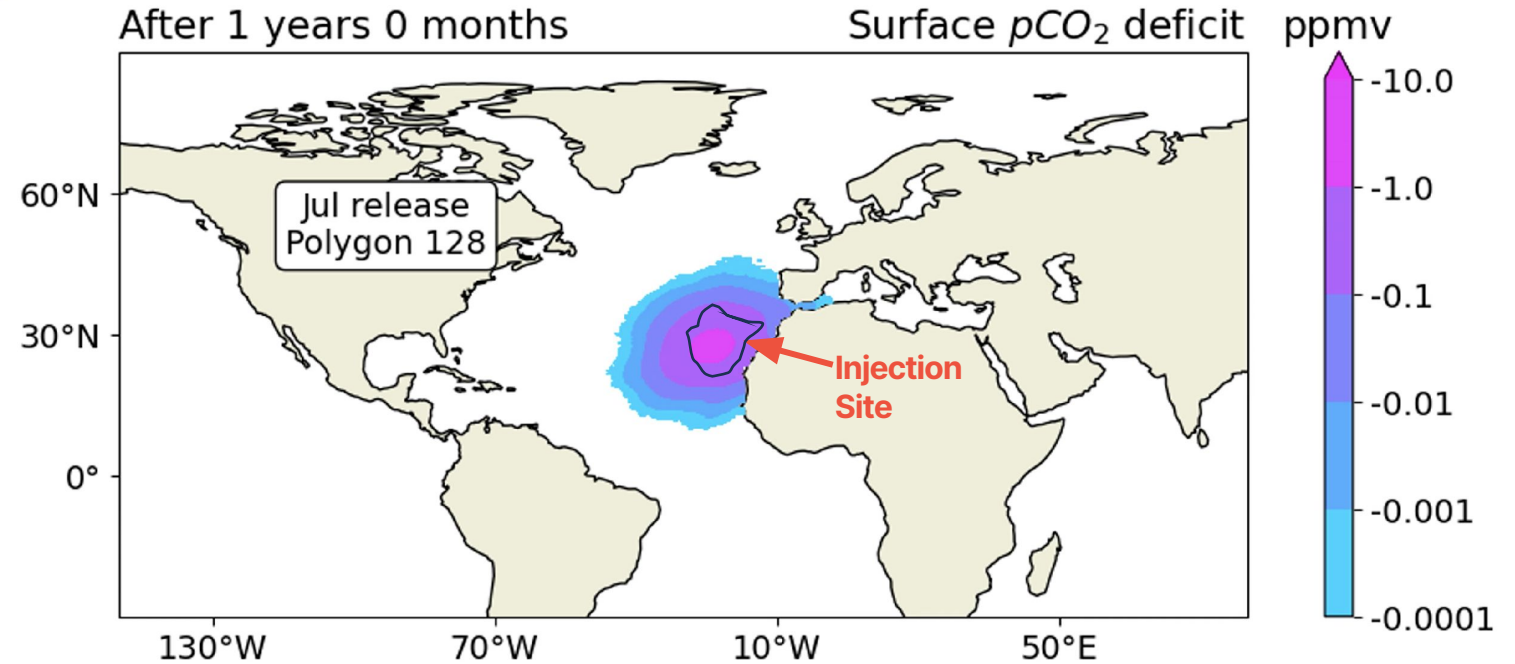
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**pCO₂ anomaly from
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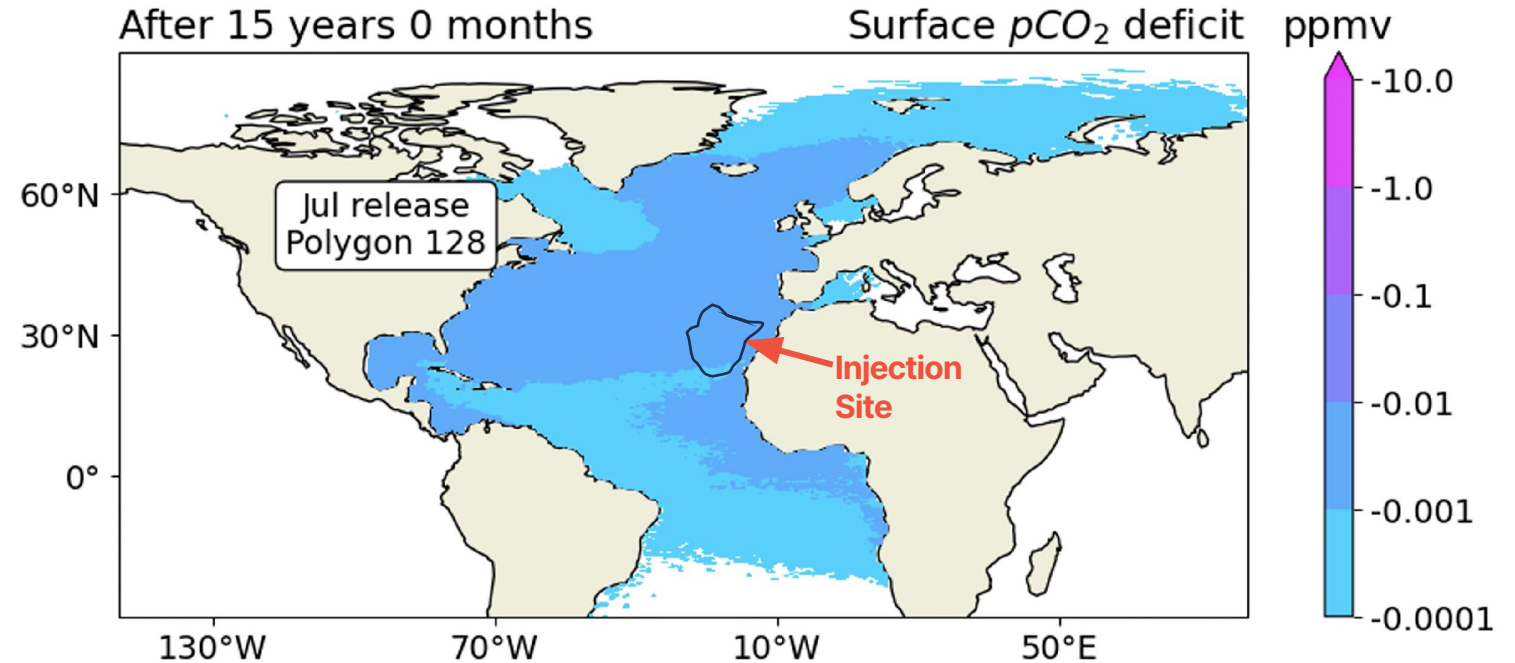
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**pCO₂ anomaly from
simulated ocean alkalinity enhancement**

Figure courtesy of Mengyang Zhou





Building the modeling tools needed to ensure safe, effective, and market-ready marine carbon dioxide removal (mCDR).



Matt Long, Ph.D.
Co-Founder,
Chief Executive Officer



Alicia Karspeck, Ph.D.
Co-Founder,
Chief Technology Officer



David Ho, Ph.D.
Co-Founder,
Chief Science Officer



Scott Bachman, Ph.D.
Technical Lead



Namy Barnett
CDR Intern



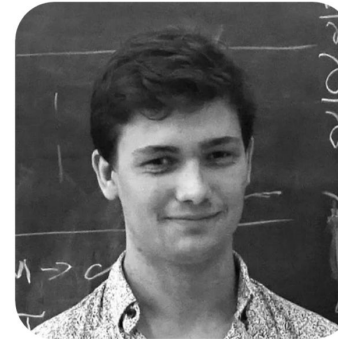
Ulla Heede, Ph.D.
Staff Scientist: mCDR Modeler



Elizabeth Yankovsky, Ph.D.
Postdoctoral Research Scientist



Dafydd Stephenson, Ph.D.
Staff Scientist: Scientific Algorithms



Tom Nicholas, Ph.D.
Staff Scientist: Data Analytics



Frank
Head of Security



[C]Worthy is a nonprofit R & D organization

Motivation for using Impulse Response Functions (IRFs)

- Infeasible to numerically simulate the fate of every mole of alkalinity □ reduce the problem
- Understand physical processes governing carbon uptake: how is carbon uptake influenced by location, seasonal/interannual variability, ocean dynamics, and deployment strategy?
- Develop a statistical tool to simplify mCDR accounting (MRV)

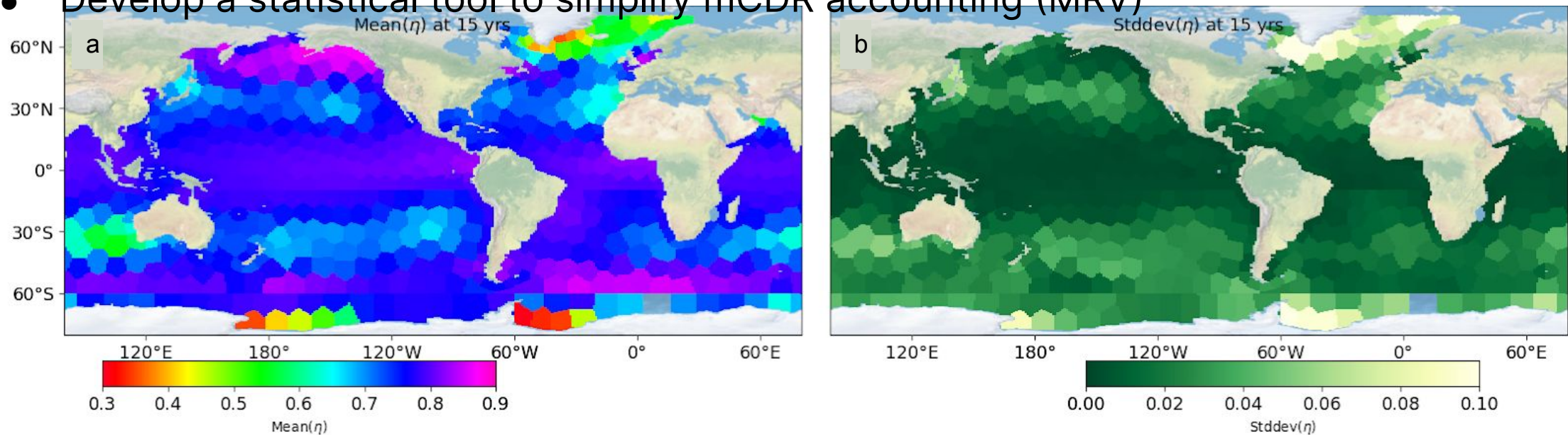
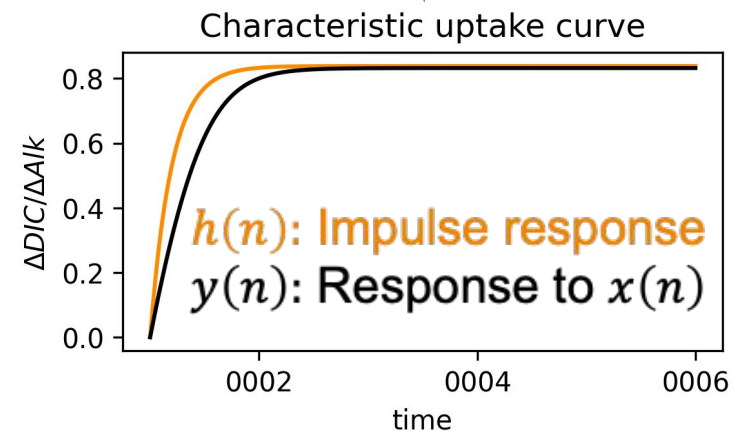
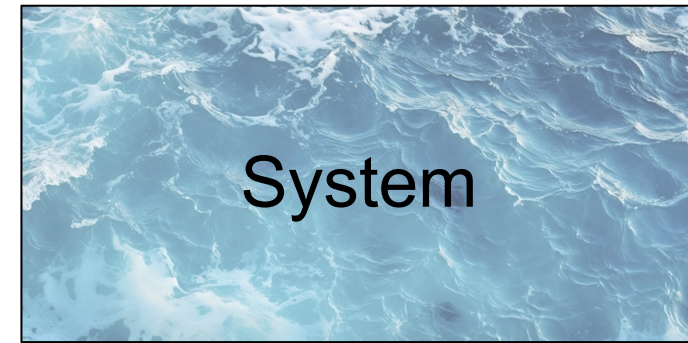
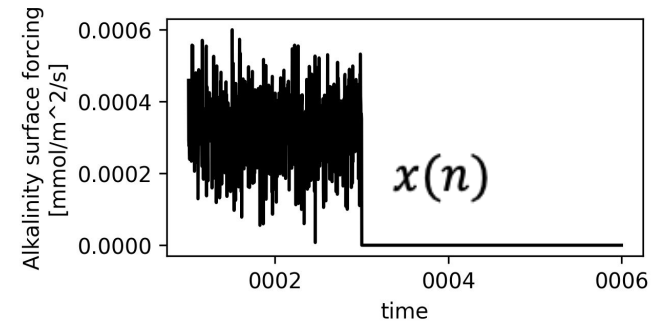
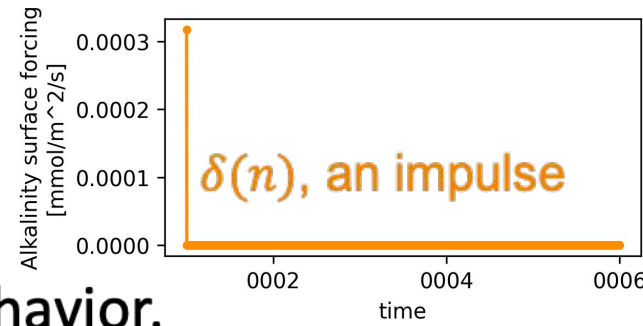


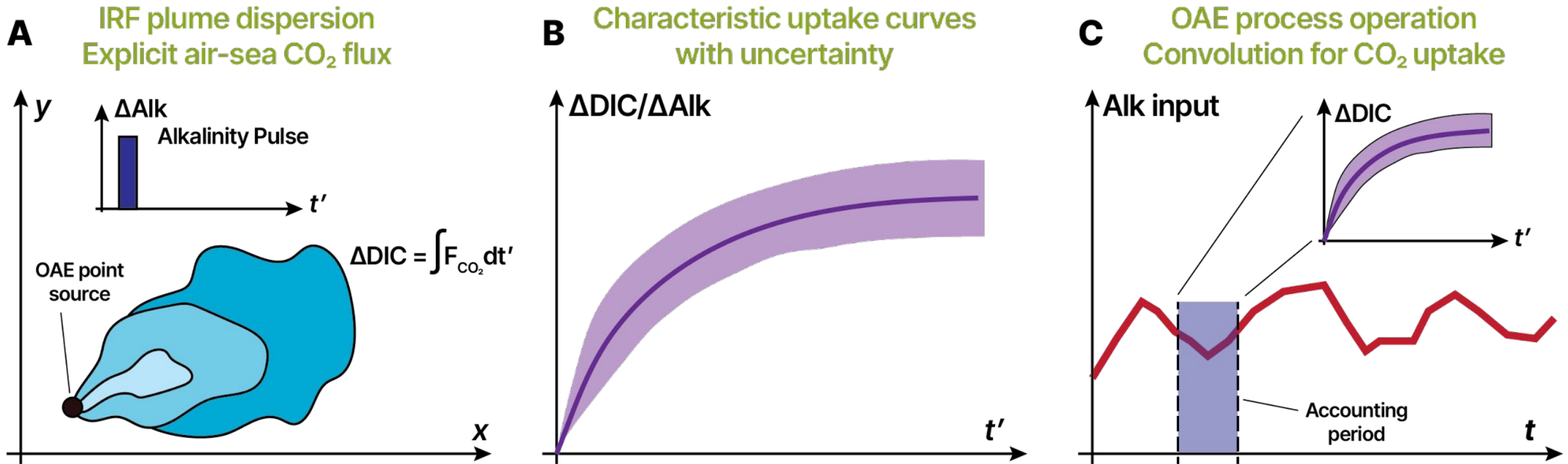
Figure by **Mengyang Zhou**, OAE Global Efficiency Atlas

What is an IRF?

- A way to encapsulate a system's behavior.
- Probe the system with impulse $\delta(n)$, obtain IRF $h(n)$.
- Requirement: a linear and time-invariant system. Then: $y(n) = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$.
- Advantage: just by knowing how the system responds to an impulse allows us to predict how the system will respond to *any*



What is an IRF?

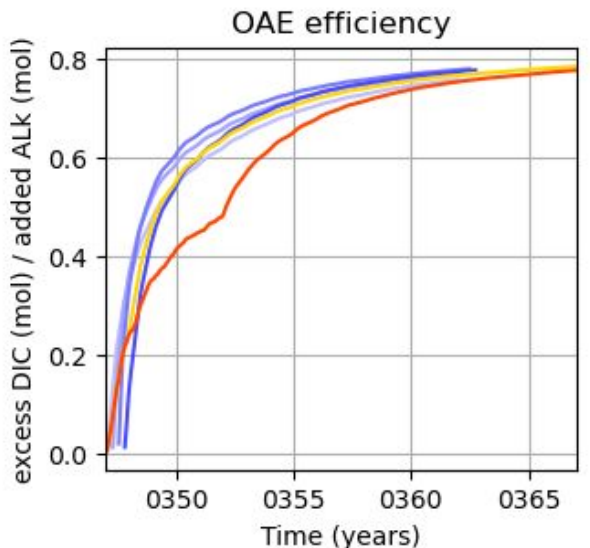
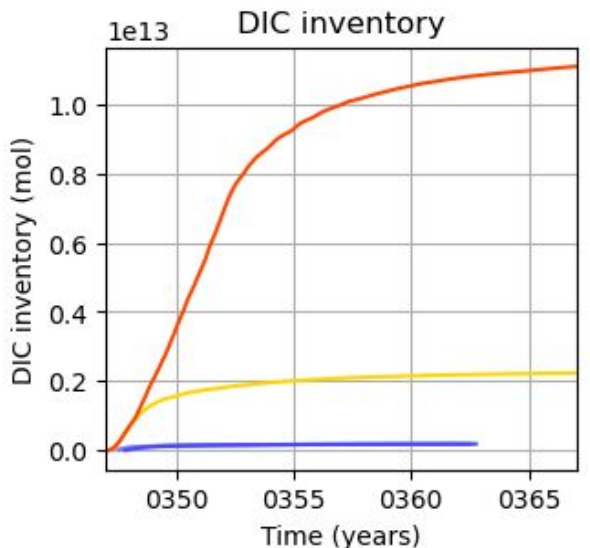
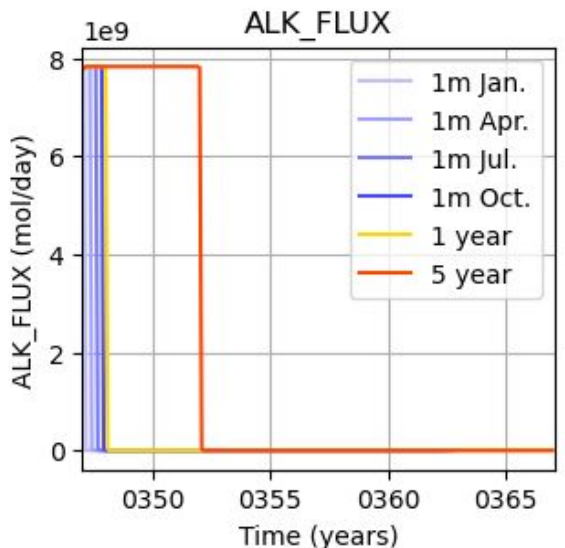
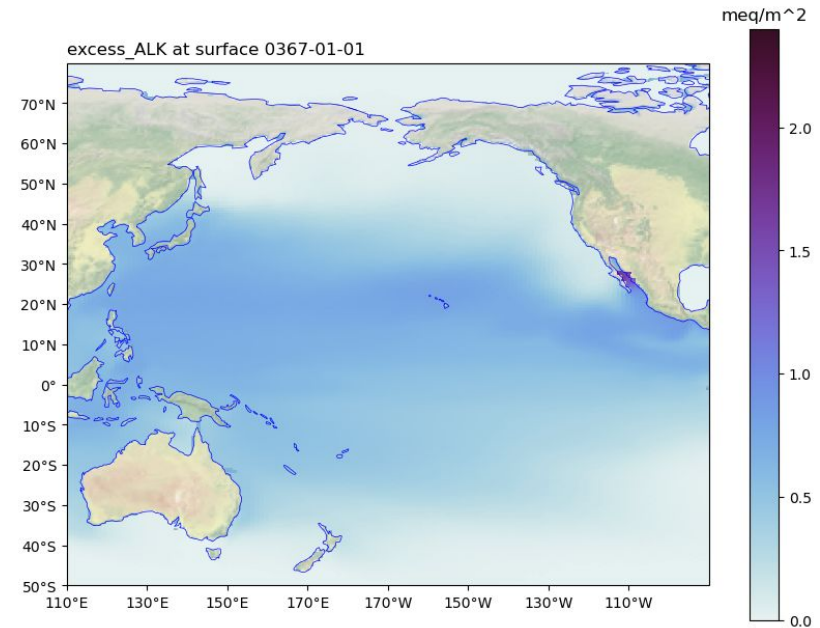
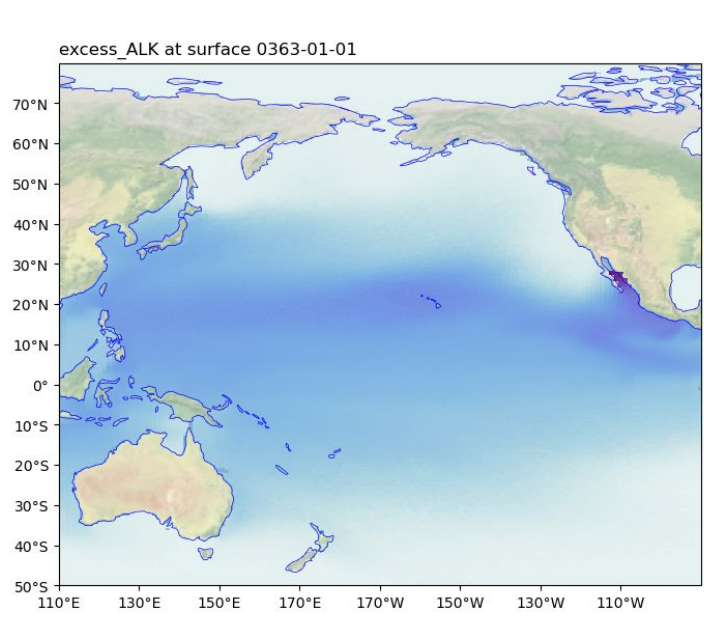
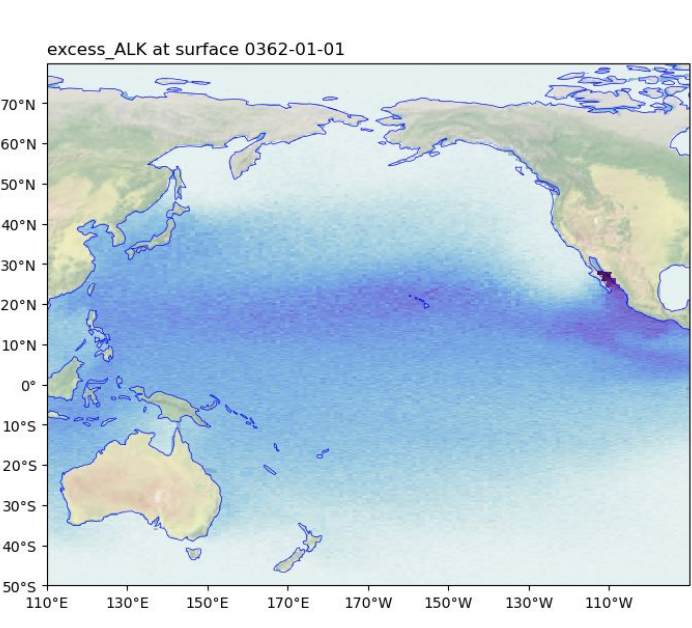


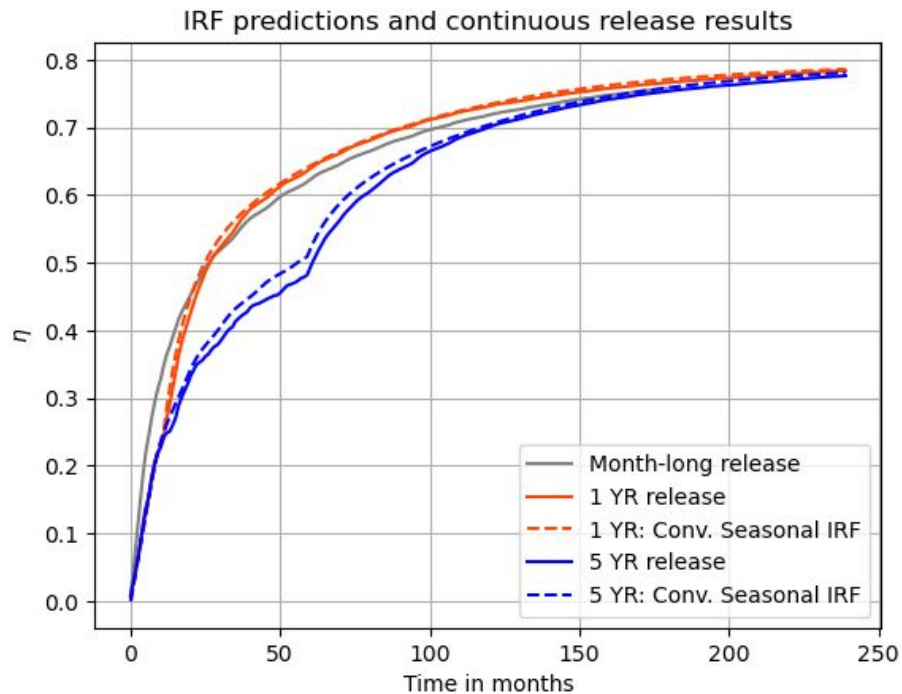
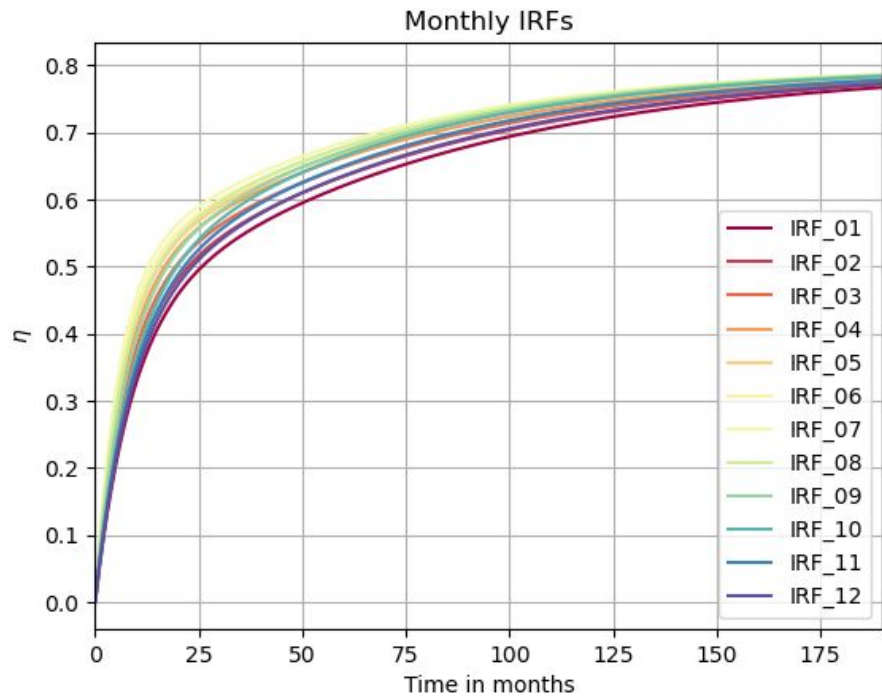
Example: Northern California

1-month release after 15 years

1-year release after 15 years

5-year release after 15 years





Applying the IRF Approach

I. Obtain IRFs

Use Global Efficiency Atlas monthly pulse experiments, interpolate over months.

II. Convolve IRFs with 1-year and 5-year alkalinity forcing.

$$y(n) = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$$

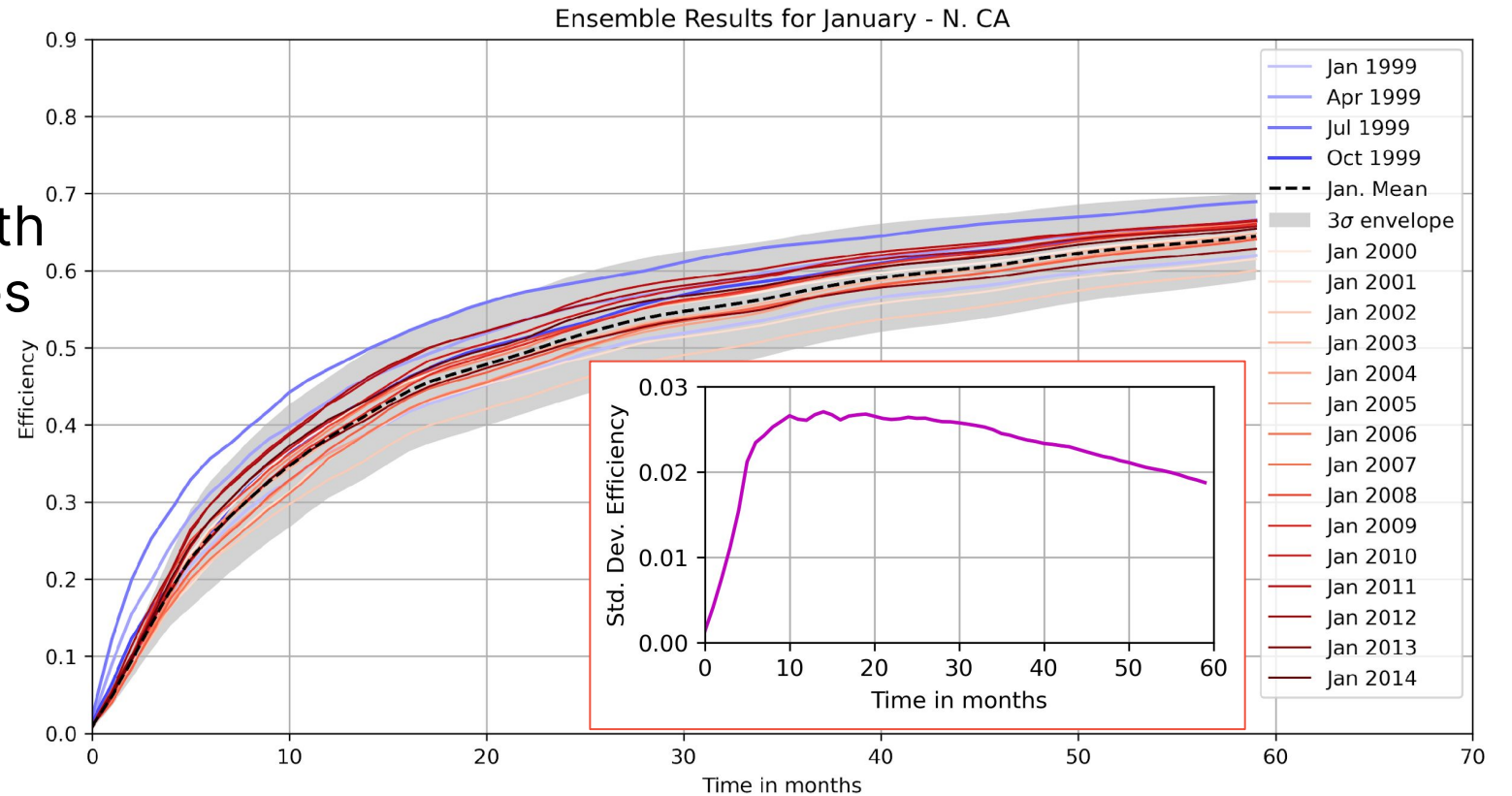
III. Compare the IRF result to the corresponding model simulation.

Observe excellent matchup!



Caveats

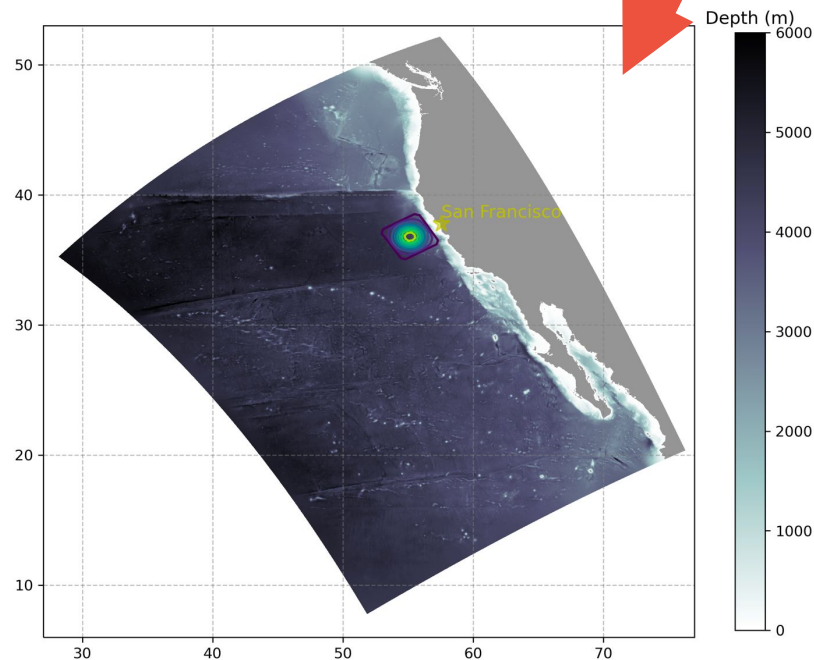
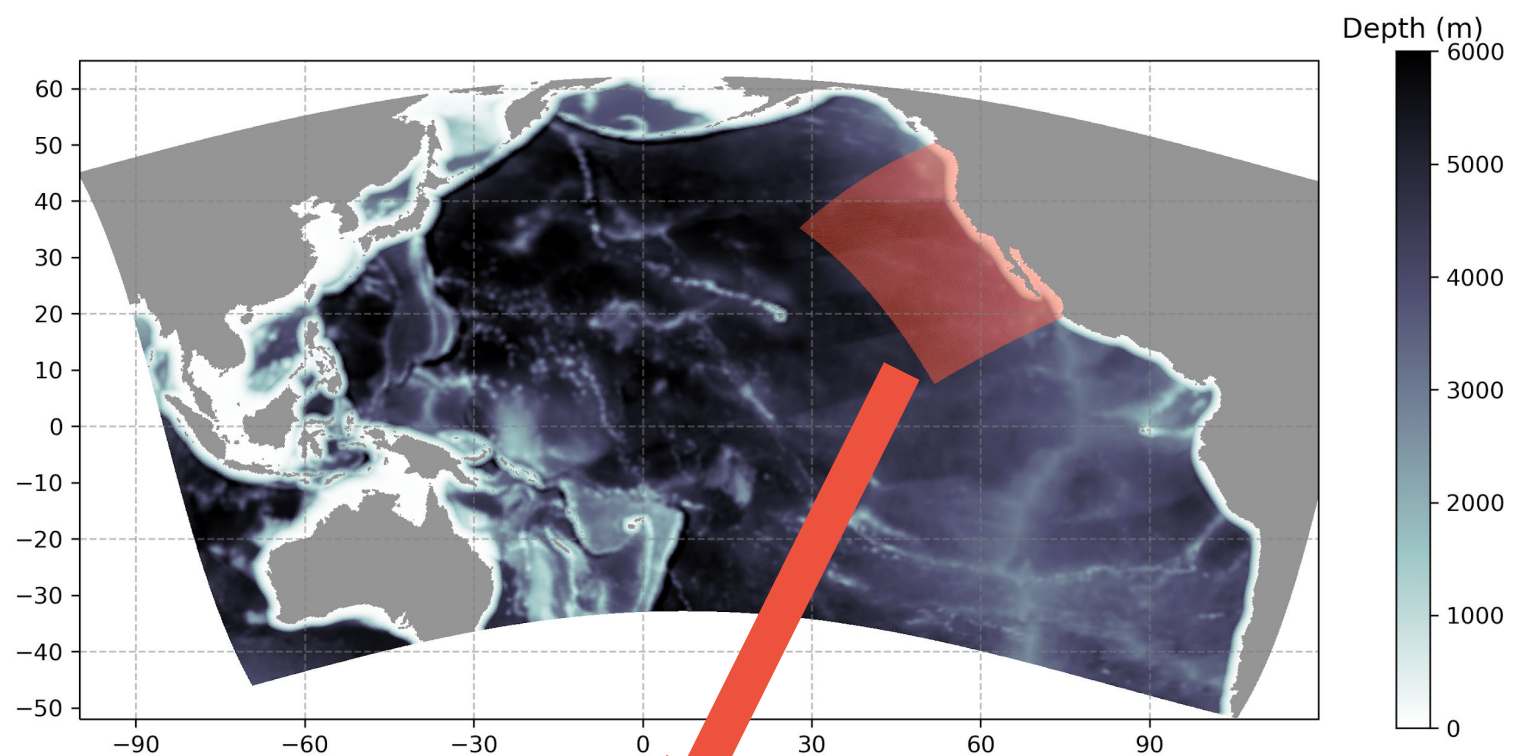
- Uncertainty associated with variability poses challenges
- Sources:
 - Seasonal (handled)
 - Inter-annual
 - Initial conditions
 - Ocean turbulence



Next Steps

Test IRFs in a high-resolution regional model (UCLA-ROMS). Examine role of grid resolution & ocean turbulence.

Gain new physical insights into mesoscale and submesoscale turbulence influences on carbon uptake.



[C] Worthy

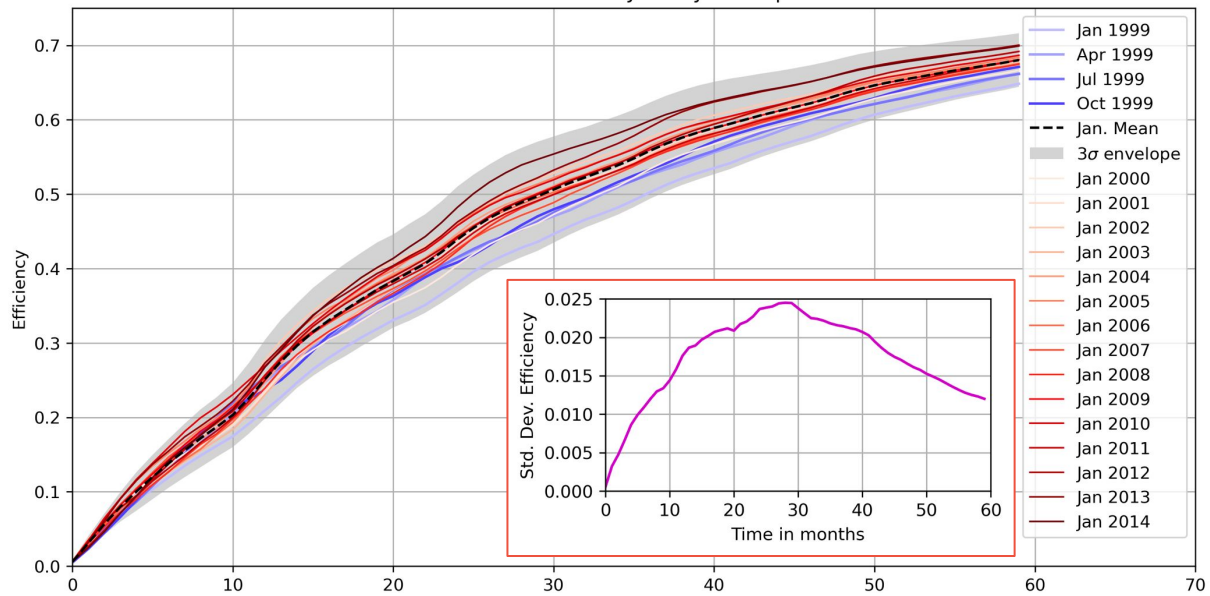
Building the modeling tools needed to ensure safe, effective, and market-ready marine carbon dioxide removal

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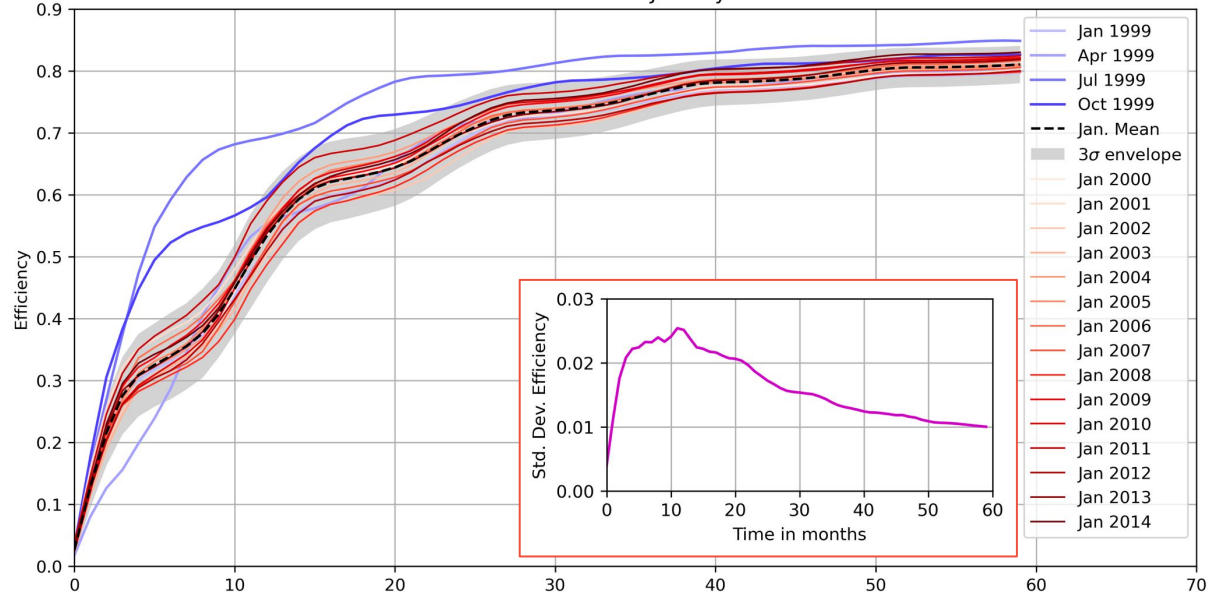


[C]Worthy is a non-profit R&D organization

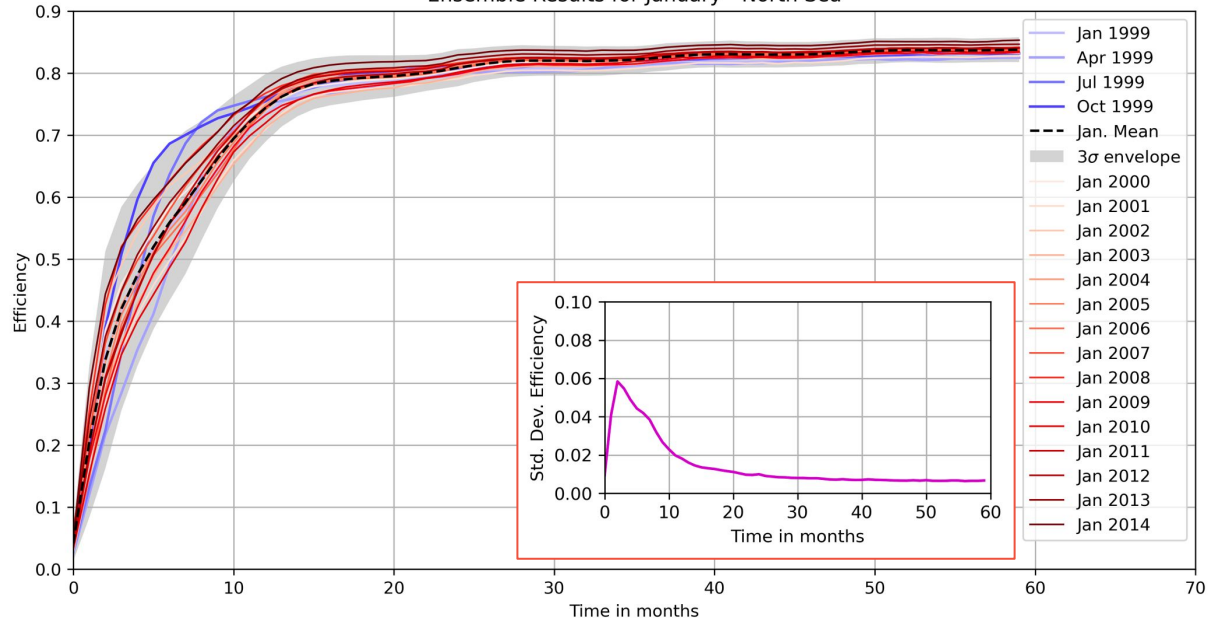
Ensemble Results for January - W. Eq. Pacific



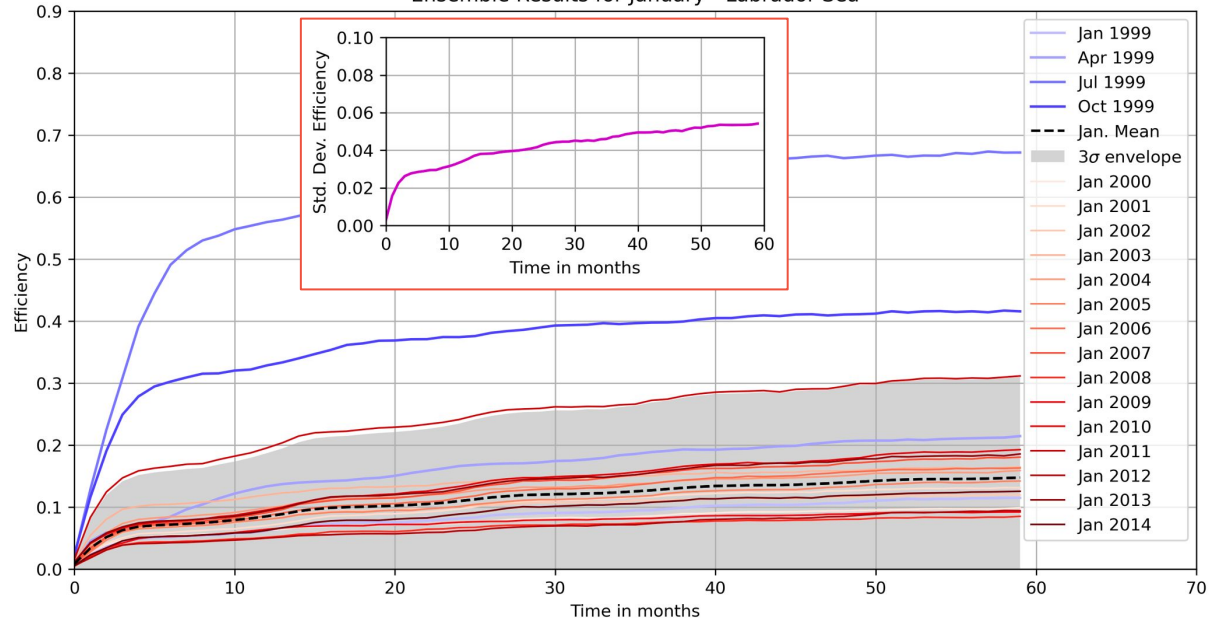
Ensemble Results for January - N. Pacific



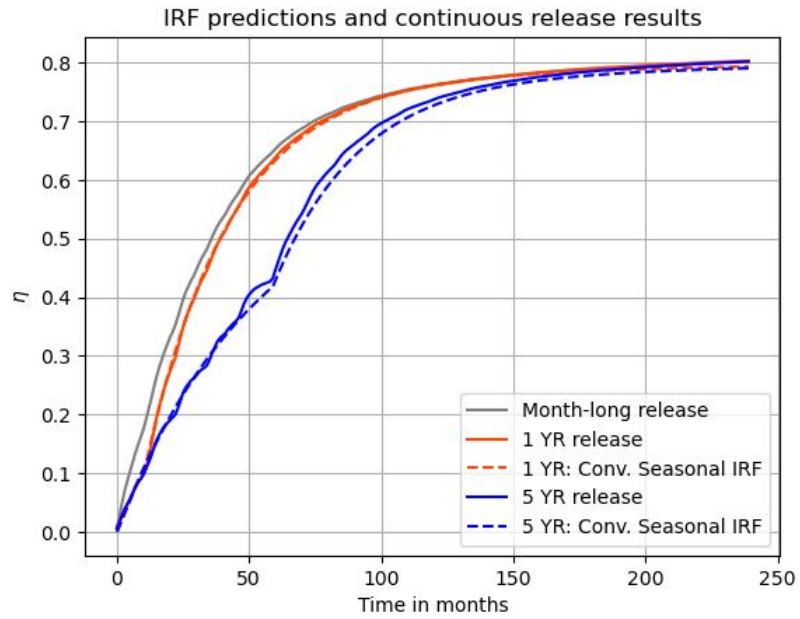
Ensemble Results for January - North Sea



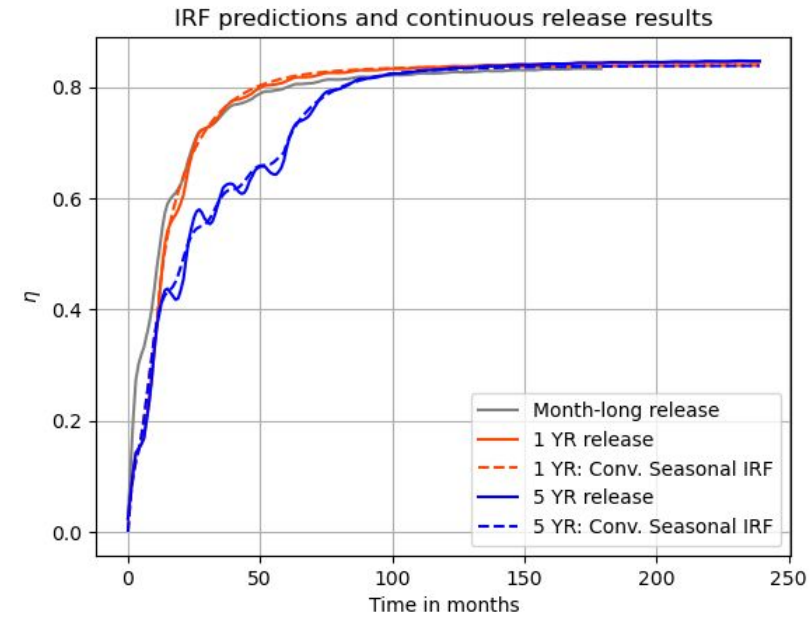
Ensemble Results for January - Labrador Sea



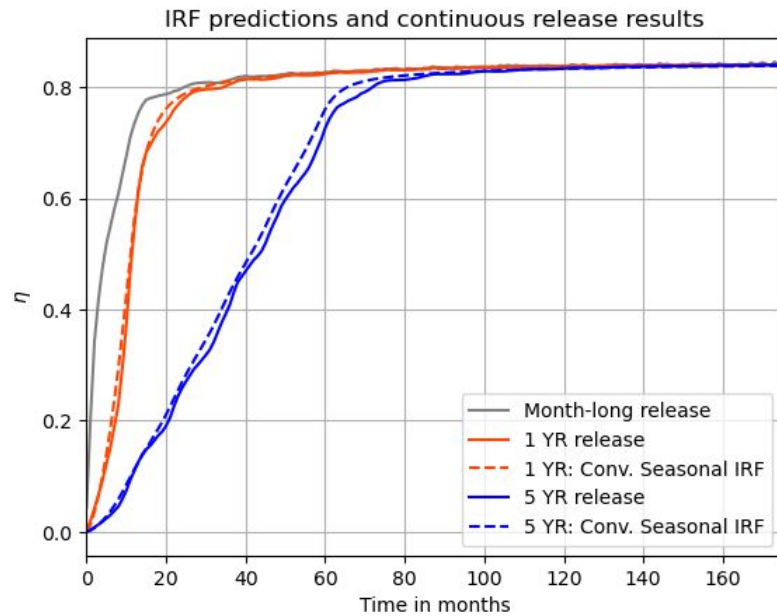
Western Eq. Pacific



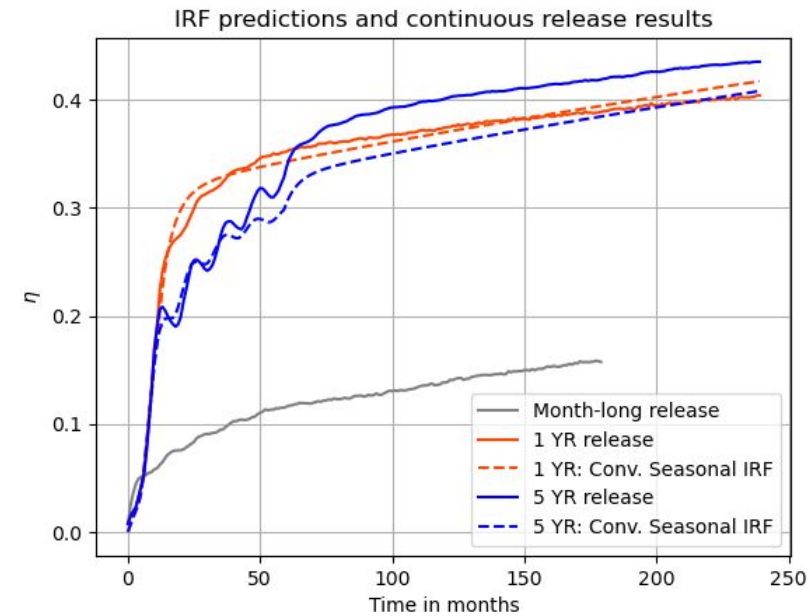
North Pacific



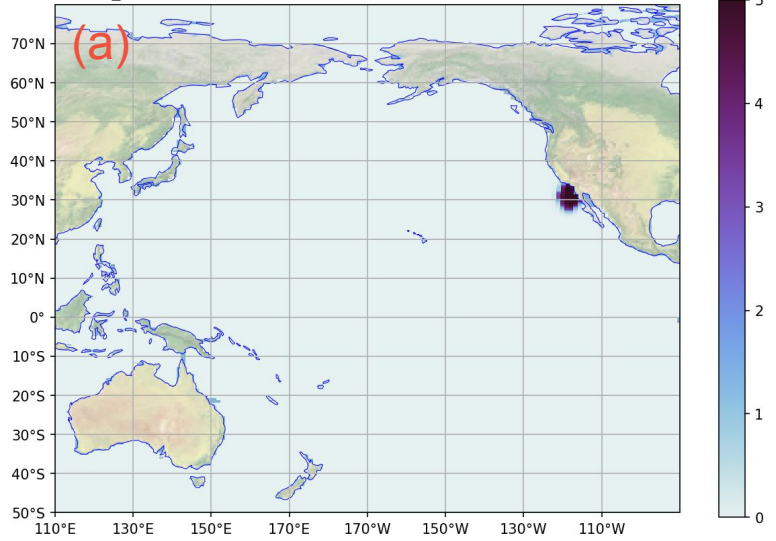
North Sea



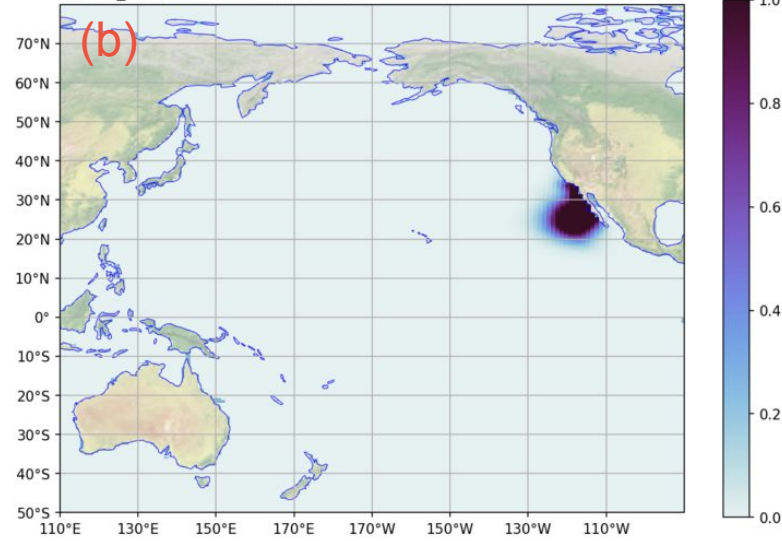
Labrador Sea



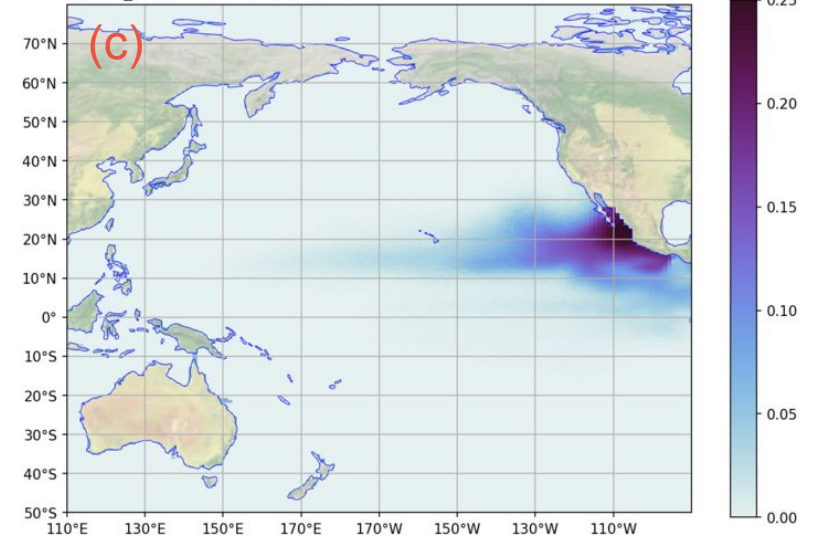
excess_ALK at surface 0347-02-01



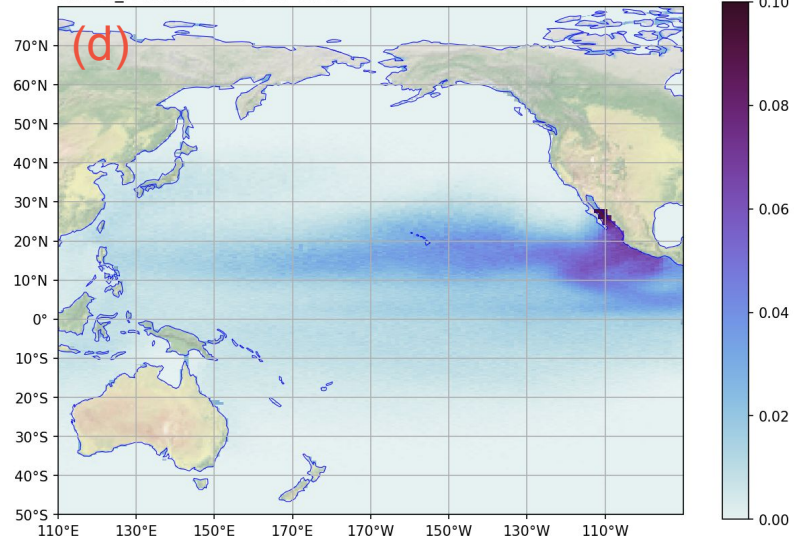
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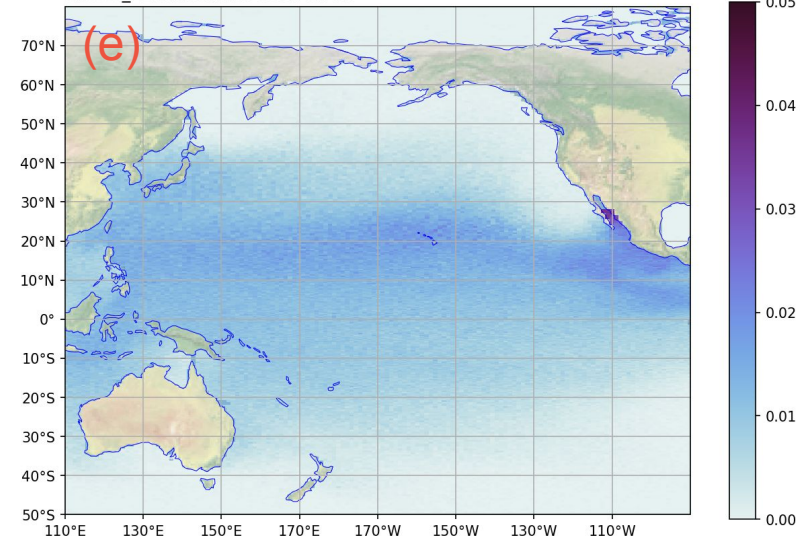
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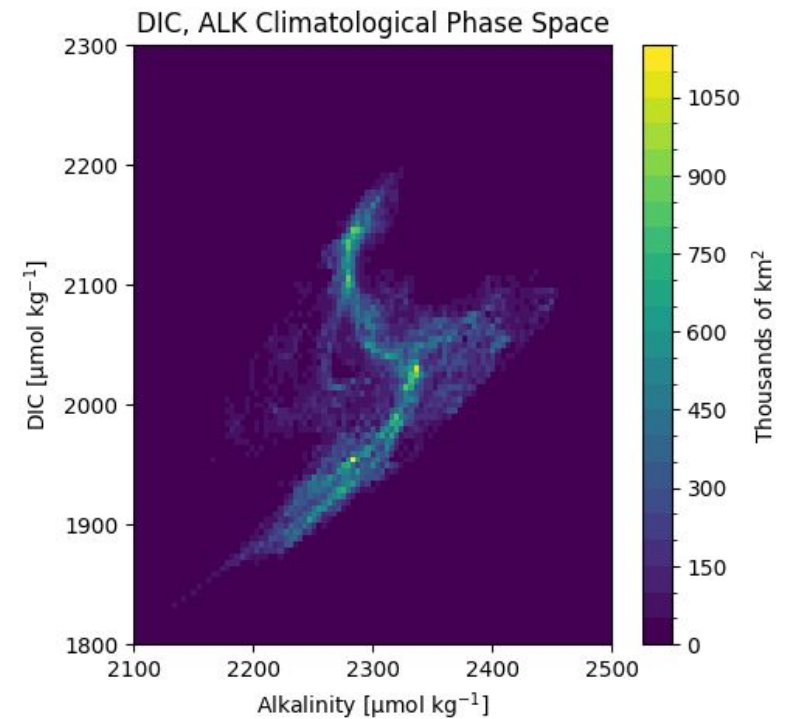
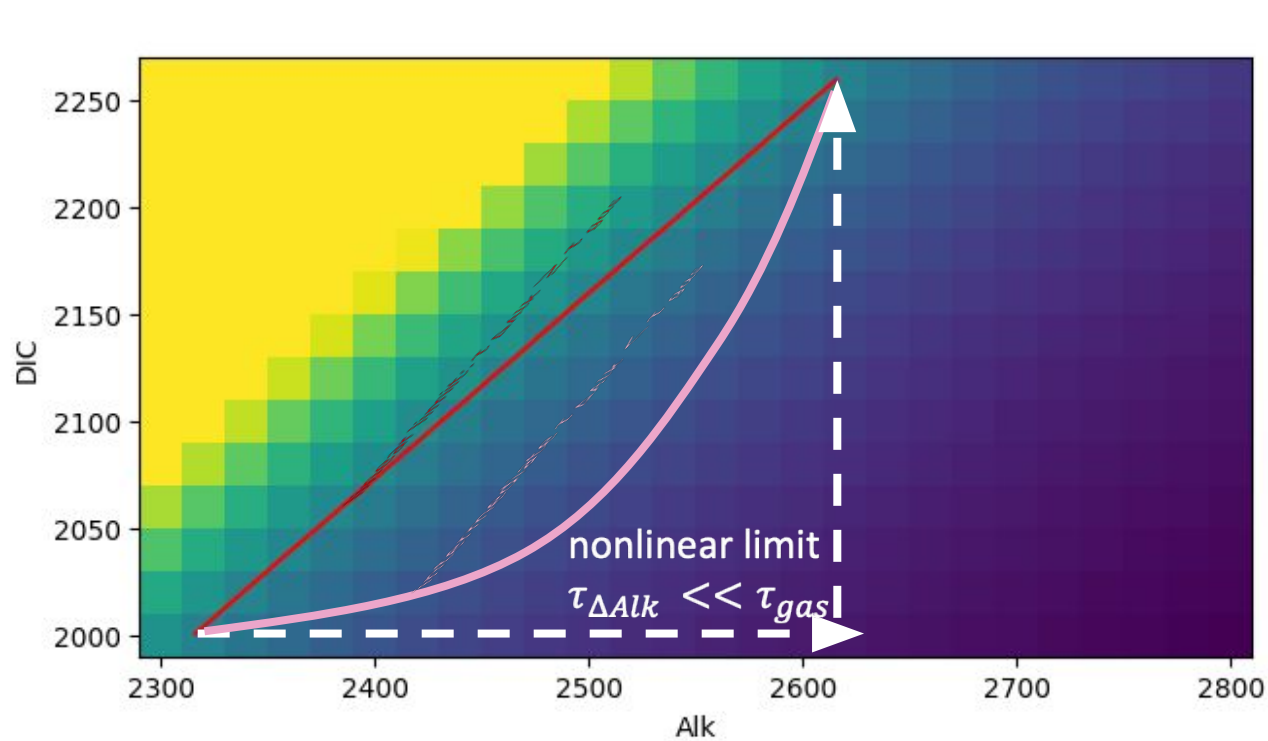


excess_ALK at surface 0357-02-01



excess_ALK at surface 0362-01-01

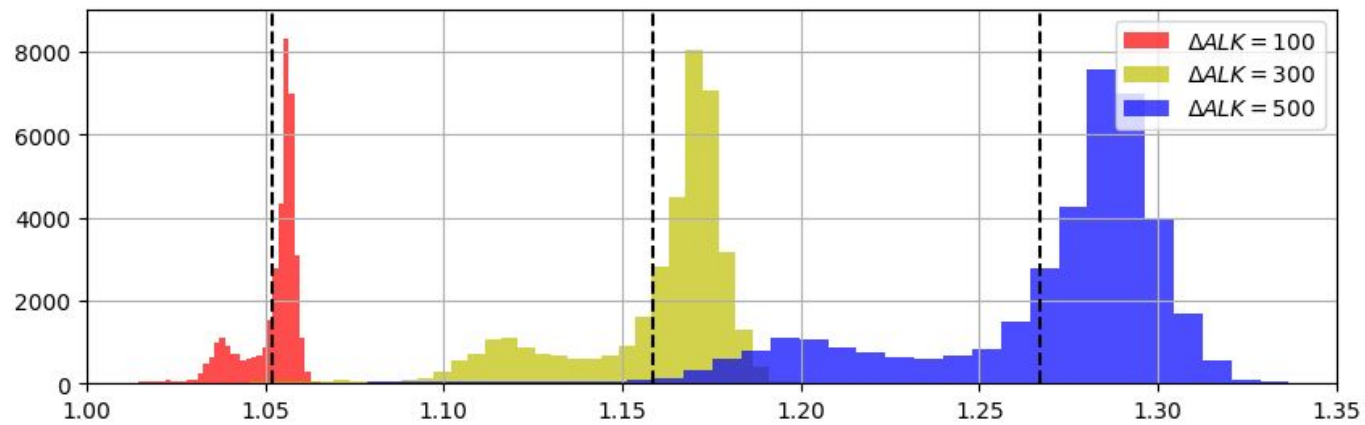
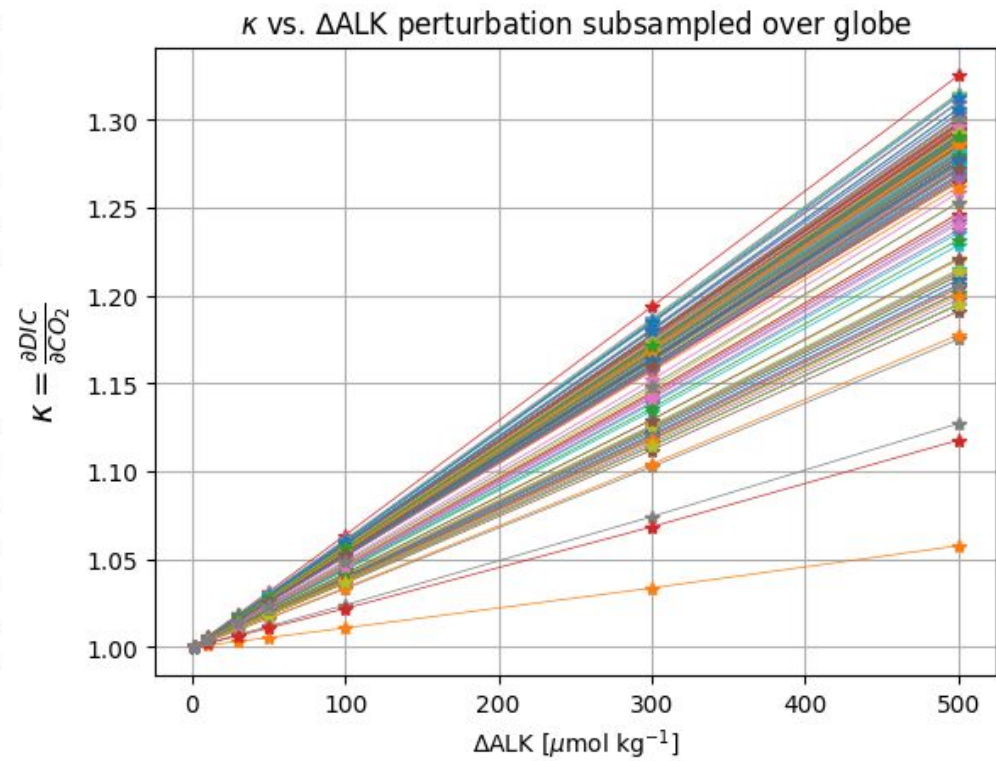
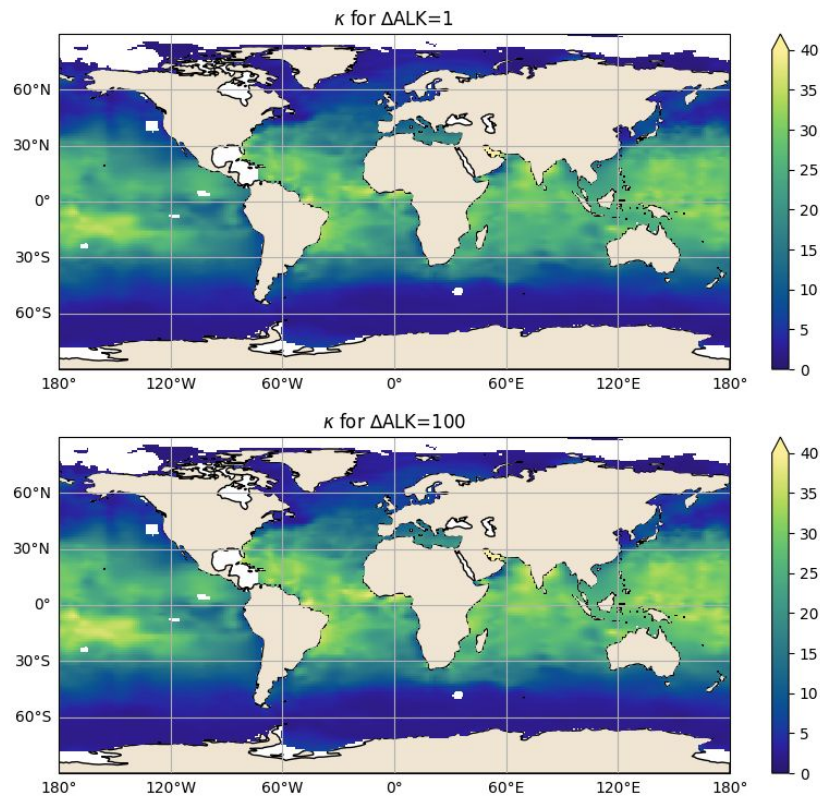




$$\frac{\partial pCO_2}{\partial DIC} = \frac{K_2}{K_0 K_1} \frac{(3 \cdot Alk - 2 \cdot DIC)(2 \cdot DIC - Alk)}{(Alk - DIC)^2} = \frac{K_2}{K_0 K_1} \kappa$$

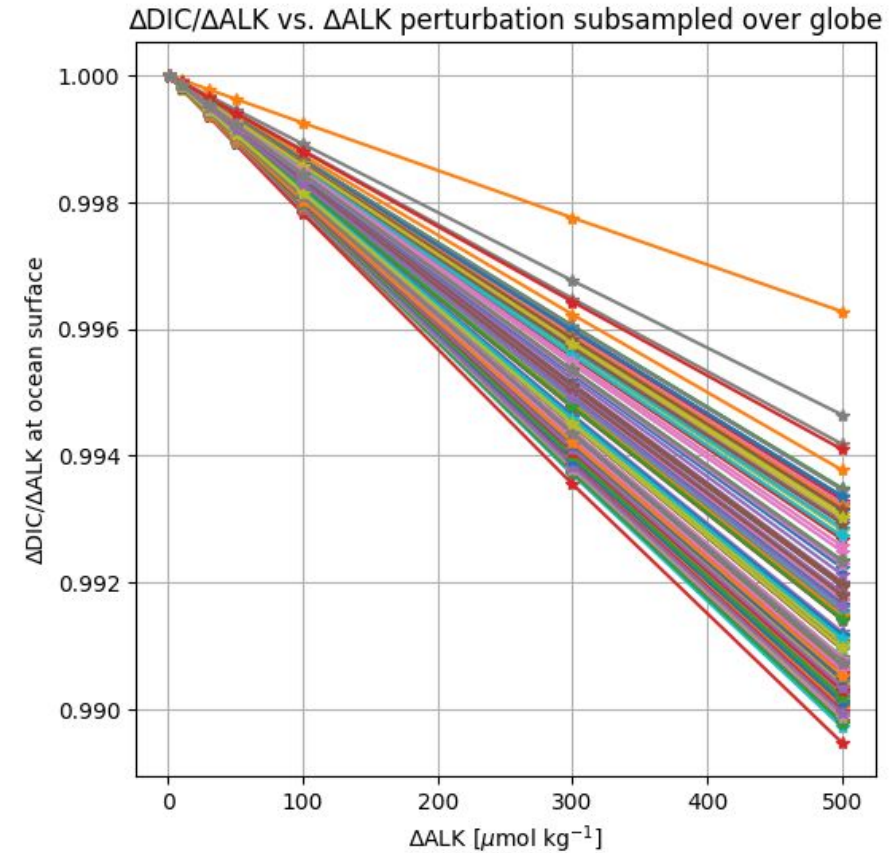
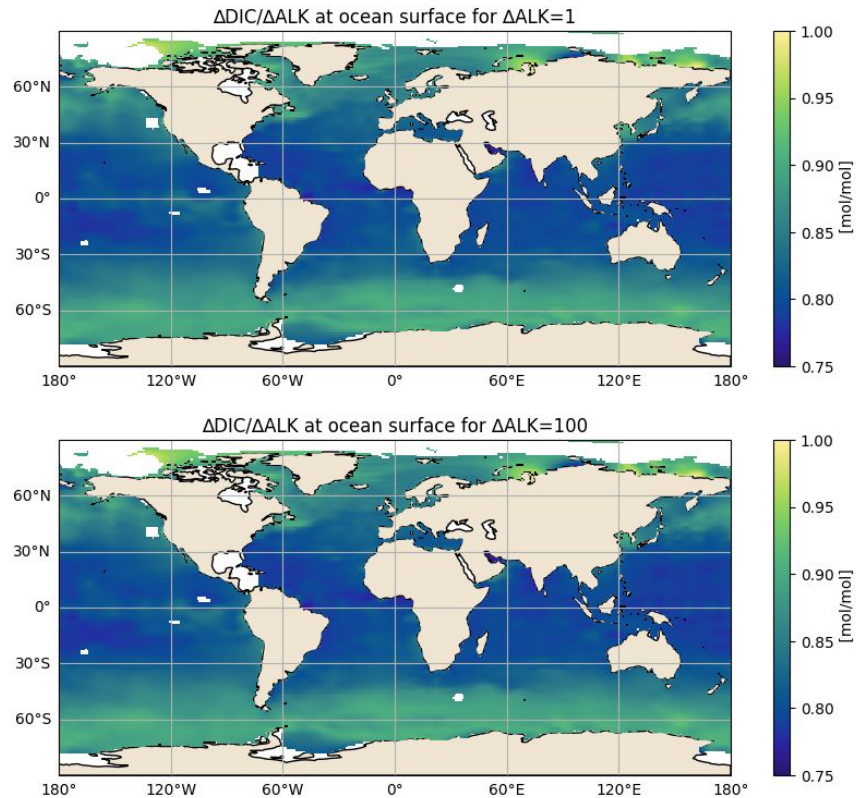
κ encapsulates the nonlinearity due to carbonate chemistry, refers to the change in how CO_2 will be taken up as a function of DIC for a given atmospheric CO_2 concentration. If we compute the IRF for a given alkalinity injection, as long as our release's κ remains close in value, the system remains





Depending on the magnitude of the alkalinity pulse, we may violate the linearity.





Goal is to show that η_{max} does not depend significantly on $|\Delta\text{Alk}|$. This means linearity with respect to the magnitude of the alkalinity pulse.

