

# The relationship between condensate lifetime and precipitating efficiency and their response to sea surface warming



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# Precipitating efficiency is a key metric connecting microphysics to its large scale environment

Precipitating efficiency is defined as:

$$\varepsilon = \frac{\textit{Precipitation}}{\textit{Total condensed water}} = \frac{P}{P + E}$$

*P*: Precipitation rate  
*E*: Hydrometeor evaporation rate

Where *P* is the mean tropical precipitation rate and *E* is the mean re-evaporation of condensate

1. Romps (2014) – PE can modulate the environment RH in the tropics
2. PE can modulate convective mass flux when radiative cooling is constrained – Jeevanjee and Zhou (2022) and Emanuel (2019)
3. Zhao et al. (2016) cmip spread in climate sensitivity can be reproduced by tuning a single model's deep convection  $\varepsilon$ .

$\varepsilon$  may encapsulate micro2macro interactions especially in the deep tropics!

# The inverse cloud lifetime as a proxy for PE?

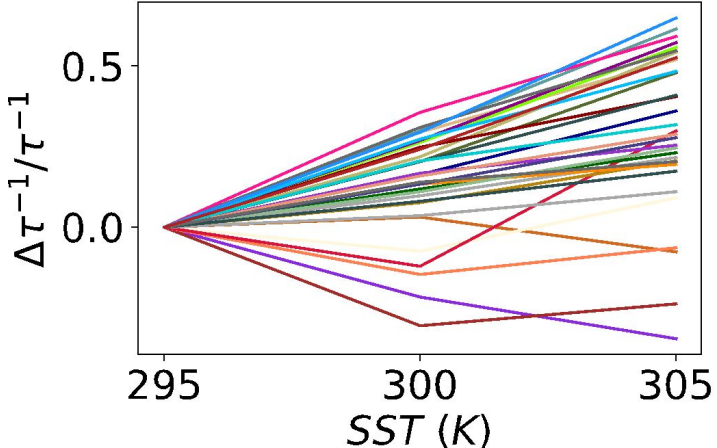
$$\varepsilon = \frac{P}{P + E}$$

Precipitating efficiency is difficult to measure and not widely available from CRM model output. An alternative that has been proposed is (Li et al., 2022):

$$\tau^{-1} = \frac{\text{Precipitation}}{\text{Total condensed water path}} = \frac{P}{TWP}$$

$\tau^{-1}$  is easy to measure from observations and standard model output

Remarkably, 32 out of 36 RCEMIP models predict an increase in  $\tau^{-1}$  with SST (Li et al., 2023):



**Does a  $\tau^{-1}$  increase indicate a  $\varepsilon$  increase? And what processes are responsible for the  $\tau^{-1}$  increase?**



# Relationship between $\varepsilon$ and lifetime and their constituent processes

$$\varepsilon = \frac{P}{P + E} = \frac{\frac{P}{TWP}}{\frac{P}{TWP} + \frac{E}{TWP}} = \frac{\tau^{-1}}{\tau^{-1} + \tau_e^{-1}}$$

Here,  $\tau_e^{-1}$  can be thought of as the inverse evaporation timescale

This equation shows that  $\varepsilon = f(\tau^{-1}, \tau_e^{-1})$  and while the former can be measured, the latter is more elusive.

How do  $\tau^{-1}$  and  $\tau_e^{-1}$  relate to microphysical inverse timescales?

$P = \text{Precip Production} - \text{Precip Evaporation}$

$E = \text{Cloud Evaporation} + \text{Precip Evaporation}$

$$\tau^{-1} = \frac{\tau_a^{-1} \tau_{sed}^{-1}}{\tau_a^{-1} + \tau_{sed}^{-1} + \tau_{pe}^{-1}}$$

$$\tau_e^{-1} = (1 - f_p) \tau_{ce}^{-1} + f_p \tau_{pe}^{-1}$$

$$f_p = \frac{\tau_a^{-1}}{\tau_a^{-1} + \tau_{sed}^{-1} + \tau_{pe}^{-1}}$$

$\tau_a^{-1}$ : inverse time scale for cloud to rain conversion

$\tau_{ce}^{-1}$ : inverse time scale for cloud evaporation

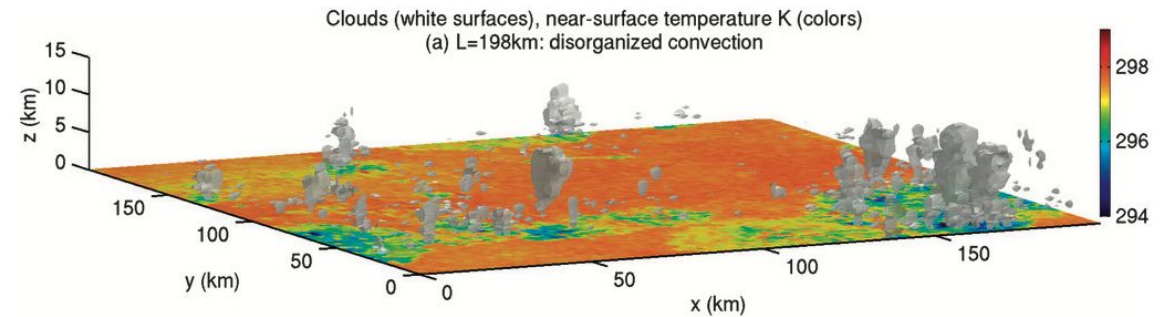
$\tau_{sed}^{-1}$ : inverse time scale for precip sedimentation

$f_p$ : fraction of condensate in precip phase

$\tau_{pe}^{-1}$ : inverse time scale for precip evaporation

# RCE simulations

- Unaggregated small domain RCE simulations. 100 km x 100 km domain with 1 km resolution
- In house SCREAM simulations with SST range of 290 K to 310K
- SAM simulations from Lutsko and Cronin (2018) – three ensemble members with different mp schemes
- Minimal recipe simulations from Jeevanjee and Zhou (2021) with the GFDL model – two ensemble members with and without accretion



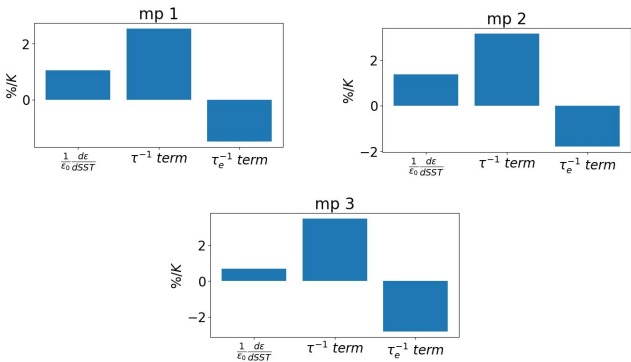
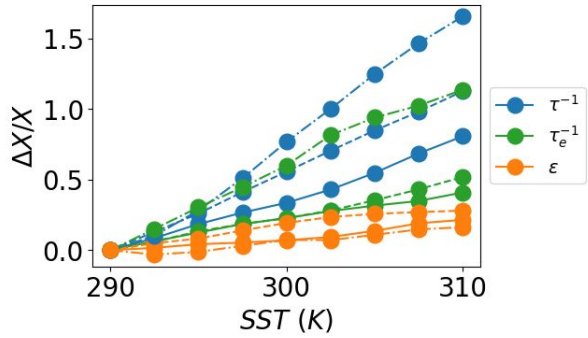
3D contour plot of an RCE simulation from Muller and Held (2014)

- For all simulation pairs, we use domain mean averages and apply the chain rule on each term within the four previously derived equations to understand SST trends.

# $\varepsilon$ breakdown

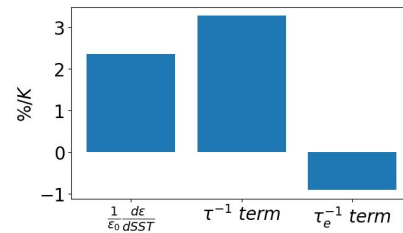
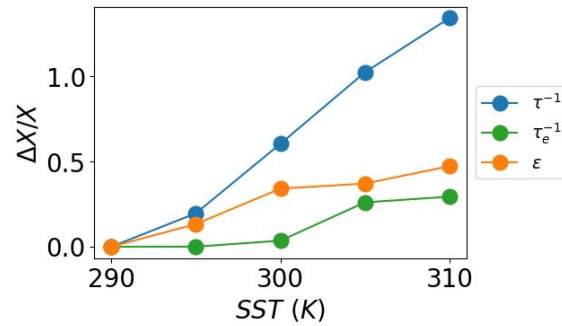
$$\varepsilon = \frac{\tau^{-1}}{\tau^{-1} + \tau_e^{-1}}$$

## SAM



$\tau^{-1}$  increase drives  $\varepsilon$  increase for all mp schemes.

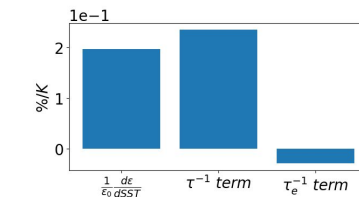
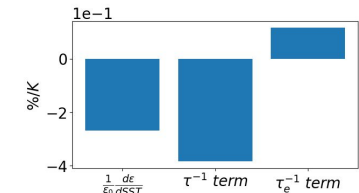
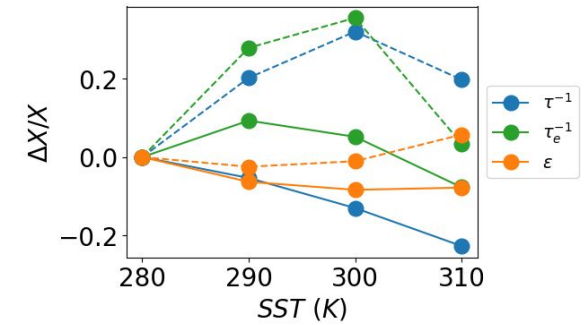
## SCREAM



$\tau^{-1}$  increase drives  $\varepsilon$  increase

SAM and SCREAM agree that  $\tau^{-1}$  increase drives  $\varepsilon$  increase however increase in  $\tau_e^{-1}$  modulates  $\varepsilon$  change

## GDFL-minimal

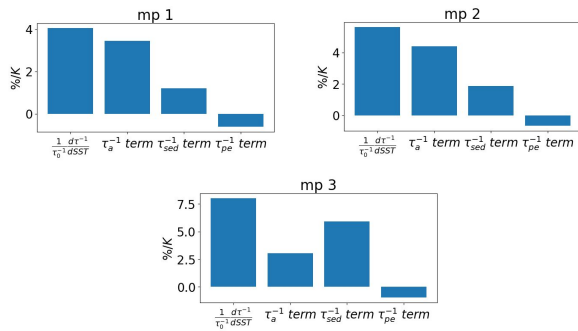
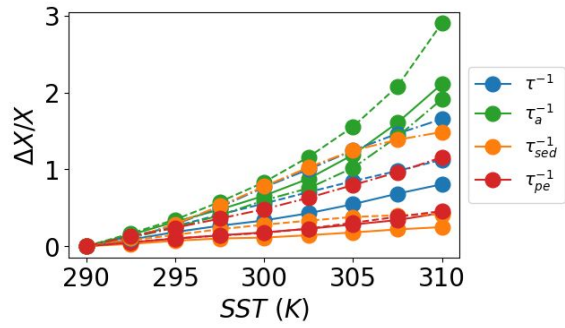


No  $\varepsilon$  change

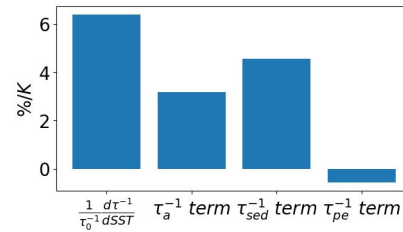
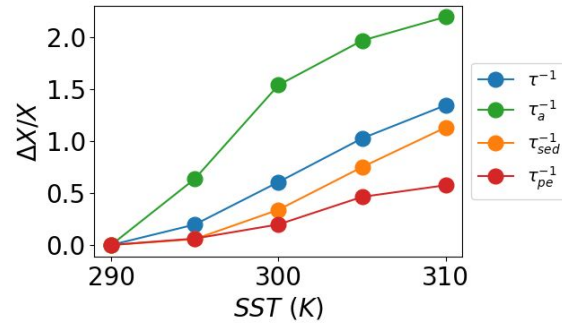
# $\tau^{-1}$ breakdown

$$\tau^{-1} = \frac{\tau_a^{-1} \tau_{sed}^{-1}}{\tau_a^{-1} + \tau_{sed}^{-1} + \tau_{pe}^{-1}}$$

## SAM

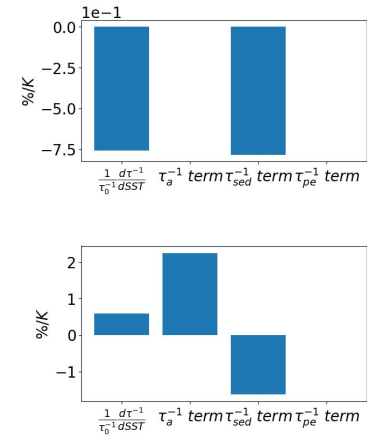
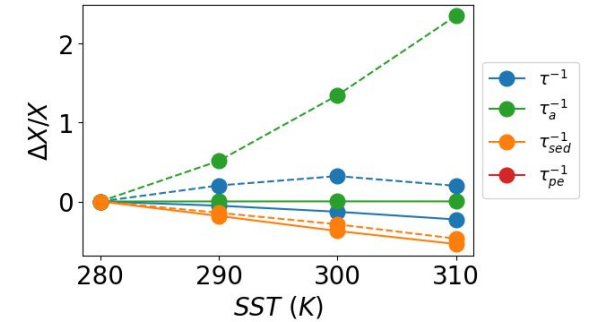


## SCREAM



$\tau^{-1}$  increase driven by  $\tau_a^{-1}$  and  $\tau_{sed}^{-1}$

## GDFL-minimal



Minimal recipe predicts decrease in  $\tau_{sed}^{-1}$  as clouds move upwards

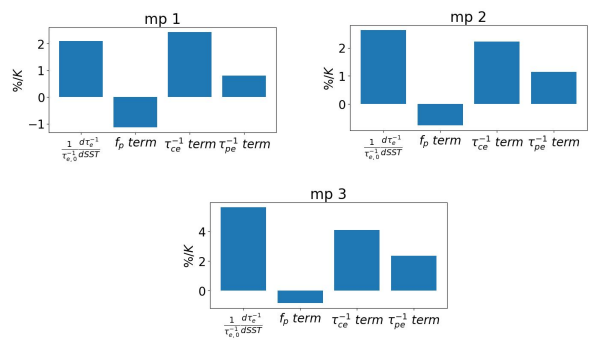
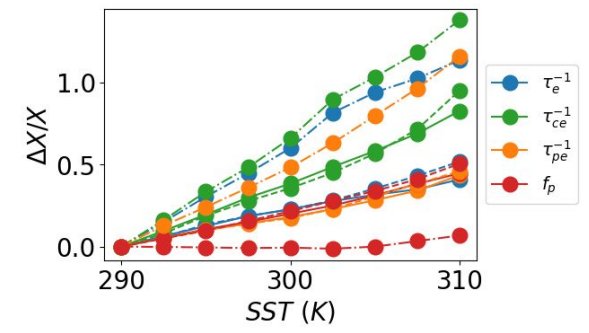
$\tau^{-1}$  increase driven by  $\tau_a^{-1}$  and  $\tau_{sed}^{-1}$

SAM and SCREAM agreement: clouds become denser and rain drops fall faster

# $\tau_e^{-1}$ breakdown

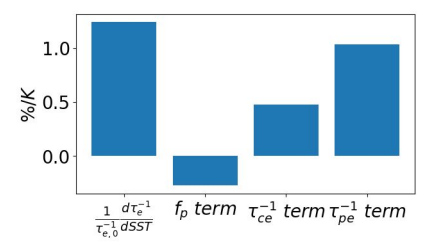
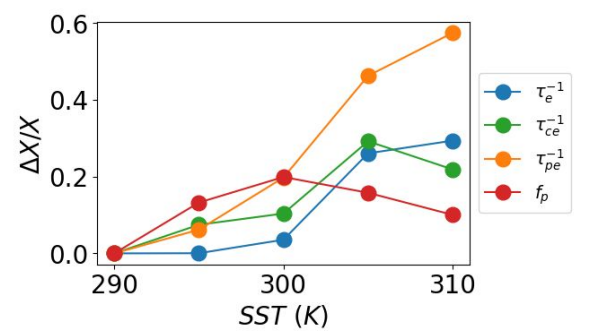
$$\tau_e^{-1} = (1 - f_p)\tau_{ce}^{-1} + f_p\tau_{pe}^{-1}$$

## SAM



$\tau_e^{-1}$  increase driven by  $\tau_{ce}^{-1}$  and  $\tau_{pe}^{-1}$

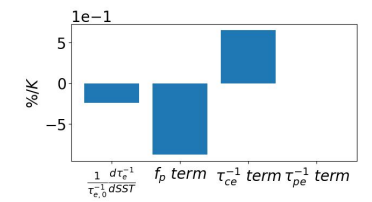
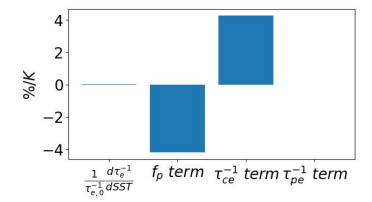
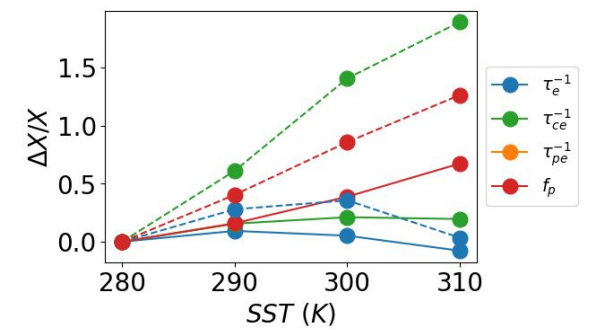
## SCREAM



$\tau_e^{-1}$  increase driven by  $\tau_{ce}^{-1}$  and  $\tau_{pe}^{-1}$

SAM and SCREAM agreement: clouds and rain drops evaporate quicker likely due to increased saturation deficit

## GDFL-minimal



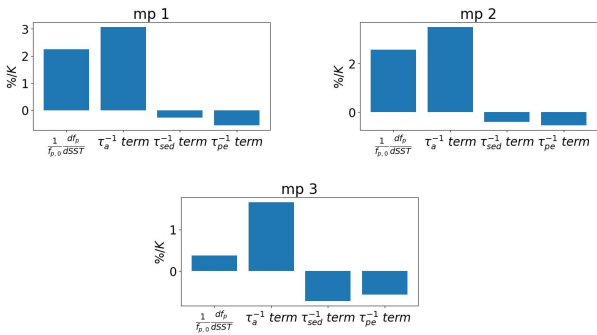
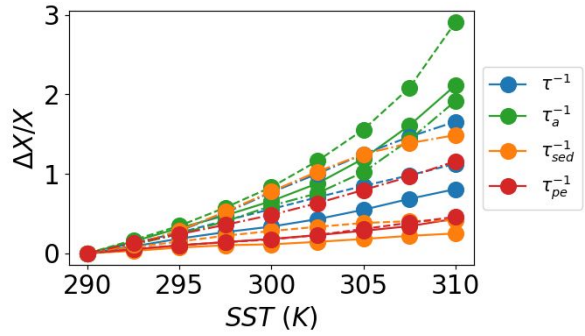
Minimal sees  $\tau_{ce}^{-1}$  increase but  $f_p$  increase counteracts



# $f_p$ breakdown

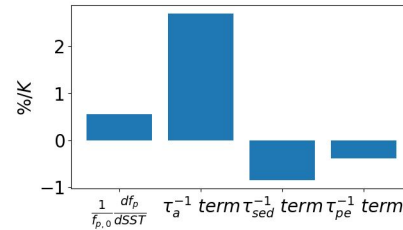
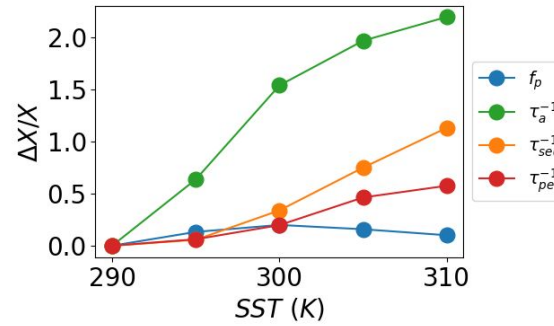
$$f_p = \frac{\tau_a^{-1}}{\tau_a^{-1} + \tau_{sed}^{-1} + \tau_{pe}^{-1}}$$

## SAM



$f_p$  increase driven by  $\tau_a^{-1}$  increase

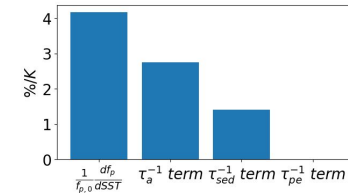
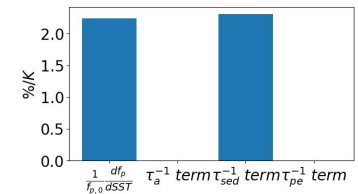
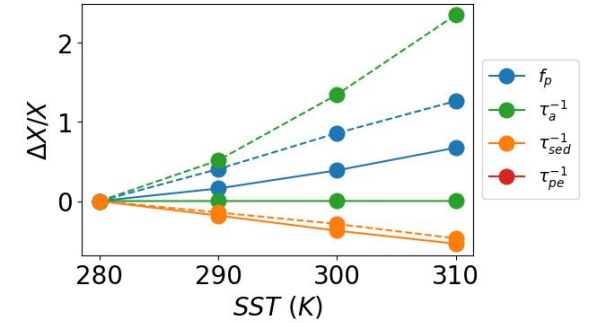
## SCREAM



$f_p$  increase driven by  $\tau_a^{-1}$  increase

SAM and SCREAM agreement:  $f_p$  increases slightly as all terms increase

## GDFL-minimal



$f_p$  increases more substantially since  $\tau_{sed}^{-1}$  decreases

# Conclusions

- $\tau^{-1}$  is not a proxy for  $\varepsilon$  but a component of it. While  $\tau^{-1}$  qualitatively drives the increase in  $\varepsilon$ , the scaling is quite different.
- $\tau^{-1}$  is still an interesting metric to examine and simulation results + diagnostic framework allow us to understand why its increasing with SST. Potential explanation for model consensus is that warmer atmospheres tend to convert clouds to rain more rapidly and create faster falling rain drops – consistent with intuition.
- Future work – as well as conversations at this meeting – should look into how best to leverage observations to affirm  $\tau^{-1}$  trends.



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