

Decomposition of CRE interannual variability with the FluxByCldTyp dataset

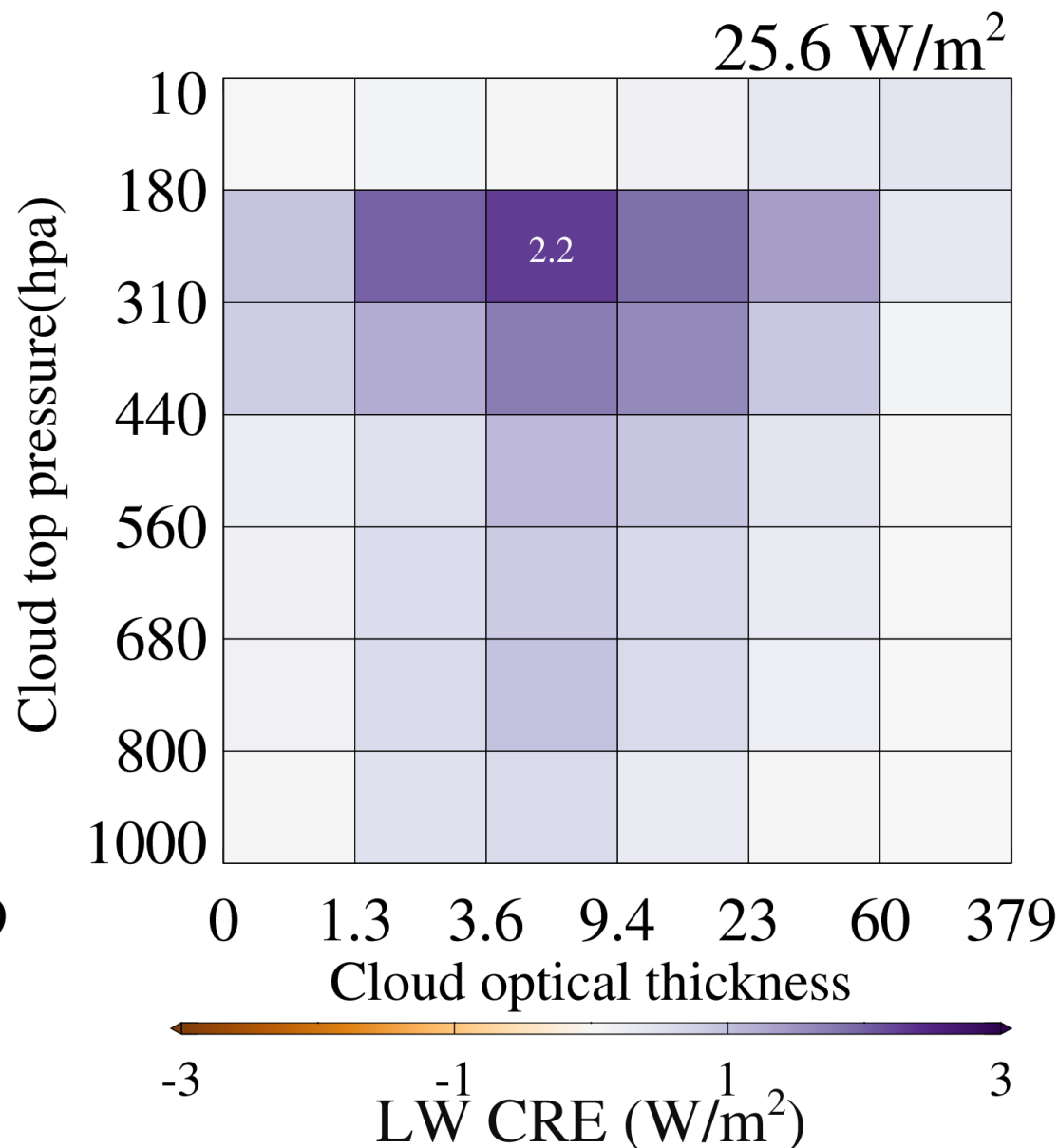
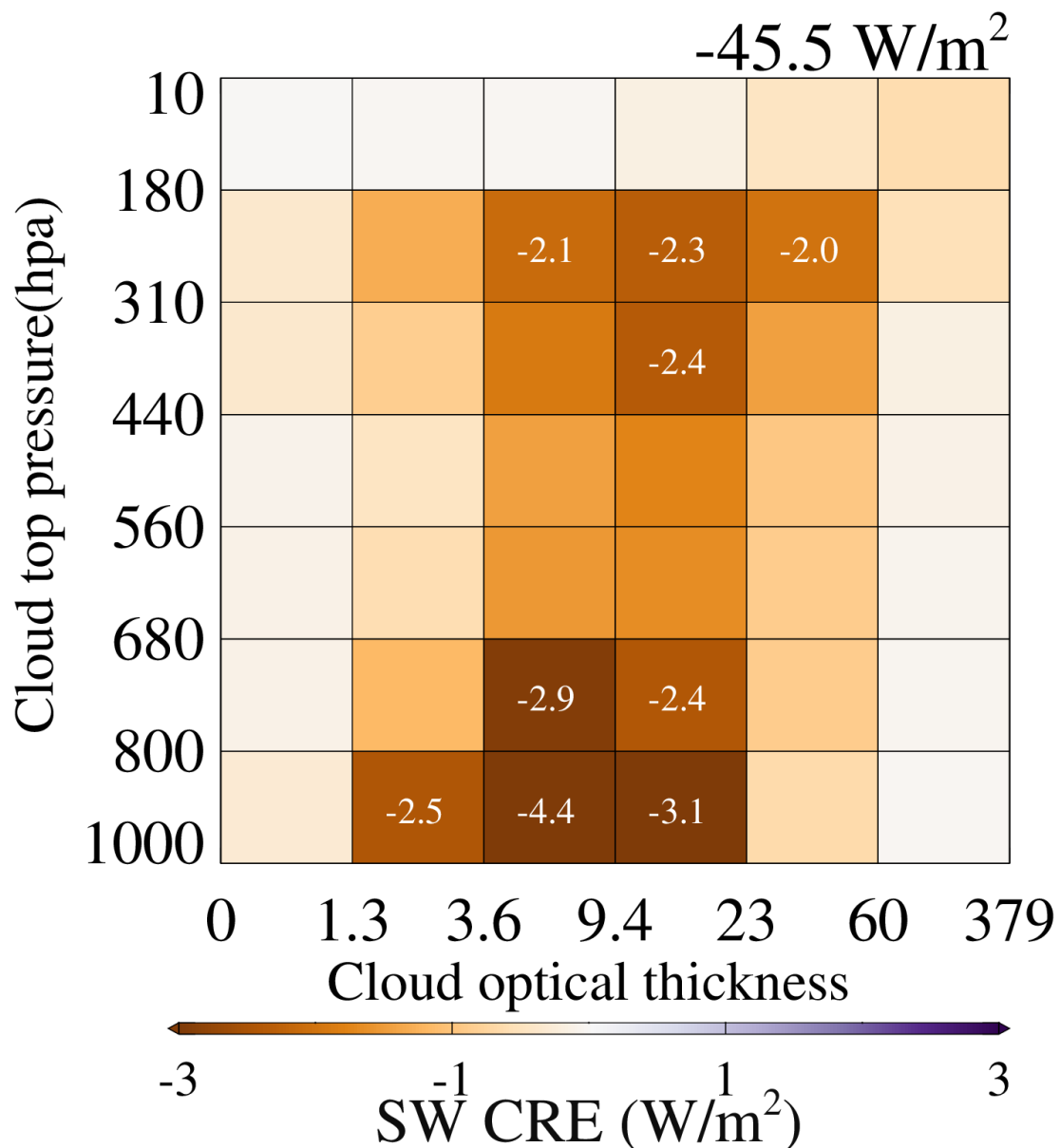
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Why FluxByCldTyp (FBCT)?

- Resolves the (subgrid) details of $CRE = F_{all-sky} - F_{clr}$
- $1^\circ CRE = F_{all-sky} - F_{clr}$ comes from contribution of many clouds within a grid cell and FBCT resolves these subgrid details.
- Provides a CRE array $CRE(i), i = 1, 42$ (7 CTP and 6 TAU bins)
 - GCMs typically do not provide subgrid CRE
- Allows us to examine the contribution of different cloud “types” (as defined by $CTP - TAU$ combinations) to CRE at various spatiotemporal scales.
- See for example the global SW and LW $\overline{CRE}(i)$ from monthly FBCT...

Global SW and LW $CRE(i)$ from monthly FBCT, 2003-2019

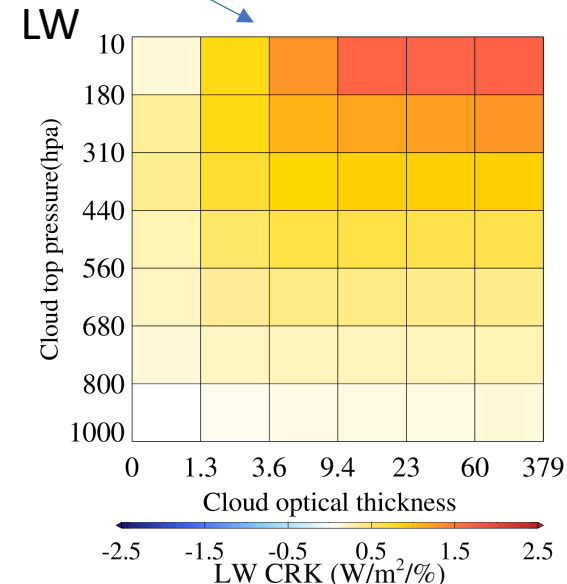
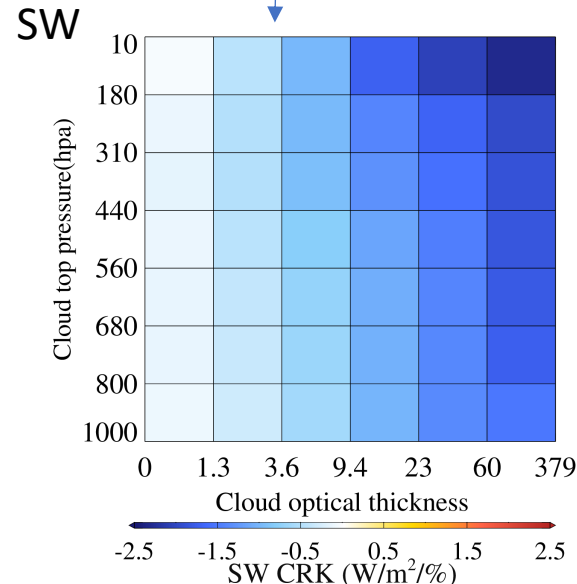
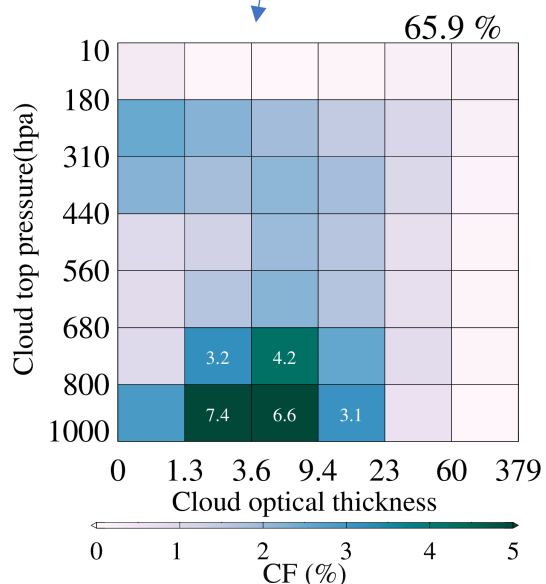


Cloud Radiative Kernels

- Following Zelinka et al. (2012) one can view $CRE(i)$ as the product of cloud fraction array $C(i), i = 1, 42$ and a “Cloud Radiative Kernel” CRK ($K(i)$)

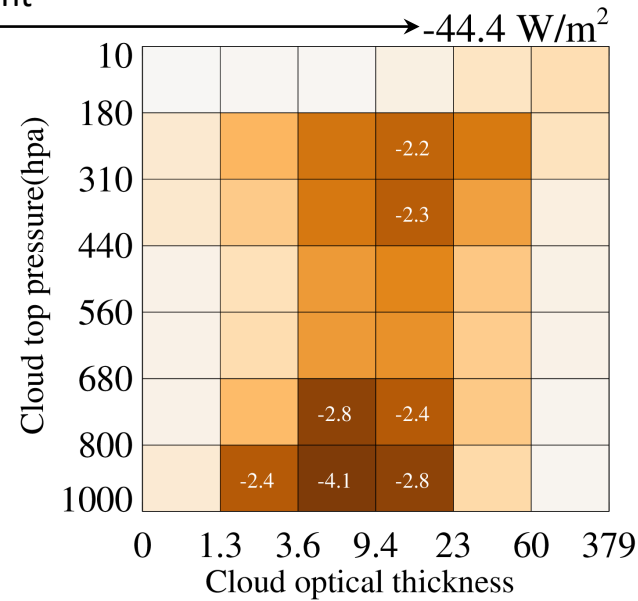
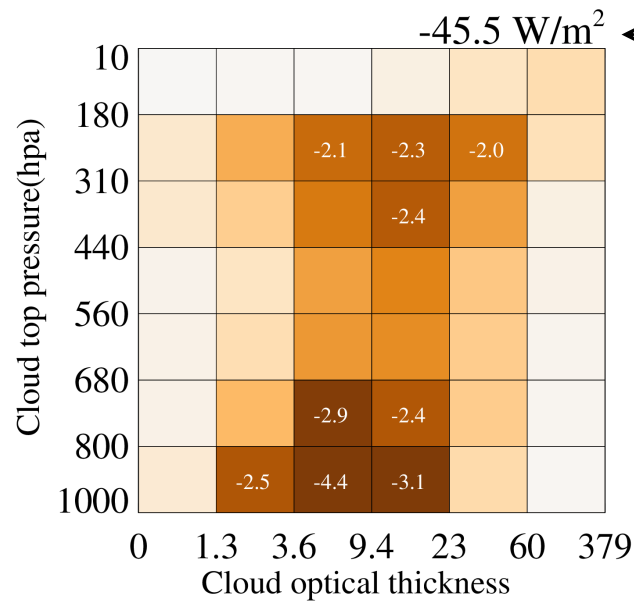
$$K(i) \equiv (F_{ovc}(i) - F_{clr}(i))/100 \text{ (often assuming that } F_{clr}(i) = \text{grid cell mean } F_{clr}\text{)}$$

- $CRE(i) = C(i) \times K(i)$
- But a single *global* $\overline{C(i)}$ multiplied by a single *global* $\overline{K(i)}$ does not give correct $\overline{CRE(i)}$
- Resolving $K(i)$ by latitude and month helps: obtain then global $\overline{CRE(i)}$ by averaging zonal values of $C(i) \times \overline{K(i)}$, $\overline{CRE(i)} \approx \overline{C(i) \times \overline{K(i)}}$
- One expects $K(i)$ to be somewhat sensitive to binning choice
- The Zelinka *CRKs* came from off-line RT computations; FBCT provides observed *CRKs* (see also Sun et al. 2022)

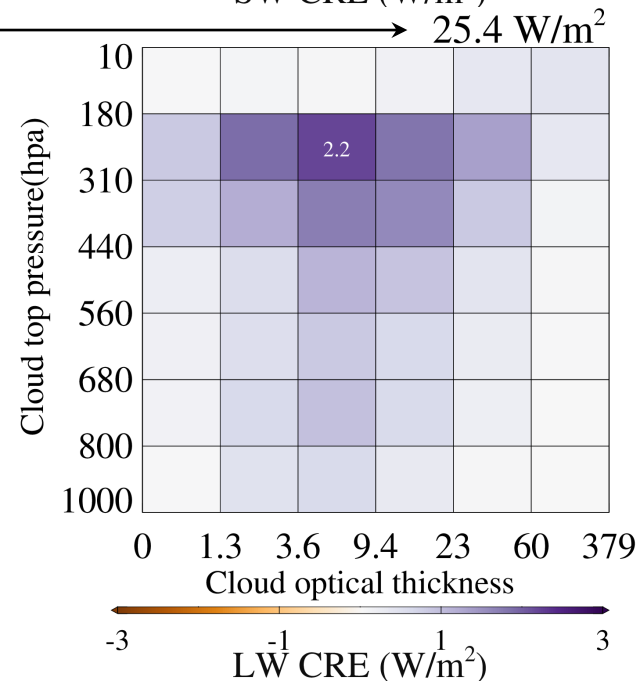
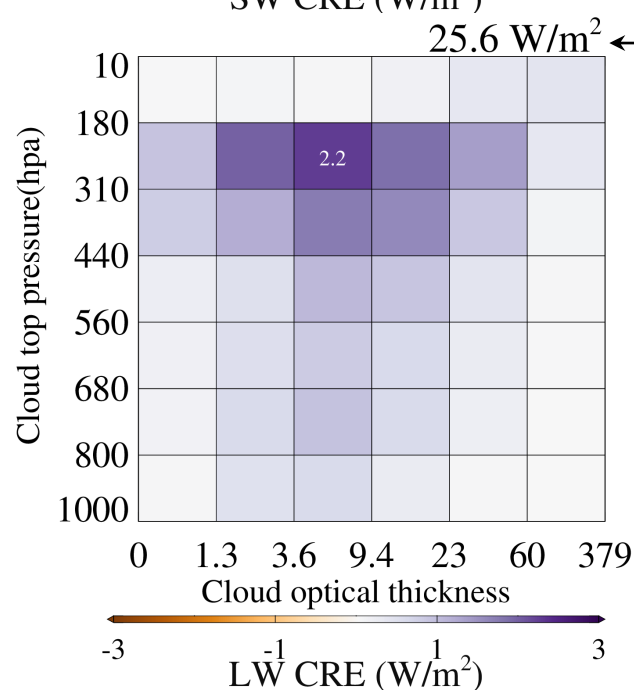


FBCT

Directly

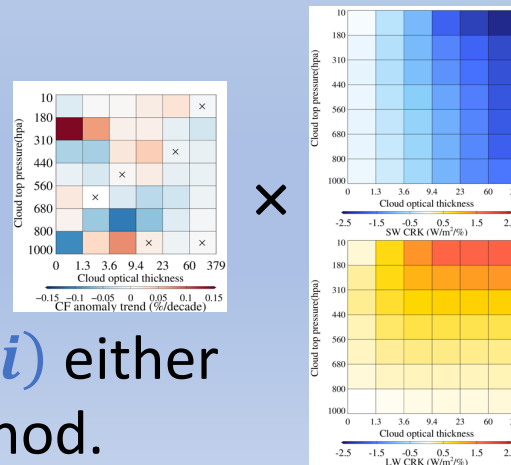


From kernels



Why is this concept useful?

- We can examine *why* CRE changes.
- The CRK is meant to be a fundamental property capturing the sensitivity of CRE to cloud changes for given certain atmospheric and surface conditions.
- To first order, ΔCRE anomalies may be driven by changes in the cloud fraction array $C(i)$ i.e., cloud amount changes within $CTP - TAU$ bins which reflect changes in C_{tot} , and/or mean CTP, TAU .
- It is thus tempting to interpret $CRE(i)$ anomalies $\Delta CRE(i)$ in terms of $C(i)$ anomalies $\Delta C(i)$: $\Delta CRE(i) \approx \Delta C(i) \times K(i)$.

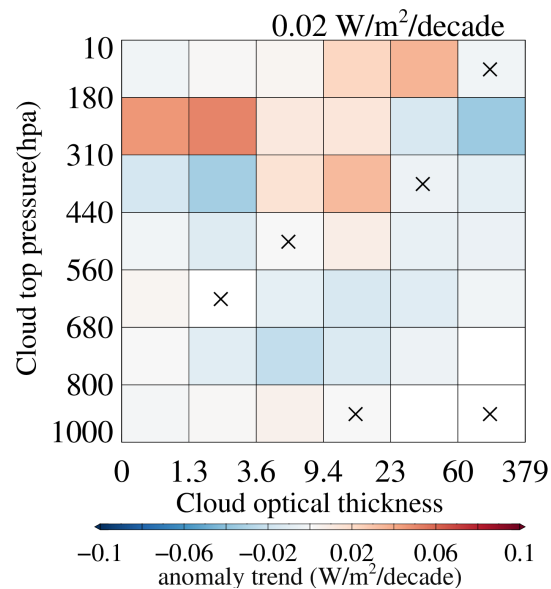
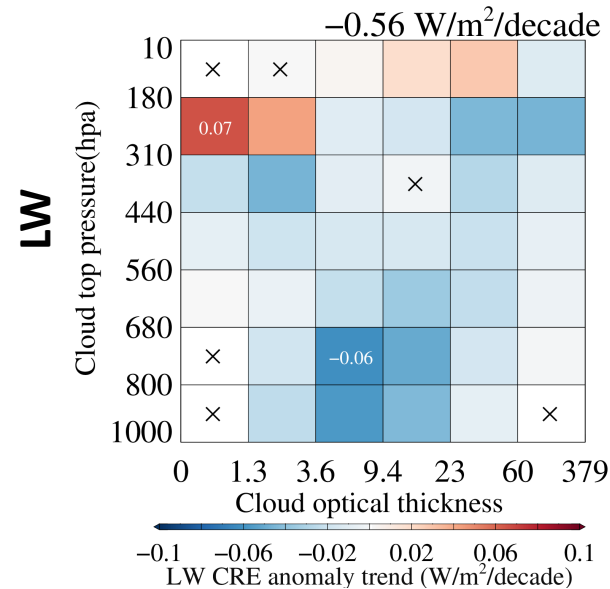
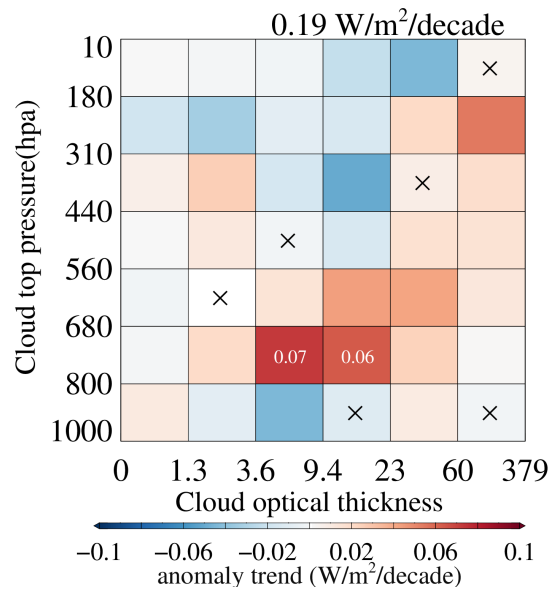
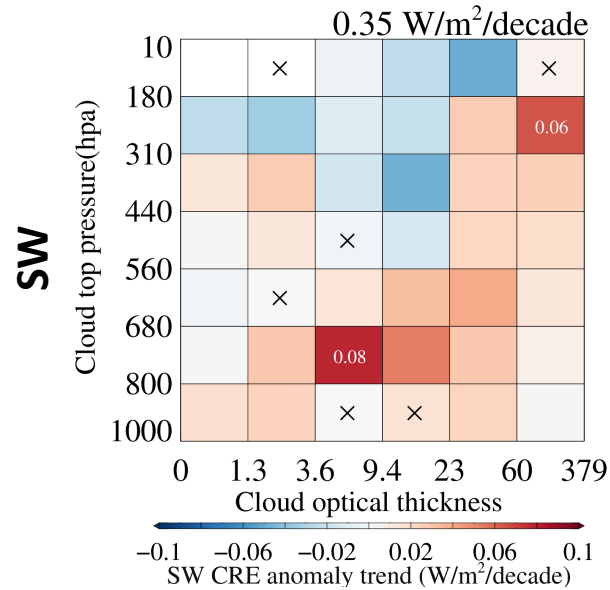


- FBCT allows us to calculate $\Delta CRE(i)$ either directly, or through the kernel method.

directly calculated
 ΔCRE

≠

from cloud anomaly
 $\Delta C \times \bar{K}$



While $\widetilde{CRE}(i) \approx \widetilde{C}(i) \times \widetilde{K}(i)$, the first-order approximation $\Delta \widetilde{CRE}(i) \approx \Delta \widetilde{C}(i) \times \widetilde{K}(i)$ does not appear to hold.

$\bar{K}(i)$ changes need to be taken into account!

(simplified notation)

$$\Delta CRE = \underbrace{\Delta C \times \bar{K}}_{\text{cloud anomaly}} + \underbrace{\bar{C} \times \Delta K}_{\text{kernel anomaly}} + \text{covariance}$$

directly calculated

ΔCRE

from cloud anomaly + kernel anomaly (no covariance)

$\Delta C \times \bar{K} + \bar{C} \times \Delta K$

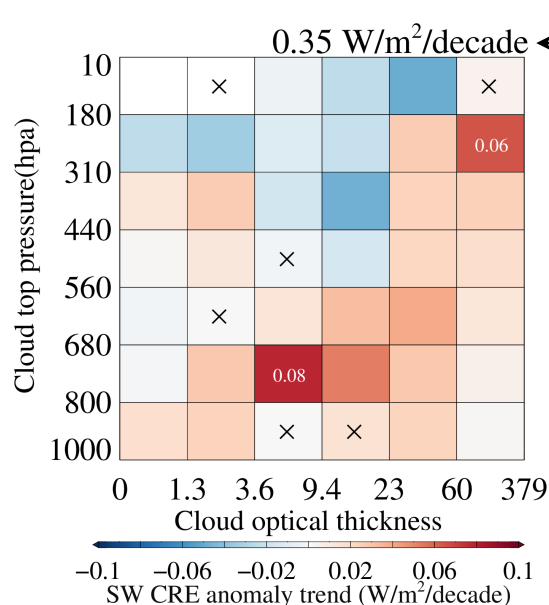
cloud anomaly term

$\Delta C \times \bar{K}$

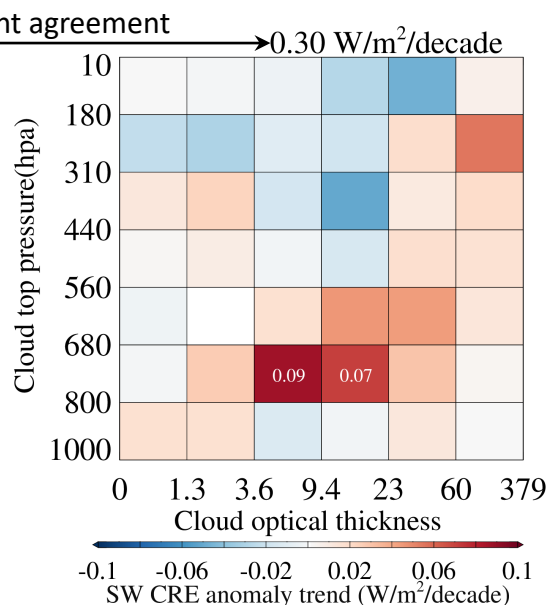
kernel anomaly term

$\bar{C} \times \Delta K$

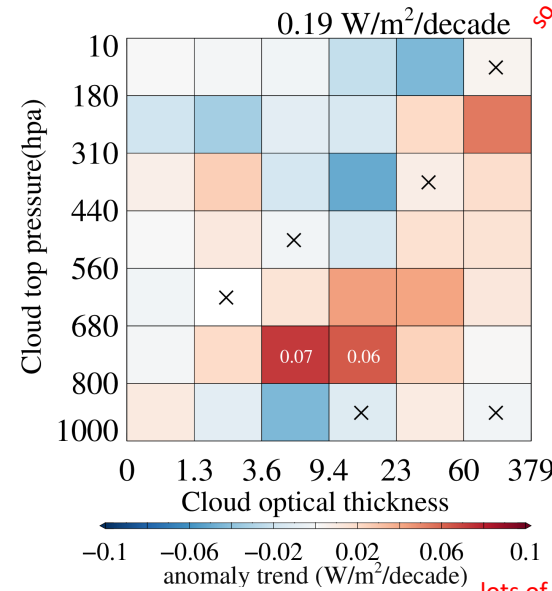
SW



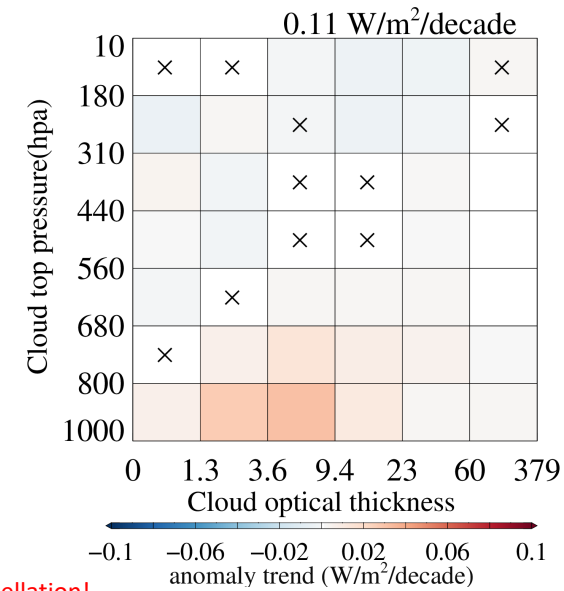
decent agreement



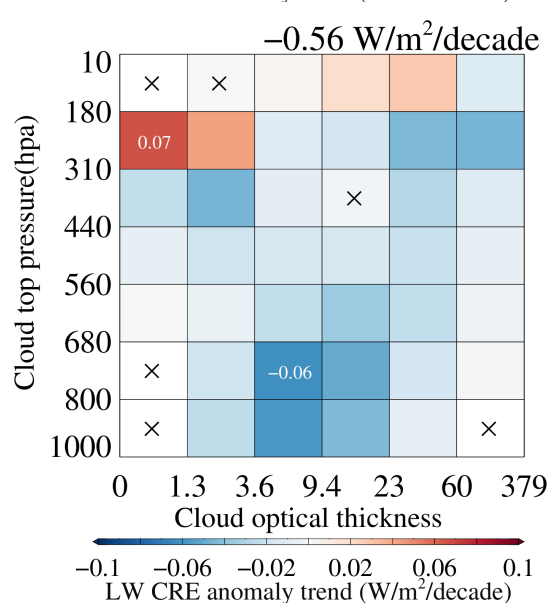
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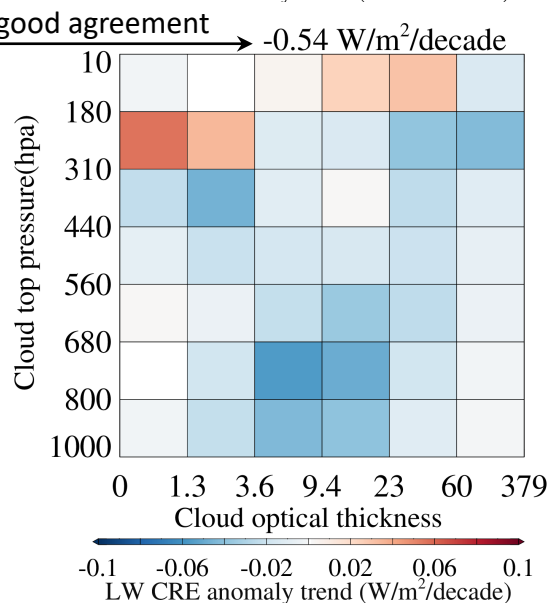
+



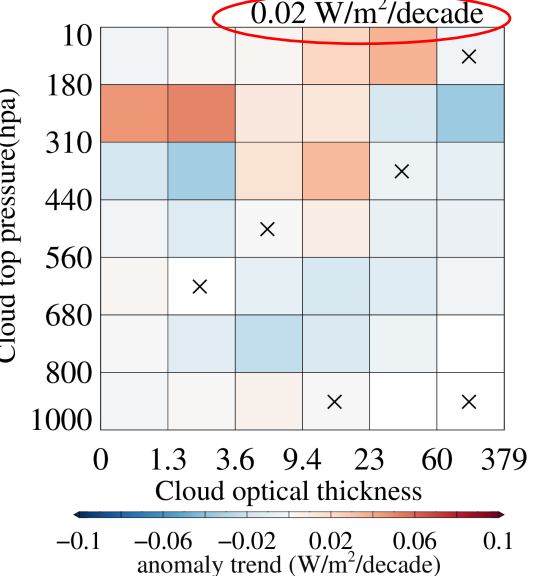
LW



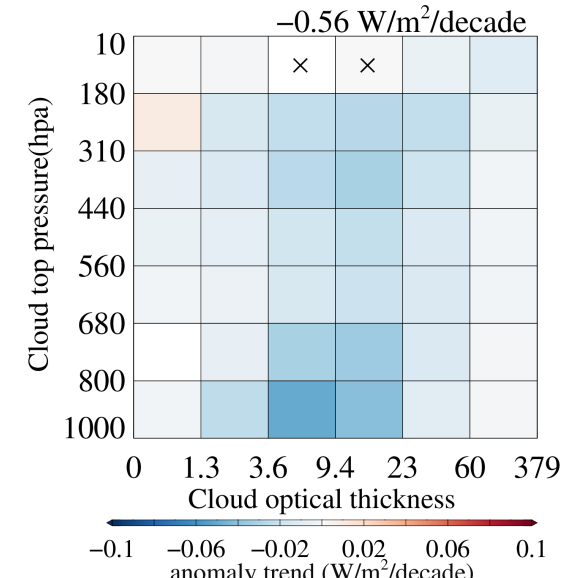
pretty good agreement



=



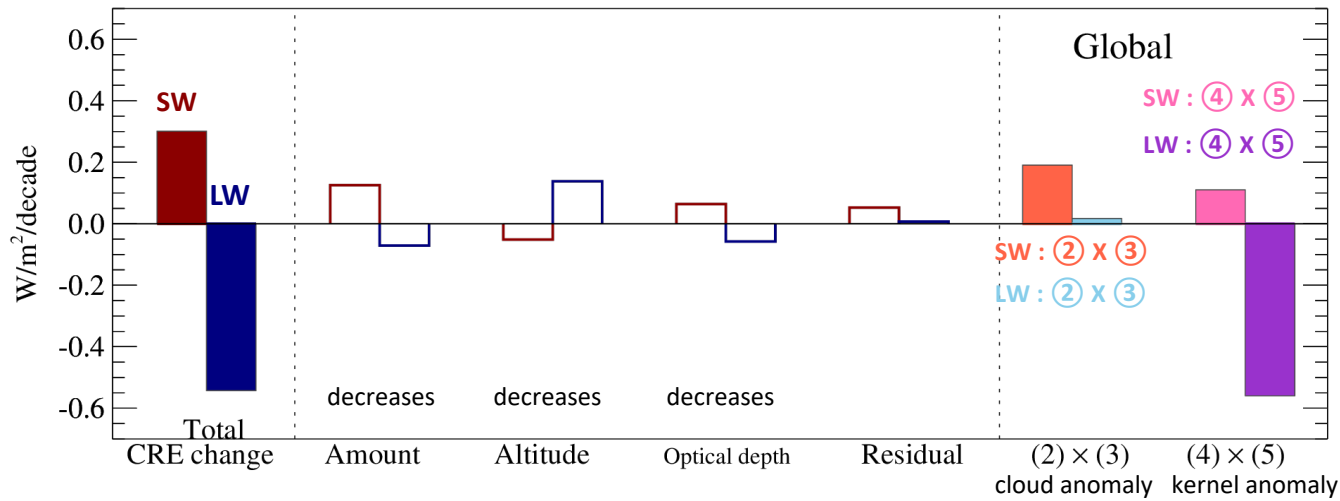
+



A closer look at the cloud anomaly term “(2) × (3)”

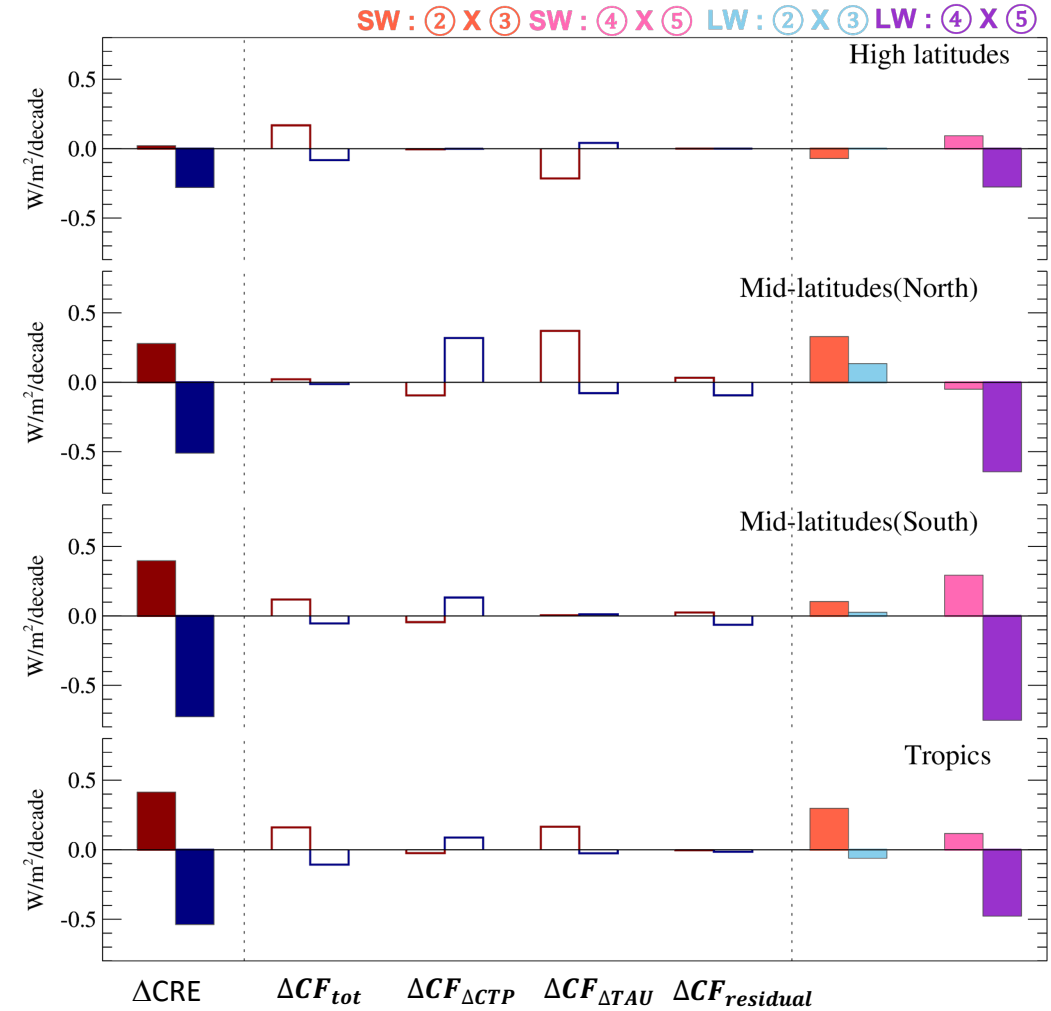
$$\Delta CRE \approx \Delta C \times \bar{K} + \bar{C} \times \Delta K$$

(Note: In the original image, the term $\Delta C \times \bar{K}$ is circled in red.)



$$\Delta C = \Delta C_{tot} + \Delta C_{\Delta CTP} + \Delta C_{\Delta TAU} + \Delta C_{residual}$$

Methodology : Zelinka et al., 2013



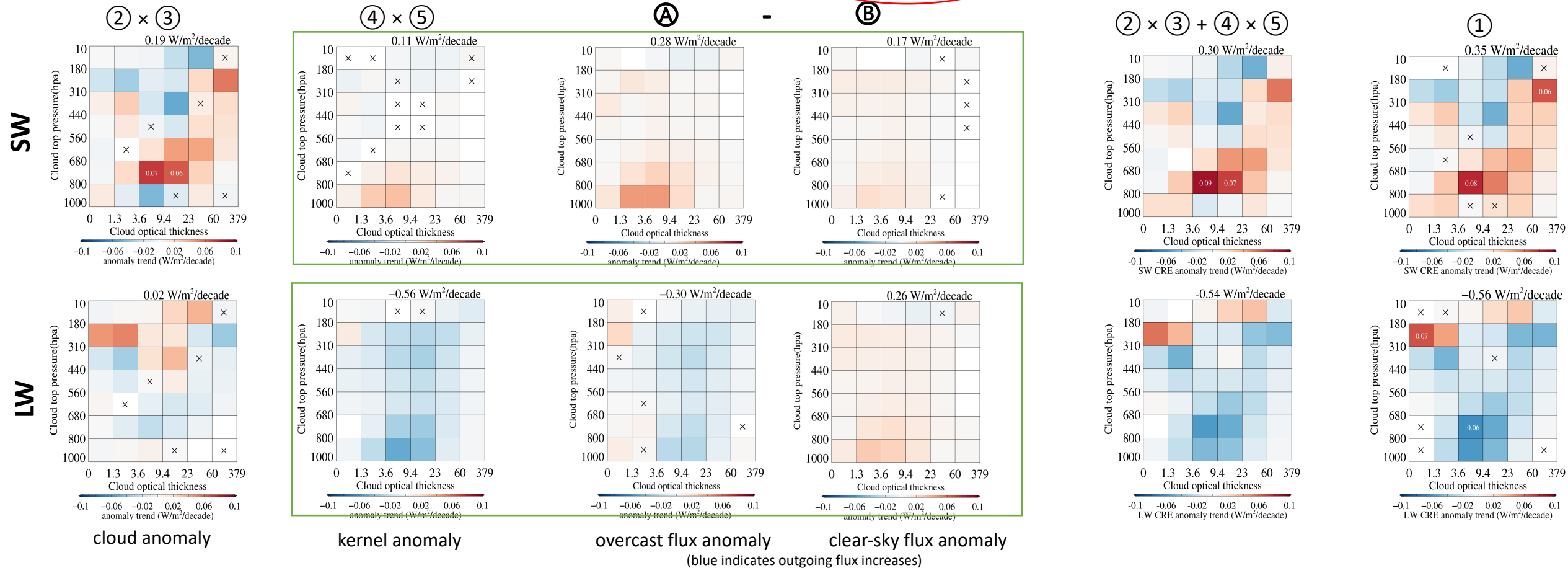
2003-2019

A closer look at the kernel anomaly term “(4) × (5)”

$$\begin{aligned} \textcircled{1} \quad \Delta CRE &\approx \textcircled{2} \quad \Delta C \quad \times \quad \textcircled{3} \quad \overline{K} \quad + \quad \textcircled{4} \quad \overline{C} \quad \times \quad \textcircled{5} \quad \Delta K \\ \Delta CRE &= \Delta C \times (\overline{F_{ovc}} - \overline{F_{clr}}) \quad + \quad \overline{C} \times \Delta(F_{ovc} - F_{clr}) \\ \Delta CRE &= \Delta C \times (\overline{F_{ovc}} - \overline{F_{clr}}) \quad + \quad \underbrace{\overline{C} \times \Delta F_{ovc}}_{\textcircled{A}} - \underbrace{\overline{C} \times \Delta F_{clr}}_{\textcircled{B}} \end{aligned}$$

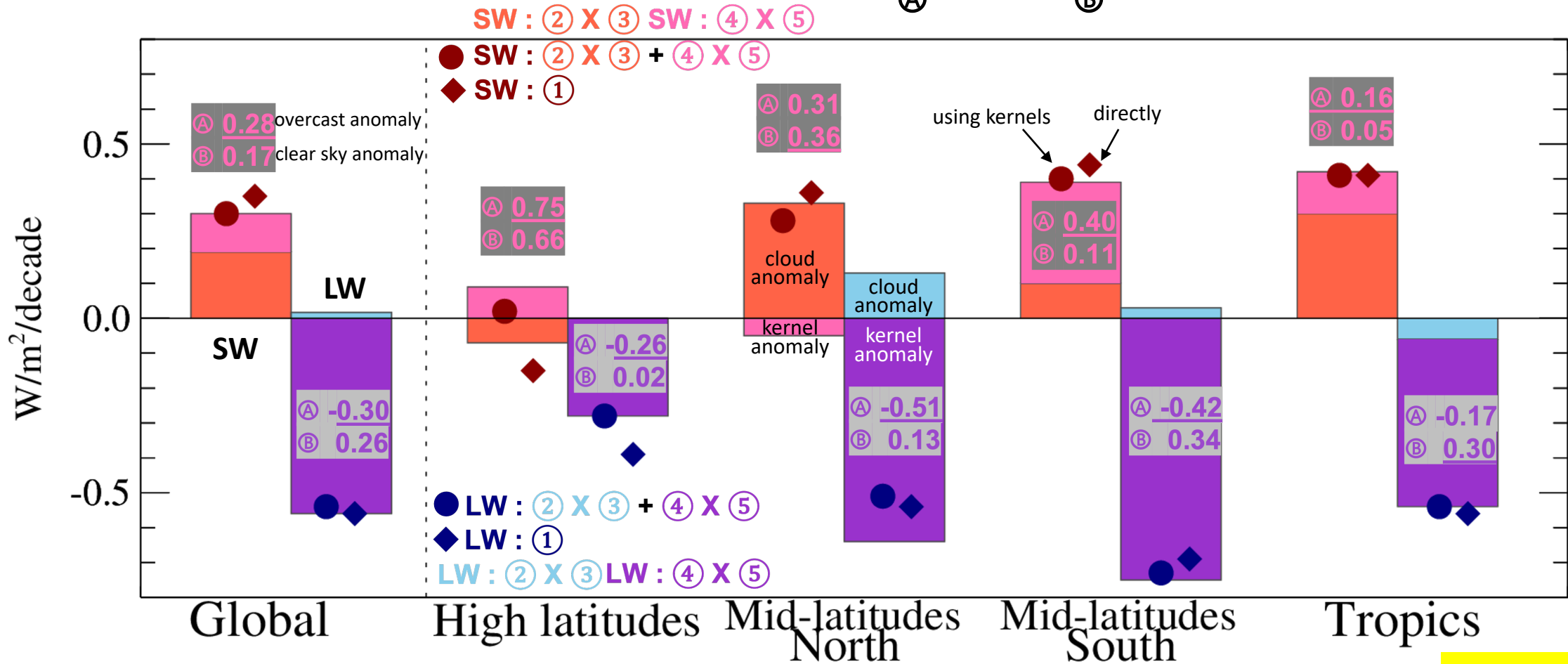
from the kernel (no covariance)

directly calculated



Everything together

$$\begin{aligned} \textcircled{1} \quad \Delta CRE &\approx \textcircled{2} \quad \Delta C \quad \times \quad \textcircled{3} \quad \bar{K} & + & \quad \textcircled{4} \quad \bar{C} \quad \times \quad \textcircled{5} \quad \Delta K \\ \Delta CRE &= \Delta C \times (\overline{F_{ovc}} - \overline{F_{clr}}) & + & \quad \textcircled{A} \quad \bar{C} \times \Delta F_{ovc} - \bar{C} \times \Delta F_{clr} & \textcircled{B} \end{aligned}$$



Potential extension by breaking down by cloud class/type/regime

$$\Delta CRE = \underbrace{\sum_{k=0}^K \bar{f}_k \Delta C_k \bar{K}_k}_{\text{within regime (can break into } \Delta CF_{tot}, \Delta CF_{\Delta T P}, \Delta CF_{\Delta T A U})}} + \underbrace{\sum_{k=0}^K \bar{f}_k \bar{C}_k \Delta K_k}_{\text{within regime (can break into ovc and clr)}} + \underbrace{\sum_{k=0}^K \Delta f_k \bar{C}_k \bar{K}_k}_{\text{across regimes}} + \textit{covariance terms}$$

② × ③
④ × ⑤
⑥

- In GCM work, Zelinka et al. (2022) considered only within-regime cloud anomaly ② × ③ and between regime frequency change ⑥ terms.
- Have to use daily FBCT data because CRs are defined daily.
- Each CR has its own K , but their ΔKs appear noisy.

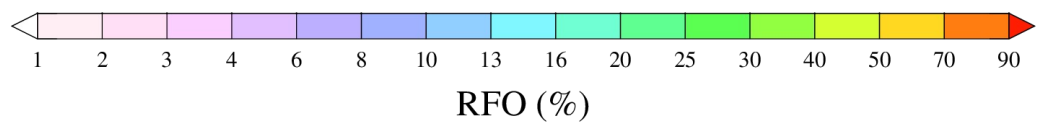
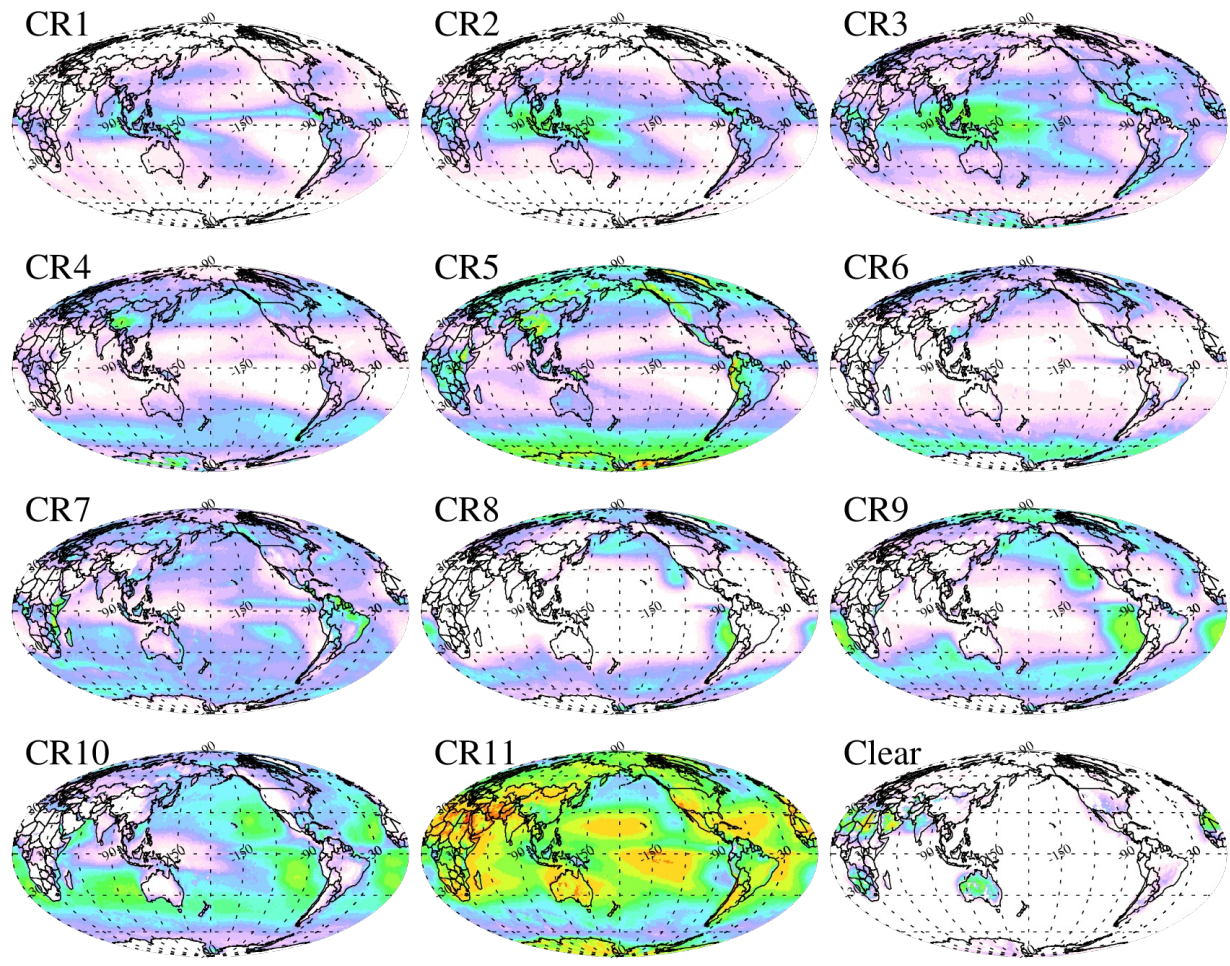
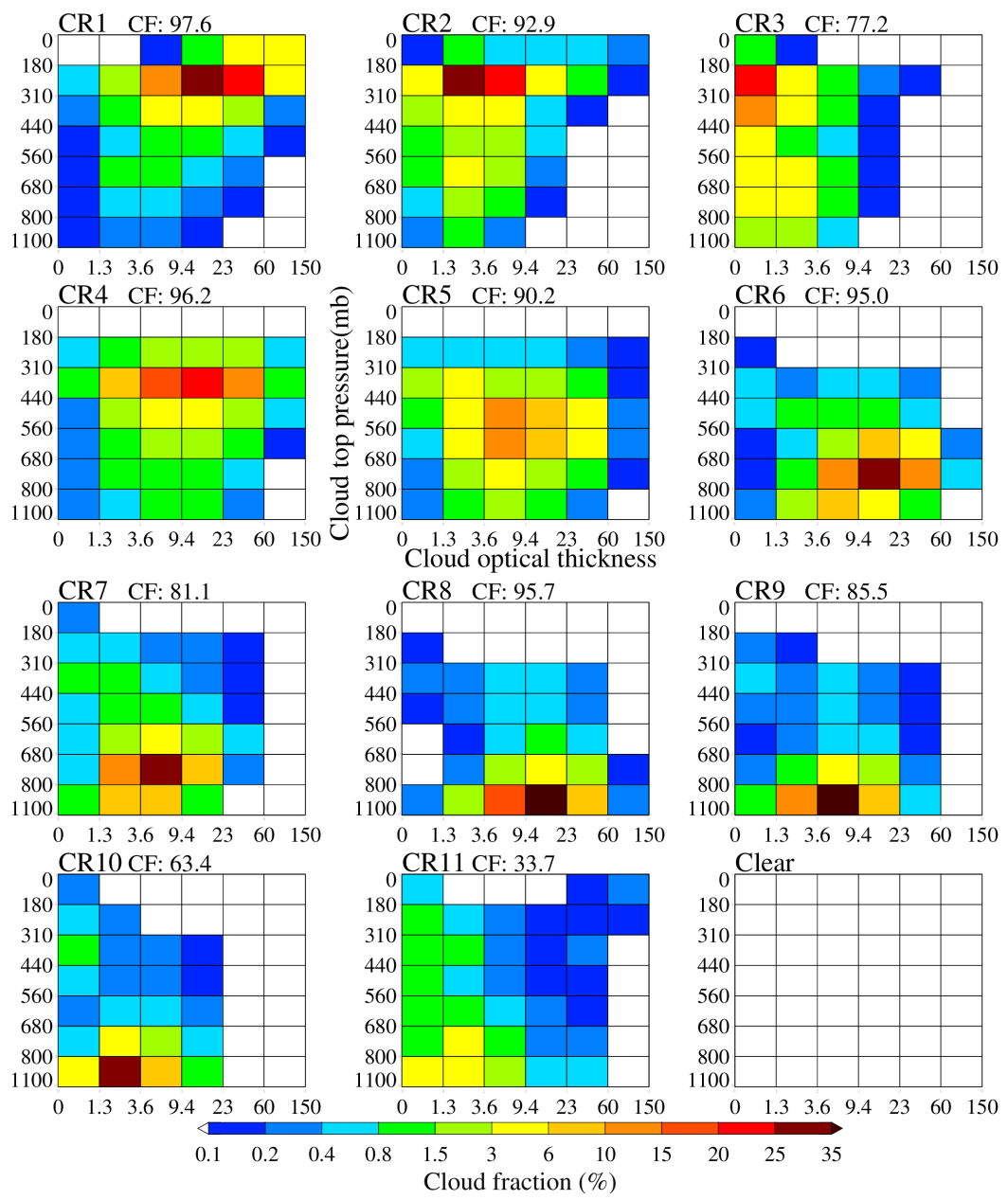
	Daily dataset		Monthly dataset	
	Wm ⁻² decade ⁻¹		Wm ⁻² decade ⁻¹	
	directly ($\Delta CRE = \sum_{k=0}^K \Delta CRE_k f_k$)	from equation above (no covariance)	directly	from cloud anomaly + kernel anomaly (no covariance)
SW	0.28	0.29	0.35	0.30
LW	-0.50	-0.45	-0.56	-0.54

Conclusions

- CERES FBCT allows for a Cloud Radiative Kernel (*CRK*) framework to be used to analyze and interpret *ΔCRE*.
- The first-order approximation of *ΔCRE* coming from cloud changes and fixed CRKs cannot reproduce directly calculated *ΔCRE*.
- When *CRK* anomalies/trends are accounted for (i.e., anomalies in overcast and clear-sky fluxes) good global agreement between direct and kernel-based estimates of *ΔCRE* is achieved.
- 2003-2019 analysis indicates a trend of weaker negative SW *CRE* values i.e., less cooling, and a trend of weaker positive values LW *CRE* values, i.e., less warming.
- A large contribution (~2/3) to SW *CRE* anomalies comes from fewer and optically thinner clouds, while LW *CRE* anomalies are dominated by overcast and clear-sky flux changes (i.e. *CRK* anomalies), as those due to cloud property anomalies largely cancel out.
 - Overcast (OVC) and clear sky (CLR) flux anomalies merit further investigation. LW outgoing fluxes increase for OVC and decrease for CLR (decreased LW *CRE*); SW outgoing fluxes decrease more for OVC than CLR (decreased SW *CRE*).

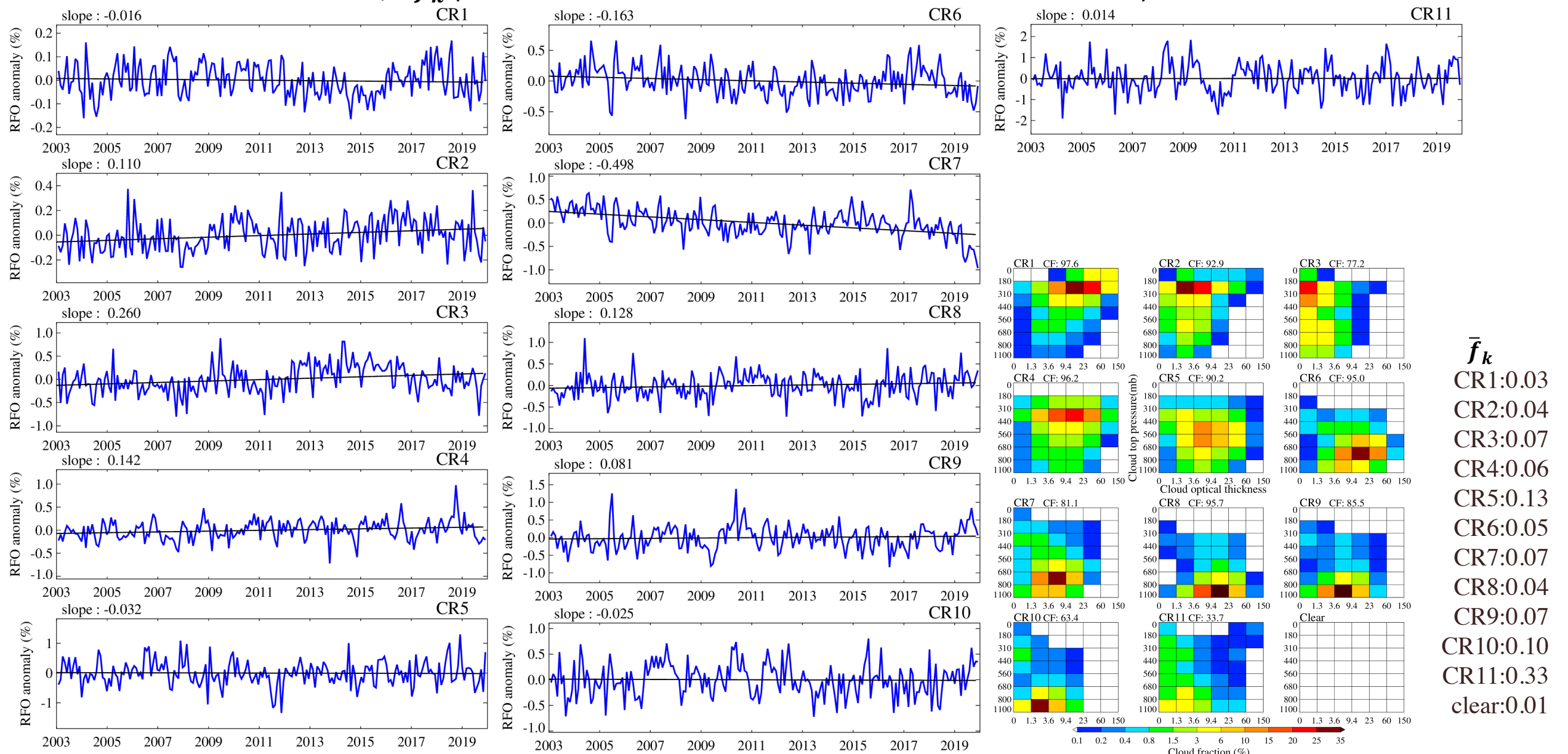
Backup Slides

CERES FBCT CRs



$$\Delta CRE = \sum_{k=0}^K (\bar{f}_k \Delta C_k + \Delta f_k \bar{C}_k) (\bar{F}_k^{ovc} - \bar{F}_k^{clr}) + \sum_{k=0}^K \bar{f}_k \bar{C}_k (\Delta F_k^{ovc} - \Delta F_k^{clr})$$

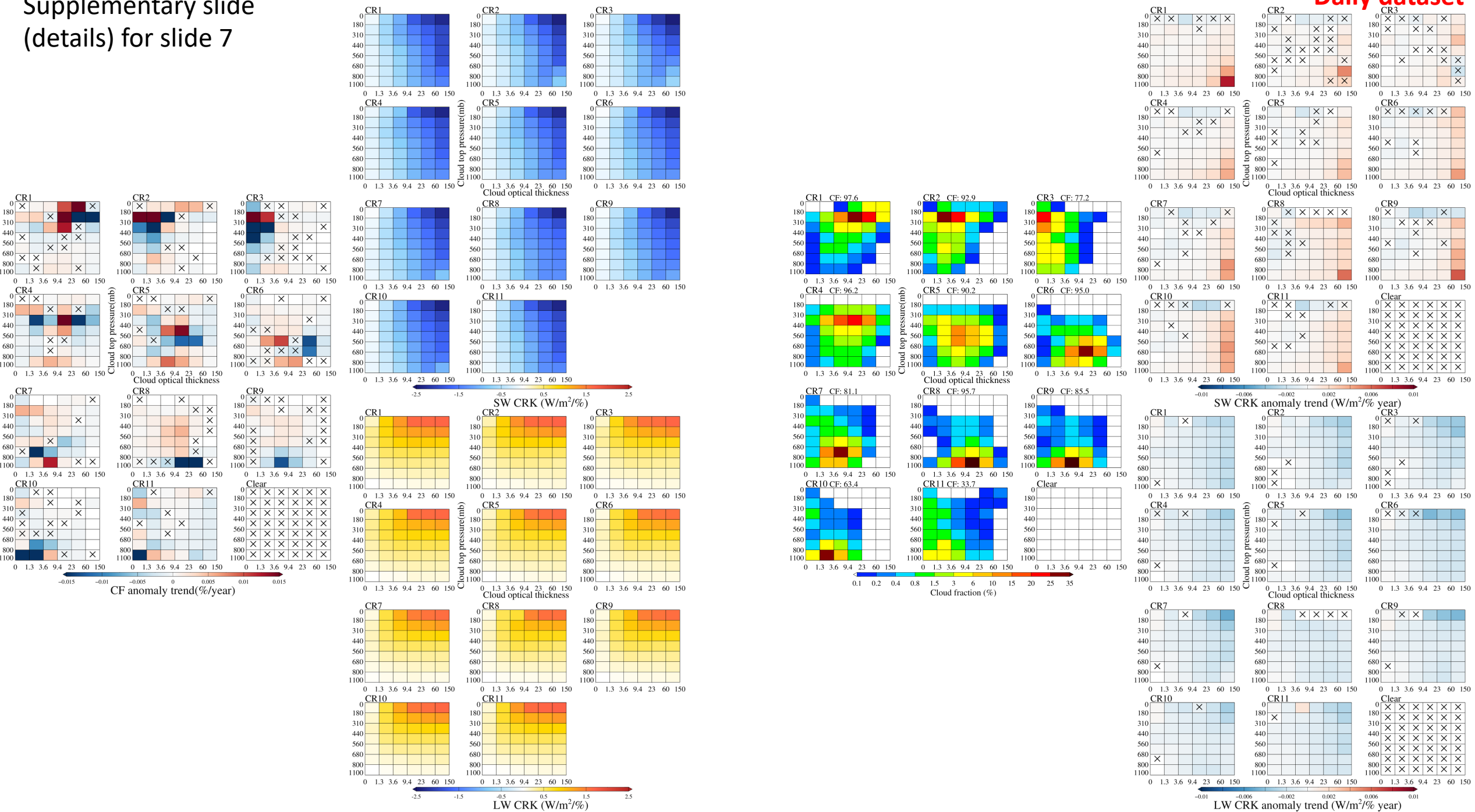
Check RFO trends for FBCT CRs, Δf_k (To confirm FBCT CRs have similar trends with MODIS CRs)



\bar{f}_k
 CR1:0.03
 CR2:0.04
 CR3:0.07
 CR4:0.06
 CR5:0.13
 CR6:0.05
 CR7:0.07
 CR8:0.04
 CR9:0.07
 CR10:0.10
 CR11:0.33
 clear:0.01

Supplementary slide (details) for slide 7

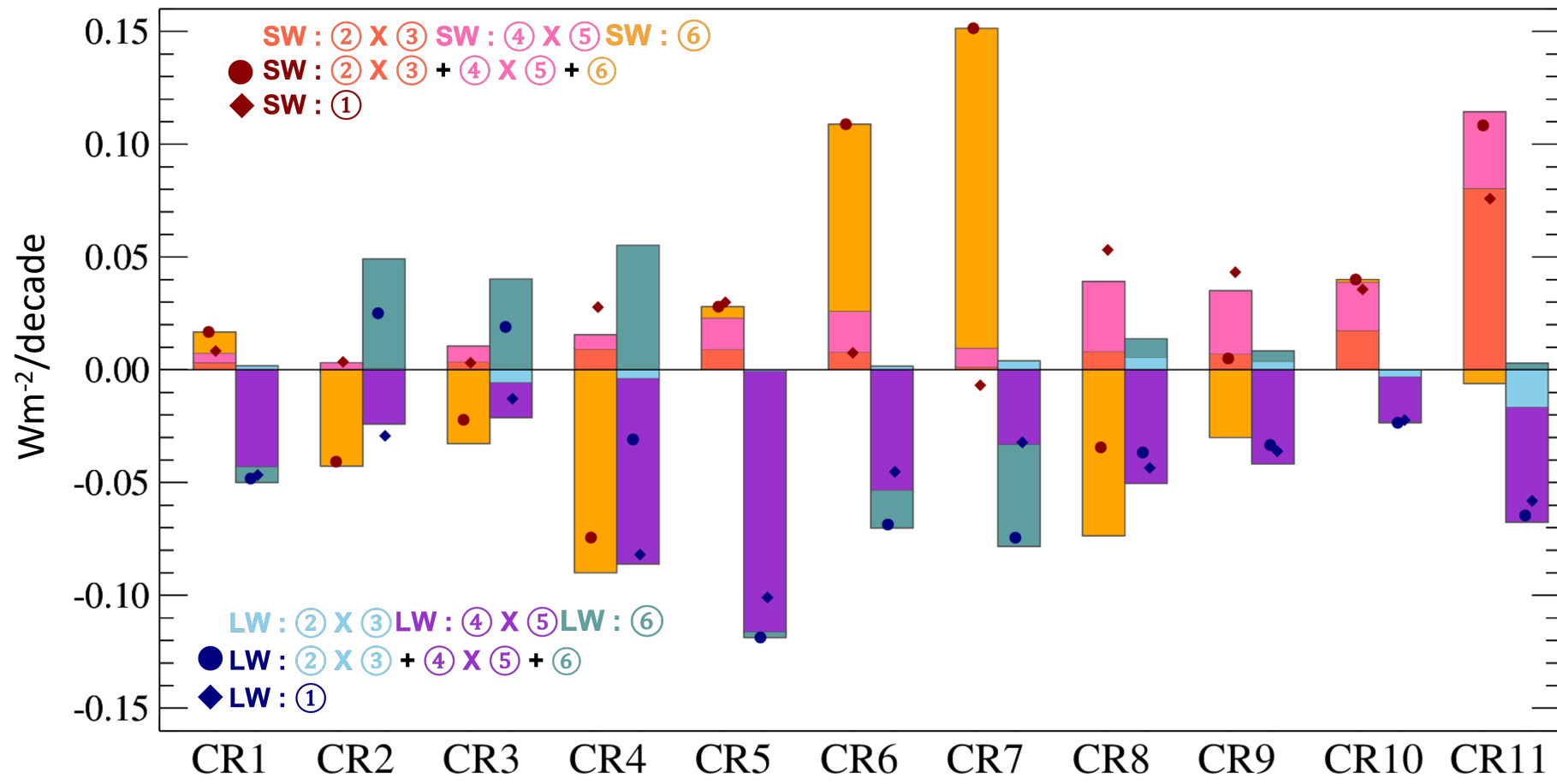
Daily dataset



Daily dataset

$$\Delta CRE = \sum_{k=0}^K \bar{f}_k \Delta C_k \overline{CRK}_k + \sum_{k=0}^K \bar{f}_k \bar{C}_k \Delta CRK_k + \sum_{k=0}^K \Delta f_k \bar{C}_k \overline{CRK}_k$$

② × ③
④ × ⑤
⑥

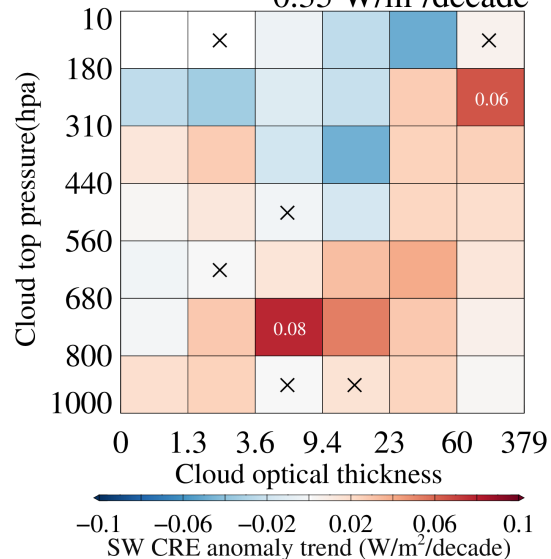


Values from
AGU poster

	Daily dataset		Monthly dataset	
	$\Delta CRE (= \sum_{k=0}^K \Delta CRE_k f_k)$	② × ③ + ④ × ⑤ + ⑥	ΔCRE	② × ③ + ④ × ⑤
SW	0.28	0.29	0.35	0.30
LW	-0.50	-0.45	-0.56	-0.54

①
 ΔCRE

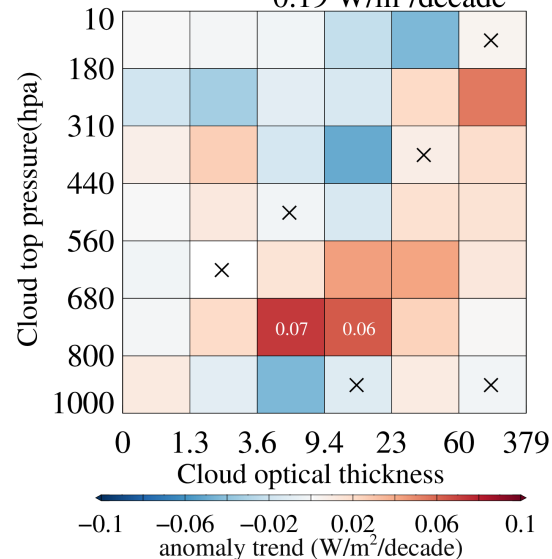
0.35 W/m²/decade



≈

② × ③
 ΔCF × \overline{CRK}

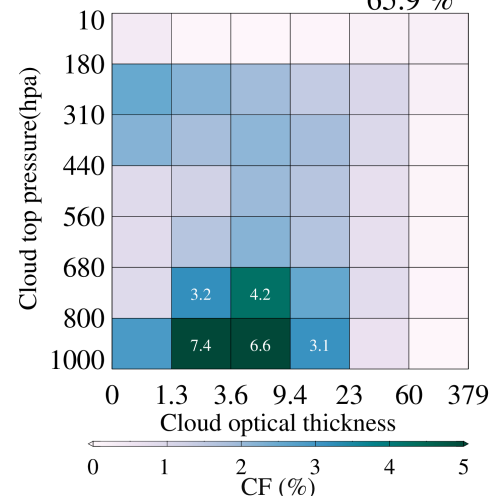
0.19 W/m²/decade



+

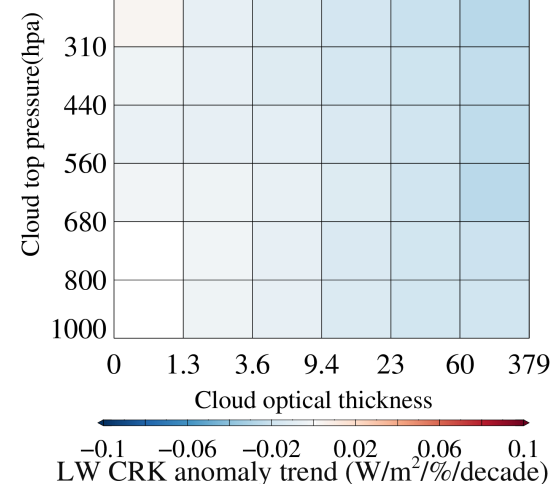
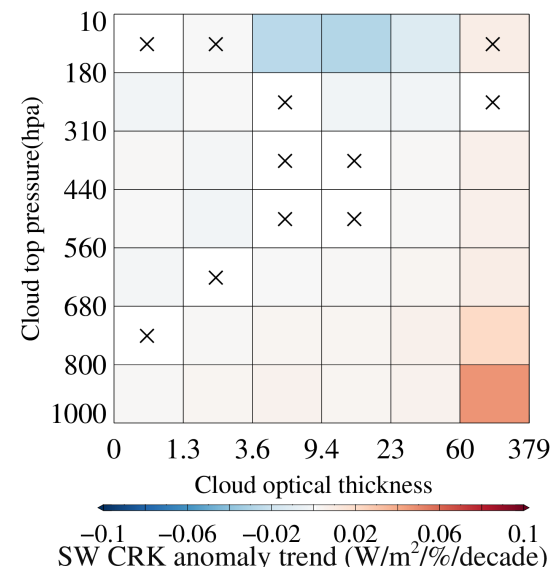
④
 \overline{CF}

65.9 %



×

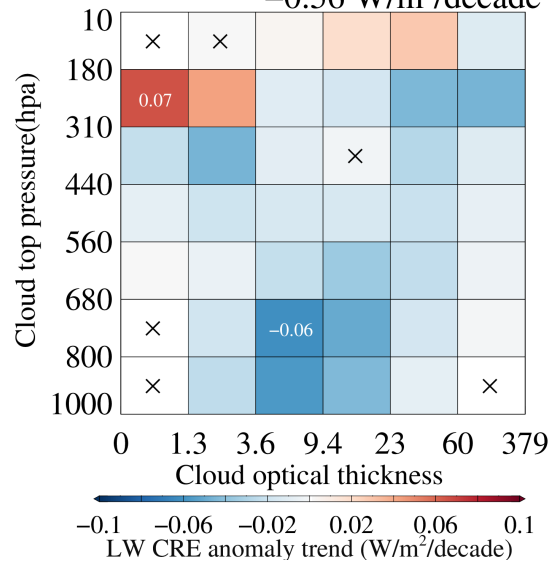
⑤
 ΔCRK



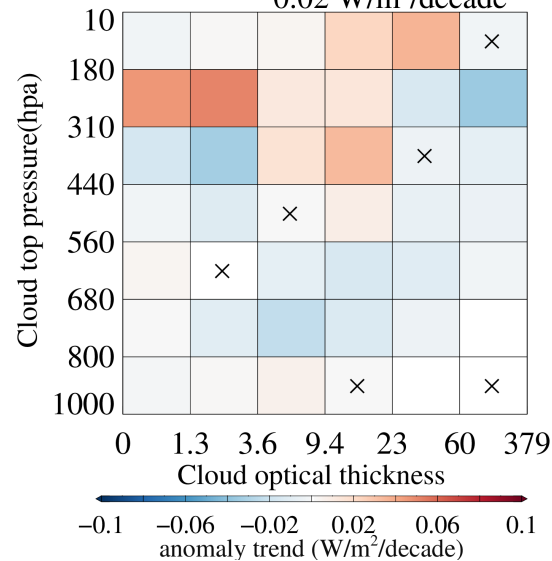
SW

LW

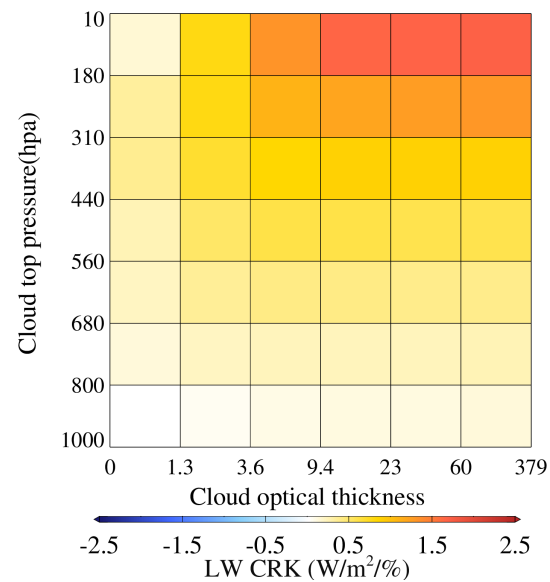
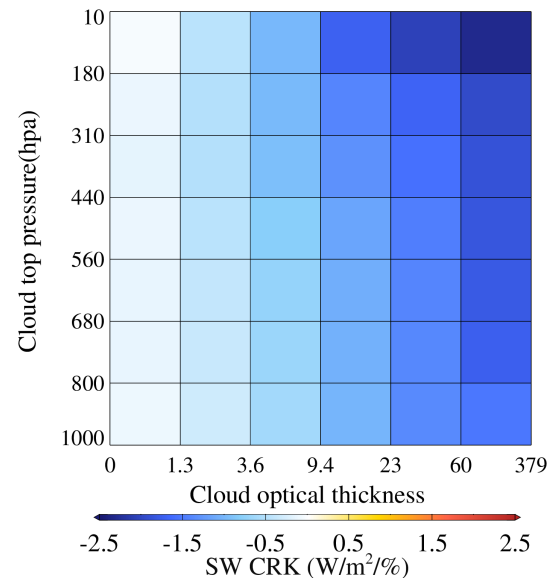
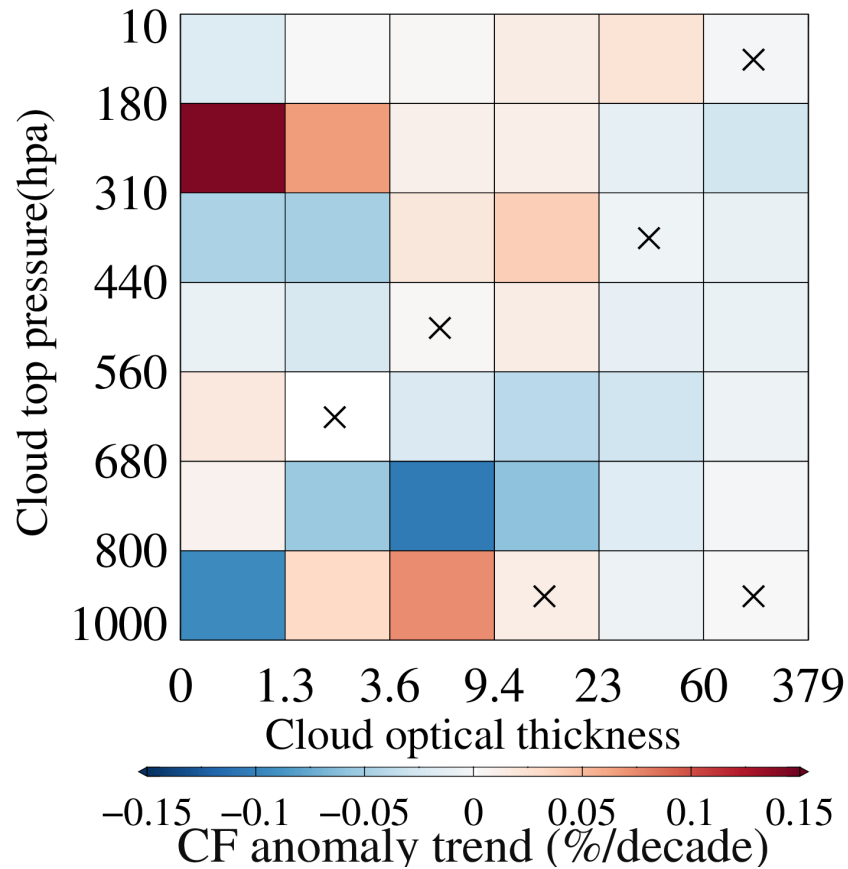
-0.56 W/m²/decade



0.02 W/m²/decade



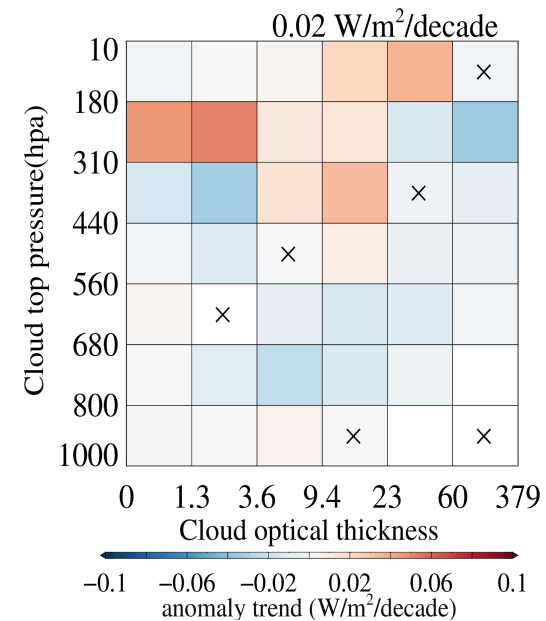
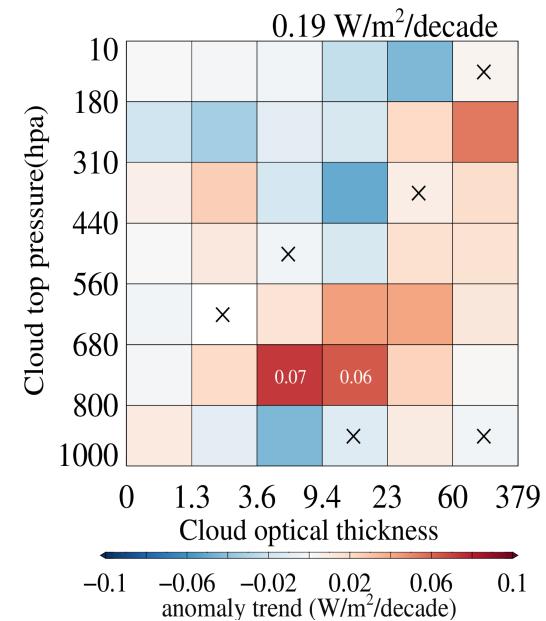
LW CRK anomaly trend (W/m²/%/decade)



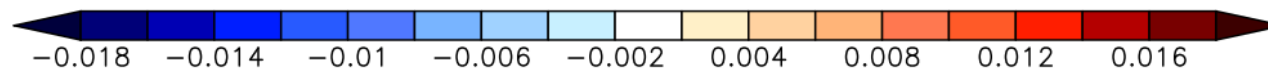
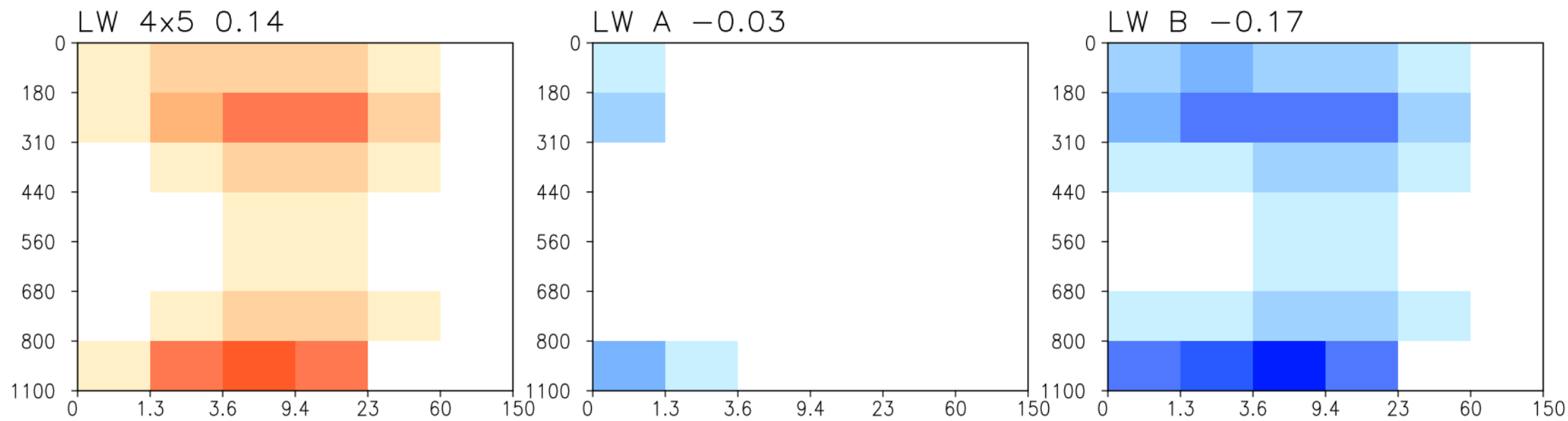
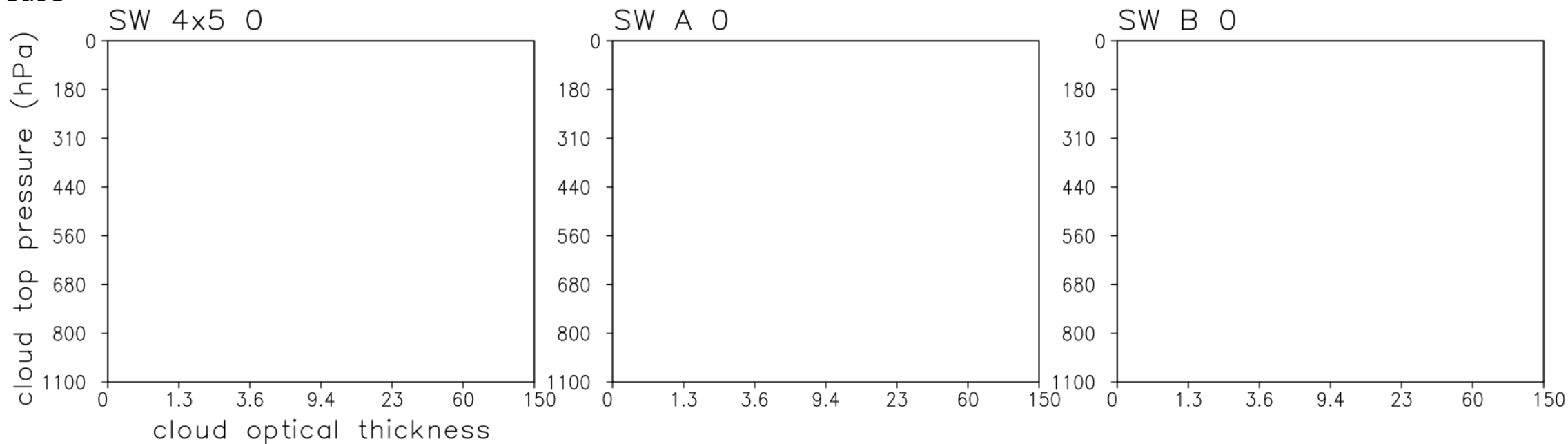
SW

LW

② × ③



Ts increase



q increase

