

Diagnosis of the Warm Bias Exhibited in CMIP5 AMIP Climate Simulations over ARM SGP site

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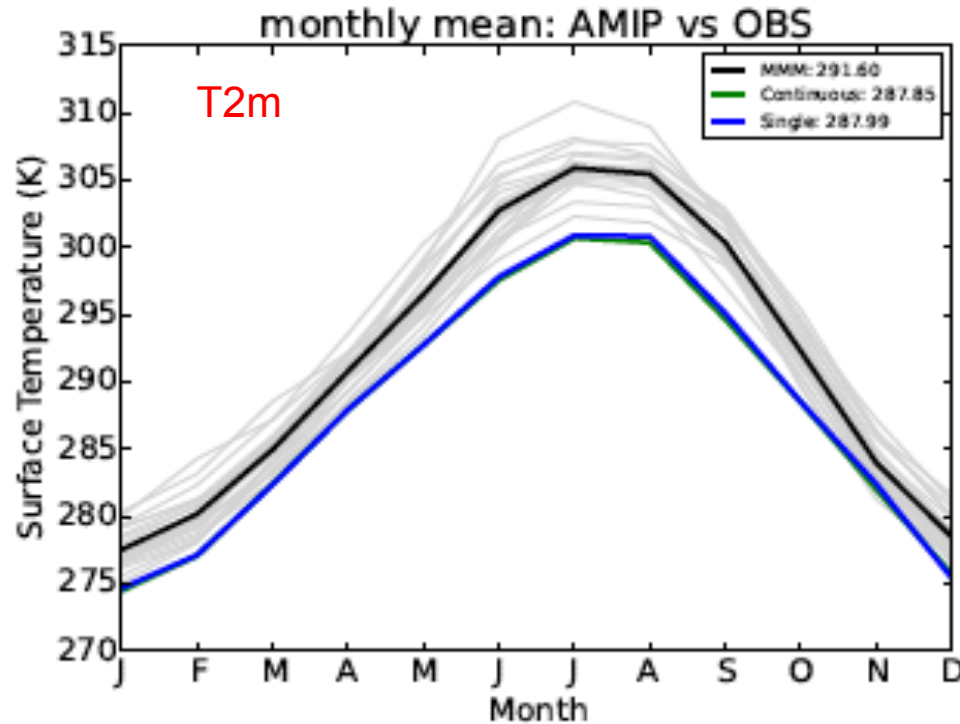


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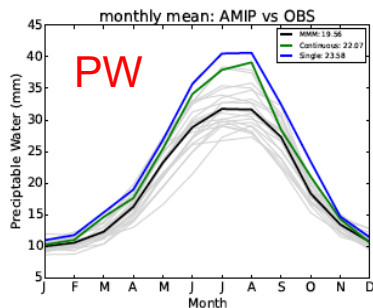
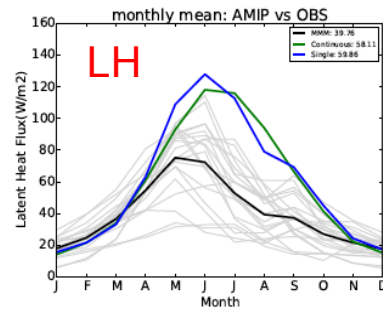
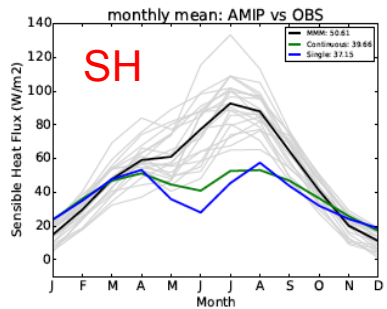
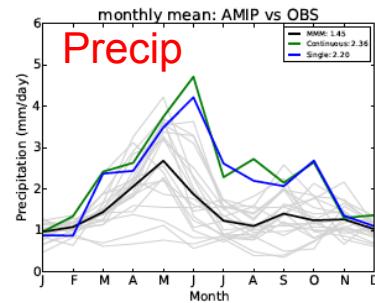
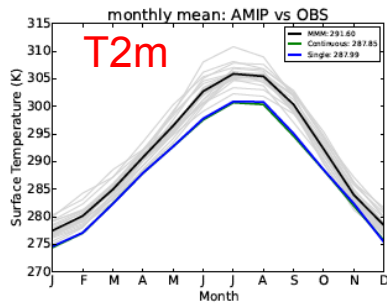
Warm bias exhibited in CMIP5 AMIP Climate Simulations over ARM SGP site



- **AMIP simulation:**
Eliminate SST bias introduced by coupled simulations
- Warm bias exhibits year around for most CMIP 5 models.
~ 3.6 K in Annual mean
~ 5.2 K in JJA
- Black: Multi-model mean
Green: Continuous forcing (Variational analysis)
3x3 Deg. domain mean
Blue: Single point at central facility (ARMBE)



Statistical Summary of Bias patterns in Climate Simulation over SGP

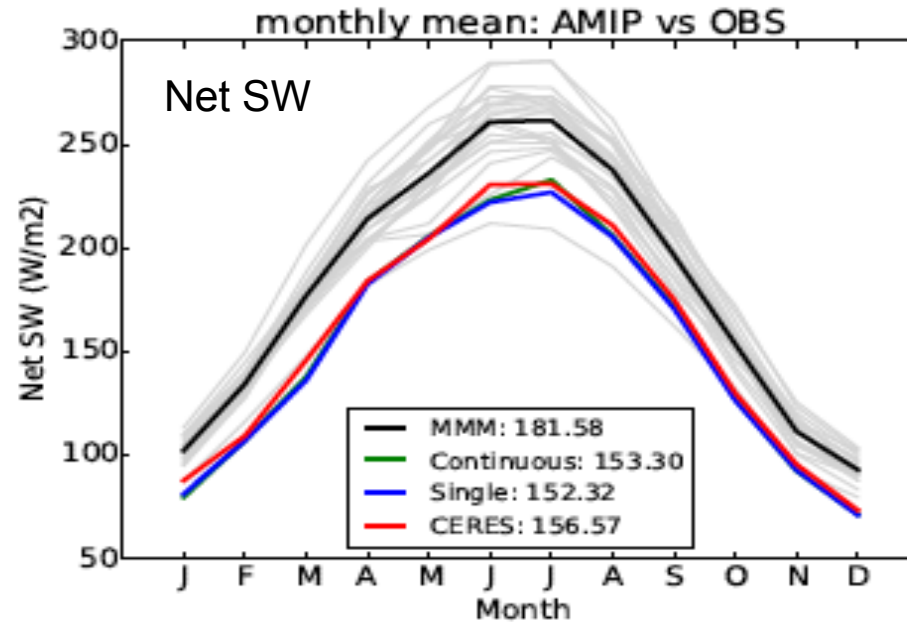


- Surface is too warm and dry. Bias is the highest during summer.
- Sensible heat is over-estimated latent heat is under-estimated in summer
- Precipitation peak bias
- The atmosphere column is too dry



What contribute to the T2m bias?

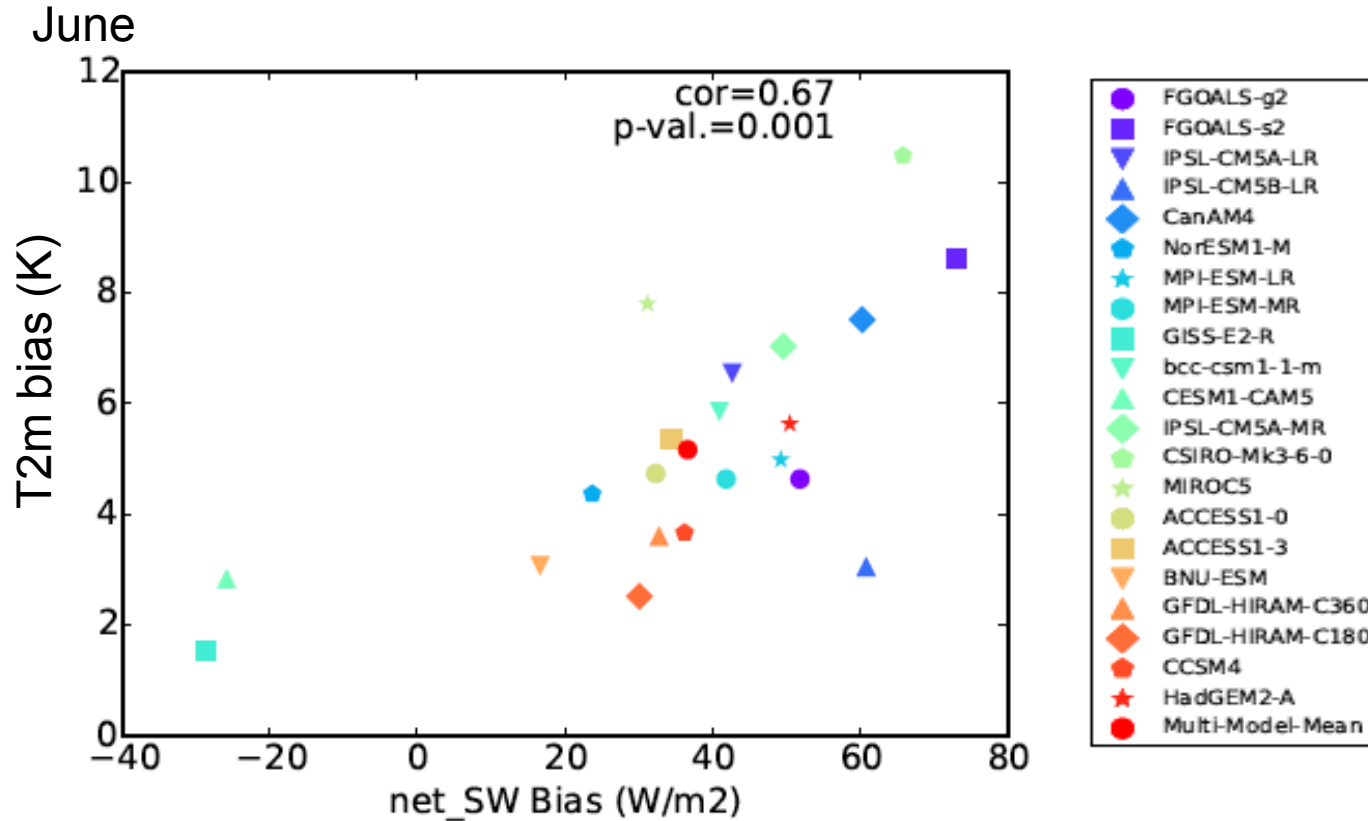
Contribution of solar radiation error at surface



Based on both **CERES (Red)** and **ARM data (Blue and Green)**, models have strong net solar radiation bias (25~29 W/m² in annual mean)



Net SW bias correlates to T2m bias across models



Partition the surface net SW error

$$\begin{aligned}
 \text{Net-SW error} &= (1 - \alpha_{\text{mod}}) * \text{SWDN clr,error} && \text{SWDNclr error} \\
 &+ (1 - \alpha_{\text{mod}}) * \text{SWDN cre,error} && \text{SWDNcre error} \\
 &+ (\alpha_{\text{mod}} - \alpha_{\text{obs}}) * (\text{SWDN clr} + \text{SWDN cre}) && \text{Albedo error}
 \end{aligned}$$

Net SW error attribution (JJA mean):

CMIP5 Mean -Observation	Net-SW error (Wm-2)	Albedo error (Wm-2)	SWDN-clr error (Wm-2)	SWDN-cre error (Wm-2)
Based on ARM	32.68	6.26	4.84	21.54
Based on CERES	29.15	0.86	14.67	13.62



ARM vs CERES vs models

- The Radiative Flux Analysis product from ARM:
Clear-sky fluxes are derived for identified clear-sky, empirical cosine fit based on solar angle to generate continuous data stream. **It reflects ACTUAL Clear-sky fluxes.**
- CERES and models:
Clear-sky fluxes are calculated by removing the cloud but retaining aerosol and water vapor at cloudy time. Clear-sky fluxes are lower than ACTUAL clear-sky values.
In order to evaluate climate models, CERES surface product is more suitable under definitional ground.

CMIP5 Mean -Observation	Net-SW error (Wm-2)	Albedo error (Wm-2)	SWDN-clr error (Wm-2)	SWDN-cre error (Wm-2)
Based on ARM	32.68	6.26	4.84	21.54
→ Based on CERES	29.15	0.86	14.67	13.62

Thanks to Seiji Kato, David Rutan and Chuck Long for helping to understand.

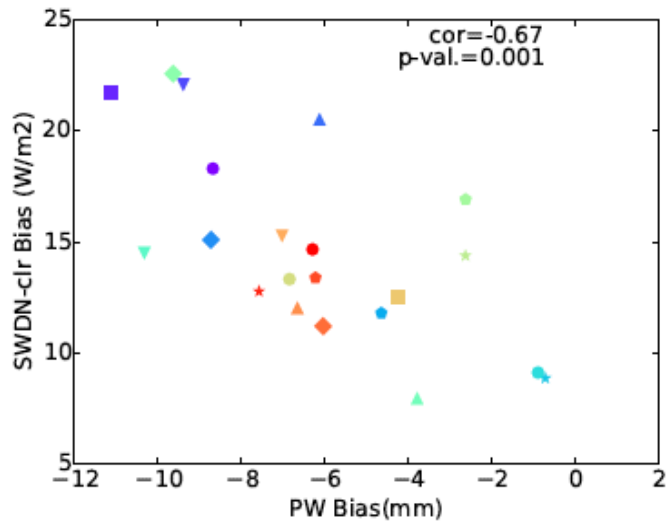


PW and [dSWA/dPW] errors contribute to clear-sky SWDN error

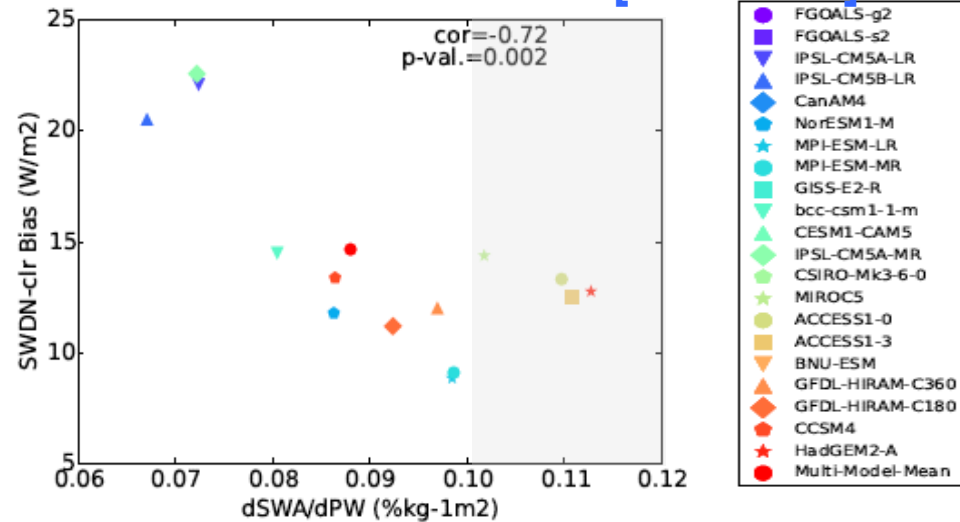
PW : Precipitable water

dSWA/dPW: SWA (SW absorption) sensitivity to PW variation

Observed dSWA/dPW = [0.10~0.13]



PW bias (mm)



dSWA/dPW (%kg⁻¹m⁻²)

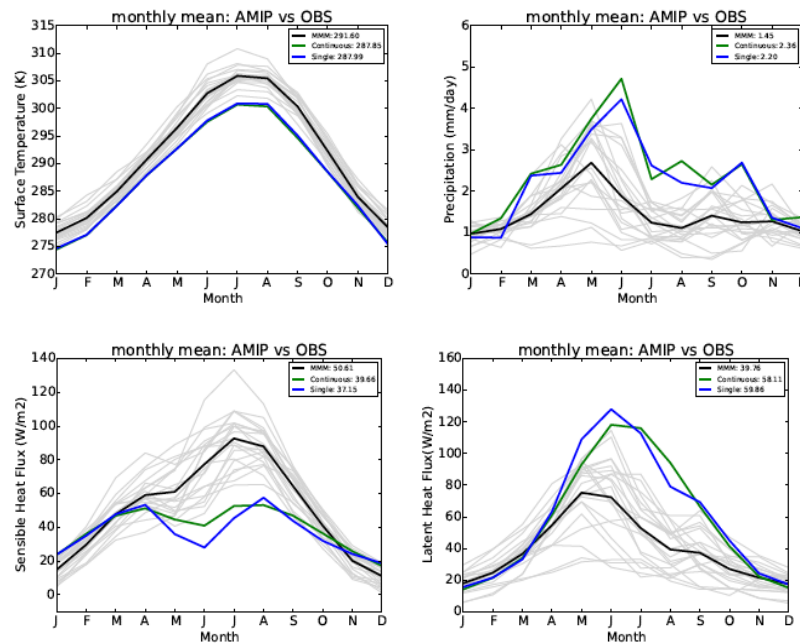
radiation transfer scheme deficiency
(DeAngelis et al. 2016)



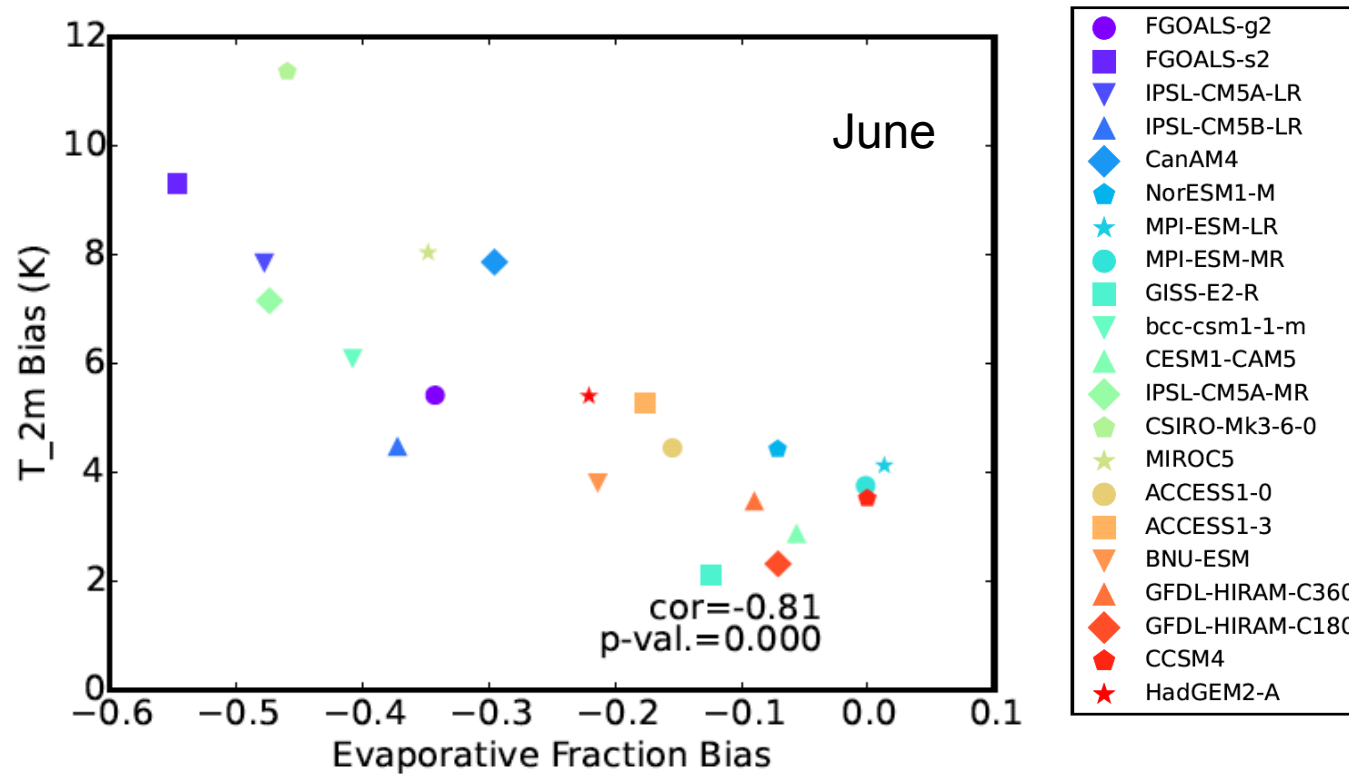
What contribute to the T2m bias? Contribution from hydrological processes

Systematic bias in summer rainfall and evaporation is found in all GCMs.

Precipitation and evaporation are too low.



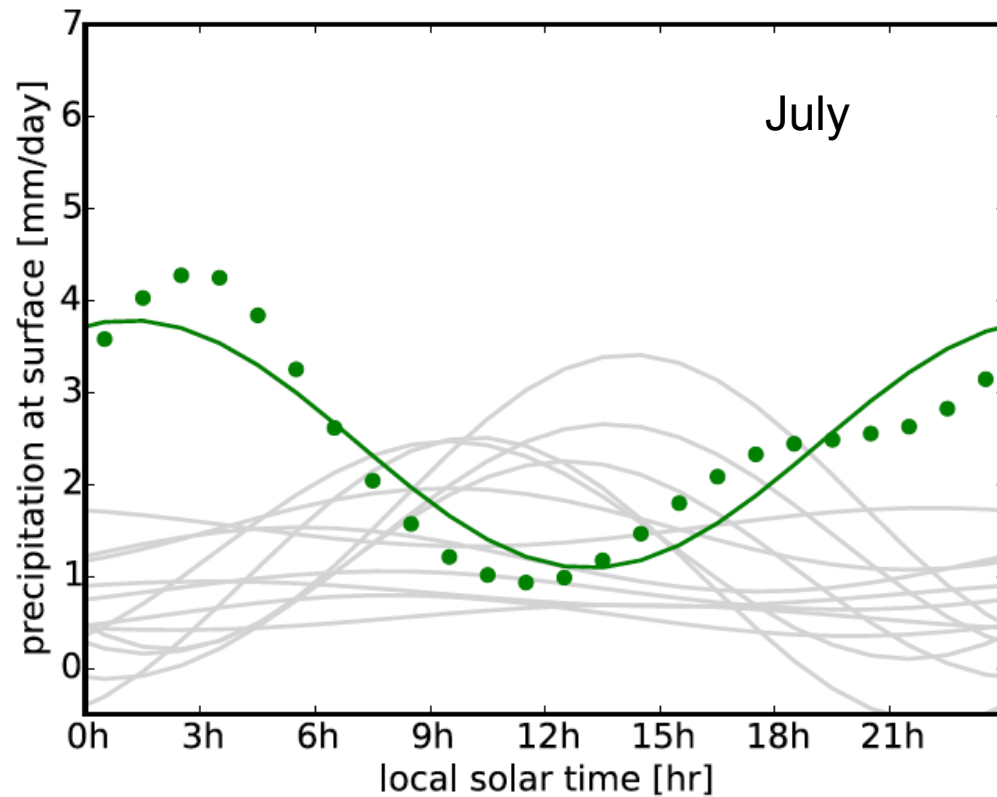
EF correlates to T2m bias across models



the partition of available energy at
the land surface: $EF = LH / (SH + LH)$



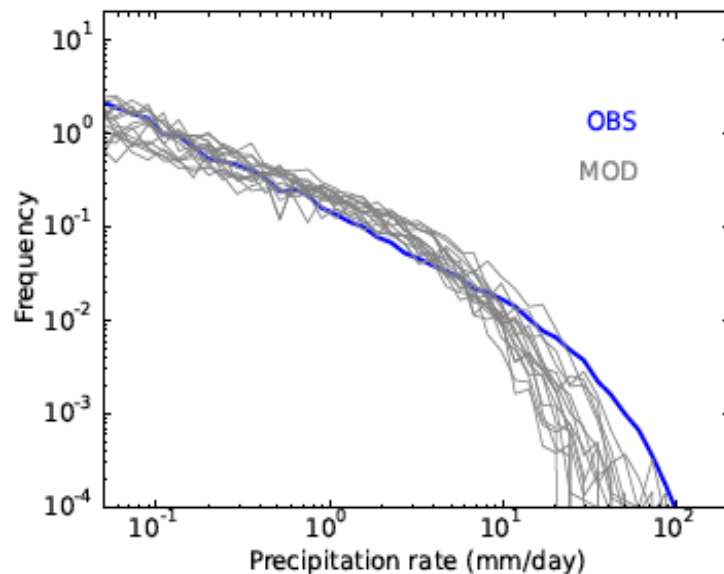
Diurnal cycle of precipitation



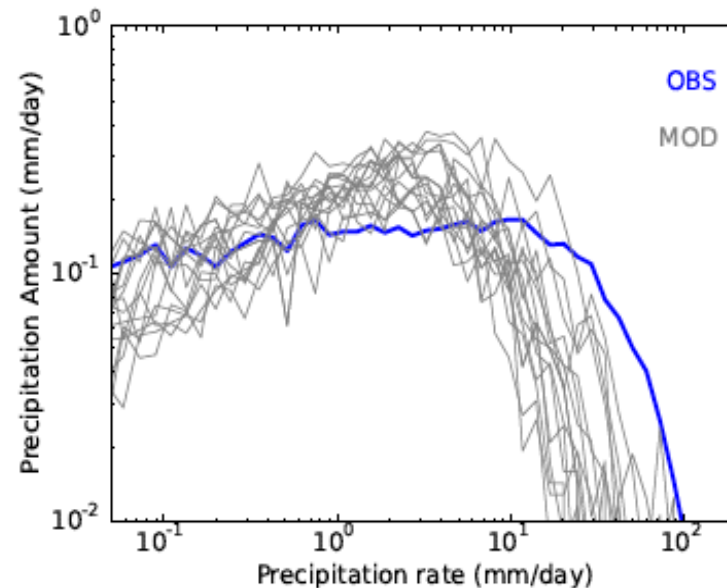
Models miss the nocturnal peak that is associated with eastward propagating convective episodes



PDF of precipitation



(a) Precipitation Frequency: JJA



(b) Precipitation Amount: JJA

- Models overestimate moderate precipitation occurrence, but underestimated light and heavy rainfall occurrence.
- This may relate to deficiencies in convective schemes in models



Summary and Discussion

- Warm bias over ARM SGP site is found in CMIP5 AMIP type of simulations
- Over-estimated net solar radiation and under-estimated Evaporative Fraction are responsible for the bias creation
- Both cloud radiative effect and clear-sky errors are important components in net solar radiation error*
- Process-oriented diagnosis using higher frequency model outputs
 - separate day time and night time processes
 - analyze what types of precipitation errors (nocturnal heavy rainfall?) that contribute most to the warm bias
- Large-scale circulation bias is inevitable: CAPT is a better tool to eliminate the large scale forcing bias and focus on the local physical processes



Extra slides



Prioritize the surface net SW error

$$\text{Net-SW error} = (1 - \alpha_{\text{mod}}) * \text{SWDN clr,error} + (1 - \alpha_{\text{mod}}) * \text{SWDN cre,error} + (\alpha_{\text{mod}} - \alpha_{\text{obs}}) * (\text{SWDN clr} + \text{SWDN cre})$$

SWDNclr error
 SWDNcre error
 Albedo error

Based on CERES EBAF:

Model	Net-SWDN error (Wm-2)	Albedo error (Wm-2)	SWDN-clr error (Wm-2)	SWDN-cre error (Wm-2)
FGOALS-g2	19.11	-9.64	18.31	10.51
FGOALS-s2	38.18	-7.93	21.76	24.32
IPSL-CM5A-LR	26.86	0.07	22.09	4.79
IPSL-CM5B-LR	41.04	-1.48	20.53	22.07
CanAM4	41.78	-1.03	15.10	27.53
NorESM1-M	35.49	6.23	11.80	17.60
MPI-ESM-LR	19.46	-7.52	8.87	18.13
MPI-ESM-MR	17.20	-7.48	9.12	15.55
GISS-E2-R	-20.10	0.49	13.34	-33.99
bcc-csm1-1-m	18.24	-8.61	14.50	12.30
CESM1-CAM5	8.21	5.42	7.96	-5.11
IPSL-CM5A-MR	40.46	3.58	22.58	14.28
CSIRO-Mk3-6-0	54.00	10.08	16.92	26.76
MIROC5	36.65	7.67	14.39	14.64
ACCESS1-0	30.93	2.49	13.33	15.23
ACCESS1-3	44.42	10.77	12.55	21.25
BNU-ESM	56.40	20.63	15.28	20.44
GFDL-HIRAM-C360	22.81	0.03	12.03	10.81
GFDL-HIRAM-C180	12.97	-7.02	11.21	8.74
CCSM4	34.80	2.02	13.39	19.52
HadGEM2-A	33.21	-0.74	12.78	21.10
Multi-Model Mean	29.15	0.86	14.67	13.62

Based on ARM: Radflux Long

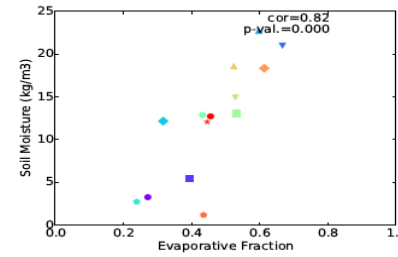
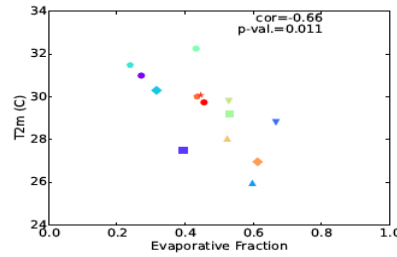
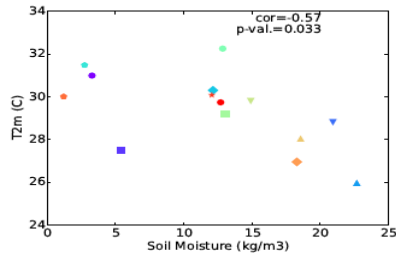
Model	Net-SWDN error (Wm-2)	Albedo error (Wm-2)	SWDN-clr error (Wm-2)	SWDN-cre error (Wm-2)
FGOALS-g2	22.65	-4.32	8.90	18.10
FGOALS-s2	41.72	-2.59	12.31	31.95
IPSL-CM5A-LR	30.39	5.46	12.27	12.71
IPSL-CM5B-LR	44.57	3.90	10.79	29.93
CanAM4	45.31	4.35	5.40	35.36
NorESM1-M	39.03	11.66	1.69	25.77
MPI-ESM-LR	22.99	-2.20	-0.60	25.77
MPI-ESM-MR	20.73	-2.15	-0.34	23.20
GISS-E2-R	-16.57	5.88	3.52	-26.06
bcc-csm1-1-m	21.78	-3.29	5.09	19.88
CESM1-CAM5	11.74	10.86	-2.06	2.97
IPSL-CM5A-MR	44.00	9.00	12.64	22.30
CSIRO-Mk3-6-0	57.53	15.60	6.77	34.91
MIROC5	40.18	13.13	4.24	22.82
ACCESS1-0	34.47	7.90	3.39	23.25
ACCESS1-3	47.95	16.25	2.26	29.56
BNU-ESM	59.93	26.19	4.61	29.05
GFDL-HIRAM-C360	26.35	5.42	2.22	18.72
GFDL-HIRAM-C180	16.50	-1.69	1.74	16.38
CCSM4	38.34	7.43	3.48	27.52
HadGEM2-A	36.74	4.65	3.04	28.97
Multi-Model Mean	32.68	6.26	4.84	21.54

With out GISS : 31.6 0.88 14.73 16.00

With out GISS : 35.14 6.27 4.9 23.92

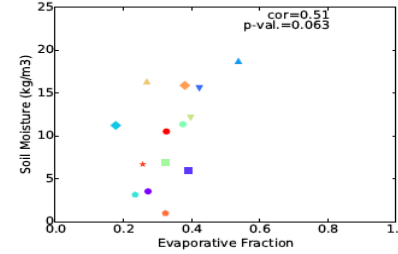
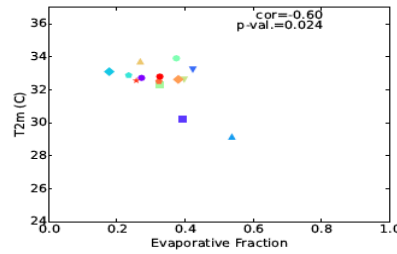
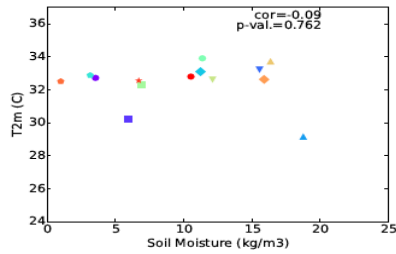


JUNE



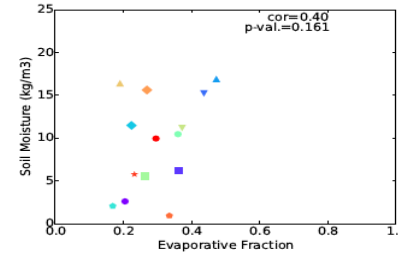
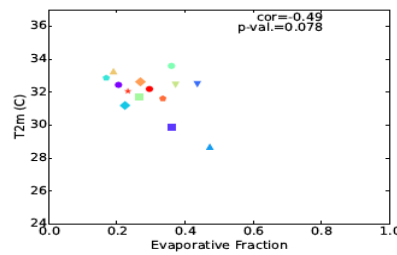
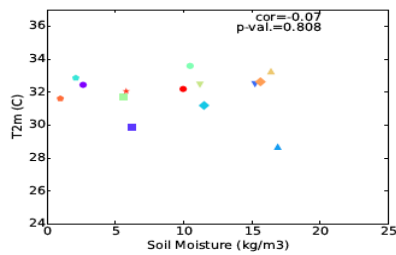
Soil moisture plays a central role in the partition of available energy at the land surface ($EF = LH / (SH + LH)$).

JULY



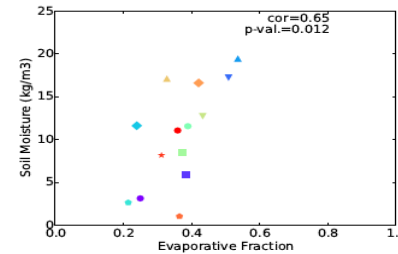
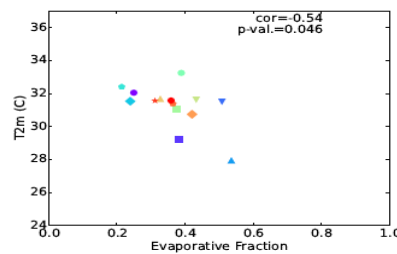
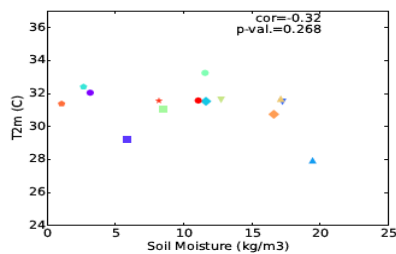
Soil moisture and EF coupling strength varies by months.

AUGUST

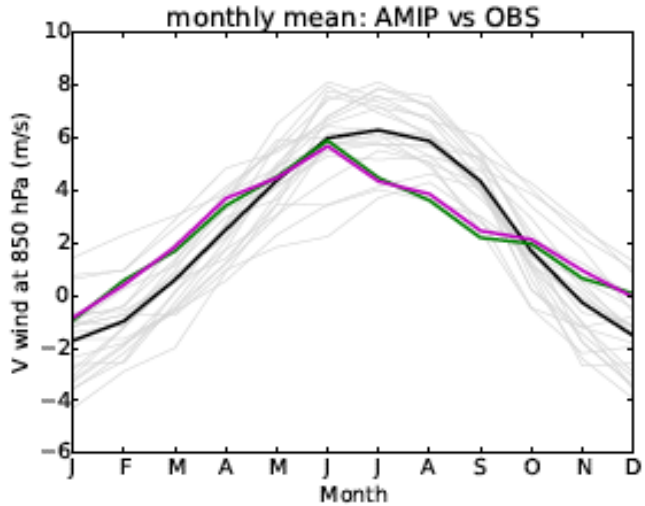


In June, coupling is strongest.

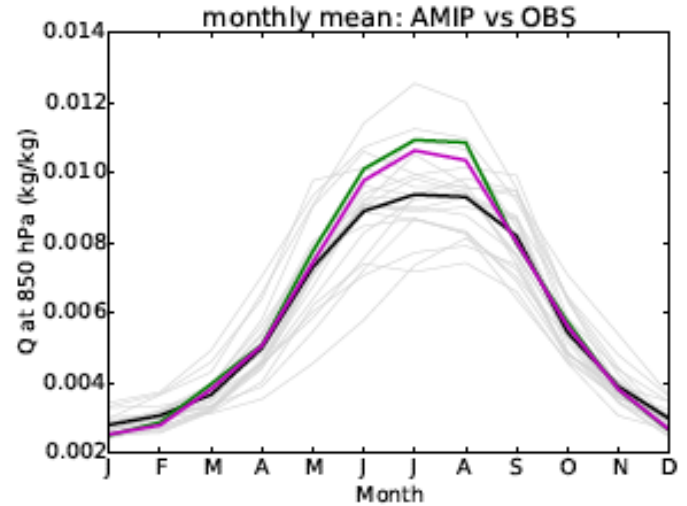
JJA



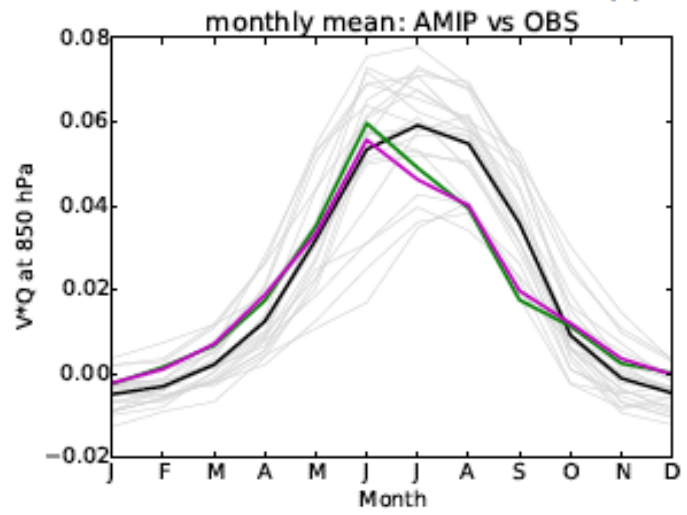
Onto moisture transport: Magenta (ECMWF) Green (Continuous Forcing)



(a) V at 850 hPa



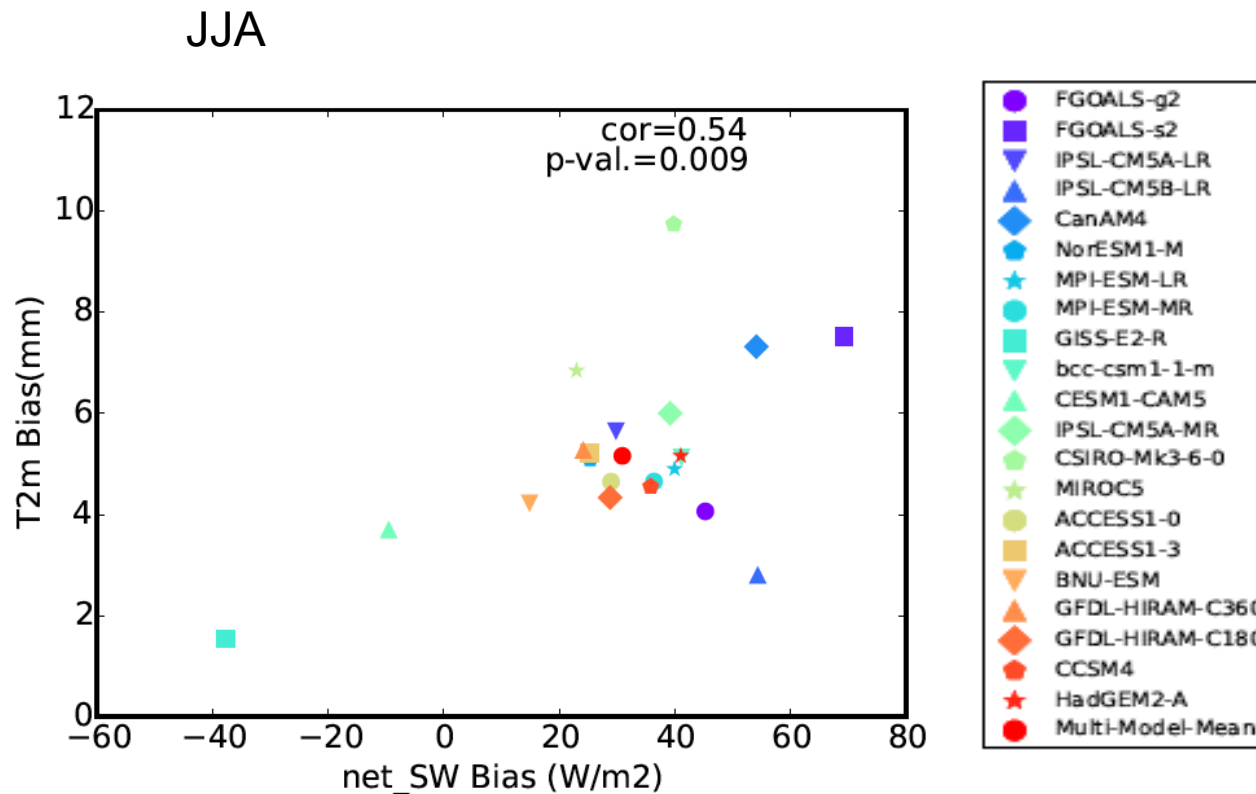
(b) q at 850 hPa



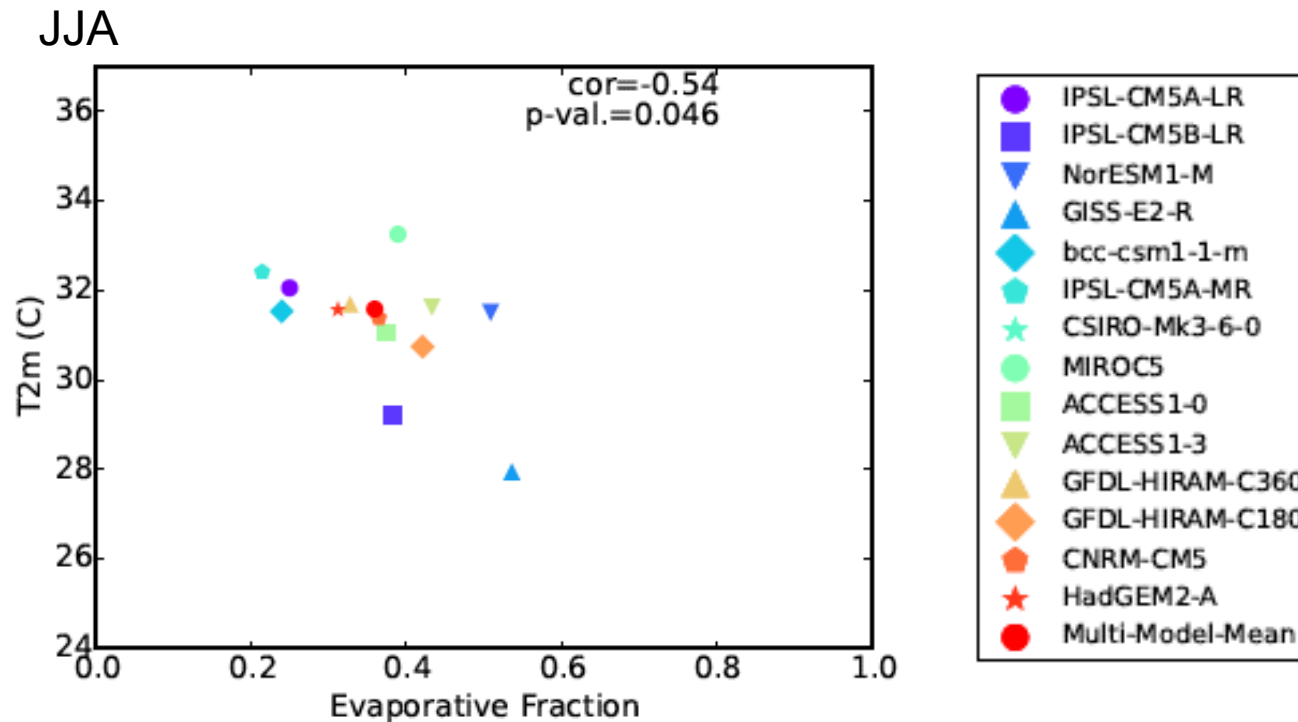
(c) Vq at 850 hPa



Net SWDN bias correlates to T2m bias across models



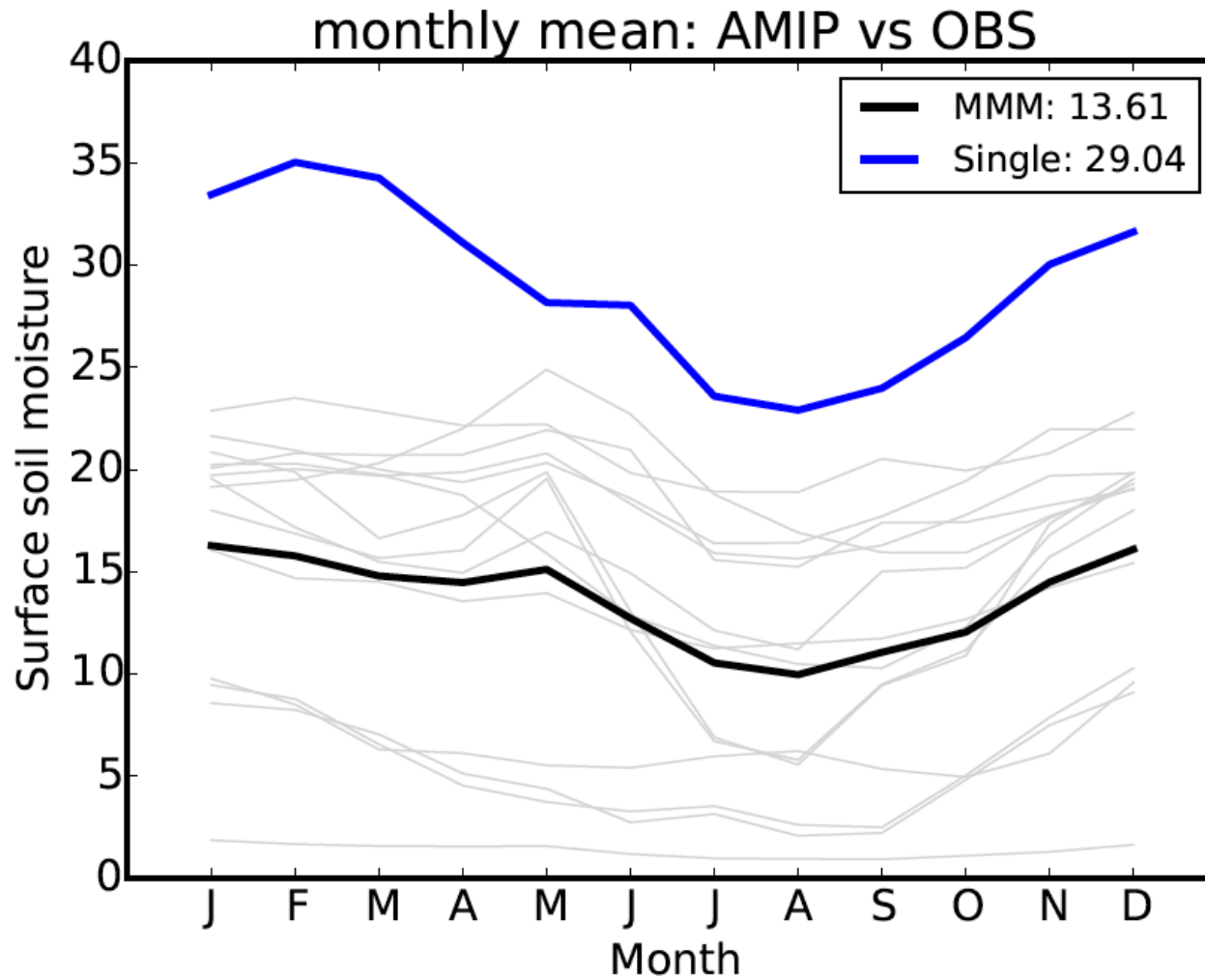
EF correlates to T2m bias across models



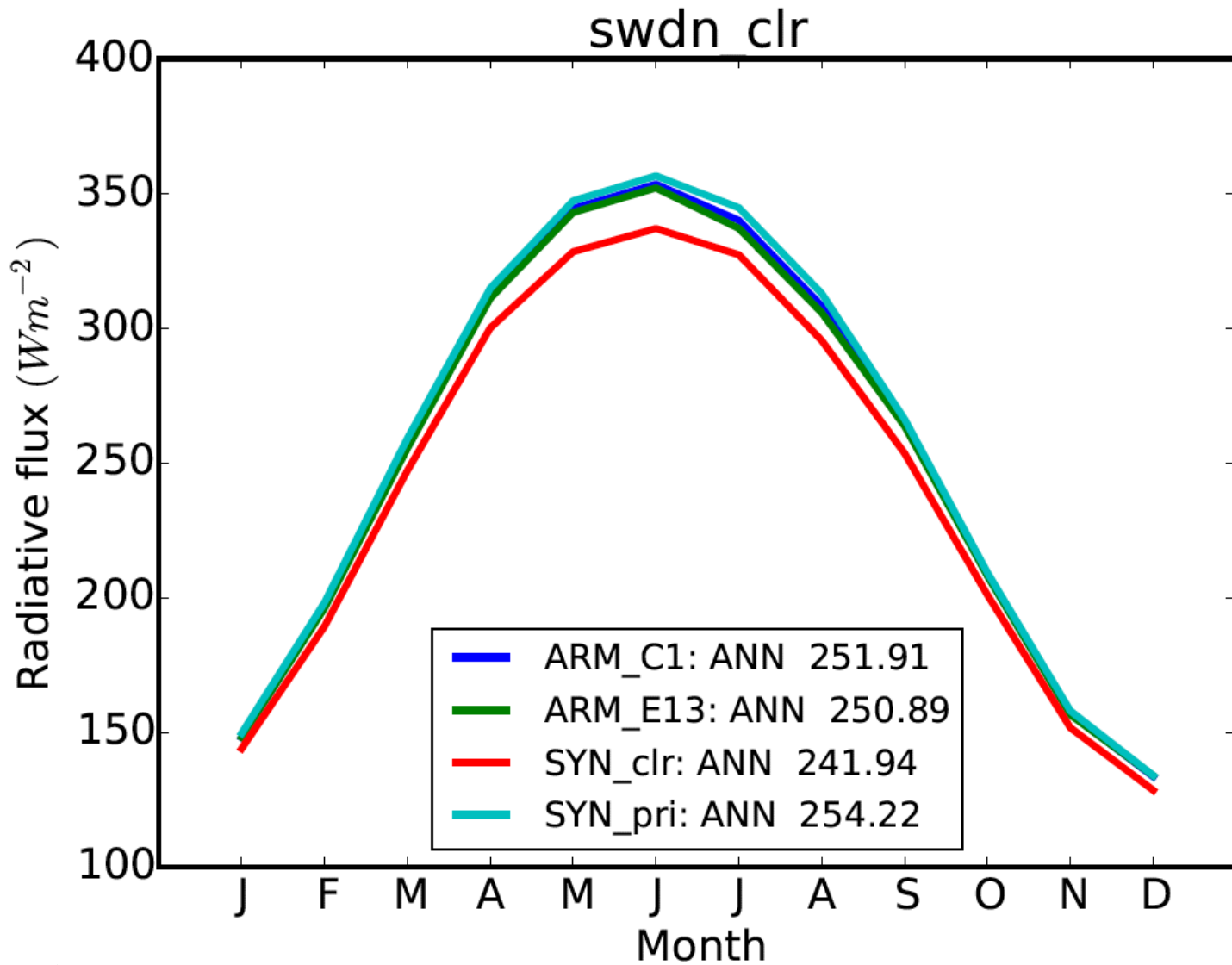
the partition of available energy at the land surface: $EF=LH/(SH+LH)$



Soil moisture



Clear-sky SW Down at two ARM sites E13 (36.66 N, 97.624 W) and C1 (36.60N, 97.485 W)



ARM Data stream:

sgpradflux1longC1 (1999~2011)
Sgpradflux1longE13 (1997~2008)

CERES Data stream:

CERES_SYN1deg (2000~2015)

