

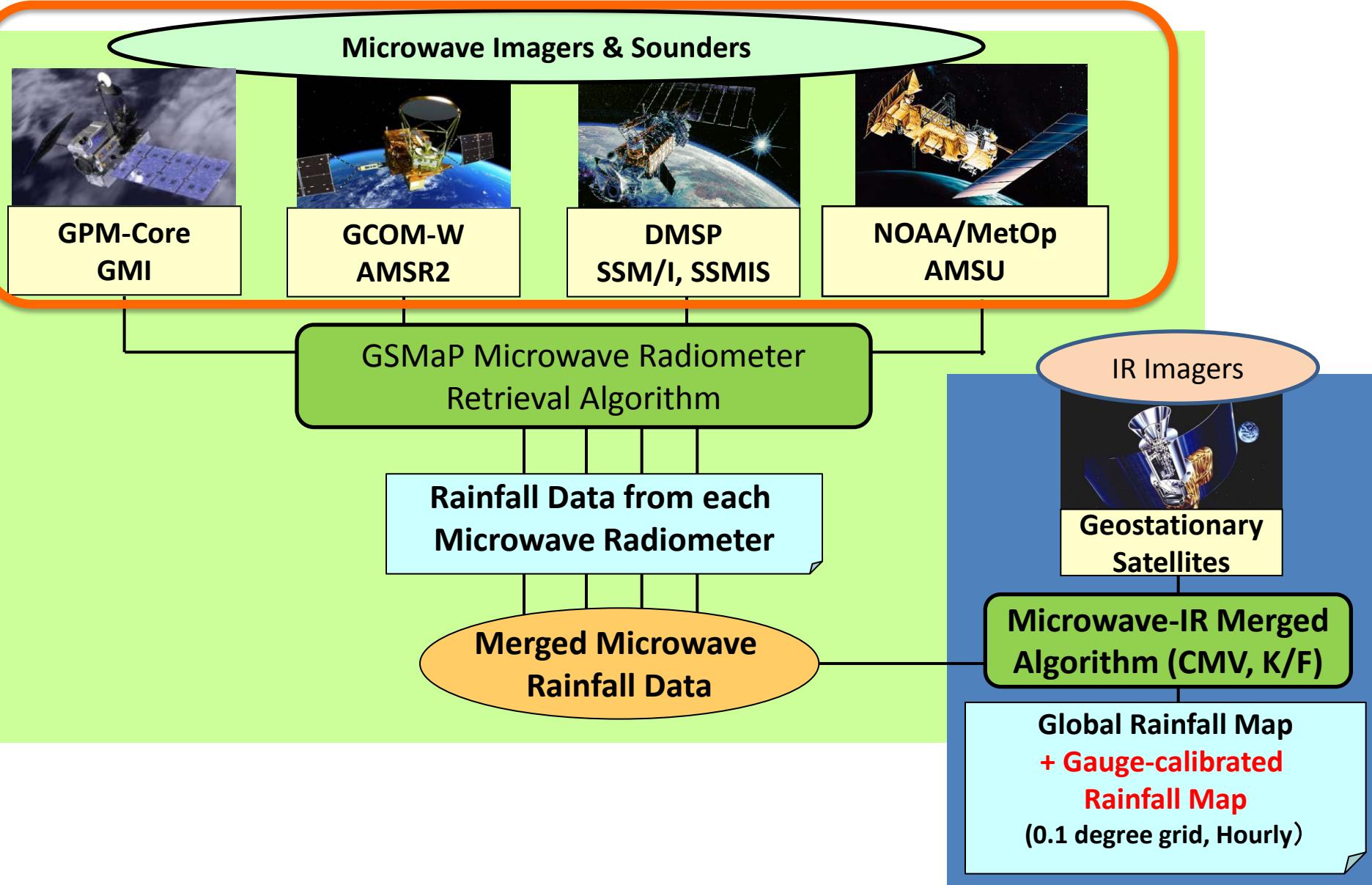
# 高分解能GSMPアルゴリズムの 構造と考え方

牛尾知雄（大阪大）

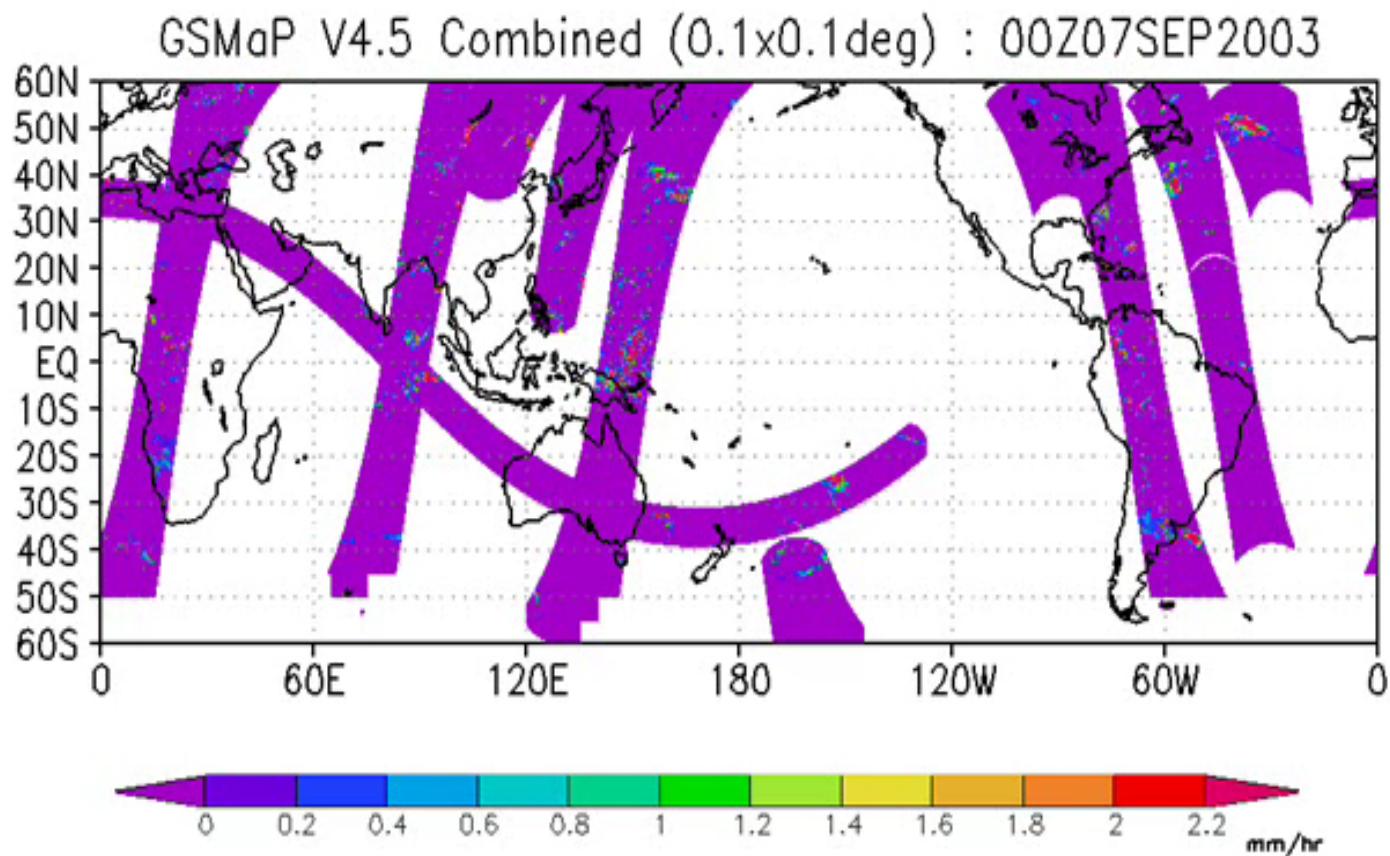
# Current GSMAp products

- GSMAp\_MWR
  - Microwave radiometer product
- GSMAp\_MVK
  - Global precipitation mapping from microwave and infrared radiometric data
- GSMAp\_Gauge
  - Gauge adjusted GSMAp\_MVK
- GSMAp\_NRT and GSMAp\_Now
  - Near real time version of the GSMAp\_MVK product
- GSMAp\_Gauge\_NRT
  - Near real time product of GSMAp\_Gauge

# GSMaP sensors

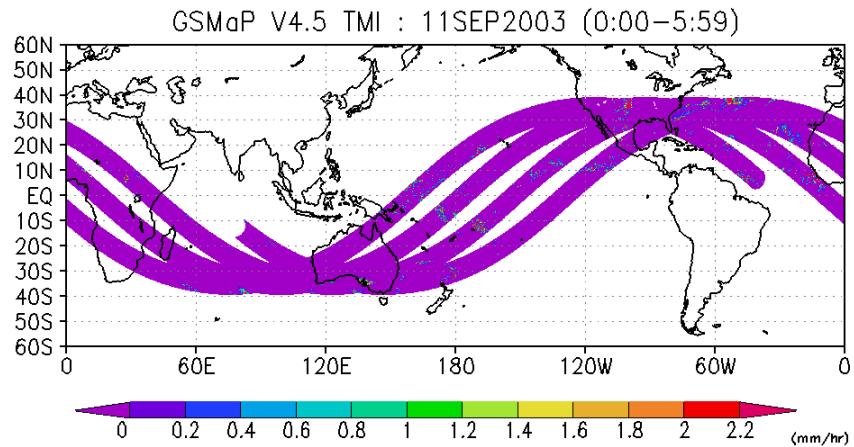


# Coverage of the 6 MWRs in 1 hour

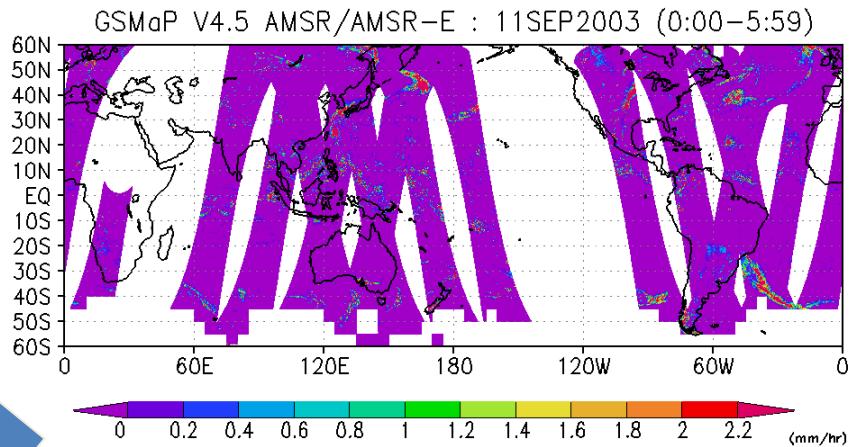


# 6 hourly MWR combined map

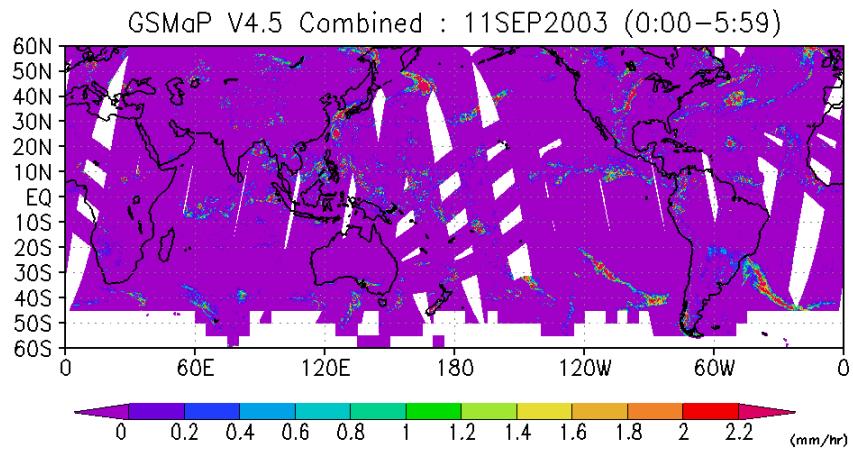
TMI



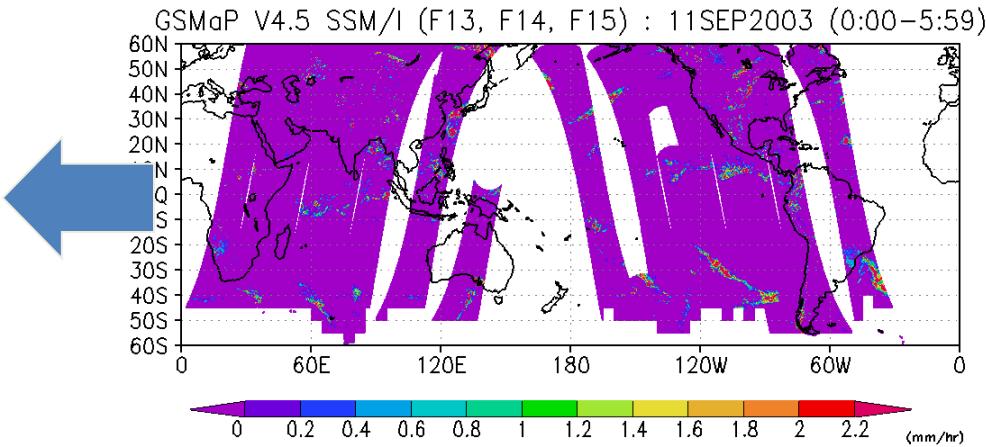
AMSR & AMSR-E



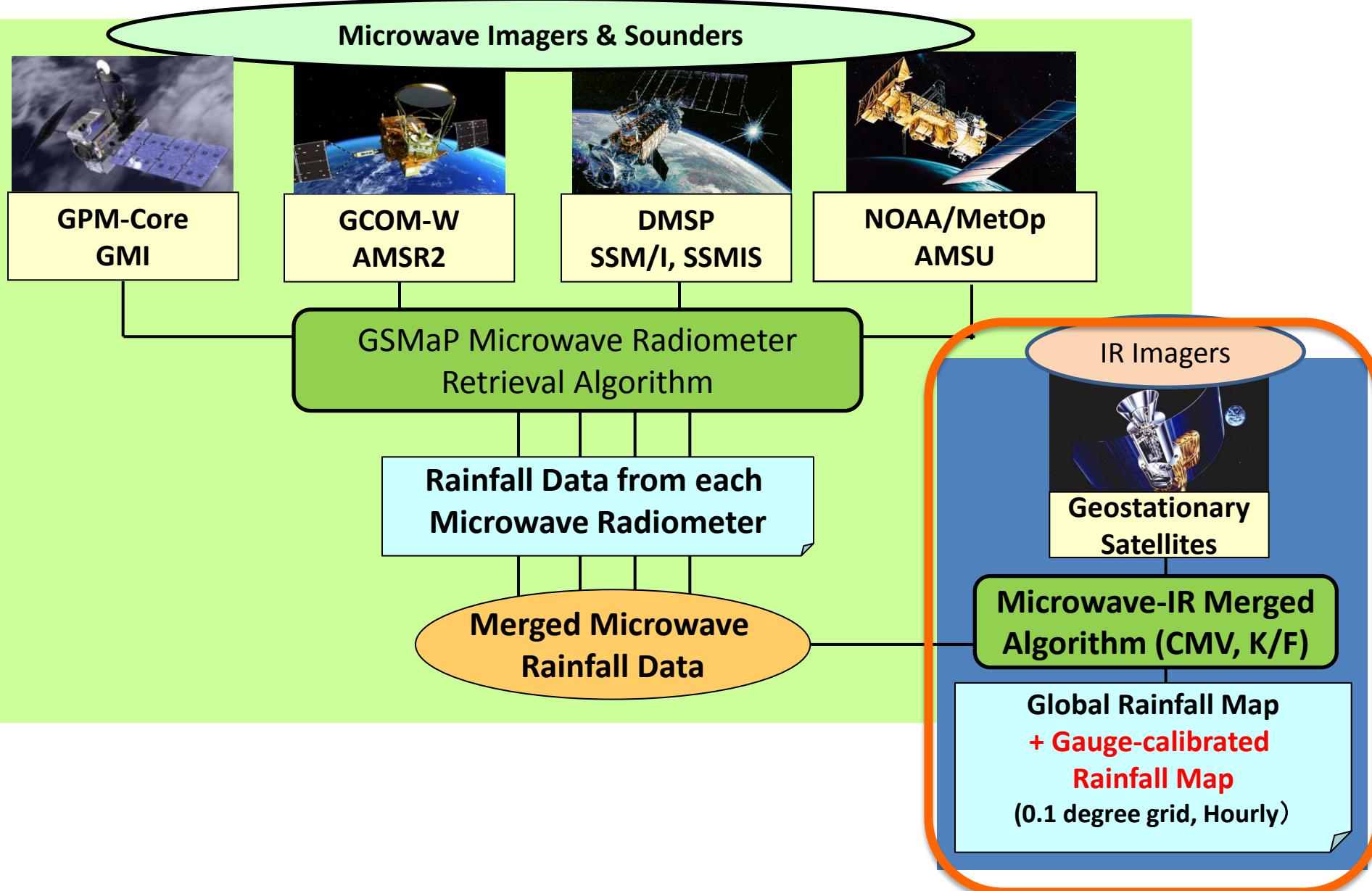
Combined  
6 hourly



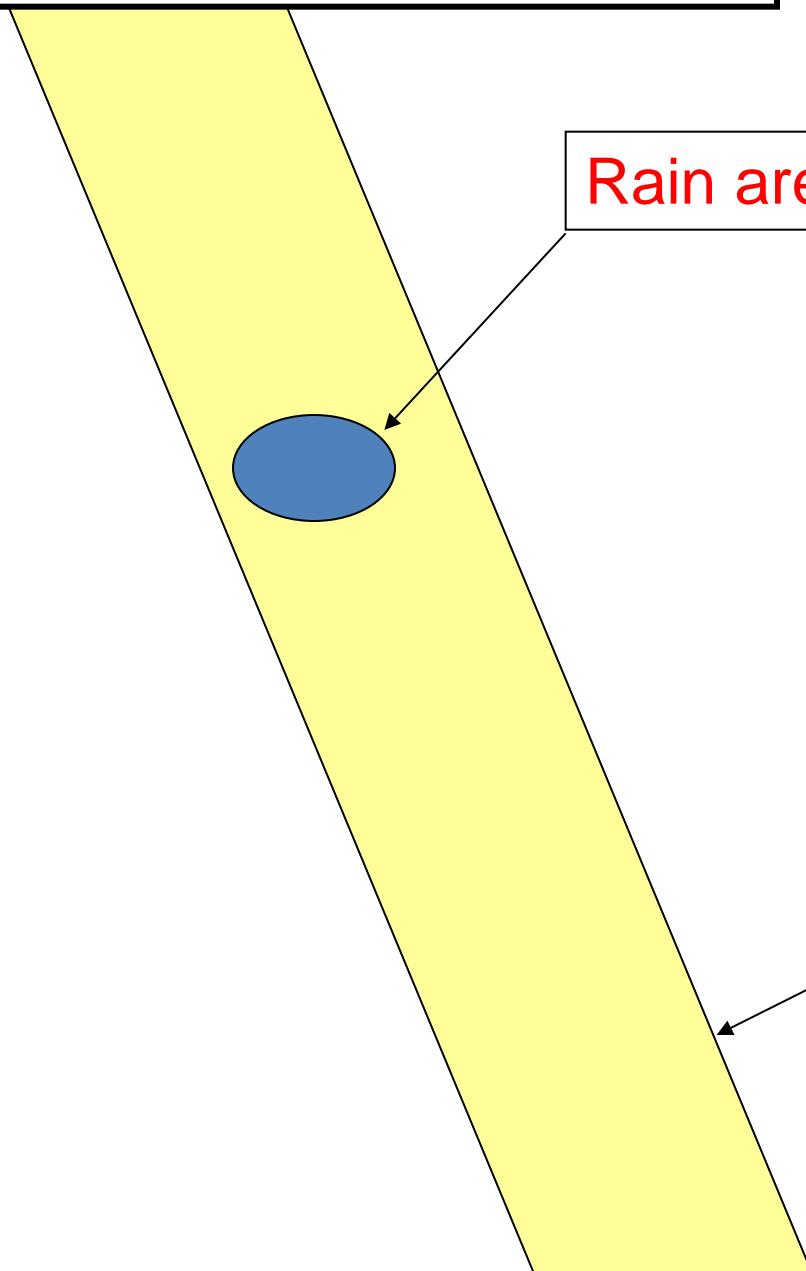
SSM/I (F13, F14, F15)



# GSMaP sensors



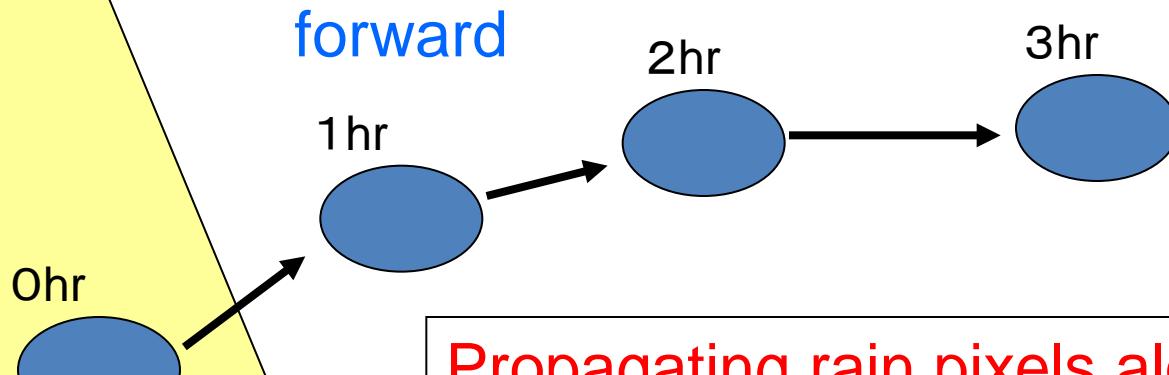
## Forward process



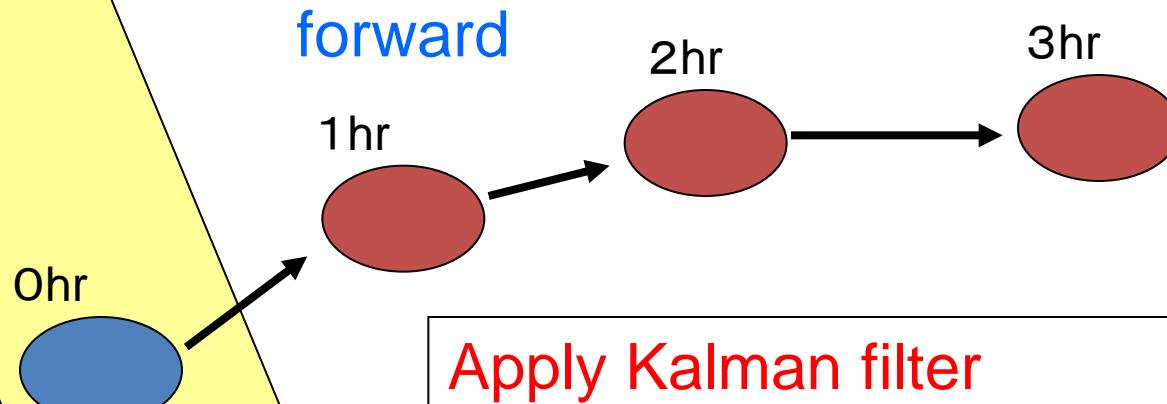
Rain area obtained from the radiometer

Swath of the sensor

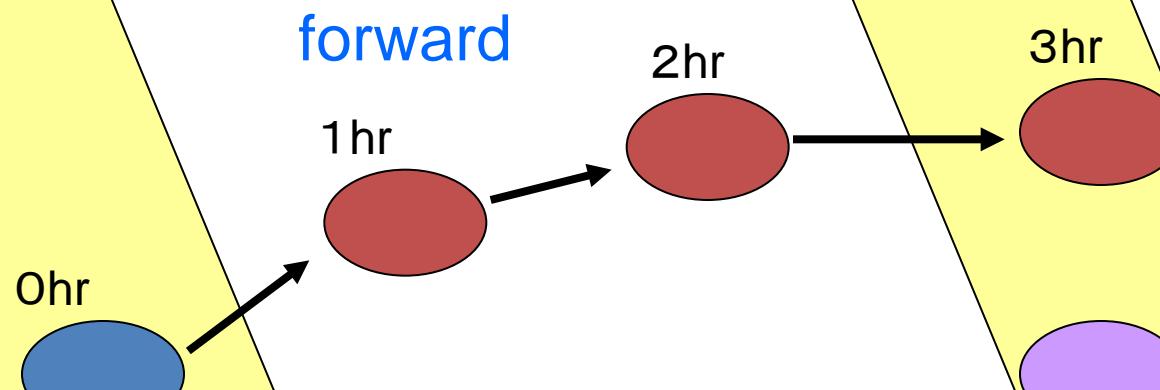
## Forward process



## Forward process



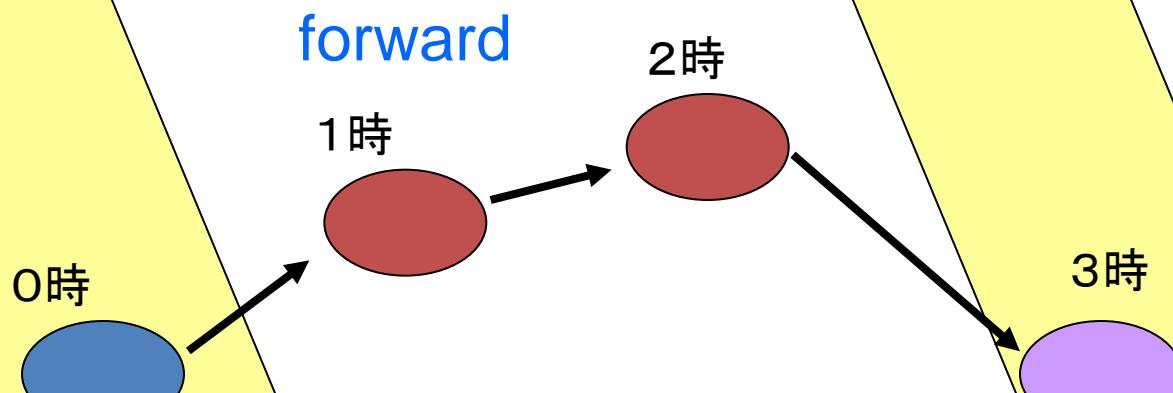
## Forward process



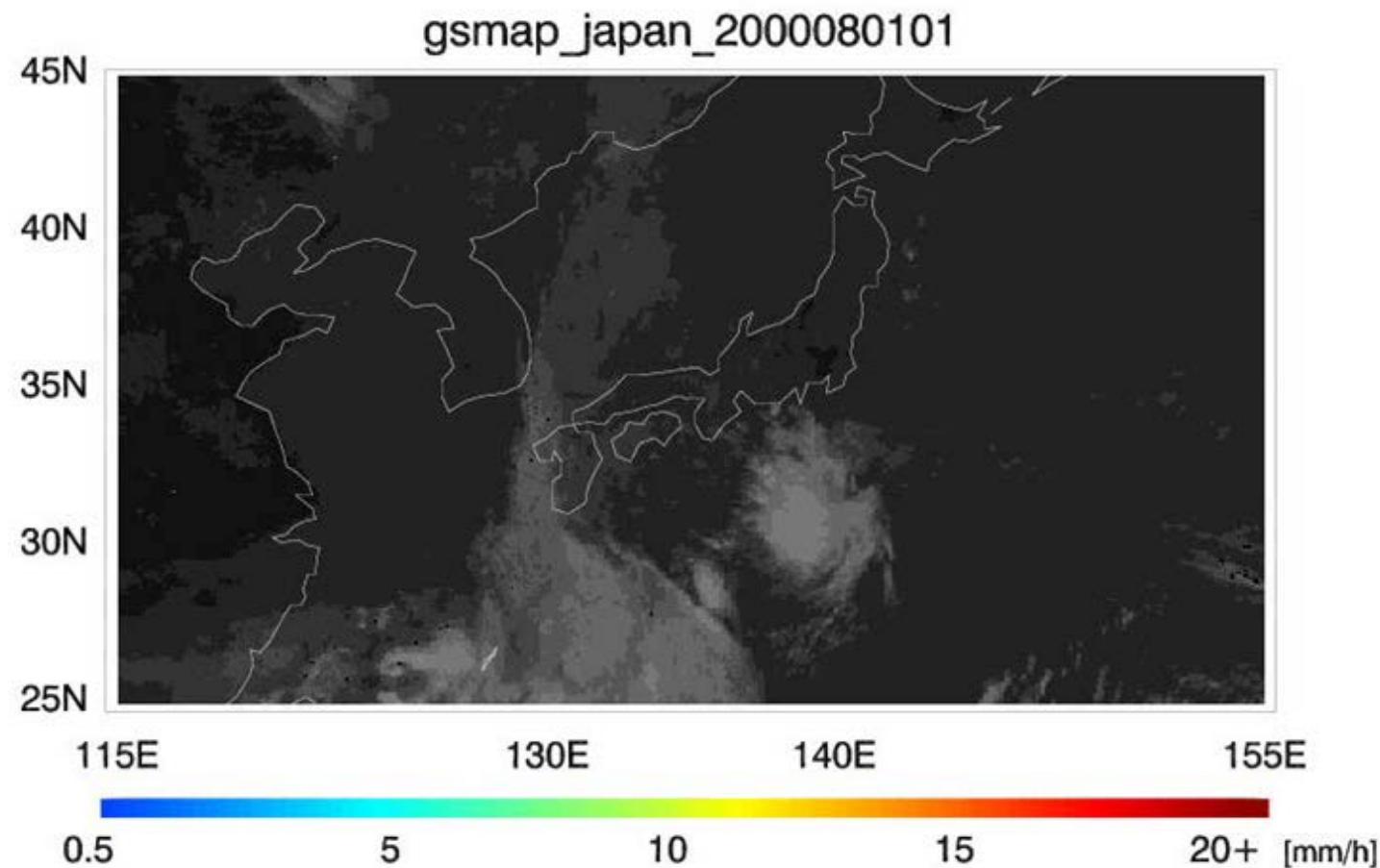
Newly identified rain area

Revisted Path of the microwave sensor

## Forward process



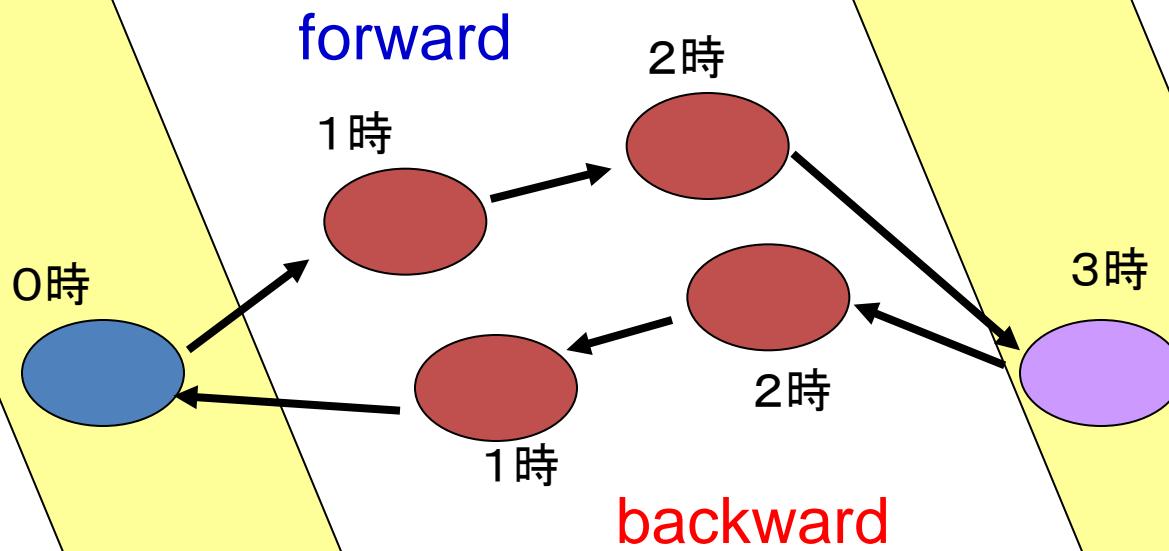
# GSMaP\_NRT



# GSMaP\_NRT

- This is the basic methodology for the GSMaP\_NRT product.
- However, sometimes the error can be large for this near real time product.
- To reduce the bias error, not only the forward process but the backward process are introduced in the GSMaP\_MVK product.

## Backward process

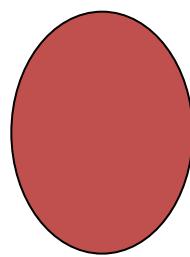


Combine

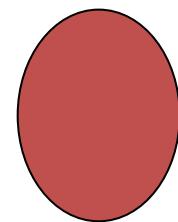
0時



1時



2時

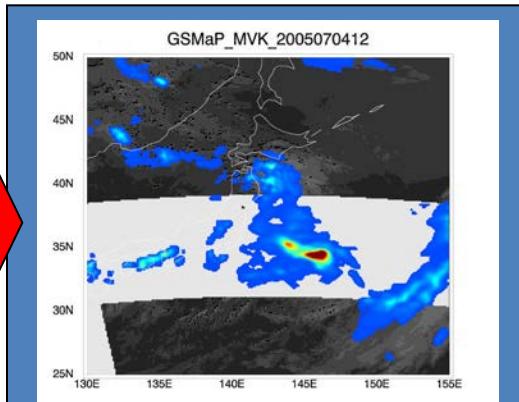


3時

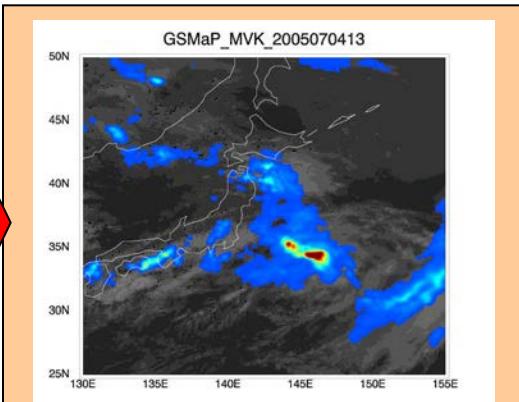


Combine

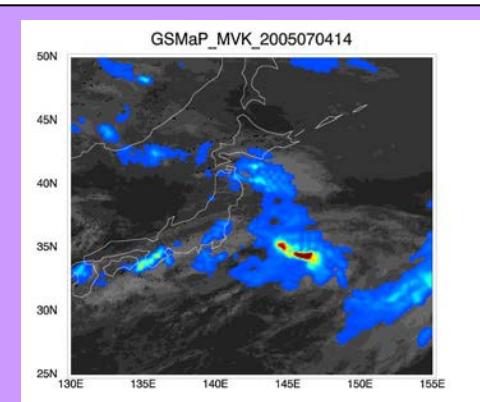
**t**



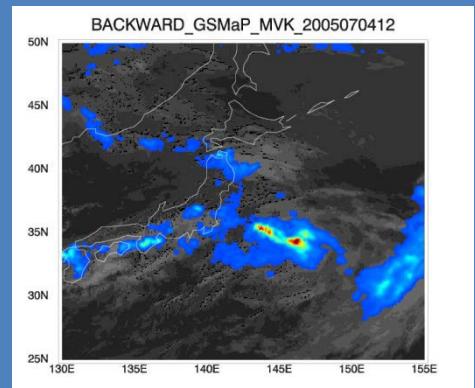
**t+1**



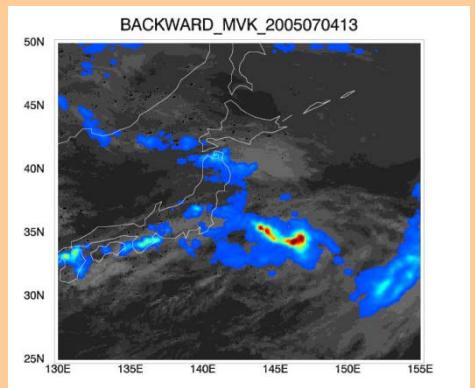
**t+2**



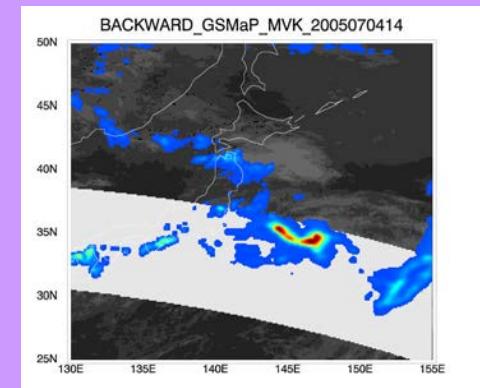
**+**



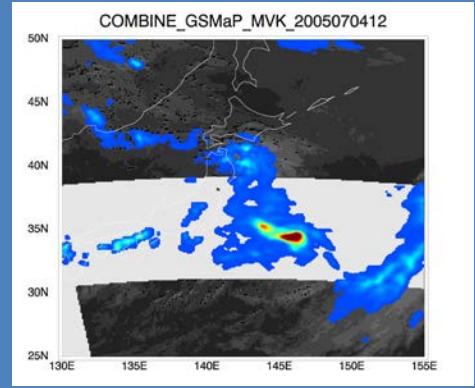
**+**



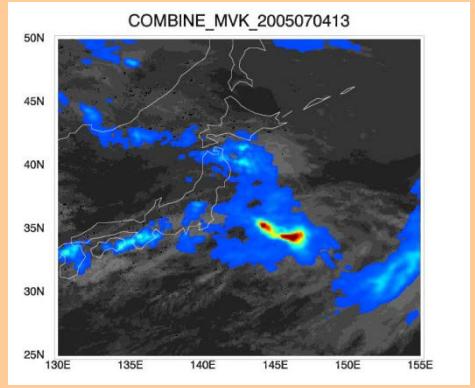
**+**



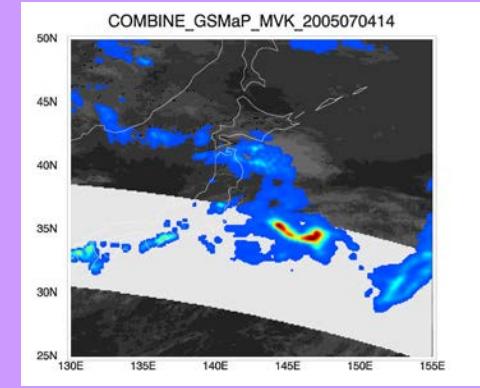
**||**



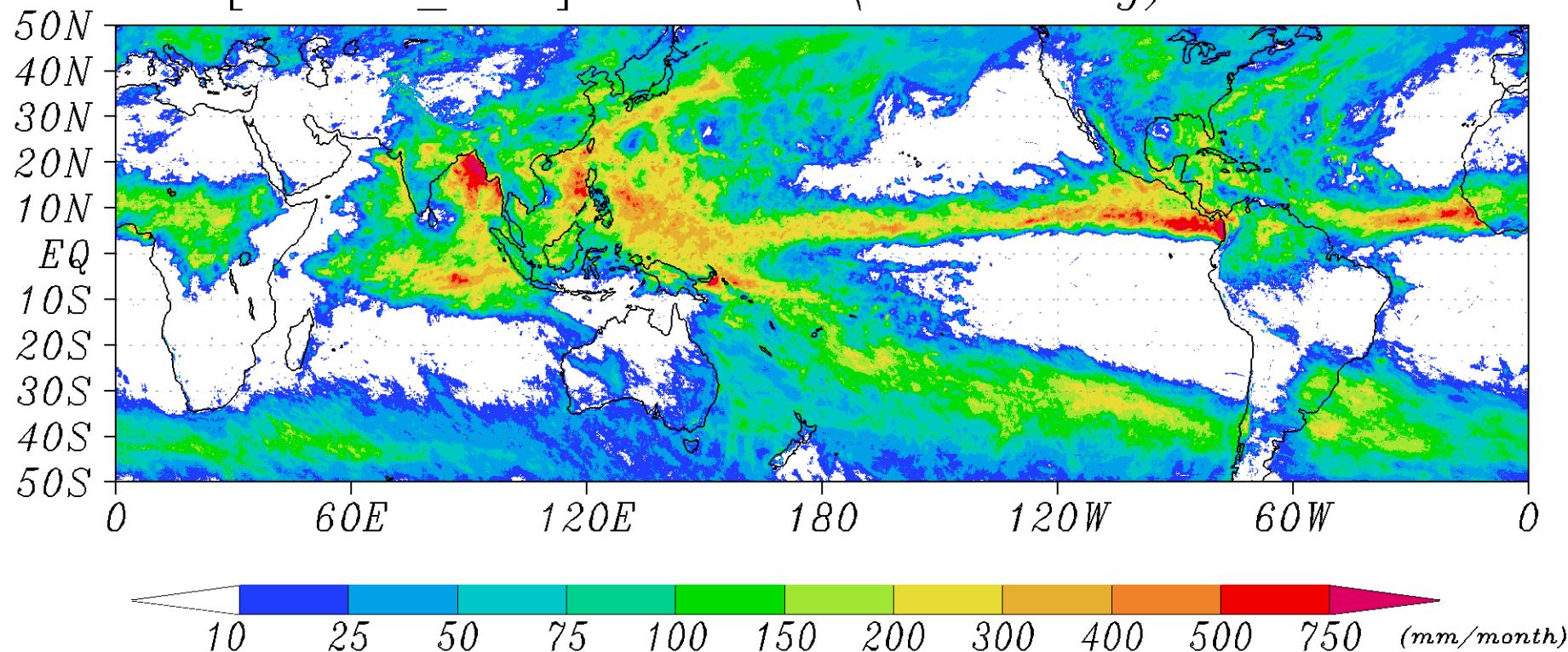
**||**



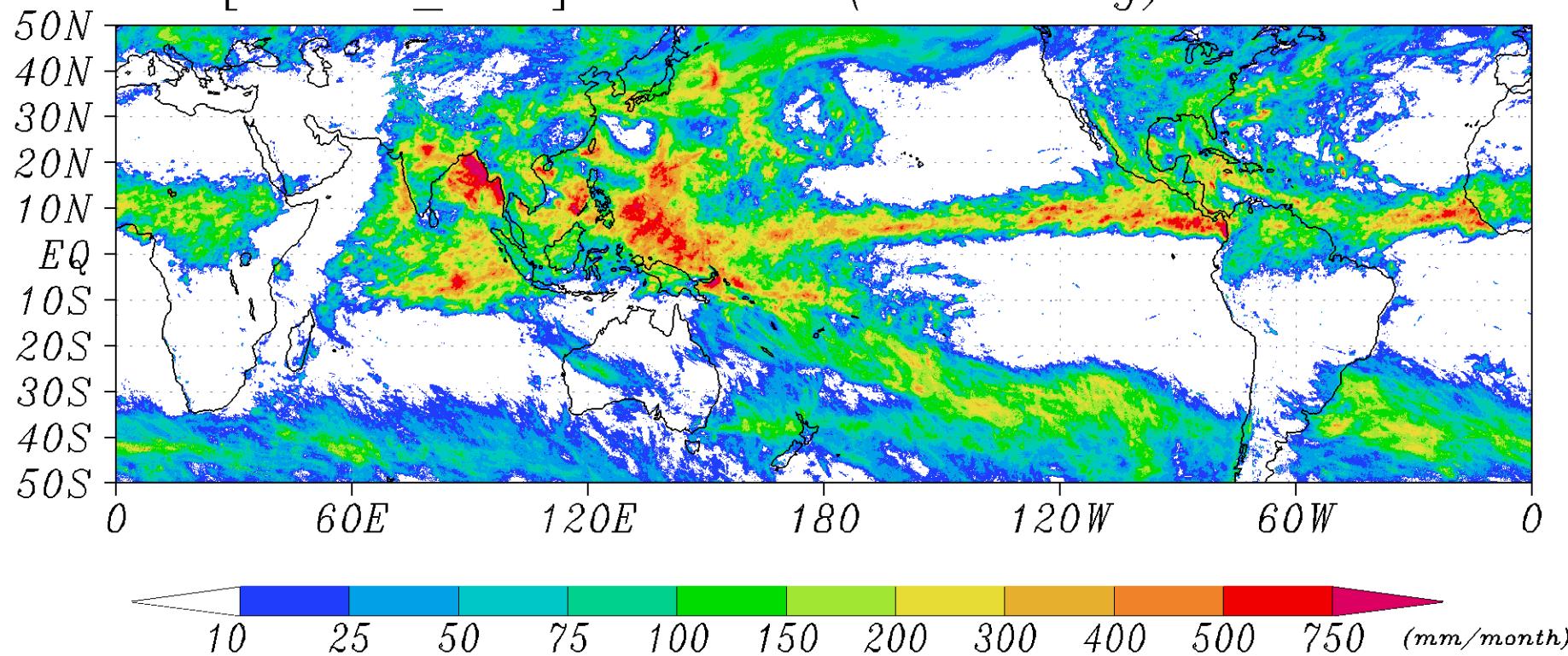
**||**



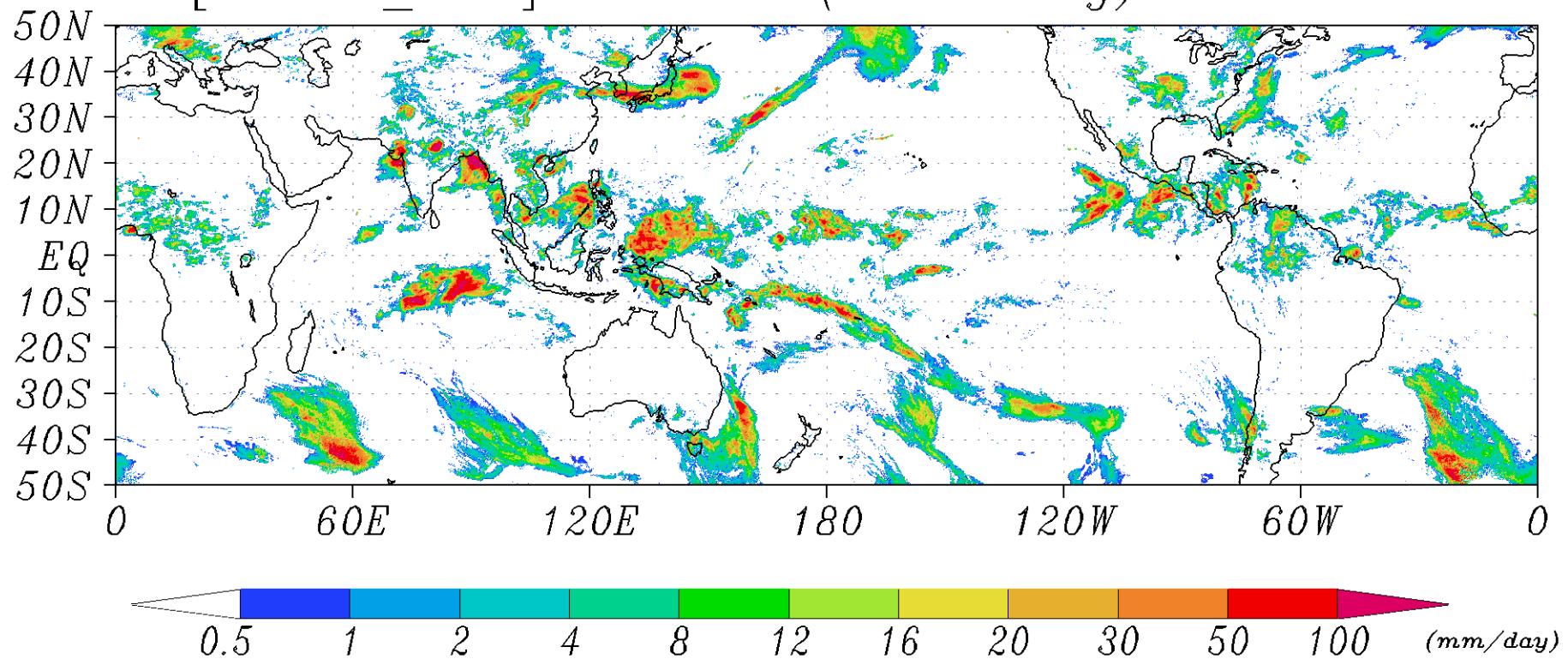
[GSMAp\_MVK] Rain rate( $0.1 \times 0.1$  deg): JJA2005



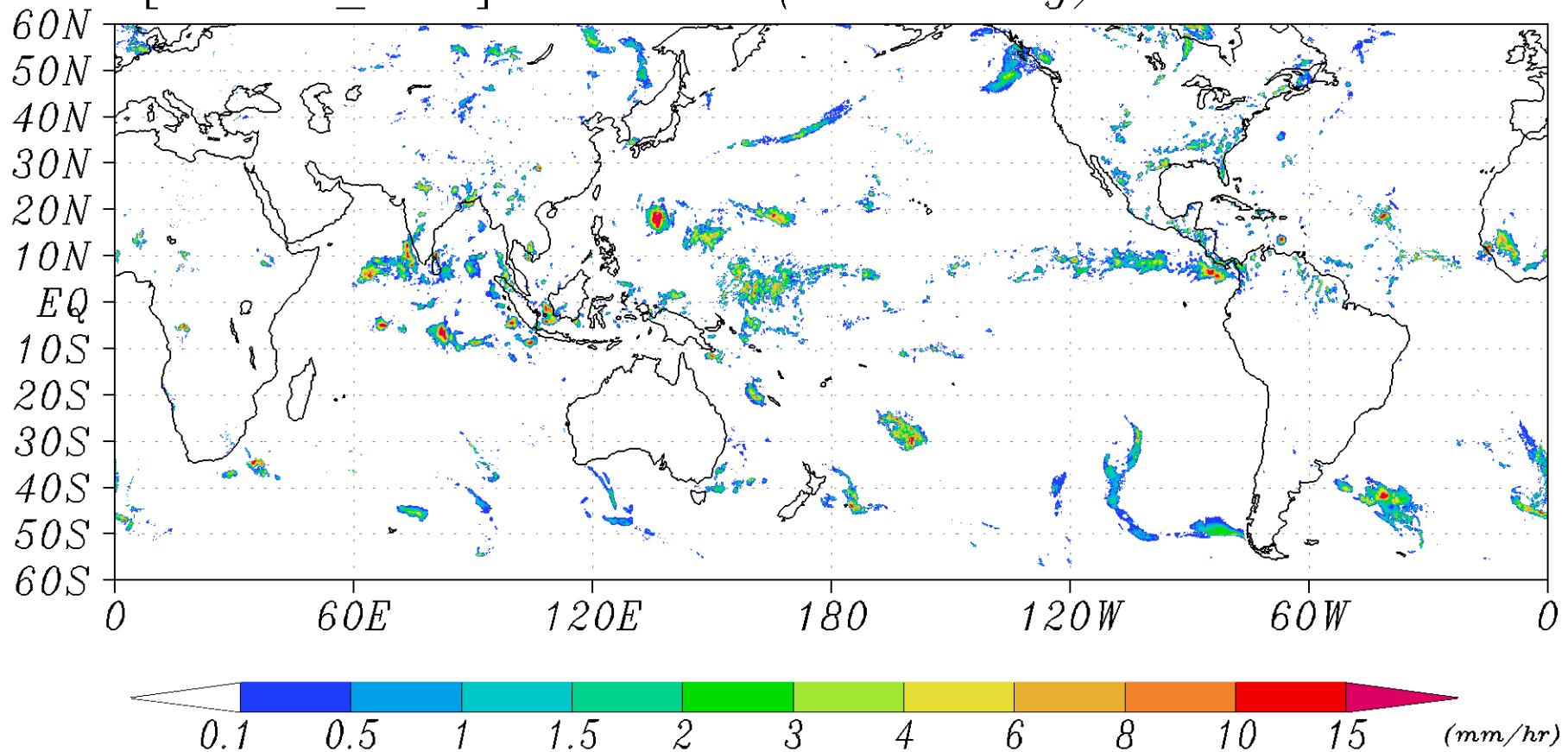
[GSMAp\_MVK] Rain rate( $0.1 \times 0.1$  deg): JUL2005



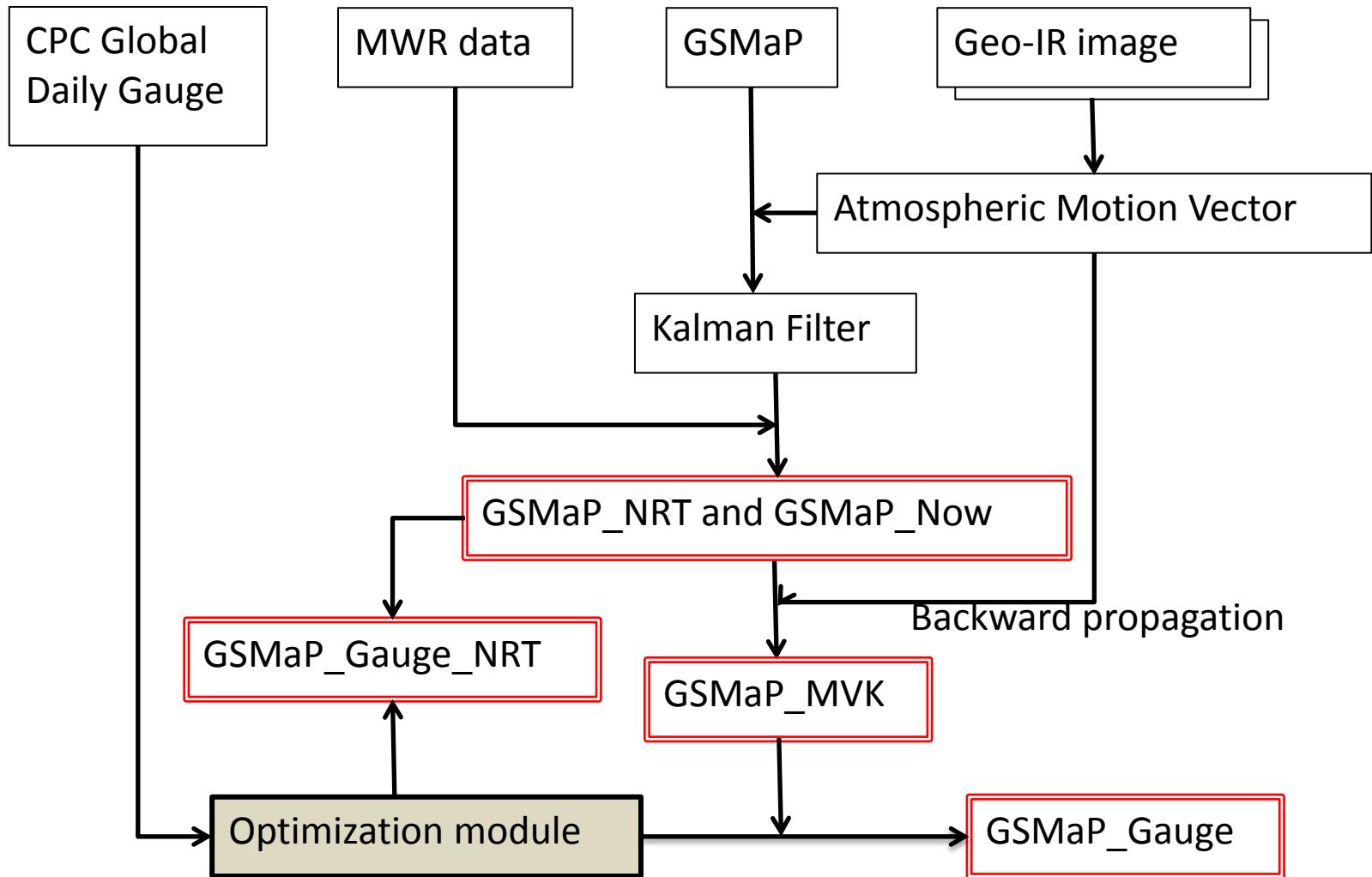
[GSMaP\_MVK] Rain rate( $0.1 \times 0.1$  deg): 01JUL2005



[GSMaP\_MVK] Rain rate( $0.1 \times 0.1$  deg): 00Z15JUL2005

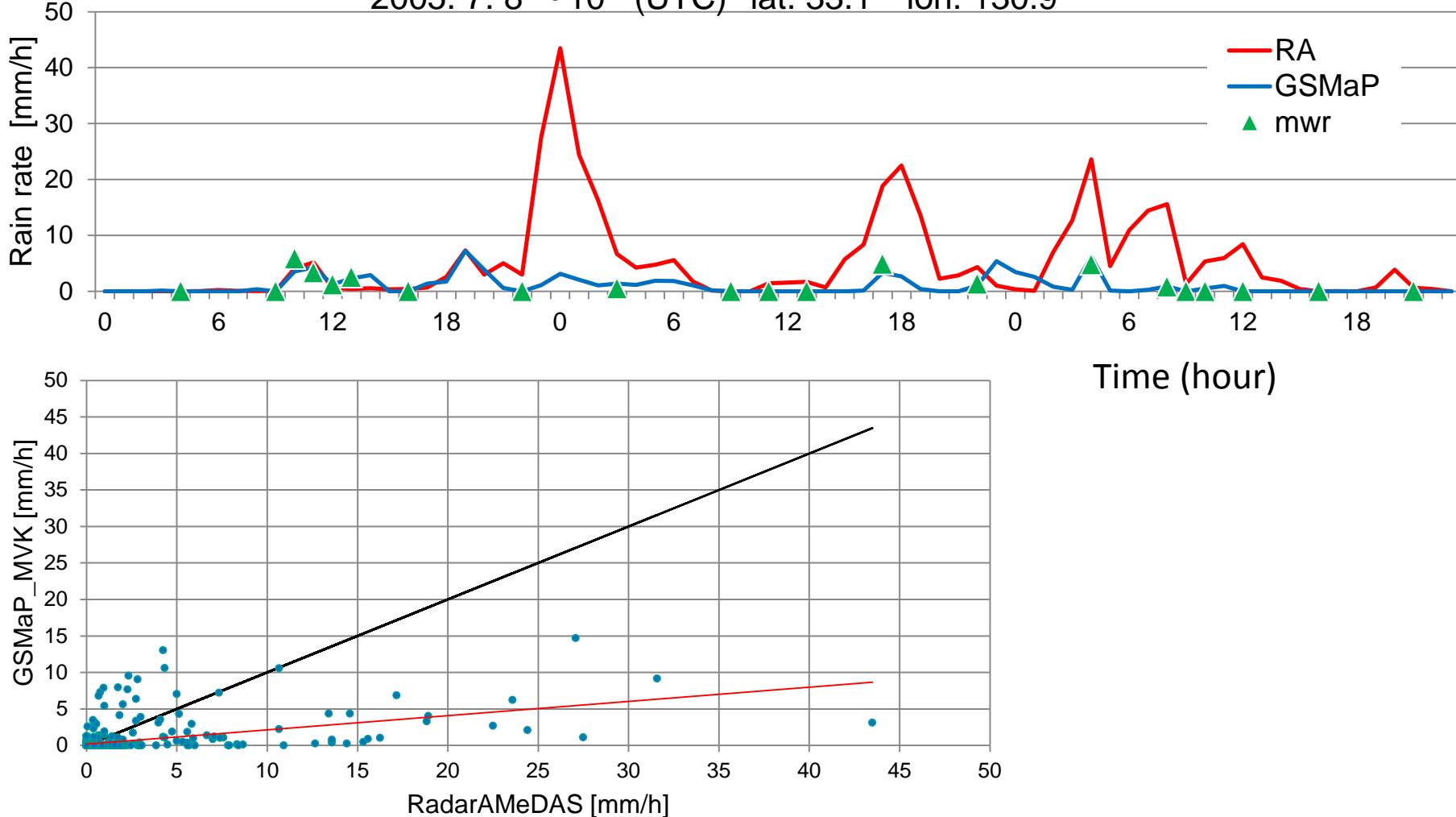


# GSMaP algorithm flow



# Comparison with radar-rain gauge network in Japan

2005. 7. 8 ~ 10<sup>th</sup> (UTC) lat. 33.1° lon. 130.9°



- GSMAp\_MVK tends to underestimate the rain fall rate with 1 hour resolution
- But, basically the GSMAp\_MVK's estimates strongly depends on the estimates from microwave radiometer

# What is our approach?

- Problem
  - Potential data sources for the gauge adjusted GSMAp\_MVK product
    - GPCC data set
      - Monthly base/ 0.5 degree grid
      - Conversion of the time resolution of Monthly -> Hourly is difficult
    - CPC global gauge data analysis
- Solution
  - CPC global gauge data analysis by Chen et al. (2008)
  - Daily base/ 0.5 degree grid

# GSMaP\_Gauge algorithm

$$\begin{cases} \mathbf{x}_{n+1} = \mathbf{x}_n + \mathcal{N}(\mu_w, \sigma_w^2) \\ \mathbf{y}_n = c\mathbf{x}_n + \mathcal{N}(\mu_v, \sigma_v^2) \end{cases}$$

$$J(\mathbf{x}) = J_1(\mathbf{x}) + \lambda J_2(\mathbf{x}, W)$$

$$J_1(\mathbf{x}) = -\ln \Pr(\mathbf{x}, \mathbf{y})$$

$$J_2(x) = \frac{1}{2} \left( \sum_{n=1}^{24} \mathbf{x}_n - W \right)^2$$

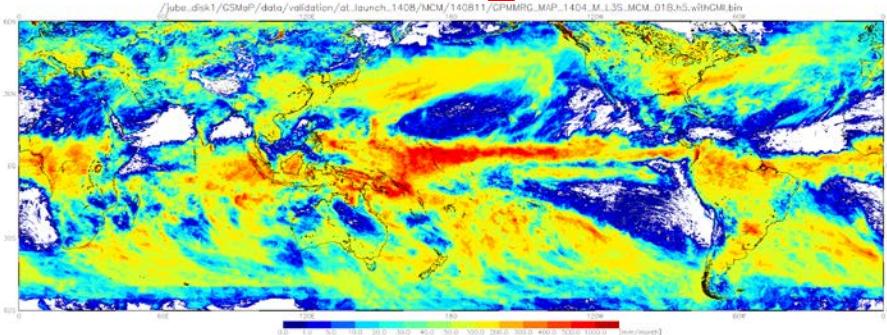
$\mathbf{y}_n$	: Estimated precipitation rate (GSMaP MVK)
$\mathbf{x}_n$	: Precipitation Rate
$c$	: Coefficient of proportionality of the estimation of precipitation
$\mathcal{N}$	: Standard distribution
$\mu_v$	: Variance of estimation error
$\mu_w$	: Change rate of precipitation
$\sigma_v$	: Estimation error
$\sigma_w$	: Variance of precipitation
$W$	: Daily precipitation
$n$	: hour
$\Pr(\mathbf{x}, \mathbf{y})$	: Probability of $\mathbf{x}$ and $\mathbf{y}$ (Gaussian distribution)
$\lambda$	: weight

- Data used in the GSMaP\_Gauge algorithm
  - NOAA CPC Unified Gauge-Based Analysis of Global Daily Precipitation (CPC).
  - The data is a daily and 0.5 grid precipitation fields on a global basis.
- Based on the assumption that the GSMaP\_Gauge – CPC Gauge data (Gauge term) has the gaussian distribution, the probability density function of the GSMaP\_Gauge estimation multiplied by the Gauge term is maximized

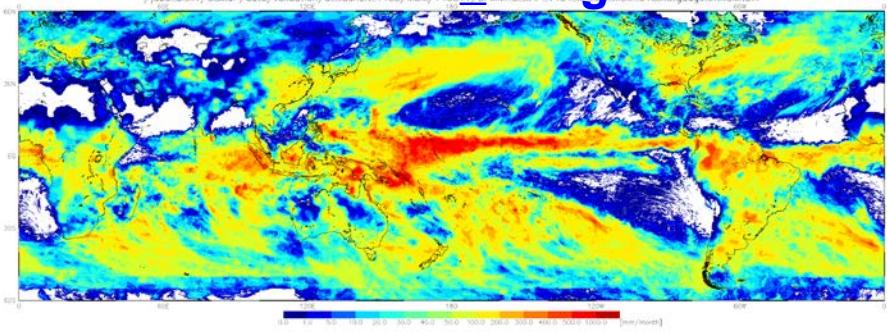
# Example of GPM-GSMaP monthly

- Monthly Mean in April 2014 (V03A)

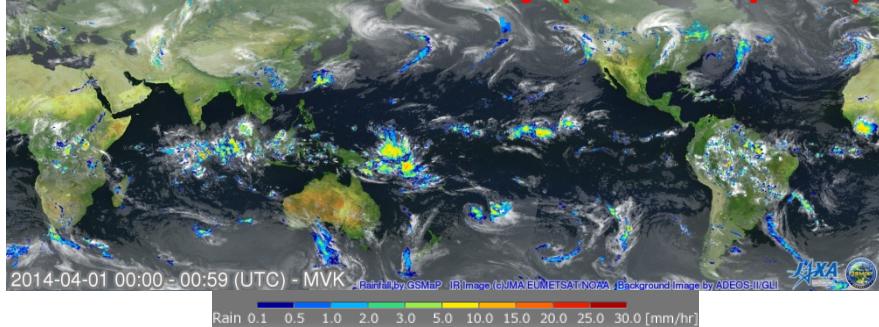
**GSMaP MVK**



**GSMaP Gauge**

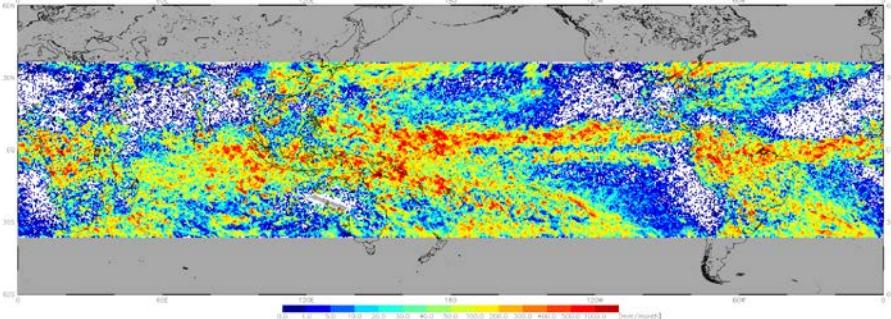


**GPM-GSMaP MVK hourly (00Z on April 1)**

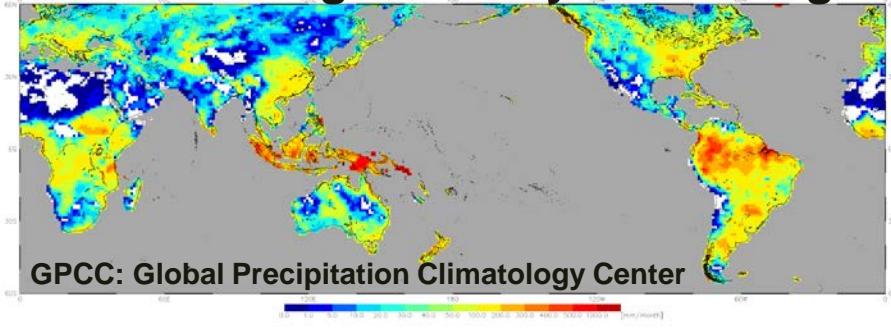


**Reference**

**TRMM/PR 3A25**

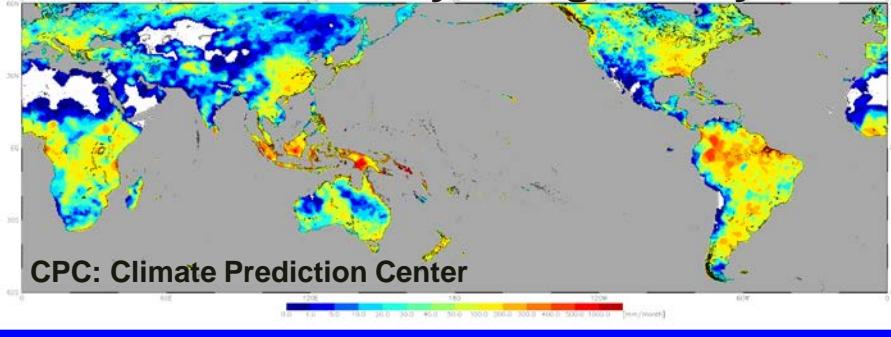


**GPCC Gauge Monthly Monitoring**



GPCC: Global Precipitation Climatology Center

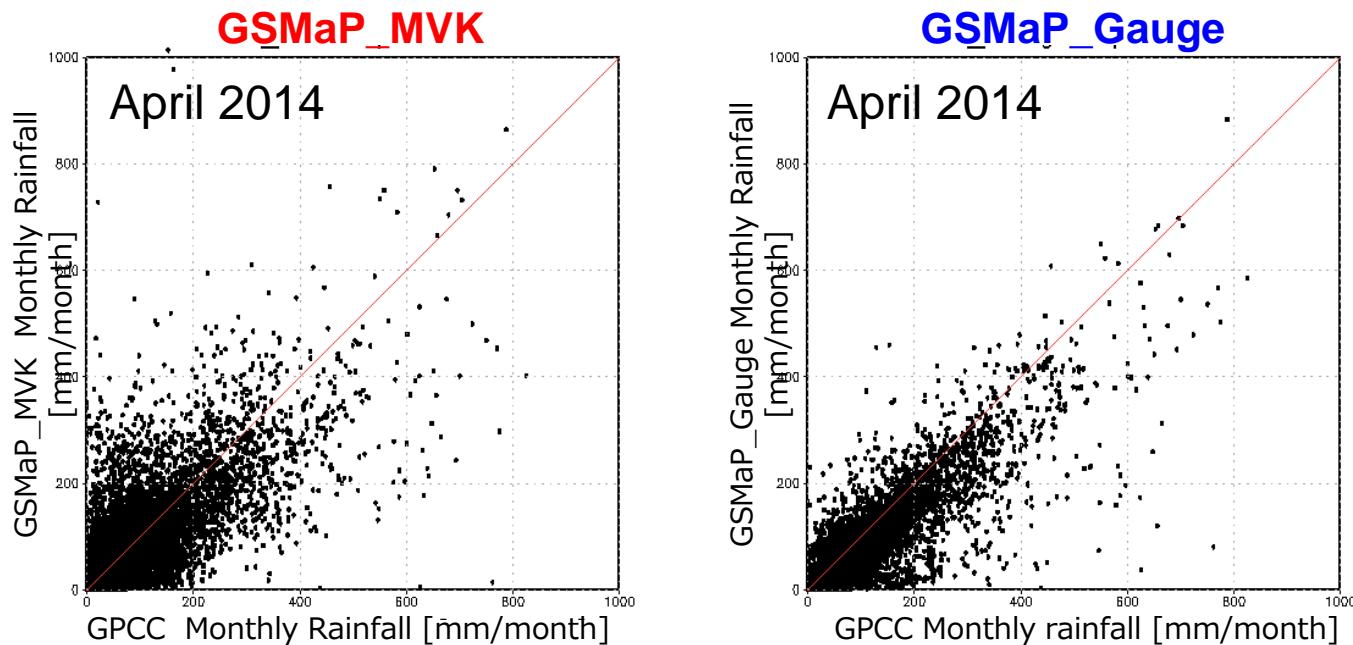
**NOAA/CPC Daily Gauge Analysis**



CPC: Climate Prediction Center

## Evaluation of GPM-GSMaP

- \* GPM-GSMaP compared with GPCC Monthly accumulated Gauge data.  
[GSMaP\\_Gauge](#) (calibrated by NOAA CPC Daily Gauge Analysis) shows 20-30% better accuracy in RMSE than GSMaP\_MVK.



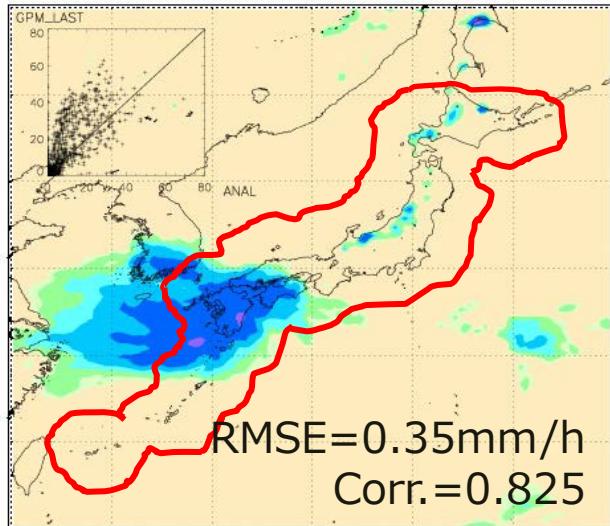
	GSMaP_MVK [mm/month]			GSMaP_Gauge [mm/month]		
	Bias	R	RMSE	Bias	R	RMSE
April 2014	-0.54	0.73	62.45	-15.58	0.88	44.26
May 2014	1.64	0.68	72.22	-17.93	0.87	47.99
June 2014	-4.40	0.69	83.10	-12.70	0.82	64.16

# Evaluation of GPM-GSMaP

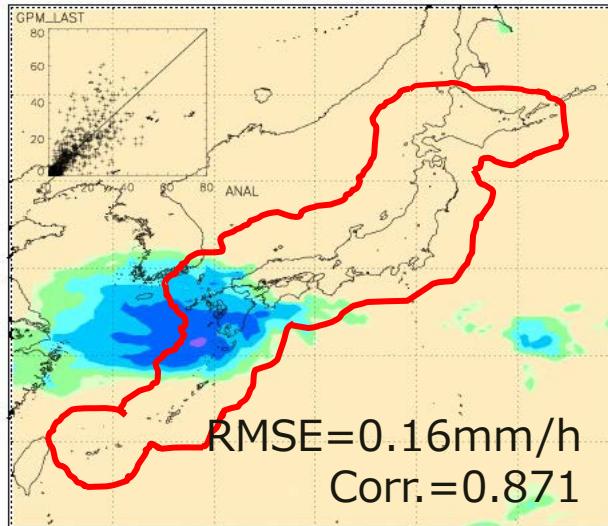
- Daily averaged rainfall around Japan in 0.25 degree grid was compared with JMA's Radar AMeDAS (gauge-calibrated radar analysis rainfall).

An example on Apr. 12, 2014

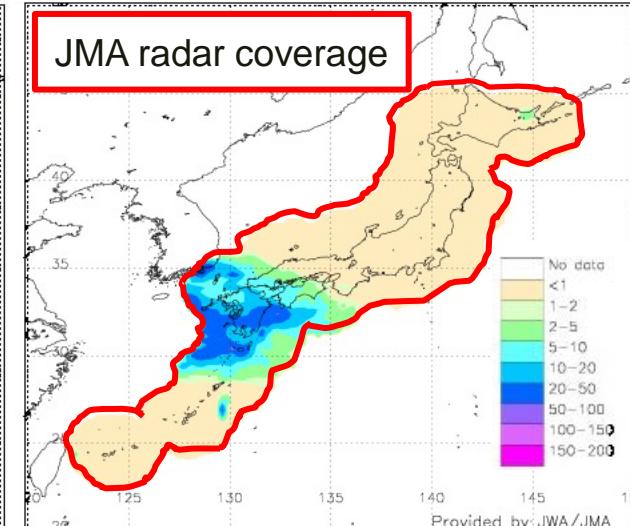
**GSMaP\_MVK**



**GSMaP\_Gauge**



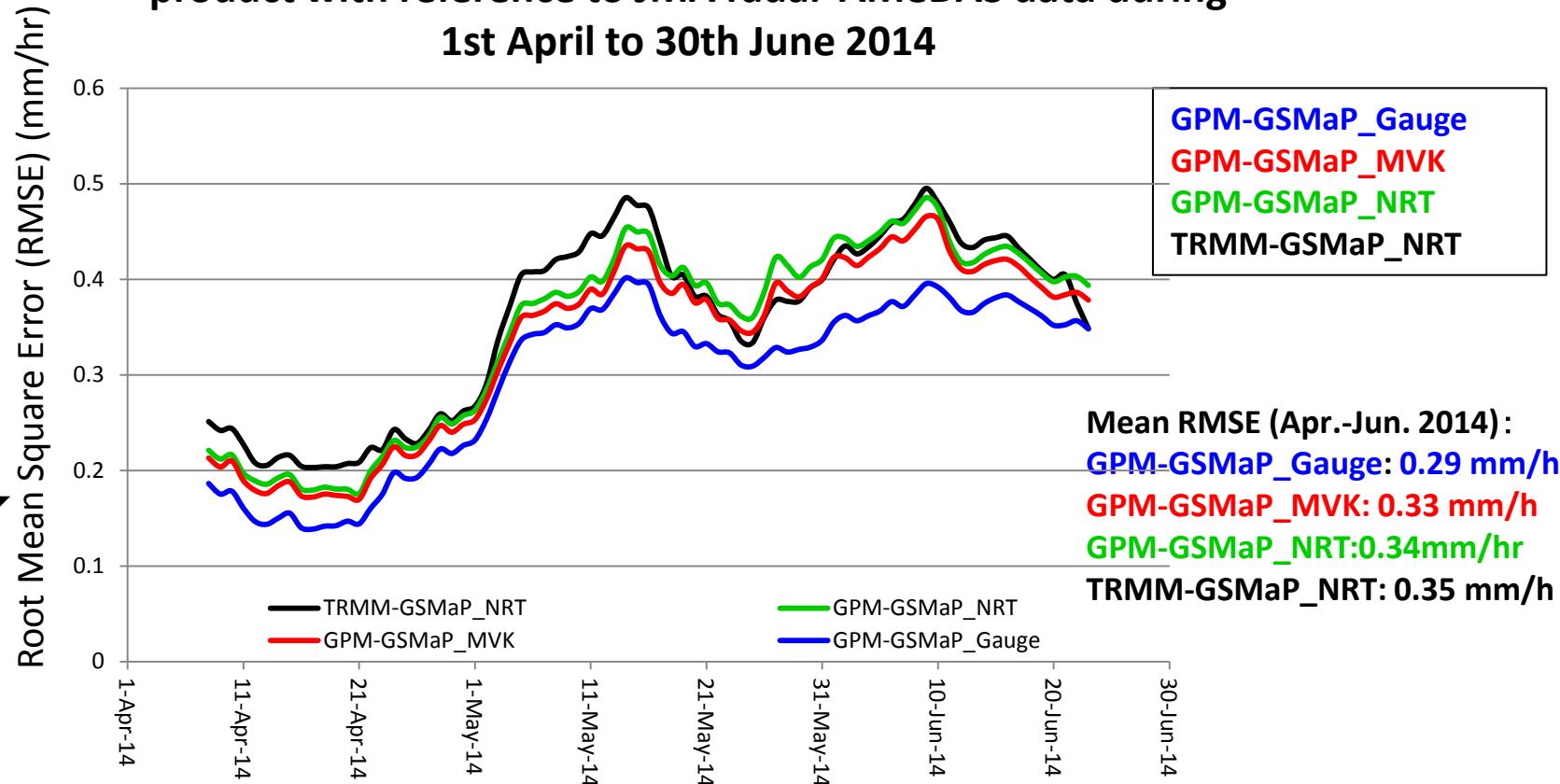
**Gauge-Radar Analysis**



→ **GSMaP\_Gauge** shows better correlation with less Root Mean Square Error (RMSE) on Apr. 12, 2014.

# Evaluations by RMSE

Daily series of Root Mean Square Error (RMSE) for GSMAp product with reference to JMA radar-AMeDAS data during  
**1st April to 30th June 2014**



- **GSMAp\_Gauge** shows the least RMSE.
- **GSMAp\_Gauge** algorithm is working well.

# However.....

- GSMAp\_Gauge has latencies of a few days.
- The latencies of the GSMAp\_Gauge sometimes cause problem for real time applications (ex. Flash flood warning system...).
- I received so many requests for real time version of the GSMAp\_Gauge product.

# GSMaP\_Gauge\_NRT algorithm

$$\begin{cases} \mathbf{x}_{n+1} = \mathbf{x}_n + \mathcal{N}(\mu_w, \sigma_w^2) \\ \mathbf{y}_n = c\mathbf{x}_n + \mathcal{N}(\mu_v, \sigma_v^2) \end{cases}$$

$$J(\mathbf{x}) = J_1(\mathbf{x}) + \lambda J_2(\mathbf{x}, W)$$

$$J_1(\mathbf{x}) = -\ln \Pr(\mathbf{x}, \mathbf{y})$$

$$J_2(\mathbf{x}) = \frac{1}{2} \left( \sum_{n=1}^{24} \mathbf{x}_n - W \right)^2$$

$\mathbf{y}_n$	: Estimated precipitation rate (GSMaP MVK)
$\mathbf{x}_n$	: Precipitation Rate
$c$	: Coefficient of proportionality of the estimation of precipitation
$\mathcal{N}$	: Standard distribution
$\mu_v$	: Variance of estimation error
$\mu_w$	: Change rate of precipitation
$\sigma_v$	: Estimation error
$\sigma_w$	: Variance of precipitation
$W$	: Daily precipitation
$n$	: hour
$\Pr(\mathbf{x}, \mathbf{y})$	: Probability of $\mathbf{x}$ and $\mathbf{y}$ (Gaussian distribution)
$\lambda$	: weight

In the real time product, unfortunately, we do not have real time gauge data. Instead, we prepare the data base which characterize the precipitation in this model. Those are sigmaw, sigmav and c.

GSMaP Gauge

$\mathbf{x}$ : True precipitation rate

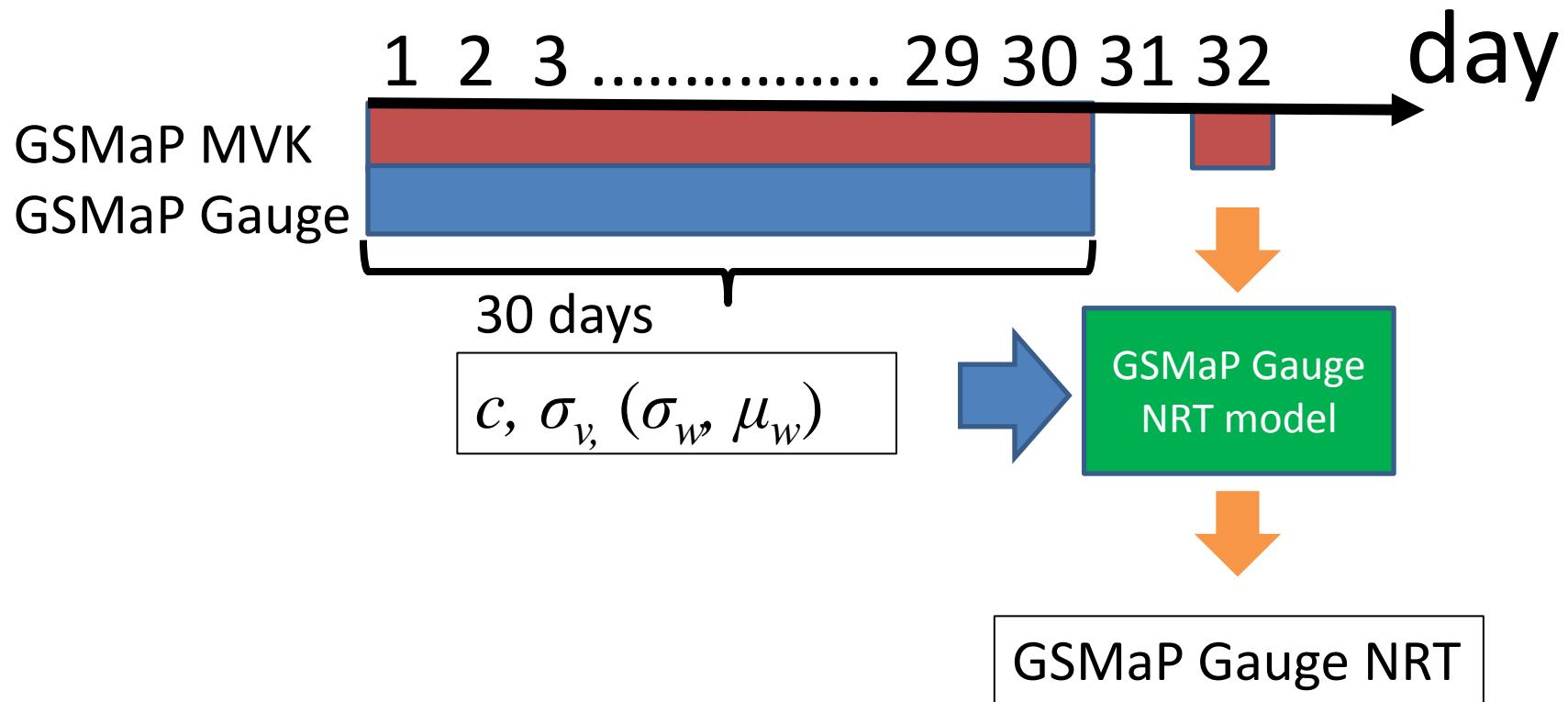
$\mathbf{y}$ : Observation



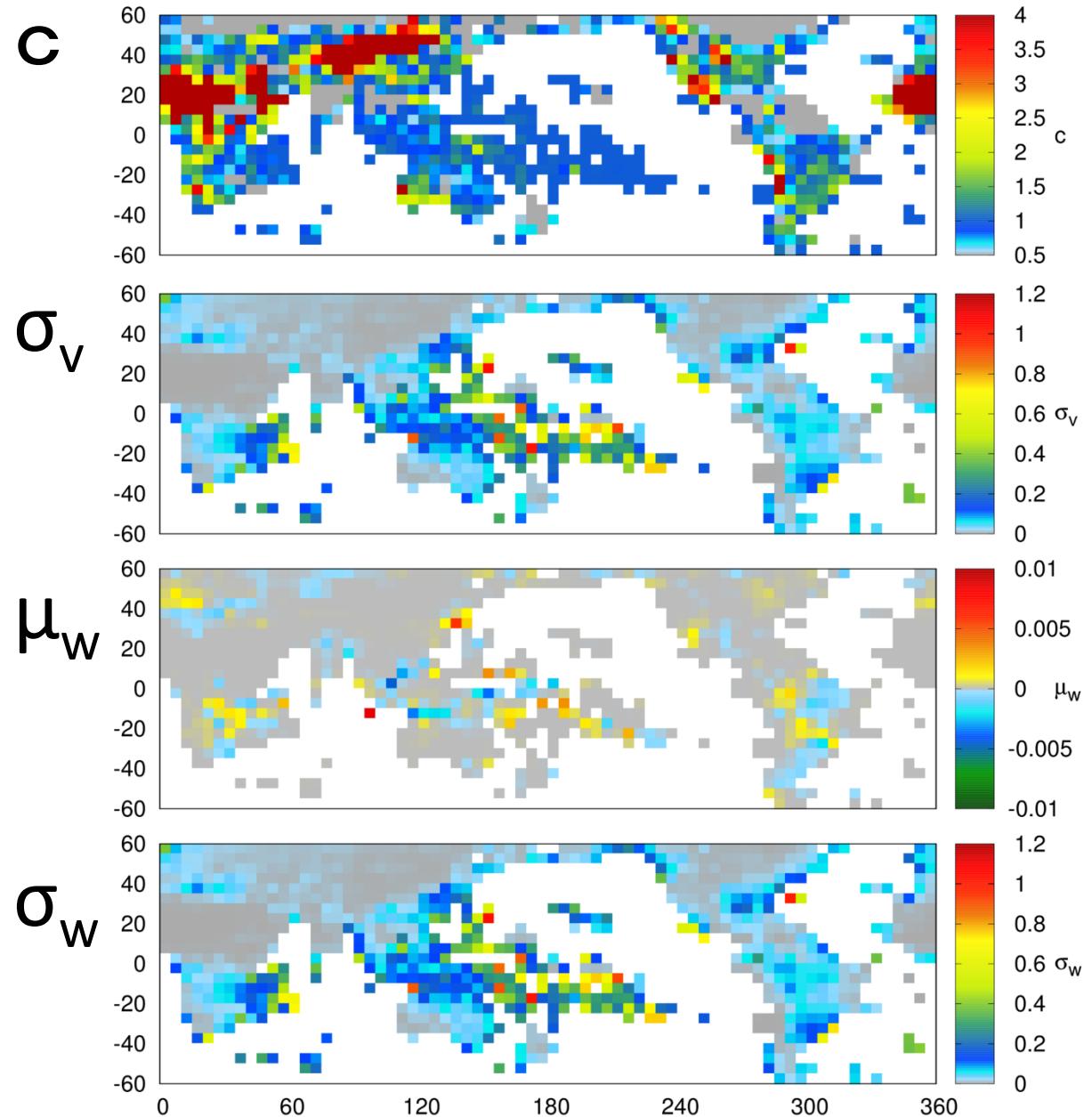
In order to have the model parameters (N, C) in GSMaP Gauge NRT,

$\mathbf{x}$ : GSMaP Gauge precipitation rate  
 $\mathbf{y}$ : GSMaP MVK precipitation rate

# How to get model parameters



# Distribution of Parameters of GSMAp Gauge NRT



GSMAp Gauge

$c : 0.7$

$\sigma_v : 1.0$

$\mu_w : 0$

$\sigma_w : 1.5$

# Zonal Mean

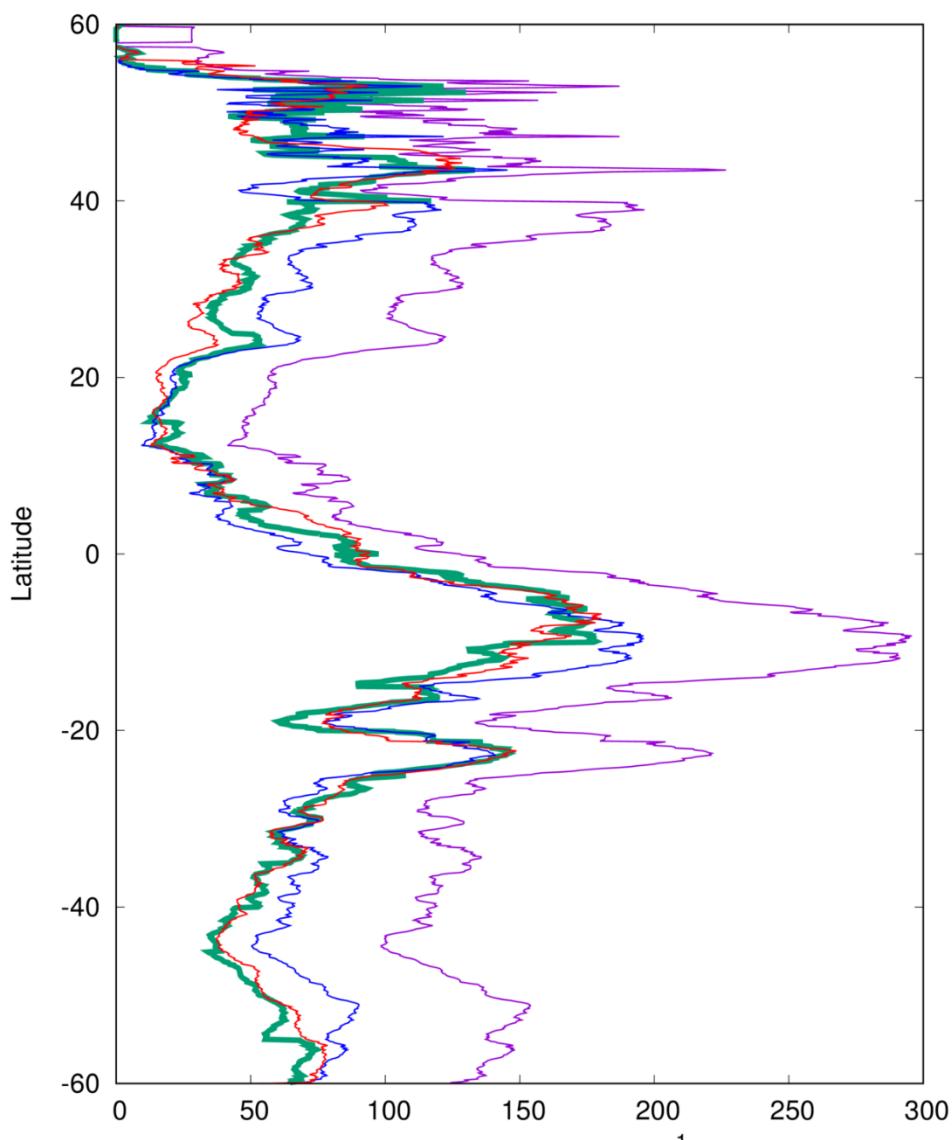
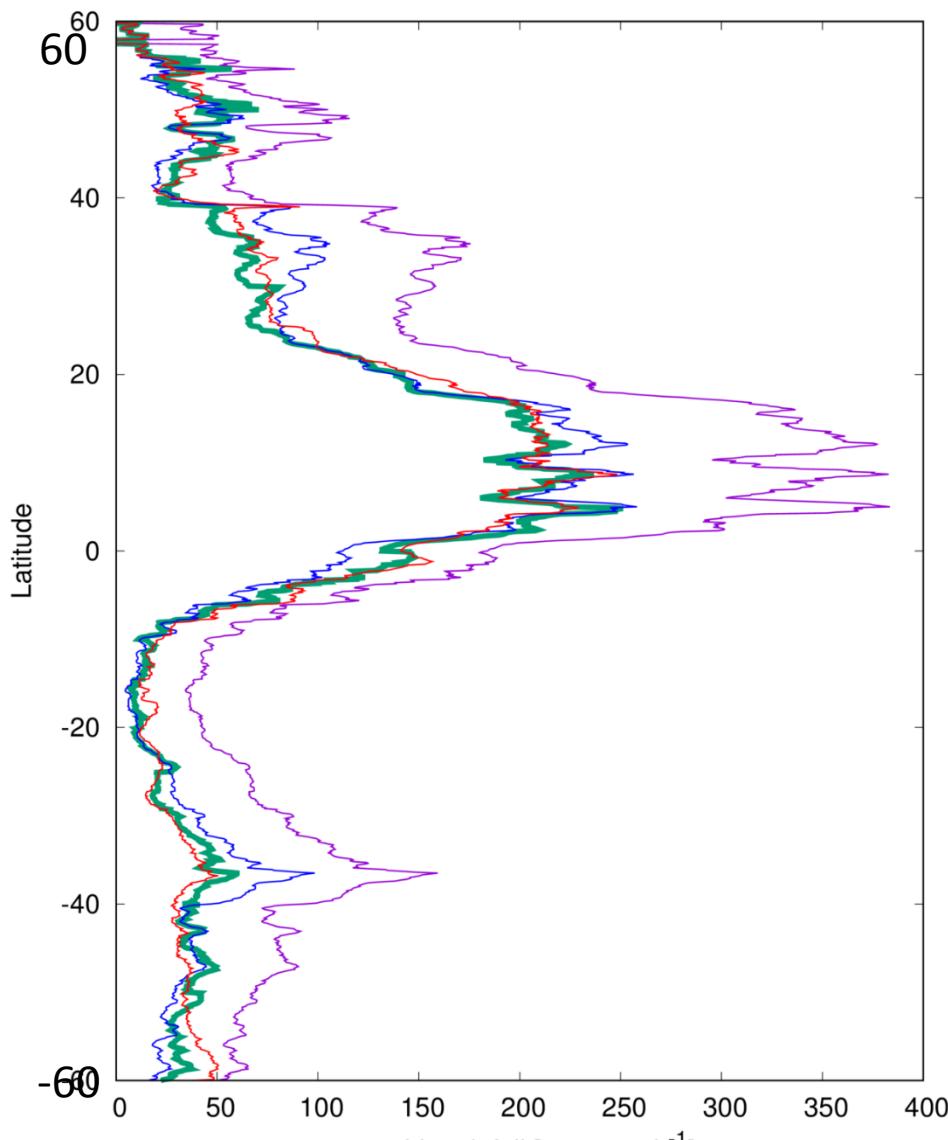
GSMaP Gauge NRT (Constant)  
GSMaP Gauge NRT ( $c, \sigma_v$ )  
GSMaP MVK  
GSMaP Gauge

much improved.

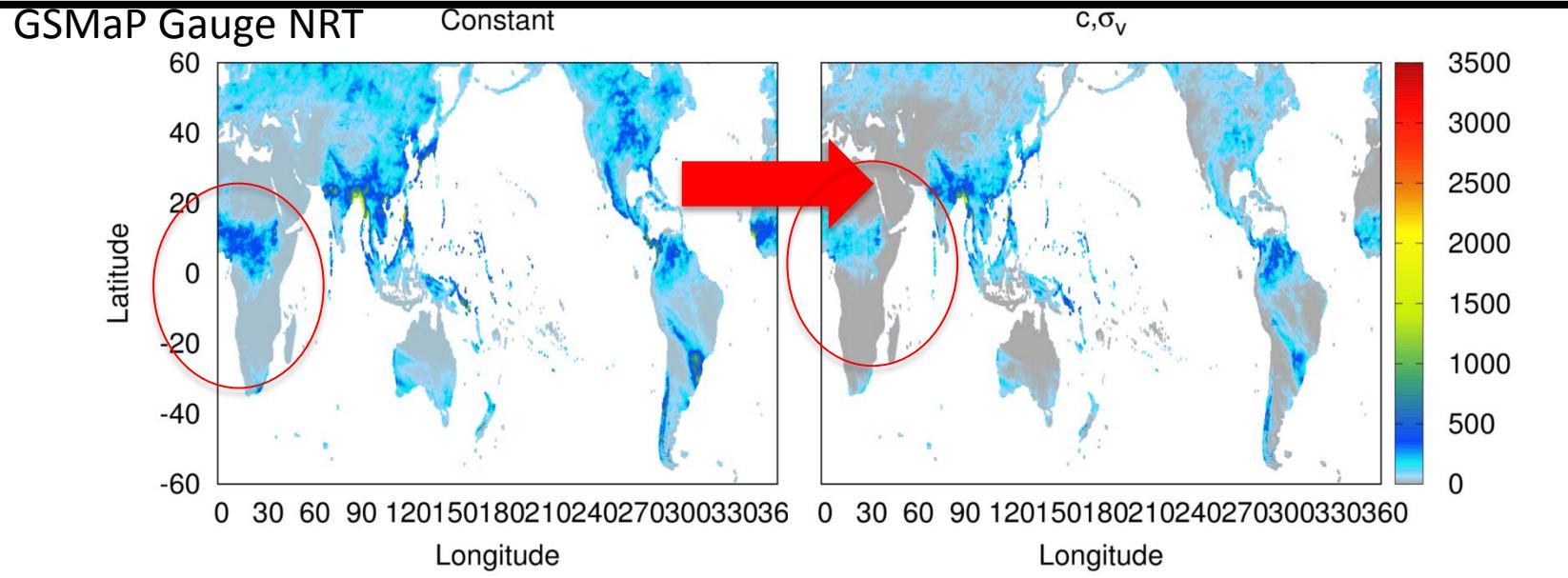
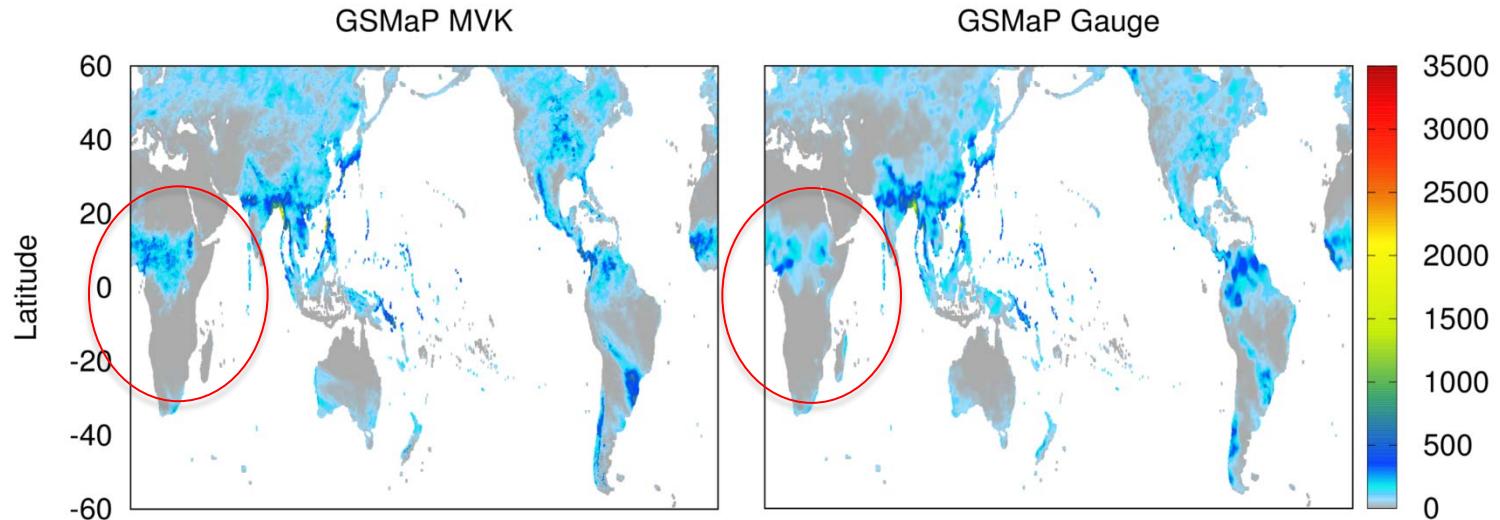
GSMaP Gauge NRT (Constant)  
GSMaP Gauge NRT ( $c, \sigma_v$ )  
GSMaP MVK  
GSMaP Gauge

201501

201507



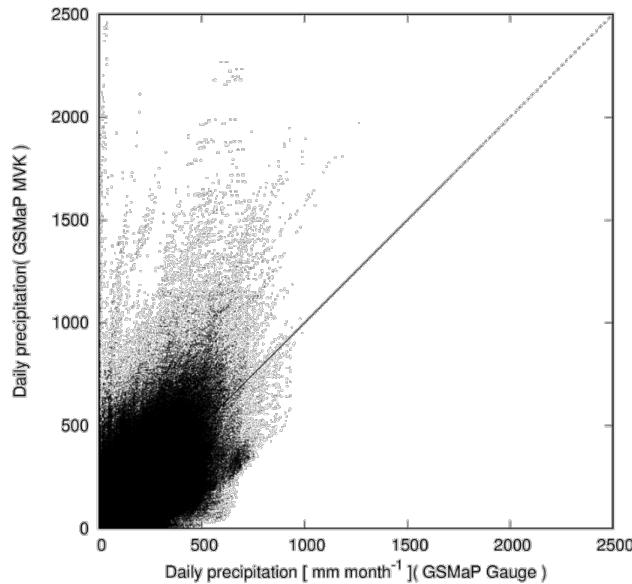
# Monthly rainfall distribution, June 2015



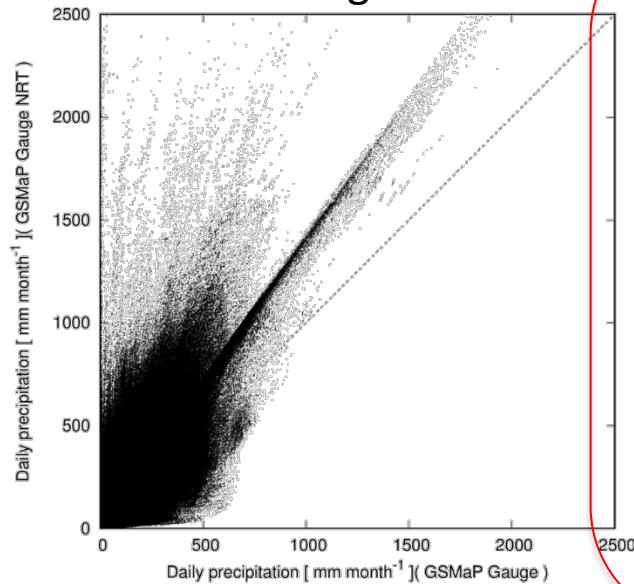
# Daily precipitation amount, January 2015

Vs GSMAp Gauge

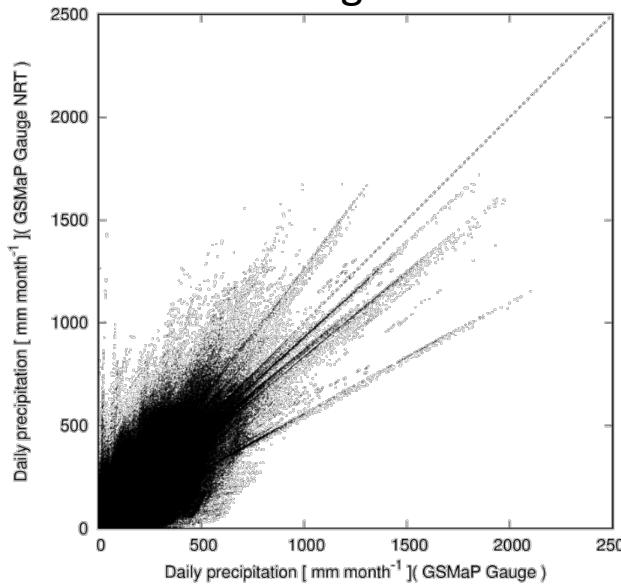
GSMAp MVK



GSMAp Gauge NRT



GSMAp Gauge NRT



	Correlation	RMSE	Coefficient of regression line
GSMAp MVK	0.844	40.3	0.894
Gauge NRT (Constant)	0.853	62.3	1.26
Gauge NRT ( $c, \sigma_v$ )	0.873	34.0	0.873

# Current GSMAp products

- GSMAp\_MWR
  - Microwave radiometer product
- GSMAp\_MVK
  - Global precipitation mapping from microwave and infrared radiometric data
- GSMAp\_Gauge
  - Gauge adjusted GSMAp\_MVK
- GSMAp\_NRT and GSMAp\_Now
  - Near real time version of the GSMAp\_MVK product
- GSMAp\_Gauge\_NRT
  - Near real time product of GSMAp\_Gauge

# Summary and Conclusions

- Methodology and algorithm structure of the high resolution GSMap was described.

# System Model of the GSMaP\_Gauge product

$$\left\{ \begin{array}{l} a_{n+1} = a_n + w_n \\ x_n = \alpha a_n + v_n \\ \sum_{n=1}^{24} a_n = W \end{array} \right.$$

a: Rain rate at each pixel

W: Rain rate from CPC global gauge data sets

n: Time, a: Rain rate from GSMAp\_Gauge

x: Rain rate from GSMAp\_MVK, v, w: Noise

$$w_n \approx N(\mu_w, \sigma_w^2) \quad v_n \approx N(\mu_v, \sigma_v^2)$$

# System Model of the GSMaP\_Gauge product

$$\left\{ \begin{array}{l} a_{n+1} = a_n + w_n \\ x_n = \alpha a_n + v_n \end{array} \right.$$

a: Rain rate at each pixel

W: Rain rate from CPC global gauge data sets

n: Time, a: Rain rate from GSMAp\_Gauge

x: Rain rate from GSMAp\_MVK, v, w: Noise

$$w_n \approx N(\mu_w, \sigma_w^2) \quad v_n \approx N(\mu_v, \sigma_v^2)$$

$$L(a) = -\ln \Pr(\mathbf{x}, \mathbf{a}) = -\ln \left[ \Pr(a_1) \prod_{m=1}^{24} \Pr(a_m | a_{m-1}) \prod_{n=1}^{24} \Pr(x_n | a_n) \right]$$

# System Model of the GSMaP\_Gauge product

$$\left\{ \begin{array}{l} a_{n+1} = a_n + w_n \\ x_n = \alpha a_n + v_n \\ \sum_{n=1}^{24} a_n = W \end{array} \right.$$

a: Rain rate at each pixel

W: Rain rate from CPC global gauge data sets

n: Time, a: Rain rate from GSMAp\_Gauge

x: Rain rate from GSMAp\_MVK, v, w: Noise

$$w_n \approx N(\mu_w, \sigma_w^2) \quad v_n \approx N(\mu_v, \sigma_v^2)$$

# Cost Function

Gauge term

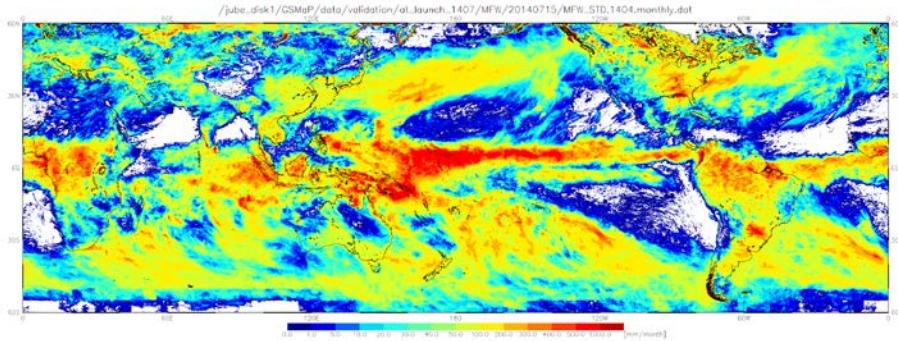
$$L(a) = -\ln \left\{ \Pr(\mathbf{x}, \mathbf{a}) \times e^{\frac{\lambda}{2} \left( \sum_{n=1}^N a_n - W \right)^2} \right\}$$

$$= -\ln \left[ \Pr(a_1) \prod_{m=1}^{24} \Pr(a_m | a_{m-1}) \prod_{n=1}^{24} \Pr(x_n | a_n) \times e^{\frac{\lambda}{2} \left( \sum_{n=1}^N a_n - W \right)^2} \right]$$

- In a word, based on the assumption that the GSMAp\_Gauge – CPC Gauge data (Gauge term) has the gaussian distribution, we maximize the probability density function of the GSMAp\_Gauge estimation multiplied by the Gauge term.
- The solution can be determined by calculating the  $dL/da = 0$  equation

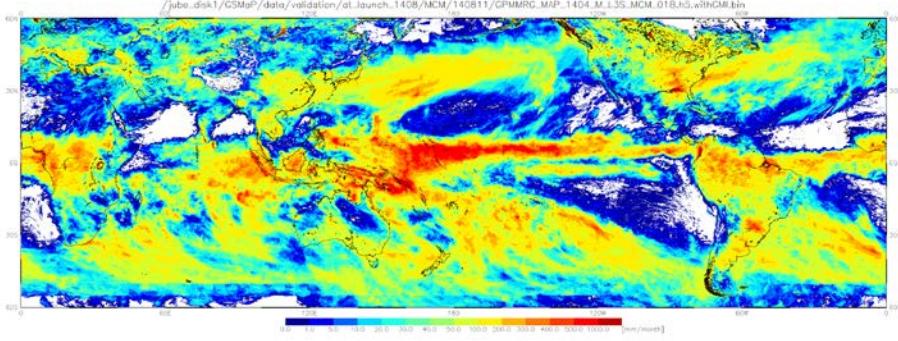
# Examples

**GSMaP\_NRT (FW)**

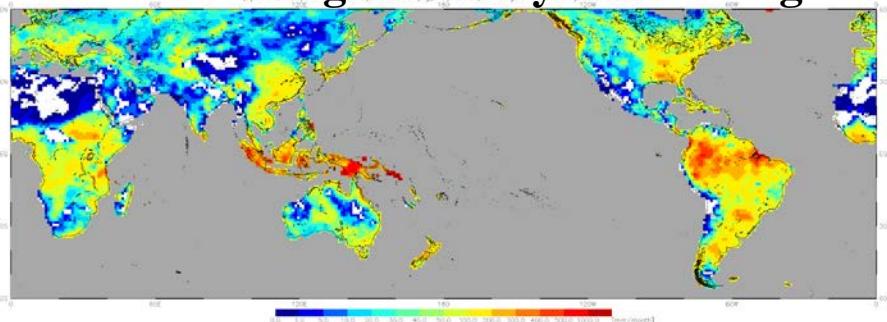


Monthly Mean in April 2014 (V03A)

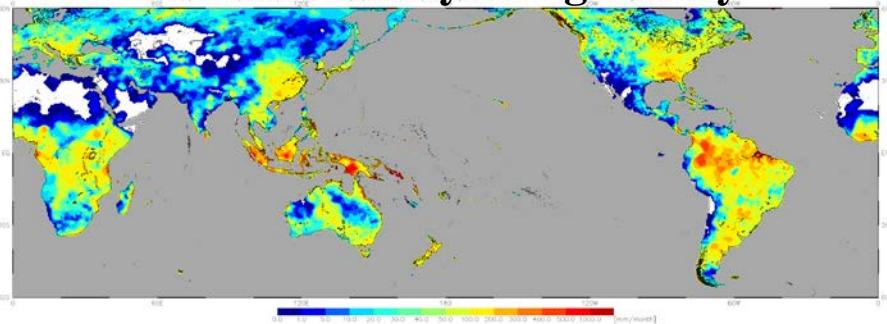
**GSMaP\_MVK**



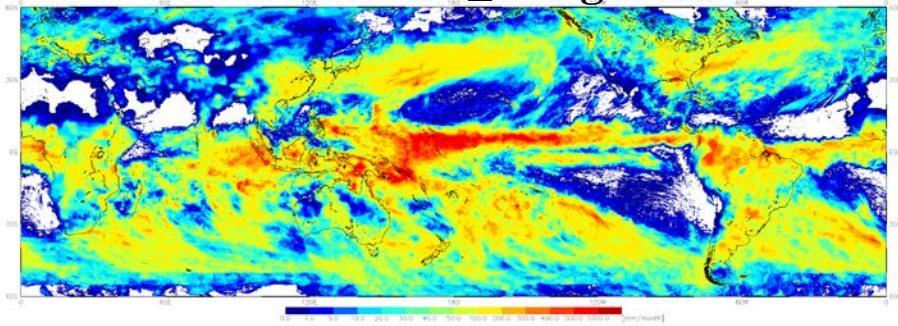
**GPCC Gauge Monthly Monitoring**



**NOAA/CPC Daily Gauge Analysis**



**GSMaP\_Gauge**

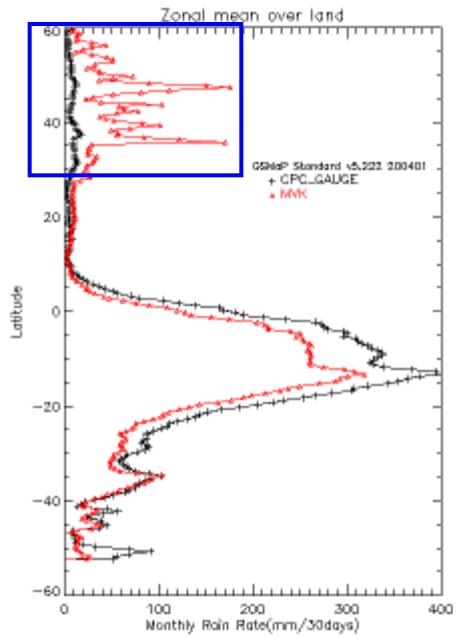


From JAXA

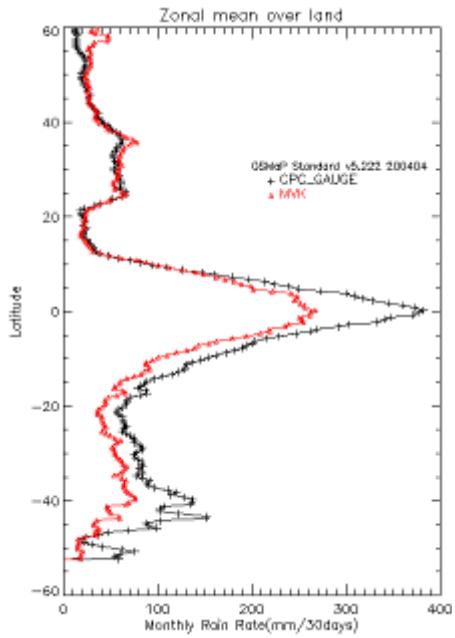
# Zonal mean analysis

## Comparison between MVK(red) and GSMAp\_Gauge(black)

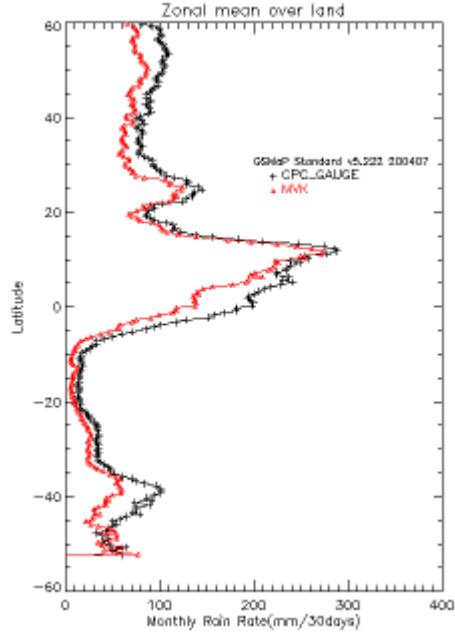
2004/01



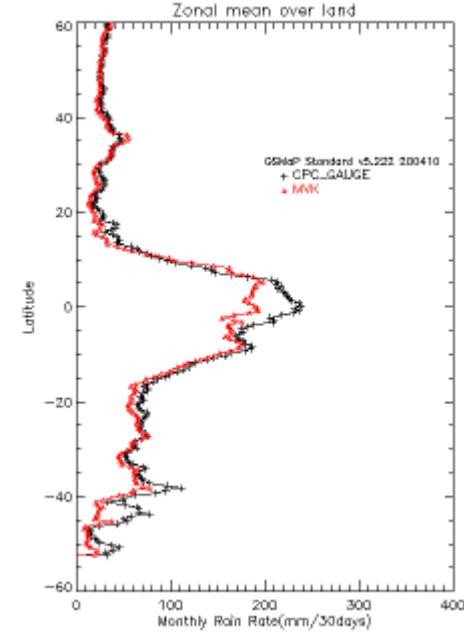
2004/04



2004/07

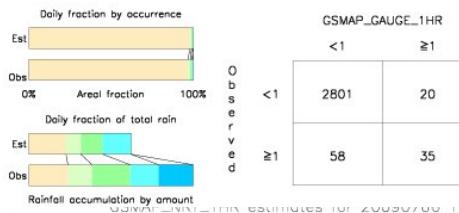
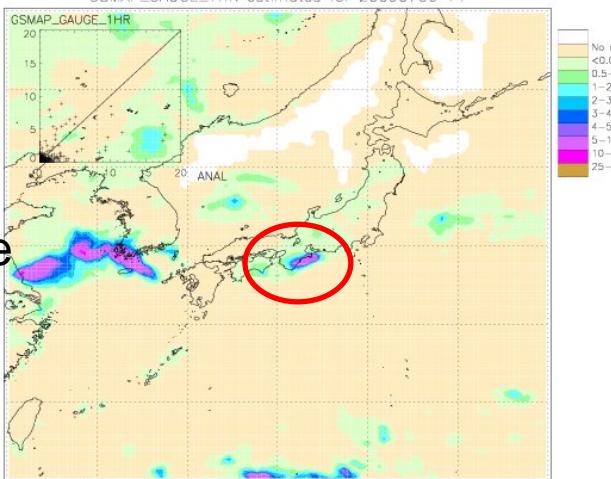


2004/10



Estimation from the GSMAp\_Gauge product tends to be  
larger than that from the GSMAp\_MVK

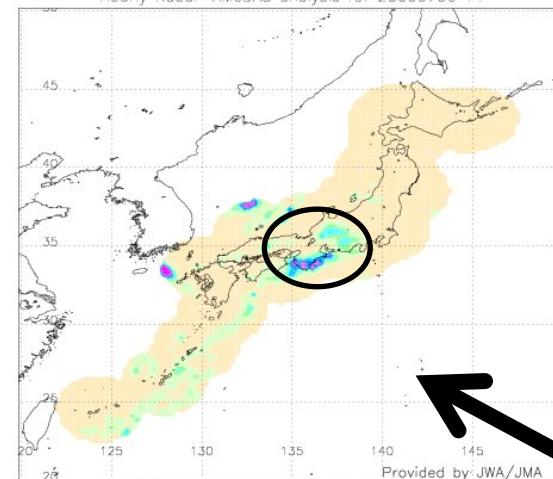
## GSMaP\_Gauge



Verification statistics for 20090706 14 n = 2914 Verif. grid=0.25° Units=mm/hr

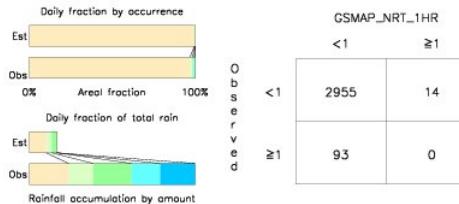
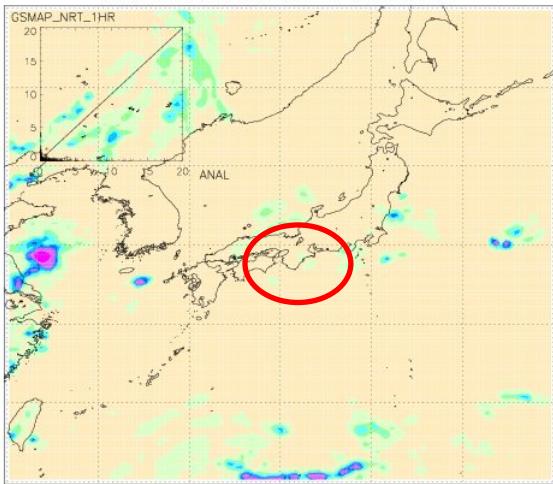
Analysed	GSMaP_GAUGE_1HR	
# gridpoints raining	93	55
Average rain	0.1	0.1
Conditional rain	4.5	4.7
Rain volume (mm·km <sup>2</sup> ·10 <sup>3</sup> )	0.3	0.2
Maximum rain	17.8	8.9

Mean abs error = 0.1  
RMS error = 0.7  
Correlation coeff = 0.599  
Frequency bias = 0.591  
Probability of detection = 0.376  
False alarm ratio = 0.364  
Hanssen & Kuipers score = 0.369  
Equitable threat score= 0.299



Radar Rain Gauge Network

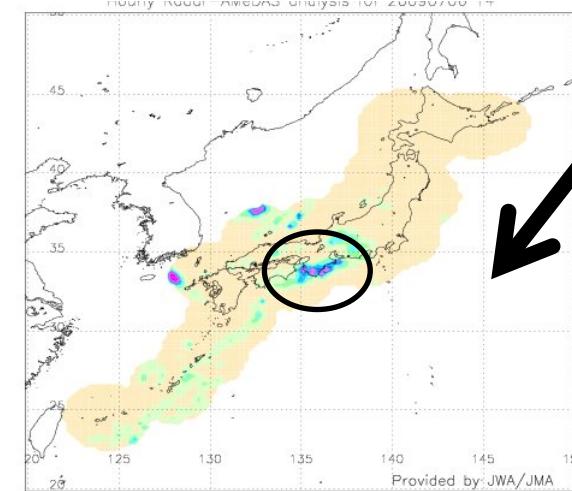
## GSMaP\_NRT



Verification statistics for 20090706 14 n = 3062 Verif. grid=0.25° Units=mm/hr

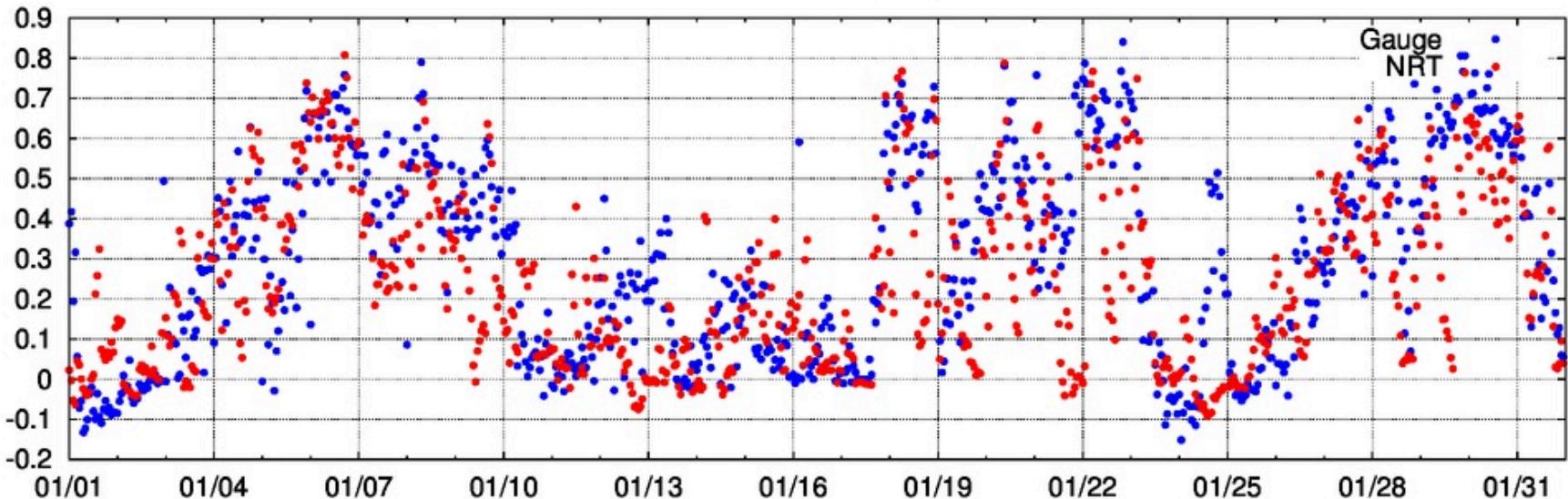
Analysed	GSMaP_NRT_1HR	
# gridpoints raining	93	14
Average rain	0.1	0.0
Conditional rain	4.5	4.9
Rain volume (mm·km <sup>2</sup> ·10 <sup>3</sup> )	0.3	0.0
Maximum rain	17.8	4.0

Mean abs error = 0.2  
RMS error = 0.9  
Correlation coeff = -0.003  
Frequency bias = 0.151  
Probability of detection = 0.000  
False alarm ratio = 1.000  
Hanssen & Kuipers score = -0.005  
Equitable threat score= -0.004

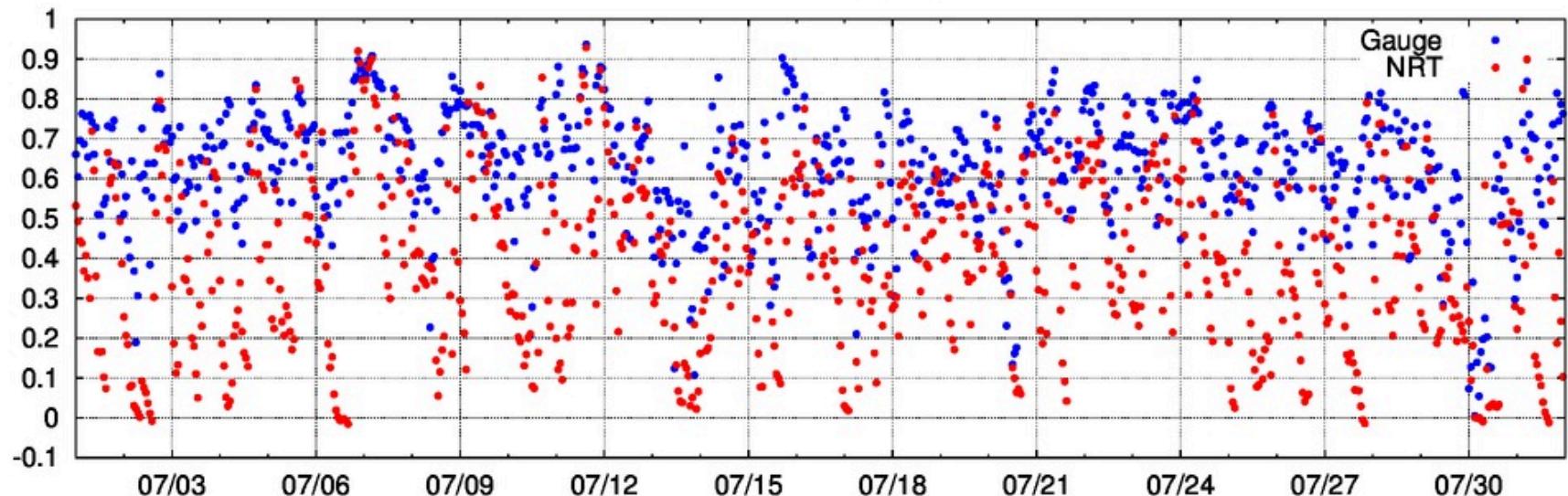


# Correlation

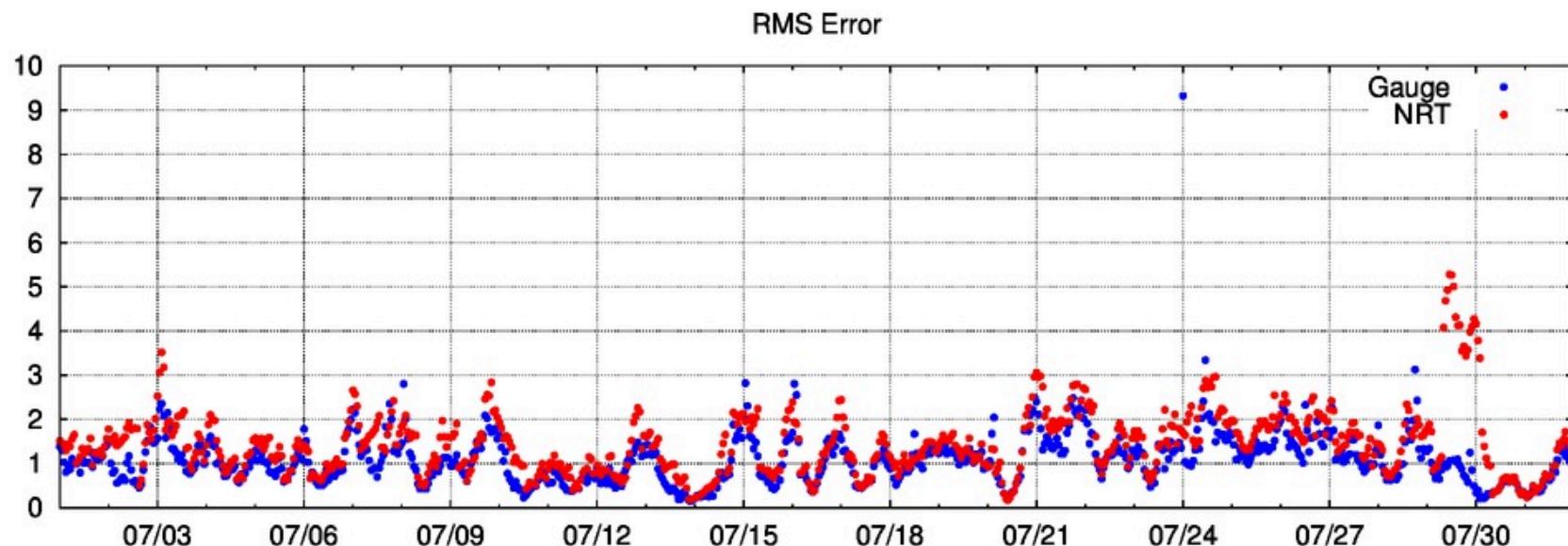
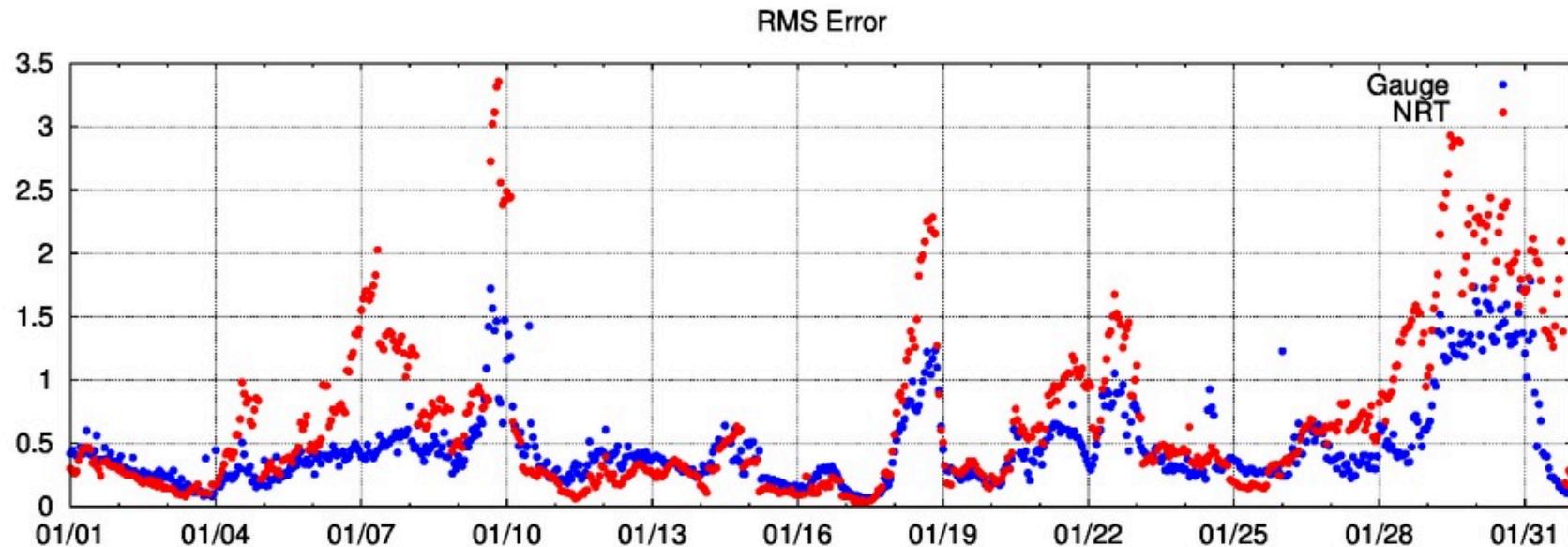
Correlation coefficient



Correlation coefficient



# RMS Error



# モデルへの期待

- 大気移動ベクトルが、過去から将来まで多層にわたって求まる
  - GSMAp\_MVK, GSMAp\_Now, GSMAp\_Futureまで使用することが可能となる
  - より効果的な高度のベクトルの使用が可能となる
- モデルによる地表面降水強度の使用
  - 高緯度では観測より精度が良い？
  - 一方、熱帯では？？？
  - MVKでのカルマンフィルタの観測方程式のyにモデルからの地表面降水量を追加することが可能

# モデルとの複合案

- モデルからの移動ベクトルを使用
  - 多層で使えるか。
  - GSMAp\_MVK
  - GSMAp\_Nowにも使用できる
  - さらに, GSMAp\_Futureも作れそう
- MVKでのカルマンフィルタ
  - 観測方程式のyにモデルからの地表面降水量を追加

# State and observation equation used in Kalman filter

$$x_{k+1} = x_k + \sigma_w \quad (\textit{State Equation})$$

$$\mathbf{y}_k = \mathbf{H}x_k + \mathbf{s}_v \quad \textit{Observation Equation}$$

$$\rightarrow \begin{bmatrix} y_k^0 \\ y_k^1 \end{bmatrix} = \begin{bmatrix} H_0 \\ H_1 \end{bmatrix} x_k + \begin{bmatrix} \sigma_v^0 \\ \sigma_v^1 \end{bmatrix}$$

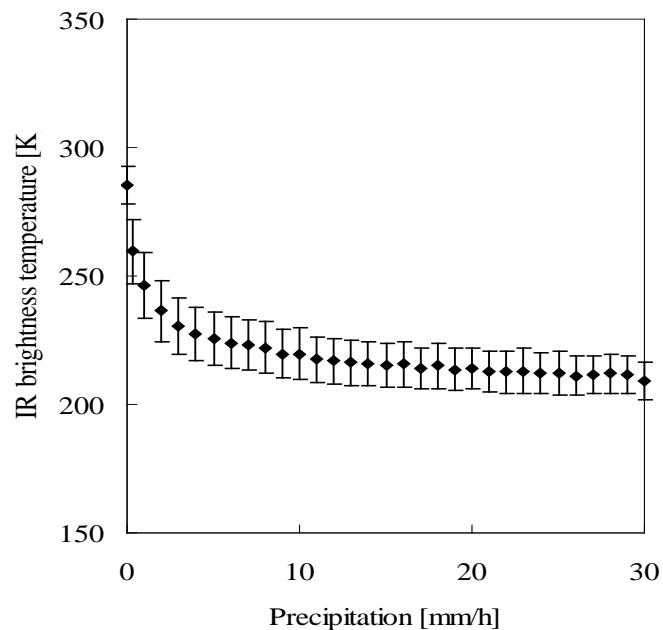
$x_k$  : Rain rate at time k

$y_k$  : Infrared Tb, R from model

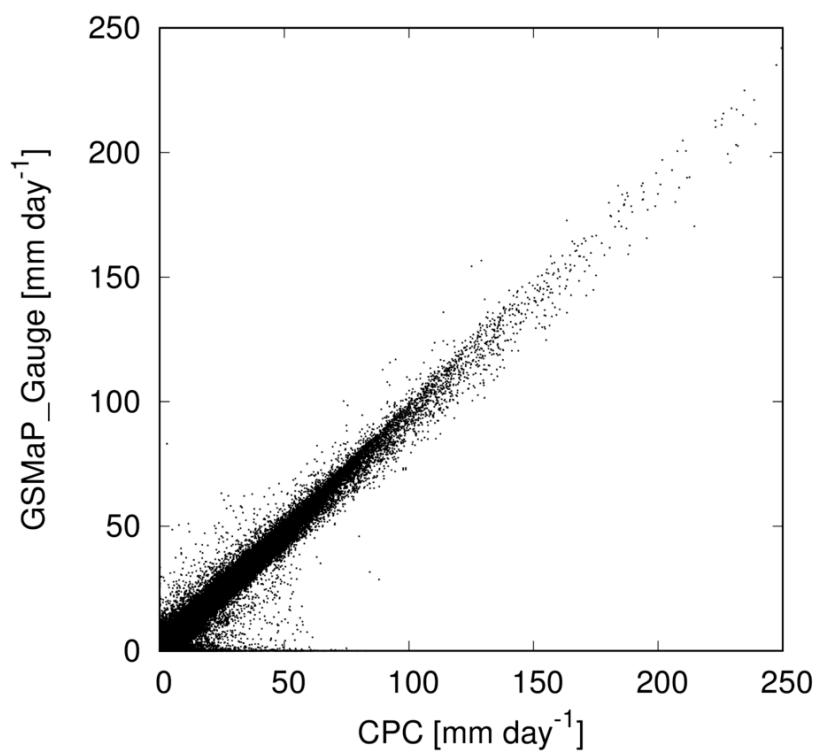
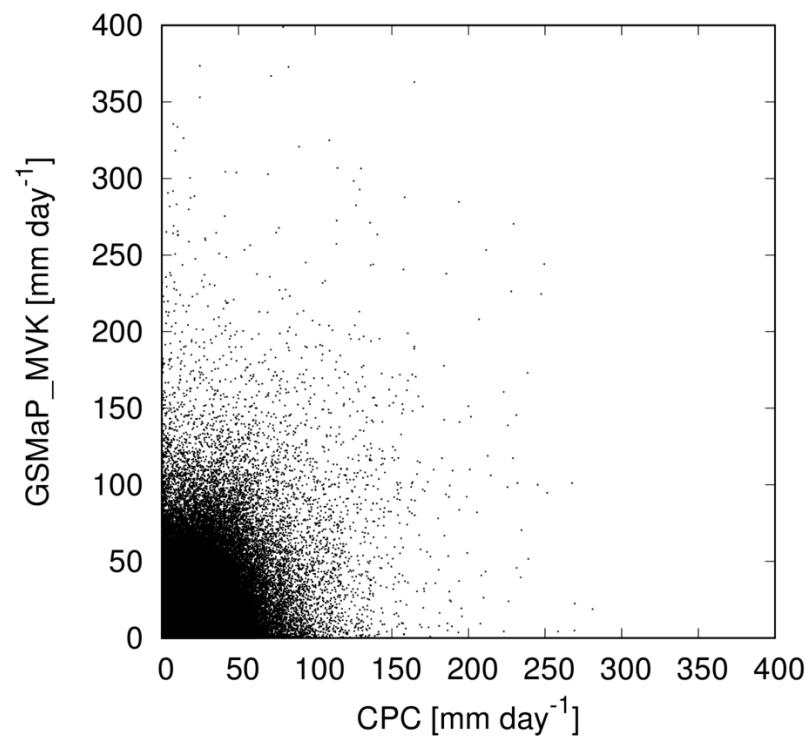
$x_{k+1}$ : Rain rate at time k+1

$w$  : System noise

$v$  : Observation noise



# CPC vs GSMAp\_MVK and GSMAp\_Gauge



# State and observation equation used in Kalman filter

$$x_{k+1} = x_k + \sigma_w \quad (\textit{State Equation})$$

$$y_k = Hx_k + \sigma_v \quad (\textit{Observation Equation})$$

$x_k$  : Rain rate at time k

$y_k$  : Infrared Tb

$x_{k+1}$ : Rain rate at time k+1

$w$  : System noise

$v$  : Observation noise

