



GOSAT データ利用
ワークショップ
虎ノ門パストラル
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GOSAT TANSO-FTS TIR (Band 4) data analysis method

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GOSAT sensors

【Satellite】

- Mass, power : Approx. **1650kg**, **3.3kw** - EOL
- Designed life span : **5 years**
- Orbit : Altitude **666km** , Orbit Inclination **98 deg.**
Sun-Synchronous Sub-Recurrent Orbit

【TANSO-FTS】

- Spectral range
 - Band 1 : **$0.75 \mu\text{m} \sim 0.78 \mu\text{m}$** ($\text{O}_2\text{-A}$)
 - Band 2 : **$1.56 \mu\text{m} \sim 1.92 \mu\text{m}$** (CO_2 column)
 - Band 3 : **$1.92 \mu\text{m} \sim 2.08 \mu\text{m}$** (CO_2 , CH_4 , cirrus cloud)
 - Band 4 : $5.5 \mu\text{m} \sim 14.3 \mu\text{m}$** (CO_2 vertical profile, CH_4)
- Spectral resolution : **0.2cm^{-1}**
- SNR : **> 300**

CO₂ retrieval from spectrum data measured from space

European group led by Dr. Chedin

4A+TIGR+3R, MSU → 3I, N.N; AIRS+AMSU.

HIRS: Chedin et al., 2002a, 2002b, 2003a, 2003b,

AIRS: Crevoisier et al., 2004, Chevallier et al., 2005, Engelen et al., 2004, 2005

and

AIRS group led by Dr. Chahine

AMSU, 4D-VAR

Chahine et al., 2008, 2005(VPD), Maddy et al. (2008), Aumann et al., 2005, Engelen et al., 2004

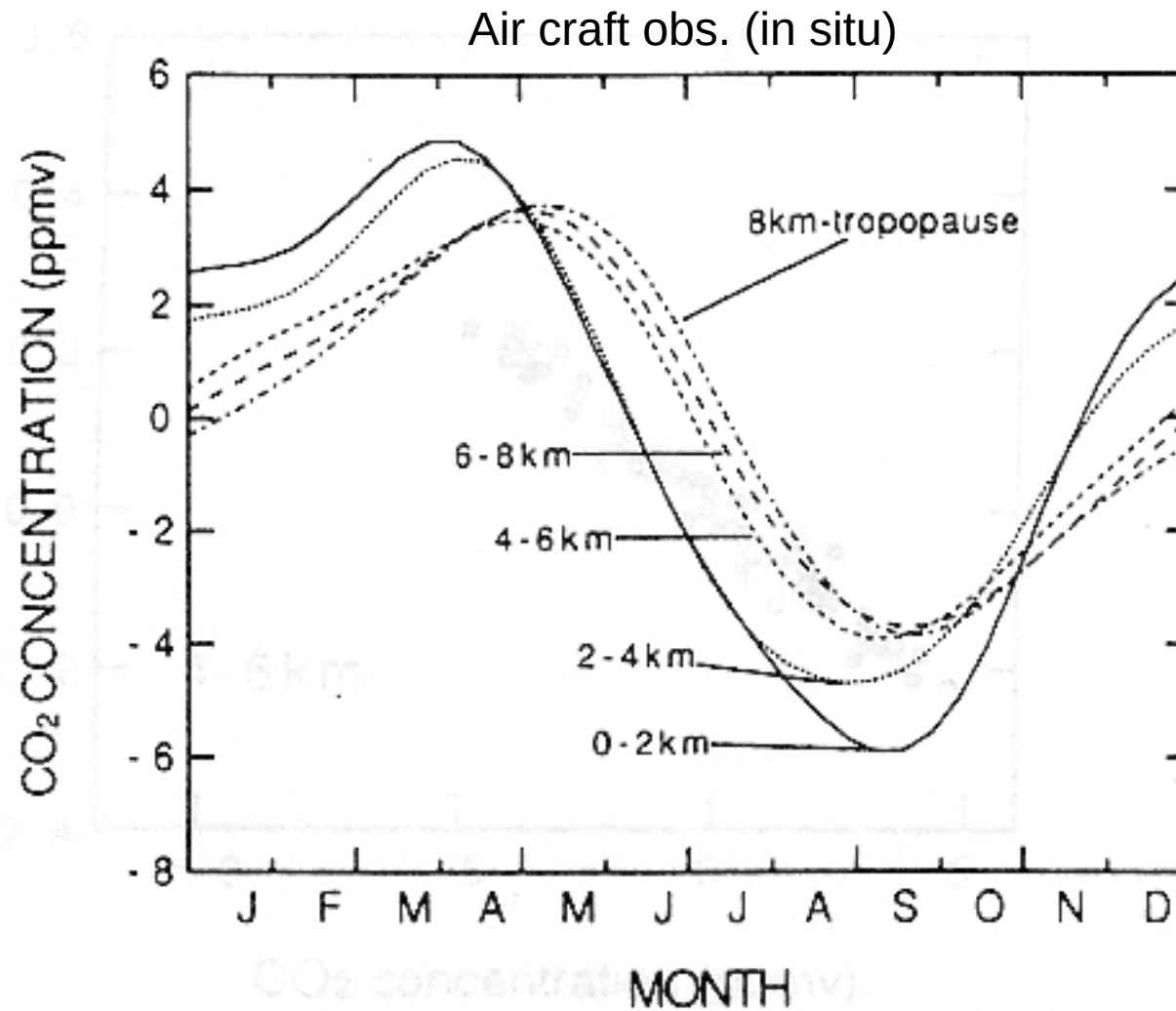
have successfully obtained global maps of CO₂ in the upper troposphere from the thermal infrared spectrum.

Next target :

Information on vertical structure of CO₂,
particularly on CO₂ concentration in PBL

→ Synergetic usage of GOSAT FTS bands (SWIR, TIR)
is effective ··· benefit of GOSAT sensors !

inter-annual variation of CO₂ at various height



(Nakazawa et al., 1996)

Calculation of Jacobian

1. Optimization of layer thickness of "retrieval grid"

In order to obtain uniform sensitivity and retrieval error, layer thickness have to be optimized

- Radiative transfer cal. at "full grid" (finest grid)
- Determining layer thickness based on "Area value" of averaging kernel
- "Linear-mapping" : "retrieval grid" $\longleftrightarrow$ "full grid"

2. Selecting retrieval channels

In order to reduce the effects of temperature estimation error, spectral channel that are used for retrieval have to be selected

- Channel selection based on "Shannon information contents"

Optimal estimation retrieval

MAP (Maximum *A Posteriori* retrieval)

- Retrieval method based on Bayesian theory.
- Total retrieval error can be estimated .

- The $(i+1)$ -th concentration, X_{i+1} , is

$$X_{i+1} = \underbrace{X_a}_{\text{a priori}} + (\underbrace{K_i^T \underbrace{S_e^{-1}}_{\text{measurement error covariance}} K_i}_{\text{a priori covariance}} + \underbrace{S_a^{-1}}_{\text{forward spectra}})^{-1} \underbrace{K_i^T}_{\text{measurement spectra}} \underbrace{S_e^{-1}}_{\text{measurement spectra}} (\underbrace{Y}_{\text{forward spectra}} - F(X_i)) + \underbrace{K_i}_{\text{Jacobian}} (x_i - x_a)$$

- Total retrieval error is

$$\hat{S} = (S_a^{-1} + K^T S_e^{-1} K)^{-1} \quad [\text{e.g., Rodgers, 2000}]$$

“full grid” ↔ “reduced grid”

- Linear-mapping between state (x) and retrieval (z) vectors is

$$\begin{aligned}x &= Wz & x : \text{state vector} \\ z &= W^* x & z : \text{retrieval vector}\end{aligned}$$

- The $(i+1)$ -th concentration, X_{i+1} , is

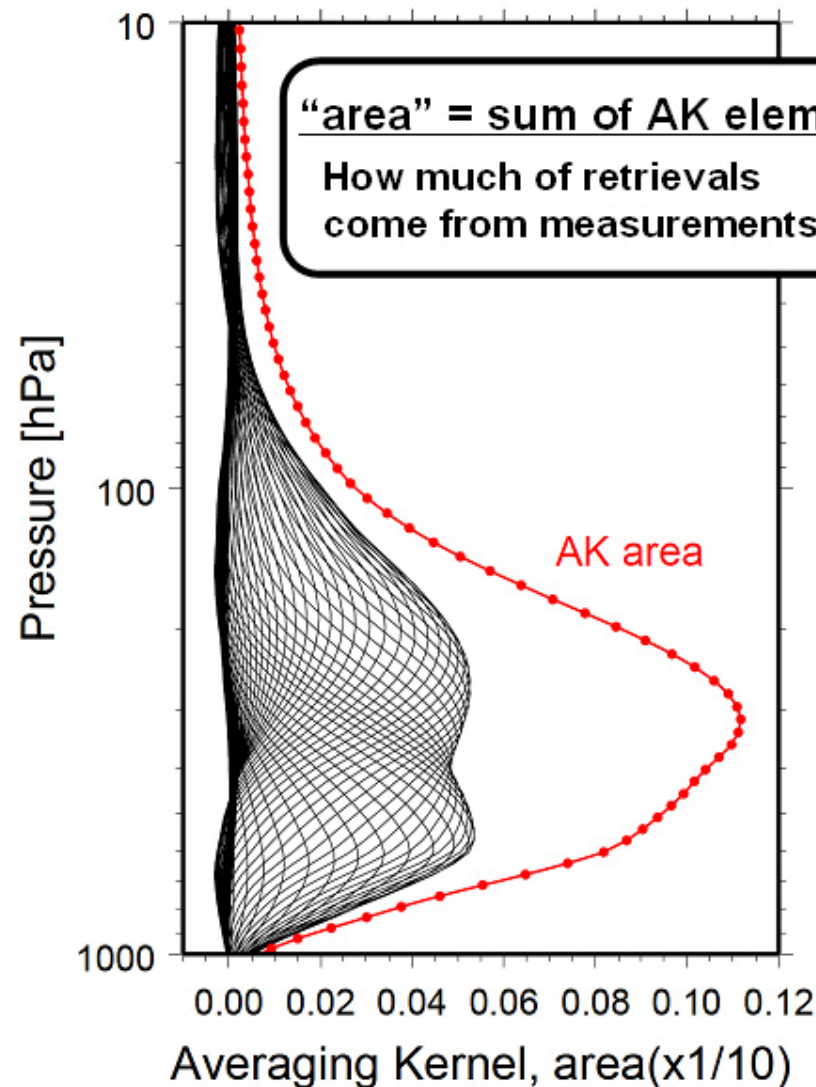
$$X_{i+1}$$

$$= W^* X_a + (W^T K_i^T S_e^{-1} K_i W + (W^* S W^{*T})^{-1})^{-1} W^T K_i^T S_e^{-1} (Y - F(X_i) + K_i W (W^* x_i - W^* x_a))$$

- Total retrieval error is

$$\hat{S} = (W^T K^T S_e^{-1} K W + (W^* S_a W^{*T})^{-1})^{-1}$$

“full grid” ↔ “reduced grid”



“full grid” (state vector)

radiative transfer &
jacobian calculations

← Atmosphere between 1100 and
0.1 hPa divided into 110 layers.

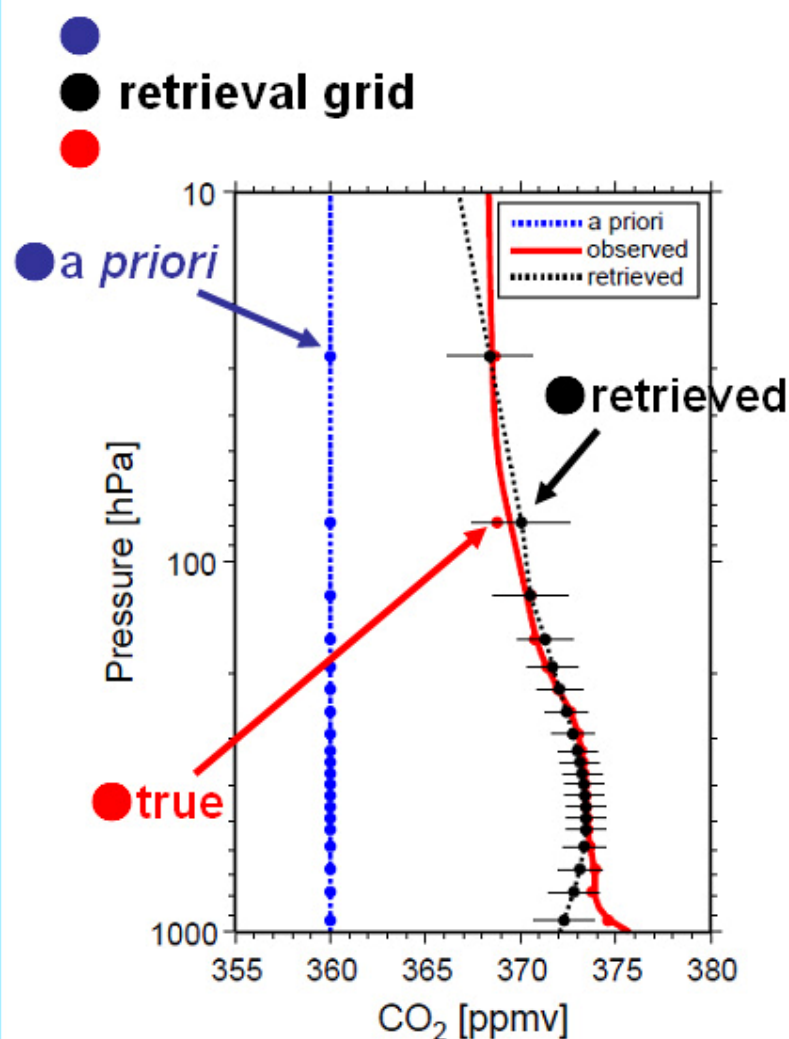
layers merged until
sum of AK areas > 1

“reduced grid” (retrieval vector)

CO₂ retrieval

← Coarser grid.
Grid inferred from AK area.

Retrieval without T uncertainties



simulations

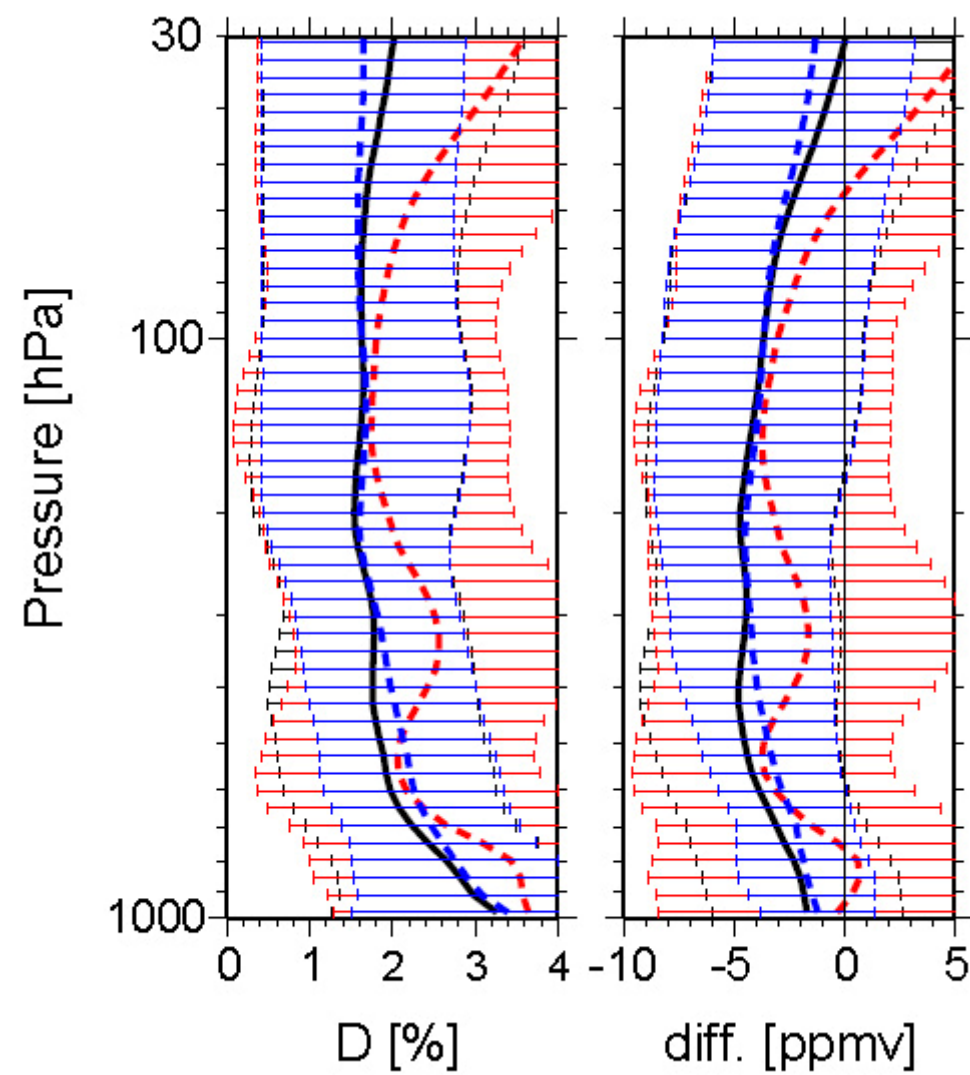
30 pseudo-spectra with random noise of GOSAT/TANSO-FTS were computed for each season and latitude.



Retrieval was performed,

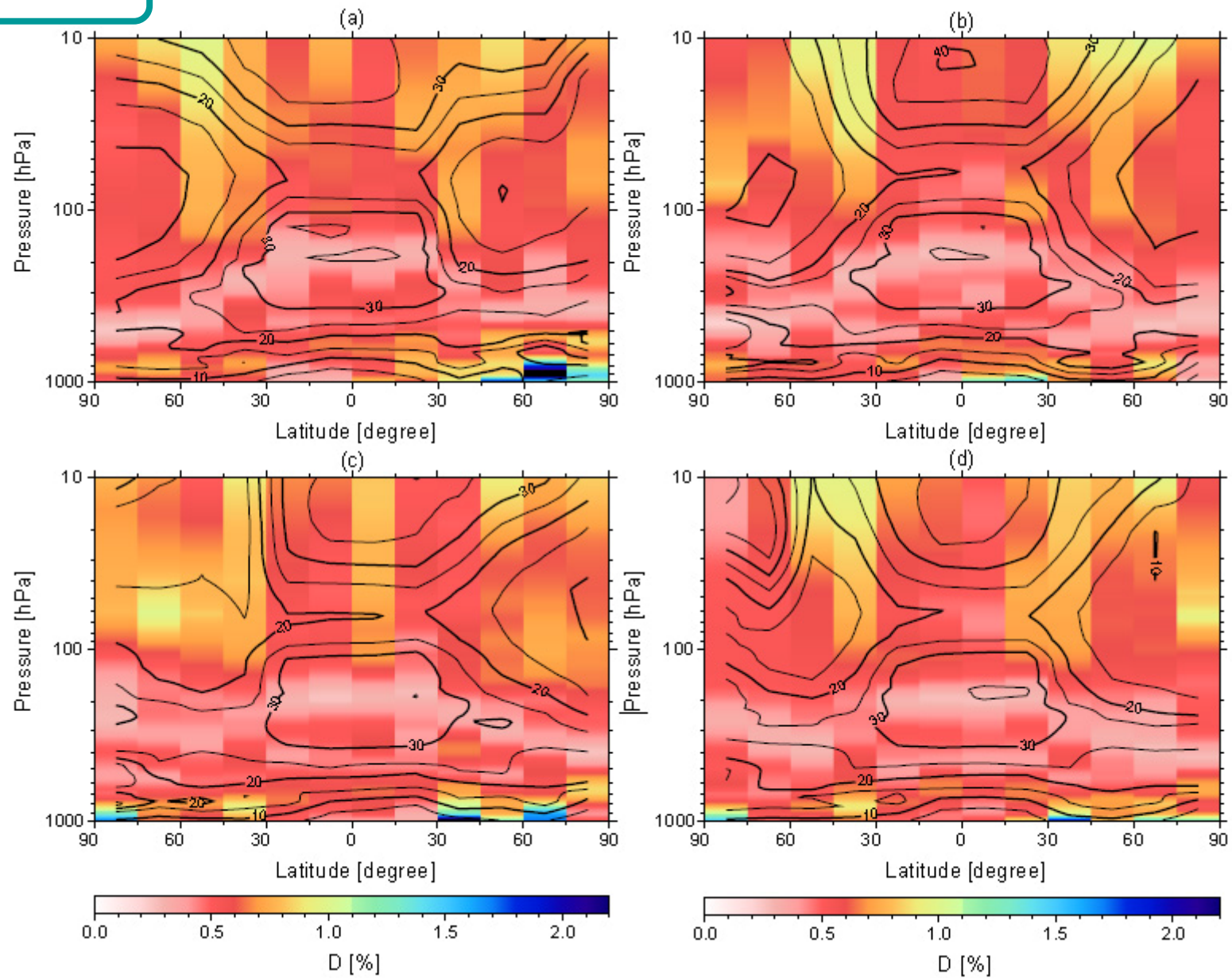
- S_a (error covariance of *a priori*)
for diagonal elements,
 - 1) $\pm 1\%$ (diagonal)
 - 2) $1-\sigma$ of NICAM CO₂ from *a priori*for non-diagonal elements,
correlation between layers decreasing to $1/e$ at 30 hPa apart from each other

Error analysis



(by Naoko Saitoh)

no T errors

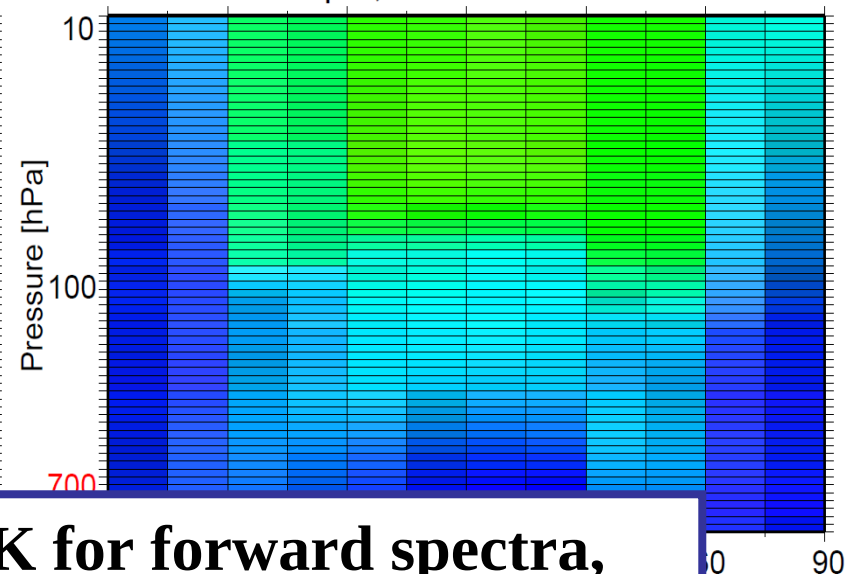
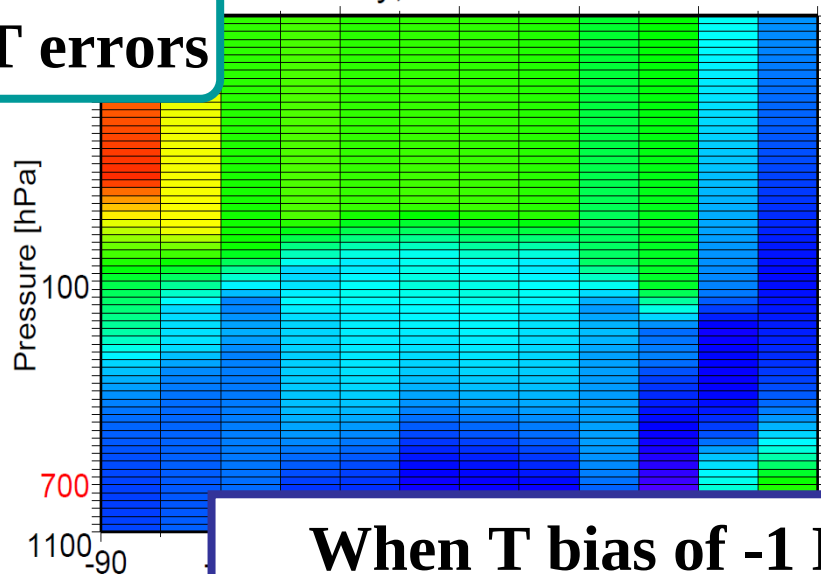


(by Naoko Saitoh)

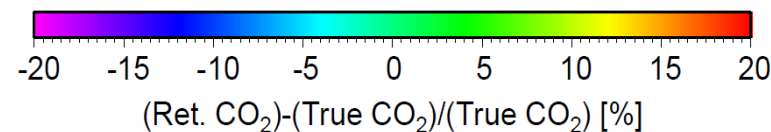
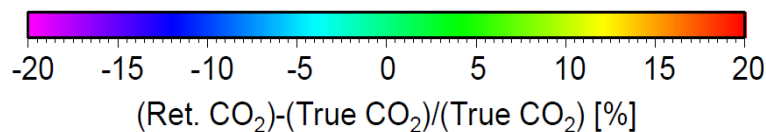
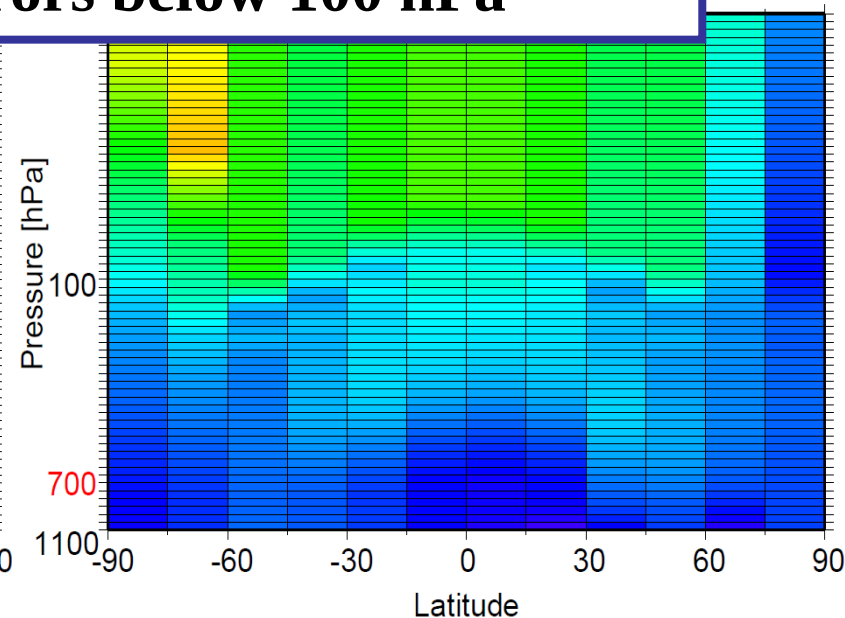
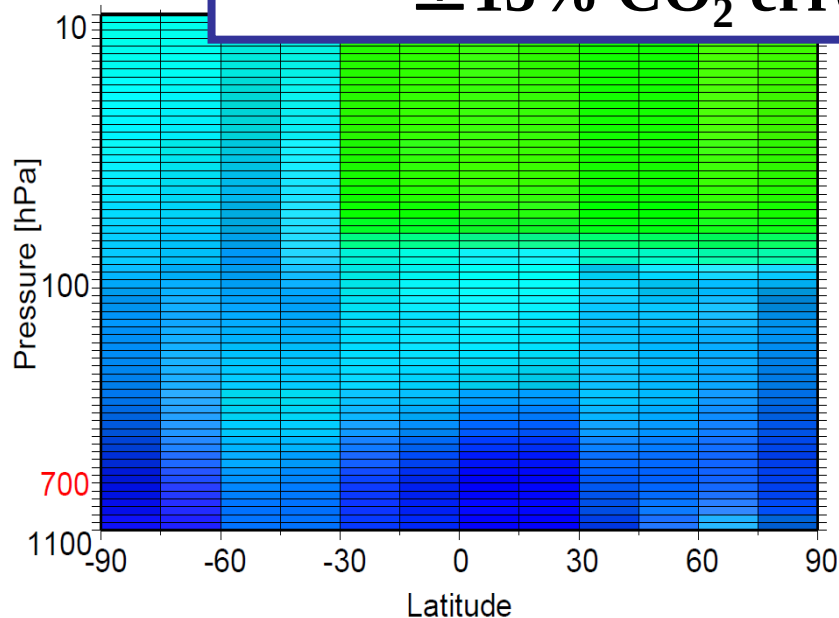
with T errors

January, Lon. E105-120

April, Lon. E105-120



**When T bias of -1 K for forward spectra,
 $\pm 15\%$ CO₂ errors below 100 hPa**



Retrieval channel selection

Appropriate retrieval channels should be selected to reduce the effect of T uncertainties on retrieved CO₂ concentrations.

- According to Shannon's information theory [Shannon and Weaver, 1949], information contents of CO₂ and temperature,

$$I = \frac{1}{2} \log_2 |S_a| - \frac{1}{2} \log_2 |\hat{S}| \quad I = \frac{1}{2} \log_2 |S_{\text{temp}}| - \frac{1}{2} \log_2 |\hat{S}|$$

[e.g., Rodgers 1996; Lerner et al., 2002]

- Temperature errors regarded as a measurement noise,

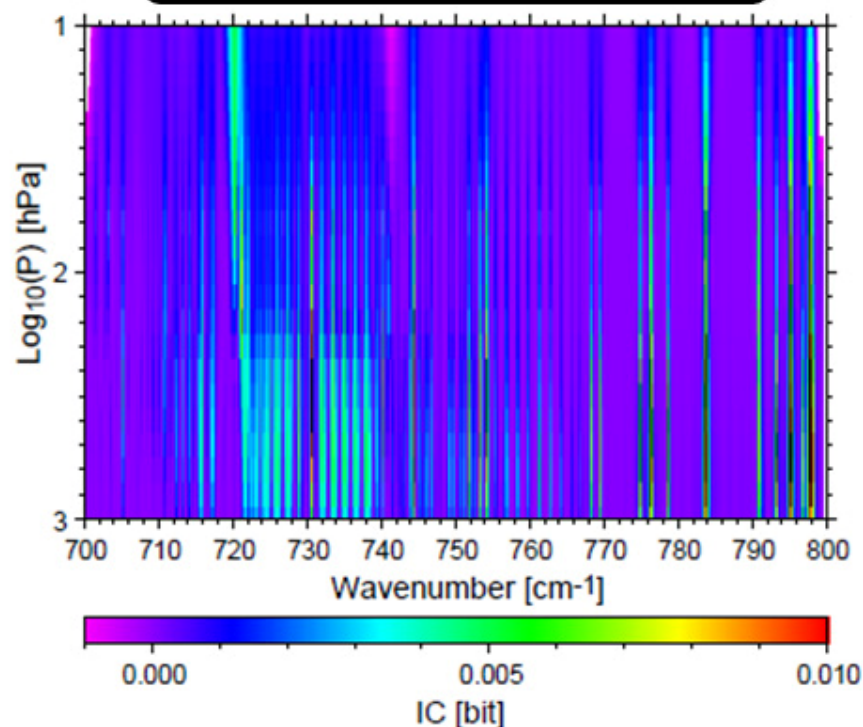
$$S_e = S_e + \underbrace{K_{\text{temp}} S_{\text{temp}} K_{\text{temp}}^T}_{\text{T random errors}}$$

T random errors

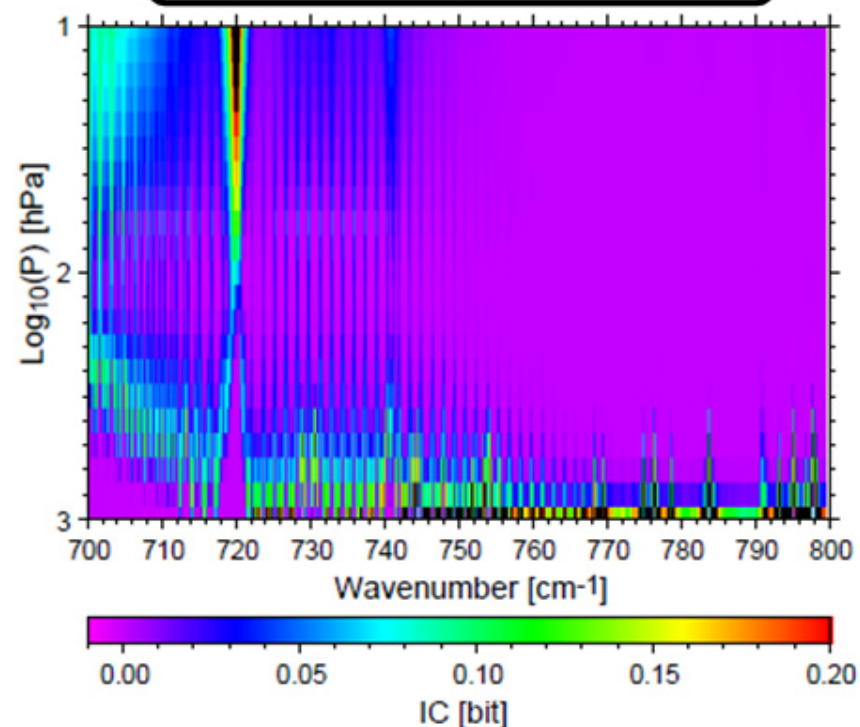
[Rodgers 2000; Ota, 2006]

Channel selection based on IC

CO₂ information content



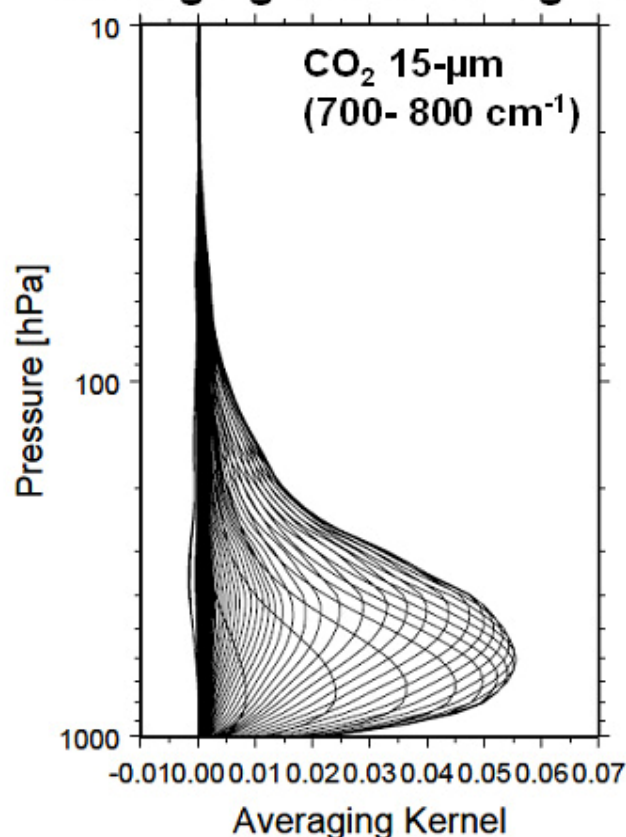
T information content



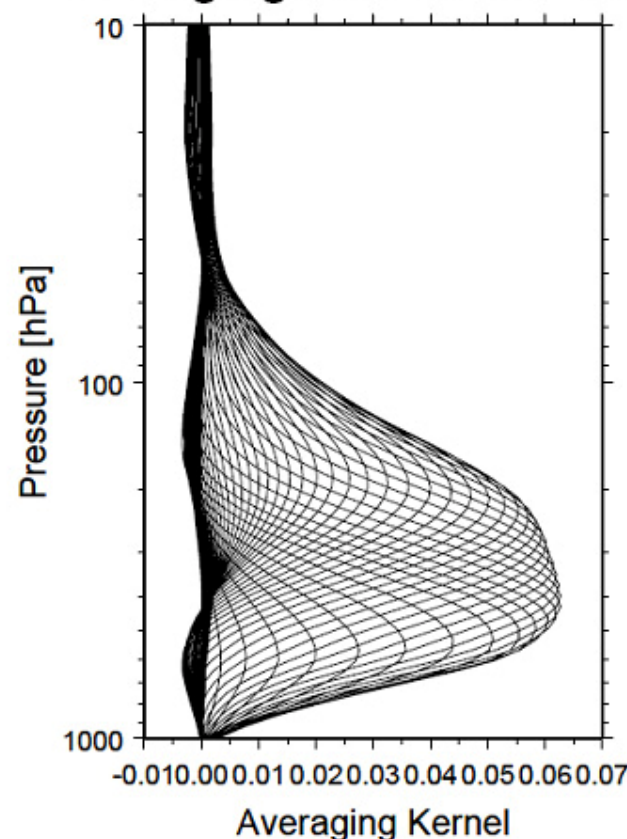
Channels sensitive to CO₂ are different from those to T.

Retrieval grid depending on latitude

averaging kernel at high lat.

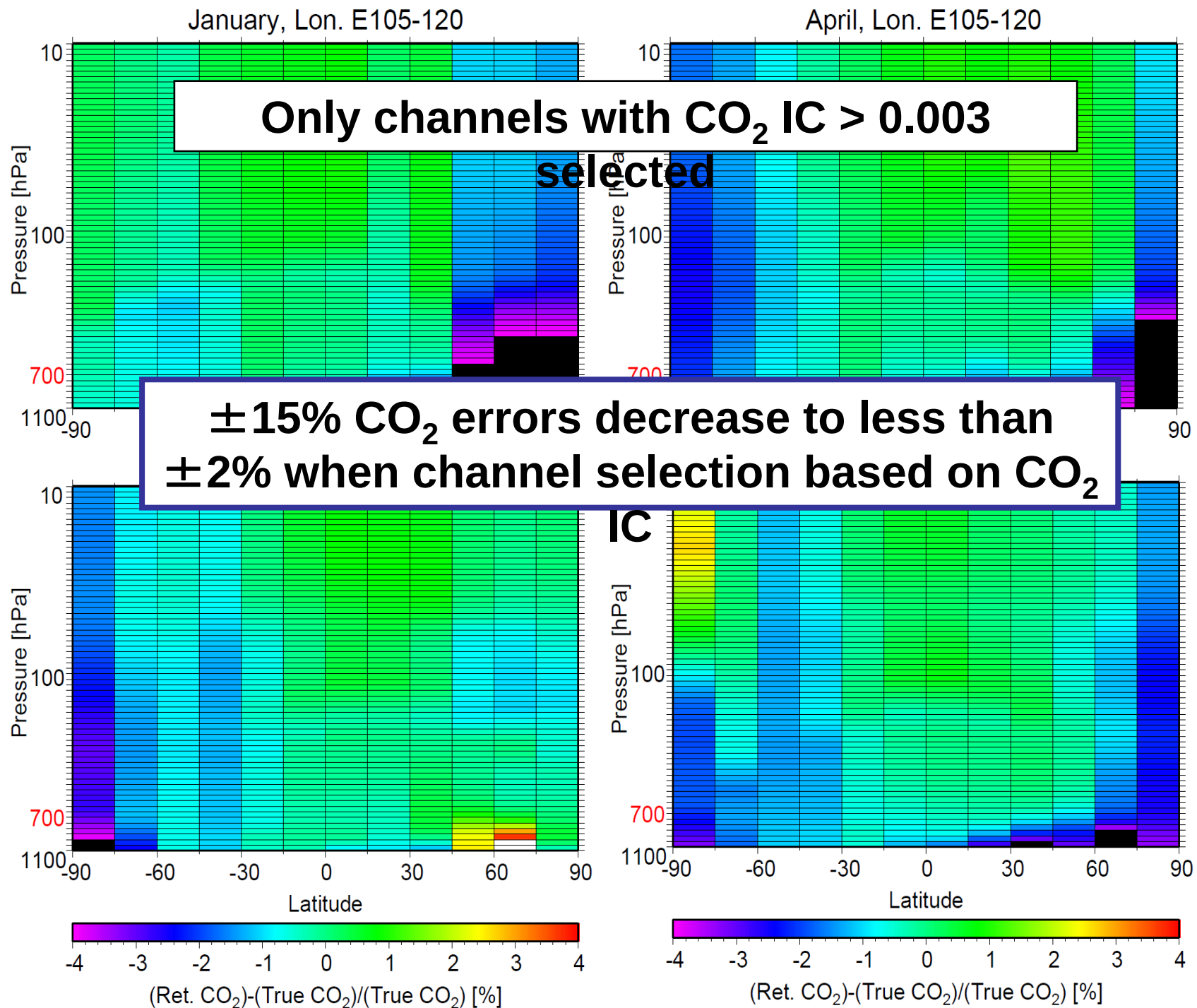


averaging kernel at low lat.



Spectra at low & mid-latitudes with large T gradient contain more CO₂ information than at high latitude.

Retrieval grid depends on latitude (and season).



In order to reduce the retrieval error down to 1 %, "averaging" is necessary

Assumption:

- Temp. estimation error : random
- CO₂ variation : systematic

(e.g., Engelen et al., 2002)



Averaging causes degradation of vertical and temporal structure of CO₂ distribution retrieved

Simulation:

How degraded by averaging even if the best performance of retrieval can be achieved.



Convolving original profile with "**Averaging Kernel**", then averaging **spatially** and **temporally**.

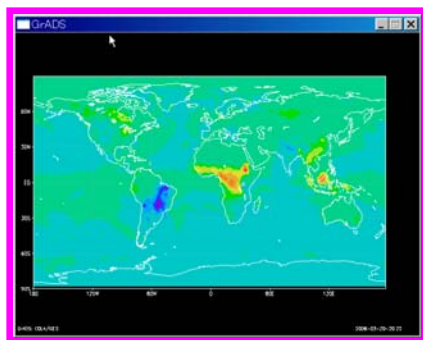
CO₂ distribution
(CO₂ trans. model:
"NICAM-CO2")

GOSAT TIR observation simulator

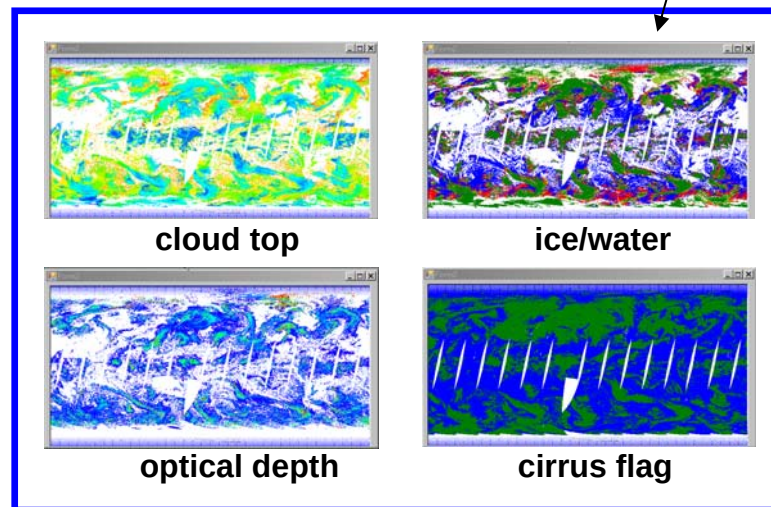
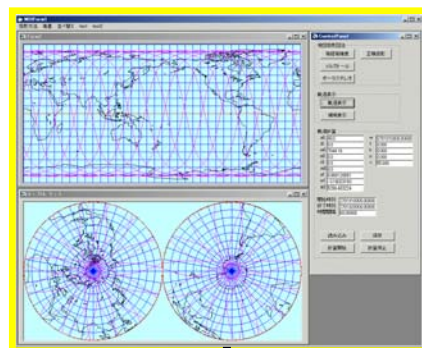
now emissivity and
vegetation data are available

Satellite orbit, obs. modes

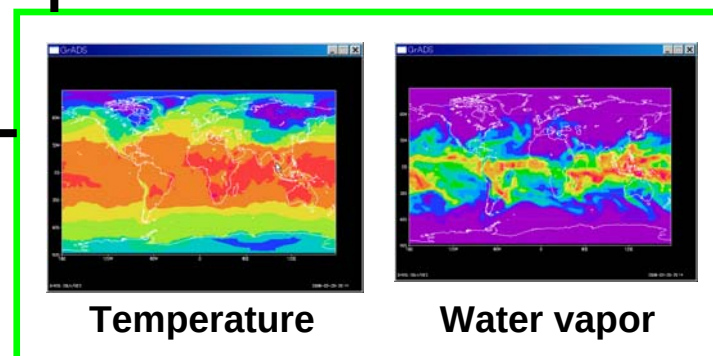
MODIS cloud data



by Yosuke Niwa

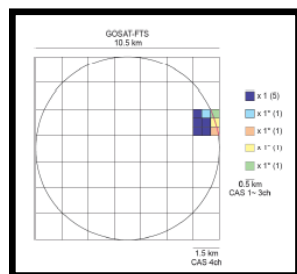


Meteorological data (ERA40)

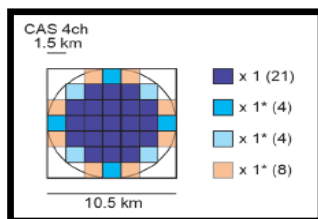


Forward models

Radiance in each IFOV



TANSO-FTS IFOV



TANSO-CAI IFOV

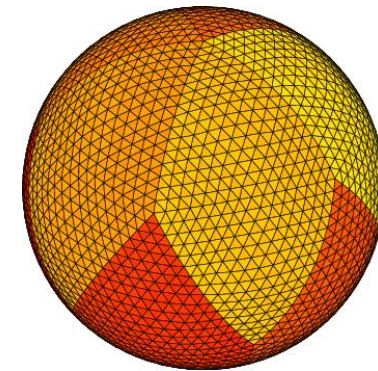
GOSAT sim. dataset
(Interferogram, spectrum)

General circulation model:
NICAM (Nonhydrostatic ICosahedral grid Model)-based
CO₂ transport model (developed by Mr. Yosuke Niwa)

NICAM was originally developed by Prof. Satoh and Dr. Tomita

Advantages

- High-resolution
 - No polar problem
- Mass conservation
 - Tracer advection is consistent with continuity
 - Tracer masses are completely conserved without a mass fixer
- Easier to develop a nest generation inverse model



Icosahedral grid

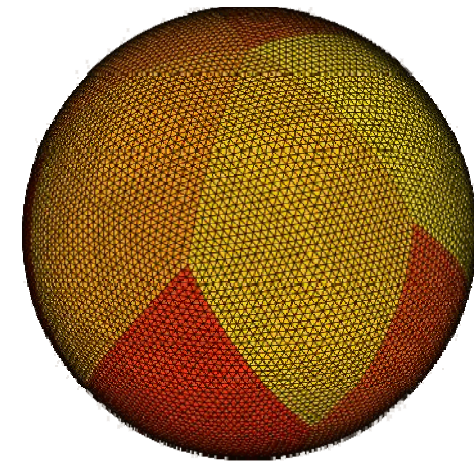
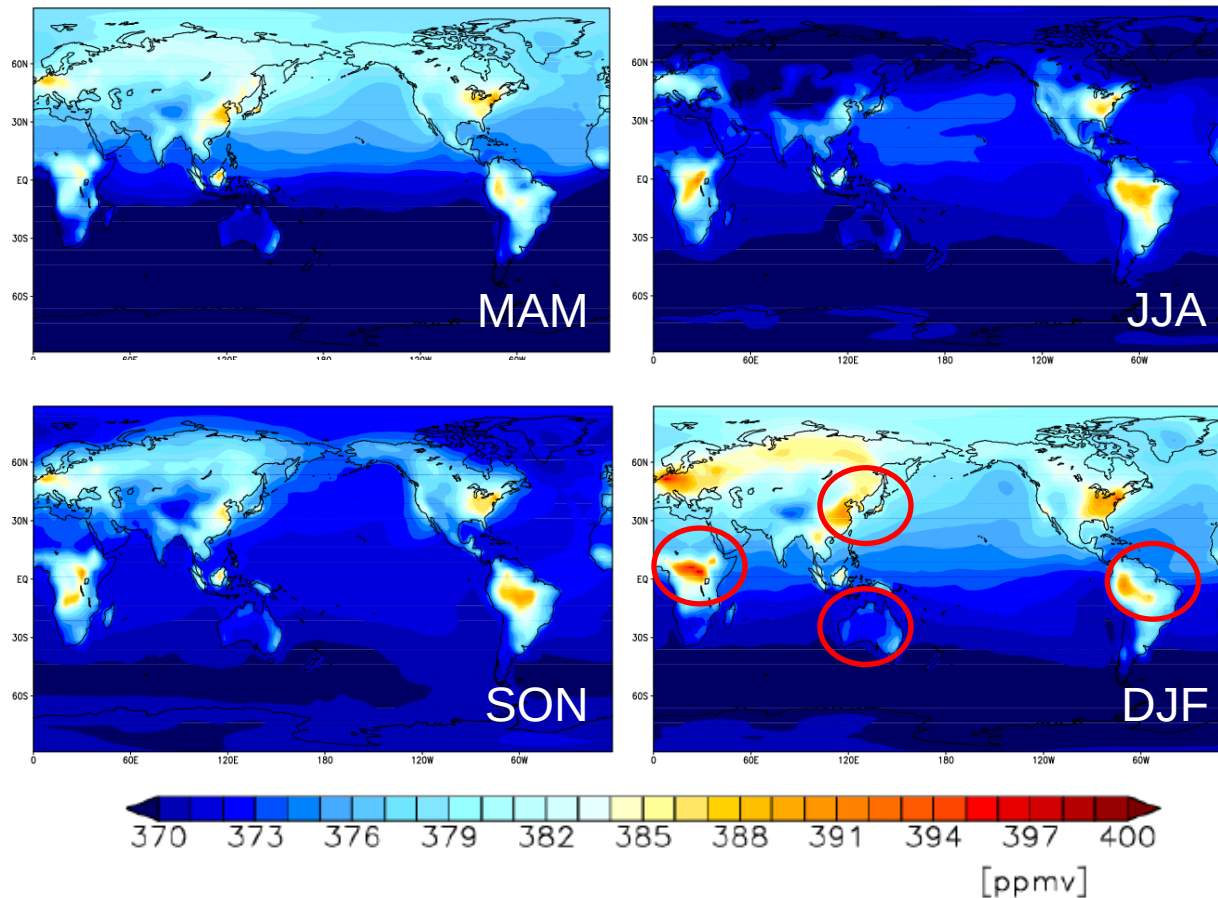
Schemes and setting parameters

- Horizontal resolution: glevel-5 (dx~240km)
- Vertical layer: 54 layers
- Advection scheme : Miura2004 scheme (no limiter version)
- Cumulus convection scheme : simplified prognostic Arakawa-Schubert
- Vertical diffusion scheme : Mellor-Yamada 2 with modification by Smith (1990)

GCM based CO₂ transport mode, NICAM-CO₂

NICAM (Nonhydrostatic ICosahedral Atmospheric Model)

Seasonal variation of surface CO₂ concentration



NICAM grid
(glevel-5:dx~240km)

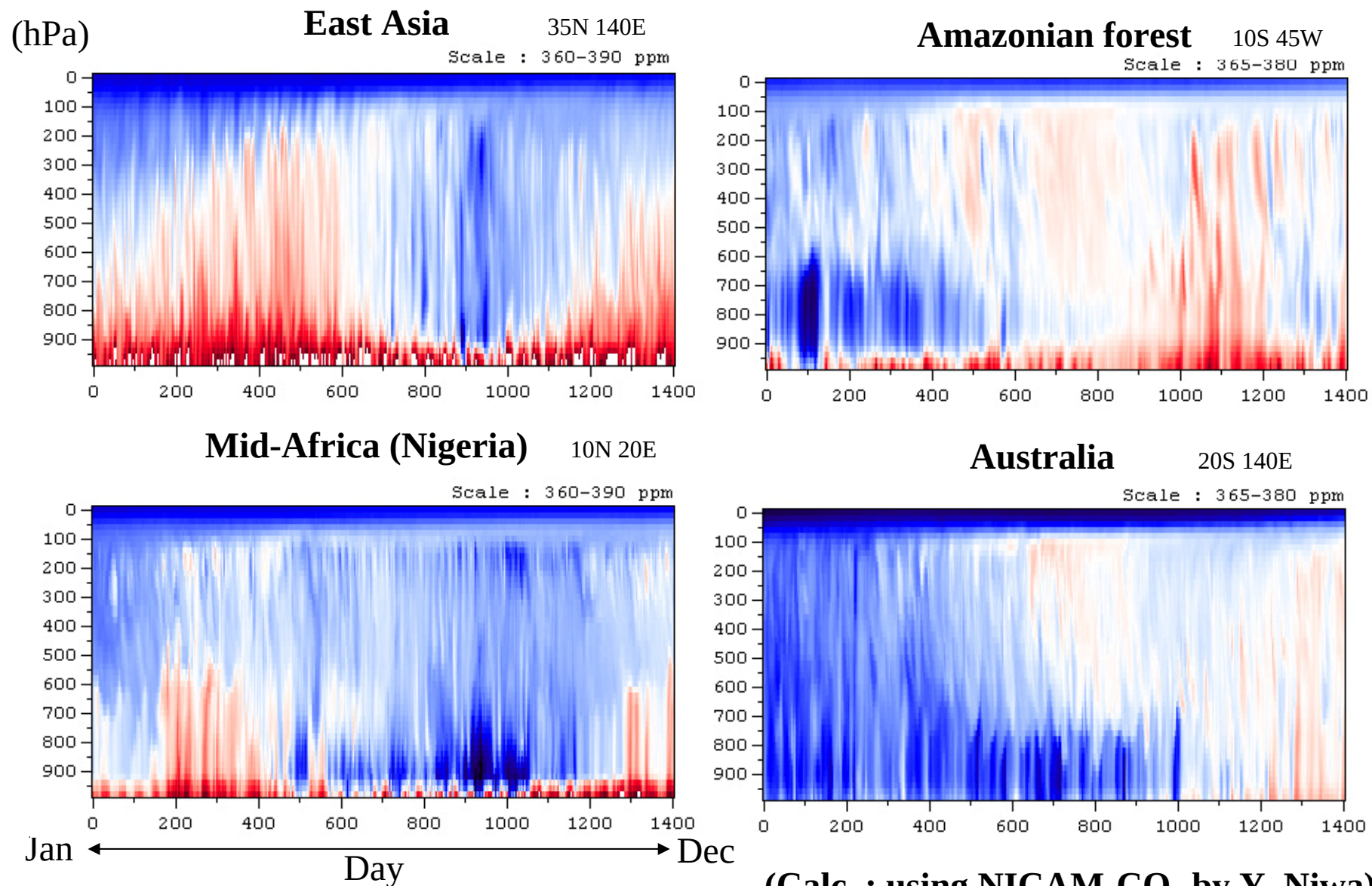
Surface fluxes

- NEP flux derived from CASA model
- Fossil fuel (CDIAC)
- Air-sea change (Takahashi et al., 2002)

Original

CO₂ concentration (original)

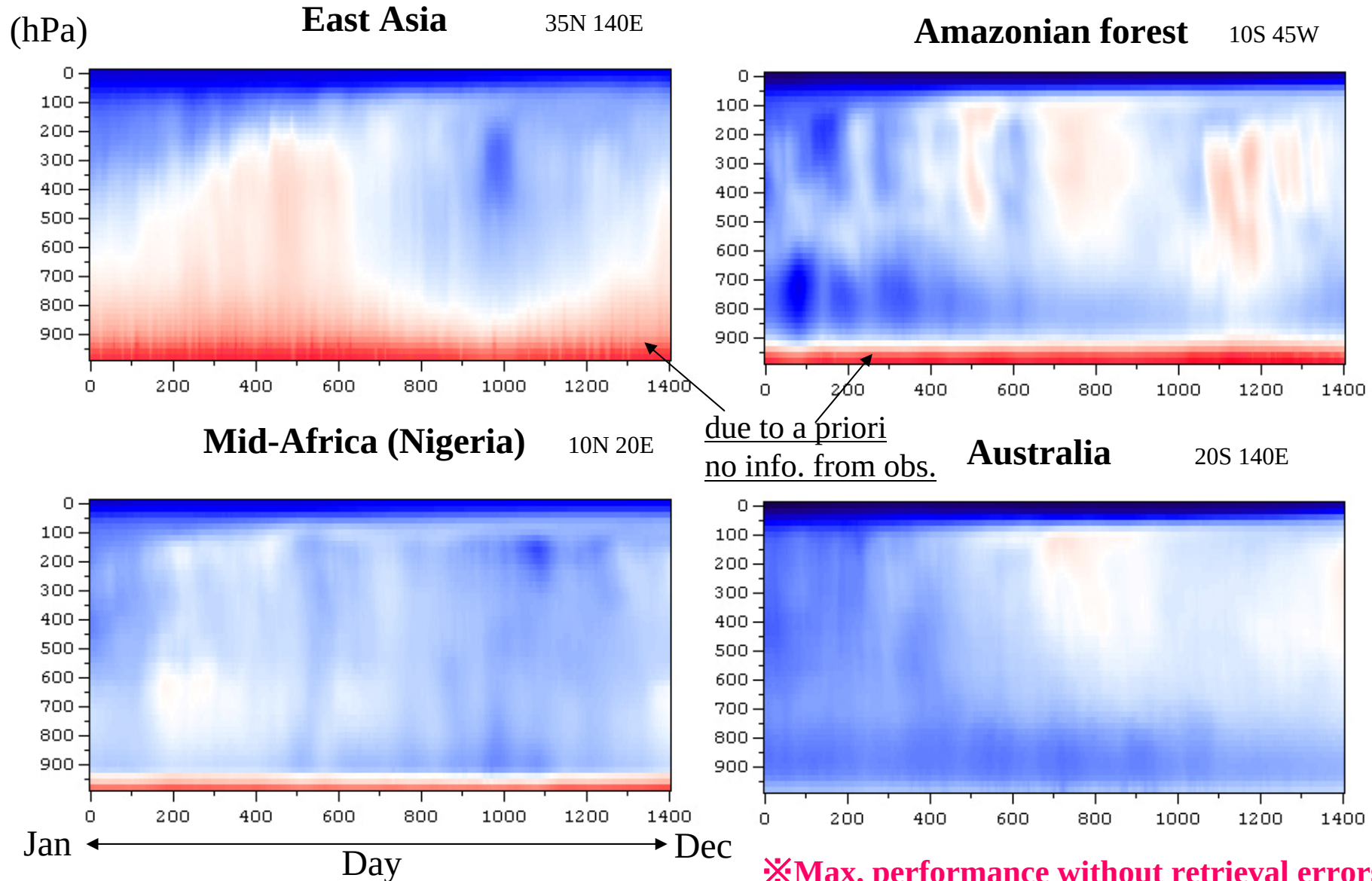
(4 times/day)



Convolved with
Averaging kernel
before averaging



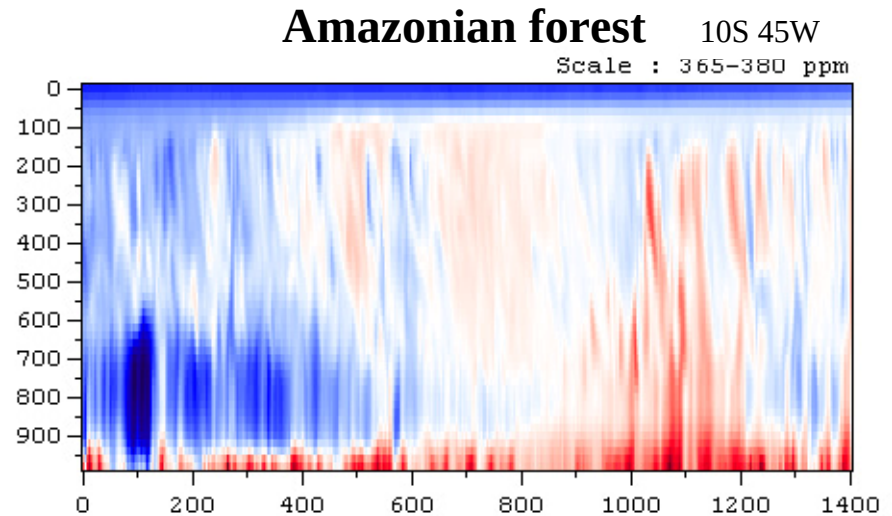
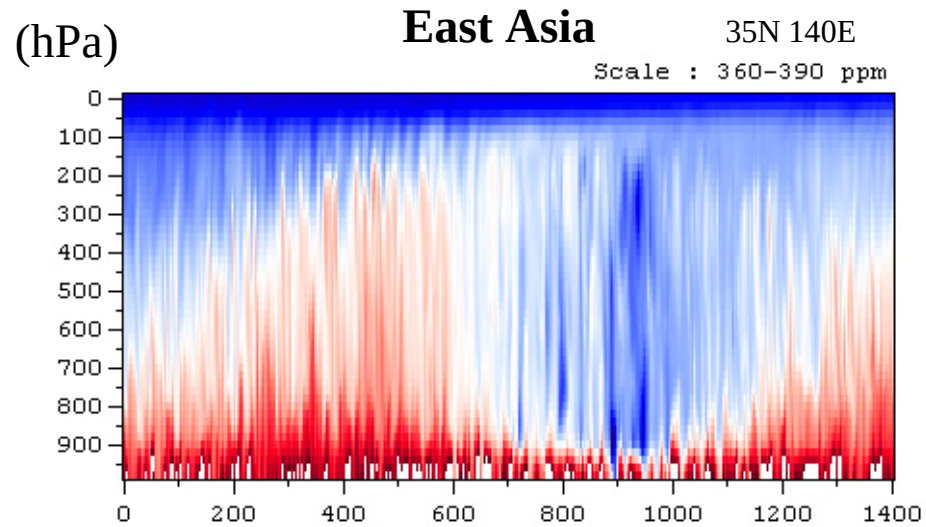
CO₂ concentration (retrieved-averaged)
(spatially averaged: $4.5^\circ \times 4.5^\circ$; temporally: 15 days)



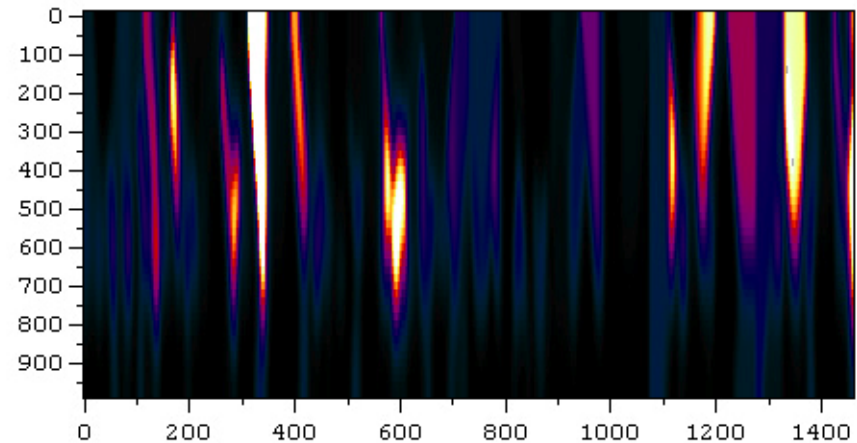
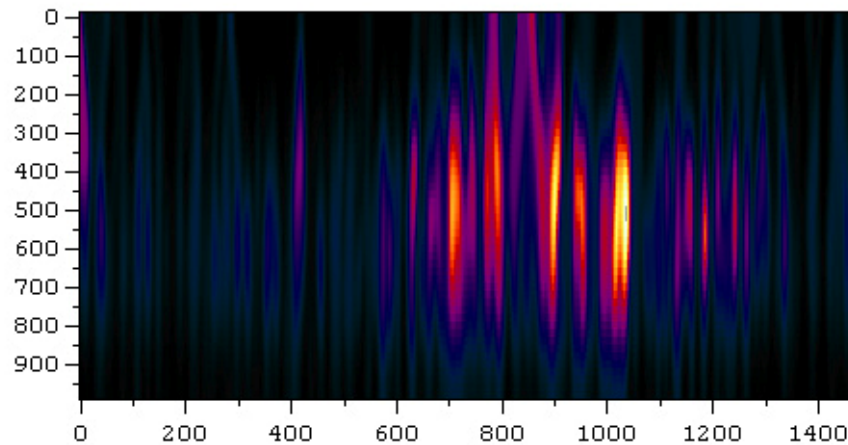
✖Max. performance without retrieval errors

CO₂ concentration (original)

(4 times/day)



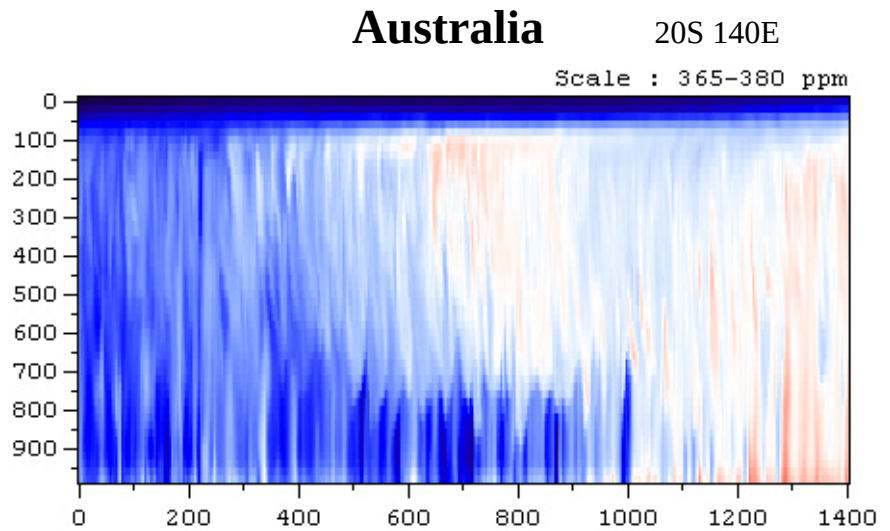
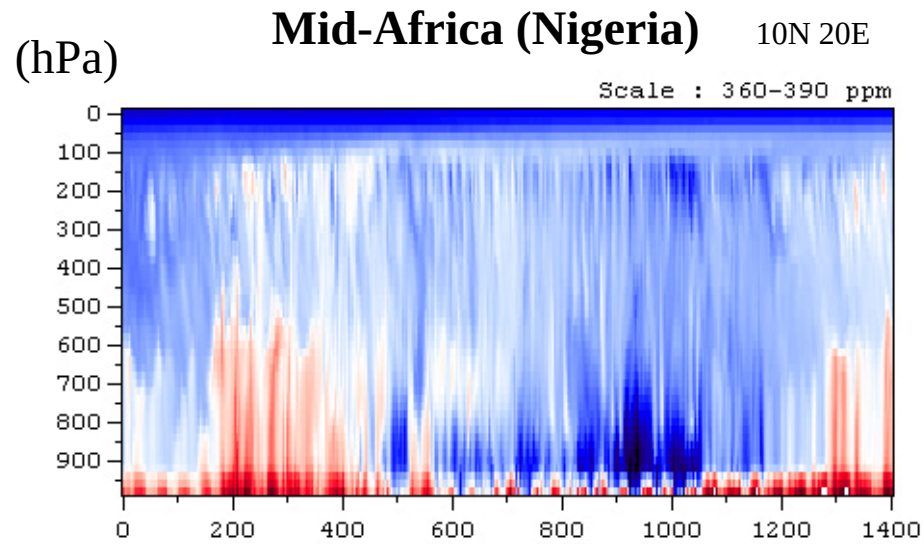
Cloud Fraction in GCM grids (NICAM-CO₂ model output)



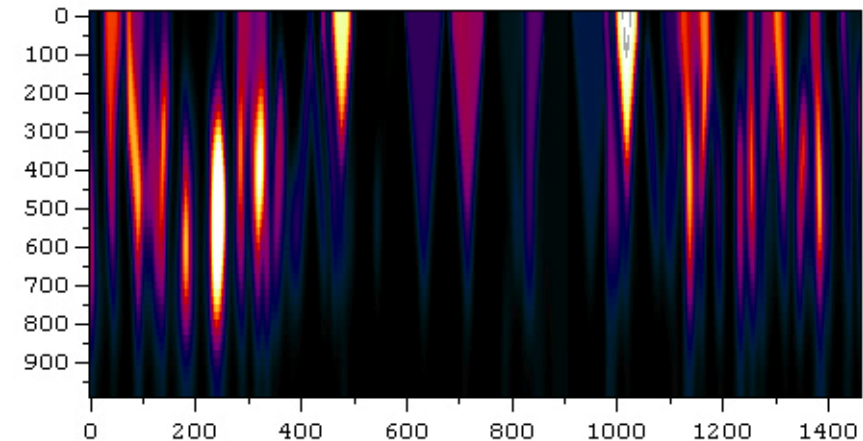
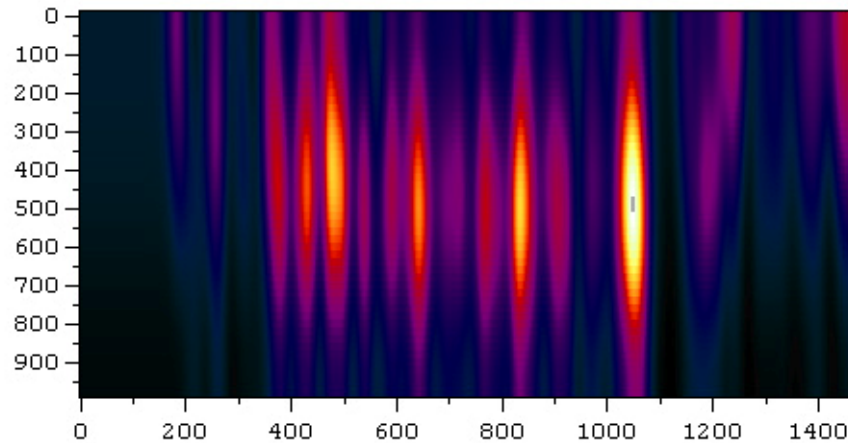
Jan ← Day → Dec

CO₂ concentration (original)

(4 times/day)

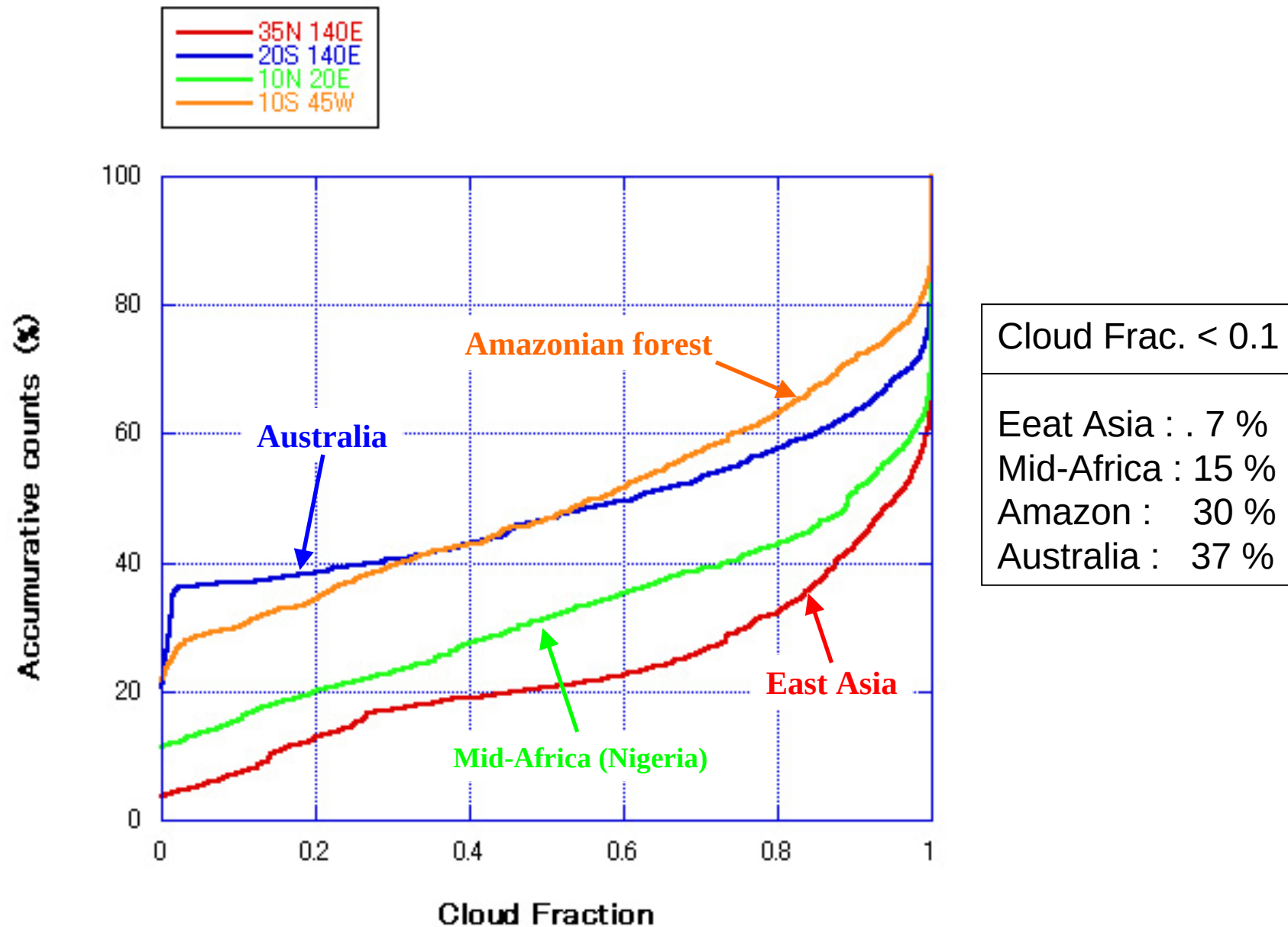


Cloud Fraction in GCM grids (NICAM-CO₂ model output)



Jan ← Day → Dec

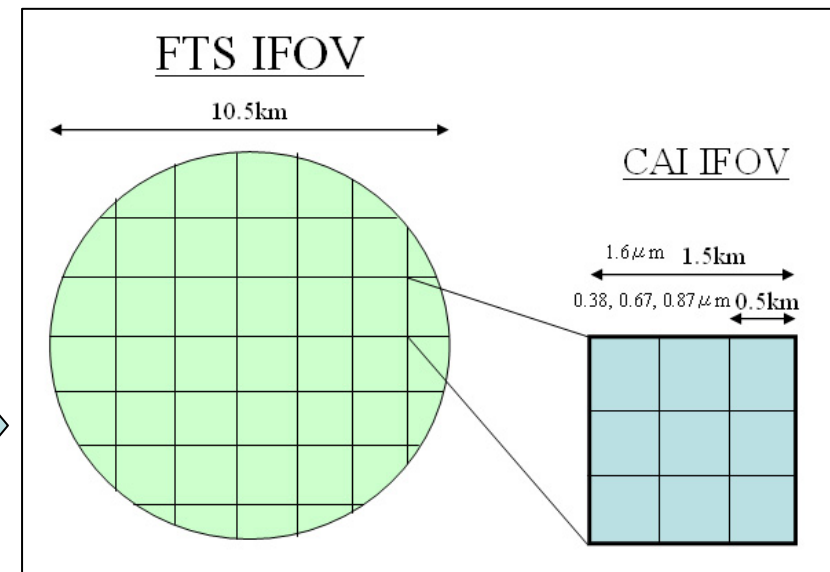
Cloud Fraction in GCM grids (NICAM-CO₂ model output)



Cloud screening

Band	IFOV	Cloud/aerosol properties	
FTS	$0.76 \mu\text{m}$ $1.6 \mu\text{m}$ $2.0 \mu\text{m}$ $5.5\text{-}14 \mu\text{m}$	10.5 km Surf. Press., Cloud top height CO_2 Cirrus clouds Cloud top height , Opt. thickness ($8\text{-}10$, $11\text{-}13 \mu\text{m}$), Ice/liquid water ($11\text{-}13 \mu\text{m}$) Cirrus clouds ($11\text{-}13 \mu\text{m}$)	Day/Night ▪ Slicing method (Z, Frac. cov.) ▪ Split window method ▪ Cloud score approach ▪ Adiabatic lapse rate analysis based on the retrieve temp.
CAI	380nm 670nm 870nm $1.6 \mu\text{m}$	0.5 km Absorbing aerosols Optical thickness Optical thickness , Ice/Vegetation 1.5 km Ice/liquid water	

Fractional cloud coverage in FTS-IFOV
can be estimated at $\sim 1\%$ resolution



Summary

Retrieval method

- Temp. data : Meteorological analysis data (sonde: land, MW: ocean)
- "Area value of averaging kernel" → layer thickness → uniform sensitivity and error
- "Shannon information contents" → channel selection → reduction temp. error

CO₂ distribution feature that will be retrieved by TANSO-FTS (TIR)

- What we want to retrieve with TANSO-FTS (TIR) is . . .
- Degrading of fine structure due to spatially and temporally averaging was estimated
- Clear sky probability = ~10% or less
- Detailed simulations considering cloud coverage have been continued using GOSAT-Sim (TIR) for source/sink inversion of CO₂

Synergetic usage of GOSAT sensor data

- Cloud screening based on CAI, SWIR(2 μ m), and TIR
- Estimation of CO₂ in the boundary layer (PBL) : "Lower = column – upper"
- Usage of columnar data for retrieving profiles
"as an additional constraint" or "scaling after retrieval"