



SatCO

Ocean Carbon from Space

Joint Science Research & Satellite-based
marine carbon monitoring and analysis system

中文 English



Marine carbon observation by satellite remote sensing

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- ❑ **Framework of satellite-based marine carbon research**
- ❑ **Inorganic Carbon : $p\text{CO}_2$**
- ❑ **Carbon flux estimation**
- ❑ **SatCO₂-- Satellite-based marine carbon monitoring and analysis system**



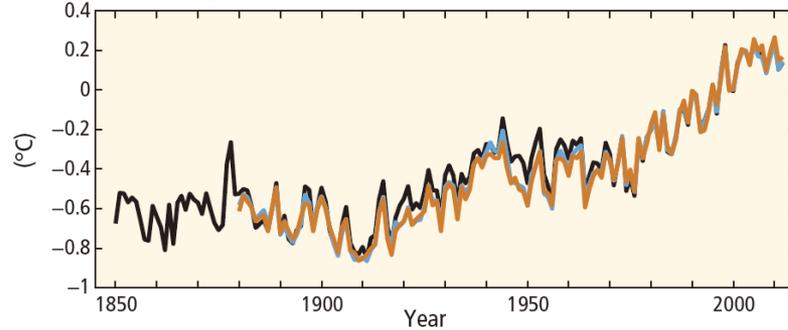
CO₂ emissions from fossil fuel combustion and industrial processes **contributed about 78 %** of the total green house gas (GHG) emission increase from 1970 to 2010 (IPCC, 2014)



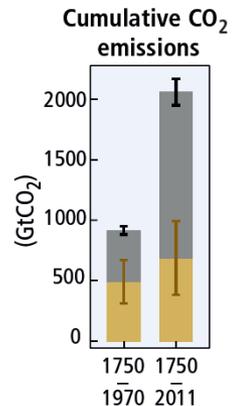
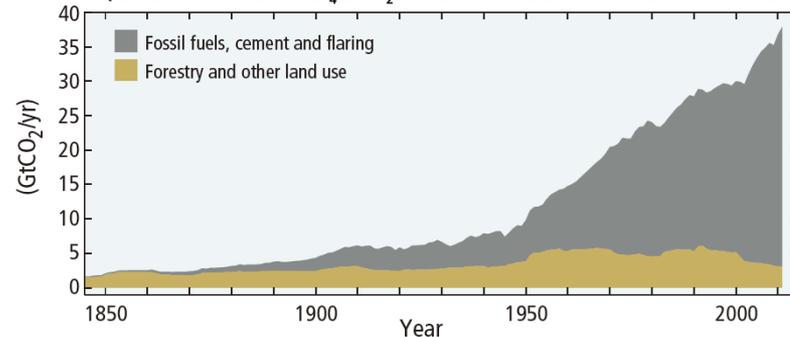
Extremely weather disasters



(a) Globally averaged combined land and ocean surface temperature anomaly



(d) Global anthropogenic CO₂ emissions
Quantitative information of CH₄ and N₂O emission time series from 1850 to 1970 is limited



The globally averaged combined land and ocean surface temperature data as calculated by a linear trend show **a warming of 0.85° C** over the period **1880 to 2012** (IPCC 2014).

The goal of the Global Carbon Project (GCP) is

to develop comprehensive, policy relevant understanding of the global carbon cycle, encompassing its natural and human dimension and their interactions.



Three Science Themes

1. Patterns and variability

What are the current temporal and geographical distributions of the **major pools and fluxes** in the global carbon cycle?

2. Processes and Interactions

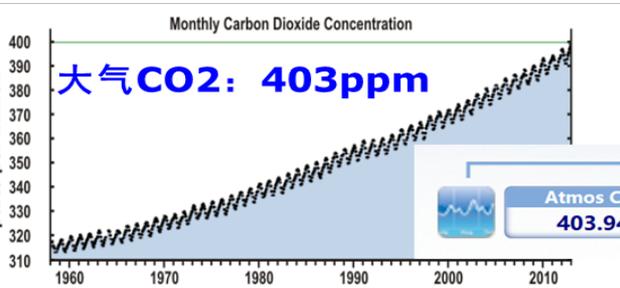
What are the **control and feedback mechanisms** - both anthropogenic and non-anthropogenic - that determine the dynamics of the carbon cycle?

3. Carbon Management

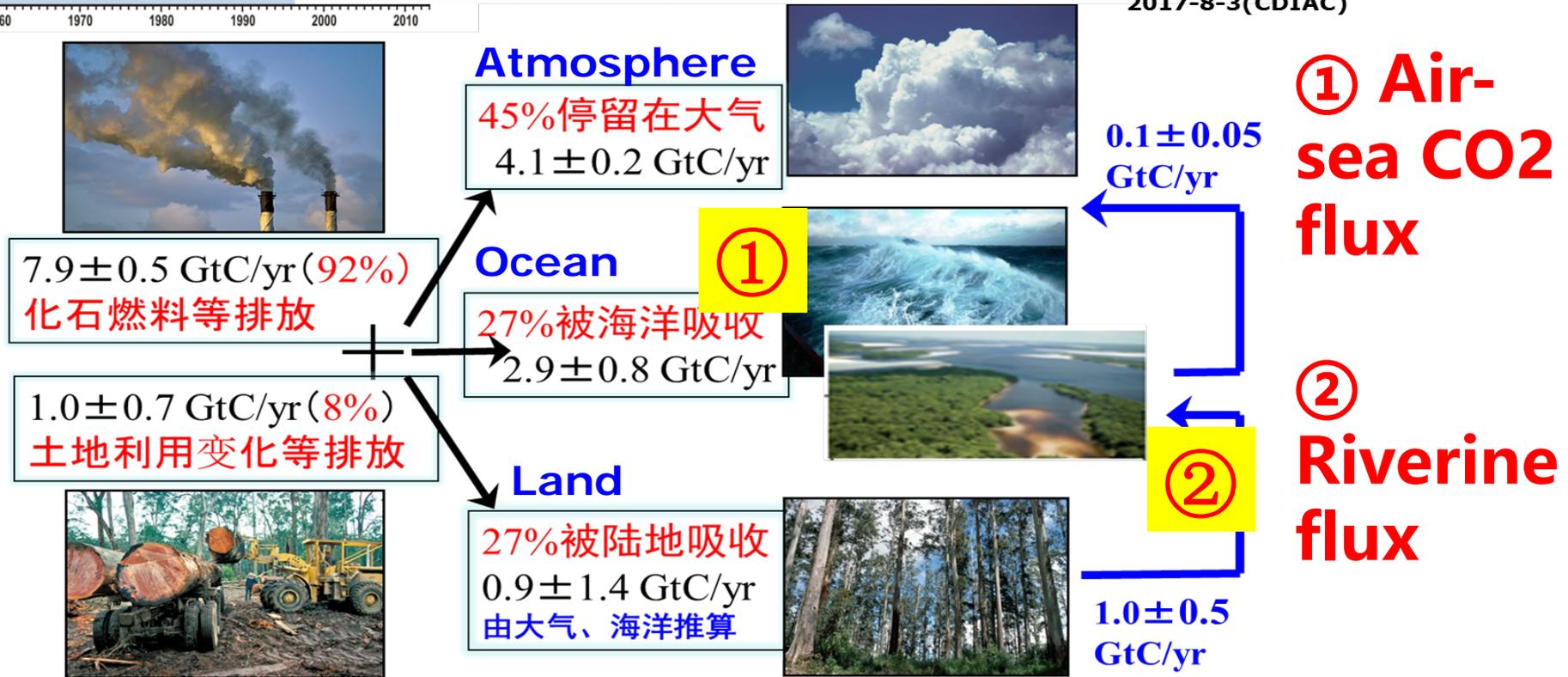
What are the likely dynamics of the carbon-climate-human system into the future, and what points of intervention and windows of opportunity exist for human societies to manage this system?

Carbon cycle and Anthropogenic CO₂

人为排放CO₂的去处



2017-8-3(CDIAC)



Source: [Le Quéré et al 2013](#); [CDIAC Data](#); [Global Carbon Project 2013](#),
[Regnier et al. Nature Geo science, 2013](#)

Gt=10亿吨

Marine Satellite Remote sensing



SatCO₂

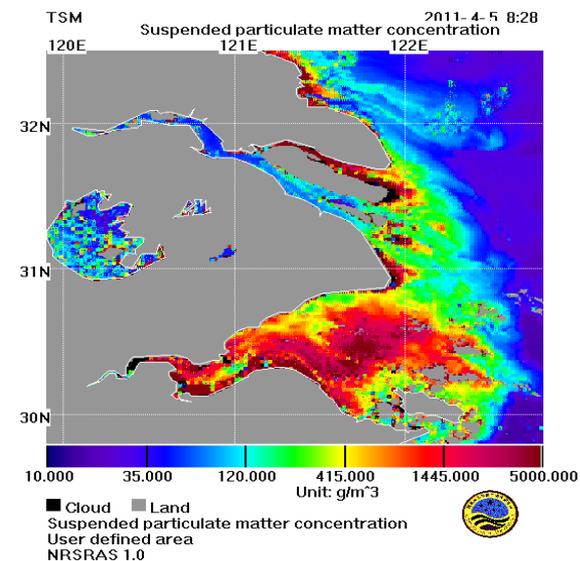
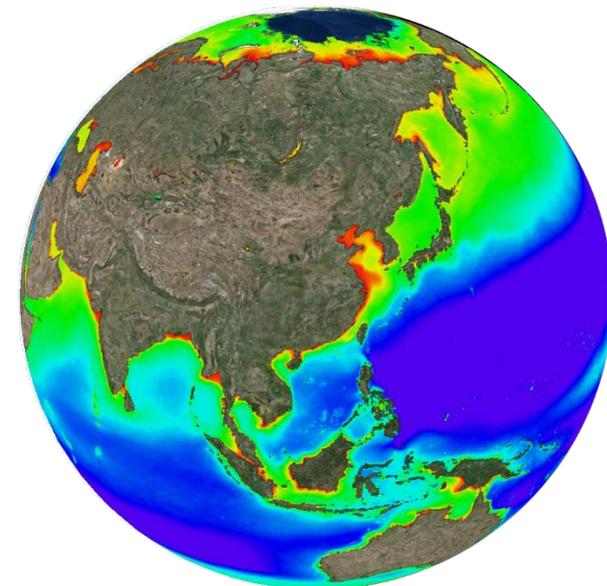
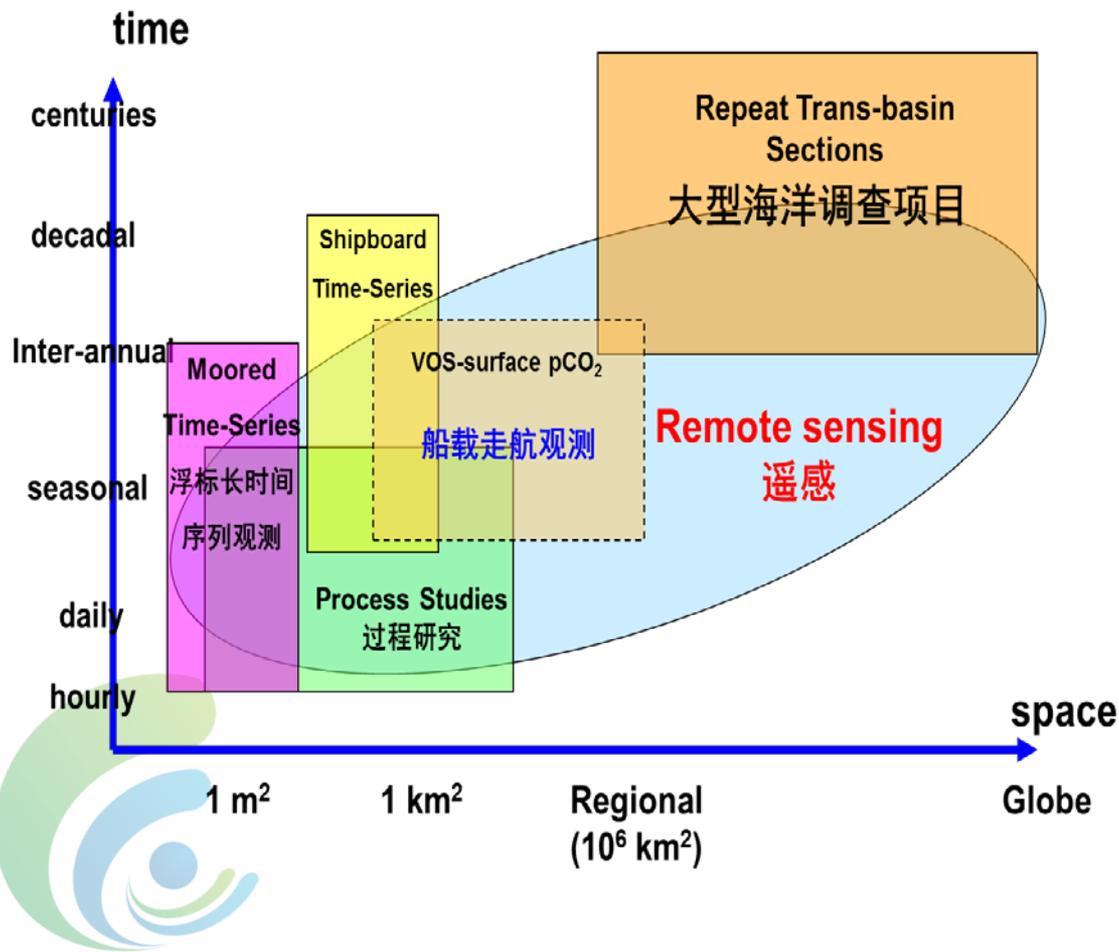
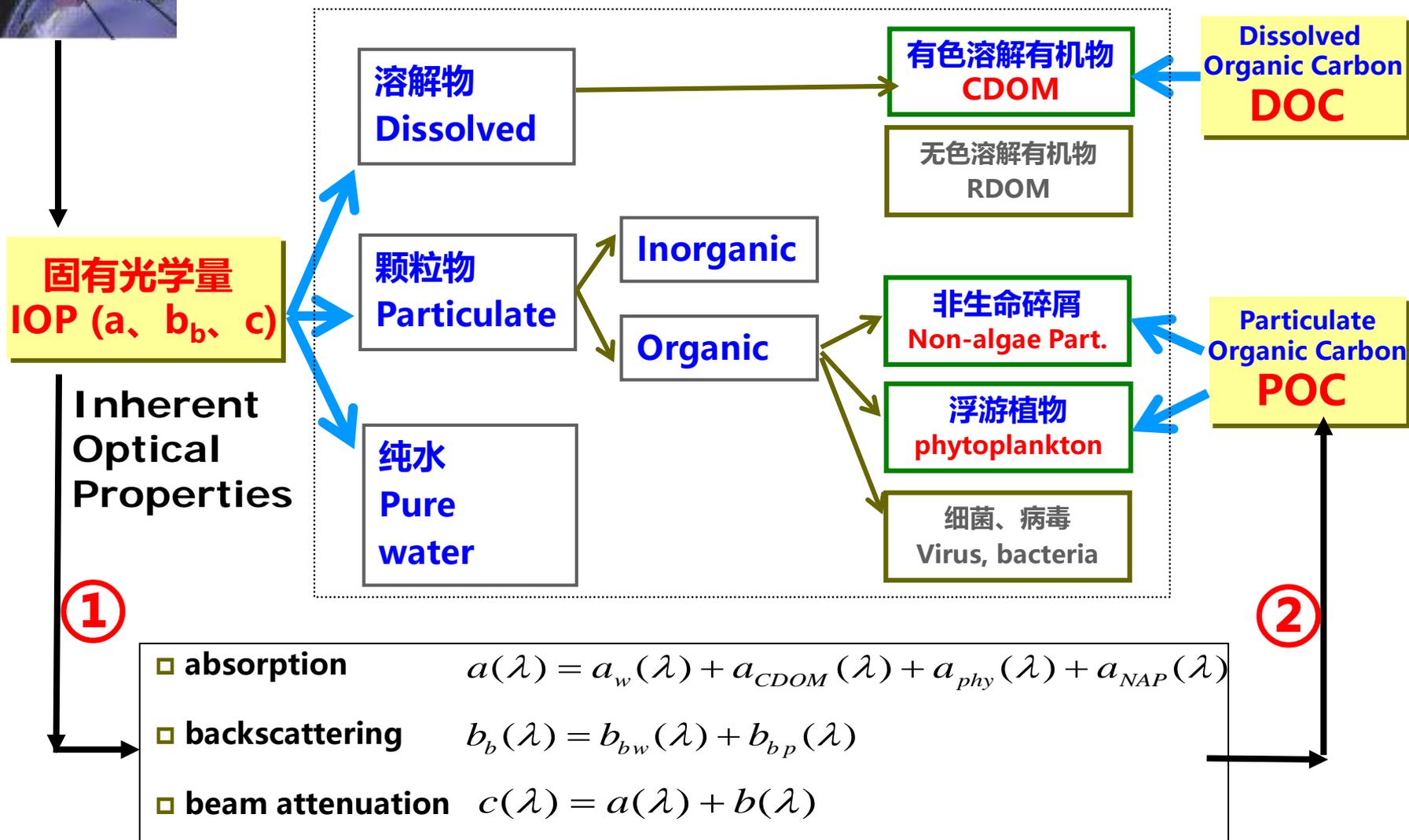




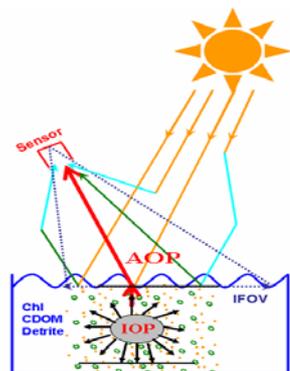
Diagram of inverse radiative transfer elements. **Many further parameters** are derived from these constituents, such as **DOC, POC** and productivity. (IOCCG 5, p9)



Framework of satellite-based marine carbon research



SatCO2



Radiation transmission
Satellite algorithms

www.satCO2.com

coasts

Marginal sea

Open Ocean

Riverine C flux

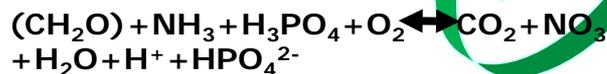


Air-sea CO₂ flux



Lateral transport

N, P, DOC, POC,
DIC, PIC, etc.



Primary production
Food-web
Acidification
Eutrophication
Hypoxia

PAR
Chl
PFT
C/Chla
Zeu
profile

Phytoplankton C

POC/DOC Inventory

POC
DOC
profile



Carbonate system

Acidification

POC export flux

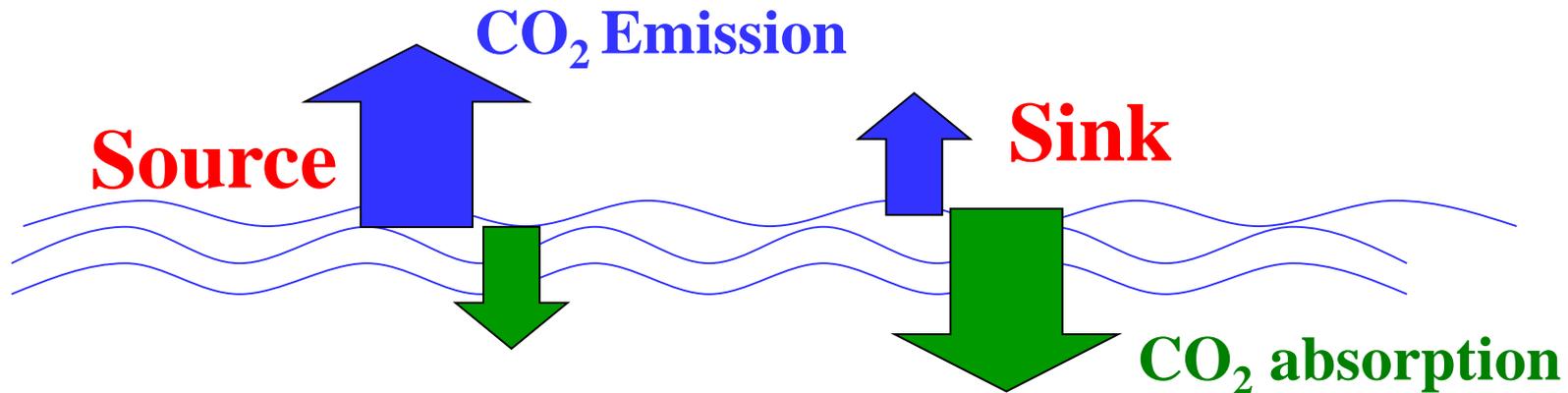
- **Framework of satellite-based marine carbon research**
- **Inorganic Carbon : $p\text{CO}_2$**
- **Carbon flux estimation**
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Estimation of CO₂ flux



SatCO₂



$$\text{Flux} = s(\text{solubility}) \times k (\text{gas transfer velocity}) \times (p\text{CO}_{2\text{sw}} - p\text{CO}_{2\text{atm}})$$

The equilibrium of gaseous and aqueous CO₂:



Subsequent hydration and dissociation reactions:



$$K_1^* = \frac{\{H^+\}[\text{HCO}_3^-]}{[\text{CO}_2]}$$

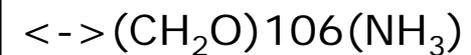
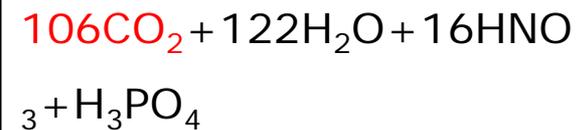


$$K_2^* = \frac{\{H^+\}[\text{CO}_3^{2-}]}{[\text{HCO}_3^-]}$$

Asterisk (*) indicates a "stoichiometric" constant

**Marine
Carbonate
System**

Photosynthesis



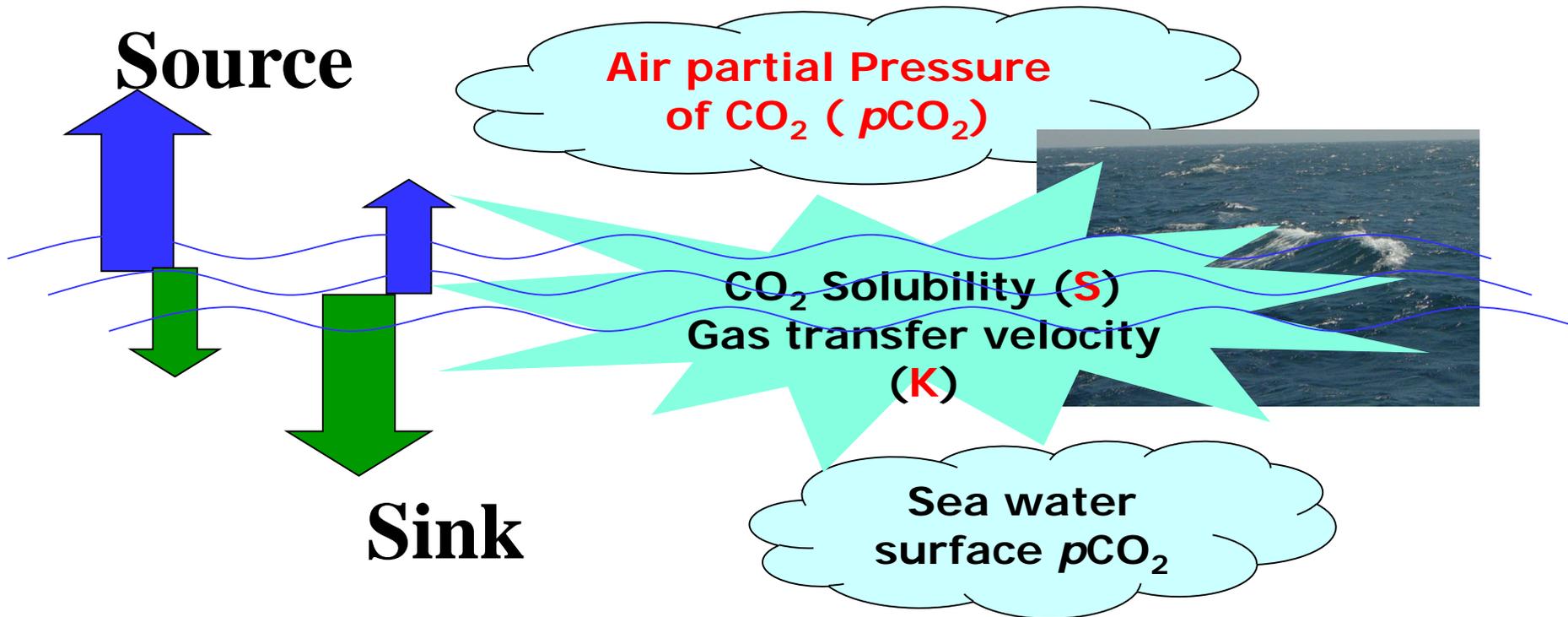
**Respiration,
Mineralization, etc.**

(pCO₂, DIC, pH, TA)

Parameters of the flux calculation

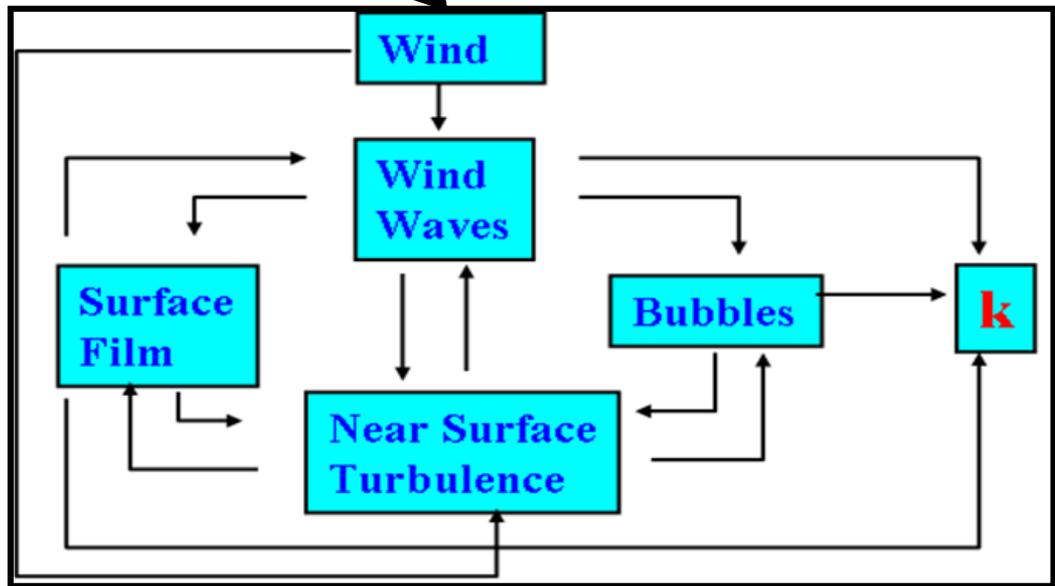
Air-sea CO₂ flux

$$F = S \times K \times (p\text{CO}_{2\text{sea}} - p\text{CO}_{2\text{air}})$$



CO₂ gas transfer velocity (K)

$$F = S \times K \times (p\text{CO}_{2\text{sea}} - p\text{CO}_{2\text{air}})$$

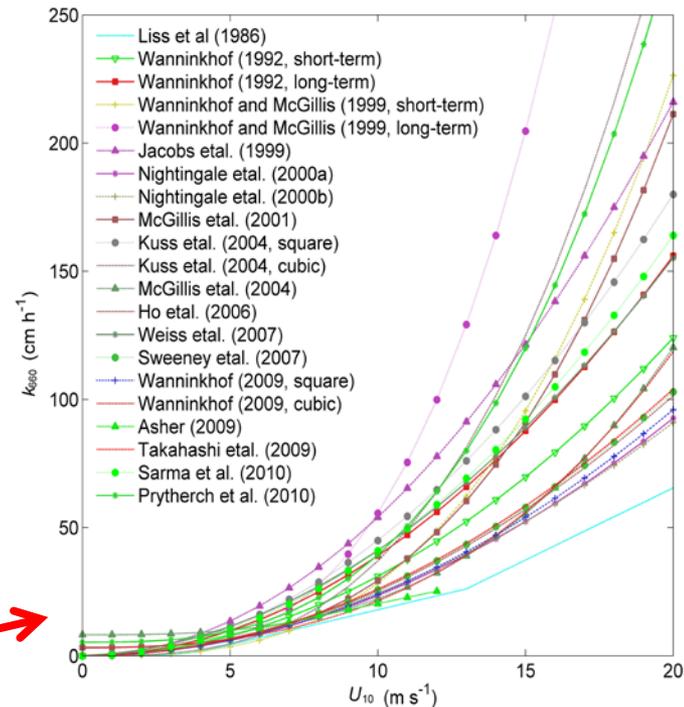


$$K = f(u^*) S_c^{-n}$$

u^* – frictional velocity

S_c – schmit number (ν/D)

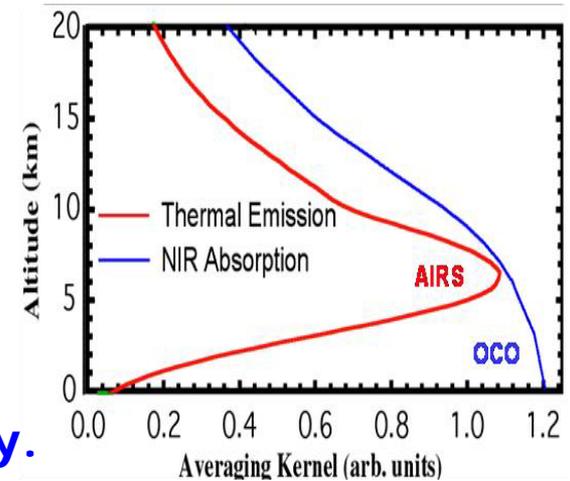
s – solubility



Parameterization of the K with **sea wind**, **wave height**, etc. with can obtained from satellite products e.g. Wanninkhof et al. (2002), Sweeney et al. (2007), Jiang et al. (2008), etc.

Satellite observation of atmosphere CO₂

- Lidar-can measure CO₂ at bottom layer, low spatial resolution.
- Thermal emission- measure CO₂ at middle layer, insensitive to the bottom layer
- Reflected Sunlight (NIR CO₂ absorption)- measure CO₂ at whole column, high accuracy.



Measurement Method	Instrument	CO ₂ Measurement	Measurement Precision	Down-track Sampling
Reflected Sunlight	OCO	Total Column	1 ppm	2.3 km
	SCIAMACHY	Total Column	3-10 ppm	60 km
	GOSAT	Total Column	4 ppm	10.5 km
Thermal Emission	AIRS	Mid-Trop	1 – 2 ppm	45 km
	IASI	Mid-Trop	38 ppm	100 km
	TES	Mid-Trop	~5 ppm	~50 km
Active (LIDAR)	ASCOPE	Lower-trop	2 – 4 ppm	~100 km
	ASCENDS	Lower-trop	2 – 4 ppm	~100 km

OCO—NASA

- NASA approved the OCO satellite (Orbiting Carbon Observatory) mission in 2002 to monitoring global atmosphere CO₂. Accuracy goal of 0.3~0.5%(1-2ppm). Reflected Sunlight method
- OCO was launched on **24 Feb. 2009**, but failed.
- NASA launched in Jul. 2014 successfully.

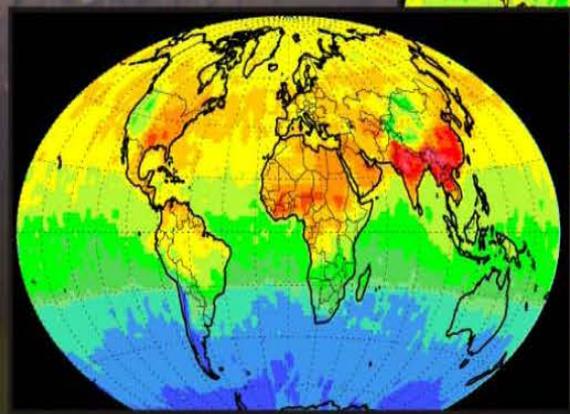
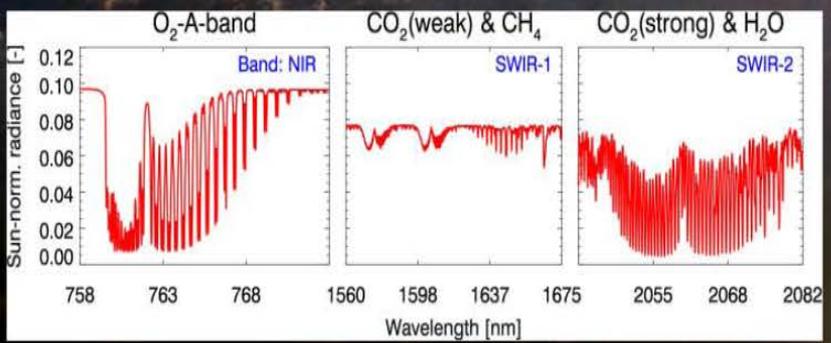
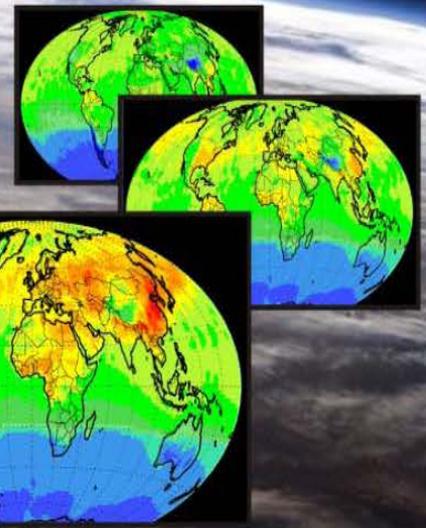
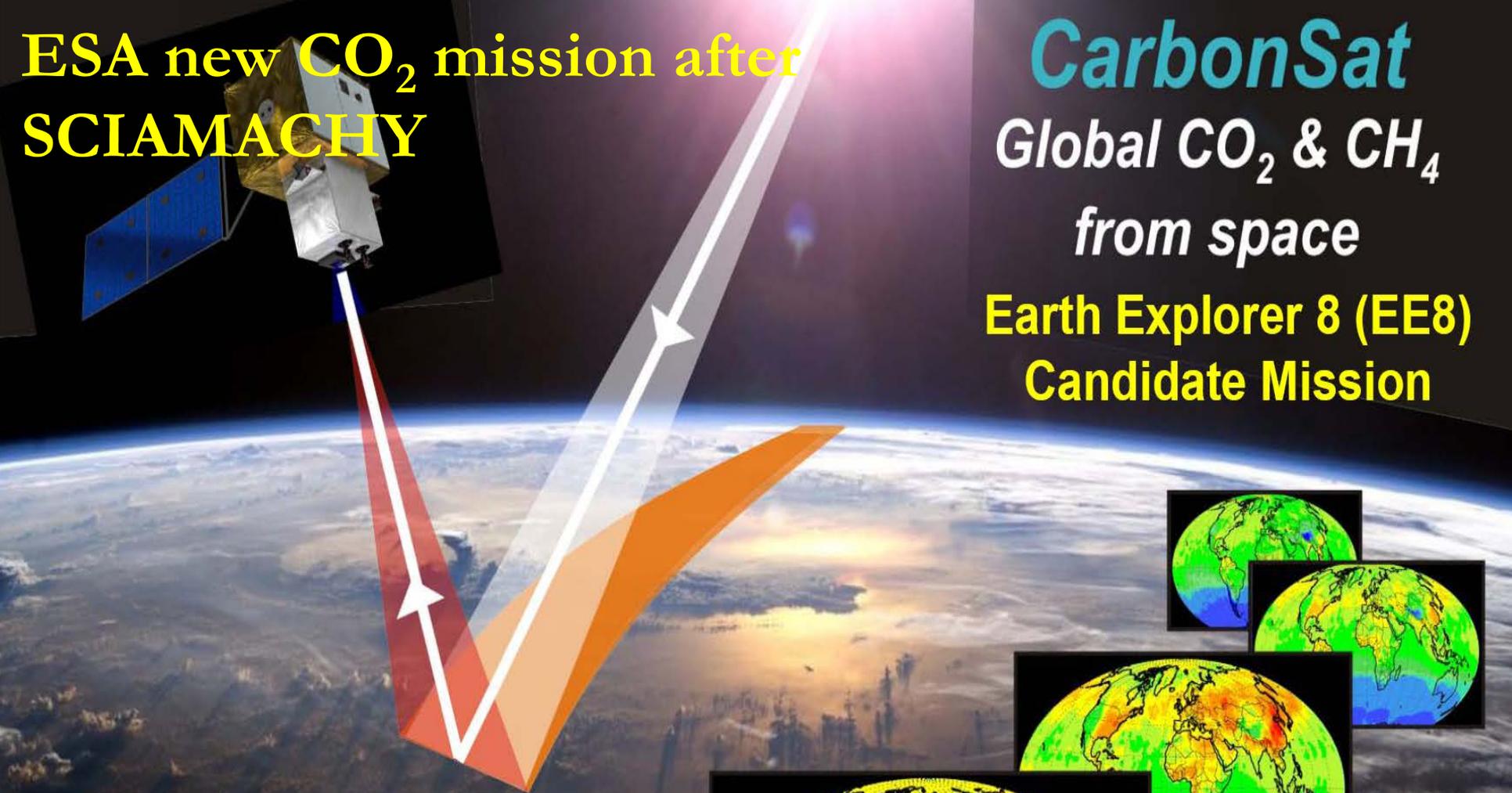
GOSAT-JAXA (Japan)

- GOSAT-**Greenhouse Gases Observing Satellite**, launched on **23 Jan. 2009**. Reflected Sunlight method.
- GOSAT-2 was **launched in Oct. 29, 2018**.

ESA new CO₂ mission after SCIAMACHY

CarbonSat Global CO₂ & CH₄ from space

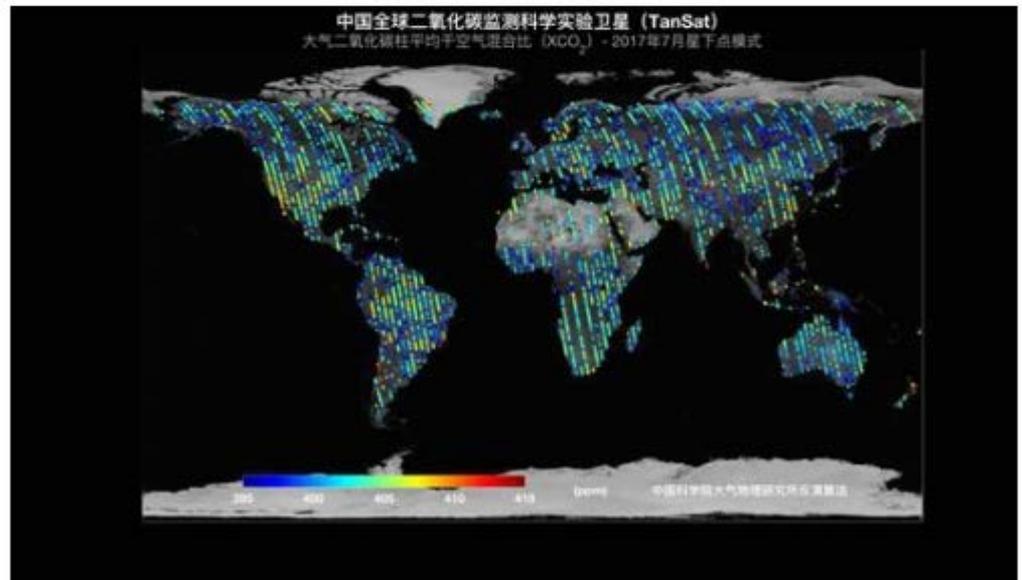
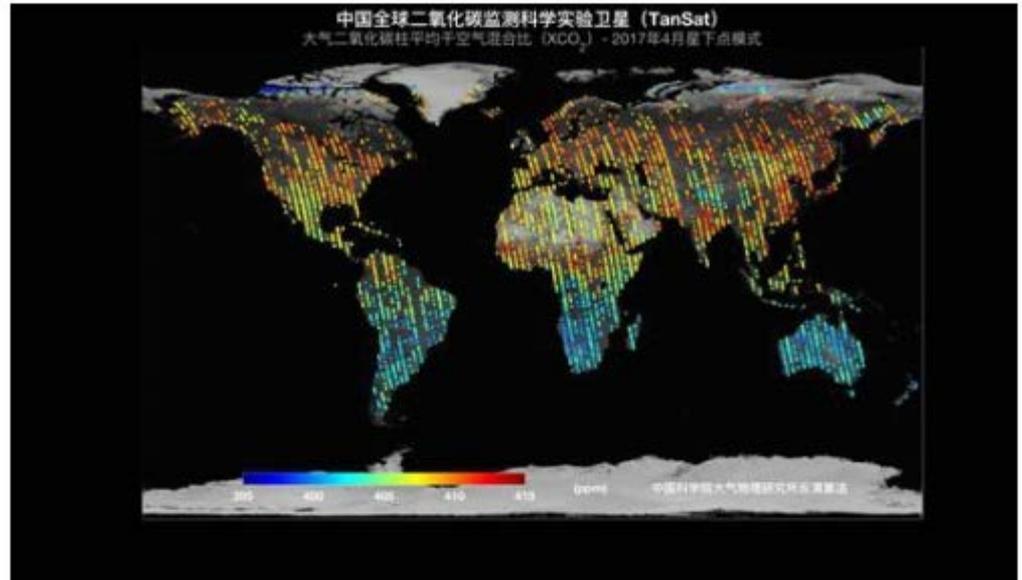
Earth Explorer 8 (EE8)
Candidate Mission



Chinese CO₂ satellite mission

- Scientific experimental satellite. Main payloads include high spectral CO₂ sensor, multiple channel cloud and aerosol.
- Launched in 2017. CO₂ measuring accuracy to be 4ppm.

The first global map of atmospheric carbon dioxide in China's carbon satellite, (above) April 2017, (below) July 2017. The color indicates the average dry air mixing ratio (XCO₂) of the atmospheric Carbon dioxide column.

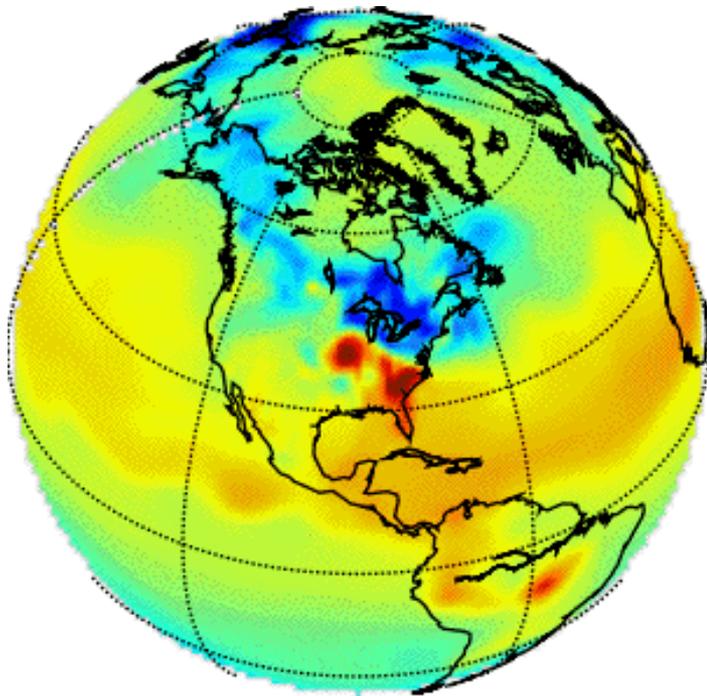


Ⓢ Air CO₂ from the model simulation data

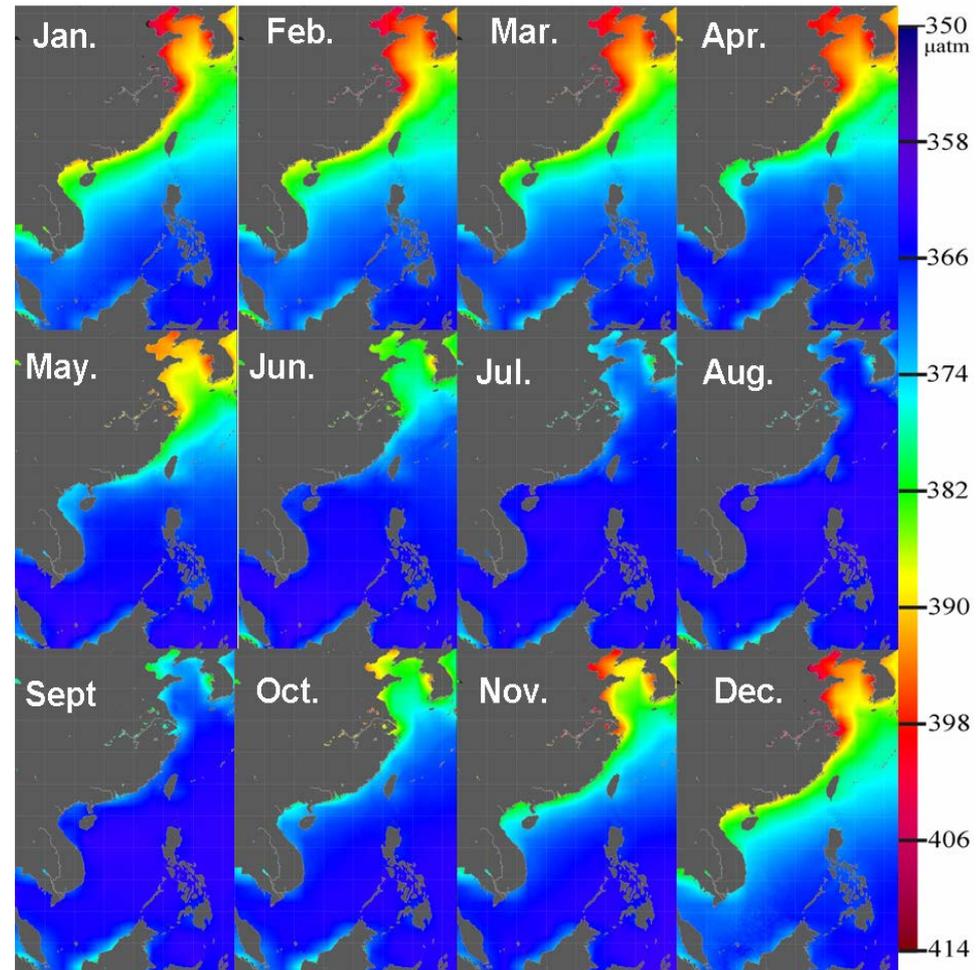
$$p\text{CO}_{2\text{atm}} = x\text{CO}_{2\text{atm}} \times (P_{\text{atm}}/1013.25 - p\text{H}_2\text{O})$$

➤ NOAA/CMDL/Carbon Tracker

➤ $x\text{CO}_{2\text{atm}}$ = CO₂ Dry Air Mole Fraction



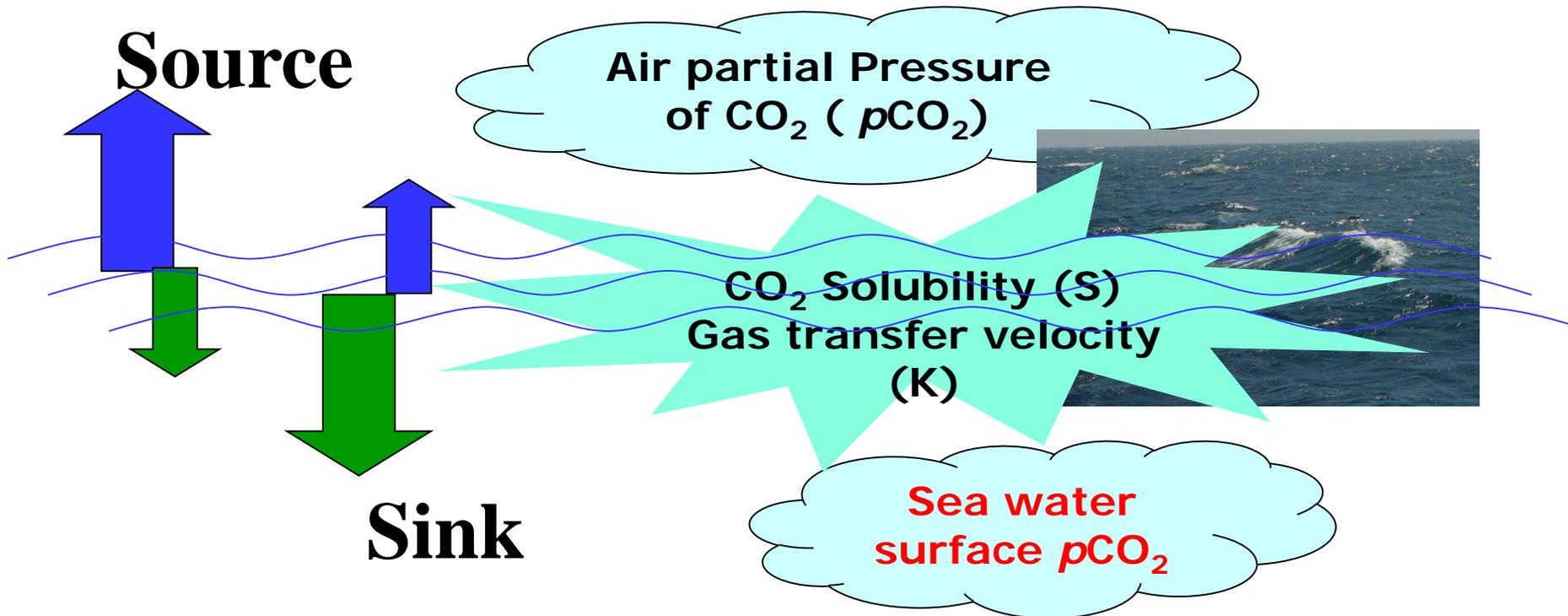
Climatologic atmospheric $p\text{CO}_2$
(2003-2009) monthly-averaged



Parameters of the flux calculation

Air-sea CO₂ flux

$$F = S \times K \times (p\text{CO}_{2\text{sea}} - p\text{CO}_{2\text{air}})$$



CO₂ Speciation

CO₂(g) has many possible transformations upon dissolution in H₂O

Major dissolved forms:

CO_{2(aq)} (*aqueous carbon dioxide* – a dissolved gas)

H₂CO₃ (*carbonic acid* – trace amount)

HCO₃⁻ (*bicarbonate ion*)

CO₃⁻² (*carbonate ion*)

The equilibrium of gaseous and aqueous CO₂:



Subsequent hydration and dissociation reactions:



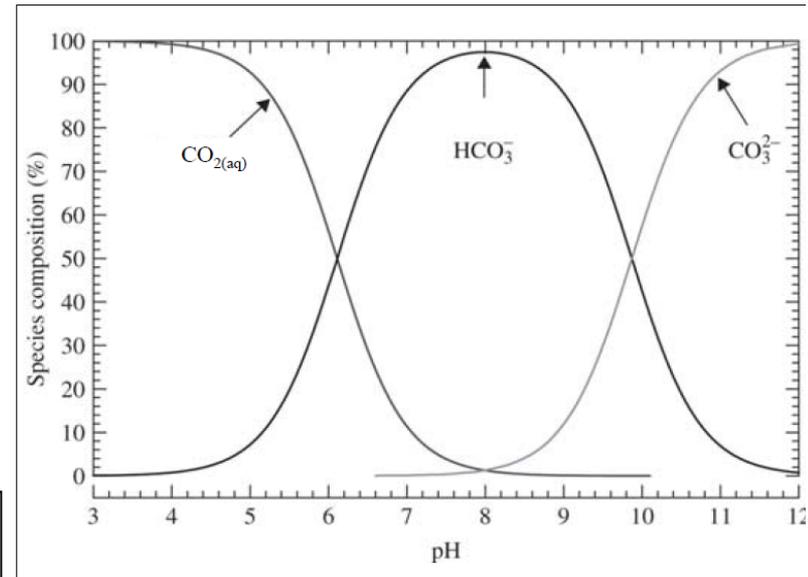
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$$K_2^* = \frac{\{H^+\} [CO_3^{2-}]}{[HCO_3^-]}$$

Asterisk (*) indicates a "stoichiometric" constant

Partial pressure of carbon dioxide (pCO₂)



Seawater values shown --- freshwater curves are shifted left

- $\text{pH} = -\log \{H^+\}$
- *Total Alkalinity (TA)* represents ability of seawater to resist pH change upon addition of acid
- Any two of the four CO₂ properties (ΣCO_2 , P_{CO_2} , pH, and *carbonate alkalinity*) can be used to determine the CO₂ system

Remote sensing algorithm of Aquatic $p\text{CO}_2$

Sea-air flux of CO_2 in the North Pacific using shipboard and satellite data

Mark P. Stephens, Geoffrey Samuels, Donald B. Olson, and Rana A. Fine
Rosenstiel School of Marine and Atmospheric Science, University of Miami, Miami, Florida

Taro Takahashi

Lamont-Doherty Earth Observatory, Columbia University, Palisades, New York

JOURNAL OF GEOPHYSICAL RESEARCH, VOL. 100, NO. C7, PAGES 13,571-13,583, JULY 15, 1995

$p\text{CO}_2$ vs. SST

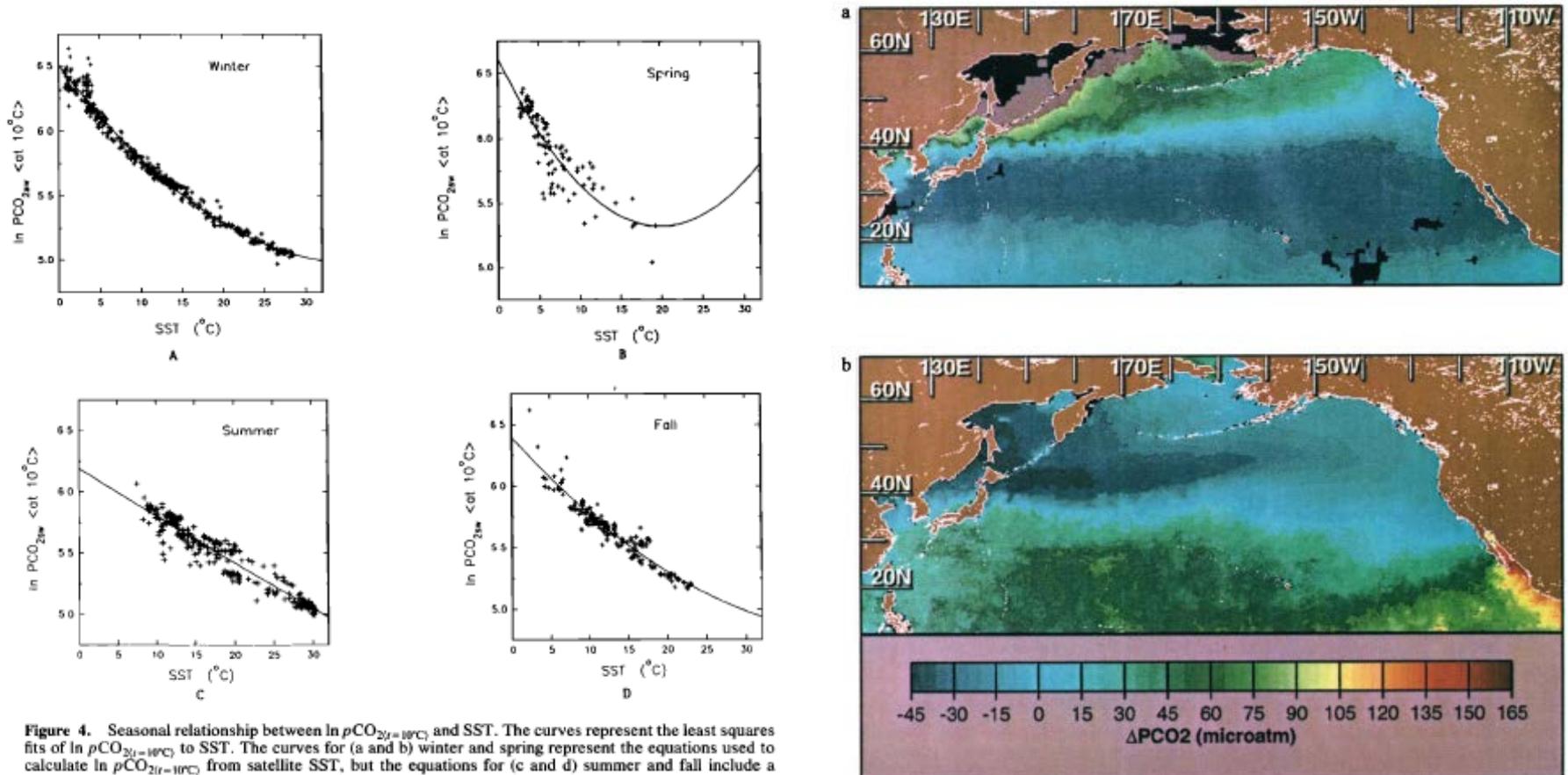


Figure 4. Seasonal relationship between $\ln p\text{CO}_{2(a=10^\circ\text{C})}$ and SST. The curves represent the least squares fits of $\ln p\text{CO}_{2(a=10^\circ\text{C})}$ to SST. The curves for (a and b) winter and spring represent the equations used to calculate $\ln p\text{CO}_{2(a=10^\circ\text{C})}$ from satellite SST, but the equations for (c and d) summer and fall include a longitude term, not included in these curves.

Remote sensing of $p\text{CO}_2$ -regression

Empirical algorithms



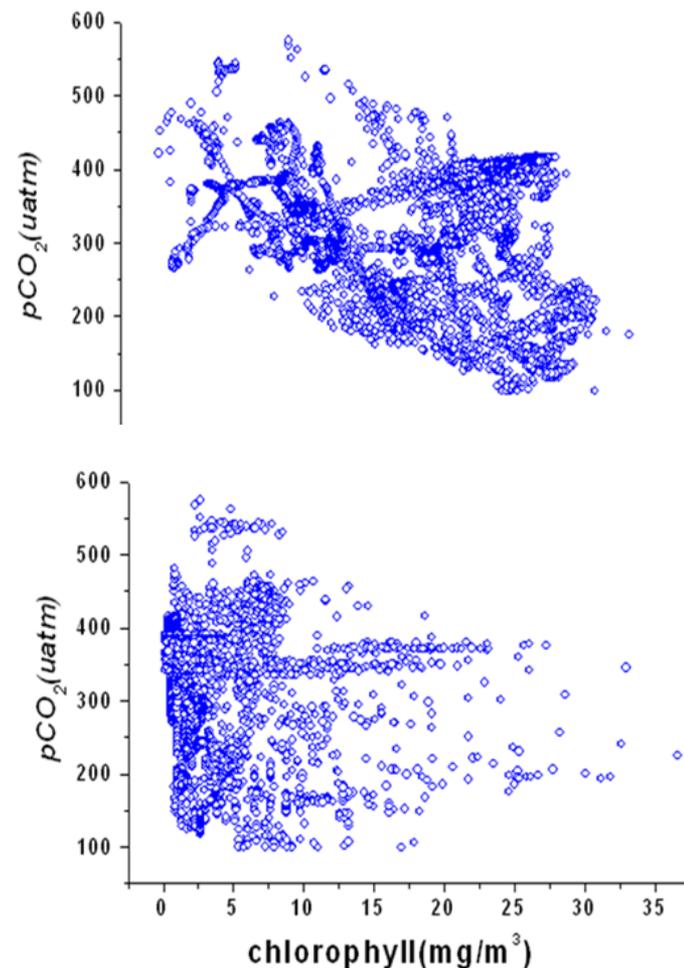
Aquatic $p\text{CO}_2$

Proxies: SST, Chla, Lon, Lat,
Salinity, Mixing Layer Depth, etc.

Estimation the Aquatic $p\text{CO}_2$ from Empirical algorithms (e.g. Linear Regression)

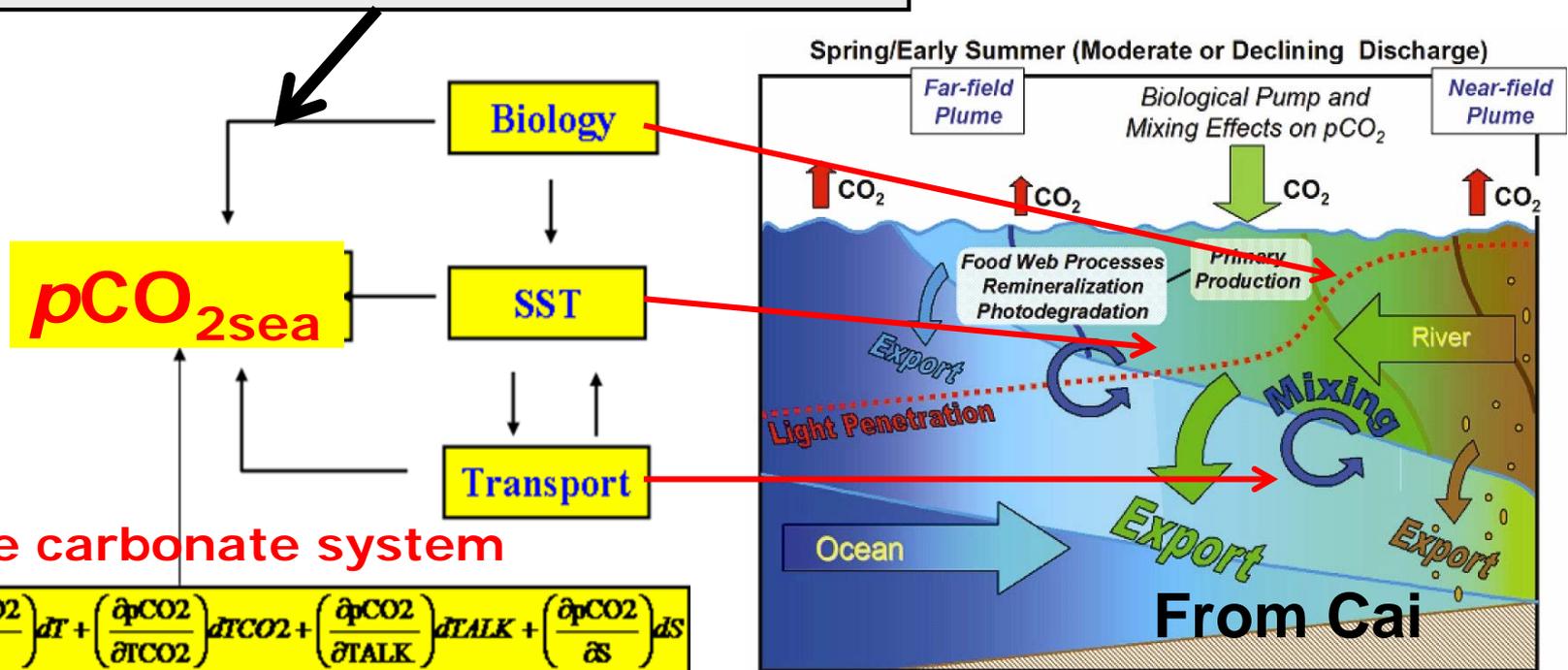
Proxy	Equation	Research area (References)
SST	$f(T)$	e.g. North Pacific (Stephens et al. 1995; Olsen et al., 2003, 2004;), Green land (Hood, et al., 1999;), Sargasso Sea (Nelson, et al., 2001), Caribbean Sea (Cosca et al., 2003;), Chile coastal (Lefevre et al., 2002), sub-Antarctic Ocean (Metzl et al., 1999;), North Atlantic (Lefèvre et al., 2004;)
Chlorophyll a	$f(T \text{ and/or } \text{Chla})$	e.g. North Pacific (Ono et al., 2004) Southern Ocean (Rangama, et al., 2005) Northern SCS (Zhu, et al., 2009)
Location (Lon, Lat)	$f(T, \text{Lon}, \text{Lat})$	e.g. Caribbean Sea (e.g. Wanninkhof, et al., 2007; Lueger, et al., 2009)
Mixing layer depth	$f(T, \text{MLD}, \text{Lon}, \text{Lat})$	e.g. North Atlantic (Lueger, et al., 2009)
CDOM	$f(T, \text{aCDOM})$	e.g. Hudson Bay (Else, et al., 2008),
Salinity	$f(S, \text{etc.})$	e.g. North Pacific (Sarma et al., 2006)
Neutral Network (T, S, Chlorophyll, ect.)		e.g. Northern SCS (Yan et al., 2011)
Principal Component analysis		e.g. Northern Gulf of Mexico (Lohrenz and Cai, 2006)
Satellite data with Model		e.g. Mediterranean (D'Ortenzio)

ECS: 2009 summer



Sea surface $p\text{CO}_{2\text{sea}}$ (Marginal Sea)

$$F = S \times K \times (p\text{CO}_{2\text{sea}} - p\text{CO}_{2\text{air}})$$



The controlling factors in the marginal sea system are very complicated, including **biological**, **thermodynamic (SST)**, and **mixing (transport)** effects, etc.

Remote sensing algorithm of Aquatic $p\text{CO}_2$

Sea-air flux of CO_2 in the North Pacific using shipboard and satellite data

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$p\text{CO}_2$ vs. SST

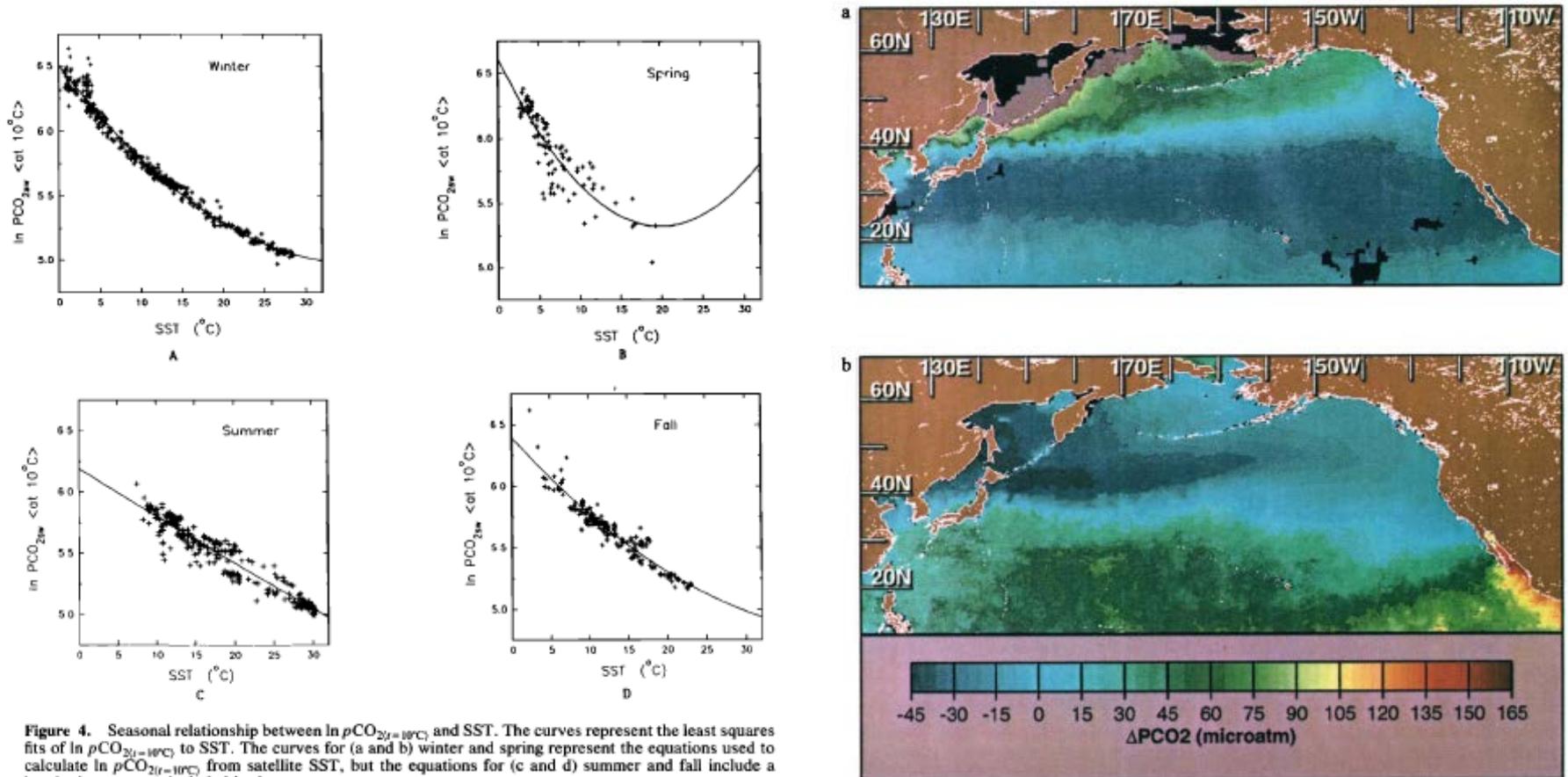
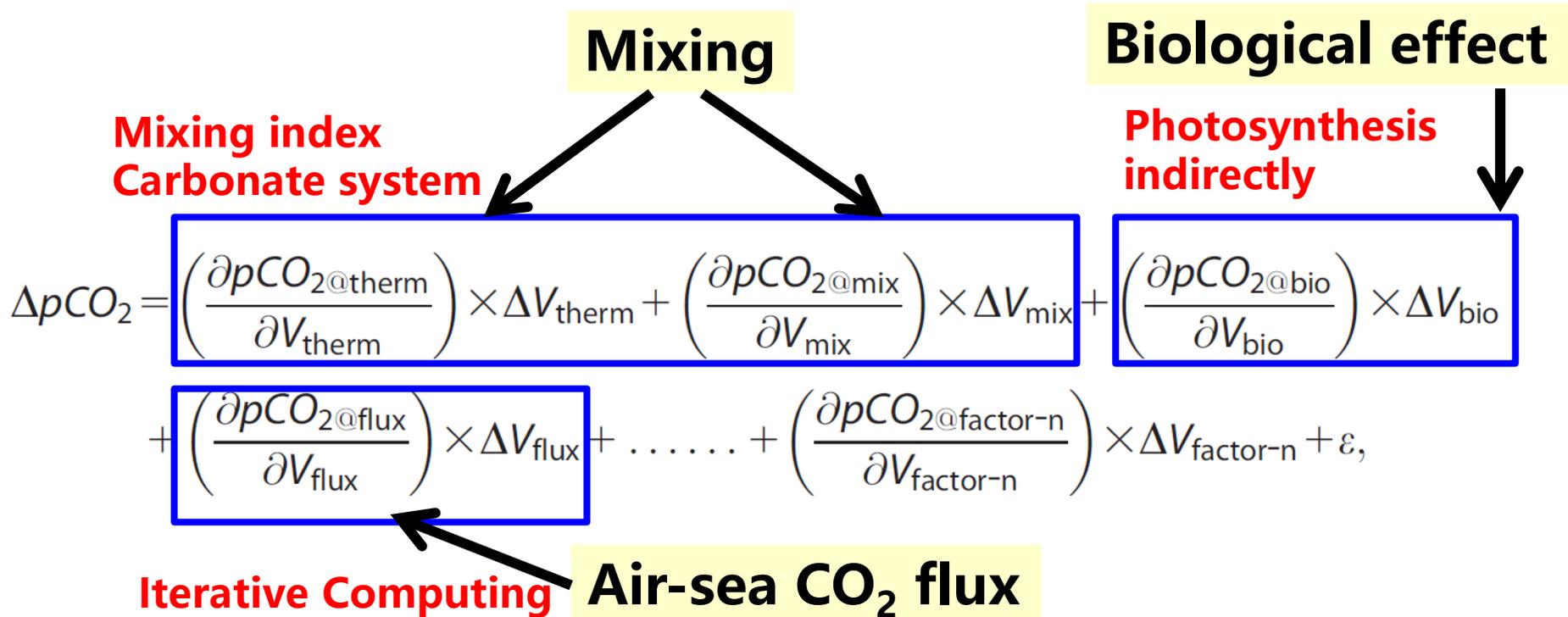


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Mechanistic-based semi-analytical algorithms (MeSAA- $p\text{CO}_2$)

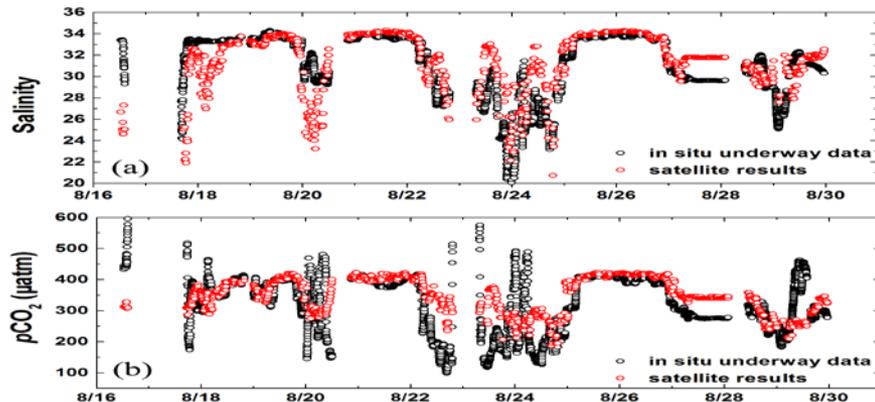
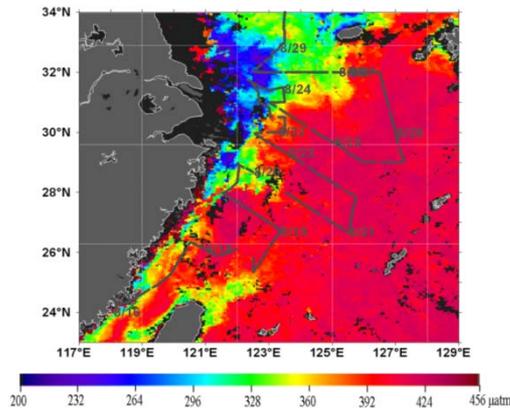


The variation of $p\text{CO}_2$ is analytically expressed as the sum of the first-order partial-difference of individual $p\text{CO}_2$ components contributed by each process or controlling factor. Hence, the critical issue is how to derive the analytical expressions of each process?

Mechanistic-based semi-analytical algorithms (MeSAA- $p\text{CO}_2$)

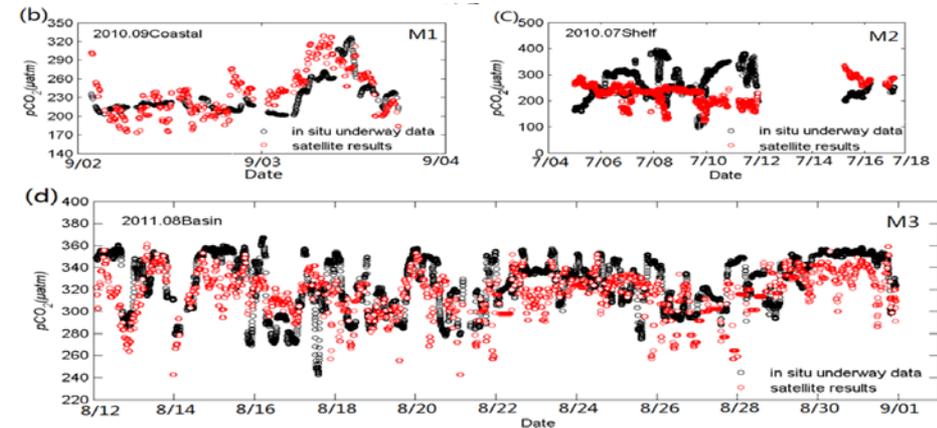
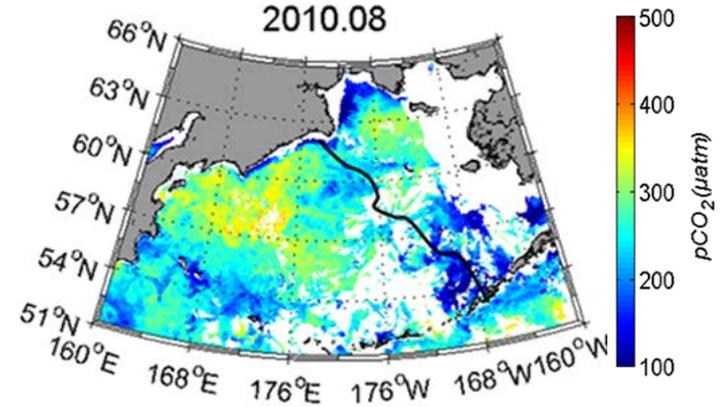
ECS (RioMar)

(Bai et al, 2013,2014 , 2015, JGR)



Bering Sea (OceMar)

(Song, Bai* et al, 2016, RS)



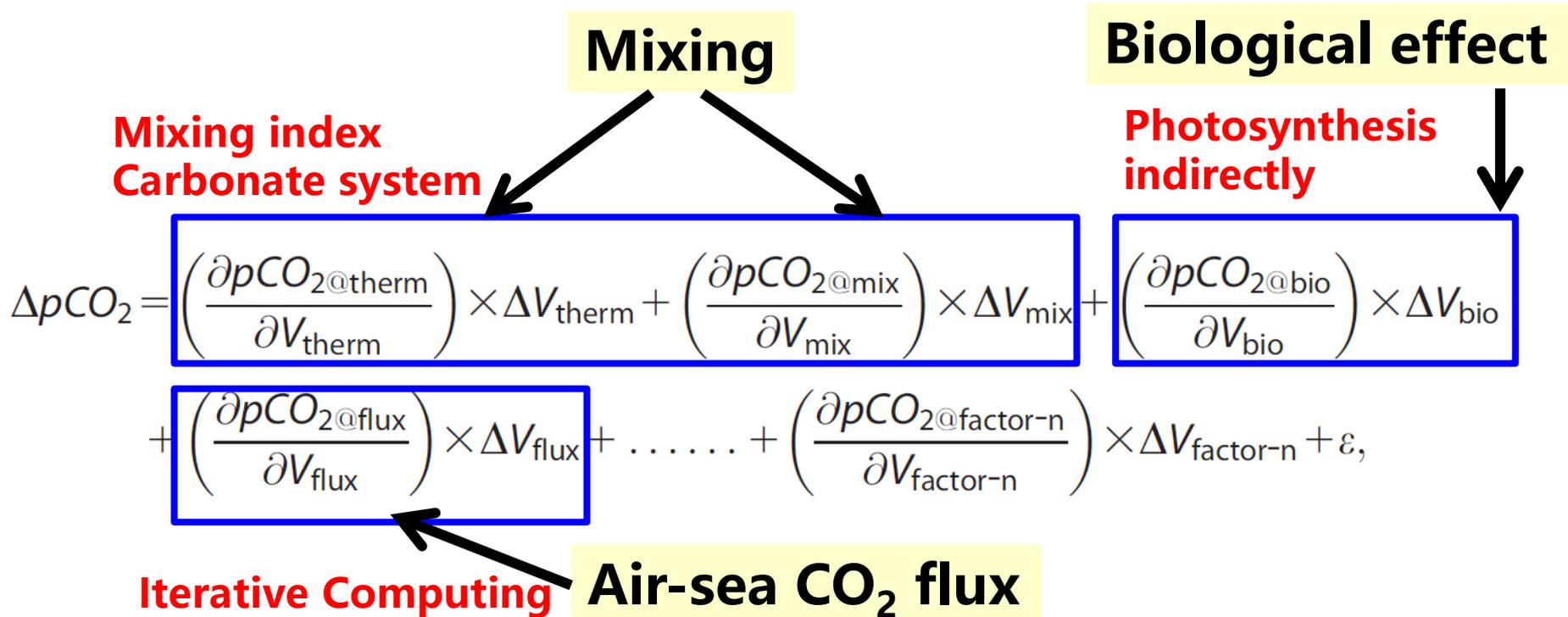
1) Prediction of thermodynamic control

- Here we **only consider the temperature-dependent component of the thermodynamic effects on sea surface $p\text{CO}_2$** . This fractional thermodynamic effect is known to have an effect of 4.23%/1° C [Takahashi *et al.*, 1993, 2002, 2009] as:

$$p\text{CO}_2 @ T_{est} = p\text{CO}_2 @ T_{obs} \times \exp[0.0423 \times (T_{est} - T_{obs})]$$

- In our method, we **combine** the variation of $p\text{CO}_2$ contributed by temperature change with that caused by the **water mass mixing process using the calculation of the carbonate system**; we do not put the thermodynamic effect as a separately item.

Mechanistic-based semi-analytical algorithms (MeSAA- $p\text{CO}_2$)



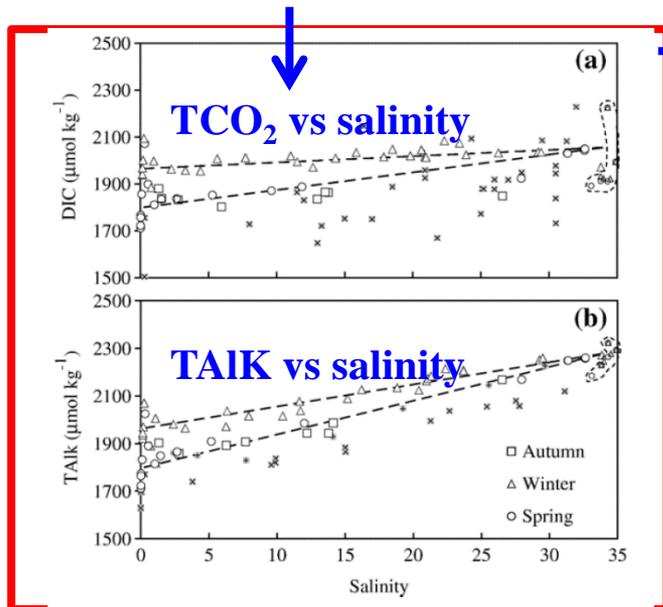
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Marine carbonate calculation

Parameterization of the two end-members mixing model (两端元混合作用)

$$dpCO_2 = \left(\frac{\partial pCO_2}{\partial T} \right) dT + \left(\frac{\partial pCO_2}{\partial TCO_2} \right) dTCO_2 + \left(\frac{\partial pCO_2}{\partial TALK} \right) dTALK + \left(\frac{\partial pCO_2}{\partial S} \right) dS$$

Fresh water (Changjiang) + Marine water (Kurishio) mixing



© The mixing and thermodynamically predicted mixing contributed pCO_2 can be calculated via marine carbonate system equilibrium equations.

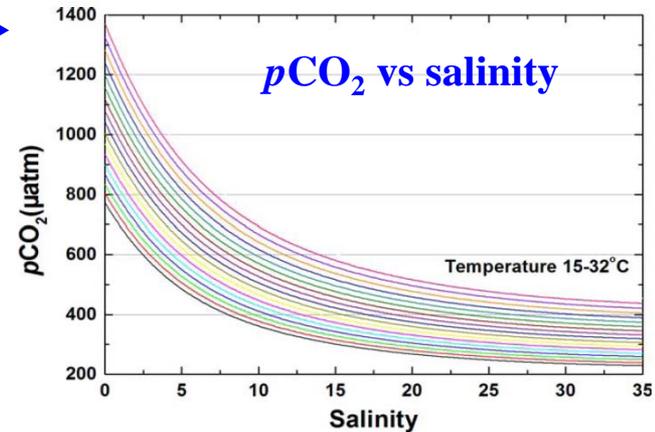


Figure 2. pCO_2 variation with salinity. Individual curves are calculated by CO2SYS with DIC and TA under the conservative mixing of two end-members (DIC = TA = 1850 $\mu\text{mol/kg}$ at salinity = 0; NDIC = 1970 $\mu\text{mol/kg}$ and NTA = 2300 $\mu\text{mol/kg}$ at salinity = 35) at given temperatures. Colored curves from bottom to top are the temperature setting from 15 to 32°C at 0.5°C step.

Conservative mixing behavior
Zhai et al., Marine Chemistry, 2007

Mixing index

Ⓢ Satellite-derived salinity by ocean color remote sensing in Changjiang River plume can be used as the mixing index in the ECS

BAI ET AL.: SATELLITE SALINITY OF CHANGJIANG PLUME

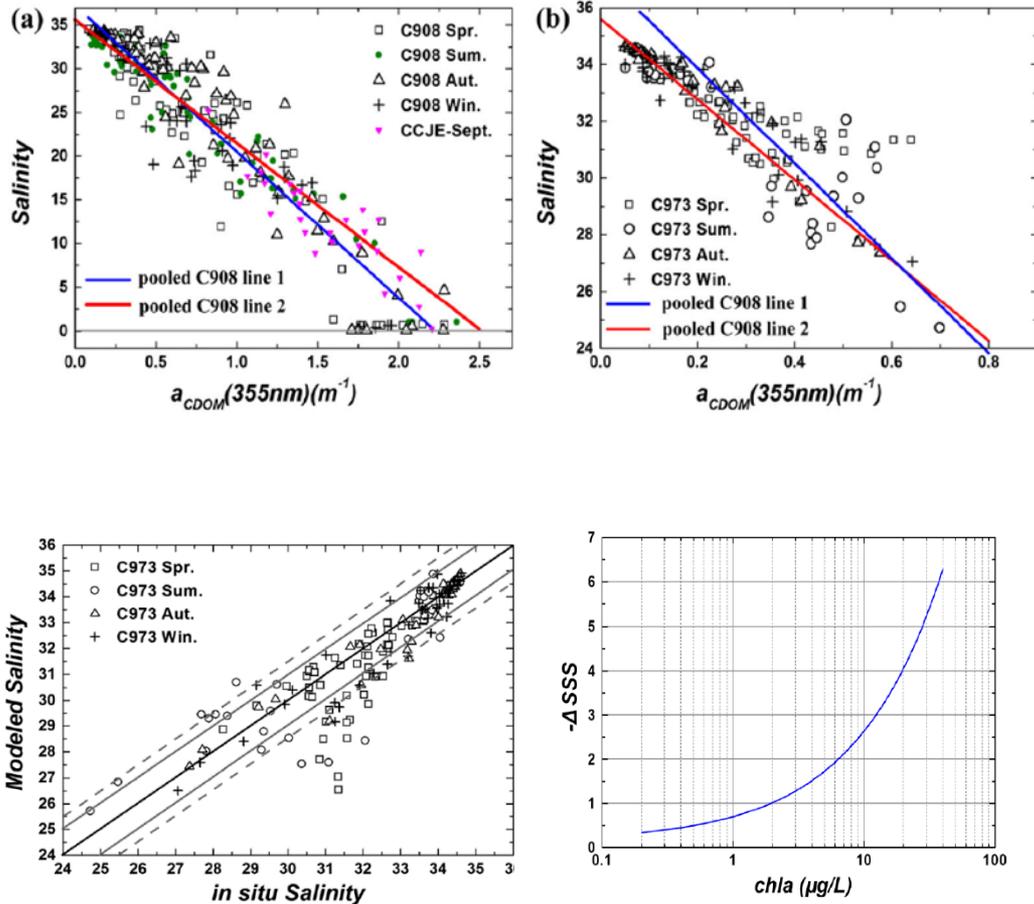
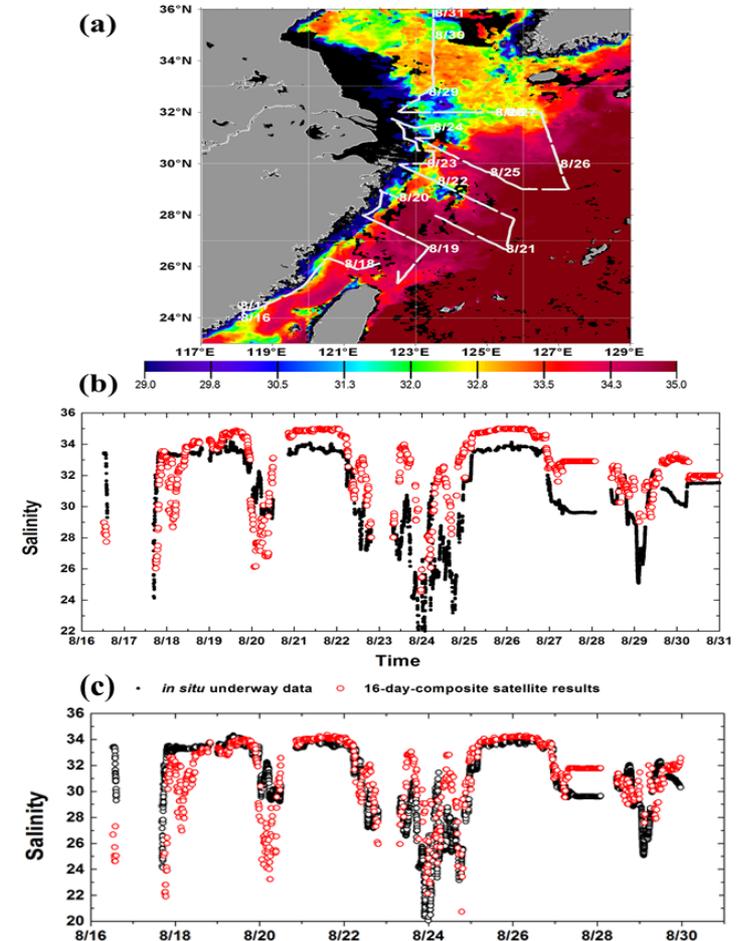


Figure 10. Underestimated salinities (ΔSSS) vary with the increase of $chl a$ concentration based on the bio-optical model of equation 10. $\Delta SSS < 1$ when $chl a \leq 2 \mu g/L$, $\Delta SSS < 1.5$ when $chl a \leq 4 \mu g/L$, and $\Delta SSS = 2.6$, when $chl a = 10 \mu g/L$.



Bio- Effect

@3) Parameterization of the biological effect

we assume that there is a general relationship between pCO_2 and $chla$,

ε represents the pCO_2 contribution from other factors, which does not overlap the contribution from the biological effects.

$$pCO_2 = A - B \times \log(chla) + \varepsilon$$

partial-difference of pCO_2 due to $chla$

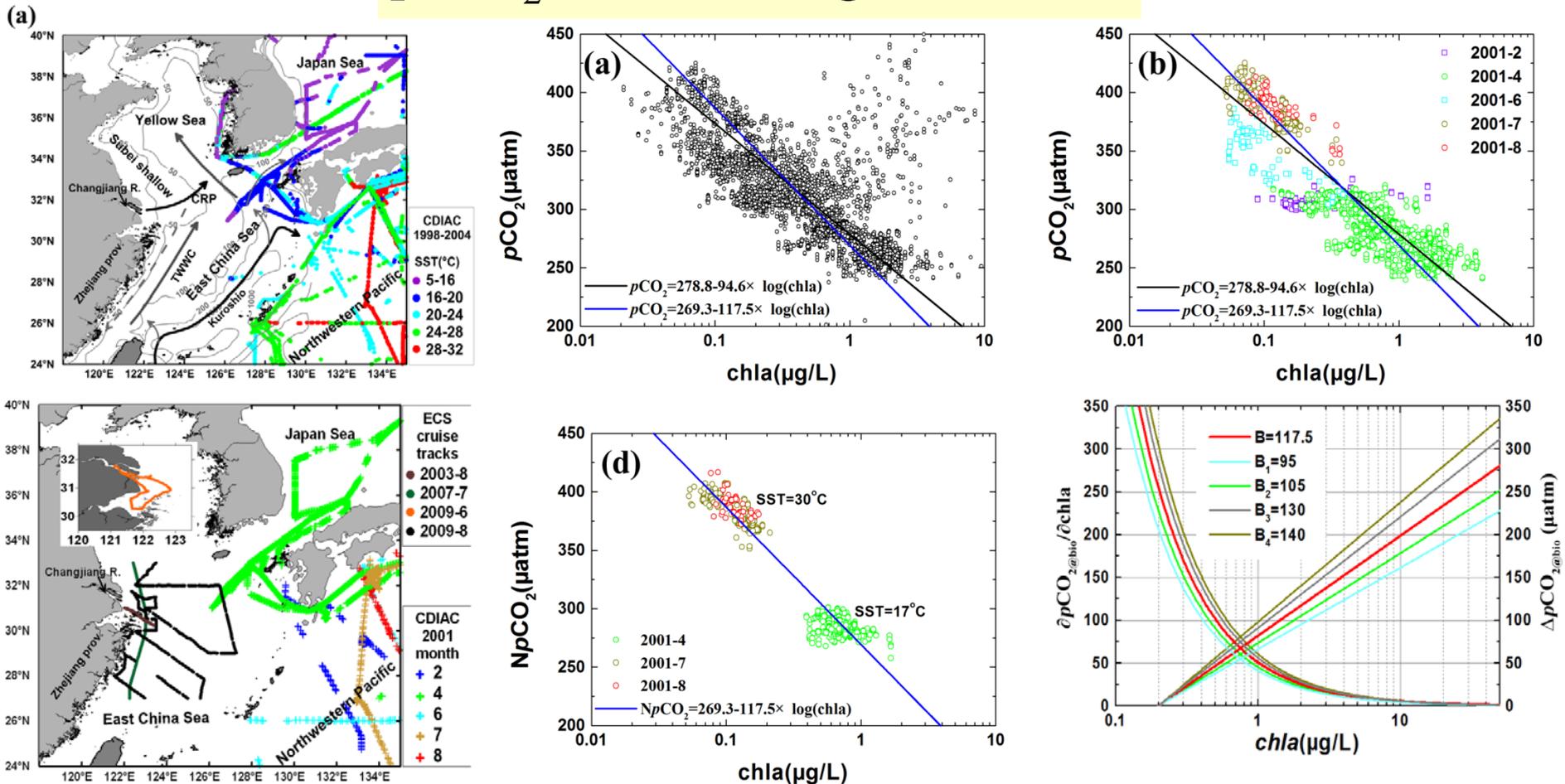
$$\frac{\partial(pCO_{2@bio})}{\partial(chla)} = \frac{\partial[A - B \times \log(chla) + \varepsilon]}{\partial(chla)} = -\frac{B}{\ln(10)} \frac{1}{chla}$$

an integration of the increment of $chla$ with time because the biological drawdown of pCO_2 is a cumulative process.

$$\begin{aligned} \Delta pCO_{2@bio} &= \frac{\partial pCO_{2@bio}}{\partial V_{bio}} \Delta V_{bio} = \int_{chla_0}^{chla_n} -\frac{B}{\ln(10)} \frac{1}{chla} d(chla) \\ &= -\frac{B}{\ln(10)} \times [\ln(chla_n) - \ln(chla_0)] \end{aligned}$$

3) Parameterization of the biological effect

$$p\text{CO}_2 = A - B \times \log(\text{chla}) + \varepsilon$$



underway $p\text{CO}_2$ from CDIAC and matching 8-day composite satellite-derived chla

we also set a $\pm 20\%$ deviation of B (95-140) to test the corresponding variation of $\Delta p\text{CO}_{2@bio}$

MeSAA algorithm in ECS in summertime

$$pCO_2 = pCO_{2@Hmix} + \Delta pCO_{2@bio}$$

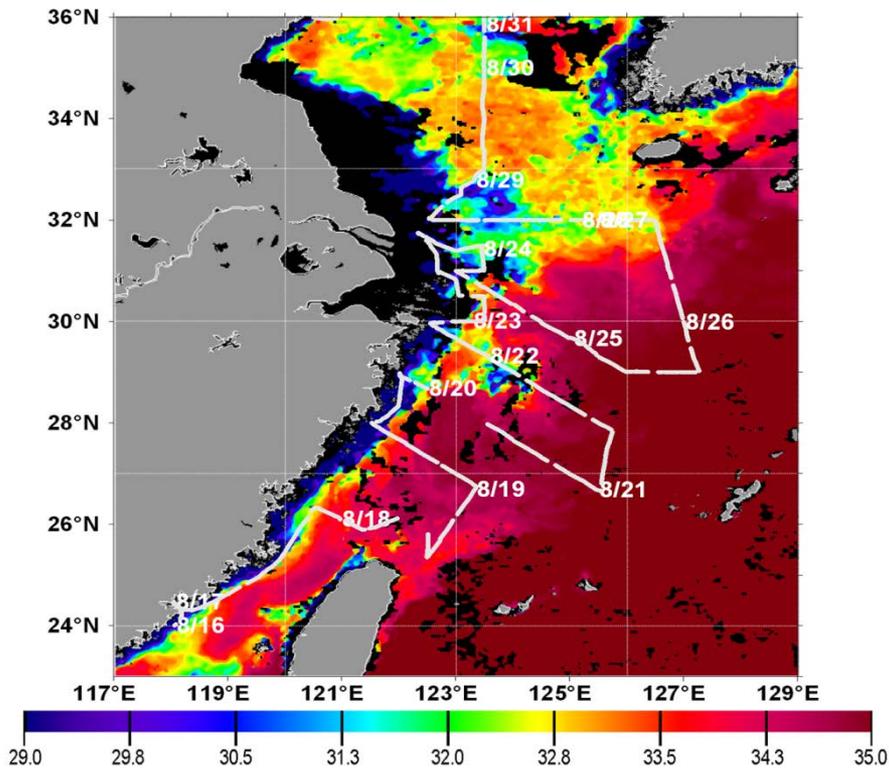
$$pCO_{2@Hmix} = \text{LUT}(TA_0, DIC_0, NTA_{35}, NDIC_{35}, SST, \text{Salinity})$$

$$\Delta pCO_{2@bio} = -117.5 \times [\log(chla) - \log(0.2)]$$

Satellite Result and validation

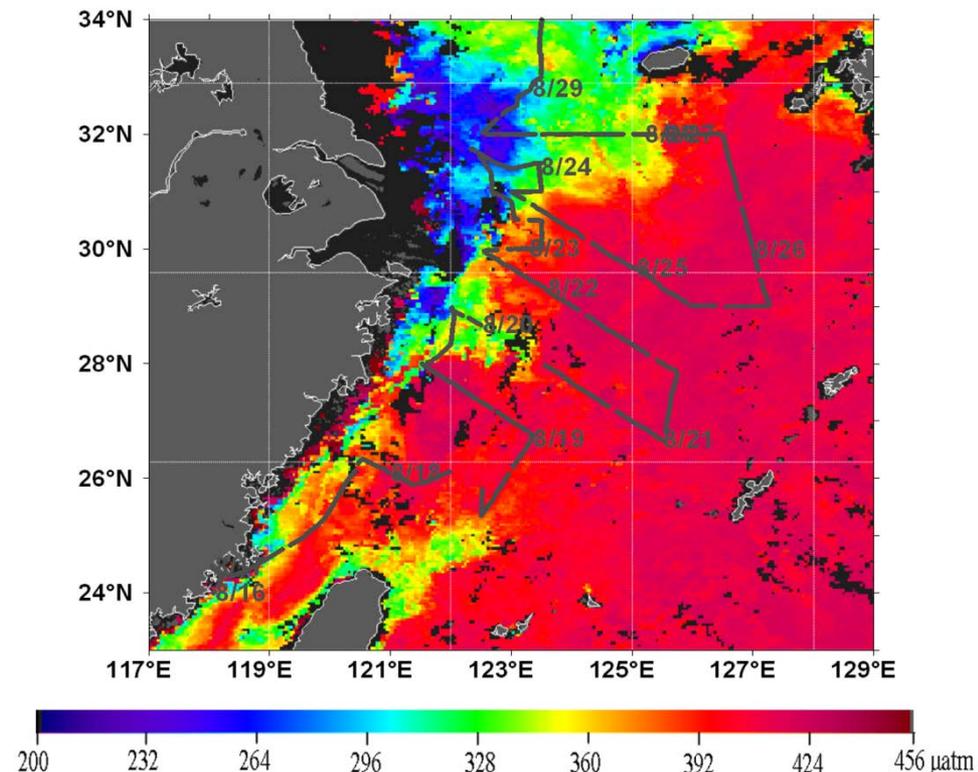
The inputs of MSAAs include satellite products of *chl_a*, *SST*, *salinity*, and *DIC* and *TA* values for two pairs of end-members.

Satellite *salinity* in August 2009



Bai, et al., 2013, JGR

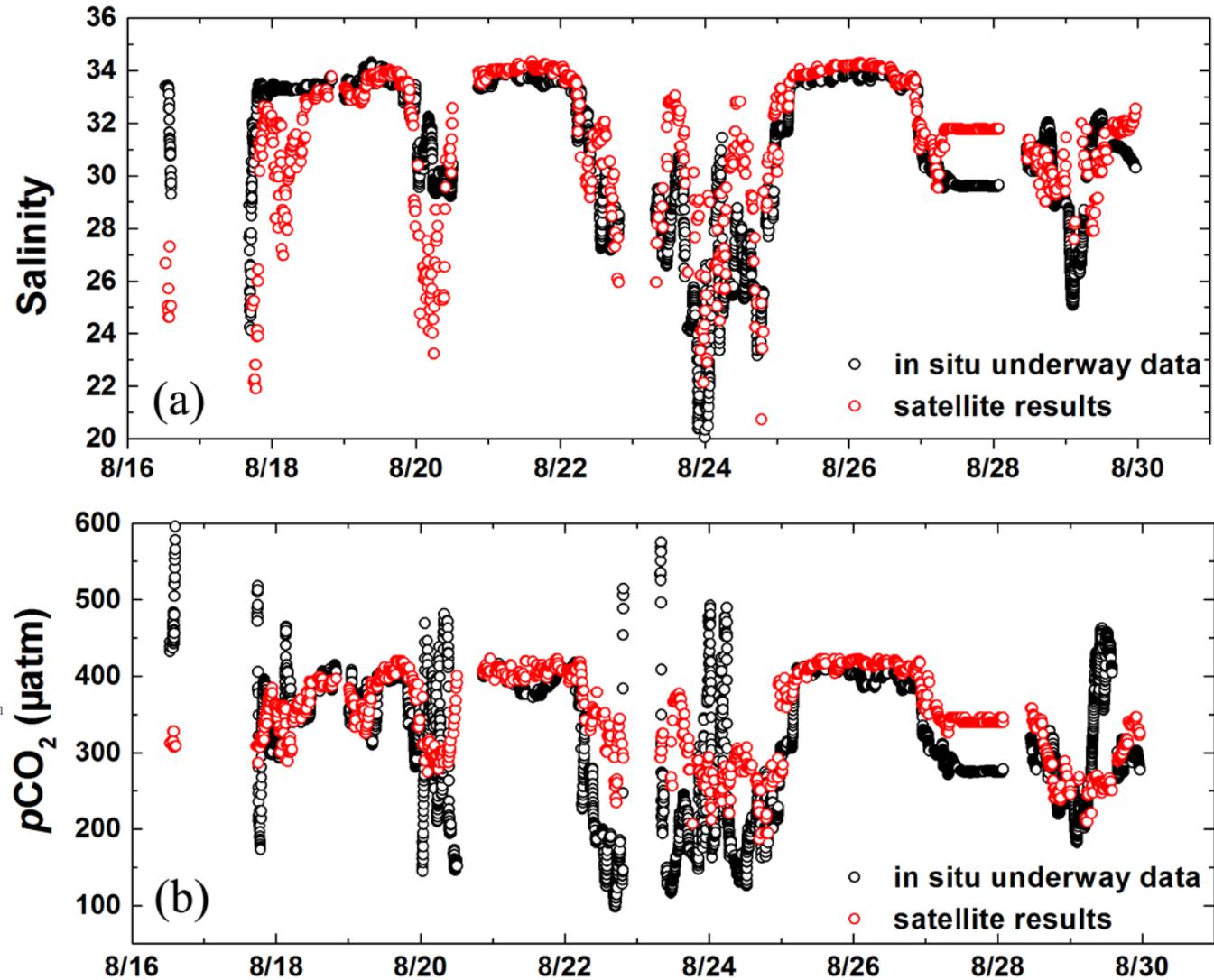
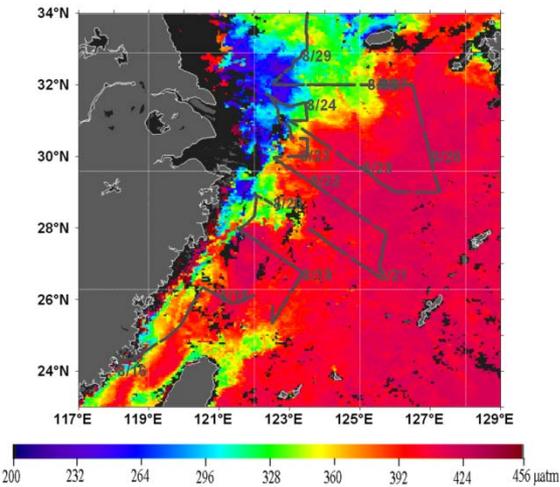
Satellite *pCO₂* in August 2009



Bai, et al., 2015, JGR

Satellite Result and validation

Satellite $p\text{CO}_2$
in August 2009
16 day composite
during the cruise



Lateral mixing

Vertical mixing

$$\Delta pCO_2 = \left(\frac{\partial pCO_{2@therm}}{\partial V_{therm}} \right) \Delta V_{therm} + \left(\frac{\partial pCO_{2@mix}}{\partial V_{mix}} \right) \Delta V_{mix} + \left(\frac{\partial pCO_{2@bio}}{\partial V_{bio}} \right) \Delta V_{bio} \\ + \left(\frac{\partial pCO_{2@flux}}{\partial V_{flux}} \right) \Delta V_{flux} + \dots + \left(\frac{\partial pCO_{2@factor-n}}{\partial V_{factor-n}} \right) \Delta V_{factor-n} + \varepsilon$$

Semi-Analytical pCO_{2sw} Algorithms

All seasons

Particle optics can denote the mixing state of water column both at vertical and horizontal directions.

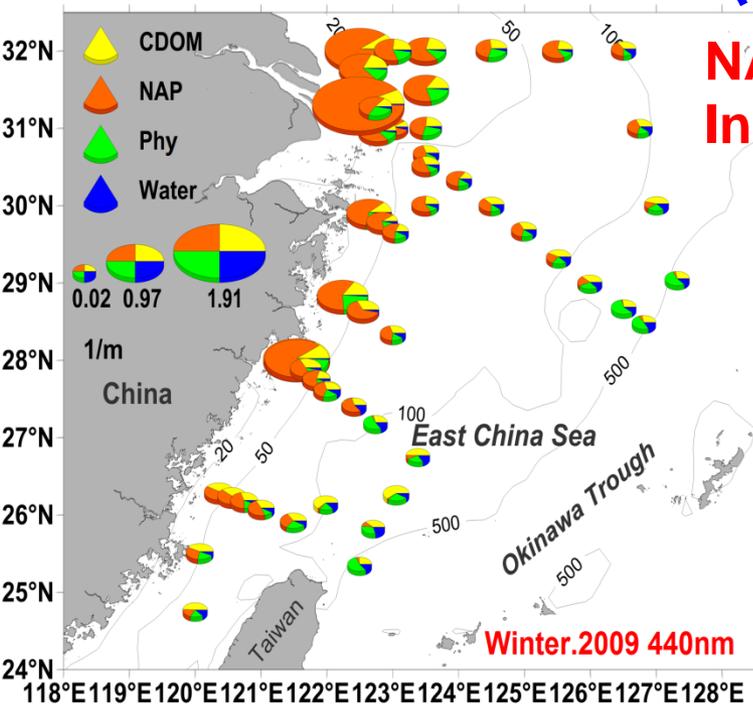
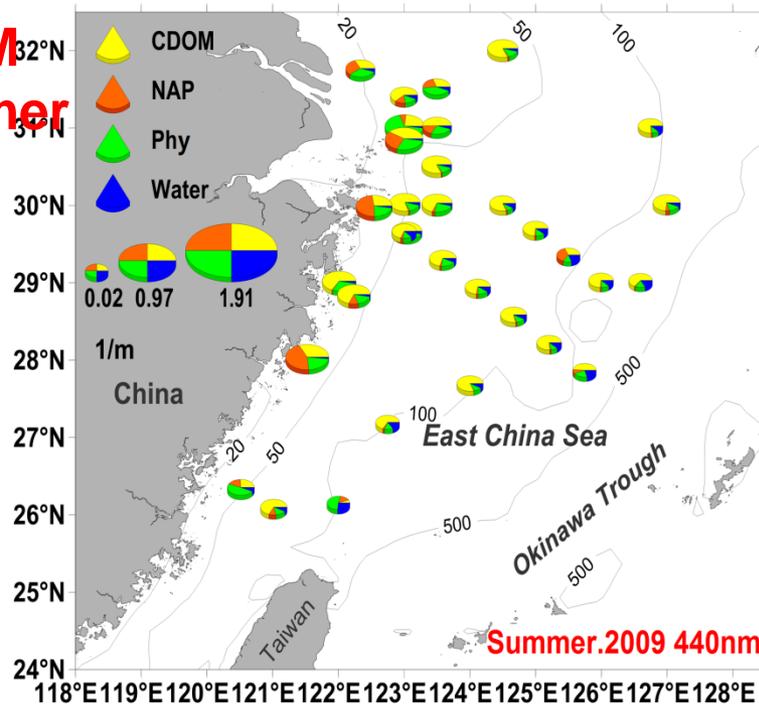
$$\Delta pCO_2 = \left(\frac{\partial pCO_{2@therm}}{\partial V_{therm}} \right) \Delta V_{therm} + \left(\frac{\partial pCO_{2@mix}}{\partial V_{mix}} \right) \Delta V_{mix} + \left(\frac{\partial pCO_{2@bio}}{\partial V_{bio}} \right) \Delta V_{bio} + \left(\frac{\partial pCO_{2@flux}}{\partial V_{flux}} \right) \Delta V_{flux} + \dots + \left(\frac{\partial pCO_{2@factor-n}}{\partial V_{factor-n}} \right) \Delta V_{factor-n} + \epsilon$$

Proxy Salinity



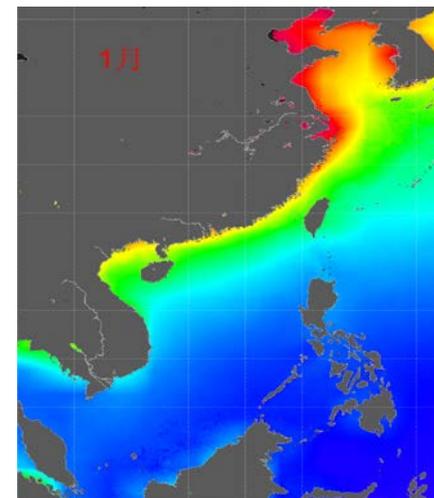
CDOM
Summer

NAP
In Winter

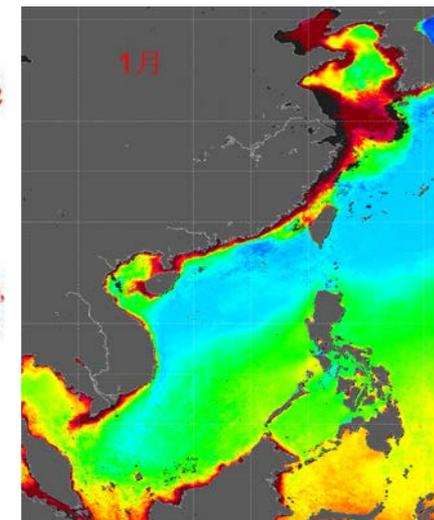


Validation- Aquatic $p\text{CO}_2$

大气 CO_2 分压

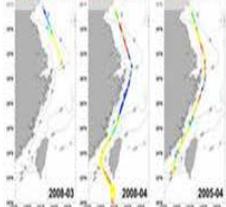


海水 CO_2 分压

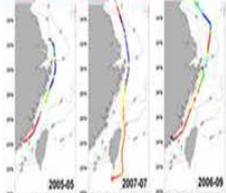


沿岸航次

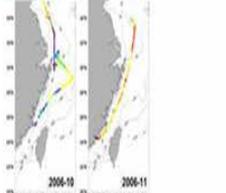
N=2336, RE=37.2
N=7603, RE=30.2
N=2821, RE=39.7



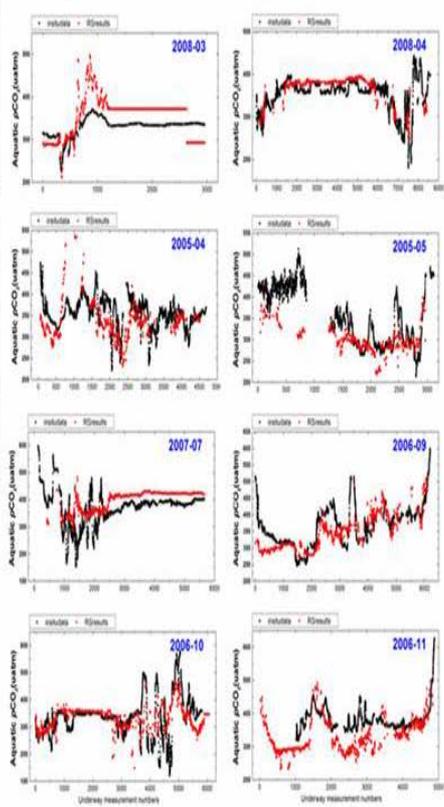
N=1868, RE=43.6
N=4551, RE=47.8
N=3689, RE=35.4



N=4551, RE=46.9
N=2942, RE=47.3

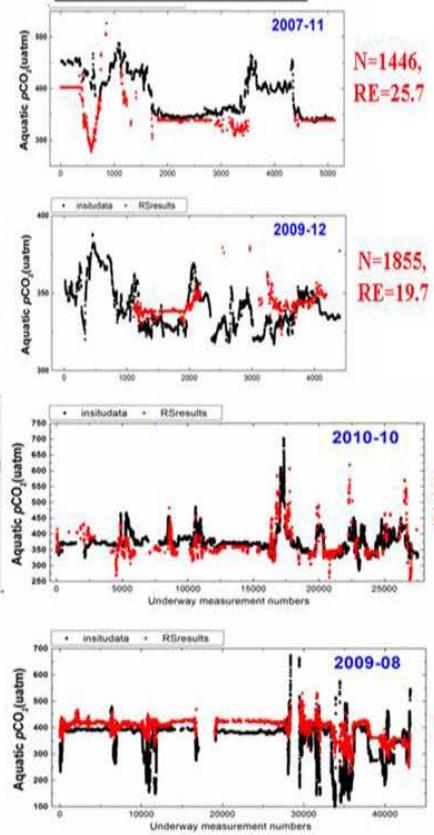
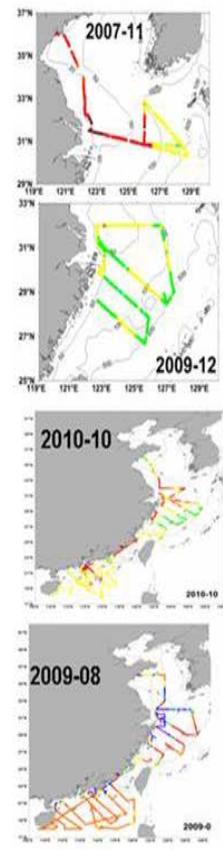


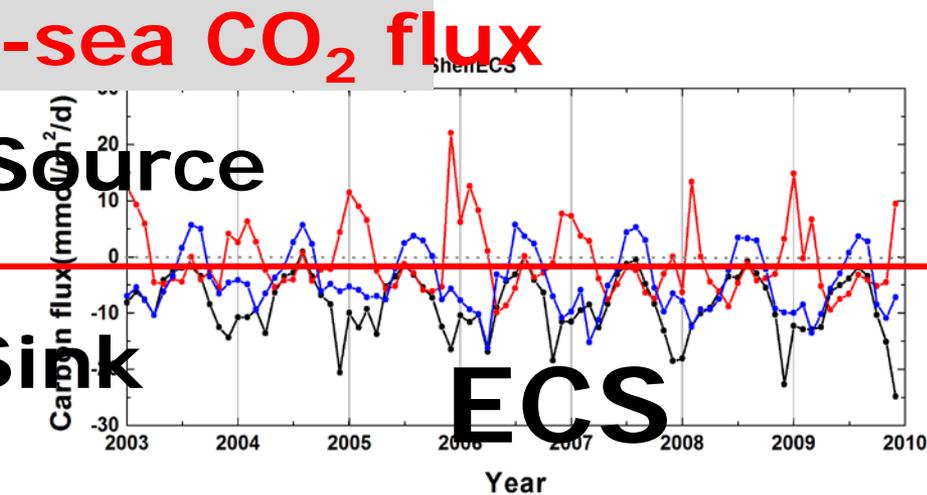
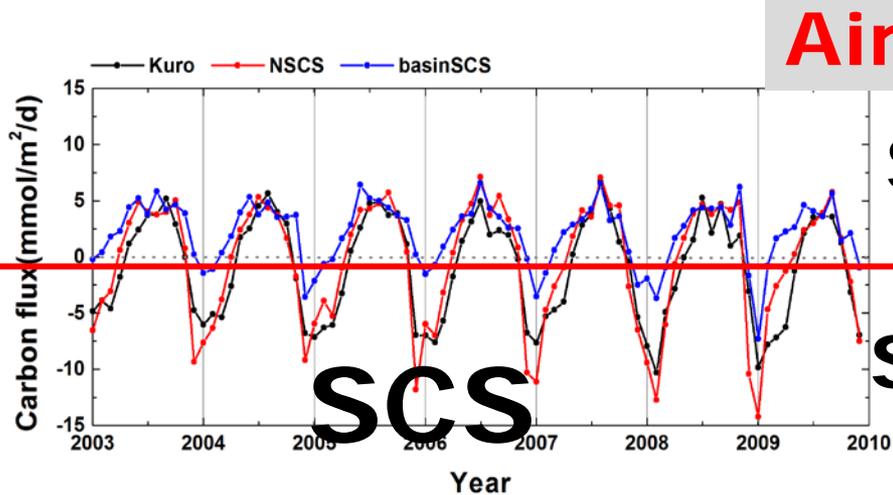
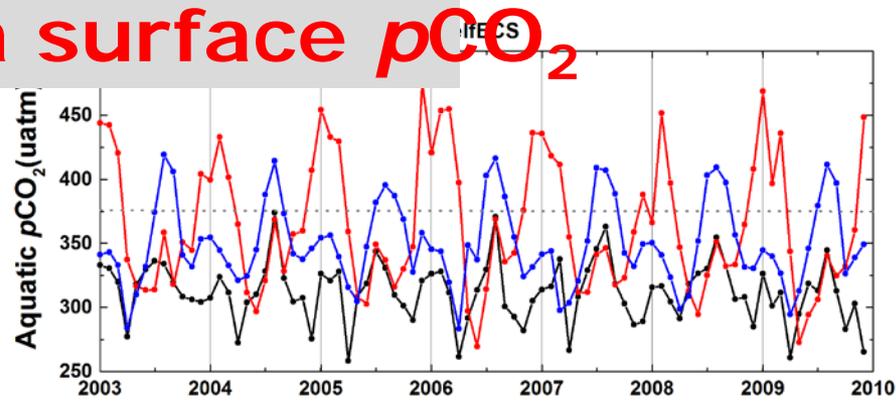
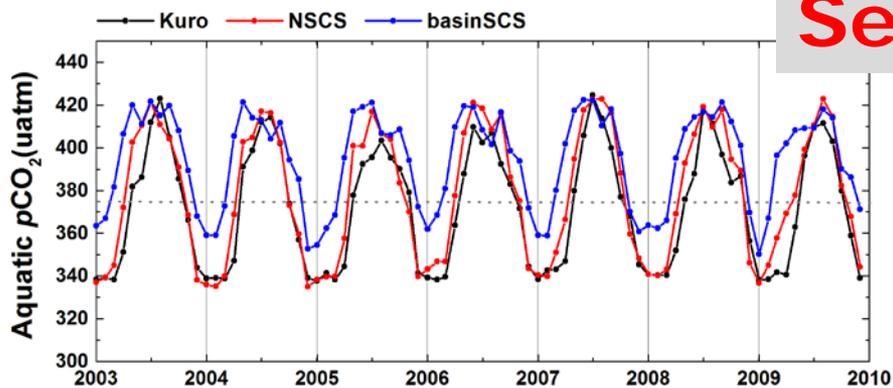
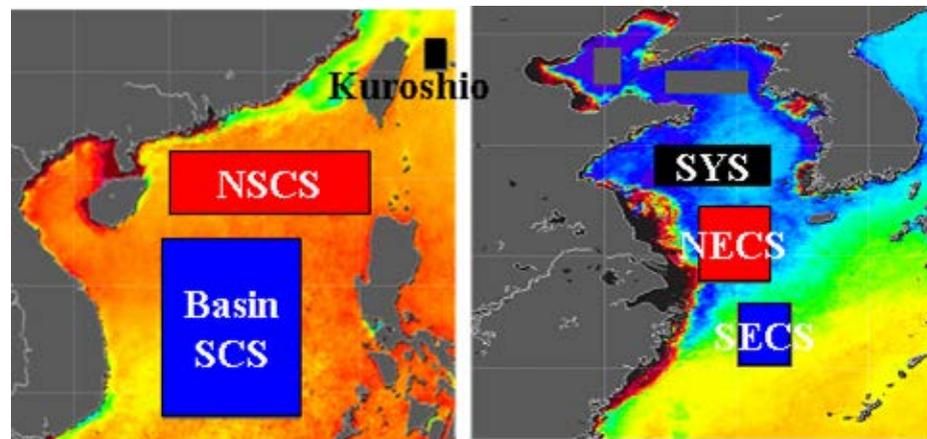
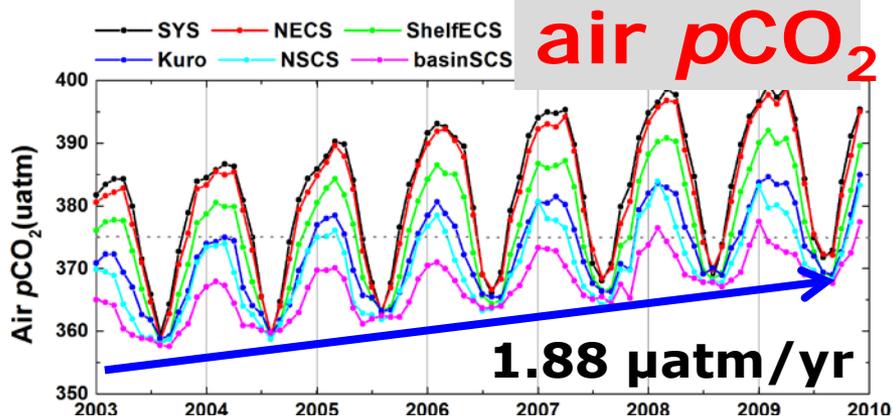
• insitadata • RSresults



陆架航次

• insitadata • RSresults

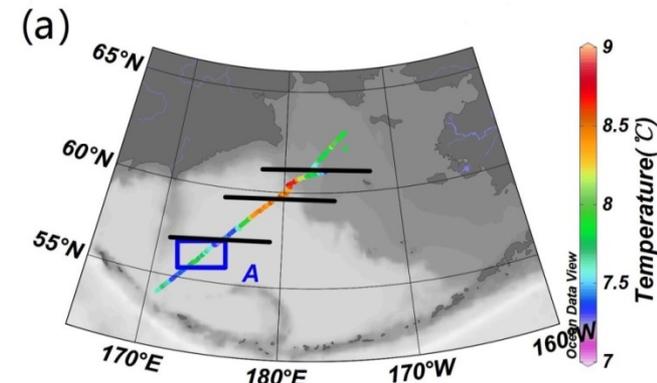
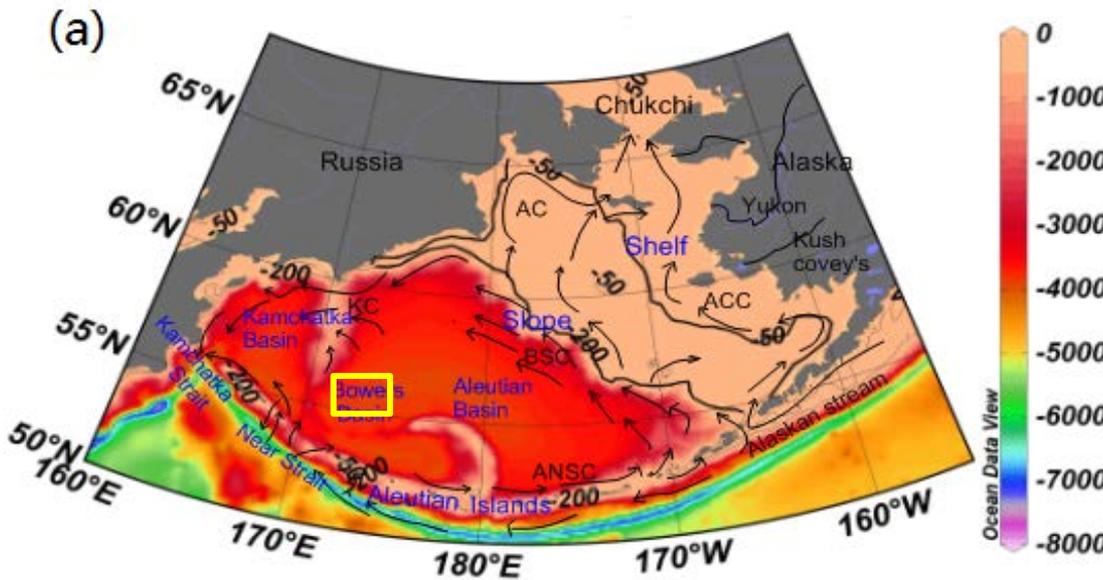




Bering Sea

Selection of reference water-mass

$$pCO_2 = pCO_{2(o)} + \left(\frac{\partial pCO_{2@therm}}{\partial V_{therm}} \right) \Delta V_{therm} + \left(\frac{\partial pCO_{2@bio}}{\partial V_{bio}} \right) \Delta V_{bio} + \varepsilon$$



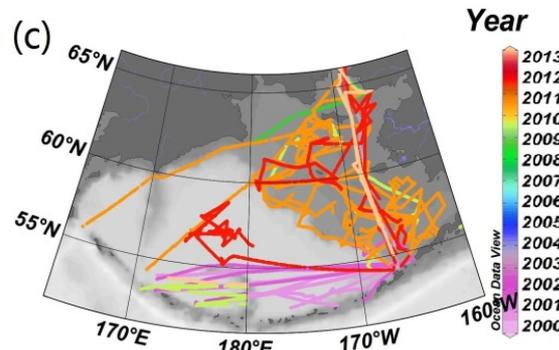
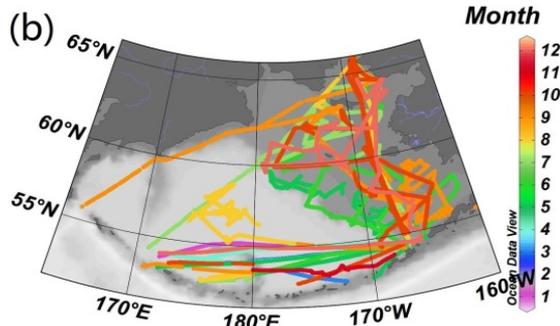
July 2010 cruise

Selected $pCO_{2(o)}$

$$= 381.8 \pm 5.08 \text{ } \mu\text{atm}$$

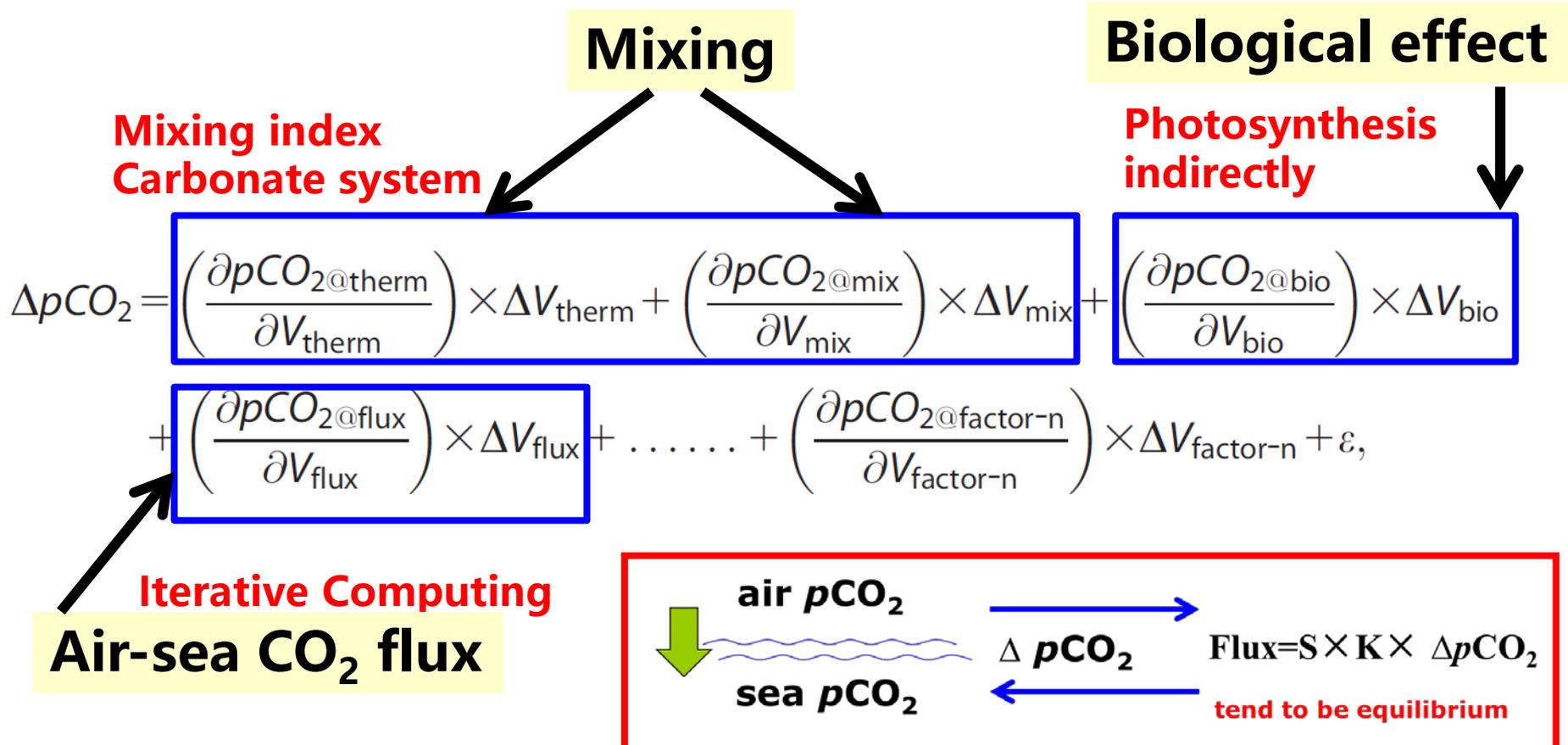
Calculated $pCO_{2(\text{summer})}$

$$= 385.9 \text{ } \mu\text{atm}$$

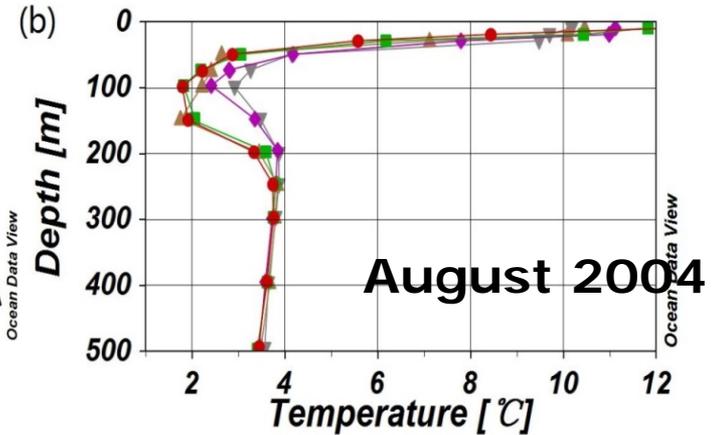
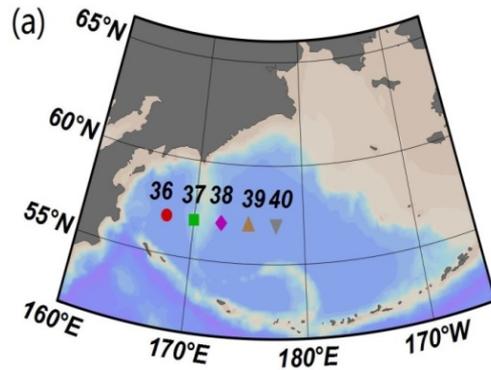
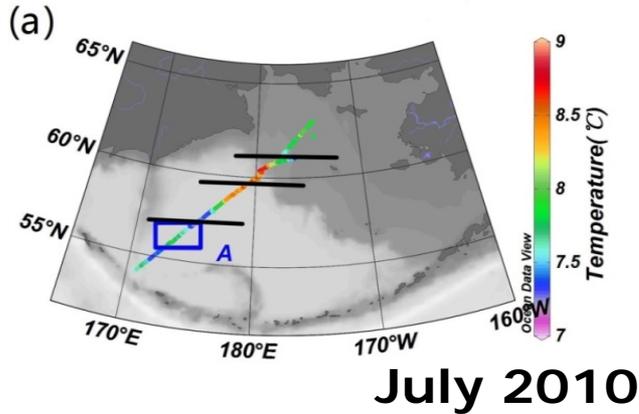


Song, Bai* et al., RS, 2016

Mechanistic-based semi-analytical algorithms (MeSAA- $p\text{CO}_2$)



© pCO₂ change from last winter to summer



385.9 μatm

381.8 μatm

Calculated pCO_{2(summer)} = Selected pCO_{2(o)}

1) biological alteration ($\Delta pCO_{2(bio)}$) = - 65 μatm

2) Warming + air-sea exchange ($\Delta pCO_{2(air-sea)}$) = 35.8 μatm

Redfield Ratio
($\Delta DIC_{bio} = 106 * AOU / 138$)



well-mixed winter water
bottom high DIC data

415.2 μatm
pCO_{2(winter)}

DIC, TA
in winter

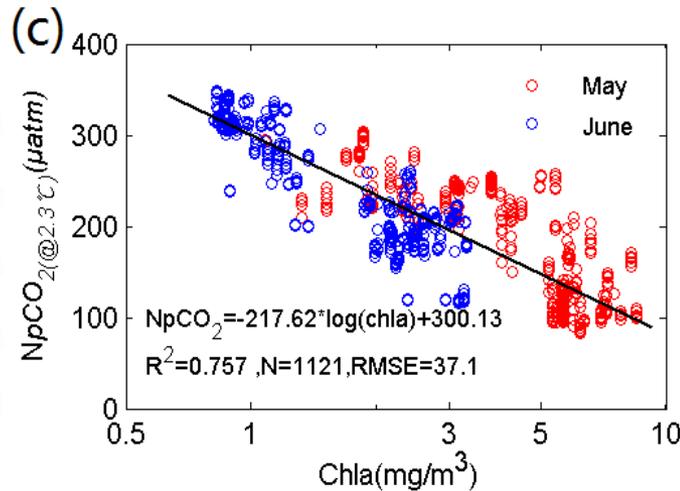
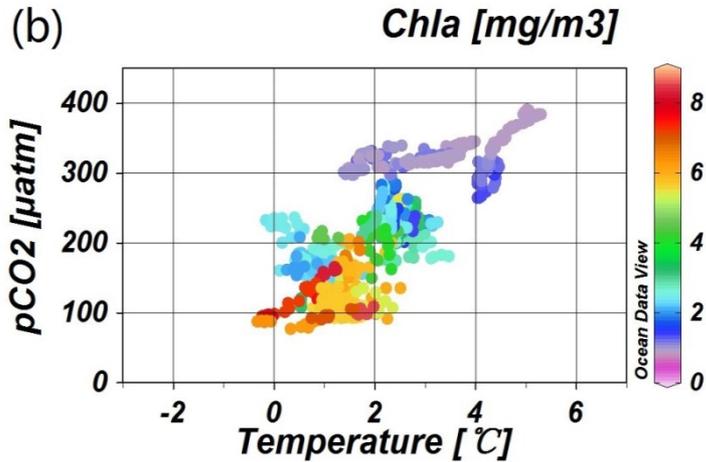
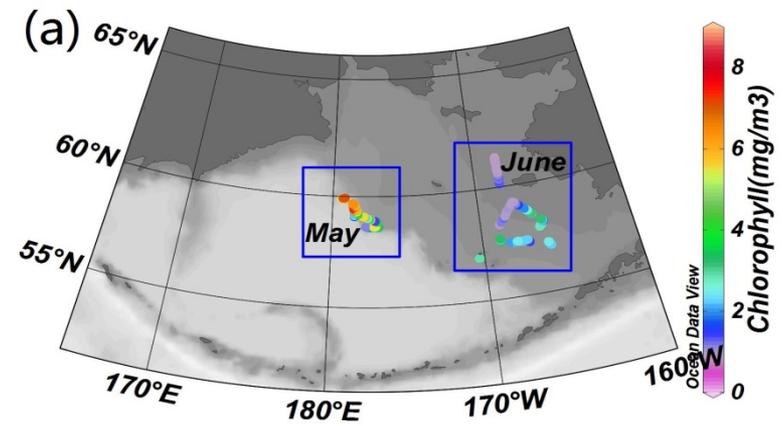
Surface pCO_{2(summer)}

The climatological monthly average **mixing layer depth** was **50.7 m** during March to July

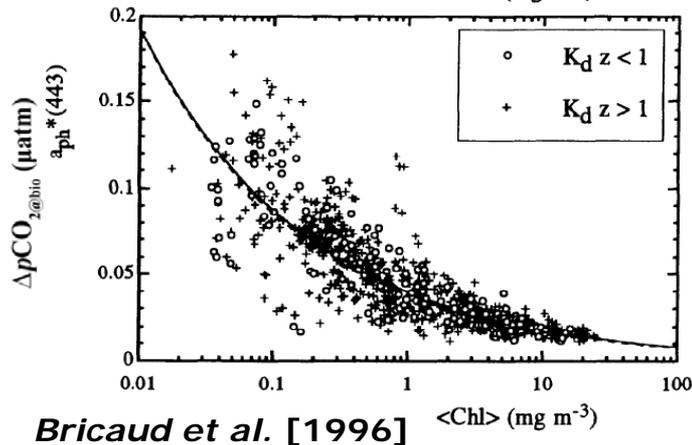
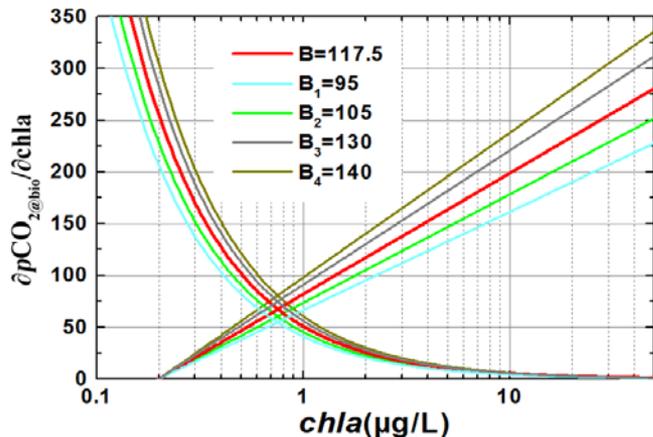
$$p\text{CO}_2 = A - B \times \log(\text{chla}) + \varepsilon$$

BS: B=217.62

ECS: B=117.5

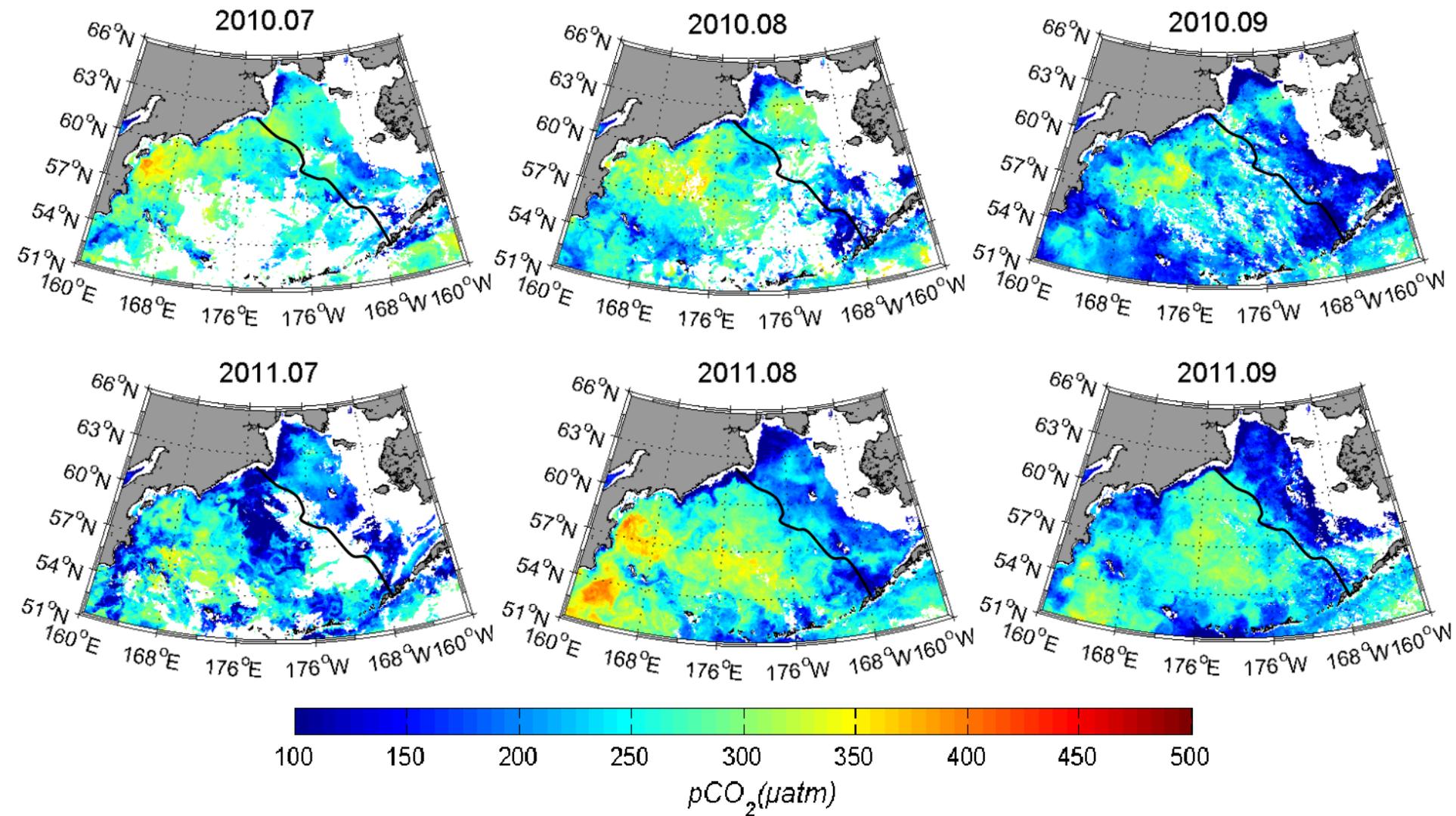


**low *chla* regime
small size
large
 $\partial p\text{CO}_2 / \partial \text{chla}$**



***chla*-specific
absorption
coefficients also
showed an
exponential decay
relationship with
log(*chla*) due to
the package
effect**

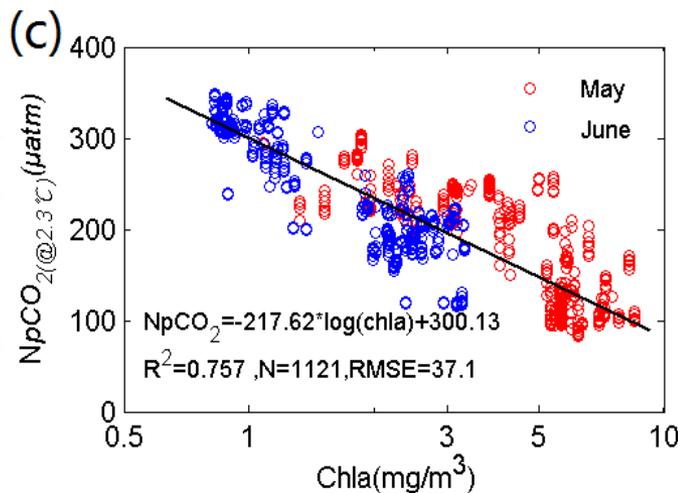
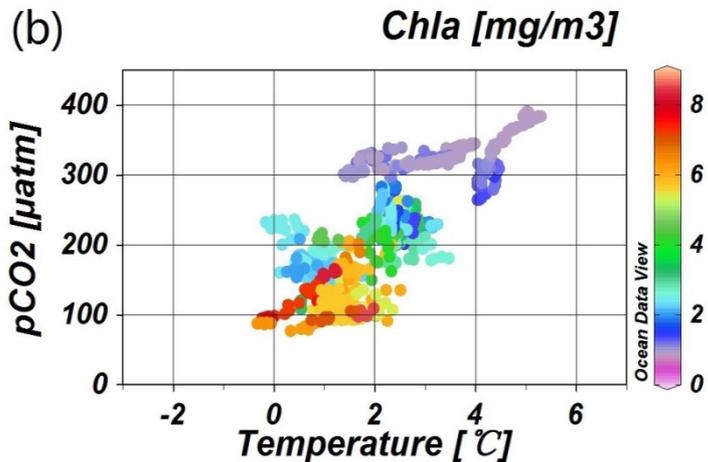
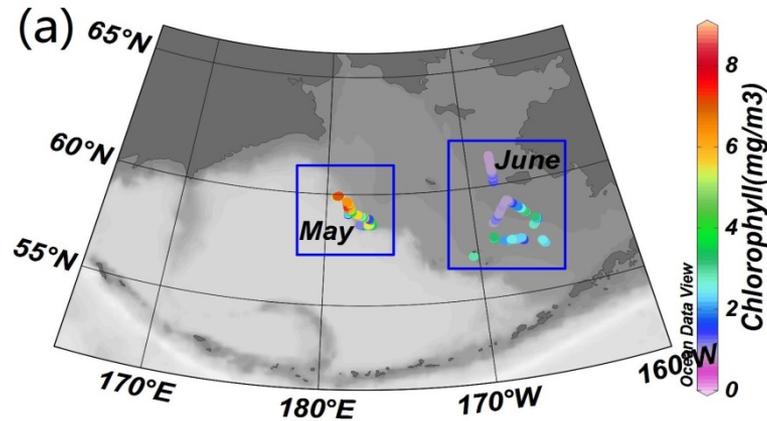
Satellite-derived $p\text{CO}_2$ in summer



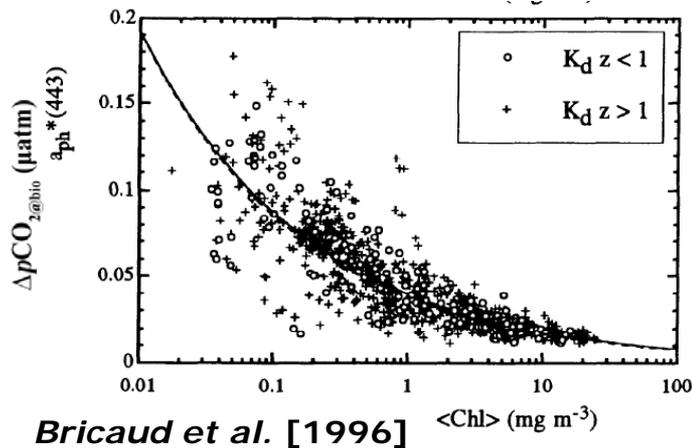
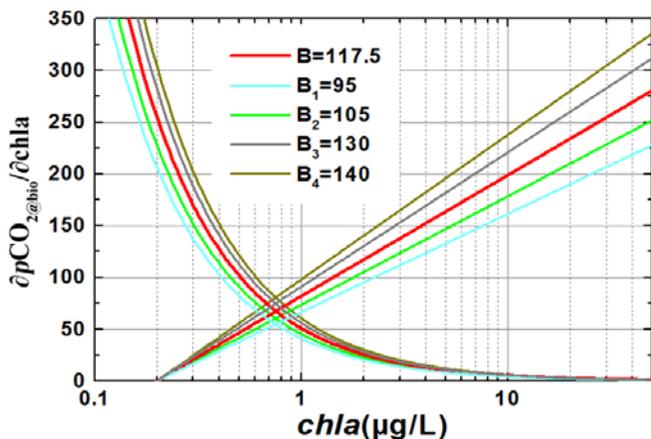
$$p\text{CO}_2 = A - B \times \log(\text{chl}a) + \varepsilon$$

BS: B=217.62

ECS: B=117.5

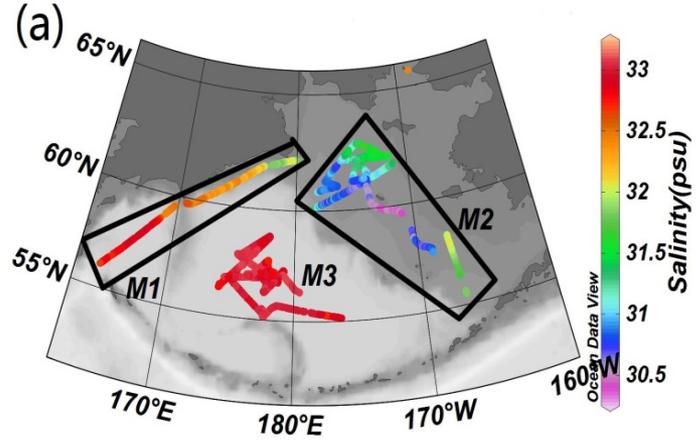


**low *chl*a regime
small size
large
 $\partial p\text{CO}_2_{2@bio} / \partial \text{chl}a$**

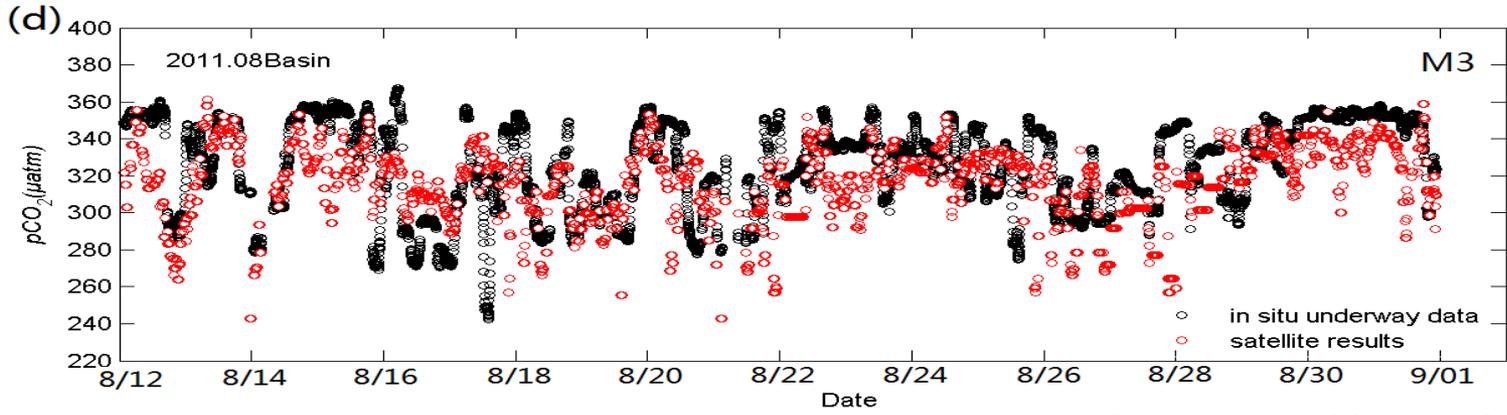
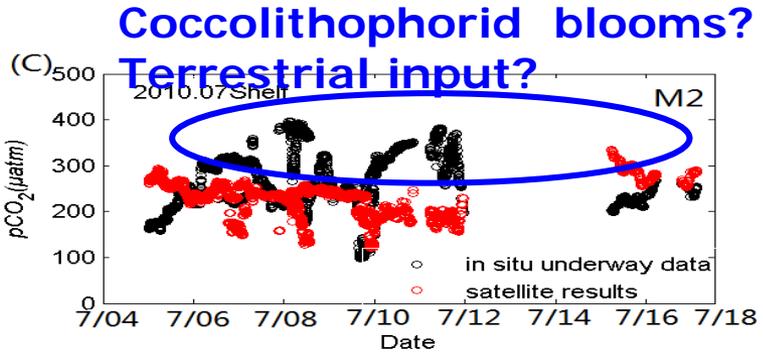
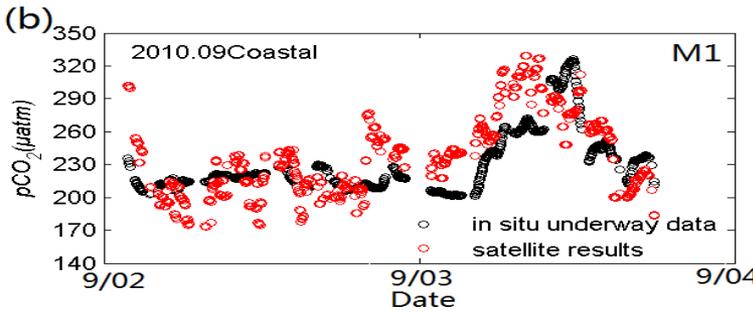


***chl*a-specific
absorption
coefficients also
showed an
exponential decay
relationship with
log(*chl*a) due to
the package
effect**

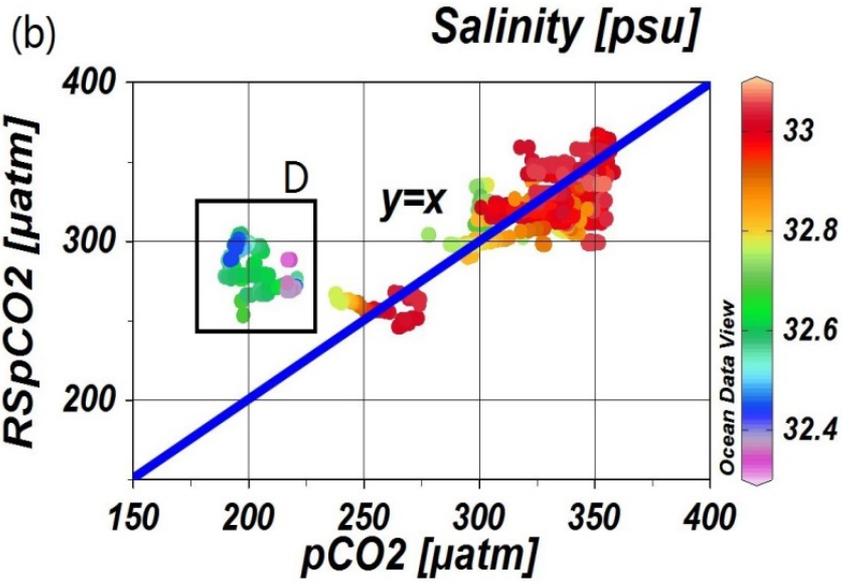
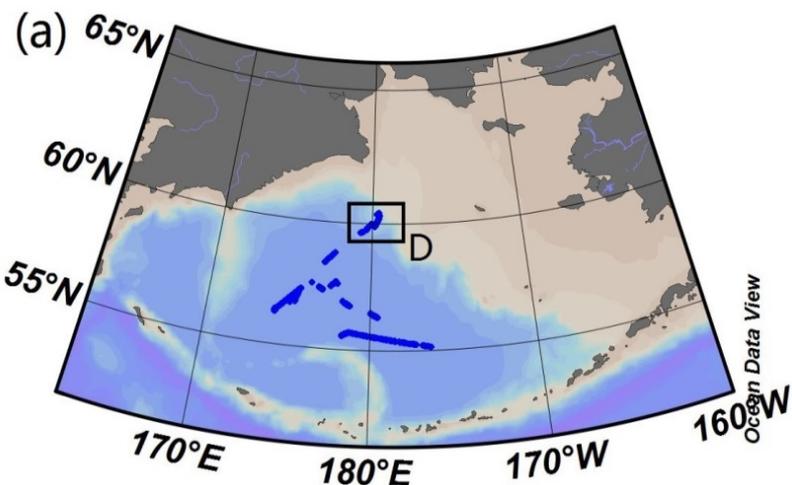
Validation (1)-monthly



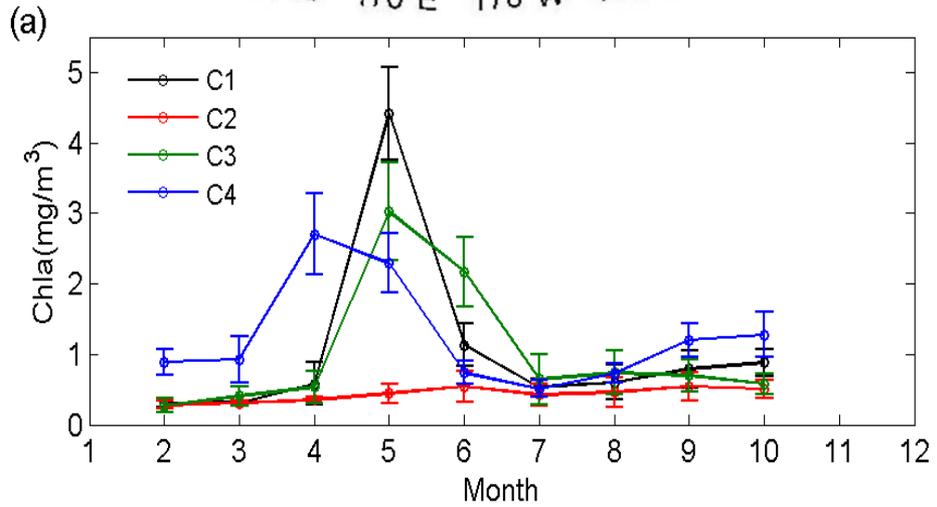
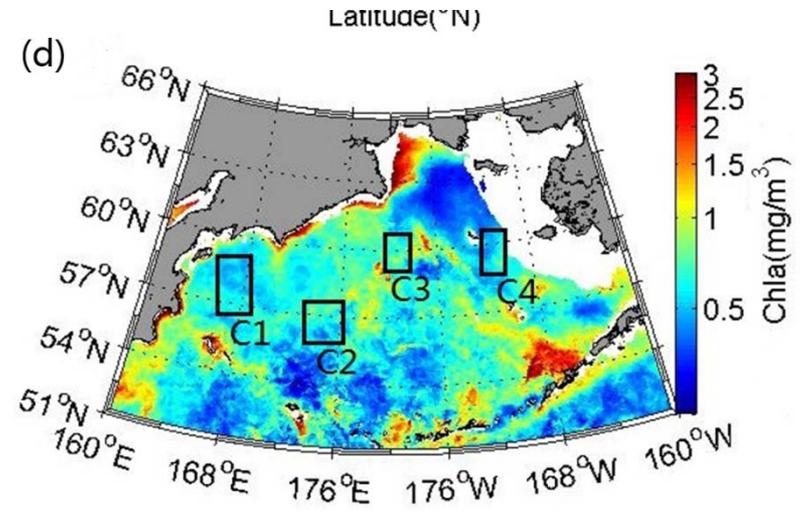
$$\Delta pCO_2 = \left(\frac{\partial pCO_{2@therm}}{\partial V_{therm}} \right) \Delta V_{therm} + \left(\frac{\partial pCO_{2@mix}}{\partial V_{mix}} \right) \Delta V_{mix} + \left(\frac{\partial pCO_{2@bio}}{\partial V_{bio}} \right) \Delta V_{bio} + \left(\frac{\partial pCO_{2@flux}}{\partial V_{flux}} \right) \Delta V_{flux} + \dots + \left(\frac{\partial pCO_{2@factor-n}}{\partial V_{factor-n}} \right) \Delta V_{factor-n} + \epsilon$$



Validation (2)-daily



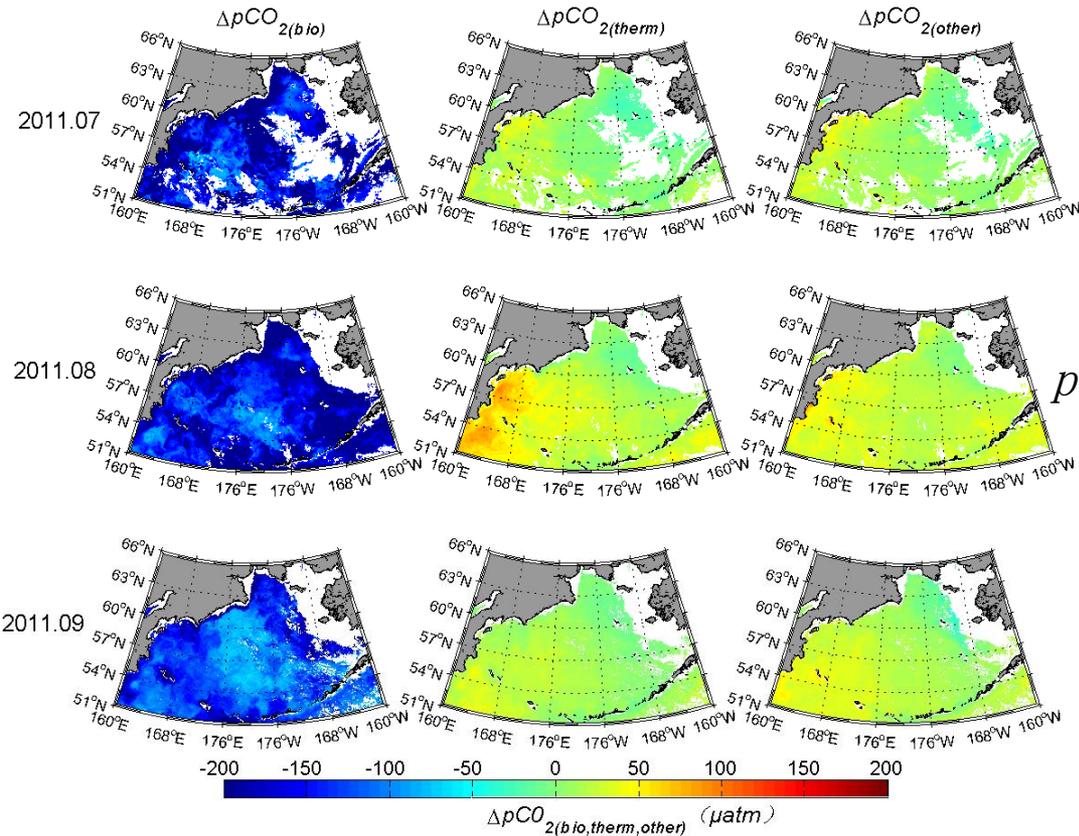
Influence by pervious spring algae bloom on summer pCO_2



The re-equilibrium time triggered by the air-sea CO_2 flux (40-60 day)

Contribution of different controlling factors on the variation of pCO_2

$$pCO_2 = pCO_{2(o)} + \left(\frac{\partial pCO_{2@therm}}{\partial V_{therm}} \right) \Delta V_{therm} + \left(\frac{\partial pCO_{2@bio}}{\partial V_{bio}} \right) \Delta V_{bio} + \varepsilon$$



$$npCO_{2(sea)} = pCO_{2(sea)@T_{obs}} * e^{(0.0423*(T(o)-T_{obs}))}$$

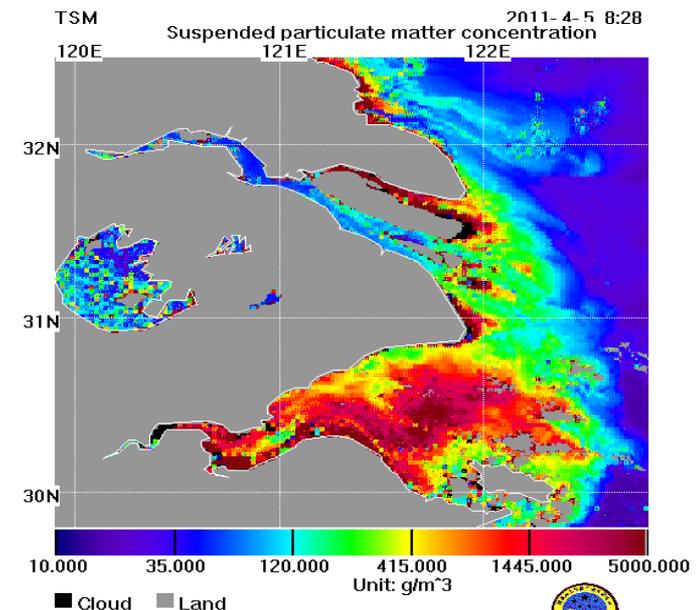
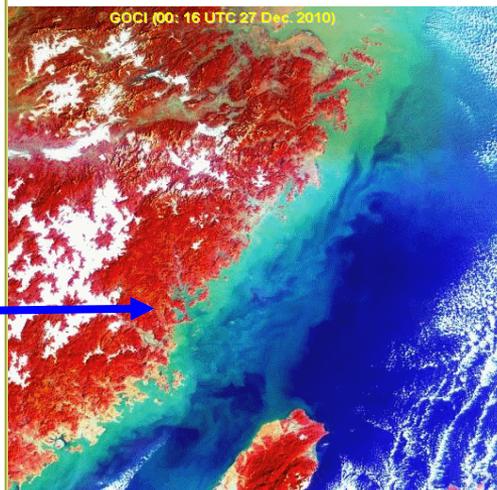
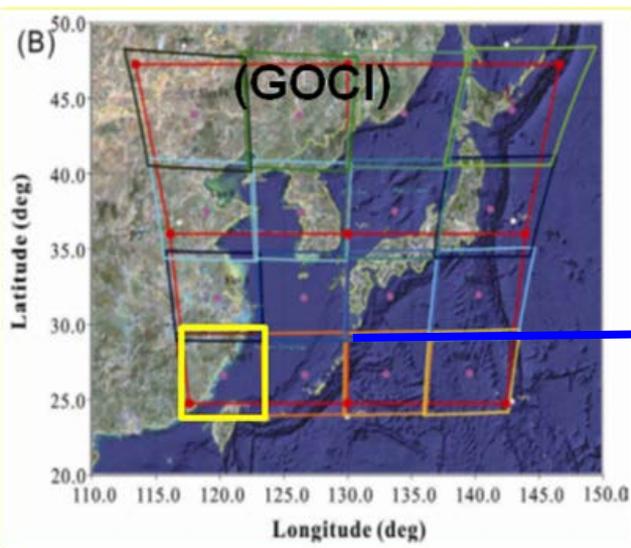
$$B^* = n\Delta pCO_{2(sea-air)}$$

$$pCO_{2(therm)} = \Delta pCO_{2(sea-air)} - npCO_{2(sea-air)}$$

$$\Delta pCO_{2(other)} = B^* - \Delta pCO_{2@bio}$$

Ⓞ Ongoing work..... (MeSAA- $p\text{CO}_2$)

1. Parameterization of the **Physical Mixing Effect** or Meso-scale processing (mixing index, SSS, SST, MLD, other proxy?)
2. Parameterization of the **Biological Effect** (C/chla, NCP, phytoplankton types, etc.)
3. Parameterization of **Processes at Different Time Scales** (re-equilibrium time?)



- **Framework of satellite-based marine carbon research**
- **Inorganic Carbon : $p\text{CO}_2$**
- **Carbon flux estimation**
- **SatCO₂-- Satellite-based marine carbon monitoring and analysis system**



From surface to the water column

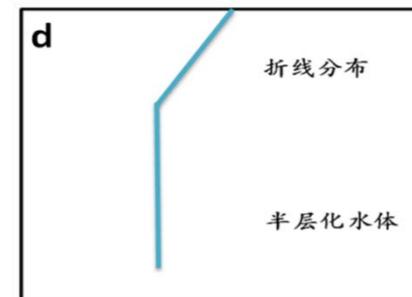
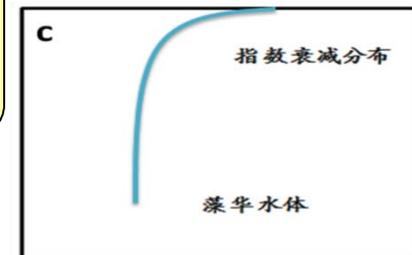
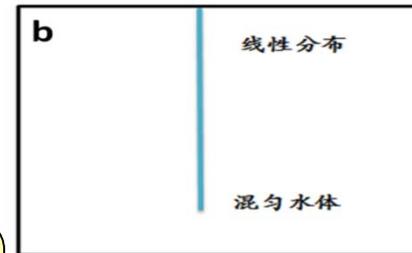
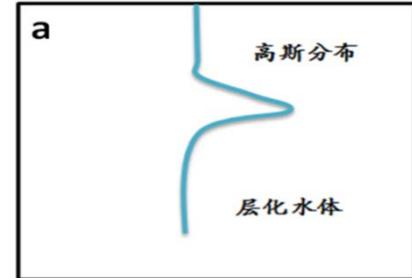
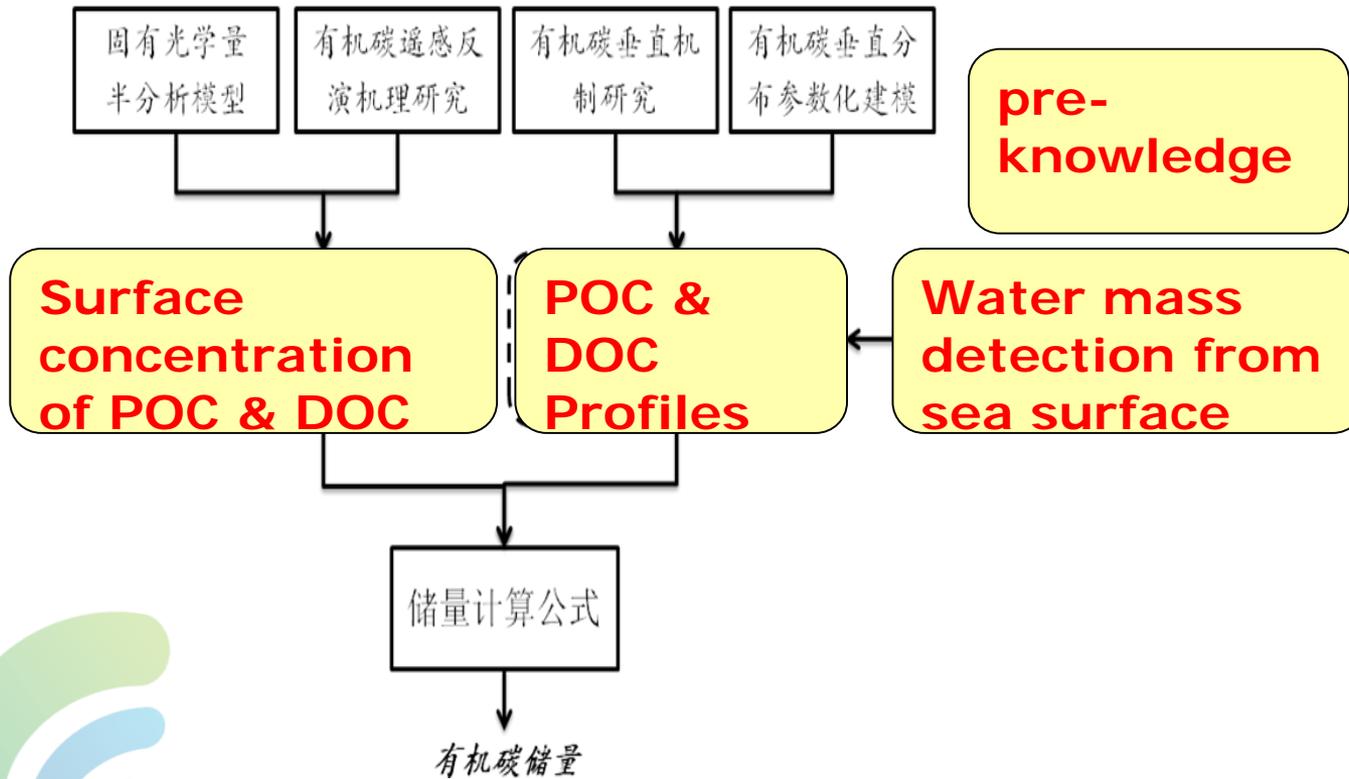
Estimation of POC inventory



SatCO₂

$$OC_{reservoir} = \int_0^{z_{eu}} OC_{surf} * f(z, L) dz$$

$$S_{poc} = \int_0^{z_m} C_{poc,s} f(z, \dots) dz,$$



Water mass and models of DOC profile



SatCO₂

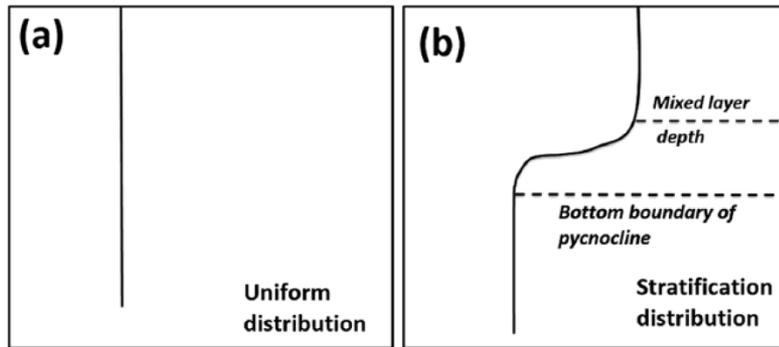
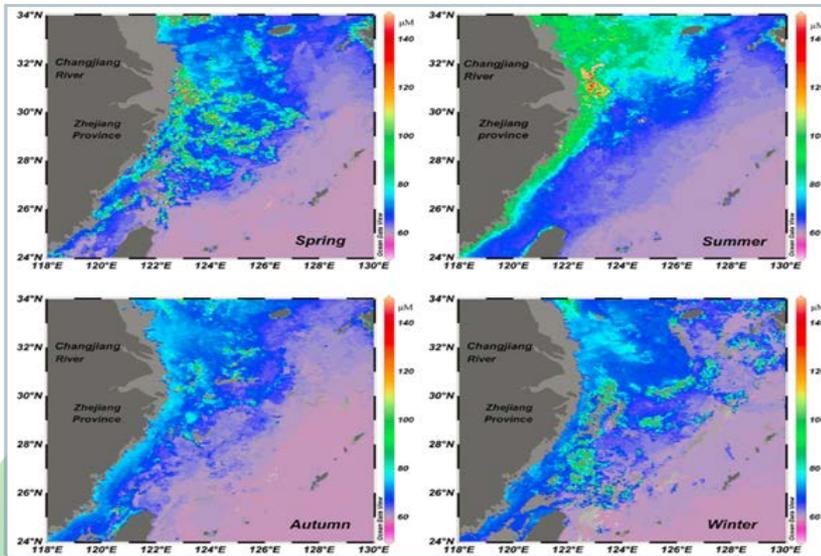
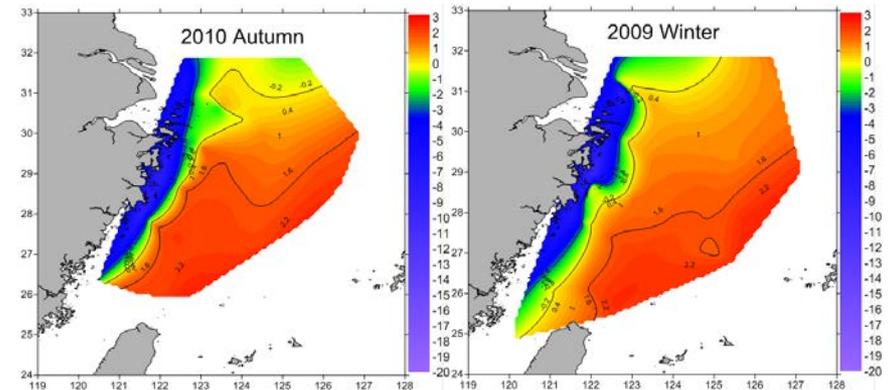
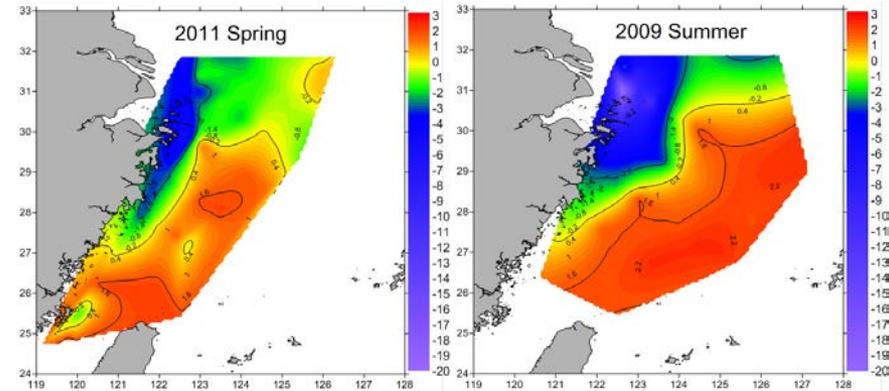


Figure 6. Two simplified models of the vertical DOC profile: (a) a uniform model and (b) a stratification model.

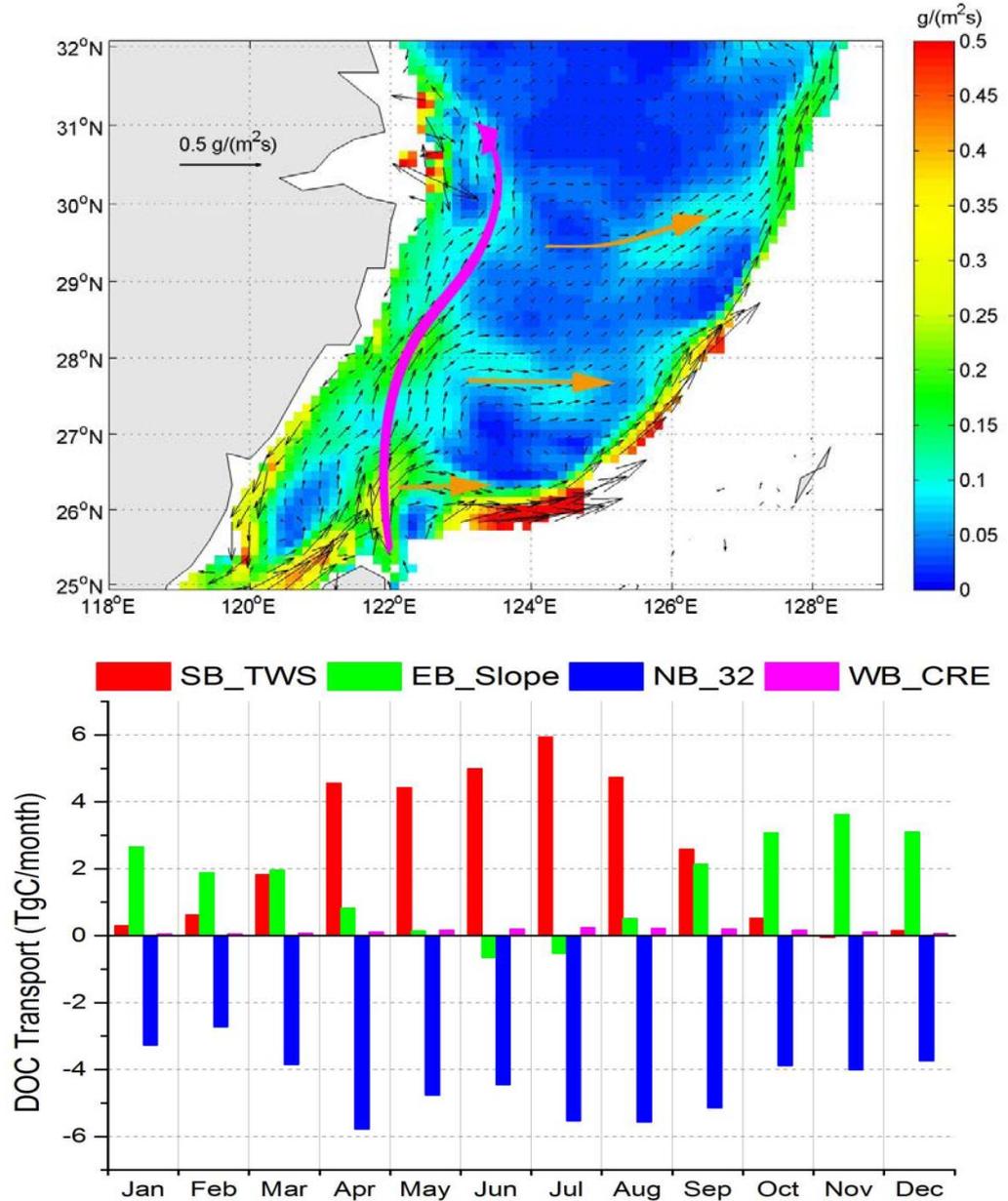
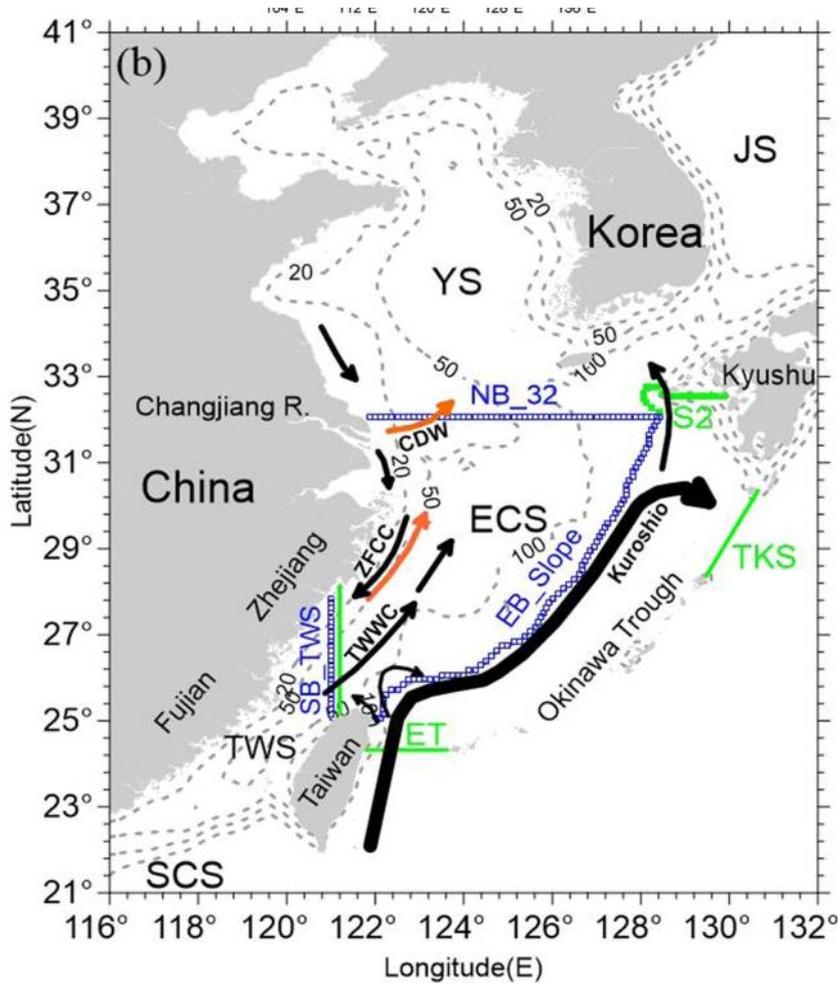


Spiciness index
(Li et al., 2006)

$$\beta = \frac{(T - T')}{\Delta T} + \frac{(S - S')}{\Delta S} \alpha,$$

Liu et al., JGR-Oceans, 2014

Estimation of lateral DOC transport in marginal sea based on remote sensing and numerical simulation



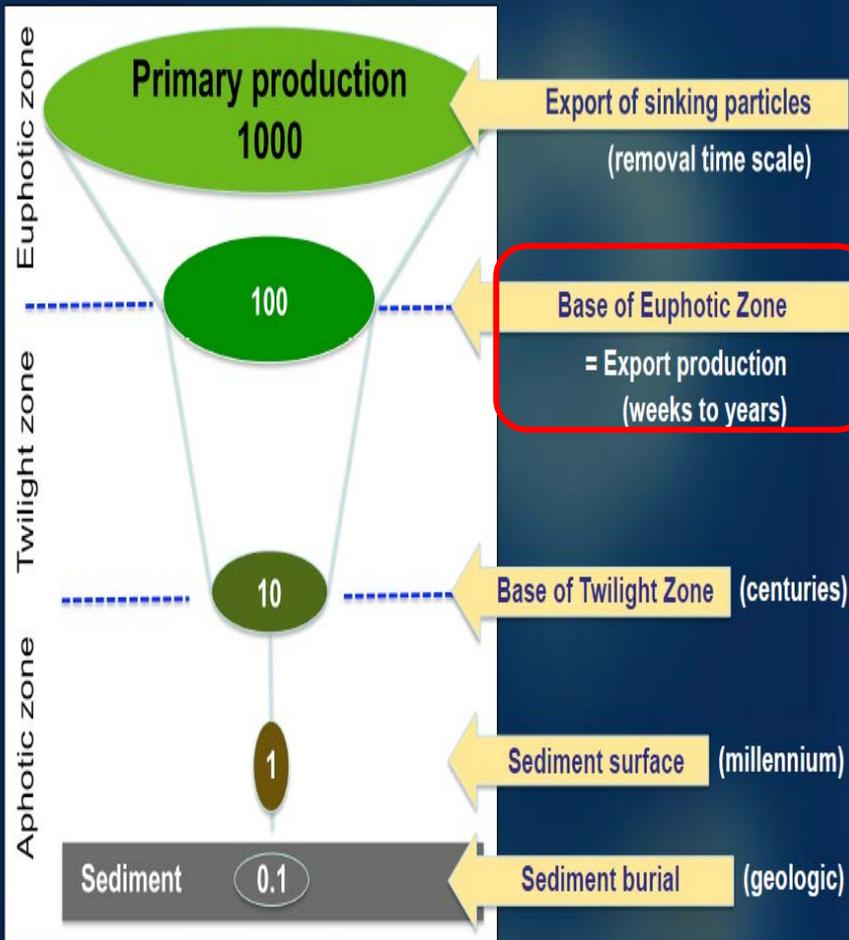
Cui, He*, et al., JGR, 2018

POC export flux (sequestration)

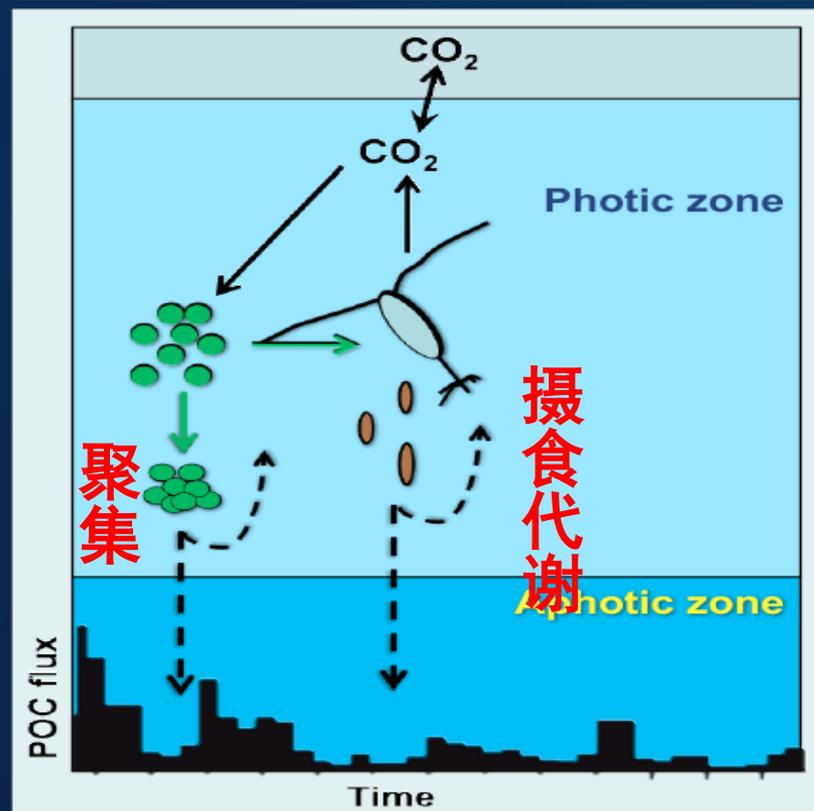
- direct sinking flux of large phytoplankton and associated aggregates by gravity
- flux of zooplankton feces by grazing

The Carbon Pumps

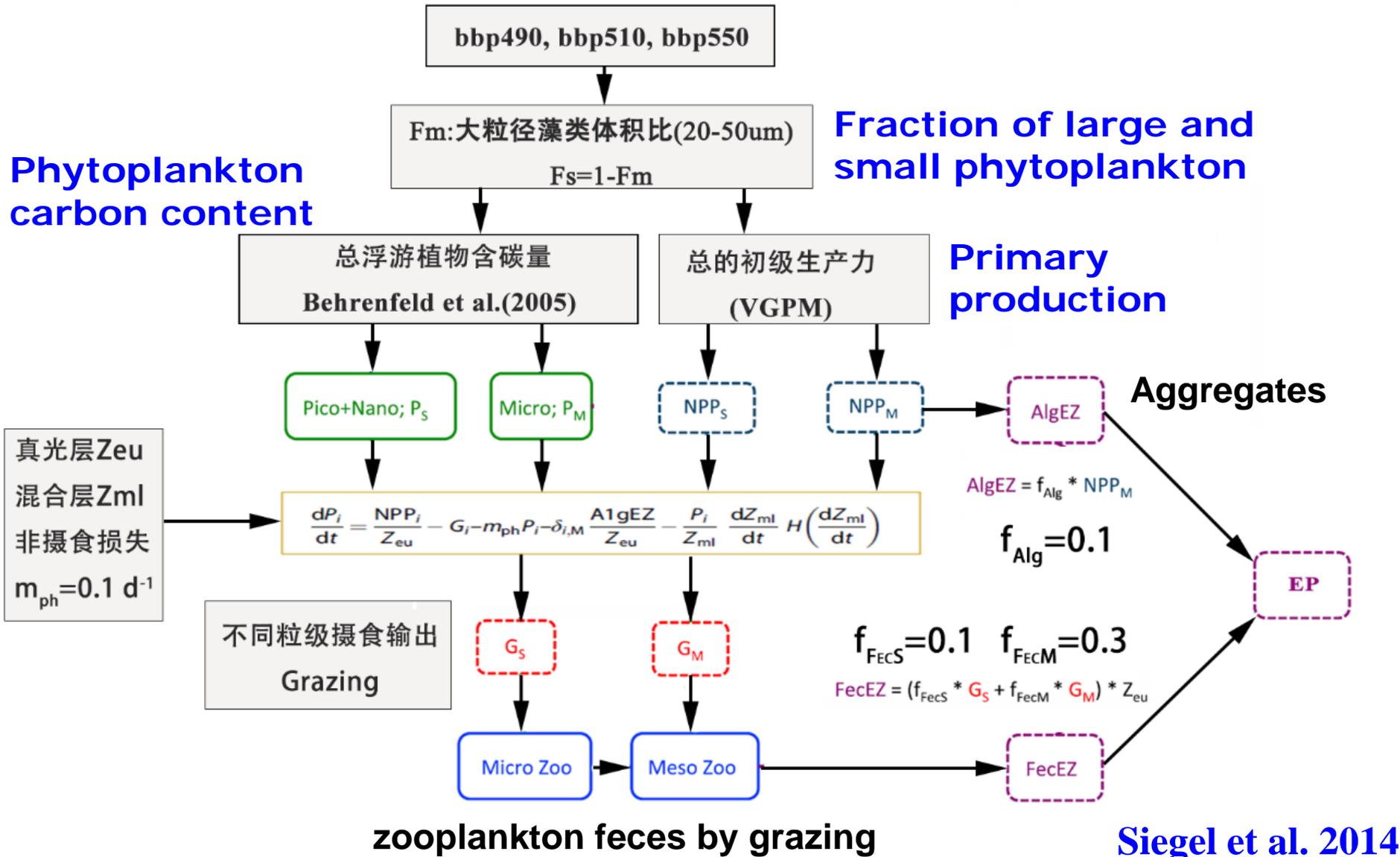
Removal Rates and Time-scales



Excursion II: Particle Sinking and Degradation

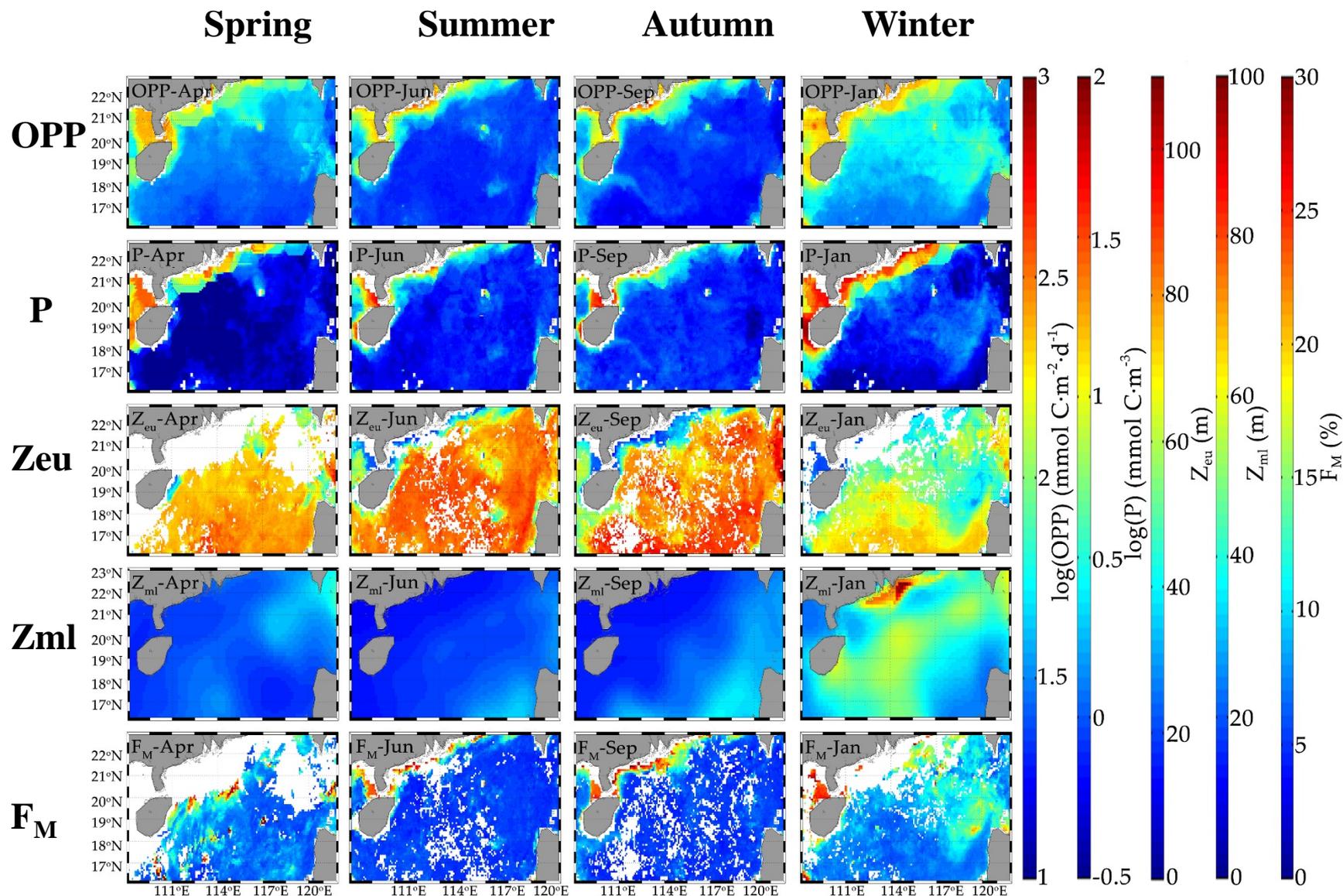


POC export flux estimated based on Food-web model



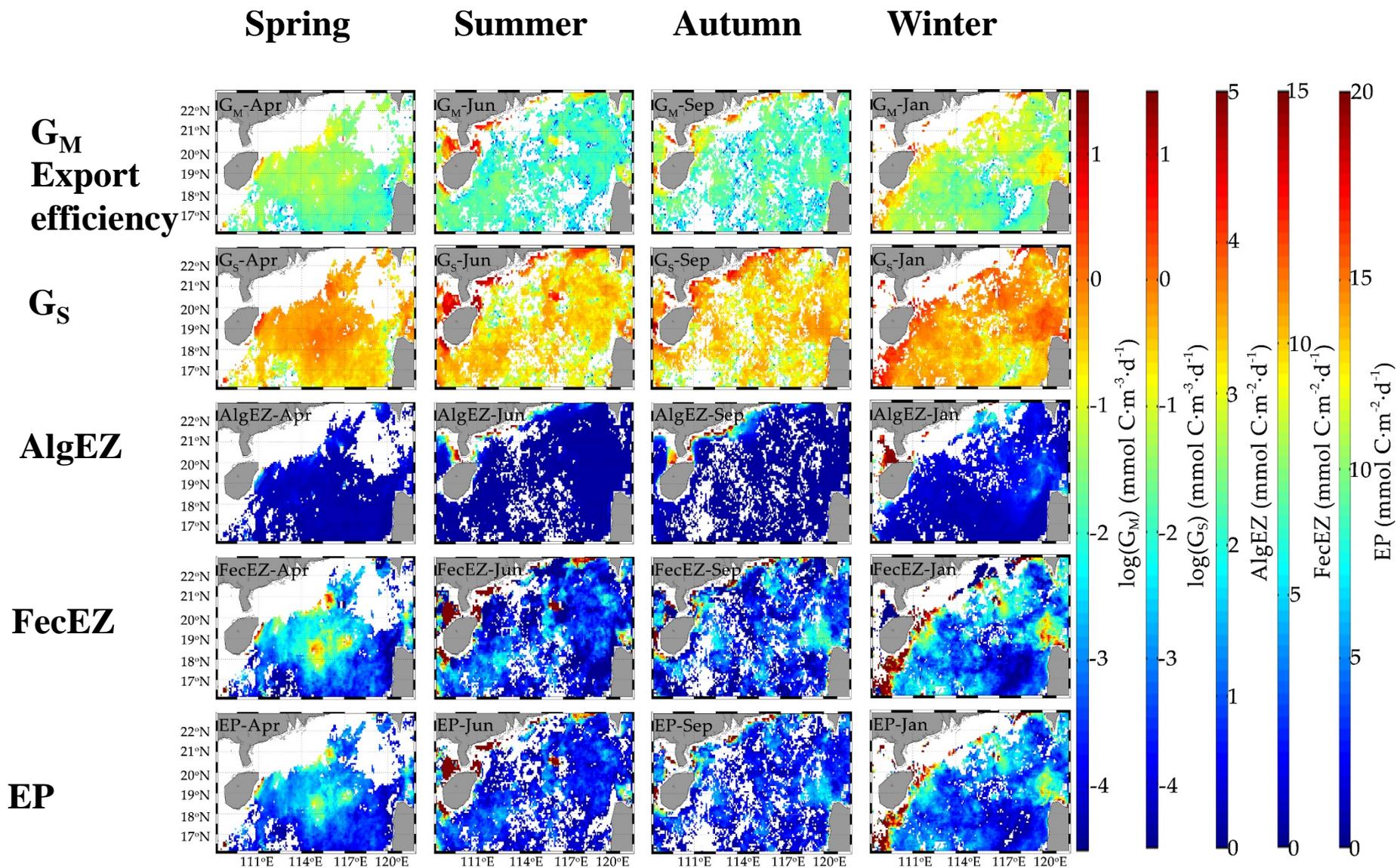
Model-input data

Li, Bai*, etc. 2018, JGR



Model results

Li, Bai*, et al. 2018, JGR



- **Framework of satellite-based marine carbon research**
- **Example: $p\text{CO}_2$ and POC export flux**
- **SatCO₂-- Satellite-based marine carbon monitoring and analysis system**



@ Satellite-based marine carbon monitoring and analysis system (SatCO2)



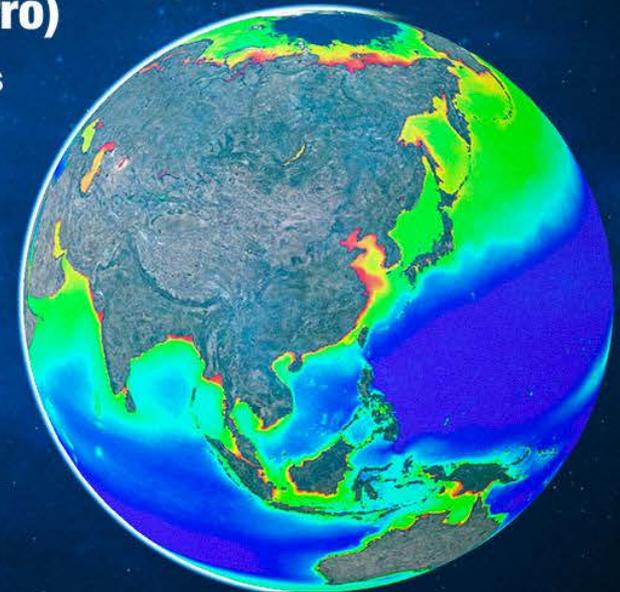
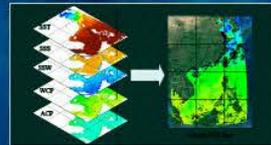
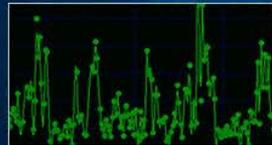
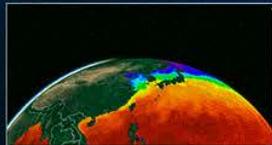
SatCO₂



Marine Satellite Data Online Analysis Platform (SatCO2-Pro)

SURPORT for multiple sources & time series data sharing and analysis

- Online access of unique satellite remote sensing data
- 3D Earth visualization and scientific computation
- Analysis and evaluation of multi-source (satellite, in situ and model) data
- User-defined algorithms and product generations
- Calculation and evaluation of ocean carbon fluxes
- Easy integration of professional modules

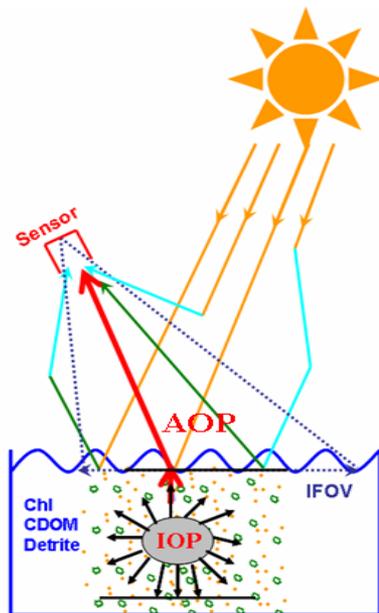


Concluding remarks

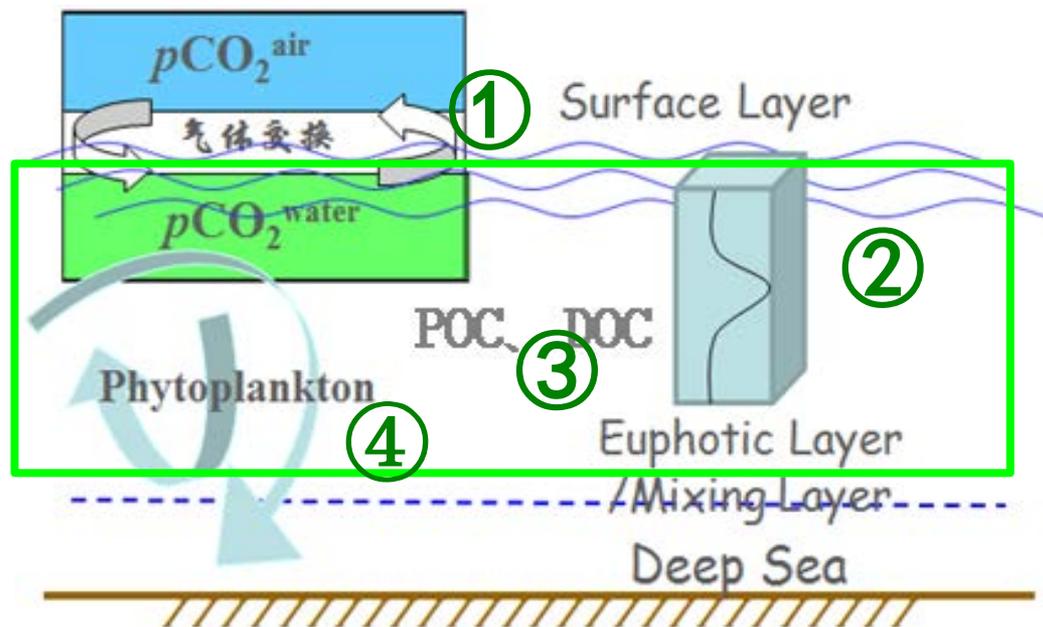


SatCO₂

1. We want to develop an **integrated marine carbon monitoring and assessing system**, to better quantify, understand and predict the changing marine carbon system, especially in the highly dynamic marginal sea.
2. It need the **joint research on multi-disciplines and the collaboration**.



0.水光学遥感





**Thank you
for your
attention!**

