

HiMAT U.S. Team meeting

Seattle, WA

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Glacial Lake Assisted Melting (GLAM)



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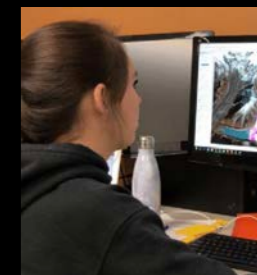
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**Acknowledgment: Thanks also to
David Shean, Dan Jantzen, and Bob Yoder**

A few key questions

How fast are Himalayan glacial lakes forming and growing?

How is suspended sediment implicated in radiative transfer, mass and energy balance of lakes and glaciers?

What is the energy and mass balance of glacial lakes and lakes' effects on glacier mass balance?

GLAM- Lakes (empirical- Shugar leads)

GLAM- Ice Flow (empirical-Haritashya leads)

GLAM- Icebergs (empirical-Watson leads)

GLAM- BioLith RT (numerical-Furfaro/Schiassi lead)

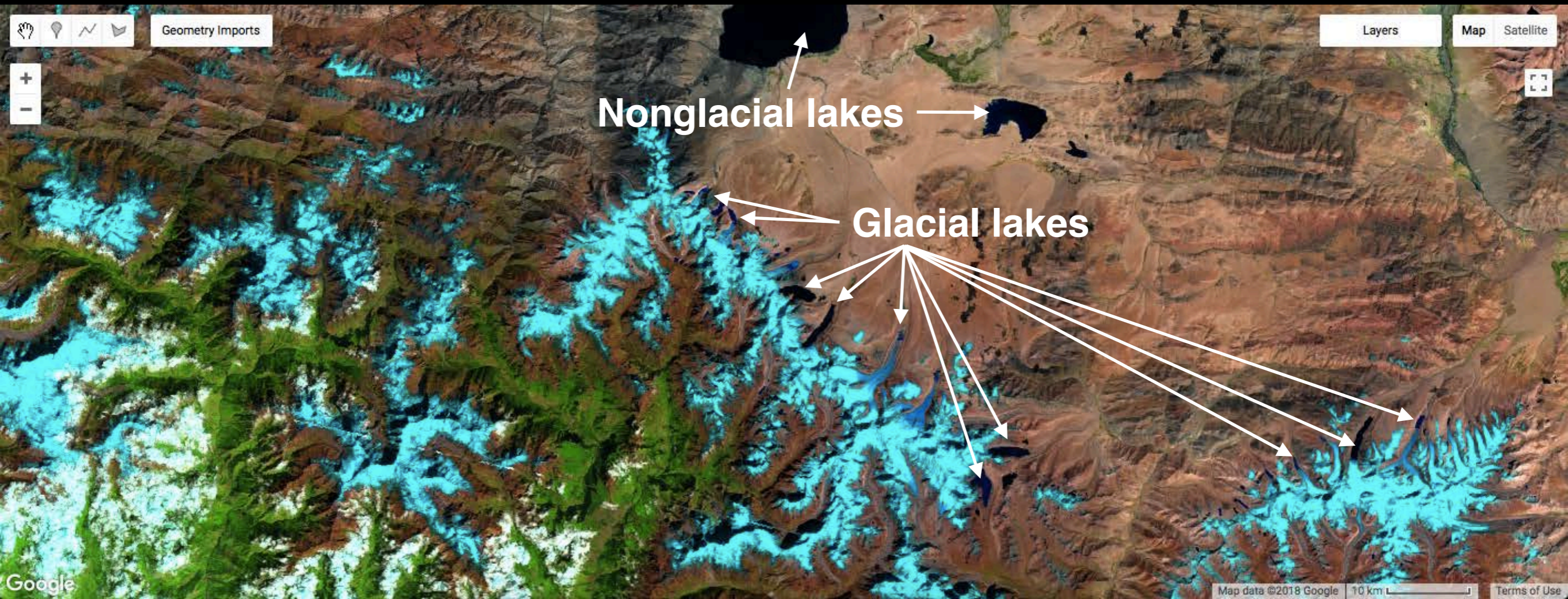
Today mostly this

GLAM- LITE (Lake, Icemass, and Thermal Energy Balance)(analytical- Kargel leads)

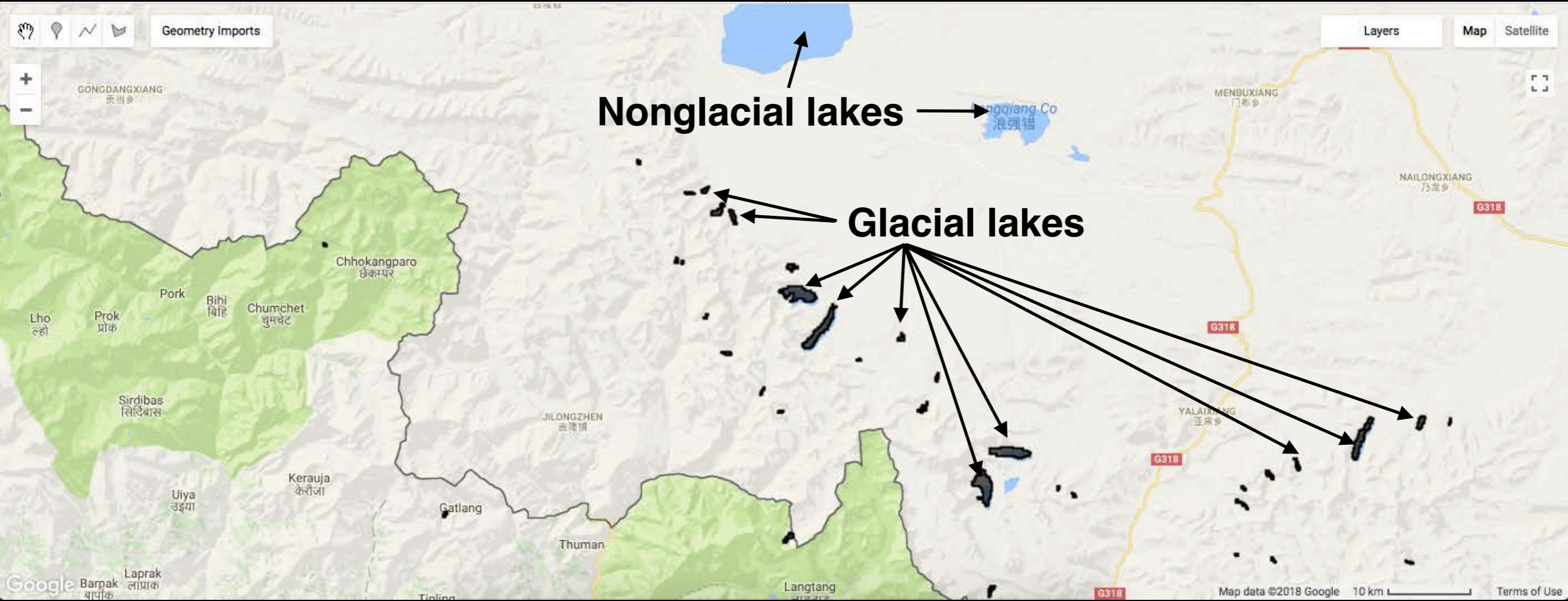
Spatio-temporal glacial lake mapping using Google Earth Engine

Dan H Shugar
& Aaron Burr

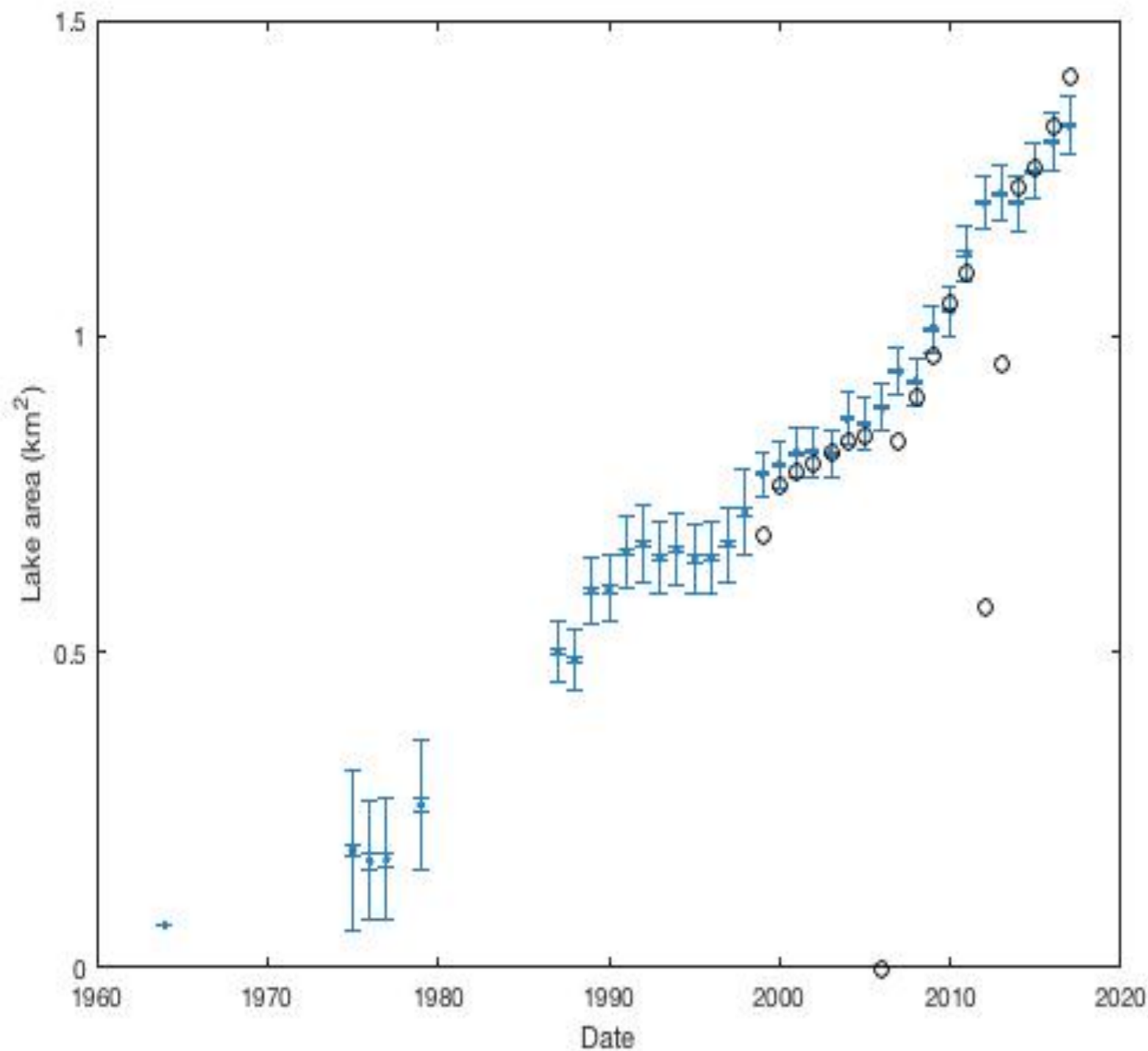
Goal: map only glacial lakes accurately, automatically



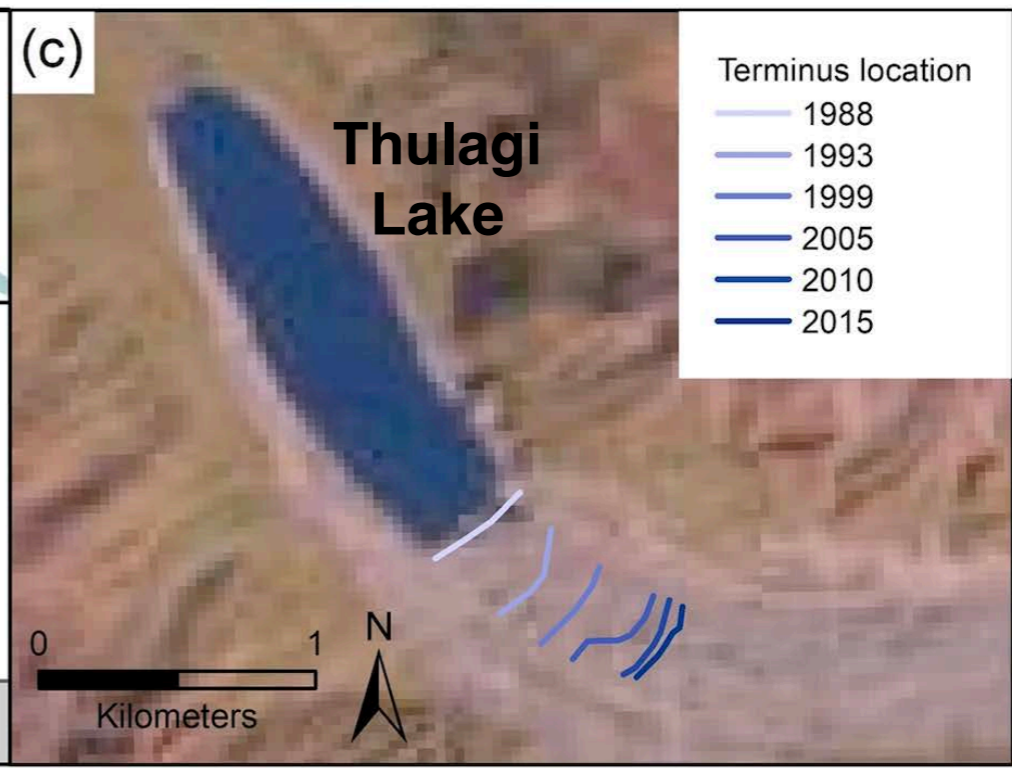
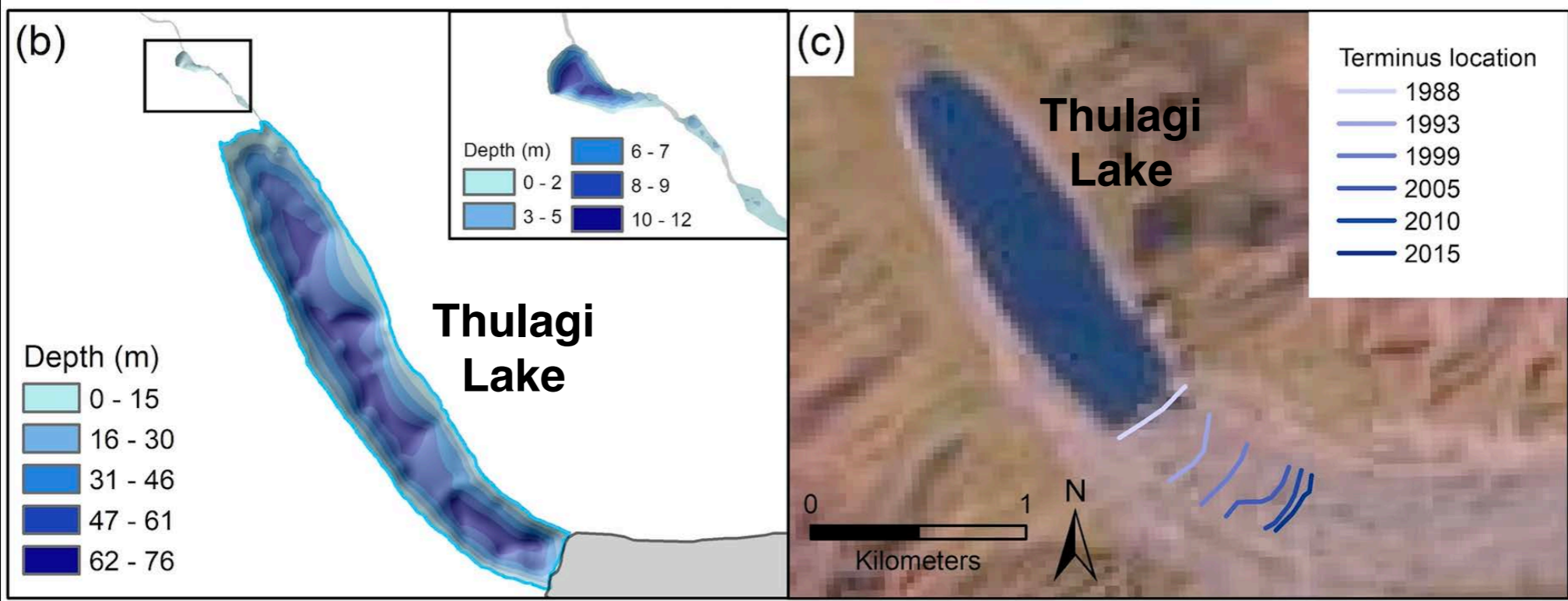
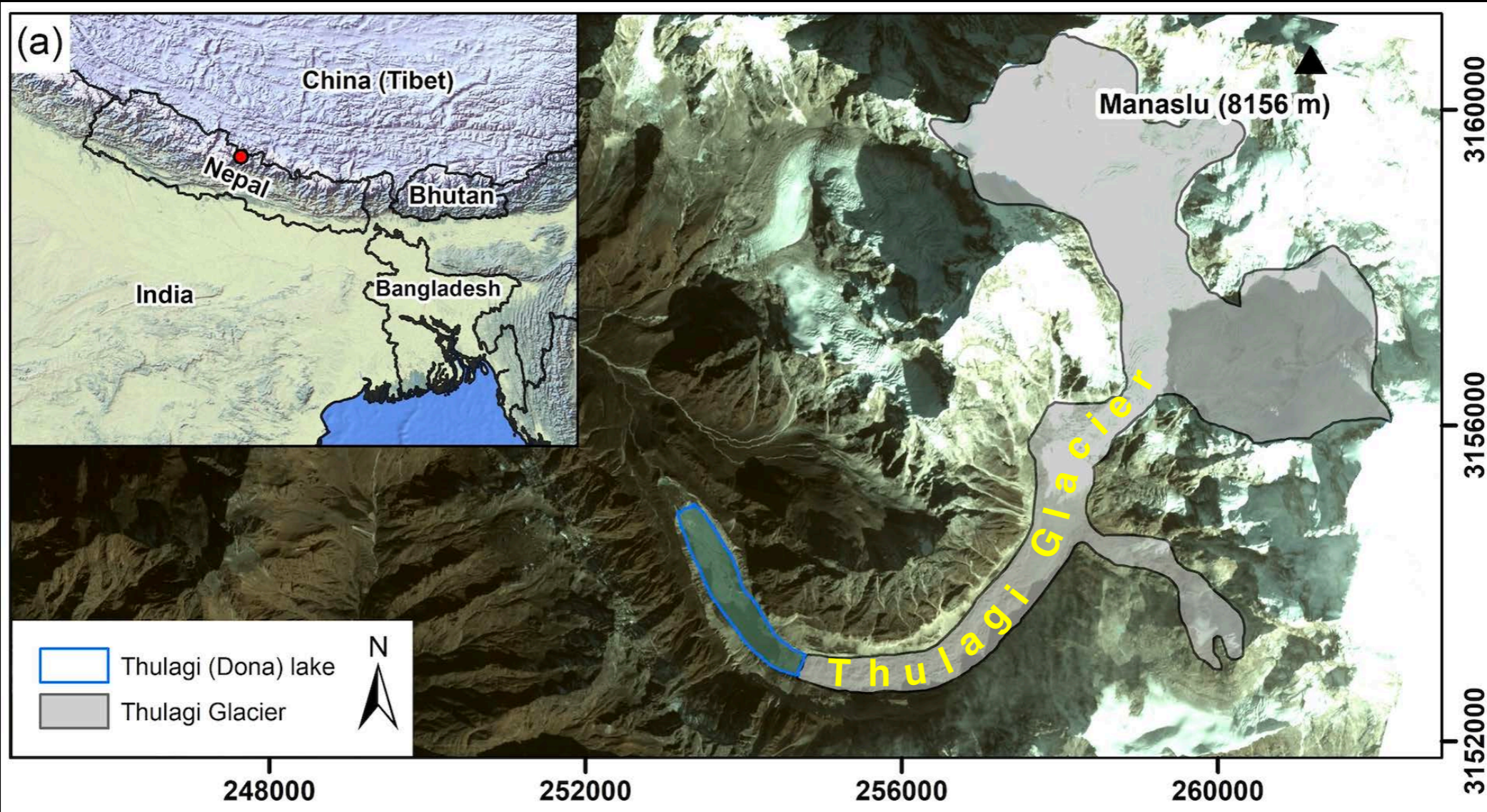
Goal: map only glacial lakes accurately, automatically



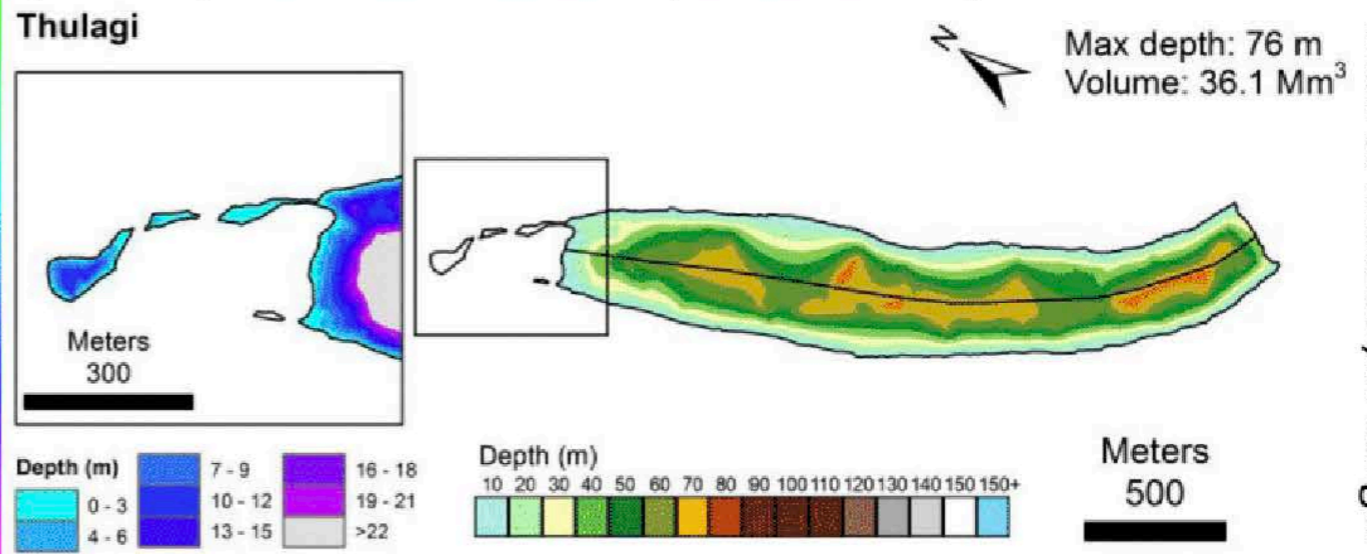
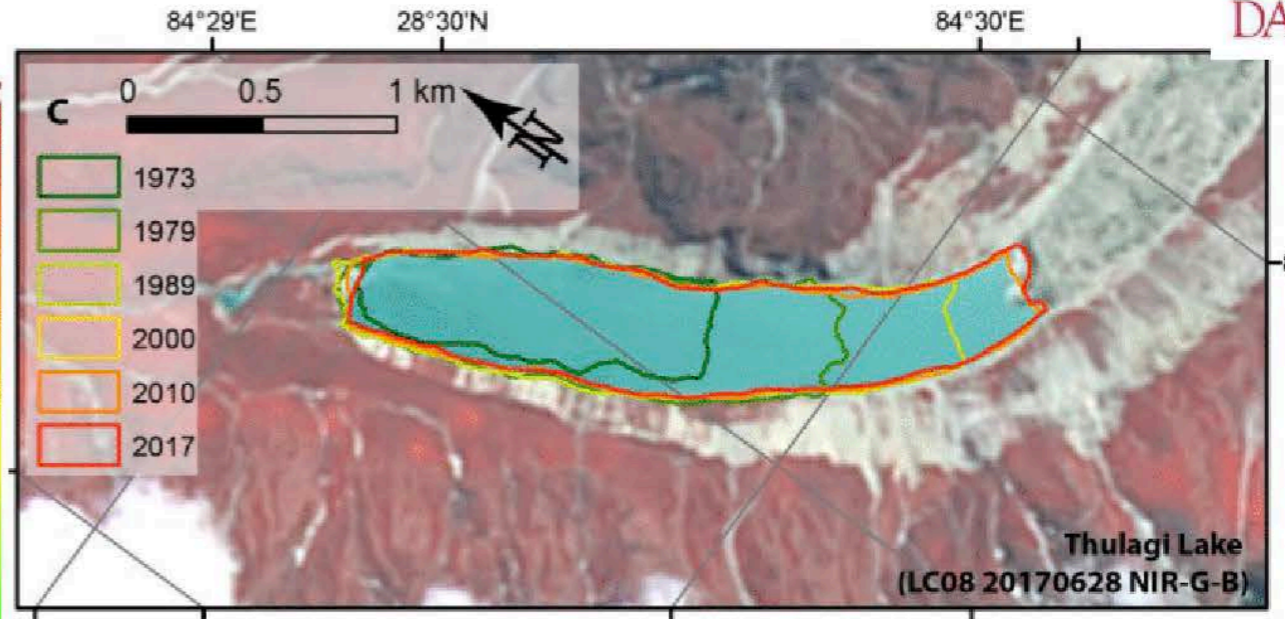
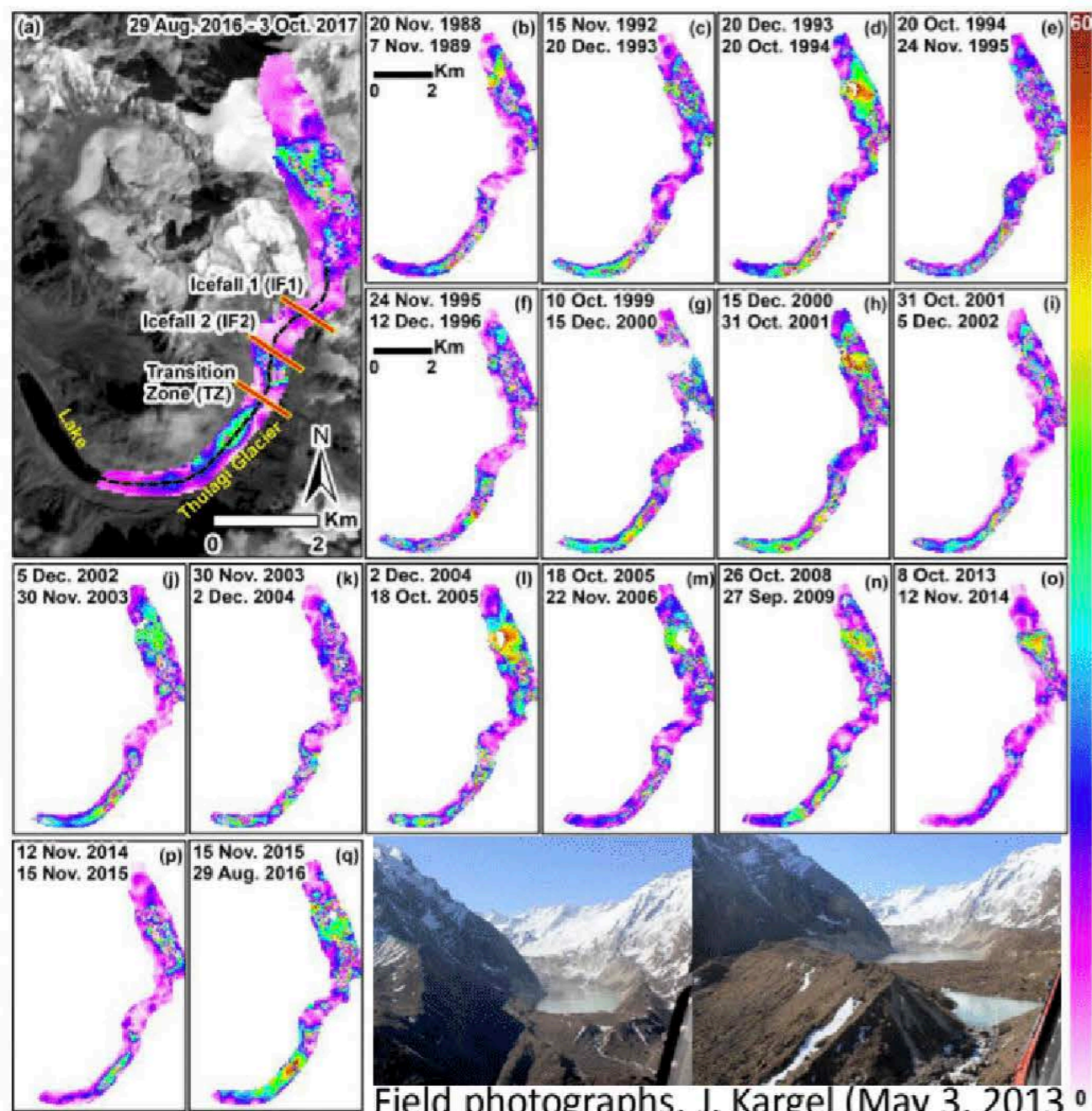
Automated & manual comparison at Imja Tsho



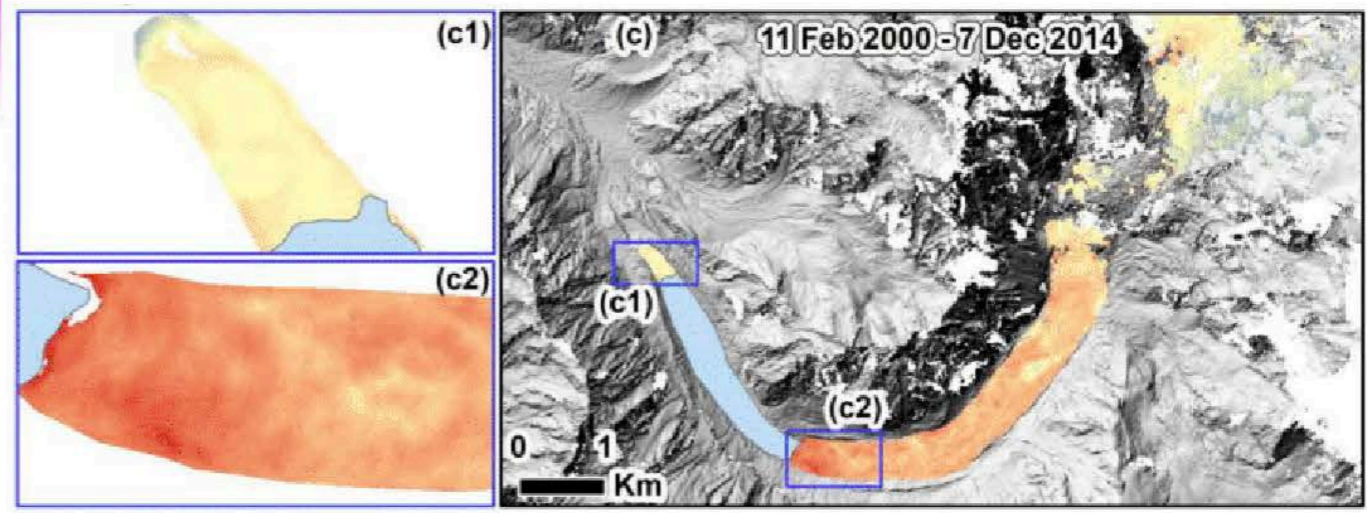
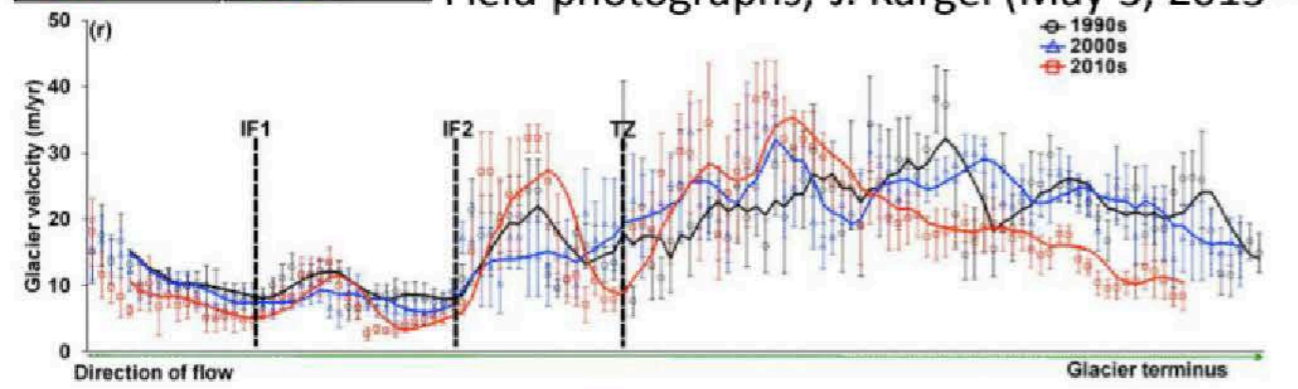
- **Generally good fit between manual (blue) and automated digitizing (black).**
- **Some errors due to brash ice, bergy bits, few cloud-free scenes in given year, etc.**
- **Next steps – incorporate ASTER, make code elegant so works across larger area through time.**



Thulagi Glacier Lake, Nepal



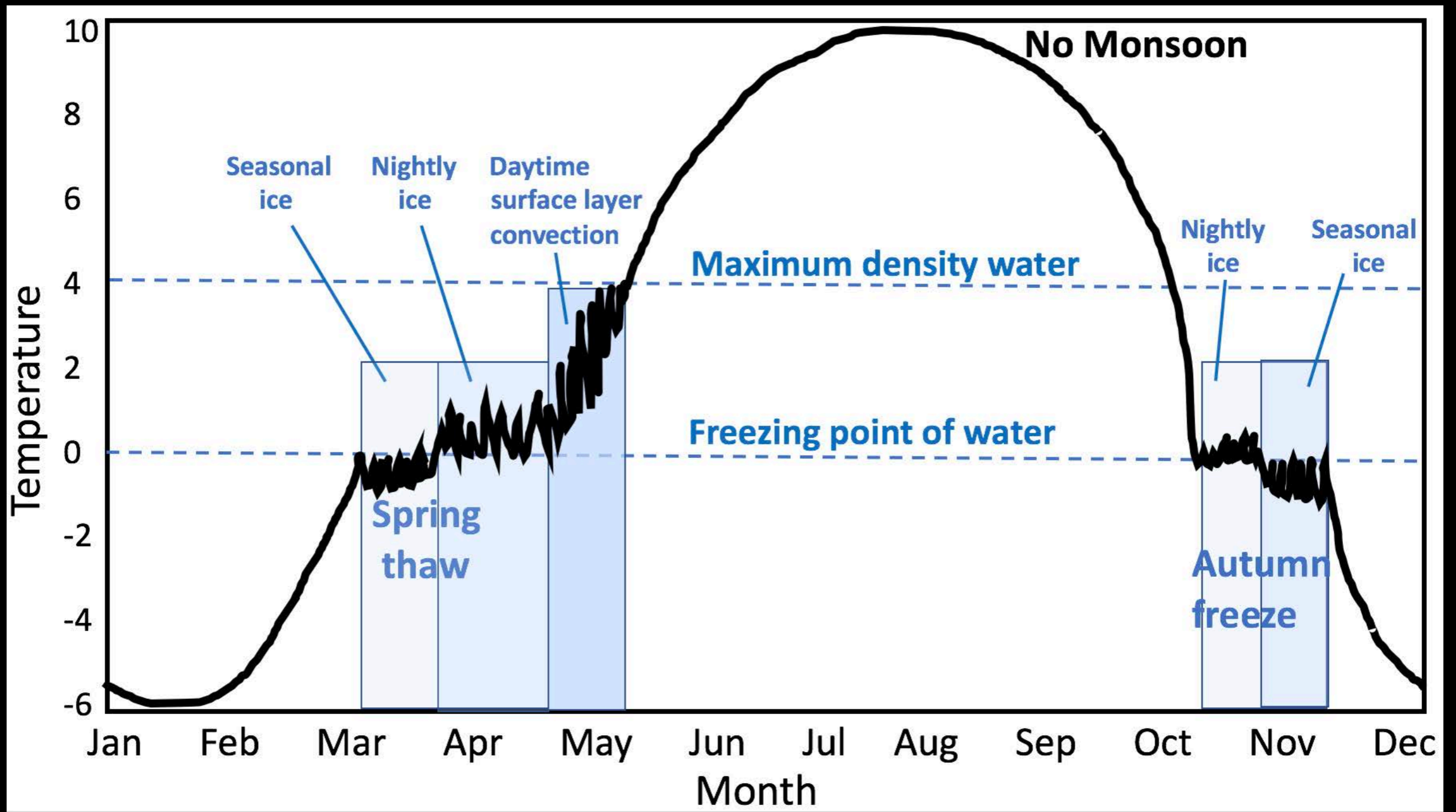
Field photographs, J. Kargel (May 3, 2013)

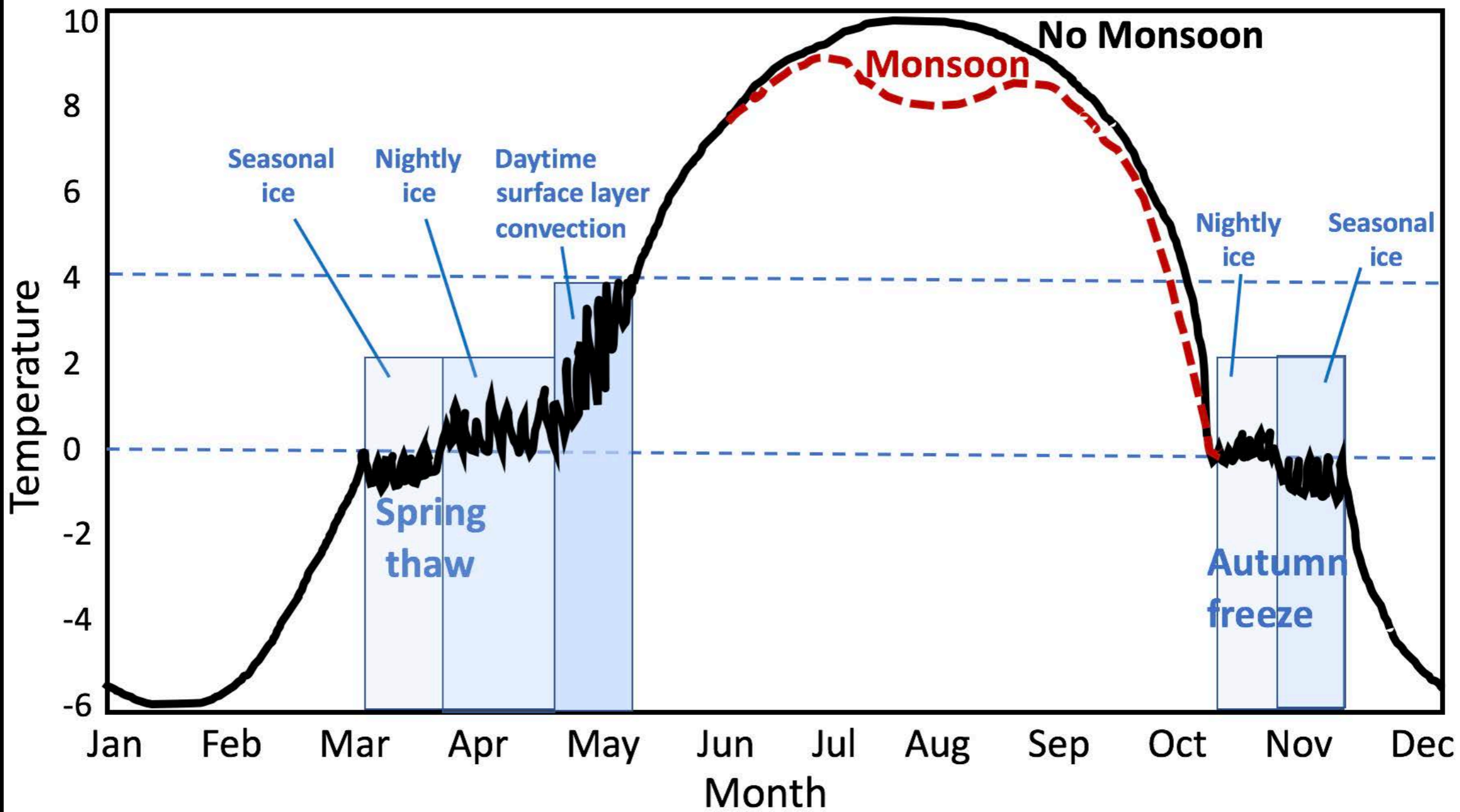


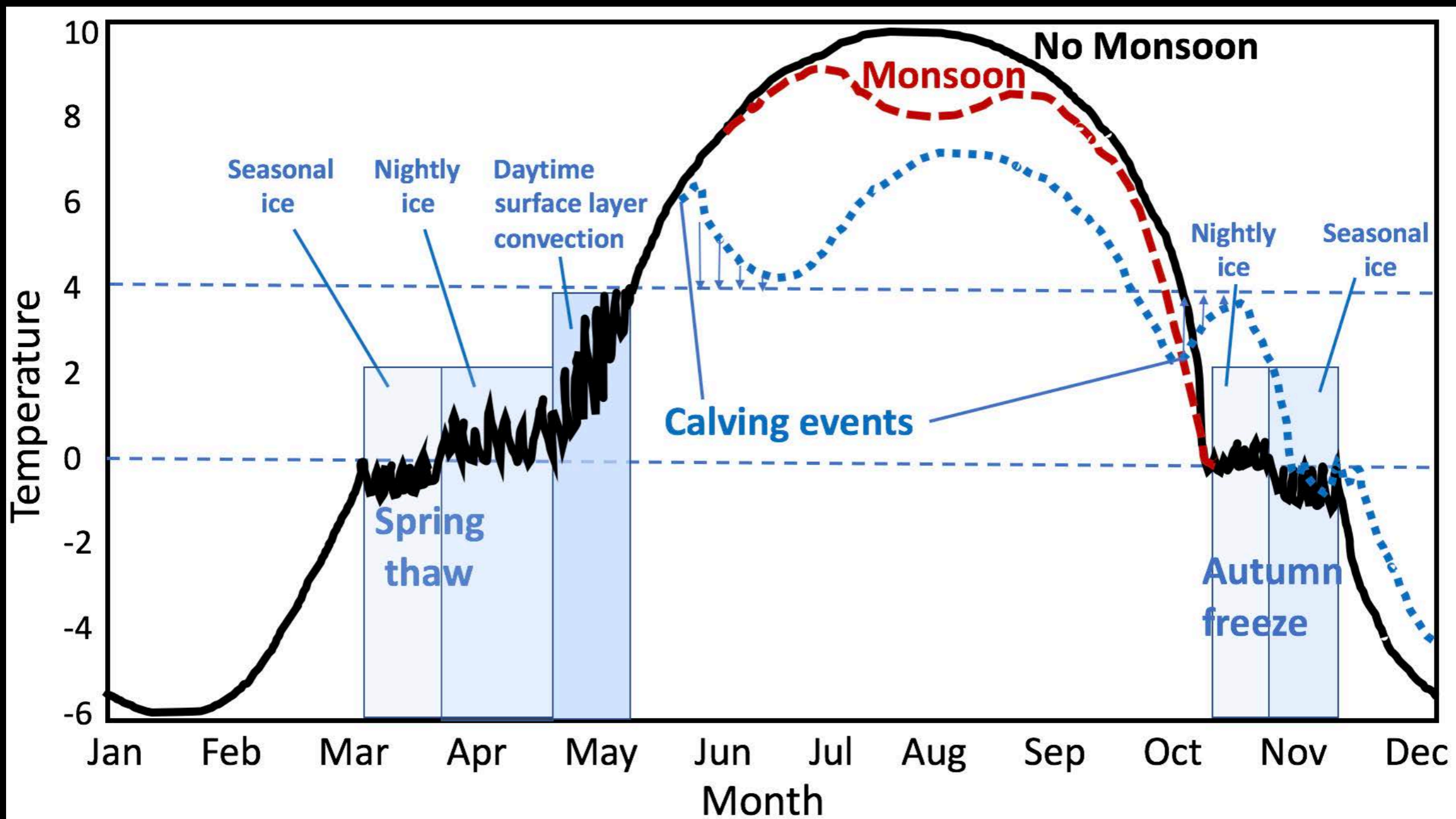
With D. Shugar

With S. Watson, J. Kargel

With D. Shean





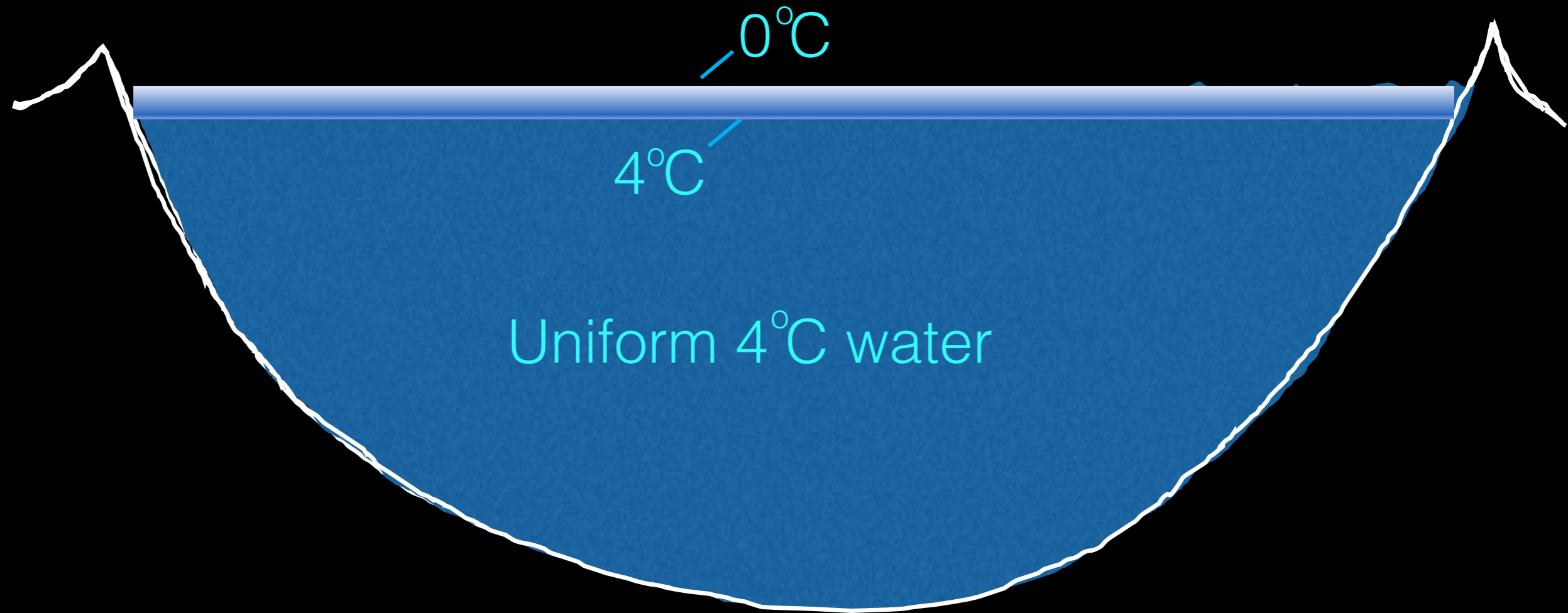


Early October

Uniform 4°C water



Early November first freeze



January thick lake ice



April thaw

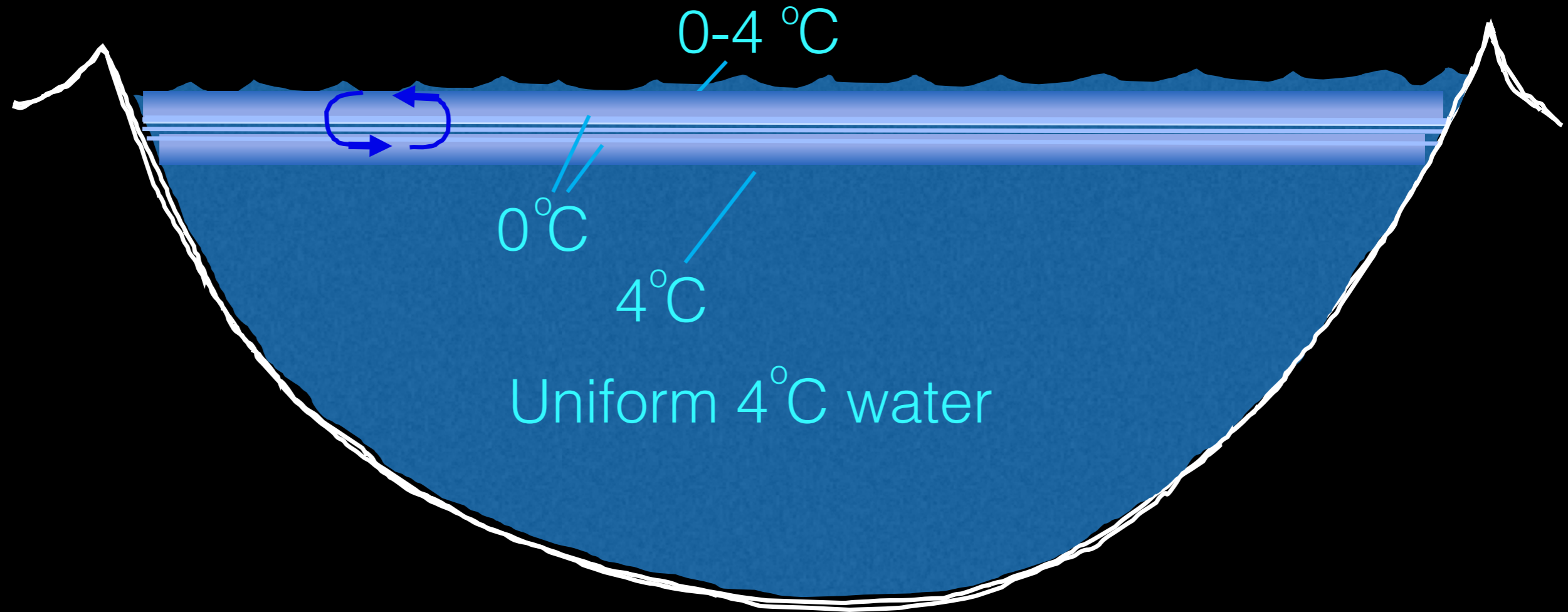
0 °C, thin melting ice

0 °C 4 °C

Uniform 4 °C water



May warmup and free thermal convection



June

Uniform 4°C water

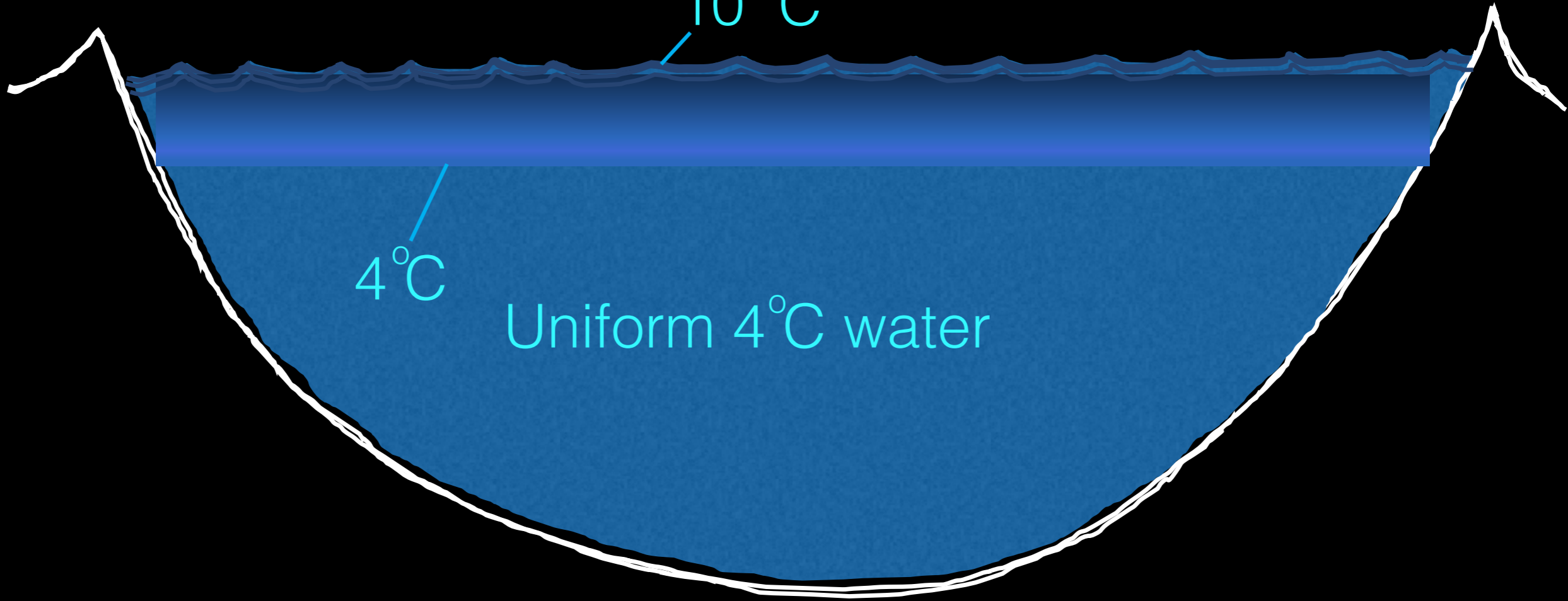


July

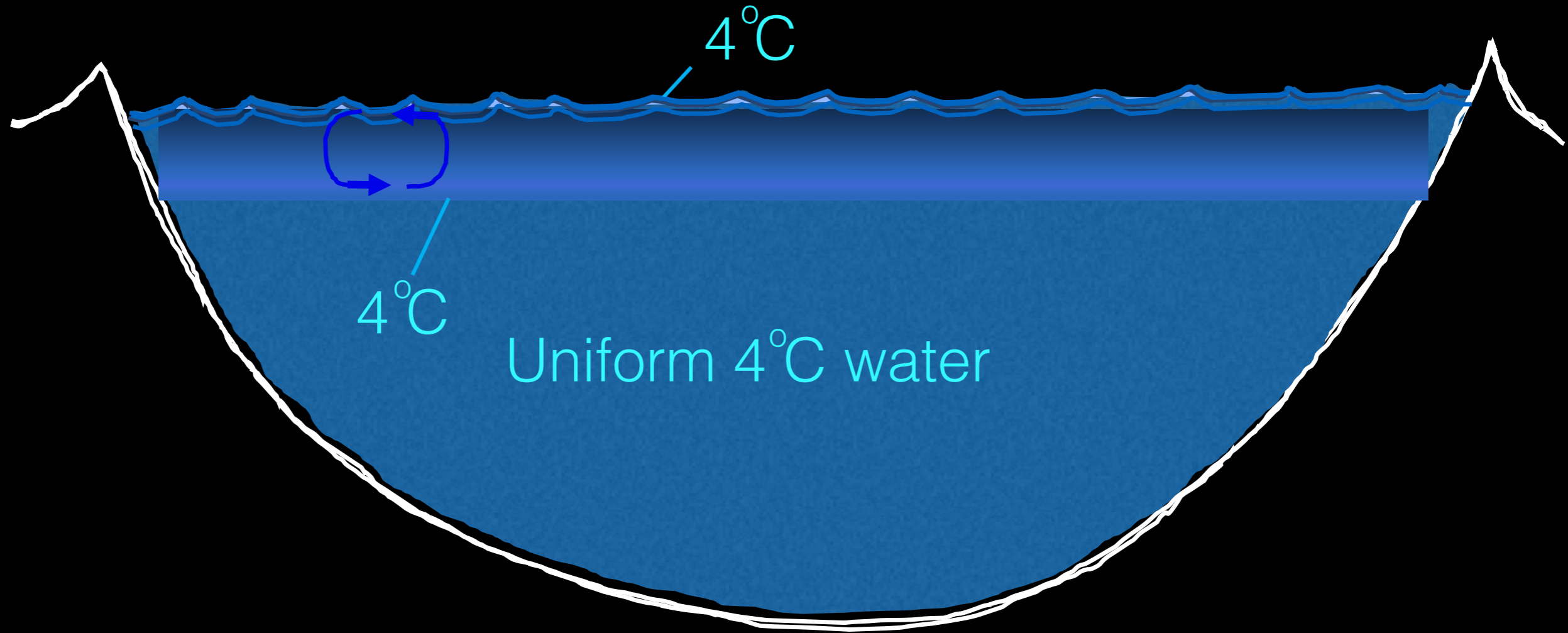
10 °C

4 °C

Uniform 4 °C water



Late September, Autumn free convection

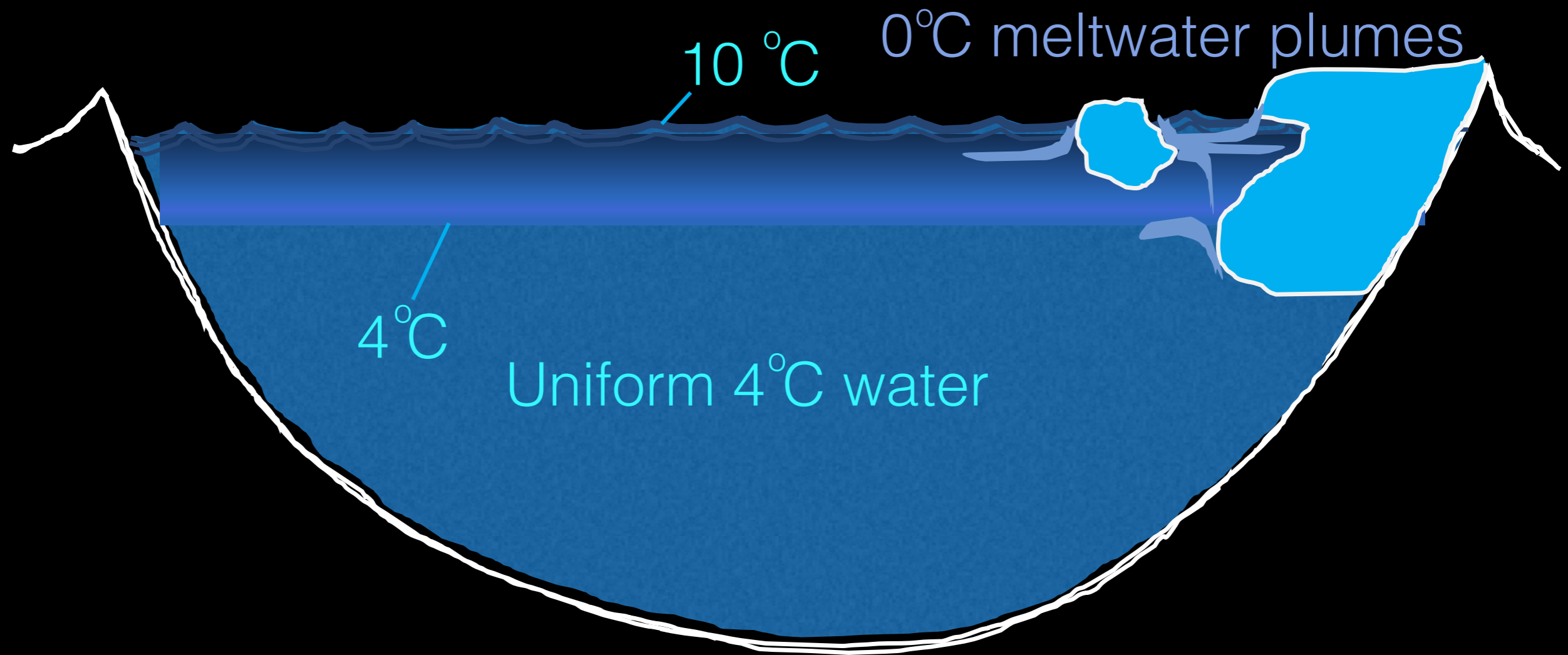


Early October

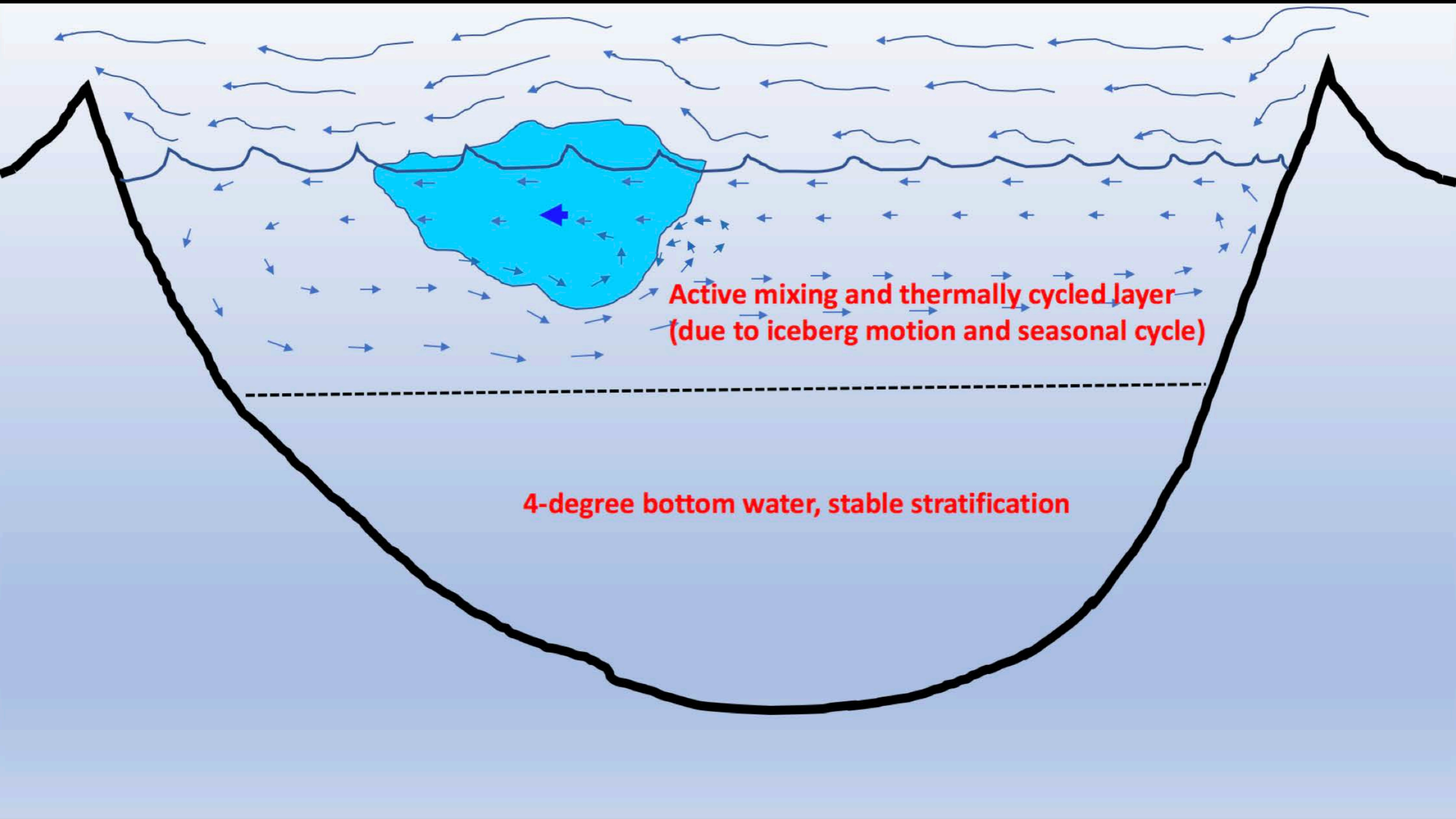
Uniform 4°C water

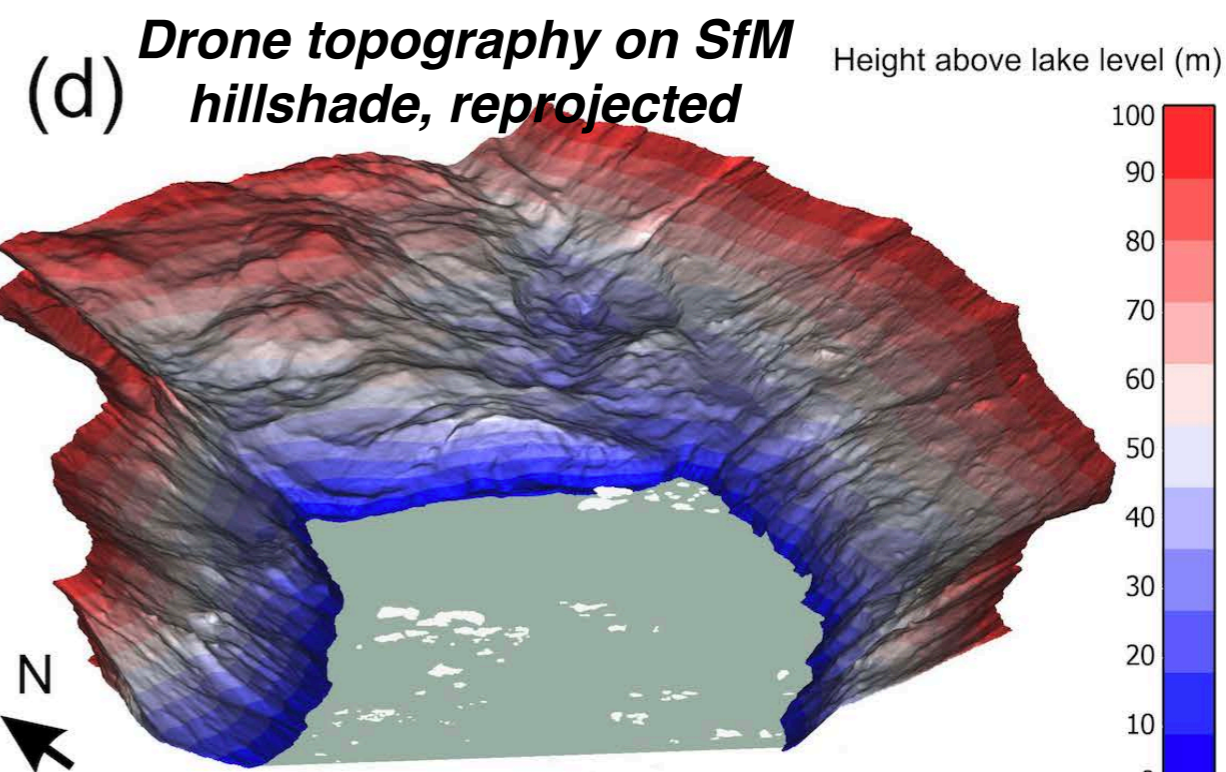
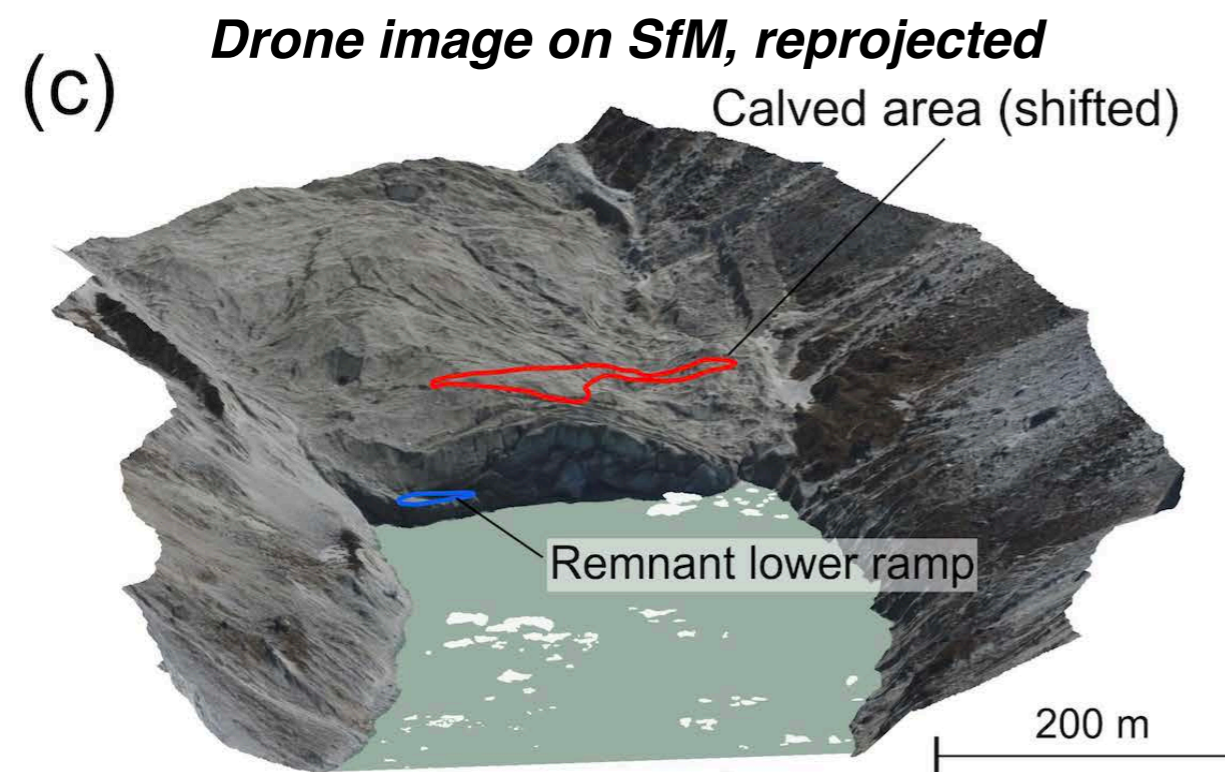
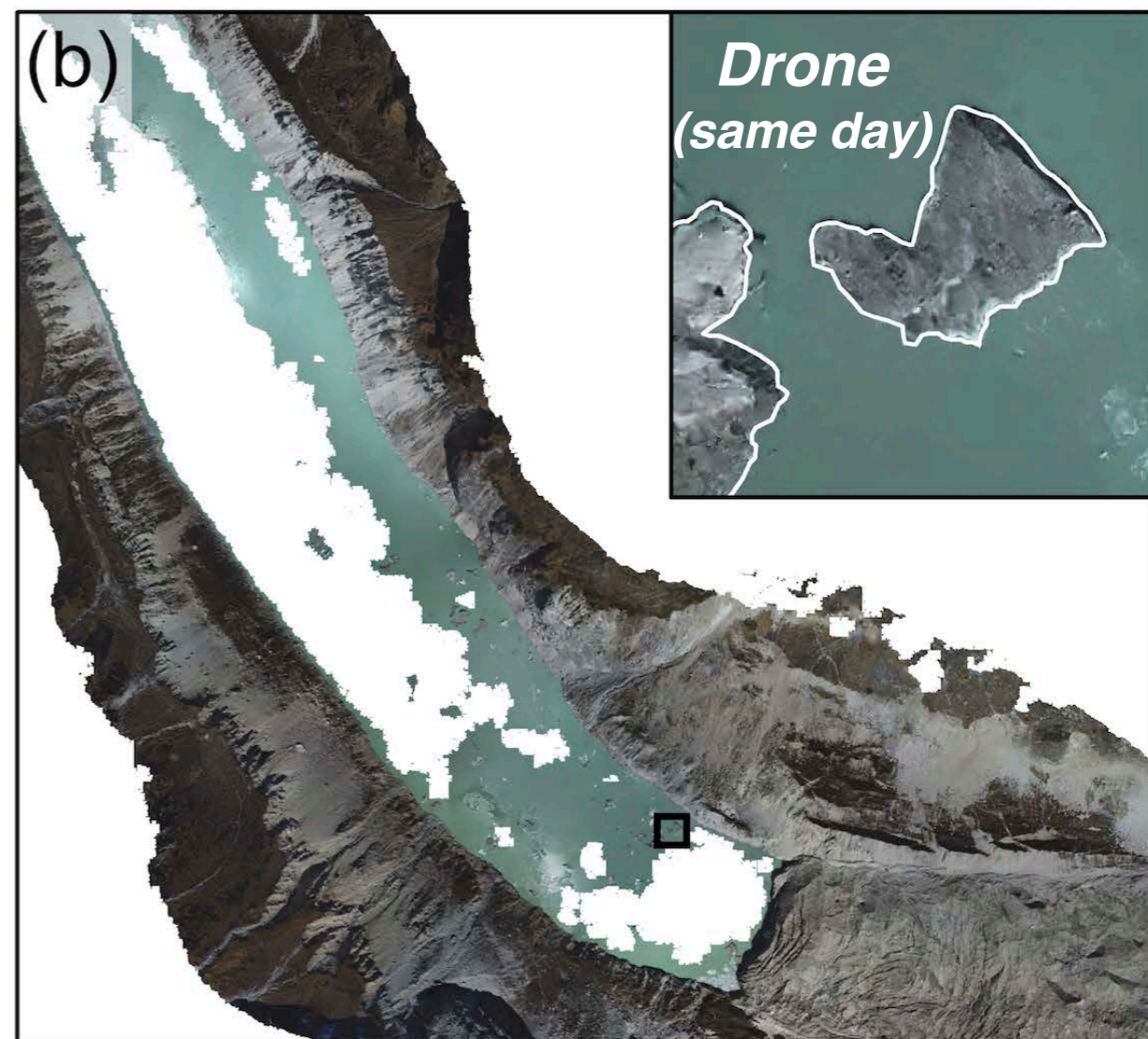
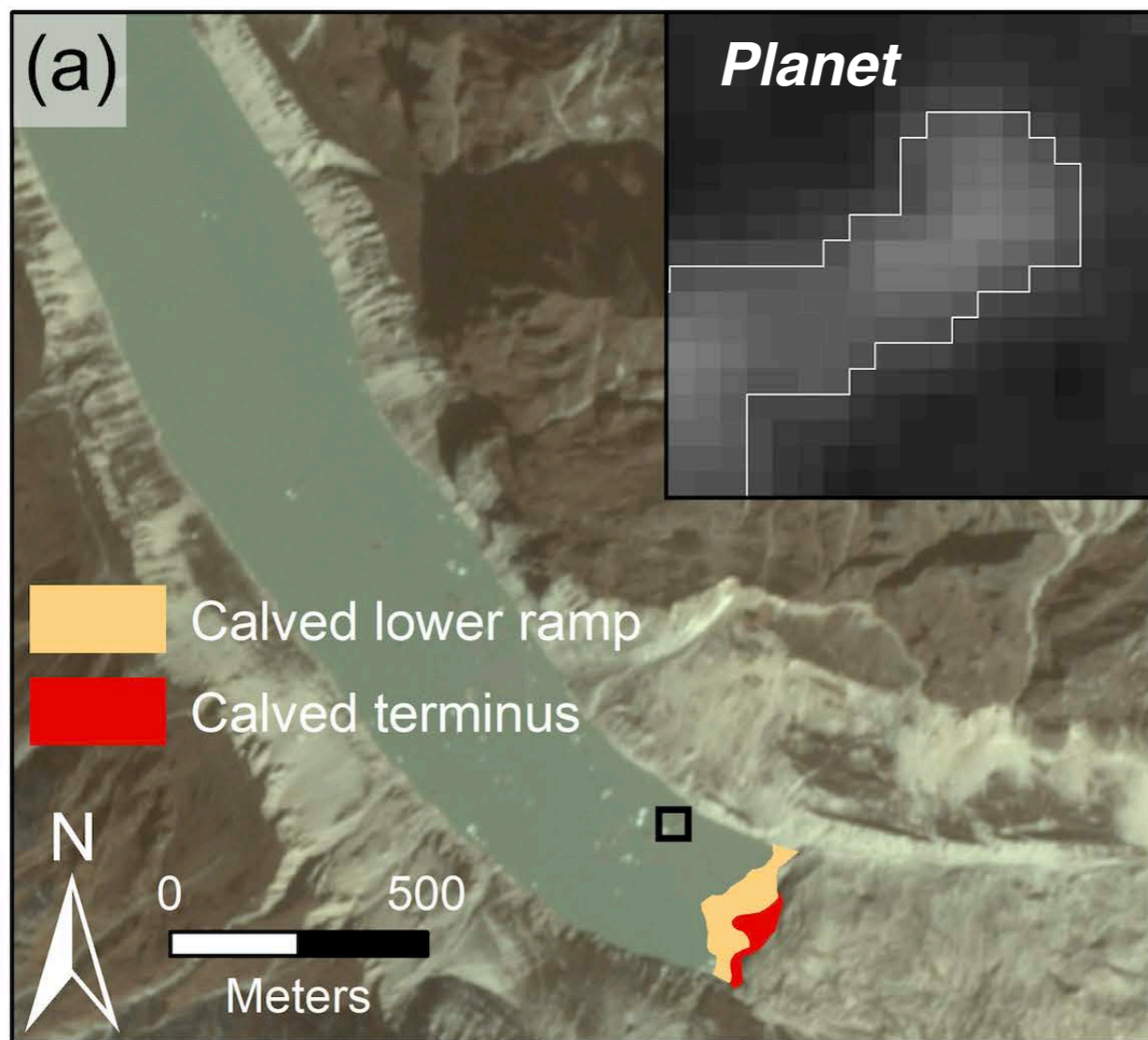


July- Glacier is present with metastable stratification

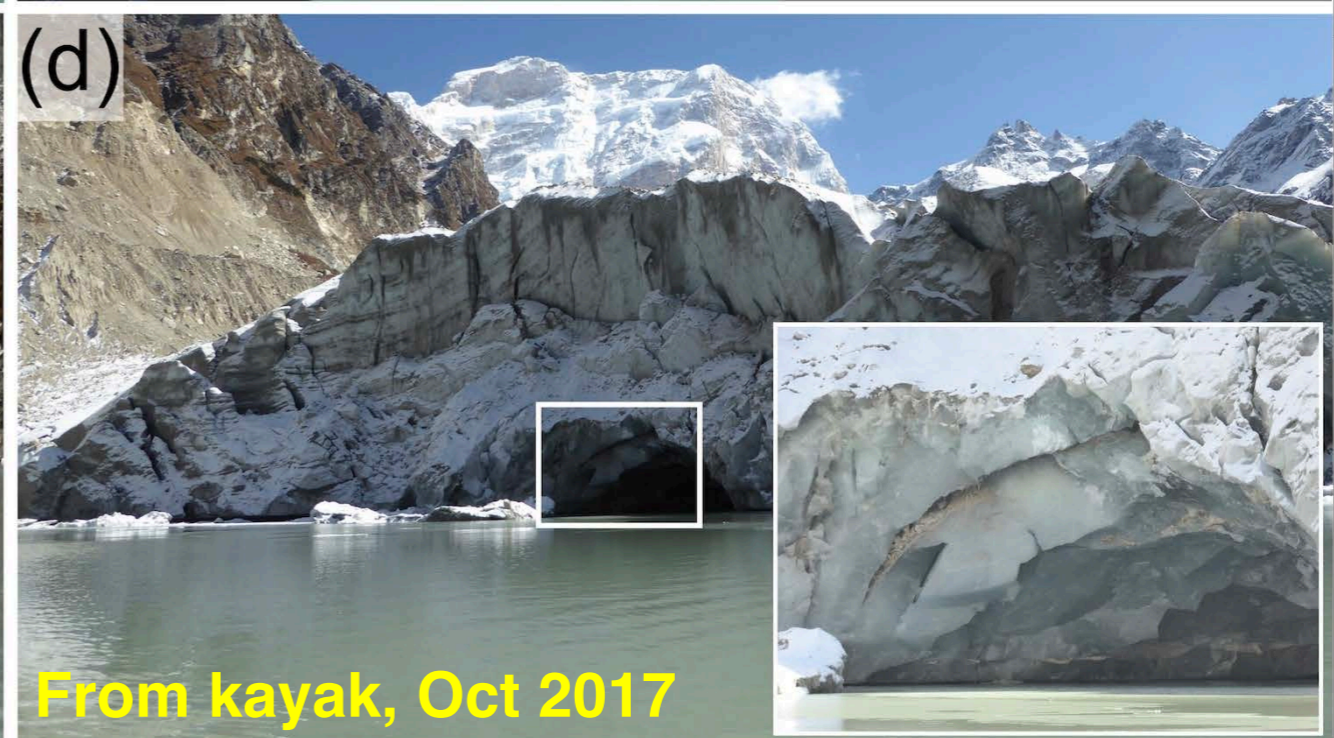
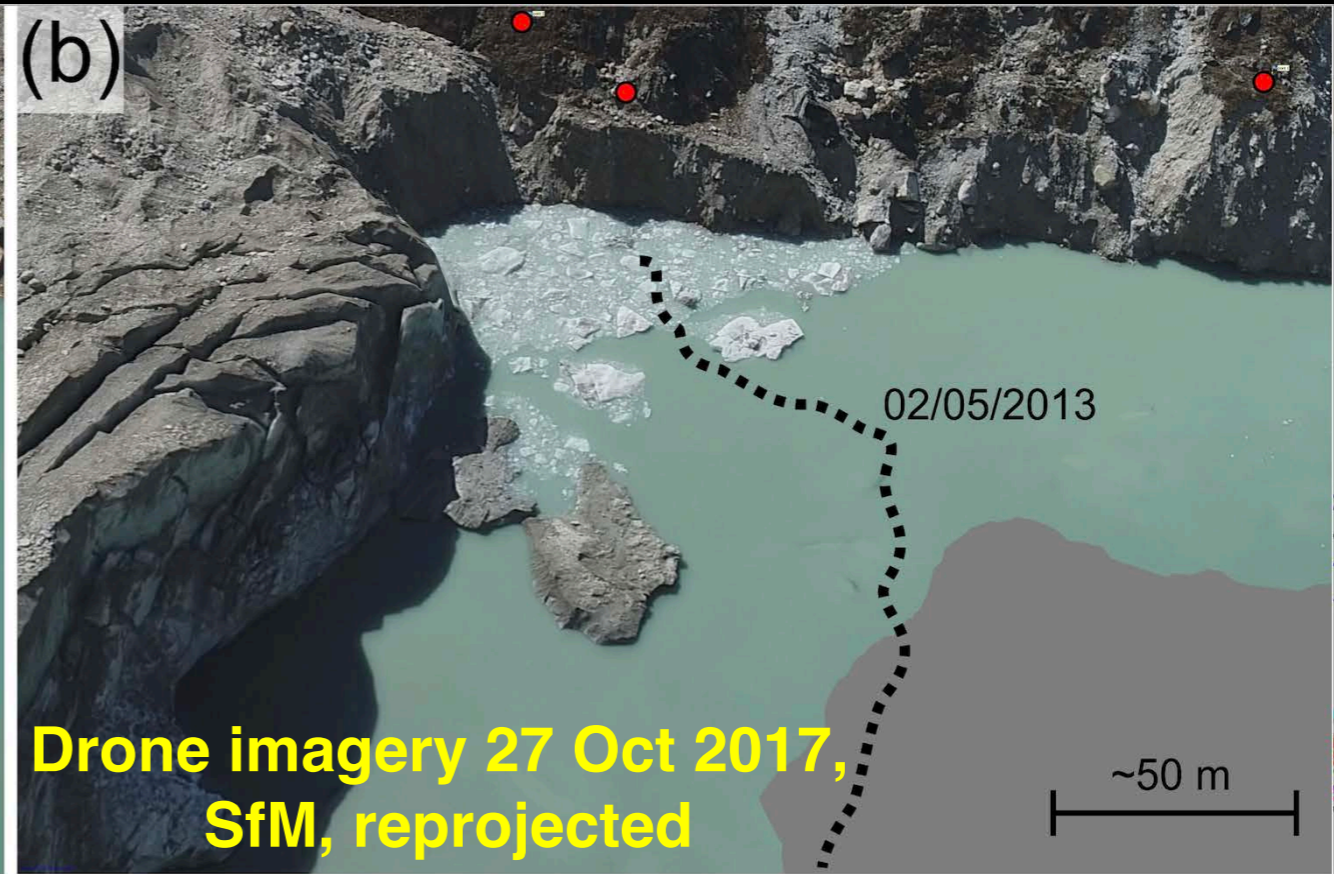


Wind-driven iceberg motion and forced convection mixes the lake down to iceberg roots

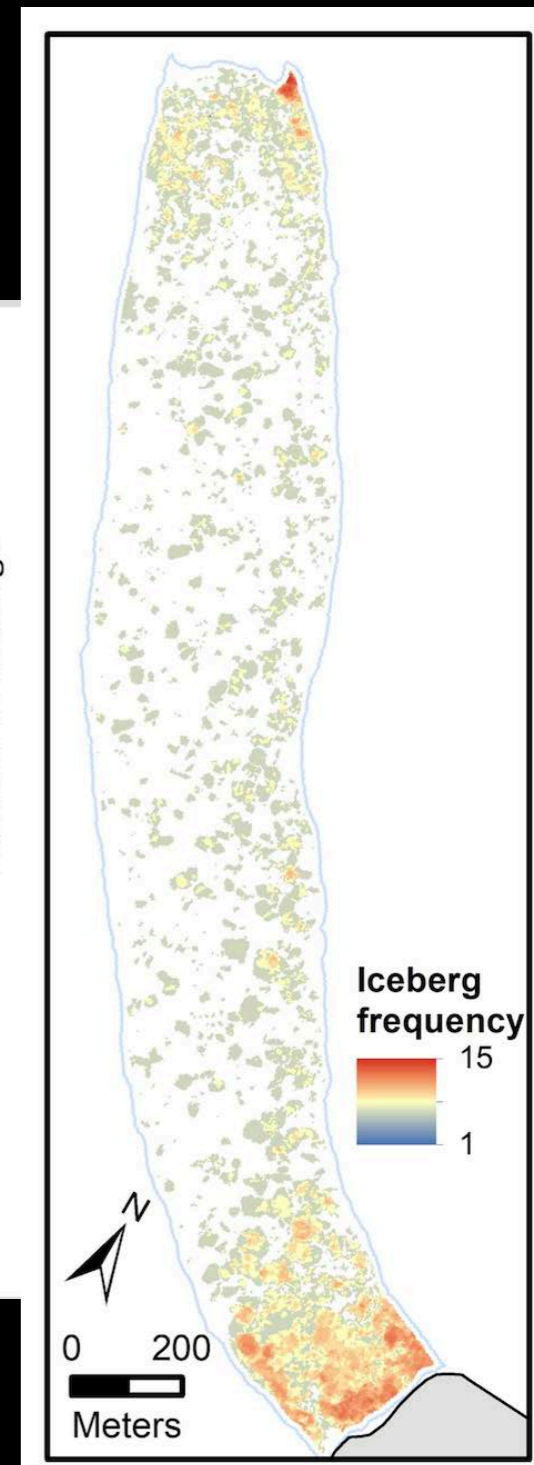
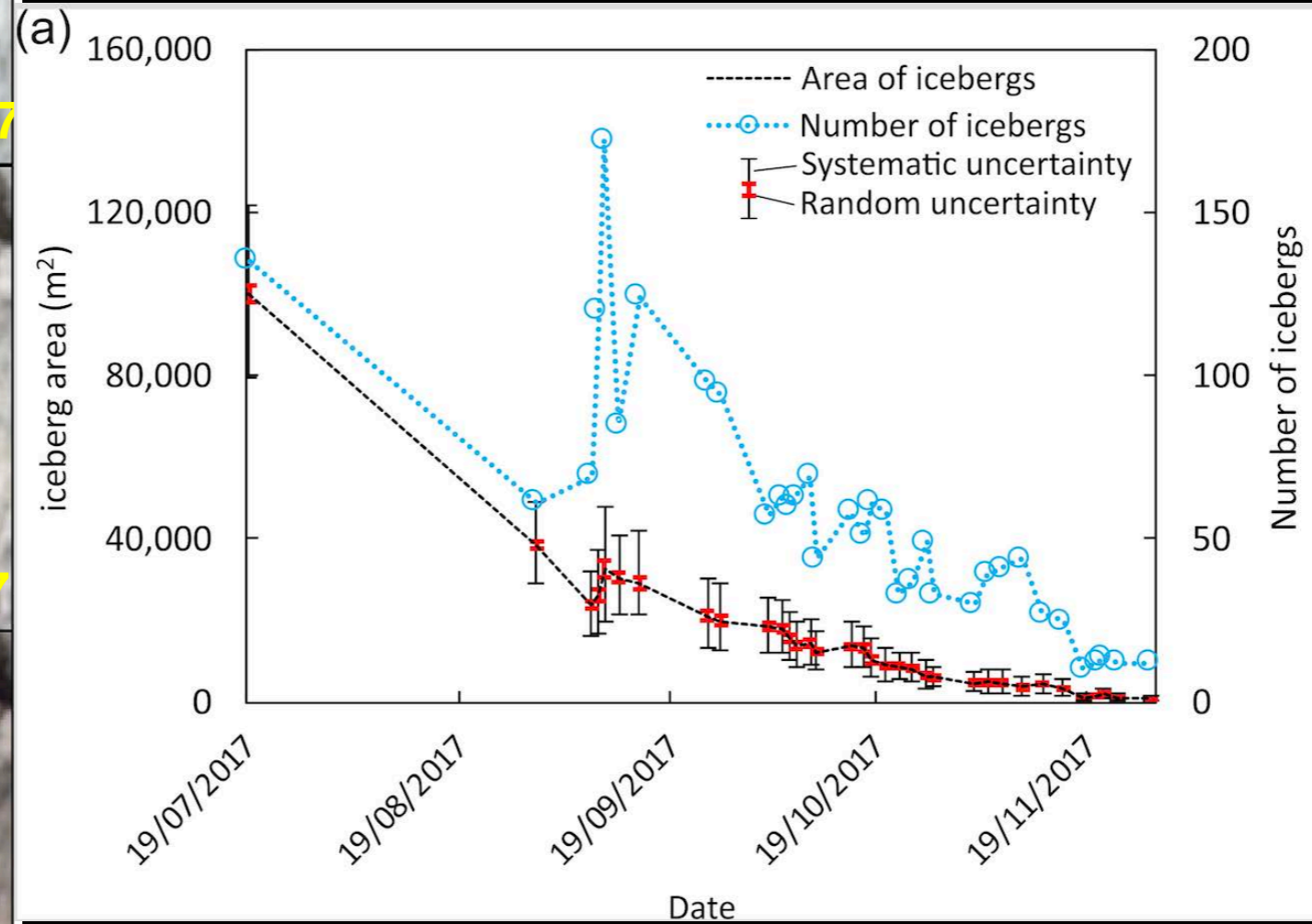
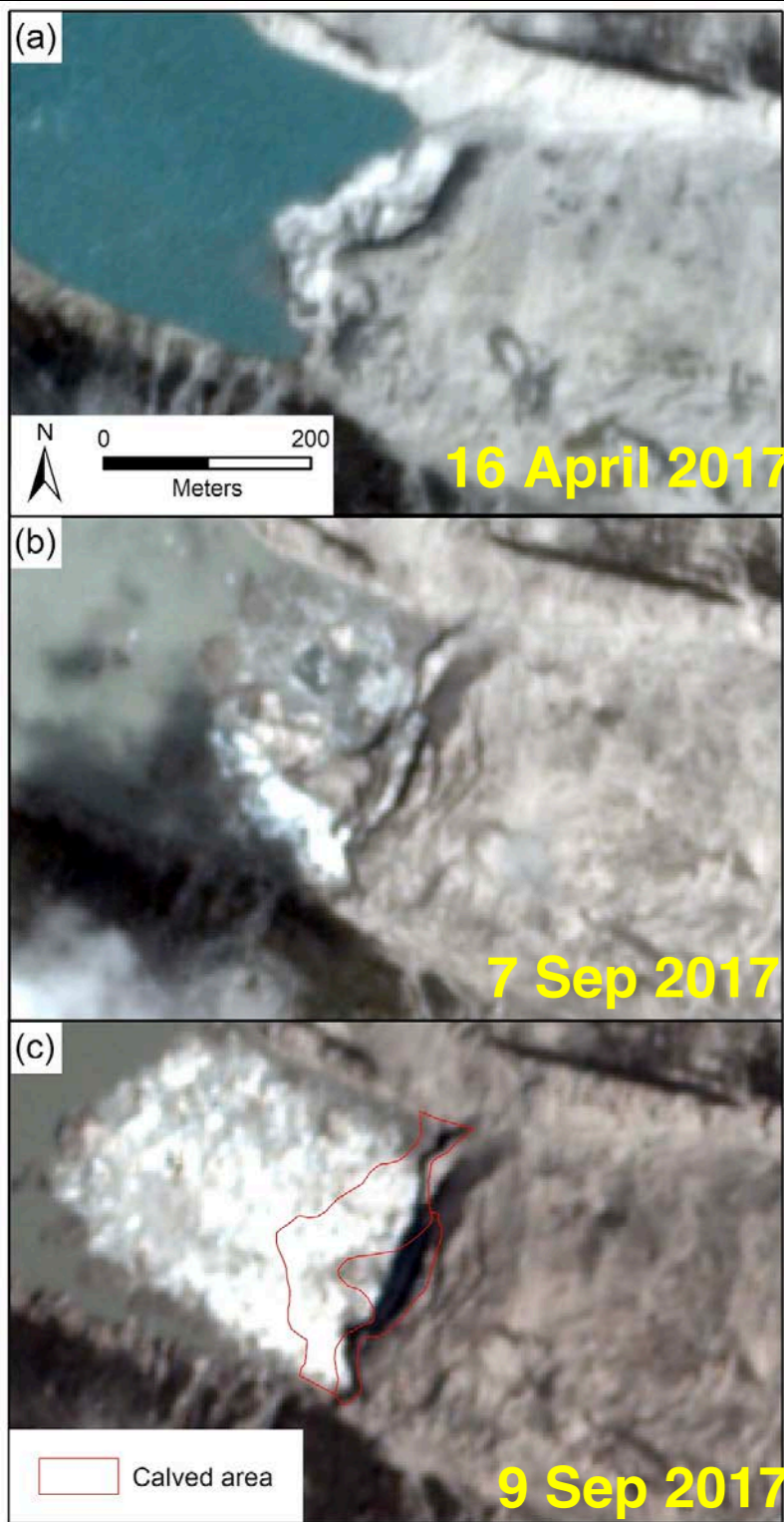




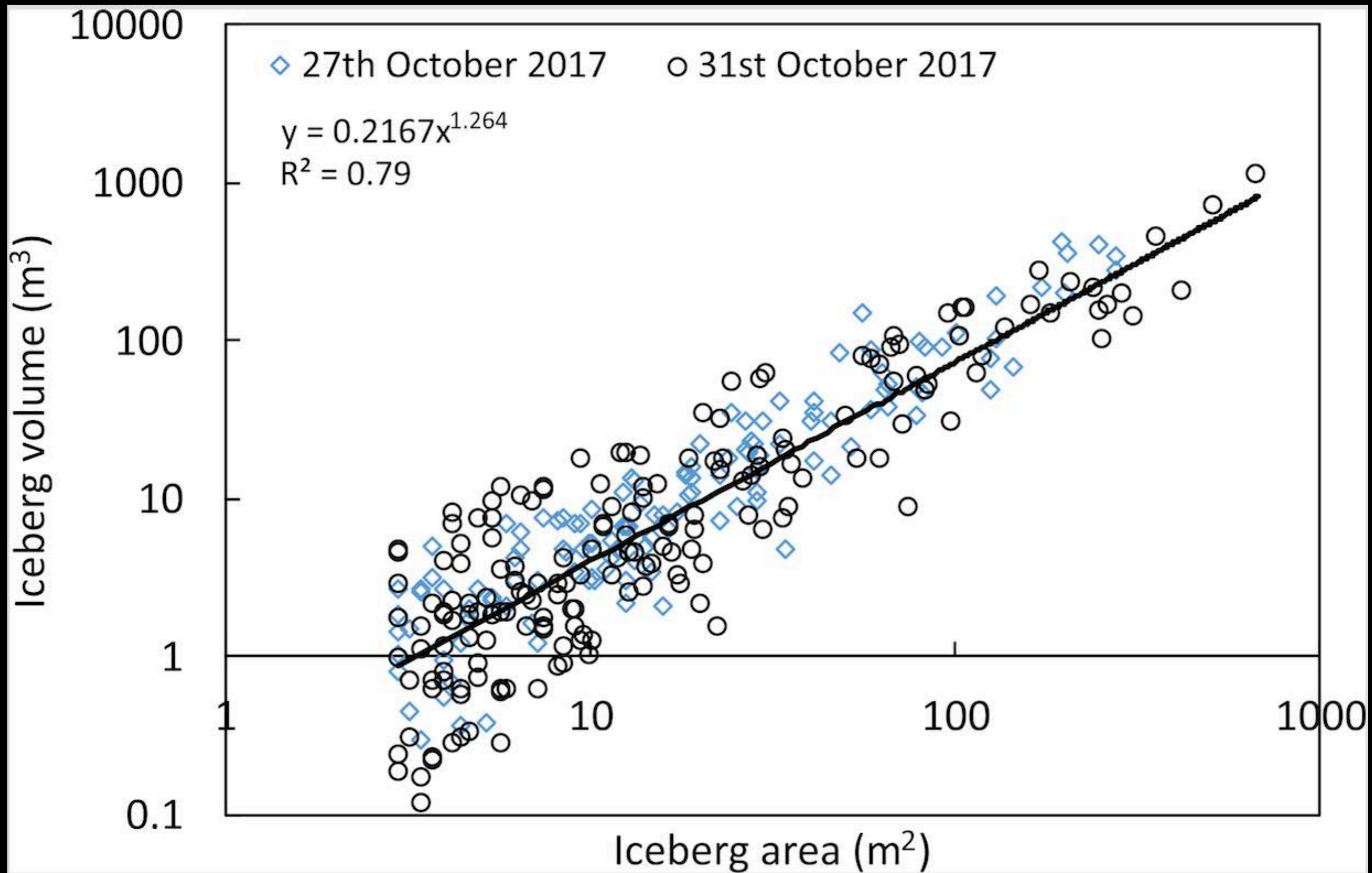
Thulagi Glacier/Lake



Thulagi Lake calving and icebergs, from *Planet* imagery



Thulagi Lake iceberg, area and above-surface volume, from drone imagery, SfM



Iceberg melting, figure of merit, I:

Thulagi Lake iceberg melting from the Sep 7-9 event. Ice mostly melted after 60 days.

The initial estimated **544,000 m³ of calved ice** = 4.99×10^8 kg
would require 333,000 J/kg to melt
= 1.66×10^{14} J.

How plausible is it that so much ice could melt in a few weeks?

MODEL 1: Heat comes from the deep lake thermal reservoir.

(1) Assume reservoir temperature = 4°C .

(2) Reservoir volume = $6.75 \times 10^7 \text{ m}^3 = 6.75 \times 10^{10} \text{ kg} = 6.75 \times 10^{13} \text{ g}$.

(3) Energy reservoir (above 0°C) = $2.7 \times 10^{14} \text{ Cal} = 1.13 \times 10^{15} \text{ J}$.

(4) CONCLUSION: The deep lake thermal reservoir is a factor of 7 greater than the energy needed to melt all that ice.

Iceberg melting, figure of merit, II:

Thulagi Lake iceberg melting from the Sep 7-9 event. Ice mostly melted after 60 days.

The initial estimated **544,000 m³ of calved ice** = 4.99×10^8 kg
would require 333,000 J/kg to melt
= 1.66×10^{14} J.

How plausible is it that so much ice could melt in a few weeks?

MODEL 2: Heat comes from concurrent solar heating.

(1) Thulagi Lake area = 900,000 m².

(2) Assume 450 W/m² absorption for 4 hours daily for 60 days.

(3) Energy absorption = 3.1×10^{14} J.

(4) CONCLUSION: Concurrent solar heating would add twice as much energy as needed to melt all the calved ice.

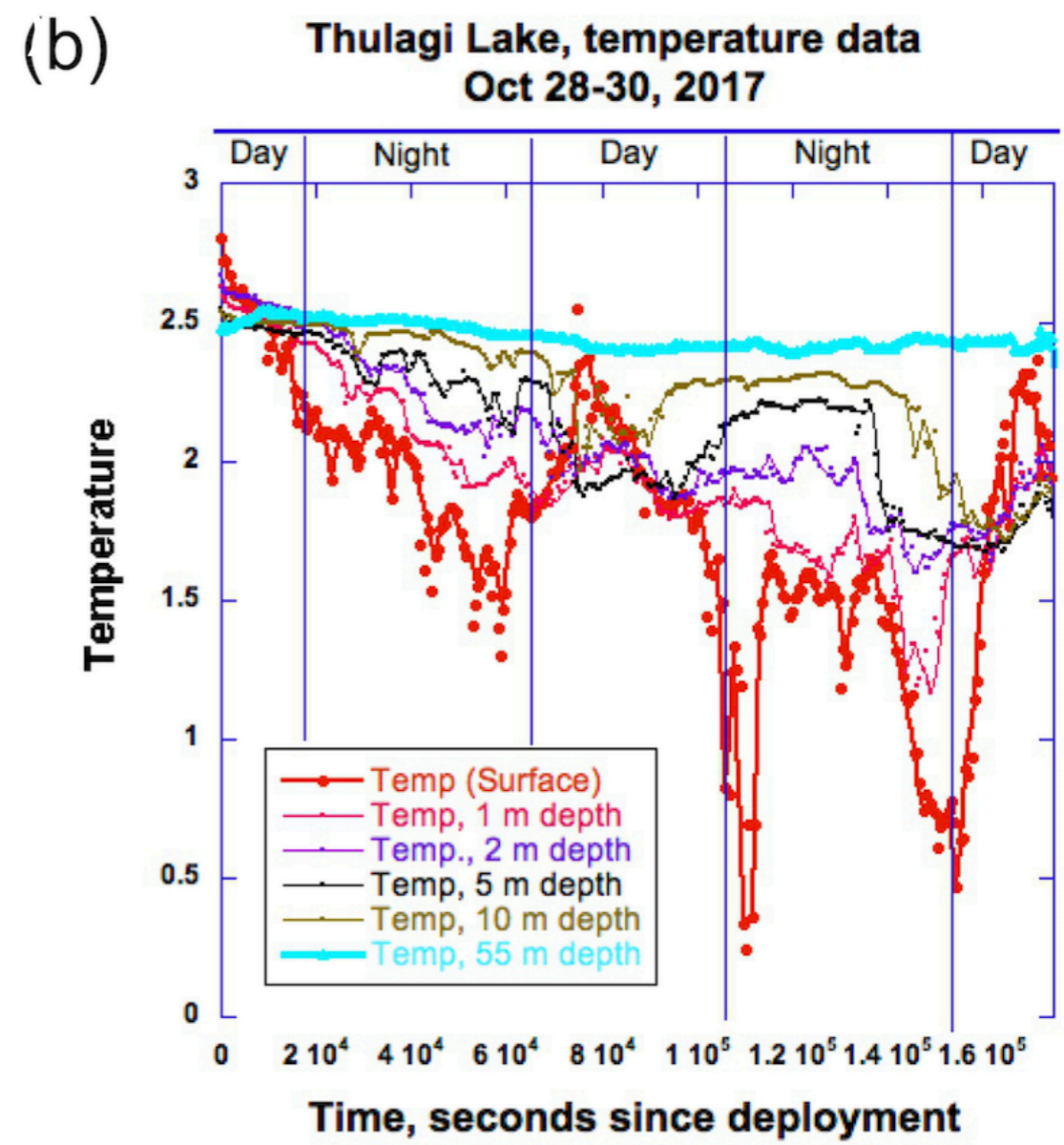
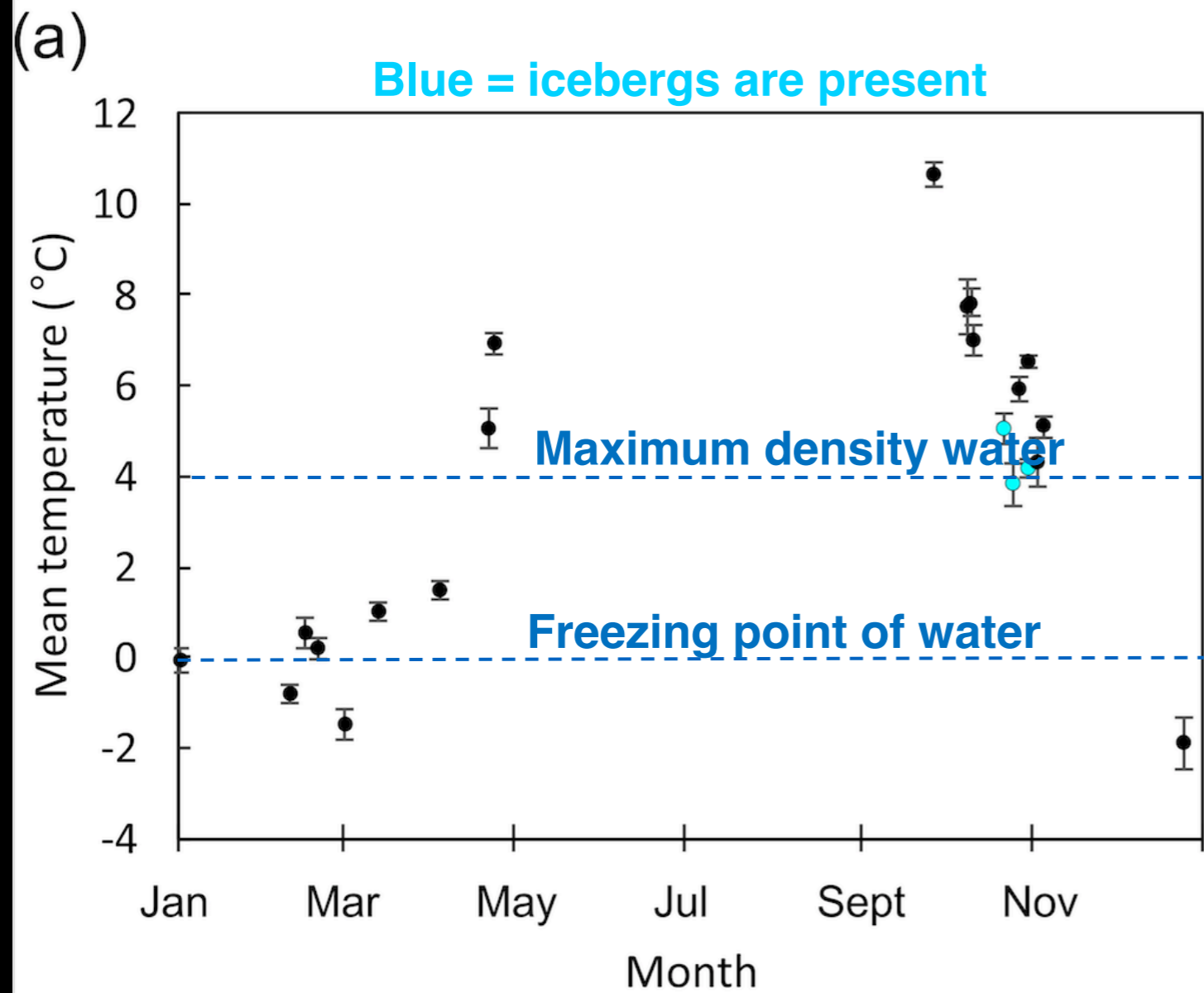
So there is plenty energy to melt the ice.

Where does the rest of the energy go?

So there is plenty energy to melt the ice.

Where does the rest of the energy go?

WE WILL TELL YOU NEXT YEAR!



Thulagi Lake-Light-Oct2017-ProfileC

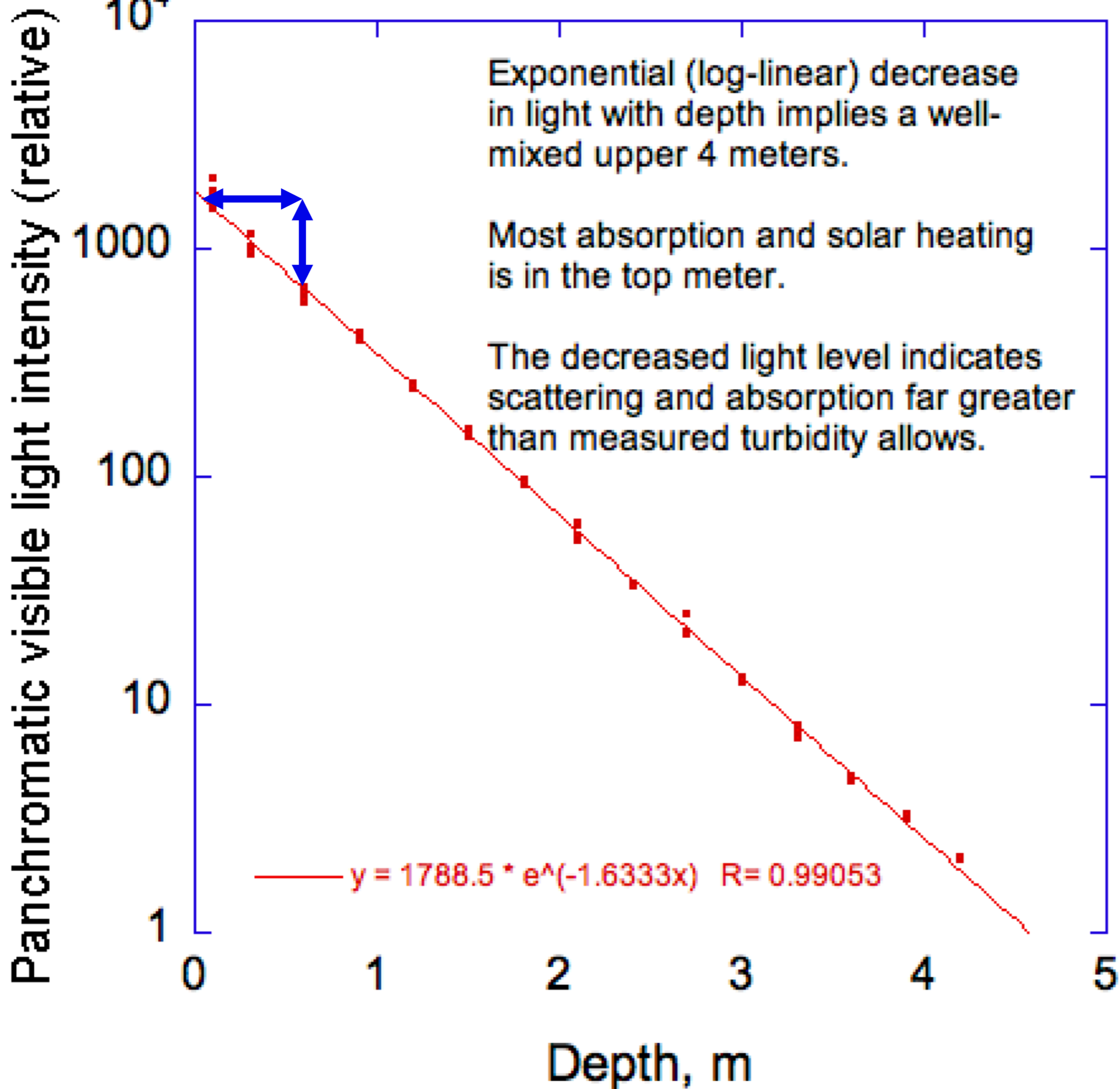


Figure of merit heating calculation.

Half of the incident unreflected sunlight is absorbed within the top 60 cm.

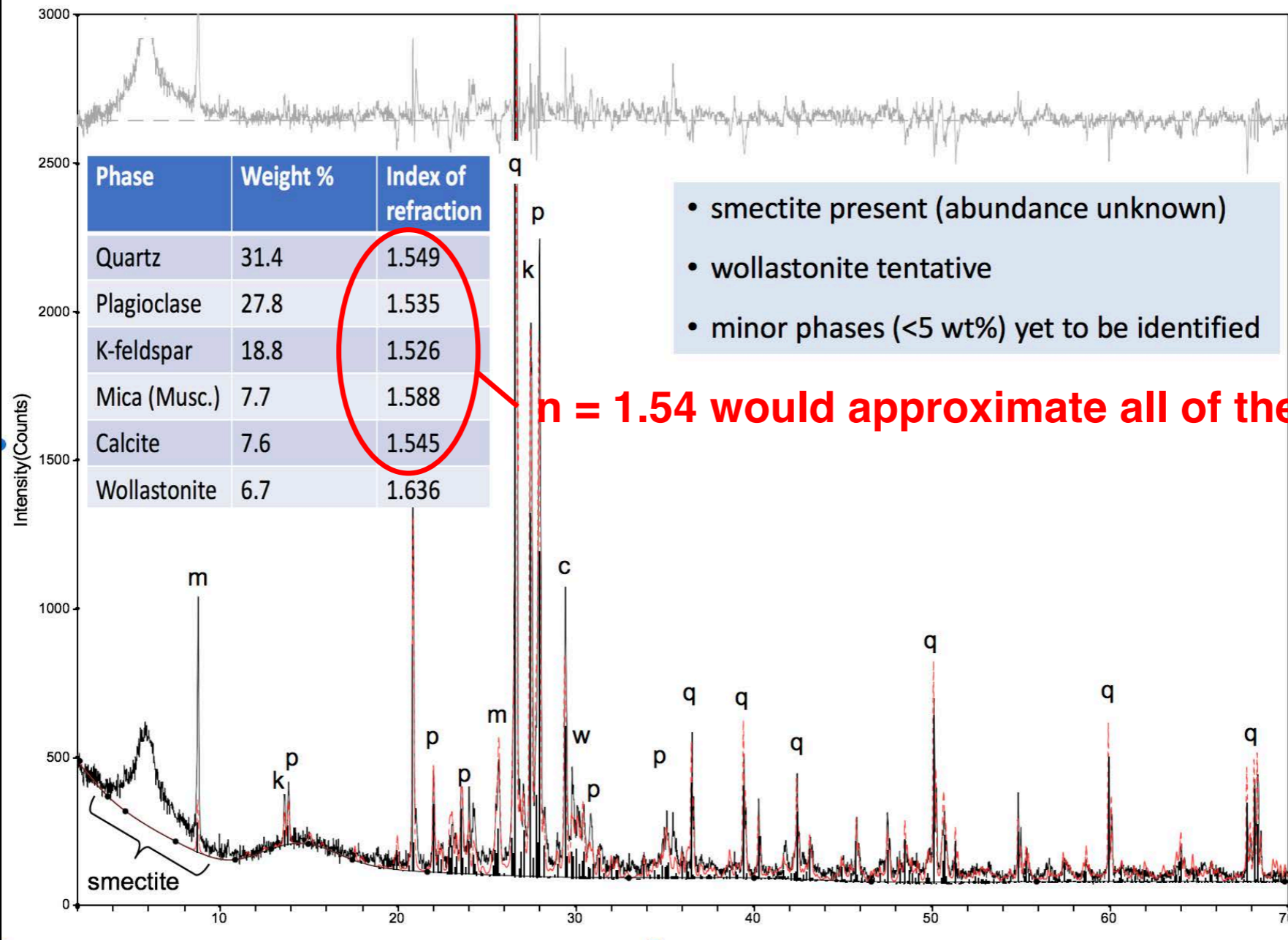
Incident refracted beam energy near the surface = 800 Watts/m²
 X 6 hours = 17.3 x 10⁶ J.

Half is absorbed in upper 60 cm = 8.65 x 10⁶ J
 = 2.06 x 10⁶ Calories.

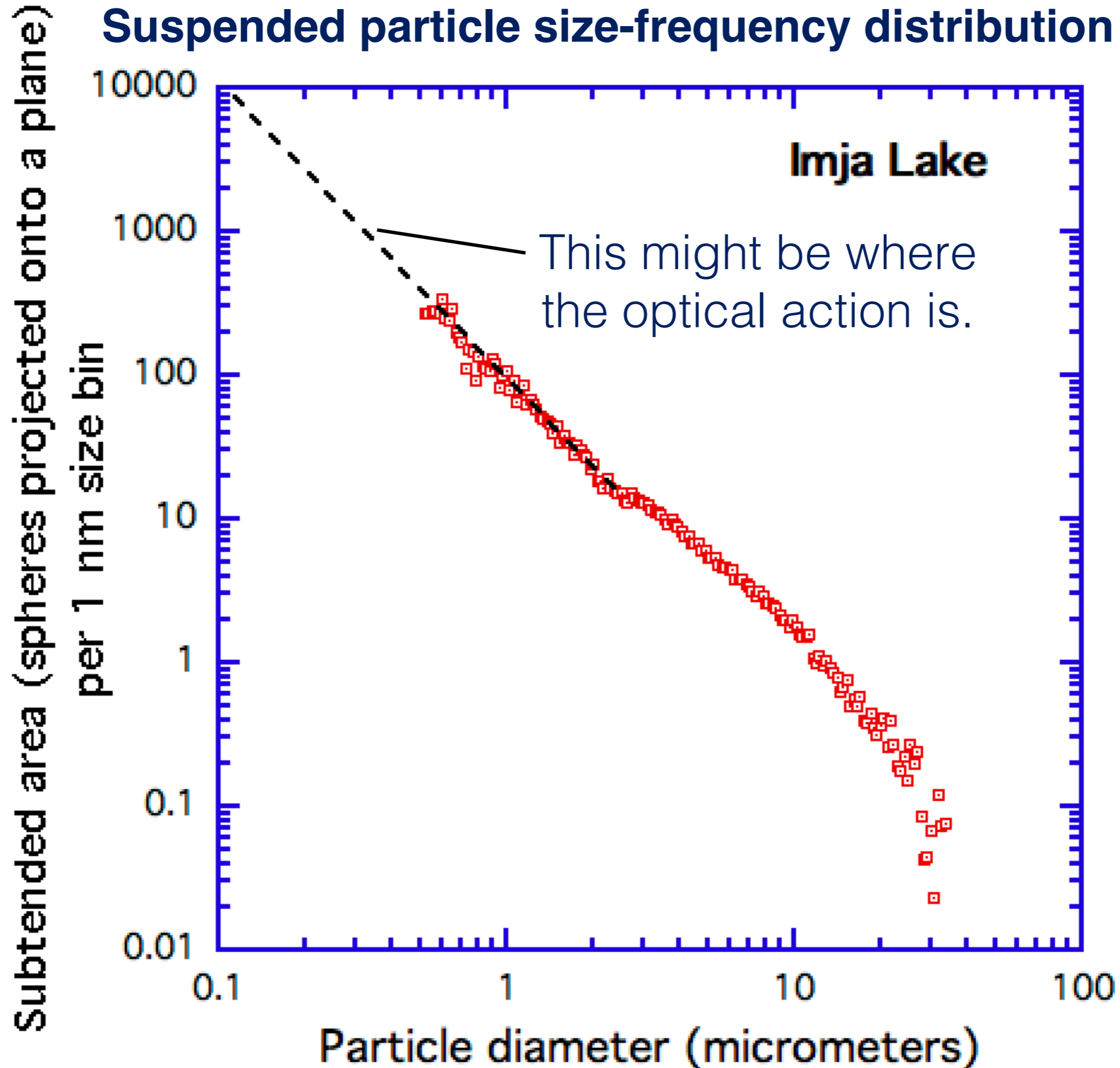
2.06 x 10⁶ C/6x10⁵ cm³ = 3.4 C/cm³.

Enough to raise the temperature by 3.4 degrees.

XRD results and optical index n for Thulagi Lake beach sand



Suspended particle size-frequency distribution



Glacier Lake Assisted Melting (GLAM)

GLAM BioLith RT

Lakes Bio-Lithological Optical/RT Modeling, Water Components Concentration Retrieval/Mapping Effort, and Lake Temperature Distribution Simulations

Overview

- **IOPs, Physics occurring in the water:** Absorption coefficient, backscattering coefficient, extinction coefficient, and single scattering albedo are **Inherent Optical Properties (IOPs)** of a water body. They depend only on the medium composition
- **AOPs, what a satellite sees:** Radiance reflectance, **remote sensing reflectance**, and irradiance reflectance are **Apparent Optical Properties (AOPs)** of a water system. They also depend on the incoming light's geometric distribution
- **Stuff in the water (as well as the water itself):** Physical components such as **Phytoplankton, detritus**, colored dissolved organic matter, and **inorganic particles** that are present in the water body influence the IOPs; and hence the AOPs of the water column
- Water components concentrations, IOPs, and AOPs are related through mathematical equations called **Bio-Lithological Optical/Radiative Transfer (RT) models** [1]. **This is what we are building.**
- ***This is why we are building it:*** The Radiative Transfer affects how much and where solar energy is absorbed by the water, hence heating of the lake:

Goals

- Development of Bio-Lithological Optical/RT models for glacier lakes water (**forward modeling**)
- **Validation** of Bio-Lithological Optical/RT models via sensitive analysis and in-situ water samples (from Imja lake and Thulagi lake, Himalaya)
- Estimating the concentration of the physical components that are deposited into the lake due to glacier dynamics via Bio-Lithological Optical/RT model inversion (**inverse modeling**)
- Using the retrieved concentrations to run **temperature distribution simulation** for the lakes of interest

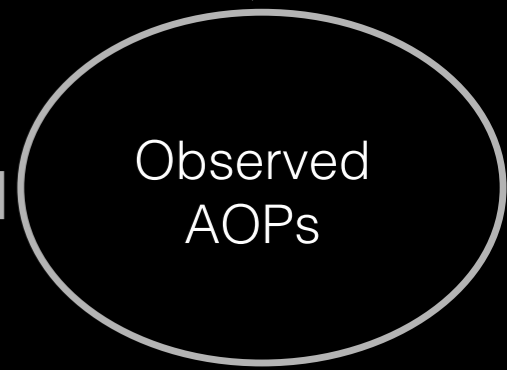
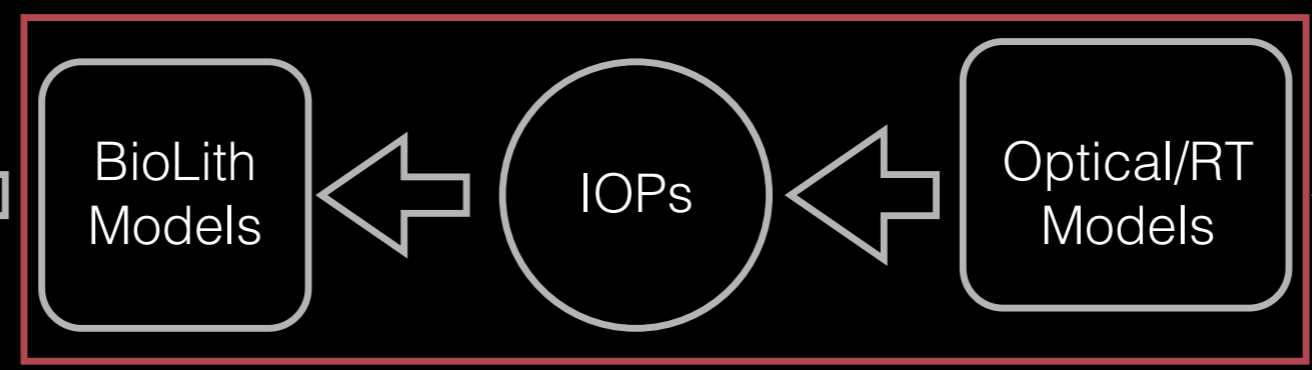
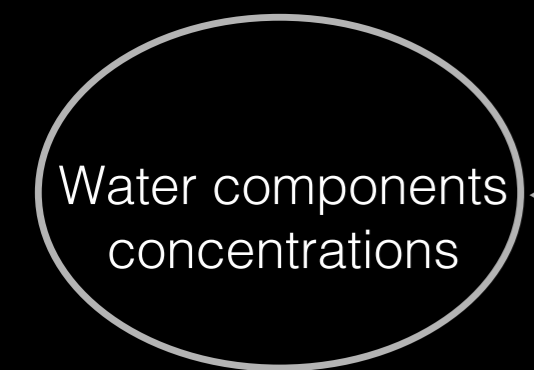
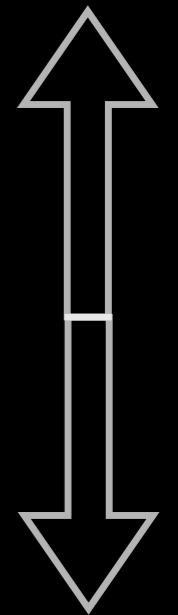
Forward Modeling



Stuff in the water

What a satellite sees

Convergence



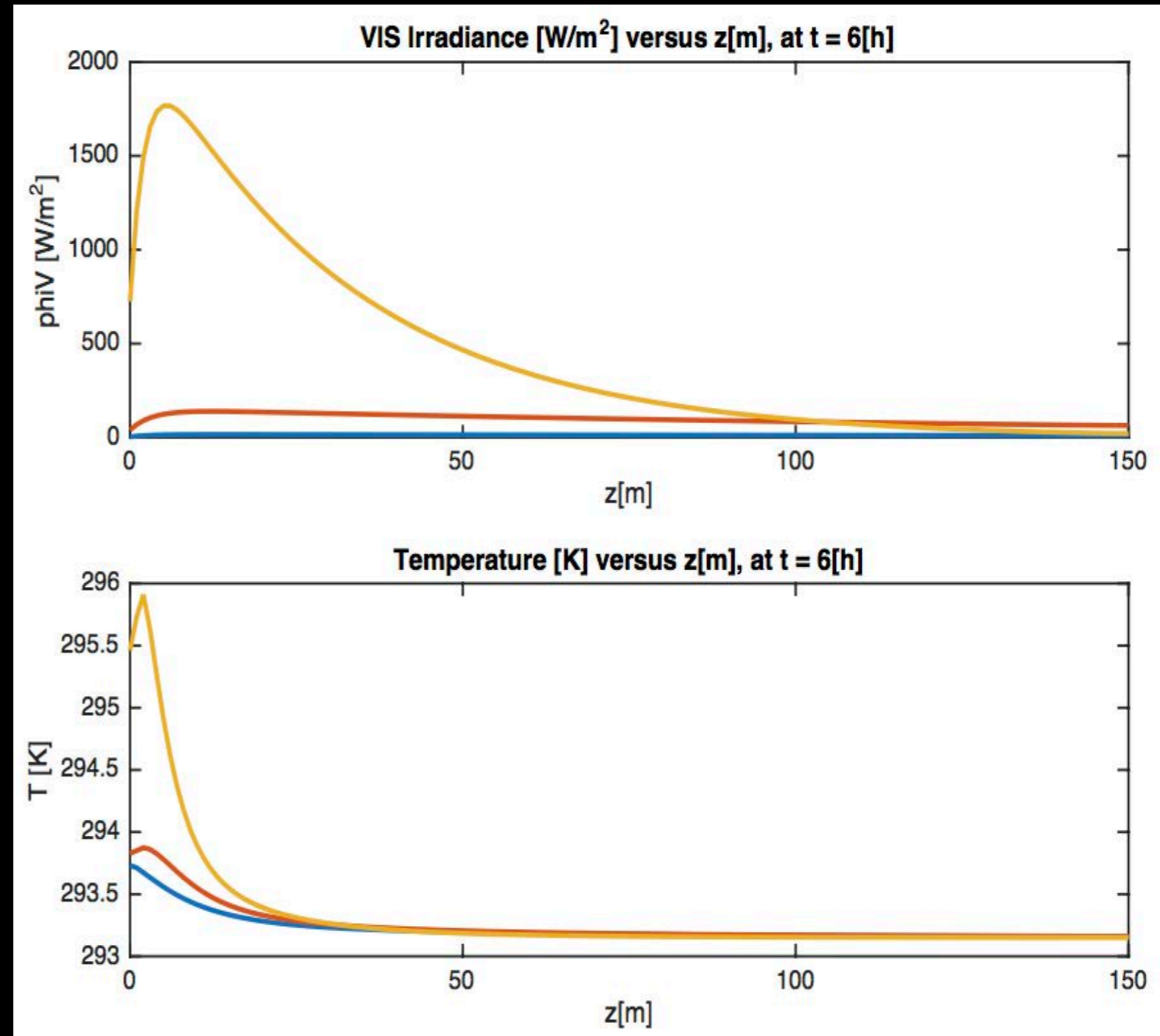
Stuff in the water

What a satellite sees

Inverse Modeling

Achievements

- **GLAM BioLith RT:** Matlab code for Bio-Lithological Optical/RT forward and inverse modeling, based on [1,2,3,4]
- **Radiative Heat Transfer Equations (RHTEs):** Matlab code for spherically visible irradiance and the temperature distributions in lake water simulations [8]



Coming next

- Water component concentrations retrieval via **GLAM-BioLith RT** inversion (classical and Bayesian)
- Temperature distribution simulation by using **RHTEs** with retrieved water component concentrations as inputs

References

- [1] Claudia Giardino, Gabriele Candiani, Mariano Bresciani, Zhongping Lee, Stefano Gagliano, Monica Pepe (2011), BOMBER: A tool for estimating water quality and bottom properties from remote sensing images, Computers & Geosciences
- [2] Claudia Giardino, Mariano Bresciani, Emiliana Valentini, Luca Gasperini, Rossano Bolpagni, Vittorio E. Brando (2014) Airborne hyperspectral data to assess suspended particulate matter and aquatic vegetation in a shallow and turbid lake, Remote Sensing of Environment
- [3] Peter Gege (2013) WASI-2D: A software tool for regionally optimized analysis of imaging spectrometer data from deep and shallow waters, Computers & Geosciences
- [4] Peter Gege (2015) The Water Color Simulator WASI, User manual for WASI version 4.1
- [5] Zhongping Lee, Kendall L. Carder, Curtis D. Mobley, Robert G. Steward, and Jennifer S. Patch (1998), Hyperspectral remote sensing for shallow waters. 1. A semianalytical model, Optical Society of America
- [6] Zhongping Lee, Kendall L. Carder, Curtis D. Mobley, Robert G. Steward, and Jennifer S. Patch (1999), Hyperspectral remote sensing for shallow waters: 2. Deriving bottom depths and water properties, by optimization, Optical Society of America
- [7] A. Albert, C.D. Mobley (2003), An analytical model for subsurface irradiance and remote sensing reflectance in deep and shallow case-2 waters, Optical Society of America
- [8] Enrico Schiassi, Roberto Furfaro, and Domiziano Mostacci (2016), Bayesian inversion of coupled radiative and heat transfer models for asteroid regoliths and lakes, Radiation Effects and Defects in Solids, 171:9-10, 736-745, DOI: 10.1080/10420150.2016.1253091