

Satellite Derived Bathymetry (SDB) Using OLI/MSI Based-On Physics-Based Algorithm

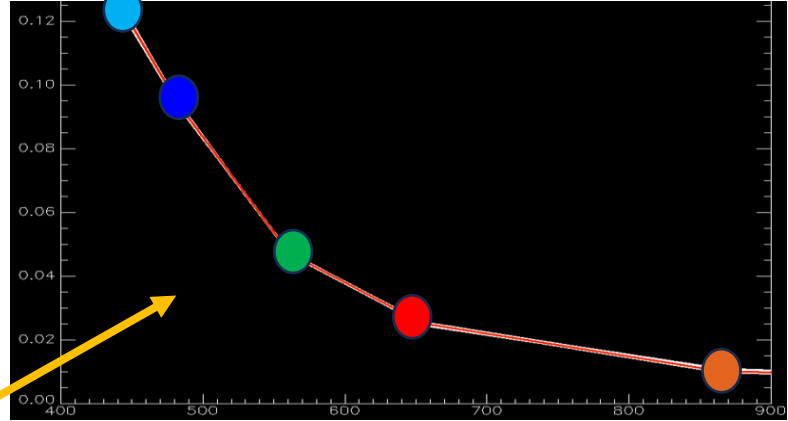
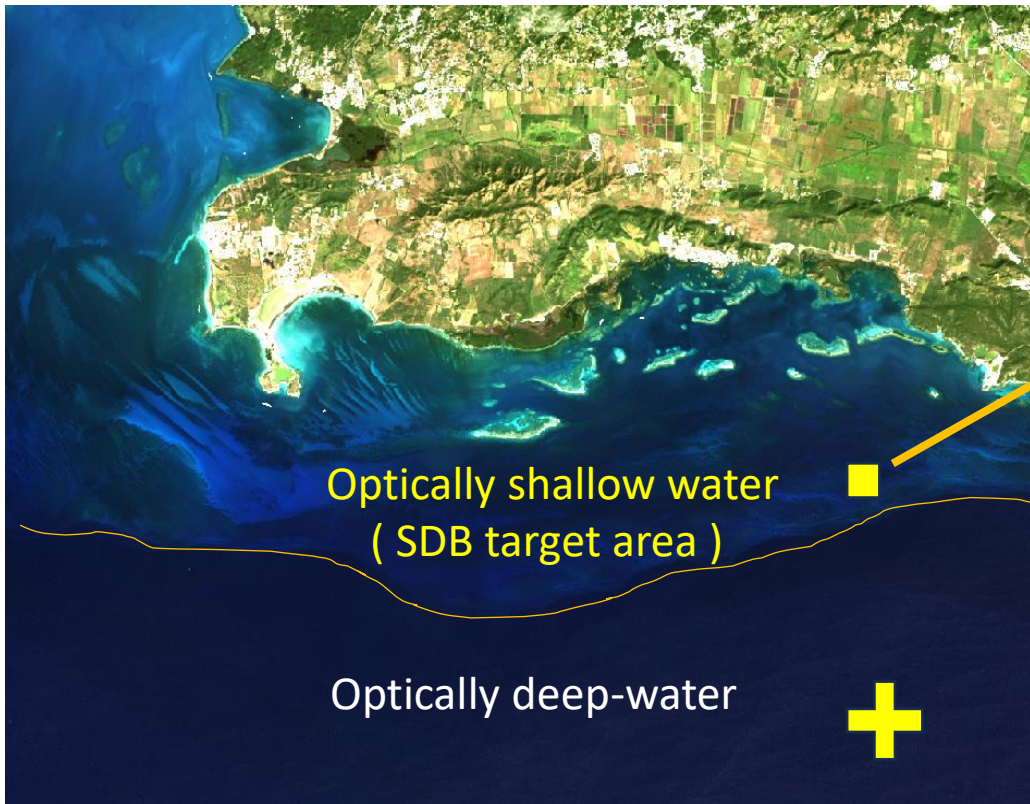
Minsu Kim, Ph D. Chief Scientist, KBR

KBR, contractor to the U.S. Geological Survey (USGS)
Earth Resources Observation and Science (EROS) Center, Sioux Falls, SD 57198, USA.
Work performed under USGS Contract [G15PCXXXXX].

Jeff Danielson, USGS EROS

ICESat-2 Bathymetry Workshop
Wilmington, NC 2025



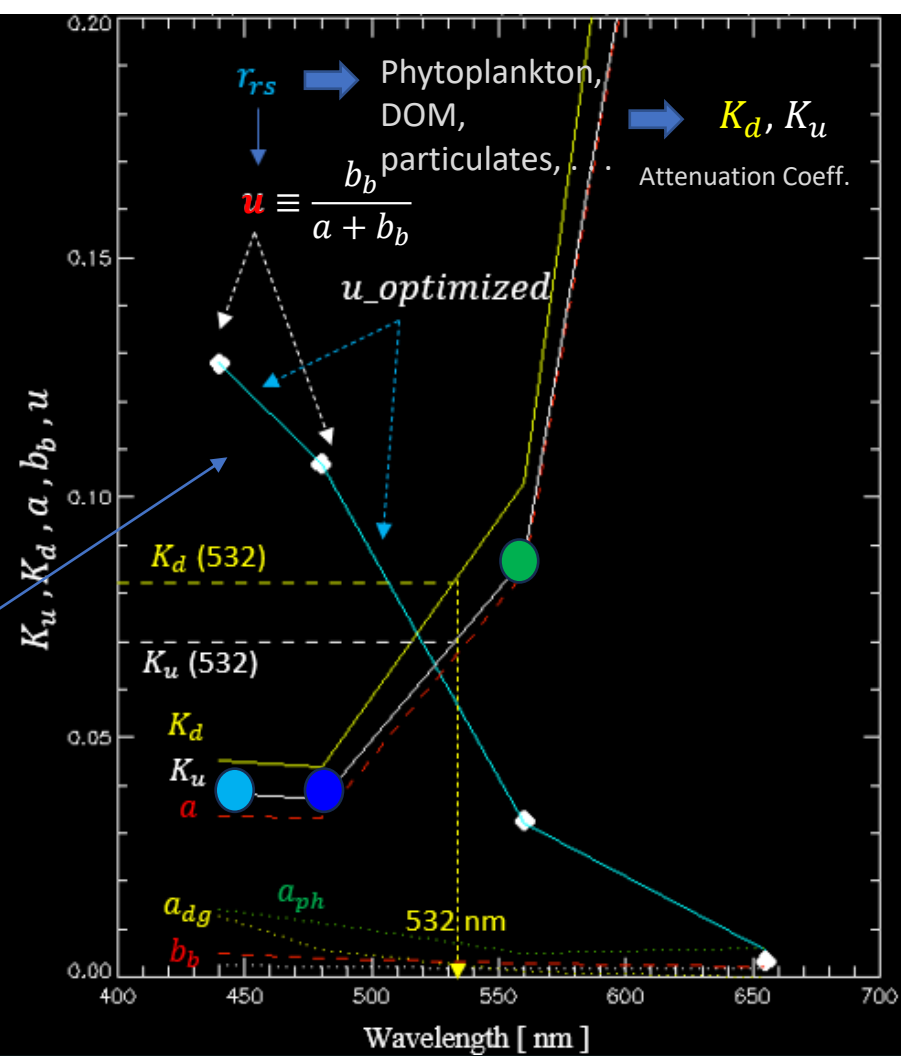
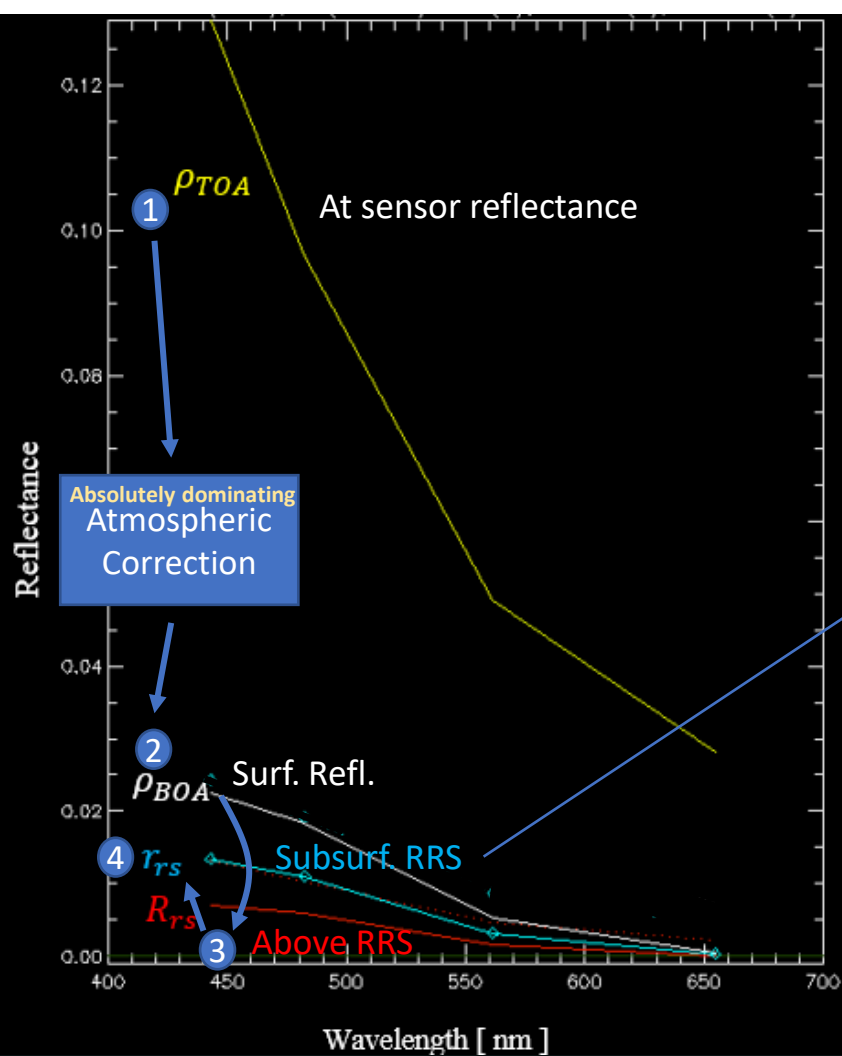


9 + unknowns >> 4 knowns
of bands

Unknowns :

1. Atmospheric (Aerosol, ozone, water vapor, . . .)
 2. Water (Phytoplankton, DOM, particulates, . . .)
 3. Depth
 4. Bottom (How many bottom types ?)
- 8 9 ...





Article

Physics-Based Satellite Derived Bathymetry (SDB) using Landsat OLI Image

Minsu Kim ^{1,*}, Jeff Danielson ², et al.

$$\rho_{TOA} = \frac{\pi \cdot L_{TOA} \cdot D_{ES}^2}{\cos SZA \cdot F_0}$$

$$a_{dg}(\lambda) = a_{dg}(\lambda_0) \cdot \exp[-\dots]$$

$$e(\lambda; \mathbf{P}) = [u_0(\lambda) - u_1(\lambda) \cdot \dots]$$

Atmospheric Optics
Atmospheric R.T.
Atmospheric Correction



Hydro (Ocean) Optics
Ocean R.T.
Bio-optical Inversion

$$\rho_b = \pi \cdot r_{rs}(z) \cdot \exp[-(K_u + K_d) \cdot z]$$

$$\Sigma = (\rho_b - \bar{\rho}_b) \cdot (\rho_b - \bar{\rho}_b)^T$$

$$J(\Phi_A; \lambda) \equiv \frac{\partial u}{\partial \Phi_A} = \frac{\partial u}{\partial a} \cdot \frac{\partial a}{\partial A} \cdot \frac{\partial A}{\partial \Phi_A} = \exp[-\dots]$$

$$J(\Phi_B; \lambda) \equiv \frac{\partial u}{\partial \Phi_B} = \frac{\partial u}{\partial b_b} \cdot \frac{\partial b_b}{\partial B} \cdot \frac{\partial B}{\partial \Phi_B} = \left(\frac{\lambda_0}{\lambda}\right)^n \cdot \frac{a(\lambda)}{(a(\lambda) + b_b(\lambda))^2} \cdot \frac{B_{min} - B_{max}}{2} \cdot \sin \Phi_B$$

$$J(\Phi_C; \lambda) \equiv \frac{\partial u}{\partial \Phi_C} = \frac{\partial u}{\partial a} \cdot \frac{\partial a}{\partial C} \cdot \frac{\partial C}{\partial \Phi_C} = 0.06 \cdot 0.65 \cdot C^{-0.35} \cdot a_c^*(\lambda) \cdot \frac{b_b(\lambda)}{(a(\lambda) + b_b(\lambda))^2} \cdot \frac{C_{max} - C_{min}}{2}$$

$$K_d = \frac{a + b_b}{\cos(SZA_w)}, \quad SZA_w = \sin^{-1}\left(\frac{\sin(SZA)}{n_w}\right)$$

$$K_u = \frac{a + b_b}{\cos(VZA_w)}, \quad VZA_w = \sin^{-1}\left(\frac{\sin(VZA)}{n_w}\right)$$

$$J(\Phi_S; \lambda) \equiv \frac{\partial r_{rs}}{\partial \Phi_S} = \frac{\partial r_{rs}}{\partial C_g} \cdot \frac{\partial C_g}{\partial \Phi_S} = \left(\frac{\rho_{bg}}{\pi}\right) \cdot \exp[-(K_u + K_d) \cdot z] \cdot \frac{G_{min} - G_{max}}{2} \cdot \sin \Phi_S$$

$$\frac{\partial r_{rs}}{\partial z} = \frac{\partial r_{rs}}{\partial z} \cdot \frac{\partial z}{\partial \Phi_z} = (K_u + K_d) \cdot \left(r_{rs,\infty} - \left(\frac{\rho_b}{\pi}\right)\right) \cdot \exp[-(K_u + K_d) \cdot z] \cdot \frac{z_{min} - z_{max}}{2} \cdot \sin \Phi_z$$

$$J(\Phi_S; \lambda) = \left(\frac{\rho_{bs} - \rho_{bg}}{\pi}\right) \cdot \exp[-(K_u + K_d) \cdot z] \cdot \frac{S_{min} - S_{max}}{2} \cdot \sin \Phi_S$$

PB-SDB
(Physics Based SDB)

$$\rho_b / \pi \cdot \exp[-(K_u + K_d) \cdot z]$$

$$\rho_{bg}$$

SDB pixel-by-pixel

Atmospheric Correction

Optically Deep

$$u(\lambda) \leftarrow r_{rs} \leftarrow R_{rs} \leftarrow SR \leftarrow TOAR$$



Each Pixel

Ocean Optical Inversion

$$r_{rs}(z) = r_{rs,\infty} \cdot \{1 - \exp[-(K_u + K_d) \cdot z]\} + (\rho_b / \pi) \cdot \exp[-(K_u + K_d) \cdot z]$$

Depth

3 unknowns

Nonlinear Optimization

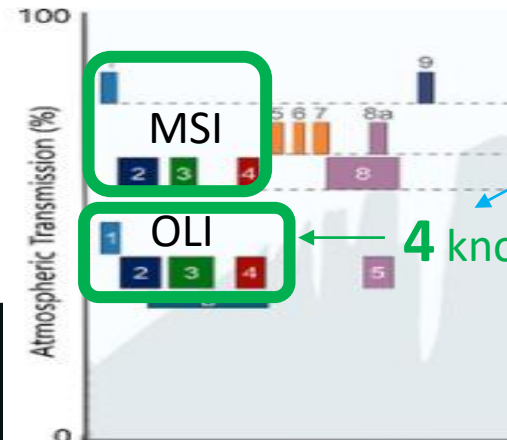
Bottom library

$$C_s \rho_{bs} + C_g \rho_{bg}$$

Sand

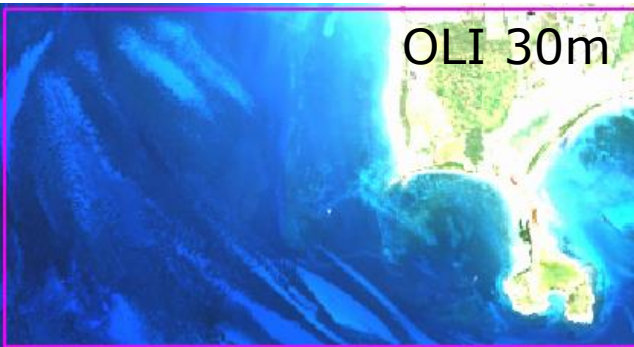
Grass

4 known bands

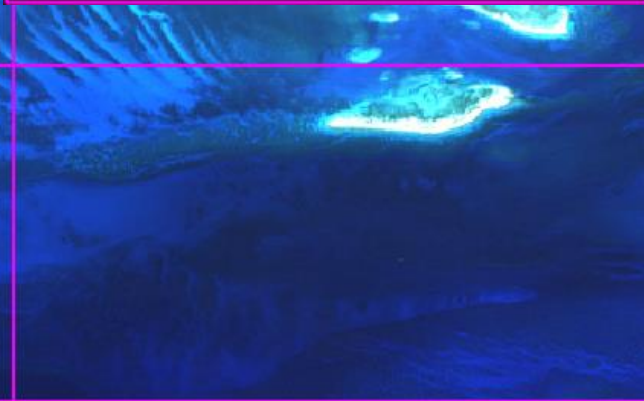
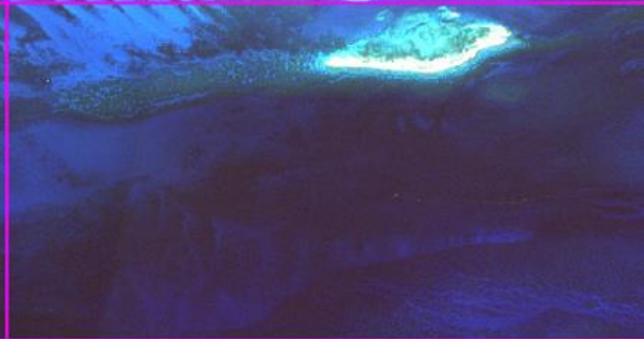
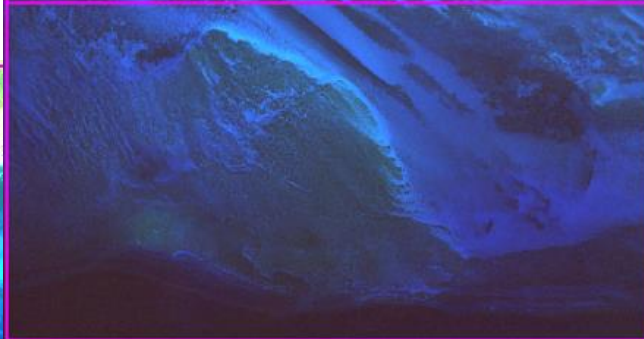


Dec 23, 2025, same day

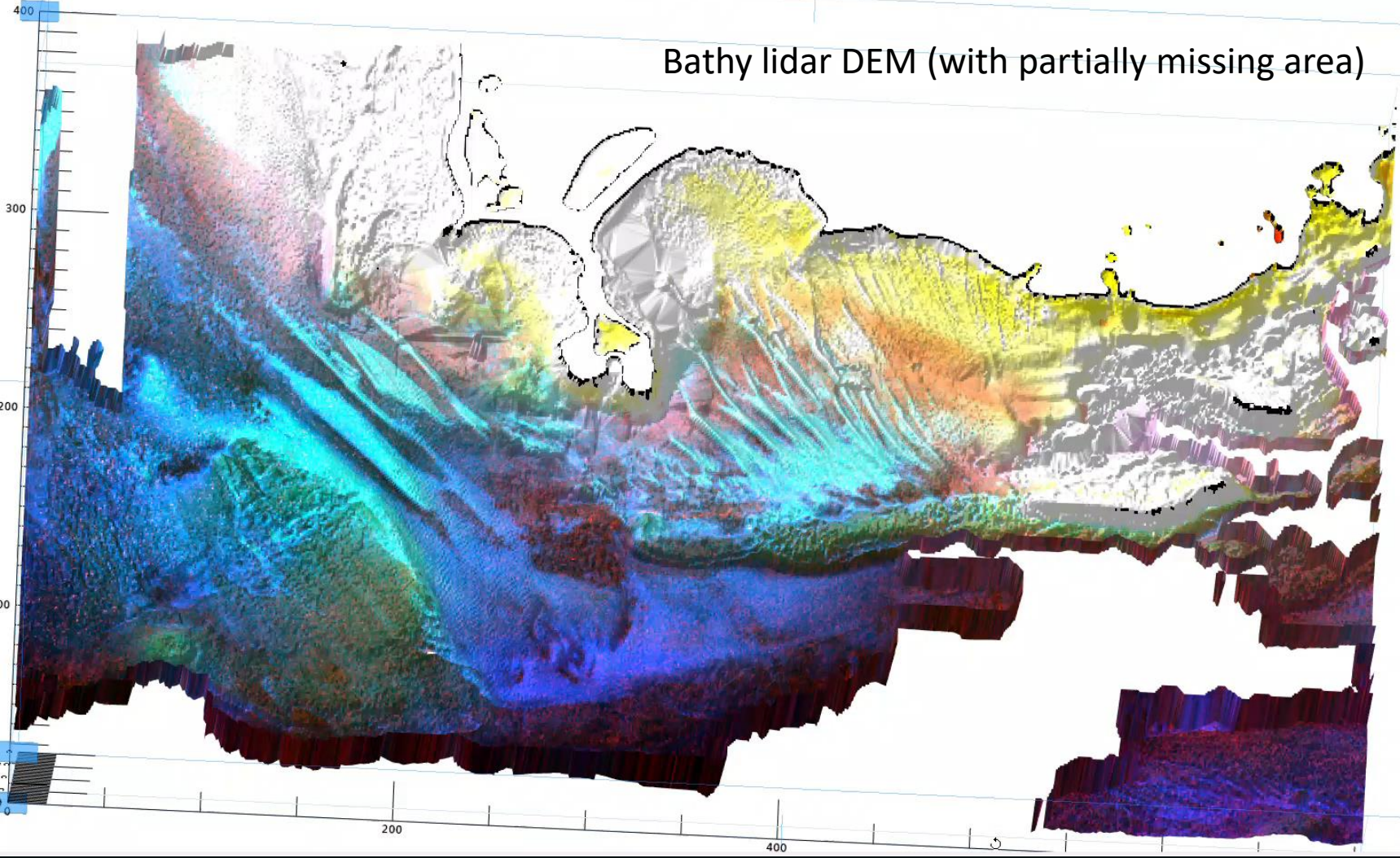
MSI 10m



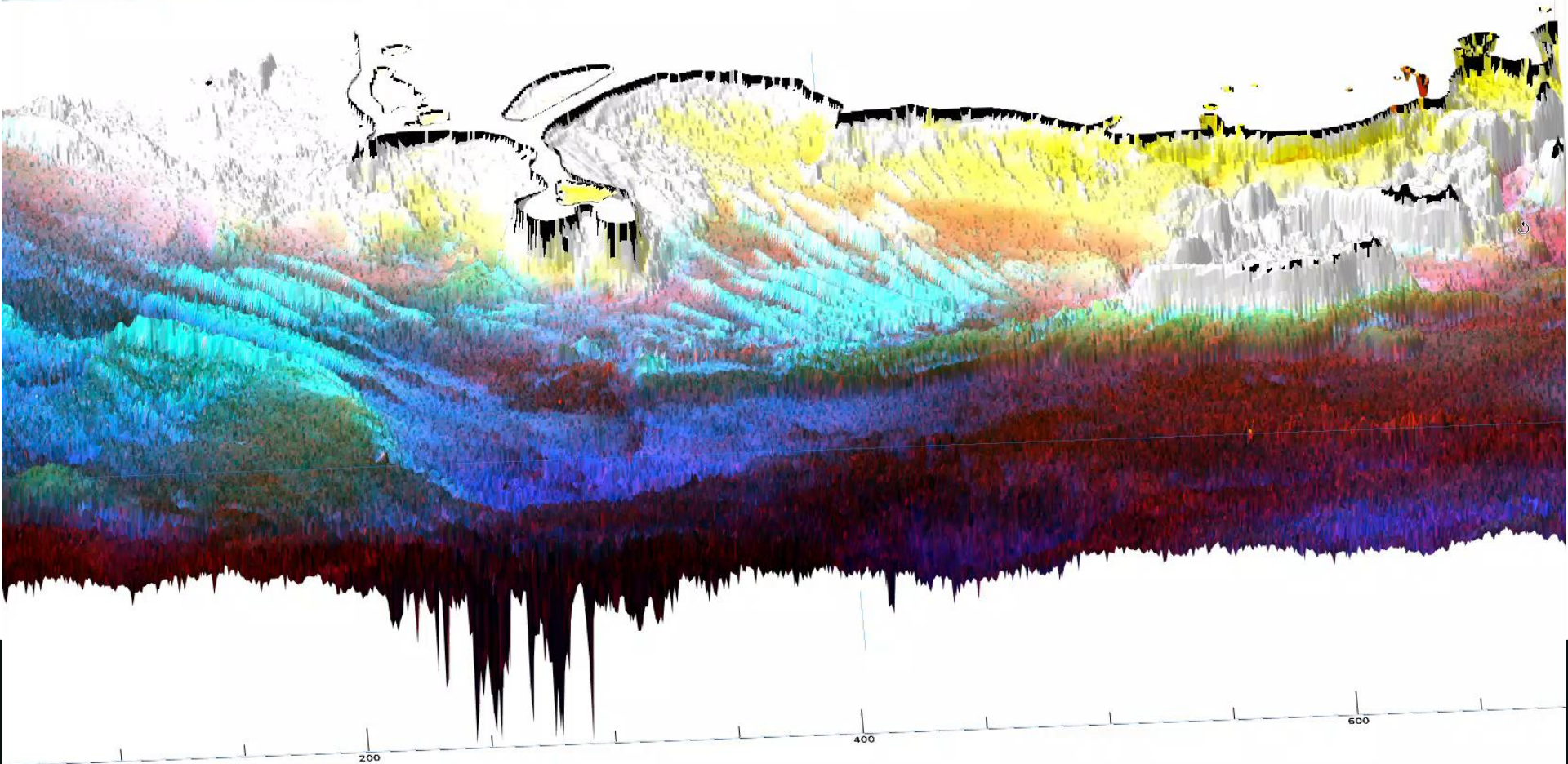
OLI 30m



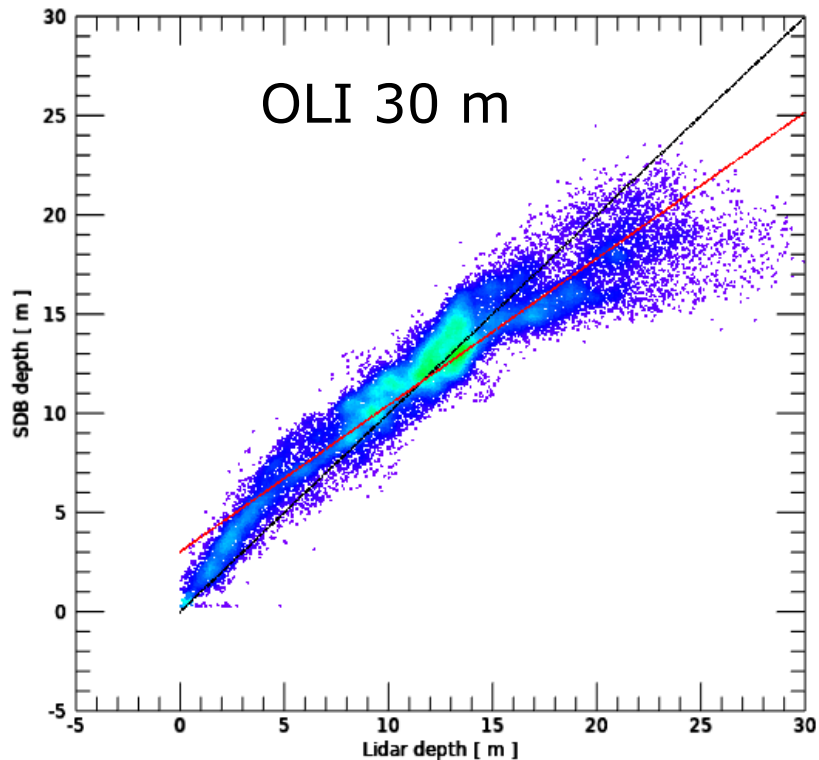
Bathy lidar DEM (with partially missing area)



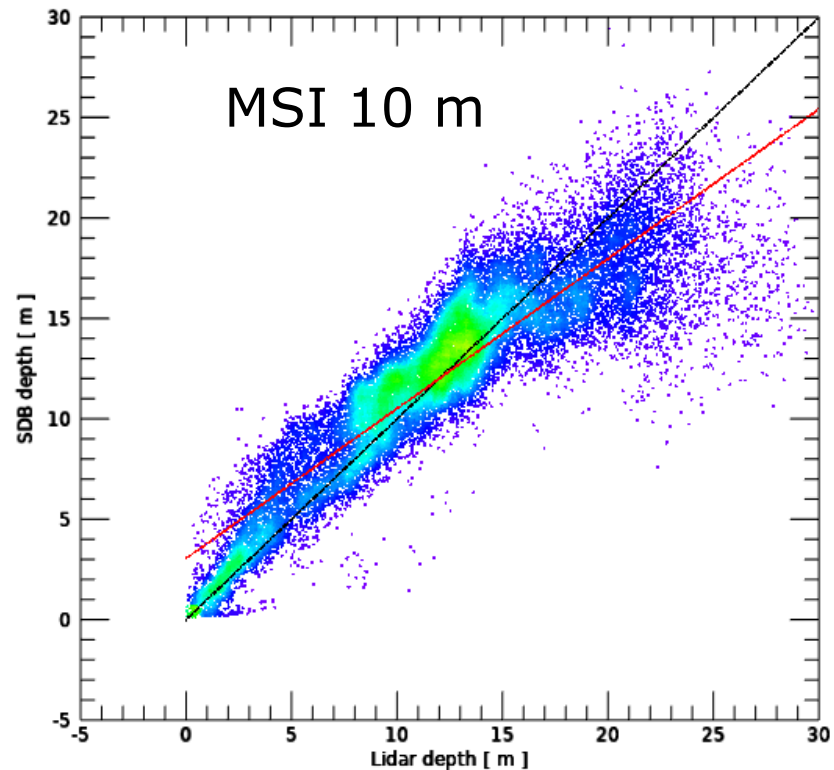
SDB DEM OLI (no missing area) — No MSI result (3M pixels → visualization issue)

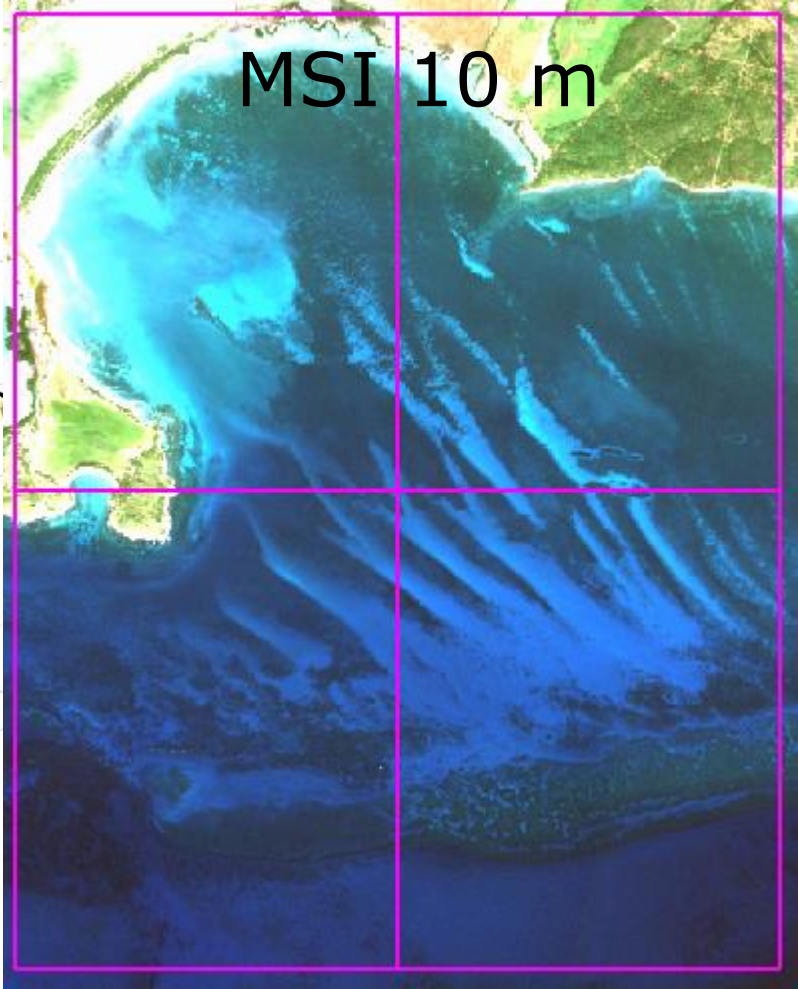
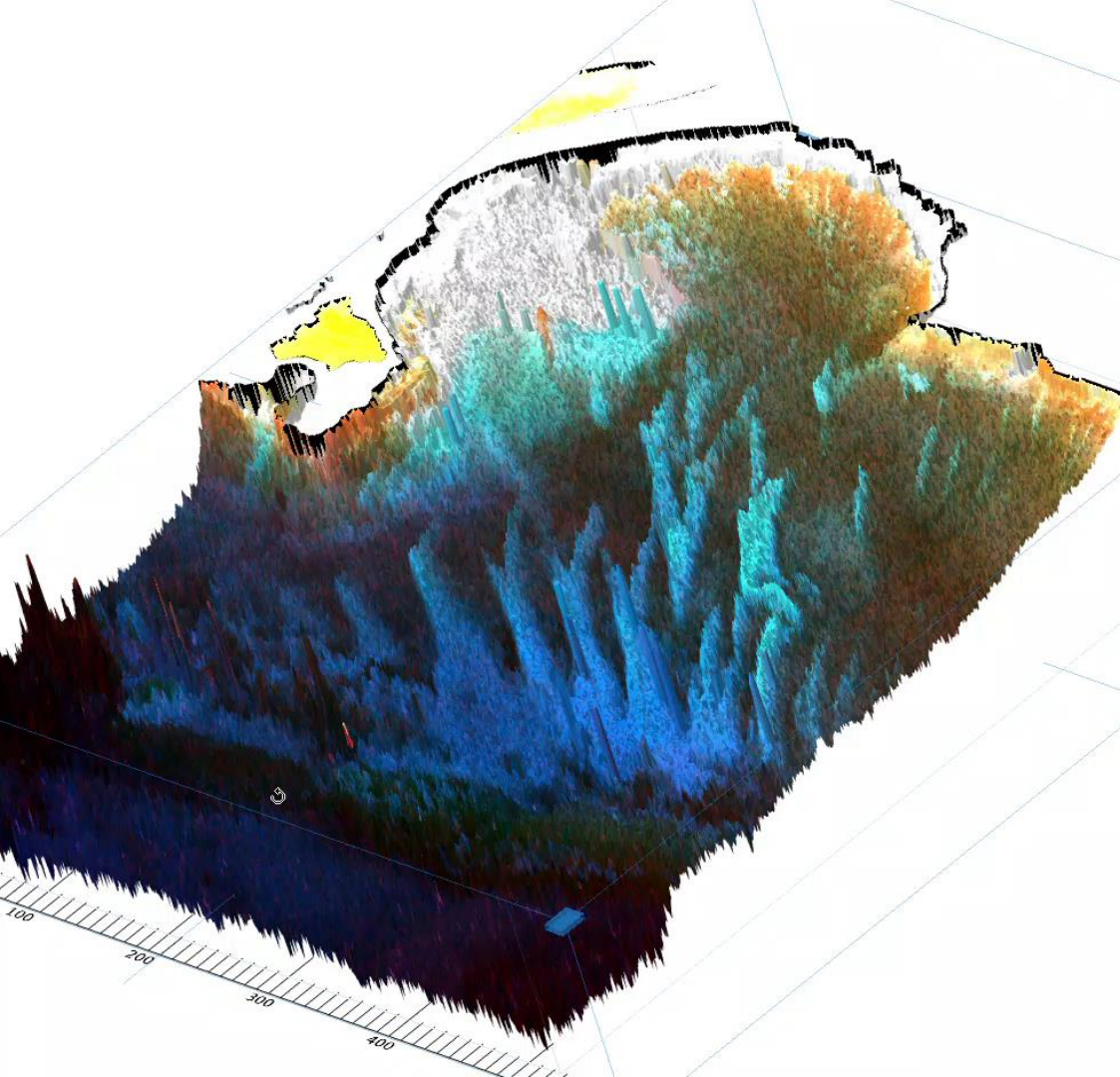


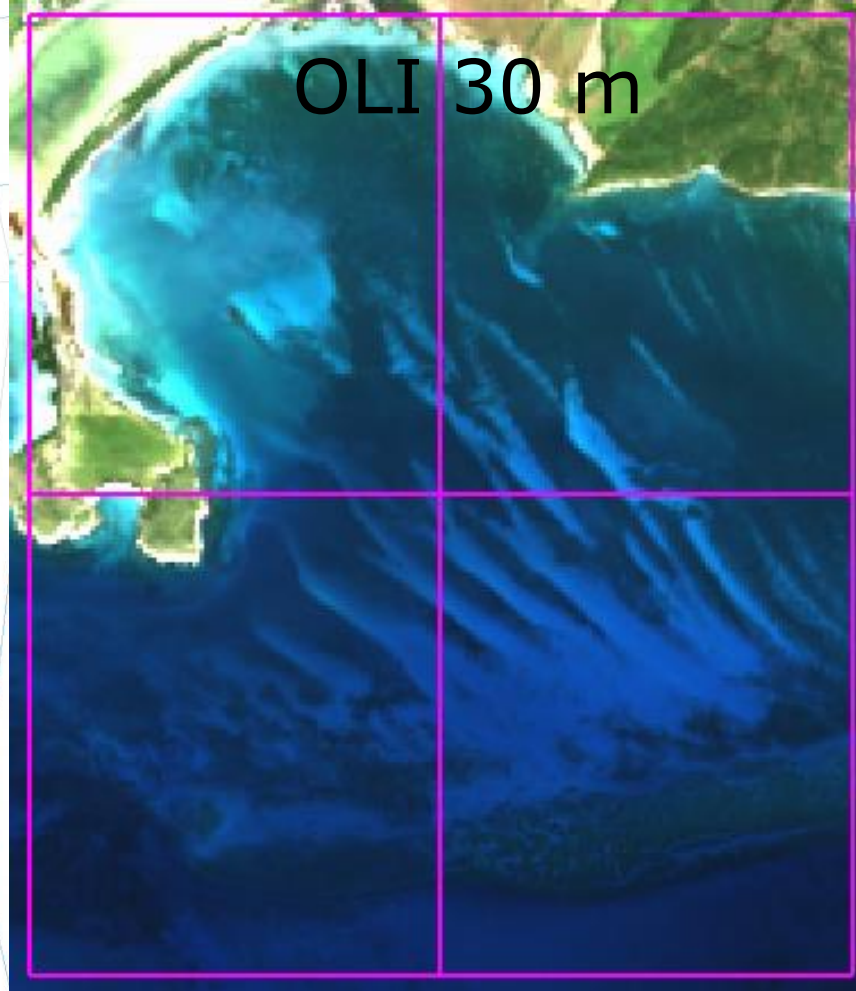
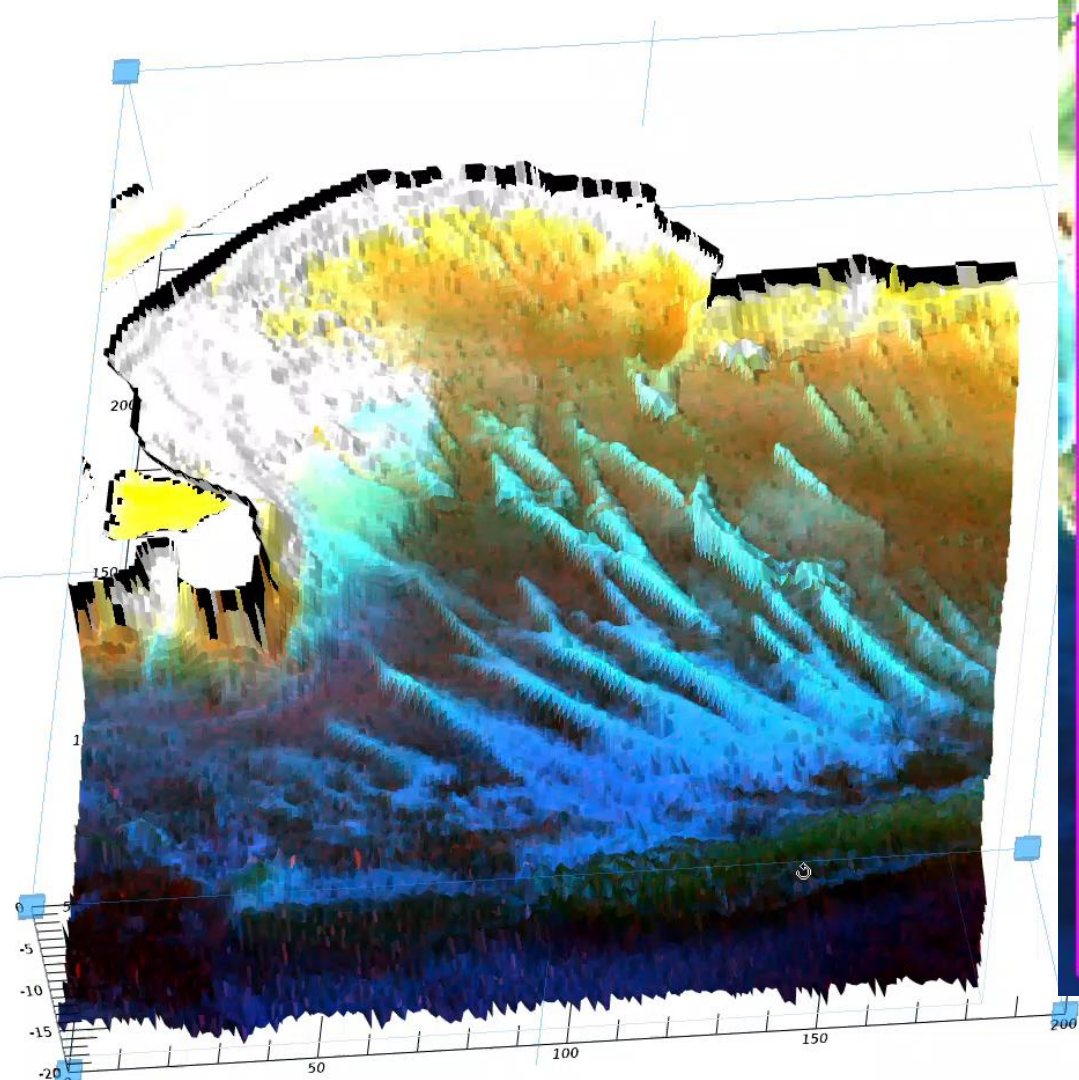
MAE = 1.322 m, RMSE = 1.678 m



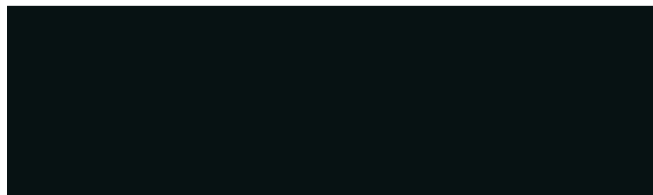
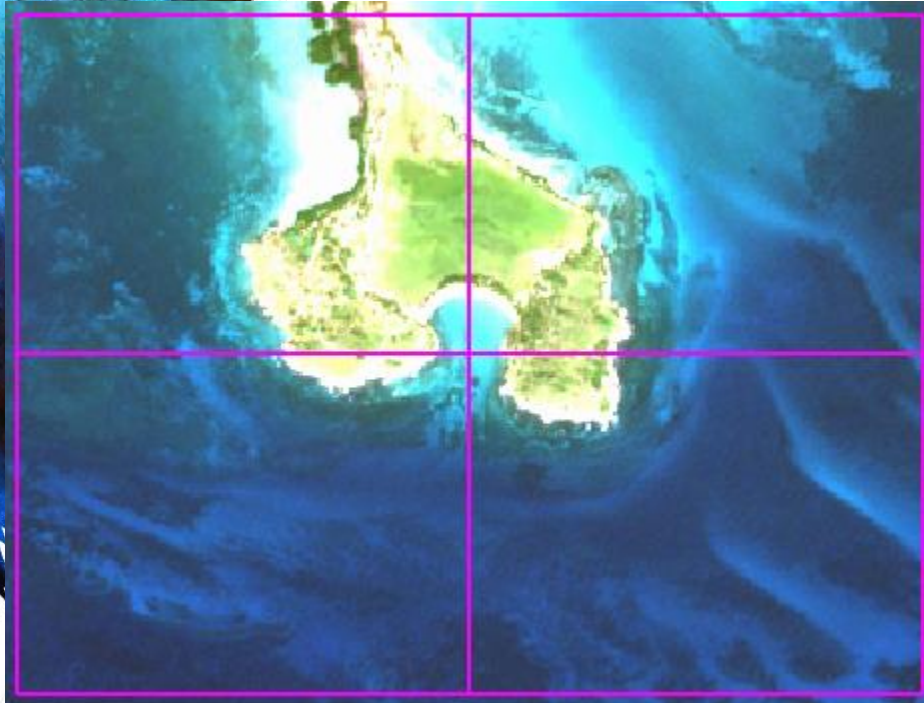
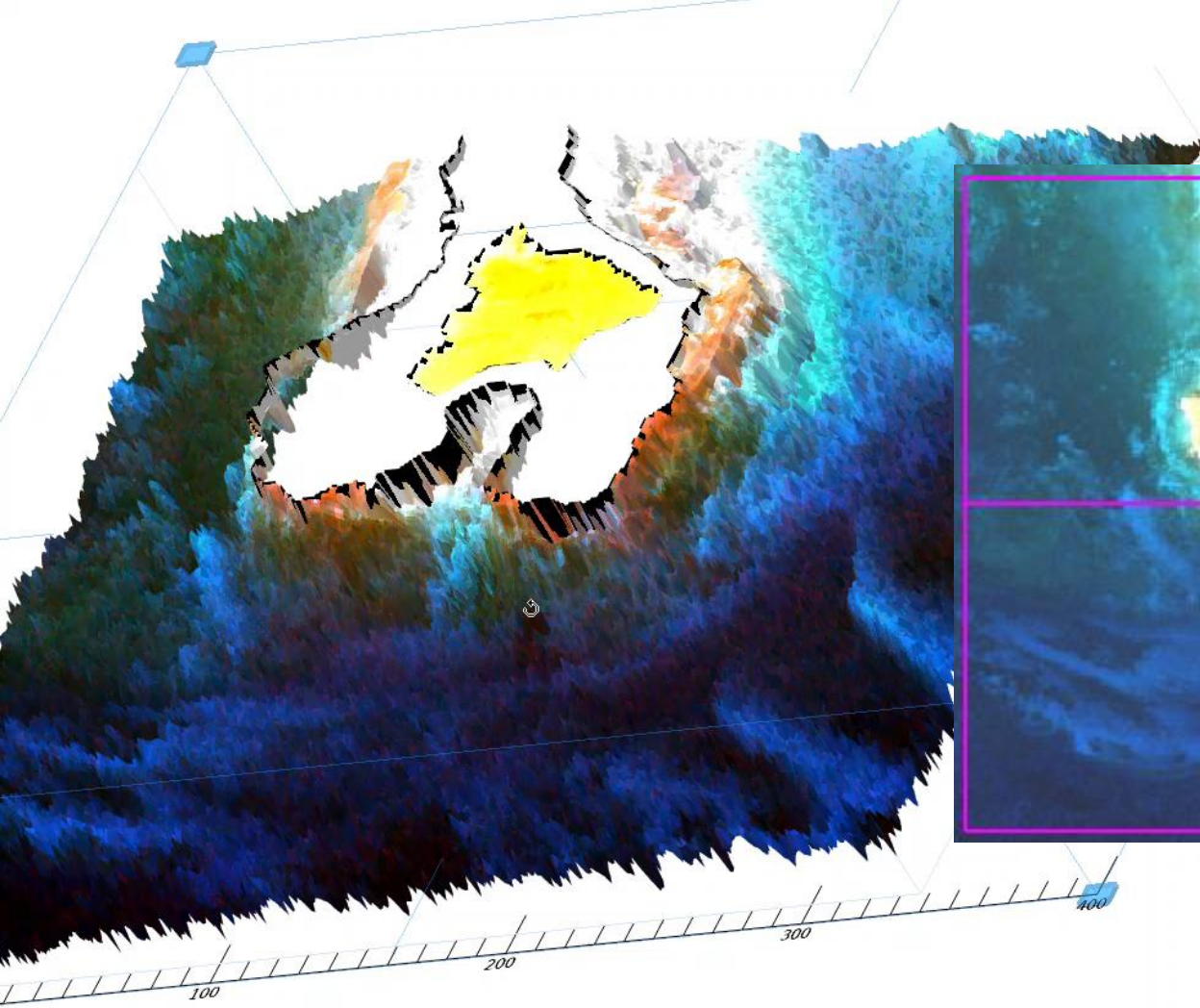
MAE = 1.687 m, RMSE = 2.070 m

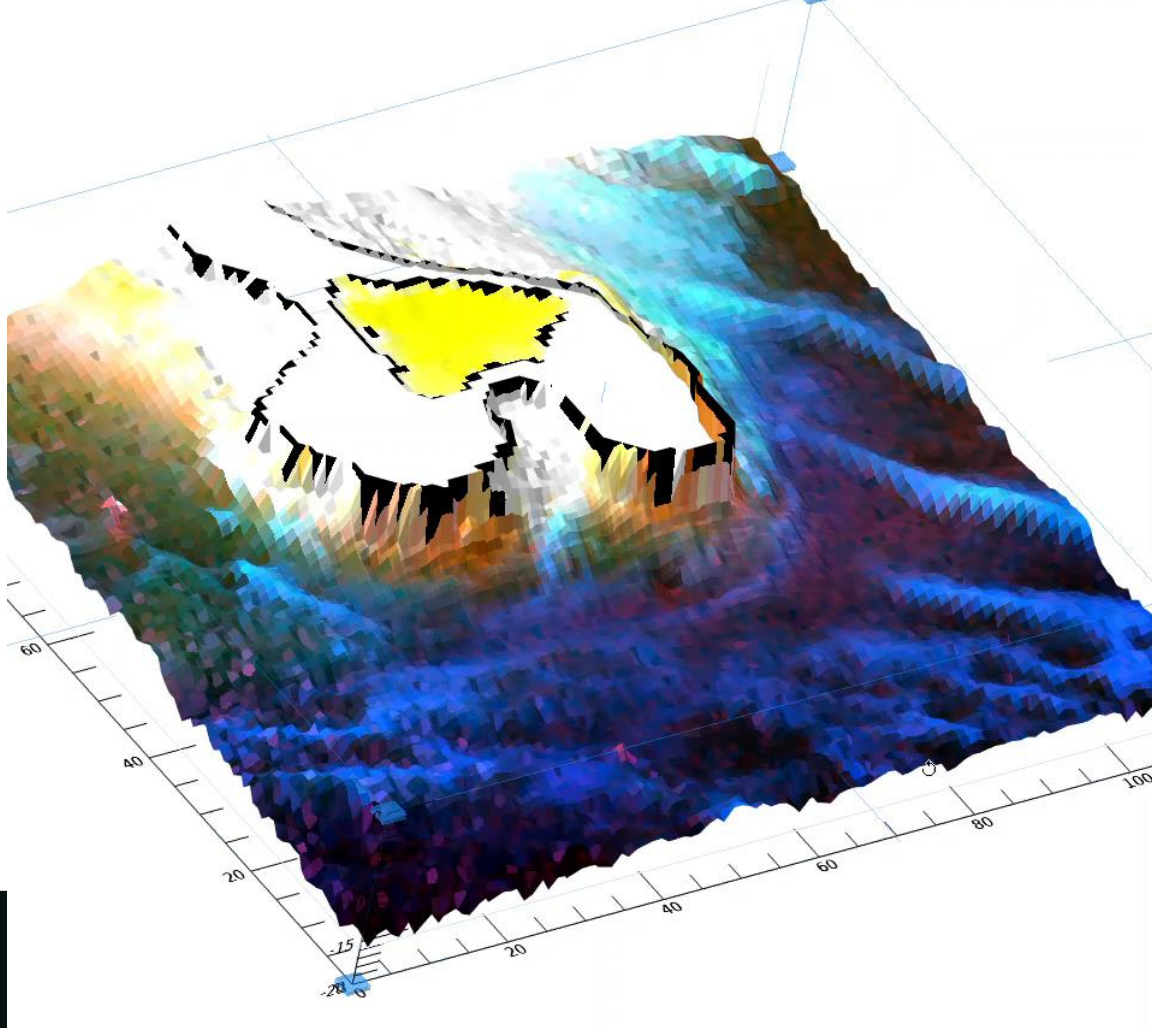




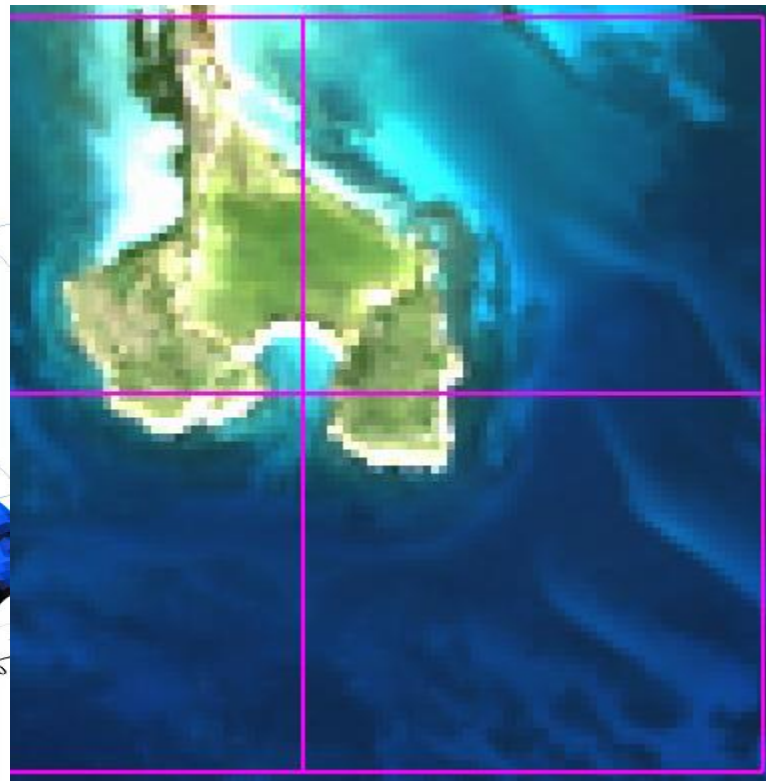


MSI 10 m



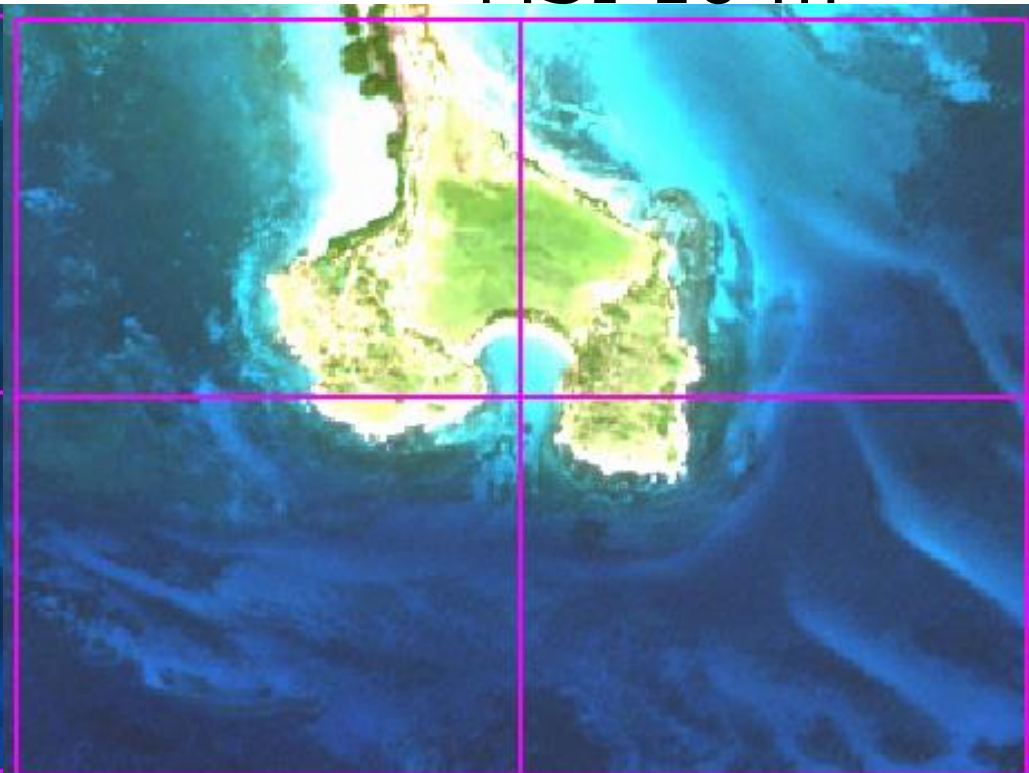
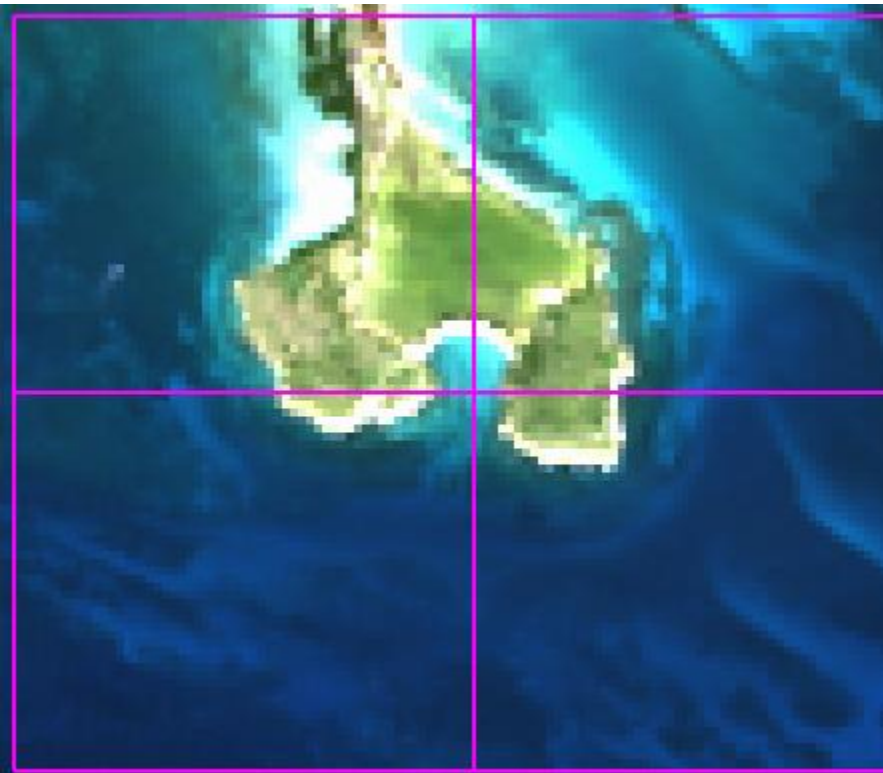


OLI 30 m

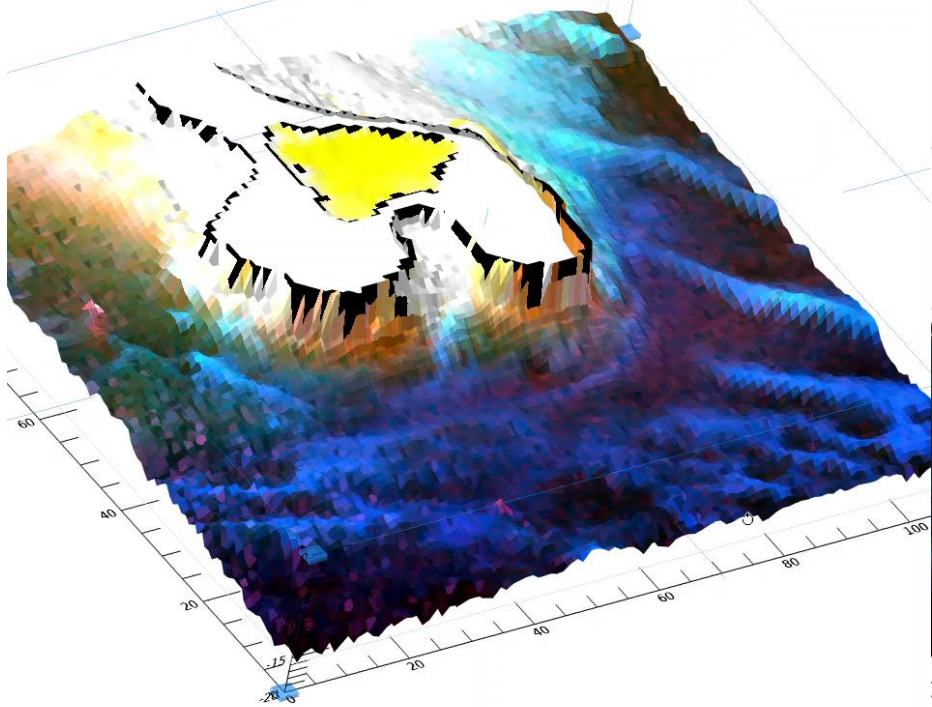


OLI 30 m

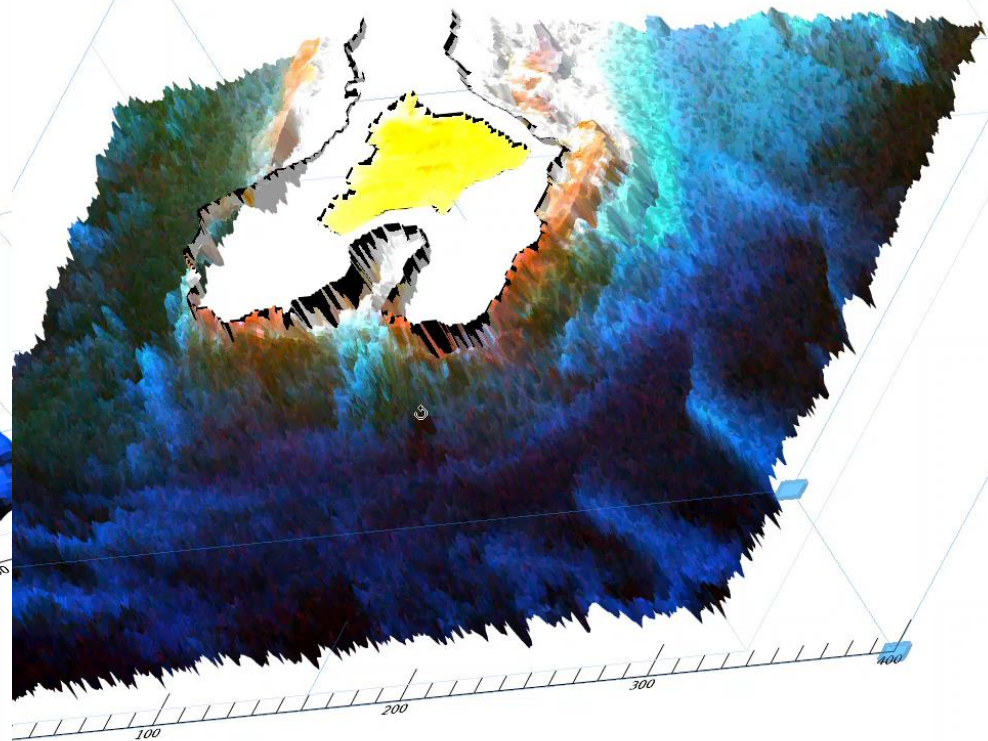
MSI 10 m



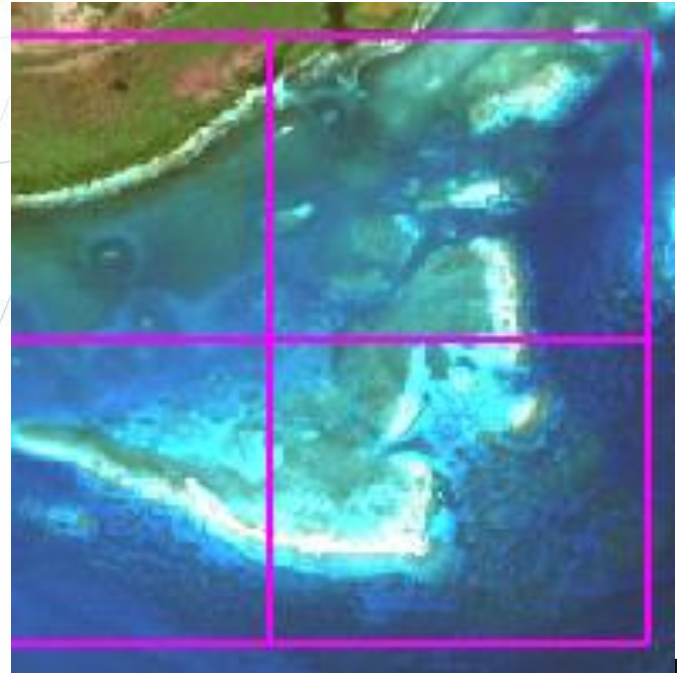
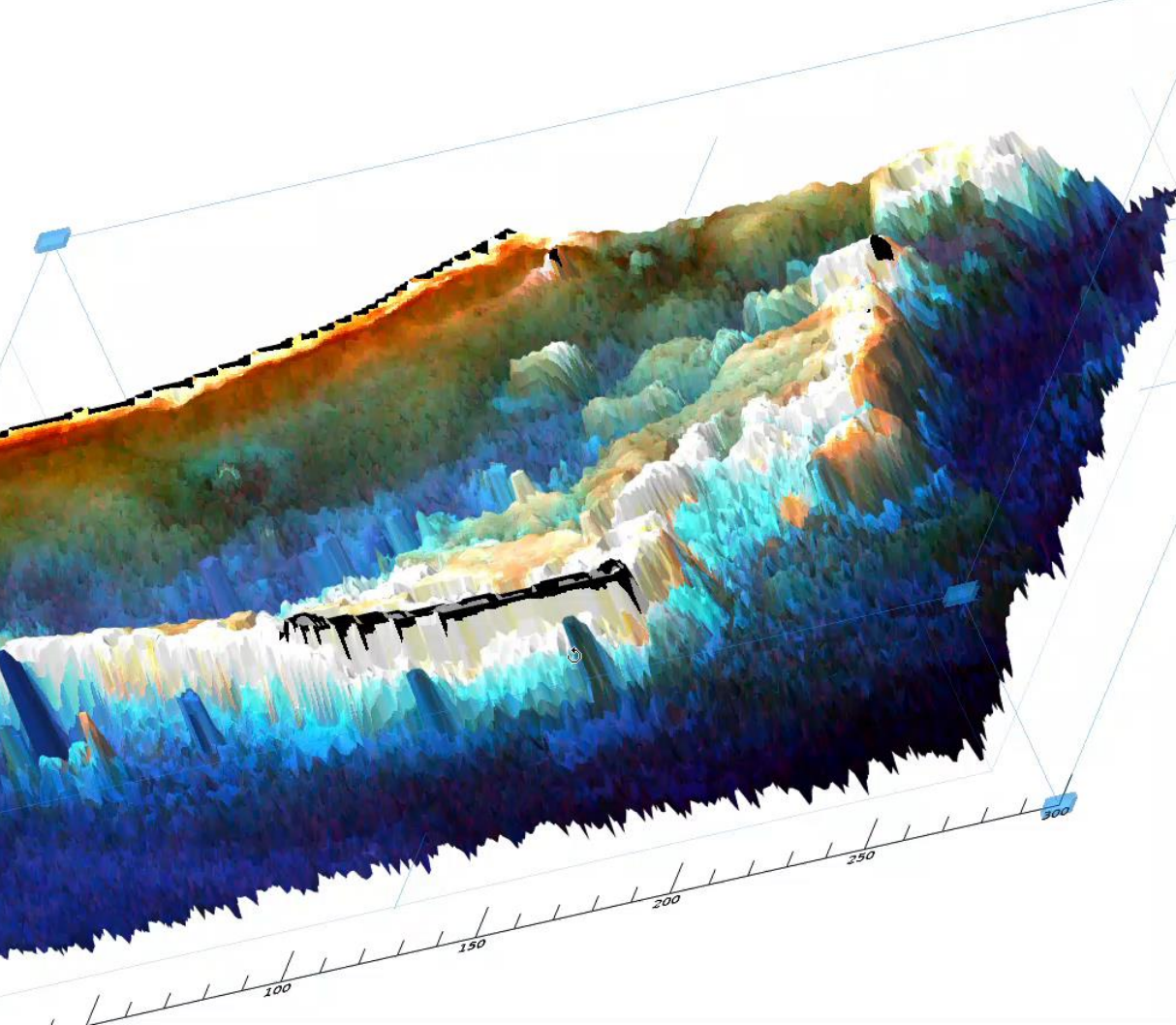
OLI 30 m

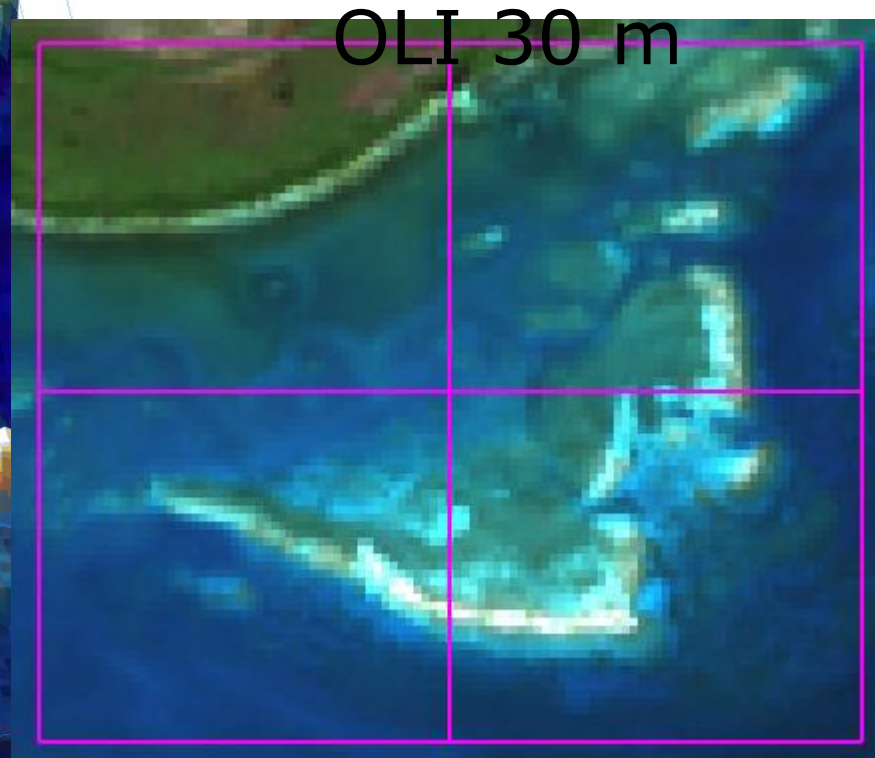
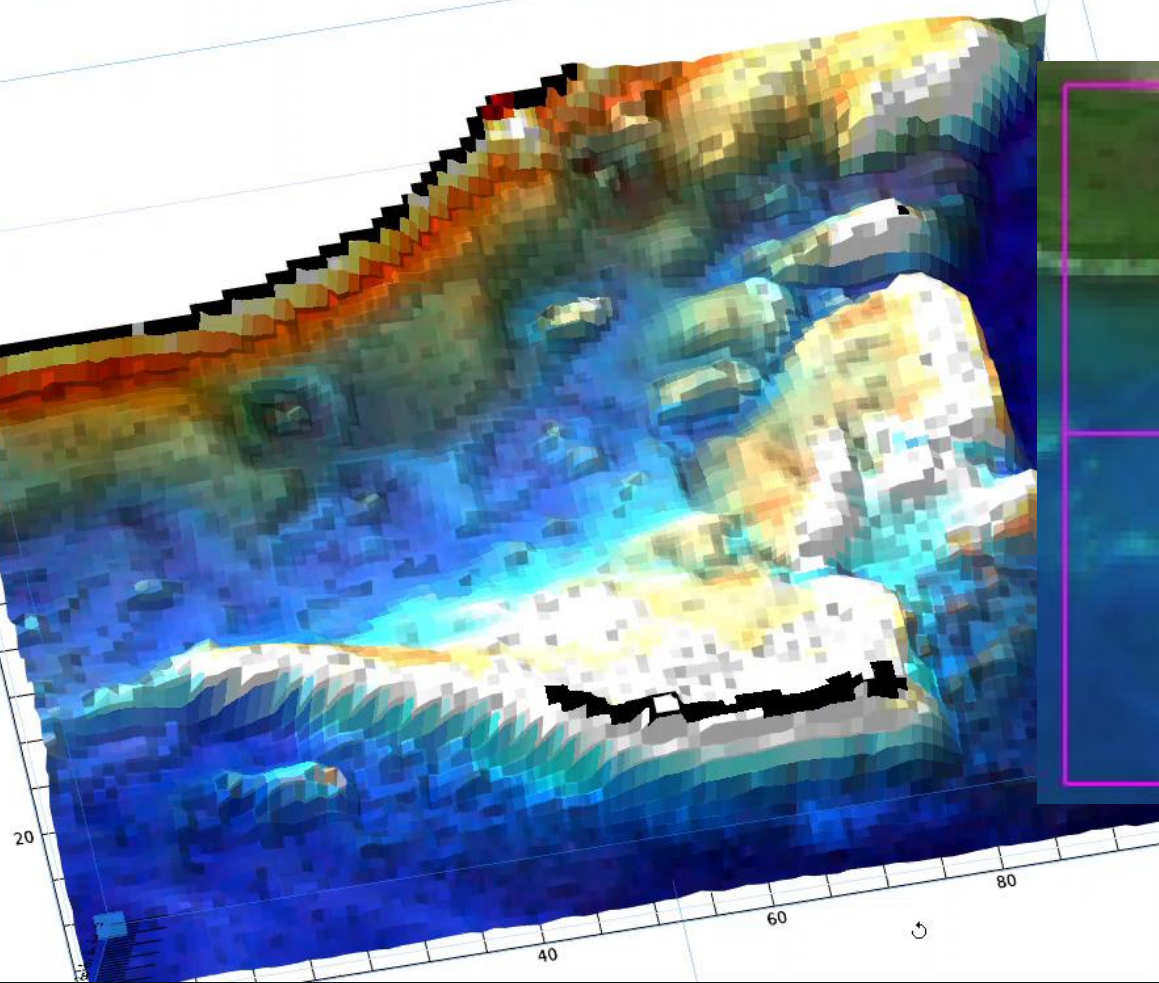


MSI 10 m



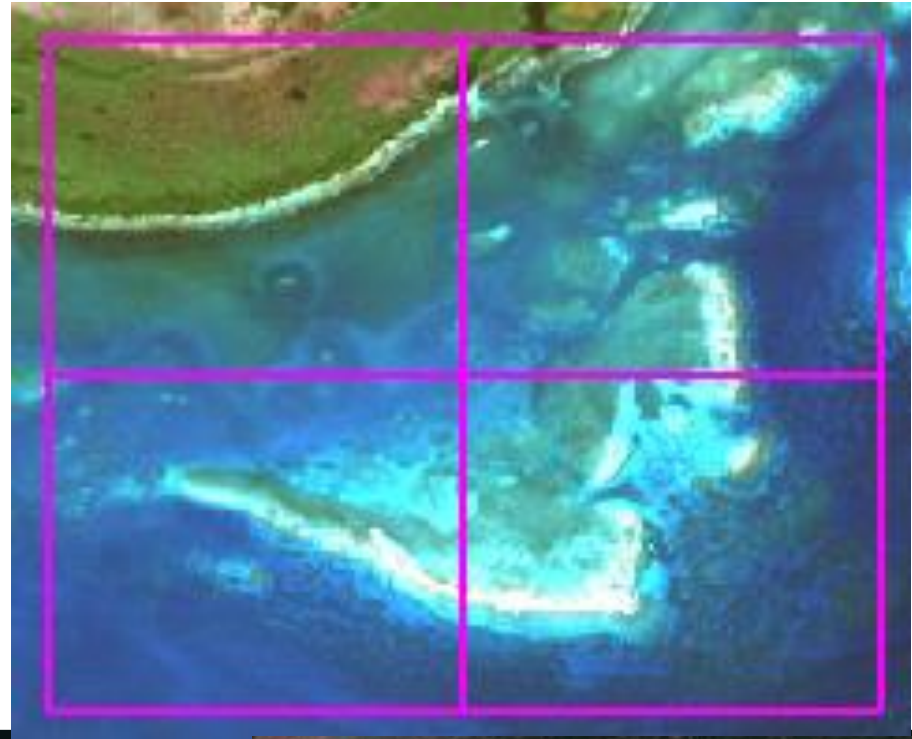
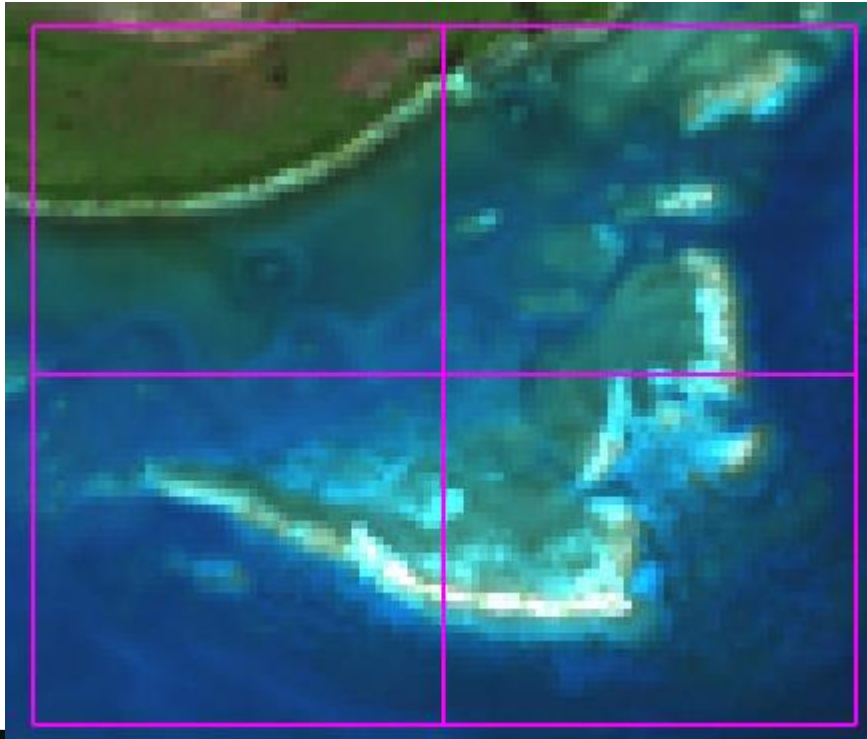
MSI 10 m



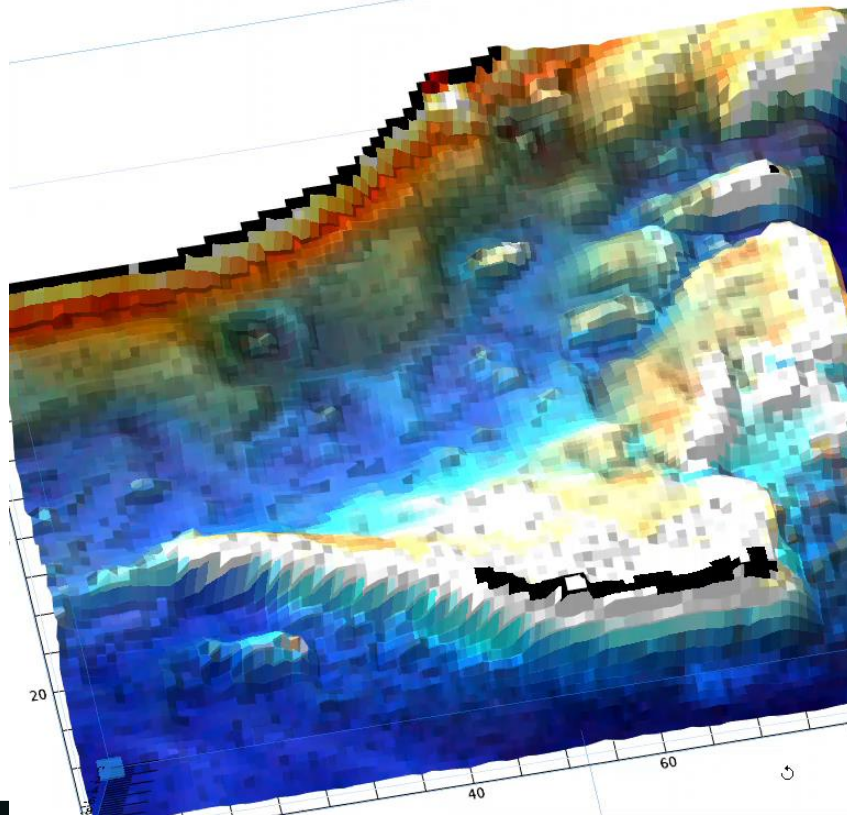


OLI 30 m

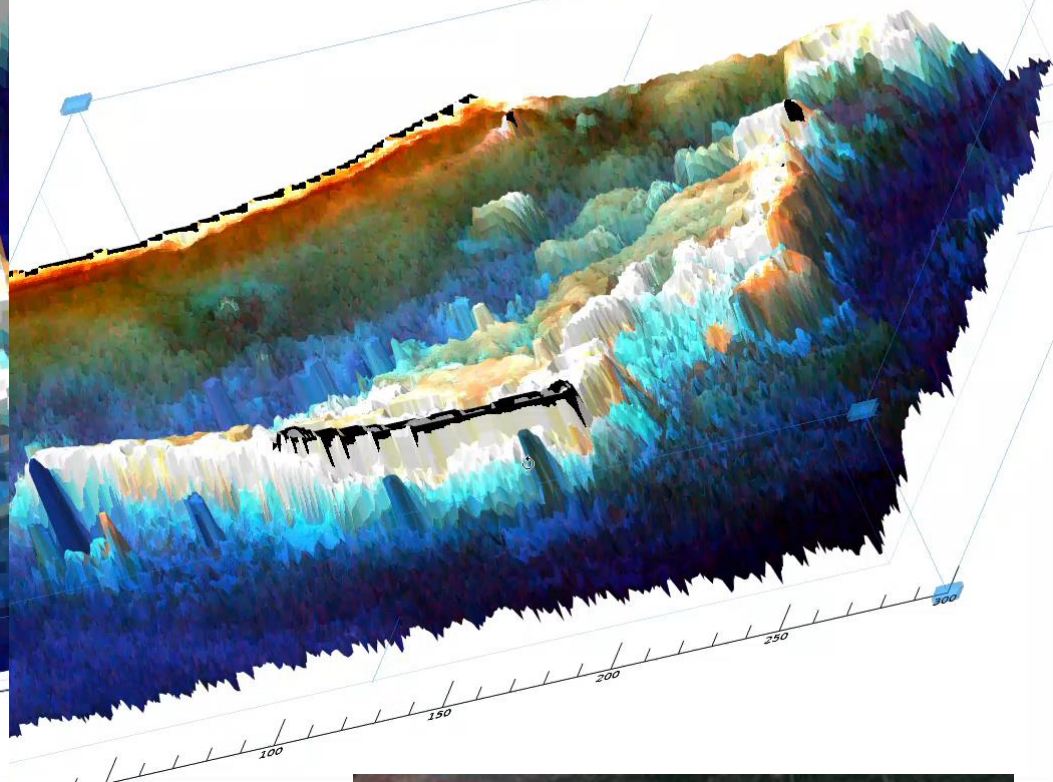
MSI 10 m

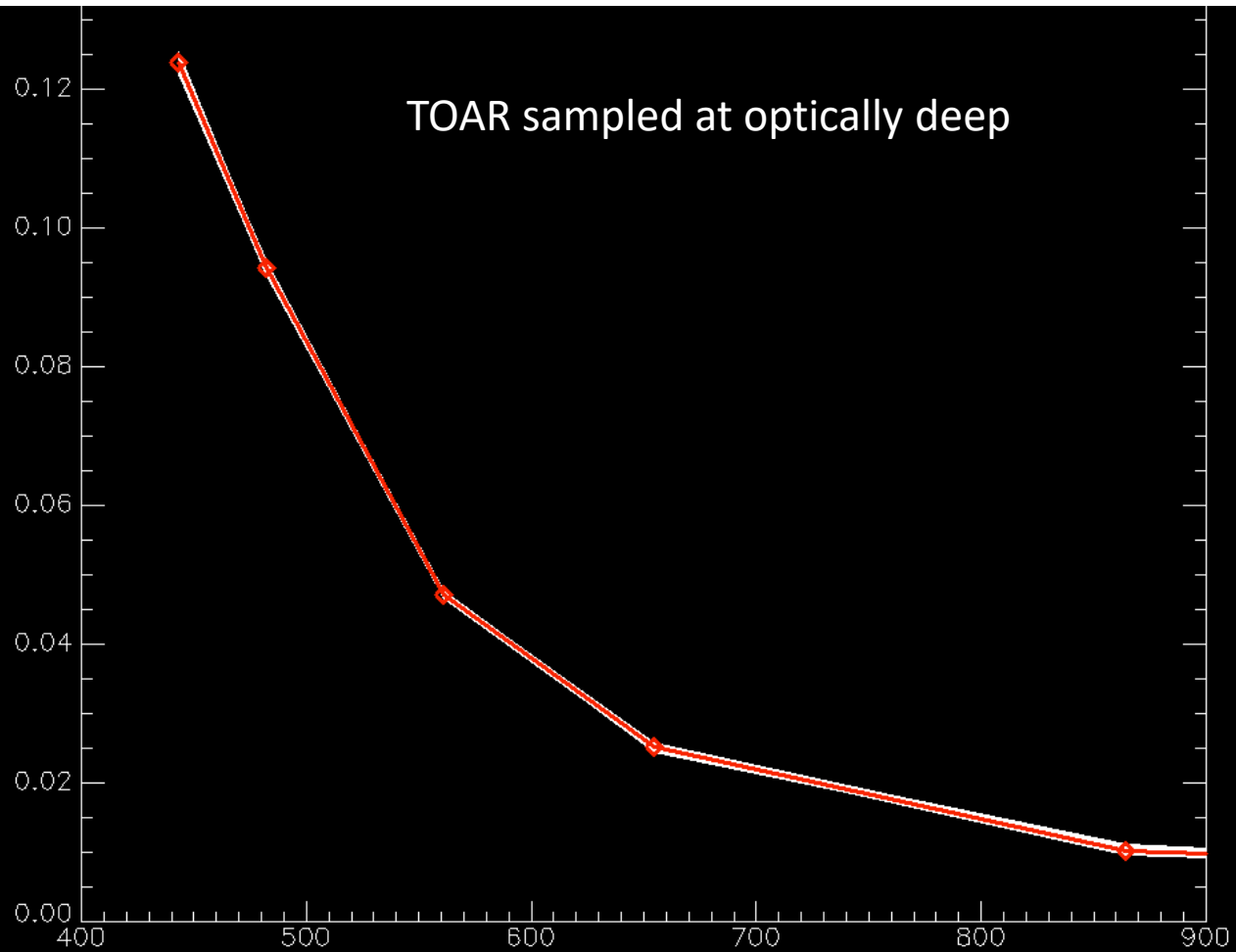
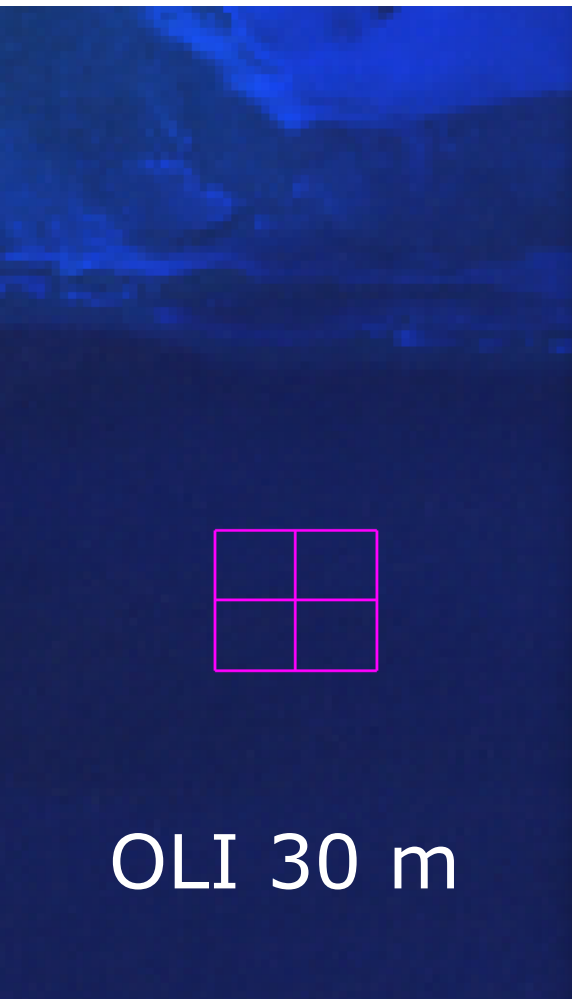


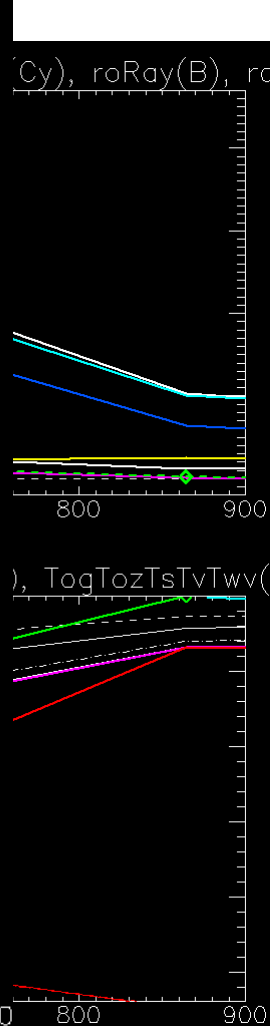
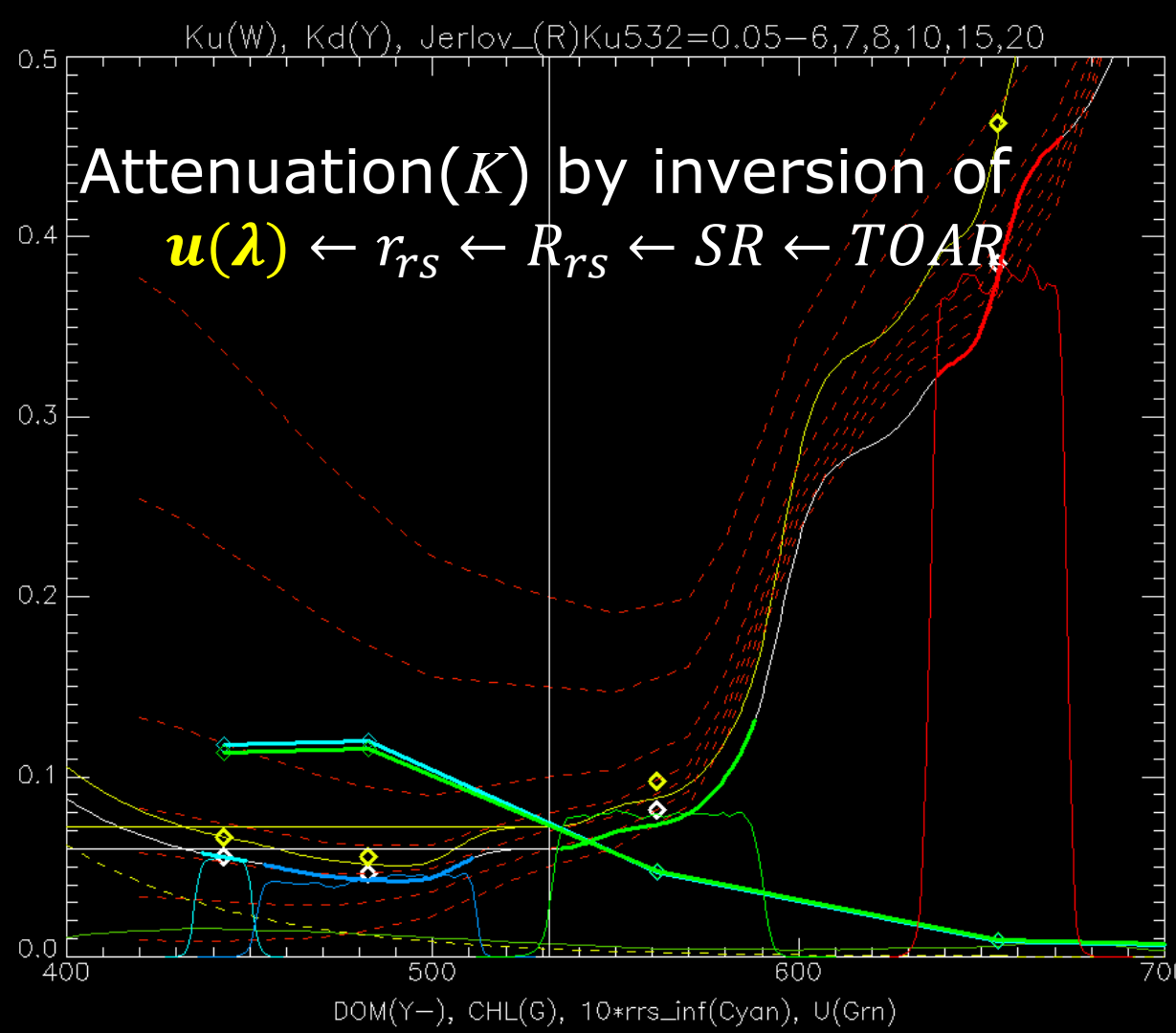
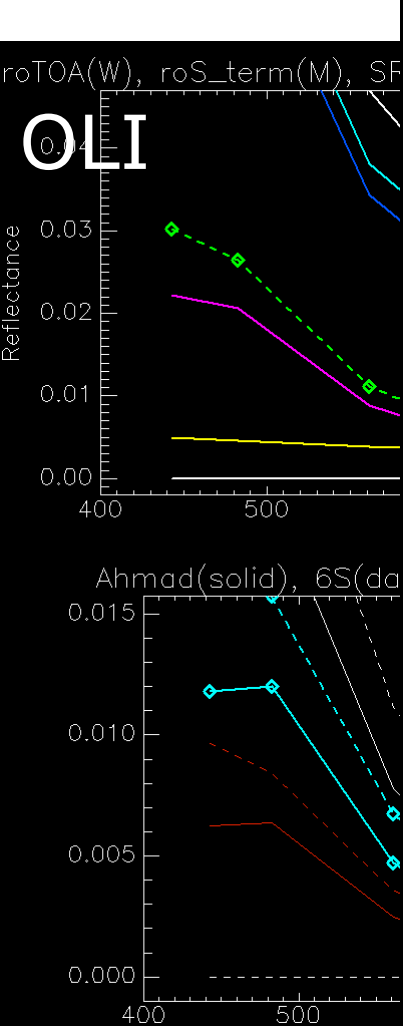
OLI 30 m

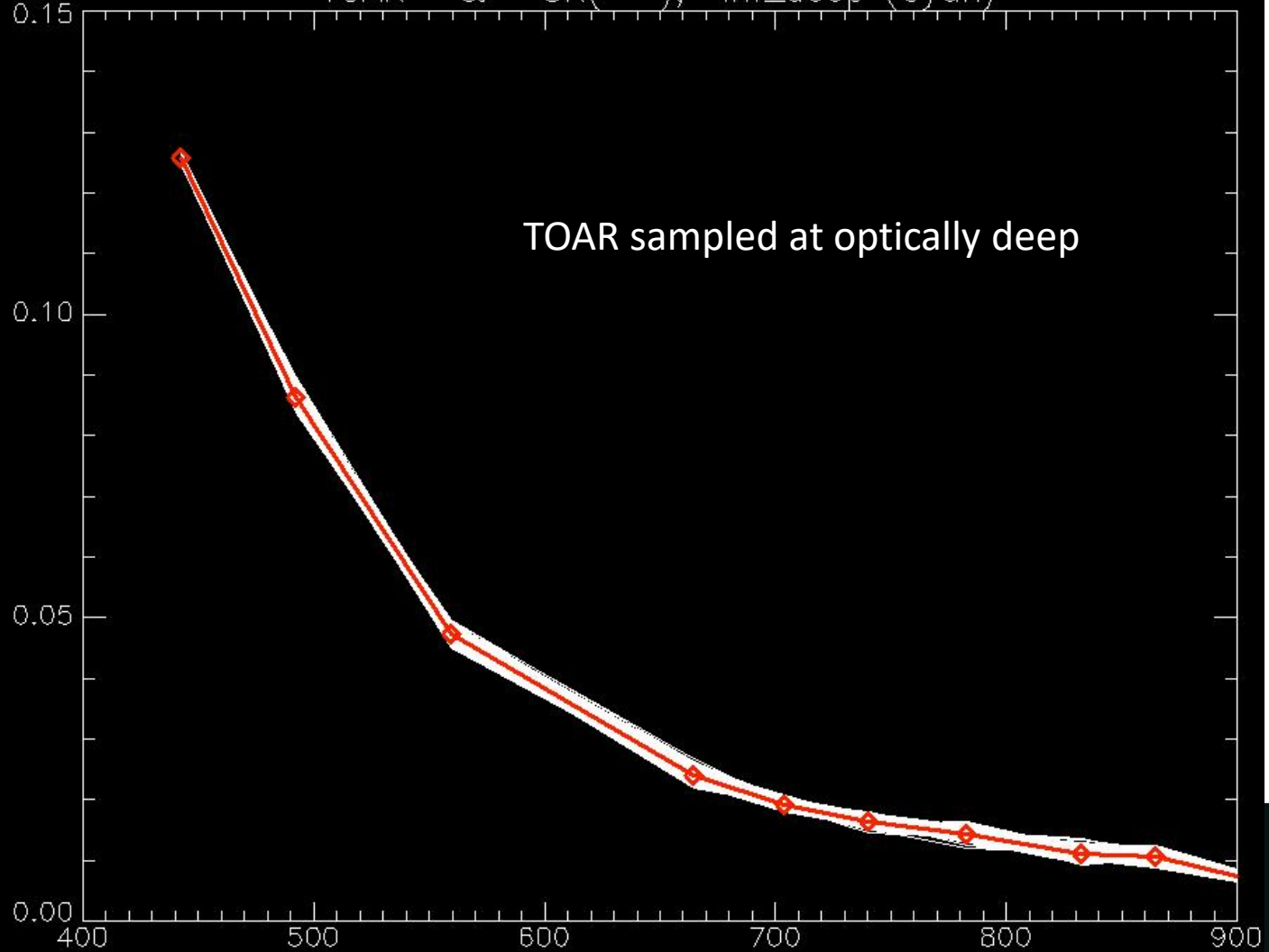
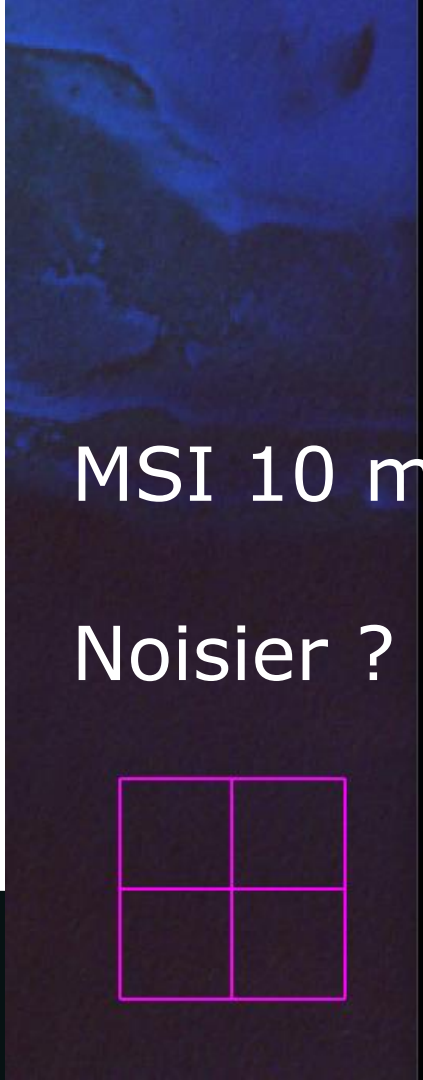


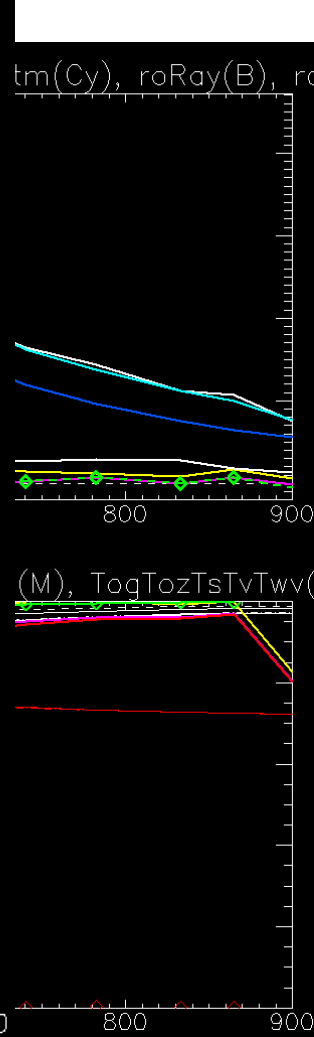
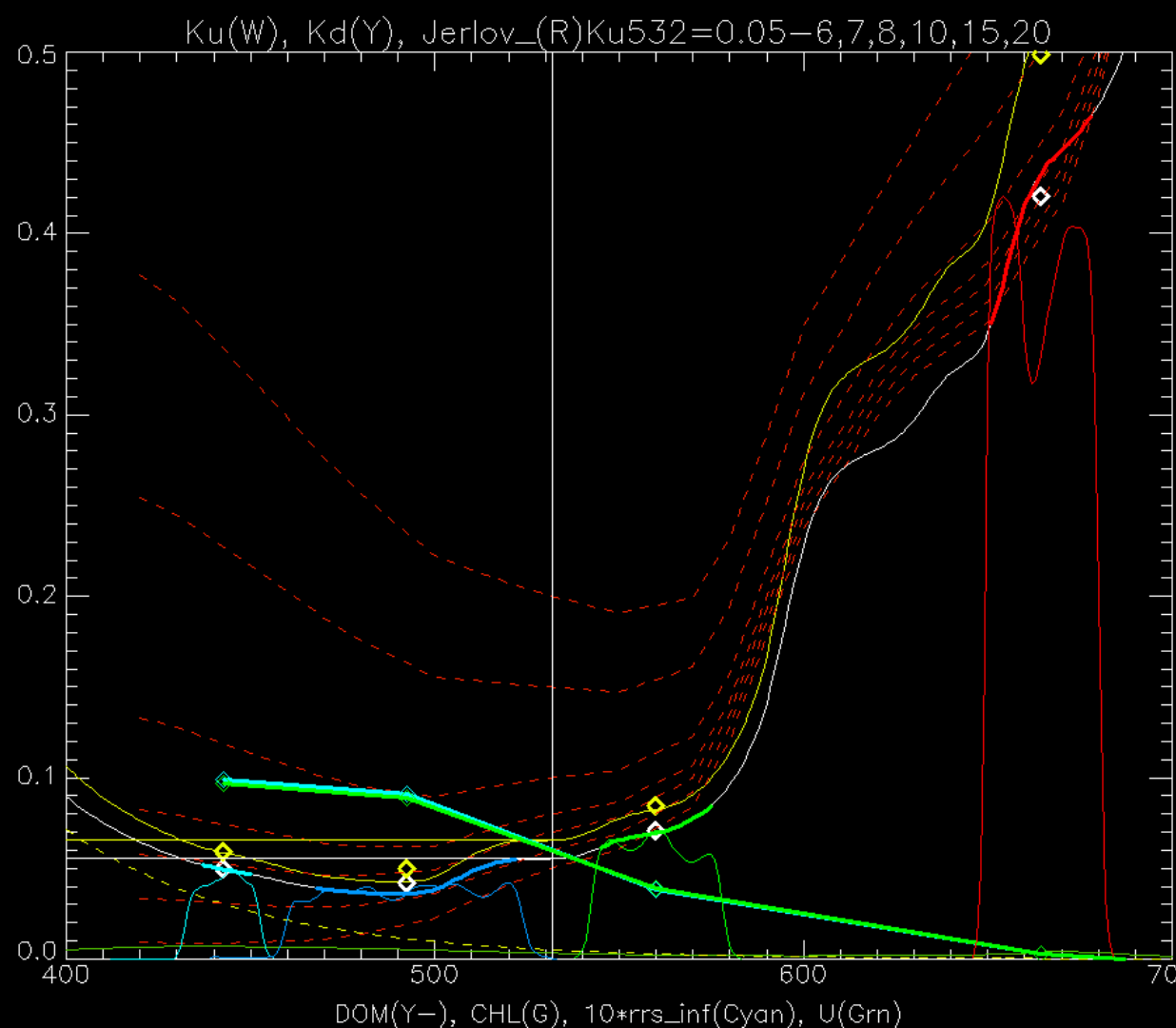
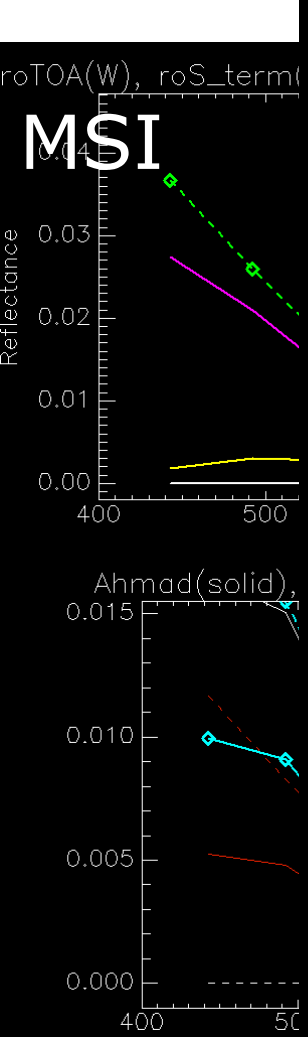
MSI 10 m





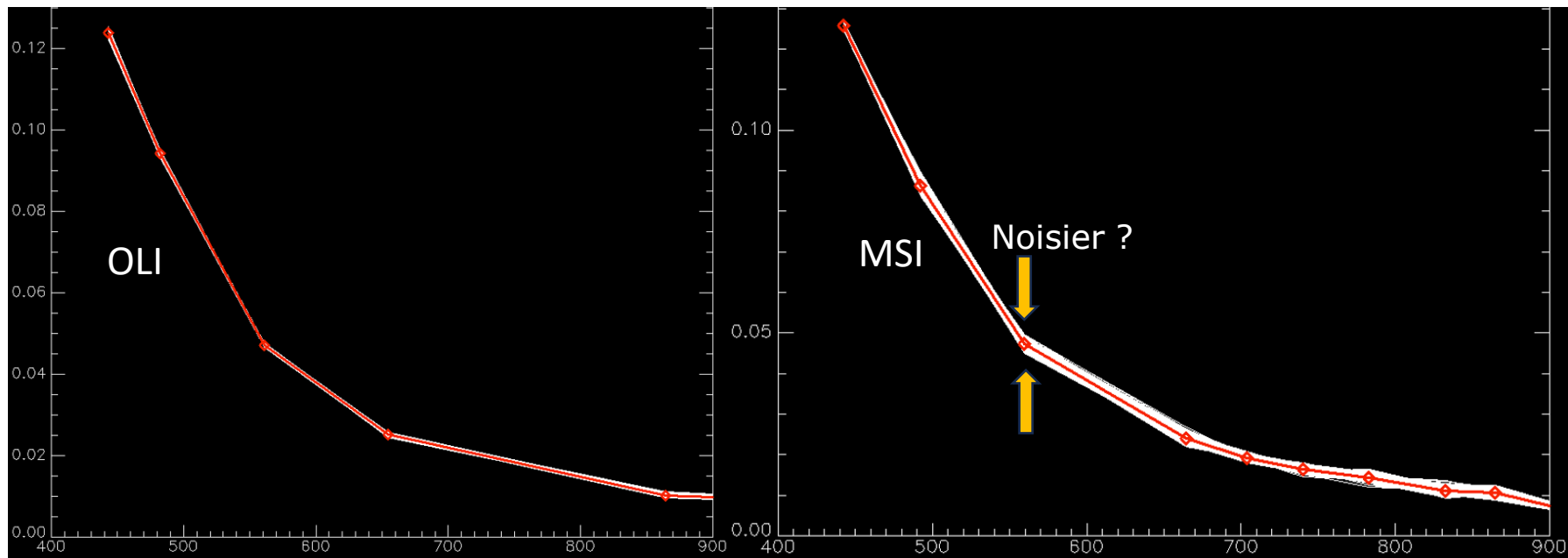




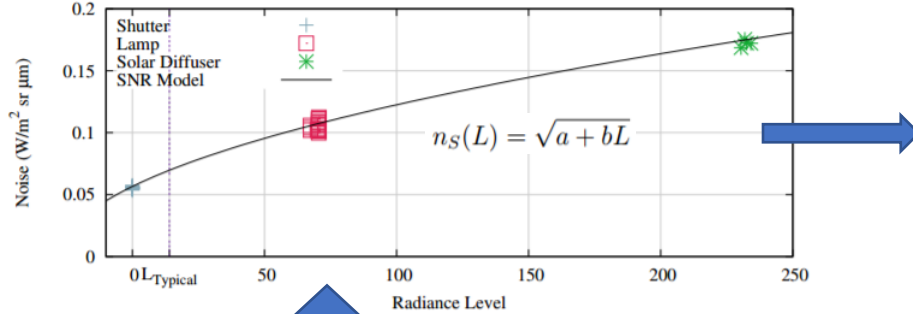


in a changing world

TOAR comparison (OLI vs MSI , same day), benefit of MSI's +4 bands ??



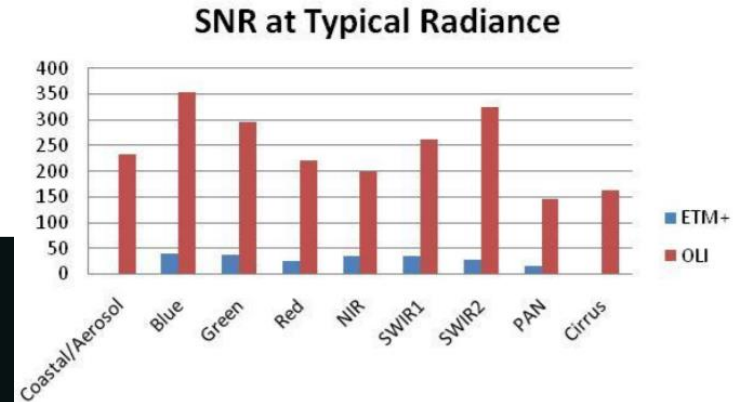
Noise of Landsat 8 OLI (Operational Land Imager)

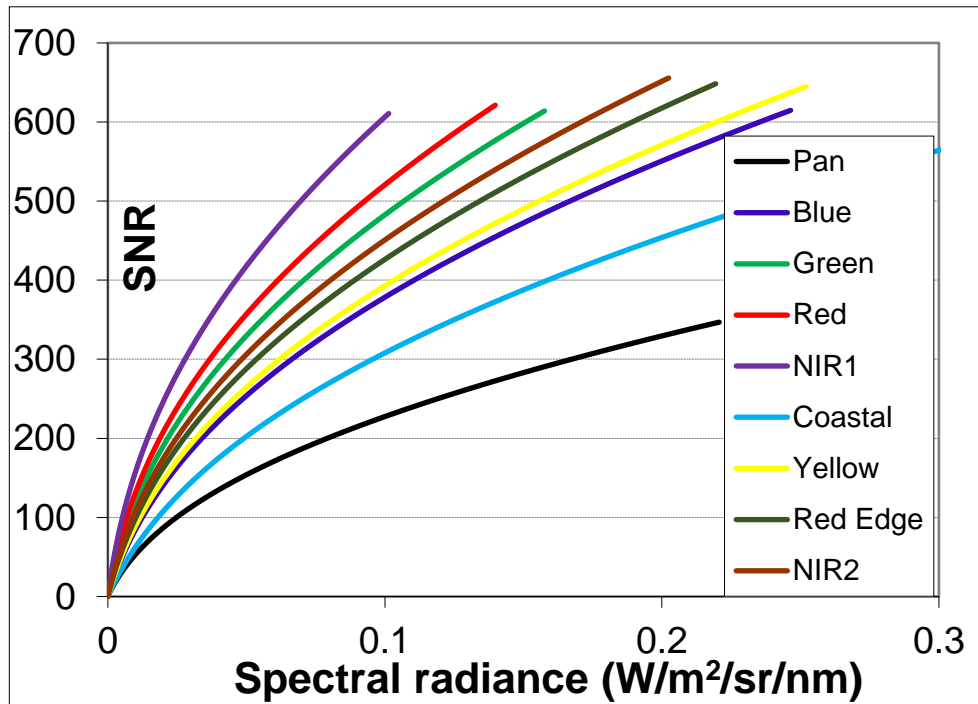


$$n_T(L) = \sqrt{0.8 \cdot n_S(L)^2 + \epsilon_q^2}$$

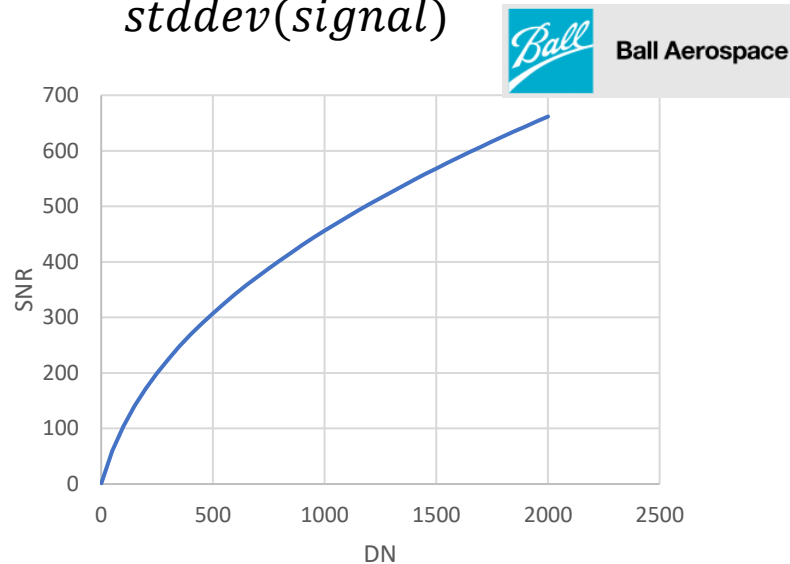
$$SNR \approx L / n_T(L)$$

Band Name	Average Dark Noise	Noise Model Coefficients		Product Quantization	Quantization Levels	
	$\left(\frac{W}{m^2 \text{ sr } \mu m}\right)$	$a \left(\left(\frac{W}{m^2 \text{ sr } \mu m}\right)^2\right)$	$b \left(\frac{W}{m^2 \text{ sr } \mu m}\right)$	Noise (ϵ_q)	$L_{\text{Typical}} \left(\frac{W}{m^2 \text{ sr } \mu m}\right)$	$L_{\text{High}} \left(\frac{W}{m^2 \text{ sr } \mu m}\right)$
Coastal Aerosol	0.11	0.012	0.00042	0.0047	40	190
Blue	0.089	0.0082	0.000094	0.0048	40	190
Green	0.084	0.0073	0.000089	0.0035	30	194
Red	0.083	0.0071	0.00011	0.0039	22	150
NIR	0.056	0.0032	0.00012	0.0028	14	150
SWIR 1	0.011	0.00013	0.000026	0.00055	4.0	32
SWIR 2	0.0034	0.000011	0.0000091	0.00025	1.7	11
Pan	0.086	0.0078	0.00069	0.0033	23	156
Cirrus	0.025	0.00059	0.00014	0.00070	6.0	N/A





$$SNR = \frac{\text{mean}(\text{signal})}{\text{stddev}(\text{signal})}$$



Radiometric resolution is routinely expressed as a bit number, typically in the range of 8 to 16 bits. The radiometric resolution of the MSI instrument is 12 bit, enabling the image to be acquired over a range of 0 to 4 095 potential light intensity values. The radiometric accuracy is less than 5% (goal 3%). Radiometric resolution is also dependent upon the Signal to Noise Ratio (SNR) of the detector.

Table 3: 10 m Spatial Resolution Bands and associated Signal to Noise ratio (SNR)

Band number	Central wavelength (nm)	Bandwidth (nm)	Lref (reference radiance) ($W m^{-2} sr^{-1} \mu m^{-1}$)	MSI SNR @ Lref
2	490	65	128	154
3	560	35	128	168
4	665	30	108	142
8	842	115	103	172



OLI L_{typical} OLI SNR L_{typical}

40 **350**

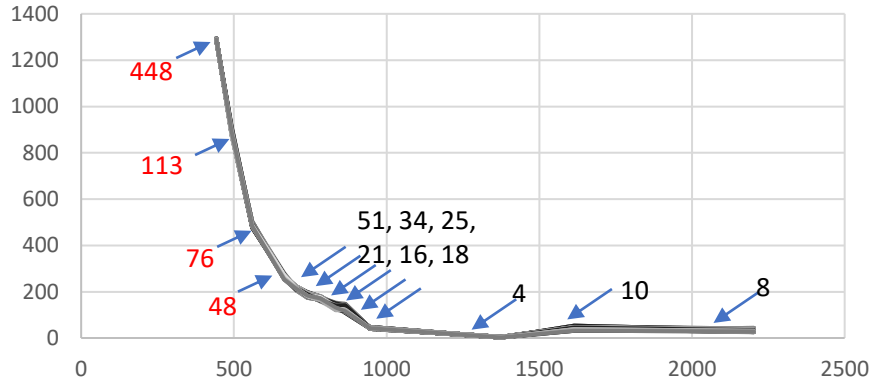
30 **300**

22 **220**

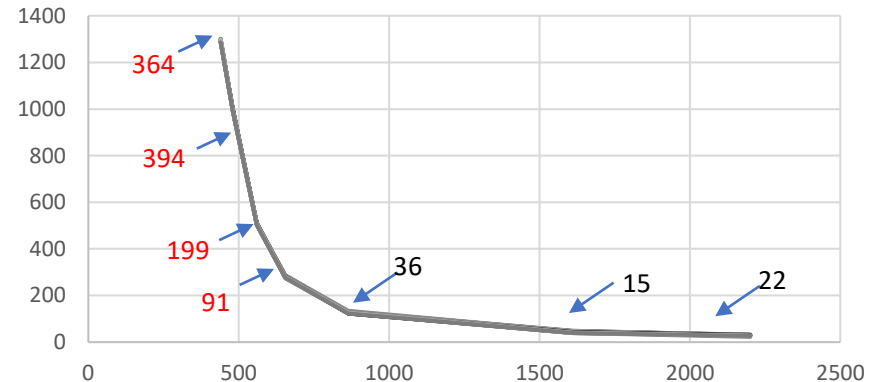
14 **200**

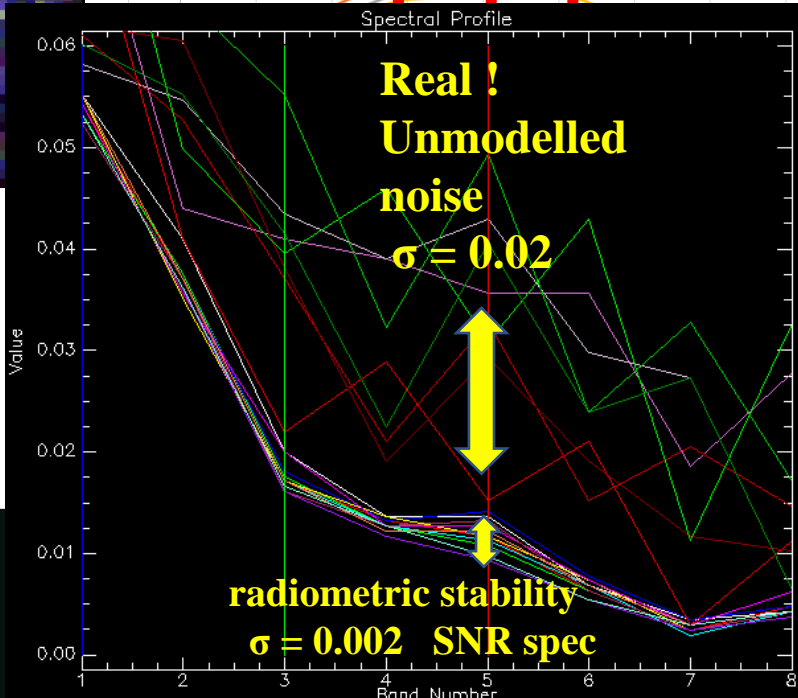
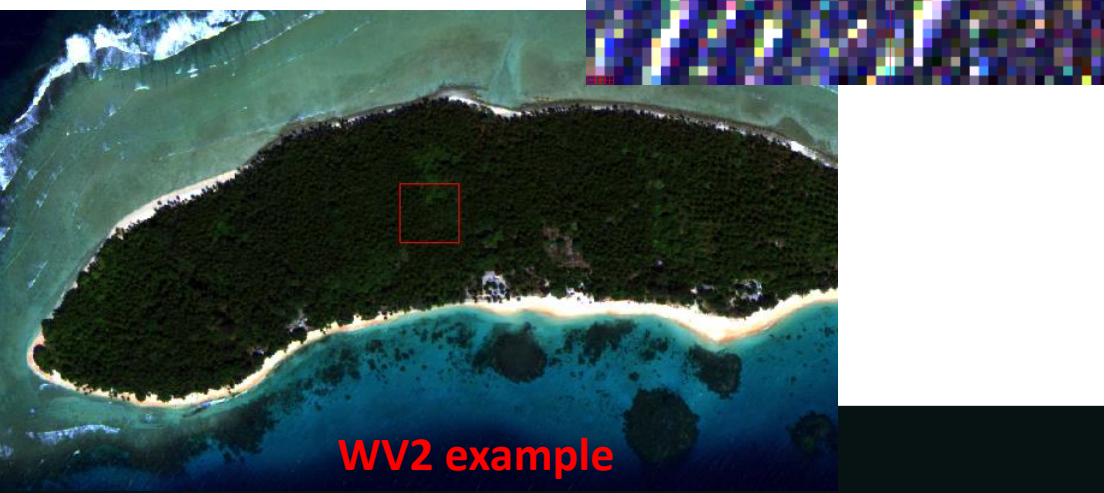
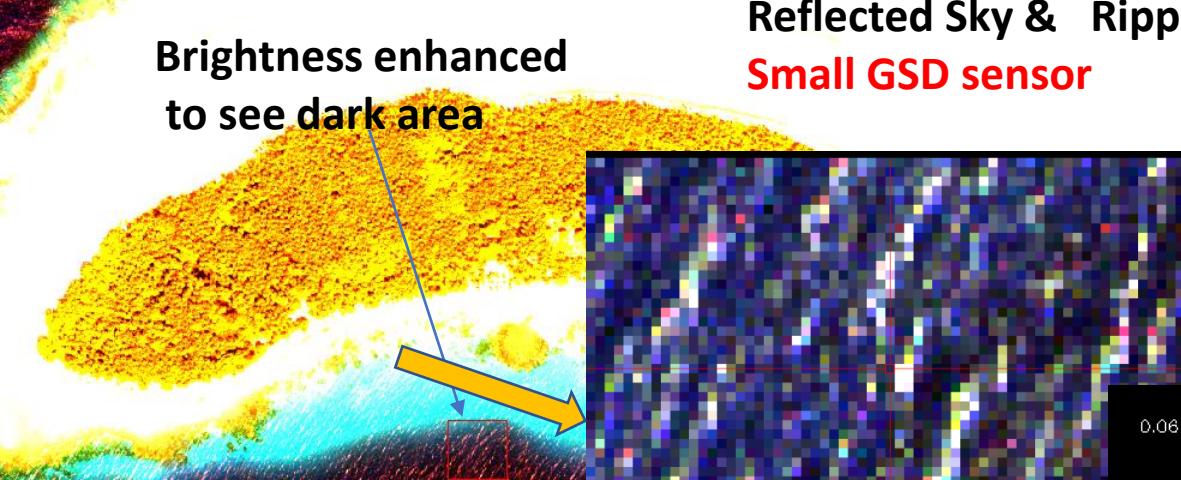
Optically Deep-Water SNR

S2-MSI (10m)

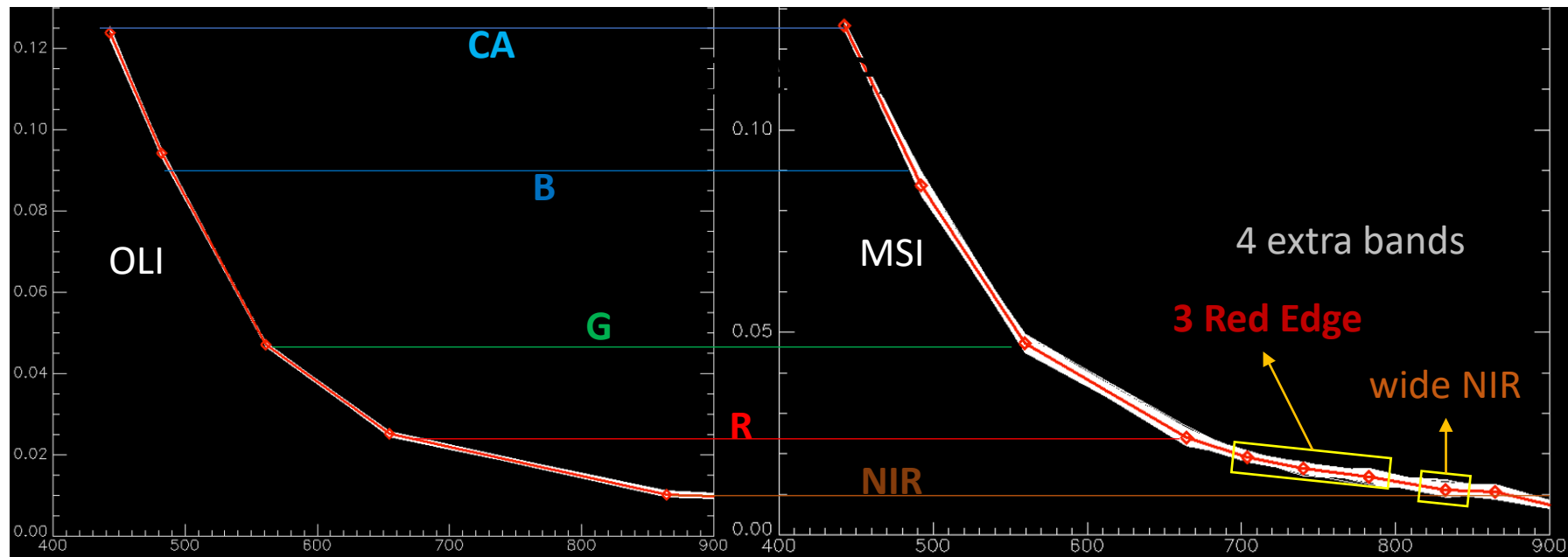


L8/9-OLI (30m)

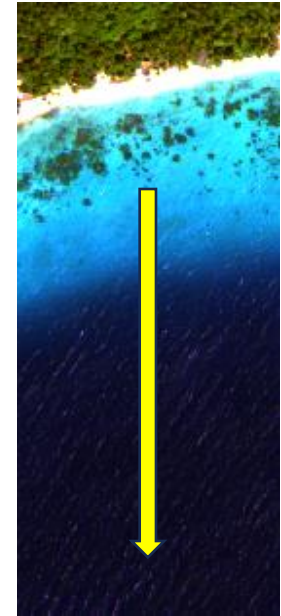
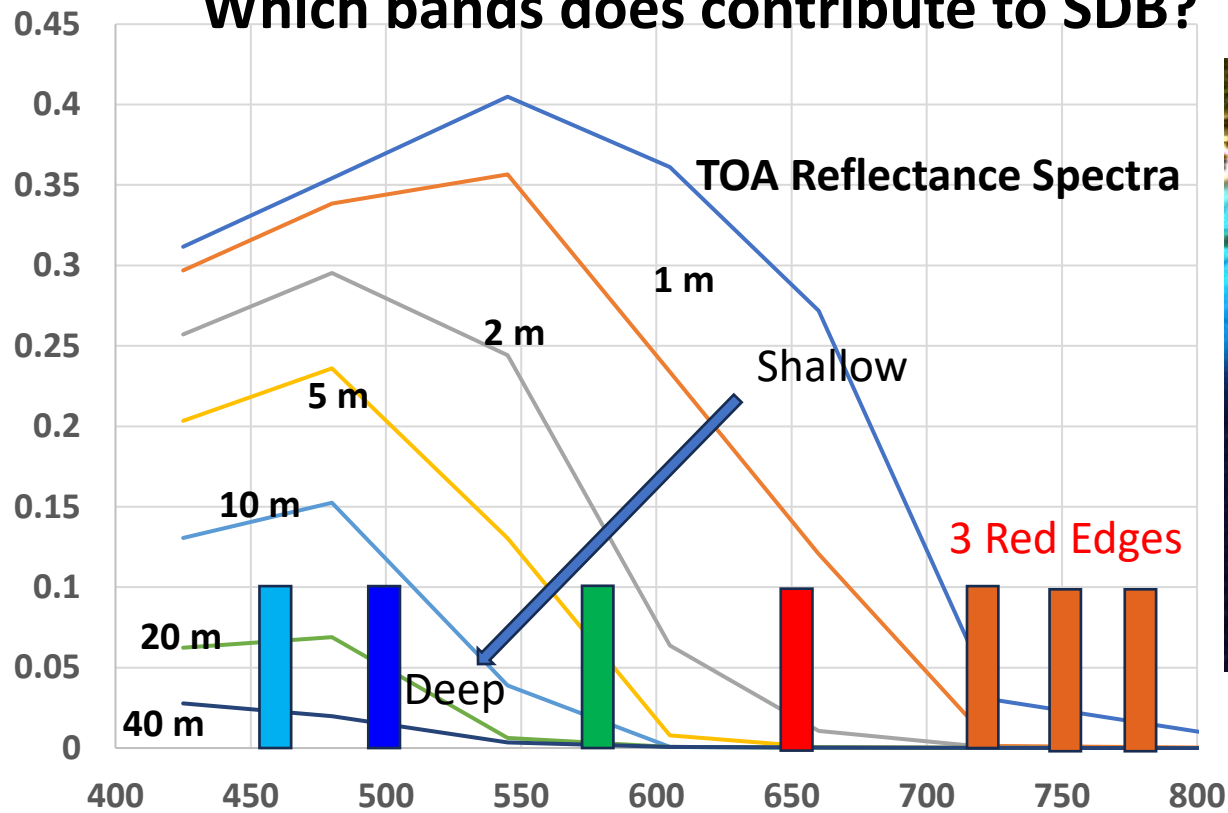




TOAR comparison (OLI vs MSI , same day), benefit of MSI's +4 bands ??

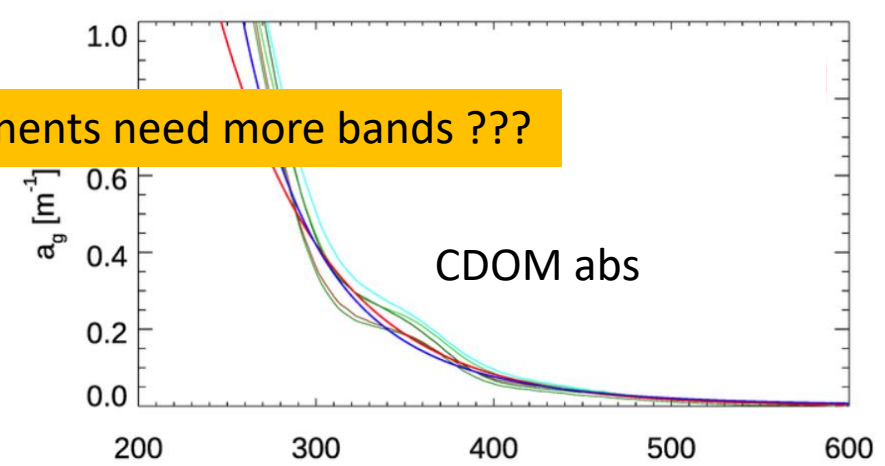
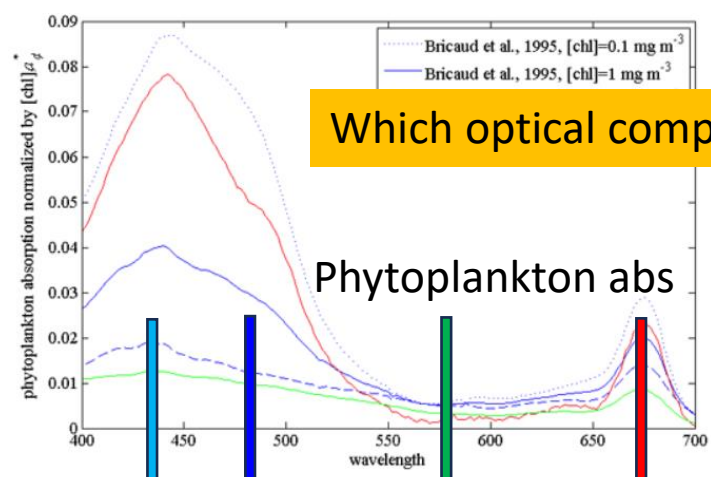


Which bands does contribute to SDB?

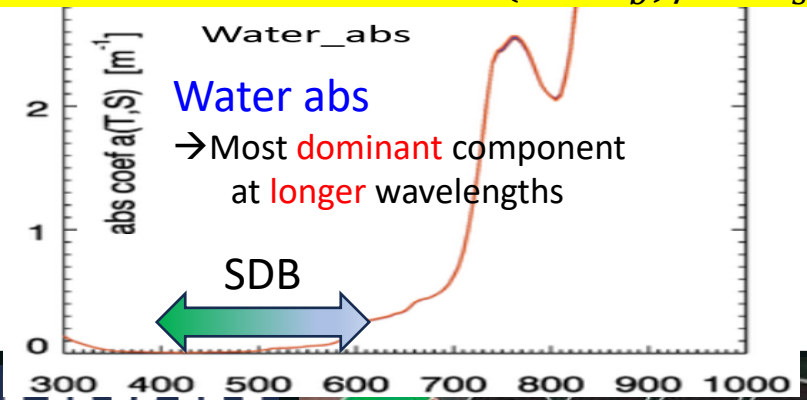
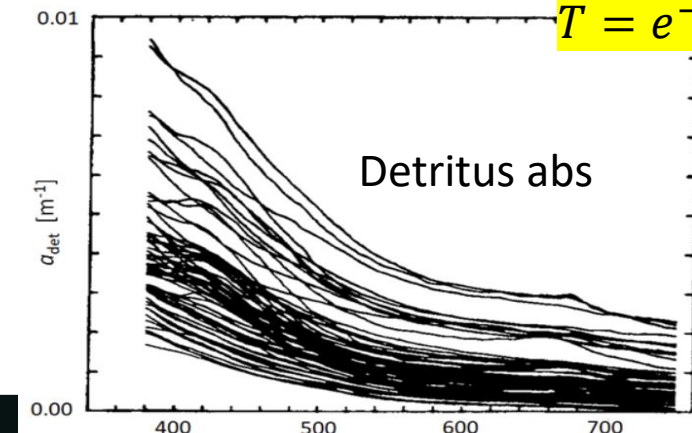


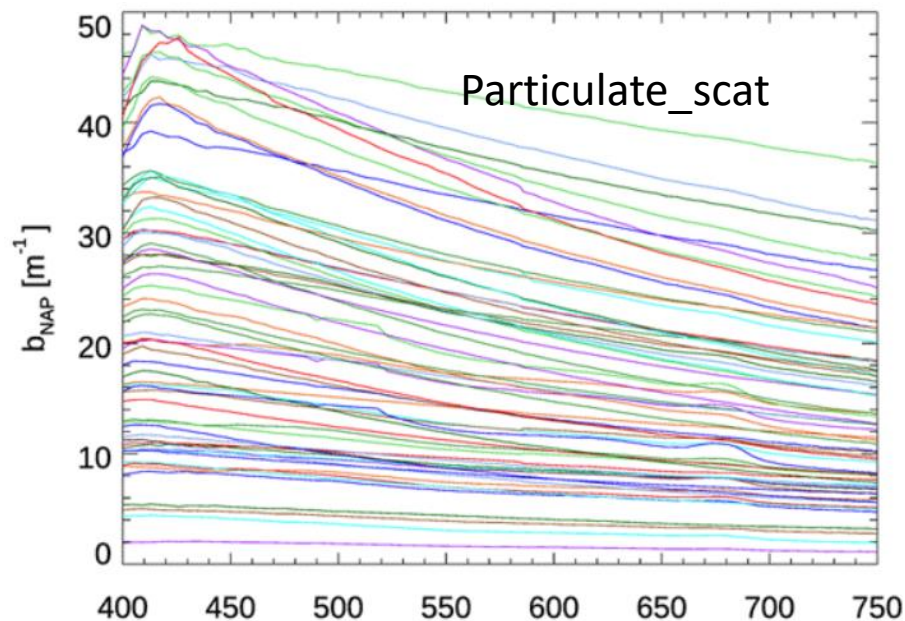
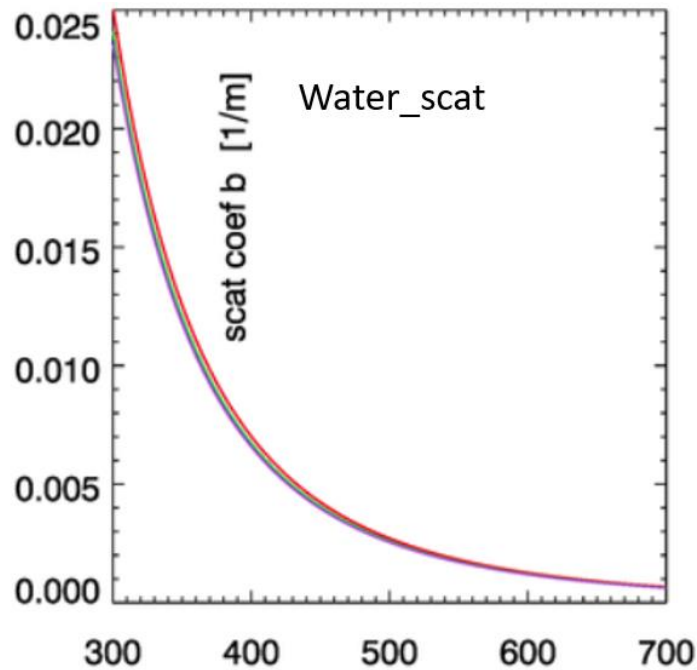
90 m

Which optical components need more bands ???



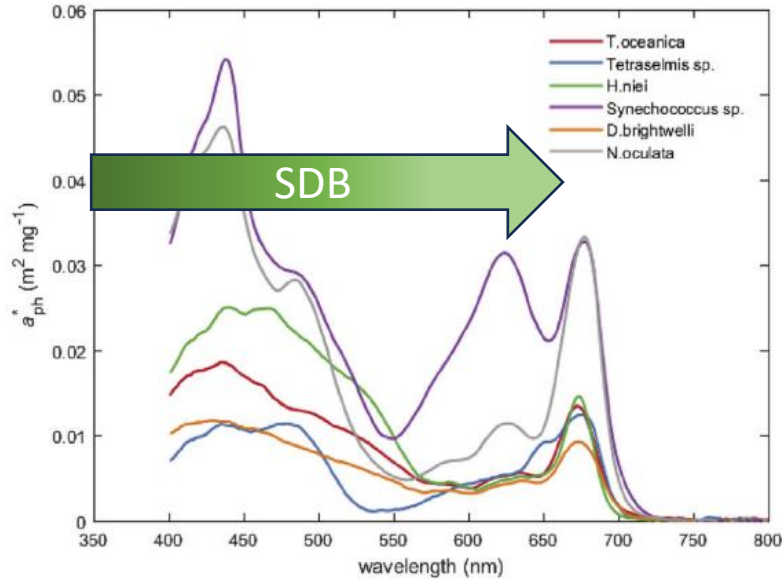
$$T = e^{-2 \cdot OD} \leftarrow OD = K \cdot z \leftarrow K = (a + b_b) / \cos\theta_s$$



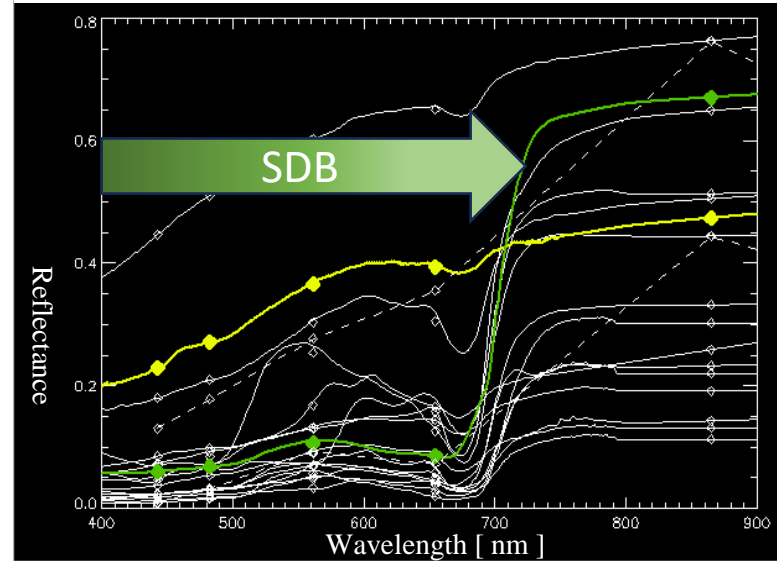


Seeking SDB-sensitive bands, what are **relevant** optical properties?

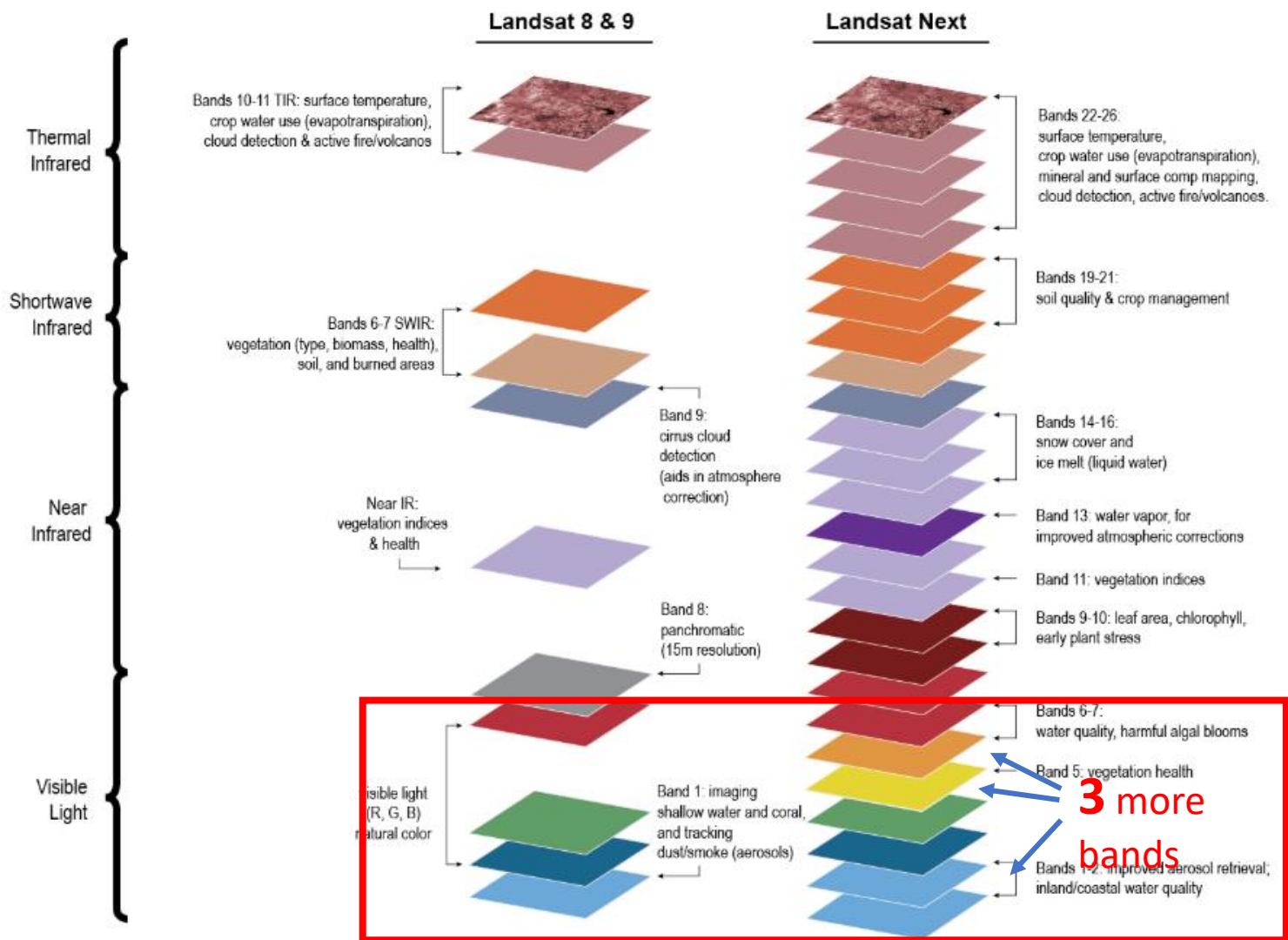
Phytoplankton species



Bottom Reflectance Library



Above two components get benefit from more bands.



More bands → better SDB ??

Sentinel2 MSI

Landsat 8/9 OLI

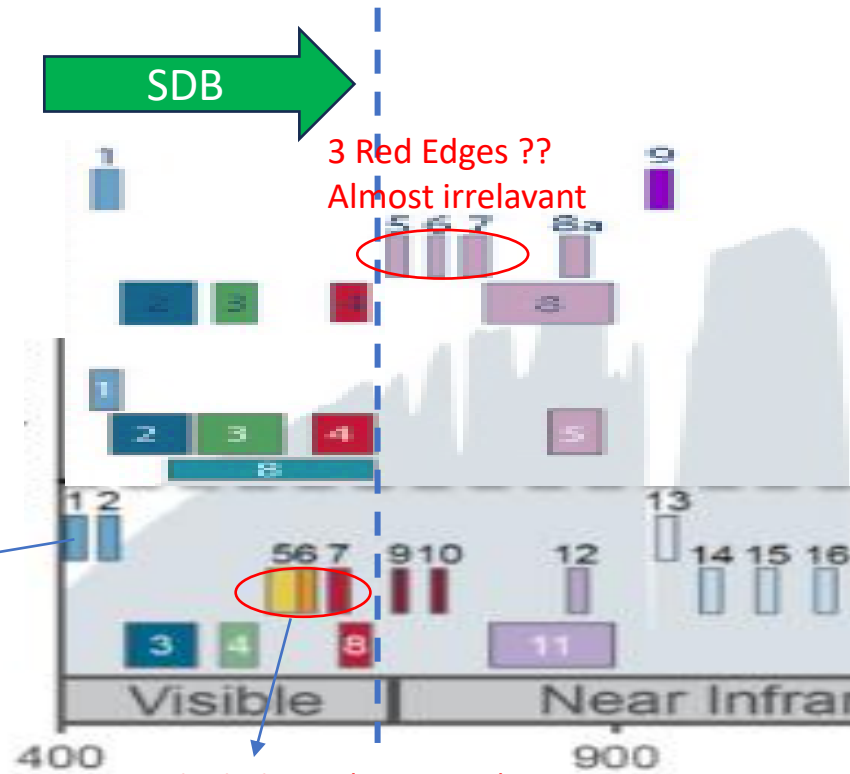
Landsat Next

Aerosol
CDOM
CHL-a

SDB

3 Red Edges ??
Almost irrelevant

Algal Bloom (Pigments)
Complex bottom



OLI/MSI SDB summary

Landsat 8/9 OLI & Sentinel-2 MSI are suitable sensors for SDB
due to high radiometric fidelity over optically shallow coastal water.

MSI 10 m can reveal more spatial details than OLI 30 m due to 3x smaller GSD.

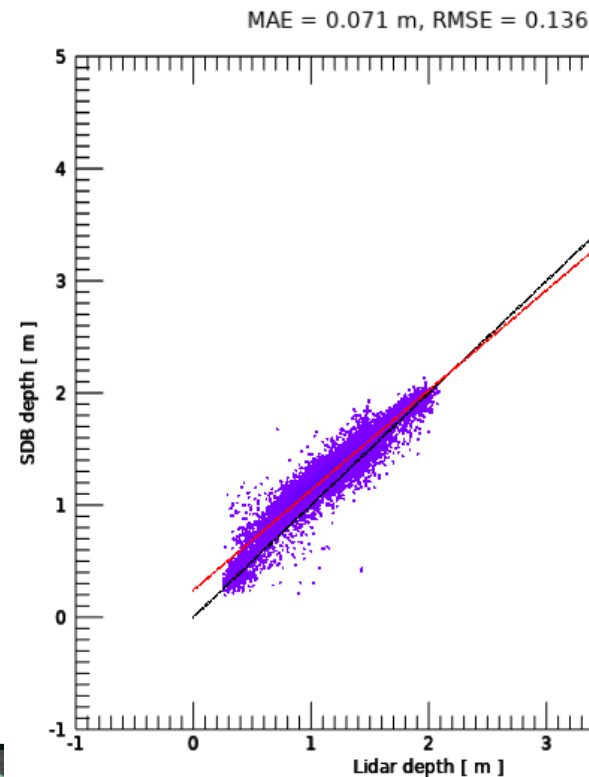
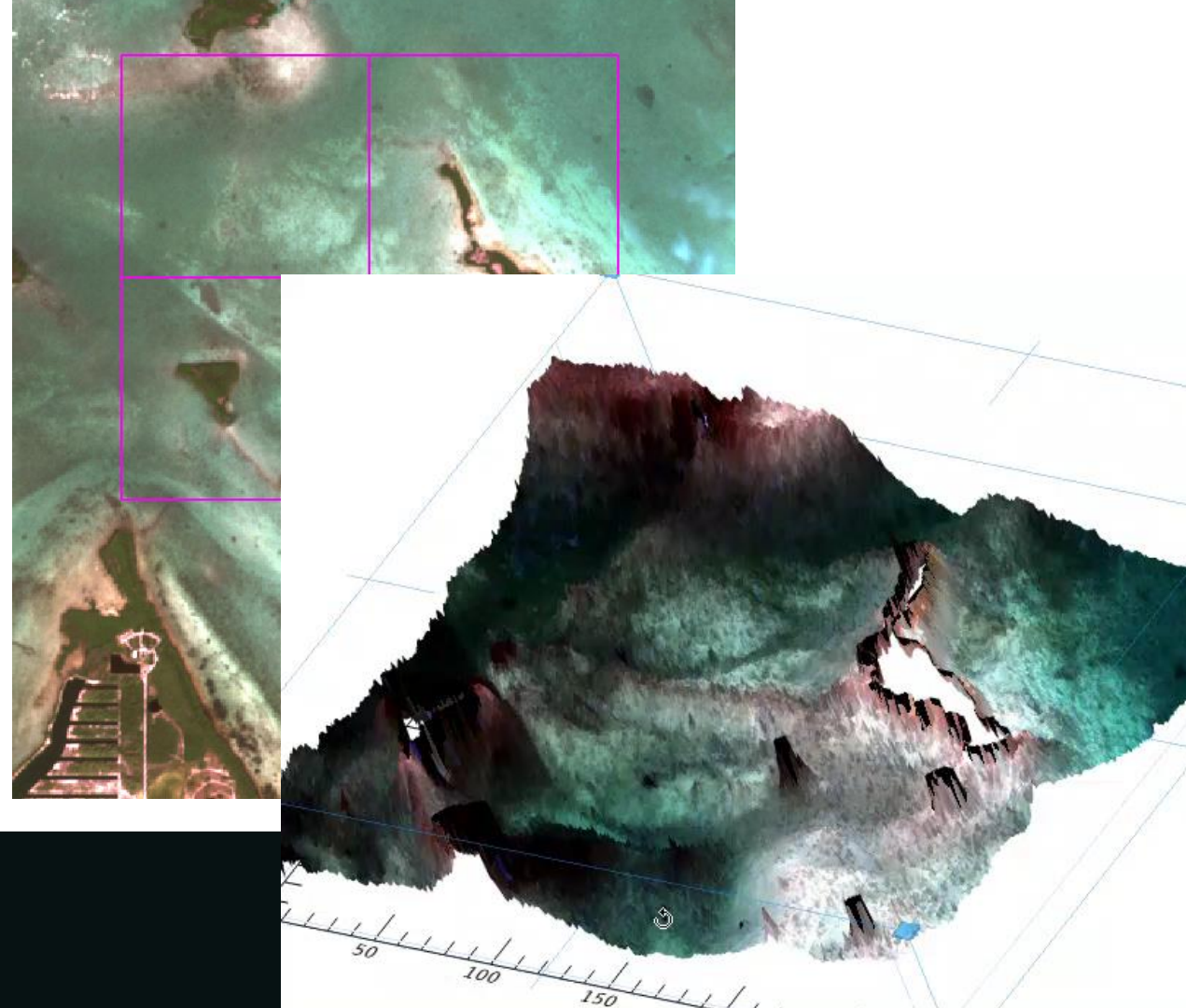
MSI scene is a bit noisier mainly due to smaller pixel size. ← MSI vs OLI SNR

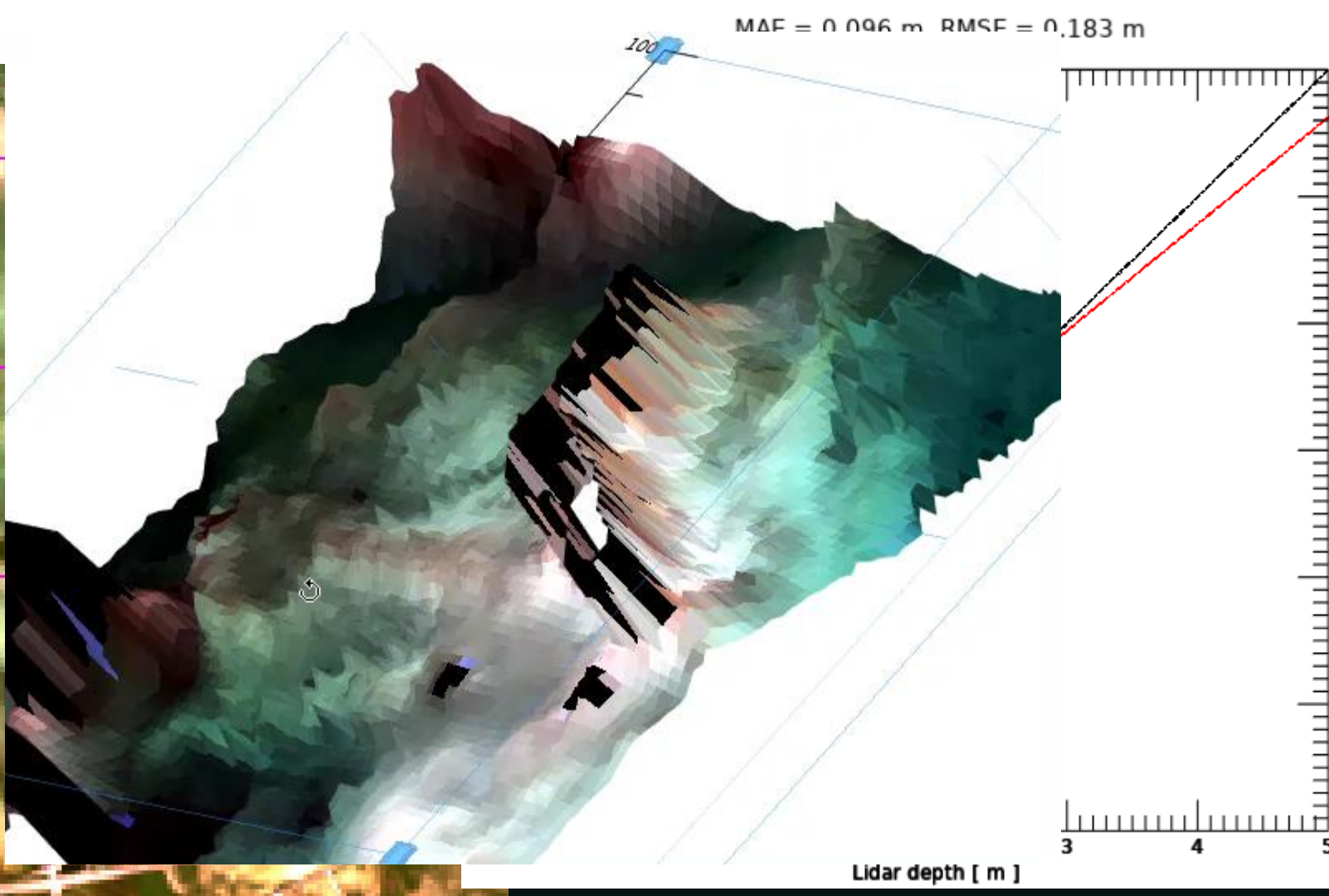
Accordingly high-res imagery will suffer from the same problem, SDB uncertainty increase
(1) Spectral smoothing ??? → multi-band too small # of points, NOT suitable for smoothing
(2) Spatial smoothing ??? → degrades the benefit of smaller GSD

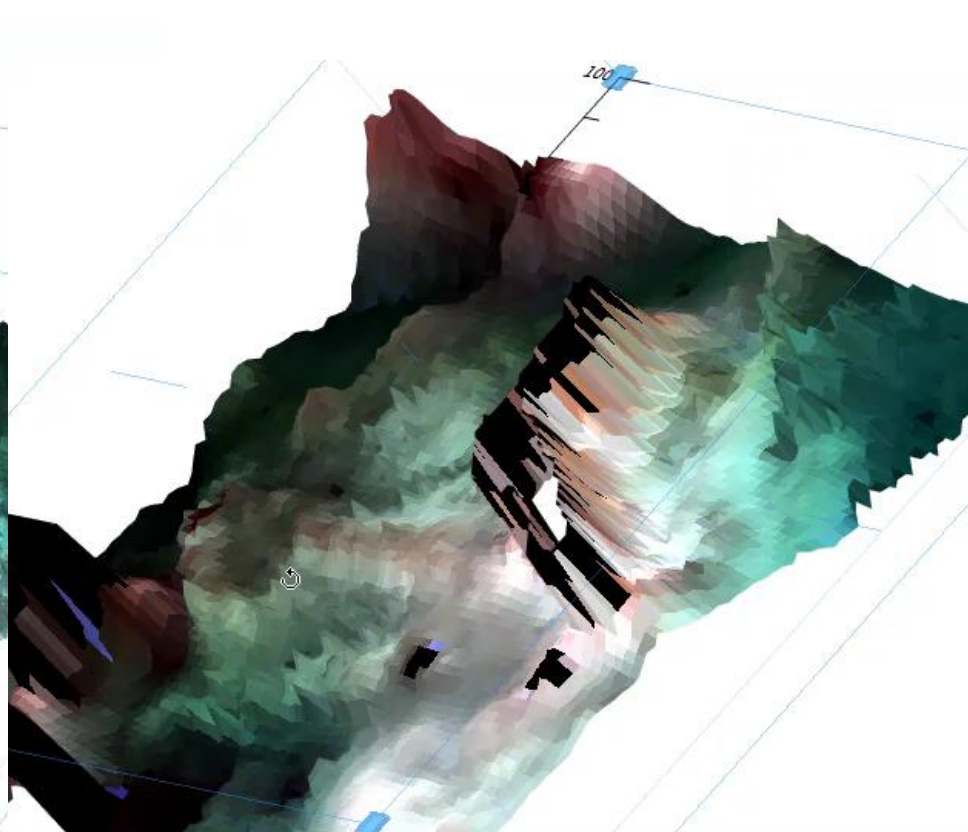
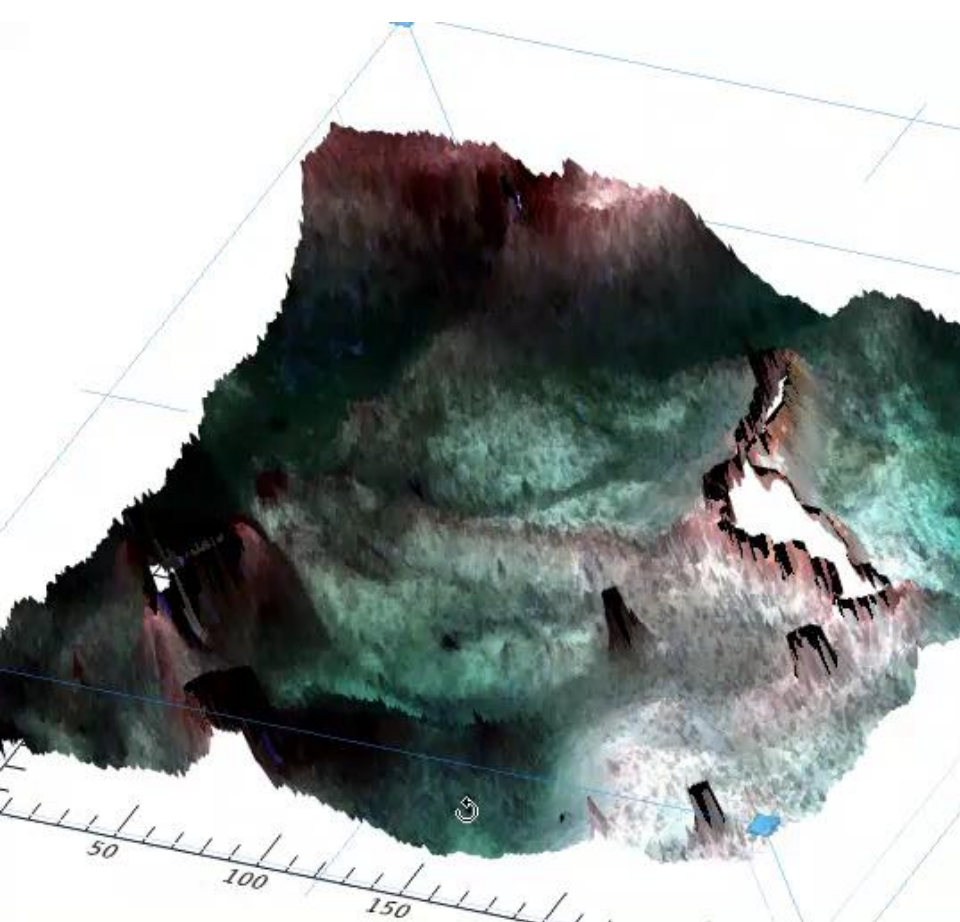
Physics-based SDB would get benefit from Landsat-Next sensor

(1) additional 3 VIS bands (2) improved GSD (10 m).



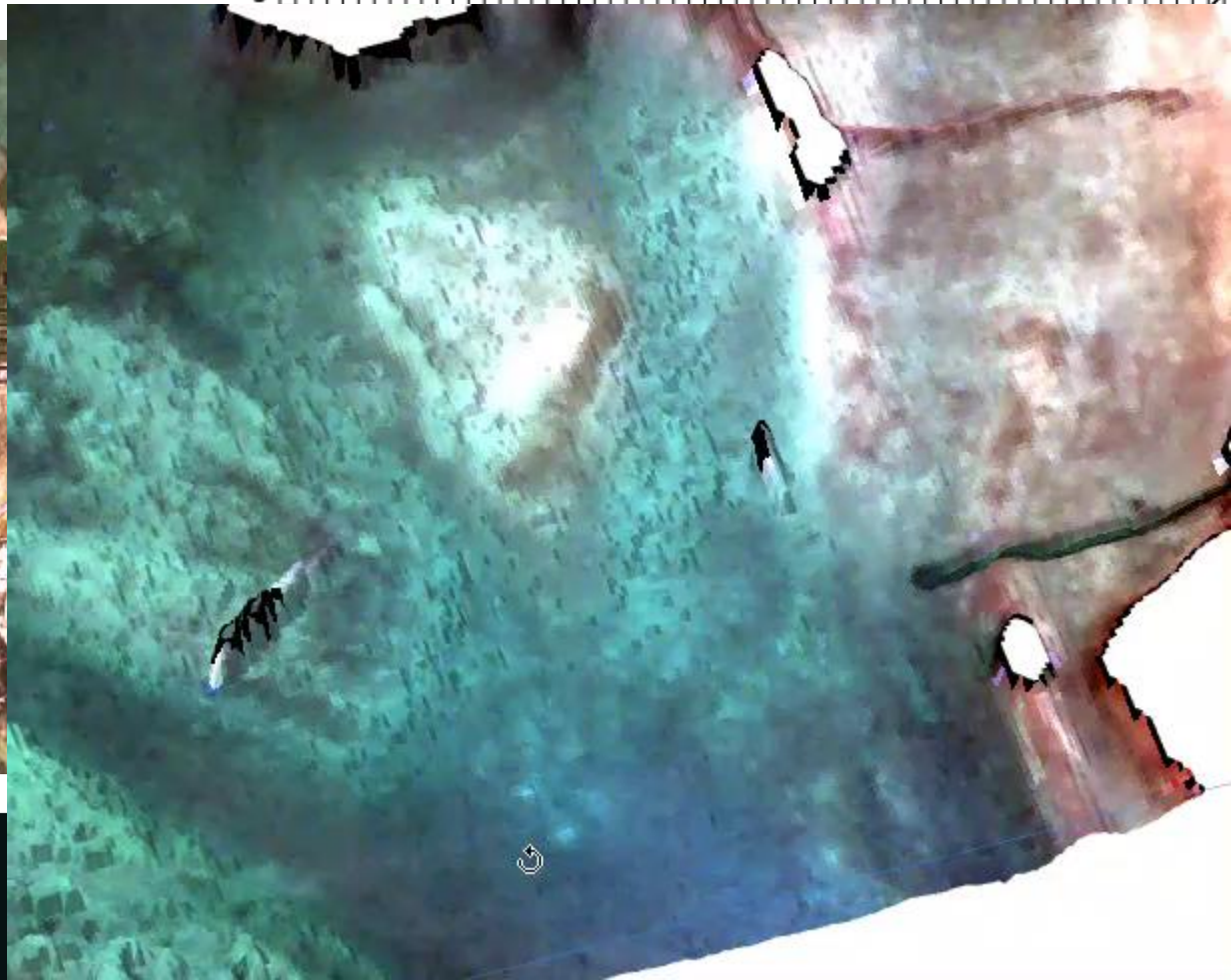


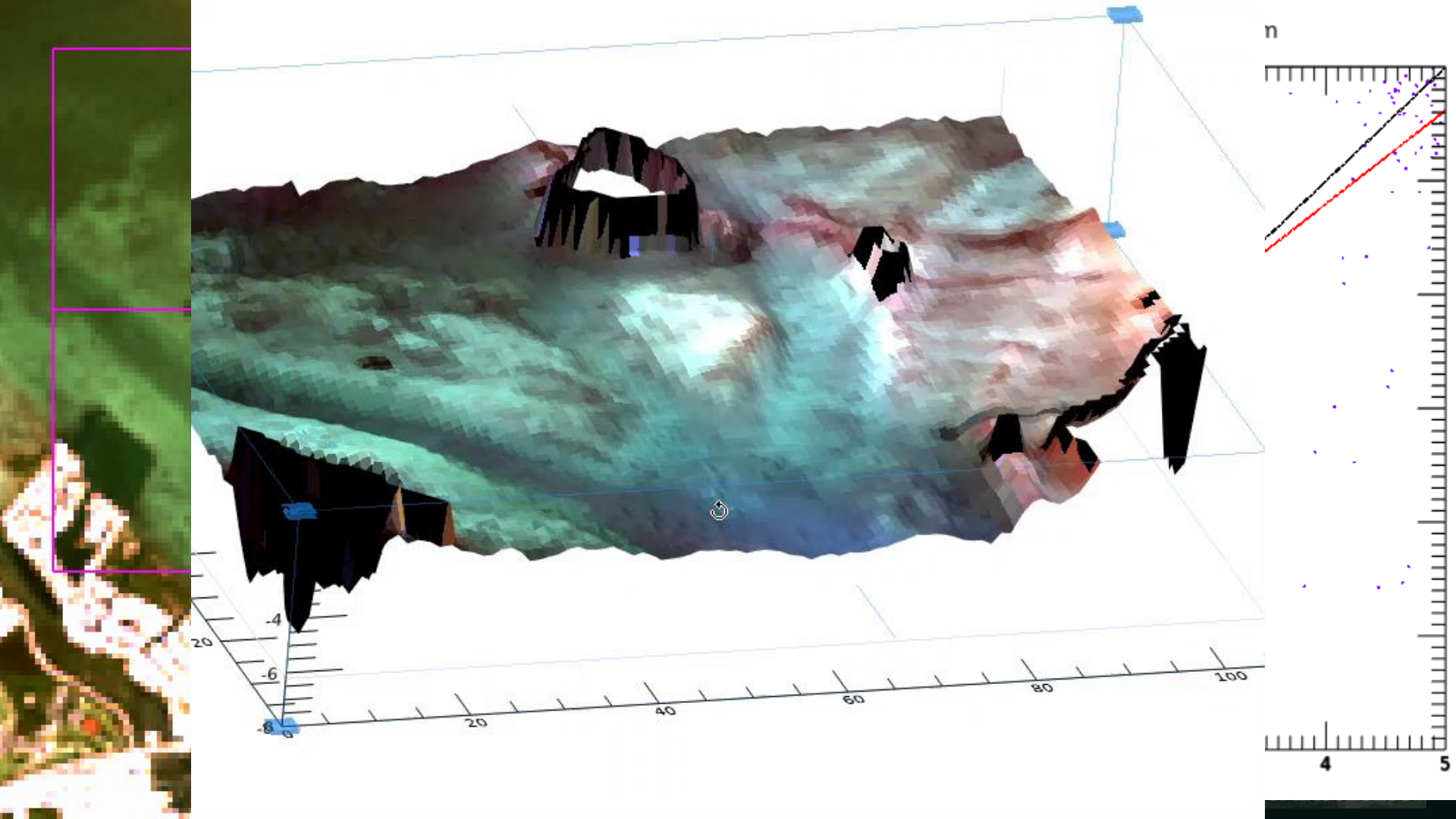


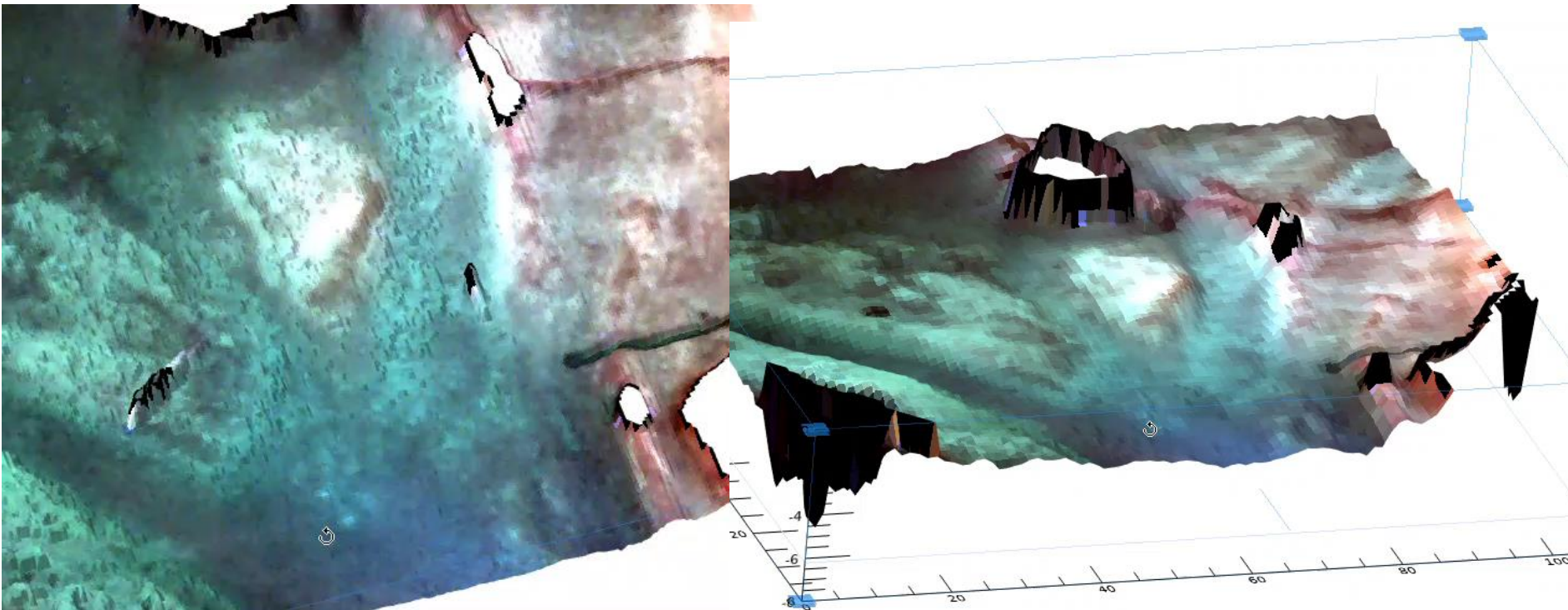


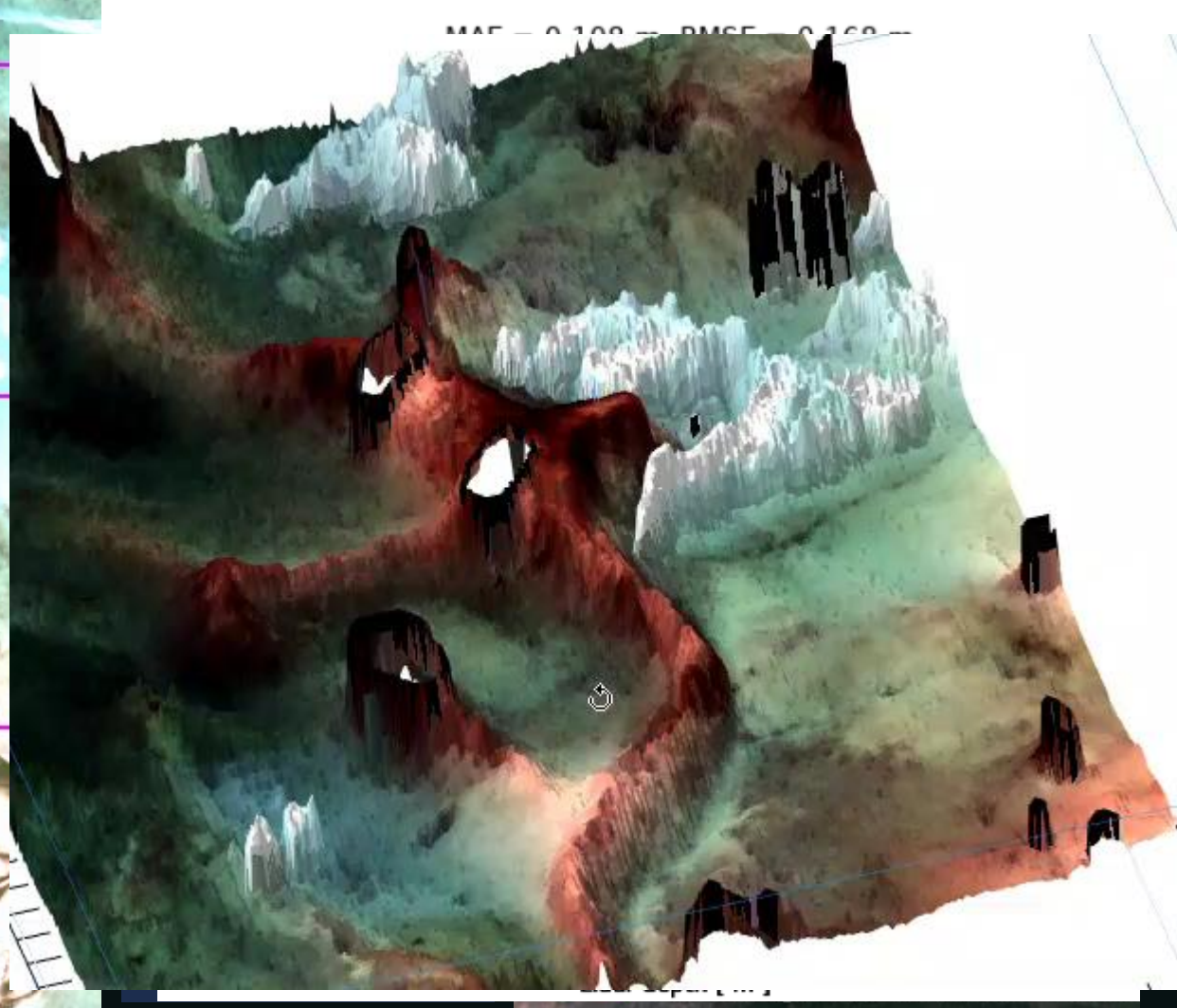
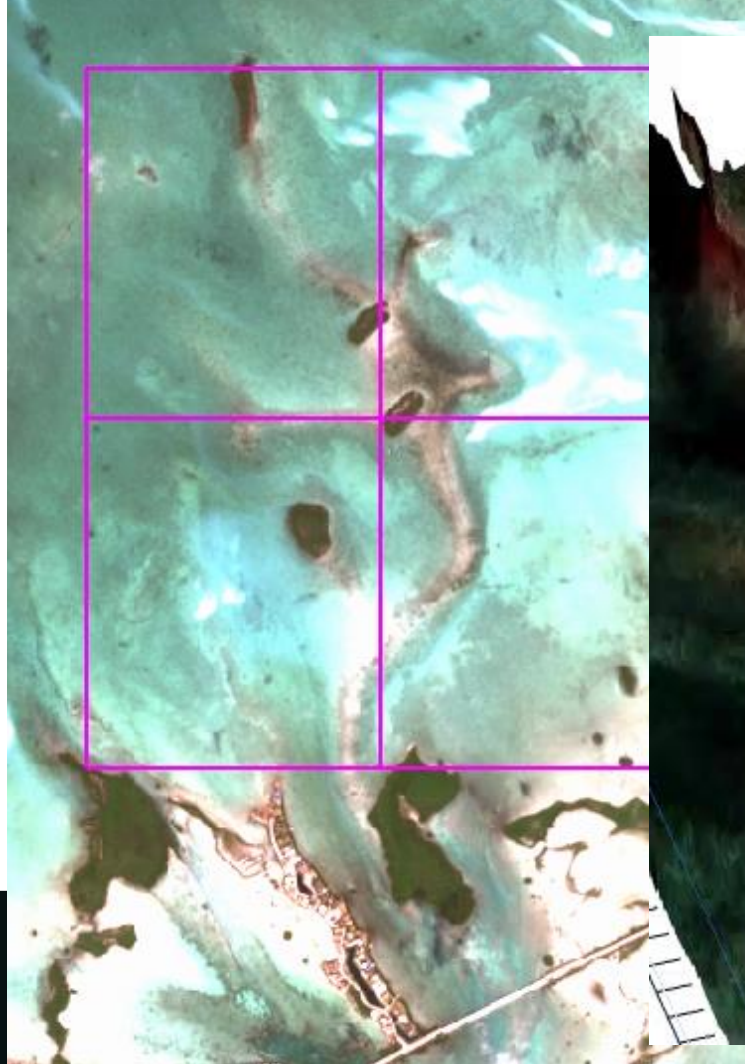
MAE = 0.215 m, RMSE = 0.226 m

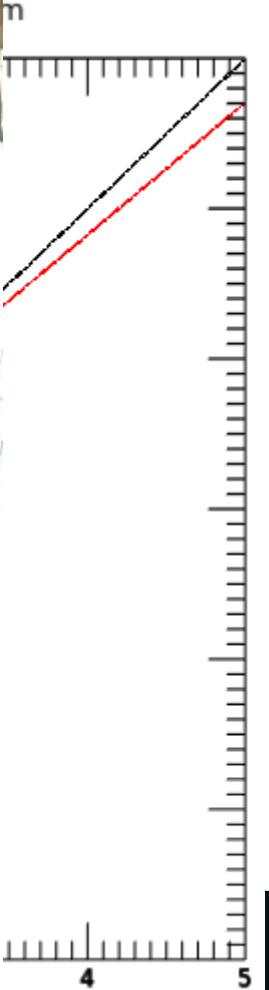
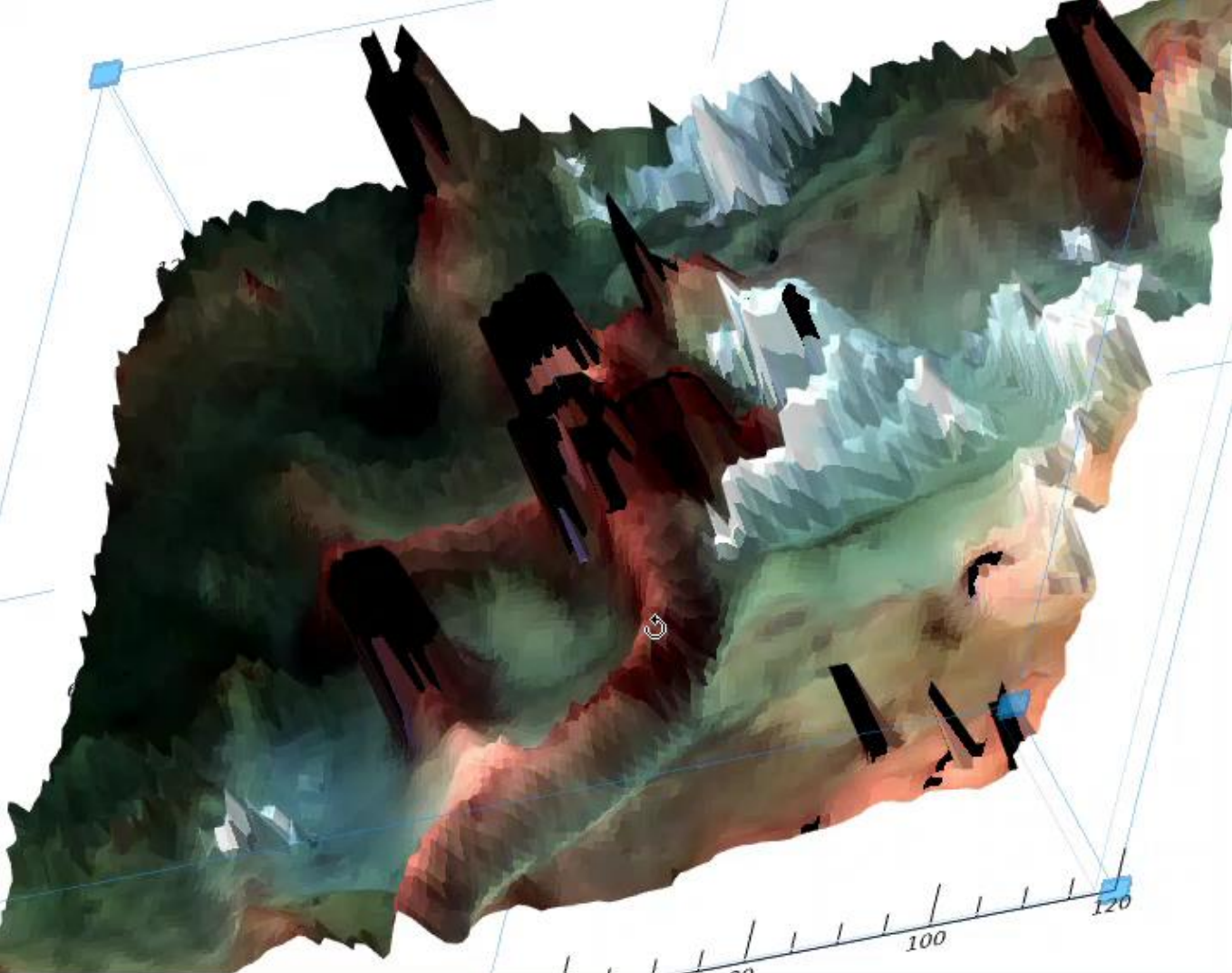
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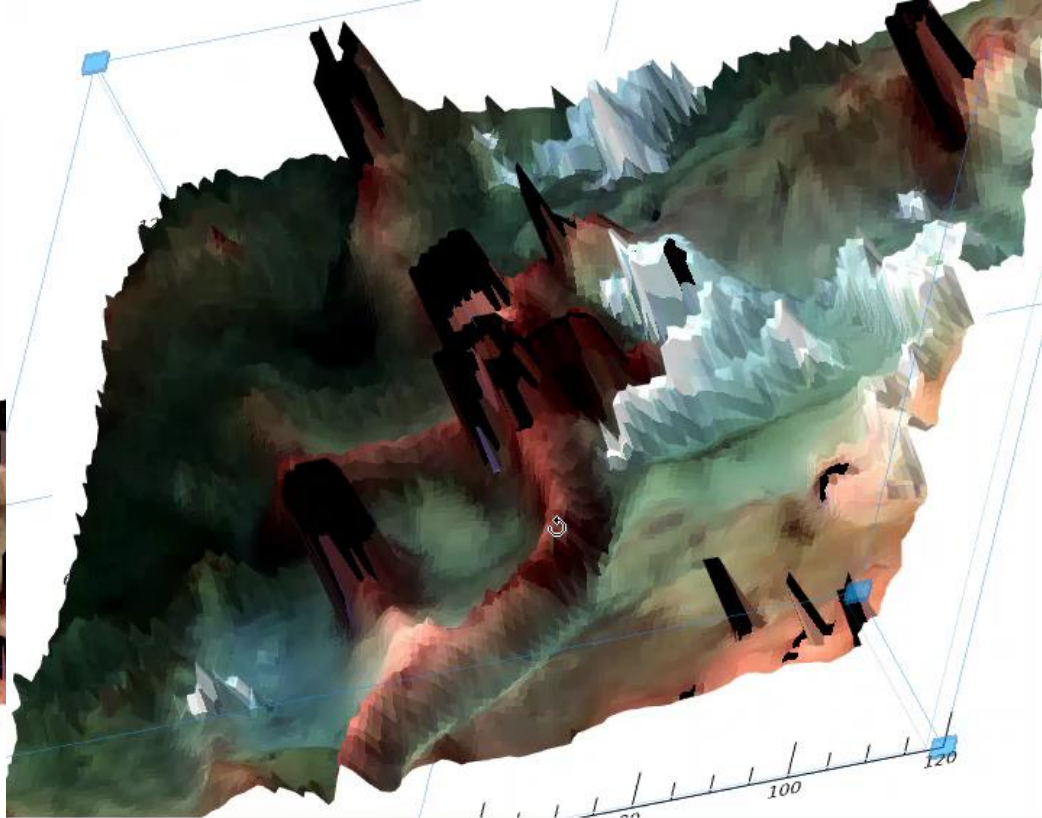
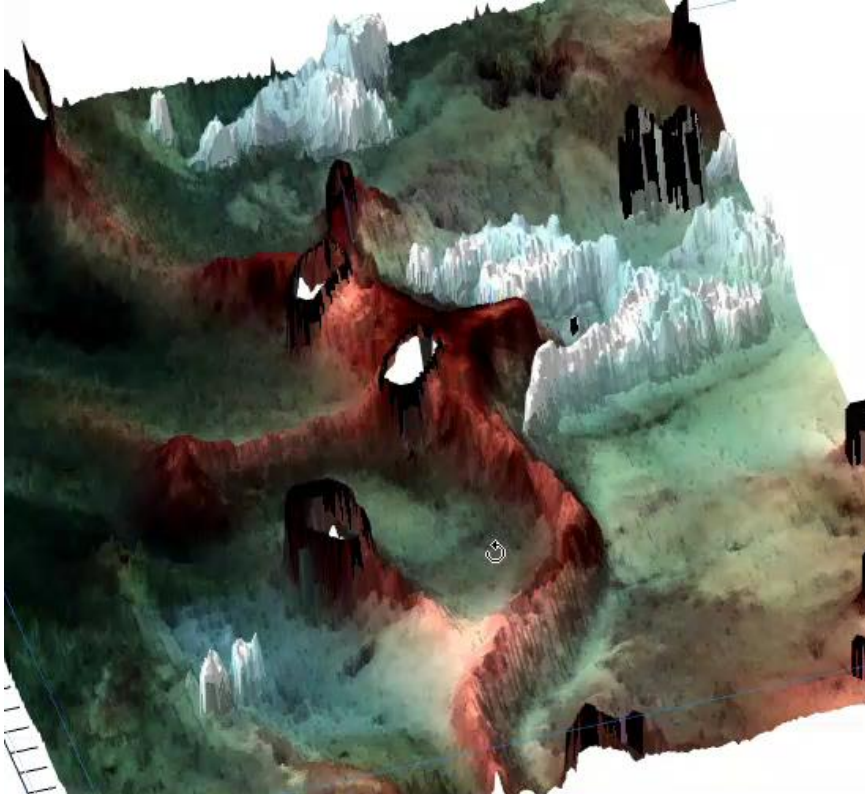


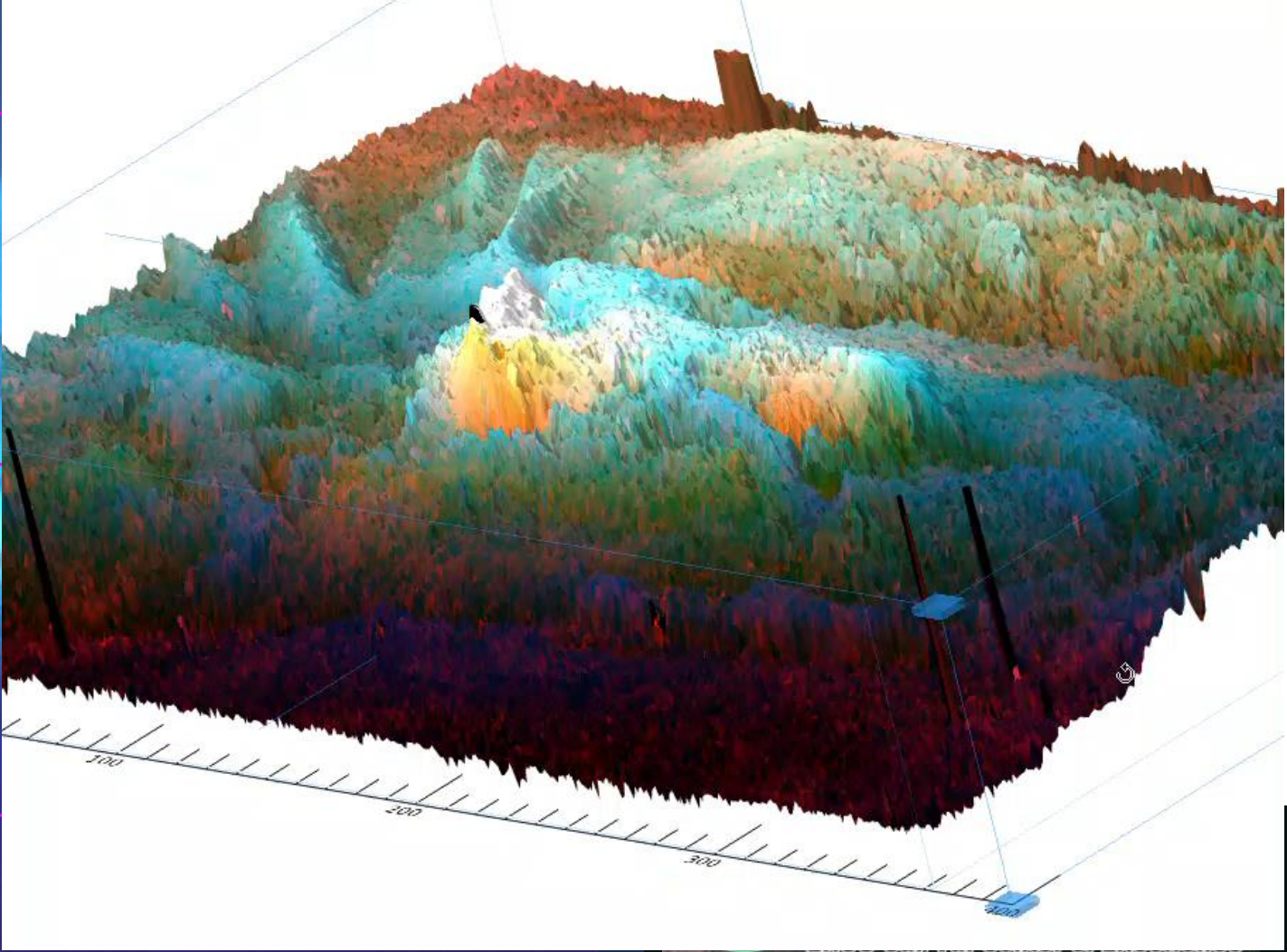
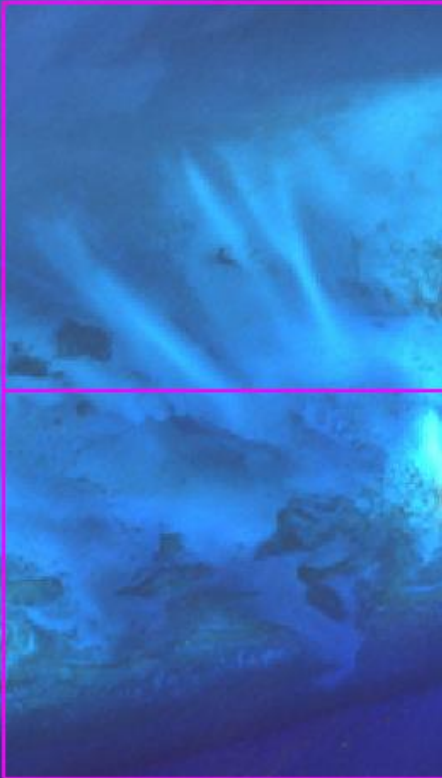


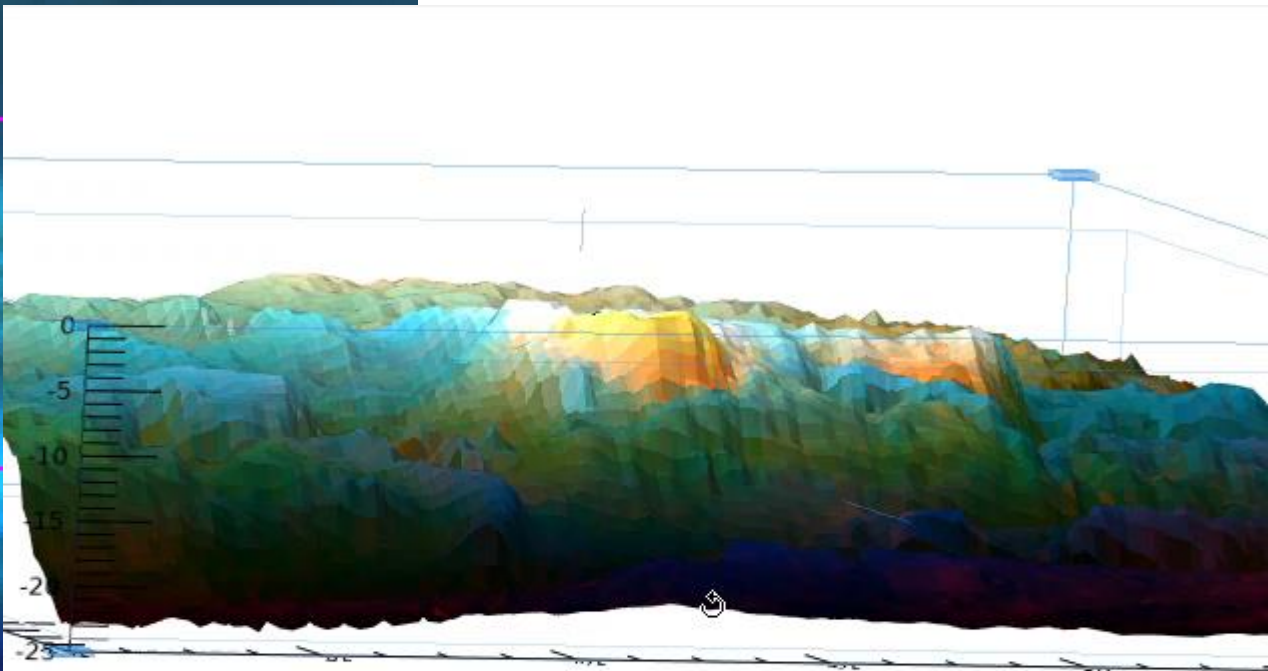
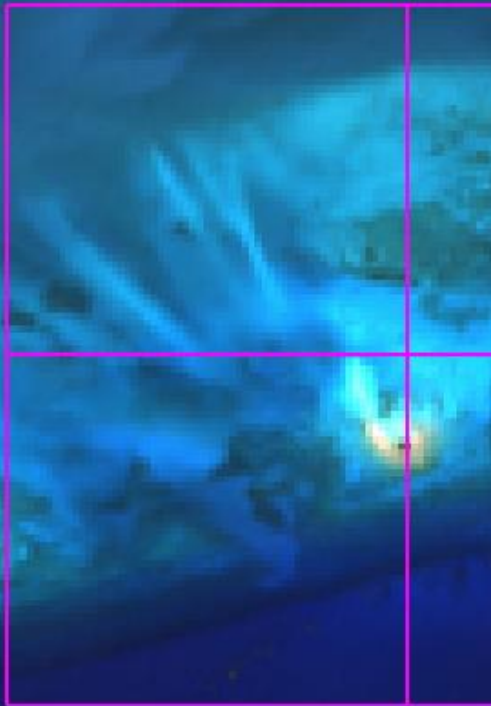


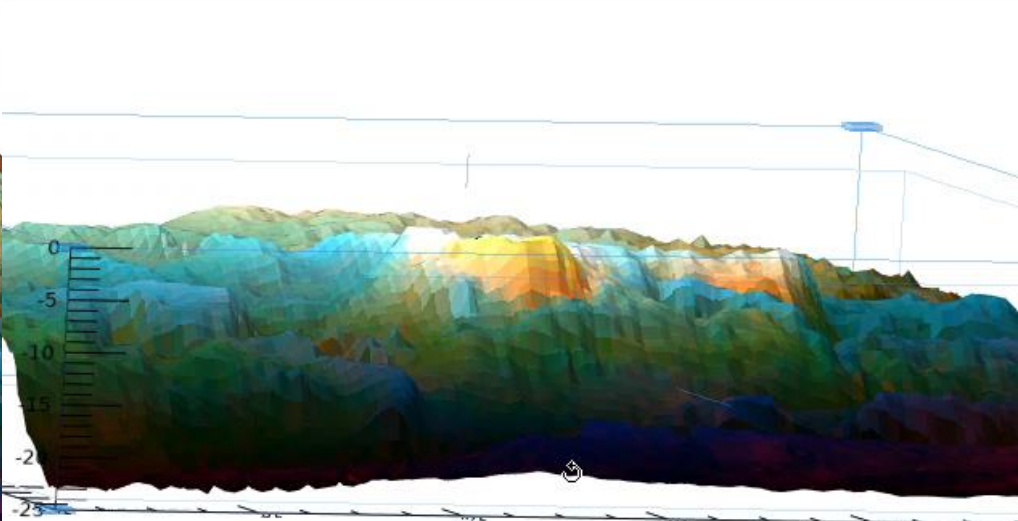
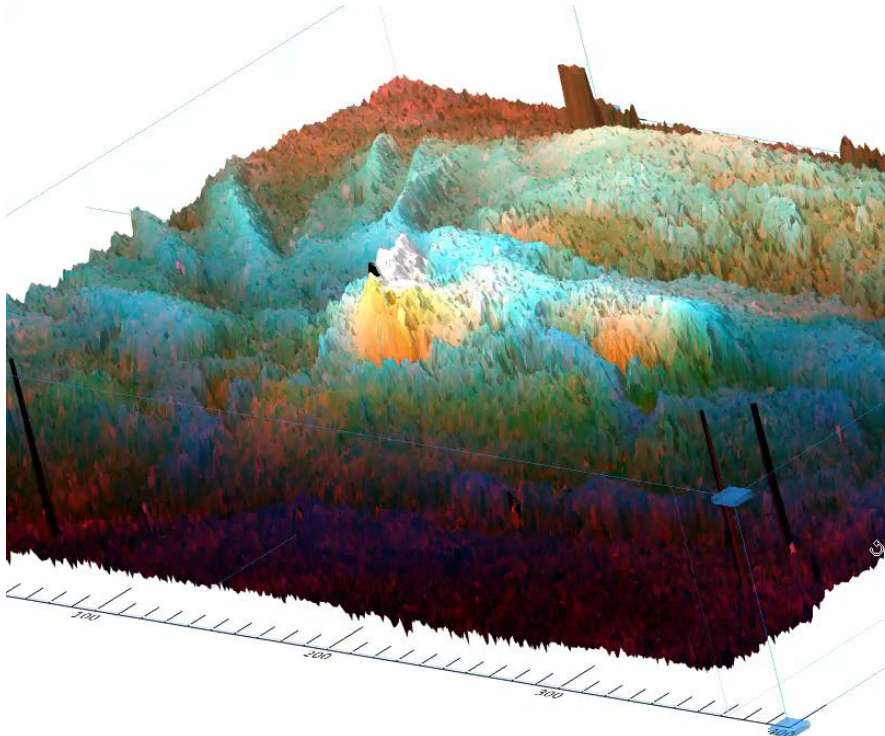


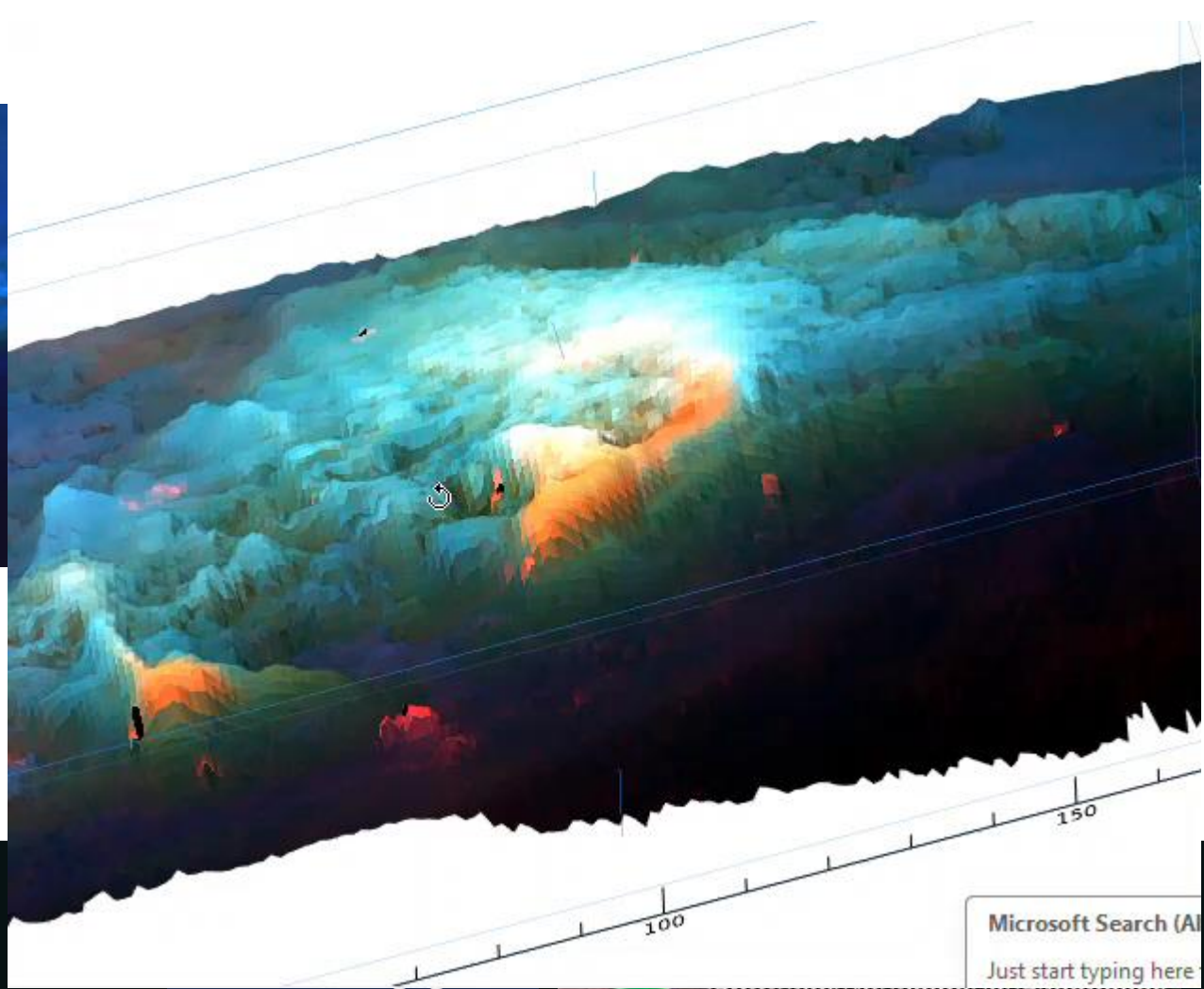
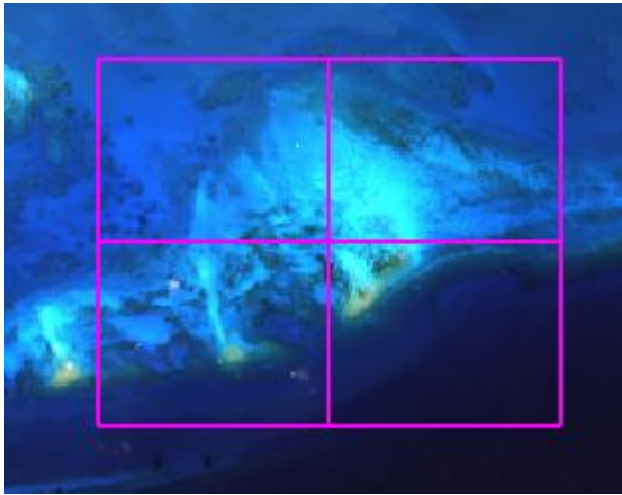


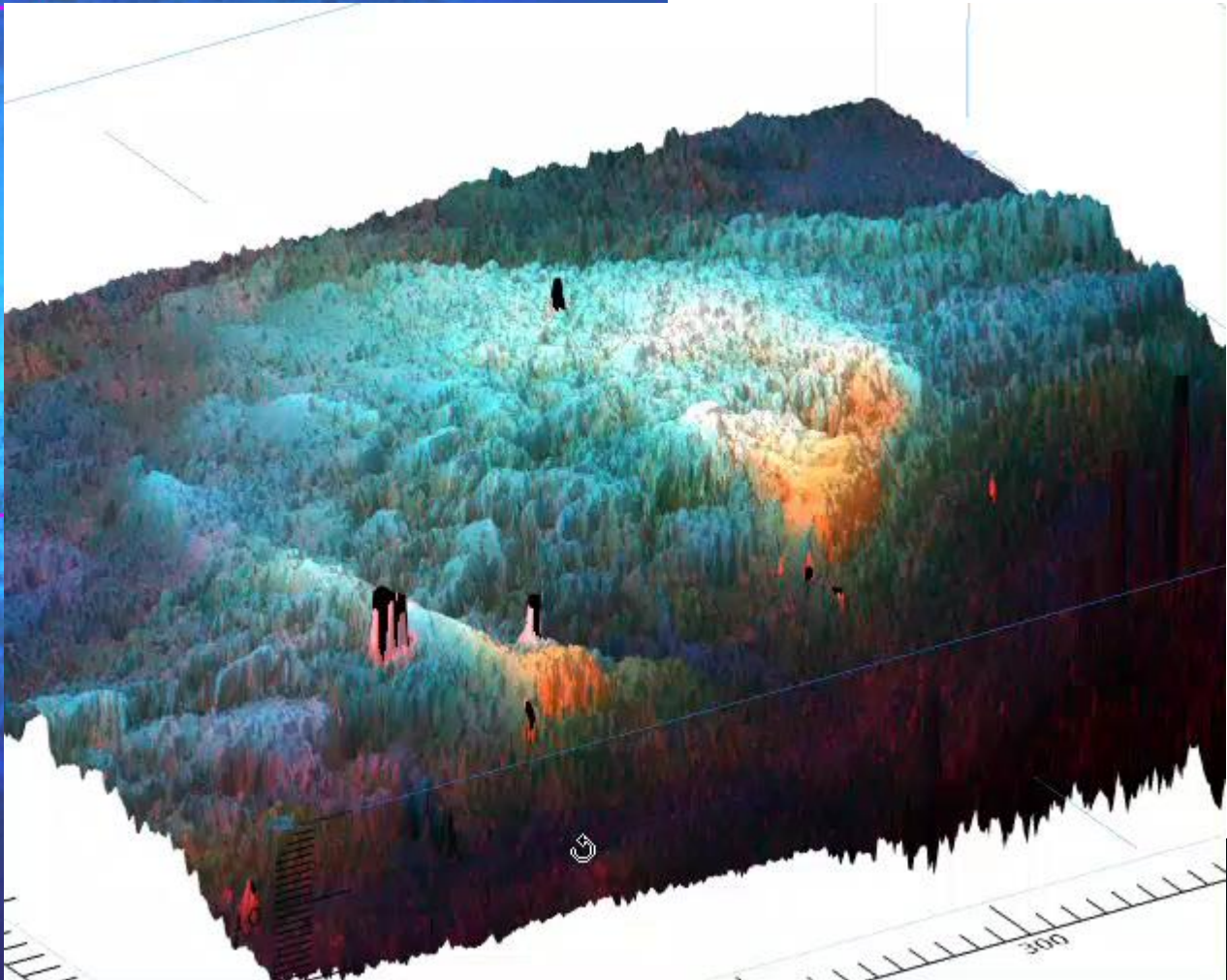
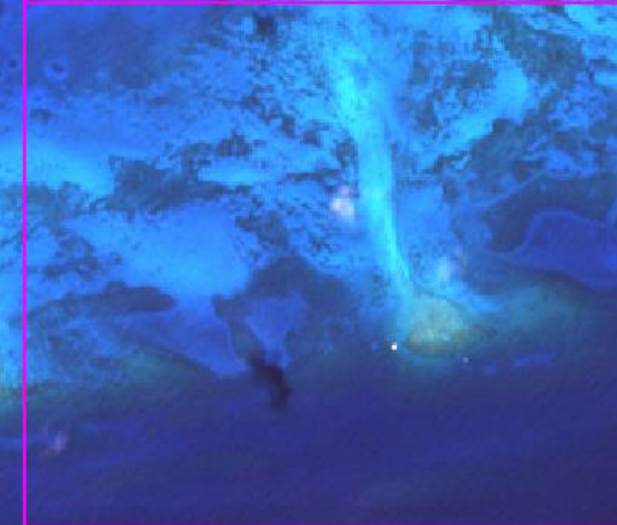
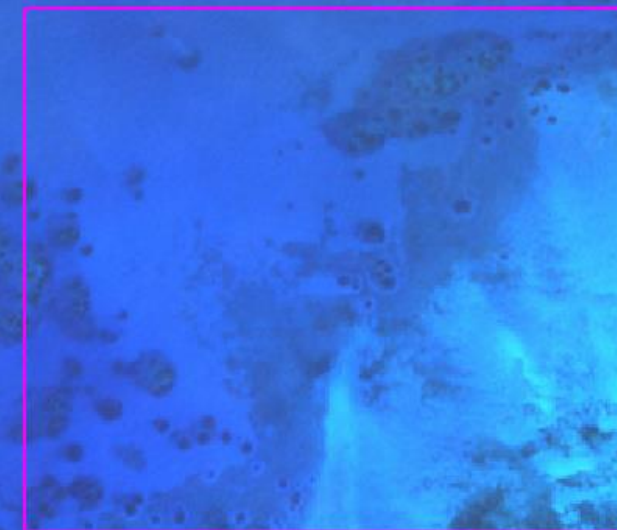


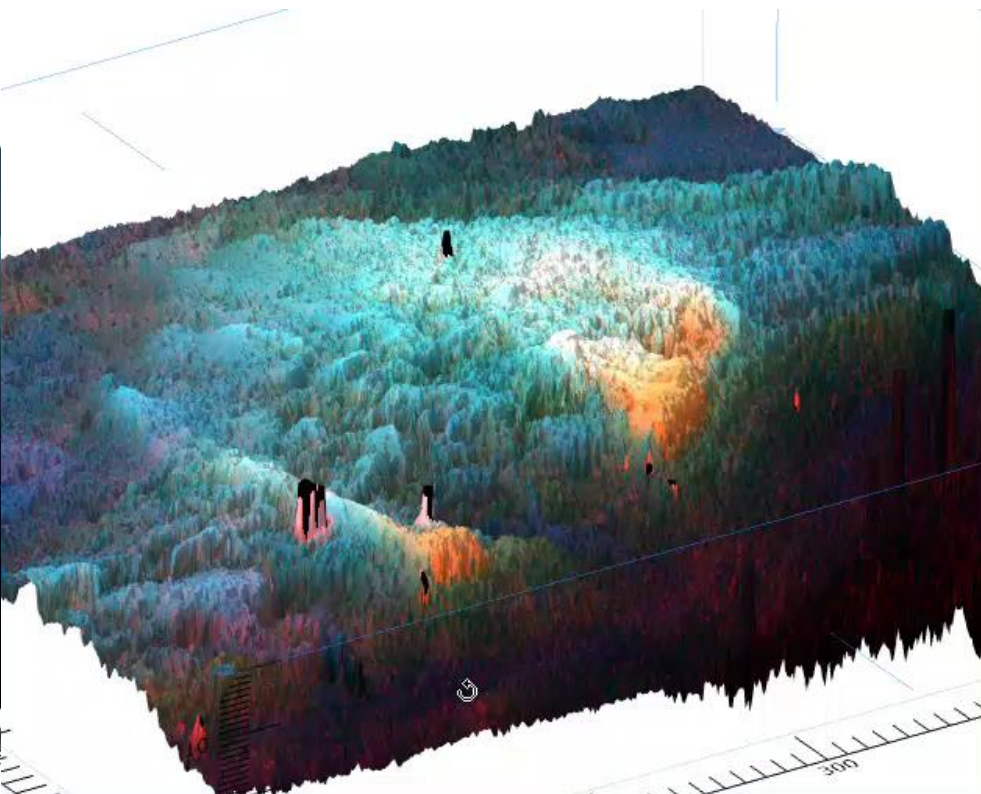
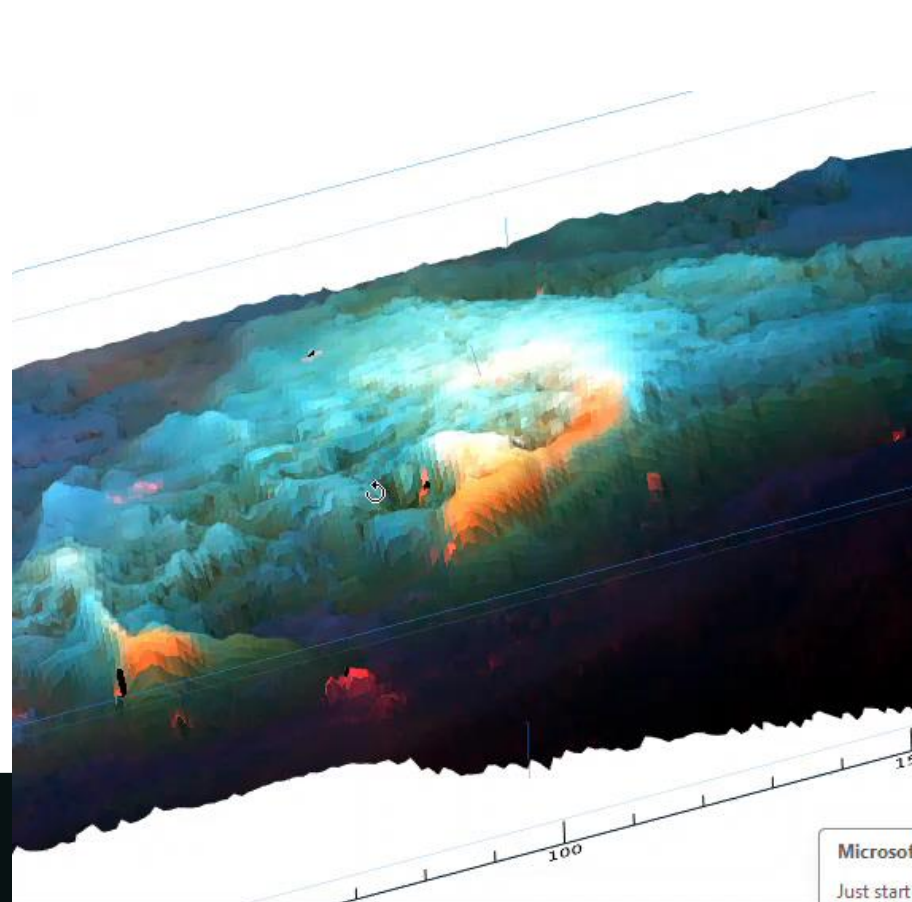






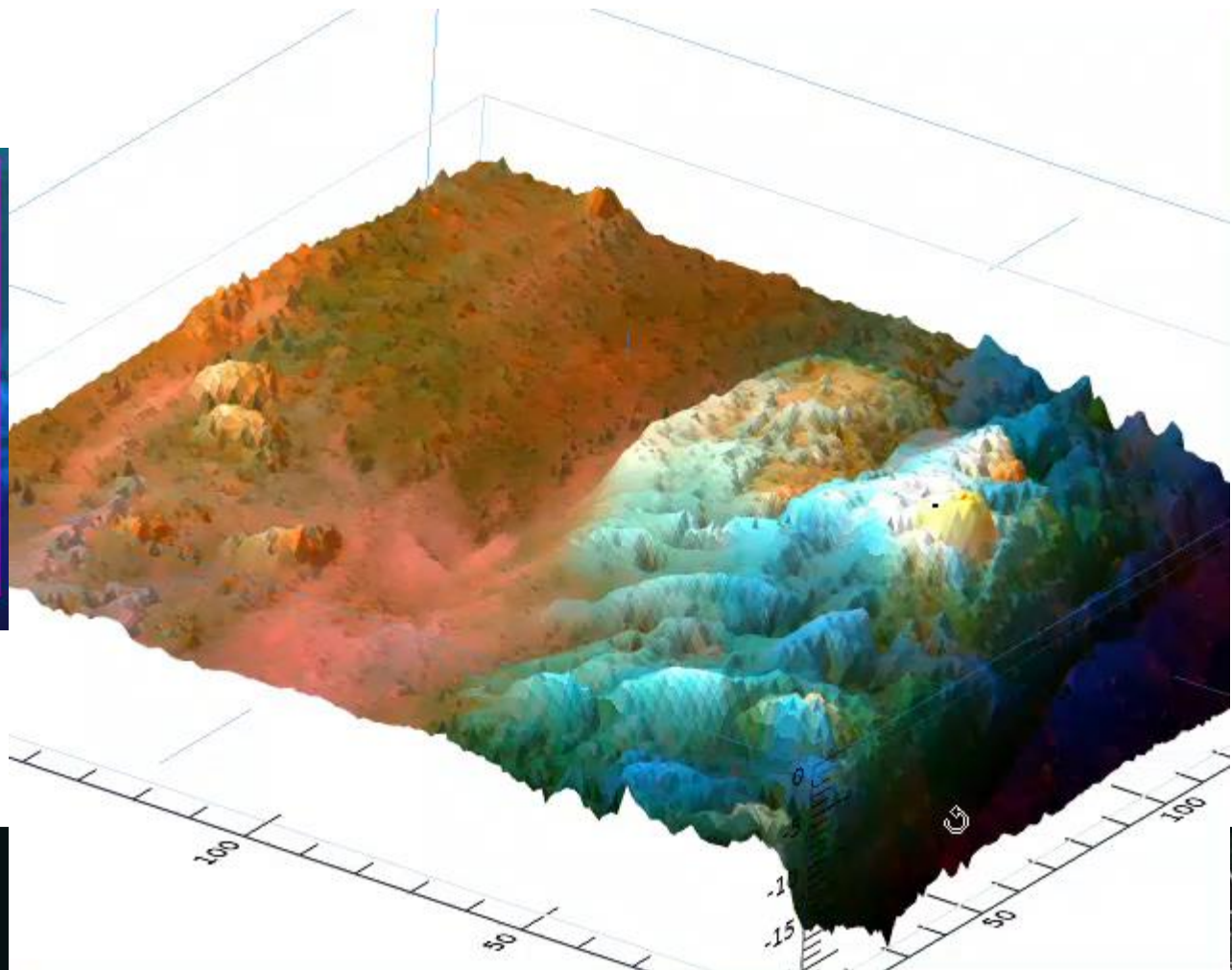
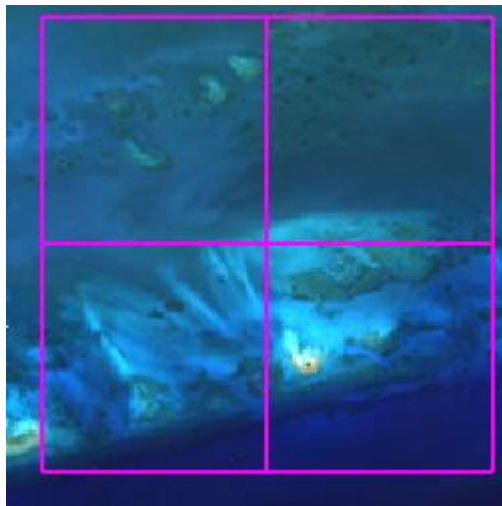


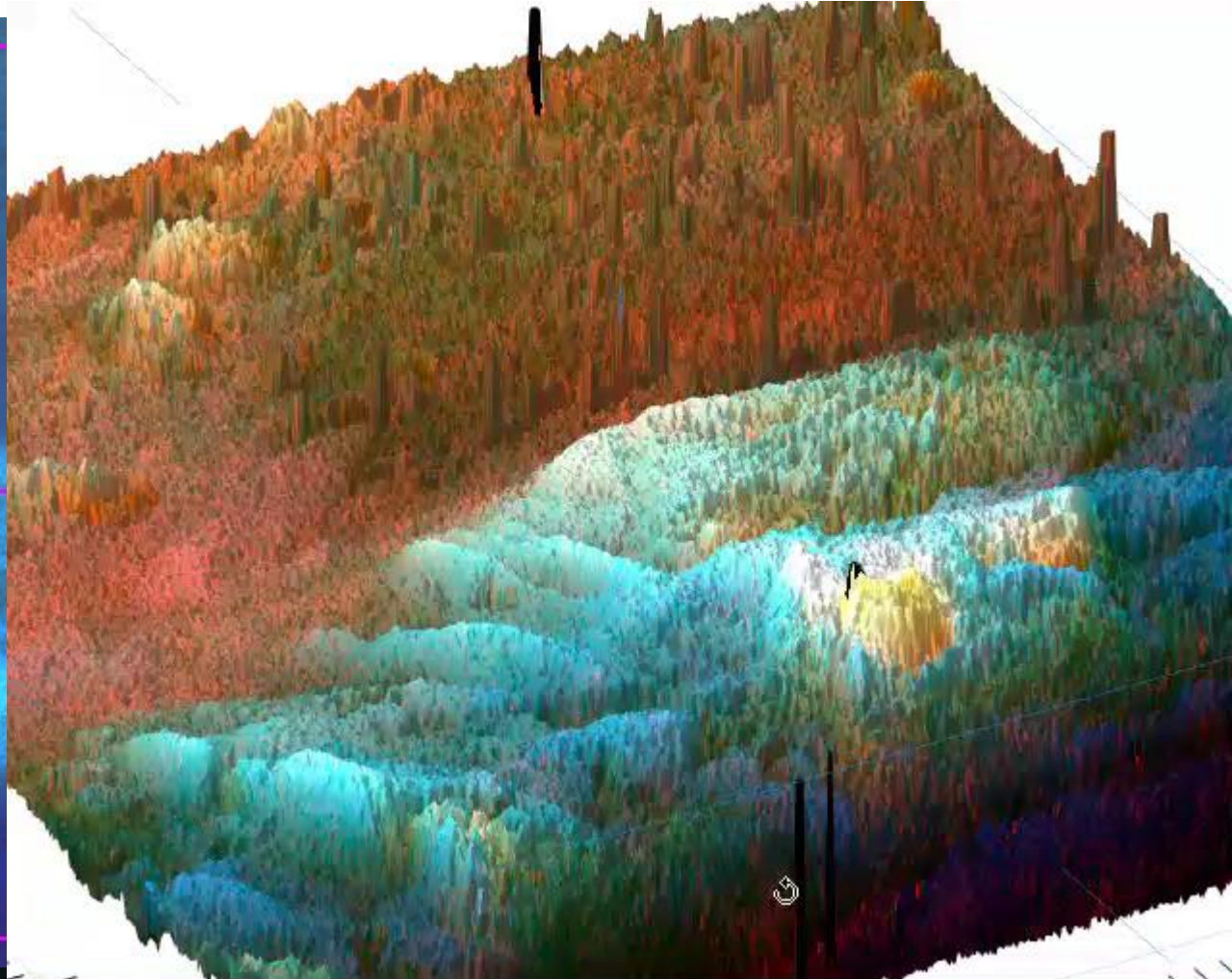
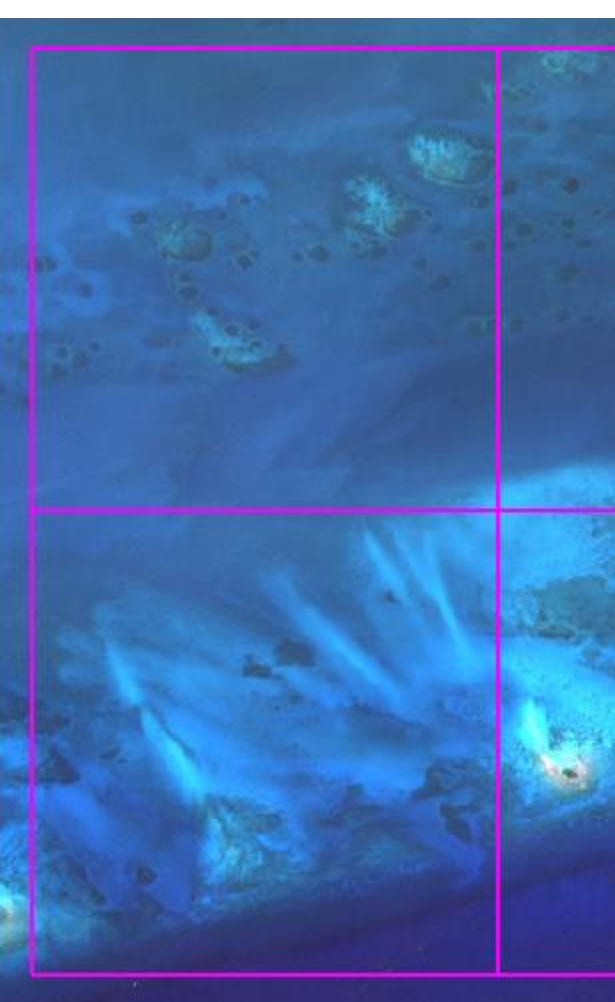




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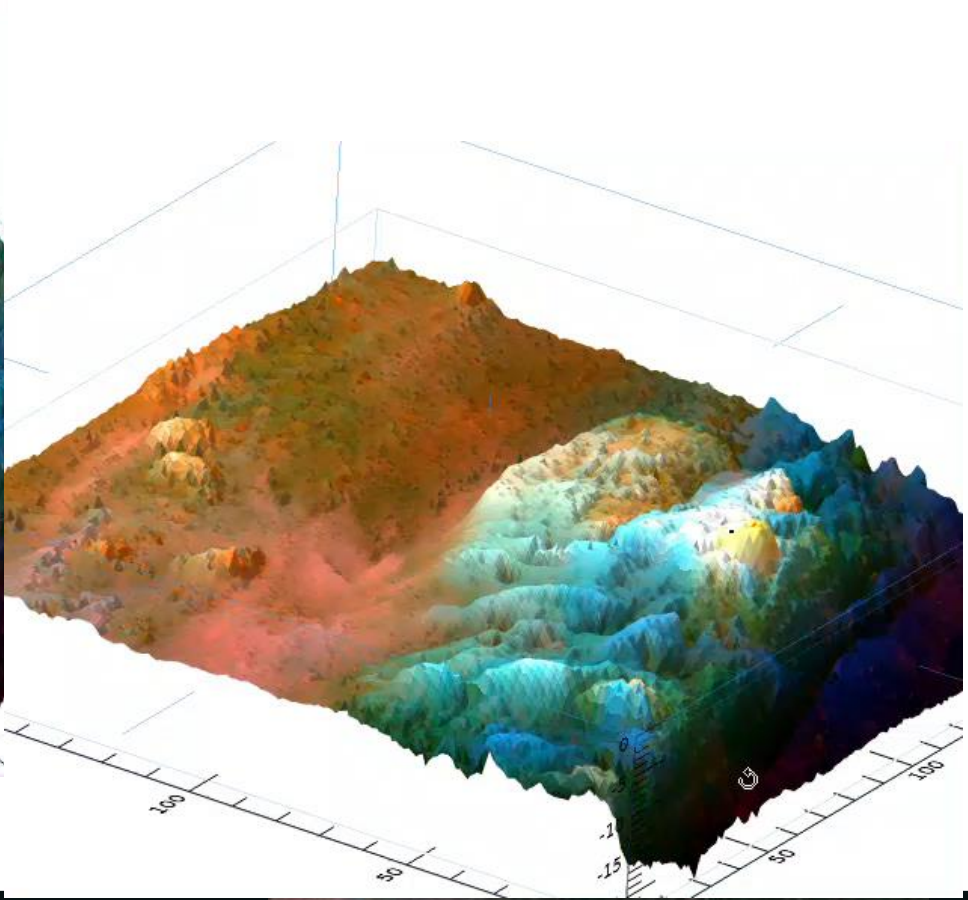
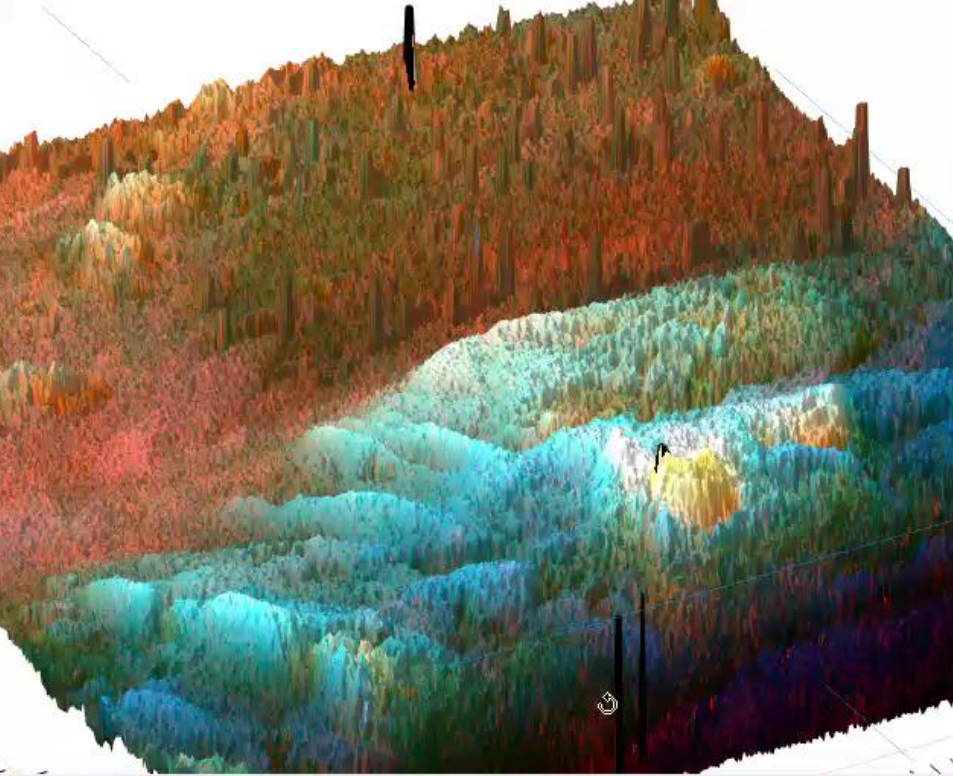


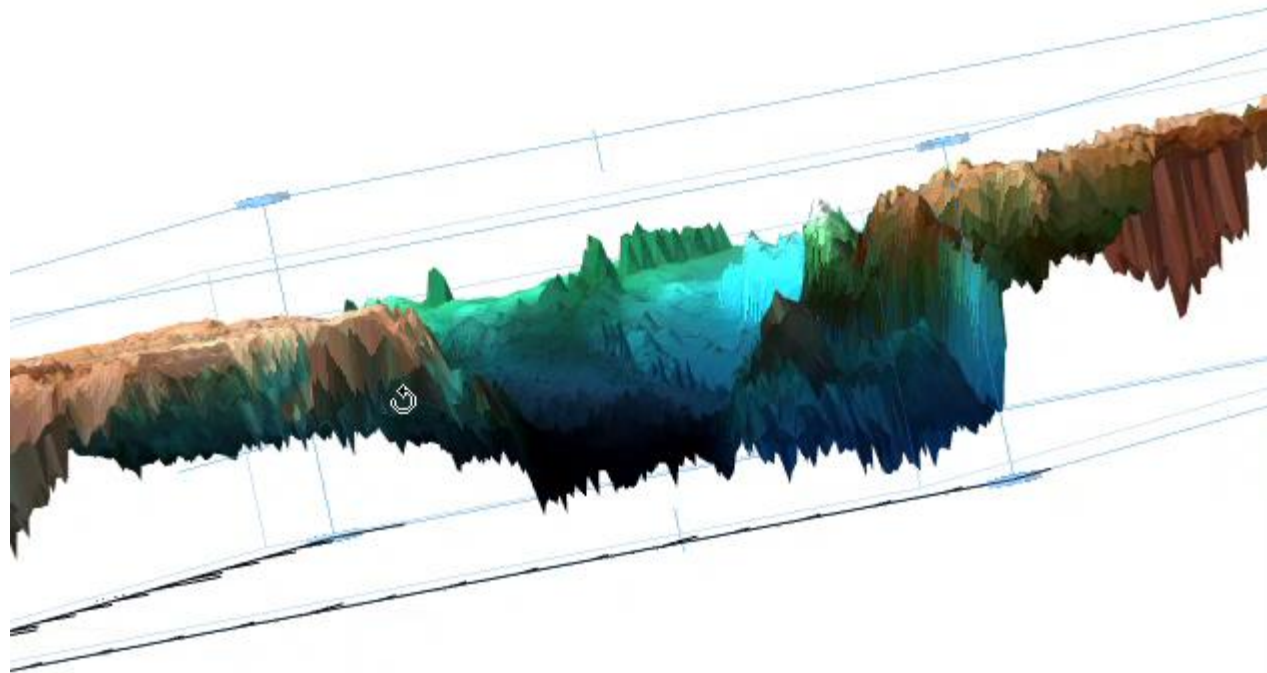
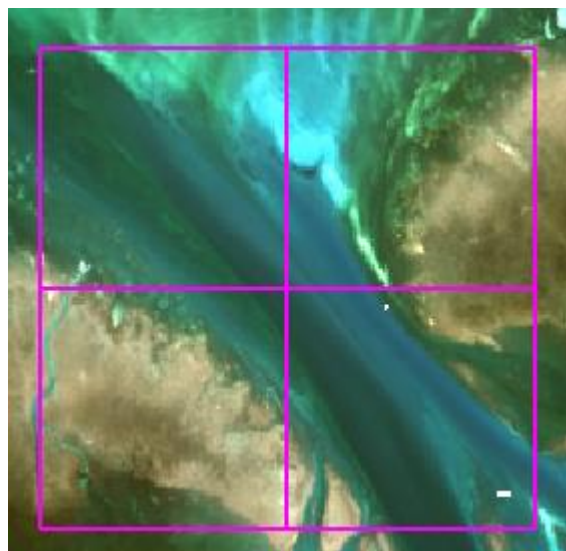


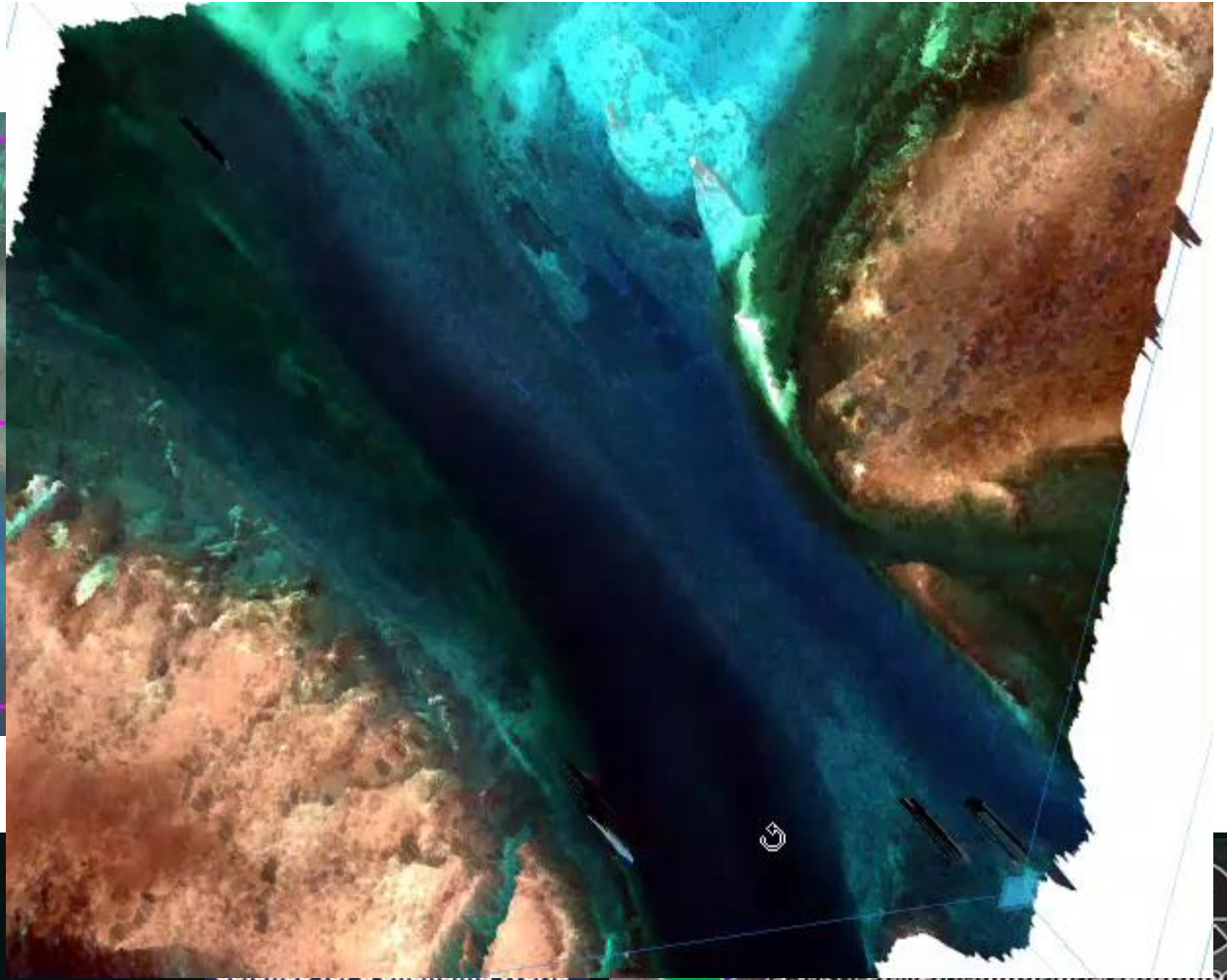
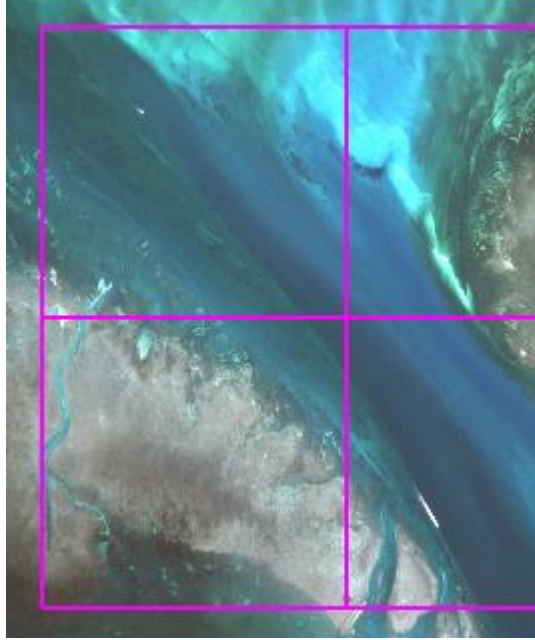
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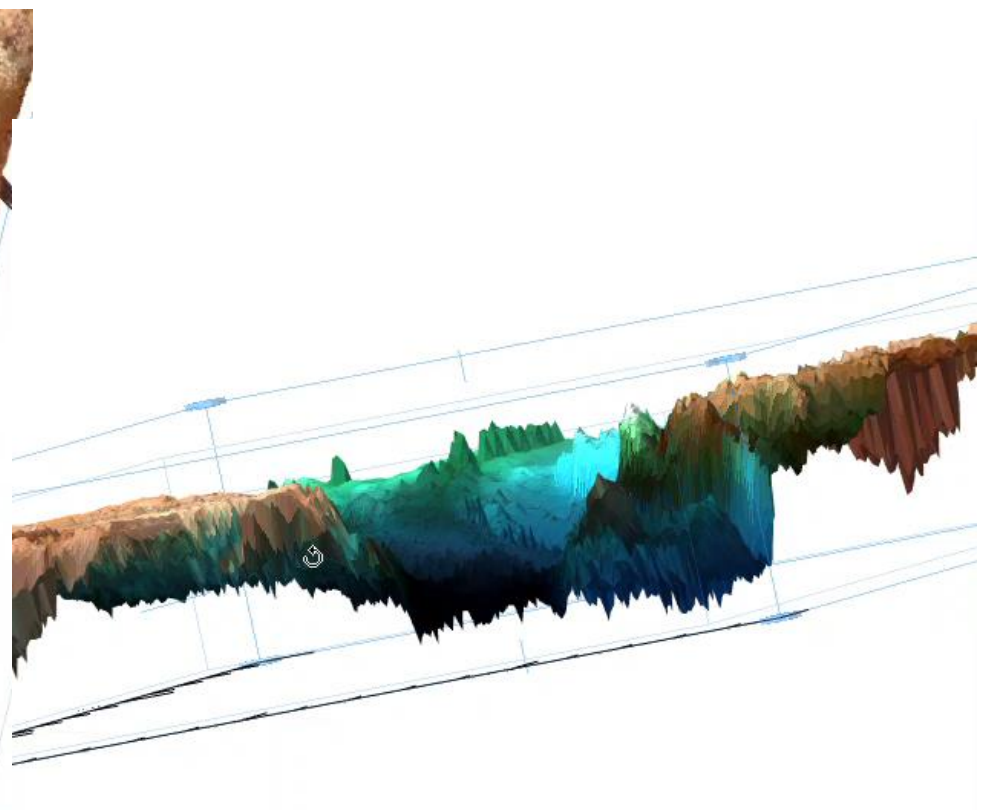
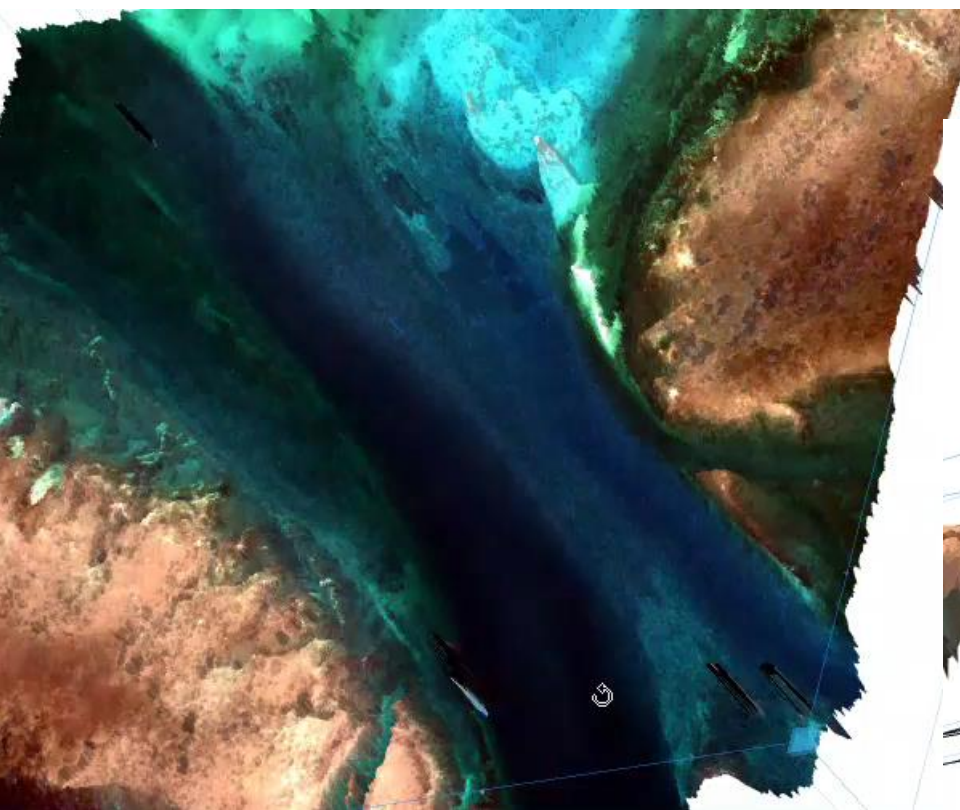


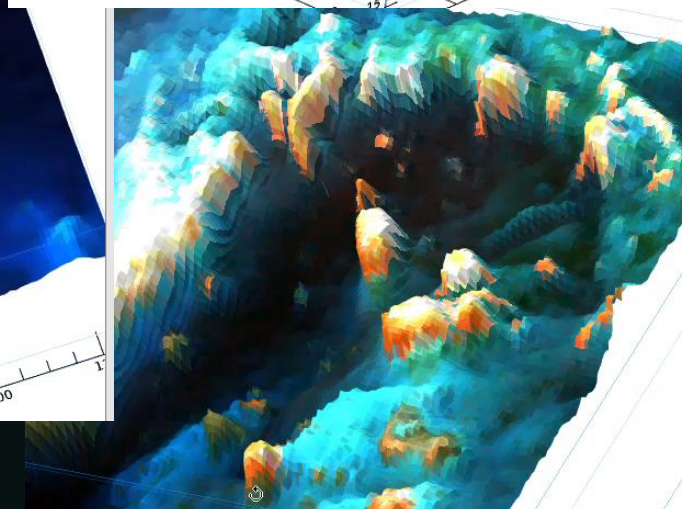
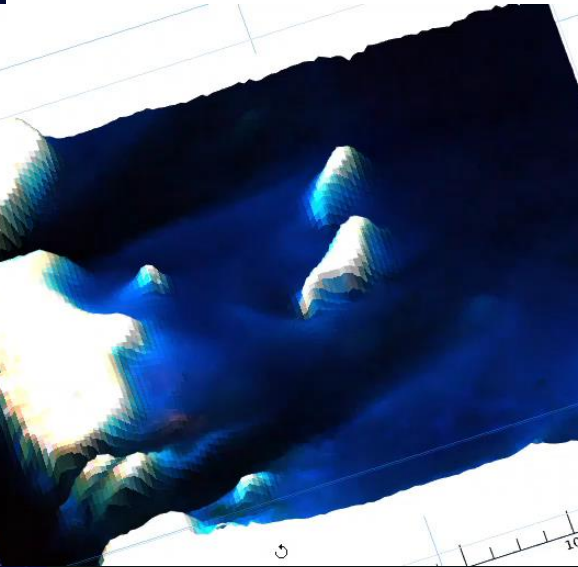
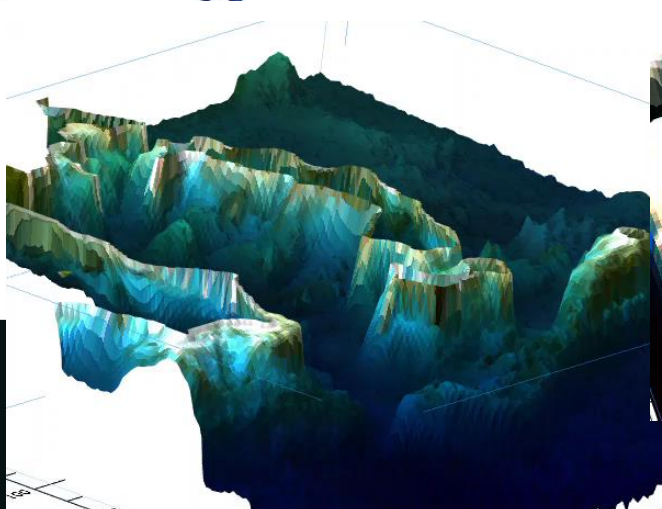
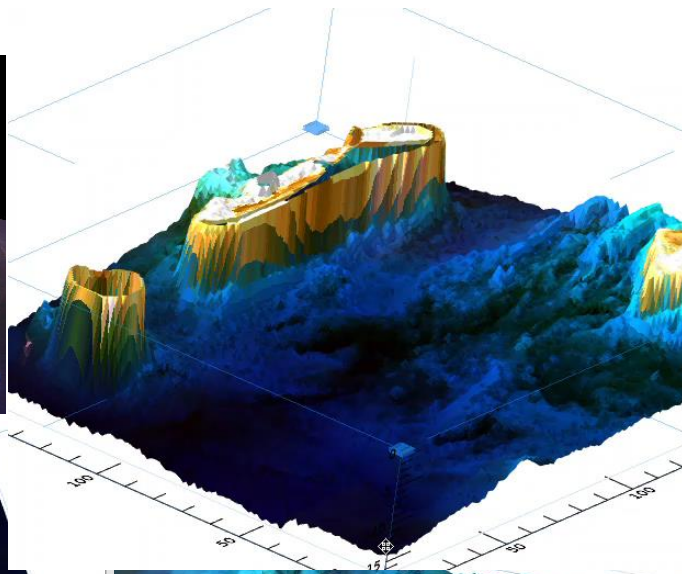
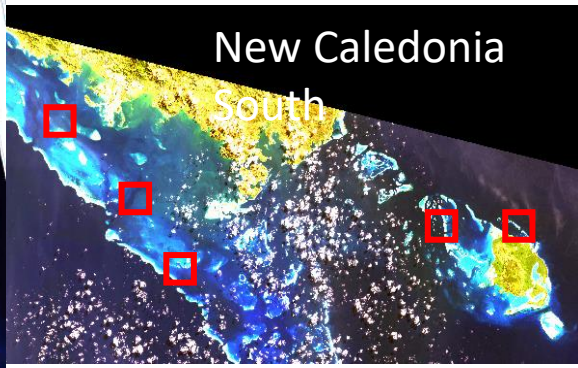
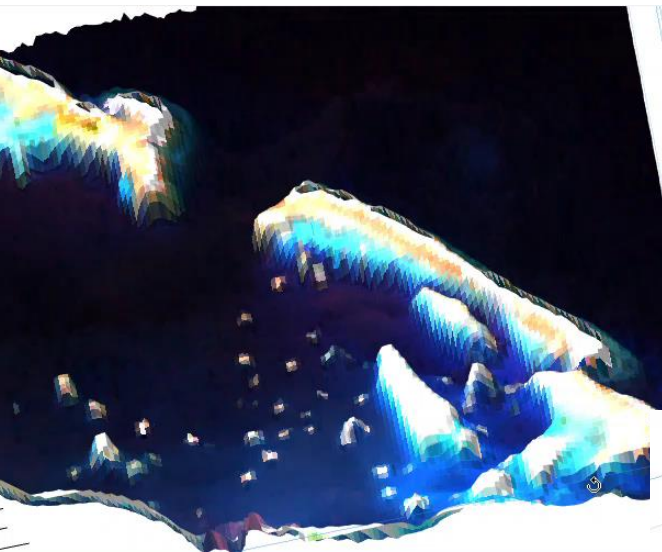
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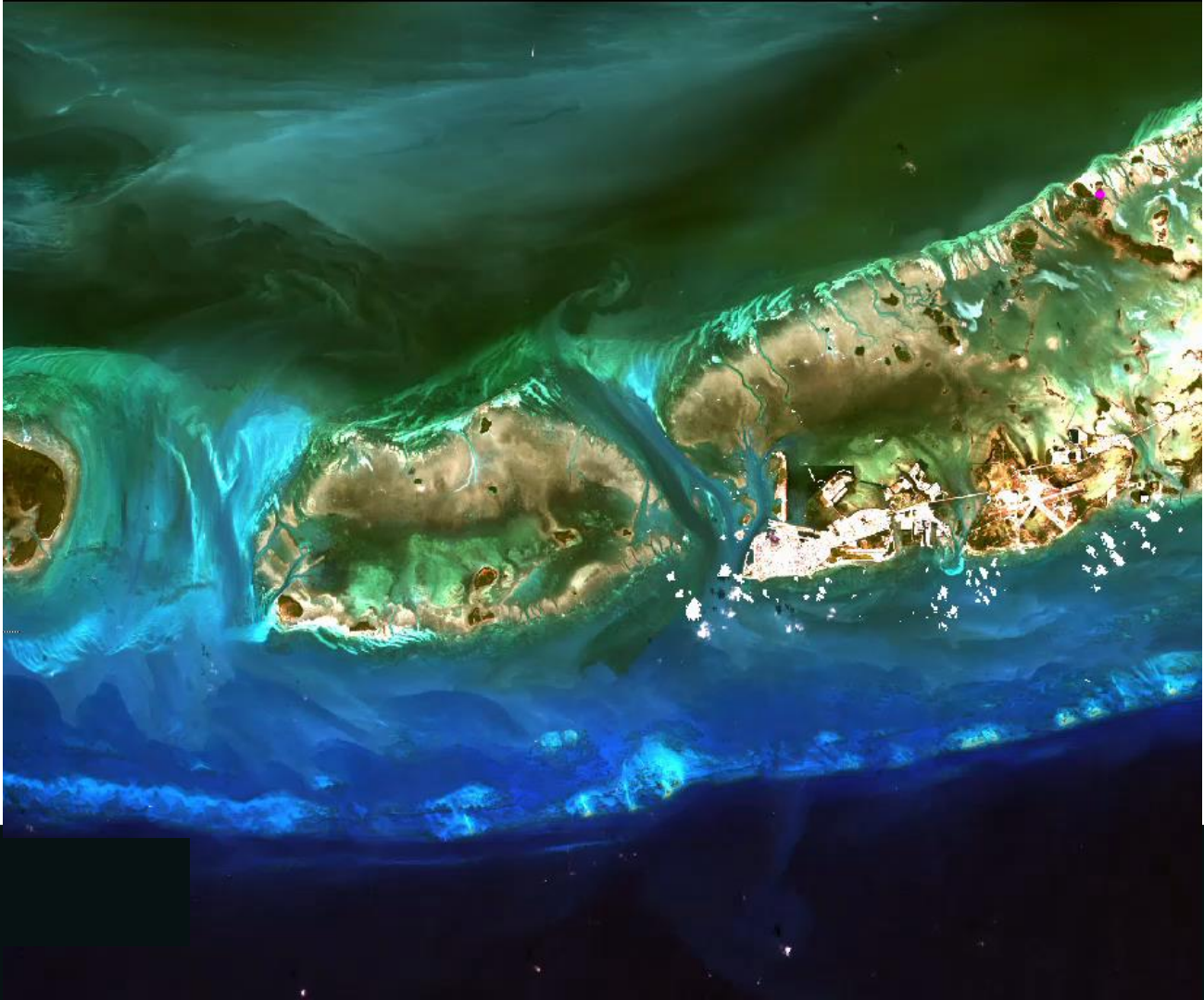


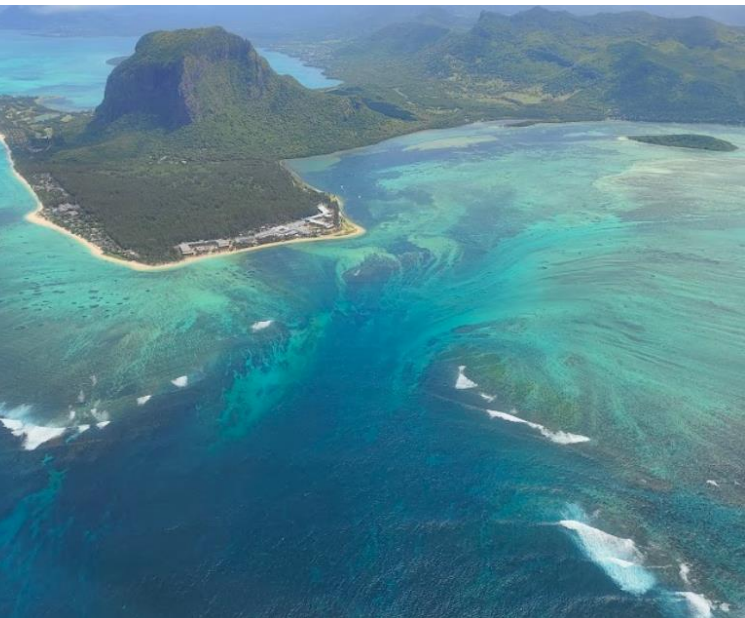






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Underwater Waterfall

