



# Towards real-time sea surface reconstruction from X-band radar and Lidar measurements

23/09/2025

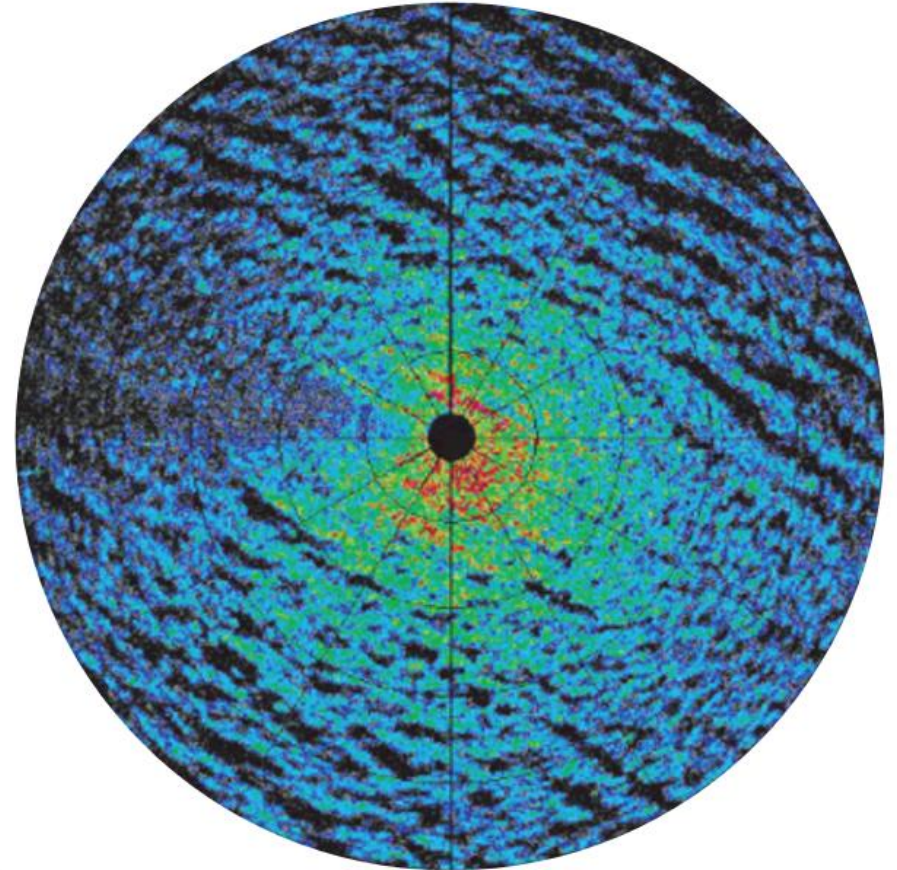
Rollon BARTHELEMY



## CONTEXT



Sea surface



Radar image

# APPLICATION OF REAL-TIME WAVE FORECASTING



## Decision support systems

in offshore operations:

- Ship-to-ship operations,
- (short) drilling operations,
- AUV/ROV launch and recovery,
- helicopter landing



## Control applications:

- heave compensation systems,
- dynamic positioning,
- predictive vessel stabilization,
- floating wind turbine control

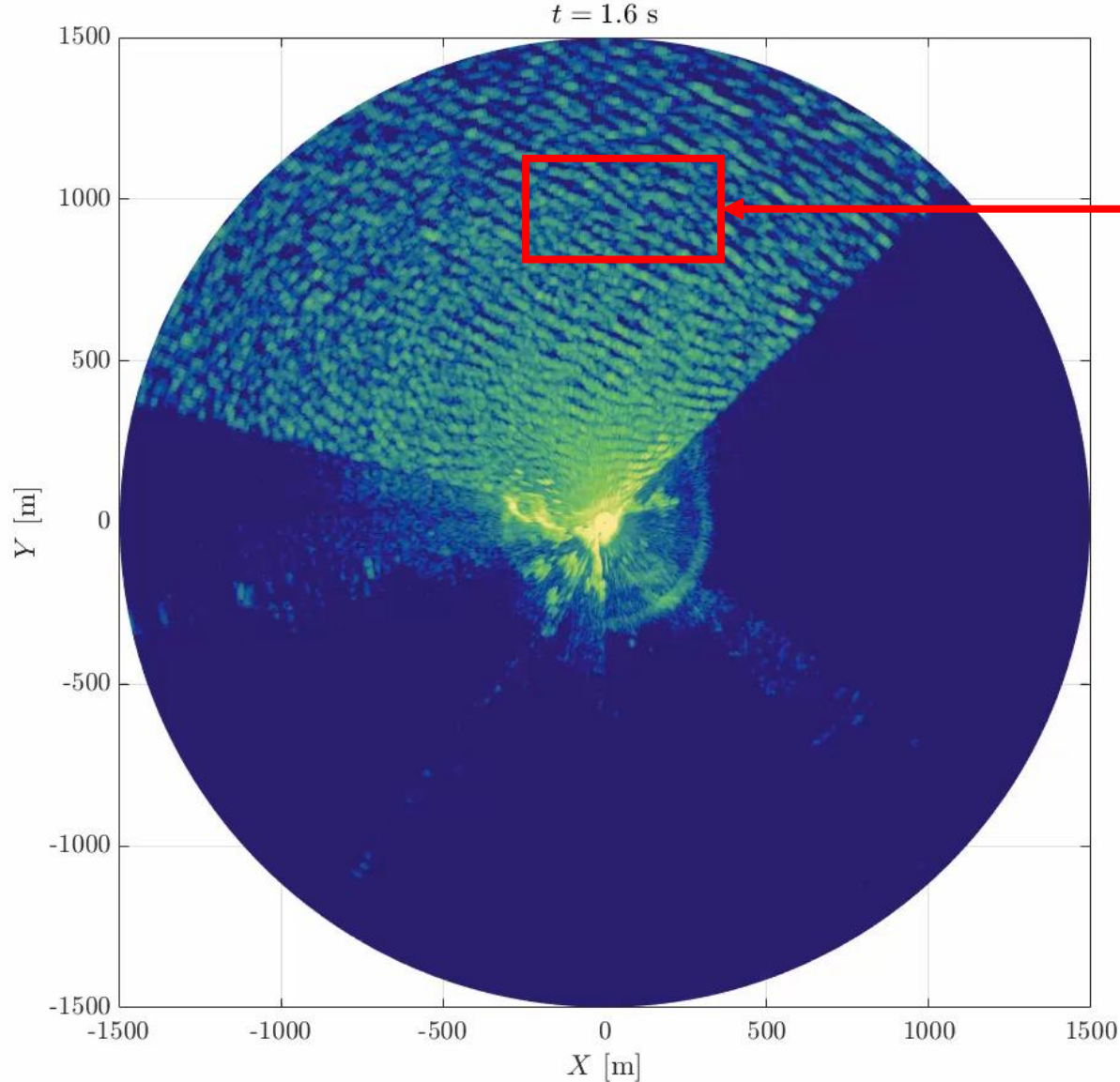


## Scientific research:

- Sea waves and current interaction,
- Wave dynamics,
- Coastal erosion,
- Validation/calibration models or satellites altimeters,
- Sea waves dynamic on high resolution and big distances



# SEA CLUTTER



Sea clutter : radar echoes returned from the rough sea surface.

The signal strength increases with the sea state and particularly to the sea waves height.

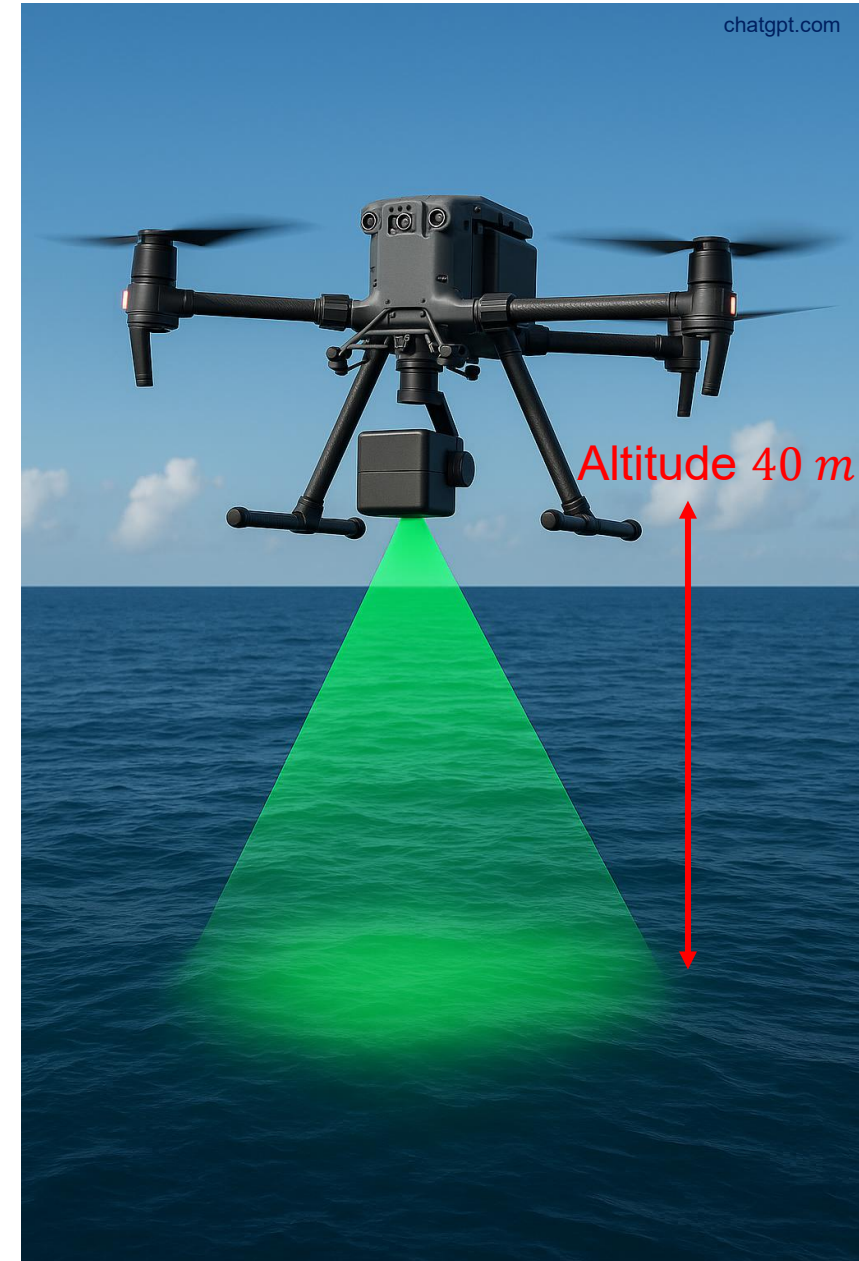
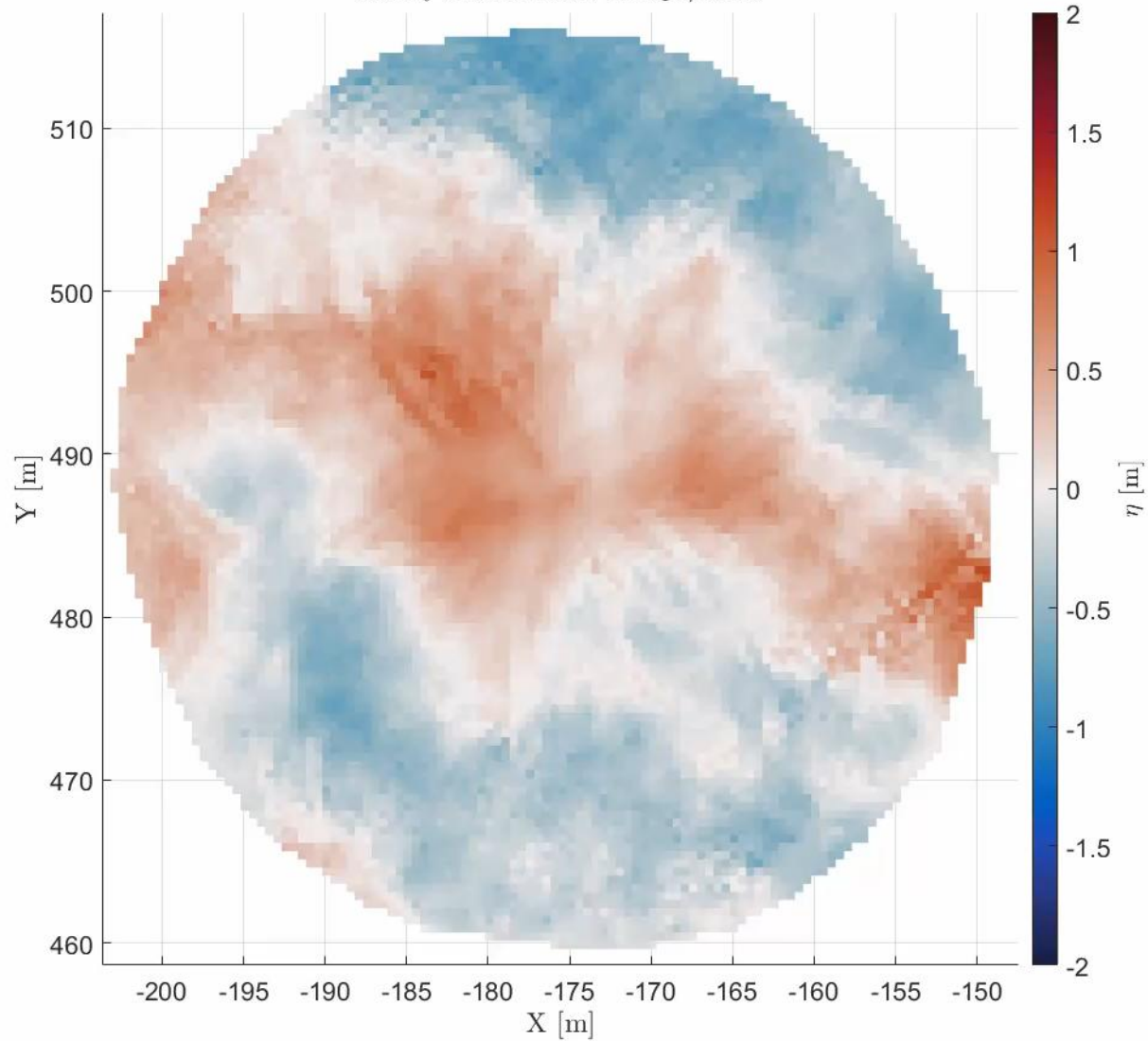


With a Lidar on a drone, stationary flight over a point to recover sea elevation over a surface area.

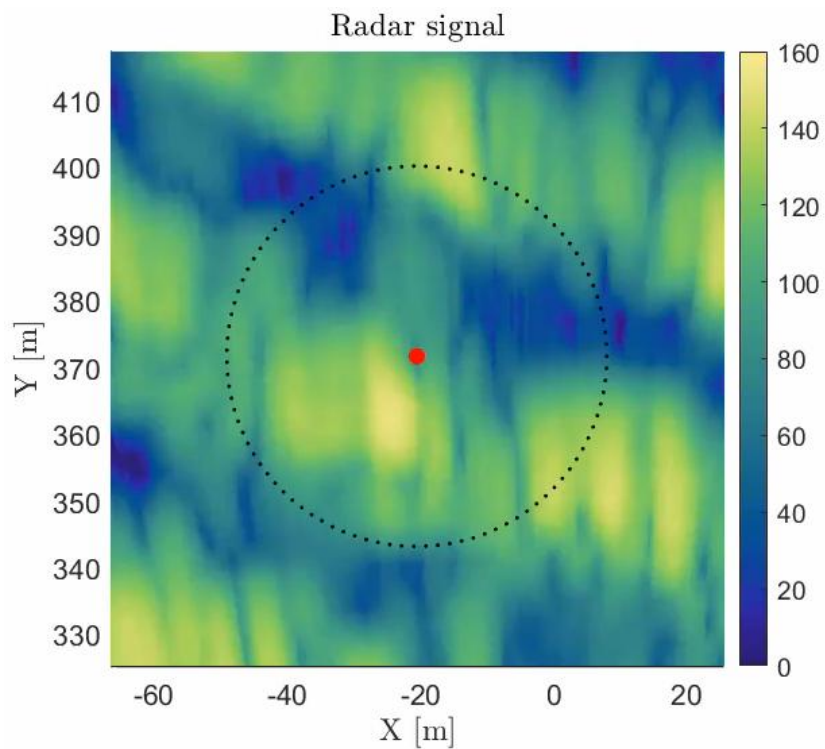
At the same time, we save the radar signal to extract the same zone later during processing.

# LIDAR DATA

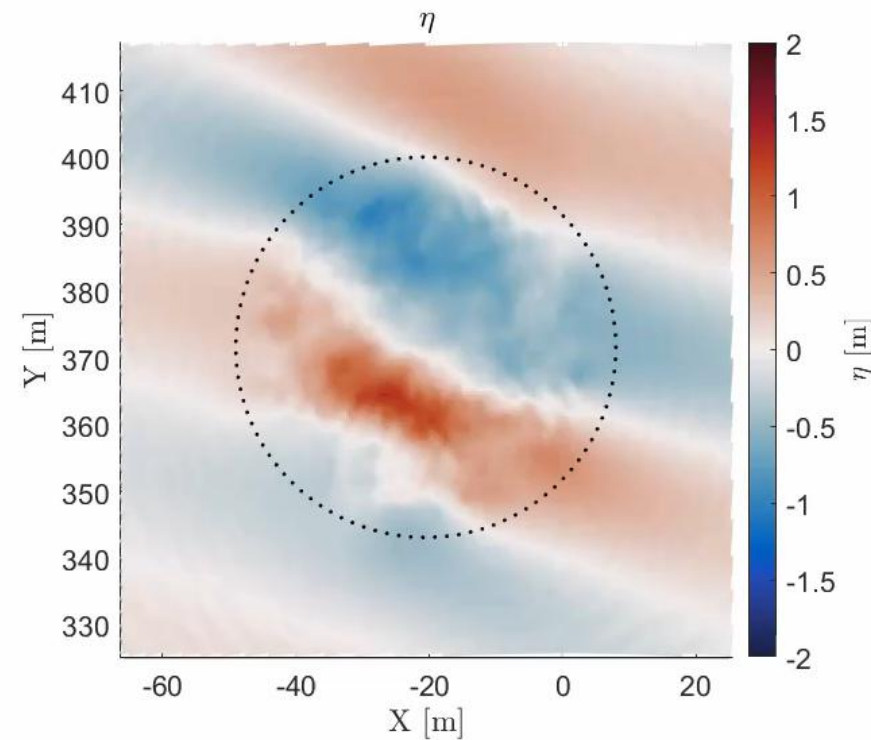
06-May-2025 17:33:28 Europe/Paris



# OUTLINE



Improving radar model

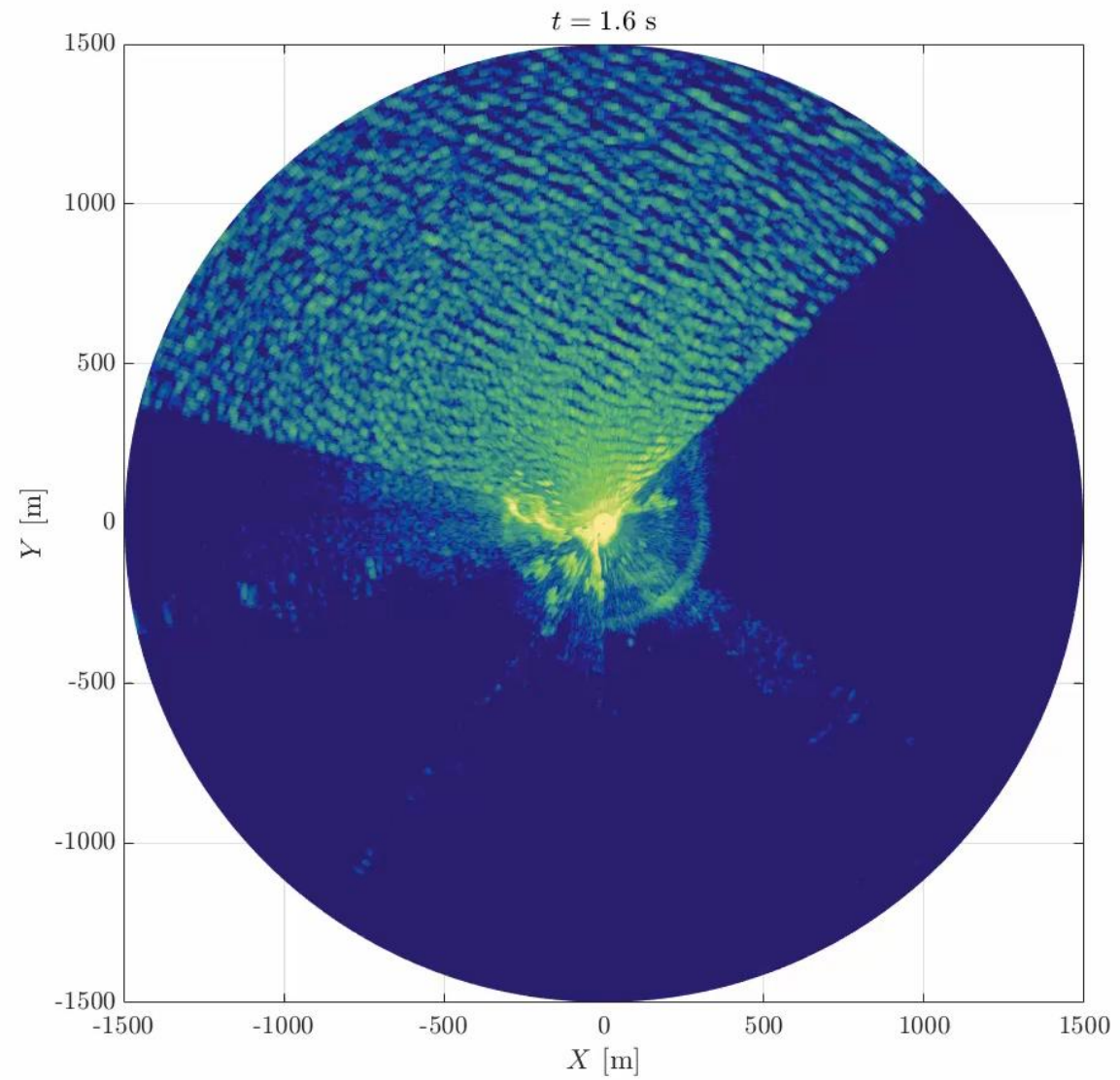
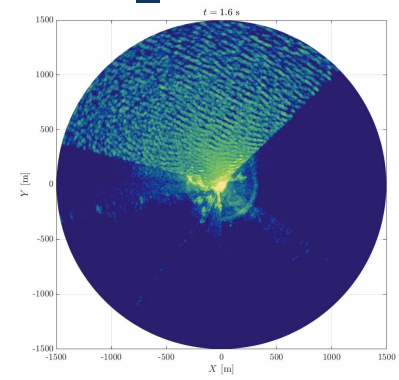


AI inversion

# IMPROVING RADAR MODEL

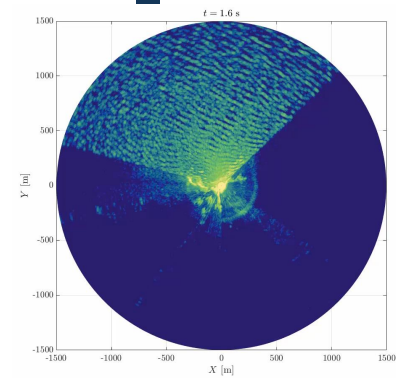


# PATHWAY

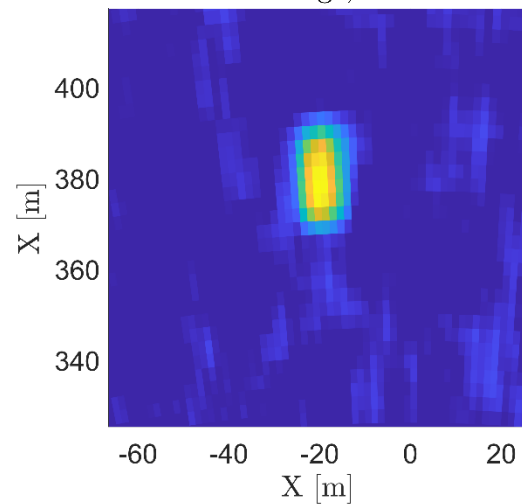




# PATHWAY

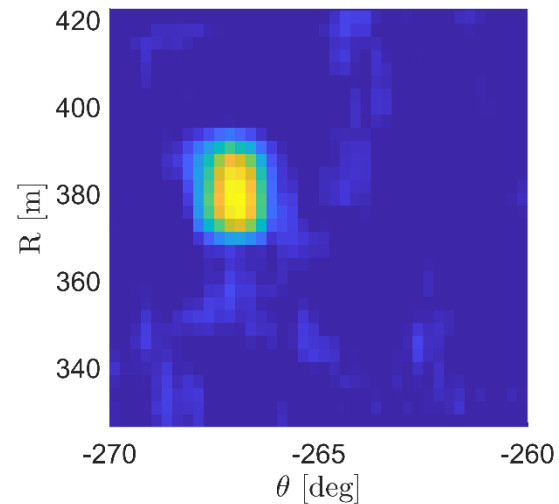


Mean radar image, drone return

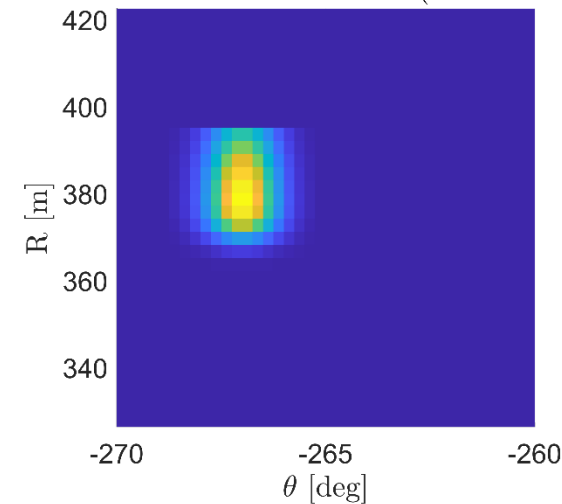


Mean radar image, drone return

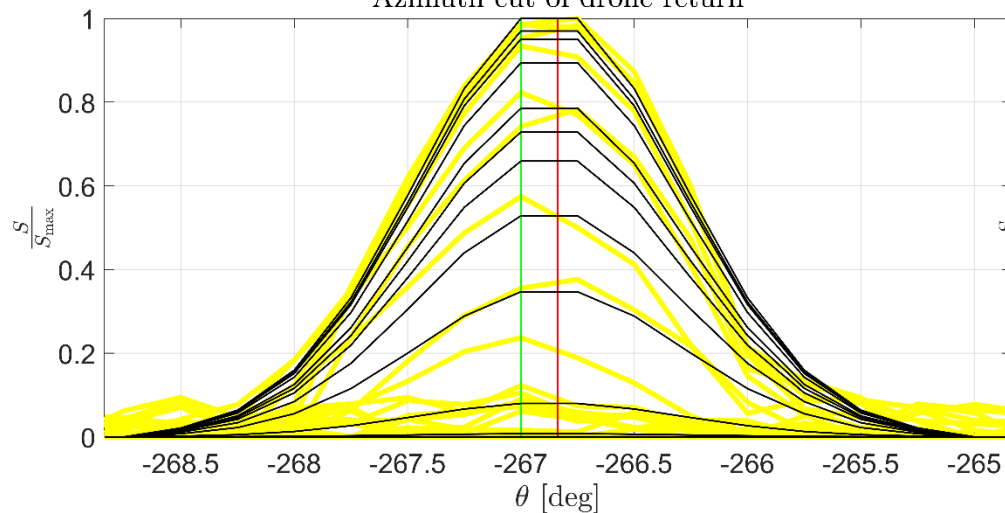
Polar coordinates



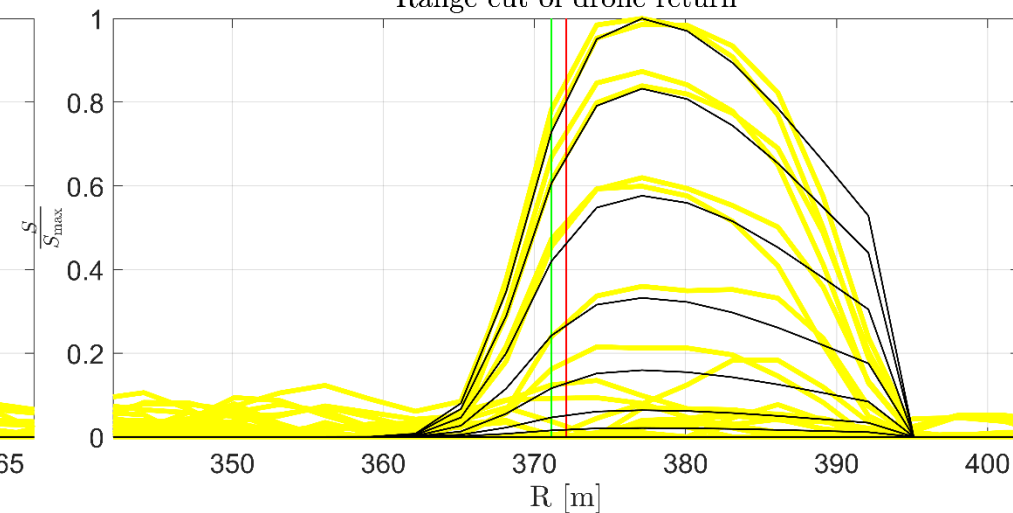
Modelisation of radar beam (drone return)



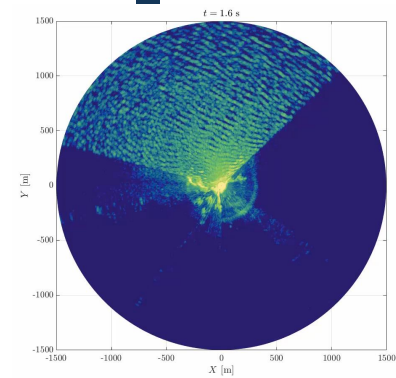
Azimuth cut of drone return



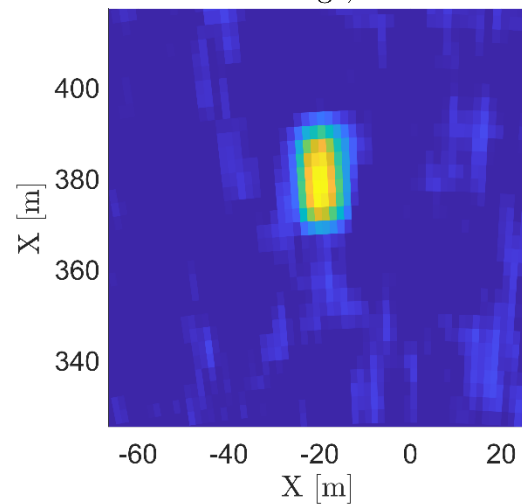
Range cut of drone return



# PATHWAY

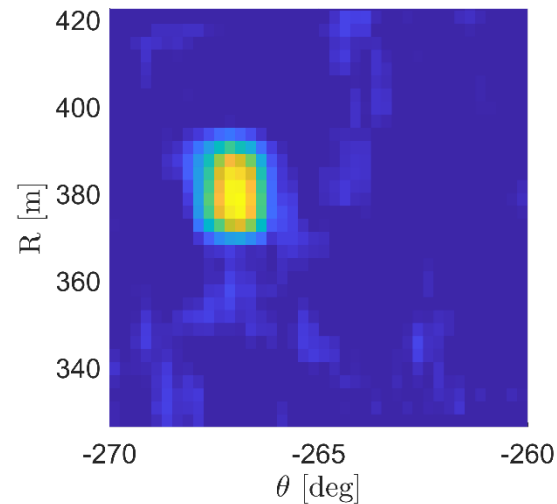


Mean radar image, drone return

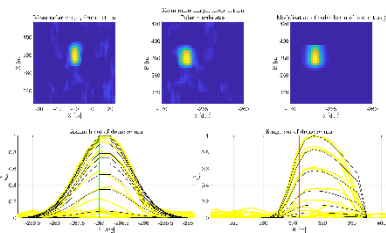
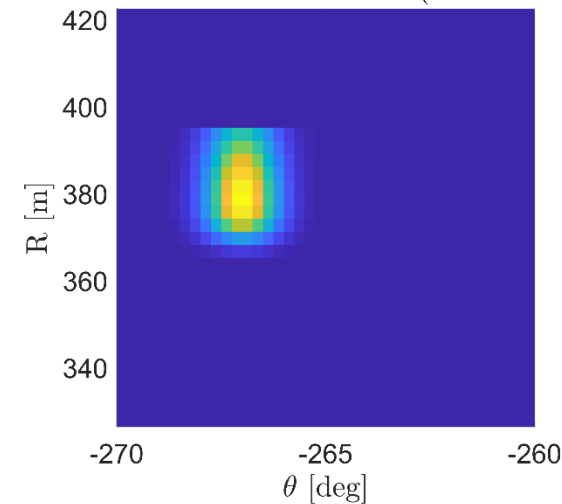


Mean radar image, drone return

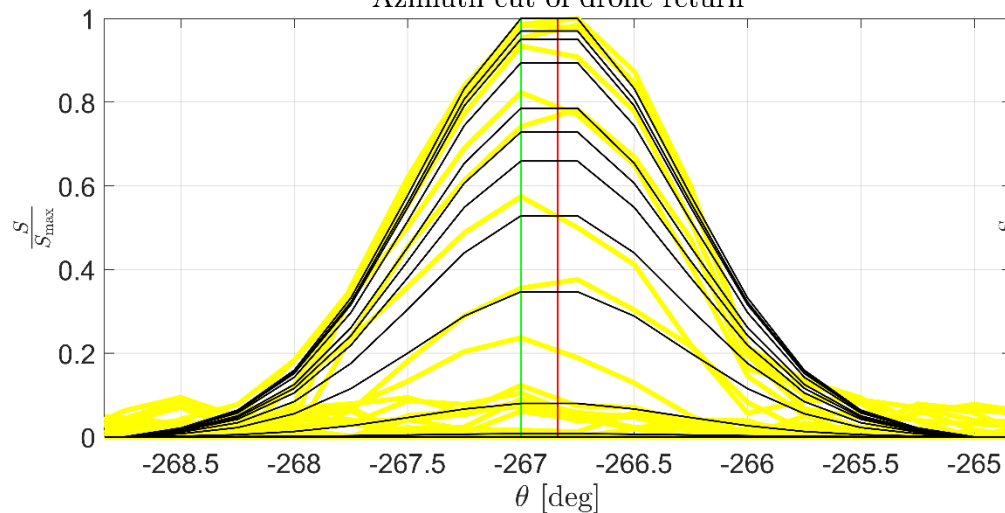
Polar coordinates



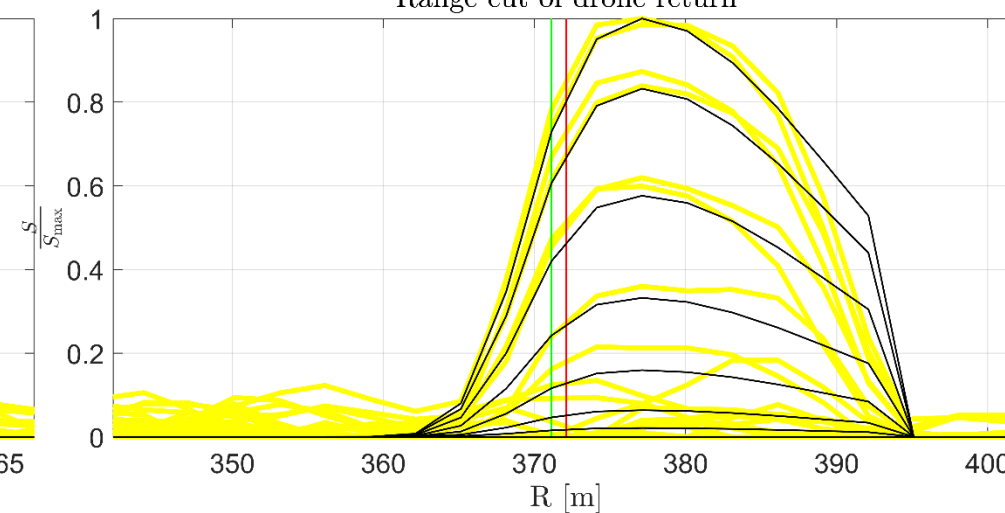
Modelisation of radar beam (drone return)



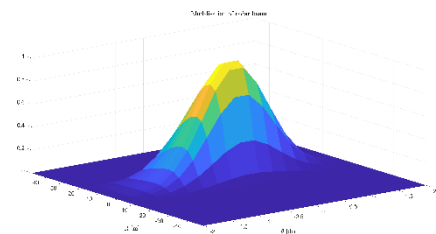
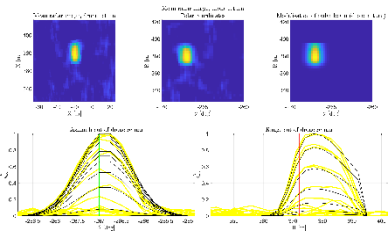
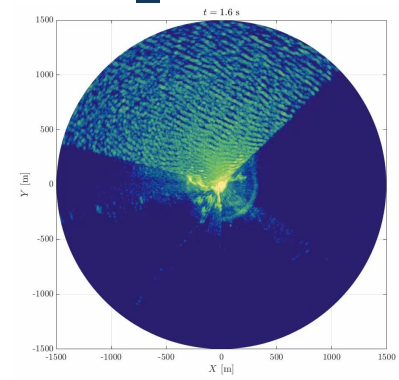
Azimuth cut of drone return



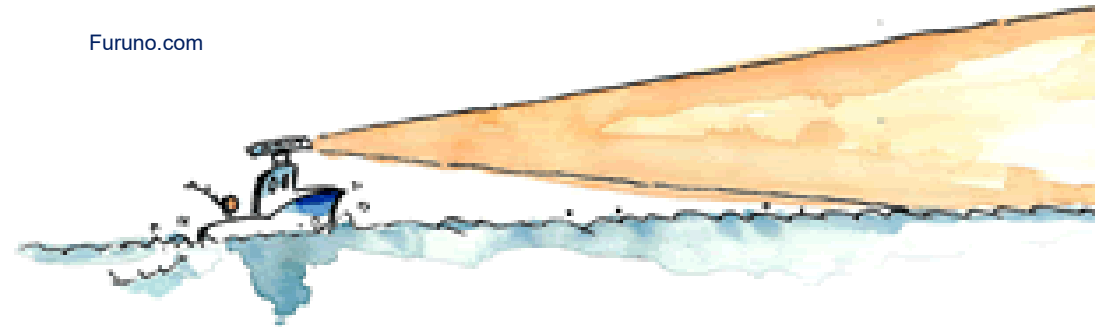
Range cut of drone return



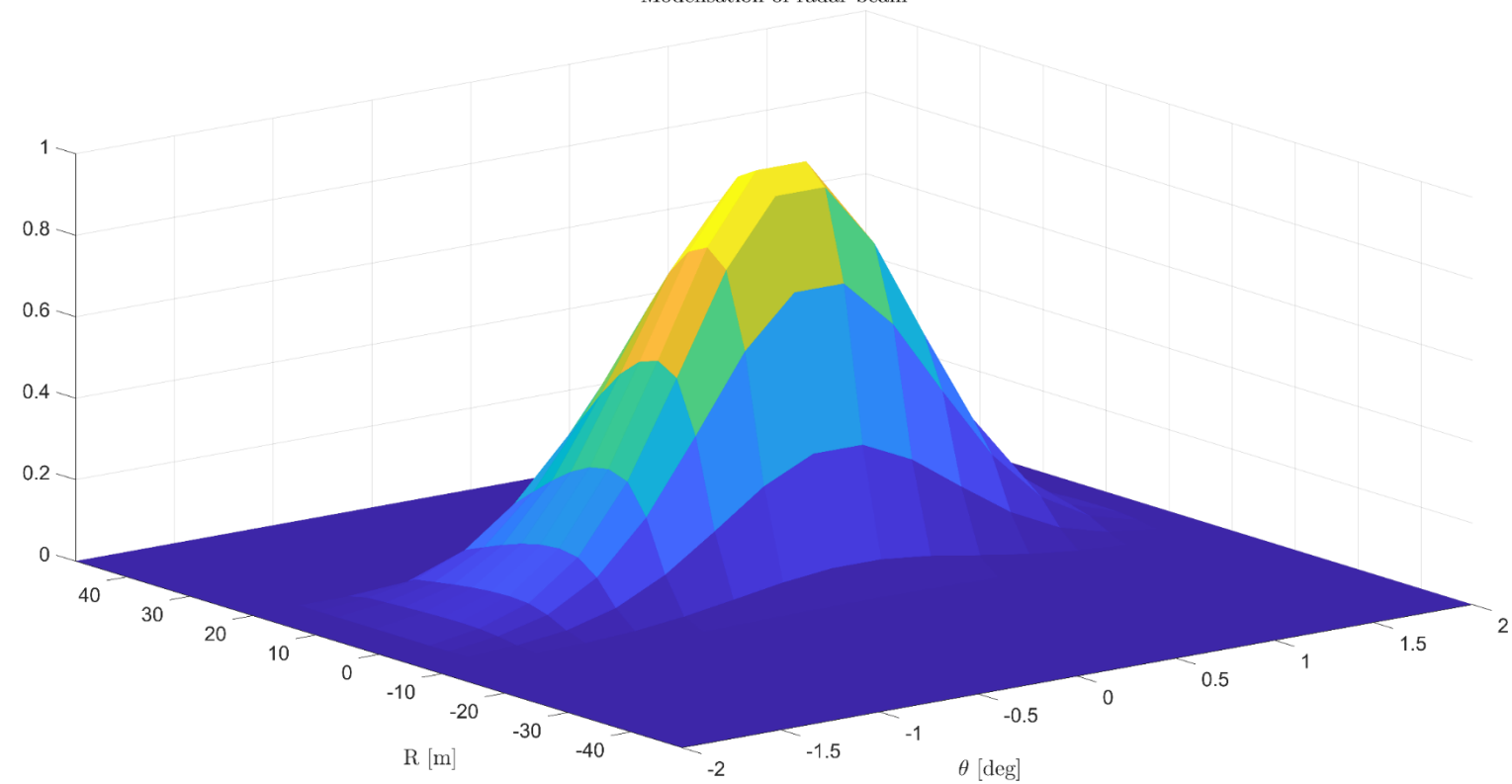
# PATHWAY



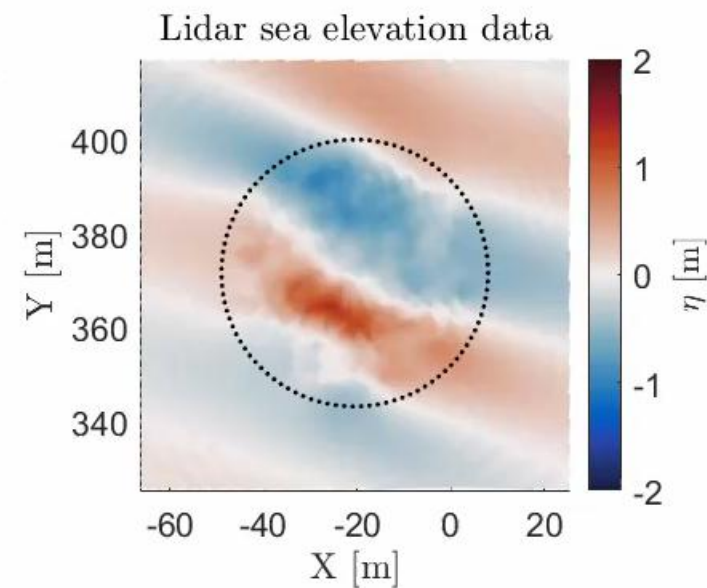
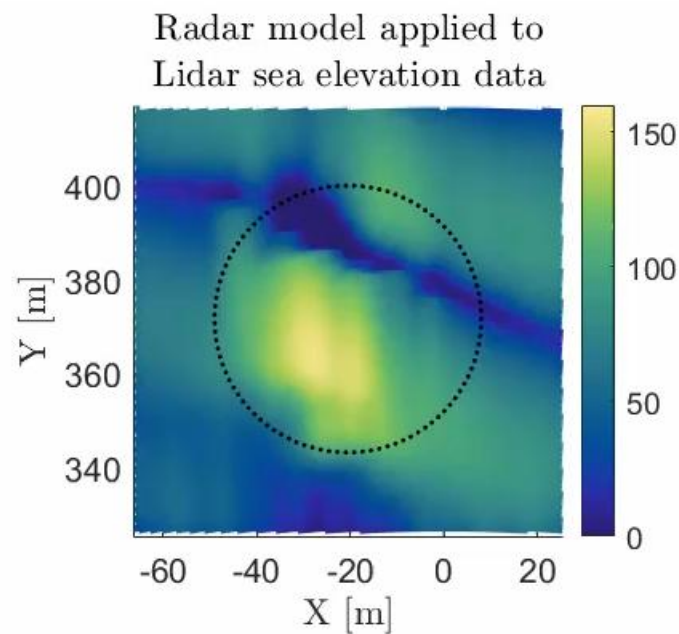
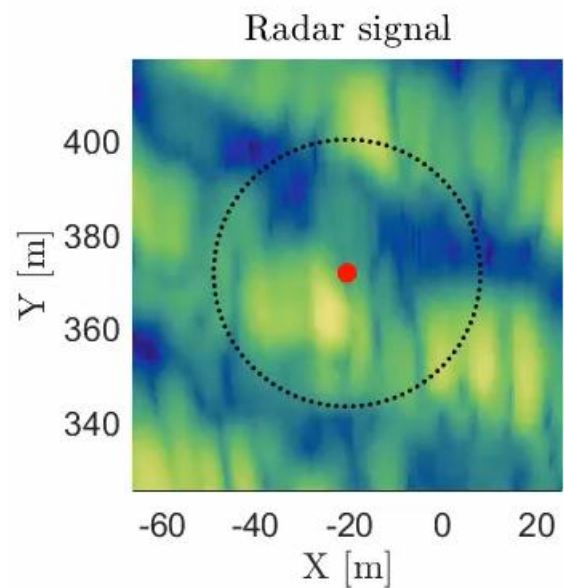
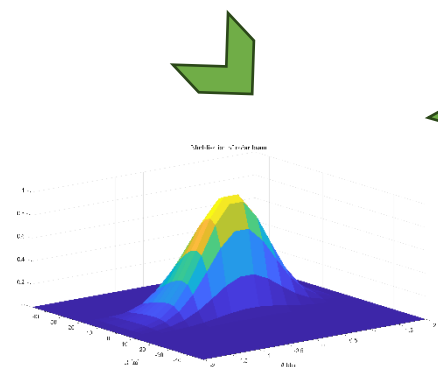
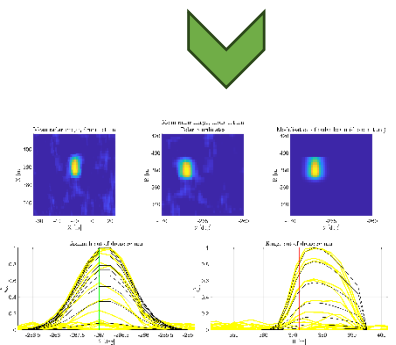
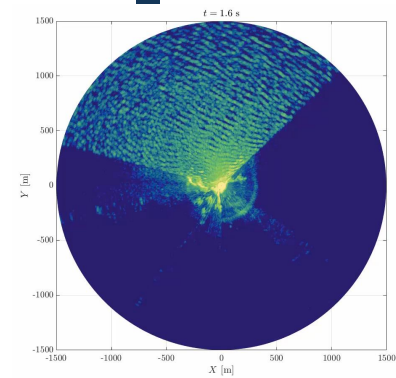
Furuno.com



Modelisation of radar beam



# PATHWAY



Sea elevation and radar signal synchronisation



# AI INVERSION

## Lidar measurement location

3 days first time campaign :

- play with a newly purchased radar
- get some lidar/radar data

3 hours of synchronised radar/lidar data

$$H_s \in [.8, 1.5] \text{ m}$$

$$T_p \in [5, 7] \text{ s}$$

Main waves direction



109 images

File 087



109 images

File 090



102 images

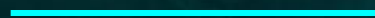
File 092



108 images

File 094

319 images



0°16'30"E

0°17'E

0°17'30"E

0°18'E

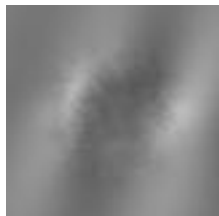
0°18'30"E

Longitude

Maxar, Microsoft

# DEEP LEARNING

Sea elevation, Lidar data

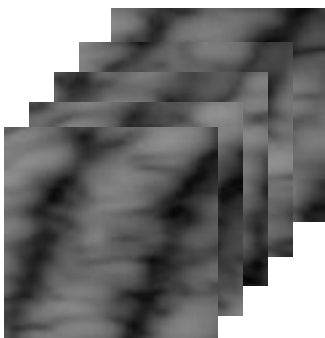


3 different acquisitions

$$H_s \approx 1.5 \text{ m}$$

$$T_p \approx 6 \text{ s}$$

Sea clutter, radar data



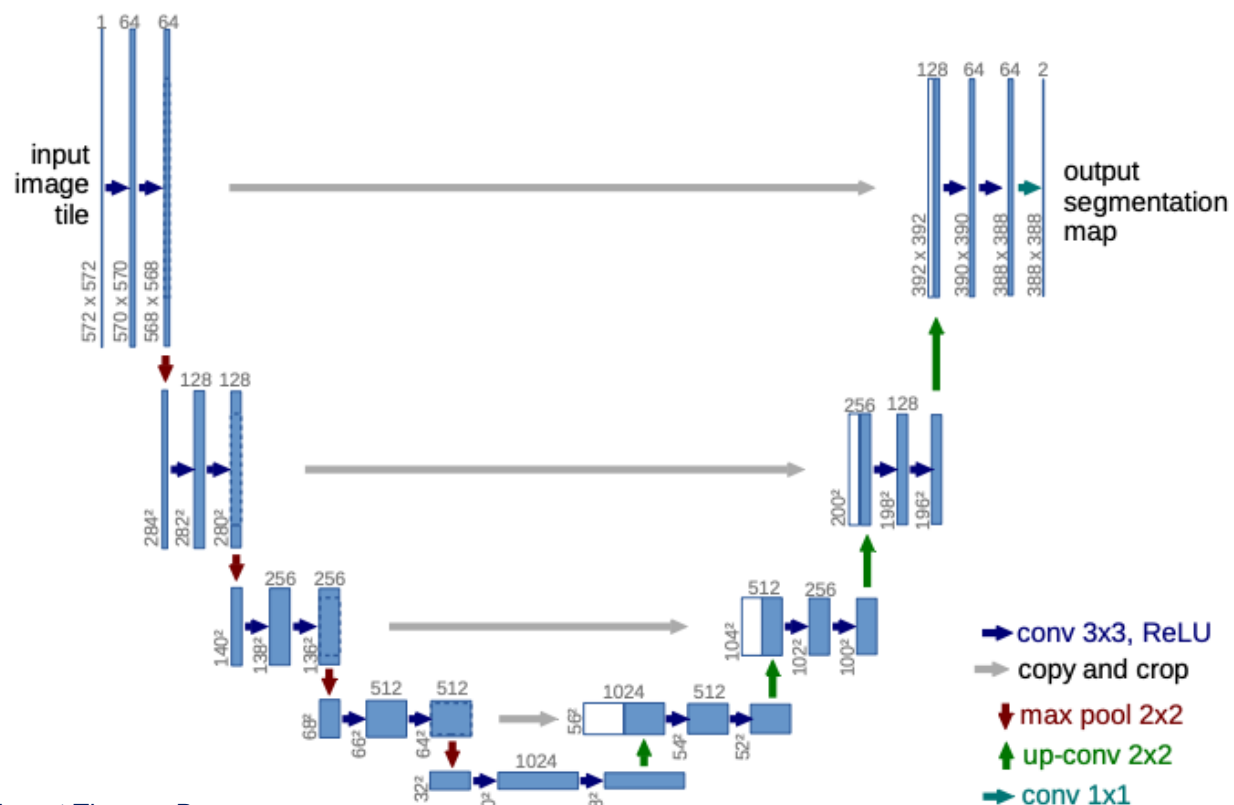
$N$  consecutive images

$$(80\text{px} \times 80\text{px}) = 80\text{m} \times 80\text{m}$$

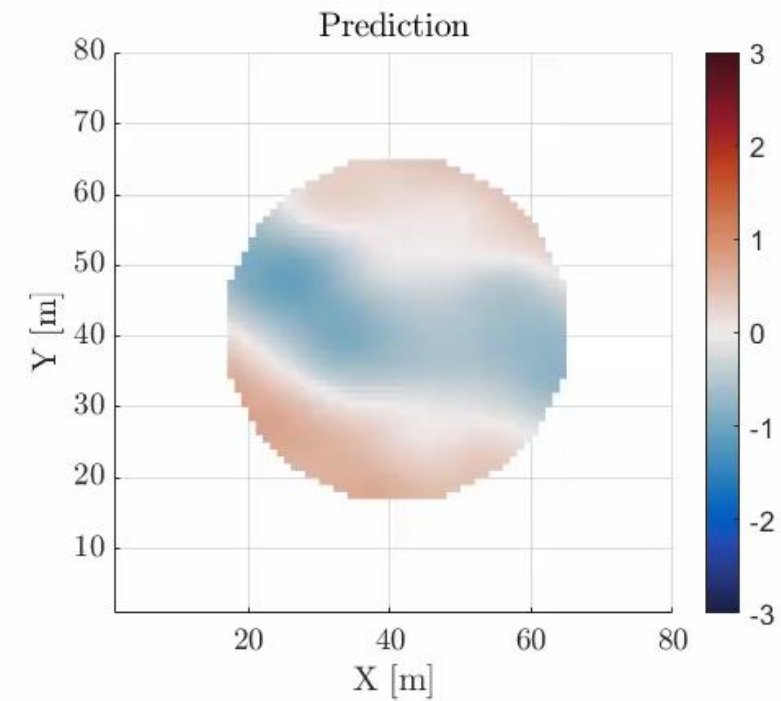
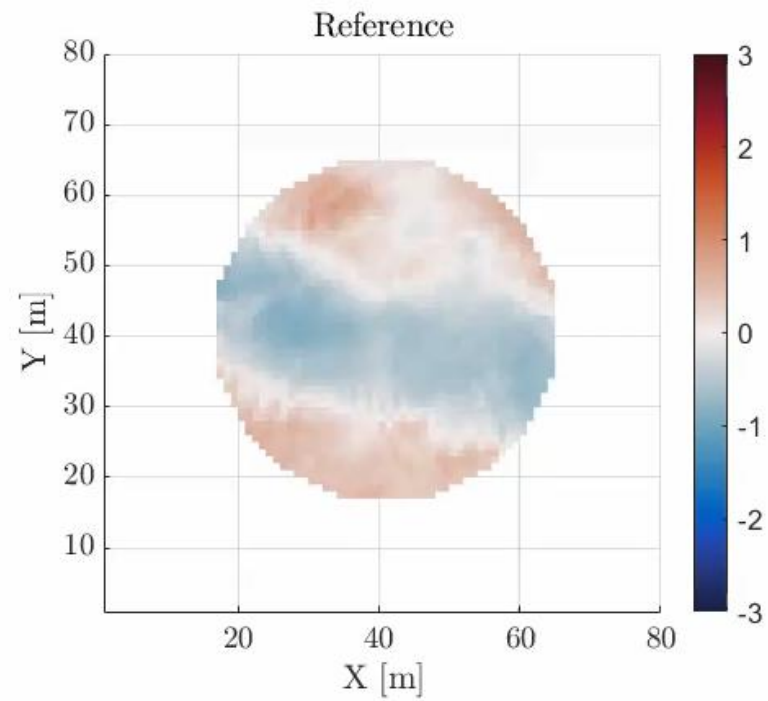
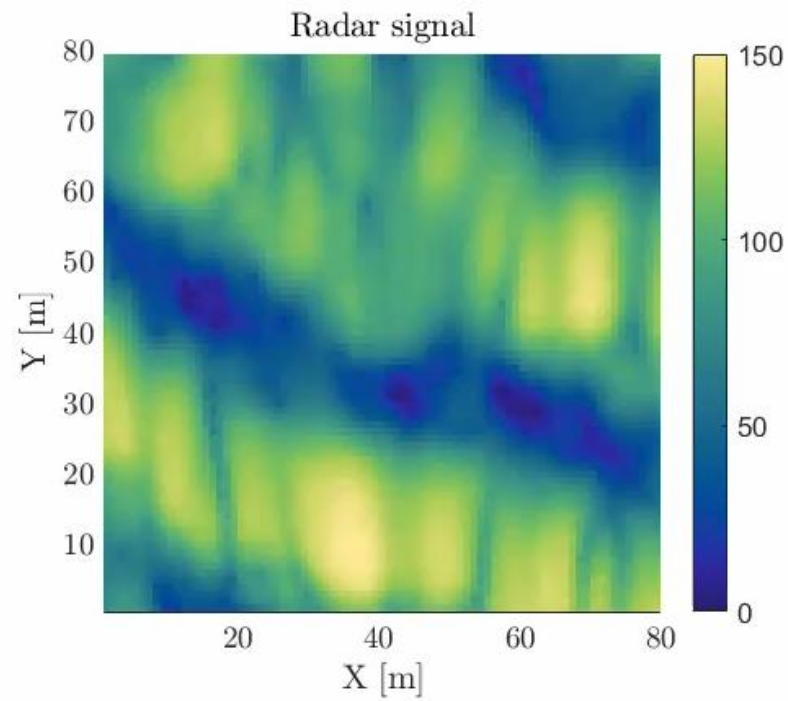
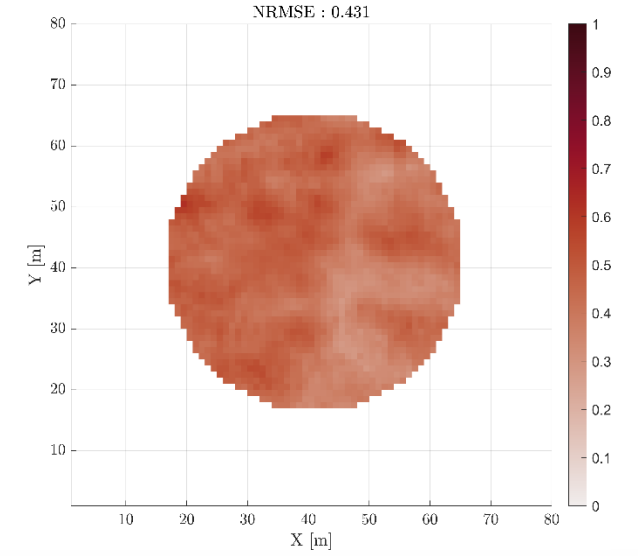
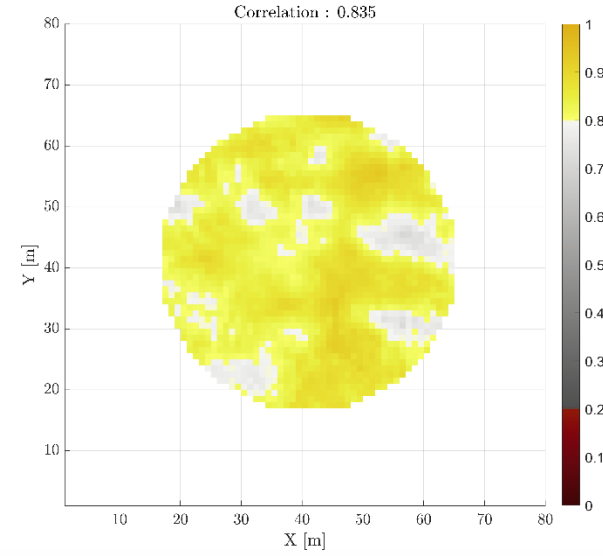
- Training on 50 or 280 pairs of images randomly chosen from  $319 - N + 1$  images
- Loss function : Structural SIMilarity (SSIM)

(Image quality assessment: from error visibility to structural similarity, Wang and al., 2004)

## U-Net Convolutional Network

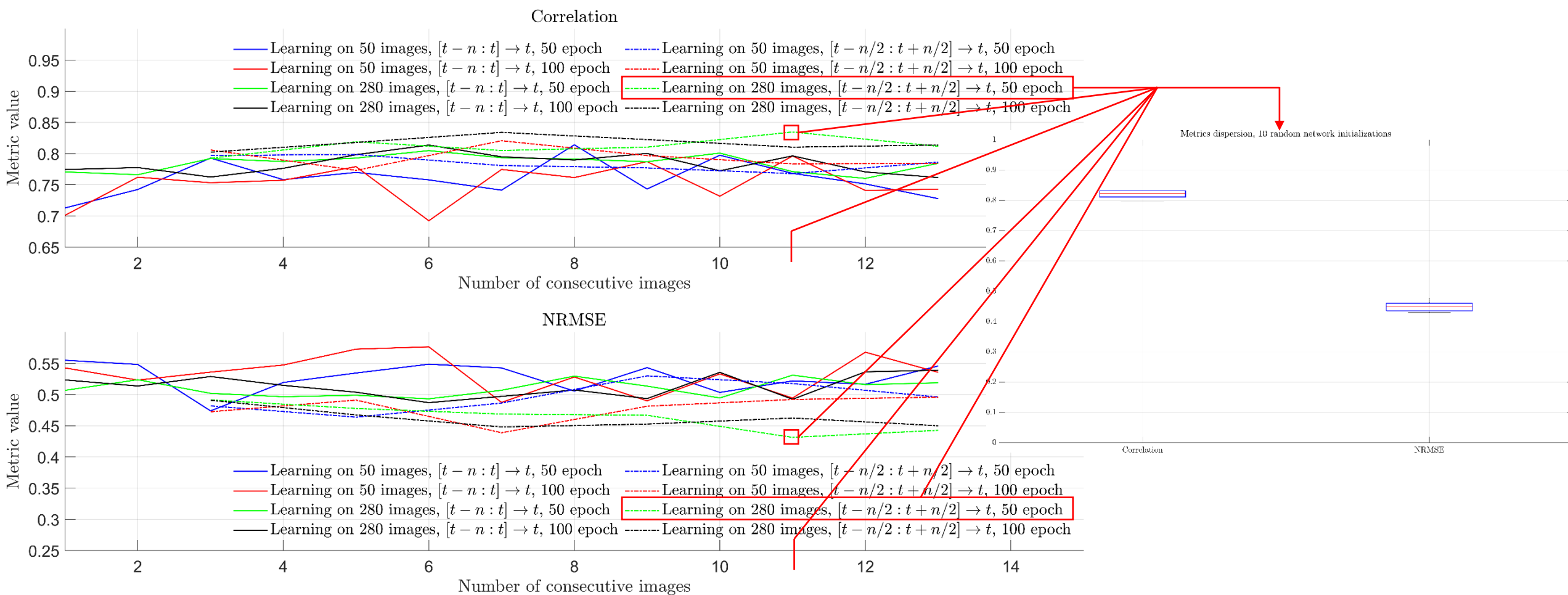


# RESULTS





# REPEATABILITY OF THE INVERSION



THANK YOU !!!

