

# Coastal altimetry at high-latitudes: the Baltic SEAL project observing sea level among jagged coastline and sea ice

Marcello Passaro<sup>1</sup>, Felix L. Müller<sup>1</sup>, Adili Abulaitijiang<sup>2</sup>, Ole B. Andersen<sup>2</sup>, Denise Dettmering<sup>1</sup>, Jacob L. Høyer<sup>3</sup>, Milla Johansson<sup>4</sup>, Kristine Skovgaard Madsen<sup>3</sup>, Laura Rautiainen<sup>4</sup>, Ida M. Ringgaard<sup>3</sup>, Eero Rinne<sup>4</sup>, Jani Särkkä<sup>4</sup>, Rory Scarrott<sup>5</sup>, Christian Schwatke<sup>1</sup>, Florian Seitz<sup>1</sup>, Eimear Tuohy<sup>5</sup>, Laura Tuomi<sup>4</sup>, Americo Ambrozio<sup>6\*</sup>, Marco Restano<sup>6\*\*</sup>, Jérôme Benveniste<sup>6</sup>

<sup>1</sup> Deutsches Geodätisches Forschungsinstitut, Technische Universität München (DGFI-TUM), Germany

<sup>2</sup> SPACE National Space Institute, Technical University of Denmark, (DTU), Denmark

<sup>3</sup> Danish Meteorological Institute (DMI), Denmark

<sup>4</sup> Finnish Meteorological Institute (FMI), Finland

<sup>5</sup> MaREI Centre, Environmental Research Institute, University College Cork (UCC), Ireland

<sup>6\*</sup> DEIMOS, c/o ESA-ESRIN, Italy

<sup>6\*\*</sup> SERCO, c/o ESA-ESRIN, Italy

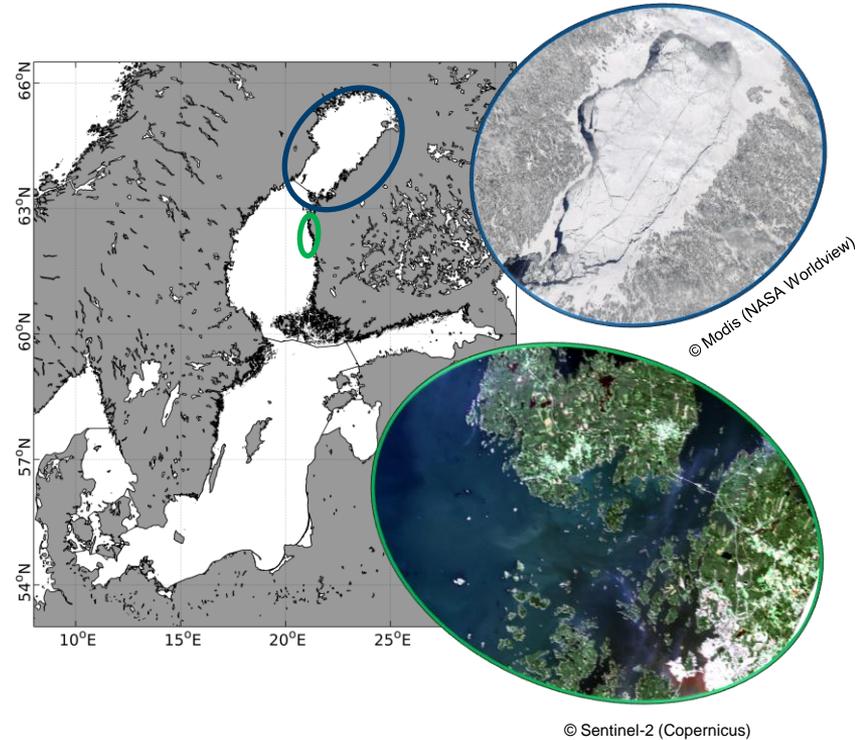
<sup>6</sup> ESA-ESRIN, Italy



# The Baltic Sea – Motivation

## What?

- Generation of a novel multi-mission sea level (MMSL) along-track and gridded product



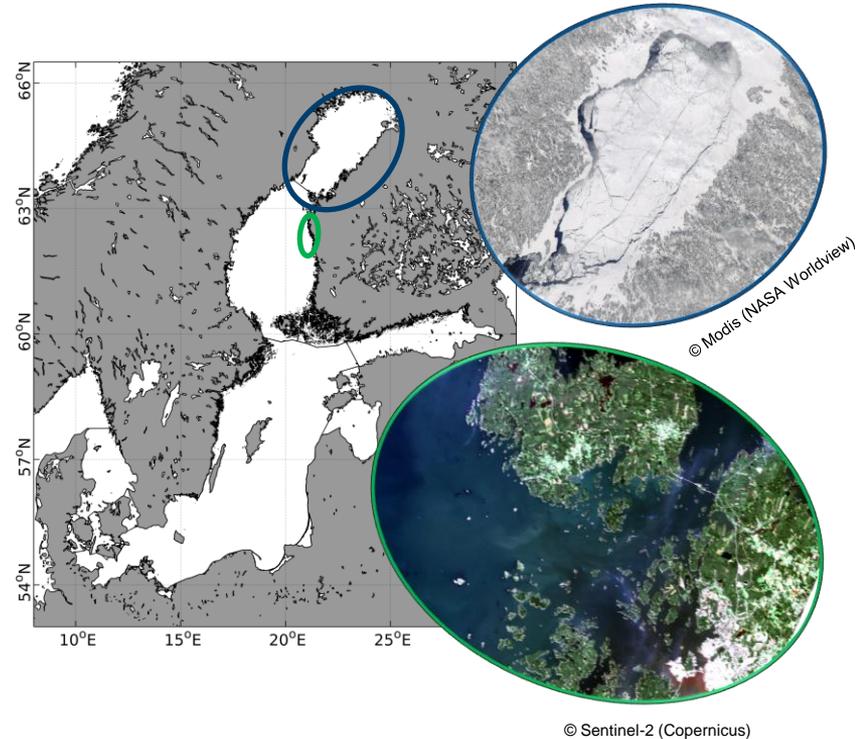
# The Baltic Sea – Motivation

## What?

- Generation of a novel multi-mission sea level (MMSL) along-track and gridded product

## Why?

- Previous products show only sparse information in the northern Baltic Sea (no sea-ice treatment)



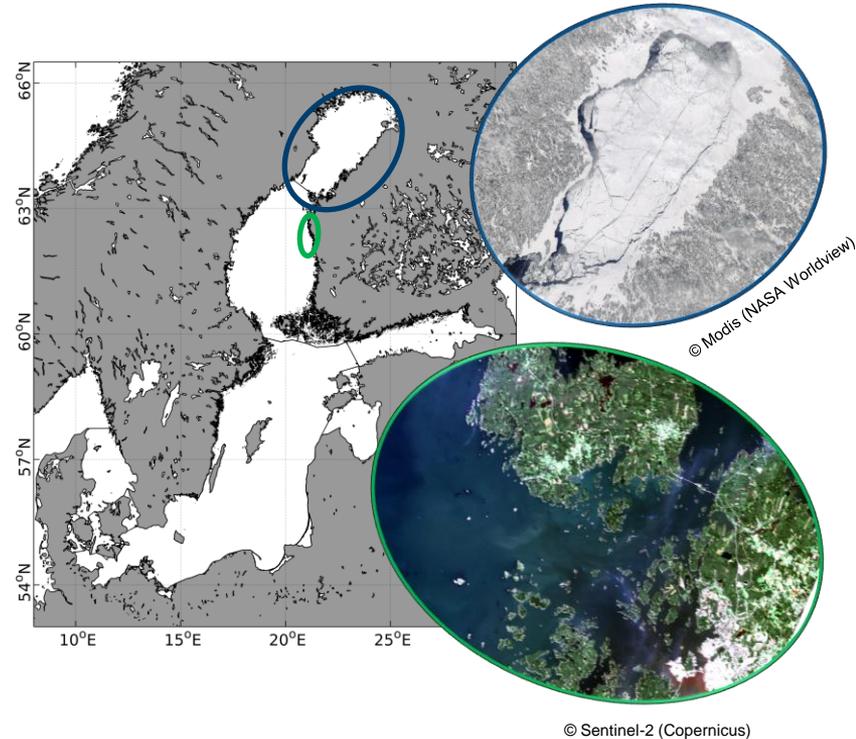
# The Baltic Sea – Motivation

## What?

- Generation of a novel multi-mission sea level (MMSL) along-track and gridded product

## Why?

- Previous products show only sparse information in the northern Baltic Sea (no sea-ice treatment)
- Improvements in algorithms (classification, retracking), geophysical adjustments and corrections, radar techniques (Delay-Doppler Altimetry)



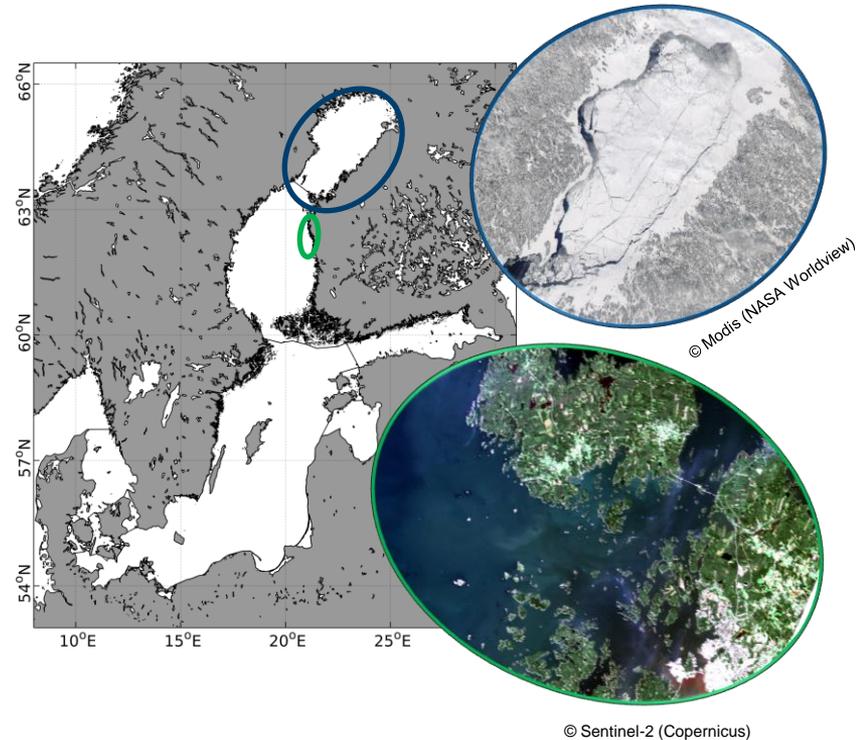
# The Baltic Sea – Motivation

## What?

- Generation of a novel multi-mission sea level (MMSL) along-track and gridded product

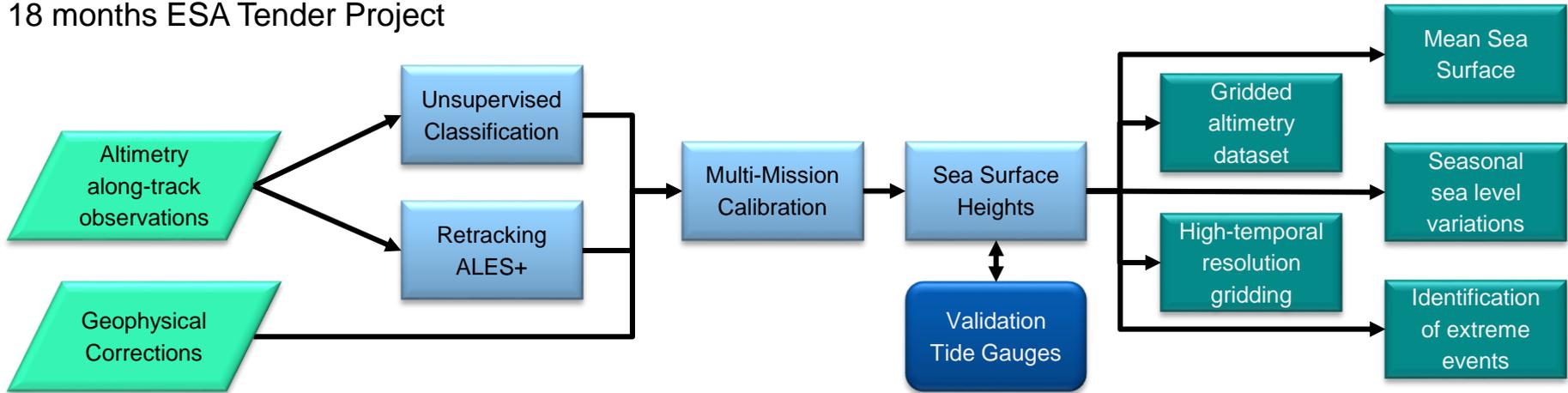
## Why?

- Previous products show only sparse information in the northern Baltic Sea (no sea-ice treatment)
- Improvements in algorithms (classification, retracking), geophysical adjustments and corrections, radar techniques (Delay-Doppler Altimetry)
- Perfect laboratory for Satellite Altimetry (challenging coastlines, sea-ice coverage, small scale variability)



# Workflow Baltic+ SEAL

18 months ESA Tender Project



## 6 Work Packages

1. Scientific Requirements (FMI)
2. Dataset collection (DMI)
3. Algorithm Development and Validation (TUM)
4. Dataset Generation and Impact Assessment (DTU)
5. Scientific Roadmap (UCC)
6. Management and Promotion (TUM)

## Unsupervised Classification to detect leads



- Usage of Artificial Intelligence Algorithms (Data-Mining)
- Definition and computation of waveform features
  - Maximum Power, waveform width, decay of trailing edge etc.  
(Parameters describe the waveform's shape and its features)

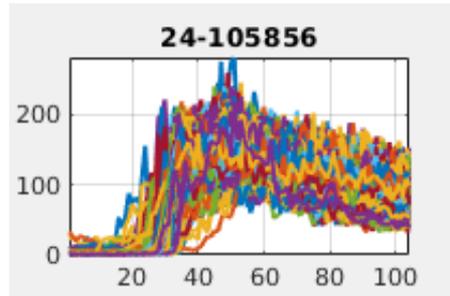
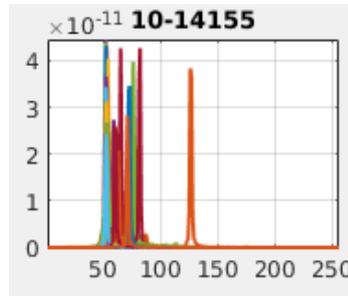


FMI



## Unsupervised Classification to detect leads

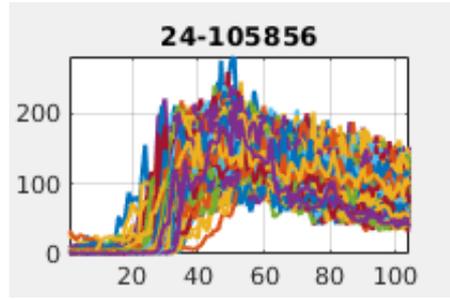
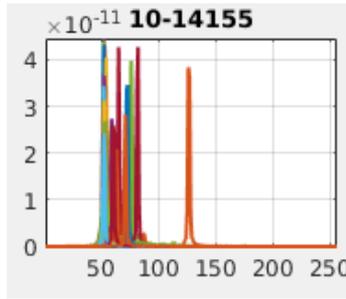
- Usage of Artificial Intelligence Algorithms (Data-Mining)
- Definition and computation of waveform features
  - Maximum Power, waveform width, decay of trailing edge etc. (Parameters describe the waveform's shape and its features)
- Clustering of waveforms applying K-medoids
  - Waveform reference model
- Assigning waveform clusters to surface conditions
  - 4 classes: lead, ocean, sea-ice and undefined



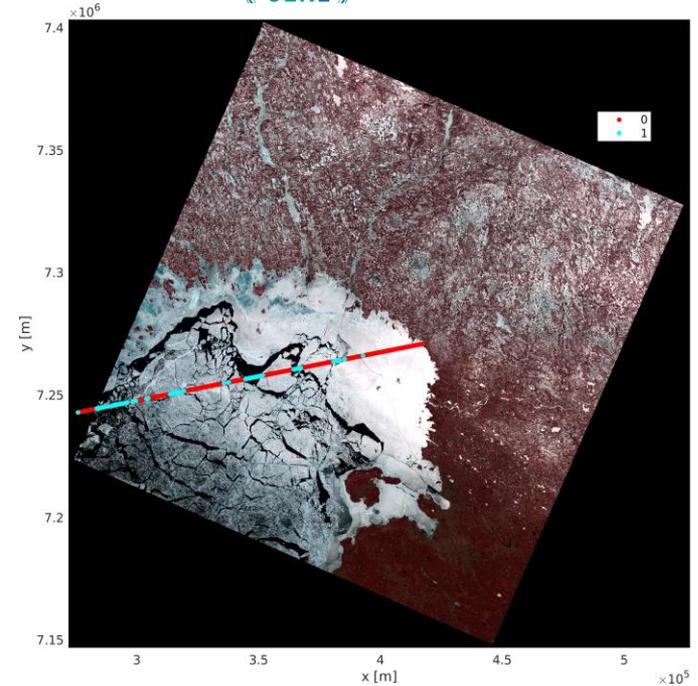
Example: a Lead cluster (left) and an Ocean cluster in Cryosat-2

# Unsupervised Classification to detect leads

- Usage of Artificial Intelligence Algorithms (Data-Mining)
- Definition and computation of waveform features
  - Maximum Power, waveform width, decay of trailing edge etc. (Parameters describe the waveform's shape and its features)
- Clustering of waveforms applying K-medoids
  - Waveform reference model
- Assigning waveform clusters to surface conditions
  - 4 classes: lead, ocean, sea-ice and undefined



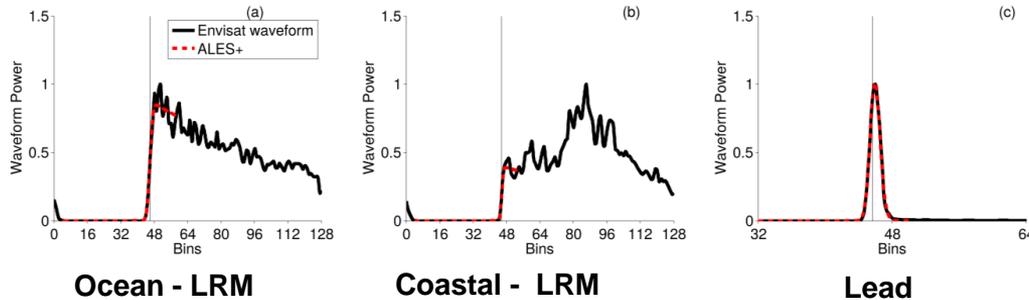
Example: a Lead cluster (left) and an Ocean cluster in Cryosat-2



More info: Müller F.L et al.: **Monitoring the Arctic Seas: How Satellite Altimetry Can Be Used to Detect Open Water in Sea-Ice Regions.** Remote Sensing, 9(6), 551, [10.3390/rs9060551](https://doi.org/10.3390/rs9060551), 2017c

# Retracking

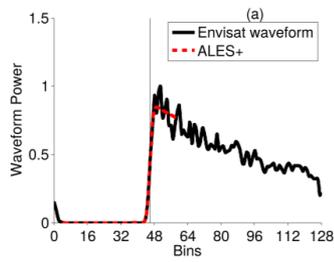
- Sub-waveform retracker to avoid coastal contamination
- Adaptive trailing edge decay to retrack peaky waveforms from leads
- Homogenous range estimation of lead/polynya, open ocean and coastal waveforms (avoids internal biases)



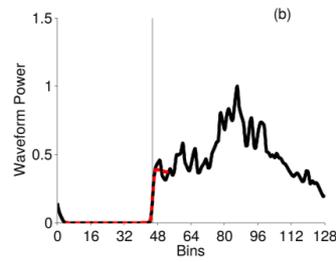
More Information: *Passaro M. et al. (2017): ALES+:  
Adapting a homogenous ocean retracker for satellite  
altimetry to sea ice leads, coastal and inland waters.,  
Remote Sensing of Environment*

# Retracking

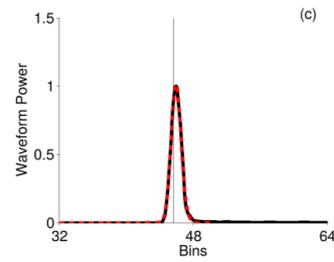
- Sub-waveform retracker to avoid coastal contamination
- Adaptive trailing edge decay to retrack peaky waveforms from leads
- Homogenous range estimation of lead/polynya, open ocean and coastal waveforms (avoids internal biases)



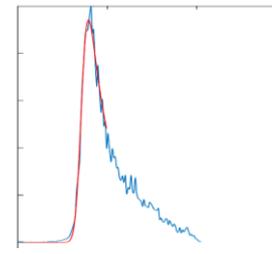
Ocean - LRM



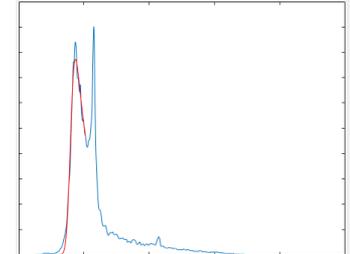
Coastal - LRM



Lead



Ocean - SAR



Coastal - SAR

More Information: *Passaro M. et al. (2017): ALES+:  
Adapting a homogenous ocean retracker for satellite  
altimetry to sea ice leads, coastal and inland waters.,  
Remote Sensing of Environment*

# Sea State Bias



- Correction applied using high-rate ALES+retracked sea state, to decrease intercorrelated errors
- Specific ALES+SAR Sea State Bias model derived

Dataset	XO var before SSB (cm <sup>2</sup> )	XO var after SSB (cm <sup>2</sup> )	Explained Variance
Gaspar et al. (1994)	127.7	120.4	6%
SGDR Jason Med Passaro et al. (2018)	135.6	108.4	20%
ALES+ SAR	106.0	84.9	20%

Variance at crossover locations before and after the application of the sea state bias correction



FMI



University College Cork, Ireland  
Coláiste na hOllscoile Corcaigh

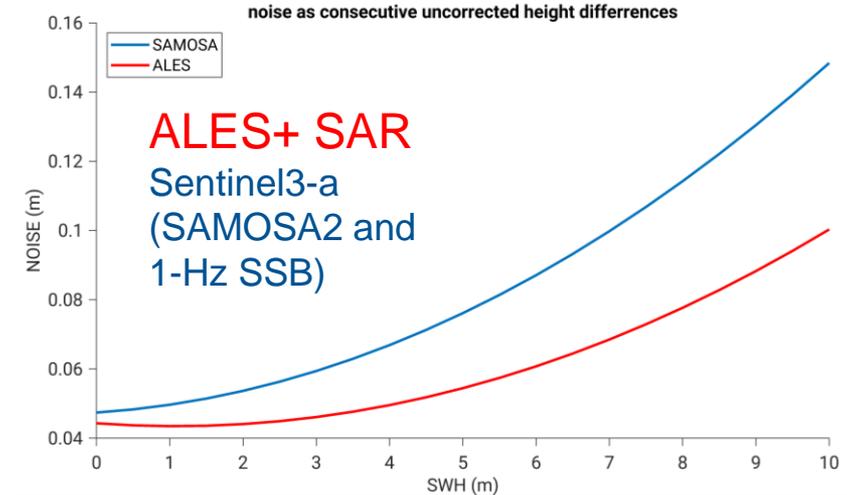
# Sea State Bias



- Correction applied using high-rate ALES+retracked sea state, to decrease intercorrelated errors
- Specific ALES+SAR Sea State Bias model derived

Dataset	XO var before SSB (cm <sup>2</sup> )	XO var after SSB (cm <sup>2</sup> )	Explained Variance
Gaspar et al. (1994)	127.7	120.4	6%
SGDR Jason Med Passaro et al. (2018)	135.6	108.4	20%
ALES+ SAR	106.0	84.9	20%

Variance at crossover locations before and after the application of the sea state bias correction



More info: *Passaro M. et al.*: **Improving the precision of sea level data from satellite altimetry with high-frequency and regional sea state bias corrections.** Remote Sensing of Environment, 245-254, [10.1016/j.rse.2018.09.007](https://doi.org/10.1016/j.rse.2018.09.007), 2018



FMI



University College Cork, Ireland  
Coláiste na hOllscoile Corcaigh



## Cross-Calibration

- Offset not always time-constant => drifts!
- Differences in sea level heights can have large-scale geographical pattern
- A location-dependent cross-calibration between all missions is needed
- Output: time series of radial errors => applied as corrections to each measurement

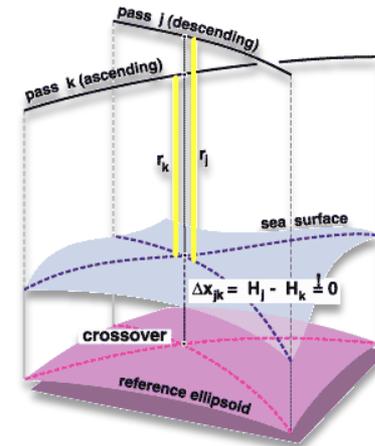
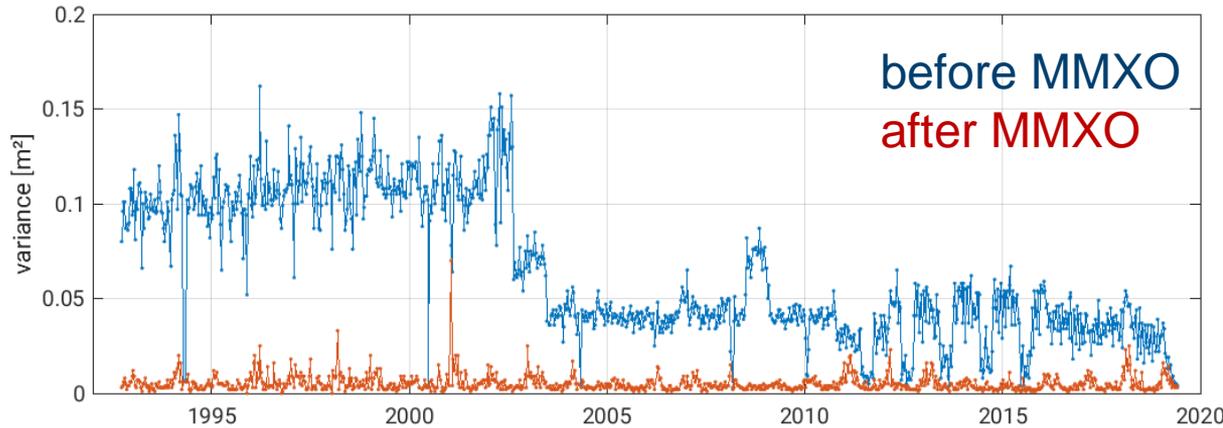


FMI



## Cross-Calibration

- Offset not always time-constant => drifts!
- Differences in sea level heights can have large-scale geographical pattern
- A location-dependent cross-calibration between all missions is needed
- Output: time series of radial errors => applied as corrections to each measurement

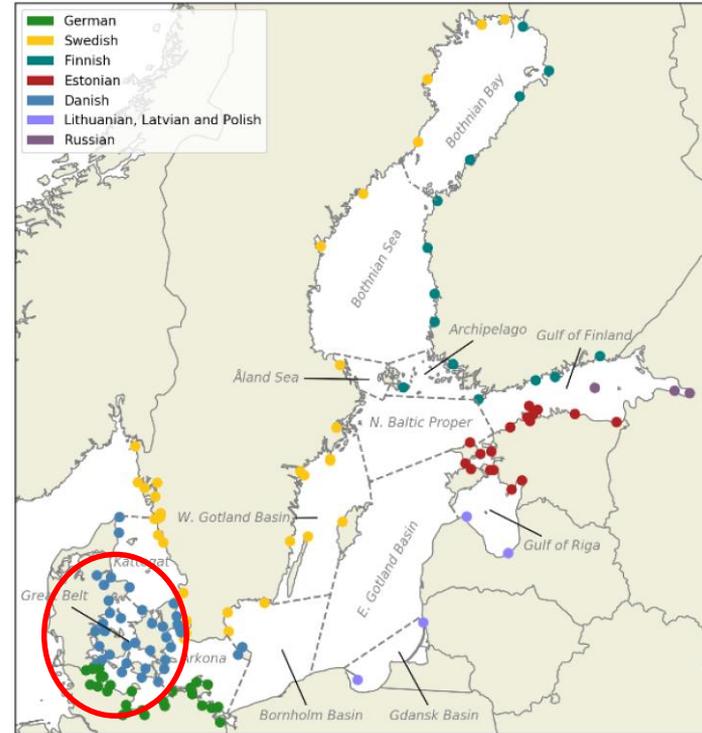


More Information: *Bosch W. et al.: Multi-mission cross-calibration of satellite altimeters: constructing a long-term data record for global and regional sea level change studies.* Remote Sensing 6(3): 2255-2281, [10.3390/rs6032255](https://doi.org/10.3390/rs6032255), 2014

# First validation

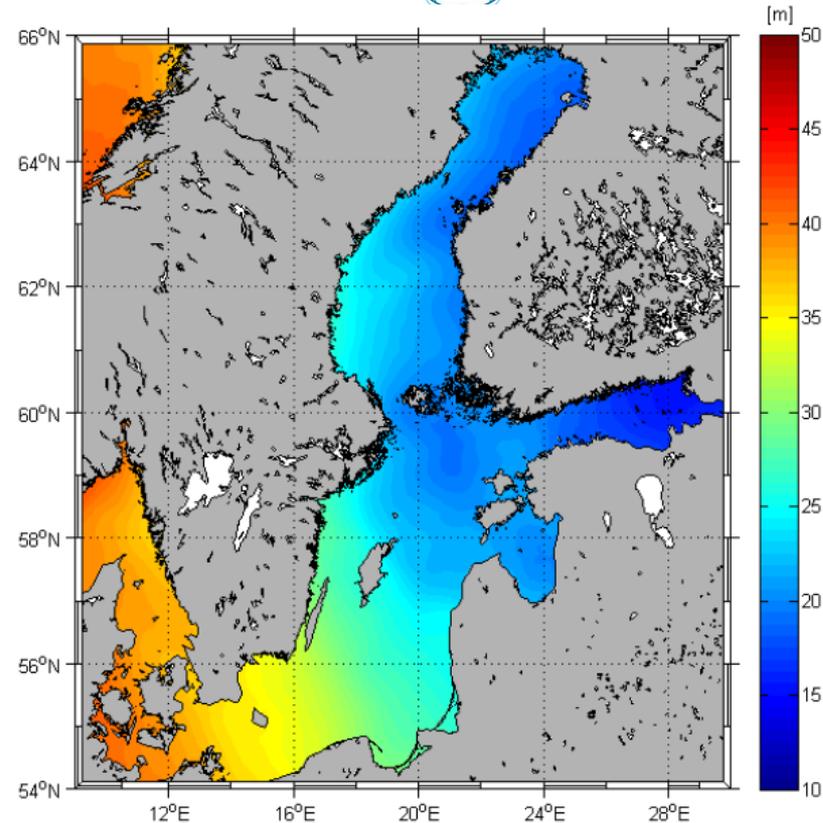
	ALES+	SAMOS2
Valid Points	727	<690
Median Correlation (only of valid points)	0.93	0.88

- 31 cycles of Sentinel-3a
- Points within 5 km from the coast averaged and compare to Tide Gauges
- Outlier check based on Tide Gauges



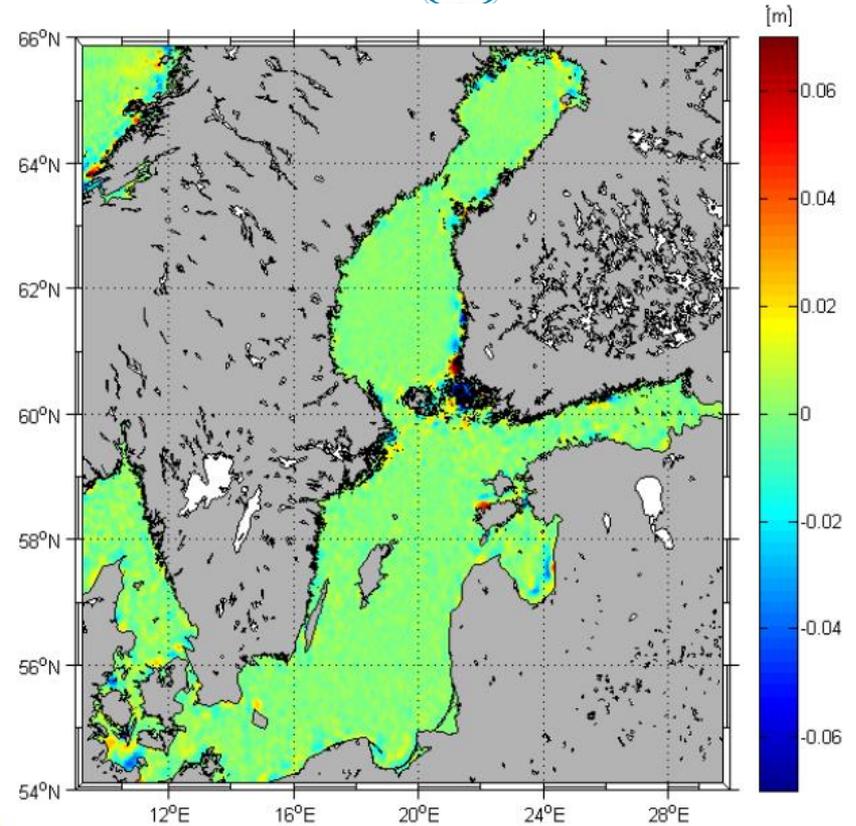
## Mean Sea Surface

- A new Mean Sea Surface is being produced within the project
- In the current version, leads among sea ice and SAR altimetry not yet included
- ...and yet, if we consider the differences with DTU15...



# Mean Sea Surface

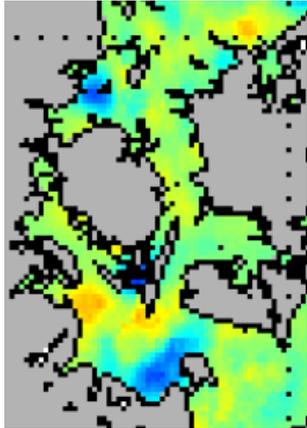
...the secret lies in the details!  
Shown are differences between new MSS and DTU15 model



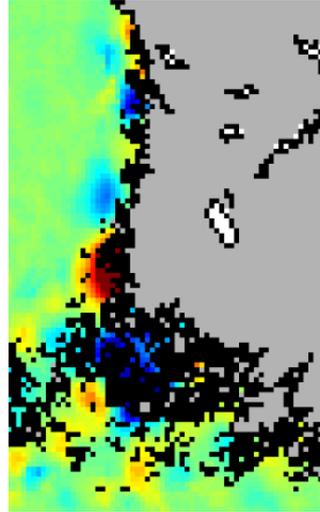
# Mean Sea Surface

...the secret lies in the details!

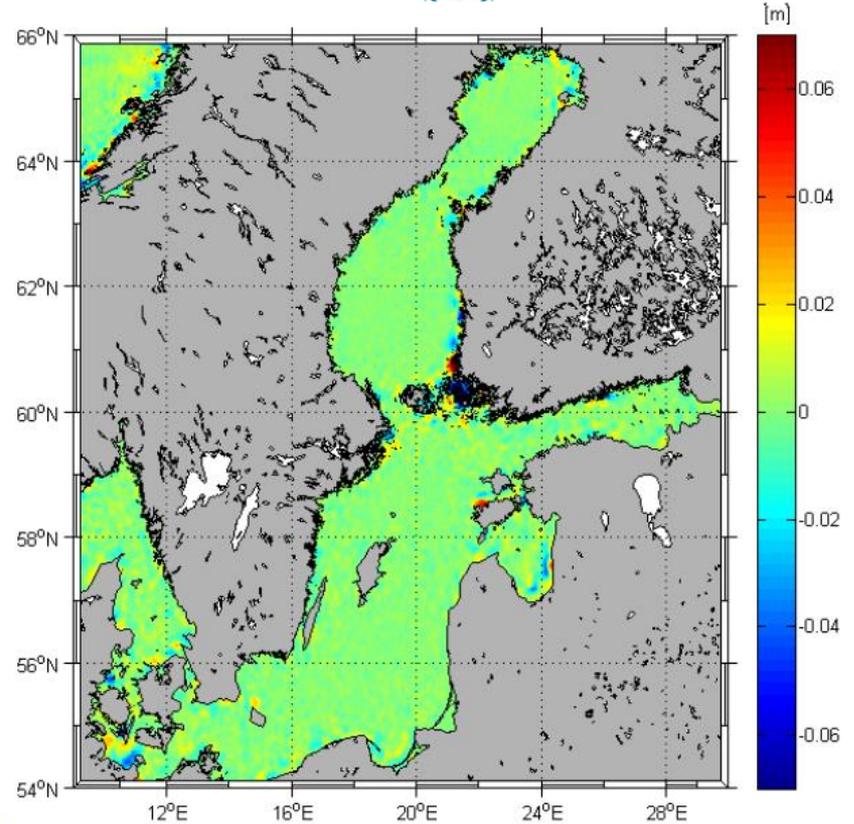
Shown are differences between new MSS and DTU15 model



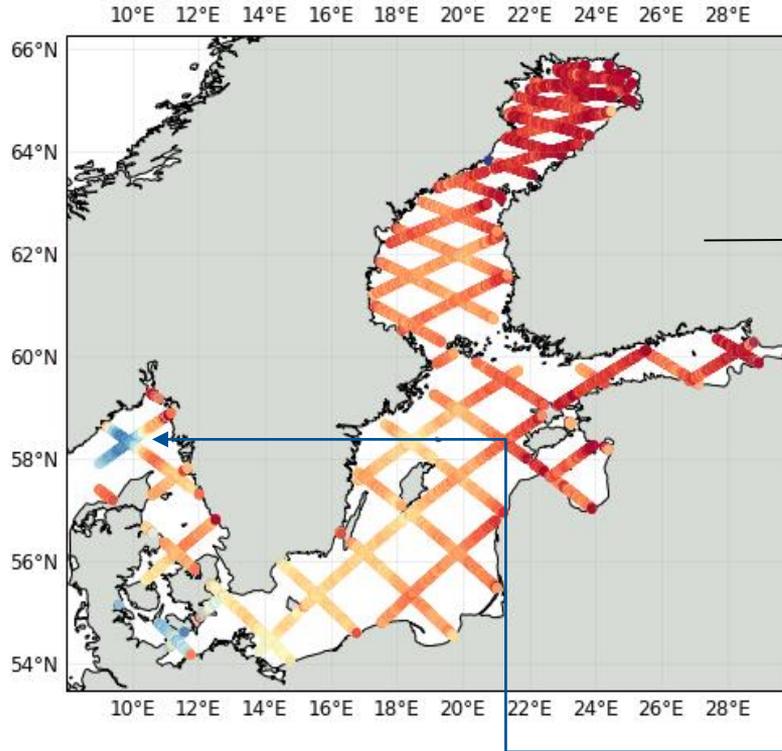
Danish Straits



Finnish Coast



# Sea Level Analysis: First Results -> Amplitude of annual cycle

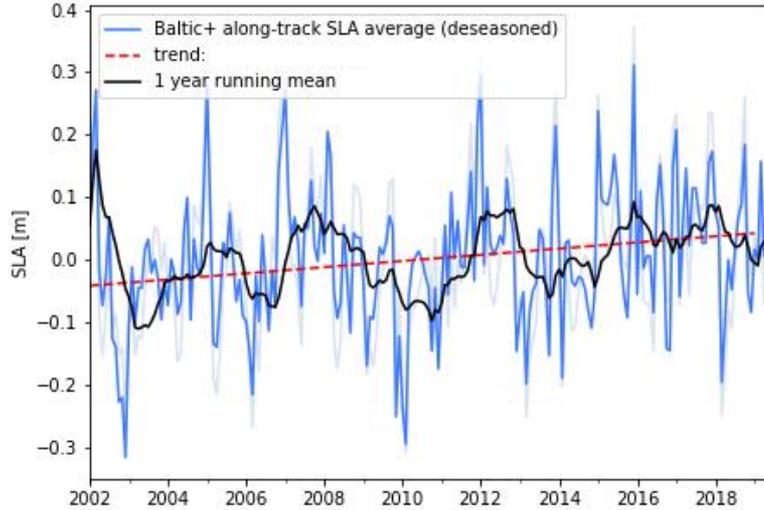


Mean annual cycle amplitude:  
9cm

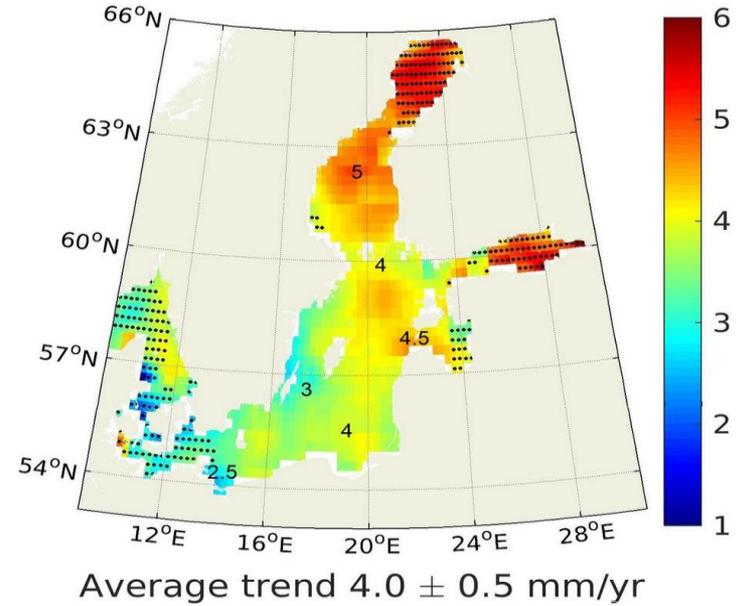
consistent with  
Stramska, M., and Chudziak, N.  
(2013)

Sloping annual cycle amplitude in the  
Skagerrak Sea [Passaro 2015]

# Sea Level Analysis: First Results



Average trend  $4.433 \pm 2.21$  mm/year



Madsen et al. 2019

Higher uncertainty: just Jason mission for now and shorter time series



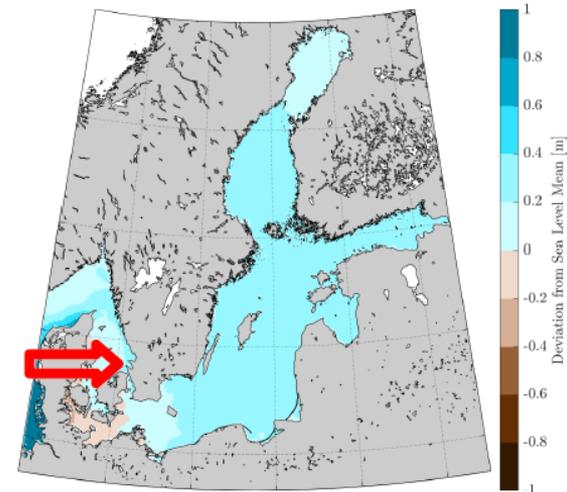
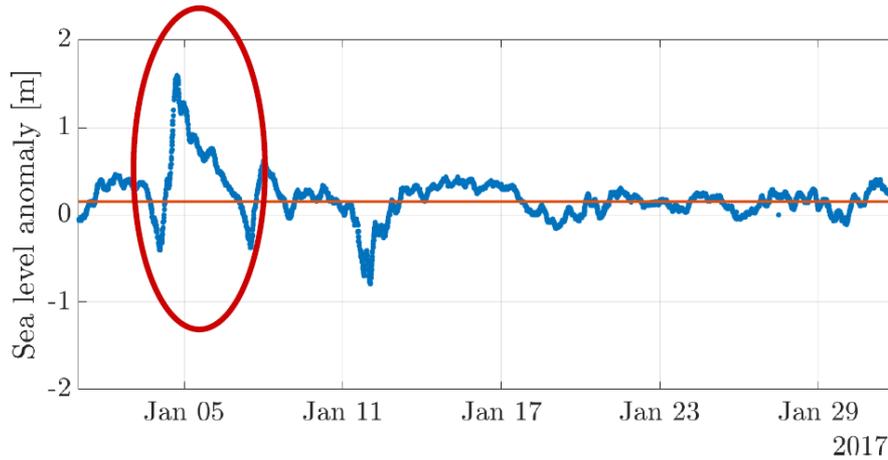
FMI



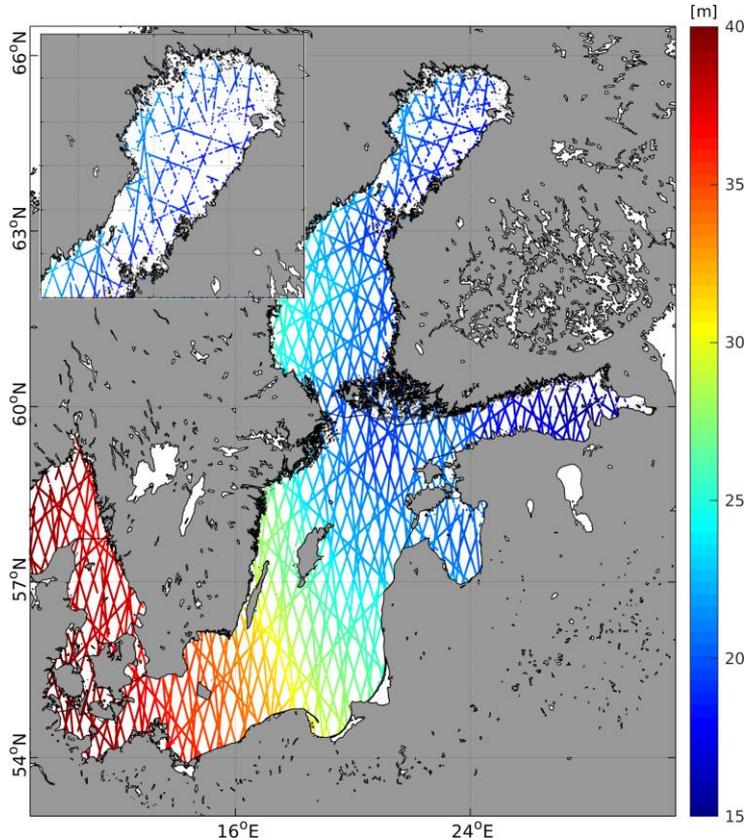
University College Cork, Ireland  
Coláiste na hOllscoile Corcaigh

## Additional products: experimental high-temporal resolution gridding

- The recent availability of several altimeters opens new possibilities for optimal interpolation every few days
- The objective is to provide a storm surge model with the best possible initial state (SSH BEFORE the surge)



## Conclusions



- Baltic SEAL is a laboratory for advanced satellite altimetry
- Regionally-dedicated analysis, featuring the latest advances at different level of products (from classification to gridding)
- By the end of 2020, a multi-mission cross-calibrated along-track product, a monthly gridded product, a new mean sea surface will be available to all users
- The techniques exploited in this framework can be easily transferred to other regions



Home About Outputs Data Access Events Contact Us



Developing improved sea level data for the Baltic Region, using satellite data

Tweets by @Baltic\_SEAL

**BalticSEAL** @Baltic\_SEAL  
So who is actually delivering Baltic SEAL?  
Meet the team at [balticseal.eu/partners/](http://balticseal.eu/partners/)  
And if you see us at an event, pop over for a chat :)  
Jan 21, 2020

**BalticSEAL** @Baltic\_SEAL  
Check out our sister project Baltic+ Geodetic SAR at [bgu.tum.de/en/lapgb/baltic/](http://bgu.tum.de/en/lapgb/baltic/)  
Jan 15, 2020

# Thank you!

Stay tuned and visit:  
**balticseal.eu**



FMI



University College Cork, Ireland  
Coláiste na hOllscoile Corcaigh



# SPARE SLIDES



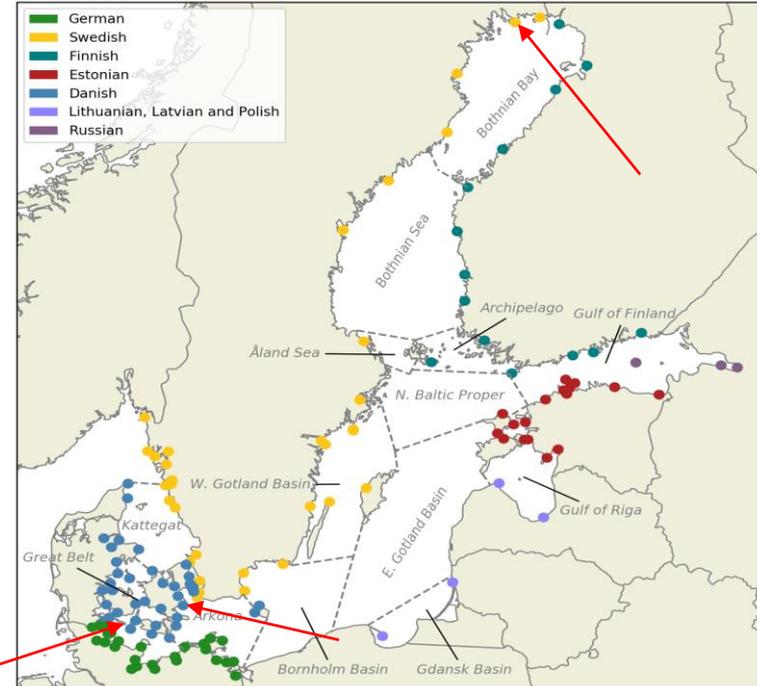
FMI



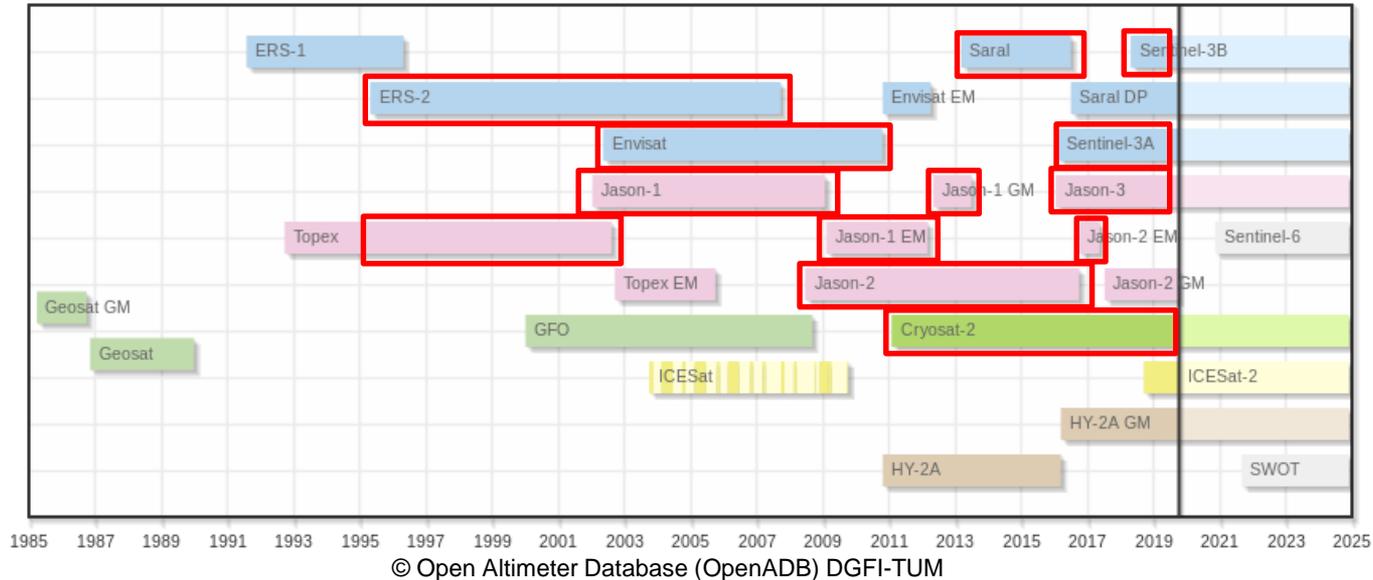
## Validation

- Data averaged in the last 10 km
- ALES+ increases the valid cycles in the sea-ice covered areas

TG	RMSE (m)	r	n	SAMOS2	ALES+
Bagenkop	0.05	0.95	27	x	
Bagenkop	0.05	0.95	28		x
Kalix-Storön	0.18	0.71	23		x
Rodvig	0.07	0.94	29	x	
Rodvig	0.07	0.95	29		x



# Dataset collection (Altimetry)

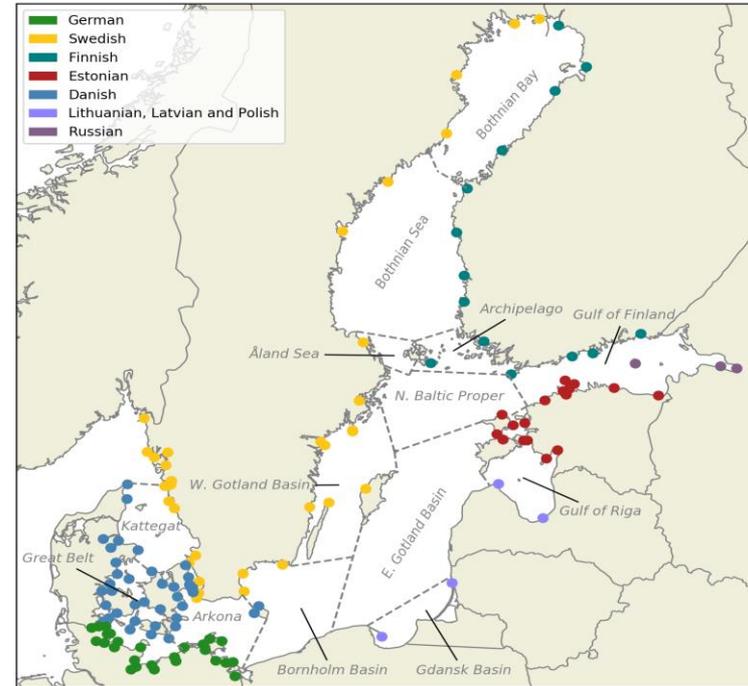


- ~ 25 years Multi-mission altimetry data (LRM & SAR)
- Usage of ALES+ retracked high-frequency along-track observations
- Multi-mission cross calibrated Sea Surface Heights
  - Regional cross calibration based on high-frequency along-track observations



## Dataset collection (Sea Surface Height Validation)

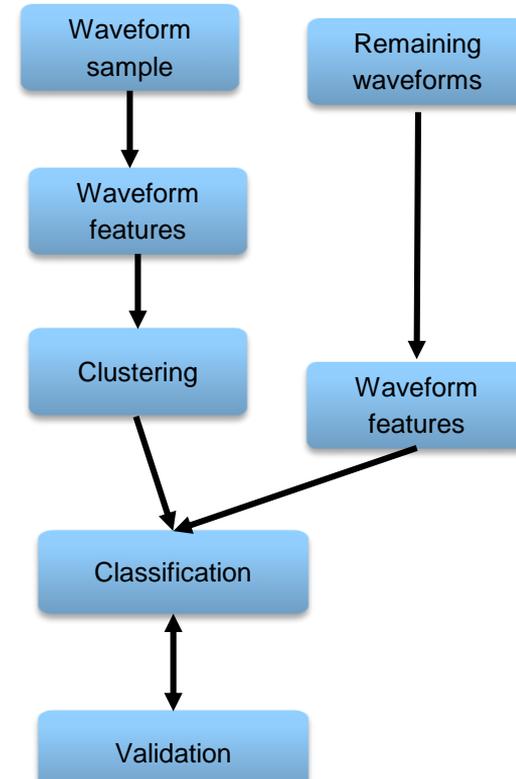
- Spatially comprehensive distribution of tide gauge stations
- All shown tide gauge stations available via CMEMS
- More tide gauge data can be requested from Finnish Meteorological Institute and Swedish Meteorological and Hydrological Institute
- Height conversion from national reference heights to EVRF2007



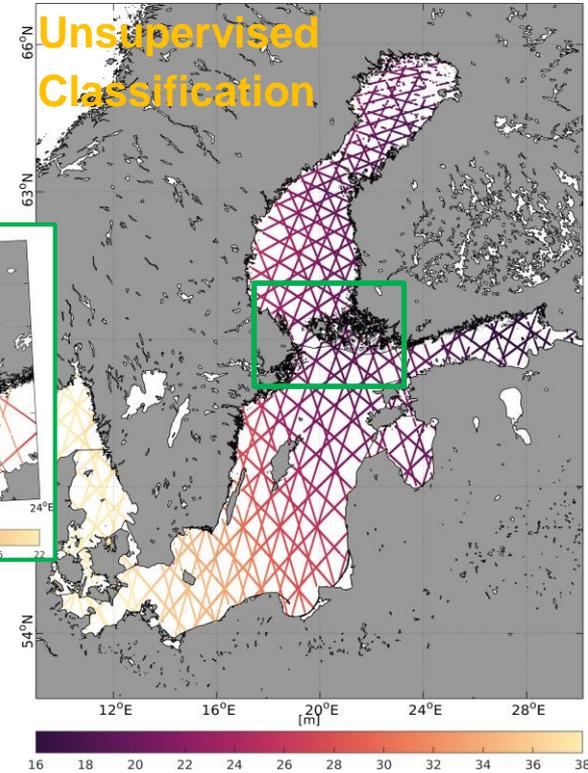
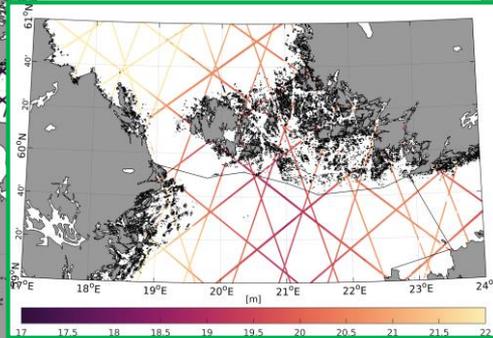
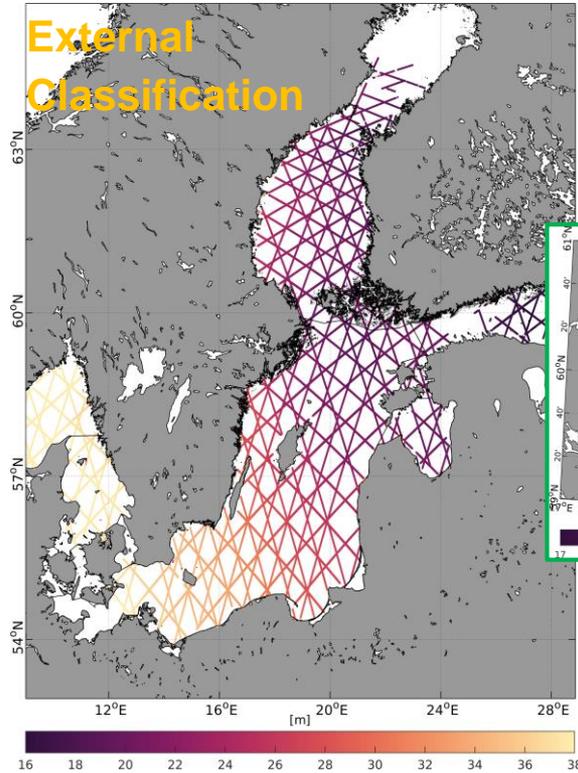
Tide gauge station locations

# Algorithm Development and Validation (Unsupervised Classification)

- Usage of Artificial Intelligence Algorithms (Data-Mining)
- Application of unsupervised (no training data) classification for detecting leads (water openings within the ice) → see Müller et al. 2017
- **Input: Original waveform data**
- Definition and computation of waveform features
  - Maximum Power, waveform width, decay of trailing edge etc. (Parameters describe the waveform's shape and its features)
- Clustering of waveforms in clusters applying K-medoids
  - Waveform reference model
- Assigning waveform clusters to surface conditions
  - 4 classes: calm water, ocean, sea-ice and undefined
- Classification of remaining waveforms using reference model and K-nearest neighbor (K-NN)
- **Classification output:** WATER [1] | ICE [0] | UNDEFINED [0] (per measurement)
- Same method for LRM and SAR missions, but slightly different feature space



# Algorithm Development and Validation (Sea Surface Heights)



Sea Surface Heights ALES+ 2009-04 (J1-EM,J2,EN)

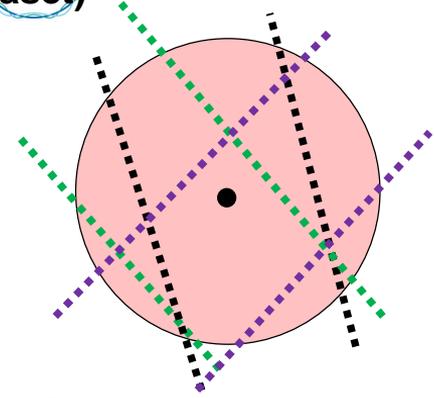


## Dataset Generation and Impact Assessment (Gridded/Meshed altimetry dataset)

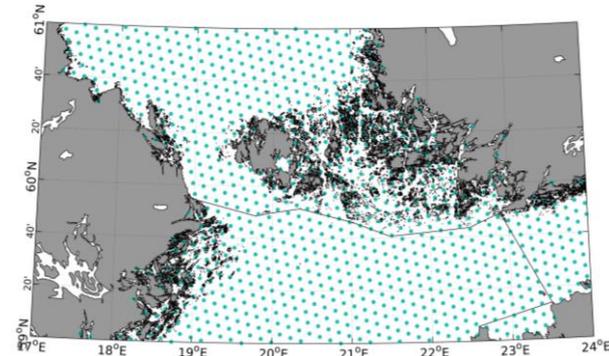
- Generation of a meshed altimetry dataset (TUM)
- Gridding profiled SSH data on an unstructured, triangulated grid
- Monthly temporal resolution 7,5 km spatial resolution
- Using least-square approach with spatio-temporally Gaussian weighting
  - Fitting a plane to the individual grid nodes
  - Defining proper weights and circular cap size

$$h(x, y) = c_0 + c_1x + c_2y$$

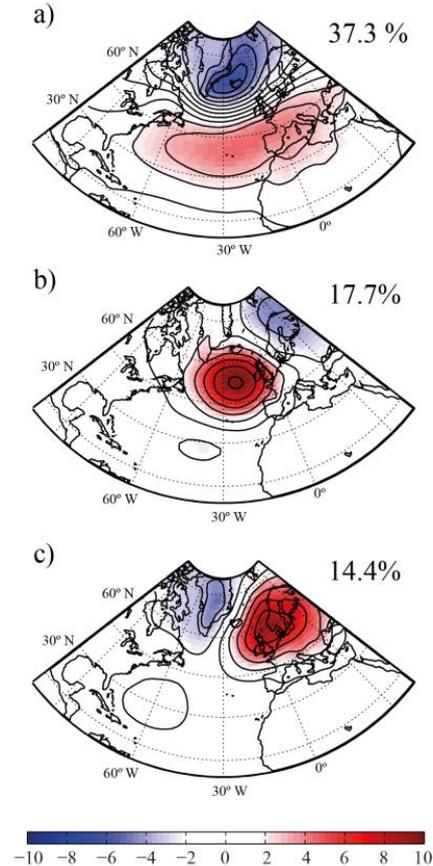
- Production of high-temporal resolution gridding (DTU/DMI)
- Development of a high-temporal (1-3 days) gridded dataset
- Combination of altimetry, tide gauges observations and the output of a hydrodynamic ocean model
- Based on Optimal Interpolation and error covariance statistics



Grid-node with cap size and sea surface height distribution



Unstructured grid



DJF – SLP leading  
modes  
[mb]

NAO

EA