## Convolutional Neural Networks for Satellite-Derived Bathymetry





#### Outline

- Introduction
- Related work
- Methodology
- Experiments and results
- Conclusions



## Introduction



### What is bathymetry?





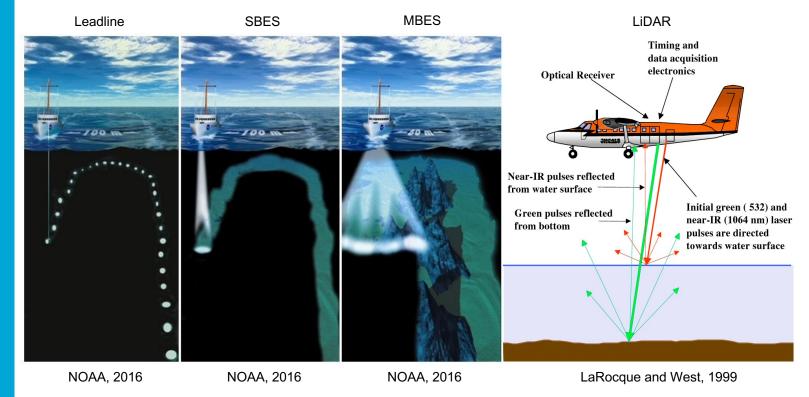
Bathymetry surveys reveal the terrain of water bed



Essential for coastal management and research

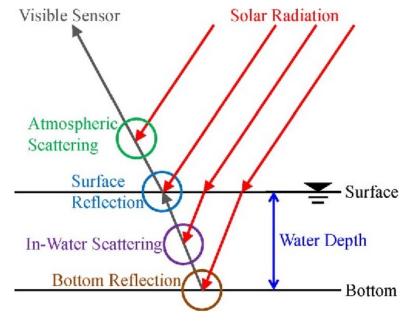


#### How to get bathymetry data?





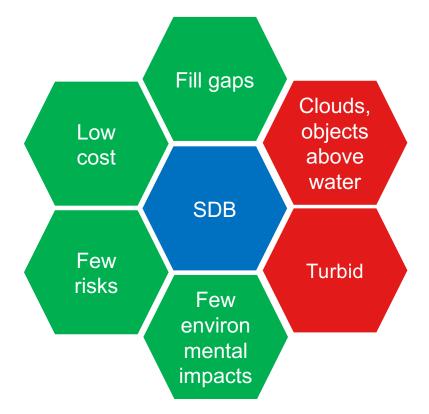
#### Satellite-Derived Bathymetry



Kanno et al., 2012



## Why SDB?

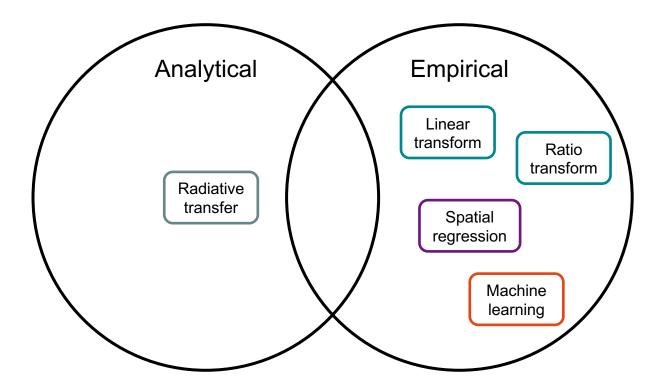




## Related work

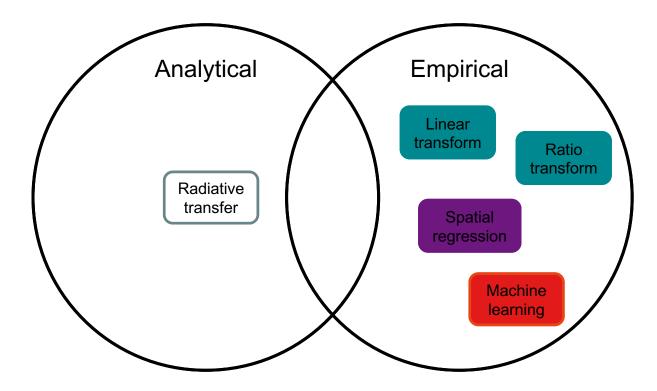


#### SDB methods





#### SDB methods





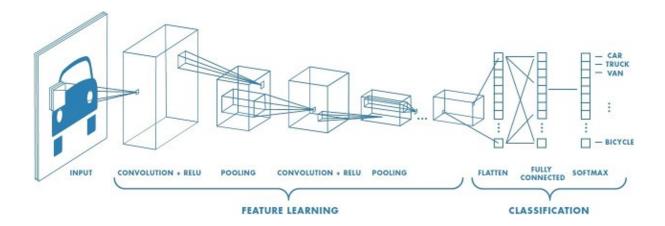
#### Factors to consider

• Nonlinearity due to bottom types/noise

Spatial correlation



#### Convolutional Neural Networks





#### Research objective

 To develop convolutional neural networks that extract accurate water depth in shallow water areas



#### Research questions

Main question:

 To what extent can convolutional neural networks be used for accurate shallow water depth extraction using Sentinel-2 satellite images?

Sub questions:

- What kind of pre-processing is needed for the data sets?
- What kind of CNN architecture can be used for SDB?
- What is the accuracy of the method?
- To what extent can the pretrained model be reused?



# Methodology



#### Study areas

- Puerto Rico
- Key West
- Hawaii







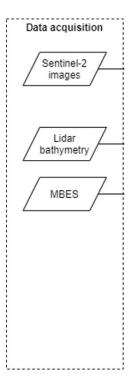
#### Data

NOAA: Bathymetric Data Viewer

Sentinel-2 Level 2A Copernicus Open Access Hub

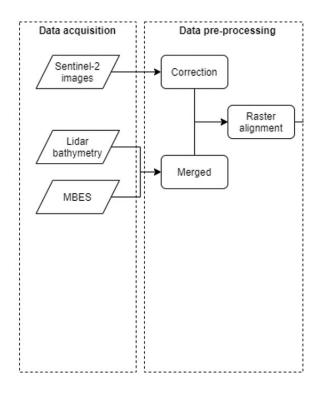


#### Workflow



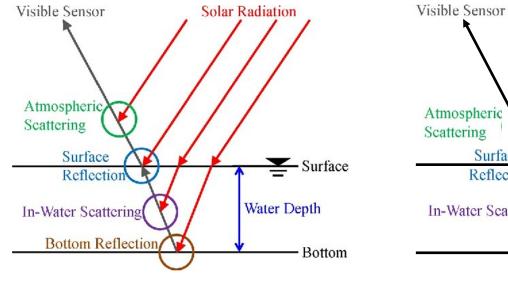


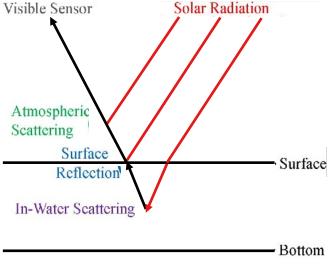
#### Workflow





#### Image correction



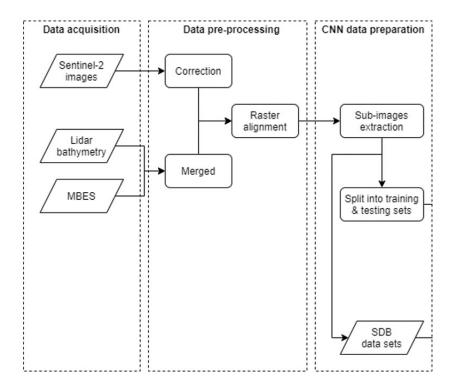


Shallow water

Deep water

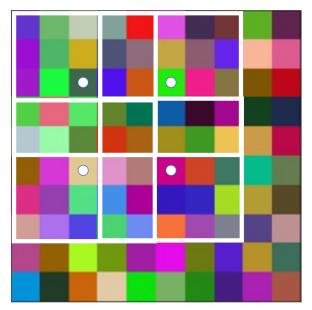


Workflow





### Sub-images extraction



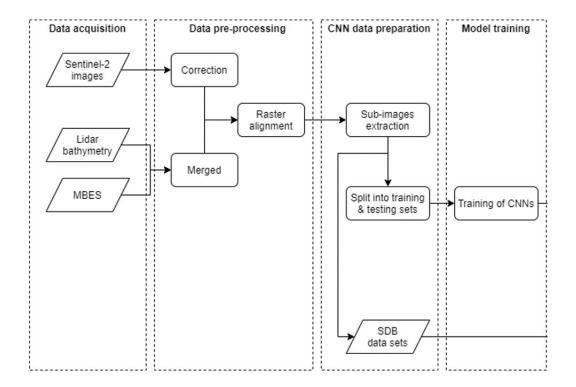
Sub-images for training and testing



Sub-images for SDB model generation

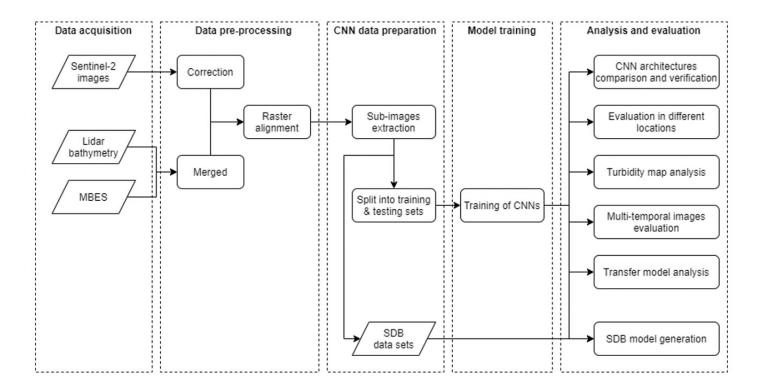


Workflow





Workflow





Experiment #1: CNN architectures comparison and verification





#### Study area CNN architecture

#### AOI-1

Layer	CNN1	CNN2	CNN3	CNN4
Conv2D	2x	2x	3x	3x
Kernel	2x2	2x2	3x3	3x3
Pooling	No	Yes	No	Yes

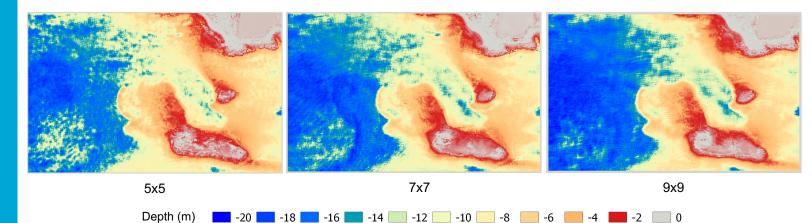
Window size Band combination 5x5, 7x7, 9x9

RGB, RGBN, RGBNSS, All bands



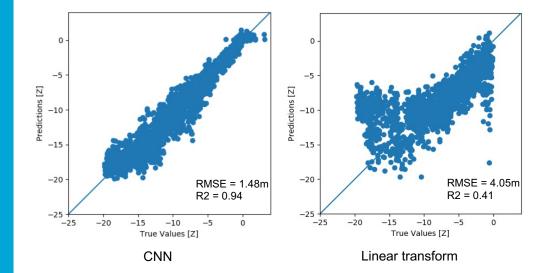
#### Window sizes

Models	5x5		7x7		9x9	
	RMSE	R2	RMSE	R2	RMSE	R2
CNN1	1.59	0.91	1.58	0.92	1.55	0.93
CNN2	1.94	0.89	1.63	0.91	1.55	0.93
CNN3			1.53	0.92	1.48	0.94
CNN4					1.64	0.90



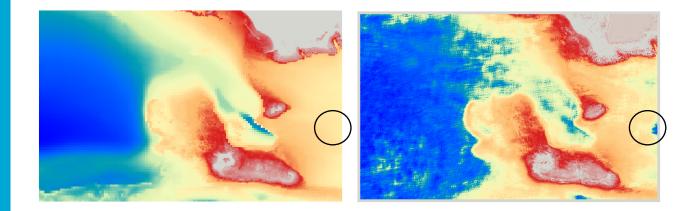


#### Comparison to the linear transform



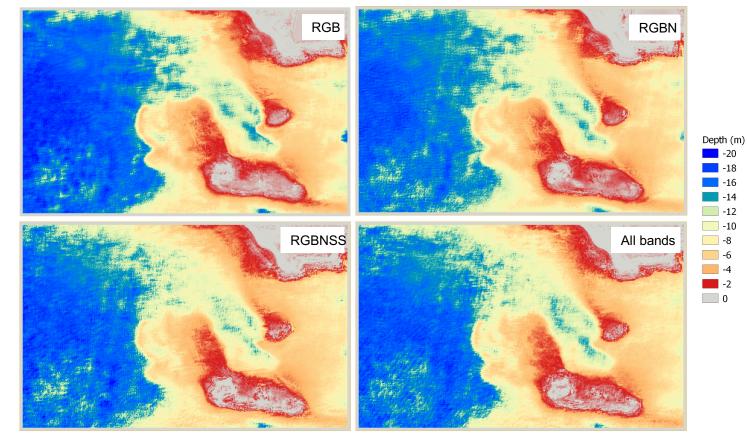


#### Prediction vs ground truth





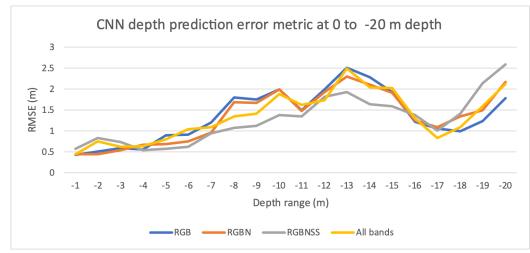
#### Band combinations



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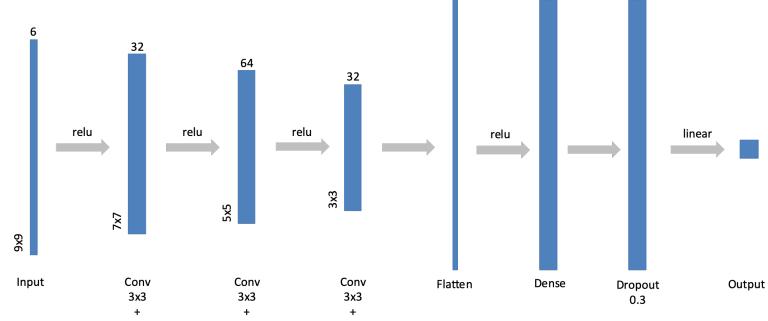
#### Band combinations

Channels	5x5		7x7		9x9	
	RMSE	R2	RMSE	R2	RMSE	R2
RGB	1.59	0.91	1.53	0.92	1.48	0.94
RGBN	1.64	0.91	1.40	0.93	1.37	0.94
RGBNSS	1.63	0.91	1.53	0.91	1.31	0.94
All bands	1.68	0.90	1.55	0.91	1.45	0.94





#### Baseline architecture



Batch normalization Batch normalization Batch normalization



Experiment #2: SDB comparison in different locations

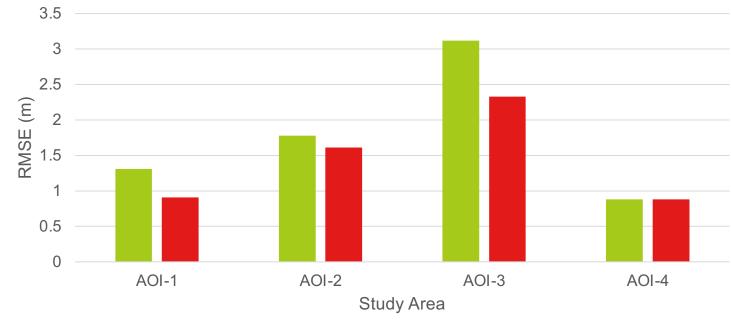


#### Setup

Study area Architecture Window size Band combination AOI-2, AOI-3, AOI-4 CNN3 9x9 RGBNSS



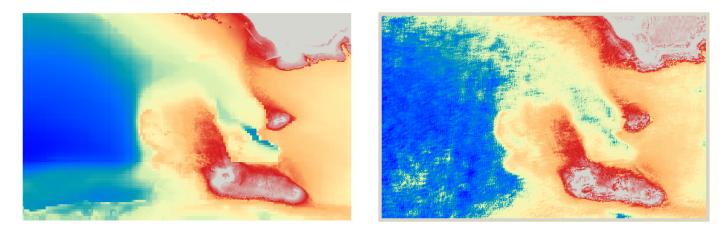
#### Accuracy



■0-20 m ■0-10 m



### Ground truth vs SDB (AOI-1)

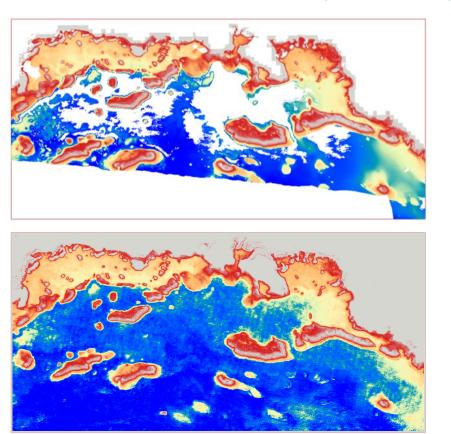


#### -20 -18 -16 -14 -12 -10 -8 -6 -4 -2 0



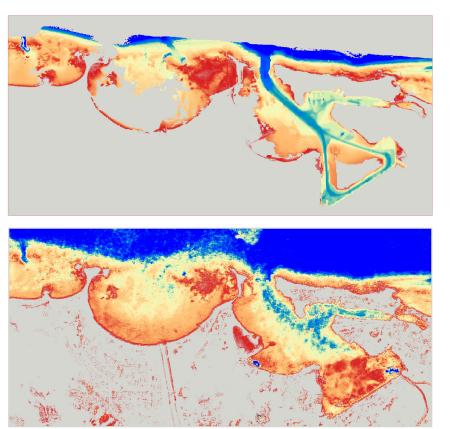
#### Ground truth vs SDB (AOI-2)

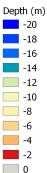
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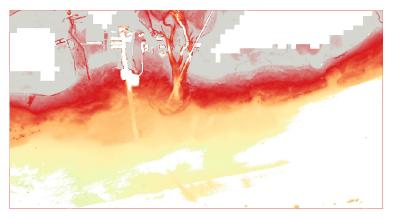
#### Ground truth vs SDB (AOI-3)

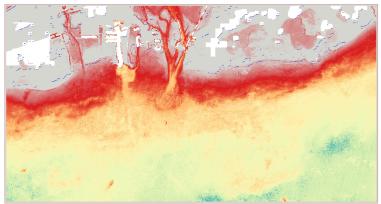






#### Ground truth vs SDB (AOI-4)





 h (m)
-20
-18
-16
-14
-12
-10
-8
-6
-4
-2
0

Accuracy (m)	CNN	Ratio Transform
MedAE	0.29	0.39



Experiment #3: SDB using uncorrected images

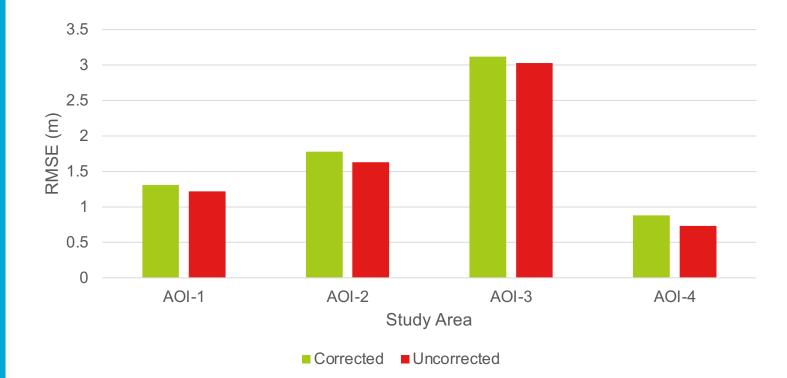


#### Motivation

 In some cases, image correction using deep water pixels encounters difficulty due to the large absorption of lights by the molecules in the water column



#### Accuracy





# Experiment #4: Multi-temporal images

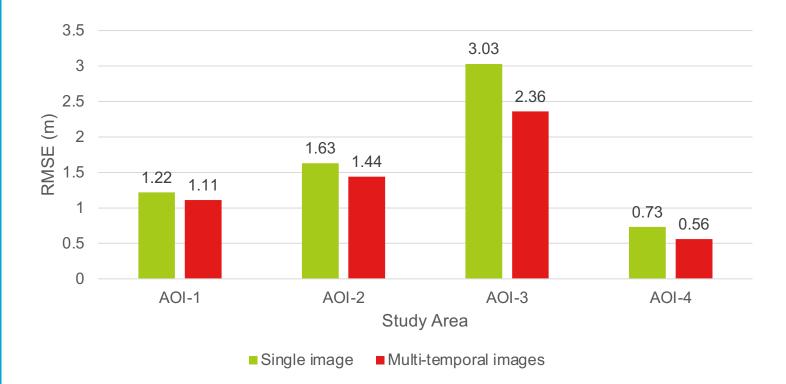


#### Setup

Study area Architecture Window size Band combination Image collection AOI-1, AOI-2, AOI-3, AOI-4 CNN3 9x9 RGBNSS 2019



#### Accuracy





Experiment #5: Transfer model



#### Setup

Study areaAOI-5, AOI-6ArchitectureCNN3Window size9x9Band combinationRGBNSS



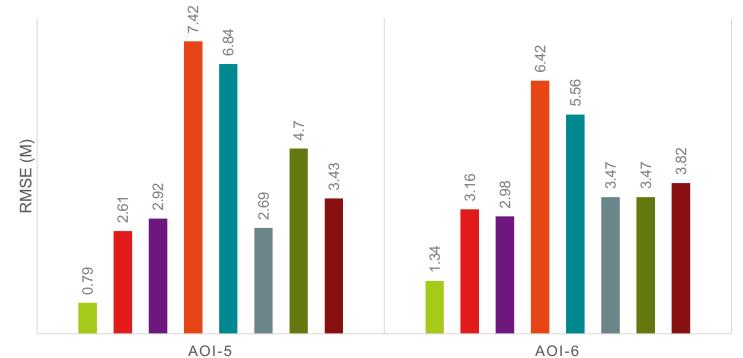
#### Accuracy (AOI-5 & AOI-6)

RMSE (m)	CNN	Other methods	
AOI-5	0.79	1.24	Random Forest
AOI-6	1.34	3.01	Radiative transfer



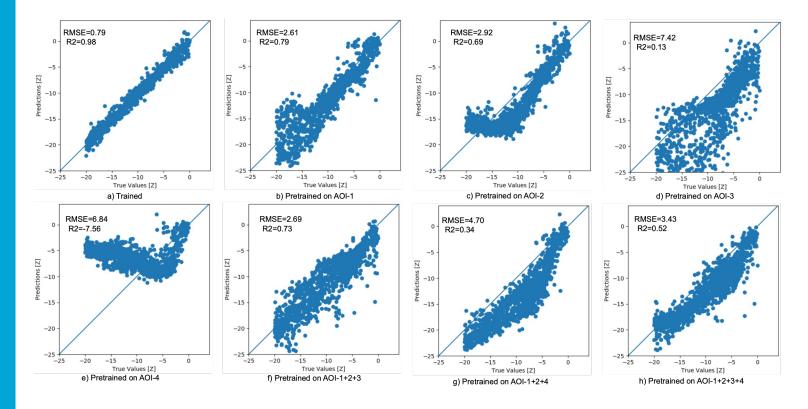


#### ■ Trained ■ AOI-1 ■ AOI-2 ■ AOI-3 ■ AOI-4 ■ Combined 123 ■ Combined 124 ■ Combined 1234



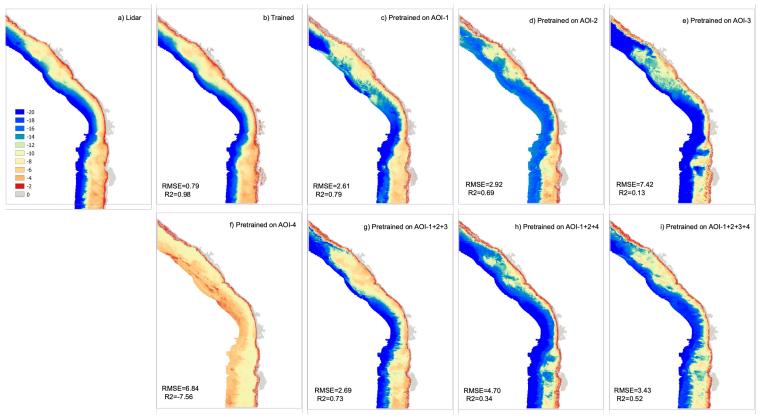


#### Prediction vs ground truth AOI-5





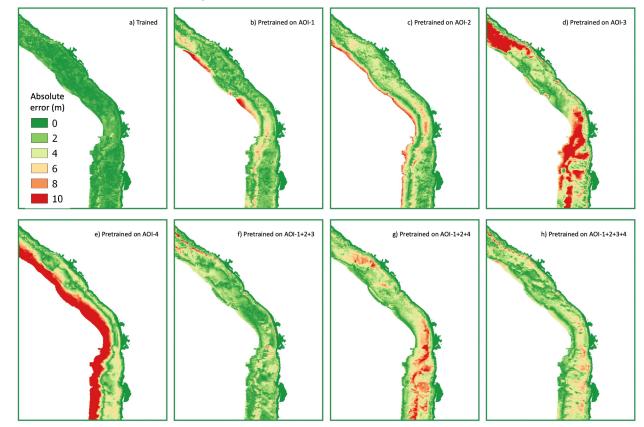
#### SDB AOI-5



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#### Error map AOI-5

**ŤU**Delft



Experiment #6: SDB production in a larger area

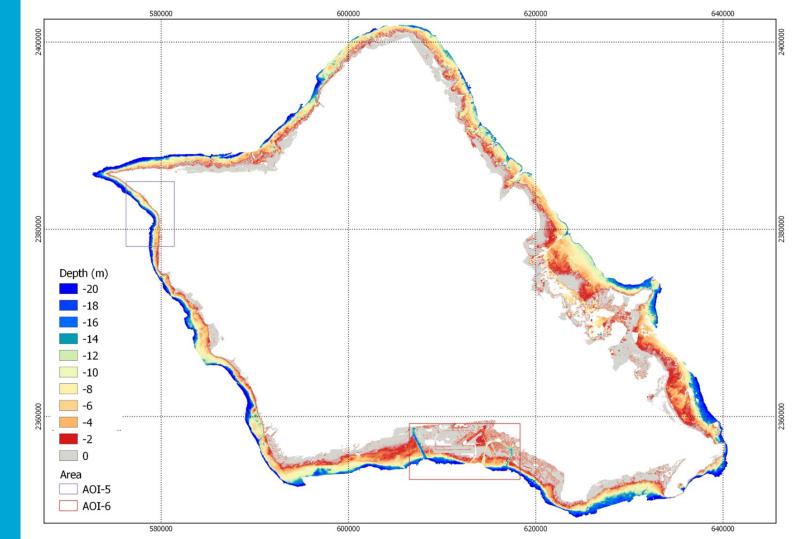


#### Setup

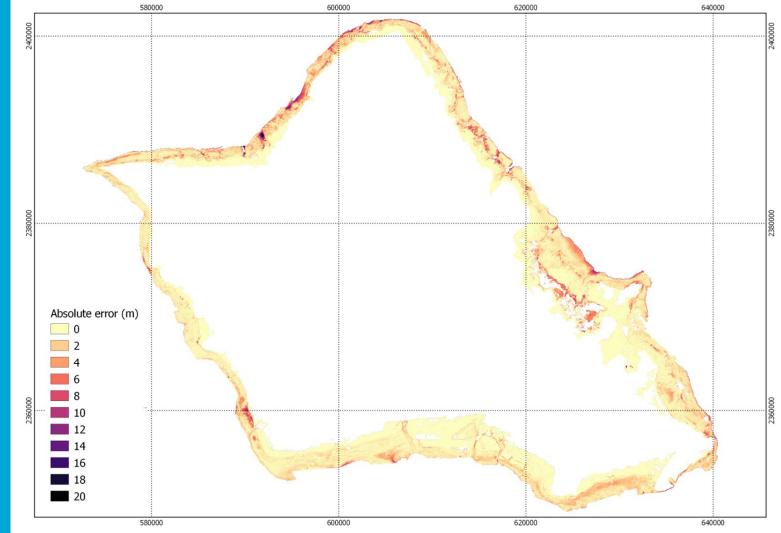
Study area Architecture Window size Band combination Training data

Oahu island CNN3 9x9 RGBNSS AOI-5 + AOI-6





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# Conclusions

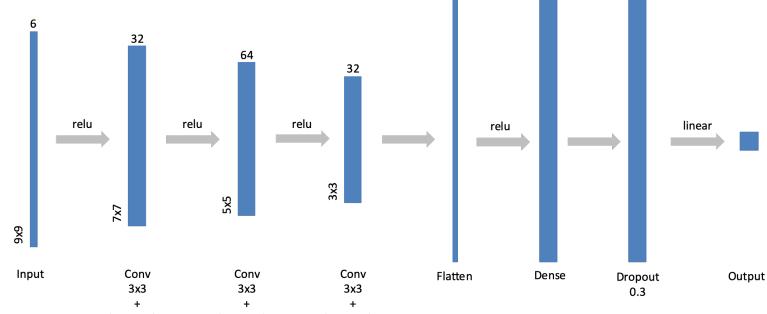


What kind of preprocessing is needed for the data sets?

No additional preprocessing is required



# What kind of CNN architecture can be used for SDB?

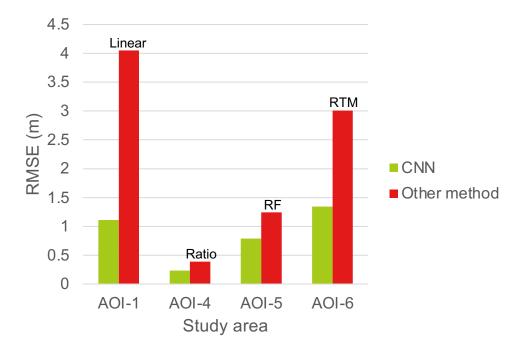


Batch normalization Batch normalization Batch normalization



#### What is the accuracy of the method?

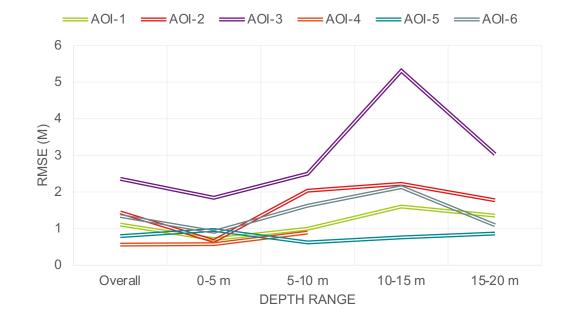
#### Compared to other methods





#### What is the accuracy of the method?

#### In different depth ranges and locations





To what extent can the pretrained model be reused?

- Should consider coastal water characteristics
- Combining multiple training data in different locations increase the variety, but need to balance the distribution
- Still need ground truth data to assess the accuracy



## Future work

Balancing the training data

Verification using another data source

SDB in the colder water and freshwater



### Thank you

