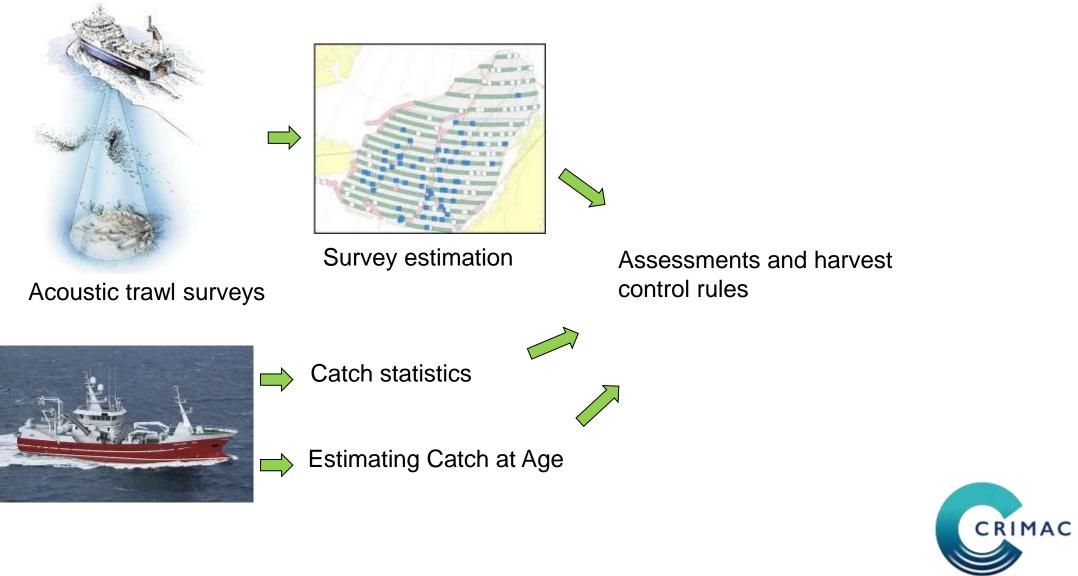


Fisheries acoustics and deep learning

Nils Olav Handegard^{*} Leif Edvard Bildøy ^{\$} Olav Brautaset [&] Tomasz Furmanek^{*} Arne Johannes Holmin^{*} Ingrid Utseth [&] Ketil Malde^{*}

*Institute of Marine Research *Kongsberg Discovery &Norwegian Computing Centre

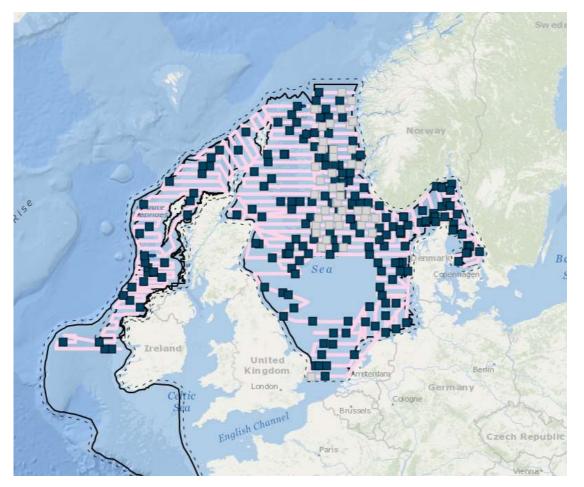
Fisheries advice based on acoustics



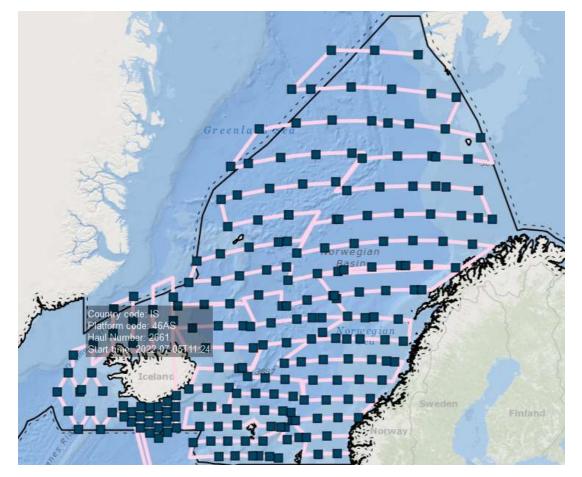


Fisheries advice based on acoustics

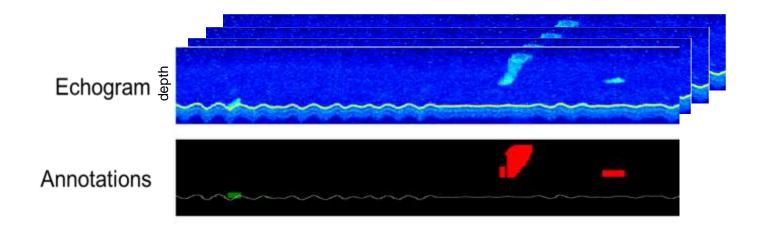
North Sea Herring and Pelagic Ecosystem Survey



International ecosystem summer survey in the Nordic Seas

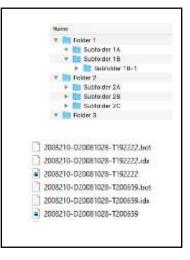


Acoustic Target Classification (ATC)

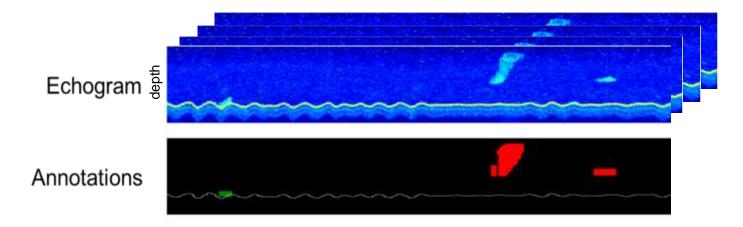






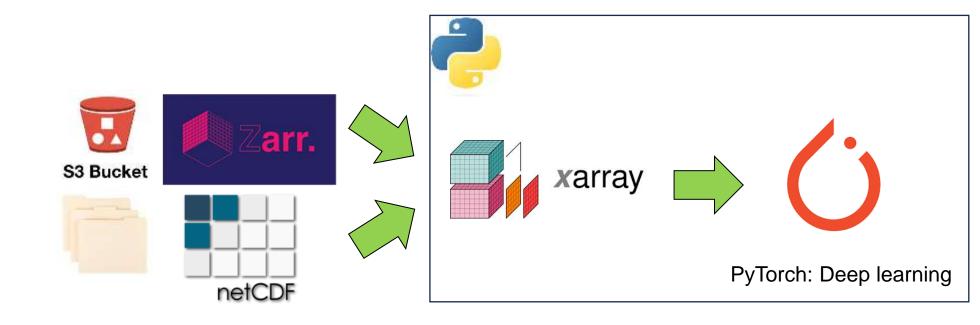


Annotated historical data from ~295 surveys



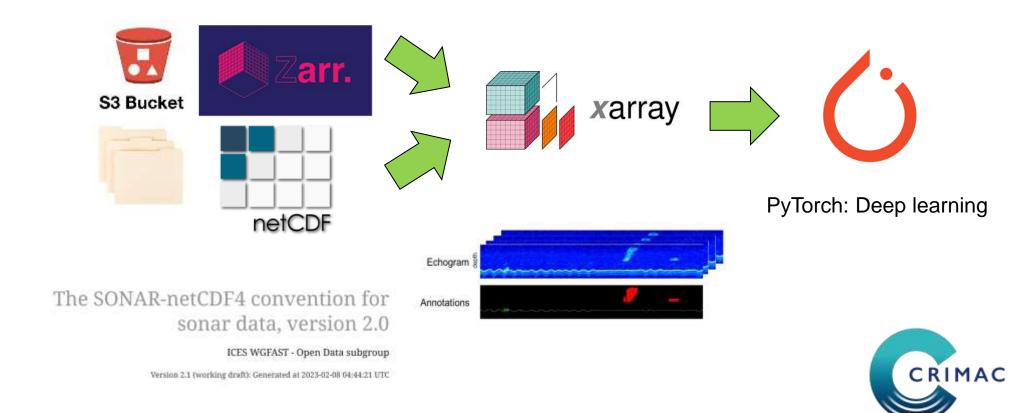




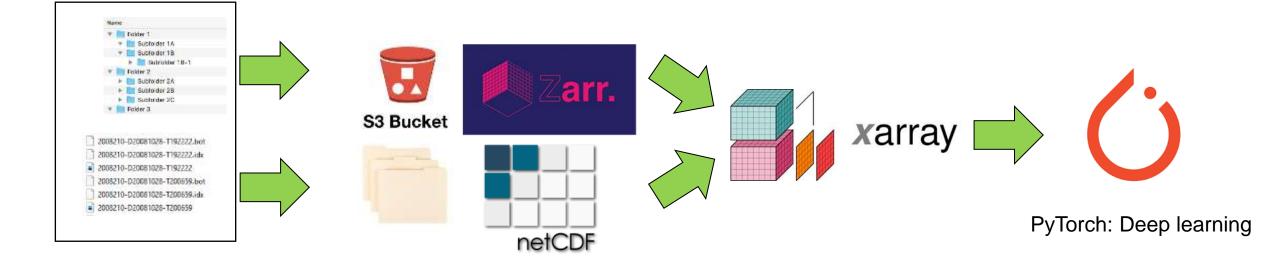














Applying calibration, missing NMEA information, logdistance errors, aligning frequencies, folder structure changing over time, proprietary data formats changing over time, extracting labels; formats changing over time, +++









2012

ImageNet Classification with Deep Convolutional Neural Networks

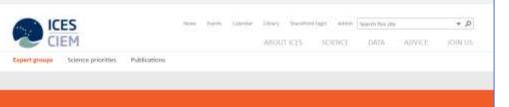
Alex Krizhevsky Ilva Sutskever Geoffrey E. Hinton University of Toronto University of Toronto University of Toronto kriz@cs.utoronto.ca ilya@cs.utoronto.ca hinton@cs.utoronto.ca mite container ship motor scooter leopard container ship mite motor scooter eopard black widow lifeboat go-kart jagua cockroach amphibian moped cheetah fireboat snow leopard tick nper car starfish drilling platform golfcart Egyptian cat mushroom cherry Madagascar cat grille dalmatian squirrel monkey convertible agaric spider monkey grille mushroom grape pickup jelly fungus elderberry tit beach wagon gill fungus (fordshire bullterrier indri fire engine dead-man's-fingers currant howler monkey





Ketil Malde (1) 1.2*, Nils Olav Handegard¹, Line Eikvil³, and Arnt-Børre Salberg³

Institute of Marine Research, Bergen, Norway ²Department of informatics, University of Bergen, Norway ¹Norwegian Computing Center, Oslo, Norway



WGMLEARN

Working Group on Machine Learning in Marine Science LINKS Machine learning designates the statistical methods that learn to perform a task (classify > View of members of this group elements, predict values, etc.) by interring properties and patterns from a set of known data (training set) and then can perform the same task to provide predictions on a new set of data > DETSG EGRESCHUTTENS (test set). These methods have the potential to considerably accelerate the processing of the > View latest WOAR FARB report mattive amount of data collected in marine science, particularly for the fisheries science process. > View WOMLEARN pooler The objectives WGMLEARN are to: · review the current applications of machine learning in marine science as well as the

- new developments in machine learning that would be of potential interest. · identify key challenges
- . Inform data collection and storage to enable the use of machine learning on marine data archives
- · identify trends and future needs, to promote the use of misvant machine learning technologies.



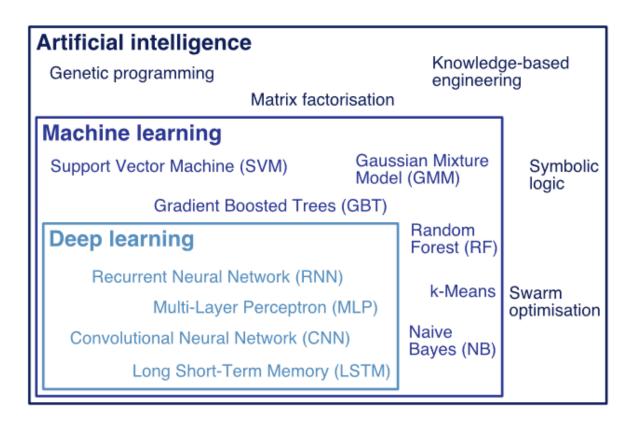
GO TO SHAREPOINT SITE

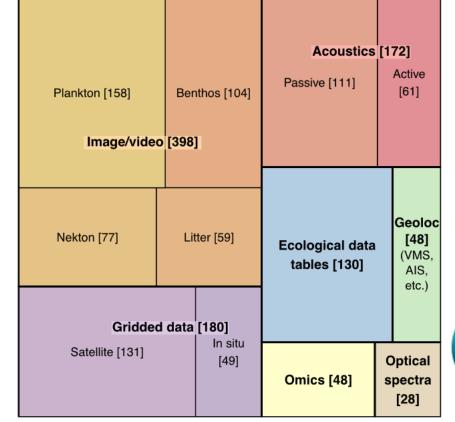
CONTACT US

ICES Journal of Marine Science, 2023, 0, 1–25 DOI: 10.1093/icesjms/fsad100 Review Article



Machine learning in marine ecology: an overview of techniques and applications



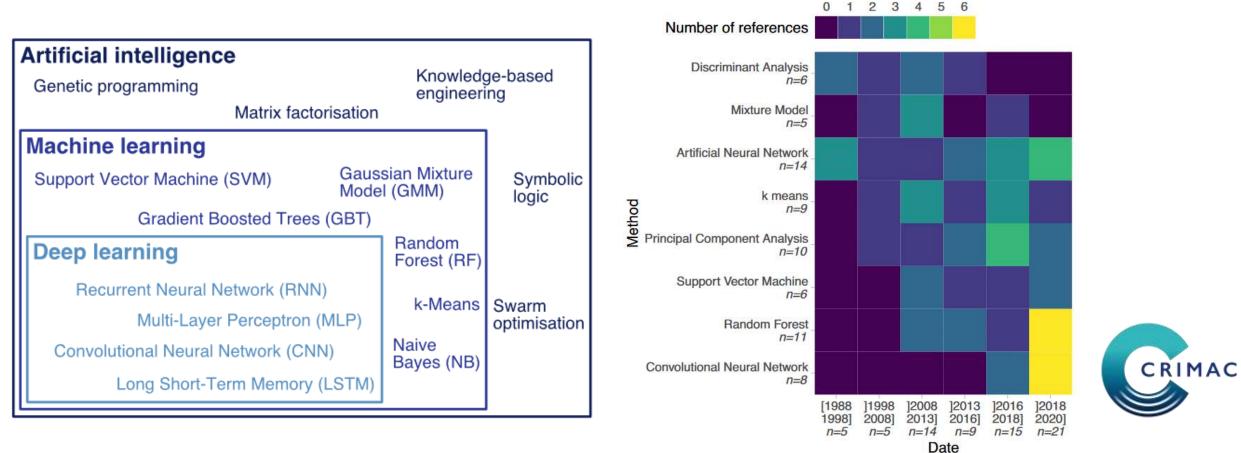




ICES Journal of Marine Science, 2023, 0, 1–25 DOI: 10.1093/icesjms/fsad100 Review Article



Machine learning in marine ecology: an overview of techniques and applications



Deep learning: The U-net

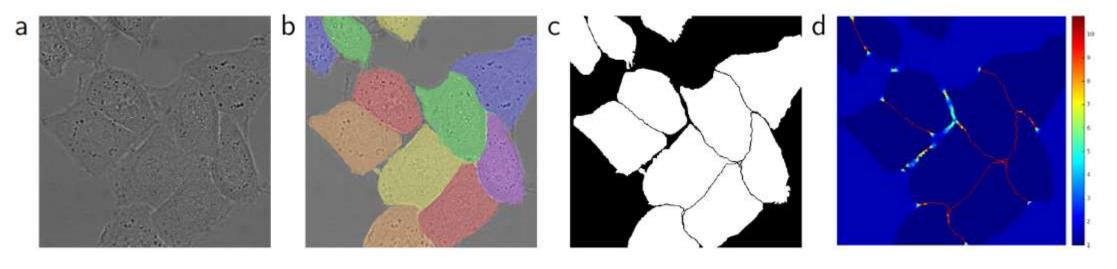


Image data

Labels

Predictions

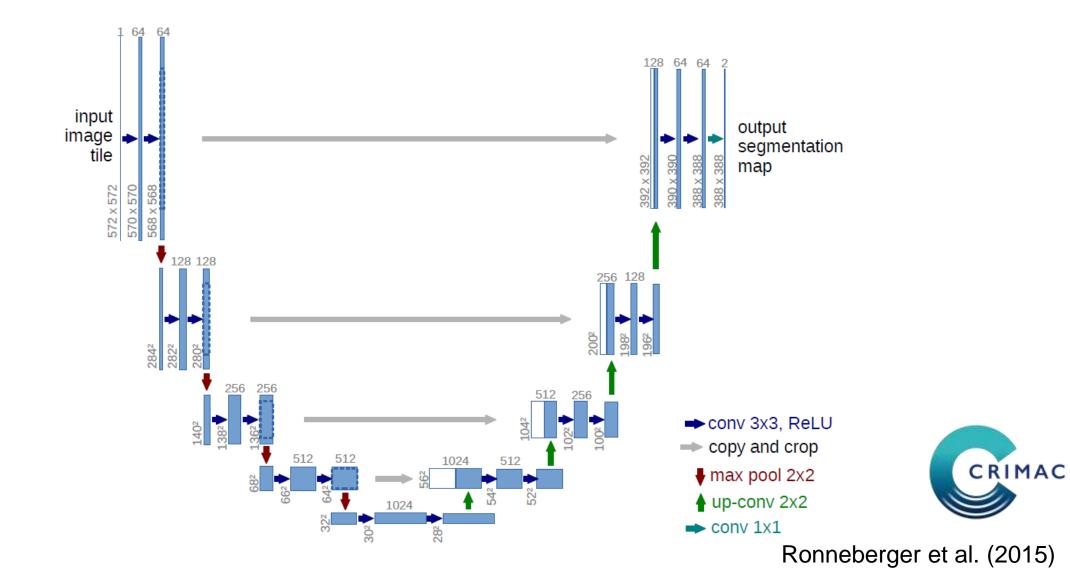
Loss weights





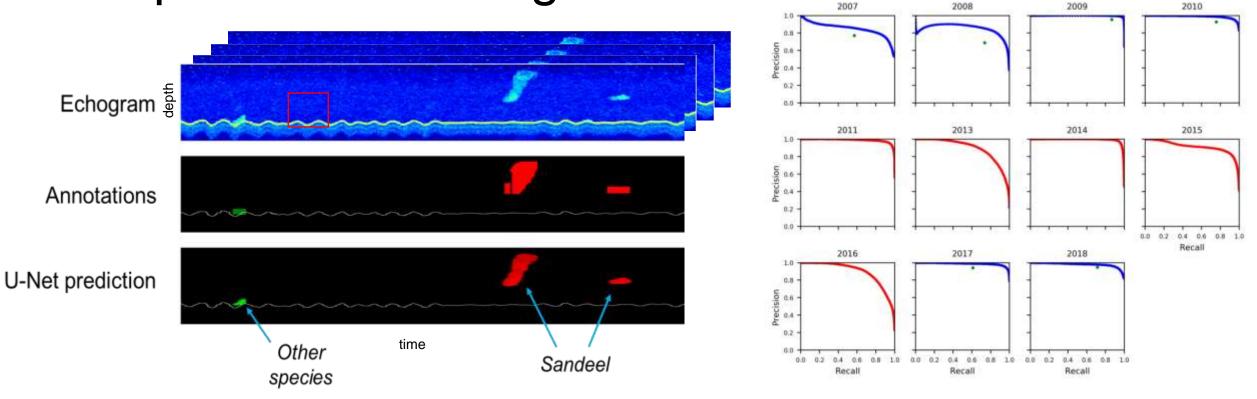
Ronneberger et al. (2015)

Deep learning: The U-net





Acoustic Target Classification (ATC) -Supervised learning

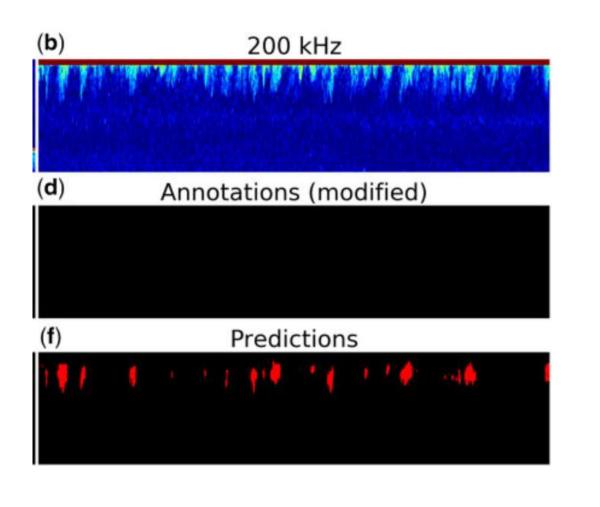






Brautaset et al, 2020

ATC – highly unbalanced data sets

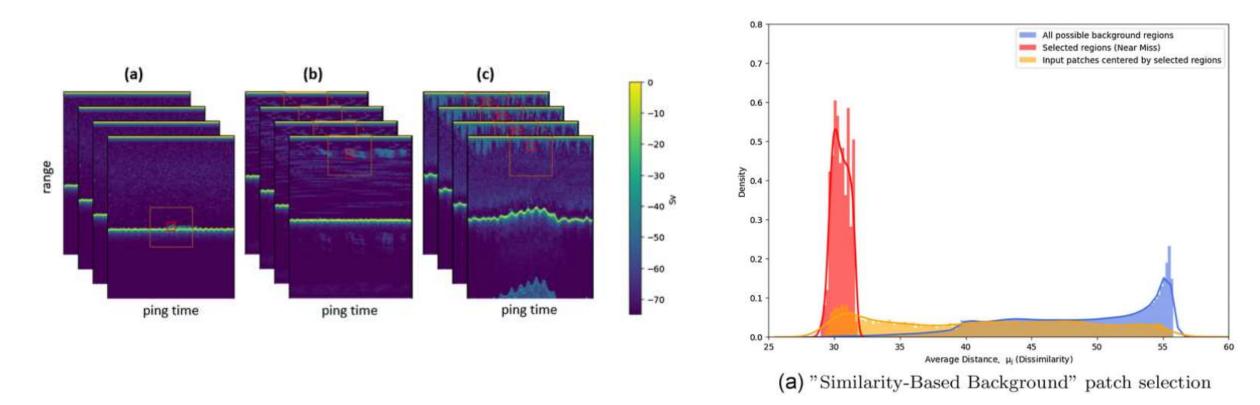






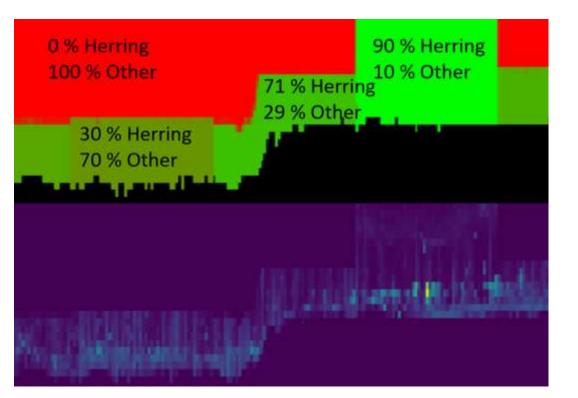
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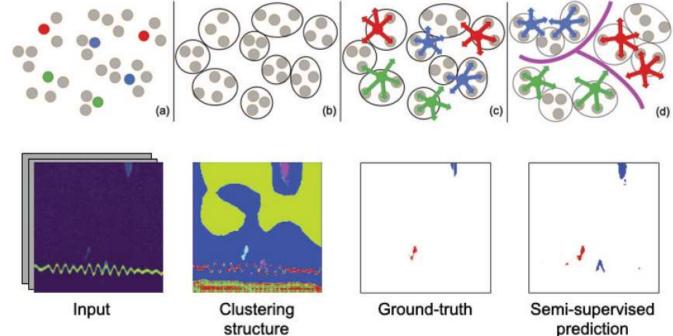
ATC – highly unbalanced data sets





Machine learning and deep learning models



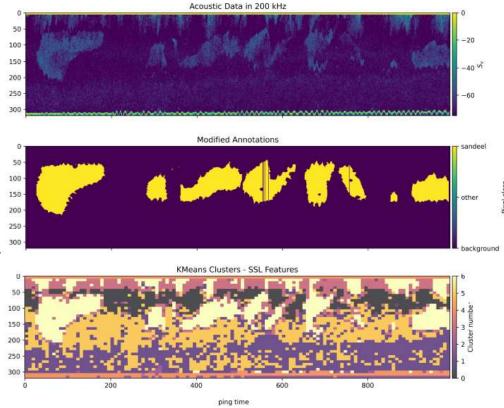




Brautaset et al. (in prep) COMPANY SHARED



Machine learning and deep learning models



Self-Supervised Learning for Annotation-Free Analysis of Acoustic Data

Ahmet Pala¹, Anna Oleynik¹, Ketil Malde², and Nils Olav Handegard²

¹Department of Mathematics, University of Bergen, Allegaten 41, 5008 Bergen, Norway ²Institute of Marine Research, Nordnesgaten 50, Bergen 5005, Norway

March 4, 2024





Machine Learning Engineering for Production

Model-centric view

Take the data you have, and develop a model that does as well as possible on it.

Data-centric view



The quality of the data is paramount. Use tools to improve the data quality; this will allow multiple models to do well.



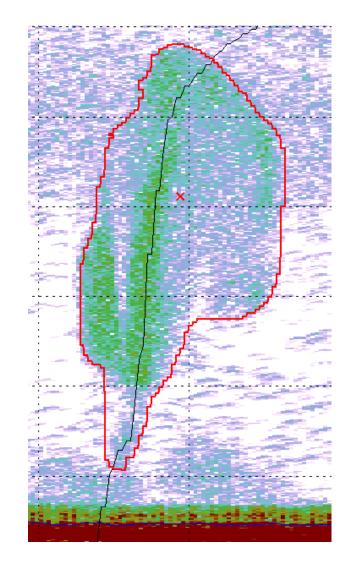
#1 Machine Learning Engineering for Production (MLOps) Specialization [Course 1, Week 1, Lesson 1]

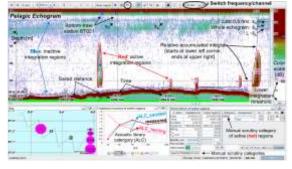




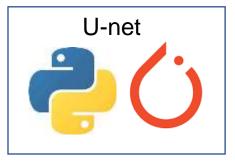
(Photo by Steve Jennings/Getty Images for TechCrunch)

Machine Learning Engineering for Production







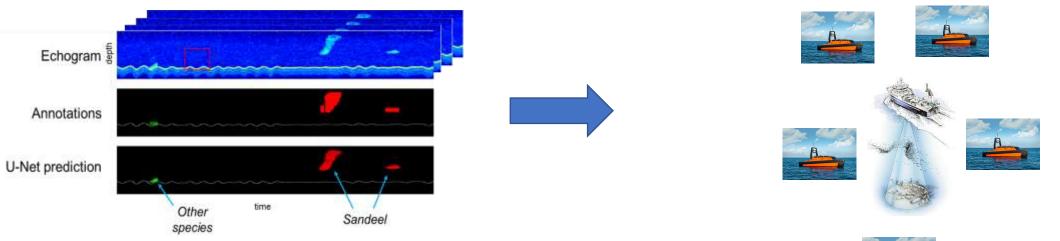






Machine Learning Engineering for Production

Not straight forward to run traditional processing on the platforms (edge)



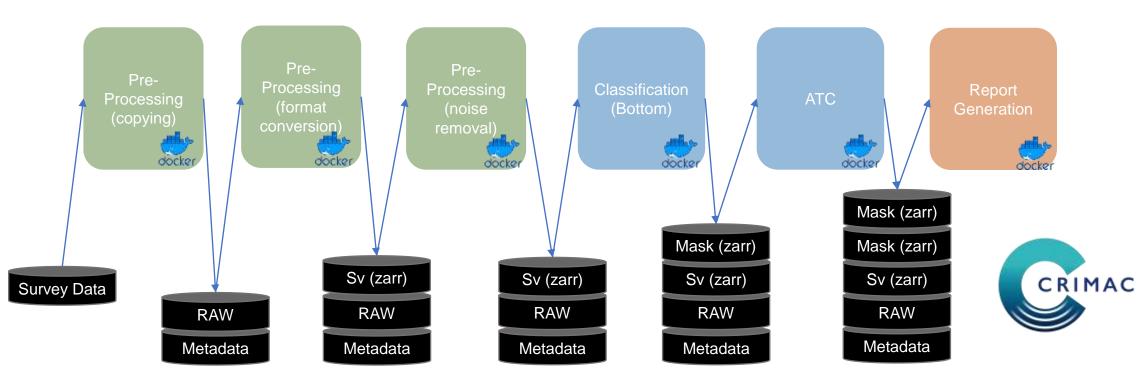


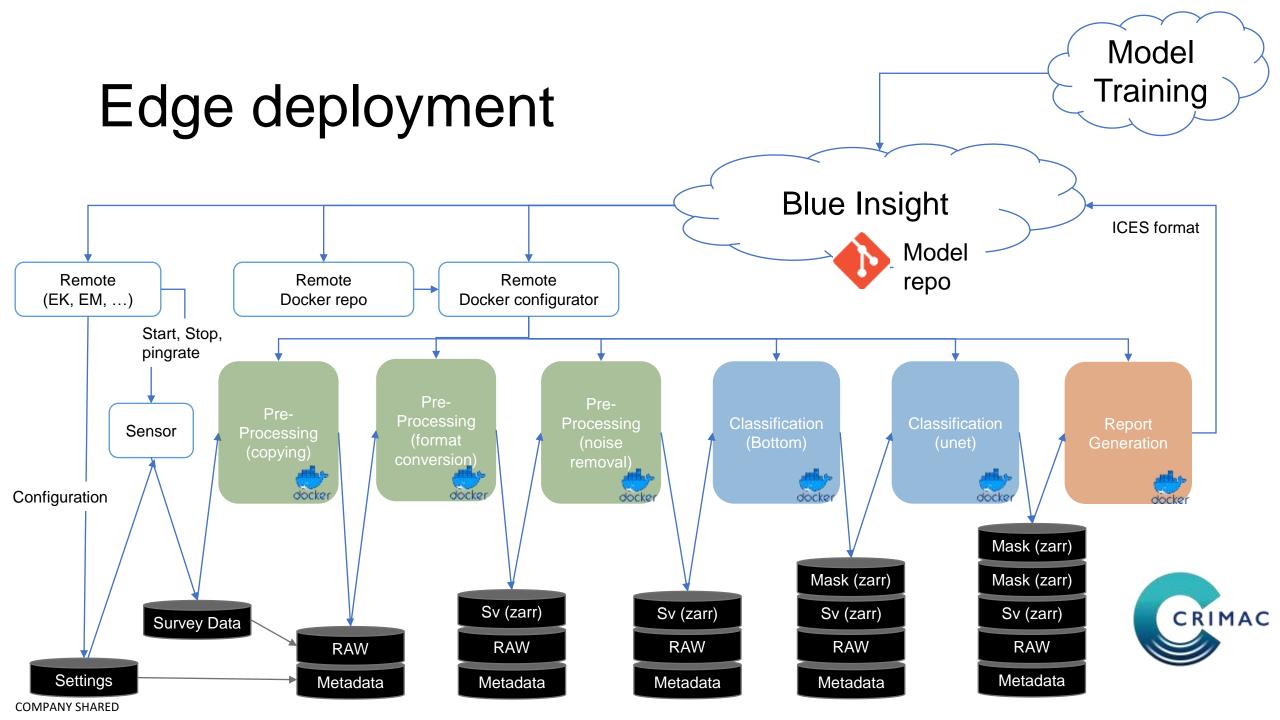




Edge deployment









"Well, Mr. Frankel, who started this program, began to suffer from the computer disease that anybody who works with computers now knows about. It's a very serious disease and it interferes completely with the work. The trouble with computers is you *play* with them. They are so wonderful." (Surely You're Joking, Mr. Feynman!: Adventures of a Curious Character"?)



Funding: CRIMAC, Norwegian Research Council, under the SFI program