Are plankton nets a thing of the past? How we can use AI for rapid plankton and ecosystem assessments

Sophie Pitois

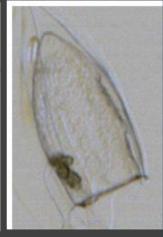
PICES – 2024 Oct 26 – Nov 1, 2024 Honolulu, USA













Centre for Environment Fisheries & Aquaculture Science



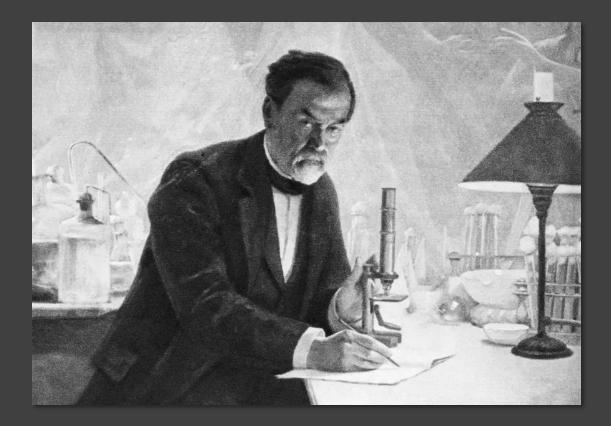
The Alan Turing Institute

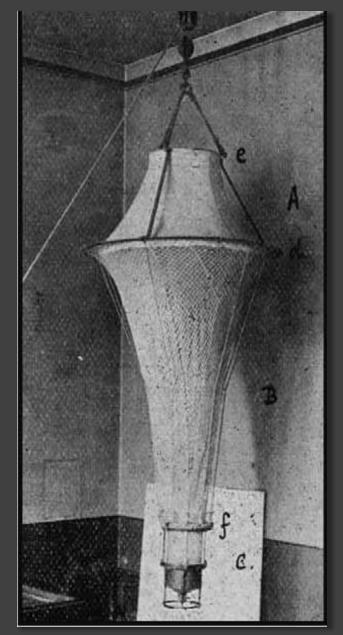




In the beginning.....

- Ring net first developed in 1880s
- Microscope analysis





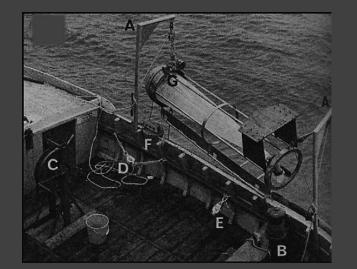
The first Hensen net developed in the 1880s (Wiebe and Benfield, 2003)

For the following 100 years....

- Nets evolved, to include high speed samplers etc...
- CPR appeared in late 1930s
- Microscopes became more performant



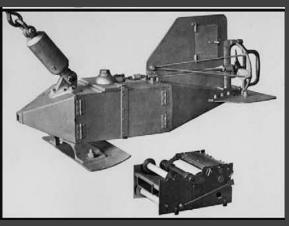
Deployment of ring net in the 1960s and 2000s



Gulf high-speed sampler



Sir Alistair Hardy looking through microscope (photo from CPR survey)



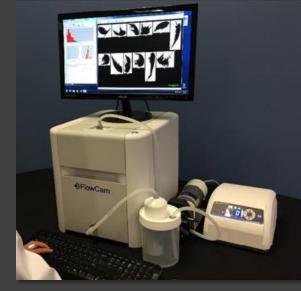
CPR in 1936 (Wiebe and Benfield, 2003)



Scientist looking through microscope (2015)

Today...

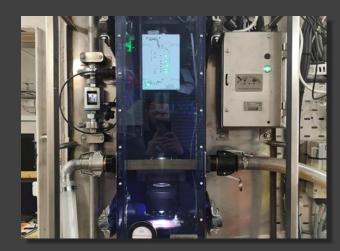




FlowCam (Fluid Imaging Techonlogies,



ZooSCAN (Hydroptic)



Plankton Imager PI-10 (Plankton Analytics)



Video Plankton Recorder (VPR)



In Situ Ichthyoplankton Imaging system (ISIIS)

Review: Lombard et al., 2019

Underwater Vision Profile (UVP)

The Plankton Imager (PI): an all-in-one tool that combine zooplankton sampling and image analysis







The Plankton Imager

The PI is a **high-speed line-scan** camera that **images all particles continuously** in a through-flow sampling system

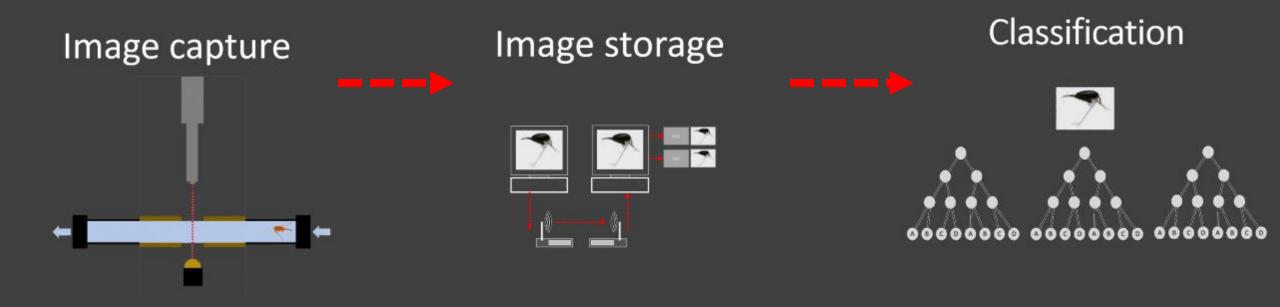


Image capture

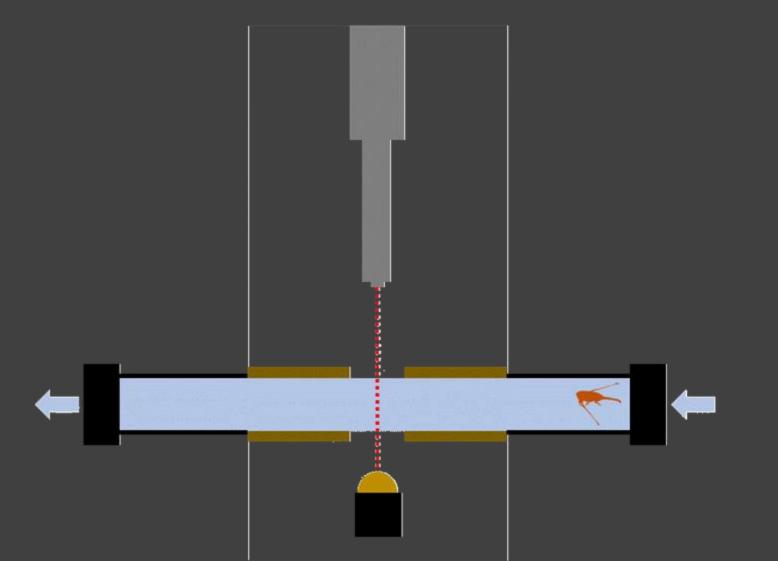
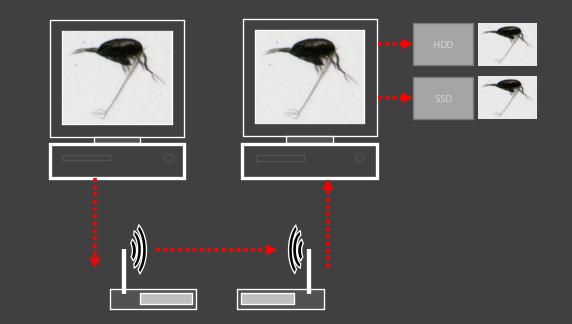
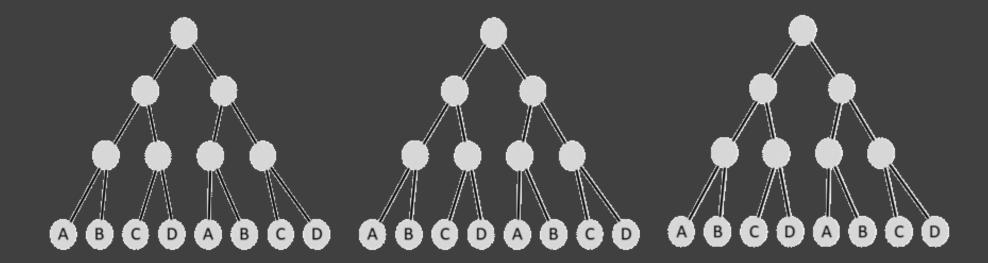


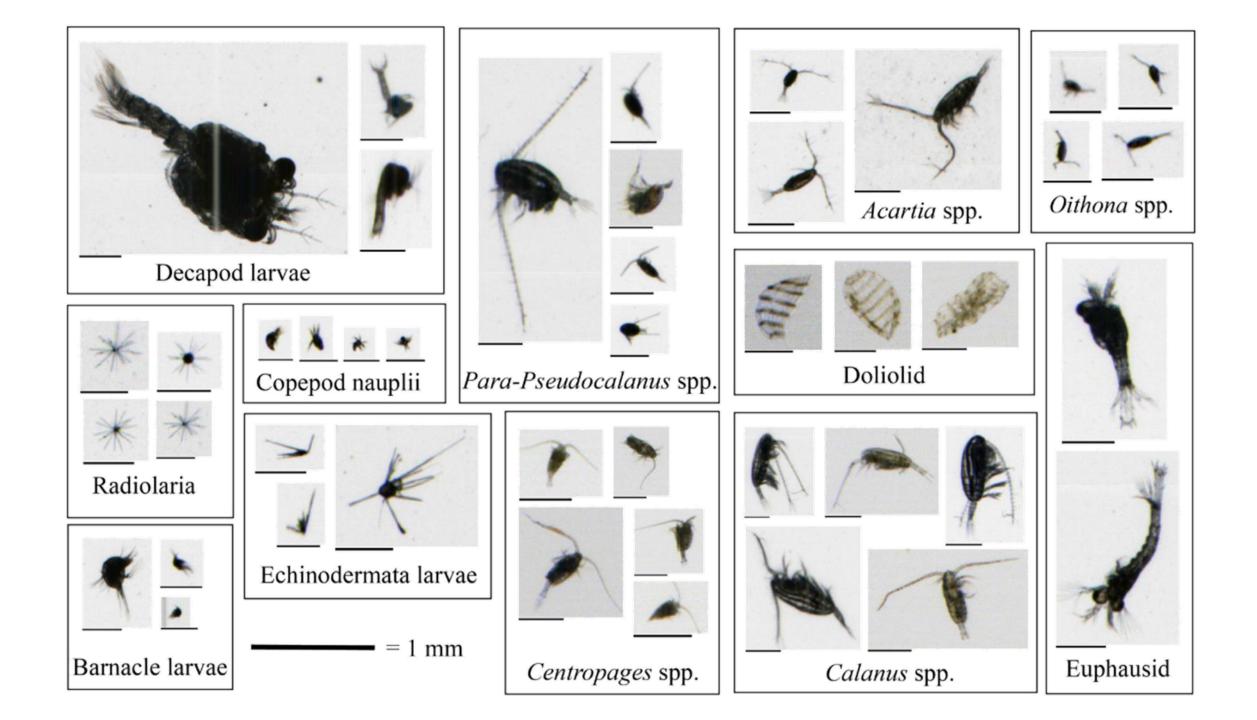
Image storage



Classification





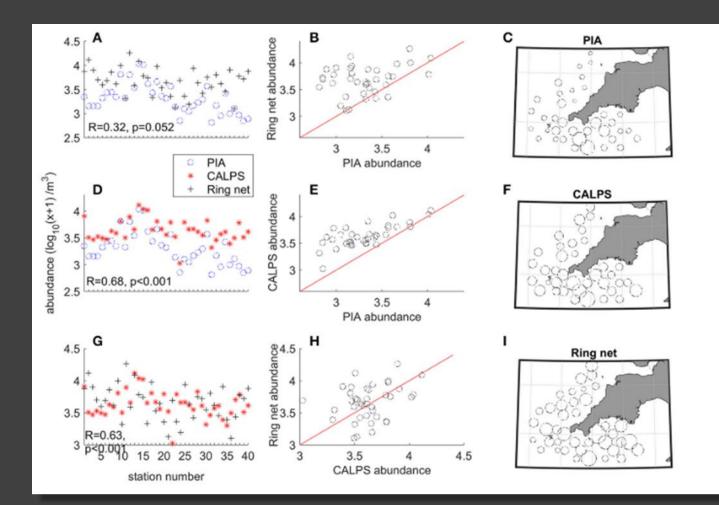


Method in numbers

- Continuous sampling in all weather, no 'down-time'
- Camera resolution = 10ym
- Current size range for recorded particles is 0.20 mm 3.5 cm (changeable)
- Tiff metadata includes GPS and ships time
- Survey 2024: 7 days = 3 billion particles captured, 128 million images saved.

Proof of concept

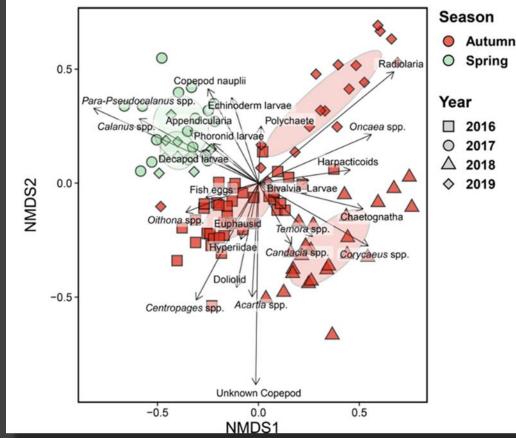
- The PI provide a robust description of zooplankton
- Comparison across 3 methods
- Differences in catchability but similar patterns of distribution



Pitois et al. 2018

Proof of concept

- the PI can be used as part of zooplankton monitoring
- Aligns well with CPR and PML L4 station:
 - similar variability in abundances between years and seasons
 - Similar variability in community structures between spring and autumn



Scott et al. 2021

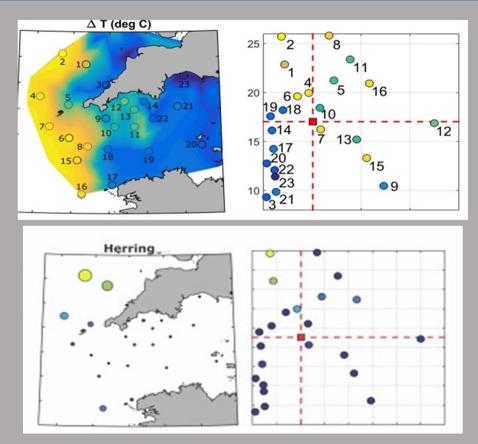


Proof of concept

 Application of in-situ size information to ecological indicators

 Colder waters characterised with low abundances of mainly small copepods

 Herring higher biomass located in areas of where copepods are larger but not necessary more abundant than average.



Link 2D indicator with physical parameter, lower and higher trophic levels (pelagic fish)

Challenges to overcome

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Need robust and quick method to identify all images

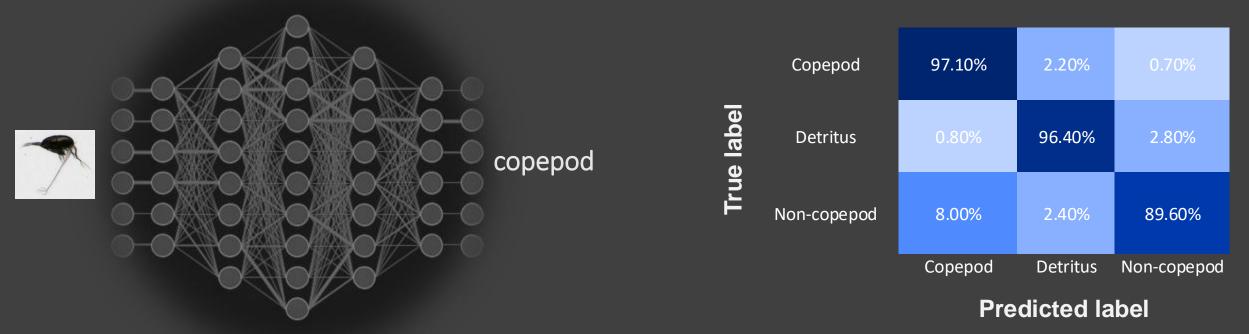
Need new methods for data intensive plankton research

Need robust and quick method to identify all images

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- •Imbalance in the training set
- Building Training set is Resource intensive
- Manual validation required

Alan Turing Quick and robust method to identify all images

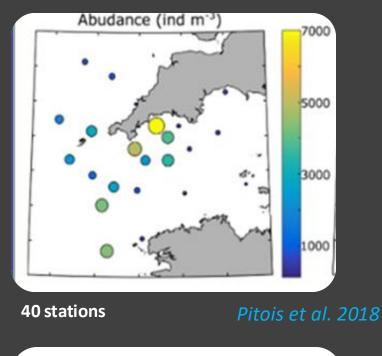


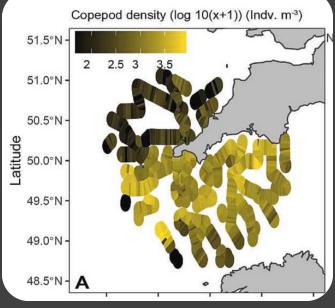
Can identity all images fast and reliably but Low taxonomic resolution

Data Study Group Final Report: doi.org/10.5281/zenodo.6799166

From low to high spatial resolution

- From 40 to 923 stations
- PI-10 station = 10 min bin of images





Need new methods for data intensive plankton research

Need new methods for data intensive plankton research

• Millions of images collected and analysed

• Resource intensive

 Data collection rate still faster than processing rate

Intensive plankton research

Real-time data pipeline



Edge AI - e.g., NVIDIA Jetson AGX Orin running a classifier and summarising plankton data in near real time.

Ship's internet, bandwidth budget < 5K per minute

Terrestrial digital dashboard showing near-real time plankton abundance and distribution.

Regular data pipeline





External storage captures data directly from instrument and uploaded to Azure



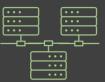
Data are classified in

learning workspace,

the classifier / new

classifier can be run multiple times

Azure Machine



Classification results returned to local servers for analysis

Cloud based

Edge-Al



Intensive plankton research

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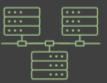


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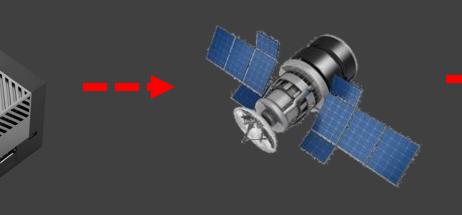


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The **Alan Turing** Cefas Institute

Real-time data pipeline



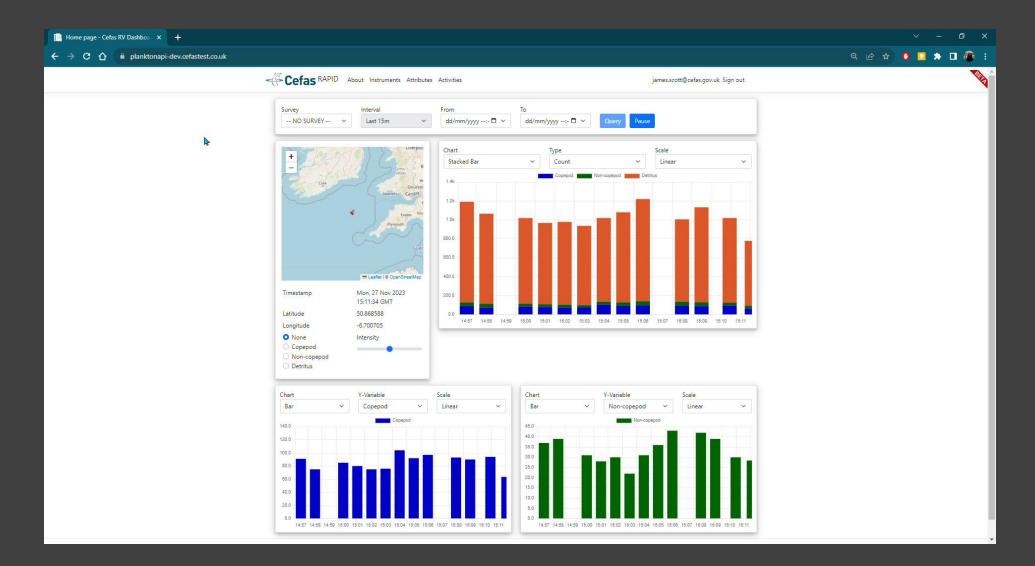
RAPID dashboard



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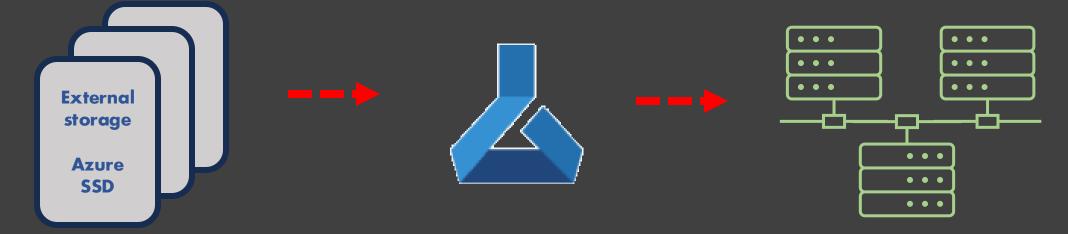
Home page - Cefas RV Dashboard (cefastest.co.uk)

RAPID dashboard



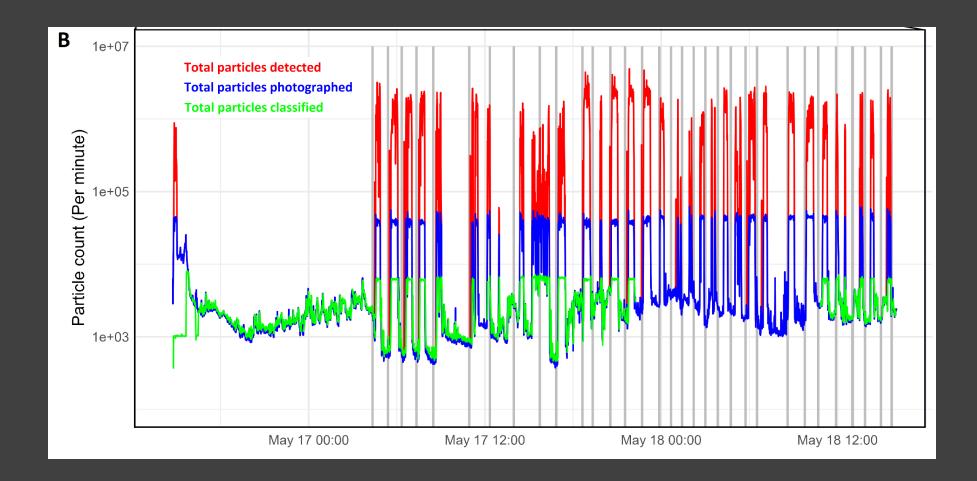
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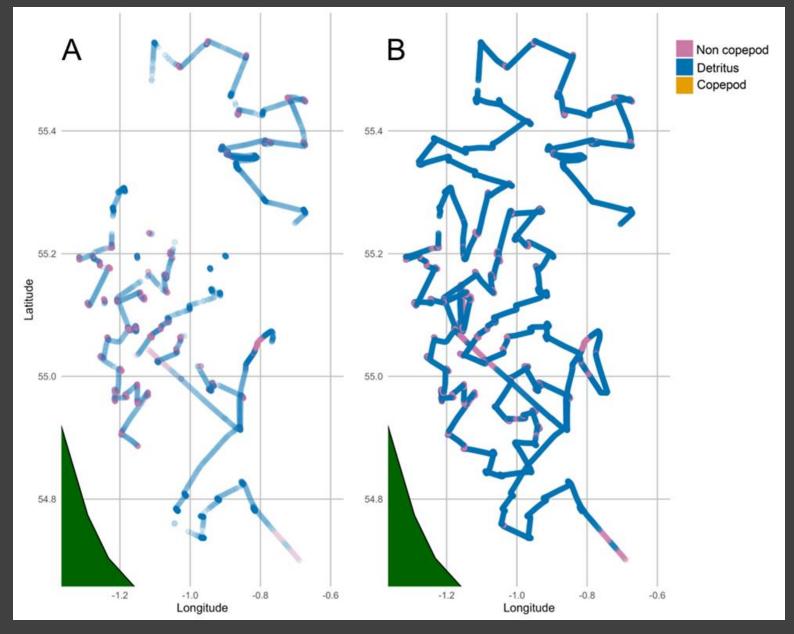
Cloud data pipeline



External storage captures data directly from instrument and uploaded to Azure Data are classified in Azure Machine learning workspace, the classifier / new classifier can be run multiple times Classification results returned to local servers for analysis

Edge AI vs Post survey computing

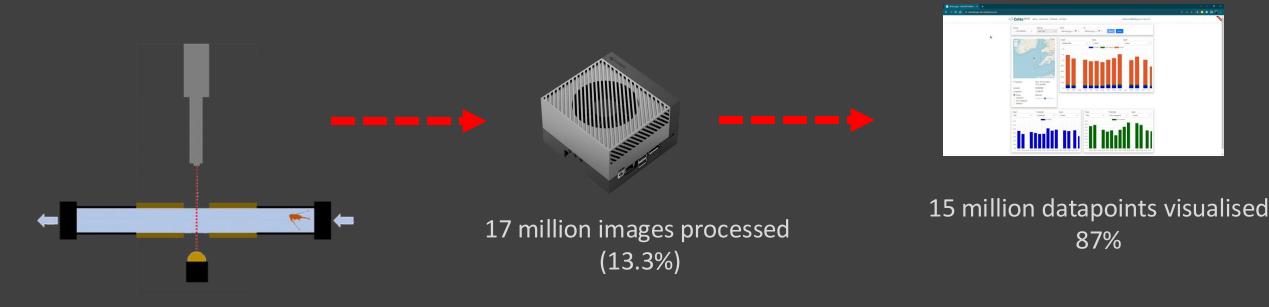




Real-time pipeline

Post-survey processing

Summary



3 billion particles captured 128 million images (4.8%)



128 million images processed (100%)

Edge Al	Cloud compute
Real-time or near real-time	Can take days / weeks to run survey
Low taxonomic resolution (broad groups)	Expanded algorithm: detailed taxonomic resolution
No computing cost	cloud computing cost
Images not saved	Archive for future ML improvements methods
processing limit (8,000 / min)	All data collected

Use of AI for future integrated monitoring

• Development of real time monitoring for zooplankton:

- Automated 24/7 collection of images
- Edge-AI for fast processing and data transfer
- Near real time visualisation of zooplankton abundances, size distribution etc...
- Adaptive sampling
- Cloud computing for in depth analysis

• Expand to other components of the pelagic system

- Physical parameters
- Phytoplankton
- Pelagic fish from acoustics

Automated Sampling

Continuous flow



Connected to ship water supply

Phytoplankton every 30-60 mins

Particle sizes and pigmentation



Live data for

Salinity
Fluorescence

Oxygen
Turbidity

Temperature



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Flowcytometer Phytoplankton

Connected to ship water supply

Phytoplankton every 30-60 mins

Particle sizes and pigmentation

Low reference Testing Concentrations of the second second

Number of particles/ml and total red and orange fluorescence/ml per size fraction



FerryBox Physical parameters

Live data for

- Temperature
- Salinity
- Fluorescence
- Oxygen
- Turbidity



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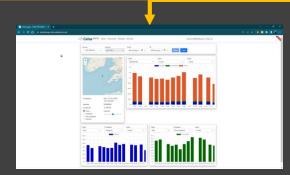
Oxygen
Turbidity



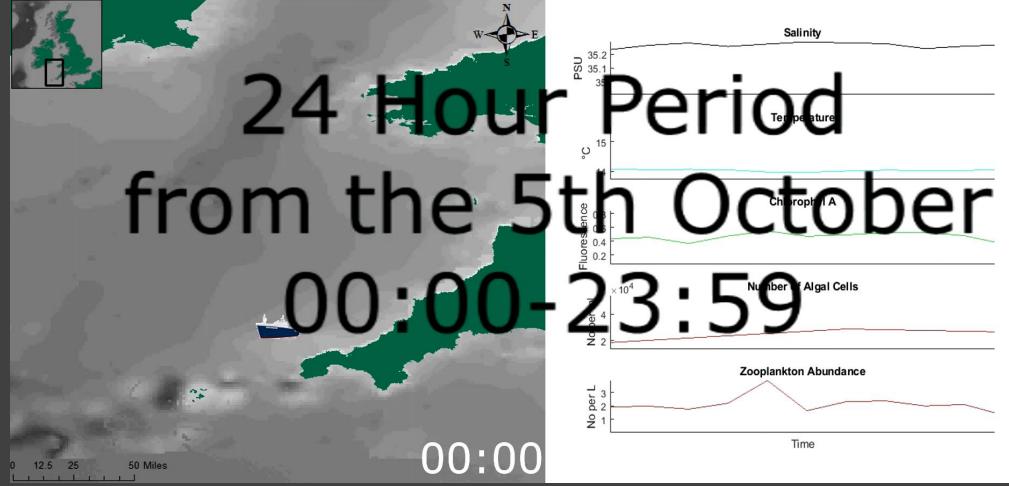
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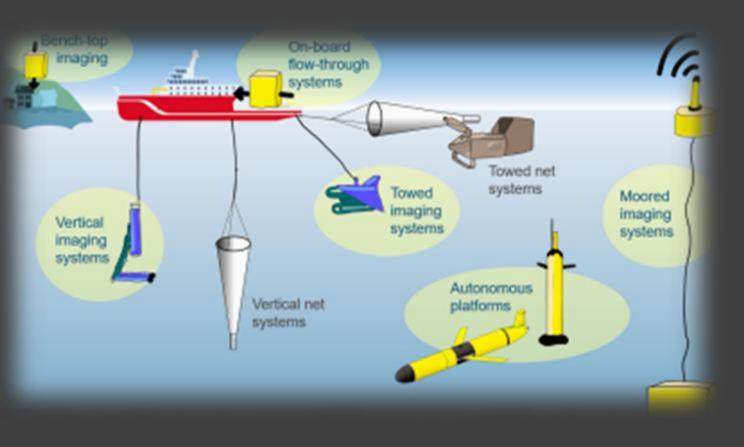






Al to change how we conduct monitoring

Many technologies

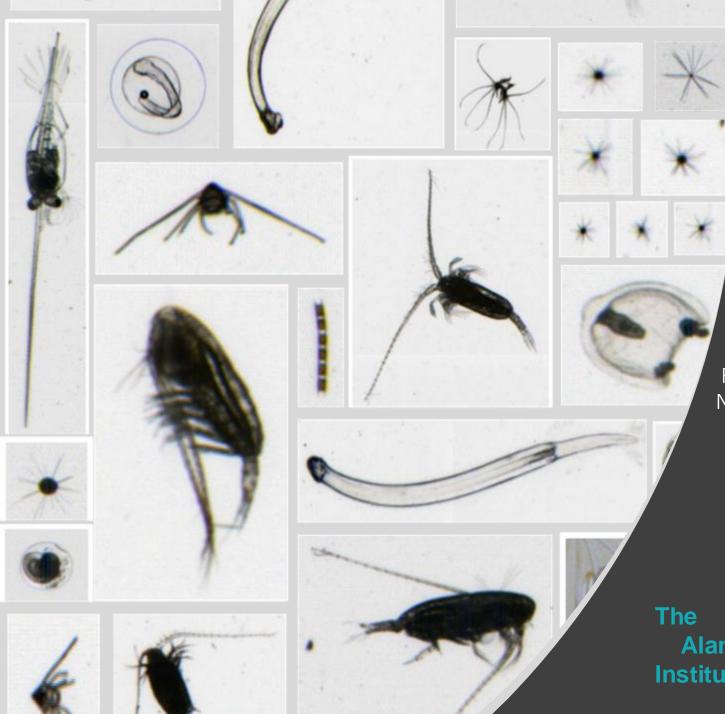


Big data analytics for multimodal data

Exponential rise in AI capabilities to provide new possibilities for fully integrated monitoring of the pelagic environment

Are nets a thing of the past?

Giering et al. Front. Mar. Sci., 2022



Thank you !

Joseph Ribeiro (Data scientist, Cefas) James Scott (Plankton ecologist, Cefas) Hayden Close (Marine scientist, Cefas) Eric Payne (Software engineer, Cefas) James Pettigrew (taxonomist, Cefas) Nevena Almeida (Taxonomist, Cefas)

Robert Blackwell (Data scientist, Alan Turing Institute) Noushin Eftekhari (Data scientist, Alan Turing Institute)

Sari Giering (National Oceanographic Centre) Mojtaba Masoudi (National Oceanographic Centre)





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