

Leveraging the power of Big Data at FAO

Applications in Fisheries and Aquaculture

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FAO

- **194 Member Countries**, two associate members and one member Organization
- Headquarters in Rome, Italy
- Presence in more than 130 countries



• FAO supports governments and their stakeholders in areas of development, in the design of adequate policies, programmes and legal frameworks to **promote food security and nutrition**



Our priorities





1. Eradicate hunger and malnutrition 2. Make agriculture, forestry and fisheries more productive and sustainable



3. Reduce rural poverty

4. Enable inclusive and efficient agricultural and food systems



5. Increase the resilience of livelihoods from disasters



Importance of statistics at FAO

Role of FAO:

- Collect, analyze, interpret and disseminate food & agriculture statistics
- Develop and implement methodologies, standards to help generate sound data
- Support for member countries: collection, dissemination, and uptake of data





Big Data in Fisheries and Aquaculture

Advantages – Limitations – Scope – Perspectives

FAO's Fisheries and Aquaculture Statistics

- Fisheries and Aquaculture as important source of food, nutrition, income and livelihoods
 - Marine and inland ecosystems and their resources under growing threat
 - Sustainability only possible with cautious and effective management
 - FAO is the only global source of fisheries and aquaculture statistics
- Our main databases:
 - Global capture and aquaculture production
 - Global trade of fisheries and aquaculture commodities
 - Consumption of Fish and Fishery Products

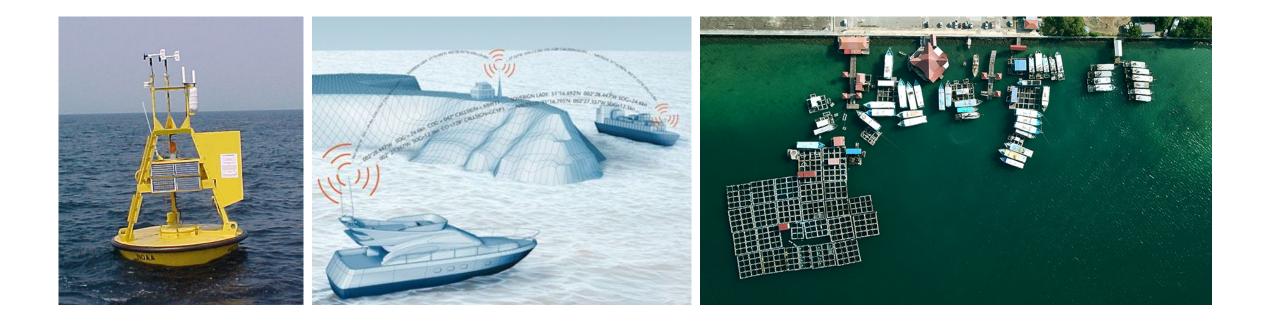


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Why Big Data in Fisheries and Aquaculture?

- Source of new data
- Alternative source to validate, complement, enhance existing datasets





Application 1 - species distribution

| Å | Goal | predicting future distribution of marine species |
|---|---------------|---|
| | Data sources | species occurrence data, marine environmental parameters (e.g. depth, temperature, salinity), habitat preferences |
| | Analysis type | ML niche modelling to compute future range under climate change scenarios |



Application 1 - species distribution

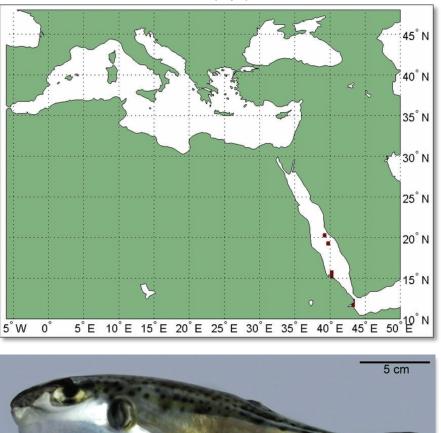


Results

- Silver-cheeked toadfish (Lagocephalus sceleratus)
- From Red Sea to Mediterranean Sea
- Without intervention, spread will continue and impact on fisheries will worsen

Example 7 Limitations

• Uncertainty of predictions unclear



Probability of species occurrence. 1950 -2050



Application 2 – AIS for fisheries monitoring

| Å | Goal | identify fishing effort location to understand impacts on environment and resources; improve fleet data |
|---|---------------|---|
| Þ | Data sources | global A utomatic Identification S ystem (AIS) data (60k vessels in 2017) |
| | Analysis type | machine learning to identify fishing gear based on movement |

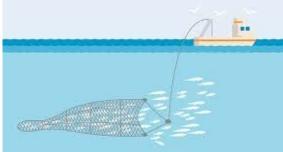




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Application 2 – AIS for fisheries monitoring Results example of output, west Africa (2017)









(Fishing hours/km2)





Application 2 – AIS for fisheries monitoring

Example 7 Limitations

AIS coverage number of vessels using AIS limited (mostly larger boats, richer countries, distant water fleet)

AIS receptionconstrained by presence of
satellites/antennae, heavy vessel traffic areas

AIS algorithm some fishing techniques are less predictable and therefore harder to identify than others (e.g. gillnets, pole and line)

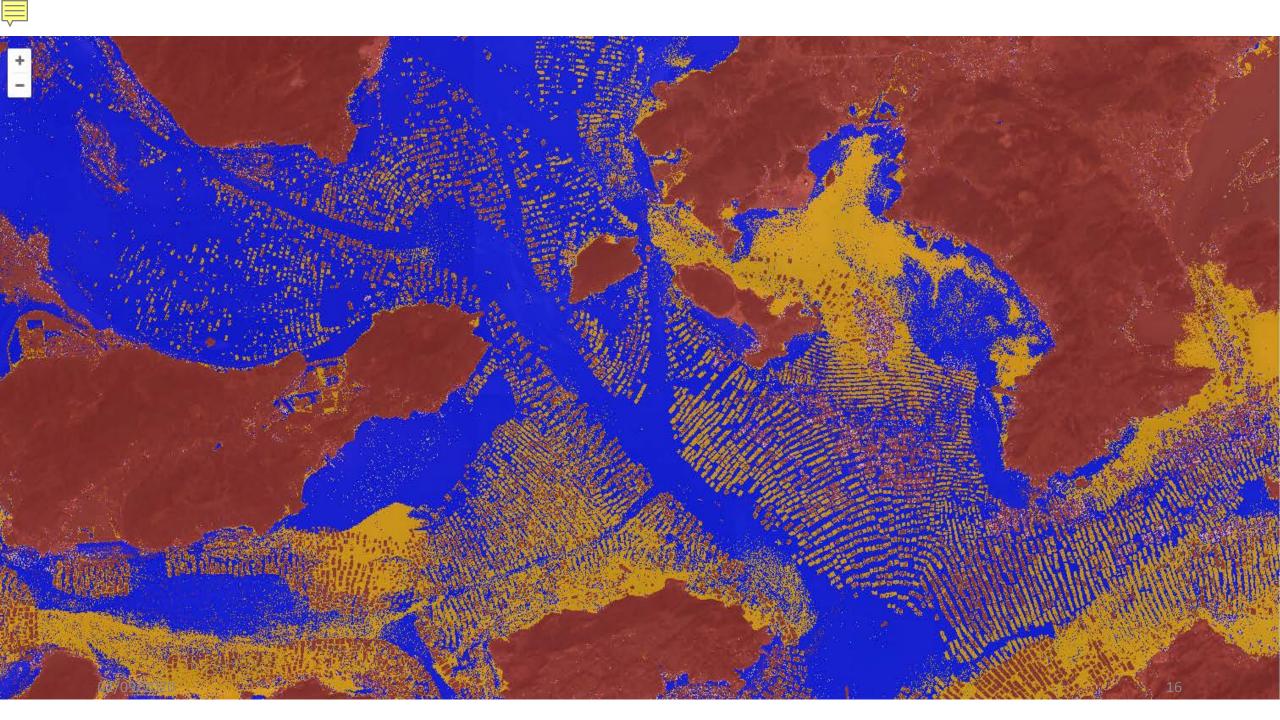




Application 3 – detection of aquaculture sites

detection and mapping of aquaculture sites for Goal improved information insights and production capacity analysis, spatial planning and potential disaster assessments Data sources satellite imagery (Sentinel II) **Analysis type** image classification algorithms Limitations imagery resolution (the better the more expensive), type of aquaculture, complex production calculation Result example in South-East China





Application 4 – SmartForms mobile app

| Å | Goal | decentralized collection of important but sparse data (e.g. bycatch, recreational catch, marine litter) |
|---|---------------|--|
| | Data sources | customizable forms designed to collect standardized data |
| | Analysis type | visualization of key data collection statistics |
| | Limitations | control over accuracy of data collection |
| Ô | Result | currently in beta version, release within months |

Application 4 – SmartForms mobile app

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Catch of the day

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Monitoring recreational fisheries in the Caribbean (Billfish project)

💷 SoFiRe

A test data collection App for Somali Fisheries Reporting

ByCatch-ABNJDeepSea

ABNJ Deep-seas Project

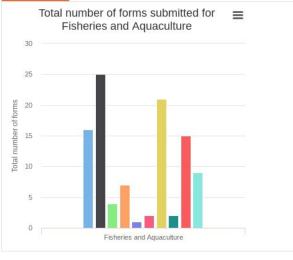
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ByCatch-ABNJDeepSea \leftarrow By-catch ABNJ Deep-seas Project Observer name Aureliano Gentile Vessel name Popeye IMO Ship Identification Number 5758908543 Trip Number 546fg Trip ID Yfg467i Tow number 2 Sharks Sharks Species Gulper shark + \triangleleft $^{\circ}$

Data Overview



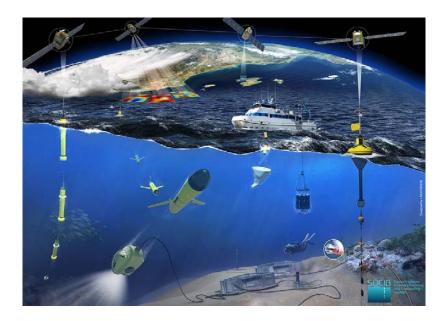
Data by Form Data by Workspace





Implementation in FAO Fisheries and Aquaculture

- Strategy on the use of Big Data under development
- Range of experimental projects
- Promising applications but no routine use of Big Data yet





Conclusion

- Very promising technology, but limitations exist
- Does not replace data collection by national statistical offices, but can be a very good complement
- Technology constantly improves, creating more and more opportunities (e.g. AIS use, satellite imagery resolution, machine learning algorithms)
- The future of fisheries and aquaculture will include these technologies and FAO is getting prepared to leverage them fully



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Annex – List of relevant publications and websites

- Advances in geographic information systems and remote sensing for fisheries and aquaculture
 - <u>http://www.fao.org/3/i3254e/i3254e.pdf</u>
- E-agriculture in Action: Big Data for Agriculture
 - <u>http://www.fao.org/e-agriculture/news/fao-itu-e-agriculture-action-big-data-agriculture</u>
- Forecasting the ongoing invasion of Lagocephalus sceleratus in the Mediterranean Sea
 - <u>https://www.sciencedirect.com/science/article/pii/S0304380018300164</u>
- Upcoming: Atlas of Fishing Activity using AIS data



Annex – List of relevant publications and websites

- FAO's Fisheries and Aquaculture statistics website: <u>http://www.fao.org/fishery/statistics/en</u>
- Global Fishing Watch website: https://globalfishingwatch.org/
- SmartForms: A mobile App platform to collect and review fishery and observer data:

http://www.fao.org/fi/static-

media/MeetingDocuments/cwp/ReferenceHarmonization/2018/S3 3
.pdf