

Quantifying the increase in fishing efficiency due to the use of drifting FADs equipped with echo-sounders in tropical tuna purse seine fisheries

David M. Kaplan, Gwenaëlle Wain, Lorelei Guéry, Daniel Gaertner

**IRD, MARBEC
Sète, France**

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Echosounder impacts on PS fishing

Fishing on a vessel's floating objects (FOBs) equipped with echosounder buoys can impact fishing in a number of ways:

- Reduce search time
- Increase catch
- Change catch composition
- Change fishing strategy

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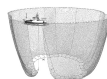
- Reduce search time
- **Increase catch**
- **Change catch composition**
- Change fishing strategy

Challenge → **Understanding FOB (\approx buoy) ownership**

Determining FOB ownership

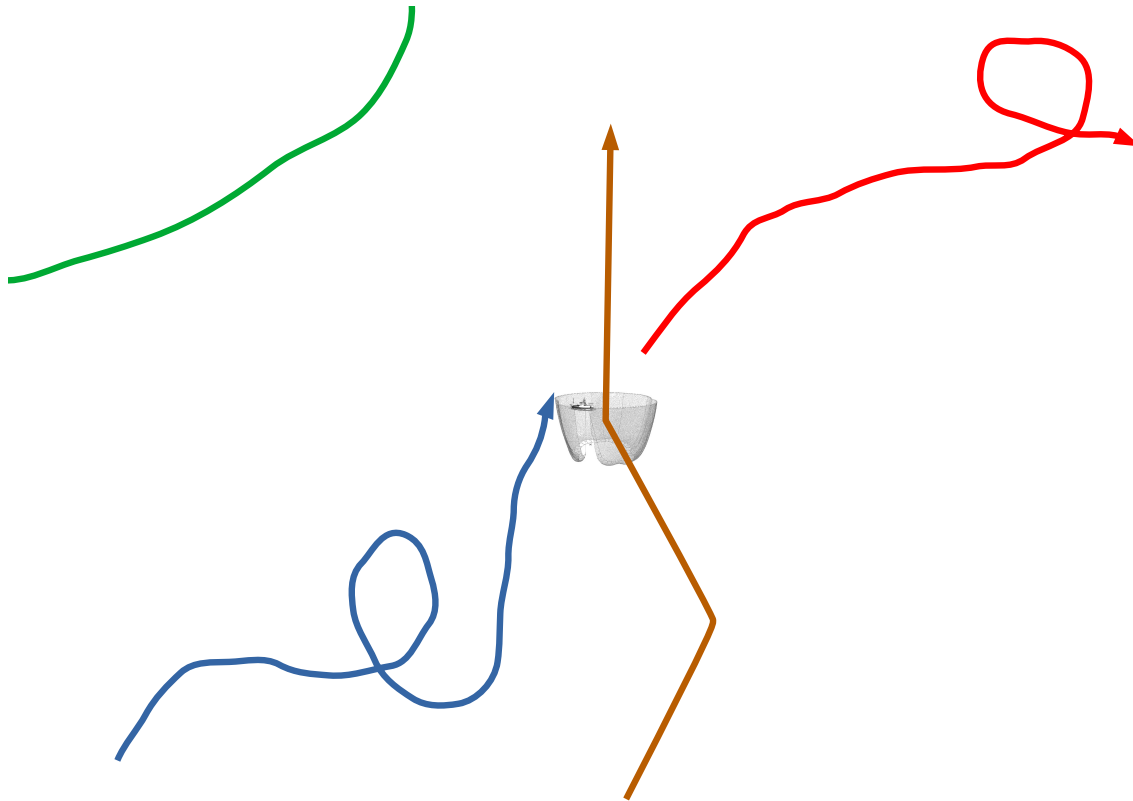
Determining FOB ownership

- Logbooks + FOB trajectories



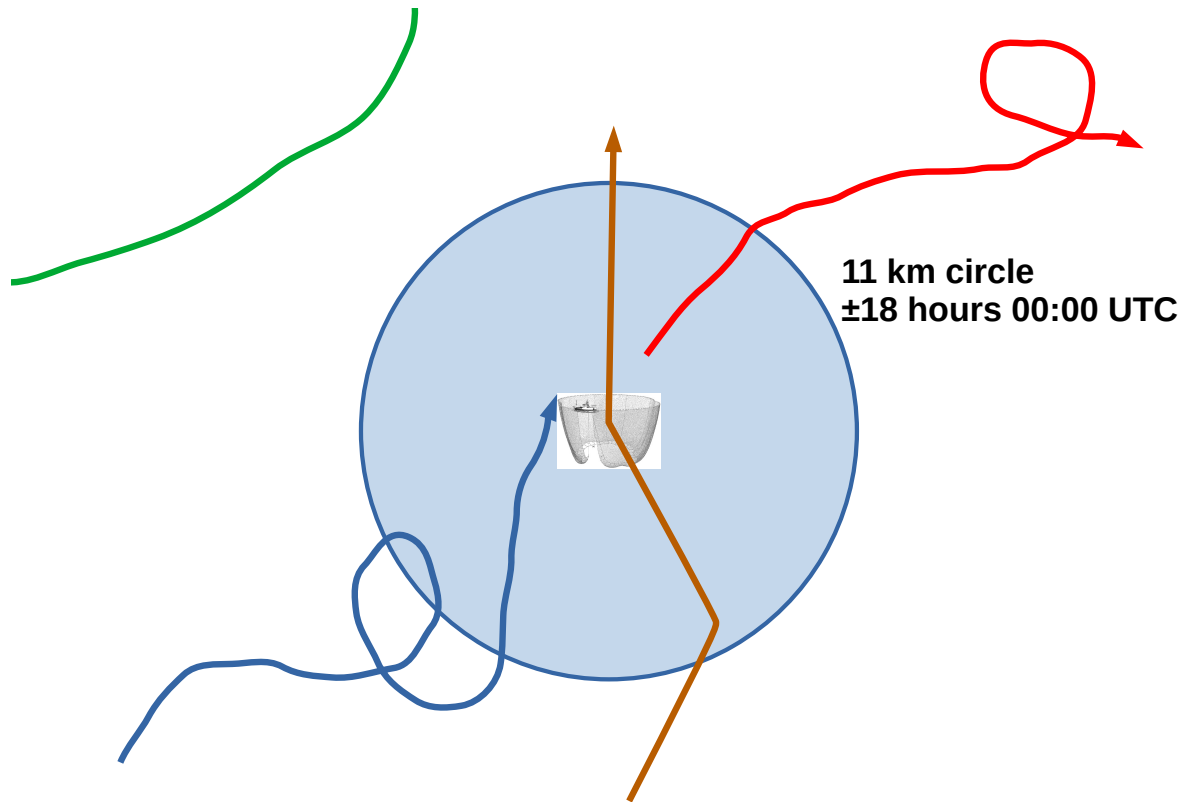
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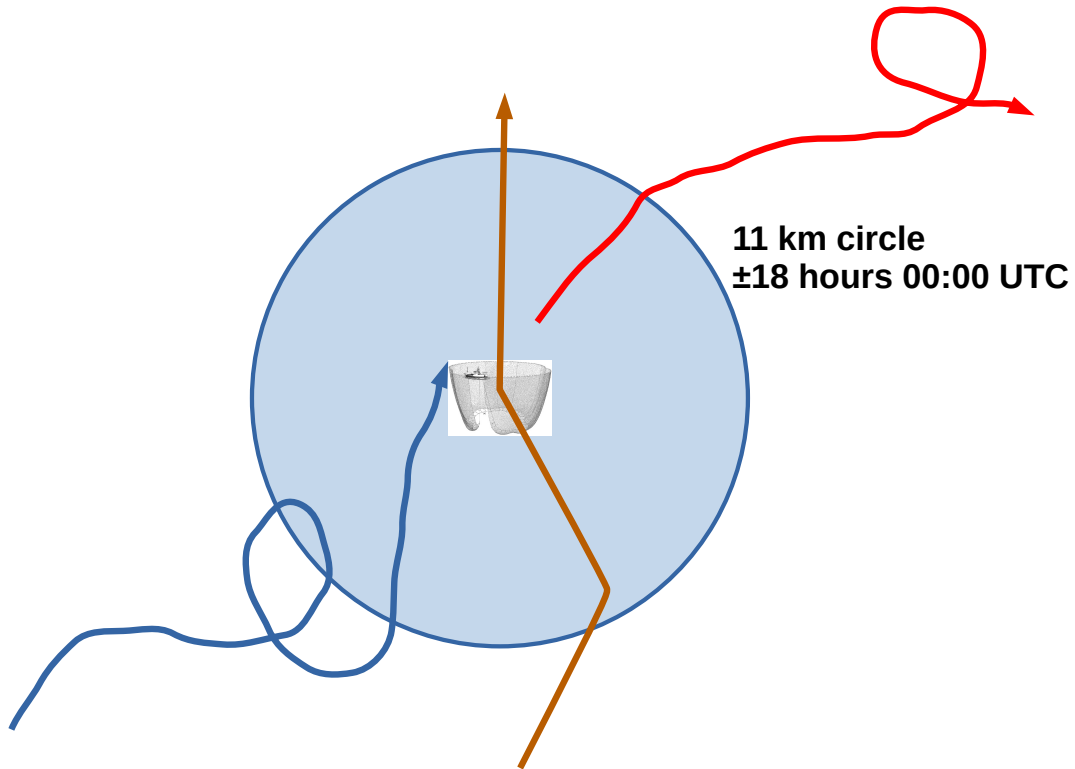
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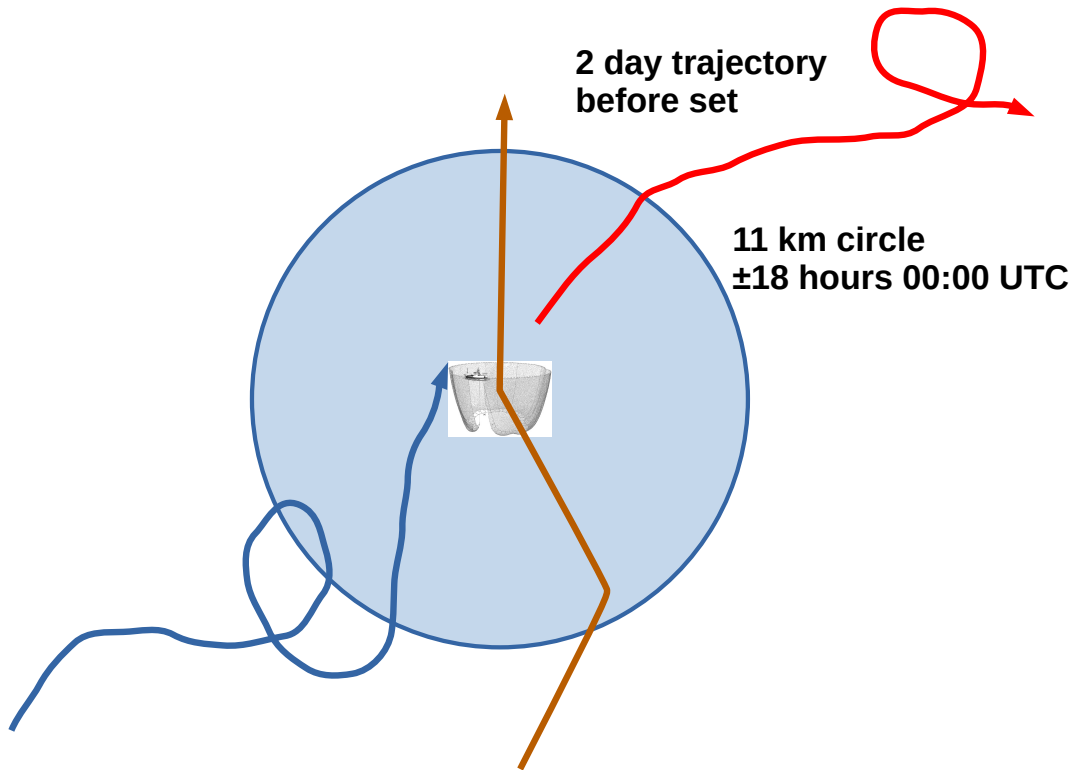
Determining FOB ownership

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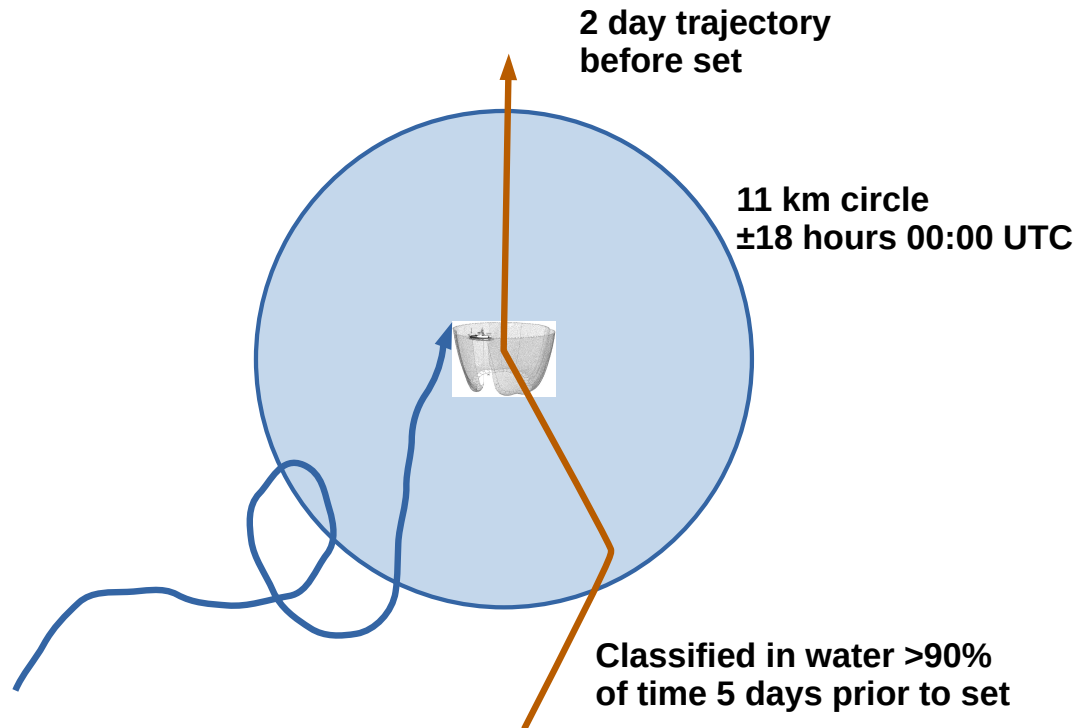
Determining FOB ownership

- Logbooks + FOB trajectories



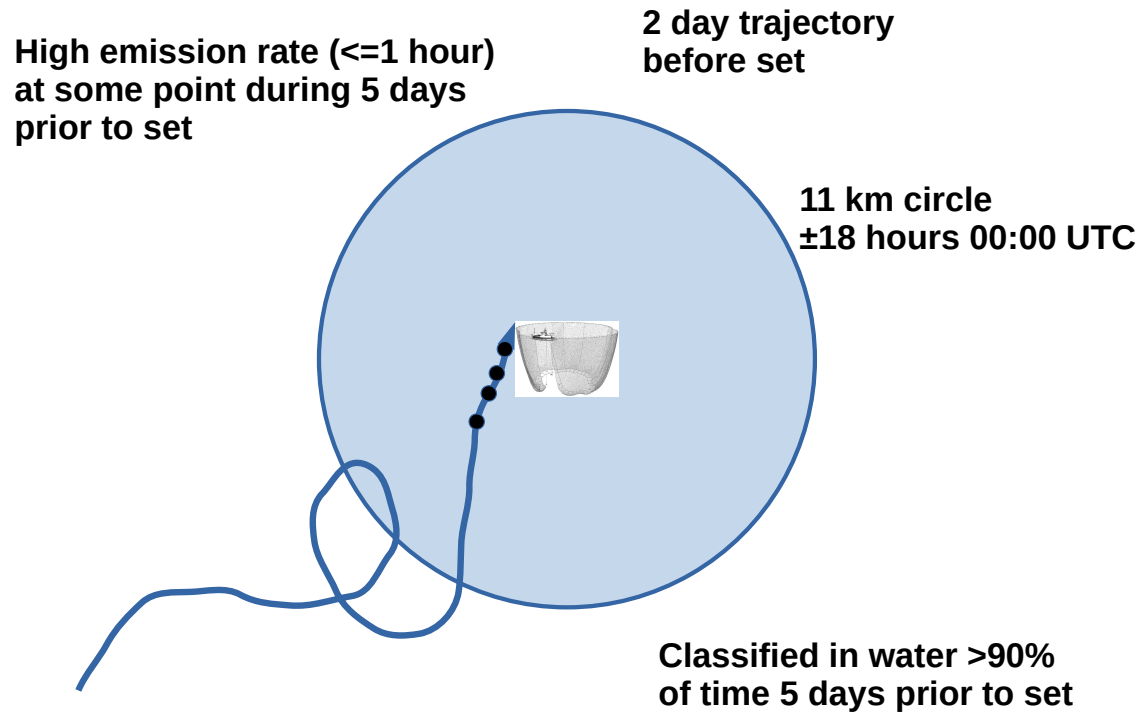
Determining FOB ownership

- Logbooks + FOB trajectories



Determining FOB ownership

- Logbooks + FOB trajectories



Determining FOB ownership

- Logbooks + FOB trajectories
- Base conditions of water trajectory, high emission rate, etc.
- **One of two additional conditions:**
 1. Matching logbook & FOB vessel names
 2. < 4 km spatial separation
- Vessel names only works for recent Marine Instruments buoys
- 4 km condition based on separation distance for FOBs meeting vessel name condition
- Error rate of $\approx 5\%$ based on free school sets

Set categories

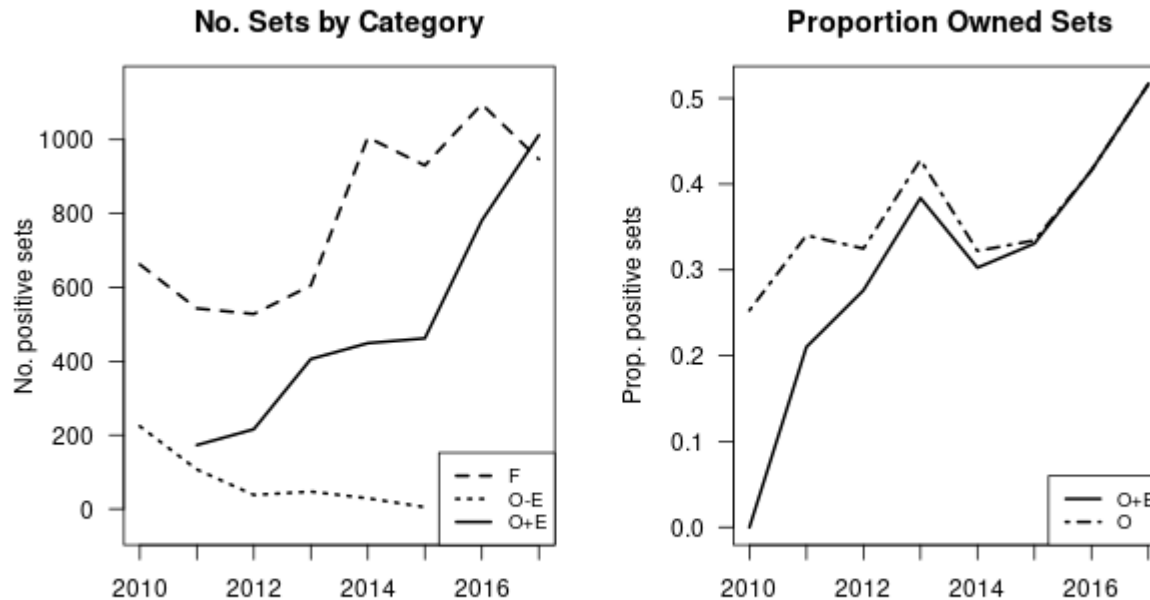
Ownership methodology + tracking buoy model allowed us to place each **French PS FOB set in the Indian Ocean between 2010 & 2017** into one of three categories:

- **Foreign (*F*):** Fishing vessel had no access to buoy tracking information
- **Owned-echosounder (*O-E*):** Fishing vessel had access to the buoy tracking data, but the buoy model was not echosounder equipped
- **Owned+echosounder (*O+E*):** Fishing vessel had access to tracking and echosounder data

→ **Examine catch of *F* versus *O+E* sets**

Analyses: Raw data

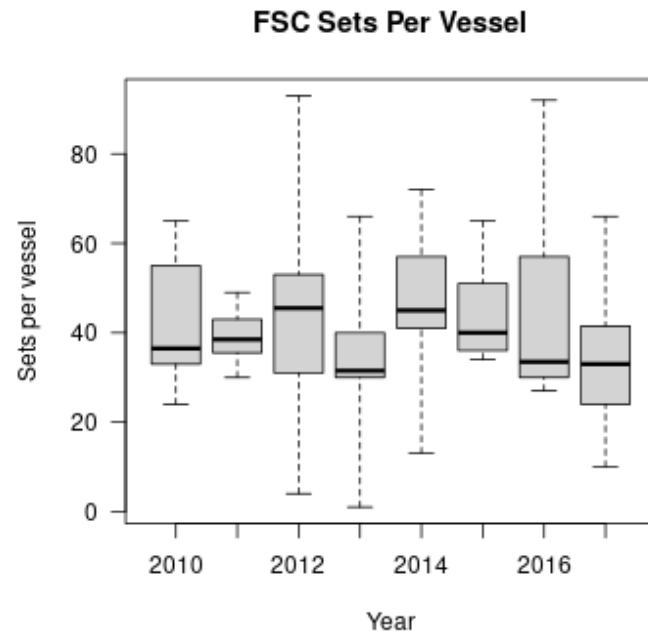
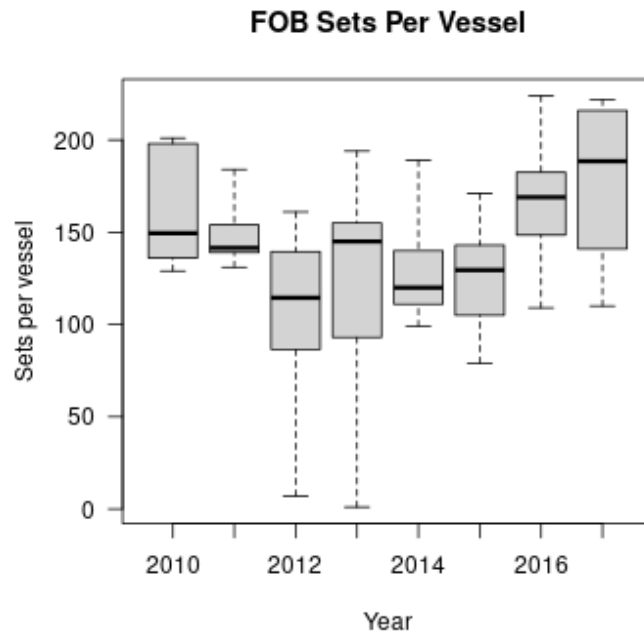
Temporal evolution



→ Recent increase in number of FOB sets

→ Increase in fraction of sets on owned FOBs

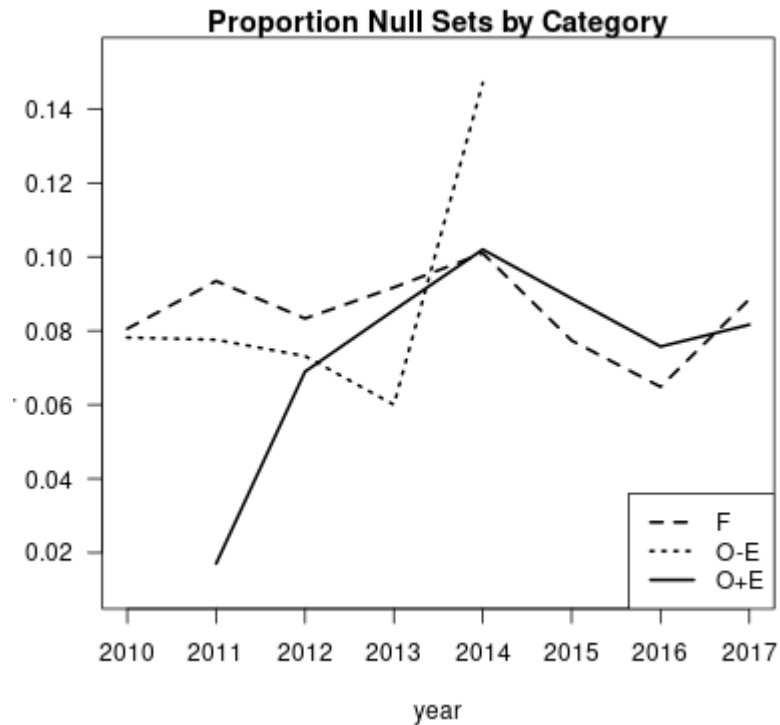
Sets per vessel



→ FOB *increase* 2014-2017: **51 sets/yr**

→ FSC *decrease* 2014-2017: **13 sets/yr**

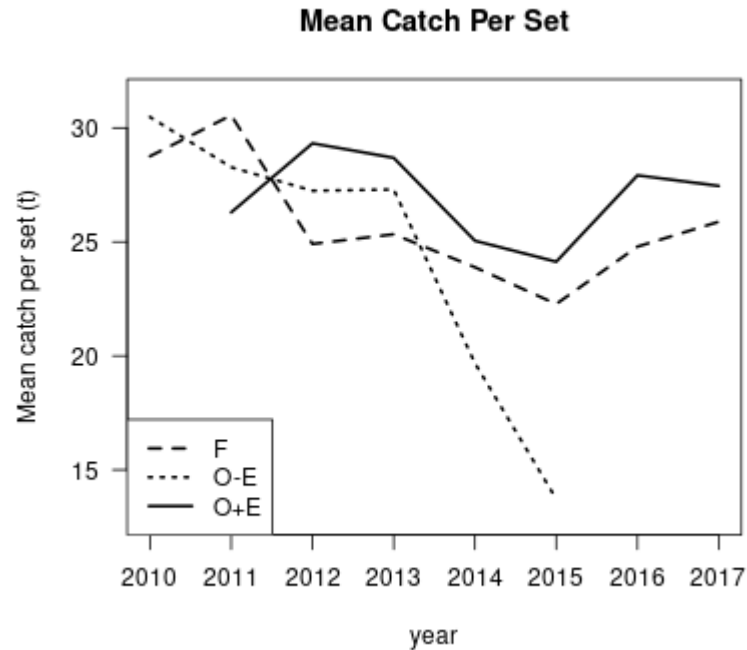
Null set reduction?



→ Sample size outliers: 2011:O+E & 2014:O-E

→ No apparent difference in null sets by category

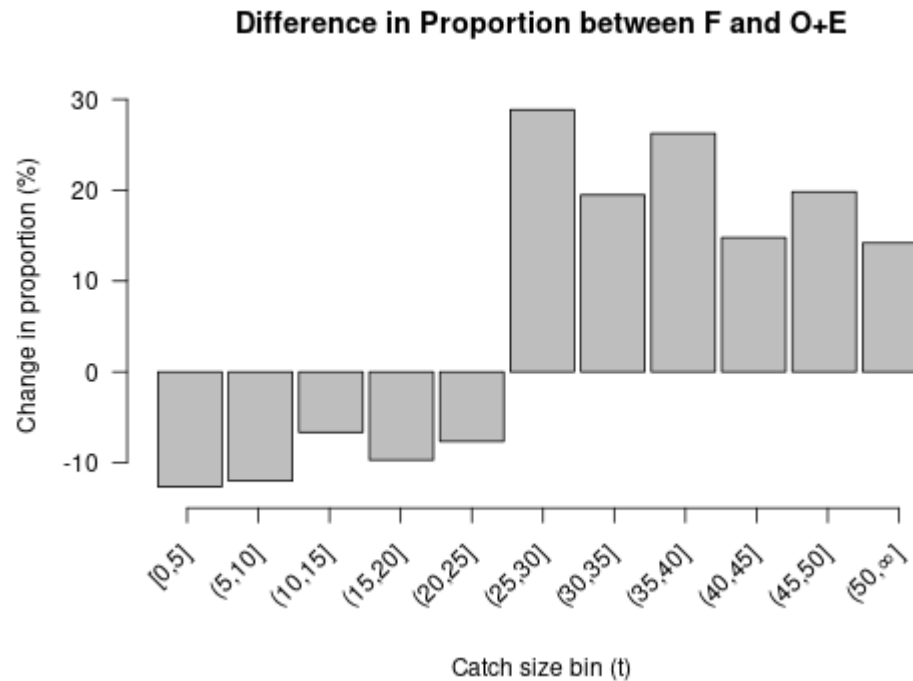
Catch per positive set?



→ Mean for O+E consistently higher than for F after 2011

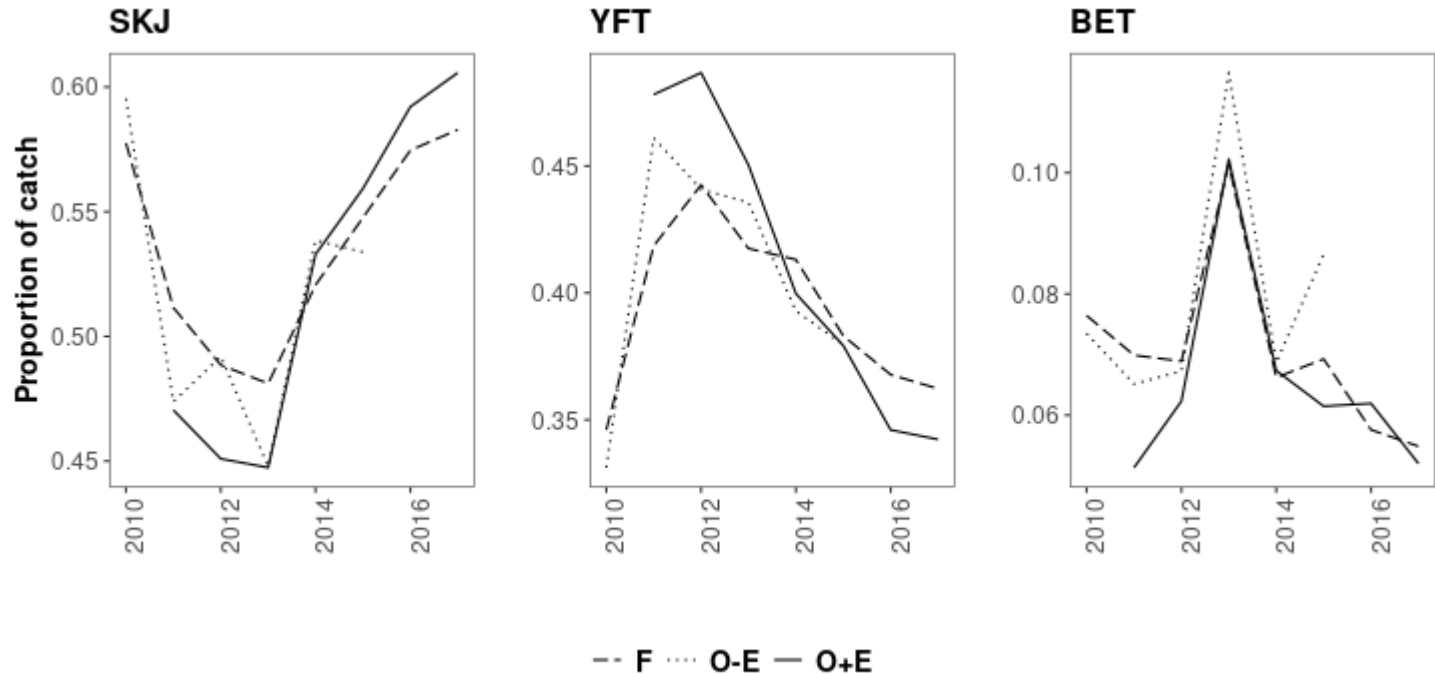
→ Approximately 2.6 tonnes per set $\approx 10\%$

Set size composition



→ **Non-linear transition around ≈ 25 tonnes \approx mean set size**

Species composition?



→ Small, but consistent, shift over time to >SKJ for O+E

→ Learning process?

Analyses: Model outputs

Model motivation

- Raw data could have spatial, temporal or vessel biases
- Standardize data to remove biases and extract echosounder effect
- Mostly GAM models (total catch, set category)
- Also Beta regression (species composition) & GLM (robustness of results)
- Focus primarily on 2012-2017
 - Avoid 2011: small sample size / learning period

Full model list

ID	Depend. var.	Data distribution	Type	Time period	Model equation
A1	Total catch	Gamma	GAM	2010-2017	$\text{te}(\text{lon}, \text{lat}, \text{by}=\text{season}) + \text{vessel} + \text{s}(\text{year}, \text{by}=\text{category})$
A2	Total catch	Gamma	GAM	2012-2017	$\text{te}(\text{lon}, \text{lat}, \text{by}=\text{season}) + \text{vessel} + \text{s}(\text{year}) + \text{category}$
L1	Total catch	Gamma	GLM	2012-2017	$\text{cwp55} + \text{vessel} + \text{season} + \text{year} + \text{category}$
A3	Total catch	Gamma	GAM	2012-2017	$\text{te}(\text{lon}, \text{lat}, \text{by}=\text{category}:\text{season}) + \text{vessel} + \text{s}(\text{year})$
B1	Proportion of SKJ	Beta	Beta regr.	2010-2017	$\text{cwp55} + \text{vessel} + \text{season} + \text{category} * \text{year} + \text{year}^2 + \text{year}^3 + \text{year}^4$
N1	Category F or E	Binomial	GAM	2012-2017	$\text{te}(\text{lon}, \text{lat}, \text{by}=\text{season}) + \text{vessel} + \text{s}(\text{year}) + \text{size class}$

Catch per set

	GAM Model A2			GLM Model L1		
	Estimate	Pr(> t)		Estimate	Pr(> t)	
Intercept (Cat. F)	22.00	0.0000	***	21.7	0.0000	***
Category O-E	-2.26	0.2241		0.2	0.9253	
Category O+E	1.98	0.0001	***	2.5	0.0000	***

Model A2: total catch ~ te(lon,lat,by=season)+vessel+s(year)+category

Model L1: total catch ~ cwp55+vessel+season+year+category

→ **Both GAM & GLM: 2.0-2.5 tonnes per set increase**

→ **GAM more sophisticated, "better" standardization**

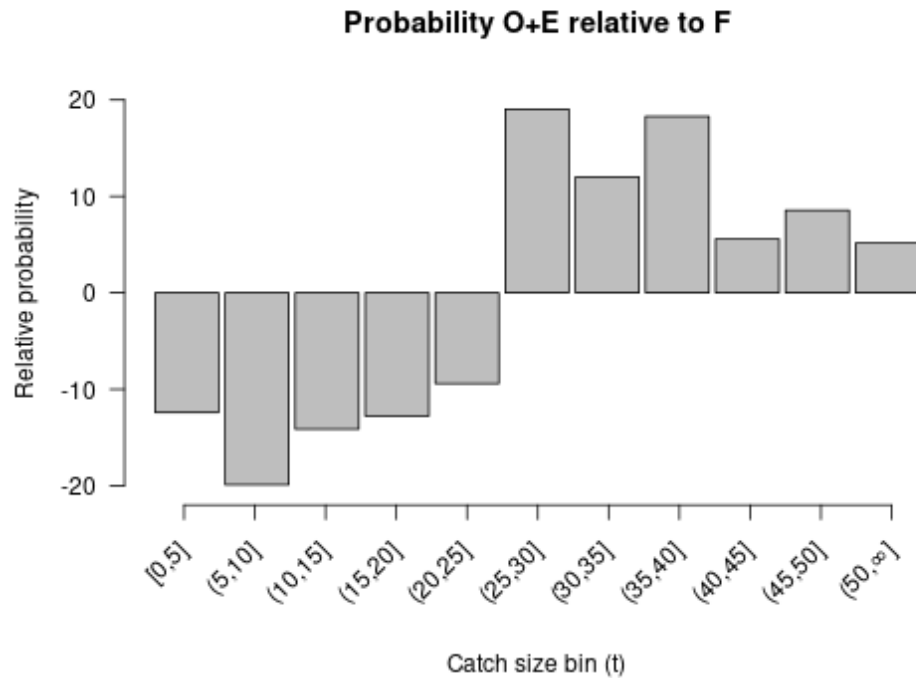
Proportion SKJ

Beta Model B1			
	Estimate	Pr(>z)	
Intercept (Cat. F)	0.05	0.5701	
Year	7.14	0.0000	***
Year ²	6.59	0.0000	***
Year ³	-7.76	0.0000	***
Year ⁴	-1.52	0.0935	.
Category O-E	0.08	0.4796	
Category O+E	0.00	0.8112	
Year : Category O-E	2.82	0.7120	
Year : Category O+E	15.39	0.0000	***

Model B1: Prop. SKJ ~ cwp55 + vessel + season + category*year + year² + year³ + year⁴

→ **Positive interaction between time and proportion SKJ**

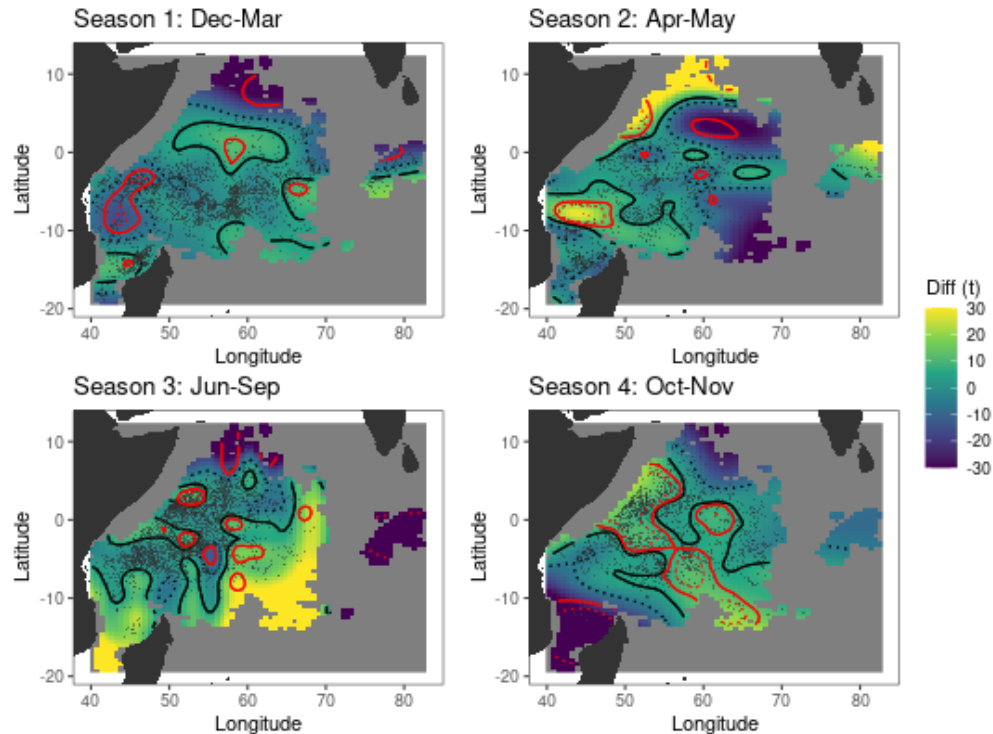
Set size composition



→ Consistent with, but somewhat less dramatic than, analyses of raw data

Model N1: Set category \sim te(lon,lat,by=season) + vessel + s(year) + size class

Spatial effects



- **Model A3:**
te(lon,lat,by=category:season)
- **Green-Yellow:**
More O+E catch
- **Black contours:**
> 5 tonnes difference
- **Red contours:**
Significant difference
- **Dots:**
Data points
- **Seasons 1 & 3:**
4 months
- **Seasons 2 & 4:**
2 months

- **Pretty noisy / no strong patterns**
- **Perhaps off Somalia Season 4 & off Tanzania Season 2**

Conclusions

- Robust methodology for assigning FOB ownership
- Depends on fine scale FOB trajectory data
- Increase in proportion sets on own FOBs
- Recent change from FSC to FOB fishing
- Echosounders increase catch per set ≈ 2 -2.5 tonnes $\approx 10\%$
 - Net gain of \approx US\$5000-7000 over life of buoy
 - Search time change to be quantified, but net 30% increase in FOB sets/vessel/year
- Accounting for foreign sets, net change in FOB fishing efficiency of 1.7%-4.0%
- Reduce sets < 25 tonnes (Baida *et al.* 2020)
- Temporal shift towards more SKJ?
- Weak spatial patterns?

Stock assessment

- Reduction in PS abundance index since onset of echosounder use
 - 2012 for French Fleet
 - ?? for Spanish Fleet
- Reduction size a function of proportion of sets on owned echosounders
 - Net 2%-4% for French Fleet
 - ?? for Spanish Fleet
 - Expert opinion estimate of proportion on owned echosounders?
- Future PS abundance indexes should also account for increases in search efficiency
 - In the meantime, constant increase in fishing efficiency?

Acknowledgements

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