

How to characterize human influence on coastal benthic ecosystems?

Elliot Dreujou, Christopher W McKindsey, Nicolas Desroy, Natalie Ban, Aurélie Foveau, Philippe Archambault November 5th 2019, Bordeaux





Connecting ocean and human activities



Spatial and temporal changes in cumulative human impacts on the world's ocean

Benjamin S. Halpern^{1,2,3}, Melanie Frazier³, John Potapenko⁴, Kenneth S. Casey⁵, Kellee Koenig⁶, Catherine Longo³, Julia Stewart Lowndes³, R. Cotton Rockwood⁷, Elizabeth R. Selig⁶, Kimberly A. Selkoe^{3,8} & Shaun Walbridge⁹

Nat Commun, 2015

Recent pace of change in human impact on the world's ocean

Benjamin S. Halpern^{1,2}, Melanie Frazier¹, Jamie Afflerbach¹, Julia S. Lowndes¹, Fiorenza Micheli^{3,4}, Casey O'Hara², Courtney Scarborough¹ & Kimberly A. Selkoe^{1,2}

Sci Rep, 2019

Review

An effective set of principles for practical implementation of marine cumulative effects assessment

A.D. Judd ^{a,*}, T. Backhaus ^b, F. Goodsir ^a

Environ Sci & Policy, 2016

Response of benthic assemblages to multiple stressors: comparative effects of nutrient enrichment and physical disturbance

Joseph M. Kenworthy^{1,2,3,*}, David M. Paterson¹, Melanie J. Bishop²

Mar Ecol-Progr Ser, 2016

Multiple Stressors in a Changing World: The Need for an Improved Perspective on Physiological Responses to the Dynamic Marine Environment

Alex R. Gunderson, Eric J. Armstrong, and Jonathon H. Stillman

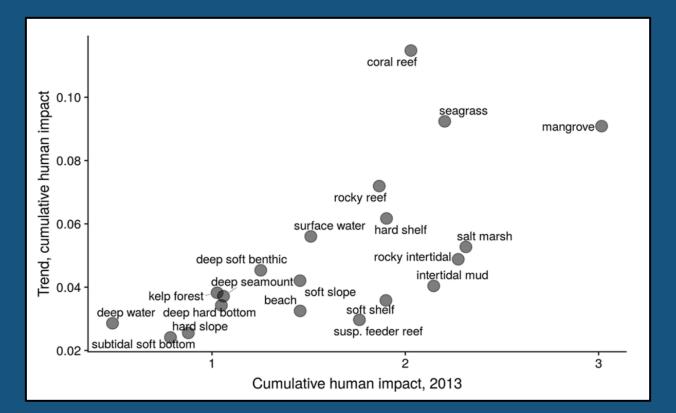
Annu Rev Mar Sci, 2016

Cumulative impact mapping: Advances, relevance and limitations to marine management and conservation, using Canada's Pacific waters as a case study Natalie C. Ban^{a,b,*}, Hussein M. Alidina^c, Jeff A. Ardron^d

Mar Policy, 2010

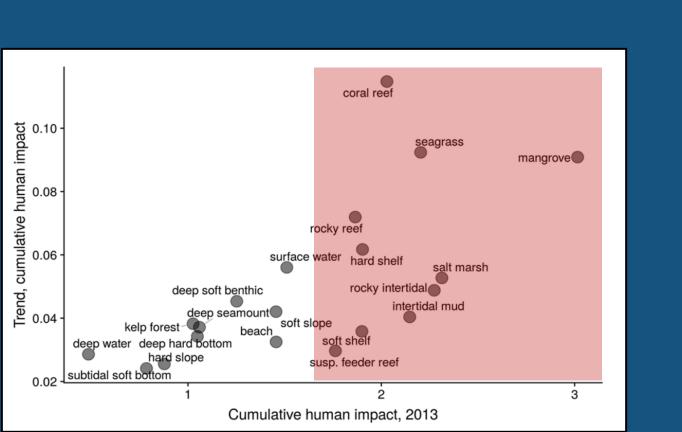
> The majority of the oceans is influenced by multiple human activities

Some ecosystems are more vulnerable





Some ecosystems are more vulnerable



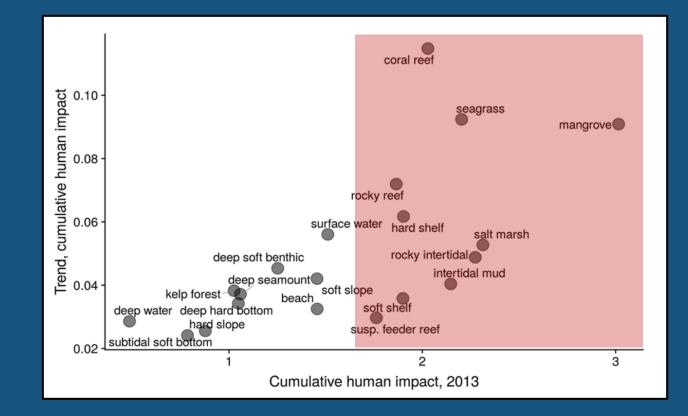
> Many coastal ecosystems are among the most influenced by cumulative impacts

Halpern et al. (2019)

Some ecosystems are more vulnerable

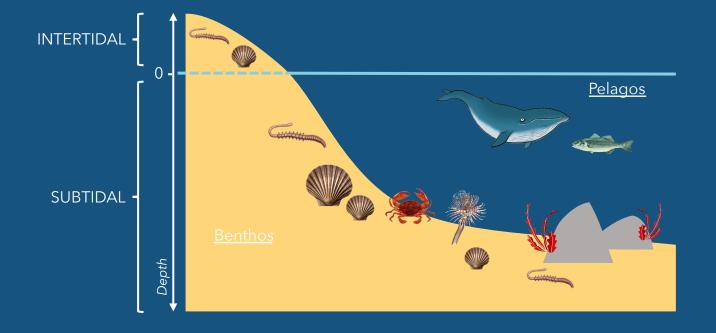


2

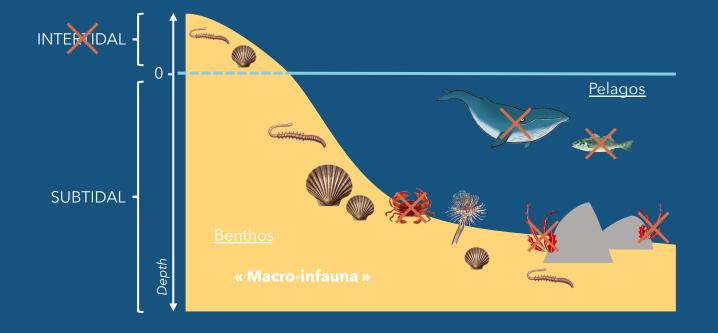


- > Many coastal ecosystems are among the most influenced by cumulative impacts
- Importance of monitoring programs and ecological studies to predict evolution (e.g. REBENT)

Halpern et al. (2019)



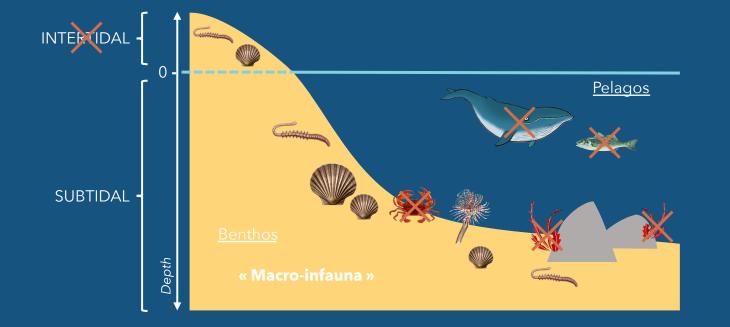








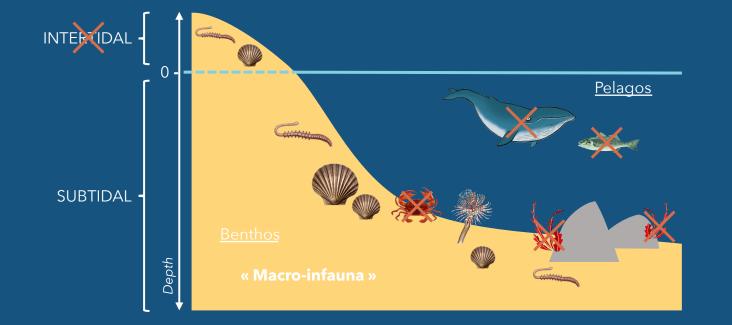
3





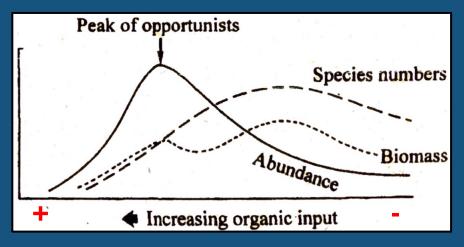
Importance for the ecosystemImportance for mankind





- Importance for the ecosystem
- Importance for mankind
- Sessile species (sensitive to perturbation)







Characterize cumulative impacts of human activities on benthic coastal ecosystems

Methods

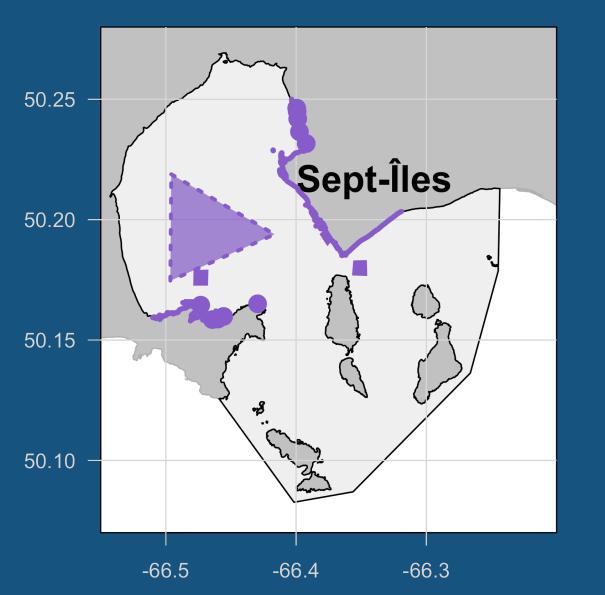






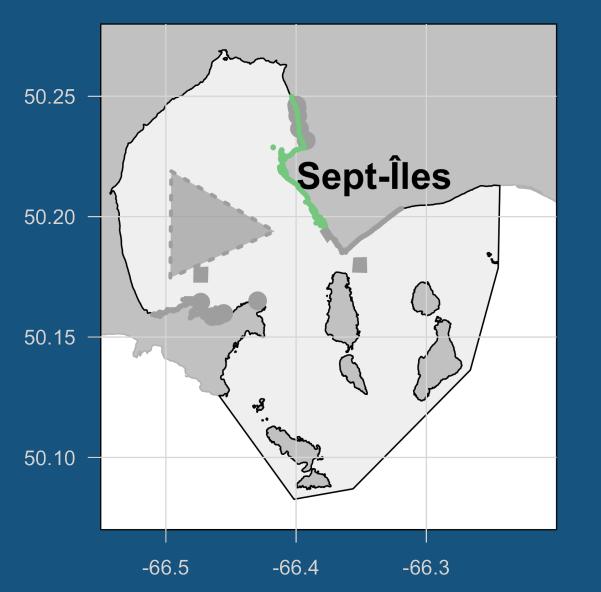
Sept-Îles, Canada

- 5th Canadian harbour
 (25.3 MT of exchanged goods in 2018)
- Various industrial and harbour activities





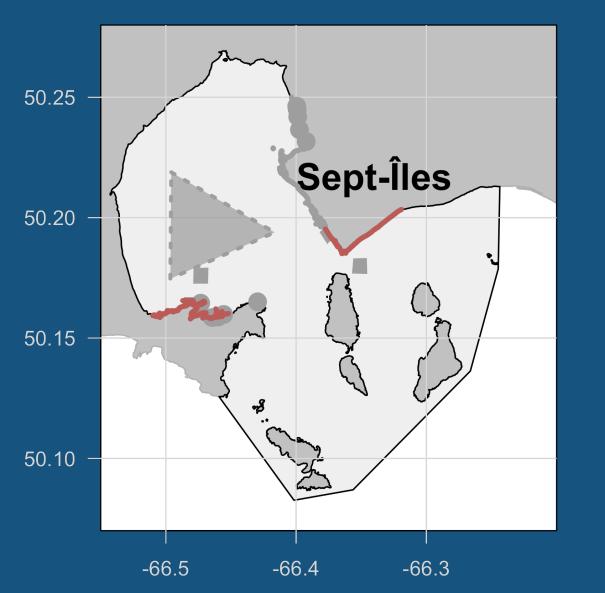
- 9 different human activities
- Classified under 4 categories





- city influence
- municipal wharves



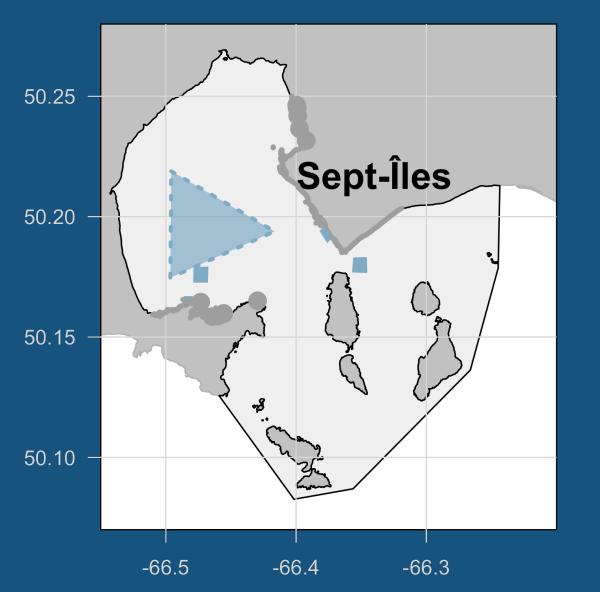






- city influence
- municipal wharves
- industries influence
- industrial wharves





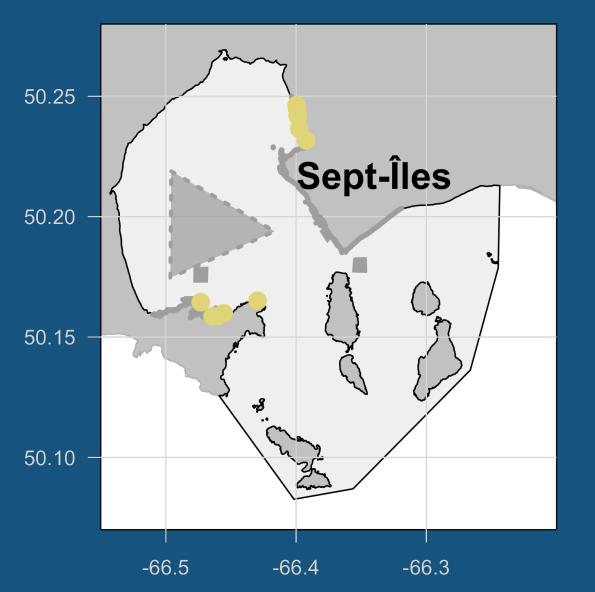






- city influence
- municipal wharves
- industries influence
- industrial wharves
- dredging dumping sites
- dredging collect sites
- commercial ship mooring sites









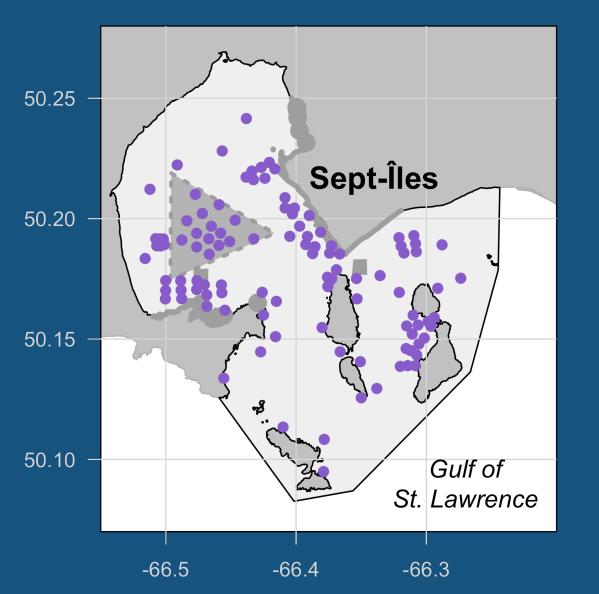




- city influence
- municipal wharves
- industries influence
- industrial wharves
- dredging dumping sites
- dredging collect sites
- commercial ship mooring sites
- rainwater sewers
- wastewater sewers



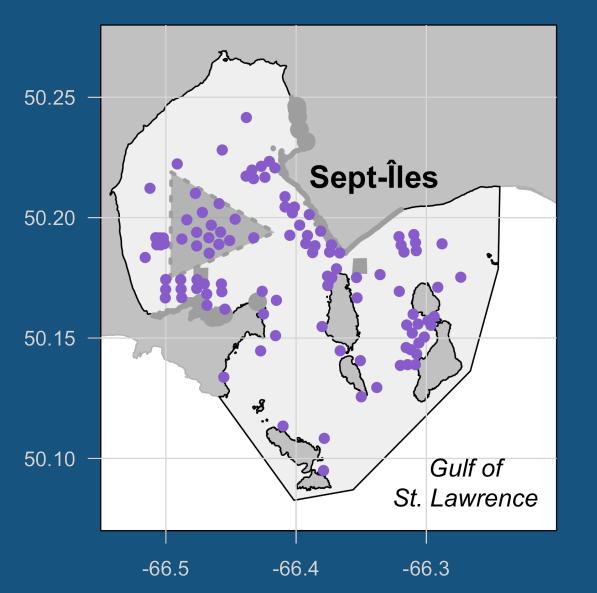
Data collection



108 stations sampled in 2017
Between 0 and 70 m deep



Data collection



- ➤ 108 stations sampled in 2017
- ➢ Between 0 and 70 m deep
- Benthic communities sieved with a
 0.5 mm mesh

- > Habitat parameters considered:
 - organic matter
 - sediment grain-size
 - heavy metal concentrations

Calculation of an index of influence with connectivity functions based on particle dispersion models:

 $I_{ij} = f_j(D_{ij}, Z_i)$

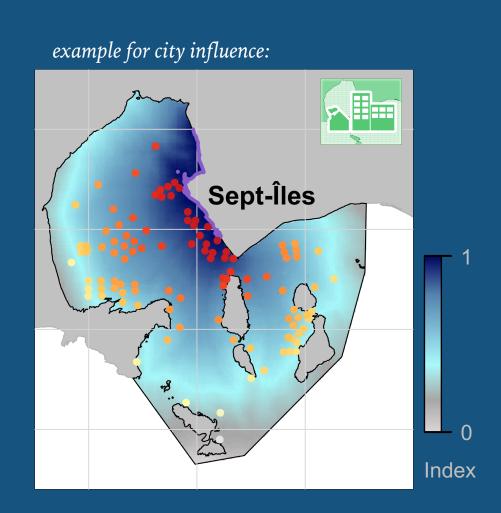
i: station, *j*: activity *D_{ij}*: distance from the source *Z_i*: depth *f_j*: connectivity function



Calculation of an index of influence with connectivity functions based on particle dispersion models:

 $I_{ij} = f_j(D_{ij}, Z_i)$

i: station, *j*: activity D_{ij} : distance from the source Z_i : depth f_j : connectivity function



Calculation of an index of influence with connectivity functions based on particle dispersion models:

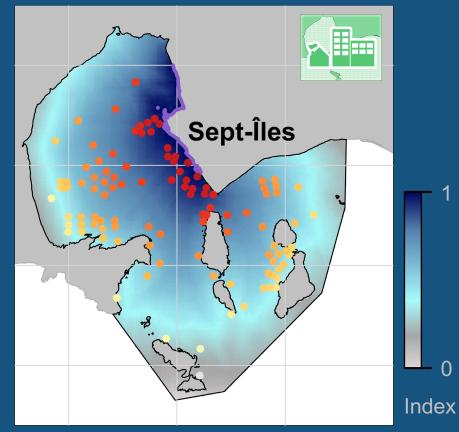
 $I_{ij} = f_j(D_{ij}, Z_i)$

i: station, *j*: activity *D_{ij}*: distance from the source *Z_i*: depth *f_j*: connectivity function

Cumulative influence computed with an additive model:

$$CI_i = \sum_j I_{ij}$$

example for city influence:



Predicting benthic communities



Organic matter



Gravel, Sand, Silt, Clay



Arsenic, Cadmium, Chromium, Copper, Iron, Manganese, Mercury, Lead, Zinc



Human influence indices

Habitat parameters

Correlation

Prediction







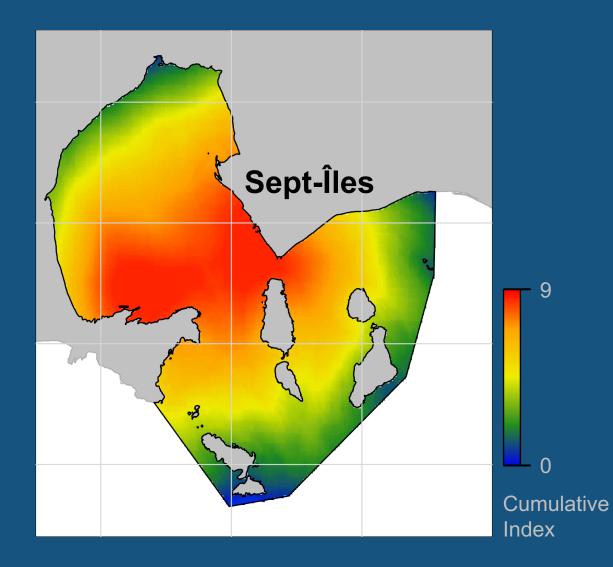


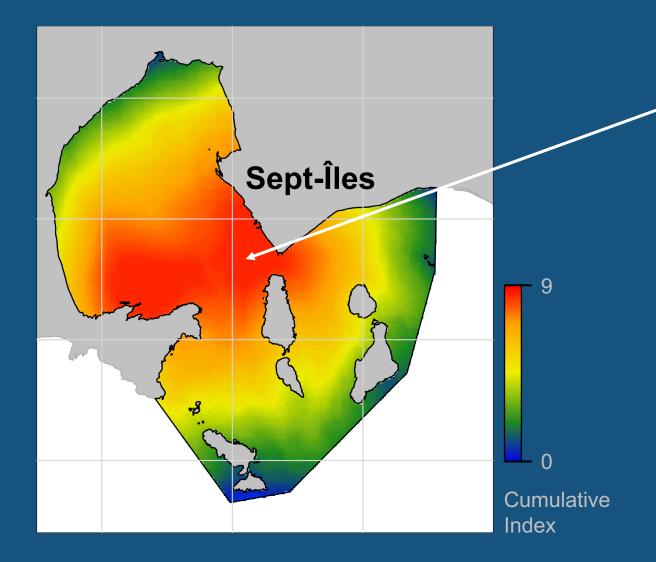


Benthic infauna

Results & Discussion



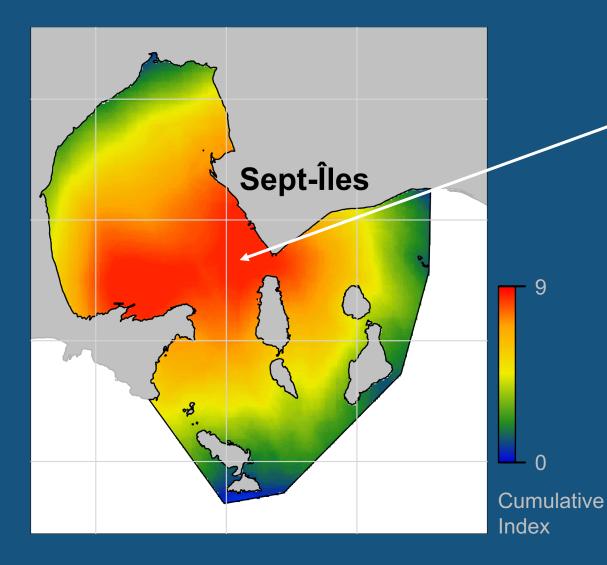




"Hotspot" of human influence

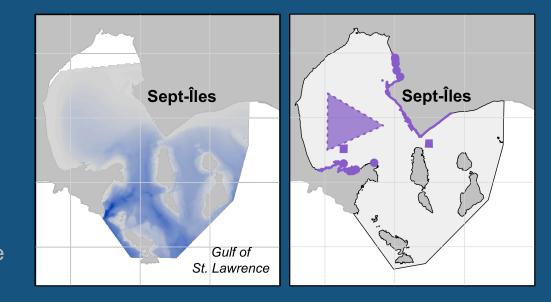


.

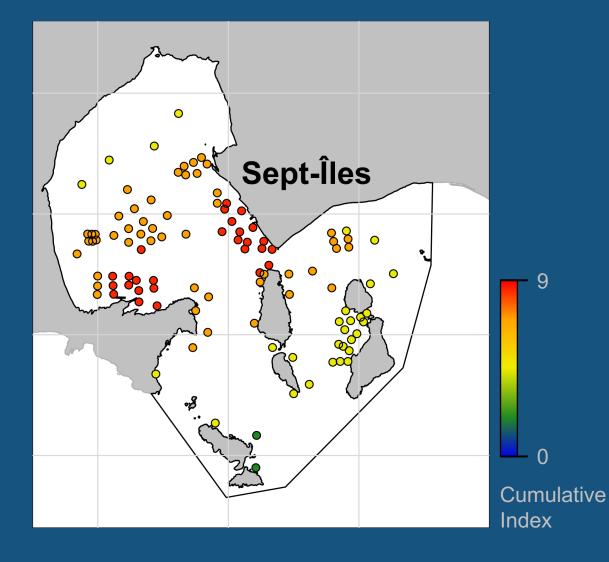


"Hotspot" of human influence

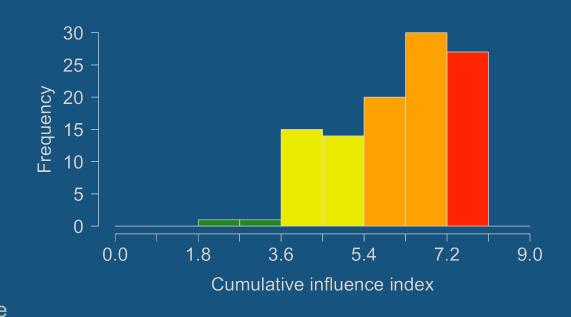
Coherent with the bathymetry and the sources of human activity :



lılı.







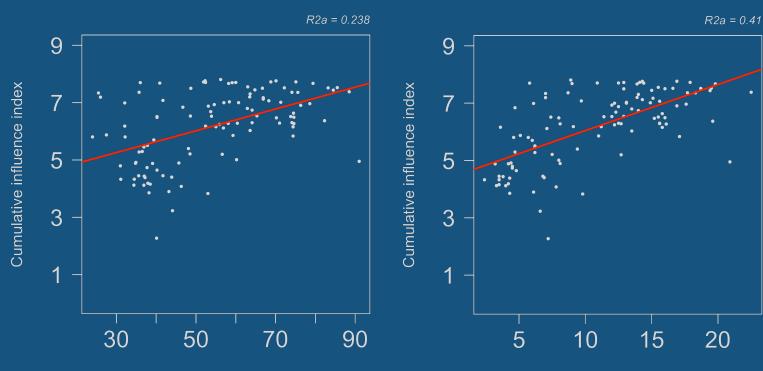
|ılı.



10



Highest correlation with metals concentrations



Chromium concentration

Copper concentration



R2a = 0.105

Δ

10

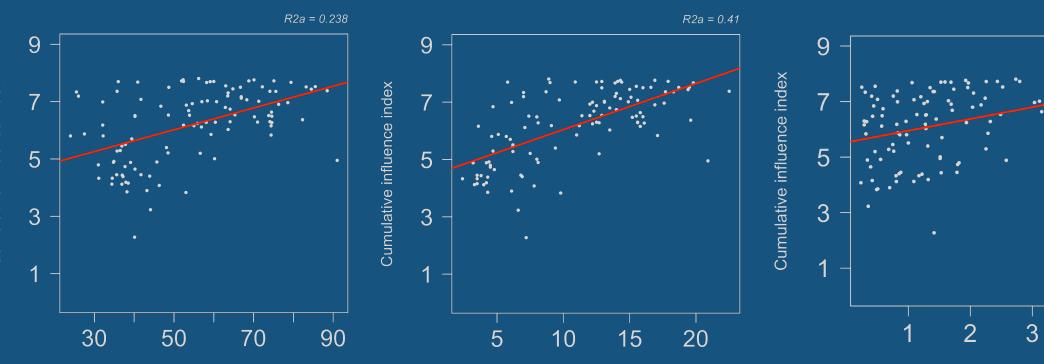


Highest correlation with metals concentrations



Low correlation with most habitat parameters

Cumulative influence index



Chromium concentration

Organic matter

Copper concentration



R2a = 0.105





Chromium concentration

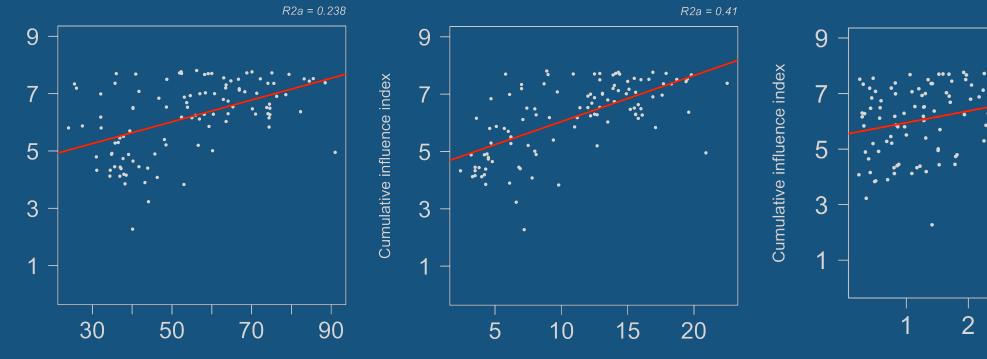
Highest correlation with metals concentrations — *Human activities?*

Copper concentration



Low correlation with most habitat parameters

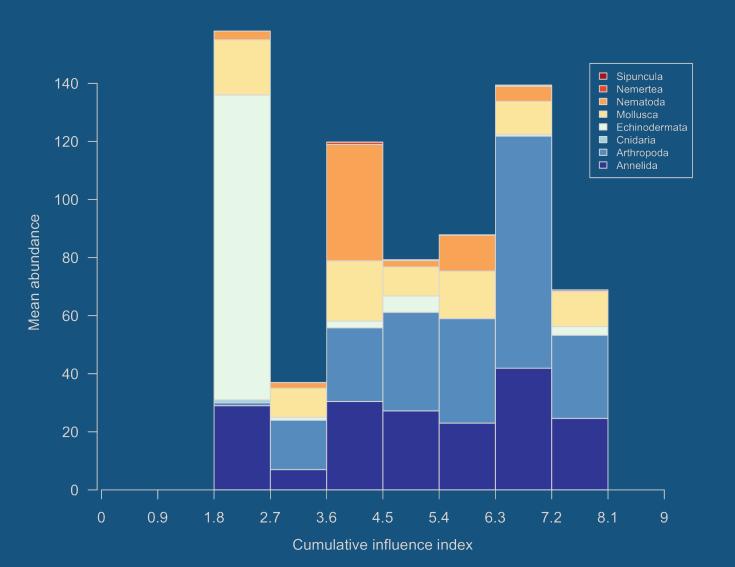
Cumulative influence index

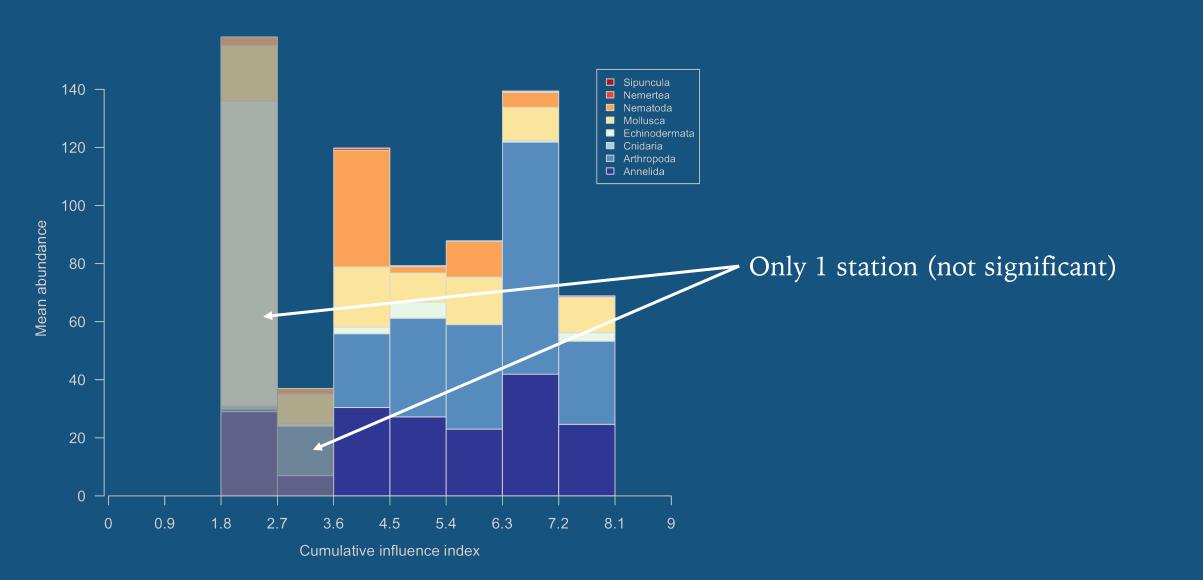


Organic matter

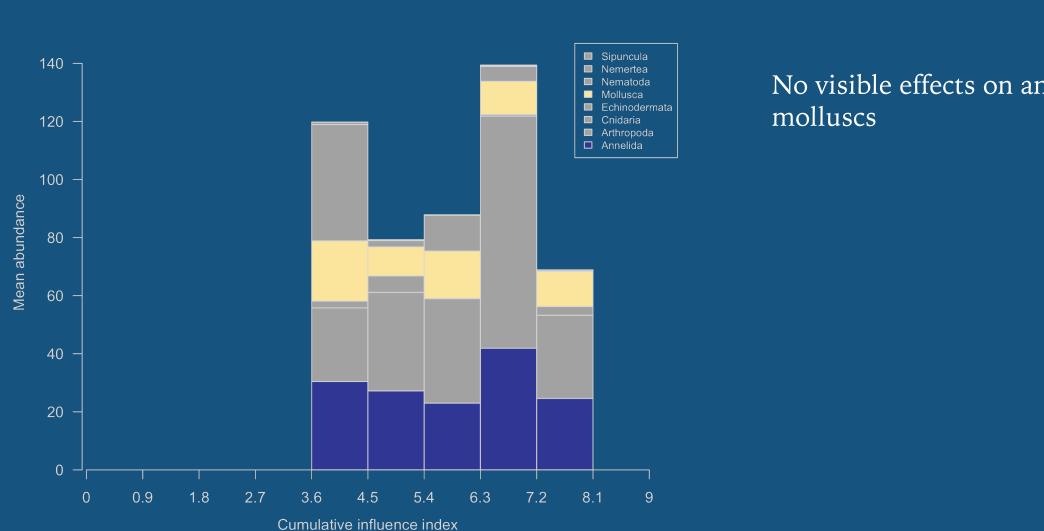
3

Δ





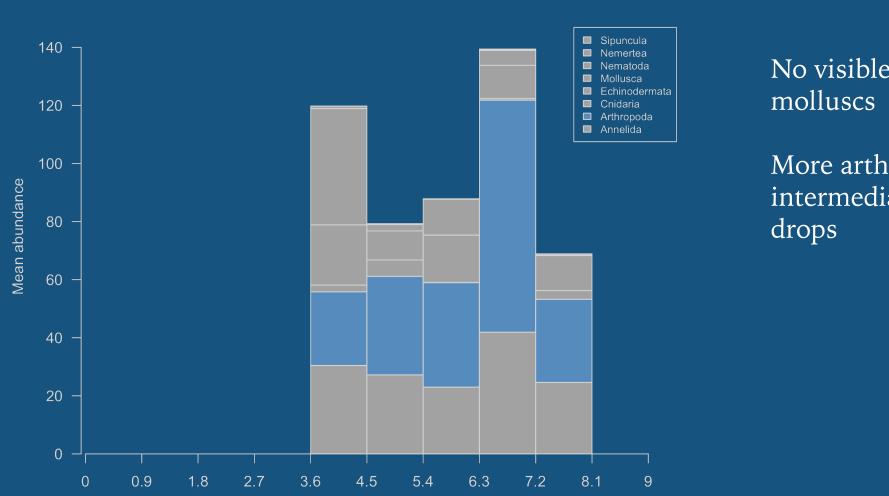
|.



No visible effects on annelids or

11

|.



Cumulative influence index

No visible effects on annelids or molluscs

More arthropods with higher intermediate influence, then drops



140 -

120

100

80

60

40

20

0

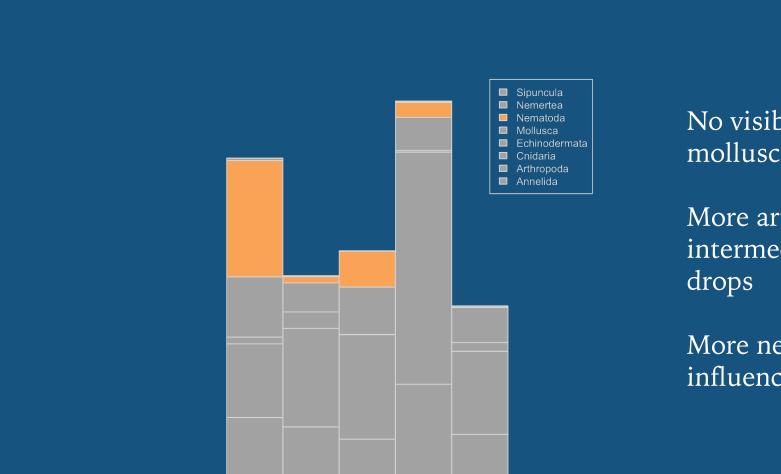
0

0.9

1.8

2.7

Mean abundance



No visible effects on annelids or molluscs

11.

11

More arthropods with higher intermediate influence, then drops

More nematodes with moderate influence, then drops

Cumulative influence index

4.5

5.4

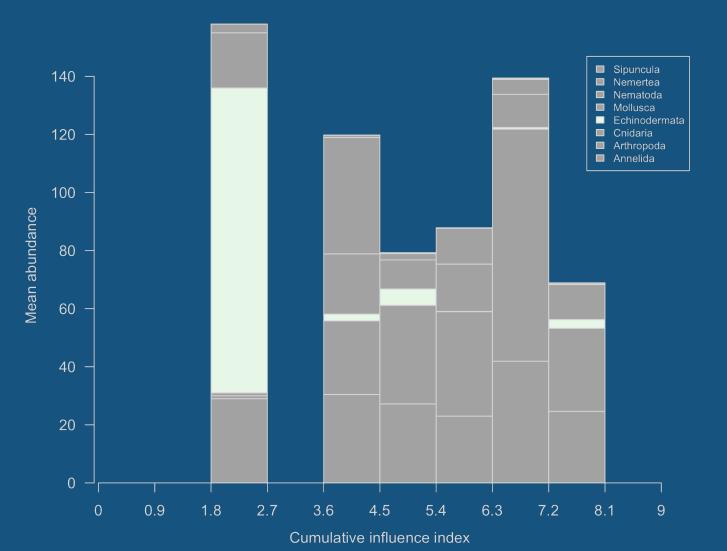
6.3

7.2

8.1

9

3.6



No visible effects on annelids or molluscs

More arthropods with higher intermediate influence, then drops

More nematodes with moderate influence, then drops

Echinoderms only with low influence?

11

11.

To conclude:

Development of an human influence score based on the activities' sources and the environment parameters

To conclude:

Development of an human influence score based on the activities' sources and the environment parameters

The influence score can be linked to habitat and communities' responses :

- correlations with organic matter and some metals
- variation of phylum abundances

Next steps...

Increase representativity of the influence index with different particle models

> Add more human activities (e.g. fisheries)

Next steps...

Increase representativity of the influence index with different particle models

> Add more human activities (e.g. fisheries)

Use prediction techniques (like HMSC) to predict change under different human activity scenarios

Next steps...

Increase representativity of the influence index with different particle models

> Add more human activities (e.g. fisheries)

Use prediction techniques (like HMSC) to predict change under different human activity scenarios

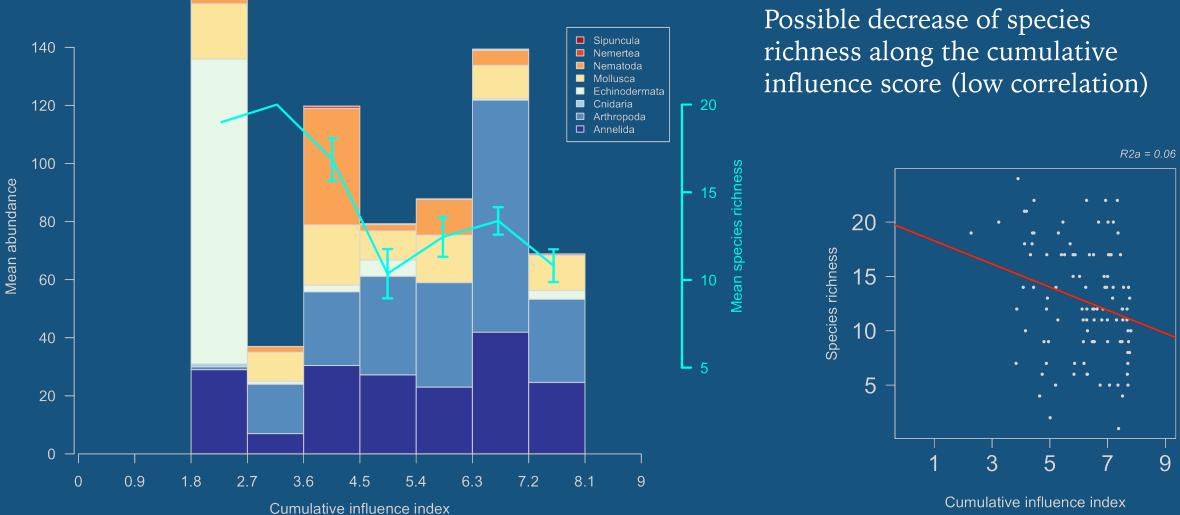
Study different spatial scales (e.g. St. Lawrence)



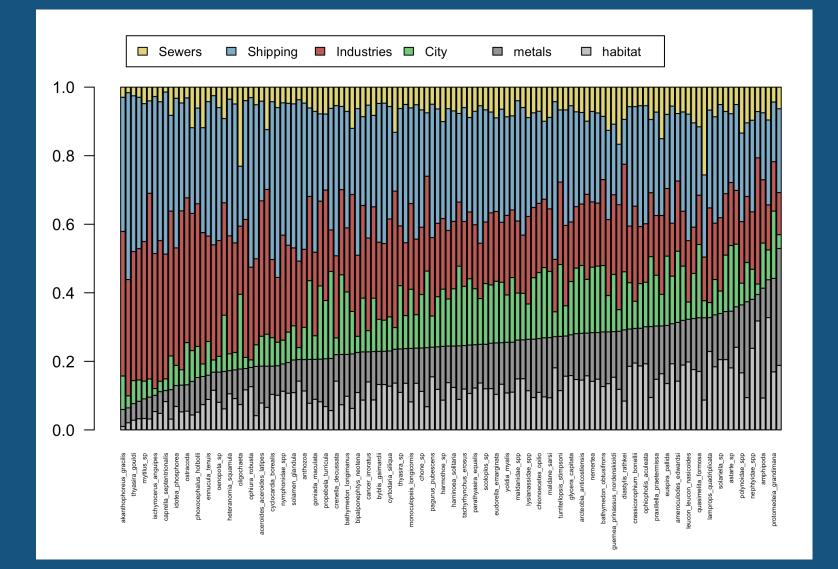
This project is supported by NSERC program CHONe II and its Partners: DFO Canada and INREST (representing Port de Sept-Îles and Ville de Sept-Îles)

Thanks for your attention!

https://eldre.github.io

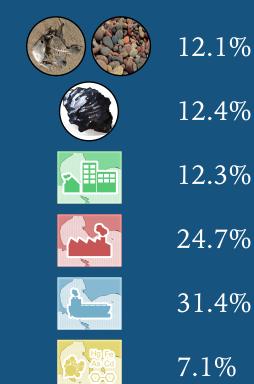


HMSC: variance partitioning



R^2 of 19.7 %

Mean species abundance variances explained by:



Graphical abstract







Chapter 3