

# Relating Wave Energy and Shoreline Change: A Uruguayan Case Study

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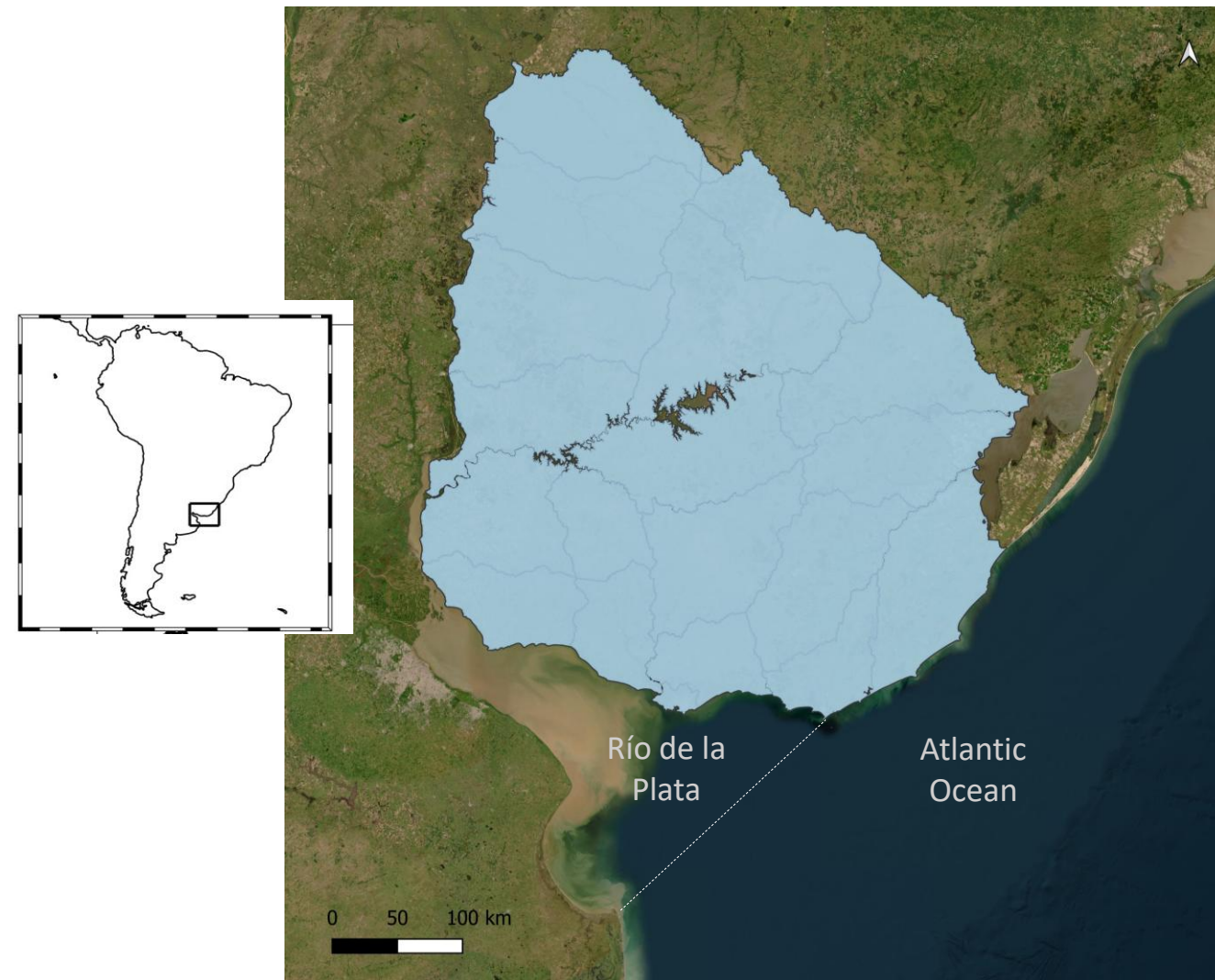
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# Study Area Motivation Databases Methodology Results Conclusions



## Uruguayan Coast

- ❖ 700 km long aprox.
  - Río de la Plata (~465 km) + Atlantic (~ 235 km)
- ❖ Mostly composed of sandy beaches.
  - Micro-tidal
  - Wave-dominated morphology
- ❖ Different wave climates between the two extremes.
  - Atlantic: Swell dominated  
(Mostly Multimodal spectra).
  - Inner Río de la Plata: Wind sea dominated  
(Mostly fetch limited).

# Motivation

## *Previous work*

To support a knowledge-based coastal management, two nationwide, high resolution and long-term databases were developed:

Local wave hindcast

Remote monitoring of  
shoreline position

## *Ongoing work*



**Analyze relationships between waves climate and  
shoreline variability**

## *This presentation*

Focus on inter-annual scale



Study Area Motivation

Databases

Methodology

Results

Conclusions

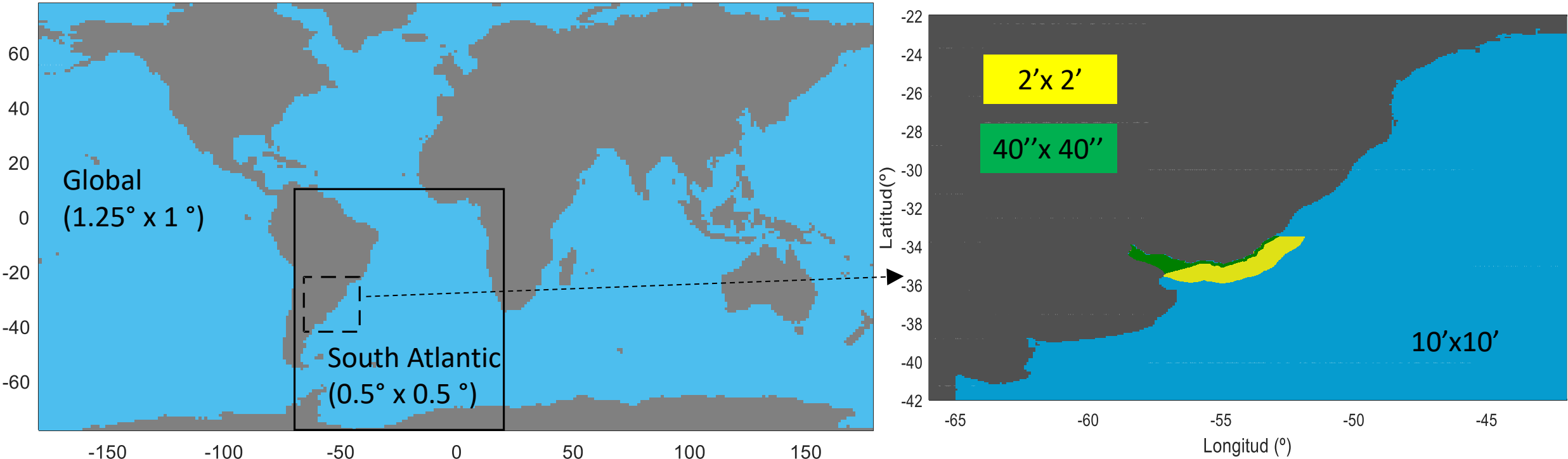
Wave hindcast

Configuration

WAVEWATCH III ® 5.16. Multi-grid mode. Two-way nesting. 5 regular grids.

Forcings

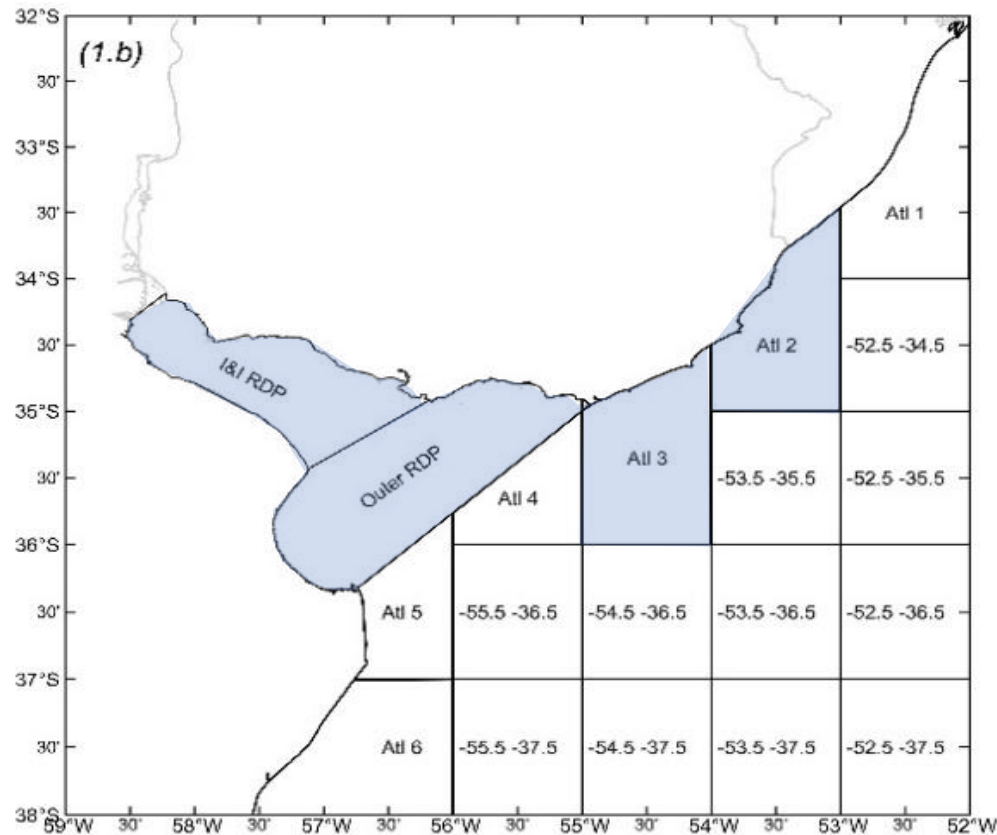
CFSR winds ~0.31° for all the grids.  
TELEMAC water levels and currents 2' for high rank grids  
(Green and yellow)



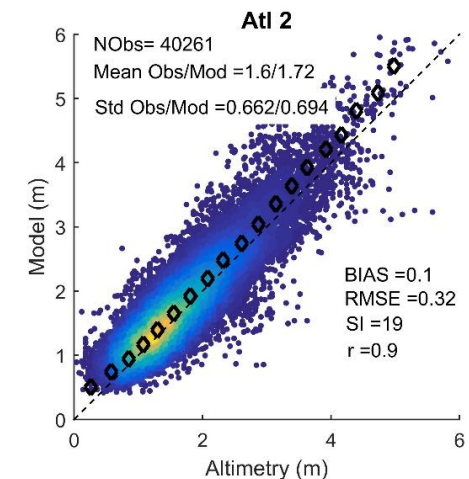
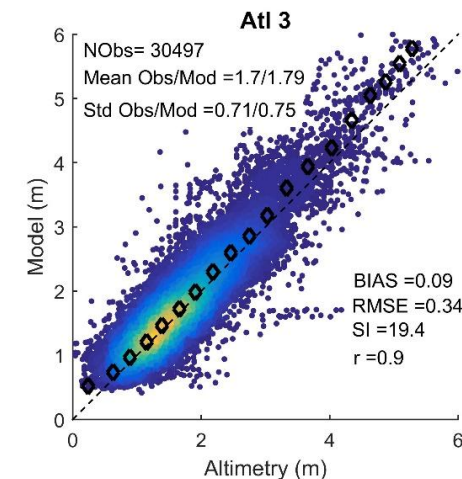
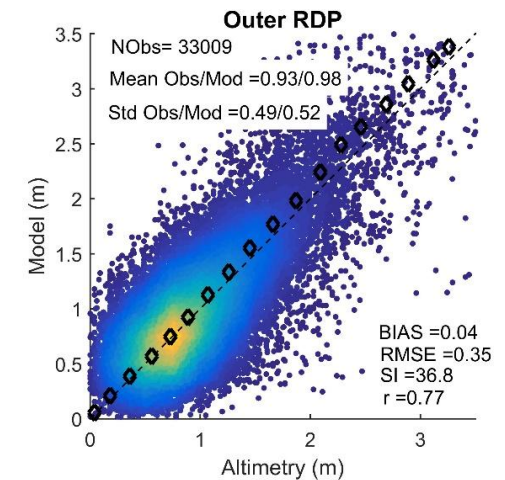
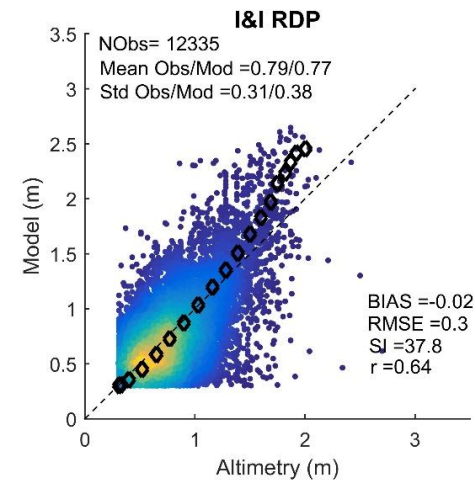
# Study Area Motivation Databases Methodology Results Conclusions

## Wave hindcast

### Calibrated with altimetry



$$\beta_{\max}(ST4)=1.55, \Gamma(S_{\text{bot}})=-0.012 \text{ m}^2\text{s}^{-3}$$



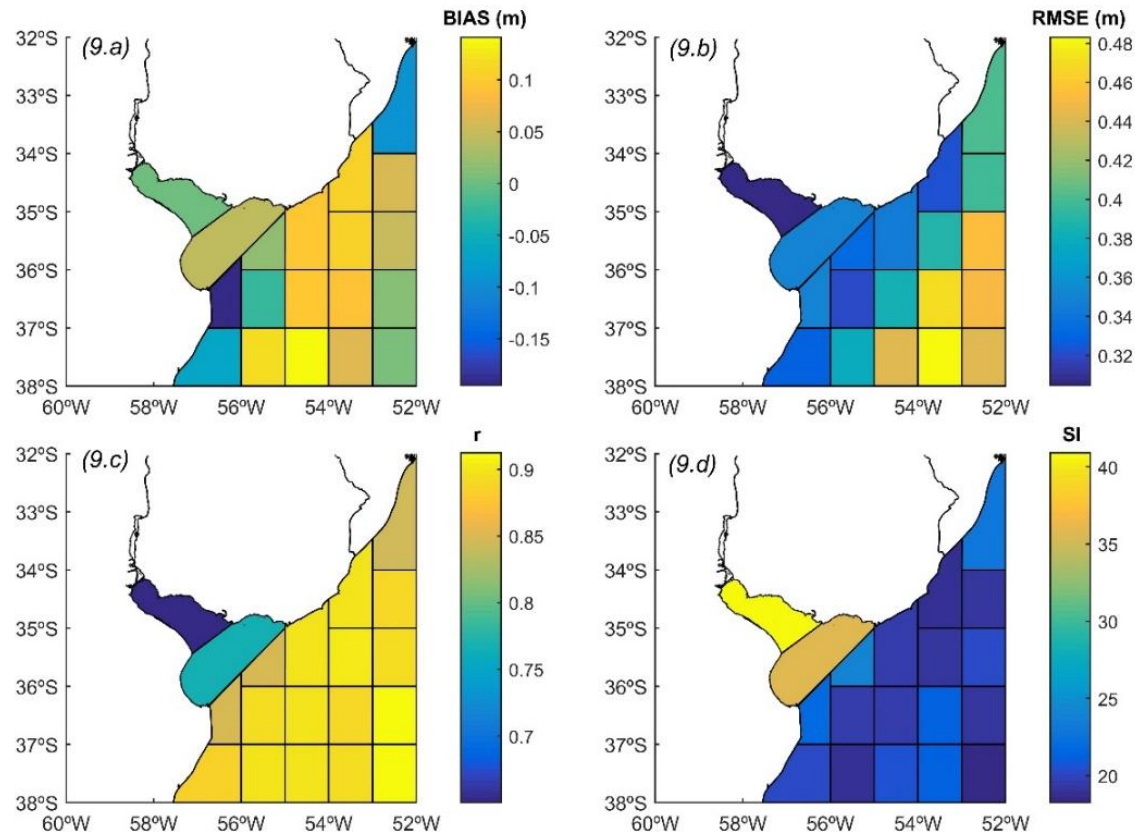


# Study Area Motivation Databases Methodology Results Conclusions

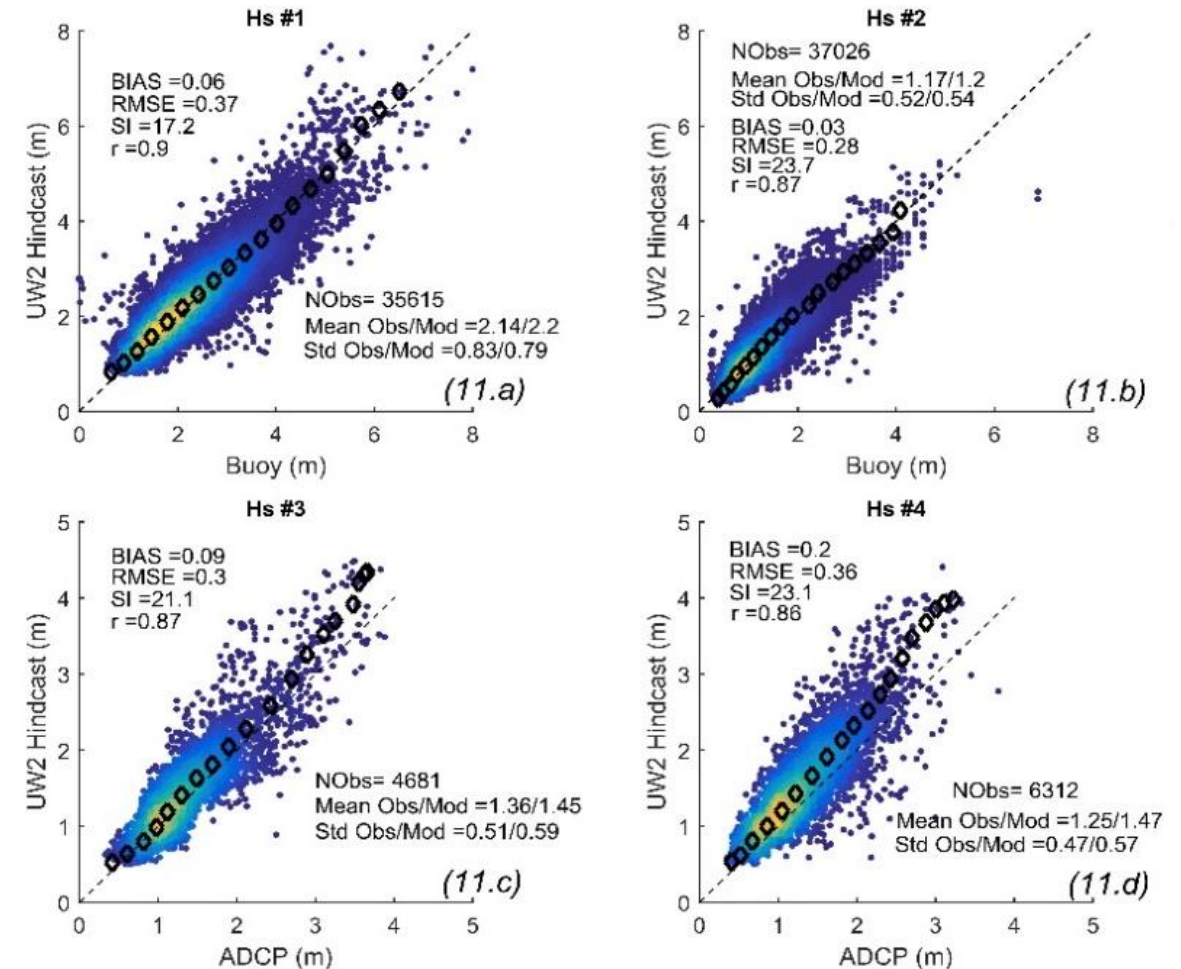
## Wave hindcast

### Validation

#### Altimetry



### In-situ



# Study Area Motivation Databases Methodology Results Conclusions

## Wave hindcast

### Products

❖ Gridded integral parameters

❖ 68 Virtual Buoys: Directional spectra (25 freq. X 36 dir.)

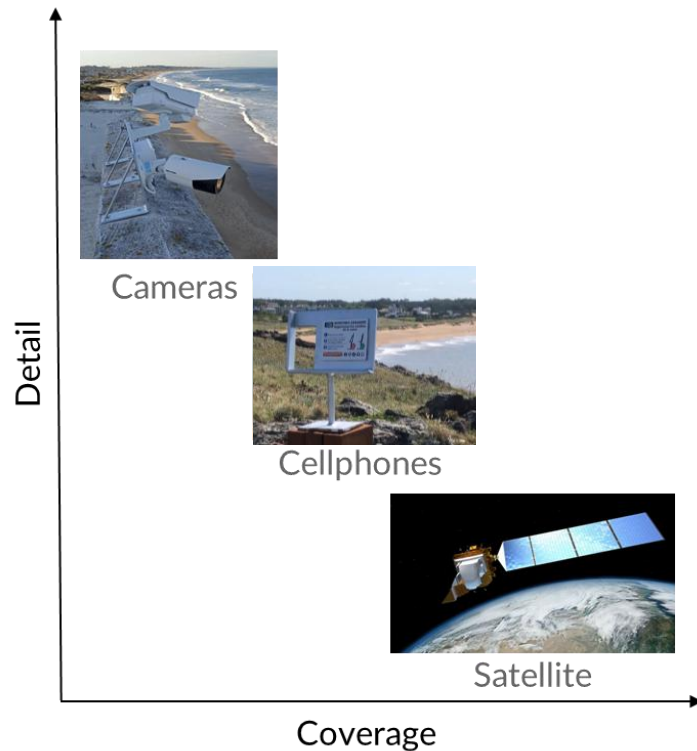




# Study Area Motivation Databases Methodology Results Conclusions

## Remote monitoring of shoreline position

- 3 sources of information

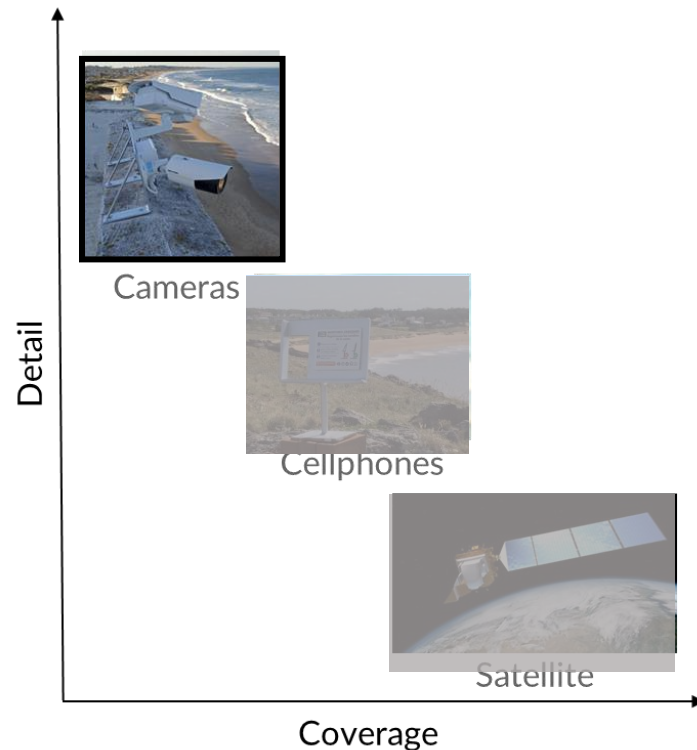




# Study Area Motivation Databases Methodology Results Conclusions

## Remote monitoring of shoreline position

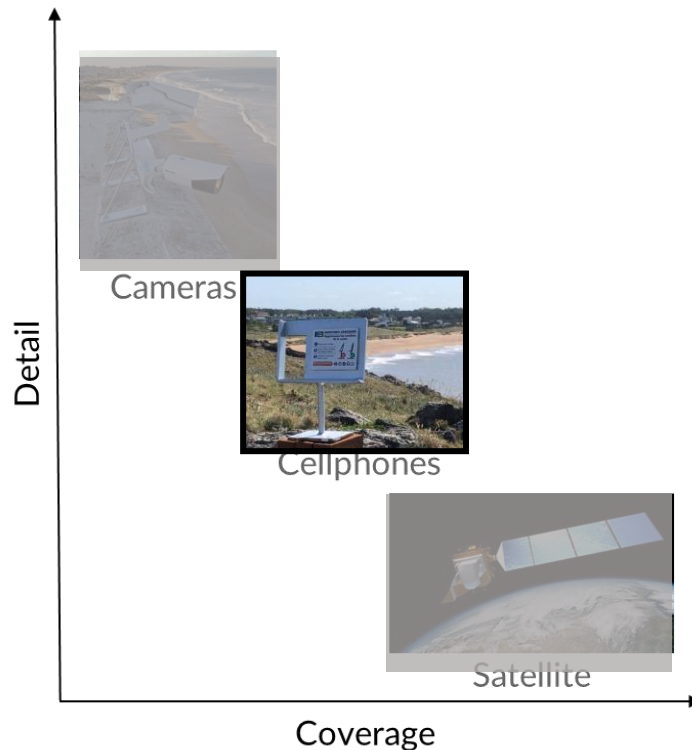
- 3 sources of information



# Study Area Motivation Databases Methodology Results Conclusions

## Remote monitoring of shoreline position

- 3 sources of information



## MONITOREO CIUDADANO LAS GRUTAS



Registremos los cambios en la costa:  
Tomá una foto y compartila



### 1 Fotografíar

Posicionamos el teléfono en el soporte

### 2 Compartir

Por redes sociales (en escala original, sin filtros e indicando fecha y hora) con el hashtag #CoastSnapUy o enviarla por mail o Whatsapp:

@coastsnapuy  
 @coastsnap\_uy  
 @coastsnap\_uy

092 957 622  
 coastsnapuy@gmail.com

### 3 Seguirnos en las redes

Para más información y ver los resultados



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Ministerio  
de Ambiente



Intendencia de Maldonado  
CONSTRUYENDO FUTURO

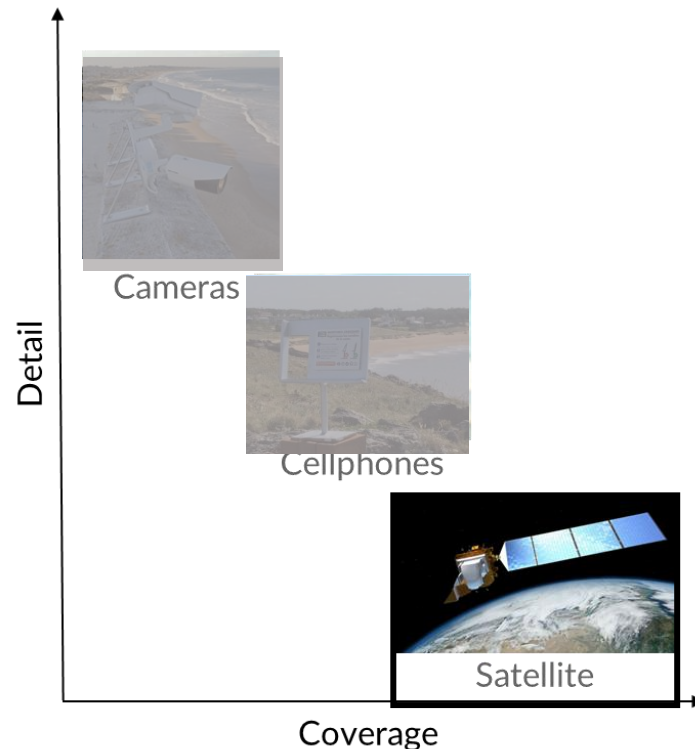
Basado en CoastSnap



# Study Area Motivation Databases Methodology Results Conclusions

## Remote monitoring of shoreline position

- 3 sources of information



- ❑ Based on the toolkit CoastSat (Vos et. al 2019)  
<https://github.com/kvos/CoastSat>.
- ❑ Uses satellite imagery available at Google Earth Engine (Landsat and Sentinel missions).
- ❑ The algorithm was adapted and applied to 90 beaches covering the entire national shoreline.
- ❑ Post-processing tools were developed to analyze results.
- ❑ Results were generated in transects each 100 m (alongshore).



## Databases:

# Remote monitoring of shoreline position



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# Study Area Motivation Databases Methodology Results Conclusions



I) Pair beaches with the nearest vitural buoy.

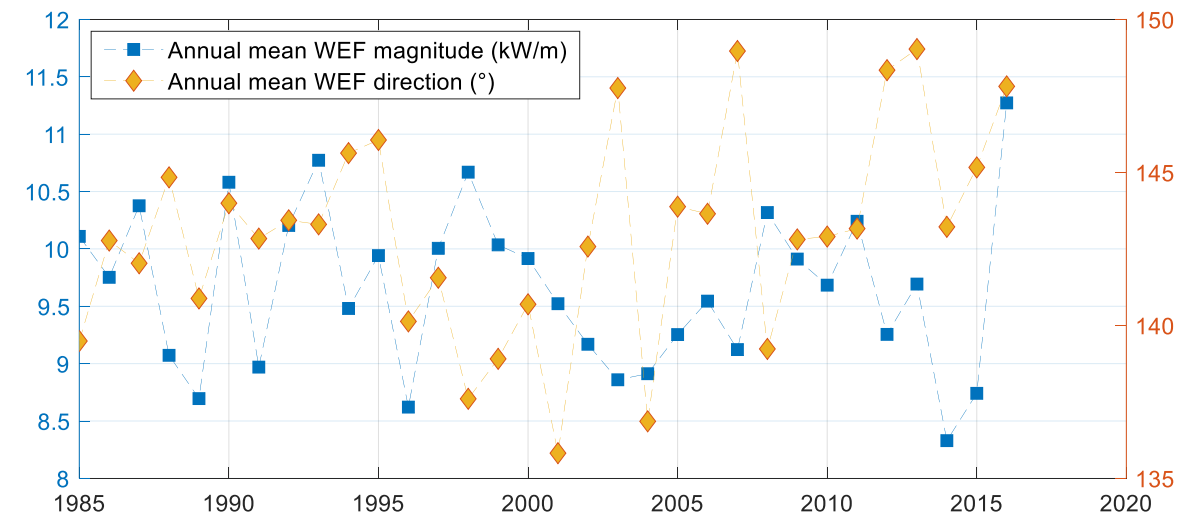
II) 2D spectra → WEF

$$WEF_x = \int_0^{2\pi} \int_0^{\infty} \cos(\theta) S(f, \theta) \cdot C_g(f, h) df d\theta$$

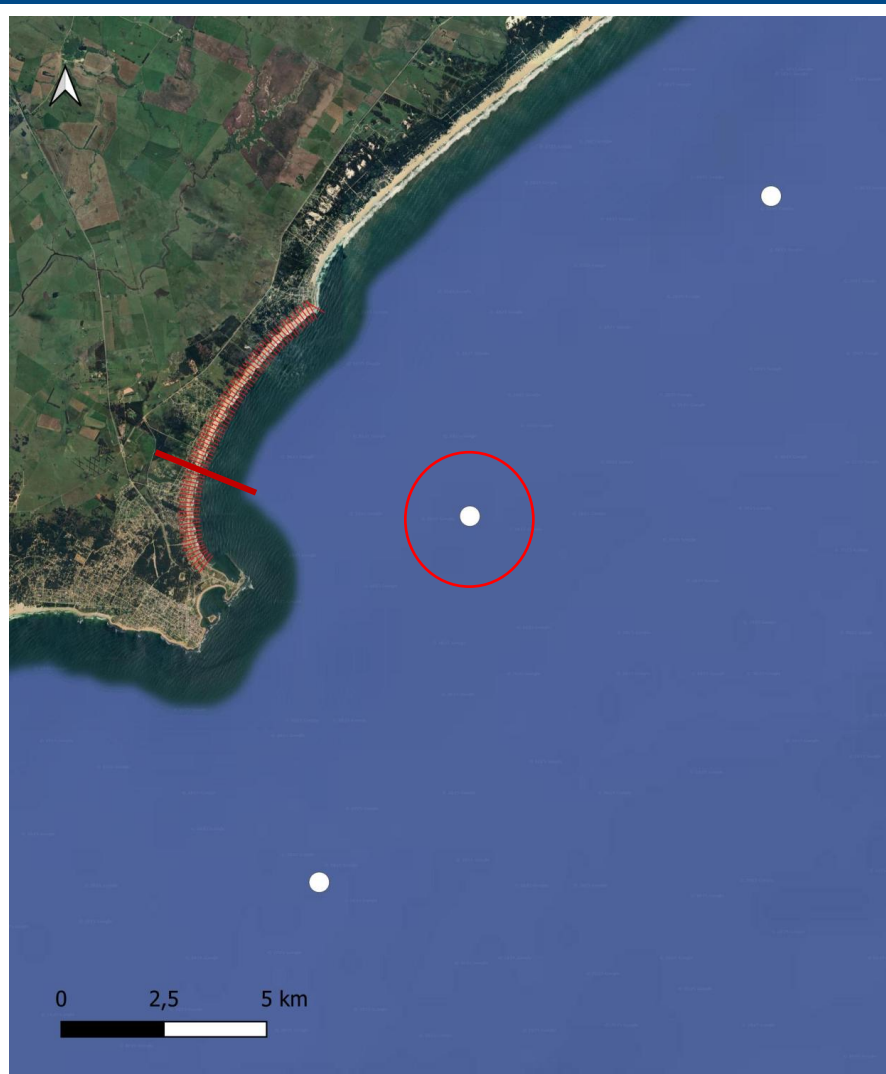
$$WEF_y = \int_0^{2\pi} \int_0^{\infty} \sin(\theta) S(f, \theta) \cdot C_g(f, h) df d\theta ;$$

$$\|WEF\| = \sqrt{WEF_x^2 + WEF_y^2}, \theta_{WEF} = \tan^{-1} \left( \frac{WEF_y}{WEF_x} \right)$$

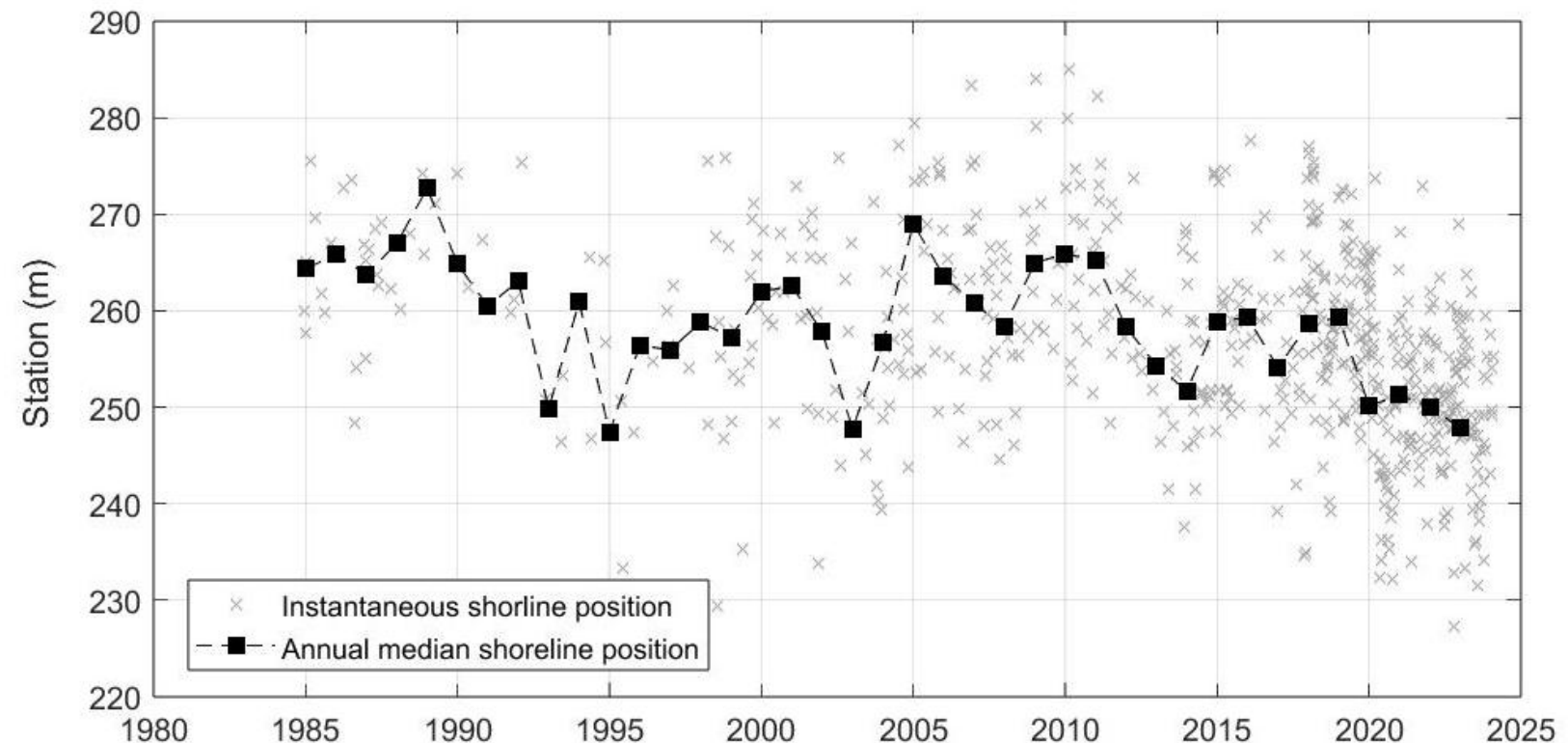
III) Annualize WEF



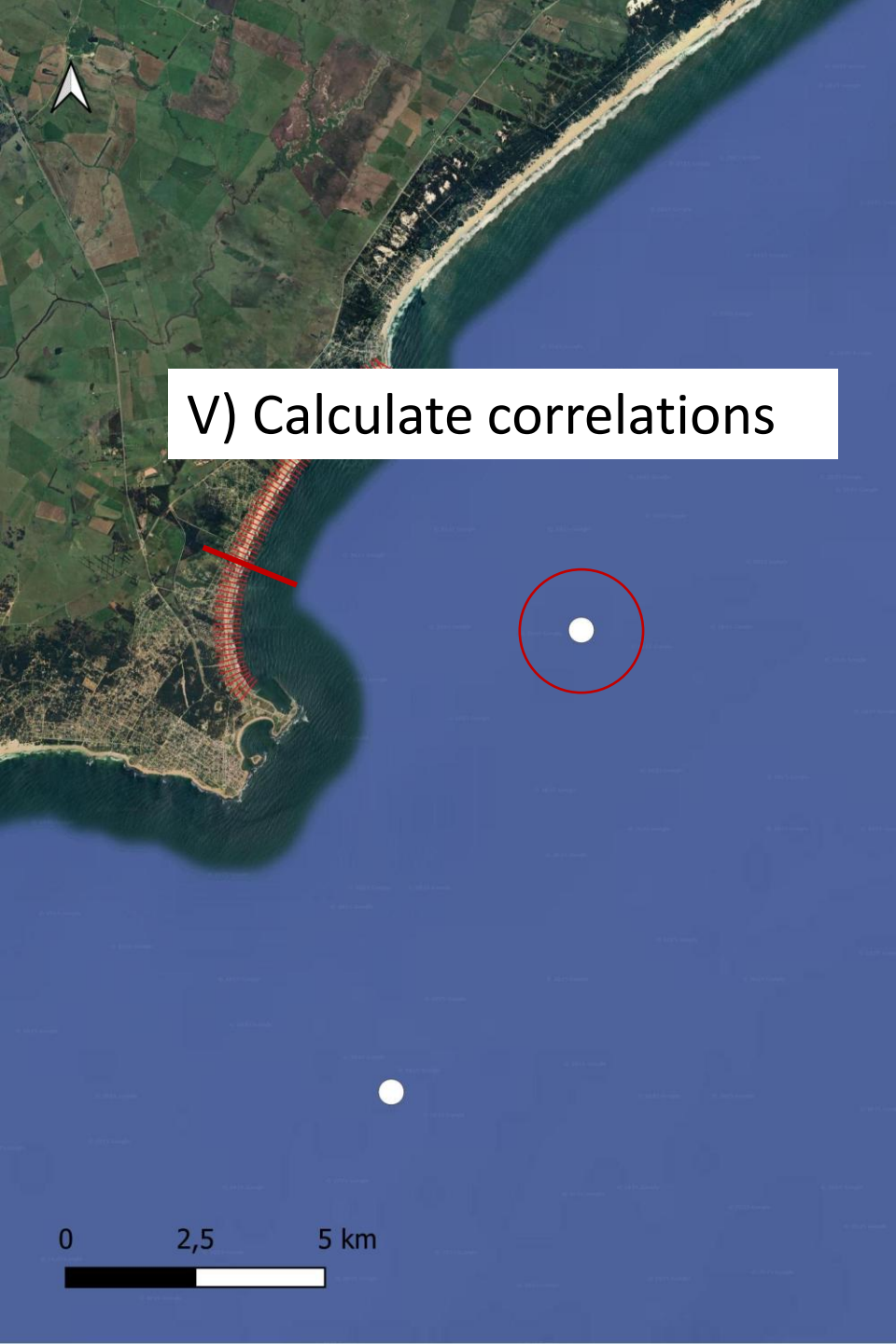
# Study Area Motivation Databases Methodology Results Conclusions



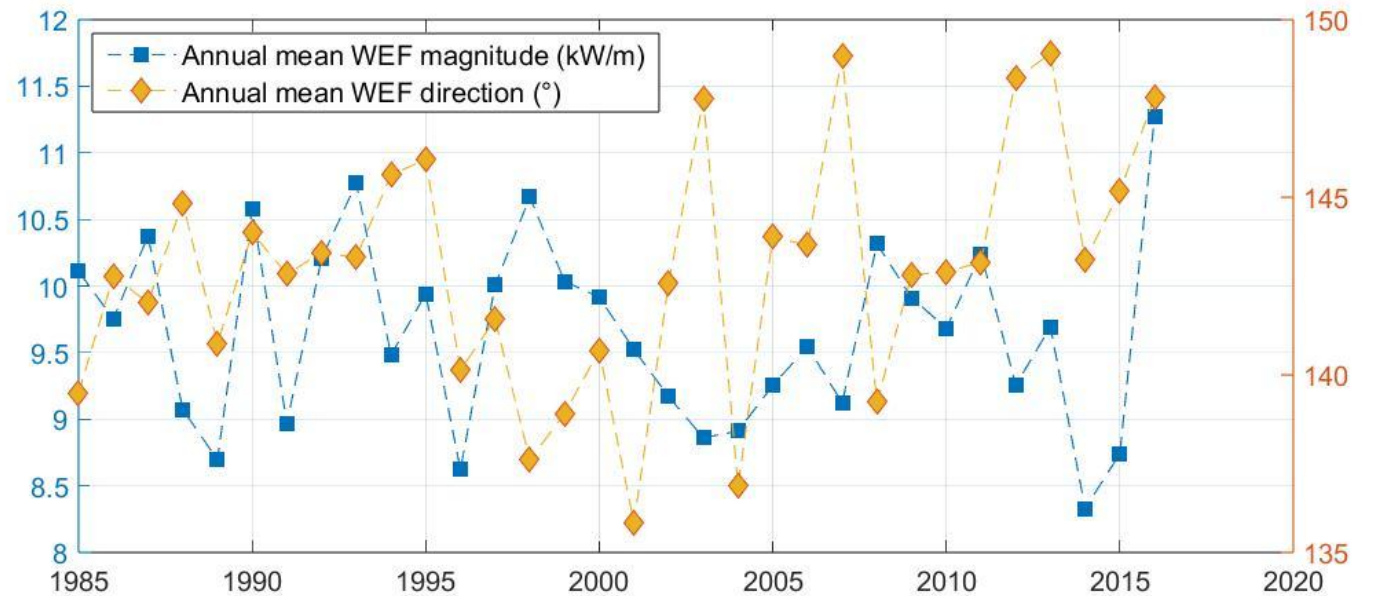
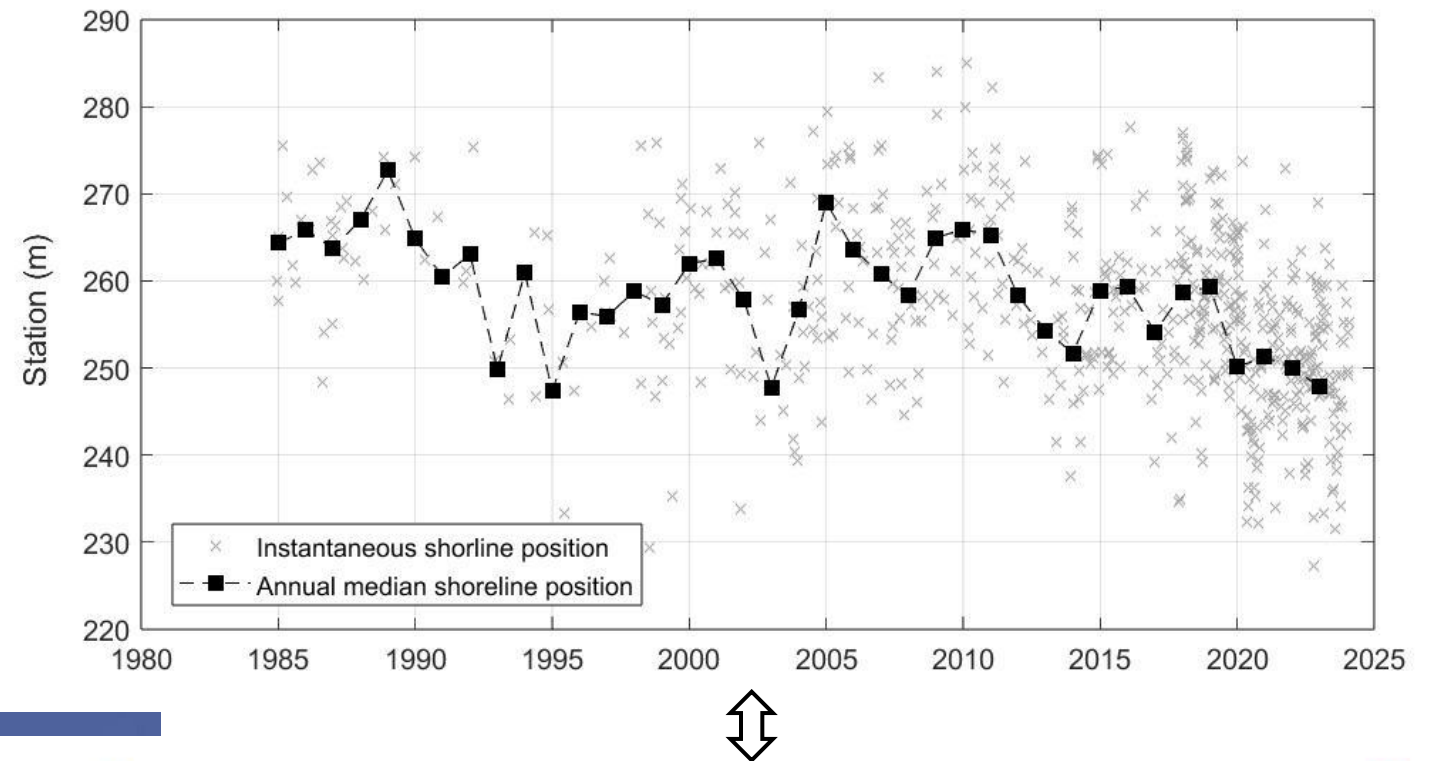
IV) Shorelines detected are traduced to stations along transects and the annual median is calculated.



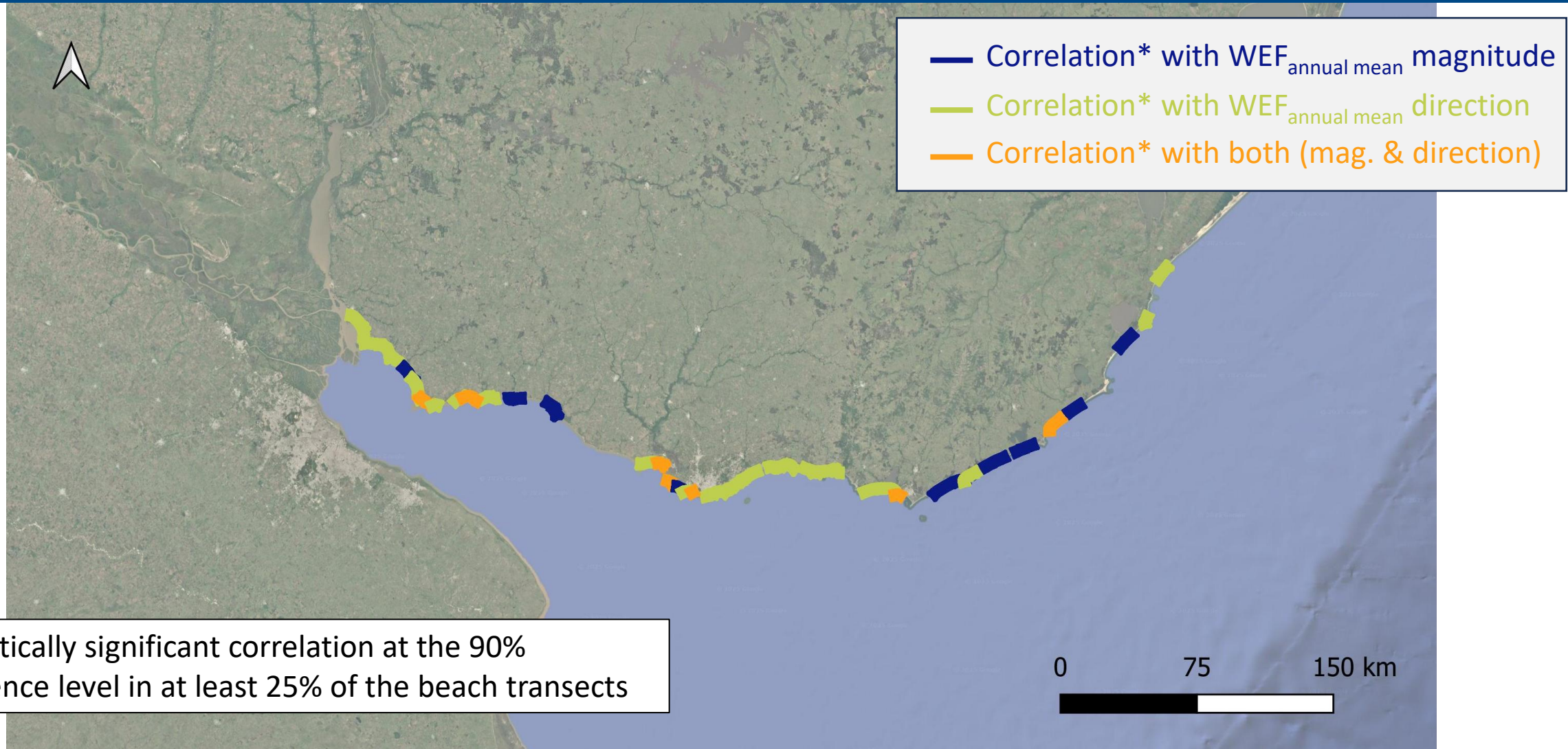




V) Calculate correlations



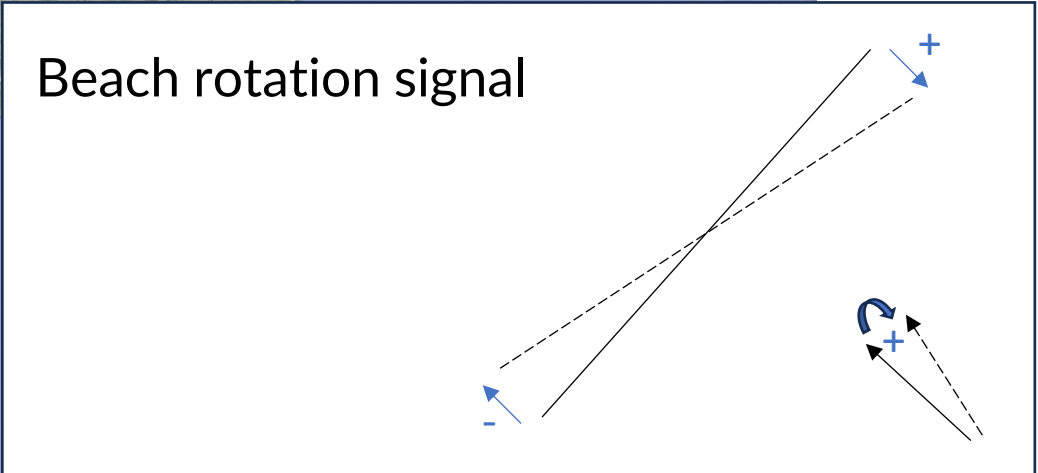
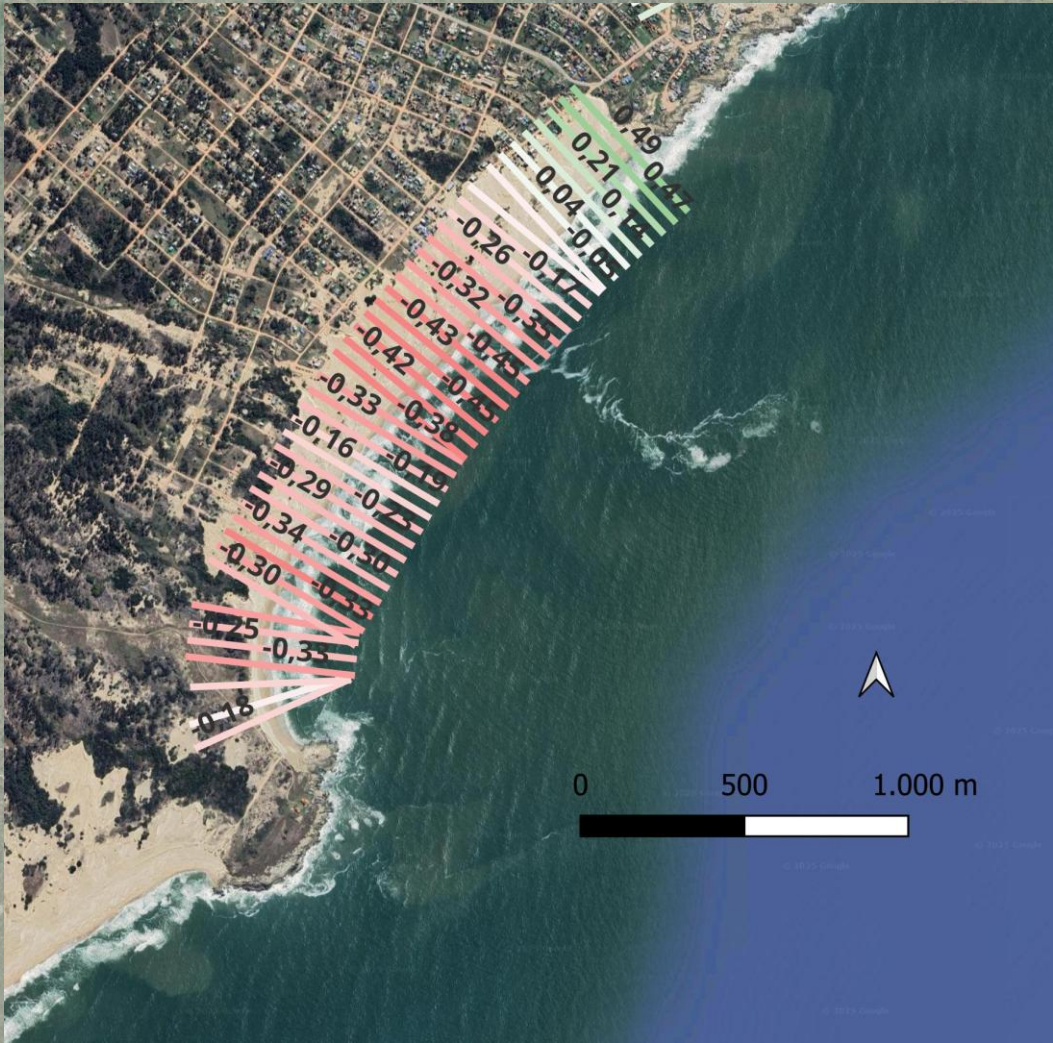
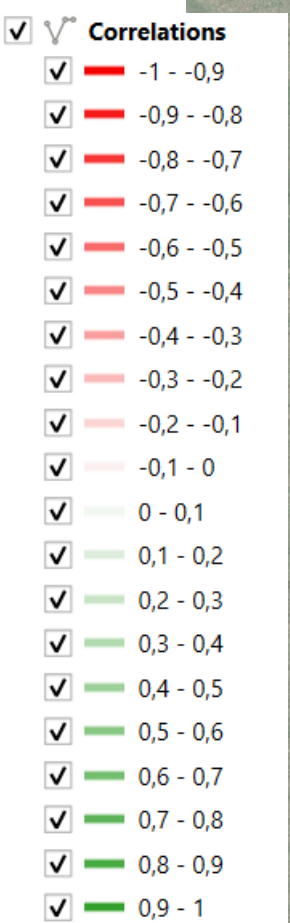
# Study Area Motivation Databases Methodology **Results** Conclusions





Study Area Motivation Databases Methodology Results Conclusions

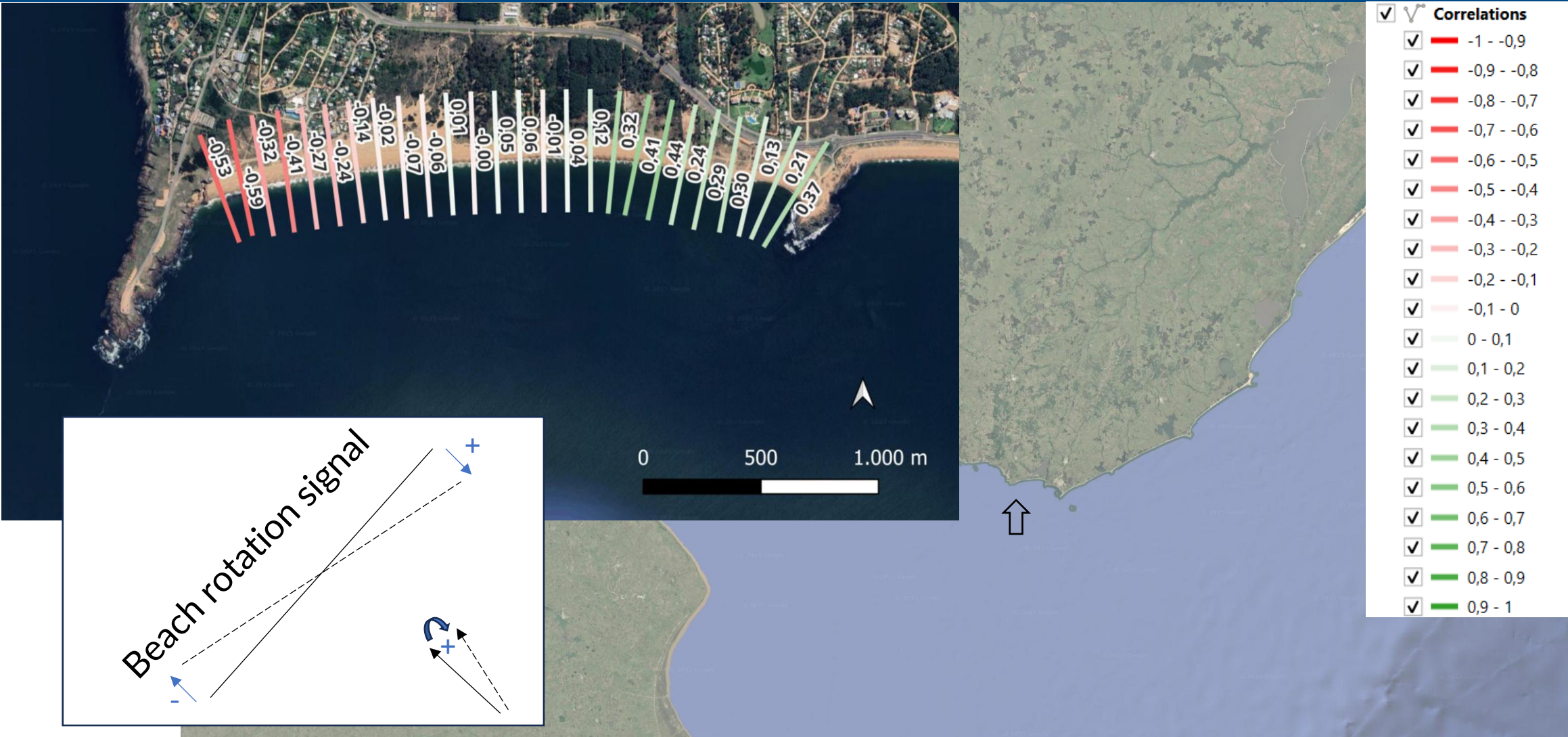
Correlation with WEF<sub>annual mean</sub> direction





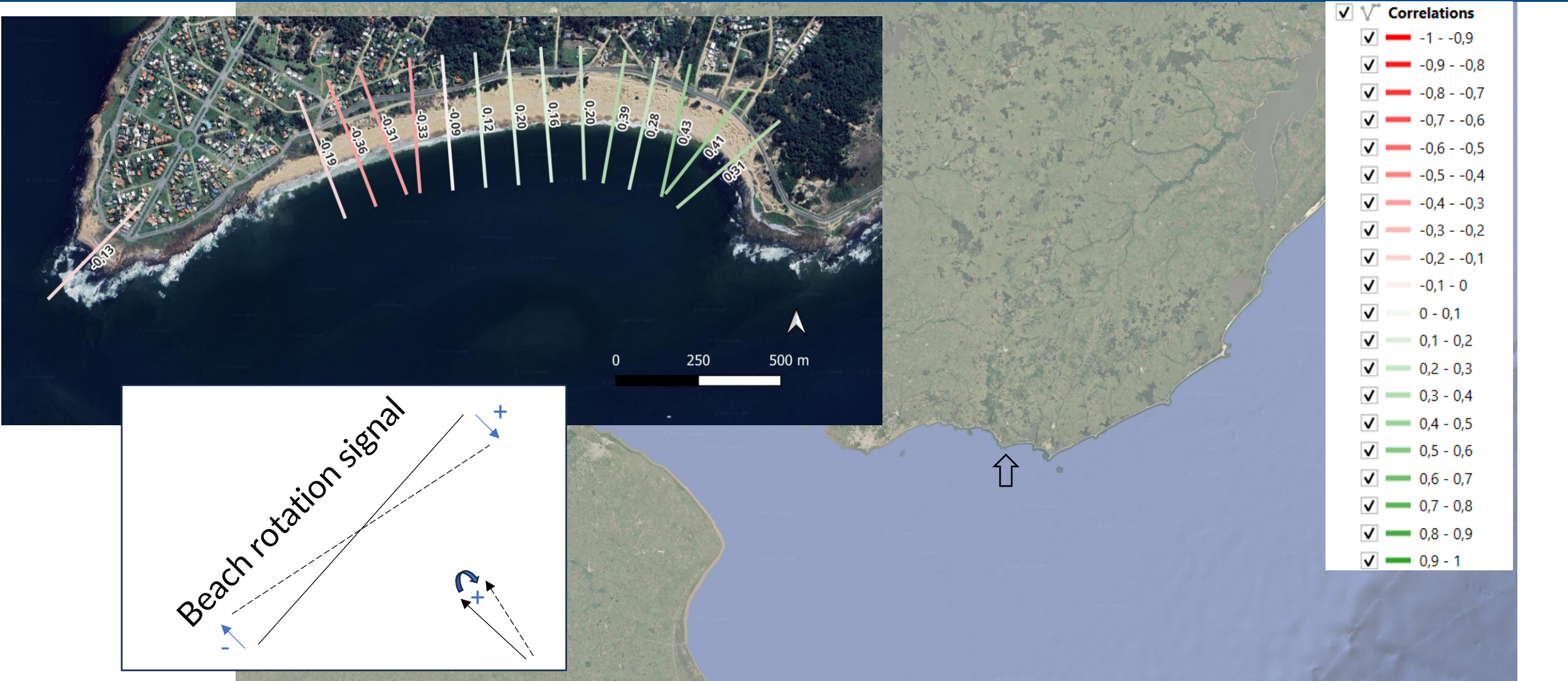
Study Area Motivation Databases Methodology Results Conclusions

Correlation with WEF<sub>annual mean</sub> direction



Study Area Motivation Databases Methodology Results Conclusions

Correlation with WEF<sub>annual mean</sub> direction

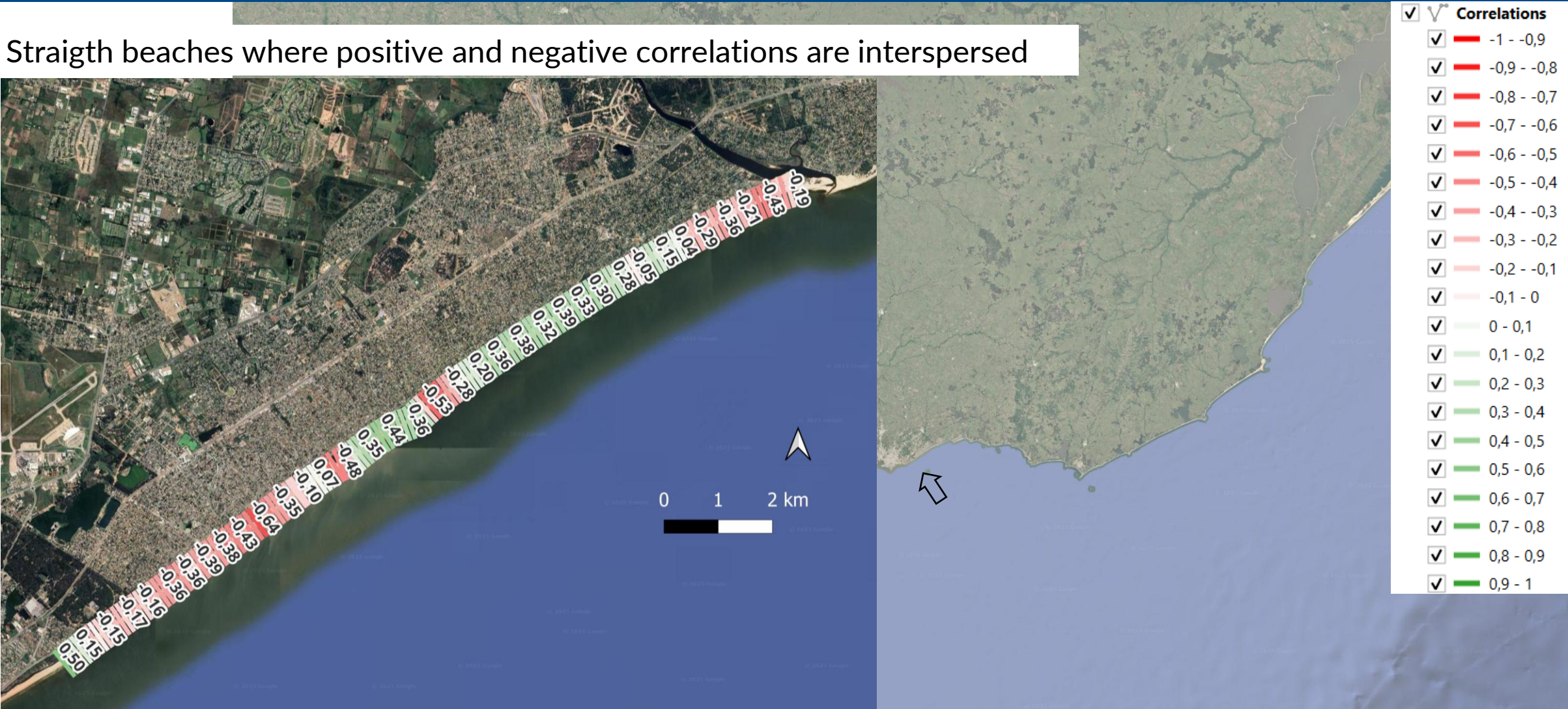




Study Area Motivation Databases Methodology **Results** Conclusions

**Correlation with WEF<sub>annual mean</sub> direction**

Straight beaches where positive and negative correlations are interspersed





Study Area Motivation Databases Methodology Results Conclusions

Correlation with WEF<sub>annual mean</sub> direction

Straight beaches where positive and negative correlations are interspersed

- ✓ V Correlations
- ✓

■

-1 - -0,9
- ✓

■

-0,9 - -0,8
- ✓

■

-0,8 - -0,7
- ✓

■

-0,7 - -0,6
- ✓

■

-0,6 - -0,5
- ✓

■

-0,5 - -0,4
- ✓

■

-0,4 - -0,3
- ✓

■

-0,3 - -0,2
- ✓

■

-0,2 - -0,1
- ✓

■

-0,1 - 0
- ✓

■

0 - 0,1
- ✓

■

0,1 - 0,2
- ✓

■

0,2 - 0,3
- ✓

■

0,3 - 0,4
- ✓

■

0,4 - 0,5
- ✓

■

0,5 - 0,6
- ✓

■

0,6 - 0,7
- ✓

■

0,7 - 0,8
- ✓

■

0,8 - 0,9
- ✓

■

0,9 - 1





Study Area Motivation Databases Methodology **Results** Conclusions

**Correlation with WEF<sub>annual mean</sub> magnitude**

Negative correlation on the most exposed beaches





Study Area Motivation Databases Methodology **Results** Conclusions

**Correlation with WEF<sub>annual mean</sub> magnitude**

Positive correlations on the less exposed beaches

- ✓ ☒ **Correlations**
- ✓

█

-1 - -0,9
- ✓

█

-0,9 - -0,8
- ✓

█

-0,8 - -0,7
- ✓

█

-0,7 - -0,6
- ✓

█

-0,6 - -0,5
- ✓

█

-0,5 - -0,4
- ✓

█

-0,4 - -0,3
- ✓

█

-0,3 - -0,2
- ✓

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-0,2 - -0,1
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- ✓

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0,6 - 0,7
- ✓

█

0,7 - 0,8
- ✓

█

0,8 - 0,9
- ✓

█

0,9 - 1





## Conclusions

Primary results from the cross-referencing of a wave hindcast and shoreline position databases were presented.

It is observed:

- ❖ Most of Uruguay's beaches (~65%) show a correlation between the inter-annual variation of shoreline position and the annual mean wave energy flux (magnitude, direction or both).
- ❖ Straight beaches with a pattern of alternating positive and negative correlations with  $WEF_{\text{annual mean}}$  direction.
- ❖ Negative (positive) correlation with  $WEF_{\text{annual mean}}$  magnitude were observed in the most (less) exposed beaches.

Future work:

Include other hindcasts, focus on higher frequency scales (e.g. seasonal), and explore other wave climate parameters (e.g. storminess).



# Thanks for your attention !!!

Acknowledgments:



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E INNOVACIÓN

Proyecto FMV\_1\_2023\_1\_176595