





A MACHINE LEARNING CLUSTERING ALGORITHM TO IDENTIFY GAPING BEHAVIOUR IN MYTILUS SPP UNDER CONTRASTING ENVIRONMENTAL CONDITIONS

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Green Aquaculture INtensification in Europe

- Support the ecological intensification of aquaculture
 - Increasing production and competitiveness of the industry
 - Ensuring sustainability and compliance with EU regulations
- Consortium
 - 10 academic and research institutes
 - 8 companies
 - 2 non-profit organisations

Research areas

- Production optimization
- Valorization of secondary inputs
- Sustainability assessment



<u>Impacts</u>

- Eco-intensification tools
- Professional developments



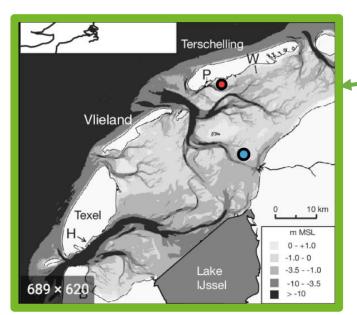




Venice

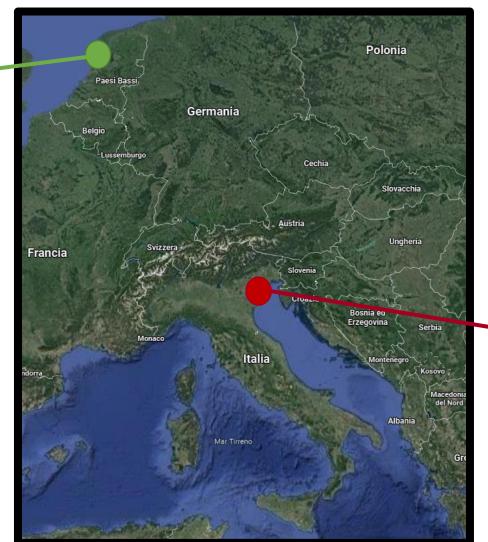






Wadden Sea

Mytilus edulis







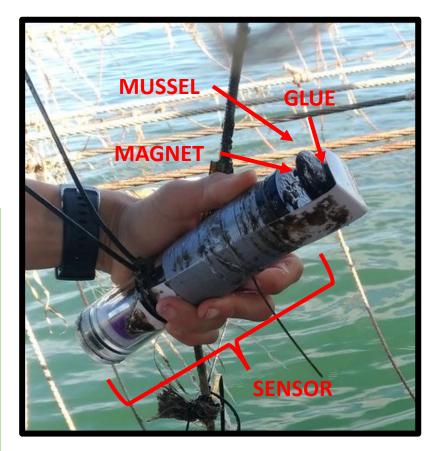


Sensors

- Venice: Multiparametric buoy (Tecnos SaS)
- Wadden sea: Logger (JFE Advantech)
- Both sites: Hall effect sensors (BIOPHYS)













Sampled data

VENICE

- **Water** (Ts = 12 min)
 - Temperature
 - Oxygen
 - Chlorophyll
 - Turbidity
- Mussel (Ts = 5min)
 - Gaping
- Time period: > 6 months

WADDEN SEA

- Water (Ts = 5 min)
 - Temperature
 - Chlorophyll
 - Turbidity
- Mussel (Ts = 1min)
 - Gaping
- Time period: One month







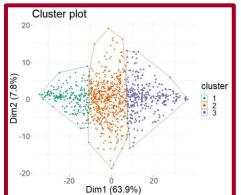
Daily Gaping

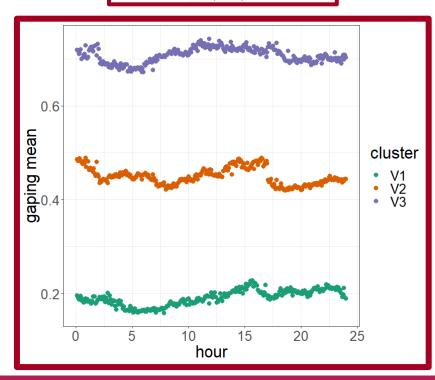
- Input data
 - Gaping timeseries (from 0 = min to 1 = max)
 - Variable to classify: daily dynamics

- Method
 - K-means clustering algorithm (unsurpervised ML)
 - Optimal number of clusters: Elbow method + Visual analysis
 - 3 clusters









3 daily average gaping value

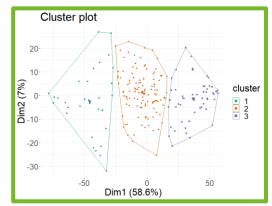
• 70 %

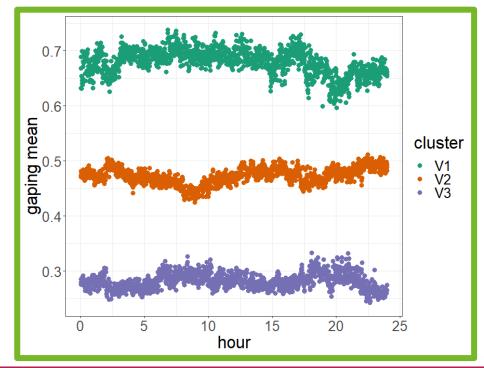
• 45 %

• 20 %



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Gaping + Water

- Input data
 - Gaping and environmental data (from 0 = min to 1 = max)
 - Variable to classify: daily means

- Methods
 - K-means clustering algorithm (unserpervised ML)
 - Optimal number of clusters: Elbow methods + Visual analysis
 - 3 clusters



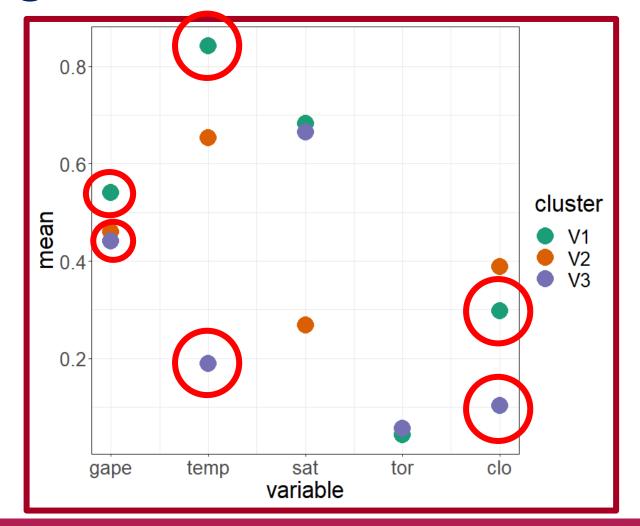




Gaping + Water

VENICE

- Wider gaping
 - Higher T°C
 - Medium Chla
- Narrower gaping
 - Lower T°C
 - Low Chla





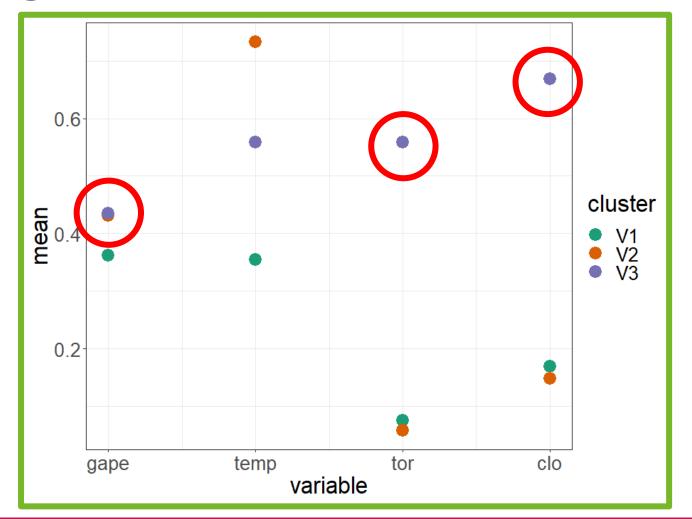




Gaping + Water

WADDEN SEA

- Wider gaping
 - High Chla
 - High Turbidity









Detrended gaping

- Input data
 - Daily detrended gaping data (= daily dynamics daily average)
 - Variable to classify: daily detrended dynamics

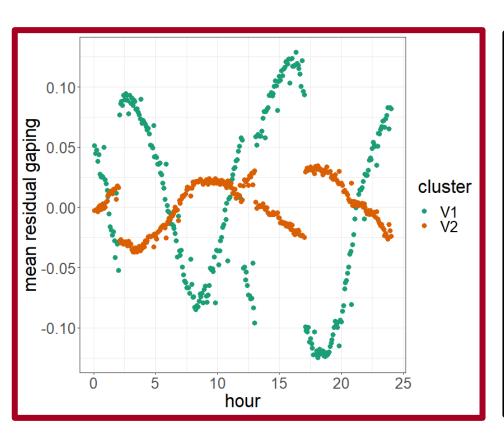
- Methods
 - K-means clustering algorithm (unserpervised ML)
 - Optimal number of clusters: Elbow methods + Visual analysis
 - 2 clusters



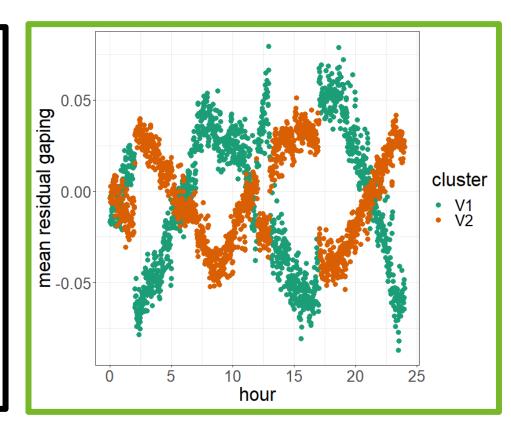




Detrended gaping



2 oscillation amplitudes as already observed by Bertolini et al. (2021) using different method (wavelet analysis)



AE 2021 - 5th October 2021







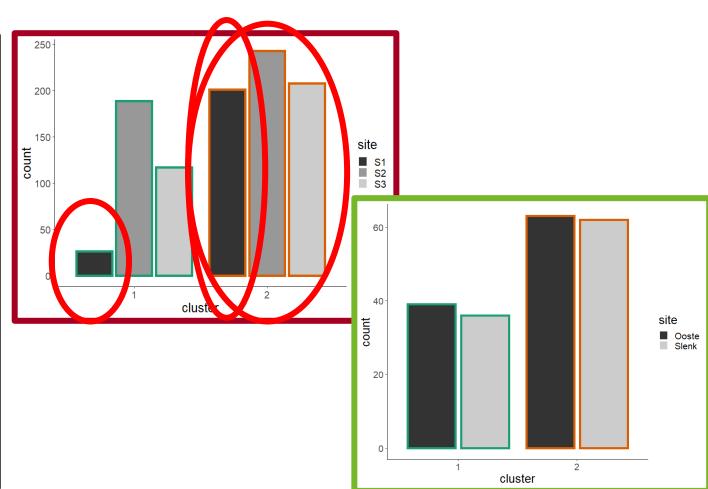
Detrended Gaping – Site effect

Venice

- Higher number of days with a lower amplitude of oscillation
- Particularly evident in S1 (near to sea inlet)

Wadden Sea

 No clear site effect or clear predominance of one behaviour









Conclusions

Results

- 3 clusters of daily mean value
- Different environmental drivers => Different growing/sampling conditions
- 2 daily detrended patterns => already observed using another method

Potential applications

- Improvement of bioenergetic models
- Precision shellfish farming: using real-time gaping monitoring in order to better forecast growth, and then productivity

AE 2021 - 5th October 2021

Thank you for your attention!

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