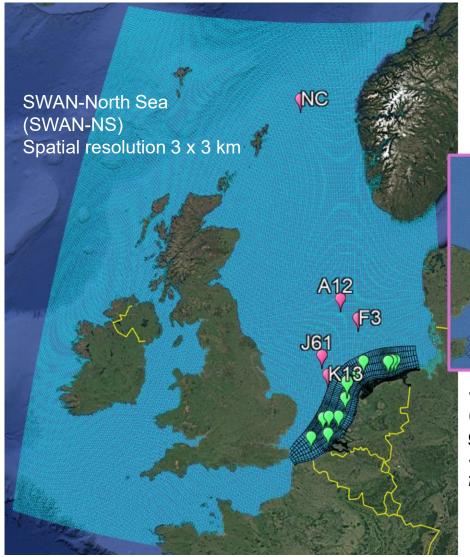


Lessons-learned: Evaluation of the operational spectral wave model with machine learning adjustments for spectral shape

Joana van Nieuwkoop, Menno de Ridder, Joost den Bieman, Marisol Irias Mata

SWAN North Sea, operational wave models RWsOS

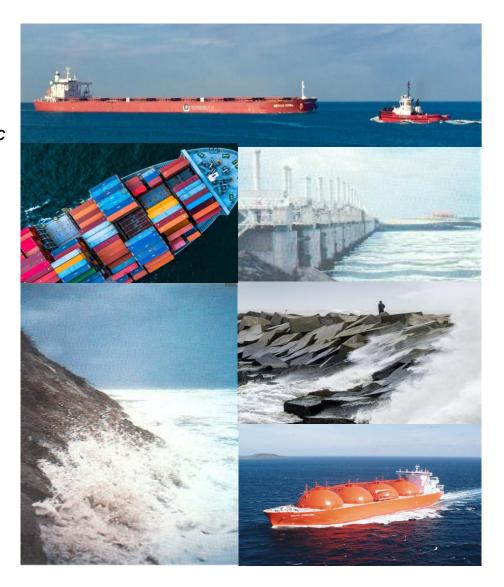


Models use:

- Harmonie wind
- ECMWF WAM boundary conditions
- D-Flow FM hydrodynamic conditions

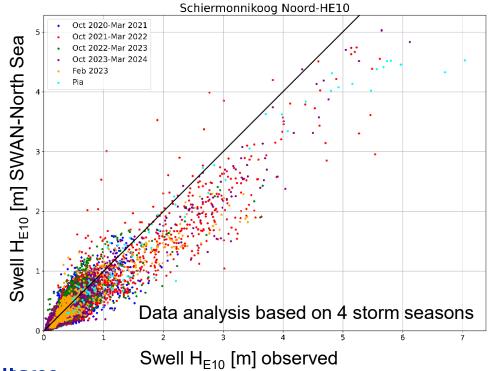


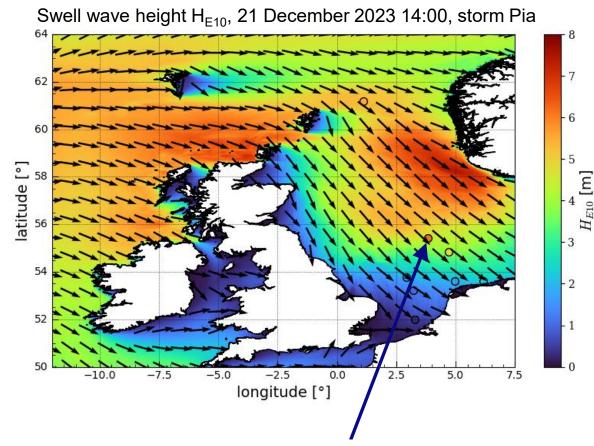
SWAN-Kuststrook (SWAN-KS) – every 10th gridline Spatial resolution varies from 35 m to 2.6 km



Performance operational wave models

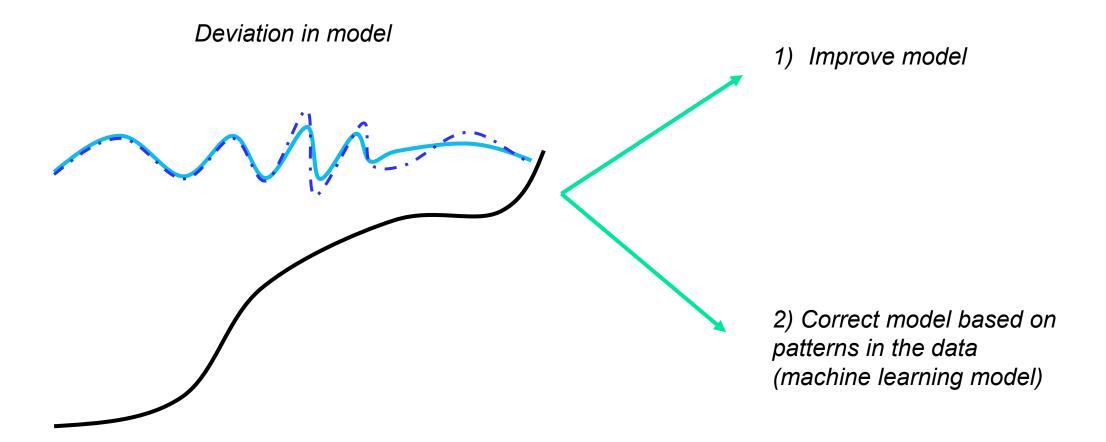
- Generally good performance for significant wave height
- For swell wave height large scatter
- Large swell events from the northwest severely underestimated

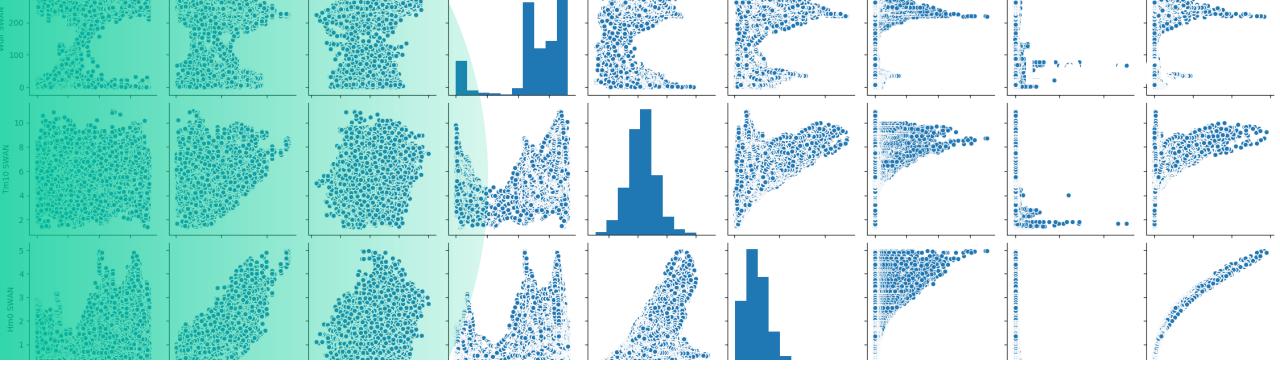




Coloured dots: wave measurements swell wave height

Options to improve results





Model correction

Machine Learning (XGBoost) to improve the SWAN-Kuststrook results

Correct the SWAN-results with data driven model

Training XGBoost model based on 15 months of data

Wind velocity

Wind direction

Water level

Significant wave height

Spectral wave period

Deep water wave steepness

Wave direction

Energy density in frequency bin

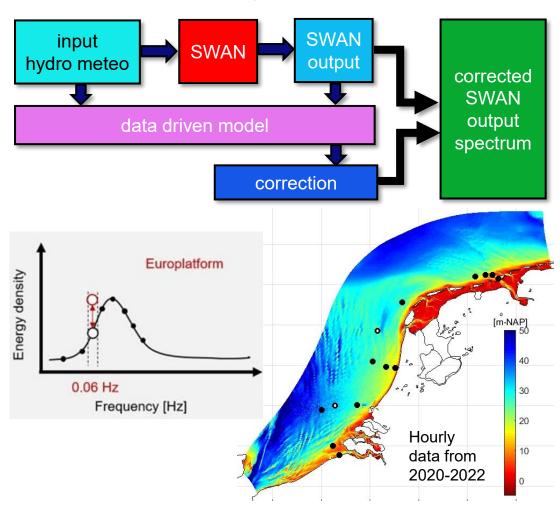
Main wave direction in frequency bin

Difference between wave and wind direction

as used in SWAN
as used in SWAN
as used in SWAN
predicted by SWAN

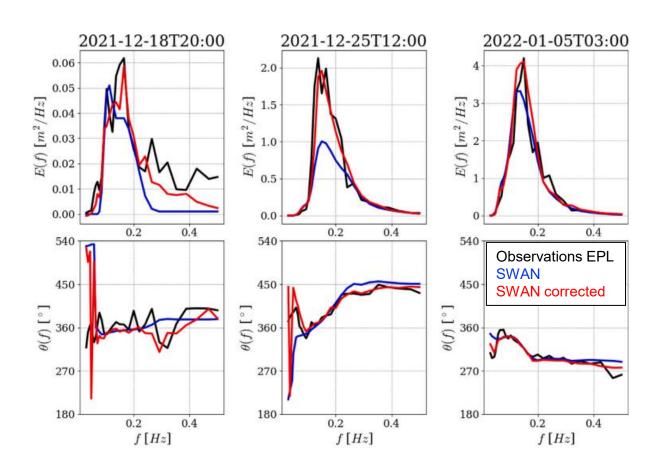
$$\Delta E(f) = \frac{E_{obs}(f) - E(f)}{E_{SWAN,total}} , \quad \Delta \theta(f) = \frac{\theta_{obs}(f) - \theta(f)}{360}$$

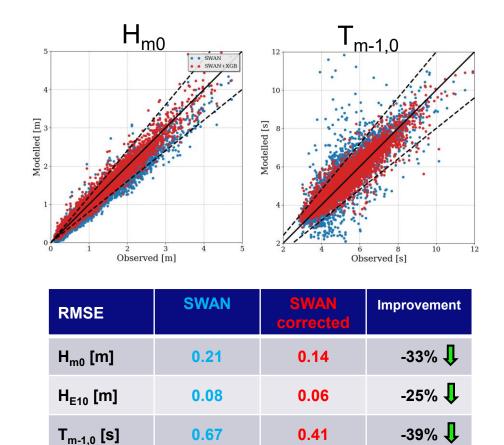
Operational machine learning correction



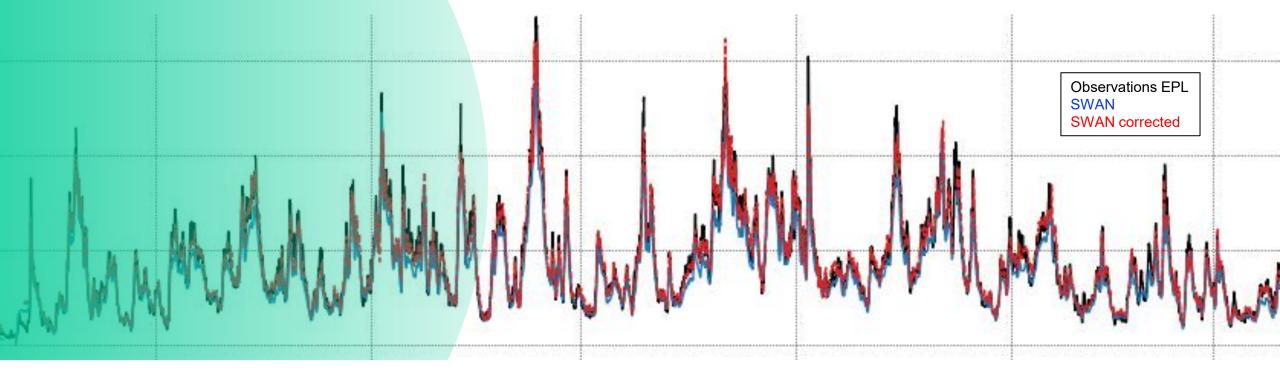
Bieman et al (2023). Hybrid modelling to improve operational wave forecasts by combining process-based and machine learning models

SWAN results after correction





Decision in 2022: implement machine learning correction in operational forecast suite Delft FEWS

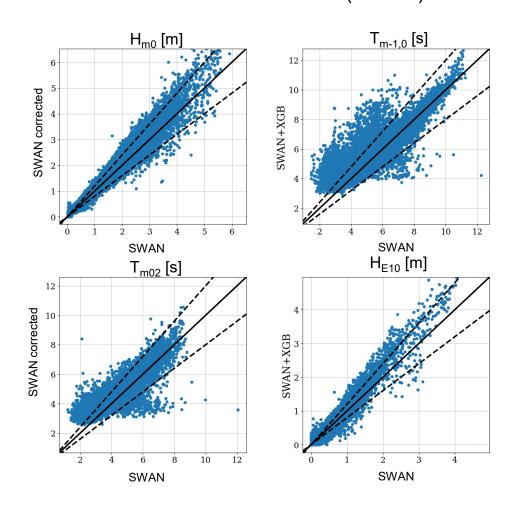


Evaluation after 1-year operational results

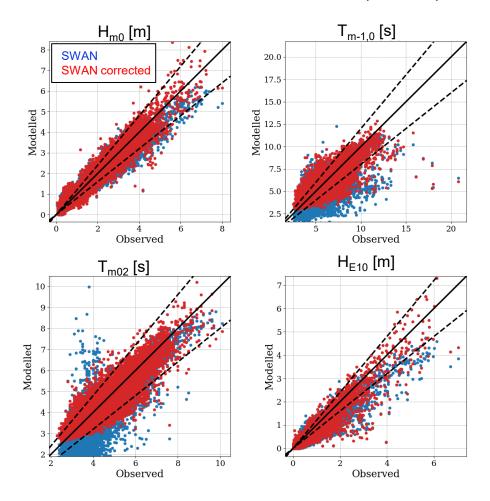
August 2023 – August 2024

Evaluation operational Machine Learning model

SWAN-ML versus SWAN (all locs)



SWAN-ML/SWAN versus obs (all locs)



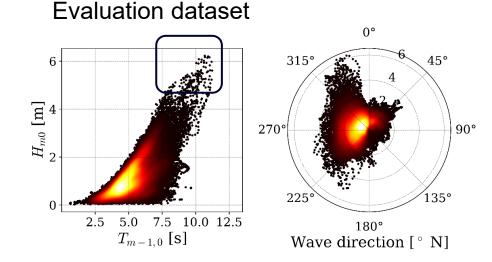
Performance comparison evaluation periods

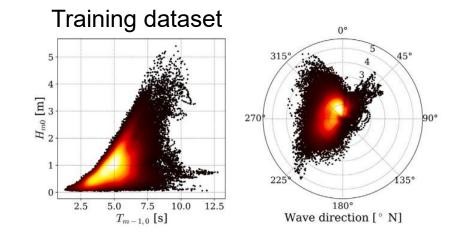
	Evaluation training period 2020 - 2022			Operational evaluation period 2023 – 2024		
RMSE	SWAN	SWAN corrected	Improvement	SWAN	SWAN corrected	Improvement
H _{m0} [m]	0.21	0.14	-33%	0.23	0.18	-22%
H _{E10} [m]	0.08	0.06	-25%	0.13	0.11	-15%
T _{m-1,0} [s]	0.67	0.41	-39%	0.65	0.53	-12%
Wavedir [º]	50	38	-25%	34	29	-15%

- Machine learning models improves the SWAN-Kuststrook results
- But operational machine learning performance suboptimal

Reasons deterioration in performance

- There have been several model changes (bug in wave boundary conditions, change model physics, transition wind input: Harmonie model version)
- The machine learning model has been trained on 15 months of data → limited amount extreme conditions and swell conditions.





Lessons-learned

- A hybrid approach (in which a numerical model is corrected by machine learning) requires regular retraining if the numerical model changes.
- Aim for a sufficient long training dataset, ~ length in our case of minimal 5 years to include 'more extreme' conditions.
- In case the machine learning model is run in an operational setting, it is important to check the operational implementation.
- It is important to think about the data management system for operationalization.
- Take along the client/ users in 'trusting' the model outcomes of the machine learning model and the added effort when retraining is necessary.



Questions, discussion



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