



# Design optimization of floating offshore wind farms using a steady state motion and flow model

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## 1. The background

- Floating wind is set to grow tremendously in the near future.
- Large FOWFs will be designed and constructed.
- Design optimization of these FOWFs will be an important task.
- Most of the current studies on FOWF design ignore the motion of the FOWTs.
- A fast model that can capture the mean (and essential) effects introduced by platform motion will be needed for energy yield assessment of FOWFs.
- How much difference will floating cause for the power production and wake flow needs to be evaluated.



## 2. The challenge





# 3. The modelling

- Python package developed by DTU Wind Energy
- Open source and available at PyPi and
  <u>https://gitlab.windenergy.dtu.dk/TOPFARM/PyWake</u>
- Simulates static flow in wind farms
- Calculate AEP and flow maps
- Modular, flexible and very fast
- Used by +35 companies











### **Turbine model**

- IEA 15 MW reference turbine •
- WindCrete floating spar buoy ٠
- 3 mooring lines •
- 200 m water depth •



### DTU Database from HAWC2 Simulations





### Surrogate inputs and outputs

# Inputs

- Wind speed
- Wind direction
- Current speed
- Current-wind
  misalignment

# Outputs

- Downwind/crosswind displacement
- Tilt/yaw rotation
- Power and thrust

Note: The procedure for modelling steady state motion and flow of floating wind farms using PyWake will be presented in TORQUE conference in May 2024, by Riccardo Riva with the title "**Incorporation of floater rotation and displacement in a static wind farm simulator**". Full paper will come out earlier.

### **Computational speed comparison**





# 3. The optimization



### **Design of Hywind Scotland**

(source: https://www.equinor.com/energy/hywind-scotland)

### **Problem formulation**

**Objective:** 

max AEP

**Design variable:** 

$$\boldsymbol{L} = [x_1, x_2, \dots, x_{N_{wt}}, y_1, y_2, \dots, y_{N_{wt}}]$$

### **Constraints:**

Wind farm boundary:  $(x_i, y_i) \in S_{feasible}$ , for  $i = 1, 2, ..., N_{wt}$ 

Minimal distance between FOWTs:

$$\sqrt{(x_i - x_j)^2 + (y_i - y_j)^2} \ge D_{min},$$
  
for  $i, j = 1, 2, ..., N_{wt}$  and  $i \ne j$ 

## DTU Optimization algorithm: Random Search



- Random search is a wind farm layout optimization algorithm first proposed by Feng and Shen [1].
- Simple and easy to implement.
- Great performance in various wind farm optimization applications [3, 4].

Feng, J., & Shen, W.Z. (2015). Solving the wind farm layout optimization problem using random search algorithm. Renewable Energy, 78, 182-192.
 Feng, J., & Shen, W.Z. (2017). Design optimization of offshore wind farms with multiple types of wind turbines. Applied Energy, 205, 1283–1297.
 Brogna, R., Feng, J., Sørensen, J. N., Shen, W. Z., & Porté-Agel, F. (2020). <u>A new wake model and comparison of eight algorithms for layout optimization of wind farms in complex terrain</u>. Applied Energy, 259, 114189.

# 4. The metocean condition

### Havbredey FOWF site in Scotland

- Climate Forecasting System Reanalysis
- DHI North Europe Spectral Wave model
- DHI 3D Hydrodynamic UK/North Sea model

Dataset	Source	Grid cell size	Temporal res.	Temporal extent
Winds	CFSR	~23 km	Hourly	2001-2021
Waves	SW <sub>NE</sub>	~3-30 km	Hourly	2001-2021
Water level	HD <sub>UKNS</sub>	~2-12 km	30 min	2001-2021
Depth- averaged	HD <sub>UKNS</sub>	~2-12 km	30 min	2001-2021
Current profiles	HD <sub>UKNS</sub>	~2-12 km	30 min	2018



Lon	Lat	Depth	Grid cell size HD <sub>uкns</sub>	Grid cell size SW <sub>NE</sub>
-5.58093°	58.84328°	85.5 m	~ 6 km	~13 km



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### **Original wind condition (** $u_{mean} = 13.58 \text{ m/s}$ **)**











- A representative current condition is selected.
- Current speed: 0.25 m/s
- Current direction: 60 deg
- Wind shear exponent: 0.14
- Turbulence intensity: 0.07



### **Reduced wind condition (** $u_{mean} = 7.29 \text{ m/s}$ **)**







### **Original wind condition**







AEP\_gross = 1073.7503 GWh

Consider as fixed:

AEP\_initi = 1050.1161 GWh, CF = 72.65 % AEP\_optim = 1055.2304 GWh, CF = 73.01 % wake\_loss\_initi = 2.2011 % wake\_loss\_optim = 1.7248 %

AEP increase percentage = **0.4870 %** 

Consider as floating: AEP\_initi = 1047.9004 GWh AEP\_optim = 1053.9225 GWh AEP increase percentage = **0.5747 %** 



### **Reduced wind condition**

 $(u_{mean} = 7.29 \text{ m/s})$ 





#### Evolutionary history of optimization

### $AEP_gross = 582.0226 GWh$

Consider as fixed: AEP\_initi = 545.1751 GWh, CF = 37.72 % AEP\_optim = 552.2446 GWh, CF = 38.21 % wake\_loss\_initi = 6.3309 % wake\_loss\_optim = 5.1163 % AEP increase percentage = **1.2967 %** 

Consider as floating: AEP\_initi = 543.5741 GWh AEP\_optim = 551.8396 GWh AEP increase percentage = **1.5206 %**  DTU

y [m]

### 23 turbines under original wind condition

Floating: 2167.33 GWh Floating: 2185.24 GWh Total AEP: 2188.06 GWh Total AEP: 2172.32 GWh 96.0 6000 6000 13 95.6 5000 5000 13 - 95.5 95.4 4000 4000  $\bigcirc^{21}$ 12  $\bigcirc^3$ 12 21 - 95.0 95.0 [ 94.5 [ 94.5 16 95.2 [4 95.2 [9] 3000 3000 16 y [m] **O**<sup>20</sup>  $\bigcirc^2$ 11 20 2000 2000 15 95.0  $\bigcirc^{10}$ 19  $\bigcirc^{19}$  $\bigcirc$ <sup>1</sup> 10 1000 1000 14 94.8 11 - 94.0 0 0 94.6 93.5 -1000-10002000 3000 4000 5000 6000 -10000 1000 2000 3000 4000 5000 6000 -10000 1000 x [m] x [m]



Evolutionary history of optimization

### $AEP_gross = 2245.1143 GWh$

Consider as fixed: AEP\_initi = 2172.3154 GWh, CF = 71.88 % AEP\_optim = 2188.0640 GWh, CF = 72.40 % wake\_loss\_initi = 3.2426 % wake\_loss\_optim = 2.5411 % AEP increase percentage = **0.7250 %** 

Consider as floating: AEP\_initi = 2167.3324 GWh AEP\_optim = 2185.2427 GWh AEP increase percentage = **0.8264 %** 

### 46 turbines under original wind condition





Evolutionary history of optimization

### $AEP_gross = 4490.2287 GWh$

Consider as fixed:

AEP\_initi = 4294.1789 GWh, CF = 71.04 %

AEP\_optim = 4339.3799 GWh, CF = 71.79 %

wake\_loss\_initi = 4.3661 %

wake\_loss\_optim = 3.3595 %

AEP increase percentage = **1.0526 %** 

# 6. Conclusions

- A methodology to account for platform motion is developed with surrogate model.
- Fast calculation of FOWF static flow and AEP can be achieved using PyWake.
- For the considered scenarios, optimization based on fixed version of modelling is feasible, since the floater displacements and motion are limited.
- With better wind condition, the relative wake loss is lower, thus, the potential of AEP improvement through layout optimization is also lower.
- For the high wind site, the relative importance of optimization of the other aspects, such as mooring systems, cables, etc. will become higher.
- More research is needed.



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EUDP C









