

How does risk aversion shape overplanting in the design of offshore wind farms?

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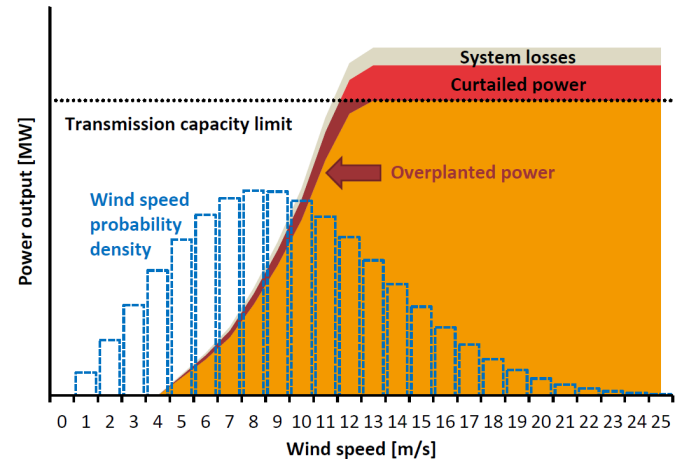
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Motivation

- Farms subjected to a maximum export capacity agreed with the TSO
- Generators can export up to their contracted maximum export capacity
- Majority of the time offshore wind farms are not generating at full power
- Can overplanting result in better overall economics despite power output being curtailed at generations' peaks?

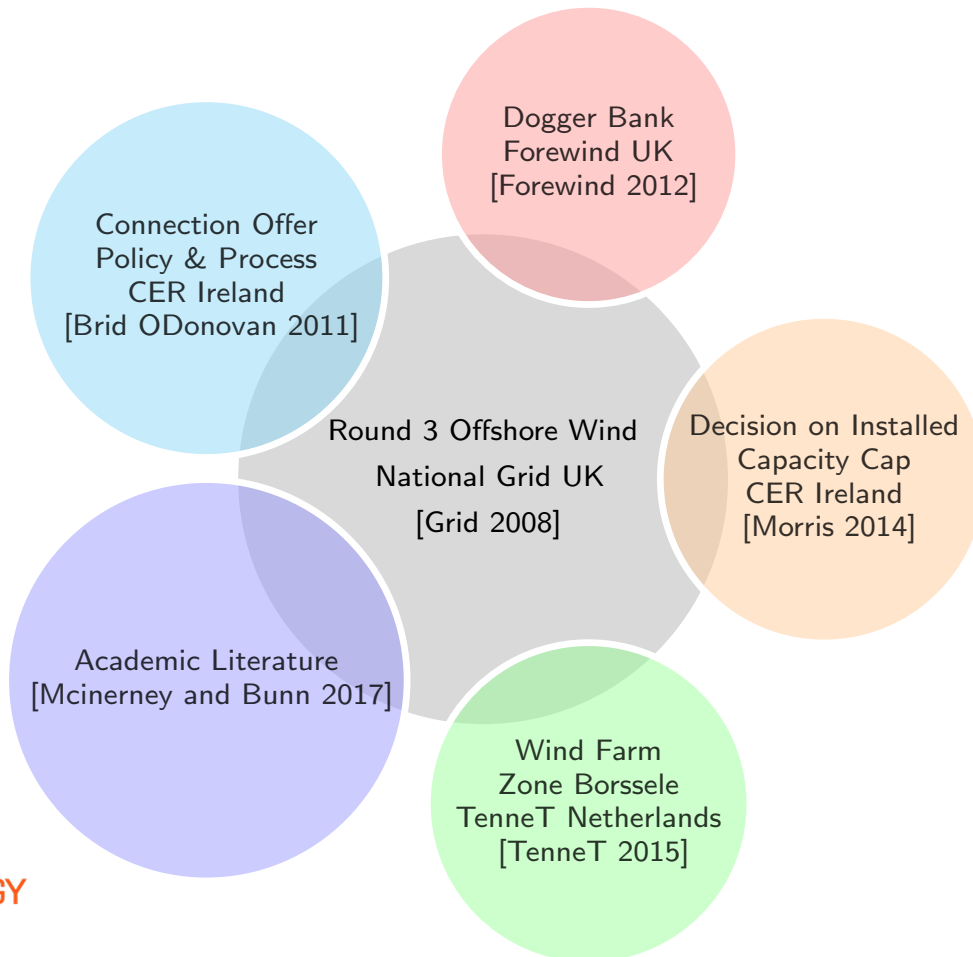


[Wolter et al. 2016]

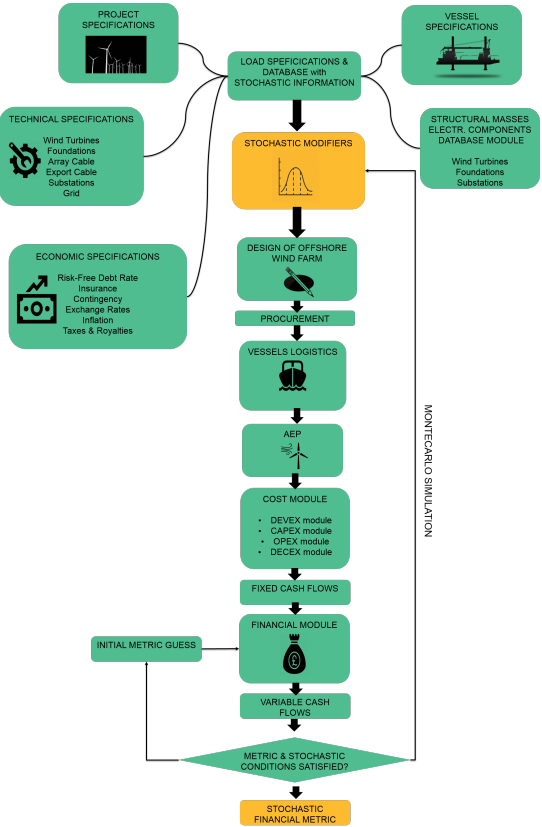
Overplanting

Over installing the offshore wind capacity to the fixed electrical infrastructure

Overplanting Studies



Offshore Wind Cost Modelling Tool



Characteristics

- **Aim** : rapidly evaluate the financial performance of a farm
- **Inputs** : project specifications, technology choices and market trends
- **Outputs** : financial metrics based on LCOE
- **Structure** : 4 main modules - Design, Cost, Financial and Stochastic
- **Stochastic Framework**: Quantitative uncertainty management, Double loop Monte Carlo Simulation - inner loop within AEP

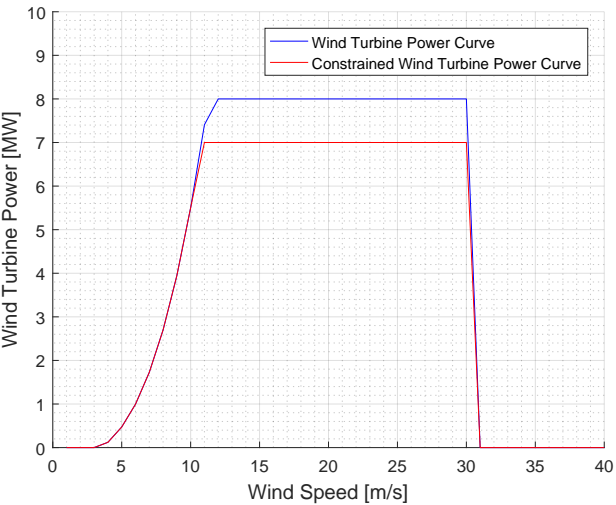
Factors Affecting Overplanting



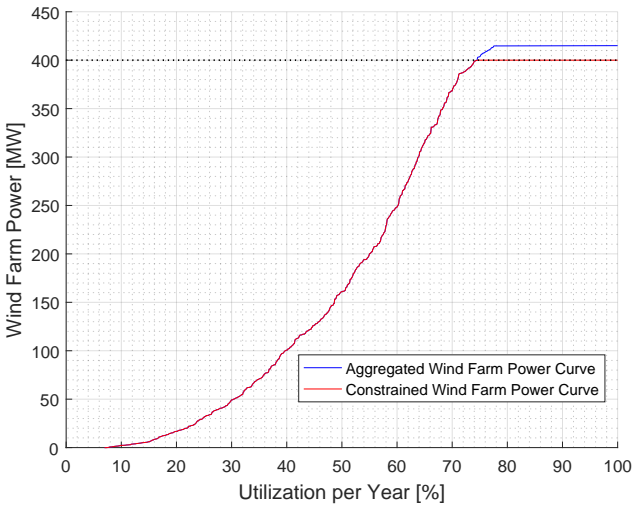
Factors

- Ratio of wind turbine expenditure to electrical infrastructure
- Wind speed distribution
- Wind turbine availability
- Inter-array cable availability
- Wake effects
- Electrical losses
- Degradation factor

Modelling of Overplanting

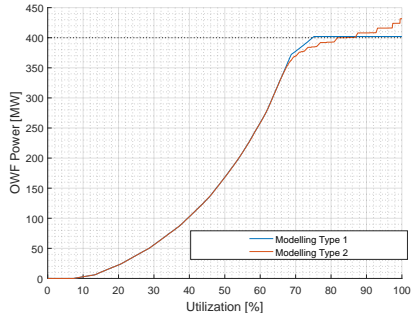
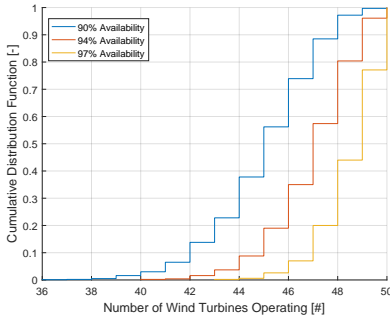
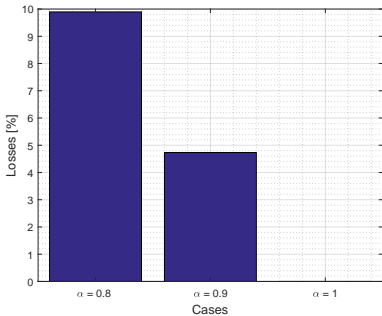
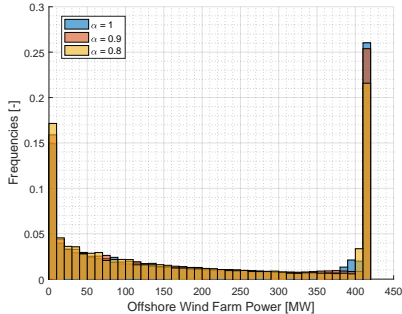
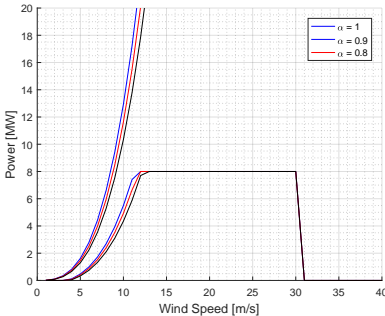
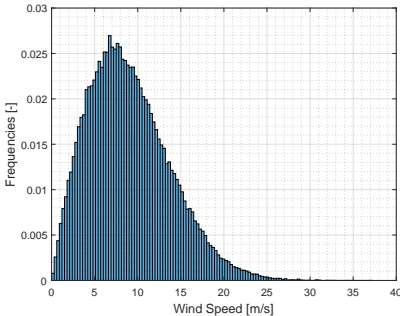
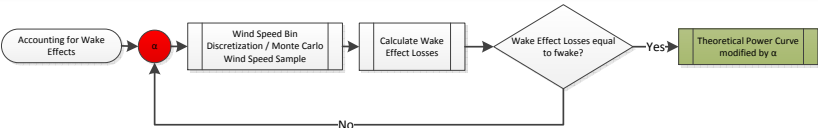


Modelling Type 1

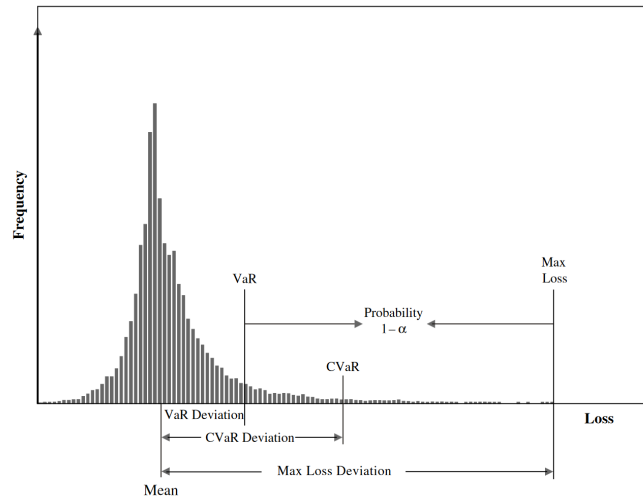


Modelling Type 2

Modelling of Overplanting



Modelling Risk Aversion



[Rockafellar and Uryasev 2002]

Risk metric

$$\rho_{\alpha}[\lambda, LCOE] = \lambda \mathbf{CVaR}_{\alpha}[LCOE] + (1 - \lambda) \mathbf{Median}[LCOE]$$

Case Study

Case Study

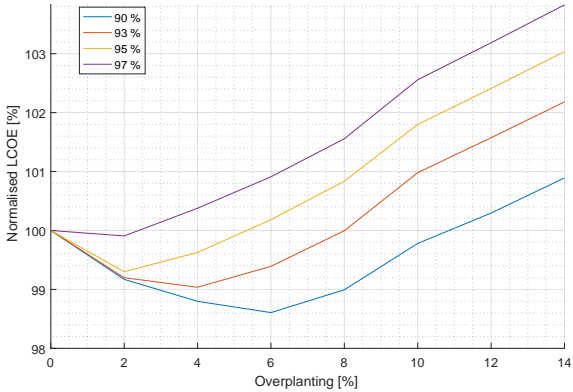
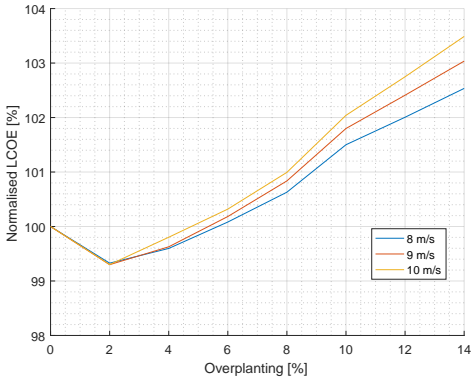
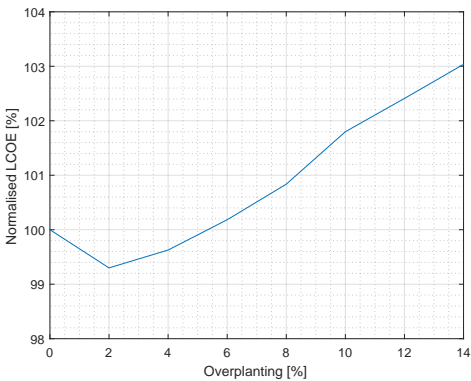
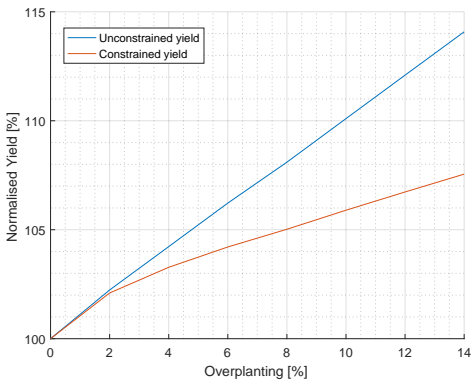
400MW commercial offshore wind farm
 400MW fixed maximum export capacity
 50-8MW WTGs
 0-14% overplanting
 2% overplanting = 1 additional WTG

Characteristic	Value	Uncertainty
Water Depth [m]	25	None
Distance from shore [km]	25	None
Mean Wind Speed @ 100m [m/s]	9	$\mathcal{N}(9, 0.1^2)$
Wind Turbine Availability [%]	95	$\mathcal{U}(90, 97)$
Inter-Array Cable Availability [%]	99	$\mathcal{U}(97, 99)$
Foundation Type [-]	Monopile	None
Electrical Infrastructure [-]	HVAC	None
Wind Turbine Type [-]	164-8 MW	None
Wake effect [%]	10	None
Degradation Factor [%]	0.5	None

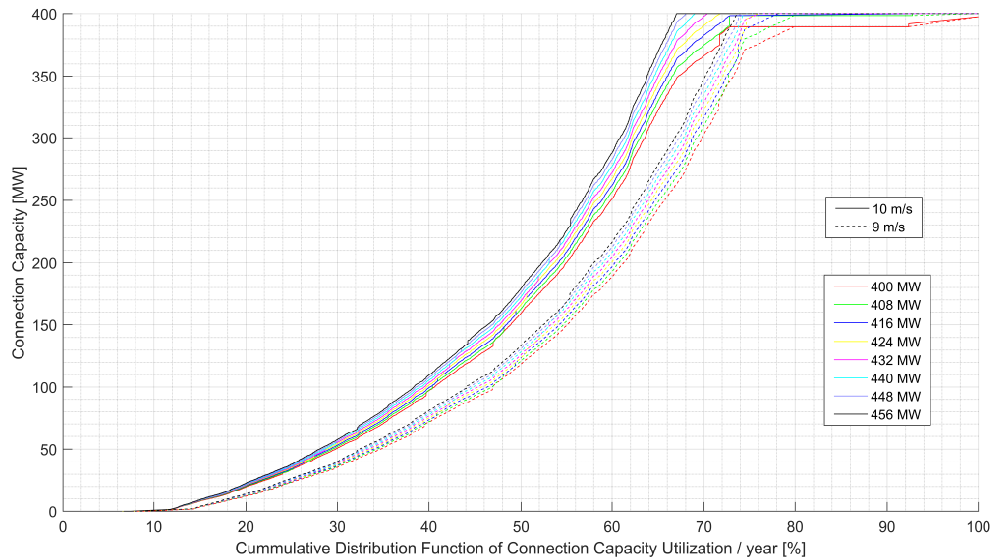


Deterministic Results

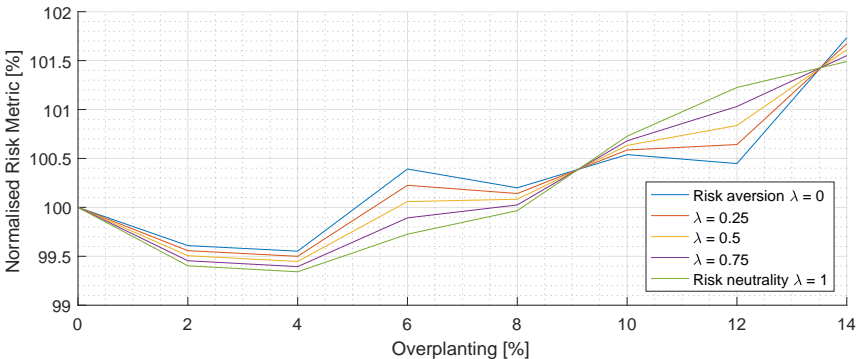
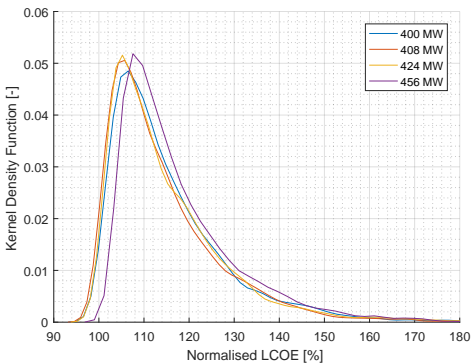
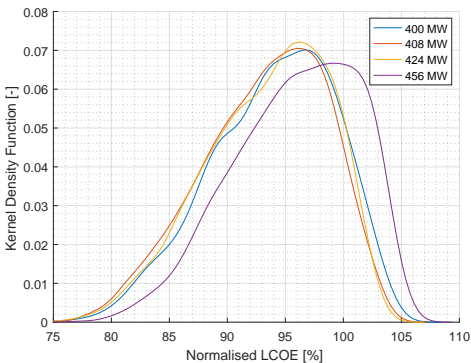
Deterministic Results & Local Sensitivity Analysis



Local Sensitivity Analysis - wind speed



Stochastic Results $\rho_\alpha[\lambda, \text{overplanting}]$



Conclusions and Future Work

Conclusions

- Development of a novel framework to evaluate overplanting
- Modelling Type 1 is easier to implement but may lead to an overestimation of the annual energy production
- Modelling Type 2 is more accurate but requires higher computational costs
- Wind turbine availability is the most sensitive parameter to overplanting
- Previous studies based on low wind turbine availabilities rates or on Modelling Type 1, need to be revisited
- Optimal overplanting setup increased when considering the uncertainty quantification framework regardless of risk appetite (from 2% to 4%)
- Overplanting the reference farm from 2% to 8% gives a better result than with no overplanting for a risk neutral setting

Future Work

- How is overplanting influence by larger turbines and sites located further from shore?
- How does risk aversion influence the decision for these new sites?

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Questions

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