

Optimising tidal turbine farms using the adjoint approach

S. W. Funke
in collaboration with
S. C. Kramer, P. E. Farrell, M. D. Piggott

Imperial College London

September 18, 2013

Tidal stream generators

- ▶ Come in different shapes:



Tidal stream generators

- ▶ Come in different shapes:



- ▶ Generate between 0.5 – 1.5 MW.

Tidal stream generators

- ▶ Come in different shapes:



- ▶ Generate between 0.5 – 1.5 MW.
- ▶ Will typically be deployed in farms of 10 – 500 turbines.

Tidal stream generators

- ▶ Come in different shapes:



- ▶ Generate between 0.5 – 1.5 MW.
- ▶ Will typically be deployed in farms of 10 – 500 turbines.
- ▶ Very new technology (compared to wind turbines).

Current tidal projects

| | | |
|----------------------------------|-------------------|----------------|
| Inner Sound, Pentland Firth, UK | 400×1 MW | starting 2014 |
| Cantick Head, Pentland Firth, UK | 200×1 MW | starting 2016 |
| Kaipara Harbour, New Zealand | 200 MW | planning stage |
| Gulf of Kutch, India | 50×1 MW | approved 2012 |
| ... | | |

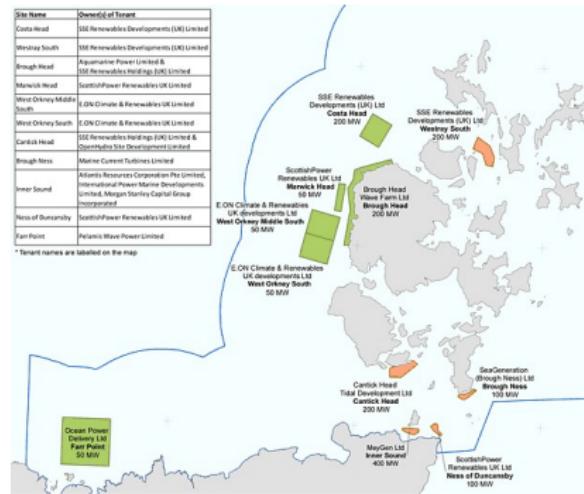


Figure: Leased tidal projects sites (orange) in Scotland, UK

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Tidal farm optimisation

Key question

How should the farm be designed to **maximise the profit?**

Design factors:

- ▶ Position and number of turbines.
- ▶ Turbine “tuning”, e.g. direction, pitch settings.
- ▶ Installation and maintenance costs
(e.g. turbines, cable, electricity transformers).
- ▶ Electricity price.

Literature overview

Tidal farm optimisation

- ▶ Analytical approach with 1-D models (*Bryden and Couch (2007), Garrett and Cummins (2008), Vennell (2010, 2011)*).
- ▶ Manual comparison of selected farm designs with (more) realistic PDE-based models (*Lee et al. (2010), Divett et al. (2013)*).

Wind farm optimisation

- ▶ Related, but has the additional uncertainty of the wind direction.
- ▶ Typically apply evolutionary algorithms on simplistic wake models (*Bilbao and Alba (2009), Wan et al. (2010)*).

Goal: Develop a software tool for optimising tidal farms based on a realistic **PDE model** which **scales** up to hundreds of turbines.

Problem formulation of the farm optimisation problem

$$\max_{u,\eta,m} \frac{1}{T} \int_0^T \int_{\Omega} \rho c_t(m) \|u\|^3 \equiv \text{Power}(u, m)$$

subject to the SWEs:

$$\frac{\partial u}{\partial t} + u \cdot \nabla u - \nu \nabla^2 u + g \nabla \eta + \frac{c_b + c_t(m)}{H} \|u\| u = 0,$$

$$\frac{\partial \eta}{\partial t} + \nabla \cdot (H u) = 0,$$

$$g(m) \leq 0,$$

where:

$$(0, T) \times \Omega$$

Computational domain,

$$m : \mathbb{R}^{2N}$$

x, y positions for N turbines,

$$u : (0, T) \times \Omega \rightarrow \mathbb{R}^2$$

Velocity,

$$\eta : (0, T) \times \Omega \rightarrow \mathbb{R}$$

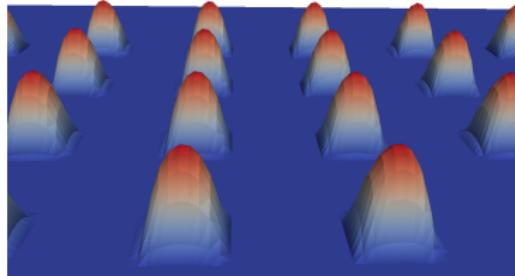
Free-surface displacement,

$$g : \mathbb{R}^{2N} \rightarrow \mathbb{R}^M$$

Constraints on m .

Turbine parametrisation

Turbines are modelled via an increased bottom friction over a small area representative of an individual turbine:

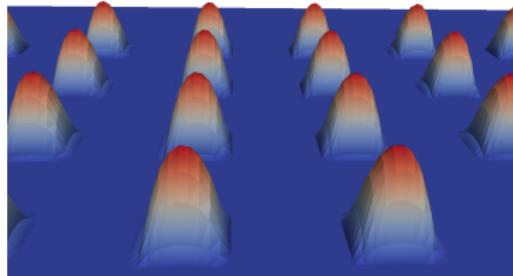


Turbine parametrisation

Turbines are modelled via an increased bottom friction over a small area representative of an individual turbine:

$$c_t(m) = \sum_{(p_x, p_y) \in m} K \psi_{p_x, r}(x) \psi_{p_y, r}(y),$$

where K is a “tuning” parameter of the turbine



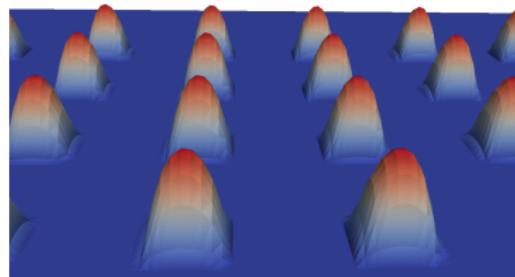
Turbine parametrisation

Turbines are modelled via an increased bottom friction over a small area representative of an individual turbine:

$$c_t(m) = \sum_{(p_x, p_y) \in m} K \psi_{p_x, r}(x) \psi_{p_y, r}(y),$$

where K is a “tuning” parameter of the turbine and

$$\psi_{p, r}(x) \equiv \begin{cases} e^{1-1/(1-\|\frac{x-p}{r}\|^2)} & \text{for } \|\frac{x-p}{r}\| < 1, \\ 0 & \text{otherwise.} \end{cases}$$



Optimisation algorithms

Optimisation methods

| Method | Gradient | Hessian | Typical # iterations |
|--------------|----------|---------|---|
| Evolutionary | No | No | $> \mathcal{O}(10^4) - \mathcal{O}(10^6)$ |
| Newton | Yes | Yes | $\mathcal{O}(100)$ |
| Quasi-Newton | Yes | No | $\mathcal{O}(100)$ |

Optimisation algorithms

Optimisation methods

| Method | Gradient | Hessian | Typical # iterations |
|--------------|----------|---------|---|
| Evolutionary | No | No | $> \mathcal{O}(10^4) - \mathcal{O}(10^6)$ |
| Newton | Yes | Yes | $\mathcal{O}(100)$ |
| Quasi-Newton | Yes | No | $\mathcal{O}(100)$ |

Gradient/Hessian computation

| Method | # PDE solves |
|-----------------------------|------------------|
| Finite differences | $\mathcal{O}(N)$ |
| Tangent linear/complex step | $\mathcal{O}(N)$ |
| Adjoint | $\mathcal{O}(1)$ |

Computing gradients with the adjoint approach

- ▶ Requires solution of one adjoint PDE (the adjoint shallow water equations), **independently of the number of turbines.**

Computing gradients with the adjoint approach

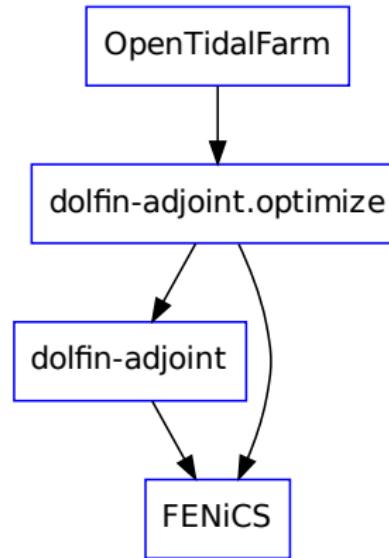
- ▶ Requires solution of one adjoint PDE (the adjoint shallow water equations), **independently of the number of turbines**.
- ▶ The adjoint PDE (i) depends on the original PDE solution, (ii) is **linear** and (iii) is solved **backwards in time**.

Computing gradients with the adjoint approach

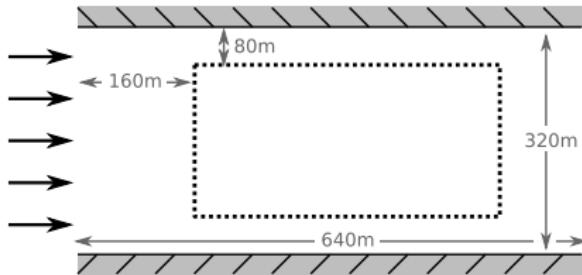
- ▶ Requires solution of one adjoint PDE (the adjoint shallow water equations), **independently of the number of turbines**.
- ▶ The adjoint PDE (i) depends on the original PDE solution, (ii) is **linear** and (iii) is solved **backwards in time**.
- ▶ Computing the gradient is **much faster** than evaluating the farm power:

| # turbines | Runtime: Power(u, m) | Runtime: dPower/dm |
|------------|--------------------------|--------------------|
| 8 | 58 s | 7.3 s |
| 64 | 56 s | 7.3 s |
| 256 | 70 s | 7.8 s |

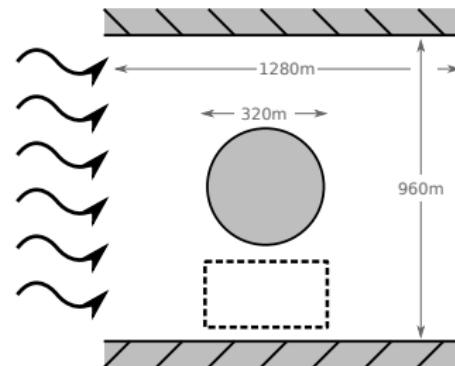
Implementation of *OpenTidalFarm*



Idealised examples



(a) Simple channel



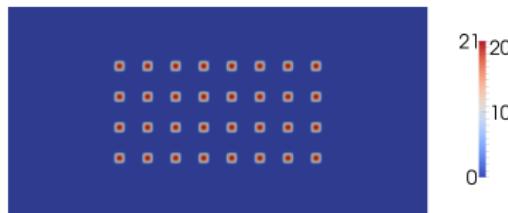
(b) Channel with island and tidal flow

Figure: Geometry of idealised examples.

Goal

Optimally deploy 32 turbines in the area marked by dashed lines.

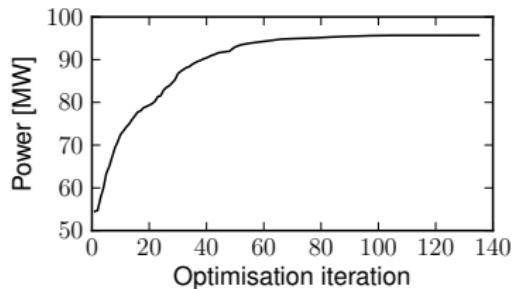
Simple channel



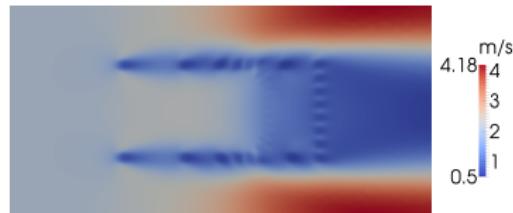
(a) Initial turbine positions



(b) Optimised turbine positions



(c) Optimisation convergence

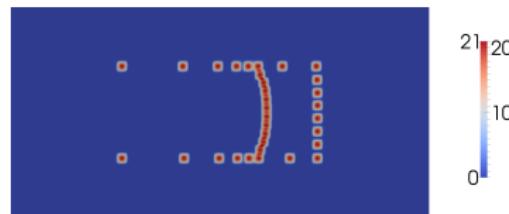


(d) Velocity magnitude

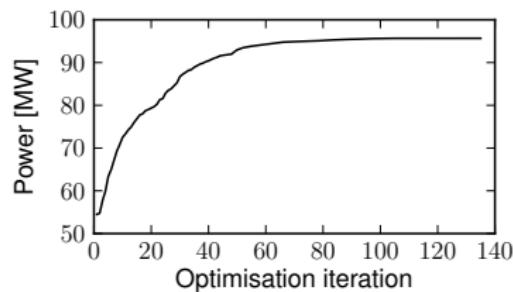
Simple channel



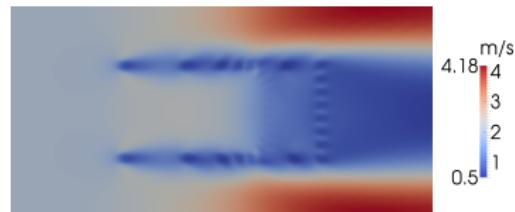
(a) Initial turbine positions



(b) Optimised turbine positions



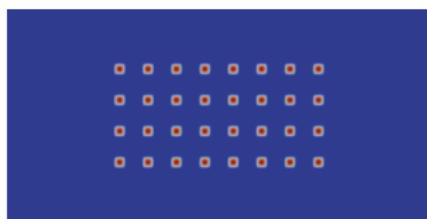
(c) Optimisation convergence



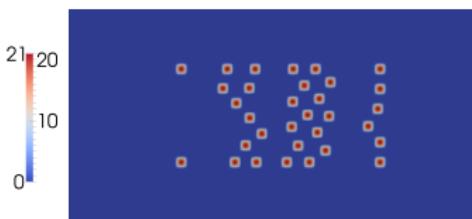
(d) Velocity magnitude

Figure: Results of the simple channel optimisation.

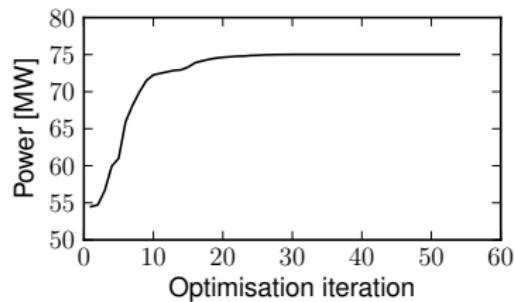
Simple channel with minimum distance constraint



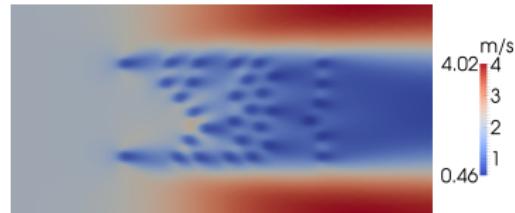
(a) Initial turbine positions



(b) Optimised turbine positions



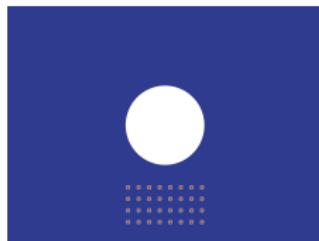
(c) Optimisation convergence



(d) Velocity magnitude

Figure: Results of the simple channel optimisation.

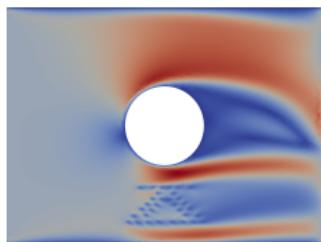
Channel with island and tidal flow



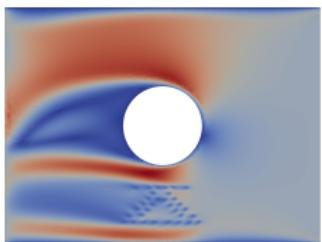
(a) Initial turbine positions

21
20
10
0

(b) Optimised turbine positions

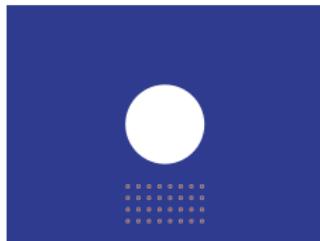
(c) Velocity magnitude
reaching its maximum from
the left

4.9
4
3
2
1
0

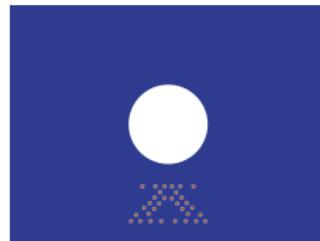
(d) Velocity magnitude
reaching its maximum from
the right

4.9
4
3
2
1
0

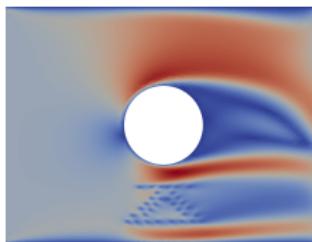
Channel with island and tidal flow



(e) Initial turbine positions

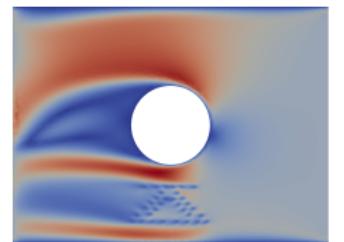


(f) Optimised turbine positions



(g) Velocity magnitude

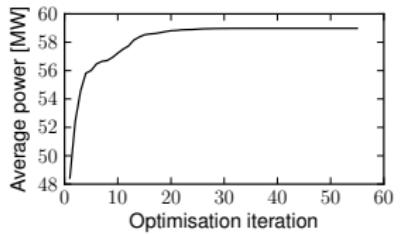
reaching its maximum from the left



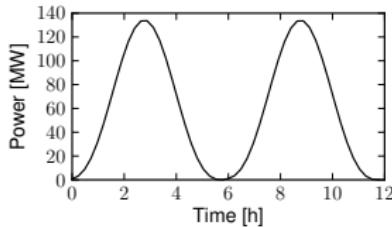
(h) Velocity magnitude

reaching its maximum from the right

Channel with island and tidal flow



(i) Optimisation convergence



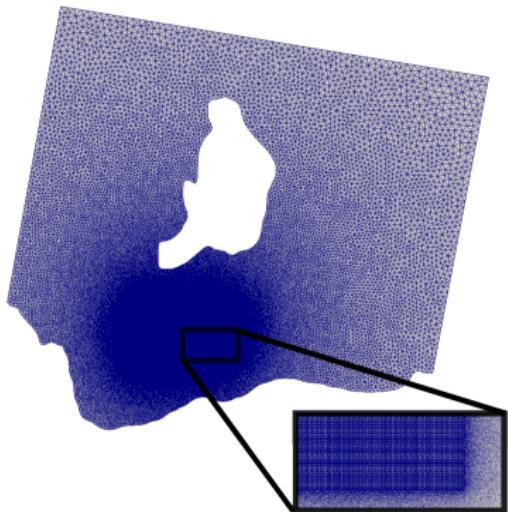
(j) Power extraction over time of the optimised configuration

Figure: Results of the channel with island and tidal flow optimisation.

Optimisation of 256 turbines in the Pentland Firth



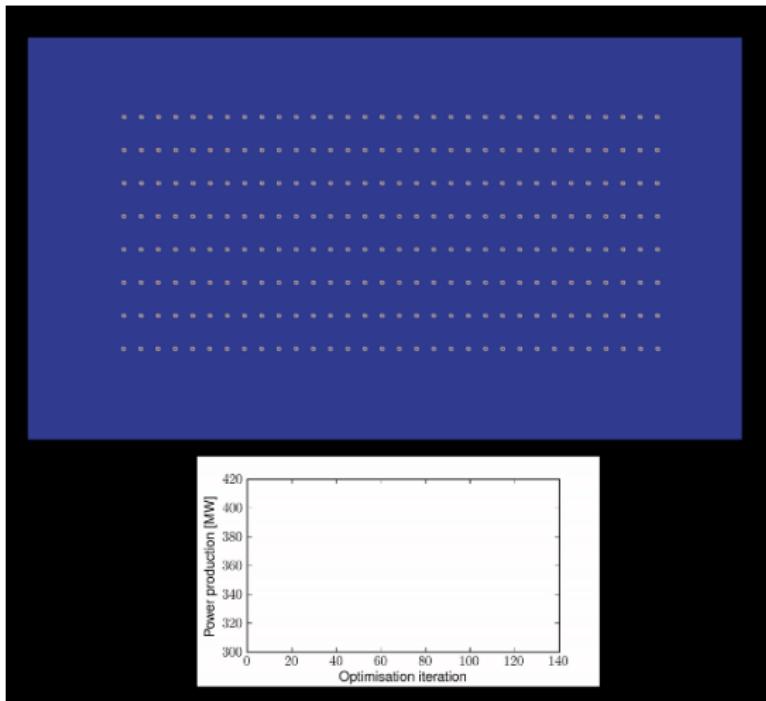
(a) Satellite image of Stroma Island and Caithness (Bing Maps, Microsoft). Turbine site marked in pink.



(b) Mesh with 1.5 – 200 m element size. 5.6×10^6 DOFs.

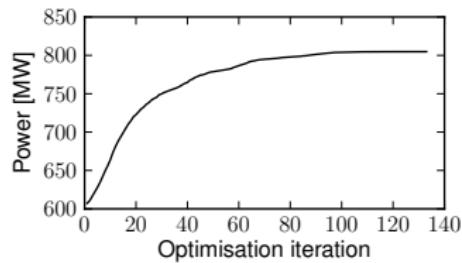
Figure: Geometry and mesh.

Optimisation of 256 turbines in the Pentland Firth

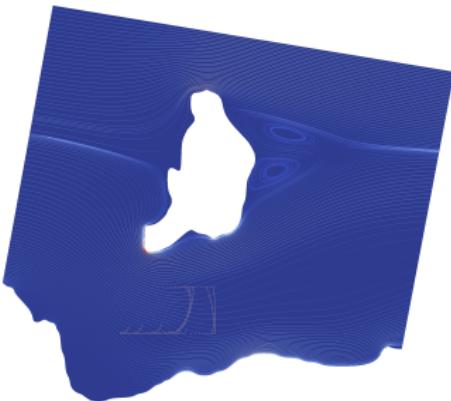


Optimisation of 256 turbines in the Pentland Firth

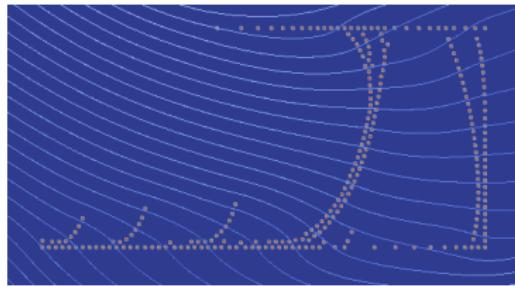
Optimisation of 256 turbines in the Pentland Firth



(a) Optimisation convergence



(b) Streamline flow



(c) Streamline flow (zoom)



(d) Turbine power map

Complex constraints

Turbine site developers need to consider complex constraints for the turbine deployment, such as:

Complex constraints

Turbine site developers need to consider complex constraints for the turbine deployment, such as:

- ▶ maximum/minimum installation depth,

Complex constraints

Turbine site developers need to consider complex constraints for the turbine deployment, such as:

- ▶ maximum/minimum installation depth,
- ▶ maximum bathymetry gradient,

Complex constraints

Turbine site developers need to consider complex constraints for the turbine deployment, such as:

- ▶ maximum/minimum installation depth,
- ▶ maximum bathymetry gradient,
- ▶ complicated (non-convex) domain sites and/or

Complex constraints

Turbine site developers need to consider complex constraints for the turbine deployment, such as:

- ▶ maximum/minimum installation depth,
- ▶ maximum bathymetry gradient,
- ▶ complicated (non-convex) domain sites and/or
- ▶ optimisation of multiple farms.

Problem formulation for complex constraints

$$\begin{aligned} & \max_{u,m} \text{Power}(u, m) \\ & \text{subject to} \\ & \frac{\partial u}{\partial t} + u \cdot \nabla u - \nu \nabla^2 u + g \nabla \eta + \frac{c_b + c_t(m)}{H} \|u\| u = 0, \\ & \frac{\partial \eta}{\partial t} + \nabla \cdot (Hu) = 0, \\ & g(m) \leq 0, \end{aligned}$$

Problem: The inequality constraint function g becomes highly non-convex.

Problem formulation for complex constraints

$$\begin{aligned}
 & \max_{u, \tilde{m}} \text{Power}(u, \tilde{m}) \\
 & \text{subject to} \\
 & \frac{\partial u}{\partial t} + u \cdot \nabla u - \nu \nabla^2 u + g \nabla \eta + \frac{c_b + \tilde{m}}{H} \|u\| u = 0, \\
 & \frac{\partial \eta}{\partial t} + \nabla \cdot (H u) = 0, \\
 & 0 \leq \tilde{m},
 \end{aligned}$$

Problem: The inequality constraint function g becomes highly non-convex.

Solution: Optimise the “concentration” of turbines $\tilde{m} : \Omega \rightarrow \mathbb{R}$ instead of parametrising turbines individually.

Multifarm optimisation in the Pentland Firth

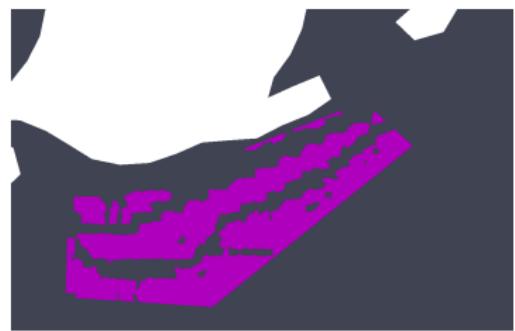


Figure: Computational domain in the wider Pentland Firth with four tidal turbine sites. The valid deployment locations (based in the bathymetry gradient) of the sites are marked in pink.

Multifarm optimisation in the Pentland Firth

Multifarm optimisation in the Pentland Firth



Conclusion

- ▶ Optimising turbine farms can **significantly increase its performance** (20 - 100%).

Conclusion

- ▶ Optimising turbine farms can **significantly increase its performance** (20 - 100%).
- ▶ Optimal configurations are **difficult to find by intuition** only.
However, recurring patterns are:
 - ▶ Arrange turbines to “barrages” perpendicular to the flow streamlines.
 - ▶ Build “walls” of turbines to “trap” flow.
 - ▶ Less (or smaller) turbines can be better!

Conclusion

- ▶ Optimising turbine farms can **significantly increase its performance** (20 - 100%).
- ▶ Optimal configurations are **difficult to find by intuition** only.
However, recurring patterns are:
 - ▶ Arrange turbines to “barrages” perpendicular to the flow streamlines.
 - ▶ Build “walls” of turbines to “trap” flow.
 - ▶ Less (or smaller) turbines can be better!
- ▶ Combined with the adjoint approach, the optimum is obtained in typically 100 – 200 **flow solves**.

Conclusion

- ▶ Optimising turbine farms can **significantly increase its performance** (20 - 100%).
- ▶ Optimal configurations are **difficult to find by intuition** only.
However, recurring patterns are:
 - ▶ Arrange turbines to “barrages” perpendicular to the flow streamlines.
 - ▶ Build “walls” of turbines to “trap” flow.
 - ▶ Less (or smaller) turbines can be better!
- ▶ Combined with the adjoint approach, the optimum is obtained in typically 100 – 200 **flow solves**.
- ▶ The farm optimisation problem is non-convex (different initial guesses yield different optimal solutions), but **relatively well behaved** (good optimal solutions are found from any initial guess).

Future work

- ▶ Combine “smooth” and individual turbine optimisation.
- ▶ Validate wakes produced by the turbines.
- ▶ Incorporate financial (e.g. cable routing costs) and ecological model in the optimisation.

OpenTidalFarm

<http://opentidalfarm.org>