

WIND FARM LAYOUT OPTIMIZATION ON A DISCRETIZED 3D DOMAIN

Federica Acerbi¹, Stefano Grassi², Davide M. Raimondo¹

¹Dept. of Electrical, Computer and Biomedical Engineering, University of Pavia, Italy

²Dept. of Civil, Environmental and Geomatic Engineering, ETH, Zurich, Switzerland



ETH
Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

PO.266

ABSTRACT

The placement of wind turbines (WTs) in a defined region is a crucial aspect since it affects the long-term energy production of the wind farm, and thus its efficiency and economic value. In order to design an optimal layout, it is necessary to take into account topographical features, wind resource, wind turbines interactions, upfront costs, O&M and financial framework. The problem can be posed as a constrained optimization, which involves a cost function, describing the objectives, and a mathematical model describing the wake effect, the interaction between WTs, and the energy generation.

APPROACH

We propose an optimization-based algorithm that, thanks to Geographic Information System (GIS) software and data, enables the design of realistic wind farm layouts over complex 3D terrains.

Elements to take into account:

- Morphology of terrain¹
- Power Curve²
- Wake effect between wind turbines³
- Wind resource⁴
- Economic performance⁵

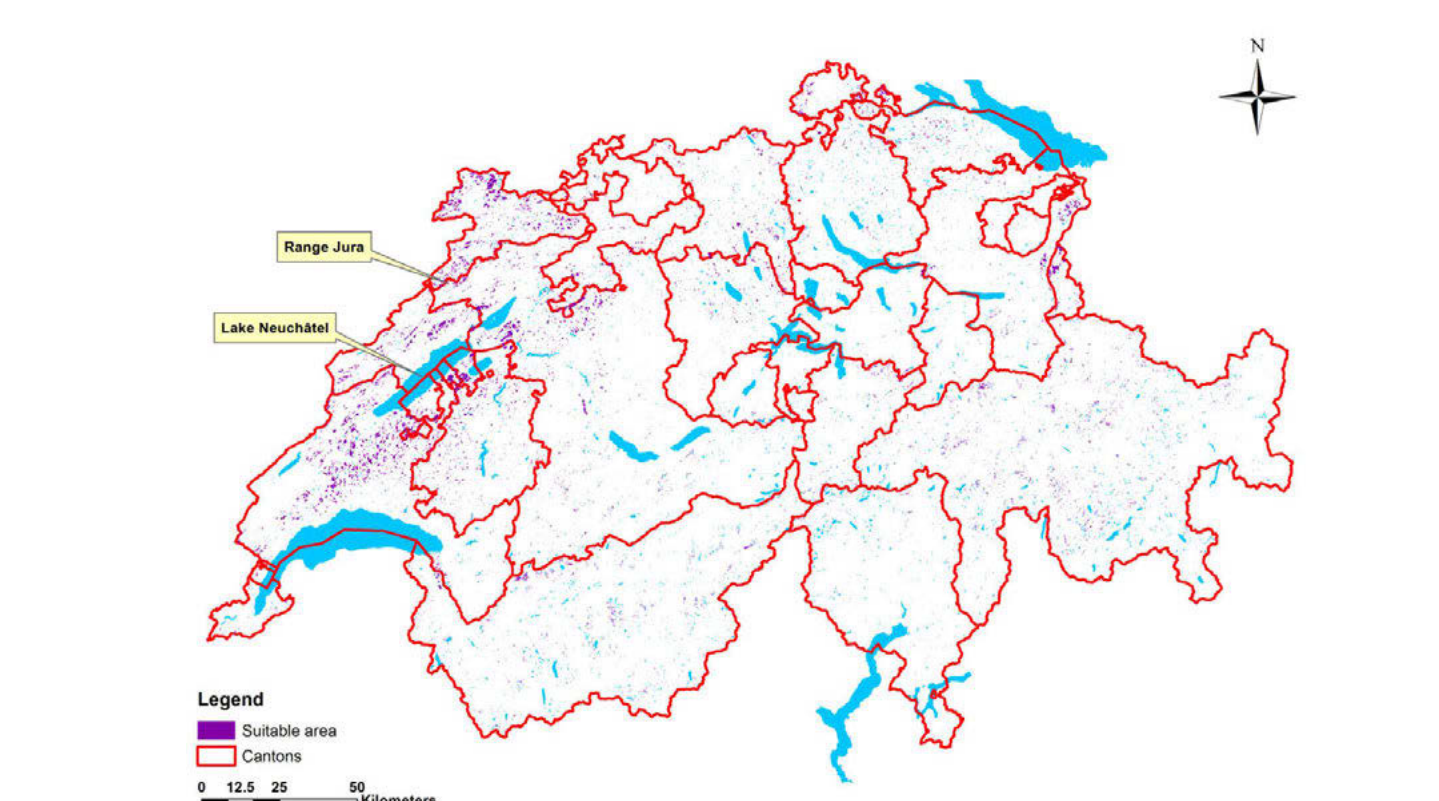
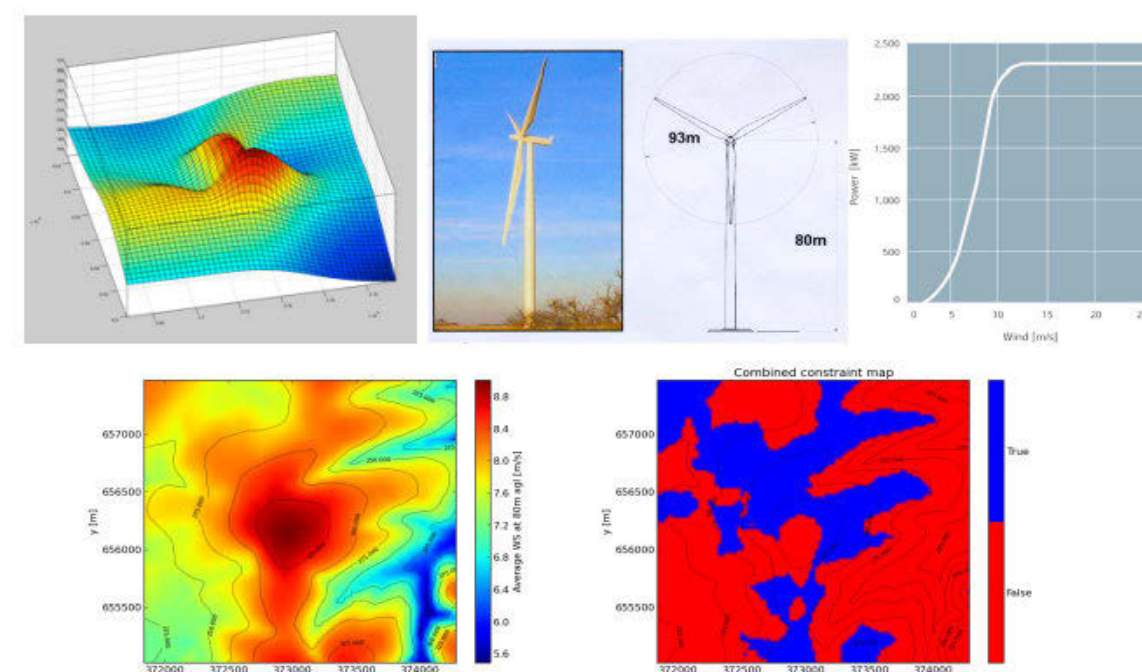


Figure 2: Wind turbine power curve

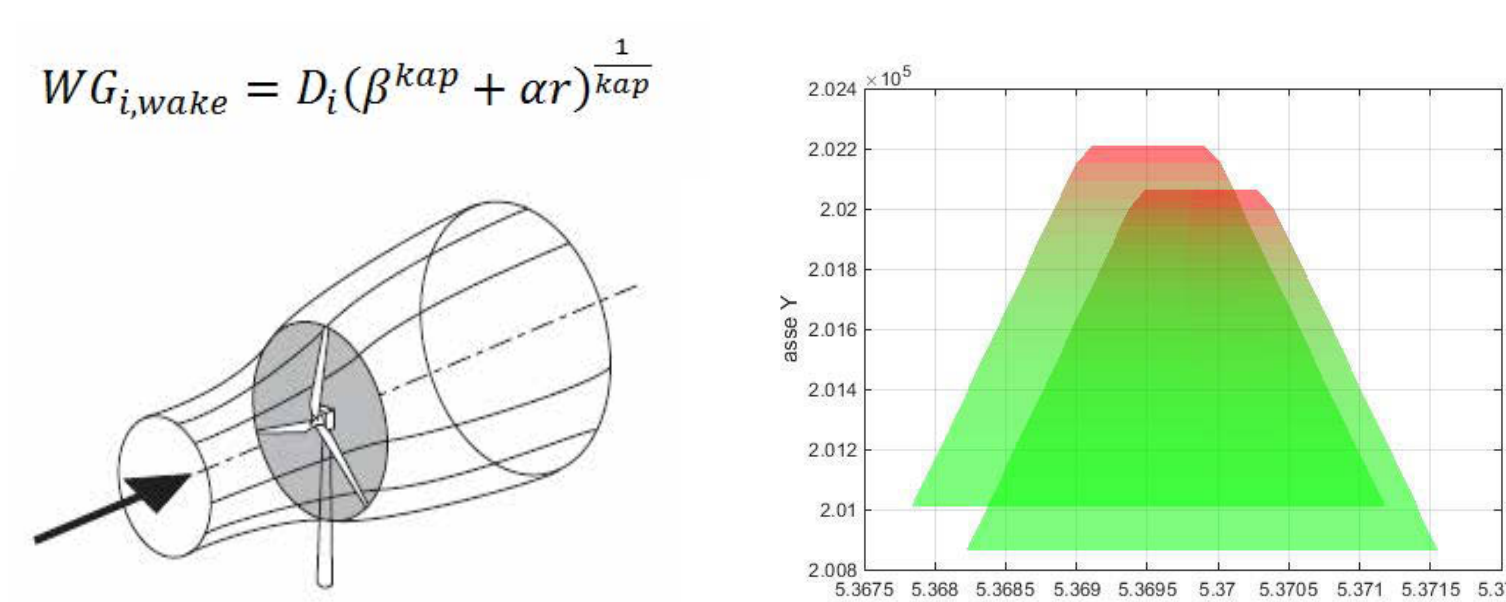
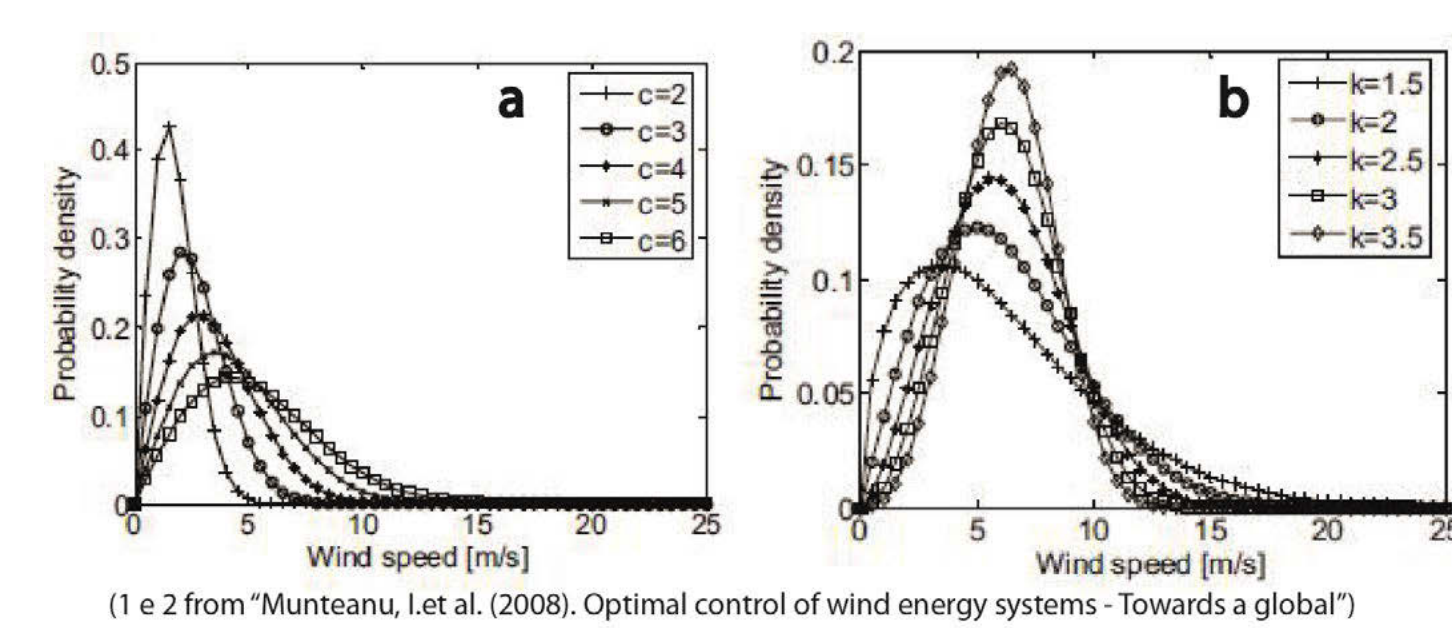


Figure 3: Example of "wake effect" in a wind farm offshore. Overlap of a wake effect with a WT rotor. This is a very important factor in the WT placement.

$$AEP(WT_i) = 8760 \beta_{avail} \beta_{loss} \sum_{j=1}^N \sum_{s=1}^{n_{dir}} \mu(\bar{u})^{(s)} P(\bar{u})^s$$

Figure 5: Annual Energy Production



$$f(x, C, k) = \begin{cases} \frac{k}{C} \left(\frac{x}{C}\right)^{k-1} e^{-\left(\frac{x}{C}\right)^k}, & \text{if } x \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

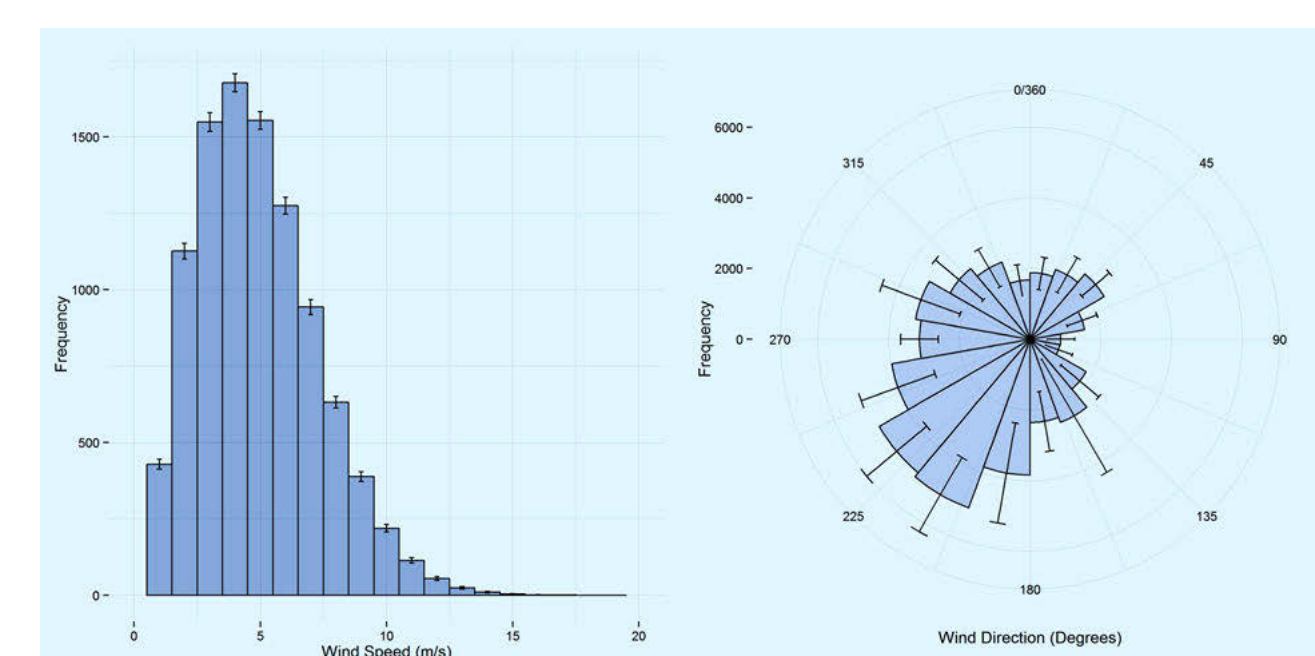


Figure 4: Examples of Weibull distributions (top). Wind speed and wind direction distributions discretized in bins (bottom). Note that each location comes with a different distribution.

ALGORITHM

The relationship that describes the aerodynamic interference between the speed in input to the i -th turbine and the overlap produced by the wake of the turbines in upstream is given by:

$$V_i^{[input]} = V_i^{[inlet]} - \sqrt{\sum_{j=1}^n \left(\frac{A_j^{overlap}}{A_i^{WT}} \right) (V_i^{[inlet]} - V_j(x))^2}$$

Nonlinear equality constraints like the one above make the problem difficult to solve. We therefore make an approximation:

$$V_i^{[input]} = V_i^{[inlet]} - \sum_{j=1}^n \sqrt{\left(\frac{A_j^{overlap}}{A_i^{WT}} \right) (V_i^{[inlet]} - V_j(x))}$$

Since $\|(x+y)\|_2 \leq \|(x+y)\|_1$ the use of the 1-norm instead of the 2-norm does not have any repercussion on the optimal placement. Binary variables are then introduced to indicate the presence/absence of a wind turbine at a given location

$$\begin{cases} V_1^{[input]} = V_1^{[inlet]} \delta_1 - \sum_{j=1}^n (C_{1j} V_1^{[inlet]} \gamma_{1j}) + \sum_{j=1}^n (K_{1j} z_{1j}) \\ \vdots \\ V_N^{[input]} = V_N^{[inlet]} \delta_N - \sum_{j=1}^n (C_{Nj} V_N^{[inlet]} \gamma_{Nj}) + \sum_{j=1}^n (K_{Nj} z_{Nj}) \end{cases}$$

The reformulation comes with a set of linear constraints which involve both continuous and binary variables

$$\begin{aligned} V_j^* &= \max(V_j^{[input]}, 0) & z_{ij} &= V_j^* \delta_i & \gamma_{ij} &= \delta_i \delta_j \\ V_j^* &= V_j^{[input]} \delta_i^* & & & & \\ \begin{cases} V_j^* \leq M \delta_i^* \\ V_j^* \geq m \delta_i^* \\ V_j^* \leq V_j^{[input]} - m(1 - \delta_i^*) \\ V_j^* \geq V_j^{[input]} - M(1 - \delta_i^*) \end{cases} & \begin{cases} z_{ij} \leq M \delta_i \\ z_{ij} \geq m \delta_i \\ z_{ij} \leq V_j^* - m(1 - \delta_i) \\ z_{ij} \geq V_j^* - M(1 - \delta_i) \end{cases} & \begin{cases} -\delta_i + \gamma_{ij} \leq 0 \\ -\delta_j + \gamma_{ij} \leq 0 \\ \delta_i + \delta_j - \gamma_{ij} \leq 1 \end{cases} & \begin{cases} V_j^{[input]} \geq m(1 - \delta_j^*) \\ V_j^{[input]} \leq -\varepsilon + (M + \varepsilon) \delta_j^* \end{cases} \end{aligned}$$

This reformulation allows to pose the problem of maximizing the energy production as an MILP with cost function $\sum_{i=1}^M V_i^{[input]}$. The evaluation of the IRR is done a posteriori. By iterating the approach with an increasing number of WT, it is possible to obtain the optimal location and WTs to be installed.

CONCLUSIONS

The following figures show the optimal placement in terms of IRR using the approach described above on a real wind farm.

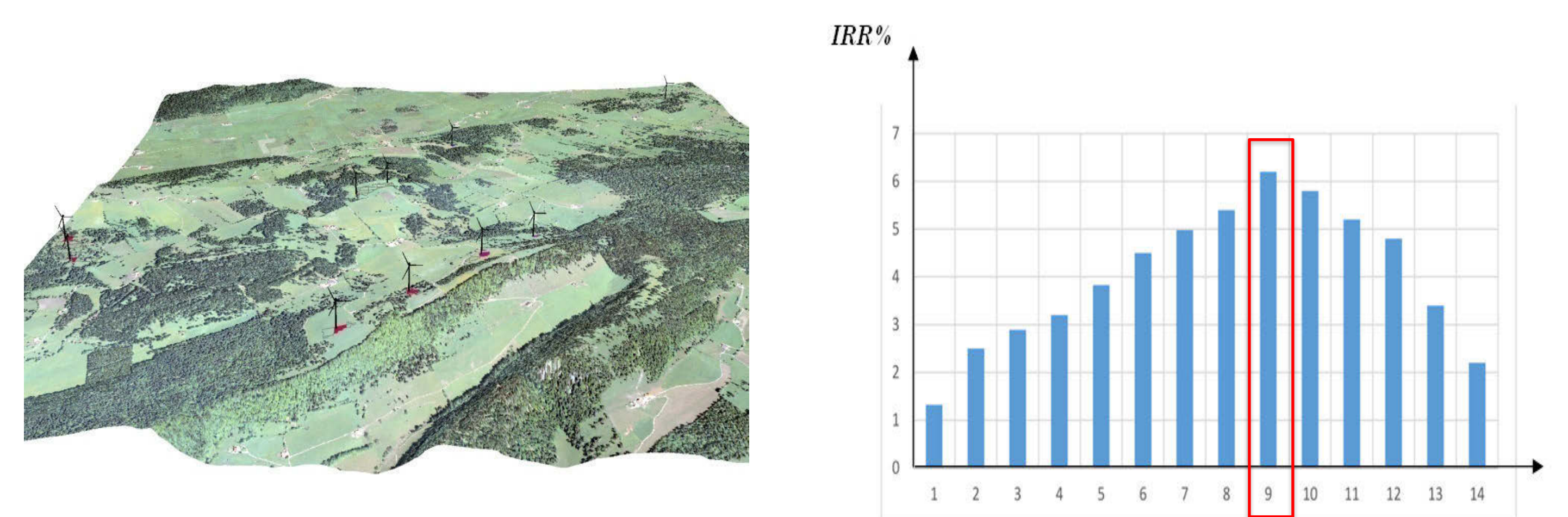


Figure 1: Visualization of the optimized WTs positions with orthophotos (oversized WTs for better visualization). Economic performance: IRR trend as a function of number of wind turbines installed.

REFERENCES

1. G. Marmidis, S.L.a.E.P., *Optimal placement of wind turbines in a wind park using Monte Carlo simulation*. Renewable Energy, 2008. 33: p. 1455-1460.
2. Şişbot, S., et al., *Optimal positioning of wind turbines on Gökçeada using multi-objective genetic algorithm*. Wind Energy, 2010. 13: p. 297-306.
3. Tzanos, et al., *Optimal wind turbine placement via randomized optimization techniques*, in *Power Systems Computation Conference (PSCC)*. 2011.
4. Veronesi, et al., *Satellite data to improve the accuracy of statistical models for wind resource assessment*. in *EWEA*. 2015.
5. Frandsen, S., et al., *Analytical modelling of wind speed deficit in large offshore wind farms*, in *Wind Energy2006*. p. 39-53.
6. Marsaglia, G., et al., *Optimal placement of wind turbines on a continuous domain: An MILP-based approach*. in *American Control Conference (ACC)*, 2015.
7. Bemporad, A. and M. Morari, *Control of systems integrating logic, dynamics, and constraints*. Automatica, 1999. 35(3): p. 407-427.

