ALFRED: The Artificial Learning Flexible Renewable Energy System **Dispatch Optimizer**

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Motivation

- The high penetration of renewable energies in the grid brings imbalance on power market prices and modify typical demand design curves⁽¹⁾.
- Renewable energy systems with storage are able to provide flexible dispatch, bringing the significant advantage of economically meeting peak demand.
- Therefore, renewable systems with storage are expected to participate in whole-sale electricity markets, enabling competition with all the players in the market and ensuring the pathway to a highly renewable energy mix

Approach & Methods

Dispatch Optimizer that derives a electricity delivery schedule for the day-ahead. It considers weather and well as specific market rules.

- forecasts
- forecasts
- Possibility of combination with several types of



- Why to optimize the dispatch?
 - To produce clean energy during electricity demand peaks
 - To reach higher selling prices of electricity produced from renewable energy systems
 - To adapt plant operation practices according to optimal production schedule
 - To allow renewable energy systems with storage to participate in the wholesale energy market
- Why to consider forecast uncertainties?
- To improve the quality of electricity scheduling by actively considering associated uncertainties in weather and pricing forecasts
- To reduce financial drawbacks related to modification of scheduled energy delivery



Results

- **Annual Simulations:**
 - Considering a 120 MW concentrated solar thermal plant with 10h storage following ALFRED's strategy
 - Three different weather forecast sets: perfect, persistence and ECMWF probabilistic sets
 - To evaluate ALFRED's benefit, simulation of the same plant following a solar driven strategy was carried out



Figure 3. Annual financial income improvement in comparison to benchmark scheduling strategy for 2016 (a) and 2017 (b).

Improvement in Financial Income: ۲

- Due to possibility of achieving more accurate delivery
- Related to the quality of the weather forecast and enhanced with the uncertainty post-processing ALFRED scheduling over performs a simple scheduling strategy

Conclusions

- Uncertainties included in dispatch optimization
- More accurate delivery scheduling •
- Quality of schedules is not exclusively ٠ bounded to quality of weather forecasts
- Possibility of combination with several • types of optimization methods and system models
- Possibility of dealing with several types of forecasts (deterministic or probabilistic)
- Flexible application for different • renewable energy systems with storage
- Flexible application for different market scenarios Improvement in financial income of

- ALFRED is able to improve revenues independently on the quality of the weather forecast
- **Illustrating Example:** financial income comparison ____
 - for different weather forecast sets
 - compared with benchmark scheduling strategy (solar driven + persistence forecast)

References

renewable energy systems with storage, which leads to possible participation in wholesale energy market

Knowledge for Tomorrow

(1) IRENA, "Adapting market design to high shares of variable renewable energy", Abu Dhabi (2017). (2) D.H. Wolpert and W.G. Macready, "No free lunch theorems for optimization", in IEEE Transactions on Evolutionary Computation, Vol.1, No. 1 (1997). Acknowledgements:

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