Deep learning-based multi-output quantile forecasting of PV generation

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Context & contributions

Renewable energy is subject to uncertainty.

Probabilistic forecasts are an important tool to equip decision-makers [1].

PV quantile forecast using quantile regression [2].

Contributions:

- 1. A tailored deep learning-based multi-output quantile PV forecaster;
- 2. An encoder-decoder architecture;
- 3. Weather forecasts of the MAR regional climate model [3];
- 4. A proper assessment of the quantile forecasts is conducted by using a **k-cross validation** methodology and **probabilistic metrics.**

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^[1] Morales, Juan M., et al. Integrating renewables in electricity markets: operational problems. Vol. 205. Springer Science & Business Media, 2013.

^[2] Koenker, Roger, and Gilbert Bassett Jr. "Regression quantiles." Econometrica: journal of the Econometric Society (1978): 33-50.

^[3] Fettweis, Xavier, et al. "Reconstructions of the 1900–2015 Greenland ice sheet surface mass balance using the regional climate MAR model." The Cryosphere 11.2 (2017): 1015-1033.

Forecasting techniques

Day-ahead models:

- Gradient Boosting (GB);
- Multi-Layers Perceptron (MLP);
- Long Short-Term Memory (LSTM);

Intraday models (encoder-decoder [4]):

- LSTM-MLP named ED-1;
- LSTM-LSTM named ED-2.

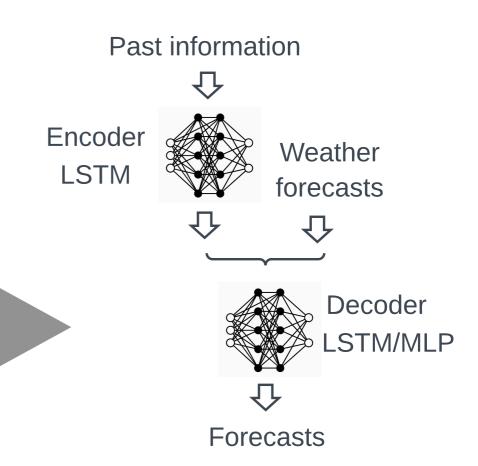


Figure 1: Encoder-decoder architecture.

[4] Bottieau, Jérémie, et al. "Very-short-term probabilistic forecasting for a risk-aware participation in the single price imbalance settlement." IEEE Transactions on Power Systems 35.2 (2019): 1218-1230.

Quality assessment

Value vs quality:

- quality: ability of the forecasts to genuinely inform of future events by mimicking the characteristics of the processes involved;
- value: benefits from using forecasts in a decision-making process such as participation to the electricity market. *Ex: in the capacity firming framework* [5].

Focus on **quality** evaluation:

- Continuous Rank Probability Score (CRPS) [6]
- Interval Score (IS) [6]
- [5] Dumas, Jonathan, et al. A Probabilistic Forecast-Driven Strategy for a Risk-Aware Participation in the Capacity Firming Market. Uliège, 2021. <u>Arxiv</u>, [under review for publication in IEEE Transactions on Sustainable Energy]
- [6] Gneiting, Tilmann, and Adrian E. Raftery. "Strictly proper scoring rules, prediction, and estimation." Journal of the American statistical Association 102.477 (2007): 359-378.

Day-ahead results: point and quantile forecasts

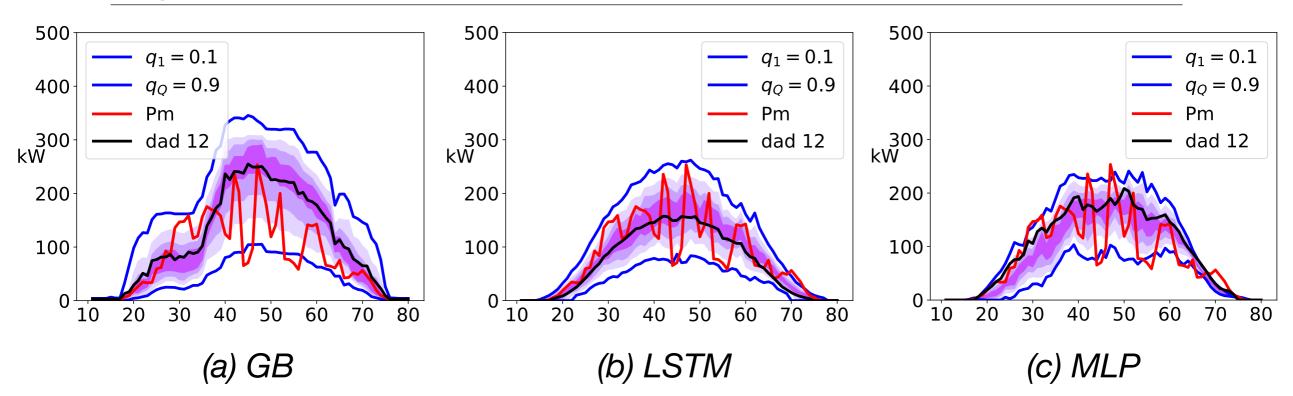


Figure 2: Quantiles vs point forecasts of day-ahead models on August 2, 2020.

Red line (Pm) = observations

Black line (dad 12) = day ahead point forecasts

Blue lines (q1, qQ) = 10 % and 90 % quantile forecasts

LSTM achieved the **best results** for both point & quantile forecasts.

Intraday results: point and quantile forecasts

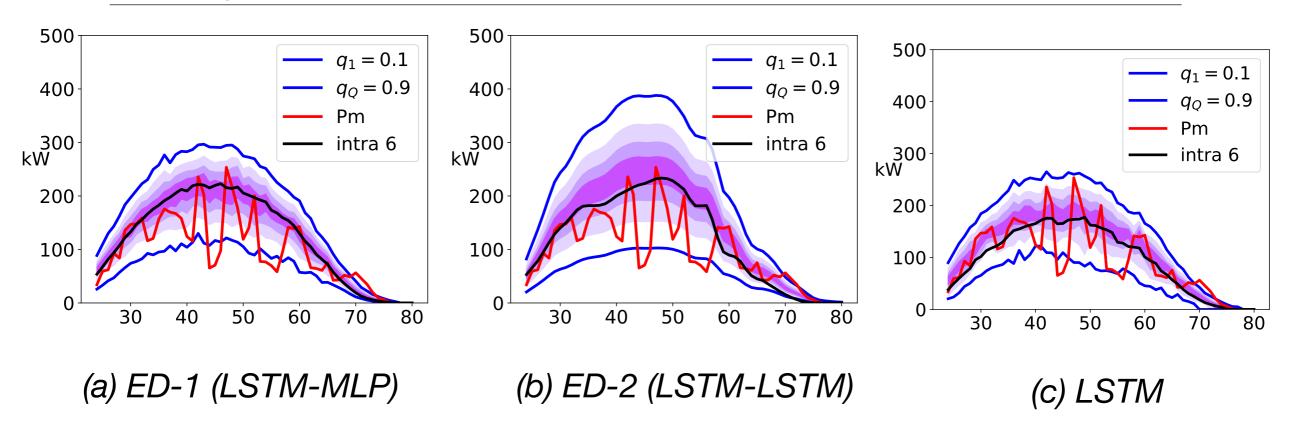


Figure 3: Quantiles vs point forecasts of intraday models of gate 6 on August 2, 2020.

Red line (Pm) = observations

Black line (intra 6) = intraday point forecasts

Blue lines (q1, qQ) = 10 % and 90 % quantile forecasts

LSTM-MLP yields accurate and calibrated forecast distributions.

Conclusions & perspectives

Conclusions:

- Day-ahead: **LSTM** achieved the **best results** for both point & quantile forecasts;
- Intraday: LSTM-MLP yields accurate and calibrated forecast distributions.

Extensions:

- A PV scenario approach based on the encoder-decoder architecture;
- Deep learning generative models such as Generative Adversial Networks (GANs), Variational AutoEncoders (VAEs), and Normalizing Flows (NFs) [under review for publication in International Journal of Forecasting].