

Empowering Generation Adequacy: The Impact of Energy Storages Evaluated with a System-wide Least-Squares Monte Carlo Approach

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Increasing uncertainties in the European power system & Increased integration of European electricity markets

- Raising shares of variable renewable generation
- Decreasing conventional power plant capacities
 Especially in Germany: nuclear phase out, coal phase out
- Uncertain availability of gas
 New concern since about two years
- Aging conventional power plant fleet with increasing availability concerns

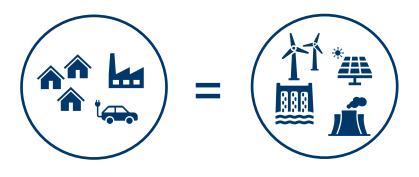
Notably French nuclear power plants, partly also other causes

Generation adequacy is the ability of the power system generation to meet the power demand at all points in time (European Commission, 2017).

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Traditional assessment using deterministic capacity balances provide unsatisfactory answers
 Need for probabilistic analyses considering stochastic events



Modeling storage operations appropriately is crucial to determine flexibility during scarcity events

- Renewable dominated energy systems have a large need for flexibility to balance fluctuations from renewable infeed and power demand
- Energy storages, e.g., battery or hydro storages or PtX technologies are flexible and can reduce supply shortages by shifting energy – but limited by their amount of stored energy
 - Battery and hydro storages: High efficiency, quick response, no carbon emissions, usually no fuel costs
 - PtX technologies: Flexible energy carriers,
- Flexible operation of storages more difficult to determine than for thermal power plants due to energy restrictions
 - Opportunity costs ("water value") determine storage operation
 - depend on future developments → comparison of continuation value (future revenues) and current revenues to determine optimal storage operation

Appropriate modeling of storage operations is crucial to determine available discharging capacities during supply shortages



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Traditionally: Storage operation is optimized based on system costs & depending on intertemporal effects

Thermal power plants: **constant** in time and **exogenously given**

Thermal power plants: **merit-order based** determination per timestep

Objective function minimizes system costs

• System costs are calculated as the product of **operational costs** and **power generation** per power plant

Storages: **opportunity costs** determine **storage operation** over the entire time period

Assumptions

- Perfect foresight
- Perfect competition
- Storage provides optimal arbitrage
- No investments

House of Energy Markets & Finance

Constraints

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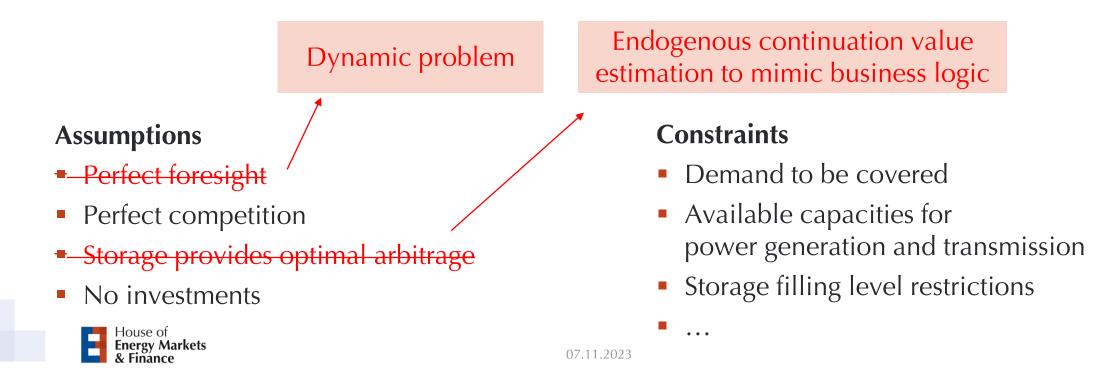
- Demand to be covered
- Available capacities for power generation and transmission
- Storage filling level restrictions

Traditionally: Storage operation is optimized based on system costs & depending on intertemporal effects

? How can we determine the **available discharging capacities** of **multiple large-scale energy storages** in each time step for a large number of Monte-Carlo paths?

Objective function minimizes system costs

• System costs are calculated as the product of **operational costs** and **power generation** per power plant



Least-Squares Monte Carlo: Conceptional introduction

Background and Usage

- Developed in finance for valuation of so-called American options by Longstaff and Schwartz (2001) to determine optimal execution of options
- Approximate continuation value (CV) as the value of not exercising the option
 - Option to charge or discharge storage
 - Least-squares regression to estimate CV
- Combination of stochastic modeling in Monte Carlo simulation and least-squares regression
- Previously used to evaluate individual storages based on exogeneous prices
- Here extension to multiple storages in a system context

1 Estimate future value (of storage) with an ordinary least-squares (OLS) regression based on a large number of Monte-Carlo paths

2 Determine storage operations for all Monte Carlo paths in a temporal forward propagation

Least-Squares Monte Carlo: Estimate future system costs

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Estimate future storage value with an OLS regression over all MC paths in a temporal **backward propagation**

for t = T-1 to 0 **do**

- B1: Load simulations, i.e., residual load D_t
- B2: Determine system costs excl. storage actions $C_t^{MR}(D_t)$
- **B3: Estimate CV for all state-of-charge (SOC) combinations**

based on $C_t^{MR}(D_t)$ and save regression coefficient $\hat{\beta}_{t,s}$

B4: Interpolate continuation values (CV) for all SOC combinations and possible actions $\mathbb{E}[C_{t+1,s'(o)}(D_t)]$ B5: Re-determine system costs incl. storage actions $C_{t,s}^{MR}(o, D_t)$

B6: Minimize total costs to determine optimal storage actions $\min_{o \in O^{Feas}(s)} C_{t,s}(o, D_t) = C_{t,s}^{MR}(o, D_t) + e^{-\delta\Delta t} \mathbb{E}[C_{t+1,s'(o)}(D_t)]$

Current

system costs

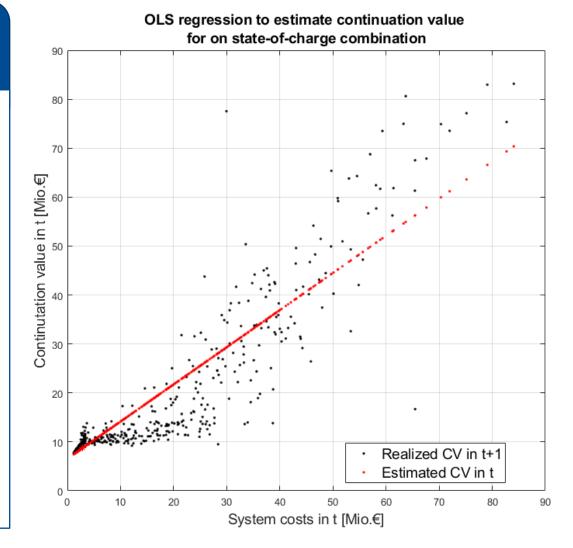
Expected future

system costs

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end for





Least-Squares Monte Carlo: Incorporate expected future system costs to determine storage operations

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Determine storage operations in all MC paths in a temporal forward propagation

for t = 1 to T do

- F1: Load simulations, i.a. residual load D_t
- F2: Determine possible storage actions o
- F3: Determine system costs excl. storage actions $C_t^{MR}(D_t)$

F4: Estimate CV for all SOC combinations based on $C_t^{MR}(D_t)$ and regression coefficients $\hat{\beta}_{t,s}$

F5: Interpolate CVs for all possible actions $\mathbb{E}[C_{t+1,o}(D_t)]$ F6: Re-determine system costs incl. storage actions $C_t^{MR}(o, D_t)$ F7: Minimize total costs to determine optimal storage actions $\min_{o \in O^{Feas}(s)} C_t(o, D_t) = C_t^{MR}(o, D_t) + e^{-\delta \Delta t} \mathbb{E}[C_{t+1,o}(D_t)]$ **Expected future**

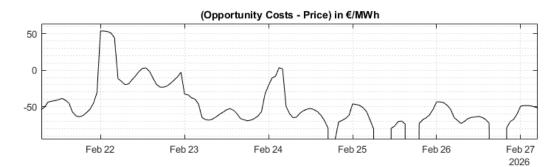
Current

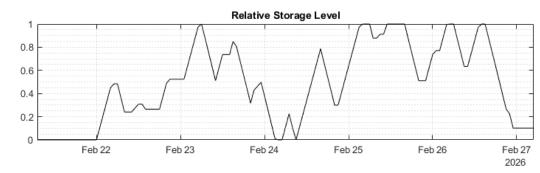
system costs

system costs

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Residual Load in GW 60 40 20 25/75%-Quantil 10/90%-Quantile Mean -20 Path X -40 Feb 22 Feb 24 Feb 25 Feb 27 Feb 23 Feb 26 2026





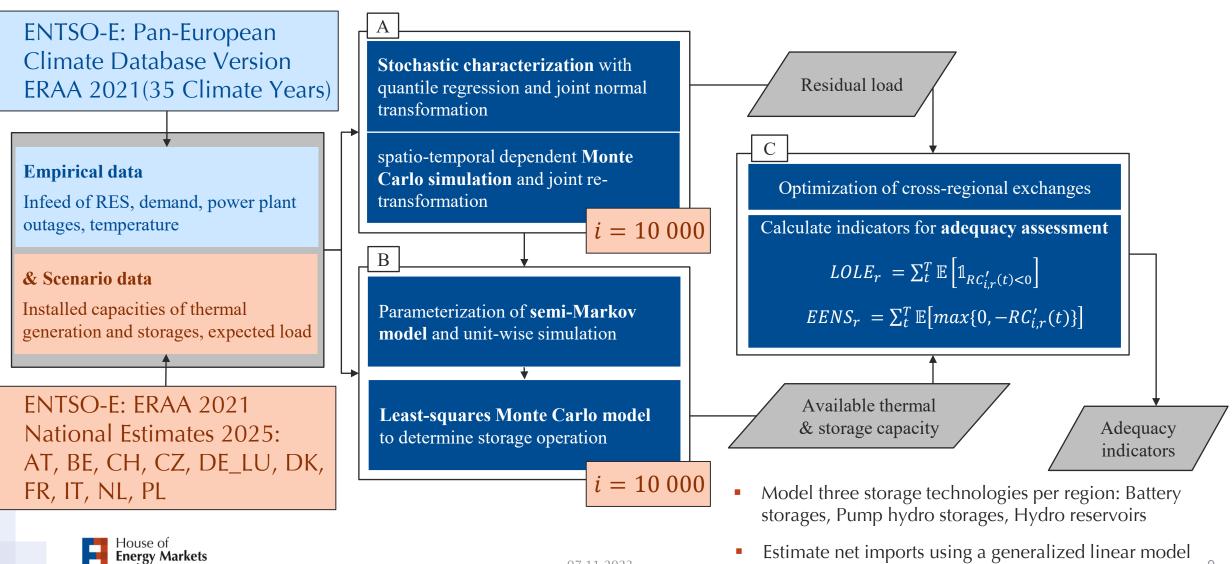
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Case Study: ERAA 2021 National Estimates 2025

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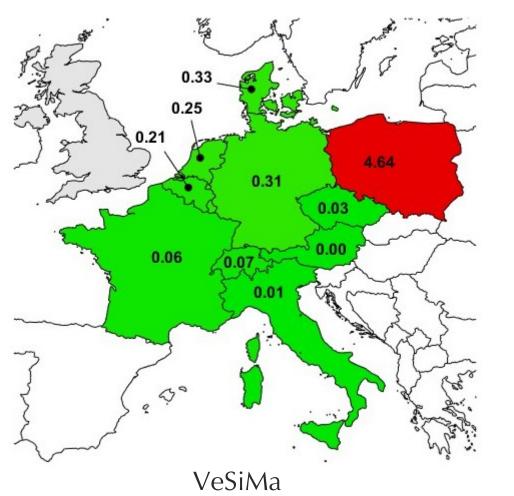


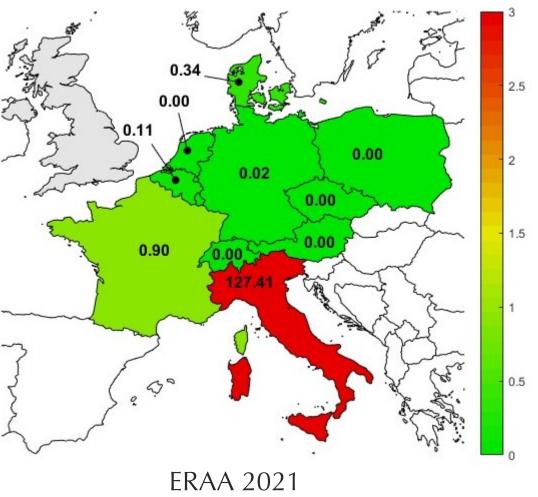
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Case Study: Loss of load expectation

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Scenario: ERAA 2021 National Estimates 2025 – No EVA & No CM





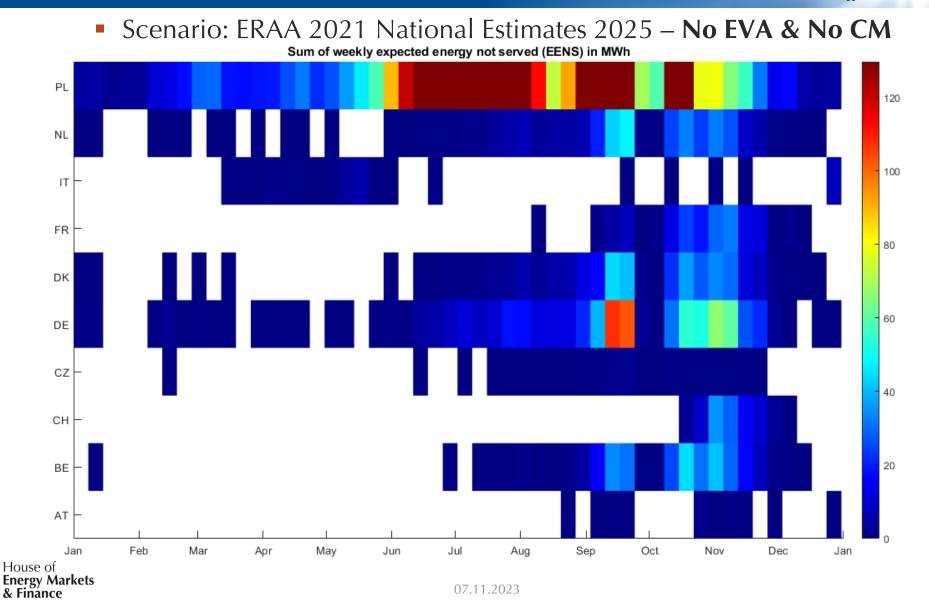


Mean national reliability standard: $LOLE \leq 3 h/a$

Case Study: Expected energy not served indicate seasonal differences

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System-wide least-squares Monte Carlo approach

- Stochastic determination of storage operations in without perfect foresight
- Computational feasibility restricts model scope: Trade-off between assumptions and stochastics
- Challenging to interpret interplay of storage technologies
- Potential to model flexibility of system components e.g., demand side management, ...

Adequacy Assessment

- Comprehensive stochastic modeling that provides a robust assessment of resource adequacy
- Insights in detailed simulation valuable for subsequent studies





Thank you for your attention!

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