

Modul: Intelligent Software Systems

Topic: A survey of stochastic optimization techniques for the unit commitment problem

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# **Summary of Literature**

## **Part 1**

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# 1 Notes to Papers

## 1.1 Paper 14

**Title of Paper** Data-Driven Adaptive Robust Unit Commitment Under Wind Power Uncertainty: A Bayesian Nonparametric Approach [2]

Adaptive Robust Optimierung für Integration von Windenergie Dirichlet-Prozessgemischmodell datengetriebener Unsicherheitssatz -  $\zeta$  unsicherheitssatz ermittelt: steht für alle unsicherheiten zusammengefasst

Ziel: robustes UC-Modell finden (mithilfe von Unsicherheitssatz und Windleistungsprognosen)

**Mathematical Formulation** Two-Stage adaptive robust UC model Model

- First Stage: Commitment Status of Generators
- Second Stage: Dispatch Decision of conventional Genrators and renewable wind power
- Goal: Minimize total operating cost
- Disadvantages:
  - No full use of complex uncertainty data information
  - Does not account correlation
  - Does not account asymmetry
  - Does not account multimodal nature of wind power forecast errors
  - Limited modeling flexibility
  - → Remedy: Data driven adaptive robust unit commitment optimization framework

**Data-Driven Adaptive Robust Unit Commitment Optimization Framework** und Text dahinter

**Dirichlet Process Mixture Model** • stochastic process

- Probability distribution over distributions
- Dirichlet distributed finite dimensional marginal distributions

- Motivation:
  - model distributions over observed data
  - unbounded complexity: underfitting is mitigated
- limited by the fact that generalizations from it are discrete distributions

**Data-Driven Uncertainty Set** Based on posterior predictive distribution

random vector: future wind forecast errors

self adaptive to underlying complexity and structure of given data

**Data-Driven Robust Unit Commitment Model**

- Solution Methodology**
- multilevel optimization structure & nonconvex nature of the proposed uncertainty set → specific solution algorithm needed
  - reduce four-level optimization problem into single-level full master problem (enumeration of all extreme points)
  - hard to calculate (large number of induced UC constraints)
  - → partial enumeration scheme of extreme points
  - identify worst-case wind forecast error scenario
  - compare optimal values of single subproblems to get largest one

## Computational Experiments

Studies on six-bus and IEEE 118-bus systems

**Illustrative Six-Bus System**

**IEEE 118-Bus System**

## 1.2 Paper 15

Title of paper: A Data-Driven Model of Virtual Power Plants in Day-Ahead Unit Commitment [1]

**Motivation** Ensuring effective integration of distributed energy resources

**Solution** Virtual power plants: condense them to single entity for wholesale market

**Problem to solve** Dependence on distributed power resources output: time varying and

not exactly known at day-ahead UC engine

**Task of this paper** Evaluating physical characteristics of VPP

- Max capacity
- Ramping capacity
- Uncertainty in wind power output
- Load consumption

### 1.3 Paper 17

**Motivation** Solving multistage stochastic unit commitment problem

**Solution** new type of decomposition algorithm (based on new framework of stochastic dual dynamic integer programming)

**label** description

### 1.4 Paper 21

**Problem** Uncertainty resulting from integration of variable renewable energy generations (wind-, solar power)

### 1.5 Paper 22

notes

[2]

## Literatur

- [1] S. Babaei, C. Zhao, and L. Fan. A data-driven model of virtual power plants in day-ahead unit commitment. *IEEE Transactions on Power Systems*, 34(6):5125–5135, Nov 2019.
- [2] C. Ning and F. You. Data-driven adaptive robust unit commitment under wind power uncertainty: A bayesian nonparametric approach. *IEEE Transactions on Power Systems*, 34(3):2409–2418, May 2019.