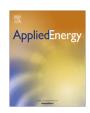
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# Economic optimization of operations for hybrid energy systems under variable markets



Jun Chen, Humberto E. Garcia\*

Idaho National Laboratory, Idaho Falls, ID 83415, USA

#### HIGHLIGHTS

- Propose method for maximizing system economics based on renewable and market info.
- Develop computational platform for performing operations optimization.
- Develop control to manage prediction error and storage to avoid energy imbalance.
- Illustrate advantages of optimizer by comparing flexible and constant operations.
- Perform sensitivity analysis under market variability and prediction error.

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#### ABSTRACT

Hybrid energy systems (HES) have been proposed to be an important element to enable increasing penetration of clean energy. This paper proposes a methodology for operations optimization to maximize their economic value based on predicted renewable generation and market information. A multi-environment computational platform for performing such operations optimization is also developed. To compensate for prediction error, a control strategy is accordingly designed to operate a standby energy storage element (ESE) to avoid energy imbalance within HES. The proposed operations optimizer allows systematic control of energy conversion for maximal economic value. Simulation results of two specific HES configurations illustrate the proposed methodology and computational capability. Economic advantages of such operations optimizer and associated flexible operations are demonstrated by comparing the economic performance of flexible operations with that of constant operations. Sensitivity analysis with respect to market variability and prediction error are also performed.

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#### 1. Introduction

#### 1.1. Background and motivation

Hybrid energy systems (HES) under flexible operations and variable energy generations/utilizations have been proposed to be an important element to enable higher penetration of clean energy generation, e.g., renewable and nuclear options, [1–9]. HES typically integrate multiple energy inputs (e.g., nuclear and renewable energy) and multiple energy outputs (e.g., electricity, gasoline, and fresh water) using complementary energy conversion processes. By enabling more than one option for energy utilization, HES configurations can change their electricity generation or consumption within a short time whenever requested.

\* Corresponding author. E-mail address: humberto.garcia@inl.gov (H.E. Garcia).

Prior works have been focused on dynamic modeling and simulation of diverse unit operations, together with their integration, control, and dynamic property characterization [5-8]. These results suggest that, from a technical point of view, HES can be operated under flexible operations schedules to accommodate the variability introduced from renewable generation, modern loads (such as electric vehicles), and markets. Such flexibility allows HES to participate in several wholesale markets, including markets for electrical energy, feedstock, and alternative energy outputs. Previous technical evaluation of HES has also shown that HES meet the requirements to bid into wholesale ancillary service (AS) market [5], to support the stability of the electric grid. A highlevel diagram of a general HES considered here is shown in Fig. 1, where HES take energy inputs from Controllable Energy Resources (CER) such as baseload generation (e.g., nuclear station), Variable Energy Resources (VER) such as wind farm, and Energy Storage Elements (ESE) such as electrical battery. HES typically include one or more Alternative Production Plants (APP) besides a Power Cycle

#### Nomenclature capital cost per unit of installed capacity of AHG amount of electrical energy sold in DAM $\alpha_{ahg}$ $P_{da,rt}^{ad}$ , $P_{da,rt}^{U}$ power held for RTM and its upper limit capital cost per unit of installed capacity of APP $\alpha_{app}$ capital cost per unit of installed capacity of ESE $P_{phg}$ power generated by PHG $\alpha_{ese}$ capital cost per unit of installed capacity of PHG $P_{ren}$ power generated by REN $\alpha_{phg}$ $P_{rt}$ capital cost per unit of installed capacity of REN amount of electrical energy sold in RTM $\alpha_{ren}$ taxation rate over CO<sub>2</sub> revenue for year k $\beta_{co_2}$ $R_k$ $\beta_{f\_ahg}$ fraction between $O\&M_{f\_ahg}$ and $C_{ahg}$ $r_R$ discount rate revenue from sale of alternative product $\beta_{f\_app}$ fraction between $O\&M_{f\_app}$ and $C_{app}$ $R_{app}$ fraction between $O\&M_{f\_ese}$ and $C_{ese}$ revenue from sale of ancillary service in DAM $R_{da.as}$ $\beta_{f\_ese}$ fraction between $O\&M_{f\_phg}$ and $C_{phg}$ $R_{da,e}$ revenue from sale of electrical energy in DAM $\beta_{f\_phg}$ fraction between $O\&M_{f\_ren}$ and $C_{ren}$ $R_{rt}$ revenue from sale of electrical energy in RTM $\beta_{f\_ren}$ price of the nth feedstock by AHG payback period $\beta_{v\_ahg,n}$ $T_{pb}$ price of the *n*th feedstock by APP installed capacity of AHG $\mathcal{N}_{ahg}$ $\beta_{\nu\_app,n}$ coefficient for computing GHG emission installed capacity of APP $\mathcal{N}_{app}$ $\gamma_{co_2}$ $\mathcal{N}_{\textit{ese},1}$ price of alternative product installed capacity of smoothing ESE $\pi_{app}$ price of ancillary service in DAM installed capacity of standby ESE $\pi_{da.as}$ $\mathcal{N}_{ese.2}$ price of electrical energy in DAM installed capacity of PHG $\pi_{da,e}$ $\mathcal{N}_{\textit{phg}}$ price of electrical energy in RTM $\mathcal{N}_{ren}$ installed capacity of REN $\pi_{rt}$ depreciation and amortization rate for year k AHG auxiliary heat generation $\rho_{da,k}$ alternative production plants tax rate APP prediction of corresponding variables ancillary service AS $B_1, B_2, B_3$ feasibility conditions CER controllable energy resources auxiliary heat generation capital cost CHP combined heat and power $C_{ahg}$ $C_{app}$ alternative production plant capital cost CM commodity market total capital cost DAM, DAO day-ahead market, day-ahead optimizer $C_{cap}$ energy storage system capital cost database $C_{ese}$ DB cost for GHG emission for year k **ESE** energy storage systems $C_{ghg,k}$ operations and maintenance cost for year k FM feedstock market $C_{0\&M.k}$ primary heat generation capital cost FMI functional mockup interface $C_{phg}$ renewable energy generation capital cost **FOM** figure of merit $CAPEX_k$ forward market capital expense for year k ForM depreciation and amortization for year k**GHG** greenhouse gas $DA_k$ $FCFF_{R,k}$ year-k real discounted free cash flow to firm GPP gasoline production plant hybrid energy systems inflation rate HES $M_{app}$ production rate of alternative product HES\_FEL HES with flexible electrical load $M_{co_2}$ combined CO2 emission HES\_FTL HES with flexible thermal load consuming rate of the nth feedstock by AHG $M_{v\_ahg,n}$ **HRES** hybrid renewable energy systems consuming rate of the nth feedstock by APP internal rate of return $M_{v\_app,n}$ IRR operations and maintenance MW. MW h megawatt and megawatt-hour 0&M $0\&M_f$ fixed operations and maintenance cost natural gas NG NPV $0\&M_v$ variable operations and maintenance cost net present value $O\&M_{f\_ahg}$ fixed O&M cost of AHG PC power cycle $O\&M_{f\_app}$ fixed O&M cost of APP PHG primary heat generation

(PC) for electricity generation. These APP allow the repurposing of energy (in form of thermal energy and/or electrical energy) for non-electricity commodity production. HES interrelate with feedstock market  $FM_i$  for procurement of feedstock material  $f_i$ , with power market PM for the sale of electricity and ancillary service, and with commodity market  $CM_j$  for the sale of commodity  $c_j$  (alternative energy output). Furthermore, each market (FM, PM and CM) in turn includes several forward and spot markets.

power generated by PHG and consumed by APP

probability of reserved capacity to be called for

minimum power consumed by APP

maximum power consumed by APP

 $P_{da,as}^{U}$ ,  $P_{da,as}^{U}$  AS sold in DAM and its upper limit

 $O\&M_{f\_ese}$  fixed O&M cost of ESE

 $O\&M_{f\_phg}$  fixed O&M cost of PHG

 $O\&M_{f\_ren}$  fixed O&M cost of REN

 $P_{app}$ 

 $P_{app}^{L}$ 

 $P_{app}^{U}$ 

 $O\&M_{\nu\_ahg}$  variable O&M cost of AHG

 $O\&M_{\nu\_app}$  variable O&M cost of APP

Hence the objective of this paper is to develop a generic methodology and computational platform for computing operations schedule among HES constituents for optimal economic performance. As shown in Fig. 2, such operations optimizer collects predicted information on VER generation and markets (denoted with dash lines), and updates the operations of the given HES through low-level controllers. Since HES participate in ancillary service market, controllers are also subject to grid system operator commands in case that reserved capacity is called upon. Note that since prediction error can cause energy imbalance within HES, an ESE is utilized to ensure energy balance at all time, as shown in Fig. 1.

power market

photovoltaics

spot market

real-time market

real-time optimizer

renewable energy input

variable energy resources

reverse osmosis desalination plant

weighted average cost of capital

PM

ΡV

REN

RODP

**RTM** 

RTO

SM

VER

WACC

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