

Cascade Hydroelectric Power System Model and its Application to an Optimal Dispatch Design

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Abstract—In this paper we propose an optimal dispatch scheme for a cascade hydroelectric power system that maximises the system efficiency, and minimises the spillage effects. Our approach proposes a methodology that has low computational burden and may be implemented for the short-term operation of a cascade hydroelectric power system. To this end, the non-linear relationships that describe the system physical constraints, e.g., power output, are transformed into affine relationships; thus reducing the computational complexity. The transformations are based on the construction of convex envelopes around bilinear functions; piecewise affine functions; and exploitation of optimisation properties. We demonstrate the efficacy of the proposed methodology with the Seven Forks system located in Kenya, and evaluate the performance of our method in terms of water volume and potential energy saved.

I. INTRODUCTION

The last years there has been an increase of renewable-based resource penetration into the electric grid that leads to the gradual de-carbonization of the energy supply sector [1]. Apart from a 40% reduction in greenhouse gas emissions and a 30% improvement in Energy Efficiency, the 2030 EU-wide targets set a minimum of at least 27% share of renewable energy resources in the energy consumption. However, renewable resources have unique characteristics and their integration raises several challenges into power systems operation. The variability of the renewable-based generation requires that the system has enough ramping capability to follow the net load variations in different time frames, ranging from seconds to hours. Furthermore, there is an increase in the need for balancing services, larger reserves for the frequency control, as well as better ramping capability. There are several technologies available that may satisfy these needs. However, they are often associated with either additional cost or partial loss of the energy output. In this regard, appropriate use of existing resources, such as hydroelectric power systems, without any extra cost are worth to be investigated. Hydroelectric power systems are fitting candidates since they have good ramping capability and energy storage possibility in form of hydro reservoirs. Thus, they may be used to smooth the output of renewable-based generation and resolve any potential problems caused to the grid due to solar output variability and intermittency.

In this regard, hydroelectric power systems may be used as “storage” devices for renewable generation. Cascade hydroelectric power systems are usually coupled both electrically,

i.e., they are used to meet same load; and hydraulically, i.e., the water outflow from one hydroelectric power plant is a significant portion of the inflow to the downstream plants [2, Ch. 7]. These relationships result in non-linear and non-convex expressions that are due to the spatial-temporal coupling among reservoirs; and for every plant, the nonlinear dependence between the power output, the water discharged, and the head of the associated reservoir. Such an event is a challenge for independent system operators (ISOs), who are responsible for the operation of hydroelectric power plants and do not usually have optimisation tools to efficiently use the generation resources [3]. There is a need to develop cascade hydroelectric power system dispatch tools that provide a balance between accuracy and complexity; and may be used for the short-term operation of cascade hydroelectric power plants.

Several papers have addressed the problem of investigating the new role of hydroelectric power systems in the power systems paradigm. The authors in [4] show that hydroelectric systems play an important role in future power systems, where renewable resources are present, and support system adequacy in case of supply shortfall. In [5] the coordination of hydroelectric systems and wind generators in order to minimise wind energy curtailments during congestion situations is analysed. A case study on using a cascaded hydropower system to firm wind generation is presented in [6]. In [7] a global optimisation of the short term scheduling for hydroelectric power generation with mixed integer nonlinear programming formulation of a cascade of hydro plants is presented. However, in a security constrained short-term hydrothermal dispatch problem for large-scale systems simplified models are also necessary as discussed in [8]. The authors in [9] propose a semidefinite programming method to solve a hydrothermal coordination problem; they reformulate and relax the non-convex constraints associated with hydroelectric power system while guaranteeing global optimality.

In this paper, we propose a short-term optimal dispatch framework for cascade hydroelectric power systems that may be used to operate a cascade hydroelectric power system taking into account the presence of renewable-based generation. The main objective is to use water efficiently, i.e., use the minimum amount of water to meet the energy target. As a result, the water stored in the reservoirs is higher thus increasing the system resiliency in dry months and benefiting irrigation. The

maximum efficiency of the hydroelectric power plant occurs when the reservoir is full because the power output for a given amount of water is higher. In this regard, we construct the cascade hydroelectric system optimal dispatch by appropriately choosing the objective function, i.e., maximise the water level in the dam, and representing the physical and power balance constraints. Unfortunately, the constructed problem contains a linear objective function but non-convex (bilinear) constraints. This formulation needs to be relaxed in an efficient manner so that its actual implementation in the short-term operation of a cascade hydroelectric power system is realistic. To this end, the non-linear terms are relaxed by the construction of convex envelopes around the bilinear functions; piecewise affine functions; and exploitation of optimisation properties of the simplex method. We demonstrate the implementation of the proposed methodology in a real cascade hydroelectric power system the Seven Forks system in Kenya, which consists of five hydroelectric power plants.

II. PRELIMINARIES

In this section, we introduce the hydropower function and the scheduling constraints that are utilised to develop our framework. We consider a hydroelectric power system with N hydroelectric power plants indexed by $\mathcal{N} = \{1, \dots, N\}$ that we wish to schedule for a time period $\mathcal{T} = \{T_1, \dots, T_T\}$, which correspond to hourly intervals.

A hydroelectric power plant $i \in \mathcal{N}$ may be characterised by its input-output curves. The input is in terms of water discharge and the output is in terms of power generation. The power generated by a hydroelectric power plant depends on the characteristics of the net hydraulic head, i.e., the difference between the level of the reservoir and the tail water, and the water discharge. In particular, the power of a hydroelectric power plant i at time t is defined as

$$P_{H_{i_t}} = \eta_i(h_{i_t}, q_{i_t}) \rho g h_{i_t} q_{i_t}, \forall i \in \mathcal{N}, \forall t \in \mathcal{T}, \quad (1)$$

where ρ is the density of the water in kg/m^3 ; g is the gravitational acceleration in m/s^2 ; h_{i_t} is the net head of water (the difference in water level between upstream and downstream of the turbine) of hydropower plant i at time t in m; q_{i_t} is the discharge of water of plant i during time t in m^3/s ; $\eta_i(h_{i_t}, q_{i_t})$ is the efficiency of the turbine generator at head h_{i_t} and discharge q_{i_t} .

The output of a hydroelectric power system is used to meet the load at every time instant $t \in \mathcal{T}$. In this regard, we have

$$\sum_{i \in \mathcal{N}} P_{H_{i_t}} = P_{L_t}, \forall t \in \mathcal{T}, \quad (2)$$

where P_{L_t} is the load at time t . We should take into account that the power output of each hydroelectric power plant $i \in \mathcal{N}$ is constrained by a minimum and a maximum output, i.e., $P_{H_i}^m \leq P_{H_{i_t}} \leq P_{H_i}^M$, for all $t \in \mathcal{T}$. Similar statements are true for the head levels and the water discharge rates. Thus, we have that $h_i^m \leq h_{i_t} \leq h_i^M$ and $q_i^m \leq q_{i_t} \leq q_i^M$ for all $t \in \mathcal{T}$ and $i \in \mathcal{N}$.

Another physical constraint that needs to be taken into consideration in the operation of a cascade hydroelectric power

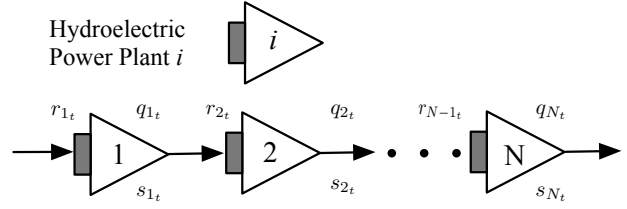


Fig. 1: Schematic representation of a hydroelectric power system.

system is the water balance between reservoirs. This balance equation relates the live volumes of the reservoirs, total discharges, spillages, and inflows. Evaporation and percolation losses may be included into the expected inflows; thus there is no need to be considered separately. Furthermore, in most cases there is a delay in the flow of water between reservoirs that needs to be accounted for in the modelling. A schematic representation of the cascade hydroelectric power system denoted by \mathcal{N} is depicted in Fig. 1. A mathematical formulation of the water balance of the cascade hydroelectric power system with the use of the hydraulic continuity equations is given below:

$$V_{1_t} = V_{1_{t-1}} + r_{1_t} - q_{1_t} - s_{1_t}, \quad (3)$$

$$V_{2_t} = V_{2_{t-1}} + r_{2_t} + q_{1_{t-\tau_1}} + s_{1_{t-\tau_1}} - q_{2_t} - s_{2_t}, \quad (4)$$

⋮

$$V_{N_t} = V_{N_{t-1}} + r_{N-1_t} + q_{N-1_{t-\tau_{N-1}}} + s_{N-1_{t-\tau_{N-1}}} - q_{N_t} - s_{N_t}, \quad (5)$$

where V_{i_t} is the live volume of hydroelectric power plant i at the end of time t in m^3 ; τ_i is the time delay between reservoir i and $i+1$, i.e., the time water needs to travel from one to the other; r_{i_t} is the inflow into hydroelectric power plant i during time t ; s_{i_t} is the spillage discharge of hydroelectric power plant i during time t . There are constraints associated with the reservoir storage volume limits of each hydroelectric power plant $i \in \mathcal{N}$, which are defined as $V_i^m \leq V_{i_t} \leq V_i^M$, for all $t \in \mathcal{T}$. Some additional physical constraints that may be taken into consideration are the initial and terminal reservoir storage volumes, i.e., $V_{i_1} = V_{i_{\text{start}}}$ and $V_{i_T} = V_{i_{\text{end}}}$ for all $i \in \mathcal{N}$.

The modelling of water stored in a reservoir and its mapping to a certain head level is important since it relates (1) with (3)-(5). This relationship in most reservoirs is determined from topographical surveys of the dam site and is highly non convex [10]. We denote this relationship by $h_{i_t} = \phi_i(V_{i_t})$, for all $i \in \mathcal{N}$ and $t \in \mathcal{T}$.

III. HYDROELECTRIC SYSTEM OPTIMAL DISPATCH

In order to formulate the hydroelectric system optimal dispatch, we first need to define a set of requirements that the system must satisfy: (i) maximise the system efficiency, and (ii) minimise the spillage effects. The maximum efficiency of the hydroelectric power plant occurs when the reservoir is full. The main reason behind this statement is that for a given water discharge the higher the head the higher the power output as

it may be seen in (1). To this end, we wish to maximise the head of each reservoir at every time instant, i.e., h_{i_t} , for all $i \in \mathcal{N}, t \in \mathcal{T}$. The minimisation of the spillage effects are accomplished by including in the objective function the term $\sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{N}} M s_{i_t}$. Then, the optimal dispatch is given by

$$\begin{aligned}
& \underset{\substack{h_{i_t}, P_{i_t}, s_{i_t} \\ V_{i_t}, q_{i_t}}}{\text{maximise}} \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{N}} h_{i_t} - \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{N}} M s_{i_t} \\
& \text{subject to } \sum_{i \in \mathcal{N}} P_{H_{i_t}} = P_{L_t}, \forall t \in \mathcal{T} \\
& P_{H_{i_t}} = \eta_i(h_{i_t}, q_{i_t}) \rho g h_{i_t} q_{i_t}, \forall i \in \mathcal{N}, \forall t \in \mathcal{T} \\
& V_{1_t} = V_{1_{t-1}} + r_{1_t} - q_{1_t} - s_{1_t}, \forall t \in \mathcal{T} \setminus \{T_1, T_T\} \\
& V_{i_t} = V_{i_{t-1}} + r_{i-1_t} + q_{i-1_{t-\tau_{i-1}}} + s_{i-1_{t-\tau_{i-1}}} \\
& \quad - q_{i_t} - s_{i_t}, \forall i \in \mathcal{N} \setminus \{1\}, \forall t \in \mathcal{T} \setminus \{T_1, T_T\} \\
& V_{i_1} = V_{i_{\text{start}}}, \forall i \in \mathcal{N} \\
& V_{i_T} = V_{i_{\text{end}}}, \forall i \in \mathcal{N} \\
& h_{i_t} = \phi_i(V_{i_t}), \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& V_{i_t}^m \leq V_{i_t} \leq V_{i_t}^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& q_{i_t}^m \leq q_{i_t} \leq q_{i_t}^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& h_{i_t}^m \leq h_{i_t} \leq h_{i_t}^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& P_{H_{i_t}}^m \leq P_{H_{i_t}} \leq P_{H_{i_t}}^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& (V_{i_t} - V_{i_t}^M) s_{i_t} = 0, \forall i \in \mathcal{N}, t \in \mathcal{T} \\
& s_{i_t} \geq 0, \forall i \in \mathcal{N}, t \in \mathcal{T},
\end{aligned} \tag{6}$$

where $\tau = \max\{\tau_1, \dots, \tau_{N-1}\}$ is the maximum of the delays and $T_1 > \tau$; and M is a large positive number. The output of (6) determines the head levels, power output, volume, spillage and water discharge for every hydroelectric power plant at every time instant in the period of interest. As it may be seen (6) has a linear objective function with linear and non-linear constraints. In order to solve (6) several algorithms may be used, e.g., the gradient method. However, this optimisation problem provides the output of a cascade hydroelectric power system in short time scales; thus, needs to be computationally efficient. To this end, we relax the original optimal dispatch problem given in (6).

A. Relaxations of Original Problem

The non-linear constraints of (6) refer to (i) the output of a hydroelectric power system, i.e., $P_{H_{i_t}} = \eta_i(h_{i_t}, q_{i_t}) \rho g h_{i_t} q_{i_t}, \forall i \in \mathcal{N}, \forall t \in \mathcal{T}$; (ii) the mapping of the head level to the volume, i.e., $h_{i_t} = \phi_i(V_{i_t}), \forall i \in \mathcal{N}, t \in \mathcal{T}$; and the spillage effects, i.e., $(V_{i_t} - V_{i_t}^M) s_{i_t} = 0, \forall i \in \mathcal{N}, t \in \mathcal{T}$. The objective is to transform these constraints and turn (6) into a linear optimisation problem, which has very low computational burden to solve with existing techniques.

The non-convex relationship of the output of a hydroelectric power system, the head and the water discharge is a bilinear function for a constant turbine efficiency. We assume the efficiency is constant, i.e., $\eta_i(h_{i_t}, q_{i_t}) = \eta_i$ and replace the remaining bilinear term with a convex envelope consisting of linear over- and underestimating inequality constraints to trans-

form the non-convex constraint into a set of linear inequality constraints. In particular, we have that $P_{H_{i_t}} = a_i h_{i_t} q_{i_t}$, with $a_i = \eta_i \rho g$, and $q_{i_t}^m \leq q_{i_t} \leq q_{i_t}^M, h_{i_t}^m \leq h_{i_t} \leq h_{i_t}^M$. By using McCormick's envelopes (e.g., [11]), we obtain:

$$P_{H_{i_t}} \geq q_{i_t}^m h_{i_t} + h_{i_t}^m q_{i_t} - h_{i_t}^m q_{i_t}^m, \tag{7}$$

$$P_{H_{i_t}} \geq q_{i_t}^M h_{i_t} + h_{i_t}^M q_{i_t} - h_{i_t}^M q_{i_t}^M, \tag{8}$$

$$P_{H_{i_t}} \leq q_{i_t}^m h_{i_t} + h_{i_t}^M q_{i_t} - h_{i_t}^M q_{i_t}^m, \tag{9}$$

$$P_{H_{i_t}} \leq q_{i_t}^M h_{i_t} + h_{i_t}^m q_{i_t} - h_{i_t}^m q_{i_t}^M. \tag{10}$$

One of the most useful applications of the piecewise linear representation is for approximating nonlinear functions. In this regard, we convert the non-linear mapping of the head level to the volume, i.e., $h_{i_t} = \phi_i(V_{i_t})$ to a piecewise affine relationship. We consider $k = 1, \dots, K$ intervals and thus have:

$$h_{i_t} = \beta_i^k v_{i_t}^k + \gamma_i^k, v_{i_t}^k \in [\zeta_k, \zeta_{k+1}], \text{ for } k = 1, \dots, K, \tag{11}$$

where $\beta_i^k, \gamma_i^k \in \mathbb{R}^+ = \{x \in \mathbb{R} | x \geq 0\}$, for $k = 1, \dots, K, \zeta_1 < \zeta_2 < \dots < \zeta_{K+1}, v_{i_t}^k \in \mathbb{R}^+$ with $V_{i_t} = \sum_{k=1}^K v_{i_t}^k$. An analysis of the piecewise affine approximation of the head to the volume of existing reservoirs based on data found in [12] showed that $\beta_i^1 > \beta_i^2 > \dots > \beta_i^K$. Usually, piecewise linear functions are formulated as mixed integer programming problems, which would increase the complexity of (6). However, a special case for representing piecewise linear functions arises when diseconomies of scale apply, i.e., when $\beta_i^1 > \beta_i^2 > \dots > \beta_i^K$ and we are maximising h_{i_t} , which is this case (e.g. [13]). Thus, (11) may be rewritten as

$$h_{i_t} = \sum_{k=1}^K (\beta_i^k v_{i_t}^k) + \gamma_i^1, v_{i_t}^k \in [0, \zeta_{k+1} - \zeta_k], \text{ for } k = 1, \dots, K. \tag{12}$$

Notice the slight abuse of notation with the term $v_{i_t}^k$, which represents different quantities in (11) and (12); the interpretation will always be clear from the context.

The spillage effects, s_{i_t} are included in the modelling with (3)-(5). In the absence of spillage (3)-(5) would be satisfied with $s_{i_t} = 0$. Thus, we may see s_{i_t} as an "artificial" variable that increases the feasible space of the problem and also has a physical meaning. The optimal dispatch objective function is designed to minimise the spillage effects, i.e., make $s_{i_t} = 0$ when it is feasible. The same concept is true for the "artificial" variables in linear programming where penalty terms, e.g., with the use of the Big-M method (e.g., [14]), are added to the objective function to push out the "artificial" variables from the basic variables, i.e., basis. With the Big-M method in a minimisation problem, if M is large, then any basis that includes a positive artificial variable will lead to a large positive value of the objective function. If there is any basic feasible solution to the constraints of the original linear program, then the corresponding basis will not include any artificial variables and its objective value will be much smaller. Because the artificial variables have a high cost associated with them, the simplex method removes them from the basis if this is at all possible. Any basic feasible solution to the penalised problem in which all the artificial variables are

nonbasic (and hence zero) is also a basic feasible solution to the original problem. The corresponding basis can be used as an initial basis for the original problem. Thus, we argue that by including s_{it} in the objective function multiplied with a large positive number M , then it will non zero only when $V_{i,t-1} + r_{i-1,t} + q_{i-1,t-\tau_{i-1}} > V_i^M \Rightarrow V_{it} > V_i^M$. When $V_{it} < V_i^M$, then $s_{it} = 0$ as explained above. As a result, the non-linear constraint $(V_{it} - V_i^M)s_{it} = 0$ is satisfied and does not need to be included in (6).

B. Overall Framework

The proposed optimal dispatch of a hydroelectric power system given in (6) may be rewritten in its relaxed form as

$$\begin{aligned} & \underset{h_{it}, P_{it}, s_{it}}{\text{maximise}} \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{N}} h_{it} - \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{N}} M s_{it} \\ & \underset{V_{it}, q_{it}, v_{it}^k}{\text{subject to}} \end{aligned}$$

$$\begin{aligned} & \sum_{i \in \mathcal{N}} P_{H_{it}} = P_{L_t}, \forall t \in \mathcal{T} \\ & V_{1t} = V_{1,t-1} + r_{1t} - q_{1t} - s_{1t}, \forall t \in \mathcal{T} \setminus \{T_1, T_T\} \\ & V_{it} = V_{i,t-1} + r_{i-1,t} + q_{i-1,t-\tau_{i-1}} + s_{i-1,t-\tau_{i-1}} \\ & \quad - q_{it} - s_{it}, \forall i \in \mathcal{N} \setminus \{1\}, \forall t \in \mathcal{T} \setminus \{T_1, T_T\} \\ & V_{i_1} = V_{i_{\text{start}}}, \forall i \in \mathcal{N} \\ & V_{i_T} = V_{i_{\text{end}}}, \forall i \in \mathcal{N} \\ & h_{it} = \sum_{k=1}^K (\beta_i^k v_{it}^k) + \gamma_i^1, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & V_{it} = \sum_{k=1}^K v_{it}^k, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & V_i^m \leq V_{it} \leq V_i^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & q_i^m \leq q_{it} \leq q_i^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & h_i^m \leq h_{it} \leq h_i^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & P_{H_{it}}^m \leq P_{H_{it}} \leq P_{H_{it}}^M, \forall i \in \mathcal{N}, t \in \mathcal{T} \\ & s_{it} \geq 0, \forall i \in \mathcal{N}, t \in \mathcal{T}, \\ & 0 \leq v_{it}^k \leq \zeta_{k+1} - \zeta_k, k = 1, \dots, K, \forall i \in \mathcal{N}, t \in \mathcal{T}, \\ & P_{H_{it}} \geq q_i^m h_{it} + h_i^m q_{it} - h_i^m q_i^m, \forall i \in \mathcal{N}, t \in \mathcal{T}, \\ & P_{H_{it}} \geq q_i^M h_{it} + h_i^M q_{it} - h_i^M q_i^M, \forall i \in \mathcal{N}, t \in \mathcal{T}, \\ & P_{H_{it}} \leq q_i^m h_{it} + h_i^M q_{it} - h_i^M q_i^m, \forall i \in \mathcal{N}, t \in \mathcal{T}, \\ & P_{H_{it}} \leq q_i^M h_{it} + h_i^m q_{it} - h_i^m q_i^M, \forall i \in \mathcal{N}, t \in \mathcal{T}. \end{aligned} \tag{13}$$

Reservoir	Masinga	Kamburu	Gitaru	Kindaruma	Kiambere
P_i^M [MW]	40	93	225	72	165
V_i^M [Mm ³]	1753	133	21	10	519
h_i^M [m]	51	78	140	35	151
h_i^m [m]	25	61	131	31	134
q_i^M [m ³ /s]	198.8	161.82	189	265.68	132
η_i	variable	0.9	0.92	0.89	0.9

TABLE I: Seven Forks system data.

IV. NUMERICAL EXAMPLES

In this section, we illustrate the proposed optimal dispatch of a cascade hydroelectric power system with the Seven Forks system in Kenya, which consists of five hydroelectric power plants [10]. The time horizon we wish to schedule its operation is for a one year period with daily schedules. The constraints of the Seven Forks systems cascade in terms of power output, live volume, head, and turbine efficiency characteristics may be found in [12] and are shown in Table I. The minimum power output, live volume, and water discharge rate for all reservoirs are zero, i.e., $P_i^m = 0$, $V_i^m = 0$, $q_i^m = 0$ for $i = 1, \dots, 5$. In order to determine P_{L_t} for $t \in \mathcal{T}$ for the entire year we use a scaled version of the national load of Kenya for 2011, which was provided to us through Kenya Electricity Generating Company (KenGen). In order to make sure that the installed capacity of the hydroelectric power system is sufficient, we constrain the load to a maximum value of 500 MW. A typical daily normalised demand profile for Kenya was used [15] is depicted in Fig. 2.

In order to construct the relaxed problem, as described in Section III-A, we approximate the relationship between the head levels and the live volume with a piecewise affine function. We use the data provided in [12] and in Fig. 3, we depict the actual relationship and the piecewise affine approximation between the head level and live volume for the Gitaru reservoir. As it may be seen the error introduced is marginal. For all the dams, we choose $K = 3$, i.e., we calculate the piecewise affine functions into three segments. For the five reservoirs we have:

$$\begin{aligned} h_{1t} &= 0.0281v_{1t}^1 + 0.0131v_{1t}^2 + 0.0084v_{1t}^3 + 25, \\ h_{2t} &= 0.3077v_{2t}^1 + 0.1351v_{2t}^2 + 0.0964v_{2t}^3 + 61, \\ h_{3t} &= 0.6349v_{3t}^1 + 0.4301v_{3t}^2 + 0.1667v_{3t}^3 + 131, \\ h_{4t} &= 0.5779v_{4t}^1 + 0.4117v_{4t}^2 + 0.2955v_{4t}^3 + 31, \\ h_{5t} &= 0.0648v_{5t}^1 + 0.0468v_{5t}^2 + 0.0398v_{5t}^3 + 134, \end{aligned}$$

with $v_{1t}^1 \in [0, 400]$, $v_{1t}^2 \in [0, 731]$, $v_{1t}^3 \in [0, 622]$; $v_{2t}^1 \in [0, 14]$, $v_{2t}^2 \in [0, 37]$, $v_{2t}^3 \in [0, 82]$; $v_{3t}^1 \in [0, 6]$, $v_{3t}^2 \in [0, 9]$, $v_{3t}^3 \in [0, 6]$; $v_{4t}^1 \in [0, 4]$, $v_{4t}^2 \in [0, 3]$, $v_{4t}^3 \in [0, 3]$; and $v_{5t}^1 \in [0, 292]$, $v_{5t}^2 \in [0, 137]$, $v_{5t}^3 \in [0, 90]$. The units for the live volumes are in Mm³ and for the head in m.

KenGen also provided hourly historical data of the inflow data for all the hydroelectric power system from July 2015-June 2016. In the Seven Forks system there are three main

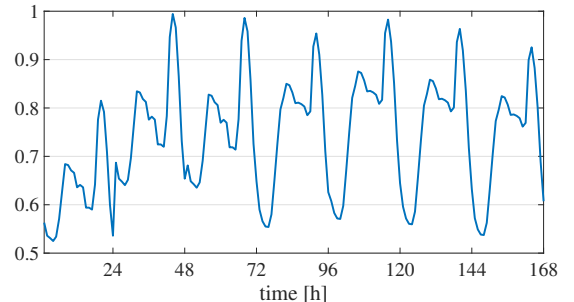


Fig. 2: Typical weekly normalised demand profile for Kenya.

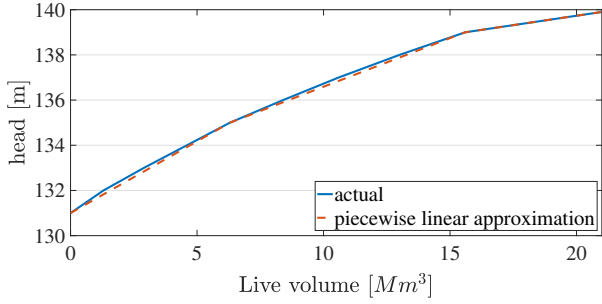


Fig. 3: Head level and live volume for Gitaru reservoir.

inflow streams; in Masinga, Kamburu and Kiambere. The remaining inflow into the dams is due to rainfall data. Based on these data, the time delay between all dams was determined to be: Masigna-Kamburu, $\tau_1 = 2$ hours; Kamburu-Gitaru, $\tau_2 = 0 < 1$ hour; Gitaru-Kindaruma, $\tau_3 = 0 < 1$ hour; and Kindaruma-Kiambere, $\tau_4 = 4$ hours. The starting volume constraint for each reservoir is: $V_{1\text{start}} = 1173 \text{ Mm}^3$; $V_{2\text{start}} = 118 \text{ Mm}^3$; $V_{3\text{start}} = 13 \text{ Mm}^3$; $V_{4\text{start}} = 4 \text{ Mm}^3$; and $V_{5\text{start}} = 420 \text{ Mm}^3$. There is no ending volume constraint.

In order to solve (13), we define M to be 10^8 . In order to test the performance of the proposed dispatch we run compare the results between two methods: (i) dispatch proportional to the maximum capacity of each hydroelectric power plant; and (ii) proposed optimal cascade hydroelectric power plant dispatch. The average daily live volume is a lot higher for method (ii) compared to (i); thus the resiliency of the system is increased. In particular, the daily live volume for method (i) is $2,006 \text{ Mm}^3$ and method (ii) $2,070 \text{ Mm}^3$ and the daily potential energy was increased by 5%. Method (ii) tries to maximise the head levels; thus keeps the Gitaru head to high value increasing the dam's efficiency. It should be noted that Gitaru is the largest in capacity dam, as depicted in Fig. 4.

V. CONCLUDING REMARKS AND EXTENSIONS

In this paper, we proposed an optimal dispatch scheme for the short-term operation of a cascade hydroelectric power system. More specifically, we constructed a non-linear optimisation problem that represents explicitly the physical constraints of hydroelectric power systems; and relaxed it to a linear optimisation problem that may be used for short-term operation of cascade hydroelectric power plants. In the

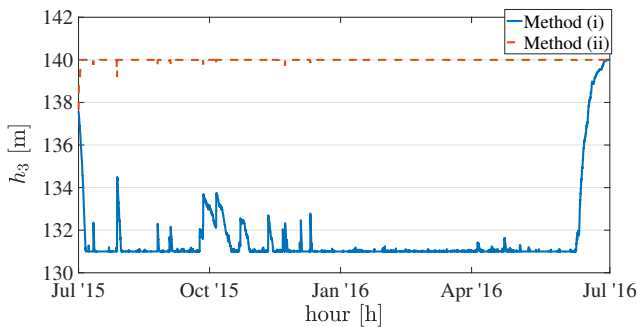


Fig. 4: Head level of Gitaru dam with the two methods.

numerical studies, we showed that the proposed approximation provides a good representation of the actual state of the hydroelectric power system and the associated benefits with using the proposed framework in terms of the live volume of water used. The excess of water increases the resiliency of the electric system in dry seasons. In the future, we will demonstrate that the coupling of hydroelectric and solar technology is beneficial due to the negative correlation of rain and sunshine. In addition, we plan on incorporating uncertainty sources into the forecasts of the load that the hydroelectric power system is required to meet, i.e., uncertainty in load variations.

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