

2 Maximizing Hydropower Generation with Observations 3 1 and Numerical Modeling of the Atmosphere

4 Yabin Miao

5 Graduate Student, Dept. of Civil and Environmental Engineering, Univ. of
6 Washington, More Hall 201, Seattle, WA 98195.

7 Xiaodong Chen, S.M.ASCE

8 Graduate Student, Dept. of Civil and Environmental Engineering, Univ. of
9 Washington, More Hall 201, Seattle, WA 98195.

10 Faisal Hossain

11 Associate Professor, Dept. of Civil and Environmental Engineering, Univ.
12 of Washington, More Hall 201, Seattle, WA 98195 (corresponding author).

13 *Forum papers are thought-provoking opinion pieces or essays*
14 *founded in fact, sometimes containing speculation, on a civil en-*
15 *gineering topic of general interest and relevance to the readership*
16 *of the journal. The views expressed in this Forum article do not*
17 *necessarily reflect the views of ASCE or the Editorial Board of*
18 *the journal.*

19 DOI: [10.1061/\(ASCE\)HE.1943-5584.0001405](https://doi.org/10.1061/(ASCE)HE.1943-5584.0001405)

20 Introduction

21 3 The ongoing drought in California seems to indicate that water
22 managers are now paying greater attention to the use of numeri-
23 cal models of the atmosphere for short-term (7–10 day) weather
24 forecasts. At the time of writing this article, a bill to Congress
25 was being formulated that essentially aims to make the rule
26 curves for large dams more adaptive through the use of numeri-
27 cal models for weather forecast. The use of such models is ex-
28 pected to reduce wastage of impounded water for dam managers
29 and allow more flexibility in water storage and release during
30 periods of anomalous/off-season big droughts or floods. The pur-
31 pose of this opinion article is to shed light on the current state of
32 hydropower generation in the United States and discuss how the
33 use of such numerical models of the atmosphere can also maxi-
34 mize energy production while conserving water or protecting
35 against floods.

36 4 As a clean and renewable energy source, hydropower has been
37 extensively exploited by human beings for over 100 years. Accord-
38 ing to the *2014 Hydropower Market Report* released by the Depart-
39 ment of Energy (DOE), there are 2,198 active hydropower plants
40 with a total operational capacity of 79.64 gigawatts (GW) in the
41 United States. Hydropower accounts approximately 7% of the total
42 power generated in the United States (Uría-Martínez et al. 2015).
43 Unfortunately, hydropower generation capacity has stagnated the
44 last two decades since the 1990s owing to lower economic growth
45 (Hall et al. 2003), stricter environmental regulations, a stagnant en-
46 ergy market, and recent breakthroughs in the shale gas and oil in-
47 dustries (such as fracking). Nevertheless, hydropower remains the
48 single largest source of renewable energy because of its relatively
49 low-cost and sustainable characteristics.

50 Compared to other renewable energy sources, hydropower has
51 several unique advantages (USBR 2005). For example, a dam,

which is normally considered an expensive investment to build, 52
has a long service life spanning at least 50–100 years. Hydropower 53
remains a more stable and durable source compared to wind or solar 54
power, which are vulnerable to changing or unpredictable weather 55
conditions. Hydropower production is relatively easier to ramp up 56
or scale back depending on the transient nature of power demand. 57
There are also no greenhouse gas emissions as byproducts during 58
hydropower generation. 59

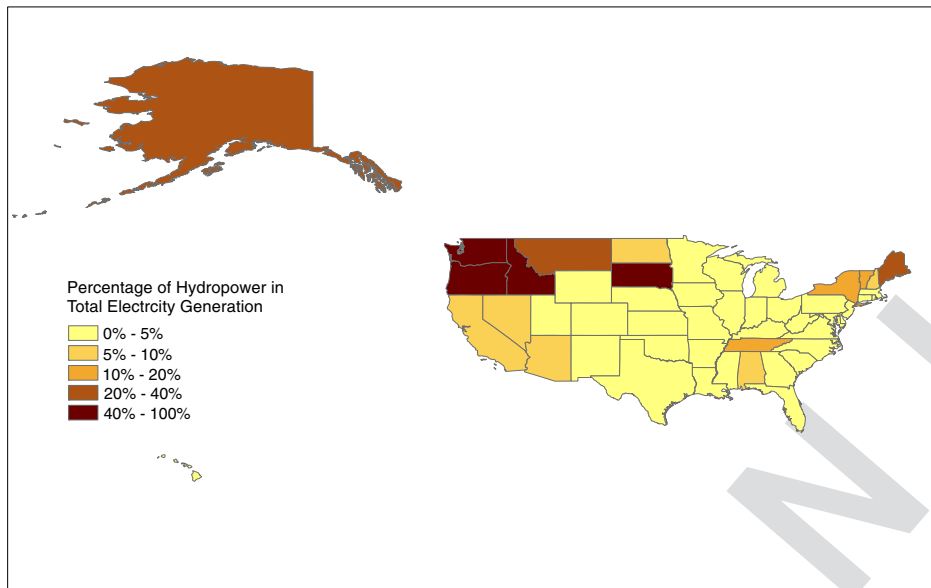
A few U.S. states that are gifted with abundant water resources 60
and topography have already harnessed hydropower as a clean and 61
reliable electricity generation source, such as Washington, Oregon, 62
and California (Fig. 1). Collectively, these three states have the 63
largest installed hydropower capacity, which is equivalent to half 64
of the total installed capacity across the United States. Among these 65
three states, Washington has the largest share with approximately 66
30.4% of total hydropower generation in the United States, which 67
also amounts to 70% of total electricity generation in the state of 68
Washington. Oregon and California contribute 13.5% and 6.3%, 69
respectively to the national grid in hydropower sector. Compared 70
to the Northwest's large amount of installed capacity, the Northeast 71
has the most hydropower facilities, which typically are aged and 72
small-capacity. 73

The states that rely more on locally available hydropower 74
apparently have lower electricity price compared to other states 75
that have limited access to hydropower generating resources (Uria- 76
Martinez et al. 2015). Because building a new power plant is 77
expensive, the low prices are usually in the states that have exten- 78
sive power facilities where owners have already paid off the capital 79
cost. The three northwestern states (Washington, Oregon, Idaho) 80
are clear examples of cheap electricity pricing due to abundant 81
hydropower resources. 82

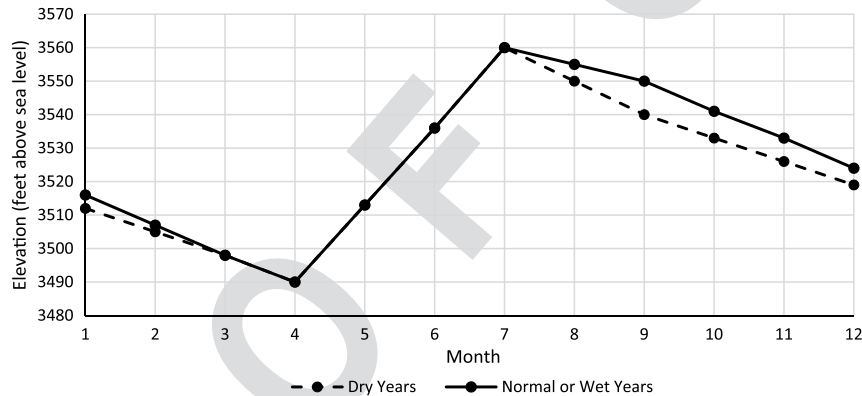
Maximizing Hydropower Generation Using Observations and Numerical Models

Every hydropower dam usually adopts a *rule curve* as an operating 85
policy based on the hydro-climatology of the river basin. The rule 86
curve represents reservoir release and storage policies (or targets) as 87
a function of season that maximizes the net benefits of the reservoir 88
system to stakeholders. For example, a hydropower dam may also 89
function as a flood-control dam requiring, as a rule curve policy, to 90
keep the pool below a maximum specified level during the onset of 91
a flood season. This way, the expected flood storage can be held 92
inside the reservoir and gradually routed downstream for flood 93
mitigation. Similarly, a hydropower dam serving the needs of water 94
supply/irrigation will need to keep the pool at the highest level 95
to maximize storage during the season of peak water demand 96
(e.g., growing or irrigation season). Such a rule curve is shown 97
in Fig. 2, for the Hungry Horse Dam in the United States. 98

Most large hydropower dams, especially if they are federally 99
operated, are required to adhere to this rule curve by minimizing 100
the deviations even though actual storage and release can be quite 101
different from the targets set in the rule curve. This is where there 102
exists, in our opinion, considerable room for maximizing hydro- 103
power generation through the use of observations and numerical 104



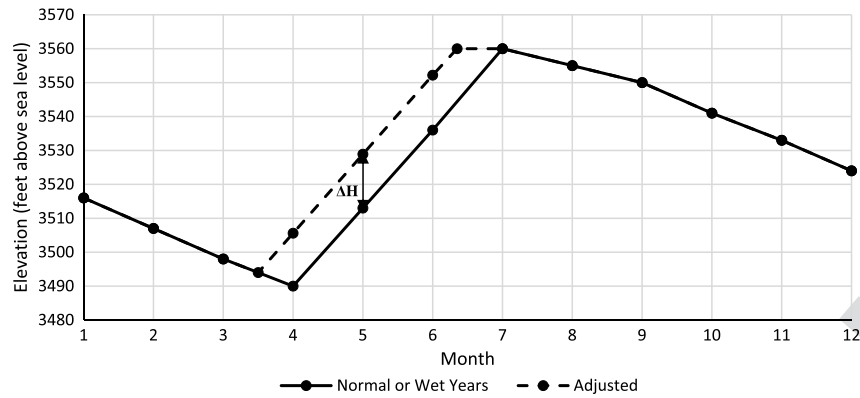
F1:1 **Fig. 1.** Percentage contribution of hydropower to total statewide electricity production in 2014 (reproduced from the U.S. Energy Information
 F1:2 **5** Administration 2015)



F2:1 **Fig. 2.** Operating rule curve of Hungry Horse Dam (USA)

105 atmospheric model without compromising other benefits or flood
 106 risks of a multipurpose hydropower dam. For example, a particular
 107 month of a flood season may be unusually less flood-intensive than
 108 what climatology may suggest, thereby making the need to lower
 109 the pool to the rule curve-recommended levels quite redundant.
 110 Vice versa, an unusually flood intensive period of a flood season
 111 may warrant the pool to be lowered more in advance to accommo-
 112 date the additional flood storage not accounted for in the rule curve
 113 storage target. Either way, the dynamics of the pertinent weather
 114 system and the season should dictate how the reservoir should
 115 be operated. Since streamflow forecast is closely related to the fore-
 116 cast of precipitation patterns, it therefore makes more sense to fore-
 117 cast the weather patterns up to 7–14 days in advance to *adjust* for a
 118 more *dynamic* rule curve or for a more transient storage/release pol-
 119 icy based on the situation. Considerably more hydropower could be
 120 generated by keeping the pool at a slightly higher level than the rule
 121 curve policy if an intense storm is not forecast within the horizon.
 122 Benefitting from advances in atmospheric science over the past
 123 40 years, numerical weather prediction models are now capable
 124 of producing short-term (approximately 7–14 days) global fore-
 125 casts at large scales with variable accuracy. Such forecast can be

126 downscaled at smaller space-time scales using numerical models of
 127 the atmosphere such as weather research and forecasting (WRF).
 128 Although the accuracy of the forecast of precipitation at scales
 129 relevant for an impounded river basin may be lower, combining
 130 the forecasts with observations of snowpack and upstream surface
 131 water (from satellite observations) as well as forecasts of other
 132 meteorological variables that have higher skill (i.e., temperature
 133 and wind) can present a practicable approach for maximizing hy-
 134 dropower operations. Models like WRF are also able to reconstruct
 135 a number of big storms in recent history (Tan 2010). Such
 136 reconstruction of storms offers physical insights into unique mech-
 137 anisms of precipitation formation that may be relevant for hydro-
 138 power operations in a priori mode during extreme events that often
 139 send a flood pulse into the reservoir.
 140 Hungry Horse Dam is a gravity dam located in Flathead County,
 141 Montana, and operated by the U.S. Bureau of Reclamation. To
 142 balance flood control and hydropower generation, a dynamic rule
 143 curve based on monthly climatology of streamflow and seasonal
 144 weather conditions (Fig. 2) is generally applied to the Hungry
 145 Horse Dam operations. Monthly water-supply forecasts indicate
 146 that inflows during the period from April 1 to August 31 are



F3:1 **Fig. 3.** Adjustment on operating rule curve of Hungry Horse Dam; the delta H is about 16 ft as the difference between model adjusted rule curve
 F3:2 (dashed) and the static rule curve (solid)

147 highest. However, a more dynamic rule curve based on a combi-
 148 nation of numerical model forecast of the atmosphere and obser-
 149 vations of snowpack and upstream surface water has capability
 150 of providing more hydropower potential without reducing the flood
 151 risk. For example, Fig. 3 shows the wet year rule curve and adjusted
 152 rule curve for use with numerical model-based forecasting of the
 153 atmosphere. With assistance from atmospheric modeling, observa-
 154 tions on snowpack and temperature and hydrological modeling
 155 (to model the anticipated inflow into a reservoir), it may be possible
 156 to extend the scheduled date of reaching minimum pool by 14 days.
 157 This may translate to an almost 14 more days of hydropower gener-
 158 ation at highest potential. With a model and observational-data
 159 driven change in the rule curve, the water elevation can be poten-
 160 tially increased by up to 16 ft during the flood season for Hungry
 161 Horse Dam and maintained 14 more days during the flood season if
 162 no major flood inflow is expected from the model-data forecast sys-
 163 tem. The additional head afforded by a 16 ft results in an increase in
 164 theoretical hydropower by about 3.7% during the flood season for
 165 the Hungry Horse Dam. Fig. 4 provides schematic of how an
 166 atmospheric model based dynamic rule curve operation could work
 167 for a dam to maximize hydropower generation.

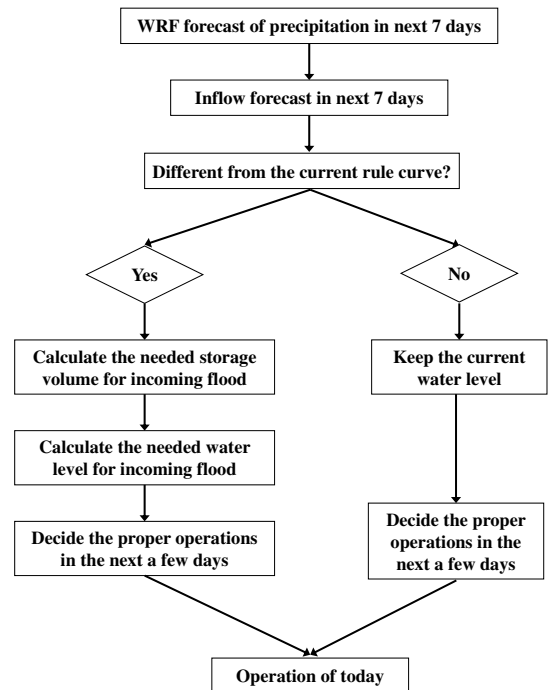
168 To compromise between flood control and hydropower gener-
 169 ation, the traditional and static rule curve sacrifices water genera-
 170 tion potential at the expense of a false sense of flood security during
 171 periods that are anomalously dry or free of extreme events. Based
 172 on the current rule curve, maximizing hydropower generation with
 173 numerical atmospheric model could be considered as an innovation
 174 that takes advantage of the progress made in atmospheric science.
 175 An even bolder strategy is to completely get rid of the rule curve
 176 and depend exclusively on short-term forecast (approximately
 177 7–14 days) of precipitation and streamflow. However, such a strat-
 178 egy may not work for large dams that impound several years' of
 179 annual runoff and are critical to a region's water security (such as
 180 Aswan High Dam on the Nile River in Egypt).

181 Future of Hydropower Dams

182 In the last decade, there has not been construction of any large-
 183 scale (>500 MW) hydropower dam project in the United States.
 184 Insufficient consideration of ecological impact has led to an enor-
 185 mous threat to salmon's existence in Washington State. Many
 186 hydropower-rich states (such as Washington) now need to take ac-
 187 tion to help local tribes claim their fishing rights (Brown et al.
 188 2013). These stricter environmental and ecological regulations
 189 and licensing procedures have resulted in a current climate where

construction of new hydropower dams is unlikely in the foreseeable
 future. In such a situation, it makes more sense to optimize current
 hydropower dam performance so that power generation can be
 maximized through the use of numerical atmospheric models of
 weather forecasting.

There is an added benefit to maximizing hydropower generation
 in states with higher electricity pricing and limited hydropower
 potential. Georgia is a typical example that experiences high retail
 price of electricity and is heavily dependent on the fossil fuel. In
 2014, 68.83% of electricity in Georgia was generated from fossil
 fuel thermal power plants and only 2.44% from conventional hy-
 dropower. The huge contrast in price and availability between fossil
 fuel and hydropower makes a good case for increasing the hydro-
 power generation efficiency in the state of Georgia. Assuming the
 current hydropower share in Georgia could be augmented by 10%



F4:1 **Fig. 4.** Dam operation decision flowchart through numerical atmo-
 F4:2 spheric model forecast; the WRF forecast refers to a combination
 F4:3 of modeling, data, and observations of snowpack, surface water, and
 F4:4 meteorological variables

205 through optimization with numerical models of the atmosphere, ap-
206 proximately 167,000 short tons of coal, 2,923 billion ft³ of natural
207 gas, and 4,700 barrels of petroleum could be reduced each year. In
208 the meantime, 522,000 short tons of CO₂ could be eliminated each
209 year. This assessment is based on the information available at the
210 Energy Information Administration.

211 Another trend now seen is environmental low-impact and
212 stream-reach hydropower developments (Kao et al. 2014). For all
213 new or purposed hydropower projects in the United States, 58% of
214 capacity (or 233 projects) are new stream-reach (non-dam) projects.
215 These are mostly concentrated in the Northwest, especially in
216 Alaska, accounting for 66% of total projects (Kao et al. 2014).

217 According to another U.S. DOE report, *New Stream-Reach*
218 *Development: A Comprehensive Assessment of Hydropower*
219 *Energy Potential in the United States* (Kao et al. 2014), there
220 are more than 50,000 nonpowered dams that are able to provide
221 another 12 GW of hydropower capacity. Additionally, rivers and
222 streams in United States are able to provide 85 GW of technical
223 hydropower potential capacity. Even excluding the lands protected
224 by federal laws, over 65 GW of potential hydropower capacity that
225 is renewable and carbon-reducing remains untapped today.

226 Conclusion

227 The current hydropower situation continues to play an indispen-
228 sable role in the whole energy generation system of the United
229 States. Although current hydropower is in a low-growth stage
230 and very large-scale dam projects are a thing of the past, more mod-
231 erate-scale and eco-friendly dams are now being given increasing
232 consideration. Additionally, due to advantages of pumped storage
233 hydropower plants, electricity power companies have a preference
234 of installing this type of power plant, which could be a trend in the

future development of the hydropower industry. With such tremen-
dous hydropower potential, both the hydropower industry and
the atmospheric science/engineering community should collaborate
effectively to design a sustainable blueprint for optimizing hydro-
power generation, using as a main tool numerical models of the
atmosphere to forecast weather patterns (ASCE 2015).

References

- ASCE. (2015). "Engineering for every drop." *Civ. Eng. Mag.*, 46–55.
- Brown, J. J., et al. (2013). "Fish and hydropower on the US Atlantic coast:
Failed fisheries policies from half-way technologies." *Conserv. Lett.*,
6(4), 280–286.
- Hall, D., Hunt, R. T., Reeves, K. S., and Carrol, G. R. (2003). "Estimation
of economic parameters of US hydropower resources, Prepared for
the U.S. Department of Energy Office of Energy Efficiency and Renew-
able Energy and Energy Information Administration Office of Inte-
grated Analysis and Forecasting Under DOE Idaho Operations
Office Contract DE-AC07-99ID13727." (<http://www1.eere.energy.gov/wind/pdfs/doewater-00662.pdf>) (Jan. 16, 2016).
- Kao, S., et al. (2014). "New stream-reach development: A comprehensive
assessment of hydropower energy potential in the United States."
- Mapes, L. V. (2013). "Tribes' salmon court win may go way beyond \$1B in
culvert repairs." *The Seattle Times*.
- Tan, E. (2010). "Development of a methodology for probable maximum
precipitation estimation over the American River watershed using
the WRF model." Ph.D. dissertation, Univ. of California, Davis, CA.
- Uría-Martínez, R., O'Connor, P. W., and Johnson, M. M. (2015). "2014
hydropower market report." (<http://energy.gov/hydropower-market-report>) (Jan. 16, 2016).
- USBR (U.S. Bureau of Reclamation). (2005). "Managing water in the
west hydroelectricity power." (<https://www.usbr.gov/power/edu/pamphlet.pdf>) (Jan. 16, 2016).