

Readme: Implementation progress and
what's next

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Margaret Chapman

See also: Submitted-Sustech2018.pdf

- For a 1-dimensional LTI system (see Setup-LTI-Dynamics.m),

We have computed $\mathcal{U}_y^r := \left\{ x \mid \min_{\pi} \text{CVaR}_y \left[\sum_{k=0}^N e^{m \cdot g(x_k)} \mid \pi, x_0 = x \right] < e^{mr} \right\}^{(1)}$ via dynamic programming, and

$$\mathcal{S}_y^r := \left\{ x \mid \min_{\pi} \text{CVaR}_y \left[\max_{k=0, \dots, N} g(x_k) \mid \pi, x_0 = x \right] < r \right\} \text{ via brute force enumeration,}$$

for several $y \in (0, 1)$ and r .

(See Results-LTISystem\compare-DPsoftmax-mis10-vs-BFmax.fig; The figure was generated using the 3rd cell of Script_Compare.m. The previous 2 cells were used to determine grid spacing and the softmax parameter, m .)

- Recall that $\mathcal{U}_y^r \subseteq \mathcal{S}_y^r$, and our goal is to find a good approximation to \mathcal{S}_y^r , and we do this by computing \mathcal{U}_y^r via dynamic programming.

- Next, we want to extend the code to a more meaningful system, and compute \mathcal{U}_y^r and \mathcal{S}_y^r for this system.

System - pond 1 of our Sustech submission (see figure 2 of Submitted-Sustech2018.pdf) equation 1a, equation 2

$$\dot{x} = \frac{\omega - q_p(x, u)}{A} \quad (1a)$$

x = water elevation in pond [ft] - state

u = valve setting $\in \{0, 1\}$ - control input

ω = surface runoff due to rain $\left[\frac{\text{ft}^3}{\text{s}} \right]$ - disturbance

A = surface area of pond [ft²]

$$q_p(x, u) = \begin{cases} C_d \pi R^2 u \sqrt{2g(x-z)} & \text{if } x \geq z \\ 0 & \text{if } x < z \end{cases} \quad \text{outflow through valve} \quad (2)$$

$\left[\frac{\text{ft}^3}{\text{s}} \right]$

- The constants are provided in Table I (e.g., $A = 28,292 \text{ ft}^2$)
(of the SuStech submission)
- The constraint set is $X = (0\text{ft}, 5\text{ft})$. In other words, if $x > 5 \text{ ft}$, then the pond has flooded.
- Water enters the pond due to surface runoff (w), and leaves the pond thru the valve. The valve has 2 settings (open, $u=1$) or (closed, $u=0$).
- The 1st step is to discretize the dynamics to get,

$$x_{k+1} = f(x_k, u_k, w_k), \text{ where } \Delta t = 5 \text{ min (perhaps).}$$

$$k = 0, 1, \dots, N-1$$

x_k = water level at time k , units of ft

u_k = valve setting at time k (0 or 1, no units)

w_k = amount of water entering the pond during $[k, k+1)$, units of ft^3

- We will assume that the probability of w_k taking on a particular value is known at each time point. This will be estimated from a design storm (which is a synthetic storm ^{after} based on historical data that is used to design storm water systems).

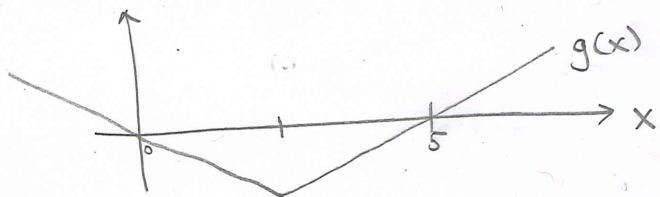
- While I'm not sure what these probabilities are yet, we should have $\mathbb{E}[w_k] \approx 45000 \text{ ft}^3$
if $\Delta t = 5 \text{ min}$, and the time duration from $k=0$ to $k=N$ is 4 hours.

- For now, let's make up a ^(time-invariant) distribution $P\{w_k = j\} = p_j$, where the expected value of w_k is about 45000 ft^3 , and has about 10 samples, so that $\sum_{j=1}^{10} p_j = 1$. We can use this distribution as a placeholder until we get a more accurate one.

- We want the distribution to have a large number of samples, if possible.

- Because the constraint set, $K = (0, 5 \text{ ft})$, the signed distance function will be

$$g(x) = |x - 2.5| - 2.5.$$



- One needs to define the discretized state space (x_s), and the soft-max parameter (m).

- I first started with $m=1$, then played around with grid size (x_s, l_s). I chose a grid size so that the computation of U_y^r via dynamic programming and the computation of U_y^r via a brute force enumeration were relatively close.
 (see top line page 1)
 perhaps use Monte Carlo instead

- After selecting the grid, then I played around with m . Theoretically, bigger m is better, but computationally big m may be problematic.

- Overall, we want to get a figure (like Results-LTI System / compare-DPsoftmax-mis10-vs-BFmax.fig), showing U_y^r and S_y^r for several y and several r , for the pond system.
 We have the results already for the LTI system.
 (including $r=0$)

- Because computation of S_y^r via brute force enumeration may not be possible, instead we'll use a Monte Carlo approach (perhaps).
- We'll use our Dynamic programming algorithm to get U_y^r for the pond system.
- In particular, we should show U_y^r, S_y^r for $r=0$.

- We would like to generate U_y^r, S_y^r for different instances of our pond system to demonstrate how risk-sensitive reachability can be used for design of infrastructure in the presense of uncertainty.

Instance 1 : $R = 1/3 \text{ ft}$, $u \in \{1\}$
 (R, passive) pond outlet radius valve always open

Instance 2 : $R = 1/3 \text{ ft}$, $u \in \{0, 1\}$ "active control"
 (R, active) valve can be either open or closed based on what's optimal

← Let's start with this instance.

Instance 3 : $R = 2/3 \text{ ft}$, $u \in \{1\}$
 (2R, passive) larger pond outlet radius

- The key challenges that I faced while coding up the LTI example:

- ① choosing $\min(x_s)$, $\max(x_s)$ and restrictions on the control input near the boundary of the discretized state space (x_s)
(to prevent the DP algorithm from interpolating outside the grid)
- ② making sure that the brute force enumeration code (Main-BruteForce.m) generated the same scenario tree as the DP algorithm (Main-DynProg.m).

- To understand the big picture of the code, please look at

- Main-DynProgram.m

- Main-BruteForce.m

- Script-Compare.m,

and the headers of every .m file in Matlab-Code.

- Please branch off the master node.