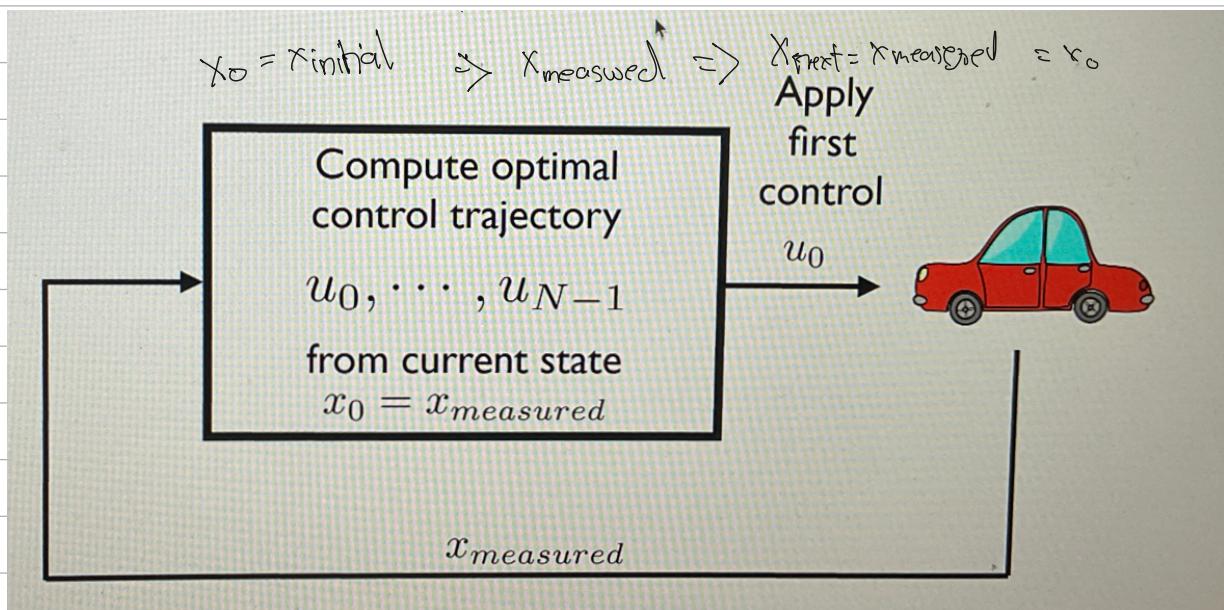


Model Predictive Control / Receding Horizon Control



Gradient estimation via Gaussian Smoothing

Suppose we have a function $f(x)$, the gaussian smoothed form of $f(x)$ is given as;

$$f_\sigma(x) = \mathbb{E}_{u \sim N(0, \sigma^2)} [f(x+u)] \quad \text{where } \begin{aligned} u &\rightarrow \text{noise vector} \\ 0 &\rightarrow \text{mean} \\ \sigma^2 &\rightarrow \text{variance} \end{aligned}$$

i.e. $f_\sigma(x)$ is the expected value of $f(x+u)$ where u is drawn from the gaussian distribution with mean 0 and σ^2 variance.

In integral form;

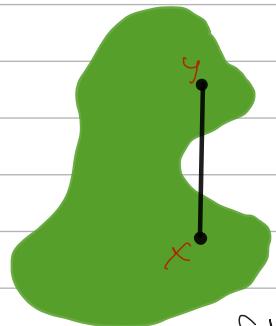
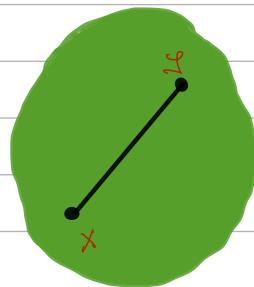
$$f_\sigma(x) = \int f(x+u) \frac{e^{-\frac{u^T u}{2\sigma^2}}}{\sqrt{(2\pi)^n}} du$$

Generally $f(x) \neq f_\sigma(x)$

But if $f(x)$ is a convex function then Jensen's Inequality condition is valid; which states:

$$f(x) \leq f_{\sigma}(x)$$

The inequality indicates that the smoothed function has an upper bound to the original function if its convex.



convex function

non convex function.

Difference Boundaries :

The difference b/w $f(x)$ & $f_{\sigma}(x)$ is bounded by certain factors that determine how quickly the function changes:

Condition 1: $|f(x) - f_{\sigma}(x)| \leq L_1 \|x - y\|$ then;

$$|f_{\sigma}(x) - f(x)| \leq \sigma \cdot L_1 \sqrt{n}$$

Condition 2: $|\nabla f(x) - \nabla f_{\sigma}(x)| \leq L_2 \|x - y\|$ then;

$L_1, \sigma, L_2 \rightarrow$ constants
 because rate of
 change of $f(x)$
 σ - smoothing control.
 $n \rightarrow$ dimension(x)

$$|f_{\sigma}(x) - f(x)| \leq \frac{\sigma^2}{2} \cdot L_1 \cdot n$$

Policies have ∞ no. of variables as shown below

Policies \rightarrow computing optimal control for every initial state

let us have a state x ;

$$x = \{0, 1, 2\} \quad \text{S. } x_{n+1} = x_n + u_n$$

where u can be any thing $\therefore x_{n+1} \in \{0, 1, 2\}$
So;

$$\begin{aligned} \text{At } x_0 = 0 \\ x_{n+1} &\Rightarrow 0 = 0 + u_0 = 0 \\ &\Rightarrow 1 = 0 + u_0 = 1 \\ &\Rightarrow 2 = 0 + u_0 = 2 \end{aligned}$$

$$\text{ie; } U_0 = \{0, 1, 2\}.$$

$$\begin{aligned} x_0 = 1 \\ x_{n+1} &\Rightarrow 0 = 1 + u_0 \Rightarrow u_0 = -1 \\ 1 &= 1 + u_0 \Rightarrow u_0 = 0 \\ 2 &= 1 + u_0 \Rightarrow u_0 = 1 \end{aligned}$$

ie;

$$\begin{aligned} x_0 = 2 \\ x_{n+1} &\Rightarrow 0 = 2 + u_0 \Rightarrow u_0 = -2 \\ 1 &= 2 + u_0 \Rightarrow u_0 = -1 \\ 2 &= 2 + u_0 \Rightarrow u_0 = 0 \end{aligned}$$

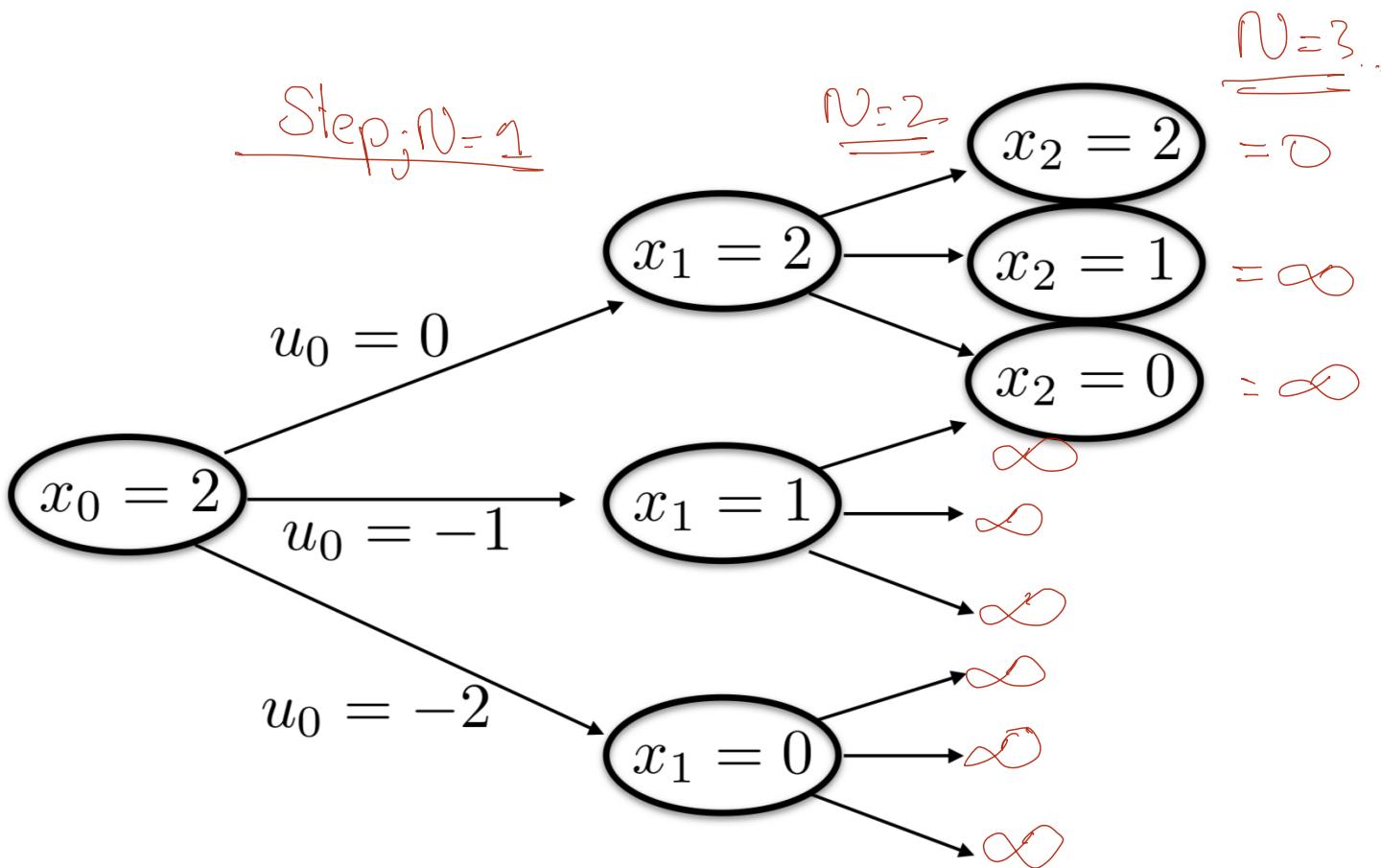
$$\text{ie; } U_0 = \{-2, -1, 0\}$$

Given function to minimize

$$\min_x \sum_{n=0} (|u_n| + x_n^2) + g_2(x_2)$$

$$\therefore g_2 = \begin{cases} 0 & \text{if } x_2 = 0 \\ \infty & \text{otherwise} \end{cases}$$

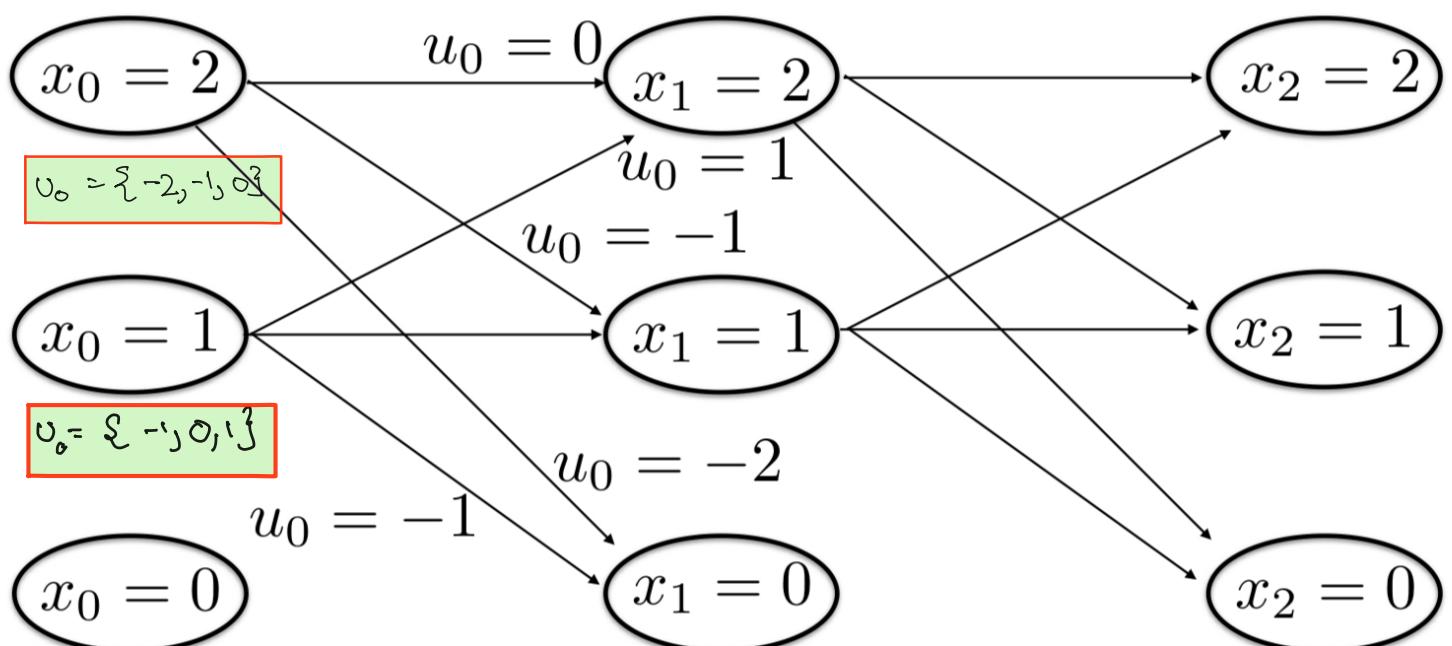
not $x = 2$, but $x_2 = 0$



What will the outcomes be of them? \uparrow

Therefore the no. of variable span infinitely at the rate of 3^{N^t} (N:step).

Additionally there are also inter dependencies



Stage 0 $\rightarrow f(x_0, u_0) \rightarrow$ Stage 1 $\rightarrow f(x_1, u_1) \rightarrow$ Stage 2 \rightarrow

So in conclusion the final stage is obtained by $f(x_{N-1}, u_{N-1})$ where $N \in \mathbb{N}$.

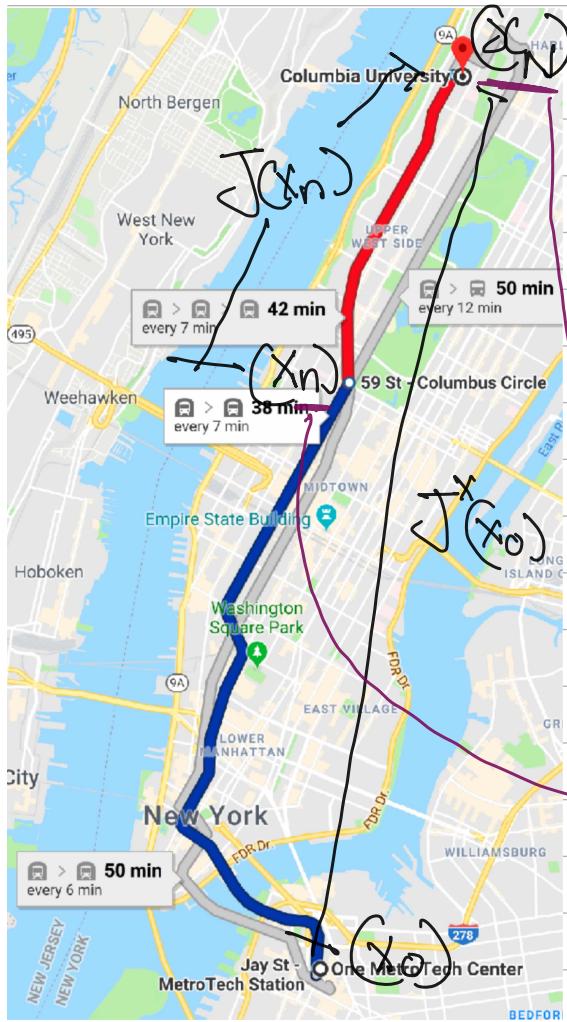
This leaves us with infinite horizon problem and we have to find a way to navigate it.

Control Law: A function denoted as $\mu_k(x_k)$ which tells us what to do at stage k. It maps every state in x_k to a control action u_k .

Control Policy: Sequence of superset of all control laws j_i :

$$\Pi = \{\mu_0(x_0), \mu_1(x_1), \mu_2(x_2), \dots, \mu_N(x_N)\}$$

Cost to go: Consider the following figure.



The path plotted is the optimal route from Jay street to Columbia Uni. This is also called optimal cost / $J^*(x_0)$.

Now any point taken along that optimal route is the most optimal route to go to Columbia Uni. This is also called optimal cost logo. Denoted as $J_n(x_n)$. It is defined as:

$$J_n(x_n) = \min_{\alpha} \sum_{i=n}^{N-1} q_i(x_i, u_i) + q_N(x_N)$$

Mnemonics:

u_k^* → optimal control law at stage k

Π^* → optimal control policy for the problem

Bellman's Principle Of Optimality :-

Let $\{u_0^*, u_1^*, \dots, u_{N-1}^*\}$ be an optimal control sequence for the original optimal control problem and assume that when we are using the sequence of control, a given state x_k occurs at time k . Consider the subproblem where we start at x_k at time k and wish to minimize the *cost-to-go* from time k to time N

$$\sum_{i=k}^{N-1} g_i(x_i, u_i) + g_N(x_N)$$

Then the truncated control sequence $\{u_k^*, \dots, u_{N-1}^*\}$ is optimal for this subproblem.

One way to tackle the infinite variable problems as we have seen above \longrightarrow **DYNAMIC PROGRAMMING**

In dynamic programming : We start from end (stage N) of the problem and compute optimal policies of the subproblems.

Dynamic Programming :-

For every initial state x_0 , the optimal cost $J^*(x_0)$ of the optimal control problem is equal to $J_0(x_0)$ (i.e. the optimal cost to go from x_0) which is computed backward in time from stage $N - 1$ to stage 0 using the following recursion:

$$J_N(x_N) = g_N(x_N)$$
$$J_k(x_k) = \min_{u_k} g_k(x_k, u_k) + J_{k+1}(f(x_k, u_k))$$

Furthermore, if the control laws $u_k^* = \mu_k^*(x_k)$ minimize the cost-to-go for each x_k and k , the policy $\pi^* = \{\mu_0^*, \dots, \mu_{N-1}^*\}$ is optimal.

$\pi^* \rightarrow$ set of optimal control laws $\rightarrow \{u_0^*, u_1^*, \dots, u_{N-1}^*\}$

Note : $N-1$ not N

Deterministic Case

Fill the table backwards with DP:

~~Not Filled~~ | Filled in green

Stage 0			Stage 1			Stage 2	
x_0	J_0^*	U_0^*	x_1	J_1^*	U_1^*	x_2	J_2^*
0	2	0	0	2	2	0	∞
1	3	0	1	2	1	1	∞
2	4	-2	2	4	0	2	0

Goal: $\min \sum_{n=0}^N (|U_n| + x_n^2) + g_n x_n$

where $g_n = 0$ at $x_2 = 0$ (∞ otherwise)

Constraints: $x_{n+1} = x_n + U_n$

- For each x_n , we
- enumerate possible U_n based on x_n
 - calculate resulting x_2
 - evaluate cost defined as (for each control):

$$\text{cost} = (|U_n| + x_n^2) + J_{n+1}^*(x_{n+1})$$

- choose the control that minimizes the cost and fill the table accordingly.

Solution:

$$\begin{aligned} J_2^*(x_2=0) &= \infty \\ J_2^*(x_2=1) &= \infty \\ J_2^*(x_2=2) &= 0 \end{aligned}$$

By Function Definition

$$\begin{aligned} \text{At } x_1=0; \quad U \in \{0, 1, 2\} \Rightarrow & |0| + 0^2 + \overline{J_2(0+0)} = \infty \\ & |1| + 0^2 + \overline{J_2(1+0)} = \infty \\ & |2| + 0^2 + \overline{J_2(2+0)} = 2 \end{aligned}$$

Hence at $x_1=0$; $J_1^*(0) = \underline{2}$ & $U_1^* = \underline{2}$

$$\begin{aligned} \text{At } x_1=1, \quad v \in \{-1, 0, 1\} \Rightarrow |1| + 1^2 + \overbrace{J_1(-1+1)}^{\infty} &= \infty \\ 0 + 1^2 + J_1(0+1) &= \infty \\ 1 + 1^2 + \overbrace{J_1(1+1)}^0 &= 2 \end{aligned}$$

Hence at $x_1=1$; $J_1^*(1) = \underline{2}$ & $U_1^* = \underline{1}$

$$\begin{aligned} \text{At } x_1=2, \quad v \in \{-2, -1, 0\} \Rightarrow |2| + (2)^2 + \overbrace{J_1(-2+2)}^{\infty} &= \infty \\ |1| + 2^2 + J_1(-1+2) &= \infty \\ 0 + 2^2 + J_1(0+2) &= 4 \end{aligned}$$

Hence at $x_1=2$; $J_1^*(2) = \underline{4}$ & $U_1^* = \underline{0}$

$$\begin{aligned} \text{At } x_0=0; \quad v \in \{0, 1, 2\} \Rightarrow |0| + 0^2 + J_0(0+0) &= \underline{2} \\ \Rightarrow |1| + 0^2 + J_0(0+1) &= 3 \\ |2| + 0^2 + J_0(0+2) &= 6 \end{aligned}$$

Hence at $x_0=0$; $J_0^*(0) = \underline{2}$ & $U_0^* = \underline{0}$

$$\begin{aligned} \text{At } x_0=1; \quad v \in \{-1, 0, 1\} \Rightarrow |1| + 1^2 + J_0(-1+1) &= 4 \\ |0| + 1^2 + J_0(0+1) &= \underline{3} \\ |1| + 1^2 + J_0(1+1) &= 6 \end{aligned}$$

Hence at $x_0=1$; $J_0^*(1) = \underline{3}$ & $U_0^* = \underline{0}$

$$\begin{aligned} \text{At } x_0=2; \quad v \in \{-2, -1, 0\} \Rightarrow |2| + 2^2 + J_0(0+0) &= \underline{4} \\ |1| + 2^2 + J_0(1+0) &= 5 \\ |0| + 2^2 + J_0(2+0) &= 8 \end{aligned}$$

Hence at $x_0=2$; $J_0^*(2) = \underline{4}$ & $U_0^* = \underline{-2}$

(Q2)

Consider the following dynamical system

$$x_{n+1} = \begin{cases} -x_n + 1 + u_n & \text{if } -2 \leq -x_n + 1 + u_n \leq 2 \\ 2 & \text{if } -x_n + 1 + u_n > 2 \\ -2 & \text{else} \end{cases}$$

where $x_n \in \{-2, -1, 0, 1, 2\}$ and $u_n \in \{-1, 0, 1\}$, and the cost function

$$J = \left(\sum_{k=0}^2 2|x_k| + |u_k| \right) + x_3^2 \quad (1)$$

Use the dynamic programming algorithm to solve the finite horizon optimal control problem that minimizes J . Show the different steps of the algorithms and present the results in a table including the cost to go and the optimal control at every stage.

Stage 0		Stage 1		Stage 2		Stage 3				
x_0	J_0^*	u_0^*	x_1	J_1^*	u_1^*	x_2	J_2^*	u_2^*	x_3	J_3^*
-2	10	0	-2	9	0	-2	8	0	-2	4
-1	6	-1	-1	5	-1	-1	4	-1	-1	1
0	3	0/1	0	2	0/1	0	1	-1	0	0
1	4	0	1	3	0	1	2	0	1	1
2	7	1	2	6	1	2	5	0/1	2	4

Given: $\left(\sum_{k=0}^2 2|x_k| + |u_k| \right) + x_3^2 \quad u \in \{-1, 0, 1\} \quad x \in \{-2, -1, 0, 1, 2\}$

Now let's start with $J_3^*(x_3) = x_3^2 \Leftrightarrow J_3^* = \{4, 1, 0, 1, 4\}$

$$\text{Cost} = 2|x_k| + |u_k| + J_{K+1}^*(x_{K+1})$$

At $x_2 = -2 \quad u \in \{-1, 0, 1\}$

$$\text{cost}(x_2 = -2) = 2|x_2| + |u_2| + J_3^*(x_3)$$

But what is x_3 ? -2

$$\begin{array}{ll}
 \text{At } v_2 = -1 & x_3 = -(-2) + (-1) + 1' = 2 \\
 v_2 = 0 & x_3 = -(-2) + 1 + 0 = 3 \Rightarrow 2 \\
 v_2 = 1 & x_3 = -(-2) + 1 + 1 = 4 \Rightarrow 2
 \end{array}
 \quad \text{Hence } x_3 = 2 \text{ at } (v_2 = -1) \\
 \quad x_3 = 2 \text{ at } (v_2 = 0) \\
 \quad x_3 = 2 \text{ at } (v_2 = 1)$$

$$\begin{aligned}
 \text{Now, cost} &= 2|(-2)| + |(-1)| + J_3(2) = 4+1+4 = 9 \\
 &= 2 \times 1 - 2 + 1 + J_3(2) = 4+0+4 = 8 \\
 &= 2 \times |(-2)| + |1| + J_3(2) = 4+1+4 = 9
 \end{aligned}$$

$$\text{Hence at } x_2 = -2 \quad v_2^* = 0 \quad \& \quad J_2^* = 8$$

$$\text{At } x_2 = -1 \quad v \in \{-1, 0, 1\}$$

$$\begin{array}{ll}
 \text{At } v_2 = -1 & x_3 = 1 + 1 - 1 = 1 \\
 v_2 = 0 & x_3 = 1 + 1 + 0 = 2 \\
 v_2 = 1 & x_3 = 1 - 1 + 1 = 1 \Rightarrow 2
 \end{array}
 \quad \text{Hence } x_3 = 1 \text{ at } (v_2 = -1) \\
 \quad x_3 = 2 \text{ at } (v_2 = 0) \\
 \quad x_3 = 2 \text{ at } (v_2 = 1)$$

$$\begin{aligned}
 \text{Now, cost} &= 2|(-1)| + |(-1)| + J_3(1) = 2+1+1 = 4 \\
 &= 2|(-1)| + |(0)| + J_3(2) = 2+0+4 = 6 \\
 &= 2|(-1)| + |1| + J_3(3) = 2+1+4 = 7
 \end{aligned}$$

$$\text{Hence at } x_2 = -1 \quad v_2^* = -1 \quad \& \quad J_2^* = 4$$

$$\text{At } x_2 = 0 \quad v \in \{-1, 0, 1\}$$

$$\begin{array}{ll}
 \text{At } v_2 = -1 & x_3 = 0 + 1 - 1 = 0 \\
 v_2 = 0 & x_3 = 0 + 1 + 0 = 1 \\
 v_2 = 1 & x_3 = 0 + 1 - 1 = 0
 \end{array}
 \quad \text{Hence } x_3 = 0 \text{ at } v_2 = -1 \\
 \quad x_3 = 1 \text{ at } v_2 = 0 \\
 \quad x_3 = 0 \text{ at } v_2 = 1$$

$$\begin{aligned}
 \text{Now, cost} &= 2 \times 0 + |(-1)| + J_3(0) = 1 \\
 &= 2 \times 0 + (0) + J_3(1) = 1 \\
 &= 2 \times 0 + (1) + J_3(2) = 5
 \end{aligned}$$

$$\text{Hence at } x_2 = 0 \quad v_2^* = -1 \quad \& \quad J_2^* = 1$$

At $x_2 = 1$ $v \in \{-1, 0, 1\}$

$$\begin{array}{lll} \text{At } v_2 = -1 & x_3 = -1 + 1 - 1 = -1 & \text{Hence } x_3 = -1 \text{ at } v_2 = -1 \\ \text{At } v_2 = 0 & x_3 = -1 + 1 + 0 = 0 & x_3 = 0 \text{ at } v_2 = 0 \\ \text{At } v_2 = 1 & x_3 = -1 + 1 + 1 = 1 & x_3 = 1 \text{ at } v_2 = 1 \end{array}$$

$$\begin{aligned} \text{Now, cost} &= 2|1| + |-1| + J_3(-1) = 2+1+1 = 4 \\ &= 2|1| + \cancel{0} + J_3(0) = 2+0+0 = \cancel{2} \\ &= 2|1| + |1| + J_3(1) = 2+1+1 = 4 \end{aligned}$$

Hence at $x_2 = 1$ $v_2^* = 0$ & $J_2^* = 2$ ~~2~~

At $x_2 = 2$ $v \in \{-1, 0, 1\}$

$$\begin{array}{lll} \text{At } v_2 = -1 & x_3 = -(2) + \cancel{1} - 1 = -2 & \text{Hence } x_3 = -2 \text{ at } v_2 = -1 \\ \text{At } v_2 = 0 & x_3 = -(2) + 1 + 0 = -1 & x_3 = -1 \text{ at } v_2 = 0 \\ \text{At } v_2 = 1 & x_3 = \cancel{-(2)} + \cancel{1} + 1 = 0 & x_3 = 0 \text{ at } v_2 = 1 \end{array}$$

$$\begin{aligned} \text{Now, cost} &= 2|2| + |-1| + J(-2) = 4+1+4 = 9 \\ &= 2|2| + \cancel{0} + J(-1) = 4+0+1 = \cancel{5} \\ &= 2|2| + |1| + J(0) = 4+1+0 = 5 \end{aligned}$$

Hence at $x_2 = 2$; $v_2^* = 0/1$; $J_2^* = 5$ ~~5~~

At $x_1 = -2$ $v \in \{-1, 0, 1\}$

$$\begin{array}{lll} \text{At } v_1 = -1 & x_2 = -(-2) + \cancel{x} - 1 = 2 & \text{Hence } x_2 = 2 \text{ at } v_1 = -1 \\ \text{At } v_1 = 0 & x_2 = -(-2) + 1 + 0 = 3 \Rightarrow 2 & x_2 = \text{ at } v_1 = 0 \\ \text{At } v_1 = 1 & x_2 = -(-2) + 1 + 1 = 4 \Rightarrow 2 & x_2 = \text{ at } v_1 = 1 \end{array}$$

$$\begin{aligned} \text{Now, cost} &= 2|-2| + |-1| + J(2) = 4+1+5 = 10 \\ &= 2|-2| + \cancel{|0|} + J(2) = 4+0+5 = \cancel{9} \\ &= 2|-2| + |1| + J(2) = 4+1+5 = 10 \end{aligned}$$

Hence at $x_1 = -2$; $v_1^* = 0$; $J_1^* = 9$ ~~9~~

8
9
2
5

At $x_1 = -1$ $U \in \{-1, 0, 1\}$

At $U_1 = -1$ $x_2 = -(-1) + 1 - 1 = 1$ Hence $x_2 = 1$ at $U_1 = -1$
 $U_1 = 0$ $x_2 = -(-1) + 1 + 0 = 2$ $x_2 = 2$ at $U_1 = 0$
 $U_1 = 1$ $x_2 = -(-1) + 1 + 1 = 3 \Rightarrow 2$ $x_2 = 2$ at $U_1 = 1$

Now, cost $= 2^{2^{-1}} | 1 + 1 | + J^2(1) = 2+1+2 = 5$
 $= 2^{2^{-1}} | 1 + 0 | + J^2(2) = 2+0+5 = 7$
 $= 2^{2^{-1}} | 1 + 1 | + J^2(2) = 2+1+5 = 8$

Hence at $x_1 = -1$; $U_1^* = -1$; $J_1^* = 5$

At $x_1 = 0$ $U \in \{-1, 0, 1\}$

At $U_1 = -1$ $x_2 = -(0) + 1 - 1 = 0$ Hence $x_2 = 0$ at $U_1 = -1$
 $U_1 = 0$ $x_2 = -(0) + 1 + 0 = 1$ $x_2 = 1$ at $U_1 = 0$
 $U_1 = 1$ $x_2 = -(0) + 1 + 1 = 2$ $x_2 = 2$ at $U_1 = 1$

Now, cost $= 2^{1^0} | 1 + 1 | + J^1(0) = 0+1+1 = 2$
 $= 2^{1^0} | 1 + 0 | + J^1(1) = 0+2 = 2$
 $= 2^{1^0} | 1 + 1 | + J^1(2) = 0+1+5 = 6$

Hence at $x_1 = 0$; $U_1^* = 1/0$; $J_1^* = 2$

At $x_1 = 1$ $U \in \{-1, 0, 1\}$

At $U_1 = -1$ $x_2 = -(1) + 1 - 1 = -1$ Hence $x_2 = -1$ at $U_1 = -1$
 $U_1 = 0$ $x_2 = -(1) + 1 + 0 = 0$ $x_2 = 0$ at $U_1 = 0$
 $U_1 = 1$ $x_2 = -(1) + 1 + 1 = 1$ $x_2 = 1$ at $U_1 = 1$

Now, cost $= 2^{1^1} | 1 + 1 - 1 | + J^1(-1) = 2+1+4 = 7$
 $= 2^{1^1} | 1 + 0 | + J^1(0) = 2+0+1 = 3$
 $= 2^{1^1} | 1 + 1 | + J^1(1) = 2+1+2 = 5$

Hence at $x_1 = 1$; $U_1^* = 0$; $J_1^* = 3$

At $x_1 = 2$ $U \in \{-1, 0, 1\}$

At $U_1 = -1$ $x_2 = -(2) + 1 - 1 = -2$ Hence $x_2 = -2$ at $U_1 = -1$
 $U_1 = 0$ $x_2 = -(2) + 1 + 0 = -1$ $x_2 = -1$ at $U_1 = 0$
 $U_1 = 1$ $x_2 = -(2) + 1 + 1 = 0$ $x_2 = 0$ at $U_1 = 1$

Now, cost $= 2^{1^2} | 1 - 1 | + J^2(-2) = 4+1+8 = 13$
 $= 2^{1^2} | 1 + 0 | + J^2(-1) = 4+0+4 = 8$
 $= 2^{1^2} | 1 + 1 | + J^2(0) = 4+1+1 = 6$

Hence at $x_1 = 2$; $U_1^* = 1$; $J_1^* = 6$

At $x_0 = -2$ $U \in \{-1, 0, 1\}$

At $U_0 = -1$ $x_1 = -(-2) + 1 - 1 = 2$
 $U_0 = 0$ $x_1 = -(-2) + 1 + 0 = 3 \Rightarrow 2$
 $U_0 = 1$ $x_1 = -(-2) + 1 + 1 = 4 \Rightarrow 2$

Hence $x_1 = 2$ at $U_0 = -1$
 $x_1 = 2$ at $U_0 = 0$
 $x_1 = 2$ at $U_0 = 1$

Now, cost $= 2| -2 | + | -1 | + J(2) = 11$
 $= 2| -2 | + | 0 | + J(2) = 10$
 $= 2| -2 | + | 1 | + J(2) = 11$

Hence at $x_0 = -2$; $U_0^* = 0$; $J_0^* = 10$

① 9

② 5

③ 2

④ 5

⑤ 6

At $x_0 = -1$ $U \in \{-1, 0, 1\}$

At $U_0 = -1$ $x_1 = -(-1) + 1 - 1 = 1$
 $U_0 = 0$ $x_1 = -(-1) + 1 + 0 = 2$
 $U_0 = 1$ $x_1 = -(-1) + 1 + 1 = 3 \Rightarrow 2$

Hence $x_1 = 1$ at $U_0 = -1$
 $x_1 = 2$ at $U_0 = 0$
 $x_1 = 2$ at $U_0 = 1$

Now, cost $= 2| -1 | + | -1 | + J(1) = 6$
 $= 2| -1 | + | 0 | + J(2) = 8$
 $= 2| -1 | + | 1 | + J(2) = 9$

Hence at $x_0 = -1$; $U_0^* = -1$; $J_0^* = 6$

At $x_0 = 0$ $U \in \{-1, 0, 1\}$

At $U_0 = -1$ $x_1 = -(0) + 1 - 1 = 0$
 $U_0 = 0$ $x_1 = -(0) + 1 + 0 = 1$
 $U_0 = 1$ $x_1 = -(0) + 1 + 1 = 2$

Hence $x_1 = 0$ at $U_0 = -1$
 $x_1 = 1$ at $U_0 = 0$
 $x_1 = 2$ at $U_0 = 1$

Now, cost $= 2| 0 | + | -1 | + J(0) = 8$
 $= 2| 0 | + | 0 | + J(1) = 3$
 $= 2| 0 | + | 1 | + J(2) = 7$

Hence at $x_0 = 0$; $U_0^* = -1/0$; $J_0^* = 3$

9

5

2

3

6

At $x_0 = 1$ $U \in \{-1, 0, 1\}$

At $U_0 = -1$ $x_1 = -(1) + 1 - 1 = -1$
 $U_0 = 0$ $x_1 = -(1) + 1 + 0 = 0$
 $U_0 = 1$ $x_1 = -(1) + 1 + 1 = 1$

Hence $x_1 = -1$ at $U_0 = -1$
 $x_1 = 0$ at $U_0 = 0$
 $x_1 = 1$ at $U_0 = 1$

Now, cost $= 2| 1 | + | -1 | + J(-1) = 8$
 $= 2| 1 | + | 0 | + J(0) = 4$
 $= 2| 1 | + | 1 | + J(1) = 6$

Hence at $x_0 = 1$; $U_0^* = 0$; $J_0^* = 4$

At $x_0 = 2$ $U \in \{-1, 0, 1\}$

At $U_0 = -1$ $x_1 = -(2) + 1 - 1 = -2$

$U_0 = 0$ $x_1 = -(2) + 1 + 0 = -1$

$U_0 = 1$ $x_1 = -(2) + 1 + 1 = 0$

Hence $x_1 = -2$ at $U_0 = -1$

$x_1 = -1$ at $U_0 = 0$

$x_1 = 0$ at $U_0 = 1$

$$\begin{aligned} \text{Now, cost} &= 2 \left| \frac{2}{4} \right| + 1 - 1 + J(-2) = 14 \\ &= 2 \left| \frac{2}{4} \right| + 1 \left| \frac{0}{5} \right| + J(-1) = 9 \\ &= 2 \left| \frac{2}{4} \right| + 1 \left| \frac{1}{1} \right| + J(0) = 7 \end{aligned}$$

Hence at $x_0 = 2$; $U_0^* = 1$; $J_0^* = 7$

Finite-Horizon optimal Control:

$$x_{n+1} = f(x_n, u_n, \omega_n)$$

Uncertainty

Then the minimisation ~~formulation~~ of the cost has to be rewritten as the ~~expectation~~ of the minimisation of the cost.

$$\min_{u_0, u_1, \dots, u_{n-1}} \sum_{i=0}^{n-1} q_i(x_i, u_i) + q_n(x_n)$$

$$\min_{u_0, u_1, \dots, u_{n-1}} \mathbb{E} \left(\sum_{i=0}^{n-1} q_i(x_i, u_i, \omega_i) + q_n(x_n) \right)$$

$x_n \rightarrow$ coffee in week n (in hand)
 $v_n \rightarrow$ coffee ~~to be ordered for week n.~~ at week n .

$w_n \rightarrow$ client demand / coffee used.

(non negative) inventory $\rightarrow \max(0, x_n + v_n - w_n)$

constraint (max storage capacity) = 2 ie $x_n + v_n \leq 2$

storage cost for period n . $(x_n + v_n - w_n)^2$

$$\Pi^* = \{v_0^*, v_1^*, \dots, v_{N-1}^*\} \quad \mathbb{E} \left(\sum_{i=k}^{N-1} g_i(x_i, v_i, w_i) + g_N(x_N) \right)$$

$$J_k(\gamma_K) = \min_{v_k} \mathbb{E} [g_k(\gamma_k, v_k) + J_{k+1}(f(\gamma_k, v_k))]$$