## MDP FROM THE DISCRETIZATION OF HJB

## 1. Problem setup

- 1.1. **HJB.** We want to solve a d-dimensions HJB given below:
  - Domain

$$O = \{ x \in \mathbb{R}^d : 0 < x_i < 1, i = 1, 2, \dots d \}.$$

• Equation on O:

$$\left(\frac{1}{2}\Delta - \lambda\right)v(x) + \inf_{a} \left\{ \sum_{i=1}^{d} b_{i}(x,a) \frac{\partial v(x)}{\partial x_{i}} + \ell(x,a) \right\} = 0.$$

• Dirichlet data on  $\partial O$ :

$$v(x) = g(x).$$

## 1.2. Examples.

1.2.1. 1-d HJB: Whittles Flypaper. This example is taken from the reference P97, Example 4 of [Roger and Williams 2000] We want to solve HJB

$$\inf_{a} \{b(x,a)v'(x) + \frac{1}{2}\sigma^{2}v''(x) - \lambda v(x) + \ell(x,a)\} = 0, \quad \text{ on } O = (l,u)$$

with Dirichlet data

$$v(x) = g(x), \quad x = l, u.$$

Let parameters be given by

$$O = (0, z), \ \sigma = 1, \ b(x, a) = a, \ \lambda = 0, \ \ell(x, a) = \frac{1}{2}(a^2 + 1), \ g(x) = -\ln(c_1 e^x + c_2 e^{-x}).$$

The value function and the optimal policy are

$$v(x) = g(x), \ a^*(x) = -g'(x).$$

Ex. In the above Whittle's "flypaper", answer the following questions:

- $\bullet$  show that v is concave.
- show that the optimal policy  $|a^*(x)| \leq 1$ .
- solve for the exact solution for terminal cost given by

$$g(x) = x^2.$$

1.2.2. Multidimensional HJB with quadratic function as its solution. Consider

$$\frac{1}{2}\Delta v + \inf_{a \in \mathbb{R}^d} \left( a \cdot \nabla v + d + 2|x|^2 + \frac{1}{2}|a|^2 \right) = 0, \ x \in O.$$

with

$$v(x) = -|x|^2, \ x \in \partial O.$$

The exact solution is

$$v(x) = -|x|^2$$
, with  $a = 2x$ .

This means that the solution is invariant if  $\inf_{a \in \mathbb{R}^d}$  is replaced by  $\inf_{a \in 3O}$ .

## 2. Discretization

2.1. **FDM.** We introduce some notions of finite difference operators. Commonly used first order finite difference operators are FFD, BFD, and CFD. Forward Finite Difference (FFD) is

$$\frac{\partial}{\partial x_i}v(x) \approx \delta_{he_i}v(x) := \frac{v(x + he_i) - v(x)}{h}.$$

Backward Finite Difference (BFD) is

$$\frac{\partial}{\partial x_i}v(x) \approx \delta_{-he_i}v(x) := \frac{v(x - he_i) - v(x)}{-h}.$$

Central Finite Difference (CFD) is

$$\frac{\partial}{\partial x_i}v(x) \approx \delta_{\pm he_i}v(x) := \frac{1}{2}(\delta_{-he_i} + \delta_{he_i})v(x) = \frac{v(x + he_i) - v(x - he_i)}{2h}.$$

Second order finite difference operators are the followings:

$$\frac{\partial^2}{\partial x_i^2}v(x) \approx \delta_{-he_i}\delta_{he_i}v(x) = \frac{v(x+he_i) - 2v(x) + v(x-he_i)}{h^2}.$$

Although the next operator will not be used below, we will write it for its completeness. If  $i \neq j$ , we use

$$\frac{\partial^2}{\partial x_i \partial x_j} v(x) \approx \delta_{\pm he_i} \delta_{\pm he_j} v(x)$$

$$= \frac{v(x + he_i + he_j) - v(x + he_i - he_j) - v(x - he_i + he_j) + v(x - he_i - he_j)}{4h^2}.$$

2.2. **CFD on HJB.** Approximations for HJB are

$$\frac{\partial v(x)}{\partial x_i} \leftarrow \delta_{\pm he_i} v(x)$$

and

$$\frac{\partial^2 v(x)}{\partial x_i^2} \leftarrow \delta_{-he_i} \delta_{he_i} v(x).$$

For simplicity, if we set

$$\gamma = \frac{d}{d+h^2\lambda}, \ p^h(x \pm he_i|x,a) = \frac{1}{2d}(1 \pm 2hb_i(x,a)), \ \ell^h(x,a) = \frac{h^2\ell(x,a)}{d},$$

then it yields DPP

$$v(x) = \gamma \inf_{a} \left\{ \ell^{h}(x, a) + \sum_{i=1}^{d} p^{h}(x + he_{i}|x, a)v(x + he_{i}) + p^{h}(x - he_{i}|x, a)v(x - he_{i}) \right\}.$$