Intern Project

Intern Name

Abstract

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1 Preliminary

2 Second-order Algorithm

Assume we define a field with $x_t = \alpha_t x_0 + \beta_t x_1$, we can manually calculate \dot{x}_t and \ddot{x} , representing the first-order and second-order gradient, respectively.

Definition 2.1. The loss function for second order method contain two parts. We define the first part which is trying to using \dot{x}_t , x_t and t to learn function $u_{1,t}$, thus the loss is

$$L_{2,1,\theta_1} := \|\dot{x}_t - u_{1,\theta_1}(x_t,t)\|_2^2$$

Next, we define the second part which is trying to use \ddot{x}_t , $u_{1,\theta_1}(x_t,t)$, x_t and t to learn u_{2,θ_2} function, thus the loss is

$$L_{2,2,\theta_2,\theta_1} := \|\ddot{x}_t - u_{2,\theta_2}(u_{1,\theta_1}(x_t,t), \mathbf{x_t}, t)\|_2^2$$

Overall the total loss is

$$L_{2,\theta} := L_{2,1,\theta_1,\theta_2} + L_{2,2,\theta_2}$$

Algorithm 1

```
1: procedure OurSecondOrderAlg()
 2:
        for each iteration do
            Random sample x_0 and time t, with target x_1
 3:
            x_t \leftarrow \alpha_t \cdot x_0 + \sqrt{1 - \alpha_t^2} \cdot x_1
 4:
            Compute gradient with respect to L_{2,\theta}
 5:
                                                                                                   \triangleright see Def. 2.1
        end for
 6:
        return u_1, u_2
                                                                                     7:
 8: end procedure
 9: /* Below is an inference algorithm that only use first-order learner u_1^*/
10: procedure INF1(u_1)
        x_0 \sim \mathcal{N}(0,1)
11:
12:
        Initial x \leftarrow x_0
        for t from 0 to 1 with step \Delta t = 0.01 do
13:
            x \leftarrow x + \Delta t \cdot u_1(x,t)
14:
15:
        end for
16:
        return x
17: end procedure
```

References

Algorithm 2

```
1: /* Option 1 */
 2: procedure INF12(u_1, u_2)
         x_0 \sim \mathcal{N}(0,1)
         Initial x \leftarrow x_0
 4:
         for t from 0 to 1 with step \Delta t_1 = 0.01 do
 5:
             y \sim \mathcal{N}(0,1)
 6:
             for t from 0 to 1 with step \Delta t_2 = 0.01 do
 7:
                  y \leftarrow y + \Delta t_2 \cdot u_2(y, x, t)
 8:
             end for
 9:
             x \leftarrow x + \Delta t_1 \cdot u_1(y,t)
10:
11:
         end for
12:
         return x
13: end procedure
```

Algorithm 3

```
1: /* Option 2 */
 2: procedure INFXX(u_1, u_2)
          x_0 \sim \mathcal{N}(0,1)
          Initial x \leftarrow x_0
 4:
          t \leftarrow 0
 5:
          \Delta t \leftarrow 0
 6:
          while t < 1 do
 7:
               \Delta u_1 \leftarrow \Delta_t \cdot u_2(u_1(x,t),x,t)
               u_1 \leftarrow u_1(x,t) + \Delta u_1
               x \leftarrow x + \Delta t \cdot u_1
10:
11:
          end while
          return x
12:
13: end procedure
```

Algorithm 4

```
1: /* Option 3 */
2: procedure INF12(u_1, u_2)
3: x_0 \sim \mathcal{N}(0, 1)
4: Initial x \leftarrow x_0
5: for t from 0 to 1 with step \Delta t = 0.01 do
6: x \leftarrow x + \Delta t \cdot u_1(x, t) + \frac{(\Delta t)^2}{2} \cdot u_2(u_1(x, t), x, t)
7: end for
8: return x
9: end procedure
```