# Reinforcement Learning Seminar

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- $\blacktriangleright$  Goal1 : Reproduce the SLIAR optimal control by uning PMP
- ► Method : PMP (Adjoint)

#### SLIAR Model

► SLIAR Model Equations and structure.

$$\begin{cases} S' &= -\beta(1-\sigma)S\Lambda - \nu S \\ L' &= \beta(1-\sigma)S\Lambda - \kappa L \\ I' &= p\kappa L - \alpha I - \tau I \\ A' &= (1-p)\kappa L - \eta A \end{cases} \text{ with } \Lambda = \epsilon L + (1-q)I + \delta A$$

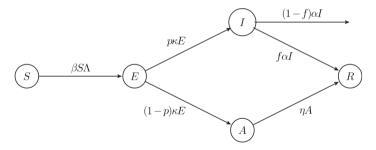
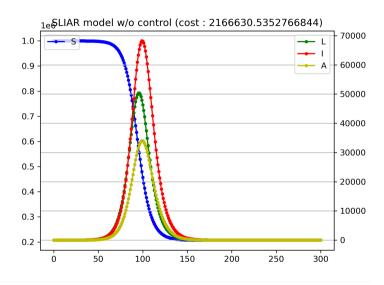


Fig. 1. SEIAR epidemic model.

# SLIAR model parameters

| Start                | 0      | $\beta$     | 7.26e-07 |
|----------------------|--------|-------------|----------|
| $\operatorname{End}$ | 300    | $\sigma$    | 0        |
| S0                   | 1e06   | $\kappa$    | 0.526    |
| L0                   | 0      | $\alpha$    | 0.224    |
| IO                   | 1      | $	au_{max}$ | 0.05     |
| A0                   | 0      | $\nu_{max}$ | 0.01     |
| $R_0$                | 1.9847 | $\epsilon$  | 0.224    |
| P                    | 1      | q           | 0.5      |
| Q                    | 1e06   | p           | 0.667    |
| R                    | 1e06   | $\delta$    | 1        |
| W                    | 0      |             |          |
|                      |        |             |          |



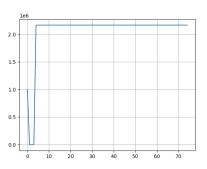
$$\min_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$$

subject to

$$\begin{cases} S' &= -\beta(1-\sigma)S\Lambda - \nu S \\ L' &= \beta(1-\sigma)S\Lambda - \kappa L \\ I' &= p\kappa L - \alpha I - \tau I \\ A' &= (1-p)\kappa L - \eta A \end{cases} \quad with \quad \Lambda = \epsilon L + (1-q)I + \delta A$$

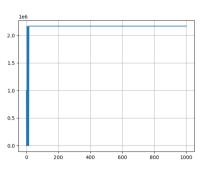
► Method : PMP(Adjoint)

ightharpoonup P = 1, Q = 1E6, R = 1E6,  $\nu_{max}$  = 0.01, learning rate : 0.1

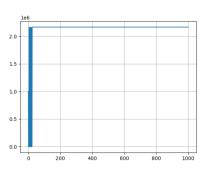


► Method : PMP(Adjoint)

▶ P = 1, Q = 1E6, R = 1E6,  $\nu_{max} = 0.01$ , learning rate : 0.001

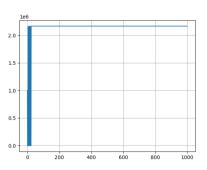


- ► Method : PMP(Adjoint)
- ▶ P = 1, Q = 1E6, R = 1E6,  $\nu_{max} = 0.01$ , learning rate : 1e-04



► Method : PMP(Adjoint)

 $\blacktriangleright$  P = 1, Q = 1E6, R = 1E6,  $\nu_{max} = 0.01,$  learning rate : 1e-06

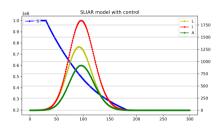


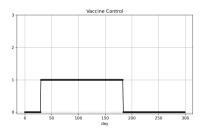
- ► Goal2 : 2-control optimal control
- ightharpoonup Method : DQN

 $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$ 

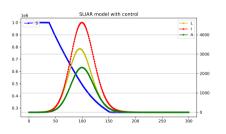
► Method : DQN

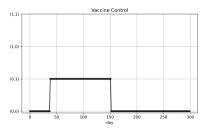
ightharpoonup P = 1, Q = 1E6, R = 1E6,  $\nu_{max}$  = 0.01,  $\tau_{max}$  = 0.05, iteration : 2000



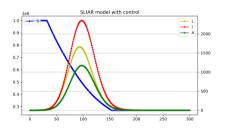


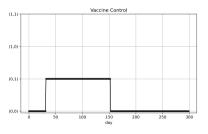
- $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$
- ► Method : DQN
- ightharpoonup P = 1, Q = 1E6, R = 1E6,  $\nu_{max}$  = 0.01,  $\tau_{max}$  = 0.05, iteration : 5000



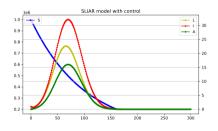


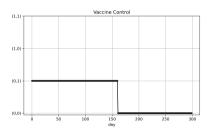
- $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$
- ► Method : DQN
- ightharpoonup P = 1, Q = 1E6, R = 1E6,  $\nu_{max}$  = 0.01,  $\tau_{max}$  = 0.05, iteration : 7000





- $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$
- ► Method : DQN
- ightharpoonup P = 1, Q = 1E6, R = 1E6,  $\nu_{max}$  = 0.01,  $\tau_{max}$  = 0.05, iteration : 10000



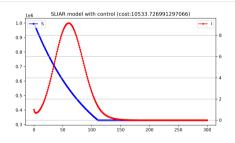


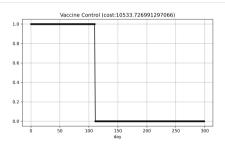
## 1-control result of DQN

 $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$ 

► Method : DQN

ightharpoonup P = 1, Q = 1e6, R = 1e6  $\nu_{max}$  = 0.01, iteration : 10,000





## 1-control result of DQN

- $\qquad \qquad \mathbf{min}_{u \in \mathcal{U}_{ad}} \int_0^T PI(t) + Q\nu^2(t) + R\tau^2(t) + W\sigma^2(t)dt$
- ► Method : DQN
- ightharpoonup P = 1, Q = 1e6, R = 1e6 $\tau_{max}$  = 0.05, iteration : 10,000

