

Models of Infectious Diseases (II)

<Last Updated on October 14, 2020>

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- Useful references:
 - Andrew Atkeson “On Using SIR Models to Model Disease Scenarios for COVID-19.”
 - <https://www.minneapolisfed.org/research/quarterly-review/on-using-sir-models-to-model-disease-scenarios-for-covid-19>
 - Ben Moll “Lockdowns in SIR Models.”
 - https://benjaminmoll.com/wp-content/uploads/2020/05/SIR_notes.pdf

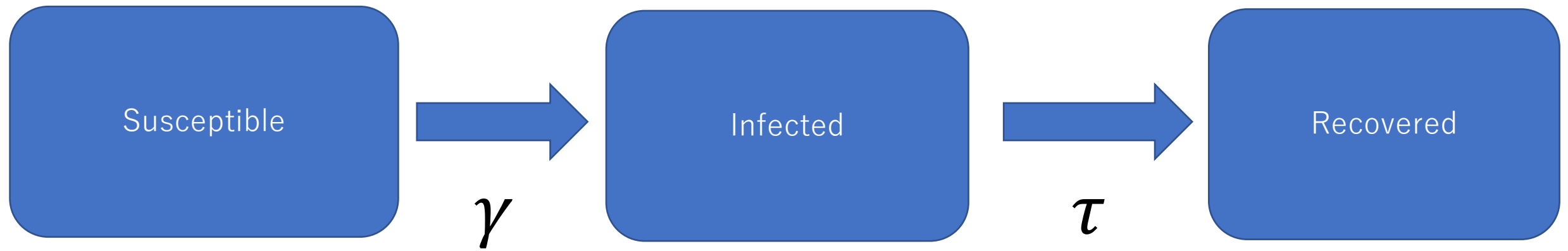
Outline

(1) SI Model

(2) SIS Model

(3) SIR Model

SIR Model



SIR Model

- Covid-19, etc.
- After you get infected, you can recover with immunity (so that you won't get infected again).
- Initially, everyone can get infected.
- A backward-looking model (easy to solve)
- Not optimization based.

- S_t : Number of people susceptible to the disease at time t .
 - I_t : Number of people with the disease at time t .
 - R_t : Number of people recovered from disease at time t .
 - Normalization: $S_t + I_t + R_t = 1$.
-
- With probability γI_t , susceptible person at time t becomes infected at time $t + 1$.
 - With probability τ , infected person at time t recover without obtaining immunity at time $t + 1$.
 - Initial condition: $I_1 = \epsilon$ and $R_1 = 0$.

- $S_{t+1} - S_t = -\gamma I_t S_t$

- $I_{t+1} - I_t = \gamma I_t S_t - \tau I_t$

- $R_{t+1} - R_t = \tau I_t$

- $S_t + I_t + R_t = 1$

Find “steady states.”

- $S_{ss} - S_{ss} = -\gamma I_{ss} S_{ss}$
- $I_{ss} - I_{ss} = \gamma I_{ss} S_{ss} - \tau I_{ss}$
- $R_{ss} - R_{ss} = \tau I_{ss}$
- $S_{ss} + I_{ss} + R_{ss} = 1$

Thus, at a steady state, we have

- $0 = -\gamma I_{ss} S_{ss}$
- $0 = \gamma I_{ss} S_{ss} - \tau I_{ss}$
- $0 = \tau I_{ss}$
- $S_{ss} + I_{ss} + R_{ss} = 1$

• From the third equation, we have

- $I_{ss} = 0$

• We are left with

- $0 = \gamma * 0 * S_{ss}$
- $S_{ss} + R_{ss} = 1$

- $0 = \gamma * 0 * S_{ss}$
- $S_{ss} + R_{ss} = 1$
- Any S_{ss} and R_{ss} satisfying the two equations above can be steady states.
- For example, $(S_{ss} = 1, R_{ss} = 0, I_{ss} = 0)$ is one steady state.
- Which steady state the economy will end up depends on the initial condition.

Compute “dynamics” (convergence towards steady state)

- Suppose that $I_1 = \epsilon$ and $R_1 = 0$ (i.e. $S_1 = 1 - \epsilon$) .

- Compute $\{I_t, S_t, R_t\}_{t=2}^{\infty}$

- How? Recursively.

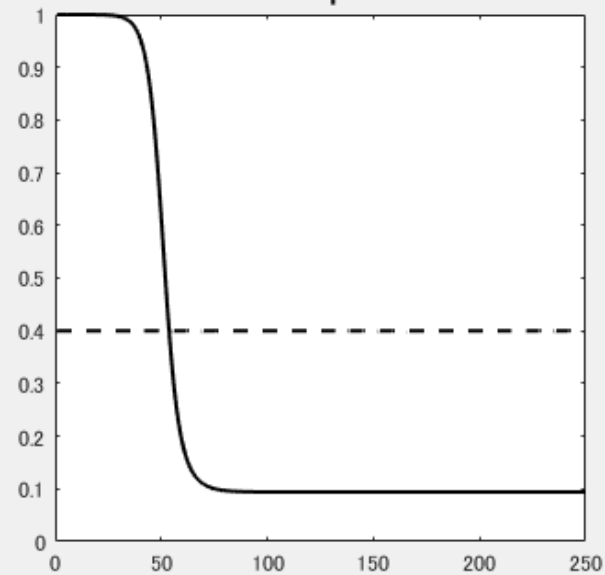
- $S_2 - S_1 = -\gamma I_1 S_1$
- $I_2 - I_1 = \gamma I_1 S_1 - \tau I_1$
- $R_2 - R_1 = \tau I_1$

- $S_2 = (1 - \epsilon) - \gamma\epsilon(1 - \epsilon) = (1 - \epsilon)(1 - \gamma\epsilon)$

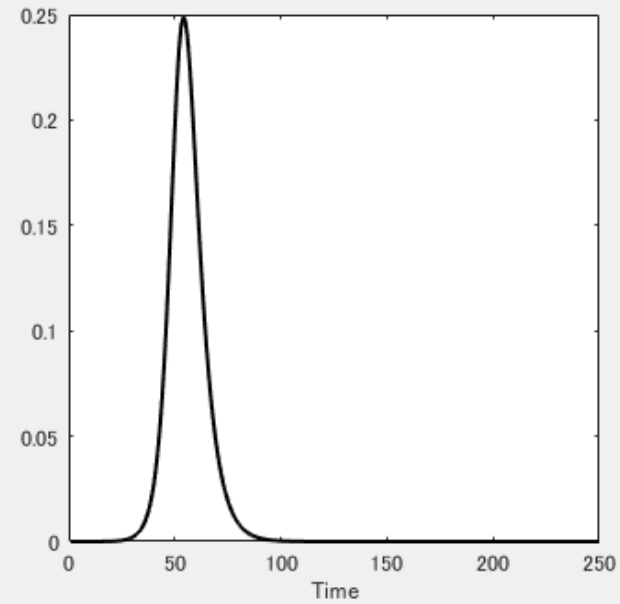
- $I_2 = \epsilon + \gamma\epsilon(1 - \epsilon) - \tau\epsilon = \epsilon(1 + \gamma(1 - \epsilon)) - \tau\epsilon$

- $R_2 = 0 + \tau\epsilon = \tau\epsilon$

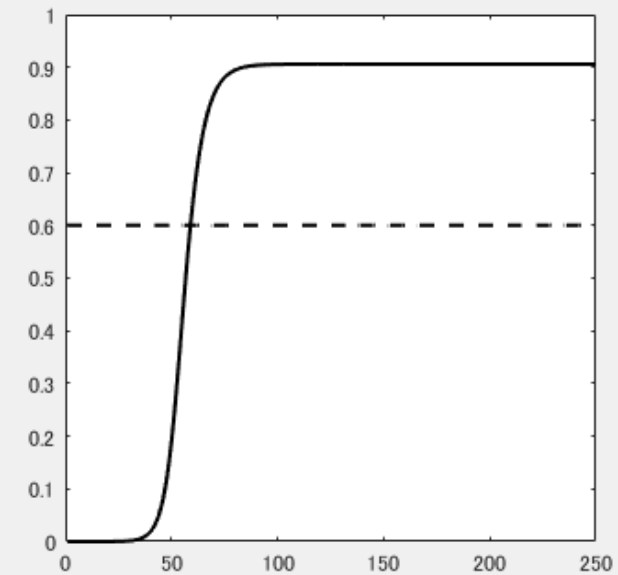
Susceptible



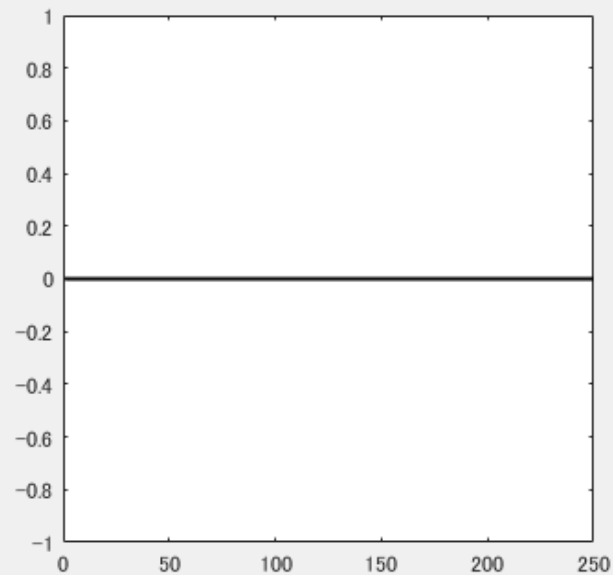
Infectious



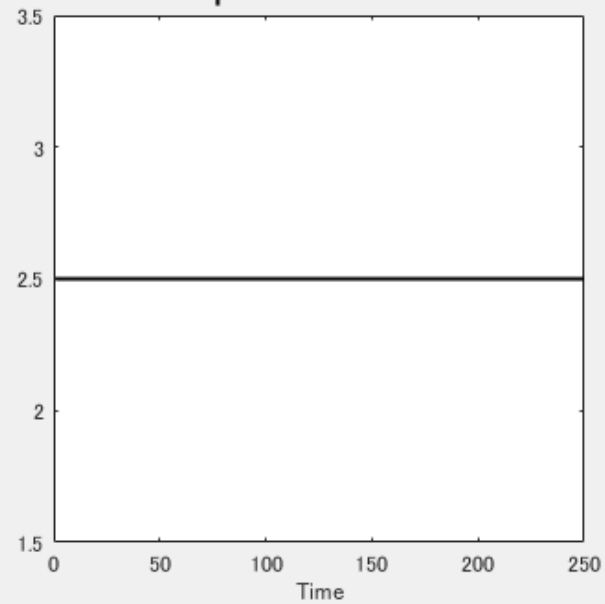
Recovered



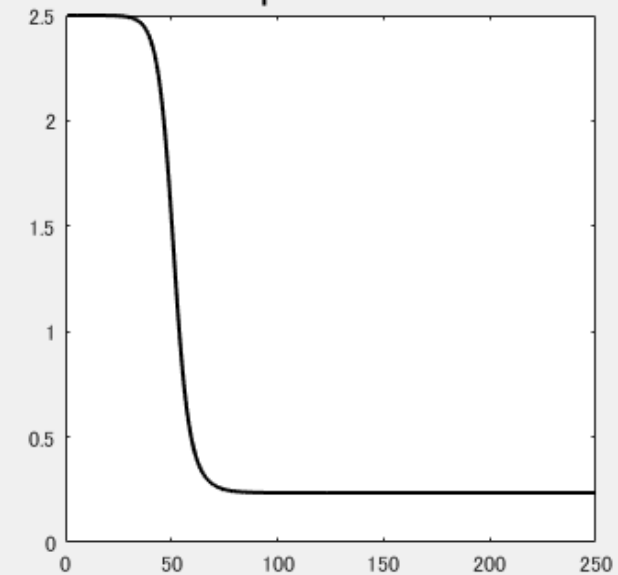
Lockdown



Basic reproductive number



Effective reproductive number



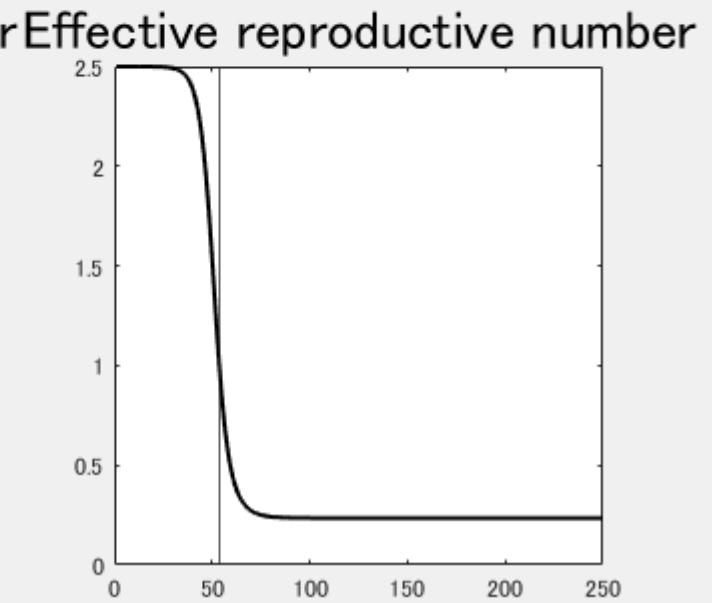
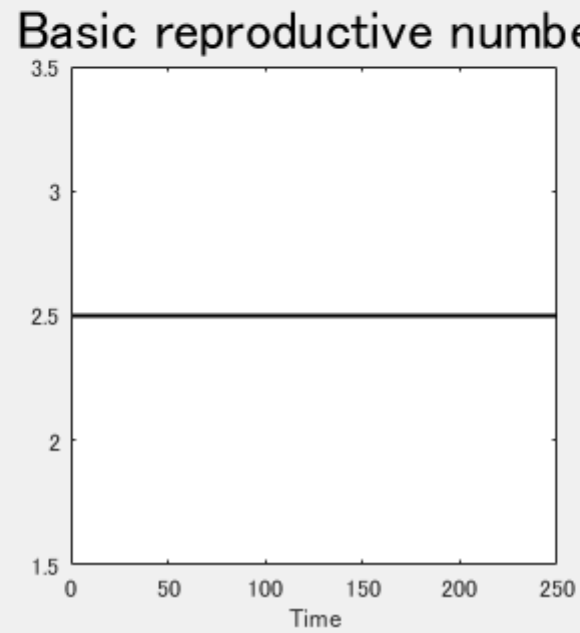
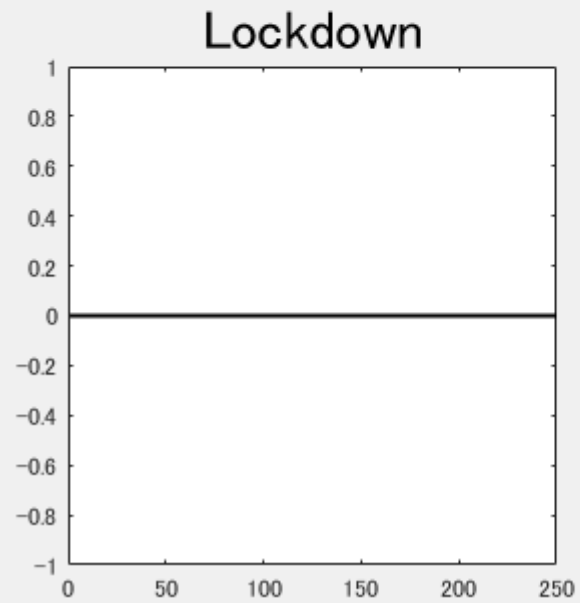
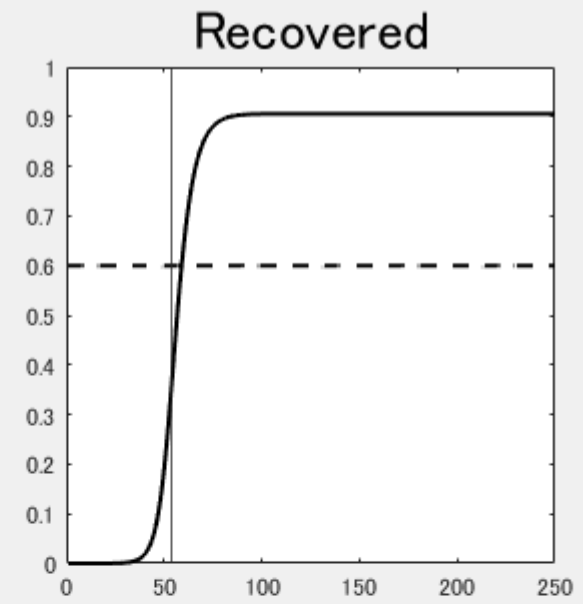
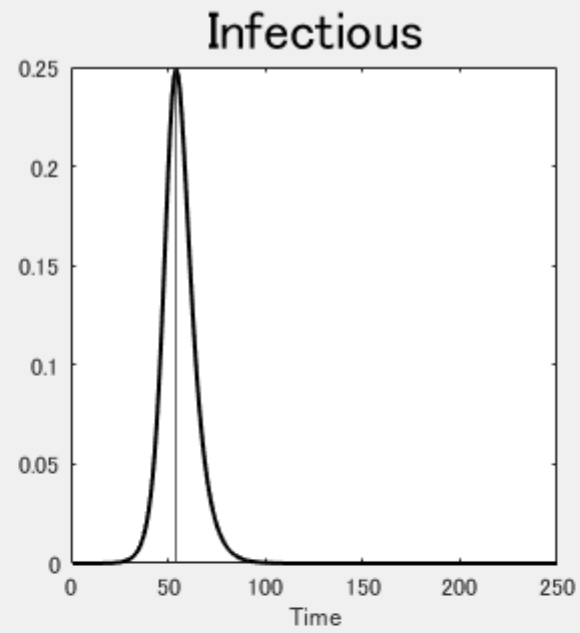
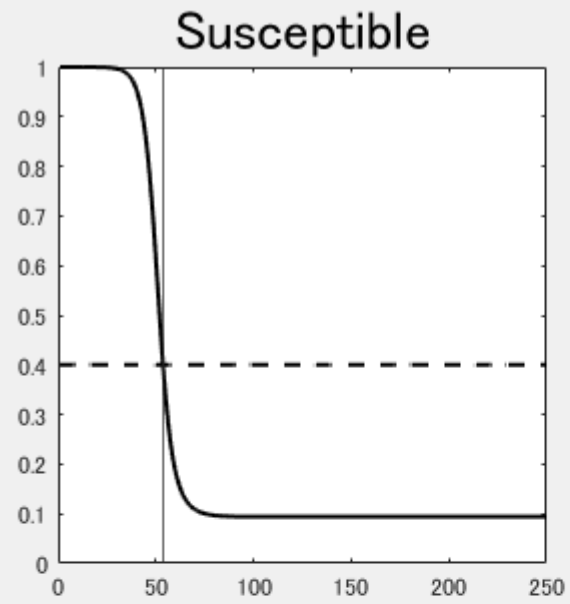
- Basic reproductive number: $X_0 = \frac{\gamma}{\tau}$
- Effective reproductive number: $X_t^e = \frac{\gamma}{\tau} S_t$
- Herd immunity threshold: $S^* = \frac{1}{X_0}$ (or $R^* = 1 - S^*$)

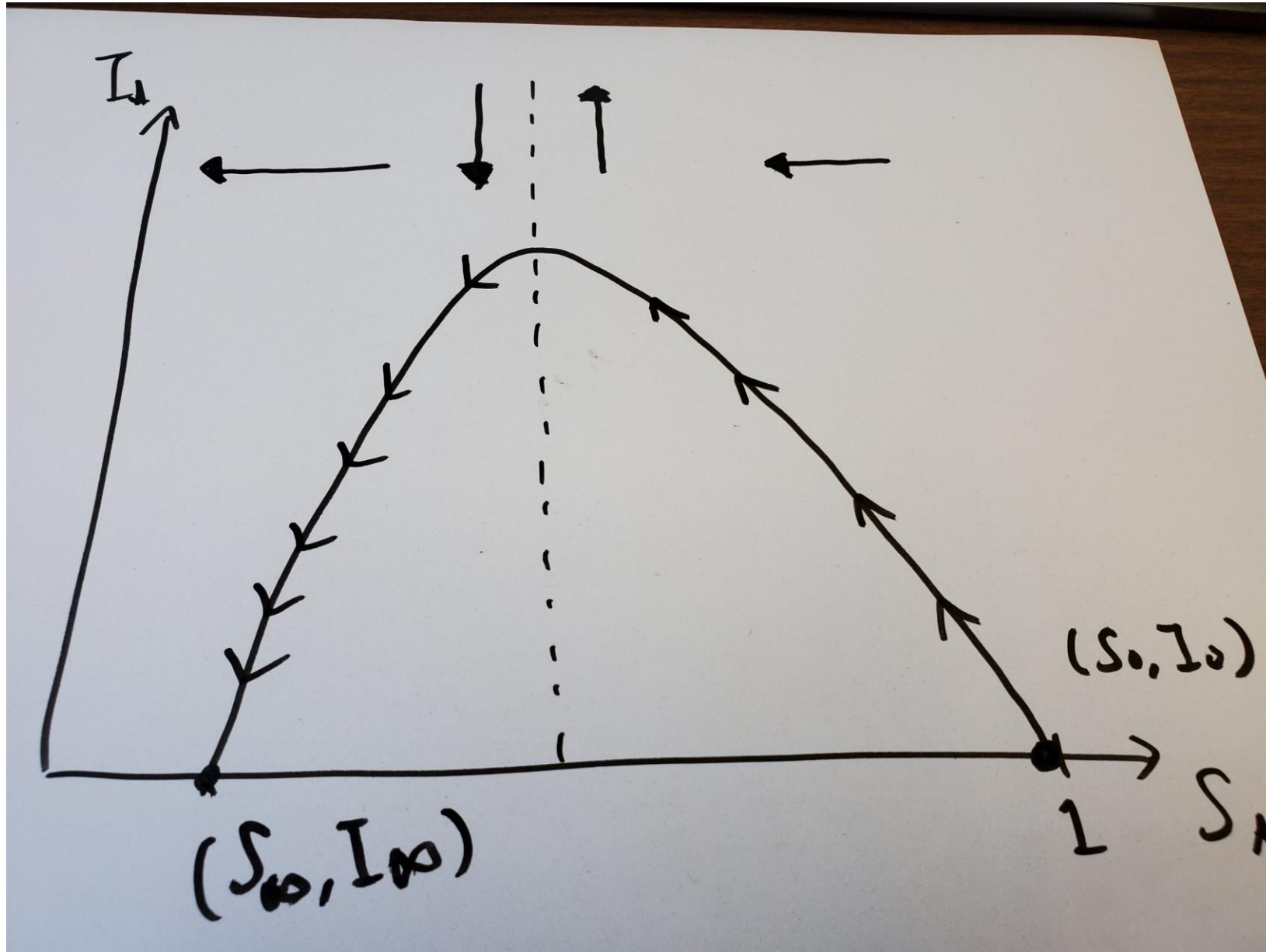
- $I_{t+1} - I_t = \gamma I_t S_t - \tau I_t > 0$

$$\Rightarrow \gamma S_t - \tau > 0$$

$$\Rightarrow S_t > \frac{\tau}{\gamma} = \frac{1}{X_0} = S^*$$

- I_t peaks when $S_t = S^*$.





- $I_{t+1} - I_t = \gamma I_t S_t - \tau I_t < 0$

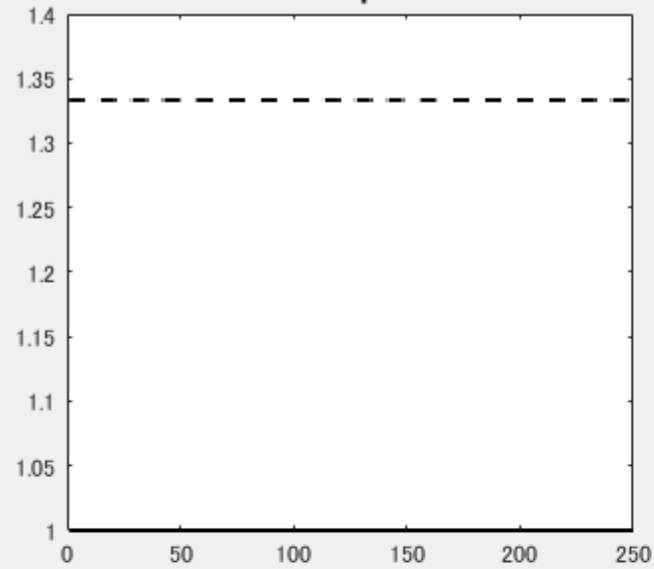
$$\Rightarrow \frac{\gamma}{\tau} < \frac{1}{S_t}$$

$$\Rightarrow X_0 < \frac{1}{S_t}$$

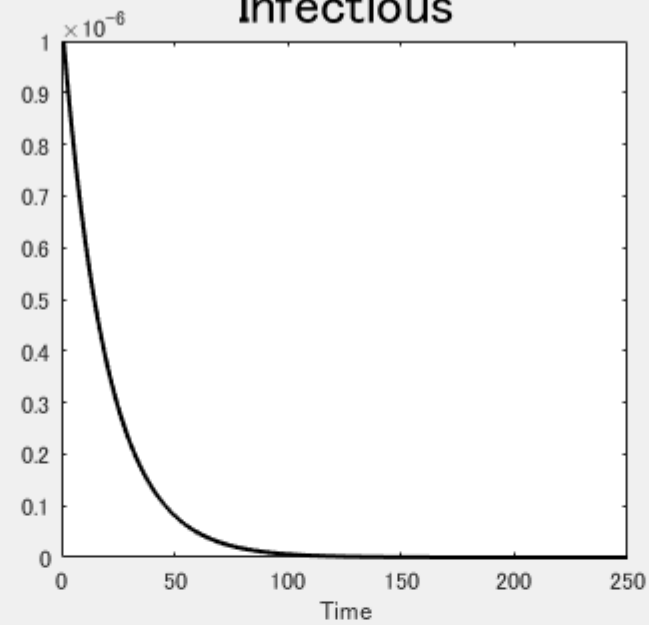
$$\Rightarrow X_0 < 1 < \frac{1}{S_t}$$

- If $X_0 < 1$, then $X_0 < \frac{1}{S_t}$ because $S_t < 1$ for any t .
- So, if $X_0 < 1$, I_t declines at any time.

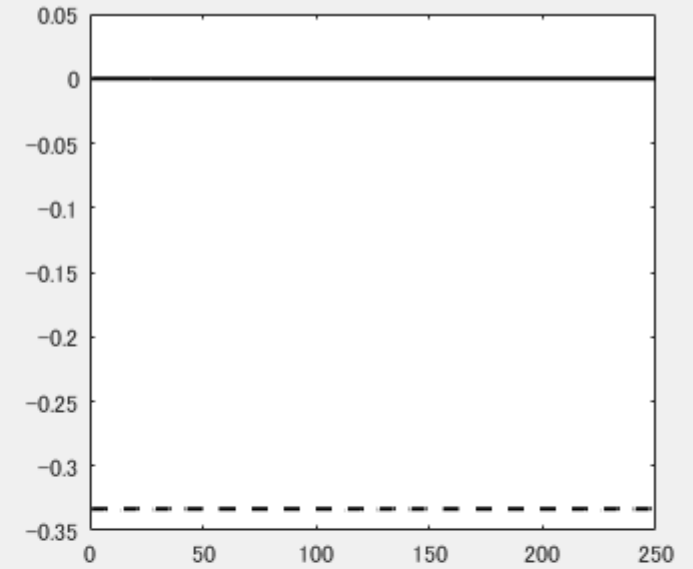
Susceptible



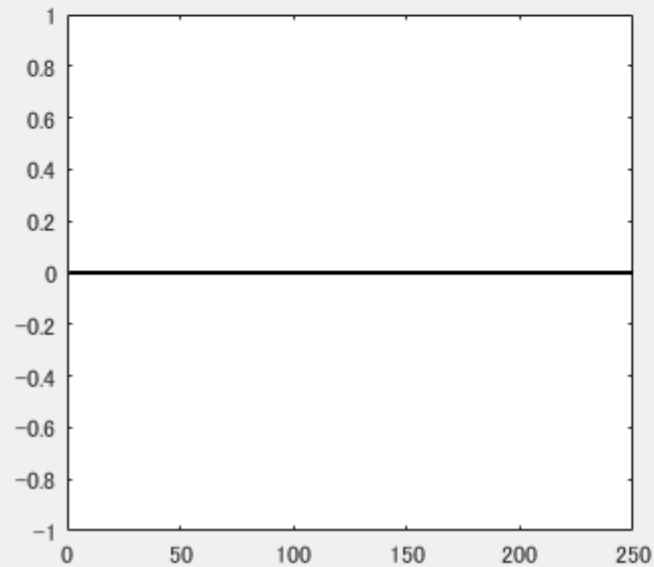
Infectious



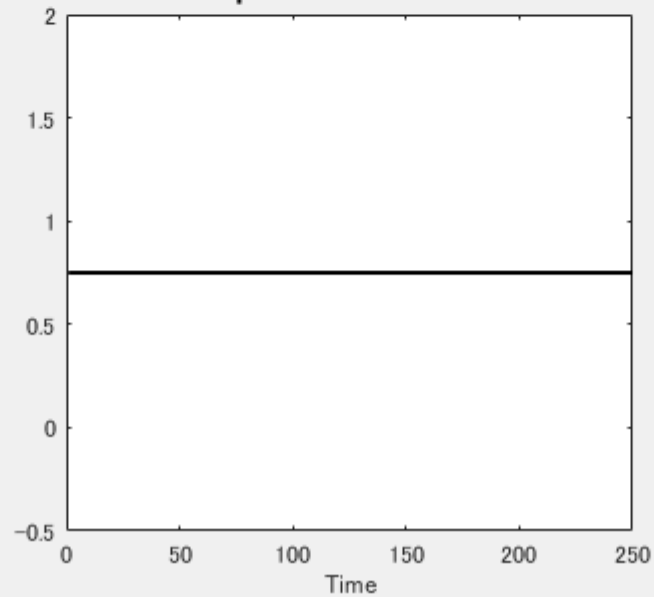
Recovered



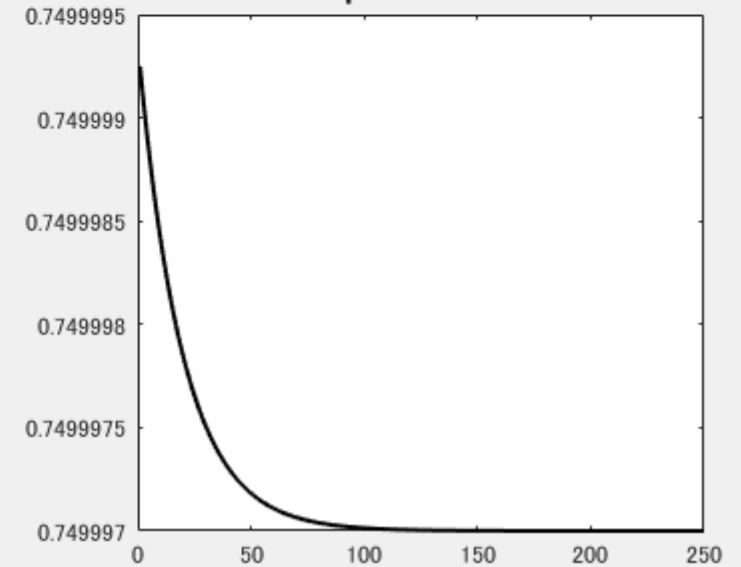
Lockdown

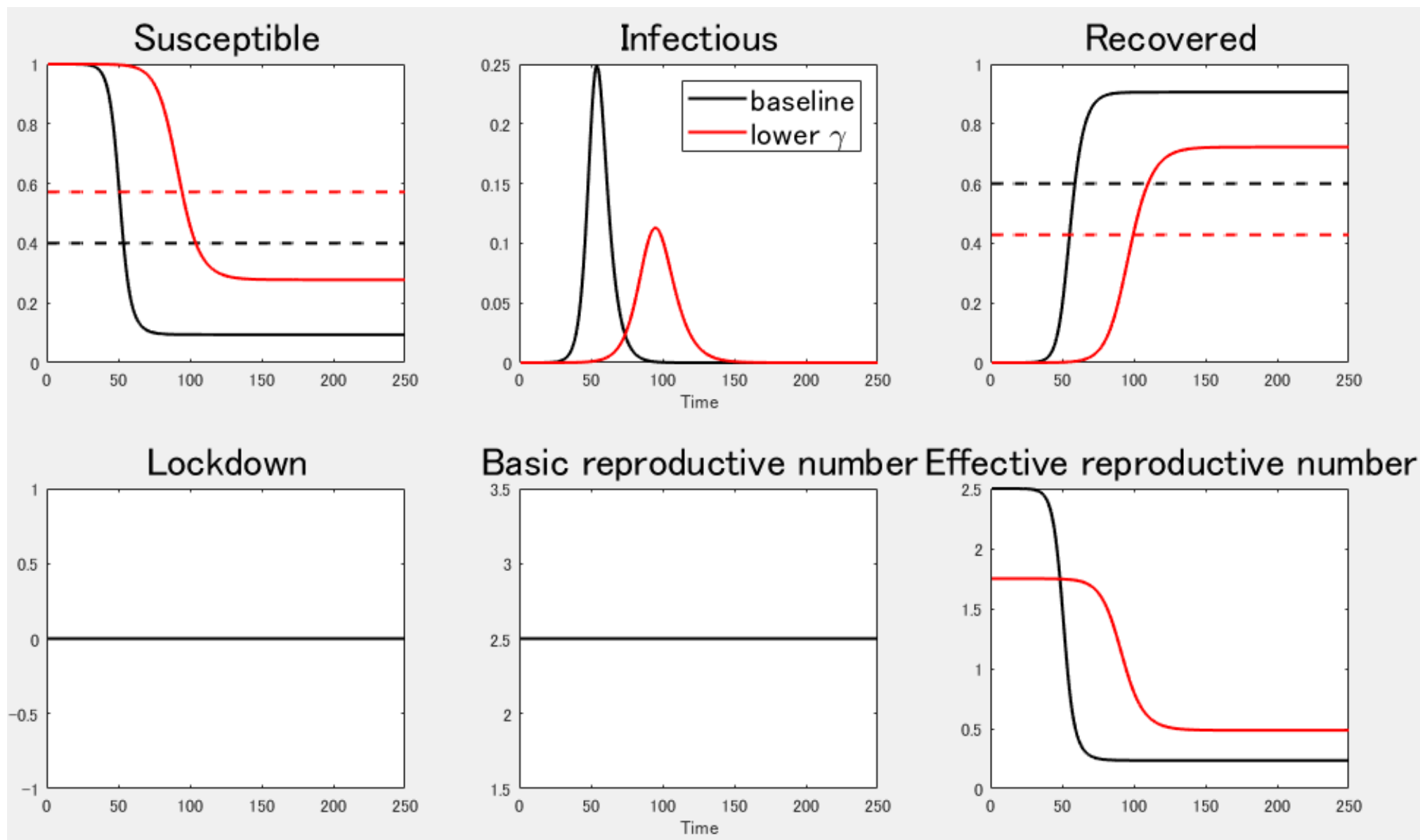


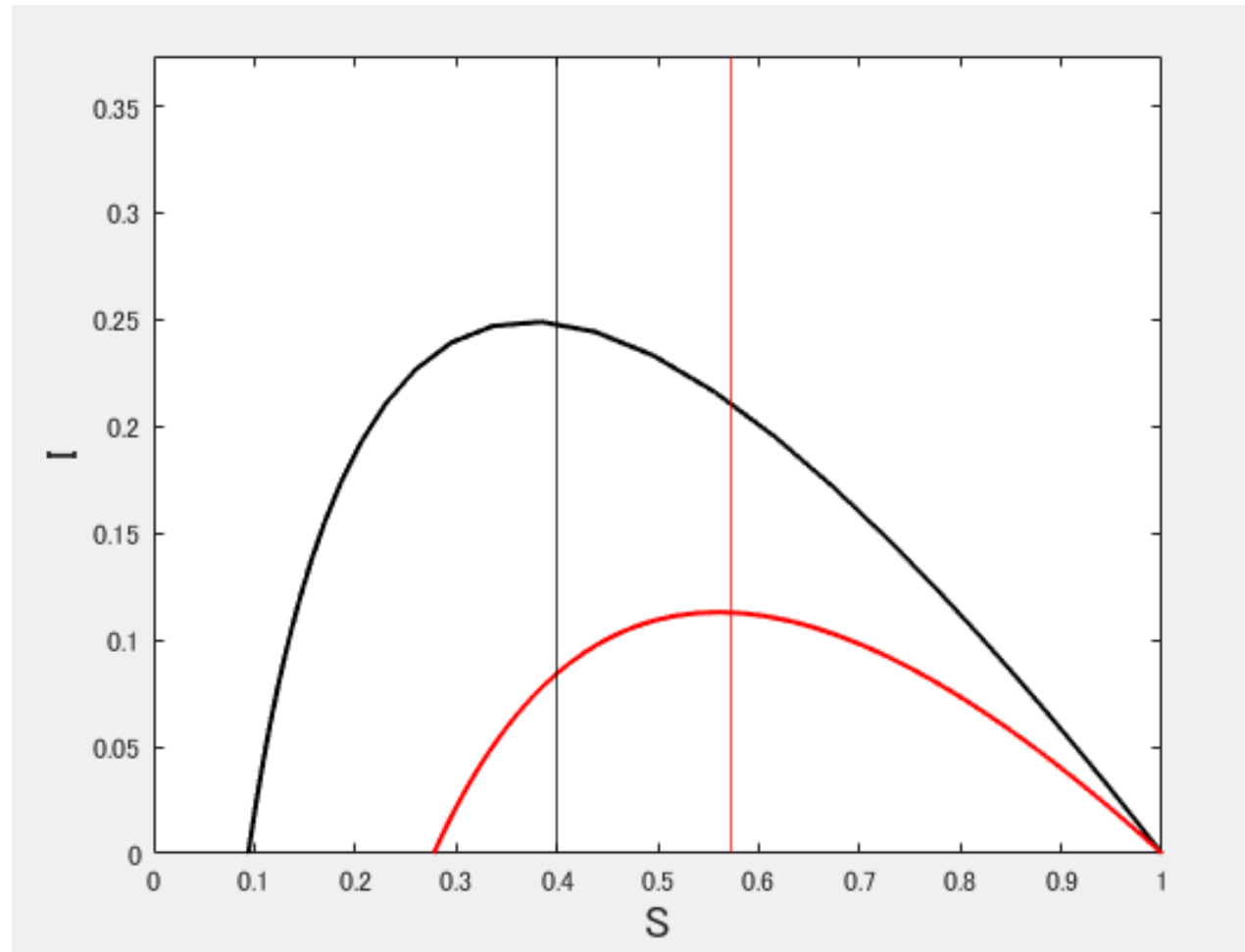
Basic reproductive number

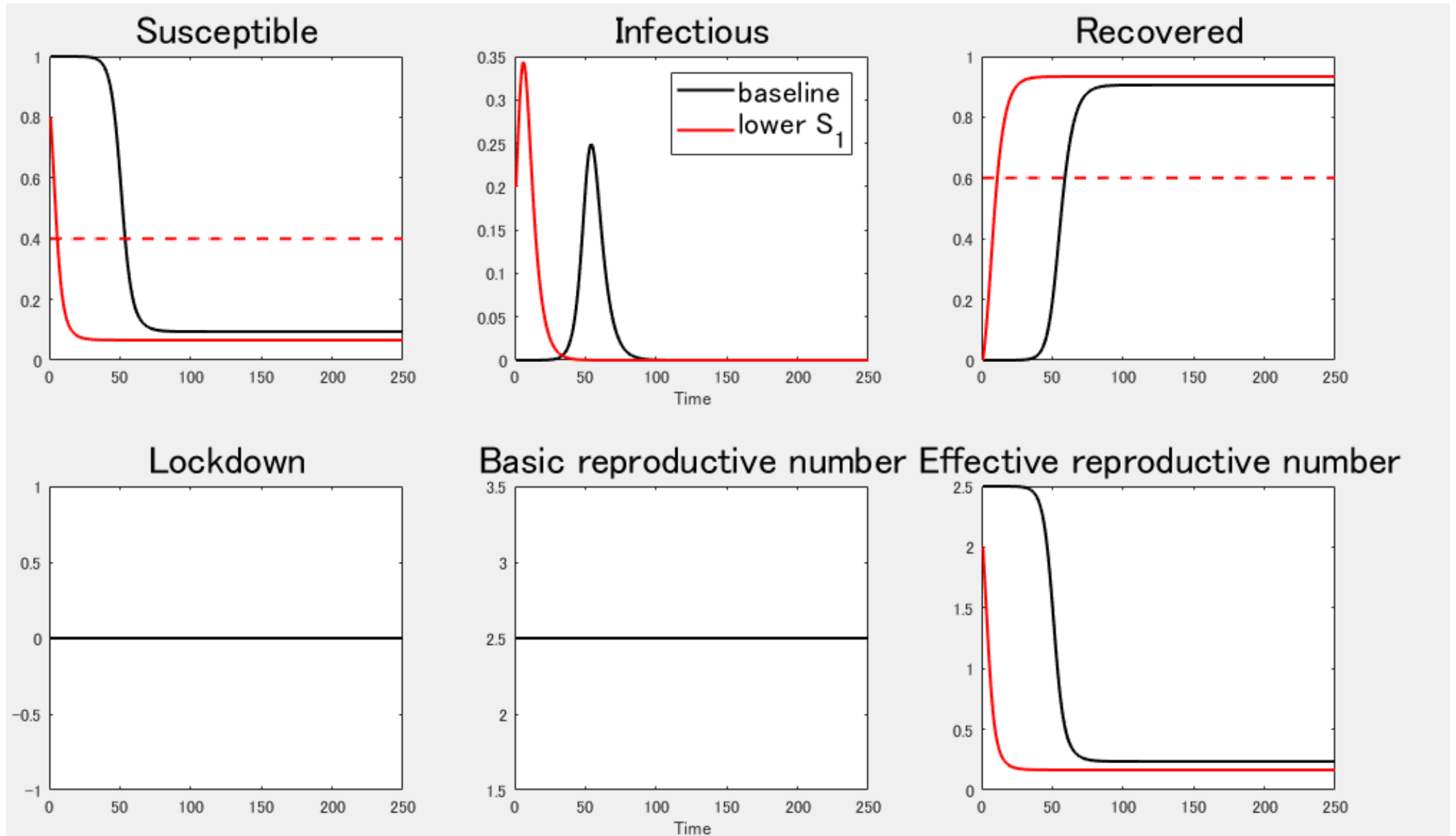


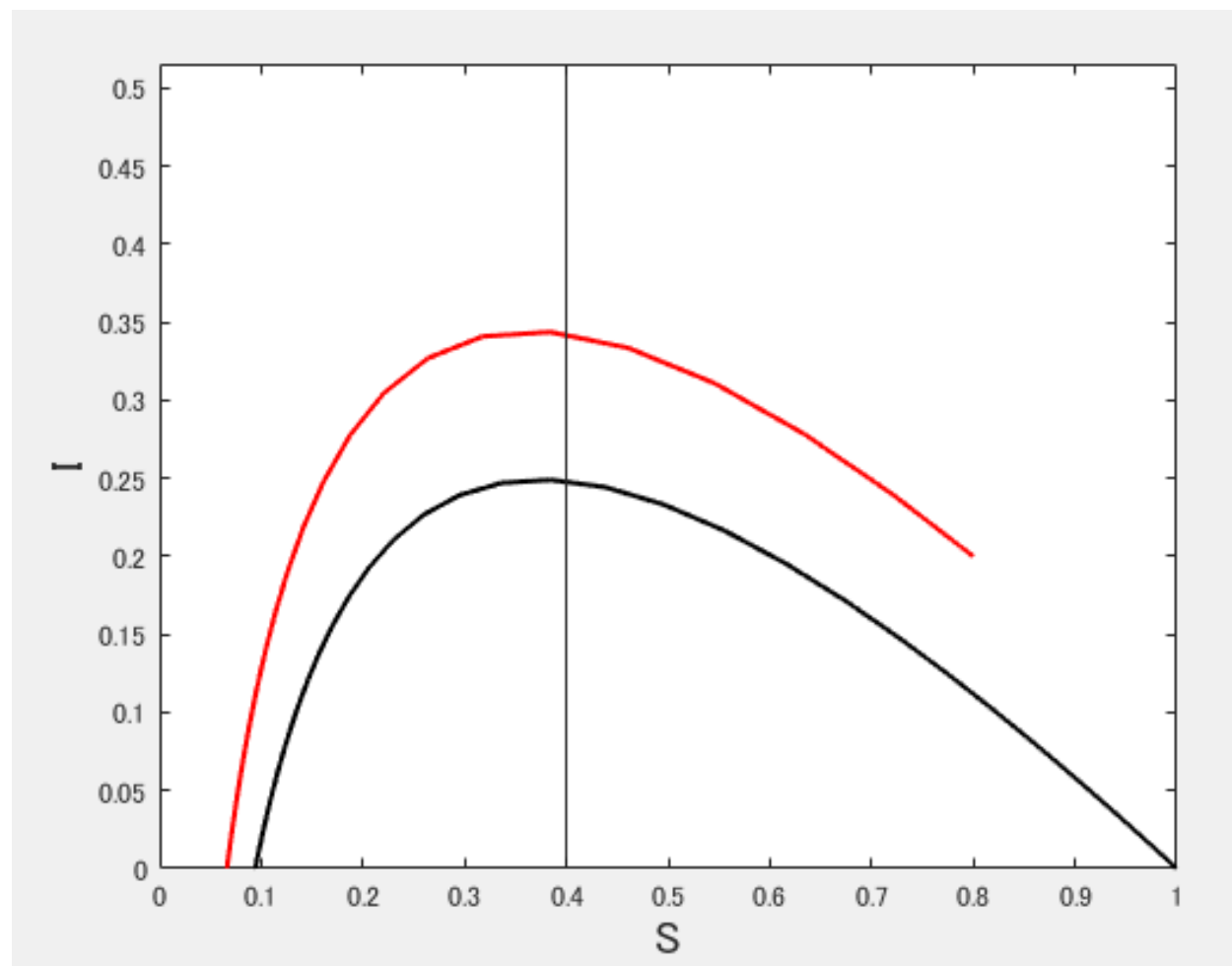
Effective reproductive number











Compute “dynamics” (exogenous shocks)

- Introduce “Lockdown” policy.
 - α_t : degree of lockdown at time t .

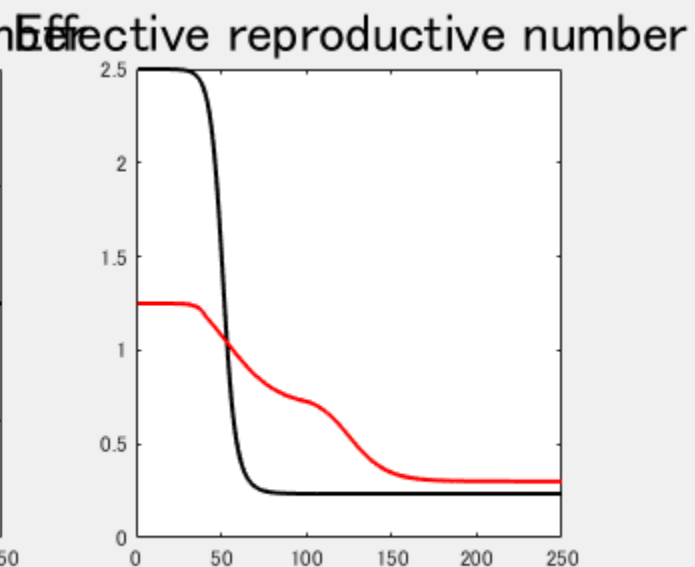
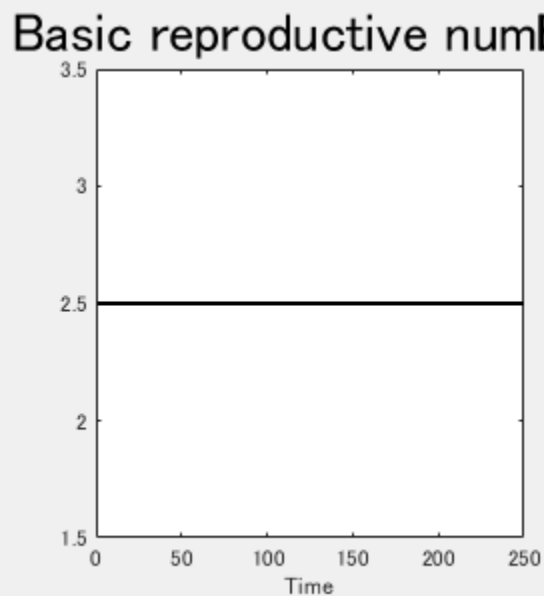
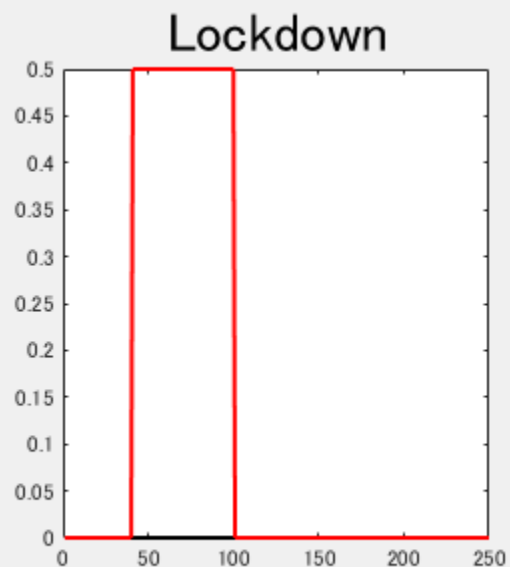
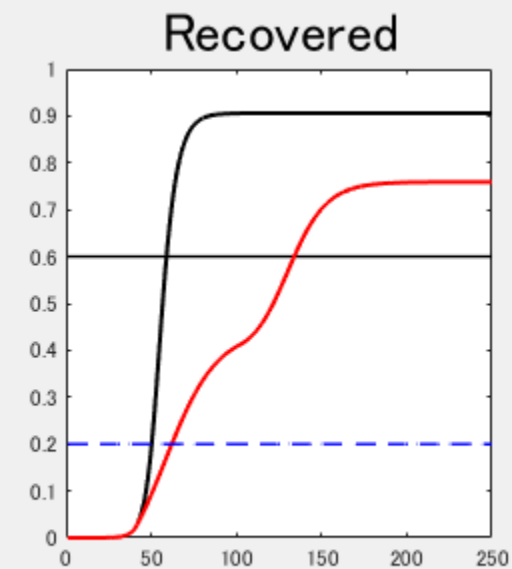
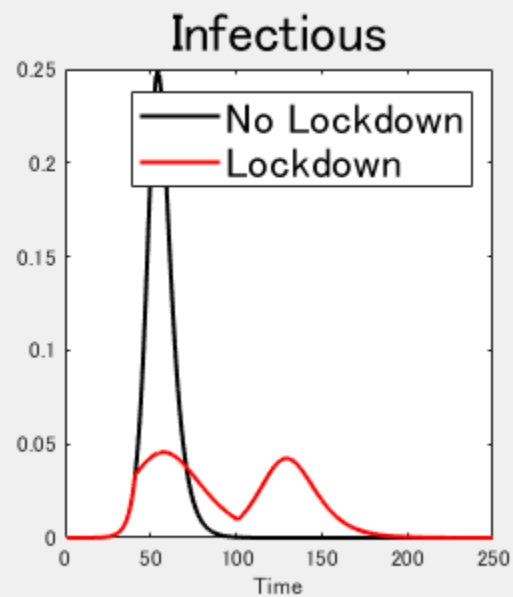
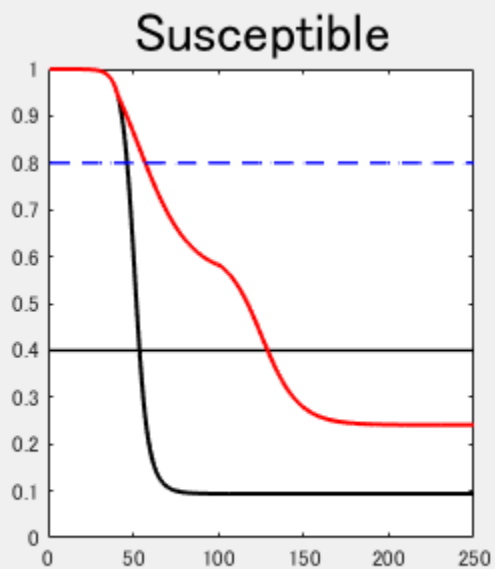
- $S_{t+1} - S_t = -\gamma(1 - \alpha_t)I_t S_t$

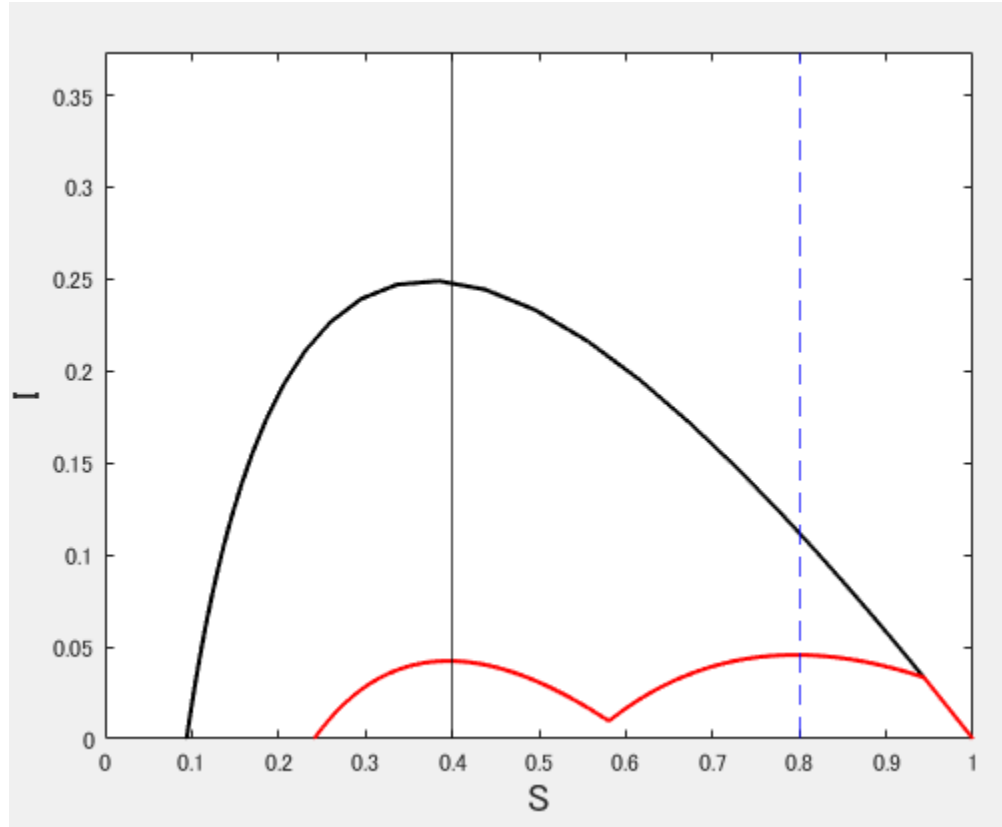
- $I_{t+1} - I_t = \gamma(1 - \alpha_t)I_t S_t - \tau I_t$

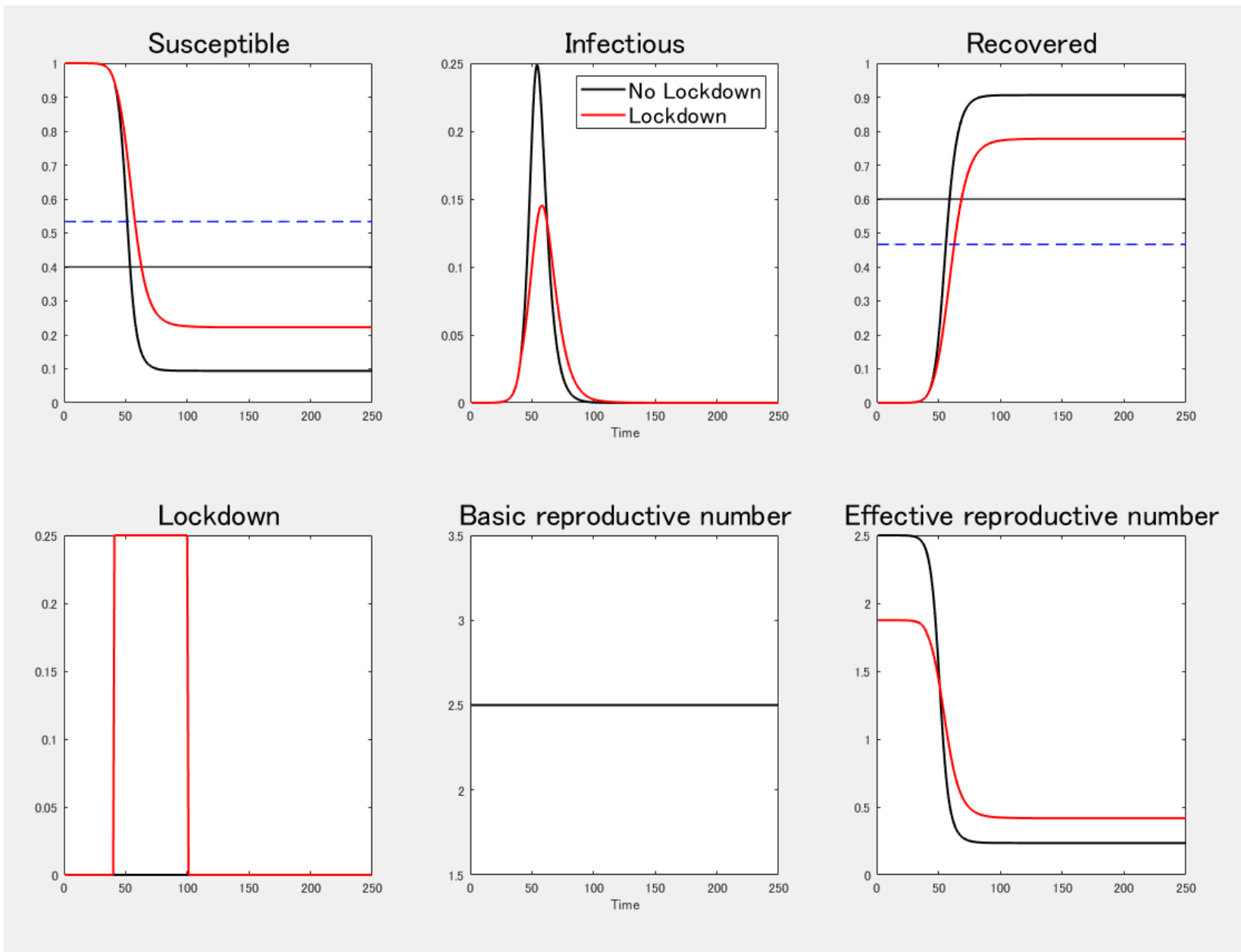
- $R_{t+1} - R_t = \tau I_t$

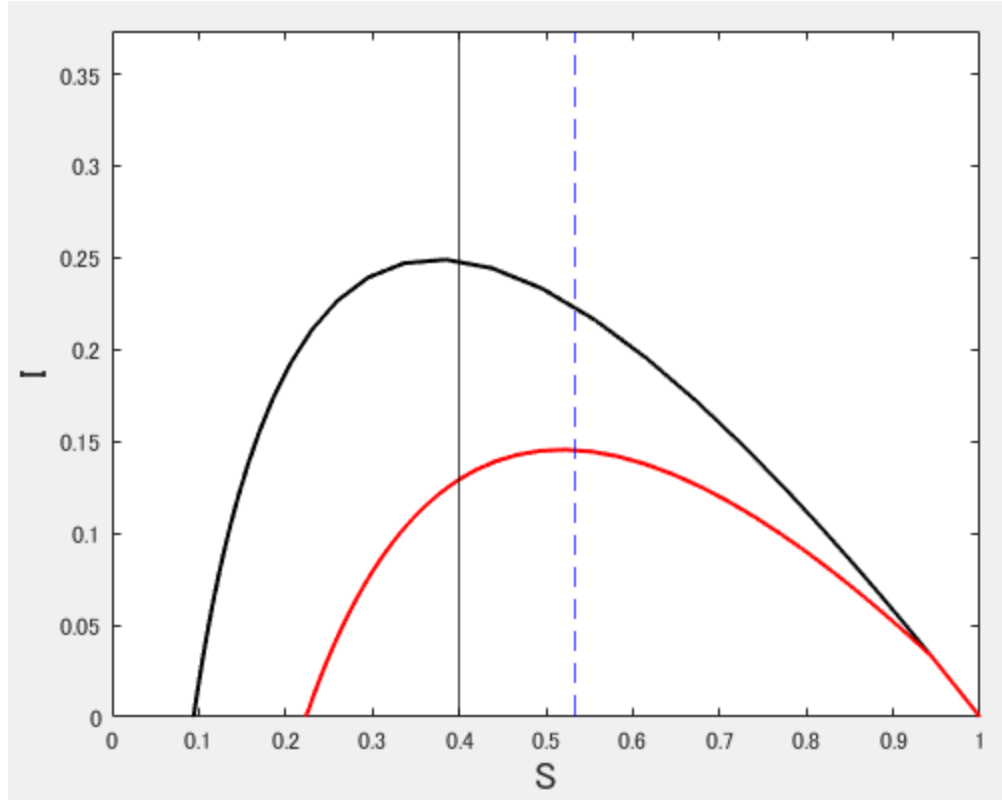
- $S_t + I_t + R_t = 1$

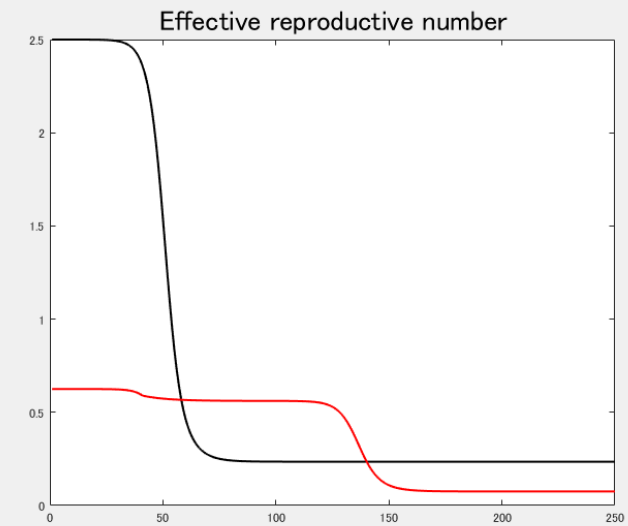
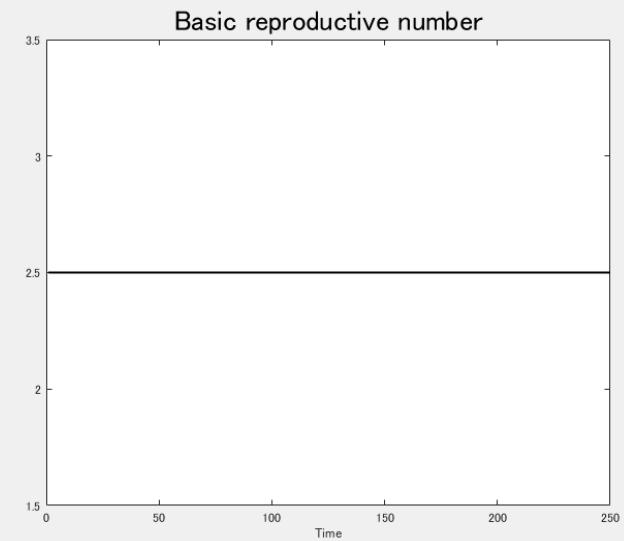
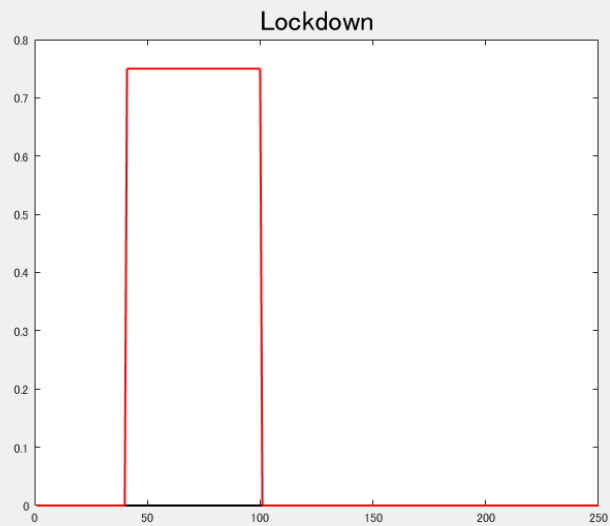
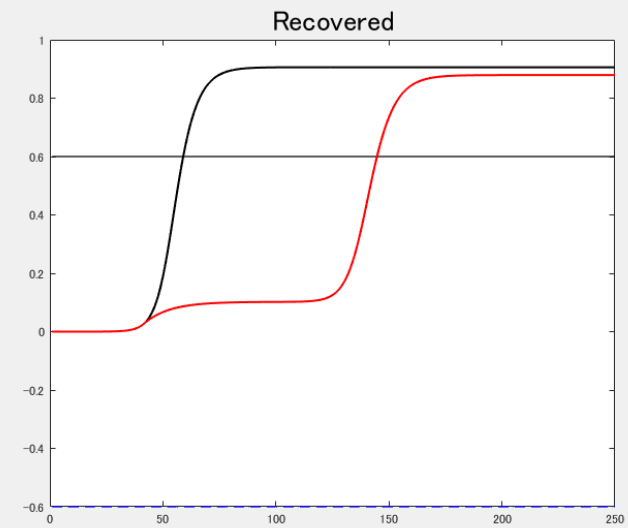
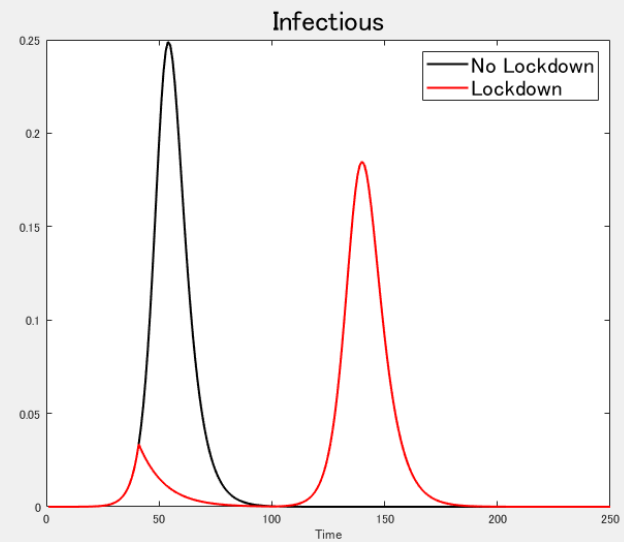
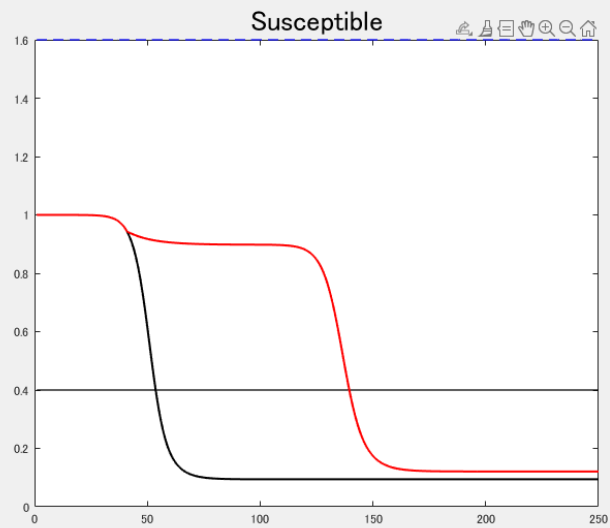
- Suppose that $\alpha_t = 0.5, 0.25, 0.75$ for all $41 \leq t \leq 100$.

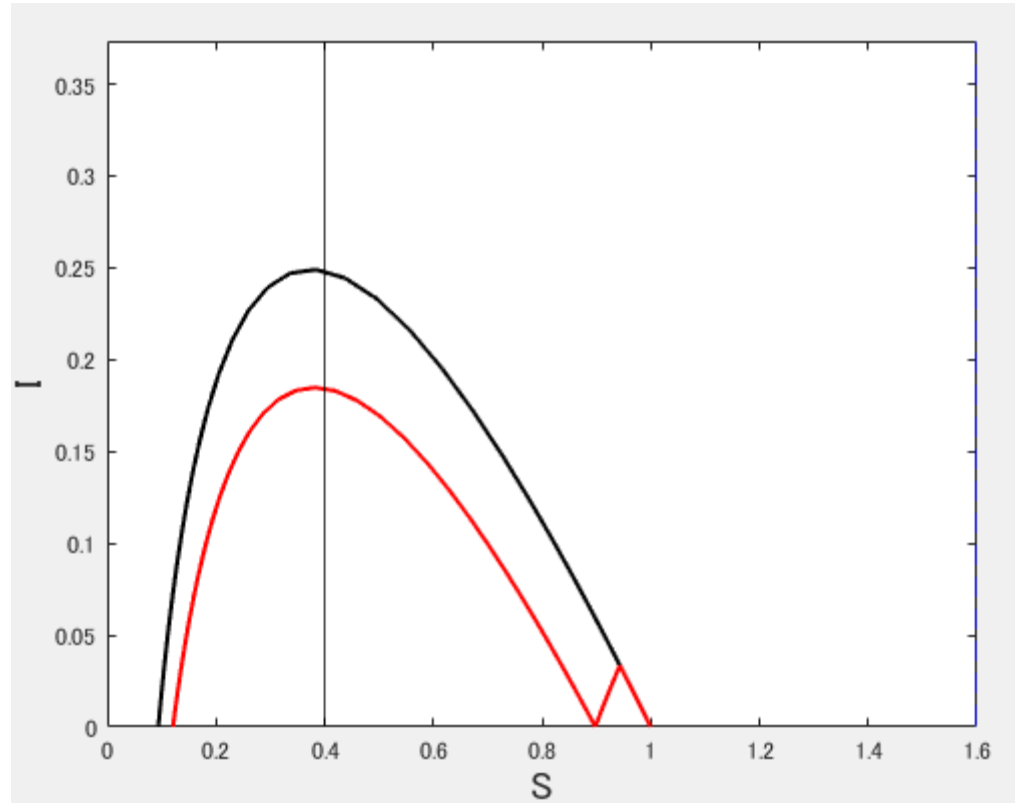












Analyze “welfare.”

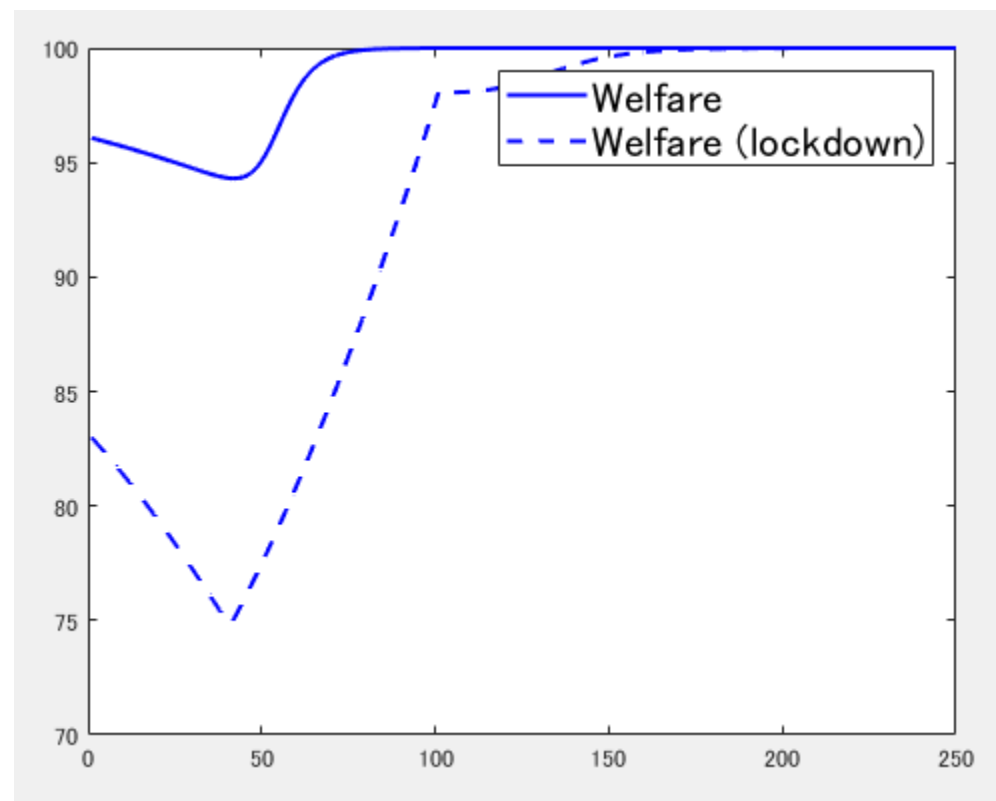
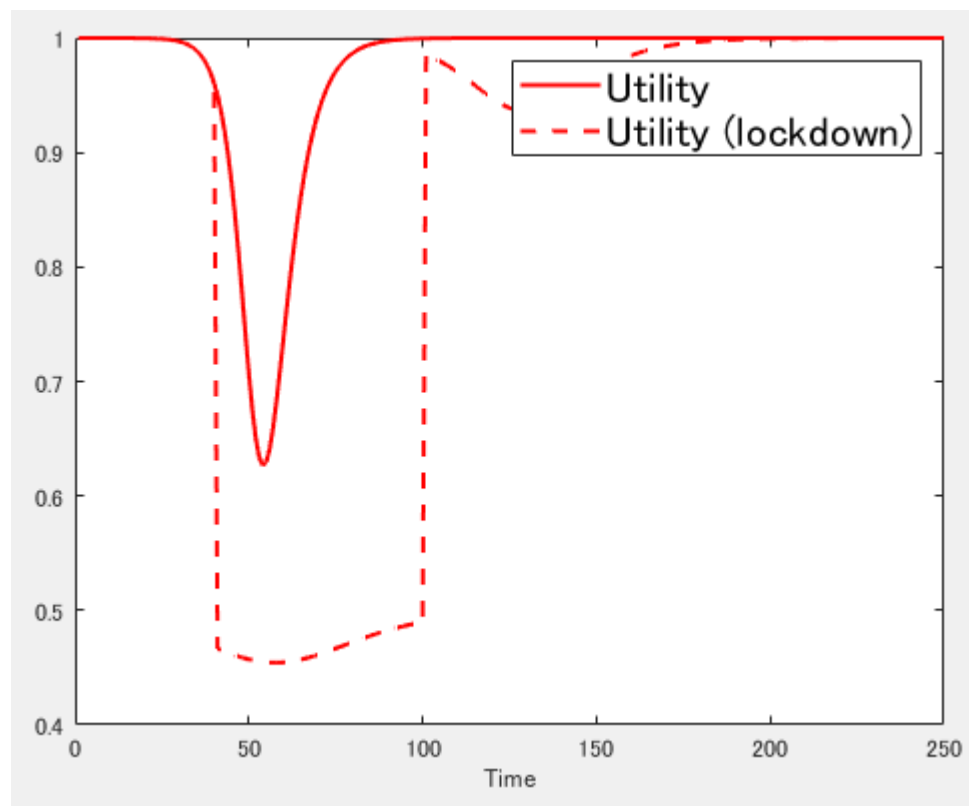
- Suppose that $y_t = (S_t + R_t)(1 - \alpha_t)$. Y_t is output.
- Define a per-period utility at time t as follows.

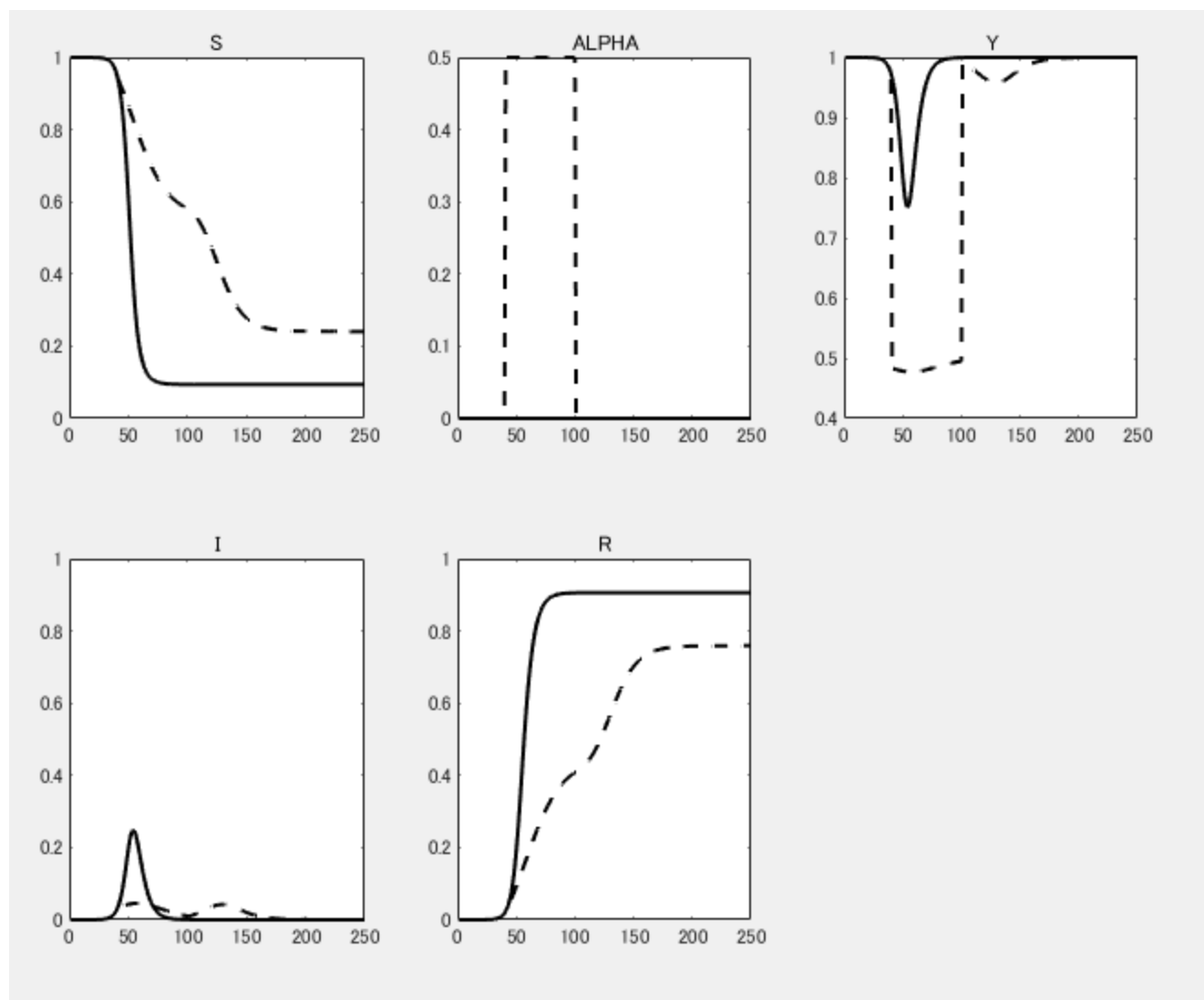
$$U_t = Y_t - \chi I_t$$

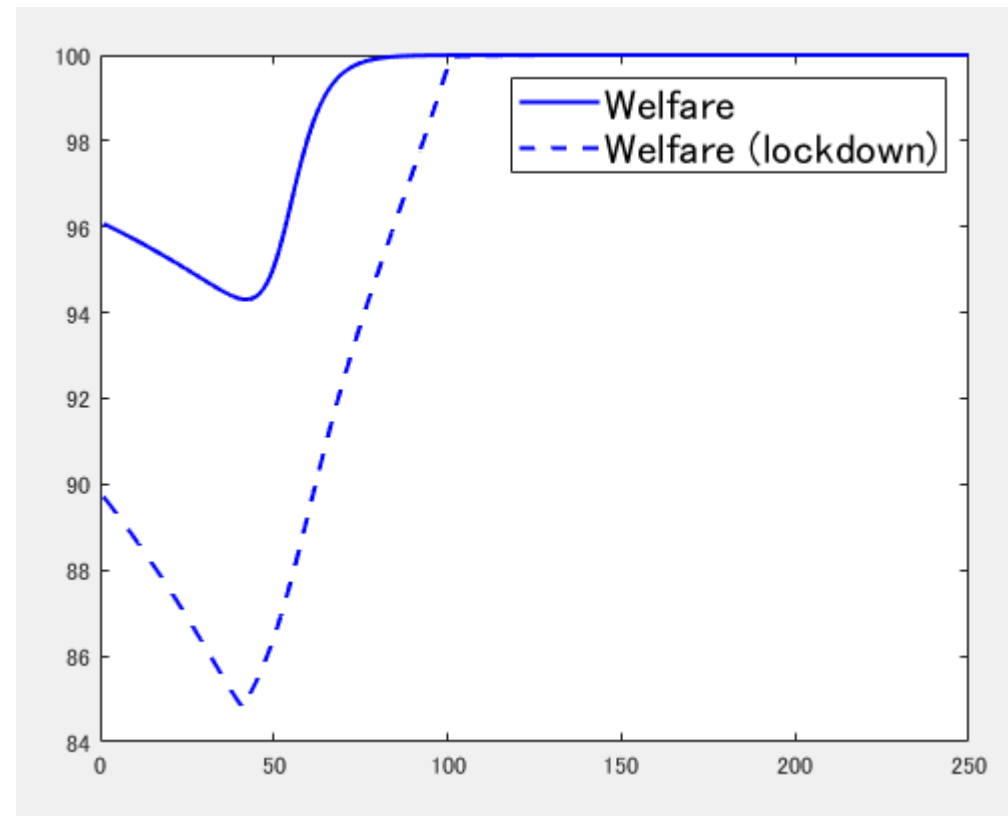
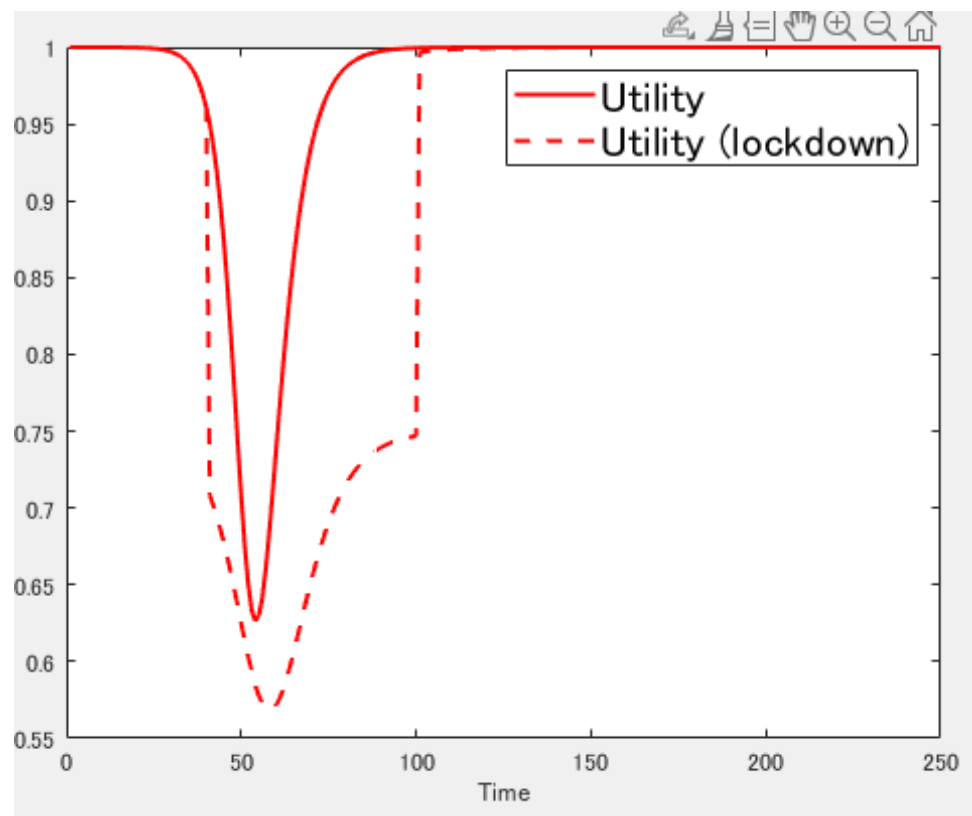
- Define welfare at time t as follows.

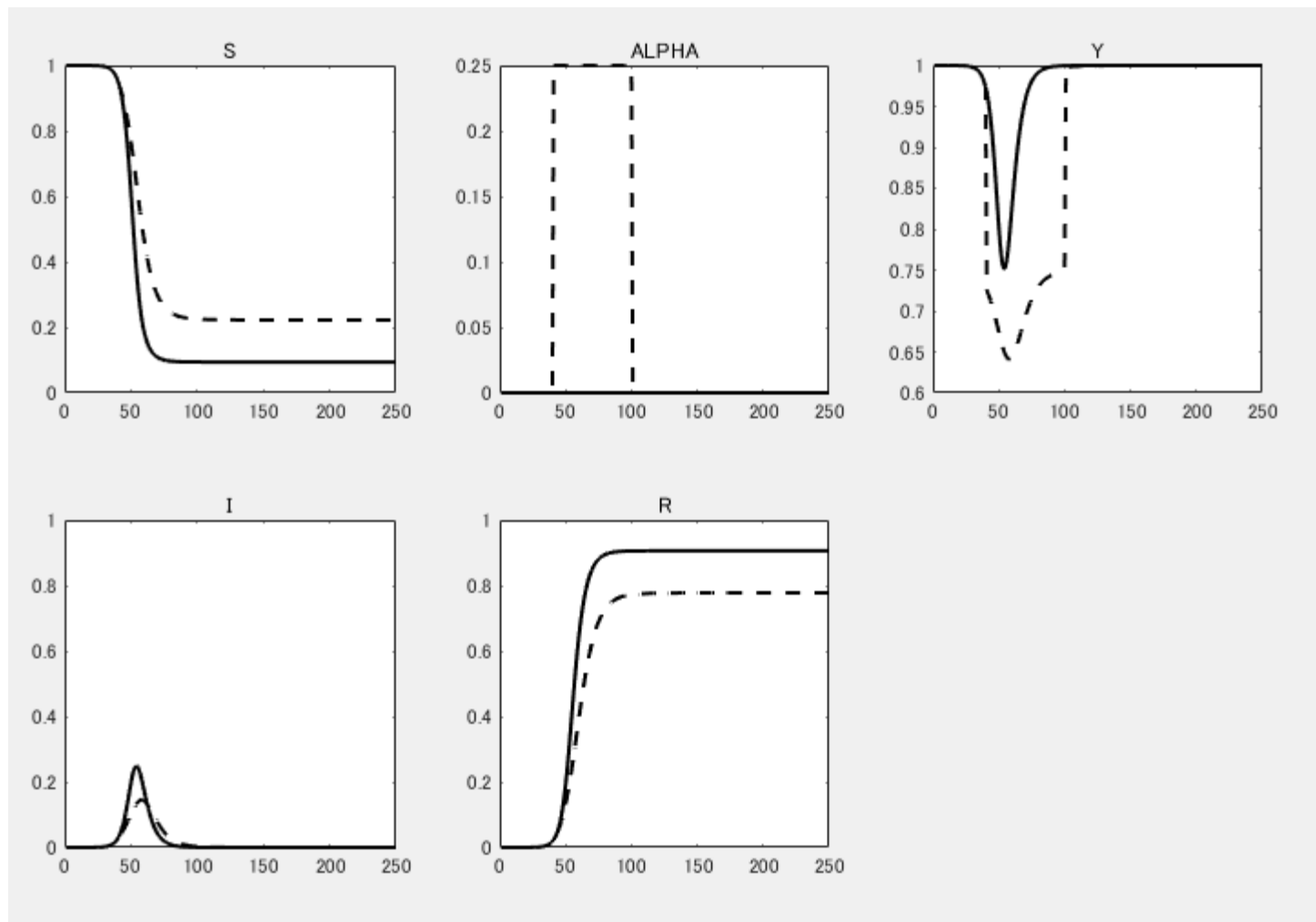
$$\begin{aligned} W_t &= \sum_{k=0}^{\infty} \beta^k U_{t+k} \\ &= U_t + \beta U_{t+1} + \beta^2 U_{t+2} + \beta^3 U_{t+3} + \dots \end{aligned}$$

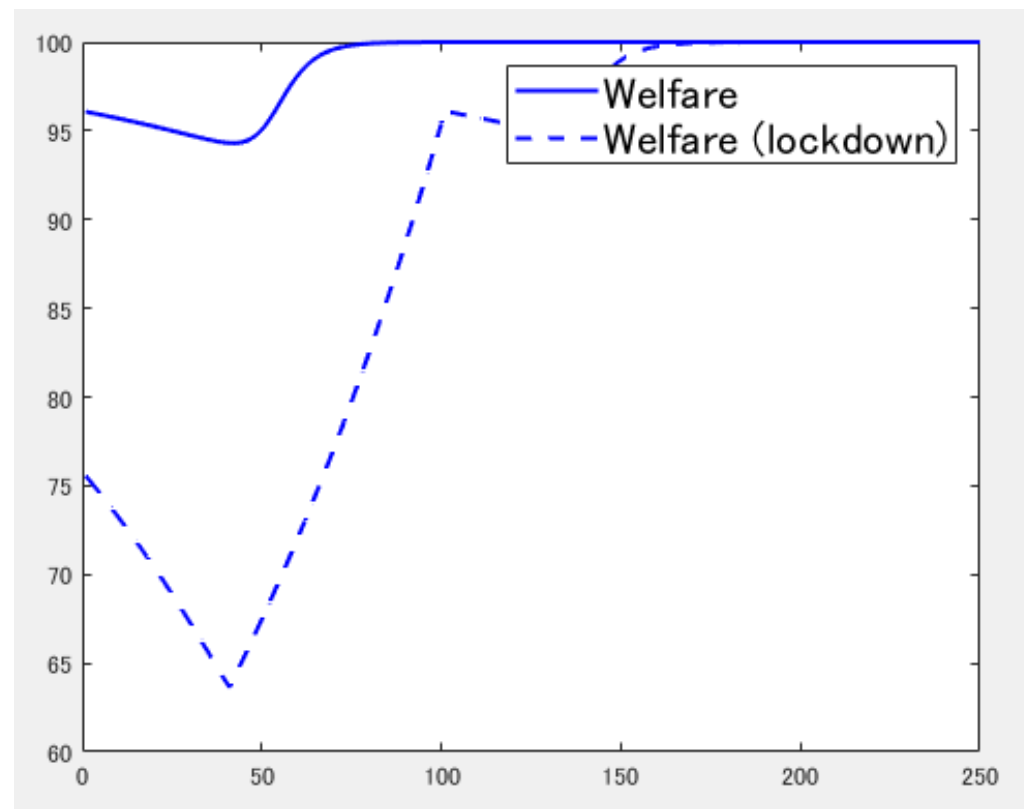
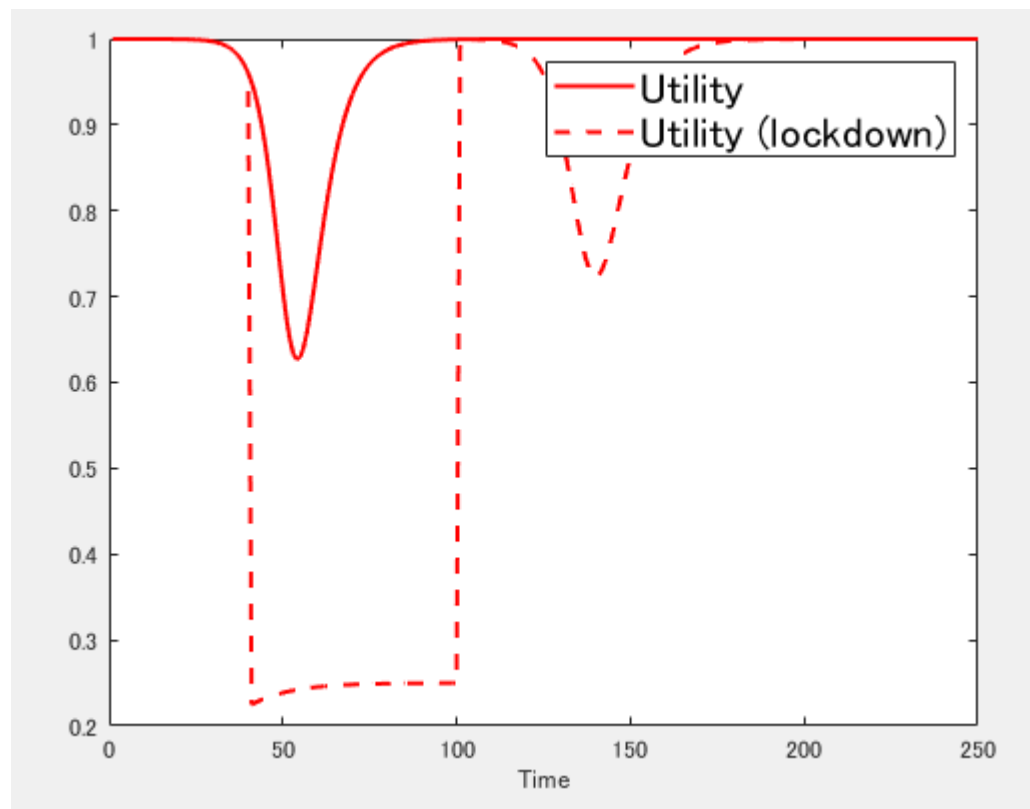
- Suppose that $\alpha_t = 0.5, 0.25, 0.75$ for all $41 \leq t \leq 100$.

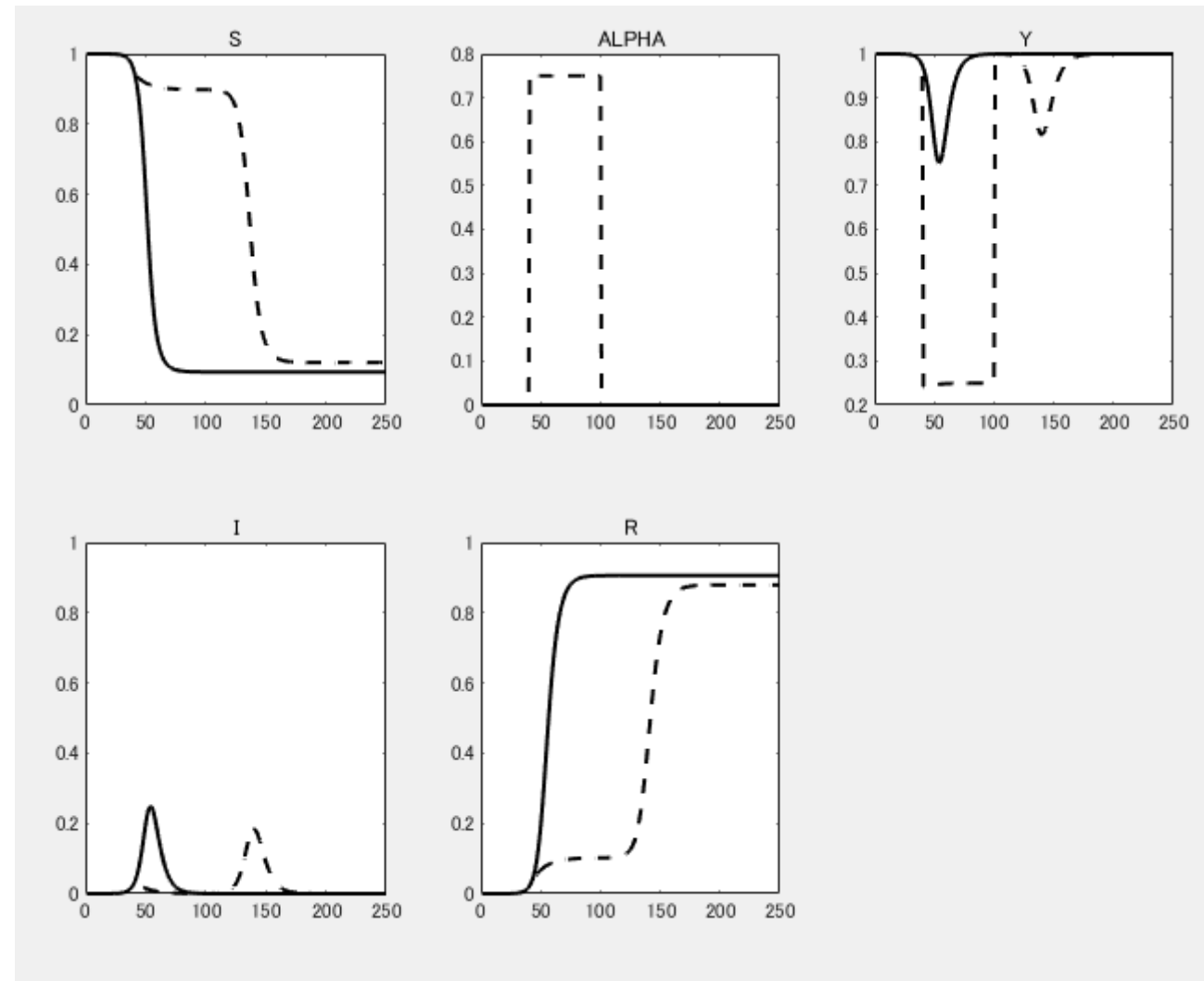










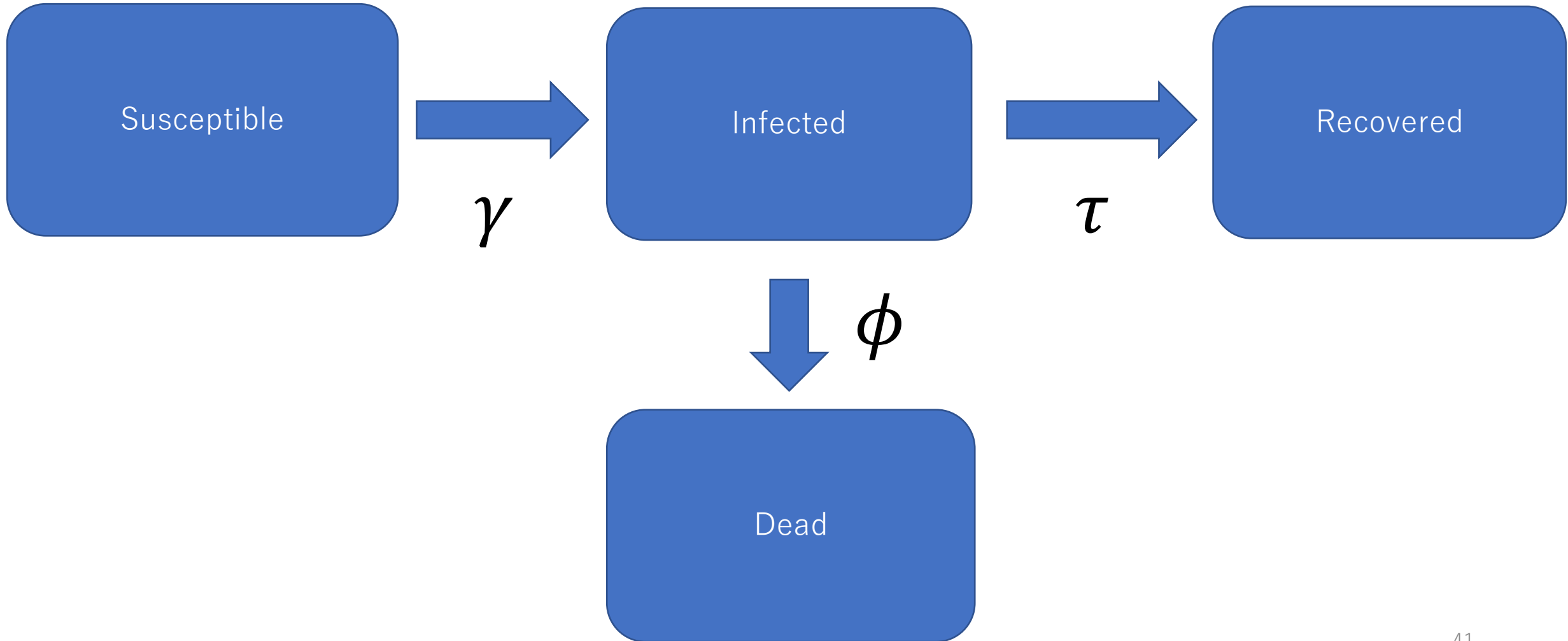


Exercises

1. SIR-D Model

- $S_{t+1} - S_t = -\gamma I_t S_t$
- $I_{t+1} - I_t = \gamma I_t S_t - \tau I_t - \phi I_t$
- $R_{t+1} - R_t = \tau I_t$
- $D_{t+1} - D_t = \phi I_t$
- $S_t + I_t + R_t + D_t = 1$

SIR-D Model

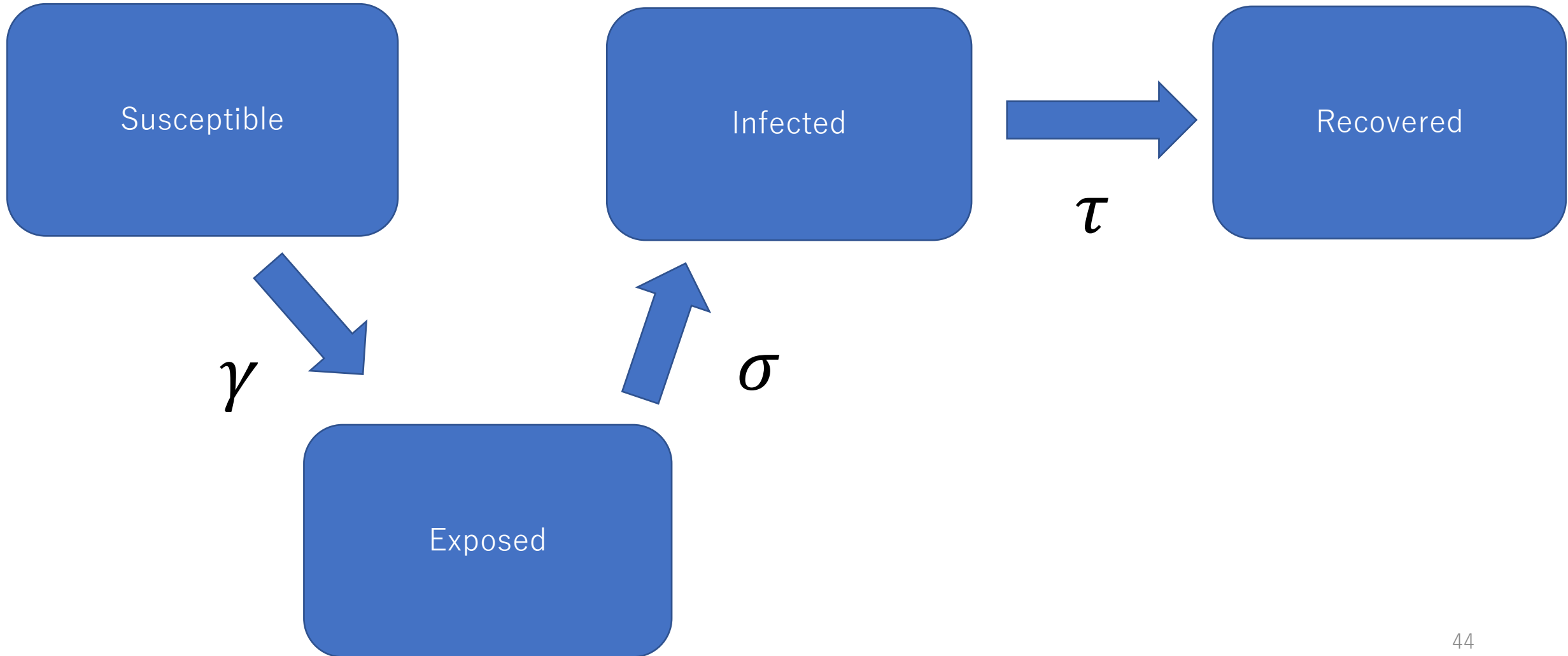


- Examine the steady state of the model.
- Compute convergence to a steady state.
- Analyze the effect of a lockdown policy.
- Analyze the welfare effect of a lockdown policy.

2. SEIR Model

- $S_{t+1} - S_t = -\gamma I_t S_t$
- $E_{t+1} - E_t = \gamma I_t S_t - \sigma E_t$
- $I_{t+1} - I_t = \sigma E_t - \tau I_t$
- $R_{t+1} - R_t = \tau I_t$
- $S_t + E_t + I_t + R_t = 1$

SEIR Model



- Examine the steady state of the model.
- Compute convergence to a steady state.
- Analyze the effect of a lockdown policy.
- Analyze the welfare effect of a lockdown policy.