

5 SIR Macro model

5.1 Introduction

- Research Question: What is the tradeoff between COVID-19 infection and economy?
- Epidemiologists use SIR model to study new coronavirus transmission. Quantitatively calculate how much social activity should be limited. But what's the economics cost?
- SIR Macro model: SIR + DSGE. incentive of preventing behavior, market equilibrium effects, externality. Rapid development since Spring 2020.
- SIR Macro models are used to answer real-time problems.
 - What kind of policies are effective? lockdown? PCR test?
 - Alpha, Delta, Omicron variants?
 - How many vaccines?
- Lecture material
 - Eichenbaum & Rebelo & Trabandt “The Macroeconomics of Epidemics”
<https://doi.org/10.1093/rfs/hhab040>
 - I wrote a note, but only in japanese.
<https://www.ihep.jp/publications/journal/>
- Key concept: Pandemic possibility frontier

5.2 Basic SIR model

- Suppose $t =$ one week.
- 4 groups of people. Total population is 1.

$$\text{Susceptible (not infected yet): } S_{t+1} = S_t - T_t \quad (1)$$

$$\text{Infected (currently): } I_{t+1} = I_t + T_t - (\pi_r + \pi_d)I_t \quad (2)$$

$$\text{Recovered (no more infection): } R_{t+1} = R_t + \pi_r I_t \quad (3)$$

$$\text{Dead: } D_{t+1} = D_t + \pi_d I_t \quad (4)$$

- Infected person recovers with probability π_r and dead with π_d each week.
- New infection

$$T_t = \pi S_t I_t$$

If each person randomly meets another one at time t , the total number meeting of S and I is $S_t I_t$. π is the transmission rate of virus.

- Reproduction Numbers

$$I_{t+1} - I_t = \pi S_t I_t - (\pi_r + \pi_d) I_t$$

$$\text{percent change: } \% \Delta I_{t+1} = \frac{I_{t+1} - I_t}{I_t} = \pi S_t - (\pi_r + \pi_d)$$

- Effective reproduction number:

$$\mathcal{R}_t \equiv \frac{\pi S_t}{\pi_r + \pi_d} \begin{matrix} > \\ = \\ < \end{matrix} 1 \Leftrightarrow \% \Delta I_{t+1} \begin{matrix} > \\ = \\ < \end{matrix} 0$$

- Basic reproduction number

$$\mathcal{R}_0 = \frac{\pi}{\pi_r + \pi_d},$$

that is, \mathcal{R}_t if $S = 1$.

5.3 SIR model with economy, theory

- Uniform lockdown: order L fraction of people to stay at home. Since the government does not identify disease, everyone is randomly locked down.

$$T_t = \pi [S_t(1 - L_t)][I_t(1 - L_t)]$$

- Economic loss is

$$\hat{Y}_t = L_t(S_t + I_t + R_t) + D_t$$

- Targetted lockdown. The government knows who is infected by the PCR test. Test-Tracing-Quarantine (TTQ) policy

$$T_t = \pi S_t [I_t(1 - Q_t)]$$

Economic loss is significantly small. The average length of disease is about two weeks, I is tiny.

$$\hat{Y}_t = L_t I_t + D_t$$

5.4 SIR model with economy, simulation

Simulation is simple because there is no optimization. Assume $I_t = 0.001$ at time 0, and then, repeatedly calculate $t + 1$ period variables from t .

%SIR model for DSGE lecture of Waseda, 2021, So Kubota

```
close all;clc;clear;
```

```
% parameters
```

```
pi_r = (7/18)*(1-0.005);
```

```
pi_d = (7/18)*0.005;
```

```
pi = 0.5852; % 60% will be infected eventually
```

```
% vectors of 100 periods
```

```
periods=100;
```

```
S=ones(periods,1);
```

```
I=zeros(periods,1);
```

```
R=zeros(periods,1);
```

```
D=zeros(periods,1);
```

```
T=zeros(periods,1);
```

```
Rt=zeros(periods,1); % effective reproduction number
```

```
Y=zeros(periods,1); % output loss
```

```
% uniform or targetted lockdowns
```

```
L=zeros(periods,1); % Fraction of lockdown
```

```
Q=zeros(periods,1); % Fraction of quarantine among infected
```

```
L(20:40) = 0.0;
```

```
Q(20:40) = 0.4;
```

```
% initial period
```

```
Rt(1) = pi*S(1)/(pi_r+pi_d); % basic reproduction rate
```

```
I(1) = 0.001; % 0.1% initial infection
```

```
Y(1) = L(1)*(S(1)+R(1)) + (L(1)*(1-Q(1))+Q(1))*I(1) + D(1);
```

```
T(1) = pi * (1-L(1))*S(1) * (1-L(1))*(1-Q(1))*I(1);
```

```
% calculation
```

```
for t=1:periods-1
```

```
    S(t+1) = S(t) - T(t);
```

```
    I(t+1) = I(t) + T(t) - pi_r*I(t) - pi_d*I(t);
```

```
    R(t+1) = R(t) + pi_r*I(t);
```

```
    D(t+1) = D(t) + pi_d*I(t);
```

```
    Y(t+1) = L(t+1)*(S(t+1)+R(t+1)) + (L(t+1)*(1-Q(t+1))+Q(t+1))*I(t+1) + D(t+1);
```

```
    Rt(t+1) = pi*S(t+1)/(pi_r+pi_d);
```

```
    T(t+1) = pi * (1-L(t+1))*S(t+1) * (1-L(t+1))*(1-Q(t+1))*I(t+1);
```

```

end

% plot
SIR_path = [I S R D T Y Rt];
titles = {'Infected', 'Susceptible', 'Recovered', 'Deceased',...
          'New Infection', 'Output loss', 'Effective Reproduction Rate'};
figure;
for x = 1:size(SIR_path,2)
    subplot(2,4,x);
    plot(SIR_path(1:periods, x), 'linewidth', 2);
    title(titles{x});
end

```

5.5 SIR Macro model, theory

- Based on Eichenbaum, Rebelo, and Trabandt (2020)
- We must consider a market equilibrium, instead of the social planner's problem. why?
- Assume infection is through economic activity (consumption) or non-economic activity.

$$T_t = \pi (S_t C_t^s) (I_t C_t^i) \quad (5)$$

C_t^s is susceptibles' total consumption, C_t^i is infecteds'. They are "Big K".

- U_t^s is a susceptible's value function. U_t^i is infected's. Suppose the infection probability as τ_t .
- Susceptible's Bellman equation

$$U_t^s = u(c_t^s, n_t^s) + \beta [(1 - \tau_t)U_{t+1}^s + \tau_t U_{t+1}^i] \quad (6)$$

c_t^s and n_t^s are each susceptible's consumption and labor.

- Why time t is needed? Because the pandemic will eventually end. It is a finite period problem.
- For simplicity, assume

$$u(c, n) = \ln c - \left(\frac{\theta}{2}\right) n^2$$

- Infection probability depends on susceptible's individual consumption c_t^s

$$\tau_t(c_t^s) = \frac{T_t}{S_t} = \pi c_t^s (I_t C_t^i) \quad (7)$$

little k, Big K trick. Each susceptible chooses c_t^s given macroeconomic level variables, including C_t^s . In the equilibrium, $c_t^s = C_t^s$.

- Budget constraint

$$(1 + \mu_t)c_t^s = A n_t^s \quad (8)$$

Assume that the production function is AN . Then the marginal labor productivity is always A , which is equivalent to the wage. Then, $A n_t^s$ is the labor income.

- How to simulate lockdown? Add a consumption tax μ_t . The government pushes down the economic activity. The tax revenue is thrown away to the pacific ocean.
- No Saving! critical assumption.
- Susceptible's problem

$$\begin{aligned} \max_{c_t^s, n_t^s} \quad & U_t^s = \ln c_t^s - \left(\frac{\theta}{2}\right) (n_t^s)^2 + \beta \{[1 - \tau_t(c_t^s)]U_{t+1}^s + \tau_t(c_t^s)U_{t+1}^i\} \\ \text{s.t.} \quad & (1 + \mu_t)c_t^s = An_t^s, \\ & \tau_t(c_t^s) = \pi c_t^s (I_t C_t^i) \end{aligned}$$

- FOC

$$\frac{1}{c_t^s} = (1 + \mu_t) \left(\frac{\theta}{A}\right) n_t^s + \beta(U_{t+1}^s - U_{t+1}^i)\pi(I_t C_t^i) \quad (9)$$

- Infected's problem

$$\begin{aligned} \max_{c_t^i, n_t^i} \quad & U_t^i = \ln c_t^i - \left(\frac{\theta}{2}\right) (n_t^i)^2 + \beta [(1 - \gamma_r - \gamma_d)U_{t+1}^i + \gamma_r U_{t+1}^r + \gamma_d \times 0], \\ \text{s.t.} \quad & (1 + \mu_t)c_t^i = An_t^i \end{aligned} \quad (10)$$

- FOCs are static.

$$n_t^i = \frac{1}{(1 + \mu_t)\sqrt{\theta}} \quad (11)$$

$$c_t^i = \frac{A}{(1 + \mu_t)\sqrt{\theta}} \quad (12)$$

- Recovered problem

$$\begin{aligned} \max_{c_t^r, n_t^r} \quad & U_t^r = \ln c_t^r - \left(\frac{\theta}{2}\right) (n_t^r)^2 + \beta U_{t+1}^r, \\ \text{s.t.} \quad & (1 + \mu_t)c_t^r = An_t^r \end{aligned} \quad (13)$$

- FOCs are the same as (11) and (12). $n_t^r = \frac{1}{(1 + \mu_t)\sqrt{\theta}}$ and $c_t^r = \frac{A}{(1 + \mu_t)\sqrt{\theta}}$
- The system of equations?
- μ_t is exogeneous
- Initial shock is $I_t = 0.001$
- 11 variables

$$S_t, I_t, R_t, D_t, T_t, \tau_t, c_t^s, n_t^s, U_t^s, U_t^i, U_t^r$$

- 11 equations

- Equations (1)-(4) about S, I, R, D
- Budget constarint (8)
- Substituting (12) to (7)

$$\tau_t = \frac{T_t}{S_t} = \pi c_t^s I_t \left(\frac{A}{\sqrt{\theta}}\right)$$

- Rewrite new infection (5): $T_t = S_t \tau_t$
- Substitute (9) to (12)

$$\frac{1}{c_t^s} = (1 + \mu_t) \left(\frac{\theta}{A}\right) n_t^s + \beta(U_{t+1}^s - U_{t+1}^i)\pi I_t \left(\frac{A}{(1 + \mu_t)\sqrt{\theta}}\right)$$

- Three Bellman equations. For infected and recovered, substitute (11) and (12)

$$U_t^s = \ln c_t^s - (\theta/2)(n_t^s)^2 + \beta [(1 - \tau_t)U_{t+1}^s + \tau_t U_{t+1}^i]$$

$$U_t^i = \ln \left(\frac{A}{(1 + \mu_t)\sqrt{\theta}} \right) - \frac{1}{2} + \beta [(1 - \gamma_r - \gamma_d)U_{t+1}^i + \gamma_r U_{t+1}^r]$$

$$U_t^r = \ln \left(\frac{A}{(1 + \mu_t)\sqrt{\theta}} \right) - \frac{1}{2} + \beta U_{t+1}^r$$

5.6 SIR Macro model, simulation

```
// Eichnbaum-Rebelo-Trandbont model
// for DSGE lecture of Waseda, 2021
// based on Krueger-Uhli-Xie
// https://github.com/tjxie/KUX_PandemicMacro
// addpath /Applications/Dynare/4.6.4/matlab
// addpath C:\dynare\4.6.4\matlab

// 0 preparation //////////////////////////////////////
close all;clc; // Do not "clear;"

// 1 variables //////////////////////////////////////

// This program will run the model many times from pi=0 to pi = target one.
// If Dynare runs with true pi from the beginning, it fails to find solution.
// But, if dynare starts from pi=0 and eventually increase pi,
// Dynare uses the last path as the initial guess for each calculation,
// then eventually find the solution under the true pi_target.

#define pi_target = 5e-7 // The true value of pi

// endogenous variables
var          ns          (long_name = 'Labor supply, Susceptible')
            cs          (long_name = 'Consumption, Susceptible')
            tau          (long_name = 'Probability of infection')
            I            (long_name = 'Infected')
            T            (long_name = 'Newly infected')
            S            (long_name = 'Susceptible')
            R            (long_name = 'Recovered')
            D            (long_name = 'Deceased')
            Ui           (long_name = 'Lifetime utility of infected')
            Us           (long_name = 'Lifetime utility of susceptible')
            Ur           (long_name = 'Lifetime utility of recovered');

// exogenous variables
```

```

varexo      eps      (long_name = 'Initial Infection')
           mu      (long_name = 'Consumption Tax');

// 2 Parameters //////////////////////////////////////

// definitions
parameters pi      (long_name = 'Probability of becoming infected through consumption')
           pi_r      (long_name = 'Probability of recovery')
           pi_d      (long_name = 'Probability of death')
           theta      (long_name = 'Labor supply parameter')
           A      (long_name = 'Productivity')
           betta      (long_name = 'Discount factor');

// values
pi_r      = (7/18)*(1-0.005);
pi_d      = (7/18)*0.005;
betta      = 0.96^(1/52);
A      = 39.835;
theta      = 0.001275;
pi      = 0; // pi in the lecture note, it is zero, not true value

// 3 model //////////////////////////////////////
model;

// SIR Equations
S - S(-1) + I - (1 - pi_r - pi_d) * I(-1) = 0;
I = T(-1) + (1 - pi_r - pi_d) * I(-1) + eps;
R = R(-1) + pi_r*I(-1);
D = D(-1) + pi_d*I(-1);

// First-order condition of susceptible
(1/cs) = (1+mu)*(theta/A)*ns + betta * (Us(+1) - Ui(+1)) * pi * (A/((1+mu)*sqrt(theta))) * I;
// Budget constraint of susceptible
(1+mu)*cs = A*ns;
// infection probability of susceptible
tau = pi * cs * (A/((1+mu)*sqrt(theta))) * I;
// total new infection;
T = S * tau;

// value functions
Us = log(cs) - (theta/2)*ns^2 + betta*((1 - tau)*Us(+1) + tau*Ui(+1));
Ui = log(A/((1+mu)*sqrt(theta))) - (1/2)*(1+mu)^(-2) + betta*((1 - pi_d - pi_r) * Ui(+1) + pi_r * Ur(+1));
Ur = log(A/((1+mu)*sqrt(theta))) - (1/2)*(1+mu)^(-2) + betta*Ur(+1);

```

```

end;

// 4 Steady state ////////////////////////////////////////

// only one steady state.
// I do not specify initial & terminal states

steady_state_model;
ns      = 1/sqrt(theta);
cs      = A*ns;
tau     = 0;
I       = 0;
T       = 0;
S       = 1;
R       = 0;
D       = 0;
Us      = (log(cs)-theta/2*ns^2)/(1-beta);
Ur      = Us;
Ui      = (log(cs)-theta/2*ns^2+beta*pi_r*Ur)/(1-beta*(1-pi_d-pi_r));
end;
steady;
check; //model_diagnostics

// 5 simulation ////////////////////////////////////////

shocks;
    var    eps;
    periods 1:1;
    values  0.001;
    var    mu;
    periods 10:30;
    values  0.3;
end;

perfect_foresight_setup(periods = 100);

// loop from pi=0 to pi=pi_target
// Dynare macro
@#for value in 0:2e-8:pi_target
    pi = @{value}; // substitute pi with each iteration value
    perfect_foresight_solver; // Dynare uses the past result as initial guess of Newton method
@#endfor

%% 6 Plot ////////////////////////////////////////

```



```

varmat = [I S R D cs ns];
titles = {'Infected', 'Susceptible', 'Recovered',...
          'Deceased', 'Susceptible Consumption', 'Susceptible Labor'};
figure;
plot_periods=100;
for x = 1:size(varmat,2)
    subplot(2,3,x);
    plot(varmat(2:(plot_periods+1), x), 'linewidth', 2);
    title(titles{x});
end

```