# Quantitative Macroeconomics

Problem Set 8 December 3, 2018

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<sup>&</sup>lt;sup>1</sup>Using the code provided by Luis Rojas.

### Model

### Asssumptions

- Small open economy with stochastic endowment  $y_t$ , which follows a Markov Process.
- The government preferences are given by  $E_0\{\sum \beta^t u(c^t)\}$ .
- Government trades one period bondsith risk neutral investors.
- Timing of the Model
  - 1. Outstanding debt  $b_t$  and the GDP realisation  $y_t$ .
  - 2. Decision to repay debt or default.
  - 3. If there is no default, the government issues new debt at  $q(b_{t+1}, y_t)$
  - 4. If there is default, the government is excluded from the financial markets and can return with a probability  $\lambda$
  - 5. The economy suffers a loss of  $\tau$  fraction of the endowment during default.

### Government's Problem - Value Functions

Value function in the case where government repays the debt.

$$V^{ND}(b,y) = \max_{b'} \{ u(y + q(b',y)' - b) + \beta E\{V(b',y'|y)\} \}$$
 (1)

The value of default is,

$$V^{D}(y) = u((1-\tau)y) + (1-\lambda)\beta E\{V^{D}(y'|y)\} + \lambda \beta E\{V(0,y'|y)\}$$
 (2)

The value function before taking the default decision is given by,

$$V(b,y) = \max_{D \in [0,1]} \{ (1-D)V^{ND}(b,y) + DV^{D}(y) \}$$
(3)

#### Creditors

The Break even condition is given by,

$$q(b',y) = \frac{E\{(1 - D(b',y')|y)\}}{R} \tag{4}$$

Where R is the opportunity cost and the creditors know the function D(b', y').

### Equilibrium

A recursive equilibrium is a price schedule q(b', y), value functions V(b, y),  $V^{ND}(b, y)$ ,  $V^{D}(y)$  and policy functions b'(b, y) and D(b, y) such that,

- 1. Given the price schedule, the value functions and policy functions correspond to the solution of the government problem.
- 2. Creditors break even in expectation.
- 3. In equilibrium the following is satisfied,

$$q(b',y) = \frac{E\{(1 - I\{V^{ND}(b',y',q) \le V^D(y',q)\}|y)\}}{R}$$
(5)

Where, I is an indicator function which is one if there is default.

# Bond price schedule

The bond-price schedule is telling us the set of quantity-price combinations for the public debt given the GDP endowments. The graph 1 shows the bond-price menu for  $\tau = 0.2$  and  $\lambda = 0.5$ , where  $\tau$  is the fraction of the GDP lost if the government defaults and  $\lambda$  is the probability of returning to the financial markets once you've been excluded.

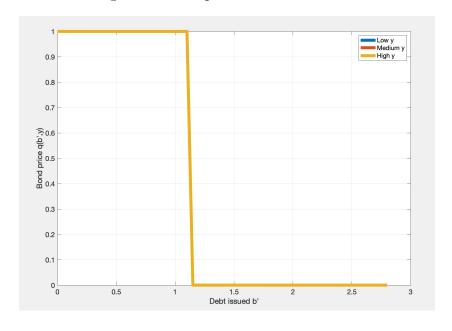


Figure 1: Bond-price menu for  $\tau = 0.2$ 

Now, we change the value of  $\tau$ , assigning different  $\tau$  to different levels of GDP endowments. In other words, we assume that the output cost of default depends on the level of output. Thus, introducing state-contingent losses makes default more responsive to GDP shocks and then, less responsive to outstanding debt.

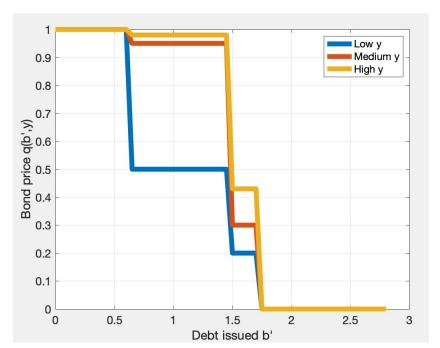


Figure 2: Bond-price menu for different  $\tau = [0.1, 0.4, 0.5]$  and  $y_t$ .

# Equilibrium default

Adjusting the parameters to have that each country is 3% in default, we get a graph with the probability of default at t+1 given the debt-to-gdp ratio.

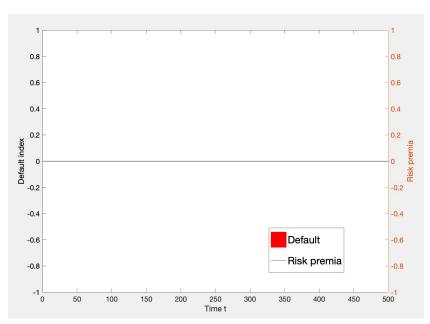


Figure 3: Probability of default for different  $\tau = 0.2$ .

Figure 4: Probability of default for different  $\tau = [0.1, 0.4, 0.5]$ .

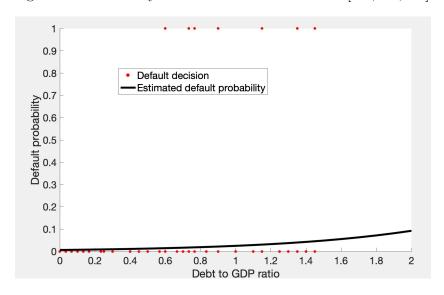
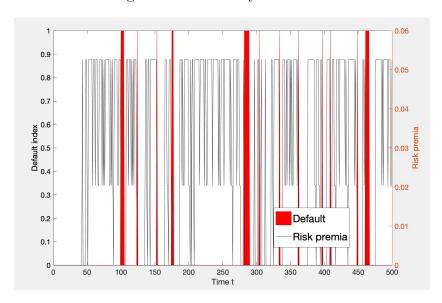


Figure 5: Probability of default.



We find that with a fixed cost of default, the country does not default. Whereas with a different cost of default, there is a higher probability of default. This could be due to the fact that now the cost it too small relative to the GDP endowment. Thus, allowing for greater felxibility when make a decision about default. As discussed before, the more variable the cost of default the higher the elasticity of default.

### Indirect inference

First, we have estimated a logit on the probability of default using the debt-to-gdp as the only regressor and controlling for country fixed effects. By doing so, we got that the coefficient of the  $debt\_to\_gdp$  is equal to 1.15.

We then simulate the model and run the same logit regression over the simulated time series. In that case, the coefficient is 1.96. To make them closer, we change  $\beta$  to 0.91; as a result, the coefficient became 1.12. Graph 5 shows the probability of default out of the simulation.

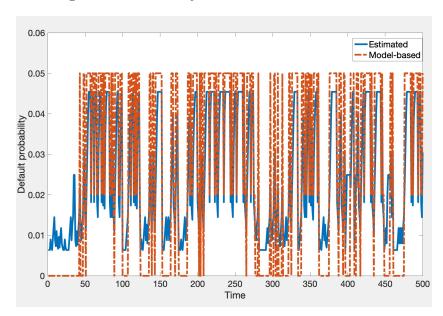


Figure 6: Probability of default of the simulation.

### STATA CODE

xtset Year Ind
xtlogit default debt\_to\_gdp, fe nolog

### MATLAB Code

```
% Code that solves the quantitative sovereign debt model % Problem Set 8 rng(13);  
% Loading parameters  
% Discount factor \theta = 0.95  
betta = 0.91;  
% % Possible values for GDP \theta = 0.91, 1.05
```

```
10
  y_g = [0.6, 1, 1.5];
11
  %%
  % Transition matrix for GDP
  \% $\pi_{yy'}=\left[\begin{array}{ccc}0.5 & 0.3 & 0.2\\0.1 & 0.6 &
      0.3 \setminus 0.2
  \% \& 0.4 \& 0.4 \ \text{end} \{ \text{array} \} \ \text{right} \ ] \$\$
17
  piy = [0.5, 0.3, 0.2; ...]
        0.05, 0.65, 0.3;...
19
        0.02, 0.55, 0.43;
20
  %%
21
  % Risk aversion parameter $\sigma=2$
23
  sig = 1.5;
24
  %%
25
  % Utility function u(c)=\frac{c^{1-\sigma}}{1-\sigma}
27
  u = @(c) c.^(1-sig)./(1-sig);
28
  %%
29
  % Risk-free interest rate $R=1$
  R=1;
32
  %%
  % Probability of regaining access to capital markets next period
  % $\lambda=0.3$ We want that the probability of regaining access is 1
      - the
  % probability of remaining in default (0.03).
  lamda = 0.5;
  %%
  % GDP loss during default. Varies with the level of GDP.
  \%tau = [0.2]
  tau = [0.1, 0.4, 0.5];
  %%
42
  M Initial values and the discretized state space
  % Possible leves of debt issuance \phi = \{0,0.05,0.1,\ldots,0.5\}
46
  B = 0:0.05:2.8;
47
48
  max_iter = 10000;
  \max_{\text{iterg}} = 10000;
51 %
```

```
% Initial values for the value functions:
  %
  % We set the inital guess by assuming that GDP and the level of debt
     do not
  % change over time.
  %
  % Value function of no default
  %
  \% \$V_{0}(b,y) = \frac{u(y)}{1-\beta}
  V = ones(size(B,2),1)*u(y_grid)/(1-betta);
  %%
  % Value function of default
  %
  \%  $${V}_{0}(y)=\frac{u(y)}{1-\beta}$$
66
  Vd = u((1-tau).*y_grid)/(1-betta);
  %
  % Equilibrium computation
  % Our Initial guess for the bond-price schedule is constructed
     assuming the
  % government never defaults q(b',y)=\frac{1}{R}
73
  q=ones(size(B,2), size(y_grid,2));
  M Iterations on the price schedule
75
  for iter_q = 1: max_iterq
77
78
79
  % Value function iterations (taking
                                          as given)
81
82
  for iter=1:max_iter
83
84
      for iy=1:size(y_grid,2)
85
86
  %%
  \% $V_{i+1}(y)=u \left( (1-\tau u)y \right) + (1-\tau u) \left( v \right) 
     v')\mid
  \% y \mid ht \mid + \quad E \mid V_{i}(0,y') \mid y \mid y \mid s
89
90
      Vd(iy)=u((1-tau(iy))*y_grid(iy))+(1-lamda)*betta*piy(iy,:)*Vd'+
91
         lamda*betta*V(1,:)*piy(iy,:)';
```

```
92
            for ib=1: size(B,2)
93
94
        V_{old}=V;
95
        Vd_old=Vd;
96
            V(ib, iy) = \max(\max(u(\max(y_grid(iy)+q(:,iy).*B'-B(ib),0))+betta
                *V*piy(iy,:)'),Vd(iy));
            end
        end
99
100
    dev = max(max(abs([V_old-V;Vd_old-Vd])));
101
   if dev <= 0.001
102
        break
103
   end
104
   end
105
106
   %%
107
   % Updating the bond price menu
108
109
    for iy=1:size(y_grid, 2)
110
        for ib=1: size(B,2)
111
112
            q(ib, iy)=1-piy(iy, :)*(V(ib, :)<=Vd)';
113
        end
114
    end
115
116
   end
117
   % Bond price menu
118
   plot_q(B,q)
   % Simulation of the model
   % recovering the policy function
121
122
    for iy=1:size(y_grid, 2)
123
           for ib=1: size(B,2)
124
           % No default value and debt issuance (conditional on no deault
125
             [V_ND(ib, iy), bp(ib, iy)] = \max(u(\max(y_grid(iy)+q(:,iy).*B'-B(ib)))
126
                ),0)+betta*V*piy(iy,:)')
            % Default decision
127
            Dp(ib, iy) = (V_ND(ib, iy) \le Vd(iy));
128
129
        end
    end
131
  %%
132
```

```
% Simuated sequence of GDP
       starting value (index)
135
136
   yt=1;
137
138
   for t = 2:500
139
        draw_t=rand;
140
        yt(t)=1+(draw_t)=piy(yt(t-1),1)+(draw_t)=sum(piy(yt(t-1),1:2));
141
   end
142
143
   %%
144
   % initial level of debt
145
146
   % index in B
147
148
   bt=ones(500,1);
149
150
   % index for the default decision =1 and the default state
151
152
   Def_b=nan(1,500);
153
   Def_state(500) = 0;
154
155
   for t = 2:500
156
157
       if Def_state(t-1)==0
158
   %%
160
           default decision (decided at t)
161
162
       Def_b(t) = Dp(bt(t-1), yt(t));
163
164
       if \operatorname{Def_b}(t) == 0
165
166
   %%
167
         Debt issuance decision (decided at t)
168
169
170
       bt(t) = bp(bt(t-1), yt(t));
171
172
            Def_state(t)=0;
173
       else
            bt(t) = 1;
175
            Def_state(t)=1;
```

```
end
177
178
       elseif rand<=lamda
179
180
            Def_b(t) = Dp(bt(t-1), yt(t));
181
182
       if \operatorname{Def_-b}(t) == 0
183
   %%
185
            Debt issuance decision (decided at t)
186
187
       bt(t) = bp(bt(t-1), yt(t));
188
       Def_state(t)=0;
189
       else
190
            bt(t) = 1;
191
            Def_state(t)=1;
192
       end
193
194
       else
195
196
            Def_state(t)=1;
197
       bt(t) = 1;
198
       end
199
200
201
   %%
202
         Observed risk spread (1/q-1)
   %
203
204
       r_{spread}(t) = 1/q(bt(t), yt(t)) - 1;
205
   %%
206
   %
          Default probability
207
208
       p_{model}(t)=1-q(bt(t),yt(t));
209
210
   end
211
212
   %%
213
       graph of default and the risk premia
214
215
   risk_premia_graph (Def_state, r_spread)
   % Estimating a logit
   %
       Arranging the data.
218
   %
219
      We have to be sure that if a default spell lasts for more than one
220
```

```
period,
  % we only include the first time default was declared (this is why we
  %
222
  %
      distinguish between the default decision and the default state)
223
  %
224
  %
225
226
  % % Number of observations
227
228
  N=sum(1-(isnan(Def_b)));
  % Regressors. Constant and debt/GDP
230
  X=[B(bt)'./y_grid(yt)'];
232
  % Default decision
233
  Y=Def_b';
234
235
  % Logit regression. Binomial outcome (0, 1)
236
   [par_est, dev, stats] = glmfit(X(1:end-1), Y(2:end), 'binomial');
   disp ('The Coefficient from the Logit')
238
   disp(par_est)
239
240
  % Estimated Default probability
241
   X_{grid} = 0:0.1:2;
   p_{est} = 1./(1 + \exp(-par_{est}(1) - par_{est}(2) * X_{grid}));
   Mean_Def_P = mean(p_est);
   disp ('Probability of Default from the Model')
   disp (Mean_Def_P)
   Fit_{model\_graph}(X(1:end-1), Y(2:end), X_{grid}, p_{est})
247
248
249
  % default probability in the simulation
250
   p_{est\_sim} = 1./(1 + \exp(-par_{est}(1) - par_{est}(2) *X));
251
   Mean_Def_Sim = mean(p_est_sim);
   disp ('Probability of Default from the Simulation')
253
   disp (Mean_Def_Sim)
254
255
  %Graph for the probability of default for the model and the
256
      Simualtion
   sim_and_model_graph([p_est_sim,p_model'])
```