

Computing a simple stochastic OLG example

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Outline

1 Concrete example

2 Computation

3 Results

Now we consider the RCE in a specific example with closed-form solution.

Example

Assume that the capital depreciation rate $\delta = 0$. Let, as before,

$$A_t F(K_t, L_t) = A_t K_t^\alpha N_t^{1-\alpha}, \quad \alpha \in (0, 1),$$

and

$$u(c) = \ln(c),$$

and

$$\ln(A_{t+1}) = (1-\gamma) \ln(A) + \gamma \ln(A_t) + \epsilon_{t+1}, \quad A > 0, \gamma \in (0, 1).$$

Example (continued)

We showed, consumer's optimal intertemporal consumption *contingent plans* satisfy:

$$\frac{1}{c_t^t} = \beta \mathbb{E}_{\mu,t} \left\{ \left[\frac{1 + r_{t+1}}{c_{t+1}^t} \right] \right\}$$

Using consumer's budget constraints, and using capital market clearing condition, this can be re-written as

$$\frac{1}{w_t - (1+n)k_{t+1}} = \beta \mathbb{E}_{\mu,t} \left\{ \left[\frac{1 + r_{t+1}}{(1 + r_{t+1})(1+n)k_{t+1}} \right] \right\}$$

Example (continued)

Simplify RHS to get

$$\frac{1}{w_t - (1+n)k_{t+1}} = \beta \mathbb{E}_{\mu,t} \left\{ \left[\frac{1}{(1+n)k_{t+1}} \right] \right\}$$

so the only stochastic element $r_{t+1} := r_{t+1}(k_{t+1}, A_{t+1})$ dropped out from the first-order condition.

Since k_{t+1} is known at time t , the condition also holds “within” the expectations operator:

$$\frac{1}{w_t - (1+n)k_{t+1}} = \beta \left[\frac{1}{(1+n)k_{t+1}} \right]$$

Example (continued)

Re-arrange for k_{t+1} , we have

$$k_{t+1} = \frac{\beta}{(1+n)(1+\beta)} w_t.$$

Looks just like its deterministic cousin we derived earlier, hey?

Not quite! Now, from firm's optimal labor demand (??), we have

$$w_t = (1-\alpha)A_t k_t^\alpha$$

which suffers from random perturbations by A_t ! So we have a **stochastic difference equation** solution to the RCE of this example:

$$k_{t+1} = \frac{\beta}{(1+n)(1+\beta)} (1-\alpha)A_t k_t^\alpha.$$

Example

The solution to this economy's RCE beginning from (k_0, A_0) is a contingent allocation plan (sequence of decision functions), $(k_{t+1}, c_t, c_{t+1}^t, y_t)(k_t, A_t)$ supported by state-contingent prices $(w_t, r_t)(k_t, A_t)$ that satisfies

- 1 $k_{t+1} = \frac{\beta}{(1+n)(1+\beta)}(1 - \alpha)A_t k_t^\alpha,$
- 2 $w_t = (1 - \alpha)A_t k_t^\alpha,$
- 3 $r_t = \alpha A_t k_t^{\alpha-1} - \delta,$
- 4 $c_t = w_t - (1 + n)k_{t+1},$
- 5 $c_{t+1}^t = (1 + r_{t+1})k_{t+1},$
- 6 $y_t = A_t k_t^\alpha,$
- 7 $i_t = s_t = (1 + n)k_{t+1}$ (investment flow).

for every possible random history of TFP levels, $\{A_t\}_{t=0}^\infty$.

Exercise

- Choose your parameter values $(\alpha, \beta, \delta, A, \gamma, M)$.
- Generate a random sequence $\{A_t\}_{t=0}^{\infty}$ according to the law of motion

$$\ln(A_{t+1}) = (1 - \gamma) \ln(A) + \gamma \ln(A_t) + \epsilon_{t+1} \equiv H(A_t, \epsilon_{t+1}).$$

For example, assume $\epsilon_{t+1} \sim U[-(1 - \gamma) \ln(A), \ln(M)]$.

- To generate a uniformly distributed random variable on an interval $[a, b]$, use the Excel command, $RAND() * (b - a) + a$. [[More on Excel](#)]
- Use Excel or any spreadsheet software to calculate a sample RCE path using these equations:

- 1 $k_{t+1} = \frac{\beta}{(1+n)(1+\beta)} (1 - \alpha) A_t k_t^\alpha \equiv G(A_t, k_t)$,
- 2 $w_t = (1 - \alpha) A_t k_t^\alpha$,
- 3 $r_t = \alpha A_t k_t^{\alpha-1} - \delta$,
- 4 $c_t = w_t - (1 + n)k_{t+1}$,
- 5 $c_{t+1}^t = (1 + r_{t+1})k_{t+1}$,
- 6 $y_t = A_t k_t^\alpha$,
- 7 $i_t = s_t = (1 + n)k_{t+1}$ (investment flow).

Pseudocode

A stylized recipe for computing a sample stochastic RCE path. First generate a random sample stored as a vector, $\mathbf{e} = \{\epsilon_t\}_{t=0}^T$.

Example (Matlab)

If \mathbf{e} is to be a uniformly distributed random vector, use the following command:

```
randyolg.m
23  a = -(1-gamma)*log(A);
24  b = log(M);
25  e = zeros(T+1,1);           % pre-allocate memory
26  e = a + (b-a).*rand(T+1,1); % populate null vector with uniform r.v.'s
27  omega = exp(e);           % take anti-log of e
```

Figure: Generating a vector containing $T + 1$ observations of uniform random variables.

Pseudocode

ALGORITHM 1. Simulating sample RCE outcomes

Input: (A_0, k_0) , Equilibrium system: (H, G) , random sample $\omega \leftarrow \{\exp(\epsilon_t)\}_{t=0}^T$, Null vectors $\mathbf{A}(\cdot) = \mathbf{k}(\cdot) = \mathbf{0}_{(T+1) \times 1}$.

set

| $\mathbf{A}(1) \leftarrow A_0$
 | $\mathbf{k}(1) \leftarrow k_0$
 | $t \leftarrow 1$

end

while $t \leq T$ **do**

| $\mathbf{A}(t+1) \leftarrow H[\mathbf{A}(t), \omega(t+1)]$
 | $\mathbf{k}(t+1) \leftarrow G[\mathbf{A}(t), \mathbf{k}(t)]$
 | **set**
 | | $t \leftarrow t+1$
 | **end**

Output: Random sample RCE path $\{k_{t+1}\}_{t=0}^T \leftarrow \mathbf{k}$

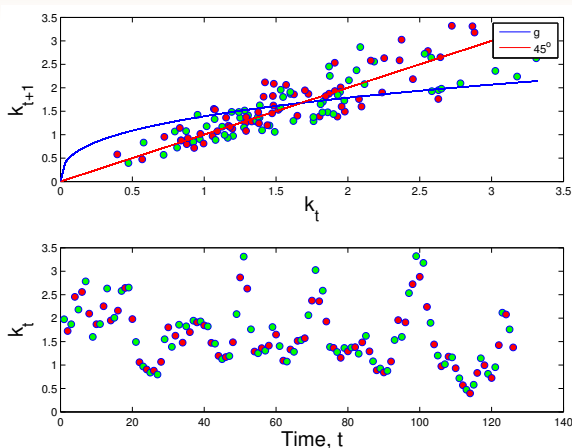


Figure: Sample RCE path for capital now appears as random fluctuations around the deterministic steady state where $k_{t+1} = g(A_t, k_t)$, because A_t is a stochastic process.

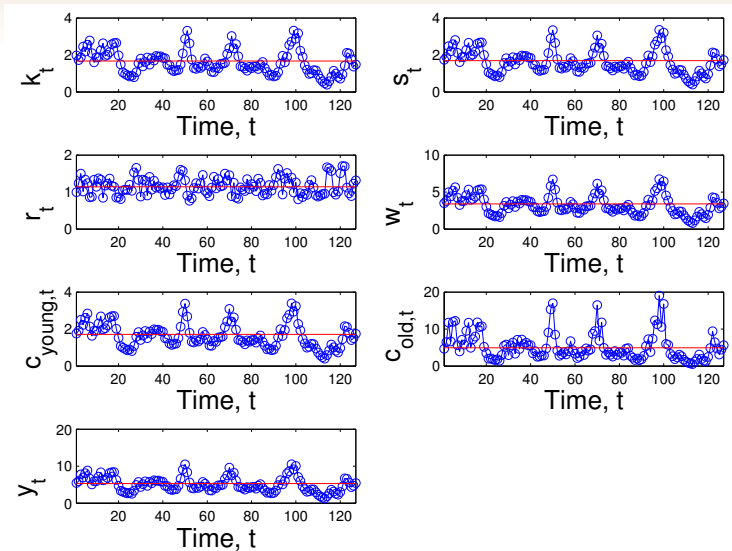


Figure: Sample RCE path for other variables which are functions of (A_t, k_t) .

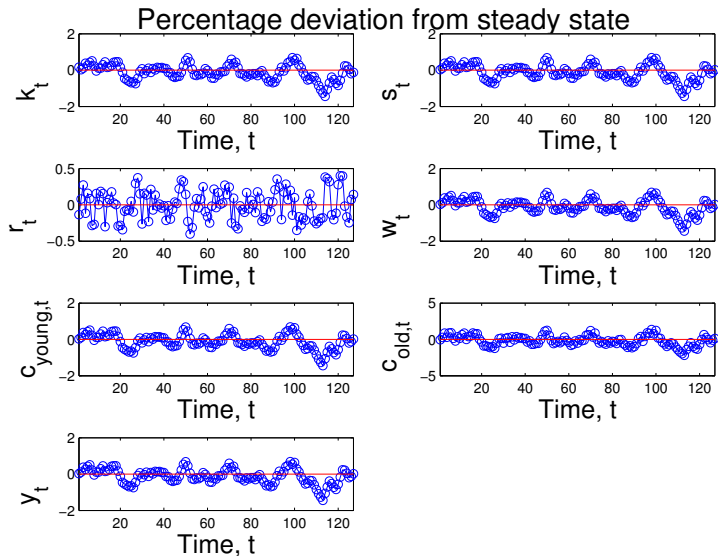


Figure: Sample RCE path for same variables transformed as percentage deviations from respective steady state values. E.g. $\ln(y_t/y_{ss})$.

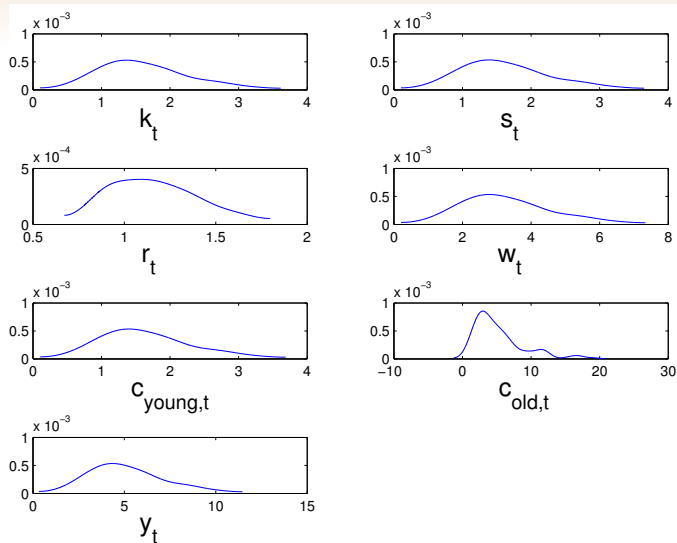


Figure: Sample distribution of RCE outcomes over time. By law of large numbers this approaches the stationary distribution (with probability 1).

Table: Mean for simulated data and steady state values

| Variable | simulated | theory |
|---------------|-----------|--------|
| s_t | 1.632 | 1.693 |
| w_t | 3.280 | 3.402 |
| $c_{young,t}$ | 1.648 | 1.710 |
| y_t | 5.125 | 5.316 |

Notice that the simulated RCE means are quite close to the theoretically calculated steady state values? If I increase the simulation sample observations, they should be the same, w.p.1.

Remark

- 1 *Example illustrates techniques in stochastic modeling and a simple economic (OLG) theory for generating theory consistent “business cycles”.*
- 2 *Should we take this model seriously as a quantitative model of real business cycles?*
- 3 *Not quite. OLG model designed for long run analysis. Notion of a “period t ” is very long. At business cycle frequencies, a period t usually is one quarter (3 months).*
- 4 *Mismatch between theory and measurement. Need a model with more refined notion of generations. Or more generally, assume infinitely lived agents.*
- 5 *Given calibration of parameters, model does not generate realistic business cycle facts*

- In principle, as we saw, the model implies an equilibrium statistical process.
- We saw the sample distribution of the model variables.
- We can use these simulated data to calculate the relevant business cycle statistics – e.g. standard deviations, means, correlations with output (measure of procyclicality), etc.
- But this model is no good, quantitatively.

Table: Standard deviations for simulated data

| Variable | std |
|---------------|-------|
| i_t | 0.635 |
| w_t | 1.277 |
| $c_{young,t}$ | 0.642 |
| y_t | 1.995 |