# Math Tools: Recursive Methods

Revised: November 24, 2014

The concept of recursion runs throughout modern mathematics and computer science. The same is true of economics and finance. There's a reason the leading PhD textbook in macroeconomics is called *Recursive Macroeconomic Theory*. Such work in economics reflects, in large part, the adoption by economists of methods developed elsewhere.

What follows is a short informal introduction to the idea and a start on the kinds of applications you'll find in economics and finance.

Warning: This contains a little linear algebra. Skip it if that's not part of your skill set.

## 1 Examples of recursion

The idea is to characterize a sequence of items, indexed by an integer n = 0, 1, 2, ..., by a rule that connects each item to the next one. If we label the items  $x_n$ , the rule might be expressed

$$x_{n+1} = g(x_n). (1)$$

If we have a starting point, say  $x_0$ , the rule tells us how to compute as many succeeding items as we wish. We would say that the set  $\{x_n\}$  is generated recursively and refer to (1) as the defining recurrence relation. A "solution" to (1) is a formula that expresses  $x_n$  as a function of n.

### Examples:

1. Linear difference equation. Let

$$x_{n+1} = ax_n. (2)$$

This has the solution  $x_n = a^n x_0$ . It converges to zero if |a| < 1, but it's the solution either way.

2. Logistic map. Let

$$x_{n+1} = ax_n(1-x_n)$$

with  $0 < a \le 4$ . If you try some experiments, you'll see that it generates wildly different behavior depending on the value of a. You might set  $x_0 = 0.3$  and a = (0.98, 1.5, 2.5, 3.25, 3.5), generate (say) 20 terms, and graph the output. See Wikipedia. The point, which we won't develop further, is that even quite simple nonlinear recurrences can generate complex behavior.

3. Fibonacci numbers. The Fibonacci numbers are generated by the second order system

$$f_{n+1} = f_n + f_{n-1}$$

starting with  $f_0 = 0$  and  $f_1 = 1$ . In matrix terms, we can write this as  $x_{n+1} = Ax_n$  with

$$x_n = \begin{bmatrix} f_n \\ f_{n-1} \end{bmatrix}, A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}.$$

This is similar to (2), but here  $x_n$  is a vector. The matrix A has eigenvalues  $(\lambda_1, \lambda_2)$  satisfying  $\lambda^2 - \lambda - 1 = 0$ . The "solution" has the form  $f_n = c_1 \lambda_1^n + c_2 \lambda_2^n$  for constants  $(c_1, c_2)$  that satisfy the initial conditions.

This is a common example in computer science courses. A recursive version of a Matlab program to compute Fibonacci numbers is

```
function answer = f(n)
if n==0
    answer = 0;
elseif n==1
    answer = 1;
else
    answer = f(n-1) + f(n-2);
end
end
```

Note that the function f refers to itself — it's recursive in the sense the word is used in computer science. In Matlab, we would save this as a file called f.m, then call it by typing (say) f(8) in the command line or as a line in another program. (If you enter a fraction, it blows up, so a better function would check and generate an error message, or perhaps convert n to an integer.)

4. Mean and variance. John Cook describes the Welford method of computing the mean and variance recursively. Consider a sequence of observations:  $x_1, x_2, \ldots$  We can compute rolling estimates of the mean and variance from

$$M_n = M_{n-1} + (x_n - M_{n-1})/n$$
  
 $S_n = S_{n-1} + (x_n - M_{n-1}) * (x_n - M_n)$ 

starting with  $M_1 = x_1$  and  $S_1 = 0$ . Do a few terms to assure yourself that  $M_n$  is the mean of the first n observations and  $S_n$  is the sum of the squared deviations from the mean. The standard estimator of the variance  $s^2$  is therefore  $S_n/(n-1)$  (although I prefer to divide by n, always).

5. Natural numbers. The natural numbers are the set  $\mathbb{N} = \{0, 1, 2, ...\}$ . We can define them recursively with the rules: (i) 0 is in  $\mathbb{N}$  and (ii) if n is in  $\mathbb{N}$  then so is n + 1.

Which reminds me of an old George Gamov story. In the story, the Hilbert Hotel has an infinite number of rooms numbered 1, 2, 3, . . . . By law, it must save one for the King, but the innkeeper fills them all anyway. When asked, he says: "No problem, I can always get

an open room by asking everyone to move over one." And if we move the person in room j to room 2j, we open up an infinite (countable) number of rooms, all the odd-numbered ones.

6. Combinatorics. Computer scientist Herbert Wilf notes that many combinatoric identities satisfy recurrences. For example, the binomial coefficients,

$$f(n,k) = {n \choose k} = \frac{n!}{k!(n-k)!},$$

are the solution to

$$f(n,k) = f(n-1,k) + f(n-1,k-1)$$

starting with f(n,0) = 1.

7. Functions. In economics and finance we often run across recursion with functions. Suppose we have a sequence of functions  $f_n(x)$  over some domain x that satisfy the recurrence

$$f_{n+1}(x) = g[f_n(x)]$$

for some g. Even better, suppose the recurrence has a fixed point:

$$f(x) = g[f(x)].$$

Here we have an equation in which the unknown is another function f rather than a number x. It's similar to (1), but we're dealing with a more complex object.

### 2 Recursion in asset pricing

Similar methods show up throughout economics and finance. The idea is to string together a series of one-period steps — recurrences — similar to those we used earlier in the course. That allows us to approach the price of (say) an n-period bond with the same methods we used to price a one-period bond.

We'll do all this in Markov settings, which require some notation. You'll recall that modern asset pricing is based on the no-arbitrage theorem: there exists a positive pricing kernel m that satisfies E(mr) = 1 for returns r on all assets. In a Markov environment, we need to keep track of the current state  $z_t$  and the possible future states  $z_{t+1}$ . The ingredients include:

- Probabilities. We have a state variable  $z_t$  and conditional probabilities  $p(z_{t+1}|z_t)$ .
- Returns. One-period returns from date t to t+1 depend on the state in both periods:  $r(z, z_{t+1})$ .
- Asset pricing. The no-arbitrage theorem becomes: there exists a positive  $m(z_t, z_{t+1})$  satisfying

$$E_t\big[m(z_t, z_{t+1})r(z_t, z_{t+1})\big] = 1$$

for all returns  $r(z_t, z_{t+1})$ . Here  $E_t$  is the expectation conditional on the current state  $z_t$ —the expectation computed from  $p(z_{t+1}|z_t)$ , in other words.

## Examples:

1. Bond pricing. A bond of maturity n is a claim to a payment of one in n periods. In a Markov setting, such bond prices are functions of the state. The question is what the functions are. We find bond prices recursively, starting with  $q^0(z_t) = 1$  for all states  $z_t$  (a dollar today is worth a dollar). Bonds of longer maturity follow from the recursion

$$q^{n+1}(z_t) = E_t [m(z_t, z_{t+1})q^n(z_{t+1})].$$
(3)

In words: an n + 1-period bond is a claim to an n-period bond in one period.

We'll spend some time with a loglinear functional form. This takes some work, but it's worth doing because we'll be spending some time with similar models. Despite how it might look at first, this is a user-friendly functional form. Suppose the pricing kernel is loglinear:

$$\log m(z_t, z_{t+1}) = \delta + az_t + bz_{t+1}$$
$$z_{t+1} = \varphi z_t + \sigma w_{t+1}$$

with  $\{w_t\}$  a sequence of independent standard normal random variables and  $0 < \varphi < 1$ . Then bond prices are loglinear functions of the state:

$$\log q^n(z_t) = A_n + B_n z_t \tag{4}$$

for coefficients  $(A_n, B_n)$  to be determined.

The solution follows from applying (3) to (4). We start with

$$\log q^{n+1}(z_t) = \log E_t \left\{ \exp \left[ \log m(z_t, z_{t+1}) + \log q^n(z_{t+1}) \right] \right\}.$$

We get the left side from (4). The right side takes some work. The inside of the square brackets on the right can be expressed

$$\log m(z_t, z_{t+1}) + \log q^n(z_{t+1}) = (\delta + az_t + bz_{t+1}) + (A_n + B_n z_{t+1})$$
$$= \delta + A_n + az_t + (b + B_n)(\varphi z_t + \sigma w_{t+1}).$$

Conditional on the state  $z_t$ , this is normal with mean and variance

$$E_t [\log m(z_t, z_{t+1}) + \log q^n(z_{t+1})] = \delta + A_n + [a + (b + B_n)\varphi]z_t$$
  
Var<sub>t</sub> [log  $m(z_t, z_{t+1}) + \log q^n(z_{t+1})$ ] =  $(b + B_n)^2 \sigma^2$ .

The usual "mean plus variance over two" gives us

$$\log E_t \left( m(z_t, z_{t+1}) q^{n+1}(z_{t+1}) \right) = \log E_t \left\{ \exp \left[ \log m(z_t, z_{t+1}) + \log q^{n+1}(z_{t+1}) \right] \right\}$$
$$= \delta + A_n + \left[ a + (b + B_n) \varphi \right] z_t + (b + B_n)^2 \sigma^2 / 2.$$

[If this isn't clear, go through it again, it's important.] By assumption, this equals  $A_{n+1} + B_{n+1}z_t$  for all values of  $z_t$ , so we must have

$$A_{n+1} = \delta + A_n + (b+B_n)^2 \sigma^2 / 2$$
  

$$B_{n+1} = a + (b+B_n)\varphi.$$

Evidently we've converted the recursion in  $q^n(z_t)$ , equation (3), into recursions in the coefficients  $(A_n, B_n)$ . They're not pretty, but we can easily compute them. The initial conditions  $A_0 = B_0 = 0$  correspond to  $\log q^0(z_t) = \log(1) = 0$ .

2. Equity pricing. A dividend paying stock is a more complicated object. In the same environment as before, let the dividend in state  $z_t$  be  $d(z_t)$ . The ex-dividend value of a share might be expressed recursively as

$$v(z_t) = E_t \{ m(z_t, z_{t+1}) [d(z_{t+1}) + v(z_{t+1})] \}.$$
 (5)

In words: equity today is a claim to two things tomorrow, a dividend and the same share of equity.

Note that the unknown in this equation is the function v. It's also recursive: you need to know v on the right to compute v on the left. You're now as ready as you'll ever be to understand the recursion joke: "To understand recursion, you need to understand recursion." Or Google "recursion." You get back: "Did you mean: recursion?"

One way to think about this is as the limit of a finite horizon. Suppose we value next period's dividend by

$$v^{1}(z_{t}) = E_{t}\{m(z_{t}, z_{t+1})d(z_{t+1})\}.$$

The superscript 1 here means we're valuing one period of dividends. We can value two periods of dividends recursively with

$$v^{2}(z_{t}) = E_{t} \{ m(z_{t}, z_{t+1}) [d(z_{t+1}) + v^{1}(z_{t+1})] \}.$$

In general, we can value n+1 periods of dividends with the recursion

$$v^{n+1}(z_t) = E_t \{ m(z_t, z_{t+1}) [d(z_{t+1}) + v^n(z_{t+1})] \},$$

starting with  $v^0(z_t) = 0$  (the value of zero dividends is zero). As we increase n, we have more and more dividends. We might imagine, if all goes well, that as n gets larger and larger, we approach (5).

3. Perpetual options. Consider the option to buy one share of stock next period for strike price k. The value today in state  $z_t$  is

$$q(z_t) = E_t \{ m(z_t, z_{t+1}) [v(z_{t+1}) - k]^+ \},$$

where  $x^+ = \max\{0, x\}$ . Evidently we exercise in states where  $v(z_{t+1}) - k$  is positive and not in other states.

A perpetual option allows us to wait: if we don't exercise now we can hold the option for another period, and do this again, forever. Valuation has a recursive form:

$$q(z_t) = \max \{v(z_t) - k, E_t[m(z_t, z_{t+1})q(z_{t+1})]\}.$$

That is: we either exercise now and get  $v(z_t) - k$  (the first branch of the max) or continue to hold the option and get the current value of the option next period (the second branch).

### 3 Bottom line

Recursive methods are at the heart of modern macroeconomics and finance. We'll use them extensively to value bonds.

# Practice problems

- 1. Discounted cash flows. Our goal here is to simplify the pricing relation (5) and derive a more conventional valuation of equity as the expected discounted value of future dividends.
  - (a) Simplify (5) using  $m(z_t, z_{t+1}) = \delta$  and replacing dependence on the state  $z_t$  with a subscript t that is, by replacing  $v(z_t)$  with  $v_t$ .
  - (b) Use this simplification to express equity's value as

$$v_t = \sum_{j=1}^n \delta^j E_t(d_{t+j}) + \delta^n E_t(v_{t+n}).$$

- (c) What happens as n gets large? What happens to the second term on the right above?
- (d) What does this example leave out that's present in (5)? Answer.
- (a) Equation (5) becomes

$$v_t = \delta E_t (d_{t+1} + v_{t+1}).$$

- (b) This follows from repeated substitution and the law of iterated expectations.
- (c) We hope that the second term goes to zero. That leaves us with the infinite sum

$$v_t = \sum_{j=1}^{\infty} \delta^j E_t(d_{t+j}).$$

- (d) If  $m(z_t, z_{t+1}) = \delta$  there are no risk premiums. The price of equity depends only on expected future dividends.
- 2. Consols. A consol pays a constant coupon c every period forever. They have been used by the British as a government financing tool since the 1700s. How would you adapt equation (5) to value such an instrument?

Answer. We set  $d(z_t) = c$ , giving us

$$v(z_t) = E_t \{ m(z_t, z_{t+1}) [c + v(z_{t+1})] \}.$$

The variation in price here comes from m, which translates roughly as variation in interest rates. This component is in equity, too, as well as variation in future dividends.