

§ 2.3 Conditional Expectation

Let $\mathbb{E}^{\mathbb{P}}$ refer to expectation in the finite probability space (Ω, \mathbb{P}) , and let X be a random variable on (Ω, \mathbb{P}) . Then we define $\mathbb{E}_n^{\mathbb{P}}[X] = \mathbb{E}^{\mathbb{P}}[X|A] = \frac{1}{\mathbb{P}(A)} \sum_{\omega \in A} X(\omega) \mathbb{P}(\omega)$.



WHAT MY OLD-FASHIONED PAL IS TRYING TO SAY IS THAT $\mathbb{E}_n^{\mathbb{P}}[X]$ IS THE EXPECTED VALUE OF X BASED ON THE INFO AVAILABLE AT TIME n .

Listen to Deadpool.
Deadpool is smart.

Fundamental Properties of Conditional Expectation

Let X, Y be R.V.'s, let $N \in \mathbb{Z}^+$, let $0 \leq n \leq N$.

(1) Linearity: $\forall c_1, c_2 \in \mathbb{R}, \mathbb{E}_n[c_1 X + c_2 Y] = c_1 \mathbb{E}_n[X] + c_2 \mathbb{E}_n[Y]$.

(2) Taking out what's known: If X only depends on the first n tosses, $\mathbb{E}_n[X Y] = X \mathbb{E}_n[Y]$.

(3) Iterated Conditioning: If $0 \leq n < m \leq N$, then $\mathbb{E}_n[\mathbb{E}_m[X]] = \mathbb{E}_n[X]$.

(4) Independence: If X depends only on tosses $n+1$ through N , then $\mathbb{E}_n[X] = \mathbb{E}[X]$.

(5) Conditional Jensen's Inequality: If φ is convex, $\mathbb{E}_n[\varphi(X)] \geq \varphi(\mathbb{E}_n[X])$.

Notation

From backward induction, we know

$$S_n(\omega_1, \dots, \omega_n) = \frac{1}{1+r} \left[\tilde{p} S_{n+1}(\omega_1, \dots, \omega_n, H) + \tilde{q} S_{n+1}(\omega_1, \dots, \omega_n, T) \right]$$

$$= \frac{1}{1+r} \tilde{\mathbb{E}}_n [S_{n+1}] (\omega_1, \dots, \omega_n) \quad (*)$$

So we can write

$$S_n = \frac{1}{1+r} \tilde{\mathbb{E}}_n [S_{n+1}] \quad (**)$$

Since (*) is true for all $(\omega_1, \dots, \omega_n) \in \{H, T\}^n$.

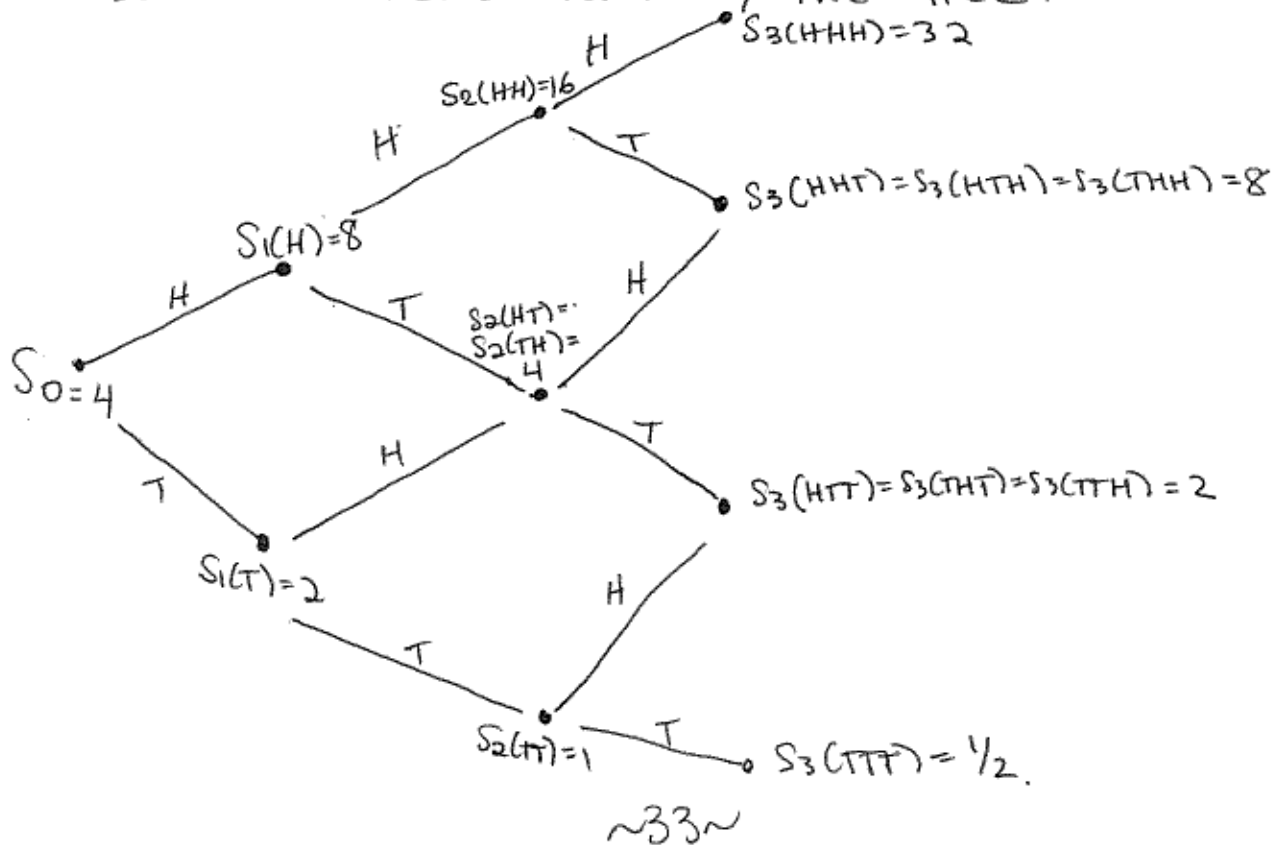
Example

Suppose $p = 2/3$, $q = 1/3$, $S_0 = 4$, $r = 5/4$, $u = 2$, $d = 1/2$

Compute $\mathbb{E}_1 [S_2](H)$, $\mathbb{E}_1 [S_3](H)$ & $\mathbb{E}_1 [\mathbb{E}_2 [S_3]](H)$.

(Note: \mathbb{E} here represents the expected value under the reference probability measure. We're not pricing anything here.)

Solution: let's start w/ the tree:



So then

$$(i) \mathbb{E}_1[S_2](H) = \frac{2}{3} \cdot 16 + \frac{1}{3} \cdot 4 = 12$$

$$(ii) \mathbb{E}_1[S_3](H) = \frac{2}{3} \left[\frac{2}{3} \cdot 32 + \frac{1}{3} \cdot 8 \right] + \frac{1}{3} \left[\frac{2}{3} \cdot 8 + \frac{1}{3} \cdot 2 \right] = 18$$

$$(iii) \mathbb{E}_1[\mathbb{E}_2[S_3]](H) = \mathbb{E}_1[S_3](H) \text{ by iterated conditioning.}$$

But let's compute it directly:

$$\mathbb{E}_2[S_3](HH) = \frac{2}{3} \cdot 32 + \frac{1}{3} \cdot 8 = 24$$

$$\mathbb{E}_2[S_3](HT) = \frac{2}{3} \cdot 8 + \frac{1}{3} \cdot 2 = 6$$

and

$$\begin{aligned} \mathbb{E}_1[\mathbb{E}_2[S_3]](H) &= \frac{2}{3} \mathbb{E}_2[S_3](HH) + \frac{1}{3} \mathbb{E}_2[S_3](HT) \\ &= \frac{2}{3} \cdot 24 + \frac{1}{3} \cdot 6 = 18, \end{aligned}$$

as expected. ■

Interpreting Equation (*) intuitively:



$$S_n = \frac{1}{1+r} \mathbb{E}_n^{\sim}[S_{n+1}] = \mathbb{E}_n^{\sim} \left[\frac{S_{n+1}}{1+r} \right]$$

i.e. OUR BEST GUESS FOR THE STOCK PRICE AT TIME $n+1$ (APPROPRIATELY DISCOUNTED) IS JUST THE STOCK PRICE AT TIME n .

We have a name for processes like this:



Processes $\{M_0, M_1, \dots, M_N\}$ (sequence of RV's) where best guess for value of M_{n+1} at time n is just M_n are called martingales.

\leadsto §2.4 in Shreve



Def. A sequence of random variables $(M_n)_{0 \leq n \leq N}$ ("stochastic process") such that $\forall n \in \{0, 1, \dots, N\}$, M_n depends only on the first n tosses (aka "adapted process") is a



- martingale $\Leftrightarrow M_n = \mathbb{E}_n[M_{n+1}] \forall n$
- submartingale $\Leftrightarrow M_n \leq \mathbb{E}_n[M_{n+1}] \forall n$
- supermartingale $\Leftrightarrow M_n \geq \mathbb{E}_n[M_{n+1}] \forall n$

(This is called the "one step ahead property!")

Remark: Note that the choice of measure matters!
An adapted process can be a martingale under one measure but not another!

Proposition Let $(M_n)_{0 \leq n \leq N}$ be a martingale. Then

$$M_n = \mathbb{E}_n[M_m] \quad \forall 0 \leq n \leq m \leq N$$

Proof Induction! Base case:

$$M_n = \mathbb{E}_n[M_{n+1}] = \mathbb{E}_n[\mathbb{E}_{n+1}[M_{n+2}]] = \mathbb{E}_n[M_{n+2}]$$

because it's a martingale \nearrow \nwarrow iterated conditioning

Induction step:

Assume $M_n = \mathbb{E}_n[M_{n+k}]$ for some k .

Then $M_n = \mathbb{E}_n[\mathbb{E}_{n+k}[M_{n+k+1}]] = \mathbb{E}_n[M_{n+k+1}]$.

So by induction, we can go as far ahead as we want. ■

Corollary $M_0 = \mathbb{E}[M_N]$.

Claim The adapted process

$$\left(\frac{S_n}{(1+r)^n} \right)_{0 \leq n \leq N}$$

is a martingale under $\tilde{\mathbb{P}}$ (the risk-neutral measure).

Proof 1. We need to show $\tilde{\mathbb{E}}_n \left[\frac{S_{n+1}}{(1+r)^{n+1}} \right] = \frac{S_n}{(1+r)^n}$ for all $n \in \{0, 1, \dots, N\}$:

Let $n \in \{0, 1, \dots, N\}$.

$$\tilde{\mathbb{E}}_n \left[\frac{S_{n+1}}{(1+r)^{n+1}} \right] = \frac{1}{(1+r)^{n+1}} \tilde{\mathbb{E}}_n [S_{n+1}]$$

taking out what is known only 2 choices \rightarrow

$$= \frac{1}{(1+r)^{n+1}} \left\{ \tilde{p} u S_n + \tilde{q} d S_n \right\}$$

algebra \rightarrow

$$= \frac{S_n}{(1+r)^{n+1}} \left\{ \tilde{p} u + \tilde{q} d \right\}$$

since $\tilde{p} u + \tilde{q} d = 1+r$ \rightarrow

$$= \frac{S_n}{(1+r)^n} \cdot \blacksquare$$

Proof 2. There's an alternate proof in Shreve.

Example. Consider an N -period binomial model with $u=1.25$, $d=0.8$, $r=0.1$. Stock S has price S_0 . Let V be a derivative security with payoff given by $V_N(\omega) = \frac{1}{S_N(\omega)}$. Find the (arbitrage-free) price V_0 .

Note that trying to do this with backward induction would suck. Let's try it a little:

At time N , there are $N+1$ possibilities: there are k heads & $N-k$ tails, for $k=0,1,2,\dots,N$.

The corresponding payouts are $\frac{1}{S_N(k \text{ H's} \& N-k \text{ T's})} =$
 $= \frac{1}{u^k d^{N-k} S_0}$, for each k .

Going back one time to $t=N-1$, what's V_{N-1} (j heads, $N-1-j$ tails)? For any $j=0,1,\dots,N-1$, the next step is either H or T. So we have

$$\begin{aligned} V_{N-1}(j \text{ H's}, N-1-j \text{ T's}) &= \frac{1}{1+r} \left[\tilde{p} \frac{1}{u^{j+1} d^{N-1-j} S_0} + \tilde{q} \frac{1}{u^j d^{N-j} S_0} \right] \\ &= \frac{1}{1+r} \frac{1}{S_0} \frac{1}{u^j d^{N-1-j}} \left[\frac{\tilde{p}}{u} + \frac{\tilde{q}}{d} \right] \\ &= \frac{1}{1+r} \frac{1}{S_0} \frac{1}{u^{j+1} d^{N-j}} \left[\tilde{p}d + \tilde{q}u \right], \end{aligned}$$

for all $j=0,1,2,\dots,N-1$.

By similar reasoning, if we have i heads at time $N-2$, the value should be

$$\begin{aligned} &\frac{1}{1+r} \left[\tilde{p} V_{N-1}(i+1 \text{ H's} \& N-i) + \tilde{q} V_{N-1}(i \text{ H's} \& N-1-i \text{ T's}) \right] \\ &= \frac{1}{1+r} \left[\tilde{p} \frac{\tilde{p}d + \tilde{q}u}{(1+r) u^{i+2} d^{N-i-1} S_0} + \tilde{q} \frac{\tilde{p}d + \tilde{q}u}{(1+r) u^{i+1} d^{N-i} S_0} \right] \\ &= \frac{(\tilde{p}d + \tilde{q}u)^2}{(1+r)^2 u^{i+2} d^{N-i} S_0}. \end{aligned}$$

OK, so we're kind of getting somewhere but this sucks. So let's do something smarter.

OK, so $(V_n)_{0 \leq n \leq N} = \left(\frac{1}{S_n}\right)_{0 \leq n \leq N}$ isn't a martingale under $\tilde{\mathbb{P}}$. But can we find some constants $\alpha_0, \alpha_1, \dots, \alpha_n, \dots, \alpha_N$ such that

$\left(\frac{\alpha_n}{S_n}\right)_{0 \leq n \leq N}$ is a martingale under $\tilde{\mathbb{P}}$.

Let $n \in \{0, 1, \dots, N-1\}$. Then

$$\tilde{\mathbb{E}}_n \left[\frac{\alpha_{n+1}}{S_{n+1}} \right] = \frac{\alpha_n}{S_n}$$

Since it's a martingale. So

$$\begin{aligned} \frac{\alpha_n}{S_n} &= \tilde{\mathbb{E}}_n \left[\frac{\alpha_{n+1}}{S_{n+1}} \right] = \alpha_{n+1} \tilde{\mathbb{E}}_n \left[\frac{1}{S_{n+1}} \right] \\ &= \alpha_{n+1} \left(\tilde{p} \frac{1}{uS_n} + \tilde{q} \frac{1}{dS_n} \right) \\ &= \frac{\alpha_{n+1}}{S_n} \left(\frac{\tilde{p}}{u} + \frac{\tilde{q}}{d} \right) \\ &= \frac{\alpha_{n+1}}{S_n} \left(\frac{\tilde{p}d + \tilde{q}u}{ud} \right) \end{aligned}$$

So we need $\alpha_n = \frac{\alpha_{n+1}}{ud} (\tilde{p}d + \tilde{q}u)$, for all $n \in \{0, 1, \dots, N-1\}$.

May as well let $\alpha_0 = 1$. So then

$$\alpha_{n+1} = \left(\frac{ud}{\tilde{p}d + \tilde{q}u} \right) \alpha_n = \left(\frac{ud}{\tilde{p}d + \tilde{q}u} \right)^{n+1}$$

So in our case,

$$\begin{aligned} \alpha_n &= \left(\frac{(0.8)(1.25)}{\frac{2}{3}(0.8) + \frac{1}{3}(1.25)} \right)^n \\ &= \left(\frac{3}{2.85} \right)^n \end{aligned}$$

So we've shown $\left(\left(\frac{3}{2.85} \right)^n \frac{1}{S_n} \right)_{0 \leq n \leq N}$ is a martingale (under $\tilde{\mathbb{P}}$).

Why is this useful? Well, $V_0 = \frac{1}{1.1^N} \tilde{\mathbb{E}} \left[\frac{1}{S_N} \right]$.

Since this is a martingale, we get

$$\frac{1}{S_0} = \frac{\alpha_0}{S_0} = \tilde{\mathbb{E}}_0 \left[\left(\frac{3}{2.85} \right)^n \frac{1}{S_n} \right]$$

for all $n=0,1,\dots,N$.

$$S_0 \frac{1}{S_0} = \left(\frac{3}{2.85} \right)^N \tilde{\mathbb{E}}_0 \left[\frac{1}{S_N} \right] = \left(\frac{3}{2.85} \right)^N \tilde{\mathbb{E}} \left[\frac{1}{S_N} \right].$$

$$S_0 V_0 = \frac{1}{1.1^N} \tilde{\mathbb{E}} \left[\frac{1}{S_N} \right] = \frac{1}{1.1^N} \left(\frac{2.85}{3} \right)^N \frac{1}{S_0} \\ = \left(\frac{2.85}{3.3} \right)^N \frac{1}{S_0}$$