

There are several ways to produce (sub/super) martingales from other martingales (or other processes).

Prop. Let $(X_n)_{n \in \mathbb{N}}$ be a martingale, and let $\varphi: \mathbb{R} \rightarrow \mathbb{R}$ be convex & increasing s.t. $\varphi(X_n) \in L^1 \forall n$. Then $(\varphi(X_n))_{n \in \mathbb{N}}$ is a submartingale.

Pf. $\varphi(X_n) = \varphi(\mathbb{E}_{\mathcal{F}_n}[X_{n+1}])$

If we only know $(X_n)_{n \in \mathbb{N}}$ is a submartingale, $X_n \leq \mathbb{E}_{\mathcal{F}_n}[X_{n+1}]$

Eg. If $(X_n)_{n \in \mathbb{N}}$ is an L^p martingale ($p \geq 1$)

then $(|X_n|^p)_{n \in \mathbb{N}}$ is a submartingale.

If $X_n \geq 0$ is an L^p submartingale, $(X_n^p)_{n \geq 0}$ is

We've seen a number of Markov chains that are also martingales. That's not general: even if a Markov chain's state space is $\subseteq \mathbb{R}$, it's usually not a martingale. But some functions of it are.

Theorem: Let $(X_n)_{n \in \mathbb{N}}$ be a time-homogeneous Markov chain in (S, \mathcal{B}) with transition operator Q . Let $f: \mathbb{N} \times S \rightarrow \mathbb{R}$ be measurable, satisfying

$$\mathbb{E}[|f(n, X_n)|] < \infty \quad \forall n.$$

Then $Z_n := f(n, X_n)$ is a martingale, provided

$$(Qf)(n+1, \cdot) = f(n, \cdot) \quad \forall n \in \mathbb{N}$$

(In particular, if $f: S \rightarrow \mathbb{R}$ doesn't depend on n , the condition is $Qf = f$)

Pf. $\mathbb{E}[Z_{n+1} | \mathcal{F}_n] = \mathbb{E}[f(n+1, X_{n+1}) | \mathcal{F}_n] = Qf(n+1, X_n)$

One way to make this work, at least for a finite time interval $0 \leq n \leq T$, is to run the Markov chain **backwards**

$$f(n, y) := (Q^{T-n} g)(y) \quad \text{for some } g: S \rightarrow \mathbb{R}.$$

So $(Z_n = (Q^{T-n} g)(X_n))_{0 \leq n \leq T}$ is a martingale.

Eg. $(X_n)_{n \in \mathbb{N}} = \text{RW}(p)$ on \mathbb{Z} , $Qf(x) = pf(x+1) + (1-p)f(x-1)$, with $X_0 = 0$
we've seen that all (and only) the functions $f(x) = \alpha + \beta \left(\frac{1-p}{p}\right)^x$ satisfy $Qf = f$ on \mathbb{Z} .

$\therefore M_n = \alpha + \beta \left(\frac{1-p}{p}\right)^{X_n}$ is a martingale $\forall \alpha, \beta \in \mathbb{R}$.

We can also verify this directly. Set $\lambda = \frac{1-p}{p}$.

$$E_{\mathcal{G}_n}[\lambda^{X_{n+1}}] =$$

(Note: $\{\text{martingales}\}$ is a vector space.)

Now, can we get more leverage if we allow f to depend on n ?

Notice: if $\lambda \neq 0$, $Q(\lambda^x) =$

So " Q^{-1} " makes sense (on the span of exponential functions)

and we can define $f_\lambda(n, x) := (p\lambda + (1-p)\lambda^{-1})^{-n} \lambda^x$

Thus $Qf_\lambda(n+1, \cdot) = f_\lambda(n, \cdot) \quad \forall n$. So, if X_0 is bounded,

$M_n = f_\lambda(n, X_n) = (p\lambda + (1-p)\lambda^{-1})^{-n} \lambda^{X_n}$ is a martingale.

$f_\theta(n, x) = (pe^\theta + (1-p)e^{-\theta})^{-n} e^{\theta x}$ satisfies $\mathbb{Q} f_\theta(n+1, \cdot) = f_\theta(n, \cdot)$, $\forall \lambda \neq 0$.

depends smoothly on θ .

$\mathbb{Q} f(x) = pf(x+1) + (1-p)f(x-1)$
action commutes with derivatives.

$\therefore \frac{d^k}{d\theta^k} f_\theta(n, x) \big|_{\theta=0}$ also satisfies the recurrence.

A careful computation, and argument, then shows

Prop: If $(X_n)_{n \in \mathbb{N}}$ is RW(p) (Markov chain; not a martingale if $p \neq \frac{1}{2}$)

then

$$M_n^{(1)} = X_n - n(p - (1-p))$$

$$M_n^{(2)} = (M_n^{(1)})^2 - 4np(1-p)$$

are martingales.

[HW]

We will soon see: this provides some very effective tools to calculate expectations (and higher moments) of some stopping times.