Lecture 6: Martingale

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1 Conditional Expectation

- **Definition** (Conditional Expectation) [Resnick, 2013] Let $(\Omega, \mathscr{F}, \mathcal{P})$ be a probability space and $\mathscr{G} \subset \mathscr{F}$ be a sub- σ -algebra. Suppose $X \in L^1(\Omega, \mathscr{F}, \mathcal{P})$. There exists a function $\mathbb{E}[X|\mathscr{G}]$, called the <u>conditional expectation</u> of X with respect to \mathscr{G} such that
 - 1. $\mathbb{E}[X|\mathcal{G}]$ is \mathcal{G} -measureable and integrable with respect to \mathcal{P} .
 - 2. $\mathbb{E}[X|\mathcal{G}]$ satisfies the functional equation:

$$\int_G X d\mathcal{P} = \int_G \mathbb{E}\left[X|\mathcal{G}\right] d\mathcal{P}, \quad \forall \, G \in \mathcal{G}.$$

- Remark To prove the existence of such a random variable,
 - 1. consider first the case of **nonnegative** X. Define a measure ν on \mathscr{G} by

$$\nu(G) = \int_G X d\mathcal{P} = \int_{\Omega} X \mathbb{1}_G d\mathcal{P}.$$

This measure is *finite* because X is *integrable*, and it is **absolutely continuous** with respect to \mathcal{P} . By the *Lebesgue-Radon-Nikodym Theorem*, there is a \mathscr{G} -measurable function f such that

$$\nu(G) = \int_G f d\mathcal{P}.$$

This f has properties (1) and (2).

- 2. If X is not necessarily nonnegative, $\mathbb{E}[X_{+}|\mathscr{G}] \mathbb{E}[X_{-}|\mathscr{G}]$ clearly has the required properties.
- Remark As \mathscr{G} increases, condition (1) becomes **weaker** and condition (2) becomes **stronger**.
- Remark Let $(\Omega, \mathscr{F}, \mathcal{P})$ be a probability space, with $\mathscr{G} \subset \mathscr{F}$ a sub- σ -algebra, define

$$\mathcal{P}[A|\mathscr{G}] = \mathbb{E}\left[\mathbb{1}_A|\mathscr{G}\right]$$

for all $A \in \mathscr{F}$.

• Remark By definition, the conditional expectation is a *Radon-Nikodym derivative* of $d\nu|_{\mathscr{G}} = Xd\mathcal{P}|_{\mathscr{G}}$ w.r.t. $d\mathcal{P}|_{\mathscr{G}}$ within \mathscr{G} .

$$\mathbb{E}\left[X|\mathscr{G}\right] := \frac{Xd\mathcal{P}|_{\mathscr{G}}}{d\mathcal{P}|_{\mathscr{G}}} = X|_{\mathscr{G}}.$$

Thus $\mathbb{E}[X|\mathcal{G}]$ is the **projection** of X on sub σ -algebra \mathcal{G} .

• Remark (Conditioning on Random Variables) By definition, conditioning on random variables $(X_t, t \in T)$ on (Ω, \mathcal{B}) can be expressed as

$$\mathbb{E}\left[X|X_t, t \in T\right] \equiv \mathbb{E}\left[X|\sigma(X_t, t \in T)\right],$$

where $\sigma(X_t, t \in T)$ is the σ -algebra generated by the cylinder set

$$C_n[A] \equiv \{\omega : (X_t(\omega), 1 \le t \le n) \in A\} \in \mathcal{B}, \quad A \in \mathcal{B}(\mathbb{R}^n), \forall n \in \mathcal{B}(\mathbb{R}^n) \}$$

• Remark (σ -Algebra Generated by Partition of Sample Space) As above, assume that the sub σ -algebra \mathscr{G} is generated by a partition B_1, B_2, \ldots of Ω , then for $X \in L^1(\Omega, \mathscr{F}, \mathcal{P})$,

$$\mathbb{E}\left[X|B_i\right] = \int X d\mathcal{P}(X|B_i) = \int_{B_i} X d\mathcal{P}/\mathcal{P}(B_i)$$

where $\mathcal{P}(X|B_i)$ is the conditional probability defined in previous section. If $\mathcal{P}(B_i) = 0$, then $\mathbb{E}[X|B_i] = 0$. We claim that

1.

$$\mathbb{E}\left[X|\mathscr{G}\right] = \sum_{i=1}^{\infty} \mathbb{E}\left[X|B_i\right] \mathbb{1}_{B_i}, \quad a.s.$$

2. For any $A \in \mathscr{F}$,

$$\mathcal{P}(A|\mathcal{G}) = \sum_{i=1}^{\infty} \mathcal{P}(A|B_i) \mathbb{1}_{B_i}, \quad a.s.$$

• Remark Both $P[A|\mathscr{F}]$ and $\mathbb{E}[X|\mathscr{F}]$ are random variables from $\Omega \to \mathbb{R}$. Formally speaking,

$$\begin{split} P\left[(X,Y) \in A | \sigma(X)\right]_{\omega} &\equiv P\left[(X(\omega),Y) \in A\right] \\ &= P\left\{\omega' : (X(\omega),Y(\omega')) \in A\right\} \\ &\equiv f(X(\omega)) \\ &= \left.\nu\right|_{\sigma(X)}(A) \\ &\mathbb{E}\left[(X,Y) | \sigma(X)\right]_{\omega} = \lim_{\substack{m(A) \to 0 \\ \omega \in A \in \sigma(X)}} \frac{P\left\{\omega' : (X(\omega),Y(\omega')) \in A\right\}}{m(A)} \end{split}$$

It is the expected value of X for someone who knows for each $E \in \mathcal{F}$, whether or not $\omega \in E$, which E itself remains unknown.

- Proposition 1.1 (Properties of Conditional Expectation) [Resnick, 2013] Let $(\Omega, \mathcal{F}, \mathcal{P})$ be a probability space and $\mathcal{G} \subset \mathcal{F}$ be a sub- σ -algebra. Suppose $X, Y \in L^1(\Omega, \mathcal{F}, \mathcal{P})$ and $\alpha, \beta \in \mathbb{R}$.
 - 1. (Linearity): $\mathbb{E}\left[\alpha X + \beta Y | \mathcal{G}\right] = \alpha \mathbb{E}\left[X | \mathcal{G}\right] + \beta \mathbb{E}\left[Y | \mathcal{G}\right];$
 - 2. (Projection): If X is \mathscr{G} -measurable, then $\mathbb{E}[X|\mathscr{G}] = X$ almost surely.
 - 3. (Conditioning on Indiscrete σ -Algebra):

$$\mathbb{E}\left[X|\left\{\emptyset,\Omega\right\}\right]=\mathbb{E}\left[X\right].$$

- 4. (Monotonicity): If $X \ge 0$, then $\mathbb{E}[X|\mathcal{G}] \ge 0$ almost surely. Similarly, if $X \ge Y$, then $\mathbb{E}[X|\mathcal{G}] \ge \mathbb{E}[Y|\mathcal{G}]$ almost surely.
- 5. (Modulus Inequality):

$$|\operatorname{\mathbb{E}}\left[X|\mathscr{G}\right]| \leq \operatorname{\mathbb{E}}\left[\left|X\right|\,|\mathscr{G}\right].$$

6. (Monotone Convergence Theorem): If $\{X_n\}_{n=1}^{\infty} \subset L^1(\Omega, \mathscr{F}, \mathcal{P}), 0 \leq X_1 \leq X_2 \leq \dots$ is a monotone sequence of non-negative random variables and $X_n \to X$ then

$$\lim_{n\to\infty} \mathbb{E}\left[X_n|\mathcal{G}\right] = \mathbb{E}\left[\lim_{n\to\infty} X_n \middle| \mathcal{G}\right] = \mathbb{E}\left[X|\mathcal{G}\right].$$

7. (**Fatou Lemma**): If $\{X_n\}_{n=1}^{\infty} \subset L^1(\Omega, \mathscr{F}, \mathcal{P})$, and $X_n \geq 0$ for all n, then

$$\mathbb{E}\left[\liminf_{n\to\infty} X_n \big| \mathscr{G}\right] \le \liminf_{n\to\infty} \mathbb{E}\left[X_n \big| \mathscr{G}\right]$$

8. (Dominated Convergence Theorem): If $\{X_n\}_{n=1}^{\infty} \subset L^1(\Omega, \mathcal{F}, \mathcal{P}) \text{ and } |X_n| \leq Z$, where $Z \in L^1(\Omega, \mathcal{F}, \mathcal{P})$ is a random variable, $X_n \to X$ almost surely, then

$$\lim_{n \to \infty} \mathbb{E}\left[X_n | \mathcal{G}\right] = \mathbb{E}\left[\lim_{n \to \infty} X_n \big| \mathcal{G}\right] = \mathbb{E}\left[X | \mathcal{G}\right], \quad a.s.$$

9. (Product Rule): If Y is \mathscr{G} -measurable,

$$\mathbb{E}\left[X\,Y|\mathcal{G}\right] = Y\,\mathbb{E}\left[X|\mathcal{G}\right], \quad a.s.$$

10. (Smoothing): For $\mathscr{F}_1 \subset \mathscr{F}_0 \subset \mathscr{F}$,

$$\mathbb{E}\left[\mathbb{E}\left[X|\mathscr{F}_{0}\right]|\mathscr{F}_{1}\right] = \mathbb{E}\left[X|\mathscr{F}_{1}\right]$$

$$\mathbb{E}\left[\mathbb{E}\left[X|\mathscr{F}_{1}\right]|\mathscr{F}_{0}\right] = \mathbb{E}\left[X|\mathscr{F}_{1}\right].$$

Note that $\mathbb{E}[X|\mathscr{F}_1]$ is **smoother** than $\mathbb{E}[X|\mathscr{F}_0]$. Moreover

$$\mathbb{E}\left[X\right] = \mathbb{E}\left[X|\left\{\emptyset,\Omega\right\}\right] = \mathbb{E}\left[\mathbb{E}\left[X|\mathscr{F}_{0}\right]|\left\{\emptyset,\Omega\right\}\right] = \mathbb{E}\left[\mathbb{E}\left[X|\mathscr{F}_{0}\right]\right].$$

11. (The Conditional Jensen's Inequality). Let ϕ be a convex function, $\phi(X) \in L^1(\Omega, \mathcal{F}, \mathcal{P})$. Then almost surely

$$\phi\left(\mathbb{E}\left[X|\mathscr{G}\right]\right) \leq \mathbb{E}\left[\phi(X)|\mathscr{G}\right]$$

2 Martingale

- 2.1 Martingale, Sub-Martingale, Super-Martingale
 - **Definition** (*Martingale*) [Resnick, 2013] Let $\{X_n, n \geq 0\}$ be a stochastic process on (Ω, \mathscr{F}) and $\{\mathscr{F}_n, n \geq 0\}$ be a *filtration*; that is, $\{\mathscr{F}_n, n \geq 0\}$ is an *increasing sub* σ -fields of \mathscr{F}

$$\mathscr{F}_0 \subseteq \mathscr{F}_1 \subseteq \mathscr{F}_2 \subseteq \ldots \subseteq \mathscr{F}$$
.

Then $\{(X_n, \mathscr{F}_n), n \geq 0\}$ is a martingale (mg) if

- 1. X_n is **adapted** in the sense that for each $n, X_n \in \mathscr{F}_n$; that is, X_n is \mathscr{F}_n -measurable.
- 2. $X_n \in L_1$; that is $\mathbb{E}[|X_n|] < \infty$ for $n \ge 0$.

3. For $0 \le m < n$

$$\mathbb{E}\left[X_n \mid \mathscr{F}_m\right] = X_m, \quad \text{a.s.} \tag{1}$$

If the equality of (1) is replaced by \geq ; that is, things are getting better on the average:

$$\mathbb{E}\left[X_n \mid \mathscr{F}_m\right] \ge X_m, \quad \text{a.s.} \tag{2}$$

then $\{X_n\}$ is called a <u>sub-martingale (submg)</u> while if things are getting worse on the average

$$\mathbb{E}\left[X_n \mid \mathscr{F}_m\right] \le X_m, \quad \text{a.s.} \tag{3}$$

 $\{X_n\}$ is called a *super-martingale* (supermg).

- Remark $\{X_n\}$ is martingale if it is both a sub and supermartingale. $\{X_n\}$ is a supermartingale if and only if $\{-X_n\}$ is a submartingale.
- Remark If $\{X_n\}$ is a *martingale*, then $\mathbb{E}[X_n]$ is *constant*. In the case of a *submartingale*, the mean increases and for a *supermartingale*, the mean decreases.
- Proposition 2.1 [Resnick, 2013] If $\{(X_n, \mathscr{F}_n), n \geq 0\}$ is a **(sub, super) martingale**, then

$$\{(X_n, \sigma(X_0, X_1, \dots, X_n)), n \ge 0\}$$

is also a (sub, super) martingale.

2.2 Martingale Difference Sequence

- Definition (Martingale Differences). [Resnick, 2013] $\{(d_j, \mathcal{B}_j), j \geq 0\}$ is a <u>(sub, super) martingale difference sequence</u> or a (sub, super) fair sequence if
 - 1. For $j \geq 0$, $\mathscr{B}_j \subset \mathscr{B}_{j+1}$.
 - 2. For $j \geq 0$, $d_j \in L_1$, $d_j \in \mathcal{B}_j$; that is, d_j is absolutely integrable and \mathcal{B}_j -measurable.
 - 3. For $j \geq 0$,

$$\mathbb{E}[d_{j+1}|\mathscr{B}_j] = 0,$$
 (martingale difference / fair sequence);
 $\geq 0,$ (submartingale difference / subfair sequence);
 $\leq 0,$ (supmartingale difference / supfair sequence)

• Theorem 2.2 (Construction of Martingale From Martingale Difference)[Resnick, 2013]

If $\{(d_j, \mathcal{B}_j), j \geq 0\}$ is (sub, super) martingale difference sequence, and

$$X_n = \sum_{j=0}^n d_j,$$

then $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a (sub, super) martingale.

• Theorem 2.3 (Construction of Martingale Difference From Martingale) [Resnick, 2013]

Suppose $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a (sub, super) martingale. Define

$$d_0 := X_0 - \mathbb{E}[X_0]$$

 $d_i := X_i - X_{i-1}, \quad j \ge 1.$

Then $\{(d_j, \mathcal{B}_j), j \geq 0\}$ is a (sub, super) martingale difference sequence.

• Theorem 2.4 (Orthogonality of Martingale Differences). [Resnick, 2013] If $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a martingale where X_n can be decomposed as

$$X_n = \sum_{j=0}^n d_j,$$

 d_j is \mathscr{B}_j -measurable and $\mathbb{E}[d_j^2] < \infty$ for $j \geq 0$, then $\{d_j\}$ are **orthogonal**:

$$\mathbb{E}\left[d_i \, d_i\right] = 0 \quad i \neq j.$$

Proof: This is an easy verification: If j > i, then

$$\begin{split} \mathbb{E}\left[d_i \, d_j\right] &= \mathbb{E}\left[\mathbb{E}\left[d_i \, d_j \, | \, \mathscr{B}_i\right]\right] \\ &= \mathbb{E}\left[d_i \mathbb{E}\left[d_j \, | \, \mathscr{B}_i\right]\right] = 0. \end{split}$$

A consequence is that

$$\mathbb{E}\left[X_n^2\right] = \mathbb{E}\left[\sum_{i=1}^n d_i^2\right] + 2\sum_{0 \le i < j \le n} \mathbb{E}\left[d_i d_j\right] = \mathbb{E}\left[\sum_{i=1}^n d_i^2\right],$$

which is **non-decreasing**. From this, it seems likely (and turns out to be true) that $\{X_n^2\}$ is a **sub-martingale**.

2.3 Examples of Martingales

• Example (Smoothing as Martingale)

Suppose $X \in L_1$ and $\{\mathscr{B}_n, n \geq 0\}$ is an increasing family of sub σ -algebra of \mathscr{B} . Define for $n \geq 0$

$$X_n := \mathbb{E}\left[X|\mathscr{B}_n\right].$$

Then (X_n, \mathcal{B}_n) is a *martingale*. From this result, we see that $\{(d_n, \mathcal{B}_n), n \geq 0\}$ is a *martingale difference sequence* when

$$d_n := \mathbb{E}\left[X|\mathscr{B}_n\right] - \mathbb{E}\left[X|\mathscr{B}_{n-1}\right], \quad n \ge 1. \tag{4}$$

Proof: See that

$$\mathbb{E}\left[X_{n+1}|\mathscr{B}_n\right] = \mathbb{E}\left[\mathbb{E}\left[X|\mathscr{B}_{n+1}\right]|\mathscr{B}_n\right]$$

$$= \mathbb{E}\left[X|\mathscr{B}_n\right] \qquad \text{(Smoothing property of conditional expectation)}$$

$$= X_n \quad \blacksquare$$

• Example (Sums of Independent Random Variables)

Suppose that $\{Z_n, n \geq 0\}$ is an *independent* sequence of integrable random variables satisfying for $n \geq 0$, $\mathbb{E}[Z_n] = 0$. Set

$$X_0 := 0,$$

$$X_n := \sum_{i=1}^n Z_i, \quad n \ge 1$$

$$\mathscr{B}_n := \sigma(Z_0, \dots, Z_n).$$

Then $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a martingale since $\{(Z_n, \mathcal{B}_n), n \geq 0\}$ is a martingale difference sequence.

• Example (*Likelihood Ratios*).

Suppose $\{Y_n, n \geq 0\}$ are *independent identically distributed* random variables and suppose the true density of Y_n is f_0 (The word "density" can be understood with respect to some fixed reference measure μ .) Let f_1 be some other probability density. For simplicity suppose $f_0(y) > 0$, for all y. For $n \geq 0$, define the likelihood ratio

$$X_n := \frac{\prod_{i=0}^n f_1(Y_i)}{\prod_{i=0}^n f_0(Y_i)}$$
$$\mathscr{B}_n := \sigma(Y_0, \dots, Y_n)$$

Then (X_n, \mathcal{B}_n) is a *martingale*.

Proof: See that

$$\mathbb{E}\left[X_{n+1}|\mathscr{B}_{n}\right] = \mathbb{E}\left[\left(\frac{\prod_{i=0}^{n} f_{1}(Y_{i})}{\prod_{i=0}^{n} f_{0}(Y_{i})}\right) \frac{f_{1}(Y_{n+1})}{f_{0}(Y_{n+1})} \mid Y_{0}, \dots, Y_{n}\right]$$

$$= X_{n} \mathbb{E}\left[\frac{f_{1}(Y_{n+1})}{f_{0}(Y_{n+1})} \mid Y_{0}, \dots, Y_{n}\right]$$

$$= X_{n} \mathbb{E}\left[\frac{f_{1}(Y_{n+1})}{f_{0}(Y_{n+1})}\right] \quad \text{(by independence)}$$

$$:= X_{n} \int \frac{f_{1}(y_{n+1})}{f_{0}(y_{n+1})} f_{0}(y_{n+1}) d\mu(y_{n+1}) = X_{n}.$$

- Example (Moment Generating Functions)
- Example (Method of Centering by Conditional Means).
- Example (Markov Chains).

2.4 Doob's Decomposition

• Definition (*Predictable and Increasing Process*) Given a process $\{X_n, n \geq 0\}$ and an increasing family of σ -algebras $\{\mathscr{B}_n, n \geq 0\}$. We call $\{X_n, n \geq 0\}$ <u>predictable</u> if $X_0 \in \mathscr{B}_0 = \{\emptyset, \Omega\}$ and X_n is \mathscr{B}_{n-1} -measurable for each $n = 1, 2, \ldots$

Call a process $\{A_n, n \geq 0\}$ an <u>increasing process</u> if $\{A_n\}$ is predictable and almost surely

$$0 = A_0 \le A_1 \le \ldots \le A_n \le \ldots$$

• Theorem 2.5 (Doob Decomposition) [Billingsley, 2008, Resnick, 2013] Any submartingale $\{(X_n, \mathcal{B}_n), n \geq 0\}$ can be written in a unique way as the sum of a martingale $\{(M_n, \mathcal{B}_n), n \geq 0\}$ and an increasing process $\{A_n, n \geq 0\}$; that is

$$X_n = M_n + A_n, \quad n \ge 0, \quad a.s.$$

2.5 Stopping Time and Optional Sampling Theorem

• **Definition** Let $(X_t)_t$ be a stochastic process and let $\{\mathscr{B}_t, n \geq 0\}$ be an increasing family of σ -algebras. Denote $\mathscr{B} = \bigcup_{t \geq 0} \mathscr{B}_t$.

A random variable $T: (\Omega, \mathscr{B}) \to (\mathbb{N}_+ \cup \{+\infty\}, 2^{\mathbb{N}_+ \cup \{+\infty\}})$ is called a <u>stopping time</u> with respect to $\{\mathscr{B}_t, t \geq 0\}$, if $\mathbb{1}_{\{T=k\}}$ is \mathscr{B}_k -measurable for $\forall k \geq 0$; i.e.

$$\{T=k\}\in\mathscr{B}_k,\quad\forall k\geq 0.$$

- Theorem 2.6 (Weak Version of Optional Sampling Theorem) [Billingsley, 2008] Let $\{(X_n, \mathcal{B}_n), n \geq 0\}$ be a martingale where $\mathcal{B}_n = \sigma(X_1, \dots, X_n)$ and let T be stopping time with respect to $\{\mathcal{B}_n, n \geq 0\}$. Suppose that at least one of the following conditions hold:
 - 1. $T \leq n_0$, a.s. for some constant $n_0 > 0$;
 - 2. $T < \infty$, a.s. and $|X_i| \le K$ where $i \le T$.

Then

$$\mathbb{E}\left[X_T\right] = \mathbb{E}\left[X_0\right].$$

• Theorem 2.7 (Weak Version of Optional Sampling Theorem, again) [Billingsley, 2008]

Let $\{(X_n, \mathcal{B}_n), n \geq 0\}$ be a martingale where $\mathcal{B}_n = \sigma(X_1, \ldots, X_n)$ and let τ be stopping time with respect to $\{\mathcal{B}_n, n \geq 0\}$. Suppose $\mathbb{E}[\tau] < \infty$, and there exists $n \geq 0$ and some constant c such that

$$\mathbb{E}\left[\left|X_{n+1}-X_{n}\right|\left|\mathscr{B}_{n}\right]\leq c,\quad a.s.\ on\ \left\{ \tau\geq n\right\} .$$

Then $\mathbb{E}[|X_{\tau}|] < \infty$ and

$$\mathbb{E}\left[X_{\tau}\right] = \mathbb{E}\left[X_{0}\right].$$

• Theorem 2.8 (Doob's Optional Sampling Theorem) [Billingsley, 2008, Resnick, 2013] Let $\{(X_n, \mathcal{B}_n), n \geq 0\}$ be a martingale and let S, T be $\{\mathcal{B}_n, n \geq 0\}$ -stopping time bounded by constant c, with $S \leq T \leq c$, a.s., then

$$\mathbb{E}\left[X_T|\mathscr{B}_S\right] = X_S, \quad a.s. \tag{5}$$

where

$$\mathscr{B}_S := \{ A \in \mathscr{B} : A \cap \{ S \leq n \} \in \mathscr{B}_n, \quad n = 0, 1, \ldots \}$$

• Theorem 2.9 (Wald's Identity) [Billingsley, 2008, Resnick, 2013] Consider random walk $S_0 = X_0$ and $S_n = X_0 + \xi_1 + \xi_2 + \ldots + \xi_n$, $n = 1, 2, \ldots$, where ξ_1, \ldots, ξ_n are i.i.d. random variables such that $\mathbb{E}[|\xi_i|] < \infty$ and $\mathbb{E}[\xi_i] = \mu$. Let τ be a stopping time with respect to $\{\sigma(S_0, S_1, \ldots, S_n), n \geq 0\}$ and $\mathbb{E}[\tau] < \infty$. Then

$$\mathbb{E}\left[S_{\tau} - S_{0}\right] = \mu \mathbb{E}\left[\tau\right]$$

• Proposition 2.10 (Wald's Equation) [Billingsley, 2008] Let X_1, \ldots, X_n be i.i.d. random variables with $\mathbb{E}[|X_i|] < \infty$ and $\mathbb{E}[X_i] = \mu < \infty$. Let τ be a stopping time with respect to $\{\sigma(X_1, \ldots, X_n), n \geq 1\}$ satisfying $\mathbb{E}[\tau] < \infty$. Then

$$\mathbb{E}\left[\sum_{i=1}^{\tau} X_i\right] = \mathbb{E}\left[\tau\right] \mathbb{E}\left[X_i\right] \tag{6}$$

• Proposition 2.11) [Resnick, 2013] Let $\{(X_n, \mathcal{B}_n), n \geq 0\}$ be a martingale and let τ be a stopping time with respect to $\{\mathcal{B}_n, n \geq 0\}$. Then

$$\{X_{\tau \wedge t}, \ t = 0, 1, \ldots\}$$

is a martingale with respect to $\{\mathscr{B}_n, n \geq 0\}$, where $t \wedge \tau = \min\{t, \tau\}$.

2.6 Martingale Inequalities

• Theorem 2.12 (Doob's Maximal Inequality) [Billingsley, 2008] Suppose $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a sub-martingale. Then for any t > 0,

$$\mathbb{P}\left\{\max_{k\leq n} X_k \geq t\right\} \leq \frac{1}{t} \mathbb{E}\left[X_n \mathbb{1}\left\{\max_{k\leq n} X_k \geq t\right\}\right]$$
 (7)

• Definition (Gambling Strategy)

A <u>gambling strategy</u> with respect to an increasing family of σ -algebras $\{\mathscr{B}_n, n \geq 0\}$ is a sequence of random variables $\gamma_1, \gamma_2, \ldots$ such that γ_n is \mathscr{B}_{n-1} -measurable for each $n = 1, 2, \ldots$, where $\mathscr{B}_0 = \{\emptyset, \Omega\}$.

The random variable sequence $(\gamma_1, \gamma_2, ...)$ is called **predictable**.

• Definition (*Upcrossing*)

Consider a sequence of random variables X_1, \ldots, X_n, \ldots such that X_n is \mathcal{B}_n -measurable for all n and two real numbers a < b. Define a gambling strategy $(\gamma_1, \gamma_2, \ldots)$ by setting:

$$\gamma_0 = 0;$$

$$\gamma_{n+1} = \begin{cases} 1 & \text{if } \gamma_n = 0 \text{ and } X_n < 0 \\ 1 & \text{if } \gamma_n = 1 \text{ and } X_n \ge b \\ 0 & \text{otherwise} \end{cases} \quad \forall n = 1, 2, \dots.$$

Such strategy is called a *upcrossing strategy*.

• Lemma 2.13 (The Upcrossing Inequality Lemma) [Billingsley, 2008] Suppose $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a super-martingale and a < b. Then

$$\mathbb{E}\left[U_n[a,b]\right] \le \frac{\mathbb{E}\left[(X_n - a)_+\right]}{b - a} = \frac{-\mathbb{E}\left[(X_n - a)_-\right]}{b - a} \tag{8}$$

where $U_n[a,b]$ denote the total number of upcrossings of [a,b] up to time n,

$$(X_n - a)_- = \min \{X_n - a, 0\} = -\max \{a - X_n, 0\};$$

 $(X_n - a)_+ = \max \{X_n - a, 0\}.$

• Theorem 2.14 (Bernstein Inequality, Martingale Difference Sequence Version) [Wainwright, 2019]

Let $\{(D_k, \mathscr{B}_k), k \geq 1\}$ be a martingale difference sequence, and suppose that

$$\mathbb{E}\left[\exp\left(\lambda D_{k}\right)\middle|\mathscr{B}_{k-1}\right] \leq \exp\left(\frac{\lambda^{2}\nu_{k}^{2}}{2}\right)$$

almost surely for any $|\lambda| < 1/\alpha_k$. Then the following hold:

1. The sum $\sum_{k=1}^{n} D_k$ is **sub-exponential** with **parameters** $\left(\sqrt{\sum_{k=1}^{n} \nu_k^2}, \alpha_*\right)$ where $\alpha_* := \max_{k=1,...,n} \alpha_k$. That is, for any $|\lambda| < 1/\alpha_*$,

$$\mathbb{E}\left[\exp\left\{\lambda\left(\sum_{k=1}^{n}D_{k}\right)\right\}\right] \leq \exp\left(\frac{\lambda^{2}\sum_{k=1}^{n}\nu_{k}^{2}}{2}\right)$$

2. The sum satisfies the concentration inequality

$$\mathbb{P}\left\{\left|\sum_{k=1}^{n} D_{k}\right| \ge t\right\} \le \begin{cases}
2\exp\left(-\frac{t^{2}}{2\sum_{k=1}^{n} \nu_{k}^{2}}\right) & \text{if } 0 \le t \le \frac{\sum_{k=1}^{n} \nu_{k}^{2}}{\alpha_{*}} \\
2\exp\left(-\frac{t}{\alpha_{*}}\right) & \text{if } t > \frac{\sum_{k=1}^{n} \nu_{k}^{2}}{\alpha_{*}}.
\end{cases} \tag{9}$$

• Corollary 2.15 (Azuma-Hoeffding Inequality, Martingale Difference)[Wainwright, 2019]

Let $\{(D_k, \mathscr{B}_k), k \geq 1\}$ be a martingale difference sequence for which there are constants $\{(a_k, b_k)\}_{k=1}^n$ such that $D_k \in [a_k, b_k]$ almost surely for all k = 1, ..., n. Then, for all $t \geq 0$,

$$\mathbb{P}\left\{ \left| \sum_{k=1}^{n} D_k \right| \ge t \right\} \le 2 \exp\left(-\frac{2t^2}{\sum_{k=1}^{n} (b_k - a_k)^2}\right) \tag{10}$$

• An important application of Azuma-Hoeffding Inequality concerns functions that satisfy a bounded difference property.

Definition (Functions with Bounded Difference Property)

Given vectors $x, x' \in \mathcal{X}^n$ and an index $k \in \{1, 2, ..., n\}$, we define a new vector $x^{(-k)} \in \mathcal{X}^n$ via

$$x_j^{(-k)} = \begin{cases} x_j & j \neq k \\ x_k' & j = k \end{cases}$$

With this notation, we say that $f: \mathcal{X}^n \to \mathbb{R}$ satisfies <u>the bounded difference inequality</u> with parameters (L_1, \ldots, L_n) if, for each index $k = 1, 2, \ldots, n$,

$$\left| f(x) - f(x^{(-k)}) \right| \le L_k, \quad \text{for all } x, x' \in \mathcal{X}^n.$$
 (11)

• Corollary 2.16 (McDiarmid's Inequality / Bounded Differences Inequality)[Wainwright, 2019]

Suppose that f satisfies **the bounded difference property** (11) with parameters (L_1, \ldots, L_n) and that the random vector $X = (X_1, X_2, \ldots, X_n)$ has **independent** components. Then

$$\mathbb{P}\left\{|f(X) - \mathbb{E}\left[f(X)\right]| \ge t\right\} \le 2\exp\left(-\frac{2t^2}{\sum_{k=1}^n L_k^2}\right). \tag{12}$$

2.7 Convergence Theorem

• Theorem 2.17 (Doob's Martingale Convergence Theorem) [Billingsley, 2008] Suppose $\{(X_n, \mathcal{B}_n), n \geq 0\}$ is a super-martingale and suppose further that

$$\sup_{n} \mathbb{E}\left[|X_n|\right] < \infty.$$

Then there exists an integrable random variable X such that

$$\lim_{n \to \infty} X_n = X \quad a.s.$$

• Remark Note that a martingale is a super-martingale, so the above theorem is **valid** for **martingale**. Also the above theorem is valid for **sub-martingale** when $\{(-X_n, \mathcal{B}_n), n \geq 0\}$ is considered.

References

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