## 270B - Homework 3

**Problem 1.** Let  $X_1, X_2, ...$  be independent random variables with means  $\mu_i$  and finite variances  $\sigma_i^2$ . Consider the sums  $S_n = X_1 + \cdots + X_n$ . Find sequences of real numbers  $(b_i)$  and  $(c_i)$  such that  $S_n^2 + b_n S_n + c_n$  is a martingale with respect to the  $\sigma$ -algebras generated by  $X_1, ..., X_n$ .

Solution. Let's start by centering the sum: define the random variable  $M_n = S_n - \sum_{i=1}^n \mu_i$ . Since the  $X_i$ 's are independent, we have  $\text{Var}[M_n] = \sum_{i=1}^n \sigma_i^2$ . We claim that

$$V_n = M_n^2 - \sum_{i=1}^n \sigma_i^2 = \left(S_n - \sum_{i=1}^n \mu_i\right)^2 - \sum_{i=1}^n \sigma_i^2$$

is a martingale with respect to the filtration  $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$ . Let's start the computation.

$$\mathbb{E}[V_{n+1}|\mathcal{F}_n] = \mathbb{E}[S_{n+1}^2] - 2\left(\sum_{i=1}^{n+1}\mu_i\right) \mathbb{E}[S_{n+1}|\mathcal{F}_n] + \left(\sum_{i=1}^{n+1}\mu_i\right)^2 - \sum_{i=1}^{n+1}\sigma_i^2$$

$$= S_n^2 + 2S_n\mu_{n+1} + \mathbb{E}[X_{n+1}^2] - 2\left(\sum_{i=1}^{n+1}\mu_i\right)(S_n + \mu_{n+1}) + \left(\sum_{i=1}^{n+1}\mu_i\right)^2 - \sum_{i=1}^{n+1}\sigma_i^2$$

$$= S_n^2 - 2\left(\sum_{i=1}^n\mu_i\right)S_n + \mathbb{E}[X_{n+1}^2] - 2\mu_{n+1}^2 + \left(\sum_{i=1}^n\mu_i\right)^2 + \mu_{n+1}^2 - \sum_{i=1}^{n+1}\sigma_i^2$$

$$= S_n^2 - 2\left(\sum_{i=1}^n\mu_i\right)S_n + \left(\sum_{i=1}^n\mu_i\right)^2 - \sum_{i=1}^n\sigma_i^2$$

$$= V_n.$$

Here we've used the fact that  $S_n$  is  $\mathcal{F}_n$ -measurable and  $X_{n+1}$  is independent of  $\mathcal{F}_n$ . The sequences we want are then

$$b_n = -2\sum_{i=1}^n \mu_i, \qquad c_n = \left(\sum_{i=1}^n \mu_i\right)^2 - \sum_{i=1}^n \sigma_i^2.$$

Problem 2.

(a) Show that if  $(X_n)$  and  $(Y_n)$  are martingales with respect to the same filtration, then  $X_n \vee Y_n$  is a submartingale.

*Proof.* We use the trusty identity

$$X_n \vee Y_n = \frac{1}{2}[(X_n + Y_n) + |X_n - Y_n|].$$

Since the sum of martingales is a martingale and conditional Jensen says the absolute value of a martingale is a submartingale, we have

$$\mathbb{E}[X_{n+1} \vee Y_{n+1} | \mathcal{F}_n] = \frac{1}{2} (\mathbb{E}[X_{n+1} + Y_{n+1} | \mathcal{F}_n] + \mathbb{E}[|X_{n+1} - Y_{n+1}| | \mathcal{F}_n])$$

$$\geq \frac{1}{2} [(X_n + Y_n) + |X_n - Y_n|]$$

$$= X_n \vee Y_n.$$

Hence,  $X_n \vee Y_n$  is a submartingale.

(b) Give an example showing that  $X_n \vee Y_n$  need not be a martingale.

**Problem 3.** Give an example of a martingale  $(X_n)$  such that  $X_n \to -\infty$  a.s.

Solution. Durrett gives a hint to let  $X_n = \xi_1 + \cdots + \xi_n$  for some independent centered  $\xi_i$ 's. The idea is to concentrate most of the mass of  $\xi_i$  around some negative value and put the rest (some summable amount) around some positive value, then apply Borel-Cantelli.

Concretely, let  $\xi_i$  be given by

$$\xi_i = \begin{cases} 2^j & \text{with probability } \frac{1}{2^j} \\ -\frac{1}{1-2^{-j}} & \text{with probability } 1 - \frac{1}{2^j} \end{cases}.$$

Clearly  $\xi_i$  is centered, so  $X_n = \xi_1 + \cdots + \xi_n$  is a martingale. Note that

$$\sum_{i=1}^{\infty} \mathbb{P}[\xi_i = 2^j] = \sum_{i=1}^{\infty} \frac{1}{2^j} = 1 < \infty.$$

By Borel-Cantelli, we have that  $\xi_i = -\frac{1}{1-2^{-j}}$  eventually with probability 1, so  $X_n \to -\infty$  a.s.

**Problem 4.** Let  $(X_n)$  be a martingale that is bounded a.s. either above or below by some constant M. Show that  $\sup_n \mathbb{E}|X_n| < \infty$ .

*Proof.* If  $X_n$  is bounded below, then  $X_n+M$  is a nonnegative martingale. By the martingale convergence theorem,  $X_n+M$  converges almost surely to some limit Y with  $\mathbb{E}|Y|<\infty$ . Consequently,  $X_n$  also converges a.s. to an integrable function, so  $\sup_n \mathbb{E}|X_n|<\infty$ . If  $X_n$  is bounded above, then  $-X_n+M$  is a nonnegative martingale and the same argument works.

**Problem 5.** Let  $Z_1, Z_2, ...$  be nonnegative iid random variables with  $\mathbb{E}[Z_i] = 1$  and  $\mathbb{P}[Z_i = 1] < 1$ . Show that as  $n \to \infty$ ,

$$\prod_{i=1}^{n} Z_i \to 0 \quad \text{a.s.}$$

*Proof.* First note that  $M_n = \prod_{i=1}^n Z_i$  is indeed a martingale:

$$\mathbb{E}[M_{n+1}|\mathcal{F}_n] = \mathbb{E}[Z_{n+1}M_n|\mathcal{F}_n] = M_{n+1},$$

where  $\mathcal{F}_n = \sigma(Z_1, \dots, Z_n)$ . Since  $M_n$  is a nonnegative martingale, it converges to some X a.s. with  $\mathbb{E}[X] \leq \mathbb{E}[Z_1] = 1$ . For any  $\epsilon > 0$  we have

$$\mathbb{P}[|M_{n+1} - M_n| > \epsilon] = \mathbb{P}[M_n|Z_{n+1} - 1| > \epsilon].$$

Now if  $M_n > \sqrt{\epsilon}$  and  $|Z_{n+1} - 1| > \sqrt{\epsilon}$ , then clearly  $M_n |Z_{n+1} - 1| > \epsilon$ . We then have by independence

$$\mathbb{P}[|M_{n+1} - M_n| > \epsilon] \ge \mathbb{P}[M_n > \sqrt{\epsilon}] \cdot \mathbb{P}[|Z_{n+1} - 1| > \sqrt{\epsilon}].$$

Now since  $\mathbb{P}[Z_{n+1}=1] < 1$ , we have that for  $\epsilon$  sufficiently small,  $\mathbb{P}[|Z_{n+1}-1| > \epsilon] \geq \delta > 0$  for some  $\delta$ . Since  $M_n$  converges almost surely, it converges in measure as well, so the left-hand side of the above inequality goes to zero. Since the  $\mathbb{P}[|Z_{n+1}-1| > \sqrt{\epsilon}]$  term is bounded below by a positive constant, we must have that  $\mathbb{P}[M_n > \sqrt{\epsilon}] \to 0$ , so  $M_n \to 0$  in probability. Since  $M_n$  converges almost surely and the a.s. limit is the same as the probability limit,  $M_n \to 0$  a.s.

**Problem 6.** Let  $(X_n)$  be a martingale and let  $\Delta_n = X_n - X_{n-1}$  be the martingale differences. Prove that if  $X_0 = 0$  and  $\sum_{n=1}^{\infty} \Delta_n^2 < \infty$  then  $X_n$  converges in  $L^2$  to some random variable X. (I tried doing the problem as stated and it didn't seem to work. I talked with Xiaowen and she said that we need to assume  $\sum \mathbb{E}[\Delta_n^2] < \infty$ . This is also how it's stated in Durrett.)

*Proof.* For any m, n we have  $X_n - X_m = \sum_{i=m+1}^n \Delta_i$ . From this we deduce

$$\mathbb{E}[|X_n - X_m|^2] = \mathbb{E}\left[\left(\sum_{i=m+1}^n \Delta_i\right)^2\right] = \sum_{i=m+1}^n \mathbb{E}[\Delta_i^2] + 2\sum_{m+1 \le i < j}^n \mathbb{E}[\Delta_i \Delta_j].$$

Since martingale increments are uncorrelated,  $\mathbb{E}[\Delta_i \Delta_j] = \mathbb{E}[\Delta_i] \mathbb{E}[\Delta_j] = 0$ . Since  $\sum \mathbb{E}[\Delta_i^2] < \infty$ , the tail sum goes to zero, so for m, n large enough, the above sum can be made arbitrarily small.  $(X_n)$  is then Cauchy in  $L^2$ , so it converges in  $L^2$  by completeness.

**Problem 7.** Construct a branching process  $(Z_n)$  as follows. Let X be a random variable with mean  $\mu$  and variance  $\sigma^2$ ; it specifies the distribution of the offspring. Set

$$Z_{n+1} = X_1^{(n+1)} + \dots + X_{Z_n}^{(n+1)},$$

to be the size of the population at time n+1, where all  $X_i^{(k)}$  are iid random variables distributed identically with X.

(a) Show that  $Y_n = Z_n/\mu^n$  defines a martingale (with respect to the filtration  $\mathcal{F}_n$  generated by  $X_j^{(k)}$ ,  $1 \le j \le Z_n, k \le n$ .)

*Proof.* We compute.

$$\mathbb{E}[Y_{n+1}|\mathcal{F}_n] = \frac{1}{\mu^{n+1}} \mathbb{E}[Z_{n+1}|\mathcal{F}_n]$$

$$= \frac{1}{\mu^{n+1}} \mathbb{E}[X_1^{(n+1)} + \dots + X_{Z_n}^{(n+1)}|\mathcal{F}_n]$$

$$= \frac{1}{\mu^{n+1}} \cdot \mu Z_n$$

$$= Y_n.$$

(b) Show that

$$\mathbb{E}[Z_{n+1}^2|\mathcal{F}_n] = \mu^2 Z_n^2 + \sigma^2 Z_n.$$

*Proof.* We compute.

$$\mathbb{E}[Z_{n+1}^2|\mathcal{F}_n] = \sum_{i=1}^{Z_n} \mathbb{E}[(X_i^{(n+1)})^2] + 2\sum_{1 \le i < j \le Z_n} \mathbb{E}[X_i^{(n+1)}X_j^{(n+1)}].$$