

# 1 Introduction

The lecture notes complement sections H and I of chapter 2 and appendix B of [1].

## 2 Four inequalities

The three inequality below are commonly used in probability and the theory of stochastic processes.

**Proposition 2.1 (Cauchy-Buniakovskii inequality).** Let  $\xi$  and  $\eta$ , two Borel measurable random variables with finite second moment. Then

$$\mathbb{E} |\xi \eta| \leq (\mathbb{E} \eta^2)^{1/2} (\mathbb{E} \xi^2)^{1/2}$$

*Proof.* By definition

$$\mathbb{E} |\eta \xi| \equiv \int_{\Omega} dP(\omega) |\xi(\omega) \eta(\omega)|$$

Set

$$\tilde{\eta} = \frac{\eta}{\sqrt{\mathbb{E} \eta^2}} \quad \& \quad \tilde{\xi} = \frac{\xi}{\sqrt{\mathbb{E} \xi^2}}$$

then

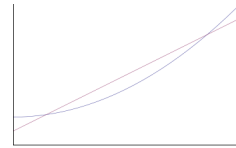
$$0 \leq \mathbb{E} (|\tilde{\xi}| - |\tilde{\eta}|)^2 \Rightarrow 2 \mathbb{E} |\tilde{\eta} \tilde{\xi}| \leq \mathbb{E} |\tilde{\eta}|^2 + \mathbb{E} |\tilde{\xi}|^2 = 1$$

whence the claim. □

**Definition 2.1 (Convexity).** A Borel (-measurable) function is said to be

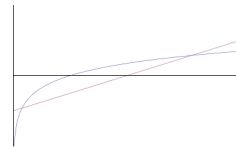
- **convex** if for any two points  $x$  and  $y$  in its domain of definition  $\Omega$  and any  $t \in [0, 1]$

$$f(tx + (1-t)y) \leq tf(x) + (1-t)f(y)$$



- **concave** if for any two points  $x$  and  $y$  in its domain of definition  $\Omega$  and any  $t \in [0, 1]$

$$f(tx + (1-t)y) \geq tf(x) + (1-t)f(y)$$



An example of concave function is the logarithm:

$$\ln(tx + (1-t)y) \leq t \ln x + (1-t) \ln y$$

**Proposition 2.2 (Jensen's inequality).** Let the Borel function  $f(x)$  be downward convex and  $\xi$  a random variable with absolutely convergent first moment. Then

$$f(\mathbb{E} \xi) \leq \mathbb{E} f(\xi)$$

*Proof.* Using the definition of convex function for each  $x_o \in \mathbb{R}$  we can find a number  $g(x_o)$  such that

$$f(x) \geq f(x_o) + g(x_o)(x - x_o)$$

The identifications  $x = \xi$  and  $x_o = \langle \xi \rangle$  yield the proof □

Jensen's inequality is important for moment estimates. We have for example

$$\langle x^2 \rangle^2 \leq \langle x^4 \rangle$$

and more generally

**Proposition 2.3** (*Lyapunov's inequality*). *Let  $0 < s < t$  then*

$$\langle |\xi|^s \rangle^{1/s} \leq \langle |\xi|^t \rangle^{1/t}$$

*Proof.* Define  $r = t/s$  and

$$\eta = |\xi|^s$$

Since  $r > 1$  the function  $f(x) = x^r$  is convex. Jensen's inequality

$$f(\langle \eta \rangle) \leq \langle f(\eta) \rangle$$

made explicit for the present case, yields

$$\langle \eta \rangle^r \leq \langle \eta^r \rangle$$

whence the claim. □

**Proposition 2.4** (*Hölder's inequality*). *Let  $1 < p, q < \infty$  two numbers such that*

$$\frac{1}{p} + \frac{1}{q} = 1$$

*If  $\langle |\xi|^p \rangle$  and  $\langle |\eta|^q \rangle$  are finite then*

$$\langle |\xi \eta| \rangle \leq (\langle |\xi|^p \rangle)^{1/p} (\langle |\eta|^q \rangle)^{1/q}$$

*Proof.* Set

$$\tilde{\xi} = \frac{\xi}{(\langle |\xi|^p \rangle)^{1/p}} \quad \& \quad \tilde{\eta} = \frac{\eta}{(\langle |\eta|^q \rangle)^{1/q}}$$

The concavity of the logarithm

$$\ln(tx + (1-t)y) \geq t \ln x + (1-t) \ln y$$

implies that for any  $x, y > 0$

$$x^{\frac{1}{p}} y^{\frac{1}{q}} \leq \frac{x}{p} + \frac{y}{q}$$

Setting

$$x = |\tilde{\xi}| \quad \& \quad y = |\tilde{\eta}|$$

yields

$$\langle |\tilde{\xi} \tilde{\eta}| \rangle \leq 1$$

which proves the claim. □

### 3 Martingales

A nice presentation of finite dimensional martingales is given in paragraph 11 of chapter 2 of [2].

**Definition 3.1 (Martingale).** Let  $\{\xi_i\}_{i=1}^{\infty}$  a sequence of real-valued random variables, such that

$$\langle |\xi_i| \rangle < \infty \quad \forall i$$

and  $\{\mathcal{F}_k\}_{k=1}^{\infty}$  a growing sequence of  $\sigma$ -algebras (so that  $\mathcal{F}_k \subseteq \mathcal{F}_{k'}$  if  $k \leq k'$ ), such that each of the  $\xi_{i+1}$  is  $\mathcal{F}_i$ -measurable. If

$$\xi_k = \langle \xi_j | \mathcal{F}_k \rangle \quad a.s \quad \forall j \geq k$$

holds true we say that  $\{\xi_i\}_{i=1}^{\infty}$  is a (discrete) martingale.

Examples of martingales:

- Let  $\{\xi_i\}_{i=1}^{\infty}$  a sequence of i.i.d. Bernoulli random variables  $\xi_i \stackrel{d}{=} \xi : \Omega \rightarrow \{-x, x\}$  and

$$P_{\xi}(x) = p$$

then

$$S_k = \sum_{i=1}^k \xi_i - (2p - 1) x k$$

is a martingale with respect to the sequence of  $\sigma$ -algebras with elements  $\mathcal{F}_k$  generated by  $\{\xi_i\}_{i=1}^k$  :

$$\mathcal{F}_1 = \{F_x, F_{-x}\}$$

$$\mathcal{F}_2 = \{F_{x,x}, F_{-x,x}, F_{x,-x}, F_{-x,-x}\}$$

$$\mathcal{F}_3 = \{F_{x,x,x}, F_{-x,x,x}, F_{x,-x,x}, F_{-x,-x,x}, F_{x,x,-x}, F_{-x,x,-x}, F_{x,-x,-x}, F_{-x,-x,-x}\}$$

etc.

Namely

$$\langle S_{k+1} | \mathcal{F}_k \rangle = \sum_{i=1}^{k-1} \xi_i - (2p - 1) x k + \langle \xi_k \rangle = \sum_{i=1}^{k-1} \xi_i - (2p - 1) x (k - 1) = S_{k-1}$$

- Let  $\{\xi_i\}_{i=1}^{\infty}$  a sequence of i.i.d. Bernoulli random variables  $\xi_i \stackrel{d}{=} \xi : \Omega \rightarrow \{-1, 1\}$  and

$$P_{\xi}(x) = p$$

then the sequence of the

$$\eta_i = \left( \frac{1-p}{p} \right)^{S_i}$$

with

$$S_i = \sum_{i=1}^k \xi_i$$

is a martingale with respect to the same sequence of  $\sigma$ -algebras as in the previous example:

$$\begin{aligned} \prec \eta_{k+1} | \mathcal{F}_k \succ &= \left( \frac{1-p}{p} \right)^{S_k} \prec \left( \frac{1-p}{p} \right)^{\xi_{k+1}} \succ \\ &= \left( \frac{1-p}{p} \right)^{S_k} \left[ \left( \frac{1-p}{p} \right) p + \left( \frac{p}{1-p} \right) (1-p) \right] = \left( \frac{1-p}{p} \right)^{S_k} \end{aligned}$$

**Definition 3.2 (Sub-Martingale and Super-Martingale).** Let  $\{\xi_i\}_{i=1}^{\infty}$  a sequence of real-valued random variables, such that

$$\prec |\xi_i| \succ < \infty \quad \forall i$$

and  $\{\mathcal{F}_k\}_{k=1}^{\infty}$  a growing sequence of  $\sigma$ -algebras (so that  $\mathcal{F}_k \subseteq \mathcal{F}_{k'}$  if  $k \leq k'$ ), such that each of the  $\xi_{i+1}$  is  $\mathcal{F}_i$ -measurable. We call  $\{\xi_i\}_{i=1}^{\infty}$  a sub-martingale if

$$\xi_k \leq \prec \xi_j | \mathcal{F}_k \succ \quad \forall j \geq k$$

we call  $\{\xi_i\}_{i=1}^{\infty}$  super-martingale if

$$\xi_k \geq \prec \xi_j | \mathcal{F}_k \succ \quad \forall j \geq k$$

An example of sub-martingale is provided by following proposition.

**Proposition 3.1.** If  $\{\xi_i, \mathcal{F}_i\}_{i=1}^{\infty}$  is a martingale then  $\{|\xi_i|, \mathcal{F}_i\}_{i=1}^{\infty}$  is a sub-martingale

*Proof.*

$$|\xi_i| = | \prec \xi_{i+1} | \mathcal{F}_i \succ | \leq \prec |\xi_{i+1}| | \mathcal{F}_i \succ$$

□

**Theorem 3.1 (Discrete martingale inequalities).** • If  $\{\xi_i\}_{i=1}^{\infty}$  is a submartingale, then

$$P(\max_{1 \leq k \leq n} \xi_k \geq x) \leq \frac{1}{x} \prec \xi_n \vee 0 \succ$$

• If  $\{\xi_i\}_{i=1}^{\infty}$  is a martingale and  $1 < p < \infty$  then

$$\prec \left( \max_{1 \leq k \leq n} |\xi_k| \right)^p \succ \leq \left( \frac{p}{p-1} \right)^p \prec |\xi_k|^p \succ$$

*Proof.* • The event

$$A = \left\{ \omega \in \Omega \mid \max_{1 \leq k \leq n} \xi_k(\omega) \geq x \right\} = \cup_{k=1}^n \{ \omega \in \Omega \mid \xi_k(\omega) \geq x \}$$

admits the decomposition into disjoint events

$$A = \sum_{k=1}^n A_k \quad \text{with} \quad A_k = \cap_{l=1}^{k-1} \{ \omega \in \Omega \mid \xi_l(\omega) < x \} \cap \{ \omega \in \Omega \mid \xi_k(\omega) \geq x \}$$

The decomposition can be formally proven observing that

$$\begin{aligned} & \{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x\} \\ &= \{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup [\{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) < x\}] \end{aligned}$$

by *associativity* of the set union operation can be re-written as

$$\begin{aligned} & \{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x\} \\ &= [\{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) \geq x\}] \cup \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) < x\} \end{aligned}$$

but

$$\begin{aligned} & \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) \geq x\} \subseteq \{\omega \in \Omega | \xi_1(\omega) \geq x\} \Rightarrow \\ & \{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) \geq x\} = \{\omega \in \Omega | \xi_1(\omega) \geq x\} \end{aligned}$$

so that

$$\begin{aligned} & \{\omega \in \Omega | \xi_1(\omega) \geq x\} \cup \{\omega \in \Omega | \xi_2(\omega) \geq x\} \\ &= \{\omega \in \Omega | \xi_1(\omega) \geq x\} + \{\omega \in \Omega | \xi_2(\omega) \geq x, \xi_1(\omega) < x\} \end{aligned}$$

etc. For any of the  $A_k$  by Chebyshev inequality we can write

$$P(A_k) \leq \frac{\xi_k}{x} \chi_{A_k}$$

so that

$$\sum_{k=1}^n P(A_k) = P(A) \leq \sum_{k=1}^n \frac{\xi_k}{x} \chi_{A_k}$$

By definition of sub-martingale we have

$$P(A) \leq \frac{1}{x} \sum_{k=1}^n \langle \xi_n | \mathcal{F}_k \rangle \chi_{A_k} = \frac{1}{x} \sum_{k=1}^n \langle \xi_n \chi_{A_k} | \mathcal{F}_k \rangle$$

But

$$\langle \xi_n \chi_{A_k} | \mathcal{F}_k \rangle = \xi_n \chi_{A_k}$$

so that

$$P(A) \leq \frac{1}{x} \langle \xi_n \chi_A \rangle = \frac{1}{x} \langle (\xi_n \vee 0) \chi_A \rangle \leq \frac{1}{x} \langle \xi_n \vee 0 \rangle$$

The first equality holds because  $x > 0$ , the second inequality because we are adding positive terms.

- By definition

$$P(\max_{1 \leq k \leq n} |\xi_k| \geq x) = 1 - F_{\max_{1 \leq k \leq n} \xi_k}(x) = \tilde{F}_{\max_{1 \leq k \leq n} \xi_k}(x)$$

so that

$$\begin{aligned} & \left\langle \left( \max_{1 \leq k \leq n} |\xi_k| \right)^p \right\rangle = \\ & \int_0^\infty dF_{\max_{1 \leq k \leq n} |\xi_k|}(x) x^p = - \int_0^\infty d\tilde{F}_{\max_{1 \leq k \leq n} \xi_k}(x) x^p = p \int_0^\infty dx P(\max_{1 \leq k \leq n} |\xi_k| \geq x) x^{p-1} \end{aligned}$$

But since the absolute value of a (sub)-martingale is a sub-martingale

$$P\left(\max_{1 \leq k \leq n} \xi_k \geq x\right) \leq \frac{1}{x} \prec |\xi_n| \chi_{\max_{1 \leq k \leq n} \xi_k \geq x} \succ$$

we have

$$\prec \left(\max_{1 \leq k \leq n} |\xi_k|\right)^p \succ \leq p \int dx x^{p-2} \prec |\xi_n| \chi_{\max_{1 \leq k \leq n} |\xi_k| \geq x} \succ$$

Regarding  $|\xi_n|$  as a function  $f_{|\xi_n|}(\bullet)$  of  $\max_{1 \leq k \leq n} |\xi_k|^p$  (they are measurable with respect to the same  $\sigma$ -algebra), we continue by writing

$$\prec \left(\max_{1 \leq k \leq n} |\xi_k|\right)^p \succ \leq p \int_{\Omega} dP f_{|\xi_n|}(x) \int_0^x dy y^{p-2} = \frac{p}{p-1} \int_{\Omega} dP f_{|\xi_n|}(x) x^{p-1}$$

whence finally by Hölder's inequality

$$\prec \left(\max_{1 \leq k \leq n} |\xi_k|\right)^p \succ \leq \left(\frac{p}{p-1}\right) \left[\int_{\Omega} dP f_{|\xi_n|}(x)^p\right]^{\frac{1}{p}} \left[\int_{\Omega} dP x^p\right]^{1-\frac{1}{p}}$$

□

## References

- [1] L.C. Evans, *An Introduction to Stochastic Differential Equations*, lecture notes, <http://math.berkeley.edu/~evans/>
- [2] A. N. Shiryaev, *Probability*, 2nd Ed. Springer (1996), <http://books.google.com/>