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## Exercise sheet 1: Gaussian vectors, random walks, conditioning.

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### Exercise 1 — Gaussian vectors.

Let  $X$  be a random vector in  $\mathbb{R}^n$ . We say that it is a Gaussian vector (i.e. has a multidimensional Gaussian distribution) if for every  $t \in \mathbb{R}^n$ , the r.v.  $\langle t, X \rangle \in \mathbb{R}$  has a (possibly degenerate) Gaussian distribution.

- (1) Recall the parameters, the characteristic function, and (when it exists) the p.d.f. of a Gaussian distribution on  $\mathbb{R}$ .
- (2) Show that  $t \mapsto \mathbb{E}[\langle t, X \rangle]$  is a linear form, and  $(s, t) \mapsto \text{Cov}[\langle s, X \rangle, \langle t, X \rangle]$  is a *positive semi-definite* bilinear form. Let them be represented by  $\langle \cdot, m \rangle$  and  $\langle \cdot, \Sigma \cdot \rangle$ . What would be the coordinates of respectively this vector and this matrix? How would you call them?
- (3) Deduce the (multidimensional) characteristic function of  $X$ , and that the distribution of  $X$  is characterized by the parameters  $m$  and  $\Sigma$ . Show that conversely any vector with a characteristic function of this form is Gaussian.
- (4) Show that a linear transform  $AX$  of a Gaussian vector  $X$  is Gaussian, and compute its parameters.
- (5) Let  $V_1$  and  $V_2$  be two subspaces of  $\mathbb{R}^n$ . Give a necessary and sufficient condition for the independence of the  $\sigma$ -algebras  $\sigma(\langle t, X \rangle, t \in V_1)$  and  $\sigma(\langle t, X \rangle, t \in V_2)$ .
- (6) Build two standard Gaussian variables  $X$  and  $Y$  that are uncorrelated yet not independent (they obviously do not form a Gaussian vector !)
- (7) Show that the vector  $(X_1, \dots, X_n)$  with  $X_1, \dots, X_n$  independent standard Gaussian variables, is Gaussian. Use it to build a Gaussian vector with arbitrary parameters. Deduce its p.d.f. when it has one.

### Exercise 2 — Limits of Gaussian variables.

Let  $(X_n)_{n \geq 0}$  be a sequence of Gaussian variables.

- (1) Give a necessary and sufficient condition for convergence in distribution, show that the limit is always Gaussian, and determine its parameters.
- (2) If  $X_n \rightarrow X$  in probability, show that for every  $p \geq 1$ ,  $X_n \rightarrow X$  in  $L^p$ .

### Exercise 3 — Some estimates.

Let  $(X_n)_{n \geq 0}$  be the simple symmetric random walk.

- (1) Compute the asymptotic for the probability of  $X_{2n} = 0$ .
- (2) Compute the probability that the first return to zero is at time  $2n$ . Compute its asymptotic.
- (3) Deduce the asymptotic for the probability that the first return to zero is after time  $n$ .

(4) Deduce the asymptotic of  $\mathbb{E}[|X_n|]$ .

**Exercise 4** — *Maximum and hitting times.*

Let  $(X_n)_{n \geq 0}$  be the simple symmetric random walk and  $M_n = \max_{1 \leq i \leq n} X_i$ .

- (1) Find a limit in distribution (after suitable rescaling) for the pair  $(X_n, M_n)$ .
- (2) For  $k \geq 0$ , let

$$\tau_k = \min\{n \geq 0 : X_n = k\}.$$

Show that  $(\tau_k)_k$  is a random walk. What is  $\mathbb{P}(\tau_k \leq n)$ ? Find a limit in distribution (after suitable rescaling) for  $\tau_k$ .

**Exercise 5** — *Conditioning and independence.*

Let  $\mathcal{G}$  be a  $\sigma$ -algebra,  $X \in \mathcal{G}$  and  $Y \perp\!\!\!\perp \mathcal{G}$  be two random variables, and  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  such that  $f(X, Y) \in L^1$ . Compute  $\mathbb{E}[f(X, Y) | \mathcal{G}]$ . Deduce the conditional distribution of  $f(X, Y)$  given  $\mathcal{G}$ .

**Exercise 6** — *Gaussian conditional distribution and Bayesian statistics 101.*

Let  $(X, Y)$  be a non-degenerate centered Gaussian vector in  $\mathbb{R}^2$  with covariance matrix  $\begin{pmatrix} \sigma_x^2 & \rho \\ \rho & \sigma_y^2 \end{pmatrix}$ .

- (1) Compute the conditional distribution of  $X$  given  $Y$ .
- (2) If you want, look up on Wikipedia the generalized version of this, where  $X \in \mathbb{R}^n$ ,  $Y \in \mathbb{R}^m$ .
- (3) Let  $\theta \sim \mathcal{N}(0, \tau^2)$  and  $X_1, \dots, X_n$  i.i.d. of distribution  $\mathcal{N}(\theta, \sigma^2)$  given  $\theta$ . In other terms,  $X_i = \theta + Y_i$  where  $Y_i$  are i.i.d, independent of  $\theta$ , and  $Y_i \sim \mathcal{N}(0, \sigma^2)$ . What is the conditional distribution of  $\theta$  given  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$ ?
- (4) Take the large  $\sigma$ , small  $\sigma$ , large  $\tau$ , small  $\tau$  limit of this and interpret it.
- (5) Find a "real-life" situation modelled by this.
- (6) (★) What about the conditional distribution of  $\theta$  given  $(X_1, \dots, X_n)$ ?