

GAUSSIAN PROCESSES
EXERCISE SHEET 6: ENTROPY AND CLT(II)

Exercise 1 (Smoothed convergence implies convergence). Let X_1, X_2, \dots be random variables of zero mean and unit variance. Let Y_1, Y_2, \dots be i.i.d. standard Gaussians, independent from X_1, X_2, \dots and define $\hat{X}_n = X_n + Y_n$. Suppose that the sequence of variables \hat{X}_n converges in law to a centered Gaussian of variance 2. Prove that X_n converges in law to a standard Gaussian.

Exercise 2 (CLT using Stein's lemma). Use Stein's lemma to prove the usual CLT for sums of i.i.d. random variables of finite variance. You may proceed in two steps:

- (1) By showing tightness, prove that there exist subsequences that converge in law.
- (2) Characterise the limit using Stein's lemma.

Assume some extra conditions on X_i if it helps!

Exercise 3 (Equivalence of EPI). Let X and Y be independent random variables with finite differential entropies, and denote by $h(\cdot)$ the differential entropy. Prove that the following statements are equivalent:

- (1) $\exp(2h(X + Y)) \geq \exp(2h(X)) + \exp(2h(Y))$.
- (2) For each $\lambda \in (0, 1)$,

$$h(\sqrt{\lambda}X + \sqrt{1-\lambda}Y) \geq \lambda h(X) + (1-\lambda)h(Y).$$

- (3) Let X' and Y' be independent Gaussian random variables with the same entropies as X and Y , respectively. Then

$$h(X + Y) \geq h(X' + Y').$$

Exercise 4 (Discretisation of the Ornstein–Uhlenbeck process). The (stationary) Ornstein–Uhlenbeck SDE is commonly written

$$dX_t = -X_t dt + \sqrt{2} dB_t.$$

To make this rigorous via discretisation, consider the time step $\epsilon > 0$ and define a discrete-time process X^ϵ on the grid $\{0, \epsilon, 2\epsilon, \dots\}$ by

$$X^\epsilon(t + \epsilon) - X^\epsilon(t) = -\epsilon X^\epsilon(t) + \sqrt{2\epsilon} Y_t,$$

where $(Y_t)_{t \in \{0, \epsilon, 2\epsilon, \dots\}}$ are i.i.d. standard Gaussian random variables and $X^\epsilon(0) = X_0$ is given.

- (a) Show that for $n \in \mathbb{N}$,

$$X^\epsilon(n\epsilon) = (1 - \epsilon)^n X_0 + \sqrt{2\epsilon} \sum_{k=1}^n (1 - \epsilon)^{n-k} Y_{k\epsilon}.$$

Use this to show that for fixed $t > 0$ and $\epsilon \downarrow 0$ (with $n = \lfloor t/\epsilon \rfloor$) the law of $X^\epsilon(t)$ converges to

$$X_t \stackrel{d}{=} e^{-t} X_0 + \sqrt{1 - e^{-2t}} Z,$$

where $Z \sim \mathcal{N}(0, 1)$ is independent of X_0 . Furthermore, try to explain why we can write

$$X_t = e^{-t} X_0 + \sqrt{2} \int_0^t e^{-(t-s)} dB_s,$$

(b) Let $p(t, x)$ denote the density of X_t (assume it exists and is smooth). Conclude that p satisfies the equation

$$\frac{\partial}{\partial t} p(t, x) = \frac{\partial}{\partial x} (x p(t, x)) + \frac{\partial^2}{\partial x^2} p(t, x),$$

and that,

$$\frac{d}{dt} \mathbb{E}[f(X_t)] = \mathbb{E}[f''(X_t)] - \mathbb{E}[X_t f'(X_t)].$$

(You can add some suitable conditions on f .)

Exercise 5 (Cramér's theorem (Not Examinable)). Let X and Y be independent random variables such that $X + Y$ is Gaussian. Prove that both X and Y are Gaussian.