

# Course no. 80573: Probability and stochastic processes, 2004-2005

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April 16, 2005

## Problem set 1: Kolmogorov theorems, Markov processes, Brownian motion

### 1 Gaussian processes

A finite family of random variables  $\xi_1, \dots, \xi_n$  is called a (nondegenerate) Gaussian family or a (nondegenerate  $n$ -dimensional) Gaussian random vector if there exists a symmetric positively definite matrix  $A = (a_{ij})$  (i.e.  $\sum_{i,j=1}^n a_{ij}x_i x_j > 0$  provided  $\sum_{i=1}^n |x_i| > 0$ ) and a vector  $a = (a_1, \dots, a_n)$  such that

$$P\{(\xi_1, \dots, \xi_n) \in \Gamma\} = \int_{\Gamma} \frac{1}{(2\pi)^{n/2}(\det A)^{1/2}} e^{-\frac{1}{2}(A^{-1}(x-a), (x-a))} dx, \quad (1)$$

where  $(\cdot, \cdot)$  denotes the scalar (inner) product of vectors. More generally, A finite family of random variables  $\xi_1, \dots, \xi_n$  is called a Gaussian family or a Gaussian random vector (may be degenerate) if for any constants  $b_1, \dots, b_n$  the sum  $\sum_{i=1}^n b_i \xi_i$  is a normal (Gaussian) random variable with a mean  $\mu$  and a variance  $\sigma^2$  which could be zero. In particular, a constant (i.e. nonrandom) vector is a (degenerate) Gaussian vector (which is excluded by the first definition which introduces nondegenerate Gaussian random vectors). We call also a Gaussian random vector degenerate if it has linearly dependent (with constant coefficients) components. An infinite family of random variables is called Gaussian if any finite subfamily is Gaussian. In particular,  $\xi_t, t \in T \subset \mathbb{R}$  is a Gaussian process if it is a Gaussian family.

#### Problems.

1. Prove that if (1) holds true then

$$E\xi_i = a_i \text{ and } \text{Cov}(\xi_i, \xi_j) = E(\xi_i - a_i)(\xi_j - a_j) = a_{ij}.$$

2. Prove that if  $\xi = (\xi_1, \dots, \xi_n)$  is a Gaussian (degenerate or not) random vector and  $L$  is a nonrandom invertible  $n \times n$  matrix then  $L\xi$  is a Gaussian (degenerate or not) random vector.

3. Let  $a(t, s)$  be a function on  $\mathbb{R}_+ \times \mathbb{R}_+ = [0, \infty)$  such that  $a(s, t) = a(t, s)$  and for any  $t_1, \dots, t_n, s_1, \dots, s_n \in \mathbb{R}_+$  and  $x_1, \dots, x_n \in \mathbb{R}$  with  $\sum_{i=1}^n |x_i| > 0$  we have

$$\sum_{i,j=1}^n a(t_i, s_j) x_i x_j > 0.$$

Prove (using Kolmogorov's theorem) that there exists a Gaussian process  $\xi_t, t \geq 0$  such that  $E\xi_t = 0, \forall t \geq 0$  and  $E\xi_t \xi_s = a(t, s) \forall t, s \geq 0$ .

4. Prove that if  $a(t, s) = \min(t, s) - ts$  then there exists a Gaussian process  $\xi_t, 1 > t > 0$  with almost surely continuous (even Hölder continuous) paths which satisfies  $E\xi_t = 0, \forall 1 > t > 0$  and  $E\xi_t \xi_s = a(t, s) \forall 1 > t, s > 0$ .

5. Suppose that  $a(t, s) = b(\min(t, s)) > 0$  where  $b(u)$  is a positive strictly increasing function. Prove that if  $\xi_t, t \geq 0$  is the Gaussian process constructed with such  $a(t, s)$  then for any  $0 \leq t_1 < t_2 < \dots < t_n$  the increments  $\xi_{t_2} - \xi_{t_1}, \xi_{t_3} - \xi_{t_2}, \dots, \xi_{t_n} - \xi_{t_{n-1}}$  are independent. Show that if  $b(u) = u$  we obtain the Brownian motion.

## 2 Brownian motion

Define the one dimensional Brownian motion  $B(t), t \geq 0$  as a Markov process on  $\mathbb{R} = (-\infty, \infty)$  with the transition probability

$$P(t, x, \Gamma) = \int_{\Gamma} p(t, y - x) dy$$

where  $p(t, z) = (2t\pi)^{-1/2} e^{-\frac{z^2}{2t}}$ .

### Problems.

6. Prove the Chapman-Kolmogorov equality

$$\int p(t, y - x) p(s, z - y) dy = p(s + t, z - x).$$

7. Assume that  $B(0) = 0$  and prove that for  $0 < t_1 < t_2 < \dots < t_n$ ,

$$B(t_1), B(t_2) - B(t_1), \dots, B(t_n) - B(t_{n-1})$$

are independent random variables. (Hint: represent this vector as a linear transformation of the vector  $B(t_1), \dots, B(t_n)$ ).

8. Prove that  $B(t) - B(s)$  is the Gaussian (normal) random variable with mean zero and the variance  $t - s$ .

9. Prove that with probability one  $B(t), t \geq 0$  is not Hölder continuous for any exponent  $\gamma > 1/2$ .

10\*. Prove that there exists no modification  $\tilde{B}$  of  $B$  (i.e.  $P\{B(t) = \tilde{B}(t)\} = 1$  for each  $t$ ) such that  $\tilde{B}(t)$  is Hölder continuous also with the exponent  $\gamma = 1/2$ . In fact, it is possible to show that with probability one there exist times  $t = t(\omega)$  depending on  $\omega$  where  $B(t, \omega)$  is Hölder continuous with the exponent  $\gamma = 1/2$  (which is not true for  $\gamma > 1/2$ ).

### 3 Poisson process

Define the Poisson process  $N(t)$ ,  $t \geq 0$  as a Markov chain (process) on  $\mathbb{N} = \{0, 1, \dots\}$  with the transition probability

$$P(t, n, \Gamma) = \sum_{m \in \Gamma} p(t, m - n)$$

where  $p(t, m - n) = e^{-\lambda t} \frac{(\lambda t)^{m-n}}{(m-n)!}$  if  $m \geq n$  and  $p(t, m - n) = 0$  if  $m < n$ .

#### Problems.

11. Prove the Chapman-Kolmogorov equality

$$\sum_n p(t, n - k)p(s, m - n) = p(t + s, m - k).$$

12. Assume that  $N(0) = 0$  and prove that for  $0 < t_1 < t_2 < \dots < t_n$ ,

$$N(t_1), N(t_2) - N(t_1), \dots, N(t_n) - N(t_{n-1})$$

are independent random variables.

### 4 Cauchy process.

Define the Cauchy process  $\xi(t)$ ,  $t \geq 0$  as a Markov process on  $\mathbb{R}$  with the transition probability

$$P(t, x, \Gamma) = \int_{\Gamma} p(t, y - x) dy$$

where  $p(t, z) = \frac{t}{\pi(t^2 + z^2)}$ .

#### Problems.

- 13\*. Prove the Chapman-Kolmogorov equality

$$\int p(t, y - x)p(s, z - y) dy = p(t + s, z - x).$$

(Hint: it is easier to integrate here using the complex variable integration).

14. Assume that  $\xi(0) = 0$  and prove that for  $0 < t_1 < t_2 < \dots < t_n$ ,

$$\xi(t_1), \xi(t_2) - \xi(t_1), \dots, \xi(t_n) - \xi(t_{n-1})$$

are independent random variables.

### 5 Stationary distribution

15. Let  $\xi_n$ ,  $n = 0, 1, 2, \dots$  be a Markov chain on the state space  $\mathbb{N} = 0, 1, 2, \dots$  with the transition probability matrix  $P = (p_{ij}, i, j \in \mathbb{N})$ . Prove that if

$$p_{ij} \geq a_j p_{kj}$$

for all  $i, j, k \in \mathbb{N}$  and some positive  $a_j$  then there exists a unique stationary distribution  $\eta$  which is an infinite probability vector satisfying  $\eta P = \eta$ .