Random processes: Theory and applications from physics to finance
 SS 2008

 Problem set 4
 2008/04/02

#### 1. Process with independent increments

By a process with independent increments we mean a family of random variables  $\hat{\mathbf{x}}(t)$  depending on the continuous time parameter t such that the increments

$$\hat{\mathbf{x}}(t_{k+1}) - \hat{\mathbf{x}}(t_k)$$

are mutually independent for any finite set of  $t_1 < t_2 < \ldots < t_n$ .

In the following, we consider a time-homogenous process with independent increments using the forward equation

$$\frac{\partial}{\partial t} p_{1|1}(\mathbf{x}, t \mid \mathbf{x}_0, t_0) = \int \mathrm{d}^r x \, \Gamma(\mathbf{x}, \mathbf{x}'; t) \, p_{1|1}(\mathbf{x}', t \mid \mathbf{x}_0, t_0) \,,$$

where the master operator can be written in the form

$$\Gamma(\mathbf{x}, \mathbf{x}', t) = \Gamma(\mathbf{x} - \mathbf{x}') = \gamma \left[ \rho(\mathbf{x} - \mathbf{x}') - \delta(\mathbf{x} - \mathbf{x}') \right] \,,$$

with  $\gamma$  being the jump rate and  $\rho$  the probability density function of the jump widths.

# 1.1. Conditional probability

Show that the conditional probability  $p_{1|1}(\mathbf{x}, t | \mathbf{x}_0, t_0)$   $(t_0 \leq t)$  consequently also depends only on the spatial difference  $\mathbf{x} - \mathbf{x}_0$  and time dependence  $t - t_0$ .

# 1.2. Solution for transition probability

Show that in terms of the Fourier transforms with respect to  $\mathbf{x}$ ,

$$p_{1|1}(\mathbf{k}, t) := \int d^r x \, \mathrm{e}^{\mathrm{i}\mathbf{k}\cdot\mathbf{x}} \, p_{1|1}(\mathbf{x}, t \mid \mathbf{0}, 0) \quad \text{and}$$
$$\rho(\mathbf{k}) := \int d^r x \, \mathrm{e}^{\mathrm{i}\mathbf{k}\cdot\mathbf{x}} \, \rho(\mathbf{x}) \,,$$

the solution of the forward equation is given by

$$p_{1|1}(\mathbf{k},t) = \exp\{\gamma \left[\rho(\mathbf{k}) - 1\right]t\}$$
.

Why does this imply that for asymptotic times  $t \to \infty$ , the process becomes Gaussian (provided that  $\rho(\mathbf{k})$  is sufficiently smooth around  $\mathbf{k} = \mathbf{0}$ )?

#### 1.3. Interpretation as compound Poisson distribution

Show that the back transformation yields

$$p_{1|1}(\mathbf{x},t \mid \mathbf{0},0) = \sum_{n=0}^{\infty} p_n(t) \left\langle \delta\left(\mathbf{x} - \sum_{i=1}^{n} \hat{\mathbf{x}}_i\right) \right\rangle_{\rho}$$

where  $p_n(t) = e^{-\gamma t} (\gamma t)^n / n!$ , the jumps  $\hat{\mathbf{x}}_i$  are independent, identically distributed according to the distribution  $\rho(\mathbf{x})$  and  $\langle \ldots \rangle_{\rho}$  denotes the corresponding average. Interpret the result in connection with a distribution from problem set 3.

## 2. Non-Gaussian white noise

Now we consider a process  $\hat{\xi}(t)$  which fulfills

$$\langle \xi(t) \rangle = \Gamma_1 , \langle \langle \hat{\xi}(t) \hat{\xi}(t') \rangle \rangle = \Gamma_2 \, \delta(t - t')$$

but is not necessarily Gaussian. Thus, the higher order cumulants do not need to vanish, but we assume that they are  $\delta$ -correlated according to

$$\langle\!\langle \hat{\xi}(t_1)\hat{\xi}(t_2)\dots\hat{\xi}(t_m)\rangle\!\rangle = \Gamma_m \,\delta(t_1-t_2)\,\delta(t_1-t_3)\dots\delta(t_1-t_m)$$

for  $m \geq 2$ . The  $\Gamma_m$  are arbitrary time-independent constants.

#### 2.1. Characteristic functional

Calculate the generating functional  $\left\langle \exp\left[i\int dt k(t)\hat{\xi}(t)\right]\right\rangle$  and its logarithm.

# 2.2. Characteristic function of integrated noise process

Mathematically better behaved is the integrated process

$$\hat{W}(t) = \int_{0}^{t} \hat{\xi}(t') \mathrm{d}t'.$$

Convince yourself that the  $\hat{W}(t)$  is a process with independent increments.

Derive the characteristic function for the increments of the integrated process from the characteristic functional of the previous section by using  $k(t) = k \ \theta(t - t_1)\theta(t - t_2)$ .

#### 2.2.1. Compound Poisson process

Discuss your results in terms of the compound Poisson process derived in exercise 1. In particular, derive the relation of the  $\Gamma_n$  with the jump rate and the moments of the jump width distribution.

Note that, vice versa, a white noise process can be constructed by taking any process with independent increments and differentiating it.

# 2.3. Example: Poisson process

Show that the Poisson process

$$p_{1|1}(n_2, t_2 \mid n_1, t_1) = \frac{(t_2 - t_1)^{n_2 - n_1}}{(n_2 - n_1)!} e^{-(t_2 - t_1)}$$

with  $p_1(n,0) = \delta_{n,0}$  has independent increments. Show that this process provides white noise with  $\Gamma_m = 1$  for all m.