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Series 1

- 1. (i) See the MATLAB file Normal.m
 - (ii) See the MATLAB files Poisson.m and PoissonPlot.m.

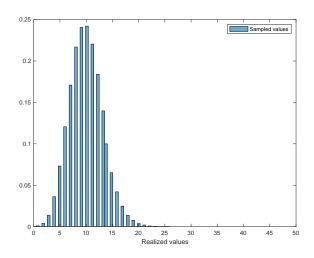


Abbildung 1: Result of a call of the MATLAB function PoissonPlot().

2. (i) Let $t \in \mathbb{T}$. We need to show that $X_t \in \mathcal{M}(\mathbb{F}_t^X, \mathcal{S})$. By definition of \mathbb{F}^X , we have that

$$\mathbb{F}_t^X = \sigma_{\Omega}((X_s)_{s \in \mathbb{T} \cap (-\infty, t]}) = \sigma_{\Omega}(\{X_s^{-1}(A) \in \mathcal{P}(\Omega) : A \in \mathcal{S}, s \in \mathbb{T} \cap (-\infty, t]\}).$$
(1)

In particular,

$$\mathbb{F}_t^X \supseteq \{X_t^{-1}(A) \in \mathcal{P}(\Omega) : A \in \mathcal{S}\}.$$

Therefore, for every $A \in \mathcal{S}$ it holds that

$$X_t^{-1}(A) \in \mathbb{F}_t^X,$$

which concludes the proof.

(ii) Let $t \in \mathbb{T}$. We want to show that $\mathbb{F}_t^X \subseteq \mathbb{F}_t$. By hypothesis, $X_s \in \mathcal{M}(\mathbb{F}_s, \mathcal{S})$ for every $s \in \mathbb{T}$. Then, for every $s \in \mathbb{T}$ and for every $A \in \mathcal{S}$ it holds that

$$X_s^{-1}(A) \in \mathbb{F}_s.$$

Furthermore, by definition of filtration we know that $\mathbb{F}_s \subseteq \mathbb{F}_t$ for every $s \in \mathbb{T} \cap (-\infty, t]$. Therefore,

$$\{X_s^{-1}(A) \in \mathcal{P}(\Omega) : A \in \mathcal{S}, s \in \mathbb{T} \cap (-\infty, t]\} \subseteq \mathbb{F}_t,$$

which by equation (1) implies that $\mathbb{F}_t^X \subseteq \mathbb{F}_t$.

(iii) Let $T \in (0, +\infty)$. We consider the stochastic process $X : [0, T] \times \Omega \to \mathbb{R}$ defined by $X(t, \omega) = tZ(\omega)$, where $Z : \Omega \to \mathbb{R}$ is a standard normal random variable. Clearly, X is a stochastic process with continuous sample paths. By definition of \mathbb{F}^X , it holds that

$$\mathbb{F}_0^X = \sigma_{\Omega}(X_0) = \sigma_{\Omega}(\{X_0^{-1}(A) \in \mathcal{P}(\Omega) : A \in \mathcal{B}(\mathbb{R})\}) = \sigma_{\Omega}(\{\Omega, \emptyset\}) = \{\Omega, \emptyset\}.$$

On the other hand, for every $t \in (0, T]$ it holds that

$$\mathbb{F}_{t}^{X} = \sigma_{\Omega} (\{X_{s}^{-1}(A) \in \mathcal{P}(\Omega) : A \in \mathcal{B}(\mathbb{R}), s \in \mathbb{T} \cap (-\infty, t]\})$$
$$= \sigma_{\Omega} (\{X_{s}^{-1}(sB) \in \mathcal{P}(\Omega) : B \in \mathcal{B}(\mathbb{R}), s \in \mathbb{T} \cap (-\infty, t]\})$$
$$= \sigma_{\Omega} (\{Z^{-1}(B) \in \mathcal{P}(\Omega) : B \in \mathcal{B}(\mathbb{R})\}) = \sigma_{\Omega} (Z),$$

and consequently

$$\mathbb{F}_0^{X^+} = \bigcap_{t \in (0,T]} \mathbb{F}_t^X = \sigma_{\Omega}(Z).$$

Therefore, $\mathbb{F}_0^X \neq \mathbb{F}_0^{X^+}$ and the filtration \mathbb{F}^X generated by X is not right-continuous.

3. (i) Let $t \in [0,T]$ and $s \in (0,T]$. By item (iii) in the definition of standard Brownian motion and by Proposition 0.4.15. in Chapter 0, it holds that

$$W_t \sim \mathcal{N}_{0,tI_{\mathbb{R}^m}} \sim \frac{\sqrt{t}}{\sqrt{s}} \mathcal{N}_{0,sI_{\mathbb{R}^m}} \sim \frac{\sqrt{t}}{\sqrt{s}} W_s.$$

(ii) We first show that W has P-independent increments. Let $n \in \mathbb{N}$ and let $t_0, \ldots, t_n \in [0, T]$ with $t_0 < t_1 < \ldots < t_n$. We have to show that the random variables

$$(W_{t_1} - W_{t_0}), (W_{t_2} - W_{t_1}), \dots, (W_{t_n} - W_{t_{n-1}})$$

are P-independent. First, by item (i) in Exercise 2, we know that W is $\mathbb{F}^W/\mathcal{B}(\mathbb{R}^m)$ -adapted. This fact together with the inclusions $\mathbb{F}^W_{t_0} \subseteq \ldots \subseteq \mathbb{F}^W_{t_{n-1}}$ implies that

$$\sigma_{\Omega}((W_{t_1} - W_{t_0}), (W_{t_2} - W_{t_1}), \dots, (W_{t_{n-1}} - W_{t_{n-2}})) \subseteq \mathbb{F}_{t_{n-1}}^W$$

Consequently, by item (iv) in the definition of standard Brownian motion,

$$\sigma_{\Omega}((W_{t_1} - W_{t_0}), (W_{t_2} - W_{t_1}), \dots, (W_{t_{n-1}} - W_{t_{n-2}}))$$
 and $\sigma_{\Omega}(W_{t_n} - W_{t_{n-1}})$

are P-independent σ -algebras. We can then repeat the above reasoning t_{n-2} times to obtain that the σ -algebras

$$\sigma_{\Omega}(W_{t_1}-W_{t_0}), \sigma_{\Omega}(W_{t_2}-W_{t_1})\dots, \sigma_{\Omega}(W_{t_n}-W_{t_{n-1}})$$

are P-independent. Therefore, for every choice of $A_0, \ldots, A_{n-1} \in \mathcal{B}(\mathbb{R}^m)$, it holds that

$$P(\{W_{t_1} - W_{t_0} \in A_0\} \cap \dots \cap \{W_{t_n} - W_{t_{n-1}} \in A_{n-1}\})$$

= $P(W_{t_1} - W_{t_0} \in A_0) \cdot \dots \cdot P(W_{t_n} - W_{t_{n-1}} \in A_{n-1}),$

which proves the first part of item (ii). Furthermore, by item (iii) in the definition of standard Brownian motion, for every $t \in [0,T]$ and $h \in [0,T-t]$ it holds that $W_{t+h} - W_t$ is $\mathcal{N}_{0,hI_{\mathbb{R}^m}}$ -distributed, which shows that W has stationary increments and concludes the proof.

(iii) Let $s, t \in [0, T]$. Without loss of generality, we suppose that $s = \min\{s, t\}$. Since $\mathbb{E}[W_s] = \mathbb{E}[W_t] = 0$ and $W_0 = 0$, we have that

$$Cov_P(W_s, W_t) = \mathbb{E}[W_s W_t] = \mathbb{E}[(W_t - W_s)(W_s - W_0)] + \mathbb{E}[W_s^2]$$
$$= \mathbb{E}[(W_t - W_s)]\mathbb{E}[(W_s - W_0)] + \mathbb{E}[W_s^2]$$
$$= \mathbb{E}[W_s^2],$$

where the last two equalities follow by the fact that W has P-independent increments and increments are normally distributed random variables with zero mean. Finally, since W_s is $\mathcal{N}_{0,s}$ -distributed and $\operatorname{Var}[W_s] = \mathbb{E}[W_s^2] - \mathbb{E}[W_s]^2 = \mathbb{E}[W_s^2]$, it holds that

$$Cov_P(W_s, W_t) = s = \min\{s, t\},\$$

which proves item (iii).

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