



NTNU
Norwegian University of
Science and Technology

Week 44: Lecture 2
Gaussian processes and covariance functions

Geir-Arne Fuglstad

October 28, 2020

Information

- **No lectures** on November 2 and 4 (week 45)
- We finish the curriculum on November 9.
- November 11: we look at winter 2019 exam together.
- November 16: Questions and answers
- November 18: **No lecture**

Section 4 (Note)

Definition (Def. 1)

The stochastic process $\{X(t) : t \geq 0\}$ with state space \mathbb{R} is called a **Gaussian process** on $[0, \infty)$ if for all $m \geq 1$, for all $0 \leq t_1 < t_2 < \dots < t_m$,

$$(X(t_1), X(t_2), \dots, X(t_m))$$

has an m -dimensional multivariate Gaussian distribution.

Example 1

Prove that Brownian motion is a Gaussian process on $[0, \infty)$.

Theorem (Thm. 1)

A Gaussian process $\{X(t) : t \in T\}$ is fully determined by two functions:

- 1) a **mean function** $m : T \rightarrow \mathbb{R}$ so that

$$E[X(t)] = m(t), \quad t \in T.$$

- 2) a **covariance function** $C : T \times T \rightarrow \mathbb{R}$ so that

$$\text{Cov}[X(t_1), X(t_2)] = C(t_1, t_2), \quad t_1, t_2 \in T.$$

Example 2

Find the mean function and the covariance function of Brownian motion with variance parameter $\sigma^2 > 0$. Are these enough to fully specify Brownian motion with variance parameter $\sigma^2 > 0$?

Definition (Def. 2)

Let $\{X(t) : t \in T\}$ be a stochastic process. The **correlation function** $r : T \times T \rightarrow [-1, 1]$ is defined by

$$\begin{aligned} r(t_1, t_2) &= \text{Corr}[X(t_1), X(t_2)] \\ &= \frac{\text{Cov}[X(t_1), X(t_2)]}{\sqrt{\text{Var}[X(t_1)]\text{Var}[X(t_2)]}} \\ &= \frac{C(t_1, t_2)}{\sqrt{C(t_1, t_1)C(t_2, t_2)}}, \end{aligned}$$

where $C : T \times T \rightarrow \mathbb{R}$ is the covariance function.

Definition (Def. 3)

A stochastic process on $[0, \infty)$ is **stationary** if

1) $m(t) = \mu_0$ for $t \in [0, \infty)$

2) $C(t_1, t_2) = \sigma^2 r(|t_1 - t_2|)$ for $t_1, t_2 \in [0, \infty)$

Here $\sigma^2 > 0$ is called the **marginal variance**, and

$r : [0, \infty) \rightarrow [-1, 1]$ is called a **stationary correlation function** and satisfies $r(0) = 1$.

Common stationary covariance functions

— Exponential:

$$C(t_1, t_2) = \sigma^2 \exp(-\phi_E |t_1 - t_2|), \quad t_1, t_2 \in \mathbb{R}.$$

— Gaussian:

$$C(t_1, t_2) = \sigma^2 \exp(-\phi_G (t_1 - t_2)^2), \quad t_1, t_2 \in \mathbb{R}.$$

— Matérn-type:

$$C(t_1, t_2) = \sigma^2 (1 + \phi_M |t_1 - t_2|) \exp(-\phi_M |t_1 - t_2|), \quad t_1, t_2 \in \mathbb{R}.$$

Example 3

Demonstration of simulation and covariance functions using R.

Section 4.2 (Note)

Simulation of Gaussian process

Input:

- $[a, b]$: interval of interest
- m : mean function
- C : covariance function

Algorithm:

1. make grid $a = t_1 < t_2 < \dots < t_n = b$
2. set $\mu = (m(t_1), m(t_2), \dots, m(t_n))$
3. set $\Sigma_{ij} = C(t_i, t_j)$ for $i, j = 1, 2, \dots, n$
4. draw $\mathbf{x} \sim \mathcal{N}_n(\mu, \Sigma)$

Output:

We have simulated values $\mathbf{x} = (x(t_1), x(t_2), \dots, x(t_n))$.

Section 4.3 (Note)

Example 4

Let $\{B(t) : t \geq 0\}$ be standard Brownian motion, and let $X(t) = (B(t) | B(1) = 1)$ for $0 < t < 1$. Find $\mu(t) = E[X(t)]$ and $\sigma(t)^2 = \text{Var}[X(t)]$ for $0 < t < 1$.

Conditional Gaussian processes

Let $\{X(t) : t \geq 0\}$ be a Gaussian process. Assume that the process has been observed at locations $B = \{s_1 < s_2 < \dots < s_m\}$ and let $\mathbf{X}_B = (X(s_1), X(s_2), \dots, X(s_m))$.

Then for any set of locations $A = \{t_1 < t_2 < \dots < t_n\}$, let $\mathbf{X}_A = (X(t_1), X(t_2), \dots, X(t_n))$. We have

$$\mathbf{X}_A | \mathbf{X}_B = \mathbf{x}_B \sim \mathcal{N}_n(\boldsymbol{\mu}_C, \Sigma_C),$$

where

$$\begin{aligned}\boldsymbol{\mu}_C &= \boldsymbol{\mu}_A + \Sigma_{AB} \Sigma_{BB}^{-1} (\mathbf{x}_B - \boldsymbol{\mu}_B) \\ \Sigma_C &= \Sigma_{AA} - \Sigma_{AB} \Sigma_{BB}^{-1} \Sigma_{BA}.\end{aligned}$$

Example 5

Let $\{B(t) : t \geq 0\}$ be standard Brownian motion, and let $X(t) = (B(t) | B(1) = 1)$. Calculate $\Pr\{X(1/2) > 1\}$.