PPE Kriging

Fall Semester 2020

Subject NR. 103-0787-00 L

## **HOMEWORK 2: RANDOM FIELDS**

## 1 Contents and Aim of Homework

- Understanding stochastic processes as special cases of Random Fields
- Calculating Expectations and Covariances to evaluate if a Random Field is stationary
- Get a feeling for how Covariances typically look as a function of distance between the locations associated to Random Variables
- Understand Gaussian Processes and build your own ones

**Note:** This homework is a group homework (2-3 people in each group). It is to be handed in till noon of 19<sup>th</sup> of November. E-mail: jemil.butt@geod.baug.ethz.ch

#### 2 Tasks

#### 2.1 Stationarity and Warm-up

a) The "Wiener Process" is just integrated white noise. Please formulate the equation for a discrete Wiener Process and explain if it is second order stationary or not.

### 2.2 Stochastic Process with dice (I)

Let a Stochastic Process  $\{X_t\}_{t=1...n}$  be defined by :

$$X_1 = 1.75 + 0.5D_1$$
$$X_t = 0.5D_{t-1} + 0.5D_t$$

where  $D_i$  is the i-th dice roll

- a) Calculate the expected value  $E[X_t]$  and the covariance  $\sigma(X_t, X_{t+p})$  and draw the covariance as a function of p . Please explain!
- b) Is this stochastic process second order stationary?
- c) Derive the relationship on slide 7 between the covariance  $\sigma(X,Y) = E[XY] E[X]E[Y]$  and the semivariance  $\gamma(X,Y) = \frac{1}{2}E[(X-Y)^2]$  for second order stationary stochastic processes
- d) Please grab a die and simulate  $X_1...X_{30}$  . Then use the relationship from c) to estimate with Matlab or Excel  $\sigma(X_t,X_{t+n})$  by using only  $\hat{\gamma}(X_t,X_{t+n})$  and  $\hat{\sigma}(X_t,X_t)$  .

#### 2.3 Stochastic Process with dice (II) (bonus points)

Let a Stochastic Process  $\{X_t\}_{t=1,n}$  be defined by :

$$X_1 = 1.75 + 0.5D_1$$

$$X_t = 0.5X_{t-1} + 0.5D_t$$
 where  $D_i$  is the i-th dice roll

- a) Give a first estimate of what you expect the covariance (as a function of distance p between two random variables) to be. Why do you think this behavior reasonable?
- b) Calculate the expected Value  $E[X_t]$ . Hint: Geometric series, see explanation on slides.
- c) Calculate the covariance  $\sigma(X_t, X_{t+p})$  and decide if  $X_t$  can be regarded as a second order stationary stochastic process. **Hint:** Geometric series, see explanation on slides.
- d) After carrying out a simulation of this dice roll game, Bob realizes that he forgot to write down  $X_{50}$ . But he has all the other X's: $X_1...X_{100}$ . Please help Bob and write down (schematically) the Kriging system for the best linear predictor for  $X_{50}$ .

#### 2.4 Gaussian Stochastic Processes (SP)

A (discrete) Gaussian process is a collection of random variables  $\{X_t\}_{t=1...n}$ ; all together jointly Gaussian distributed.

- a) Assume  $X_t$  to be a stationary Gaussian process. Is the expected value  $\mathrm{E}[X_t]$  together with the covariance  $\sigma(X_t, X_{t+1})$  enough to determine the whole process completely? Please explain (or give a counterexample).
- b) Please define two second order stationary Gaussian processes  $\{X_t\}_{t=1...300}$  ,  $\{Y_t\}_{t=1...300}$  with:

$$E[X_{t}] = 0 E[Y_{t}] = 0$$

$$\sigma(X_{t}, X_{t+p}) = \sigma_{0}^{2} e^{-\frac{|p|}{30}} \sigma(Y_{t}, Y_{t+p}) = \sigma_{0}^{2} e^{-\left(\frac{|p|}{30}\right)^{2}} \sigma_{0}^{2} = 1$$

(Hint: Think of what Information you need to fix to completely determine a Gaussian process -> a))

- c) Write a Matlab program that simulates the two stochastic processes you just invented! You can use the pre-built Matlab function mynrnd (generate multivariate Gaussian random numbers see slides 18, 19). Please plot the values of your Gaussian processes and also the correlation coefficients as a function of the distance between the random variables.
- d) Compare the two different stochastic processes by describing and explaining their different behavior by relating it to the two covariance functions.

## 2.5 Gaussian Random Fields (RF)

Let  $\{X_s\}_{s\in\{1\dots30\}\times\{1\dots30\}}$ ,  $\{Y_s\}_{s\in\{1\dots30\}\times\{1\dots30\}}$  be two 2D discrete Gaussian random fields. The goal is now to simulate drawing realizations from these random fields. This can be done in a straightforward way by considering the 2D matrix to be simulated as a single long vector with appropriate covariance matrix calculated from the respective 2D distances.

a) Write a Matlab program that simulates the two random fields  $X_{\cdot}$  and  $Y_{\cdot}$  with properties as sketched below.

$$E[X_s] = 0$$

$$E[Y_s] = 0$$

$$\sigma(X_s, X_{s+h}) = \sigma_0^2 e^{\frac{-||h||}{5}}$$

$$\sigma(Y_s, Y_{s+h}) = \sigma_0^2 e^{-\frac{-||h||}{5}}$$

$$\sigma_0^2 = 1$$

where s and h are now vectors and  $\|\cdot\|$  denotes the euclidean distance

- b) Look at the covariance matrix of the random field you have just sampled from. Can you explain, why it looks how it looks?
- c) Try to invent a random field with the weirdest behavior you can come up with.

## 2.6 Spectral Theory of Stochastic Processes (bonus points)

As is well known, Covariance matrices are positive definite and symmetric (see slide 18). Therefore, they have a unique eigenvalue-eigenvector decomposition. Many interesting properties like the mean energy of a stochastic processes and random fields or their smoothness properties can be traced back to these spectral properties.

a) Let  $X_t$  be a discrete mean 0 SP and denote by  $\langle X_t, X_t \rangle = \|X_t\|^2 = E\left[\sum_{t=1}^n X_t^2\right]$  its squared norm. Show that  $\|X_t\|^2 = \sum_{k=1}^n \lambda_k$  where  $\lambda_k$  are the eigenvalues of the covariance matrix.

**Hint:** trace(AB) = trace(BA)

- b) Look at the sequence of eigenvalues of the covariance matrices you used for simulation in 2.4 b). Can you somehow derive some further info about these processes?
- c) Let  $X_i$  and  $Y_i$  be two SP's and let  $\langle X_i, Y_i \rangle = E\left[\sum_{t=1}^n X_t Y_t\right]$  be their inner product. Show that  $\langle \cdot, \cdot \rangle$  is bilinear, symmetric, positive definite; i.e. has the same properties as the standard scalar product  $a \bullet b = \sum a_i b_i$  that you know from linear algebra and geometry. What implications does this have?

# **Good Luck!**