## Computer exercise 3 in Stationary stochastic processes, HT 20.

The purpose with this computer exercise is to examine the connection between the covariance function and the spectral density of a stationary stochastic process and to study different AR- and MA-processes.

Please work in groups of 2, whenever possible, during the computer exercise.

## 1 Passing the computer exercise

In order to pass the computer exercise you have to:

- Attend your scheduled session. Be on time and make sure that you are in the correct computer room.
- Finish and present the computer exercise within your scheduled 2 hour session. Prepare well.
- Present your work to a teaching assistant before the end of the session. Call the attention of a teaching assistant, who will ask you questions relating to the exercises. Passing requires understanding of the main concepts of the exercise.

Please note that you are always welcome to ask questions during your computer exercise session, just make sure you have prepared well (see next section).

A valid reason for missing a scheduled session, e.g. sickness, needs to be reported to the course responsible (lecturer) BEFORE the start of your session.

## 2 Preparations

- Read through the entire computer exercise.
- Study chapters 7.1-7.3 in the course book.
- Answer the exercises for Computer exercise 3 in Mozquizto at latest Monday October 12 at 12.00 noon.

Mozquizto is found at http://quizms.maths.lth.se/ where you login with your Stil-login. You need to pass the test in order to attend the computer exercise! Prepare well, you might be asked to discuss and explain your answers during the exercise.

### 3 ARMA-models in Matlab

This section is an introduction to how AR, MA and ARMA models are handled in MATLAB.

An ARMA(p,q)-model

$$X_t + a_1 X_{t-1} + \dots + a_p X_{t-p} = e_t + c_1 e_{t-1} + \dots + c_q e_{t-q}$$

can also be written as

$$A(z^{-1})X_t = C(z^{-1})e_t,$$

where

$$A(z^{-1}) = 1 + a_1 z^{-1} + \dots + a_p z^{-p}$$
  

$$C(z^{-1}) = 1 + c_1 z^{-1} + \dots + c_q z^{-q}$$

and where the operator  $z^{-1}$  is the unit time delay. If you choose  $C(z) \equiv 1$ , the process becomes an AR(p)-process, whereas if you choose  $A(z) \equiv 1$ , it becomes an MA(q)-process. In MATLAB, polynomials are represented as vectors; e.g.  $z^p A(z^{-1}) = z^p + \cdots + a_{p-1}z + a_p$  is represented as [1 a1 ... ap].

### 3.1 Study of AR(1)-processes

Create an AR(1)-model in MATLAB according to the previous section, i.e., C(z) = 1, and  $A(z^{-1}) = 1 + a_1 z^{-1}$  represented as C=1 and A=[1 a1] for your choice of the parameter. Remember to save your MATLAB code in scripts. One may compute the spectral density  $R_X(f)$  for the process as

```
>> [H,w]=freqz(C,A);
>> R=abs(H).^2;
```

where freqz is a MATLAB function that calculates the frequency function corresponding to the polynomials C and A in a number of equally spaced frequency values (default 512) between  $0 \le \omega < \pi$ , with  $\omega = 2\pi f$ . Plot the spectral density using

```
>> plot(w/2/pi,R)
```

To calculate the covariance function, the inverse Fourier transform of the spectral density is computed. The frequency range must be the whole period with the frequency range  $0 \le f < 1$  when the command ifft is used. The frequency function is computed for this range using whole in the function freqz,

```
>> H=freqz(C,A,512,'whole');
>> Rd=abs(H).^2;
>> r=ifft(Rd);
>> stem([0:49],r(1:50))
```

where the 50 first values of the covariance function is plotted. Observe that the vector element  $\mathbf{r}(1)$  is equal to the covariance value at zero, i.e.  $r_X(0) = \mathbf{r}(1)$ , and more generally  $r_X(k) = \mathbf{r}(k+1)$  in MATLAB. Try a positive and a negative value of  $a_1$ , e.g.  $a_1 = 0.5$  and  $a_1 = -0.5$ , and study the results.

 $Q.\ Draw\ rough\ sketches\ of\ the\ covariance\ functions\ for\ a_1<0\ and\ a_1>0.$  How do they differ?  $Q.\ Sketch\ the\ spectral\ densities.\ At\ what\ frequency\ does\ the\ spectral\ density\ reaches\ the\ maximum\ value\ for\ a_1<0\ and\ a_1>0,\ respectively?$ 

To simulate a realisation of the AR-process, we generate a sequence of independent and Gaussian<sup>1</sup> distributed stochastic variables,  $\{e_t, t = 1, ..., n\}$ . This can be made as

```
>> m = 0;
>> sigma = 1;
>> e = normrnd(m, sigma, 1, n);
or, alternatively,
>> e = randn(1,n);
```

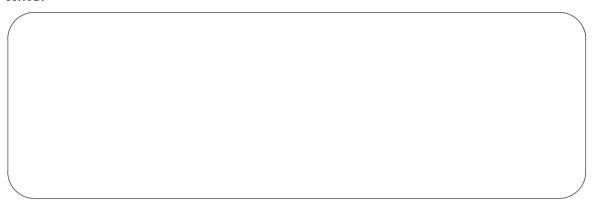
Each call to the above function will generate a row-vector  ${\tt e}$  with  ${\tt n}$  Gaussian distributed random numbers; here, choose  ${\tt n} \approx 400$ . The simulation of the AR(1)-process is made using

 $<sup>^{1}</sup>$ We note that, in general,  $e_{t}$  does not need to be Gaussian distributed.

```
>> x = filter(C, A, e);
```

which result in the vector  $\mathbf{x}$  with the simulated time-series as the result, where  $\mathbf{A}$  and  $\mathbf{C}$  are defined as earlier. Create one realization for the process where  $a_1 < 0$  and one for  $a_1 > 0$ .

Q. Sketch the view of the realizations. Express  $X_t$  in  $X_{t-1}$  and  $e_t$  for the two cases and explain the differences in how the realizations evolve with time.



### 3.2 Study of an ARMA(2,2)-process

Specify an ARMA(2,2)-process using A=[1 -1 0.5] and C=[1 1 0.5]. It is often convenient to specify the roots as poles and zeros instead of the coefficients of the A-and C-polynomials. Compute and plot the poles and zeros of the ARMA(2,2)-process using

```
>> P = roots(A)
>> Z = roots(C)
and
>> zplane(Z,P)
```

Note: When using the function zplane, the poles must be located in column vectors. If row vectors are used as input, zplane interprets this as polynomial vectors and plots roots(A). Verify this by using the commands zplane(Z,P) and zplane(C,A).

If the poles P and zeros Z are specified instead and you want to calculate the polynomials A and C, one may use the command poly

```
>> A=poly(P)
>> C=poly(Z)
```

Sketch the spectral density just from the view of the pole-zero plot. Verify by computing and plotting the spectral density according to the previous subsection.

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# 4 Study of ARMA-processes using armagui

Use the function armagui to investigate the connections between poles, zeros, covariance functions and spectral densities for different ARMA-models. Type

### >> help armagui

to get information on how to use it, and start the function by typing

### >> armagui

The help text is also included at the end of this exercise.

## 4.1 Study of AR(2)-processes

Use armagui to produce an AR(2)-model with a complex conjugated pair of poles. Restart by removing the poles with the button "Remove all". Add a complex pole, (gives you the pole as well as the conjugate). Move the poles (easily done with the mouse button). Change the angle frequency from the real axis and keep quite close to the unit circle. Study the spectral density and the covariance function. Change the distances of the poles from the unit circle and study the covariance function and the spectral density.

from the unit circle?			
Move your poles very close to	the origin.		
Q.	$ss, \ can \ you \ g$	uess what the	$a_k$ coefficients are:
			)
To see what happens when	the generating L	AR-filter is unstab	le, i.e., if the roots to
the characteristic polynomial stability" in armagui and mother beginning of the realization	are outside the ve one pole/pol	unit circle, de-selec e pair outside the t	t the button "Demand unit circle. Zoom in a
Q. What happens with th	ne process rea	lization? Why?	

### 4.2 Study of MA(q)-processes

MA-processes behave a bit different from AR-processes and we will study MA-processes of some different orders. Restart by removing the poles with the button "Remove all". Add a complex zero pair, generating an MA(2)-process. Add more zeros to form MA(3)-, MA(4)- and MA(5)-processes.

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	appearance of	_		_	•	
$Q.\ Sketch\ t$ $circle\ and\ c$	ally what happed with a second	lensities, on complex	one with co	omplex zer se to the o	os close to	the unit

## 5 Speech modeling in cell phones

An early method for modeling speech is that it locally can be assumed to be an AR(20)-process. A refined version of this model is actually still used when transmitting sound between cellphones (or mobile phones if you prefer British English). When you are speaking in a cellphone, the sound is divided into 20 millisecond long time frames. The parameters of the model are estimated; these parameters of the AR-model are then sent to the other persons phone. By running white noise through the model, the speech is reconstructed.

#### 5.1 Parameter estimation

Open the file lab3\_extra.m. The code loads an audio file where someone is saying "Why were you away a year Roy?". You can listen the file, once it's loaded, by typing >>sound(x,Fs):

Plot the process and check that the process is zero mean! The first section of the code in lab3\_extra.m divides the sound file into into sections of 20 ms; for each section we want to estimate the parameters of the AR(20)-model. Your task is to write that code into the for-loop. It's a very useful exercise to write this code yourself; for help see pages 180-181 in the book. However if you are short on time, there are built-in functions in MATLAB that you can look up. The function arcov solves exactly this problem and you can also use this function to verify that your programming is correct. Check help arcov to understand how the function works. Note that you need two output values.

Once the parameters are estimated, one can see the sizes of audio file x and the AR parameters in ar\_param by writing:

#### >> whos x ar\_param

This will show you the sizes of the audio-file and the size of the parameters in bytes.

$Q.\ How\ much\ storage\ is\ saved\ by\ sending\ the\ AR$ -parameters instante audio itself?	ead of

#### 5.2 Evaluation

The second part of the code in lab3\_extra.m will estimate the spectrum of one section of the audio file and compare it with the spectrum of the estimated AR(20)-model. If your code work they should show some similarities, although they are not exactly the same.

The third section of the code will reconstruct the audio using your estimated model parameters. The original audio and the reconstructed audio will be plotted in figure one. NOTE: Before playing the reconstructed audio, check the plotted reconstruction and make sure it is stable!

If the reconstruction looks stable and close to the original audio you can listen to it by typing

#### >> sound(x\_rec,Fs)

As you (hopefully) can hear, the reconstruction is quite noisy and sounds somewhat inhuman. This is, as mentioned, an early model; if you want to know how some easy changes can be made, reducing the order to AR(10)) with improved sound quality, you

can learn this in the followup course  $\mathit{Time\ series\ analysis}.$ 

### 6 Matlab-functions

#### armagui

```
function armagui(action)
% ARMAGUI
% armagui opens a window for ARMA(p,q)-modeling.
% Use the buttons to add poles and zeros. Move them by using the mouse.
% Delete poles and zeros with the right mouse button.
% If the button "Demand stability" is marked, poles and zeros
% are not allowed outside the unit circle.
% The button "Simulate" gives a realization of the process.
\mbox{\ensuremath{\mbox{\%}}} A- and C-polynomial can be loaded into ARMAGUI by creating, e.g.,
\% >> modell.A=[1 -0.5];
% >> modell.C=[1 0.4 0.7];
% and write the name (i.e.
% "model") in the "Import"-box and push the button.
% Covariance function, spectral density and realizations can be
% exported to Matlab with the "Export"-button.
% Then a variable with the name written in the "Export"-box is created,
% which includes poles, zeros etc..
```