

**NAME: GEORGIOS DIMITROPOULOS**

**STUDENT NUMBER: 4727657**

**EMAIL ADDRESS:**

**[G.Dimitropoulos-1@student.tudelft.nl](mailto:G.Dimitropoulos-1@student.tudelft.nl)**

**RANDOM SIGNAL PROCESSING: FINAL MATLAB  
ASSIGNMENT**

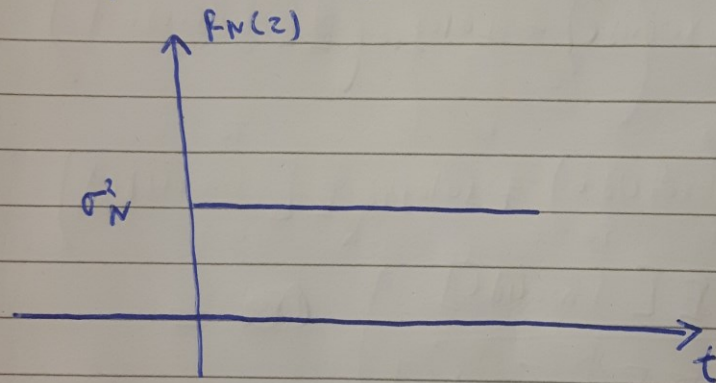
## EXERCISE 1

### Exercise 01

We are given that  $E[N(t)] = 0$  and  $\text{Var}[N(t)] = \sigma_N^2$  and also it is assumed that the noise  $N(t)$  is a stochastic stationary and uncorrelated signal. This implies that:

$$R_N(t, z) = R_N(z) = E[N(t) \cdot N(t+z)] \stackrel{t=0}{=} E[N^2(z)] = \sigma_N^2.$$

So, the sketch of the autocorrelation function of the noise process  $N(t)$  is given by the next figure:



## EXERCISE 2

## Exercise 02

We should compute the quantity  $SNR_e - SNR$ .

$$\text{So, } SNR_e - SNR = 10 \log_{10} \frac{\int_{-T_0}^{T_0} x_0^2(\tau) d\tau}{E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right]} -$$

$$- 10 \log_{10} \left( \frac{\int_{-T_0}^{T_0} x_0^2(\tau) d\tau}{E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right]} \right)$$

$$= 10 \log_{10} \left( \int_{-T_0}^{T_0} x_0^2(\tau) d\tau \right) - 10 \log_{10} \left( E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right] \right)$$

$$= 10 \log_{10} \left( \int_{-T_0}^{T_0} x_0^2(\tau) d\tau \right) + 10 \log_{10} \left( E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right] \right)$$

$$= 10 \log_{10} \left( \frac{E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right]}{E \left[ \int_{-T_0}^{T_0} \tilde{n}^2(\tau) d\tau \right]} \right) =$$

$$= 10 \log_{10} \frac{\int_{-T_0}^{T_0} E[\tilde{n}^2(\tau)] d\tau}{\int_{-T_0}^{T_0} E[\tilde{n}^2(\tau)] d\tau} \quad (*)$$

Let compute the average times that are inside the integrals separately.

$$E[n^2(t)] = \text{Var}[n(t)] + (E[n(t)])^2 = \text{Var}[n(t)] = \sigma_n^2 \quad (1)$$

$$E[\bar{n}^2(t)] = \text{Var}[\bar{n}(t)] + E[\bar{n}(t)]^2 \quad (2)$$

We have that:

$$E[\bar{n}(t)] = E\left[\frac{1}{10} (n(t) + n(t+T_0) + \dots + n(t+9T_0))\right] =$$

$$= \frac{1}{10} \{E[n(t)] + E[n(t+T_0)] + \dots + E[n(t+9T_0)]\}$$

$$(3)$$

$$= 0 \quad ; \text{ as we have that } E[n(t)] = E[n(t+T_0)] = \dots = E[n(t+9T_0)] = 0$$

$$\text{Var}[\bar{n}(t)] = \text{Var}\left[\frac{1}{10} [n(t) + \dots + n(t+9T_0)]\right] =$$

$$\overset{n(t) \text{ uncorrelated}}{=} \frac{1}{100} [\text{Var}[n(t)] + \dots + \text{Var}[n(t+9T_0)]] =$$

$$= \frac{1}{100} (\underbrace{\sigma_n^2 + \dots + \sigma_n^2}_{10 \text{ times}}) = \frac{10 \sigma_n^2}{100} = \frac{\sigma_n^2}{10} \quad (4)$$

$$\text{So, } (2) \xrightarrow{(4)} E[\bar{n}^2(t)] = \frac{\sigma_n^2}{10} \quad (5)$$

$$\text{Thus, } (1) \xrightarrow{(5)} \boxed{\text{SNR}_{\text{e}} - \text{SNR}} = 10 \log_{10} \frac{\int_0^{T_0} \sigma_n^2 dt}{\int_0^{T_0} \frac{\sigma_n^2}{10} dt}$$

$$= 10 \log_{10} \left( \frac{T_0 \sigma_n^2}{T_0 \frac{\sigma_n^2}{10}} \right) = 10 \log_{10} 10 = \boxed{10}$$

### EXERCISE 3



### Exercise 03

In order to obtain an SNR increase of 90 dB using the epoch folding, we should calculate the value of the variable  $k$  for given  $T_0 = 0.5$ .

$$\text{So, } \text{SNR}_e - \text{SNR} = 10 \log_{10} \left( \frac{\int_0^{0.5} \sigma^2 N dt}{\int_0^{0.5} \frac{\sigma^2 N}{k} dt} \right) = 90$$

$$\Rightarrow 10 \log_{10} \left( \frac{\frac{e \cdot 50^2 N}{1}}{\frac{e \cdot 50^2 N}{k}} \right) = 90$$

$$\Rightarrow 10 \log_{10} k = 90 \Rightarrow \log_{10} k = 9$$

$$\Rightarrow 10^{\log_{10} k} = 10^9$$

$$\Rightarrow \boxed{k = 10^9}$$

#### Exercise 04

These signals that are given are sampled at 10 kHz. So, we can obtain that the frequency  $f_s = 10 \cdot 10^3 = 10^4$  Hz.

Thus, the periodicity  $T_0$  in samples by observing signal  $x$  is given by:  $T_0 = \frac{1}{f_s} = \frac{1}{10^4} = 10^{-4}$  s

Also, we computed the SNR of the original signal using signals  $x$  and  $n$  by using the MATLAB command  $\text{SNR}(x, n)$ . The result is  $-34.9972$ .

#### EXERCISE 5

The epoch\_folding gives as an output the epoch folded signal  $z$  by taking as an input a signal  $y$ , the period  $T_o$  and the number of frames  $K$ .

The accompanying code follows:

```
function [z] = Epoch_folding (y, To, k)

z=zeros (length(y), 1)
for i= 1:length(y)-(k-1)*To
    for j=0:k-1
        z(i)=(1/k)*y(i+(To*j))+z(i);
    end
end

for i=1:length(z)
    time(i)=i;
end

for i=1:length(z)
    a(i)=z(i);
end

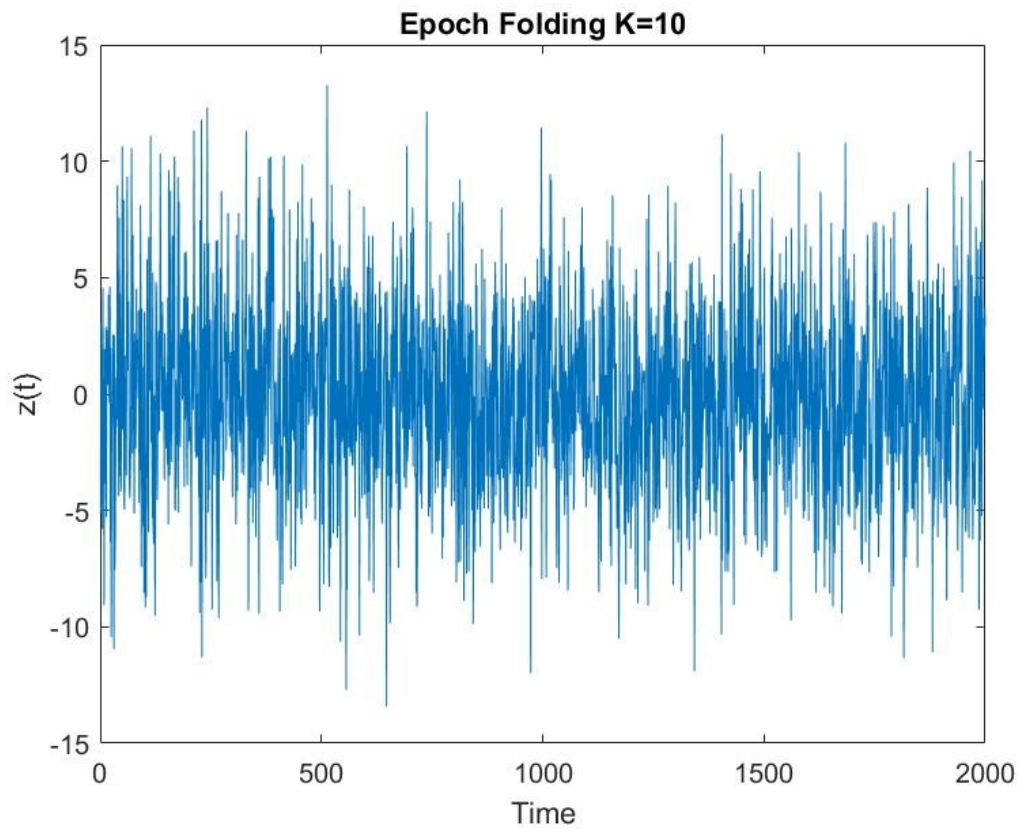
plot(time, a);
xlabel ('Time');
ylabel ('z(t)');
end
```

## EXERCISE 6

We use the above function epoch\_folding in order to detect the pulse in the pulsar signal. An increment of SNR of 10, 20, and 30 dB implies that the  $K$  will be 10, 100, 1000 respectively.

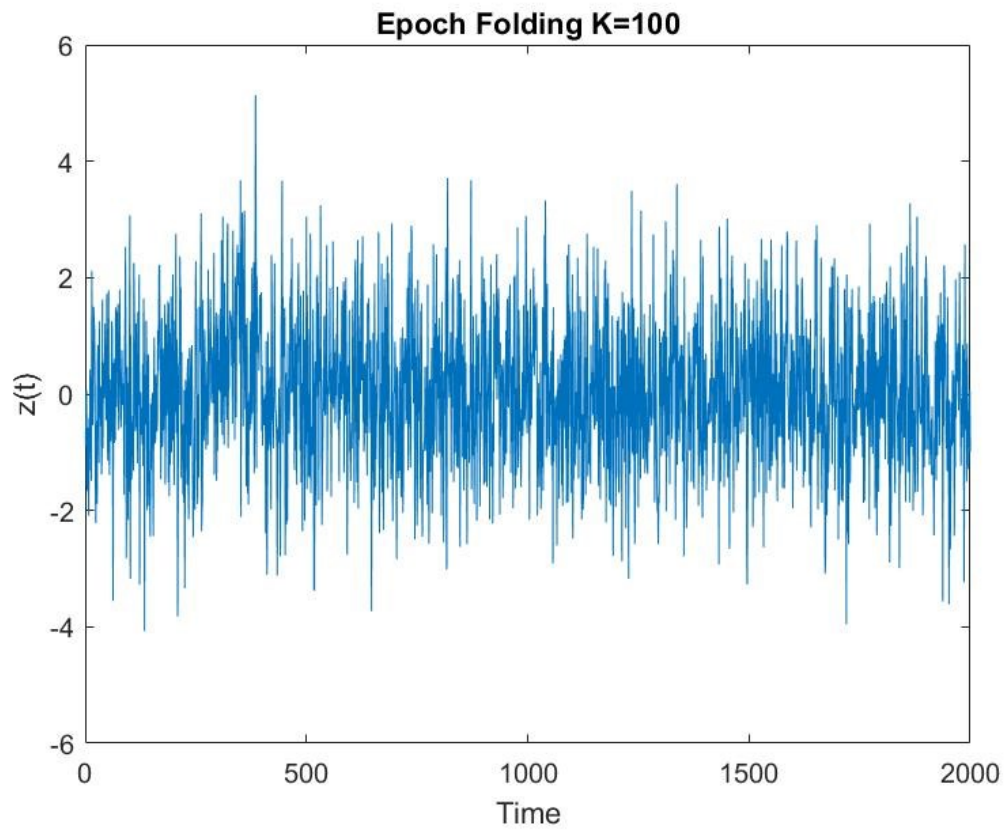
Hence, for the corresponding values of  $k$ :

**$k=10$**

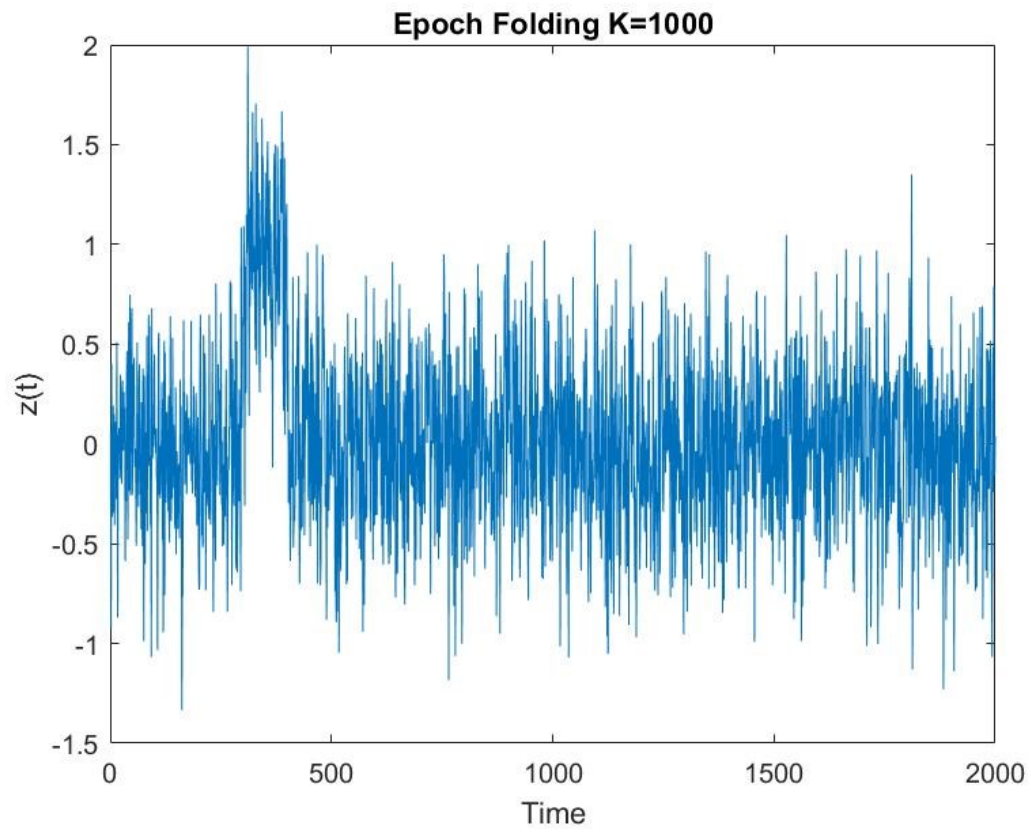


**k=100**





**k=1000**



## EXERCISE 7

### Exercise 07

It is given that  $z_m(t) = \int_0^{T_0} x_0(s) x(-t+ks) ds + \int_0^{T_0} \tilde{n}(u) x(-t+u) du$

By assuming that  $x_0(t)$  consists of a perfect pulse and taking into account that for  $k \rightarrow \infty$  we have that

$$\lim_{k \rightarrow \infty} \frac{1}{k} \sum_{k=0}^{k-1} n(t+ks) = \lim_{k \rightarrow \infty} \tilde{n}(t) \rightarrow 0$$

So, as the noise is assumed to be uncorrelated, it is going to cancel out as  $k \rightarrow \infty$ .

Thus our sketch consists of the graph of our ~~initial~~ initial signal  $y(t)$ . We notice that it contains a lot of noise. However later on, as we use techniques such as ~~epoch~~ epoch folding and matched filtering and we are able to have better results.

### EXERCISE 8

## EXERCISE 8

### Exercise 08

We need to prove that  $E[z_m(t)] = T_0 R_{X_0} X(-t)$

We have that  $z_m(t) = \int_0^{T_0} x_0(s) x(-t+s) ds + \int_0^{T_0} \tilde{n}(s) x(-t+s) ds$

$$\text{So, } E[z_m(t)] = E \left[ \int_0^{T_0} x_0(s) x(-t+s) ds + \int_0^{T_0} \tilde{n}(s) x(-t+s) ds \right]$$

$$= \int_0^{T_0} E[x_0(s) x(-t+s)] ds + \int_0^{T_0} E[\tilde{n}(s) x(-t+s)] ds \quad (x)$$

We notice that:

$$E[\tilde{n}(s) \cdot x(-t+s)] \stackrel{\tilde{n}, x \text{ independent}}{=} E[\tilde{n}(s)] \cdot E[x(-t+s)]$$

Also, we proved in exercise 2 that  $E[\tilde{n}(s)] = 0$ .

$$\text{Thus, } E[\tilde{n}(s) \cdot x(-t+s)] = 0 \quad (1)$$

$$(x) \stackrel{(1)}{\Rightarrow} E[z_m(t)] = \int_0^{T_0} E[x_0(s) x(-t+s)] ds \quad (2)$$

$$\text{By definition, } E[x_0(s) x(-t+s)] = R_{X_0} X(s, -t) =$$

$$\stackrel{(3)}{=} R_{X_0} X(-t) \quad (4)$$

$x_0, x$   
jointly WSS

~~scribbles~~



In general we have that:

$$R_{XY}(t, z) = E[X(t) \cdot Y(t+z)]$$

Let  $\boxed{t=s}$ , then  $t+z = -t+s$   
 and  $s+z = -t+s \Rightarrow \boxed{z=-t}$

So,  $\boxed{\begin{matrix} t=s \\ z=-t \end{matrix}}$  (3) ~~(4)~~

So, (2)  $\Rightarrow E[Z_n(t)] = \int_0^{T_0} R_{X_0}(-t) ds =$

$$= R_{X_0}(-t) [s]_0^{T_0} = \boxed{T_0 R_{X_0}(-t)} \quad \blacksquare$$

Hence

~~So~~, we made the assumption that the variables  $\tilde{n}, x$  are independent and the variables  $x_0$  and  $x$  are Wide Sense Stationary jointly.

## EXERCISE 9

### Exercise 09

It is given in equation (18) that:

$$z_m(L) = \frac{T_0}{N} \sum_{m=0}^{N-1} z_e(m) x(-L+m) \quad (1)$$

We know that  $z_e(m) = \frac{1}{K} \sum_{k=0}^{K-1} x(m + kL) + \frac{1}{K} \sum_{k=0}^{K-1} \tilde{n}(m + kL) \quad (2)$

Thus, (1)  $\Rightarrow$   $z_m(L) = \frac{T_0}{N} \sum_{k=0}^{K-1} x_0(m) x(-L+m) + \frac{T_0}{N} \sum_{k=0}^{K-1} \tilde{n}(m) x(-L+m)$

$$= \frac{T_0}{N} R_{x_0} x(-L) + \frac{T_0}{N} R_{\tilde{n}} x(-L)$$

So, for large  $N$ ,  $K \rightarrow \infty$ ,  $z_m(L)$  converges to:

$$\lim_{N \rightarrow \infty} \frac{1}{N} T_0 R_{x_0} x(-L) + \lim_{N \rightarrow \infty} \frac{1}{N} T_0 R_{\tilde{n}} x(-L) =$$

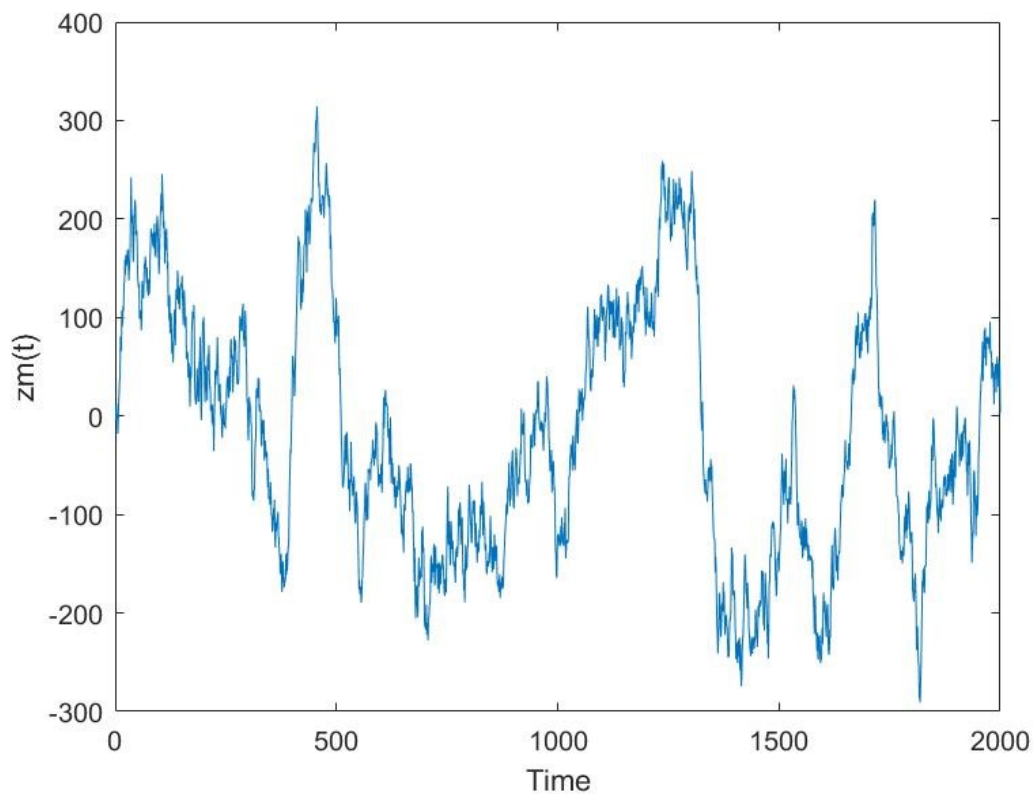
$$= T_0 R_{x_0} x(-L) + T_0 R_{\tilde{n}} x(-L).$$

Taking into consideration that  $\tilde{n}, x$  are independent (the property related to estimation of autocorrelation functions that the underlying processes shall have), we have that  $T_0 R_{\tilde{n}} x(-L) = 0$ .

Thus  $z_m(L) \approx T_0 R_{x_0} x(-L)$

### EXERCISE 10

The plot of the results:



**Accompanying code:**

```
function [zm] = Matched_Filter (ze,y)
```

```
zm=zeros(length(x),1);
a(1)=x(1);
for i=0:length(x)-2
    a(i+2)=x(length(x)-i);
end
a=a';
Tp = toeplitz(a,x);
zm=Ta*ze ;
for i=1:length(x)
    t(i)=i;
end
for i=1:length(x)
    s(i)=zm(i);
end
plot(t,s);
xlabel('Time');
```

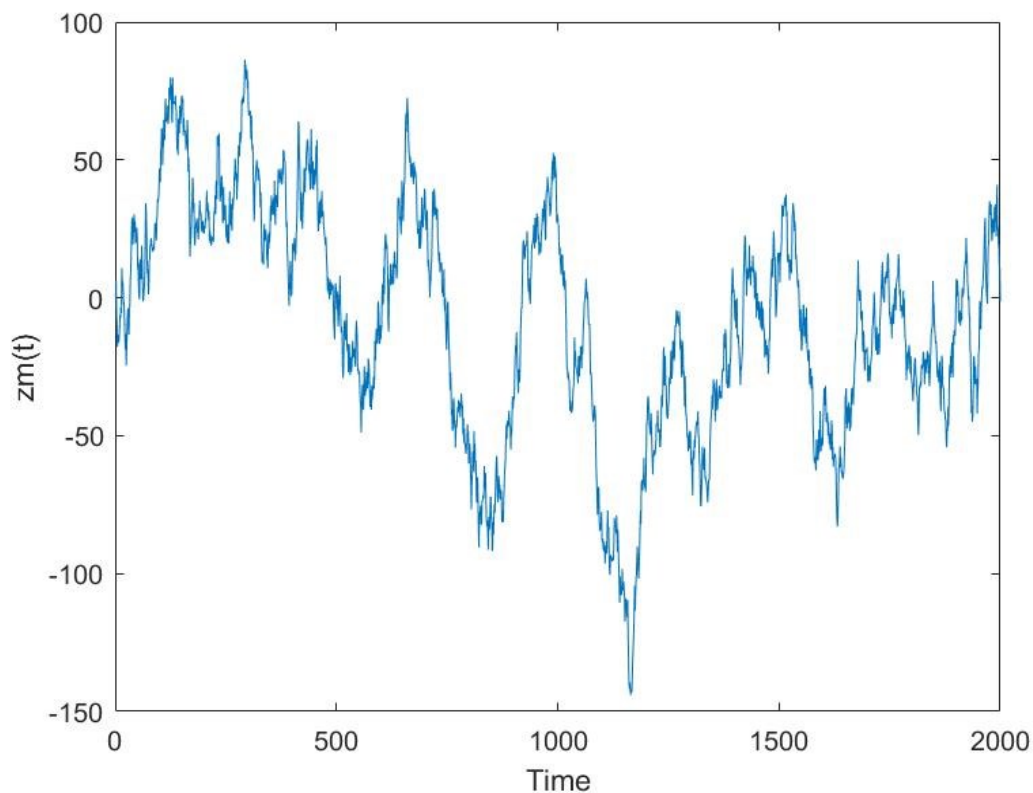
```
ylabel('zm(t)');  
end
```

We assumed that the  $z_e$  will be the  $y$  variable and the  $x$  equals to template in order to compute the output of the matched filter.

## EXERCISE 11

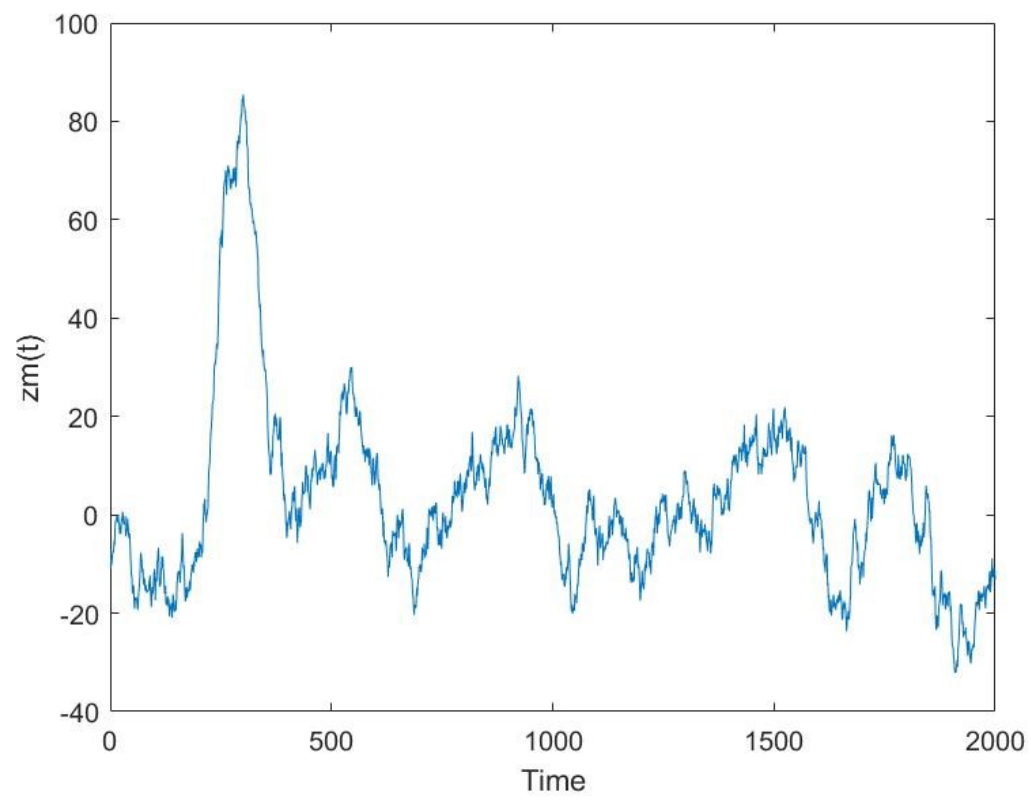
First we generate epoch folded signals using the function epoch folding on the noisy signal  $y$  for the  $K$ -values which we computed in exercise 6 in order to determine the pulse location of the pulsar signal. Then we use each of these epoch folded signals as input  $z_e$  in the function matched filter. Hence, for the corresponding values of  $k$  we attach the results:

**k=10**



**k=100**





**k=1000**

