

Power Spectra & Spectral Factorization

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Introduction

In this live script, we offer a few examples of solving, using MATLAB code, simple problems involving computation of the **power spectra** of stationary stochastic processes, the **spectral factorization** and the so-called **canonical spectral factorization**.

What you will learn:

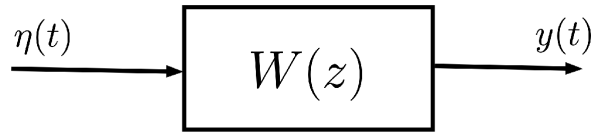
- How to cope with a spectral factorization problem using MATLAB;
- How to manage a spectrum of a stochastic proces with the aim to obtain a canonical spectral factorization;
- How to evaluate the autocorrelation function of a stationary stochastic process, given its complex spectrum, and viceversa.

For a while we'll use the Symbolic Toolbox

```
clear variables
close all
clc

sympref('HeavisideAtOrigin', 1); % setting the Symbolic Toolbox
```

A Stationary Stochastic Process from a White Noise Response of an LTI Filter - A First Example



Consider a stationary stochastic process $y(t)$, generated by an LTI dynamic system with transfer function

$$W(z) = \frac{1}{z - 0.5}$$

and white noise $\eta(t)$ as its input. Assume that $\eta(t)$ has a zero-mean and a unitary variance: $\eta(\cdot) \sim \text{WN}(0, 1)$

Find the power spectrum $\Gamma_y(\omega)$ and the autocorrelation function $\gamma_y(\tau)$ of $y(t)$.

Solution

Since $y(t)$ is the output of the LTI system $W(z)$ (the filter) and the input is the unit-variance and zero-mean white noise $\eta(t)$, then the spectrum $\Phi(z)$ can be expressed as

$$\Phi_y(z) = W(z) \cdot W\left(\frac{1}{z}\right) \cdot \Phi_\eta(z) = W(z) \cdot W\left(\frac{1}{z}\right) \cdot 1 = W(z) \cdot W\left(\frac{1}{z}\right)$$

```
syms z n
```

```
Wz = 1/(z-0.5) % the LTI system
```

```
Wz =
```

$$\frac{1}{-\frac{1}{2} + z}$$

```
sigma2_eta = 1; % the variance of the white noise
```

Thus the process spectrum is

```
W_z1 = subs(Wz, z, (z^(-1)))
```

$$W_{z1} =$$

$$\frac{1}{\frac{1}{z} - \frac{1}{2}}$$

$$\text{PhiZ} = Wz * W_{z1} * \text{sigma2_eta}$$

$$\text{PhiZ} =$$

$$\frac{1}{\left(\frac{1}{z} - \frac{1}{2}\right) \left(-\frac{1}{2} + z\right)}$$

Evaluating for $z = e^{i\omega}$ we obtain $\Gamma_y(\omega)$

```
syms omega
Gamma_y = subs(PhiZ, z, exp(1i*omega))
```

$$\text{Gamma}_y =$$

$$\frac{1}{\left(e^{-i\omega} - \frac{1}{2}\right) \left(e^{i\omega} - \frac{1}{2}\right)}$$

Rewrite $\Gamma_y(\omega)$ exploiting the Euler's formula

$$e^{i\omega} = \cos \omega + i \sin \omega$$

where i is the imaginary unit. Thus

$$\cos \omega = \frac{e^{i\omega} + e^{-i\omega}}{2}$$

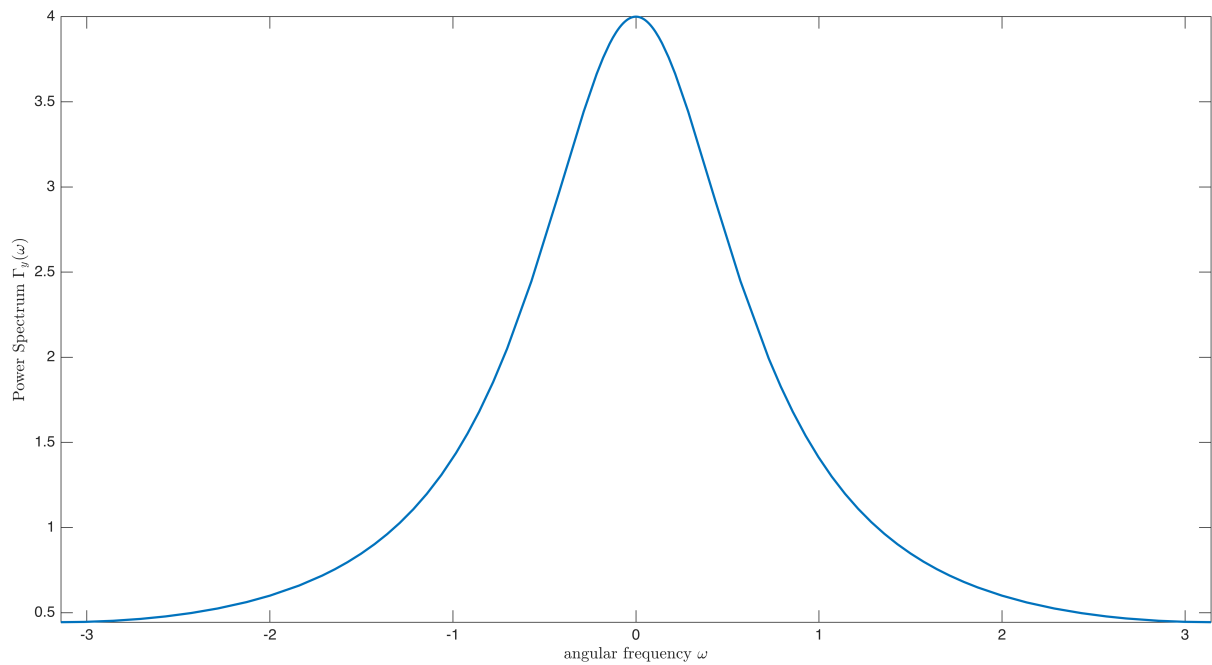
```
Gamma_y = simplify(rewrite(Gamma_y, "cos"))
```

$$\text{Gamma}_y =$$

$$-\frac{4}{4 \cos(\omega) - 5}$$

The spectrum is

```
figure('Units','normalized','Position',[0.1,0.1,0.9,0.8]);
fplot(gca, Gamma_y, [-pi, pi], 'LineStyle','-', 'LineWidth',1.5);
% plot the curve defined by the function y = Gamma_y(omega) over the
% interval [-pi , +pi]
xlabel('angular frequency $\omega$', 'Interpreter','latex');
ylabel('Power Spectrum $\Gamma_y(\omega)$', 'Interpreter','latex')
```



Q&A

- **Q1:** Given the expression of $W(z)$, to which family of rational spectrum processes does the analyzed stochastic process belong?
- **Q2a:** Is the spectral factorisation canonical or not? **Q2b:** Why? **Q2c:** If not, what can the canonical spectral factorisation be?
- **A2c:** the canonical spectral form is $\hat{W}(z) = \frac{z}{z - 0.5}$ $\xi(\cdot) \sim \text{WN}(0, 1)$
- **Q3:** What is the variance of the process?
- **A3:** $\sigma_y^2 = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \Gamma_y(\omega) d\omega$ or, exploiting the spectral canonical form $\sigma_y^2 = \frac{\lambda_\xi^2}{(1 - a^2)} = \frac{1}{1 - 0.25} = \frac{4}{3}$

```
sigma2_y = double((1/(2*pi))*int(Gamma_y,omega, [-pi,pi] ))
```

```
sigma2_y = 1.3333
```

```
sym(sigma2_y)
```

```
ans =
```

```
 $\frac{4}{3}$ 
```

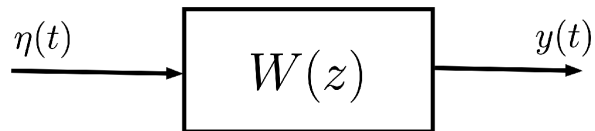
- **Q4:** Given the spectral canonical factorization, what is the autocorrelation function?
- **A4:** For an AR(1) stationary process the autocorrelation function is $\gamma_y(\tau) = \frac{\lambda_\xi^2}{(1 - a^2)} a^{|\tau|} = \frac{4}{3} \left(\frac{1}{2}\right)^{|\tau|}$

A Spectral Shaping Filter

Suppose we want to **generate a stationary stochastic process** characterised by the following **power spectrum**

$$\Gamma_y(\omega) = \frac{5 + 4 \cos \omega}{13 + 12 \cos \omega} \quad \omega \in [-\pi, +\pi]$$

by **filtering** unit-variance and zero-mean white noise $\eta(t)$ with an asymptotically stable LTI system



Find the transfer function $W(z)$ of the filter.

```
clear variables
```

Let us define the angular frequency ω as symbolic variable, and then assign the spectrum

```
syms omega
Gamma_y = (5+4*cos(omega))/(13+12*cos(omega))
```

```
Gamma_y =  
4 cos(ω) + 5  
12 cos(ω) + 13
```

Evaluating the Process Variance

Given the expression of the power spectrum $\Gamma_y(\omega)$, the variance of the r.v. $y(t)$ can be determined evaluating the integral

$$\sigma_y^2 = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \Gamma_y(\omega) d\omega$$

```
sigma2_y = ( (1/(2*pi))*int(Gamma_y,omega, [-pi,pi] ) )
```

```
sigma2_y =  
13379709324852871 π  
90071992547409920
```

Not so clearly readable. A better way to obtain a readable result is

```
sigma2_y_ALT = double(sigma2_y)
```

```
sigma2_y_ALT = 0.4667
```

```
Sigma2Y = sym(sigma2_y_ALT)
```

```
Sigma2Y =
```

$$\frac{7}{15}$$

Now the result is clearly readable.

The Complex Spectrum $\Phi(z)$

Rewrite $\Gamma_y(\omega)$ exploiting the Euler's formula

$$e^{i\omega} = \cos \omega + i \sin \omega$$

where i is the imaginary unit. Thus

$$\cos \omega = \frac{e^{i\omega} + e^{-i\omega}}{2}$$

and we obtain

```
Gamma_y = rewrite(Gamma_y, "exp")
```

```
Gamma_y =
```

$$\frac{2e^{-\omega i} + 2e^{\omega i} + 5}{6e^{-\omega i} + 6e^{\omega i} + 13}$$

Symmetric Polynomials

Now, replacing $e^{i\omega}$ with the complex variable z we obtain the expression of the desired complex spectrum $\Phi(z)$, as the bilateral Z-transform of the autocorrelation function $\gamma_y(\tau)$

```
syms z
```

```
PhiZ = subs(Gamma_y, exp(1i*omega), z)
```

```
PhiZ =
```

$$\frac{2z + \frac{2}{z} + 5}{6z + \frac{6}{z} + 13}$$

Note the peculiar structure of the polynomials at the numerator and the denominator in the expression of the spectrum. They are **symmetric polynomials** in the variables z and z^{-1} . In fact, permuting the variable z with z^{-1} does not change the expression of both the polynomials.

Remark

The feature just highlighted is in fact a peculiar property of rational spectra: one can always write the two polynomials that appear in the spectrum as symmetrical polynomials in z and z^{-1} . More details are available in the [live function applying the canonical spectral factorization algorithm](#).

Applying the Spectral Factorization Algorithm

Note that in this live script we are applying the "*Spectral Factorization Algorithm*", summarising the steps of the algorithm itself, but without any sketch of proof or any explanation. For details on the algorithm and some explanation, please refer to the [live function implementing the spectral factorization algorithm](#).

The Spectral Factorization Theorem

Given a process with rational spectrum $\Phi(z)$, there exists one and only one representation of the process as the output of an LTI system driven by a white process $\xi(\cdot)$ and with transfer function $W(z) = \frac{N(z)}{D(z)}$ if the following conditions are imposed on $W(z)$:

- $N(z)$ and $D(z)$ monic, co-prime and of the same degree;
- all roots of $N(z)$ (zeros of $W(z)$) have $|\cdot| \leq 1$;
- all roots of $D(z)$ (poles of $W(z)$) have $|\cdot| < 1$.

$$\Phi_y(z) = W(z) \cdot W\left(\frac{1}{z}\right) \cdot \Phi_\xi(z) = W(z) \cdot W\left(\frac{1}{z}\right) \cdot \lambda_\xi^2$$

An Usefull Trick

In order to apply the "*Spectral Factorization Algorithm*" we have to determine the coefficients of both the [symmetric polynomials](#) at the numerator and the denominator in the rational spectrum expression.

A simple trick to obtain the needed polynomial coefficients is based on the symbolic variable substitution $(z + z^{-1}) \rightarrow x$ into the expression of $\Phi(z)$

```
syms x
PhiX = subs(PhiZ, (z+z^(-1)), x) % the first step of the trick: substituting the expres
```

```
PhiX =
    5 + 2 x
    13 + 6 x
```

```
% the second step of the trick: get numerator and denominator of the
% symbolic expression PhiX
[betaX, alphaX] = numden(PhiX)
```

```
betaX = 5 + 2 x
alphaX = 13 + 6 x
```

Finally, extract the coefficient array from each polynomial, ordered from the highest to the lowest degree of the monomials in the instrumental variable x

```
betaPoly = sym2poly(betaX) % the coefficients of the numerator of the spectrum, as a s
```

```
betaPoly = 1x2
           2   5
```

```
% betaPoly contains the floating-point description of the coefficients,
% from the highest to the lowest degree monomial of the auxiliary variable x
```

```
alphaPoly = sym2poly(alphaX) % the coefficients of the denominator of the spectrum, as
```

```
alphaPoly = 1x2
           6   13
```

```
% alphaPoly contains the floating-point description of the coefficients,
% from the highest to the lowest degree monomial of the auxiliary variable x
```

Now we re ready to apply the *Spectral Factorization Algorithm*.

The Spectral Factorization Algorithm: a Summary

Given the rational spectrum

$$\Phi(z) = \frac{\beta(z)}{\alpha(z)} = \frac{\sum_{k=-m}^m \beta_k z^k}{\sum_{k=-n}^n \alpha_k z^k}$$

with

$$\begin{cases} \alpha_p = \alpha_{-p} & p = -n, -n+1, \dots, n-1, n \\ \beta_q = \beta_{-q} & q = -m, -m+1, \dots, m-1, m \\ \alpha_n \neq 0, \beta_m \neq 0 & \text{i.e. both polynomials are symmetric polynomials in } z \text{ and } z^{-1} \\ \Phi(e^{j\omega}) \geq 0 \quad \forall \omega & \text{i.e. } \Phi(z) \text{ is a rational spectrum} \end{cases}$$

then there is only one factorization of the spectrum $\Phi(z)$ such that

$$\Phi(z) = \frac{C(z)}{A(z)} \cdot \frac{C(z^{-1})}{A(z^{-1})} \cdot r$$

with

$$\begin{cases} r > 0 \\ A(z) = 1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_n z^{-n} & \text{with } z_p : A(z_p) = 0, |z_p| < 1 \quad \forall p \\ C(z) = 1 + c_1 z^{-1} + c_2 z^{-2} + \dots + c_m z^{-m} & \text{with } z_q : C(z_q) = 0, |z_q| \leq 1 \quad \forall q \end{cases}$$

First step of the algorithm: exploiting the fact that both $\alpha(z)$ and $\beta(z)$ are symmetric polynomials, factor both the polynomials

$$\begin{cases} \alpha(z) = r_1 A(z) A(z^{-1}) \\ \beta(z) = r_2 C(z) C(z^{-1}) \end{cases}$$

where $r_1 > 0$, $r_2 > 0$ and the polynomials $A(z)$ and $C(z)$ are monic polynomials as defined above.

Second step of the algorithm: the spectral factorization of $\Phi(z)$ is

$$\Phi(z) = \frac{\beta(z)}{\alpha(z)} = \left(\frac{r_2}{r_1} \right) \cdot \left[\frac{C(z)}{A(z)} \right] \cdot \left[\frac{C(z^{-1})}{A(z^{-1})} \right] = r \cdot W(z) \cdot W(z^{-1})$$

The Resulting Factorization

Given the description of both the polynomials $\beta(z)$ and $\alpha(z)$ as symmetric polynomials, we can obtain the spectral factorization simply using the live function `L8_spectrFactAlg()`

```
[r, Cz_coeffs, Az_coeffs] = L8_spectrFactAlg(betaPoly, alphaPoly)
```

```
r = 0.4444
Cz_coeffs = 1x2
```

```

    1.0000    0.5000
Az_coeffs = 1x2
    1.0000    0.6667

```

Now rewrite the result as symbolic rational function

```

% sort the display output of polynomial function in ascending order, i.e.
% according to the order of each monomial term
sympref('PolynomialDisplayStyle','ascend');

syms z1 % a trick to represent in a compact way 1/z

lambda2xi = sym(r) % the white noise spectrum

```

```

lambda2xi =
    4
    9

```

```

Az1 = poly2sym(fliplr(Az_coeffs), z1) % the polynomial A(z)

```

```

Az1 =
    1 + 2*z1
         3

```

```

Az = subs(Az1,z1,z^(-1))

```

```

Az =
    2
    3*z + 1

```

```

Cz1 = poly2sym(fliplr(Cz_coeffs), z1) % the polynomial C(z)

```

```

Cz1 =
    1 + z1
         2

```

```

Cz = subs(Cz1,z1,z^(-1))

```

```

Cz =
    1
    2*z + 1

```

```

Wz = Cz1/Az1 % the LTI system transfer function as symbolic object

```

```

Wz =
    1 + z1
    -----
    1 + 2*z1
         3

```

The Spectral Shaping Filter

Finally, the transfer function of the LTI filter $W(z)$ is

```
W_z = tf(Cz_coeffs, Az_coeffs, 1, 'Variable', 'z^-1')
```

W_z =

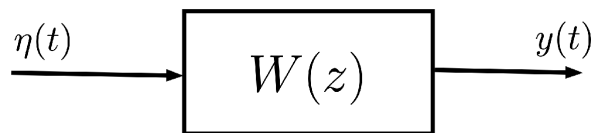
$$\frac{1 + 0.5 z^{-1}}{1 + 0.6667 z^{-1}}$$

Sample time: 1 seconds
Discrete-time transfer function.

In conclusion, the power spectrum

$$\Gamma_y(\omega) = \frac{5 + 4 \cos \omega}{13 + 12 \cos \omega} \quad \omega \in [-\pi, +\pi]$$

is associated to the following model



- a stationary white noise process $\eta(\cdot) \sim \text{WN}\left(0, \frac{4}{9}\right)$
- an LTI stable system, with transfer function $W(z) = \frac{1 + \frac{1}{2}z^{-1}}{1 + \frac{2}{3}z^{-1}}$

Thus we found the **spectral canonical form** of an **ARMA(1, 1) stationary stochastic process**.