

Auto-regressive AR(2) Random Process

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Autoregressive random process (CompEx 3.4)

We consider the autoregressive AR(2) random process given by the difference equation as follow:

$$x(n) = a(1)x(n-1) + a(2)x(n-2) + b(0)v(n) \quad (1)$$

Here $v(n)$ is a unit variance white noise.

1. with $a(1) = 0$, $a(2) = -0.81$ and $b(0) = 1$, generate a 24 sample of the random process $x(n)$. Generate $v(n)$ with the function `randn()`, as the gaussian white noise.

```
N = 24
```

```
testrand = 0.5377
```

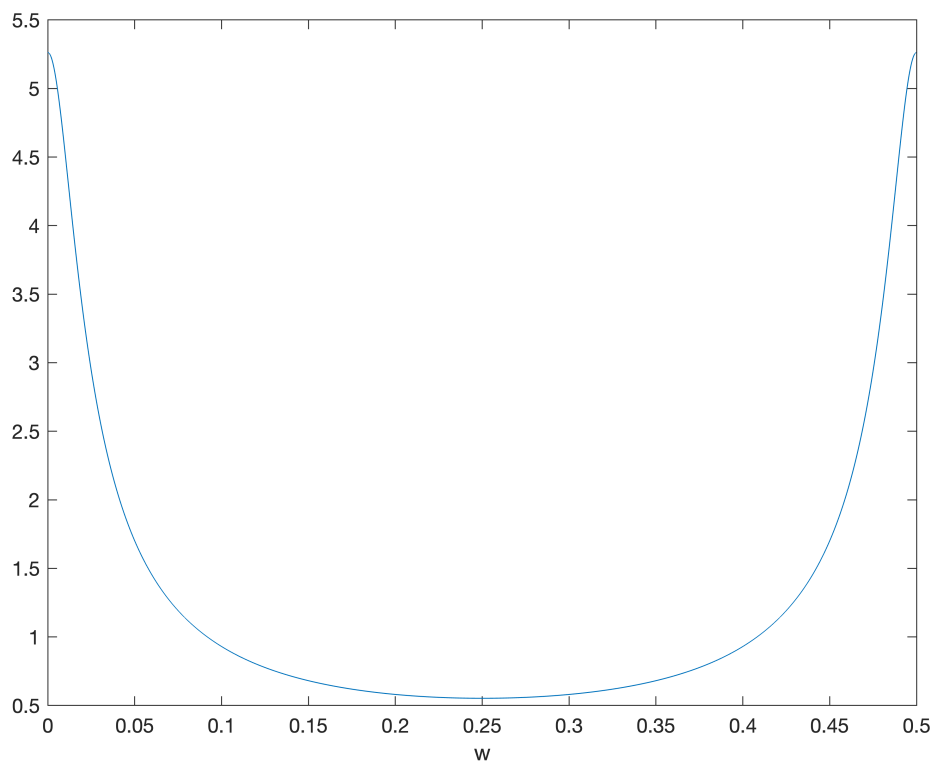
```
vn = 1x24
```

```
1.8339 -2.2588 0.8622 0.3188 -1.3077 -0.4336 0.3426 3.5784 ...
```

```
b = 1
```

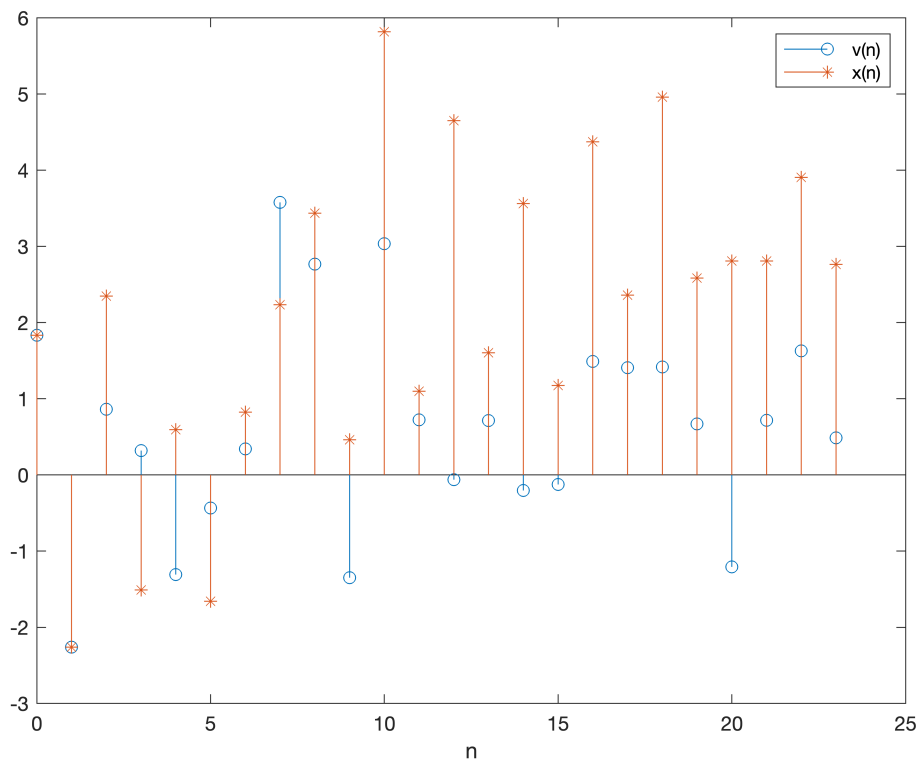
```
a = 1x3
```

```
1.0000 0 -0.8100
```



Compute $x(n)$ with the filter function:

$xn = 1 \times 24$
 1.8339 -2.2588 2.3476 -1.5109 0.5939 -1.6574 0.8237 2.2359 \dots

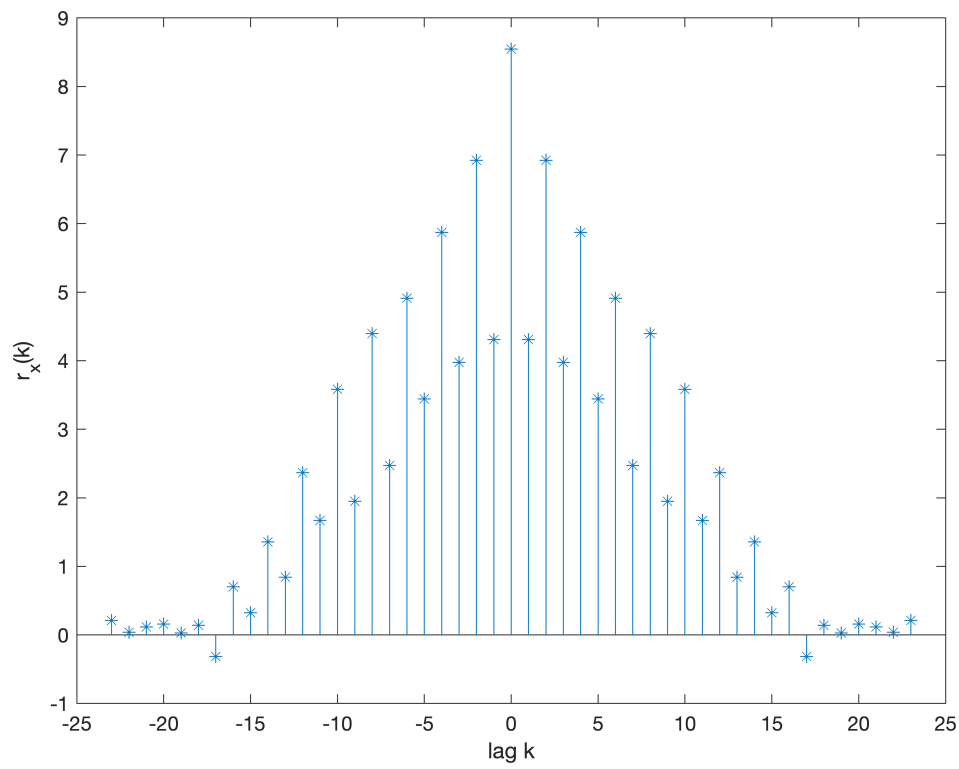


2. Estimate the autocorrelation sequence of $x(n)$ using the sample autocorrelation :

`[rxhat,k] = SampleAutocorr(xn)`

```
rxhat = 1x47
    5.0714    0.9180    2.8214    3.7992    0.7281    3.3481   -7.5824   16.8412 ...
k = 1x47
   -23   -22   -21   -20   -19   -18   -17   -16   -15   -14   -13   -12   -11 ...

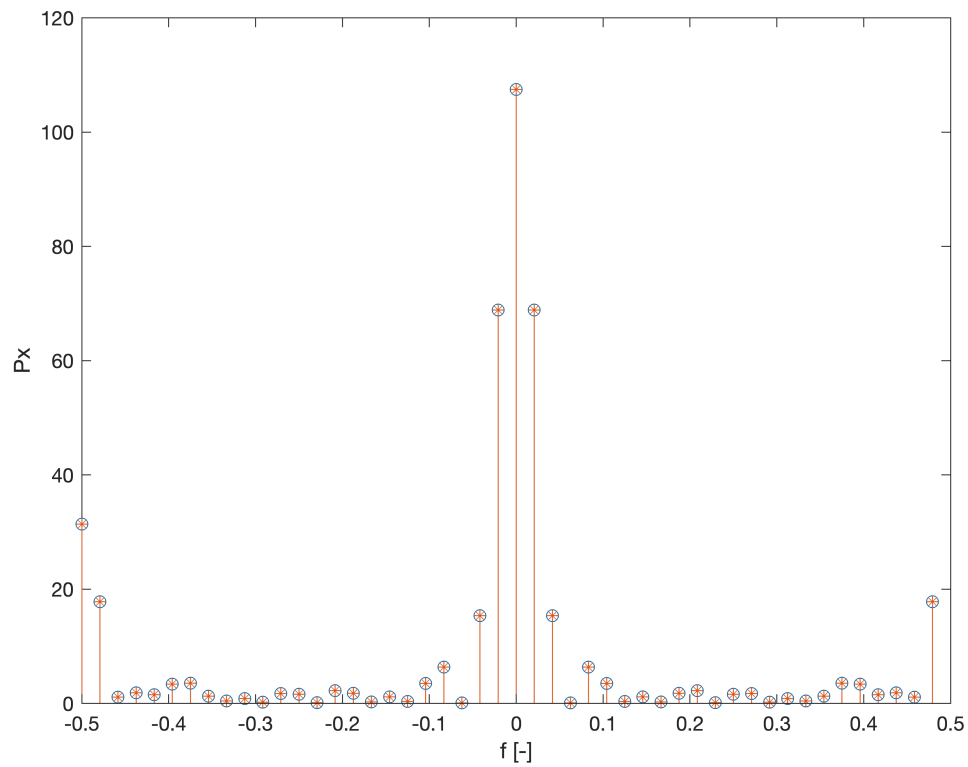
rxhat = 1x47
    0.2113    0.0383    0.1176    0.1583    0.0303    0.1395   -0.3159    0.7017 ...
```



3. Estimate the power spectrum from the sample autocorrelation:

```
w = 1x48
    -3.1416    -3.0107    -2.8798    -2.7489    -2.6180    -2.4871    -2.3562    -2.2253 ...

Px2 = 1x48
     31.3891    17.8194     1.1140     1.8736     1.5195     3.3949     3.5446     1.2839 ...
```



4. From the estimation of the autocorrelation $\hat{r}_x(k)$ use the Youle Walker equations to find the values of the coefficients $a(1)$ and $a(2)$ and $b(0)$ and comment the accuracy of the estimation:

$$\begin{bmatrix} r_x(0) & r_x(-1) & r_x(-2) \\ r_x(1) & r_x(0) & r_x(-1) \\ r_x(2) & r_x(1) & r_x(0) \end{bmatrix} \begin{bmatrix} 1 \\ a(1) \\ a(2) \end{bmatrix} = \sigma_v^2 |b(0)|^2 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad (2)$$

In this case the process is real value and stationary so $r_x(-1) = r_x(1)$.

Take the values computed for \hat{r}_x for $k = 0, 1, 2$ and $\sigma_v^2 = 1$

```
Rx = 3x3
      8.5457    4.3075    6.9229
      4.3075    8.5457    4.3075
      6.9229    4.3075    8.5457
```

```
a2 = -0.7454
```

```
a1 = -0.1283
```

```
b0 = 2.8324
```