

Auto-regressive Random Process

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Estimation of autocorrelation of random process (CompEx 3.2)

In many applications it is necessary to be able to efficiently estimate the autocorrelation sequence of a random process from a finite number of samples e.g. $x(n)$ for $n = 0 \dots N-1$.

The autocorrelation may be estimated using the sample autocorrelation:

$$\hat{r}_x(k) = \frac{1}{N} \sum_{n=0}^{N-1} x(n)x^*(n-k) \quad (1)$$

For this estimation we assume the samples outside the interval $0 \dots N-1$ are equal to 0. Since this is assumed we could write the sample autocorrelation as a convolution :

$$\hat{r}_x(k) = \frac{1}{N} [x(k) * x^*(-k)] \quad (2)$$

The discrete-time Fourier transform of $\hat{r}_x(k)$ is the magnitude squared of the discrete-time Fourier transform of $x(n)$ scaled by $1/N$:

$$\sum_{k=-N+1}^{N-1} \hat{r}_x(k) e^{-jk\omega} = \frac{1}{N} |X(e^{-jk\omega})|^2 \quad (3)$$

where $X(e^{-jk\omega}) = \sum_{n=0}^{N-1} x(n)e^{-jn\omega}$

First experiment : test with a simple signal $x(n)$

with the signal $x(n) = 1$ (deterministic) for $n = 1 \dots N-1$ with $N = 8$ compute the sample autocorrelation using (1) write a function `[rxhat,k]=SampleAutocorr(xn)`

1. Compute the sample autocorrelation by the convolution $1/N (x(k)*x(-k))$ and show that give the same result.
2. Verify the property given by (3) , for that compute the DFT of the sample autocorrelation `rxhat` and compare to the squared of the module DFT of $x(n)$
3. Find the sample autocorrelation using the FFT of $x(n)$ with $N = 8$ or 16 , or 32 samples. Compare the results.
4. Try all the precedent steps with a gaussian white process: $x(n) = \text{randn}(1,N)$

`N = 8;`

`xn = ones(1,N)`

```

xn = 1x8
    1    1    1    1    1    1    1    1

% sample autocorrelation using (1) with a four loop

```

```

[rxhat, k] = SampleAutocorr(xn);

```

```

[k ; rxhat]

```

```

ans = 2x15
    -7.0000    -6.0000    -5.0000    -4.0000    -3.0000    -2.0000    -1.0000         0 ...
         0.1250         0.2500         0.3750         0.5000         0.6250         0.7500         0.8750         1.0000

```

Compute with the function conv() and the function flipr() the autocorrelation of xn:

```

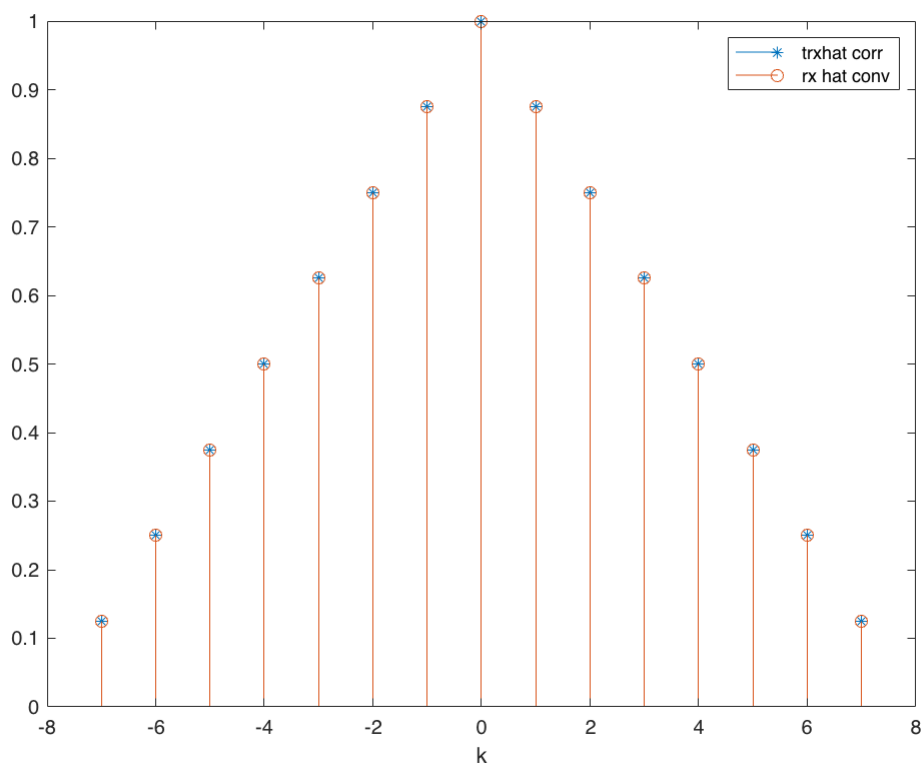
[k;rxhatconv]

```

```

ans = 2x15
    -7.0000    -6.0000    -5.0000    -4.0000    -3.0000    -2.0000    -1.0000         0 ...
         0.1250         0.2500         0.3750         0.5000         0.6250         0.7500         0.8750         1.0000

```



Compute the DFT of the signal x(n) (N samples) with NFFT sample of the DFT (try 8, 16, 32), compute the DFT and compute the square of the magnitude:

For ex for NFFT = 16 we have:

```
NFFT = 16
```

```
w = -pi: 2*pi/NFFT : pi-2*pi/NFFT;
f = w/2/pi
```

```
f = 1x16
-0.5000 -0.4375 -0.3750 -0.3125 -0.2500 -0.1875 -0.1250 -0.0625 ...
Xsq_fft = (abs(fftshift(fft(xn,NFFT)))).^2 /N
[f;Xsq_fft]
```

```
Xsq_fft = 1x16
0 0.1299 0 0.1808 0 0.4050 0 3.2843 ...
```

Compute the DFT of the sample autocorrelation sequence P_x and verify that is real:

For FFT use it is important to have $rxhat$ as a entire periode of a periodic sequence of length $2N$, $rxhat$ must be symetric and has $2N-1$ samples so to ensure simmetricity and have a right computation of the DFT we need to add a 0 sample at the beginning of $rxhat$ previously computed :

$\hat{r}'_x(k) = [0, \hat{r}_x(-N/2 + 1, \dots, \hat{r}_x(0), \dots, \hat{r}_x(N/2 - 1))$ this vector now has $2N$ samples and has the same energy than $\hat{r}_x(k)$

```
rxhat_p = [zeros(1,NFFT/2 - N+1),rxhat,zeros(1,NFFT/2 - N)]
```

```
rxhat_p = 1x16
0 0.1250 0.2500 0.3750 0.5000 0.6250 0.7500 0.8750 ...
Pxfft =fftshift( fft(fftshift(rxhat_p),NFFT))
```

```
Pxfft = 1x16
0 0.1299 0 0.1808 0 0.4050 0 3.2843 ...
```

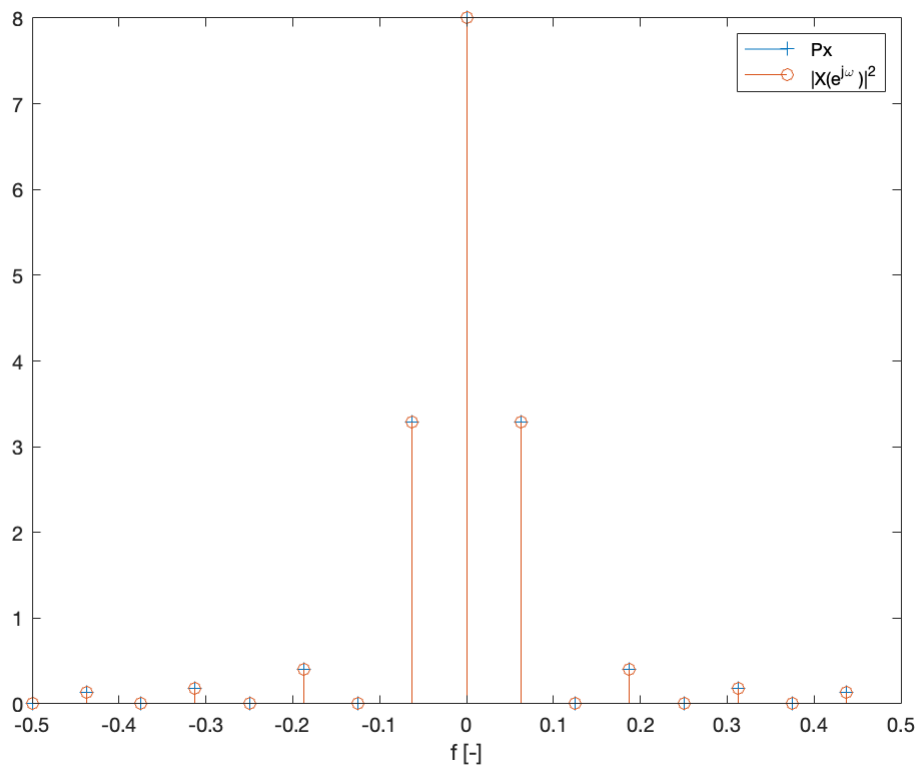
Must be real values and positives ! So take real part after verification.

```
Pxfftr = 1x16
0 0.1299 0 0.1808 0 0.4050 0 3.2843 ...
```

```
[f;Pxfft] = ....
```

```
ans = 2x16
-0.5000 -0.4375 -0.3750 -0.3125 -0.2500 -0.1875 -0.1250 -0.0625 ...
0 0.1299 0 0.1808 0 0.4050 0 3.2843
```

Plot the power spectrum P_x computed with $x(n)$ directly or with the sample autocorrelation sequence:



Compute the sample autocorrelation by the inverse DFT of the power spectrum:

We could compute the sample autocorrelation estimation by the inverse of the squared of the DFT of x_n :

To obtain the same autocorrelation vector you need to remove the useless zeros at the beginning and the end of the vector they were added when the power spectrum was computed.

```
rxhatfft = fftshift(ifft(fftshift(Pxfft)))
```

```
rxhatfft = 1×16
0 0.1250 0.2500 0.3750 0.5000 0.6250 0.7500 0.8750 ...
```

Plot the sample autocorrelation directly computed from x_n and from the power spectrum P_x computed with the square of the DFT magnitude of $x(n)$. Try with $N_{FFT} = 8, 16$ or 32 .

