

2D Auto correlation function

In this document, we explain the method of calculating the auto-correlation function, it is used in statistical studies, signal and image processing, estimation filtering ...etc.

the Wiener-Khinchine theorem [1] (also known as the **Wiener–Khinchine theorem** and sometimes as the **Wiener–Khinchin–Einstein theorem** or the **Khinchin–Kolmogorov theorem**) states that the power spectral density of wide-sense-stationary-random process is the Fourier transform of the corresponding auto-correlation function,

[Norbert Wiener](#) first published this [theorem](#) in 1930, and [Aleksandr Khinchin](#) did so independently in 1934. [Albert Einstein](#) had probably anticipated the idea in a brief two-page memo in 1914 .

Let's consider a 2d signal X , its auto-correlation function $R_x(u, v)$ is related to the power spectral density $S_x(a, b)$ via the theorem [1] :

$$S_x(a, b) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} R_x(a, b) e^{-2\pi i(au+bv)} du dv = F[R_x(a, b)] \quad (1)$$

On the other hand, the power spectral density is the square modulus of the Fourier transform of the 2d signal X :

$$S_x = |F(X)|^2 = F(X) \cdot F(\bar{X}) \quad (2)$$

with $F(X)$, the Fourier transform, and $F(\bar{X})$ the conjugate transform .

$$F(X) = Y(a, b) = \sum_{u=1}^N \sum_{v=1}^M X(u, v) \cdot \exp\left(\frac{-2\pi i ua}{N} + \frac{-2\pi i vb}{M}\right) \quad (3)$$

$$F(\bar{X}) = Y(\bar{a}, b) = \sum_{u=1}^N \sum_{v=1}^M X(u, v) \cdot \exp\left(\frac{2\pi i ua}{N} + \frac{2\pi i vb}{M}\right) \quad (4)$$

From the equations (1) and (2) the Auto-correlation function of the Signal X is given by :

$$R(X) = F^{-1}[F(X) \cdot F(\bar{X})] \quad (5)$$

Application in MATLAB:

Let's take two examples, First let's try to visualize the auto-correlation function of random gray scale image , we choose the 'circuit.tif' :

```
I=imread('circuit.tif');  
I=im2double(I); % convert from uint8 to double
```

Now we implement the equation (5) :

```
B=abs(fftshift(iff2(fft2(I).*conj(fft2(I))))));
```

Note that B and I have the same size 280x272, and for better representation we normalize the auto-correlation function :

```
[n p]=size(I);  
B=B/(n*m);  
figure,imshow(I);  
figure, surf(B), shading interp
```

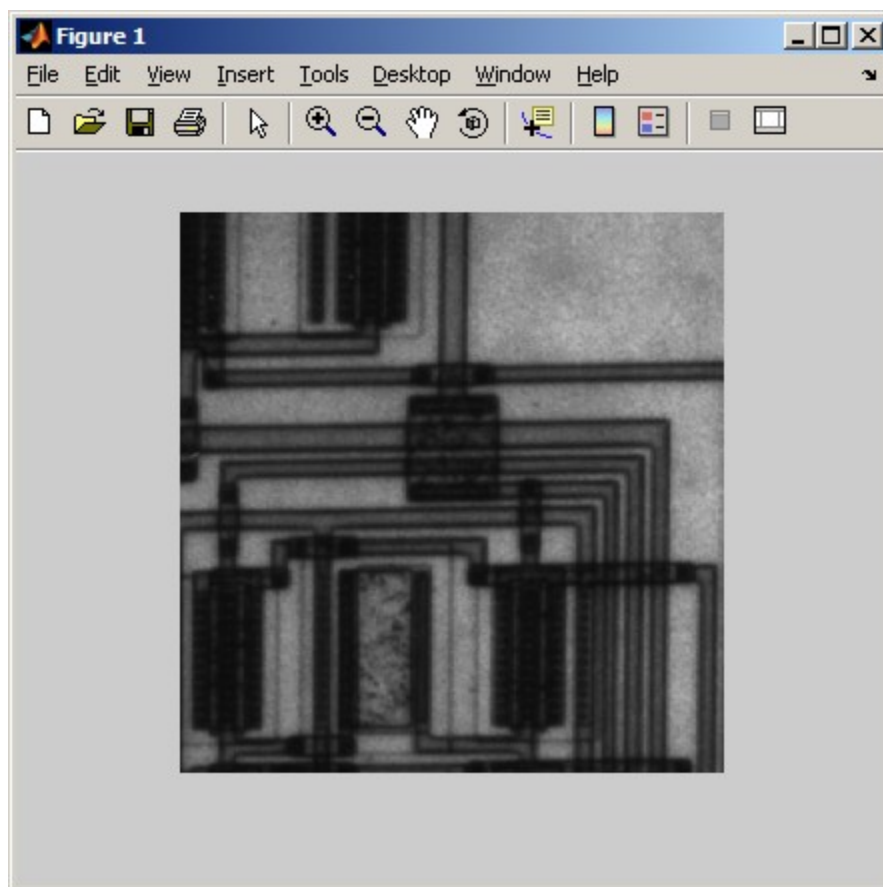


Figure 1 : Image 'circuit.tif'

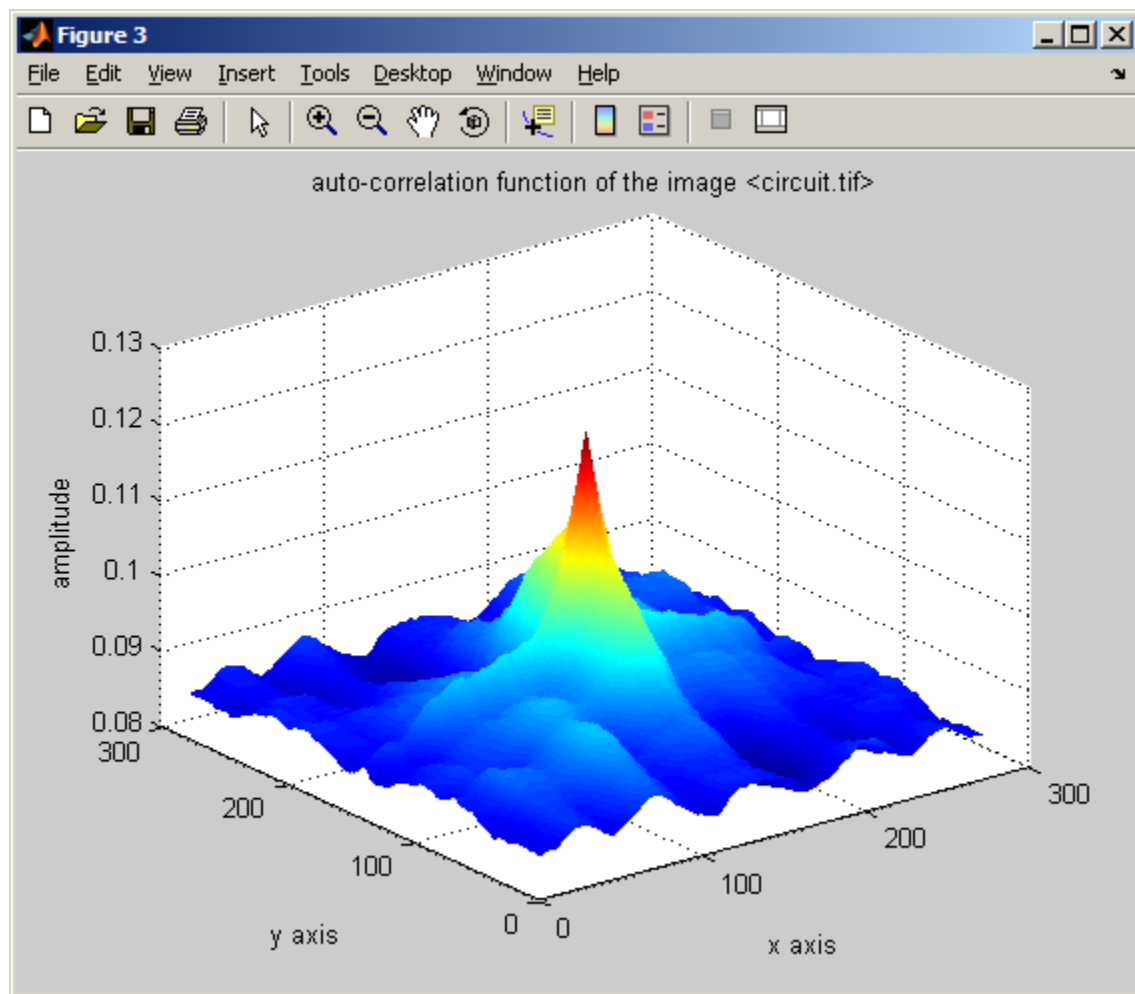


Figure 2 : Auto-correlation of the image 'circuit.tif'

Now , we take special case, a two dimensional Gaussian process, lets' consider a 2d Gaussian signal of size(460,480) with mean=0.0036 and variance = 1.0046 :

```
I2=randn(460,480);
```

```
mean(I2(:))
```

```
ans =
```

```
0.0036
```

```
>> var(I2(:))
```

```
ans =
```

```
1.0046
```

The signal fluctuates around the value zeros as we can see in the figure n° 3:

`surf(I2);`

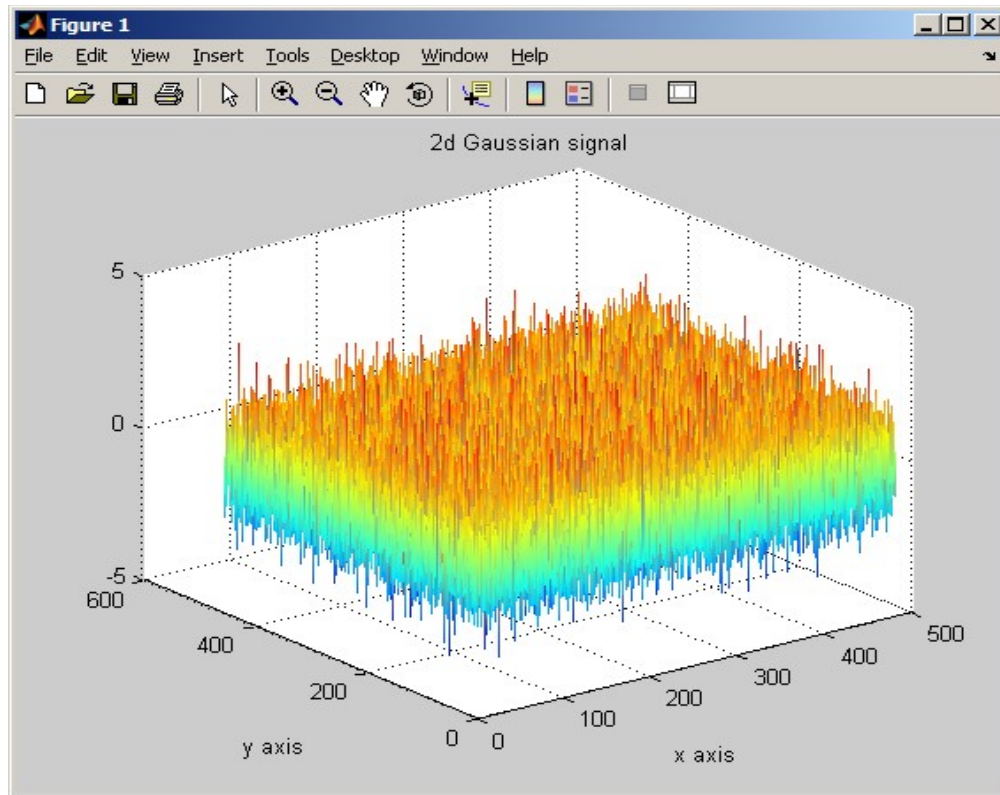


Figure 3 : 2D Gaussian signal

The signal can be seen as an image that contains noise if we set the azimuth to 0 and elevation to 90:
`view(0,90)` :

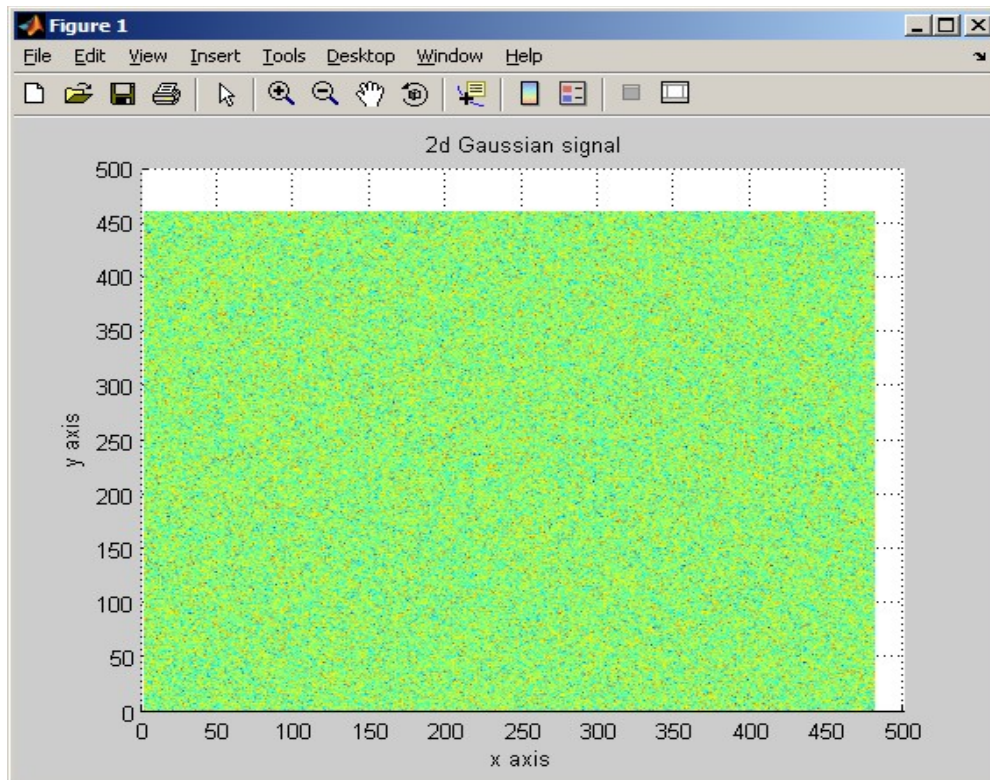
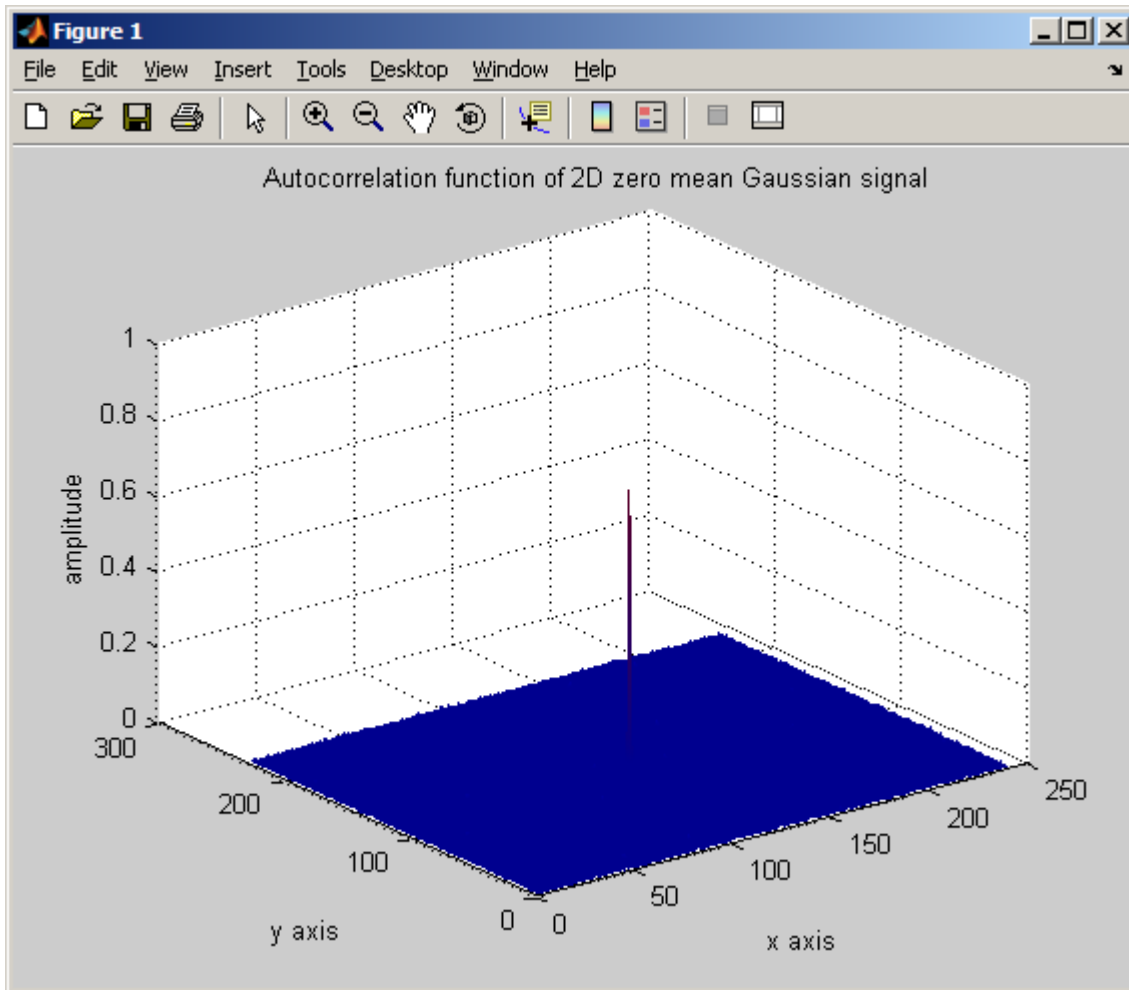


Figure 4 : Image view of the Gaussian signal

Then , if we calculate the auto-correlation of this signal, we find that is has a special form : it is symmetric , with no correlation except in the center which has a 2d Dirac impulsion .

Note : for better visualization we divide the size of signal to have X [230,240] :

```
I2=randn(230,240);
B2=abs(fftshift(iff2(fft2(I2).*conj(fft2(I2)))))./(230*240);
surf(B2)
shading interp
xlabel(' x axis')
ylabel(' y axis')
zlabel(' amplitude')
title(' Autocorrelation function of 2D zero mean Gaussian signal')
```



The auto-correlation of the X signal is : $r(n, m) = \sigma^2 \cdot \delta(n, m)$ with :

σ : the constant power spectral density.

$\delta(n, m)$: 2d Dirac impulsion .

References :

- [1] : http://en.wikipedia.org/wiki/Wiener%E2%80%93Khinchin_theorem#cite_note-0
- [2] :Dennis Ward Ricker (2003). *Echo Signal Processing*. Springer. ISBN 1-4020-7395-X.
- [3] : <http://mathworld.wolfram.com/Wiener-KhinchinTheorem.html>