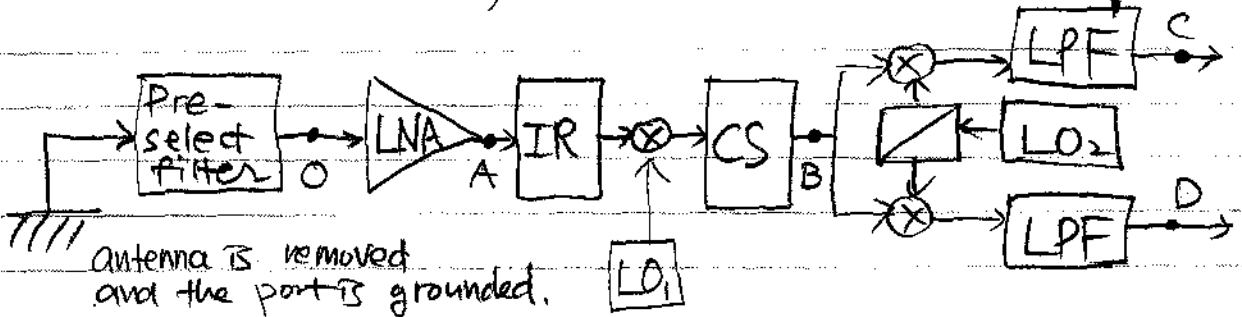


Complex baseband representation of real-valued WSS bandpass random processes Page 1

○ Why real-valued WSS bandpass random processes?

- Consider a heterodyne receiver with **no input**



At point O, there is no signal.

At point A, there is noise generated by the LNA

At point B, there exists the noise component that is within the desired frequency band.

At point C, there is the in-phase component of the noise at B.

At point D, there is " quadrature " " " " " " " B.

• **Noise** generated by the **LNA** (From Dr. M.P. Fitz's lecture note.)

The noise at point A is known to be well modeled as a

**real-valued
stationary-
Gaussian**

} random process with the PSD

given by
$$S_{NN}(f) = \frac{2Rhf|f|}{e^{\frac{hf|f|}{KT}} - 1}$$

$\left\{ \begin{array}{l} S_{NN}(f) \rightarrow 0 \text{ as } f \rightarrow \infty \\ f \ll \infty \\ S_{NN}(f) = 2RKT \end{array} \right.$

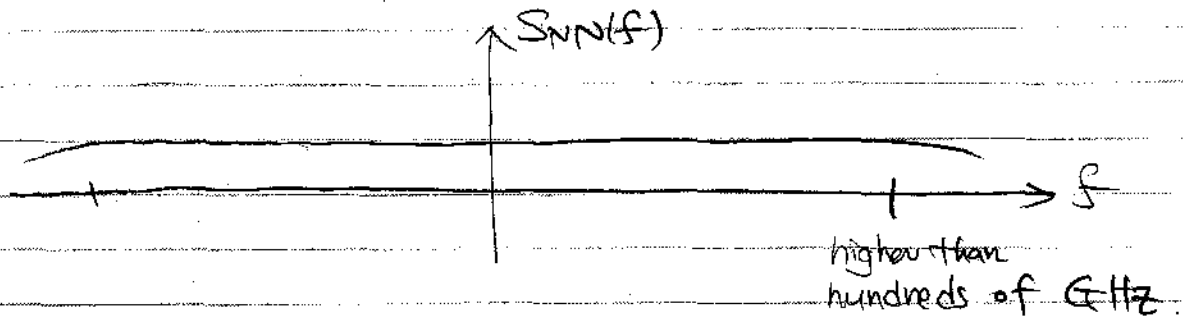
where $h = 6.62 \times 10^{-34}$ is Planck's constant,

$K = 1.3799 \times 10^{-23}$ is Boltzmann's "

R is the resistance

T is the temperature in Kelvin.

If we plot a typical PSD, then



Note that $e^x \approx 1+x$ for small x . (Why?)

So, within the frequency band of interest, the PSD is flat. That's why often we model the noise as a **white noise**. By convention, $S_{NN}(f) = \frac{N_0}{2}$, N_0 is called the one-sided PSD, $\frac{N_0}{2}$ is called the two-sided PSD.

- Noise at point B.

Due to the IR and CS filtering, the noise present at point B is the component of $N(t)$ that is within the desired frequency band. Let it be $X(t)$.

Then, $X(t)$ is a real-valued stationary Gaussian bandpass random process.

- Noise at point C & D.

If we define $X_c(t, \omega)$ and $X_s(t, \omega)$ as the in-phase and quadrature components of $X(t, \omega)$, then

$$X(t) = X_c(t) \cos 2\pi f_c t - X_s(t) \sin 2\pi f_c t$$

Thus, at point C, we have $X_c(t)$,
and at " D, " " " $X_s(t)$

- Noise is one of the most important system impairment in communications. Especially, the **thermal** noise is unavoidable as far as an LNA is used.

Since $X(t)$ is a $\left(\begin{array}{l} \text{real-valued} \\ \text{stationary} \\ \text{Gaussian} \\ \text{bandpass} \end{array} \right)$ random process,

We first study what is the relation between $X(t)$, $X_c(t)$, and $X_s(t)$ when $X(t)$ is modeled as a $\left(\begin{array}{l} \text{real-valued} \\ \text{wide-sense stationary} \\ \text{bandpass} \end{array} \right)$ random process

We then include the Gaussian assumption. ^{Note that} if Gaussianity is assumed, wide-sense stationarity implies stationarity. then the

○ The main results (in time domain)

- Let $X(t)$ be a real-valued, WSS, bandpass random process given by

$$X(t) = X_c(t) \cos 2\pi f_c t - X_s(t) \sin 2\pi f_c t$$

- Then, (i) $E[X_c(t)] = E[X_s(t)] = 0, \forall t$
 (ii) $E[X_c(t)X_c(t+\tau)] = E[X_s(t)X_s(t+\tau)] = R_{X_c X_c}(\tau)$
 and (iii) $E[X_c(t)X_s(t+\tau)] = R_{X_c X_s}(\tau) = -R_{X_c X_s}(-\tau), \forall t, \tau$

All three results are interesting.

(i) A real-valued WSS bandpass process has **mean zero**

(ii) $X_c(t)$ and $X_s(t)$ are **WSS** and have the **same autocorrelation** function

(iii) $X_c(t)$ and $X_s(t)$ are **jointly WSS** and have the **cross-correlation** function that is **odd**.

Proof of (i)

Let $E[X_c(t)] \triangleq \mu_c(t)$ and $E[X_s(t)] \triangleq \mu_s(t)$.
Then $\mu_c(t)$ & $\mu_s(t)$ have bandwidth less than f_c .

Thus,

$$E[X(t)] = E[X_c(t) \cos 2\pi f_c t - X_s(t) \sin 2\pi f_c t]$$

$$= \text{Re}[\mu_x(t) e^{j2\pi f_c t}] \quad \dots (*)$$

where $\mu_x(t) \triangleq \mu_c(t) + j\mu_s(t)$.

By the wide-sense stationarity, (*) must be a constant μ for all t . Since (*) does not have any DC term, μ must be zero $\forall t$.

Thus, $\mu_x(t) = \mu_c(t) + j\mu_s(t) = 0, \forall t$

which implies $\mu_c(t) = \mu_s(t) = \mu_x(t) = \mu(t) = 0, \forall t$.

Think it in the freq. domain.

Proof of (ii) & (iii)

As we used the wide-sense stationarity to prove (i), we use the fact ^{again} to prove (ii) & (iii)

$$\begin{aligned}
 R_{xx}(\tau) &= E[x(t)x(t+\tau)] \\
 &= E\left[\frac{x_e(t)e^{j2\pi f_c t} + x_e(t)^*e^{-j2\pi f_c t}}{2} \times \right. \\
 &\quad \left. \frac{x_e(t+\tau)e^{j2\pi f_c(t+\tau)} + x_e(t+\tau)^*e^{-j2\pi f_c(t+\tau)}}{2}\right] \\
 &= \frac{1}{2} E\left[\frac{(x_e(t)x_e(t+\tau)e^{j2\pi f_c(2t+\tau)} + x_e(t)^*x_e(t+\tau)^*e^{-j2\pi f_c(2t+\tau)})}{2} + \right. \\
 &\quad \left. \frac{(x_e(t)x_e(t+\tau)^*e^{-j2\pi f_c\tau} + x_e(t)^*x_e(t+\tau)e^{j2\pi f_c\tau})}{2}\right] \\
 &= \frac{1}{2} \operatorname{Re}\{E[x_e(t)x_e(t+\tau)]e^{j2\pi f_c(2t+\tau)}\} \\
 &\quad + \frac{1}{2} \operatorname{Re}\{E[x_e(t)^*x_e(t+\tau)]e^{j2\pi f_c\tau}\} \quad \forall t, \tau.
 \end{aligned}$$

By the defn of WSS,

$$\begin{aligned}
 \therefore E[x_e(t)x_e(t+\tau)] &= 0 \quad \forall t, \tau \text{ \& } \\
 E[x_e(t)^*x_e(t+\tau)] &= R_{x_e}(\tau)
 \end{aligned}$$

... (*)
 ... (**)
 pseudo-autocorrelation function of the complex envelope $x_e(t)$.

Note that (*) implies

$$\begin{aligned}
 E[(x_c(t) + jx_s(t))(x_c(t+\tau) + jx_s(t+\tau))] \\
 = R_{x_c x_c}(t, t+\tau) - R_{x_s x_s}(t, t+\tau) \\
 + jR_{x_s x_c}(t, t+\tau) + jR_{x_c x_s}(t, t+\tau) \\
 = 0, \quad \forall t, \tau.
 \end{aligned}$$

Thus, $R_{x_c x_c}(t, t+\tau) = R_{x_s x_s}(t, t+\tau)$ and \dots (***)
 $R_{x_s x_c}(t, t+\tau) = -R_{x_c x_s}(t, t+\tau)$

Note also that (**) implies

$$\begin{aligned}
 E[(x_c(t) - jx_s(t))(x_c(t+\tau) + jx_s(t+\tau))] \\
 = R_{x_c x_c}(t, t+\tau) + R_{x_s x_s}(t, t+\tau) \\
 - jR_{x_s x_c}(t, t+\tau) + jR_{x_c x_s}(t, t+\tau) \\
 = R_{x_e}(\tau) \quad \forall t, \tau.
 \end{aligned}$$

Thus, combined w/ (144), we obtain

$$\begin{cases} R_{xx}(t) = 2 R_{xc}x_c(t) + j 2 R_{cx}x_s(t) \\ R_{xc}x_c(t) = R_{xs}x_s(t) \\ R_{cx}x_s(t) = -R_{xs}x_c(t) = -R_{cx}x_s(-t) \\ R_{xx}(t) = R_{xc}x_c(t) \cos 2\pi f_c t \\ \quad - R_{cx}x_c(t) \sin 2\pi f_c t \end{cases}$$

Remarks

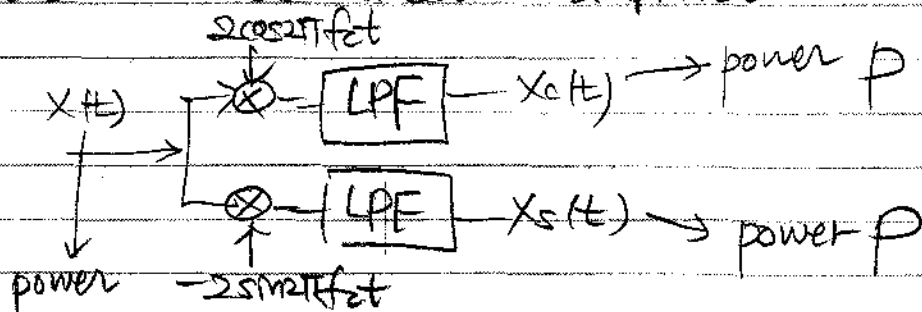
- We naturally define in this course the autocorrelation function of a complex-valued random process $X_c(t)$ as

$$R_{X_c X_c}(t, t+\tau) \triangleq R_{X_c X_c}(t, t+\tau)$$

- The autocorrelation function of $X(t)$ can be written as

$$\begin{aligned} R_{XX}(t) &= \text{Re} \left\{ \left(R_{xc}x_c(t) + j R_{cx}x_s(t) \right) e^{j 2\pi f_c t} \right\} \\ &= \text{Re} \left\{ \frac{R_{X_c X_c}(t)}{2} e^{j 2\pi f_c t} \right\} \end{aligned}$$

- The term $\frac{1}{2}$ comes from the fact that the ideal IQ demodulator amplifies the power by 2.



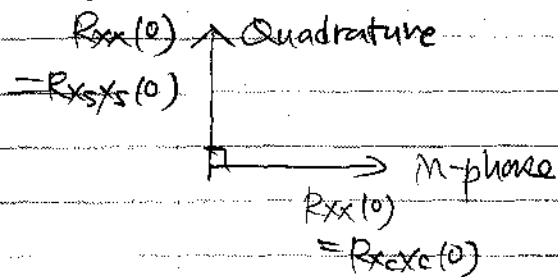
$$P \triangleq R_{XX}(0) = \frac{1}{2} \text{Re} \left\{ R_{X_c X_c}(0) \right\} = \frac{1}{2} \left(2 R_{xc}x_c(0) + j 2 R_{cx}x_s(0) \right) = R_{xc}x_c(0)$$

The cross power is zero, i.e. $E[X_c(t)X_s(t)] = 0, \forall t$

$$\therefore R_{X_c X_s}(\tau) = -R_{X_c X_s}(-\tau) \Rightarrow R_{X_c X_s}(0) = 0$$

To the contrary,

$$R_{X_c X_c}(0) = R_{X_s X_s}(0) = R_{XX}(0)$$



If $X(t)$ is Gaussian, then $X_c(t)$ & $X_s(t)$ are also Gaussian w/

$$E\{X_c(t)^2\} = E\{X_s(t)^2\} = E\{X(t)^2\} \triangleq \sigma^2$$

$$E\{X_c(t)X_s(t)\} = 0$$

$$\therefore f_{X_c(t)X_s(t)}(x, y) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right)$$

↑
circularly symmetric 2-D Gaussian random vector $X_c(t)$ & $X_s(t)$ for each t are zero-mean & i.i.d Gaussian.

$$R_{X_c X_c}(\tau) = E[X_c(t)^* X_c(t+\tau)] = \left(E[X_c(t+\tau)^* X_c(t)]\right)^*$$

$$= R_{X_c X_c}(-\tau)^*$$

The auto-correlation function is conjugate symmetrical. (actually, for any complex-valued WSS random process.)

$$R_{X_c X_c}(\tau) \triangleq E[X_c(t) X_c(t+\tau)] = 0$$

The pseudo-auto-correlation function is always zero. (not for arbitrary complex-valued WSS random process. A sufficient condition for this to be true is

"the complex-valued WSS process is the complex envelope of a real-valued WSS bandpass process.")

○ Proper-complex random

- Def. A complex-valued random process $X(t)$ is called proper-complex if its pseudo-autocorrelation function is zero, i.e., $E\{X(t)X(t+\tau)\} = 0$, $\forall t, \tau$.
- Def. A complex-valued random vector X is called proper-complex if its pseudo-covariance matrix is zero, i.e.,

$$E\{(X-\bar{X})(X-\bar{X})^T\} = \underline{0}$$

- We will study ^{later} more about proper-complex random vectors & processes, especially for Gaussian cases.

○ The main result (in frequency domain)

- We interpret the main results derived in the time domain now in the frequency domain.

main results

$$\begin{cases} R_{x_c x_c}(\tau) = R_{x_s x_s}(\tau) = R_{x_c x_c}(-\tau) & \dots (*) \\ R_{x_c x_s}(\tau) = -R_{x_c x_s}(-\tau) = -R_{x_s x_c}(\tau) & \dots (**) \\ R_{xx}(\tau) = R_{x_c x_c}(\tau) \cos 2\pi f_c \tau \\ \quad - R_{x_c x_c}(\tau) \sin 2\pi f_c \tau & \dots (***) \end{cases}$$

The PSD of $X_c(t)$ & $X_s(t)$

From $(*)$, $S_{X_c X_c}(f) = S_{X_s X_s}(f) = S_{X_c X_c}(-f)$ real & even $(*)'$

The cross PSD b/w $X_c(t)$ & $X_s(t)$

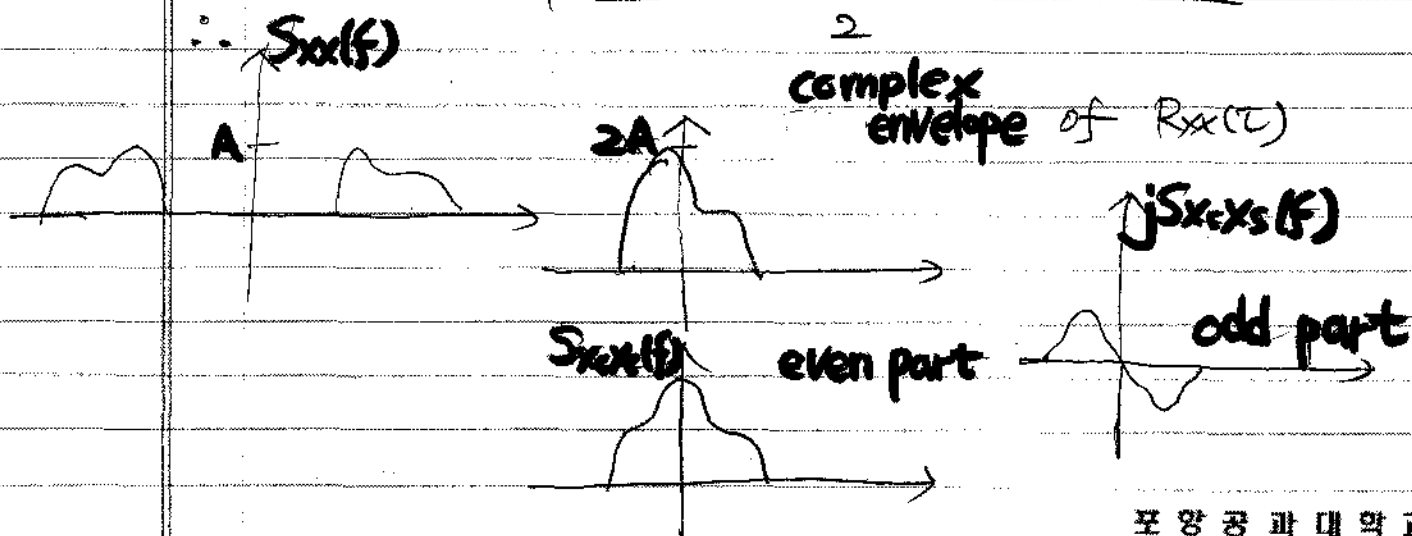
From $(**)$, $S_{X_c X_s}(f) = -S_{X_c X_s}(-f) = -S_{X_s X_c}(f)$ pure imaginary & odd $(***)'$

Relation among $S_{XX}(f)$, $S_{X_c X_c}(f)$, and $S_{X_c X_s}(f)$

From $(***)$, $(*)'$, and $(***)'$, we obtain

$$S_{XX}(f) = \frac{S_{X_c X_c}(f-f_c) + S_{X_c X_c}(f+f_c)}{2} - \frac{S_{X_c X_s}(f-f_c) - S_{X_c X_s}(f+f_c)}{2j}$$

$$= \frac{S_{X_c X_c}(f-f_c) + j S_{X_c X_s}(f-f_c)}{2} + \frac{S_{X_c X_c}(-(f+f_c)) + j S_{X_c X_s}(-(f+f_c))}{2}$$



- We can view $R_{xx}(\tau)$ as a real-valued bandpass signal. Then, $R_{xxc}(\tau)$ is the in-phase component and $R_{xxs}(\tau)$ is the quadrature component.

Thus,

$$\mathcal{F}\{R_{xxc}(\tau)\} = \text{even part of } \mathcal{F}\{R_{xx}(\tau)\}$$

$$j\mathcal{F}\{R_{xxs}(\tau)\} = \text{odd part of } \mathcal{F}\{R_{xx}(\tau)\}$$

Caution

$$R_{xx}(\tau) \triangleq \text{Re}\{R_{xxc}(\tau)e^{j2\pi f_c\tau}\}$$

$$= \text{Re}\left\{\frac{R_{xxc}(\tau)}{2}e^{j2\pi f_c\tau}\right\}$$

the complex envelope of the autocorrelation $R_{xx}(\tau)$

the autocorrelation of the complex envelope $X_c(t)$.

$$R_{xxc}(\tau) = \frac{R_{xx}(\tau)}{2}$$

