

Principles of Digital Communication

May 21, 2021



"Appendices" of Chapter 7, complex valued RVs.
(& random vectors).

Z is a complex RV $\equiv \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix}$ is a Real
random
vector.

$Z \in \mathbb{C}^n$ " " " vector $\equiv \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix}$ is a $2n$ dim
Real R-vector.

Z is a Gaussian $\mathbb{C}^n$

$\equiv \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix}$ is $\mathbb{R}^{2n}$ - Gaussian.

Then, to specify the statistical properties of Z
we should specify

①. $E \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix}$ and covariance matrix,

$$\textcircled{2}. E \left[\begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix} \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix}^T \right] = \left(E \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix} \right) \left(E \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix} \right)^T$$

Assume, for simplicity notation that

$E \begin{bmatrix} \text{Re}(Z) \\ \text{Im}(Z) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ so the covariance
matrix is

$$\equiv \begin{pmatrix} \boxed{\text{Re}(z) \text{Re}(z)^T} & \boxed{\text{Re}(z) \text{Im}(z)^T} \\ \text{Im}(z) \text{Re}(z)^T & \boxed{\text{Im}(z) \text{Im}(z)^T} \end{pmatrix}$$

we need four $n \times n$ real matrices to specify.

$$E[\hat{z} \hat{z}^T] \quad \text{with} \quad \hat{z} = \begin{bmatrix} \text{Re}(z) \\ \text{Im}(z) \end{bmatrix}$$

compare this with

$$\begin{aligned} E[\hat{z} \hat{z}^*] &= E[(\text{Re}(z) + j \text{Im}(z))(\text{Re}(z)^T - j \text{Im}(z)^T)] \\ &= \underbrace{E[\text{Re}(z) \text{Re}(z)^T] + E[\text{Im}(z) \text{Im}(z)^T]}_{\text{}} \\ &\quad + j \underbrace{[-E(\text{Re}(z) \text{Im}(z)^T) + E(\text{Im}(z) \text{Re}(z)^T)]}_{\text{}} \end{aligned}$$

Knowing $E[\hat{z} \hat{z}^*]$ is $\equiv$ to know:

$$\left\{ \begin{aligned} &E(\text{Re}(z) \text{Re}(z)^T) + E(\text{Im}(z) \text{Im}(z)^T) \\ &\& E(\text{Re}(z) \text{Im}(z)^T) - E(\text{Im}(z) \text{Re}(z)^T). \end{aligned} \right\}$$

Suppose we also knew

$$\begin{aligned} E[\hat{z} \hat{z}^T] &= E[(\text{Re}(z) + j \text{Im}(z))(\text{Re}(z)^T + j \text{Im}(z)^T)] \\ &= \underbrace{(E(\text{Re}(z) \text{Re}(z)^T) - E(\text{Im}(z) \text{Im}(z)^T))}_{\text{}} \\ &\quad + j \underbrace{(E(\text{Re}(z) \text{Im}(z)^T) + E(\text{Im}(z) \text{Re}(z)^T))}_{\text{}} \end{aligned}$$

So, we also know

$$\rightarrow \left\{ \begin{array}{l} E(\operatorname{Re}(z)\operatorname{Re}(z)^T) - E(\operatorname{Im}(z)\operatorname{Im}(z)^T) \\ E(\operatorname{Re}(z)\operatorname{Im}(z)^T) + E(\operatorname{Im}(z)\operatorname{Re}(z)^T) \end{array} \right\}$$

So, to specify the statistics of z , all we

need is $\underbrace{E[zz^*]}_{\text{variance}} \text{ and } \underbrace{E[zz^T]}_{\text{zero mean}}.$

We define a special class of RVs:

We say that a RV (or Rvec) is "proper" if

$$\underbrace{E[zz^T]}_{\text{zero mean}} = 0.$$

Thus, for a proper Rvec, $E[zz^*]$ is sufficient to specify its statistics,
 ↑ Gaussian, zero-mean

More generally, we say that a Rvector z is "circularly symmetric" if for every $\theta \in [0, 2\pi)$

$$z \sim \underbrace{ze^{j\theta}}$$

if z is circ. sym, then:

$$E[z] = 0. \quad (E[z] = E[ze^{j\theta}] = \underbrace{E[z]}_{=0} e^{j\theta})$$

$$E[zz^T] = 0 \quad (E[zz^T] = E[ze^{j\theta}(ze^{j\theta})^T])$$

$$\underbrace{E(z z^T)} = \underbrace{E(z z^T)} \cdot e^{j 2\theta} \Rightarrow E(z z^T) = 0.$$

So a circ-sym R vect is zero-mean & proper.

~~For a Gaussian R Vector~~

Moreover if $z \in \mathbb{C}^n$ is Gaussian, zero-mean & proper then z is circ symmetric.

why: $z e^{j\theta}$ is also gaussian,

has zero mean, $E((z e^{j\theta})(z e^{j\theta})^T) =$

$$E(z z^T) e^{j 2\theta} = 0 = \underline{\underline{E(z z^T)}}.$$

$$E(\underbrace{(z e^{j\theta})}_{\text{}})(z e^{j\theta})^* = E(\underbrace{z z^*}_{\text{}})$$

So (z) & $(z e^{j\theta})$ have the same 1st & 2nd order statistics $\Rightarrow$ they have the same distrib.
↑
Gaussian

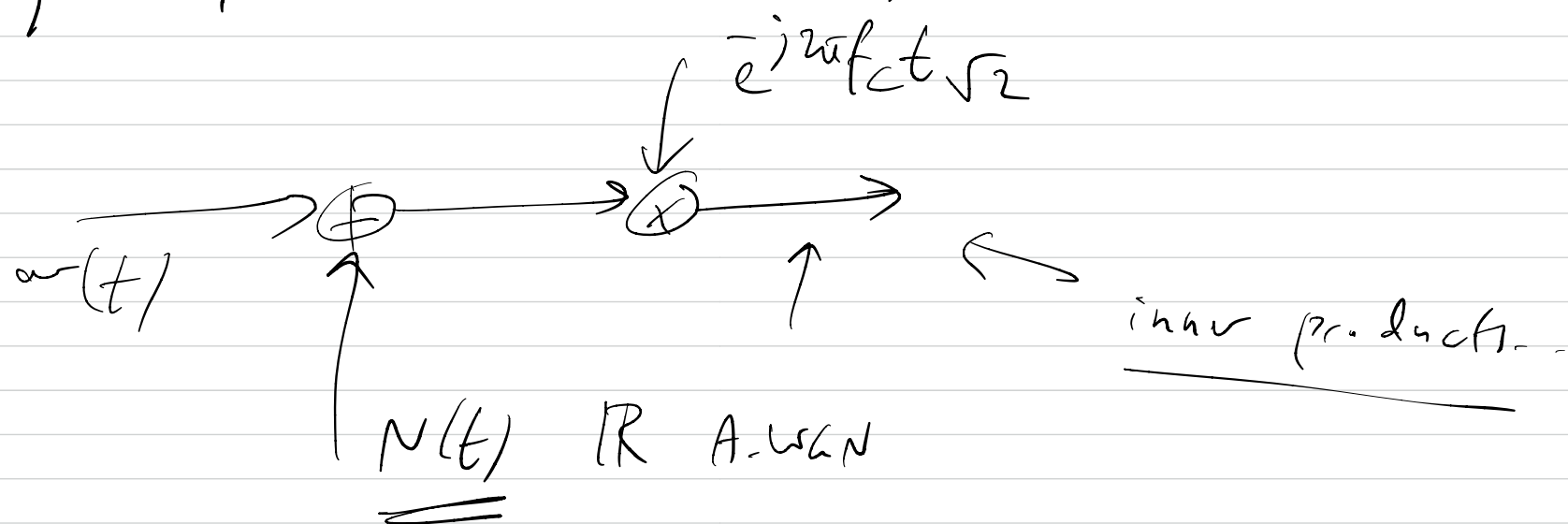
Suppose z is a circ-sym Gaussian with

$K = E(z z^*)$, the pdf of z is given then

$$\text{by: } f(z) = (\det(\pi K))^{-1} \exp(-z^* K^{-1} z), \quad \leftarrow$$

[Compare to the real case $z \sim N(0, K)$ (Real)
 $f_z(z) = \det(\pi K)^{-1/2} \exp(-\frac{1}{2} z^T K^{-1} z)$]

Why worry about the circ. sym Gaussian?



the noise after the inner product:

$$z_j = (\underbrace{N(t)} e^{j2\pi f_c t \sqrt{2}}, \underbrace{g_j(t)}) \quad j=1 \dots n$$

if $N(t)$ is R-WGN

$g_j(t)$ Bandlimited to $[-B, B)$ with $B \ll f_c$

then $\begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$ is circ. sym. Gaussian & rand. vector,

with $\underline{E[z z^*]} =$

$$\begin{bmatrix} \langle g_1, g_1 \rangle & \dots & \langle g_1, g_n \rangle \\ \vdots & & \vdots \\ \langle g_n, g_1 \rangle & \dots & \langle g_n, g_n \rangle \end{bmatrix}$$