

Revision

- gaussian

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$f_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Lambda|^{\frac{1}{2}}} e^{-\frac{1}{2|\Lambda|} \sum_{i=1}^n \sum_{j=1}^n \Lambda_{ij} (x_i - \mu_i)(x_j - \mu_j)}$$

$$f_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_{X_1}\sigma_{X_2}\sqrt{1-r^2}} e^{-\frac{\left(\frac{x_1 - EX_1}{\sigma_{X_1}}\right)^2 - 2r\left(\frac{x_1 - EX_1}{\sigma_{X_1}}\right)\left(\frac{x_2 - EX_2}{\sigma_{X_2}}\right) + \left(\frac{x_2 - EX_2}{\sigma_{X_2}}\right)^2}{2(1-r^2)},$$

$$\text{where } r = \frac{\text{Cov}(X_1, X_2)}{\sigma_{X_1}\sigma_{X_2}} = \frac{E(X_1X_2) - EX_1EX_2}{\sigma_{X_1}\sigma_{X_2}}$$

- If $\text{Cov}(X_1, X_2) = 0$ (or $X_1 \perp X_2$ which also make $\text{Cov}(X_1, X_2) = 0$),

$$f_{X_1, X_2}(x_1, x_2) = f_{X_1}(x_1) \times f_{X_2}(x_2) = \frac{1}{2\pi\sigma_{X_1}\sigma_{X_2}} e^{-\frac{\left(\frac{x_1 - EX_1}{\sigma_{X_1}}\right)^2 + \left(\frac{x_2 - EX_2}{\sigma_{X_2}}\right)^2}{2}}$$

- Independent Gaussian random variables are always jointly Gaussian.

Complex Random Processes

- **Complex random variable** : $Z = X + jY$
where X and Y are real random variable
- **Complex random process**: $\{Z(t)\} = \{X(t) + jY(t)\}$
- **Gaussian**
 - Z is a **complex Gaussian random variable** if X and Y are jointly Gaussian
 - $\{Z(t)\}$ is a **complex Gaussian random process** if
any vector $\underline{Z} = (Z(t_1), \dots, Z(t_n)) = (X(t_1) + jY(t_1), \dots, X(t_n) + jY(t_n))$
is built from $2n$ real jointly Gaussian random variable.
- **Vanderkulk's Lemma** : The complex r.v. $Z = X + jY$ is zero mean and Gaussian, then
 $E e^{jvZ} = e^{-\frac{1}{2}v^2 E Z^2}$

The complex r.v. $Z = X + jY$ is zero mean and Gaussian.

First, prove that $X|Y \sim N\left(\frac{s_X}{s_Y} r y, s_X^2 (1-r^2)\right)$.

$$\text{We have } f_{X,Y}(x,y) = \frac{1}{2\pi s_X s_Y \sqrt{1-r^2}} e^{-\frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + \left(\frac{y}{s_Y}\right)^2}{2(1-r^2)}}$$

$$\text{where } r = \frac{\text{Cov}(X,Y)}{s_X s_Y} = \frac{E(XY) - EXEY}{s_X s_Y} = \frac{E(XY)}{s_X s_Y}.$$

$$\text{And } f_Y(y) = \frac{1}{\sqrt{2\pi} s_Y} e^{-\frac{y^2}{2s_Y^2}}.$$

$$\begin{aligned} \text{Then, } f_{X|Y}(x|y) &= \frac{f_{X,Y}(x,y)}{f_Y(y)} = \frac{\frac{1}{2\pi s_X s_Y \sqrt{1-r^2}} e^{-\frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + \left(\frac{y}{s_Y}\right)^2}{2(1-r^2)}}}{\frac{1}{\sqrt{2\pi} s_Y} e^{-\frac{y^2}{2s_Y^2}}} \\ &= \frac{1}{\sqrt{2\pi} s_X \sqrt{1-r^2}} e^{-\left(\frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + \left(\frac{y}{s_Y}\right)^2}{2(1-r^2)} - \frac{y^2}{2s_Y^2}\right)} \end{aligned}$$

$$\begin{aligned} \text{Note that } & \frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + \left(\frac{y}{s_Y}\right)^2}{2(1-r^2)} - \frac{y^2}{2s_Y^2} \\ &= \frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + \left(\frac{y}{s_Y}\right)^2 - (1-r^2)\frac{y^2}{s_Y^2}}{2(1-r^2)} \\ &= \frac{\left(\frac{x}{s_X}\right)^2 - 2r\left(\frac{xy}{s_X s_Y}\right) + r^2 \frac{y^2}{s_Y^2}}{2(1-r^2)} = \frac{\left(\frac{x}{s_X} - \frac{r y}{s_Y}\right)^2}{2(1-r^2)} \end{aligned}$$

Thus,

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{1}{\sqrt{2\pi} s_X \sqrt{1-r^2}} e^{-\frac{\left(\frac{x}{s_X} - \frac{r y}{s_Y}\right)^2}{2(1-r^2)}} = \frac{1}{\sqrt{2\pi} s_X \sqrt{1-r^2}} e^{-\frac{\left(\frac{x - \frac{s_X}{s_Y} r y}{s_X}\right)^2}{2s_X^2 (1-r^2)}} \\ &= f_G(x) \quad ; \text{where } G \sim N\left(\frac{s_X}{s_Y} r y, s_X^2 (1-r^2)\right) \end{aligned}$$

$$\begin{aligned}
Ee^{jvZ} &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{jv(x+jy)} f_{X,Y}(x,y) dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{jvx-vy} f_Y(y) f_{X|Y}(x|y) dx dy \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{jvx} e^{-vy} f_Y(y) f_{X|Y}(x|y) dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{jvx} e^{-vy} f_Y(y) f_G(x) dx dy \\
&= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{jvx} f_G(x) dx \right) e^{-vy} f_Y(y) dy = \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{jvg} f_G(g) dg \right) e^{-vy} f_Y(y) dy \\
&= \int_{-\infty}^{\infty} (Ee^{jvG}) e^{-vy} f_Y(y) dy = \int_{-\infty}^{\infty} f_G(v) e^{-vy} f_Y(y) dy \\
&= \int_{-\infty}^{\infty} \left(e^{jv \frac{s_X}{s_Y} r y} e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} \right) e^{-vy} \frac{1}{\sqrt{2p} s_Y} e^{-\frac{y^2}{2s_Y^2}} dy \\
&= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} \frac{1}{\sqrt{2p} s_Y} \int_{-\infty}^{\infty} e^{j \frac{s_X}{s_Y} r v y} e^{-vy} e^{-\frac{y^2}{2s_Y^2}} dy = e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} \frac{1}{\sqrt{2p} s_Y} \int_{-\infty}^{\infty} e^{j \frac{s_X}{s_Y} r v y} e^{-\left(\frac{y^2}{2s_Y^2} + vy\right)} dy
\end{aligned}$$

Consider the exponent,

$$\frac{y^2}{2s_Y^2} + vy = \frac{1}{2s_Y^2} (y^2 + 2y(s_Y^2 v)) = \frac{1}{2s_Y^2} ((y + s_Y^2 v)^2 - s_Y^4 v^2) = \frac{1}{2s_Y^2} (y + s_Y^2 v)^2 - \frac{1}{2} v^2 s_Y^2$$

Hence,

$$\begin{aligned}
Ee^{jvZ} &= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} \frac{1}{\sqrt{2p} s_Y} e^{\frac{1}{2} v^2 s_Y^2} \int_{-\infty}^{\infty} e^{j \frac{s_X}{s_Y} r v y} e^{-\frac{1}{2s_Y^2} (y + s_Y^2 v)^2} dy \\
&= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} \int_{-\infty}^{\infty} e^{j \left(\frac{s_X}{s_Y} r v \right) y} \frac{1}{\sqrt{2p} s_Y} e^{-\frac{1}{2s_Y^2} (y + s_Y^2 v)^2} dy \\
&= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} \int_{-\infty}^{\infty} e^{j \left(\frac{s_X}{s_Y} r v \right) y'} \frac{1}{\sqrt{2p} s_Y} e^{-\frac{1}{2s_Y^2} (y' + s_Y^2 v)^2} dy' \\
&= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} \int_{-\infty}^{\infty} e^{j \left(\frac{s_X}{s_Y} r v \right) y'} f_{Y'}(y') dy'
\end{aligned}$$

where $Y' \sim N(-s_Y^2 v, s_Y^2)$.

$$\begin{aligned}
Ee^{jvZ} &= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} f_Y \left(\frac{s_X}{s_Y} r v \right) = e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} e^{-j \left(\frac{s_X}{s_Y} r v \right) (-s_Y^2 v)} e^{-\frac{1}{2} s_Y^2 \left(\frac{s_X}{s_Y} r v \right)^2} \\
&= e^{-\frac{1}{2} v^2 s_X^2 (1-r^2)} e^{\frac{1}{2} v^2 s_Y^2} e^{-j s_X s_Y r v^2} e^{-\frac{1}{2} s_X^2 r^2 v^2} = e^{-\frac{1}{2} v^2 (s_X^2 - s_Y^2 - 2j s_X s_Y r)}
\end{aligned}$$

$$EZ^2 = E(X + jY)^2 = E(X^2 - Y^2 + j2XY) = EX^2 - EY^2 + 2jEXY$$

$$= s_X^2 - s_Y^2 - 2j s_X s_Y r$$

Therefore, $Ee^{jvZ} = e^{-\frac{1}{2} v^2 EZ^2}$ QED.

- Let W and Z be two complex random variable
 - $\text{Cov}(W, Z) = E(W - EW)(\overline{Z - EZ}) = \overline{\text{Cov}(Z, W)}$

- $Z = X + jY$
 - mean** $EZ = EX + jEY$
 - variance** $s_Z^2 = \text{Var}(Z) = \text{Cov}(Z, Z) = E|Z - EZ|^2 = s_X^2 + s_Y^2$ (real)

$$s_Z^2 = \text{Var}(Z) = \text{Cov}(Z, Z) = E(Z - EZ)(\overline{Z - EZ}) = E|Z - EZ|^2 \geq 0$$

$$= E[(X - EX)^2 + (Y - EY)^2] = \text{VAR}(X) + \text{VAR}(Y) = s_X^2 + s_Y^2$$

- $Z = X + jY$ and $z = x + jy$
 X, Y iid. $\mathcal{N}(0, \sigma^2)$

- $f_Z(z) = f_{X,Y}(x, y) = \frac{1}{2\pi s^2} e^{-\frac{|z|^2}{2s^2}} = \frac{1}{\pi s_Z^2} e^{-\frac{|z|^2}{s_Z^2}}$

$$f_{X,Y}(x, y) = \left(\frac{1}{\sqrt{2\pi} s} e^{-\frac{x^2}{2s^2}} \right) \left(\frac{1}{\sqrt{2\pi} s} e^{-\frac{y^2}{2s^2}} \right) = \frac{1}{2\pi s^2} e^{-\frac{x^2 + y^2}{2s^2}}$$

Let $z = x + jy$

Then $x^2 + y^2 = |z|^2$, so

$$f_{X,Y}(x, y) = \frac{1}{2\pi s^2} e^{-\frac{|z|^2}{2s^2}}$$

$f_Z(z)$ is a shorthand for $f_{X,Y}(x, y)$

$$s_Z^2 = s_X^2 + s_Y^2 = 2s^2$$

- Note: Though not as compact, any $f_{X,Y}(x, y)$ can be expressed only in terms of z by let

$$x = \frac{z + \bar{z}}{2} \text{ and } y = \frac{z - \bar{z}}{2j}.$$

- $\{Z(t)\} = \{X(t) + jY(t)\}$
 - $m_Z(t) = m_X(t) + jm_Y(t)$
 - $K_Z(s, t) = \text{Cov}(Z(s), Z(t)) = E(Z(s) - m_Z(s))(\overline{Z(t) - m_Z(t)}) = \overline{K_Z(t, s)}$ complex

$$K_Z(t, s) = E(Z(t) - m_Z(t))(\overline{Z(s) - m_Z(s)}) = \overline{E(Z(t) - m_Z(t))(\overline{Z(s) - m_Z(s)})}$$

$$= \overline{E(Z(s) - m_Z(s))(\overline{Z(t) - m_Z(t)})} = \overline{K_Z(s, t)}$$

- If $K_Z(t, t - \tau)$ = function of τ only, we write $K_Z(t, t - \tau) = K_Z(\tau)$

- $R_Z(s, t) = EZ(s) \overline{Z(t)} = \overline{R_Z(s, t)}$
- If $R_Z(t, t - \tau) = \text{function of } \tau \text{ only}$, we write $R_Z(t, t - \tau) = R_Z(\tau)$
- If $m_Z(t) = \text{constant}$ and $R_Z(t, t - \tau) = R_Z(\tau)$, $\{Z(t)\}$ is w.s.s.

Proper / circular / phase-invariant complex Gaussian

- $\underline{Z}$ is **proper (circular, phase-invariant) Gaussian** iff
pseudo-covariance $E(Z_i - EZ_i)(Z_j - EZ_j) = 0$ for all $1 \leq i, j \leq n$ including when $i = j$

- same as $\text{Cov}(X_i, X_j) = \text{Cov}(Y_i, Y_j)$ and $\text{Cov}(X_i, Y_j) = -\text{Cov}(Y_i, X_j)$

$$\begin{aligned}
 E(Z_i - EZ_i)(Z_j - EZ_j) &= E((X_i - EX_i) + j(Y_i - EY_i))((X_j - EX_j) + j(Y_j - EY_j)) \\
 &= E \left(\begin{aligned} &(X_i - EX_i)(X_j - EX_j) - (Y_i - EY_i)(Y_j - EY_j) \\ &+ j(Y_i - EY_i)(X_j - EX_j) + j(X_i - EX_i)(Y_j - EY_j) \end{aligned} \right) \\
 &= (E(X_i - EX_i)(X_j - EX_j) - E(Y_i - EY_i)(Y_j - EY_j)) \\
 &\quad + j(E(Y_i - EY_i)(X_j - EX_j) + E(X_i - EX_i)(Y_j - EY_j)) \\
 &= (\text{Cov}(X_i, X_j) - \text{Cov}(Y_i, Y_j)) + j(\text{Cov}(X_i, Y_j) + \text{Cov}(Y_i, X_j)) \\
 &= 0 + j0
 \end{aligned}$$

- For $E\underline{Z} = \underline{0}$ ($E\underline{X} = E\underline{Y} = \underline{0}$)

$$\begin{aligned}
 EZ_i Z_j &= E(X_i + jY_i)(X_j + jY_j) = E(X_i X_j - Y_i Y_j + jY_i X_j + jX_i Y_j) \\
 &= (EX_i X_j - EY_i Y_j) + j(EY_i X_j + EX_i Y_j) = 0 + j0
 \end{aligned}$$

Then, $EX_i X_j - EY_i Y_j = 0$ and $EY_i X_j + EX_i Y_j = 0$

- $\text{VAR}(X_i) = \text{VAR}(Y_i)$

$$\text{Cov}(X_i, X_j) = \text{Cov}(Y_i, Y_j) \rightarrow \text{Cov}(X_i, X_i) = \text{Cov}(Y_i, Y_i)$$

- $s_{Z_i}^2 = \text{Var}(Z_i) = E|Z_i|^2 = 2s_{X_i}^2 = 2s_{Y_i}^2$

$$s_{Z_i}^2 = \text{Var}(Z_i) = E|Z_i|^2 = s_{X_i}^2 + s_{Y_i}^2 = 2s_{X_i}^2 = 2s_{Y_i}^2$$

- $\text{Cov}(X_i, Y_i) = 0$

$$\text{Cov}(X_i, Y_j) = -\text{Cov}(X_j, Y_i) \rightarrow \text{Cov}(X_i, Y_i) = -\text{Cov}(X_i, Y_i)$$

- $\{Z(t) = X(t) + jY(t)\}$ is **proper (circular, phase-invariant) Gaussian** iff
all n-dimensional vector of the form $(Z(t_1), \dots, Z(t_n))$ must be proper Gaussian

- Condition for properness: $E(Z(s) - m_z(s))(Z(t) - m_z(t)) = 0$

- **n = 1:** Suppose

- $Z = X + jY$ is Gaussian
- $s_X^2 = s_Y^2$
- $\text{Cov}(X, Y) = 0$ (or $X \perp\!\!\!\perp Y$ which also make $\text{Cov}(X, Y) = 0$)

Then,
$$f_{X,Y}(x, y) = \frac{1}{ps_Z^2} e^{-\frac{|z-m|^2}{s_Z^2}} = f_Z(z)$$

$$s_X^2 = s_Y^2 = \sigma^2 \Rightarrow s_Z^2 = s_X^2 + s_Y^2 = 2s^2$$

$$\begin{aligned} f_{X,Y}(x, y) &= \left(\frac{1}{\sqrt{2ps}} e^{-\frac{(x-EX)^2}{2s^2}} \right) \left(\frac{1}{\sqrt{2ps}} e^{-\frac{(y-EY)^2}{2s^2}} \right) = \frac{1}{2ps^2} e^{-\frac{(x-EX)^2 + (y-EY)^2}{2s^2}} \\ &= \frac{1}{ps_Z^2} e^{-\frac{(x-EX)^2 + (y-EY)^2}{s_Z^2}} \end{aligned}$$

Consider the exponent

$$(x-EX)^2 + (y-EY)^2 = (x^2 - 2xEX + (EX)^2) + (y^2 - 2yEY + (EY)^2)$$

Let

- $m = m_z = EX + jEY \Rightarrow |m_z|^2 = (EX)^2 + (EY)^2$
- $z = x + jy \Rightarrow |z|^2 = x^2 + y^2$

$$\text{Since } \overline{mz} + z\overline{m} = \overline{mz} + \overline{mz} = 2\text{Re}\{\overline{mz}\} = 2\text{Re}\{(EX + jEY)(x - jy)\} = 2(xEX + yEY)$$

$$\begin{aligned} (x-EX)^2 + (y-EY)^2 &= (x^2 + y^2) + ((EX)^2 + (EY)^2) - (2xEX + 2yEY) \\ &= |z|^2 + |m|^2 - \overline{mz} - z\overline{m} \end{aligned}$$

$$\text{Since } |z-m|^2 = (z-m)(\overline{z-m}) = |z|^2 + |m|^2 - \overline{mz} - z\overline{m}$$

$$(x-EX)^2 + (y-EY)^2 = |z-m|^2$$

Hence,
$$f_{X,Y}(x, y) = \frac{1}{ps_Z^2} e^{-\frac{|z-m|^2}{s_Z^2}} = f_Z(z)$$

- Let

$$\underline{Z} = (Z_1, \dots, Z_n)^T = (X_1 + jY_1, \dots, X_n + jY_n)^T$$

$$z_k = x_k + jy_k \text{ for } 1 \leq k \leq n \text{ and } \underline{z} = (z_1, \dots, z_n)^T$$

If

1) $(X_1, \dots, X_n, Y_1, \dots, Y_n)$ is $2n$ -dimensional jointly Gaussian

2) $\underline{Z}$ is proper. $\Rightarrow E(Z_j - EZ_j)(Z_k - EZ_k) = 0$ for all $1 \leq j, k \leq n$ including when $j = k$

then

$$f_{\underline{X}, \underline{Y}}(\underline{x}, \underline{y}) = f_{\underline{Z}}(\underline{z}) = \frac{1}{p^n |\det(K_Z)|} e^{-(\underline{z} - \underline{m})^\dagger K_Z^{-1} (\underline{z} - \underline{m})}$$

where

$$\underline{m} = E\underline{Z} \text{ and}$$

$$K_Z = \left(\text{Cov}(Z_i, Z_j) \right)_{n \times n}$$

- To see that it reduces correctly when $n = 1$,

$$f_Z(z) = \frac{1}{p^1 s_Z^2} e^{-\frac{(z-m)^2}{s_Z^2}} = \frac{1}{p^1 s_Z^2} e^{-\frac{1}{s_Z^2} (z-m)(z-m)} = \frac{1}{p^1 s_Z^2} e^{-\frac{1}{s_Z^2} |z-m|^2}$$

Only have to consider one case where $j = k = 1$

Obviously, X, Y is jointly Gaussian

So, have to show that

$$E(Z_1 - EZ_1)(Z_1 - EZ_1) = E(Z_1 - EZ_1)^2 = E(Z - EZ)^2 = 0 \text{ is equivalent to}$$

$$s_X^2 = s_Y^2 \text{ and } \text{Cov}(X, Y) = 0$$

This can be shown from

$$\begin{aligned} E(Z - EZ)^2 &= E((X - EX) - j(Y - EY))^2 \\ &= E\left(\left((X - EX)^2 - (Y - EY)^2\right) - 2j(X - EX)(Y - EY)\right) \\ &= (s_X^2 - s_Y^2) - 2j(\text{Cov}(X, Y)) = 0 + j0 \end{aligned}$$

- For $n = 2$,

$(Z_1, Z_2)^T$ is proper Gaussian iff

0) X_1, X_2, Y_1, Y_2 are jointly Gaussian

1) Z_1 is proper Gaussian

2) Z_2 is proper Gaussian

3) $E(Z_1 - EZ_1)(Z_2 - EZ_2) = 0$

$$f_{X_1, Y_1, X_2, Y_2}(x_1, y_1, x_2, y_2) = \frac{1}{p^2 \det(\Lambda_Z)} e^{-(z)^\dagger \Lambda_Z^{-1} (z)}$$

where

$$EZ = 0$$

$$r = \frac{1}{s_1 s_2} EZ_1 \overline{Z_2}$$

$$\Lambda_Z = E \underline{Z} \underline{Z}^\dagger = \begin{pmatrix} EZ_1 \overline{Z_1} & EZ_1 \overline{Z_2} \\ EZ_2 \overline{Z_1} & EZ_2 \overline{Z_2} \end{pmatrix} = \begin{pmatrix} s_1^2 & EZ_1 \overline{Z_2} \\ EZ_2 \overline{Z_1} & s_2^2 \end{pmatrix} = \begin{pmatrix} s_1^2 & r s_1 s_2 \\ \overline{r} s_1 s_2 & s_2^2 \end{pmatrix}$$

$$\Lambda_Z^{-1} = \frac{1}{(1-|r|^2) s_1 s_2} \begin{pmatrix} \frac{s_2}{s_1} & -r \\ -\overline{r} & \frac{s_1}{s_2} \end{pmatrix}$$

$$\det(\Lambda_Z) = (1-|r|^2) s_1^2 s_2^2 \geq 0$$

$$\left((1-|r|^2) s_1 s_2 \right) \underline{z}^\dagger \Lambda_Z^{-1} \underline{z} = \frac{s_2}{s_1} |z_1|^2 - 2\text{Re}\{r \overline{z_1} z_2\} + \frac{s_1}{s_2} |z_2|^2$$

- To see that this is true for $n = 2$ and $E\underline{Z} = 0$

Bivariate Proper Gaussian Density

Define

- $Z_1 = X_1 + jY_1$, $Z_2 = X_2 + jY_2$ and $\underline{Z} = (Z_1, Z_2)^T$
- $E\underline{Z} = 0$, so $EX_1 = EX_2 = EY_1 = EY_2 = 0$
- $\underline{Z}$ is proper Gaussian, so $EZ_1 Z_2 = 0$

This gives

$$\text{Cov}(X_1, X_2) = \text{Cov}(Y_1, Y_2) \Rightarrow EX_1 X_2 = EY_1 Y_2$$

$$\text{Cov}(X_i, X_i) = \text{Cov}(Y_i, Y_i) \Rightarrow s_{X_i}^2 = s_{Y_i}^2$$

$$\text{Cov}(X_i, Y_j) = -\text{Cov}(Y_i, X_j) \Rightarrow EX_i Y_j = -EY_i X_j$$

$$\text{And } EX_i Y_i = -EY_i X_i \Rightarrow EX_i Y_i = 0$$

$$\bullet \quad s_1^2 = \text{Var}(Z_1) = E|Z_1|^2 = s_{X_1}^2 + s_{Y_1}^2 = 2s_{X_1}^2$$

$$s_2^2 = \text{Var}(Z_2) = E|Z_2|^2 = s_{X_2}^2 + s_{Y_2}^2 = 2s_{X_2}^2$$

$$\text{So, we have } s_{X_1}^2 = s_{Y_1}^2 = \frac{1}{2} s_1^2 \text{ and } s_{X_2}^2 = s_{Y_2}^2 = \frac{1}{2} s_2^2$$

$$\bullet \quad r = \frac{1}{s_1 s_2} EZ_1 \overline{Z_2}$$

$$\text{a) } s_1^2 s_2^2 = (2s_{X_1}^2)(2s_{X_2}^2) \Rightarrow s_1 s_2 = 2s_{X_1} s_{X_2}$$

$$\begin{aligned} EZ_1 \overline{Z_2} &= E(X_1 + jY_1)(X_2 - jY_2) \\ &= E(X_1 X_2 + Y_1 Y_2 + jY_1 X_2 - jX_1 Y_2) \\ &= E(X_1 X_2 + X_1 X_2 + jY_1 X_2 + jY_1 X_2) \\ &= 2EX_1 X_2 + 2jEY_1 X_2 \end{aligned}$$

$$r_{X_1, X_2} = \frac{\text{Cov}(X_1, X_2)}{s_{X_1} s_{X_2}} = \frac{E(X_1 - \cancel{EX_1})(X_2 - \cancel{EX_2})}{s_{X_1} s_{X_2}} = \frac{EX_1 X_2}{s_{X_1} s_{X_2}}.$$

$$\text{So, } EX_1 X_2 = s_{X_1} s_{X_2} r_{X_1, X_2}$$

$$r_{Y_1, X_2} = \frac{\text{Cov}(Y_1, X_2)}{s_{Y_1} s_{X_2}} = \frac{E(Y_1 - \cancel{EY_1})(X_2 - \cancel{EX_2})}{s_{Y_1} s_{X_2}} = \frac{EY_1 X_2}{s_{Y_1} s_{X_2}} = \frac{EY_1 X_2}{s_{X_1} s_{X_2}}.$$

$$\text{So, } EY_1 X_2 = s_{X_1} s_{X_2} r_{Y_1, X_2}$$

Thus,

$$\begin{aligned} r &= \frac{EZ_1 \bar{Z}_2}{s_1 s_2} = \frac{\cancel{EZ_1 X_2} + jEY_1 X_2}{\cancel{s_{X_1} s_{X_2}}} \\ &= \frac{s_{X_1} s_{X_2} r_{X_1, X_2} + j s_{X_1} s_{X_2} r_{Y_1, X_2}}{s_{X_1} s_{X_2}} \end{aligned}$$

$$\boxed{r = r_{X_1, X_2} + j r_{Y_1, X_2}}$$

$$\text{b) } \Lambda_Z = E \underline{Z} \underline{Z}^\dagger = E \begin{pmatrix} Z_1 \\ Z_2 \end{pmatrix} \begin{pmatrix} \bar{Z}_1 & \bar{Z}_2 \end{pmatrix} = \begin{pmatrix} EZ_1 \bar{Z}_1 & EZ_1 \bar{Z}_2 \\ EZ_2 \bar{Z}_1 & EZ_2 \bar{Z}_2 \end{pmatrix} = \begin{pmatrix} s_1^2 & EZ_1 \bar{Z}_2 \\ EZ_2 \bar{Z}_1 & s_2^2 \end{pmatrix}$$

$$r = \frac{EZ_1 \bar{Z}_2}{s_1 s_2} \Rightarrow EZ_1 \bar{Z}_2 = r s_1 s_2, \quad EZ_2 \bar{Z}_1 = \overline{EZ_1 \bar{Z}_2} = \bar{r} s_1 s_2$$

$$\text{So, } \Lambda_Z = \begin{pmatrix} s_1^2 & r s_1 s_2 \\ \bar{r} s_1 s_2 & s_2^2 \end{pmatrix}$$

$$\det(\Lambda_Z) = s_1^2 s_2^2 - r \bar{r} s_1^2 s_2^2$$

$$\boxed{\det(\Lambda_Z) = (1 - |r|^2) s_1^2 s_2^2}$$

$$\text{Note that for } A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, \quad A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Therefore,

$$\Lambda_Z^{-1} = \frac{1}{(1 - |r|^2) s_1^2 s_2^2} \begin{pmatrix} s_2^2 & -s_1 s_2 r \\ -s_1 s_2 \bar{r} & s_1^2 \end{pmatrix}$$

$$\boxed{\Lambda_Z^{-1} = \frac{1}{(1 - |r|^2) s_1 s_2} \begin{pmatrix} \frac{s_2}{s_1} & -r \\ -\bar{r} & \frac{s_1}{s_2} \end{pmatrix}}$$

$$\text{c) Define } \underline{R} = (X_1, Y_1, X_2, Y_2)^T$$

$$\begin{aligned}
\Lambda_R &= \underline{E} \underline{R} \underline{R}^T = E \begin{pmatrix} X_1 \\ Y_1 \\ X_2 \\ Y_2 \end{pmatrix} (X_1, Y_1, X_2, Y_2) = \begin{pmatrix} EX_1X_1 & EX_1Y_1 & EX_1X_2 & EX_1Y_2 \\ EY_1X_1 & EY_1Y_1 & EY_1X_2 & EY_1Y_2 \\ EX_2X_1 & EX_2Y_1 & EX_2X_2 & EX_2Y_2 \\ EY_2X_1 & EY_2Y_1 & EY_2X_2 & EY_2Y_2 \end{pmatrix} \\
&= \begin{pmatrix} s_{X_1}^2 & 0 & EX_1X_2 & -EY_1X_2 \\ 0 & s_{Y_1}^2 & EY_1X_2 & EX_1X_2 \\ EX_1X_2 & EY_1X_2 & s_{X_2}^2 & 0 \\ -EY_1X_2 & EX_1X_2 & 0 & s_{Y_2}^2 \end{pmatrix} \\
&= \begin{pmatrix} s_{X_1}^2 & 0 & s_{X_1}s_{X_2}r_{X_1,X_2} & -s_{X_1}s_{X_2}r_{Y_1,X_2} \\ 0 & s_{Y_1}^2 & s_{X_1}s_{X_2}r_{Y_1,X_2} & s_{X_1}s_{X_2}r_{X_1,X_2} \\ s_{X_1}s_{X_2}r_{X_1,X_2} & s_{X_1}s_{X_2}r_{Y_1,X_2} & s_{X_2}^2 & 0 \\ -s_{X_1}s_{X_2}r_{Y_1,X_2} & s_{X_1}s_{X_2}r_{X_1,X_2} & 0 & s_{Y_2}^2 \end{pmatrix} \\
\Lambda_R &= \frac{1}{2} s_1 s_2 \begin{pmatrix} s_1/s_2 & 0 & r_{X_1,X_2} & -r_{Y_1,X_2} \\ 0 & s_1/s_2 & r_{Y_1,X_2} & r_{X_1,X_2} \\ r_{X_1,X_2} & r_{Y_1,X_2} & s_2/s_1 & 0 \\ -r_{Y_1,X_2} & r_{X_1,X_2} & 0 & s_2/s_1 \end{pmatrix}
\end{aligned}$$

It can be shown that

$$\begin{pmatrix} a & 0 & c & -d \\ 0 & a & d & c \\ c & d & b & 0 \\ -d & c & 0 & b \end{pmatrix}^{-1} = \frac{1}{ab - (c^2 + d^2)} \begin{pmatrix} b & 0 & -c & d \\ 0 & b & -d & -c \\ -c & -d & a & 0 \\ d & -c & 0 & a \end{pmatrix}$$

$$\Lambda_R^{-1} = \frac{2}{(1 - |r|^2) s_1 s_2} \begin{pmatrix} s_2/s_1 & 0 & -r_{X_1,X_2} & r_{Y_1,X_2} \\ 0 & s_2/s_1 & -r_{Y_1,X_2} & -r_{X_1,X_2} \\ -r_{X_1,X_2} & -r_{Y_1,X_2} & s_1/s_2 & 0 \\ r_{Y_1,X_2} & -r_{X_1,X_2} & 0 & s_1/s_2 \end{pmatrix}$$

It can also be shown that

$$\begin{aligned}
\det \begin{pmatrix} a & 0 & c & -d \\ 0 & a & d & c \\ c & d & b & 0 \\ -d & c & 0 & b \end{pmatrix} &= a^2 b^2 - 2abd^2 - 2abc^2 + c^4 + 2c^2 d^2 + d^4 \\
&= (ab - (c^2 + d^2))^2
\end{aligned}$$

$$\text{Thus, } \det(\Lambda_R) = \left(\frac{1}{2} s_1 s_2 \right)^4 (1 - |r|^2)^2$$

$$\boxed{\det(\Lambda_R) = \frac{s_1^4 s_2^4}{16} (1 - |r|^2)^2}$$

$$d) \sqrt{\det(\Lambda_R)} = \frac{s_1^2 s_2^2}{4} (1 - |r|^2)$$

From part (b) we have $\det(\Lambda_Z) = (1 - |r|^2) s_1^2 s_2^2$

$$\text{Therefore, } \sqrt{\det(\Lambda_R)} = \frac{\det(\Lambda_Z)}{4}$$

e) We want to show that $f_{x_1, y_1, x_2, y_2}(x_1, y_1, x_2, y_2) = f_{z_1, z_2}(z_1, z_2)$ where

$$f_{z_1, z_2}(z_1, z_2) = \frac{1}{p^2 \det(\Lambda_Z)} e^{-(z)^\dagger \Lambda_Z^{-1} (z)}$$

$$f_{x_1, y_1, x_2, y_2}(x_1, y_1, x_2, y_2) = \frac{1}{(2p)^2 \sqrt{\det(\Lambda_R)}} e^{-\frac{1}{2} \underline{r}^T \Lambda_R^{-1} \underline{r}}$$

$$\underline{z} = (z_1, z_2)^T = (x_1 + jy_1, x_2 + jy_2)^T$$

$$\underline{r} = (x_1, y_1, x_2, y_2)^T$$

I. First, compare $p^2 \det(\Lambda_Z)$ and $(2p)^2 \sqrt{\det(\Lambda_R)}$

$$(2p)^2 \sqrt{\det(\Lambda_R)} = 4p^2 \sqrt{\det(\Lambda_R)} = 4p^2 \frac{\det(\Lambda_Z)}{4} = p^2 \det(\Lambda_Z)$$

II. Next, compare $\underline{z}^\dagger \Lambda_Z^{-1} \underline{z}$ and $\frac{1}{2} \underline{r}^T \Lambda_R^{-1} \underline{r}$

$$\left((1 - |r|^2) s_1 s_2 \right) \underline{z}^\dagger \Lambda_Z^{-1} \underline{z} = (\overline{z_1}, \overline{z_2}) \begin{pmatrix} \frac{s_2}{s_1} & -r \\ -\overline{r} & \frac{s_1}{s_2} \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \frac{s_2}{s_1} \overline{z_1} z_1 - r \overline{z_1} z_2 - \overline{r} z_2 z_1 + \frac{s_1}{s_2} \overline{z_2} z_2$$

$$= \frac{s_2}{s_1} |z_1|^2 - 2 \operatorname{Re}\{r \overline{z_1} z_2\} + \frac{s_1}{s_2} |z_2|^2$$

$$\begin{aligned} \operatorname{Re}\{r \overline{z_1} z_2\} &= \operatorname{Re}\{(r_{x_1, x_2} + j r_{y_1, x_2})(x_1 - jy_1)(x_2 + jy_2)\} \\ &= r_{x_1, x_2} x_1 x_2 + r_{x_1, x_2} y_1 y_2 - r_{y_1, x_2} x_1 y_2 + r_{y_1, x_2} y_1 x_2 \\ &= r_{x_1, x_2} (x_1 x_2 + y_1 y_2) + r_{y_1, x_2} (y_1 x_2 - r_{x_1, x_2}) \end{aligned}$$

$$\begin{aligned}
(1-|r|^2)s_1s_2\left(\frac{1}{2}\underline{r}^T\Lambda_R^{-1}\underline{r}\right) &= (x_1, y_1, x_2, y_2) \begin{pmatrix} s_2/s_1 & 0 & -r_{x_1, x_2} & r_{y_1, x_2} \\ 0 & s_2/s_1 & -r_{y_1, x_2} & -r_{x_1, x_2} \\ -r_{x_1, x_2} & -r_{y_1, x_2} & s_1/s_2 & 0 \\ r_{y_1, x_2} & -r_{x_1, x_2} & 0 & s_1/s_2 \end{pmatrix} \begin{pmatrix} x_1 \\ y_1 \\ x_2 \\ y_2 \end{pmatrix} \\
&= (x_1^2 + y_1^2) \frac{s_2}{s_1} + (x_2^2 + y_2^2) \frac{s_1}{s_2} \\
&\quad - 2((y_1x_2 + x_1y_2)r_{x_1, x_2} + (y_1x_2 - x_1y_2)r_{y_1, x_2}) \\
&= |z_1|^2 \frac{s_2}{s_1} + |z_2|^2 \frac{s_1}{s_2} \\
&\quad - 2((y_1x_2 + x_1y_2)r_{x_1, x_2} + (y_1x_2 - x_1y_2)r_{y_1, x_2}) \\
&= |z_1|^2 \frac{s_2}{s_1} + |z_2|^2 \frac{s_1}{s_2} - 2\text{Re}\{r\bar{z}_1z_2\}
\end{aligned}$$

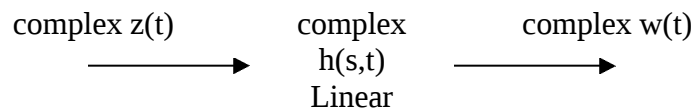
Therefore, $\left((1-|r|^2)s_1s_2\right)\underline{z}^\dagger\Lambda_Z^{-1}\underline{z} = (1-|r|^2)s_1s_2\left(\frac{1}{2}\underline{r}^T\Lambda_R^{-1}\underline{r}\right).$

And $\underline{z}^\dagger\Lambda_Z^{-1}\underline{z} = \frac{1}{2}\underline{r}^T\Lambda_R^{-1}\underline{r}$

From I and II, we have $\frac{1}{p^2\det(\Lambda_Z)}e^{-(z)^\dagger\Lambda_Z^{-1}(z)} = \frac{1}{(2p)^2\sqrt{\det(\Lambda_R)}}e^{-\frac{1}{2}\underline{r}^T\Lambda_R^{-1}\underline{r}}.$

Hence, $\boxed{f_{x_1, y_1, x_2, y_2}(x_1, y_1, x_2, y_2) = \frac{1}{p^2\det(\Lambda_Z)}e^{-(z)^\dagger\Lambda_Z^{-1}(z)}} \text{ QED}$

- Proper Gaussian is preserved under linear filter:



$$w(t) = \int h(s, t) z(s) ds$$

If $\{Z(t)\}$ is proper Gaussian, so is $\{w(t)\}$

- If $\{Z(t)\}$ is 0-mean proper Gaussian, so is $\{w(t)\}$

$$\begin{aligned}
EW(u)W(v) &= E \int h(s, u) z(s) ds \int h(s', v) z(s') ds' \\
&= E \iint h(s, u) h(s', v) z(s) z(s') ds ds' \\
&= \iint h(s, u) h(s', v) \underbrace{(E(z(s)z(s')))}_0 ds ds' \\
&= 0
\end{aligned}$$

- For non-zero mean case, we have

$$E(z(s)z(s')) - Ez(s)Ez(s') = 0$$

$$EW(u) = \int h(s, u) Ez(s) ds$$

$$EW(v) = \int h(s', v) Ez(s') ds'$$

$$EW(u)W(v) = \iint h(s, u)h(s', v)(E(z(s)z(s')))) ds ds' \text{ as before.}$$

$$\begin{aligned} EW(u)EW(v) &= \int h(s, u)Ez(s)ds \int h(s', v)Ez(s')ds' \\ &= \iint h(s, u)h(s', v)(Ez(s)Ez(s')) ds ds' \end{aligned}$$

$$\begin{aligned} &EW(u)W(v) - EW(u)EW(v) \\ &= \iint h(s, u)h(s', v) \underbrace{(E(z(s)z(s')) - Ez(s)Ez(s'))}_0 ds ds' \\ &= 0 \end{aligned}$$