

Contrast Functions for Non-circular and Circular Sources Separation in Complex-Valued ICA

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Abstract—In this paper, the complex-valued ICA problem is studied in the context of blind complex-source separation. We formulate the complex ICA problem in a general setting, and define the superadditive functional that may be used for constructing a contrast function for circular complex sources separation. We propose several contrast functions and study their properties. Finally, we also discuss relevant issues and present the convex analysis of a specific contrast function.

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$$\begin{aligned}
& \bullet \quad \text{Linearity of expectation:} \\
& \mathbb{E}[\mathbf{X}] = \mathbb{E}[\mathbf{X}_1 + \mathbf{X}_2] = \mathbb{E}[\mathbf{X}_1] + \mathbb{E}[\mathbf{X}_2]. \\
& \bullet \quad \text{Variance of a sum:} \\
& \mathbb{E}[\mathbf{X}^2] = \mathbb{E}[\mathbf{X}_1^2] + \mathbb{E}[\mathbf{X}_2^2] + 2\mathbb{E}[\mathbf{X}_1\mathbf{X}_2]. \\
& \bullet \quad \text{Covariance of a sum:} \\
& \text{var}[\mathbf{X}] = \mathbb{E}[(\mathbf{X} - \mathbb{E}[\mathbf{X}])^2] = \mathbb{E}[\|\mathbf{X} - \mathbb{E}[\mathbf{X}]\|^2]. \\
& \bullet \quad \text{Covariance matrix:} \\
& C_{ij} = \mathbb{E}[(\mathbf{x}_i - \mathbb{E}[\mathbf{x}_i])(\mathbf{x}_j^* - \mathbb{E}[\mathbf{x}_j^*])] = \mathbb{E}[\mathbf{x}_i \mathbf{x}_j^*] - \mathbb{E}[\mathbf{x}_i] \mathbb{E}[\mathbf{x}_j^*]. \\
& \text{uncorrelated} \quad C_{ij} = 0. \\
& \bullet \quad \text{Covariance matrix of a vector:} \\
& \mathbf{z} = [z_1, \dots, z_n]^T, \quad \mathbf{z}^H = [z_1^*, \dots, z_n^*]^T \equiv (\mathbf{z}^*)^T \\
& \text{Cov}[\mathbf{z}] \equiv \mathbb{E}[(\mathbf{z} - \mathbb{E}[\mathbf{z}])(\mathbf{z} - \mathbb{E}[\mathbf{z}])^H]; \quad \text{pseudo-covariance matrix} \\
& \text{pcov}[\mathbf{z}] \equiv \mathbb{E}[(\mathbf{z} - \mathbb{E}[\mathbf{z}])(\mathbf{z} - \mathbb{E}[\mathbf{z}])^T]. \\
& \text{Definition 1: A random vector } \mathbf{z} \text{ is } \alpha\text{-circular if} \\
& \mathbb{E}[\mathbf{z} e^{j\alpha}] = e^{j\alpha} \mathbb{E}[\mathbf{z}]. \\
& \text{order circular} \quad \mathbf{z} \text{ is second-order circular if} \\
& \mathbb{E}[\mathbf{z}] = 0, \quad \mathbb{E}[\mathbf{z}^2] = 0. \\
& \bullet \quad \text{Covariance matrix of a circular vector:} \\
& \mathbb{E}[\mathbf{z}\mathbf{z}^H] \text{ is symmetric; } \mathbb{E}[\mathbf{z}\mathbf{z}^T] = 0. \\
& \bullet \quad \text{Covariance matrix of a zero-mean vector:} \\
& \mathbb{E}[\mathbf{z}\mathbf{z}^T] = 0.
\end{aligned}$$

$$\text{skewness}(\cdot) = \mathbb{E}[|\cdot|^3]/(\mathbb{E}[|\cdot|^2])^{3/2}, \quad (1)$$

$$\text{kurtosis}(\cdot) = \mathbb{E}[\|\cdot\|^4] - 2(\mathbb{E}[\|\cdot\|^2])^2 - |\mathbb{E}[\cdot]|^2. (2)$$

Definition 2: Let $\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$, $\mathbf{B} = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$. Then

$${}^1_1({}^2_1|{}^1_1|, {}^2_2|{}^1_2|) = ({}^1_1|{}^1_1|)({}^2_2|{}^1_2|).$$

$$J(\mathbf{z}) = \frac{1}{2} \sum_{i=1}^N \|\mathbf{z} - \mathbf{y}_i\|^2, \quad \mathbf{z} \in \mathbb{C}^N,$$

$$\nabla J \equiv \frac{\partial J(\mathbf{z})}{\partial \mathbf{z}^*} = \frac{1}{2} \left(\frac{\partial J(\mathbf{z})}{\partial \mathbf{z}} + \frac{\partial J(\mathbf{z})}{\partial \bar{\mathbf{z}}} \right) = 0.$$

$$, \frac{\partial J(\mathbf{z})}{\partial \mathbf{x}} = \frac{\partial J(\mathbf{z})}{\partial \mathbf{z}} = 0.$$

Definition 3: A function $J(\mathbf{z})$ ($\mathbf{z} \in \mathbb{C}^N$) is called a *quasi-convex* function if

$$J(\lambda \mathbf{z}_1 + (1 - \lambda) \mathbf{z}_2) \leq \lambda J(\mathbf{z}_1) + (1 - \lambda) J(\mathbf{z}_2)$$

$$\mathbf{z}_1, \mathbf{z}_2 \in \mathbb{C}^N \quad 0 \leq \lambda \leq 1.$$

$$\mathbf{H} = \frac{\partial J(\mathbf{z})}{\partial \mathbf{z} \partial \mathbf{z}^H},$$

$$J(\mathbf{z})$$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix}, \quad x = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \quad y = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

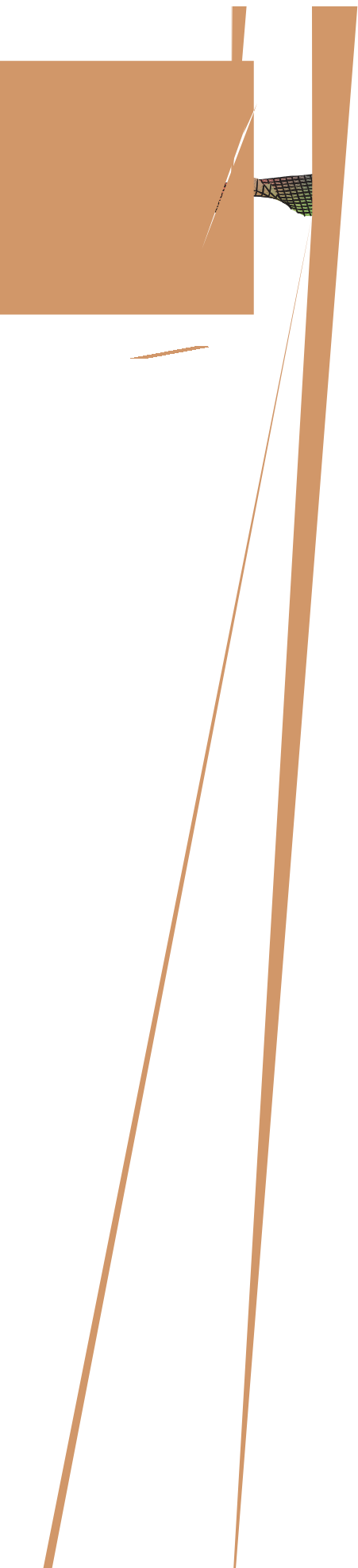
$$\mathbf{s} \in \mathbb{C}^n, \quad \mathbf{x} \in \mathbb{C}^n, \quad \mathbf{A} \in \mathbb{C}^{n \times n} \quad (\text{..})$$

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$$Q(\mathbf{1} + \mathbf{2}) = Q(\mathbf{1}) + Q(\mathbf{2}) \quad (19)$$

$$Q(i) > 0$$

$$\begin{aligned} Q^2(\gamma_1 + \gamma_2) &= Q^2(\gamma_1) + Q^2(\gamma_2) + 2Q(\gamma_1)Q(\gamma_2) \\ &\geq Q^2(\gamma_1) + Q^2(\gamma_2), \end{aligned}$$

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$$(\theta_1, \theta_2 \in \mathbb{R})$$

$$\theta = \arctan \frac{\frac{1}{\theta_1} - \frac{1}{\theta_2}}{\frac{\theta_1 + 2}{\theta_1 - \theta_2} - \frac{\theta_2}{\theta_1 - \theta_2}}. \quad , \quad \left| \left| \frac{1}{\theta_1} - \frac{1}{\theta_2} \right| \right| \leq$$

Theorem 3: Q (2-)

$$\overline{Q}(\cdot) = Q(|\cdot|) = \sqrt{Q^2(|\cdot| \cos \theta) + Q^2(|\cdot| \sin \theta)}; \quad = |e^{j\theta} \quad (\theta \in \mathbb{R}),$$

$$2 = |2|e^{j\theta_2}$$

$$Q_1 + Q_2 = Q_1 + Q_2; \quad \square$$

$$Q^2(|\mathbf{r}| \sin \theta) \geq Q^2(|\mathbf{r}_1| \sin \theta_1) + Q^2(|\mathbf{r}_2| \sin \theta_2).$$

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$$\begin{aligned} \tilde{\mathbf{z}} &\in \mathbb{R}^2; & (1) \\ Q(\alpha + \tilde{\mathbf{z}}) &= Q(\tilde{\mathbf{z}}) \ (\forall \alpha \in \mathbb{R}^2); & (2) \\ Q(\alpha \tilde{\mathbf{z}}) &= |\alpha|Q(\tilde{\mathbf{z}}) \ (\forall \alpha \in \mathbb{R}). \end{aligned}$$

$|\alpha|e^{j0}$ (i.e., $\alpha = 1$). A 15, $\alpha = 1$.

$$-\sum_{i=1}^n Q^k(i) - \sum_{i=1}^n Q^{2k}(i) \geq 2, \\ \vdots \\ -\sum_{i=1}^n Q^k(i) - \sum_{i=1}^n Q^{2k}(i) > 2, \quad \{j\}, \\ \vdots \\ -\sum_{i=1}^n Q^k(i) - \sum_{i=1}^n Q^{2k}(i) > 2, \quad Q(j) = 0.$$
$$\begin{aligned} \sum_{j=1}^n Q^2(j) &\leq \sum_{j=1}^n |ij|^2 Q^2(j) \leq \sum_{j=1}^n |ij| \leq \sum_{k=1}^n |ik| = 1; \\ \sum_{j=1}^n Q^k(i) &\leq \sum_{j=1}^n |ij|^k Q^k(j) \leq \sum_{j=1}^n Q^k(j), \\ -\sum_{j=1}^n Q^k(i) &\geq -\sum_{j=1}^n Q^k(j), \end{aligned}$$
$$\begin{aligned} \mathbf{A} \mathbb{E}[\mathbf{s}\mathbf{s}^H] \mathbf{A}^H &= \mathbf{U} \Sigma \mathbf{U}^H, & \mathbf{z} &= \mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{x} = \mathbf{A} \mathbf{s}, \\ \mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{A} \mathbf{s} &\equiv \tilde{\mathbf{A}} \mathbf{s}, & \mathbb{E}[\mathbf{z}\mathbf{z}^H] &= \tilde{\mathbf{A}} \mathbb{E}[\mathbf{s}\mathbf{s}^H] \tilde{\mathbf{A}}^H = \mathbf{I}; \\ & & \mathbb{E}[\mathbf{s}\mathbf{s}^H] &= \mathbf{I}. \end{aligned}$$

1) Range Function:

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$$R(\cdot) = d, \quad \{(\cdot) \geq 0 \mid |\cdot - 0| \leq d\} \quad (21)$$

Theorem 5: $d \in \mathbb{R}^+$

$$R(\cdot_1 + \cdot_2) = R(\cdot_1) + R(\cdot_2). \quad (22)$$

Proof:

$$\text{Corollary 2: } \alpha, \beta \in \mathbb{C},$$

$$R(\cdot) = |\alpha| \cdot R(\cdot_1) + |\beta| \cdot R(\cdot_2). \quad (23)$$

$$Q(\cdot_i) = R(\cdot_i), \quad (17)$$

$$J(\mathbf{W}) = \sum_{i=1}^n \log \left(\sum_{j=1}^n |_{ij}| R(\cdot_j) \right) - \log |(\mathbf{W})|. \quad (24)$$

Definition 6:

$$\overline{Q}(\cdot_i) = Q(|\cdot_i|) = \exp(H(|\cdot_i|)), \quad (25)$$

$$\text{entropy power inequality,} \quad (25)$$

$$\{|\cdot_i|\}, \quad (25) \quad (17) \quad (6)$$

3) Rényi Entropy Function:

$$H_k(\cdot_i) = \frac{1}{1-k} \log \left(\int (\cdot_i)^k d\cdot_i \right). \quad (26)$$

2 extension entropy

$$H_2(\cdot_i) = -\log \left(\int (\cdot_i)^2 d\cdot_i \right).$$

Jensen inequality, $H_2(\cdot_i) \leq H(\cdot_i).$

$$H_k(\cdot_i) \geq H_r(\cdot_i) >$$

Lemma 2: $\{ \cdot : (\cdot) =$

$$\exp(H_2(\cdot)), \quad H_2(\cdot) = H_2(\cdot + \alpha)$$

$$H_2(\alpha) = -\log \left(\int g(|\alpha|)^2 d|\alpha| \right)$$

$$\exp(H_2(\alpha)) = |\alpha| \exp(H_2(\cdot)).$$

$$\exp(H_2(|\cdot_i|)) = \frac{1}{\int p(|y_i|)^2 dy_i},$$

$$\overline{Q}(\cdot_i) = Q(|\cdot_i|) =$$

4) Fisher Information Function:

$$\mathbf{G} = \mathbb{E}[\psi(\cdot)^2] = \mathbb{E} \left[\left(\frac{d \log(\cdot)}{d} \right)^2 \right], \quad (27)$$

$$\psi(\cdot) = \frac{d p(\cdot)}{d} = \frac{p'(\cdot)}{p(\cdot)}$$

$$Q(\cdot) = \mathbf{G}^{-1/2},$$

$$\overline{Q}(\cdot) = \mathbb{E} \left[\left(\frac{p'(|\cdot|)}{p(|\cdot|)} \right)^2 \right]^{-1/2}$$

$$\psi_k(\cdot) =$$

$$Q(\mathbf{\Sigma}) = \left(\mathbf{\Sigma}\right)^{1/2} \equiv \left(\text{cov}[\tilde{\mathbf{z}}]\right)^{1/2},$$

$$\begin{aligned} Q(\omega) &= \left(\text{cov}[\tilde{\mathbf{z}}']\right)^{1/2} = \left(\text{cov}[\Omega\tilde{\mathbf{z}}]\right)^{1/2} \\ &= \left(\Omega\text{cov}[\tilde{\mathbf{z}}]\right)^{1/2} = (\Omega)^{1/2} \left(\text{cov}[\tilde{\mathbf{z}}]\right)^{1/2} \\ &= (\omega^2 + \omega^2)^{1/2} (\mathbf{\Sigma})^{1/2} = |\omega| (\mathbf{\Sigma})^{1/2}. \end{aligned}$$

$$Q(\mathbf{\Sigma}) = \left(\mathbf{\Sigma}\right)^{1/2} = \left(\text{cov}[\tilde{\mathbf{z}}]\right)^{1/2}; \quad Q(\omega) = \left(\omega^2 + \omega^2\right)^{1/2} (\mathbf{\Sigma})^{1/2} = |\omega| (\mathbf{\Sigma})^{1/2}.$$

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$$A = \begin{pmatrix} 1 & 2 & 5 \\ 2 & 1 & 5 \\ 5 & 5 & 1 \end{pmatrix} \quad (1)$$