

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

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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- $\mathbb{E}[a + b] = \mathbb{E}[a] + \mathbb{E}[b]$.
- $\mathbb{E}[a^2] = \mathbb{E}[a^2] - \mathbb{E}[a]^2 + 2\mathbb{E}[a]\mathbb{E}[a]$.
- $\text{var}[a] = \mathbb{E}[|a - \mathbb{E}[a]|^2] = \mathbb{E}[|a|^2] - |\mathbb{E}[a]|^2$.

$$C_{ij} = \mathbb{E}[(a_i - \mathbb{E}[a_i])(a_j^* - \mathbb{E}[a_j^*])] = \mathbb{E}[a_i a_j^*] - \mathbb{E}[a_i]\mathbb{E}[a_j^*].$$

uncorrelated $C_{ij} = 0$.

$$\mathbf{z} = [z_1, \dots, z_n]^T, \quad \mathbf{z}^H = [z_1^*, \dots, z_n^*]^T \equiv (\mathbf{z}^*)^T$$

(\dots),

$$\mathbb{E}[\mathbf{z}], \quad \text{cov}[\mathbf{z}] \equiv \mathbb{E}[(\mathbf{z} - \mathbb{E}[\mathbf{z}])(\mathbf{z} - \mathbb{E}[\mathbf{z}])^H];$$

pseudo-covariance matrix

$$\text{pcov}[\mathbf{z}] \equiv \mathbb{E}[(\mathbf{z} - \mathbb{E}[\mathbf{z}])(\mathbf{z} - \mathbb{E}[\mathbf{z}])^T].$$

Definition 1: \mathbf{A} is *second-order circular* if $\mathbb{E}[z_i] = 0$, $\mathbb{E}[z_i^2] = 0$, and $\mathbb{E}[z_i z_j^*] = \mathbb{E}[z_j z_i^*]$.

\mathbf{z} is *strongly uncorrelated* if $\mathbb{E}[\mathbf{z}\mathbf{z}^H] = \mathbf{I}$.

\mathbf{z} is *symmetric* if $\mathbb{E}[\mathbf{z}\mathbf{z}^T] = \mathbf{0}$.

\mathbf{z} is *zero-mean* if $\mathbb{E}[\mathbf{z}] = \mathbf{0}$.

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

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

Definition 2: $(\mathbf{z}_1, \mathbf{z}_2)$ is *independent* if $\mathbb{E}[z_1 z_2^*] = \mathbb{E}[z_1] \mathbb{E}[z_2^*]$.

$J(\mathbf{z})$ is *real-valued* if $J(\mathbf{z}) = J(\mathbf{z})^*$.

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

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

Definition 3: \mathbf{A} is *Hermitian* if $\mathbf{A} = \mathbf{A}^H$.

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

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

\mathbf{A} is *positive semi-definite* if $\mathbf{H} = \frac{\partial^2 J(\mathbf{z})}{\partial \mathbf{z} \partial \mathbf{z}^H}$ is *positive semi-definite*.

$$\mathbf{x} = \mathbf{A}\mathbf{s}, \quad \mathbf{s} \in \mathbb{C}^n, \quad \mathbf{x} \in \mathbb{C}^n$$

$\mathbf{A} \in \mathbb{C}^{n \times n}$

- \mathbf{A} is *invertible* if $\det(\mathbf{A}) \neq 0$.
- \mathbf{A} is *orthogonal* if $\mathbf{A}^{-1} = \mathbf{A}^H$.
- \mathbf{A} is *unitary* if $\mathbf{A}^{-1} = \mathbf{A}^H$.

\mathbf{A} is *symmetric* if $\mathbf{A} = \mathbf{A}^T$.

\mathbf{A} is *skew-symmetric* if $\mathbf{A} = -\mathbf{A}^T$.

\mathbf{A} is *Hermitian* if $\mathbf{A} = \mathbf{A}^H$.

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$$\mathbb{E}[\mathbf{x}] = 0 \quad \mathbb{E}[\mathbf{x}\mathbf{x}^H] = \mathbf{I},$$

$$\Delta \mathbf{W} = \eta(\mathbf{I} - \psi(\mathbf{y})\mathbf{y}^H)\mathbf{W}. \quad (9)$$

strong-uncorrelating transform 10, 11 .

Liouville's theorem,
boundedness analyticity

formation, mutual in-

9, 13, 14); $\psi(\cdot)$ (. . . , 7 -

$$I(1, \dots, n) = \mathbb{E}_{p(\mathbf{y})} \left[\log \frac{p(\mathbf{y})}{\prod_{i=1}^n p(i)} \right] \\ = \int p(\mathbf{y}) \log \frac{p(\mathbf{y})}{\prod_{i=1}^n p(i)} d\mathbf{y} \quad (4)$$

$$e_i = \mathbf{e}_i - \mathbf{e}_i, \\ H(i | i)$$

$$H(i | i) = H(i - i) \equiv H(e_i) \quad (10)$$

$$(4) \quad \{1, \dots, n\} \\ \{1, \dots, n\}$$

$$H(i) \equiv H(i, i)$$

$$I(1, \dots, n) = -\frac{1}{2} \log \left(\frac{|\mathbf{C}_y|}{\prod_{i=1}^n C_{ii}} \right) = -\frac{1}{2} \sum_{i=1}^n \log(\lambda_i) \quad (5)$$

$$H(i) = H(i | i) + H(i) \equiv H(-e_i) + H(i) \\ = H(i | i) + H(i) \equiv H(e_i) + H(i) \quad (11)$$

$$C_{ij} = \mathbb{E}[(i - \mathbb{E}[i])(j^* - \mathbb{E}[j^*])], \quad \lambda_i \\ \mathbf{C}_y \mathbf{u} = \lambda \mathbf{\Lambda} \mathbf{u}, \\ \mathbf{C}_y = \text{cov}[\mathbf{y}], \quad \mathbf{\Lambda} = \nabla \{C_{11}, C_{22}, \dots, C_{nn}\}.$$

$$e_i \quad ; \quad i \quad i$$

$$I(1, \dots, n) = \sum_{i=1}^n H(i) - H(\mathbf{y}),$$

$$\log |\mathbf{W}|, \quad \{1, \dots, n\}; \quad H(\mathbf{y}) = H(\mathbf{x}) + \quad (4)$$

$$J(\mathbf{W}) = \sum_{i=1}^n H(i) - \log |\mathbf{W}| - H(\mathbf{x}), \quad (6)$$

$$H(\mathbf{x}) \quad (\dots) \mathbf{W}. \quad (6) \dots$$

\mathbf{W}^*

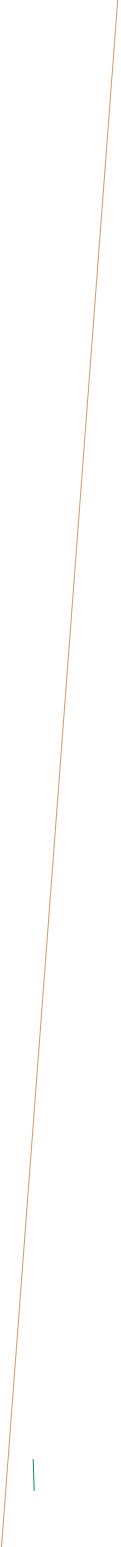
$$\nabla_{\mathbf{W}^*} J(\mathbf{W}) = (\mathbb{E}_y[\psi(\mathbf{y})\mathbf{y}^H] - \mathbf{I})\mathbf{W}^{-H}, \quad (7)$$

$$\mathbf{W}^{-H} \quad \mathbf{W}^{-1}, \\ \psi(\mathbf{y}) = [\psi(1), \dots, \psi(n)]^T$$

$$\psi(i) = -\frac{d \log p(i)}{d i^*} = -\frac{\frac{\partial p(y_i)}{\partial y_i} + \frac{\partial p(y_i)}{\partial y_i}}{p(i)} \\ = \frac{\partial \log p(i)}{\partial i} + \frac{\partial \log p(i)}{\partial i^*}, \quad (8)$$

$$\psi(\cdot) \quad \text{complex score function.} \\ (i) \equiv (i, i) \quad (8)$$

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$Q(\alpha) > 0$ for all $\alpha \in \mathbb{R}$.
 Corollary 1: $Q(\alpha_1 + \alpha_2) = Q(\alpha_1) + Q(\alpha_2)$ (19)

$$Q(\alpha_1 + \alpha_2) = Q(\alpha_1) + Q(\alpha_2) \quad (17)$$

Proof:

$$Q^2(\alpha_1 + \alpha_2) = Q^2(\alpha_1) + Q^2(\alpha_2) + 2Q(\alpha_1)Q(\alpha_2) \geq Q^2(\alpha_1) + Q^2(\alpha_2)$$

$$Q(\alpha_1) = Q(\alpha_2) = 0.$$

2. \square sufficient

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

$$\alpha_1 + \alpha_2 = | \alpha_1 | e^{j\theta_1} + | \alpha_2 | e^{j\theta_2} = \sqrt{| \alpha_1 |^2 + | \alpha_2 |^2 + 2| \alpha_1 | | \alpha_2 | \cos(\theta_1 - \theta_2)} e^{j\theta}$$

$$\theta = \arctan \frac{| \alpha_1 | \sin \theta_1 + | \alpha_2 | \sin \theta_2}{| \alpha_1 | \cos \theta_1 + | \alpha_2 | \cos \theta_2}$$

$$| \alpha_1 + \alpha_2 | \leq | \alpha_1 | + | \alpha_2 |$$

Theorem 3: $Q(\alpha) = \sqrt{Q^2(| \alpha | \cos \theta) + Q^2(| \alpha | \sin \theta)}$ (20)

$$\bar{Q}(\alpha) = Q(| \alpha |) = \sqrt{Q^2(| \alpha | \cos \theta) + Q^2(| \alpha | \sin \theta)}$$

Proof:

$$Q(\alpha_1 + \alpha_2) = \sqrt{Q^2(| \alpha_1 + \alpha_2 | \cos \theta) + Q^2(| \alpha_1 + \alpha_2 | \sin \theta)}$$

$Q(\alpha_1 + \alpha_2)$

$$Q^2(| \alpha | \cos \theta) \geq Q^2(| \alpha_1 | \cos \theta_1) + Q^2(| \alpha_2 | \cos \theta_2)$$

$$Q^2(| \alpha | \sin \theta) \geq Q^2(| \alpha_1 | \sin \theta_1) + Q^2(| \alpha_2 | \sin \theta_2)$$

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$$\bar{Q}^2(\alpha) = Q^2(| \alpha |) = Q^2(| \alpha | \cos \theta) + Q^2(| \alpha | \sin \theta) \geq Q^2(| \alpha_1 |) + Q^2(| \alpha_2 |) = \bar{Q}^2(\alpha_1) + \bar{Q}^2(\alpha_2), \quad (20)$$

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D. \square

Lemma 1: $Q(\alpha) (\alpha \in \mathbb{C})$

$$\tilde{\mathbf{z}} = [\alpha_1, \alpha_2]^T, \quad Q(\alpha)$$

$\tilde{\mathbf{z}} \in \mathbb{R}^2$;

$$Q(\alpha + \tilde{\mathbf{z}}) = Q(\tilde{\mathbf{z}}) (\forall \alpha \in \mathbb{R}^2);$$

$$Q(\alpha \tilde{\mathbf{z}}) = | \alpha | Q(\tilde{\mathbf{z}}) (\forall \alpha \in \mathbb{R}).$$

Proof:

$$| \alpha | e^{j\theta} (\dots)$$

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Theorem 4:

$$- \sum_{i=1}^n Q^k(\alpha_i) - \sum_{i=1}^n Q^{2k}(\alpha_i)$$

$$\geq 2,$$

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{ j },

$$Q(\alpha_j) = 0.$$

Proof:

$$\sum_{j=1}^n Q^2(\alpha_j) \leq \sum_{j=1}^n | \alpha_j |^2 Q^2(\alpha_j) \leq \sum_{j=1}^n | \alpha_j | \leq \sum_{k=1}^n | \alpha_k |^2 = 1; \quad Q^k(\alpha_i) \leq \sum_{j=1}^n | \alpha_j |^k Q^k(\alpha_j) \leq \sum_{j=1}^n Q^k(\alpha_j).$$

$$- \sum_{i=1}^n Q^k(\alpha_i) \geq - \sum_{j=1}^n Q^k(\alpha_j),$$

Remark:

$$\mathbf{A} \mathbb{E}[\mathbf{ss}^H] \mathbf{A}^H = \mathbf{U} \Sigma \mathbf{U}^H, \quad \Sigma = \mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{x} \mathbf{x}^H \mathbf{U} = \mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{x} \mathbf{x}^H \mathbf{U}$$

$$\mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{A} \mathbf{s} \equiv \tilde{\mathbf{A}} \mathbf{s}, \quad \tilde{\mathbf{A}} = \mathbf{D}^{-1/2} \mathbf{U}^H \mathbf{A}$$

$$\mathbb{E}[\mathbf{z} \mathbf{z}^H] = \tilde{\mathbf{A}} \mathbb{E}[\mathbf{ss}^H] \tilde{\mathbf{A}}^H = \mathbf{I}; \quad \mathbb{E}[\mathbf{ss}^H] = \mathbf{I}.$$

B. Examples of Contrast Functions

1) Range Function:

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

Jensen inequality, $H_2(|i|) \leq H(|i|)$.

$$H_k(|i|) \geq H_r(|i|) > \dots \{ : () =$$

Lemma 2: $g(|i|)$, $\exp(H_2(|i|))$,

Proof: $H_2(\alpha) = H_2(\alpha)$

$$\begin{aligned} H_2(\alpha) &= -\log \left(\int g(|\alpha|)^2 d|\alpha| \right) \\ &= -\log \left(\int g(|i|)^2 d|i| \right) + \log |\alpha|. \end{aligned}$$

$$\exp(H_2(\alpha)) = |\alpha| \exp(H_2(|i|)). \quad \square$$

$$\exp(H_2(|i|)) = \frac{1}{\int p(|y_i|)^2 dy_i}, \quad \bar{Q}(|i|) = Q(|i|) =$$

$$26,27; \quad \bar{Q} \{ |i| \},$$

18.

4) Fisher Information Function:

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

$$\begin{aligned} \psi(|i|) &= \frac{d \log p(|i|)}{d} = \frac{p'(|i|)}{p(|i|)} \\ Q(|i|) &= \mathbf{G}^{-1/2}, \quad Q \end{aligned}$$

15.

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

$$\psi_k(|i|) = (|i|)^T \dots j/T \quad T /$$

$$R(|i|),$$

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

$$d \in \mathbb{R}^+$$

Theorem 5:

$$1 \quad 2$$

$$R(|i_1 + i_2|) = R(|i_1|) + R(|i_2|). \quad (22)$$

Proof:

$$|i_1 + i_2| \leq \square$$

Corollary 2: $= \alpha |i_1| + \beta |i_2|$

$\alpha, \beta \in \mathbb{C}$,

$$R(|i|) = |\alpha| \cdot R(|i_1|) + |\beta| \cdot R(|i_2|). \quad (23)$$

$$|\beta| \alpha |i_1| (|\beta| |i_2|) \quad |\alpha| (|i_1|)$$

$$Q(|i|) = R(|i|), \quad (17)$$

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

Definition 6:

$$\int_0^d (|i|)^d d|i| \approx 1 \quad (0 < d < +\infty).$$

2) Shannon Entropy Function: A

$$H(|i|).$$

$$\bar{Q}(|i|) = Q(|i|) = \exp(H(|i|)), \quad (25)$$

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

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$$(25) \quad (17) \quad (6)$$

$$\{ |i| \}, \quad \{ |i| \}$$

16, 17.

3) Rényi Entropy Function:

($0 < \in \mathbb{R}$):

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

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$$f(x) \approx \mathcal{N}(x) \left(1 + \frac{\kappa^3}{6} \mathcal{H}_3(x) + \frac{\kappa^4}{24} \mathcal{H}_4(x) \right)$$