

# Syntheses Hyvarinen

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## 1 Results

**ICA** Assume that we observe  $n$  linear mixtures  $x_1, \dots, x_n$  of  $n$  independent components:

$$x_j = a_{j1}s_1 + a_{j2}s_2 + \dots + a_{jn}s_n, \forall j$$

where all  $x_j, s_j$  are random variables with zero mean and the  $s_j$  are statistically independent. Equivalently:

$$x = As$$

Given the entry  $x$ , the goal is to estimate both  $A$  and  $s$ .

**Constraint** The vector  $s$  must be *nongaussian*. One can prove that the distribution of any orthogonal transformation of the Gaussian  $(x_1, x_2)$  has exactly the same distribution as  $(x_1, x_2)$ , and that  $x_1$  and  $x_2$  are independent. Thus, in the case of Gaussian variables, we can only estimate the ICA model up to an orthogonal transformation. In other words, the matrix  $A$  is not identifiable for Gaussian independent components.

**Notations**  $y = w^T x$  and  $z = A^T w$  such that:  $y = z^T s$

**Idea** Use the Central Limit Theorem to determine  $w$  so that it would equal one of the rows of the inverse of  $A$ . Choose  $w$  such that it corresponds to a  $z$  with only one nonzero coordinate. Consequently,  $z^T s$  is equal to one of the wanted components up to a multiplicative sign.

**Kurtosis** This quantity defined for a zero-mean random variable with unit variance by:

$$kurt(y) = \mathbb{E}(y^4) - 3(\mathbb{E}(y^2))^2$$

is commonly used to measure nongaussianity as is it zero for a Gaussian variable and often nonzero for a nongaussian variable. Besides, for independent variables:

$$\begin{aligned} kurt(x_1 + x_2) &= kurt(x_1) + kurt(x_2) \\ kurt(\alpha x_1) &= \alpha^4 kurt(x_1) \end{aligned}$$

**Problem formulation**

$$\begin{aligned} &\text{maximize } |kurt(y)| \\ &\text{st } y = z^T s \\ &\text{and } Var(y) = 1 \end{aligned}$$

The drawback is that the kurtosis measure is not robust.

**Negentropy** Another measure of the nongaussianity of a random variable is its negentropy:

$$J(y) = H(y_{gauss}) - H(y)$$

Where  $y_{gauss}$  is the Gaussian random variable with the same covariance matrix than  $y$ . The entropy has the property of being maximal for a Gaussian variable among all variables of equal variance. Thus, the negentropy is zero for a Gaussian variable and nonnegative for other variables. Besides, negentropy is invariant for invertible linear transformations. in this form, the problem consists in maximizing the negentropy.

### Preprocessing for ICA

- Centering
- Whitening
- Dimension reduction (PCA)
- Band-Pass filtering

## 2 FastICA

**Idea** Use Gram-Schmidt-like decorrelation to compute independent  $w_i$ .

**Algorithm** Assuming that the data is whitened, initialize  $W$  at random and compute for any norm different of the Frobenius norm:

1.  $W = W / \sqrt{\|WW^T\|}$
2. repeat until convergence:  $W = \frac{3}{2}W - \frac{1}{2}WW^TW$

## 3 Applications

- Separation of Artifacts in MEG Data
- Finding Hidden Factors in Financial Data (data available on the challenges website)
- Reducing Noise in Natural Images