

ICA

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under

Given samples $x = As$

unknown $A \in \mathbb{R}^{n \times d}$

$s = \begin{pmatrix} s_1 \\ \vdots \\ s_d \end{pmatrix}$ has independent coordinates

We can assume $\mathbb{E}(s) = 0$ (else recenter x)

and $\mathbb{E}(s_i^2) = 1$

() () since scaling s_i is same as scaling $A^{(i)}$.

Problem Recover columns of A up to sign.

Q. Is this uniquely identifiable?

Suppose $s_1, s_2 \sim N(0, 1)$

then for any rotation A , $x = As \sim N(0, I_2)$

Thm. If at most one component of s is Gaussian, then A is uniquely identifiable

Gaussian, then A is unitary up to signs of its columns.

How?

Let's try moments -

$$\mathbb{E}(x) = \mathbb{E}(As) = A\mathbb{E}(s) = 0$$

$$\mathbb{E}(xx^T) = A\mathbb{E}(ss^T)A^T = AA^T$$

So we can again apply a linear transformation to make $\mathbb{E}(xx^T) = I = AA^T$ if A has full rank n .

But this is not enough

$\mathbb{E}(x \otimes x \otimes x)$ might be 0. e.g. if s_i are symmetric.

$$\mathbb{E}(x \otimes x \otimes x \otimes x) = \mathbb{E}(\otimes^4 x)$$

$$= \mathbb{E}(\otimes^4 As)$$

$$\begin{aligned} \mathbb{E}(\otimes^4 As)_{ijkl} &= \mathbb{E}((As)_i (As)_j (As)_k (As)_l) \\ &= \mathbb{E}\left(\sum_i A_{ii'} s_{i'}, \sum_j A_{jj'} s_{j'}, \sum_k A_{kk'} s_{k'}, \sum_l A_{ll'} s_{l'}\right) \\ &= \sum_{i'i'j'k'l'} A_{ii'} A_{jj'} A_{kk'} A_{ll'} \mathbb{E}(s_{i'} s_{j'} s_{k'} s_{l'}) \end{aligned}$$

$\nu_{\mu} \nu_{\tau}, \nu_{\tau}$

$$\mathbb{E}(A_i, A_j, A_k, A_{l'}) = \begin{cases} \mathbb{E}(\Delta_{i'}^4) & \text{if } i' = j' = k' = l' \\ \mathbb{E}(A_{i'})\mathbb{E}(A_{k'}) & \text{if } i' = j' (\#) k' = l' \\ 0 & \text{if } i' = k', j' = l' \\ 0, \omega. & i' = l', j' = k' \end{cases}$$

$$= \sum_{i'} A_{ii'} A_{ji'} A_{ki'} A_{li'} \mathbb{E}(\Delta_{i'}^4)$$

$$+ \sum_{i' \neq k'} A_{ii'} A_{ji'} A_{kk'} A_{lk'}$$

$$+ \sum_{i' \neq j'} A_{ii'} A_{jj'} A_{ki'} A_{lj'}$$

$$+ \sum_{i' \neq l'} A_{ii'} A_{jl'} A_{kl'} A_{li'}$$

$$= \sum_{i'} A_{ii'} A_{ji'} A_{ki'} A_{li'} (\mathbb{E}(\Delta_{i'}^4) - 3)$$

$$+ \sum_{i', k'} A_{ii'} A_{ji'} A_{kk'} A_{lk'}$$

$$+ \sum_{i', j'} A_{ii'} A_{jj'} A_{ki'} A_{lj'}$$

$$+ \sum_{i', l'} A_{ii'} A_{jl'} A_{kl'} A_{li'}$$

(

$$E(x_i x_j) E(x_k x_e) + E(x_i x_k) E(x_j x_e)$$

$$+ E(x_i x_e) E(x_j x_k) = M_{ijkl}$$

So,

$$E(\otimes^4 x) - M = \sum_i (E(x_i^4) - 3) A^{(i)} \otimes A^{(i)} \otimes A^{(i)} \otimes A^{(i)}$$

A tensor decomposition into orthogonal vectors!
We can solve this by Tensor power iteration.

OR. pick random $v \sim N(0, I)$

$$\text{Apply } T[\cdot, \cdot, v, v] = \sum_i (E(x_i^4) - 3) \underbrace{(A^{(i)} \cdot v)^2}_{\alpha_i} A^{(i)} A^{(i)T}$$

$$= A \begin{bmatrix} \alpha_1 & & 0 \\ & \ddots & \\ 0 & & \alpha_d \end{bmatrix} A^T$$

eigenvalues are columns of A !

eigenvalues are distinct w/HP!