

ICA

Sunday, September 12, 2021 4:08 PM

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 $E(s) = 0$ (else recenter x)

and $E(s_i^2) = 1$

$\left(\right) \left(\right)$ 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 uniquely determined up to signs of its columns.

How?

Let's try moments.

$$E(x) = E(As) = A E(s) = 0$$

$$E(x x^T) = A E(s s^T) A^T = A A^T$$

So we can again apply a linear transformation to make $E(x x^T) = I = A A^T$ if A has full rank n .

But this is not enough

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

$$E(x \otimes x \otimes x \otimes x) = E(\otimes^4 x)$$

$$= E(\otimes^4 As)$$

$$E(\otimes^4 As)_{i'j'k'l} = E((As)_{i'} (As)_{j'} (As)_{k'} (As)_{l'})$$

$$= E\left(\sum_{i''} A_{i'i''} s_{i''} \sum_{j''} A_{j'j''} s_{j''} \sum_{k''} A_{k'k''} s_{k''} \sum_{l''} A_{l'l''} s_{l''}\right)$$

$$= \sum_{i'',j'',k'',l''} A_{i'i''} A_{j'j''} A_{k'k''} A_{l'l''} E(s_{i''} s_{j''} s_{k''} s_{l''})$$

i, j, k, l

$$E(\Delta_{i'} \Delta_{j'} \Delta_{k'} \Delta_{l'}) = \begin{cases} E(\Delta_{i'}^4) & \text{if } i'=j'=k'=l' \\ E(\Delta_{i'})E(\Delta_{k'}) & \text{if } i'=j' \neq k'=l' \\ & \text{or } i'=k', j'=l' \\ & \text{or } i'=l', j'=k' \\ 0 & \text{o.w.} \end{cases}$$

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

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

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

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

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

$$\left[\begin{aligned} &+ \sum_{i', k'} A_{ii'} A_{jj'} A_{kk'} A_{lk'} \\ &+ \sum_{i', j'} \quad \quad \quad + \sum_{i', l'} \end{aligned} \right]$$



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

So,

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

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

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

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

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

eigenvectors are columns of A !
eigenvalues are distinct WHP. !