

The Derivation of TCA

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1. How to get the formulation of TCA (Transfer Component Analysis)?

Solution:

We give two kinds of expressions: the first one is to show JDA paper, while the second one is for TCA paper.

$$\begin{aligned}
 & \left\| \frac{1}{n_s} \sum_{i=1}^{n_s} \mathbf{A}^T \mathbf{x}_i - \frac{1}{n_t} \sum_{j=n_s+1}^{n_s+n_t} \mathbf{A}^T \mathbf{x}_j \right\|^2 \\
 &= \left\| \frac{1}{n_s} \mathbf{A}^T \begin{bmatrix} \mathbf{x}_1 & \mathbf{x}_2 & \cdots & \mathbf{x}_{n_s} \end{bmatrix}_{1 \times n_s} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}_{n_s \times 1} - \frac{1}{n_t} \mathbf{A}^T \begin{bmatrix} \mathbf{x}_1 & \mathbf{x}_2 & \cdots & \mathbf{x}_{n_t} \end{bmatrix}_{1 \times n_t} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}_{n_t \times 1} \right\|^2 \\
 &= \text{tr} \left(\frac{1}{n_s^2} \mathbf{A}^T \mathbf{X}_s \mathbf{1} (\mathbf{A}^T \mathbf{X}_s \mathbf{1})^T + \frac{1}{n_t^2} \mathbf{A}^T \mathbf{X}_t \mathbf{1} (\mathbf{A}^T \mathbf{X}_t \mathbf{1})^T - \frac{1}{n_s n_t} \mathbf{A}^T \mathbf{X}_s \mathbf{1} (\mathbf{A}^T \mathbf{X}_t \mathbf{1})^T - \frac{1}{n_s n_t} \mathbf{A}^T \mathbf{X}_t \mathbf{1} (\mathbf{A}^T \mathbf{X}_s \mathbf{1})^T \right) \\
 &= \text{tr} \left(\frac{1}{n_s^2} \mathbf{A}^T \mathbf{X}_s \mathbf{1} \mathbf{1}^T \mathbf{X}_s^T \mathbf{A} + \frac{1}{n_t^2} \mathbf{A}^T \mathbf{X}_t \mathbf{1} \mathbf{1}^T \mathbf{X}_t^T \mathbf{A} - \frac{1}{n_s n_t} \mathbf{A}^T \mathbf{X}_s \mathbf{1} \mathbf{1}^T \mathbf{X}_t^T \mathbf{A} - \frac{1}{n_s n_t} \mathbf{A}^T \mathbf{X}_t \mathbf{1} \mathbf{1}^T \mathbf{X}_s^T \mathbf{A} \right) \\
 &= \text{tr} \left[\mathbf{A}^T \left(\frac{1}{n_s^2} \mathbf{1} \mathbf{1}^T \mathbf{X}_s^T \mathbf{X}_s + \frac{1}{n_t^2} \mathbf{1} \mathbf{1}^T \mathbf{X}_t^T \mathbf{X}_t - \frac{1}{n_s n_t} \mathbf{1} \mathbf{1}^T \mathbf{X}_s^T \mathbf{X}_t - \frac{1}{n_s n_t} \mathbf{1} \mathbf{1}^T \mathbf{X}_t^T \mathbf{X}_s \right) \mathbf{A} \right] \\
 &= \text{tr} \left(\mathbf{A}^T \begin{bmatrix} \mathbf{X}_s & \mathbf{X}_t \end{bmatrix} \begin{bmatrix} \frac{1}{n_s^2} \mathbf{1} \mathbf{1}^T & \frac{-1}{n_s n_t} \mathbf{1} \mathbf{1}^T \\ \frac{-1}{n_s n_t} \mathbf{1} \mathbf{1}^T & \frac{1}{n_t^2} \mathbf{1} \mathbf{1}^T \end{bmatrix} \begin{bmatrix} \mathbf{X}_s \\ \mathbf{X}_t \end{bmatrix} \mathbf{A} \right) \\
 &= \text{tr} (\mathbf{A}^T \mathbf{X} \mathbf{M} \mathbf{X}^T \mathbf{A})
 \end{aligned}$$

Important property:

1. $\|\mathbf{A}\|^2 = \text{tr}(\mathbf{A} \mathbf{A}^T)$, which is used in the second equation.
2. $\text{tr}(\mathbf{A} \mathbf{B}) = \text{tr}(\mathbf{B} \mathbf{A})$, which is used in the fourth equation.

If used in kernel trick, then

$$\begin{aligned}
& \left\| \frac{1}{n_s} \sum_{i=1}^{n_s} \phi(\mathbf{x}_i) - \frac{1}{n_t} \sum_{j=1}^{n_t} \phi(\mathbf{x}_j) \right\|^2 \\
&= \text{tr} \left(\begin{bmatrix} \phi(\mathbf{x}_s) & \phi(\mathbf{x}_t) \end{bmatrix} \begin{bmatrix} \frac{1}{n_s^2} \mathbf{1}\mathbf{1}^T & \frac{-1}{n_s n_t} \mathbf{1}\mathbf{1}^T \\ \frac{-1}{n_s n_t} \mathbf{1}\mathbf{1}^T & \frac{1}{n_t^2} \mathbf{1}\mathbf{1}^T \end{bmatrix} \begin{bmatrix} \phi(\mathbf{x}_s) \\ \phi(\mathbf{x}_t) \end{bmatrix} \right) \\
&= \text{tr} \left(\begin{bmatrix} \phi(\mathbf{x}_s) \\ \phi(\mathbf{x}_t) \end{bmatrix} \begin{bmatrix} \phi(\mathbf{x}_s) & \phi(\mathbf{x}_t) \end{bmatrix} \begin{bmatrix} \frac{1}{n_s^2} \mathbf{1}\mathbf{1}^T & \frac{-1}{n_s n_t} \mathbf{1}\mathbf{1}^T \\ \frac{-1}{n_s n_t} \mathbf{1}\mathbf{1}^T & \frac{1}{n_t^2} \mathbf{1}\mathbf{1}^T \end{bmatrix} \right) \\
&= \text{tr} \left(\begin{bmatrix} \langle \phi(\mathbf{x}_s), \phi(\mathbf{x}_s) \rangle & \langle \phi(\mathbf{x}_s), \phi(\mathbf{x}_t) \rangle \\ \langle \phi(\mathbf{x}_t), \phi(\mathbf{x}_s) \rangle & \langle \phi(\mathbf{x}_t), \phi(\mathbf{x}_t) \rangle \end{bmatrix} \mathbf{M} \right) \\
&= \text{tr} \left(\begin{bmatrix} K_{s,s} & K_{s,t} \\ K_{t,s} & K_{t,t} \end{bmatrix} \mathbf{M} \right)
\end{aligned}$$

where

$$(M)_{ij} = \begin{cases} \frac{1}{n_s n_s}, & \mathbf{x}_i, \mathbf{x}_j \in \mathcal{D}_s \\ \frac{1}{n_t n_t}, & \mathbf{x}_i, \mathbf{x}_j \in \mathcal{D}_t \\ \frac{-1}{n_s n_t}, & \text{otherwise} \end{cases}$$