

## PCA algorithms proof

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Suppose we have the following collection of  $m$  points

$$\{x^{(1)}, x^{(2)}, \dots, x^{(m)}\} \text{ in } \mathbb{R}^n$$

We like to reduce each  $x^{(i)}$  in the collection, i.e.,

$$f: \mathbb{R}^n \rightarrow \mathbb{R}^l \quad l \leq n$$

$$f(x) = c$$

So  $f$  is a function that map  $x$  to a low dimension vector space, so  $f$  is our coding function. That means we'll need a decode function for  $f$   $g$

$$g: \mathbb{R}^l \rightarrow \mathbb{R}^n \quad l \leq n \text{ and } g(f(x)) \approx x$$

the most simple transformation is define  $g(c) = Dc$ ,  $D \in \mathbb{R}^{n \times l}$

to keep the decoding function easy let's define some properties for  $D$

1) the columns of  $D$  define an orthonormal basis

let be  $d^{(1)}, d^{(2)}, \dots, d^{(l)}$

$$d^{(i)T} d^{(j)} = \begin{cases} 0, & i \neq j \\ 1, & i = j \end{cases}$$

$$2) D^T D = I_l$$

this properties give the problem a unique solution

Now we need a definition for  $f(x) = c$ , and we know

$$a(x) = Dc \text{ and } a(f(x)) \approx x \Rightarrow c = D^T x$$

$g(c) = Dc$  so we can obtain an problem

$$c^* = \underset{c}{\operatorname{argmin}} \|x - g(c)\|_2$$

because  $L_2$  norm is always positive then optimize the before is the same that optimize the square of it because square function are monotonic for positive arg

$$c^* = \underset{c}{\operatorname{argmin}} \|x - g(c)\|_2^2$$

$$\begin{aligned} \|x - g(c)\|_2^2 &= (x - g(c))^T (x - g(c)) \\ &= x^T x - x^T g(c) - g(c)^T x + g(c)^T g(c) \end{aligned}$$

$$x^T g(c) \in \mathbb{R}$$

$$\Rightarrow x^T g(c) = (x^T g(c))^T \Rightarrow = x^T x - 2x^T g(c) + g(c)^T g(c)$$

$$= g(c)^T x \Rightarrow x^T x \text{ doesn't depend so is constant}$$

Using the definition of  $g(c) = Dc$

$$\Rightarrow c^* = \underset{c}{\operatorname{argmin}} (-2x^T g(c) + g(c)^T g(c))$$

$$\Rightarrow c^* = \underset{c}{\operatorname{argmin}} (-2x^T Dc + (Dc)^T (Dc))$$