## Derivative of the first order SVD

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Consider computing the first order SVD expansion. By the Eckart–Young–Mirsky theorem, this is equivalent to solving

$$\underset{\sigma, u, v}{\text{minimize}} \frac{1}{2} \|A - \sigma u v^{\top}\|_F^2 \quad \text{s.t.} \quad \|u\| = 1 \quad \text{and} \quad \|v\| = 1$$

Any triplet  $(\sigma, u, v)$  for a **unique** singular value satisfies

The jacobian and lagrangian The derivative of the objective function is

$$\mathbf{J}_f(A, \begin{bmatrix} \sigma \\ u \\ v \end{bmatrix}) = \begin{bmatrix} \sigma - u^\top A v & \sigma - u^\top A v \\ A - \sigma u v^\top & \sigma^2 u - \sigma A v \\ \sigma^2 v - \sigma A^\top u \end{bmatrix} \implies \mathbf{H}_f(\begin{bmatrix} \sigma \\ u \\ v \end{bmatrix}) = \begin{bmatrix} 1 & 2\sigma u - A v & 2\sigma v - A^\top u \\ -A v & \sigma^2 \mathbb{I}_m & -\sigma A \\ -A^\top u & -\sigma A^\top & \sigma^2 \mathbb{I}_n \end{bmatrix}$$

Consider the function

$$f(A, \begin{bmatrix} \sigma \\ u \\ v \end{bmatrix}) = \begin{pmatrix} \sigma - u^{\top} A v \\ \sigma^{2} u - \sigma A v \\ \sigma^{2} v - \sigma A^{\top} u \end{pmatrix} \equiv \mathbf{0} \implies \mathbf{J}_{f}(A, \begin{bmatrix} \sigma \\ u \\ v \end{bmatrix}) = \begin{bmatrix} -\xi v u^{\top} & 1 & 2\sigma u - A v & 2\sigma v - A^{\top} u \\ -\sigma v \phi^{\top} & -A v & \sigma^{2} \mathbb{I}_{m} & -\sigma A \\ -\sigma u \psi^{\top} & -A^{\top} u & -\sigma A^{\top} & \sigma^{2} \mathbb{I}_{n} \end{bmatrix}$$

Thus, gradient descent schema is

$$\sigma' = \sigma - \eta_{\sigma}(\sigma - u^{\top}Av)$$
  

$$u' = u - \eta_{u}(\sigma^{2}u - \sigma Av)$$
  

$$v' = v - \eta_{v}(\sigma^{2}v - \sigma A^{\top}u)$$

And the newton step with diagonal approximation of the hessian:

$$\begin{aligned} \sigma' &= \sigma - 1(\sigma - u^{\top}Av) &= u^{\top}Av \\ u' &= u - \frac{1}{\sigma^2}(\sigma^2u - \sigma Av) &= \frac{1}{\sigma}Av \\ v' &= v - \frac{1}{\sigma^2}(\sigma^2v - \sigma A^{\top}u) &= \frac{1}{\sigma}A^{\top}u \end{aligned}$$

$$\sigma = u^{\top} A v$$
$$A v = \sigma u$$
$$A^{\top} u = \sigma v$$

from this we can derive

$$\Delta \sigma = \Delta u^{\top} A v + u^{\top} \Delta A v + u^{\top} A \Delta v$$
$$= \Delta u^{\top} u + u^{\top} \Delta A v + v^{\top} \Delta v$$
$$= u^{\top} \Delta A v$$

Where in the last step we used  $\Delta u \perp u$  and  $\Delta v \perp v$ , which follows from the side condition. Further we have:

$$\begin{array}{c} \Delta A v + A \Delta v = \Delta \sigma u + \sigma \Delta u \\ \Delta A^\top u + A^\top \Delta u = \Delta \sigma v + \sigma \Delta v \end{array} \Longleftrightarrow \underbrace{ \begin{bmatrix} \sigma \mathbb{I}_m & -A \\ -A^\top & \sigma \mathbb{I}_n \end{bmatrix}}_{=:K} \cdot \begin{bmatrix} \Delta u \\ \Delta v \end{bmatrix} = \begin{bmatrix} \Delta A v - \Delta \sigma u \\ \Delta A^\top u - \Delta \sigma v \end{bmatrix}$$

## 1 The VJP

The last equation allows us to compute the VJP at ease:

$$\begin{split} \left\langle \begin{bmatrix} \phi \\ \psi \end{bmatrix} \middle| \begin{bmatrix} \Delta u \\ \Delta v \end{bmatrix} \right\rangle &= \left\langle \begin{bmatrix} \phi \\ \psi \end{bmatrix} \middle| K^{-1} \begin{bmatrix} \Delta A v - \Delta \sigma u \\ \Delta A^{\top} u - \Delta \sigma v \end{bmatrix} \right\rangle \\ &= \left\langle K^{-\top} \begin{bmatrix} \phi \\ \psi \end{bmatrix} \middle| \begin{bmatrix} \Delta A v - \Delta \sigma u \\ \Delta A^{\top} u - \Delta \sigma v \end{bmatrix} \right\rangle \\ &= \left\langle \begin{bmatrix} \tilde{\phi} \\ \tilde{\psi} \end{bmatrix} \middle| \begin{bmatrix} \Delta A v - \Delta \sigma u \\ \Delta A^{\top} u - \Delta \sigma v \end{bmatrix} \right\rangle \end{split}$$

Now, we compute the terms individually:

$$\begin{split} \langle \tilde{\phi} \mid \Delta A v - \Delta \sigma u \rangle &= \langle \tilde{\phi} v^\top \mid \Delta A \rangle - \langle u^\top \tilde{\phi} \mid \Delta \sigma \rangle \\ &= \langle \tilde{\phi} v^\top \mid \Delta A \rangle - \langle u^\top \tilde{\phi} \mid u^\top \Delta A v \rangle \\ &= \langle (\mathbb{I}_m - u u^\top) \tilde{\phi} v^\top \mid \Delta A \rangle \end{split}$$

And for the second term we get

$$\langle \tilde{\psi} \mid \Delta A^{\top} u - \Delta \sigma v \rangle = \langle \tilde{\psi} u^{\top} \mid \Delta A^{\top} \rangle - \langle v^{\top} \tilde{\psi} \mid \Delta \sigma \rangle$$
$$= \langle u \tilde{\psi}^{\top} \mid \Delta A \rangle - \langle \tilde{\psi}^{\top} v \mid u^{\top} \Delta A v \rangle$$
$$= \langle u \tilde{\psi} (\mathbb{I}_n - v v^{\top}) \mid \Delta A \rangle$$

Using the formula for inverting a block-matrix, we can give an explicit solution to  $K^{-\top}$ :

$$\begin{bmatrix} \sigma \mathbb{I}_m & -A \\ -A^\top & \sigma \mathbb{I}_n \end{bmatrix}^{-1} = \begin{bmatrix} (\sigma \mathbb{I}_m - \frac{1}{\sigma} A A^\top)^{-1} & \mathbf{0}_{m \times n} \\ \mathbf{0}_{n \times m} & (\sigma \mathbb{I}_n - \frac{1}{\sigma} A^\top A)^{-1} \end{bmatrix} \cdot \begin{bmatrix} \mathbb{I}_m & \frac{1}{\sigma} A \\ \frac{1}{\sigma} A^\top & \mathbb{I}_n \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{\sigma} (\mathbb{I}_m - \frac{1}{\sigma^2} A A^\top)^{-1} & \frac{1}{\sigma^2} (\mathbb{I}_m - \frac{1}{\sigma^2} A A^\top)^{-1} A \\ \frac{1}{\sigma^2} (\mathbb{I}_n - \frac{1}{\sigma^2} A^\top A)^{-1} A^\top & \frac{1}{\sigma} (\mathbb{I}_n - \frac{1}{\sigma^2} A^\top A)^{-1} \end{bmatrix}$$

$$= \frac{1}{\sigma} \begin{bmatrix} (\mathbb{I}_m - \tilde{A} \tilde{A}^\top)^{-1} & (\mathbb{I}_m - \tilde{A} \tilde{A}^\top)^{-1} \tilde{A} \\ (\mathbb{I}_n - \tilde{A}^\top \tilde{A})^{-1} \tilde{A}^\top & (\mathbb{I}_n - \tilde{A}^\top \tilde{A})^{-1} \end{bmatrix}$$

And we see it's basically projection operators with respect to the image/kernel of  $\tilde{A} = \frac{1}{\sigma}A$ . In summary, we obtain the following formula for the VJP:

$$\begin{bmatrix} \sigma \mathbb{I}_{m} & -A \\ -A^{\top} & \sigma \mathbb{I}_{n} \end{bmatrix}^{-\top} \begin{bmatrix} \tilde{\phi} \\ \tilde{\psi} \end{bmatrix} = \begin{bmatrix} \phi \\ \psi \end{bmatrix} \iff \begin{bmatrix} \tilde{\phi} \\ \tilde{\psi} \end{bmatrix} = \frac{1}{\sigma} \begin{bmatrix} (\mathbb{I}_{m} - \tilde{A}\tilde{A}^{\top})^{-1} & (\mathbb{I}_{m} - \tilde{A}\tilde{A}^{\top})^{-1}\tilde{A} \\ (\mathbb{I}_{n} - \tilde{A}^{\top}\tilde{A})^{-1}\tilde{A}^{\top} & (\mathbb{I}_{n} - \tilde{A}^{\top}\tilde{A})^{-1} \end{bmatrix}^{\top} \begin{bmatrix} \phi \\ \psi \end{bmatrix}$$
$$\iff \begin{bmatrix} \tilde{\phi} \\ \tilde{\psi} \end{bmatrix} = \frac{1}{\sigma} \begin{bmatrix} (\mathbb{I}_{m} - \tilde{A}\tilde{A}^{\top})^{-1} & \tilde{A}(\mathbb{I}_{n} - \tilde{A}^{\top}\tilde{A})^{-1} \\ \tilde{A}^{\top}(\mathbb{I}_{m} - \tilde{A}\tilde{A}^{\top})^{-1} & (\mathbb{I}_{n} - \tilde{A}^{\top}\tilde{A})^{-1} \end{bmatrix} \begin{bmatrix} \phi \\ \psi \end{bmatrix}$$

Thus, we need to solve 4 linear systems:

$$(\mathbb{I}_m - \tilde{A}\tilde{A}^\top)x = \phi \qquad (\mathbb{I}_m - \tilde{A}^\top\tilde{A})y = \tilde{A}\psi$$
  
$$(\mathbb{I}_m - \tilde{A}\tilde{A}^\top)z = \tilde{A}^\top\phi \qquad (\mathbb{I}_m - \tilde{A}^\top\tilde{A})w = \psi$$

Then  $\tilde{\phi} = \frac{1}{\sigma}(x+y)$  and  $\tilde{\psi} = \frac{1}{\sigma}(z+w)$ , and the VJP are given by the previous equations:

$$\xi^{\top} \frac{\partial \sigma}{\partial A} = \xi u v^{\top}$$

$$\phi^{\top} \frac{\partial u}{\partial A} = (\mathbb{I}_m - u u^{\top}) \tilde{\phi} v^{\top}$$

$$\psi^{\top} \frac{\partial v}{\partial A} = u \tilde{\psi} (\mathbb{I}_n - v v^{\top})$$