

A Simple Exercise on Matrix Derivatives

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Abstract

This is an exercise to remind myself of how to do simple matrix derivatives.

Thomas Minka has an excellent discussion of derivatives of matrices [1]. Let $U \in \mathbb{R}^{n \times k}$, $V \in \mathbb{R}^{m \times k}$, and $Y \in \mathbb{R}^{n \times m}$. Let

$$J(U, V) = \|UV^T - Y\|_{\text{Fro}}^2 + \frac{\lambda}{2}(\|U\|_{\text{Fro}}^2 + \|V\|_{\text{Fro}}^2) \quad (1)$$

$$= \sum_{i,j} \left(\sum_a U_{ia} V_{ja} - Y_{ij} \right)^2 + \frac{\lambda}{2} \left(\sum_{i,a} U_{ia}^2 + \sum_{j,a} V_{ja}^2 \right) \quad (2)$$

Let U_i be the i^{th} row of U , let V_j be the j^{th} row of V , let Y_i be the i^{th} row of Y , and let V_a be the a^{th} column of V . Then,

$$\frac{\partial J}{\partial U_{ia}} = 2 \sum_j \left(\sum_a U_{ia} V_{ja} - Y_{ij} \right) V_{ja} + \lambda U_{ia}, \quad (3)$$

$$= 2 \sum_j (U_i V_j^T - Y_{ij}) V_{ja} + \lambda U_{ia}, \quad (4)$$

$$= 2 (U_i V^T - Y_i) V_a + \lambda U_{ia}, \quad (5)$$

$$\frac{\partial J}{\partial U_i} = 2 (U_i V^T - Y_i) V + \lambda U_i, \quad (6)$$

$$\frac{\partial J}{\partial U} = 2 (UV^T - Y) V + \lambda U. \quad (7)$$

Similarly,

$$\frac{\partial J}{\partial V} = 2 (UV^T - Y)^T U + \lambda V. \quad (8)$$

Note that this does *not* follow the convention of swapping indices as discussed in [1].

References

- [1] T. Minka. Old and new matrix algebra useful for statistics.
<http://research.microsoft.com/minka/papers/matrix/>, December 2000.