Supplementary Material: Efficient Stochastic Optimization for Low-Rank Distance Metric Learning

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Proof of Theorem 2

Note that $S_+ \cap \{W | \|W\|_2 < \tau\}$ [X] is the optimal solution to the following problem

The Lagrangian function associated with (6) is

$$L(A; Y_1; Y_2) = \frac{jjA - Xjj_F^2}{2} tr(AY_1) + tr((A - I)Y_2)$$

where $Y_1 \supseteq \mathbb{R}^{d \times d}$ and $Y_2 \supseteq \mathbb{R}^{d \times d}$ are dual variables for constraints $A \supseteq \mathbb{S}_+$ and $A \supseteq \mathbb{S}_+$. Let A^*, Y_1^*, Y_2^* be the optimal primal and dual solutions. The KKT conditions are

$$A^* 2 S_+;$$

$$I A^* 2 S_+;$$

$$A^* = X + Y_1^* Y_2^*;$$

$$tr(A^* Y^* X + Y^* Y_2^*) A Td_= f(Y)]TJ/F38 9.9626 Tf 9.409 0 Td [(S)]$$

$$A^* \ 2 \, \mathbb{S}_+;$$

$$kA^* k_F$$

$$A^* = \frac{1}{1 + *} (X + Y_1^*);$$

$$\text{tr}(A^* Y_1^*) = 0;$$

$$^* (kA^* k_F^2 \quad ^2) = 0;$$

$$Y_1^* \ 2 \, \mathbb{S}_+; \quad ^* \quad 0:$$

We complete the proof by noticing that

$$Y_{1}^{*} = \begin{cases} \mathbf{X} \\ i \mathbf{u}_{i} \mathbf{u}_{i}^{\top}, \\ i \cdot \lambda_{i} < 0 \\ \mathbf{X} \\$$

satisfy these KKT conditions.