

Without loss of generality, let us instantiate the objective function with the regularizer λ of empirical error on the labeled data for a family of hash codes.

$$f(W, Q) = \mathcal{L}$$

$$\min \sum_i \sum_k \left\| W_k^T P_k^i - H_k^i \right\|_F^2 + \lambda \sum_i \sum_k \sum_{u,v} -s_{k,w}^i r_{uv}^i \quad (\text{A-1})$$

One can express the second term of the above objective function in a compact matrix form by defining matrices $S_k^i \in R^{N_i \times N_i}$ and $R^i \in R^{N_i \times N_i}$ incorporating the pairwise similarity of P_k^i and the pairwise relationship of P^i for labeled information respectively. Then, $S\left(\left\{W_k\right\}_{k=1}^K\right)$ can be represented as

$$S\left(\left\{W_k\right\}_{k=1}^K\right) = \lambda \sum_i \sum_k \text{tr}(-S_k^i R^i) \quad (\text{A-2})$$

and using (1) and absorbing the constant term into λ we can also represent it as

$$S\left(\left\{W_k\right\}_{k=1}^K\right) = \lambda \sum_i \sum_k \text{tr}(-H_k^i R^i H_k^{iT}). \quad (\text{A-3})$$

The objective function becomes the follows.

$$f(W, Q) = \min \sum_i \sum_k \left\| W_k^T P_k^i - \text{sign}(Q_k^i) \right\|_F^2 + \mathcal{L} \\ \lambda \sum_i \sum_k \text{tr}(-\text{sign}(Q_k^i) R^i \text{sign}(Q_k^i)^T) \quad (\text{A-4})$$

Here, we concern two possible problems due to the sign function for Q . First, Q may not be a unique solution and thus the objective function is difficult to converge without considering any regularizer about Q . We add a Frobenius norm regularizer η . In addition, the objective function $f(W, Q)$ is nondifferentiable in terms of Q . We can approximate the sign function with the surrogate function (A-5).

$$S_L(Q_k^i) = (Q_k^i \circ Q_k^i + \xi)^{\frac{-1}{2}} \circ Q_k^i \quad (\text{A-5})$$

Here, ξ is a positive constant close to zero and \circ is the hadamard (elementwise) product. Finally, we have the following objective function.

$$f(W, Q) = \min \sum_i \sum_k \left\| W_k^T P_k^i - S_L(Q_k^i) \right\|_F^2 + \mathcal{L}$$

$$\lambda \sum_i \sum_k \text{tr}(-S_L(Q_k^i) R^i S_L(Q_k^i)^T) + \eta \sum_i \sum_k \|Q_k^i\|_F^2 \quad (\text{A-6})$$

To minimize the objective function applying (A-6), we use a Newton-Raphson algorithm [55] and iteratively solve W and Q . When updating W with Q fixed we consider the first derivative of W in terms of k without loss of generality.

$$\frac{\partial f}{\partial W_k} = \sum_j 2P_k^j \{W_k^T P_k^j - (Q_k^i \circ Q_k^i + \xi)^{-\frac{1}{2}} \circ Q_k^i\} \quad (\text{A-7})$$

This approach allows us to update W_k for all $k(1 \leq k \leq K)$ simultaneously

$$W_k^{\text{new}} = W_k - \left(\frac{\partial^2 f}{\partial W_k^2} \right)^{-1} \frac{\partial f}{\partial W_k}. \text{ Similarly with } W \text{ fixed we can update } Q. \text{ To be specific,}$$

we update Q_k^i for all combinations of (i, k) , $1 \leq i \leq M$, $1 \leq k \leq K$ at the same time

$$Q_k^{i, \text{new}} = Q_k^i - \left(\frac{\partial^2 f}{\partial Q_k^i{}^2} \right)^{-1} \frac{\partial f}{\partial Q_k^i}. \text{ The first derivative of } Q_k^i \text{ is}$$

$$\frac{\partial f}{\partial Q_k^i} = -2 \{W_k^T P_k^i - S_L(Q_k^i)\} \circ S_L'(Q_k^i) - \lambda \nabla \text{tr} + 2\eta Q_k^i \quad (\text{A-8})$$

where ∇tr is defined in an elementwise manner

$$(\nabla \text{tr})_{p,q} = S_L'(Q_k^i; p, q) \left\{ \left(\sum_{t=1}^{N_i} S_L(Q_k^i; t, q) R_{t,p}^i \right) + S_L(Q_k^i; p, q) \right\} \quad (\text{A-9})$$

and $S_L'(Q_k^i) = (Q_k^i \circ Q_k^i + \xi)^{-\frac{3}{2}}$. Moreover, we calculate second derivatives for the Hessian matrix. We first consider the W_k

$$\frac{\partial^2 f}{\partial W_k^2} = \sum_i 2P_k^i P_k^{i T} \quad (\text{A-10})$$

In terms of Q_k^i , $\frac{\partial^2 f}{\partial Q_k^i{}^2}$ is derived as

$$\left(\left[-2(W_k^T P_k^i) S_L''(Q_k^i) + 2 \{S_L'(Q_k^i) \circ S_L'(Q_k^i) + S_L(Q_k^i) \circ S_L''(Q_k^i)\} \right] (I_{(N, b_k)})^T - \lambda \nabla^2 \text{tr} + 2\eta I_{(N, b_k)} \right) \quad (\text{A-11})$$

Where

$$\nabla^2 tr = \left[I_{(b_k)} \otimes \left((J_{(N_i)} + I_{(N_i)}) \circ R^i \right) \right] \circ \left\{ \left(S'_L(Q_k^i) \right) \left(S'_L(Q_k^i) \right)^T \right\} + \text{diag}(\vec{A}) \quad . \quad (\text{A-12})$$

Here, \otimes is the Kronecker product [59] and \circ and diag are to transform a matrix to a vector and a vector to a diagonal matrix respectively. In addition, J is a matrix of ones and A is defined in an elementwise manner as follows

$$(A)_{p,q} = S''_L(Q_k^i; p, q) \left\{ \left(\sum_{t=1}^{N_i} S_L(Q_k^i; t, q) R_{t,p}^i \right) + S_L(Q_k^i; p, q) \right\} \quad (\text{A-13})$$

where $S''_L(Q_k^i; p, q) = (Q_k^i \circ Q_k^i + \xi) - \frac{5}{2} Q_k^i$.