

APPENDIX A

DERIVATION OF THE SOLUTION IN (15)

Since the cost in (15) is quadratic, *i.e.*, convex, we differentiate it with respect to \mathcal{X} and set it to $\mathbf{0}$, *i.e.*,

$$\begin{aligned} & \frac{\partial}{\partial \mathcal{X}} \left(\frac{1}{2} \|\sqrt{\mathcal{F}_{\text{att}}(\mathcal{D})} \odot (\mathcal{L}_k * \mathcal{R}_k - \mathcal{X})\|_F^2 \right. \\ & \quad \left. + \frac{\mu_k + \alpha_k}{2} \|\mathcal{X} - \Psi_{\mathcal{X},k}\|_F^2 \right) \\ & = -\sqrt{\mathcal{F}_{\text{att}}(\mathcal{D})} \odot \left(\sqrt{\mathcal{F}_{\text{att}}(\mathcal{D})} \odot (\mathcal{L}_k * \mathcal{R}_k - \mathcal{X}) \right) \\ & \quad + (\mu_k + \alpha_k)(\mathcal{X} - \Psi_{\mathcal{X},k}) = \mathbf{0} \quad (34) \\ \Leftrightarrow & (\mathcal{F}_{\text{att}}(\mathcal{D}) + \mu_k \mathbf{1} + \alpha_k \mathbf{1}) \odot \mathcal{X} \\ & = \mathcal{F}_{\text{att}}(\mathcal{D}) \odot (\mathcal{L}_k * \mathcal{R}_k) + (\mu_k + \alpha_k) \Psi_{\mathcal{X},k}. \quad (35) \end{aligned}$$

Since each element of \mathcal{X} on the left-hand side of (35) is multiplied by a scalar value $([\mathcal{F}_{\text{att}}(\mathcal{D})](i_1, i_2, i_3) + \mu_k + \alpha_k)$, we obtain the closed-form solution in (15) by performing element-wise division.

APPENDIX B

DERIVATION OF THE SOLUTION IN (21)

The optimization problem in (21) can be reformulated as

$$\begin{aligned} & \underset{\mathcal{S}}{\text{minimize}} \quad \|\mathcal{S} - \Psi_{\mathcal{S},k}\|_F^2 \\ & \text{subject to} \quad \|\mathcal{P}_{\Omega}(\mathcal{S})\|_F \leq \delta_k. \end{aligned} \quad (36)$$

We solve this problem by using Theorem 1 from [1]. For the elements $(i_1, i_2, i_3) \notin \Omega$, the solution is given by

$$\mathcal{P}_{\Omega^c}(\mathcal{S}_{k+1}) = \mathcal{P}_{\Omega^c}(\Psi_{\mathcal{S},k}), \quad (37)$$

since it is apparent that $\mathcal{S}_{k+1}(i_1, i_2, i_3) = \Psi_{\mathcal{S},k}(i_1, i_2, i_3)$. For the elements $(i_1, i_2, i_3) \in \Omega$, we first denote $\mathcal{Z} = \mathcal{P}_{\Omega}(\mathcal{S})$ and $\mathcal{W} = \mathcal{P}_{\Omega}(\Psi_{\mathcal{S},k})$ for simpler notations. Then, the optimization in (36) can be rewritten as

$$\begin{aligned} & \underset{\mathcal{Z}}{\text{minimize}} \quad \|\mathcal{Z} - \mathcal{W}\|_F^2 \\ & \text{subject to} \quad \|\mathcal{Z}\|_F^2 \leq \delta_k^2. \end{aligned} \quad (38)$$

We define the Lagrangian function for (38) as

$$\mathcal{L}(\mathcal{Z}, \nu) = \|\mathcal{Z} - \mathcal{W}\|_F^2 + \nu(\|\mathcal{Z}\|_F^2 - \delta_k^2),$$

where ν is the Lagrange multiplier for the constraint. Then, the optimal \mathcal{Z} can be obtained by solving the Karush-Kuhn-Tucker conditions [2], *i.e.*,

$$\|\mathcal{Z}\|_F - \delta_k \leq 0, \quad (39)$$

$$\nu \geq 0, \quad (40)$$

$$\nu(\|\mathcal{Z}\|_F - \delta_k) = 0, \quad (41)$$

$$(1 + \nu)\mathcal{Z} - \mathcal{W} = \mathbf{0}, \quad (42)$$

where $\mathbf{0}$ is the zero tensor. By substituting \mathcal{Z} from (42) into (41), we obtain

$$\nu(\|\mathcal{Z}\|_F - \delta_k) = \frac{\nu}{1 + \nu} \{\|\mathcal{W}\|_F - (1 + \nu)\delta_k\} = 0. \quad (43)$$

We consider two cases.

Case 1: $\|\mathcal{W}\|_F < \delta_k$. Then, $\|\mathcal{W}\|_F - (1 + \nu)\delta_k < 0$, and thus $\nu = 0$ from (40). Therefore, we have $\mathcal{Z} = \mathcal{W}$.

Case 2: $\|\mathcal{W}\|_F \geq \delta_k$. In this case, $\|\mathcal{W}\|_F - (1 + \nu)\delta_k = 0$. Then, from (42), we have $\mathcal{Z} = \frac{\delta_k}{\|\mathcal{W}\|_F} \mathcal{W}$.

By combining the two cases, we obtain the optimal \mathcal{Z} as

$$\mathcal{Z} = \mathcal{P}_{\Omega}(\mathcal{S}_{k+1}) = \min \left\{ \frac{\delta_k}{\|\mathcal{P}_{\Omega}(\Psi_{\mathcal{S},k})\|_F}, 1 \right\} \mathcal{P}_{\Omega}(\Psi_{\mathcal{S},k}). \quad (44)$$

Since $\mathcal{S}_{k+1} = \mathcal{P}_{\Omega^c}(\mathcal{S}_{k+1}) + \mathcal{P}_{\Omega}(\mathcal{S}_{k+1})$, from (37) and (44), we obtain the solution in (21).

REFERENCES

- [1] C. Lee and E. Y. Lam, "Computationally efficient truncated nuclear norm minimization for high dynamic range imaging," *IEEE Trans. Image Process.*, vol. 25, no. 9, pp. 4145–4157, Sep. 2016.
- [2] S. Boyd and L. Vandenberghe, *Convex Optimization*. Cambridge, U.K.: Cambridge Univ. Press, 2004.