

Consider the optimization problem:

$$\begin{aligned} \min_{\mathbf{x}, \mathbf{y}, \mathbf{z}} \quad & \|\mathbf{y}\|_1 + \lambda \|\mathbf{z}\|_1 \\ \text{subject to} \quad & \mathbf{z} - \mathbf{x} = 0 \\ & \mathbf{y} - (\mathbf{s} - \mathbf{D}\mathbf{x}) = 0 \end{aligned} \quad (1)$$

This optimization problem has the Lagrangian function:

$$\mathcal{L}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\gamma}, \boldsymbol{\eta}) = \|\mathbf{y}\|_1 + \lambda \|\mathbf{z}\|_1 + \boldsymbol{\gamma}^H (\mathbf{z} - \mathbf{x}) + \boldsymbol{\eta}^H (\mathbf{y} - \mathbf{s} + \mathbf{D}\mathbf{x}) \quad (2)$$

Now, we can augment the Lagrangian. I will use the modification from the normalization document in my notes for one of the augmentation terms.

$$\mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = \|\mathbf{y}\|_1 + \lambda \|\mathbf{z}\|_1 + \boldsymbol{\gamma}^H (\mathbf{z} - \mathbf{x}) + \boldsymbol{\eta}^H (\mathbf{y} - \mathbf{s} + \mathbf{D}\mathbf{x}) + \frac{\rho}{2} \|\Lambda^{\frac{1}{2}} (\mathbf{z} - \mathbf{x})\|_2^2 + \frac{\rho}{2} \|\mathbf{y} - \mathbf{s} + \mathbf{D}\mathbf{x}\|_2^2 \quad (3)$$

$$\nabla_{\mathbf{x}} \mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = -\boldsymbol{\gamma} - \mathbf{D}^H \boldsymbol{\eta} + \rho \Lambda \mathbf{x} - \rho \Lambda \mathbf{z} + \rho \mathbf{D}^H \mathbf{D} \mathbf{x} - \rho \mathbf{D}^H (\mathbf{y} - \mathbf{s}) \quad (4)$$

$$\min_{\mathbf{x}} \mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = (\rho \Lambda + \rho \mathbf{D}^H \mathbf{D})^{-1} (\rho \mathbf{D}^H (\mathbf{y} - \mathbf{s} + \frac{\boldsymbol{\eta}}{\rho}) + \rho \Lambda \mathbf{z} + \boldsymbol{\gamma}) \quad (5)$$

$$\min_{\mathbf{x}} \mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = (\Lambda + \mathbf{D}^H \mathbf{D})^{-1} (\mathbf{D}^H (\mathbf{y} - \mathbf{s} + \frac{\boldsymbol{\eta}}{\rho}) + \Lambda \mathbf{z} + \frac{\boldsymbol{\gamma}}{\rho}) \quad (6)$$

$$\min_{\mathbf{y}} \mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = \mathcal{S}_{\frac{1}{\rho}}(\mathbf{s} - \mathbf{D}\mathbf{x} - \frac{\boldsymbol{\eta}}{\rho}) \quad (7)$$

$$\min_{\mathbf{z}} \mathcal{L}_{\rho, \Lambda}(\mathbf{x}, \mathbf{y}, \mathbf{z}, \boldsymbol{\eta}, \boldsymbol{\gamma}) = \mathcal{S}_{\frac{\lambda \Lambda^{-1}}{\rho}}(\mathbf{x} - \frac{\Lambda^{-1} \boldsymbol{\gamma}}{\rho}) \quad (8)$$

From this, we can get the update equations:

$$\mathbf{x}^{(k+1)} = (\Lambda + \mathbf{D}^H \mathbf{D})^{-1} (\mathbf{D}^H (\mathbf{y}^{(k)} - \mathbf{s} + \frac{\boldsymbol{\eta}^{(k)}}{\rho}) + \Lambda \mathbf{z}^{(k)} + \frac{\boldsymbol{\gamma}^{(k)}}{\rho}) \quad (9)$$

$$\mathbf{y}^{(k+1)} = \mathcal{S}_{\frac{1}{\rho}}(\mathbf{s} - \mathbf{D}\mathbf{x}^{(k+1)} - \frac{\boldsymbol{\eta}^{(k)}}{\rho}) \quad (10)$$

$$\mathbf{z}^{(k+1)} = \mathcal{S}_{\frac{\lambda \Lambda^{-1}}{\rho}}(\mathbf{x}^{(k+1)} - \frac{\Lambda^{-1} \boldsymbol{\gamma}^{(k)}}{\rho}) \quad (11)$$

$$\boldsymbol{\gamma}^{(k+1)} = \boldsymbol{\gamma}^{(k)} + \rho \Lambda (\mathbf{z}^{(k+1)} - \mathbf{x}^{(k+1)}) \quad (12)$$

$$\boldsymbol{\eta}^{(k+1)} = \boldsymbol{\eta}^{(k)} + \rho (\mathbf{y}^{(k+1)} - \mathbf{s} + \mathbf{D}\mathbf{x}^{(k+1)}) \quad (13)$$

For mask decoupling, only one change is needed:

$$\mathbf{y}^{(k+1)} = \begin{cases} \mathbf{S}_{\perp} \left(\mathbf{s} - \mathbf{D}\mathbf{x}^{(k+1)} - \frac{\boldsymbol{\eta}^{(k)}}{\rho} \right) & \text{inside signal domain} \\ \mathbf{s} - \mathbf{D}\mathbf{x}^{(k+1)} - \frac{\boldsymbol{\eta}^{(k)}}{\rho} & \text{outside signal domain} \end{cases} \quad (14)$$