**Nuclear-norm Regularized Model** Let  $D \in \mathbb{R}^{n \times t}$  be the reshaped video with n pixels and t frames. Assume that  $\Phi_s \in \mathbb{R}^{n \times n}$  and  $\Phi_t \in \mathbb{R}^{t \times t}$  are graph Laplacians in the spatial and temporal domains, respectively. Now we consider a foreground-background separation model of the form

$$\min_{L,S \in \mathbb{R}^{n \times t}} \frac{1}{2} \|D - L - S\|_F^2 + \lambda_1 \|L\|_* + \lambda_2 \|S\|_1 + \frac{\gamma_1}{2} \operatorname{tr}(L^T \Phi_s L) + \frac{\gamma_2}{2} \operatorname{tr}(L \Phi_t L^T).$$

By introducing an auxiliary variable U, we rewrite the above problem

$$\min_{L,S,U} \frac{1}{2} \|D - L - S\|_F^2 + \lambda_1 \|U\|_* + \lambda_2 \|S\|_1 + \frac{\gamma_1}{2} \operatorname{tr}(L^T \Phi_s L) + \frac{\gamma_2}{2} \operatorname{tr}(L \Phi_t L^T), \quad \text{s.t.} \quad U = L.$$

Define the augmented Lagrangian

$$\mathcal{L}(L, S, U) = \frac{1}{2} \|D - L - S\|_F^2 + \lambda_1 \|U\|_* + \lambda_2 \|S\|_1 + \frac{\gamma_1}{2} \operatorname{tr}(L^T \Phi_s L) + \frac{\gamma_2}{2} \operatorname{tr}(L \Phi_t L^T) + \frac{\rho}{2} \|U - L + \widetilde{L}\|_F^2.$$

Based on the alternating direction method of multipliers (ADMM), we derive the algorithm

$$\begin{cases} L \leftarrow \underset{L}{\operatorname{argmin}} \frac{1}{2} \|D - L - S\|_F^2 + \frac{\gamma_1}{2} \operatorname{tr}(L^T \Phi_s L) + \frac{\gamma_2}{2} \operatorname{tr}(L \Phi_t L^T) + \frac{\rho}{2} \left\| U - L + \tilde{L} \right\|_F^2 \\ S \leftarrow \underset{S}{\operatorname{argmin}} \lambda_2 \|S\|_1 + \frac{1}{2} \|D - L - S\|_F^2 \\ U \leftarrow \underset{U}{\operatorname{argmin}} \lambda_1 \|U\|_* + \frac{\rho}{2} \left\| U - L + \tilde{L} \right\|_F^2 = \underset{U}{\operatorname{argmin}} \frac{\lambda_1}{\rho} \|U\|_* + \frac{1}{2} \left\| U - L + \tilde{L} \right\|_F^2 \\ \tilde{L} \leftarrow \tilde{L} + (U - L) \end{cases}$$

The first L-subproblem is a least-square problem with the normal equation

$$L + S - D + \gamma_1 \Phi_s L + \gamma_2 L \Phi_t + \rho (L - U - \widetilde{L}) = 0$$

Note that  $\frac{d}{dX}\operatorname{tr}(X^TAX)=(A+A^T)X$ ,  $\frac{d}{dX}\operatorname{tr}(XAX^T)=X(A+A^T)$  and both  $\Phi_t$  and  $\Phi_s$  are symmetric. The above equation can be rewritten in the Sylvester equation form

$$(I + \gamma_1 \Phi_s + \gamma_2 I + \rho I)L + L(\gamma_2 \Phi_t) = D - S + \rho (U + \widetilde{L}).$$

This can give us a unique solution under certain condition via scipy.linalg.solve\_sylvester(A,B,Q) for solving the standard Sylvester equation of the form AX + XB = Q

$$L = sylvester\_solver(A, B, Q),$$

where  $A = (1 + \gamma_2 + \rho)I + \gamma_1\Phi_s$ ,  $B = \gamma_2\Phi_t$ , and  $Q = D - S + \rho(U + \widetilde{L})$ . The S-subproblem has the closed-form solution

$$S \leftarrow \text{shrink}(D - L, \lambda_2).$$

Here the shrinkage operator is defined as  $(\operatorname{shrink}(A,\mu))_{ij} = \operatorname{sign}(a_{ij}) \cdot \max\{|a_{ij}| - \mu, 0\}$  where  $a_{ij}$  is the (i,j)-th entry of A. The U-subproblem has the closed-form solution via the singular value thresholding operator (SVT)

$$U \leftarrow A\widetilde{D}B, \quad \widetilde{D} = \operatorname{shrink}(D, \lambda_1/\rho)$$

where ADB is the SVD of the matrix  $(L - \widetilde{L})$ .