Gradient Computation

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This document contains the gradient formula for the linear and nonlinear lasso problem.

1 Gradient computation

Define two points in spherical coordinates as follows:

$$\mathbf{Q} = (1, \theta, \psi), \ \mathbf{P_i} = (r_i, \theta_i, \psi_i). \tag{1}$$

The normalized forward solution reads:

$$\hat{\Phi}(\mu)(\mathbf{Q}) = \sum_{i=1}^{N} \frac{I_i}{G_{\mathbf{P}_i}} \phi_{\mathbf{P}_i}(\mathbf{Q}), \ \mathbf{Q} \in \partial\Omega, \ where,$$
(2)

$$G_{\mathbf{P_i}} = \sqrt{\int_{\partial\Omega} |F(\mathbf{Q}, \mathbf{P_i})|^2 d\mathbf{Q}} = \sqrt{\int_{\partial\Omega} \phi_{\mathbf{P_i}}^2(\mathbf{Q}) d\mathbf{Q}}$$
(3)

where,

$$\phi_{\boldsymbol{P_i}}(\boldsymbol{Q}) = \frac{2}{||\boldsymbol{Q} - \boldsymbol{P_i}||_2} - \log(1 - \langle \boldsymbol{P_i}, \boldsymbol{Q} \rangle + ||\boldsymbol{Q} - \boldsymbol{P_i}||_2)$$

$$= \frac{2}{\sqrt{1 + r_i^2 - 2r_i \cos \gamma_i}} - \log(1 - r_i \cos \gamma_i + \sqrt{1 + r_i^2 - 2r_i \cos \gamma_i}), \text{ where,}$$

$$\cos \gamma_i = \cos \theta \cos \theta_i + \sin \theta \sin \theta_i \cos(\psi - \psi_i), \ \langle \boldsymbol{P_i}, \boldsymbol{Q} \rangle = r_i \cos \gamma_i, \ ||\boldsymbol{Q} - \boldsymbol{P_i}||_2 = \sqrt{1 + r_i^2 - 2r_i \cos \gamma_i}.$$
(4)

Define the source parameter vector as follows:

$$\mathbf{V} = \{ \overbrace{I_1, I_2, \cdots, r_1, r_2, \cdots, \theta_1, \theta_2, \cdots, \psi_1, \psi_2, \cdots}^{N} \},$$
 (5)

$$I_{all} = \{I_1, I_2, \cdots, I_N\}. \tag{6}$$

And in this report, we use the symbol D^i to represent the partial derivative with respect to the ith component of the unknown parameter vector.

1.1 Linear lasso problem

For the linear lasso problem, locations of the source (i.e., P_i) are given, we define the forward estimation as follows:

$$Est(\mathbf{I_{all}}, \mathbf{Q}) = \sum_{i=1}^{N} \frac{I_i}{G_{\mathbf{P_i}}} \phi_{\mathbf{P_i}}(\mathbf{Q}), \ \mathbf{Q} \in \partial\Omega,$$
 (7)

The objective function reads:

$$Ojb = \frac{1}{2}||Est(\boldsymbol{I_{all}}, \boldsymbol{Q}) - \phi^{d}(\boldsymbol{Q})||_{2}^{2} + \lambda||\boldsymbol{I_{all}}||_{1},$$
(8)

$$= \int_{\partial\Omega} [Est(\mathbf{I}_{all}, \mathbf{Q}) - \phi^d(\mathbf{Q})]^2 d\mathbf{Q} + \lambda \sum_i |I_i|.$$
(9)

For convenience, we introduce the following two terms:

$$f_{des}(\boldsymbol{I_{all}}) := \frac{1}{2} ||Est(\boldsymbol{I_{all}}, \boldsymbol{Q}) - \phi^{d}(\boldsymbol{Q})||_{2}^{2} = \frac{1}{2} \int_{\partial \Omega} [Est(\boldsymbol{I_{all}}, \boldsymbol{Q}) - \phi^{d}(\boldsymbol{Q})]^{2} d\boldsymbol{Q}, \tag{10}$$

$$f_{reg} := \lambda || \mathbf{I}_{all} ||_1. \tag{11}$$

1.2 Subgradient of the regularization term

$$D^{i}f_{reg} = \lambda \ sign(I_{i}). \tag{12}$$

FBS

For the FBS solver, only the gradient of the discrepancy term is required, and locations of the source are given. Therefore we only compute:

$$Grad_{i} = D^{i} f_{des}(\mathbf{Q}) = \int_{\partial\Omega} (Est(\mathbf{I}_{all}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})) D^{i} Est(\mathbf{I}_{all}, \mathbf{Q}) d\mathbf{Q},$$

$$= \int_{\partial\Omega} (Est(\mathbf{I}_{all}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})) \frac{1}{G_{\mathbf{P}_{i}}} \phi_{\mathbf{P}_{i}}(\mathbf{Q}) d\mathbf{Q},$$

$$i = 1, 2, \dots, N.$$
(13)

In the minimization iteration, source location P_i is provided by the solver in each search step.

Quasi Newton

For quasi newton solver, gradient of the whole objective function needs to be computed.

$$PseudoGrad_{i} = \int_{\partial\Omega} (Est(\mathbf{I}_{all}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})) \frac{1}{G_{\mathbf{P}_{i}}} \phi_{\mathbf{P}_{i}}(\mathbf{Q}) d\mathbf{Q} + \lambda \ sign(I_{i}). \tag{14}$$

The Lipschitz constant of the discrepancy gradient

Proposition 1.1. Let $I_{all,1}$, $I_{all,2}$ be two intensity solution vectors and $I_{i,1}$, $I_{i,2}$ be their components, and let:

$$||x|| := \sqrt{x_1^2 + x_2^2 + \dots + x_N^2}, \text{ for any } x \in \mathbb{R}^N.$$

Define a vector as follows:

$$\mathbf{A}(\mathbf{Q}) := \{ D^1 Est(\mathbf{I}_{all}, \mathbf{Q}), D^2 Est(\mathbf{I}_{all}, \mathbf{Q}), \cdots, D^N Est(\mathbf{I}_{all}, \mathbf{Q}) \}$$
$$= \{ \frac{1}{G_{\mathbf{P_1}}} \phi_{\mathbf{P_1}}(\mathbf{Q}), \frac{1}{G_{\mathbf{P_2}}} \phi_{\mathbf{P_2}}(\mathbf{Q}), \cdots, \frac{1}{G_{\mathbf{P_N}}} \phi_{\mathbf{P_N}}(\mathbf{Q}) \}.$$

, then the gradient $\nabla f_{des}(\mathbf{I_{all}})$ has a Lipschitz constant as follows:

$$L = |\sum_{i} \int_{\partial \Omega} ||A(\mathbf{Q})|| |A_{i}(\mathbf{Q}) d\mathbf{Q}|.$$
 (15)

Proof. Notice that:

$$||\nabla f_{des}(\boldsymbol{I_{all,1}}) - \nabla f_{des}(\boldsymbol{I_{all,2}})|| = ||\int_{\partial\Omega} [Est(\boldsymbol{I_{all,1}}, \boldsymbol{Q}) - Est(\boldsymbol{I_{all,2}}, \boldsymbol{Q})] \cdot \nabla Est(\boldsymbol{I_{all}}, \boldsymbol{Q}) d\boldsymbol{Q} ||$$

$$= ||\int_{\partial\Omega} \langle \boldsymbol{A}(\boldsymbol{Q}), \boldsymbol{I_{all,1}} - \boldsymbol{I_{all,2}} \rangle \boldsymbol{A}(\boldsymbol{Q}) d\boldsymbol{Q} ||, \qquad (16)$$

Using the Cauchy Schwarz inequality to the vector 12 norm, we get:

$$\int_{\partial \Omega} \langle \boldsymbol{A}(\boldsymbol{Q}), \boldsymbol{I}_{all,1} - \boldsymbol{I}_{all,2} \rangle \boldsymbol{A}(\boldsymbol{Q}) d\boldsymbol{Q} \le \int_{\partial \Omega} ||\boldsymbol{A}(\boldsymbol{Q})|| ||\boldsymbol{I}_{all,1} - \boldsymbol{I}_{all,2}||\boldsymbol{A}(\boldsymbol{Q}) d\boldsymbol{Q}.$$
(17)

Return to Equation (16), we derive the following:

$$||\nabla f_{des}(\boldsymbol{I_{all,1}}) - \nabla f_{des}(\boldsymbol{I_{all,2}})|| = ||\int_{\partial\Omega} \langle \boldsymbol{A}(\boldsymbol{Q}), \boldsymbol{I_{all,1}} - \boldsymbol{I_{all,2}} \rangle \boldsymbol{A}(\boldsymbol{Q}) d\boldsymbol{Q}||$$
(18)

$$\leq \sqrt{\sum_{i} \left[||\boldsymbol{I}_{all,1} - \boldsymbol{I}_{all,2}|| \int_{\partial \Omega} ||\boldsymbol{A}(\boldsymbol{Q})|| |A_{i}(\boldsymbol{Q}) d\boldsymbol{Q}| \right]^{2}}$$
 (19)

$$= |\sum_{i} \int_{\partial \Omega} ||A(\mathbf{Q})|| |A_{i}(\mathbf{Q}) d\mathbf{Q}| ||I_{all,1} - I_{all,2}||$$
(20)

therefore, we obtain an estimation of the Lipschitz constant as follows,

$$L = |\sum_{i} \int_{\partial \Omega} ||A(\mathbf{Q})|| |A_{i}(\mathbf{Q}) d\mathbf{Q}|.$$
(21)

1.3 Nonlinear lasso problem

For the nonlinear lasso problem, both locations and intensities are unknown, the forward estimation reads:

$$Est(\boldsymbol{V}, \boldsymbol{Q}) = \sum_{i=1}^{N} \frac{I_i}{G_{\boldsymbol{P_i}}} \phi_{\boldsymbol{P_i}}(\boldsymbol{Q}), \ \boldsymbol{Q} \in \partial\Omega,$$
(22)

The objective function reads:

$$Ojb = \frac{1}{2} ||Est(\mathbf{V}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})||_{2}^{2} + \lambda ||I_{all}||_{1},$$
(23)

$$= \int_{\partial\Omega} [Est(\boldsymbol{V}, \boldsymbol{Q}) - \phi^d(\boldsymbol{Q})]^2 d\boldsymbol{Q} + \lambda \sum_i |I_i|.$$
 (24)

The pseudo gradient of the nonlinear lasso problem consists of two component: the gradient of the discrepancy term and the subgradient of the l1 regularization term.

The discrepancy gradient

In this section we compute the gradient term ∇f_{des} .

For
$$i = 1 \rightarrow N$$
,

$$Grad_{i} = D^{i} f_{des}(\mathbf{Q}) = D^{i} f_{des}(\mathbf{Q}) = \int_{\partial \Omega} (Est(\mathbf{V}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})) \frac{1}{G_{\mathbf{P}_{i}}} \phi_{\mathbf{P}_{i}}(\mathbf{Q}) d\mathbf{Q},$$
(25)

for $i = N + 1 \rightarrow 4N$,

$$Grad_{i} = D^{i} f_{des}(\mathbf{Q}) = D^{i} f_{des}(\mathbf{Q}) = \int_{\partial \Omega} (Est(\mathbf{V}, \mathbf{Q}) - \phi^{d}(\mathbf{Q})) D^{i} Est(\mathbf{V}, \mathbf{Q}) d\mathbf{Q},$$
(26)

$$D^{i}Est(\boldsymbol{V},\boldsymbol{Q}) = I_{i} \frac{G_{\boldsymbol{P}_{i}}D^{i}\phi_{\boldsymbol{P}_{i}} - \phi_{\boldsymbol{P}_{i}}D^{i}G_{\boldsymbol{P}_{i}}}{G_{\boldsymbol{P}_{i}}^{2}},$$
(27)

Recall that:

$$\phi_{P_i}(Q) = \frac{2}{||Q - P_i||_2} - \log(1 - \langle P_i, Q \rangle + ||Q - P_i||_2),$$

$$G_{\boldsymbol{P_i}} = \sqrt{\int_{\partial\Omega} |F(\boldsymbol{Q}, \boldsymbol{P_i})|^2 d\boldsymbol{Q}} = \sqrt{\int_{\partial\Omega} \phi_{\boldsymbol{P_i}}^2(\boldsymbol{Q}) d\boldsymbol{Q}}.$$

Thus,

$$D^{i}\phi_{\mathbf{P}_{i}} = -\frac{1}{\|\mathbf{Q} - \mathbf{P}_{i}\|_{2}^{3}}D^{i}\|\mathbf{Q} - \mathbf{P}_{i}\|_{2} - \frac{1}{1 - \langle \mathbf{P}_{i}, \mathbf{Q} \rangle + \|\mathbf{Q} - \mathbf{P}_{i}\|_{2}}D^{i}(-\langle \mathbf{P}_{i}, \mathbf{Q} \rangle + \|\mathbf{Q} - \mathbf{P}_{i}\|_{2}). \tag{28}$$

$$D^{i}G_{\mathbf{P}_{i}} = \frac{\int_{\partial\Omega} \phi_{\mathbf{P}_{i}}(\mathbf{Q}) D^{i} \phi_{\mathbf{P}_{i}}(\mathbf{Q}) d\mathbf{Q}}{G_{\mathbf{P}_{i}}}.$$
(29)

Next we compute the following terms:

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2}, \quad D^{i}\langle\mathbf{Q}, \mathbf{P}_{i}\rangle, \text{ where},$$

$$\langle\mathbf{P}_{i}, \mathbf{Q}\rangle = r_{i}\cos\gamma_{i}, \quad ||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \sqrt{1 + r_{i}^{2} - 2r_{i}\cos\gamma_{i}},$$

$$\cos\gamma_{i} = \cos\theta\cos\theta_{i} + \sin\theta\sin\theta_{i}\cos(\psi - \psi_{i}).$$

For $i = N + 1 \rightarrow 2N$:

$$D^i\langle Q, P_i \rangle = \cos \gamma_i,$$
 (30)

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial r_{i}} = \frac{r_{i} - \cos\gamma_{i}}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}},$$
(31)

for $i = 2N + 1 \rightarrow 3N$:

$$D^{i}\langle \mathbf{Q}, \mathbf{P}_{i}\rangle = r_{i}(-\cos\theta\sin\theta_{i} + \sin\theta\cos\theta_{i}\cos(\psi - \psi_{i})), \tag{32}$$

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial\theta_{i}} = \frac{-r_{i}(-\cos\theta\sin\theta_{i} + \sin\theta\cos\theta_{i}\cos(\psi - \psi_{i}))}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}},$$
(33)

for $i = 3N + 1 \to 4N$:

$$D^{i}\langle \mathbf{Q}, \mathbf{P}_{i}\rangle = r_{i}(\cos\theta\cos\theta_{i} - \sin\theta\sin\theta_{i}\sin(\psi - \psi_{i}), \tag{34}$$

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial\psi_{i}} = \frac{-r_{i}(-\sin\theta\sin\theta_{i}\sin(\psi_{i} - \psi))}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}}.$$
(35)

Pseudo Gradient

Whenever the gradient of the whole objective function is required, we compute the discrepancy gradient and add the subgradient of the regularization term to it, i.e.,

$$PsudoGrad_{i} = \begin{cases} Grad_{i} + \lambda * sign(I_{i}), \ i = 1 \to N, \\ Grad_{i}, \ i > N. \end{cases}$$
(36)