Gradient Computation

ss2571

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

1 Setup

Define two points in spherical coordinates:

$$\mathbf{Q} := (1, \theta, \psi) \in \partial\Omega, \ \mathbf{P}_i := (r_i, \theta_i, \psi_i) \in \Omega. \tag{1}$$

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:

$$V = \{\overline{I_1, I_2, \cdots, P_1, P_2, \cdots, P_N}\},\tag{5}$$

$$I = \{I_1, I_2, \cdots, I_N\}.$$
 (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.

2 Linear lasso problem

For the linear lasso problem, locations of the source (for example, P_i) are given, we define the estimation function as follows:

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

Linear forward operator and its adjoint operator

Define a vector function $\mathbf{A}(\mathbf{Q}) \in \mathbb{R}^N$ such that,

$$\mathbf{A}(\mathbf{Q}) = \{A_1(\mathbf{Q}), A_2(\mathbf{Q}), \cdots, A_N(\mathbf{Q})\}, \text{ and,}$$
(8)

$$A_i(\mathbf{Q}) := \frac{\phi_{\mathbf{P}_i}(\mathbf{Q})}{G_{\mathbf{P}_i}} \in L^2(\partial\Omega). \tag{9}$$

We define the linear forward operator as follows:

$$\Phi: \mathbb{R}^N \to L^2(\partial\Omega), \ s.t.,$$

$$\Phi(\mathbf{I}) = \langle \mathbf{A}(\mathbf{Q}), \mathbf{I} \rangle = \sum_{i=1}^{N} I_i A_i(\mathbf{Q}).$$
(10)

Then, the estimation function becomes:

$$Est(\mathbf{I}, \mathbf{Q}) = \Phi(\mathbf{I}) \tag{11}$$

Next we define the adjoint of Φ , denote it as Φ^* :

$$\Phi^*: L^2(\partial\Omega) \to R^N, \ s.t., \tag{12}$$

$$\Phi^*(p(\mathbf{Q})) = \int_{\partial\Omega} \mathbf{A}(\mathbf{Q})p(\mathbf{Q})d\mathbf{Q}, \ p(\mathbf{Q}) \in L^2(\partial\Omega).$$
(13)

The composition map $\Phi^*\Phi$ derives from the following steps, by definition,

$$\Phi^*\Phi: R^N \to R^N. \tag{14}$$

Take arbitrary vector $a \in \mathbb{R}^N$, and apply the map, we get:

$$\Phi^*\Phi(\mathbf{a}) = \int_{\partial\Omega} \mathbf{A}(\mathbf{Q}) \sum_{i=1}^N A_i(\mathbf{Q}) a_i d\mathbf{Q} \in \mathbb{R}^N,$$
(15)

take the kth entry of $\Phi^*\Phi(a)$, we derive:

$$[\Phi^*\Phi(\boldsymbol{a})]_k = \int_{\partial\Omega} A_k(\boldsymbol{Q}) A_i(\boldsymbol{Q}) a_i d\boldsymbol{Q} = \sum_{i=1}^N (\int_{\partial\Omega} A_k(\boldsymbol{Q}) A_i(\boldsymbol{Q}) d\boldsymbol{Q}) a_i = \sum_{i=1}^N M_{k,i} a_i$$
(16)

, from above we derive that,

$$\mathcal{M}_{i,j} := (\Phi^* \Phi)_{i,j} = \int_{\partial \Omega} A_i(\mathbf{Q}) A_j(\mathbf{Q}) d\mathbf{Q}. \tag{17}$$

Therefore,

$$\Phi^*\Phi = \mathcal{M} \in \mathbb{R}^{N,N}. \tag{18}$$

Objective function

The objective function reads:

$$Obj(\mathbf{I}) = \frac{1}{2} ||\Phi(\mathbf{I}) - \phi^d(\mathbf{Q})||_{\partial\Omega}^2 + \lambda ||\mathbf{I}||_1,$$
(19)

$$= \int_{\partial\Omega} [\Phi(\mathbf{I}) - \phi^d(\mathbf{Q})]^2 d\mathbf{Q} + \lambda \sum_i |I_i|.$$
 (20)

For convenience, we introduce the following two terms:

$$f_{des}(\mathbf{I}) := \frac{1}{2} ||\Phi(\mathbf{I}) - \phi^d(\mathbf{Q})||_{\partial\Omega}^2 = \frac{1}{2} \int_{\partial\Omega} [\Phi(\mathbf{I}) - \phi^d(\mathbf{Q})]^2 d\mathbf{Q}, \tag{21}$$

$$f_{reg}(\mathbf{I}) := \lambda ||\mathbf{I}||_1. \tag{22}$$

2.1 The gradient and hessian of f_{des}

Through a direct calculation, we find that,

$$\nabla f_{des}(\mathbf{I}) = \Phi^* \Phi(\mathbf{I}) - \Phi^* (\phi^d) = \mathcal{M} \mathbf{I} - \mathbf{b}, \text{ where, } \mathbf{b} = \Phi^* (\phi^d). \tag{23}$$

Follows from above, we find the hessian of f_{des} ,

$$\mathcal{H}(f_{des}(\mathbf{I})) = \mathcal{M}. \tag{24}$$

2.1.1 The Lipschitz constant of ∇f_{des}

Proposition 2.1. The gradient $\nabla f_{des}(I)$ has a Lipschitz constant as follows:

$$L = ||\mathbf{\Phi}^*\mathbf{\Phi}||_2, \text{ where } ||\cdot||_2 \text{ is the matrix } l2 \text{ norm.}$$
(25)

Proof. Let $I_a, I_b \in \mathbb{R}^N$ be two arbitrary vectors, then,

$$||\nabla f_{des}(\mathbf{I}_a) - \nabla f_{des}(\mathbf{I}_b)||_2 = ||\mathcal{M}(\mathbf{I}_a - \mathbf{I}_b)||_2 \le ||\mathcal{M}||_2 ||\mathbf{I}_a - \mathbf{I}_b||_2.$$
(26)

2.2 The subgradient of f_{reg}

Denote the subgradient of f_{reg} as ∂f_{reg} , we have that,

$$\partial f_{reg}(\mathbf{I}) = \lambda \ sign(\mathbf{I}).$$
 (27)

2.3 Gradient calculation for linear solvers

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:

$$\nabla f_{des}(\mathbf{I}) = \mathcal{M}\mathbf{I} - \Phi^*(\phi^d(\mathbf{Q})), \text{ where,}$$
(28)

$$\mathcal{M}_{i,j} = \int_{\partial\Omega} A_i(\mathbf{Q}) A_j(\mathbf{Q}) d\mathbf{Q}, \text{ and, } \Phi^*(\phi^d(\mathbf{Q})) = \int_{\partial\Omega} \mathbf{A}(\mathbf{Q}) \phi^d(\mathbf{Q}) d\mathbf{Q}.$$
 (29)

Quasi Newton

For quasi newton solver, we compute the gradient of the entire objective function, i.e.,

$$\nabla_{nseudo}Obj(I) = \nabla f_{des} + \partial f_{reg}. \tag{30}$$

Proof.

3 Nonlinear lasso problem

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

$$Est(\boldsymbol{V}, \boldsymbol{Q}) = \sum_{i=1}^{N} \frac{I_i}{G_{\boldsymbol{P_i}}} \phi_{\boldsymbol{P_i}}(\boldsymbol{Q}), \ \boldsymbol{Q} \in \partial\Omega,$$
(31)

The objective function reads:

$$Ojb = \frac{1}{2}||Est(\boldsymbol{V}, \boldsymbol{Q}) - \phi^{d}(\boldsymbol{Q})||_{\partial\Omega}^{2} + \lambda||\boldsymbol{I}||_{1},$$
(32)

$$= \frac{1}{2} \int_{\partial \Omega} [Est(\boldsymbol{V}, \boldsymbol{Q}) - \phi^{d}(\boldsymbol{Q})]^{2} d\boldsymbol{Q} + \lambda \sum_{i} |I_{i}|.$$
(33)

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
$$i = 1 \rightarrow N$$
,

$$(\nabla f_{des})_i = D^i f_{des} = \int_{\partial \Omega} (Est(\boldsymbol{V}, \boldsymbol{Q}) - \phi^d(\boldsymbol{Q})) \frac{1}{G_{\boldsymbol{P_i}}} \phi_{\boldsymbol{P_i}}(\boldsymbol{Q}) d\boldsymbol{Q}, \tag{34}$$

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

$$(\nabla f_{des})_i = D^i f_{des} = \int_{\partial \Omega} (Est(\boldsymbol{V}, \boldsymbol{Q}) - \phi^d(\boldsymbol{Q})) D^i Est(\boldsymbol{V}, \boldsymbol{Q}) d\boldsymbol{Q},$$
(35)

$$D^{i}Est(\boldsymbol{V},\boldsymbol{Q}) = I_{i}\frac{G_{\boldsymbol{P}_{i}}D^{i}\phi_{\boldsymbol{P}_{i}} - \phi_{\boldsymbol{P}_{i}}D^{i}G_{\boldsymbol{P}_{i}}}{G_{\boldsymbol{P}_{i}}^{2}},$$
(36)

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.

$$D^{i}\phi_{P_{i}} = -\frac{2}{\|Q - P_{i}\|_{2}^{2}}D^{i}\|Q - P_{i}\|_{2} - \frac{1}{1 - \langle P_{i}, Q \rangle + \|Q - P_{i}\|_{2}}D^{i}(-\langle P_{i}, Q \rangle + \|Q - P_{i}\|_{2}).$$
(37)

$$D^{i}G_{\mathbf{P}_{i}} = \frac{\int_{\partial\Omega} \phi_{\mathbf{P}_{i}}(\mathbf{Q})D^{i}\phi_{\mathbf{P}_{i}}(\mathbf{Q})d\mathbf{Q}}{G_{\mathbf{P}_{i}}}.$$
(38)

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
$$i = N + 1: 3: 4N - 2$$
, $V_i = r_i$, $D^i = \frac{\partial}{\partial r_i}$ therefore:

$$D^i\langle Q, P_i\rangle = \cos\gamma_i,$$
 (39)

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial r_{i}} = \frac{r_{i} - \cos\gamma_{i}}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}},$$
(40)

for i = N + 2: 3: 4N - 1, $V_i = \theta_i$, $D^i = \frac{\partial}{\partial \theta_i}$ therefore:

$$D^{i}\langle \mathbf{Q}, \mathbf{P}_{i}\rangle = r_{i}(-\cos\theta\sin\theta_{i} + \sin\theta\cos\theta_{i}\cos(\psi - \psi_{i})), \tag{41}$$

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial\theta_{i}} = \frac{-r_{i}(-\cos\theta\sin\theta_{i} + \sin\theta\cos\theta_{i}\cos(\psi - \psi_{i}))}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}},$$
(42)

for i = N + 3: 3: 4N, $V_i = \psi_i$, $D^i = \frac{\partial}{\partial \psi_i}$ therefore:

$$D^{i}\langle \mathbf{Q}, \mathbf{P}_{i}\rangle = r_{i}(-\sin\theta\sin\theta_{i}\sin(\psi_{i} - \psi)), \tag{43}$$

$$D^{i}||\mathbf{Q} - \mathbf{P}_{i}||_{2} = \frac{\partial||\mathbf{Q} - \mathbf{P}_{i}||_{2}}{\partial\psi_{i}} = \frac{-r_{i}(-\sin\theta\sin\theta_{i}\sin(\psi_{i} - \psi))}{||\mathbf{Q} - \mathbf{P}_{i}||_{2}}.$$
(44)

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.,

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

$$(45)$$