

# Full waveform inversion of isotropic elastic media using an adjoint state method in finite difference time-domain

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## Abstract

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## 1. Introduction

## 2. Theory

### 2.1 Adjoint method for general discretized linear systems

This section derives in simple steps how to obtain the adjoint equations and gradients given a certain system of equations. These are cast in a general form usually present in acoustic, elastic or electromagnetic FWI applications. The proof of validity of this method is presented in ??, in which we derived the method for an more general set of equations based on the notes of Bradley[1].

#### 2.1.1 Formulation

Say we have a model representing the earth (in seismic applications) or tissue (in medical applications) for some parameter like density  $\rho$  or speed of sound  $v$ , which we discretized in a grid in space into  $N_g (= N_x \times N_z$  in 2D) points. If we want to model  $d_m$  parameters over this grid we can write the values of these parameters over the grid as a vector  $\vec{m}$  of dimension  $d_m \times N_g$ . We want to model  $d_s$  fields also discretized in space on the same grid we can write them as a vector  $\vec{s}$  of dimension  $d_s \times N_g$ . useful simplification for already (only in space!)

discretized wave equations (e.g. electromagnetic, elastic or acoustic), whether written as a set of multiple first or second order PDE's in time and space, is the following form

$$\begin{aligned} \min_{\vec{m}} \chi(\vec{s}, \vec{m}) &= \int_0^T f(\vec{s}, \vec{m}) dt \\ \text{subject to } T(\vec{m}) \ddot{\vec{s}} - C(\vec{m}) \dot{\vec{s}} - A(\vec{m}) \vec{s} - b(\vec{m}) &= 0 \\ \text{with B.C. } \vec{s}(0) &= 0 \\ \dot{\vec{s}}(0) &= 0 \end{aligned} \quad (1)$$

where the  $T, C, A$  are big matrices of coefficients which follow from the discretization in space of a PDE system and  $b(\vec{m})$  contains the source terms.

### 2.1.2 Adjoint formulation

Using ?? and applying it to our system of ?? we obtain

$$T^T(\vec{m}) \ddot{\vec{\lambda}} = -C^T(\vec{m}) \dot{\vec{\lambda}} + A^T(\vec{m}) \vec{\lambda} - \frac{\partial f}{\partial \vec{s}} \quad (2)$$

### 2.1.3 Gradients

We can apply ?? to the system of ?? and find the following expression for the total gradient with respect to *all* model parameters

$$\frac{d\chi}{d\vec{m}} = \int_0^T \frac{\partial f}{\partial \vec{m}} + \vec{\lambda}^T \left( \frac{\partial T}{\partial \vec{m}} \ddot{\vec{s}} - \frac{\partial C}{\partial \vec{m}} \dot{\vec{s}} - \frac{\partial A}{\partial \vec{m}} \vec{s} - \frac{\partial b}{\partial \vec{m}} \right) dt \quad (3)$$

Note that the gradients  $\frac{\partial T}{\partial \vec{m}}, \frac{\partial C}{\partial \vec{m}}, \frac{\partial A}{\partial \vec{m}}$  can be seen as a sum of all the  $3N$  derivatives of each component of the matrices  $T(\vec{m}), C(\vec{m}), A(\vec{m})$  to the discrete variables  $\vec{m}_i$ , but in practice we can write this into simple compact expressions as shown in the examples below.

### 2.1.4 Least squares cost function

In most cases the cost function is given by the misfit between observed data  $d$  at certain receiver positions  $y_R$  and the modeled data at the same positions  $s_{y_R}$ , so  $f(\vec{s})$  is given by

$$f(\vec{s}) = \frac{1}{2} \sum_R^{N_R} ||s_{y_R} - d_{y_R}||^2 \quad (4)$$

where the sum goes over all receiver positions. Note that  $f(\vec{s})$  does not depend directly on the model parameters  $\vec{m}$  making the gradient expression simpler since  $\frac{\partial f}{\partial \vec{m}} = 0$ . We thus obtain

$$\frac{\partial f}{\partial \vec{s}} = \sum_R^{N_R} ||s_{y_R} - d_{y_R}|| \quad (5)$$

## 2.2 Adjoint and gradients of the elastic wave equation

Elastic waves in 2D isotropic media can be modeled by using only three parameters, the density  $\rho$  and the two Lamé parameters  $\mu, \lambda$  and follow the following system of equations

$$\begin{aligned}
 \rho \frac{\partial v_x}{\partial t} &= \frac{\partial \tau_{xx}}{\partial x} + \frac{\partial \tau_{xz}}{\partial z} \\
 \rho \frac{\partial v_z}{\partial t} &= \frac{\partial \tau_{xz}}{\partial x} + \frac{\partial \tau_{zz}}{\partial z} \\
 \frac{\partial \tau_{xx}}{\partial t} &= (\lambda + 2\mu) \frac{\partial v_x}{\partial x} + \lambda \frac{\partial v_z}{\partial z} + b_{xx} \\
 \frac{\partial \tau_{zz}}{\partial t} &= (\lambda + 2\mu) \frac{\partial v_z}{\partial z} + \lambda \frac{\partial v_x}{\partial x} + b_{zz} \\
 \frac{\partial \tau_{xz}}{\partial t} &= \mu \left( \frac{\partial v_x}{\partial z} + \frac{\partial v_z}{\partial x} \right)
 \end{aligned} \tag{6}$$

where we only added the source to the stress as to resemble an earthquake. The form of  $b$  is independent of the model parameters and usually taken as a ricker wavelet. writing this in the form of ?? gives  $T(m) = 0$  and

$$s = \begin{bmatrix} v_x \\ v_z \\ \tau_{xx} \\ \tau_{xz} \\ \tau_{zx} \end{bmatrix}, \quad C(m) = - \begin{bmatrix} \vec{\rho} & 0 & 0 & 0 & 0 \\ 0 & \vec{\rho} & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \tag{7}$$

and

$$A(m) = \begin{bmatrix} 0 & 0 & \mathcal{D}_x & 0 & \mathcal{D}_z \\ 0 & 0 & 0 & \mathcal{D}_z & \mathcal{D}_x \\ (\lambda + 2\mu) \mathcal{D}_x & \lambda \mathcal{D}_z & 0 & 0 & 0 \\ \lambda \mathcal{D}_x & (\lambda + 2\mu) \mathcal{D}_z & 0 & 0 & 0 \\ \mu \mathcal{D}_z & \mu \mathcal{D}_x & 0 & 0 & 0 \end{bmatrix} \tag{8}$$

Thus we have  $C^T(m) = C(m)$  and

$$A^T(m) = \begin{bmatrix} 0 & 0 & -\mathcal{D}_x(\lambda + 2\mu) & -\mathcal{D}_x\lambda & -\mathcal{D}_z\mu \\ 0 & 0 & -\mathcal{D}_z\lambda & -\mathcal{D}_z(\lambda + 2\mu) & -\mathcal{D}_x\mu \\ -\mathcal{D}_x & 0 & 0 & 0 & 0 \\ 0 & -\mathcal{D}_z & 0 & 0 & 0 \\ -\mathcal{D}_z & -\mathcal{D}_x & 0 & 0 & 0 \end{bmatrix} \tag{9}$$

which in turn (when transforming back to the continuous domain from the discrete variables) results in the following adjoint equations (naming the adjoint variable  $l$ ) as defined in ??

$$\begin{aligned}
 \rho \frac{\partial l_1}{\partial t} &= \frac{\partial ((\lambda + 2\mu) l_3)}{\partial x} + \frac{\partial (\lambda l_4)}{\partial x} + \frac{\partial (\mu \lambda_5)}{\partial z} \\
 \rho \frac{\partial l_2}{\partial t} &= \frac{\partial (\lambda l_3)}{\partial z} + \frac{\partial ((\lambda + 2\mu) l_4)}{\partial z} + \frac{\partial (\mu \lambda_5)}{\partial x} \\
 \frac{\partial l_3}{\partial t} &= \frac{\partial l_1}{\partial x} \\
 \frac{\partial l_4}{\partial t} &= \frac{\partial l_2}{\partial z} \\
 \frac{\partial l_5}{\partial t} &= \frac{\partial l_1}{\partial z} + \frac{\partial l_2}{\partial x}
 \end{aligned} \tag{10}$$

where the order of differentiation and multiplication with the model parameters is important. The gradients then follow from ??, giving

$$\begin{aligned}
 \frac{d\chi}{d\rho} &= \int_0^T -l^T \frac{\partial C}{\partial \rho} s \, dt \\
 &= \int_0^T \begin{bmatrix} \vec{l}_1 & \vec{l}_2 & \vec{l}_3 & \vec{l}_4 & \vec{l}_5 \end{bmatrix} \begin{bmatrix} I & 0 & 0 & 0 & 0 \\ 0 & I & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \vec{v}_x \\ \vec{v}_y \\ \vec{\tau}_{xx} \\ \vec{\tau}_{zz} \\ \vec{\tau}_{xz} \end{bmatrix} \\
 &= \int_0^T \vec{l}_1 \vec{v}_x + \vec{l}_2 \vec{v}_z \, dt
 \end{aligned} \tag{11}$$

,

$$\begin{aligned}
 \frac{d\chi}{d\lambda} &= \int_0^T -l^T \frac{\partial A}{\partial \lambda} s \, dt \\
 &= - \int_0^T \begin{bmatrix} \vec{l}_1 & \vec{l}_2 & \vec{l}_3 & \vec{l}_4 & \vec{l}_5 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ \mathcal{D}_x & \mathcal{D}_z & 0 & 0 & 0 \\ \mathcal{D}_x & \mathcal{D}_z & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \vec{v}_x \\ \vec{v}_y \\ \vec{\tau}_{xx} \\ \vec{\tau}_{zz} \\ \vec{\tau}_{xz} \end{bmatrix} \\
 &= - \int_0^T (\vec{l}_3 + \vec{l}_4) \left( \frac{\partial \vec{v}_x}{\partial x} + \frac{\partial \vec{v}_z}{\partial z} \right)
 \end{aligned} \tag{12}$$

and

$$\begin{aligned}
 \frac{d\chi}{d\mu} &= \int_0^T -l^T \frac{\partial A}{\partial \mu} s \, dt \\
 &= - \int_0^T \begin{bmatrix} \vec{l}_1 & \vec{l}_2 & \vec{l}_3 & \vec{l}_4 & \vec{l}_5 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 2\mathcal{D}_x & 0 & 0 & 0 & 0 \\ 0 & 2\mathcal{D}_z & 0 & 0 & 0 \\ \mathcal{D}_z & \mathcal{D}_x & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \vec{v}_x \\ \vec{v}_y \\ \vec{\tau}_{xx} \\ \vec{\tau}_{zz} \\ \vec{\tau}_{xz} \end{bmatrix} \, dt \\
 &= - \int_0^T 2\vec{l}_3 \frac{\partial \vec{v}_x}{\partial x} + 2\vec{l}_4 \frac{\partial \vec{v}_z}{\partial z} + \vec{l}_5 \left( \frac{\partial \vec{v}_x}{\partial z} + \frac{\partial \vec{v}_z}{\partial x} \right) \, dt
 \end{aligned} \tag{13}$$

### 3. Numerical implementation

### 4. Simulation results

### 5. Conclusion

### Appendices

#### 1. Lagrangian based derivation of the gradient via the adjoint state method

In this section the adjoint equations and the gradients for a general first or second order PDE system will be derived following the approach outlined in [1].

##### A.1 General system of equations

We can write the most general minimization problem of a cost function subject any first order or second order (in time) set of partial differential equations(PDE) in the following form

$$\begin{aligned}
 \min_m \quad & \chi(s, m) = \int_0^T f(s, m) \, dt \\
 \text{subject to} \quad & h(\ddot{s}, \dot{s}, s, m, t) = 0 \\
 \text{with B.C} \quad & g((s(0), m) = 0 \\
 & k((\dot{s}(0), m) = 0
 \end{aligned} \tag{14}$$

where  $s$  is a discretized vector of the fields and the dots denote derivatives to time,  $m$  is the vector of all the discretized model parameters,  $t$  denotes time,  $h(\ddot{s}, \dot{s}, s, m, t) = 0$  is the system of equations, e.g the wave equation in 2D or 3D,  $T$  is the final time of integration and  $g((s(0), m)$ ,  $k((\dot{s}(0), m)$  denote initial conditions for the field vector  $s$  and its derivative  $\dot{s}$ . Note that this is similar to [1] but with the added explicit dependence of both  $\dot{s}$  and  $\ddot{s}$  giving a more general expression for the adjoint equations and the gradients.

## A.2 Derivative to model parameters

When solving the minimization problem numerically, one often uses a method similar to Newton's gradient descend. This thus requires knowledge of the derivative of the cost function  $\chi(s, m)$  to all of the model parameters  $m$ . Using the chain rule we obtain:

$$\frac{d\chi}{dm} = \frac{\partial\chi}{\partial s} \frac{\partial s}{\partial m} + \frac{\partial\chi}{\partial m} \quad (15)$$

This depends on the Frechet derivatives  $\frac{\partial s}{\partial m}$  which require at least  $3N$  evaluations of the forward model in order to obtain an estimate, we thus want to avoid the calculation of this term. To facilitate this we define the Lagrangian  $\mathcal{L}$  by

$$\mathcal{L} = \int_0^T [f(s, m) + \lambda^T h(\dot{s}, \ddot{s}, s, m, t)] dt + \mu^T g((s(0), m) + \eta^T k((\dot{s}(0), m) \quad (16)$$

where the auxiliary variables  $\lambda, \mu, \eta$  have the same length as the discretized field vector  $s$ . Note that due to the initial conditions of ?? we have  $\frac{d\chi}{dm} = \frac{d\mathcal{L}}{dm}$ , where using the chain rule repetitively

$$\begin{aligned} \frac{d\mathcal{L}}{dm} = & \int_0^T \left[ \frac{\partial f}{\partial s} \frac{\partial s}{\partial m} + \frac{\partial f}{\partial m} \right. \\ & + \lambda^T \left( \frac{\partial h}{\partial \ddot{s}} \frac{\partial \ddot{s}}{\partial m} + \frac{\partial h}{\partial \dot{s}} \frac{\partial \dot{s}}{\partial m} + \frac{\partial h}{\partial s} \frac{\partial s}{\partial m} + \frac{\partial h}{\partial m} \right) \Big] dt \\ & + \mu^T \left( \frac{\partial g}{\partial s(0)} \frac{\partial s(0)}{\partial m} + \frac{\partial g}{\partial m} \right) \\ & + \eta^T \left( \frac{\partial k}{\partial \dot{s}(0)} \frac{\partial \dot{s}(0)}{\partial m} + \frac{\partial k}{\partial m} \right) \end{aligned} \quad (17)$$

This looks intimidating and not really much simpler, but we are still free to choose the expressions for  $\lambda, \mu$  and  $\eta$ . We will choose this such that we can avoid calculating the computationally difficult derivatives  $\frac{\partial s}{\partial m}$ . But first we need to do some partial integration to get rid of  $\frac{\partial \dot{s}}{\partial m}$  and  $\frac{\partial \ddot{s}}{\partial m}$ . With a single partial integration we can write

$$\int_0^T \lambda^T \frac{\partial h}{\partial \dot{s}} \frac{\partial \dot{s}}{\partial m} = \left[ \lambda^T \frac{\partial h}{\partial \dot{s}} \frac{\partial s}{\partial m} \right]_0^T - \int_0^T \frac{\partial s}{\partial m} \left( \dot{\lambda}^T \frac{\partial h}{\partial \dot{s}} + \lambda^T \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} \right) \quad (18)$$

and with a double partial integration we can write

$$\begin{aligned} \int_0^T \lambda^T \frac{\partial h}{\partial \ddot{s}} \frac{\partial \ddot{s}}{\partial m} = & \left[ \lambda^T \frac{\partial h}{\partial \ddot{s}} \frac{\partial \dot{s}}{\partial m} - \frac{\partial s}{\partial m} \left( \dot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} + \lambda^T \frac{\partial}{\partial t} \frac{\partial h}{\partial \ddot{s}} \right) \right]_0^T \\ & + \int_0^T \frac{\partial s}{\partial m} \left( \ddot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} + 2\dot{\lambda}^T \frac{\partial}{\partial t} \frac{\partial h}{\partial \ddot{s}} + \lambda^T \frac{\partial^2}{\partial t^2} \frac{\partial h}{\partial \ddot{s}} \right) dt \end{aligned} \quad (19)$$

So filling this back into ?? and regrouping terms gives

$$\begin{aligned}
\frac{d\mathcal{L}}{dm} = & \int_0^T \left[ \frac{\partial s}{\partial m} \left( \frac{\partial f}{\partial s} + \ddot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} + \dot{\lambda}^T \left( 2 \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} - \frac{\partial h}{\partial s} \right) \right. \right. \\
& \left. \left. + \lambda^T \left( \frac{\partial h}{\partial s} + \frac{\partial^2}{\partial t^2} \frac{\partial h}{\partial \ddot{s}} - \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} \right) \right) + \frac{\partial f}{\partial m} + \lambda^T \frac{\partial h}{\partial m} \right] dt \\
& + \left( \mu^T \frac{\partial g}{\partial s(0)} - \lambda^T \frac{\partial h}{\partial \dot{s}} + \dot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} + \lambda^T \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} \right) \frac{\partial s}{\partial m} \Big|_0 \\
& + \left( \eta^T \frac{\partial h}{\partial \dot{s}(0)} - \lambda^T \frac{\partial h}{\partial \ddot{s}} \right) \frac{\partial \dot{s}}{\partial m} \Big|_0 \\
& + \lambda^T \frac{\partial h}{\partial \ddot{s}} \frac{\partial \dot{s}}{\partial m} \Big|_T + \left( \lambda^T \frac{\partial h}{\partial \dot{s}} - \dot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} - \lambda^T \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} \right) \frac{\partial s}{\partial m} \Big|_T \\
& + \mu^T \frac{\partial g}{\partial m} + \eta^T \frac{\partial k}{\partial m}
\end{aligned} \tag{20}$$

we can then choose values for the auxiliary variables such that undesired terms drop out of  $\frac{d\mathcal{L}}{dm}$ . If we set

$$\begin{aligned}
\mu^T &= \left( \dot{\lambda}(0) \frac{\partial h}{\partial \dot{s}} + \lambda^T(0) \left( \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} - \frac{\partial h}{\partial s} \right) \right) \left( \frac{\partial g}{\partial s(0)} \right)^{-1} \\
\eta^T &= \left( \lambda(0) \frac{\partial h}{\partial \dot{s}} \right) \left( \frac{\partial k}{\partial \dot{s}(0)} \right)^{-1} \\
\lambda(T) &= 0 \\
\dot{\lambda}(T) &= 0
\end{aligned} \tag{21}$$

and let  $\lambda$  satisfy the following so called *adjoint* equation

$$\ddot{\lambda}^T \frac{\partial h}{\partial \ddot{s}} + \dot{\lambda}^T \left( 2 \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} - \frac{\partial h}{\partial s} \right) + \lambda^T \left( \frac{\partial h}{\partial s} + \frac{\partial^2}{\partial t^2} \frac{\partial h}{\partial \ddot{s}} - \frac{\partial}{\partial t} \frac{\partial h}{\partial \dot{s}} \right) = - \frac{\partial f}{\partial s} \tag{22}$$

we can see that the  $\frac{\partial s}{\partial m}$  terms drop out of  $\frac{d\mathcal{L}}{dm}$  and we remain for the *gradients* only with

$$\frac{d\mathcal{L}}{dm} = \int_0^T \frac{\partial f}{\partial m} + \lambda^T \frac{\partial h}{\partial m} dt + \mu^T \frac{\partial g}{\partial m} + \eta^T \frac{\partial k}{\partial m} \tag{23}$$

where the two rightmost terms are zero if the initial conditions do not depend directly on the model parameters  $m$ . ?? is called the *adjoint* equation to the original system of equations  $h(\ddot{s}, \dot{s}, s, m, t)$



## 2. More example applications

### B.1 The second order electromagnetic wave equation

Now for an example with second order terms as well we take the following second order electromagnetic wave equation

$$\varepsilon \frac{\partial^2 E_y}{\partial t^2} = \frac{\partial}{\partial x} \left( \frac{1}{\mu} \frac{\partial E_y}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\mu} \frac{\partial E_y}{\partial z} \right) - \sigma \frac{\partial E_y}{\partial t} \quad (24)$$

which models the same as ???. We can see that now  $s = \vec{E}_y$  a single field variable ( which is still a vector of length N)

$$T(m) = \varepsilon, C(m) = -\sigma, A(m) = \mathcal{D}_x \frac{1}{\mu} \mathcal{D}_x + \mathcal{D}_z \frac{1}{\mu} \mathcal{D}_z \quad (25)$$

All of these operators are self-adjoint, thus  $A^T(m) = A(m)$ ,  $C^T(m) = C(m)$ ,  $T^T(m) = T(m)$ , we thus get for the adjoint equation

$$\varepsilon \frac{\partial^2 \lambda}{\partial t^2} = \frac{\partial}{\partial x} \left( \frac{1}{\mu} \frac{\partial \lambda}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\mu} \frac{\partial \lambda}{\partial z} \right) + \sigma \frac{\partial \lambda}{\partial t} \quad (26)$$

where there is only a sign change in front of  $\sigma$  due to ???. The gradients then become according to ??

$$\begin{aligned} \frac{d\chi}{d\varepsilon} &= \int_0^T \vec{\lambda}^T \vec{E}_y dt \\ \frac{d\chi}{d\mu} &= \int_0^T \vec{\lambda}^T \left( \frac{\partial}{\partial x} \left( \frac{1}{\mu^2} \frac{\partial \vec{E}_y}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\mu^2} \frac{\partial \vec{E}_y}{\partial z} \right) \right) dt \\ \frac{d\chi}{d\sigma} &= \int_0^T \vec{\lambda}^T \vec{E}_y dt \end{aligned} \quad (27)$$

### B.2 The second order acoustic wave equation

The acoustic wave equation with variable density  $\rho$  and velocity  $c$  is given by

$$\frac{1}{\rho c^2} \frac{\partial^2 p}{\partial t^2} = \left( \frac{\partial}{\partial x} \left( \frac{1}{\rho} \frac{\partial p}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\rho} \frac{\partial p}{\partial z} \right) \right) \quad (28)$$

we thus get

$$T(m) = \frac{1}{\rho c^2}, C(m) = 0, A(m) = \mathcal{D}_x \frac{1}{\rho} \mathcal{D}_x + \mathcal{D}_z \frac{1}{\rho} \mathcal{D}_z \quad (29)$$

which looks very similar to the electromagnetic case, again  $A^T(m) = A(m)$ ,  $C^T(m) = C(m)$ ,  $T^T(m) = T(m)$ , so following ??? the adjoint equation is the same as the original equation

$$\frac{1}{\rho c^2} \frac{\partial^2 \lambda}{\partial t^2} = \left( \frac{\partial}{\partial x} \left( \frac{1}{\rho} \frac{\partial \lambda}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\rho} \frac{\partial \lambda}{\partial z} \right) \right) \quad (30)$$

According to ?? we then obtain

$$\begin{aligned}\frac{d\chi}{dc} &= -\frac{2}{\rho c^3} \int_0^T \vec{\lambda}^T \vec{p} dt \\ \frac{d\chi}{d\rho} &= \int_0^T \vec{\lambda}^T \left( -\frac{1}{\rho^2 c^2} \vec{p} + \frac{\partial}{\partial x} \left( \frac{1}{\rho^2} \frac{\partial \vec{p}}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\rho^2} \frac{\partial \vec{p}}{\partial z} \right) \right) dt\end{aligned}\quad (31)$$

### B.3 The first order acoustic wave equation

The first order acoustic wave equation with variable density  $\rho$  and velocity  $c$  is given by

$$\begin{aligned}\frac{1}{\rho c^2} \frac{\partial p}{\partial t} &= -\frac{\partial v_x}{\partial x} - \frac{\partial v_z}{\partial z} \\ \frac{\partial v_x}{\partial t} &= -\frac{1}{\rho} \frac{\partial p}{\partial x} \\ \frac{\partial v_z}{\partial t} &= -\frac{1}{\rho} \frac{\partial p}{\partial z}\end{aligned}\quad (32)$$

this gives

$$T(m) = 0, \quad C(m) = -\begin{bmatrix} \frac{1}{\rho c^2} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad A(m) = \begin{bmatrix} 0 & -\mathcal{D}_x & -\mathcal{D}_z \\ -\frac{1}{\rho} \mathcal{D}_x & 0 & 0 \\ -\frac{1}{\rho} \mathcal{D}_z & 0 & 0 \end{bmatrix}\quad (33)$$

we thus get  $C^T(m) = C$  and

$$A^T(m) = \begin{bmatrix} 0 & \mathcal{D}_x \frac{1}{\rho} & \mathcal{D}_z \frac{1}{\rho} \\ \mathcal{D}_x & 0 & 0 \\ \mathcal{D}_z & 0 & 0 \end{bmatrix}\quad (34)$$

so the adjoint equations according to ?? become

$$\begin{aligned}\frac{1}{\rho c^2} \frac{\partial \lambda_1}{\partial t} &= -\frac{\partial \left( \frac{1}{\rho} \lambda_2 \right)}{\partial x} - \frac{\partial \left( \frac{1}{\rho} \lambda_3 \right)}{\partial z} \\ \frac{\partial \lambda_2}{\partial t} &= -\frac{\partial \lambda_1}{\partial x} \\ \frac{\partial \lambda_3}{\partial t} &= -\frac{\partial \lambda_1}{\partial z}\end{aligned}\quad (35)$$

and the gradients according to ??

$$\begin{aligned}\frac{d\chi}{dc} &= -\frac{2}{\rho c^3} \int_0^T \vec{\lambda}_1^T \vec{p} dt \\ \frac{d\chi}{d\rho} &= \int_0^T \frac{-1}{\rho^2 c^2} \vec{\lambda}_1^T \vec{p} + \frac{1}{\rho^2} \left( \vec{\lambda}_2^T \frac{\partial \vec{v}_x}{\partial x} + \vec{\lambda}_3^T \frac{\partial \vec{v}_z}{\partial z} \right) dt\end{aligned}\quad (36)$$

## References

- <sup>[1]</sup> Andrew M. Bradley. PDE-constrained optimization problems and the adjoint method. 2012(2):1–6, 2012.