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Sensitivity Analysis : A Variational Approach

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1 / 23

Overview

- General Sensitivity Analysis
- Sensitivity and Data Assimilation
- Second Order Analysis
- A 1-D Example
- An Application to a pollution problem

Sensitivity Analysis: Deterministic Approach

• Model: \mathcal{F} :

$$\mathcal{F}(\mathcal{X}, \mathcal{U}) = 0 \tag{1}$$

• Scalar Response Function G:

$$\mathcal{G}(\mathcal{X}, \mathcal{U})$$
 (2)

ullet Sensitivity ${\mathcal S}$ is by definition the gradient of ${\mathcal G}$ with respect to ${\mathcal U}$:

$$S = \nabla \mathcal{G}(\mathcal{X}(\mathcal{U}), \mathcal{U}) \tag{3}$$

Computing Sensitivity: Variational Method

ullet An adjoint variable ${\mathcal P}$ is introduced as the solution of :

$$\left[\frac{\partial \mathcal{F}}{\partial \mathcal{X}}\right]^{t} . \mathcal{P} = \left[\frac{\partial \mathcal{G}}{\partial \mathcal{X}}\right] \tag{4}$$

• Then we get :

$$S = \left[\frac{\partial \mathcal{G}}{\partial \mathcal{U}}\right] - \left[\frac{\partial \mathcal{F}}{\partial \mathcal{U}}\right]^{t} . \mathcal{P} \tag{5}$$

Data Assimilation for Pollution Modeling

• X is the state variable (velocity, surface elevation) governed by :

$$\begin{cases}
\frac{dX}{dt} = F(X) \\
X(0) = U
\end{cases}$$
(6)

• The concentration of pollutant *C*, produced by sources *S* verifies:

$$\begin{cases}
\frac{dC}{dt} = G(X, C, S) \\
C(0) = V
\end{cases}$$
(7)

• U and V are unknonw. The VDA problem is to evaluate them from observation X_{obs} and C_{obs} , in order to minimize the cost function J defined by:

$$J(U,V) = \frac{1}{2} \int_0^T \|EX - X_{obs}\|^2 dt + \frac{1}{2} \int_0^T \|DC - C_{obs}\|^2 dt$$
 (8)

 For sake of simplicity regularization terms, of great practical importance, are not displayed

Data Assimilation for Pollution Modeling: Optimality System

 P and Q adjoint variables are introduced as the solution of the system :

$$\begin{cases}
\frac{dP}{dt} + \left[\frac{\partial F}{\partial X}\right]^t . P + \left[\frac{\partial G}{\partial X}\right]^t . Q = E^t(EX - X_{obs}) \\
P(T) = 0;
\end{cases}$$
(9)

.;

$$\begin{cases}
\frac{dQ}{dt} + \left[\frac{\partial G}{\partial C}\right]^t . Q = D^t(DC - C_{obs}); \\
Q(T) = 0,
\end{cases} (10)$$

ullet Then the gradient of J with respect to U and V are given by :

$$\nabla J_U = -P(0) \tag{11}$$

$$\nabla J_V = -Q(0) \tag{12}$$

Sensitivity with respect to Observations and Sources

- If some response function S is introduced, how to evaluate the sensitivity with respect to observations? For instance how to evaluate the impact of an error of observation on a prediction?
- ullet What should be the "model" ${\mathcal F}$ of the general sensitivity analysis?
- Because only the Optimality System contains the observation, the sensitivity analysis must be carried out on the O.S. considered as a Generalized Model
- Deriving the O.S. leads to carry out a Second Order Analysis.

Computing the sensitivity with respect to sources : second order adjoint.

• We need to introduce four second order adjoint variables Γ , Λ , Φ and Ψ as the solution of :

$$\begin{cases}
\frac{d\Gamma}{dt} + \left[\frac{\partial F}{\partial X}\right]^{t} \cdot \Gamma + \left[\frac{\partial F}{\partial X}\right]^{t} \cdot \Lambda + \left[\frac{\partial^{2} F}{\partial X^{2}} P\right]^{t} \cdot \Phi \\
+ \left[\frac{\partial^{2} G}{\partial X^{2}} Q\right]^{t} \cdot \Phi + \left[\frac{\partial^{2} G}{\partial C \partial X} Q\right]^{t} \cdot \Psi - E^{t} E \Phi = 0; \quad (13)
\end{cases}$$

$$\Gamma(0) = 0; \quad \Gamma(T) = 0,$$

Computing the sensitivity with respect to sources 2

•

$$\begin{cases}
\frac{d\Lambda}{dt} + \left[\frac{\partial F}{\partial C}\right]^{t} \cdot \Lambda + \left[\frac{\partial^{2} G}{\partial C \partial X} Q\right]^{t} \cdot \Phi \\
+ \left[\frac{\partial^{2} G}{\partial X^{2}} Q\right]^{t} \cdot \Psi - D^{t} D \Psi = \frac{\partial \varphi}{\partial C}; \\
\Lambda(0) = 0; \\
\Lambda(T) = 0,
\end{cases} (14)$$

$$\frac{d\Phi}{dt} + \left[\frac{\partial F}{\partial X}\right]^t \cdot \Phi = 0, \tag{15}$$

$$\frac{d\Psi}{dt} + \left[\frac{\partial G}{\partial C}\right]^t \cdot \Psi = 0, \tag{16}$$

Then it comes :

$$\nabla \varphi = \left[\frac{\partial F}{\partial S} \right]^t \cdot \Lambda + \left[\frac{\partial^2 G}{\partial X^2} Q \right]^t \cdot \Phi + \left[\frac{\partial^2 G}{\partial C \partial S} Q \right]^t \cdot \Psi + \frac{\partial \varphi}{\partial S}$$
 (17)

Remarks

- The sensitivity is obtained by solving the coupled system of four equations
- The System involves second order terms.
- We found a non-standard problem: two equations have two conditions an initial condition and a final condition, the other two equations have no condition

Solving the Non-Standard problem

The Non-Standard problem can be symbolically written :

$$\begin{cases}
\frac{dX}{dt} = K(X,Y), & t \in [0,T]; \\
\frac{dY}{dt} = L(X,Y), & t \in [0,T]
\end{cases}$$
(18)

with:

$$\begin{cases}
X(0) = 0; \\
X(T) = 0
\end{cases}$$
(19)

and no condition on Y.

NSP is transformed into a problem of optimal control by introducing the control U and a cost-function $J_P(U)$ with :

$$\begin{cases} X(0) = 0; \\ Y(0) = U. \end{cases} \tag{20}$$

Solving the Non-Standard problem 2

A cost function $J_P(U)$ is defined by:

$$J_P(U) = \frac{1}{2} ||X(T, U)||^2 + \frac{1}{2} ||U||^2$$
 (21)

If Z and W are defined as the solution of:

$$\frac{dW}{dt} + \left[\frac{\partial K}{\partial X}\right]^t \cdot W + \left[\frac{\partial L}{\partial X}\right]^t \cdot Z = 0; \tag{22}$$

$$\frac{dZ}{dt} + \left[\frac{\partial K}{\partial Y}\right]^t \cdot W + \left[\frac{\partial L}{\partial Y}\right]^t \cdot Z = 0; \tag{23}$$

$$Z(T) = 0; W(T) = X(T),$$
 (24)

then we get

$$\nabla J_P(U) = -Z(0) + U \tag{25}$$

Solving the Non-Standard problem 3

This problem involved third derivatives of the original model. Recent developments on the NSP have been recently carrier out by V. Shutyaev and F.-X. Le Dimet
The existence of a solution is demonstrated
Another method to solve NSP is proposed.

A 1-D Example 1

Let us assume that the one dimensional velocity field u = u(x, t) evolves according to the Burgers equation given by :

$$\begin{cases}
\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} - \nu \frac{\partial^2 u}{\partial x^2} &= f, x \in \Omega =]-1, 1[, t \in [0, T]; \\
u &= u_0, t = 0, \\
u &= u_1, x \in \{-1, 1\},
\end{cases} (26)$$

Evolution of the pollutant's concentration:

$$\begin{cases}
\frac{\partial c}{\partial t} + u \frac{\partial c}{\partial x} &= \eta \frac{\partial^2 c}{\partial x^2} + s, x \in]-1, 1[, t \in [0, T]] \\
c &= c_0, t = 0; \\
c &= c_1, x \in \{-1, 1\}
\end{cases}$$
(27)

A 1-D Example: cost function

The cost function takes the form (with continuous observation in space and time):

$$J(u_0, c_0) = \frac{1}{2} \int_0^T \|u - u_{obs}\|_{\Omega}^2 dt + \frac{1}{2} \int_0^T \|c - c_{obs}\|_{\Omega}^2 dt.$$
 (28)

where
$$||f||_{\Omega}^2 = \int_{\Omega} f(x)f(x)dx = \int_0^1 f(x)f(x)dx$$
.

A 1-D Example: adjoint model

The adjoint variables p and q are introduced as the solutions of :

$$\begin{cases}
\frac{\partial p}{\partial t} + u \frac{\partial p}{\partial x} + \nu \frac{\partial^2 p}{\partial x^2} + q \frac{\partial c}{\partial x} &= u - u_{obs} \\
p(t = T) &= 0 \\
p &= 0, x \in \{-1, 1\}
\end{cases} (29)$$

$$\begin{cases}
\frac{\partial q}{\partial t} + \frac{\partial uq}{\partial x} + \eta \frac{\partial^2 q}{\partial x^2} &= c - c_{obs} \\
q(t = T) &= 0 \\
q &= 0, x \in \{-1, 1\}
\end{cases}$$
(30)

And the gradient of the cost function is given by:

$$\nabla_{u_0} J = -p(0)
\nabla_{c_0} J = -q(0)$$

Sensitivity of a response function

Let φ be a function a the concentration and the source functions, the response function is given by:

$$\Phi_{A}(t,s) = \int_{\Omega_{A}} \varphi(c,s) dx \tag{31}$$

where $\Omega_A\subset\Omega$ is the response region. Following the guidelines of the derivation of the gradient, we introduces the adjoint variables Γ ϕ , ψ and Λ as the solution of:

Sensitivity of a response function

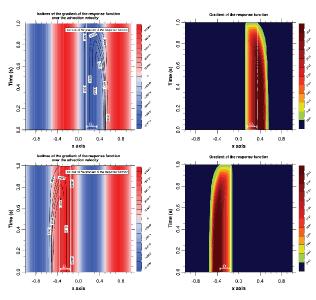
$$\begin{cases}
\frac{\partial \Gamma}{\partial t} + u \frac{\partial \Gamma}{\partial x} + \nu \frac{\partial^2 \Gamma}{\partial x^2} - \Lambda \frac{\partial c}{\partial x} \\
-\phi \frac{\partial p}{\partial x} + q \frac{\partial \psi}{\partial x} - \phi &= 0 \\
\Gamma &= 0, t \in \{0, T\} \\
\Gamma &= 0, x \in \{-1, 1\}
\end{cases}$$
(32)

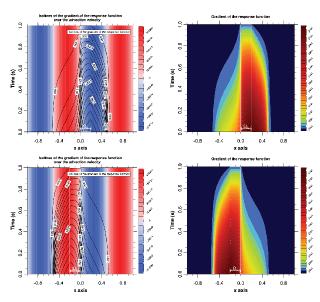
$$\begin{cases}
\frac{\partial \Lambda}{\partial t} + \frac{\partial u \Lambda}{\partial x} + \eta \frac{\partial^2 \Lambda}{\partial x^2} + \frac{\partial q \phi}{\partial x} - \psi &= -\frac{\partial \varphi}{\partial c} \\
\Lambda &= 0, t \in \{0, T\} \\
\Lambda &= 0, x \in \{-1, 1\}
\end{cases}$$
(33)

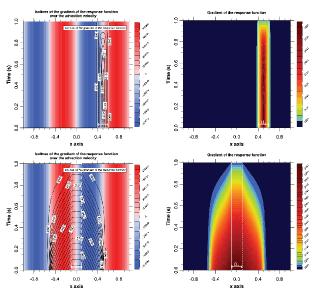
Where the function 1_{Ω_A} is:

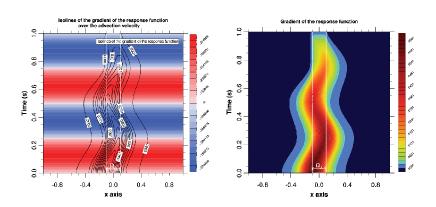
$$1_{\Omega_A}(x) = \left\{ egin{array}{ll} 1, & ext{if } x \in \Omega_A \ 0, & ext{if not.} \end{array}
ight.$$











The End