

# User Guide (MATLAB Version R2019a)

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## 1 Introduction

This code produces a solution of an ill-posed nonlinear problem of the form:

$$u(x) = g(z(w)) \quad (1)$$

$$= \int f(z(w))dw \quad (2)$$

where  $g$  is given and we are trying to recover continuous  $z(w)$  with discrete observation of  $u(x)$ .

Considering discrete data with continuous parameter case, the authors of [1] use the following back-propagation to recover  $z(w)$ . They devised the following iterative algorithm that is continuous in  $w$ .

$$\begin{aligned} \hat{z}_{k+1}(w) = z_0 + \int dw' \sum_i \sum_j C_z(w, w') G_k^i(w') (S^{-1})^{ij} [d^j - g^j(\hat{\mathbf{z}}_k) \\ + \int dw'' G_k^j(w'') \cdot [\hat{z}_k(w'') - z_0(w'')]] \end{aligned} \quad (3)$$

We are coding the above equation (3). Here  $G_k^i(w)$  is the derivative of  $g(x^i, w, z_k(w))$ ,  $z_0$  represents the prior parameter estimate and  $C_z(w, w')$  is the prior covariance. The authors define  $C_z(w, w')$  as follows:

$$C_z(w, w') = \sigma^2 \exp\left[-\frac{1}{2} \frac{(w - w')^2}{\Delta^2}\right] \quad (4)$$

Note that here  $\mathbf{G}^T$  in the continuous setting is the more general adjoint  $G^*$ . That adjoint  $G^*$  is defined by  $\langle Gr, t \rangle = \langle t, G^*g \rangle$  for all  $r, t$  in a Hilbert space [1].

## 2 User Interface

If you want to run the code, you need to run the "`frontier.m`" file only. Note that the "`frontier.m`" is calling the "`inv_DDCP.m`" file so that we could obtain our output `zhat` along with its validations and plot those. For more details about `zhat`, please see the Tarantola's paper [1] and "`inv_DDCP.m`" file where we have coded the main part of Tarantola's algorithm. More general, the name "`inv_DDCP.m`" was chosen to represent "inverse solution of discrete data and continuous parameter".

The user interface of obtaining the "inverse solution of discrete data and continuous parameter",  $\hat{z}(w)$  (see equation (3)) between two media of different density using as data the anomaly is as follows:

```
zhat=inv_DDCP(w,x,d,sigma_d,sigma,theta,K,@Gfun,@ffun)
```

where

### Output Variable :

`zhat` : Output vector  $\hat{z}_{k+1}(w)$  (see equation (3)). Here we are recovering the  $\hat{z}(w)$  at the  $(K + 1)$ -th iteration.

### Input Variables :

`w` : Discrete grid of the independent variable for the continuous parameter. A finer grid will produce more accurate approximation to  $z(w)$ . This  $w$  needs to be a column vector. For more details, please see the `run.m` file.

`x` : Location where data are observed. This  $x$  needs to be a column vector. For more details, please see the `run.m` file.

`d` : Data vector considering it as a single column floating point numbers in units. This data is imported from `.txt` file. In my case, the name has been specified as "`data.txt`". Please change your datafile path if needed.

`sigma_d` : Constant standard deviation of the data error.

`sigma` : Prior uncertainty of  $z_0(w)$  in units. See  $\sigma$  in equation (4). Changing `sigma` is the reason of having different prior covariances.

**theta** : Spread in the prior covariance. See  $\Delta$  in equation (4).

**K** : Number of iteration of the algorithm (3).

**Gfun** : Model which has been implemented as the above equation (3).  $G_k^i(w)$  is the derivative of  $g(x^i, w, z_k(w))$ . Here "Gfun.m" file contains the model. Writing a different model in the "Gfun.m" file but keeping the "inv\_DDCP.m" file unchanged would provide a different result. For example: if we change the value of the "Depth between the surface and the subsurface", then we can easily form a different Gfun and end up having a different result.

**ffun** : Integrand. See equation (1). Writing a different "ffun.m" file but keeping the "inv\_DDCP.m" file unchanged would provide a different result. For example: if we change the value of the "Depth between the surface and the subsurface", then we can easily form a different ffun and end up having a different result.

### 3 Example

For an example, we are considering "the gravitational inverse problem of obtaining the shape of a frontier  $z(w)$  between two media of different densities using as data the anomaly at the surface" [1]. It is the example of an ill-posed non-linear problem from [2].

- We have considered the Tarantola paper's

$$u(x) = \int_a^b \log \frac{(x-w)^2 + H^2}{(x-w)^2 + [H-z(w)]^2} dw \quad (5)$$

where  $a = wmin$ ,  $b = wmax$ ,  $H$  is the depth between the surface and the subsurface.

Also, according to Tarantola's example, we have considered the following  $g(x^i, w, z_k(w))$  and **ffun** i.e,  $f$

$$g = \int f dw = \int \log \frac{(x-w)^2 + H^2}{(x-w)^2 + [H-z(w)]^2} dw \quad (6)$$

- We have created a continuous true  $z(w)$  called  $ztrue$  using the concept

of the Tarantola paper [1] where

$$z_{true} = \begin{cases} 0 & \text{if } 0 < w < 40 \\ 2.5 e^{-5 \frac{(w-50)^2}{m}} & \text{if } 40 < w < 60 \\ 0 & \text{if } 60 < w < 100. \end{cases}$$

- We form a true  $u(x)$  (see equation (1)) called  $u_{true}$  where  $u_{true} = \int f(z_{true})dw$ . Then we create our own data set  $d$  where

$$d = u(x) + \varepsilon \quad (7)$$

and  $\varepsilon$  follows a Gaussian distribution with mean 0 and standard deviation  $\sqrt{0.001}$ . Users can create their own data set  $d$  using the above equation (7) according to their particular problem set. Note that in the "inv\_DDCP" folder we also have "creatingNEWdata.m" file so that user could create other different data sets according to my particular example. With new data  $d$ , our results will be different. To avoid replacement issue, we have chosen different names in "creatingNEWdata.m" file. If you are using different data with a different filename, then please change the filename in the "Loading data" command at our "frontier.m" file.

- We save our data set  $d$  as "data.txt" file which is a column vector and it is inside our "inv\_DDCP" folder.
- We form the `zhat=inv_DDCP(w,x,d,sigma_d,sigma,theta,K,@Gfun,@ffun)` and plot this "zhat" (see Figure 1) considering the following values of our variables.

### An example: Recovering the frontier

```

m= 100; %Points of integration grid of Tarantola paper's algorithm
wmin=0; % [km] Lower limit of the integration domain
wmax=100; % [km] Upper limit of the integration domain
w = linspace(wmin,wmax,m); % Defining the integration grid w
% Loading data
d= load("data.txt"); %[km] data vector (column vector)
n = length(d); %Total number of data
x = linspace(wmin,wmax,n)'; %Location where data are observed
sigma_d= sqrt(0.001); %[km] Constant standard deviation of the data
error
sigma = 5; %[km] Prior uncertainty
theta = 1; %[km] Spread in the prior covariance
H = 10; %[km] Depth between the surface and the subsurface
K=10; %Number of iteration

```

Note that we have considered our  $G_{\text{fun}}$  i.e.,  $G_k^i(w)$  according to [1] where

$$G_k^i(w) = \frac{2(H - z_k(w))}{(x^i - w)^2 + [H - z_k(w)]^2} \quad (8)$$

The plot of our `zhat` looks like:

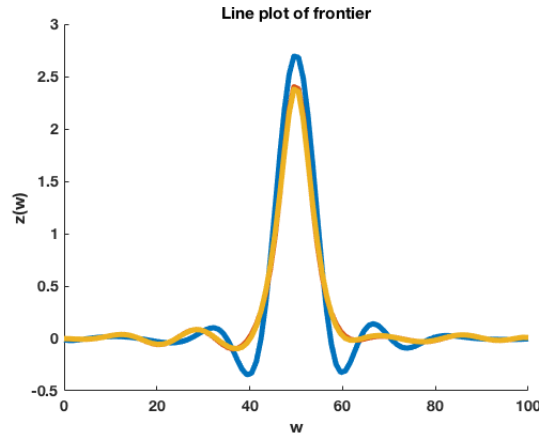


Figure 1: Example: Recovering the frontier  $z(w)$ .

### Validation of my example

To validate our `zhat`, we have created a section in our "frontier.m" file. When users run this "frontier.m" file, they will see the plot of our result-

ing frontier,  $\mathbf{zhat}$  along with its validations. In this file, we have formed a ground truth i.e.,  $\mathbf{ztrue}$  and plotted this  $\mathbf{ztrue}$  with our estimated  $\mathbf{zhat}$  to see the accuracy in  $z$  (see figure ??) . Similarly, we have created a  $\mathbf{uttrue}$ , and estimated a  $\mathbf{uhat}$  and plotted this  $\mathbf{uttrue}$  with our estimated  $\mathbf{uhat}$  (see figure ??). The file named as " $\mathbf{g\_small\_fun.m}$ " is also a part of our " $\mathbf{frontier.m}$ " file. We have formed this " $\mathbf{g\_small\_fun.m}$ " according to the equation (1). For more details, please read [1].

The plots to check the accuracy in  $z$  (1st graph) and prediction (2nd graph) are as follows:

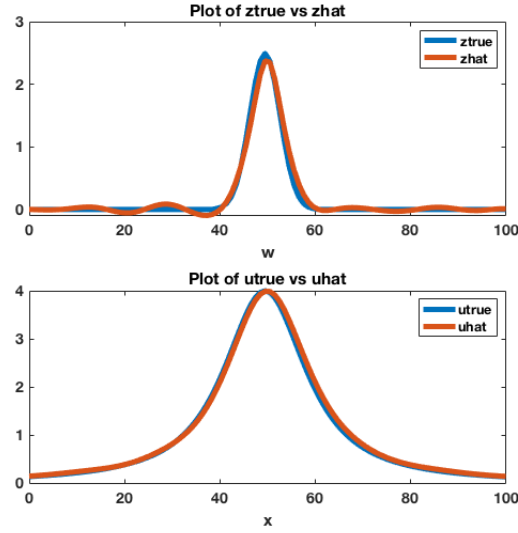


Figure 2: Accuracy checking in  $z$  (1st graph) and Accuracy checking in prediction (2nd graph)

In addition, we have calculated the "relative error in  $z$ " and the "relative error in prediction" as follows:

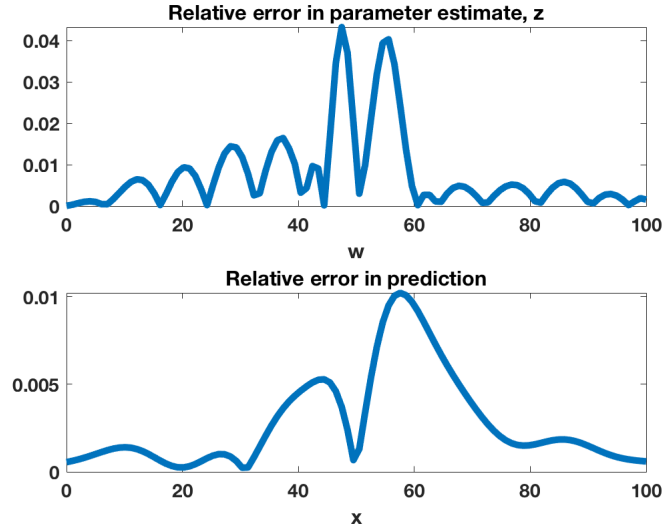


Figure 3: Relative Error in  $z$  (1st graph) and Relative Error in prediction (2nd graph)

## References

- [1] A. Tarantola, B. Valette. *Generalized Nonlinear Inverse Problems Solved Using the Least Squares Criterion*. Institut de Physique du Globe de Paris, 75005 Paris, France. Reviews of Geophysics and Space Physics, Vol. 20, No. 2, pages 219 – 232; May 1982.
- [2] A. Tikhonov, V. Arsenine. *Methodes de resolution de problemes mal poses*. Editions MIR, Moscow, 1976.