# The code on solving Fredholm equation of the first kind: usage instructions

#### I.I.Antokhin

Sternberg State Astronomical Institute, Lomonosov Moscow State University

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#### 1 General notes

- 1. While some code is implemented to prevent most evident user errors (like setting the initial approximation for z outside the region defined by the constraints etc.), the code is not fool proof. Making it such would take a lot of efforts and time which I do not have.
- 2. The code "as is" can be used for solving Fredholm equation with kernels appropriate for solving WR+O light curves, or for solving Abel equation. If you need other kernels for Fredholm, you should provide your own functions computing the kernels (see below).

## 2 Files in the package

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- functions to solve Fredholm and Abel equations
prgrad_reg.c
wr_o_common.c
                     - equation kernels for Fredholm eqs. solving WR+O light curves
                       and Abel eq, also routines computing elliptical functions
test_prgrad.c
                     - test program showing how to run functions from prgrad_reg.c
Makefile
                     - make file to compile test_prgrad
                     - The output of the test program, solution on a compact set, no
test_compact.dat
                       Tikhonov's regularization
                     - Same, on the same compact set plus Tikhonov's regularization
test_reg_W21.dat
                       in the metric space W_2^1
                     - Same, metric space W_2^2
test_reg_W22.dat
prgrad_reg_usage.pdf - this file
```

# 3 Compiling

The code is developed in Linux and compiled with gcc 12.3.1. There should be no problem compiling it with any ANSI C compiler. In a non-Linux OS you will have to replace the random generator functions srandom and random with your local versions. These functions are used in the test code only, to add Gaussian noise to the simulated input data.

To compile the test program in Unix environment, run make (correct the supplied Makefile if needed). After compiling, run the test program as

#### test\_prgrad

More on the test code and a simulated model used in it, below. Note that in UNIX/Linux lines of text files are ended with <LF> while in Windows with <CR><LF>. If you use Windows you may want to convert the files to <CR><LF> format, otherwise you will see a file as a single line.

# 4 The structure of the calling (main) program

It is linear and very simple. See comments in the source code test\_prgrad.c. Outline:

Define necessary variables.

Read input data.

Calculate the uncertainty of input data and the sum of weights.

Compute the constraints matrix and its right-hand part.

Set initial approximation for z.

Call the main function solving the equation (ptizr\_proj).

Calculate model Az.

Output the results.

Variables: In my code the memory for all arrays is allocated dynamically. If you prefer, you may set most of 1D arrays (u, z, az, grids, weights etc.) in the calling program as static. Exceptions are the constraints matrix con and its right-hand part b which are allocated within the function ptilrb and must be defined as pointers (see the source code in test\_prgrad.c and comments below).

Read input data: For simplicity and readability, in test\_prgrad.c I hard-coded reading them in the body of the main function. You may wish to write a separate function and call it from main. Same for setting algorithm parameters. I actually do this in my practical applications of the code.

Uncertainty of input data and the sum of weights: These are numeric representations of the uncertainty

$$\delta^2 = \int_c^d \sigma^2(x) dx$$

(where  $\sigma(x)$  is uncertainty of u(x)) and the integral of weights

$$sumv = \int_{a}^{d} v(x)dx$$

where v(x) is weight of u(x), e.g.  $\sim 1/\sigma^2(x)$ . See the test code for further details.

# 5 Description of functions to call from your main program

#### 5.1 Constraints matrix

A priori constrains have the form

$$con \cdot \vec{z} < \vec{b}$$

where con is the constraints matrix and  $\vec{b}$  is the right-hand part. They are computed by

Parameters:

kernel\_type - 1,2, or 3. 1 and 2 are kernels for Fredholm eq. when
solving light curves of WR+O binaries. For Abel equation,

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you must set it to 3. You can add your own kernels and
                 enumerate them as 4, 5, ...
switch_contype - type of constraints on the unknown function:
                 = 1 - monotonically non-increasing and non-negative
                       (constraints 2,3,4 also include this constraint)
                 = 2 - concave z" <= 0
                 = 3 - convex z" >= 0
                 = 4 - concave-convex with inflection point index 1
                 = 5 - non-negative (this is to use Tikhonov's regularization
                       only without specifying any compact set)
               - the number of rows in con, computed within the function
n_con
               - constraints matrix. Note that in the calling program you must
con
                 declare it as "double **con". Memory for con is allocated within
                 the function. The cell values are also computed within the function.
               - right-hand part of constraints. In the calling program must be
h
                 declared as "double *b". Allocated and computed within the function.
               - dimension of z.
n
               - The value of z(0) (if known, see below). If unknown, set to
                 an arbitrary value. However, you will need it for initial
                 approximation, so may set to a reasonable value before caling
                 ptilrb (see below).
               - If it is known that z(s)=c2 at s<s0, where s0 is some known value,
icore
                 icore sets the maximal index of a grid knot where, in numeric
                 representation of z, z[i<=icore]=c2 (see below). For Abel equation,
                 MUST be set to 0: icore=0.
1
               - Index of inflection point for switch_contype=4. For other
                 constraint types, not used, set to an arbitrary value.
               - return code. O if all is ok, 202 if memory for con or b could
ierr
                 not be allocated.
```

Note that when calling ptilrb, you must use &con, &b and not just con, b.

#### 5.2 Initial approximation

#### 5.3 The ptizr\_proj function

Here is the main function solving Fredholm or Abel equations. I provide a list of all its parameters and then give explanations on some of them.

void ptizr\_proj( int kernel\_type, int switch\_contype, int n\_con, double \*\*con, double \*b, double rstar, double \*u0, double \*v, double sumv, double \*s, double \*x, double xmin, double xmax, int n, int m, double \*z, double c2, double d12, double eps, double h, int adjust\_alpha, double \*alpha, char \*metric, int 1, int icore, double ax, double \*del2, int imax\_reg, int \*iter\_reg, int imax\_minim, int \*iter\_minim, int verbose, int \*ierr ); - Same as above. kernel\_type switch\_contype - Same as above. n\_con - Same as above. con - Same as above. - same as above. b - When solving a set of Fredholm eqs. for WR+O light curves, sets the rstar radius of the O star (used in kernels). In Abel equation, or if you use your own kernels in Fredholm, set to an arbitrary value. u0 - Input data, right-hand part of Fredholm eq. Array of size m. The same as u from the paper. - Weights of u0 data points. - Sum of weights (see the code in test\_prgrad.c). sumv - Same as above. - Argument of u0. Array of size m. х - Lower limit of x. xmin - Upper limit of x. xmax - Dimension of z. n - Dimension of u0. The number of input data points.  $\mathbf{m}$ - Unknown function. Array of size n. When calling the function, must z contain initial approximation. On return, contains the solution. - Same as above. - Uncertainty of input data (see above). delta2 - Threshold for solving ||Az-u||^2-delta2=0. See below. eps - If known, uncertainty of the A operator. The difference between h exact Az and its numeric approximation. At reasonably large n, it is safe to set h=0.0. - If 1, search for optimal regularization parameter. If 0, solve adjust\_alpha at fixed alpha. - Either alpha value to use or its initial value. alpha metric - Metric space to use. May be "L2", "W21", or "W22". - Index of the inflection point as above. icore - As above. - A parameter used when solving light curves of WR+O (needed for ax computing the kernels. For your own kernels or in Abel eq., set to an arbitrary value. del2 - Residual of the model relative to the input data. - Max number of regularization iterations. imax\_reg iter\_reg - Actual number of regularization iterations made. - Max number of conjugate gradients minimization iterations made in imax\_minim the last regularization iteration. - Actual number of conjugate gr. iterations made. iter\_minim - If set to 1, some additional stdout printing is done. Set to 0 to verbose skip verbose printing. ierr - Return code. 1\*\* - normal end, 2\*\* - various errors = 100 - normal end, exact minimum is found

= 101 - iterations finished by residual value

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= 102 - iterations finished by the norm of the gradient
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- = 103 alpha became equal to zero while doing iterations
- = 200 initial approximation outside allowable range.
- = 201 inconsistent adjust\_alpha and alpha
- = 202 errors allocating memory for working arrays
- = 203 singular matrix of active constraints (when computing the projector)
- = 204 initial alpha < 0
- = 205 while searching for the interval on alpha containing the solution, made imax\_reg multiplications by 2.0, residual still negative. Initial regularization parameter too small.
- = 207 in bisection method, imax\_reg iterations done, still not reached exit criteria.
- =  $208 kernel_type < 1 or > 4$ .

Hopefully, the meaning of most of the parameters above is clear. A few comments on some of them:

#### Parameters controlling Tikhonov's regularization:

If adjust\_alpha=0 and alpha=0, no Tikhonov's regularization is done. The equation is solved on a compact set. If adjust\_alpha=0 and alpha>0, Tikhonov's regularization at this fixed value of alpha is performed. If adjust\_alpha=1, a search for the optimal alpha is performed. Before calling ptizr\_proj alpha must be set to some positive value, otherwise the function will return an error. This positive value is used as a starting point in searching for the optimal alpha.

Optimal alpha is defined by the condition

$$an4 \equiv ||Az - u||^2 - \delta^2 = 0$$

The function on the left is a monotonically increasing function of alpha (see Fig. 1). So the solution of the above equation is located between the points alpha=0 and some alpha\_1 such that the left hand part of the above equation is positive. The search is performed as follows:

- 1. Compute an4 at alpha=0. If an4<0, proceed, otherwise exit (see below).
- 2. Compute an4 at alpha set before calling ptizr\_proj.
- 3. If an4 is negative, multiply alpha by 2 and repeat (2).
- 4. Repeat (3) until an4 becomes positive (at some alpha\_1).
- 5. Now, the solution is between alpha=0 and alpha=alpha\_1.
- 6. Use the bisection method to solve the above equation.

In practical terms, the solution is considered to be found if

$$||Az - u||^2 - \delta^2 < \epsilon \delta^2$$

Thus, variable eps sets the threshold for solution of the above equation.

There may be a situation when even at alpha=0

. This may happen e.g. if a priori constraints do not allow the model to fit the data very well. Or, if the uncertainty of input data  $\delta^2$  is underestimated. No Tikhonov's regularization is possible in this case and the function exits with the errorcode ierr=104. z contains the solution at alpha=0.

#### Inflection point

If you want to use the concave-convex type of constraints, you must set the position (index) of the inflection point. If it is known, just set it once and call ptizr\_proj. If not, you can search for it by repeatedly calling ptizr\_proj with various values of 1 and choosing the optimal value by the minimum of de12. Clearly, 1>=1 and 1<=n-2 (recall that in C array indexes start from 0 so the first and last indexes of an array of the length n are 0 and n-1). Otherwise, z will be simply convex or concave. You can use these values of 1 as the end points of the possible 1 interval and find the optimal 1 by e.g. the golden section method. I have the code for the method but did not include it in the test code, as I wanted to keep it simple. In any case, it is easy to implement.

In reality the situation may be more complicated than that. For instance, in case of the simulated unknown function I included in this package, you can see (Fig.2 below) that in the range  $s \sim 0.2-0.4$  the function is only slightly curved (not very different from a straight line). So if you choose various s(l) to be between 0.2 and 0.4 the deviations from the data will be about equal. Only at  $s(l) > \sim 0.6$  the concave part of model z will be very different from the true (exact) function and the deviation will become large. Moreover, as the central (small s) parts of z give relatively small contribution to the total value of Fredholm integral, at small 1 the deviations will increase not so much. This is illustrated in Fig.2. Note that the true value of 1 in this example (the inflection point of the exact simulated z) is equal to 40, corresponding to s=0.295.

This illustration shows that it would be a bad idea to blindly choose optimal 1 by the minimum of the norm. The dependency of the norm on 1 is strongly defined by the shape of the unknown function. For simulated functions used in my examples, parts of which are nearly straight lines, there will be no clear "parabola"-like shape of the dependency unambiguously defining 1. If real z(s) is a curved function with the second derivative significantly different from zero at all points, the situation may be different.

On the other hand, if a part of your unknown function is nearly straight line, there is no much difference which value of 1 you choose. As long as its is located within the straight part of the function, the shape of the solutions with different 1-s will be nearly identical. Still, I make these comments so that you will understand the possible behaviour of the algorithm.

In practical terms, this means that it is a good idea to run the program for all possible values of 1 and produce a plot like my Fig.2. If the plot shows well defined minimum, the corresponding value of 1 may be easily choosen. If not, you might use some independent considerations to choose the optimal 1, e.g. if you have some quesses on possible shape of your unknown function.

One further note has to be made. If you want to use Tikhonov's regularization with the concave-convex constraint, do not use it when searching for the optimal 1. Instead, while searching for such 1, solve your equation on a compact set. The reason is that with Tikhonov's regularization switched on, you will always get  $del2 = \delta^2$  and will not be able to choose the optimal 1 by the minimum of del2. Also, at different 1 the values of optimal alpha will be different and the corresponding solutions are hard to compare. So search for the optimal 1 not using Tikhonov's regularization, and then run ptizr\_proj once more with the optimal value of 1 (just found) AND Tikhonov's regularization switched on.

#### icore and c2

Whatever is the kernel of Fredholm equation, c2 is used to set z[0] in the initial approximation of z. However, in my WR+O code it also has another use. In this code, I solve not just one Fredholm equation but a system of three eqs., two Fredholm ones and one algebraic eq. setting the normalization. After solving the first Fredholm eq., by using the normalization equation, I get c2 such that in the second Fredholm equation z(0) must be equal to c2. Moreover, in that latter equation z(s) is the opacity of the WR disk  $z(s) = c2(1 - e^{-\tau(s)})$ , where  $\tau(s)$  is the optical depth at the impact distance s from the disk center. Clearly, even though a WR star has semi-transparent wind, it also has a completely non-transparent core  $\tau(s < rcore) = \infty$ . So z(s < rcore) = c2. icore is computed as rcore (set in the input file) divided by ds, the grid step size. Thus, icore is the index of the last point of the grid on s, where

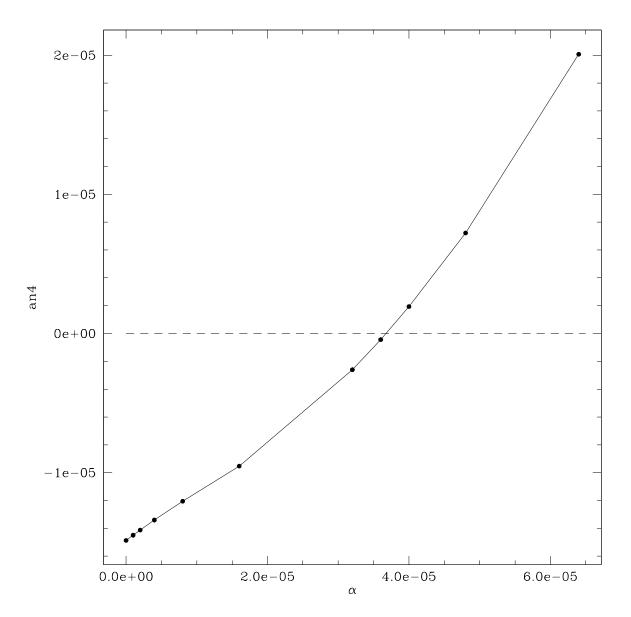


Figure 1: Dependence of an 4 from  $\alpha$ . Black dots show the sequencial steps in bisection search for the optimal  $\alpha$ .

z[i] = c2. Important: if icore>0, the unknown function z(i) is fixed to c2 at all grid knots with indexes from 0 to icore.

This means that if you do not have such constraint on your unknown function, set c2 to some reasonable value (see above) before calling initial approx and set icore=0. If icore>0, your solution z(s) will be fixed to c2 at all z[0<=i<=i].

#### 5.4 Computing model u (Az)

This is simply calculating the Fredholm integral with model z. To do this, first compute the kernel (of course, it is computed in ptizr\_proj, but it is a local variable not seen outside ptizr\_proj).

Allocate memory for the kernel. Function "matrix'" allocates memory for a 2D array with m rows and n columns,

double \*\*a = matrix( int m, int n )

Compute Az (the model function to compare with the input data):

void pticr3( double \*\*a, double \*z, double \*az, int n, int m )

### 6 The test program

In the test program I tried to keep the code to bare essentials for readability and easier understanding. All parameters and data are hard-coded in the "main" function. All output goes to stdout. Redirect it to a file if you wish.

The test program solves the Fredholm equation

$$Az = \int_{a}^{b} K(x, s)z(s) ds = u(x), \quad s \in [a, b], x \in [c, d]$$

The simulated model is defined as the kernel:

$$K(x,s) = \frac{1}{1 + 100(s - x)^2}$$

The intervals [a,b] = [0,1], [c,d] = [0,1]. The exact z is set to be a concave-convex function

$$z_0(s) = \frac{3}{2} \left( \cos(\pi \frac{s}{2s_l}) + 1 \right)$$

so that  $z_0(0) = 3$ . The position of the inflection point  $s_l = 0.4$ . With the given kernel and  $z_0$ , the integral in the left hand side is computed, resulting in the exact right hand part  $u_0(x)$ . Gaussian noise with  $\sigma = 0.01$  is then added to  $u_0(x)$  resulting in u(x) which is used as input data for test solutions.

The test solutions are provided for three cases:

- 1. Solution on a compact set of non-negative, monotonically non-increasing concave-convex functions without Tikhonov's regularization.
- 2. The same, with added Tikhonov's regularization in the metric space  $W_2^1$ .
- 3. The same, with added Tikhonov's regularization in the metric space  $W_2^2$ .

The results are shown in Fig.3-5. To get a feeling on how the algorithm works, you may wish to play with the test model e.g. by changing the value of Gaussian noise, kernel parameters, using different constraint types (e.g. switch\_contype=1 for solving the equation on a set of monotonic functions) etc.

# 7 Adding your own kernel

The kernel of the Fredholm's equation is computed by the pticr0 function in prgrad\_reg.c. However, since I have more than one kernel and in every one there is some specifics, ptirc0 does not include the code computing the various kernels, directly. Instead, it calls functions which do this for various cases. These functions for the 4 current kernels (WR+O light curve eclipses 1 and 2, Abel equation, and the test problem) are locates in wr\_o\_common.c. To use your own kernel, you may add kernel number 5 by adding a call to your own function from pticr0 and adding the function itself to wr\_o\_common.c.

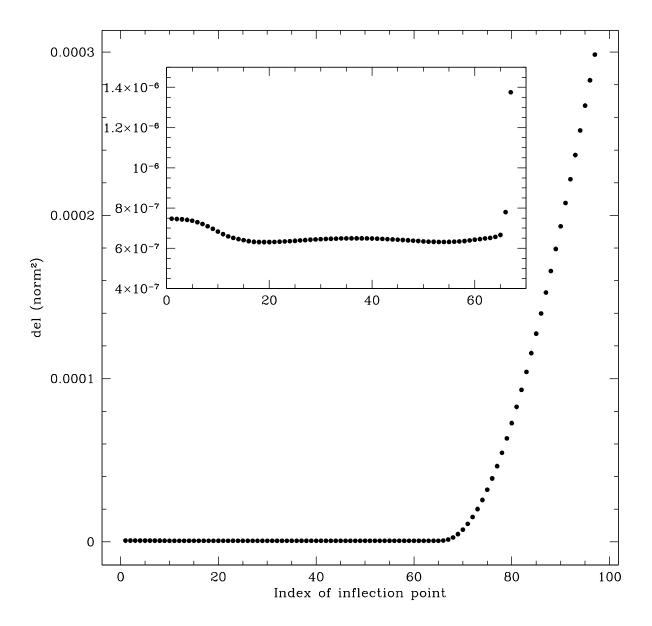


Figure 2: Dependency of the norm  $||Az - u||^2$  from the position of the inflection point l, for a simulated function with known solution, with added Gaussian noise (sigma=0.001) and without Tikhonov's regularization. See comments in text. In the insert, the lower-left part of the plot is shown at an increased scale.

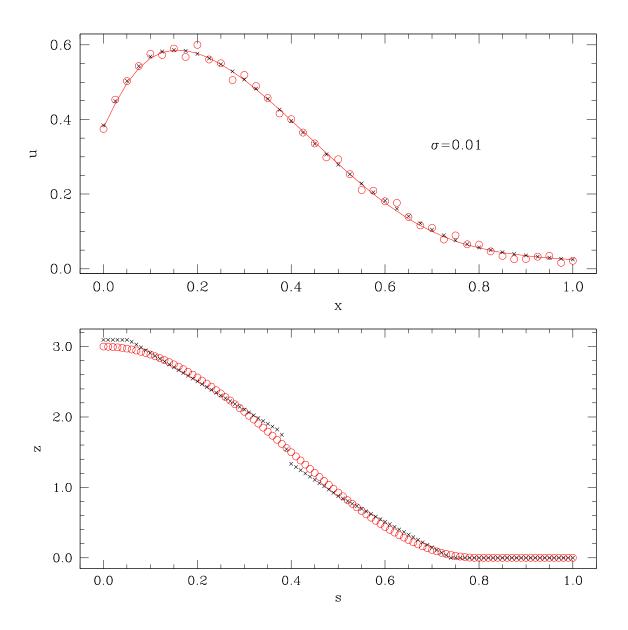


Figure 3: Solution 1.

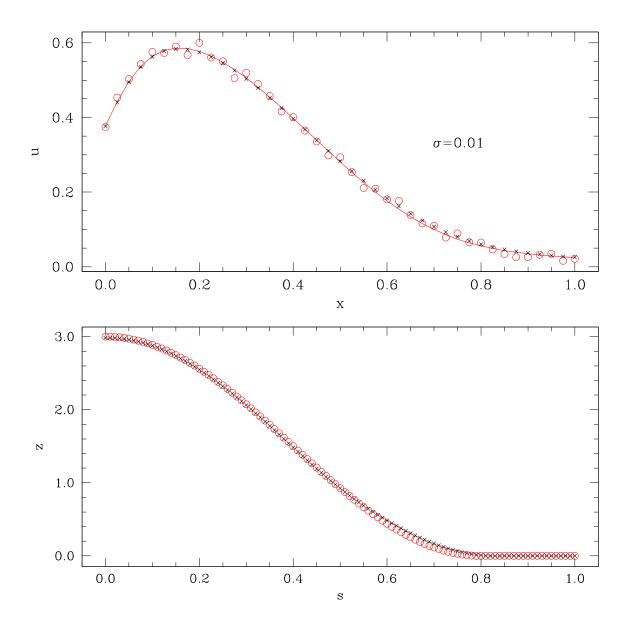


Figure 4: Solution 2.

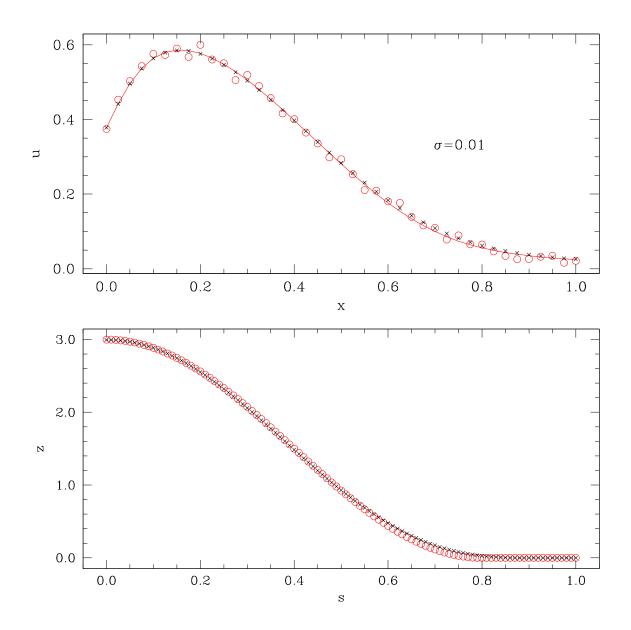


Figure 5: Solution 3.