

User's Guide

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Getting started

1.1. Prerequisites

The ESTER libraries depend on some external libraries that should be installed in the system, namely:

- BLAS, CBLAS and LAPACK, for matrix algebra. There are several alternatives available, as for example:
 - Netlib. This is the original implementation. The LAPACK library can be found at http://www.netlib.org/lapack, and already contains BLAS, but CBLAS should be downloaded separately from http://www.netlib.org/blas.
 - ATLAS (Automatically Tuned Linear Algebra Software). An implementation of LA-PACK/BLAS that is automatically optimized during the compilation process. It can be found at http://math-atlas.sourceforge.net/. It contains LAPACK, BLAS and CBLAS.
 - Intel MKL. Contains an optimized version of LAPACK, BLAS and CBLAS for Intel processors.
- Python
- HDF5 for standardized model output (optional). HDF5 is available at http://www.hdfgroup.org/downloads/.

1.1.1. A NOTE ABOUT THE PERFORMANCE OF THE CODE

The performance of the ESTER code depends strongly on LAPACK. To get the best results, use an optimized (and parallelized) version.

1.2. Installation

Please go to the wiki pages to get the latest instructions for installation. Now we prefer to use the *cmake* software to do the installation (Jan. 2019).

The latest version of ESTER is available from the git repository:

```
$ git clone https://github.com/ester-project/ester.git
```

or from a source tarball: http://ester-project.github.io/ester/.

The tarball source is for users who need a stable version and develop their own applications from it. The git repository version is for users who want to keep up with the lastest version of the code or even want to contribute to its development.

If you choose to get ESTER from the git repository, you will need to have libtool, autoconf and automake installed. The first step after cloning the repository is to run the bootstrap script:

```
$ cd ester
ester$ ./bootstrap
```

This will create the configure script.

1.2.1. Configure

In this step, the **configure** script will detect the libraries and compiler installed on the system. It us preferred to configure and compile in a different directory than the top source directory:

```
ester$ mkdir BUILD
ester$ cd BUILD
ester/BUILD$ ../configure [OPTION]... [VAR=VALUE]...
```

The most important configure options are:

-prefix: selects the installation directory for binaries, ester libraries, etc. (default is \$HOME/local.

-enable-hdf5: to enable HDF5 support (this requires to have HDF5 library installed on the system).

-help: prints help and the full list of configure options.

The following variables can be used to tune the ESTER build configuration:

FC: Fortran compiler to be used (e.g., FC=ifort)

CC: C compiler

 \mathbf{CXX} : C++ compiler

For instance, is you want to compile with Intel compilers and install ESTER in \$HOME/ester_local, you should use:

```
ester/BUILD$ ../configure --prefix=$HOME/ester\_local CC=icc CXX=icpc FC=ifort
```

1.2.2. Build and Install

After the configure step, building and installing ESTER is straightforward:

```
ester/BUILD$ make && make install
```

make will build the ESTER's libraries and binaries. And make install will copy the libraries into \$prefix/lib and binaries into \$prefix/bin.

Make sure you add the install directory to you PATH environment variable to be able to launch ester without specifying the full path to the binary. If you are using bash, you can add the following line to your .bashrc:

export PATH="\$HOME/local/bin:\$PATH"

1.2.3. Updating the code

If you chose to download ESTER from the git repository, you can update the code with:

```
ester$ git pull
```

and compile the new version by going to your build directory and running make install:

ester\$ cd BUILD
ester/BUILD\$ make install

1.3. Checking the Installation

To check the installation, a series of test runs is proposed: following the previous example, in the directory \$HOME/ester_local/bin, where binaries have been built, just run ester-check-models. These tests first compute a 1D 5 M_{\odot} model both with native and hdf5 output and compare the results to a reference output. A 2D 5 M_{\odot} model with $\omega = 0.5\omega_k$ is then computed and checked again for the two kinds of output. The same check is repeated for a $10M_{\odot}$ model but with $\omega = 0.3\omega_k$.

Basic usage

The ESTER code calculates the axisymmetric structure and mean flows of an isolated (non-magnetic) rotating star. It uses realistic physics (tabulated opacities and EOS like those of OPAL) and completely accounts for the deformation of the star. The mean flows are calculated self-consistently in the limit of low viscosity, so there is no need to impose an arbitrary prescription for the differential rotation of the star. Surface convection is not included yet, so the computations are limited to early-type stars, that is to stars with mass typically above $\sim 2 M_{\odot}$.

Presently, chemical evolution is not included, but it can be faked by tweaking the fractional abundance of hydrogen in the convective core.

For a detailed description of the physics involved in the models, see Rieutord & Espinosa Lara 2013 (arXiv:1208.4926, LNP 865, p.49), Espinosa Lara & Rieutord 2013 (arXiv:1212.0778, in A&A 552, A35) and Rieutord et al. 2016 (), in J. Comp. Phys. vol. 318, 277-304. The code is still in development, so new functionality will be added in future versions.

To execute the program, use the following syntax:

\$ ester command [options]

where command can be:

1d : Calculate the structure of a 1D non-rotating star

2d : Calculate the structure of a 2D rotating star

output: Generate a custom output file

info: Get information about a model file

help: Get help

2.1. Configuration files

The main configuration file is located at \$prefix/share/ester/config/star.cfg. This file contains the main options for the program, which are

• maxit (default 200). Maximum number of iterations. After maxit iterations, the program exits normally and the output file is saved, even if it has not completely converged.

- minit (default 1). Minimum number of iterations. It may occur that the value of the error for the first iteration is not representative. With this parameter we force the solver to do at least minit iterations. This parameter is superseded by maxit, for example if maxit=5 and minit=10, the solver will do only 5 iterations.
- tol (default 1e-8). The relative tolerance for checking the convergence of the model.
- newton_dmax (default 0.5). After one step of the Newton's method, the maximum relative change allowed for a variable is given by newton_dmax. If necessary the iteration is relaxed by a parameter h

$$\boldsymbol{x}^{N+1} = \boldsymbol{x}^N + h\delta\boldsymbol{x}^N$$

according to this value. This parameter can be used to stabilize the convergence when the initial estimation is far from the solution.

- output_file (default star.out). Name of the output file.
- output_mode (default b). Type of the output file b for binary and t for text output.
- verbose (default 1). Level of verbosity, from 0 (quiet) to 4.

All these options can be specified in the file <code>\$prefix/share/ester/config/star.cfg</code> in the form <code>option_name=option_value</code> (one per line) and in the command line as <code>-option_name</code> <code>option_value</code>. The options specified in the command line have precedence over those specified in the configuration file.

There are some additional options that can be included in the command line:

- -input_file infile. Use the file infile as the starting point for the iteration.
- -i infile. Same as -input_file infile.
- -o outfile. Same as -output_file outfile.
- -param_file file. Where file contains the parameters of the stellar model to be calculated (see below).
- -p file. Same as -param_file file.
- -ascii. Same as -output_mode $\ensuremath{\text{t.}}$
- -binary. Same as -output_mode b.
- -noplot. disable runtime plotting
- -vn. Same as -verbose n.

2.2. Default values

Default values to be used by star1d or star2d may be set up with the files

- \$prefix/share/ester/config/1d_default.par
- \$prefix/share/ester/config/2d_default.par.

In the distribution of ESTER, the proposed default values are such that the star is divided in 8 domains with 30 points in each domains. Opacities and equation of state are computed through OPAL tables. These inputs allow the calculation of a 1D 3 $\rm M_{\odot}$ model (but not only of course) from scratch. 2D-models cannot be computed from scratch and need a first 1D model to start with.

2.2.1. Chemical composition

The default chemical composition is the solar mixture of Grevesse & Noels (1993) as given in tables GN93hz. Presently (nov. 2017), it cannot be changed (easily!).

2.3. ester 1d INPUT PARAMETERS

The input parameters for ester 1d can be passed in the command line or in a text file specified with the option -param_file file (or just -p file). It can also be used simultaneously, in this case the parameters given in the command line take precedence over those specified in the file. In the text file they are written in the form param_name=param_value and in the command line as -param_name param_value. Here is the list of valid parameters

- ndomains. The number of subdomains to use.
- npts. Number of points in each subdomain. It is specified as a comma-separated list. If only one value is specified, it will be used for all the subdomains, for example:

```
$ star1d -ndomains 4 -npts 20,20,20,20 is equivalent to
```

- \$ star1d -ndomains 4 -npts 20
- M. The mass in units of solar mass.
- X. Mass fraction of hydrogen.
- Z. Mass fraction of metals.
- Xc. Fraction of the hydrogen abundance present in the convective core. The profile of hydrogen abundance will be in the form

$$X(r) = \left\{ egin{array}{ll} {\tt X} imes {\tt Xc} & {
m if} \ r \ {
m is} \ {
m in} \ {
m the} \ {
m convective} \ {
m core} \ {
m otherwise} \end{array}
ight.$$

If there is no convective core, this parameter is ignored.

- surff. This parameter is used for truncating the stellar model at some point below the surface. The surface pressure will be surff times the "real value and the boundary conditions will be adjusted correspondingly. This parameter is provided only for testing purposes as it does not produce an accurate representation of the internal layers of the star. For regular calculations it should be surff=1.
- Tc. Initial estimation of the central temperature. To be updated during the calculation.
- pc. Initial estimation of the central pressure. To be updated during the calculation.

- opa. Type of opacity law. Possible values are:
 - opal. OPAL opacities.
 - houdek. Houdek's interpolation of OPAL opacities (smoother), see Houdek and Rogl (1996), "On the accuracy of opacity interpolation schemes", Bull. Ast. Soc. India, 24, 317.
 - kramer. Kramer's opacity.
- eos. Type of equation of state. Possible values are:
 - opal. OPAL equation of state.
 - ideal. Ideal gas.
 - ideal+rad. Ideal gas with radiation.
- nuc. Type of nuclear reactions. Possible values are:
 - simple. Simplified formulation of pp and CNO cycles.
 - cesam. NACRE reaction rates as implemented in the ppcno9 chain of the CESAM code.
- atm. Type of atmosphere. At the moment, only simple is implemented.
- core_convec (default 1). Use 0 to disable core convection.
- min_core_size (default 0.01). The minimum size of the convective core in fraction of the polar radius).

If some parameters are omitted, the program will take the value from the input file (set with -input_file or -i) or from the default parameters file in ester/config/1d_default.par when no input file is specified.

Chemical composition: The chemical composition is

Element	Mass fraction
H He3 He4 C12 C13 N14 N15 O16 O17	X 3.15247417638132e-04*(1X-Z) (1X-Z)-X(He3) Z*1.71243418737847e-01 Z*2.06388003380057e-03 Z*5.29501630871695e-02 Z*2.08372414940812e-04 Z*4.82006487350336e-01 Z*1.95126448826986e-04

which actually corresponds to the one given by Grevesse & Noels (1993)

2.4. ester 2d INPUT PARAMETERS

ester 2d needs a non-rotating 1D model calculated with ester 1d or a previous 2D-model.

The program ester 2d admits the same parameters than ester 1d plus some extra specific options:

- nth. The number of grid points in latitude.
- nex. Number of radial points in the external domain.
- Omega_bk. Angular velocity at the equator in units of the critical velocity $\Omega_c = \sqrt{\frac{GM}{R_o^3}}$.
- Ekman. Ekman number (gives the amplitude of the meridional circulation).

2.5. ester evol input parameters

At the moment ester evol is the poor man recipe for stellar evolution of intermediate mass stars. It assumes that the convective core is the only part of the star that burns hydrogen. It also assumes that the star evolves at constant angular momentum. This latter assumption is therefore inappropriate for massive stars that are known to lose a lot of mass. It should be fine for stars with masses less than say $8M_{\odot}$.

From a ZAMS model, Xc is set to 1 (same hydrogen content in the core and in the envelope). Usage is

```
$ ester evol -dXc 0.01 -i M4_04 -o M4_04_evol
```

where dXc gives the variation of Xc from one step to the next. Output files are numbered M4_04_evol_0000, M4_04_evol_0001,.. Note that the output files give an evolution but not a time-evolution since the nuclear clock is not installed.

Note that dXc may also be negative so as to permit an evolution backward in time.

2.6. Some recipes

The typical workflow to calculate a model starts with the calculation of the corresponding 1D model and using it as an input for star2d. For example, to calculate the structure of a $5M_{\odot}$ star with OPAL opacity rotating at with $\Omega = 0.5\Omega_c$ we can do:

```
$ ester 1d -M 5 -opa opal -o model1d
$ ester 2d -i model1d -nth 24 -Omega_bk 0.5 -o model2d
```

As the code uses the Newton's method, it is not possible to converge to a solution if the initial estimation is too far from it. In this case we can use some intermediate steps. For example, if we want to calculate the structure of a $2.5M_{\odot}$ star rotating with $\Omega=0.9\Omega_c$, we can do

Executing ester 2d with maxit=0 can be used to interpolate a model without recalculating it, like

\$ ester 2d -i model -npts npts_new -nth nth_new -o model_interp -maxit 0

Pressing Ctrl-C at any time during the execution of ester 2d will terminate the program, giving the possibility of finishing the current iteration and write the result in the output file.

2.7. Spatial resolution and memory requirements

The ESTER code uses a direct method to solve the equations of structure of a star. This type of method involves the factorization of a big matrix that arises from the discretization of the equations. The main drawback is that memory requirements are high, but the stiffness of the equations prevents the convergence of an iterative (matrix-free) method that would be more memory efficient.

The memory needed by the calculation can be estimated as:

RAM Used
$$\gtrsim 25 \times n_d \times n_r^2 \times n_\theta^2 \times 8$$
 bytes

where

 n_d : Number of domains

 n_r : Number of "radial" points per domain

 n_{θ} : Number of points in latitude

Of course, this is a lower limit, the actual memory used can be between $\sim 10\%$ and $\sim 50\%$ higher than this value. This overhead increases with the number of domains and decreases with the overall number of points. As an example, the following table shows the memory usage for some configurations (this is only approximated, real values are machine-dependent):

n_d	n_r	n_{θ}	RAM (estimated)	RAM (real)	Overhead
8	30	24	791 Mb	986 Mb	25%
8	50	32	3.81 Gb	$4.27~\mathrm{Gb}$	12%
32	25	32	$3.81~\mathrm{Gb}$	4.61 Gb	21%

The first case correspond to the default values and, in most cases, gives a decent representation of the structure of the star for moderate rotation rates. If more precision is required, we need to increase the resolution. In particular, to achieve a good precision for high rotation values, the number of points in latitude should be incremented.

As seen in the second and third cases in the table, the total number of radial points $n_d \times n_r$ can be increased, without affecting considerably the memory usage, just by increasing the number of domains. Or equivalently, we can reduce the memory requirements by distributing the radial points over more domains.

As a rule of thumb, if the number of domains is doubled, keeping the total number of radial points, the memory required is reduced in a factor $\sqrt{2}$. However this does not necessarily imply an improvement in precision, as the order of the integration method is also reduced, so it should be used carefully.

2.7.1. Estimating the precision of the output model

The ESTER code uses spectral methods to calculate the structure of the star. The star is subdivided in a certain number of domains and, in each domain, the variables are represented by a double truncated series of orthogonal polynomials in the "radial" (ζ) and "horizontal" θ directions, for instance

$$\rho(\zeta, \theta) = \sum_{i=0}^{n_r - 1} \sum_{j=0}^{n_\theta - 1} \rho_{ij} T_i(\zeta) P_j(\cos \theta)$$

where T_i and P_j are Chebyshev and Legendre polynomial respectively. When n_r and n_θ are high enough, the spectral representation converges to the exact function. To check the convergence of the solution ρ_{ij} , the corresponding normalised spectra for the density (ρ) are shown in the graphics output of ESTER (see 2.1 and 2.2).

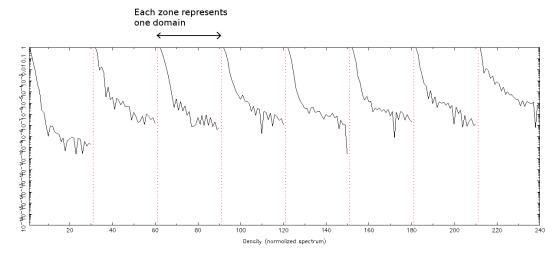


Figure 2.1: Example of spectrum plot showing the normalised coefficients ρ_i in a logarithmic scale in the graphical output of ester 1d.

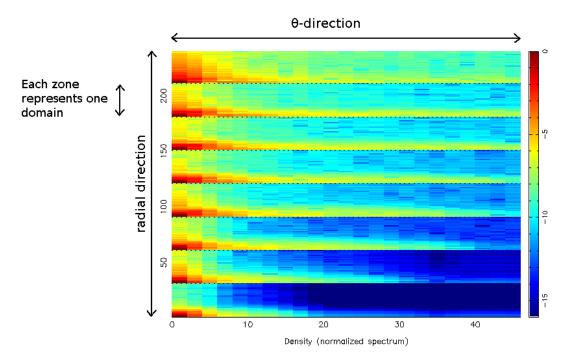


Figure 2.2: Example of 2D-spectrum showing the logarithm of the normalised coefficients ρ_{ij} in the graphical output of ester 2d.

There are two other indicators that can be used to estimate the quality of the solution:

- The virial test. It is the normalised residual resulting from the virial theorem. Ideally, it should be zero. It is mainly influenced by the internal layers and even with low resolution we can get very good values $\lesssim 10^{-9}$.
- The energy test. It is the relative difference between the luminosity of the star obtained as the integral over the volume of the energy generation rate and that obtained as the integral of the energy flux at the surface. It is highly influenced by the quality of the solution in the most external layers. Due to the use of tabulated opacities which are not quite smooth, the energy test will be always higher than 10^{-5} – 10^{-6} .

In the following table, we can see examples of the precision achieved for three different configurations. The three configurations use approximately 1 Gb of RAM, but have different number of domains. We have used a model with 1600 radial points distributed over 16 domains and $n_{\theta} = 48$ (approx. 70 Gb of RAM) as a reference to calculate the errors.

n_d	n_r	$\mid n_{\theta} \mid$	Virial test	Energy test	$\Delta ho/ ho$	$\Delta R/R$	$\Delta T_{ m eff}/T_{ m eff}$	$\Delta\Omega/\Omega$
8	30	24	1.848e-09	1.192e-03	1.613e-04	5.157e-05	3.338e-04	3.739e-04
16	21	24	1.851e-11	1.726e-04	1.065e-04	4.702e-05	1.215e-04	4.446e-04
32	15	24	3.377e-10	3.447e-05	2.780e-05	2.021e-06	1.736e-05	3.095e-04

2.8. Generating custom output files

The output files generated by ester 1d and ester 2d contain just the minimal information necessary to reconstruct the model. However, sometimes a more detailed output is required. This can be done using ester output. This program reads a template from the standard input and write the result in the standard output. A typical call would be

\$ ester output $model_file < template_file > output_file$

The template file is a regular text file with the following rules:

- Plain text is copied from the template to the output file. It cannot contain the reserved characters \$ and \.
- Line breaks are ignored. To insert a line break in the output file you have to insert a blank line in the template.
- Variables from the model are written in the form \$\{var, fmt\}\, where var is the code for the variable (see table below) and fmt is a valid format for the C function printf (e.g. %d for an integer, %f for float, %e for exponential notation). If fmt is omitted \$\{var\}\$ the variable is written in binary format.

Table 2.1: Non-exhaustive list of variables codes for the model in the template file. Dimensional quantities are in cgs

Code	Description	star1d	star2d
nr	# of radial points	*	*
nth	# of points in latitude		*
ndomains	# of domains	*	*
npts	# of radial points in each domain	*	*
xif	Position of each domain	*	*
nex	# of radial points in the external domain		*
surff	Parameter surff (see above)	*	*
conv	# of convective domains	*	*
Omega	Angular velocity at the equator (rd/s)		*
Omegac	Critical angular velocity $\Omega_c = \sqrt{\frac{GM}{R_e^3}} \; (\text{rd/s})$		*
Omega_bk	Ω/Ω_c		*
X	Hydrogen initial abundance	*	*
Z	Metal abundance	*	*
Xc	Fraction of X in the convective core	*	*
rhoc	Central density (g/cm ³)	*	*
Тс	Central temperature (K)	*	*
pc	Central pressure (cgs)	*	*
M	Mass (grams)	*	*
R	(Polar) Radius (cm)	*	*
Rp	Polar radius (cm)		*
Re	Equatorial radius (cm)		*
L	Luminosity (erg/s)	*	*
M/M_SUN	Mass in solar units	*	*

R/R_SUN	(Polar) Radius in solar units	*	*
Rp/R_SUN	Polar radius in solar units		*
Re/R_SUN	Equatorial radius in solar units		*
L/L_SUN	Luminosity in solar units	*	*
r	Radius (non-dimensional)	*	*
rt	$\partial_{\theta} r$ (non-dimensional)	*	*
rz	$\partial_{\mathcal{E}} r$ (non-dimensional)	*	*
rzt	$\partial_{\zeta\theta}r$ (non-dimensional)	*	*
rzz	$\partial_{\zeta\zeta}r$ (non-dimensional)	*	*
gzz	$g_{\zeta\zeta}$ (non-dimensional)	*	*
gzt	$g_{\zeta\zeta}$ (non-dimensional)	*	*
Z	Radial variable (non-dimensional)	*	*
th	Colatitude (rd) (2D grid)		*
	External radius		*
rex phi	Gravitational potential	*	*
*	1		*
phiex	Gravitational potential of the external domain	*	*
rho	Density (non-dimensional)	*	*
p	Pressure (non-dimensional)	*	*
Т	Temperature (non-dimensional)	^	
W	Angular velocity (rd/s)		*
G	Stream function for the meridional circulation		*
Xr	Hydrogen abundance $X(r,\theta)$ not in use (yet)	*	*
N2	Squared Brunt-Väisälä frequency (in rd ² /s ²)	*	*
opa	Type of opacity	*	*
opa.k	Rosseland mean opacity (cm ² /g)	*	*
opa.xi	Thermal conductivity (χ)	*	*
opa.dlnxi_lnT	$\left(\frac{\partial \log \chi}{\partial \log T} \right)_{ ho, \mu}$	*	*
opa.dlnxi_lnrho	$\left(\frac{\partial \log \chi}{\partial \log \rho}\right)_{T,\mu}$	*	*
eos	Type of equation of state	*	*
eos.G1	Γ_1	*	*
eos.cp	c_p (cgs)	*	*
eos.del_ad	∇_{ad}	*	*
eos.G3_1	$\Gamma_3 - 1$	*	*
eos.cv	c_v (cgs)	*	*
eos.prad	Radiation pressure	*	*
eos.chi_T	$\chi_T = \left(\frac{\partial \log p}{\partial \log T}\right)$	*	*
eos.chi_rho	$v_{-} = \left(\frac{\partial \log p}{\partial \log p}\right)$	*	*
eos.d	$d = \frac{\chi_T}{\chi_\rho} = -\left(\frac{\partial \log \rho}{\partial \log T}\right)_{T,\mu}$	*	*
nuc.eps	Energy generation rate per unit mass (cgs)	*	*
	Energy generation rate per unit mass (cgs) Energy generation rate per unit mass (pp-chain)	*	*
nuc.pp	Energy generation rate per unit mass (pp-chain) Energy generation rate per unit mass (CNO cycle)	*	*
nuc.cno Teff	Surface effective temperature $T_{\text{eff}}(\theta)$ (Kelvin)	*	*
	Surface effective temperature $I_{\text{eff}}(\theta)$ (Keivin) Surface effective gravity surface $g_{\text{eff}}(\theta)$ (cm/s ²)	*	*
gsup	Surface effective gravity surface $g_{\text{eff}}(\theta)$ (cm/s ²)		<u> </u>

D	Radial differentiation matrix $\frac{\partial}{\partial \zeta}$ for 2D models, $\frac{d}{dr}$	*	*
	for 1D models		
I	Radial integration matrix	*	*
Dex	Radial differentiation matrix for the external domain		*
Dt	Angular differentiation matrix $\frac{\partial}{\partial \theta}$ for symmetric vari-		*
	ables		
Dtodd	Angular differentiation matrix for antisymmetric		*
	variables		
Dt2	Second order angular differentiation matrix for sym-		*
	metric variables		
It	Angular integration matrix over $\mu = \cos \theta$		*

For 2D variables, their values at the collocation points are written in the output file in matrix form. Each line corresponds to a different value of the colatitude θ starting at the equator. Namely, if nth=8, we have

```
\begin{array}{lll} \theta_0 & = & 1.47564028, \; \theta_1 = 1.28533144, \; \theta_2 = 1.0950334, \; \theta_3 = 0.90475753, \; \theta_4 = 0.71452525, \\ \theta_5 & = & 0.52438664, \; \theta_6 = 0.33449864, \; \theta_7 = 0.14572468 \end{array}
```

while the values of a field are ordered as

$$\begin{array}{ccccc} p(\zeta_0,\theta_0) & p(\zeta_1,\theta_0) & p(\zeta_2,\theta_0) & \cdots \\ p(\zeta_0,\theta_1) & p(\zeta_1,\theta_1) & p(\zeta_2,\theta_1) & \cdots \\ p(\zeta_0,\theta_2) & p(\zeta_1,\theta_2) & p(\zeta_2,\theta_2) & \cdots \\ \vdots & \vdots & \vdots & \vdots \end{array}$$

 ζ being the radial spheroidal coordinate. Similarly, 1D variables can be seen as a column vector and are written in one line in the output file, terminated by a new line character. This behavior can be inverted by writing this line in the template file

\conf{transpose=1}

After this command, the variables will be written row wise, i.e. one line for each value of the radial coordinate. Note that it does not affect variables written in binary format, which are always column wise. To recover the original behaviour we use

\conf{transpose=0}

The original grid does not contain points at the equator nor at the pole. If we want the values at these points we should write

```
\conf{equator=1}
\conf{pole=1}
```

By default, the output uses cgs units. If we want the normalized values used internally by the code, we simply put

\conf{dim=0}

These control commands can be written anywhere in the template file, in separated lines, affecting only the code that appears below them.

Let's see an example.

Template file:

```
Model of ${M/M_SUN,%.2f} solar masses and R=${R,%e} cm
rotating with Omega=${Omega_bk,%f} Omegac

${nr,%d} radial points and
${nth,%d} latitudinal points

\conf{pole=1}
\conf{equator=1}
r:

${r,%e}
Pressure:

${p,%.14e}
```

Output file:

```
Model of 2.50 solar masses and R=1.219822e+11 cm
rotating with Omega=0.900000 Omegac
240 radial points and 32 latitudinal points
r:
0.000000e+00 4.944313e+07 1.971944e+08 4.415355e+08 7.796539e+08 ...
0.000000e+00 4.944313e+07 1.971944e+08 4.415355e+08 7.796539e+08 ...
0.000000e+00 4.944313e+07 1.971944e+08 4.415354e+08 7.796533e+08 ...
0.000000e+00 4.944313e+07 1.971944e+08 4.415352e+08 7.796523e+08 ...
[...]
Pressure:
1.61049808835808e+17 1.61048890365891e+17 1.61035199104197e+17 ...
1.61049808835808e+17 1.61048890354742e+17 1.6103519927083e+17 ...
1.61049808835808e+17 1.61048890265707e+17 1.61035194697311e+17 ...
[...]
```

2.9. Python module

2.9.1. Using ESTER Native format

A basic interface python module for reading the models is included in the distribution. It is located in ester/python/. The interface between the C/C++ and python variables is based on SWIG. The C++.variables of ESTER are made visible using the interface file ester.in, which can be modified by the (advanced) user if necessary. In principles all fields have been made

available. If the ester.in file is modified then a run of make install in the build directory is necessary.

The variables in the models are defined as *numpy* arrays. Here is a little example extracted from the postprocessing directory where various python scripts are available:

```
import sys
from ester import *  # Loads the module
import matplotlib.pyplot as plt
import numpy as np

A=star1d('model_file') # Loads a model
# or A=star2d('model_file') if you use a 2D-model

print A.p[0,0] # Prints the central pressure

plt.plot(A.r[:],np.log10(A.rho[:]))

show() # Needed in non-interactive mode of matplotlib
```

Other examples are available, especially meridional cuts like the differential rotation (see P_rot.py). Note that "dotted variables" like opa.k are accessed via A.opa_k under python.

2.9.2. Using HDF5

If you installed ESTER with HDF5 (Hierarchical Data Format) support, writing/reading to/from a file ending with .hdf5 or .h5 (-i and -o options) will write/read the file in HDF5 format.

You can visualize these files with a variety of tools such as hdfview or h5dump, and you can easily read them from several languages such as python, Fortran or IDL.

FILE FORMAT

An ESTER's model written in HDF5 will have the following structure:

- a group named star attached to the root of the file
- the parameters of the simulation are attributes attached to this group
 - for instance ndomains or npts
- the different fields are datasets attached to the /star group

Fields present in ESTER's models are:

- \bullet /star/G: the stream function Φ
- /star/N2: the squared Brunt-Visl frequency (in rd^2/s^2).
- /star/R: radius of domain's boundary.
- /star/T: temperature.

- /star/X: hydrogen abundance.
- /star/Y: helium abundance.
- /star/Z: metal abundance.
- /star/nuc.eps: nuclear reaction.
- /star/p: pressure.
- /star/phi: gravitational potential.
- /star/phiex: gravitational potential in the external domain.
- /star/r: radius of collocation points.
- /star/rho: density.
- /star/th: colatitude.
- /star/w: angular velocity (ω).
- /star/z: radial variable (ζ).

Code Examples

Python: You need to have the h5py package installed.

Fortran: You need to compile with the h5fc compiler wrapper.

```
program read_hdf5
    use hdf5
    implicit none

    character*100 :: file_name = "star.hdf5"
    integer status, error
    integer(hsize_t) :: dims(2)
    integer :: nr, nth
    integer(HID_T) :: fid, gid, aid, did
    double precision, allocatable :: T(:,:)
```

```
! init interface
call h5open_f(status)
! open the HDF5 file
call h5fopen_f(file_name, H5F_ACC_RDWR_F, fid, status)
! open the 'star' group
call h5gopen_f(fid, "star", gid, error)
! open the 'nr' attribute
call h5aopen_f(gid, "nr", aid, error)
! read the attribute
dims(1) = 1
dims(2) = 0
call h5aread_f(aid, H5T_NATIVE_INTEGER, nr, dims, error)
! close the attribute
call h5aclose_f(aid, error)
! open the 'nth' attribute
call h5aopen_f(gid, "nth", aid, error)
! read the attribute
dims(1) = 1
dims(2) = 0
call h5aread_f(aid, H5T_NATIVE_INTEGER, nth, dims, error)
! close the attribute
call h5aclose_f(aid, error)
print *, "nr: ", nr
print *, "nth:", nth
! allocate memory for the temperature field
allocate(T(nr, nth))
! open the 'T' dataset
call h5dopen_f(gid, "T", did, error)
! read the field
dims(1) = nr
dims(2) = nth
call h5dread_f(did, H5T_NATIVE_DOUBLE, T, dims, error)
print *, "T at the center: ", T(1, 1)
print *, "T at the equator:", T(nr, 1)
```

```
deallocate(T)
    ! close dataset, group and file
   call h5dclose_f(did, error)
   call h5gclose_f(gid, error)
    call h5fclose_f(fid, error)
end program
IDL:
:-----
fid = h5f_open('star2d05.h5') ; open the file
gid = h5g_open(fid, '/star') ; give the group id to the object star
T = h5d_read(h5d_open(gid, 'T')); in the group id read the field T
r = h5d_read(h5d_open(gid, 'r'))
th = h5d_read(h5d_open(fid, '/star/th'))
; in the group id read the parameters nth:
nth = h5a_read(h5a_open_name(gid, 'nth'))
; in the group id read the parameters nr:
nr = h5a_read(h5a_open_name(gid, 'nr'))
; make a contour plot of the temperature
x=fltarr(nr,nth) & y=fltarr(nr,nth)
for i=0,nr-1 do begin
 for j=0, nth-1 do begin
 x(i,j)=r(i,j)*cos(th(j))
 y(i,j)=r(i,j)*sin(th(j))
 endfor
endfor
print, 'Central temperature: ', T[0, 0]
contour,T,x,y,nlev=40,xtit='r',ytit='z'
h5f_close, fid ; close the file
end
```

More on the use of hdf5 data by idl on http://www.exelisvis.fr/docs/routines-102.html

General structure of the code

3.1. Equations to be solved

3.1.1. WITH DIMENSIONAL VARIABLES

We consider a lonely rotating star in a steady state. The star is governed by the following equations for macroscopic quantities:

$$\Delta \phi = 4\pi G \rho \tag{3.1}$$

$$\rho T \boldsymbol{v} \cdot \nabla s = -\nabla \cdot \boldsymbol{F} + \rho \varepsilon_* \tag{3.2}$$

$$\rho \boldsymbol{v} \cdot \nabla \boldsymbol{v} = -\nabla P - \rho \nabla \phi + \boldsymbol{F}_{v} \tag{3.3}$$

$$\nabla \cdot (\rho \mathbf{v}) = 0 \tag{3.4}$$

which need to be completed by the equations of microphysics:

$$\begin{cases}
P \equiv P(\rho, T) \\
\chi_r \equiv \chi_r(\rho, T) \\
\varepsilon_* \equiv \varepsilon_*(\rho, T)
\end{cases}$$
(3.5)

and the expressions of the viscous force, which could be (for instance)

$$\boldsymbol{F}_{v} = \mu(\Delta \boldsymbol{v} + \frac{1}{3}\nabla\nabla\cdot\boldsymbol{v}) \tag{3.6}$$

for a compressible, constant viscosity fluid, and of the heat flux

$$\mathbf{F} = -\chi_r \nabla T - \frac{\chi_{\text{turb}} T}{\mathcal{R}_M} \nabla s \tag{3.7}$$

where χ_{turb} is the turbulent diffusion of heat and s the entropy. χ_r is the radiative conductivity (usually dominating over the collision one).

This set of equations is completed by boundary conditions (discussed below).

3.1.2. Simplifications

We simplify the system of equations by first neglecting entropy advection by meridional circulation, which is justified at low viscosity (e.g. Rieutord 2006 or Rieutord et al. 2016). We also neglect the convective flux: thus we avoid computing stars with an outer convective envelope where the convective flux is non-negligible. Core convection is simplified in assuming an isentropic core. Hence, the energy/entropy equation just reads:

$$-\nabla \cdot \mathbf{F} + \rho \varepsilon_* = 0 \tag{3.8}$$

in a radiative zone and

$$\nabla s = \mathbf{0} \tag{3.9}$$

in the convective core. This equation gives the temperature gradient as a function of the pressure gradient thanks to the thermodynamics relation

$$ds = \frac{c_p}{T}dT - \frac{\alpha_t}{\rho T}dP$$

where

$$\alpha_t = -\left(\frac{\partial \ln \rho}{\partial \ln T}\right)_P \tag{3.10}$$

is the isobar expansion coefficient. Hence, (3.9) leads to

$$\nabla T = \frac{\alpha_t}{\rho c_p} \nabla P \tag{3.11}$$

which also means that

$$\nabla_a = \left(\frac{\partial \ln T}{\partial \ln P}\right)_s = \frac{P\alpha_t}{\rho c_p T} \tag{3.12}$$

As for the momentum equation, we split it into its azimuthal and meridional components. The meridional components of the equation may be reduced to

$$\rho s \Omega^2 \mathbf{e}_s = \nabla p + \rho \nabla \phi \tag{3.13}$$

where s is the cylindrical radial coordinate (e.g. Espinosa Lara & Rieutord, 2013, hereafter referred to as ELR). The vorticity equation reduces to

$$s\frac{\partial\Omega^2}{\partial z} = \boldsymbol{e}_{\varphi} \cdot \frac{\nabla p \times \nabla \rho}{\rho^2} . \tag{3.14}$$

In these equation, the advection of the 'meridian momentum' has been neglected in view of the smallness of the meridianal flow.

The meridian circulation is important in the advection of angular momentum as it balances its diffusion by viscosity. So the e_{φ} component of the momentum equation reads:

$$\nabla \cdot (\rho s^2 \Omega \boldsymbol{u}) = \nabla \cdot (\mu s^2 \nabla \Omega) \tag{3.15}$$

where u is the meridional circulation and μ the dynamical viscosity.

3.1.3. Scaled equations

First step is to move scaled equations with scaled quantities. We choose to scale pressure, density and temperature by their central values and other quantities as follows:

Length scale \equiv polar radius	?
Pressure scale \equiv central pressure	c
Density scale \equiv central density	c
Temperature scale \equiv central temperature	c
Gravitational potential scale $\frac{P_{\theta}}{\rho_{\theta}}$	2
Angular velocity scale $\frac{1}{R}\sqrt{\frac{P_c}{\rho_c}}$	_
Meridional velocity scale $E\sqrt{\frac{P_0}{\rho_0}}$	2

where E is the Ekman number defined as:

$$E = \frac{\mu_c}{\rho_c \Omega_0 R^2} \quad \text{with} \quad \Omega_0 = \sqrt{\frac{P_c}{R^2 \rho_c}} . \tag{3.16}$$

With these scalings, the Poisson equation now reads:

$$\Delta \phi = \pi_c \rho \tag{3.17}$$

where

$$\pi_c = \frac{4\pi G \rho_c^2 R^2}{P_c} \tag{3.18}$$

Energy equation can be written

$$\Delta T + \nabla \ln \chi_r \cdot \nabla T + \Lambda \rho \frac{\varepsilon_*}{\chi_r} = 0 \tag{3.19}$$

where

$$\Lambda = \frac{\rho_c R^2}{T_c} \tag{3.20}$$

is a dimensional constant since ε_* and χ_r are dimensional.

The momentum equation leads to

$$\rho s \Omega^2 \mathbf{e}_s = \nabla p + \rho \nabla \phi \tag{3.21}$$

and the vorticity equation to

$$s\frac{\partial\Omega^2}{\partial z} = \mathbf{e}_{\varphi} \cdot \frac{\nabla p \times \nabla \rho}{\rho^2} \ . \tag{3.22}$$

while angular momentum flux balance reads

$$\nabla \cdot (\rho s^2 \Omega \boldsymbol{u}) = E \nabla \cdot (\mu s^2 \nabla \Omega) \tag{3.23}$$

Mass conservation remains the same

$$\nabla \cdot (\rho \mathbf{u}) = 0 \tag{3.24}$$

3.1.4. Boundary conditions

Before presenting the boundary conditions, we should define the surface of the star: we take as its definition, the isobar where the polar pressure is

$$P_s = \tau_s \frac{g_{\text{pole}}}{\kappa_{\text{pole}}},\tag{3.25}$$

On this isobar, $T = T_{\text{eff}}$ only at the pole. This definition permits a smooth continuity with the non-rotating models. This surface will be associated with the value $\zeta = 1$ of the pseudo-radial coordinate (see the chapter on the mapping).

• On the gravitational potential: regularity at the centre of the star and vanishing at infinity. However, imposing this condition on the surface of the star is cumbersome and leads to problem of convergence for very flattened stars. Different solutions to this problem are possible. We can encompass the star within an empty domain whose outer boundary is a sphere. On this outer sphere the boundary conditions on spherical harmonics components of the gravitational potential are simply

$$\frac{\partial \phi_{\ell}}{\partial r} + \frac{(\ell+1)\phi_{\ell}}{r} = 0$$

which ensure the matching with a field vanishing at infinity. This solution was applied successfully in the first versions of the code but is not appropriate when boundary conditions are implemented in real (θ) space. We now prefer using an outer domain $\zeta \in [1, +\infty[$ that is mapped to [0, 1[by an appropriate change of variable (see chap. 6) so as to simply set the potential to zero on the outer boundary of the encompassing outer domain.

• On the velocity, we demand stress-free conditions, namely

$$\boldsymbol{v} \cdot \boldsymbol{n} = 0$$
 and $([\sigma]\boldsymbol{n}) \wedge \boldsymbol{n} = \boldsymbol{0}$

where $[\sigma]$ is the stress tensor. Stars are in the limit of small Ekman numbers and it is interesting (numerically) not to have to compute the ensuing Ekman layer. However, it is necessary to take this layer into account for computing the azimuthal velocity, which is otherwise undefined (see ELR). ELR have shown that the effect of the Ekman layer on the interior flow can be mimicked by the boundary condition:

$$\mu s^2 \hat{\boldsymbol{\xi}} \cdot \nabla \Omega + \psi \hat{\boldsymbol{\tau}} \cdot \nabla (s^2 \Omega) = 0 \quad \text{on the surface} . \tag{3.26}$$

where $\hat{\boldsymbol{\xi}}$ is a unit vector perpendicular to the surface while $\hat{\boldsymbol{\tau}}$ is tangent to it. ψ is the stream function of the meridional flow.

• The rotation speed of the star must be specified. For this we impose the equatorial angular velocity as a fraction of the critical angular velocity, namely:

$$\Omega(r = R_{\rm eq}, \theta = \pi/2) = \omega_k \sqrt{\frac{GMR^2 \rho_c}{P_c R_{\rm eq}^3}}$$

where ω_k is chosen by the user. Another way of specifying the angular velocity of the star is to impose its total angular momentum (see integral constraints).

• On temperature: with the adopted definition of the stellar surface and thanks to a simple model described in ELR, we can ascribe to the surface of the star a temperature profile, such that

$$T_b(\theta) = \left(\frac{g_{\text{pole}}}{g_{\text{eff}}(\theta)} \frac{\kappa(\theta)}{\kappa_{\text{pole}}}\right)^{1/(n+1)} \left(\frac{-\chi_r \boldsymbol{n} \cdot \nabla T}{\sigma}\right)^{1/4} . \tag{3.27}$$

where the polytropic index n is set to n = 3. So the temperature boundary condition are simply:

$$T(0) = 1$$
 and $T(\zeta = 1, \theta) = T_b(\theta)/T_c$

3.1.5. Integral constraints

• Mass is an input parameter that is related to the preceding quantities by

$$M = \rho_c R^3 \int_{(V)} \rho dV = \rho_c R^3 m \tag{3.28}$$

• Angular momentum is another integral quantity that is useful to monitor or to impose:

$$L_z = \rho_c R^4 \sqrt{\frac{P_c}{\rho_c}} \int_{(V)} r^2 \sin^2 \theta \Omega \rho dV$$
 (3.29)

3.1.6. The mapping

The foregoing equations and boundary conditions should be completed by the definition of the mapping that maps the spheroidal star to a sphere. This thorny subject is described at length in a separate chapter (see chap. 6).

3.2. The algorithm

3.2.1. Discretization

This system of partial differential equation is solved using a spectral method or more precisely spectral elements. The star divided in "onion" layers where the vertical direction is discretized on the Gauss-Lobatto collocation grid (associated with Chebyshev polynomials) and the Gauss-Legendre collocation grid for the horizontal dependence. This is associated with Legendre polynomials (or axisymmeric spherical harmonics).

Let us recall the ordering of pressure discretized values, namely

$$\begin{array}{cccc} p(\zeta_0,\theta_0) & p(\zeta_1,\theta_0) & p(\zeta_2,\theta_0) & \cdots \\ p(\zeta_0,\theta_1) & p(\zeta_1,\theta_1) & p(\zeta_2,\theta_1) & \cdots \\ p(\zeta_0,\theta_2) & p(\zeta_1,\theta_2) & p(\zeta_2,\theta_2) & \cdots \\ \vdots & \vdots & \vdots & \vdots \end{array}$$

Scalar fields can therefore be represented by a 2D matrix

$$p_{ij}$$

where the first index represents the radial variations and the second the angular variations. Hence, operators acting on ζ apply on "left" while operators acting on θ apply on right of the matrix [p]. Hence, we shall write

$$[\partial_{\zeta\theta}p] = [D_{\zeta}][p][D_{\theta}]$$

or

$$(\partial_{\zeta\theta}p)_{ij} = (D_{\zeta})_{ik}p_{kl}(D_{\theta})_{lj}$$

3.2.2. Iterations

We solve the resulting discretized equations using Newton's iterative scheme, namely, if we write the problem in the form

$$F(x) = 0, (3.30)$$

where the vector function F represents the equations that we want to solve and x is the vector containing all the independent variables of the problem (pressure, temperature, ...) including the shape of the surface which is not known a priori. The equations are linearized using the Jacobian matrix \mathcal{J} of F(x) such that

$$\delta \mathbf{F}(\mathbf{x}) = \mathcal{J}(\mathbf{x})\delta \mathbf{x}.\tag{3.31}$$

The correction to the solution $\delta x^{(k)}$ at the k-th iteration is calculated solving the linear system

$$\mathcal{J}(\boldsymbol{x}^{(k)})\delta\boldsymbol{x}^{(k)} = -\boldsymbol{F}(\boldsymbol{x}^{(k)}) \tag{3.32}$$

and the solution is updated with $\boldsymbol{x}^{(k+1)} = \boldsymbol{x}^{(k)} + \delta \boldsymbol{x}^{(k)}$.

With an appropriate initial approximation $x^{(0)}$, Newton's method has quadratic convergence. In practice, a rotating stellar model can be calculated in approximately 10 iterations starting with the corresponding non-rotating model.

3.3. Implementation

3.3.1. The vector

The solution vector \boldsymbol{x} is built from the discretized variables as follows Real ordering of variables:

"Phi"; "log_p"; "pi_c"; "log_T"; "Lambda"; "eta"; "deta"; "Ri"; "dRi"; "Omega"; "log_pc"; "log_Tc"; "log_R"; "m"; "ps"; "Ts"; "lum"; "Frad"; "Teff"; "gsup"; "w"; "G"; "gamma";

Note that this list is completed by a list of secondary variable (e.g. Tab. 3.2). These variables are, in the end, substituted by their expression in the final jacobian. They are used to facilitate the programming. They are declared as regvar_dep. However, declared regvar variables can be used as shortcuts when they do not increase the size of the Jacobian too much. These are ps and Ts for instance (see the equation for atmosphere solve_atm).

Variable	Dependence	C++ quantity	Comments
ϕ	$(\zeta, heta)$	Phi	Gravitation potential
$\ln P$	(ζ, θ)	\log_{-p}	Natural log of pressure
$\ln T$	(ζ, θ)	$\log_{-}T$	Natural log of temperature
w	(ζ, θ)	W	Differential Rotation
ψ	(ζ, θ)	G	Stream function
$T_{ m eff}$	θ	Teff	Effective temperature
$g_{ m sup}$	heta	gsup	Effective surface gravity
P_s	heta	ps	surface pressure
T_s	heta	Ts	surface temperature
γ	heta	gamma	$\gamma = \sqrt{g^{\zeta\zeta}} = \sqrt{1 + r_{\theta}^2/r^2}$ at the surface
$F_{\rm rad,i}$	(θ, i)	Frad	F_{ζ} Radiative flux component at each domain boundary
R_i	(θ,i)	Ri	boundaries of the domains
dR_i	(θ,i)	dRi	$R_{i+1} - R_i$
η_i		eta	Polar radius of the domains
$d\eta_i$		$\det a$	$\eta_{i+1} - \eta_i$
π_c		$\mathrm{pi}_{-}\mathrm{c}$	Non-dimensional central pressure
Λ		Lambda	Dimensional constant
Ω		Omega	Equatorial angular velocity
$\ln P_c$		\log_{-pc}	log of central pressure
$\ln T_c$		$log_{-}Tc$	log of central temperature
$\log R$		$\log_{-}R$	log of polar radius
m		m	non-dimensional mass
Lum		lum	Luminosity

Table 3.1: This table lists all the primary variables contained in \boldsymbol{x} but not in their actual order in the code.

C++ Variable	Dependence (C++ quantity)	Comments
${ m T}$	$\log_{-}T$	Temperature
\mathbf{r}	$R_i, dR_i, \eta_i, d\eta_i$	radius
rz	$R_i, dR_i, \eta_i, d\eta_i$	$\partial_\zeta r$
\log_rhoc	\log_{-pc} , \log_{-Tc}	log of central density
rho	$\log_{-}T, \log_{-}p$	density
opa.xi	$\log_{-}T, \log_{-}p$	thermal conductivity
opa.k	$\log_{-}T, \log_{-}p$	opacity
nuc.eps	$\log_{-}T, \log_{-}p$	nuclear heating
S	$\log_{-}T, \log_{-}p$	entropy

Table 3.2: This table lists all the secondary variables. They depend on the primary variables and are used as short-cut in the programming.

3.3.2. The equations

• Initialization fill

Uses (3.28) and that

$$m = 2\pi \int_0^1 \int_0^{\pi} \rho r^2 r_{\zeta} d\zeta \sin\theta d\theta$$

Set the definition of π_c as (3.18), and the definition of Ω_c , namely the scaled keplerian velocity. We have

$$\Omega_c = \frac{\Omega_k}{\Omega_0}$$
 with $\Omega_k = \sqrt{\frac{GM}{R_e^3}}$

and $\Omega_0^2 = P_c/\rho_c/R^2$.

• Constitutive relations solve_definitions

Secondary variables are used to simplify the writing of the Jacobian matrix.

The EOS is written as the dependence of ρ_* with respect to P_* and T_* (* quantities are dimensional). Thus we have:

$$\delta \rho_* = \frac{\rho_*}{P_*} \left(\frac{\partial \ln \rho}{\partial \ln P} \right)_T \delta(P_c P) + \frac{\rho_*}{T_*} \left(\frac{\partial \ln \rho}{\partial \ln T} \right)_P \delta(T_c T)$$

After some manipulations, we get:

$$\frac{\delta \rho}{\rho} = \frac{1}{\chi_{\rho}} \delta \ln P + \frac{1}{\chi_{\rho}} \delta \ln P_c - \delta \ln \rho_c - d_p \delta \ln T_c - d_p \delta \ln T$$

where we introduced

$$\chi_{\rho} = \left(\frac{\partial \ln P}{\partial \ln \rho}\right)_T$$
 and $d_p = -\left(\frac{\partial \ln \rho}{\partial \ln T}\right)_P = \chi_t/\chi_{\rho}$

following the definitions of Rogers et al. (1996) and using the thermodynamic triangular relation.

Opacity and nuclear heat production depend on ρ and T. So for both of these quantities we have the formal relation $f = f(\rho, T)$. For instance, for opacity, denoted κ (denoted opa.k in the code) we get

$$\delta\kappa = \frac{\partial\kappa}{\partial\rho_*}\delta\rho_* + \frac{\partial\kappa}{\partial T_*}\delta T_*$$

Noting that $\rho_* = \rho_c \rho$ and $T_* = T_c T$, after some manipulations we get:

$$\delta \kappa = \frac{\kappa}{\rho} \frac{\partial \ln \kappa}{\partial \ln \rho} \delta \rho + \kappa \frac{\partial \ln \kappa}{\partial \ln \rho} \delta \ln \rho_c + \kappa \frac{\partial \ln \kappa}{\partial \ln T} \delta \ln T + \kappa \frac{\partial \ln \kappa}{\partial \ln T} \delta \ln T_c$$

For radiative conductivity (denoted opa.xi in the code), its expression

$$\chi_r = \frac{16\sigma T^3}{3\rho\kappa}$$

leads to the following contribution

$$\frac{\delta \chi_r}{\chi_r} = 3\delta \ln T + 3\delta \ln T_c - \frac{1}{\rho} \delta \rho - \delta \ln \rho_c - \frac{1}{\kappa} \delta \kappa$$

where we took into account that ρ and T are non-dimensional.

As far as entropy is concerned, we express its dependence with temperature and pressure:

$$\delta s = \frac{c_p}{T_*} \delta T_* + \left(\frac{\partial s}{\partial P_*}\right)_T \delta P_*$$

We note that

$$\left(\frac{\partial s}{\partial P_*}\right)_T = -\frac{\alpha}{\rho} = -\frac{c_p \nabla_a}{P}$$

where we used (3.10) and (3.12). Thus, after some manipulations we find

$$\delta s = c_p \left(\delta \ln T_c + \delta \ln T - \nabla_a \delta \ln P_c - \nabla_a \delta \ln P \right)$$

• Poisson equation solve_poisson

Poisson equation reads

$$\Delta\phi = \pi_c \rho \qquad \text{with} \quad \pi_c = \frac{4\pi G \rho_c^2 R^2}{p_c}$$

so that its variational expression is

$$\Delta\delta\phi + (\delta\Delta)\phi - \rho\delta\pi_c - \pi_c\delta\rho = -(\Delta\phi - \pi_c\rho)_n$$

where $(\delta\Delta)$ represents the variation of the Laplacian operator due to the change of the metric. Note that the last domain is void of matter.

Together with Poisson equation, we need to impose the boundary conditions and the interface conditions at each domain boundary. The boundary conditions are

$$\nabla \phi(\mathbf{0}) = \mathbf{0}$$
 and $\phi(\infty) = 0$

The first condition is equivalent to $\partial_{\zeta} \phi = 0$.

At the interface of the domains we demand that ϕ and $n \cdot \nabla \phi$ are continuous. The continuity of the normal component of the gradient implies the continuity of

$$\frac{1}{r_\zeta}\frac{\partial \phi}{\partial \zeta}$$

Even though we have chosen a continuous r_{ζ} -field, we keep the dependence of the interface condition with r_{ζ} in case we would like to test other kinds of mapping (see chapter 6 for more details).

• Momentum equation solve_mov

We start from the meridional components of the (inviscid) momentum equation (3.21), (3.22) and (3.23). For the angular momentum equation (3.23), we impose boundary condition (3.26).

The equation are solved for the log of the pressure. At the domain interfaces we impose the continuity of pressure, while at the center the non-dimensional pressure is set to unity.

Let us discuss the implementation of the vorticity equation. We first rewrite (3.21) as

$$\nabla p + \rho \nabla \phi - \rho s \Omega^2 \mathbf{e}_s = \mathbf{0}$$

and introduce the eqmov equation as

from which we derive the equation of vorticity (3.22), symbolically name eq_vort, and defined as

which means the scalar product of e_{φ} with the curl of the equation eqmov by ρ .

At the interface of the domain, since we are working in the inviscid limit, Ω is not necessarily continuous. It is continuous only if ρ is continuous (see discussion in Espinosa Lara & Rieutord 2013). In general we only know that pressure is continuous and so is its horizontal derivative. Thus the continuity of Ω is ruled by the continuity of the horizontal component of $\rho \nabla \phi - \rho s \Omega^2 e_s$ as explained by Eq. (63) in Espinosa Lara & Rieutord (2013). Thus the programme introduces:

```
ic_w=covariant(eqmov-grad(p))(1);
```

namely the θ (as 1) covariant component of $\rho \nabla \phi - \rho s \Omega^2 \mathbf{e}_s$, since \mathbf{E}_{θ} is parallel to $\zeta = \mathrm{Cst}$ surfaces.

The vorticity equation, which is called "w", is then plugged into the big Jacobian matrix by just saying that functional variations are on variables "p", "w", "rho" and "r" like

Equation "w" is completed by the interface and boundary conditions that say that:

$$\partial_{\zeta}\Omega(\zeta=0)=0$$
, at center

$$\mu s^2 \mathbf{n} \cdot \nabla \Omega + G \mathbf{t} \cdot \nabla (s^2 \Omega) = 0$$
, at surface
$$\Omega(\zeta = 1, \theta = \pi/2) = \Omega_e$$

where G is the stream function ψ in (3.26). The central boundary condition reads:

```
op->bc_bot2_add_1(0, "w", "w", ones(1, nth), D.block(0).row(0));
```

Then we loop on the domains to impose the continuity of ic_w

```
for(int n=0;n<ndomains-1;n++) {
    ic_w.bc_top1_add(op,n,"w","w");
    ic_w.bc_top1_add(op,n,"w","rho");
    ic_w.bc_top1_add(op,n,"w","Phi");
    ic_w.bc_top1_add(op,n,"w","r");
    ic_w.bc_top2_add(op,n,"w","w",-ones(1,nth));
    ic_w.bc_top2_add(op,n,"w","rho",-ones(1,nth));
    ic_w.bc_top2_add(op,n,"w","Phi",-ones(1,nth));
    ic_w.bc_top2_add(op,n,"w","Phi",-ones(1,nth));
    ic_w.bc_top2_add(op,n,"w","r",-ones(1,nth));
    rhs.setrow(j0+map.gl.npts[n]-1,
-ic_w.eval().row(j0+map.gl.npts[n]-1)+ic_w.eval().row(j0+map.gl.npts[n]));
    j0+=map.gl.npts[n];
}</pre>
```

Then we impose the condition on the value of the velocity at equator and the special BC that suppresses the Ekman layer: (3.26) is inserted as

```
bc=mu*s*s*(nvec,grad(w))+G*(tvec,grad(s*s*w));
```

and plugged in the BC of the last domain. So the outer boundary condition on Ω imposes the angular velocity at equator. Since Ω obeys a first order equation the central condition and its fixed equatorial value determines Ω everywhere provided we lift the degeneracy $\Omega^2 \to \Omega^2 + f(s)$ where f is arbitrary. This is done by (3.26) which also provides the second BC for G the stream function which obeys a second order PDE. Hence, noting that $\Omega_{\rm eq}$ is a linear function of the values of Ω on the latitudinal grid, BC (3.26) is imposed on all the grid points except the first one where the equatorial angular velocity is imposed: hence,

Of course the equation for Ω is strongly coupled to that of the meridional circulation that is represented by the stream function G, which is actually the stream function of ρv . Hence, one introduces

```
V=curl(G*phivec)/rho;
```

namely the meridional circulation, and the angular momentum equation:

```
eq_phi=div(rho*s*s*w*V)-div(mu*s*s*grad(w));
```

Again the equation of angular momentum is plugged into the Jacobian matrix by:

```
eq_phi.add(op,"G","w");
eq_phi.add(op,"G","G");
eq_phi.add(op,"G","rho");
eq_phi.add(op,"G","r");
op->set_rhs("G",-eq_phi.eval());
```

Then it is indicated that G is zero at the center and continuous at the interfaces. For non-dimensional parameters like π_c , it is continuous at the domain interface:

$$1 \times \delta \pi_c(1) - 1 \times \delta \pi_c(2) = 0$$

"1" means "part 1 of the interface condition" and "2" means "part 2 of the interface condition" (see Fig. 3.1).

• Temperature equation solve_temp

First, the luminosity at each domain boundary is computed through

$$L_n = 2\pi\Lambda \int_0^{\pi} \int_0^{\eta_n} \rho \varepsilon_* r^2 r_{\zeta} \sin\theta d\theta d\zeta$$

The definition of the luminosity

$$L = \Lambda \int_{(V)} \rho \varepsilon_* dV$$
 with $\Lambda = \frac{\rho_c R^2}{T_c}$

is differentiated and yields

$$\delta L - \Lambda \int_{(V)} (\delta \rho \varepsilon_* + \rho \delta \varepsilon_*) dV - \delta \Lambda \int_{(V)} \rho \varepsilon_* dV - \Lambda \int_{(V)} \rho \varepsilon_* \delta dV = 0$$

with L continuous at domain interfaces.

Let us examine the coding of one of the terms of this equation, namely

$$-\Lambda \int_{(V)} \delta \rho \varepsilon_* dV = -\Lambda 2\pi \int_0^{\pi} \sin \theta d\theta \int_{r_-}^{\eta_{n+1}} \delta \rho \varepsilon_* r^2 r_\zeta d\zeta$$

This terms needs an operation on the θ -coordinate and on the ζ coordinate. Recalling that left indices are for the ζ_i grid and right indices are for the θ_j grid the discretized form of this term is like

$$\sum_{l} \left(\sum_{l} (D_{ij} * L_{ik} * (\varepsilon_* r^2 r_\zeta)_{kl} * R_{lj}) \right)$$

where D is just a scalar function, here $D = -2\pi\Lambda$, L is the left-operator that does the integration over ζ and R the right operator that does the integration over θ . In terms of coding it reads:

```
bc_bot2_add_lri(n,"lum","rho",-2*PI*Lambda*ones(1,1),
map.gl.I.block(0,0,j0,j0+map.gl.npts[n]-1),map.leg.I_00,
  (r*r*rz*nuc.eps).block(j0,j0+map.gl.npts[n]-1,0,-1));
```

with

$$\begin{array}{lll} \textbf{D} & \equiv & -2*PI*Lambda*ones(1,1)\\ \textbf{L} & \equiv & \texttt{map.gl.I.block}(0,0,j0,j0+\texttt{map.gl.npts}[n]-1)\\ \textbf{R} & \equiv & \texttt{map.leg.I}_{00}\\ \textbf{I} & \equiv & (r*r*rz*nuc.eps).block(j0,j0+\texttt{map.gl.npts}[n]-1,0,-1)) \end{array} \eqno(3.33)$$

Then we introduce

$$F_{\rm rad} = -\chi_r \mathbf{E}^{\zeta} \cdot \nabla T$$

so the projection of the radiative flux on E^{ζ} , the normal direction of the ζ = Cst surfaces (χ_r is the dimensional thermal radiative conductivity). The basic property of this scalar quantity is that it gives the radiative luminosity when integrated over a surface, namely

$$L = \int_{(S)} F_{\rm rad} 2\pi r^2 r_{\zeta} \sin \theta d\theta$$

Explicit expression of $F_{\rm rad}$ (a comfort equation, see 3.3.3) is

$$F_{\rm rad} = -\chi_r \left(g^{\zeta\zeta} \partial_{\zeta} T + g^{\zeta\theta} \partial_{\theta} T \right)$$

so that the functional derivative is given by

$$\delta F_{\rm rad} + \frac{F_{\rm rad}}{\chi_r} \delta \chi_r + \chi_r \left(g^{\zeta\zeta} \partial_{\zeta} \delta T + g^{\zeta\theta} \partial_{\theta} \delta T + \delta g^{\zeta\zeta} \partial_{\zeta} T + \delta g^{\zeta\theta} \partial_{\theta} T \right) = 0$$

where we have

$$\delta g^{\zeta\zeta} = -2r(g^{\zeta\theta})^2 \delta r - \frac{2g^{\zeta\zeta}}{r_{\zeta}} \delta r_{\zeta} - \frac{2g^{\zeta\theta}}{r_{\zeta}} \delta r_{\theta} \quad \text{and} \quad \delta g^{\zeta\theta} = -\frac{2g^{\zeta\theta}}{r} \delta r - \frac{g^{\zeta\theta}}{r_{\zeta}} \delta r_{\zeta} - \frac{\delta r_{\theta}}{r^2 r_{\zeta}} \delta r_{\zeta}$$

Then we introduce the masks of the convective zones (CZ) qconv and of the radiative zones (RZ) qrad.

$$egin{array}{lll} \mathtt{qconv}(m{r}) &=& 1 & & \mathrm{if} & m{r} \in \mathrm{CZ} \ & \mathtt{qrad}(m{r}) &=& 1 & & \mathrm{if} & m{r} \in \mathrm{RZ} \ \end{array}$$

and of course qconv+qrad=1 $\forall r$.

Hence, the temperature equation can be written

$$q_{\text{conv}}(\boldsymbol{n} \cdot \nabla s) + q_{\text{rad}} \left(\frac{\nabla \cdot (\chi_r \nabla T)}{\chi_r} + \Lambda \rho \frac{\varepsilon_*}{\chi_r} \right) = 0$$
 (3.34)

where n is the unit vector opposite to local effective gravity. Note that this expression of the temperature equation assumes that the temperature profile of the convective region is such that entropy is constant along a vertical. This assumes complete vertical mixing and no horizontal mixing. If the core is convective, the equation is equivalent to saying that the entropy is the same all over the core and $n \cdot \nabla s$ can be replaced by $\partial_{\zeta} s$. Equation (3.34) is then differentiated functionally.

Co	nvective	9		Radiative					Convective			Radiative			
T=1	Т	Т	Т	∂Τ	Т	∂Τ	Т	∂Τ	Т	Т	Т	F	∂Τ	Т	$T=T_s$
Λ	Λ	Λ	F		Λ		Λ		Λ	Λ		Λ		Λ	

	Radiative Convective					Radiative											
T=1	∂T	Т	∂T	Т	∂T	T	T	T	Τ	F	∂T	Т	∂T	T	∂T	Τ	$T=T_s$
∂ T=0)	Λ		Λ		Λ	Λ	Λ		Λ		Λ		Λ		Λ	

Table 3.3: Two tables showing the way boundary and interface conditions are imposed in the different domains. The left position of a symbol means the continuity is imposed at the bottom of the domain, right is for Top. Λ is for the continuity of the Λ constant, while F means that $L = \int_{(S)} F_{\rm rad} dS$. T and ∂T are for the continuity of T and heat flux.

Convective	Radiative					Conv	Convective Radiative															
0	1		2		3		4		5		6		7		8		9		10		11	
T=1	Т	∂Т	Т	∂T	T	θТ	Т	θТ	T	T	F	∂T	Т	θТ	Т	∂T	Т	θТ	T	θТ	T	$T=T_s$
Λ	F		Λ		Λ		Λ		Λ		Λ		Λ		Λ		Λ		Λ		Λ	

Table 3.4: A case that is working (useful for calibration....).

Then, we need to impose the interface conditions which reduce to demanding the continuity of T and $\partial_{\zeta}T/r_{\zeta}$. If the interface separates a CZ and a RZ, then $\partial_{\zeta}T$ is not continuous (in a CZ temperature is derived from entropy and therefore, if pressure is known, temperature is known; if entropy is assumed constant there is no differential equation and imposing the continuity of the fields is sufficient).

Note that the continuity of $\partial_{\zeta} T/r_{\zeta}$ is equivalent to the continuity of $F_{\rm rad}$ (provided there is no chemical jump) since our mapping insures the continuity of r_{ζ} . In the last domain the boundary condition imposes that $T = T_s$.

The implementation of the boundary and interface conditions is a bit tricky especially if convective layers appear in the middle. If we follow the propagation of the information (a bit strange for an elliptic equation like the temperature one!), at the bottom of radiative layer lying above a convective layer, the temperature gradient is actually only defined by the total flux (if the boundary between radiative and convective zones is allowed to move then the temperature gradient converges to the adiabatic one). This is of course the 1D-picture. It is summarized by the two tables that present two explicit cases.

In 2d the problem is more delicate and we are not able to impose an arbitrary position to the convection zone, otherwise we cannot distribute the flux in latitude (to be checked!)...

• Surface effective gravity solve_gsup

Comfort equation (e.g. 3.3.3): Effective gravity is defined as

$$\mathbf{g}_{\text{eff}} = -\nabla \phi + s\Omega^2 \mathbf{e}_s$$

Thus from the simplified meridional momentum equation (3.21), we get

$$oldsymbol{g}_{ ext{eff}} = rac{1}{
ho}
abla p$$

The variable gsup is defined as $g_s = -\mathbf{n} \cdot \mathbf{g}_{\text{eff}}$ so as to be positive. Thus, from the equation

$$g_s = \frac{P_c}{\rho_c R} \frac{-\boldsymbol{n} \cdot \nabla p}{\rho}$$

We get the following functional equation

$$\delta g_s - g_s \delta \ln P_c + g_s \delta \ln \rho_c + g_s \delta \ln R + \frac{g_s}{\rho} \delta \rho + \frac{P_c}{R \rho_c \rho} \delta(\boldsymbol{n} \cdot \nabla P) = 0$$
 (3.35)

• Surface effective temperature solve_Teff

This is a new comfort equation (e.g. 3.3.3). Effective temperature is defined as

$$\sigma T_{c}^{4} = -\chi \boldsymbol{n} \cdot \nabla T$$

Showing the dimensional and non-dimensional variables we rewrite the previous equality as:

$$\sigma T_e^4 = -\chi \boldsymbol{n} \cdot \nabla T \frac{T_c}{R}$$

Taking the log and making the functional derivative, we get

$$4T_e^3 \delta T_e - T_e^4 \delta \ln T_c + T_e^4 \delta \ln R - \frac{T_e^4}{\chi} \delta \chi + \frac{\chi T_c}{\sigma R} \delta(\boldsymbol{n} \cdot \nabla T) = 0$$
(3.36)

• Atmosphere solve_atm

The atmosphere of the star is described in a simple manner by a polytropic atmosphere as described in Espinosa Lara & Rieutord (2013) after Eq. (9). It follows that the surface temperature $T_s(\theta)$ is derived from the pressure field by the polytropic relation $P \propto T^{n+1}$. Namely,

$$T_s(\theta) = T_e \left(\frac{P_s}{P_e}\right)^{1/(n+1)} \tag{3.37}$$

In this expression P_s is the pressure of the isobar on which the boundary conditions are taken. In Espinosa Lara & Rieutord (2013), it is assumed to be the pressure at the pole given by $\tau_s g_{\text{pole}}/\kappa_{\text{pole}}$ but higher values are possible. P_e is the pressure at the top of the atmosphere, where presumably $T = T_{\text{eff}}$ and $P_e = \tau_s g_{\text{eff}}/\kappa_e \sim \tau_s g_{\text{eff}}/\kappa_s$.

In the simple model, $P_s = \tau_s g_s / \kappa_s$, which is translated as

$$\ln P_s + \ln P_c = \ln \tau_s + \ln g_s - \ln \kappa_s$$

assuming that τ_s is a fixed quantity, then the variational equation is

$$\delta P_s/P_s + \delta \ln P_c - \delta g_s/g_s + \delta \kappa_s/\kappa_s = 0 \tag{3.38}$$

while for the temperature of the last bounding surface, we translate (3.37) into

$$\ln T_s + \ln T_c = \ln T_e + \frac{1}{n+1} (\ln P_s - \ln P_e)$$

and gives the variational equation

$$\delta T_s/T_s + \delta \ln T_c - \delta T_e/T_e - \frac{1}{n+1} (\delta P_s/P_s + \delta P_e/P_e) = 0$$
(3.39)

where P_e is approximated by P.

These two equations are also "comfort" equations.

3.3.3. Comfort equations

Basically, some quantities can be introduced to simplify the writing of equations. For instance, the use the surface effective gravity is convenient. Since it plays a role at the surface only, we can introduce this quantity as a regular variable without increasing too much the size of the Jacobian. Such "comfort" variables are different than the _dep variables like rho, which do not exist at all in the final Jacobian. These latter variables are replaced symbolically by their dependance before the construction of the Jacobian.

The comfort equations do not need a RHS. Let us see an example to explain this point. If we consider the Emden-like equation

$$y'' + \Lambda y^n = 0$$

with some boundary conditions, Λ being an unkown quantity part of the solution, the associated variational form is

$$\delta y'' + y^n \delta \Lambda + n\Lambda y^{n-1} \delta y = -(y'' + \Lambda y^n)$$

Now, let us introduce a "comfort" equation namely

$$Y = \Lambda y^n$$

so that the former equation reads

$$\delta y'' + \delta Y = -(y'' + \Lambda y^n)$$

much simpler indeed, but we need to add the new variational equation

$$\delta Y - u^n \delta \Lambda - n \Lambda u^{n-1} \delta u = 0$$

that has no RHS.

3.3.4. Setting interface and boundary conditions

At the interface of two domains we need to impose the continuity of the functions and of their normal derivative (when the operators are of second order). The conditions are imposed on the last grid point of the lower domain and on the first grid point of the upper domain. To be clear let us take an example of a 1D problem, where at some radius $\zeta = \eta$ we impose

$$f(\zeta_{\text{interf}}^n) = f(\zeta_{\text{interf}}^{n+1})$$
 and $f'(\zeta_{\text{interf}}^n) = f'(\zeta_{\text{interf}}^{n+1})$

where the prime is for the derivative. n designates the lower domain and n+1 the upper domain. Of course $\zeta_{\text{interf}}^n = \zeta_{\text{interf}}^{n+1} = \eta$. The equation to be solved may be Laplace equation for instance. In figure 3.1 we display the structure of the matrix using 20 grid points, with 10 points in each domain. We use the strong formulation of the differential problem (Rieutord et al., 2016). On the first line of the matrix we have, for instance, the boundary condition at the bottom of the first domain. On row 10, the last grid point of the first domain, we impose the continuity of the solution, so the first dot is just "1" and the second dot is "-1". In the ESTER-nomenclature, the "1" represents the first part of the interface condition while the "-1" is the second part. Hence

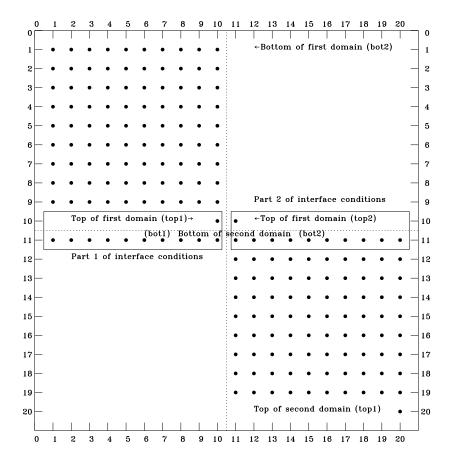


Figure 3.1: Example of discretized-operator matrix, in 1D, with two domains and interface conditions.

the "1" is inserted in the matrix via a bc_top1_add operation, while the "-1" is inserted in the matrix via a bc_top2_add operation.

The second domain, in our example, starts at row 11 by the other interface condition imposing the continuity of the derivative of the solution. In the ESTER-nomenclature, the lower domain (n-1) part is inserted by a bc_bot1_add operation, while the actual domain n condition need a bc_bot2_add operation (see figure 3.1). In row 20, we impose the "surface" boundary condition saying $f(\zeta_{\text{surf}}) = 0$, so the dot is just a "1".

Meaning of the symbols (but see also Fig. 3.1):

- $d \equiv \text{diagonal} \equiv \text{essentially multiplication by a function of } (\zeta, \theta).$
- $\ell \equiv \text{left operator} \equiv \text{Operator}$ on the radial grid
- $r \equiv \text{right operator} \equiv \text{Operator on the } \theta\text{-grid}$
- $i \equiv \text{inside} \equiv \text{multiplication}$ inside the operator, namely before applying the operator (see (3.33))
- $ex \equiv$ concerning the exterior (vacuum) domain.

		Optio	ns		Example	Argument
	top	1		$egin{pmatrix} \ell \ r \end{bmatrix}$		
bc			add	lri	$bc_top1_add_1$	(idom,"Eq","var",coef,mat)
	bot	2		d		
		_		ex		

Table 3.5: Reading guide for the implementation of boundary conditions.

• 1 and 2: If we consider the nth domain, interface conditions with the n-1-th domain are set with bot1 and bot2 (bot1 for the n-1th domain), and interface conditions with the n+1-th domain are set with top1 and top2 (top2 for the n+1th domain).

3.3.5. Implementation of Newton's algorithm

The construction of ESTER's models follows the following steps:

- 1. Read the initial model or build it (only in 1D)
- 2. Build the Jacobian matrix
- 3. LU factorization of the Jacobian matrix
- 4. CGS solve (or refine) for the correction δx
- 5. Update \boldsymbol{x}
- 6. recompute the RHS, $F(x_k)$ and the Jacobian matrix
- 7. Attempt a Conjugate Gradient Squared (CGS) solution to derive the new δx using the former LU matrices as a preconditionner. Namely, if L_k and U_k are the factors of \mathcal{J}_k , the k+1 equation is solved with the (split-preconditioned) CGS solver as

$$J_k^{k+1} U_k \delta x_{k+1} = -L_k^{-1} F_{k+1}$$

where the $J_k^{k+1} = L_k^{-1} \mathcal{J}_{k+1} U_k^{-1}$ -matrix is almost diagonal. If the convergence of this iterative method is less than say N iterates, then the algorithm continues at step 5. N is chosen such that the N iterations are faster than a LU factorization. If convergence is not reached, then the algorithm continues on step 3

8. if $\|\delta x\|/\|x\| \le \text{tolerance}$, then stop.

3.4. Modifying the equations

Two kinds of modifications are conceivable: either to modify existing equations or to add new equations coupled to the existing ones.

3.4.1. Adding a new force or a heat source

 $Under\ construction$

3.4.2. Adding a new equation and a new variable

Under construction

The code is divided in several libraries. Each library implements one or more classes designed to handle one particular aspect of the calculation.

- matrix. Matrix algebra.
- numdiff. Implements Gauss-Legendre and multi-domain Gauss-Lobatto numerical differentiation.
- mapping. Defines the mapping in spheroidal coordinates $r(\zeta, \theta)$.
- solver. Resolution of systems of linear differential equations in 2D.
- **physics**. Calculation of physical quantities (opacity, equation of state, nuclear reaction rates).
- star. Provides objects and functions to calculate the structure of a star in 1D and 2D.
- global. Definition of global variables, e.g. physical and mathematical constants.
- graphics. Provides graphical output through pgplot.
- parser. Parsing of configuration files and command-line arguments and file input/output.

3.5. Heading towards time evolution

3.5.1. Simple scheme for hydrogen burning

Step 1

During hydrogen burning, heat is basically produced by the transformation of protons into alpha particles (⁴He nuclei) and thus, not to bother with the nuclear reaction network, we can write

$$\frac{dX}{dt} = -\frac{4m_p\varepsilon}{Q}$$

where m_p is the proton mass, and

$$Q = (4m_p - m_{\rm He})c^2 = 4.3 \times 10^{-12} \text{ J}$$

is the energy released by the transformation of the 4 protons¹. We thus deduced the discretized time evolution for $\ln X$ during Δt , namely

$$\ln X + \frac{4m_p \varepsilon \Delta t}{QX} = \ln X_{n-1} \tag{3.40}$$

 $^{^{1}}arepsilon/Q$ is the rate of He generation per unit mass and 4arepsilon/Q the number of protons that disappear per unit mass and time. Thus $4m_{H}arepsilon/Q$ is the number of kg of H that disappear per kg of matter and per second; this is precisely \dot{X} .

where the LHS variables refer to the time-step n. The variational form for solving (3.40) is

$$\delta \ln X + \frac{4m_p \varepsilon \Delta t}{QX} \left[\left(\frac{\partial \ln \varepsilon}{\partial \ln T} \right) \delta \ln T + \left(\frac{\partial \ln \varepsilon}{\partial \ln \rho} \right) \delta \ln \rho + \left(\frac{\partial \ln \varepsilon}{\partial \ln X} - 1 \right) \delta \ln X \right] = RHS \quad (3.41)$$

where

$$RHS = \ln X_{n-1} - \left(\ln X + \frac{4m_p \varepsilon \Delta t}{QX}\right)$$

we note that when the CNO-cycle dominates the third term in [...] can be neglected. Indeed, $\varepsilon = \varepsilon_{\rm pp} + \varepsilon_{\rm cno}$ and $\varepsilon_{\rm pp} \propto X^2$ while $\varepsilon_{\rm cno} \propto X$, it follows that

$$\frac{\partial \ln \varepsilon}{\partial \ln X} - 1 = \frac{\varepsilon_{\mathrm{pp}}}{\varepsilon}$$

which can be neglected at high temperature.

So as a first step and for early-type stars we adopt

$$\delta \ln X + \frac{4m_p \varepsilon \Delta t}{QX} \left[\left(\frac{\partial \ln \varepsilon}{\partial \ln T} \right) \delta \ln T + \left(\frac{\partial \ln \varepsilon}{\partial \ln \rho} \right) \delta \ln \rho \right] = RHS \tag{3.42}$$

In the convective core, we assume a uniform distribution of X, namely

$$\overline{X} = \frac{1}{M_{\text{core}}} \int_{(V)} X \rho dV$$

Its evolution is described (basically) by

$$\frac{d\overline{X}}{dt} = -\frac{4m_p}{Q} \frac{L_{\text{core}}}{M_{\text{core}}} \tag{3.43}$$

so that we need to solve

$$\overline{X} + \frac{4m_p}{Q} \frac{L_{\text{core}}}{M_{\text{core}}} = \overline{X}_{n-1}$$

As a first step, we solve for a variational formulation neglecting the core mass variation, namely

$$\delta \overline{X} + \frac{4m_p}{QM_{\text{core}}} \int_0^{\zeta_c} \delta(\rho \varepsilon r^2 r_{\zeta}) d\zeta = \overline{X}_{n-1} - \left(\overline{X} + \frac{4m_p}{Q} \frac{L_{\text{core}}}{M_{\text{core}}} \right)$$

in 1D.

Some numerical values for orders of magnitude

Simplifying the expression given by Kippenhahn et al. (2012), we take

$$\varepsilon_{\text{CNO}} = 8.24 \, 10^{25} \rho X X_{\text{CNO}} T_9^{-2/3} e^{-15.231/T_9^{1/3}} \text{ cgs}$$

Let's take $X_{\rm CNO}\simeq Z/2\simeq 0.01,~X=0.7,~T_c=2.76\times 10^7{\rm K},~\rho_c=21.3~{\rm g/cm^3}$ (from a 1D ESTER model of a 5M $_\odot$ ZAMS star). Then,

$$\varepsilon_c = 1.74 \times 10^4 \text{ erg/s/g}$$

Hence, if $\Delta t = 1$ Myr, then

$$\frac{4m_p\varepsilon_c\Delta t}{Q} = 0.085$$

So the decrease is a little more than 10% (of 0.7) every million years for this star.

Step 2

While the star consumes its hydrogen it shrinks thus generating a quasi-steady flow towards the center (if no rotation). Let's consider this case first.

In the Euler formulation (unlike lagrangian) we have to include the velocity field

$$\frac{\partial X}{\partial t} + \boldsymbol{v} \cdot \nabla X = -f\varepsilon \tag{3.44}$$

1D-case:

$$\frac{\partial X}{\partial t} + V_r \frac{\partial X}{\partial r} = -f\varepsilon \tag{3.45}$$

If we discretize in time

$$X_{n+1} - X_n + \Delta t V_r \frac{\partial X_{n+1/2}}{\partial r} + f \varepsilon = 0$$
 (3.46)

In 1D, V_r is just the motion of the mass-coordinate. In ESTER, mass-coordinate does not exist, domains are attached either to pressure or temperature, not to mass. We need the radial velocity that insures mass-conservation:

$$\frac{\partial \rho}{\partial t} + \nabla \cdot \rho \boldsymbol{v} = 0 \quad \implies \quad v_r = -\frac{1}{r^2 \rho} \frac{\partial}{\partial t} \int_0^r r^2 \rho dr$$

or

$$v_r(r) = -\frac{1}{4\pi r^2 \rho} \left(\frac{\partial M}{\partial t}\right)_r$$

So to estimate v_r we need to evaluate the mass function M(r,t) at successive times. So if $\Delta M = M'(r_i) - M(r_i)$ then

$$v_r(r) = -\frac{1}{4\pi r^2 \rho} \frac{\Delta M}{\Delta t}$$

Matrix Algebra. The matrix library.

To facilitate the work with matrices in C++, the matrix library provides two classes:

- matrix for regular matrices
- matrix_block_diag for block diagonal matrices

The function prototypes are defined in the header file matrix.h.

4.1. MATRIX CREATION AND MANIPULATION

Regular matrices are defined as objects of the matrix class. For example, the sentence:

```
matrix a(3,4);
creates a matrix a with 3 rows and 4 columns. If the size is not specified,
matrix a;
a 1x1 matrix is created. The size of the matrix can be modified using the method dim
a.dim(3,4);
```

or, if the total number of elements does not change, using ${\tt redim}$

```
a.redim(1,12);
```

With redim the element values are also preserved. The number of rows and columns of a matrix object can be retrieved using the methods nrows() and ncols(). For example

```
int n,m;
matrix a(3,4);
n=a.nrows();
m=a.ncols();
```

in this example n=3 and m=4.

The elements of the matrix can be indexed using parenthesis. Note that, as regular C arrays, the index of the first element is 0. There are also methods for extracting parts of the matrix. Let's see an example

```
matrix a(3,3),row,col,block;
double elem;

a(0,0)=1;a(0,1)=2;a(0,2)=3;
a(1,0)=4;a(1,1)=5;a(1,2)=6;
a(2,0)=7;a(2,1)=8;a(2,2)=9;

elem=a(1,2); // elem=6
row=a.row(1); // Extracts the second row
col=a.col(0); // Extracts the first column
block=a.block(0,1,1,2); // Extracts the block (0-1,1-2)
```

After running the example, the contents of the different matrices will be

$$\mathbf{a} = \left(\begin{array}{ccc} 1 & 2 & 3\\ 4 & 5 & 6\\ 7 & 8 & 9 \end{array}\right)$$

$$\mathsf{row} = \left(\begin{array}{ccc} 4 & 5 & 6 \end{array}\right) \qquad \mathsf{col} = \left(\begin{array}{ccc} 1\\ 4\\ 7 \end{array}\right) \qquad \mathsf{block} = \left(\begin{array}{ccc} 2 & 3\\ 5 & 6 \end{array}\right)$$

We can also insert parts of the matrix using the methods setrow, setcol and setblock.

```
matrix a(3,3),b;
b=ones(1,3); // Creates a 1x3 array of all ones
a.setrow(0,b);
b=ones(3,1);
a.setcol(2,b);
b=ones(2,3);
a.setblock(1,2,0,2,b);
```

Negative indices are interpreted starting from the end of the matrix. For example a.row(-1) returns the last row of the matrix a.

Indexing with only one parameter is also possible, being a(i,j) equivalent to a(j*a.nrows()+i). This makes sense when working with row or column vectors, if we define

```
matrix row(1,5),col(5,1);
```

then row(i) is equivalent to row(1,i) and col(i) is equivalent to col(i,1).

4.2. FILE INPUT/OUTPUT

The method write writes a matrix in a file, the syntax is

```
write(FILE *fp, char mode)
```

Here, mode can be 't' for text output or 'b' for binary output. Default is 't'. The matrix is written in column-wise order, i.e. each line represents a column of the matrix. When called without arguments write(), it writes the matrix in the standard output.

To read a matrix from a file we use the method read.

```
read(int nrow, int ncol, FILE *fp, char mode)
```

Where we must specify the size of the matrix.

In the following example, we will write a matrix to a file and read it again.

```
#include<stdio.h>
#include"matrix.h"
int main() {
    FILE *fp;
    matrix a(2,3);
    a(0,0)=1;a(0,1)=2;a(0,2)=3;
    a(1,0)=4; a(1,1)=5; a(1,2)=6;
    // Write the matrix to a file in binary mode
    fp=fopen("matrix.dat", "wb");
    a.write(fp, 'b');
    fclose(fp);
    // Read the matrix from file
    fp=fopen("matrix.dat", "rb");
    a.read(2,3,fp,'b');
    fclose(fp);
    return 0;
}
```

We can write a matrix on the screen in a more convenient format using write_fmt. For the previous example the sentence

```
a.write_fmt("%.2f");
will produce the following output
1.00 2.00 3.00
4.00 5.00 6.00
```

4.3. Operators

Element-wise operators for the matrix class:

a=b	Assignment
a+b	Addition
a-b	Subtraction
a*b	Element-wise multiplication
a/b	Element-wise division
a+=b	Equivalent to a=a+b
a-=b	Equivalent to a=a-b
a*=b	Equivalent to a=a*b
a/=b	Equivalent to a=a/b
+a	Unary plus
-a	Unary minus
a==b	Comparison: Equal to
a!=b	Comparison: Not equal to
a>b	Comparison: Greater than
a <b< td=""><td>Comparison: Less than</td></b<>	Comparison: Less than
a>=b	Comparison: Greater than or equal to
a<=b	Comparison: Less than or equal to
a&&b	Logical AND
allb	Logical OR

The operands ${\tt a}$ and ${\tt b}$ can be either matrices or scalars. Element-wise operators are performed element by element. For example if we define

c=a*b

the elements of the new matrix c will be

$$c(i,j)=a(i,j)*b(i,j)$$

obviously, the two matrices must have the same size. There is one exception, when one or both of the dimensions are one, for example if ${\tt a}$ is $({\tt n,m})$ and ${\tt b}$ is $({\tt 1,m})$ the matrix ${\tt c}$ will be $({\tt n,m})$ with elements

$$c(i,j)=a(i,j)*b(j)$$

also if a is $(n,1)$ and b is $(1,m)$, c will be (n,m) with $c(i,j)=a(i)*b(j)$

The comparison operator == compares two matrices element by element, so the result is a new matrix whose elements are 1 if the corresponding elements of a and b are equal or 0 if they are different. If we want to know if two matrices are completely equal, we can use the function isequal(a,b) that returns 1 if a and b are the same and 0 otherwise.

Matrix product are indicated with a comma ",". The product of matrices a and b are expressed as:

```
c=(a,b);
```

The operation should be put in parentheses when necessary to avoid ambiguity. Note that the operator "," in C has the lowest precedence, for example

```
(2*a,b+c,d)
is equivalent to
( (2*a) , ( (b+c) ,d ) )
```

4.4. Block diagonal matrices

Another type of object included in the library are the block diagonal matrices. An object of this class has the following structure

$$M = \begin{pmatrix} M_0 & 0 & \cdots & 0 \\ 0 & M_1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & M_{n-1} \end{pmatrix}$$

where the M_i are also matrices. Although the definition of a block diagonal matrix requires the blocks M_i to be square, in the current implementation they are allowed to have any size.

A block diagonal matrix is defined using the sentence

```
matrix_block_diag D;
```

An optional argument can be included to specify the number of blocks in the matrix (default is 1)

```
matrix_block_diag D(4);
```

Alternatively, the number of blocks can be changed using the sentence

```
D.set_nblocks(4);
```

In order to access the different blocks, we use the method block(int i), for exmaple

```
matrix_block_diag D(3);
matrix a,b;

a=ones(2,2);
D.block(0)=a;
b=D.block(0);
```

Individual elements can also be indexed using parentheses D(i,j), as with regular matrices. A number of operators are defined in the matrix_block_diag class:

Operator	Operands type	Return type	Description		
a=b	matrix_block_diag	matrix_block_diag	Assignment		
a+b	matrix_block_diag	matrix_block_diag	Addition		
a-b	matrix_block_diag	matrix_block_diag	Subtraction		
+a	matrix_block_diag	matrix_block_diag	Unary plus		
-a	matrix_block_diag	matrix_block_diag	Unary minus		
	matrix_block_diag	matrix_block_diag			
	matrix_block_diag & matrix	matrix_block_diag			
a*b	matrix $\&$ matrix_block_diag	matrix_block_diag	Element-wise multiplication		
	matrix_block_diag & double	matrix_block_diag			
	double $\&$ matrix_block_diag	matrix_block_diag			
a/b	matrix_block_diag & matrix	matrix_block_diag	Element-wise division		
a/D	matrix_block_diag $\&$ double	matrix_block_diag	Element-wise division		
	matrix_block_diag	matrix_block_diag			
(a,b)	matrix_block_diag & matrix	matrix_block_diag	Matrix product		
	matrix $\&$ matrix_block_diag	matrix_block_diag			

For element-wise operators between matrix_block_diag objects, both objects must have exactly the same structure. Matrix product is also performed block by block, so the structure of the operands must be compatible.

A matrix_block_diag object can be converted in a matrix object using type casting.

```
matrix a;
matrix_block_diag D;
a=(matrix) D;
```

4.5. Reference

Matrix manipulation

```
\dim(n,m)
                                                setblock(n1,n2,m1,m2,A)
redim(n,m)
                                                setblock\_step(n1,n2,nstep,m1,m2,mstep,A)
nrows()
                                                transpose()
                                                fliplr()
ncols()
row(n)
                                                flipud()
col(n)
                                                data()
block(n1,n2,m1,m2)
                                                swap()
block\_step(n1,n2,nstep,m1,m2,mstep)
                                                zero(n,m)
setrow(n,A)
setcol(n,A)
```

File input/output

```
write(fp,mode)
read(n,m,fp,mode)
write_fmt(format,fp)
```

Special matrices

```
\begin{array}{ll} eye(n) & vector(x0,x1,n) \\ zeros(n,m) & vector\_t(x0,x1,n) \\ ones(n,m) & \\ random\_matrix(n,m) & \end{array}
```

Matrix functions

$\max(A)$	exist(A)
$\min(A)$	isequal(A,B)
sum(A)	solve(b)
$\operatorname{mean}(A)$	inv()
$\max(A,B)$	
$\min(A,B)$	

Mathematical functions

$\cos(x)$	$\exp(x)$
$\sin(x)$	$\log(x)$
$\tan(x)$	$\log 10(x)$
acos(x)	$\operatorname{sqrt}(x)$
asin(x)	abs(x)
atan(x)	pow(x,y)
atan2(y,x)	round(x)
$\cosh(x)$	floor(x)
$\sinh(x)$	ceil(x)
tanh(x)	

Block diagonal matrices

```
\begin{array}{lll} set\_nblocks(n) & row(n) \\ block(n) & transpose() \\ nblocks() & eye(D) \\ nrows() & & \\ ncols() & & \end{array}
```

4.5.1. A NOTE ABOUT METHODS AND FUNCTIONS

The subroutines are divided in two types: functions and methods. Contrary to functions, methods belong to the object and they are called using a different syntax. For example if met is a method of the object a that takes one argument b and returns a value c, we use the sentence

```
c=a.met(b)
```

The same subroutine implemented as a function will be

```
c=met(a,b)
```

When using pointers, the dot is replaced by \rightarrow , then if p=&a the sentence above is equivalent to

c=p->met(b)

The parenthesis are needed even if the method takes no arguments, i.e. a.method_without_args().

4.5.2. Matrix manipulation

dim(n,m)

Type: Method

Inputs: n (int): Number of rows

m (int): Number of columns

Output: Reference to current object

Description: Changes the dimensions of the matrix object.

redim(n,m)

Type: Method

Inputs: n (int): Number of rows

m (int): Number of columns

Output: Reference to current object

Description: Changes the dimensions of the matrix object. The total number elements

must not change. Element values are preserved.

nrows()

Type: Method Inputs: None Output: int

Description: Returns the number of rows of the matrix.

ncols()

Type: Method Inputs: None Output: int

Description: Returns the number of columns of the matrix.

row(n)

Type: Method

Inputs: n (int): Row index

Output: matrix

Description: Extracts row n from matrix.

col(n)

Type: Method

Inputs: n (int): Column index

Output: matrix

Description: Extracts column n from matrix.

block(n1,n2,m1,m2)

Type: Method

Inputs: n1 (int): First row index

n2 (int): Last row indexm1 (int): First column indexm2 (int): Last column index

Output: matrix

Description: Extracts the block contained between the rows n1 and n2 and the columns

m1 and m2.

$block_step(n1,n2,nstep,m1,m2,mstep)$

Type: Method

Inputs: n1 (int): First row index

n2 (int): Last row index nstep (int): Row increment m1 (int): First column index m2 (int): Last column index mstep (int): Column increment

Output: matrix

Description: Extracts the block contained between the rows n1 and n2 and the columns

m1 and m2 using increments nstep and mstep.

setrow(n,A)

Type: Method

Inputs: n (int): Row index

A (matrix)

Output: Reference to current object Description: Inserts matrix A at row n.

setcol(n,A)

Type: Method

Inputs: n (int): Column index

A (matrix)

Output: Reference to current object Description: Inserts matrix A at column n.

setblock(n1,n2,m1,m2,A)

Type: Method

Inputs: n1 (int): First row index

n2 (int): Last row indexm1 (int): First column indexm2 (int): Last column index

A (matrix)

Output: Reference to current object

Description: Inserts matrix A between the rows n1 and n2 and the columns m1 and m2.

$setblock_step(n1,n2,nstep,m1,m2,mstep,A)$

Type: Method

Inputs: n1 (int): First row index

n2 (int): Last row index nstep (int): Row increment m1 (int): First column index m2 (int): Last column index mstep (int): Column increment

A (matrix)

Output: Reference to current object

Description: Inserts matrix A between between the rows n1 and n2 and the columns m1

and m2 using increments nstep and mstep.

transpose()

Type: Method Inputs: None Output: matrix

Description: Returns the transpose of the object. Does not modify the original matrix.

fliplr()

Type: Method Inputs: None Output: matrix

Description: Flip columns in the left-right direction. Does not modify the original matrix.

flipud()

Type: Method Inputs: None Output: matrix

Description: Flip rows in the up-down direction. Does not modify the original matrix.

data()

 $Type: \qquad \qquad \text{Method} \\ Inputs: \qquad \qquad \text{None} \\$

Output: Pointer to double

Description: Returns a pointer to the first element in the matrix. The elements are stored

consecutively in column order.

swap()

Type: Method Inputs: matrix Output: None

Description: Swaps the contents of the current matrix object and the one used as argument.

zero(n,m)

Type: Method

Inputs: n (int): Number of rows

m (int): Number of columns

Output: None

Description: Creates a nxm matrix of all zeros. Note that a.zero(n,m) is equivalent to

a=zeros(n,m) but avoids the creation of an intermediate object, saving mem-

ory for large matrices.

4.5.3. FILE INPUT/OUTPUT

write(fp,mode)

Type: Method

 $Inputs: \hspace{1cm} \text{fp (FILE *): File pointer (optional, default=stdout)} \\$

mode (char): Write mode (optional, default='t')

Output: int

Description: Writes a matrix in the file pointed by fp in text mode (mode='t') or binary

mode (mode='b'). The matrix is written in column order. Returns 1 on

success, 0 otherwise.

read(n,m,fp,mode)

Type: Method

Inputs: n (int): Number of rows

m (int): Number of columns fp (FILE *): File pointer

mode (char): Write mode (optional, default='t')

Output: int

Description: Reads a nxm matrix from the file pointed by fp in text mode (mode='t') or

binary mode (mode='b'). The matrix is read in column order. Returns 1 on

success, 0 otherwise.

write_fmt(format,fp)

Type: Method

Inputs: format (char *): Format string

fp (FILE *): File pointer (optional, default=stdout)

Output: None

Description: Writes a matrix in the file pointed by fp using given format. The matrix is

ordered such that each line represents a row.

4.5.4. Special matrices

eye(n)

Type: Function

Inputs: n (int): Number of rows

Output: matrix

Description: Returns the nxn identity matrix.

zeros(n,m)

Type: Function

Inputs: n (int): Number of rows

m (int): Number of columns

Output: matrix

Description: Returns a nxm matrix of all zeros.

ones(n,m)

Type: Function

Inputs: n (int): Number of rows

m (int): Number of columns

Output: matrix

Description: Returns a nxm matrix of all ones.

random_matrix(n,m)

Type: Function

Inputs: n (int): Number of rows

m (int): Number of columns

Output: matrix

Description: Returns a nxm matrix with random values between 0 and 1.

vector(x0,x1,n)

Type: Function

Inputs: x0 (double): Minimum value

x1 (double): Maximum value n (int): Number of elements

Output: matrix

Description: Returns a row vector with n equally spaced elements between x0 and x1.

$vector_t(x0,x1,n)$

Type: Function

Inputs: x0 (double): Minimum value

x1 (double): Maximum value n (int): Number of elements

Output: matrix

Description: Returns a column vector with n equally spaced elements between x0 and x1.

4.5.5. Matrix functions

max(A)

Type: Function
Inputs: A (matrix)
Output: double

Description: Returns the maximum value.

$\min(\mathbf{A})$

Type: Function
Inputs: A (matrix)
Output: double

Description: Returns the minimum value.

sum(A)

Type: Function Inputs: A (matrix) Output: double

Description: Returns the sum of the matrix elements.

mean(A)

Type: Function
Inputs: A (matrix)
Output: double

Description: Returns the mean value of the matrix elements.

max(A,B)

Type: Function
Inputs: B (matrix)

B (matrix)

Output: matrix

Description: Compares the matrices a and b and returns a new matrix C containing the

larger values of each pair of elements $C(i,j)=\max(A(i,j),B(i,j))$.

min(A,B)

Type: Function
Inputs: A (matrix)

B (matrix)

Output: matrix

Description: Compares the matrices A and B and returns a new matrix C containing the

smaller values of each pair of elements $C(i,j)=\min(A(i,j),B(i,j))$.

exist(A)

Type: Function Inputs: A (matrix)

Output: int

Description: Returns 1 if any of the elements of A is not zero, 0 otherwise. It is often used

in constructions of type if (exist(condition))..., where condition is a

valid comparison. For example if (exist(A<0))....

isequal(A,B)

Type: Function
Inputs: A (matrix)

B(matrix)

Output: int

Description: Returns 1 if matrices A and B contain exactly the same values, 0 otherwise.

solve(b)

Description: x=A.solve(b) solves the linear system Ax=b and returns matrix x.

inv()

Type: Method Inputs: None Output: matrix

Description: Returns the inverse of the current matrix object. The original matrix is not

modified.

4.5.6. Mathematical functions

$\cos(x)$

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the cosine of x. x must be expressed in radians.

$\sin(x)$

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the sine of x. x must be expressed in radians.

tan(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the tangent of x. x must be expressed in radians.

acos(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the arc cosine of x in radians.

asin(x)

Type: Function Inputs: $x ext{ (matrix)}$ Output: matrix

Description: Returns the arc sine of x in radians.

atan(x)

Type: Function
Inputs: x (matrix)
Output: matrix

Description: Returns the arc tangent of x in radians.

atan2(y,x)

Type: Function

Inputs: y (matrix or double)

x (matrix or double)

Output: matrix

Description: Returns the arc tangent of y/x in radians. Uses the sign of y and x to

determine the quadrant.

cosh(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the hyperbolic cosine of x. x must be expressed in radians.

sinh(x)

Type: Function
Inputs: x (matrix)
Output: matrix

Description: Returns the hyperbolic sine of x. x must be expressed in radians.

tanh(x)

Description: Returns the hyperbolic tangent of x. x must be expressed in radians.

$\exp(x)$

Type:FunctionInputs:x (matrix)Output:matrixDescription:Returns e^x .

log(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns $\log(x)$.

log10(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns $\log_{10}(x)$.

sqrt(x)

Type: Function
Inputs: x (matrix)Output: matrixDescription: Returns \sqrt{x} .

abs(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Returns the absolute value of x.

pow(x,y)

Type: Function

Inputs: x (matrix or double)

y (matrix or double)

Output: matrix Description: Returns x^y .

round(x)

Type: Function Inputs: x (matrix) Output: matrix

Description: Rounds the elements of x to the nearest integer.

floor(x)

Type: Function
Inputs: x (matrix)
Output: matrix

Description: Rounds the elements of x to the nearest integer below the current value.

ceil(x)

Type: Function
Inputs: x (matrix)
Output: matrix

Description: Rounds the elements of x to the nearest integer above the current value.

4.5.7. Block diagonal matrices

set_nblocks(n)

Type: Method

Inputs: n (int): Number of blocks
Output: Reference to current object

 $Description: \quad \hbox{Changes the number of blocks of the matrix_block_diag object.}$

block(n)

Type: Method

Inputs: n (int): Block number
Output: Reference to matrix

Description: Returns a reference to the matrix in the block number n.

nblocks()

Description: Returns the number of blocks.

nrows()

Description: Returns the total number of rows.

ncols()

Type: Method Inputs: None Output: int

Description: Returns the total number of columns.

row(n)

Type: Method

Inputs: n (int): Row number

Output: int

Description: Extracts the row n.

transpose()

Type: Method Inputs: None

Output: matrix_block_diag

 $Description: \quad \hbox{Calculates the transpose}.$

eye(D)

Type: Function

Inputs: D (matrix_block_diag)
Output: matrix_block_diag

Description: Returns the identity block matrix with the same structure as D.

5

Numerical differentiation

5.1. Introduction

Numerical differentiation refers to the algorithms for estimating the derivative of a function using only its values at certain evaluation points.

The simplest method for evaluating the derivative is the finite difference method. Given a function f sampled at points x_i (i = 0, ..., n - 1), its derivative is estimated using the formula

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i}$$

For equally spaced points $x_{i+1} - x_i = h$ and the formula becomes

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h}$$

There are some variations of the finite difference formula, as for example the central difference

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h}$$

When solving differential equations, it is a good idea to write this expression in matrix form:

$$\begin{pmatrix} f'(x_0) \\ f'(x_1) \\ f'(x_2) \\ f'(x_3) \\ \vdots \\ f'(x_{n-3}) \\ f'(x_{n-2}) \\ f'(x_{n-1}) \end{pmatrix} = \frac{1}{2h} \begin{pmatrix} -2 & 2 & 0 & 0 & \cdots & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & \cdots & 0 & 0 & 0 \\ 0 & -1 & 0 & 1 & \cdots & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & \cdots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & \cdots & 0 & -1 & 0 & 1 \\ 0 & 0 & 0 & 0 & \cdots & 0 & -2 & 2 \end{pmatrix} \begin{pmatrix} f(x_0) \\ f(x_1) \\ f(x_2) \\ f(x_3) \\ \vdots \\ f(x_{n-3}) \\ f(x_{n-2}) \\ f(x_{n-1}) \end{pmatrix}$$

or, in more compact form

$$f'(x_i) = \sum_{j=0}^{n-1} D_{ij} f(x_j)$$

where D_{ij} is the differentiation matrix.

The finite difference method has order 2, which means that the error in the estimation of the derivative is proportional to h^2 . In fact, it is possible to construct higher order methods using more points to estimate the derivative. As a rule of thumb, the order of the method is at least equal to the number of points used in the estimation of the derivative.

Unfortunately, high order methods using equally spaced points are affected by the Runge's phenomenon. Indeed, finite difference formulas of order n are obtained by interpolating the function between the points of interest using a polynomial of degree n-1. One of the main problems of polynomial interpolation using polynomials of high degree is the apparition of oscillations near the edges of the interval between the interpolation points. The amplitude of the oscillations increase with the degree of the polynomial and quickly degrades the derivative estimation for high order methods. This effect is known as the Runge's phenomenon.

Collocation or pseudospectral methods attempt to suppress the Runge's phenomenon by choosing a set of non-equally spaced points called collocation points.

5.1.1. Collocation/Pseudospectral methods

Collocation methods are high order methods for estimating the derivative of a function knowing its values at certain points called collocation points. The position of this points is different for each collocation method and is designed to suppress the Runge's phenomenon. A pseudospectral method with n points has order 2n.

Each particular collocation method is associated with a family of orthogonal functions $P_l(x)$. This functions form a basis so, any arbitrary function $\phi(x)$ can be expressed as a linear combination of the basis functions

$$\phi(x) = \sum_{l=0}^{\infty} a_l P_l(x)$$

In practice, we will work with a finite discretization using n points, so the expansion is truncated to use only n basis functions, the fuction $\phi(x)$ is then approximated by

$$\phi^{(n)}(x) = \sum_{l=0}^{n-1} \phi_l P_l(x)$$

For regular functions and a well-adapted choice of the basis functions, collocation methods have exponential convergence, which means that the error in the approximation decreases exponentially with the number of basis functions n.

The functions $P_l(x)$ are orthogonal against some scalar product $\langle P_l, P_m \rangle = \delta_{lm}$, then the coefficients on the expansion of $\phi(x)$ can be calculated as

$$\phi_l = \langle \phi(x), P_l(x) \rangle$$

For each family of basis functions it exists a formula of gaussian quadrature for calculating this scalar product, then

$$\phi_l = \sum_{j=0}^{n-1} w_j P_l(x_j) \phi(x_j)$$

where x_j and w_j are the nodes and weights of the corresponding gaussian quadrature. Note that x_j are the collocation points. The estimation of the first derivative at the collocation points can

be obtained as

$$\phi'(x_i) = \sum_{l=0}^{n-1} \phi_l P_l'(x_i)$$

$$= \sum_{l=0}^{n-1} \left(\sum_{j=0}^{n-1} w_j P_l(x_j) \phi(x_j) \right) P_l'(x_i)$$

$$= \sum_{j=0}^{n-1} \left(\sum_{l=0}^{n-1} w_j P_l(x_j) P_l'(x_i) \right) \phi(x_j)$$

$$= \sum_{j=0}^{n-1} D_{ij} \phi(x_j)$$

where $D_{ij} = \sum_{l=0}^{n-1} w_j P_l(x_j) P'_l(x_i)$ is the differentiation matrix. We see that the derivative of a discretized function can be calculated by doing a matrix product.

$$\Phi' = D\Phi$$

Similarly, the second derivative will be

$$\Phi'' = DD\Phi$$

5.1.2. Relation with spectral methods

Collocation methods are intimately related with spectral methods and share most of their properties. The main difference is that in spectral methods we work with the coefficients ϕ_l in the expansion of the functions contrarily to collocation methods that deal directly with the values of the function at the collocation points. This has a clear advantage when solving differential equations with variable coefficients. For example, the equation

$$\phi'(x) + a(x)\phi(x) = b(x)$$

will be discretized using spectral methods as

$$\sum_{m} \langle P_{l}, P'_{m} \rangle \phi_{m} + \sum_{m,k} \langle P_{l}, P_{m} P_{k} \rangle a_{m} \phi_{k} = b_{l}$$

While the first product $\langle P_l, P'_m \rangle$ use to be easy to calculate, this is not the case for the second one $\langle P_l, P_m P_k \rangle$. By contrast, for collocation methods the discretization is just

$$\sum_{j} D_{ij}\phi(x_j) + a(x_i)\phi(x_i) = b(x_i)$$

It is possible to pass from one representation to the other using projection matrices. To get the spectral coefficients of a function ϕ , we multiply by the projection matrix \mathcal{P}_{ij} .

$$\phi_l = \sum_{j=0}^{n-1} \mathcal{P}_{li} \phi(x_i)$$

where $\mathcal{P}_{li} = w_i P_l(x_i)$. To recover the values at the collocation points we do

$$\phi(x_i) = \sum_{l=0}^{n-1} \mathcal{P}_{il}^{-1} \phi_l$$

where $\mathcal{P}_{il}^{-1} = P_l(x_i)$ is the matrix inverse of \mathcal{P}_{li} .

5.1.3. Multi-domain

One of the main drawbacks of pseudospectral collocation methods is that they do not deal correctly with non-regular functions. If the function that we want approximate has discontinuities, even in its first derivatives, the exponential convergence is lost and the approximated function can show oscillations around the discontinuity. This is known as the Gibbs phenomenon.

There are multiple ways to deal with this problem, one of them is to use a multi-domain approach. It consists in dividing the integration domain in multiple subintervals. A division is placed at the points where there is a discontinuity in the function. So now the function is continuous in each subinterval and the pseudospectral approximation works properly.

5.1.4. Numerical differentiation in ESTER

At the moment, ESTER provides two classes for numerical differentiation:

- diff_gl: Multi-domain Gauss-Lobatto numerical differentiation.
- diff_leg: Gauss-Legendre numerical differentiation for axisymmetric functions on the sphere with a defined type of symmetry (pole,equator).

The function prototypes are defined in numdiff.h.

5.2. Multi-domain Gauss-Lobatto numerical differentiation

In the Gauss-Lobatto (or more properly Gauss-Lobatto-Chebyshev) collocation method, the basis functions are Chebyshev polynomials of the first kind

$$T_l(x) = \cos(l\arccos(x))$$

defined in the interval (-1,1). The collocation points are

$$x_i = -\cos(\frac{i\pi}{n})$$

The end points of the interval are also collocation points, which make this method well-suited for boundary value problems.

In the ESTER library, multi-domain Gauss-Lobatto numerical differentiation is implemented in the diff_gl class. To work with this class we should first create an object using

The argument n is optional (default 1) and indicates the number of domains. To change the number of domains we can do

```
gl.set_ndomains(n);
```

After setting the number of domains we must indicate the number of points per domain and the position of the domains. Let's see an example using three domains and the following set-up.

- First domain in the interval (0,5) with 30 points.
- \bullet Second domain in the interval (5,7.5) with 20 points.
- Third domain in the interval (7.5,10) with 20 points.

The diff_gl object can be initialized using the code

```
diff_gl gl;
gl.set_ndomains(3); // Use 3 domains
gl.set_xif(0.,5.,7.5,10.); // Set the limits between domains
gl.set_npts(30,20,20); // Set the number of points in each domain
gl.init(); // Initialize the object
```

The limits between the subdomains and the number of points are stored in C arrays that are accessible from outside the class, so the code above is equivalent to

```
diff_gl gl;
gl.set_ndomains(3);
gl.xif[0]=0;gl.xif[1]=5;gl.xif[2]=7.5;gl.xif[3]=10;
gl.npts[0]=30;gl.npts[1]=20;gl.npts[2]=20;
gl.init();
```

During the initialization, the following objects are created:

Name	Type	Size	Description
ndomains	int		Number of domains
N	int		Total number of points
x	matrix	(N,1)	Collocation points x_i
D	$matrix_block_diag$	(N,N) in blocks of (npts[i],npts[i])	Differentiation matrix
I	matrix	(1,N)	Integration matrix
P	matrix_block_diag	(N,N) in blocks of (npts[i],npts[i])	Projection matrix \mathcal{P}_{ij}
P1	matrix_block_diag	(N,N) in blocks of (npts[i],npts[i])	Inverse projection matrix \mathcal{P}_{ij}^{-1}

The following example illustrates the use of these objects:

```
[...] // Initialization (previous example)
matrix y,dy,ddy;
y=cos(gl.x); // Definition of function y
dy=(gl.D,y); // First derivative
```

A function can also be interpolated at any point within the whole domain using the method

```
eval(y,x)
```

where x can be of type double or matrix and the return value will be of always of type matrix. It returns the value of y at the point(s) x. We can use also a third argument of type matrix

```
eval(y,x,T)
```

where T is modified during the call and can be used to interpolate other functions. For the previous example

A diff_gl object can be copied using the assignment operator. We can do

```
diff_gl gl1,gl2;
gl1.set_ndomains(2);
gl1.set_npts(10,10);
gl1.set_xif(0.,0.5,1.);
gl2=gl1;
```

Now gl2 is an independent copy of gl1 and, as gl1 has already been initialized, this is also the case of gl2.

5.2.1. Example

Let's see a full featured example that uses the class diff_gl. We are going to solve the ordinary differential equation

$$x^2 \frac{\mathrm{d}^2 y}{\mathrm{d}x^2} + x \frac{\mathrm{d}y}{\mathrm{d}x} - y = x^2$$

within the interval [0,1] with boundary conditions

$$y(0) = 0$$
 $y(1) = 0$

whose exact solution is

$$y = \frac{1}{3}x(x-1)$$

To simplify the task, we will use only 1 domain. Later, we will see the class solver that allows to solve more complicated problems using several domains and several variables.

The code is the following

```
// The following example solves the differential equation
         x^2*y'' + x*y' - y = x^2
// with boundary conditions y(0)=0 and y(1)=0
// whose exact solution is
       y = x*(x-1)/3
#include<stdio.h>
#include"numdiff.h"
int main() {
    //Initialize a diff_gl object with 1 domain
    int n=20; // Number of points.
        //In fact, this example can be solved using only 3 points.
    diff_gl gl(1);
    gl.set_npts(n);
    gl.set_xif(0.,1.);
    gl.init();
    // We will work with only 1 domain, so we create a reference to the
    // first (and only) block of gl.D
   matrix &D=gl.D.block(0);
   matrix &x=gl.x;
    // Set up the operator matrix and the right hand side
   matrix op, rhs;
    op=x*x*(D,D)+x*D-eye(n);
    rhs=x*x;
```

```
// Introduce boundary conditions
    op.setrow(0,zeros(1,n));op(0,0)=1;
    rhs(0)=0;
    op.setrow(-1,zeros(1,n));op(-1,-1)=1;
    rhs(-1)=0;
    // Solve the system
    matrix y;
    y=op.solve(rhs);
    // Interpolate the solution into a finer grid
   matrix x_fine,y_fine;
    x_fine=vector_t(0,1,100);
    y_fine=gl.eval(y,x_fine);
    // Compare with the exact solution
   matrix y_exact;
    y_exact=x_fine*(x_fine-1)/3;
    printf("Solved using %d points\n",gl.N);
   printf("Max. error=%e\n", max(abs(y_fine-y_exact)));
    return 0;
}
```

To run the example, just copy the code above to a file, and compile with <code>ester_build</code>. If the file is called <code>example.cpp</code> we will do

```
$ ester_build example.cpp -o example
and then execute using
$ ./example
The output will be something like
Solved using 20 points
Max. error=5.329938e-16
```

5.3. Gauss-Legendre numerical differentiation

The Gauss-Legendre collocation method uses Legendre polynomials $P_l(x)$ as basis functions. For n points, the collocation points are defined as the roots of $P_n(x)$.

Legendre collocation is particularly adapted to deal with axisymmetric functions on the surface of a sphere, that depend only on the colatitude θ , just by doing $x = \cos \theta$.

The current implementation in ESTER considers only one domain, limited to one hemisphere $\theta \in [0, \pi/2]$. This is more efficient when dealing with functions that have a defined type of symmetry with respect to the equator $(\theta = \pi/2)$, which is the case for all of the variables used in the ESTER code. Legendre polynomials $P_l(\cos \theta)$ are symmetric with respect to $\theta = 0$ (the pole). When dealing with antisymmetric functions with respect to the pole, the derivatives $\frac{\mathrm{d}P_l}{\mathrm{d}\theta}$ are used as basis functions instead.

The implementation considers four types of symmetry. Each type is indicated by its own suffix.

Suffix	Pole	Equator	Basis functions
00	Symmetric	Symmetric	$P_l(\cos\theta)$ with l even
01	Symmetric	Antisymmetric	$P_l(\cos\theta)$ with l odd
10	Antisymmetric	Symmetric	$\frac{\mathrm{d}P_l}{\mathrm{d}\theta}$ with l odd
11	Antisymmetric	Antisymmetric	$\frac{\mathrm{d}\theta}{\mathrm{d}\theta}$ with l even

Legendre numerical differentiation is implemented in the class diff_leg. In order to use it, we must start by creating an object

diff_leg leg;

Then we set the number of points by setting the variable npts, for example

leg.npts=20;

and initialize the object

leg.init();

The following objects are created:

Name	Type	Size	Description
th	matrix	(1, npts)	Collocation points x_i
D_00, D_01, D_10, D_11	matrix	$(\mathtt{npts},\mathtt{npts})$	Differentiation matrices
D2_00, D2_01, D2_10, D2_11	matrix	(npts,npts)	Second differentiation matrices
lap_00, lap_01, lap_10, lap_11	matrix	$(\mathtt{npts},\mathtt{npts})$	Laplacian operator $\frac{1}{\sin \theta} \frac{d}{d\theta} \left(\sin \theta \frac{d}{d\theta} \right)$
I_00	matrix	$(\mathtt{npts},\!1)$	Integration matrix
P_00, P_01, P_10, P_11	matrix	(npts,npts)	Projection matrices \mathcal{P}_{ij}
P1_00, P1_01, P1_10, P1_11	matrix	(npts,npts)	Inverse projection matrices \mathcal{P}_{ij}^{-1}

Contrary to the diff_gl class, in which the functions was supposed to be column vectors, the diff_leg class expects functions to be defined as row vectors. This means that all the operators are applied using right multiplication. For example, the derivative will be

 $dy=(y,D_0);$

The reason for that is more clear when working with 2D functions. Consider the code

```
int nr=30;
int nth=20;
// Inititalize a diff_gl object with nr points
diff_gl gl(1);
gl.set_xif(0.,1.);gl.set_npts(nr);
gl.init();
// Initialize a diff_leg object with nth points
diff_leg leg;
leg.npts=nth;
leg.init();
// Define a 2D function
matrix y;
y=gl.x*sin(leg.th)*sin(leg.th);
    // gl.x is (nr,1) and leg.th is (1,nth), then y is (nr,nth)
//Compute derivatives
matrix dy_x,dy_th,dy_x_th;
dy_x=(gl.D,y); // Derivative with respect to x
dy_th=(y,leg.D_00); // Derivative with respect to th
dy_x_th=(gl.D,y,leg.D_00); // Second derivative with respect to x and th
```

Note that the derivative of a function will have a different type of symmetry. For example, for a symmetric-symmetric 00 function, the first derivative is

```
dy=(y,leg.D_00)
```

which is of type 11. Then to calculate the second derivative we should do

```
ddy=(y,leg.D_00,leg.D_11)
```

or, using the second derivative matrix

where ddy has type 00.

The integration matrix is defined only for type 00 functions and computes the integral between 0 and π with weight function $\sin \theta$

(y,leg.I_00) =
$$\int_0^{\pi} y \sin \theta d\theta$$

We can interpolate functions at any point using eval_xx, where xx is the type of symmetry, for example

```
eval_00(y,th)
```

gives the value of y at the point(s) th. Here, th can be either double or int and the returned value is always of type matrix. We may also use a third argument

```
eval_00(y,th,T)
```

where T can be used to interpolate additional functions at the same point(s) by doing (y2,T)

A diff_leg object can be copied using the assignment operator. We can do diff_leg leg1,leg2;
leg1.npts=20;leg1.init();
leg2=leg1;
Now leg2 is an independent copy of leg1.

5.3.1. Example

Let's see a more complete example. We will consider axisymmetric functions in spherical coordinates, that is functions that depend only on r and θ . For the radial direction we use Gauss-Lobatto differentiation in the interval (0,1), and Legendre differentiation for the latitudinal direction. We will write two functions, one for calculating the value of the laplacian at a certain point and another one to evaluate the volume integral within the whole sphere. For simplicity, we will consider only type 00 functions.

The code is as follows

```
/* The following example illustrates the use of the numerical differentiation
library in 2D in spherical coordinates */
#include<stdio.h>
#include"numdiff.h"
#include"constants.h" //For the defintion of PI
//Function prototypes
double laplacian(matrix y,double r0,double th0);
double integral(matrix y);
// Define diff_gl and diff_leg objects as global variables
diff_gl gl;
diff_leg leg;
// Create references for spherical coordinates
matrix &r=gl.x,&th=leg.th;
int main() {
    //Initialize gl. In the example we will use 2 domains
    gl.set_ndomains(2);
    gl.set\_xif(1e-3,0.2,1.); // Use 1e-3 as the interior limit (instead of 0)
                            // to avoid a division by zero in the
                            // calculation of the laplacian
    gl.set_npts(100,100);
    gl.init();
    //Initialize leg
    leg.npts=50;
```

```
leg.init();
    matrix y;
    //Define the function y
    y=r*r*r*(1+sin(th)*sin(th));
    double lap_y,int_y;
    double r0=0.3, th0=PI/3;
    lap_y=laplacian(y,r0,th0);
    int_y=integral(y);
    printf("The value of the laplacian at (%f, %f) is %e\n", r0, th0, lap_y);
    printf("The volume integral is %e\n",int_y);
    return 0;
}
// Function for calculating the laplacian of y at (r0,th0)
double laplacian(matrix y,double r0,double th0) {
   matrix lap_y;
   lap_y=(gl.D,r*r*gl.D,y)/r/r+(y,leg.lap_00)/r/r;
    //Interpolate in the direction of r
   lap_y=gl.eval(lap_y,r0); // Now lap_y is (1,nth)
    //Interpolate in the direction of theta
    lap_y=leg.eval_00(lap_y,th0); // Now lap_y is (1,1)
    return lap_y(0);
    // lap_y has only 1 element, but we must include (0)
    // at the end to return a double
// Function for calculating the volume integral of y
double integral(matrix y) {
    return 2*PI*(gl.I,r*r*y,leg.I_00)(0);
}
```

After running the code, the output should be

```
The value of the laplacian at (0.300000, 1.047198) is 6.150000e+00 The volume integral is 3.490659e+00
```

5.4. Reference

Gauss-Lobatto differentiation

Data members

ndomains P xif

 $\begin{array}{ccc} N & & P1 \\ x & & I \\ D & & npts \end{array}$

FUNCTIONS

 $\begin{array}{lll} set_ndomains(n) & set_xif(x0,x1,\dots) & eval(y,x,T) \\ set_npts(n0,n1,\dots) & init() & \end{array}$

Legendre differentiation

Data members

 $p1_00, P1_01, P1_10, P1_11 \\ lap_00, lap_01, lap_10, lap_11 \\ lap_00, lap_01, lap_11 \\ lap_00, lap_01, lap_11 \\ lap_00, lap_01, lap_01, lap_11 \\ lap_00, lap_01, lap_01,$

th D_00,D_01,D_10,D_11 I_00

P_00,P_01,P_10,P_11 D2_00,D2_01,D2_10,D2_11

FUNCTIONS

 $\begin{array}{ll} \operatorname{init}() & \operatorname{eval_10}(y, \operatorname{th}, T) \\ \operatorname{eval_00}(y, \operatorname{th}, T) & \operatorname{eval_11}(y, \operatorname{th}, T) \end{array}$

 $eval_01(y,th,T)$ $eval(y,th,T,par_pol,par_eq)$

5.4.1. Gauss-Lobatto differentiation

Data members

ndomains

Type: int

Description: Number of domains (read-only).

 $\overline{\mathbf{N}}$

Type: int

Description: Total number of points, including all the domains (read-only).

 $\overline{\mathbf{x}}$

Type: matrix

Description: Collocation points (N x 1).

 $\overline{\mathbf{D}}$

Type: matrix_block_diag

Description: Differentiation matrix.

 $\overline{\mathbf{P}}$

Type: matrix_block_diagDescription: Projection matrix.

 $\overline{\mathbf{P1}}$

Type: matrix_block_diag

Description: Inverse projection matrix.

Ī

Type: matrix

Description: Integration matrix.

npts

Type: int *

Description: Array of size ndomains() with the number of points in each domain.

xif

Type: double *

Description: Array of size ndomains()+1 with the limits between domains.

FUNCTIONS

set_ndomains(n)

Type: Method

Inputs: n (int): Number of domains

Output: None

Description: Change the number of domains.

$set_npts(n0,n1,...)$

Type: Method

Inputs: n0,n1,...(int): Number of points

Output: None

Description: Change the number of points in each domain.

$set_xif(x0,x1,...)$

Type: Method

Inputs: n0,n1,...(double): Number of points

Output: None

Description: Change the position of the limits between domains.

init()

Type: Method Inputs: None Output: None

Description: Initializes the object.

eval(y,x,T)

Type: Method

Inputs: y (matrix): Function to evaluate

x (matrix or double): Evaluation point(s)

T (matrix): Interpolating matrix (optional, output)

Output: matrix

Description: Evaluate function y at point(s) x. If y is NxM and x is Kx1, the returned

matrix is KxM. The optional matrix T can be used to interpolate additional

functions by multiplying (T,y2).

5.4.2. Legendre differentiation

Data members

npts

Type: int

Description: Number of points.

\mathbf{th}

Type: matrix

Description: θ values of the collocation points (1 x npts).

P_00,P_01,P_10,P_11

Type: matrix

Description: Projection matrices for each type of symmetry.

P1_00,P1_01,P1_10,P1_11

Type: matrix

Description: Inverse projection matrices for each type of symmetry.

D_-00,D_-01,D_-10,D_-11

Type: matrix

Description: Differentiation matrices for each type of symmetry.

$D2_{-}00,D2_{-}01,D2_{-}10,D2_{-}11$

Type: matrix

Description: Second differentiation matrices for each type of symmetry.

$lap_00, lap_01, lap_10, lap_11$

Type: matrix

Description: Laplacian operator matrices for each type of symmetry.

$$(\mathtt{lap_xx},\mathtt{f}) \equiv \frac{1}{\sin\theta} \frac{\mathrm{d}}{\mathrm{d}\theta} \left(\sin\theta \frac{\mathrm{d}\mathtt{f}}{\mathrm{d}\theta} \right)$$

$\overline{I_{-}00}$

Type: matrix

Description: Integration matrix (npts x 1) for symmetric functions.

$$(I_00, f) \equiv \int_0^{\pi} f \sin(\theta) d\theta$$

FUNCTIONS

init()

Type: Method Inputs: None Output: None

 $Description : \quad \text{Initializes the object.}$

$eval_00(y,th,T)$

Type: Method

Inputs: y (matrix): Function to evaluate

th (matrix or double): Evaluation point(s)

T (matrix): Interpolating matrix (optional, output)

Output: matrix

Description: Evaluate function $y(\theta)$ at point(s) th. The function y should be symmetric

at the pole and the equator. If y is MxN and x is 1xK, the returned matrix is MxK. The optional matrix T can be used to interpolate additional functions

by multiplying (T,y2).

$eval_01(y,th,T)$

Type: Method

Inputs: y (matrix): Function to evaluate

th (matrix or double): Evaluation point(s)

T (matrix): Interpolating matrix (optional, output)

Output: matrix

Description: Evaluate function $y(\theta)$ at point(s) th. The function y should be symmetric

at the pole and antisymmetric the equator. If y is MxN and x is 1xK, the returned matrix is MxK. The optional matrix T can be used to interpolate

additional functions by multiplying (T,y2).

$eval_10(y,th,T)$

Type: Method

Inputs: y (matrix): Function to evaluate

th (matrix or double): Evaluation point(s)

T (matrix): Interpolating matrix (optional, output)

Output: matrix

Description: Evaluate function $y(\theta)$ at point(s) th. The function y should be antisymmetric

at the pole and symmetric at the equator. If y is MxN and x is 1xK, the returned matrix is MxK. The optional matrix T can be used to interpolate

additional functions by multiplying (T,y2).

$eval_11(y,th,T)$

Type: Method

Inputs: y (matrix): Function to evaluate

th (matrix or double): Evaluation point(s)

T (matrix): Interpolating matrix (optional, output)

Output: matrix

Description: Evaluate function $y(\theta)$ at point(s) th. The function y should be antisymmetric

at the pole and the equator. If y is MxN and x is 1xK, the returned matrix is MxK. The optional matrix T can be used to interpolate additional functions

by multiplying (T,y2).

eval(y,th,T,par_pol,par_eq)

Type: Method

Inputs: y (matrix): Function to evaluate

th (matrix or double): Evaluation point(s) T (matrix): Interpolating matrix (output) par_pol (int): Type of symmetry at the pole par_eq (int): Type of symmetry at the equator

Output: matrix

Description: Evaluate function $y(\theta)$ at point(s) th. par_pol and par_eq can be 0 (symmet-

ric) or 1 (antisymmetric). If y is MxN and x is 1xK, the returned matrix is MxK. The matrix T can be used to interpolate additional functions by

multiplying (T,y2).

Mapping. Spheroidal coordinates

We have seen at the end of the previous chapter that we can work with 2D problems combining diff_gl and diff_leg objects. This is exactly the purpose of the mapping class, that contains all the elements to work with spherical, and more important, with deformed spheroidal domains.

6.1. Introduction

Rotating stars are not spherical. The centrifugal force flattens the star, and this flattening increases with the angular velocity. For this reason, we use a discretization of the stellar variables in a deformed spheroidal domain.

Note that the problems in which we are interested involves only axisymmetric quantities, thus essentially 2D. For that reason, we restrict the discussion to axisymmetric 2D problems in a spherical-like domain, but it can be generalized to a non-axisymmetric spheroidal geometry as shown in Bonazzola *et al.* (1998).

6.1.1. Coordinate mapping

Let (r, θ, φ) be the spherical coordinates and \mathcal{D} be an axisymmetric domain centered at the origin of coordinates (r = 0) and whose outer boundary $\partial \mathcal{D}$ can be represented by a function $R(\theta)$ that depends only on colatitude. We define a new set of coordinates $(\zeta, \theta', \varphi')$ such that the new radial-like coordinate ζ is constant over $\partial \mathcal{D}$. These new spheroidal coordinates are defined by the transformation:

$$\begin{cases}
 r = r(\zeta, \theta') \\
 \theta = \theta' \\
 \varphi = \varphi'
\end{cases}$$
(6.1)

The problem reduces to find a suitable form for the function $r(\zeta, \theta)$.

In our case we are going a little bit further. We split the domain \mathcal{D} in n subdomains \mathcal{D}_i with $i = 0, \ldots, n-1$. The frontiers between this subdomains are represented by a series of functions $R_i(\theta)$, $i = 0, \ldots, n$, such that the $\mathcal{D}_i \in [R_i(\theta), R_{i+1}(\theta)]$. Note that $R_n(\theta) = R(\theta)$ is the outer boundary of the whole domain and, if the domain contains the origin of coordinates $R_0(\theta) = 0$.

We also use an external domain \mathcal{D}_{ex} that extends from the outer boundary $R_n(\theta)$ to infinity. It will be useful for writing boundary conditions for the gravitational potential.

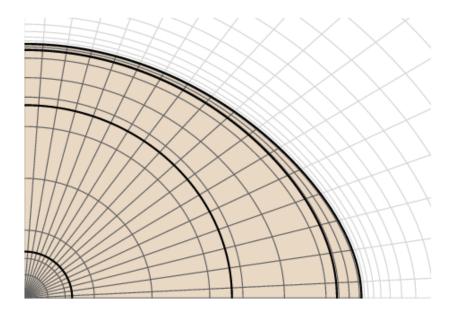


Figure 6.1: Coordinate mapping.

We use a technique adopted from Bonazzola et al. (1998), in each subdomain \mathcal{D}_i we use a mapping in the form

$$r(\zeta, \theta) = a_i \xi \Delta \eta_i + R_i(\theta) + A_i(\xi) (\Delta R_i(\theta) - a_i \Delta \eta_i) \quad \text{for } \zeta \in [\eta_i, \eta_{i+1}]$$
 (6.2)

where we have defined:

$$\begin{array}{rcl} \eta_i & = & R_i(\theta=0) \\ \Delta \eta_i & = & \eta_{i+1} - \eta_i \\ \Delta R_i(\theta) & = & R_{i+1}(\theta) - R_i(\theta) \\ \xi & = & \frac{\zeta - \eta_i}{\Delta \eta_i} \quad \text{so that} \quad 0 \le \xi \le 1 \end{array}$$

so the η_i are the polar radii of the interfaces between the domains. Since we choose the polar radius as the unit length $R_n(\theta = 0) = 1$. Hence, inside the star

$$0 \le \zeta \le 1$$

From the definition we also have $\eta_0 = 0$ and $\eta_n = 1$.

The function(s) $A_i(\xi)$ and the constant(s) a_i determine the final form of the mapping. In particular, the function $A_i(\xi)$ should verify the following conditions:

$$r(\eta_i, \theta) = R(\theta) \longrightarrow A_i(\xi = 0) = 0$$

 $r(\eta_{i+1}, \theta) = R_{i+1}(\theta) \longrightarrow A_i(\xi = 1) = 1$

The simplest possibility is a linear mapping where $A_i(\xi) = \xi$ and $a_i = 0$ giving $r(\zeta, \theta) = R_i(\theta) + \xi \Delta R_i(\theta)$. This mapping however is not equivalent to the spherical geometry near the center

and therefore not adapted to the use of spherical harmonics basis for the angular dependence of the solution. This is why Bonazzola et al. (1998) proposed a mapping that satisfies some extra conditions to make it suitable for spectral methods. For that reason they set $A_i'(\xi=0)=0$ and $A_i'(\xi=1)=0$. Doing this, the first derivative of the mapping

$$r_{\zeta} = \frac{\partial r}{\partial \zeta} = a_i + A_i'(\xi) \left(\frac{\Delta R_i(\theta)}{\Delta \eta_i} - a_i \right)$$
 (6.3)

is constant over the boundaries of the domains. This facilitates the writing of interface conditions for the derivatives of the variables in the problem when they are expressed in terms of their spectral coefficients. In this case, we set

$$A_i(\xi) = -2\xi^3 + 3\xi^2$$
 for $i = 1, ..., n-1$ (6.4)

$$A_i(\xi) = -2\xi^3 + 3\xi^2$$
 for $i = 1, ..., n - 1$ (6.4)
 $A_0(\xi) = -1.5\xi^5 + 2.5\xi^3$ (6.5)

thus

$$A'_i(\xi) = 6\xi(1-\xi)$$
 and $A'_0 = 7.5\xi^2(1-\xi^2)$ (6.6)

The constants a_i deserve some special attention. We first note that the Jacobian of the transformation (6.1) is non-singular, namely

$$J = \begin{vmatrix} \frac{\partial r}{\partial \zeta} & \frac{\partial r}{\partial \theta'} \\ \frac{\partial \theta}{\partial \zeta} & \frac{\partial \theta}{\partial \theta'} \end{vmatrix} = r_{\zeta} \neq 0$$

Actually, we want $r_{\zeta} > 0$, so that r is monotonically increasing with ζ . Then a_i should satisfy the condition

$$a_i(A_i'(\xi) - 1) < A_i'(\xi) \frac{\Delta R_i(\theta)}{\Delta \eta_i}$$
(6.7)

For $A'_i(\xi) = 0$ (the boundaries), the condition states $a_i > 0$. When $A'(\xi) < 1$, the condition is automatically satisfied (note that $A'_i(\xi)$ is always positive). But, since $A(\xi)$ should go from 0 at $\xi = 0$ to 1 at $\xi = 1$, we know that $\max(A_i(\xi)) \ge 1$, where the equality corresponds to the linear mapping that we have seen before. So, in the worst case, the condition becomes

$$a_i < \frac{1}{1 - 1/\max(A_i'(\xi))} \frac{\min(\Delta R_i(\theta))}{\Delta \eta_i}$$
(6.8)

or, using (6.4)

$$a_{i} < 3 \frac{\min(\Delta R_{i}(\theta))}{\Delta \eta_{i}} \quad \text{for } i = 1, \dots, n - 1$$

$$a_{0} < \frac{15}{7} \frac{\min(\Delta R_{0}(\theta))}{\Delta \eta_{0}}$$

$$(6.9)$$

In practice, we take

$$a_i = 1. (6.10)$$

This is motivated by the fact that we work with oblate domains. Indeed, the flattening increases for the successive subdomains so that $\min(\Delta R_i(\theta)) = \Delta R_i(0) = \Delta \eta_i$. Hence, conditions (6.9) are fully satisfied.

In addition, working with a fixed value of a_i makes easier to work with problems where the frontiers between the subdomains are not known a priori, and the Jacobian of the mapping becomes a smooth function suitable for iterative methods.

The general form of the Jacobian of the mapping is defined by the expression

$$\delta r^{i} = J_0^{(i)}(\zeta, \theta)\delta \eta_i + J_1^{(i)}(\zeta, \theta)\delta \Delta \eta_i + J_2^{(i)}(\zeta, \theta)\delta R_i(\theta) + J_3^{(i)}(\zeta, \theta)\delta \Delta R_i(\theta)$$

$$(6.11)$$

and using (6.2), for fixed a_i

$$J_0^{(i)} = 0$$

$$J_1^{(i)} = a_i(\xi - A_i(\xi))$$

$$J_2^{(i)} = 1$$

$$J_3^{(i)} = A_i(\xi)$$
(6.12)

For the external domain, we take

$$\xi_{ex} = \frac{\xi}{1 - \xi}$$
 with $\xi \in [0, 1[$ and $\xi_{ex} \in [0, \infty[$ (6.13)

so that

$$r_{ex}(\zeta,\theta) = \xi_{ex} + R(\theta) \tag{6.14}$$

with Jacobian

$$J_0^{(ex)} = 0$$

$$J_1^{(ex)} = 0$$

$$J_2^{(ex)} = 1$$

$$J_3^{(ex)} = 0$$
(6.15)

6.1.2. Spheroidal coordinates

In the previous section, we have defined a system of spheroidal coordinates $(\zeta, \theta', \varphi')$, where $\theta' = \theta$ and $\varphi' = \varphi$ correspond to the usual spherical coordinates and ζ is defined by a relation $r = r(\zeta, \theta)$. These spheroidal coordinates are non-orthogonal, which means that the surfaces of constant ζ are not perpendicular to those of constant θ .

Before we continue, we should clarify a point. We have set $\theta' = \theta$, so hereafter we will remove the tilde (') to simplify the notation. But when working with spheroidal coordinates, the partial derivative $\frac{\partial}{\partial \theta}$ refers to the derivative with respect to θ with ζ constant, that is not the habitual derivative in spherical coordinates that is done holding r constant.

$$\left. \frac{\partial}{\partial \theta} \right|_{\zeta,\varphi \, \text{const.}} \neq \left. \frac{\partial}{\partial \theta} \right|_{r,\varphi \, \text{const.}}$$

The same can be said for the azimuthal coordinate φ but, in this case $\frac{\partial}{\partial \varphi}\Big|_{\zeta,\theta \text{ const.}} = \frac{\partial}{\partial \varphi}\Big|_{r,\theta \text{ const.}}$.

NATURAL BASIS

We will start by defining the natural basis for the spheroidal coordinates, we have two sets of basis vectors:

• Covariant basis vectors: $\mathbf{E}_i = \frac{\partial \mathbf{r}}{\partial x^i}$

$$\mathbf{E}_{\zeta} = r_{\zeta}\hat{\mathbf{r}}, \quad \mathbf{E}_{\theta} = r_{\theta}\hat{\mathbf{r}} + r\hat{\boldsymbol{\theta}}, \quad \mathbf{E}_{\varphi} = r\sin\theta\hat{\boldsymbol{\varphi}},$$
 (6.16)

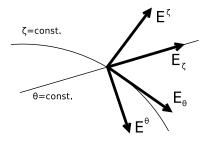
• Contravariant basis vectors: $\mathbf{E}^i = \nabla x^i$

$$\mathbf{E}^{\zeta} = \frac{\hat{r}}{r_{\zeta}} - \frac{r_{\theta}}{rr_{\zeta}}\hat{\boldsymbol{\theta}}, \quad \mathbf{E}^{\theta} = \frac{\hat{\boldsymbol{\theta}}}{r}, \quad \mathbf{E}^{\varphi} = \frac{\hat{\boldsymbol{\varphi}}}{r\sin\theta}, \tag{6.17}$$

where $\hat{r}, \hat{\theta}, \hat{\varphi}$ are the usual unit vectors in spherical coordinates, and

$$r_{\zeta} = \frac{\partial r}{\partial \zeta} \quad r_{\theta} = \frac{\partial r}{\partial \theta}$$

The vectors of the natural basis are not unit vectors. The covariant vector \mathbf{E}_i is parallel to the line $x^j = \text{const.}$ with $j \neq i$, while the contravariant vector \mathbf{E}^i is perpendicular to the surface $x^i = \text{const.}$ For orthogonal coordinates $\mathbf{E}_i \parallel \mathbf{E}^i$, but this is not the case for non-orthogonal coordinates.



The basis vectors satisfy

$$\mathbf{E}_i \cdot \mathbf{E}^j = \mathbf{E}^i \cdot \mathbf{E}_j = \delta_{ij} \tag{6.18}$$

where δ_{ij} is the Kronecker's delta.

Using the basis vectors, we can calculate the metric tensor

$$g_{ij} = \mathbf{E}_i \cdot \mathbf{E}_j = \begin{pmatrix} r_{\zeta}^2 & r_{\zeta} r_{\theta} & 0\\ r_{\zeta} r_{\theta} & r^2 + r_{\theta}^2 & 0\\ 0 & 0 & r^2 \sin^2 \theta \end{pmatrix}$$
(6.19)

or, in contravariant form

$$g^{ij} = \mathbf{E}^{i} \cdot \mathbf{E}^{j} = \begin{pmatrix} \frac{r^{2} + r_{\theta}^{2}}{r^{2}r_{\zeta}^{2}} & \frac{-r_{\theta}}{r^{2}r_{\zeta}} & 0\\ \frac{-r_{\theta}}{r^{2}r_{\zeta}} & \frac{1}{r^{2}} & 0\\ 0 & 0 & \frac{1}{r^{2}\sin^{2}\theta} \end{pmatrix}$$
(6.20)

Note that g^{ij} is the matrix inverse of g_{ij}

$$g_{ij}g^{jk} = \delta_{ij}$$

where we have used the Einstein's summation convention, that implies summation over repeated indices.

Given two points x^i and $x^i + dx^i$, the distance (ds) between them is given by the metric tensor:

$$ds^{2} = g_{ij}dx^{i}dx^{j} = r_{\zeta}^{2}d\zeta^{2} + 2r_{\zeta}r_{\theta}d\zeta d\theta + (r^{2} + r_{\theta}^{2})d\theta^{2} + r^{2}\sin^{2}\theta d\varphi^{2}$$

$$(6.21)$$

The basis vectors verify

$$\mathbf{E}_i \cdot (\mathbf{E}_j \times \mathbf{E}_k) = \epsilon_{ijk} \tag{6.22}$$

and

$$\mathbf{E}^i \cdot (\mathbf{E}^j \times \mathbf{E}^k) = \epsilon^{ijk} \tag{6.23}$$

where ϵ^{ijk} is the Levi-Civita tensor

$$\epsilon_{ijk} = \sqrt{|g|}[i, j, k] \tag{6.24}$$

$$\epsilon^{ijk} = \frac{1}{\sqrt{|g|}} [i, j, k] \tag{6.25}$$

where $|g| = \det(g_{ij}) = r^4 r_{\zeta}^2 \sin^2 \theta$ and

$$[i, j, k] = \begin{cases} 1 & \text{the arguments are an even permutation of } 1,2,3 \\ -1 & \text{the arguments are an odd permutation of } 1,2,3 \\ 0 & \text{two or more arguments are equal} \end{cases}$$
 (6.26)

Representation of vectors

A vector \mathbf{v} can be represented either in covariant or contravariant form:

- Covariant form: $\mathbf{v} = V_{\zeta} \mathbf{E}^{\zeta} + V_{\theta} \mathbf{E}^{\theta} + V_{\varphi} \mathbf{E}^{\varphi}$
- Contravariant form: $\mathbf{v} = V^{\zeta} \mathbf{E}_{\zeta} + V^{\theta} \mathbf{E}_{\theta} + V^{\varphi} \mathbf{E}_{\varphi}$

Here, V_i are the covariant components of the vector v and V^i the contravariant components. Note that

$$\mathbf{E}_i \cdot \mathbf{v} = V_i$$
 and $\mathbf{E}^i \cdot \mathbf{v} = V^i$

We can use the metric tensor to pass from one representation to the other, indeed

$$V_i = \mathbf{E}_i \cdot \mathbf{v} = \mathbf{E}_i \cdot (\mathbf{E}_i V^j) = g_{ij} V^j \tag{6.27}$$

and similarly

$$V^i = g^{ij}V_j (6.28)$$

Let $(v_r, v_\theta, v_\varphi)$ be the spherical components of a vector \mathbf{v} such that $\mathbf{v} = v_r \hat{\mathbf{r}} + v_\theta \hat{\boldsymbol{\theta}} + v_\varphi \hat{\boldsymbol{\varphi}}$. Its spheroidal components will be

$$V_{\zeta} = r_{\zeta} v_r, \quad V_{\theta} = r_{\theta} v_r + r v_{\theta}, \quad V_{\varphi} = r \sin \theta v_{\varphi}$$
 (6.29)

and

$$V^{\zeta} = \frac{v_r}{r_{\zeta}} - \frac{r_{\theta}}{rr_{\zeta}} v_{\theta}, \quad V^{\theta} = \frac{v_{\theta}}{r}, \quad V^{\varphi} = \frac{v_{\varphi}}{r \sin \theta}$$
 (6.30)

We can see from this expressions that V^{θ} and V^{φ} are in fact angular velocities.

Using the properties of the basis vectors it can be shown that the scalar product of two vectors is given by

$$\mathbf{a} \cdot \mathbf{b} = A_i B^i = A^i B_i \tag{6.31}$$

and the cross product is

$$(\mathbf{a} \times \mathbf{b})^i = \epsilon^{ijk} A_j B_k (\mathbf{a} \times \mathbf{b})_i = \epsilon_{ijk} A^j B^k$$
 (6.32)

We have presented the basics of the representation of vectors in spheroidal coordinates, let'see now a little example. Consider a surface \mathcal{S} defined by $\zeta = \mathrm{const.}$ as for example the surface of a star or the frontier between two subdomains. We want to calculate the normal and tangential projections of a vector \mathbf{v} with respect to \mathcal{S} . First, we define a unit vector $\hat{\mathbf{n}}$, perpendicular to \mathcal{S} . For that, we just recall that \mathbf{E}^{ζ} is perpendicular to the surfaces $\zeta = \mathrm{const.}$, but it is not a unit vector, so

$$\hat{\mathbf{n}} = \frac{\mathbf{E}^{\zeta}}{|\mathbf{E}^{\zeta}|} = \frac{\mathbf{E}^{\zeta}}{\sqrt{\mathbf{E}^{\zeta} \cdot \mathbf{E}^{\zeta}}} = \frac{\mathbf{E}^{\zeta}}{\sqrt{q^{\zeta\zeta}}}$$
(6.33)

then, the normal projection is

$$\hat{\mathbf{n}} \cdot \mathbf{v} = \frac{V^{\zeta}}{\sqrt{g^{\zeta\zeta}}} = \frac{r_{\zeta}V^{\zeta}}{\sqrt{1 + \frac{r_{\theta}^2}{r^2}}} \tag{6.34}$$

For the parallel projection we have two vectors, the first one, in the direction of φ is just the spherical unit vector $\hat{\varphi}$, in the latitudinal direction, however, it will be

$$\hat{\mathbf{t}} = \frac{\mathbf{E}_{\theta}}{|\mathbf{E}_{\theta}|} = \frac{\mathbf{E}_{\theta}}{\sqrt{\mathbf{E}_{\theta} \cdot \mathbf{E}_{\theta}}} = \frac{\mathbf{E}_{\theta}}{\sqrt{g_{\theta\theta}}}$$
(6.35)

so, the parallel projections over $\mathcal S$ are

$$\hat{\mathbf{t}} \cdot \mathbf{v} = \frac{V_{\theta}}{\sqrt{g_{\theta\theta}}} = \frac{1}{\sqrt{1 + \frac{r_{\theta}^2}{r^2}}} \frac{V_{\theta}}{r}$$
(6.36)

and

$$\hat{\boldsymbol{\varphi}} \cdot \mathbf{v} = \frac{V_{\varphi}}{r \sin \theta} \tag{6.37}$$

Tensors

A second order tensor \mathcal{T} is represented using 2 indices

$$\mathcal{T} = T^{ij} \mathbf{E}_i \mathbf{E}_j = T_{ij} \mathbf{E}^i \mathbf{E}^j = T^i j \mathbf{E}_i \mathbf{E}^j = T_i^j \mathbf{E}^i \mathbf{E}_j$$
(6.38)

Again, we can use the metric tensor to lower and raise indices

$$T^{ij} = g^{ik} T_k{}^j = g^{jl} T^i{}_l = g^{ik} g^{jl} T_{kl}$$

$$T_{ij} = g_{ik} T^k{}_j = g_{jl} T_i{}^l = g_{ik} g_{jl} T^{kl}$$
(6.39)

The tensor product of 2 vectors is a tensor

$$(\mathbf{a}\ \mathbf{b})^{ij} = a^i b^j \tag{6.40}$$

The dot product between a tensor and a vector is

$$(\mathcal{T} \cdot \mathbf{v})^i = T^{ij} V_i \tag{6.41}$$

and between a vector and a tensor

$$(\mathbf{v} \cdot \mathcal{T})^j = T^{ij} V_i \tag{6.42}$$

Finally, the double dot product is a scalar

$$\mathcal{T}: \mathcal{T} = T^{ij}T_{ij} \tag{6.43}$$

All of this can be generalized to higher order tensors. Note that vectors are in fact tensors of order 1.

DIFFERENTIAL OPERATORS

Our goal is to be able to write differential equations using spheroidal coordinates. We will start finding the relation between the partial derivatives with respect to the spherical coordinates and those calculated with respect to the spheroidal coordinates. We will add the primes (') in the notation for the spheroidal θ' and φ' coordinates to clarify the notation, so the derivative with respect to a spheroidal coordinate is done holding the other spheroidal coordinates constant. Following the chain rule

$$\frac{\partial}{\partial r} = \frac{\partial \zeta}{\partial r} \frac{\partial}{\partial \zeta} + \frac{\partial \theta'}{\partial r} \frac{\partial}{\partial \theta'} + \frac{\partial \varphi'}{\partial r} \frac{\partial}{\partial \varphi'}$$
(6.44)

Obviously, $\frac{\partial \theta'}{\partial r} = \frac{\partial \varphi'}{\partial r} = 0$, and

$$dr = r_{\zeta} d\zeta + r_{\theta} d\theta$$

$$\mathrm{d}\zeta = \frac{1}{r_{\zeta}}\mathrm{d}r - \frac{r_{\theta}}{r_{\zeta}}\mathrm{d}\theta$$

where we see $\frac{\partial \zeta}{\partial r} = \frac{1}{r_{\zeta}}$ and $\frac{\partial \zeta}{\partial \theta} = -\frac{r_{\theta}}{r_{\zeta}}$. Then

$$\frac{\partial}{\partial r} = \frac{1}{r_{\zeta}} \frac{\partial}{\partial \zeta} \tag{6.45}$$

The other partial derivatives are calculated in the same way

$$\frac{\partial}{\partial \theta} = \frac{\partial}{\partial \theta'} - \frac{r_{\theta}}{r_{\zeta}} \frac{\partial}{\partial \zeta} \tag{6.46}$$

$$\frac{\partial}{\partial \varphi} = \frac{\partial}{\partial \varphi'} \tag{6.47}$$

Of course, we could take this expressions and substitute them into the expressions for the differential operators corresponding to the spherical coordinates, but there is a much more efficient way to do it. First, let's define the general form of the gradient of a scalar quantity. The gradient will be a vector, whose covariant components are

$$(\nabla \phi)_i = \frac{\partial \phi}{\partial x^i} = \phi_{,i} \tag{6.48}$$

where we have introduced the comma notation for the partial derivative. The contravariant components of the gradient will be

$$(\nabla \phi)^i = g^{ij}\phi_{,j} \tag{6.49}$$

We can also derive a component of a vector V^i in the same way. However, this derivative

$$\frac{\partial V^i}{\partial x^j} = V^i{}_{,j} \tag{6.50}$$

is not a tensor, as it does not transform correctly under a change of coordinates. That's why we will introduce the covariant derivative

$$\nabla_{i} V^{i} = V^{i}_{:j} = V^{i}_{:j} + \Gamma^{i}_{kj} V^{k} \tag{6.51}$$

Where $\Gamma^i_{kj} = \mathbf{E}^i \cdot \frac{\partial \mathbf{E}_k}{\partial x^j}$ is a Christoffel symbol of the second kind. The covariant derivative of a vector $V^i_{;j}$ is a tensor that represents the gradient of the vector.

$$(\nabla \mathbf{v})^{ij} = g^{jk} (\nabla \mathbf{v})^i_{\ k} = g^{jk} V^i_{\ :k} \tag{6.52}$$

We can also calculate the covariant derivative using the covariant components of the vector

$$\nabla_i V_i = V_{i,j} = V_{i,j} - \Gamma^k_{ij} V_k \tag{6.53}$$

The Christoffel symbols can be calculated using the following relation

$$\Gamma_{jk}^{i} = \frac{1}{2}g^{il}(g_{lj,k} + g_{lk,j} - g_{jk,l})$$
(6.54)

where we can see that they are symmetric with respect to the second and third indices $\Gamma^i_{jk} = \Gamma^i_{kj}$. They also verify

$$\Gamma_{ji}^{i} = \frac{\log \sqrt{|g|}}{\partial x^{j}} \tag{6.55}$$

The covariant derivative of second order tensors is done in a similar way

$$\nabla_k T^{ij} = T^{ij}_{:k} = T^{ij}_{:k} + \Gamma^i_{lk} T^{lj} + \Gamma^j_{lk} T^{il}$$
(6.56)

If one of the indices is covariant, then we do

$$\nabla_k T^i_{\ j} = T^i_{\ j,k} = T^i_{\ j,k} + \Gamma^i_{\ lk} T^l_{\ j} - \Gamma^l_{\ jk} T^i_{\ l} \tag{6.57}$$

where we can see the general rule valid also for higher order tensors, the covariant derivative is equal to the regular derivative plus:

- For each contravariant index, we add $\Gamma^{i}_{lk}T^{...l...}$
- For each covariant index, we subtract $\Gamma^l_{ik}T_{...l...}$

Using the covariant derivative, we can calculate all the differential operators in spheroidal coordinates. We have already see the gradient of a scalar and a vector, similarly, the divergence of a vector will be

$$\nabla \cdot \mathbf{v} = \nabla_i V^i = V^i_{:i} \tag{6.58}$$

and for a tensor

$$(\nabla \cdot \mathcal{T})^i = \nabla_j T^{ij} = T^{ij}_{;j} \tag{6.59}$$

Note that some authors prefer the definition $(\nabla \cdot \mathcal{T})^j = \nabla_i T^{ij} = T^{ij}_{,i}$. Using the expression for the cross product, we can calculate the curl of a vector

$$(\nabla \times \mathbf{v})^i = \epsilon^{ijk} \nabla_j V_k = \epsilon^{ijk} V_{k;j} \tag{6.60}$$

The laplacian of a scalar field will be

$$\Delta \phi = \nabla \cdot (\nabla \phi) = \nabla_i (g^{ij} \nabla_i \phi) = (g^{ij} \phi_{,i})_{:i} \tag{6.61}$$

and for a vector field

$$(\Delta \mathbf{v})^i = \nabla_j (g^{jk} \nabla_k V^i) = (g^{jk} V^i_{:k})_{:j}$$

$$(6.62)$$

The material derivative is

$$[(\mathbf{v} \cdot \nabla)\mathbf{v}]^i = V^j \nabla_j V^i = V^j V^i_{:j} \tag{6.63}$$

USEFUL RELATIONS

- Line, area and volume elements
 - Line element

$$ds^{2} = g_{ij}dx^{i}dx^{j} = r_{\zeta}^{2}d\zeta^{2} + 2r_{\zeta}r_{\theta}d\zeta d\theta + (r^{2} + r_{\theta}^{2})d\theta^{2} + r^{2}\sin^{2}\theta d\varphi^{2}$$

$$(6.64)$$

$$d\mathbf{r} = \mathbf{E}_i dx^i = \mathbf{E}_{\zeta} d\zeta + \mathbf{E}_{\theta} d\theta + \mathbf{E}_{\omega} d\varphi \tag{6.65}$$

- Area element in a surface $\zeta = \text{const.}$

$$d\mathbf{S} = (\mathbf{E}_{\theta} \times \mathbf{E}_{\omega}) d\theta d\varphi = r^2 r_{\zeta} \sin \theta \mathbf{E}^{\zeta} d\theta d\varphi \qquad (6.66)$$

$$dS = |d\mathbf{S}| = \sqrt{g^{\zeta\zeta}} r^2 r_{\zeta} \sin\theta d\theta d\varphi = r^2 \sqrt{1 + \frac{r_{\theta}^2}{r^2}} \sin\theta d\theta d\varphi \qquad (6.67)$$

- Area element in a surface of constant $p = p(\zeta, \theta)$

$$d\mathbf{S} = r^2 r_{\zeta} \sin \theta \left(\mathbf{E}^{\zeta} + \frac{p_{,\theta}}{p_{,\zeta}} \mathbf{E}^{\theta} \right) d\theta d\varphi \tag{6.68}$$

$$dS = |d\mathbf{S}| = r^2 r_{\zeta} \sin \theta \sqrt{g^{\zeta\zeta} + 2\frac{p_{,\theta}}{p_{,\zeta}}g^{\zeta\theta} + \left(\frac{p_{,\theta}}{p_{,\zeta}}\right)^2 g^{\theta\theta}} d\theta d\varphi$$
 (6.69)

- Volume element

$$dV = \mathbf{E}_{\zeta} \cdot (\mathbf{E}_{\theta} \times \mathbf{E}_{\varphi}) d\zeta d\theta d\varphi = r^{2} r_{\zeta} \sin \theta d\zeta d\theta d\varphi$$
 (6.70)

• Differential operators

- Gradient

$$\nabla \phi = \phi_{,i} \mathbf{E}^{i} = \frac{\partial \phi}{\partial \zeta} \mathbf{E}^{\zeta} + \frac{\partial \phi}{\partial \theta} \mathbf{E}^{\theta} + \frac{\partial \phi}{\partial \varphi} \mathbf{E}^{\varphi}$$
 (6.71)

- Divergence

$$\nabla \cdot \mathbf{v} = V^{i}_{;i} = \frac{\partial V^{i}}{\partial x^{i}} + \frac{\partial \log \sqrt{|g|}}{\partial x^{k}} V^{k} =$$

$$= \frac{\partial V^{\zeta}}{\partial \zeta} + \left(\frac{2r_{\zeta}}{r} + \frac{r_{\zeta\zeta}}{r_{\zeta}}\right) V^{\zeta} + \frac{\partial V^{\theta}}{\partial \theta} + \left(\frac{2r_{\theta}}{r} + \frac{\cos \theta}{\sin \theta} + \frac{r_{\zeta\theta}}{r_{\zeta}}\right) V^{\theta} + \frac{\partial V^{\varphi}}{\partial \varphi}$$
(6.72)

- Laplacian

$$\Delta \phi = \operatorname{div}(\nabla \phi) = (g^{ij}\phi_{,j})_{;i} = \frac{1}{\sqrt{|g|}} \frac{\partial}{\partial x^{i}} \left(\sqrt{|g|} g^{ij} \frac{\partial \phi}{\partial x^{j}} \right) =$$

$$= g^{\zeta\zeta} \frac{\partial^{2}\phi}{\partial \zeta^{2}} + 2g^{\zeta\theta} \frac{\partial^{2}\phi}{\partial \zeta \partial \theta} + \frac{1}{r^{2}} \frac{\partial^{2}\phi}{\partial \theta^{2}} + \frac{1}{r^{2}\sin^{2}\theta} \frac{\partial^{2}\phi}{\partial \varphi^{2}} +$$

$$+ \left[\frac{2}{rr_{\zeta}} - \frac{r_{\theta\theta}}{r^{2}r_{\zeta}} - g^{\zeta\zeta} \frac{r_{\zeta\zeta}}{r_{\zeta}} - g^{\zeta\theta} \left(\frac{2r_{\zeta\theta}}{r_{\zeta}} - \frac{\cos\theta}{\sin\theta} \right) \right] \frac{\partial\phi}{\partial\zeta} + \frac{\cos\theta}{r^{2}\sin\theta} \frac{\partial\phi}{\partial\theta} \tag{6.73}$$

- Curl

$$\nabla \times \mathbf{v} = \epsilon^{ijk} V_{k;j} \mathbf{E}_i =$$

$$= \frac{1}{r^2 r_{\zeta} \sin \theta} \left[\left(\frac{\partial V_{\varphi}}{\partial \theta} - \frac{\partial V_{\theta}}{\partial \varphi} \right) \mathbf{E}_{\zeta} + \left(\frac{\partial V_{\zeta}}{\partial \varphi} - \frac{\partial V_{\varphi}}{\partial \zeta} \right) \mathbf{E}_{\theta} + \left(\frac{\partial V_{\theta}}{\partial \zeta} - \frac{\partial V_{\zeta}}{\partial \theta} \right) \mathbf{E}_{\varphi} \right]$$
(6.74)

- Material derivative

$$(\mathbf{a} \cdot \nabla)\mathbf{b} = A^j B^i_{\ \ i} \mathbf{E}_i =$$

$$= \left[A^{\zeta} \frac{\partial B^{\zeta}}{\partial \zeta} + A^{\theta} \frac{\partial B^{\zeta}}{\partial \theta} + A^{\varphi} \frac{\partial B^{\zeta}}{\partial \varphi} + \frac{r_{\zeta\zeta}}{r_{\zeta}} A^{\zeta} B^{\zeta} + \left(\frac{r_{\zeta\theta}}{r_{\zeta}} - \frac{r_{\theta}}{r} \right) \left(A^{\zeta} B^{\theta} + A^{\theta} B^{\zeta} \right) + \right. \\ \left. + \frac{1}{r_{\zeta}} \left(r_{\theta\theta} - \frac{2r_{\theta}^{2}}{r} - r \right) A^{\theta} B^{\theta} + \frac{\sin \theta}{r_{\zeta}} \left(r_{\theta} \cos \theta - r \sin \theta \right) A^{\varphi} B^{\varphi} \right] \mathbf{E}_{\zeta} + \\ \left. + \left[A^{\zeta} \frac{\partial B^{\theta}}{\partial \zeta} + A^{\theta} \frac{\partial B^{\theta}}{\partial \theta} + A^{\varphi} \frac{\partial B^{\theta}}{\partial \varphi} + \frac{r_{\zeta}}{r} \left(A^{\zeta} B^{\theta} + A^{\theta} B^{\zeta} \right) + \frac{2r_{\theta}}{r} A^{\theta} B^{\theta} - \sin \theta \cos \theta A^{\varphi} B^{\varphi} \right] \mathbf{E}_{\theta} + \\ \left. + \left[A^{\zeta} \frac{\partial B^{\varphi}}{\partial \zeta} + A^{\theta} \frac{\partial B^{\varphi}}{\partial \theta} + A^{\varphi} \frac{\partial B^{\varphi}}{\partial \varphi} + \frac{r_{\zeta}}{r} \left(A^{\zeta} B^{\varphi} + A^{\varphi} B^{\zeta} \right) + \left(\frac{r_{\theta}}{r} + \frac{\cos \theta}{\sin \theta} \right) \left(A^{\theta} B^{\varphi} + A^{\varphi} B^{\theta} \right) \right] \mathbf{E}_{\varphi} \right.$$

$$(6.75)$$

• Christoffel symbols (different from 0)

$$\Gamma_{\zeta\zeta}^{\zeta} = \frac{r_{\zeta\zeta}}{r_{\zeta}} \qquad \Gamma_{\zeta\theta}^{\zeta} = \frac{r_{\zeta\theta}}{r_{\zeta}} - \frac{r_{\theta}}{r} \qquad \Gamma_{\theta\theta}^{\zeta} = \frac{1}{r_{\zeta}} \left(r_{\theta\theta} - \frac{2r_{\theta}^{2}}{r} - r \right) \\
\Gamma_{\varphi\varphi}^{\zeta} = \frac{\sin\theta}{r_{\zeta}} \left(r_{\theta}\cos\theta - r\sin\theta \right) \qquad \Gamma_{\zeta\theta}^{\theta} = \frac{r_{\zeta}}{r} \qquad \Gamma_{\theta\theta}^{\theta} = \frac{2r_{\theta}}{r} \\
\Gamma_{\varphi\varphi}^{\theta} = -\sin\theta\cos\theta \qquad \Gamma_{\zeta\varphi}^{\varphi} = \frac{r_{\zeta}}{r} \qquad \Gamma_{\theta\varphi}^{\varphi} = \frac{r_{\theta}}{r} + \frac{\cos\theta}{\sin\theta}$$
(6.76)

	Continuous between subdomains			
	Bonazzola	Linear mapping		
\overline{r}	Yes	Yes		
r_{ζ}	Yes	No		
$r_{\zeta} \ r_{\zeta\zeta}$	No	Yes		
$r_{ heta}$	Yes	Yes		
$r_{ heta heta}$	Yes	Yes		
$r_{\zeta\theta}$	Yes	No		

Table 6.1: Note that if the a_i 's change from one domain to the next, r_{ζ} is not continuous even in the Bonazzola et al. mapping.

where $\Gamma^{i}_{jk} = \Gamma^{i}_{kj}$ should be used for the remaining ones.

6.1.3. Multidomain and continuity conditions

When using multidomain one faces the problem of writing continuity conditions for the different variables on the boundaries between contiguous subdomains. The main issue is that the mapping presented in 6.1.1 has discontinuities in some of its derivatives, so we should be careful of using the correct expression for the continuity conditions.

In the case of a scalar field $\phi(r,\theta)$, if ϕ is continuous between subdomains, the condition is simply

$$\phi^{(+)} = \phi^{(-)} \tag{6.77}$$

where (+) and (-) represent each side of the boundary. If we want ϕ to be derivable across the boundary, we have to write a condition on the normal derivative $\hat{\mathbf{n}} \cdot \nabla \phi$ (and not on $\frac{\partial \phi}{\partial \zeta}$), namely,

$$\hat{\mathbf{n}} \cdot \nabla^{(+)} \phi^{(+)} = \hat{\mathbf{n}} \cdot \nabla^{(-)} \phi^{(-)} \tag{6.78}$$

where

$$\hat{\mathbf{n}} \cdot \nabla \phi = \frac{\mathbf{E}^{\zeta}}{\sqrt{g^{\zeta\zeta}}} \cdot \left(\frac{\partial \phi}{\partial \zeta} \mathbf{E}^{\zeta} + \frac{\partial \phi}{\partial \theta} \mathbf{E}^{\theta} + \frac{\partial \phi}{\partial \varphi} \mathbf{E}^{\varphi} \right)$$

$$= \sqrt{g^{\zeta\zeta}} \frac{\partial \phi}{\partial \zeta} + \frac{g^{\zeta\theta}}{\sqrt{g^{\zeta\zeta}}} \frac{\partial \phi}{\partial \theta}$$

$$= \sqrt{1 + \frac{r_{\theta}^{2}}{r^{2}}} \left(\frac{1}{r_{\zeta}} \frac{\partial \phi}{\partial \zeta} - \frac{r_{\theta}}{r^{2} + r_{\theta}^{2}} \frac{\partial \phi}{\partial \theta} \right)$$
(6.79)

We know that r and r_{θ} are continuous across the boundary and if ϕ is continuous then so is $\frac{\partial \phi}{\partial \theta}$, thus condition (6.78) becomes

$$\frac{1}{r_{\zeta}^{(+)}} \left(\frac{\partial \phi}{\partial \zeta} \right)^{(+)} = \frac{1}{r_{\zeta}^{(-)}} \left(\frac{\partial \phi}{\partial \zeta} \right)^{(-)} \tag{6.80}$$

that is equivalent to saying $\left(\frac{\partial \phi}{\partial r}\right)^{(+)} = \left(\frac{\partial \phi}{\partial r}\right)^{(-)}$. In general $r_{\zeta}^{(+)} \neq r_{\zeta}^{(-)}$.

In the case of a vector field, the conditions are

$$r_{\zeta}^{(+)}V^{\zeta^{(+)}} = r_{\zeta}^{(-)}V^{\zeta^{(-)}}$$

$$V^{\theta^{(+)}} = V^{\theta^{(-)}}$$

$$V^{\varphi^{(+)}} = V^{\varphi^{(-)}}$$
(6.81)

for continuity and

$$\left(\frac{\partial V^{\zeta}}{\partial \zeta}\right)^{(+)} + \frac{1}{r_{\zeta}^{(+)}} \left(r_{\zeta\zeta}^{(+)} V^{\zeta^{(+)}} + r_{\zeta\theta}^{(+)} V^{\theta^{(+)}}\right) = \left(\frac{\partial V^{\zeta}}{\partial \zeta}\right)^{(-)} + \frac{1}{r_{\zeta}^{(-)}} \left(r_{\zeta\zeta}^{(-)} V^{\zeta^{(-)}} + r_{\zeta\theta}^{(-)} V^{\theta^{(-)}}\right)
\frac{1}{r_{\zeta}^{(+)}} \left(\frac{\partial V^{\theta}}{\partial \zeta}\right)^{(+)} = \frac{1}{r_{\zeta}^{(-)}} \left(\frac{\partial V^{\theta}}{\partial \zeta}\right)^{(-)}
\frac{1}{r_{\zeta}^{(+)}} \left(\frac{\partial V^{\varphi}}{\partial \zeta}\right)^{(+)} = \frac{1}{r_{\zeta}^{(-)}} \left(\frac{\partial V^{\varphi}}{\partial \zeta}\right)^{(-)}$$
(6.82)

for derivability. These conditions are based on the fact that the spherical components of the vector field are continuous and derivable scalar functions. For a physical boundary, it could happen that some components of the vector field are continuous (or derivable) and others are not, then none of the above conditions is correct.

6.2. Coordinate mapping in ESTER

The class mapping combines a diff_gl object with a diff_leg object to perform calculations in 2D. The prototype of this class is defined in mapping.h.

Let's see an example of initialization of a mapping object

```
mapping map;
map.set_ndomains(3);
map.set_npts(20);
map.set_nt(25);
map.set_nex(20);
map.init();

map.R.setrow(1,0.5*ones(1,map.nt));
map.R.setrow(2,1+0.1*sin(map.th)*sin(map.th));
map.R.setrow(3,2+0.3*sin(map.th)*sin(map.th));
map.remap();
```

First, we have declared the object

```
mapping map;
```

Then we set the number of domains

```
map.set_ndomains(3);
```

Our mapping will have 3 domains plus one external domain. We choose the number of points in the internal domains

```
map.set_npts(20);
```

That is, the 3 domains will have each one 20 points. We can also set a different number of points for each domain, for that we should do

```
int npts[3];
npts[0]=10;npts[1]=20;npts[2]=30;
map.set_npts(npts);
```

After that, we set the number of points in latitude (25) and in the external domain (20)

```
map.set_nt(25);
map.set_nex(20);
```

If we are not using the external domain, there is no need to change its number of points (It is 10 by default). There is one optional parameter that we can set before initialize the mapping

```
map.mode=MAP_BONAZZOLA;
```

to use the mapping with constant r_{ζ} at the boundaries presented in the previous section (this is the default), or

```
map.mode=MAP_LINEAR;
```

to use a linear mapping. Now we can proceed with the initialization

map.init();

If we change the resolution after this point, we should call init() again. Now we have a working object, with 3 spherical domains distributed uniformly from r=0 to r=1. If this setup is fine for us, then we are done with the initialization, but usually we will want to change to boundaries between domains. This is done using the matrix R with size $(n_{\text{domains}}+1)\times n_{\theta}$. Each row of R defines a different boundary, starting with the row 0 that corresponds to the inner boundary. In the previous example, we have not changed the inner boundary, so our mapping contains the center of coordinates, but if we want the mapping to start, for example, at r=0.3, we could just add

map.R.setrow(0,0.3*ones(1,map.nt));

In the definition of the boundaries we have used map.th that contains the values of the θ coordinate. Once we have set the boundaries we just call remap()

map.remap();

We can call remap() as many times as we want if we need to change the boundaries again. After the initialization, a mapping object contains the following variables

Name	Type	Size	Description
ndomains	int		Number of domains
nr	int		Number of radial points
nt	int		Number of latitudinal points
nex	int		Number of radial points in the external domain
npts	<pre>int[ndomains]</pre>		Number of points in each domain
R	matrix	(ndomains+1,nt)	Domain boundaries
eta	matrix	(ndomains+1,nt)	$R(\theta = 0)$
z	matrix	(nr,1)	Spheroidal radial coordinate ζ
th	matrix	(1,nt)	Colatitude θ
r	matrix	(nr,nt)	Spherical radial coordinate r
ex.r	matrix	(nex,nt)	Spherical radial coordinate r in the external domain
$\mathtt{rz},\mathtt{rt},\mathtt{rzz},\mathtt{rtt},\mathtt{rzt}$	matrix	(nr,nt)	$r_{\zeta}, r_{\theta}, r_{\zeta\zeta}, r_{\theta\theta}, r_{\zeta\theta}$
ex.rz, ex.rt, ex.rzz, ex.rtt, ex.rzt	matrix	(nex,nt)	$r_{\zeta}, r_{\theta}, r_{\zeta\zeta}, r_{\theta\theta}, r_{\zeta\theta}$ in the external domain
gzz, gzt, gtt	matrix	(nr,nt)	Elements of the metric tensor $g^{\zeta\zeta}, g^{\zeta\theta}$ and $g^{\theta\theta}$
ex.gzz, ex.gzt, ex.gtt	matrix	(nex,nt)	Elements of the metric tensor $g^{\zeta\zeta}$, $g^{\zeta\theta}$ and $g^{\theta\theta}$ in the external domain
D	$matrix_block_diag$	(nr,nr)	Differentiation matrix $\frac{\partial}{\partial \zeta}$

ex.D	matrix	(nex,nex)	Differentiation matrix $\frac{\partial}{\partial \zeta}$ in the external domain
Dt, Dt_11, Dt_01, Dt_10	matrix	(nt,nt)	Differentiation matrix $\frac{\partial}{\partial \theta}$ for each type of symmetry
Dt2, Dt2_11, Dt2_01, Dt2_10	matrix	(nt,nt)	Second order differentiation matrix $\frac{\partial^2}{\partial \theta^2}$ for each type of symmetry
I	matrix	(1,nr)	Integration matrix $\int_{\eta_0}^{\eta_{\text{ndom.}}} d\zeta$
It	matrix	(nt,1)	Integration matrix $\int_0^{\pi} \sin \theta d\theta$ for symmetric functions
J	matrix[4]	(nr,nt)	Jacobian of the mapping
ex.J	matrix[4]	(nex,nt)	Jacobian of the mapping in the external domain
gl	$diff_{-}gl$		Numerical differentiation object for the radial direction
ex.gl	diff_gl		Numerical differentiation object for the radial direction in the ex- ternal domain
leg	${\tt diff_leg}$		Numerical differentiation object for the latitudinal direction

A given scalar field $\phi(\zeta, \theta)$ will be represented by a 2D matrix with size $\mathtt{nr} \times \mathtt{nt}$, each element being the value of the function at each collocation point $\phi_{ij} = \phi(\zeta_i, \theta_j)$. With this representation, the operators acting on the radial direction are implemented using left multiplication while those acting in the latitudinal direction use right multiplication. Let's see some examples

$$egin{array}{lll} rac{\partial \phi}{\partial \zeta} & : & (exttt{map.D,phi}) \ \int_{\eta_0}^{\eta_{ exttt{ndom.}}} \phi \, \mathrm{d} \zeta & : & (exttt{map.I,phi}) \ & rac{\partial \phi}{\partial \theta} & : & (exttt{phi,map.Dt}) \ & \int_0^{\pi} \phi \sin \theta \, \mathrm{d} \theta & : & (exttt{phi,map.It}) \end{array}$$

We can also use operators in both sides at the same time

$$\frac{\partial^2 \phi}{\partial \zeta \partial \theta} \qquad : \qquad \qquad (\texttt{map.D,phi,map.Dt}) \\ \int_{\eta_0}^{\eta_{\text{ndom.}}} \int_0^\pi \phi r^2 r_\zeta \sin\theta \, \mathrm{d}\zeta \mathrm{d}\theta \quad : \quad (\texttt{map.I,phi*map.r*map.r*map.rz,map.It})$$

Most of the members of a mapping object are in fact references to the corresponding member of the diff_gl or diff_leg object presented in the previous chapter, for example map.D is equivalent to map.gl.D.

We can use the interpolation functions defined in diff_gl and diff_leg for interpolation in ζ and θ respectively. The mapping class provides also a function for interpolating at some points (r_{ij}, θ_{ij}) given in spherical coordinates

```
map.eval(phi,ri,thi,parity)
```

where ri and thi are matrices containing the interpolation points and parity is an integer representing the type of symmetry of phi (00, 01, 10 or 11). For symmetric functions (00), this parameter can be omitted.

6.2.1. Example

Let's take the example in section 5.3.1 and rewrite it for spheroidal coordinates, now using the mapping class.

```
/* The following example illustrates the use of the mapping
library in spheroidal coordinates */
#include<stdio.h>
#include"mapping.h"
#include"constants.h" //For the defintion of PI
//Function prototypes
double laplacian(matrix y,double r0,double th0);
double integral(matrix y);
// Define mapping objects as global variable
mapping map;
// Create references for spherical coordinates
matrix &r=map.r,&th=map.th;
int main() {
    //Initialize map. In the example we will use 2 domains
    map.set_ndomains(2);
   map.set_npts(100);
   map.set_nt(50);
    //map.set_nex(20); // We won't use the external domain
    map.init();
    map.R.setrow(0,1e-3*ones(1,map.nt));
                            // Use 1e-3 as the interior limit (instead of 0)
                            // to avoid a division by zero in the
                            // calculation of the laplacian
    map.R.setrow(1,0.5+0.1*sin(th)*sin(th));
    map.R.setrow(2,ones(1,map.nt));
    map.remap();
    matrix y;
    //Define the function y
    y=r*r*r*(1+sin(th)*sin(th));
    double lap_y,int_y;
```

```
double r0=0.3,th0=PI/3;
    lap_y=laplacian(y,r0,th0);
    int_y=integral(y);
    printf("The value of the laplacian at (%f, %f) is %e\n", r0, th0, lap_y);
    printf("The volume integral is %e\n",int_y);
    return 0;
}
// Function for calculating the laplacian of y at (r0,th0)
double laplacian(matrix y,double r0,double th0) {
   matrix lap_y;
   matrix &gzz=map.gzz,&gzt=map.gzt,&gtt=map.gtt;
   matrix &rz=map.rz,&rzz=map.rzz,&rzt=map.rzt,&rt=map.rt,&rtt=map.rtt;
   matrix &Dt=map.Dt,&Dt2=map.Dt2;
   matrix_block_diag &D=map.D;
    lap_y=gzz*(D,D,y)+2*gzt*(D,y,Dt)+(y,Dt2)/r/r
        +(2./r/rz-rtt/r/r/rz-gzz*rzz/rz-gzt*(2*rzt/rz-cos(th)/sin(th)))*(D,y)
        +\cos(th)/r/r/\sin(th)*(y,Dt);
    lap_y=map.eval(lap_y,r0*ones(1,1),th0*ones(1,1));
    return lap_y(0);
// Function for calculating the volume integral of y
double integral(matrix y) {
    return 2*PI*(map.I,r*r*map.rz*y,map.It)(0);
```

The output will be the same that in the previous example

The value of the laplacian at (0.300000, 1.047198) is 6.150000e+00 The volume integral is 3.490659e+00

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