STRUCTURE OF THE LEEDS DYNAMO CODE

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Governing equations

$$(Ro \partial_t - E \nabla^2) \mathbf{u} = \mathbf{N}_u - \mathbf{\nabla} \hat{p},$$

$$(\partial_t - \nabla^2) \mathbf{B} = \mathbf{N}_B,$$

$$(\partial_t - q \nabla^2) C = N_C.$$

Toriodal-poloidal decomposition

$$\boldsymbol{u} = \boldsymbol{\nabla} \times (T\boldsymbol{r}) + \boldsymbol{\nabla} \times \boldsymbol{\nabla} \times (P\boldsymbol{r}),$$

 $\boldsymbol{B} = \boldsymbol{\nabla} \times (\mathcal{T}\boldsymbol{r}) + \boldsymbol{\nabla} \times \boldsymbol{\nabla} \times (\mathcal{P}\boldsymbol{r}).$

Numerical formulation

$$(\partial_t - \nabla^2)\mathcal{P} = N_{\mathcal{P}}$$

$$(\partial_t - \nabla^2)\mathcal{T} = N_{\mathcal{T}}$$

$$(Ro \, \partial_t - E\nabla^2)\mathcal{T} = N_{\mathcal{T}},$$

$$(Ro \, \partial_t - E\nabla^2)\mathcal{P} = g$$

$$-\nabla^2 g = N_{\mathcal{P}}$$

$$(\partial_t - q\nabla^2)\mathcal{C} = N_{\mathcal{C}}.$$

Model equation

$$(a\,\partial_t - b\nabla^2)f = N$$

- A lot of code-sharing.
- narrow f.d. stencil or higher-order approximation for a given width.
- good stability.
- Each 2nd-order + 2 BCs, except system

$$\begin{cases} (Ro \partial_t - E\nabla^2)P = g \\ -\nabla^2 g = N_P \end{cases} (r_i < r < r_o) \qquad P = \partial_r P = 0 \ (r = r_i, r_o),$$

which has 4 BCs on P, none on g.

Green's function or influence-matrix method

$$\begin{cases} (Ro \, \partial_t - E \nabla^2) P &= g \\ -\nabla^2 g &= N_P \end{cases} (r_i < r < r_o) \qquad P = \partial_r P = 0 \ (r = r_i, r_o).$$

Write as system

$$\begin{cases} \mathbf{X} P = g \\ \mathbf{Y} g = N \end{cases} \qquad P = 0, 0 \quad \partial_r P = 0, 0 \quad (r = \mathbf{r_i}, \mathbf{r_o}).$$

Let $P = \bar{P} + a P_i + b P_o$, where

$$\begin{cases} \mbox{X} P_i &= g_i & \partial_r P_i = 0, 0 \\ \mbox{Y} g_i &= 0 & g_i = 1, 0 \end{cases}$$

$$\begin{cases} \mbox{X} P_o &= g_o & \partial_r P_o = 0, 0 \\ \mbox{Y} g_o &= 0 & g_o = 0, 1 \end{cases}$$

$$\begin{cases} \mbox{X} \bar{P} &= \bar{g} & \partial_r \bar{P} = 0, 0 \\ \mbox{Y} \bar{g} &= N & \bar{g} = 0, 0 \end{cases}$$

- P_i and P_o both precomputable.
- Only inverted for \bar{P} each timestep.
- ullet P satisfies two BCs automatically: $\partial_r P = 0, 0$.
- Scalars a and b set by remaining two BCs: P=0,0. (Requires inversion of a 2×2 matrix.)
- BCs on another variable or coupling BCs satisfied to numerical precision.
- Computational overhead is nominal.

Modules

Somewhere to gather together related data and functions that work on that data. Don't worry about syntax here!

```
module rotation
! ****************************
 use parameters, timestep ! use data from other modules
 implicit none
 save
                    ! data other mods can inherit
 double precision :: rot_omega
 double precision :: rot_torque
                             ! private data
 double precision, private :: rot_inertia
contains
I -----
! initialise
 subroutine rot_precompute()
   rot_omega = 0d0
   rot_inertia = (8d0*d_PI/15d0) * d_ICradius**5
 end subroutine rot_precompute
timestep
1-----
                -----
 subroutine rot_predictor()
   rot_omega = rot_omega &
     + (tim_dt/rot_inertia) * rot_torque
 end subroutine rot_predictor
end module rotation
```

Modules:

codensity, (IC) rotation, these variables stored here velocity, magnetic + predictor-corrector functions

timestep functions for setting up matrices for model eqn

 $(a\partial_t - b\nabla^2) f = N.$

nonlinear evaluate nonlinear (coupling) terms N.

mesh selection of radial points r_n

legendre weights for transforms.

variables definition of data types (coll), (spec), (phys)

transform between spherical harmonic coeffs

and data at points in physical space.

The 'main' program:

parameters physical and numerical parameters

main the main time-stepping loop

io input / output

A 'derived' data type:

Define type:

Declare a variable: type (phys) :: p

Set value of an element: p%Re(j,i,n) = 1d0

- No need to specify the dimension of every variable with the same size
- Very handy when passing between functions.
- Fortran column dominant keep 'close' data in first index, split over last index...

Legendre transform strongly linked to data types

$$A(\theta, \phi) = \sum_{m=0}^{M-1} \sum_{l=m}^{L-1} A_{lm} \hat{Y}_l^m(\theta, \phi)$$

$$A(r_n, \theta, \phi) = \sum_{m} e^{im\phi} \sum_{l} A_{lmn} P_l^{|m|}(\cos \theta)$$

$$A(r_n, \theta_j, \phi) = \sum_{m} e^{im\phi} A_{jmn}$$

$$A(r_n, \theta_j, \phi_i) = A_{jin}$$

(coll) $(A_n)_{lm}$ Data at collocation point for each harmonic. All n on same CPU (split accross modes lm)

 $\rightarrow \text{transpose} \rightarrow$

(spec) $(A_{lm})_n$ All spectral coeffs on same CPU (split accross radial pts) sum over l at each $\theta_j \to (A_{jm})_n$

sum over t at each $\theta_j \to (A_{jm})$ sum over m at ϕ_i (FFT) \to

(phys) $(A_{ij})_n$ data evaluated on points in physical space.

Strengths

- Short 3000 lines excluding IO. Readable modular, function-based.
- Data easily manipulated function-based, netCDF.
- Fast in serial.
- Timestep control (improved by Chris Jones).
- Parallelised, linear scaling with number of CPUs.

Weakness

Number of CPUs currently limited to number of radial points.
 (See index n on (spec))

Possible way out:

$$(A_n)_{lm}$$
; (transpose) \to $(A_l)_{mn}$; (sum over l) \to $(A_j)_{mn}$; (transpose) \to $(A_m)_{jn}$; (FFT) \to $(A_i)_{jn}$.

The 'derived' types:

Currently split over radial points, $n \in [1, N]$, and harmonics lm mapped to a single index $nh \in [0:H-1]$.

```
nh = -1
do m = 0, M-1
do l = m, L-1
nh = nh + 1
```