

# Example Programs for IDA v5.7.0

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September 1, 2021



UCRL-SM-208112

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This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under Contract DE-AC52-07NA27344.

## CONTRIBUTORS

The SUNDIALS library has been developed over many years by a number of contributors. The current SUNDIALS team consists of Cody J. Balos, David J. Gardner, Alan C. Hindmarsh, Daniel R. Reynolds, and Carol S. Woodward. We thank Radu Serban for significant and critical past contributions.

Other contributors to SUNDIALS include: James Almgren-Bell, Lawrence E. Banks, Peter N. Brown, George Byrne, Rujeko Chinomona, Scott D. Cohen, Aaron Collier, Keith E. Grant, Steven L. Lee, Shelby L. Lockhart, John Loffeld, Daniel McGreer, Slaven Peles, Cosmin Petra, H. Hunter Schwartz, Jean M. Sexton, Dan Shumaker, Steve G. Smith, Allan G. Taylor, Hilari C. Tiedeman, Chris White, Ting Yan, and Ulrike M. Yang.



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# 1 Introduction

This report is intended to serve as a companion document to the User Documentation of IDA [6]. It provides details, with listings, on the example programs supplied with the IDA distribution package.

The IDA distribution contains examples of four types: serial C examples, parallel C examples, FORTRAN examples, PETSc examples, and Trilinos examples. With the exception of “demo”-type example files, the names of all the examples distributed with SUNDIALS are of the form `[slv][PbName]_[ls]_[prec]_[p]`, where

`[slv]` identifies the solver (for IDA examples this is `ida`, while for FIDA examples, this is `fida`);

`[PbName]` identifies the problem;

`[ls]` identifies the linear solver module used;

`[prec]` indicates the IDA preconditioner module used (if applicable — for examples using a Krylov linear solver and the IDABBDPRE module, this will be `bbd`);

`[p]` indicates an example using the parallel vector module `NVECTOR_PARALLEL`.

The following lists summarize all examples distributed with IDA.

The IDA distribution contains, in the `srcdir/examples/ida/serial` directory, the following nine serial examples (using the `NVECTOR_SERIAL` module):

- `idaRoberts_dns` solves the Robertson chemical kinetics problem [8], which consists of two differential equations and one algebraic constraint. It also uses the rootfinding feature of IDA.

The problem is solved with the `SUNLINSOL_DENSE` linear solver using a user-supplied Jacobian.

- `idaRoberts_klu` is the same as `idaRoberts_dns` but uses the `KLU` sparse direct linear solver.
- `idaRoberts_sps` is the same as `idaRoberts_dns` but uses the `SuperLUMT` sparse direct linear solver (with one thread).
- `idaSlCrank_dns` solves a system of index-2 DAEs, modeling a planar slider-crank mechanism.

The problem is obtained through a stabilized index reduction (Gear-Gupta-Leimkuhler) starting from the index-3 DAE equations of motion derived using three generalized coordinates and two algebraic position constraints.

- `idaHeat2D_bnd` solves a 2-D heat equation, semidiscretized to a DAE on the unit square. This program solves the problem with the `SUNLINSOL_BAND` linear solver and the default difference-quotient Jacobian approximation. For purposes of illustration, `IDACalcIC` is called to compute correct values at the boundary, given incorrect values as input initial guesses. The constraint  $u > 0.0$  is imposed for all components.

- `idaHeat2D_kry` solves the same 2-D heat equation problem as `idaHeat2D_bnd`, with the Krylov linear solver `SUNLINSOL_SPGMR`. The preconditioner uses only the diagonal elements of the Jacobian.
- `idaHeat2D_klu` solves the same 2-D heat equation problem as `idaHeat2D_bnd`, with sparse linear solver `SUNLINSOL_KLU`.
- `idaHeat2D_sps` solves the same 2-D heat equation problem as `idaHeat2D_bnd`, with sparse linear solver `SuperLUMT`.
- `idaFoodWeb_bnd` solves a system of PDEs modelling a food web problem, with predator-prey interaction and diffusion, on the unit square in 2-D, using the band linear solver.
- `idaFoodWeb_kry` solves the same problem as `idaFoodWeb_bnd`, but with `SUNLINSOL_SPGMR` and a user-supplied preconditioner.

The PDEs are discretized in space to a system of DAEs which are solved using the `SUNLINSOL_BAND` linear solver with the default difference-quotient Jacobian approximation.

- `idaKrylovDemo_ls` solves the same problem as `idaHeat2D_kry`, with three Krylov linear solvers `SUNLINSOL_SPGMR`, `SUNLINSOL_SPBCGS`, and `SUNLINSOL_SPTFQMR`. The preconditioner uses only the diagonal elements of the Jacobian.

In the `srcdir/examples/ida/parallel` directory, the IDA distribution contains the following four parallel examples (using the `NVECTOR_PARALLEL` module):

- `idaHeat2D_kry_p` solves the same 2-D heat equation problem as `idaHeat2D_kry`, with `SUNLINSOL_SPGMR` in parallel, and with a user-supplied diagonal preconditioner,
- `idaHeat2D_kry_bbd_p` solves the same problem as `idaHeat2D_kry_p`.

This program uses the Krylov linear solver `SUNLINSOL_SPGMR` in parallel, and the band-block-diagonal preconditioner `IDABBDPRE` with half-bandwidths equal to 1.

- `idaFoodWeb_kry_p` solves the same food web problem as `idaFoodWeb_bnd`, but with `SUNLINSOL_SPGMR` and a user-supplied preconditioner.

The preconditioner supplied to `SUNLINSOL_SPGMR` is the block-diagonal part of the Jacobian with  $n_s \times n_s$  blocks arising from the reaction terms only ( $n_s$  = number of species).

- `idaFoodWeb_kry_bbd_p` solves the same food web problem as `idaFoodWeb_kry_p`.

This program solves the problem using `SUNLINSOL_SPGMR` in parallel and the `IDABBDPRE` preconditioner.

As part of the FIDA module, in the four subdirectories `fcmix_serial`, `fcmix_parallel`, `fcmix_openmp`, and `fcmix_pthreads`, within the directory `srcdir/examples/ida`, are the following four examples for the FORTRAN-C interface:

- `fidaRoberts_dns` is a serial chemical kinetics example (`DENSE`) with rootfinding, equivalent to `idaRoberts_dns`.



- `fidaHeat2D_kry_bbd_p` is a parallel example (SPGMR/IDABBPRE) equivalent to the example `idaHeat2D_kry_bbd_p`.
- `fidaRoberts_dns_openmp` is the same as `fidaRoberts_dns` but uses the NVECTOR module NVECTOR\_OPENMP.
- `fidaRoberts_dns_pthreads` is the same as `fidaRoberts_dns` but uses the NVECTOR module NVECTOR\_PTHREADS.

Finally, in the subdirectory `petsc` of `examples/ida` are the following examples:

- `idaHeat2D_kry_petsc` solves the same problem as `idaHeat2D_kry` (with SPGMR) but using the PETSc vector module.
- `idaHeat2D_jac_petsc` solves the same problem as `idaHeat2D_kry` but using the default PETSc Krylov solver and the PETSc vector module.

In the following sections, we give detailed descriptions of some (but not all) of these examples. We also give our output files for each of these examples, but users should be cautioned that their results may differ slightly from these. Solution values may differ within tolerances, and differences in cumulative counters, such as numbers of steps or Newton iterations, may differ from one machine environment to another by as much as 10% to 20%.

In the descriptions below, we make frequent references to the IDA User Document [6]. All citations to specific sections (e.g. §4.2) are references to parts of that User Document, unless explicitly stated otherwise.

**Note.** The examples in the IDA distribution are written in such a way as to compile and run for any combination of configuration options during the installation of SUNDIALS (see Appendix A in the User Guide). As a consequence, they contain portions of code that will not be typically present in a user program. For example, all example programs make use of the variables `SUNDIALS_EXTENDED_PRECISION` and `SUNDIALS_DOUBLE_PRECISION` to test if the solver libraries were built in extended or double precision, and use the appropriate conversion specifiers in `printf` functions.

## 2 Serial example problems

### 2.1 A dense example: `idaRoberts_dns`

This example, due to Robertson [8], is a model of a three-species chemical kinetics system written in DAE form. Differential equations are given for species  $y_1$  and  $y_2$  while an algebraic equation determines  $y_3$ . The equations for the species concentrations  $y_i(t)$  are:

$$\begin{cases} y_1' &= -.04y_1 + 10^4y_2y_3 \\ y_2' &= +.04y_1 - 10^4y_2y_3 - 3 \cdot 10^7y_2^2 \\ 0 &= y_1 + y_2 + y_3 - 1. \end{cases} \quad (1)$$

The initial values are taken as  $y_1 = 1$ ,  $y_2 = 0$ , and  $y_3 = 0$ . This example computes the three concentration components on the interval from  $t = 0$  through  $t = 4 \cdot 10^{10}$ . While integrating the system, the program also use the rootfinding feature to find the points at which  $y_1 = 10^{-4}$  or at which  $y_3 = 0.01$ .

We give a rather detailed explanation of the parts of the program and their interaction with IDA.

Following the initial comment block, this program has a number of `#include` lines, which allow access to useful items in IDA header files. The `sundials_types.h` file provides the definition of the type `realtype` (see §4.2 in the user guide [6] for details). For now, it suffices to read `realtype` as `double`. The `ida.h` file provides prototypes for the IDA functions to be called (excluding the linear solver selection function), and also a number of constants that are to be used in setting input arguments and testing the return value of `IDASolve`. The `nvector_serial.h` file is the header file for the serial implementation of the `NVECTOR` module and includes definitions of the `N_Vector` type, a macro to access vector components, and prototypes for the serial implementation specific machine environment memory allocation and freeing functions. Finally, note that the include files `sunmatrix_dense.h` and `sunlinsol_dense.h` include definition of the dense matrix and linear solver modules, as well as a macro for accessing matrix elements.

This program includes the user-defined accessor macro `IJth` that is useful in writing the problem functions in a form closely matching the mathematical description of the DAE system, i.e. with components numbered from 1 instead of from 0. The `IJth` macro is used to access elements of a dense matrix of type `sunmatdense`. It is defined using the accessor macro `SM_ELEMENT_D` which numbers matrix rows and columns starting with 0. The macro `SM_ELEMENT_D` is fully described in §8.3.

The program prologue ends with prototypes of the three user-supplied functions that are called by IDA and the prototypes of five private functions. Of the latter, the four `Print***` functions perform printing operations, and `check_flag` tests the return flag from the IDA user-callable functions.

After various declarations, the `main` program begins by allocating memory for the `yy`, `yp`, and `avtol` vectors using `N_VNew_Serial` with a length argument of `NEQ` ( $= 3$ ). The lines following that load the initial values of the dependent variable vectors into `yy` and `yp`, and set the relative tolerance `rtol` and absolute tolerance vector `avtol`. Serial `N_Vector` values are set by first accessing the pointer to their underlying data using the macro `NV_DATA_S` defined by `NVECTOR_SERIAL` in `nvector_serial.h`.

The calls to `N_VNew_Serial`, and also later calls to `IDA***` functions, make use of a private function, `check_flag`, which examines the return value and prints a message if there was a failure. This `check_flag` function was written to be used for any serial SUNDIALS application.

The call to `IDACreate` creates the IDA solver memory block. The return value of this function is a pointer to the memory block for this problem. In the case of failure, the return value is `NULL`. This pointer must be passed in the remaining calls to IDA functions.

The call to `IDAInit` allocates the solver memory block. Its arguments include the name of the C function `resrob` defining the residual function  $F(t, y, y')$ , and the initial values of  $t$ ,  $y$ , and  $y'$ . The call to `IDASVtolerances` specifies a vector of absolute tolerances, and this call includes the relative tolerance `rtol` and the absolute tolerance vector `avtol`. See §4.5.1 and §4.5.2 for full details of these calls. (The `avtol` vector is then freed, because IDA keeps a separate copy of it.)

The call to `IDARootInit` specifies that a rootfinding problem is to be solved along with the integration of the DAE system, that the root functions are specified in the function `grob`, and that there are two such functions. Specifically, they are set to  $y_1 - 0.0001$  and  $y_3 - 0.01$ , respectively. See §4.5.6 for a detailed description of this call.

The calls to `SUNDenseMatrix` (see §8.3), `SUNLinSol_Dense` (see §9.5), `IDASetLinearSolver` (see §4.5.3) and `IDASetJacFn` (see §4.5.8.2) specify the `SUNLINSOL_DENSE` linear solver with an analytic Jacobian supplied by the user-supplied function `jacrob`.

The actual solution of the DAE initial value problem is accomplished in the loop over values of the output time `tout`. In each pass of the loop, the program calls `IDASolve` in the `IDA_NORMAL` mode, meaning that the integrator is to take steps until it overshoots `tout` and then interpolate to  $t = \text{tout}$ , putting the computed value of  $y(\text{tout})$  and  $y'(\text{tout})$  into `yy` and `yp`, respectively, with `tret = tout`. The return value in this case is `IDA_SUCCESS`. However, if `IDASolve` finds a root before reaching the next value of `tout`, it returns `IDA_ROOT_RETURN` and stores the root location in `tret` and the solution there in `yy` and `yp`. In either case, the program prints  $t$  ( $= \text{tret}$ ) and `yy`, and also the cumulative number of steps taken so far, and the current method order and step size. In the case of a root, the program calls `IDAGetRootInfo` to get a length-2 array `rootsfound` of bits showing which root function was found to have a root. If `IDASolve` returned any negative value (indicating a failure), the program breaks out of the loop. In the case of a `IDA_SUCCESS` return, the value of `tout` is advanced (multiplied by 10) and a counter (`iout`) is advanced, so that the loop can be ended when that counter reaches the preset number of output times, `NOUT = 12`. See §4.5.7 for full details of the call to `IDASolve`.

Finally, the main program calls `PrintFinalStats` to extract and print several relevant statistical quantities, such as the total number of steps, the number of residual and Jacobian evaluations, and the number of error test and convergence test failures. It then calls `IDAFree` to free the IDA memory block and `N_VDestroy_Serial` to free the vectors `yy` and `yp`.

The function `PrintFinalStats` used here is actually suitable for general use in applications of IDA to any problem with a dense Jacobian. It calls various `IDAGet***` functions to obtain the relevant counters, and then prints them. Specifically, these are: the cumulative number of steps (`nst`), the number of residual evaluations (`nre`) (excluding those for difference-quotient Jacobian evaluations), the number of residual evaluations for Jacobian evaluations (`nreLS`), the number of Jacobian evaluations (`nje`), the number of nonlinear (Newton) iterations (`nni`), the number of local error test failures (`netf`), the number of nonlinear convergence failures (`ncfn`), and the number of `grob` (root function) evaluations (`nge`). These optional outputs are described in §4.5.10.

The functions `resrob` (of type `IDAResFn`) and `jacrob` (of type `IDALsJacFn`) are straightforward expressions of the DAE system (1) and its system Jacobian. The function `jacrob` makes use of the macro `IJth` discussed above. See §4.6.1 for detailed specifications of `IDAResFn`. Similarly, the function `grob` defines the two functions,  $g_0$  and  $g_1$ , whose roots

are to be found. See §4.6.4 for a detailed description of the `grob` function.

The output generated by `idaRoberts_dns` is shown below. It shows the output values at the 12 preset values of `tout`. It also shows the two root locations found, first at a root of  $g_1$ , and then at a root of  $g_0$ .

idaRoberts_dns sample output						
idaRoberts_dns: Robertson kinetics DAE serial example problem for IDA Three equation chemical kinetics problem.						
Linear solver: DENSE, with user-supplied Jacobian.						
Tolerance parameters: rtol = 0.0001 atol = 1e-08 1e-06 1e-06						
Initial conditions y0 = (1 0 0)						
Constraints and id not used.						
t	y1	y2	y3	nst	k	h
2.6402e-01	9.8997e-01	3.4706e-05	1.0000e-02	27	2	4.4012e-02
rootsfound[] = 0 1						
4.0000e-01	9.8517e-01	3.3864e-05	1.4794e-02	29	3	8.8024e-02
4.0000e+00	9.0553e-01	2.2406e-05	9.4452e-02	43	4	6.3377e-01
4.0000e+01	7.1579e-01	9.1838e-06	2.8420e-01	68	4	3.1932e+00
4.0000e+02	4.5044e-01	3.2218e-06	5.4956e-01	95	4	3.3201e+01
4.0000e+03	1.8320e-01	8.9444e-07	8.1680e-01	126	3	3.1458e+02
4.0000e+04	3.8992e-02	1.6221e-07	9.6101e-01	161	5	2.5058e+03
4.0000e+05	4.9369e-03	1.9842e-08	9.9506e-01	202	3	2.6371e+04
4.0000e+06	5.1674e-04	2.0684e-09	9.9948e-01	250	3	1.7187e+05
2.0788e+07	1.0000e-04	4.0004e-10	9.9990e-01	280	5	1.0513e+06
rootsfound[] = -1 0						
4.0000e+07	5.2009e-05	2.0805e-10	9.9995e-01	293	4	2.3655e+06
4.0000e+08	5.2012e-06	2.0805e-11	9.9999e-01	325	4	2.6808e+07
4.0000e+09	5.1850e-07	2.0740e-12	1.0000e+00	348	3	7.4305e+08
4.0000e+10	4.8641e-08	1.9456e-13	1.0000e+00	362	2	7.5480e+09
Final Run Statistics:						
Number of steps = 362						
Number of residual evaluations = 537						
Number of Jacobian evaluations = 60						
Number of nonlinear iterations = 537						
Number of error test failures = 15						
Number of nonlinear conv. failures = 0						
Number of root fn. evaluations = 404						

## 2.2 A banded example: idaFoodWeb.bnd

This example is a model of a multi-species food web [3], in which predator-prey relationships with diffusion in a 2-D spatial domain are simulated. Here we consider a model with  $s = 2p$  species:  $p$  predators and  $p$  prey. Species  $1, \dots, p$  (the prey) satisfy rate equations, while species  $p + 1, \dots, s$  (the predators) have infinitely fast reaction rates. The coupled PDEs for the species concentrations  $c^i(x, y, t)$  are:

$$\begin{cases} \partial c^i / \partial t = R_i(x, y, c) + d_i(c_{xx}^i + c_{yy}^i) & i = 1, 2, \dots, p \\ 0 = R_i(x, y, c) + d_i(c_{xx}^i + c_{yy}^i) & i = p + 1, \dots, s, \end{cases} \quad (2)$$

with

$$R_i(x, y, c) = c^i \left( b_i + \sum_{j=1}^s a_{ij} c^j \right).$$

Here  $c$  denotes the vector  $\{c^i\}$ . The interaction and diffusion coefficients  $(a_{ij}, b_i, d_i)$  can be functions of  $(x, y)$  in general. The choices made for this test problem are as follows:

$$a_{ij} = \begin{cases} -1 & i = j \\ -0.5 \cdot 10^{-6} & i \leq p, j > p \\ 10^4 & i > p, j \leq p \\ 0 & \text{all other } (i, j), \end{cases}$$

$$b_i = b_i(x, y) = \begin{cases} (1 + \alpha xy + \beta \sin(4\pi x) \sin(4\pi y)) & i \leq p \\ -(1 + \alpha xy + \beta \sin(4\pi x) \sin(4\pi y)) & i > p, \end{cases}$$

and

$$d_i = \begin{cases} 1 & i \leq p \\ 0.5 & i > p. \end{cases}$$

The spatial domain is the unit square  $0 \leq x, y \leq 1$ , and the time interval is  $0 \leq t \leq 1$ . The boundary conditions are of homogeneous Neumann type (zero normal derivatives) everywhere. The coefficients are such that a unique stable equilibrium is guaranteed to exist when  $\alpha = \beta = 0$  [3]. Empirically, a stable equilibrium appears to exist for (2) when  $\alpha$  and  $\beta$  are positive, although it may not be unique. In this problem we take  $\alpha = 50$  and  $\beta = 1000$ . For the initial conditions, we set each prey concentration to a simple polynomial profile satisfying the boundary conditions, while the predator concentrations are all set to a flat value:

$$c^i(x, y, 0) = \begin{cases} 10 + i[16x(1-x)y(1-y)]^2 & i \leq p, \\ 10^5 & i > p. \end{cases}$$

We discretize this PDE system (2) (plus boundary conditions) with central differencing on an  $L \times L$  mesh, so as to obtain a DAE system of size  $N = sL^2$ . The dependent variable vector  $C$  consists of the values  $c^i(x_j, y_k, t)$  grouped first by species index  $i$ , then by  $x$ , and lastly by  $y$ . At each spatial mesh point, the system has a block of  $p$  ODE's followed by a block of  $p$  algebraic equations, all coupled. For this example, we take  $p = 1, s = 2$ , and  $L = 20$ . The Jacobian is banded, with half-bandwidths  $\text{mu} = \text{ml} = sL = 40$ .

The `idaFoodWeb_bnd.c` program includes the files `sunmatrix_band.h` and `sunlinsol_band.h` in order to use the `SUNLINSOL_BAND` linear solver. The former of these files contains the definition for the band matrix type `SUNMATRIX_BAND`, and the `SM_COLUMN_B` and `SM_COLUMN_ELEMENT_B` macros for accessing matrix elements. See §8.4. The main IDA header file `ida.h` is included for the prototypes of the solver user-callable functions and IDA constants, while the file `nvector_serial.h` is included for the definition of the serial `N_Vector` type. The header file `sundials_dense.h` is included for the `newDenseMat` function used in allocating memory for the user data structure.

The include lines at the top of the file are followed by definitions of problem constants which include the  $x$  and  $y$  mesh dimensions, `MX` and `MY`, the number of equations `NEQ`, the scalar relative and absolute tolerances `RTOL` and `ATOL`, and various parameters for the food-web problem.

Spatial discretization of the PDE naturally produces a DAE system in which equations are numbered by mesh coordinates  $(i, j)$ . The user-defined macro `IJth_Vptr` isolates the translation for the mathematical two-dimensional index to the one-dimensional `N_Vector` index and allows the user to write clean, readable code to access components of the dependent variable. `IJ_Vptr(v, i, j)` returns a pointer to the location in `v` corresponding to the species with index `is = 0`, x-index `ix = i`, and y-index `jy = j`.

The type `UserData` is a pointer to a structure containing problem data used in the `resweb` function. This structure is allocated and initialized at the beginning of `main`. The pointer to it, called `webdata`, is then passed to `IDASetUserData` and as a result it will be passed back to the `resweb` function each time it is called.

The `main` program is straightforward and very similar to that for `idaRoberts_dns`. The differences come from the use of the `SUNLINSOL_BAND` linear solver and from the use of the consistent initial conditions algorithm in IDA to correct the initial values. The call to `SUNBandMatrix` includes the half-bandwidths `m1` and `mu`. `IDACalcIC` is called with the option `IDA_YA_YDP_INIT`, meaning that IDA is to compute the algebraic components of  $y$  and differential components of  $y'$ , given the differential components of  $y$ . This option requires that the `N_Vector` `id` be set through a call to `IDASetId` specifying the differential and algebraic components. In this example, `id` has components equal to 1 for the prey (indicating differential variables) and 0 for the predators (algebraic variables).

Next, the `IDASolve` function is called in a loop over the output times, and the solution for the species concentrations at the bottom-left and top-right corners is printed, along with the cumulative number of time steps, current method order, and current step size.

Finally, the `main` program calls `PrintFinalStats` to get and print all of the relevant statistical quantities. It then calls `N_VDestroy_Serial` to free the vectors `cc`, `cp`, and `id`, and `IDAFree` to free the IDA memory block.

The function `PrintFinalStats` used here is actually suitable for general use in applications of IDA to any problem with a banded Jacobian. It calls various `IDAGet***` functions to obtain the relevant counters, and then prints them. Specifically, these are: the cumulative number of steps (`nst`), the number of residual evaluations (`nre`) (excluding those for difference-quotient Jacobian evaluations), the number of residual evaluations for Jacobian evaluations (`nreLS`), the number of Jacobian evaluations (`nje`), the number of nonlinear (Newton) iterations (`nni`), the number of local error test failures (`netf`), and the number of nonlinear convergence failures (`ncfn`). These optional outputs are described in §4.5.10.

The function `resweb` is a direct translation of the residual of (2). It first calls the private function `Fweb` to initialize the residual vector with the right-hand side of (2) and then it loops over all grid points, setting residual values appropriately for differential or algebraic components. The calculation of the interaction terms  $R_i$  is done in the function `WebRates`.

Sample output from `idaFoodWeb_bnd` follows.

```

----- idaFoodWeb_bnd sample output -----

idaFoodWeb_bnd: Predator-prey DAE serial example problem for IDA

Number of species ns: 2      Mesh dimensions: 20 x 20      System size: 800
Tolerance parameters:  rtol = 1e-05   atol = 1e-05
Linear solver: BAND,   Band parameters mu = 40, m1 = 40
CalcIC called to correct initial predator concentrations.

-----
t          bottom-left  top-right  | nst  k      h

```

-----					
0.00e+00	1.0000e+01	1.0000e+01		0 0	1.6310e-08
	1.0000e+05	1.0000e+05			
1.00e-03	1.0318e+01	1.0827e+01		32 4	1.0823e-04
	1.0319e+05	1.0822e+05			
1.00e-02	1.6188e+02	1.9735e+02		127 4	1.4203e-04
	1.6189e+06	1.9734e+06			
1.00e-01	2.4019e+02	2.7072e+02		235 1	3.9160e-02
	2.4019e+06	2.7072e+06			
4.00e-01	2.4019e+02	2.7072e+02		238 1	3.1328e-01
	2.4019e+06	2.7072e+06			
7.00e-01	2.4019e+02	2.7072e+02		239 1	6.2655e-01
	2.4019e+06	2.7072e+06			
1.00e+00	2.4019e+02	2.7072e+02		239 1	6.2655e-01
	2.4019e+06	2.7072e+06			
-----					
Final run statistics:					
Number of steps				= 239	
Number of residual evaluations				= 3339	
Number of Jacobian evaluations				= 36	
Number of nonlinear iterations				= 421	
Number of error test failures				= 3	
Number of nonlinear conv. failures				= 0	

### 2.3 A Krylov example: idaHeat2D\_kry

This example solves a discretized 2D heat PDE problem. The DAE system arises from the Dirichlet boundary condition  $u = 0$ , along with the differential equations arising from the discretization of the interior of the region.

The domain is the unit square  $\Omega = \{0 \leq x, y \leq 1\}$  and the equations solved are:

$$\begin{cases} \partial u / \partial t = u_{xx} + u_{yy} & (x, y) \in \Omega \\ u = 0 & (x, y) \in \partial\Omega. \end{cases} \quad (3)$$

The time interval is  $0 \leq t \leq 10.24$ , and the initial conditions are  $u = 16x(1-x)y(1-y)$ .

We discretize the PDE system (3) (plus boundary conditions) with central differencing on a  $10 \times 10$  mesh, so as to obtain a DAE system of size  $N = 100$ . The dependent variable vector  $u$  consists of the values  $u(x_j, y_k, t)$  grouped first by  $x$ , and then by  $y$ . Each discrete boundary condition becomes an algebraic equation within the DAE system.

In this case, `sunlinsol_spgmr.h` is included for the definitions of constants and function prototypes associated with the SUNLINSOL\_SPGMR linear solver module.

After various initializations (including a vector of constraints with all components set to 1, imposing all solution components to be non-negative), the main program creates and initializes the IDA memory block. It then creates the SUNLINSOL\_SPGMR linear solver using



the default `MODIFIED_GS` Gram-Schmidt orthogonalization algorithm, and updates the number of allowed SPGMR restarts to 5. It then attaches this linear solver module to IDA with a call to `IDASetLinearSolver`.

The user-supplied preconditioner setup and solve functions, `PsetupHeat` and `PsolveHeat`, and the pointer to user data (`data`) are specified in a call to `IDASetPreconditioner`. In a loop over the desired output times, `IDASolve` is called in `IDA_NORMAL` mode and the maximum solution norm is printed. Following this, three more counters are printed.

The main program then re-initializes the IDA solver and the `SUNLINSOL_SPGMR` linear solver and solves the problem again, this time using the `CLASSICAL_GS` Gram-Schmidt orthogonalization algorithm. Finally, memory for the IDA solver and for the various vectors used is deallocated.

The user-supplied residual function `resHeat`, of type `IDAResFn`, loads the DAE residual with the value of  $u$  on the boundary (representing the algebraic equations expressing the boundary conditions of (3)) and with the spatial discretization of the PDE (using central differences) in the rest of the domain.

The user-supplied functions `PsetupHeat` and `PsolveHeat` together define the left preconditioner matrix  $P$  approximating the system Jacobian matrix  $J = \partial F / \partial u + \alpha \partial F / \partial u'$  (where the DAE system is  $F(t, u, u') = 0$ ), and solve the linear systems  $Pz = r$ . Preconditioning is done in this case by keeping only the diagonal elements of the  $J$  matrix above, storing them as inverses in a vector `pp`, when computed in `PsetupHeat`, for subsequent use in `PsolveHeat`. In this instance, only `cj =  $\alpha$`  and `data` (the user data structure) are used from the `PsetupHeat` argument list.

Sample output from `idaHeat2D_kry` follows.

idaHeat2D_kry sample output										
idaHeat2D_kry: Heat equation, serial example problem for IDA										
Discretized heat equation on 2D unit square.										
Zero boundary conditions, polynomial initial conditions.										
Mesh dimensions: 10 x 10                      Total system size: 100										
Tolerance parameters:    rtol = 0    atol = 0.001										
Constraints set to force all solution components >= 0.										
Linear solver: SPGMR, preconditioner using diagonal elements.										
Case 1: gsytpe = MODIFIED_GS										
Output Summary (umax = max-norm of solution)										
time	umax	k	nst	nni	nje	nre	nreLS	h	npe	nps
0.01	8.24106e-01	2	12	14	7	14	7	2.56e-03	8	21
0.02	6.88134e-01	3	15	18	12	18	12	5.12e-03	8	30
0.04	4.70711e-01	3	18	24	21	24	21	6.58e-03	9	45
0.08	2.16509e-01	3	22	29	30	29	30	1.32e-02	9	59
0.16	4.57687e-02	4	28	36	44	36	44	1.32e-02	9	80
0.32	2.09938e-03	4	35	44	67	44	67	2.63e-02	10	111
0.64	5.54028e-21	1	39	51	77	51	77	1.05e-01	12	128
1.28	3.85107e-20	1	41	53	77	53	77	4.21e-01	14	130
2.56	5.00523e-20	1	43	55	77	55	77	1.69e+00	16	132
5.12	1.58940e-19	1	44	56	77	56	77	3.37e+00	17	133
10.24	5.12685e-19	1	45	57	77	57	77	6.74e+00	18	134



```
Error test failures           = 1
Nonlinear convergence failures = 0
Linear convergence failures   = 0
```

Case 2: gstype = CLASSICAL\_GS

Output Summary (umax = max-norm of solution)

time	umax	k	nst	nni	nje	nre	nreLS	h	npe	nps
0.01	8.24106e-01	2	12	14	7	14	7	2.56e-03	8	21
0.02	6.88134e-01	3	15	18	12	18	12	5.12e-03	8	30
0.04	4.70711e-01	3	18	24	21	24	21	6.58e-03	9	45
0.08	2.16509e-01	3	22	29	30	29	30	1.32e-02	9	59
0.16	4.57687e-02	4	28	36	44	36	44	1.32e-02	9	80
0.32	2.09938e-03	4	35	44	67	44	67	2.63e-02	10	111
0.64	2.15648e-20	1	39	51	77	51	77	1.05e-01	12	128
1.28	5.77661e-20	1	41	53	77	53	77	4.21e-01	14	130
2.56	7.50780e-20	1	43	55	77	55	77	1.69e+00	16	132
5.12	2.26547e-19	1	44	56	77	56	77	3.37e+00	17	133
10.24	6.95460e-19	1	45	57	77	57	77	6.74e+00	18	134

```
Error test failures           = 1
Nonlinear convergence failures = 0
Linear convergence failures   = 0
```

## 2.4 A sparse direct example: idaHeat2D\_klu

We provide an example of using IDA with the KLU sparse direct solver module SUNLINSOL\_KLU that solves the same 2D heat PDE problem as `idaHeat2D_kry` with the same zero Dirichlet boundary conditions and central differencing but with no preconditioner. This example is mainly based off of the `idaHeat2D.bnd` example program.

Due to the nature of the Jacobian matrix of the 2D heat PDE problem in column major format, in order to store the Jacobian in compressed sparse column (CSC) format, it was necessary to have two separate user-supplied Jacobian functions. The function `jacHeat3` sets up the Jacobian in the special case that MGRID, the number of node points used in the central difference method, is 3. For  $\text{MGRID} \geq 4$ , we use the function `jacHeat`.

The main program is written in the same way it was written in `idaHeat2D_kry` and `idaHeat2D.bnd` but with a few exceptions. In order to use the SUNLINSOL\_KLU solver and associated SUNMATRIX\_SPARSE matrix type, the user must determine the number of non-zero (`nnz`) variables and there is a conditional statement to check the size of MGRID in order to determine which `jacHeat` function to use.

The user-supplied function `jacHeat3` specifies the values of the Jacobian matrix for the  $\text{MGRID}=3$  case for each of the three datatypes needed for CSC format: column pointers (`colptrs`), actual data values (`data`), and row value of the data stored (`rowvals`).

The user-supplied function `jacHeat` defines the structure of the Jacobian matrix for a general MGRID size greater than or equal to 4 in CSC format and fills in the three datatypes as needed. The system Jacobian matrix is defined as  $J = \partial F / \partial u + \alpha \partial F / \partial u'$  with `cj` =  $\alpha$  as before. The column-based structure, which was determined heuristically, was generalized for any size in the allowable range and to allow for the appropriate number of repeats within the

structure of the Jacobian matrix. The structure's pattern was found by splitting the matrix into MGRID blocks and determining the pattern within each block separately for each of the datatypes.

The IDA package also includes support for SUPERLU\_MT, the parallel sparse direct solver. The `idaHeat2D_sps` example has been included to demonstrate SUPERLU\_MT. It is very similar to `idaHeat2D_klu`.

Sample output from `idaHeat2D_klu` follows.

```

idaHeat2D_klu sample output

idaHeat2D_klu: Heat equation, serial example problem for IDA
  Discretized heat equation on 2D unit square.
  Zero boundary conditions, polynomial initial conditions.
  Mesh dimensions: 10 x 10      Total system size: 100

Tolerance parameters:  rtol = 0    atol = 1e-08
Constraints set to force all solution components >= 0.
Linear solver: KLU, sparse direct solver
  difference quotient Jacobian
IDACalcIC called with input boundary values = 0

  Output Summary (umax = max-norm of solution)

time      umax      k  nst  nni  nje  nre      h
. . . . .
0.00      9.75461e-01  0    0    0    2    2    5.15e-10
0.01      8.24056e-01  5   53   63   23   65    5.55e-04
0.02      6.88097e-01  5   69   81   24   83    9.99e-04
0.04      4.70961e-01  5   90  106   27  108    1.91e-03
0.08      2.16312e-01  5  113  130   27  132    1.72e-03
0.16      4.53210e-02  5  137  155   28  157    3.43e-03
0.32      1.98864e-03  5  173  193   29  195    6.18e-03
0.64      3.83238e-06  5  210  233   31  235    2.22e-02
1.28      0.00000e+00  1  227  255   34  257    1.78e-01
2.56      0.00000e+00  1  230  258   37  260    1.42e+00
5.12      0.00000e+00  1  231  259   38  261    2.85e+00
10.24     0.00000e+00  1  232  260   39  262    5.69e+00

netf = 2,    ncfn = 0

```

## 3 Parallel example problems

### 3.1 A user preconditioner example: `idaHeat2D_kry_p`

As an example of using IDA with the parallel MPI `NVECTOR_PARALLEL` module and the Krylov linear solver `SUNLINSOL_SPGMR` with user-defined preconditioner, we provide the example `idaHeat2D_kry_p` which solves the same 2-D heat PDE problem as `idaHeat2D_kry`.

In the parallel setting, we can think of the processors as being laid out in a grid of size  $NPEX \times NPEY$ , with each processor computing a subset of the solution vector on a submesh of size  $MXSUB \times MYSUB$ . As a consequence, the computation of the residual vector requires that each processor exchange boundary information (namely the components at all interior subgrid boundaries) with its neighboring processors. The message-passing (implemented in the function `rescomm`) uses blocking sends, non-blocking receives, and receive-waiting, in routines `BSend`, `BRecvPost`, and `BRecvWait`, respectively. The data received from each neighboring processor is then loaded into a work array, `uext`, which contains this ghost cell data along with the local portion of the solution.

The local portion of the residual vector is then computed in the routine `reslocal`, which assumes that all inter-processor communication of data needed to calculate `rr` has already been done. Components at interior subgrid boundaries are assumed to be in the work array `uext`. The local portion of the solution vector `uu` is first copied into `uext`. The diffusion terms are evaluated in terms of the `uext` array, and the residuals are formed. The zero Dirichlet boundary conditions are handled here by including the boundary components in the residual, giving algebraic equations for the discrete boundary conditions.

The preconditioner (implemented in `PsetupHeat` and `PsolveHeat`) uses the diagonal elements of the Jacobian only and therefore involves only local calculations.

The `idaHeat2D_kry_p` main program begins with MPI calls to initialize MPI and to set multi-processor environment parameters `npes` (number of processes) and `thispe` (local process index). Then the local and global vector lengths are set, the user-data structure `Userdata` is created and initialized, and `N_Vector` variables are created and initialized for the initial conditions (`uu` and `up`), for constraints, for the vector `id` specifying the differential and algebraic components of the solution vector, and for the preconditioner (`pp`). As in `idaHeat2D_kry`, constraints are passed to IDA through the `N_Vector` `constraints` and the function `IDASetConstraints`, with all components set to 1.0 to impose non-negativity on each solution component. A temporary `N_Vector` `res` is also created here, for use only in `SetInitialProfiles`. In addition, for illustration purposes, `idaHeat2D_kry_p` also excludes the algebraic components of the solution (specified through the `N_Vector` `id`) from the error test by calling `IDASetSuppressAlg` with a flag `SUNTRUE`.

Sample output from `idaHeat2D_kry_p` follows.

```
idaHeat2D_kry_p sample output

idaHeat2D_kry_p: Heat equation, parallel example problem for IDA
                  Discretized heat equation on 2D unit square.
                  Zero boundary conditions, polynomial initial conditions.
                  Mesh dimensions: 10 x 10          Total system size: 100

Subgrid dimensions: 5 x 5          Processor array: 2 x 2
Tolerance parameters:  rtol = 0    atol = 0.001
Constraints set to force all solution components >= 0.
SUPPRESSALG = SUNTRUE to suppress local error testing on all boundary components.
Linear solver: SUNLinSol_SPGMR  Preconditioner: diagonal elements only.
```

Output Summary (umax = max-norm of solution)										
time	umax	k	nst	nni	nli	nre	nreLS	h	npe	nps
0.00	9.75461e-01	0	0	0	0	0	0	0.00e+00	0	0
0.01	8.24106e-01	2	12	14	7	14	7	2.56e-03	8	21
0.02	6.88134e-01	3	15	18	12	18	12	5.12e-03	8	30
0.04	4.70711e-01	3	18	24	21	24	21	6.58e-03	9	45
0.08	2.16509e-01	3	22	29	30	29	30	1.32e-02	9	59
0.16	4.57687e-02	4	28	36	44	36	44	1.32e-02	9	80
0.32	2.09938e-03	4	35	44	67	44	67	2.63e-02	10	111
0.64	0.00000e+00	1	39	51	77	51	77	1.05e-01	12	128
1.28	0.00000e+00	1	41	53	77	53	77	4.21e-01	14	130
2.56	0.00000e+00	1	43	55	77	55	77	1.69e+00	16	132
5.12	0.00000e+00	1	44	56	77	56	77	3.37e+00	17	133
10.24	0.00000e+00	1	45	57	77	57	77	6.74e+00	18	134
Error test failures				= 1						
Nonlinear convergence failures				= 0						
Linear convergence failures				= 0						

### 3.2 An IDABBDPRE preconditioner example: idaFoodWeb\_kry\_bbd\_p

In this example, we solve the same food web problem as with `idaFoodWeb_bnd`, but in parallel and with the `SUNLINSOL_SPGMR` linear solver and using the `IDABBDPRE` module, which generates and uses a band-block-diagonal preconditioner.

As with `idaHeat2D_kry_p`, we use a  $NPEX \times NPEY$  processor grid, with an  $MXSUB \times MYSUB$  submesh on each processor. Again, the residual evaluation begins with the communication of ghost data (in `rescomm`), followed by computation using an extended local array, `cext`, in the `reslocal` routine. The exterior Neumann boundary conditions are explicitly handled here by copying data from the first interior mesh line to the ghost cell locations in `cext`. Then the reaction and diffusion terms are evaluated in terms of the `cext` array, and the residuals are formed.

The Jacobian block on each processor is banded, and the half-bandwidths of that block are both equal to `NUM_SPECIES * MXSUB`. This is the value supplied as `mudq` and `mldq` in the call to `IDABBDPrecInit`. But in order to reduce storage and computation costs for preconditioning, we supply the values `mukeep = mlkeep = 2` ( $= \text{NUM\_SPECIES}$ ) as the half-bandwidths of the retained band matrix blocks. This means that the Jacobian elements are computed with a difference quotient scheme using the true bandwidth of the block, but only a narrow band matrix (bandwidth 5) is kept as the preconditioner.

The function `reslocal` is also passed to the `IDABBDPRE` preconditioner as the `Gres` argument, while a `NULL` pointer is passed for the `Gcomm` argument (since all required communication for the evaluation of `Gres` was already done for `resweb`).

In the `idaFoodWeb_kry_bbd_p` main program, following MPI initializations and creation of user data block `webdata` and `N_Vector` variables, the initial profiles are set, the IDA memory block is created, the `SUNLINSOL_SPGMR` linear solver is created and attached to the IDA solver, and the `IDABBDPRE` preconditioner is initialized. The call to `IDACalcIC` corrects the initial values so that they are consistent with the DAE algebraic constraints.

In a loop over the desired output times, the main solver function `IDASolve` is called, and selected solution components (at the bottom-left and top-right corners of the computational

domain) are collected on processor 0 and printed to `stdout`. The main program ends by printing final solver statistics, freeing memory, and finalizing MPI.

Sample output from `idaFoodWeb_kry_bbd_p` follows.

```

idaFoodWeb_kry_bbd_p sample output

idaFoodWeb_kry_bbd_p: Predator-prey DAE parallel example problem for IDA

Number of species ns: 2      Mesh dimensions: 20 x 20      Total system size: 800
Subgrid dimensions: 10 x 10  Processor array: 2 x 2
Tolerance parameters: rtol = 1e-05  atol = 1e-05
Linear solver: SUNLinSol_SPGMR      Max. Krylov dimension maxl: 16
Preconditioner: band-block-diagonal (IDABBDPRE), with parameters
      mudq = 20,  mldq = 20,  mukeep = 2,  mlkeep = 2
CalcIC called to correct initial predator concentrations

-----
      t      bottom-left  top-right  |  nst  k      h
-----
0.00e+00    1.0000e+01    1.0000e+01  |    0  0    1.6310e-08
              1.0000e+05    1.0000e+05  |
1.00e-03    1.0318e+01    1.0827e+01  |   33  4    9.7404e-05
              1.0319e+05    1.0822e+05  |
1.00e-02    1.6189e+02    1.9735e+02  |  118  4    1.7533e-04
              1.6189e+06    1.9735e+06  |
1.00e-01    2.4019e+02    2.7072e+02  |  175  1    3.0682e-02
              2.4019e+06    2.7072e+06  |
4.00e-01    2.4019e+02    2.7072e+02  |  178  1    2.4545e-01
              2.4019e+06    2.7072e+06  |
7.00e-01    2.4019e+02    2.7072e+02  |  179  1    4.9091e-01
              2.4019e+06    2.7072e+06  |
1.00e+00    2.4019e+02    2.7072e+02  |  179  1    4.9091e-01
              2.4019e+06    2.7072e+06  |
-----

Final statistics:

Number of steps                = 179
Number of residual evaluations = 946
Number of nonlinear iterations = 222
Number of error test failures  = 0
Number of nonlinear conv. failures = 0

Number of linear iterations    = 722
Number of linear conv. failures = 0

Number of preconditioner setups = 24
Number of preconditioner solves = 946
Number of local residual evals. = 1008

```

## 4 PETSc example problems

### 4.1 A nonstiff example: idaHeat2D\_kry\_petsc

This example is the same as the one in 3.1, except it uses PETSc vector instead of SUNDIALS native parallel vector implementation. The output of the two examples is identical. In the following, we will describe only the implementation differences between the two.

Before PETSc functions can be called, the library needs to be initialized. In this example we use initialization without arguments:

```
PetscInitializeNoArguments();
```

Alternatively, a call that takes PETSc command line arguments could be used. At the end of the program, `PetscFinalize()` is called to clean up any objects that PETSc may have created automatically. We use PETSc data management library (DM) to create 2D grid and set the partitioning. In our implementation we follow Example 15 from PETSc Time Stepping component (TS) documentation [1]. We store a pointer to thus created PETSc distributed array object in user defined structure `data`.

```
ierr = DMDACreate2d(comm,
                    DM_BOUNDARY_NONE, /* NONE, PERIODIC, GHOSTED */
                    DM_BOUNDARY_NONE,
                    DMDA_STENCIL_STAR, /* STAR, BOX */
                    MX,
                    MY,
                    NPEX,
                    NPEY,
                    1, /* degrees of freedom per node */
                    1, /* stencil width */
                    NULL,
                    NULL,
                    &(data->da));
```

This call will create  $M_X \times M_Y$  grid on MPI communicator `comm` with Dirichlet boundary conditions, using 5-point star stencil. Once the distributed array is created, we create PETSc vector by calling:

```
ierr = DMCreateGlobalVector(data->da, &uvec);
```

Template vector `uu` is created as a wrapper around PETSc vector `uvec` using `N_VMake_petsc` constructor. All other vectors are created by cloning the template to ensure the same partitioning and 2D data mapping is used everywhere. One should note that the template vector does not own the underlying PETSc vector, and it is user's responsibility to delete it after the template vector is destroyed.

To use PETSc vector wrapper in user supplied functions such as `resHeat`, one needs first to extract PETSc vector with `N_VGetVector_petsc`, and then use PETSc methods to access vector elements. Providing PETSc tutorial is beyond the scope of this document, and interested reader should consult [2]. Instead, we provide a brief description of functions used in this example.

- `PetscFunctionBeginUser;`  
First executable line of user supplied PETSc function. It should precede any other PETSc call in the user supplied function.
- `DMGetLocalVector(da, &localU)`  
Allocates a local vector `localU` with space for ghost values, based on partitioning in distributed array `da`. Vector `localU` is an object equivalent to array `uext` in function `reslocal` in example in Section 4.1.
- `DMDAGetInfo(da, ..., &Mx, &My, ...)`  
Function to get information about data array `da`. In this example it is used only to get the grid size  $M_X \times M_Y$ .
- `DMGlobalToLocalBegin(da, U, INSERT_VALUES, localU)`  
Moves data (including ghosts) from the global vector `U` to the local vector `localU`.
- `DMGlobalToLocalEnd(da, U, INSERT_VALUES, localU)`  
Barrier for `DMGlobalToLocalBegin(...)`.
- `DMDAVecGetArray(da, F, &f)`  
Gets a handle to data array `f` that shares data with vector `F` and is indexed using global dimensions from distributed array object `da`. This is logically collective call.
- `DMDAVecGetArrayRead(da, U, &u)`  
Gets a handle to data array `u` that shares data with vector `U` and is indexed using global dimensions from distributed array object `da`. This is *not* a collective call. Elements of the data array `u` are accessed by indexing `u[i][j]`, where  $i \in 0, \dots, M_X$  and  $j \in 0, \dots, M_Y$  are global mesh indices.
- `DMDAGetCorners(da, &xs, &ys, NULL, &xm, &ym, NULL)`  
Gets boundaries of grid defined in distributed array object `da`. Returns the global indices of the lower left corner  $(x_s, y_s)$ , and size of the local region  $x_m \times y_m$ , excluding ghost points.
- `DMDAVecRestoreArray(da, F, &f)`  
“Restores” array `f`. This function needs to be called after reading/writing to `f` is done. Similar holds for functions `DMDAVecRestoreArrayRead` and `DMRestoreLocalVector`.
- `PetscFunctionReturn(0)`  
This function should be used instead of `return` call in user supplied PETSc functions. It is used for error handling.

Using PETSc library when dealing with a structured grid problem like this allows one to use global indices when implementing the model and thus separate the model from the parallelization scheme. Also, note that PETSc functions used here replace private functions `rescomm`, `reslocal`, `BSend`, `BRecvPost`, `BRecvWait` and `InitUserData` from the `idaHeat2D_kry_p` example in Section 3.1, and therefore simplify the implementation.

## Notes

- At this point interfaces to PETSc solvers and preconditioners are not available. They will be added in subsequent SUNDIALS releases.





## 5 Trilinos example problems

### 5.1 A nonstiff shared memory parallel example: idaHeat2D\_kry\_tpetra

This example is the same as 2.3, except it uses the Tpetra [7] vector from the Trilinos library [5]. The Tpetra vector is built on top of the Kokkos framework [4], which provides different shared memory parallelism options. The output of the two examples is identical. In the following, we will describe only the implementation differences between the two. We assume the user is familiar with the Trilinos packages Kokkos, Teuchos, and Tpetra.

Before Tpetra methods can be called, the Tpetra scope guard needs to be instantiated.

```
/* Start an MPI session */
Tpetra::ScopeGuard tpetraScope(&argc, &argv);
```

The scope guard will initialize an MPI session and create a Tpetra communicator within the current scope. The scope guard will ensure the MPI session is finalized and all related objects are destroyed upon leaving the scope. The user does not need to make any MPI calls directly. If Tpetra is built without MPI support, the scope guard will create a dummy (serial) communicator.

Once the Tpetra communicator is created, a mapping from global to local vectors needs to be created:

```
/* Create Tpetra communicator */
auto comm = Tpetra::getDefaultComm();

/* Choose zero-based (C-style) indexing. */
const sunindex_type index_base = 0;

/* Construct an MPI Map */
Teuchos::RCP<const map_type> mpiMap =
    Teuchos::rcp(new map_type(global_length, index_base, comm,
                              Tpetra::GloballyDistributed));
```

The constructor above will create a map that will evenly partition the global vectors and assign local vector lengths to each MPI rank. This example is designed to run in a shared memory environment on a single MPI rank, so the partitioning is trivial. If Trilinos is built without MPI support, the Tpetra serial communicator will be used and the MPI size will be set to one rank. If Trilinos is built with MPI support, the user has to run the example with one rank only, otherwise the example will exit with an error message. The advantage of this approach is that this example can be linked to a Trilinos library built with or without MPI support, without changing the example code. Once the communicator and map are set, a Tpetra vector is created as:

```
/* Create a Tpetra vector and return reference counting pointer to it. */
Teuchos::RCP<vector_type> rcpuu =
    Teuchos::rcp(new vector_type(mpiMap));
```

With the Tpetra vector instantiated, the template N\_Vector is created by invoking

```
uu = N_VMake_Trilinos(rcpuu);
```

All other vectors are created by cloning the template vector `uu` to ensure they are all of the same size and have the same partitioning. The rest of the main body of the example is the same as in the corresponding serial example in 2.3.

User supplied functions `resHeat` and `PsetupHeat` are implemented using Kokkos kernels. They will be executed on the default Kokkos node type. Available Kokkos node types in Trilinos 12.14 release are serial (single thread), OpenMP, Pthread, and CUDA. The default node type is selected when building the Kokkos package.

## 5.2 A nonstiff MPI+X parallel example: `idaHeat2D_kry_p_tpetra`

This example is the same as the one in 3.1, except it uses the Tpetra vector instead of the native SUNDIALS parallel vector implementation. The output of the two examples is identical. In the following, we describe only the implementation differences between the two.

The template `N_Vector` is created the same way as in 5.1. All other vectors are created by cloning the template `N_Vector`. This example is hard-wired to use 4 MPI partitions, and will return an error if it is not. Because of this, the SUNDIALS CMake system will build this example only if the Trilinos library is built with MPI support. The Tpetra vector provides different on-node (shared memory) parallelization options in addition to MPI (distributed memory) parallelism. The `N_Vector_Trilinos` will use the Kokkos default on-node parallelism, which is selected when building the Kokkos package.

This example uses Kokkos 1D views [4] as MPI buffers. The internal boundaries of the four subgrids in the example are copied to the buffers using custom built Kokkos kernels. Each buffer has its host mirror. The buffer data is passed from the host (CPU memory) to MPI functions in the same way as described in 3.1. If the buffer is in CPU memory, the buffer and its host mirror are views of the same data. If the buffer is on a GPU device, then its host mirror is a copy of the buffer data in CPU memory. Before passing the buffer to an MPI call, the host mirror is updated using `Kokkos::deep_copy`. If the buffer is on the host, the `Kokkos::deep_copy` call to update the buffer host mirror will not do anything, and therefore will not create unnecessary overhead.

### Notes



- At this point interfaces to Trilinos solvers and preconditioners are not available. They will be added in subsequent SUNDIALS releases.

## 6 Fortran example problems

The FORTRAN example problem programs supplied with the IDA package are all written in standard FORTRAN77 and use double precision arithmetic. Before running any of these examples, the user should make sure that the FORTRAN data types for real and integer variables appropriately match the C types. See §5.4 in the IDA User Document for details.

### 6.1 A serial example: fidaRoberts\_dns

The `fidaRoberts_dns` example is a FORTRAN equivalent of the `idaRoberts_dns` example.

The main program begins with declarations and initializations. It calls the routines `FNVINITS`, `FIDAMALLOC`, `FIDAROOTINIT`, `FSUNDENSEMATINIT`, `FSUNDENSELINSOLINIT`, `FIDALSINIT`, and `FIDADENSESETJAC`, to initialize the `NVECTOR_SERIAL` module, the main solver memory, the rootfinding module, the `SUNMATRIX_DENSE` module, the `SUNLINSOL_DENSE` module, attach these to IDA, and to specify user-supplied Jacobian routine, respectively. It calls `FIDASOLVE` in a loop over `TOUT` values, with printing of the solution values and performance data (current order and step count from the `IOUT` array, and current step size from the `ROUT` array). In the case of a root return, an extra line is printed with the root information from `FIDAROOTINFO`. At the end, it prints a number of performance counters, and frees memory with calls to `FIDAROOTFREE` and `FIDAFREE`.

In `fidaRoberts_dns.f`, the `FIDARESFUN` routine is a straightforward implementation of Eqns. (1). In `FIDADJAC`, the  $3 \times 3$  system Jacobian is supplied. The `FIDAROOTFN` routine defines the two root functions, which are set to determine the points at which  $y_1 = 10^{-4}$  or  $y_3 = .01$ . The final two routines are for printing a header and the final run statistics.

The following is sample output from `fidaRoberts_dns`. The performance of FIDA here is similar to that of IDA on the `idaRoberts_dns` problem, with somewhat lower cost counters owing to the larger absolute error tolerances.

fidaRoberts_dns sample output						
fidaRoberts_dns: Robertson kinetics DAE serial exampleproblem for IDA Three equation chemicalkinetics problem.						
Tolerance parameters: rtol = 0.10E-03 atol = 0.10E-05 0.10E-09 0.10E-05						
Initial conditions y0 = ( 0.10E+01 0.00E+00 0.00E+00)						
t	y1	y2	y3	nst	k	h
0.2640E+00	0.9900E+00	0.3471E-04	0.1000E-01	75	2	0.5716E-01
Above is a root, INFO() = 0 1						
0.4000E+00	0.9852E+00	0.3386E-04	0.1480E-01	77	3	0.1143E+00
0.4000E+01	0.9055E+00	0.2240E-04	0.9447E-01	91	4	0.3704E+00
0.4000E+02	0.7158E+00	0.9185E-05	0.2842E+00	127	4	0.2963E+01
0.4000E+03	0.4505E+00	0.3223E-05	0.5495E+00	177	3	0.1241E+02
0.4000E+04	0.1832E+00	0.8940E-06	0.8168E+00	228	3	0.2765E+03
0.4000E+05	0.3899E-01	0.1622E-06	0.9610E+00	278	5	0.2614E+04
0.4000E+06	0.4939E-02	0.1985E-07	0.9951E+00	324	5	0.2770E+05
0.4000E+07	0.5176E-03	0.2072E-08	0.9995E+00	355	4	0.3979E+06
0.2075E+08	0.1000E-03	0.4000E-09	0.9999E+00	374	4	0.1592E+07
Above is a root, INFO() = -1 0						
0.4000E+08	0.5191E-04	0.2076E-09	0.9999E+00	380	3	0.6366E+07
0.4000E+09	0.5882E-05	0.2353E-10	0.1000E+01	394	1	0.9167E+08
0.4000E+10	0.7054E-06	0.2822E-11	0.1000E+01	402	1	0.1467E+10
0.4000E+11	-0.7300E-06	-0.2920E-11	0.1000E+01	407	1	0.2347E+11

Final Run Statistics:

Number of steps	= 407
Number of residual evaluations	= 557
Number of Jacobian evaluations	= 65
Number of nonlinear iterations	= 557
Number of error test failures	= 6
Number of nonlinear conv. failures	= 0
Number of root function evals.	= 437

## 6.2 A parallel example: fidaHeat2D\_kry\_bbd\_p

This example, `fidaHeat2D_kry_bbd_p`, is the FORTRAN equivalent of `idaHeat2D_kry_bbd_p`. The heat equation problem is described under the `idaHeat2D_kry` example above, but here it is solved in parallel, using the `IDABBDPRE` (band-block-diagonal) preconditioner module. The decomposition of the problem onto a processor array is identical to that in the `idaHeat2D_kry_p` example above.

The problem is solved twice — once with half-bandwidths of 5 in the difference-quotient banded preconditioner blocks, and once with half-bandwidths of 1 (which results in lumping of Jacobian values). In both cases, the retained banded blocks are tridiagonal, even though the true Jacobian is not.

The main program begins with initializations, including MPI calls, a call to `FNVINITP` to initialize `NVECTOR_PARALLEL`, and a call to `SETINITPROFILE` to initialize the `UU`, `UP`, `ID`, and `CONSTR` arrays (containing the solution vector, solution derivative vector, the differential/algebraic bit vector, and the constraint specification vector, respectively). A call to `FIDASETIIN` and two calls to `FIDASETVIN` are made to suppress error control on the algebraic variables, and to supply the `ID` array and constraints array (making the computed solution non-negative). The call to `FIDAMALLOC` initializes the FIDA main memory. The calls to `FSUNSPGMRINIT`, `FSUNSPGMRSETMAXRS`, `FIDALSINIT` and `FIDABBDINIT` create and initialize the SPGMR solver and FIDABBD module.

In the first loop over `TOUT` values, the main program calls `FIDASOLVE` and prints the max-norm of the solution and selected counters. When finished, it calls `PRNTFINALSTATS` to print a few more counters.

The second solution is initialized by resetting `mudq` and `mldq`, followed by a second call to `SETINITPROFILE`, and by calls to `FIDAREINIT` and `FIDABBDREINIT`. After completing the second solution, the program frees memory and terminates MPI.

The `FIDARESFUN` routine simply calls two other routines: `FIDACOMMFN`, to communicate needed boundary data from `U` to an extension of it called `UEXT`; and `FIDAGLOCFN`, to compute the residuals in terms of `UEXT` and `UP`.

The following is a sample output from `fidaHeat2D_kry_bbd_p`, with a  $10 \times 10$  mesh and `NPES = 4` processors. The performance is similar for the two solutions. The second case requires more linear iterations, as expected, but their cost is offset by the much cheaper preconditioner evaluations.

fidaHeat2D\_kry\_bbd\_p sample output

```
fidaHeat2D_kry_bbd_p: Heat equation, parallel example for FIDA
                      Discretized heat equation on 2D unit square.
                      Zero boundary conditions, polynomial conditions.
```

```

Mesh dimensions: 10 x 10          Total system size: 100

Subgrid dimensions: 5 x 5          Processor array: 2 x 2
Tolerance parameters: rtol = 0.00E+00  atol = 0.10E-02
Constraints set to force all solution components >= 0.
SUPPRESSALG = SUNTRUE to remove boundary components from the error test.
Linear solver: SPGMR.      Preconditioner: BBDPRE - Banded-block-diagonal.

Case 1
  Difference quotient half-bandwidths = 5
  Retained matrix half-bandwidths = 1

Output Summary
  umax = max-norm of solution
  nre = nre + nreLS (total number of RES evals.)

    time          umax          k nst  nni  nli  nre  nge          h          npe  nps
-----
0.1000E-01    0.82411E+00    2  12   14   7  14+ 7   96    0.26E-02    8   21
0.2000E-01    0.68812E+00    3  15   18   12  18+12   96    0.51E-02    8   30
0.4000E-01    0.47075E+00    3  18   24   22  24+22  108    0.66E-02    9   46
0.8000E-01    0.21660E+00    3  22   29   30  29+30  108    0.13E-01    9   59
0.1600E+00    0.45659E-01    4  28   37   43  37+43  120    0.26E-01   10   80
0.3200E+00    0.21095E-02    4  35   45   59  45+59  120    0.24E-01   10  104
0.6400E+00    0.34044E-04    1  40   54   71  54+71  156    0.19E+00   13  125
0.1280E+01    0.36151E-18    1  42   56   71  56+71  180    0.76E+00   15  127
0.2560E+01    0.81974E-20    1  43   57   71  57+71  192    0.15E+01   16  128
0.5120E+01    0.17133E-19    1  44   58   71  58+71  204    0.30E+01   17  129
0.1024E+02    0.36660E-19    1  45   59   71  59+71  216    0.61E+01   18  130

Error test failures          = 1
Nonlinear convergence failures = 0
Linear convergence failures  = 0

Case 2
  Difference quotient half-bandwidths = 1
  Retained matrix half-bandwidths = 1

Output Summary
  umax = max-norm of solution
  nre = nre + nreLS (total number of RES evals.)

    time          umax          k nst  nni  nli  nre  nge          h          npe  nps
-----
0.1000E-01    0.82411E+00    2  12   14   7  14+ 7   32    0.26E-02    8   21
0.2000E-01    0.68812E+00    3  15   18   12  18+12   32    0.51E-02    8   30
0.4000E-01    0.47093E+00    3  19   23   20  23+20   36    0.10E-01    9   43
0.8000E-01    0.21655E+00    3  23   27   32  27+32   36    0.10E-01    9   59
0.1600E+00    0.45225E-01    4  27   33   44  33+44   40    0.20E-01   10   77
0.3200E+00    0.21868E-02    3  34   41   67  41+67   44    0.41E-01   11  108
0.6400E+00    0.79056E-20    1  39   49   86  49+86   52    0.16E+00   13  135
0.1280E+01    0.18819E-19    1  41   51   86  51+86   60    0.66E+00   15  137
0.2560E+01    0.20662E-18    1  42   52   86  52+86   64    0.13E+01   16  138
0.5120E+01    0.20095E-17    1  43   53   86  53+86   68    0.26E+01   17  139
0.1024E+02    0.12941E-16    1  44   54   86  54+86   72    0.52E+01   18  140

Error test failures          = 0

```

Nonlinear convergence failures = 0
Linear convergence failures = 0

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