

Calcolo Parallelo: Lezione 3

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Calcolo Parallelo 1 / 30

- Some advanced instructions
 - Derived datatypes

- Numerical integration of the Hénon–Heiles model
- (3) "They're moving in herds. They do move in herds."
- Conclusions

Calcolo Parallelo Outline 2 / 38

All the communication procedures we have seen until now are built on the assumption that

▶ all the buffers contain a sequence of identical *basic* datatypes.

This is constraining, sometimes we need more flexibility, we may want to

- pass messages that contain values with different datatypes,
- send noncontiguous data

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Such **sequence of pairs** is called then a type map:

Typemap = {(type₀, disp₀), ..., (type_{$$n-1$$}, disp _{$n-1$})},

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where type $_i$ are basic types, and disp $_i$ are displacements.

Given a type map

$$\mathsf{Typemap} = \{(\mathsf{type}_0 \; \mathsf{,} \; \mathsf{disp}_0 \; \mathsf{)}, \; \ldots, \; (\mathsf{type}_{n-1} \; \mathsf{,} \; \mathsf{disp}_{n-1} \; \mathsf{)}\},$$

then

the extent of a datatype is defined to be the span from the first byte to the last byte occupied by entries in this datatype, i.e.,

```
\mathsf{Ib}(\mathsf{Typemap}) = \min_i \mathsf{disp}_i
ub(Typemap) = max_i(disp_i + sizeof(tipe_i)) + \varepsilon,
extent(Typemap) = ub(Typemap) - lb(Typemap),
```

in which the ε is used to to satisfy alignment requirements.

Given a type map

$$\mathsf{Typemap} = \{(\mathsf{type}_0 \; , \; \mathsf{disp}_0 \;), \; \ldots, \; (\mathsf{type}_{n-1} \; , \; \mathsf{disp}_{n-1} \;)\},$$

then

- the extent of a datatype is defined to be the span from the first byte to the last byte occupied by entries in this datatype,
- ► The definition of extent is motivated by the assumption that the amount of *padding added* at the end of each structure in an array of structures is *the least needed* to fulfill alignment constraints.

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- the extent of a datatype is defined to be the span from the first byte to the last byte occupied by entries in this datatype,
- ▶ The definition of extent is motivated by the assumption that the amount of *padding added* at the end of each structure in an array of structures is *the least needed* to fulfill alignment constraints.
- We need some constructor routines to build and define the new datatypes.

Datatype constructors - MPI_TYPE_CONTIGUOUS

Allows for the replication of a datatype into contiguous locations.

int MPI_Type_contiguous(int count, MPI_Datatype oldtype,
 MPI_Datatype *newtype)

int count replication count,

MPI_Datatype oldtype old datatype,

MPI_Datatype *newtype new datatype.

- newtype is the datatype obtained by concatenating count copies of oldtype,
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- ► Example: If the old datatype has *type map* {(double, 0), (char, 8)}, and we choose count=2;, then the new datatype has *type map* {(double, 0), (char, 8), (double, 16), (char, 24)}.

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 - int MPI_Type_commit(MPI_Datatype *datatype)
- ▶ after a datatype has been committed, it can be repeatedly reused to communicate *any* buffer of that datatype.
- ► When we have ended using the new datatype we can free (deallocate) it by doing
 - int MPI_Type_free(MPI_Datatype *datatype)
 observe that, any communication that is currently using this
 datatype will complete normally, and that derived datatype
 depending on the one you are deallocating will not be touched.

Datatype constructors - An example

We want to modify the simple send and receive program from Lecture 2:

The fundamental part of the code:

```
char message[20];
if (myrank == 0){
    strcpy(message, "Hello, there");
MPI_Send(message, strlen(message)+1,
    MPI_CHAR, 1, 99, MPI_COMM_WORLD);
}
else if (myrank == 1){
MPI_Recv(message, 20, MPI_CHAR, 0,
    99, MPI_COMM_WORLD, &status);
printf("received :%s:\n", message);
}
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We want to introduce a new type for strings made of 20 char, and use it in our communications.

Note: a string is exactly a contiguous set of char.

We define the new type by doing:

```
MPI_Datatype mystring;
MPI_Type_contiguous(20,MPI_CHAR,
&mystring);
then we commit it by
MPI_Type_commit(&mystring);
and rewrite the send/receive as
 if (myrank == 0){
strcpy(message,"Hello, there");
MPI_Send(message, 1, mystring, 1, 99,
MPI COMM WORLD):
else if (myrank == 1){
MPI_Recv(message, 1, mystring, 0, 99,
MPI_COMM_WORLD, &status);
printf("received :%s:\n", message);
before finalizing we deallocate the
```

MPI_Type_free(&mystring);

type

Datatype constructors — MPI_TYPE_VECTOR

There is also a more general construct available in MPI that allows replication of a datatype into locations of equally spaced blocks.

int MPI_Type_vector(int count, int blocklength, int stride,
 MPI_Datatype oldtype, MPI_Datatype *newtype)

int count number of blocks

int blocklength number of elements in each block

int stride number of elements between start of each block

MPI_Datatype oldtype old datatype

MPI_Datatype *newtype new datatype

- ► Each block is obtained by concatenating the same number of copies of the old datatype.
- ► The spacing between blocks is a multiple of the extent of the old datatype.

We try now to use the MPI_TYPE_VECTOR by starting from a MPI_TYPE_CONTIGUOUS type.

- 1. we define an MPI_TYPE_CONTIGUOUS made up of 3 int:
 MPI_Type_contiguous(3, MPI_INT, &my3int);
- 2. we commit the new type
 MPI_Type_commit(&my3int);
- 3. Then we define (and commit) a vector type consisting in 3 blocks, of size 2 and a stride of 3 elements:

```
MPI_Type_vector(3, 2, 3, my3int, &myvector);
MPI_Type_commit(&myvector);
```

4. On rank 0 we create an array int buffer [24]; and populate it with the number $i=0,\dots,23$. Then we send it to rank 1 as an element of type myvector

```
MPI_Send(buffer, 1, myvector, 1, 666, MPI_COMM_WORLD);
```

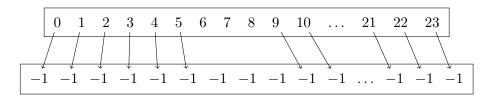
5. We receive the message on rank 1 on a buffer initialized to -1 for (i=0; i<24; i++) buffer[i] = -1; MPI_Recv(buffer, 1, myvector, 0, 666, MPI_COMM_WORLD, &status);</p>

We can depict the communication as:

0 1 2 3 4 5 6 7 8 9 10 ... 21 22 23

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We can depict the communication as:

0 1 2 3 4 5 6 7 8 9 10 ... 21 22 23

 $0 \ 1 \ 2 \ 3 \ 4 \ 5 \ -1 \ -1 \ -1 \ 9 \ 10 \ \dots \ 21 \ 22 \ 23$

Datatype constructors - MPI_TYPE_CREATE_SUBARRAY

int MPI_Type_create_subarray(int ndims,

const int array_of_sizes[],

dimension

This constructor creates a datatype describing an n-dimensional subarray of an n-dimensional array.

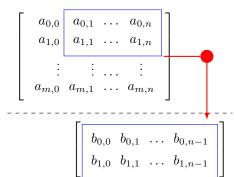
Calcolo Parallelo Some advanced instructions 11 / 3

const int array_of_starts[] starting coordinates of the subarray in each

Datatype constructors — MPI_TYPE_CREATE_SUBARRAY

This constructor creates a datatype describing an n-dimensional subarray of an n-dimensional array.

```
int MPI_Type_create_subarray(int ndims,
  const int array_of_sizes[],
  const int array_of_subsizes[],
  const int array_of_starts[],
  int order, MPI_Datatype oldtype, MPI_Datatype *newtype)
```



- The subarray may be situated anywhere within the full array,
- ► The subarray may be of any (nonzero) size up to the size of the larger array (it has to be confined within the original array!)
- ► Note that a C program may use Fortran order and a Fortran program may use C order.

As a first example we extract and communicate a subarray of a 1D array.

We start by producing a subarray for a 1D array int array [9]

```
int myrank;
MPI_Status status;
MPI_Datatype subarray;
int array[9] = { -1, 1, 2, 3, -2,
    -3, -4, -5, -6 };
int array_size[] = {9};
int array_subsize[] = {3};
int array_start[] = {1};
```

As a first example we extract and communicate a subarray of a 1D array.

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int myrank;
MPI_Status status;
MPI_Datatype subarray;
int array[9] = \{-1, 1, 2, 3, -2,
-3. -4. -5. -6 }:
int array_size[] = {9};
int array_subsize[] = {3};
int array_start[] = {1};
MPI_Init(&argc, &argv);
MPI_Type_create_subarray(1, array_size,
 array_subsize, array_start,
 MPI_ORDER_C, MPI_INT, &subarray);
MPI_Type_commit(&subarray);
```

- We start by producing a subarray for a 1D array int array[9]
- ▶ We select
 the subarray made of 3 elements
 (int array_subsize[]={3};),
 starting from the first position
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 from an array of 9 elements
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MPI_Init(&argc, &argv);
MPI_Type_create_subarray(1, array_size,
array_subsize, array_start,
MPI_ORDER_C, MPI_INT, &subarray);
MPI_Type_commit(&subarray);
MPI_Comm_rank(MPI_COMM_WORLD, &myrank);
if (mvrank == 0){
MPI_Send(array, 1, subarray, 1, 123,
 MPI_COMM_WORLD);
}
```

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 the subarray made of 3 elements
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 (int array_size[]={9});
- ► From the process with myrank = 0; we send the subarray taken from the variable array to the process with rank 1.

As a first example we extract and communicate a subarray of a 1D array.

```
else if (myrank == 1){
  for (int i=0; i<9; i++)
    array[i] = 0;
MPI_Recv(array, 1, subarray, 0, 123,
    MPI_COMM_WORLD, &status);
  for (int i=0; i<9; i++)
    printf("array[%d] = %d\n", i, array[i]);
  fflush(stdout);
}</pre>
```

► The process with myrank = 1; receives the variable and stores it into its version of the variable array.

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    printf("array[%d] = %d\n", i, array[i]);
  fflush(stdout);
}
MPI_Type_free(&subarray);
MPI_Finalize();</pre>
```

- ► The process with myrank = 1; receives the variable and stores it into its version of the variable array.
- ► In conclusion we deallocate the subarray type, and finalize the MPI environment.

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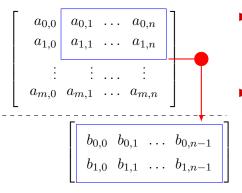
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- ► The process with myrank = 1; receives the variable and stores it into its version of the variable array.
- ► In conclusion we deallocate the subarray type, and finalize the MPI environment.
- When we run the code we print in output the array with values array[0] = 0 array[1] = 1 array[2] = 2 array[3] = 3 array[4] = 0 ...
 that are the elements in position 2 to 4 of

Calcolo Parallelo Some advanced instructions 13 / 36

 $array[9] = \{ -1, 1, 2, 3, -2, -3, -4, -5, -6 \};$

▶ We could have been more clever in the receveing phase:



- ▶ If we look back at the figure the receveing array was instead an array object of the received size,
- ► Nevertheless, if we expect to use the the new datatype subarray on the receiving end we have to use an array with global size equal to the one used to build the type
- ► We can do it in a slightly different way to match what is depicted in the figure:

```
int receivearray[3];
MPI_Recv(receivearray, 3, MPI_INT, 0, 123,
    MPI_COMM_WORLD, &status);
```

Datatype constructors - Exercise

Let us use the MPI_Type_create_subarray routine to extract a 2D subarray from a 2D array.

- 1. We want to be a little more general than we have been for the 1D array case, thus let us define first two functions:
 - ightharpoonup a function int **allocarray(int n) that allocates a $n \times n$ array;
 - ▶ a function void printarr(int **array, int n, char *str) that takes as input the address of a 2D array int **array, its size int n, and a string with the name of the array for printing everything to screen.
- Then processor 0 initializes (and populates) the bigarray, build the new data type with the MPI_Type_create_subarray routine, prints it in output and sends the data to a (smaller) array on processor 1.
- On the other side, processor 1 initializes the memory for the subarray, receives the message from process 0, prints it in output and terminates.
- ➤ You have an "skeleton code" in the subarray2Darray.c file in the code folder.

The function to dynamically allocate a 2D array can be written as:

```
int **allocarray(int n) {
int *rows = malloc(n*n*sizeof(int));
int **arr2D = malloc(n*sizeof(int *));
for (int i=0: i<n: i++)
  arr2D[i] = &(rows[i*n]);
return arr2D;
}
while the function to print a given n \times n array can be obtained as
void printarr(int **array, int n, char *str) {
 printf("-- %s --\n", str);
for (int i=0; i<n; i++) {
  for (int j=0; j<n; j++) {
   printf("%3d ", array[i][j]);
  printf("\n");
```

Calcolo Parallelo

On the processor with rank = 0; we initialize (and populate) the bigarray, build the new data type with the MPI_Type_create_subarray routine, then we print it in output and sends the data to a (smaller) array on processor 1.

```
int **bigarray = allocarray(bigsize);
for (int i=0; i < bigsize; i++)</pre>
 for (int j=0; j<bigsize; j++)</pre>
  bigarray[i][j] = i*bigsize+j;
printarr(bigarray, bigsize, " Sender: Big array ");
and then the main part:
MPI_Datatype mysubarray;
// We choose the coordinate from which we start the communication:
int starts[] = {}:
int subsizes[] = {}; // The size of the square subarray to
                       // communicate
int bigsizes[] = {}; // The size of the big square array
MPI_Type_create_subarray(); // We create the subarray type
MPI_Type_commit(); // We commit it
MPI_Send(); // We send the subarray to process 1
MPI_Type_free(); // We deallocate the type
```

Some advanced instructions

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  bigarray[i][j] = i*bigsize+j;
printarr(bigarray, bigsize, " Sender: Big array ");
and then the main part:
MPI_Datatype mysubarray;
int starts[2] = {5,3};
int subsizes[2] = {subsize, subsize};
int bigsizes[2] = {bigsize,bigsize};
MPI_Type_create_subarray(2, bigsizes, subsizes, starts,
MPI_ORDER_C, MPI_INT, &mysubarray);
MPI_Type_commit(&mysubarray);
MPI_Send(&(bigarray[0][0]),1,mysubarray,1,tag,MPI_COMM_WORLD);
MPI_Type_free(&mysubarray);
```

Processor 1 initializes the memory for the subarray, receives the message from process 0, prints it in output and terminates.

```
// Now we are rank 1: we first allocate the array for
// receiving the message from rank 0 (it has to be a
// square array of size subsize^2):
int **subarray = allocarray();
// with a double for loop we put to zero the entries of the
// receiveing array:
for (int i=0; i<subsize; i++)</pre>
 for (int j=0; j<subsize; j++)</pre>
  subarray[i][j] = 0;
// We receive the data in the subarray we have allocated
MPI_Recv();
// We print what we have received:
printarr(subarray, subsize, "Receiver: Subarray -- after receive");
```

Calcolo Parallelo Some advanced instructions 18 / 38

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for (int i=0; i<subsize; i++)</pre>
 for (int j=0; j<subsize; j++)</pre>
  subarray[i][j] = 0;
// We receive the data in the subarray we have allocated
MPI_Recv(&(subarray[0][0]), subsize*subsize, MPI_INT, 0,
 tag, MPI_COMM_WORLD, &status);
// We print what we have received:
printarr(subarray, subsize, "Receiver: Subarray -- after receive");
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Calcolo Parallelo Some advanced instructions 18 / 38

Datatype constructors – (Further) Exercise(s)

Observe that:

- we have freed the mysubarray (MPI_Type_free(&mysubarray);) right after sending a data with this type, this operation is safe since it does not interfere with instantiated communications.
- to free the dynamically allocated arrays we do it in the order: free(bigarray[0]); free(bigarray);
 - i.e., we first free the memory allocated for the rows, then we free the pointer to the whole array. We need to free the memory from inside/out, otherwise we loose the inner pointers!

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Some extensions of the previous exercise could be:

- ▶ Generalize the construction of the array to a rectangular array of size $n \times m$,
- ► Try allocating and sending array with more than 2D dimensions,

Calcolo Parallelo Some advanced instructions 19 / 3

As a concluding example we consider the problem of integrating an (a system of) ODE(s) for different values of the parameters defining it.

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Parallel integration of ODEs

There exists algorithms (and libraries) that permits to actually integrate in parallel an ODE, they exploit

- parallelism across the method, e.g., substantially rewrite the sequential nature of integration procedure: exploit independent stages of multi-stage algorithms,
- parallelism across the system, e.g., different evolution time for the ODEs: waveform relaxation.
- ▶ parallelism across the steps, e.g., we have a huge time domain that we want to divide into parallel processes: parallel in time algorithm.

As a concluding example we consider the problem of integrating an (a system of) ODE(s) for different values of the parameters defining it.

Parallel integration of ODEs

There exists algorithms (and libraries) that permits to actually integrate in parallel an ODE, they exploit

- ▶ parallelism across the method, e.g., substantially rewrite the sequential nature of integration procedure: exploit independent stages of multi–stage algorithms,
- ▶ parallelism across the system, e.g., different evolution time for the ODEs: waveform relaxation.
- ▶ parallelism across the steps, e.g., we have a huge time domain that we want to divide into parallel processes: parallel in time algorithm.

We are going to treat a much simpler usage of parallelism: having to integrate the same ODE(s) for different values of some parameters.

Let us remain on the concrete, the ODEs we want to integrate is obtained from the Hamiltonian

$$H(\mathbf{p}, \mathbf{q}) = \frac{\omega_1}{2}(q_1^2 + p_1^2) + \frac{\omega_2}{2}(q_2^2 + p_2^2) + p_1^2q_2 - \frac{1}{3}p_2^3, \quad \omega_1, \omega_2 \in \mathbb{R},$$

where the q_i denotes the position coordinates, the p_i the momentum coordinates, and from which we obtain the Hamilton's equations

$$\begin{cases} \dot{q}_1 = \frac{\partial H}{\partial p_1} = 2p_1q_2 + \omega_1 p_1, \\ \dot{q}_2 = \frac{\partial H}{\partial p_2} = \omega_2 p_2 - \frac{2p_2}{3}, \\ \dot{p}_1 = -\frac{\partial H}{\partial q_1} = -\omega_1 q_1, \\ \dot{p}_2 = -\frac{\partial H}{\partial q_2} = -p_1^2 - \omega_2 q_2. \end{cases}$$

and we are interested in integrating it for different values of the constants ω_1, ω_2 , and different values of the energy E.

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What are the operation that our (parallel) software should perform?

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The "general idea" for performing such task can be summarized as:

- 1. Processor 0 reads the list of parameters in input for each instance of the problem,
- 2. Processor 0 scatters such data to every other process in the pool,
- 3. Each processor from 0,...,size-1 executes its own integration,
- 4. Processor 0 gathers the results from all the processors,
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- ► What type of method do we need?

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- ► To decide what parameters we need, we first need to decide what method we want to use for the integration,
- ► What type of method do we need? We need conservation of energy, so we need a *symplectic* method.

Symplectic integrators

A symplectic integrator is a numerical integration scheme for Hamiltonian systems, i.e., it is used to integrate equations of the form:

$$\dot{q}_i = \frac{\partial H}{\partial p_i}, \quad \dot{p}_i = -\frac{\partial H}{\partial q_i}, \quad i = 1, \dots, n.$$

This is a method for which, under opportune constraints on its parameters, a solution on a symplectic manifold is computed.

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- ightharpoonup a symplectic integrator conserves the value of H along the computed trajectories,
- ▶ it conserves the two-from $d\mathbf{p} \wedge d\mathbf{q}$.
- we are doing it numerically thus "conserves" means up to the accuracy of method, i.e., it conserves a numerical Hamiltonian (... usually a perturbation of the original one)

For our test program we will use the leapfrog method 1 . We divide the integration interval $[0,T_{\max}]$ in n intervals, i.e., $dt=T_{\max}/n$, and consider the evaluation of $\{p_i^{(n)},q_i^{(n)}\}_{i=1}^2$, for $n=0,\ldots,n$, on the points $t_i=idt$.

The method is then obtained as:

$$\begin{cases} p_1^{(n+1/2)} = p_1^{(n-1)} + \frac{\omega_1}{2} q_1^{(n-1)} dt, \\ p_2^{(n+1/2)} = p_2^{(n-1)} + \frac{\omega_2}{2} q_2^{(n-1)} dt, \\ q_1^{(n)} = q_1^{(n-1)} - (\omega_1 p_1^{(n+1/2)} + 2 p_1^{(n+1/2)} p_2^{(n+1/2)}) dt, \\ q_2^{(n)} = q_2^{(n-1)} - (\omega_2 p_2^{(n+1/2)} + [p_1^{(n+1/2)}]^2 + [p_2^{(n+1/2)}]^2) dt, \\ p_1^{(n)} = p_1^{(n+1/2)} + \frac{\omega_1}{2} q_1^{(n)} dt, \\ p_2^{(n)} = p_2^{(n+1/2)} + \frac{\omega_2}{2} q_2^{(n)} dt, \end{cases}$$

This is a symplectic method of order 1!

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¹See, e.g., J. Laskar and P. Robutel, High order symplectic integrators for perturbed Hamiltonian systems, Celestial Mech. Dynam. Astronom. **80** (2001), no. 1, 39–62.

▶ Observe that the first and the last block are the same computation on different inputs, therefore we can perform it as:

```
void kin_flow(double p[2], double q[2],
double omega[2], double dt){
  p[0] += omega[0] * q[0] * 0.5 * dt;
  p[1] += omega[1] * q[1] * 0.5 * dt;}
```

 \blacktriangleright then the central block for updating the q_i can be coded as

```
void pot_flow(double p[2], double q[2], double omega[2],
  double dt){
  q[0] -= ( omega[0] * p[0] + 2 * p[0] * p[1] ) * dt;
  q[1] -= ( omega[1] * p[1] + pow(p[0],2) - pow(p[1],2) ) * dt;}
```

► A complete step of the integration is then given by

```
void leapfrog(double p[2], double q[2], double omega[2],
  double dt){
kin_flow(p, q, omega, dt);
pot_flow(p, q, omega, dt);
kin_flow(p, q, omega, dt);}
```

Then to integrate our Hamiltonian system we repeat the routine performing one step n times (int numberofsteps times)

```
for(int i=1; i<=numberofsteps; i++) {
  leapfrog(p, q, omega, dt);
  energy_at_time_t = compute_energy(p, q, omega);
  delta_energy[i] = energy_at_time_t - initialenergy;
}</pre>
```

- we have coded the algorithm in a way that does not store the values of $\{p_i^{(j)},q_i^{(j)}\}_{i=1,2}^{j=1,\dots,n}$, therefore we need to compute the energy at each integration step,
- ▶ the *energy* is nothing more than the value of the Hamiltonian on the current values of p_i and q_i , i.e.,

```
double compute_energy(double p[2], double q[2], double omega[2]){
int i; double ris = 0;
for(i=0; i<2; i++)
  ris += omega[i] * ( pow(p[i],2) + pow(q[i],2) ) / 2;
ris += p[1] * ( pow(p[0],2) - pow(p[1],2) / 3 );
return ris;
}</pre>
```

The last part of the sequential algorithm we need to precise regards the initial conditions and the initial energy:

in which we are assuming to know:

- ▶ the values of p0[1], and q0[1],
- the initial value of the energy initialenergy,
- ▶ to complete the information we need as input also the values omega[0], omega[1], the maximum time of integration tmax, and either the value of dt or the number of intervals n.

Let us look again at our list:

- 1. Processor 0 reads the list of parameters in input for each instance of the problem,
- 2. Processor 0 scatters such data to every other process in the pool,
- 3. Each processor from 0,...,size-1 executes its own integration,
- 4. Processor 0 gathers the results from all the processors,
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we have solved how to perform the integration at step 3, and we know what parameters in input we need, namely:

- ▶ omega[0], omega[1], initialenergy, p0[1], q0[1], tmax, dt
- we want to read the the input for all the processors on rank 0, therefore we decide to produce our input file in the following form:
 - ▶ the first line will tell us how many sets of parameters we have,
 - the following lines, in groups by seven, will give us all the various parameters in the order above, one for each line.

```
Input file:
henonheiles.inp
                   if(myrank == 0){
                    FILE *ifp; char line[300];
3
                    int numberoftests, i=0;
                    ifp = fopen("henonheiles.inp","r");
0.6180339887498948482
0.03
                    if (ifp == NULL) {
0.35
                      fprintf(stderr,
0.
                       "Can't open input file henonheiles.inp!\n");
1.
0.1
                      MPI_Finalize(); return 1;
                    }
0.6180339887498948482
                    fgets(line, 300, ifp);
0.03
                    sscanf(line,"%d",&numberoftests);
0.35
0.
                    readdata = (double *)
1.
                       malloc( sizeof(double)*7*numberoftests );
0.1
                    while(fgets(line, 300, ifp)!=NULL) {
                    readdata[i] = atof(line);
0.6180339887498948482
0.03
                    i++;
0.35
0.
                    fclose(ifp);
1.
0.01
```

At this point the array readdata contains all the input value, we need now to send to each processor in MPI_COMM_RANK for 0,...,size-1 the relative seven values:

- ▶ this function has to be called by each processor,
- ▶ after the completion of the communication processor j has the jth group of 7 inputs (omega[0], omega[1], initialenergy, p0[1], q0[1], tmax, dt),

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- ▶ after the completion of the communication processor j has the jth group of 7 inputs (omega[0], omega[1], initialenergy, p0[1], q0[1], tmax, dt),

On each processor we can populate the (local) inputs with the values:

```
double omega[2],initialenergy,p0[2],q0[2],tmax,dt;
omega[0] = input[0]; omega[1] = input[1];
initialenergy = input[2]; p0[1] = input[3]; q0[1] = input[4];
tmax = input[5]; dt = input[6];
and compute the number of steps by doing:
int numberofsteps = tmax/dt;
```

We need now to gather the array delta_energy containing the error on the energy from each processor.

▶ Note: the size of delta_energy may be different on each processor!

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- we can use the MPI_Gatherv function to accommodate this issue, but this requires knowing the length of each array!

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- ▶ Note: the size of delta_energy may be different on each processor!
- we can use the MPI_Gatherv function to accommodate this issue, but this requires knowing the length of each array!
- we can first share this information between all the processors by using an MPI_Allgather that will store this information on the array

```
int *intervals_for_process;
by calling:
intervals_for_process=(int *) malloc(sizeof(int)*size);
MPI_Allgather(&numberofsteps,1,MPI_INT,
    intervals_for_process,1,MPI_INT,MPI_COMM_WORLD);
after this call every processor will have the information regarding
```

the number of nodes computed by each processor.

We put together the ingredients needed for the MPI_Gatherv

```
if (myrank == 0){
 stride = (int *)
   malloc(sizeof(int)*size);
for(int i = 0; i < size; i++){</pre>
  grandtotal +=
    intervals_for_process[i];
 if(i==0){
  stride[i] = 0:
}else{
  stride[i] = stride[i-1]
   +intervals_for_process[i-1];
every_delta_energy = (double *)
malloc(sizeof(double)*
  (grandtotal+size));
}
```

- Observe that we have allocated the memory for the receive only on the root processor, it would have been a waste of resources allocating it everywhere!
- We will need to put the received vectors with the correct spacing in the receiving buffer (every_delta_energy), thus we need to define the opportune stride (recevbuf + stride[i]*extent[recevtype])
- ▶ all this information are needed only on the root process, so it is safe to define their value only on this rank.

We have all the pieces needed for the MPI_Gatherv:

```
MPI_Gatherv(delta_energy,numberofsteps,MPI_DOUBLE,
  every_delta_energy, intervals_for_process,stride,
  MPI_DOUBLE,0,MPI_COMM_WORLD);
```

- each process will contribute with its own local delta_energy (that is an array of numberofsteps doubles),
- everything will end up in the every_delta_energy, it is going to receive from jth processor intervals_for_process[j] doubles,
- the received values will be placed contiguously as dictated by the stride array.

We have all the pieces needed for the MPI_Gatherv:

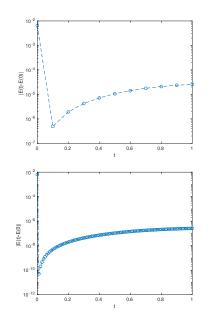
```
MPI_Gatherv(delta_energy,numberofsteps,MPI_DOUBLE,
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  MPI_DOUBLE,0,MPI_COMM_WORLD);
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- the received values will be placed contiguously as dictated by the stride array.

To conclude our code, we need only to print every result in its own file.

The writing is completely mechanical at this point:

```
if (myrank == 0){
 FILE *ofp;
 int glob_counter = 0;
 char filename[200]:
 for(int i = 0; i < size; i++){
  sprintf(filename, "energy_process_%d.dat",i);
  ofp = fopen(filename,"w+");
  for(int j=0; j < intervals_for_process[i]; j++){</pre>
   fprintf(ofp, " %le %le\n",
   j * readdata[i*7+6],
   every_delta_energy[glob_counter]);
   glob_counter++;
  fclose(ofp);
 free(readdata); free(every_delta_energy);
free(intervals_for_process);
```



As usual, this code is not perfect, and there are several possible improvement

- Written in this way we need to pay attention to the ratio between the number of processors and parameter sets,
- ▶ We could use a reduce operation (MPI_Reduce) to get (instead of all the array of the errors) a norm or some other meaningful statistics,
- We may modify it to get instead the value of the positions p_i , or the moments q_i ,
- ...

"They're moving in herds. They do move in herds."



"They're moving in herds. They do move in herds."

There exists many libraries with parallel capabilities for scientific computing.

LA Parallel Basic Linear Algebra Subprograms

- ScaLAPACK: www.netlib.org/scalapack
- ► PSBLAS: github.com/sfilippone/psblas3
- ► PETSc: www.mcs.anl.gov/petsc

Solvers Solvers for Linear and Nonlinear equations

- Hypre: computing.llnl.gov/projects/hypre-scalable-linear-solversmultigrid-methods
- ► AMG4PSBLAS: github.com/sfilippone/amg4psblas
- ► MUMPS: mumps.enseeiht.fr/
- ► Trilinos: trilinos.github.io/
- ► SUNDIALS: computing.llnl.gov/projects/sundials

Repos

- software.llnl.gov
- developer.nvidia.com/gpu-accelerated-libraries
- en.wikipedia.org/wiki/List_of_numerical_libraries

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I am a collaborator on the PSCToolkit project...so give it a try!

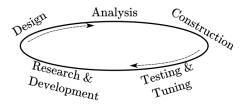
If you have either to solve large linear systems in parallel, or have programs based on BLAS, then these tools can be useful.

- ► For an introduction on how to use the libraries there are several recorded webinars: psctoolkit.github.io/talks/
- ► The website psctoolkit.github.io contains up-to-date information on software release, webinars, and new developments.

Conclusions

What do we do now?

- Find a problem you want to solve *computationally* and in *parallel*,
- ▶ Look for an MPI-enabled library implementing the basic routine you may need,
- ► Enter the never-ending cycle of scientific software development:



Things to read:

- ▶ Rouson, D., Xia, J., & Xu, X. (2011). Scientific software design: the object-oriented way. Cambridge University Press.
- ▶ Leiserson, Charles E., et al. There's plenty of room at the Top: What will drive computer performance after Moore's law?. Science, 2020, 368.6495.

Calcolo Parallelo Conclusions 38 / 38