

# 3D Cartesian Transport Sweep for Massively Parallel Architectures on top of PaRSEC

ANR SOLHAR, Toulouse

S. Moustafa, M. Faverge, L. Plagne, and P. Ramet

S. Moustafa, M. Faverge L. Plagne, and P. Ramet LabRI – Inria Bordeaux Sud-Ouest R&D – SINETICS 1 Context and goals



#### Guideline

Context and goals

Parallelization Strategies

Sweep Theoritical Model

DOMINO on top of PARSEC

Results

Conclusion and future works



#### Context

- EDF R&D is looking for a Fast Reference Solver
- PhD Student: Salli Moustafa
- Industrial solvers:
  - ▶ diffusion approximation ( $\approx$  SP1);
  - COCAGNE (SPN).
- ► Solution on more than 10<sup>11</sup> degrees of freedom (DoFs) involved
  - probabilistic solvers (very long computation time);
  - deterministic solvers.

DOMINO (SN) is designed for this validation purpose.



#### DOMINO: Discrete Ordinates Method In NeutrOnics

- Deterministic, Cartesian, and 3D solver;
- 3 levels of discretization:
  - energy (G): multigroup formalism;
  - ▶ angle  $(\vec{\Omega})$ : Level Symmetric Quadrature, N(N+2) directions
  - ▶ space (x, y, z): Diamond Differencing scheme (order 0);
- 3 nested levels of iterations:
  - power iterations + Chebychev acceleration;
  - multigroup iterations: Gauss–Seidel algorithm;
  - scattering iterations + DSA acceleration (using the SPN solver):
    - $\rightarrow$  spatial sweep, which consumes most of the computation time.



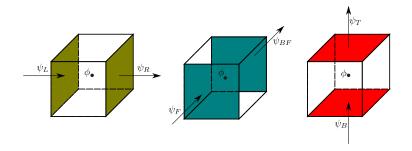
# The Sweep Algorithm

```
 \begin{aligned} & \text{forall the } o \in \textit{Octants} \text{ do} \\ & \text{forall the } c \in \textit{Cells} \text{ do} \\ & \triangleright c = (i,j,k) \\ & \text{forall the } d \in \textit{Directions[o]} \text{ do} \\ & \triangleright d = (\nu,\mu,\xi) \\ & \epsilon_x = \frac{2\nu}{\Delta x}; \quad \epsilon_y = \frac{2\eta}{\Delta y}; \quad \epsilon_z = \frac{2\xi}{\Delta z}; \\ & \psi[o][c][d] = \frac{\epsilon_x \psi_L + \epsilon_y \psi_B + \epsilon_z \psi_F + S}{\epsilon_x + \epsilon_y + \epsilon_z + \Sigma_t}; \\ & \psi_R[o][c][d] = 2\psi[o][c][d] - \psi_L[o][c][d]; \\ & \psi_T[o][c][d] = 2\psi[o][c][d] - \psi_B[o][c][d]; \\ & \psi_B[o][c][d] = 2\psi[o][c][d] - \psi_B[o][c][d]; \\ & \psi_B[b][j][i] = \phi[k][j][i] + \psi[o][c][d] * \omega[d]; \\ & \text{end} \end{aligned}
```

- 9 add or sub;
- 11 mul;
- ▶ 1 div (5 flops)
  - ightarrow 25 flops per cell, per direction, per energy group.



# The Spatial Sweep ( $Diamond\ Differencing\ scheme$ ) (1/2)



3D regular mesh with per cell, per angle, per energy group:

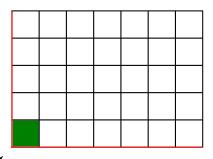
- ▶ 1 moment to update
- 3 incoming fluxes
- 3 outgoing fluxes



# The Spatial Sweep (Diamond Differencing scheme) (2/2)

2D example of the spatial mesh for one octant

At the beginning, data are known only on the incoming faces



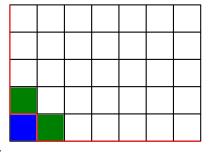
ready cell

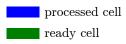




# The Spatial Sweep (Diamond Differencing scheme) (2/2)

2D example of the spatial mesh for one octant





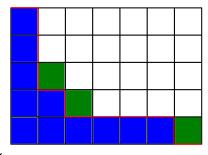




# The Spatial Sweep (Diamond Differencing scheme) (2/2)

2D example of the spatial mesh for one octant

#### ... after a few steps









# 2 Parallelization Strategies

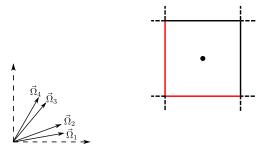


### Many opportunities for parallelism

- Each level of discretization is a potentially independent computation:
  - energy group
  - angles
  - space
- All energy groups are computed together
- All angles are considered independent
  - ightarrow This is not true when problems have boundary conditions
- ► All cell updates on a front are independent



# Angular Parallelization Level (Very Low Level)



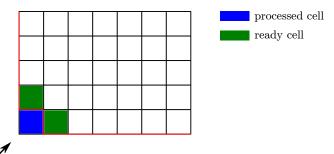
Several directions belong to the same octant:

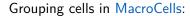
- Vectorization of the computation
- ▶ Use of SIMD units at processor/core level
  - $\rightarrow$  improve kernel performance



# Spatial Parallelization

First level: granularity



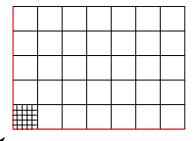


- Reduces thread scheduling overhead
- Similar to exploiting BLAS 3
- ► Reduces overall parallelism



# Spatial Parallelization

First level: granularity





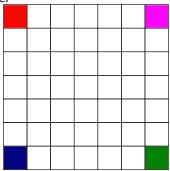
#### Grouping cells in MacroCells:

- Reduces thread scheduling overhead
- Similar to exploiting BLAS 3
- ► Reduces overall parallelism



#### Case of Vacuum Boundary Conditions

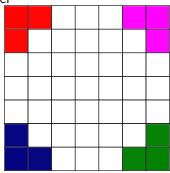
When using vacuum boundary conditions, all octants are independent from each other





#### Case of Vacuum Boundary Conditions

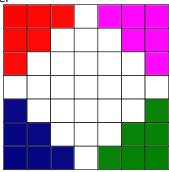
When using vacuum boundary conditions, all octants are independent from each other





#### Case of Vacuum Boundary Conditions

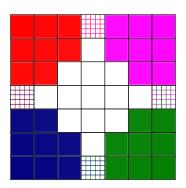
When using vacuum boundary conditions, all octants are independent from each other





#### Case of Vacuum Boundary Conditions

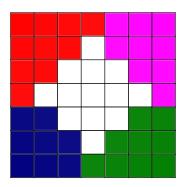
Concurrent access to a cell (or MacroCell) are protected by mutexes.





#### Case of Vacuum Boundary Conditions

Concurrent access to a cell (or MacroCell) are protected by mutexes.





3 Sweep Theoritical Model



### Basic formulas

We define the efficiency of the sweep algorithm as follow:

$$\begin{array}{lcl} \epsilon & = & \frac{T_{task} N_{tasks}}{\left(N_{tasks} + N_{idle}\right) * \left(T_{task} + T_{comm}\right)} \\ & = & \frac{1}{\left(1 + N_{idle} / N_{tasks}\right) * \left(1 + T_{comm} / T_{task}\right)} \end{array}$$

Objective: **Minimize**  $N_{idle}$ 



#### For 3D block distribution

The minimal number of idle steps are those required to reach the cube center:

$$N_{idle}^{min} = P_x + \delta_x - 2 + P_y + \delta_y - 2 + P_z + \delta_z - 2$$

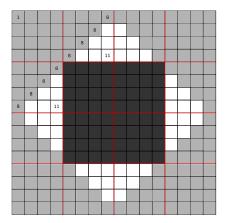
where  $\delta_u = 0$ , if  $P_u$  is even, 1 otherwise.

Objective: **Minimize the sum** P + Q + R, where  $P \times Q \times R$  is the process grid.

→ Hybrid MPI-Thread implementation allows this



# Hybrid Model







#### Implementation

- Only one kind of task:
  - Associated to one MacroCell
  - All energy group
  - All directions included in one octant
    - ightarrow 8 tasks per MacroCell
  - No dependencies from one octant to another
    - $\rightarrow$  protected by mutexes
- Simple algorithm to write in JDF
- Require a data distribution:
  - ▶ Independent from the algorithm: 2D, 3D, cyclic or not, ...
  - ▶ For now: Block-3D (Non cyclic) with a  $P \times Q \times R$  grid
- ► Fluxes on faces are dynamically allocated/freed by the runtime



# DOMINO JDF Representation (2D)

```
CellUpdate(a, b)
    /* Execution Space */
    a = 0 \dots ncx-1
    b = 0 \dots ncv-1
    /* Task Locality (Owner Compute) */
    : mcg(a, b)
    /* Data dependencies */
    RWX \leftarrow (a != aBeg) ? X CellUpdate(a-aInc, b) : X READ_X(b)
12
          —> (a != aEnd) ? X CellUpdate(a+alnc, b)
    RW Y \leftarrow (b \mid = bBeg) ? Y CellUpdate(a, b-blnc) : Y READ_Y(a)
          -> (b != bEnd) ? Y CellUpdate(a, b+blnc)
14
    RW MCG <- mcg(a, b)
16
            -> mcg(a, b)
17
    BODY
18
19
      solve ( MCG, X, Y, ... ):
20
21
    END
```

 aBeg, aEnd, aInc, bBeg, bEnd and bInc are octant dependent variables.



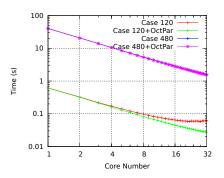
# 5 Results



# Scalability of the existing implementation with Intel TBB

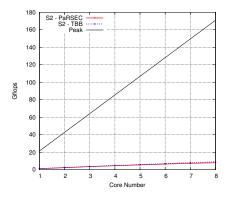
32-core Nehalem node with two 4-way SIMD units running at 2.26 Ghz

- 2 energy groups calculation;
- S8 Level Symmetric quadrature (80 angular directions);
- ▶ spatial mesh:  $120 \times 120 \times 120$  and  $480 \times 480 \times 480$ .





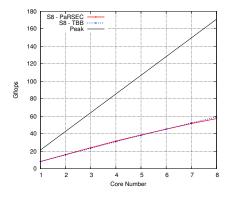
Shared Memory Results: Comparison with Intel TBB



- 1 energy group;
- ▶ mesh size: 480 × 480 × 480;
- Level Symmetric S2;
- ▶ 7.9 Gflops (4.6%)



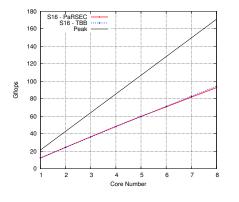
#### Shared Memory Results: Comparison with Intel TBB



- 1 energy group;
- mesh size: 480 × 480 × 480;
- Level Symmetric S8;
- ▶ 57.2 Gflops (33.5%)



#### Shared Memory Results: Comparison with Intel TBB

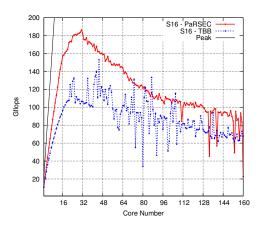


- 1 energy group;
- mesh size: 480 × 480 × 480;
- ► Level Symmetric S16;
- ▶ 92.6 Gflops (54.2%)



Shared Memory Results: Comparison with Intel TBB

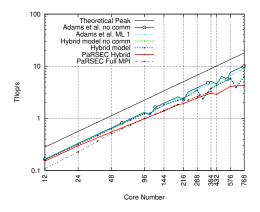
Test on manumanu NUMA node: 160 cores.





#### Distributed Memory Results (Ivanoe)

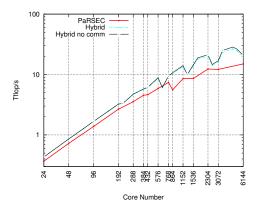
▶ 1 energy group; mesh size: 480 × 480 × 480; *Level Symmetric* S16:





#### Distributed Memory Results (Athos)

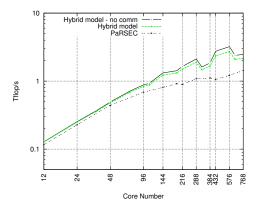
▶ 1 energy group; mesh size: 480 × 480 × 480; *Level Symmetric* S16:





Distributed Memory Results (Athos)

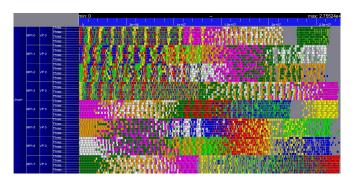
▶ 1 energy group; mesh size: 120 × 120 × 120; *Level Symmetric* S16:





Distributed Memory Results

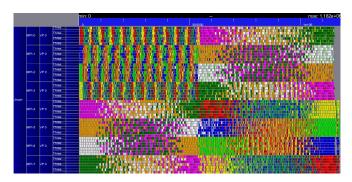
Execution trace for a run on 8 nodes (2, 2, 2) (Bad scheduling).





Distributed Memory Results

Execution trace for a run on 8 nodes (2, 2, 2) (Good scheduling).





Scheduling by front

Disco NoPrio Prio



# 6 Conclusion and future works



#### Conclusion and Future Work

#### Conclusion

- Efficient implementation on top of PaRSEC
  - Less than 2 weeks to be implemented
  - Comparable to Intel TBB in shared memory
- multi-level implementation:
  - Code vectorization (angular direction)
  - Block algorithm (MacroCells)
  - ▶ Hybrid MPI-Thread implementation

#### Future work

- ► Finish the hybrid model to get better evaluation of the performance
- Experiments on Intel Xeon Phi



# Thanks!

