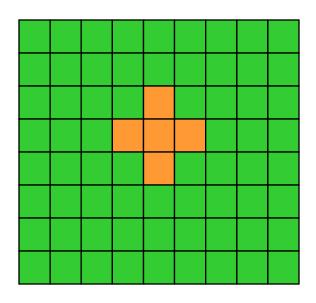
Project 2016-17

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Regular computation, iterative, stencil-based

Loop until some condition is true
Perform computation
which involves
communicating with
N,E,W,S neighbors
of a point
(5 point stencil)

[Convergence test?]



- Jacobi is an iterative method to solve a system of n linear equations in n unknowns Ax=b
 - where the i-th equations is the following:

$$a_{i,1}x_1 + a_{i,1}x_1 + \dots + a_{i,n}x_n = b_i$$

– alternatively, we can say that the i-th equations is:

$$x_i = \frac{1}{a_{i,i}} \left[b_i - \sum_{j \neq i} a_{i,j} x_j \right]$$

The various a's and b's are constants, and we have to find the unkonwns x

 The Jacobi method to solve the linear system is iterative, till we arrive at convergence:

$$x^{k}_{i} = \frac{1}{a_{i,i}} \left[b_{i} - \sum_{j \neq i} a_{i,j} x^{k-1}_{j} \right]$$

where the values of the unknowns at the (k-1)-th iteration are used to compute the values at iteration k

- The method converges for specific classes of matrixes
 - Usually the solution is approximated till a desired error threshold by computing the difference between the values of the unknowns at the (k-1)-th and k-th iterations

- The linear system Ax=b can be easily and quickly solved with Jacobi when
 A is diagonally dominant
 - i.e., the magnitude of the diagonal entry in a row is larger than or equal to the sum of the magnitudes of all the other (non-diagonal) entries in that row

$$x^{k}_{i} = \frac{1}{a_{i,i}} \left[b_{i} - \sum_{j \neq i} a_{i,j} x^{k-1}_{j} \right]$$

 Jacobi can be used to solve the differential equation of Laplace in two variables (2D):

$$\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} = 0$$

- The equation di Laplace models the steady state of a function f defined in a physical 2D space, where f is a given physical quantity
- For example, f(x,y) could represent heat as measured over a metal plate
 - Given a metal plate, for which we know the temperature at the edges, what is the temperature distribution inside the plate?

- For our purposes, it is enough to find the values of f(x,y) for a suitable 2D discretization of the space
- Let Δ be the uniform distance between the discrete grid along the two Cartesian dimensions
 - If Δ is enough small, we can approximate the 2nd order derivatives with the finite difference method. Using the Taylor series, we can thus approximate the 2nd order derivatives as follows:

$$\frac{\partial^2 f}{\partial x^2} = \frac{1}{\Delta^2} \Big[f(x + \Delta, y) - 2f(x, y) + f(x - \Delta, y) \Big]$$
$$\frac{\partial^2 f}{\partial y^2} = \frac{1}{\Delta^2} \Big[f(x, y + \Delta) - 2f(x, y) + f(x, y - \Delta) \Big]$$

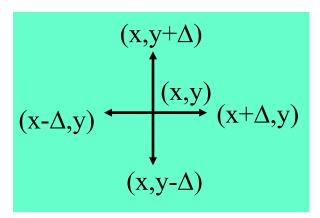
Substituting both in the Laplace equation:

$$f(x,y) = \frac{\left[f(x-\Delta,y) + f(x,y-\Delta) + f(x+\Delta,y) + f(x,y+\Delta)\right]}{4}$$

Note the relations between the values of the functions:

$$f(x,y) = \frac{\left[f(x-\Delta,y) + f(x,y-\Delta) + f(x+\Delta,y) + f(x,y+\Delta)\right]}{4}$$

We can note a cross stencil of 4 points, which is the same for each (x,y)



STENCIL: a thin sheet of cardboard, plastic, or metal with a pattern or letters cut out of it, used to produce the cut design on the surface below by the application of ink or paint through the holes.

- To solve the Laplace equation we have to find all the values of f(x,y) (steady state) wrt a discrete grid with 2 indexes
- To apply Jacobi:
 - we need to solve a linear system, by finding the values of a 1D vector x(i) of unknowns.
 - we thus convert from f(x,y) to x(i)
 - We associate the variables x(i) of our linear system Ax=b with the discrete values of f(x,y) with respect to a discrete 2D grid of n^2 points
 - Row-major distribution of the 2D discrete matrix derived from f(x,y):

We thus have n² unknowns x(i) with matrix A of size n² × n²
$$\begin{bmatrix} x_1 & x_2 & x_3 \\ x_4 & x_5 & x_6 \\ x_7 & x_8 & x_9 \end{bmatrix}$$

• Since the vector of unknowns $\underline{x(i)}$ of our linear system Ax=b are distributed row-major over the discrete 2D grid of f(x,y)

$$\begin{bmatrix} x_1 & x_2 & x_3 \\ x_4 & x_5 & x_6 \\ x_7 & x_8 & x_9 \end{bmatrix}$$

then:

$$f(x,y) = \frac{\left[f(x-\Delta,y) + f(x,y-\Delta) + f(x+\Delta,y) + f(x,y+\Delta)\right]}{4}$$

becomes:

$$x_{i} = \frac{\left[x_{i-n} + x_{i-1} + x_{i+1} + x_{i+n}\right]}{4}$$

- We have to find matrix A and vector b of the linear system, on which we have to apply Jacobi
- Our solution to the Laplace equation determined a linear system of equations of this form:

$$x_{i} = \frac{\left[x_{i-n} + x_{i-1} + x_{i+1} + x_{i+n}\right]}{4}$$

that we can re-write as follows:

$$-x_{i-n} - x_{i-1} + 4x_i - x_{i+1} - x_{i+n} = 0$$

- Then, the various b(i) are 0
- and matrix A?

- Matrix A of the linear system is sparse and is diagonally dominant
 - Each row has at most 5 non-zero values
 - All the values on the diagonal are equal to 4, that is > (or >=) than the sum of all the values on the row

$$\begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$N=9 \Rightarrow n=3$$

$$\begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$A = \begin{bmatrix} 4 & -1 & 0 & -1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 4 & -1 & 0 & -1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 4 & -1 & 0 & -1 & 0 & 0 & 0 \\ -1 & 0 & -1 & 4 & -1 & 0 & -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 0 & -1 & 4$$

Jacobi to solve the generated linear system

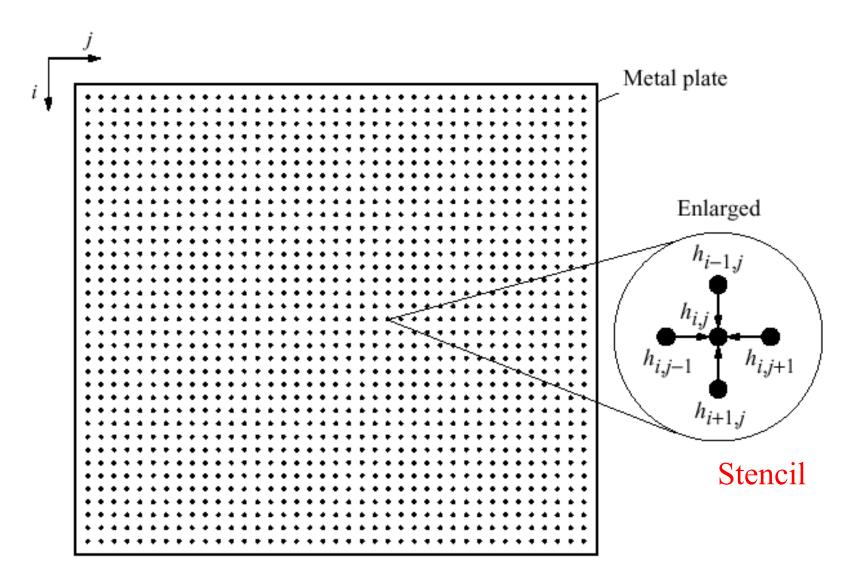
- Initialize the values of the unknowns x(i), e.g., x(i) = b(i)
- The new x(i) are computed iteratively by the Jacobi method as follows:

$$x^{t}_{i} = \frac{\left[x^{t-1}_{i-n} + x^{t-1}_{i-1} + x^{t-1}_{i+1} + x^{t-1}_{i+n}\right]}{4}$$

- The new updated values of x(i) at iteration t are used at iteration t+1
- Termination condition:

$$\left| x^{t}_{i} - x^{t-1}_{i} \right| < \text{error threshold}(\forall \mathbf{x}_{i})$$

Example: Heat diffusion with finite differences



Pseudo-codice of Message Passing Jacobi

```
[Initialize boundary regions, read-only boundary condition]
for (i=1; i<=N; i++)
     x[0][i] = north[i];
     x[N+1][i] = south[i];
     x[i][0] = west[i];
     x[i][N+1] = east[i];
[Initialize matrix]
for (i=1; i<=N; i++)
     for (j=1; j <= N; j++)
           x[i][i] = initvalue;
[Iterative refinement of x until values converge]
while (maxdiff > CONVERG)
     [Update x array]
     for (i=1; i<=N; i++)
           for (i=1; i \le N; i++)
                 newx[i][i] = \frac{1}{4}(x[i-1][j] + x[i][j+1] + x[i+1][j] + x[i][j-1]);
     [Convergence test]
     maxdiff = 0:
     for (i=1; i<=N; i++)
         for (j=1; j <= N; j++)
             maxdiff = max(maxdiff, |newx[i][i]-x[i][i]);
             x[i][i] = newx[i][i]
```

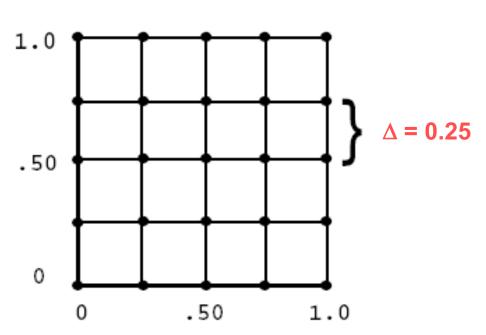
Example

- If we know the boundary conditions of f(x,y), i.e., we know the values of f(x,y) on the edges of the grid and they are constant
 - We can compute the steady state by applying Laplace, and use Jacoby to compute the values of f(x,y) for all the internal points
- Heat diffusion on a metal plate, where the edge of the plate = 1
- Border conditions:

$$- f(x, 0) = 300$$

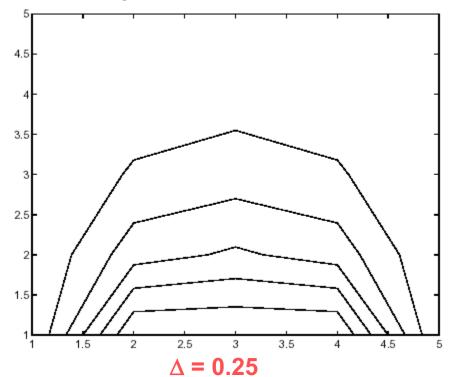
$$- f(x, 1) = f(0, y) = f(1, y) = 0$$

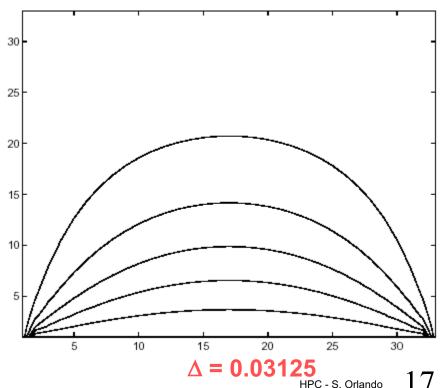
• Discretize using a uniform mesh where Δ = 0.25, we obtain the grid on the right-hand



Example

- Once Jacoby solved the Laplace equation, for the discrete points of the 2D meah, we can plot some iso-temperature curves as follow
 - Remember that the temperature at the bottom edge is kept constant at a value of 300, while is 0 on all the other three edges
- The isocurves that connect all the points at the same temperature t are smoother for denser grids, but their computation is more complex in time and space

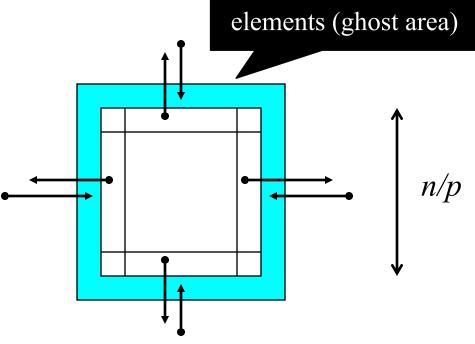




MPI parallelization: Block data distribution

- Given n² points and p² processors
 - Every processor is given n^2/p^2 points
 - The four borders, each of size n/p need to be exchanged with the neighbors
 - These four borders are calls ghost area
 - The communication volume is proportional to 4* n/p
 - The computational cost is proportional to n^2/p^2

What if we increase granularity?



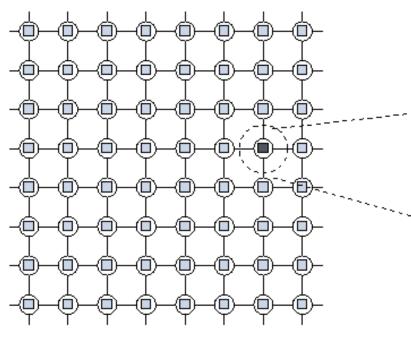
Allocate $(n/p +2)^2$

Heat diffusion: SPMD with block data distribution

- Matrix of n^2 elements distributed (block, block) on a grid of p^2 processes
- Each process $P_{i,j}$ is thus assigned a block of n^2/p^2 elements
- Code executed by a generic SPMD process P_{i,i}

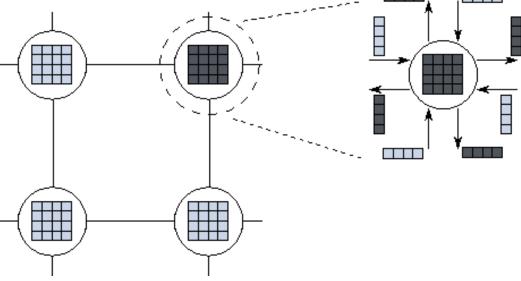
- We have to differentiate the code for blocks/processes on the grid borders
 - P_{ij} where i, j=0,p-1
 - less partners with which to exchange communications
- With asynchronous communications, we can postpone the receives
 - the receive must however be completed before accessing the ghost area

Increasing granularity



- 64 points
- 64 computations
- 4 points sent per processor
- 4 * n_procs = 4*64 =communication volume256

- 64 points
- 64 computations
- 16 points sent per processor
- 16 * n_procs = 16*4 =
 communication volume 64

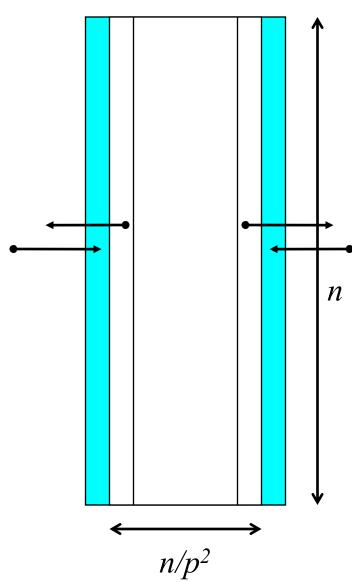


Surface / Volume

- The communication cost is proportional to the surface (perimeter) of the data partition
- The computational cost is proportional to the volume (area) of the data partition
- In two dimensions:
 - The surface increases as n
 - The volume increases as n^2
- In three dimensions:
 - The surface increase as n^2
 - The volume increases as n^3
- The ratio communication/computation decreases when increasing the partition size
 - Good!
 - This effect is less visible for high dimensional problems

MPI parallelization: Stripe data distribution

- Given n² points and p² processors
 - Every processor is given n²/p² points
 - Those points form a n * n/p² rectangle
 - Two borders, each of size n need to be exchanged with the neighbors
 - The data exchanged are proportional to n
 (2 sends and 2 receives)
- In the (block,block) distribution
 - The data exchanged are proportional to n/p
 (4 sends and 4 receives)
- To minimize the communication volume:
 - (block,block) is better

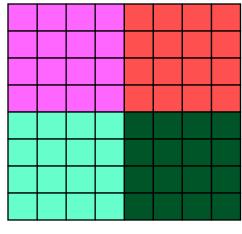


Project

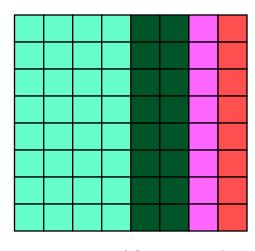
- Design Jacobi in parallel and evaluate performance plots
 - Speedup, the same problem size, ...
 - Scalability, how the execution time gets larger as we increases the size of the problem (e.g., by reducing Δ)
- MPI parallelization
- Choose some of these
 - OpenMP parallelization
 - OpenMP & MPI parallelization
 - SIMD parallelization
 - GPU parallelization

Some issues in implementing Jacobi

- Syncronization
 - Can we avoid to synchronize at each iteration?
 - Measure the effect on convergence, and if this strategy save time
- Message passing (MPI) implementation
 - Can we avoid to update the borders owned by other tasks at each iteration?
 Effect on convergence and execution time.
- Data partitioning
 - Block or strip?



Block Uniform Strip



Non-uniform Strip