

# Scalable Cloud-in-Cell Interpolation Using OpenMP

## 1 Problem Statement

You are given a set of scattered points in a 2D domain with known values:

$$S = \{(x_i, y_i, f_i) \mid i = 1, 2, \dots, N\}$$

The objective is to interpolate these values onto a structured  $Nx \times Ny$  mesh using the Cloud-in-Cell (CIC) method. For simplicity, each point has  $f_i = 1$ .

## 2 Structured Mesh Description

The target structured mesh consists of regularly spaced grid points:

$$G = \{(X_i, Y_j) \mid X_i = i\Delta x, Y_j = j\Delta y, i, j = 0, 1, \dots, M-1\}$$

where:

$$\Delta x = \frac{X_{\max}}{M}, \quad \Delta y = \frac{Y_{\max}}{M}$$

The domain is normalized such that  $0 \leq x_i \leq X_{\max}$  and  $0 \leq y_i \leq Y_{\max}$ .

## 3 Interpolation Technique

Using bilinear interpolation, each scattered point contributes to four surrounding grid points in the mesh. For a point  $(x, y)$  in a cell with corners:

- $(X_i, Y_j)$
- $(X_{i+1}, Y_j)$
- $(X_i, Y_{j+1})$
- $(X_{i+1}, Y_{j+1})$

Weights are computed as:

$$\begin{aligned} dx &= \frac{x - X_i}{\Delta x}, & dy &= \frac{y - Y_j}{\Delta y} \\ w_{i,j} &= (1 - dx)(1 - dy) \\ w_{i+1,j} &= dx(1 - dy) \\ w_{i,j+1} &= (1 - dx)dy \\ w_{i+1,j+1} &= dx \cdot dy \end{aligned}$$

Each of these weights is used to proportionally distribute the value of  $f_i$  to the corresponding grid points.

## 4 Parallelization Objectives

The primary goal of this project is to optimize the interpolation process for high-performance computing platforms by:

- Exploiting thread-level parallelism using OpenMP
- Avoiding race conditions through thread-local data buffers
- Improving spatial locality to enhance cache performance
- Achieving strong scalability across increasing thread counts

## 5 Implementation Steps

1. Read input binary file containing grid and particle information.
2. For each iteration, load a new set of scattered points.
3. Use bilinear interpolation to deposit values onto the grid using thread-local arrays.
4. Reduce per-thread contributions to obtain the final interpolated mesh.
5. Write the output mesh to a file.

## 6 Applications

- Scientific simulations: e.g., fluid dynamics, astrophysical simulations
- Computer graphics and surface reconstruction from point clouds
- Meteorological and oceanographic visualization from sensor data
- Medical imaging: reconstruction from CT or MRI scans
- Data preprocessing in machine learning for sparse sensor networks

## 7 Input Cases

For performance evaluation, five input cases were tested with varying grid sizes and number of particles. The following table summarizes the configurations:

Case	Grid Size ( $N_x \times N_y$ )	Number of Particles	Iterations (Maxiter)
A	$250 \times 100$	0.9 million	10
B	$250 \times 100$	5 million	10
C	$500 \times 200$	3.6 million	10
D	$500 \times 200$	20 million	10
E	$1000 \times 400$	14 million	10

Table 1: Summary of input cases used for interpolation performance evaluation