Monte Carlo methods for Solving PDEs Computational Physics Final Project

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Overview

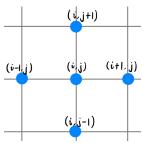
- Introduction
 - Laplace's Equation with Dirichlet Boundary Condition
 - Simple Random Walk method
 - Walk on Spheres method
 - Characteristics of Monte Carlo methods
- 2 Implementation & Analysis
 - Parallelization
 - Square Boundary
 - Circle Boundary
 - Analysis of WoS method
- Future Work

Laplace's Equation with Dirichlet Boundary Condition

 $abla^2 u = 0$ on G, u = f(x) on ∂G .

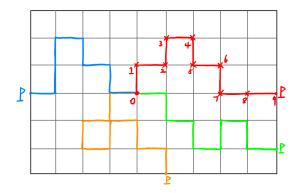
• Discrete form with centered finite difference approximation,

$$u_{i,j} = \frac{1}{4}(u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1}).$$



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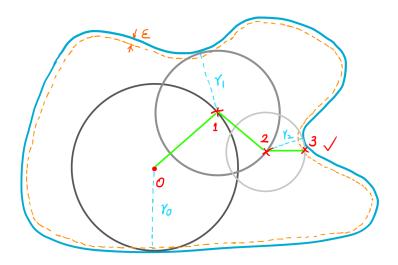
Simple Random Walk method



- For one random walk, we can get one estimate. Average over a large amount of estimates, we can get a precise value.
- The simple random walk is not very efficient, and the path doesn't really matter! We can improve it!

4 / 13

Walk on Spheres method



• Much faster than the Simple Random Walk method.

Characteristics of Monte Carlo methods

- Independence of points: The points to be evaluated are totally independent. This makes the Monte Carlo methods very suitable and efficient for evaluating values at certain points.
- Boundary & Dimension: It's easier for the Monte Carlo methods to handle complex boundaries and to be extended to higher dimensions.
- Parallel Computing: Not only different points, but also different estimates for a given point are independent. This makes the Monte Carlo methods naturally parallel.

Parallelization

- The two straightforward ways to parallelize the program are,
 - Parallelize the different estimates of one point, then average the results from different processes.
 - Parallelize the set of points to be evaluated.
- For the first way,
 - Allocate the estimates you want to get for every point to all available cores;
 - ② Initialize the random number generator of different processes with different seeds, with numpy.random.RandomState(seed).
 - Wait for all the processes to finish, then collect and average the data.
- Python package multiprocessing.

Square Boundary

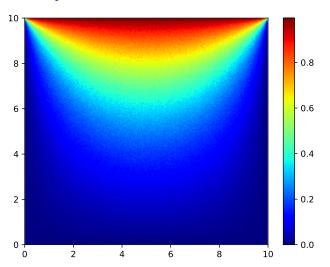


Figure: WoS method on 2D Square Boundary, 200×200 points, 2,000 estimates for each point.

Circle Boundary

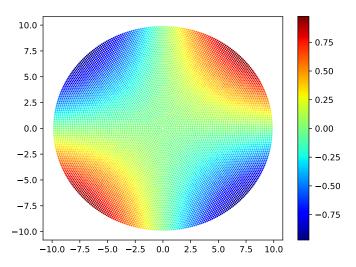


Figure: WoS method on 2D Circle Boundary, 10,620 points, 1,000 estimates for each point.

Analysis of WoS method - Running time vs. ϵ

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$$t = -k \log \epsilon + t_0$$

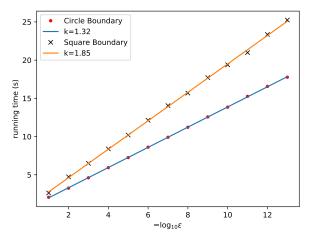


Figure: The running time - ϵ relation on both square and circle boundaries.

Analysis of WoS method - Convergence

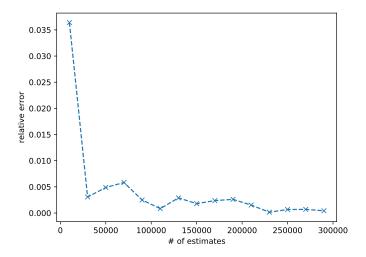


Figure: The convergence of the WoS method at the center point of the square boundary.

Future Work

• Green's Function First Passage:

$$p(x_0,x) = \frac{\partial G(x_0,x)}{\partial n}$$

If we know the Green's Function for a domain, we can directly simulate the transition from the starting point to the boundary. Then we can get all the estimates directly.

- Parallelization on GPU:
 The Monte Carlo methods here are naturally parallel, so they may be accelarated a lot with GPU computing.
- How things work for other PDEs:
 e.g. Diffusion Equation etc.

Thank you! Questions?