Ant colony algorithm for driving variance reduction techniques in Monte Carlo simulations

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Introduction

The Monte Carlo simulation is a useful tool in the study of radiation transport.

High degree of agreement with experimental measurements.

Variance reduction techniques can solve the problem.

Variance reduction techniques:

- •Russian Roulette, splitting, interaction forcing, etc.
- •Statistical weight is assigned to every particle for keeping the simulation unbiased.
- •Used properly they can increase efficiency. Otherwise, efficiency could even decrease.

How to use them properly?

Introduction

Aim of this work: To find an algorithm that permits the application of these techniques optimizing the simulation with a minimal intervention from the user.

The algorithm has been developed studying different situations regarding medical applications of ionizing radiation.

We have chosen the PENELOPE Monte Carlo code for the radiation transport simulations.

Widely used in near-surface treatments.

In the presence of heterogeneities MC is the best choice of calculation.

Essential a good characterization of the beam.

12 MeV electron beam PDD in a water phantom.

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Typical simulation parameters. PC Pentium 4 (1.6 GHz). Time to reach 2% uncertainty $(k = 3)$: 220 h.

The problem

Most of that time is spent simulating electrons that are absorbed by the jaws $($ \sim 80%).

Possible solutions for electrons:

•Russian Roulette: Reduce time but increases variance.

•Splitting: Reduce variance but increases simulation time.

The problem

Ant Colony Optimization Algorithms

Algorithms based on ant behavior:

- •Ants look for food following a random walk.
- •If they find food, then they come back to the nest depositing pheromone.
- •The other ants tend to follow the pheromone trail.
- •The overall effect is an increased deposition of pheromone on the optimal path between food and nest.

Analogies with our problem:

First step:

- •The entire geometry is divided into virtual cells.
- •The simulation starts with no use of VRTs.
- •The ratio of particles that passing through every cell, reach the ROI is registered during the simulation.
- •Importance *I* in each cell is defined as a function of that ratio (Importance map).

Second step:

•Once the importance map has enough information, VRTs can be used.

•Each time a particle arrives to a new cell, VRTs are applied according to the particle weight *w* and the cell's importance *I.*

•If *I* increases *Splitting* in *w∙I* particles with *w*' =*I* -1 .

•If *I* increases *Splitting* in *w∙I* particles with *w*' =*I* -1 . \cdot If *I* decreases \rightarrow Russian roulette with probability of survival *w∙I*. If it survives, *w*' =*I* -1 .

•If *I* increases *Splitting* in *w∙I* particles with *w*' =*I* -1 . \cdot If *I* decreases \longrightarrow Russian roulette with probability of survival *w∙I*. If it survives, *w*' =*I* -1 .

 \cdot Defining $I = 2^k$, with k as integer, all particles in the same cell have the same weight.

Importance map

$$
I=I(x, y, z, E, m)
$$

Virtual cubic cells of side 1 cm.

Two values for energy.

Two values for the material.

Importance map

When there is few information on the map, the usefulness of the algorithm is reduced.

$$
I = I(x=1, y, z, E=1, m=1)
$$

N = 100

Importance map

When there is few information on the map, the usefulness of the algorithm is reduced.

$$
I = I(x=1, y, z, E=1, m=1)
$$

N = 1,000

Importance map

When there is few information on the map, the usefulness of the algorithm is reduced.

 $I = I(x = 1, y, z, E = 1, m = 1)$ N = $10,000$

Importance map

When there is few information on the map, the usefulness of the algorithm is reduced.

 $I = I(x = 1, y, z, E = 1, m = 1)$ $N = 500,000$

Results applying the optimization algorithm

Time simulation is reduced from 220 h to 4.4 h. Efficiency x 50.

The optimization algorithm allows the efficient and automatic use of variance reduction techniques.

Tested on a particular problem.

But is this algorithm general enough?

They are very narrow beams used for treatment of small lesions near healthy structures that has to be preserved.

The characterization of these beams is very complex due to their small size.

Monte Carlo simulations can be used as a complementary tool to experimental measurements.

Again, the simulation time can be huge.

Characterization of the beams

Circular fields generated by a Varian accelerator 2100C with conical collimators.

Characterization for the treatment planner:

- •Depth dose distributions and lateral profiles in water phantom.
- •Output factors for each cone.

Characterization of the beams

•Tuning of the electron beam incident on a target for reproducing experimental measurements.

•We need to apply the optimization algorithm to both electrons and photons.

Application of the optimization algorithm

Importance maps:

 $I_{_{1}}$ = I ($x,$ $y,$ $z,$ $E,$ M) for electrons. $I_2 = I(x, y, z, E, M, \theta, \phi)$ $= I(x,y,z,E,M,\theta,\phi)$ for photons.

Using new variance reduction techniques with photons:

•Russian roulette and splitting \rightarrow No gain in efficiency. •Directional bremsstrahlung splitting.

Directional bremsstrahlung splitting

Whenever an event that produces photons occurs, the event is repeated *w⋅N*_{Br} times.

Russian roulette is applied on each generated photon according to *w*' *∙I*.

It is used throughout the geometry and when photons are scattered.

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Left: map for high energy photons pointed to the phantom. Right: High energy electrons. Brighter colors correspond to higher importance.

Simulation Time

Computer Intel Quad Core Harpertown E5405 (2.0 GHz).

Version 2008 of PENELOPE.

Uncertainty 2% $(k = 2)$:

 \cdot Cone 10 mm: 9 h. •Cone 20 mm: 3.6 h. •Cone 30 mm: 0.9 h.

Conclusions

We have developed an optimization algorithm based on ant colonies that allows the efficient implementation of variance reduction techniques in different situations.

It makes use of information registered on importance maps.

Minimum intervention by the user is required.

Other applications

In addition to the former situations, the optimization algorithm has been applied in solving other problems:

•Calculation of specific absorbed doses to organs by nuclear medicine procedures. Efficiency \times 10.

•Computation of correction factors of micro-ionization chambers. Efficiency \times 100.

Perspectives

•Application of the optimization algorithm to other problems that use the Monte Carlo simulation of radiation transport.

- •Implementation in other simulation codes.
- •To increase the degree of automation.

•Study of applications of the information stored in the importance maps.

Papers

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