Quantum Gravity Simulations with MIXMAX Generator

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Introduction

Discretization
Numerical setup
Minimal triangulation

Measurements

Observables
Phase structure

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Parallel rejection
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Introduction to Causal Dynamical Triangulations

What is Causal Dynamical Triangulations?

Causal Dynamical Triangulations (CDT) is a background independent approach to quantum gravity. It provides a lattice regularization of the formal gravitational path integral via a sum over causal triangulations.

$$\int D[g]e^{iS^{EH}[g]} \longrightarrow \sum_{\mathcal{T}} e^{-S^{R}[\mathcal{T}]}$$
continuous
discrete

MIXMAX and CDT

For the last three years we have been using the standard version of the MIXMAX pseudorandom number generator with N=256, s=-1, m=0 (and discarding the first component) in our CDT code.

It shows many advantages:

- ► Great quality (passes Big Crush from *TestU01*)
- lacktriangle Very fast (although it's a small part, < 0.5% CPU time)
- ▶ Convenient vector size $(255 = 3 \times 85)$
- Convenient output size (61-bits, double)

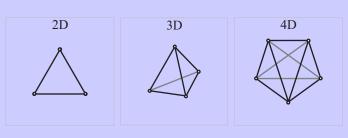
It just works!

We encountered no issues which could be related to the RNG.

Regularization via triangulation

- ▶ A four-dimensional simplicial manifold is obtained by gluing pairs of four-simplices along their three-faces (tetrahedra).
- ► The metric is **flat** inside each 4-simplex.
- Curvature is localized at triangles.
- ▶ Global proper-time foliation. Spatial states are three-dimensional geometries. Discretized states are build from equilateral tetrahedra.
- ▶ Length of time links a_t and space links a_s is constant.

Fundamental building blocks



Regge action

The Einstein-Hilbert action has a natural realization on piecewise linear geometries called Regge action

$$S^{E}[g] = -\frac{1}{G} \int \mathrm{d}t \int \mathrm{d}^{D}x \sqrt{g}(R - 2\Lambda)$$

$$N_0$$
 r

 N_4 r

 N_{14} r

 N_{14} r

 N_{14} r

 N_{14} r

 N_{14} r

 N_{15} r

 N_{16} r



Regge action

The Einstein-Hilbert action has a natural realization on piecewise linear geometries called Regge action

$$S^{R}[\mathcal{T}] = -K_0N_0 + K_4N_4 + \Delta(N_{14} - 6N_0)$$

 N_0 number of vertices

N₄ number of simplices

 N_{14} number of simplices of type $\{1,4\}$

 K_0 K_4 Δ bare coupling constants $(G, \Lambda, a_t/a_s)$



$$Z = \int \mathrm{D}[g] e^{iS^{EH}[g]}$$

- ► To make sense of the gravitational path integral one uses the standard method of regularization discretization.
- ▶ The path integral is written as a nonperturbative sum over all causal triangulations \mathcal{T} .
- Wick rotation is well defined due to global proper-time foliation. $(a_t \rightarrow i a_t)$
- Using Monte Carlo techniques we can approximate expectation values of observables.

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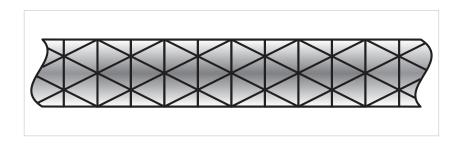
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To start the simulations we need to construct an initial triangulation with the desired topology.

Spherical topology $(S^3 \times S^1)$

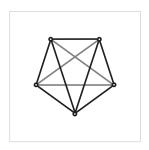
- ▶ Minimal triangulation: five tetrahedra. Toroidal topology $T^3 \times S^1$
- ▶ Regular initial triangulation: 1024 four-simplices per slice
- ► Minimal triangulation: layered and interlaced structure, 15 vertices (not 3³) and 90 tetrahedra.



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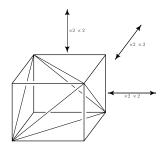
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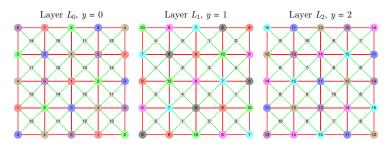


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Minimal triangulation: five tetrahedra.
 Toroidal topology T³ × S¹

▶ Regular initial triangulation: 1024 four-simplices per slice

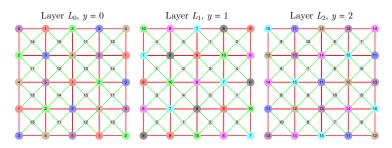
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To start the simulations we need to construct an initial triangulation with the desired topology. Spherical topology $(S^3 \times S^1)$

► Minimal triangulation: five tetrahedra. Toroidal topology $T^3 \times S^1$

- ▶ Regular initial triangulation: 1024 four-simplices per slice
- Minimal triangulation Stronger finite size effects! re, 15 vertices (not 3³) and 90 tetra



Monte Carlo simulations - Pachner moves

Random walk over configuration space:

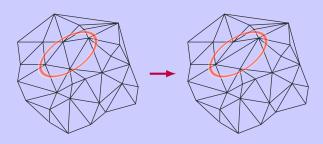
Ergodicity all possible configurations can be generated by moves

► Fixed topology moves don't change the topology

Causality moves preserve the foliation

▶ 4D CDT set of 7 moves

Example of a 2D Monte Carlo move



Monte Carlo Markov Chain

- We perform a random walk in the phase-space of configurations (space of piecewise linear geometries).
- ► Each step is one of the 4D CDT moves.
- ▶ The weight (acceptance probability) $W(A \rightarrow B)$ of a move from configuration A to B is determined (not uniquely) by the detailed balance condition:

$$P(A)W(A \rightarrow B) = P(B)W(B \rightarrow A)$$

- ▶ The Monte Carlo algorithm ensures that we probe the configurations with the probability P(A).
- ► After sufficiently long time, the configurations are independent.
- All we need, is the probability functional for configurations P(A) up to the normalization (Regge action).

Dynamical lattice

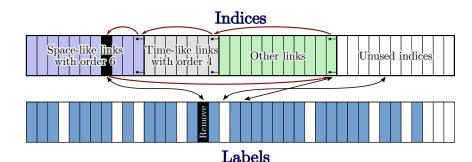
- The Monte Carlo moves alter the underlying spacetime lattice making it dynamical. The (sub)-simplices appear and disappear modifying the connectivity.
- To effectively deal with such system various techniques are used.

Basic stages of a single move attempt:

- Choose which move type (with adjusted probabilities).
- Choose where to make the move (index).
- Verify move validity (hash table) and calculate its weight (maximal information).
- ► Check whether to make the move (detailed balance). Reject the move, or accept the move.
- Update the triangulation if move was accepted.

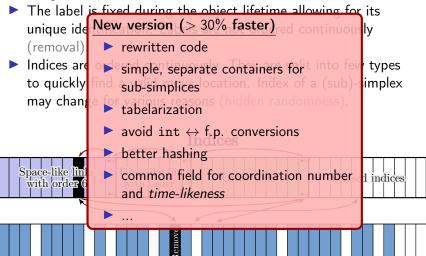
Internals of CDT code: indices and labels

- ► Each object (point, link, triangle or simplex) is assigned two integer numbers: a label and an index.
- ► The label is fixed during the object lifetime allowing for its unique identification. Labels are not ordered continuously (removal).
- ▶ Indices are ordered continuously. They are split into few types to quickly find a valid move location. Index of a (sub)-simplex may change for various reasons (hidden randomness).



Internals of CDT code: indices and labels

► Each object (point, link, triangle or simplex) is assigned two integer numbers: a label and an index.



Labels

Internals of CDT code: data structures

- Simulations are limited by CPU time consumption, while memory usage is not a bottle-neck.
- ➤ To enhance the acceptance rate of the moves and gain fast access to necessary data maximal information philosophy is adopted.
- ➤ Various information about points, links, triangles and simplices (but not tetrahedra) are stored:

```
struct triangle
{
   int p[3];  /* Vertex labels */
   int n;  /* Coordination number */
   int s;  /* Simplex which owns this triangle */
   int i;  /* Index */
   int h;  /* Hash code of triangle */
   int next;  /* Next triangle in hash table list */
}
```

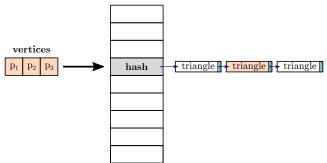
Internals of CDT code: hash tables

- ➤ To perform a fast check if a *link* or *triangle* with given vertices exists, a **hash table** is used.
- Verification of tetrahedron existence is more complicated and demands making a loop around a triangle.

Given labels of vertices p[0] < p[1] < p[2] we calculate the hash code,

$$h = (p[0] \ll 8) \text{ xor } (p[1] \ll 4) \text{ xor } p[2]$$

Triangles with the same *hash code* are stored in a singly-linked list.

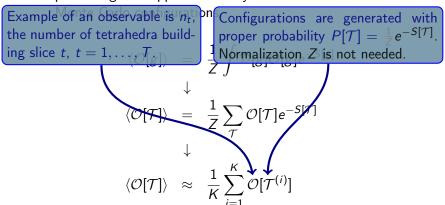


Numerical setup

- ▶ Monte Carlo algorithm probes the space of configurations with the probability $P[\mathcal{T}] = \frac{1}{7}e^{-S[\mathcal{T}]}$.
- ➤ To calculate the expectation value of an observable, the path integral is approximated by a sum over a finite set of Monte Carlo configurations.

Numerical setup

- ▶ Monte Carlo algorithm probes the space of configurations with the probability $P[\mathcal{T}] = \frac{1}{Z}e^{-S[\mathcal{T}]}$.
- ► To calculate the **expectation value of an observable**, the path integral is approximated by a sum over a finite set of



What can we measure?

From the computational point of view, the basic observables are the parameters which appear in the bare action:

 N_0 number of vertices

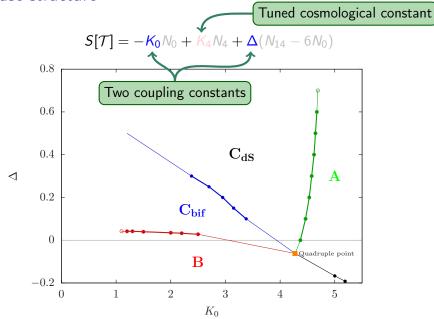
 N_{41} number of simplices of type $\{4,1\}$

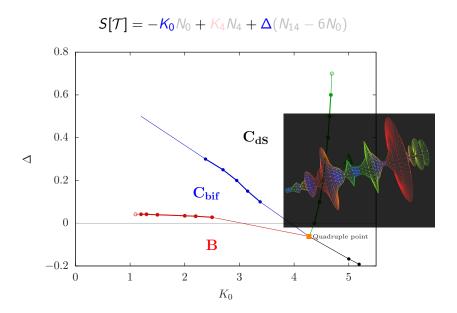
 N_{32} number of simplices of type $\{3,2\}$

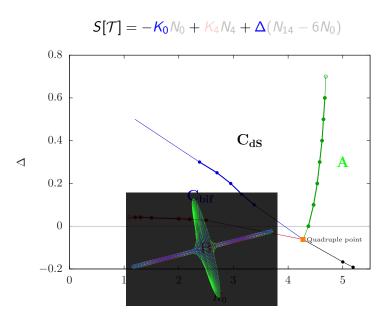
The number of all remaining sub-simplices can be expressed as a linear combination of above parameters.

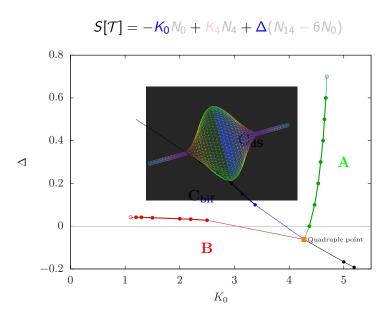
The simplest observable providing information about spacetime geometry is **spatial volume** n_t defined as a number of tetrahedra building a three-dimensional slice $t = 1 \dots T$.

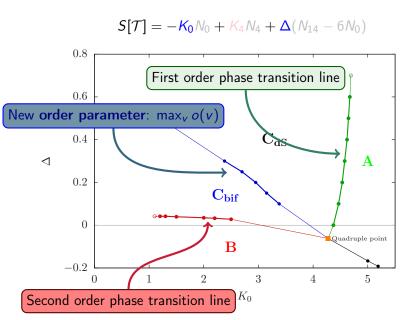
(Proper definition of e.g. curvature is non-trivial.)









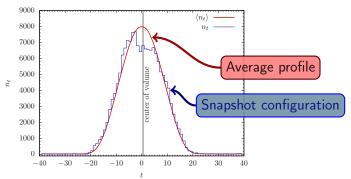


De Sitter phase - background geometry

- In phase C the time translation symmetry is spontaneously broken and the three-volume profile n_t is bell-shaped.
- ▶ The average volume $\langle n_t \rangle$ is with high accuracy given by formula

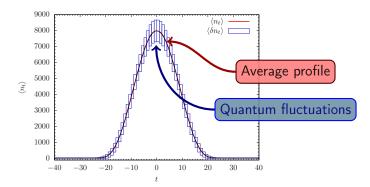
$$\langle n_t \rangle = H \cos^3 \left(\frac{t}{W} \right)$$

a classical vacuum solution.



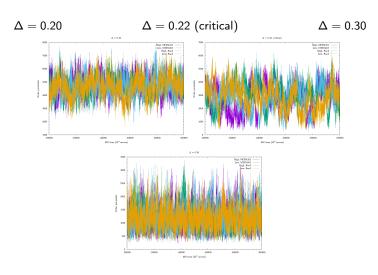
De Sitter phase - properties

- Phase C, the so called de Sitter phase, is physically most interesting.
- ▶ A four-dimensional background geometry emerges dynamically. The background geometry corresponds to Euclidean de Sitter space (S^4) , a classical vacuum solution.
- It is also possible to study quantum fluctuations around it.



Autocorrelation: MIXMAX vs Ran3

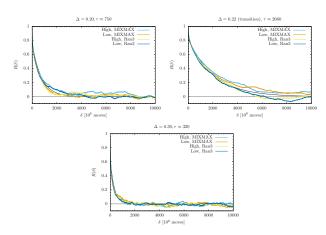
The MC moves introduce only small local changes and the configurations may be correlated for a long time.



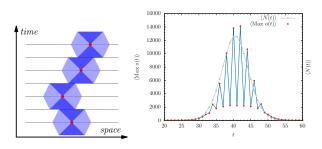
Autocorrelation time: MIXMAX vs Ran3

Often in the vicinity of phase transition, a phenomenon of critical slowing down is observed.

It takes longer for the system to thermalize, amplitude of fluctuations increases and the autocorrelation time grows.

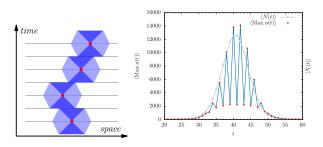


The new bifurcation phase



- ▶ The spatial volume profile $\langle n_t \rangle$ is similar as in phase C. But the agreement with $\cos^3(t)$ is broken.
- ► The transfer matrix bifurcates into two branches. At some volume the kinetic term splits into a sum of two shifted Gausses.
- Every second slice **singular** vertices of very high order appear. New **order parameter**: $\max_{v} o(v)$.
- Not captured by global properties of the triangulation (N_0, N_{32}) .

The new bifurcation phase

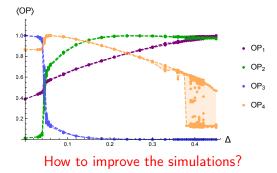


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- ▶ Not captured by global properties of the triangulation (N_0, N_{32}) .

Toroidal topology - phase transitions

For the toroidal topology, the C_{bif}/C_{dS} transitions reveals a hysteresis and very long **thermalization time**. Reason?

- ► Strong finite size effects
- Phase transition order depends on topology?



Motivation

- Limited single-core performance (gates, frequency, voltage, power)
- Cheap access to many cores

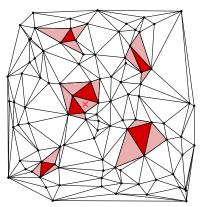
Naive parallelization is used to multiply the statistic, but ...

- ... a thermalized configuration is needed.
- ► Thermalization stage is a bottleneck
- Critical slowing down near phase transition
- Need to speed-up a single Monte Carlo Markov chain ⇒ parallelization
- ▶ Large minimal triangulation ⇒ strong finite size effects ⇒ large triangulations
- Simple parallelization does not work ...

Simple parallelization

A simple parallel algorithm, where many threads modify the same triangulation, has a very large overhead

- Detect collisions
- ► Thread synchronization: withdraw a move
- Move weight depends on global parameters threads are coupled



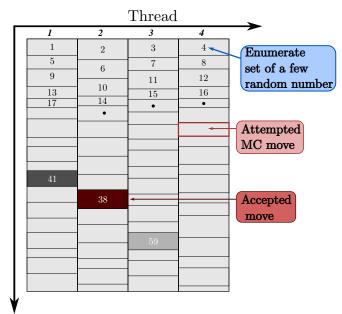
Parallel rejection

(by Jack Laiho, Phys. Rev. D 96 (2017) 064015)

- ► In some regions, the acceptance rate of MC moves is very low (critical slowing down)
- ▶ In such case a **parallel rejection** algorithm can be applied
- Multiple threads execute a series of attempted moves
- When one of them succeeds, rest of the threads are stopped and the triangulation is updated
- ► No collision detection is necessary
- ➤ To guarantee compatibility with the scalar version, *youngest* move has to be chosen (wait)
- ▶ Each attempted move is identified by three *random* numbers:
 - which move
 - **where** to make the move
 - whether to make the move (detailed balance)
- ▶ A block of random numbers is precomputed and split between threads. Synchronization with the RNG (separate thread).

Parallel rejection - diagram

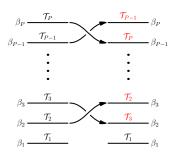
CPU time



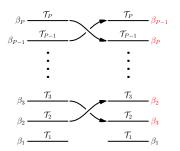
Parallel rejection - issues

- For three threads we gained 100% speed up for very small acceptance rate $\frac{1}{400}$
- Different move types significantly vary in acceptance rate
- Fixed number of random numbers (3) consumed per attempted move
- Replicability, compatibility with scalar version
- Thread synchronization is expensive
- Low level libraries, single barrier from pthreads
- Atomic operations (counter), busy waiting

- ▶ In this method we run a number of processes (replicas) with slightly different values of coupling constants.
- Periodically, the configurations are exchanged between processes with a certain probability (detailed balance).
- A *CDT configuration* is a heavy object (triangulation + additional data). Instead, the sets of a few coupling constants (denoted as β) are exchanged.



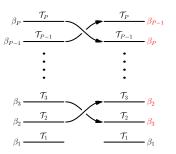
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Exchange probability fulfills the detailed balance condition,

$$\begin{split} P(A) &= P_1(\mathcal{T}_1) P_2(\mathcal{T}_2), \quad P(B) = P_2(\mathcal{T}_1) P_1(\mathcal{T}_2), \\ W(A \to B) &= \min\{1, \frac{P_2(\mathcal{T}_1) P_1(\mathcal{T}_2)}{P_1(\mathcal{T}_1) P_2(\mathcal{T}_2)}\} \end{split}$$

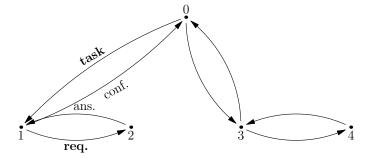
► Implementation: Managing process and *Linux message queues* to notify *working* processes when and with whom exchange



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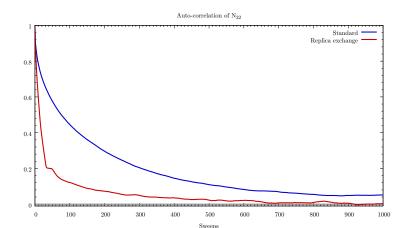
$$P(A) = P_1(T_1)P_2(T_2), \quad P(B) = P_2(T_1)P_1(T_2),$$
 $W(A \to B) = \min\{1, \frac{P_2(T_1)P_1(T_2)}{P_1(T_1)P_2(T_2)}\}$

► Implementation: Managing process and Linux message queues to notify working processes when and with whom exchange



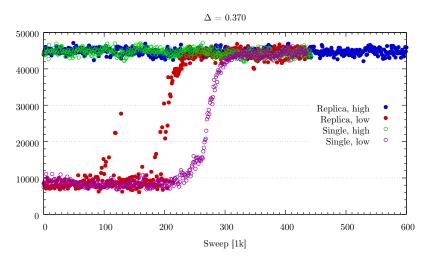
Autocorrelation time

- ► History of one configuration (Markov chain) contributes to many simulations and reduces the autocorrelation time.
- Exchange rate has to be higher than the autocorrelation time
- ► For 31 replicas a significant decrease of autocorrelation time is observed. A point near A/C transition (toroidal topology):



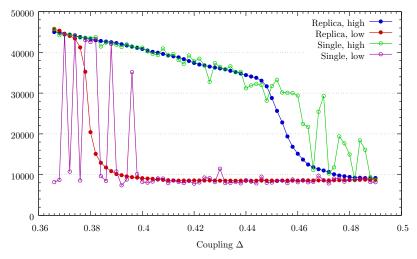
Replica exchange - hysteresis

Monte Carlo history of the relevant order parameter $\operatorname{Max} o(p)$ for single value of coupling constants. Visible jump and exchanges.



Replica exchange - hysteresis

Average value of the order parameter as a function of coupling constant. Comparison: replica exchange vs single threaded.



Thank You!