

Title: Tutorial: Monte Carlo methods in Dynamical Triangulations

Date: Jun 21, 2017 02:00 PM

URL: <http://pirsa.org/17060079>

Abstract:

Config

```
In[1]:= (* Provide the location of the randgeom executable *)
(* On windows change randgeom to randgeom.exe *)
programLocation = FileNameJoin[{NotebookDirectory[], "linux", "randgeom"}]

Out[1]:= /home/timothy/Documents/web/homepage/randgeom/linux/randgeom

In[2]:= (* Test the program. If it return False, check the programLocation provided. If it still does not work, first test randgeom from the console. *)
FileExistsQ[programLocation] && ListQ[RunThrough["" <> programLocation <> "", ""]]

Out[2]:= True
```

Useful functions

```
In[3]:= (* the following just runs randgeom with the specified parameters and parses the output as Mathematica code *)
generateMaps[type_, size_, number_] := RunThrough["" <> programLocation <> " -t" <> type <> " -s" <> ToString[size] <> " -n" <> ToString[number], ""];
generateMap[type_, size_] := First@generateMaps[type, size, 1];

In[5]:= (* given a permutation p of {1,2,...,n}, cycles[p] gives the partition of {1,2,...,n} into cycles *)
cycles[p_] := PermutationCycles[p, Identity];
(* given a list plist of permutations, orbits[plist] gives the partition of {1,2,...,n} into orbits under the permutations *)
orbits[plist_] := GroupOrbits@PermutationGroup[PermutationCycles /@ plist]

In[7]:= (* edges, vertices and faces correspond to cycles of halfedge-permutations *)
edgecycles[map_] := cycles[map][[All, 3]];
facecycles[map_] := cycles[map][[All, 1]];
vertexcycles[map_] := cycles[map][[map][[All, 3]], 1]];
(* We may assign id's to the vertices of map according to their position in vertexcycles[map] *)
halfedgeToVertexId[map_] := Dispatch[Join@@MapIndexed[#1 -> #2[[1]] &, vertexcycles[map], {2}]];
halfedgeToFaceId[map_] := Dispatch[Join@@MapIndexed[#1 -> #2[[1]] &, facecycles[map], {2}]];

In[12]:= (* functions to construct a Mathematica Graph object *)
uniqueEdges[map_] := Union[Sort /@ (edgecycles[map] /. halfedgeToVertexId[map])];
uniqueDualEdges[map_] := Union[Sort /@ (edgecycles[map] /. halfedgeToFaceId[map])];
mapGraph[map_] := With[{edges = uniqueEdges[map]}, Graph[Union@@edges, #[[1]] -> #[[2]] & /@ edges, GraphLayout -> None]];
mapDualGraph[map_] := With[{edges = uniqueDualEdges[map]}, Graph[Union@@edges, #[[1]] -> #[[2]] & /@ edges, GraphLayout -> None]];


```

Plotting

```
In[16]:= (* the following is a bit of a hack to extract coordinates from GraphPlot3D's embedding of a graph *)
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```

Plotting

```

In[16]:= (* the following is a bit of a hack to extract coordinates from GraphPlot3D's embedding of a graph *)
get3DGraphEmbedding[edg, method_] := (VertexCoordinateRules /. Cases[GraphPlot3D[#, # [[1]] -> # [[2]] & /@ edg, Method -> method, _Rule, Infinity]][Ordering @ DeleteDuplicates[Join @@ (List @@ edg)]]);

In[17]:= (* one can play with the options here to adapt the embedding *)
coordinates[map_] := get3DGraphEmbedding[uniqueEdges[map], {"SpringElectricalEmbedding", "InferentialDistance" -> Automatic, "RepulsiveForcePower" -> -2.4}];

In[18]:= (* assign colors to faces according to closeness centrality *)
colorFunction[x_] := ColorData["Rainbow"][Round[Max[Min[0.55 - 0.15 x, 1], 0], 1/40]];
centralityColors[map_] := colorFunction /@ Standardize @ ClosenessCentrality[mapDualGraph[map]];

In[20]:= plotMap3D[map_, coord_] :=
Graphics3D[{EdgeForm[None], {FaceForm[#, # [[1]]], Polygon[coord][#, # [[2], ;; 3]]}}, If[Length[#, # [[2]]] == 4, Polygon[coord][#, # [[2], (1, 3, 4)]]], {}] & /@
Transpose[{centralityColors[map], facecycles[map] /. halfedgeToVertexId[map]}], Boxed -> False]

In[21]:= map = generateMap["C", 500];
plotMap3D[map, coordinates[map]]

```

(* Task: plot geometries of various sizes and models. What are the qualitative differences between the models? *)

(* Task: produce some nice pictures. To save a nice picture, one may use something like Export["picture.png", ImageCrop @ Rasterize[plot, ImageSize -> 800, Background -> None]] *)

Geodesic two-point function

(* Mathematica has built in support for graph distances: this returns a list of distances from all vertices to a randomly chosen vertex *)

```
distanceListFromRandomPoint[graph_] := GraphDistance[graph, RandomChoice @ VertexList[graph]];
```

(* produce a histogram with the fraction of points at distance 0, 1, 2, ... *)

```

halfEdgeToVertexId[map_] := Dispatch[Join @@ MapIndexed[#1 → #2[[1]] &, vertexcycles[map], {2}]];
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mapGraph[map_] := With[{edges = uniqueEdges[map]}, Graph[Union @@ edges, #[[1]] ↔ #[[2]] & /@ edges, GraphLayout → None]];
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Transpose[{centralityColors[map], facecycles[map] /. halfEdgeToVertexId[map]}], Boxed → False]

In[21]:= map = generateMap["C", 500];
plotMap3D[map, coordinates[map]]

```

Out[22]=



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```

halfedgeToVertexId[map_] := Dispatch[Join @@ MapIndexed[#1 -> #2[[1]] &, vertexcycles[map], {2}]];
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coordinates[map_] := get3DGraphEmbedding[uniqueEdges[map], {"SpringEmbedding", "ForceDirectedEmbedding" -> Automatic, "SpringLength" -> 1.2}];

In[18]:= (* assign colors to faces according to closeness centrality *)
colorFunction[x_] := ColorData["Rainbow"][Round[Max[Min[0.55 - 0.15 x, 0], 1, 0.9]]];
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Transpose[{centralityColors[map], facecycles[map] /. halfedgeToVertexId[map]}]}, Boxed -> False];

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```



Out[22]=

timothy@timothy-UX31A ~

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timothy@timothy-UX31A ~ \$

100%

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```
(* Task: plot geometries of various sizes and models. What are the qualitative differences between the models? *)
(* Task: produce some nice pictures. To save a nice picture, one may use something like Export["picture.png",ImageCrop@Rasterize[plot,ImageSize->800,Background->None]] *)
```

Geodesic two-point function

```
In[23]:= (* Mathematica has built in support for graph distances: this returns a list of distances from all vertices to a randomly chosen vertex *)
distanceListFromRandomPoint[graph_] := GraphDistance[graph, RandomChoice@VertexList[graph]];
```

```
In[24]:= (* produce a histogram with the fraction of points at distance 0,1,2,3,... *)
distanceProfile[map_, max_] := BinCounts[#, {0, max}] / Length[#] & @ distanceListFromRandomPoint @ mapGraph[map];
dualDistanceProfile[map_, max_] := BinCounts[#, {0, max}] / Length[#] & @ distanceListFromRandomPoint @ mapDualGraph[map];
```

```
In[26]:= (* An example of a distance profile (from a random vertex) for a single random geometry *)
ListPlot[distanceProfile[generateMap["C", 500], 30], Joined -> True, Filling -> Axis, InterpolationOrder -> 0, AxesLabel -> {"r", "\rho(r)"}]
```

```
(* the same but for dual graph distance *)
ListPlot[dualDistanceProfile[generateMap["C", 500], 30], Joined -> True, Filling -> Axis, InterpolationOrder -> 0, AxesLabel -> {"r", "\rho(r)"}]
(* Task: Proceed to gather data for the average distance profile for different system sizes and models, and attempt finite-size scaling to extract the Hausdorff dimension *)
```

Spectral dimension

The spectral dimension d_s of a map is related to the probability $p(t)$ that a simple random walk on the map (or its dual) returns to its starting point after t steps: $p(t) \sim t^{-d_s/2}$ for $1 \ll t \ll n$ (where n is the system size). There are various ways to measure this return probability: one can simulate a random walker and just record its returns; study a heat diffusion process; or use linear algebra as follows. The return probability $\langle p(t) \rangle$ averaged over all starting points of the map is related to the normalized adjacency matrix A via $\langle p(t) \rangle = \text{Tr}(A^t) / n = \sum \lambda_i^t / n$, where λ_i are the eigenvalues of A .

```
dualAdjacencyMatrix[map_] := Module[{dualEdges, matrixEntries, facedegrees},
  dualEdges = edgeCycles[map] /. halfEdgeToFaceId[map];
  facedegrees = Length /@ faceCycles[map];
  matrixEntries = Tally @ Join[dualEdges, Reverse /@ dualEdges];
  SparseArray[#, [[1] -> # [[2]] / facedegrees[#[[1, 1]]] & /@ matrixEntries]
]
(* plot the adjacency matrix (for fun) *)
MatrixPlot[dualAdjacencyMatrix[generateMap["D", 300]]]
```

- (* Task: plot geometries of various sizes and models. What are the qualitative differences between the models? *)
- (* Task: produce some nice pictures. To save a nice picture, one may use something like `Export["picture.png",ImageCrop@Rasterize[plot,ImageSize->800,Background->None]]` *)

Geodesic two-point function

In[23]:= (* Mathematica has built in support for graph distances: this returns a list of distances from all vertices to a randomly chosen vertex *)

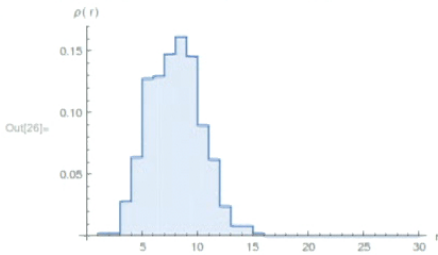
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```

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distanceProfile[map_, max_] := BinCounts[#, {0, max}] / Length[#] & @ distanceListFromRandomPoint @ mapGraph[map];
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In[26]:= (* An example of a distance profile (from a random vertex) for a single random geometry *)

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ListPlot[distanceProfile[generateMap["C", 500], 30], Joined -> True, Filling -> Axis, InterpolationOrder -> 0, AxesLabel -> {"r", "\rho(r)"}]
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(* the same but for dual graph distance *)

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ListPlot[dualDistanceProfile[generateMap["C", 500], 30], Joined -> True, Filling -> Axis, InterpolationOrder -> 0, AxesLabel -> {"r", "\rho(r)"}]
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(* Task: Proceed to gather data for the average distance profile for different system sizes and models, and attempt finite-size scaling to extract the Hausdorff dimension *)

Spectral dimension

The spectral dimension d_s of a map is related to the probability $p(t)$ that a simple random walk on the map (or its dual) returns to its starting point after t steps: $p(t) \sim t^{-d_s/2}$ for $1 \ll t \ll n$ (where n is the system size). There are various ways to measure this return probability: one can simulate a random walker and just record its returns; study a heat diffusion process; or use linear algebra as follows. The return probability $\langle p(t) \rangle$ averaged over all starting points of the map is related to the normalized adjacency matrix

(*) Task: plot geometries of various sizes and models. What are the qualitative differences between the models? (*)

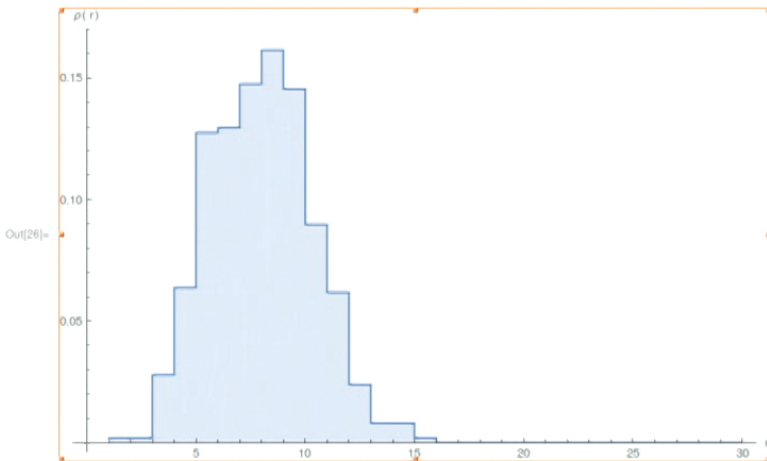
(*) Task: produce some nice pictures. To save a nice picture, one may use something like `Export["picture.png", ImageCrop@Rasterize[plot, ImageSize -> 800, Background -> None]]` (*)

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In[26]:= (*) An example of a distance profile (from a random vertex) for a single random geometry *)
`ListPlot[distanceProfile[generateMap["C", 500], 30], Joined -> True, Filling -> Axis, InterpolationOrder -> 0, AxesLabel -> {"r", " $\rho(r)$ "}]`



Usage of randgeom

randgeom takes three parameters:

- -t followed by A, B, C, or D: the requested model type.
- -s followed by positive integer S: the requested size S of the planar map measured by number of faces.
- -n followed by positive integer N: the number N of independent configurations to be returned.

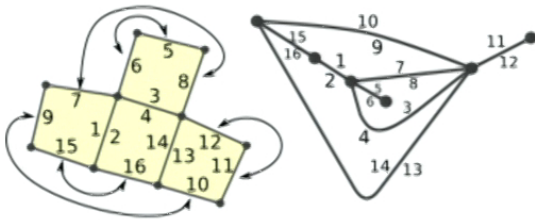
The following returns a single random planar map sampled from model A with 4 faces.

```

$ ./randgeom -tA -s4 -n1
{{ {7, 15, 2}, {16, 4, 1}, {8, 6, 4}, {2, 14, 3}, {6, 8, 6}, {3, 5, 5}, {9, 1, 8}, {5, 3, 7}, {15, 7, 10}, {11, 13, 9}, {12, 10, 12},
{13, 11, 11}, {10, 12, 14}, {4, 16, 13}, {1, 9, 16}, {14, 2, 15}}

```

The output is formatted as a Mathematica-style nested list of length N, one entry per configuration. Each configuration corresponds to a list of triples of integers describing a planar map through permutations: the i 'th triple corresponds to $\{n(i), n^{-1}(i), a(i)\}$, where n and a are the "next" and "adjacent" permutations (see lecture slides). The output above corresponds to the following quadrangulation displayed both as a gluing prescription and as a planar map.

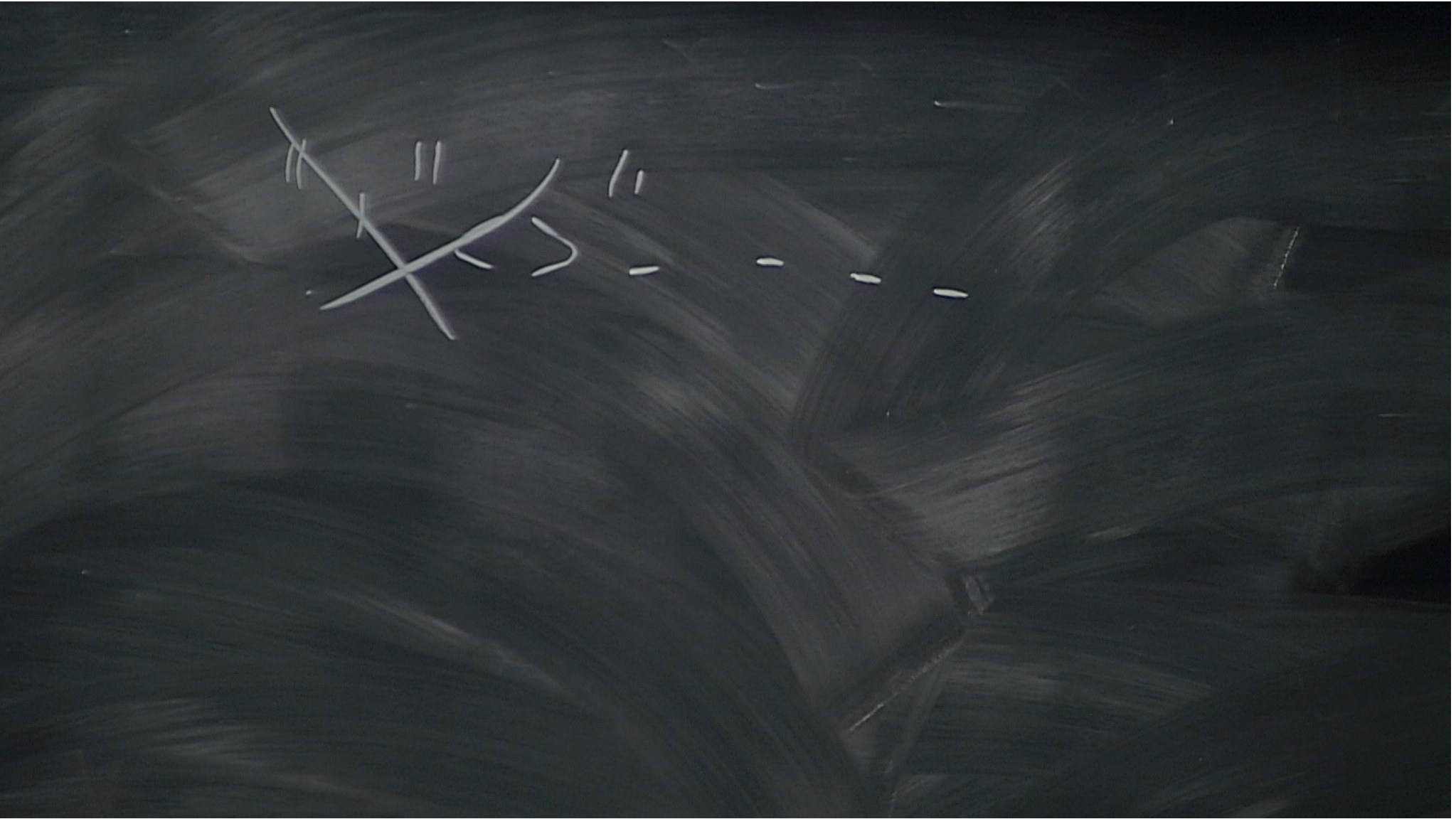


In case you prefer to read the output from a different program than Mathematica, you may prefer to receive the data as a space-separated list. To this end one may use the option `--space-separated`:

```

$ ./randgeom -tA -s2 -n2 --space-separated
2
8
3 7 2
8 4 1
5 1 4
2 6 3
7 3 6
4 8 5
1 5 8

```

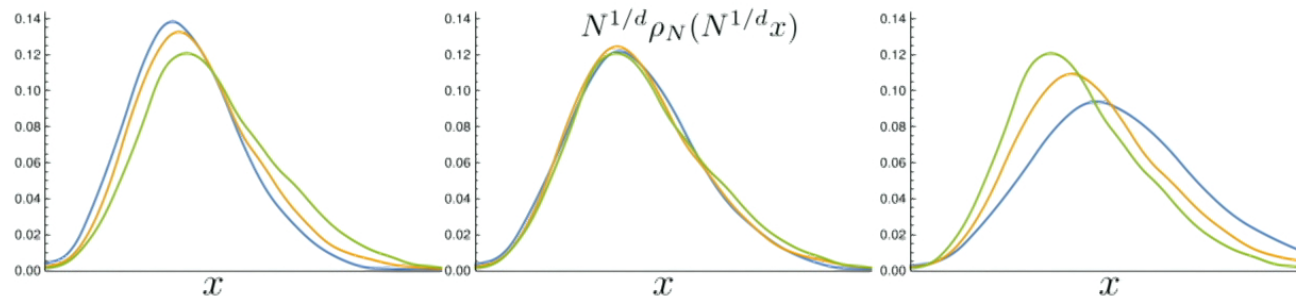


Observables: geodesic distances (continued)

- ▶ Instead of taking $N \rightarrow \infty$ and then $r \rightarrow \infty$, it is usually better to use *finite-size scaling methods*: one expects

$N^{1/d_H} \rho_N(N^{1/d_H} x)$ to converge as $N \rightarrow \infty$ for any fixed $x \in \mathbb{R}$.

- ▶ Equivalently, we expect the distribution of the distance between two random points to converge as $N \rightarrow \infty$ provided we take edge lengths $\sim N^{-1/d_H}$.
- ▶ Estimate d_H by “collapsing curves”:



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jupyter Quantum-Cuboid-renormalization Last Checkpoint: 8 hours ago (autosaved) Julia 0.5.0

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Tutorial: Quantum cuboid renormalisation

Basic definitions

```
In [ ]: function DE(j1::Float64,j2::Float64,j3::Float64,j4::Float64,j5::Float64,j6::Float64)
    res = (-2) * (j1^2 * (j2 + j4) + j2 * j4 * (j2 + j4) + j1 * (j2^2 + (1 + 1im) * j2 * j4 + j4^2))
    res += (j1^2 * (j3 + j5) + j3 * j5 * (j3 + j5) + j1 * (j3^2 + (1 + 1im) * j3 * j5 + j5^2))
    res += (j3 * j4 * j5 + j2 * (j4 * j5 + j3 * (j4 + j5)))
    res += (j2^2 * (j3 + j6) + j3 * j6 * (j3 + j6) + j2 * (j3^2 + (1 + 1im) * j3 * j6 + j6^2))
    res += (j4^2 * (j5 + j6) + j5 * j6 * (j5 + j6) + j4 * (j5^2 + (1 + 1im) * j5 * j6 + j6^2))
    res += (j3 * j4 * j6 + j1 * (j4 * j6 + j3 * (j4 + j6)))
    res += (j2 * j5 * j6 + j1 * (j5 * j6 + j2 * (j5 + j6)))

    res
end

function DF(j1::Float64,j2::Float64,j3::Float64,j4::Float64,j5::Float64,j6::Float64)
    res = (-2) * (j1^2 * (j2 + j4) + j2 * j4 * (j2 + j4) + j1 * (j2^2 + (1 - 1im) * j2 * j4 + j4^2))
    res += (j1^2 * (j3 + j5) + j3 * j5 * (j3 + j5) + j1 * (j3^2 + (1 - 1im) * j3 * j5 + j5^2))
    res += (j3 * j4 * j5 + j2 * (j4 * j5 + j3 * (j4 + j5)))
    res += (j2^2 * (j3 + j6) + j3 * j6 * (j3 + j6) + j2 * (j3^2 + (1 - 1im) * j3 * j6 + j6^2))
    res += (j4^2 * (j5 + j6) + j5 * j6 * (j5 + j6) + j4 * (j5^2 + (1 - 1im) * j5 * j6 + j6^2))
    res += (j3 * j4 * j6 + j1 * (j4 * j6 + j3 * (j4 + j6)))
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    res
end

function Ampl(a::Float64,j1::Float64,j2::Float64,j3::Float64,j4::Float64,j5::Float64,j6::Float64)
    res = 1/((2. * pi)^3) * 2.^(24)
    res += (1/(sqrt(-DE(j1,j2,j3,j4,j5,j6))) + 1i)/(sqrt(-DF(j1,j2,j3,j4,j5,j6)))
    res += (1/(sqrt(-DE(j1,j2,j3,j4,j5,j6))) + 1)/(sqrt(-DF(j1,j2,j3,j4,j5,j6)))
```

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