Computational Astrophysics Monsoon 2022

1 n-Body Tree

Constructing the octree data structure has several steps. A broad outline of the algorithm would be -

- a. Construct a function to split space into equal octants.
- b. Construct a function to identify which octant a particle resides in.
- c. Define a new data structure to store tree-node information.
- d. Define a top-leaf (in other words, a parent node) in the form of the newly created Tree-node data type.
- e. Construct a function which recursively fills in all the tree nodes by repeatedly splitting space until there are zero or one particles left in the node.

Once the tree structure has been created, computing any quantities of interest (such as the angle subtended by a node diagonal on a particle) would also involve recursively tracing the branches of the tree. I have created two main modules (treeGenerator.jl and treeAttributes.jl) which generate the desired tree and acquires desired attributes from the tree. I will quickly summarize each of their functions and more details about low-level implementations can be gathered by following the comments in my Julia code.

1.1 Code Walk-through

Following is the code-listing of treeGenerator.jl. A vital step is to define a new datastructure TreeNode. Each TreeNode has five properties - its spatial span, positions of the particles inside that spatial span, the center of mass of the particles inside the node, total mass of the node and a tree of daughter tree nodes. The next function definition in the module is to divide a 3D space of a defined size into equal octants. There are two types of spans which constitute the octant bounds - either the bound spans between minimum to mid-point (labelled as 1) or from mid-point to maximum (labelled as 2). Using an array of these span types, we can uniquely label each of the octants. The next function, uses particle's location and these label types to identify what octant of the given node would the particle reside in. Finally, we have the function which populates the entire tree recursively. The base cases are identified when there is one or no particle and recursion into further nodes stops there. In case of large data-sets, we also define a maximum recursion depth as an upper bound of the tree-layers.

```
### treeGenerator.jl Module ###
2 ################################
  "For turning n-body position data into octrees"
5 # creating custom julia structure to store tree type data
6 struct TreeNode
      bounds # stores the spatial bounds of the node
      bodies # stores the particles in the node
             # stores center of mass of the node
      mass
             # stores the mass of the node (in solar masses)
      nodes # stores the child nodes
11
12 end:
13
# function to divide the node into 8 equal octants
15 function canvas_split(bound)
16
      # unpacking bounds of the full region
17
      xMin, xMax, yMin, yMax, zMin, zMax = bound
19
      # Calculating midpoints of the total region
20
      xMid = (xMax + xMin)/2.0
      yMid = (yMax + yMin)/2.0
22
      zMid = (zMax + zMin)/2.0
23
24
      # Defining the new bounds of the 8 octants
25
      new_bounds = [(xMin, xMid, yMin, yMid, zMin, zMid), # label (1, 1, 1)
26
                      (xMid, xMax, yMin, yMid, zMin, zMid), \# label (2, 1, 1)
27
                      (xMin, xMid, yMid, yMax, zMin, zMid), # label (1, 2, 1) (xMin, xMid, yMin, yMid, zMid, zMax), # label (1, 1, 2)
28
```

```
(xMid, xMax, yMid, yMax, zMin, zMid), # label (2, 2, 1)
                     (xMid, xMax, yMin, yMid, zMid, zMax), \# label (2, 1, 2)
31
                     (xMin, xMid, yMid, yMax, zMid, zMax), # label (1, 2, 2) (xMid, xMax, yMid, yMax, zMid, zMax) # label (2, 2, 2)
32
33
      ٦
34
35
      return new_bounds # output is a list of 8 tuples
36
37 end;
39
40 # function to locate which octant a particle belongs to
41 function which_node(particle_location, nodes_boundaries)
42.
43
      xMid = nodes_boundaries[1][2]
                                       # extracting node boundary information
      yMid = nodes_boundaries[1][4] # extracting node boundary information
44
                                       # extracting node boundary information
      zMid = nodes_boundaries[1][6]
45
46
      x, y, z = particle_location # unpacking particle locations
47
48
                            # checking if x-coordinate passes the y-z plane
49
      yPass = y < yMid
                           # checking if y-coordinate passes the x-z plane
50
      zPass = z < zMid
                           # checking if z-coordinate passes the x-y plane
51
52
      identifier = (xPass, yPass, zPass) # creating an identifier to check which octant the particle
53
54
      if identifier == (true, true, true)
55
56
      elseif identifier == (false, true, true)
57
          n = 2
58
      elseif identifier == (true, false, true)
59
60
          n = 3
61
      elseif identifier == (true, true, false)
         n = 4
62
      elseif identifier == (false, false, true)
63
64
      elseif identifier == (false, true, false)
65
          n = 6
66
      elseif identifier == (true, false, false)
67
         n = 7
68
      else
          n = 8
70
71
      end
72
73
      return n
74 end;
76 # The following function builds the tree recursively
77 # Input parameters: parent node, particles in the parent node, boundaries of the parent node, max
      allowed recursion depth
_{78} # Output: a tree structure with the parent node as the root
80 function build_tree(particles, nodes_boundaries, depth)
81
82
      # Base case: if there is only one particle in the node or the depth of the tree is \emptyset
      if length(particles) <= 1 || depth == 0</pre>
83
           return TreeNode(nodes_boundaries, particles, compute_COM(particles), length(particles),
84
      nodes_boundaries)
85
      end
      # keeping track of which particle belongs to which octant
87
      child = [which_node(particles[i], nodes_boundaries) for i in eachindex(particles)]
88
89
      # Recursively building the tree
90
      new_nodes = [build_tree(particles[child .== i], canvas_split(nodes_boundaries[i]), depth-1)
91
      for i in 1:8]
92
      return TreeNode(nodes_boundaries, particles, compute_COM(particles), length(particles),
93
       new_nodes)
94 end:
```

The second module involves functions to compute center of mass of a node, total mass of the node, angle subtended by a node on a particle etc. and methods to find these quantities recursively for all non-empty leafs.

```
# angles subtended on a particle by a node
function angle_subtended(particle_location, node)
# unpacking particle location
```

```
x, y, z = particle_location
6
      # unpacking node boundaries
      xMin, xMax, yMin, yMax, zMin, zMax = node.bounds[1][1], node.bounds[8][2], node.bounds[1][3],
8
      node.bounds[8][4], node.bounds[1][5], node.bounds[8][6]
      d = sqrt((xMax - xMin)^2 + (yMax - yMin)^2 + (zMax - zMin)^2)
                                                                    # distance between the
10
      two corners of the node
      xMid, yMid, zMid = (xMin + xMax)/2, (yMin + yMax)/2, (zMin + zMax)/2 # center of the node
13
      r = sqrt((x - xMid)^2 + (y - yMid)^2 + (z - zMid)^2) # distance between the center of the node
      and the particle
14
15
      # calculating the angle subtended by the node on the particle
      angle = 2*atan(d/r)
16
17
      return angle *180/pi # output is in degrees
19 end:
20
21
23 ####### COM LIST ###########
24 ################################
26 # Center of Mass computing function for equal mass particles
27 function compute_COM(particle_list)
      N = length(particle_list)
                                     # total number of particles
29
      vecAdd = [0,0,0]
                                     # initiating array to compute center of mass
30
31
      i = 1
                                     # initiating iterating index
32
      while i <= N
33
34
          vecAdd += particle_list[i] # adding each particle's contribution to the total
          i += 1
35
      end
36
37
                                     # dividing my total number of particles (note it only works
      return vecAdd/N
38
      with equal mass particles)
39 end:
40
_{\rm 41} # recursive function to get the centers of mass of all nodes
42 function get_COMs(tree, COM_list)
      \# if node has only one particle storing particle position as COM of the node
43
      if length(tree.bodies) == 1
         push!(COM_list, tree.bodies[1])
45
46
47
      push!(COM_list, tree.COM)
                                               # appending the center of mass position to the list
48
49
      for i in 1:8
                                                # iterating over the 8 octants
         if typeof(tree.nodes[i]) == TreeNode # convoluted way of checking if the node is empty
50
51
              get_COMs(tree.nodes[i], COM_list) # recursive step
          end
52
53
      end
54 end;
55
_{56} # Recursively tracing the tree to find the centers of mass of the non-empty nodes
57 # Input: tree structure
58 # Output: list of center of mass locations for all "non-empty" nodes
60 function list_COMs(tree)
61
      {\tt COM\_list} = []  # intiating an empty list to store the centers of mass
62
63
      get_COMs(tree, COM_list)
                                                        # updating list
64
      COM_list = [x for x in COM_list if !isnan(x[1])] # removing the NaNs from the list
65
67
      return COM_list
68 end;
69
71 ####### MASS LIST ###########
74 # recursive function to get the masses of all nodes
75 function get_mass(tree, mass_list)
^{77} # if node has only one particle, storing particle mass as mass of the node
```

```
if length(tree.bodies) == 1
           push!(mass_list, tree.mass)
79
80
       end
81
       push!(mass_list, tree.mass)
                                                    # appending the center of mass position to the
82
       for i in 1:8
                                                   # iterating over the 8 octants
83
           if typeof(tree.nodes[i]) == TreeNode
                                                  # convoluted way of checking if the node is empty
84
               get_mass(tree.nodes[i], mass_list) # recursive step
86
       end
87
88 end:
90 # Recursively tracing the tree to find the centers of mass of the non-empty nodes
91 # Input: tree structure
_{92} # Output: list of masses of corresponding to the COMs generated by list_COM() function
94 function list_mass(tree)
       mass_list = []  # intiating an empty list to store the centers of mass
96
97
       get_mass(tree, mass_list)
                                                            # updating list
98
       mass_list = [x for x in mass_list if x > 0]
                                                           # removing the NaNs from the list
99
100
       return mass_list
102 end;
103
104
106 ######## NODE LIST ##########
107 ###################################
108
109 # recusrively finding angle subtended by each non-empty node on a particle
function get_nodes(particle, tree, theta_cutoff, outLists)
       # base case: angle_subtended is less than theta_cutoff
       if angle_subtended(particle, tree) <= theta_cutoff || length(tree.bodies) == 1</pre>
114
           return push!(outLists, tree.COM)
116
       # recursively finding the centers of mass of the nodes to be considered
       for i in 1:8
                     # iterating through octants
118
119
          if typeof(tree.nodes[i]) == TreeNode
               push!(outLists, get_nodes(particle, tree.nodes[i], theta_cutoff, outLists))
120
           end
121
122
       end
123
      return outLists
124
125 end;
126
function list_nodes(p, cutoff, tree)
       consider_COM = []
129
       consider_COM = get_nodes(particles[p], tree, cutoff, consider_COM)
130
131
       # removing non-vector elements from the list
132
       consider_COM = [x \text{ for } x \text{ in consider}_COM \text{ if typeof}(x) == Vector{Float64} && !isnan(x[1])];
134
135
       return consider_COM
```

Running the code is straightforward. We start with particle locations stored in a clean CSV file (developed from the provided data). Then using the farthest distances of the given particles, we compute the total size of the space. We divide the total space into octants manually to initiate the top leaf. Then we provide the particle list and top leaf data to treeGenerator script which fills up all the branches.

```
include("treeGenerator.jl");
include("treeAttributes.jl");

using CSV # importing libraries to read the data

# reading position of equal mass n-bodies from a csv file
pos = CSV.read("nbody.csv", DataFrame, header=true);
particles = [[pos.X[i], pos.Y[i], pos.Z[i]] for i in 1:Integer(nrow(pos))]; # creating a list of particles positions

# size of the parent node
```

Finally, a fairly ugly function had to be defined to visualize node boundaries as it served as a useful check during the code development process. We will see the results of this visualization code (and others) in action in the next section.

```
# special funcion to plot octant division of a node
function plot_nodes(nodes_boundaries, color, alpha)
         for i in 1:8
              xMin, xMax, yMin, yMax, zMin, zMax = nodes_boundaries[i]
              plot!([xMin, xMax], [yMin, yMin], [zMin, zMin], color=color, alpha=alpha, linewidth=2)
              plot!([xMin, xMax], [yMax, yMax], [zMin, zMin], color=color, alpha=alpha, linewidth=2)
plot!([xMin, xMax], [yMin, yMin], [zMax, zMax], color=color, alpha=alpha, linewidth=2)
6
              plot!([xMin, xMax], [yMax, yMax], [zMax, zMax], color=color, alpha=alpha, linewidth=2)
              plot!([xMin, xMin], [yMin, yMax], [zMin, zMin], color=color, alpha=alpha, linewidth=2) plot!([xMin, xMin], [yMin, yMax], [zMax, zMax], color=color, alpha=alpha, linewidth=2)
10
11
              plot!([xMax, xMax], [yMin, yMax], [zMin, zMin], color=color, alpha=alpha, linewidth=2)
              plot!([xMax, xMax], [yMin, yMax], [zMax, zMax], color=color, alpha=alpha, linewidth=2)
plot!([xMin, xMin], [yMin, yMin], [zMin, zMax], color=color, alpha=alpha, linewidth=2)
              plot!([xMin, xMin], [yMax, yMax], [zMin, zMax], color=color, alpha=alpha, linewidth=2)
14
              plot!([xMax, xMax], [yMin, yMin], [zMin, zMax], color=color, alpha=alpha, linewidth=2)
plot!([xMax, xMax], [yMax, yMax], [zMin, zMax], color=color, alpha=alpha, linewidth=2)
15
16
        end
17
18 end:
```

1.2 Results

Once the tree is generated, we can use some specially defined plotting functions to visualizing the data in three dimensions. For example, the following code snippet chooses the first particle from the particle list and computes the angle subtended on the particle by a node generated on the third level of the tree. Then, it proceeds to plot the node boundaries to illustrate division into octants, positions of all n-bodies and the angle created between desired node diagonal and chosen particle. Using this functionality, the list (added in Appendix 1) of all 312 nodes with non-zero masses and their center of mass coordinates were enumerated .

```
_{
m 1} # illustrative visualization of subtended angle calculation for a single particle and single node
 p p, n = 1, 1 #particle index, node index
 4 node = tree.nodes[5].nodes[3].nodes[n]
 s xMin, xMax, yMin, yMax, zMin, zMax = node.bounds[1][1], node.bounds[8][2], node.bounds[1][3], node
               .bounds[8][4], node.bounds[1][5], node.bounds[8][6];
 7 angle = round(angle_subtended(particles[p], node))
 9 scatter(pos.X, pos.Y, pos.Z, legend = false, markersize=0.6, alpha=0.8, label="All Particles");
# plotting subtended angle
12 scatter!((particles[p][1], particles[p][2], particles[p][3]), color="red", legend = false,
               markersize=2) # plotting the particle
13 plot!([xMin, xMax], [yMin, yMax], [zMin, zMax], color="blue", linewidth=3) # plotting the node
               diagonal
15 # plotting lines connecting the particle to edges of the node diagonal
16 plot!([particles[p][1], xMax], [particles[p][2], yMax], [particles[p][3], zMax], color="blue",
               linewidth=3) # plotting the subtended angle
17 plot!([particles[p][1], xMin], [particles[p][2], yMin], [particles[p][3], zMin], color="blue",
               linewidth=3) # plotting the subtended angle
18
20 # plotting the node boundaries
plot_nodes(tree.bounds, "red", 0.4);
plot_nodes(tree.nodes[5].bounds, "red", 0.2);
plot_nodes(tree.nodes[5].nodes[3].bounds, "black", 0.3);
{\tt 24~display(plot!(dpi=200,~size=(900,~800),~camera=(15,~25,~250),~background\_color="white",~title="align: color="white",~title="align: color="align: color="align: color="white",~title="align: color="align: co
               Canvas Spliting (node depth = 3) and Subtended Angle \tilde{\ } $angle",
                                        xlabel="x", ylabel="y", zlabel="z"))
```



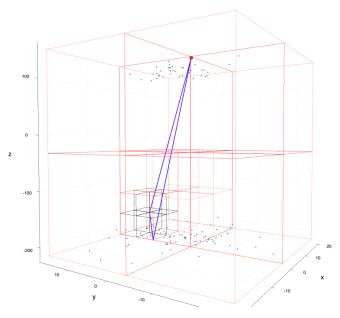
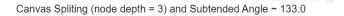


Figure 1: Plotting the 16° angle (bold lines) subtended by a third layer leaf of the tree on the first particle (bold spot). Other particles from the list plotted as smaller dots and the node boundaries are plotted as light lines.

We can run the exact same code while choosing some other particle to see what angle is subtended by the same node on the particle. By comparing the results of Fig. 2 and Fig. 1, we can clearly see how the angle subtended on a particle closer to a particle node is much greater than the angle subtended on a particle farther away.



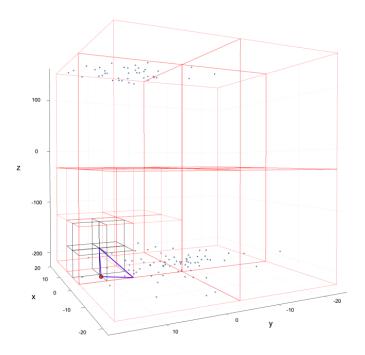


Figure 2: Plotting the 113^o angle (bold lines) subtended by a third layer leaf of the tree on the 50th particle (bold spot). Other particles from the list plotted as smaller dots and the node boundaries are plotted as light lines.

We can set a cut-off angle such that only nodes subtending an angle smaller than the cutoff would be listed for further computations. This is achieved by calling the <code>list_nodes(p, cutoff, tree)</code> function which takes the

particle index of interest, cutoff angle and the data tree as its input. Though I noticed the particles are broadly situated on two parallel planes, with $\theta_o=1^o$, list_nodes() output is 120 entries long (119 nodes to consider + the particle's self node). However, Fig. 3 shows how the number of nodes to be considered during force calculations falls as we increase the cutoff angle for a random selection of a few particles.

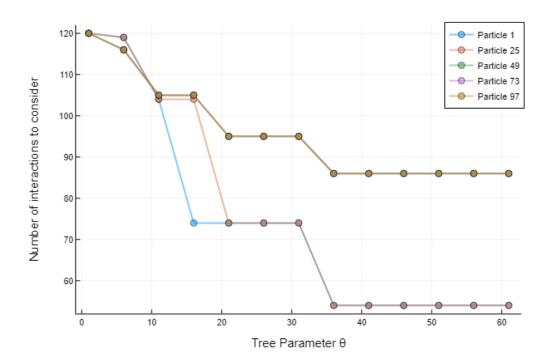


Figure 3: Relation between cutoff angle and number of nodes to consider

Now that the novel aspects of writing an octree data structure and recursively parsing through it have been developed, I look forward to developing a complete Barnes-Hut n-body solver in the coming weeks.

Appendix: Node COM Locations and Masses

To generate the desired list of center of mass locations for each non-empty node and their respective masses, we generate the tree and call the following functions

```
1 COM_list = list_COMs(tree); # list of centers of mass of all non-empty nodes
2 mass_list = list_mass(tree); # list of masses of all non-empty nodes
3 out = [(COM_list[i], mass_list[i]) for i in eachindex(mass_list)] # array to store
4 using DelimitedFiles # required for writing ouput files
6 writedlm("out.txt", out) # writes to a text file ([COM location, mass] for each node)
```

The data is too big to be meaningfully visualized in a 2D black and white non-interactive medium. Therefore, Fig. 4 displays center of mass locations along with node masses for the first two layers of the tree (top leaf and the first splitting of space into octants) to demonstrate the output matches the intuition.

Node COMs for two levels of tree

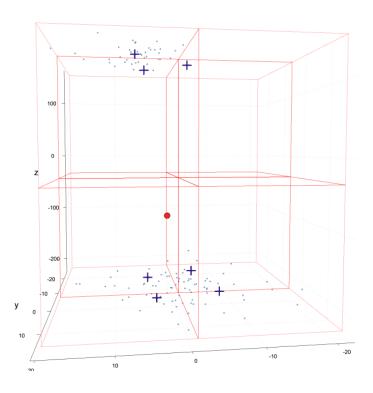


Figure 4: Node Center of masses positions. The big circle corresponds to the COM of the total n-body system and the crosses represent the nodes of the first layer of the octree. Smallest dots correspond to the equal mass particles in space.

This generates 312 entries corresponding to 312 non-empty leafs at various levels of the tree.