

# The Barnes Hut Algorithm: Implementation and Simulation in Julia

20602 - Computer Science (Algorithms)

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20th October 2021



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# The N-body problem

The setting:

- Delimited region of space  
in  $\mathbb{R}^2$



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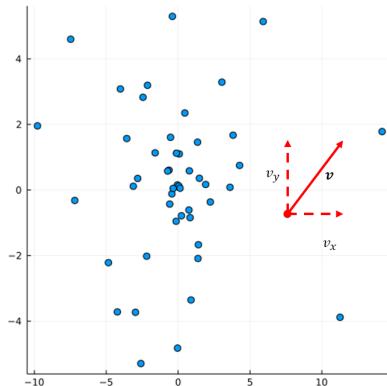
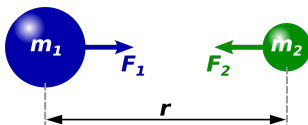


Figure: A cloud of 50 bodies

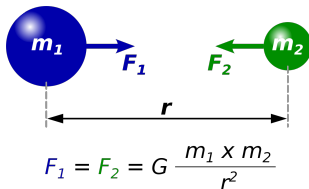
# The N-body problem

The points in space are subject to a force of attraction, which in this case is the **gravitational pull**, and it behaves according to Newton's law:


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# The N-body problem

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Then, the point is moved according to Newton's second law of mechanics:

$$\vec{a}_i = \frac{\vec{F}_i^{net}}{m_i}$$

# Brute Force Approach

Therefore, to compute the net force acting on a body one needs to compute all the pairwise gravitational interactions between points.

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- Such time complexity is feasible only for small  $N$
- For example: Earth–Sun simulation, Solar System ( $N \sim 20$ )



# Brute Force Approach

```
function onestepBrute(time::Float64,stars::Array{Star,1},spaceScale::Int64)
    new_stars = copy(stars)
    # prepare a NxNx2 tensor for the interactions
    F_mat = zeros(length(stars),length(stars),2)
    for i in 1:length(stars)
        # compute the net force
        for j in i:length(stars)
            # compute force for all pairs not already computed
            if i != j
                F = newton(stars[j],stars[i])
                d_j = stars[j]-stars[i]
                cos_0j, sin_0j = get_cos_sin(d_j)
                f_j = [F * cos_0j, F * sin_0j]
                F_mat[i,j,:] = f_j # * 10^9
                F_mat[j,i,:] = -f_j # * 10^9
            end
        end
        # move the body according to the net force
        net_force = [sum(F_mat[i,:,1]),sum(F_mat[i,:,2])]
        new_stars[i] = moveStar(net_force,stars[i],time,spaceScale)
    end
    # return a list of all stars with updated pos & vel
    return new_stars
end
```

Complexity analysis annotations:

- $O(N)$  for the outer loop `for i in 1:length(stars)`.
- $O(N)$  for the inner loop `for j in i:length(stars)`.
- $O(1)$  for the innermost block (calculating force between two stars).
- The overall complexity is  $O(N^2)$ , indicated by a large bracket on the right.
- $O(N)$  for the net force calculation `net_force = [sum(F_mat[i,:,1]),sum(F_mat[i,:,2])]`.

Figure: Naive brute force approach

# Earth-Sun Simulation

Why is there a spaceScale?

$$m_{\text{sun}} \sim 1\text{e}30$$

$$m_{\text{earth}} \sim 1\text{e}24$$

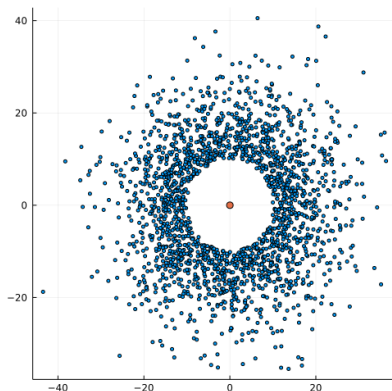
$$d_{e,s}^2 \sim (1\text{e}11)^2$$

$$G \sim 1\text{e}-11$$

$$\Rightarrow F_{e,s} \sim 1\text{e}21$$

Divide the masses by  $1\text{e}20$  and the distance by  $1\text{e}10$ , so that

$$F_{e,s}^{\text{scaled}} \sim 10^1$$



[Earth Sun Simulation](#)

B

# The Barnes Hut Algorithm

## MAIN IDEA:

Approximate the net force on a body in a *clever* way.

When a point is "**far enough**" from a cluster, the gravitational field generated by the cloud of points is *roughly* the same as if it were generated by a **single point** having as mass the sum of the masses and located in the **center of mass**<sup>1</sup>.

---

<sup>1</sup>The center of mass is the average of the coordinates weighted by the mass of the points,  $(x_{CM}, y_{CM}) = (\sum m_i)^{-1}(\sum m_i x_i; \sum m_i y_i)$

# The Barnes Hut Algorithm

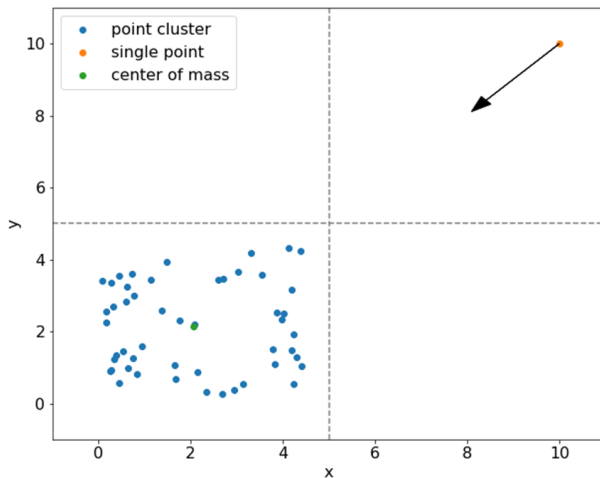


Figure: Approximating the net force with Barnes Hut

# The Barnes Hut Algorithm

Problems:

- ① How do I define a cluster of points?
- ② What classifies as "**far enough**"?



# The Barnes Hut Algorithm

Problem 1: *how to define a cluster of points*

- find a way to characterize the space of particles and organize their positions;
- optimize the necessary retrieval operations and computations.





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- find a way to characterize the space of particles and organize their positions;
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IDEA: divide the space using a tree structure called **Quadtree** (or Octree in 3D)



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A **Quadtree** is a “map” of space that helps us model groups of points as a single center of mass. Formally, it is a tree structure with the following properties:

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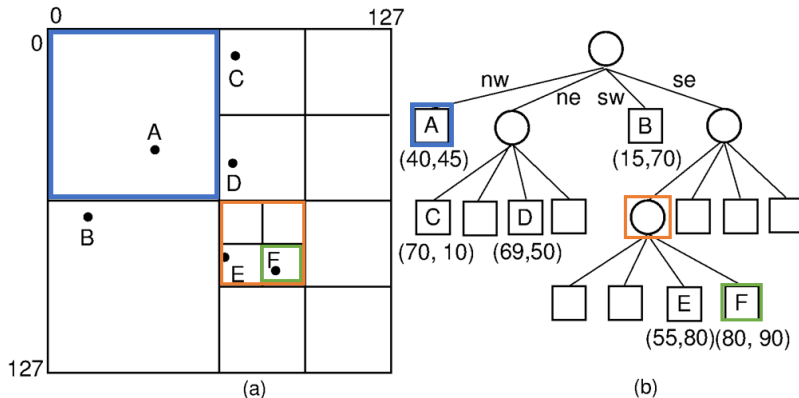
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- Each child node is recursively defined in the same way.



# The Quadtree

An example:<sup>2</sup>



**Figure:** To the left the space carved by the tree, and to the right the tree structure of the points in the space

<sup>2</sup>source: [website](#)

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- ③ At the end, return the root.



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Once we constructed the tree, we will only need to traverse it from the root to the leaves, in order to compute the force acting on a point.



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Once we constructed the tree, we will only need to traverse it from the root to the leaves, in order to compute the force acting on a point.

Hence, we can store the quadtree through Node2D objects which contain four pointers to the children nodes and also information about the quadrant.

```
mutable struct Node2D
  parent::Union{Nothing,Node2D}
  children::Dict{String,Union{Nothing,Node2D}}
  stars::Array{Star,1}
  regCenter::Array{Float64,1}
  x_lim::Array{Float64,1}
  y_lim::Array{Float64,1}
  n::Int64
  centerOfMass::Array{Float64,1}
  totalMass::Float64
  Node2D(parent,children,stars,regCenter,x_lim,
    y_lim,n,centerOfMass,totalMass) =
    new(parent,children,stars,regCenter,x_lim,
    y_lim,n,centerOfMass,totalMass)
end
```



# Quadtree Data Structure

Therefore, it is enough to keep track of the root of the tree, and from there we can explore all the other nodes.

This data structure has the following operation times:

	Insert	Get
Average	$O(\log N)$	$O(\log N)$
Worst	$O(N)$	$O(N)$

**Table:** Time complexity of operations

As for space complexity, the number of nodes in the tree is

$$\frac{4N - 1}{3} = O(N)$$



# Build Quadtree in Julia

```
function buildQTree(root::Union{Nothing,Node2D},stars::Union{Nothing,Array{Star,1}},
    x_lim::Array{Float64,1},y_lim::Array{Float64,1})
    if isnothing(root)
        root = createRoot(stars,x_lim,y_lim) ← O(1)
    end
    # Loop on every star
    for s in stars
        Q = getQuadrant(root.regCenter,s,s) ← O(1)
        if isnothing(root.children[Q])
            # If the quadrant is still empty, create the node
            center, x_lim, y_lim = getCenterLimits(root,Q)
            root.children[Q] = Node2D(root,center,x_lim,y_lim)
            root.children[Q] = updateNode!(root.children[Q],s) } O(1)
        else
            # Otherwise add the star to the node
            root.children[Q] = updateNode!(root.children[Q],s) ← O(1)
            if root.children[Q].n == 2
                # If there was already a star, then we need to go down a level and create other nodes
                stars_ = root.children[Q].stars
            else
                # otherwise we just need to sort the new star
                stars_ = [s]
            end
            # Then recursively call the constructor for that node and its corresponding quadrant
            root.children[Q] = buildQTree(root.children[Q], stars_,
                root.children[Q].x_lim, root.children[Q].y_lim) ← O(logN)
        end
    end
    # At the end it is enough to return the root, as it contains pointers to all its children
    return root
end
```

$O(N \log N)$





# The Multipole Acceptance Criterion

Problem 2: *What classifies as "far enough"?*

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## The Multipole Acceptance Criterion (MAC)

If the size of the cluster  $s$  divided by the distance of the cluster's CoM  $d$  from the point is lower than a parameter  $\theta > 0$ , then consider the cluster as a whole, i.e. approximate if:

$$\theta > \frac{s}{d}$$



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- If  $\theta$  is large instead, we gradually give up precision in the result to gain speed in the computation.

A usual good starting point is  $\theta = 1$ , which can be then adjusted according to the interest of the simulation.



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At the end of each iteration, we'll have  $\vec{F}_i^{net}$  and we are able to move the point accordingly.



# Computing the force with the Quadtree

```
function computeForceTree(node::Node2D,star::Star,theta::Float64)
    F_s = [0.,0.]
    s = abs(node.x_lim[2]-node.x_lim[1])
    d = distance(star.s,node.centerOfMass)
    if node.n == 1
        if star ∉ node.stars
            f = newton(star,node.stars[1])
            dir = node.stars[1]-star
            cosθ, sinθ = get_cos_sin(dir)
            F_s += [f * cosθ, f * sinθ] # * 10^9
        else
            F_s += [0.,0.]
        end
    elseif s/d < theta
        cm = Star(node.centerOfMass,[0.,0.],node.totalMass)
        f = newton(star,cm)
        dir = cm-star
        cosθ, sinθ = get_cos_sin(dir)
        F_s += [f * cosθ, f * sinθ] # * 10^9
    else
        for k in keys(node.children)
            if ! isnothing(node.children[k])
                F_s += computeForceTree(node.children[k],star,theta)
            end
        end
    end
    return F_s
end
```

$O(1)$

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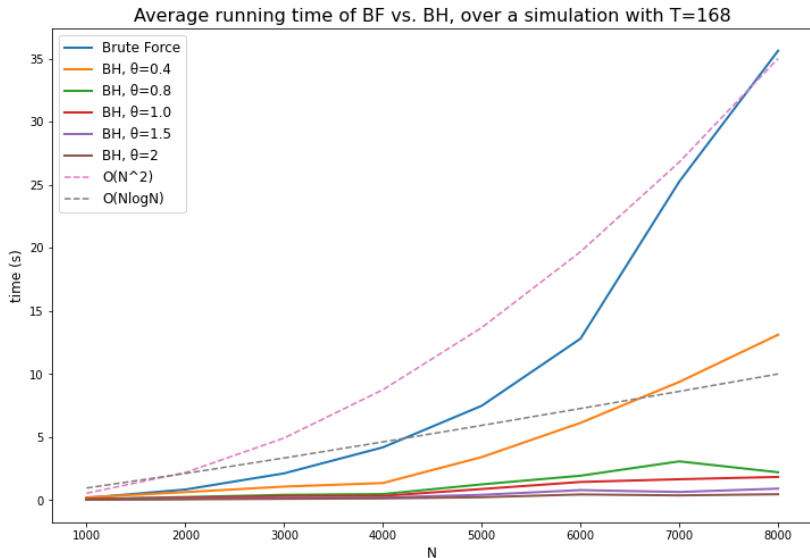
$O(\log N)$

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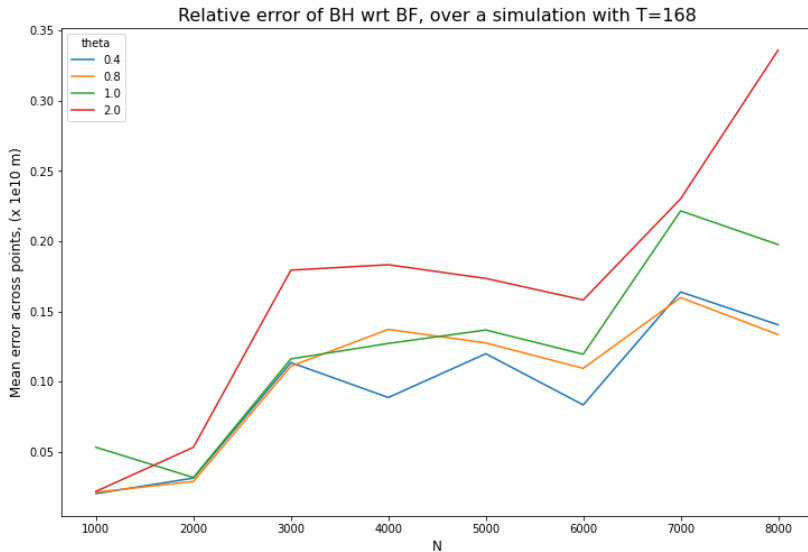
Galaxy clash simulation

B

# Simulations and Benchmarking



# Simulations and Benchmarking



- Barnes, J., Hut, P. A hierarchical  $O(N \log N)$  force-calculation algorithm. Nature 324, 446–449 (1986).  
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- [en.wikipedia.org/wiki/Barnes–Hut\\_simulation](https://en.wikipedia.org/wiki/Barnes–Hut_simulation)
- [jheer.github.io/barnes-hut](https://jheer.github.io/barnes-hut)
- [arborjs.org/docs/barnes-hut](https://arborjs.org/docs/barnes-hut)
- [beltoforion.de/en/barnes-hut-galaxy-simulator](https://beltoforion.de/en/barnes-hut-galaxy-simulator)