

# Parallel Implement of Convex Hull Problem

The background features a complex geometric pattern of blue lines and arrows. Solid blue lines intersect at various angles, creating a grid-like structure. Dashed blue lines form curved paths, some with arrows indicating direction. Small circles, some solid and some hollow, are scattered throughout the design, often positioned at the ends of lines or at points of intersection.

CSE 633 - Parallel Algorithms (Spring 2020)

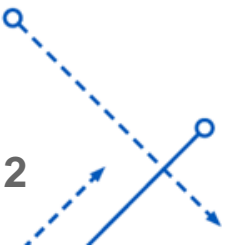
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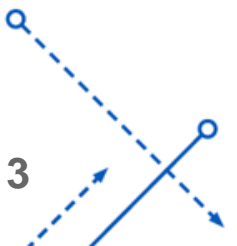
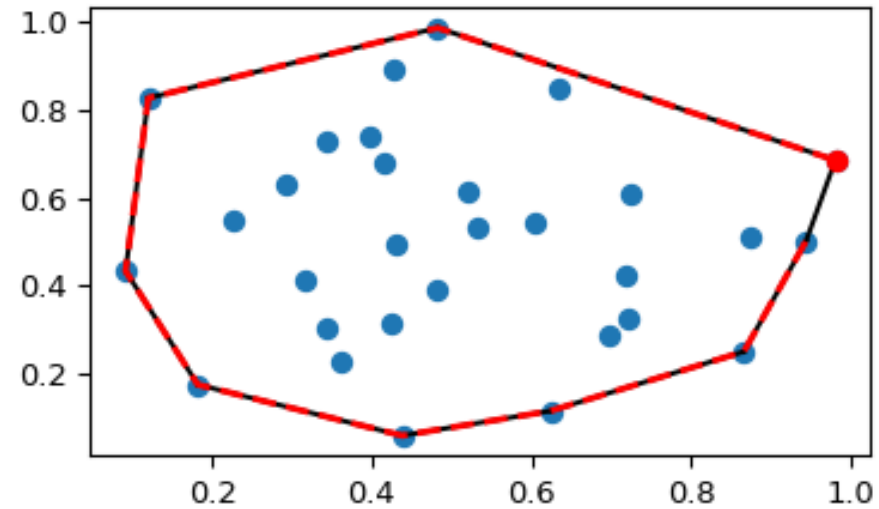
# Outline

- Convex Hull Definition
- Graham' Scan Aglorithm
- Combine Two Convex Hull
- Parallel Implementation
- Running Time Analysis
- Conclusion
- Future Work
- Reference



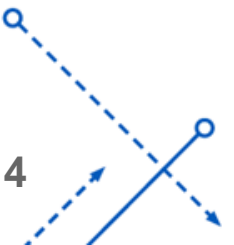
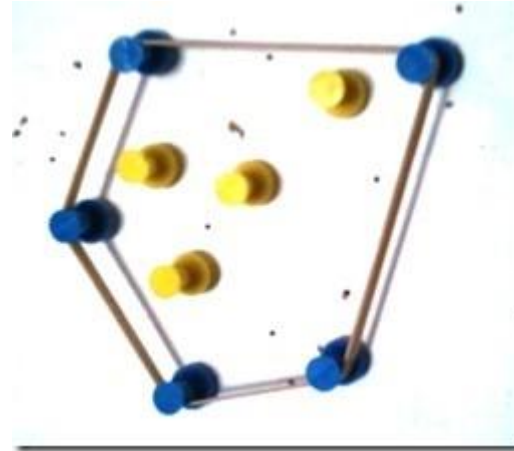
# Convex Hull Definition

Assume we have a set of points in the plane. The convex hull of set of points is defined as the smallest convex polygon, that encloses all the points in the set



# Convex Hull Example

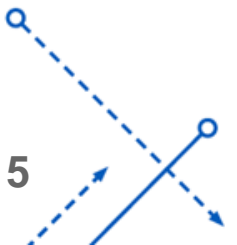
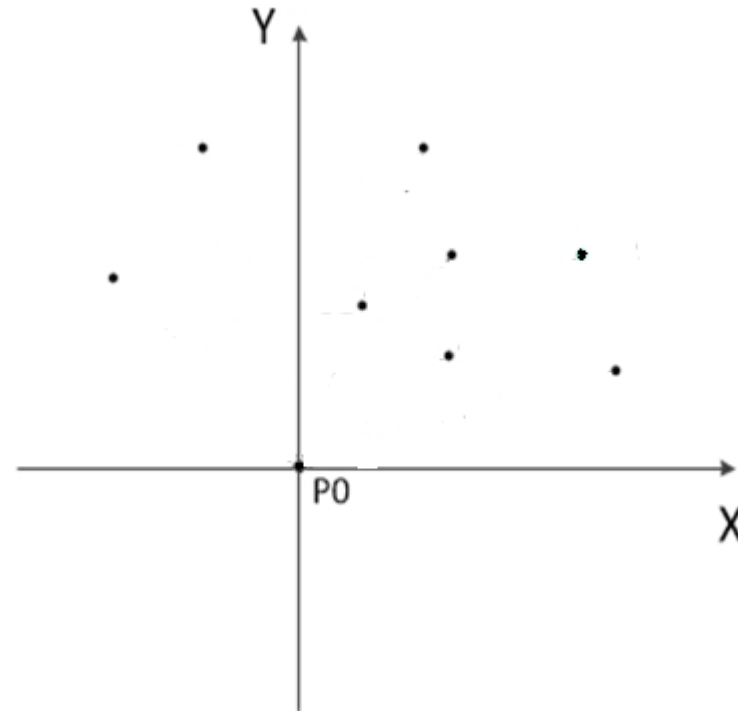
Imagine all the points as being pegs. The convex hull is the shape of rubber-band stretched around the pegs.



# Graham's Scan Algorithm

**First :**

Find a starting point that in the convex hull. (Lowermost point, if there is more than two lowermost points, then select leftmost one)



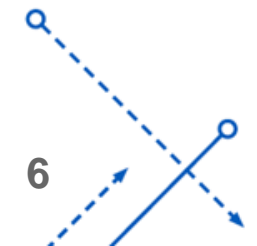
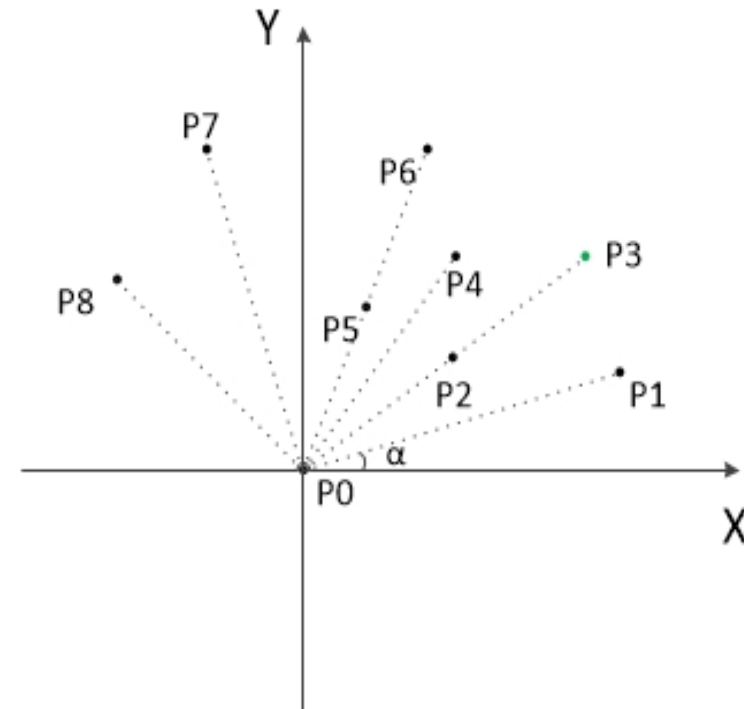
**Second :**

Compute remaining n - 1 points' angle with respect to the starting point.

Formula :

$$\phi = \cos^{-1} \frac{(x - x_0)}{\sqrt{(x - x_0)^2 + (y - y_0)^2}}$$

After Computing, Sort the points by angle with respect to the starting point. (If two points have same angle, then sort by the distance respect to the starting point from small to big)

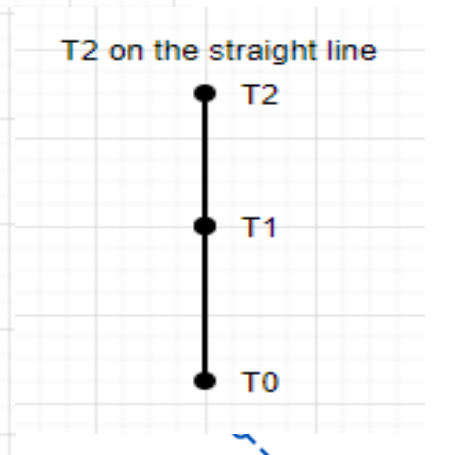
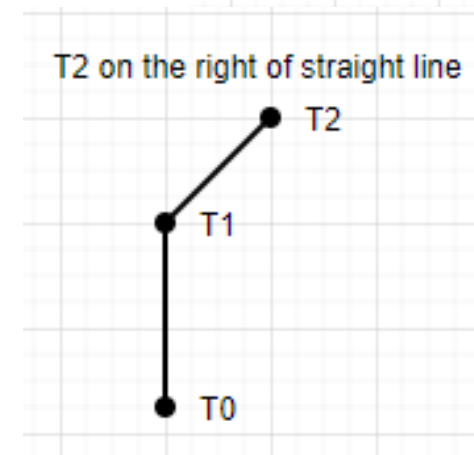
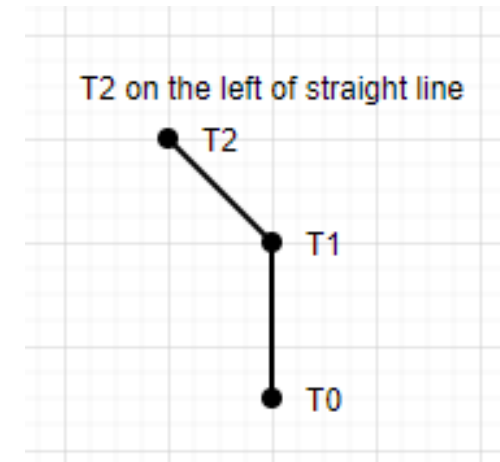


# Graham's Scan Algorithm Continue

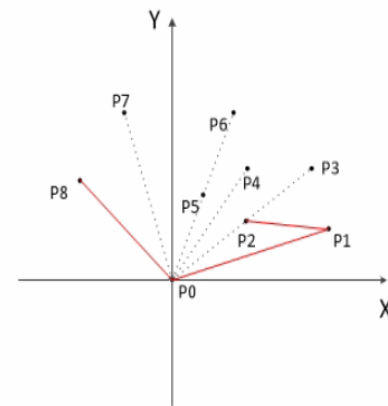
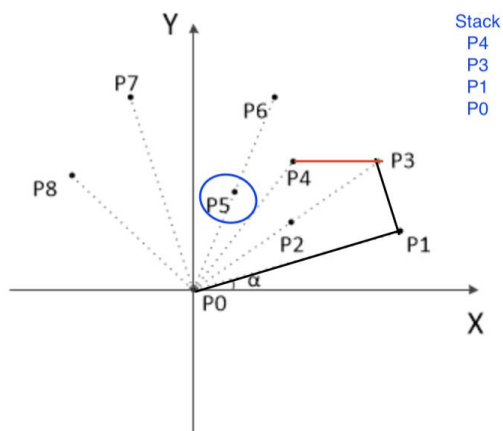
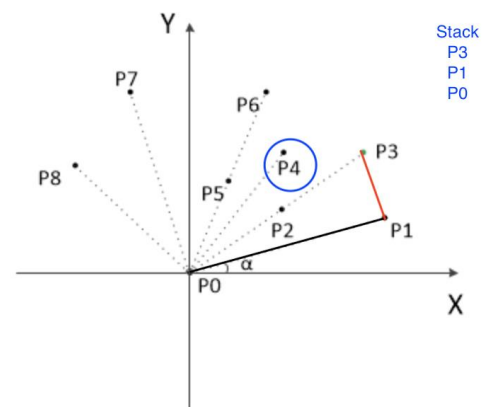
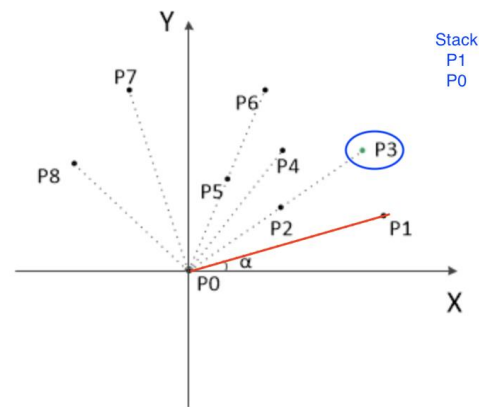
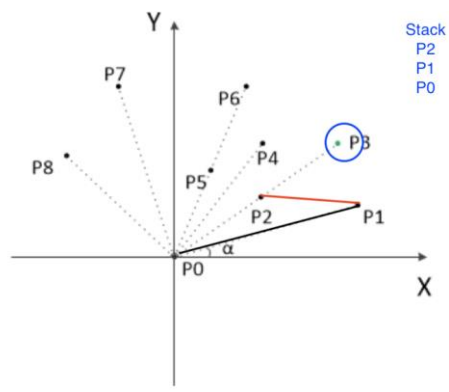
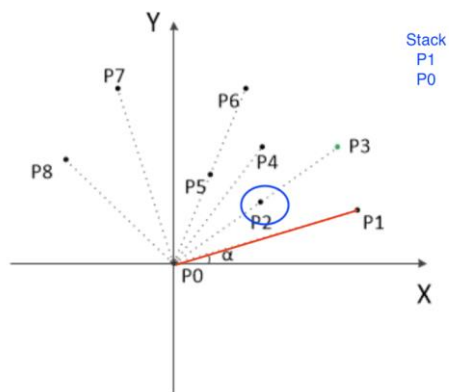
How to determine a point belongs to convex hull?

Scan:

1. Using stack to store, add starting point and second point to the stack.
2. Start Scan from third point:
  - a. Set T1 and T0 as top two points in the stack, and current scan point as T2
  - b. Connect point T1 and T0 as straight line and to determine T2 is located on left or right of straight line. (Cross Product)
  - c. If T2 on the left of straight line or on the straight line, then push the T2 into stack. and start next scan.
  - d. If T0, T1, T2 are collinear then it means T1 is not extreme point, so remove T1, and start next scan
  - e. If T2 on the right of straight line, then pop top point from the stack which is T1 and repeat from step a.
3. When we finish scanning, we will get a stack that contains all the points of convex hull in order.



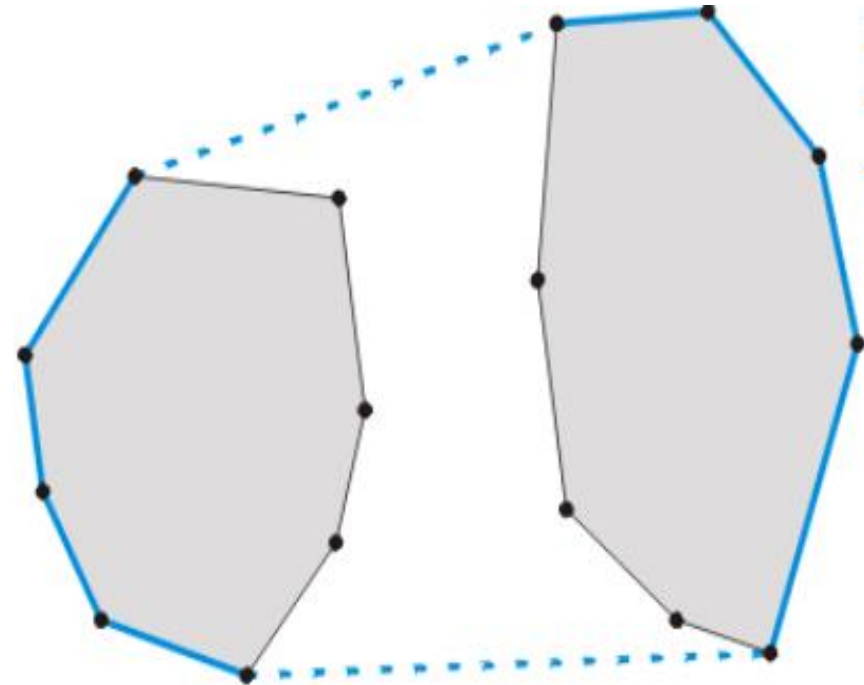
# Graham's Scan Algorithm Continue





# Merge Two Convex Hull Algorithm

The idea to merge two convex hull together is to find top and bottom tangent line, and connect them by the tangent line and remove all the points between two tangent lines.



# Merge Two Convex Hull Algorithm Continue

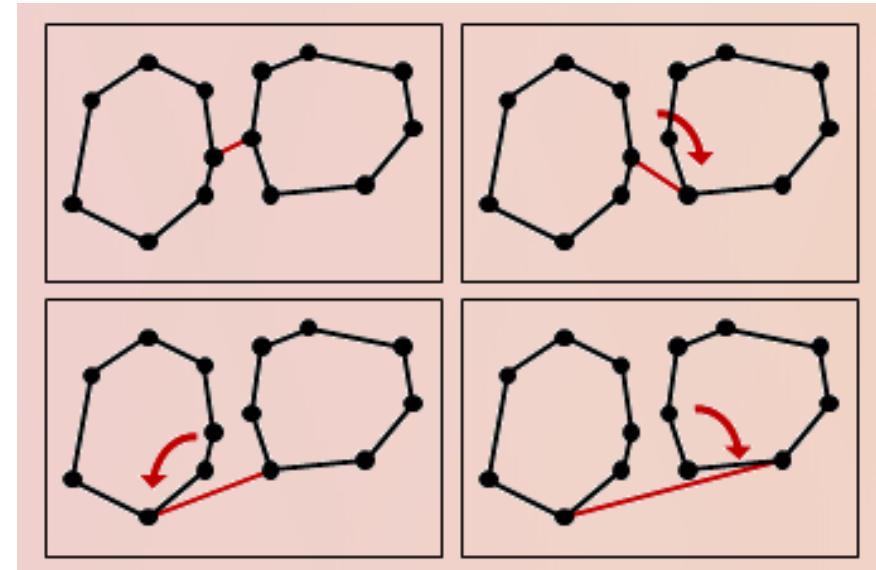
How to find the tangent line between two convex hull?

Bottom Tangent Line:

1. Find left convex hull's most right point P1, and right convex hull's most left point P2 and connect together.
2. Find the new P1 that is next point in the clockwise. And find new P2 that is next point in the counterclockwise.
3. Stop until it is fix.(Means if P1 or P2 go next, the connect line will cross to other line in the left or right convex hull)

Top Tangent Line:

Same to the Bottom tangent line, but P1 goes as counterclockwise and P2 goes as clockwise.



# Parallel Implementation - Tree Structure

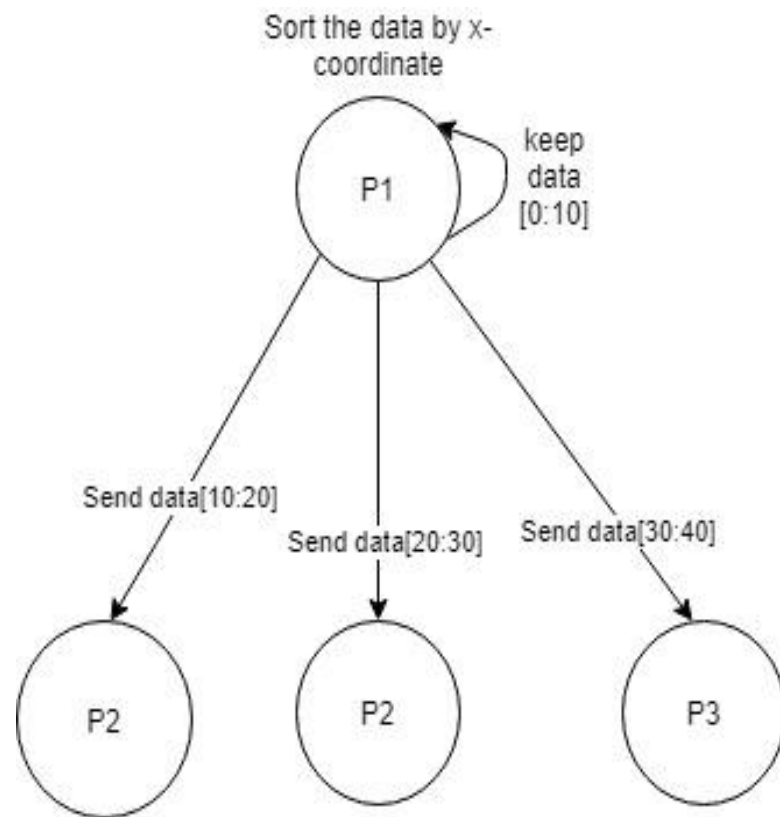
- Input a point data set with size  $n$  and  $p$  processors.
- Using master processor to sort the data by x-coordinate and separate the data to each processor equally.
- Each processor receive  $n/p$  size data from master processor, and parallelly find local convex hull by Graham's Scan.
- Recursively merge convex hull together between two processors by finding their top and bottom tangent line.



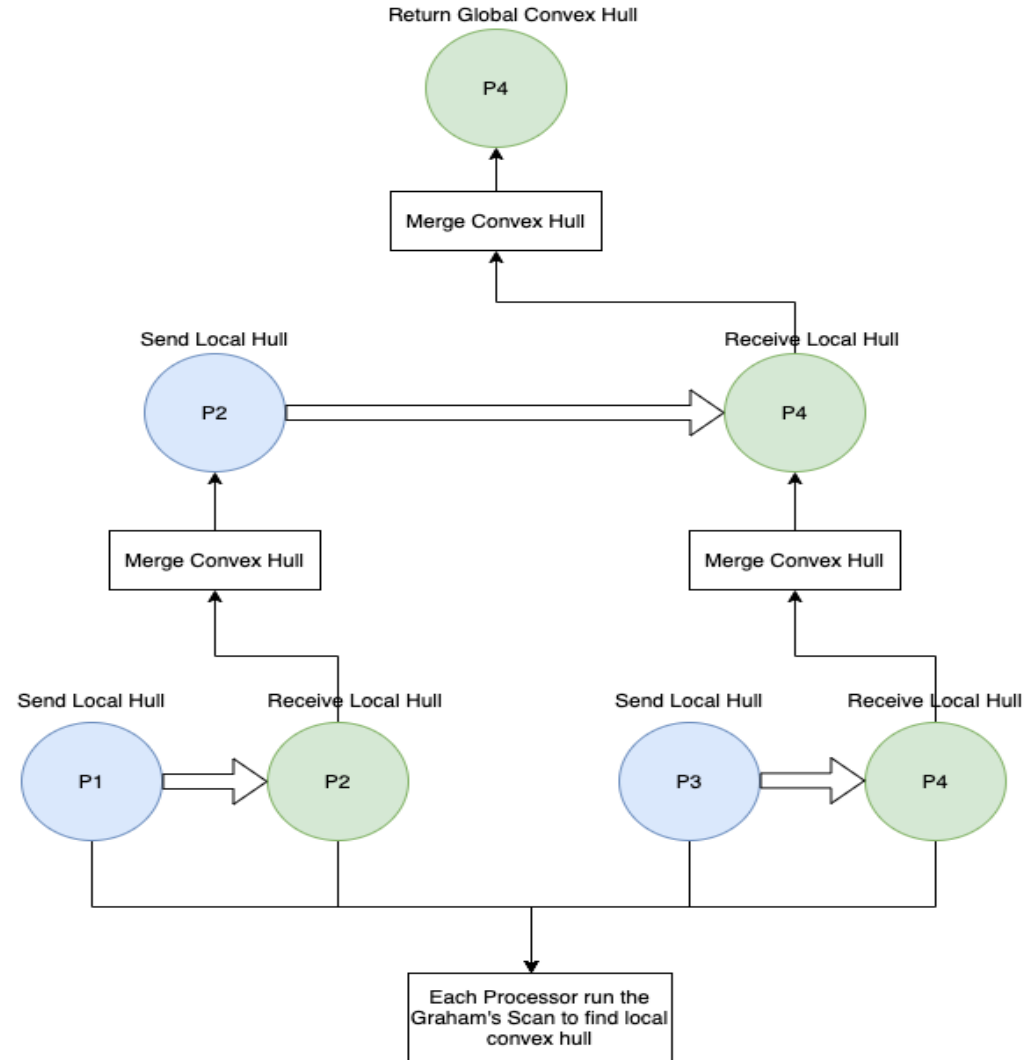
# PARALLEL EXAMPLE

As Example :

Data Size = 40  
# of Processor = 4



# PARALLEL EXAMPLE



# Parallel Implementation - Code

```
def treeBuildMPI(comm,data):
    size = comm.Get_size()
    rank = comm.Get_rank() + 1
    level = 1
    while(size>=(2**level)):
        if(rank%(2**level)==0):
            s = rank - (2**(level-1)) - 1
            get = comm.recv(source=s,tag=level)

            a,b = findBottomTangentLine(get,data)
            c,d = findTopTangentLine(get,data)
            data = combineTwoConvexHull(a,b,c,d,get,data)

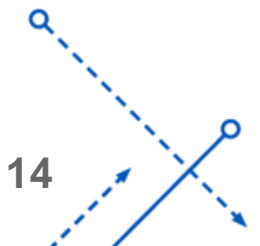
            print("Rank %d Receive Data From Rank %d, And Combine Convex Hull In Level %d"%(rank,s+1,level))

        elif (rank+(2**(level-1)))%(2**level)==0:
            dest = rank + (2**(level-1)) -1
            comm.send(data,dest,tag=level)
            print("Rank %d Send Data To Rank %d In Level : %d"%(rank,dest+1,level))
        else:
            print("Rank %d Out In Level %d....."%(rank,level))
    level += 1
```

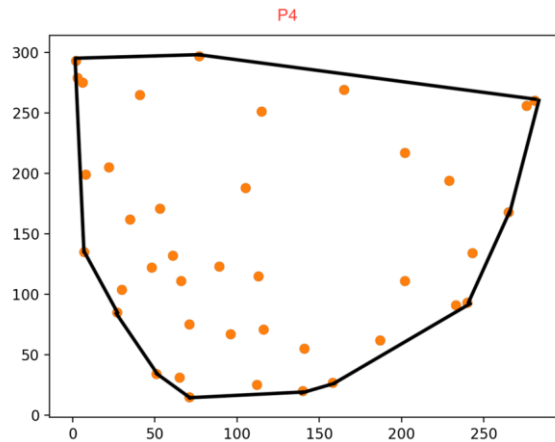
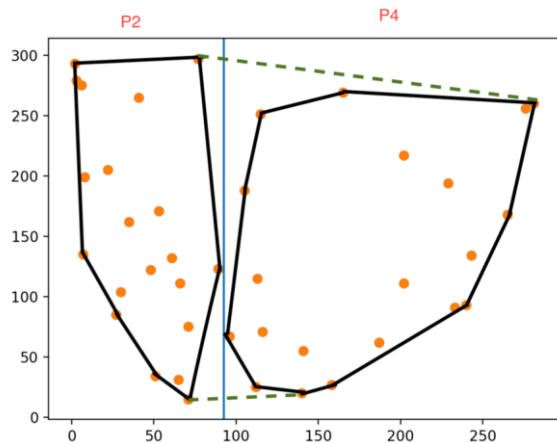
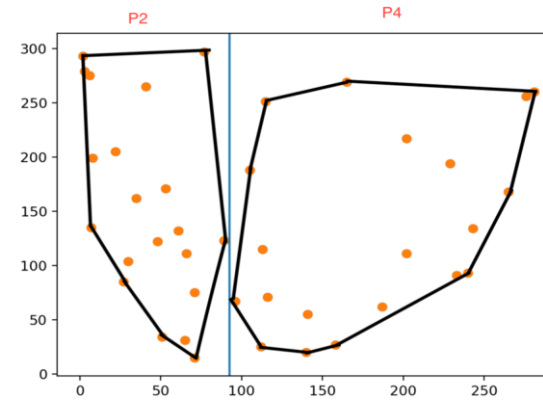
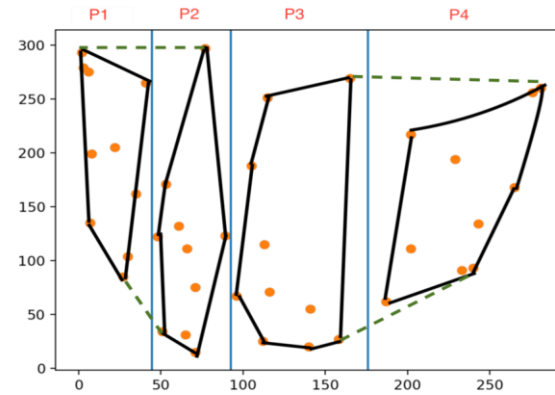
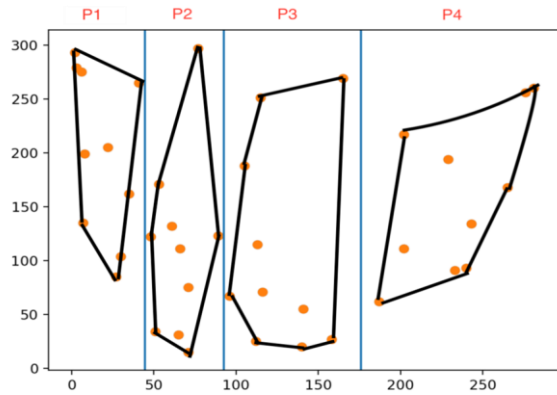
Receive Rank:  
 $R = \text{multiple of } 2^{\text{level}}$

distance between Send and Receive :  
 $D = 2^{(\text{level}-1)}$

Send Rank:  
 $S = R - D$

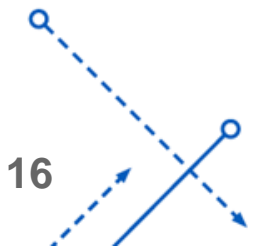


# Image Example



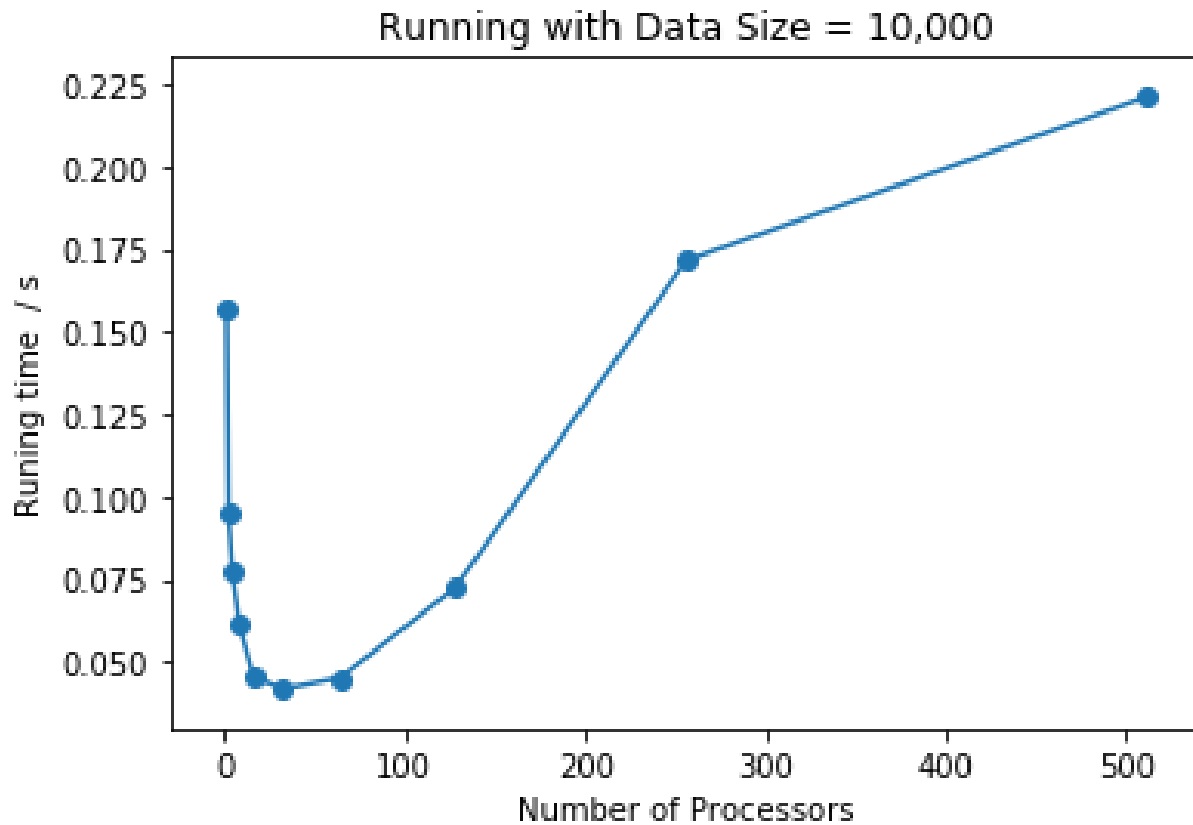
# Running Time Analysis

- Running Time vs. # processor with 10,000 points
- Running Time vs. # processor with 100,000 points
- Running Time vs. # processor with 1,000,000 points
- Running Time vs. # processor with 10,000,000 points
- Speedup Rate vs. # processor with four different size data



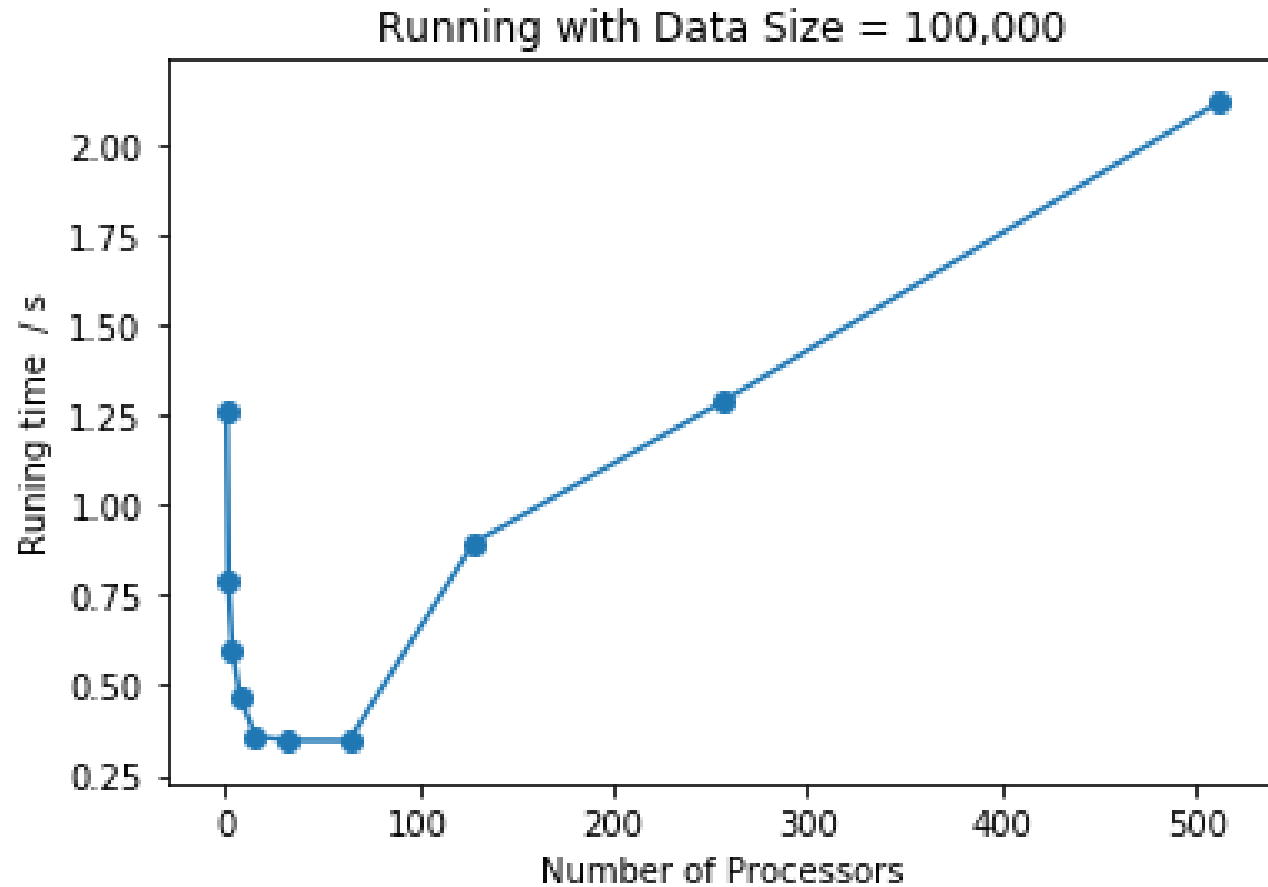


# Graph of Running Time with Data Size = 10,000



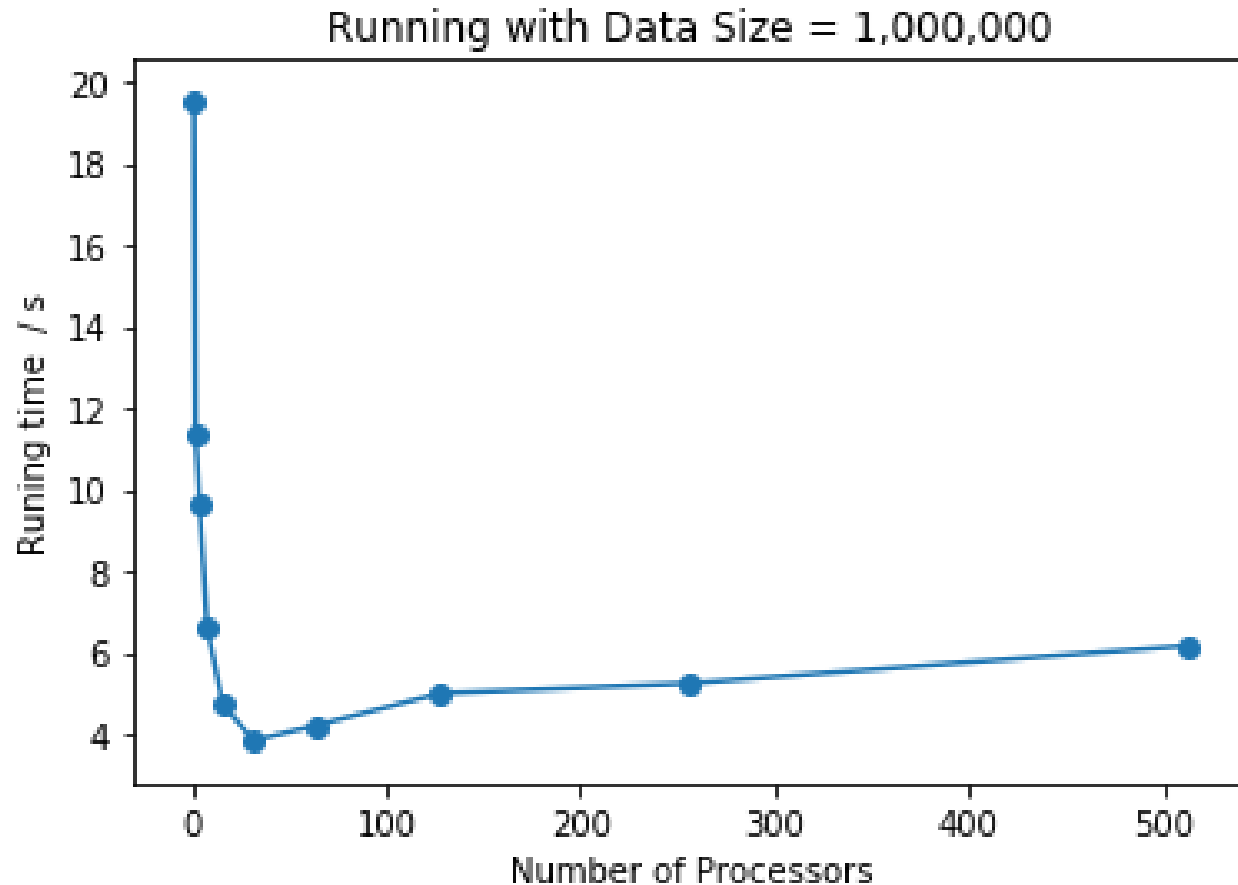
Processor #	Time(sec)
<b>1</b>	0.1565
<b>2</b>	0.0947
<b>4</b>	0.0775
<b>8</b>	0.0616
<b>16</b>	0.0456
<b>32</b>	0.0419
<b>64</b>	0.0452
<b>128</b>	0.0725
<b>256</b>	0.1715
<b>512</b>	0.2211

# Graph of Running Time with Data Size = 100,000

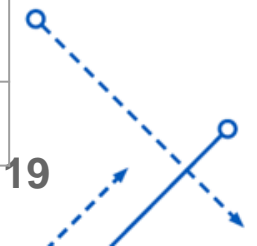


Processor #	Time(sec)
<b>1</b>	1.260
<b>2</b>	0.790
<b>4</b>	0.600
<b>8</b>	0.466
<b>16</b>	0.359
<b>32</b>	0.348
<b>64</b>	0.347
<b>128</b>	0.895
<b>256</b>	1.287
<b>512</b>	2.121

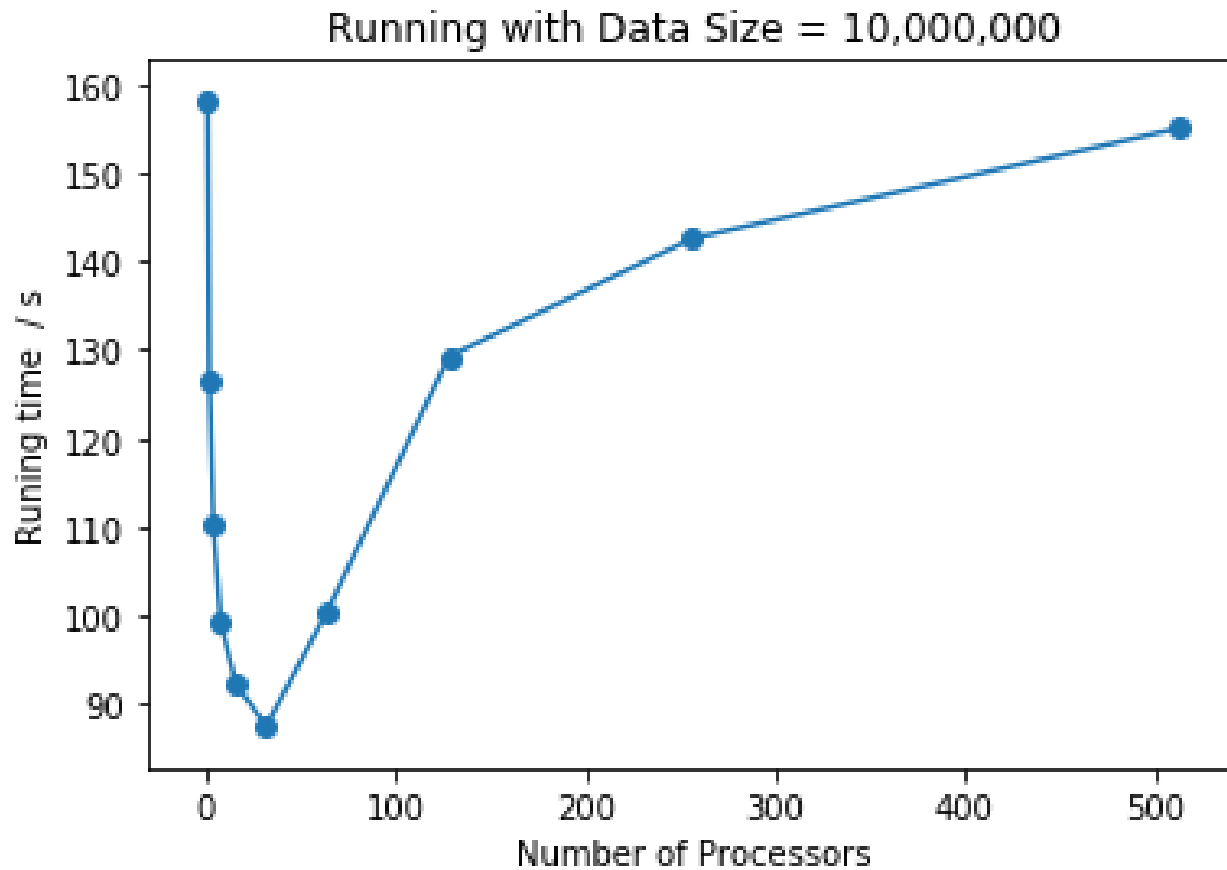
# Graph of Running Time with Data Size = 1,000,000



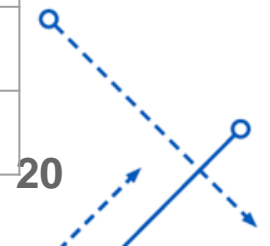
Processor #	Time(sec)
<b>1</b>	19.498
<b>2</b>	11.371
<b>4</b>	9.619
<b>8</b>	6.590
<b>16</b>	4.710
<b>32</b>	3.839
<b>64</b>	4.205
<b>128</b>	4.992
<b>256</b>	5.234
<b>512</b>	6.155



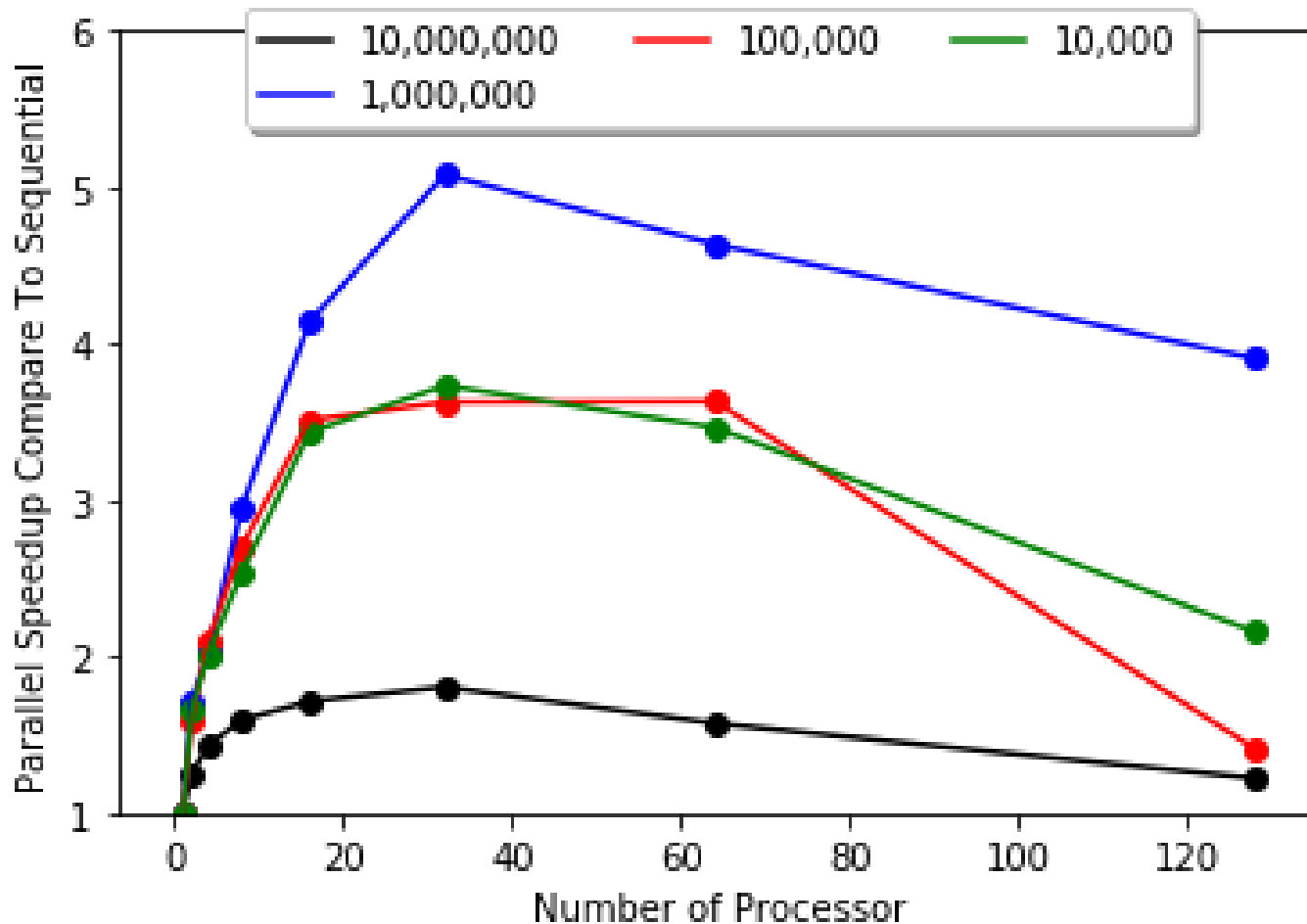
# Graph of Running Time with Data Size = 10,000,000



Processor #	Time(sec)
<b>1</b>	158.06
<b>2</b>	126.20
<b>4</b>	110.33
<b>8</b>	99.12
<b>16</b>	92.24
<b>32</b>	87.52
<b>64</b>	100.53
<b>128</b>	129.25
<b>256</b>	142.58
<b>512</b>	155.02



# Speedup



# Conclusion

- As observation, more processors does not mean better running time performance.
  - Running Time decreasing when we adding processor before 32 processor.
  - Running Time increasing when continue adding processor after 32 processor.
- Parallel algorithm can save time a lot. When we have 10,000,000 size data. Although the best speedup is about 180% compare to sequential running time, it save about 70 sec in real time.

## Future Work

For now, I implemented the Tree Structure Parallel Algorithm which only accept the number of processor is equal to power of 2. And in the future I will implement the Mesh Structure Parallel Algorithm to accept more different number of processor.



# Reference

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- Russ Miller and Quentin F. Stout : Efficient Parallel Convex Hull Algorithm ; IEEE Transaction on Computers  
vol. 37 no. 12; December 1988.
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<https://medium.com/@pascal.sommer.ch/a-gentle-introduction-to-the-convex-hull-problem-62dfcabee90c>
- MPI For Python : <https://mpi4py.readthedocs.io/en/stable/tutorial.html>
- CCR Tutorial : <https://ubccr.freshdesk.com/support/solutions>



Thank You!

