1.

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from which pyqt import PYQT VER
if PYOT VER == 'PYOT5':
 from PyQt5.QtCore import QLineF, QPointF, QThread, pyqtSignal
elif PYQT VER == 'PYQT4':
  from PyQt4.QtCore import QLineF, QPointF, QThread, pyqtSignal
  raise Exception('Unsupported Version of PyQt: {}'.format(PYQT VER))
# Using the left hull find its highest x value
def find rightmost(leftside hull):
  assert(type(leftside hull) == list and type(leftside hull[0] == QPointF))
  for i in range(len(leftside hull)):
    if ((i + 1) >= len(leftside hull)): # If adding one to the index makes it bigger or equal to the length we already
    if (leftside hull[i + 1].x() < leftside hull[i].x()): # If the next x value is lower than the current x value then we
      return i
 def find slope(point1, point2):
  assert(type(point1) == QPointF and type(point2) == QPointF)
  return ((point2.y() - point1.y()) / (point2.x() - point1.x()))
checked
 # n is the size of the hull being passed in (number of points)
 † The space complexity is O(1) because it is a constant stack being used,
def find tangent(leftside hull, rightside hull, lefthull index, righthull index):
  slope = find slope(leftside hull[lefthull index], rightside hull[righthull index])
  update = True
  while (update):
    update = False
    left decreasing = True
    while (left decreasing):
       newleftpt index = lefthull index - 1
       if (newleftpt_index < 0): # If the index becomes negative wrap around to end of list
         newleftpt index = len(leftside hull) - 1
       new slope = find slope(leftside hull[newleftpt index], rightside hull[righthull index])
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if (new slope > slope): # Slope needs to decrease in the left hull
         left decreasing = False
         update = True
         lefthull index = newleftpt index
         slope = new slope
    right increasing = True
     while (right_increasing):
       newrightpt_index = righthull_index + 1
       if (newrightpt_index == len(rightside_hull)): # Wrap to beginning of list if index is equal to length
         newrightpt index = 0
       new slope = find slope(leftside hull[lefthull index], rightside hull[newrightpt index])
         right increasing = False
         update = True
         righthull index = newrightpt index
  return lefthull index, righthull index
Combines the two arrays of QPointF points into one hull. The array returned is still in clockwise order
 Time complexity here is O(n) because it calls find tangent and find rightmost which both of them are O(n)
def combine hulls(leftside hull, rightside hull):
  lefthull index = find rightmost(leftside hull) # The right most can be anywhere in the array but it can be found by
  righthull_index = 0 # The right's left most point will always be the first index because of the clockwise order of how
  leftpointidx uppertangent, rightpointidx uppertangent = find tangent(leftside hull, rightside hull, lefthull index,
righthull index)
 rightpointidx lowertangent, leftpointidx lowertangent = find tangent(rightside hull, leftside hull, righthull index,
lefthull index)
  merged hull = leftside hull[0:leftpointidx uppertangent + 1] # Grab the first to the leftpoint
  merged hull += (rightside hull rightpointidx uppertangent:rightpointidx lowertangent + 1]) # Grab the points from
  if (rightpointidx lowertangent == 0):
    merged hull += rightside hull[rightpointidx uppertangent:] # Grab all of the points since 0 is the start of the right
    merged hull.append(rightside hull[0])
  if not (leftpointidx lowertangent == 0): # Grab all of the rest of the points in the left hull
    merged hull += (leftside hull[leftpointidx lowertangent:])
  return merged hull
 Time complexity is O(nlogn) where n is the number of sorted points
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def make convex(sorted points):
  if (len(sorted points) <= 3): # Base case
    # If the length is 1 or 2 the array will already be in clockwise order since it doesn't matter which
    if (len(sorted points) == 3): # Big O is constant time here to look up and switch
       slope = find slope(sorted points[0], sorted points[1])
       next_slope = find_slope(sorted_points[0], sorted_points[2])
       if (next_slope > slope): # To go in clockwise order the slopes should be decreasing so need to swap points here
         point holder = sorted points[2]
         sorted_points[2] = sorted_points[1]
         sorted_points[1] = point_holder
       elif (sorted points[1].y() == sorted points[2].y()):
         if (sorted points[0].y() > sorted points[1].y()):
           point holder = sorted points[2]
           sorted points[2] = sorted points[1]
           sorted_points[1] = point_holder
    return sorted points
  middle = math.floor(len(sorted points) / 2)
  leftside hull = make convex(sorted points[:middle]) # Takes points from 0 to middle - 1
  rightside hull = make convex(sorted points[middle:]) # Takes points from middle to end
  combined hull = combine hulls(leftside hull, rightside hull)
  return combined hull
class ConvexHullSolverThread(QThread):
 def __init__ (self, unsorted points, demo):
    self.points = unsorted points
    self.pause = demo
    QThread. init (self)
  show_hull = pyqtSignal(list, tuple)
  display text = pyqtSignal(str)
  show tangent = pyqtSignal(list, tuple)
  erase hull = pyqtSignal(list)
  erase tangent = pyqtSignal(list)
    assert (type(self.points) == list and type(self.points[0]) == QPointF)
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sorted points = sorted(self.points, key=lambda p: p.x()) # Sorts the points by x value
    t2 = time.time()
    print('Time Elapsed (Sorting): {:3.3f} sec'.format(t2 - t1))
    t3 = time.time()
    convex hull = make convex(sorted points) # Makes an array of the convex hull points
    USE DUMMY = False
    if USE DUMMY:
       polygon = [QLineF(self.points[i], self.points[(i + 1) % 3]) for i in range(3)]
       assert (type(polygon) == list and type(polygon[0]) == QLineF)
       self.show hull.emit(polygon, (255, 0, 0))
       # Store the lines of the hull in convex lines array to be sent to the GUI to draw them
the first point
       convex\_lines = [QLineF(convex\_hull[i], convex\_hull[(i + 1) \% len(convex\_hull))]) for i in
range(len(convex hull))]
       assert (type(convex lines) == list and type(convex lines[0]) == QLineF)
       self.show hull.emit(convex lines, (255, 0, 0))
    self.display text.emit('Time Elapsed (Convex Hull): {:3.3f} sec'.format(t4 - t3))
    print('Time Elapsed (Convex Hull): {:3.3f} sec'.format(t4 - t3))
```

Time and Space Complexity

Find rightmost function

This function takes in the left hull that is a list of n points. It iterates over the list looking for which point has the highest x-value. Therefore its time complexity is O(n) where n is the size of the leftside_hull list of points. It is just iterating over the given list and doesn't store anything new. Thus, its space complexity is constant O(1).

<u>Find_slope function</u>

This function takes in two points to figure out what the slope is. Time complexity here is constant O(1) this is because the slope is found by subtracting two coordinates which has nothing to do with the number of points taken in by the program. The number of points is what causes this program to take a certain amount of time not subtracting bits. The space complexity is O(1) as well because we aren't storing anything new here just returning a value.

Find tangent function

This function takes in two hulls, the left and right that are both a list of n points, and the indices of the rightmost point in the left hull list and the leftmost point in right hull list. Its time complexity is O(n) because the worst case is that it has to move through all the list of points till the correct tangent is found. N is the size of the hull list. The space complexity is O(1) since it is a constant stack being used. In other words there are no recursive calls that are increasing the memory being used.

Combine hulls function

This function takes in two hulls, the left and right which are both a list of n points. The time complexity is O(n) because it calls find_tangent and find_rightmost which are both O(n) (see above). The space complexity is O(n) where n is the size of the new hull created from merging the left and right hulls together.

Make convex function

This function takes in the list of sorted_points. It recursively calls itself to keep dividing them by the midpoint of the list. Since it does this the time complexity is O(nlogn) where n is the size of the sorted_points list. The space complexity is O(n) where n is the number of points being stored in the left or right hull with each recursive call.

ConvexHullSolverThread

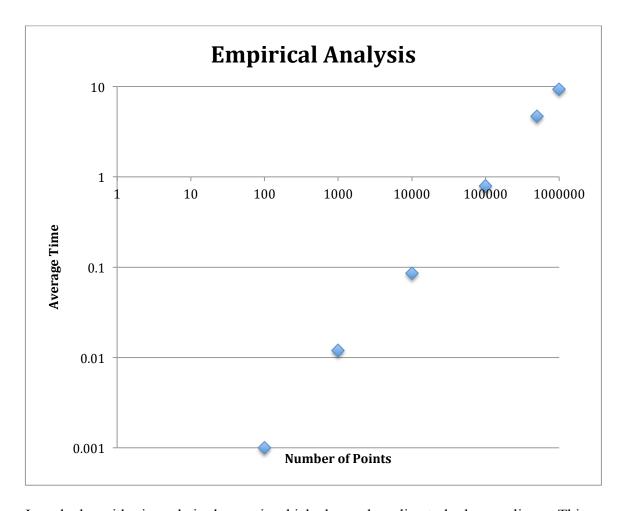
This is the main function that calls everything else. It takes in a QThread that contains all of the points n. Thus the overall time complexity is O(nlogn) this comes from the sorted function that sorts the points as well as the call to make_convex where n is the number of points passed in. The overall space complexity is O(n) where n is the number of points passed in. This is because the points are stored in sorted_points by their x value and there are n of them.

Theoretical Analysis with Recurrence Relation

Worst-case time efficiency is O(nlogn) this is cause of the divide and conquer algorithm used. This theoretical time efficiency comes from the Master Theorem/recurrence relation which is: $T(n) = aT(n / b) + O(n^d)$. The a stands for the number of subproblems. The n is the size of the original problem. The b is how splits are made. And the d is the work at each level. From my algorithm I will get a = 2 (a left and right hull is made at each level) b = 2 (splitting the n in half by cutting the list in half each time) and d = 1 (putting the hulls together will be O(n) because it will require iterating through both n sized hulls). Therefore my master theorem is T(n) = 2T(n / 2) + O(n). log2(2) = 1 since logb(a) = d by the master theorem we can tell that the time complexity will be O(nlogn)

3. Empirical Analysis

N points	10	100	1,000	10,000	100,000	500,000	1,000,000
Test 1	0	0.001	0.012	0.085	0.786	4.7	9.333
Test 2	0	0.001	0.012	0.089	0.806	4.644	9.345
Test 3	0	0.001	0.012	0.084	0.786	4.653	9.401
Test 4	0	0.001	0.012	0.084	0.782	4.666	9.392
Test 5	0	0.001	0.012	0.084	0.782	4.738	9.332
Average	0	0.001	0.012	0.0852	0.7884	4.6802	9.3606

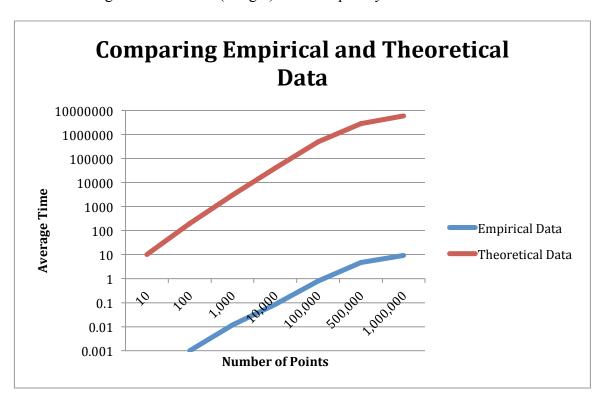


I used a logarithmic scale in the y axis which changed my line to look more linear. This happens because I am normalizing the data for a polynomial space. Normalizing the data helps see the points at 100, and 1000 because if I kept it as a normal graph these points would all look like they don't move at all and stay at zero.

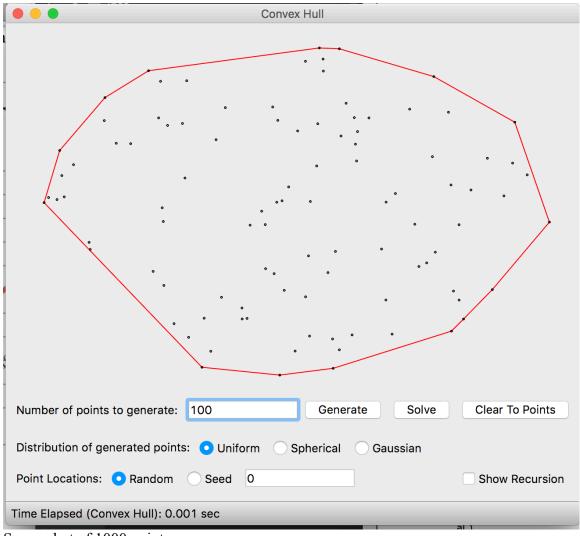
My plot looks very close to a n log n plot as can be seen when graphing both my numbers and what n log n gives (see below). Thus the order of grow that fits best is n log n. As can be seen the graphs differ by a constant. This constant can be found as so. When n is 10 n log n equals 10. The time it took for 10 points was 0. So, 10x = 0 the constant here would be zero. For 100 n log n is 200x = 0.001. The constant here is $5x10^{-6}$. For 1000 n log n is 3000x = 0.012. The constant here is $4x10^{-6}$. For 10000 n log n is 40000x = 0.0852. The constant here is $2.13x10^{-6}$. For 100000 n log n is 500000x = 0.7884. The constant here is $1.58x10^{-6}$. For 500000 n log n is 2849485x = 4.6802. The constant here is $1.64x10^{-6}$. For 1000000 n log n is 6000000x = 9.3606. The constant here is $1.56x10^{-6}$. Thus, as more points are calculated for the convex hull the proportionality constant rarely changes. In fact for all of the points it hardly changes at all since all of them are 10^{-6} which is a very small number. Thus I will take the average of them all to calculate my proportionality constant. This comes out to be $2.27x10^{-6}$.

4.

There were no differences seen from my theoretical and empirical analyses. It makes sense that the plot would look n log n because through the master theorem I deduced that in theory it should be. Then through my empirical analysis I was able to plot and compare my data to what an n log n graph would look like. They matched up perfectly and only differed slightly by a very small constant between the two lines. Therefore through my observations of both my theoretical and empirical analyses I can firmly say that my convex hull algorithm runs in O(n log n) time complexity.



5. Screenshot of 100 points



Screenshot of 1000 points

