## Hierarchical Clustering

## September 19, 2021

[1]: import numpy as np

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1.1.1

return self.left\_child

def set\_root\_data(self, data):

self.root\_data = data

def get\_root\_data(self):

```
import pandas as pd
     import math
     import matplotlib.pyplot as plt
[2]: class BinaryTree:
        111
         def __init__(self, root_data=None, left_child=None, right_child=None):
             111
             111
             self.root_data = root_data
             self.left_child = left_child
             self.right_child = right_child
         def get_right_child(self):
             111
             111
             return self.right_child
         def get_left_child(self):
```

```
return self.root_data
def is_empty(self):
    I I I
    111
    return self.root_data == None
def insert_left(self, new_data):
    111
    if self.left_child == None:
        self.left_child = BinaryTree(new_data)
    else:
        t = BinaryTree(new_data)
        t.left_child = self.left_child
        self.left_child = t
def insert_right(self, new_data):
    111
    111
    if self.right_child == None:
        self.right_child = BinaryTree(new_data)
    else:
        t = BinaryTree(new_data)
        t.right_child = self.right_child
        self.right_child = t
def pre_order_traversal(self):
    I I I
    111
    if not self.is_empty():
        self.pre_order_helper(self)
        print()
    else:
        print("Empty tree")
def pre_order_helper(self, tree):
    111
    111
    if tree is not None:
        print(tree.root_data, end=" ")
```

```
self.pre_order_helper(tree.left_child)
        self.pre_order_helper(tree.right_child)
def post_order_traversal(self):
    111
    if not self.is_empty():
        self.post_order_helper(self)
    else:
        print("Empty tree")
def post_order_helper(self, tree):
    111
    , , ,
    if tree is not None:
        self.post_order_helper(tree.left_child)
        self.post_order_helper(tree.right_child)
        print(tree.root_data, end=" ")
def level_order_traversal(self):
    if not self.is_empty():
        queue = [self.root_data]
        self.level_order_helper(self, queue)
        for data in queue:
            print(data, end=" ")
        print()
    else:
        print("Empty tree")
def level_order_helper(self, tree, queue):
    111
    111
    if tree is not None:
        if tree.left child is not None:
            queue.append(tree.left_child.root_data)
        if tree.right_child is not None:
            queue.append(tree.right_child.root_data)
        self.level_order_helper(tree.left_child, queue)
        self.level_order_helper(tree.right_child, queue)
```

```
def in_order_traversal(self):
    '''

if not self.is_empty():
        self.in_order_helper(self)
        print()

else:
        print("Empty tree")

def in_order_helper(self, tree):
    '''

if tree is not None:
        self.in_order_helper(tree.left_child)
        print(tree.root_data, end="")
        self.in_order_helper(tree.right_child)
```

```
[3]: class HierachicalCluster(BinaryTree):
         def __init__(self, name, root=None, left_child=None, right_child=None,
      →distance=0):
             self.name = name
             self.root = root
             self.left_child = left_child
             self.right_child = right_child
             self.distance = distance
         def __getitem__(self, index):
             return self.root[index]
         def len (self):
             return len(self.root)
         def compute_centroid(self):
             centroid = []
             if self.left_child is None and self.right_child is None:
                 return self.root
             else:
                 rc = self.right_child.get_leaf_count()
                 lc = self.left_child.get_leaf_count()
                 for i in range(len(self.right_child)):
                     val = ((self.right_child[i]*rc + self.left_child[i]*lc) /__
      \hookrightarrow (rc+lc))
                     centroid.append(val)
             self.root = centroid
```

```
return self.root
   def get_leaf_count(self):
       if self.left_child == None:
           return 1
       else:
           return self.left_child.get_leaf_count() + self.right_child.
→get_leaf_count()
   def set_right_child(self, r_child):
       self.right_child = r_child
   def set_left_child(self, l_child):
       self.left_child = l_child
   def is_leaf(self):
       return self.left_child == None and self.right_child == None
   def get_root(self):
       return self.root
   def get_distance(self):
       return self.distance
   def get_display_name(self):
       if self.is_leaf():
           return self.name
       else:
           return str(round(self.distance, 2))
   def print(self):
       thislevel = [self]
       while thislevel:
           nextlevel = []
           for n in thislevel:
               print(n.get_display_name()+' ', end='')
               if n.left_child:
                   nextlevel.append(n.left_child)
               if n.right_child:
                   nextlevel.append(n.right_child)
           print('')
           thislevel = nextlevel
   def get_all_leaves(self):
       leaves = []
       self.get_all_leaves_rc(self, leaves)
```

```
return leaves

def get_all_leaves_rc(self, hc, leaves):
    if hc is None:
        return
    hc.get_all_leaves_rc(hc.left_child, leaves)
    hc.get_all_leaves_rc(hc.right_child, leaves)
    if hc.left_child is None and hc.right_child is None:
        leaves.append(hc)
```

```
[4]: class Pair:
         def __init__(self, HC_member1, HC_member2):
             self.HCmember1 = HC_member1
             self.HCmember2 = HC_member2
             self.distance = self.compute_distance()
         def __lt__(self, p2):
             return self.distance < p2.get_distance()</pre>
         def compute_distance(self):
             x = 0
             for i in range(len(self.HCmember1)):
                 x += (self.HCmember1[i] - self.HCmember2[i])**2
             return math.sqrt(x)
         def get_distance(self):
             return self.distance
         def get_hc_member1(self):
             return self.HCmember1
         def get_hc_member2(self):
             return self.HCmember2
```

```
[5]: class BinaryHeap:
    def __init__(self):
        '''
        heap_list[0] = 0 is a dummy value (not used)
        '''
        self.heap_list = [0]
        self.size = 0

def __str__(self):
        return str(self.heap_list)

def __len__(self):
```

```
return self.size
   def __contains__(self, item):
       return item in self.heap_list
   def is_empty(self):
       return self.size == 0
   def find min(self):
       the smallest item is at the root node (index 1)
       if self.size > 0:
           min_val = self.heap_list[1]
           return min_val
       return None
   def insert(self, item):
       append the item to the end of the list (maintains complete tree_
\hookrightarrow property)
       violates the heap order property
       call percolate up to move the new item up to restore the heap order
\hookrightarrow property
       111
       self.heap_list.append(item)
       self.size += 1
       self.percolate_up(self.size)
   def del_min(self):
       111
       min item in the tree is at the root
       replace the root with the last item in the list (maintains complete \sqcup
violates the heap order property
       call percolate down to move the new root down to restore the heap_{\sqcup}
\hookrightarrow property
       min_val = self.heap_list[1]
       self.heap_list[1] = self.heap_list[self.size]
       self.size = self.size - 1
       self.heap_list.pop()
       self.percolate_down(1)
       return min_val
   def min_child(self, index):
```

```
return the index of the smallest child
       if there is no right child, return the left child
       if there are two children, return the smallest of the two
       if index * 2 + 1 > self.size:
           return index * 2
       else:
           if self.heap_list[index * 2] < self.heap_list[index * 2 + 1]:</pre>
               return index * 2
           else:
               return index *2 + 1
   def build heap(self, alist):
       111
       build a heap from a list of keys to establish complete tree property
       starting with the first non leaf node
       percolate each node down to establish heap order property
       index = len(alist) // 2 # any nodes past the half way point are leaves
       self.size = len(alist)
       self.heap_list = [0] + alist[:]
       while (index > 0):
           self.percolate_down(index)
           index -= 1
   def percolate_up(self, index):
       compare the item at index with its parent
       if the item is less than its parent, swap!
       continue comparing until we hit the top of tree
       (can stop once an item is swapped into a position where it is greater_
\hookrightarrow than its parent)
       111
       while index // 2 > 0:
           if self.heap_list[index] < self.heap_list[index // 2]:</pre>
               temp = self.heap_list[index // 2]
               self.heap_list[index // 2] = self.heap_list[index]
               self.heap_list[index] = temp
           index //= 2
   def percolate_down(self, index):
       111
       compare the item at index with its smallest child
       if the item is greater than its smallest child, swap!
       continue continue while there are children to compare with
       (can stop once an item is swapped into a position where it is less than \Box
\hookrightarrow both children)
```

```
while (index * 2) <= self.size:
    mc = self.min_child(index)
    if self.heap_list[index] > self.heap_list[mc]:
        temp = self.heap_list[index]
        self.heap_list[index] = self.heap_list[mc]
        self.heap_list[mc] = temp
    index = mc
```

```
[6]: def main():
         df = pd.read_csv(r'cancer.csv', header=None)
         df_size = len(df.values)
         df_list = []
         for i in range(df_size):
             df_sample = toList(df, i)
             df_list.append(df_sample)
         hc_list = []
         for samp in df_list:
             hc = HierachicalCluster(samp[0], samp[1:])
             hc_list.append(hc)
         while len(hc_list) > 1:
             bmh = BinaryHeap()
             pairs_list = all_pairs(hc_list)
             for pair in pairs list:
                 bmh.insert(pair)
             min = bmh.del_min()
             new_cluster = HierachicalCluster('branch', None, distance=min.
      →compute_distance())
             new_cluster.set_left_child(min.get_hc_member1())
             new_cluster.set_right_child(min.get_hc_member2())
             new_cluster.compute_centroid()
             hc_list.append(new_cluster)
             hc_list.remove(new_cluster.get_left_child())
             hc_list.remove(new_cluster.get_right_child())
     def toList(df, ri):
         a = df.loc[ri]
         b = a.to_numpy()
         c = b.tolist()
         return c
     def all_pairs(hc_list):
         pairs = []
         for i in range(len(hc_list)-1):
```

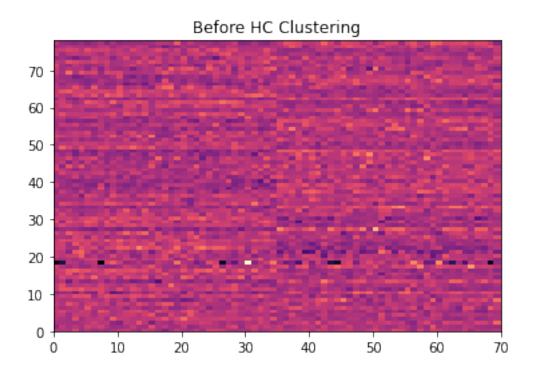
```
for j in range(i+1,len(hc_list)):
    pair = Pair(hc_list[i],hc_list[j])
    pairs.append(pair)
    return pairs

main()
```

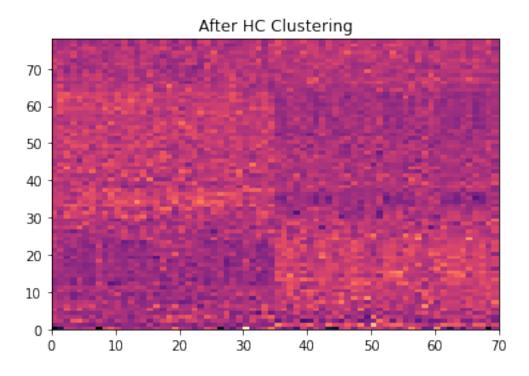
```
[7]: df = pd.read_csv(r'cancer.csv', header=None)
    df_size = len(df.values)
    df_list = []
    for i in range(df size):
        df_sample = toList(df, i)
        df_list.append(df_sample)
    hc_list = []
    for samp in df_list:
        hc = HierachicalCluster(samp[0], samp[1:])
        hc_list.append(hc)
    hc1 = []
    for i in range(len(hc_list)):
        hc1.append(hc_list[i].get_root())
    print('clustering in process...')
    while len(hc list) > 1:
        bmh = BinaryHeap()
        pairs list = all pairs(hc list)
        for pair in pairs_list:
            bmh.insert(pair)
        min = bmh.del_min()
        new_cluster = HierachicalCluster('branch', None, distance=min.
     new_cluster.set_left_child(min.get_hc_member1())
        new_cluster.set_right_child(min.get_hc_member2())
        new_cluster.compute_centroid()
        hc_list.append(new_cluster)
        hc_list.remove(new_cluster.get_left_child())
        hc_list.remove(new_cluster.get_right_child())
    leaves = hc_list[0].get_all_leaves()
    roots = []
    for i in range(len(leaves)):
        roots.append(leaves[i].get_root())
    df = pd.DataFrame(data=roots)
```

clustering in process...

```
[8]: df
[8]:
    0 -6.743823 -5.055445 0.131029 -0.179264 -0.905948 -0.099369 0.171932
    1 - 1.569939 - 0.976442 - 0.776502 - 0.903095 - 1.530858 - 1.691495 - 2.058896
        0.504952 0.193136 -1.430769 1.992230 2.415003 0.555135 -0.886940
        0.688423 - 0.664907 - 0.032537 - 0.265434 - 0.178374 - 1.381730 0.331276
    3
    4 -0.362366 -0.532900 -1.130017 -1.638415 -1.325531 -0.975372 -1.210770
       0.128002 -0.052396 -0.243591 1.119036 -0.638129 0.035529 -0.686474
    73
    75 0.001240 -0.572502 -0.021985 -0.713520 0.129617 0.808276 0.937815
    76 -0.482456 -0.878758 -0.354395 0.785844 -1.035394 0.293666 -0.665914
    77 0.051278 -0.884038 -0.539067 0.027545 -1.164839 -0.525712 -1.102827
              7
                       8
                                 9
                                             60
                                                       61
                                                                62
                                                                          63
                                                                             \
      -6.667616 1.012630 2.135013
                                    ... 3.629780 -2.003763 -4.489750 -0.832883
    1 \quad -0.854075 \quad -1.392909 \quad -1.389785 \quad \dots \quad 0.239395 \quad 2.538446 \quad 2.015366 \quad 2.125202
    2
        0.542794 - 0.639944 - 1.572939 \dots -2.271288 - 1.692297 0.177905
                                                                   1.025060
    3 -1.491776 0.223004 -0.723611 ... 1.042043 0.238790 0.466701 0.493270
    4 -1.218476 0.169423 0.228067
                                       1.048464 -0.731945 -0.728190 -0.706582
    73 -0.317596 -0.775309 -0.382444 ... 0.554033 -0.041529 -0.406905 -0.297769
    74 0.175020 -1.170122 -0.475816 ... 0.740247 0.072675 -0.172259 -0.533751
    75 -0.155640 0.169423 -0.168765 ... -0.807258 0.705989 0.322303 0.423472
    76 -0.469430 -0.823250 -0.059232 ... 0.438452 0.742327 0.195955 0.616246
    77 -0.631386 0.251205 -0.511728 ... 0.085286 0.789047 -0.168649 -0.134908
              64
                       65
                                 66
                                          67
                                                             69
                                                    68
      -3.425178 0.029338 -0.789247 -0.383363 -6.173325 -1.197897
        1.456331 -0.688572 2.237339 2.230814 -0.081404 1.270326
      -1.832721 -2.023436 -0.376697 -1.279755 0.227700 -2.085043
        3
      -0.635125  0.982812 -1.409898 -0.822631  1.312783  0.130099
    73 0.497385 -0.677355 0.050457 -0.447646 -0.155460 0.059345
    74 0.380229 -0.868050 -0.500827 -0.286939 1.422257 -0.229113
    75 0.085169 -0.666138 0.868256 0.430889 0.002312 -1.401995
    76 0.740376 -0.419356 0.295067 0.166614 0.710675 -0.065835
    77 0.414942 -0.660529 -0.077324 -0.086947 1.522072 -0.468588
    [78 rows x 70 columns]
[9]: plt.pcolor(hc1,cmap='magma')
    plt.title('Before HC Clustering')
    plt.show()
```



```
[10]: plt.pcolor(df,cmap='magma')
  plt.title('After HC Clustering')
  plt.show()
```



[]: