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Capacitated k-means clustering?





I'm newbie to algorithm and optimization.

I'm trying to implement capacitated k-means, but getting unresolved and poor result so far.

This is used as part of a CVRP simulation (capacitated vehicle routing problem).

I'm curious if I interprets the referenced algorithm wrong.

Ref: "Improved K-Means Algorithm for Capacitated Clustering Problem" (Geetha, Poonthalir, Vanathi)

The simulated CVRP has 15 customers, with 1 depot.

Each customer has Euclidean coordinate (x,y) and demand.

There are 3 vehicles, each has capacity of 90.

So, the capacitated k-means is trying to cluster 15 customers into 3 vehicles, with the total demands in each cluster must not exceed vehicle capacity.

UPDATE:

In the referenced algorithm, I couldn't catch any information about what must the code do when it runs out of "next nearest centroid". That is, when all of the "nearest centroids" has been examined, in the step 14.b below, while the <code>customers[1]</code> is still unassigned.

This results in the customer with index 1 being unassigned.

Note: customer[1] is customer with largest demand (30).

Q: When this condition is met, what the code should do then?

Here is my interpretation of the referenced algorithm, please correct my code, thank you.

- 1. Given n requesters (customers), n = customerCount, and a depot
- n demands.
- 3. n coordinates (x,y)
- 4. calculate number of clusters, k = (sum of all demands) / vehicleCapacity
- 5. select initial centroids.
 - 5.a. sort customers based on $\mbox{\tt demand}$, in descending order = $\mbox{\tt d_customers}$,
 - 5.b. select k first customers from d_customers as initial centroids = centroids[0 .. k-1],
- 6. Create binary matrix bin_matrix, dimension = (customerCount) x (k),

6.a. Fill bin matrix with all zeros

7. start WHILE loop, condition = WHILE not converged .

7.a. converged = False

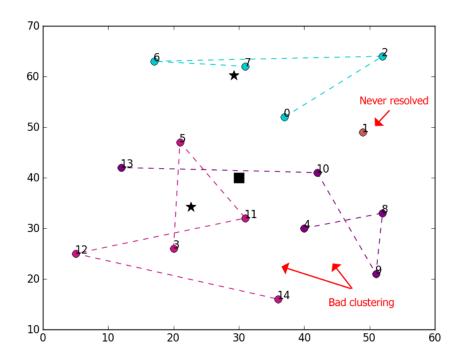
- 8. start FOR loop, condition = FOR each customers,
 - 8.a. index of customer = i
- 9. calculate Euclidean distances from <code>customers[i]</code> to all <code>centroids => edist</code>
 - 9.a. sort edist in ascending order,
 - 9.b. select first centroid with closest distance = closest_centroid
- 10. start WHILE loop, condition = while customers[i] is not assigned to any cluster.
- 11. group all the other unassigned customers = G,
 - 11.a. consider closest_centroid as centroid for ${\tt G}$.
- 12. calculate priorities Pi for each customers of G,
 - 12.a. Priority Pi = (distance from customers[i] to closest_cent) / demand[i]
 - 12.b. select a customer with highest priority Pi .
 - 12.c. customer with highest priority has index = hpc
 - 12.d. Q: IF highest priority customer cannot be found, what must we do?
- 13. assign customers[hpc] to centroids[closest_centroid] if possible.
 - 13.a. demand of customers[hpc] = d1,
 - 13.b. sum of all demands of centroids' members = dtot,
 - 13.C. IF (d1 + dtot) <= vehicleCapacity, THEN \dots
 - 13.d. assign customers[hpc] to centroids[closest_centroid]
 - 13.e. update bin_matrix , row index = hpc , column index = closest_centroid , set to 1 .

- 14. IF customers[i] is (still) not assigned to any cluster, THEN..
 - 14.a. choose the \mbox{next} nearest centroid , with the next nearest distance from \mbox{edist} .
 - 14.b. Q: IF there is no next nearest centroid, THEN what must we do?
- 15. calculate converged by comparing previous matrix and updated matrix bin_matrix.
 - 15.a. IF no changes in the bin_matrix , then set converged = True .
- 16. otherwise, calculate new centroids from updated clusters.
 - 16.a. calculate new centroids' coordinates based on members of each cluster.
 - 16.b. $sum_x = sum of all x$ -coordinate of a cluster members,
 - 16.c. num_c = number of all customers (members) in the cluster,
 - 16.d. new centroid's x-coordinate of the cluster = sum_x / num_c .
 - 16.e. with the same formula, calculate new centroid's y-coordinate of the cluster = sum_y / num_c .
- 17. iterate the main WHILE loop.

My code is always ended with unassigned customer at the step 14.b.

That is when there is a <code>customers[i]</code> still not assigned to any centroid, and it has run out of "next nearest centroid".

And the resulting clusters is poor. Output graph:



-In the picture, star is centroid, square is depot.

In the pic, customer labeled "1", with demand=30 always ended with no assigned cluster.

Output of the program,

```
k_cluster 3 idx [ 1 -1 1 0 2 0 1 1 2 2 2 0 0 2 0] centroids [(22.6, 29.2), (34.25, 60.25), (39.4, 33.4)] members [[3, 14, 12, 5, 11], [0, 2, 6, 7], [9, 8, 4, 13, 10]] demands [86, 65, 77]
```

First and third cluster is poorly calculated.

idx with index ' 1 ' is not assigned (-1)

Q: What's wrong with my interpretation and my implementation?

Any correction, suggestion, help, will be very much appreciated, thank you in advanced.

Here is my full code:

```
[8, 31.0, 62.0, 23], [9, 52.0, 33.0, 11], \
               [10, 51.0, 21.0, 5], [11, 42.0, 41.0, 19], \
[12, 31.0, 32.0, 29], [13, 5.0, 25.0, 23], \
[14, 12.0, 42.0, 21], [15, 36.0, 16.0, 10]]
customerCount = 15
vehicleCount = 3
vehicleCapacity = 90
assigned = [-1] * customerCount
# number of clusters
k cluster = 0
# binary matrix
bin_matrix = []
# coordinate of centroids
centroids = []
# total demand for each cluster, must be <= capacity
tot_demand = []
# members of each cluster
members = [j
# coordinate of members of each cluster
xy_members = []
def distance(p1, p2):
     return math.sqrt((p1[0] - p2[0])**2 + (p1[1] - p2[1])**2)
# capacitated k-means clustering
# http://www.dcc.ufla.br/infocomp/artigos/v8.4/art07.pdf
def cap_k_means():
     global k_cluster, bin_matrix, centroids, tot_demand
     global members, xy_members, prev_members
     # calculate number of clusters
     tot_demand = sum([c[3] for c in customers])
     k_cluster = int(math.ceil(float(tot_demand) / vehicleCapacity))
     print 'k_cluster', k_cluster
     # initial centroids = first sorted-customers based on demand
     d_customers = sorted(customers, key=itemgetter(3), reverse=True)
     centroids, tot_demand, members, xy_members = [], [], [], []
      \begin{tabular}{ll} for i & in & range(k_cluster): \\ \end{tabular} 
         centroids.append(d_customers[i][1:3]) # [x,y]
         # initial total demand and members for each cluster
         tot_demand.append(0)
         members.append([])
         xy members.append([])
     # binary matrix, dimension = customerCount-1 x k_cluster
     bin_matrix = [[0] * k_cluster for i in range(len(customers))]
     converged = False
     while not converged: # until no changes in formed-clusters
         prev_matrix = deepcopy(bin_matrix)
         for i in range(len(customers)):
              edist = [] # list of distance to clusters
              if assigned[i] == -1: # if not assigned yet
                   # Calculate the Euclidean distance to each of k-clusters
                  for k in range(k_cluster):
   p1 = (customers[i][1], customers[i][2]) # x,y
   p2 = (centroids[k][0], centroids[k][1])
   edist.append((distance(p1, p2), k))
                   # sort, based on closest distance
                   edist = sorted(edist, key=itemgetter(0))
              closest_centroid = 0
                                         # first index of edist
              # loop while customer[i] is not assigned
              while assigned[i] == -1:
                   # calculate all unsigned customers (G)'s priority
                   max_prior = (0, -1) # value, index
for n in range(len(customers)):
                       pc = customers[n]
                        if assigned[n] == -1: # if unassigned
                            # get index of current centroid
                            c = edist[closest_centroid][1]
                            cen = centroids[c]
                            # distance_cost / demand
                            p = distance((pc[1], pc[2]), cen) / pc[3]
                            # find highest priority
                            if p > max_prior[0]:
                                 max_prior = (p, n) # priority, customer-index
                   # if highest-priority is not found, what should we do ???
                   if max_prior[1] == -1:
                        break
                   # try to assign current cluster to highest-priority customer
                   hpc = max_prior[1]  # index of highest-priority customer
c = edist[closest_centroid][1]  # index of current cluster
                   # constraint, total demand in a cluster <= capacity
                   if tot_demand[c] + customers[hpc][3] <= vehicleCapacity:
    # assign new member of cluster
    members[c].append(hpc) # add index of customer</pre>
```

```
xy = (customers[hpc][1], customers[hpc][2]) # x,y
                           xy_members[c].append(xy)
                           tot_demand[c] += customers[hpc][3]
                           assigned[hpc] = c  # update cluster to assigned-customer
                           # update binary matrix
                           bin_matrix[hpc][c] = 1
                     # if customer is not assigned then,
                     if assigned[i] == -1:
                          if closest_centroid < len(edist)-1:</pre>
                                # choose the next nearest centroid
                                closest_centroid += 1
                           # if run out of closest centroid, what must we do ???
                           else:
                               break # exit without centroid ???
                # end while
          # end for
          # Calculate the new centroid from the formed clusters
           for j in range(k_cluster):
               xj = sum([cn[0] for cn in xy_members[j]])
yj = sum([cn[1] for cn in xy_members[j]])
xj = float(xj) / len(xy_members[j])
yj = float(yj) / len(xy_members[j])
                centroids[j] = (xj, yj)
          # calculate converged
converged = numpy.array_equal(numpy.array(prev_matrix),
numpy.array(bin_matrix))
      # end while
def clustering():
     cap k means()
     # debug plot
     idx = numpy.array([c for c in assigned])
     xy = numpy.array([(c[1], c[2]) for c in customers])
     COLORS = ["Blue", "DarkSeaGreen", "DarkTurquoise",
"IndianRed", "MediumVioletRed", "Orange", "Purple"]
     for i in range(min(idx), max(idx)+1):
    clr = random.choice(COLORS)
     pylab.plot(xy[idx==i, 0], xy[idx==i, 1], color=clr, \
    linestyle='dashed', \
    marker='o', markerfacecolor=clr, markersize=8)
pylab.plot(centroids[:][0], centroids[:][1], '*k', markersize=12)
     pylab.plot(depot[1], depot[2], 'sk', markersize=12)
     for i in range(len(idx)):
          pylab.annotate(str(i), xy[i])
     pylab.savefig('clust1.png')
     pylab.show()
     return idx
def main():
     idx = clustering()
print 'idx', idx
print 'centroids', centroids
print 'members', members
print 'demands', tot_demand
if __name__ == '__main__':
     main()
python algorithm optimization cluster-analysis
                                                            k-means
edited Aug 1 '13 at 13:16
                                                               asked Aug 1 '13 at 10:59
                                                               silo
25
       Dukeling
       32.4k
                7 30 63
```

I am resisting the urge to downvote - this is way too much info. Would you settle for sorting index 1 being unassigned? - doctorlove Aug 1 '13 at 11:56

@doctorlove Thank you very much for the response. I've updated my question above, right after "UPDATE" section. (sorry for possible misunderstand your comment, I'm not native English speaker, so I have little doubt about the meaning of "settle for sorting", even after I Google translate the phrase, very sorry, thank you.) – silo Aug 1 '13 at 12:24

s/sorting/"deal with the problem of" - doctorlove Aug 1 '13 at 12:35

@doctorlove Thank you, does my update satisfy your question? Please correct me if I still provided less information needed, thanks. - silo Aug 1 '13 at 12:45

@Dukeling Thank you very much for editing my question, so that the picture of output graph can be

displayed. - silo Aug 1 '13 at 13:32

2 Answers

When the total demand is close to the total capacity, this problem begins to take on aspects of bin packing. As you've discovered, this particular algorithm's greedy approach is not always successful. I don't know whether the authors admitted that, but if they didn't, the reviewers should have caught it.

If you want to continue with something like this algorithm, I would try using integer programming to assign requesters to centroids.

I understand it. Thank you very much. I'll switch to mixed integer programming then. - silo Aug 1 '13 at 14:20

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Without going through all the details, the paper you cite says

```
if ri is not assigned then
    choose the next nearest centroid
end if
```

in the algorithm at the end of section 5.

There must be a next nearest centroid - if two are equidistant I presume it doesn't matter which you choose.

answered Aug 1 '13 at 12:39



Thank you very much. I store the next nearest centroids in a list. There are at most 3 nearest centroids. In my case, all 3 as the next nearest centroids has been chosen, while the customer[1] is still unassigned. It does "cross" assignment, based on the "self and other" unassigned-customers with highest priority, so customer[1] never has a chance to be assigned, due to it has lower priority, even until there is no more "next nearest centroids" (already hit the third index of nearest centroids). Thanks. — silo Aug 1 '13 at 13:12

Ouch - hopefully someone else who knows this variant of the algo will step in and help - doctorlove Aug 1 '13 at 13:32

Thank you so much because you've examined, answered and commented my problem, I really appreciate it. - silo Aug 1 '13 at 13:43