```
procedure communityDetection(k, data, alpha, delta, epsilon)
   /* COMMENTS:
    * Line 2: Read in data and implement graph as an adjacency list with n raws (= num of
    nodes).
    * Line 3: Create the particle objects and save them in a vector P, \dim(P) = k \times 1.
    * Lines 4, 5: Disperse all k particles randomly in graph G. Calculate initial particle
    energy levels (attribute in particle struct).
    * Line 6: Initialize particle state (attribute in particle struct). Initially all
    particles have state ACTIVE (vs EXHAUSTED).
    * Line 7: Initialize matrix N, containing the number of times each particle visits EACH
    node in G. dim(N) = n \times k.
    * Line 8: Compute matrix NBar, containing the relative visit frequencies of all particles
    to all nodes in G. dim(NBar) = n \times k.
    * Line 9: Compute Prand, the matrix containing the probability that a particle p,
    occupying node i at time t, would move to node j at time t+1.
    * Prand is Markovian, proportional to the edge weight linking nodes i and j. It is
    time-invariant and thus computed only once. */
2:
       G <- buildGraph(data)</pre>
3:
       P <- initParticles(k)
4:
       initParticleDisperse (G)
5:
       calcInitE(P, k)
       initParticleState (P,k)
6:
7:
       N <- initMatrixN(G,P)
8:
       NBar <- compMatrixNBar(N)
9:
       Prand <- compPrand(G)</pre>
   /* COMMENTS:
    * Lines 10-24: From time t=1 until a fixed point is reached, move each particle on G as
    follows -
    * Lines 13, 14: If the particle is ACTIVE, compute the preferential movement probabilities
                   for all neighbors of P[k].current node. That is, calculate the preferential
                   probability that particle k will move to node j, a neighbor of node i, k's
                   current residence.
                   The probabilities are stored in a vector of size G[i].number neighbors,
                   where i = P[k].current node.
                   The relative visit frequencies from matrix NBar are used in the computation.
                   dim(pref prob) = 1 \times number neighbors.
    * Line 15: Compute the transition vector. That is,
               tran vector[i] = (1-alpha)*rand prob[i] +alpha*pref prob[i],
               where i is the i-th neighbor of P[k].current node and rand prob[i] is
               the i-th entry in raw P[k].current node in Prand.
     * Line 16: Move particle k to a new node based on tran vector.
     * Line 17: Update the energy level of particle k (+-delta).
     * Lines 18, 19: If particle k is EXHAUSTED, reanimate it by teleporting it
               to a node it dominates (selected based on weighted rand distrib.)
     * Line 21: Update matrix N to increment the number of node visits where appropriate.
     * Line 22: Update matrixNBar to reflect the change in visit frequencies.
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*/ 10: t <- 1 11: do 12: for k = 1 to |P| do 13: if(P[k].state == ACTIVE) 14: pref prob = computePrefProb(P[k]) 15: tran vector = compTransitionVector() 16: selectNextNode(P[k], tran_vector, num_neighbors) 17: updateE(P[k]) 18: else if(P[k].state == EXHAUSTED) 19: particleRean(P[k]) 20: end **for** 21: updateMatrixN() 22: compMatrixNBar(N) 23: t <- t+1 24: 25: end procedure

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