TSP ACO

April 28, 2022

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[1]: import numpy as np
     import pandas as pd
     df = pd.read_csv('./adj_mat_kota.csv')
[20]: def generate_1_semut(ndim):
         return [ [], list(range(ndim)) ]
     def generate_individus(n_individu,ndim):
         return [ generate_1_semut(ndim) for i in range(n_individu) ]
     def generate pheromones(ndim,tau awal):
         pheromones = np.ones(shape=(ndim,ndim)) * tau_awal
         np.fill diagonal(pheromones,0)
         return pd.DataFrame(pheromones)
     def generate_delta_pheromones(ndim):
         return generate_pheromones(ndim,0)
     def f(X,adj_mat,tipe='full'):
         return sum([ adj_mat[ X[i],X[i+1] ] for i in range(len(X) - 1) ]) + (__
       →adj_mat[ X[-1],X[0] ] if tipe=='full' else 0)
     def ro_1(el_c,el_t,pher,adj_mat,params):
         return ( pher[el_c,el_t] )**params['alpha'] * ( 1/adj_mat[el_c,el_t]_
       ⇔)**params['beta']
     def ro_1_all(semut,pher,adj_mat,params):
         return [ ro_1(semut[0][-1],el_t,pher,adj_mat,params) for el_t in semut[1] ]
     def transition_prob(semut,pher,adj_mat,params):
         r = ro_1_all(semut,pher,adj_mat,params)
         total prob = sum(r)
         return [ p/total_prob for p in r ]
     def choose_next_kota(semut,pher,adj_mat,params):
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df = pd.DataFrame( {'next':semut[1] , 'prob':
 stransition_prob(semut,pher,adj_mat,params) } )
    choosen = df.sample(n=1,weights='prob').iloc[0,0]
   return nv_to_v(semut,choosen)
def choose next kota all(semuts,pher,adj mat,params):
   return [ choose_next_kota(s,pher,adj_mat,params) for s in semuts ]
def update_delta_pher(semuts,delta_pher,adj_mat,params):
   temp = generate_delta_pheromones(params['ndim'])
   for s in semuts:
       Lk = f(s[0],adj_mat,tipe='partial')
       temp[s[0][-2]][s[0][-1]] = delta_pher[s[0][-2]][s[0][-1]] +_{L}
 return temp
def update_pheromones(pher,delta_pher,semuts,adj_mat,params):
   return params['rho'] * pher +
 →update_delta_pher(semuts,delta_pher,adj_mat,params)
def sol(semuts):
   return np.array( [ s[0] for s in semuts])
def calc_fitness(semuts,adj_mat):
   return np.array( [ f(s,adj_mat,'full') for s in semuts ] )
def find best(sl,fitness):
   idxs = np.argsort(fitness)
   return np.append( sl[idxs][0],fitness[idxs][0] )
def satu_full(pheromones,adj_mat,params):
    semuts = generate_individus(params['n_individu'],params['ndim'])
    semuts = inisialisasi(semuts)
   delta_pheromones = generate_delta_pheromones(params['ndim'])
   t = 1
   while t < params['ndim']:</pre>
        semuts = choose_next_kota_all(semuts,pheromones.values,adj_mat,params)
       pheromones =
 -update_pheromones(pheromones,delta_pheromones,semuts,adj_mat,params)
       delta_pheromones = generate_delta_pheromones(params['ndim'])
       t = t + 1
   sl = sol(semuts)
   fitness = calc_fitness(sl,adj_mat)
   return find_best(sl,fitness) , pheromones
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def ACO(params,df):
         adj_mat = df.values
         generasi = 0
         bests = []
         pheromones = generate_pheromones(params['ndim'],2)
         delta_pheromones = generate_delta_pheromones(params['ndim'])
         while generasi < params['max_generasi']:</pre>
             best_ , pheromones = satu_full(pheromones,adj_mat,params)
             bests.append(best_)
             generasi = generasi + 1
         return np.array(bests)
     def nv_to_v(semut,el):
         temp = semut.copy()
         temp[0].append(el)
         temp[1].remove(el)
         return temp
     def inisialisasi(semuts):
         return [ nv_to_v(s,s[1][ np.random.randint(0,len(s[1]))] ) for s in semuts ]
[23]: params = {
         'alpha':1,
         'beta':1,
         'rho':0.9,
         'Q':1,
          'n_individu':10,
         'ndim':10,
          'max_generasi':5
     }
     hasil = ACO(params,df)
     df_hasil = pd.DataFrame(hasil)
[24]: df_hasil
[24]:
                   2
                                           7
         0
                        3
                             4
                                  5
                                                8
                                                     9
                                                               10
       4.0 8.0 9.0 1.0 3.0 6.0
                                     2.0 7.0 0.0 5.0 3.003576
     1 0.0 5.0 6.0
                       2.0 1.0 9.0 8.0 4.0 3.0 7.0 3.191468
     2 6.0 2.0 4.0 3.0 1.0 9.0 8.0 0.0 7.0 5.0 3.557272
     3 2.0 6.0 5.0 0.0 7.0 3.0 1.0 9.0 8.0 4.0 3.218540
     4 3.0 4.0 2.0 7.0 0.0 5.0 6.0 8.0 9.0 1.0 3.321142
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