```
In [1]:
         from gerrychain import Graph
In [2]:
         # Read Alabama county graph from the json file "COUNTY 01.json"
         filename = 'COUNTY 01.json'
         # GerryChain has a built-in function for reading graphs of this type:
         G = Graph.from json( filename )
In [3]:
         # For each node, print the node #, county name, and its population
         for node in G.nodes:
             name = G.nodes[node]["NAME10"]
             population = G.nodes[node]['TOTPOP']
             x coordinate = G.nodes[node]['C X']
             y coordinate = G.nodes[node]['C Y']
             print("Node",node,"is",name,"County, which has population",population,"and is cente
        Node 0 is Barbour County, which has population 27457 and is centered at ( -85.3931969717
        4619 , 31.869603217898423 )
        Node 1 is Clay County, which has population 13932 and is centered at ( -85.8605470594419
        4 , 33.269085886198276 )
        Node 2 is Marengo County, which has population 21027 and is centered at ( -87.7895192852
        3522 , 32.24761342168653 )
        Node 3 is Houston County, which has population 101547 and is centered at ( -85.302519754
        26631 , 31.15318422009409 )
        Node 4 is Cherokee County, which has population 25989 and is centered at ( -85.603793825
        52829 , 34.175955543616034 )
        Node 5 is Baldwin County, which has population 182265 and is centered at ( -87.749844694
        53685 , 30.6609738924624 )
        Node 6 is Conecuh County, which has population 13228 and is centered at ( -86.9936813437
        0104 , 31.42926727158543 )
        Node 7 is Cleburne County, which has population 14972 and is centered at ( -85.518767919
        86888 , 33.67455784656897 )
        Node 8 is Jefferson County, which has population 658466 and is centered at ( -86.8964902
        800165 , 33.55431475658097 )
        Node 9 is Henry County, which has population 17302 and is centered at ( -85.241406300126
        81 , 31.51469440693819 )
        Node 10 is Crenshaw County, which has population 13906 and is centered at ( -86.31354765
        901722 , 31.731486183501875 )
        Node 11 is Madison County, which has population 334811 and is centered at ( -86.55020653
        641006 , 34.76308970093465 )
        Node 12 is Limestone County, which has population 82782 and is centered at ( -86.9813695
        0223683 , 34.81007650885704 )
        Node 13 is Dallas County, which has population 43820 and is centered at ( -87.1064787764
        9095 , 32.325974538937146 )
        Node 14 is Covington County, which has population 37765 and is centered at ( -86.4512478
        8997703 , 31.248492845464902 )
        Node 15 is Shelby County, which has population 195085 and is centered at ( -86.660648630
        41424 , 33.2642778204166 )
        Node 16 is Bibb County, which has population 22915 and is centered at ( -87.126439083023
        29 , 32.998644313019 )
        Node 17 is Butler County, which has population 20947 and is centered at ( -86.6802892156
        2546 , 31.752433942560142 )
        Node 18 is Macon County, which has population 21452 and is centered at ( -85.69266850175
        498 , 32.38597359119281 )
        Node 19 is Autauga County, which has population 54571 and is centered at ( -86.642757249
        25504 , 32.53492069021946 )
        Node 20 is Franklin County, which has population 31704 and is centered at ( -87.84381198
        716348 , 34.441670632318164 )
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Node 21 is Marshall County, which has population 93019 and is centered at (-86.30663696
580778 , 34.36696193383658 )
Node 22 is Talladega County, which has population 82291 and is centered at ( -86.1659069
1520001 , 33.38005742632039 )
Node 23 is Walker County, which has population 67023 and is centered at ( -87.2973574325
2813 , 33.803334781108035 )
Node 24 is Lauderdale County, which has population 92709 and is centered at ( -87.654019
46382268 , 34.90114017896794 )
Node 25 is Morgan County, which has population 119490 and is centered at ( -86.852927813
64645 , 34.45347499534302 )
Node 26 is Pickens County, which has population 19746 and is centered at ( -88.088692609
80207 , 33.28087688365432 )
Node 27 is Etowah County, which has population 104430 and is centered at ( -86.034762204
99286 , 34.045254128332594 )
Node 28 is Tuscaloosa County, which has population 194656 and is centered at ( -87.52502
994601569 , 33.289549847357435 )
Node 29 is Dale County, which has population 50251 and is centered at ( -85.611031961504
93 , 31.431819470587527 )
Node 30 is Tallapoosa County, which has population 41616 and is centered at ( -85.797502
22941729 , 32.8624049444577 )
Node 31 is Geneva County, which has population 26790 and is centered at ( -85.8389848908
043 , 31.095016702449865 )
Node 32 is Mobile County, which has population 412992 and is centered at (-88.197529576
3405 , 30.685146910580293 )
Node 33 is Wilcox County, which has population 11670 and is centered at ( -87.3081990893
2153 , 31.98924196018949 )
Node 34 is Sumter County, which has population 13763 and is centered at ( -88.1987913473
9342 , 32.591014077360576 )
Node 35 is Lawrence County, which has population 34339 and is centered at ( -87.31104472
929573 , 34.521650917020324 )
Node 36 is Calhoun County, which has population 118572 and is centered at ( -85.82603486
914782 , 33.77142782210847 )
Node 37 is Jackson County, which has population 53227 and is centered at ( -85.999300681
75878 , 34.77941470747724 )
Node 38 is Marion County, which has population 30776 and is centered at ( -87.8871384347
8739 , 34.13654972020097 )
Node 39 is DeKalb County, which has population 71109 and is centered at ( -85.8041414499
2365 , 34.45977281989643 )
Node 40 is Pike County, which has population 32899 and is centered at ( -85.940920740428
94 , 31.802715045380932 )
Node 41 is Perry County, which has population 10591 and is centered at ( -87.29440084137
56 , 32.63846568673937 )
Node 42 is Colbert County, which has population 54428 and is centered at ( -87.805296766
72091 , 34.70025404740263 )
Node 43 is Elmore County, which has population 79303 and is centered at ( -86.1491463397
6642 , 32.59664788700988 )
Node 44 is Washington County, which has population 17581 and is centered at ( -88.207876
81246034 , 31.407603796087137 )
Node 45 is Chambers County, which has population 34215 and is centered at (-85.39204294
373201 , 32.9143722172032 )
Node 46 is Clarke County, which has population 25833 and is centered at ( -87.8308098505
0333 , 31.676659580123818 )
Node 47 is St. Clair County, which has population 83593 and is centered at ( -86.3146877
4720817 , 33.71570290572818 )
Node 48 is Chilton County, which has population 43643 and is centered at ( -86.718813730
83016 , 32.84785313410404 )
Node 49 is Bullock County, which has population 10914 and is centered at ( -85.715697201
64433 , 32.10055420532796 )
Node 50 is Greene County, which has population 9045 and is centered at ( -87.95223434205
52 , 32.853137408728344 )
Node 51 is Blount County, which has population 57322 and is centered at ( -86.5673709599
9279 , 33.98086739482323 )
Node 52 is Coosa County, which has population 11539 and is centered at ( -86.24765921602
953 , 32.936226896286094 )
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Node 53 is Winston County, which has population 24484 and is centered at ( -87.373683521

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54319 , 34.14922498766865 )
         Node 54 is Russell County, which has population 52947 and is centered at ( -85.184962597
         87717 , 32.288380056218536 )
         Node 55 is Coffee County, which has population 49948 and is centered at ( -85.9882069708
         7443 , 31.40262716953991 )
         Node 56 is Monroe County, which has population 23068 and is centered at ( -87.3654308965
         5846 , 31.570836608980876 )
         Node 57 is Lamar County, which has population 14564 and is centered at ( -88.09689885856
         152 , 33.77920579845164 )
         Node 58 is Choctaw County, which has population 13859 and is centered at ( -88.263201266
         21178 , 32.01961076277522 )
         Node 59 is Cullman County, which has population 80406 and is centered at ( -86.867619397
         44032 , 34.13194106743531 )
         Node 60 is Montgomery County, which has population 229363 and is centered at ( -86.20761
         429347573 , 32.22025793851669 )
         Node 61 is Escambia County, which has population 38319 and is centered at ( -87.16161983
         379405 , 31.126122544455463 )
         Node 62 is Randolph County, which has population 22913 and is centered at ( -85.45907017
         313048 , 33.29378740802176 )
         Node 63 is Fayette County, which has population 17241 and is centered at ( -87.738861962
         61972 , 33.72120614857144 )
         Node 64 is Lee County, which has population 140247 and is centered at ( -85.355562714324
         5 , 32.601136470581956 )
         Node 65 is Lowndes County, which has population 11299 and is centered at ( -86.650107703
         8513 , 32.15475001138599 )
         Node 66 is Hale County, which has population 15760 and is centered at ( -87.629117736684
         48 , 32.76266425862635 )
In [12]:
          # pip install geopy
          !pip install geopy
         Collecting geopy
           Downloading geopy-2.1.0-py3-none-any.whl (112 kB)
         Collecting geographiclib<2,>=1.49
           Downloading geographiclib-1.50-py3-none-any.whl (38 kB)
         Installing collected packages: geographiclib, geopy
         Successfully installed geographiclib-1.50 geopy-2.1.0
In [13]:
          import geopy
In [14]:
          # what is the "distance" between Barbour County (node 0), Shelby County (node 15), and
          from geopy.distance import geodesic
          # Store centroid location as ( long, lat )
          Barbour = (G.nodes[0]['C_Y'], G.nodes[0]['C_X'])
          Shelby = ( G.nodes[15]['C_Y'], G.nodes[15]['C_X'] )
          Walker = ( G.nodes[23]['C_Y'], G.nodes[23]['C_X'] )
          # Print the distance in miles
          print("Barbour -> Shelby:",geodesic(Barbour, Shelby).miles)
          print("Shelby -> Walker:",geodesic(Shelby, Walker).miles)
          print("Walker -> Barbour:",geodesic(Walker, Barbour).miles)
         Barbour -> Shelby: 121.2644116391872
         Shelby -> Walker: 52.255718525283626
         Walker -> Barbour: 173.27867565607963
In [15]:
          # create distance dictionary
          dist = dict()
```

```
for i in G.nodes:
              for j in G.nodes:
                  loc_i = (G.nodes[i]['C_Y'], G.nodes[i]['C_X'])
                  loc_j = (G.nodes[j]['C_Y'], G.nodes[j]['C_X'])
                  dist[i,j] = geodesic(loc_i,loc_j).miles
In [16]:
          # check the dictionary by printing the Barbour County -> Shelby County distance
          print("Barbour -> Shelby:",dist[0,15])
         Barbour -> Shelby: 121.2644116391872
In [17]:
          # Let's impose a 1% population deviation (+/- 0.5%)
          deviation = 0.01
          import math
                         # number of districts
          k = 7
          total population = sum(G.nodes[node]['TOTPOP'] for node in G.nodes)
          L = math.ceil((1-deviation/2)*total_population/k)
          U = math.floor((1+deviation/2)*total population/k)
          print("Using L =",L,"and U =",U,"and k =",k)
         Using L = 679406 and U = 686233 and k = 7
In [18]:
          import gurobipy as gp
          from gurobipy import GRB
          # create model
          m = gp.Model()
          # create x[i,j] variable which equals one when county i
               is assigned to (the district centered at) county j
          x = m.addVars(G.nodes, G.nodes, vtype=GRB.BINARY)
         Academic license - for non-commercial use only - expires 2021-06-24
         Using license file C:\Users\Louisa\gurobi.lic
In [19]:
          # objective is to minimize the moment of inertia: d^2 * p * x
          m.setObjective( gp.quicksum( dist[i,j]*dist[i,j]*G.nodes[i]['TOTPOP']*x[i,j]
                                       for i in G.nodes for j in G.nodes), GRB.MINIMIZE )
In [20]:
          # add constraints saying that each county i is assigned to one district
          m.addConstrs( gp.quicksum(x[i,j] for j in G.nodes) == 1 for i in G.nodes)
          # add constraint saying there should be k district centers
          m.addConstr( gp.quicksum( x[j,j] for j in G.nodes ) == k )
          # add constraints that say: if j roots a district, then its population is between L and
          m.addConstrs( gp.quicksum( G.nodes[i]['TOTPOP'] * x[i,j] for i in G.nodes) >= L * x[j,j]
          m.addConstrs( gp.quicksum( G.nodes[i]['TOTPOP'] * x[i,j] for i in G.nodes) <= U * x[j,j]</pre>
          # add coupling constraints saying that if i is assigned to j, then j is a center.
          m.addConstrs(x[i,j] \leftarrow x[j,j]  for i in G.nodes for j in G.nodes )
          m.update()
```

```
In [21]: | # Add contiguity constraints
          import networkx as nx
          DG = nx.DiGraph(G)
          # Add variable f[j,u,v] which equals the amount of flow (originally from j) that is sen
          f = m.addVars( DG.nodes, DG.edges, vtype=GRB.CONTINUOUS)
          M = DG.number of nodes()-1
          # Add constraint saying that node j cannot receive flow of its own type
          m.addConstrs( gp.quicksum( f[j,u,j] for u in DG.neighbors(j) ) == 0 for j in DG.nodes )
          # Add constraints saying that node i can receive flow of type j only if i is assigned t
          m.addConstrs(gp.quicksum(f[j,u,i] for u in DG.neighbors(i)) <= M * x[i,j] for i in DG
          # If i is assigned to j, then i should consume one unit of j flow.
               Otherwise, i should consume no units of j flow.
          m.addConstrs(gp.quicksum(f[j,u,i] - f[j,i,u] for u in DG.neighbors(i)) == x[i,j] for
          m.update()
In [22]:
          # solve, making sure to set a 0.00% MIP gap tolerance(!)
          m.Params.MIPGap = 0.0
          m.optimize()
         Changed value of parameter MIPGap to 0.0
            Prev: 0.0001 Min: 0.0 Max: inf Default: 0.0001
         Gurobi Optimizer version 9.1.1 build v9.1.1rc0 (win64)
         Thread count: 4 physical cores, 8 logical processors, using up to 8 threads
         Optimize a model with 13602 rows, 27403 columns and 99280 nonzeros
         Model fingerprint: 0x8806bbad
         Variable types: 22914 continuous, 4489 integer (4489 binary)
         Coefficient statistics:
           Matrix range [1e+00, 7e+05]
           Objective range [4e+06, 4e+10]
           Bounds range [1e+00, 1e+00]
           RHS range
                           [1e+00, 7e+00]
         Warning: Model contains large objective coefficients
                  Consider reformulating model or setting NumericFocus parameter
                  to avoid numerical issues.
         Presolve removed 1027 rows and 2760 columns
         Presolve time: 1.12s
         Presolved: 12575 rows, 24643 columns, 90388 nonzeros
         Variable types: 20420 continuous, 4223 integer (4223 binary)
         Root relaxation: objective 5.661390e+09, 1722 iterations, 0.37 seconds
                          Current Node
                                                 Objective Bounds
             Nodes
                                                                              Work
          Expl Unexpl | Obj Depth IntInf | Incumbent BestBd Gap | It/Node Time
                                     0 175
              0
                    0 5.6614e+09
                                                     - 5.6614e+09
                                                                                  2s
                                               - 5.7467e+09

- 5.7636e+09

- 5.7637e+09

- 5.7992e+09

- 5.8036e+09

- 5.8284e+09

- 5.8447e+09

- 5.8597e+09

- 5.8641e+09

- 5.8671e+09
                    0 5.7467e+09
              0
                                     0 241
                                                   - 5.7467e+09
                                                                                  3s
                    0 5.7636e+09 0 250
                                                                                  4s
                    0 5.7637e+09 0 250
              0
                                                                                  4s
                    0 5.7992e+09 0 241
              0
                                                                                  5s
                    0 5.8036e+09 0 254
              0
                                                                                  5s
                    0 5.8284e+09 0 258
              0
                                                                                  5s
              0
                    0 5.8447e+09 0 278
                                                                                6s
              0
                   0 5.8449e+09 0 279
                                                                                6s
              0
                    0 5.8597e+09 0 282
                                                                                6s
                    0 5.8641e+09 0 280
              0
                                                                                  7s
```

0 250

7s

0 5.8671e+09

0	0 5 96735:00	0	256		F 067300			7-
0	0 5.8673e+09	0	256		5.8673e+09	-	-	7s
0	0 5.8686e+09	0	264	-	5.8686e+09	-	-	7s
0	0 5.8688e+09	0	260	-	5.8688e+09	_	_	7s
0	0 5.8689e+09	0	262	_	5.8689e+09	_	_	7s
	0 5.8689e+09		264			_	_	
0		0			5.8689e+09			7s
0	0 5.8898e+09	0	306	-	5.8898e+09	-	-	8s
0	0 5.8916e+09	0	321	_	5.8916e+09	_	-	8s
0	0 5.8921e+09	0	321	_	5.8921e+09	_	_	8s
			321					
0	0 5.8921e+09	0			5.8921e+09	-	-	8s
0	0 5.8921e+09	0	322	-	5.8921e+09	-	-	8s
0	0 5.9102e+09	0	294	-	5.9102e+09	-	-	8s
0	0 5.9117e+09	0	308	_	5.9117e+09	_	_	9s
0	0 5.9119e+09	0	310		5.9119e+09	_	_	9s
0	0 5.9158e+09	0	284		5.9158e+09	-	-	9s
0	0 5.9158e+09	0	285	-	5.9158e+09	-	-	9s
0	0 5.9191e+09	0	284	_	5.9191e+09	_	_	10s
0	0 5.9191e+09	0	288		5.9191e+09	_	_	10s
						_		
0	0 5.9200e+09	0	271		5.9200e+09	-	-	10s
0	0 5.9207e+09	0	296	-	5.9207e+09	-	-	11s
0	0 5.9213e+09	0	287	_	5.9213e+09	_	-	11s
0	0 5.9215e+09	0	284	_	5.9215e+09	_	_	11s
0	0 5.9215e+09	0	285		5.9215e+09		_	11s
						_		
0	0 5.9238e+09	0	306		5.9238e+09	-	-	11s
0	0 5.9241e+09	0	314	-	5.9241e+09	-	-	12s
0	0 5.9241e+09	0	314	_	5.9241e+09	_	_	12s
0	0 5.9248e+09	0	316		5.9248e+09	_	_	12s
0	0 5.9248e+09	0	316		5.9248e+09	_	_	12s
0	0 5.9249e+09	0	326		5.9249e+09	-	-	13s
0	0 5.9251e+09	0	332	-	5.9251e+09	-	-	13s
0	0 5.9251e+09	0	317	-	5.9251e+09	_	-	13s
0	0 5.9255e+09	0	333		5.9255e+09	_	_	14s
0	0 5.9255e+09	0	331		5.9255e+09	-	-	14s
0	0 5.9262e+09	0	310	-	5.9262e+09	-	-	14s
0	0 5.9263e+09	0	316	-	5.9263e+09	-	-	15s
0	0 5.9263e+09	0	308	_	5.9263e+09	_	_	15s
0	0 5.9305e+09	0	333		5.9305e+09	_	_	15s
0	0 5.9306e+09	0	331		5.9306e+09	-	-	15s
0	0 5.9331e+09	0	317	-	5.9331e+09	-	-	16s
0	0 5.9336e+09	0	317	-	5.9336e+09	-	-	16s
0	2 5.9336e+09	0	317	_	5.9336e+09	_	_	20s
108	117 6.6698e+09	21	22	_	5.9663e+09	_	158	25s
620	466 7.2493e+09	46	12		5.9818e+09	-	88.6	30s
1044	729 6.5538e+09	18	317	-	6.0132e+09	-	85.1	37s
1046	730 8.2062e+09	35	175	_	6.0132e+09	-	85.0	40s
1056	737 6.5616e+09	23	295	_	6.0132e+09	_	84.2	45s
1066	744 8.4801e+09	35	316		6.0132e+09	_	83.4	50s
1069	746 6.9744e+09	18	351		6.0132e+09	-	83.1	59s
1070	749 6.0132e+09	15	281	-	6.0132e+09	-	5.5	66s
1072	753 6.0304e+09	16	267	_	6.0132e+09	_	6.3	70s
1131	803 6.5993e+09	24	19		6.0303e+09	_	18.1	75s
1415	902 6.0668e+09	18	209		6.0313e+09	-	27.4	80s
1490	954 6.1919e+09	27	19		6.0313e+09	-	36.0	85s
1892	1163 6.1273e+09	22	133	-	6.0669e+09	-	43.3	90s
2203	1253 6.5566e+09	32	20	_	6.0669e+09	_	48.7	95s
2613	1461 6.2920e+09	24	61		6.0766e+09	_	55.5	100s
3244	1715 7.6653e+09	55	6		6.0965e+09	-	59.5	105s
3670	2079 6.3004e+09	22	111		6.1107e+09	-	64.1	110s
* 3804	672	39	6	.613137e+09	6.1170e+09	7.50%	65.4	111s
H 3957	297		6	.409756e+09	6.1170e+09	4.57%	65.4	112s
4140	300 6.3664e+09	26			6.1340e+09		66.5	115s
			1)1					
4564	313 cutoff	20			6.1958e+09	3.34%	68.8	120s
5005	261 6.3430e+09	26		6.4098e+09		2.68%	68.7	125s
5512	0 6.3370e+09	26	215	6.4098e+09	6.3107e+09	1.55%	68.7	130s

```
MIR: 18
            StrongCG: 8
            Flow cover: 123
            GUB cover: 10
            Zero half: 6
            Mod-K: 2
            Network: 8
            RLT: 1
          Explored 5718 nodes (485813 simplex iterations) in 130.86 seconds
          Thread count was 8 (of 8 available processors)
          Solution count 2: 6.40976e+09 6.61314e+09
          Optimal solution found (tolerance 0.00e+00)
          Best objective 6.409755608475e+09, best bound 6.409755608475e+09, gap 0.0000%
In [23]:
           print("The moment of inertia objective is",m.objval)
           # retrieve the districts and their populations
           centers = [j \text{ for } j \text{ in } G.nodes \text{ if } x[j,j].x > 0.5]
           districts = [ [i for i in G.nodes if x[i,j].x > 0.5] for j in centers]
           district_counties = [ [ G.nodes[i]["NAME10"] for i in districts[j] ] for j in range(k)]
           district populations = [ sum(G.nodes[i]["TOTPOP"] for i in districts[j]) for j in range
           # print district info
           for j in range(k):
               print("District",j,"has population",district populations[j],"and contains counties"
          The moment of inertia objective is 6409755608.475345
          District 0 has population 681381 and contains counties ['Jefferson', 'Bibb']
          District 1 has population 680437 and contains counties ['Madison', 'Limestone', 'Frankli
          n', 'Lauderdale', 'Jackson', 'Marion', 'Colbert']
          District 2 has population 684669 and contains counties ['Cherokee', 'Cleburne', 'Marshal
          l', 'Morgan', 'Etowah', 'Lawrence', 'DeKalb', 'St. Clair', 'Blount', 'Cullman']
         District 3 has population 683280 and contains counties ['Dallas', 'Shelby', 'Walker', 'Pickens', 'Tuscaloosa', 'Sumter', 'Perry', 'Chilton', 'Greene', 'Winston', 'Lamar', 'Choc
          taw', 'Fayette', 'Hale']
          District 4 has population 684512 and contains counties ['Clay', 'Macon', 'Autauga', 'Tal
          ladega', 'Tallapoosa', 'Calhoun', 'Elmore', 'Chambers', 'Bullock', 'Coosa', 'Russell',
          'Randolph', 'Lee']
          District 5 has population 682766 and contains counties ['Marengo', 'Baldwin', 'Mobile',
          'Washington', 'Clarke', 'Monroe']
          District 6 has population 682691 and contains counties ['Barbour', 'Houston', 'Conecuh',
          'Henry', 'Crenshaw', 'Covington', 'Butler', 'Dale', 'Geneva', 'Wilcox', 'Pike', 'Coffe
          e', 'Montgomery', 'Escambia', 'Lowndes']
In [24]:
           # Let's draw it on a map
           import geopandas as gpd
In [25]:
           # Read Oklahoma county shapefile from "OK county.shp"
           filename = 'AL_counties.shp'
           # Read geopandas dataframe from file
           df = gpd.read file( filename )
In [26]:
           # Which district is each county assigned to?
```

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```
assignment = [ -1 for u in G.nodes ]

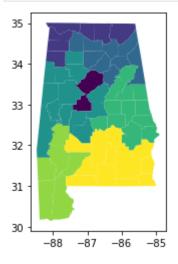
# for each district j
for j in range(len(districts)):

# for each node i in this district
for i in districts[j]:

# What is its GEOID?
geoID = G.nodes[i]["GEOID10"]

# Need to find this GEOID in the dataframe
for u in G.nodes:
    if geoID == df['GEOID10'][u]: # Found it
        assignment[u] = j # Node u from the dataframe should be assigned to dis

# Now add the assignments to a column of the dataframe and map it
df['assignment'] = assignment
my_fig = df.plot(column='assignment').get_figure()
```



In [ ]: