

```
In [44]: import os
import json
from networkx.readwrite import json_graph

def read_graph_from_json(json_file):
    with open(json_file) as f:
        data = json.load(f)
    return json_graph.adjacency_graph(data)
```

```
In [45]: from gerrychain import Graph

#read the iowa json county graph
# Define the file path and file name
filepath = '/Users/adenijibabalola/Desktop/DeterministicOperationsResearch/I
filename = 'IA_counties.json'

# Use os.path.join to correctly construct the file path
full_path = os.path.join(filepath,filename)
# Load the graph
G = Graph.from_json(full_path)

#Make sure the file is read correctly by printing the node #, and its popula
for node in G.nodes:
    print(node, G.nodes[node].get('POP100', 'Population not found'))
```

```
0 12943
1 10330
2 12138
3 10033
4 98537
5 10565
6 12329
7 20070
8 492401
9 9814
10 40105
11 17488
12 9110
13 15627
14 5674
15 20646
16 16878
17 26715
18 37813
19 25698
20 93667
21 5896
22 15211
23 11658
24 38910
25 174669
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26 33555
27 6605
28 16662
29 14334
30 9388
31 105941
32 16384
33 16525
34 20565
35 6192
36 43127
37 8771
38 35437
39 20482
40 11934
41 3704
42 8751
43 9597
44 10837
45 17135
46 7577
47 43235
48 8634
49 15039
50 10019
51 24988
52 14182
53 10623
54 22565
55 35872
56 14061
57 10679
58 7005
59 13127
60 7645
61 18505
62 20760
63 7078
64 14828
65 14582
66 36999
67 15663
68 16548
69 9469
70 8996
71 4663
72 46460
73 9927
74 7443
75 131144
76 9748
77 19509
78 52403

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79 7496
80 12012
81 10795
82 99266
83 18662
84 11746
85 25575
86 6497
87 19485
88 230299
89 17703
90 17043
91 20823
92 14484
93 99678
94 33414
95 22190
96 12317
97 7203
98 152854

```

```

In [47]: ## Let's impose a 1% population deviation (+/-0.5%)
         deviation = 0.01

         import math
         # No of districts
         k = 4
         #Summing up the total population in counties
         total_population = sum(G.nodes[node]['POP100'] for node in G.nodes)

         # Equations for bounds
         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 = 793605 and U = 801580 and K = 4

```

In [48]: # Begin with model
         import gurobipy as gp
         from gurobipy import GRB

         # create model
         m = gp.Model()

         # create variables
         x = m.addVars(G.nodes, k, vtype=GRB.BINARY) # x[i,j] equals one when county
         y = m.addVars(G.edges, vtype=GRB.BINARY)    # y[u,v] equals one when edge {u

         # objective is to minimize cut edges
         m.setObjective( gp.quicksum( y[u,v] for u,v in G.edges ), GRB.MINIMIZE )

```

```
In [49]: # add constraints saying that each county i is assigned to one district
m.addConstrs( gp.quicksum(x[i,j] for j in range(k)) == 1 for i in G.nodes)

# add constraints saying that each district has population at least L and at
m.addConstrs( gp.quicksum( G.nodes[i]['POP100'] * x[i,j] for i in G.nodes) >
m.addConstrs( gp.quicksum( G.nodes[i]['POP100'] * x[i,j] for i in G.nodes) <

# add constraints saying that edge {u,v} is cut if u is assigned to district
m.addConstrs( x[u,j] - x[v,j] <= y[u,v] for u,v in G.edges for j in range(k)

m.update()
```

```
In [50]: # Now, let's add contiguity constraints and re-solve the model.
# We will use the contiguity constraints of Hojny et al. (MPC, 2021)
# https://link.springer.com/article/10.1007/s12532-020-00186-3

# Add root variables: r[i,j] equals 1 if node i is the "root" of district j
r = m.addVars( G.nodes, k, vtype=GRB.BINARY )

# Add flow variables: f[u,v] = amount of flow sent across arc uv
# Flows are sent across arcs of the directed version of G which we call DG

import networkx as nx
DG = nx.DiGraph(G)          # directed version of G

f = m.addVars( DG.edges )
```

```
In [51]: # The big-M proposed by Hojny et al.
M = G.number_of_nodes() - k + 1

# Each district j should have one root
m.addConstrs( gp.quicksum( r[i,j] for i in G.nodes ) == 1 for j in range(k)

# If node i is not assigned to district j, then it cannot be its root
m.addConstrs( r[i,j] <= x[i,j] for i in G.nodes for j in range(k) )

# if not a root, consume some flow.
# if a root, only send out (so much) flow.
m.addConstrs( gp.quicksum( f[j,i] - f[i,j] for j in G.neighbors(i) )
               >= 1 - M * gp.quicksum( r[i,j] for j in range(k) ) for i in G.n

# do not send flow across cut edges
m.addConstrs( f[i,j] + f[j,i] <= M * ( 1 - y[i,j] ) for i,j in G.edges )

m.update()
```

```
In [52]: # solve IP model
m.optimize()
```

Gurobi Optimizer version 10.0.3 build v10.0.3rc0 (mac64[arm])

CPU model: Apple M2

Thread count: 8 physical cores, 8 logical processors, using up to 8 threads

Optimize a model with 1716 rows, 1458 columns and 6990 nonzeros

Model fingerprint: 0xe75d93b4

Variable types: 444 continuous, 1014 integer (1014 binary)

Coefficient statistics:

Matrix range [1e+00, 5e+05]

Objective range [1e+00, 1e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 8e+05]

Presolve time: 0.02s

Presolved: 1716 rows, 1458 columns, 6990 nonzeros

Variable types: 444 continuous, 1014 integer (1014 binary)

Root relaxation: objective 0.000000e+00, 745 iterations, 0.02 seconds (0.02 work units)

Nodes			Current Node			Objective Bounds			Work	
Expl	Unexpl		Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
	0	0	0.00000		0 412	-	0.00000	-	-	0s
	0	0	0.13229		0 420	-	0.13229	-	-	0s
	0	0	3.89100		0 415	-	3.89100	-	-	0s
	0	0	4.39046		0 425	-	4.39046	-	-	0s
	0	0	4.44167		0 429	-	4.44167	-	-	0s
	0	0	4.45279		0 427	-	4.45279	-	-	0s
	0	0	4.46568		0 430	-	4.46568	-	-	0s
	0	0	4.49784		0 426	-	4.49784	-	-	0s
	0	0	4.50690		0 424	-	4.50690	-	-	0s
	0	0	4.50918		0 423	-	4.50918	-	-	0s
	0	0	4.50918		0 421	-	4.50918	-	-	0s
	0	0	4.50918		0 421	-	4.50918	-	-	0s
	0	0	4.50918		0 421	-	4.50918	-	-	0s
	0	0	4.50918		0 421	-	4.50918	-	-	0s
	0	2	4.52615		0 421	-	4.52615	-	-	0s
H	785	663				37.0000000	9.01575	75.6%	152	2s
H	1327	854				35.0000000	10.14711	71.0%	164	4s
	1502	977	32.68564	28	435	35.00000	10.14711	71.0%	165	5s
	2360	1336	19.52204	19	418	35.00000	14.67644	58.1%	175	10s
	4335	2033	20.03843	22	414	35.00000	17.14163	51.0%	172	15s
	7097	3762	26.60019	25	436	35.00000	18.60925	46.8%	168	20s
	9960	5399	23.49082	21	439	35.00000	19.48936	44.3%	165	25s
	13963	7646	33.28237	41	306	35.00000	20.23112	42.2%	164	30s
	14504	9084	27.38567	30	311	35.00000	20.33970	41.9%	164	37s
	19191	10621	32.84419	44	330	35.00000	21.05841	39.8%	162	40s
	23175	12813	30.29846	32	380	35.00000	21.56821	38.4%	160	45s
	26498	14540	28.14321	34	315	35.00000	21.89176	37.5%	159	50s
	29071	15916	32.35565	36	429	35.00000	22.09372	36.9%	158	64s
	30122	16266	cutoff	38		35.00000	22.18979	36.6%	157	65s
	34888	18687	32.79401	34	341	35.00000	22.54188	35.6%	155	70s
	38969	20681	32.93731	30	315	35.00000	22.80566	34.8%	154	75s
	42100	22306	28.07201	30	426	35.00000	22.98507	34.3%	153	80s

46246	24537	cutoff	51		35.00000	23.22425	33.6%	153	85s
49482	26189	26.85835	28	419	35.00000	23.38478	33.2%	151	90s
54408	28418	cutoff	37		35.00000	23.62722	32.5%	150	95s
58474	30390	27.77101	34	373	35.00000	23.79131	32.0%	150	101s
60668	31456	29.05549	30	292	35.00000	23.87242	31.8%	149	105s
64400	33318	31.31166	36	379	35.00000	24.01657	31.4%	148	110s
68752	35315	25.80558	29	190	35.00000	24.13502	31.0%	148	115s
72945	37403	25.87921	30	336	35.00000	24.25644	30.7%	147	120s
76188	39009	33.33893	42	280	35.00000	24.34214	30.5%	146	125s
80382	41086	30.45402	34	315	35.00000	24.46442	30.1%	146	130s
84743	43145	cutoff	43		35.00000	24.55677	29.8%	145	135s
89031	45097	cutoff	29		35.00000	24.66427	29.5%	144	140s
93756	47391	31.24341	37	308	35.00000	24.78137	29.2%	143	145s
96840	48869	26.53343	32	286	35.00000	24.84809	29.0%	143	150s
101786	51074	33.69711	37	263	35.00000	24.96028	28.7%	143	155s
106057	52806	32.12102	32	397	35.00000	25.04343	28.4%	142	160s
110181	54551	28.80701	28	448	35.00000	25.11854	28.2%	142	165s
114393	56458	29.55163	31	300	35.00000	25.19519	28.0%	141	170s
119278	58480	29.41319	34	231	35.00000	25.27284	27.8%	141	175s
123210	60141	cutoff	54		35.00000	25.33620	27.6%	140	180s
127050	61915	27.17192	30	371	35.00000	25.39278	27.4%	140	185s
131403	63675	31.24369	38	204	35.00000	25.45028	27.3%	139	190s
135763	65508	33.58586	34	347	35.00000	25.52004	27.1%	139	195s
139555	67175	27.25432	31	386	35.00000	25.57117	26.9%	139	200s
144519	69464	cutoff	33		35.00000	25.64277	26.7%	138	205s
148895	71098	cutoff	39		35.00000	25.69861	26.6%	138	210s
152568	72718	28.51571	33	318	35.00000	25.74158	26.5%	137	215s
157662	74830	cutoff	42		35.00000	25.79969	26.3%	137	220s
162742	76965	32.70012	32	399	35.00000	25.85932	26.1%	136	225s
166891	78724	33.54547	39	217	35.00000	25.90103	26.0%	136	230s
170712	80275	32.62345	37	371	35.00000	25.93956	25.9%	136	235s
175345	82133	30.95576	41	236	35.00000	25.98855	25.7%	135	240s
179599	84023	31.78475	31	305	35.00000	26.03657	25.6%	135	245s
184214	85981	30.87069	33	392	35.00000	26.07887	25.5%	135	250s
189157	88077	28.94397	32	383	35.00000	26.13242	25.3%	135	255s
192194	89525	28.94644	28	388	35.00000	26.15756	25.3%	134	260s
197047	91438	27.95455	32	397	35.00000	26.19905	25.1%	134	265s
200485	92835	cutoff	34		35.00000	26.23501	25.0%	134	270s
205526	94914	29.32311	28	411	35.00000	26.28537	24.9%	133	275s
209886	96740	cutoff	51		35.00000	26.32722	24.8%	133	280s
214782	98525	cutoff	35		35.00000	26.37682	24.6%	133	285s
218788	100092	cutoff	41		35.00000	26.40910	24.5%	133	290s
223032	101818	33.84527	35	254	35.00000	26.44607	24.4%	132	295s
227429	103527	32.22844	43	243	35.00000	26.47672	24.4%	132	300s
231968	105259	cutoff	36		35.00000	26.51263	24.2%	132	305s
235969	106841	31.90975	34	393	35.00000	26.54838	24.1%	132	310s
240461	108458	30.70625	28	409	35.00000	26.58113	24.1%	131	315s
245234	110428	29.85883	35	309	35.00000	26.62007	23.9%	131	320s
248987	111753	31.21761	39	328	35.00000	26.64634	23.9%	131	325s
253234	113397	31.56452	40	369	35.00000	26.68020	23.8%	131	330s
257143	114892	32.37363	37	383	35.00000	26.70660	23.7%	131	335s
261369	116472	33.90007	35	340	35.00000	26.73480	23.6%	131	340s
265424	118042	32.22041	38	343	35.00000	26.76521	23.5%	130	345s

270312	119799	30.71576	29	345	35.00000	26.79471	23.4%	130	350s
274414	121400	29.90779	32	313	35.00000	26.82340	23.4%	130	355s
278422	122888	cutoff	43		35.00000	26.84936	23.3%	130	360s
282768	124672	32.43335	39	259	35.00000	26.87829	23.2%	130	365s
286925	126274	29.00347	32	358	35.00000	26.90545	23.1%	129	370s
290813	127708	30.75098	33	409	35.00000	26.93083	23.1%	129	375s
295140	129443	cutoff	32		35.00000	26.95543	23.0%	129	380s
299301	130780	cutoff	33		35.00000	26.97987	22.9%	129	385s
304078	132595	cutoff	32		35.00000	27.01039	22.8%	129	390s
308179	133890	cutoff	40		35.00000	27.03179	22.8%	128	395s
312292	135195	cutoff	32		35.00000	27.05415	22.7%	128	400s
316401	136562	29.33346	33	353	35.00000	27.08089	22.6%	128	405s
319824	138048	cutoff	40		35.00000	27.09853	22.6%	128	410s
322758	139201	28.33413	28	388	35.00000	27.11371	22.5%	128	415s
327045	140899	28.15586	41	290	35.00000	27.13822	22.5%	128	420s
331052	142220	33.63282	32	330	35.00000	27.16033	22.4%	128	425s
334899	143596	cutoff	29		35.00000	27.18076	22.3%	128	430s
339090	145069	28.97371	35	361	35.00000	27.20520	22.3%	127	435s
343581	146429	32.48662	34	358	35.00000	27.22739	22.2%	127	440s
347377	147831	cutoff	45		35.00000	27.25202	22.1%	127	445s
352465	149724	30.86873	29	428	35.00000	27.28155	22.1%	127	450s
356701	151131	cutoff	35		35.00000	27.30239	22.0%	127	455s
360667	152578	33.17712	39	285	35.00000	27.32506	21.9%	127	460s
364569	154051	cutoff	37		35.00000	27.34632	21.9%	127	465s
368819	155375	33.91904	36	333	35.00000	27.36679	21.8%	126	470s
373374	157025	30.69474	40	268	35.00000	27.38797	21.7%	126	475s
376941	157972	31.96930	38	344	35.00000	27.40422	21.7%	126	480s
380382	159516	33.18478	30	424	35.00000	27.42193	21.7%	126	485s
384884	160897	cutoff	42		35.00000	27.44218	21.6%	126	490s
389235	162377	32.91974	39	321	35.00000	27.45905	21.5%	126	495s
394010	163978	31.36474	35	426	35.00000	27.48212	21.5%	126	500s
397755	165187	30.91252	34	327	35.00000	27.49911	21.4%	126	505s
402025	166644	30.71045	31	362	35.00000	27.51894	21.4%	125	510s
406106	168104	29.46169	28	439	35.00000	27.53531	21.3%	125	515s
410111	169382	29.25569	32	290	35.00000	27.55311	21.3%	125	520s
414419	170669	cutoff	37		35.00000	27.57077	21.2%	125	525s
418131	171982	cutoff	29		35.00000	27.58494	21.2%	125	530s
423010	173637	cutoff	37		35.00000	27.60554	21.1%	125	535s
427248	174980	33.98781	36	337	35.00000	27.62363	21.1%	125	540s
431548	176285	cutoff	34		35.00000	27.64061	21.0%	125	545s
435463	177577	30.81218	35	312	35.00000	27.65583	21.0%	125	550s
439763	179102	cutoff	42		35.00000	27.67254	20.9%	124	555s
445182	180921	32.60452	38	375	35.00000	27.69168	20.9%	124	560s
449251	182122	cutoff	41		35.00000	27.70703	20.8%	124	565s
453368	183353	30.09552	34	331	35.00000	27.72575	20.8%	124	570s
457560	184767	cutoff	36		35.00000	27.73986	20.7%	124	575s
462756	186513	cutoff	45		35.00000	27.76272	20.7%	124	580s
466703	187820	29.86952	32	367	35.00000	27.77676	20.6%	124	585s
471836	189392	32.39715	38	437	35.00000	27.79473	20.6%	123	591s
475373	190379	cutoff	46		35.00000	27.80810	20.5%	123	595s
479329	191636	29.53675	35	373	35.00000	27.82523	20.5%	123	600s
483611	192818	29.76693	30	328	35.00000	27.83996	20.5%	123	605s
486813	193826	30.85290	34	260	35.00000	27.85110	20.4%	123	610s

491158	195004	33.60104	42	293	35.00000	27.86524	20.4%	123	615s
495869	196495	cutoff	34		35.00000	27.87984	20.3%	123	620s
500083	197834	29.83190	41	264	35.00000	27.89529	20.3%	123	625s
504511	199206	29.55898	37	313	35.00000	27.90976	20.3%	123	630s
508579	200271	cutoff	36		35.00000	27.92433	20.2%	123	635s
512195	201574	cutoff	37		35.00000	27.93612	20.2%	123	640s
517381	203128	30.33917	29	320	35.00000	27.95152	20.1%	122	645s
520824	204109	33.96907	34	297	35.00000	27.96264	20.1%	122	650s
526087	205584	28.91999	29	399	35.00000	27.97896	20.1%	122	655s
529504	206763	29.37531	35	322	35.00000	27.99011	20.0%	122	660s
534159	208105	30.71242	32	241	35.00000	28.00458	20.0%	122	665s
538172	209336	31.42570	32	372	35.00000	28.01619	20.0%	122	671s
542274	210431	33.74150	46	215	35.00000	28.03013	19.9%	122	675s
546113	211522	32.05245	35	313	35.00000	28.04409	19.9%	122	680s
550438	212843	cutoff	37		35.00000	28.05733	19.8%	122	685s
553258	213678	30.97821	27	345	35.00000	28.06657	19.8%	122	690s
557609	215028	cutoff	41		35.00000	28.08105	19.8%	122	695s
562132	216303	32.39627	37	408	35.00000	28.09380	19.7%	122	700s
566533	217585	31.98235	40	343	35.00000	28.10655	19.7%	121	705s
570816	218699	32.48763	33	433	35.00000	28.12020	19.7%	121	710s
575157	219830	33.25283	41	286	35.00000	28.13192	19.6%	121	715s
579487	221141	32.23223	32	399	35.00000	28.14552	19.6%	121	720s
583419	222336	cutoff	36		35.00000	28.15750	19.6%	121	725s
587767	223502	32.76089	32	338	35.00000	28.17146	19.5%	121	730s
592946	225033	infeasible	45		35.00000	28.18732	19.5%	121	736s
597209	225967	32.78623	36	299	35.00000	28.19910	19.4%	121	740s
600546	226915	cutoff	45		35.00000	28.20743	19.4%	121	745s
605200	228115	30.67344	34	368	35.00000	28.22286	19.4%	121	750s
609538	229370	cutoff	44		35.00000	28.23693	19.3%	121	755s
613511	230470	31.83382	39	225	35.00000	28.24924	19.3%	121	760s
618731	232142	cutoff	32		35.00000	28.26505	19.2%	120	765s
623279	233395	cutoff	35		35.00000	28.27565	19.2%	120	770s
626630	234368	32.40789	34	371	35.00000	28.28643	19.2%	120	775s
631198	235540	33.74528	36	227	35.00000	28.29707	19.2%	120	780s
636491	237129	31.79914	40	286	35.00000	28.31176	19.1%	120	785s
H637643	218540				34.0000000	28.31426	16.7%	120	788s
637654	219480	cutoff	44		34.00000	28.31464	16.7%	120	794s
642790	219504	31.95443	31	301	34.00000	28.33195	16.7%	120	796s
645935	220249	29.24911	38	242	34.00000	28.34214	16.6%	120	800s
649715	220958	29.92034	43	196	34.00000	28.35474	16.6%	120	805s
654263	221814	31.72929	36	310	34.00000	28.36898	16.6%	120	810s
658582	222560	cutoff	45		34.00000	28.38290	16.5%	120	815s
662748	223229	31.43541	32	248	34.00000	28.39622	16.5%	120	820s
666722	224002	31.01846	29	371	34.00000	28.40925	16.4%	120	825s
671144	224814	32.14599	27	389	34.00000	28.42244	16.4%	120	830s
676314	225667	32.22421	38	326	34.00000	28.43757	16.4%	119	836s
679634	226334	cutoff	33		34.00000	28.44890	16.3%	119	840s
684085	227079	30.88490	29	351	34.00000	28.46285	16.3%	119	845s
688516	227811	32.31138	35	272	34.00000	28.47479	16.3%	119	850s
692871	228576	31.01274	38	385	34.00000	28.48884	16.2%	119	855s
697083	229367	32.04092	32	402	34.00000	28.50169	16.2%	119	860s
701494	230131	cutoff	32		34.00000	28.51543	16.1%	119	865s
705759	230824	cutoff	37		34.00000	28.52964	16.1%	119	870s

709985	231507	32.42137	33	353	34.00000	28.54214	16.1%	119	875s
712174	231974	cutoff	47		34.00000	28.54826	16.0%	119	880s
716847	232664	32.79425	43	294	34.00000	28.56068	16.0%	119	885s
720646	233343	30.05282	36	443	34.00000	28.57151	16.0%	119	890s
725169	233982	30.93792	34	381	34.00000	28.58488	15.9%	119	896s
728798	234538	30.53164	41	347	34.00000	28.59450	15.9%	119	900s
733124	235142	cutoff	34		34.00000	28.60726	15.9%	119	905s
737481	235729	31.78585	36	195	34.00000	28.61908	15.8%	118	910s
741861	236443	cutoff	37		34.00000	28.63166	15.8%	118	915s
746497	237111	cutoff	38		34.00000	28.64430	15.8%	118	920s
750850	237728	28.94301	31	314	34.00000	28.65557	15.7%	118	925s
755511	238293	cutoff	37		34.00000	28.66865	15.7%	118	930s
760019	238847	32.80348	36	277	34.00000	28.68029	15.6%	118	935s
764264	239430	32.48031	34	412	34.00000	28.69279	15.6%	118	940s
768504	239996	32.39182	41	107	34.00000	28.70385	15.6%	118	945s
772210	240515	cutoff	37		34.00000	28.71411	15.5%	118	950s
776365	241056	31.94366	31	309	34.00000	28.72627	15.5%	118	955s
780631	241651	cutoff	31		34.00000	28.73771	15.5%	118	960s
785054	242370	32.64039	40	249	34.00000	28.75065	15.4%	118	965s
788024	242755	31.31489	32	306	34.00000	28.75796	15.4%	118	970s
792545	243385	30.70009	31	387	34.00000	28.77033	15.4%	118	975s
796364	243948	32.38027	29	402	34.00000	28.78113	15.3%	118	980s
800914	244570	32.43753	32	422	34.00000	28.79343	15.3%	118	985s
805265	245119	30.62013	33	363	34.00000	28.80494	15.3%	117	990s
809623	245706	31.82100	32	373	34.00000	28.81670	15.2%	117	995s
814238	246271	cutoff	37		34.00000	28.82722	15.2%	117	1000s
817621	246695	32.71006	36	293	34.00000	28.83586	15.2%	117	1005s
821525	247204	30.80031	38	291	34.00000	28.84612	15.2%	117	1010s
826143	247726	29.15866	35	372	34.00000	28.85963	15.1%	117	1015s
829985	248183	31.63050	33	335	34.00000	28.87063	15.1%	117	1020s
834688	248811	31.70993	35	373	34.00000	28.88349	15.0%	117	1025s
839048	249365	30.36886	42	319	34.00000	28.89455	15.0%	117	1030s
843589	249879	cutoff	31		34.00000	28.90599	15.0%	117	1035s
847018	250319	32.74612	31	341	34.00000	28.91444	15.0%	117	1040s
851792	250941	31.89926	44	320	34.00000	28.92631	14.9%	117	1045s
856349	251394	cutoff	35		34.00000	28.93728	14.9%	117	1050s
860853	252008	30.95189	37	316	34.00000	28.94824	14.9%	117	1055s
865461	252608	32.21538	32	395	34.00000	28.95898	14.8%	117	1060s
869962	253137	cutoff	33		34.00000	28.97112	14.8%	117	1065s
873717	253616	32.38655	31	461	34.00000	28.98012	14.8%	117	1070s
878227	254002	32.38304	45	347	34.00000	28.99048	14.7%	116	1075s
882802	254591	30.63113	32	406	34.00000	29.00126	14.7%	116	1080s
887167	255117	30.37983	37	300	34.00000	29.01166	14.7%	116	1085s
891918	255658	32.44275	38	318	34.00000	29.02188	14.6%	116	1090s
896200	256143	32.53225	44	166	34.00000	29.03236	14.6%	116	1095s
900634	256752	31.73896	33	425	34.00000	29.04292	14.6%	116	1100s
904526	257138	32.80578	38	371	34.00000	29.05242	14.6%	116	1105s
909073	257644	cutoff	34		34.00000	29.06346	14.5%	116	1110s
913312	258065	cutoff	34		34.00000	29.07354	14.5%	116	1115s
917924	258610	31.22891	30	427	34.00000	29.08380	14.5%	116	1120s
922640	259161	32.86844	44	254	34.00000	29.09465	14.4%	116	1125s
927058	259572	31.26740	45	283	34.00000	29.10434	14.4%	116	1130s
931456	260067	32.56347	36	267	34.00000	29.11315	14.4%	116	1135s

936229	260567	32.20304	31	302	34.00000	29.12430	14.3%	116	1140s
H939064	202446				33.0000000	29.13108	11.7%	116	1143s
939860	202381	cutoff	32		33.00000	29.13393	11.7%	116	1145s
941165	202300	31.43135	36	386	33.00000	29.13806	11.7%	116	1150s
944972	202173	cutoff	36		33.00000	29.15300	11.7%	116	1155s
949600	201937	cutoff	35		33.00000	29.16958	11.6%	115	1160s
954460	201785	cutoff	30		33.00000	29.18653	11.6%	115	1165s
959161	201602	cutoff	35		33.00000	29.20214	11.5%	115	1170s
962769	201440	31.26730	39	255	33.00000	29.21458	11.5%	115	1175s
968209	201261	cutoff	36		33.00000	29.23479	11.4%	115	1181s
972531	201019	cutoff	42		33.00000	29.25063	11.4%	115	1185s
976285	200845	cutoff	34		33.00000	29.26393	11.3%	115	1190s
981204	200689	31.70201	34	334	33.00000	29.27973	11.3%	115	1195s
984759	200499	30.97863	32	447	33.00000	29.29105	11.2%	115	1200s
989386	200269	31.94406	32	375	33.00000	29.30630	11.2%	115	1205s
993788	199954	31.77997	40	272	33.00000	29.32270	11.1%	115	1210s
998306	199647	30.92026	31	268	33.00000	29.33811	11.1%	115	1215s
1002831	199366	cutoff	37		33.00000	29.35283	11.1%	115	1220s
S									
1007362	199073	cutoff	38		33.00000	29.36870	11.0%	115	1225s
S									
1011808	198736	31.11063	33	394	33.00000	29.38358	11.0%	115	1230s
S									
1016485	198437	31.87153	31	407	33.00000	29.39989	10.9%	115	1235s
S									
1021087	198143	31.94314	29	420	33.00000	29.41476	10.9%	114	1240s
S									
1025698	197835	cutoff	38		33.00000	29.43071	10.8%	114	1245s
S									
1030157	197512	30.46166	46	286	33.00000	29.44588	10.8%	114	1250s
S									
1034645	197169	cutoff	35		33.00000	29.46084	10.7%	114	1255s
S									
1039361	196777	cutoff	35		33.00000	29.47640	10.7%	114	1260s
S									
1043828	196395	30.61832	37	372	33.00000	29.49098	10.6%	114	1265s
S									
1048567	195974	cutoff	33		33.00000	29.50765	10.6%	114	1270s
S									
1053017	195614	31.74468	36	318	33.00000	29.52422	10.5%	114	1277s
S									
1055002	195330	31.82572	35	363	33.00000	29.53062	10.5%	114	1280s
S									
1059659	194932	30.60691	30	363	33.00000	29.54706	10.5%	114	1285s
S									
1064014	194469	cutoff	36		33.00000	29.56214	10.4%	114	1290s
S									
1067913	194085	31.76121	36	278	33.00000	29.57493	10.4%	114	1295s
S									
1072708	193547	31.42854	37	427	33.00000	29.59176	10.3%	114	1300s
S									
1077388	193121	31.44226	42	185	33.00000	29.60726	10.3%	114	1305s
S									

1082136	192611	cutoff	34		33.00000	29.62272	10.2%	114	1310
S									
1086715	192122	cutoff	43		33.00000	29.63786	10.2%	114	1315
S									
1091505	191658	30.97463	39	363	33.00000	29.65275	10.1%	113	1320
S									
1096241	191045	cutoff	35		33.00000	29.66877	10.1%	113	1325
S									
1100838	190464	cutoff	32		33.00000	29.68382	10.0%	113	1330
S									
1105774	189901	31.85304	31	302	33.00000	29.69940	10.0%	113	1335
S									
1110533	189273	30.89193	43	312	33.00000	29.71395	10.0%	113	1340
S									
1115173	188657	cutoff	32		33.00000	29.72920	9.91%	113	1346
S									
1119036	188205	31.36064	29	387	33.00000	29.74089	9.88%	113	1350
S									
1123844	187521	31.91942	33	342	33.00000	29.75701	9.83%	113	1355
S									
1128305	186846	31.35067	31	325	33.00000	29.77221	9.78%	113	1360
S									
1133338	186179	31.42791	46	213	33.00000	29.78820	9.73%	113	1365
S									
1137911	185503	31.92634	46	191	33.00000	29.80406	9.68%	113	1370
S									
1142924	184879	cutoff	35		33.00000	29.81970	9.64%	113	1375
S									
1146615	184289	31.44727	31	342	33.00000	29.83057	9.60%	113	1380
S									
1151492	183497	cutoff	38		33.00000	29.84676	9.56%	113	1385
S									
1156227	182758	cutoff	32		33.00000	29.86298	9.51%	113	1390
S									
1160992	181945	31.83515	31	273	33.00000	29.87843	9.46%	112	1395
S									
1165872	181090	cutoff	38		33.00000	29.89474	9.41%	112	1400
S									
1169971	180486	cutoff	33		33.00000	29.90871	9.37%	112	1405
S									
1174247	179659	cutoff	40		33.00000	29.92181	9.33%	112	1410
S									
1179402	178799	cutoff	35		33.00000	29.93908	9.28%	112	1415
S									
1184064	177921	cutoff	40		33.00000	29.95556	9.23%	112	1420
S									
1188724	177136	cutoff	35		33.00000	29.97063	9.18%	112	1425
S									
1192898	176424	31.69949	31	343	33.00000	29.98419	9.14%	112	1430
S									
1197888	175555	30.42157	32	340	33.00000	30.00085	9.09%	112	1435
S									
1202993	174612	31.76865	44	199	33.00000	30.01722	9.04%	112	1440

S										
	1207795	173705	cutoff	36		33.00000	30.03351	8.99%	112	1445
S										
	1212671	172702	cutoff	33		33.00000	30.05018	8.94%	112	1450
S										
	1217678	171709	cutoff	36		33.00000	30.06635	8.89%	112	1455
S										
	1221518	170937	cutoff	42		33.00000	30.07913	8.85%	112	1460
S										
	1226693	169888	31.06893	38	225	33.00000	30.09540	8.80%	112	1465
S										
	1231357	168938	cutoff	33		33.00000	30.11073	8.76%	111	1470
S										
	1235383	168151	cutoff	33		33.00000	30.12544	8.71%	111	1475
S										
	1239776	167035	cutoff	31		33.00000	30.14126	8.66%	111	1480
S										
	1244982	165970	infeasible	39		33.00000	30.15835	8.61%	111	1485
S										
	1250194	164876	31.98217	38	370	33.00000	30.17434	8.56%	111	1490
S										
	1254077	163982	cutoff	35		33.00000	30.18662	8.53%	111	1495
S										
	1259108	162847	31.42824	36	155	33.00000	30.20360	8.47%	111	1500
S										
	1263953	161621	cutoff	28		33.00000	30.21962	8.43%	111	1505
S										
	1267545	160753	cutoff	32		33.00000	30.23247	8.39%	111	1510
S										
	1272567	159448	cutoff	37		33.00000	30.25000	8.33%	111	1515
S										
	1276686	158380	cutoff	33		33.00000	30.26411	8.29%	111	1520
S										
	1281720	157063	cutoff	30		33.00000	30.28190	8.24%	111	1525
S										
	1287102	155618	31.79839	35	269	33.00000	30.30084	8.18%	111	1530
S										
	1291041	154480	cutoff	45		33.00000	30.31494	8.14%	111	1535
S										
	1296414	153018	31.99021	27	352	33.00000	30.33395	8.08%	110	1540
S										
	1301532	151588	31.74992	45	343	33.00000	30.35202	8.02%	110	1545
S										
	1305585	150438	31.78531	33	362	33.00000	30.36553	7.98%	110	1550
S										
	1310810	148935	cutoff	49		33.00000	30.38397	7.93%	110	1555
S										
	1316130	147236	31.65126	30	335	33.00000	30.40444	7.87%	110	1560
S										
	1321400	145640	cutoff	42		33.00000	30.42383	7.81%	110	1565
S										
	1326542	144140	cutoff	35		33.00000	30.44182	7.75%	110	1570
S										

1331697	142760	cutoff	32		33.00000	30.46235	7.69%	110	1577
S									
1333869	141782	31.94244	38	361	33.00000	30.47051	7.67%	110	1580
S									
1340213	139688	cutoff	27		33.00000	30.49448	7.59%	110	1585
S									
1344182	138394	cutoff	34		33.00000	30.50941	7.55%	110	1590
S									
1349421	136645	30.87259	46	177	33.00000	30.52891	7.49%	110	1595
S									
1353845	135173	cutoff	37		33.00000	30.54499	7.44%	109	1600
S									
1357897	133772	31.63878	34	352	33.00000	30.56166	7.39%	109	1605
S									
1363460	131868	cutoff	41		33.00000	30.58187	7.33%	109	1610
S									
1367151	130473	cutoff	34		33.00000	30.59690	7.28%	109	1615
S									
1372565	128450	31.78089	42	265	33.00000	30.61781	7.22%	109	1620
S									
1377835	126443	cutoff	40		33.00000	30.63917	7.15%	109	1625
S									
1381945	124807	cutoff	41		33.00000	30.65656	7.10%	109	1630
S									
1386543	123005	cutoff	34		33.00000	30.67350	7.05%	109	1635
S									
1390643	121291	31.57359	44	184	33.00000	30.69254	6.99%	109	1640
S									
1395774	119139	cutoff	33		33.00000	30.71602	6.92%	109	1645
S									
1401215	116786	31.10054	39	163	33.00000	30.73900	6.85%	109	1650
S									
1406655	114447	cutoff	45		33.00000	30.76237	6.78%	108	1655
S									
1411936	112079	cutoff	42		33.00000	30.78479	6.71%	108	1660
S									
1418296	109137	cutoff	28		33.00000	30.81410	6.62%	108	1665
S									
1423534	106561	cutoff	35		33.00000	30.83935	6.55%	108	1670
S									
1429132	103854	31.17267	35	261	33.00000	30.86564	6.47%	108	1675
S									
1434371	101325	cutoff	37		33.00000	30.89064	6.39%	108	1680
S									
1439470	98695	cutoff	30		33.00000	30.91733	6.31%	108	1685s
1444537	96023	cutoff	47		33.00000	30.94344	6.23%	108	1690s
1450944	92668	cutoff	32		33.00000	30.97594	6.13%	108	1695s
1456372	89687	31.32656	46	344	33.00000	31.00545	6.04%	107	1700s
1462618	86152	31.73002	33	348	33.00000	31.03859	5.94%	107	1705s
1468928	82539	cutoff	39		33.00000	31.07354	5.84%	107	1710s
1475437	78614	cutoff	40		33.00000	31.11195	5.72%	107	1715s
1481596	74813	cutoff	34		33.00000	31.14905	5.61%	107	1720s
1488766	70302	cutoff	30		33.00000	31.19352	5.47%	107	1725s

1494905	66249	cutoff	30	33.00000	31.23432	5.35%	106	1730s	
1501874	61297	cutoff	33	33.00000	31.28269	5.20%	106	1735s	
1509014	56126	cutoff	40	33.00000	31.33756	5.04%	106	1740s	
1516047	50790	cutoff	36	33.00000	31.39368	4.87%	106	1745s	
1524543	43996	cutoff	38	33.00000	31.46423	4.65%	105	1750s	
1531302	38470	31.81784	42	296	33.00000	31.52435	4.47%	105	1755s
1541216	29917	cutoff	41	33.00000	31.62003	4.18%	105	1760s	
1550755	21267	cutoff	41	33.00000	31.72209	3.87%	104	1765s	
1564598	8198	cutoff	40	33.00000	31.88192	3.39%	104	1770s	

Cutting planes:

Gomory: 29

Flow cover: 8

Zero half: 1

RLT: 22

Explored 1573757 nodes (162462278 simplex iterations) in 1772.22 seconds (33 25.92 work units)

Thread count was 8 (of 8 available processors)

Solution count 5: 33 33 34 ... 37

Optimal solution found (tolerance 1.00e-04)

Best objective 3.3000000000000e+01, best bound 3.3000000000000e+01, gap 0.000 0%

```
In [53]: print("The number of cut edges is",m.objval)

# retrieve the districts and their populations
districts = [ [i for i in G.nodes if x[i,j].x > 0.5] for j in range(k)]
district_counties = [ [ G.nodes[i]["NAMELSAD20"] for i in districts[j] ] for
district_populations = [ sum(G.nodes[i]["POP100"] for i in districts[j]) for

# print district info
for j in range(k):
    print("District",j,"has population",district_populations[j],"and contain
    print("")
```

The number of cut edges is 33.0

District 0 has population 797302 and contains counties ['Wright County', 'Story County', 'Sac County', 'Hardin County', 'Boone County', 'Plymouth County', 'Cherokee County', 'Emmet County', 'Woodbury County', 'Clay County', 'Crawford County', 'Osceola County', 'Cerro Gordo County', 'Greene County', 'Lyon County', 'Monona County', 'Humboldt County', 'Hamilton County', 'Franklin County', 'O'Brien County', 'Guthrie County', 'Sioux County', 'Winnebago County', 'Ida County', 'Carroll County', 'Pocahontas County', 'Kossuth County', 'Webster County', 'Palo Alto County', 'Calhoun County', 'Worth County', 'Hancock County', 'Dickinson County', 'Buena Vista County', 'Dallas County']

District 1 has population 796929 and contains counties ['Keokuk County', 'Marshall County', 'Davis County', 'Jasper County', 'Des Moines County', 'Scott County', 'Lee County', 'Iowa County', 'Wapello County', 'Henry County', 'Louisa County', 'Monroe County', 'Muscatine County', 'Lucas County', 'Washington County', 'Cedar County', 'Jefferson County', 'Poweshiek County', 'Wayne County', 'Marion County', 'Mahaska County', 'Appanoose County', 'Van Buren County', 'Johnson County']

District 2 has population 798070 and contains counties ['Mitchell County', 'Grundy County', 'Winneshiek County', 'Delaware County', 'Floyd County', 'Jones County', 'Butler County', 'Buchanan County', 'Tama County', 'Bremer County', 'Allamakee County', 'Howard County', 'Clinton County', 'Black Hawk County', 'Fayette County', 'Chickasaw County', 'Dubuque County', 'Benton County', 'Jackson County', 'Linn County', 'Clayton County']

District 3 has population 798068 and contains counties ['Montgomery County', 'Union County', 'Polk County', 'Audubon County', 'Pottawattamie County', 'Taylor County', 'Page County', 'Fremont County', 'Adams County', 'Cass County', 'Decatur County', 'Harrison County', 'Madison County', 'Ringgold County', 'Clarke County', 'Warren County', 'Adair County', 'Shelby County', 'Mills County']

```
In [54]: # To check if satisfies contiguity constraint
for district in districts:
    print("Is district =", district, "connected?", nx.is_connected(G.subgraph
```

```
Is district = [0, 4, 9, 16, 17, 19, 23, 30, 31, 32, 33, 35, 36, 37, 40, 42,
43, 49, 50, 52, 53, 55, 57, 58, 62, 63, 64, 66, 70, 73, 74, 81, 89, 91, 93]
connected? True
Is district = [3, 10, 12, 18, 24, 25, 26, 28, 38, 39, 44, 46, 47, 48, 54, 61,
67, 83, 86, 94, 95, 96, 97, 98] connected? True
Is district = [5, 6, 7, 11, 13, 15, 29, 34, 45, 51, 56, 69, 72, 75, 77, 80,
82, 85, 87, 88, 90] connected? True
Is district = [1, 2, 8, 14, 20, 21, 22, 27, 41, 59, 60, 65, 68, 71, 76, 78,
79, 84, 92] connected? True
```

```
In [55]: # Let's draw it on a map
import geopandas as gpd

# Read Iowa counties shapefile from "IA_counties.shp"
filepath = '/Users/adenijibabalola/Desktop/DeterministicOperationsResearch/I
filename = 'IA_counties.shp'

# Use os.path.join to correctly construct the file path
full_path = os.path.join(filepath,filename)

# Read geopandas dataframe from file
df = gpd.read_file(full_path)

# Which district is each county assigned to?
assignment = [ -1 for i in G.nodes ]

labeling = { i : j for i in G.nodes for j in range(k) if x[i,j].x > 0.5 }

# Now add the assignments to a column of the dataframe and map it
node_with_this_geoid = { G.nodes[i]['GEOID20'] : i for i in G.nodes }

# pick a position u in the dataframe
for u in range(G.number_of_nodes()):

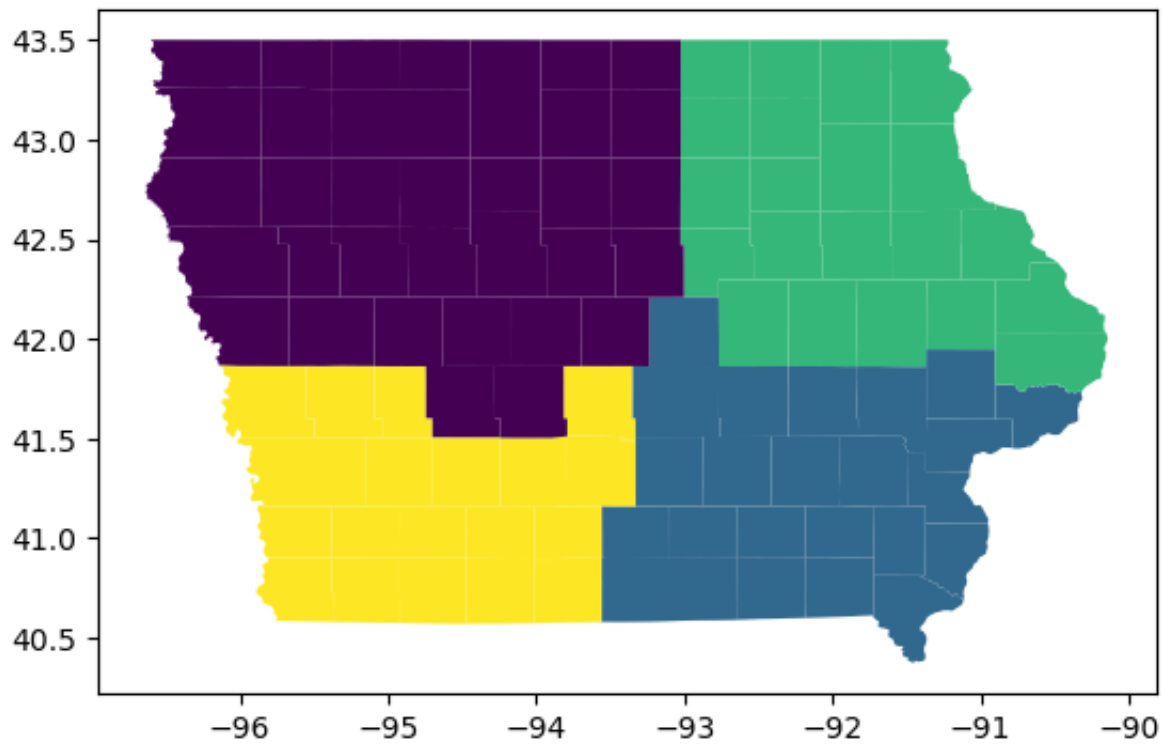
    geoid = df['GEOID20'][u]

    # what node in G has this geoid?
    i = node_with_this_geoid[geoid]

    # position u in the dataframe should be given
    # the same district # that county i has in 'labeling'
    assignment[u] = labeling[i]

# now add the assignments to a column of our dataframe and then map it
df['assignment'] = assignment

my_fig = df.plot(column='assignment').get_figure()
```

In []: