```
import pandas as pd
In [1]:
        grand 1 = pd.read csv('C:/1 grandata seoul single.csv')
        grand_2 = pd.read_csv('C://2_grandata_한글.csv', index_col =0)
        grand 1 = grand 1.drop(['기준시점'],axis=1)
        grand_1 = grand_1.groupby('아파트단지코드').sum()
         grand 1 = grand 1.reset index()
         grand 1 2 = pd.merge(grand 1,grand 2)
        grand 1 2
Out[1]:
                       라
                       0
                                                                                  스포츠/
                아파트
                      프
                                                                        의류/잡
                                                     대형할인
                                                              소형유통
                           요식이용
                                     유흥이
                                            백화점이
                                                                                  문화/레
                                                                                           숙박이
                단지코 스
                                                              점이용금
                                                     점이용금
                                                                       화이용금
                              금액
                                     용금액
                                              용금액
                                                                                 저이용금
                                                                                           용금액
                      테
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                       0
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           0 A100064
                       4 28357000
                                    367000
                                            5115000 10665000 19474000
                                                                        5128000
                                                                                 5354000
                                                                                          526000
           1 A100070
                       4 38529000
                                   5068000
                                            3967000
                                                    10025000 26564000
                                                                        1590000
                                                                                10145000
                                                                                         1816000
           2 A100079
                       4 29840000
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                                                    10090000 19777000
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                                    287000
                                                                        1626000
                                                                                          909000
           3 A100086
                       4 74321000
                                   5025000
                                            6308000
                                                     8156000 41960000
                                                                        5354000
                                                                                17270000
                                                                                         2833000
              A100163
                       4 11962000
                                     27000
                                            2519000
                                                     2422000 10843000
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                                   2704000
                                                     12147000 61765000
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               A99883
                       4 94290000
                                           12962000
                                                                                19173000
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               A99893
        3606
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                                                      5029000
                                                             25728000
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        3607
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               A99915
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                                            2762000
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        3609
               A99978
                       4 32380000
                                    259000
                                            4072000
                                                     5538000 17540000
                                                                        3894000
                                                                                 6660000
                                                                                          879000
        3610 rows × 116 columns
        home = pd.read_excel('C:/fsecdata/수정본 포함(한투, 그랜데이터)/백업/bizcell코드 매칭테이
In [2]:
        grand 1 2 = pd.merge(grand 1 2,home,left on = '아파트단지코드',right on = 'HOME BIZ CEL
In [3]:
        grand 1 2
```

Out[3]:

	아파트 단지코 드	라 이 프 스 테 이 지	요식이용 금액	유흥이 용금액	백화점이 용금액	대형할인 점이용금 액	소형유통 점이용금 액	의류/잡 화이용금 액	스포츠/ 문화/레 저이용금 액	숙박이 용금액
	<b>0</b> A100064	4	28357000	367000	5115000	10665000	19474000	5128000	5354000	526000
	<b>1</b> A100070	4	38529000	5068000	3967000	10025000	26564000	1590000	10145000	1816000
	<b>2</b> A100079	4	29840000	287000	4257000	10090000	19777000	1626000	4085000	909000
3	<b>3</b> A100086	4	74321000	5025000	6308000	8156000	41960000	5354000	17270000	2833000
	<b>4</b> A100163	4	11962000	27000	2519000	2422000	10843000	649000	2597000	267000
	····									
360	<b>5</b> A99883	4	94290000	2704000	12962000	12147000	61765000	10792000	19173000	3183000
360	<b>6</b> A99893	4	41169000	1735000	3242000	5029000	25728000	4589000	12297000	1682000
360	<b>7</b> A99911	4	33623000	429000	10684000	7730000	23120000	4402000	6861000	585000

```
라
            이
                                                                스포츠/
      아파트 프
                                       대형할인
                                               소형유통
                                                       의류/잡
               요식이용
                        유흥이
                               백화점이
                                                                문화/레
                                                                        숙박이
      단지코 스
                                       점이용금
                                               점이용금
                                                       화이용금
                                                               저이용금
                  금액
                        용금액
                                용금액
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         드 테
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3608
     A99915 4 11926000
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                                               8923000
                                                        318000
                                                               2883000 2414000
3609
     A99978 4 32380000
                       259000
                              4072000
                                       5538000 17540000
                                                       3894000
                                                               6660000
                                                                       879000
```

3610 rows × 121 columns

```
grand 1 2 = grand 1 2.drop(['HOME BIZ CELL','시도명','아파트단지코드','행정동코드','행정
In [4]:
       grand_1_2 = grand_1_2.drop(['라이프스테이지'],axis=1)
In [5]:
       grand 1 2 = grand 1 2.groupby('시군구명').median()
In [6]:
      grand 2.columns
      Index(['아파트단지코드', '고객수', '하이앤드소비수준', '하이앤드소득수준', '하이앤드법인대
Out[6]:
       표/전문직', '추정연소득'
             '연간카드소비금액', '순자산평가금액', '최근차량할부약정액', '차량보유(국산)', '차량
       보유(수입)', '20대',
            '30대', '40대', '50대', '60대', '70대이상', '싱글', '신혼및어린이자녀', '청소년자
       녀', '성인자녀',
             '실버', '명품구매여부', '앱구매등급', '디지털음악이용', '게임이용등급', '골프이용등
      급', '애완동물관심등급',
                        '자동차이용여부', '대형할인점', '백화점', '브랜드', '슈퍼마켓', '아
             '국내여행등급',
       울렛', '고궁', '공연장'
            '관광지', '국립공원', '관광_놀이시설', '동물원', '관광_레프팅', '미술관', '박물관/
       기념관', '식물원'
             '자연휴양림', '골프장', '낚시터', '레저 놀이시설', '레저 레프팅', '볼링장', '스키
       장', '실내골프', '체험장',
            '캠핑장', '헬스스포츠'],
           dtype='object')
       grand 1 2= grand 1 2.drop(['행정코드','고객수','20대','30대','40대','50대','60대','70대C
In [7]:
             '실버'],axis=1)
In [8]: from sklearn.metrics.pairwise import cosine similarity
       total_cosine = cosine_similarity(grand_1_2,grand_1_2)
In [9]: ind = grand 1 2.index
       cosine result=pd.DataFrame(total cosine,columns=ind, index= ind)
```

```
002.[grand 1,2]지역별 유사도분석및시각화
                                                  from matplotlib import font_manager, rc
In [10]:
                                                    import matplotlib
                                                    matplotlib.rcParams['font.family'] = 'Malgun Gothic'
                                                   matplotlib.rcParams['axes.unicode minus'] = False
                                                   import matplotlib.pyplot as plt
In [11]:
                                                    import seaborn as sns
                                                    plt.figure(figsize=(20,20))
                                                     ax = sns.heatmap(cosine_result,annot=True,fmt='.4')
                                                     plt.title('구별 코사인 유사도 전체 column')
                                                    plt.show()
                                                                                                                                                                                                                         구별 코사인 유사도 전체 column
                                                                                                                                                                                                                                                                                                             961 0.9935 0.9944 0.994 0.9944 <mark>0.9996</mark>
                                                                                   1.0 0.9991 0.9999 0.9998 0.9986 0.9995 0.9985 0.9992 0.9992 0.9997 0.9992 0.9981 0.9998 0.9993 0.9988 0.9993 0.9988 0.9995 0.9988 0.9993 0.9988 0.9993 0.9988 0.9993 0.9988 0.9993 0.9988 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9998 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9998 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988 0.9988
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                                                                                                                                             0.9985 0.9996 0.9985 0.999 0.999 0.9998 0.9987 0.9976 0.9995
                                                                                                                                                                                                                                                                                             957 0.9999 0.9989 0.9985 0.9993 0.9991 0.9972 0.9983 0.9982
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                                                                                                                                                                                                                                                                                                         0.9959 0.998 0.9968 0.9977 0.9974 0.988 0.9982 0.9921 0.9969 0.9987
                                                                                 0.9985 0.9994 0.999 0.9985
                                                                                                                                                                                             10 09995 09993 09973 09957 09993 09849 09966 09983 09967 09984 09969 09891 09993
                                                                                 0.9992 0.9994 0.9993 0.999
                                                                                                                                                                                                                                                              9963 0.9993 0.9862 0.9971 0.9988 0.9979 0.9986 0.9978 0.9896 0.9989 0
                                                                                 0.9992 0.9998 0.9996 0.999 0.9966 0.9997 0.9994 0.9995 1.0 0.9996 0.9977
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                - 0 994
                                                                                0997 0995 0999 0998 0998 0998 0998 0992 0993 0996 1.0 0998 0997 0997 0989 0998 0998 0994 0998 09994 0998 0999 0992 0995
                                                                               0.9992 0.9979 0.9991 0.9987 0.9995 0.9979 0.9965 0.9973 0.9977 0.9988 1.0 0.9992 0.999
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                                                                                                                                                                                                                                                                                                      0.9997 0.9989 0.9989 0.9992 0.9991 0.9982 0.9977 0.9985 0.9953 0.9975
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                 0.992
                                                                                    9998 0.9991 0.9998 0.9995 0.9985 0.9993 0.9984 0.9993 0.9993 0.9997 0.999 0.9982 1.0
                                                                                   9903 0.9866 0.99 0.9895
```

0.990 0.998 0.9979 0.9986 0.9976 0.9985 0.9973 0.9968 0.9967 0.9979 0.9984 0.9989 0.9989 0.9983 9935 0.999 0.9995 1.0 0.9992 0.9991 956 0.9993 0.9968 - 0.988 0.9988 0.999 0.9987 0.9991 0.9981 0.9974 0.9969 0.9978 0.9989 0.9992 0.9991 0.9986 0.9941 0.9995 0.9994 0.9991 0.9993 1.0 957 0.9983 0.9981 0.9983 0.9987 9932 09897 09929 09924 09972 961 0.9982 0.9935 0.999 0.9971 0.995 0.9955 0.9956 0.995 1.0 19935 09921 0.992 0.9934 0.9954 0.997 0.9985 0.9954 0.9971 0.9984 0.9966 0.9972 0.9968 0.9981 0.9975 0.9949 0.997 0.9968 0.9972 0.9974 0.996 0.9971 0.9969 0.9951 0.9964 0.9974 0.9963 0.9953 0.9961 0.9893 0.9966 0.9967 0.9959 0.9962 0.9983 0.9903 09998 0999 09997 09998 09982 09996 09987 09988 09992 09997 09986 **09975** 09994 **09993** 09984 09989 **09976** 0999 09987 **09921** 09994 강납구 강봉구 강복구 강서구 관학구 광진구 구보구 급천구 노원구 도봉구동대문구동작구 마보구서대문구서초구 성봉구 성복구 송파구 양천구영등포구용산구 은형구 종보구

grand\_1\_2.div(grand\_1\_2.sum(axis=1),axis=0) In [12]:

의류/잡

대형할인 소형유통

Out[12]:

	요식이용 금액	유흥이용 금액	백화점이 용금액	점이용금	조영류용 점이용금 액	의류/섭 화이용금 액	문화/레 저이용금 액	숙박이용 금액	여행이용 금액	교통이용 금액
시 군 구 명										
강 남 구	0.032482	0.000373	0.008822	0.002563	0.014082	0.005275	0.007466	0.001700	0.000751	0.002412
강 동 구	0.030743	0.000435	0.004835	0.004519	0.016546	0.003881	0.004969	0.000921	0.000368	0.001543
강 북 구	0.027275	0.000513	0.004581	0.004103	0.017190	0.003221	0.004954	0.000985	0.000475	0.000924
강 서 구	0.029273	0.000386	0.005226	0.004737	0.017387	0.002789	0.005035	0.000914	0.000776	0.001506
관 악 구	0.030635	0.000295	0.003847	0.003278	0.019912	0.004144	0.004552	0.000928	0.000511	0.001624
광 진 구	0.032769	0.000271	0.007329	0.003537	0.017194	0.003477	0.005740	0.000942	0.000595	0.001747
구 로 구	0.028459	0.000448	0.005849	0.005317	0.019267	0.003224	0.004679	0.001026	0.000550	0.001662
금 천 구	0.024977	0.000491	0.003431	0.007789	0.017574	0.003880	0.004952	0.000963	0.000262	0.001539
노 원 구	0.024572	0.000412	0.003985	0.003927	0.015513	0.002689	0.004479	0.000826	0.000437	0.001278
도 봉 구	0.031052	0.000560	0.004848	0.006137	0.018701	0.002929	0.005094	0.001087	0.000467	0.001327
동 대 문 구	0.028835	0.000528	0.004405	0.004669	0.020310	0.003444	0.005286	0.001293	0.000503	0.001801
동 작 구	0.029753	0.000402	0.005868	0.003743	0.015299	0.003480	0.005259	0.001244	0.000617	0.002199

의류/잡

대형할인 소형유통

	요식이용 금액	유흥이용 금액	백화점이 용금액	점이용금	조영유공 점이용금 액	의류/집 화이용금 액	문화/레 저이용금 액	숙박이용 금액	여행이용 금액	교통이용 금액
시 군 무 명										
마 포 구	0.030327	0.000295	0.005378	0.004580	0.016292	0.003523	0.005997	0.001107	0.000584	0.001964
서 대 문 구	0.026802	0.000339	0.004016	0.004120	0.017981	0.002833	0.004715	0.001045	0.000697	0.001945
서 초 구	0.033139	0.000272	0.010816	0.003950	0.014682	0.004085	0.007276	0.001440	0.000714	0.002034
성 동 구	0.031166	0.000396	0.008055	0.004867	0.016882	0.003879	0.006315	0.001169	0.000743	0.001888
성 북 구	0.027238	0.000343	0.006351	0.004937	0.018074	0.003113	0.004286	0.000995	0.000441	0.001587
송 파 구	0.029795	0.000429	0.007827	0.004115	0.016010	0.003501	0.006572	0.001198	0.000570	0.001932
양 천 구	0.027893	0.000395	0.006618	0.004079	0.018324	0.002803	0.005265	0.000877	0.000665	0.001711
영 등 포 구	0.029689	0.000420	0.007000	0.006247	0.016776	0.003928	0.005308	0.001048	0.000619	0.002181
용 산 구	0.026833	0.000265	0.011837	0.003908	0.014533	0.003939	0.007258	0.001816	0.000638	0.003031
은 평 구	0.027841	0.000362	0.005170	0.003948	0.016360	0.002337	0.004719	0.000771	0.000653	0.001383
종 로 구	0.033761	0.000424	0.010203	0.002920	0.025326	0.003924	0.007853	0.001858	0.000669	0.002657
중 구	0.032287	0.001180	0.007793	0.005030	0.016813	0.004050	0.005607	0.001864	0.001375	0.004188
중 랑 구	0.033502	0.000722	0.004204	0.005993	0.022264	0.004101	0.006009	0.001088	0.000729	0.001622

25 rows × 101 columns

```
In [13]: from sklearn.preprocessing import MinMaxScaler

minmax_scaler = MinMaxScaler()

minmax_scaled = minmax_scaler.fit_transform(grand_1_2)

In [14]: minmax_scale = cosine_similarity(minmax_scaled,minmax_scaled)

ind = grand_1_2.index

cosine_result=pd.DataFrame(minmax_scale,columns=ind, index= ind)

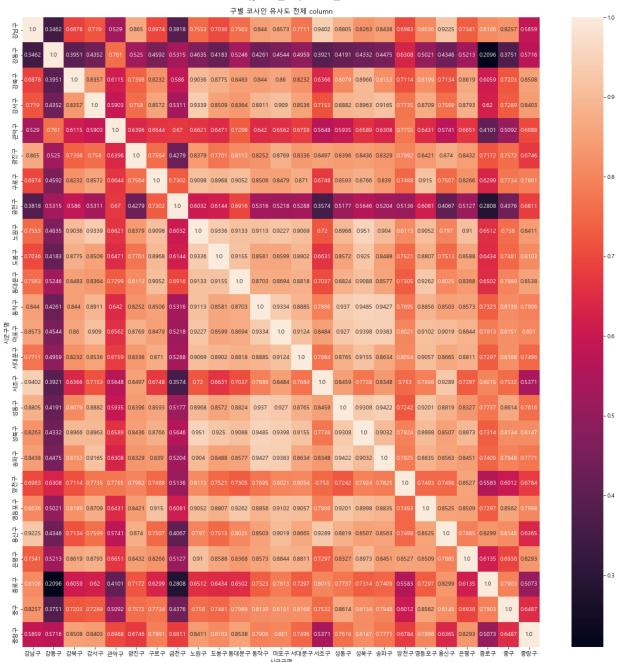
In [60]: a = pd.DataFrame(minmax_scale)

In [15]: plt.figure(figsize=(20,20))

ax = sns.heatmap(cosine_result,annot=True,fmt='.4')

plt.title('구별 코사인 유사도 전체 column')

plt.show()
```



In [16]: cosine\_result

Out[16]: 시

시 군 구 명	강남구	강동구	강북구	강서구	관악구	광진구	구로구	금천구	노원구	도봉구
시 군 무 명										
강 남 구	1.000000	0.346214	0.687755	0.719036	0.528974	0.865050	0.697364	0.381813	0.755299	0.703584
강 동 구	0.346214	1.000000	0.395074	0.435163	0.761013	0.524993	0.459178	0.531463	0.463490	0.418301
강 북 구	0.687755	0.395074	1.000000	0.835707	0.611474	0.739779	0.823165	0.585956	0.903618	0.877497
강 서 구	0.719036	0.435163	0.835707	1.000000	0.590287	0.758004	0.857233	0.531086	0.933850	0.850926
관 악 구	0.528974	0.761013	0.611474	0.590287	1.000000	0.639649	0.664436	0.670025	0.662149	0.647085
광 진 구	0.865050	0.524993	0.739779	0.758004	0.639649	1.000000	0.756442	0.427886	0.837920	0.770119
구 로 구	0.697364	0.459178	0.823165	0.857233	0.664436	0.756442	1.000000	0.730185	0.909820	0.896828
금 천 구	0.381813	0.531463	0.585956	0.531086	0.670025	0.427886	0.730185	1.000000	0.603242	0.614370
노 원 구	0.755299	0.463490	0.903618	0.933850	0.662149	0.837920	0.909820	0.603242	1.000000	0.933627
도 봉 구	0.703584	0.418301	0.877497	0.850926	0.647085	0.770119	0.896828	0.614370	0.933627	1.000000
동 대 문 구	0.756292	0.524645	0.848291	0.836351	0.729850	0.811336	0.905226	0.691598	0.913311	0.915502
동 작 구	0.843973	0.426058	0.843981	0.891056	0.641982	0.825230	0.850618	0.531550	0.911272	0.858075
마 포 구	0.857284	0.454411	0.860036	0.908951	0.656229	0.876933	0.847858	0.521793	0.922734	0.859936

시 군 구 명	강남구	강동구	강북구	강서구	관악구	광진구	구로구	금천구	노원구	도봉구
시 군 구 명										
서 대 문 구	0.771134	0.495864	0.823225	0.853559	0.675890	0.833555	0.871044	0.528785	0.906898	0.890247
서 초 구	0.940246	0.392098	0.636551	0.715286	0.564804	0.849652	0.674789	0.357415	0.720003	0.663135
성 동 구	0.880450	0.419057	0.807903	0.888206	0.593525	0.839626	0.859316	0.517746	0.896805	0.857205
성 북 구	0.826311	0.433193	0.896579	0.896322	0.658890	0.843554	0.876577	0.564555	0.950972	0.925047
송 파 구	0.843765	0.447497	0.815339	0.916504	0.630832	0.832892	0.838993	0.520407	0.903966	0.848786
양 천 구	0.698301	0.630840	0.711430	0.773483	0.775524	0.799154	0.748807	0.513594	0.811332	0.752125
영 등 포 구	0.803555	0.502088	0.819889	0.870926	0.643120	0.842077	0.915048	0.608091	0.905185	0.880721
용 산 구	0.922490	0.434645	0.713411	0.759874	0.574090	0.874024	0.750659	0.406687	0.797003	0.751325
은 평 구	0.734102	0.521324	0.861937	0.879300	0.665085	0.843179	0.826639	0.512679	0.910025	0.858765
종 로 구	0.810604	0.209636	0.605942	0.619974	0.410117	0.717230	0.629925	0.280775	0.651218	0.643446
중 구	0.825705	0.375073	0.720290	0.728931	0.509231	0.757236	0.773377	0.437623	0.758049	0.748059
중 랑 구	0.585862	0.571603	0.850793	0.840261	0.698784	0.674561	0.789115	0.681081	0.841055	0.810305

25 rows × 25 columns

```
In [24]: cosine_result.columns

Out[24]: Index(['강남구', '강동구', '강북구', '강서구', '관악구', '광진구', '구로구', '금천구', '노원구', '도봉구', '동대문구', '동작구', '마포구', '서대문구', '서초구', '성동구', '성북구', '송파구', '양천구', '영등포구', '용산구', '은평구', '종로구', '중구', '중랑구'], dtype='object', name='시군구명')

In [36]: cosine_result
```

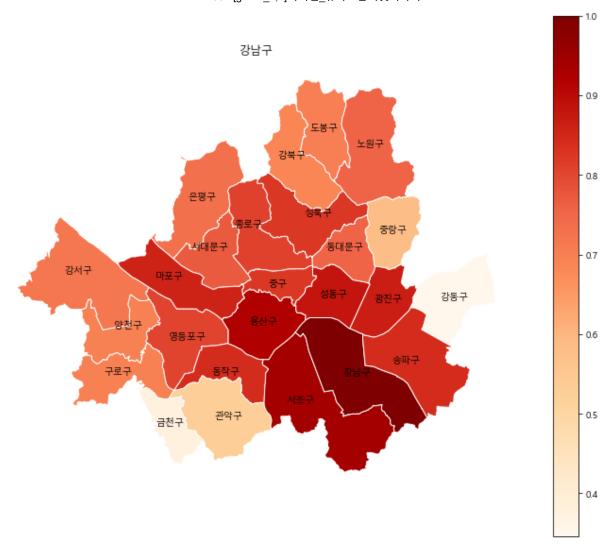
Out[36]: 시

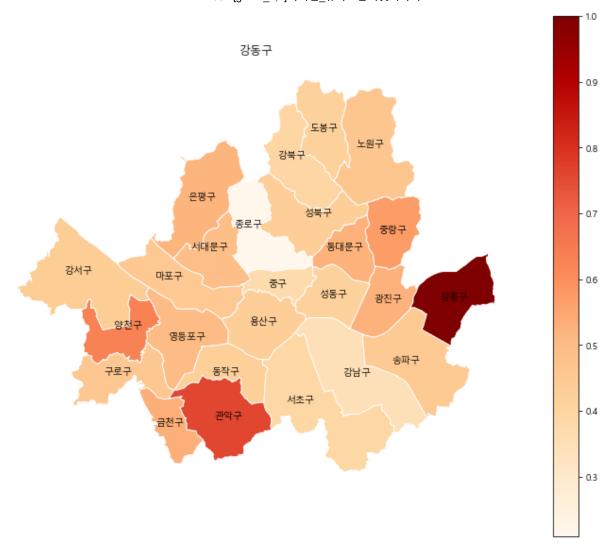
시 군 무 명	강남구	강동구	강북구	강서구	관악구	광진구	구로구	금천구	노원구	도봉구
시 군 구 명										
강 남 구	1.000000	0.346214	0.687755	0.719036	0.528974	0.865050	0.697364	0.381813	0.755299	0.703584
강 동 구	0.346214	1.000000	0.395074	0.435163	0.761013	0.524993	0.459178	0.531463	0.463490	0.418301
강 북 구	0.687755	0.395074	1.000000	0.835707	0.611474	0.739779	0.823165	0.585956	0.903618	0.877497
강 서 구	0.719036	0.435163	0.835707	1.000000	0.590287	0.758004	0.857233	0.531086	0.933850	0.850926
관 악 구	0.528974	0.761013	0.611474	0.590287	1.000000	0.639649	0.664436	0.670025	0.662149	0.647085
광 진 구	0.865050	0.524993	0.739779	0.758004	0.639649	1.000000	0.756442	0.427886	0.837920	0.770119
구 로 구	0.697364	0.459178	0.823165	0.857233	0.664436	0.756442	1.000000	0.730185	0.909820	0.896828
금 천 구	0.381813	0.531463	0.585956	0.531086	0.670025	0.427886	0.730185	1.000000	0.603242	0.614370
노 원 구	0.755299	0.463490	0.903618	0.933850	0.662149	0.837920	0.909820	0.603242	1.000000	0.933627
도 봉 구	0.703584	0.418301	0.877497	0.850926	0.647085	0.770119	0.896828	0.614370	0.933627	1.000000
동 대 문 구	0.756292	0.524645	0.848291	0.836351	0.729850	0.811336	0.905226	0.691598	0.913311	0.915502
동 작 구	0.843973	0.426058	0.843981	0.891056	0.641982	0.825230	0.850618	0.531550	0.911272	0.858075
마 포 구	0.857284	0.454411	0.860036	0.908951	0.656229	0.876933	0.847858	0.521793	0.922734	0.859936

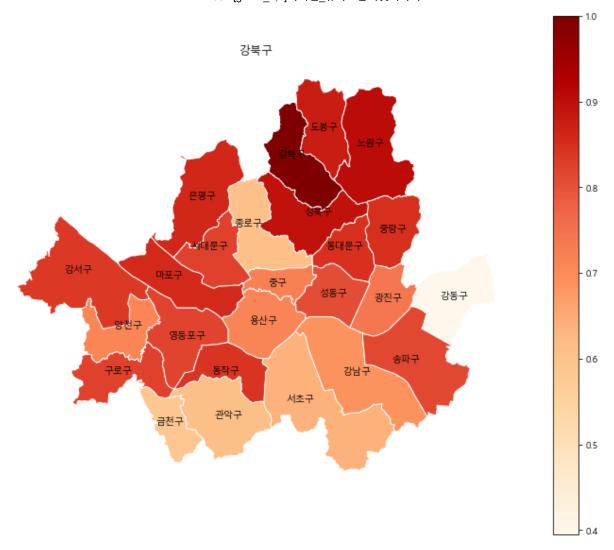
시 군 구 명	강남구	강동구	강북구	강서구	관악구	광진구	구로구	금천구	노원구	도봉구
시 군 구 명										
서 대 문 구	0.771134	0.495864	0.823225	0.853559	0.675890	0.833555	0.871044	0.528785	0.906898	0.890247
서 초 구	0.940246	0.392098	0.636551	0.715286	0.564804	0.849652	0.674789	0.357415	0.720003	0.663135
성 동 구	0.880450	0.419057	0.807903	0.888206	0.593525	0.839626	0.859316	0.517746	0.896805	0.857205
성 북 구	0.826311	0.433193	0.896579	0.896322	0.658890	0.843554	0.876577	0.564555	0.950972	0.925047
송 파 구	0.843765	0.447497	0.815339	0.916504	0.630832	0.832892	0.838993	0.520407	0.903966	0.848786
양 천 구	0.698301	0.630840	0.711430	0.773483	0.775524	0.799154	0.748807	0.513594	0.811332	0.752125
영 등 포 구	0.803555	0.502088	0.819889	0.870926	0.643120	0.842077	0.915048	0.608091	0.905185	0.880721
용 산 구	0.922490	0.434645	0.713411	0.759874	0.574090	0.874024	0.750659	0.406687	0.797003	0.751325
은 평 구	0.734102	0.521324	0.861937	0.879300	0.665085	0.843179	0.826639	0.512679	0.910025	0.858765
종 로 구	0.810604	0.209636	0.605942	0.619974	0.410117	0.717230	0.629925	0.280775	0.651218	0.643446
중 구	0.825705	0.375073	0.720290	0.728931	0.509231	0.757236	0.773377	0.437623	0.758049	0.748059
중 랑 구	0.585862	0.571603	0.850793	0.840261	0.698784	0.674561	0.789115	0.681081	0.841055	0.810305

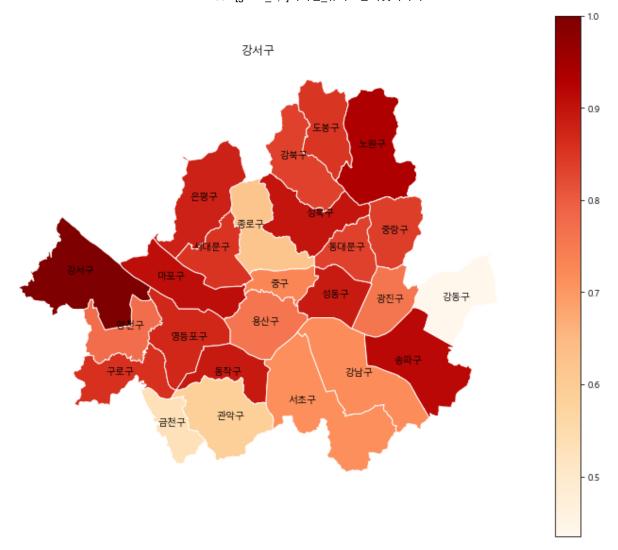
25 rows × 25 columns

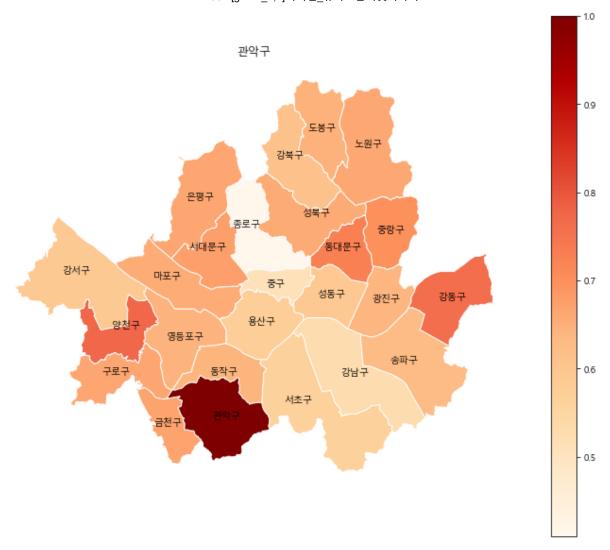
```
cosine result
In [56]:
         Index(['index', '시군구명', '강남구', '강동구', '강북구', '강서구', '관악구', '광진구',
Out[56]:
         '구로구', '금천구'
                '노원구', '도봉구', '동대문구', '동작구', '마포구', '서대문구', '서초구', '성동구',
         '성북구', '송파구',
               , '양천구', '영등포구', '용산구', '은평구', '종로구', '중구', '중랑구'],
              dtype='object', name='시군구명')
        a.index= ['강남구', '강동구', '강북구', '강서구', '관악구', '광진구', '구로구', '금천구',
In [63]:
                '동대문구', '동작구', '마포구', '서대문구', '서초구', '성동구', '성북구', '송파구',
                '용산구', '은평구', '종로구', '중구', '중랑구']
        a.columns=['강남구', '강동구', '강북구', '강서구', '관악구', '광진구', '구로구', '금천구', '동대문구', '동작구', '마포구', '서대문구', '서초구', '성동구', '성북구', '송파구',
In [65]:
                '용산구', '은평구', '종로구', '중구', '중랑구']
         a c =pd.melt(a.reset index(),id vars='index')
In [68]:
         a c = a c.rename(columns={'index':'구 1','variable':'구 2'})
In [85]:
         a_c.to_csv('구별_코사인유사도 목록.csv')
In [86]:
         import numpy as np
In [17]:
         import geopandas as gpd
         sgg file = 'C:/Users/fsecuser213/Downloads/지리정보/SIG 20220324/sig.shp'
         sgg = gpd.read_file(sgg_file,encoding='euc-kr')
         seoul = sgg[sgg['SIG CD'].str.startswith('11')]
         seoul_viz = pd.merge(seoul,cosine_result,left_on = 'SIG_KOR_NM',right_on='시군구명')
In [22]:
         import shapely
         seoul_viz['coords']=seoul_viz['geometry'].apply(lambda x: x.representative_point().coc
         seoul viz['coords'] = [coords[0] for coords in seoul viz['coords']]
         for i in seoul viz.columns[4:]:
In [23]:
             ax =seoul viz.plot(column=i,legend=True,figsize=(10,8),edgecolor='w',cmap='OrRd')
             for idx,row in seoul viz.iterrows():
                plt.annotate(text=row['SIG KOR NM'],xy=row['coords'],horizontalalignment='cent
             ax.axis('off')
             ax.set title(i)
             ax.set axis off()
             plt.tight_layout()
             plt.show()
```

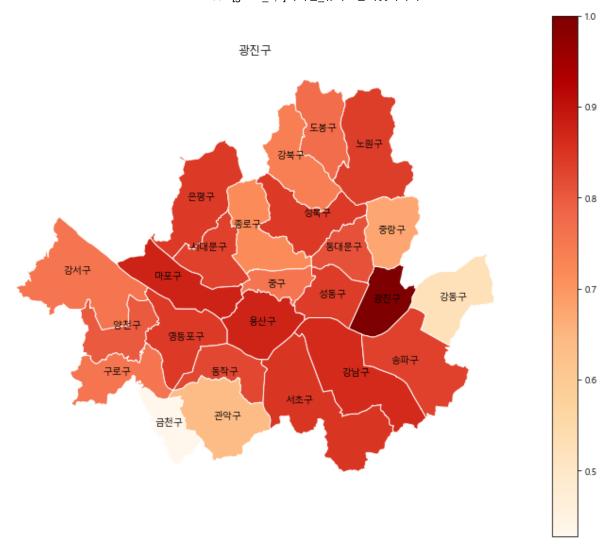


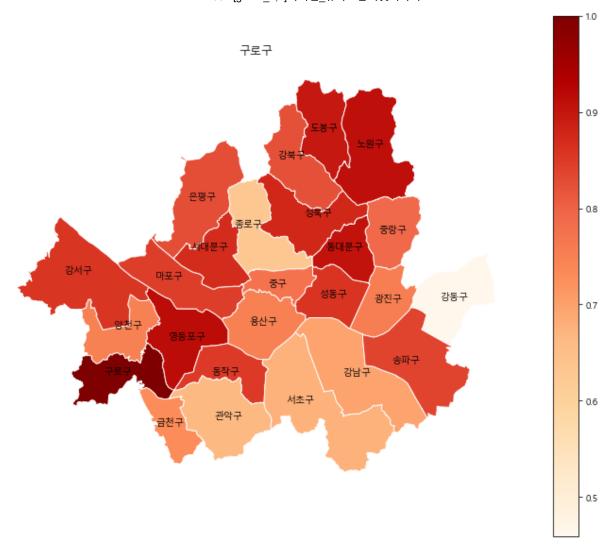


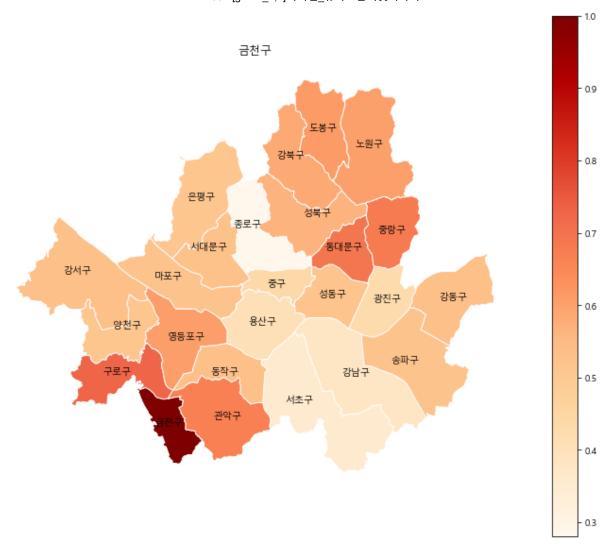


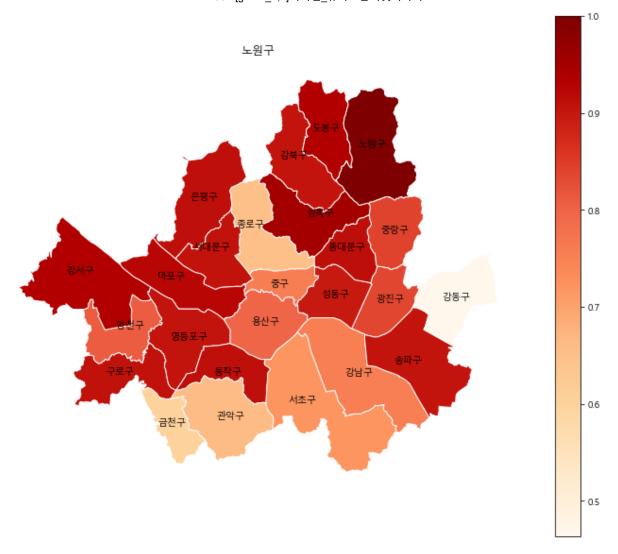


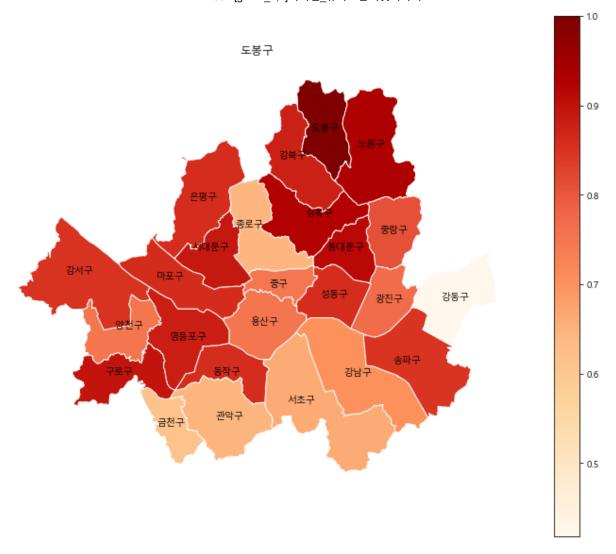


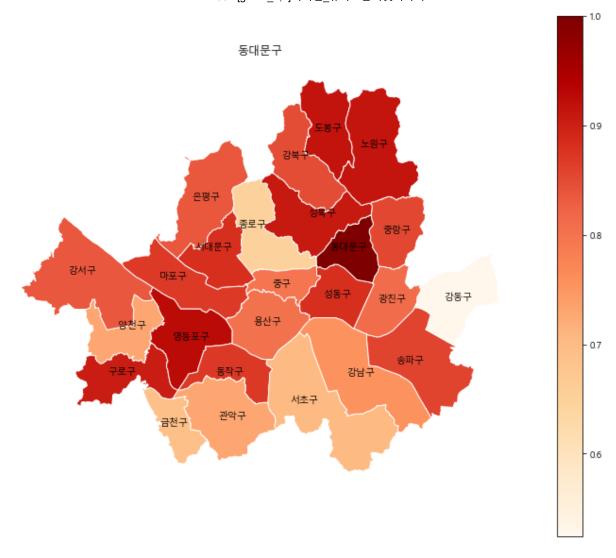


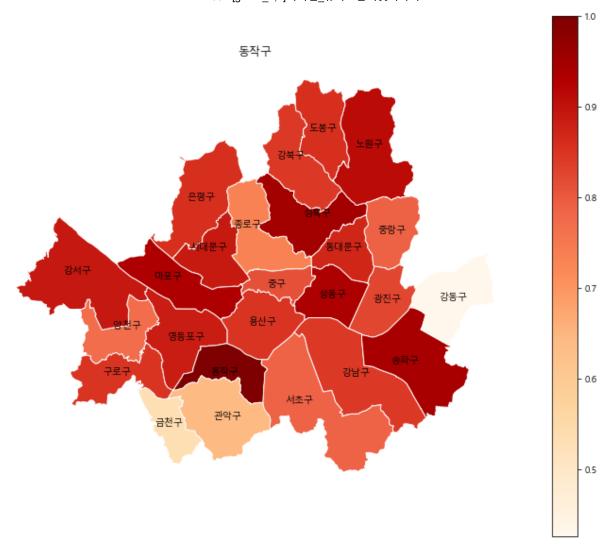


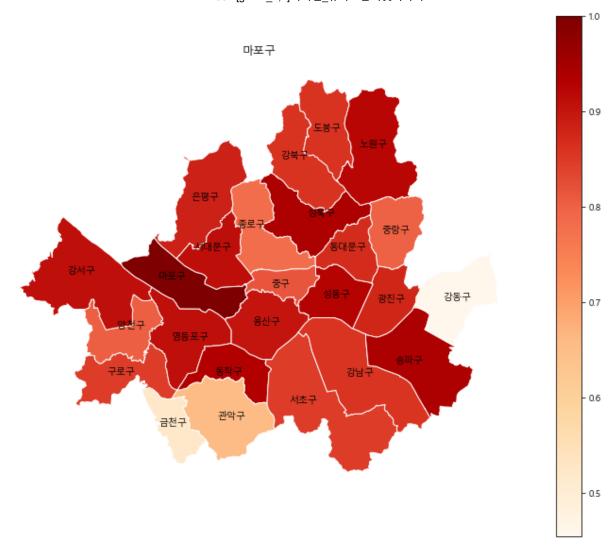


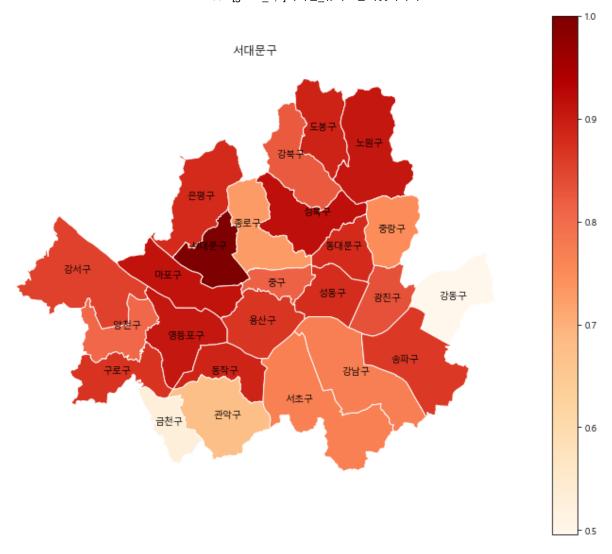


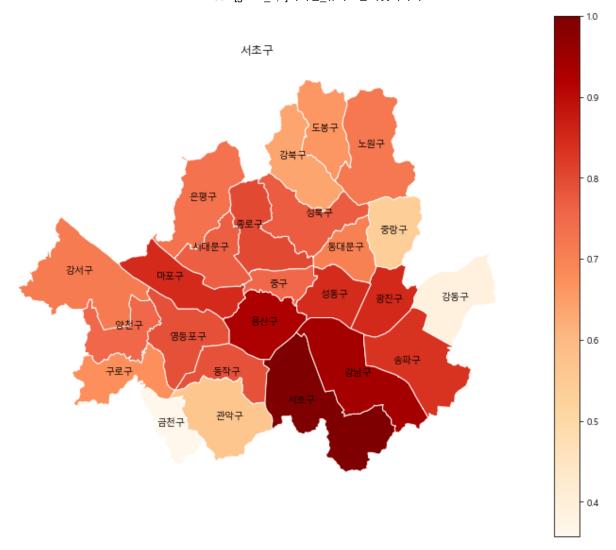


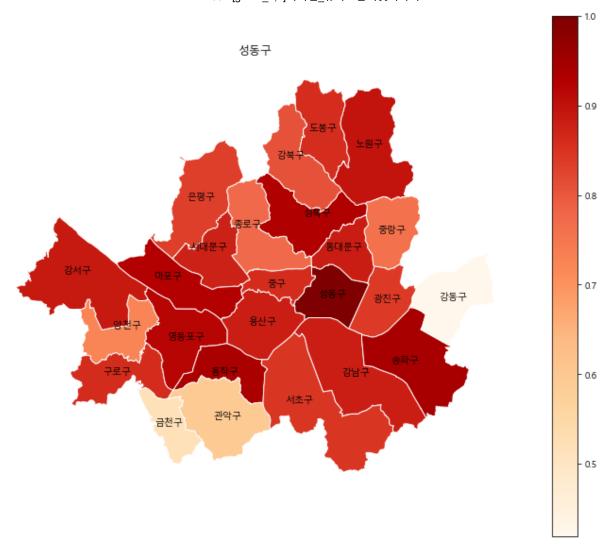


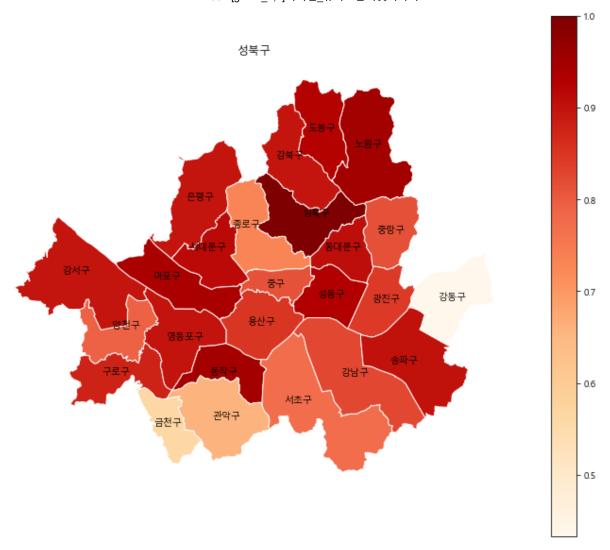


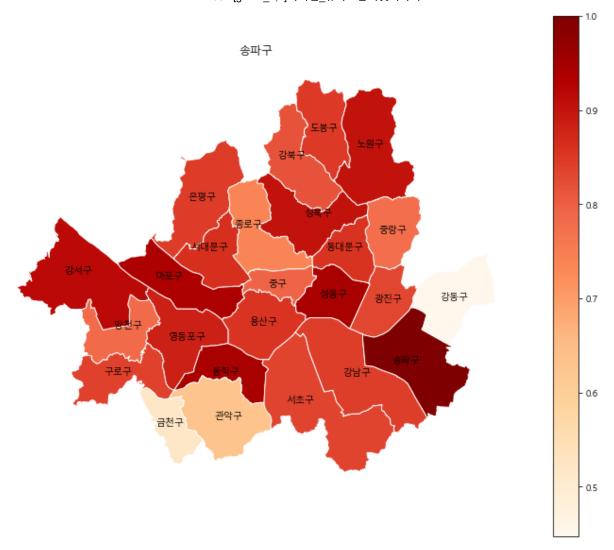


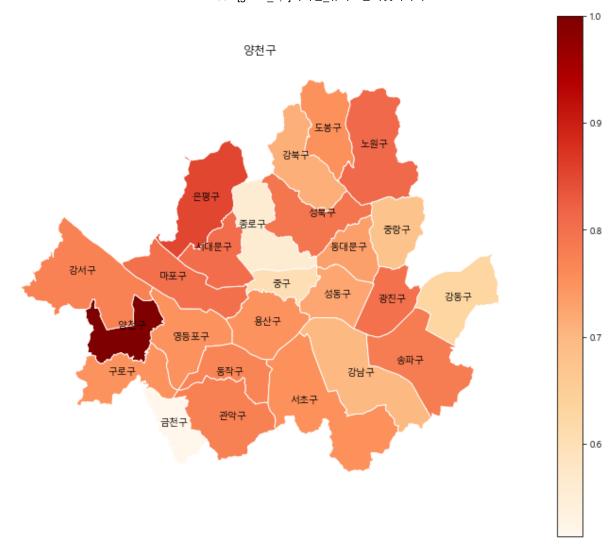


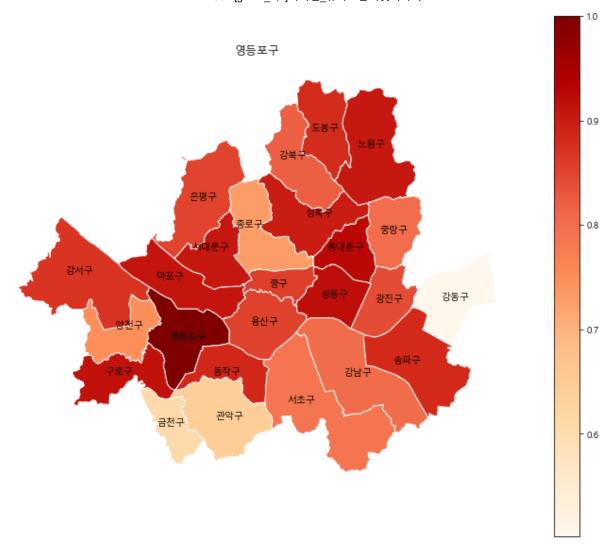


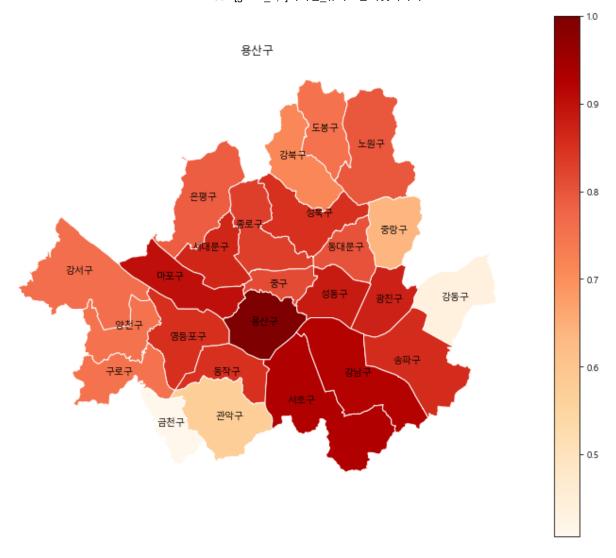


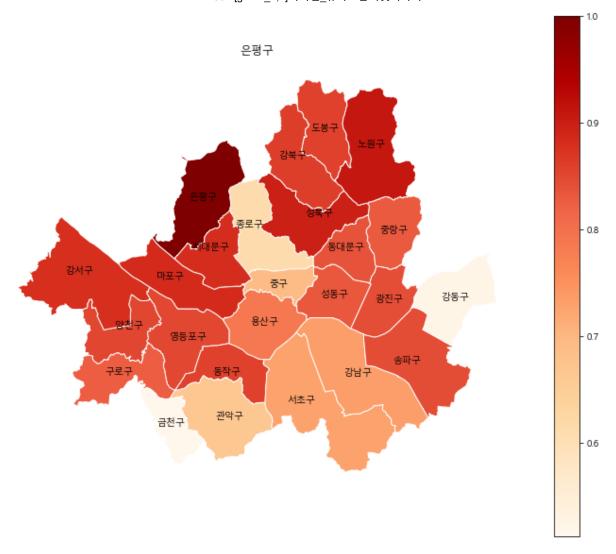


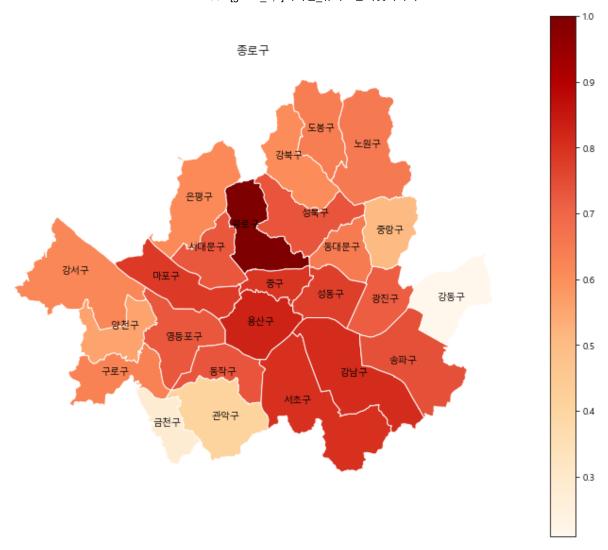


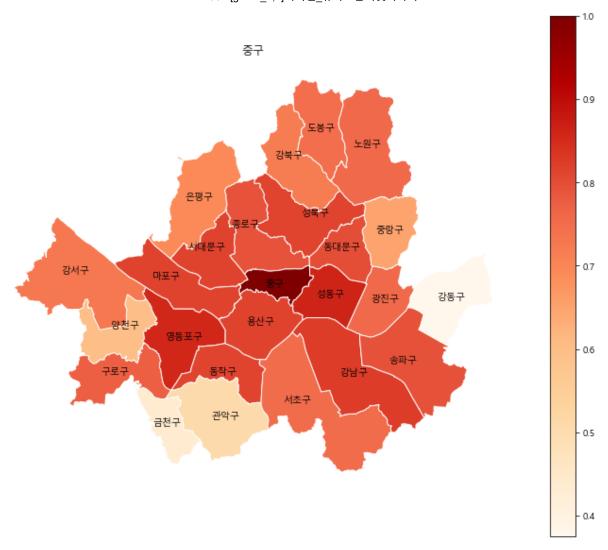


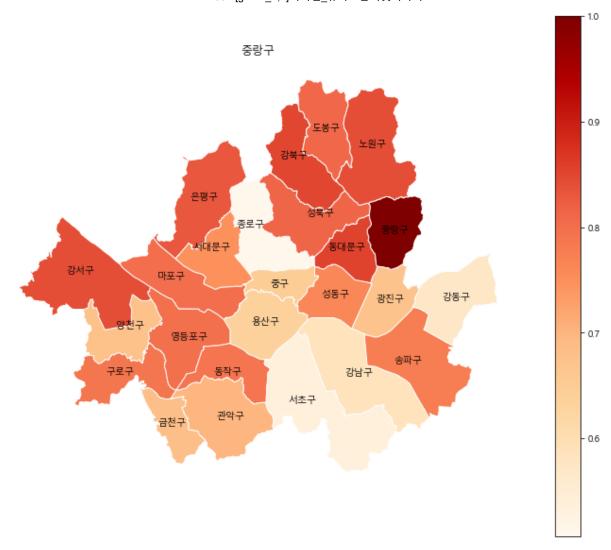


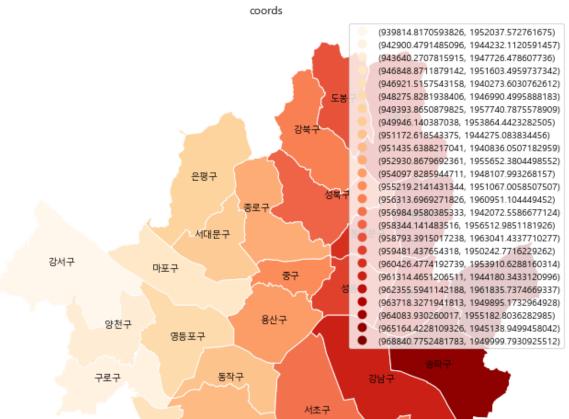












In [19]: cosine\_result.to\_csv('cosine\_similarity.csv',encoding='utf-8')

관악구

금천구