MDA HW3 KMeans

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在這次的作業中我主要共用到了 6 個 mapper, 2 個 reducer

mapperE_dist:在這個 mapper 中,我是將每個點去與 centroid 算出他們之間的 Euclidean distance 後,以我希望的形式包好後回傳

mapperE_min: 在這個 mapper 中,傳進來的資料會是 data 中的點,後面接著十個 centroid,我用迴圈掃過 10 個 centroid 後,會得到是哪個 centroid 距離最短,並且距離為多少,最後要回傳時,我除了回傳是哪個 centroid 之外,我還會回傳這個 centroid 的 index 以利於後面可以 sortByKey,讓最後 csv 檔的輸出可以順利一點。並且我在回傳時,我在最後的 KV pair 就會直接回傳這個點所產生的 cost 為多少,這樣最後我對 data 所有點的這個值用 api sum()起來,就可以得到這次的總 cost 為多少。除此之外,由於發現 data 中有重複出現的點會影響到 cost,所以我還會去算若同樣的點後面還接了幾個,最後的總 cost 會去乘上點的數量。

mapperM_dist:在這個 mapper 中,我是將每個點去與 centroid 算出他們之間的 Manhattan distance 後,以我希望的形式包好後回傳

mapperM_min:在這個 mapper 中,與 mapperE_min 實作的方式幾乎一樣,我

只是為了要回傳不同的 cost。所以這邊只有在回傳的 cost 算法不一樣,其他都跟 mapperE_min 一樣。

mapperNew_cen:在這個 mapper 中,傳進來的資料會是其中一個 centroid 接上,現在屬於這個 centroid,或者說新的這個 cluster 的點。所以要計算新的 cluster 的中心時,就把後面用迴圈全部掃過,並且把各 dimension 的值加起來 平均後,即可得到新的 centroid。

mapperinit:在這個 mapper 中,我為了讓每次 iteration 完,紀錄 centroid 的 rdd 回復成我一開始的型態所用的。會回傳 (0, centroid)。

Reducer:兩個 reducer 的功用都蠻像的,都是為了讓我可以將資料串接在一起。一次是在每次 iteration 一開始,我的 centroid 和 data 這兩個 rdd 的 kv pair 都長成 (0, centroid) (0, data),的形式,所以我先把他們 join 起來後,算完每一個點和 centroid 的距離後,為了要去計算 min distance,要把同一個點對不同的 centroid 的這些 pair 串起來,故我這裡用了一次 reduceByKey(lambda a, b:a+b)。第二次是在把每個 centroid 的與某個 data point 的 cost 算完後,為了要算出總 cost 要把所有屬於這個 cluster 得 data point 串在 centroid 的後面,於是我在這邊用了第二次的 reduceByKey(lambda a, b:a+b)。

(a)

(1) A plot of cost vs. iteration for 2 initialization strategies

	C1	C2
(6.236603e+08	4.387478e+08
	1 5.098629e+08	2.498039e+08
:	2 4.854807e+08	1.944948e+08
;	4.639970e+08	1.698048e+08
	4.609693e+08	1.562957e+08
	5 4.605378e+08	1.490942e+08
(6 4.603131e+08	1.425085e+08
	7 4.600035e+08	1.323039e+08
;	4.595705e+08	1.171710e+08
,	9 4.590211e+08	1.085474e+08
10	4.584907e+08	1.022372e+08
1	1 4.579442e+08	9.827802e+07
1:	2 4.575580e+08	9.563023e+07
13	3 4.572901e+08	9.379331e+07
14	4 4.570506e+08	9.237713e+07
1	5 4.568922e+08	9.154161e+07
10	6 4.567036e+08	9.104557e+07
1	7 4.564042e+08	9.075224e+07
18	3 4.561778e+08	9.047017e+07
19	9 4.559869e+08	9.021642e+07

(b) Percentage improvement values and your explanation

Change with c1 26.885383292517258 Change with c2 79.437750291599

由於 Euclidean 計算 cost 的方式為距離的平方,所以對於 cluster 距離遠近、離群值較為敏感,故在這種情況,使用 c2 事先刻意挑好讓最遠的點可以自己形成一種 cluster 的方式,最後得到結果以及下降的幅度也較好。

(c)

(1) Euclidean distance for all pairs of centroids, with c1 started

Euclidean	1	2	3	4	5	6	7	8	9	10
	0	692.157887	3490.25864	205.750279	346.718823	512.612247	444.731001	566.201992	1282.77084	307.669128
		0	2798.80105	897.658986	1038.82689	1204.0782	1136.32734	1257.44953	669.890228	412.076077
			0	3695.11419	3836.90664	4002.68908	3934.87156	4056.13557	2294.57964	3195.9239
				0	142.438874	309.506324	241.730115	363.262895	1474.94542	504.634116
					0	167.1498	99.5455433	220.901784	1615.85235	646.930564
						0	67.9118611	53.7898912	1782.20305	814.07615
							0	121.63372	1715.2532	746.335559
								0	1835.63967	867.823079
									0	975.320423
										0

(2) Manhattan distance for all pairs of centroids, with c1 started

A	В	C	D	Е	F	G	H	I	J	K
Manhattan	1	2	3	4	5	6	7	8	9	10
	0	728.924314	3797.89908	212.18109	374.890422	577.402076	499.157894	645.769777	1731.06431	406.701225
		0	3072.88869	935.885338	1100.83309	1303.89572	1225.35171	1372.09221	1005.29305	490.928058
			0	4001.03805	4170.30453	4372.78872	4294.95283	4440.71977	2513.42266	3396.42
				0	171.365154	375.247921	296.254724	443.498445	1934.08696	609.749322
Ī					0	204.522924	125.596786	272.934913	2102.86492	779.397227
						0	79.4016844	69.5898763	2306.38025	983.019681
							0	147.865709	2227.55586	904.37025
								0	2374.54543	1050.91622
									0	1327.58398
										0

(3) Euclidean distance for all pairs of centroids, with c2 started

Euclidean	1	2	3	4	5	6	7	8	9	10
	0	15760.1225	14110.8344	9045.32023	5567.68452	1924.62408	1100.85905	402.89055	2105.44258	3169.00377
		0	11524.5057	6743.8841	10192.525	14455.1194	14682.451	15362.418	13674.7075	12597.0396
			0	9545.8794	10883.3822	12233.9598	13208.0029	13786.4842	12508.9574	11938.3761
				0	3494.22242	7718.22201	7957.77595	8644.80704	6947.82064	5876.3302
					0	4404.56259	4492.45821	5169.93729	3488.15852	2407.91879
						0	1182.86419	1615.78824	1313.32749	2153.77147
							0	698.488136	1010.19767	2085.46068
								0	1702.79266	2768.60772
									0	1080.53494
										0

(4) Manhattan distance for all pairs of centroids, with c2 started

Manhattan	1	2	3	4	5	6	7	8	9	10
	0	15772.6149	20215.646	9533.17085	5604.20049	3088.05432	1311.03916	471.26572	2369.41216	3349.65709
		0	16003.499	7219.19667	10221.031	16105.3475	14909.1695	15434.46	13950.576	12776.8831
			0	10690.4843	14613.552	17509.9028	18912.6054	19748.9357	17851.8068	16873.2437
				0	3935.29267	8896.38921	8228.35508	9065.40433	7168.73296	6190.67931
					0	5893.07013	4696.97538	5221.25281	3737.707	2564.17054
						0	1781.82267	2619.81139	2162.80215	3337.74626
							0	840.722524	1068.93997	2137.78826
								0	1901.20876	2883.73454
									0	1176.45043
										0

(b)

(1) A plot of cost vs. iteration for 2 initialization strategies

	C1	C2
0	550117.142000	1.433739e+06
1	464869.275879	1.084489e+06
2	470897.382277	9.734317e+05
3	483914.409173	8.959346e+05
4	489216.071003	8.651283e+05
5	487629.668550	8.458466e+05
6	483711.923214	8.272196e+05
7	475330.773493	8.035903e+05
8	474871.238846	7.560395e+05
9	457232.920115	7.173329e+05
10	447494.386197	6.945879e+05
11	450915.012577	6.844445e+05
12	451250.367073	6.745747e+05
13	451974.595540	6.674095e+05
14	451570.364070	6.635566e+05
15	452739.011366	6.601628e+05
16	453082.730287	6.560413e+05
17	450583.670860	6.530368e+05
18	450368.749317	6.511124e+05
	430300.743317	0.0111210100

(b) Percentage improvement values and your explanation

Change with c1: 18.378954327236887 Change with c2: 54.685694348134085

由於 Manhattan 在計算 cost 的方式相對來說比較不敏感,Manhattan 更希望你能將集中很多點的大群即使距離很近,也可以分成不同 cluster,故在這種情況,一開始還特地將距離拉遠來看,最後達到的效果沒有比較好。

(c)

(1) Euclidean distance for all pairs of centroids, with c1 started

Euclidean	1	2	3	4	5	6	7	8	9	10
	0	2219.17728	9948.04408	528.699758	413.365061	827.718886	681.03499	917.127383	832.147434	729.056349
		0	7767.9456	2734.04985	2628.49081	3044.47787	2898.71289	3133.46013	1812.45457	1491.35735
			0	10433.0614	10361.3675	10773.5308	10626.4886	10862.9658	9340.27523	9236.84002
				0	221.372794	375.156188	249.379188	457.259653	1156.58338	1251.15835
					0	415.989985	270.748792	505.071067	1171.96421	1137.13527
						0	147.046974	89.4909166	1529.46401	1553.12381
							0	236.514622	1391.55042	1407.4044
								0	1613.55579	1642.12869
									0	709.407786
										0

(2) Manhattan distance for all pairs of centroids, with c1 started

Manhattan	1	2	3	4	5	6	7	8	9	10
	0	2341.01722	11929.3002	651.187488	496.331521	947.743236	770.737383	1056.7995	1260.51056	737.713573
		0	9597.44119	2778.94576	2830.14453	3280.35917	3104.28577	3388.98265	2380.46096	1605.27013
			0	12323.2876	12421.2631	12871.4834	12695.5542	12979.1332	10775.9392	11196.787
				0	335.951213	558.469258	382.46333	667.53323	1653.82589	1379.16517
					0	452.861331	276.326491	561.849249	1755.10553	1226.66035
						0	177.593162	110.217624	2205.30738	1677.66686
							0	287.429708	2028.90162	1500.99341
								0	2314.66745	1786.81132
									0	1006.36783
										0

(3) Euclidean distance for all pairs of centroids, with c2 started

Euclidean	1	2	3	4	5	6	7	8	9	10
	0	15747.2342	14100.1447	9032.33302	5554.78669	2006.70267	1338.16113	514.627038	1571.24342	3022.66088
		0	11524.5057	6743.8841	10192.525	14474.5541	14412.0566	15239.8771	14328.2262	12731.3976
			0	9545.8794	10883.3822	12167.7939	13125.351	13684.6068	12643.9856	12006.3946
				0	3494.22242	7742.62812	7694.2767	8521.19786	7588.40454	6009.82022
					0	4452.97168	4219.76057	5047.51626	4167.63653	2542.56935
						0	1405.10908	1637.72944	910.994388	2124.26336
							0	827.840658	566.551017	1684.51601
								0	1081.37933	2511.45886
									0	1649.38917
										0

(4) Manhattan distance for all pairs of centroids, with c2 started

Manhattan	1	2	3	4	5	6	7	8	9	10
	0	15757.6913	20200.2594	9517.66823	5588.85363	3281.48825	1430.20868	602.954849	2102.55398	3211.45576
		0	16003.499	7219.19667	10221.031	16325.2705	14506.4859	15335.9574	14980.0561	12922.9314
			0	10690.4843	14613.552	17521.5177	18775.1215	19602.2628	18111.8854	16995.1335
				0	3935.29267	9116.0245	8090.51019	8918.81312	7771.22208	6312.53001
					0	6110.8325	4293.5019	5123.06681	4768.923	2710.0565
						0	1855.57991	2682.56923	1358.79589	3413.03618
							0	833.430282	674.82757	1784.51205
								0	1500.82488	2613.99731
									0	2062.25107
										0