



UNIVERSITÀ DI PISA

University of Pisa

Laurea Magistrale (MSc) in Artificial Intelligence and Data
Engineering

Project

Data Mining and Machine Learning

FEDERATED DBSCAN BASED ON GRID

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**[https://github.com/ValeGian/](https://github.com/ValeGian/DMML_FederatedDBSCAN)
DMML_FederatedDBSCAN**

Academic year 2020-2021

INTRODUCTION TO THE PROBLEM

Federated Learning (FL): Can we train a model, in a "collaborative" way, without transferring the data to a central processing server?

Local Data Owner:

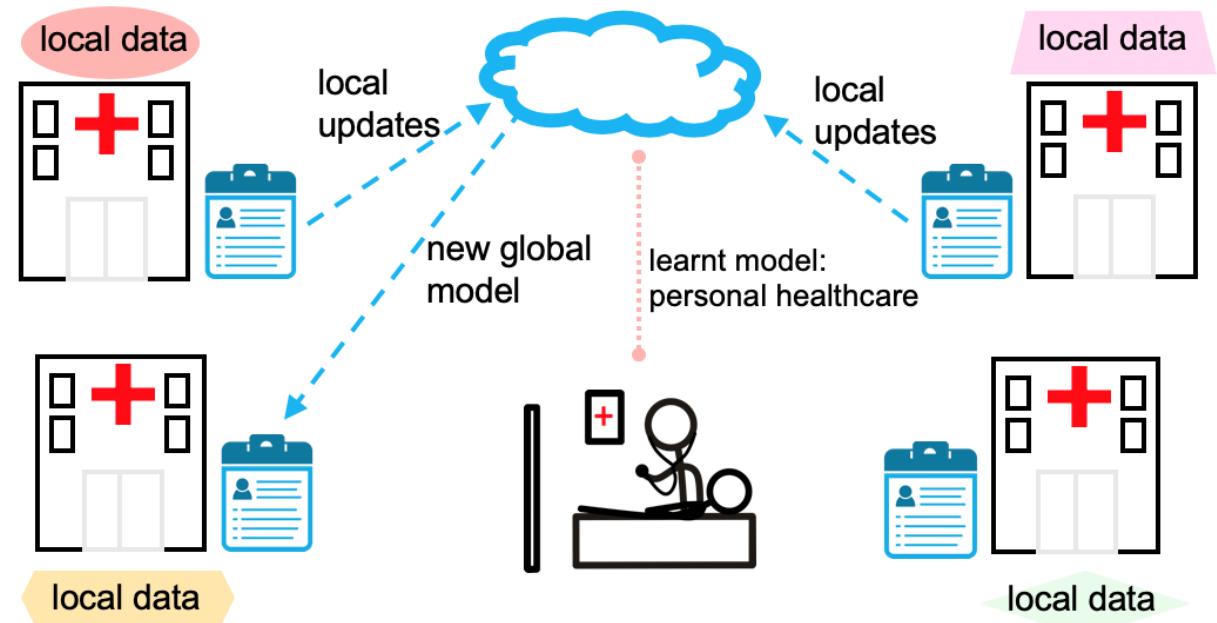
- Contains several samples described with same attributes.
- Can perform local processing.
- Can assign points to clusters depending on the result they receive from the server.

Central Server:

- Aggregate local models and consolidate the global model.
- Send results to individual Data Owners.

Practical example:

- **Healthcare domain** in which patient data cannot be transmitted.



STEPS OF THE ALGORITHM

LOCAL	SERVER
Partition the space with a granularity fixed (L), assuming the same range of features for all nodes.	
Evaluates the number of points in each cell and transmits information about non-empty cells to the server.	
	For each cell, add the contributions of all owner of the data.
	Define dense cell the cell with at least MinPts.
	Evaluate clustering by expanding a cluster along adjacent dense cells.
	Return to each local information on cluster membership of each cell.
Assign all the points relating to the cells dense to its cluster.	
Assign the remaining points to the cluster of dense adjacent cell closest to the point. Otherwise, the point is considered an	

PARAMETERS TO SET

Real parameter of federated algorithm:

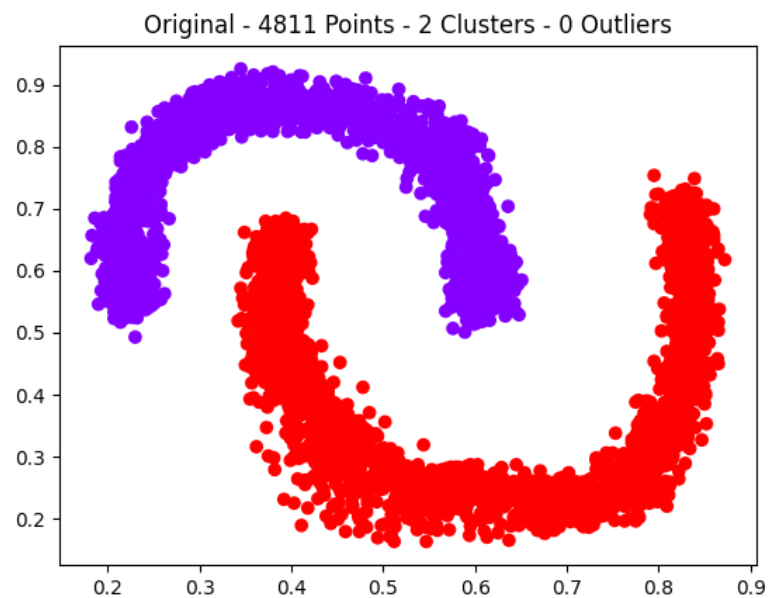
- L = fix the granularity of the cell.
- MinPts = determine the minimum number of points in a dense cell.

Parameters to simulate locally a distributed execution:

- M = Number of nodes.
- Partitioning methods.

DATASET ANALYZED

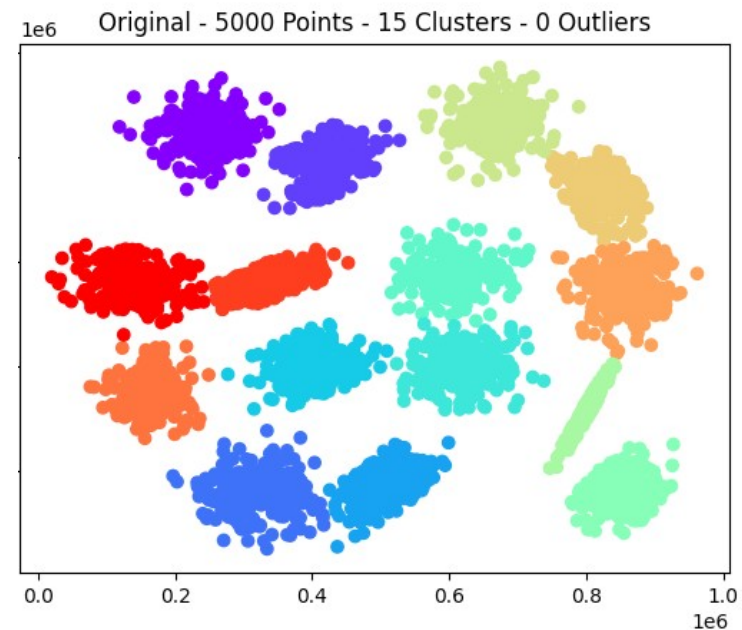
BANANA (4800)



**CLUTO-T8.8K
(8000)**

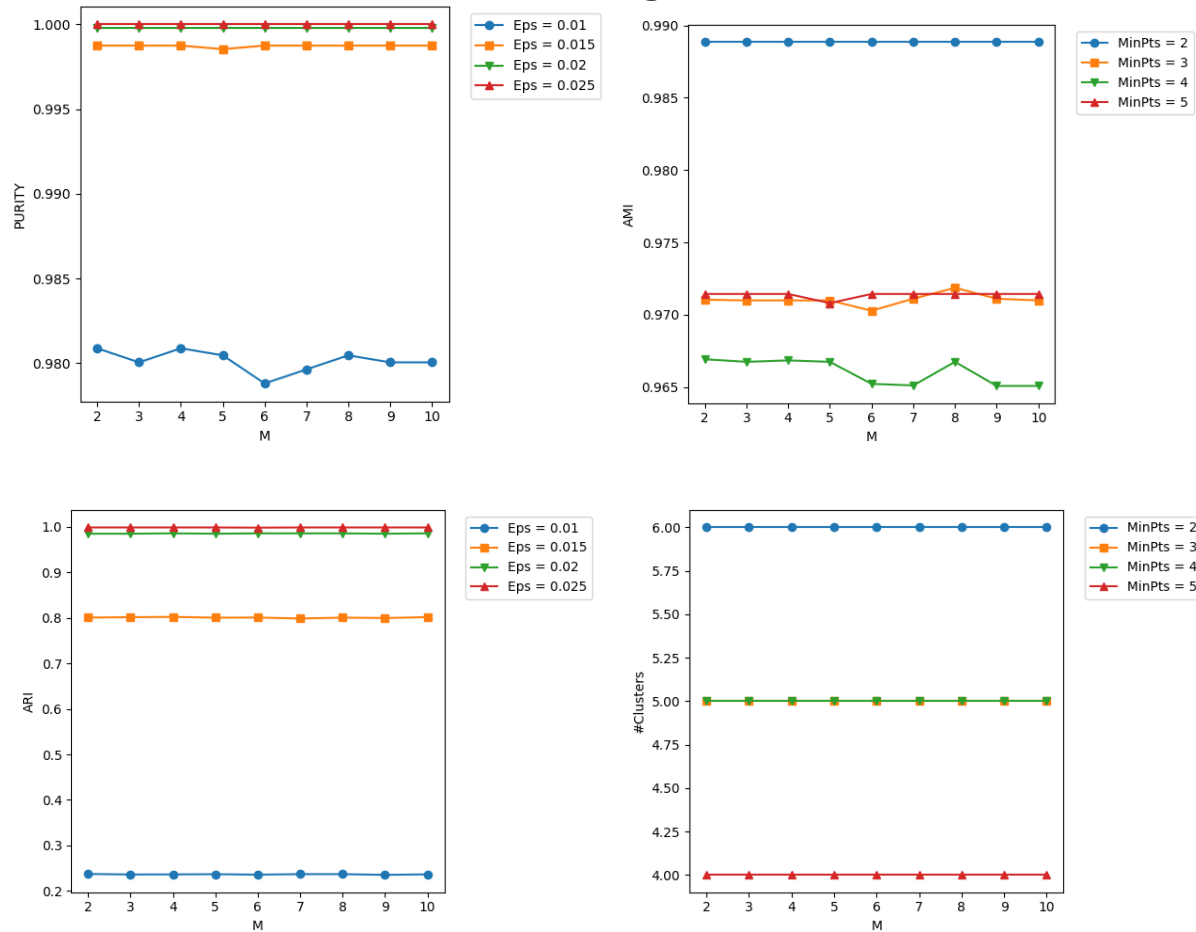


S-SET-1 (5000)

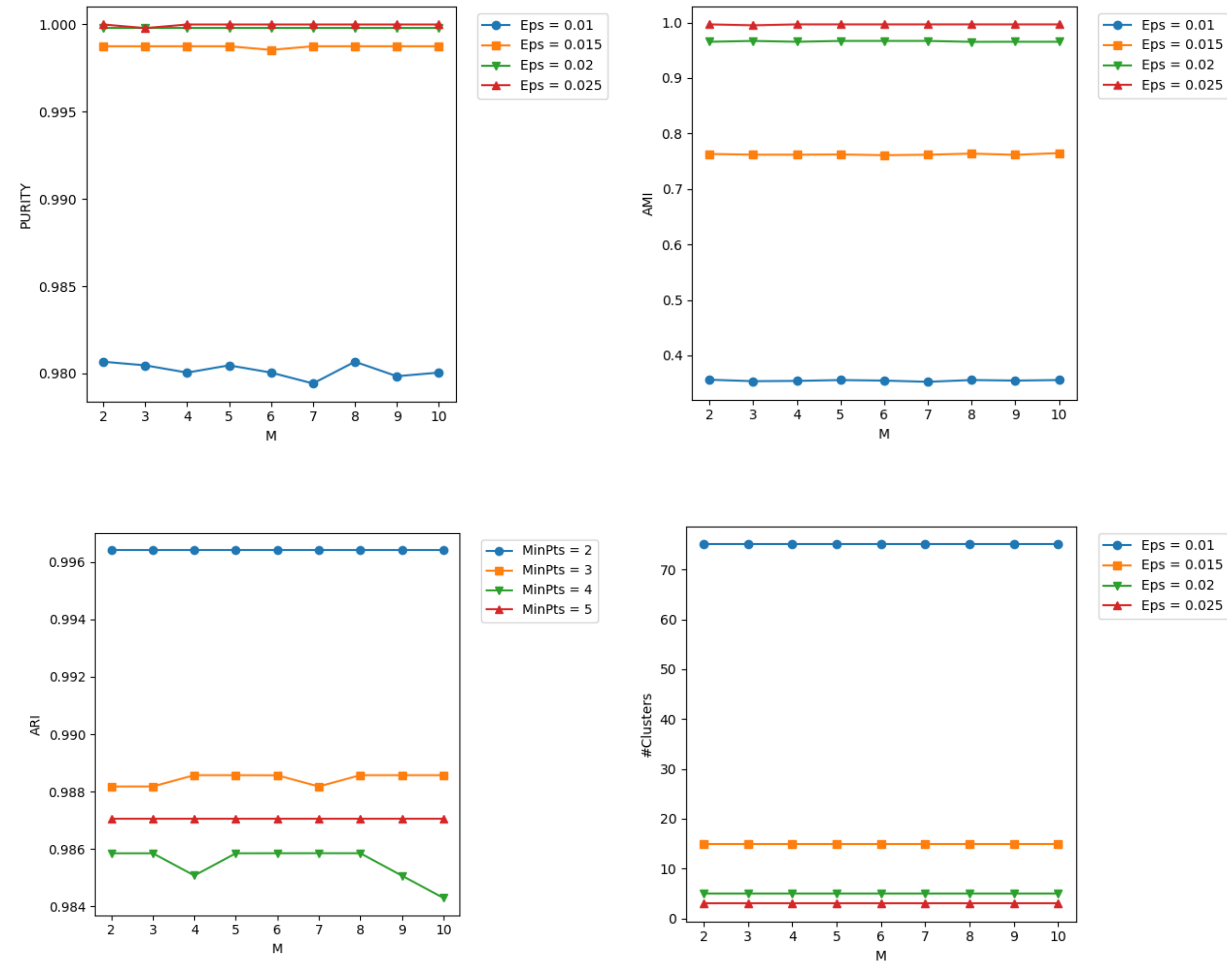


PARAMETERS REDUCTION

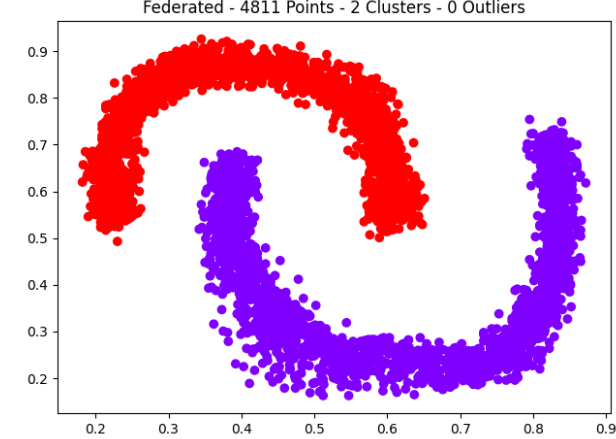
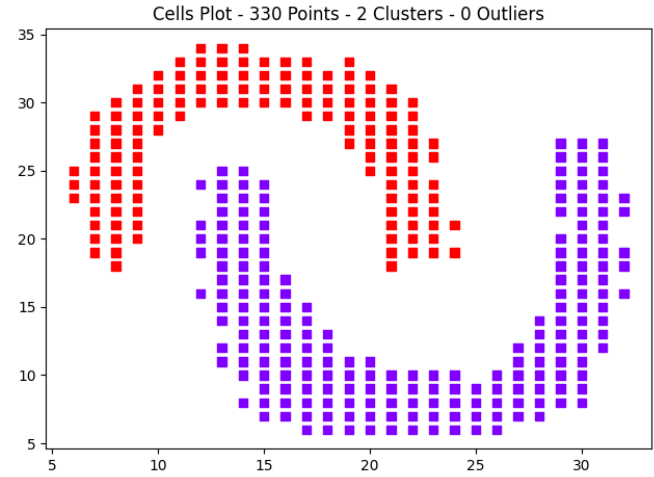
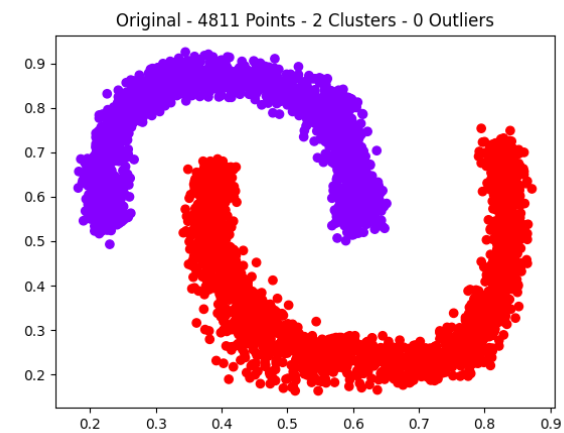
Stratified Partitioning



Separated Partitioning



RESULT WITH BANANA



$L = 0.03$ MinPts = 2

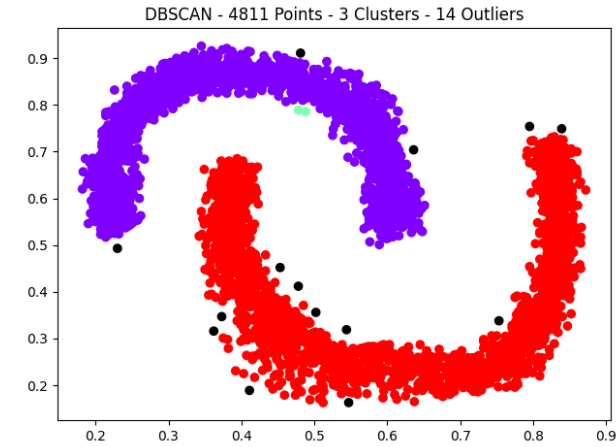
PURITY: 1.0
ARI: 1.0
AMI: 1.0
PRECISION-BCUBED: 1.0
RECALL-BCUBED: 1.0

Outliers

MinPts \ L	0.002	0.004	0.006	0.008	0.01	0.012	0.014	0.016	0.018	0.02	0.022	0.024	0.026	0.028	0.03	0.032	0.034	0.036	0.038
2.0	3902.0	1622.0	549.0	203.0	89.0	48.0	23.0	12.0	7.0	3.0	3.0	1.0	2.0	1.0	0.0	1.0	0.0	0.0	0.0
3.0	4725.0	3379.0	1460.0	585.0	213.0	100.0	46.0	31.0	15.0	9.0	4.0	1.0	2.0	1.0	0.0	1.0	0.0	0.0	0.0
4.0	4793.0	4460.0	2639.0	1089.0	442.0	232.0	83.0	38.0	24.0	21.0	5.0	1.0	3.0	2.0	0.0	1.0	0.0	0.0	0.0
5.0	4811.0	4787.0	3516.0	1752.0	873.0	362.0	193.0	80.0	51.0	28.0	12.0	3.0	3.0	2.0	1.0	1.0	0.0	0.0	0.0
6.0	4811.0	4811.0	4229.0	2605.0	1362.0	523.0	297.0	130.0	59.0	40.0	25.0	19.0	5.0	2.0	0.0	1.0	0.0	0.0	0.0
7.0	4811.0	4811.0	4649.0	3271.0	1745.0	816.0	479.0	200.0	85.0	49.0	27.0	19.0	21.0	6.0	3.0	1.0	0.0	0.0	0.0

Outliers DBSCAN

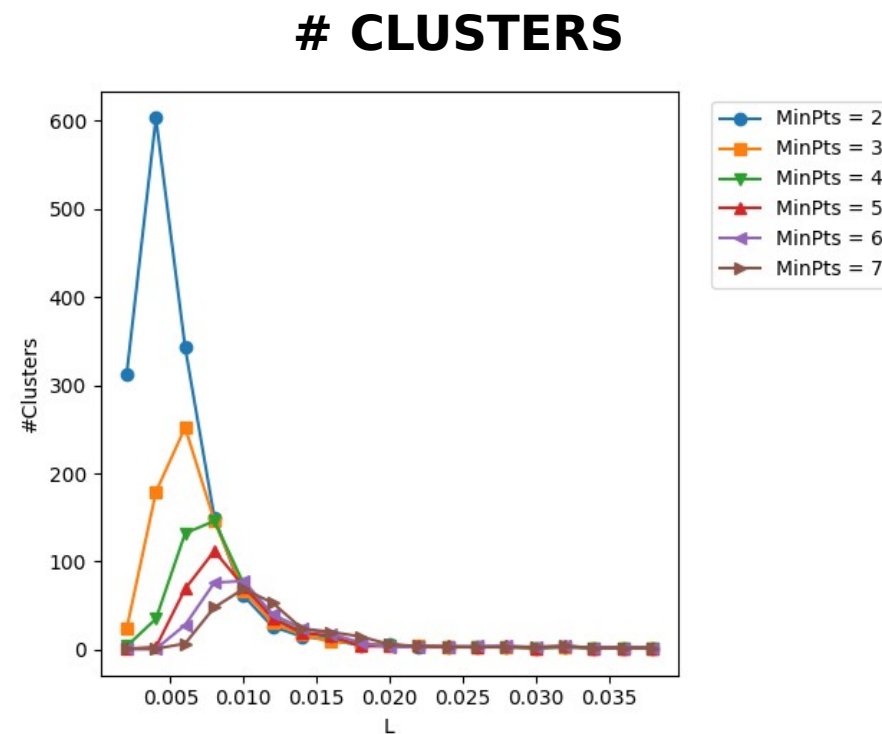
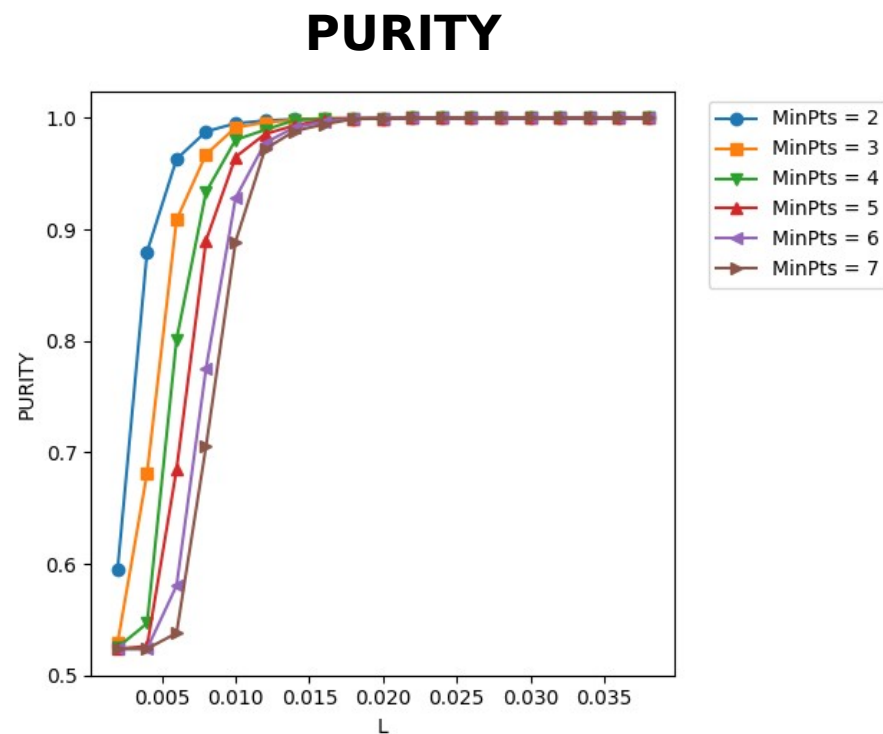
MinPts \ Eps	0.001	0.002	0.003	0.004	0.005	0.006	0.007	0.008	0.009	0.01	0.011	0.012	0.013	0.014	0.015	0.016	0.017	0.018	0.019
2.0	4639.0	3375.0	2021.0	1180.0	757.0	433.0	298.0	204.0	139.0	104.0	70.0	56.0	45.0	34.0	27.0	22.0	19.0	17.0	15.0
3.0	4811.0	4357.0	3035.0	1902.0	1245.0	681.0	470.0	298.0	213.0	152.0	98.0	74.0	59.0	42.0	37.0	32.0	23.0	21.0	17.0
4.0	4811.0	4747.0	3864.0	2694.0	1805.0	1037.0	692.0	458.0	292.0	208.0	141.0	102.0	78.0	56.0	43.0	33.0	27.0	25.0	20.0
5.0	4811.0	4806.0	4429.0	3468.0	2432.0	1417.0	938.0	654.0	398.0	284.0	196.0	151.0	112.0	79.0	56.0	44.0	31.0	27.0	21.0
6.0	4811.0	4811.0	4708.0	4056.0	3010.0	1840.0	1275.0	878.0	602.0	373.0	259.0	193.0	139.0	102.0	74.0	54.0	45.0	31.0	25.0
7.0	4811.0	4811.0	4795.0	4411.0	3596.0	2314.0	1586.0	1108.0	749.0	499.0	344.0	244.0	178.0	136.0	101.0	70.0	52.0	39.0	31.0



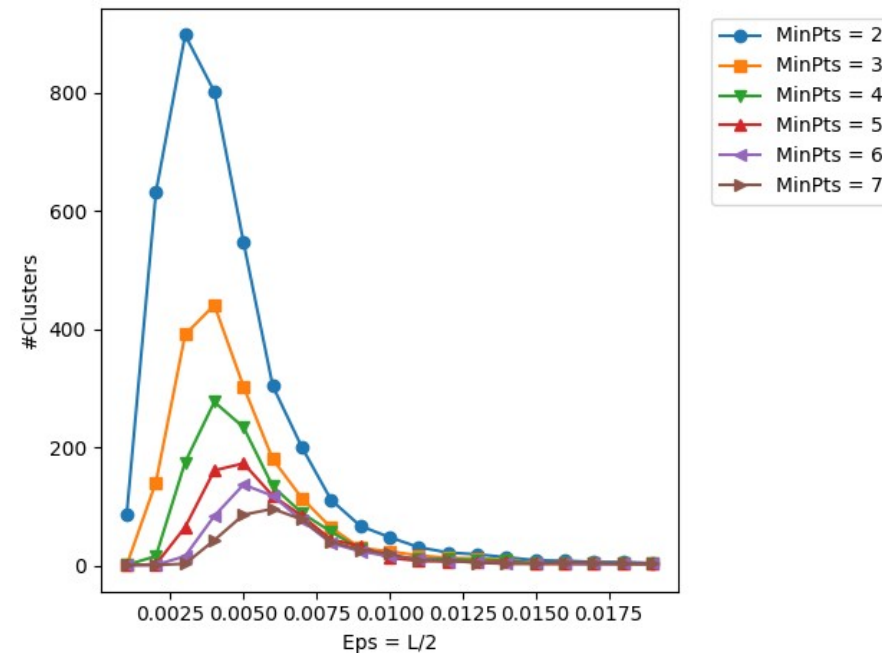
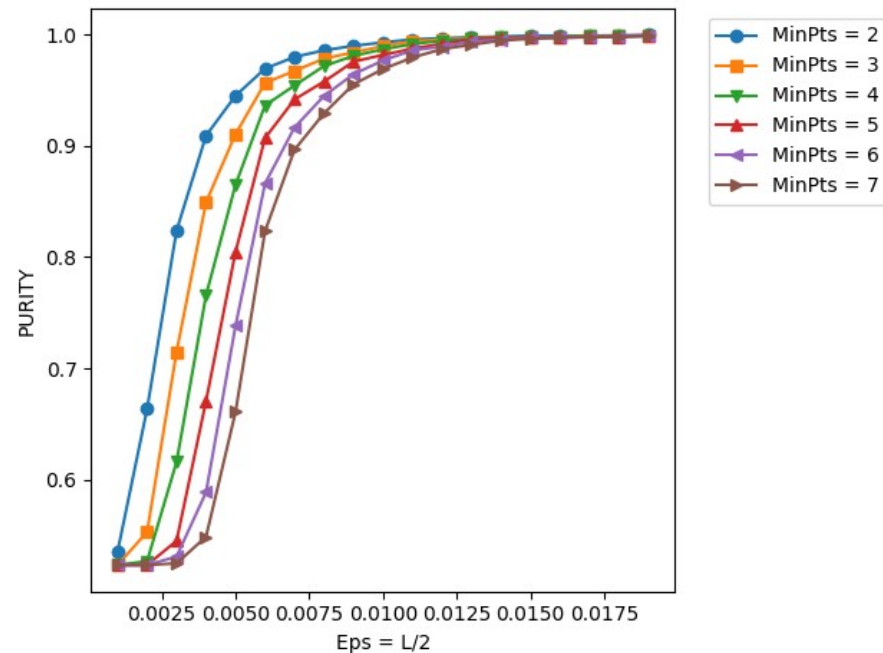
$\epsilon = 0.019$ MinPts = 2

PURITY: 0.9994
ARI: 0.9935
AMI: 0.9828
PRECISION-BCUBED: 0.9990

FEDERATED

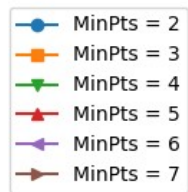
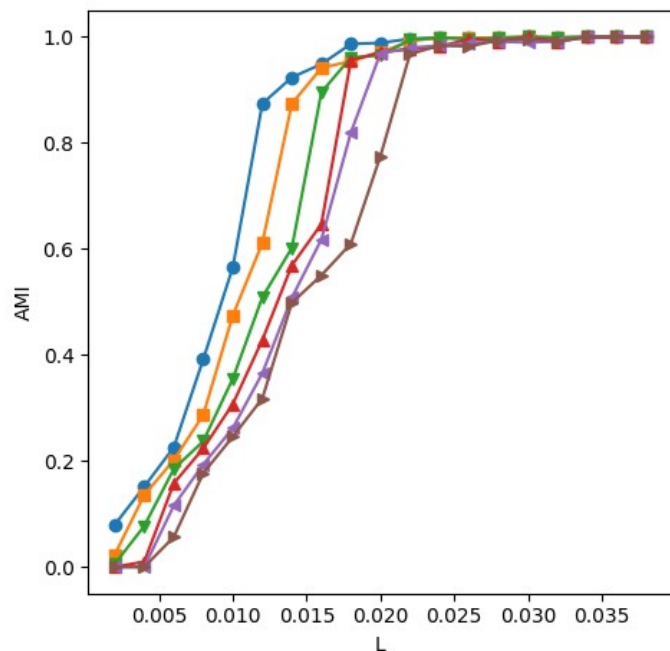


DBSCAN

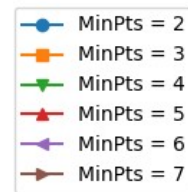
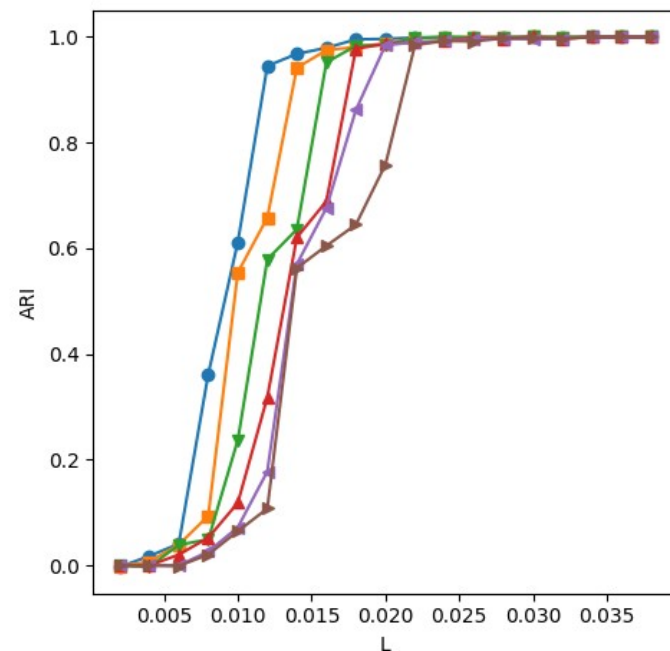


FEDERATED

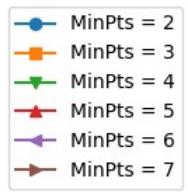
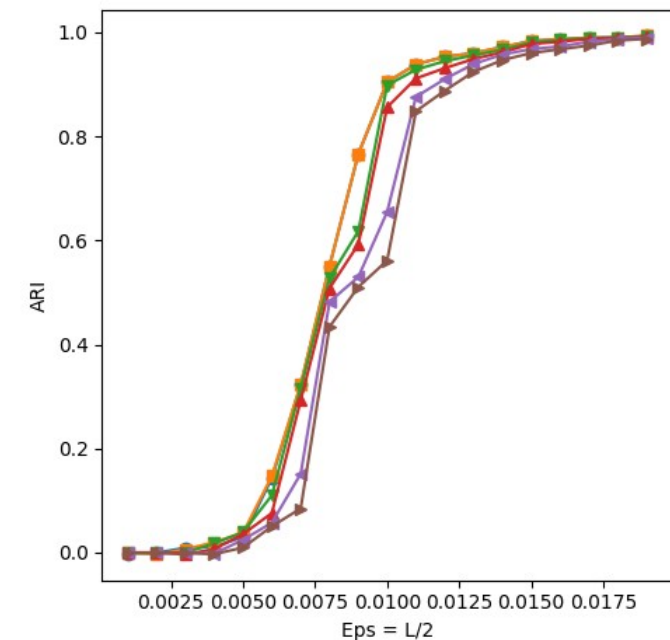
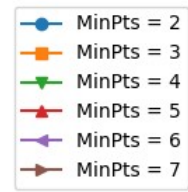
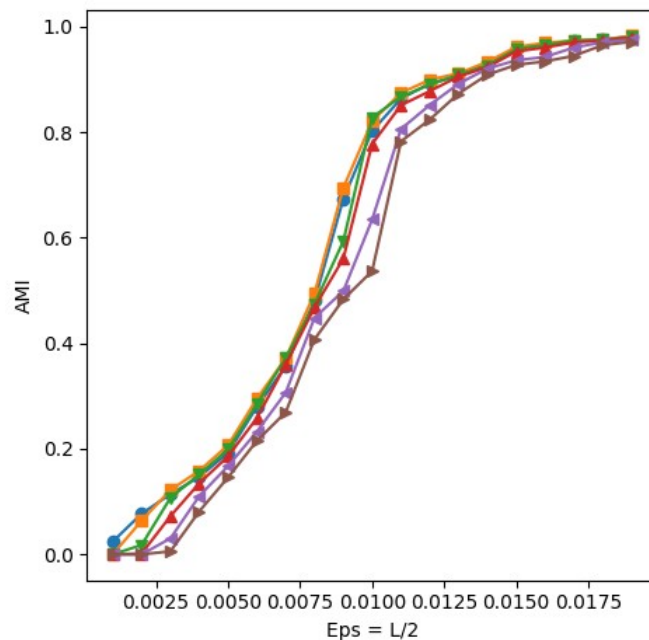
AMI



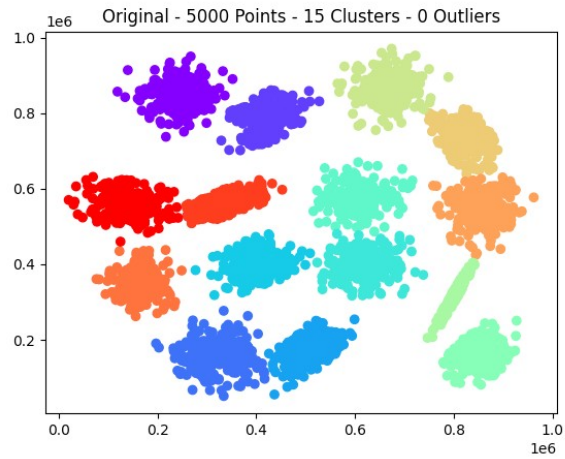
ARI



DBSCAN

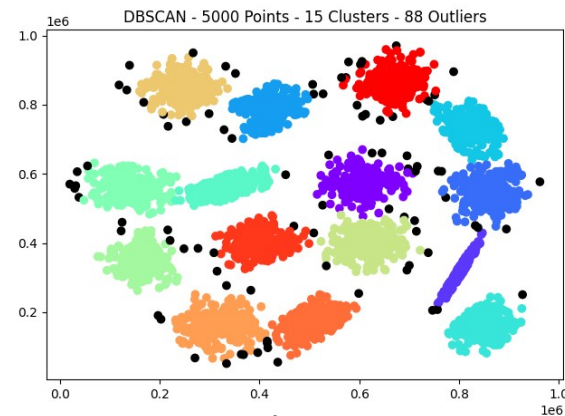
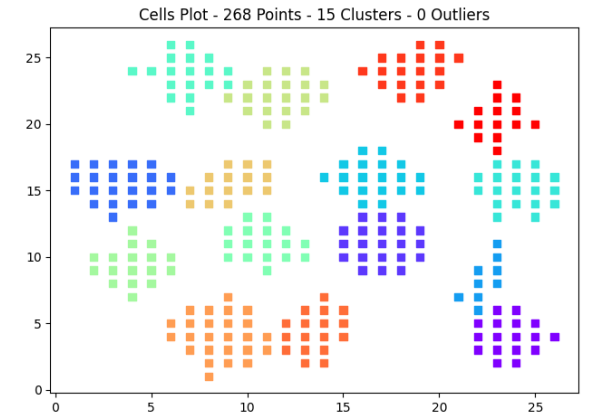


RESULT WITH S-SET-1



Outliers

MinPts \ L	15000	20000	25000	30000	35000	40000	45000	50000	55000	60000	65000
8.0	1148.0	448.0	166.0	72.0	18.0	11.0	0.0	0.0	0.0	0.0	0.0
9.0	1285.0	552.0	217.0	95.0	21.0	11.0	0.0	0.0	0.0	0.0	0.0
10.0	1541.0	676.0	274.0	104.0	29.0	10.0	6.0	0.0	0.0	0.0	0.0
11.0	1552.0	719.0	313.0	119.0	52.0	10.0	7.0	3.0	2.0	0.0	0.0
12.0	1718.0	789.0	356.0	146.0	42.0	12.0	8.0	3.0	0.0	2.0	0.0
13.0	1828.0	924.0	366.0	148.0	57.0	17.0	7.0	3.0	0.0	0.0	0.0
14.0	1937.0	899.0	412.0	188.0	57.0	21.0	13.0	3.0	0.0	0.0	0.0
15.0	2259.0	946.0	461.0	242.0	88.0	44.0	30.0	5.0	7.0	0.0	0.0

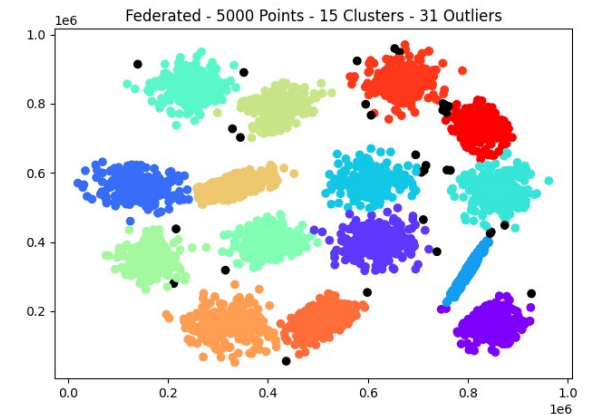


$\epsilon = 31500$ MinPts = 15

PURITY: 0.9842
ARI: 0.9763
AMI: 0.9741
PRECISION-BCUBED:
 0.9818
RECALL-BCUBED:

Outliers

MinPts \ Eps	7500.0	10000.0	12500.0	15000.0	17500.0	20000.0	22500.0	25000.0	27500.0	30000.0	32500.0	35000.0
8.0	2202.0	1412.0	835.0	569.0	367.0	238.0	178.0	125.0	85.0	54.0	35.0	22.0
9.0	2392.0	1588.0	991.0	623.0	416.0	270.0	207.0	146.0	96.0	61.0	41.0	23.0
10.0	2567.0	1740.0	1091.0	723.0	473.0	306.0	220.0	160.0	108.0	75.0	49.0	28.0
11.0	2717.0	1882.0	1215.0	808.0	518.0	350.0	237.0	174.0	120.0	84.0	49.0	32.0
12.0	2853.0	2023.0	1330.0	876.0	571.0	383.0	271.0	190.0	132.0	89.0	52.0	35.0
13.0	2956.0	2146.0	1421.0	953.0	633.0	414.0	290.0	199.0	140.0	100.0	56.0	37.0
14.0	3070.0	2267.0	1520.0	1043.0	687.0	474.0	321.0	224.0	153.0	111.0	64.0	39.0
15.0	3178.0	2332.0	1614.0	1110.0	744.0	498.0	348.0	244.0	159.0	113.0	77.0	47.0

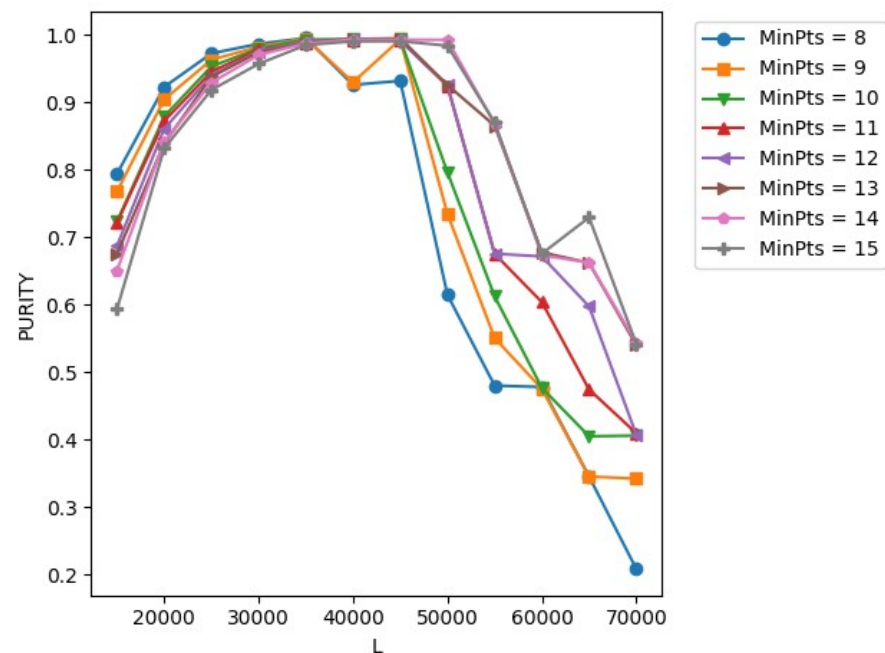


$L = 34750$ MinPts = 9

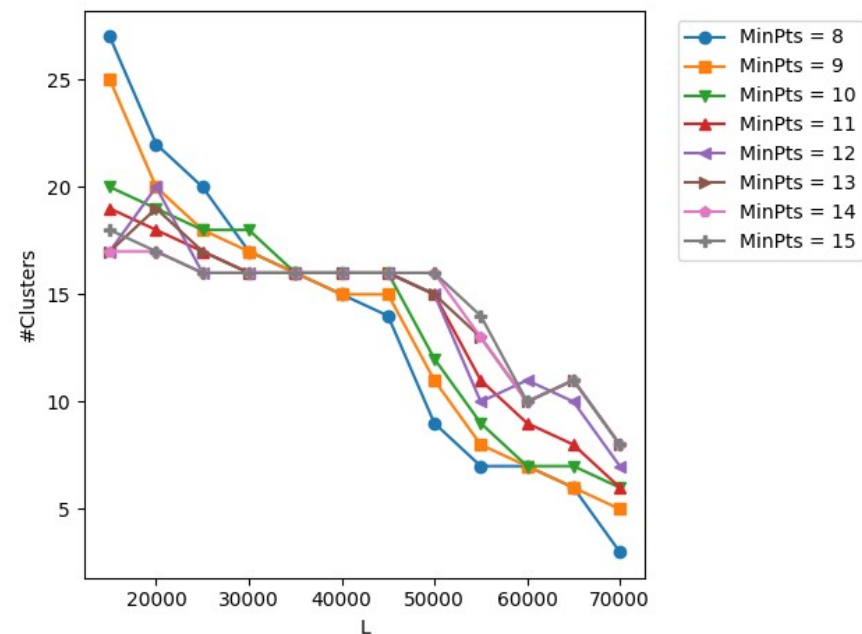
PURITY: 0.9962
ARI: 0.9885
AMI: 0.9860
PRECISION-BCUBED:
 0.9902
RECALL-BCUBED:

FEDERATED

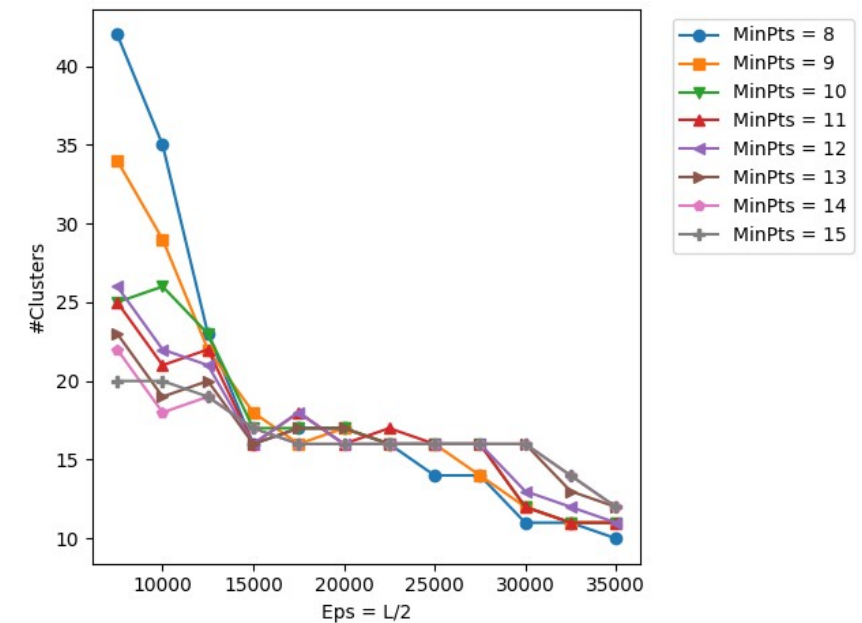
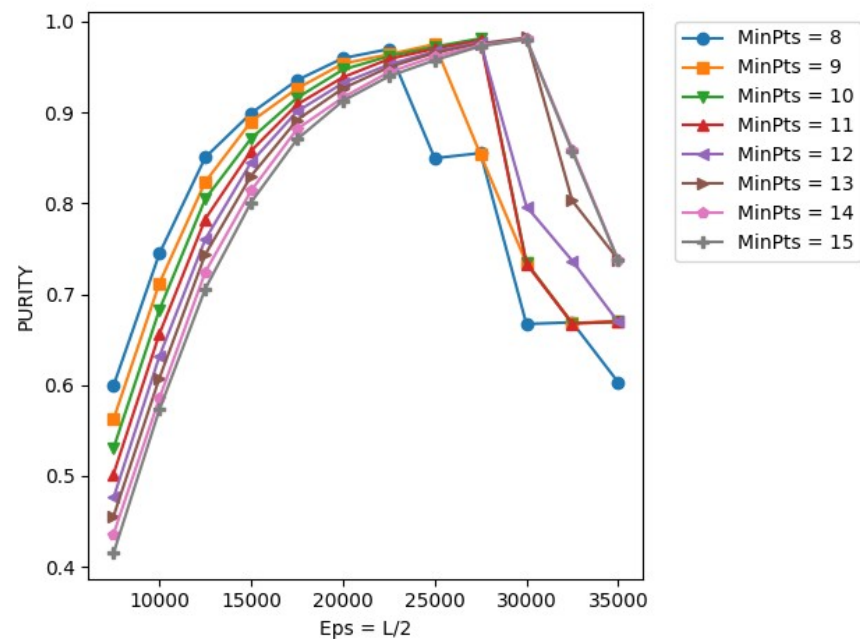
PURITY



CLUSTERS

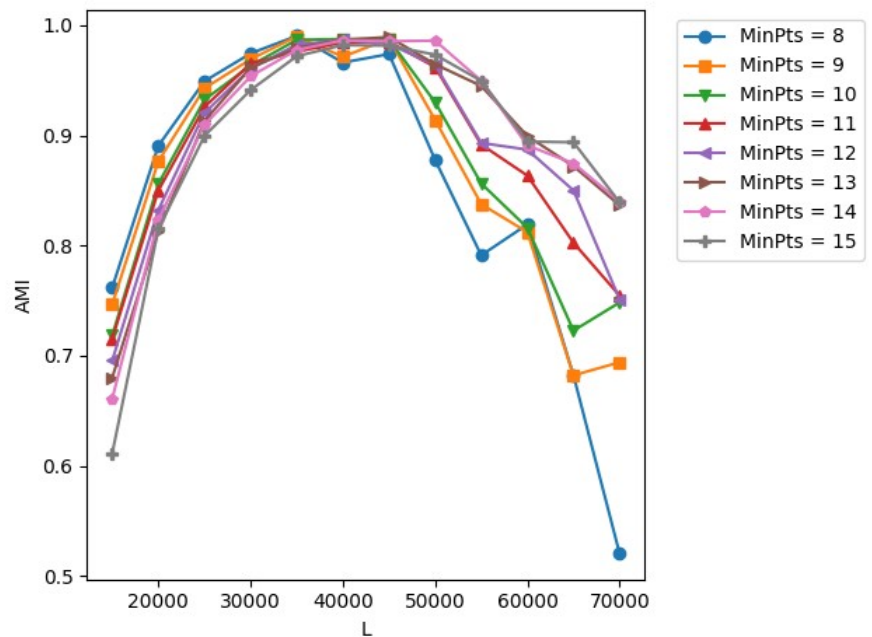


DBSCAN

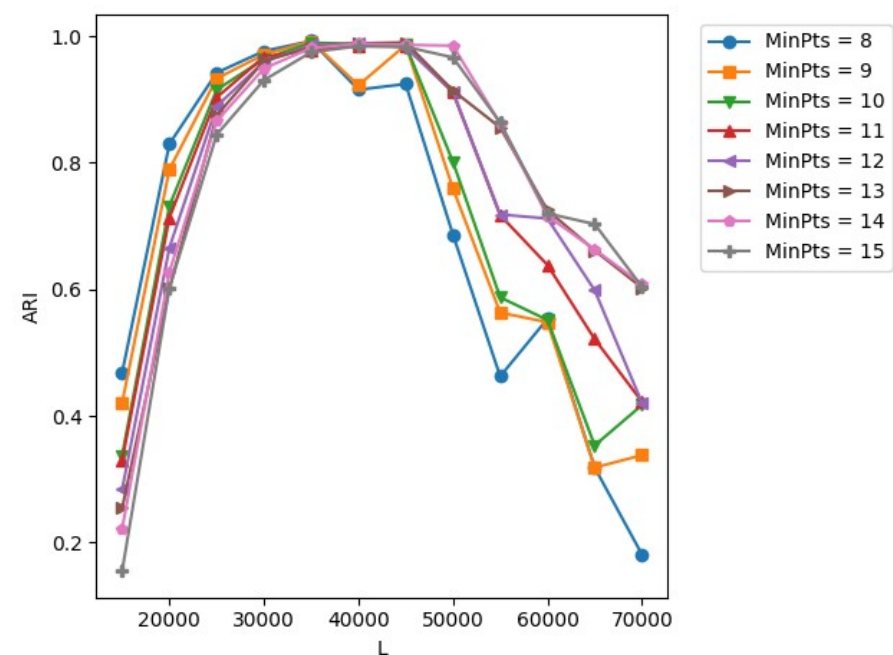


FEDERATED

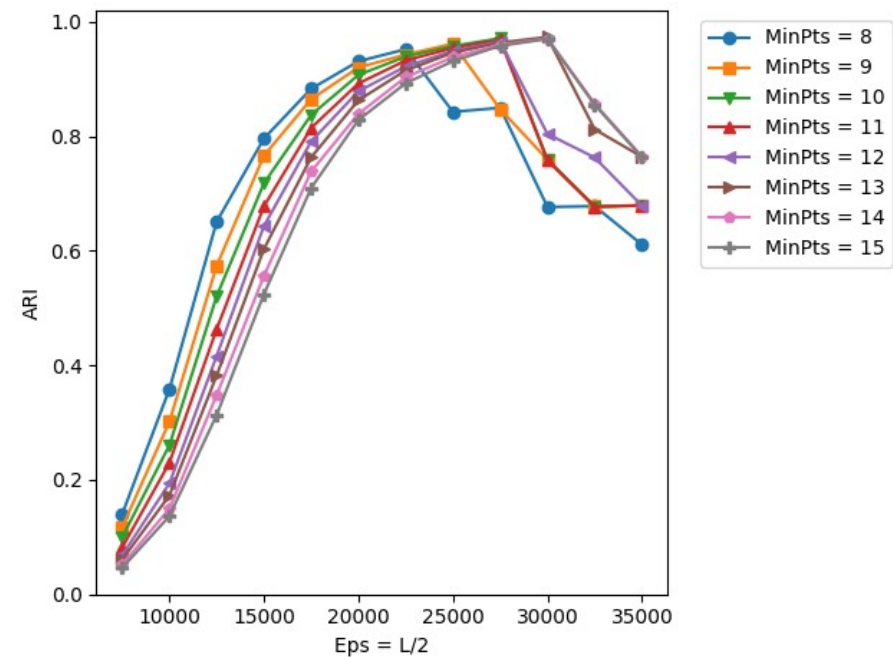
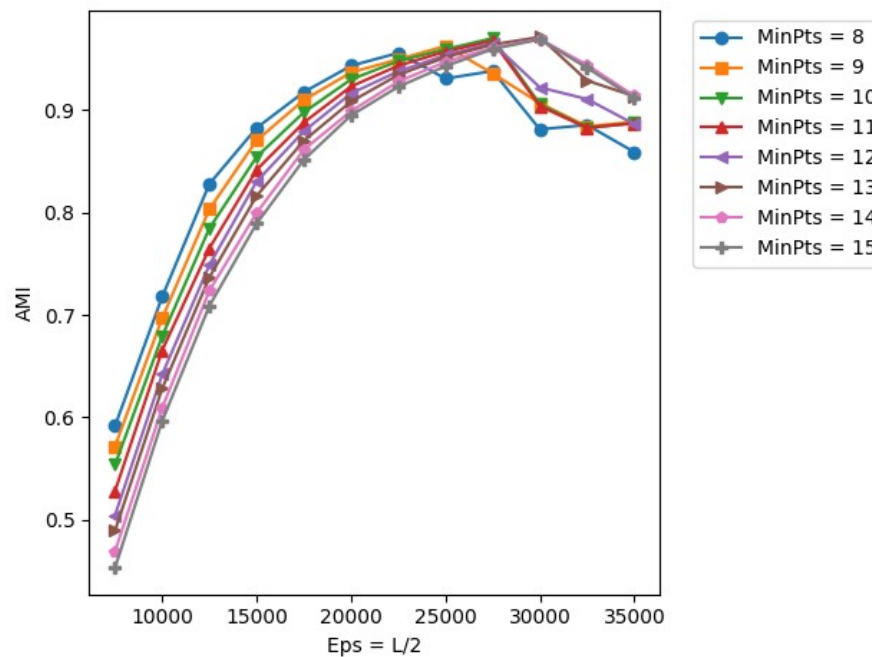
AMI



ARI

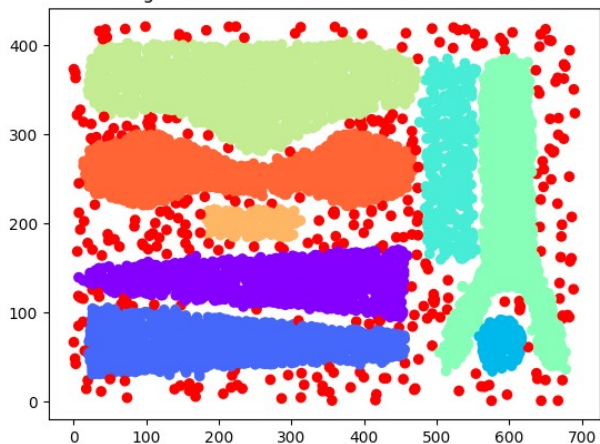


DBSCAN



RESULTS WITH CLUTO

Original - 8000 Points - 9 Clusters - 0 Outliers



PURITY_DBSCAN

MinPts\Ep	5.5	5.75	6.0	6.25	6.5	6.75	7.0	7.25	7.5	7.75	8.0	8.25
3.0	0.9608	0.9687	0.9753	0.9795	0.8783	0.8755	0.8231	0.8131	0.8121	0.8115	0.8107	0.8103
4.0	0.9287	0.9457	0.9568	0.9641	0.973	0.9776	0.8358	0.8113	0.8115	0.8113	0.8106	0.8103
5.0	0.8802	0.9017	0.9238	0.939	0.9532	0.9642	0.9713	0.9776	0.8493	0.8117	0.8107	0.8106

AMI_DBSCAN

MinPts\Ep	5.5	5.75	6.0	6.25	6.5	6.75	7.0	7.25	7.5	7.75	8.0	8.25
3.0	0.663	0.7373	0.7685	0.8021	0.8061	0.8162	0.8556	0.8711	0.8761	0.8836	0.8856	0.893
4.0	0.6303	0.674	0.716	0.7578	0.8228	0.8346	0.8353	0.8649	0.8697	0.8852	0.8859	0.8887
5.0	0.5383	0.6118	0.6527	0.7044	0.7441	0.8104	0.8565	0.882	0.8453	0.8643	0.8709	0.8752

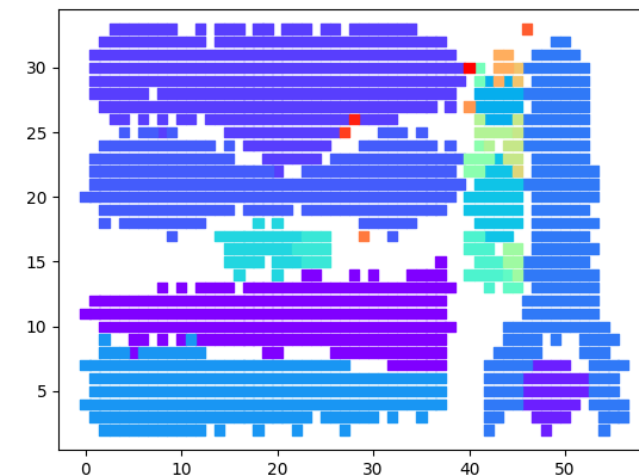
PURITY_FEDERATED

MinPts\L	11.0	11.5	12.0	12.5	13.0	13.5	14.0	14.5	15.0	15.5	16.0	16.5
3.0	0.8193	0.6341	0.6332	0.6086	0.5878	0.5853	0.391	0.3893	0.3872	0.3877	0.2027	0.385
4.0	0.8467	0.93	0.9742	0.657	0.653	0.6238	0.6326	0.4118	0.3887	0.4065	0.3851	0.3861
5.0	0.9706	0.971	0.9681	0.9727	0.834	0.815	0.6482	0.6093	0.3912	0.5818	0.3863	0.406

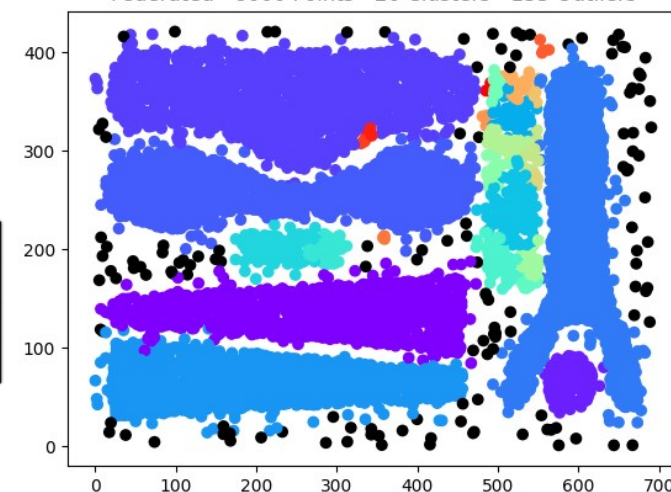
AMI_FEDERATED

MinPts\L	11.0	11.5	12.0	12.5	13.0	13.5	14.0	14.5	15.0	15.5	16.0	16.5
3.0	0.8484	0.7716	0.771	0.7409	0.733	0.7277	0.4929	0.4979	0.4872	0.494	0.025	0.4825
4.0	0.8057	0.8485	0.9148	0.7759	0.7807	0.7524	0.7537	0.5426	0.4891	0.5341	0.4811	0.4843
5.0	0.7458	0.8081	0.827	0.8453	0.8172	0.7989	0.7643	0.7331	0.4901	0.7036	0.4829	0.5319

Cells Plot - 1302 Points - 26 Clusters - 0 Outliers



Federated - 8000 Points - 26 Clusters - 133 Outliers



$\epsilon = 7.25$ MinPts = 5

PURITY: 0.9776

ARI: 0.8305

AMI: 0.882

PREC-BCUBED: 0.9427

REC-BCUBED: 0.5400

F-SCORE: 0.6867

$L = 12$ MinPts = 4

PURITY: 0.9748

ARI: 0.9536

AMI: 0.9148

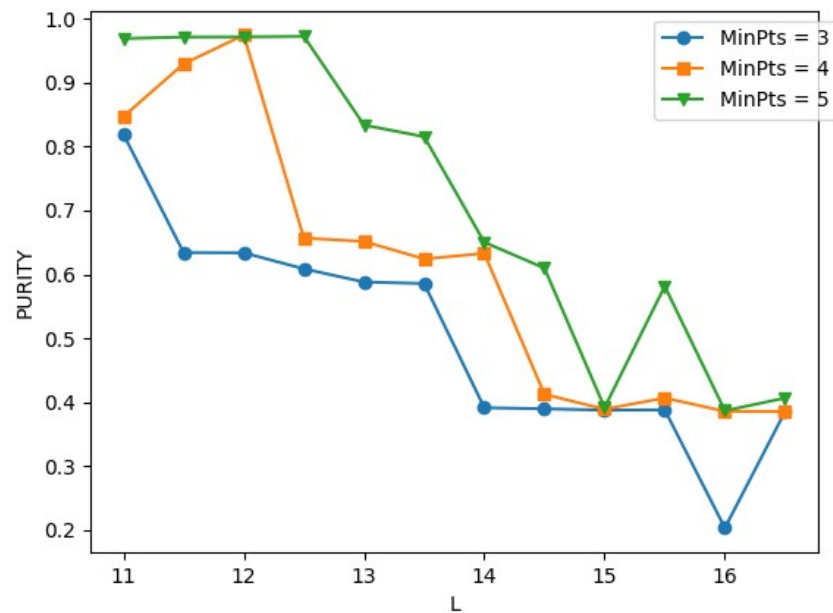
PREC-BCUBED:

0.9504

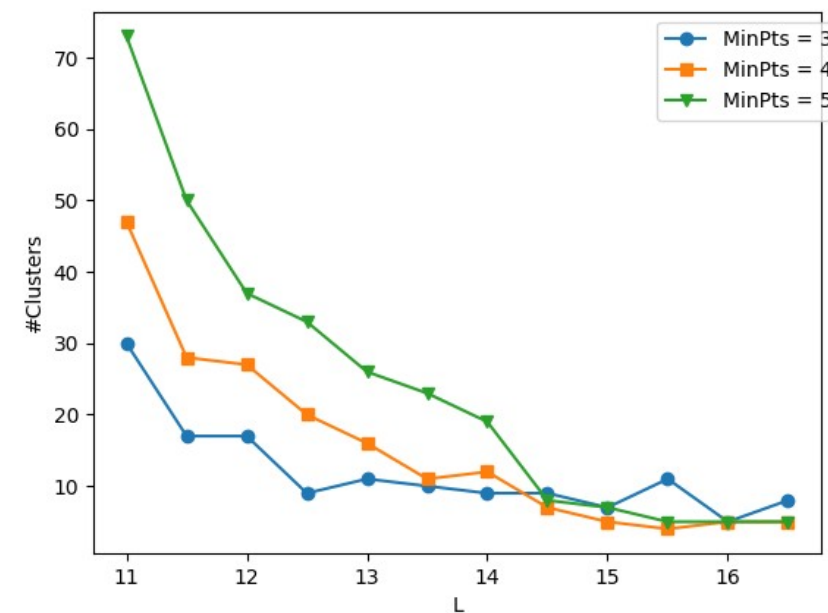
REC-BCUBED: 0.9148

FEDERATED

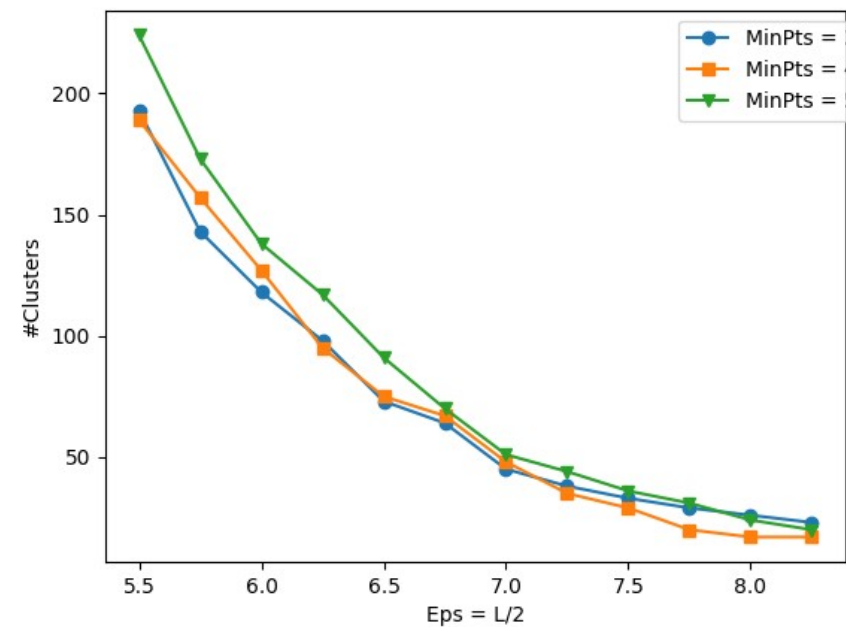
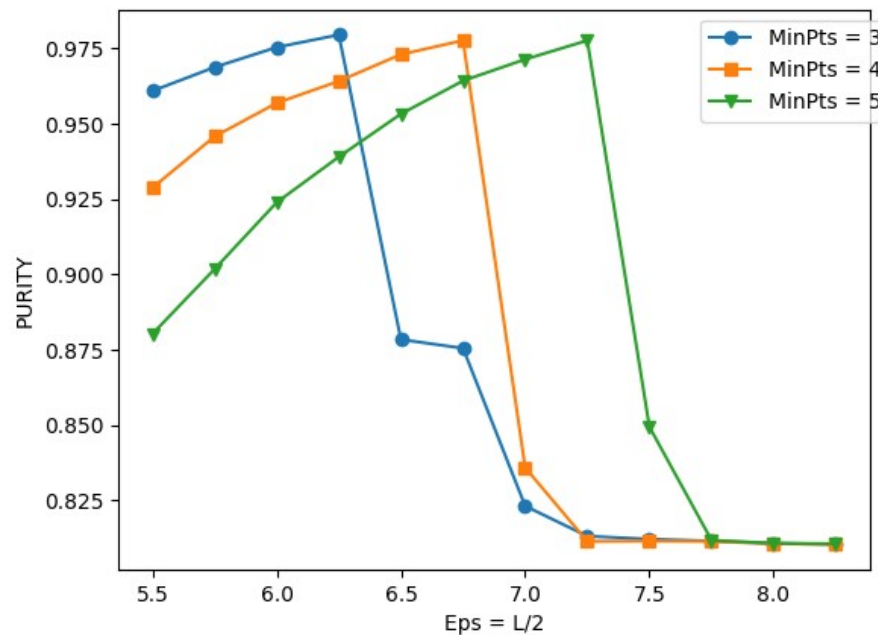
PURITY



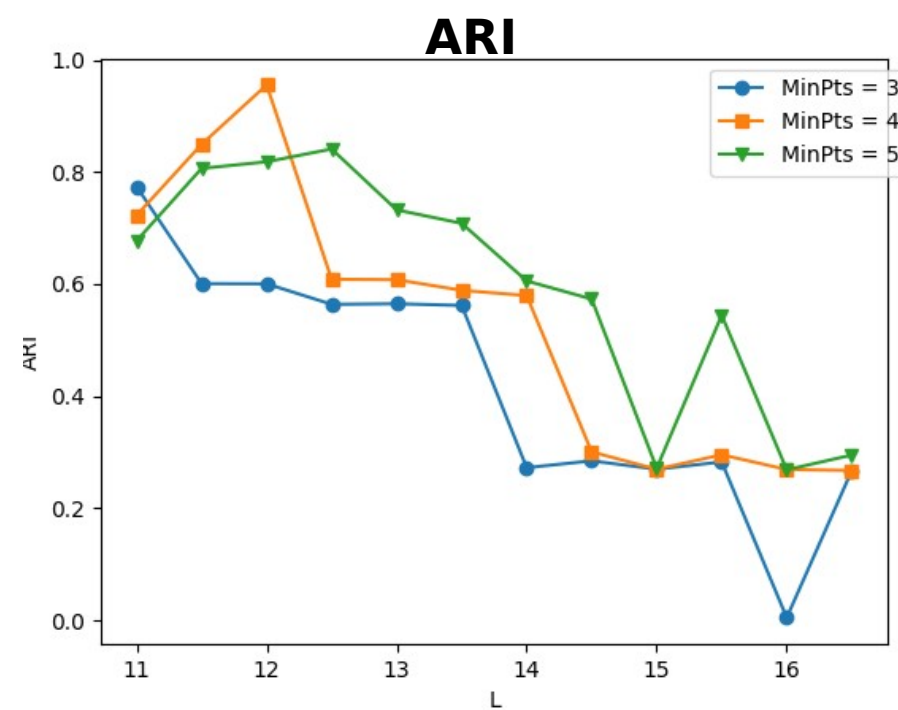
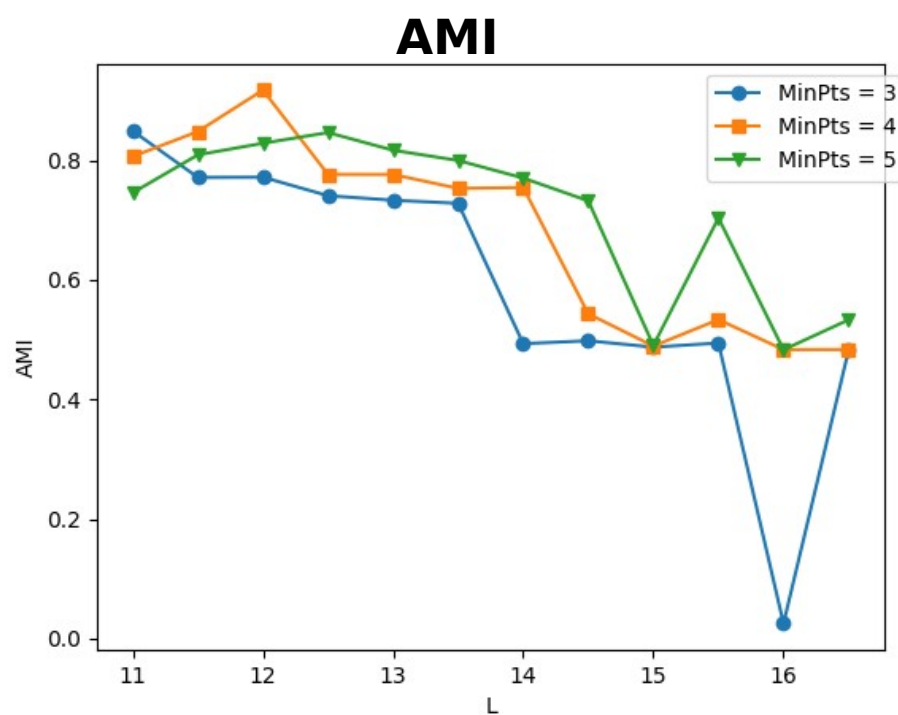
CLUSTERS



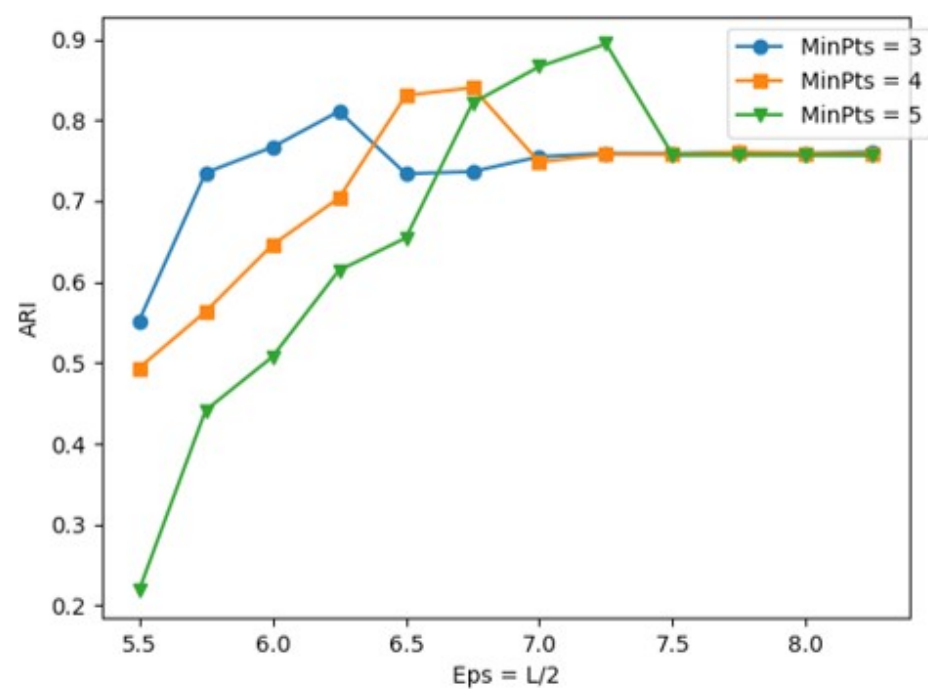
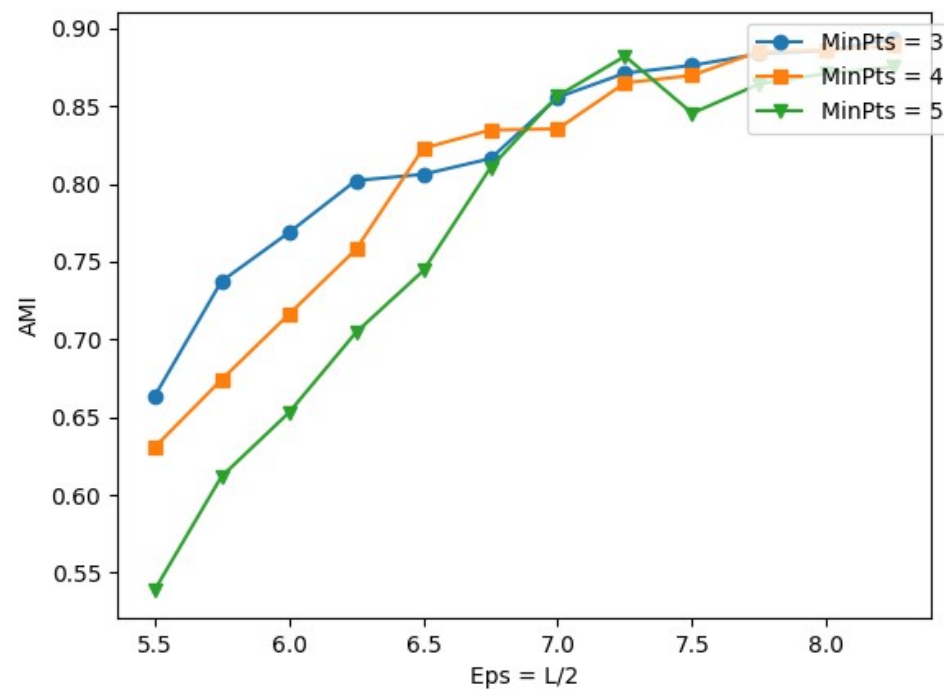
DBSCAN



FEDERATED



DBSCAN



RESULTS CHAINLINK

