



# **Generalized centroid index**

## Cluster level quality measure

**Pasi Fränti and Mohammad Rezaei**

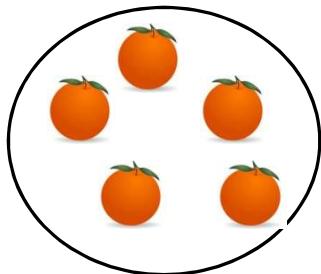
30.11.2016

# **Clustering accuracy**

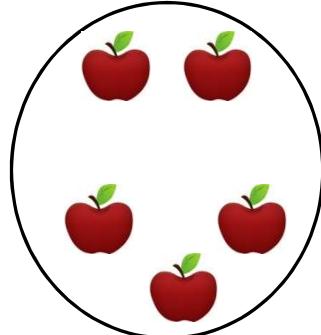
# Classification accuracy

Known class labels

Solution A:

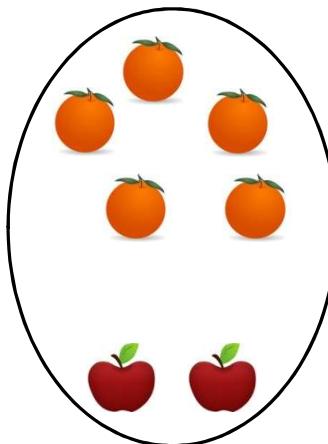


**Oranges**  
100 %



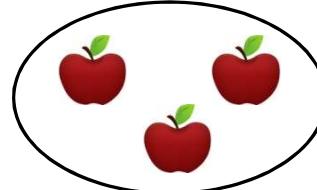
**Apples**  
100 %

Solution B:



**Oranges**  
Precision  
Recall

$$\begin{aligned} &= 5/7 = 71\% \\ &= 5/5 = 100\% \end{aligned}$$



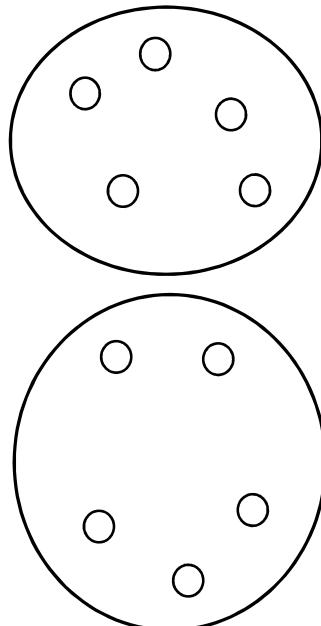
**Apples**  
Precision  
Recall

$$\begin{aligned} &= 3/3 = 100\% \\ &= 3/5 = 60\% \end{aligned}$$

# Clustering accuracy

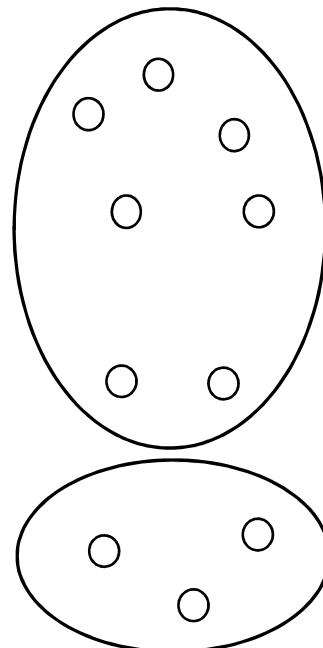
No class labels!

Solution A:



???

Solution B:

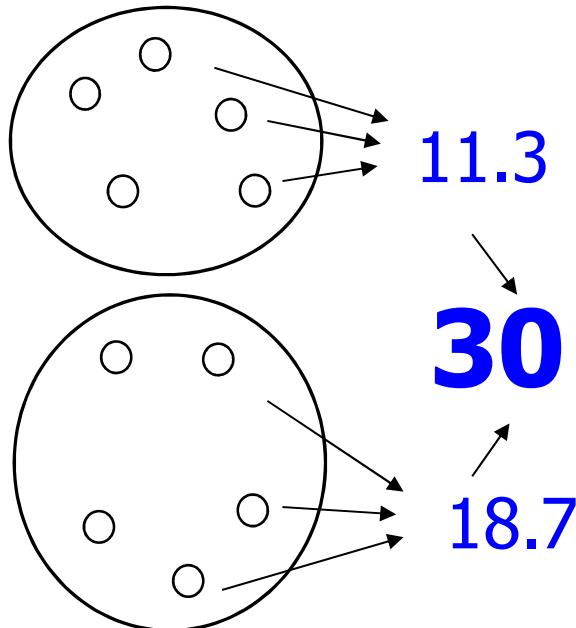


# Internal index

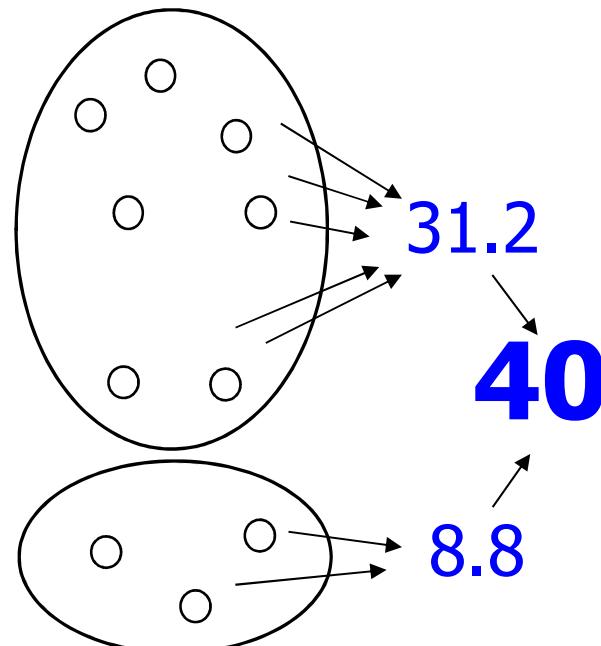
## Sum-of-squared error (MSE)

$$f = \frac{1}{N} \sum_{i=1}^N \sum_{x_i \in c_k} \|x_i - c_k\|^2$$

Solution A:



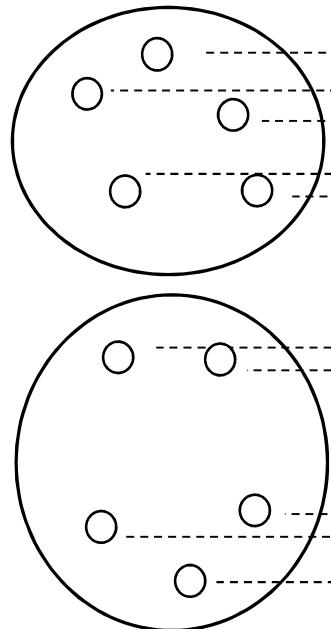
Solution B:



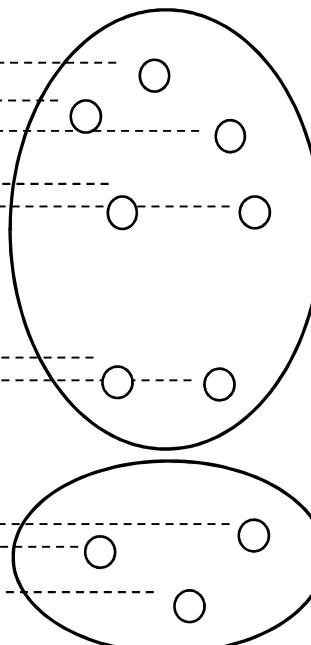
# External index

Compare two solutions

Solution A:



Solution B:



$5/7 = 71\%$

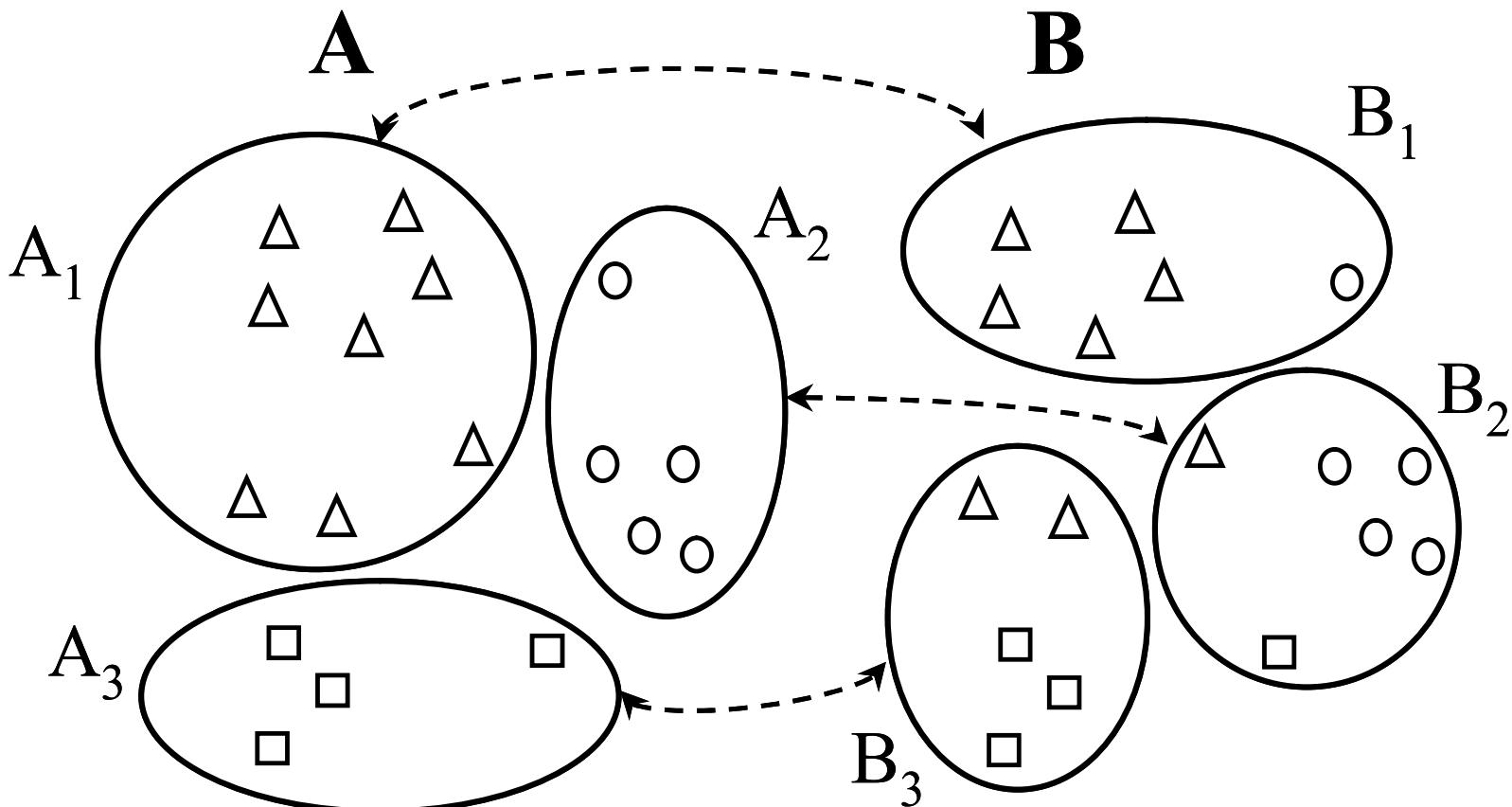
$2/5 = 40\%$

$3/5 = 60\%$

- Two clustering (A and B)
- Clustering against ground truth

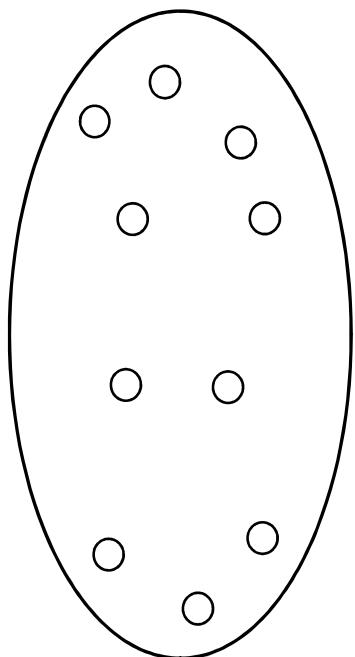
# Set-matching based methods

M. Rezaei and P. Fränti, "Set-matching methods for external cluster validity",  
*IEEE Trans. on Knowledge and Data Engineering*, 28 (8), 2173-2186, August 2016.



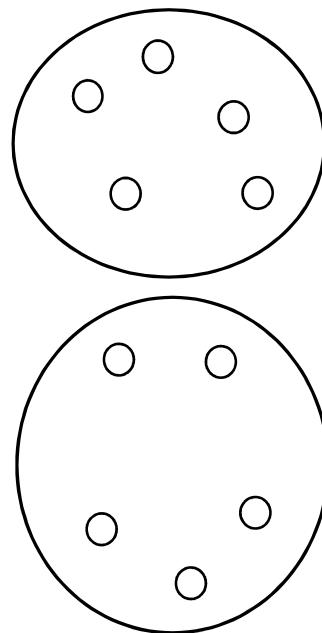
# What about this...?

Solution 1:



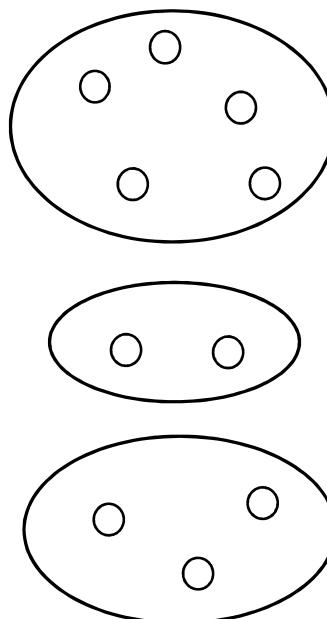
?

Solution 2:



?

Solution 3:



# **External index**

## Selection of existed methods

### Pair-counting measures

- Rand index (RI) [Rand, 1971]
- Adjusted Rand index (ARI) [Hubert & Arabie, 1985]

### Information-theoretic measures

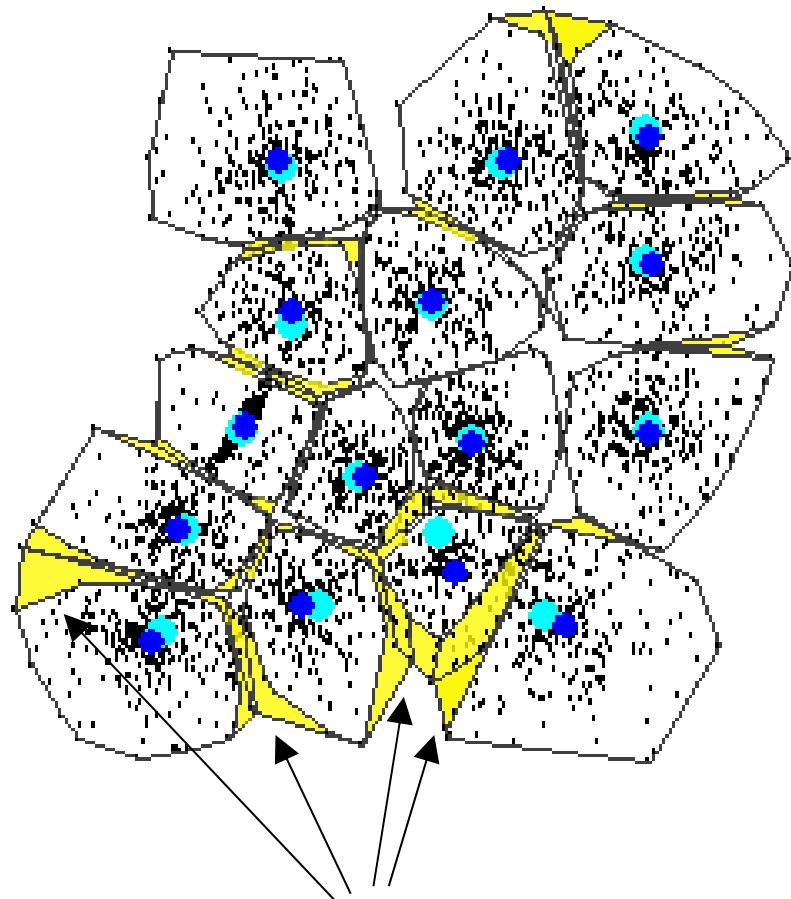
- Mutual information (MI) [Vinh, Epps, Bailey, 2010]
- Normalized Mutual information (NMI) [Kvalseth, 1987]

### Set-matching based measures

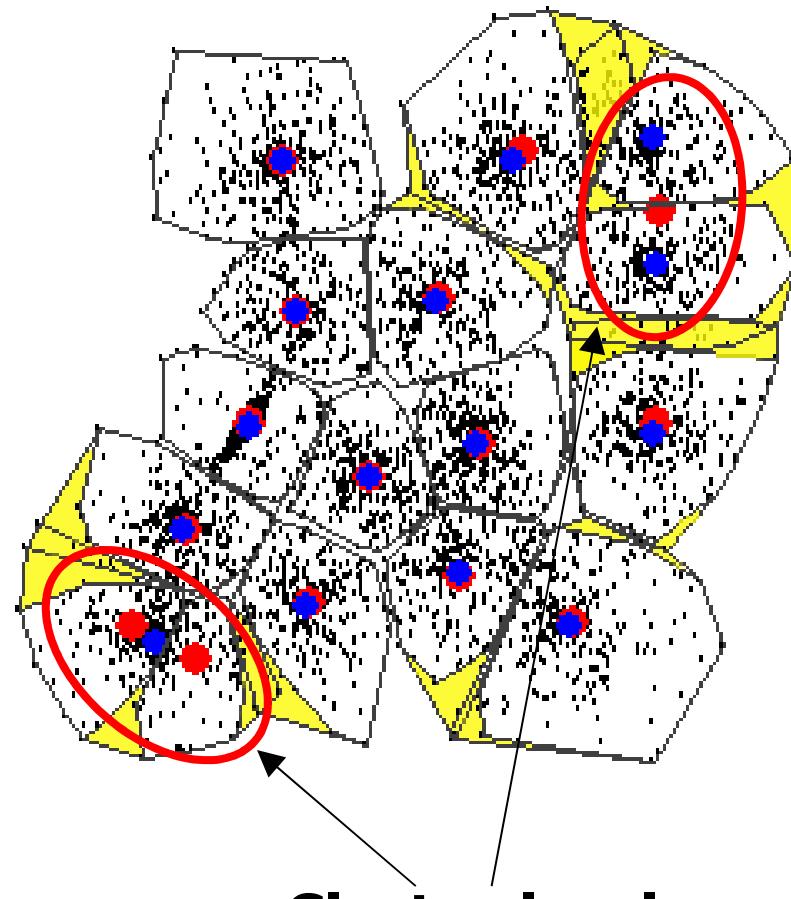
- Normalized van Dongen (NVD) [Kvalseth, 1987]
- Criterion H (CH) [Meila & Heckerman, 2001]
- Purity [Rendon et al, 2011]
- Centroid index (CI) [Fränti, Rezaei & Zhao, 2014]

# **Cluster level measure**

# Point level vs. cluster level



**Point-level  
differences**

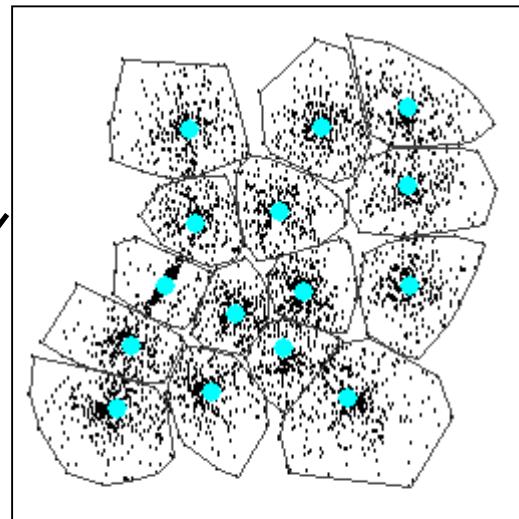


**Cluster-level  
mismatches**

# Point level vs. cluster level

Agglomerative (AC)

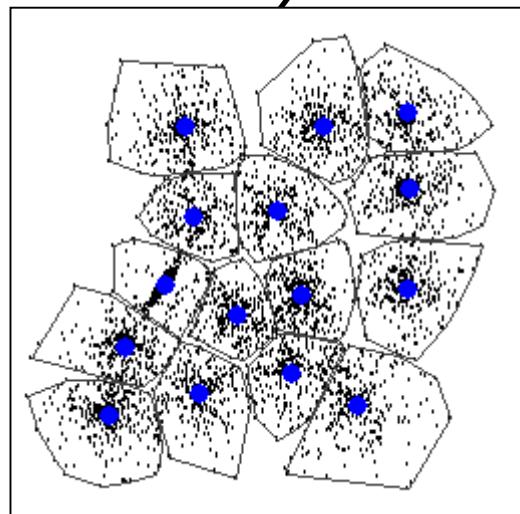
**ARI=0.91**



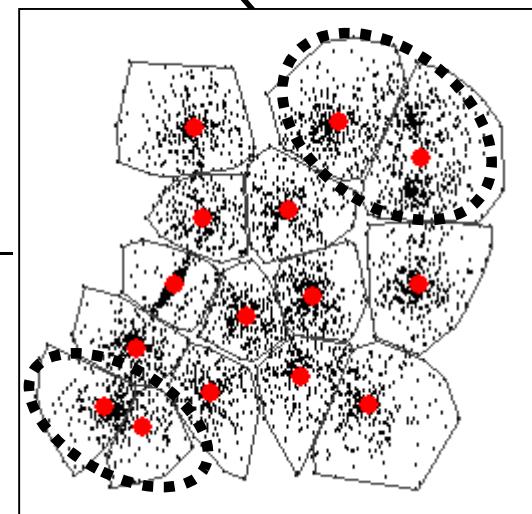
**ARI=0.82**

Random Swap (RS)

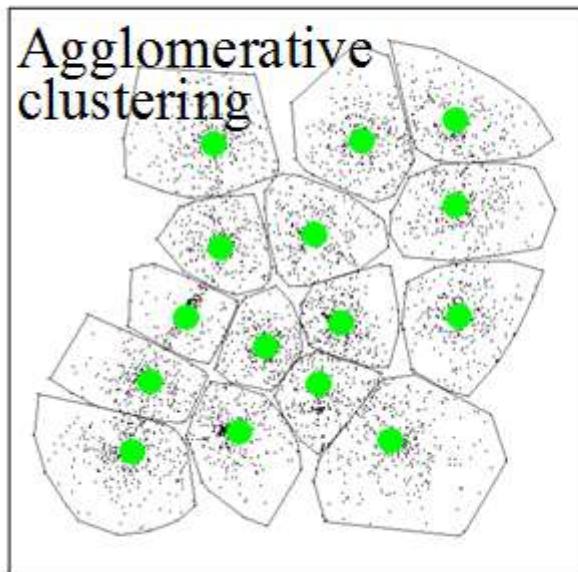
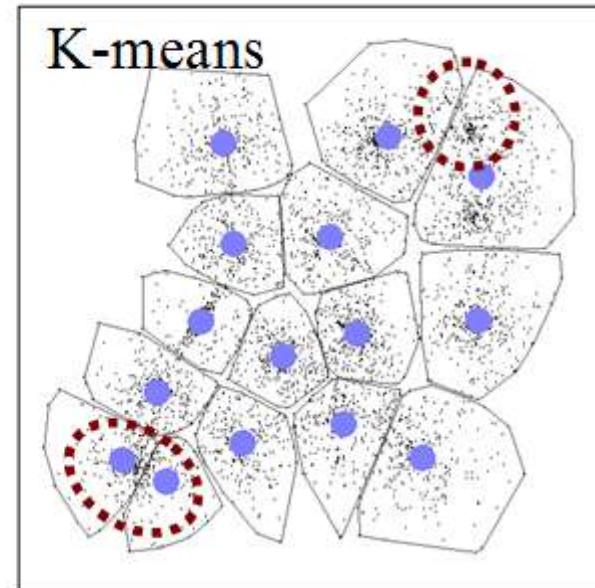
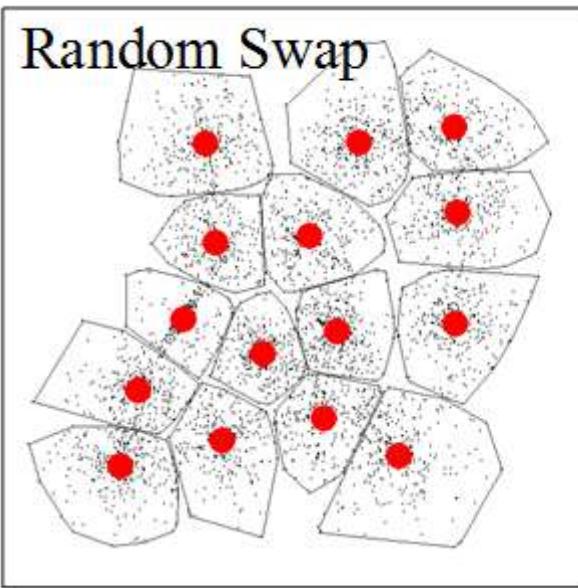
**ARI=0.88**



K-means



# Point level vs. cluster level

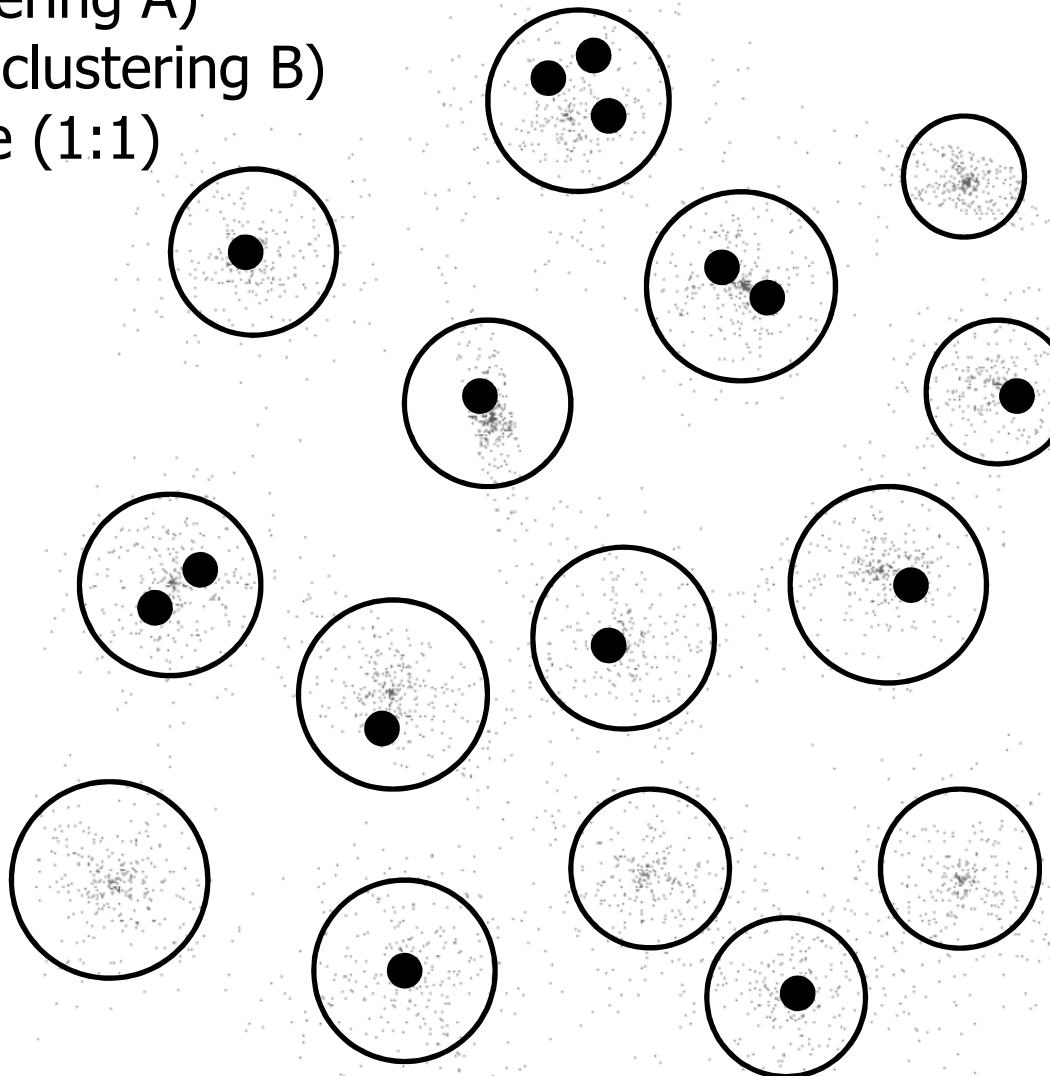


	RS AC	RS KM	AC KM
RI	0.99	0.99	0.98
ARI	0.91	0.88	0.82
MI	3.64	3.64	3.48
NMI	0.93	0.94	0.90
NVD	0.05	0.07	0.10
CH	0.05	0.10	0.14
CI	0	1	1

# **Centroid index**

# Pigeon hole principle

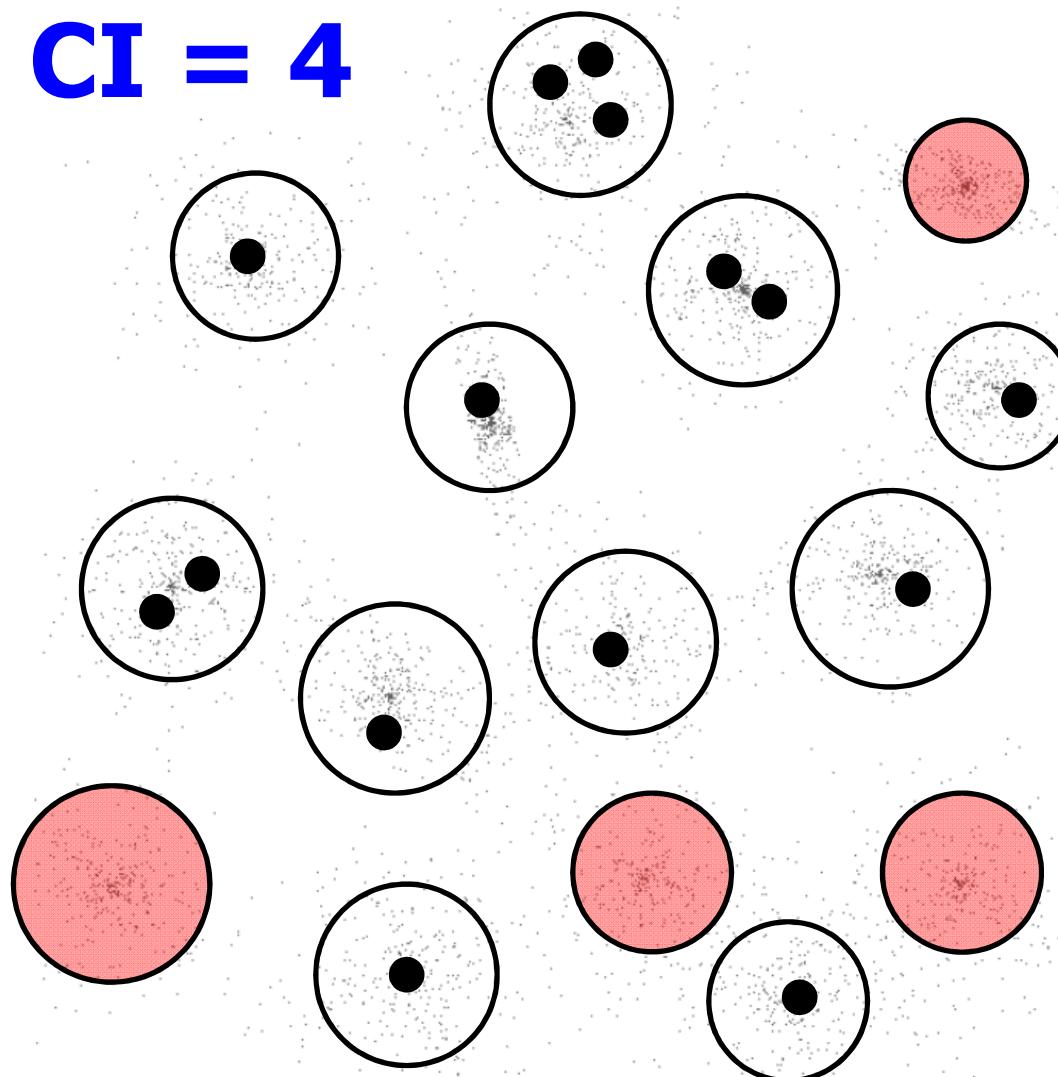
- 15 pigeons (clustering A)
- 15 pigeon holes (clustering B)
- Only one bijective (1:1) mapping exists



# Centroid index (CI)

[Fränti, Rezaei, Zhao, *Pattern Recognition* 2014]

$$\text{CI} = 4$$



# Definitions

Find nearest centroids ( $A \rightarrow B$ ):

$$NN(A_i) = \arg \min_{1 \leq j \leq k} \|c[A_i] - c[B_j]\|$$

Detect prototypes with no mapping:

$$Orphan(B) = \begin{cases} 1 & InDegree(A) = 0 \\ 0 & InDegree(A) > 0 \end{cases}$$

Number of orphans:

$$CI_1(A \rightarrow B) = \sum_{j=1}^k Orphan(B_j)$$

Number of zero mappings!

Centroid index:

$$CI(A, B) = \max \{CI_1(A \rightarrow B), CI_1(B \rightarrow A)\}$$

Mapping both ways

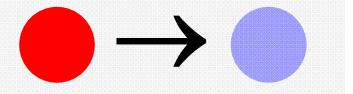
# Example

$S_2$

**Indegree  
Counts**

**CI = 2**

**Mappings**

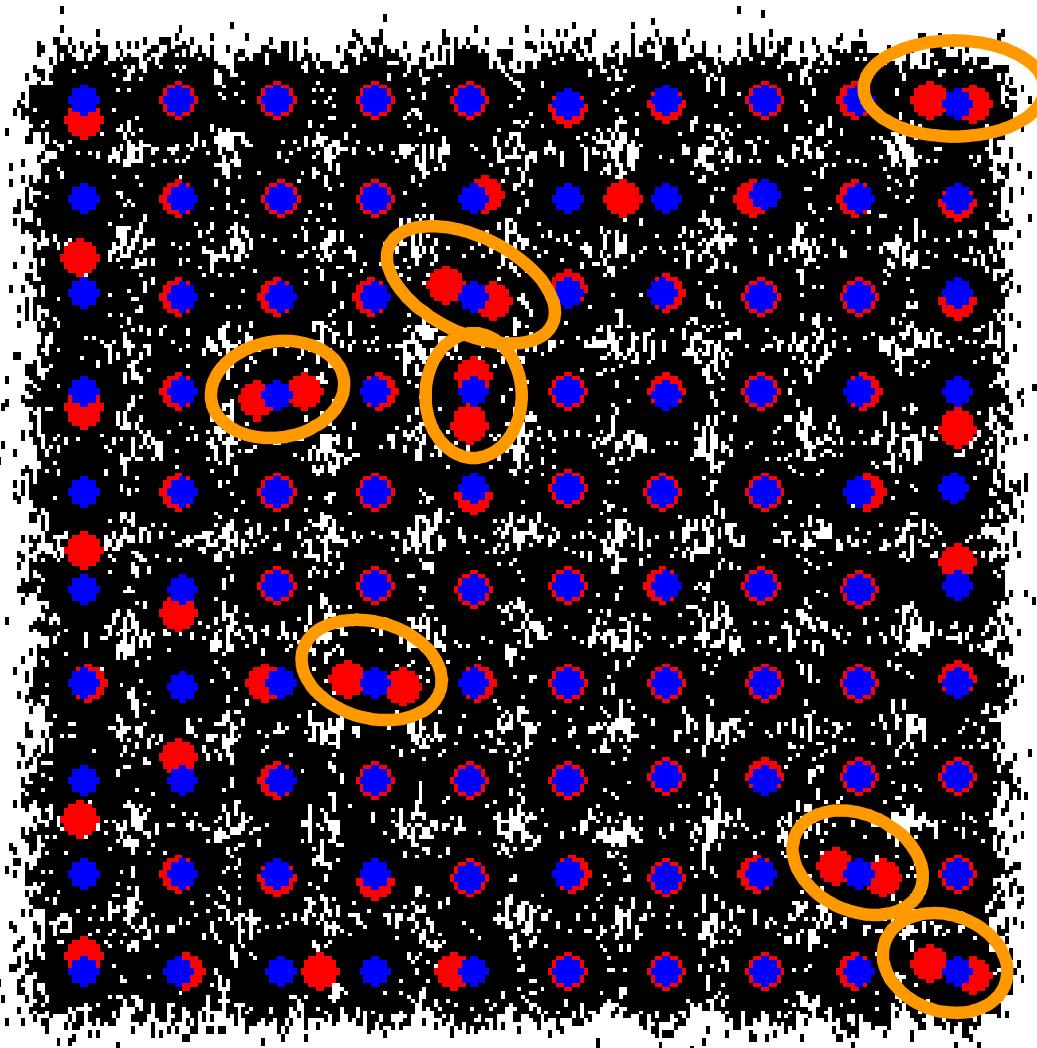
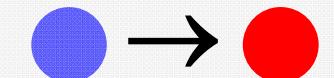


**Value 0 indicates  
an orphan cluster**

# Example

## Birch1

Mappings

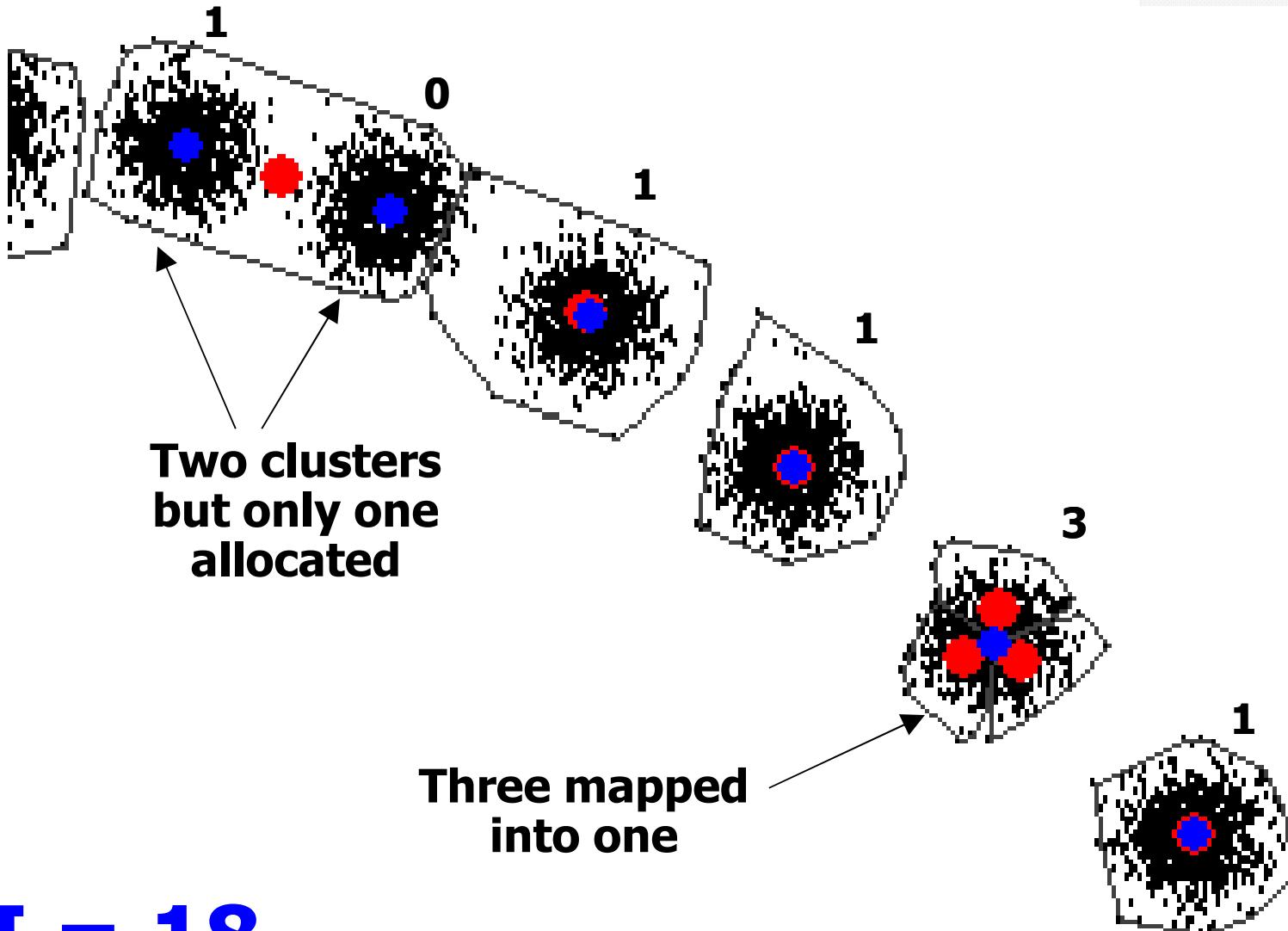
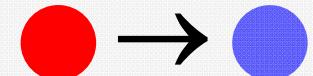


**CI = 7**

# Example

## Birch2

Mappings

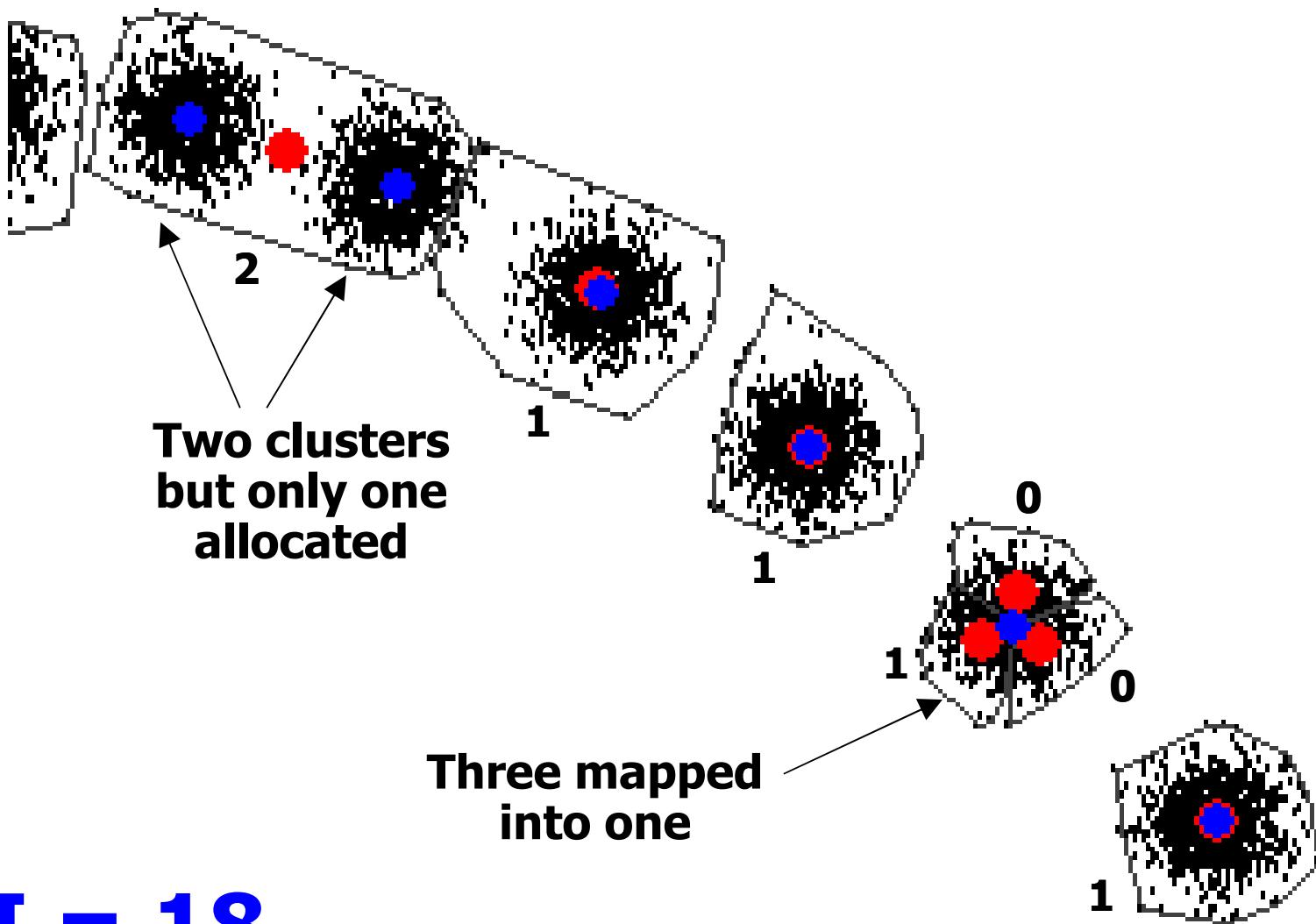
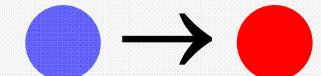


**CI = 18**

# Example

## Birch2

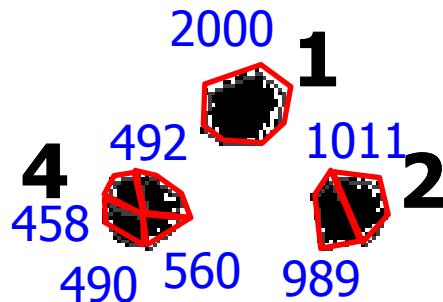
Mappings



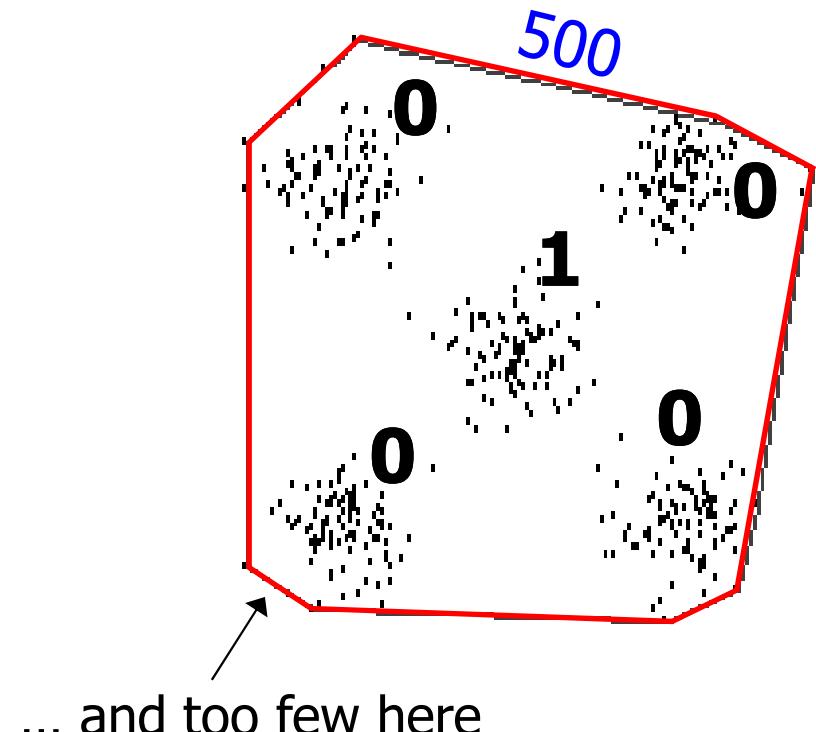
# Unbalanced example

K-means result

**KM→GT**  
**CI = 4**



K-means tend to put  
too many clusters here ...

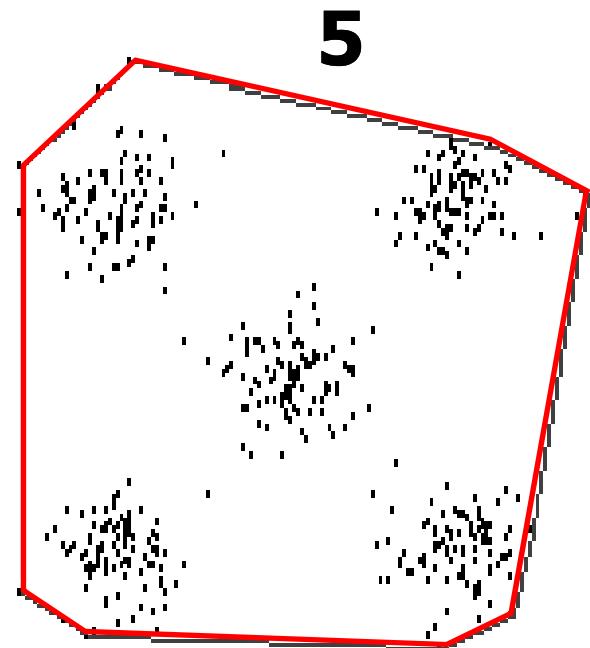
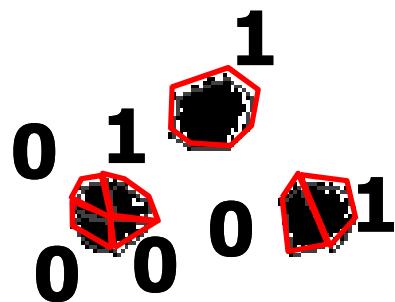


# Unbalanced example

K-means result

**GT→KM**

**CI = 4**



# **Experiments**

# Mean Squared Errors

Green = Correct clustering structure

Data set	Clustering quality (MSE)							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	179.76	176.92	173.64	179.73	168.92	164.64	164.78	161.47
<i>House</i>	6.67	6.43	6.28	6.20	6.27	5.96	5.91	5.87
<i>Miss America</i>	5.95	5.83	5.52	5.92	5.36	5.28	5.21	5.10
<i>House</i>	3.61	3.28	2.50	3.57	2.62	2.83	-	2.44
<i>Birch</i> <sub>1</sub>	5.47	5.01	4.88	5.12	4.73	4.64	-	4.64
<i>Birch</i> <sub>2</sub>	7.47	5.65	3.07	6.29	2.28	2.28	-	2.28
<i>Birch</i> <sub>3</sub>	2.51	2.07	1.92	2.07	1.96	1.86	-	1.86
<i>S</i> <sub>1</sub>	19.71	8.92	8.92	8.92	8.93	8.92	8.92	8.92
<i>S</i> <sub>2</sub>	20.58	13.28	13.28	15.87	13.44	13.28	13.28	13.28
<i>S</i> <sub>3</sub>	19.57	16.89	16.89	16.89	17.70	16.89	16.89	16.89
<i>S</i> <sub>4</sub>	17.73	15.70	15.70	15.71	17.52	15.70	15.71	15.70

Raw numbers  
don't tell much

# Adjusted Rand Index

[Hubert & Arabie, 1985]

Data set	Adjusted Rand Index (ARI)							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	0.38	0.40	0.39	0.37	0.43	0.52	0.50	1
<i>House</i>	0.40	0.40	0.44	0.47	0.43	0.53	0.53	1
<i>Miss America</i>	0.19	0.19	0.18	0.20	0.20	0.20	0.23	1
<i>House</i>	0.46	0.49	0.52	0.46	0.49	0.49	-	1
<i>Birch</i> <sub>1</sub>	0.85	0.93	0.98	0.91	0.96	1.00	-	1
<i>Birch</i> <sub>2</sub>	0.81	0.86	0.95	0.86	1	1	-	1
<i>Birch</i> <sub>3</sub>	0.74	0.82	0.87	0.82	0.86	How high is good?		1
<i>S</i> <sub>1</sub>	<b>0.83</b>	1.00	1.00	1.00	1.00	How high is good?		1.00
<i>S</i> <sub>2</sub>	<b>0.80</b>	0.99	0.99	<b>0.89</b>	0.98	0.99	0.99	0.99
<i>S</i> <sub>3</sub>	<b>0.86</b>	0.96	0.96	0.96	0.92	0.96	0.96	0.96
<i>S</i> <sub>4</sub>	<b>0.82</b>	0.93	0.93	0.94	<b>0.77</b>	0.93	0.93	0.93

# Normalized Mutual information

[Kvalseth, 1987]

Data set	Normalized Mutual Information (NMI)							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	0.77	0.78	0.78	0.77	0.80	0.83	0.82	1.00
<i>House</i>	0.80	0.80	0.81	0.82	0.81	0.83	0.84	1.00
<i>Miss America</i>	0.64	0.64	0.63	0.64	0.64	0.66	0.66	1.00
<i>House</i>	0.81	0.81	0.82	0.81	0.81	0.82	-	1.00
<i>Birch</i> <sub>1</sub>	0.95	0.97	0.99	0.96	0.98	1.00	-	1.00
<i>Birch</i> <sub>2</sub>	0.96	0.97	0.99	0.97	1.00	1.00	-	1.00
<i>Birch</i> <sub>3</sub>	0.90	0.94	0.94	0.93	0.93	0.96	-	1.00
<i>S</i> <sub>1</sub>	<b>0.93</b>	1.00	1.00	1.00	1.00	1.00	1.00	1.00
<i>S</i> <sub>2</sub>	<b>0.90</b>	0.99	0.99	<b>0.95</b>	0.99	0.93	0.99	0.99
<i>S</i> <sub>3</sub>	<b>0.92</b>	0.97	0.97	0.97	0.94	0.97	0.97	0.97
<i>S</i> <sub>4</sub>	<b>0.88</b>	0.94	0.94	0.95	<b>0.85</b>	0.94	0.94	0.94

# Normalized Van Dongen

[Kvalseth, 1987]

Data set	Normalized Van Dongen (NVD)							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	0.45	0.42	0.43	0.46	0.38	0.32	0.33	0.00
<i>House</i>	0.44	0.43	0.40	0.37	0.40	0.33	0.31	0.00
<i>Miss America</i>	0.60	0.60	0.61	0.59	0.57	0.55	0.53	0.00
<i>House</i>	0.40	0.37	0.34	0.39	0.39	0.34	-	0.00
<i>Birch</i> <sub>1</sub>	0.09	0.04	0.01	0.06	0.02	0.00	-	0.00
<i>Birch</i> <sub>2</sub>	0.12	0.08	0.03	0.09	0.00	0.00	-	0.00
<i>Birch</i> <sub>3</sub>	0.19	0.12	0.10	0.13	0.13	<b>Lower is better</b>		0.00
<i>S</i> <sub>1</sub>	<b>0.09</b>	0.00	0.00	0.00	0.00	<b>0.00 better</b>		0.00
<i>S</i> <sub>2</sub>	<b>0.11</b>	0.00	0.00	<b>0.06</b>	0.01	0.04	0.00	0.00
<i>S</i> <sub>3</sub>	<b>0.08</b>	0.02	0.02	0.02	0.05	0.00	0.00	0.02
<i>S</i> <sub>4</sub>	<b>0.11</b>	0.04	0.04	0.03	<b>0.13</b>	0.04	0.04	0.04

# Centroid Similarity Index

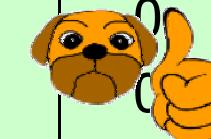
[Fränti, Rezaei, Zhao, 2014]

Data set	Centroid Similarity Index (CSI)							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	0.47	0.51	0.49	0.45	0.57	0.62	0.63	1.00
<i>House</i>	0.49	0.50	0.54	0.57	0.55	0.63	0.66	1.00
<i>Miss America</i>	0.32	0.32	0.32	0.33	0.38	0.40	0.42	1.00
<i>House</i>	0.54	0.57	0.63	0.54	0.57	0.62	---	1.00
<i>Birch</i> <sub>1</sub>	0.87	0.94	0.98	0.93	0.99	1.00	---	1.00
<i>Birch</i> <sub>2</sub>	0.76	0.84	0.94	0.83	1.00	Ok but lacks threshold		
<i>Birch</i> <sub>3</sub>	0.71	0.82	0.87	0.81	0.86	0.00		
<i>S</i> <sub>1</sub>	<b>0.83</b>	1.00	1.00	1.00	1.00	1.00	1.00	1.00
<i>S</i> <sub>2</sub>	<b>0.82</b>	1.00	1.00	<b>0.91</b>	1.00	1.00	1.00	1.00
<i>S</i> <sub>3</sub>	<b>0.89</b>	0.99	0.99	0.99	0.98	0.99	0.99	0.99
<i>S</i> <sub>4</sub>	<b>0.87</b>	0.98	0.98	0.99	<b>0.85</b>	0.98	0.98	0.98

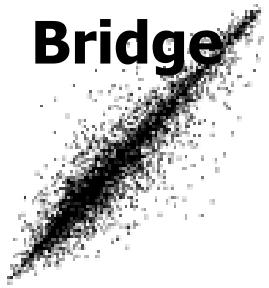
# Centroid Index

[Fränti, Rezaei, Zhao, 2014]

Data set	C-Index ( $CI_2$ )							
	KM	RKM	KM++	XM	AC	RS	GKM	GA
<i>Bridge</i>	74	63	58	81	33	33	35	0
<i>House</i>	56	45	40	37	31	22	20	0
<i>Miss America</i>	88	91	67	88	38	43	36	0
<i>House</i>	43	39	22	47	26	23	---	0
<i>Birch</i> <sub>1</sub>	7	3	1	4	0	0	---	0
<i>Birch</i> <sub>2</sub>	18	11	4	12	0	0	---	0
<i>Birch</i> <sub>3</sub>	23	11	7	10	7	2	---	0
$S_1$	<b>2</b>	0	0	0	0	0	0	0
$S_2$	<b>2</b>	0	0	<b>1</b>	0	0	0	0
$S_3$	<b>1</b>	0	0	0	0	0	0	0
$S_4$	<b>1</b>	0	0	0	<b>1</b>	0	0	0



**Going deeper...**



# Accurate clustering

GAIS-2002 similar to GAIS-2012 ?

	Method	MSE
GKM	Global K-means	164.78
RS	Random swap (5k)	164.64
GA	Genetic algorithm	161.47
RS <sub>8M</sub>	Random swap (8M)	161.02
GAIS-2002	GAIS	160.72
+ RS <sub>1M</sub>	GAIS + RS (1M)	160.49
+ RS <sub>8M</sub>	GAIS + RS (8M)	160.43
GAIS-2012	GAIS	160.68
+ RS <sub>1M</sub>	GAIS + RS (1M)	160.45
+ RS <sub>8M</sub>	GAIS + RS (8M)	160.39
+ PRS	GAIS + PRS	160.33
+ RS <sub>8M</sub> +	GAIS + RS (8M) + PRS	160.28

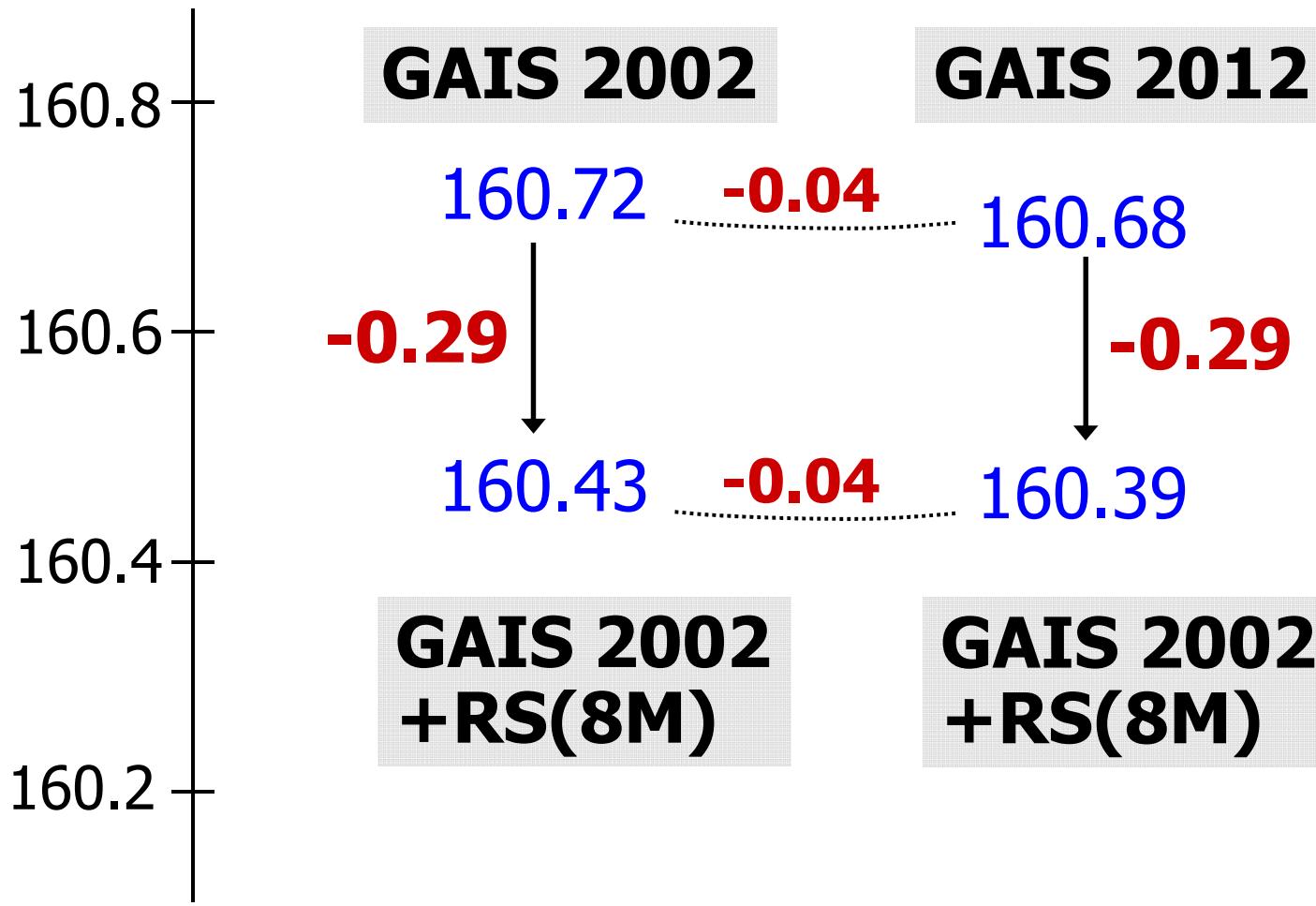
-0.04

-0.29

-0.29

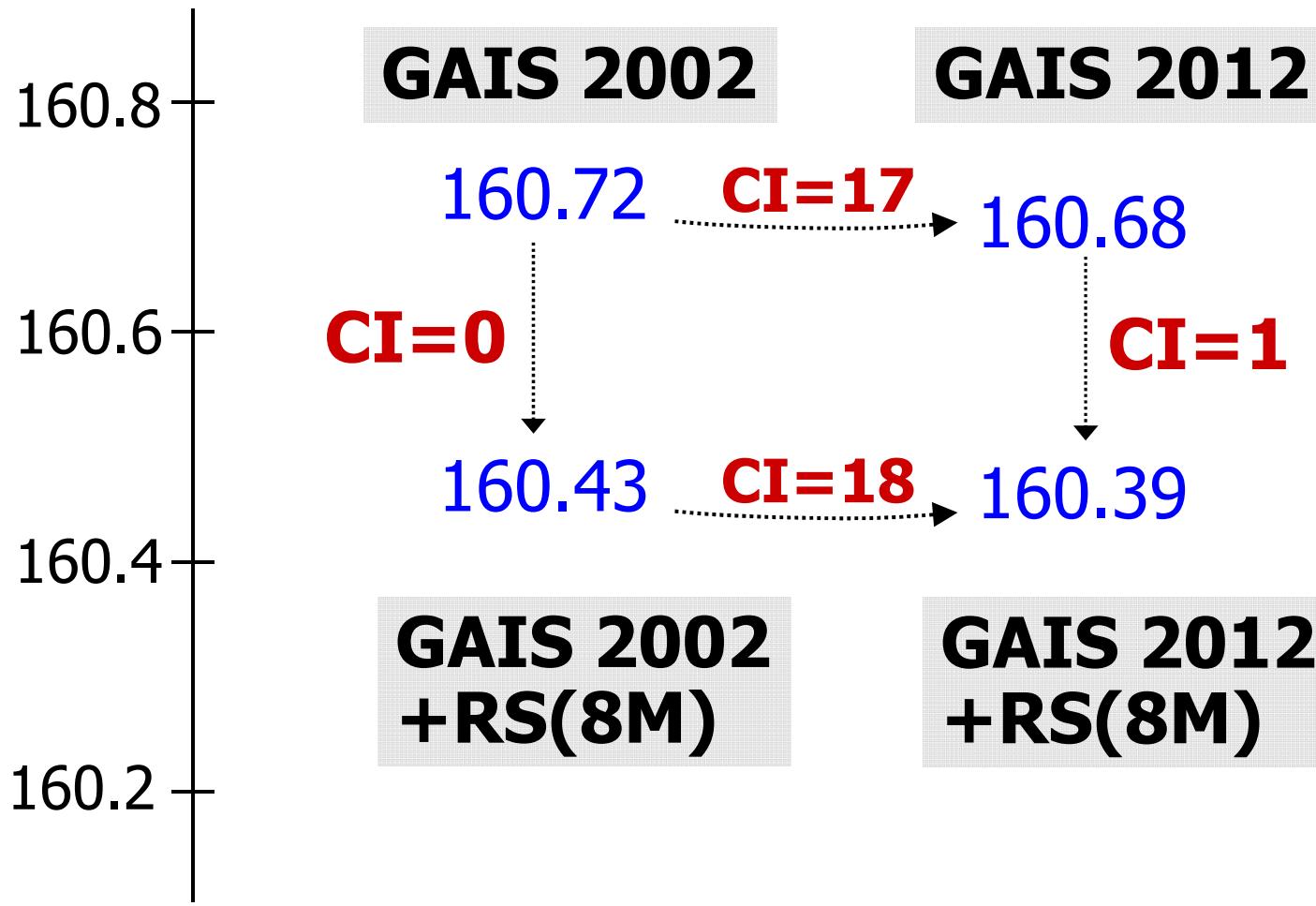
# GAIS'02 and GAIS'12 the same?

Virtually the same MSE-values



# GAIS'02 and GAIS'12 the same?

But different structure!



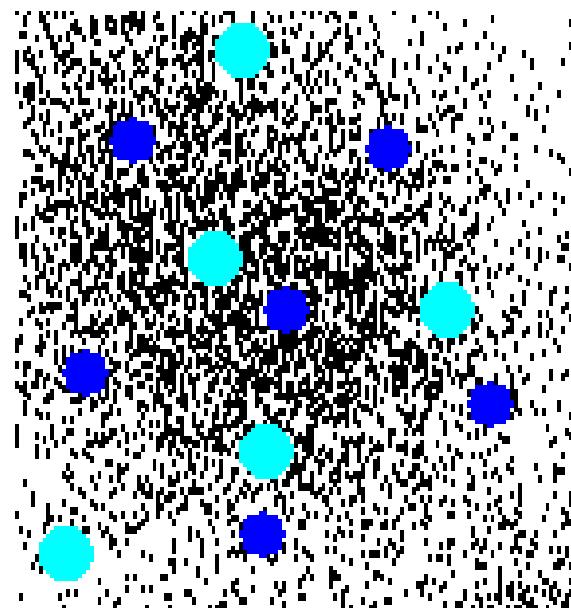
# Seemingly the same solutions

Same structure  
“same family”

<b>Main algorithm:</b> + Tuning 1 + Tuning 2	RS <sub>8M</sub>	GAIS 2002				GAIS 2012			
		RS <sub>1M</sub>	RS <sub>8M</sub>	RS <sub>1M</sub>	RS <sub>8M</sub>	RS <sub>1M</sub>	RS <sub>8M</sub>	RS <sub>8M</sub>	RS <sub>8M</sub>
RS <sub>8M</sub>	---	19	19	19	23	24	24	23	22
<b>GAIS (2002)</b>	23	---	0	0	14	15	15	14	16
+ RS <sub>1M</sub>	23	0	---	0	14	15	15	14	13
+ RS <sub>8M</sub>	23	0	0	---	14	15	15	14	13
<b>GAIS (2012)</b>	25	17	18	18	---	1	1	1	1
+ RS <sub>1M</sub>	25	17	18	18	1	---	0	0	1
+ RS <sub>8M</sub>	25	17	18	18	1	0	---	0	1
+ PRS	25	17	18	18	1	0	0	---	1
+ RS <sub>8M</sub> + PRS	24	17	18	18	1	1	1	1	---

# But why?

- Real cluster structure missing
- Clusters allocated like well optimized “grid”
- Several grids results different allocation
- Overall clustering quality can still be the same

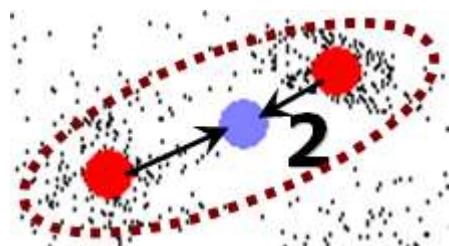


# **Generalization**

# Three alternatives

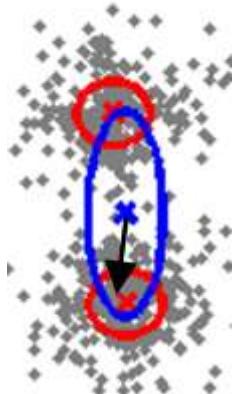
## 1. Prototype similarity

Prototype must exist

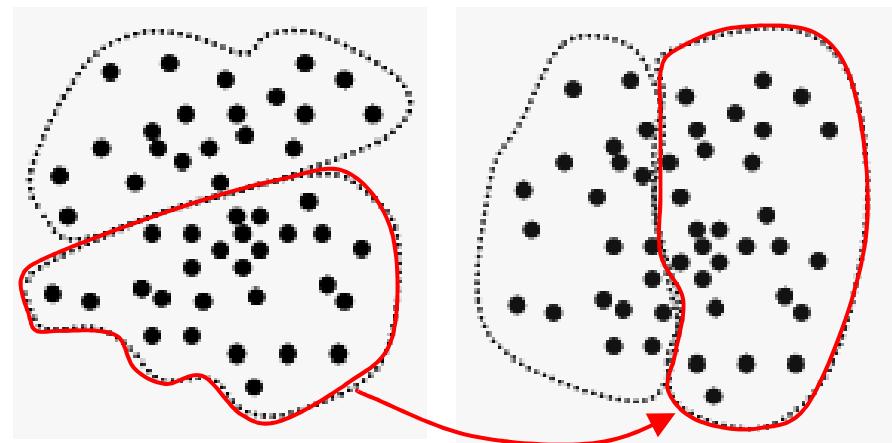


## 2. Model similarity

Derived from model



## 3. Partition similarity

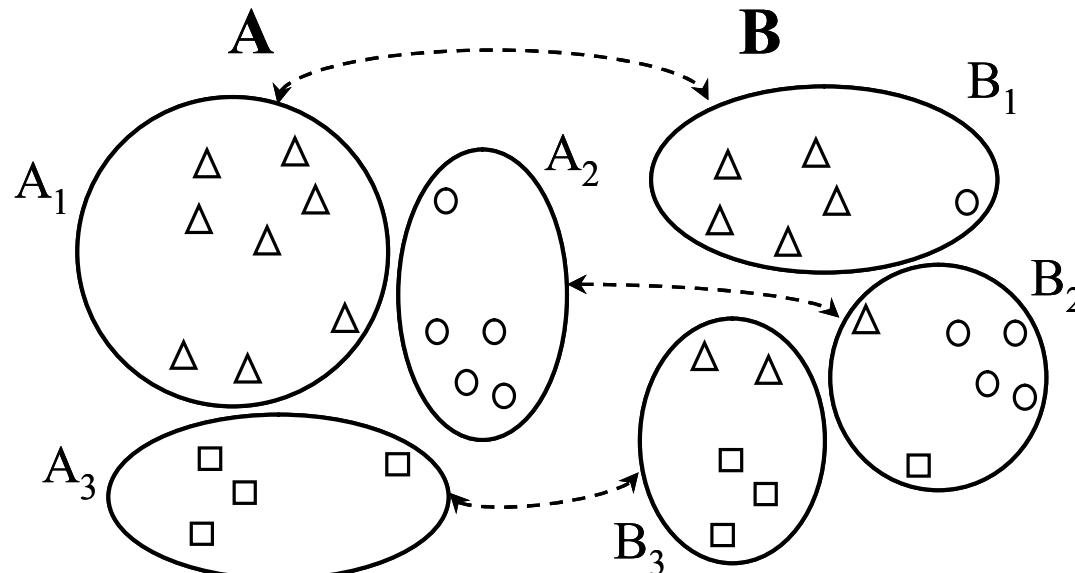


$$J = \frac{|P_i \cap G_j|}{|P_i \cup G_j|}$$

# Partition similarity

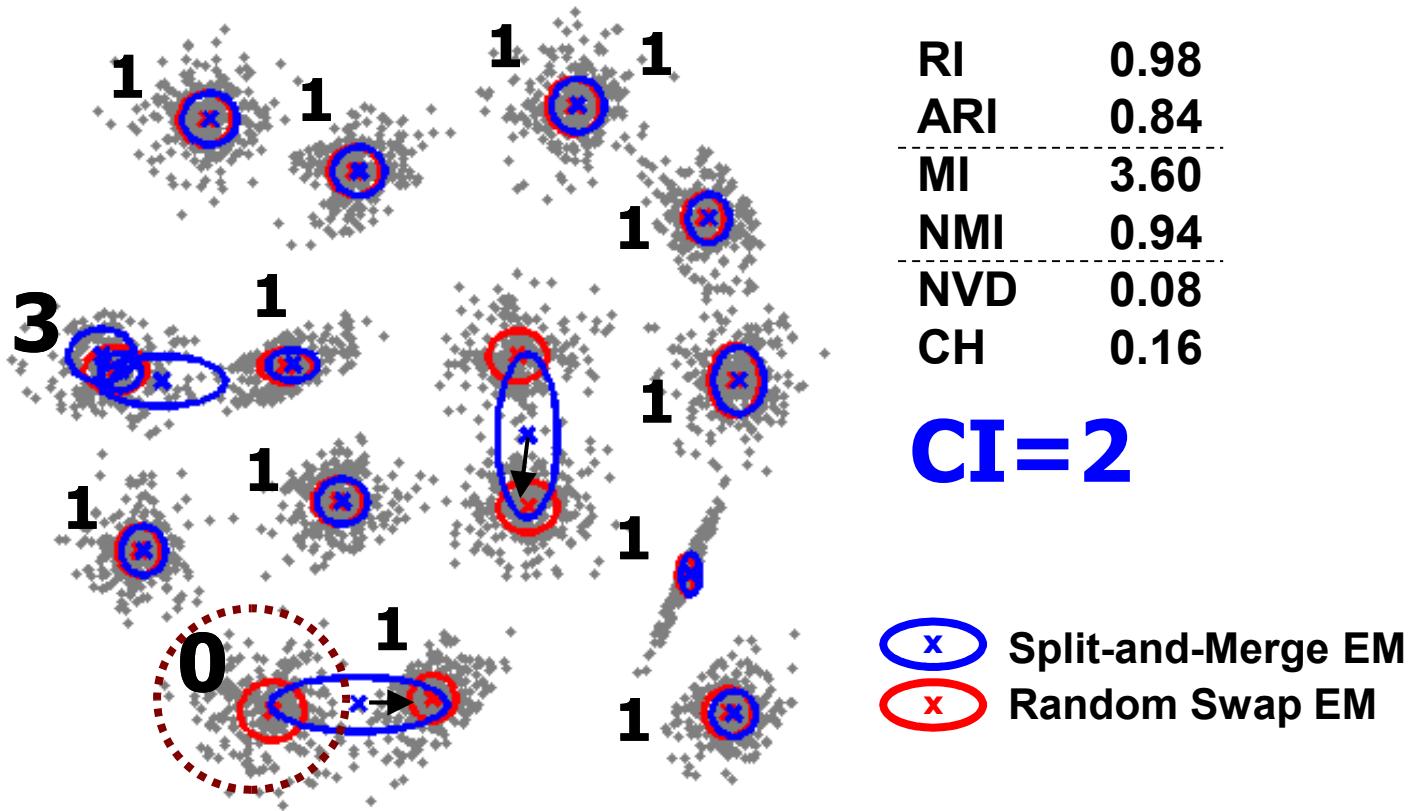
- Cluster similarity using Jaccard
- Calculated from contingency table

	$B_1$	$B_2$	$B_3$	$\Sigma$
$A_1$	5	1	2	8
$A_2$	1	4	0	5
$A_3$	0	1	3	4
$\Sigma$	6	6	5	17



# Gaussian mixture model

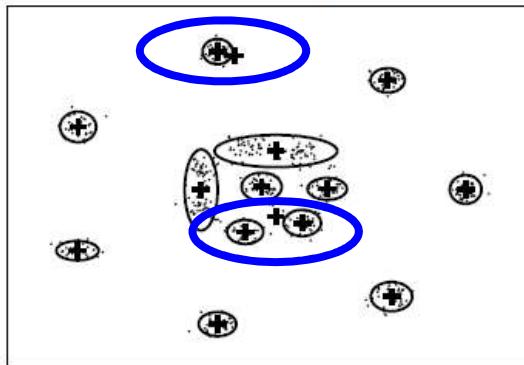
$$S_{BC} = \frac{1}{8} \left( c[A_i] - c[B_j] \right)^T \Sigma^{-1} \left( c[A_i] - c[B_j] \right) + \frac{1}{2} \ln \left( \frac{|\Sigma|}{\sqrt{|\Sigma_1||\Sigma_2|}} \right)$$



# Gaussian mixture model

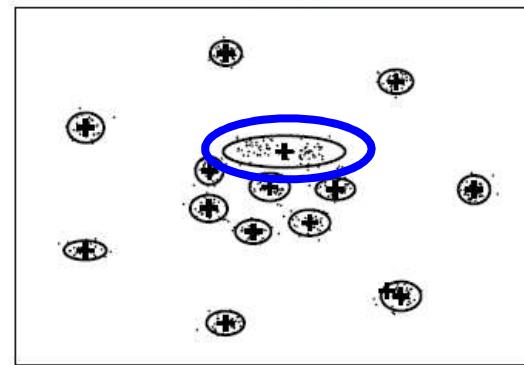
**CI=2**

REM:-6.42



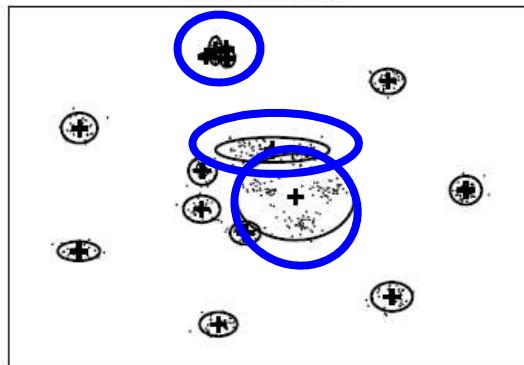
**CI=1**

REM:-6.37



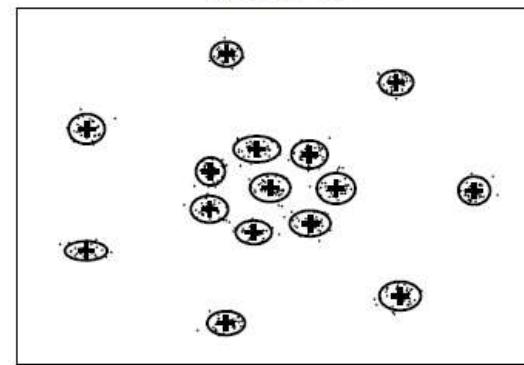
**CI=3**

SMEM:-6.53



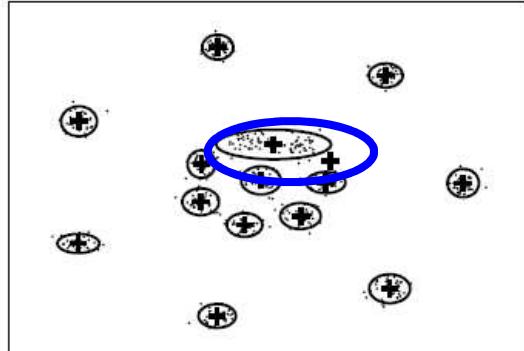
**CI=0**

SMEM:-6.33

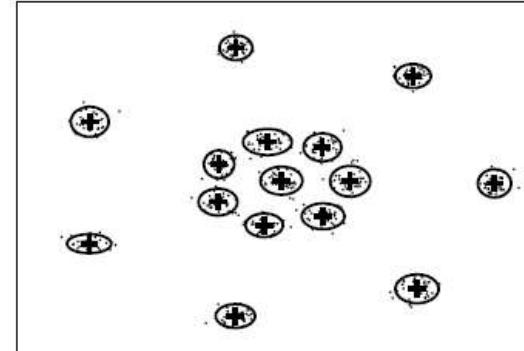


**CI=1**

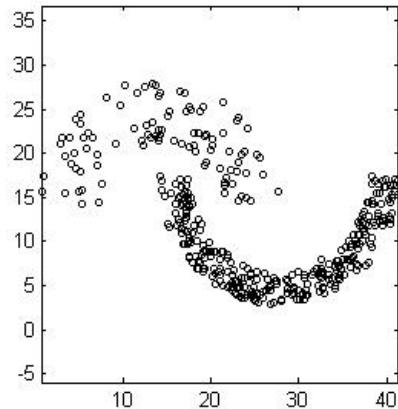
RSEM:-6.37



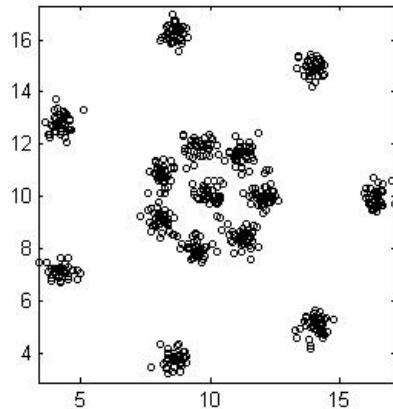
**CI=0**



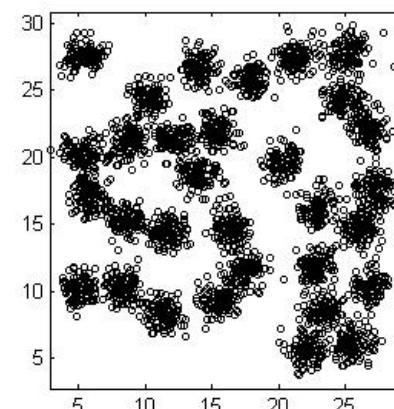
# Arbitrary-shape data



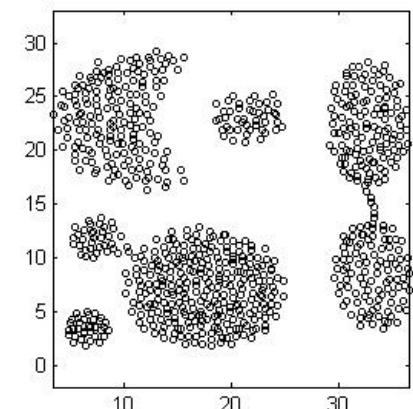
A.K. Jain's Toy problem



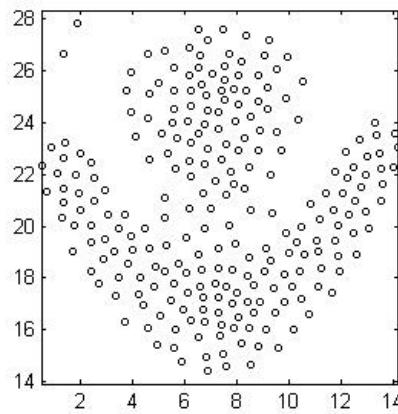
R15



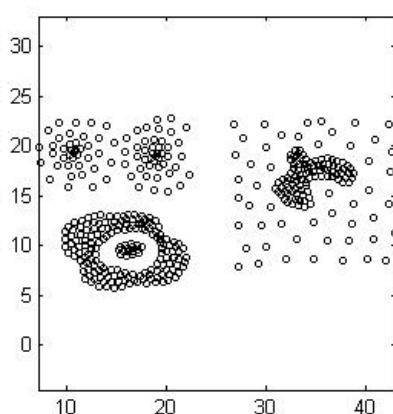
D31



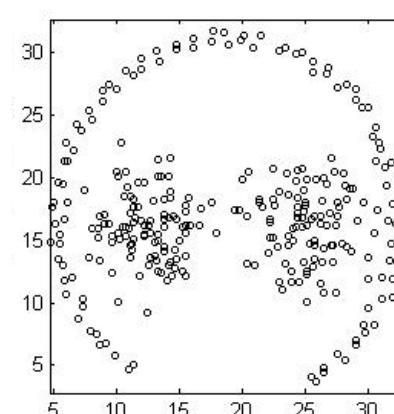
Aggregation



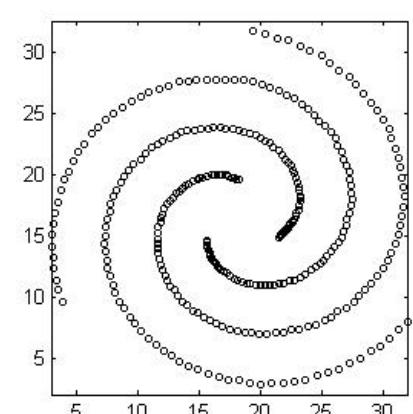
Flame



Zahn's Compound



Path-based1



path-based2: spiral

**KM $\rightarrow$ GT**

**2**

**1**

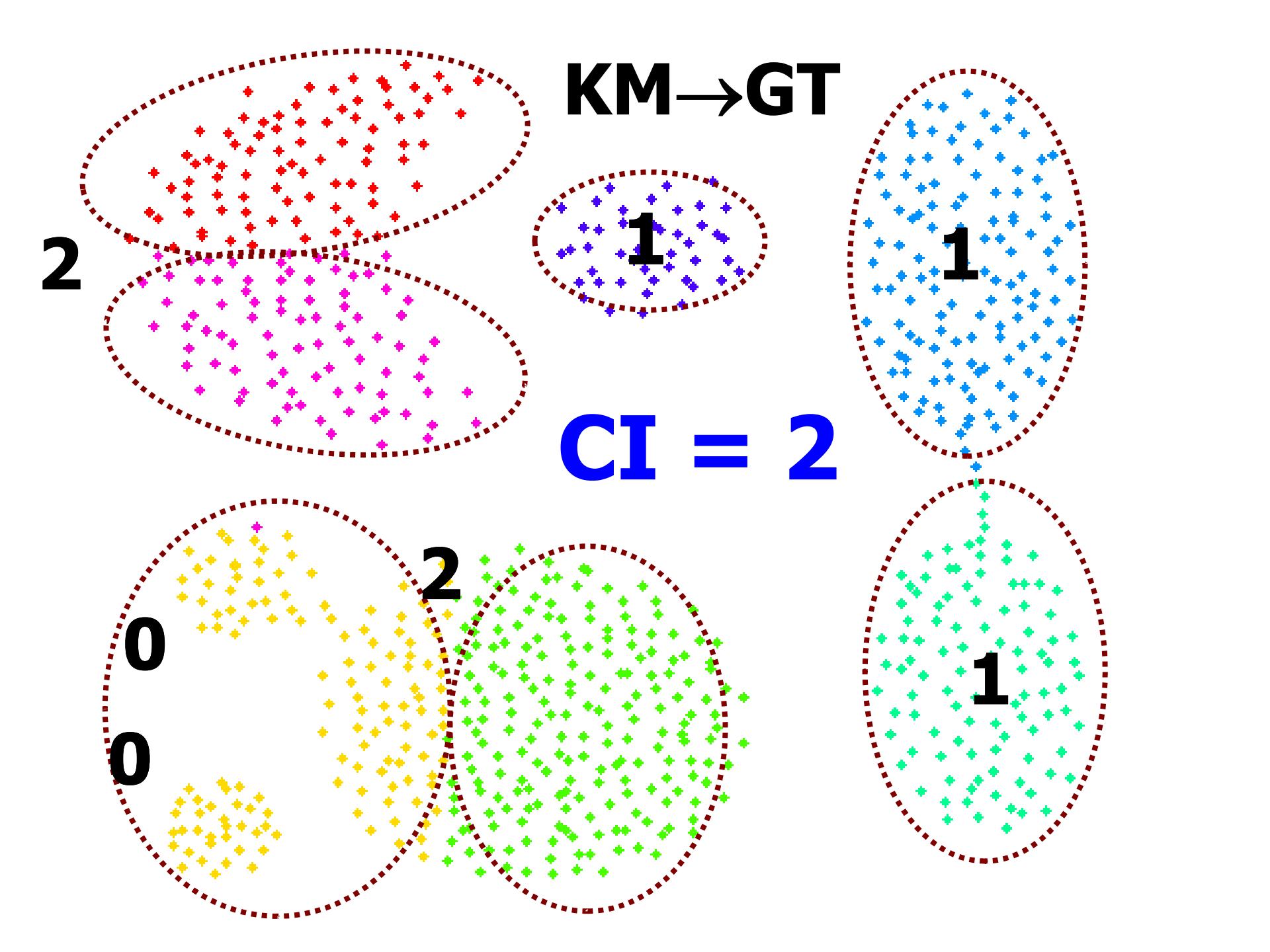
**1**

**CI = 2**

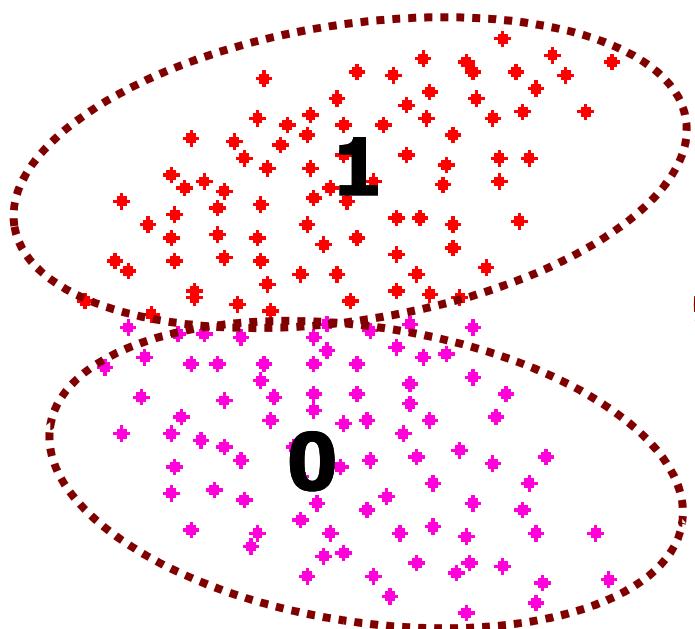
**0  
0**

**2**

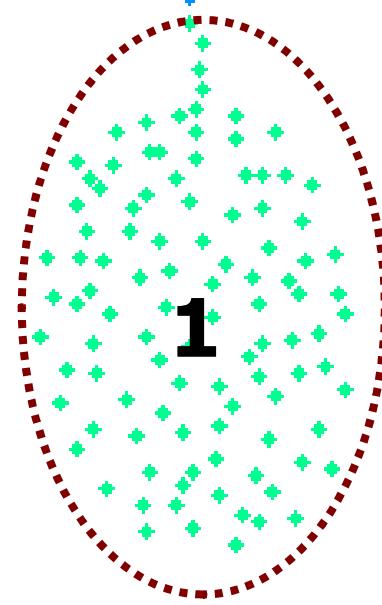
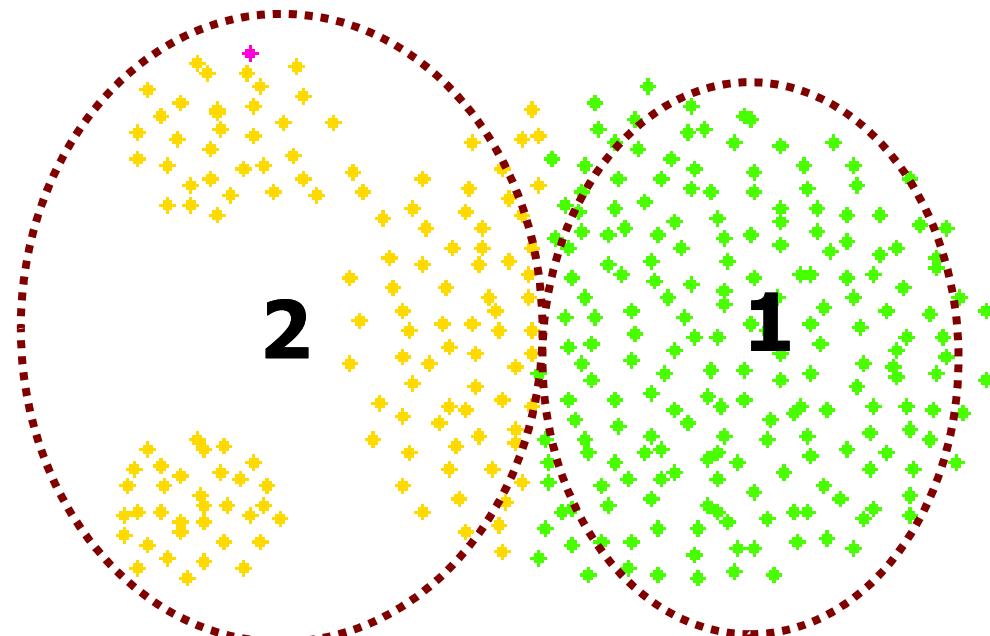
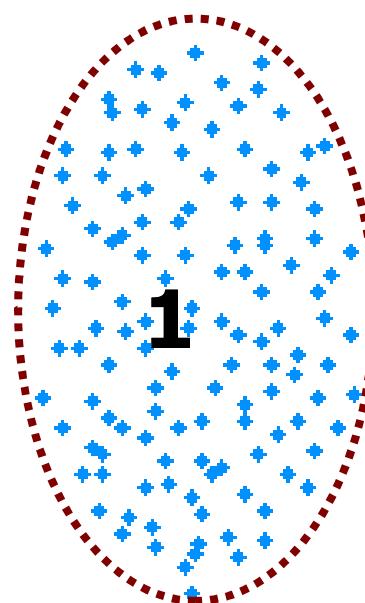
**1**



**GT → KM**



**CI = 1**



**SL $\rightarrow$ GT**

**0**

**1**

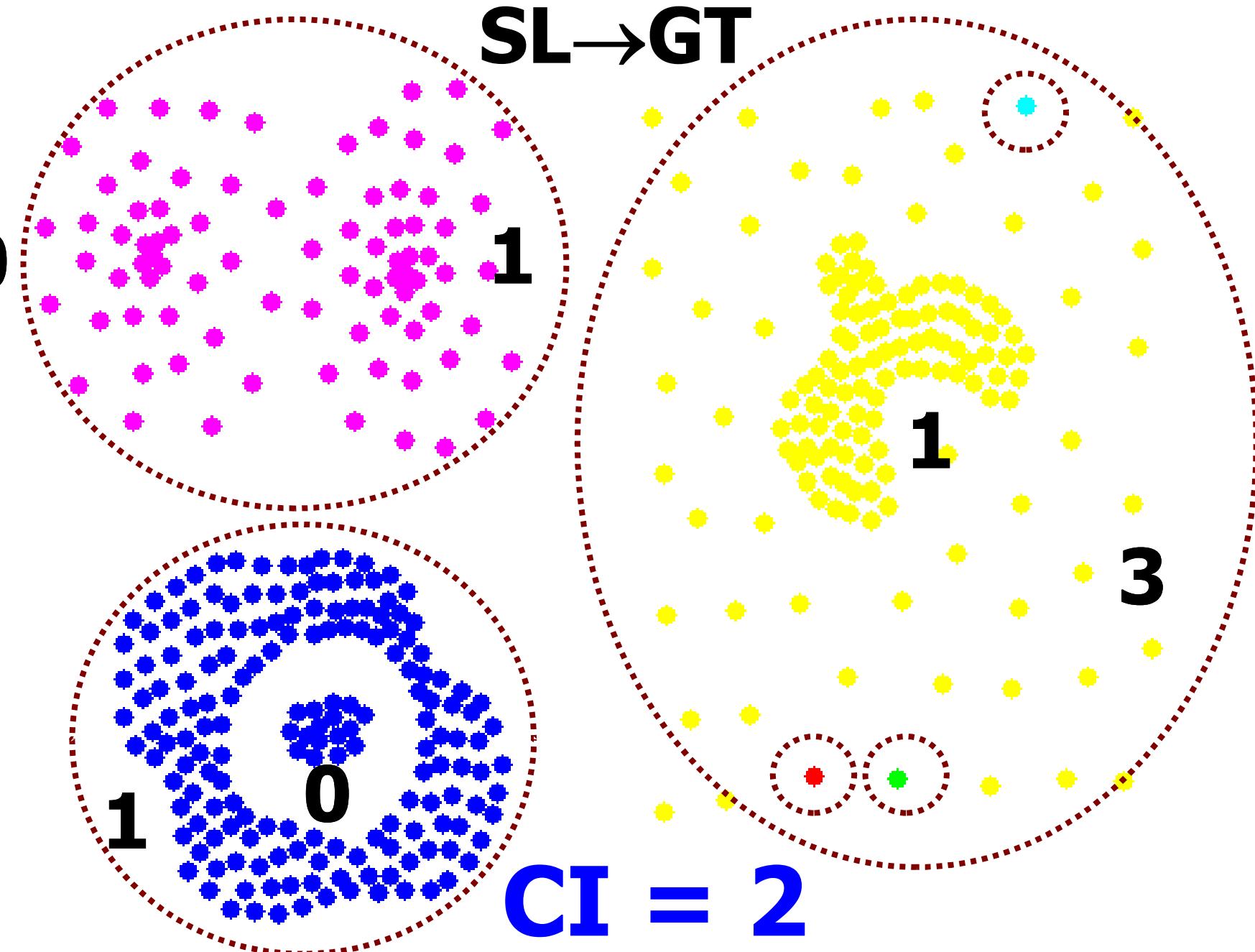
**1**

**0**

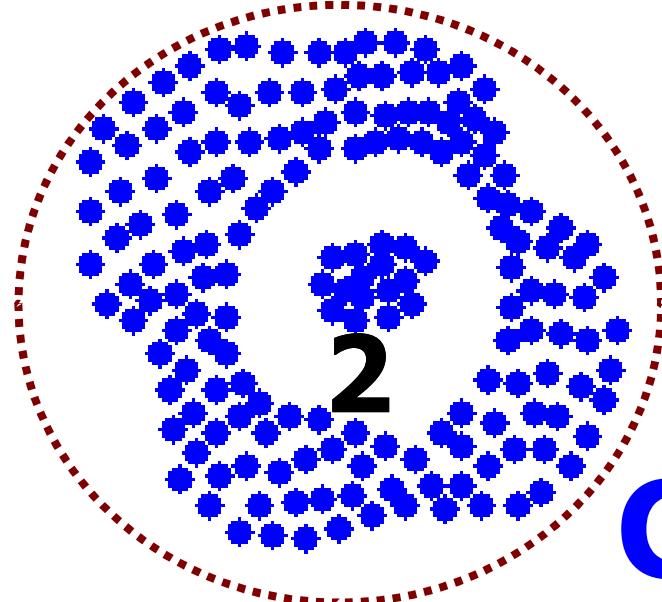
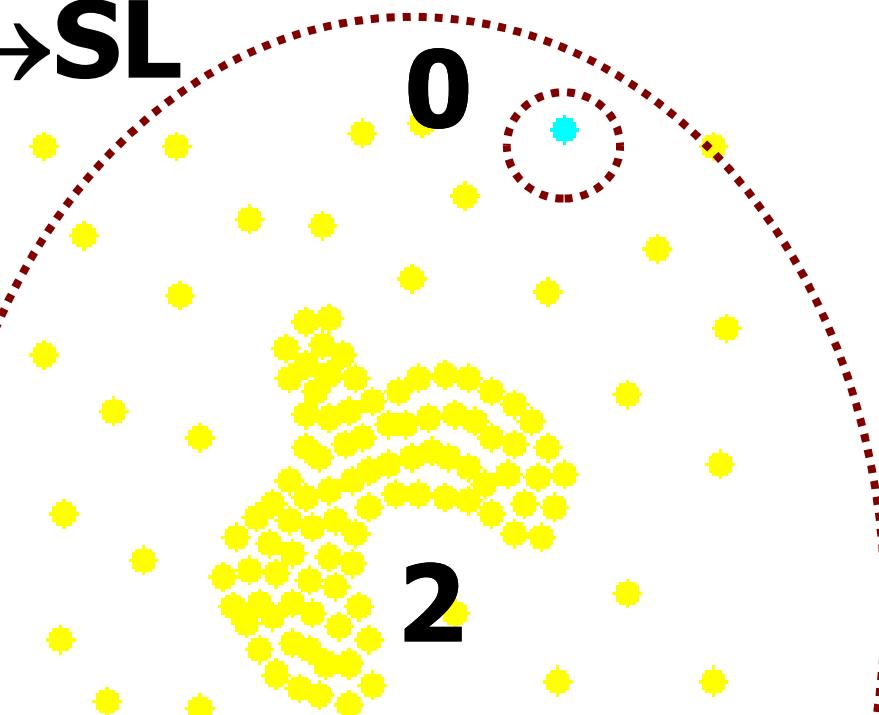
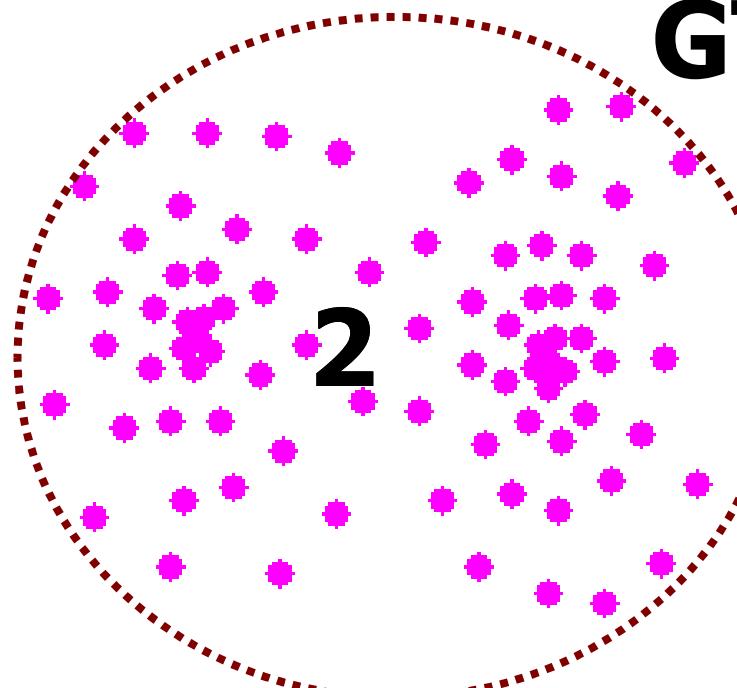
**CI = 2**

**1**

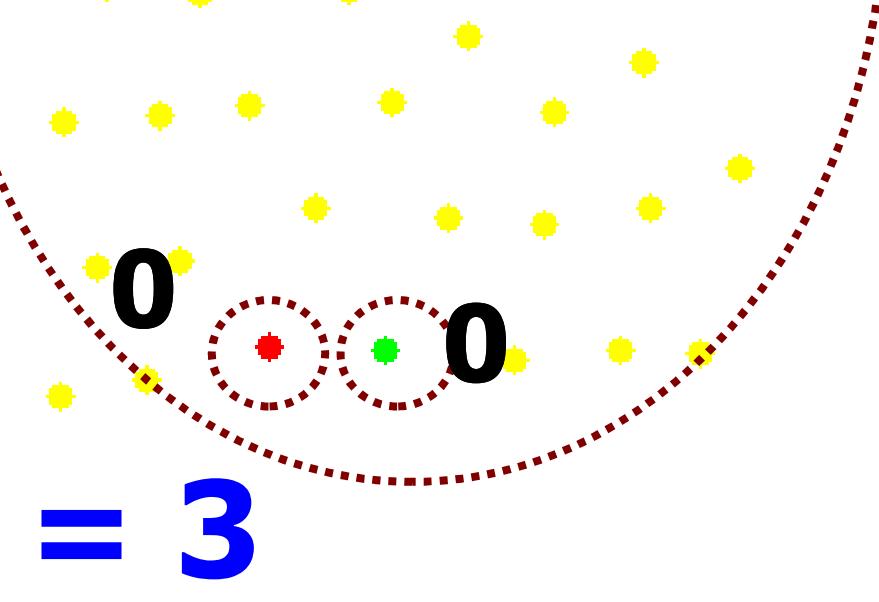
**3**



**GT→SL**



**CI = 3**



# Summary of experiments

## prototype similarity

	RI	ARI	MI	NMI	NVD	CH	CSI	CI
<b>Birch2</b>								
<b>KM</b>	1.00	0.81	6.26	0.96	0.12	0.24	0.88	18
<b>KM++</b>	1.00	0.95	6.54	0.99	0.03	0.06	0.97	4
<b>RS</b>	1.00	1.00	6.64	1.00	0.00	0.00	1.00	0
<b>GA</b>	1.00	1.00	6.64	1.00	0.00	0.00	1.00	0
<b>S1</b>								
<b>KM</b>	0.98	0.82	3.57	0.93	0.09	0.17	0.83	2
<b>KM++</b>	1.00	1.00	3.90	0.98	0.00	0.00	1.00	0
<b>RS</b>	1.00	1.00	3.90	0.98	0.00	0.00	1.00	0
<b>GA</b>	1.00	1.00	3.90	0.98	0.00	0.00	1.00	0
<b>S2</b>								
<b>KM</b>	0.97	0.80	3.46	0.90	0.11	0.18	0.82	2
<b>KM++</b>	1.00	0.99	3.87	0.99	0.00	0.00	1.00	0
<b>RS</b>	1.00	0.99	3.87	0.99	0.00	0.00	1.00	0
<b>GA</b>	1.00	0.99	3.87	0.99	0.00	0.00	1.00	0

# Summary of experiments

## partition similarity

	RI	ARI	MI	NMI	NVD	CH	CSI	CI
<b>Unbalanced</b>								
<b>KM</b>	0.92	0.79	1.85	0.81	0.14	0.29	0.86	4
<b>KM++</b>	1.00	1.00	2.03	1.00	0.00	0.00	1.00	0
<b>RS</b>	1.00	1.00	2.03	1.00	0.00	0.00	1.00	0
<b>GA</b>	1.00	1.00	2.03	1.00	0.00	0.00	1.00	0
<b>SL</b>	1.00	0.99	1.91	0.97	0.02	0.05	0.98	3
<b>DBSCAN</b>	1.00	1.00	2.02	0.99	0.00	0.00	1.00	0
<b>SAM</b>	0.93	0.81	1.85	0.82	0.12	0.25	0.88	4
<b>Aggregate</b>								
<b>KM</b>	0.91	0.71	2.16	0.84	0.14	0.24	0.86	2
<b>SL</b>	0.93	0.80	1.96	0.88	0.09	0.18	0.91	2
<b>DBSCAN</b>	0.99	0.98	2.41	0.98	0.01	0.01	0.99	0
<b>SAM</b>	1.00	1.00	2.45	1.00	0.00	0.00	1.00	0
<b>Compound</b>								
<b>KM</b>	0.84	0.54	1.71	0.72	0.25	0.34	0.75	2
<b>SL</b>	0.89	0.74	1.54	0.80	0.13	0.26	0.87	3
<b>DBSCAN</b>	0.95	0.88	1.90	0.87	0.10	0.12	0.90	2
<b>SAM</b>	0.83	0.53	1.78	0.76	0.19	0.34	0.81	2

# Conclusions

- Simple is cluster level measure
- Generalized to GMM and arbitrary-shaped data
- Value has clear interpretation:

