Semi-supervised clustering: a review

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Semi-supervised clustering : a Machine Learning technique



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Semi-supervised clustering

Determines groups of objects using a data set and some constraints



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Outline : semi-supervised clustering





Outline



2 Methodology to add constraints

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Constraint types



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Instance level constraints



Cluster level constraints



Ratio objects per class / balanced clusters : $\frac{n_k}{n} = \tau_k \qquad / \qquad \frac{n_k}{n} = \frac{1}{c}$



 $\begin{array}{l} {\sf Min} \ / \ {\sf Max} \\ {\sf distance \ inter-class} \ (\delta) \\ {\sf distance \ intra-class} \ (\epsilon) \end{array}$

Attribute level constraints

Caution [2]

Protected attributes can not be explicitly used in decision making !



Fairness

Percentage *p*%-rule on the modality of an attribute

Attribute level constraints

Caution [2]

Protected attributes can not be explicitly used in decision making !



Fairness

Percentage *p*%-rule on the modality of an attribute

Model level constraints



Negative information

A specific partition should not be the final solution

Remark

Close to singular alternative clustering problem

Semi-supervised clustering history



Outline





Adding constraints in unsupervised learning



Preprocessing

- learning step
 - centroids, distance,...
- constraints influence
 - feasability solution w.r.t constraints ?
 - augmenting constraints
 - reducing constraints to informative ones

Adding constraints in unsupervised learning



Semi-supervised clustering

- learning step
 - partition, distance
- constraints
 - hard / soft respect of the constraints
 - noisy constraints ?

Adding constraints in unsupervised learning



Active learning scheme

- Originally from semi-supervised learning
- Ask expert few selected constraints
 - exploration / consolidation

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Conclusion

Semi-supervised clustering

- ✓ adding constraints improves the partition accuracy
- ✓ several methods have been already implemented

Perspectives

- Real application
- Use of soft constraints

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Thank you