



AIDEme: An active learning based system for interactive exploration of large datasets

Enhui Huang, Luciano Di Palma, Laurent Cetinsay, Yanlei Diao, Anna Liu

► To cite this version:

Enhui Huang, Luciano Di Palma, Laurent Cetinsay, Yanlei Diao, Anna Liu. AIDEme: An active learning based system for interactive exploration of large datasets. 2019. hal-02430750

HAL Id: hal-02430750

<https://inria.hal.science/hal-02430750>

Submitted on 7 Jan 2020

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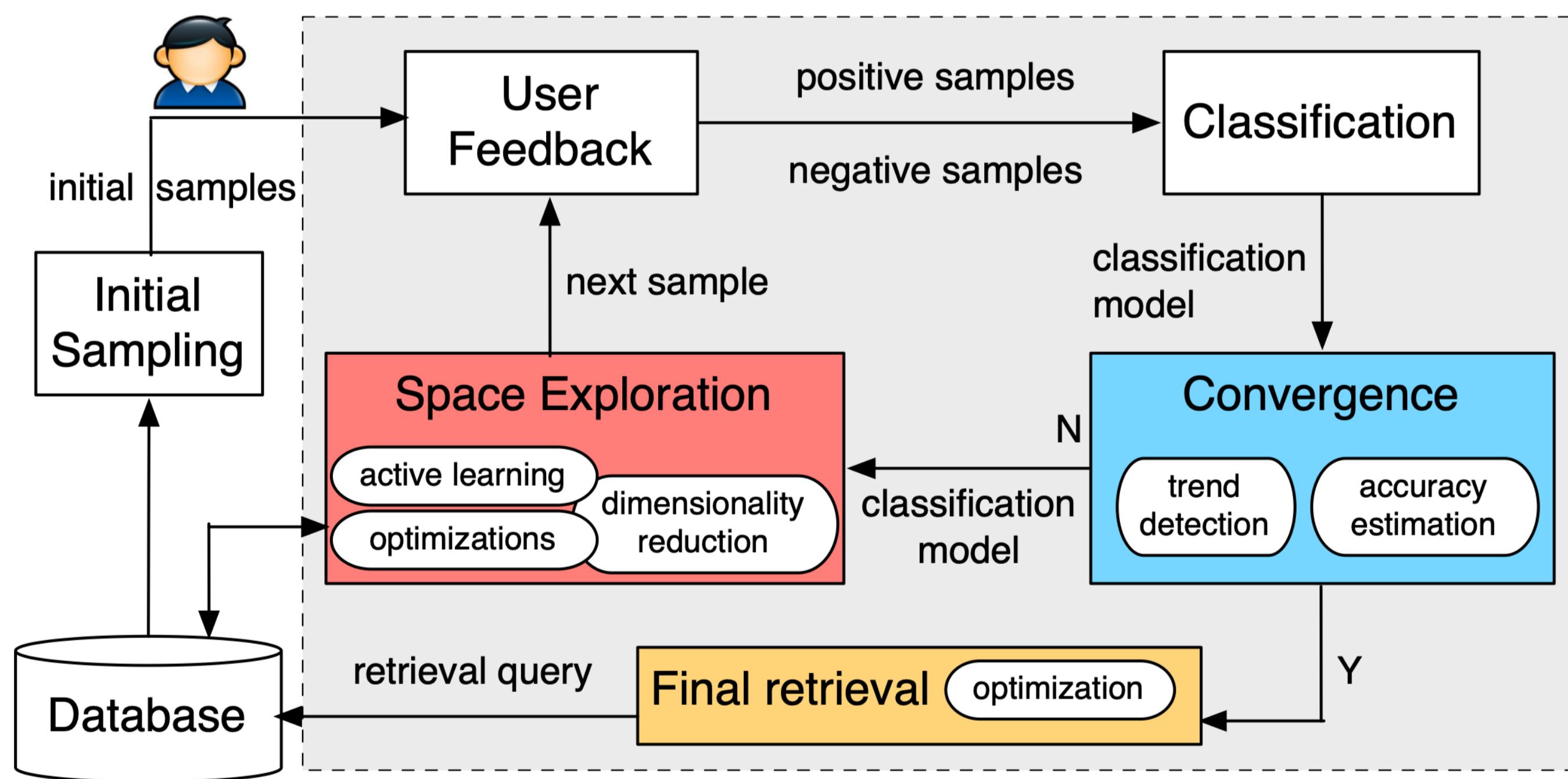
AIDEme is a scalable interactive data exploration system for efficiently learning a user interest pattern over a large dataset

Motivation



- An increasing gap between fast growth of data and limited human ability to comprehend data.
- A growing demand of data analytics tools that can bridge this gap and help the user retrieve high-value content from data more effectively.

System Overview



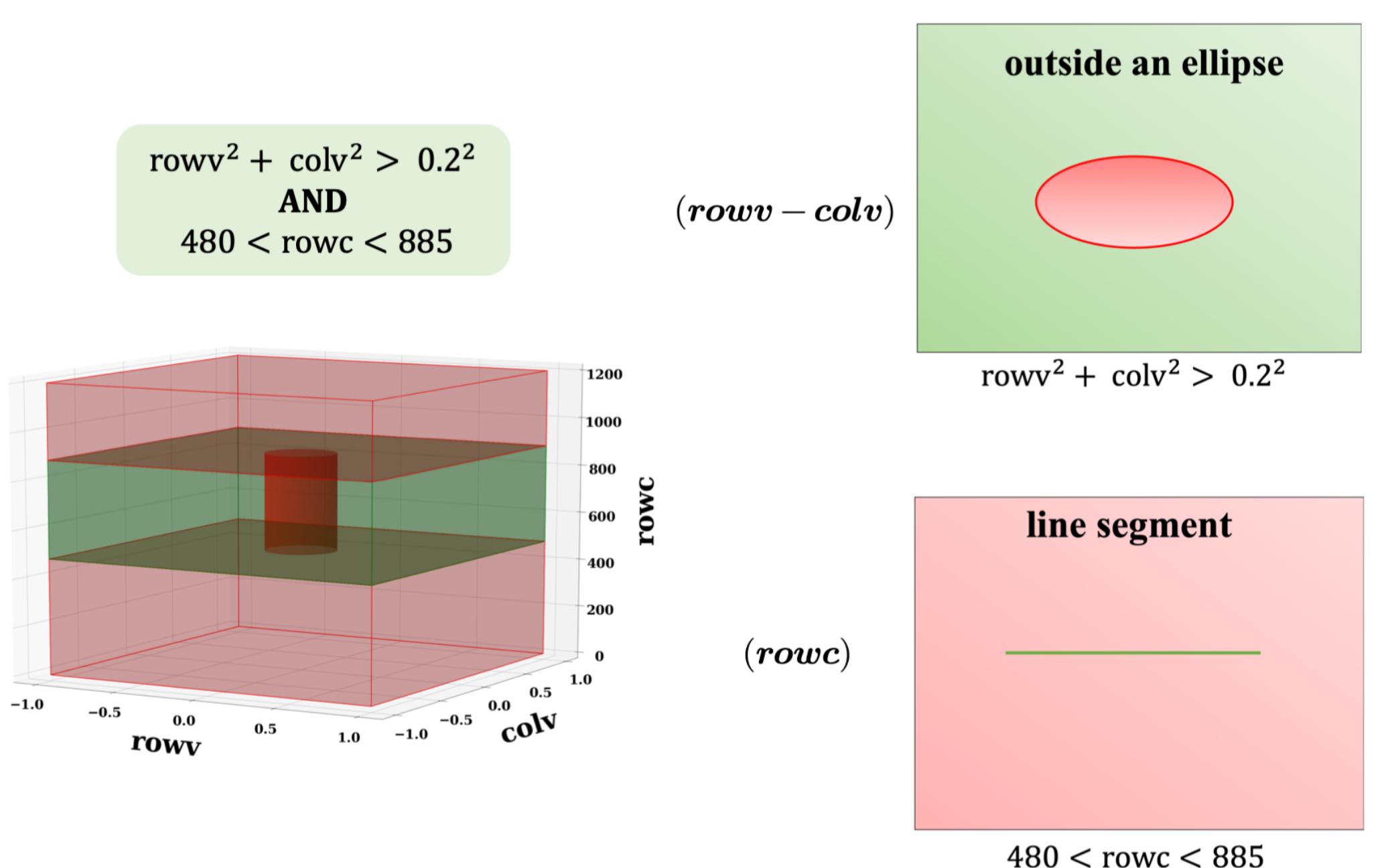
- Consider the data content as a set of records, and the user is interested in some of them but not all.
- In each iteration,
 - the user labels a record as "interesting" or "not interesting",
 - a classification model is built,
 - active learning techniques are employed to select a new record from the unlabeled data source.
- Construct an increasingly-more-accurate model of the user interest.
- Upon convergence, the model is run through the entire data source to retrieve all relevant records.

Key Techniques

Challenge: Slow Convergence

Novel techniques in AIDEme:

1. Factorization



– Factorized Version Space

Version Space V (size = 16)			Version Space V (size = 8)		
Color (B/R)	Size (S/L)	Label	Color (B/R)	Size (S/L)	Label
B(lack)	L(arge)	{-, +}	B(lack)	L(arge)	{-}
B(lack)	S(mall)	{-, +}	B(lack)	S(mall)	{-, +}
R(ed)	L(arge)	{-, +}	R(ed)	L(arge)	{-, +}
R(ed)	S(mall)	{-, +}	R(ed)	S(mall)	{-, +}

A labeled example → (BL, '-')

(a) Without factorization

Version Space V (size = 16)		Version Space V (size = 4)	
Color (B/R)	Label	Color (B/R)	Label
B(lack)	{-, +}	B(lack)	{-}
R(ed)	{-, +}	R(ed)	{-, +}

A labeled example → ((B, '-'), (L, '+'))

(b) With factorization

2. Formal results on convergence

- Theoretical results on the convergence of our proposed techniques.
- Detect convergence and terminate the exploration process.

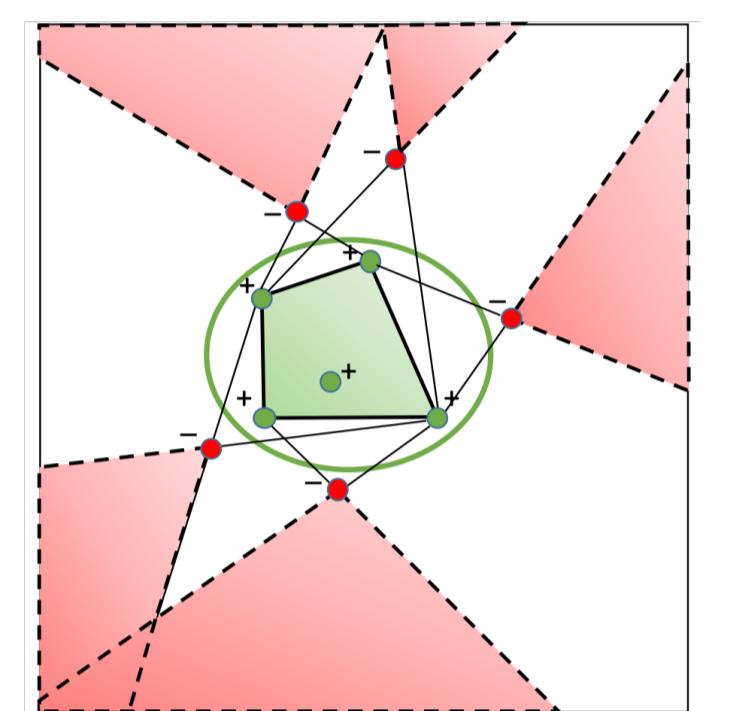
3. Scaling to large datasets

- Subsampling procedures
- Provide provable results that guarantee the performance of the model learned from the sample over the entire data source

4. Optimization using Class Distribution

- Subspatial convex property:** the user interest pattern projected onto a subspace often entails a convex object.
- When the subspatial convex property holds, we introduce a Dual-Space Model (DSM).

$$\text{DSM} \xleftarrow[\text{sample next}]{\text{predict}} \text{Classifier} + \text{Polytope model.}$$

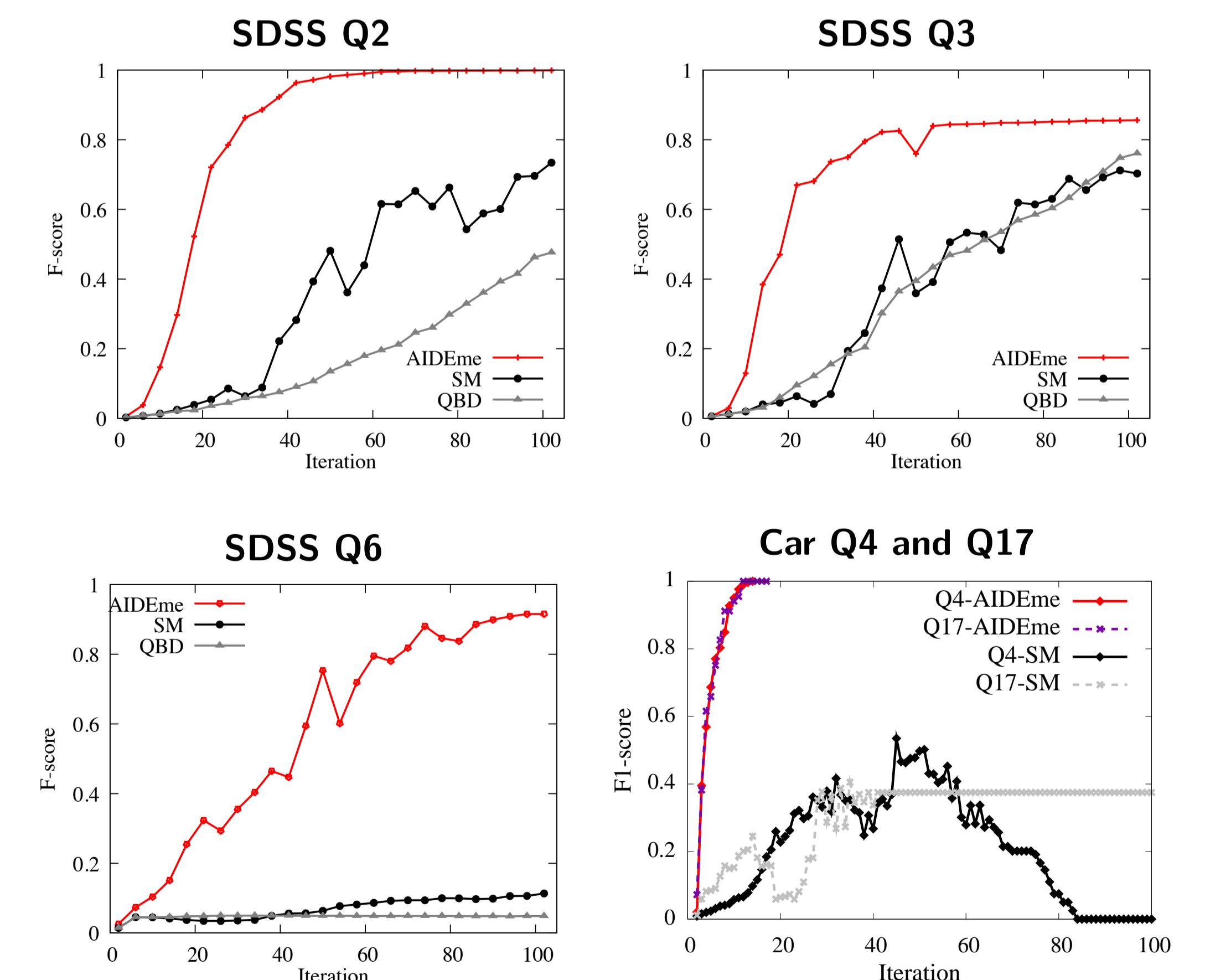


Demonstration

• Demonstration

Setup ⇒ Iterative Exploration ⇒ Final retrieval ⇒ Comparison

• Comparison



References

- Huang, E., Peng, L., Di Palma, L., Abdelkafi, A., Liu, A. & Diao, Y. Optimization for active learning-based interactive database exploration. *Proceedings of the VLDB Endowment (PVLDB)*, **12**(1), 71-84, September 2018.
- Di Palma, L., Diao, Y. & Liu, A. A Factorized Version Space Algorithm for "Human-In-the-Loop" Data Exploration. *IEEE International Conference on Data Mining (ICDM)*, November 2019.