

# Towards Comparing Learned Classifiers

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**The most similar real  
face of ID 8 (no access)**



**MLDiff Witness:**  
MLP: ID 8 (no access)  
LogReg: access



**The most similar real  
face with access**

# Introduction & Example

- Numerous complex, real-world applications rely on Machine Learning (ML) classifiers
- Example classification problem
  - A Support Vector Machine (SVM) & Decision Tree (DT) are both trained on the Iris dataset
  - Accuracy: SVM=96% and DT=96%
- Should we use the SVM or the DT?
- How are the two classifiers different?
- Is it relevant which one to use if they agree on known data (the train/test dataset)?

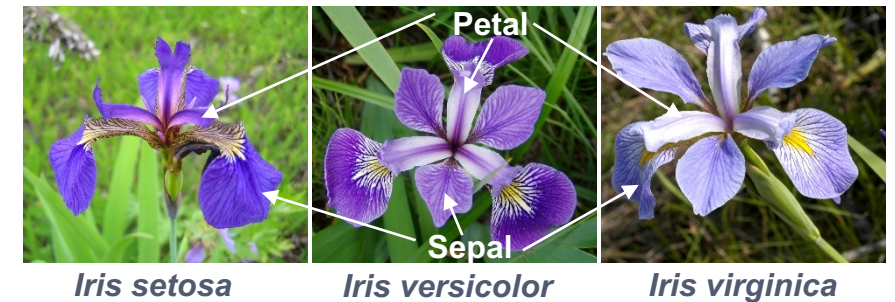


Figure: Iris flower data set  
([https://en.wikipedia.org/wiki/Iris\\_flower\\_data\\_set](https://en.wikipedia.org/wiki/Iris_flower_data_set))

DT classifies some instances as *Virginica* (**medical use**), while SVM classifies them as *Versicolor* (**poisonous**)

# Motivation of MLDiff

- State-of-the-art formal analysis of ML models lacks systematic methods to compare multiple classifiers
- Understanding classifier variants during software design and evolution is crucial for improving model quality and trust
- MLDiff aims to uncover and present differences (witnesses), i.e., disagreements, of classifiers (even those not observable in the dataset)

The LogReg classifier grants access to employee ID 8 when the MLP classifier denies it.



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The most similar real face with access

Figure: A conflict of a Multi-Layer Perceptron and a Logistic Regression classifier detected by MLDiff  
(Source: The Olivetti faces dataset from scikit-learn)

# MLDiff Implementation

- For two classifiers  $cl_1$  on features  $X_1$  and  $cl_2$  on features  $X_2$  we **encode SMT assertions** for

$$\forall d \in \mathbb{R}^{|X_1 \cup X_2|}: cl_1 \oplus cl_2 (d) = cl_1(d|_{X_1}) \times cl_2(d|_{X_2})$$

```
(declare-const x0 Real) ; one constant for each feature
; ...
(declare-const xn Real)
(declare-const cls1 Int) ; predicted class of first classifier
(declare-const cls2 Int) ; predicted class of second classifier
; assertion for classifier 1 relating x1..xn to cls1
; assertion for classifier 2 relating x1..xn to cls2
(assert (not (= cls1 cls2))) ; example query for disagreement
```

Currently supports

- Decision Trees
- Logistic Regression
- Multi Layer Perceptron (ReLU, identity)
- Support Vector Machine (linear kernels)

- Use Cases and Queries

- Differences:  $cl_1(d|_{X_1}) \neq cl_2(d|_{X_2})$
- Extension with **custom/domain constraints**:

*classifier disagreement*

$$\underbrace{x_1 = 1}_{\text{mammal (categorical } x_1\text{)}} \wedge \underbrace{x_3 = 4}_{\text{4 legs } (x_3)} \wedge \underbrace{x_6 \leq 0.2}_{\text{small (weight } x_6\text{)}} \wedge cl_1(d|_{X_1}) \neq cl_2(d|_{X_2})$$

# Challenges and Open Problems

- Supporting a larger set of functions in classifiers (SMT's arithmetic limitations)
- Scaling to complex models and queries (approximation and decomposition)
- Developing a domain-expert-friendly query language
- Generating relevant and interesting in-domain examples
- Exploring examples and explanations (summarizing and explainable AI)