

# Learning to Aggregate on Structured Data

## Master Thesis Proposal & Work Plan

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## 1 Motivation

Most of the commonly used supervised machine learning techniques assume that instances are represented by  $d$ -dimensional feature vectors  $x \in \mathcal{X} = \mathcal{X}_1 \times \dots \times \mathcal{X}_d$  for which some target value  $y \in \mathcal{Y}$  should be predicted. In the regression setting the target domain  $\mathcal{Y}$  is continuous, typically  $\mathcal{Y} = \mathbb{R}$ , whereas  $\mathcal{Y}$  is some discrete set of classes in the classification setting.

Since not all data is well-suited for a fixed-dimensional vector representation, approaches that directly consider the structure of the input data might be more appropriate in such cases. One such case is the class of so-called “*learning to aggregate*” (LTA) problems as described by Melnikov and Hüllermeier [1]. There the instances are represented by compositions  $\mathbf{c}$  of constituents  $c_i \in \mathbf{c}$ , i.e. variable-size multisets. The assumption in LTA problems is that for all constituents  $c_i$  a local valuation  $y_i \in \mathcal{Y}$  is either given or computable. The set of those local valuations should be indicative of the overall valuation  $y \in \mathcal{Y}$  of the entire composition  $\mathbf{c}$ . The goal of LTA is to learn a variadic aggregation function  $A : \mathcal{Y}^* \rightarrow \mathcal{Y}$  that estimates such composite valuations, i.e.  $\hat{y} = A(y_1, \dots, y_n)$  for a composition with  $n$  constituents. Additionally the aggregation function  $A$  should be associative and

commutative to fit with the multiset-structure of compositions.

## **2 Related Work**

## **3 Goals**

### **3.1 Required Goals**

### **3.2 Optional Goals**

## **4 Approach**

## **5 Preliminary Document Structure**

1. Introduction
2. ...

## **6 Time-Schedule**

Figure 1: Sketch of the time schedule for the work on the thesis

## References

- [1] Vitalik Melnikov and Eyke Hüllermeier. “Learning to Aggregate Using Uninorms.” In: *Machine Learning and Knowledge Discovery in Databases*. Springer International Publishing, 2016, pp. 756–771 (cit. on p. 1).

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Supervisor

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