

TECHNISCHE UNIVERSITÄT MÜNCHEN

Master's Thesis in Informatics

Neural Network Hyperparameter Optimization with Sparse Grids

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Parameteroptimierung von neuronalen Netzen mit dünnen Gittern

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Submission Date:



I confirm that this master's thesis in informatics is my own work and I have documented all sources and material used.				
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Abstract

In recent years, machine learning has gained much importance due to the increasing amount of available data. The models that are performing very different tasks have a thing in common. They have parameters that are fixed before being trained on the data. The right choice of those hyperparameters can have a huge impact on the performance which is why they have to be optimized. Different techniques like grid search, random search, and bayesian optimization tackle this problem.

In this thesis, a new approach called adaptive sparse grid search for hyperparameter optimization is introduced. This new technique allows to adapt to the hyperparameter space and the model which leads to less training and evaluation runs compared to normal grid search while still finding the optimal model configuration for the best model results.

We compare the new approach to the other three techniques mentioned regarding execution time and resulting model performance using different machine learning tasks. The results show that adaptive sparse grid search is very efficient with a model performance similar to bayesian optimization and grid search.

Zusammenfassung

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1 Introduction

2 Related Work

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- Definition machine learning
- Examples machine learning tasks and architectures
- Definition hyperparameters
- Goal of hyperparameter optimization
- 2.1.1 Grid Search
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3 Hyperparameter optimization with sparse grids

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