

Master Thesis: Compare Dataset Distillation's Impacts with Long-Rage Dataset Distillation Methods

Many projects and research endeavors encounter challenges when dealing with the training of extensive datasets. Dataset distillation emerges as a solution to substantially decrease the volume of data while maintaining comparable test accuracy. A common practice in dataset distillation involves the widespread use of data augmentation techniques.

However, there remains a gap in understanding the impact of integrating two distinct data augmentation pipelines in various sections. Therefore, in this master's thesis, we aim to investigate the effects of employing a two-stage data augmentation approach with different combinations.

It's important to note that prior knowledge of dataset distillation is not a prerequisite; a foundational understanding of deep learning and a Python framework for deep learning is sufficient for this research.

Tasks:

- Get familiar with dataset distillation method MTT and reproduce its result on images
- Design experiments with data augmentation on images
- Apply the method on time-series dataset HPC-ODA
- Design experiments with data augmentation on the time-series dataset
- Investigate into the difference

Recommended knowledge and experience:

- Experience in programming with Python
- Experience in deep learning(classes and projects)

Benefits:

- Involve in the academic environment of chair of Computer Architecture and Parallel Systems
- Research with new research topic dataset distillation

Application:

If you are interested in this topic, get in contact with Dai Liu (find the contact details below) through email.

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