

## Initial Notes:

Define the correct paths in the various scripts for saving the dataset (`dataPATH`) and the results (`resultsPATH`):

- **dataPATH**: Directory where the MNIST dataset will be stored.
- **resultsPATH**: Directory where experiment results will be saved. This must include the following subfolders:
  - `clustering`
  - `collapse`
  - `coloring`
  - `plots`
  - `training`

## Script Descriptions:

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- **coloring.py**: Contains functions for computing fibrations and opfibrations in linear layers.
  - **model.py**: Defines the MLP model (a 3-hidden-layer neural network). The `MLP` class (a `nn.Module`) includes methods for computing fibrations, opfibrations, and coverings using the network's current parameters.

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- **training.py**: Script for training the MLP model on the MNIST classification task.
    - In the `training` subfolder of `resultsPATH`, the model weights are saved throughout training as `weight\_batch\_bb.pth` (where `bb = 0, ..., 599`).
    - Training performance (accuracy) is saved in `accuracy.pth` (a 1D tensor with 600 elements).
  - **generate\_activity.py**: Generates node activity using:
    1. Samples from the evaluation subset of the dataset (10,000 samples).
    2. Randomly generated samples (200 samples).

This is computed after the 599 training steps. The results are saved in:

- `activity\_batch\_599.pth`
- `activity\_random\_input\_batch\_599.pth`

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- **fibration\_coloring.py**: Computes fibrations for the MLP model at different training steps ( $bb = 0, \dots, 599$ ).  
A `threshold` parameter determines fibrations:
    - `threshold = 0` → Each node is in its own fiber (no symmetries).
    - `threshold = 2` → All nodes are in a single large fiber (fully symmetric).

Results are stored in `coloring/fibration\_batch\_bb.pth` for different thresholds (0, 0.01, ..., 1.00).

- **synchronization\_clusters.py:** Computes synchronization clusters per layer based on network activity from random inputs (see `generate\_activity.py`). Clustering requires an `epsilon` parameter (100 values between `0` and `30`). Note: `epsilon` depends on the network's activity distribution. (\*Work in progress: Normalization of `epsilon`.\*) Results are saved in `clustering/clusters\_batch\_599.pth` for the network at step `bb = 599` across different `epsilon` values.

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- **collapse\_during\_training.py, collapse\_post\_training.py, plt\_during\_training.py, plt\_post\_training.py:** These scripts compare the performance of the original model against collapsed models (using fibration symmetries) at different training stages and thresholds. They also evaluate the reduction in network size.

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- **matching.py:** Compares synchronization clusters and fibrations, establishing a relationship between `epsilon` (synchronization) and `threshold` (fibration).