# **RSIM PROJECT**

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This is the technical documentation of our project with RSIM of TU BERLIN.

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**CHAPTER** 

ONE

**SETUP** 

First install the conda environment at cluster\_instructions/conda\_env.yml and activate it.

## 1.1 Data Set (Deep Globe Patches)

Then download the Deepglobe dataset from KAGGLE. Place it in the project with the path data/deepglobe. Then run the patch-sampling pipeline: data\_pipeline/deepglobe/patch\_sampling.py. This creates LMDB files for the train, test and valid set at data/deepglobe\_patches/[train/test/valid]/.

## 1.2 Word Embeddings

The folder data/glove contains the embeddings for the deepglobe-labels for embeddings spaces of 50 and 300. If other embeddings are necessary, or a different dataset is used. These can be created by modifying the src/wordembedding/glove.py and running it. To use this download the glove txt files from Stanford. Currently d = [50,100,200,300] are available there. For different embedding size, retrain the glove model.

## 1.3 Run Example

The most important parameters are contained in this exemplary run. For all parameters check out src/config\_args. py, for our parameterized runs check out cluster\_instructions/<model>\_<loss>.sh

```
python main.py -model CbMLC -loss weighted_bce -optim sgd -d_model 50-lr 0.0001
-add_noise 0.1 -sub_noise 0.1
```

Chapter 1. Setup

## **TECHNICAL DOCUMENTATION**

For the documentation of specific methods consider the docstrings and inline comments of these methods.

#### 2.1 Main

The main method is structured as follows:

- 1. Argument Parsing
- 2. Data Loading
- 3. Model Preparation and loading
- 4. Optimizer and Loss Setup
- 5. CUDA Setup
- 6. Training or Predict

#### **2.2 LAMP**

The Model can be found in the src/lamp/Models.py

Note: In the LAMP directory we only contributed the RESNETs in the Decoders.py and Models.py

### 2.3 Evaluation

To reconstruct our evaluation process, use the plots/plot\_training.ipynb file. The averaged results are attached in the results directory as CSV files.