Technical documentation

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1 Lumen Geoguesser

1.1 Notices:

Although you might be reading this documentation in the form of a PDF file, we highly recommend that you open the README. md file in a markdown editor (GitHub, VSCode, PyCharm, IDE...). As for the API documentation, after setting up the environment, we recommend you run the server with the python3 src/app/main.py command after which you can inspect API endpoints in browser (and execute them too!). Essentially, the technical documentation PDF is rendered from the README.md markdown file and export of the in-browser API documentation.

Few more notes:

- the documentation assumes you are located at the .lumen-geoguesser directory when running Python scripts
- all global variables are defined in src/config.py and src/paths.py
- other directories have their own README.md files which hopefully come useful
- you can run most python files with the python3 program.py -h to the sense of which arguments you can/must send and what the script actually does

1.2 Directory structure

Directory	Description
data	dataset, csvs, country shapefiles
docs	documentation
figures	figures
models	model checkpoints, model metadata
references	research papers and competition guidelines
reports	model checkpoints and model metadata
src	python source code

1.3 Setup

1.3.1 Virtual environment

Create and populate the virtual environment. Simply put, the virtual environment allows you to install Python packages only for this project (which you can easily delete later). This way, we won't clutter your global Python packages.

Step 1: Execute the following command: - the command will initialize the venv if it doesn't yet exist

```
[ ! -d "venv" ] && (echo "Creating python3 virtual environment"; python3 -m venv venv)
. venv/bin/activate
pip install -r requirements.txt
```

1.3.2 Dataset setup

This project allows multiple datasets, therefore multiple dataset directories can usually be sent to *.py programs

Step 1: If needed, rename directory data (which contains unid subdirectories) to images - The original dataset structure has a directory data (e.g dataset_original_subset/data) which contains subdirectories with unids of locations (dataset_original_subset/data/6bde8efe-a565-4f05-8c60-ae2ffb32ee9b).

Dataset structure should look like this:

```
dataset original subset/
   images
       6bde8efe-a565-4f05-8c60-ae2ffb32ee9b
        ├─ 0.jpg
        ├─ 180.jpg
          — 270.jpg
        └─ 90.jpg
      - 6c0ed2ea-b31b-4cfd-9828-4aec22bc0b37
          - 0.jpg
        ├─ 180.jpg
   data.csv
dataset_external_subset/
   images
       e61b6e5f-db0d-4f57-bbe3-4d31f16c5bc3
        ├─ 0.jpg
        ├─ 180.jpg
```

Step 2: Setup datasets with src/preprocess_setup_datasets.py - Before running other scripts you have to properly setup new dataset structure using the dataset directories as an argument and it will make sure to merge the datasets correctly. No changes will be done to your original directories.

```
python3 src/preprocess_setup_datasets.py -h
usage: preprocess_setup_datasets.py [-h] [--dataset-dirs dir [dir ...]] [--out-dir dir] [--copy-images] [--spacing SPACING
optional arguments:
 -h, --help
                        show this help message and exit
 --dataset-dirs dir [dir ...]
                        Dataset root directories that will be transformed into a single dataset
  --out-dir dir
                        Directory where the complete dataset will be placed
                        Copy images from dataset directories to the new complete directory.
 --copy-images
                        You don't need to do this as later on you will be able to pass multiple dataset directories to var
 --spacing SPACING
                        Spacing that will be used to create a grid of polygons.
                        Different spacings produce different number of classes
                        0.7 spacing => ~31 classes
                        0.5 spacing => ~55 classes
                        0.4 spacing => ~75 classes
                        0.3 spacing => ~115 classes
```

python3 src/preprocess_setup_datasets.py --dataset-dirs data/dataset_original_subset data/dataset_external_subset --out-di

preprocess_setup_datasets.py does all the necessary preprocessing. However, underneath the hood it calls other preprocessing scripts. What happens when you run this script?

- 1. directory of the new (complete) dataset is created, images are copied if --copy-images flag was passed
- 2. preprocess_csv_concat.main() is called which concatenates multiple data.csv s into a single data.csv
- 3. this new (complete) data.csv is enriched by preprocess_csv_create_rich_static.main() . Here, regions (future classes) and their information (centroids, crs centroids...) are attached to each location. Enriched data is saved to a *Rich static CSV* file created called data_rich_static_spacing_<float>_classes_<int> .
- 4. Directory images in all directories (including the complete one) will be split into train, val and test directories.

 Note: directory images won't be deleted.

New dataset structure:

```
dataset complete subset/
 — data.csv
  - images <= exists if the --copy-images flag was passed
  data_rich_static__spacing_0.5_classes_55.csv
dataset original subset/
  — data.csv
  images
  – test
      — c4a74f0d-7f30-4966-9b92-f63279139d68
        \vdash 0.jpg
        ├─ 180.jpg
    train
  val
dataset_external_subset/
  data.csv
  - images
   train
   val
```

1.4 Training

After you prepared that new dataset structure you can start the quick version of training:

```
python3 src/train.py --dataset-dirs data/dataset_external_subset/ data/dataset_original_subset/ \
--csv-rich-static data/dataset_complete_subset/data_rich_static__spacing_0.7_classes_31.csv \
--quick
```

--csv-rich-static can be left out which fill force the *Rich static CSV* creation in the runtime (this will somewhat slow down the initial setup). You can perform the full training by removing the --quick flag. Some additional interesting arguments are listed below. Run the python src/train.py -h command to see all supported arguments.

```
--image-size
--num-workers
--lr
--dataset-dirs [dir1, dir2, ...]
--csv-rich-static
--unfreeze-blocks
--pretrained
--quick
```

```
--batch-size
--optimizer
--regression
```

Example of production training (in our case):

```
python3 src/train.py \
--accelerator gpu --devices 1 --num-workers 32 \
--dataset-dir data/raw/ data/external/ \
--csv-rich-static data/complete/data_huge_spacing_0.21_num_class_211.csv \
--batch-size 8 --image-size 224 --lr 0.00002 \
--unfreeze-at-epoch 1 --scheduler plateau
```

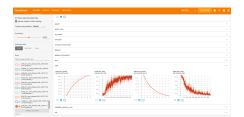
During the training a few things will occur in the reports/ directory:

- 1. reports/train_*.txt files will be create which log everything that's outputted to the standard output
- 2. subdirectory reports/<model_name> will be created in which:
 - 1. data_runtime.csv will be created, serves as backup
 - 2. version/0 directory which contains:
 - 1. hparams.yaml: details of hyperparameters
 - 2. events.out.tfevents* : log file which tensorboard consumes
 - 3. checkpoints: the most important subdirectory, contains model checkpoints (trained models)

1.5 Logs - Tensorboard

Tensorboard logging is enabled by default. To see training and validation logs, run the command bellow. Logs should be available in browser at http://localhost:6006/. For more options check tensorboard -h.

tensorboard --port 6006 --logdir reports/



1.6 Inference

After you trained a model and have the model checkpoint file (.ckpt) ready, you can run inference with predict.py file.

```
python3 src/predict.py --checkpoint <path_to_model.ckpt> \
    --images-dir data/dataset_original_subset/test
    --out data/dataset_original_subset/predictions.csv
```

1.7 Local server

Local server is useful when you are trying to do inference on a trained model. The sever code and config live in src/app directory.

Before running the sever set the variable MODEL_DIRECTORY in src/app/.env to a directory which contains (or will contain) model
checkpoints (.ckpt). Models outside of this directory can't be used for inference via endpoints. We recommend creating new
directory called models and copying model checkpoint files (e.g. reports/<model_name>/version_0/checkpoints/mymodel.ckpt
) to this directory.

Step 1. copy model checkpoints to /models/

```
mkdir models
cp -r reports/<model_name>/version_0/checkpoints/* models/
```

Step 1.1. ensure that the MODEL DIRECTORY variable is set in src/app/.env file:

```
cat src/app/.env

MODEL_DIRECTORY = "models"  # relative to the lumen-geoguesser directory
MODEL_EXTENSION = "ckpt"
PREDICT_BATCH_SIZE = 16
HOST = "localhost"
PORT = 8090
HOT_RELOAD = 0  # please don't enable hot reloading as it's unstable
```

Step 2. run the server:

python3 src/app/main.py



Before consuming the endpoints set the variable MODEL_DIRECTORY in .env to a directory that contains model checkpoints (.ckpt). Models outside of this directory can't be used for inference via endpoints.

