Usage of DSIPTS

This repo collect some examples related to the use of [dsipts] (https://gitlab.fbk.eu/dsip/dsip_dlresearch/timeseries). Using this repo you can train some DL models for timeseries forecasting using public datasets like Monarch or (six_dataset)[https://drive.google.com/drive/folders/1ZOYpTUa82_jCcxIdTmyr0LXQfvaM9vIy] used for benchmarking timeseries models.

Here we report a complete approach using Hydra that allows to better manage multiple experiments.

This particular repo is structured for working with the six datasets (that are 9 because ethh has different resolution and periods): ['electricity','etth1','etth2','ettm1','ettm2','exchange_rate','illness','traffic','weather']. The task is to predict y using past data and other past covariates for a variable number of steps depending on the paper used as reference.

Installation

In a pre-generated environment install pytorch and pytorch-lightning (pip install pytorch-lightning==1.9.4) then go inside the lib folder and execute:

```
python setup_local.py install --force
```

Alternatively, you can install it from the package registry:

pip install --force dsipts --index-url https://dsipts:glpat-98SR11neR7hzxy__SueG@gitlab.fbk.eu/api/v4/projects/4571/packages/pypi

Configuration

- copy the folder all_six_datasets inside a data folder (in what follows /home/agobbi/Projects/ExpTS/data).
- place yoursel in bash_examples
- train the models
- create the folders csv and plots in the pathdir in this case /home/agobbi/Projects/ExpTS

Hydra

In you environment install hydra and the joblib launcher for the paralellization tasks:

```
pip install hydra-core
pip install hydra-joblib-launcher  ## if you have a big gpu
```

pip install hydra-submitit-launcher ## if you are in a slurm envirionment

The script used are: - train.py for trainint - inference.py for inference - compare.py for comparing different models

Hydra is used for composing configuration files. In our case most of the parameter can be reused among the different models and are collected under the general configuration file config/config.yaml. In what follows the weather dataset is used, and notice that this dataset has a frequency of 10 minutes. The parameters here are the same described in the dsitps documentation but clearly some of them can not be modified since they depend on the selected time series. The configuration files related to this experiment can be found in config_weather; a generic config folder contains:

```
config_gpu.yaml  # containing the global configuration usually for gpu or local train, see below one example config_slurm.yaml  # containing the global configuration for slurm training see below one example compare.yaml  # instructions for comparing different models architecture/  # the folder containing the configurations specific for all the models to test config_used/  # this folder will be populated while training the models, and will be used in the comparison phase
```

The config file in the case of the weather dataset is reported and commented below.

dataset:

```
dataset: 'weather'
  path: '/home/agobbi/Projects/ExpTS/data' ##path to data. In the folder data must be present the folder six_dataset
scheduler config:
  gamma: 0.1
  step_size: 100
optim config:
  lr: 0.0005
  weight_decay: 0.01
model_configs:
  past_steps: 16
  future_steps: 16
  quantiles: [0.1,0.5,0.9] ##if you want to use quantile loss, otherwise set it to []
  past_channels : null #dataset dependent hydra expect you to set it anyway also if it depends on data
  future_channels : null #dataset dependent
  embs: null #dataset dependent
  out_channels: null #dataset dependent
```

```
split_params:
 perc_train: 0.7
 perc_valid: 0.1
 range_train: null
 range_validation: null
 range_test: null
  shift: 0
  starting_point: null
  skip_step: 1
  past_steps: model_configs@past_steps ##this is a convinient what to reuse previous information!
 future_steps: model_configs@future_steps
train_config:
 dirpath: "/home/agobbi/Projects/ExpTS"
 num_workers: 0
  auto_lr_find: true
  devices: [0]
inference:
  output_path: "/home/agobbi/Projects/ExpTS"
 load_last: true
 batch_size: 200
 num_workers: 4
  set: "validation"
 rescaling: false (sometimes you want to get the errors on normalized datasets)
#since now standard things, these two sessions are the most crucial and useful
defaults:
 - _self_
                                     # take all this configuration
 - architecture: null
                                     # and let the use specify the architecture to use (be aware that the filed here is the same
 - override hydra/launcher: joblib # use joblib for multiprocess allowing parallelization in case of multirun in a gpu/cpu envi
```

hydra:

```
launcher:
    n jobs: 2
                                # parameters indicate the number of parallel jobs in case of multirun
  output subdir: null
                                # do not save any file
  sweeper:
    params:
      architecture: glob(*)
                                # this is a way to train all the models in the architecure folder
If you are using a SLURM cluster:
defaults:
  - _self_
  - architecture: null
  - override hydra/launcher: submittit_slurm ## use slurm launcher
hydra:
  launcher:
    submittit_folder: ${hydra.sweep.dir}/.submittit/%j
    timeout min: 600
    partition: gpu-V100 ##partition to use REQURED
    mem gb: 6
                         ##gb requires
                                             REQURED
    nodes: 1
                         ##number of GPU
                                             REQURED
    gres: gpu:1
    name: ${hydra.job.name}
    target : hydra plugins.hydra submitit launcher.submitit launcher.SlurmLauncher
    setup:
      - conda activate tt ##activate the conda environment first!
In the config_weather/architecture folder there are the selected models that have the following structure:
# @package _global_ ##care this must be present!
#the specified parameters below overwrite the default configuration having a more compact representation
model:
  type: 'linear'
ts:
  name: 'weather'
```

```
version: 1
                          # if you need to versioning a model
  enrich: ['hour']
  use covariates: false # if true all the columns of the dataset will be used as past features
## for more information about models please look at the documentation [here] (https://dsip.pages.fbk.eu/dsip_dlresearch/timeserie
model configs:
  cat emb dim: 32
                          # dimension of categorical variables
  kernel_size: 5 # kernel size
  sum_emb: true
                          # if true each embdedding will be summed otherwise stacked
                          # hidden size of the fully connected block
  hidden_size: 256
  kind: 'linear'
                          # model type
  dropout_rate: 0.2
                          # dropout
                          # use or not bn layers in the first layers
  use_bn: false
  activation: 'selu'
                          # activation function
train config:
  batch size: 128
  max epochs: 250
  gradient_clip_val: null
                                   # pytorch lightening gradient clipping procedure
  gradient clip algorithm: 'norm' # pytorch lightening gradient clipping procedure
Hydra allows us to train a specific model using if you are in a gpu environment
python train.py architecture=linear --config-dir=config_weather --config-name=config_gpu
or, if you are in a slurm gpu cluster
python train.py architecture=linear --config-dir=config_weather --config-name=config_slurm -m
or a list of models in paralle:
python train.py -m architecture=linear, dlinear --config-dir=config_weather --config-name=config_slurm
or all the implemented models:
python train.py -m --config-dir=config weather --config-name=config slurm
```

In case of parallel experiment you should see at display something like:

```
(tt) agobbi@frontend:~/Projects/timeseries/bash_examples$ python train.py --config-dir=config_weather --config-name=config_slurm -m
[2023-05-03 16:46:59,679][HYDRA] Submitit 'slurm' sweep output dir : multirun/2023-05-03/16-46-58
[2023-05-03 16:46:59,680][HYDRA]
                                       #0 : architecture=attention
[2023-05-03 16:46:59,690][HYDRA]
                                       #1 : architecture=d3vae
[2023-05-03 16:46:59,696][HYDRA]
                                       #2 : architecture=dlinear
[2023-05-03 16:46:59,700][HYDRA]
                                       #3: architecture=gru
[2023-05-03 16:46:59,710][HYDRA]
                                       #4: architecture=informer
[2023-05-03 16:46:59,715][HYDRA]
                                       #5 : architecture=linear
[2023-05-03 16:46:59,720][HYDRA]
                                       #6 : architecture=lstm
[2023-05-03 16:46:59,733][HYDRA]
                                       #7 : architecture=mymodel
[2023-05-03 16:46:59,738][HYDRA]
                                       #8 : architecture=mymodel v2
[2023-05-03 16:46:59,742][HYDRA]
                                       #9 : architecture=mymodel_v3
[2023-05-03 16:46:59,747][HYDRA]
                                       #10 : architecture=nlinear
[2023-05-03 16:46:59,757][HYDRA]
                                       #11 : architecture=persistent
```

Hydra will create a folder called multirun with all the experiments lauched nested as date/time/x where x indicates the id of the lauched joib. Inside date/time/x there will be a file called train.log containing all the information logged by the system and useful for debugging.

If the row override hydra/launcher: joblib is commented the train will be consecutive, otherwise in parallel. In the latter case the output in the terminal will be a mess, please check all is woking fine. In the future the logging will be more efficient.

Once the models are trained, the relative full configurations are saved in config_used and can be used for inference or comparison:

```
python compare.py -c config/compare.yaml
```

where the compare file is:

```
models: 'config_weather'  ## path to the main config folder or list of configuration files dirpath: "/home/agobbi/Projects/ExpTS" ## where are store the models and where to put the results set: 'test'  ## set to test name: 'prova' rescaling: false  ## sometimes want to get the MSE on the scaled data or:

python compare_slurm.py --config-dir=config_weather --config-name=compare_slurm -m if you are in a SLURM cluster.
```

In the dirpath folder /home/agobbi/Projects/ExpTS/ there are three folder now: weights containing the model and the weights, plots containing some plots coming from the compare script and the csv forder containing the files.

A typical example of plot is displayed below and shows the MSE at different lags in the test set for different models:

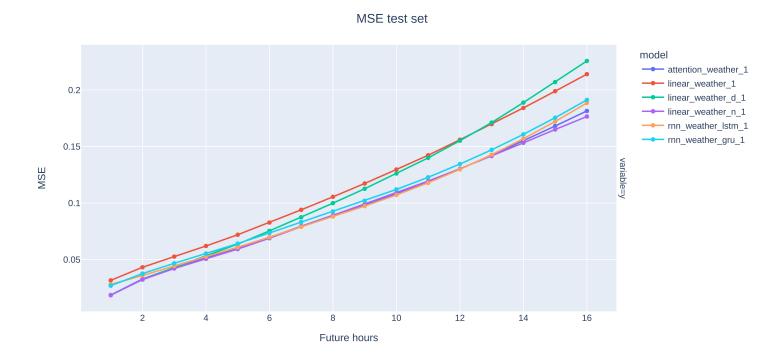


Figure 1: plot

The loss plot is currenty broken on server, you can reproduce it form the notebook 4- results (see the notebook section)

Testing

You can use the config_test for testing your models. In this case you can use smaller model with fewer epochs:

python train.py -m --config-dir=config_test --config-name=config_gpu