
EasyPL
Release 0.3

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**CHAPTER
ONE**

INSTALL GUIDE

You can install this library using pip:

```
pip install easyplib
```

Note: Sorry for the mismatch between the library name in the pypi index and the documentation. The pypi project name normalization algorithms does not allow you to specify an easypl project name.

Also you can install library manually:

```
git clone https://github.com/tam2511/EasyPL.git
cd EasyPL
python setup.py install
```


2.1 Callbacks

2.1.1 SequentialFinetuning

```
class easyp1.callbacks.finetuners.sequential_tuner.SequentialFinetuning(sequence: Dict)
```

Callback for sequence unfreezing model

sequence
Dict of dicts with unfreezing information for epochs. Dict must be like: { “epoch_num”: EPOCH_INFO, ... }

EPOCH_INFO: Dict

layers: List
List of layers names, which will be able to unfreeze

lr_gamma: float
Multiple gamma for previous param group learning rate

Type
Dict

Examples

```
>>> from easyp1.callbacks import SequentialFinetuning
... sequence = {
...     '0': {
...         'layers': ['block1.layer_name1', ...],
...     },
...
...     ...
...     '12': {
...         'layers': ['block12.layer_name13', ...],
...     },
...     '14': {
...         'layers': ['block14.layer_name3', ...],
...         'lr_gamma': 0.1
...     }
... }
... finetuner = SequentialFinetuning(sequence=sequence)
```

2.1.2 Loggers

BaseImageLogger

```
class easyp1.callbacks.loggers.base_image.BaseImageLogger(phase: str = 'train', max_samples: int = 1, class_names: Optional[List] = None, mode: str = 'first', sample_key: Optional = None, score_func: Optional[Callable] = None, largest: bool = True, dir_path: Optional[str] = None, save_on_disk: bool = False)
```

Abstract callback class for logging images

phase

Phase which will be used by this Logger. Available: [“train”, “val”, “test”, “predict”].

Type

str

max_samples

Maximum number of samples which will be logged at one epoch.

Type

int

class_names

List of class names for pretty logging. If None, then class_names will set range of number of classes.

Type

Optional[List]

mode

Mode of sample generation. Available modes: [“random”, “first”, “top”].

Type

str

sample_key

Key of batch, which define sample. If None, then sample_key will parse *learner.data_keys*.

Type

Optional

score_func

Function for score evaluation. Necessary if “mode” = “top”.

Type

Optional[Callable]

largest

Sorting order for “top” mode

Type

bool

dir_path

If defined, then logs will be writed in this directory. Else in lighting_logs.

Type

Optional[str]

save_on_disk

If true, then logs will be writed on disk to “dir_path”.

Type

bool

_post_init(trainer: Trainer, pl_module: LightningModule)

Method for initialization inverse transforms.

trainer

Trainer of pytorch-lightning

Type

pytorch_lightning.Trainer

pl_module

LightningModule of pytorch-lightning

Type

pytorch_lightning.LightningModule

ClassificationImageLogger

```
class easypl.callbacks.loggers.image_classification.ClassificationImageLogger(phase: str =  
    'train',  
    max_samples: int = 1,  
    class_names: Optional[List] = None,  
    num_classes: Optional[int] = None,  
    max_log_classes: Optional[int] = None, mode:  
    str = 'first', sample_key:  
    Optional = None, score_func:  
    Optional[Callable] = None, largest: bool =  
    True, dir_path:  
    Optional[str] = None, save_on_disk:  
    bool = False)
```

Callback class for logging images in classification task

phase

Phase which will be used by this Logger. Available: [“train”, “val”, “test”, “predict”].

Type	str
max_samples	
Type	int
class_names	
List of class names for pretty logging. If None, then class_names will set range of number of classes.	
Type	Optional[List]
num_classes	
Number of classes. Necessary if <i>class_names</i> is None.	
Type	Optional[int]
max_log_classes	
Max of number classes, which will be logged in one sample.	
Type	Optional[int]
mode	
Mode of sample generation. Available modes: [“random”, “first”, “top”].	
Type	str
sample_key	
Key of batch, which define sample. If None, then sample_key will parse <i>learner.data_keys</i> .	
Type	Optional
score_func	
Function for score evaluation. Necessary if “mode” = “top”.	
Type	Optional[Callable]
largest	
Sorting order for “top” mode	
Type	bool
dir_path	
If defined, then logs will be writed in this directory. Else in lighting_logs.	
Type	Optional[str]
save_on_disk	
If true, then logs will be writed on disk to “dir_path”.	
Type	bool

_log_on_disk(*samples*: List, *dataloader_idx*: int = 0)

Method for logging on disk.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_tensorboard(*samples*: List, *dataloader_idx*: int = 0)

Method for tensorboard logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_wandb(*samples*: List, *dataloader_idx*: int = 0)

Method for wandb logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

SegmentationImageLogger

```
class easyp1.callbacks.loggers.image_segmentation.SegmentationImageLogger(phase: str = 'train',
                                                               max_samples: int = 1,
                                                               class_names: Optional[List] = None,
                                                               num_classes: Optional[int] = None,
                                                               max_log_classes: Optional[int] = None,
                                                               mode: str = 'first',
                                                               sample_key: Optional = None,
                                                               score_func: Optional[Callable] = None,
                                                               largest: bool = True,
                                                               dir_path: Optional[str] = None,
                                                               save_on_disk: bool = False,
                                                               back_ground_class=0,
                                                               dpi=100)
```

Callback class for logging images in segmentation task

phase

Phase which will be used by this Logger. Available: [“train”, “val”, “test”, “predict”].

Type

str

max_samples

Maximum number of samples which will be logged at one epoch.

Type

int

class_names

List of class names for pretty logging. If None, then class_names will set range of number of classes.

Type

Optional[List]

num_classes

Number of classes. Necessary if *class_names* is None.

Type

Optional[int]

max_log_classes

Max of number classes, which will be logged in one sample.

Type

Optional[int]

mode

Mode of sample generation. Available modes: [“random”, “first”, “top”].

Type

str

sample_key

Key of batch, which define sample. If None, then sample_key will parse *learner.data_keys*.

Type

Optional

score_func

Function for score evaluation. Necessary if “mode” = “top”.

Type

Optional[Callable]

largest

Sorting order for “top” mode

Type

bool

dir_path

If defined, then logs will be writed in this directory. Else in lighting_logs.

Type

Optional[str]

save_on_disk

If true, then logs will be writed on disk to “dir_path”.

Type

bool

background_class

Index of background class

Type

int, default: 0

dpi

Dots per inch

Type

int, default: 100

_log_on_disk(samples: List, dataloader_idx: int = 0)

Method for logging on disk.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_tensorboard(*samples*: List, *dataloader_idx*: int = 0)

Method for tensorboard logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_wandb(*samples*: List, *dataloader_idx*: int = 0)

Method for wandb logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

DetectionImageLogger

```
class easypl.callbacks.loggers.image_detection.DetectionImageLogger(phase: str = 'train',
                                                               max_samples: int = 1,
                                                               class_names:
                                                               Optional[List] = None,
                                                               num_classes: Optional[int]
                                                               = None,
                                                               max_detections_per_image:
                                                               Optional[int] = None,
                                                               confidence: Optional[float]
                                                               = None, mode: str = 'first',
                                                               sample_key: Optional =
                                                               None, score_func:
                                                               Optional[Callable] = None,
                                                               largest: bool = True,
                                                               dir_path: Optional[str] =
                                                               None, save_on_disk: bool
                                                               = False, dpi=100)
```

Callback class for logging images in detection task

phase

Phase which will be used by this Logger. Available: ["train", "val", "test", "predict"].

Type

str

max_samples

Maximum number of samples which will be logged at one epoch.

Type	int
class_names	
	List of class names for pretty logging. If None, then class_names will set range of number of classes.
Type	
	Optional[List]
num_classes	
	Number of classes. Necessary if <i>class_names</i> is None.
Type	
	Optional[int]
max_detections_per_image	
	Max of number detections, which will be logged in one sample.
Type	
	Optional[int]
confidence	
	Min confidence of predictions, which will be logged in one sample.
Type	
	Optional[float]
mode	
	Mode of sample generation. Available modes: [“random”, “first”, “top”].
Type	
	str
sample_key	
	Key of batch, which define sample. If None, then sample_key will parse <i>learner.data_keys</i> .
Type	
	Optional
score_func	
	Function for score evaluation. Necessary if “mode” = “top”.
Type	
	Optional[Callable]
largest	
	Sorting order for “top” mode
Type	
	bool
dir_path	
	If defined, then logs will be writed in this directory. Else in lighting_logs.
Type	
	Optional[str]
save_on_disk	
	If true, then logs will be writed on disk to “dir_path”.
Type	
	bool

_log_on_disk(*samples*: List, *dataloader_idx*: int = 0)

Method for logging on disk.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_tensorboard(*samples*: List, *dataloader_idx*: int = 0)

Method for tensorboard logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_wandb(*samples*: List, *dataloader_idx*: int = 0)

Method for wandb logging.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

GANImageLogger

class easyp1.callbacks.loggers.image_gan.GANImageLogger(*phase*: str = 'train', *max_samples*: int = 1, *mode*: str = 'first', *sample_key*: Optional[None], *score_func*: Optional[Callable] = None, *largest*: bool = True, *dir_path*: Optional[str] = None, *save_on_disk*: bool = False, *dpi*=100)

Callback class for logging images in gan task

phase

Phase which will be used by this Logger. Available: ['train', 'val', 'test', 'predict'].

Type

str

max_samples

Maximum number of samples which will be logged at one epoch.

Type

int

mode

Mode of sample generation. Available modes: [“random”, “first”, “top”].

Type

str

sample_key

Key of batch, which define sample. If None, then sample_key will parse *learner.data_keys*.

Type

Optional

score_func

Function for score evaluation. Necessary if “mode” = “top”.

Type

Optional[Callable]

largest

Sorting order for “top” mode

Type

bool

dir_path

If defined, then logs will be writed in this directory. Else in lighting_logs.

Type

Optional[str]

save_on_disk

If true, then logs will be writed on disk to “dir_path”.

Type

bool

_log_on_disk(samples: List, dataloader_idx: int = 0)

Method for logging on disk.

samples

List of returns from *get_log*.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_tensorboard(samples: List, dataloader_idx: int = 0)

Method for tensorboard logging.

samples

List of returns from `get_log`.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

_log_wandb(samples: List, dataloader_idx: int = 0)

Method for wandb logging.

samples

List of returns from `get_log`.

Type

List

dataloader_idx

Index of dataloader.

Type

int, default: 0

class easyp1.callbacks.loggers.base.BaseSampleLogger(phase: str = 'train', max_samples: int = 1, class_names: Optional[List] = None, mode: str = 'first', sample_key: Optional = None, score_func: Optional[Callable] = None, largest: bool = True, dir_path: Optional[str] = None, save_on_disk: bool = False)

Abstract callback class for logging any objects

phase

Phase which will be used by this Logger. Available: [“train”, “val”, “test”, “predict”].

Type

str

max_samples

Maximum number of samples which will be logged at one epoch.

Type

int

class_names

List of class names for pretty logging. If None, then `class_names` will set range of number of classes.

Type

Optional[List]

mode

Mode of sample generation. Available modes: [“random”, “first”, “top”].

Type

str

sample_key

Key of batch, which define sample. If None, then `sample_key` will parse `learner.data_keys`.

Type	Optional
score_func	
	Function for score evaluation. Necessary if “mode” = “top”.
Type	Optional[Callable]
largest	
	Sorting order for “top” mode
Type	bool
dir_path	
	If defined, then logs will be writed in this directory. Else in lighting_logs.
Type	Optional[str]
save_on_disk	
	If true, then logs will be writed on disk to “dir_path”.
Type	bool
_log_on_disk (samples: List, dataloader_idx: int = 0)	
	Abstract method for logging on disk.
samples	
	List of returns from <i>get_log</i> .
Type	List
dataloader_idx	
	Index of dataloader.
Type	int, default: 0
_log_tensorboard (samples: List, dataloader_idx: int = 0)	
	Abstract method for tensorboard logging.
samples	
	List of returns from <i>get_log</i> .
Type	List
dataloader_idx	
	Index of dataloader.
Type	int, default: 0
_log_wandb (samples: List, dataloader_idx: int = 0)	
	Abstract method for wandb logging.
samples	
	List of returns from <i>get_log</i> .

Type
List

dataloader_idx
Index of dataloader.

Type
int, default: 0

_post_init(trainer: Trainer, pl_module: LightningModule)
Abstract method for initialization in first batch handling.

trainer
Trainer of pytorch-lightning

Type
pytorch_lightning.Trainer

pl_module
LightningModule of pytorch-lightning

Type
pytorch_lightning.LightningModule

2.1.3 Mixers

Mixup

```
class easypl.callbacks.mixers.mixup.Mixup(on_batch: bool = True, alpha: float = 0.4, num_workers: int = 1, p: float = 0.5, domen: str = 'classification')
```

Callback for mixup data operations

on_batch

If True generate samples from batch otherwise from dataset.

Type
bool

p

Mix probability.

Type
float

num_workers

Number of workers for mixing operation.

Type
int

alpha

Parameter of mixup operation.

Type
float

domen

Name of task, in which will be mixed samples. Available: ["classification", "segmentation"].

Type
str

mix(*sample1*: Dict, *sample2*: Dict) → Dict

Mixup method for two samples.

sample1

Sample of batch, which will be sampled with sample from *sample2*.

Type

Dict

sample2

Sample from batch or dataset.

Type

Dict

Returns

Mixed sample.

Return type

Dict

Cutmix

```
class easyp1.callbacks.mixers.cutmix.Cutmix(on_batch: bool = True, alpha: float = 0.4, num_workers: int = 1, p: float = 0.5, domen: str = 'classification')
```

Callback for cutmixing data operations

on_batch

If True generate samples from batch otherwise from dataset.

Type

bool

p

Mix probability.

Type

float

num_workers

Number of workers for mixing operation.

Type

int

alpha

Parameter of cutmix operation.

Type

float

domen

Name of task, in which will be mixed samples. Available: [“classification, segmentation”].

Type

str

mix(*sample1*: Dict, *sample2*: Dict) → Dict

Cutmix method for two samples.

sample1

Sample of batch, which will be sampled with sample from *sample2*.

Type

Dict

sample2

Sample from batch or dataset.

Type

Dict

Returns

Mixed sample.

Return type

Dict

Mosaic

```
class easyp1.callbacks.mixers.mosaic.Mosaic(on_batch: bool = True, n_mosaics: Union[int, List[int]] = 4, num_workers: int = 1, p: float = 0.5, domen: str = 'classification')
```

Callback for mosaic data operations

on_batch

If True generate samples from batch otherwise from dataset.

Type

bool

p

Mix probability.

Type

float

num_workers

Number of workers for mixing operation.

Type

int

n_mosaics

Number of mosaics. If it's list, then number of mosaic will be selected from this list.

Type

Union[int, List[int]]

domen

Name of task, in which will be mixed samples. Available: ["classification", "segmentation"].

Type

str

mix(*sample1*: Dict, *sample2*: Dict) → Dict

Mosaic mix method for two samples.

sample1

Sample of batch, which will be sampled with samples from *sample2*.

Type

Dict

sample2

Samples from batch or dataset.

Type

Dict

Returns

Mixed sample.

Return type

Dict

```
class easyp1.callbacks.mixers.base.MixBaseCallback(on_batch=True, samples_per: Union[int,
List[int]] = 1, p: float = 0.5, num_workers: int = 1)
```

Abstract callback for mixing data operations

on_batch

If True generate samples from batch otherwise from dataset.

Type

bool

samples_per

Number generating samples for one sample.

Type

Union[int, List[int]]

p

Mix probability.

Type

float

num_workers

Number of workers for mixing operation.

Type

int

mix(*sample1: Dict, sample2: Dict*) → Dict

Abstract method for mix operation of two samples. *sample2* can be list of samples.

sample1

Sample of batch, which will be sampled with sample/samples from *sample2*.

Type

Dict

sample2

Sample/samples from batch or dataset.

Type

Dict

Returns

Mixed sample.

Return type

Dict

2.1.4 Predictors

BaseImageTestTimeAugmentation

```
class easypl.callbacks.predictors.base_image.BaseImageTestTimeAugmentation(n: int,  
augmentations:  
List, augmentation_method: str =  
'first',  
phase='val')
```

Abstract image base callback for test-time-augmentation.

n

Number of augmentations.

Type

int

augmentations

List of augmentations, which will be used.

Type

List

augmentation_method

Method of selecting augmentations from list. Available: [“first”, “random”]

Type

str

phase

Phase which will be used by this predictor callback. Available: [“val”, “test”, “predict”].

Type

str

post_init(trainer: Trainer, pl_module: LightningModule)

Method for initialization in first batch handling. [NOT REQUIRED]

trainer

Trainer of pytorch-lightning

Type

pytorch_lightning.Trainer

pl_module

LightningModule of pytorch-lightning

Type

pytorch_lightning.LightningModule

ClassificationImageTestTimeAugmentation

```
class easypl.callbacks.predictors.image_classification.ClassificationImageTestTimeAugmentation(n:
                                            int,
                                            aug-
                                            men-
                                            ta-
                                            tions:
                                            List,
                                            aug-
                                            men-
                                            ta-
                                            tion_metho-
                                            str
                                            =
                                            'first',
                                            phase='val'
                                            re-
                                            duce_metho-
                                            Union[str,
                                            Callable]
                                            =
                                            'mean')
```

Image classification callback for test-time-augmentation

n

Number of augmentations.

Type

int

augmentations

List of augmentations, which will be used.

Type

List

augmentation_method

Method of selecting augmentations from list. Available: [“first”, “random”]

Type

str

phase

Phase which will be used by this predictor callback. Available: [“val”, “test”, “predict”].

Type

str

reduce_method

Method of result reducing

Type

Union[str, Callable]

augment(sample: Dict, augmentation) → Dict

Method for augmentation apply.

sample

Any sample of batch

Type

Dict

augmentation

Transform object

Returns

Augmented sample

Return type

Dict

metric_formatting(outputs: Any, targets: Any) → Tuple

Preparing before metric pass.

outputs

Output from model

Type

Any

targets

Targets from batch

Type

Any

Returns

Formatted outputs and targets

Return type

Tuple

post_init(trainer: Trainer, pl_module: LightningModule)

Method for initialization in first batch handling. [NOT REQUIRED]

trainer

Trainer of pytorch-lightning

Type

pytorch_lightning.Trainer

pl_module

LightningModule of pytorch-lightning

Type

pytorch_lightning.LightningModule

postprocessing(sample: Dict, dataloader_idx: int = 0) → Dict

Method for postprocessing sample

sample

Any sample of batch

Type

Dict

dataloader_idx

Index of dataloader

Type
int

Returns

Postprocessed sample

Return type

Dict

preprocessing(sample: Dict, dataloader_idx: int = 0) → Dict

Method for preprocessing sample

sample

Any sample of batch

Type
Dict

dataloader_idx

Index of dataloader

Type
int

Returns

Preprocessed sample

Return type

Dict

reduce(tensor: Tensor) → Tensor

Method for reducing of results.

tensor

Any tensor with size [batch_size X ...]

Type
torch.Tensor

Returns

Reduced tensor

Return type

torch.Tensor

```
class easyp1.callbacks.predictors.base.BaseTestTimeAugmentation(n: int, augmentations: List,
                                                               augmentation_method: str =
                                                               'first', phase='val')
```

Base callback for test-time-augmentation

n

Number of augmentations.

Type
int

augmentations

List of augmentations, which will be used.

Type

List

augmentation_method

Method of selecting augmentations from list. Available: [“first”, “random”]

Type

str

phase

Phase which will be used by this predictor callback. Available: [“val”, “test”, “predict”].

Type

str

augment(*sample*: Dict, *augmentation*) → Dict

Abstract method for augmentation apply.

sample

Any sample of batch

Type

Dict

augmentation

Transform object

Returns

Augmented sample

Return type

Dict

metric_formatting(*outputs*: Any, *targets*: Any) → Tuple

Preparing before metric pass. On default, return passed values.

outputs

Output from model

Type

Any

targets

Targets from batch

Type

Any

Returns

Formatted outputs and targets

Return type

Tuple

post_init(*trainer*: Trainer, *pl_module*: LightningModule)

Abstract method for initialization in first batch handling. [NOT REQUIRED]

trainer
Trainer of pytorch-lightning
Type
pytorch_lightning.Trainer

pl_module
LightningModule of pytorch-lightning
Type
pytorch_lightning.LightningModule

postprocessing(*sample*: Dict, *dataloader_idx*: int = 0) → Dict
Abstract method for postprocessing sample

sample
Any sample of batch
Type
Dict

dataloader_idx
Index of dataloader
Type
int

Returns
Postprocessed sample

Return type
Dict

preprocessing(*sample*: Dict, *dataloader_idx*: int = 0) → Dict
Abstract method for preprocessing sample

sample
Any sample of batch
Type
Dict

dataloader_idx
Index of dataloader
Type
int

Returns
Preprocessed sample

Return type
Dict

reduce(*tensor*: Tensor) → Tensor
Abstract method for reducing of results.

tensor
Any tensor with size [batch_size X ...]
Type
torch.Tensor

Returns
Reduced tensor

Return type
torch.Tensor

2.2 Datasets

For EasyPL to work correctly, your dataset must return a dict. For ease of creating such a class, we prepared the base class PathBaseDataset.

```
class easyp1.datasets.base.PathBaseDataset(image_prefix: str = "", path_transform: Optional[Callable] = None, transform: Optional = None)
```

Abstract class of path based dataset

image_prefix

path prefix which will be added to paths of images in csv file

Type
str

path_transform

None or function for transform of path. Will be os.path.join(image_prefix, path_transform(image_path))

Type
Optional[Callable]

transform

albumentations transform class or None

Type
Optional

__len__ () → int

Return length of dataset

Return type
int

__getitem__ (idx: int) → Dict

Read object of dataset by index

idx

index of object in dataset

Type
int

Returns

object of dataset if dict format

Return type
Dict

_read_image (image_id: Any, image_prefix: Optional[str] = None, **image_read_kwargs) → Any

Read image from disk

image_id
Any image identifier
Type
Any

image_prefix
prefix identifier with os.path.join if is str
Type
Optional

image_read_kwargs
Additional arguments for function *easypl.datasets.utils.read_image*

Returns
image object

Return type
Any

For correctly using *PathBaseDataset* you should override *__len__* and *__getitem__* methods. You can use *_read_image* method for simply image loading.

We create simply examples of datasets for classification and segmentation tasks. See below.

2.2.1 Classification

CSVDatasetClassification

```
class easypl.datasets.classification.csv.CSVDatasetClassification(csv_path: str, image_prefix: str = '', path_transform: Optional[Callable] = None, transform=None, return_label: bool = True, image_column: Optional[str] = None, target_columns: Optional[Union[str, List[str]]] = None)
```

Csv dataset for classification

csv_path

path to csv file with paths of images

Type
str

return_label

if True return dict with two keys (image, target), else return dict with one key (image)

Type
bool

image_column

column name or None. If None then will be getting the first column

Type
Optional[str]

target_columns

column name/names or None. If None then will be getting all but the first column

Type

Optional[Union[str, List[str]]]

image_prefix

path prefix which will be added to paths of images in csv file

Type

str

path_transform

None or function for transform of path. Will be os.path.join(image_prefix, path_transform(image_path))

Type

Optional[Callable]

transform

albumentations transform class or None

Type

Optional

__len__() → int

Return length of dataset

Return type

int

__getitem__(idx: int) → Dict

Read object of dataset by index

idx

index of object in dataset

Type

int

Returns

{“image”: …} or {“image”: …, “target”: …}

Return type

Dict

DirDatasetClassification

```
class easypl.datasets.classification.dir.DirDatasetClassification(root_path: str, label_parser:  
                                                                    Callable, transform: Optional  
                                                                    = None, return_label: bool =  
                                                                    True)
```

Dataset implementation for images in directory on disk (stored images paths in RAM). Require root_path/.../image_path structure.

root_path

path of directory with images

Type

str

transform

albumentations transform or None

Type

Optional

return_label

if True return dict with two keys (image, target), else return dict with one key (image)

Type

bool

label_parser

function for parsing label from relative path

Type

Callable

__len__() → int

Return length of dataset

Return type

int

__getitem__(idx: int) → Dict

Read object of dataset by index

idx

index of object in dataset

Type

int

Returns

{“image”: …} or {“image”: …, “target”: …}

Return type

Dict

2.2.2 Segmentation

CSVDatasetSegmentation

```
class easypl.datasets.segmentation.csv.CSVDatasetSegmentation(csv_path: str, image_prefix: str = '',
                                                               mask_prefix: str = '',
                                                               path_transform:
                                                               Optional[Callable] = None,
                                                               transform: Optional = None,
                                                               return_label: bool = True,
                                                               image_column: Optional[str] =
                                                               None, target_column: Optional[str]
                                                               = None)
```

Csv dataset for segmentation

csv_path

path to csv file with paths of images

Type
str

return_label
if True return dict with two keys (image, mask), else return dict with one key (image)

Type
bool

image_column
column name or None. If None then will be getting the first column

Type
Optional[str]

target_column
column name or None. If None then will be getting all but the second column

Type
Optional[str]

image_prefix
path prefix which will be added to paths of images in csv file

Type
str

mask_prefix
path prefix which will be added to paths of masks in csv file

Type
str

path_transform
None or function for transform of path. Will be os.path.join(image_prefix, path_transform(image_path))

Type
Optional[Callable]

transform
albumentations transform class or None

Type
Optional

__len__() → int
Return length of dataset

Return type
int

__getitem__(idx: int) → Dict
Read object of dataset by index

idx
index of object in dataset

Type
int

Returns
{“image”: …} or {“image”: …, “mask”: …}

Return type
Dict

2.2.3 Detection

CSVDatasetDetection

```
class easypl.datasets.detection.csv.CSVDatasetDetection(csv_path: str, image_prefix: str = "",  
path_transform: Optional[Callable] = None, transform=None, return_label: bool = True, image_column: Optional[str] = None, target_column: Optional[str] = None)
```

Csv dataset for detection

csv_path

path to csv file with paths of images

Type
str

return_label

if True return dict with two keys (image, annotations), else return dict with one key (image)

Type
bool

image_column

column name or None. If None then will be getting the first column

Type
Optional[str]

target_column

column name/names or None. If None then will be getting all but the second column

Type
Optional[str]

image_prefix

path prefix which will be added to paths of images in csv file

Type
str

path_transform

None or function for transform of path. Will be os.path.join(image_prefix, path_transform(image_path))

Type
Optional[Callable]

transform

albumentations transform class or None

Type
Optional

__len__ () → int
Return length of dataset

Return type
int

__getitem__ (idx: int) → Dict
Read object of dataset by index

idx
index of object in dataset

Type
int

Returns
{“image”: …} or {“image”: …, “annotations”: …}

Return type
Dict

2.3 Learners

2.3.1 ClassificationLearner

```
class easypl.learners.classification.ClassificationLearner(model: Optional[Union[Module,
List[Module]]] = None, loss: Optional[Union[Module,
List[Module]]] = None, optimizer: Optional[Union[WrapperOptimizer,
List[WrapperOptimizer]]] = None, lr_scheduler: Optional[Union[WrapperScheduler,
List[WrapperScheduler]]] = None, train_metrics: Optional[List[Metric]] =
None, val_metrics: Optional[List[Metric]] = None, test_metrics: Optional[List[Metric]] =
None, data_keys: Optional[List[str]] = None, target_keys: Optional[List[str]] =
None, multilabel: bool = False)
```

Classification learner.

model

torch.nn.Module model.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

loss

torch.nn.Module loss function.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

optimizer

Optimizer wrapper object.

Type

Optional[Union[*WrapperOptimizer*, List[*WrapperOptimizer*]]]

lr_scheduler

Scheduler object for lr scheduling.

Type

Optional[Union[*WrapperScheduler*, List[*WrapperScheduler*]]]

train_metrics

List of train metrics.

Type

Optional[List[Metric]]

val_metrics

List of validation metrics.

Type

Optional[List[Metric]]

test_metrics

List of test metrics.

Type

Optional[List[Metric]]

data_keys

List of data keys

Type

Optional[List[str]]

target_keys

List of target keys

Type

Optional[List[str]]

multilabel

If classification task is multilabel.

Type

bool

forward(*samples*: Tensor) → Tensor

Standart method for forwarding model. ... attribute:: samples

Image tensor.

type

`torch.Tensor`

Returns

Output from model.

Return type

`torch.Tensor`

get_outputs(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

get_targets(batch: Dict, optimizer_idx: int = 0) → Dict

Method for selecting and preprocessing targets from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

loss_step(outputs: Tensor, targets: Tensor, optimizer_idx: int = 0) → Dict

Method for loss evaluating.

outputs

Outputs from model

Type

torch.Tensor

targets

Targets from batch

Type

torch.Tensor

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “log”]

Return type

Dict

2.3.2 RecognitionLearner

```
class easypl.learners.recognition.RecognitionLearner(model: Optional[Union[Module,
    List[Module]]] = None, loss:
    Optional[Union[Module, List[Module]]] = None, optimizer:
    Optional[Union[WrapperOptimizer,
    List[WrapperOptimizer]]] = None, lr_scheduler:
    Optional[Union[WrapperScheduler,
    List[WrapperScheduler]]] = None, train_metrics: Optional[List[Metric]] = None,
    val_metrics: Optional[List[Metric]] = None, test_metrics: Optional[List[Metric]] = None,
    data_keys: Optional[List[str]] = None, target_keys: Optional[List[str]] = None,
    multilabel: bool = False)
```

Recognition learner.

model

`torch.nn.Module` model.

Type

`Optional[Union[torch.nn.Module, List[torch.nn.Module]]]`

loss

`torch.nn.Module` loss function.

Type

`Optional[Union[torch.nn.Module, List[torch.nn.Module]]]`

optimizer

Optimizer wrapper object.

Type

`Optional[Union[WrapperOptimizer, List[WrapperOptimizer]]]`

lr_scheduler

Scheduler object for lr scheduling.

Type

`Optional[Union[WrapperScheduler, List[WrapperScheduler]]]`

train_metrics

List of train metrics.

Type

`Optional[List[Metric]]`

val_metrics

List of validation metrics.

Type
Optional[List[Metric]]

test_metrics

List of test metrics.

Type
Optional[List[Metric]]

data_keys

List of data keys

Type
Optional[List[str]]

target_keys

List of target keys

Type
Optional[List[str]]

multilabel

If recognition task is multilabel.

Type
bool

forward(samples: Tensor) → Tensor

Standart method for forwarding model. .. attribute:: samples

Image tensor.

type
torch.Tensor

Returns

Output from model.

Return type

torch.Tensor

get_outputs(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type
Dict

optimizer_idx

Index of optimizer

Type
int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

get_targets(batch: Dict, optimizer_idx: int = 0) → Dict
Method for selecting and preprocessing targets from batch

batch
Batch in step
Type
Dict

optimizer_idx
Index of optimizer
Type
int

Returns
Dict with keys: [“loss”, “metric”, “log”]

Return type
Dict

loss_step(outputs: Tensor, targets: Tensor, optimizer_idx: int = 0) → Dict
Method for loss evaluating.

outputs
Outputs from model
Type
torch.Tensor

targets
Targets from batch
Type
torch.Tensor

optimizer_idx
Index of optimizer
Type
int

Returns
Dict with keys: [“loss”, “log”]

Return type
Dict

2.3.3 SegmentationLearner

```
class easyp1.learners.segmentation.SegmentationLearner(model: Optional[Union[Module,
    List[Module]]] = None, loss:
    Optional[Union[Module, List[Module]]] = None, optimizer:
    Optional[Union[WrapperOptimizer,
    List[WrapperOptimizer]]] = None, lr_scheduler:
    Optional[Union[WrapperScheduler,
    List[WrapperScheduler]]] = None, train_metrics: Optional[List[Metric]] =
    None, val_metrics: Optional[List[Metric]] = None, test_metrics: Optional[List[Metric]] =
    None, data_keys: Optional[List[str]] = None, target_keys: Optional[List[str]] =
    None, multilabel: bool = False)
```

Segmenatation learner.

model

torch.nn.Module model.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

loss

torch.nn.Module loss function.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

optimizer

Optimizer wrapper object.

Type

Optional[Union[WrapperOptimizer, List[WrapperOptimizer]]]

lr_scheduler

Scheduler object for lr scheduling.

Type

Optional[Union[WrapperScheduler, List[WrapperScheduler]]]

train_metrics

List of train metrics.

Type

Optional[List[Metric]]

val_metrics

List of validation metrics.

Type

Optional[List[Metric]]

test_metrics

List of test metrics.

Type

Optional[List[Metric]]

data_keys

List of data keys

Type

Optional[List[str]]

target_keys

List of target keys

Type

Optional[List[str]]

multilabel

If segmentation task is multilabel.

Type

bool

forward(samples: Tensor) → Tensor

Standart method for forwarding model. .. attribute:: samples

Image tensor.

type

torch.Tensor

Returns

Output from model.

Return type

torch.Tensor

get_outputs(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

get_targets(batch: Dict, optimizer_idx: int = 0) → Dict

Method for selecting and preprocessing targets from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type
int**Returns**

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

loss_step(outputs: Tensor, targets: Tensor, optimizer_idx: int = 0) → Dict

Method fow loss evaluating.

outputs

Outputs from model

Type
torch.Tensor**targets**

Targets from batch

Type
torch.Tensor**optimizer_idx**

Index of optimizer

Type
int**Returns**

Dict with keys: [“loss”, “log”]

Return type

Dict

2.3.4 DetectionLearner

```
class easyp1.learners.detection.DetectionLearner(model: Optional[Union[Module, List[Module]]] = None, loss: Optional[Union[Module, List[Module]]] = None, optimizer: Optional[Union[WrapperOptimizer, List[WrapperOptimizer]]] = None, lr_scheduler: Optional[Union[WrapperScheduler, List[WrapperScheduler]]] = None, train_metrics: Optional[List[Metric]] = None, val_metrics: Optional[List[Metric]] = None, test_metrics: Optional[List[Metric]] = None, data_keys: Optional[List[str]] = None, target_keys: Optional[List[str]] = None, image_size_key: Optional[str] = None, image_scale_key: Optional[str] = None, postprocessing: Optional[BasePostprocessing] = None)
```

Detection learner.

model

torch.nn.Module model.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

loss

torch.nn.Module loss function.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

optimizer

Optimizer wrapper object.

Type

Optional[Union[*WrapperOptimizer*, List[*WrapperOptimizer*]]]

lr_scheduler

Scheduler object for lr scheduling.

Type

Optional[Union[*WrapperScheduler*, List[*WrapperScheduler*]]]

train_metrics

List of train metrics.

Type

Optional[List[Metric]]

val_metrics

List of validation metrics.

Type

Optional[List[Metric]]

test_metrics

List of test metrics.

Type

Optional[List[Metric]]

data_keys

List of data keys

Type

Optional[List[str]]

target_keys

List of target keys

Type

Optional[List[str]]

postprocessing

If postprocessing is not None then this

Type

Optional

forward(*samples: Tensor*) → *Tensor*

Standart method for forwarding model. ... attribute:: samples

Image tensor.

type

torch.Tensor

Returns

Output from model.

Return type

torch.Tensor

get_outputs(*batch: Dict*, *optimizer_idx: int = 0*) → *Dict*

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

get_targets(*batch: Dict*, *optimizer_idx: int = 0*) → *Dict*

Method for selecting and preprocessing targets from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

loss_step(*outputs: Tensor*, *targets: Tensor*, *optimizer_idx: int = 0*) → *Dict*

Method fow loss evaluating.

outputs
 Outputs from model
Type
 torch.Tensor

targets
 Targets from batch
Type
 torch.Tensor

optimizer_idx
 Index of optimizer
Type
 int

Returns
 Dict with keys: [“loss”, “log”]

Return type
 Dict

2.3.5 GANLearner

```
class easyp1.learners.gan.GANLearner(model: Optional[List[Module]] = None, loss:
                                         Optional[List[Module]] = None, optimizer:
                                         Optional[List[WrapperOptimizer]] = None, lr_scheduler:
                                         Optional[Union[WrapperScheduler, List[WrapperScheduler]]] =
                                         None, train_metrics: Optional[List[Metric]] = None, val_metrics:
                                         Optional[List[Metric]] = None, test_metrics:
                                         Optional[List[Metric]] = None, data_keys: Optional[List[str]] =
                                         None, target_keys: Optional[List[str]] = None)
```

Generative adversarial networks learner.

model
 Generative adversarial networks.
Type
 Optional[List[torch.nn.Module]]

loss
 torch.nn.Module losses function.

Type
 Optional[List[torch.nn.Module]]

optimizer
 Optimizers wrapper object.

Type
 Optional[List[WrapperOptimizer]]

lr_scheduler
 Scheduler object for lr scheduling.

Type
 Optional[Union[WrapperScheduler, List[WrapperScheduler]]]

train_metrics

List of train metrics.

Type

Optional[List[Metric]]

val_metrics

List of validation metrics.

Type

Optional[List[Metric]]

test_metrics

List of test metrics.

Type

Optional[List[Metric]]

data_keys

List of data keys

Type

Optional[List[str]]

target_keys

List of target keys

Type

Optional[List[str]]

forward(samples: Tensor) → Tensor

Standart method for forwarding model. .. attribute:: samples

Image tensor.

type

torch.Tensor

Returns

Output from model.

Return type

torch.Tensor

get_outputs(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type

Dict

optimizer_idx

Index of optimizer

Type

int

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type
Dict

get_targets(batch: Dict, optimizer_idx: int = 0) → Dict
Method for selecting and preprocessing targets from batch

batch
Batch in step
Type
Dict

optimizer_idx
Index of optimizer
Type
int

Returns
Dict with keys: [“loss”, “metric”, “log”]

Return type
Dict

loss_step(outputs: Dict, targets: Dict, optimizer_idx: int = 0) → Dict
Method for loss evaluating.

outputs
Outputs from model
Type
torch.Tensor

targets
Targets from batch
Type
torch.Tensor

optimizer_idx
Index of optimizer
Type
int

Returns
Dict with keys: [“loss”, “log”]

Return type
Dict

```
class easyppl.learners.base.BaseLearner(model: Optional[Union[Module, List[Module]]] = None, loss: Optional[Union[Module, List[Module]]] = None, optimizer: Optional[Union[WrapperOptimizer, List[WrapperOptimizer]]] = None, lr_scheduler: Optional[Union[WrapperScheduler, List[WrapperScheduler]]] = None, train_metrics: Optional[List[Metric]] = None, val_metrics: Optional[List[Metric]] = None, test_metrics: Optional[List[Metric]] = None, data_keys: Optional[List[str]] = None, target_keys: Optional[List[str]] = None)
```

Abstract base learner

model

torch.nn.Module model.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

loss

torch.nn.Module loss function.

Type

Optional[Union[torch.nn.Module, List[torch.nn.Module]]]

optimizer

Optimizer wrapper object.

Type

Optional[Union[*WrapperOptimizer*, List[*WrapperOptimizer*]]]

lr_scheduler

Scheduler object for lr scheduling.

Type

Optional[Union[*WrapperScheduler*, List[*WrapperScheduler*]]]

train_metrics

List of train metrics.

Type

Optional[List[Metric]]

val_metrics

List of validation metrics.

Type

Optional[List[Metric]]

test_metrics

List of test metrics.

Type

Optional[List[Metric]]

data_keys

List of data keys

Type

Optional[List[str]]

target_keys

List of target keys

Type

Optional[List[str]]

get_outputs(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing outputs from batch

batch

Batch in step

Type

Dict

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

get_targets(batch: Dict, optimizer_idx: int = 0) → Dict

Abstract method for selecting and preprocessing targets from batch

batch

Batch in step

Type

Dict

Returns

Dict with keys: [“loss”, “metric”, “log”]

Return type

Dict

loss_step(outputs: Any, targets: Any, optimizer_idx: int = 0) → Dict

Abstract method for loss evaluating.

outputs

Any outputs from model

Type

Any

targets

Any targets from batch

Type

Any

Returns

Dict with keys: [“loss”, “log”]

Return type

Dict

2.4 Losses

2.4.1 Segmentation Losses

DiceLoss

```
class easyp1.losses.segmentation.diceloss.DiceLoss(weight: Optional[Tensor] = None, ignore_index: Optional[int] = None, **kwargs)
```

Dice loss, need one hot encode input. Taken from: <https://github.com/hubutui/DiceLoss-PyTorch/blob/master/loss.py>. Inputs must be a tensor of shape [N, C, *].

weight

An array of shape [num_classes,].

Type
Optional[torch.Tensor]

ignore_index
Class index to ignore.

Type
Optional[int]

kwargs
Additional arguments.

Returns
Loss value

Return type
torch.Tensor

2.5 Lr schedulers

```
class easypl.lr_schedulers.wrapper.WrapperScheduler(scheduler_class, **kwargs)
```

Wrapper for pytorch Learning rate Scheduler class.

scheduler_class
Pytorch Learning rate Scheduler class.

__call__(optimizer: Optimizer)
Return lr_scheduler

optimizer
WrapperOptimizer object
Type
Optimizer

2.6 Metrics

2.6.1 Classification metrics

SearchAccuracy

```
class easypl.metrics.classification.search_accuracy.SearchAccuracy(k: Union[int, List] = 1,  
batch_size: int = 512,  
distance: Union[str,  
Callable] = 'L2', largest:  
bool = True,  
dist_sync_on_step: bool =  
False, compute_on_step:  
bool = True)
```

Version of accuracy for search case

k

SearchAccuracy return top k (top (k[0], k[1], ...) if k is list) accuracy rate.

Type

Union[int, List]

batch_size

Batch size for evaluate distance operations.

Type

int

distance

Name or function of distance.

Type

Union[str, Callable]

largest

If True metric evaluate top largest samples, else evaluate smallest samples.

Type

bool

SearchMAP

```
class easypl.metrics.classification.search_mean_average_precision.SearchMAP(k: Union[int,
list] = 1,
batch_size: int =
512, distance:
Union[str,
Callable] = 'L2',
largest: bool =
True,
dist_sync_on_step:
bool = False,
com-
pute_on_step:
bool = True)
```

Version of mean average precision for search case

k

SearchMAP return top k (top (k[0], k[1], ...) if k is list) accuracy rate.

Type

Union[int, List]

batch_size

Batch size for evaluate distance operations.

Type

int

distance

Name or function of distance.

Type

Union[str, Callable]

largest

If True metric evaluate top largest samples, else evaluate smallest samples.

Type

bool

2.6.2 Segmentation metrics

Pixel level metrics

Pixel level Accuracy

```
class easypl.metrics.segmentation.pixel_level.PixelLevelAccuracy(average: str = 'macro',
                                                               num_classes: int = 0,
                                                               threshold: float = 0.5)
```

Pixel-level accuracy segmentation metric.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

Pixel level Precision

```
class easypl.metrics.segmentation.pixel_level.PixelLevelPrecision(average: str = 'macro',
                                                               num_classes: int = 0,
                                                               threshold: float = 0.5, epsilon:
                                                               float = 1e-08)
```

Pixel-level precision segmentation metric.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

epsilon

Epsilon for correct evaluating metric.

Type

float

Pixel level Recall

```
class easypl.metrics.segmentation.pixel_level.PixelLevelRecall(average: str = 'macro',  

num_classes: int = 0, threshold:  

float = 0.5, epsilon: float = 1e-08)
```

Pixel-level recall segmentation metric.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

epsilon

Epsilon for correct evaluating metric.

Type

float

Pixel level FBeta

```
class easypl.metrics.segmentation.pixel_level.PixelLevelFBeta(average: str = 'macro',  

num_classes: int = 0, threshold:  

float = 0.5, beta: float = 1.0,  

epsilon: float = 1e-08)
```

Pixel-level f-beta segmentation metric.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

beta

Param of metric F-beta

Type

float

epsilon

Epsilon for correct evaluating metric.

Type

float

Pixel level F1

```
class easypl.metrics.segmentation.pixel_level.PixelLevelF1(average: str = 'macro', num_classes:  
int = 0, threshold: float = 0.5, epsilon:  
float = 1e-08)
```

Pixel-level f1 segmentation metric.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

epsilon

Epsilon for correct evaluating metric.

Type

float

```
class easypl.metrics.segmentation.pixel_level.PixelLevelBase(average: str = 'macro', num_classes:  
int = 0, threshold: float = 0.5)
```

Abstract class for pixel-level segmentation metrics.

average

Method of averaging.

Type

str

num_classes

Number of classes.

Type

int

threshold

Threshold for probabilities of pixels.

Type

float

2.6.3 Detection metrics

MAP

```
class easypl.metrics.detection.mean_average_precision.MAP(iou_thresholds: Optional[List[float]] = None, rec_thresholds: Optional[List[float]] = None, max_detection_thresholds: Optional[List[int]] = None, class_metrics: bool = False, **kwargs)
```

Wrapper for MeanAveragePrecision from torchmetrics.

iou_thresholds

Iou threshold/thresholds for boxes.

Type

Optional[List[float]]

rec_thresholds

Recall thresholds for evaluation.

Type

Optional[List[float]]

max_detection_thresholds

Thresholds on max detections per image.

Type

Optional[List[int]]

class_metrics

Option to enable per-class metrics for mAP and mAR_100.

Type

bool

kwargs

Torchmetrics Metric args.

reset()

This method automatically resets the metric state variables to their default value.

BaseDetectionMetric

```
class easypl.metrics.detection.base.BaseDetectionMetric(iou_threshold: Union[float, List[float]],  
                                                       confidence: Union[float, List[float]],  
                                                       num_classes: Optional[int] = None,  
                                                       **kwargs)
```

Base detection metric. Compute true positive, false negative and false positive metrics.

iou_threshold

Iou threshold/thresholds for boxes.

Type

Union[float, List[float]]

confidence

Confidence/confidences thresholds.

Type

Union[float, List[float]]

num_classes

Number of classes.

Type

Optional[int]

kwargs

Torchmetrics Metric args.

reset()

This method automatically resets the metric state variables to their default value.

FBetaDetection

```
class easypl.metrics.detection.f_beta.FBetaDetection(iou_threshold: Union[float, List[float]],  
                                                       confidence: Optional[List[float]] = None,  
                                                       num_classes: Optional[int] = None, beta: float  
                                                       = 1.0, eps: float = 1e-09, **kwargs)
```

Evaluate optimal confidence by F beta metric and return with precision, recall values.

iou_threshold

Iou threshold/thresholds for boxes.

Type

Union[float, List[float]]

confidence

Confidence/confidences thresholds or None. If is None then evaluate as arange from 0 to 1 with step 0.05.

Type

Optional[List[float]]

num_classes

Number of classes.

Type

Optional[int]

beta

Parameter of F metric.

Type

float

eps

Epsilon.

Type

float

kwargs

Torchmetrics Metric args.

```
class easypl.metrics.base.MetricsList(dist_sync_on_step: bool = False, compute_on_step: bool = True)
```

List of metrics

clone()

Make a copy of the metric.

```
class easypl.metrics.torch.TorchMetric(metric: Metric, class_names: Optional[List] = None)
```

Wrapper for metrics from torchmetrics

metric

Metric object from torchmetrics.

Type

Metric

class_names

Names of classes.

Type

Optional[List]

Examples

```
>>> from torchmetrics import F1
... from easypl.metrics import TorchMetric
... result_metric = TorchMetric(F1(), class_names=None)
```

2.7 Optimizers

```
class easypl.optimizers.wrapper.WrapperOptimizer(optimizer_cls, **kwargs)
```

Wrapper for pytorch Optimizer class.

optimizer_cls

Pytorch Optimizer class.

kwargs

Additional arguments.

__call__(params) → Optimizer

Return optimizer.

params

Model params

Returns

Optimizer object

Return type

Optimizer

EXAMPLES

3.1 Simple multiclass classification example

We can observe simple example for classification task with two classes (cats and dogs).

First, you should import common libraries and packages below. (If you don't have some package, than install it).

```
import cv2
from torch.utils.data import Dataset, DataLoader
import torch
import torch.optim as optim
import torch.nn as nn
from pytorch_lightning import Trainer
from pytorch_lightning.callbacks import ModelCheckpoint
from albumentations.augmentations import *
from albumentations.core.composition import *
from albumentations.pytorch.transforms import *
from timm import create_model
import random
from torchmetrics import *
import shutil
```

Than you can import EasyPL packages, as like:

```
from easypl.learners import ClassificationLearner
from easypl.metrics import TorchMetric
from easypl.optimizers import WrapperOptimizer
from easypl.lr_schedulers import WrapperScheduler
from easypl.datasets import CSVDatasetClassification
from easypl.callbacks import ClassificationImageLogger
from easypl.callbacks import Cutmix
from easypl.callbacks import ClassificationImageTestTimeAugmentation
```

Than you should define datasets and dataloaders. You can use this simple example:

```
train_transform = Compose([
    HorizontalFlip(p=0.5),
    Rotate(p=0.5),
    LongestMaxSize(max_size=224),
    PadIfNeeded(min_height=224, min_width=224, border_mode=cv2.BORDER_CONSTANT, value=0, u
    ↵mask_value=0),
```

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```

Normalize(),
ToTensorV2(),
])

val_transform = Compose([
    LongestMaxSize(max_size=600),
    PadIfNeeded(min_height=600, min_width=600, border_mode=cv2.BORDER_CONSTANT, value=0,
    mask_value=0),
    Normalize(),
    ToTensorV2(),
])

train_dataset = CSVDatasetClassification('../input/cat-dog-test/train.csv', image_prefix=
    '../input/cat-dog-test/train', transform=train_transform, return_label=True)
val_dataset = CSVDatasetClassification('../input/cat-dog-test/val.csv', image_prefix='../
    input/cat-dog-test/val', transform=val_transform, return_label=True)

train_dataloader = DataLoader(train_dataset, batch_size=16, shuffle=True, pin_
    memory=True, num_workers=2)
val_dataloader = DataLoader(train_dataset, batch_size=16, shuffle=False, pin_memory=True,
    num_workers=2)

```

Than we should define model (used timm), loss function, optimizer and metrics:

```

model = create_model('resnet18', pretrained=True, num_classes=2)

loss_f = nn.CrossEntropyLoss()

optimizer = WrapperOptimizer(optim.Adam, lr=1e-4)
lr_scheduler = WrapperScheduler(optim.lr_scheduler.StepLR, step_size=2, gamma=1e-1,
    interval='epoch')

train_metrics = []
val_metrics = [TorchMetric(F1(num_classes=2, average='none'), class_names=['cat', 'dog
    '])]

```

If you need in callbacks, you can use our simple realization. Creating of callbacks looks like:

```

# Logger of outputs (images)
image_logger = ClassificationImageLogger(
    phase='train',
    max_samples=10,
    class_names=['cat', 'dog'],
    max_log_classes=2,
    dir_path='images',
    save_on_disk=True,
)

# Cutmix callback
cutmix = Cutmix(
    on_batch=True,
    p=1.0,

```

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```

        domen='classification',
    )

# Test time augmentation callback
tta = ClassificationImageTestTimeAugmentation(
    n=2,
    augmentations=[VerticalFlip(p=1.0)],
    phase='val'
)

```

In finally, we should define learner and trainer, and than run training.

```

learner = ClassificationLearner(
    model=model,
    loss=loss_f,
    optimizer=optimizer,
    lr_scheduler=lr_scheduler,
    train_metrics=train_metrics,
    val_metrics=val_metrics,
    data_keys=['image'],
    target_keys=['target'],
    multilabel=False
)
trainer = Trainer(
    gpus=1,
    callbacks=[image_logger, cutmix, tta],
    max_epochs=3,
    precision=16
)
trainer.fit(learner, train_dataloaders=train_dataloader, val_dataloaders=[val_
    ↴dataloader])

```

3.2 Simple semantic segmentation example

We can observe simple example for segmentation task with one class (text).

First, you should import common libraries and packages below. (If you don't have some package, than install it).

```

import cv2
import numpy as np
from torch.utils.data import Dataset, DataLoader
import torch
import torch.optim as optim
import torch.nn as nn
from pytorch_lightning import Trainer
from pytorch_lightning.callbacks import ModelCheckpoint
from albumentations.augmentations import *
from albumentations.core.composition import *
from albumentations.pytorch.transforms import *
from timm import create_model
import random

```

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```
from torchmetrics import *
import shutil
```

Than you can import EasyPL packages, as like:

```
from easypl.learners import SegmentationLearner
from easypl.metrics import TorchMetric
from easypl.metrics.segmentation import PixelLevelF1
from easypl.optimizers import WrapperOptimizer
from easypl.lr_schedulers import WrapperScheduler
from easypl.datasets import CSVDatasetSegmentation
from easypl.callbacks import SegmentationImageLogger
from easypl.callbacks import Mixup
from easypl.losses.segmentation import DiceLoss
```

Than you should define datasets and dataloaders. You can use this simple example:

```
train_transform = Compose([
    HorizontalFlip(p=0.5),
    Rotate(p=0.5),
    LongestMaxSize(max_size=224),
    PadIfNeeded(min_height=224, min_width=224, border_mode=cv2.BORDER_CONSTANT, value=0,
    mask_value=0),
    Normalize(),
    ToTensorV2(),
])

val_transform = Compose([
    LongestMaxSize(max_size=600),
    PadIfNeeded(min_height=600, min_width=600, border_mode=cv2.BORDER_CONSTANT, value=0,
    mask_value=0),
    Normalize(),
    ToTensorV2(),
])

dataset = CSVDatasetSegmentation(
    csv_path='../input/lan-segmentation-1/train.csv',
    image_prefix='../input/lan-segmentation-1/train/images',
    mask_prefix='../input/lan-segmentation-1/train/masks',
    image_column='path',
    target_column='path'
)

class WrapperDataset(Dataset):
    def __init__(self, dataset, transform=None, idxs=[]):
        self.dataset = dataset
        self.transform = transform
        self.idxs = idxs

    def __getitem__(self, idx):
        idx = self.idxs[idx]
        row = self.dataset[idx]
```

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```

row['mask'][row['mask'] < 150] = 0
row['mask'][row['mask'] > 150] = 1
if self.transform:
    result = self.transform(image=row['image'], mask=row['mask'])
    row['image'] = result['image']
    row['mask'] = result['mask'].to(dtype=torch.long)
row['mask'] = one_hot(row['mask'], num_classes=2).permute(2, 0, 1)
return row

def __len__(self):
    return len(self.idxs)

image_size = 768

train_transform = A.Compose([
    A.HorizontalFlip(p=0.5),
    A.ColorJitter(p=0.7),
    A.LongestMaxSize(max_size=image_size),
    A.PadIfNeeded(min_height=image_size, min_width=image_size, border_mode=cv2.BORDER_CONSTANT, value=0, mask_value=0),
    A.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]),
    ToTensorV2()
])
val_transform = A.Compose([
    A.LongestMaxSize(max_size=image_size),
    A.PadIfNeeded(min_height=image_size, min_width=image_size, border_mode=cv2.BORDER_CONSTANT, value=0, mask_value=0),
    A.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]),
    ToTensorV2()
])

size_dataset = len(dataset)
val_size = int(size_dataset * 0.1)
train_dataset = WrapperDataset(dataset, transform=train_transform, idxs=list(range(val_size, size_dataset)))
val_dataset = WrapperDataset(dataset, transform=val_transform, idxs=list(range(val_size)))

train_dataloader = DataLoader(train_dataset, batch_size=4, shuffle=True, num_workers=0, drop_last=True)
val_dataloader = DataLoader(val_dataset, batch_size=8, shuffle=False, num_workers=0)

```

Than we should define model (used timm), loss function, optimizer and metrics:

```

model = smp.UnetPlusPlus(
    encoder_name="resnet18",
    encoder_weights="imagenet",
    in_channels=3,
    decoder_use_batchnorm=False,
    classes=2,
)

```

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```
loss_f = DiceLoss(weight=torch.tensor([1, 10]))  
  
optimizer = WrapperOptimizer(optim.Adam, lr=1e-4)  
  
num_epochs = 7  
num_gpus = 1  
  
lr_scheduler = WrapperScheduler(  
    torch.optim.lr_scheduler.OneCycleLR, max_lr=3e-4, pct_start=1 / (num_epochs),  
    total_steps=int(len(train_dataloader)) * num_epochs / num_gpus + 10, div_factor=1e+3,  
    final_div_factor=1e+4,  
    anneal_strategy='cos', interval='step'  
)  
  
class_names = ['background', 'text']  
  
train_metrics = [  
]  
  
val_metrics = [  
    TorchMetric(PixelLevelF1(average='none', num_classes=len(class_names)), class_names),  
]
```

If you need in callbacks, you can use our simple realization. Creating of callbacks looks like:

```
# Logger of outputs (images)  
logger = SegmentationImageLogger(  
    phase='val',  
    max_samples=10,  
    num_classes=2,  
    save_on_disk=True,  
    dir_path='images'  
)  
  
# Cutmix callback  
mixup = Mixup(  
    on_batch=True,  
    p=1.0,  
    domen='segmentation',  
)
```

In finally, we should define learner and trainer, and than run training.

```
learner = SegmentationLearner(  
    model=model,  
    loss=loss_f,  
    optimizer=optimizer,  
    lr_scheduler=lr_scheduler,  
    train_metrics=train_metrics,  
    val_metrics=val_metrics,
```

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```

data_keys=['image'],
target_keys=['mask'],
multilabel=False
)
trainer = pl.Trainer(gpus=num_gpus, callbacks=[logger, mixup, checkpoint_callback], max_
↪epochs=num_epochs)
trainer.fit(learner, train_dataloaders=train_dataloader, val_dataloaders=[val_
↪dataloader])

```

3.3 Simple detection train example

We can observe simple example for detection in pascal dataset using effdet project (<https://github.com/rwightman/efficientdet-pytorch>).

First, you should import common libraries and packages below. (If you don't have some package, than install it).

```

import cv2
from torch.utils.data import Dataset, DataLoader
import torch
import torch.optim as optim
import torch.nn as nn
from pytorch_lightning import Trainer
from pytorch_lightning.callbacks import ModelCheckpoint
from albumentations.augmentations import *
from albumentations.core.composition import *
from albumentations.pytorch.transforms import *
from timm import create_model
import random
from torchmetrics import *
import shutil

```

Than you can import EasyPL packages, as like:

```

from easypl.learners.detection import DetectionLearner
from easypl.metrics.detection import FBetaDetection
from easypl.optimizers import WrapperOptimizer
from easypl.lr_schedulers import WrapperScheduler
from easypl.datasets import CSVDatasetDetection
from easypl.callbacks.loggers import DetectionImageLogger
from easypl.callbacks.mixers import Mixup, Cutmix, Mosaic
from easypl.callbacks.finetuners import OptimizerInitialization
from easypl.utilities.detection import BasePostprocessing

```

Than you should define datasets and dataloaders. You can use this simple example:

```

train_transform = Compose([
    HorizontalFlip(p=0.5),
    LongestMaxSize(max_size=512),
    PadIfNeeded(min_height=512, min_width=512, border_mode=cv2.BORDER_CONSTANT, value=0, ↵
    ↵mask_value=0),
    Normalize(),
]

```

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```

    ToTensorV2(),
], bbox_params=BboxParams(format='pascal_voc', min_visibility=0.1))

val_transform = Compose([
    LongestMaxSize(max_size=512),
    PadIfNeeded(min_height=512, min_width=512, border_mode=cv2.BORDER_CONSTANT, value=0, u
    ↵mask_value=0),
    Normalize(),
    ToTensorV2(),
], bbox_params=BboxParams(format='pascal_voc', min_visibility=0.1))

test_transform = Compose([
    Normalize(),
    ToTensorV2(),
], bbox_params=BboxParams(format='pascal_voc', min_visibility=0.1))

def collate_fn(batch):
    images = torch.stack([_[ 'image'] for _ in batch])
    max_anno_size = max(len(_[ 'annotations'][0]) for _ in batch)
    image_sizes = torch.from_numpy(np.stack([_[ 'image_size'] for _ in batch]))
    image_scales = torch.from_numpy(np.stack([_[ 'image_scale'] for _ in batch]))
    annotations = torch.ones(len(batch), max_anno_size, 5, dtype=torch.float) * -1
    for i in range(len(batch)):
        annotations[i][:len(batch[i][ 'annotations'][0])] = batch[i][ 'annotations'][0]
    return {
        'image': images,
        'annotations': annotations,
        'image_size': image_sizes,
        'image_scale': image_scales
    }

train_dataset = CSVDatasetDetection('../input/pascal/train.csv', image_prefix='../input/
    ↵pascal/train', transform=train_transform, return_label=True)
val_dataset = CSVDatasetDetection('../input/pascal/val.csv', image_prefix='../input/
    ↵pascal/val', transform=val_transform, return_label=True)

train_dataloader = DataLoader(train_dataset, batch_size=16, shuffle=True, pin_
    ↵memory=True, num_workers=2)
val_dataloader = DataLoader(val_dataset, batch_size=16, shuffle=False, pin_memory=True,
    ↵ num_workers=2)

```

Than we should define model (used effnet: <https://github.com/rwightman/efficientdet-pytorch>), loss function, optimizer and metrics:

```

from effdet import EfficientDet, get_efficientdet_config

num_classes = 20

config = get_efficientdet_config('tf_efficientdet_d0')

model = EfficientDet(config, pretrained_backbone=True)

```

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```

model.reset_head(num_classes=num_classes)

from effdet.anchors import Anchors, AnchorLabeler, generate_detections
from effdet.loss import DetectionLoss

class EfficientDetLoss(nn.Module):
    def __init__(self, model, create_labeler=False):
        super().__init__()
        self.model = model
        self.config = model.config # FIXME remove this when we can use @property
        self.num_levels = model.config.num_levels
        self.num_classes = model.config.num_classes
        self.anchors = Anchors.from_config(model.config)
        self.max_detection_points = model.config.max_detection_points
        self.max_det_per_image = model.config.max_det_per_image
        self.soft_nms = model.config.soft_nms
        self.anchor_labeler = AnchorLabeler(self.anchors, self.num_classes, match_
        threshold=0.5)
        self.loss_fn = DetectionLoss(model.config)

    def forward(self, x, target):
        class_out, box_out = x
        cls_targets, box_targets, num_positives = self.anchor_labeler.batch_label_
        anchors(target[:, :, :4], target[:, :, 4])
        loss, class_loss, box_loss = self.loss_fn(class_out, box_out, cls_targets, box_
        targets, num_positives)
        output = {'loss': loss, 'class_loss': class_loss, 'box_loss': box_loss}
        return output

loss_f = EfficientDetLoss(model=model, create_labeler=True)

num_epochs = 5
num_gpus = 1

optimizer = WrapperOptimizer(optim.Adam, lr=1e-4)

lr_scheduler = WrapperScheduler(
    torch.optim.lr_scheduler.OneCycleLR, max_lr=3e-4, pct_start=1 / (num_epochs),
    total_steps=int(len(train_dataloader)) * num_epochs / num_gpus + 10, div_factor=1e+3,
    final_div_factor=1e+4,
    anneal_strategy='cos', interval='step'
)

train_metrics = []
val_metrics = [FBetaDetection([0.5])]
```

If you need callbacks, you can use our simple realization. Creating of callbacks looks like:

```
# Logger of outputs (images)
```

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```
image_logger = DetectionImageLogger(phase='val', num_classes=num_classes)
```

In finally, we should define learner and trainer, and than run training.

```
learner = DetectionLearner(  
    model=model,  
    loss=loss_f,  
    optimizer=optimizer,  
    lr_scheduler=lr_scheduler,  
    train_metrics=train_metrics,  
    val_metrics=val_metrics,  
    data_keys=['image'],  
    target_keys=['annotations'],  
    postprocessing=EfficientdetPostprocessing(model),  
    image_size_key='image_size',  
    image_scale_key='image_scale'  
)  
trainer = Trainer(  
    accelerator='gpu',  
    devices=1,  
    callbacks=[image_logger],  
    max_epochs=num_epochs,  
#    precision=32  
)  
trainer.fit(learner, train_dataloaders=train_dataloader, val_dataloaders=[val_  
    ↴dataloader])
```

Above there are few examples of code, which you can use for yourself projects.

**CHAPTER
FOUR**

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