

Costantino Grana - UNIMORE

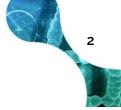
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From ECVL to EDDL



- ECVL allows easy integration and exchange of data with EDDL.
- The ECVL EDDL interfacing is based on two main functions: ImageToTensor and TensorToImage.
- ImageToTensor transforms an ECVL Image in a EDDL Tensor, converting its data to floating point numbers and rearranging its channels to "xy[c/z/o]", which is how EDDL Tensor handles data in color images. Finally, Image data are copied into Tensor data.

```
import pyecvl.ecvl as ecvl
from pyeddl.tensor import Tensor

image = ecvl.ImRead("/mnt/data/winter_school/lena.png")
t = ecvl.ImageToTensor(image)
```





From EDDL to ECVL



- TensorToImage creates a float "xyo" raw Image with none color space, and copies Tensor data into the Image data.
- TensorToView is also available in order to avoid the copy of the data.

```
import pyecvl.ecvl as ecvl
from pyeddl.tensor import Tensor

image = ecvl.ImRead("/mnt/data/winter_school/lena.png")
t = ecvl.ImageToTensor(image)
image_from_tensor = ecvl.TensorToImage(t)
view_from_tensor = ecvl.TensorToView(t)
```





Dataset Format



- A simple and flexible YAML syntax to describe a dataset for the DeepHealth libraries (EDDL/ECVL), in order to provide a unified way for loading data.
- Complete description: <u>DeepHealth-</u> Toolkit-Dataset-Format
- Optional elements: name, description, classes (mandatory in case of a classification task), features (additional information related to each image), split.

```
# Example of DeepHealth toolkit dataset format
# Arrays are always 0 based
# Descriptive string used just for pretty reporting (optional)
name: dataset name
# Descriptive string to document the file (optional)
description: >
  This is an example of long
  text which describes the use of this dataset and
  whatever I want to annotate.
  You can also write multiple paragraphs with the only
  care of indenting them correctly.
# Array of class names (optional)
classes: [class a, class b, class c]
# Array of features names (optional)
features: [feature 1, feature 2, feature 3, feature 4]
# Split (optional) is a dictionary with a custom number of arrays.
# They list the indexes of the images to be used in different phases.
split:
 training: [0, 1]
 validation: [2]
 test: [3]
```



Dataset Format



Mandatory
 element: the list of
 the images with
 their location and
 optionally their
 label (a class or a
 path) and values.

```
images:
# label can be a class name (string)...
# values can be an array with a positional correspondence with the features array...
  - location: image path and name 1
    label: class b
    values: [value 1, null, value 3, null]
# ... or the class index (integer) wrt the classes array
# ... or a dictionary with the name of the feature coupled with its value
  - location: image_path_and_name_2
    label: 2
    values: { feature_1: value_1, feature_3: value_3 }
# In the case of multi class problems, label can be an array of class names (array of strings)...
  - location: image path and name 3
    label: [class a, class c]
# ... or an array of class indexes (array of integers)
  - location: image path and name 4
    label: [0, 2]
# label can be a path (string) to an image in case of a segmentation task
  - location: image path and name 5
    label: path_to_ground_truth_image
# Remember that labels are optional
  - location: image path and name 6
  - location: image path and name 7
```



Dataset



 ECVL also provides a Generator which creates a Dataset file from a directory tree (<u>Dataset-Generator</u>)

```
gcd = ecvl.GenerateClassificationDataset(input_directory)
cls_d = gcd.GetDataset()
cls_d.Dump("classification_dataset.yml")
```

ECVL allows to parse the Dataset YAML file to create a Dataset object.

```
input_dataset = "/mnt/data/winter_school/mnist.yml"
d = ecvl.Dataset(input_dataset)
print("name:", d.name_)
print("classes:", d.classes_)
print("features:", d.features_)
print("n. samples:", len(d.samples_))
```







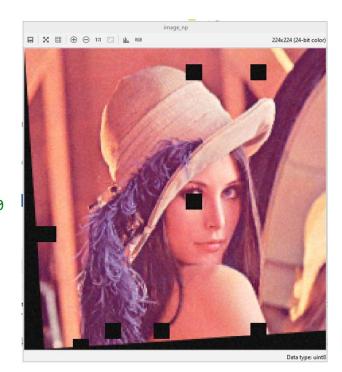
Data Augmentations can be loaded from text...

```
import pyecvl.ecvl as ecvl

image = ecvl.ImRead("/mnt/data/winter_school/lena.png")

AUG_TXT = '''\
SequentialAugmentationContainer
    AugResizeDim dims=(224,224) interp="linear"
    AugRotate angle=[-5,5] center=(0,0) interp="linear"
    AugAdditiveLaplaceNoise std_dev=[0,0.51]
    AugCoarseDropout p=[0,0.05] drop_size=[0.02,0.1] per_channel=0
    AugAdditivePoissonNoise lambda=[0,40]
end
'''

augs_from_text = ecvl.AugmentationFactory.create(AUG_TXT)
augs from text.Apply(image)
```











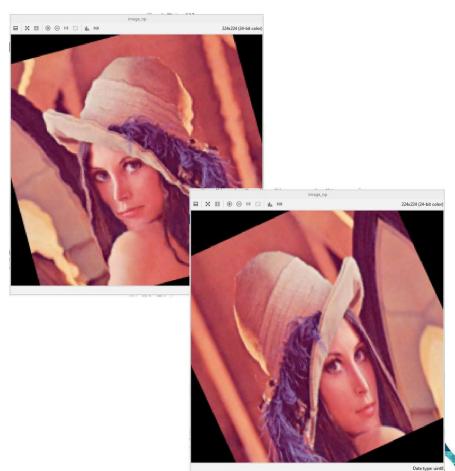
...or directly written in the code

```
import pyecvl.ecvl as ecvl

image = ecvl.ImRead("/mnt/data/winter_school/lena.png")
disp_image = ecvl.Image.empty()

training_augs = ecvl.SequentialAugmentationContainer([
    ecvl.AugResizeDim([224, 224]),
    ecvl.AugMirror(.5),
    ecvl.OneOfAugmentationContainer(
          0.3,
        [ecvl.AugElasticTransform([10, 20], [2, 4]),
          ecvl.AugGridDistortion([2, 5], [-0.3, 0.3]),
          ecvl.AugOpticalDistortion([-0.3, 0.3], [-0.1, 0.1])]
    ),
    ecvl.AugRotate([-30, 30]),
])

training_augs.Apply(image)
```







- The list of all the augmentations is available at https://deephealthproject.github.io/ecvl/master/classecvl_1_1_augmentation.html
- Augmentations must be inserted in a Container, which can be
 - Sequential \rightarrow all augmentations in the container will be performed
 - OneOf \rightarrow only one of the augmentations in the container will be performed, with a given probability
- Containers can be nested one inside the other.

```
training_augs = ecvl.SequentialAugmentationContainer([
    ecvl.AugResizeDim([224, 224]),
    ecvl.AugMirror(.5),
    ecvl.OneOfAugmentationContainer(
        0.3,
        [ecvl.AugElasticTransform([10, 20], [2, 4]),
        ecvl.AugGridDistortion([2, 5], [-0.3, 0.3]),
        ecvl.AugOpticalDistortion([-0.3, 0.3], [-0.1, 0.1])]
    ),
    ecvl.AugRotate([-30, 30]),
])
```







Different augmentations can be applied to each split defined in the Dataset

```
training augs = ecvl.SequentialAugmentationContainer([
    ecvl.AugResizeDim([224, 224]),
    ecvl.AugMirror(.5),
    ecvl.OneOfAugmentationContainer(
        0.3,
        [ecvl.AugElasticTransform([10, 20], [2, 4]),
        ecvl.AugGridDistortion([2, 5], [-0.3, 0.3]),
        ecvl.AugOpticalDistortion([-0.3, 0.3], [-0.1, 0.1])]
    ),
    ecvl.AugRotate([-30, 30]),
])
validation_augs = ecvl.SequentialAugmentationContainer([
    ecvl.AugResizeDim([224, 224]),
1)
test augs = ecvl.SequentialAugmentationContainer([
    ecvl.AugResizeDim([224, 224]),
])
ds_augs = ecvl.DatasetAugmentations([training_augs, validation_augs, test_augs])
```





Deep Learning Dataset



 The Dataset object is extended by DLDataset, which contains information for the deep learning functionalities, specifically for generating data to feed the neural network

- In order to create samples in a non-blocking scenario, *num_workers* threads work in parallel to:
 - loading the images
 - apply augmentations
 - convert them to Tensors
 - push the Tensors to a queue from which the main thread can pull a batch for the network







ProduceImageLabel

These steps are performed by the function *ProduceImageLabel*, which in C++ can be overwritten to customize the creation of the batches. For example, this dataset consists of 3D volumes that we want to push into the queue slice by slice





Generate batches



- With the *Start* method the threads are spawned and begin to fill the queue, until the *Stop* method is called.
- The GetBatch function will retrieve a vector of samples, which contains information like the name of the files in the batch, the Tensor containing the batch of the prepared images and the Tensor containing the corresponding labels.





