

## Exercise 3: Solution

I2DL: Prof. Leal-Taixe

### ImageFolderDataset: \_\_len\_\_()

```
def len (self):
 length = None
 # TODO:
                          #
 # Return the length of the dataset (number of images)
 length = len(self.images)
 #
           FND OF YOUR CODE
                          #
 return length
```

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## ImageFolderDataset: \_\_getitem\_\_()

```
def getitem (self, index):
  data dict = None
   # TODO:
  # create a dict of the data at the given index in your dataset
  # The dict should be of the following format:
  # {"image": <i-th image>, "label": <label of i-th image>}
  # Hints:
    - use load image as numpy() to load an image from a file path
     - If applicable, make sure to apply self.transform to the image:
       image transformed = self.transform(image)
   label = self.labels[index]
  path = self.images[index]
  image = self.load image as numpy(path)
  if self.transform is not None:
     image = self.transform(image)
  data dict = {
     "image": image,
     "label": label,
   END OF YOUR CODE
   return data dict
```

#### Hints:

- self.images[index] contains the full name of the image we want to retrieve (we don't want to keep all images in memory at the same time - we only read them when it's required)
- Self.labels[index] contains the label of the image we want to retrieve
- We only apply the transformation if it's not None

#### RescaleTransform: \_\_call\_\_()

```
def call (self, images):
  # TODO:
  # Rescale the given images:
    - from (self. data min, self. data max)
   to (self.min, self.max)
  images = images - self. data min # normalize to (0, data max-data min)
  images /= (self._data_max - self._data_min) # normalize to (0, 1)
  images *= (self.max - self.min) # norm to (0, target max-target min)
  images += self.min # normalize to (target min, target max)
  FND OF YOUR CODE
  return images
```

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## compute\_image\_mean\_and\_std()

```
def compute image mean and std(images):
   .....
  Calculate the per-channel image mean and standard deviation of given images
  :param images: numpy array of shape NxHxWxC
     (for N images with C channels of spatial size HxW)
  :returns: per-channels mean and std; numpy array of shape C
  mean, std = None, None
  # TODO:
  # Calculate the per-channel mean and standard deviation of the images #
  # Hint: You can use numby to calculate mean and standard deviation
  mean = np.mean(images, axis=(0, 1, 2))
  std = np.std(images, axis=(0, 1, 2))
  END OF YOUR CODE
  return mean, std
```

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#### Dataloader: \_\_len\_\_()

```
def len (self):
  length = None
  # TODO:
  # Return the length of the dataloader
  # Hint: this is the number of batches you can sample from the dataset. #
  # Don't forget to check for drop last!
  if self.drop last:
    length = len(self.dataset) // self.batch size
  else:
    length = int(np.ceil(len(self.dataset) / self.batch size))
  #
                 END OF YOUR CODE
  return length
```

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#### Dataloader: \_\_iter\_\_()

```
def _ iter_ (self):
    # TODO:
   # Define an iterable function that samples batches from the dataset.
   # Each batch should be a dict containing numpy arrays of length
   # batch_size (except for the last batch if drop_last=True)
   # Hints:
    # - np.random.permutation(n) can be used to get a list of all
         numbers from 0 to n-1 in a random order
       - To load data efficiently, you should try to load only those
         samples from the dataset that are needed for the current batch. #
         An easy way to do this is to build a generator with the yield
         keyword, see https://wiki.python.org/moin/Generators
       - Have a look at the "DataLoader" notebook first. This function is #
         supposed to combine the functions:
           - combine_batch_dicts

    batch_to_numpy

           - build batch iterator
         in the notebook.
    def combine_batch_dicts(batch):
       Combines a given batch (list of dicts) to a dict of numpy arrays
       :param batch: batch, list of dicts
           e.g. [{k1: v1, k2: v2, ...}, {k1:, v3, k2: v4, ...}, ...]
       :returns: dict of numpy arrays
           e.g. {k1: [v1, v3, ...], k2: [v2, v4, ...], ...}
       batch dict = {}
       for data dict in batch:
           for key, value in data dict.items():
              if key not in batch dict:
                  batch dict[key] = []
               batch dict[key].append(value)
       return batch dict
```

#### Hints:

We create two helper functions: one for merging a batch of dictionaries as well as a convenient way to convert those dictionaries to numpy arrays which we will then feed to our networks later.

#### Dataloader: \_\_iter\_\_()

```
def batch_to_numpy(batch):
    """Transform all values of the given batch dict to numpy arrays"""
    numpy batch = {}
    for key, value in batch.items():
        numpy batch[key] = np.array(value)
    return numpy batch
if self.shuffle:
    index iterator = iter(np.random.permutation(len(self.dataset)))
else:
    index iterator = iter(range(len(self.dataset)))
batch = []
for index in index iterator:
    batch.append(self.dataset[index])
    if len(batch) == self.batch size:
        vield batch to numpy(combine batch dicts(batch))
        batch = []
if len(batch) > 0 and not self.drop last:
    yield batch to numpy(combine batch dicts(batch))
```

#### Hints:

- Shuffling is implemented here using numpy's random permutation but there are multiple possible solutions
- We iterate over the dataset and use yield to properly invoke our iterator
- Finally we have to check for the last batch size in order to account for "drop\_last".

# Questions? Campuswire ©