

le DigitalPathology library

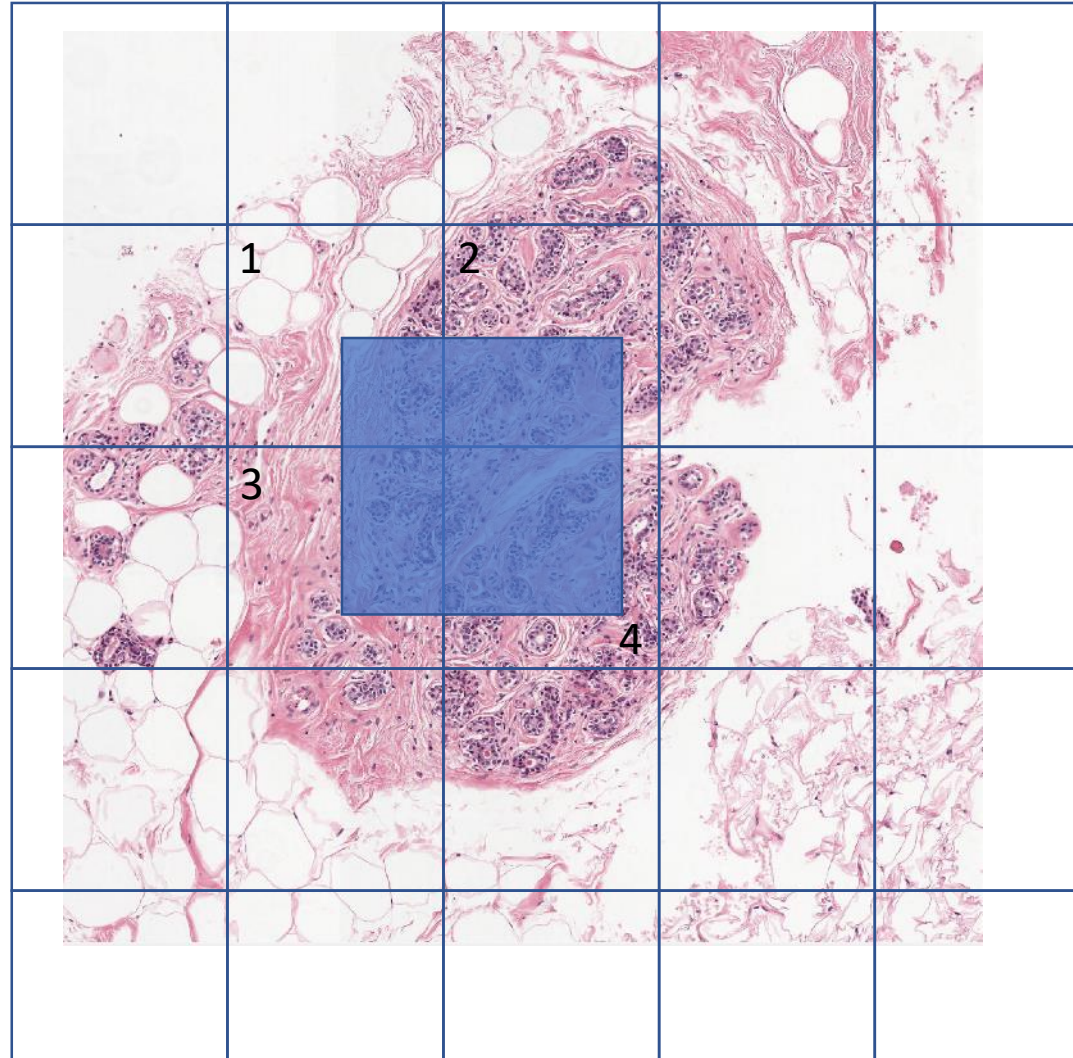
DigitalPathology library

- <https://github.com/DIAGNijmegen/DigitalPathology>
- 110 files
- 27032 lines of code
- Main purpose:
 1. **Data extraction** from multiresolution images
 2. **Saving/loading network models**
 3. **Executing experiments**
 4. **Collection of handy scripts**
 5. **Some examples**
- Documentation: docstring only (*for now*)
- Testing: No (*for now*)

Main components

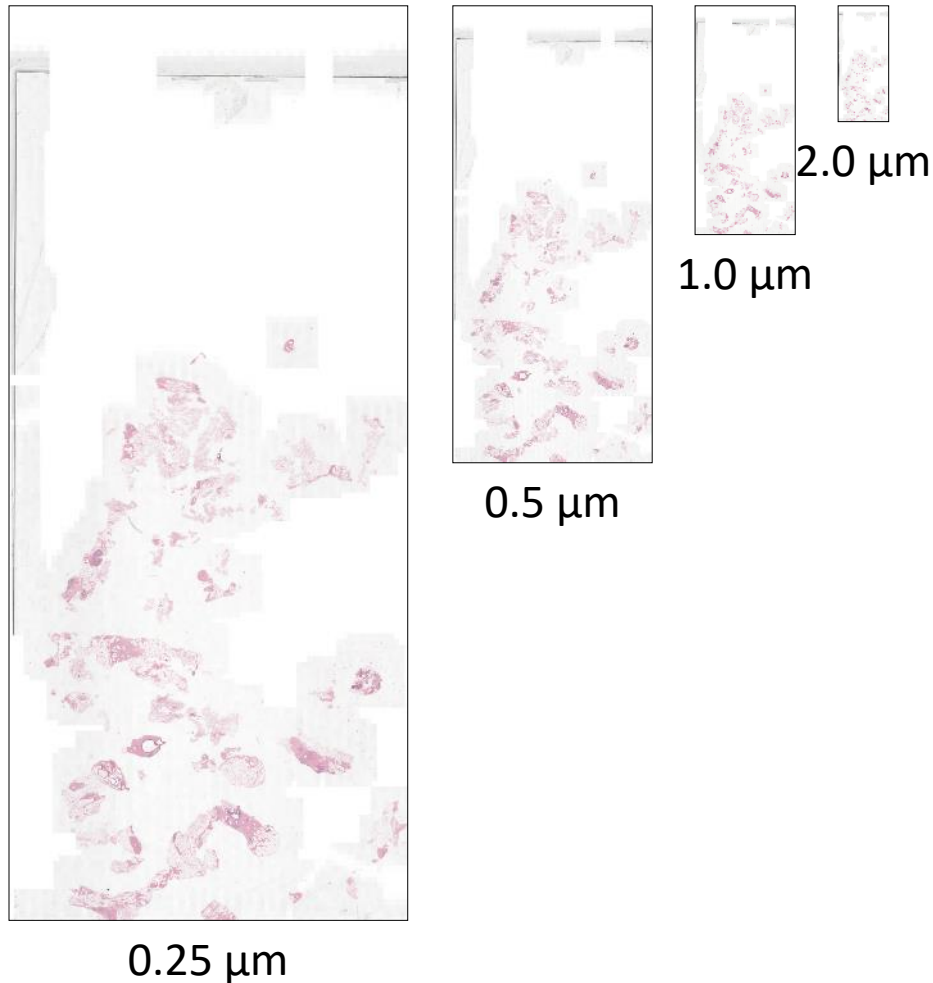
1. **BatchGenerator:** for loading patch of images from collection of multiresolution images.
2. **ModelBase:** reusable representation of networks.
3. **NetworkTrainer:** for executing, logging and saving deep learning experiments.
4. **scripts:** collection of useful scripts.

Multi-resolution images



- Images are stored in tiles
- Tiles are JPEG compressed
- One collection of tiles per level

Multi-resolution images



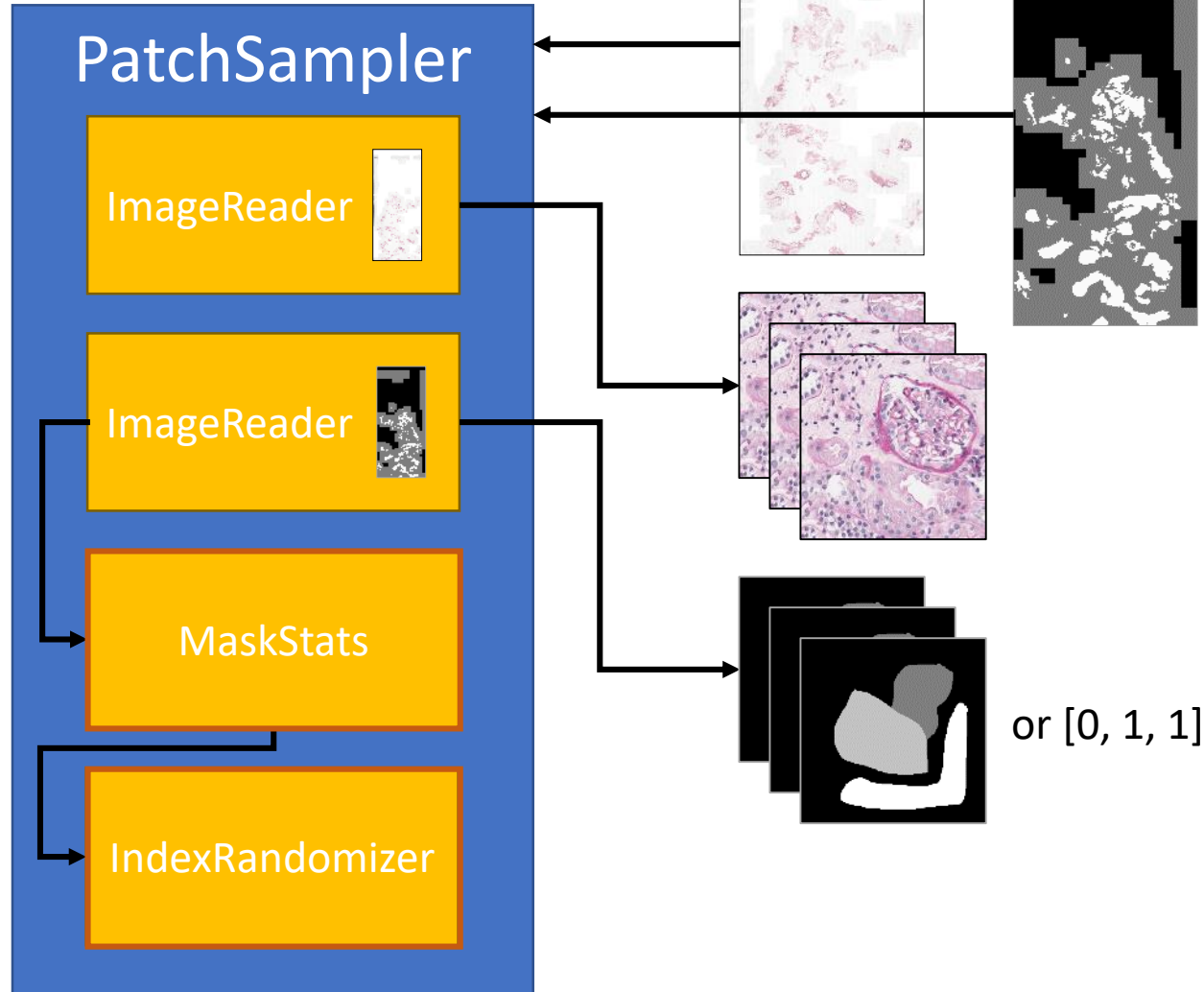
- The levels cover the same spatial area, but with increasing pixel size
- Use pixel spacing instead of zoom
- Levels: arbitrary
- Identify levels by pixel spacing

Multi-resolution images

- multiresolutionimageinterface of ASAP
 - Uses OpenSlide and LibTIFF
 - C++ library
 - Not pythonic
- digitalpathology.image.imagereader.ImageReader
digitalpathology.image.imagewriter.ImageWriter
 - Pythonic
 - Exceptions
 - Returns numpy array of the right size and data type

1. BatchGenerator

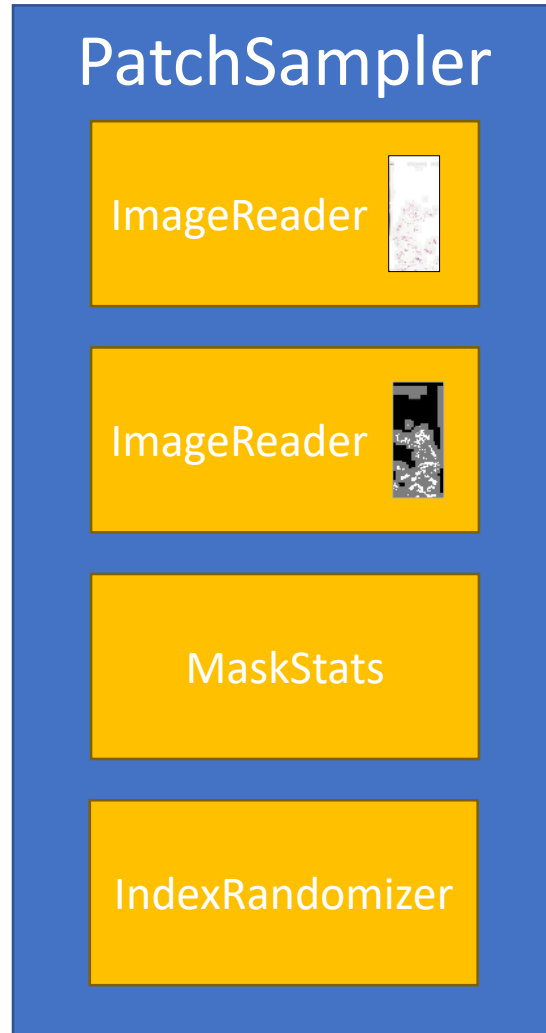
Generating batch of image patches: PatchSampler



```
def sample(self, counts, shapes, label_mode):
```

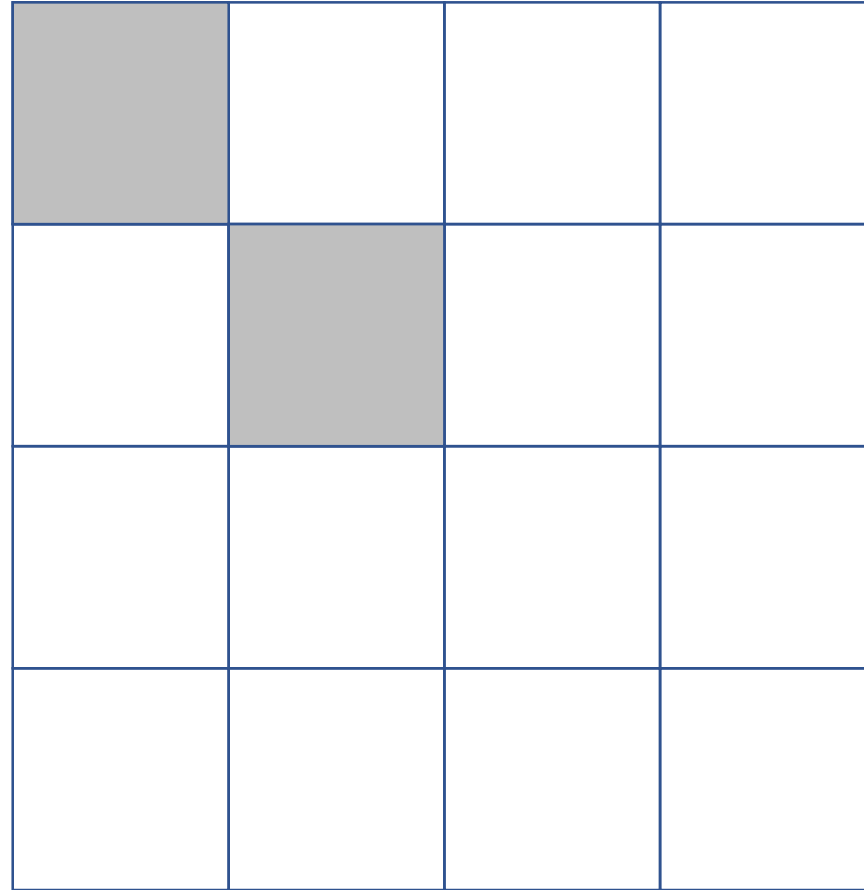
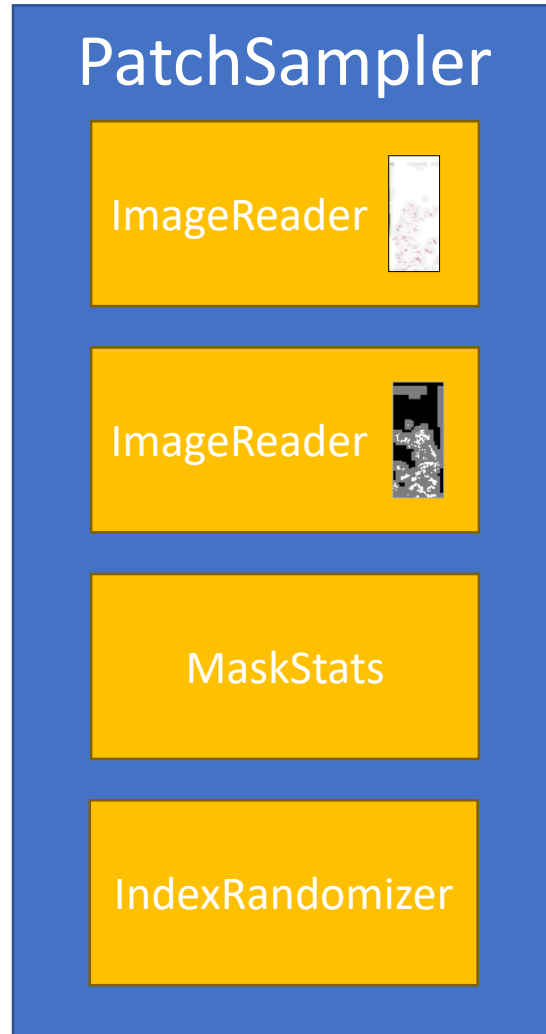
- **counts:** label value to label count mapping
- **shapes:** pixel spacings to patch shape mapping
- **label_mode:** label generation mode
 - central
 - synthesize
 - load

Generating batch of image patches: PatchSampler

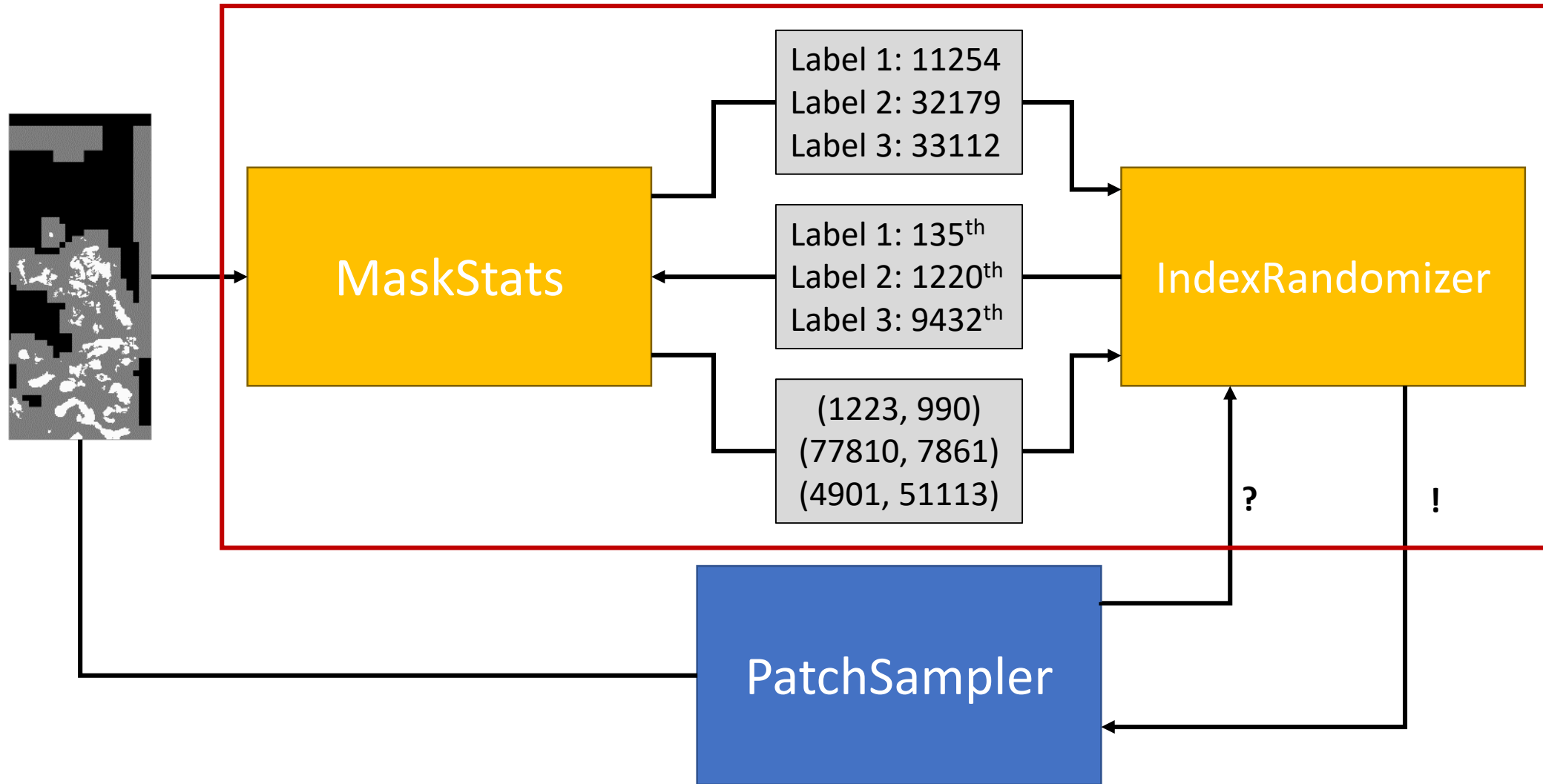


- Mask level is arbitrary
- The lower the level to more memory the MaskStats consume
- Matching of image and mask is based on pixel spacing
- Fallback to image shape matching
- Parsing of mask files is slow, MaskStats can be saved/loaded (.stat files)
- The extracted image patch is centered on the randomized position. Always truncated to the upper left.
- In case of pixel spacing mismatch, the central pixel is used.

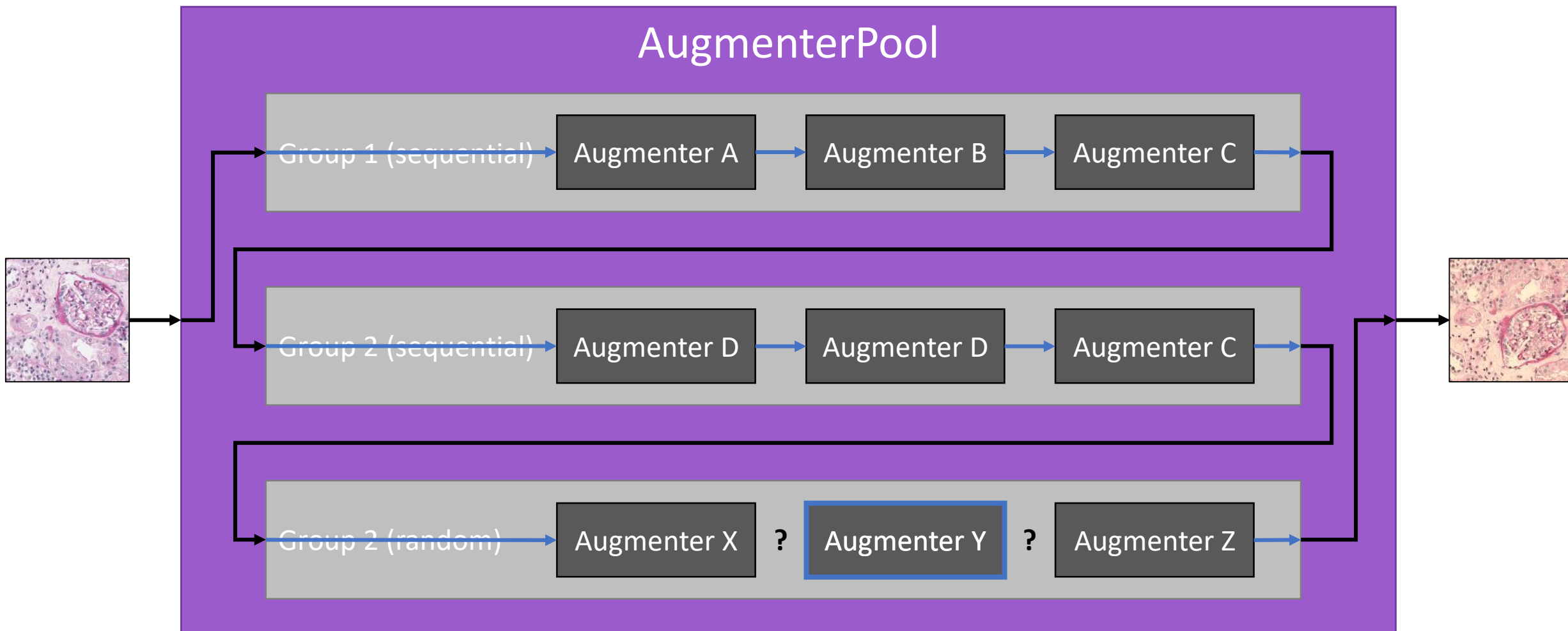
Generating batch of image patches: PatchSampler



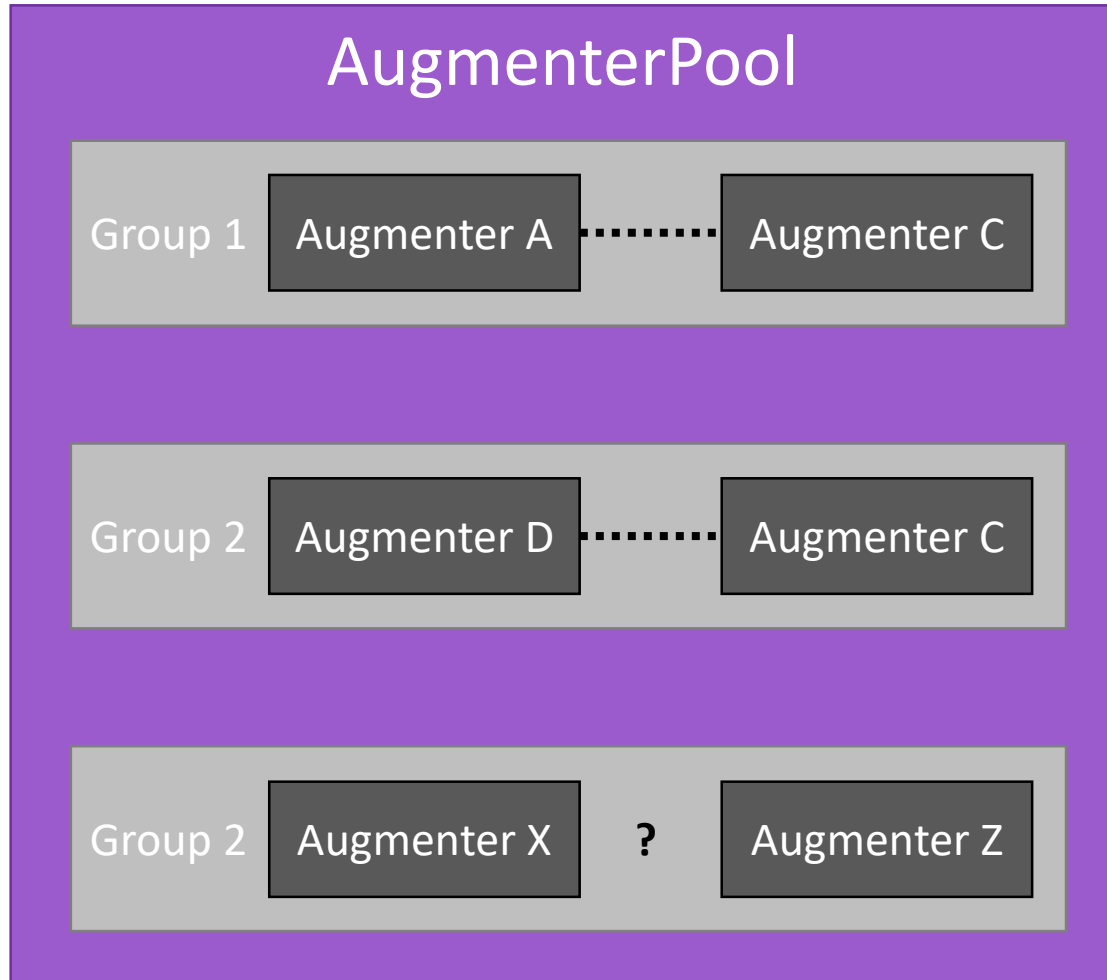
Generating batch of image patches: MaskStats and IndexRandomizer



Generating batch of image patches: AugmenterPool



Generating batch of image patches: AugmenterPool



```
def appendgroup(self, group, randomized):
```

- Append an empty group to the pool
- Groups are executed in order
- Groups can be sequential or random

```
def appendaugmentor(self, augmenter, group, ratio=0.0):
```

- Append an augmenter to an existing group
- Set probability for random selection

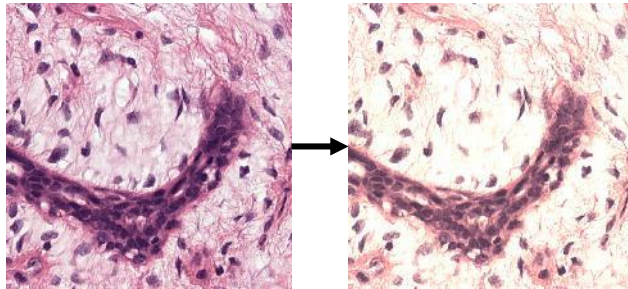
```
def process(self, patches, shapes=None, randomize=True):
```

→ Process a patch collection (as extracted by PatchSampler)

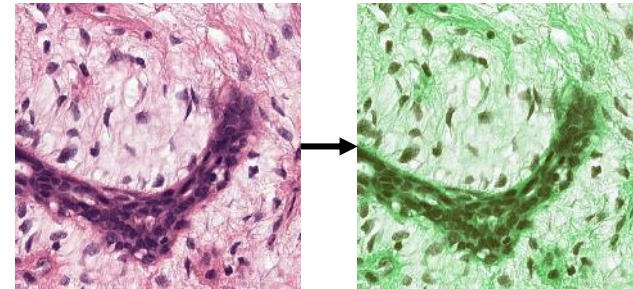
```
def randomize(self):
```

→ Randomize parameters for each augmenter object

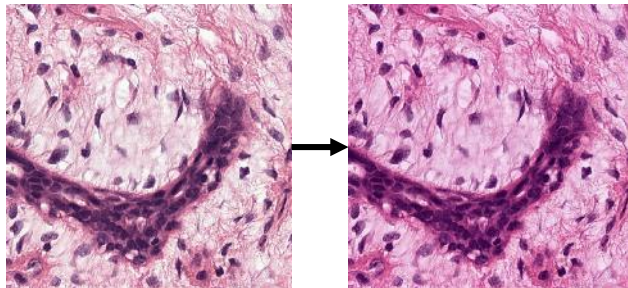
Generating batch of image patches: HedColorAugmenter, HsbColorAugmenter



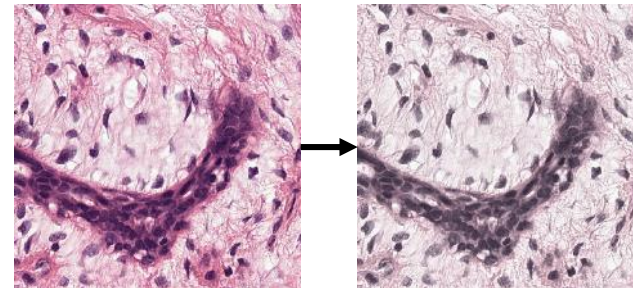
HedColorAugmenter
Haematoxylin adjustment



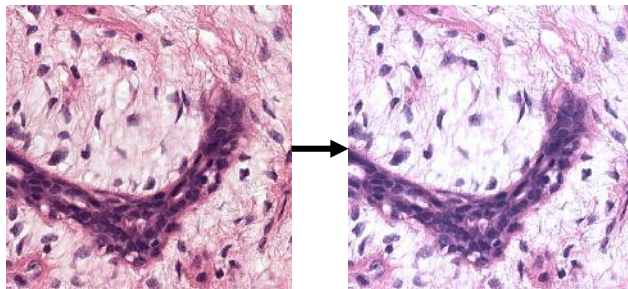
HsbColorAugmenter
Hue adjustment



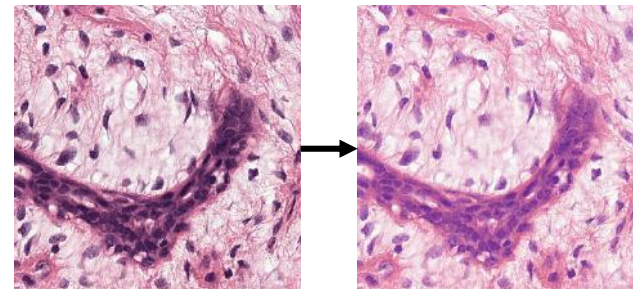
HedColorAugmenter
Eusin adjustment



HsbColorAugmenter
Saturation adjustment



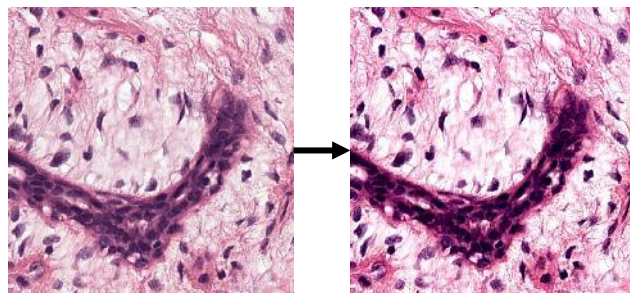
HedColorAugmenter
DAB adjustment



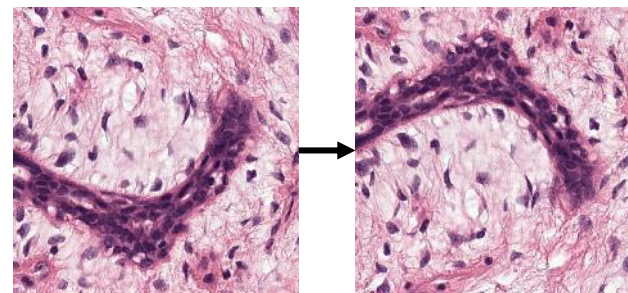
HsbColorAugmenter
Brightness adjustment

Generating batch of image patches:

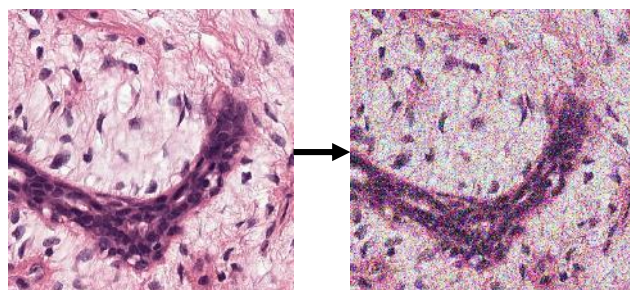
ContrastAugmenter, AdditiveGaussianNoiseAugmenter, GaussianBlurAugmenter, FlipAugmenter, Rotate90Augmenter, ScalingAugmenter



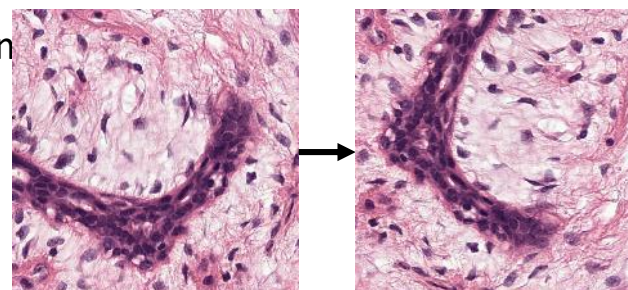
ContrastAugmenter
Contrast adjustment



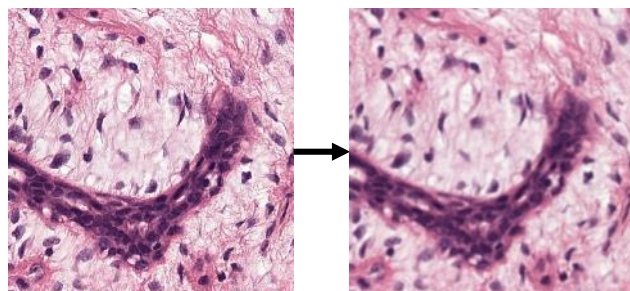
FlipAugmenter
Flipping



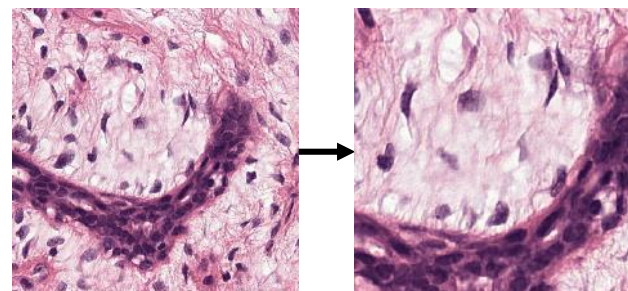
AdditiveGaussianNoiseAugmenter
Gaussian noise addition



Rotate90Augmenter
Rotate by 90°, 180°, 270°

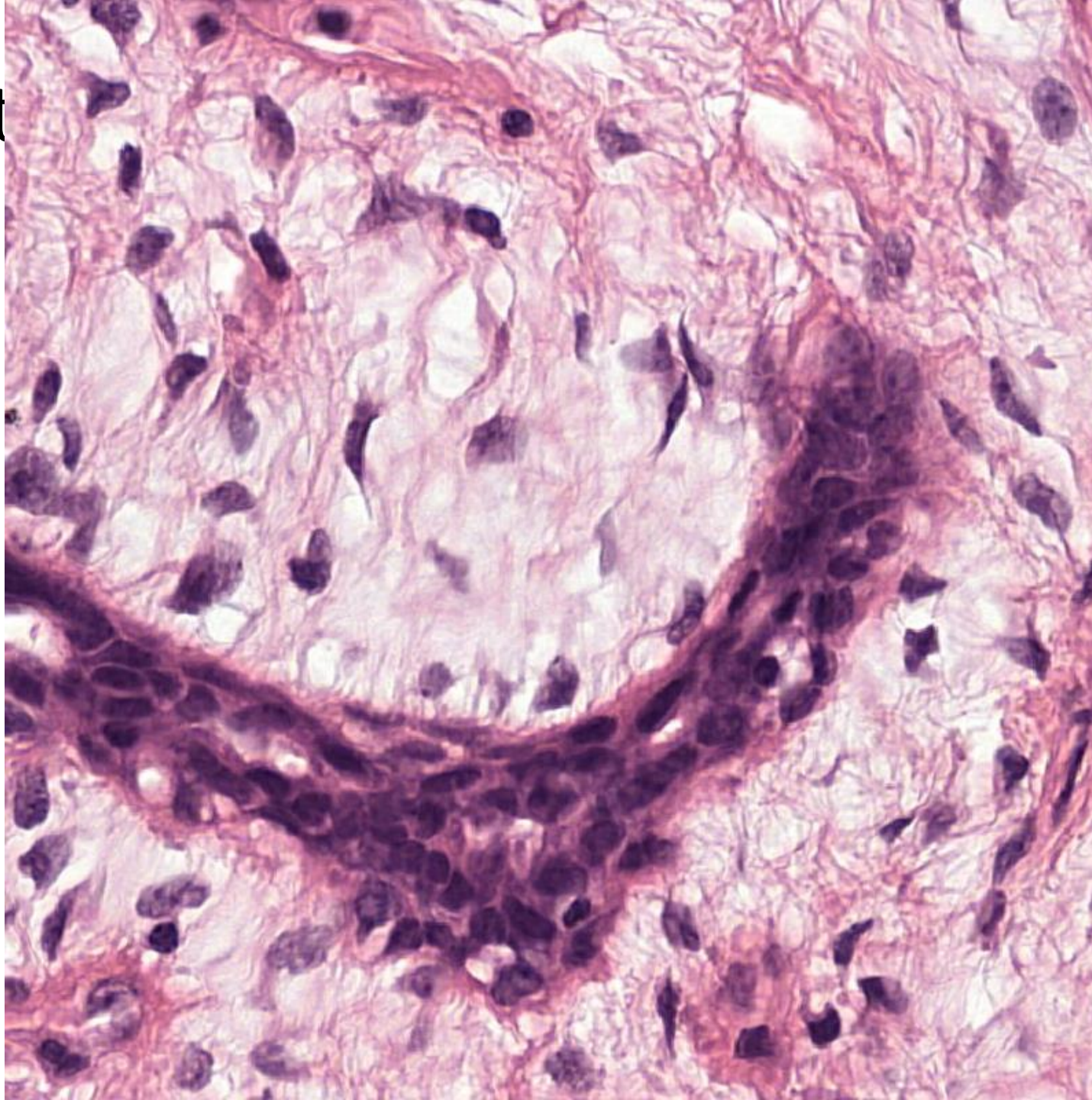


GaussianBlurAugmenter
Gaussian blurring

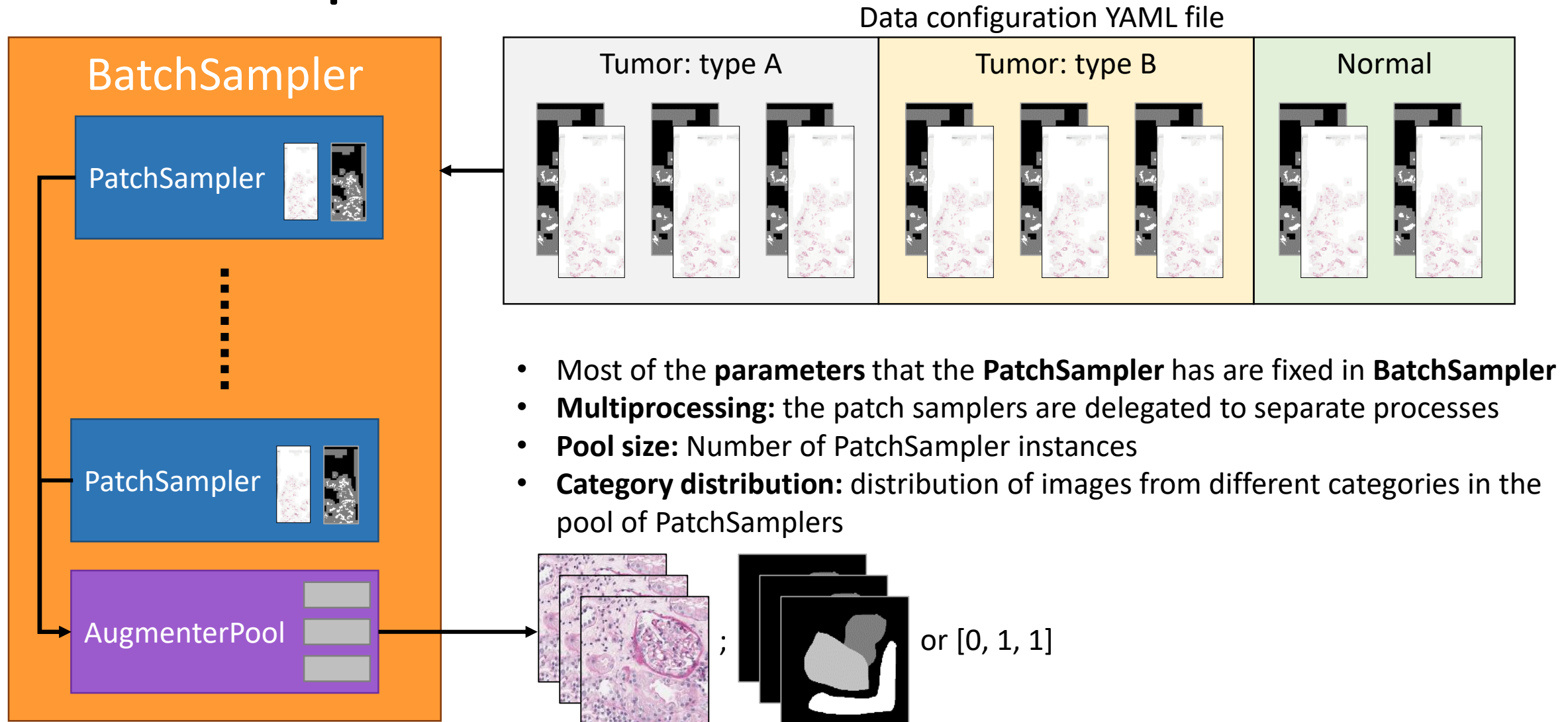


ScalingAugmenter
Scaling

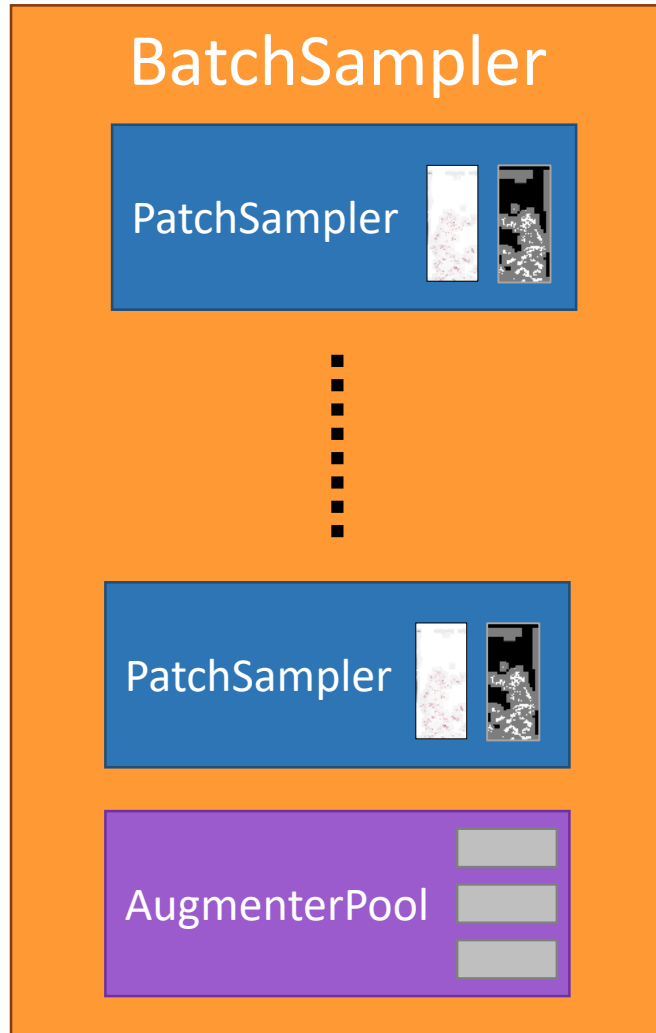
Generat
ElasticA



Generating batch of image patches: BatchSampler



Generating batch of image patches: BatchSampler



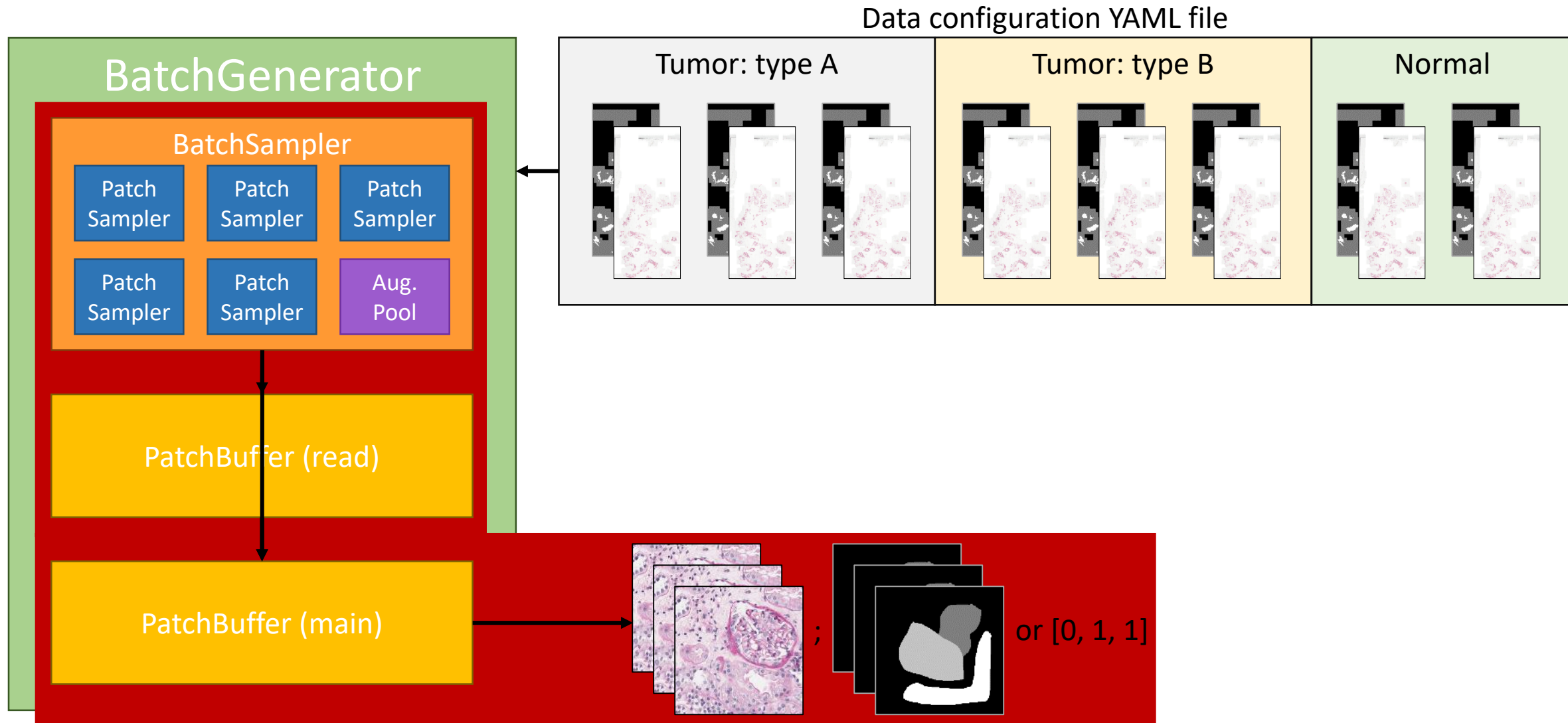
```
def step(self):
```

- After instantiation the BatchSampler is in **invalid state**: it has no PatchSampler instances
- The step function **randomizes the images** to open based on the distribution of categories
- Opens a collection images, the number is limited by the **pool size**

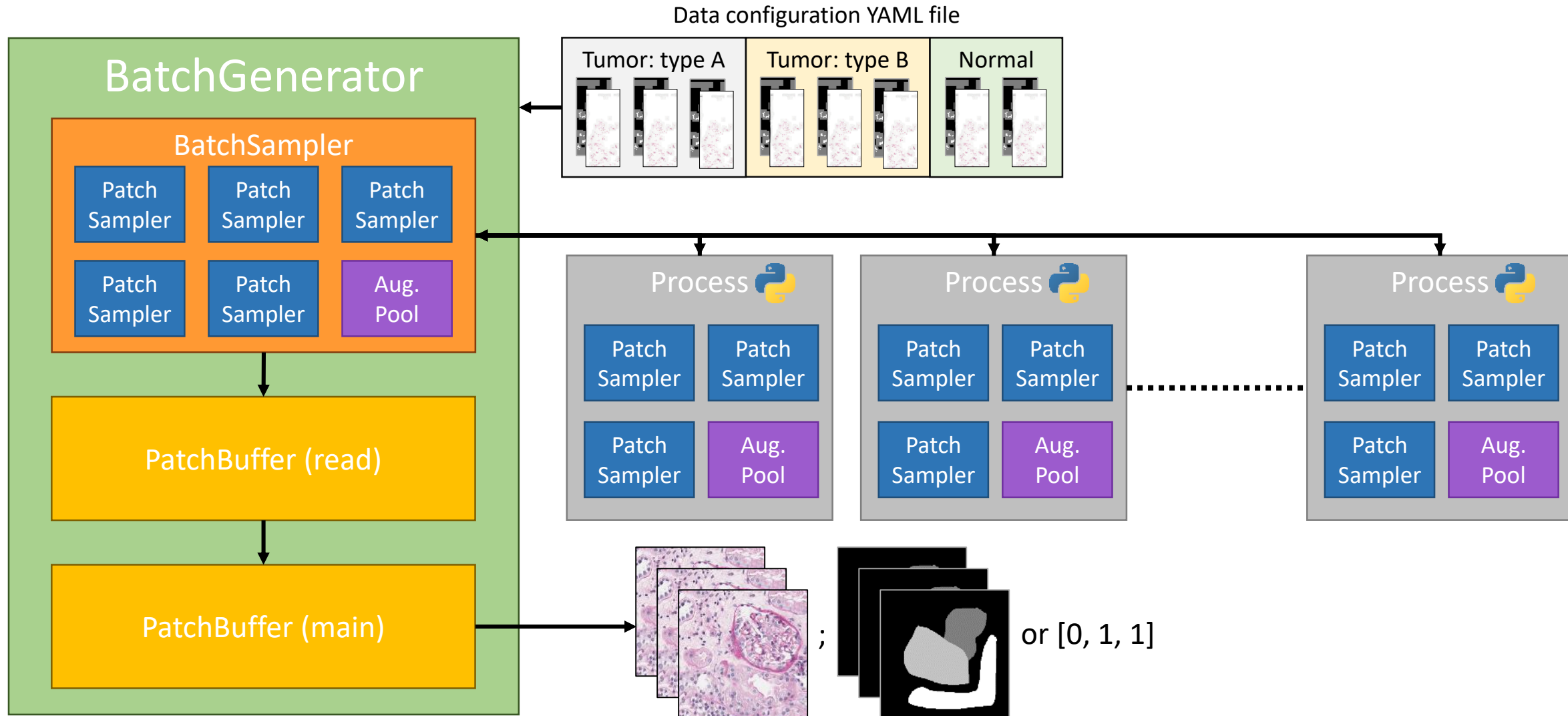
```
def batch(self, batch_size):
```

- **Load a batch of patches** from the pool of PatchSamplers
- All other parameters for the PatchSampler are fixed by the constructor.

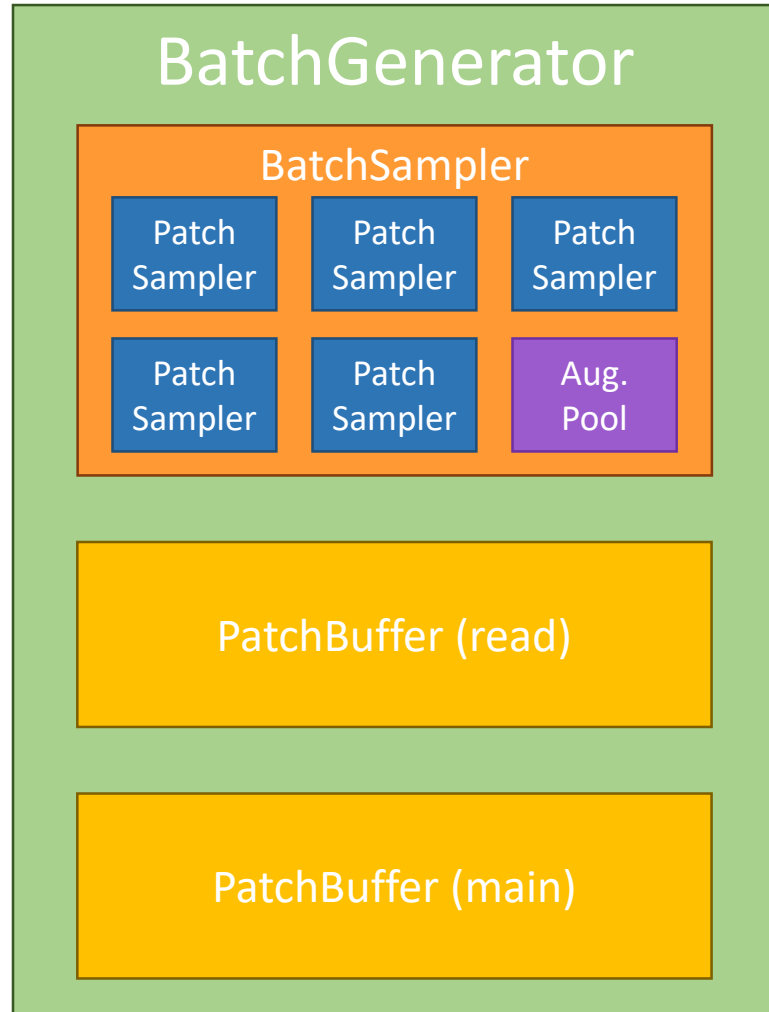
Generating batch of image patches: BatchGenerator



Generating batch of image patches: BatchGenerator



Generating batch of image patches: BatchGenerator



`def start(self):` → Instantiate BatchSampler

`def stop(self):` → Shut down everything

`def step(self):` → BatchSampler:step, open a randomized set of images

`def batch(self, batch_size):` → Get a batch from the main buffer

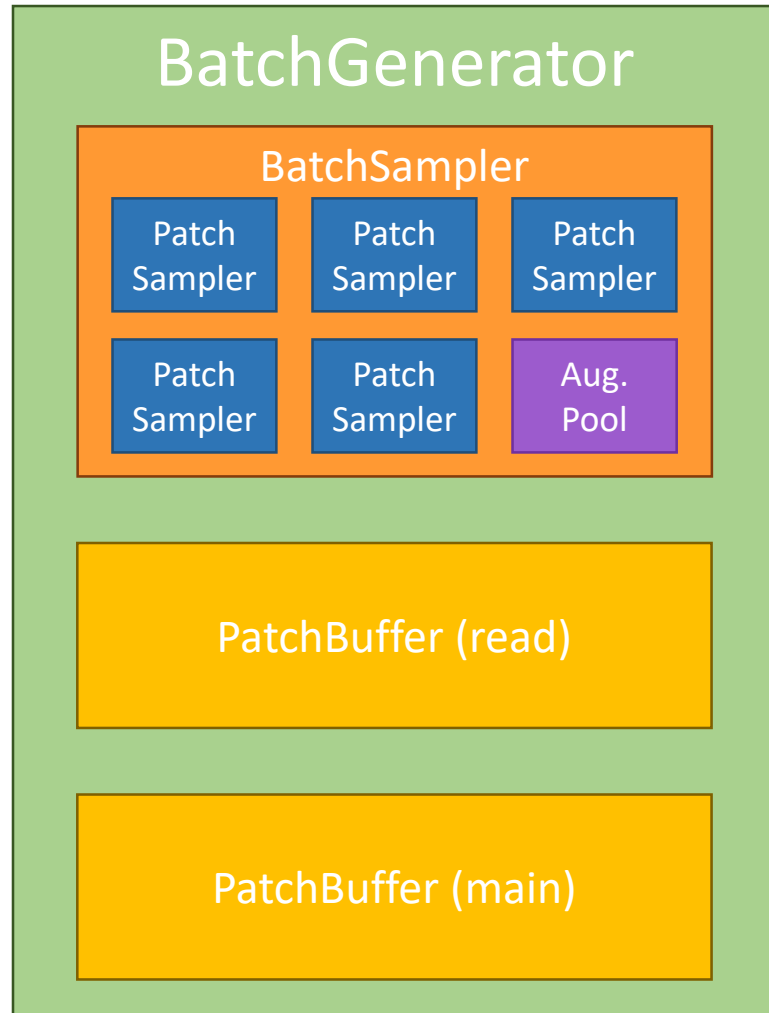
`def fill(self):` → Fill up the **main** buffer directly

`def load(self, batch_size=0):` → Load patches to the **read** buffer

`def transfer(self, batch_size=0, difficult_threshold=0.0):`
↳ Transfer patches from the **read** to the **main** buffer

`def wait(self):` → Wait for the previous async job to finish

Generating batch of image patches: BatchGenerator



```
# Initialize the generators.
self.__training_gen.start()
self.__training_gen.step()
self.__training_gen.wait()

# Loading initial data batch generators.
self.__training_gen.fill()
self.__training_gen.wait()

# Execute epochs.
for epoch_index in range(epoch_count):
    # Start loading in the background.
    self.__training_gen.load(batch_size=0)
    self.__training_gen.ping()

    # Extract patches from the buffer.
    for iter_index in range(iter_count):
        batch, indices = self.__training_gen.batch(batch_size=batch_size)
        # Use the patches for training...

    # Wait the loading to finish.
    self.__training_gen.wait()
    self.__training_gen.transfer(batch_size=0, difficult_threshold=0.0)

    # Step sources.
    self.__training_gen.step()
    self.__training_gen.wait()

# Terminate the batch generator objects to let the program shut down in a clean way.
self.__training_gen.stop()
```

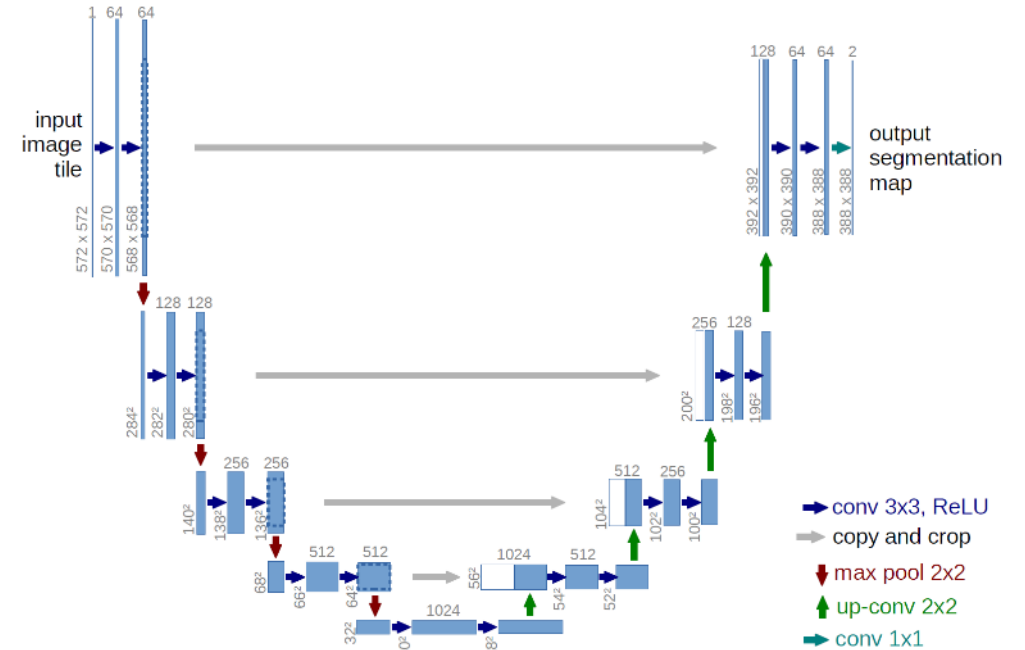
2. ModelBase

Base class for network models: **ModelBase**

- Base class for network models
- Wrapper around libraries, currently:
 - Theano/Lasagna (probably already broken)
 - Keras + TensorFlow
- In the future:
 - TensorFlow
 - PyThorc
- The library provides a standard way to **build, save, load, train and use** the networks.
- Provides an easy way to configure existing network architectures.

Base class for network models: ModelBase

```
class UNet(KerasModelBase):  
    """U-Net model in Keras"""  
  
    def configure(self,  
                  input_shape,  
                  depth,  
                  classes,  
                  branching_factor,  
                  batch_norm,  
                  dropout_count,  
                  dropout_prob,  
                  l2_lambda,  
                  padding,  
                  residual,  
                  downsampling,  
                  upsampling,  
                  channels_first):  
        """  
        Save the network configuration parameters.
```



Base class for network models: **ModelBase**

`def build(self):` → Build a network

`def save(self, file_path):` → Save the network to file

`def load(self, file):` → Load the network from file

`def update(self, x, y, sample_weight=None, class_weight=None, *args, **kwargs):` → Update (forward + backward pass)

`def validate(self, x, y, sample_weight=None, *args, **kwargs):` → Validate (forward pass)

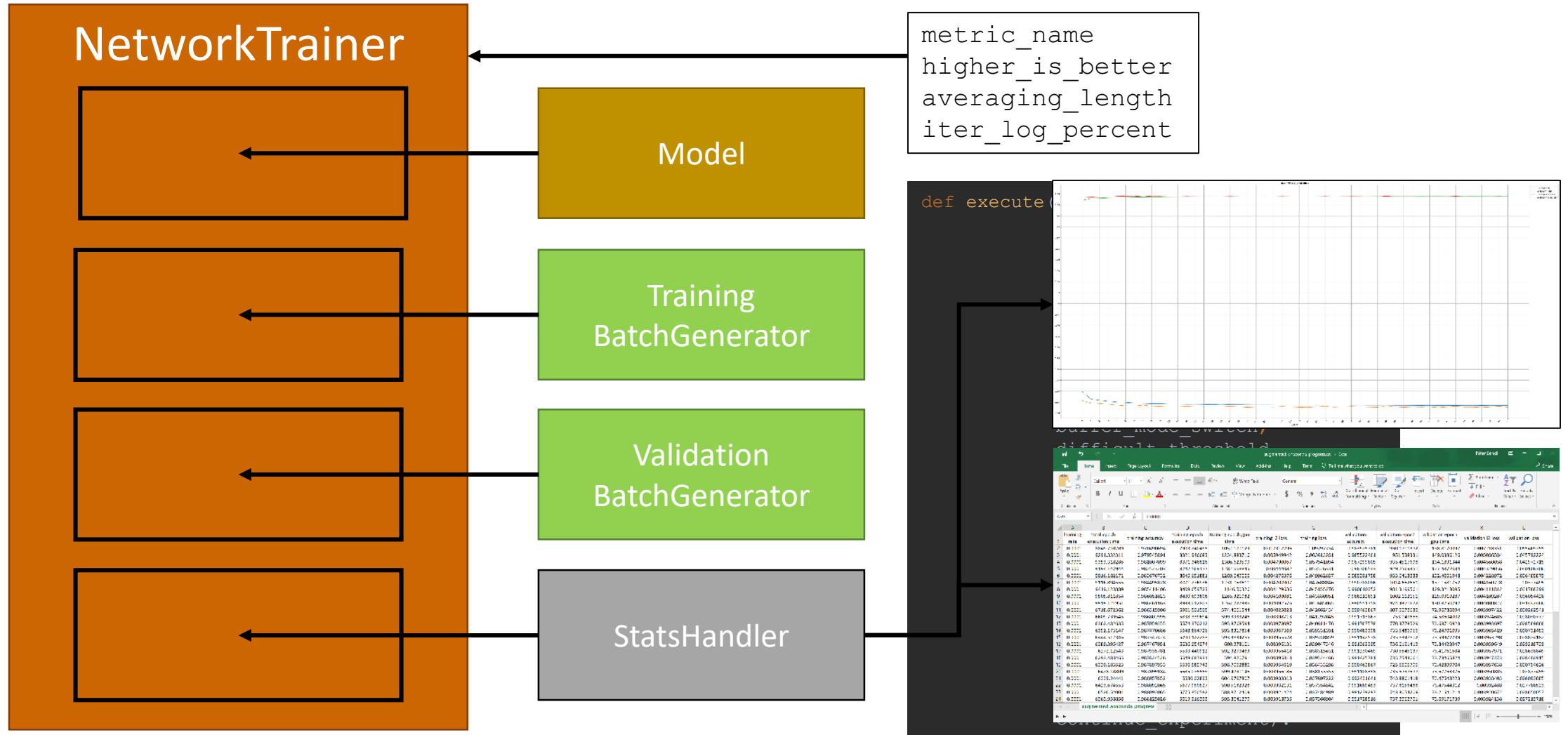
`def predict(self, x, *args, **kwargs):` → Predict (forward pass, predictions only)

`def getreconstructioninformation(self, input_shape=None):` → Calculate the scale factor and padding to reconstruct the input shape.

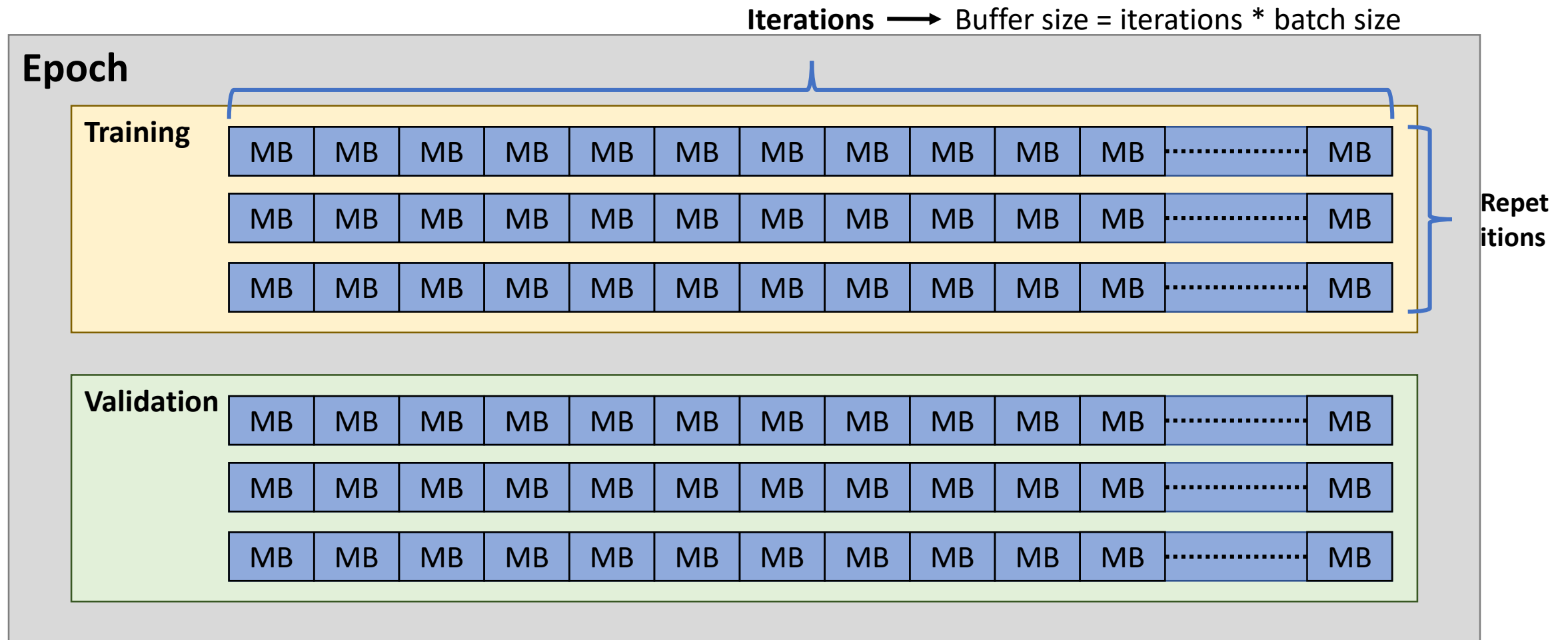
`def _restoremodelparameters(self, parameters):` → Helper function after loading

3. NetworkTrainer

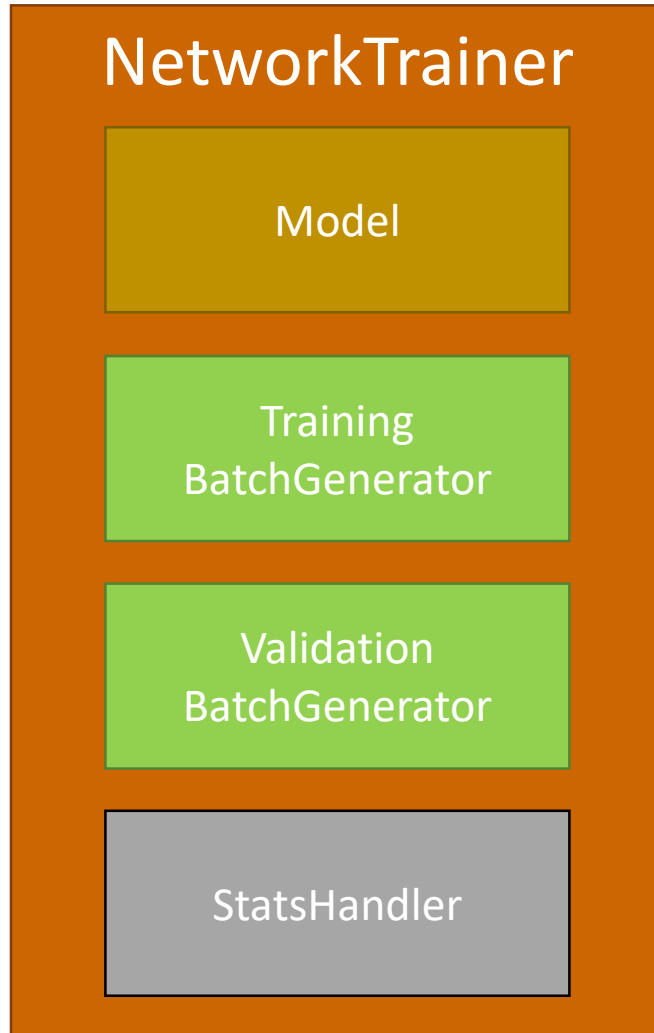
Training networks: NetworkTrainer



Training networks: NetworkTrainer

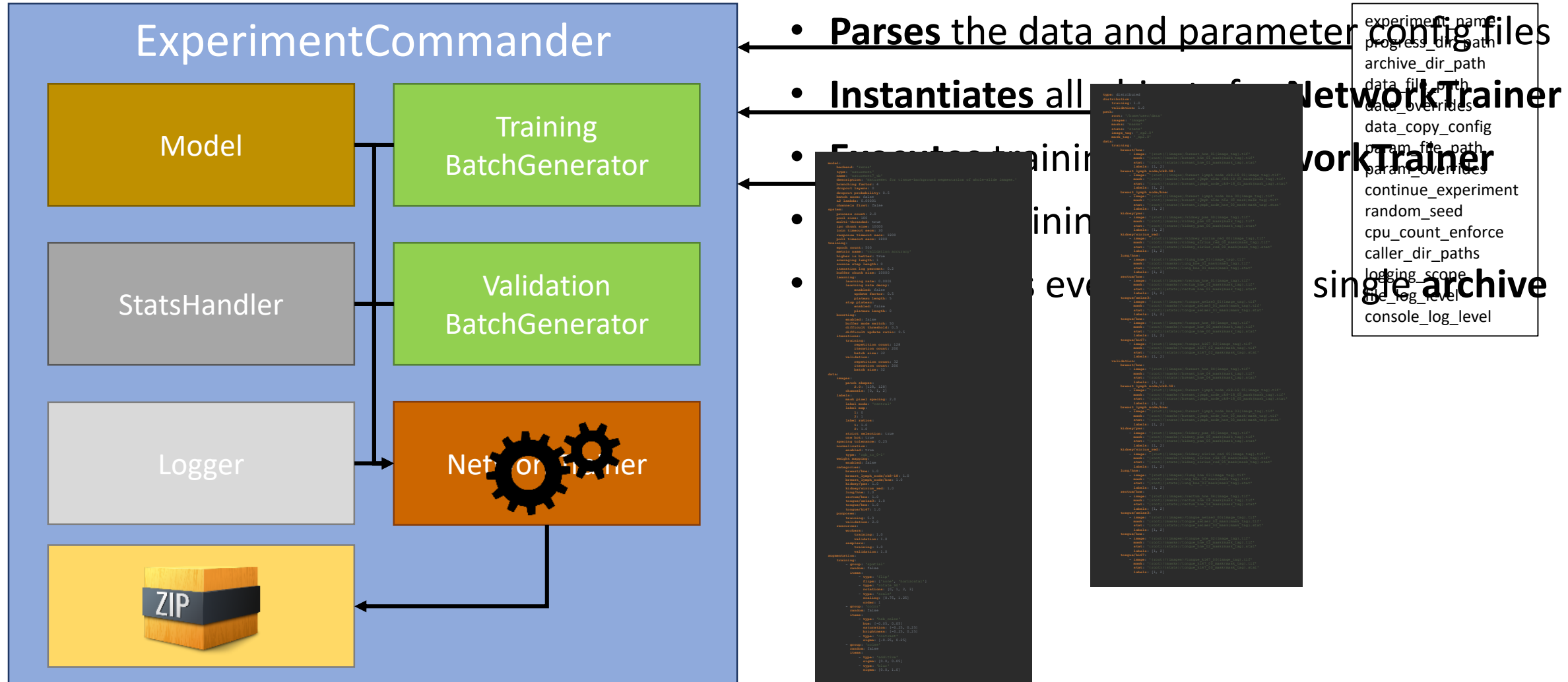


Training networks: NetworkTrainer



- **Executes** the experiment
- Saves the **best** and the **last** network **model**
- Saves the **metrics** table and the plot
- Saves the **status** of the training
- Can pick up an **continue** an experiment

Training networks: ExperimentCommander



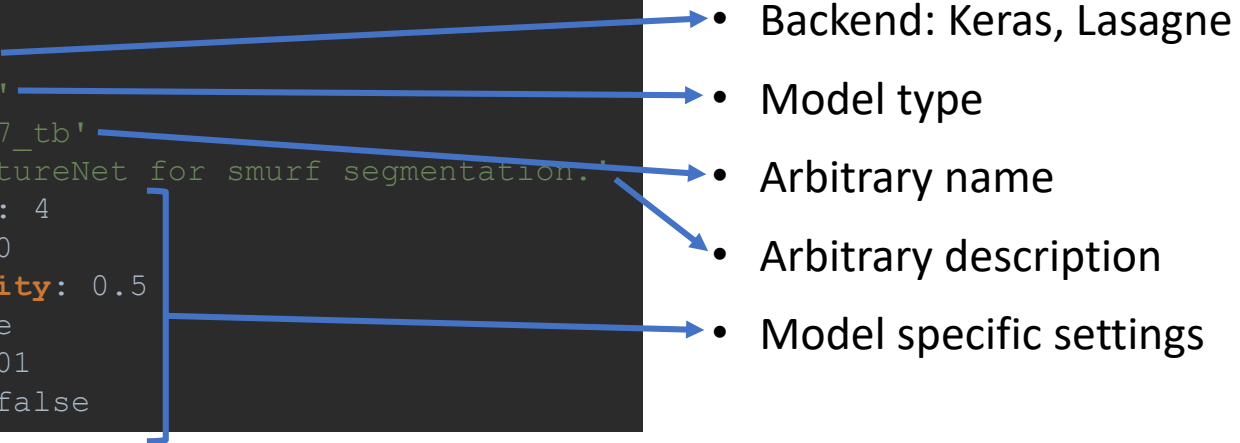
Training networks: data config

```
type: distributed
distribution:
  training: 1.0
  validation: 1.0
path:
  root: '/home/user/data'
  images: 'images'
  masks: 'masks'
  stats: 'stats'
  image_tag: '_sp2.0'
  mask_tag: '_sp2.0'
data:
  training:
    breast/hne:
      - image: '{root}/{images}/breast_hne_01{image_tag}.tif'
        mask: '{root}/{masks}/breast_hne_01_mask{mask_tag}.tif'
        stat: '{root}/{stats}/breast_hne_01_mask{mask_tag}.stat'
        labels: [1, 2]
    kidney/pas:
      - image: '{root}/{images}/kidney_pas_00{image_tag}.tif'
        mask: '{root}/{masks}/kidney_pas_00_mask{mask_tag}.tif'
        stat: '{root}/{stats}/kidney_pas_00_mask{mask_tag}.stat'
        labels: [1, 2]
```

- Distributed, list (e.g. training, validation)
- Distribution ratio
- Replacement strings
- Data part
- Purpose: “training”
- Class: “breast/hne”
- Image, mask, stat files
- Labels in mask

Training networks: parameters: model

```
model:
  backend: 'keras'
  type: 'naturenet'
  name: 'naturenet7_tb'
  description: 'NatureNet for smurf segmentation.'
  branching factor: 4
  dropout layers: 0
  dropout probability: 0.5
  batch norm: false
  L2 lambda: 0.00001
  channels first: false
```



The diagram illustrates the mapping between model parameters and their descriptions. Blue arrows point from specific parameters in the code block to their corresponding descriptions in the list:

- Backend: Keras, Lasagne (points to `backend: 'keras'`)
- Model type (points to `type: 'naturenet'`)
- Arbitrary name (points to `name: 'naturenet7_tb'`)
- Arbitrary description (points to `description: 'NatureNet for smurf segmentation.'`)
- Model specific settings (points to a bracket grouping `branching factor: 4`, `dropout layers: 0`, `dropout probability: 0.5`, `batch norm: false`, `L2 lambda: 0.00001`, and `channels first: false`)

Training networks: parameters: system

```
system:
  process count: 2.0
  pool size: 16
  multi-threaded: true
  ipc chunk size: 10000
  join timeout secs: 30
  response timeout secs: 600
  poll timeout secs: 1800
```

- Worker process count
 - Float: multiplier for CPU count
 - Integer: fixed number
 - Use **--cpu** option with float
- Pool size: Max number of PatchSampler objects at any time
- Multi threaded: BatchSampler runs on separate thread
- Chunk size for transferring patches between processes
- Timeouts to prevent the docker hanging

```

training:
  epoch count: 4
  metric name: 'validation accuracy'
  higher is better: true
  averaging length: 1
  source step length: 0
  iteration log percent: 0.2
  buffer chunk size: 10000
  learning:
    learning rate: 0.0005
    learning rate decay:
      enabled: false
      update factor: 0.5
      plateau length: 5
    stop plateau:
      enabled: false
      plateau length: 0
  boosting:
    enabled: false
    buffer mode switch: 2
    difficult threshold: 0.5
    difficult update ratio: 0.5
  iterations:
    training:
      repetition count: 4
      iteration count: 100
      batch size: 32
    validation:
      repetition count: 2
      iteration count: 100
      batch size: 32

```

ks: parameters: training

- Epoch count
- Metric name, higher is better
- Average the last X epoch of the metric
- Step the sources in every X epoch
- Log (and print) the progress in every 20% of the iterations
- Chunk size for transferring the read buffer to the main
- Learning rate, and learning rate decay settings
- Stop criteria
- Boosting settings
- Training iteration and repetition counts and batch sizes
- Validation iteration and repetition counts and batch sizes

```

data:
  images:
    patch shapes:
      2.0: [128, 128]
    channels: [0, 1, 2]
  labels:
    mask pixel spacing: 2.0
    label mode: 'central'
    label map:
      1: 0
      2: 1
    label ratios:
      1: 1.0
      2: 1.0
    strict selection: true
    one hot: true
  spacing tolerance: 0.25
  normalization:
    enabled: true
    type: 'rgb_to_0-1'
  weight mapping:
    enabled: false
  categories:
    breast/hne: 1.0
    kidney/pas: 1.0
    tongue/ki67: 1.0
  purposes:
    training: 1.0
    validation: 1.0
  resources:
    workers:
      training: 1.0
      validation: 1.0
    samplers:
      training: 1.0
      validation: 1.0

```

works: parameters: data

- Patch shapes, per pixel spacing (same location)
- Channels to extract from image
- Mask processing pixel spacing
- Label mode: **central**, **synthesize**, or **load**
- Label mapping from mask values to network values
- Label ratios in buffer (~batch)
- Check if all labels can be load, at all time
- Convert labels to **one hot** representation
- Spacing tolerance (percentage)
- Normalization of image values:
 - `rgb_to_0-1`: `[0, 255] → [0.0, 1.0]`
 - `rgb`: `[0, 255] → [?, ?]`
 - `general`: `[?, ?] → [?, ?]`
- Weight mapping (for U-Net like networks)
- **Distribution of patches** in the buffer from categories
- **Distribution of images** for different purposes. The input data YAML file should be either be list type, or the ratios should match the ratios in the YAML file.
- **Distribution of resources** (workers: processes, sampler: PatchSampler objects) for different purposes.

parameters: augmentation

```
augmentation:
  training:
    - group: 'spatial'
      random: false
      items:
        - type: 'flip'
          flips: ['none', 'horizontal']
        - type: 'rotate_90'
          rotations: [0, 1, 2, 3]
        - type: 'scale'
          scaling: [0.75, 1.25]
          order: 1
    - group: 'color'
      random: true
      items:
        - type: 'hsb_color'
          hue: [-0.05, 0.05]
          saturation: [-0.25, 0.25]
          brightness: [-0.25, 0.25]
          ratio: 1.0
        - type: 'contrast'
          sigma: [-0.25, 0.25]
          ratio: 1.0
    - group: 'noise'
      random: false
      items:
        - type: 'additive'
          sigma: [0.0, 0.05]
        - type: 'blur'
          sigma: [0.0, 1.0]
```

- Separate settings for training and (possibly) validation
- Group with arbitrary name and the type:
 - Sequential (random = false)
 - Random (random = true)
- Augmenter list
- Augmenter type and its parameters

4. Scripts

Scripts 2/1

```
argument_parser.add_argument('-i', '--input',
argument_parser.add_argument('-l', '--input_spacing',
argument_parser.add_argument('-m', '--mask',
argument_parser.add_argument('-o', '--output',
argument_parser.add_argument('-g', '--output_spacing',
argument_parser.add_argument('-n', '--model',
argument_parser.add_argument('-v', '--interval',
argument_parser.add_argument('-p', '--patch_size',
argument_parser.add_argument('-c', '--output_class',
argument_parser.add_argument('-u', '--num_classes',
argument_parser.add_argument('-a', '--channels',
argument_parser.add_argument('-f', '--confidence',
argument_parser.add_argument('-b', '--diagonal_threshold',
argument_parser.add_argument('-t', '--tolerance',
argument_parser.add_argument('-z', '--order',
argument_parser.add_argument('-r', '--purposes',
argument_parser.add_argument('-e', '--categories',
argument_parser.add_argument('-d', '--input_override',
argument_parser.add_argument('-cp', '--copy_directory',
argument_parser.add_argument('-wp', '--work_directory',
argument_parser.add_argument('-nr', '--normalizer',
argument_parser.add_argument('-tr', '--source_range',
argument_parser.add_argument('-sr', '--target_range',
argument_parser.add_argument('-s', '--soft',
argument_parser.add_argument('-q', '--quantize',
argument_parser.add_argument('-x', '--no_mask',
argument_parser.add_argument('-cf', '--channels_first',
argument_parser.add_argument('-ki', '--keep_intermediates',
argument_parser.add_argument('-kc', '--keep_copies',
argument_parser.add_argument('-w', '--overwrite',

required=True, type=str,
required=True, type=float,
required=False, type=str, default=None,
required=True, type=str,
required=False, type=float, default=None,
required=True, type=str,
required=False, type=str, default=None,
required=False, type=int, default=1024,
required=False, type=int, default=-1,
required=False, type=int, default=-1,
required=False, type=str, default='(0, 1, 2)',
required=False, type=float, default=0.0,
required=False, type=float, default=0.0,
required=False, type=float, default=0.25,
required=False, type=int, default=0,
required=False, type=str, default=None,
required=False, type=str, default=None,
required=False, type=str, default=None,
required=False, type=str, default=None,
required=False, type=str, default='rgb_to_0-1',
required=False, type=str, default='[]',
required=False, type=str, default='[]',

help='image file or directory path to process')
help='input image pixel spacing (micrometer)')
help='mask file or directory path to use')
help='output file or directory path')
help='target output pixel spacing (micrometer)')
help='network model')
help='quantization interval file path')
help='processing patch size')
help='output class')
help='number of output classes of the network')
help='list of input channel indices')
help='network confidence for thresholding')
help='region size filter (micrometer)')
help='pixel spacing tolerance (percentage)')
help='interpolation order')
help='list of purpose identifiers to use from da
help='list of category identifiers to use from d
help='data config source overrides')
help='data copy target directory path')
help='work directory path')
help='normalizer to use for preprocessing')
help='source-range for normalizer')
help='target-range for normalizer')
help='get soft classification from the network')
help='quantize result to [0, 255]')
help='do not use mask entries from the data conf
help='channels first')
help='keep intermediate files')
help='keep copied image files')
help='overwrite existing results')
```

Scripts 2/2

- `imageinfo.py`: Print image shape and spacing information.
- `normalizemasks.py`: Map mask values to different values.
- `preprocessmasks.py`: Calculate STAT files from mask images.
- `saveimagesatlevel.py`: Save TIFF image at a given level (still multiresolution).
- `savemrimageasimage.py`: Save TIFF image at a given level as PNG.
- `setspacing.py`: Set the pixel spacing on level 0 for a TIFF image (recompression).
- `summarizelogs.py`: Get the best values from the logs of ExperimentCommander.
- `thresholdimage.py`: Low-threshold a TIFF image at a given value.
- `trainnetwork.py`: Callable wrapper around ExperimentCommander.
- `updatenetwork.py`: Update an old network file format to the newest one.
- `zoomimage.py`: Zoom image.

Developing DigitalPathology

Developing DigitalPathology

- GitHub: <https://github.com/DIAGNijmegen/DigitalPathology>
- Latest stable branch: **master**
- Latest development branch: **develop**
- To add something to DigitalPathology:
 1. Create a branch from develop, called
 - `feature/feature_name`
 - `experiment/feature_name`
 2. Commit changes
 3. Create pull request
- Follow **PEP 8** recommendations **and** [DigitalPathology/documents/style.md](#)

Configuration advices

Configuration advices

1. Look for examples in DigitalPathology/examples
2. The example annotation configuration is **not a good one!**
3. The resource configuration is a **good one**
 - *system: process count*
 - *system: pool size* (set it for your project!)
 - *data: resources*
4. Configuring the resource use:
 - There is only **either** the training **or** the validation BatchGenerator active at once
 - Use *process count: 2.0* setting to use all the available CPUs
 - Use *--cpu=X* command line option for the trainnetwork.py script on the cluster
 - Each worker process has its own python interpreter (~200Mb of memory)
 - Test how much memory a single image/mask pair consumes when loaded to a PatchSampler
 - Log in to the executing machine and use *docker stats* command to monitor the memory usage

Errors

Errors

1. Look for the error reporting in the logs
2. Most of the errors has their unique exception
3. The stack trace is reported with the exception
4. If a process disappears during training it is most likely due to **insufficient memory**
5. Always find the first error in the logs
6. Try remote debugging if you cannot reproduce the error locally

Errors

```
[training] Exception raised: 'FailedSourceSelectionError("Not all [1, 2, 3] labels can be sampled from the current source selection:
['c:\TissueBackground\masks\tongue_aelae3_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\kidney_pas_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\tongue_ki67_02_mask_sp2.0.tif',
'c:\TissueBackground\masks\breast_lymph_node_ck8-18_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\breast_lymph_node_hne_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\tongue_hne_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\rectum_hne_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\lung_hne_01_mask_sp2.0.tif'].",))'.
```

```
Trace: 'batchsamplerdaemon.py:130:batchsampler_daemon_loop/batchsampler.py:938:step/batchsampler.py:757: __samplesources'.
Shutdown is imminent
```

```
[training] Unknown response from batch sampler:
{'response': 'error',
 'exception': 'FailedSourceSelectionError("Not all [1, 2, 3] labels can be sampled from the current source selection:
['c:\TissueBackground\masks\tongue_aelae3_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\kidney_pas_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\tongue_ki67_02_mask_sp2.0.tif',
'c:\TissueBackground\masks\breast_lymph_node_ck8-18_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\breast_lymph_node_hne_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\tongue_hne_00_mask_sp2.0.tif',
'c:\TissueBackground\masks\rectum_hne_01_mask_sp2.0.tif',
'c:\TissueBackground\masks\lung_hne_01_mask_sp2.0.tif'].",))',
 'trace':
'batchsamplerdaemon.py:130:batchsampler_daemon_loop/batchsampler.py:938:step/batchsampler.py:757: __samplesources',
 'command': {'command': 'step'},
 'tid': 15396}
```

Questions?