
SLAYER PyTorch

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SLAYER PYTORCH MAIN

This module bundles various SLAYER PyTorch modules as a single package. The complete module can be imported as

```
>>> import slayerSNN as snn
```

- The **spike-layer** module will be available as `snn.layer`.
- The **yaml-parameter** module will be available as `snn.params`.
- The **spike-loss** module will be available as `snn.loss`.
- The **spike-classifier** module will be available as `snn.predict`.
- The **spike-IO** module will be available as `snn.io`.

SLAYER MODULE

class `slayer.spikeLayer` (*neuronDesc, simulationDesc, fullRefKernel=False*)

This class defines the main engine of SLAYER. It provides necessary functions for describing a SNN layer. The input to output connection can be fully-connected, convolutional, or aggregation (pool). It also defines the psp operation and spiking mechanism of a spiking neuron in the layer.

Important: It assumes all the tensors that are being processed are 5 dimensional. (Batch, Channels, Height, Width, Time) or NCHWT format. The user must make sure that an input of correct dimension is supplied.

*If the layer does not have spatial dimension, the neurons can be distributed along either Channel, Height or Width dimension where Channel * Height * Width is equal to number of neurons. It is recommended (for speed reasons) to define the neurons in Channels dimension and make Height and Width dimension one.*

Arguments:

- **neuronDesc (slayer.yamlParams):** spiking neuron descriptor.

```
neuron:
  type:      SRMALPHA  # neuron type
  theta:     10         # neuron threshold
  tauSr:     10.0       # neuron time constant
  tauRef:    1.0        # neuron refractory time constant
  scaleRef:  2          # neuron refractory response scaling (relative_
↪to theta)
  tauRho:    1          # spike function derivative time constant_
↪(relative to theta)
  scaleRho:  1          # spike function derivative scale factor
```

- **simulationDesc (slayer.yamlParams):** simulation descriptor

```
simulation:
  Ts: 1.0
  tSample: 300
  nSample: 12
```

- **fullRefKernel** (bool, optional): high resolution refractory kernel (the user shall not use it in practice)

Usage:

```
>>> snnLayer = slayer.spikeLayer(neuronDesc, simulationDesc)
```

conv (*inChannels, outChannels, kernelSize, stride=1, padding=0, dilation=1, groups=1*)

Returns a function that can be called to apply conv layer mapping to input tensor per time instance. It behaves same as `torch.nn.conv2d` applied for each time instance.

Arguments:

- `inChannels (int)`: number of channels in input
- `outChannels (int)`: number of channels produced by convolution
- `kernelSize (int or tuple of two ints)`: size of the convolving kernel
- `stride (int or tuple of two ints)`: stride of the convolution. Default: 1
- `padding (int or tuple of two ints)`: zero-padding added to both sides of the input. Default: 0
- `dilation (int or tuple of two ints)`: spacing between kernel elements. Default: 1
- `groups (int or tuple of two ints)`: number of blocked connections from input channels to output channels. Default: 1

The parameters `kernelSize`, `stride`, `padding`, `dilation` can either be:

- a single `int` – in which case the same value is used for the height and width dimension
- a `tuple` of two `ints` – in which case, the first `int` is used for the height dimension, and the second `int` for the width dimension

Usage:

```
>>> conv = snnLayer.conv(2, 32, 5) # 32C5 filter
>>> output = conv(input)           # must have 2 channels
```

dense (*inFeatures*, *outFeatures*)

Returns a function that can be called to apply dense layer mapping to input tensor per time instance. It behaves similar to `torch.nn.Linear` applied for each time instance.

Arguments:

- **inFeatures (int, tuple of two ints, tuple of three ints)**: dimension of input features (Width, Height, Channel) that represents the number of input neurons.
- **outFeatures (int)**: number of output neurons.

Usage:

```
>>> fcl = snnLayer.dense(2048, 512)           # takes (N, 2048, 1, 1, T) tensor
>>> fcl = snnLayer.dense((128, 128, 2), 512) # takes (N, 2, 128, 128, T)
↳ tensor
>>> output = fcl(input)                       # output will be (N, 512, 1, 1,
↳ T) tensor
```

pool (*kernelSize*, *stride=None*, *padding=0*, *dilation=1*)

Returns a function that can be called to apply pool layer mapping to input tensor per time instance. It behaves same as `torch.nn.sum` pooling applied for each time instance.

Arguments:

- `kernelSize (int or tuple of two ints)`: the size of the window to pool over
- `stride (int or tuple of two ints)`: stride of the window. Default: `kernelSize`
- `padding (int or tuple of two ints)`: implicit zero padding to be added on both sides. Default: 0
- `dilation (int or tuple of two ints)`: a parameter that controls the stride of elements in the window. Default: 1

The parameters `kernelSize`, `stride`, `padding`, `dilation` can either be:

- a single `int` – in which case the same value is used for the height and width dimension

- a tuple of two ints – in which case, the first *int* is used for the height dimension, and the second *int* for the width dimension

Usage:

```
>>> pool = snnLayer.pool(4) # 4x4 pooling
>>> output = pool(input)
```

psp (*spike*)

Applies psp filtering to spikes. The output tensor dimension is same as input.

Arguments:

- spike: input spike tensor.

Usage:

```
>>> filteredSpike = snnLayer.psp(spike)
```

spike (*membranePotential*)

Applies spike function and refractory response. The output tensor dimension is same as input. *membranePotential* will reflect spike and refractory behaviour as well.

Arguments:

- membranePotential: subthreshold membrane potential.

Usage:

```
>>> outSpike = snnLayer.spike(membranePotential)
```

class `slayer.yamlParams` (*parameter_file_path*)

This class reads yaml parameter file and allows dictionary like access to the members.

Usage:

```
import slayerSNN as snn
netParams = snn.params('path_to_yaml_file') # OR
netParams = slayer.yamlParams('path_to_yaml_file')

netParams['training']['learning']['etaW'] = 0.01
print('Simulation step size', netParams['simulation']['Ts'])
print('Spiking neuron time constant', netParams['neuron']['tauSr'])
print('Spiking neuron threshold', netParams['neuron']['theta'])

netParams.save('filename.yaml')
```


SPIKE CLASSIFIER

class spikeClassifier.**spikeClassifier**

It provides classification modules for SNNs. All the functions it supplies are static and can be called without making an instance of the class.

static getClass (*spike*)

Returns the predicted class label. It assigns single class for the SNN output for the whole simulation runtime.

Usage:

```
>>> predictedClass = spikeClassifier.getClass(spikeOut)
```


SPIKE LOSS

class spikeLoss.**spikeLoss** (*networkDescriptor*, *slayerClass*=<class 'slayer.spikeLayer'>)

This class defines different spike based loss modules that can be used to optimize the SNN.

Usage:

```
>>> error = spikeLoss.spikeLoss(networkDescriptor)
>>> error = spikeLoss.spikeLoss(errorDescriptor, neuronDesc, simulationDesc)
```

numSpikes (*spikeOut*, *desiredClass*, *numSpikesScale*=1)

Calculates spike loss based on number of spikes within a *target region*. The *target region* and *desired spike count* is specified in `error.errorDescriptor['tgtSpikeRegion']` Any spikes outside the target region are penalized with `error.spikeTime` loss..

$$e(t) = \begin{cases} \frac{actualSpikeCount - desiredSpikeCount}{targetRegionLength} & \text{for } t \in targetRegion \\ (\varepsilon * (output - desired))(t) & \text{otherwise} \end{cases}$$
$$E = \int_0^T e(t)^2 dt$$

Arguments:

- `spikeOut` (torch.tensor): spike tensor
- `desiredClass` (torch.tensor): one-hot encoded desired class tensor. Time dimension should be 1 and rest of the tensor dimensions should be same as `spikeOut`.

Usage:

```
>>> loss = error.numSpikes(spikeOut, target)
```

spikeTime (*spikeOut*, *spikeDesired*)

Calculates spike loss based on spike time. The loss is similar to van Rossum distance between output and desired spike train.

$$E = \int_0^T (\varepsilon * (output - desired))(t)^2 dt$$

Arguments:

- `spikeOut` (torch.tensor): spike tensor
- `spikeDesired` (torch.tensor): desired spike tensor

Usage:

```
>>> loss = error.spikeTime(spikeOut, spikeDes)
```

SPIKE INPUT/OUTPUT

`spikeFileIO.encode1DnumSpikes` (*filename, nID, tSt, tEn, nSp*)

Writes binary spike file given a tuple specifying neuron, start of spike region, end of spike region and number of spikes.

The binary file is encoded as follows:

- Number of spikes data is represented by an 80 bit number
- First 16 bits (bits 79-64) represent the neuronID
- Next 24 bits (bits 63-40) represents the start time in microseconds
- Next 24 bits (bits 39-16) represents the end time in microseconds
- Last 16 bits (bits 15-0) represents the number of spikes

Arguments:

- `filename` (string): path to the binary file
- `nID` (numpy array): neuron ID
- `tSt` (numpy array): region start time (in milliseconds)
- `tEn` (numpy array): region end time (in milliseconds)
- `nSp` (numpy array): number of spikes in the region

Usage:

```
>>> spikeFileIO.encode1DnumSpikes(file_path, nID, tSt, tEn, nSp)
```

`spikeFileIO.encode1Dspikes` (*filename, TD*)

Writes one dimensional binary spike file from a TD event.

The binary file is encoded as follows:

- Each spike event is represented by a 40 bit number.
- First 16 bits (bits 39-24) represent the neuronID.
- Bit 23 represents the sign of spike event: 0=>OFF event, 1=>ON event.
- the last 23 bits (bits 22-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename` (string): path to the binary file.
- `TD` (an `spikeFileIO.event`): TD event.

Usage:

```
>>> spikeFileIO.write1Dspikes(file_path, TD)
```

`spikeFileIO.encode2Dspikes(filename, TD)`

Writes two dimensional binary spike file from a TD event. It is the same format used in neuromorphic datasets NMNIST & NCALTECH101.

The binary file is encoded as follows:

- Each spike event is represented by a 40 bit number.
- First 8 bits (bits 39-32) represent the xID of the neuron.
- Next 8 bits (bits 31-24) represent the yID of the neuron.
- Bit 23 represents the sign of spike event: 0=>OFF event, 1=>ON event.
- The last 23 bits (bits 22-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename (string)`: path to the binary file.
- `TD (an spikeFileIO.event)`: TD event.

Usage:

```
>>> spikeFileIO.write2Dspikes(file_path, TD)
```

`spikeFileIO.encode3Dspikes(filename, TD)`

Writes binary spike file for TD event in height, width and channel dimension.

The binary file is encoded as follows:

- Each spike event is represented by a 56 bit number.
- First 12 bits (bits 56-44) represent the xID of the neuron.
- Next 12 bits (bits 43-32) represent the yID of the neuron.
- Next 8 bits (bits 31-24) represents the channel ID of the neuron.
- The last 24 bits (bits 23-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename (string)`: path to the binary file.
- `TD (an spikeFileIO.event)`: TD event.

Usage:

```
>>> spikeFileIO.write3Dspikes(file_path, TD)
```

class `spikeFileIO.event(xEvent, yEvent, pEvent, tEvent)`

This class provides a way to store, read, write and visualize spike event.

Members:

- `x (numpy int array)`: *x* index of spike event.
- `y (numpy int array)`: *y* index of spike event (not used if the spatial dimension is 1).
- `p (numpy int array)`: *polarity* or *channel* index of spike event.
- `t (numpy double array)`: *timestamp* of spike event. Time is assumend to be in ms.

Usage:


```
>>> TD = spikeFileIO.event(xEvent, yEvent, pEvent, tEvent)
```

toSpikeArray (*samplingTime=1, dim=None*)

Returns a numpy tensor that contains the spike events sampled in bins of *samplingTime*. The array is of dimension (channels, height, time) or “CHT” for 1D data. The array is of dimension (channels, height, width, time) or “CHWT” for 2D data.

Arguments:

- *samplingTime*: the width of time bin to use.
- *dim*: the dimension of the desired tensor. Assigns dimension itself if not provided.

Usage:

```
>>> spike = TD.toSpikeArray()
```

toSpikeTensor (*emptyTensor, samplingTime=1*)

Returns a numpy tensor that contains the spike events sampled in bins of *samplingTime*. The tensor is of dimension (channels, height, width, time) or “CHWT”.

Arguments:

- *emptyTensor* (numpy or torch tensor): an empty tensor to hold spike data
- *samplingTime*: the width of time bin to use.

Usage:

```
>>> spike = TD.toSpikeTensor( torch.zeros((2, 240, 180, 5000)) )
```

spikeFileIO.read1DnumSpikes (*filename*)

Reads a tuple specifying neuron, start of spike region, end of spike region and number of spikes from binary spike file.

The binary file is encoded as follows:

- Number of spikes data is represented by an 80 bit number.
- First 16 bits (bits 79-64) represent the neuronID.
- Next 24 bits (bits 63-40) represents the start time in microseconds.
- Next 24 bits (bits 39-16) represents the end time in microseconds.
- Last 16 bits (bits 15-0) represents the number of spikes.

Arguments:

- *filename* (string): path to the binary file

Usage:

```
>>> nID, tSt, tEn, nSp = spikeFileIO.read1DnumSpikes(file_path)
``tSt`` and ``tEn`` are returned in milliseconds
```

spikeFileIO.read1Dspikes (*filename*)

Reads one dimensional binary spike file and returns a TD event.

The binary file is encoded as follows:

- Each spike event is represented by a 40 bit number.
- First 16 bits (bits 39-24) represent the neuronID.

- Bit 23 represents the sign of spike event: 0=>OFF event, 1=>ON event.
- the last 23 bits (bits 22-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename (string)`: path to the binary file.

Usage:

```
>>> TD = spikeFileIO.read1Dspikes(file_path)
```

`spikeFileIO.read2Dspikes (filename)`

Reads two dimensional binary spike file and returns a TD event. It is the same format used in neuromorphic datasets MNIST & NCALTECH101.

The binary file is encoded as follows:

- Each spike event is represented by a 40 bit number.
- First 8 bits (bits 39-32) represent the xID of the neuron.
- Next 8 bits (bits 31-24) represent the yID of the neuron.
- Bit 23 represents the sign of spike event: 0=>OFF event, 1=>ON event.
- The last 23 bits (bits 22-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename (string)`: path to the binary file.

Usage:

```
>>> TD = spikeFileIO.read2Dspikes(file_path)
```

`spikeFileIO.read3Dspikes (filename)`

Reads binary spike file for spike event in height, width and channel dimension and returns a TD event.

The binary file is encoded as follows:

- Each spike event is represented by a 56 bit number.
- First 12 bits (bits 56-44) represent the xID of the neuron.
- Next 12 bits (bits 43-32) represent the yID of the neuron.
- Next 8 bits (bits 31-24) represents the channel ID of the neuron.
- The last 24 bits (bits 23-0) represent the spike event timestamp in microseconds.

Arguments:

- `filename (string)`: path to the binary file.

Usage:

```
>>> TD = spikeFileIO.read3Dspikes(file_path)
```

`spikeFileIO.showTD (TD, frameRate=24, preComputeFrames=True, repeat=False)`

Visualizes TD event.

Arguments:

- `TD`: spike event to visualize.
- `frameRate`: framerate of visualization.

- `preComputeFrames`: flag to enable precomputation of frames for faster visualization. Default is `True`.
- `repeat`: flag to enable repeat of animation. Default is `False`.

Usage:

```
>>> showTD(TD)
```

`spikeFileIO.spikeArrayToEvent` (*spikeMat*, *samplingTime=1*)

Returns TD event from a numpy array (of dimension 3 or 4). The numpy array must be of dimension (channels, height, time) or “CHT” for 1D data. The numpy array must be of dimension (channels, height, width, time) or “CHWT” for 2D data.

Arguments:

- `spikeMat`: numpy array with spike information.
- `samplingTime`: time width of each time bin.

Usage:

```
>>> TD = spikeFileIO.spikeArrayToEvent(spike)
```


LEARNING STATISTICS

class learningStats.learningStat

This class collect the learning statistics over the epoch.

Usage:

This class is designed to be used with learningStats instance although it can be used separately.

```
>>> trainingStat = learningStat()
```

accuracy()

Returns the average accuracy calculated from the point the stats was reset.

Usage:

```
>>> loss = trainingStat.accuracy()
```

loss()

Returns the average loss calculated from the point the stats was reset.

Usage:

```
>>> loss = trainingStat.loss()
```

reset()

Reset the learning staistics. This should usually be done before the start of an epoch so that new statistics counts can be accumulated.

Usage:

```
>>> trainingStat.reset()
```

update()

Updates the stats of the current session and resets the measures for next session.

Usage:

```
>>> trainingStat.update()
```

class learningStats.learningStats

This class provides mechanism to collect learning stats for training and testing, and displaying them efficiently.

Usage:

```
stats = learningStats()
```

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```

for epoch in range(100):
    tSt = datetime.now()

    for i in trainingLoop:
        # other main stuffs
        stats.training.correctSamples += numberOfCorrectClassification
        stats.training.numSamples    += numberOfSamplesProcessed
        stats.training.lossSum       += currentLoss
        stats.print(epoch, i, (datetime.now() - tSt).total_seconds())
    stats.training.update()

    for i in testingLoop:
        # other main stuffs
        stats.testing.correctSamples += numberOfCorrectClassification
        stats.testing.numSamples    += numberOfSamplesProcessed
        stats.testing.lossSum       += currentLoss
        stats.print(epoch, i)
    stats.training.update()

```

print (epoch, iter=None, timeElapsed=None)

Prints the available learning statistics from the current session on the console. For Linux systems, prints the data on same terminal space (might not work properly on other systems).

Arguments:

- epoch: epoch counter to display (required).
- iter: iteration counter to display (not required).
- timeElapsed: runtime information (not required).

Usage:

```

# prints stats with epoch index provided
stats.print(epoch)

# prints stats with epoch index and iteration index provided
stats.print(epoch, iter=i)

# prints stats with epoch index, iteration index and time elapsed information_
↪provided
stats.print(epoch, iter=i, timeElapsed=time)

```

update ()

Updates the stats for training and testing and resets the measures for next session.

Usage:

```
>>> stats.update()
```

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7.1 Usage:

```
>>> import slayerSNN as snn
```

- The **spike-layer** module will be available as `snn.layer`.
- The **yaml-parameter** module will be available as `snn.params`.
- The **spike-loss** module will be available as `snn.loss`.
- The **spike-classifier** module will be available as `snn.predict`.
- The **spike-IO** module will be available as `snn.io`.

7.2 Example:

The SNN parameters are stored in a yaml file. The structure of the yaml file follows the same hierarchy as the C++ SLAYER framework (see Network Description)

```
1 import slayerSNN as snn
2 # other imports and definitions
3
4 class Network(torch.nn.Module):
5     def __init__(self, netParams, device=device):
6         super(Network, self).__init__()
7         # initialize slayer
8         slayer = snn.layer(netParams['neuron'], netParams['simulation'], device=device)
9         self.sl = slayer
10        # define network functions
11        self.conv1 = slayer.conv(2, 16, 5, padding=1)
12        self.conv2 = slayer.conv(16, 32, 3, padding=1)
13        self.conv3 = slayer.conv(32, 64, 3, padding=1)
14        self.pool1 = slayer.pool(2)
15        self.pool2 = slayer.pool(2)
16        self.fcl = slayer.dense((8, 8, 64), 10)
```

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```

17
18     def forward(self, spikeInput):
19         spikeLayer1 = self.sl.spike(self.conv1(self.sl.psp(spikeInput))) # 32, 32, 16
20         spikeLayer2 = self.sl.spike(self.pool1(self.sl.psp(spikeLayer1))) # 16, 16, 16
21         spikeLayer3 = self.sl.spike(self.conv2(self.sl.psp(spikeLayer2))) # 16, 16, 32
22         spikeLayer4 = self.sl.spike(self.pool2(self.sl.psp(spikeLayer3))) # 8, 8, 32
23         spikeLayer5 = self.sl.spike(self.conv3(self.sl.psp(spikeLayer4))) # 8, 8, 64
24         spikeOut = self.sl.spike(self.fcl (self.sl.psp(spikeLayer5))) # 10
25         return spikeOut
26
27     # network
28     net = Network(snn.params('path to yaml file'))
29
30     # cost function
31     error = snn.loss(netParams)
32
33     # dataloader not shown. input and target are assumed to be available
34     output = net.forward(input)
35     loss = error.numSpikes(output, target)

```

Important: It is assumed that all the tensors that are being processed are 5 dimensional. (Batch, Channels, Height, Width, Time) or NCHWT format. The user must make sure that an input of correct dimension is supplied.

*If the layer does not have spatial dimension, the neurons can be distributed along either Channel, Height or Width dimension where Channel * Height * Width is equal to number of neurons. It is recommended (for speed reasons) to define the neurons in Channels dimension and make Height and Width dimension one.*

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