
MemTorch

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MemTorch is a simulation framework for memristive deep learning systems that integrates directly with the well-known PyTorch Machine Learning (ML) library. MemTorch is formally described in *MemTorch: An Open-source Simulation Framework for Memristive Deep Learning Systems*, which is openly accessible [here](#).

The best place to get started is [here](#).

CHAPTER 1

Documentation

We provide documentation in the form of a complete Python API, and numerous interactive tutorials. In addition, a Gitter chatroom is available for discussions:

1.1 Python API

MemTorch consists of various submodules, as defined below:

1.1.1 memtorch.bh

Submodule containing various memristive device behavioral models and methods to simulate non-ideal device and circuit behavior.

memtorch.bh.memristor

All memristor models and window functions are encapsulated and documented in memtorch.bh.memristor.

memtorch.bh.nonideality

All non-idealities modelled by MemTorch are encapsulated and documented in memtorch.bh.nonideality.

memtorch.bh.crossbar.Crossbar

Class used to model memristor crossbars and to manage modular crossbar tiles.

```
import torch
import memtorch

crossbar = memtorch.bh.crossbar.Crossbar(memtorch.bh.memristor.VTEAM,
                                         {"r_on": 1e2, "r_off": 1e4},
                                         shape=(100, 100),
                                         tile_shape=(64, 64))
crossbar.write_conductance_matrix(torch.zeros(100, 100).uniform_(1e-2, 1e-4),  
    ↪transistor=True)
crossbar.devices[0][0][0].set_conductance(1e-4)
crossbar.update(from_devices=True, parallelize=True)
```

Note: `use_bindings` is enabled by default, to accelerate operation using C++/CUDA (if supported) bindings.

Warning: As of version 1.1.6, the `write_conductance_matrix` method exhibits different behavior when `self.use_bindings` is True, CUDA operation is enabled, and the **Data_Driven2021** memristor model is used.

When `self.use_bindings` is True, CUDA operation is enabled, and the **Data_Driven2021** memristor model is used, the programming voltage is force adjusted by `force_adjustment_voltage` when a device becomes stuck. For all others models, or when CUDA operation is not enabled or `self.use_bindings` is false, the conductance state of the device being modelled is adjusted using `force_adjustment` when it becomes stuck.

This behavior will made consistent across Python, C++, and CUDA bindings, in a future release.

```
class memtorch.bh.crossbar.Crossbar(memristor_model, memris-  
                                     tor_model_params, shape,  
                                     tile_shape=None, use_bindings=True,  
                                     cuda_malloc_heap_size=50, ran-  
                                     dom_crossbar_init=False)
```

Bases: `object`

Class used to model memristor crossbars.

Parameters

- `memristor_model` (`memtorch.bh.memristor.Memristor`) – Memristor model.
- `memristor_model_params` (**kwargs) – **kwargs to instantiate the memristor model with.
- `shape` (`int`, `int`) – Shape of the crossbar.
- `tile_shape` (`int`, `int`) – Tile shape to use to store weights. If None, modular tiles are not used.
- `use_bindings` (`bool`) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- `random_crossbar_init` (`bool`) – Determines if the crossbar is to be initialized at random values in between Ron and Roff

`update` (`from_devices=True, parallelize=False`)

Method to update either the layers conductance_matrix or each devices conductance state.

Parameters

- **from_devices** (`bool`) – The conductance matrix can either be updated from all devices (True), or each device can be updated from the conductance matrix (False).
- **parallelize** (`bool`) – The operation is parallelized (True).

`write_conductance_matrix`(`conductance_matrix`, `transistor=True`, `programming_routine=None`,
`programming_routine_params={}`)

Method to directly program (alter) the conductance of all devices within the crossbar.

Parameters

- **conductance_matrix** (`torch.FloatTensor`) – Conductance matrix to write.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 0T1R arrangement (False) is simulated.
- **programming_routine** – Programming routine (method) to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.

`class memtorch.bh.crossbar.Crossbar.Scheme`

Bases: `enum.Enum`

Scheme enumeration.

`DoubleColumn = 2`

`SingleColumn = 1`

`memtorch.bh.crossbar.Crossbar.init_crossbar`(`weights`, `memristor_model`, `memristor_model_params`, `transistor`, `mapping_routine`, `programming_routine`, `programming_routine_params={}`, `p_l=None`, `scheme=<Scheme.DoubleColumn: 2>`, `tile_shape=(128, 128)`, `use_bindings=True`, `cuda_malloc_heap_size=50`, `random_crossbar_init=False`)

Method to initialise and construct memristive crossbars.

Parameters

- **weights** (`torch.Tensor`) – Weights to map.
- **memristor_model** (`memtorch.bh.memristor.Memristor.Memristor`) – Memristor model.
- **memristor_model_params** (**kwargs) – **kwargs to instantiate the memristor model with.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **mapping_routine** (`function`) – Mapping routine to use.
- **programming_routine** (`function`) – Programming routine to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_l** (`float`) – If not None, the proportion of weights to retain.
- **scheme** (`memtorch.bh.Scheme`) – Scheme enum.
- **tile_shape** (`int`, `int`) – Tile shape to use to store weights. If None, modular tiles are not used.

- **use_bindings** (`bool`) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- **random_crossbar_init** (`boolean`) – Determines if the crossbar is to be initialized at random values in between Ron and Roff

Returns The constructed crossbars and forward() function.

Return type tuple

```
memtorch.bh.crossbar.Crossbar.simulate_matmul(input, crossbar, nl=True,  
    tiles_map=None, crossbar_shape=None,  
    max_input_voltage=None,  
    ADC_resolution=None,  
    ADC_overflow_rate=0.0,  
    quant_method=None,  
    use_bindings=True)
```

Method to simulate non-linear IV device characterisits for a 2-D crossbar architecture given scaled inputs.

Parameters

- **input** (`torch.Tensor`) – Scaled input tensor.
- **crossbar** (`memtorch.bh.Crossbar`) – Crossbar containing devices to simulate.
- **nl** (`bool`) – Use lookup tables rather than simulating each device (True).
- **tiles_map** (`torch.Tensor`) – Tiles map for devices if tile_shape is not None.
- **crossbar_shape** (`int, int`) – Crossbar shape if tile_shape is not None.
- **max_input_voltage** (`float`) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **ADC_resolution** (`int`) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (`float`) – Overflow rate threshold for linear quanitzation (if ADC_resolution is not None).
- **quant_method** – Quantization method. Must be in memtorch.bh.Quantize.quant_methods.
- **use_bindings** (`bool`) – Used to determine if C++/CUDA bindings are used (True) or not (False).

Returns Output tensor.

Return type torch.Tensor

memtorch.bh.crossbar.Program

Methods to program (alter) the conductance devices within a crossbar or modular crossbar tiles.

```
memtorch.bh.crossbar.Program.gen_programming_signal(number_of_pulses,  
    pulse_duration, refactory_period, voltage_level,  
    time_series_resolution)
```

Method to generate a programming signal using a sequence of pulses.

Parameters

- **number_of_pulses** (`int`) – Number of pulses.

- **pulse_duration** (*float*) – Duration of the programming pulse (s).
- **refactory_period** (*float*) – Duration of the refractory period (s).
- **voltage_level** (*float*) – Voltage level (V).
- **time_series_resolution** (*float*) – Time series resolution (s).

Returns Tuple containing the generated time and voltage signals.

Return type `tuple`

```
memtorch.bh.crossbar.Program.naive_program(crossbar, point, conductance, rel_tol=0.01,
                                             pulse_duration=0.001, refactory_period=0,
                                             pos_voltage_level=1.0, neg_voltage_level=-1.0,
                                             timeout=5, force_adjustment=0.001,
                                             force_adjustment_rel_tol=0.1,
                                             force_adjustment_pos_voltage_threshold=0,
                                             force_adjustment_neg_voltage_threshold=0,
                                             failure_iteration_threshold=1000, simulate_neighbours=True)
```

Method to program (alter) the conductance of a given device within a crossbar.

Parameters

- **crossbar** (`memtorch.bh.crossbar.Crossbar`) – Crossbar containing the device to program.
- **point** (`tuple`) – Point to program (row, column).
- **conductance** (*float*) – Conductance to program.
- **rel_tol** (*float*) – Relative tolerance between the desired conductance and the device's conductance.
- **pulse_duration** (*float*) – Duration of the programming pulse (s).
- **refactory_period** (*float*) – Duration of the refractory period (s).
- **pos_voltage_level** (*float*) – Positive voltage level (V).
- **neg_voltage_level** (*float*) – Negative voltage level (V).
- **timeout** (*int*) – Timeout (seconds) until stuck devices are unstuck.
- **force_adjustment** (*float*) – Adjustment (resistance) to unstick stuck devices.
- **force_adjustment_rel_tol** (*float*) – Relative tolerance threshold between a stuck device's conductance and high and low conductance states to force adjust.
- **force_adjustment_pos_voltage_threshold** (*float*) – Positive voltage level threshold (V) to enable force adjustment.
- **force_adjustment_neg_voltage_threshold** (*float*) – Negative voltage level threshold (V) to enable force adjustment.
- **failure_iteration_threshold** (*int*) – Failure iteration threshold.
- **simulate_neighbours** (`bool`) – Simulate neighbours (True).

Returns Programmed device.

Return type `memtorch.bh.memristor.Memristor.Memristor`

memtorch.bh.crossbar.Tile

```
class memtorch.bh.crossbar.Tile(tile_shape, patch_num=None)
Bases: object
```

Class used to create modular crossbar tiles to represent 2D matrices.

Parameters

- **tile_shape** (*int*, *int*) – Tile shape to use to store weights.
- **patch_num** (*int*) – Patch number.

update_array (*new_array*)

Method to update the tile's weights.

Parameters **new_array** (*torch.Tensor*) – New array to construct the tile with.

```
memtorch.bh.crossbar.Tile.gen_tiles(tensor, tile_shape, input=False, use_bindings=True)
```

Method to generate a set of modular tiles representative of a tensor.

Parameters

- **tensor** (*torch.Tensor*) – Tensor to represent using modular crossbar tiles.
- **tile_shape** (*int*, *int*) – Tile shape to use to store weights.
- **input** (*bool*) – Used to determine if a tensor is an input (True).

Returns Tiles and tile_map.

Return type *torch.Tensor*, *torch.Tensor*

```
memtorch.bh.crossbar.Tile.tile_matmul(mat_a_tiles, mat_a_tiles_map, mat_a_shape,
                                       mat_b_tiles, mat_b_tiles_map, mat_b_shape,
                                       source_resistance=None, line_resistance=None,
                                       ADC_resolution=None, ADC_overflow_rate=0.0,
                                       quant_method=None, transistor=True,
                                       use_bindings=True, cuda_malloc_heap_size=50)
```

Method to perform 2D matrix multiplication, given two sets of tiles.

Parameters

- **mat_a_tiles** (*torch.Tensor*) – Tiles representing matrix A.
- **mat_a_tiles_map** (*torch.Tensor*) – Tiles map for matrix A.
- **mat_a_shape** (*int*, *int*) – Shape of matrix A.
- **mat_b_tiles** (*torch.Tensor*) – Tiles representing matrix B.
- **mat_b_tiles_map** (*torch.Tensor*) – Tiles map for matrix B.
- **mat_b_shape** (*int*, *int*) – Shape of matrix B.
- **source_resistance** (*float*) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (*float*) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (*int*) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (*float*) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).

- **quant_method** (*str*) – Quantization method. Must be in memtorch.bh.Quantize.quant_methods.
- **transistor** (*bool*) – TBD.
- **use_bindings** (*bool*) – Use C++/CUDA bindings to parallelize tile_matmul operations (True).
- **cuda_malloc_heap_size** (*int*) – cudaLimitMallocHeapSize (in MB) to determine allocatable kernel heap memory if CUDA is used.

Returns Output tensor.

Return type torch.Tensor

```
memtorch.bh.crossbar.Tile.tile_matmul_row(mat_a_row_tiles, mat_a_tiles_map,
                                            mat_b_tiles, mat_b_tiles_map,
                                            mat_b_shape, source_resistance=None,
                                            line_resistance=None, ADC_resolution=None,
                                            ADC_overflow_rate=0.0, quant_method=None,
                                            transistor=True)
```

Method to perform row-wise tile matrix multiplication, given two sets of tiles, using a pythonic approach.

Parameters

- **mat_a_row_tiles** (*torch.Tensor*) – Tiles representing a row of matrix A.
- **mat_a_tiles_map** (*torch.Tensor*) – Tiles map for matrix A.
- **mat_b_tiles** (*torch.Tensor*) – Tiles representing matrix B.
- **mat_b_tiles_map** (*torch.Tensor*) – Tiles map for matrix B.
- **mat_b_shape** (*int*, *int*) – Shape of matrix B.
- **source_resistance** (*float*) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (*float*) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (*int*) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (*float*) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).
- **quant_method** (*str*) – Quantization method. Must be in memtorch.bh.Quantize.quant_methods.
- **transistor** (*bool*) – TBD.

Returns Output tensor.

Return type torch.Tensor

```
memtorch.bh.crossbar.Tile.tiled_inference(input, m, transistor)
```

Method to perform tiled inference.

Parameters

- **input** (*torch.Tensor*) – Input tensor (2-D).
- **m** (*memtorch.mn*) – Memristive MemTorch layer.

Returns Output tensor.

Return type torch.Tensor

memtorch.bh.Quantize

Wrapper for C++ quantization bindings.

```
memtorch.bh.Quantize.quantize(tensor, quant, overflow_rate=0.0, quant_method=None, min=nan,  
                               max=nan, override_original=False)
```

Method to quantize a tensor.

Parameters

- **tensor** (`torch.Tensor`) – Input tensor.
- **quant** (`int`) – Bit width (if quant_method is not None) or the number of discrete quantization levels (if quant_method is None).
- **overflow_rate** (`float`, *optional*) – Overflow rate threshold for linear quantization.
- **quant_method** (`str`, *optional*) – Quantization method. Must be in quant_methods.
- **min** (`float` or `tensor`, *optional*) – Minimum value(s) to clip numbers to.
- **max** (`float` or `tensor`, *optional*) – Maximum value(s) to clip numbers to.
- **override_original** (`bool`, *optional*) – Whether to override the original tensor (True) or not (False).

Returns Quantized tensor.

Return type `torch.Tensor`

memtorch.bh.StochasticParameter

Methods to model stochastic parameters.

memtorch.bh.StochasticParameter is most commonly used to define stochastic parameters when defining behavioural memristor models, as follows:

```
import torch  
import memtorch  
  
crossbar = memtorch.bh.crossbar.Crossbar(memtorch.bh.memristor.VTEAM,  
                                         {"r_on": memtorch.bh.  
                                         ↳StochasticParameter(min=1e3, max=1e2), "r_off": 1e4},  
                                         shape=(100, 100),  
                                         tile_shape=(64, 64))
```

```
class memtorch.bh.StochasticParameter.Dict2Obj(dictionary)  
Bases: object
```

Class used to instantiate a object given a dictionary.

```
memtorch.bh.StochasticParameter.StochasticParameter(distribution=<class  
                                                 'torch.distributions.normal.Normal'>,  
                                                 min=0, max=inf, function=True,  
                                                 **kwargs)
```

Method to model a stochastic parameter.

Parameters

- **distribution** (`torch.distributions`) – torch distribution.
- **min** (`float`) – Minimum value to sample.

- **max** (*float*) – Maximum value to sample.
- **function** (*bool*) – A sampled value is returned (False). A function to return a sampled value or mean is returned (True).

Returns A sampled value of the stochastic parameter, or a sample-value generator.

Return type `float` or function

```
memtorch.bh.StochasticParameter.unpack_parameters(local_args, r_rel_tol=None,
                                                 r_abs_tol=None, resample_threshold=5)
```

Method to sample from stochastic sample-value generators

Parameters

- **local_args** (`locals()`) – Local arguments with stochastic sample-value generators from which to sample from.
- **r_rel_tol** (*float*) – Relative threshold tolerance.
- **r_abs_tol** (*float*) – Absolute threshold tolerance.
- **resample_threshold** (*int*) – Number of times to resample `r_off` and `r_on` when their proximity is within the threshold tolerance before raising an exception.

Returns `locals()` with sampled stochastic parameters.

Return type `**`

1.1.2 memtorch.map

Submodule containing various mapping, scaling, and encoding methods.

memtorch.map.Input

Encapsulates internal methods to encode (scale) input values as bit-line voltages. Methods can either be specified when converting individual layers:

```
from memtorch.map.Input import naive_scale

m = memtorch.mn.Linear(torch.nn.Linear(10, 10),
                      memtorch.bh.memristor.VTEAM,
                      {},
                      tile_shape=(64, 64),
                      scaling_routine=naive_scale)
```

or when converting `torch.nn.Module` instances:

```
import copy
from memtorch.mn.Module import patch_model
from memtorch.map.Input import naive_scale
import Net

model = Net()
patched_model = patch_model(copy.deepcopy(model),
                           memtorch.bh.memristor.VTEAM,
                           {},
                           module_parameters_to_patch=[torch.nn.Linear],
                           scaling_routine=naive_scale)
```

```
memtorch.map.Input.naive_scale(module, input, force_scale=False)
```

Naive method to encode input values as bit-line voltages.

Parameters

- **module** (`torch.nn.Module`) – Memristive layer to tune.
- **input** (`torch.tensor`) – Input tensor to encode.
- **force_scale** (`bool, optional`) – Used to determine if inputs are scaled (True) or not (False) if they no not exceed max_input_voltage.

Returns Encoded voltages.

Return type `torch.Tensor`

Note: `force_scale` is used to specify whether inputs smaller than or equal to `max_input_voltage` are scaled or not.

memtorch.map.Module

Encapsulates internal methods to determine relationships between readout currents of memristive crossbars and desired outputs.

Warning: Currently, only `naive_tune` is supported. In a future release, externally-defined methods will be supported.

```
memtorch.map.Module.naive_tune(module, input_shape, verbose=True)
```

Method to determine a linear relationship between a memristive crossbar and the output for a given memristive module.

Parameters

- **module** (`torch.nn.Module`) – Memristive layer to tune.
- **input_shape** (`int, int`) – Shape of the randomly generated input used to tune a cross-bar.
- **verbose** (`bool, optional`) – Used to determine if verbose output is enabled (True) or disabled (False).

Returns Function which transforms the output of the crossbar to the expected output.

Return type function

memtorch.map.Parameter

Encapsulates internal methods to naively map network parameters to memristive device conductance values. Methods can either be specified when converting individual layers:

```
from memtorch.map.Parameter import naive_map

m = memtorch.mn.Linear(torch.nn.Linear(10, 10),
                      memtorch.bh.memristor.VTEAM,
                      {},
                      tile_shape=(64, 64),
                      mapping_routine=naive_map)
```

or when converting `torch.nn.Module` instances:

```
import copy
from memtorch.mn.Module import patch_model
from memtorch.map.Parameter import naive_map
import Net

model = Net()
patched_model = patch_model(copy.deepcopy(model),
                            memtorch.bh.memristor.VTEAM,
                            {},
                            module_parameters_to_patch=[torch.nn.Linear],
                            mapping_routine=naive_map)
```

`memtorch.map.Parameter.naive_map`(*weight*, *r_on*, *r_off*, *scheme*, *p_l*=None)

Method to naively map network parameters to memristive device conductances, using two crossbars to represent both positive and negative weights.

Parameters

- **weight** (`torch.Tensor`) – Weight tensor to map.
- **r_on** (`float`) – Low resistance state.
- **r_off** (`float`) – High resistance state.
- **scheme** (`memtorch.bh.crossbar.Scheme`) – Weight representation scheme.
- **p_l** (`float, optional`) – If not None, the proportion of weights to retain.

Returns Positive and negative crossbar weights.

Return type `torch.Tensor, torch.Tensor`

1.1.3 memtorch.mn

Memristive `torch.nn` equivalent submodule.

memtorch.mn.Module

Encapsulates `memtorch.bmn.Module.patch_model`, which can be used to convert `torch.nn` models.

```
import copy
import Net
from memtorch.mn.Module import patch_model
from memtorch.map.Parameter import naive_map
from memtorch.map.Input import naive_scale

model = Net()
reference_memristor = memtorch.bh.memristor.VTEAM
patched_model = patch_model(copy.deepcopy(model),
                            memristor_model=reference_memristor,
                            memristor_model_params={},
                            module_parameters_to_patch=[torch.nn.Linear, torch.nn.Conv2d],
                            mapping_routine=naive_map,
                            transistor=True,
                            programming_routine=None,
                            tile_shape=(128, 128),
```

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```
max_input_voltage=0.3,
scaling_routine=naive_scale,
ADC_resolution=8,
ADC_overflow_rate=0.,
quant_method='linear')
```

Warning: It is strongly suggested to copy the original model using `copy.deepcopy` prior to conversion, as some values are overridden by-reference.

```
memtorch.mn.Module.patch_model(model, memristor_model, memristor_model_params, module_parameters_to_patch={}, mapping_routine=<function naive_map>, transistor=True, programming_routine=None, programming_routine_params={'rel_tol': 0.1}, p_l=None, scheme=<Scheme.DoubleColumn: 2>, tile_shape=None, max_input_voltage=None, scaling_routine=<function naive_scale>, scaling_routine_params={}, source_resistance=None, line_resistance=None, ADC_resolution=None, ADC_overflow_rate=0.0, quant_method=None, use_bindings=True, random_crossbar_init=False, verbose=True, **kwargs)
```

Method to convert a torch.nn model to a memristive model.

Parameters

- **model** (`torch.nn.Module`) – torch.nn.Module to patch.
- **memristor_model** (`memtorch.bh.memristor.Memristor.Memristor`) – Memristor model.
- **memristor_model_params** (**kwargs) – Memristor model keyword arguments.
- **module_parameters_to_patch** (`module_parameter_patches`) – Model parameters to patch.
- **mapping_routine** (`function`) – Mapping routine to use.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **programming_routine** (`function`) – Programming routine to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_l** (`float`) – If not None, the proportion of weights to retain.
- **scheme** (`memtorch.bh.Scheme`) – Weight representation scheme.
- **tile_shape** ((`int`, `int`)) – Tile shape to use to store weights. If None, modular tiles are not used.
- **max_input_voltage** (`float`) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **scaling_routine** (`function`) – Scaling routine to use in order to scale batch inputs.
- **scaling_routine_params** (**kwargs) – Scaling routine keyword arguments.

- **source_resistance** (*float*) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (*float*) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (*int*) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (*float*) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).
- **quant_method** – Quantization method. Must be in ['linear', 'log', 'log_minmax', 'minmax', 'tanh'], or None.
- **use_bindings** (*bool*) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- **random_crossbar_init** (*bool*) – Determines if the crossbar is to be initialized at random values in between Ron and Roff
- **verbose** (*bool*) – Used to determine if verbose output is enabled (True) or disabled (False).

Returns Patched torch.nn.Module.

Return type torch.nn.Module

The following layer/module types are currently supported:

memtorch.mn.Linear

torch.nn.Linear equivalent.

```
class memtorch.mn.Linear.Linear(linear_layer, memristor_model, memristor_model_params,
                                mapping_routine=<function naive_map>, transistor=True,
                                programming_routine=None, programming_routine_params={}, p_l=None,
                                scheme=<Scheme.DoubleColumn: 2>, tile_shape=None,
                                max_input_voltage=None, scaling_routine=<function naive_scale>,
                                scaling_routine_params={}, source_resistance=None, line_resistance=None,
                                ADC_resolution=None, ADC_overflow_rate=0.0,
                                quant_method=None, use_bindings=True, random_crossbar_init=False,
                                verbose=True, *args, **kwargs)
```

Bases: torch.nn.modules.linear.Linear

nn.Linear equivalent.

Parameters

- **linear_layer** (*torch.nn.Linear*) – Linear layer to patch.
- **memristor_model** (*memtorch.bh.memristor.Memristor.Memristor*) – Memristor model.
- **memristor_model_params** (**kwargs) – Memristor model keyword arguments.
- **mapping_routine** (*function*) – Mapping routine to use.
- **transistor** (*bool*) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **programming_routine** (*function*) – Programming routine to use.

- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_1** (*float*) – If not None, the proportion of weights to retain.
- **scheme** (*memtorch.bh.Scheme*) – Weight representation scheme.
- **tile_shape** ((*int*, *int*)) – Tile shape to use to store weights. If None, modular tiles are not used.
- **max_input_voltage** (*float*) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **scaling_routine** (*function*) – Scaling routine to use in order to scale batch inputs.
- **scaling_routine_params** (**kwargs) – Scaling routine keyword arguments.
- **source_resistance** (*float*) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (*float*) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (*int*) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (*float*) – Overflow rate threshold for linear quanitzation (if ADC_resolution is not None).
- **quant_method** (*string*) – Quantization method. Must be in ['linear', 'log', 'log_minmax', 'minmax', 'tanh'], or None.
- **use_bindings** (*bool*) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- **random_crossbar_init** (*bool*) – Determines if the crossbar is to be initialized at random values in between Ron and Roff
- **verbose** (*bool*) – Used to determine if verbose output is enabled (True) or disabled (False).

forward (*input*)

Method to perform forward propagations.

Parameters **input** (*torch.Tensor*) – Input tensor.

Returns Output tensor.

Return type *torch.Tensor*

tune (*input_shape=4098*)

Tuning method.

memtorch.mn.Conv1d

torch.nn.Conv1d equivalent.

```
class memtorch.mn.Conv1d(convolutional_layer,          memristor_model,      memris-
                        tor_model_params,        mapping_routine=<function
                        naive_map>, transistor=True, programming_routine=None,
                        programming_routine_params={}, p_l=None,
                        scheme=<Scheme.DoubleColumn: 2>, tile_shape=None,
                        max_input_voltage=None,    scaling_routine=<function
                        naive_scale>,           scaling_routine_params={},
                        source_resistance=None,   line_resistance=None,
                        ADC_resolution=None,      ADC_overflow_rate=0.0,
                        quant_method=None,        use_bindings=True,      ran-
                        dom_crossbar_init=False, verbose=True, *args, **kwargs)
```

Bases: torch.nn.modules.conv.Conv1d

nn.Conv1d equivalent.

Parameters

- **convolutional_layer** (`torch.nn.Conv1d`) – Convolutional layer to patch.
- **memristor_model** (`memtorch.bh.memristor.Memristor.Memristor`) – Memristor model.
- **memristor_model_params** (**kwargs) – Memristor model keyword arguments.
- **mapping_routine** (`function`) – Mapping routine to use.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **programming_routine** (`function`) – Programming routine to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_l** (`float`) – If not None, the proportion of weights to retain.
- **scheme** (`memtorch.bh.Scheme`) – Weight representation scheme.
- **tile_shape** ((`int`, `int`)) – Tile shape to use to store weights. If None, modular tiles are not used.
- **max_input_voltage** (`float`) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **scaling_routine** (`function`) – Scaling routine to use in order to scale batch inputs.
- **scaling_routine_params** (**kwargs) – Scaling routine keyword arguments.
- **source_resistance** (`float`) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (`float`) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (`int`) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (`float`) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).
- **quant_method** (`string`) – Quantization method. Must be in ['linear', 'log', 'log_minmax', 'minmax', 'tanh'], or None.
- **use_bindings** (`bool`) – Used to determine if C++/CUDA bindings are used (True) or not (False).

- **random_crossbar_init** (`bool`) – Determines if the crossbar is to be initialized at random values in between Ron and Roff
- **verbose** (`bool`) – Used to determine if verbose output is enabled (True) or disabled (False).

forward(*input*)

Method to perform forward propagations.

Parameters **input** (`torch.Tensor`) – Input tensor.

Returns Output tensor.

Return type `torch.Tensor`

tune(*input_batch_size*=8, *input_shape*=32)

Tuning method.

memtorch.mn.Conv2d

`torch.nn.Conv2d` equivalent.

```
class memtorch.mn.Conv2d.Conv2d(convolutional_layer, memristor_model, memris-
                                tor_model_params, mapping_routine=<function
                                naive_map>, transistor=True, programming_routine=None,
                                programming_routine_params={}, p_l=None,
                                scheme=<Scheme.DoubleColumn: 2>, tile_shape=None,
                                max_input_voltage=None, scaling_routine=<function
                                naive_scale>, scaling_routine_params={},
                                source_resistance=None, line_resistance=None,
                                ADC_resolution=None, ADC_overflow_rate=0.0,
                                quant_method=None, use_bindings=True, ran-
                                dom_crossbar_init=False, verbose=True, *args, **kwargs)
```

Bases: `torch.nn.modules.conv.Conv2d`

`nn.Conv2d` equivalent.

Parameters

- **convolutional_layer** (`torch.nn.Conv2d`) – Convolutional layer to patch.
- **memristor_model** (`memtorch.bh.memristor.Memristor.Memristor`) – Memristor model.
- **memristor_model_params** (**kwargs) – Memristor model keyword arguments.
- **mapping_routine** (`function`) – Mapping routine to use.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **programming_routine** (`function`) – Programming routine to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_l** (`float`) – If not None, the proportion of weights to retain.
- **scheme** (`memtorch.bh.Scheme`) – Weight representation scheme.
- **tile_shape** ((`int`, `int`)) – Tile shape to use to store weights. If None, modular tiles are not used.

- **max_input_voltage** (*float*) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **scaling_routine** (*function*) – Scaling routine to use in order to scale batch inputs.
- **scaling_routine_params** (***kwargs*) – Scaling routine keyword arguments.
- **source_resistance** (*float*) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (*float*) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (*int*) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (*float*) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).
- **quant_method** (*string*) – Quantization method. Must be in ['linear', 'log', 'log_minmax', 'minmax', 'tanh'], or None.
- **use_bindings** (*bool*) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- **random_crossbar_init** (*bool*) – Determines if the crossbar is to be initialized at random values in between Ron and Roff
- **verbose** (*bool*) – Used to determine if verbose output is enabled (True) or disabled (False).

forward(*input*)

Method to perform forward propagations.

Parameters **input** (*torch.Tensor*) – Input tensor.

Returns Output tensor.

Return type *torch.Tensor*

tune(*input_batch_size=8, input_shape=32*)

Tuning method.

memtorch.mn.Conv3d

torch.nn.Conv3d equivalent.

```
class memtorch.mn.Conv3d.Conv3d(convolutional_layer, memristor_model, memris-
tor_model_params, mapping_routine=<function
naive_map>, transistor=True, programming_routine=None,
programming_routine_params={}, p_l=None,
scheme=<Scheme.DoubleColumn: 2>, tile_shape=None,
max_input_voltage=None, scaling_routine=<function
naive_scale>, scaling_routine_params={}, line_resistance=None,
source_resistance=None, ADC_overflow_rate=0.0,
ADC_resolution=None, quant_method=None, use_bindings=True, ran-
dom_crossbar_init=False, verbose=True, *args, **kwargs)
```

Bases: *torch.nn.modules.conv.Conv3d*

nn.Conv3d equivalent.

Parameters

- **convolutional_layer** (`torch.nn.Conv3d`) – Convolutional layer to patch.
- **memristor_model** (`memtorch.bh.memristor.Memristor.Memristor`) – Memristor model.
- **memristor_model_params** (**kwargs) – Memristor model keyword arguments.
- **mapping_routine** (`function`) – Mapping routine to use.
- **transistor** (`bool`) – Used to determine if a 1T1R (True) or 1R arrangement (False) is simulated.
- **programming_routine** (`function`) – Programming routine to use.
- **programming_routine_params** (**kwargs) – Programming routine keyword arguments.
- **p_1** (`float`) – If not None, the proportion of weights to retain.
- **scheme** (`memtorch.bh.Scheme`) – Weight representation scheme.
- **tile_shape** ((`int`, `int`)) – Tile shape to use to store weights. If None, modular tiles are not used.
- **max_input_voltage** (`float`) – Maximum input voltage used to encode inputs. If None, inputs are unbounded.
- **scaling_routine** (`function`) – Scaling routine to use in order to scale batch inputs.
- **scaling_routine_params** (**kwargs) – Scaling routine keyword arguments.
- **source_resistance** (`float`) – The resistance between word/bit line voltage sources and crossbar(s).
- **line_resistance** (`float`) – The interconnect line resistance between adjacent cells.
- **ADC_resolution** (`int`) – ADC resolution (bit width). If None, quantization noise is not accounted for.
- **ADC_overflow_rate** (`float`) – Overflow rate threshold for linear quantization (if ADC_resolution is not None).
- **quant_method** (`string`) – Quantization method. Must be in ['linear', 'log', 'log_minmax', 'minmax', 'tanh'], or None.
- **use_bindings** (`bool`) – Used to determine if C++/CUDA bindings are used (True) or not (False).
- **random_crossbar_init** (`bool`) – Determines if the crossbar is to be initialized at random values in between Ron and Roff
- **verbose** (`bool`) – Used to determine if verbose output is enabled (True) or disabled (False).

forward (`input`)

Method to perform forward propagations.

Parameters **input** (`torch.Tensor`) – Input tensor.

Returns Output tensor.

Return type `torch.Tensor`

tune (`input_batch_size=4, input_shape=32`)

Tuning method.

1.2 Tutorials

To learn how to use MemTorch using interactive tutorials, and to reproduce simulations presented in ‘MemTorch: An Open-source Simulation Framework for Memristive Deep Learning Systems’ [1], we provide numerous Jupyter notebooks.

Jupyter Note-book	Description	Google Colab Link
Tutorial	Introductory Tutorial- Start Here	
Exemplar Simulations	Various Exemplar Simulations, As Presented In [1]	
Case Study	(Legacy) An epileptic Seizure Detection Case Study	
Novel Simulations	(Legacy) Novel Simulations Using CIFAR-10	

The development of more Jupyter notebooks and tutorials is currently ongoing.

[1] C. Lammie, W. Xiang, B. Linares-Barranco, and Azghadi, Mostafa Rahimi, “MemTorch: An Open-source Simulation Framework for Memristive Deep Learning Systems,” arXiv.org, 2020. <https://arxiv.org/abs/2004.10971>.

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