

# Continuum

**Simple** Management of Complex Continual Learning Scenarios



```
$ pip3 install continuum
```

<https://github.com/Continuumvm/continuum>

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# Data Loading for Continual Learning

Continual data loading is **complex**

Each paper requires **many different settings**

Reinventing the wheel

```
dataset = MNIST("my/data/path", download=True, train=True)
scenario = ClassIncremental(dataset, increment=2)

for task_id, taskset in enumerate(scenario):
    train_taskset, val_taskset = split_train_val(taskset, val_split=0.1)
    train_loader = DataLoader(train_taskset, batch_size=32, shuffle=True)
    val_loader = DataLoader(val_taskset, batch_size=32, shuffle=True)

    for x, y, t in train_loader:
        # Do your cool stuff here
```

# UNIX Philosophy:

Minimal

One Single Goal

Modular

1. Choose a dataset,
2. Choose a scenario,
3. and use *torchvision* loaders!

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    for x, y, t in train_loader:
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But wait there is already many libraries...

Sequoia

FACIL

Avalanche

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Continuum

Released to public and used by several researchers since **April 2020**

**Light-weight** with only goal in mind: data loading, no models!

Easiest to **plug in** existing codebase

Also **good synergy** with large codebase such as Sequoia!

# But wait there is already many libraries...

Sequoia

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**Different goal than existing libraries**

Continuum

Released to public and used by several researchers since **April 2020**

**Light-weight** with only goal in mind: data loading, no models!

Easiest to **plug in** existing codebase

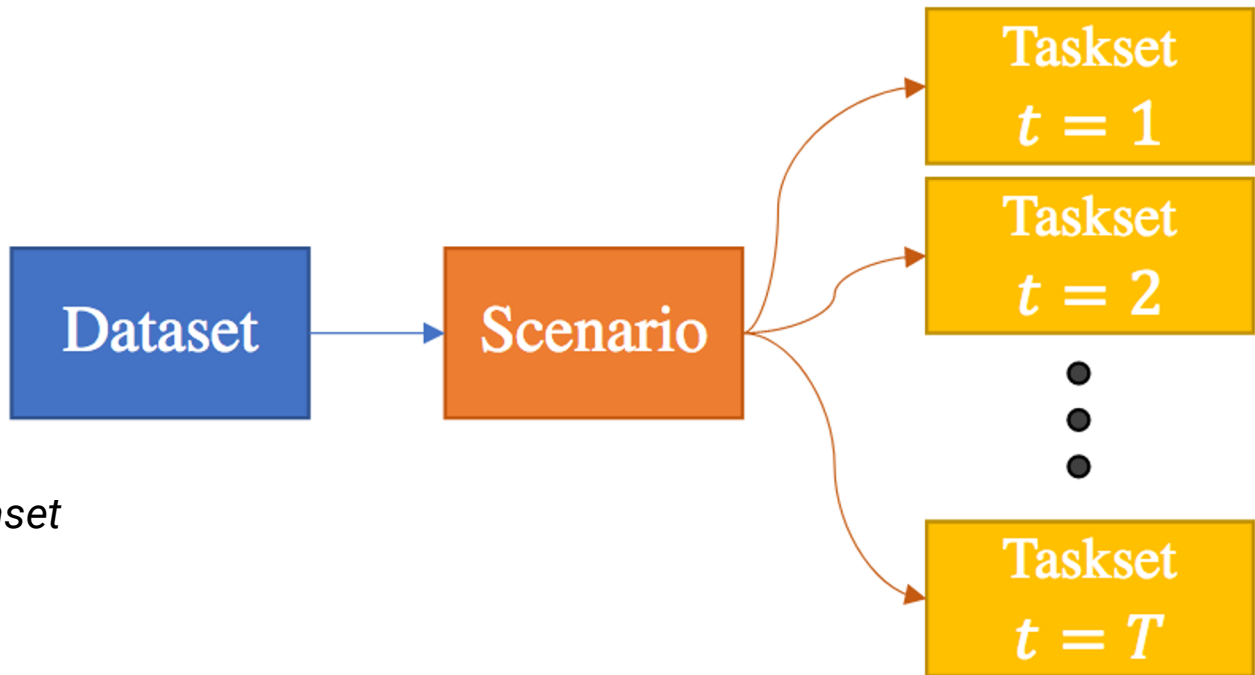
Also **good synergy** with large codebase such as Sequoia!

# No big diagram


More than 20 datasets

5 different scenarios

Tasksets are *torchvision.Dataset*



# Smooth integration in Pytorch



```
from torch.utils.data import DataLoader
from continuum import ClassIncremental
from continuum.datasets import MNIST

dataset = MNIST("/data", train=True)
scenario = ClassIncremental(dataset, increment=2)

for task_id, train_taskset in enumerate(scenario):
    train_loader = DataLoader(train_taskset, batch_size=32, shuffle=True)

    for x, y, t in train_loader:
        # x --> data
        # y --> labels
        # t --> task ids
```



# Class-Incremental vs Task-Incremental

Use task ids,  
or not.

Your choice!

```
from torch.utils.data import DataLoader
from continuum import ClassIncremental
from continuum.datasets import MNIST

dataset = MNIST("/data", train=False)
scenario = ClassIncremental(dataset, increment=2)

for task_id, test_taskset in enumerate(scenario):
    test_loader = DataLoader(test_taskset, batch_size=32)

    for x, y, t in test_loader:
        # x --> data
        # y --> labels
        # t --> task ids
```

# Tasks Flexibility



```
dataset = MNIST("/data", train=False)
scenario = ClassIncremental(dataset, increment=2)

third_taskset = scenario[2]
all_seen_tasksets = scenario[:3]
```

All usual python fancy indexing for scenario

# Split MNIST



```
from continuum import ClassIncremental
from continuum.datasets import MNIST

dataset = MNIST("/path", train=True)
scenario = ClassIncremental(dataset, increment=2)
```

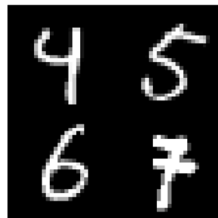
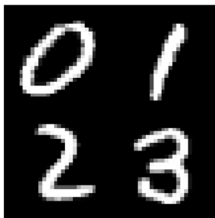


# Split MNIST



```
from continuum import ClassIncremental
from continuum.datasets import MNIST

dataset = MNIST("/path", train=True)
scenario = ClassIncremental(dataset, increment=[4, 4, 2])
```



# New Instances



```
from continuum import InstanceIncremental
from continuum.datasets import Core50v2_79
```

```
dataset = Core50v2_79("/path", train=True, download=True)
scenario = InstanceIncremental(dataset)
```



...

# Segmentation



```
dataset = PascalVOC2012("/path", train=True, download=True)
scenario = SegmentationClassIncremental(
    dataset,
    increment=[19, 1], # Learning 19 classes then 1
    nb_classes=20,
    mode="overlap",
    transformations=[Resize((512, 512)), ToTensor()]
)

for taskset in scenario:
    loader = DataLoader(taskset, batch_size=12)
    for x, y, t in loaders:
        # Same interface, your model goes here
```



# Transformations Scenario



```
from torchvision.transforms import RandomAffine

dataset = MNIST("/path", train=True)
scenario = TransformationIncremental(
    dataset,
    [
        [RandomAffine(degrees=0)],
        [RandomAffine(degrees=45)],
        [RandomAffine(degrees=90)]
    ]
)
```

**Task 0:** 0° degree



**Task 1:** 45° degree



**Task 2:** 90° degree



# Rehearsal Learning

Handle rehearsal memory as you want

Use memory in the taskset you want

Oversample with **native** pytorch samplers

```
from torch.utils.data import DataLoader
from continuum import ClassIncremental
from continuum.datasets import MNIST

dataset = MNIST("/path", train=True)
scenario = ClassIncremental(dataset, increment=2)

taskset = scenario[2]
# Add rehearsal memory to current task
taskset.add_samples(memory_x, memory_y)

loader = DataLoader(
    taskset,
    sampler=my_pytorch_sampler
)
```



# Future

- 🚫 We don't want to increase complexity:
  - No addition of tons of options, attributes, etc.
  - No models, no losses, etc.
- ✅ We want to offer a large choice of simple interfaces:
  - More datasets (more segmentation, more NLP, etc.)
  - More scenarios

## Don't use Continuum

- if you want a whole ecosystem,
- many models,
- and a strict way to do continual learning

## Do use Continuum

- if you want a **light-weight** library,
- easily **pluggable** in any codebase, no matter how small or large it is
- with the most **pytorchnic** interface

# Want more?

 Documentation: [continuum.readthedocs.io](https://continuum.readthedocs.io)

 Github: [github.com/Continuumio/continuum](https://github.com/Continuumio/continuum)

 Colab tutorial: [colab.research.google.com](https://colab.research.google.com)



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