

Use with NumPy

This document is a quick introduction to using datasets with NumPy, with a particular focus on how to get

numpy.ndarray objects out of our datasets, and how to use them to train models based on NumPy such as scikit-learn models.

Dataset format

By default, datasets return regular Python objects: integers, floats, strings, lists, etc..

To get NumPy arrays instead, you can set the format of the dataset to numpy:

[!TIP]

A Dataset object is a wrapper of an Arrow table, which allows fast reads from arrays in the dataset to NumPy arrays.

Note that the exact same procedure applies to <code>DatasetDict</code> objects, so that when setting the format of a <code>DatasetDict</code> to <code>numpy</code>, all the <code>Dataset</code> s there will be formatted as <code>numpy</code>:

```
>>> from datasets import DatasetDict
>>> data = {"train": {"data": [[1, 2], [3, 4]]}, "test": {"data": [[5, 6], [7, 8]]}}
>>> dds = DatasetDict.from_dict(data)
>>> dds = dds.with_format("numpy")
>>> dds["train"][:2]
{'data': array([
        [1, 2],
        [3, 4]])}
```

N-dimensional arrays

If your dataset consists of N-dimensional arrays, you will see that by default they are considered as the same array if the shape is fixed:

```
>>> from datasets import Dataset
>>> data = [[[1, 2],[3]], [[4, 5, 6],[7, 8]]] # varying shape
>>> ds = Dataset.from_dict({"data": data})
>>> ds = ds.with_format("numpy")
>>> ds[0]
{'data': array([array([1, 2]), array([3])], dtype=object)}
```

However this logic often requires slow shape comparisons and data copies.

To avoid this, you must explicitly use the Array feature type and specify the shape of your tensors:

Other feature types

ClassLabel data is properly converted to arrays:

```
>>> from datasets import Dataset, Features, ClassLabel
>>> labels = [0, 0, 1]
>>> features = Features({"label": ClassLabel(names=["negative", "positive"])})
>>> ds = Dataset.from_dict({"label": labels}, features=features)
>>> ds = ds.with_format("numpy")
>>> ds[:3]
{'label': array([0, 0, 1])}
```

String and binary objects are unchanged, since NumPy only supports numbers.

The Image and Audio feature types are also supported.

```
[!TIP]

To use the Image feature type, you'll need to install the vision extra as pip install datasets[vision].
```

```
>>> from datasets import Dataset, Features, Image
>>> images = ["path/to/image.png"] * 10
>>> features = Features({"image": Image()})
>>> ds = Dataset.from_dict({"image": images}, features=features)
>>> ds = ds.with format("numpy")
>>> ds[0]["image"].shape
(512, 512, 3)
>>> ds[0]
{'image': array([[[ 255, 255, 255],
              [ 255, 255, 255],
              . . . ,
              [ 255, 255, 255],
              [ 255, 255, 255]]], dtype=uint8)}
>>> ds[:2]["image"].shape
(2, 512, 512, 3)
>>> ds[:2]
{'image': array([[[[ 255, 255, 255],
              [ 255, 255, 255],
              [ 255, 255, 255],
              [ 255, 255, 255]]]], dtype=uint8)}
```

[!TIP]

To use the Audio feature type, you'll need to install the audio extra as pip install datasets[audio].

Data loading

NumPy doesn't have any built-in data loading capabilities, so you'll either need to materialize the NumPy arrays like x, y to use in scikit-learn or use a library such as PyTorch to load your data using a DataLoader.

Using with_format('numpy')

The easiest way to get NumPy arrays out of a dataset is to use the with_format('numpy') method. Lets assume

that we want to train a neural network on the MNIST dataset available at the HuggingFace Hub at https://huggingface.co/datasets/mnist.

Once the format is set we can feed the dataset to the model based on NumPy in batches using the Dataset.iter()
method: