
convclasses Documentation

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CONTENTS

1	convclasses	3
1.1	Features	5
1.2	Credits	5
2	Installation	7
2.1	Stable release	7
2.2	From sources	7
3	Common Usage Examples	9
3.1	Using Pendulum for Dates and Time	9
4	Converters	11
4.1	Global converter	11
4.2	Converter objects	11
5	What you can structure and how	13
5.1	Primitive values	13
5.2	Collections and other generics	14
5.3	<code>dataclasses</code> classes	17
5.4	Registering custom structuring hooks	18
6	What you can unstructure and how	21
6.1	Primitive types and collections	21
6.2	<code>dataclasses</code> classes	21
7	Modifiers	23
7.1	Change structuring/unstructuring name in dict	23
8	Contributing	25
8.1	Types of Contributions	25
8.2	Get Started!	26
8.3	Pull Request Guidelines	27
8.4	Tips	27
9	History	29
9.1	2.0.0	29
9.2	1.1.0	29
9.3	1.0.0	29
9.4	0.9.1 (2019-10-26)	29
9.5	0.9.0 (2018-07-22)	30
9.6	0.8.1 (2018-06-19)	30

9.7	0.8.0 (2018-04-14)	30
9.8	0.7.0 (2018-04-12)	30
9.9	0.6.0 (2017-12-25)	30
9.10	0.5.0 (2017-12-11)	30
9.11	0.4.0 (2017-07-17)	31
9.12	0.3.0 (2017-03-18)	31
9.13	0.2.0 (2016-10-02)	31
9.14	0.1.0 (2016-08-13)	31
10	Indices and tables	33

Contents:

CONVCLASSES

`convclasses` is an open source Python library for structuring and unstructuring data. `convclasses` works best with `dataclasses` classes and the usual Python collections, but other kinds of classes are supported by manually registering converters.

Python has a rich set of powerful, easy to use, built-in data types like dictionaries, lists and tuples. These data types are also the lingua franca of most data serialization libraries, for formats like json, msgpack, yaml or toml.

Data types like this, and mappings like `dict`s in particular, represent unstructured data. Your data is, in all likelihood, structured: not all combinations of field names and values are valid inputs to your programs. In Python, structured data is better represented with classes and enumerations. `dataclasses` is an excellent library for declaratively describing the structure of your data, and validating it.

When you're handed unstructured data (by your network, file system, database...), `convclasses` helps to convert this data into structured data. When you have to convert your structured data into data types other libraries can handle, `convclasses` turns your classes and enumerations into dictionaries, integers and strings.

Here's a simple taste. The list containing a float, an int and a string gets converted into a tuple of three ints.

```
>>> import convclasses
>>> from typing import Tuple
>>>
>>> convclasses.structure([1.0, 2, "3"], Tuple[int, int, int])
(1, 2, 3)
```

`convclasses` works well with `dataclasses` classes out of the box.

```
>>> import convclasses
>>> from dataclasses import dataclass
>>> from typing import Any
>>> @dataclass(frozen=True) # It works with normal classes too.
... class C:
...     a: Any
...     b: Any
```

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```
...
>>> instance = C(1, 'a')
>>> convclasses.unstructure(instance)
{'a': 1, 'b': 'a'}
>>> convclasses.structure({'a': 1, 'b': 'a'}, C)
C(a=1, b='a')
```

Here's a much more complex example, involving dataclasses classes with type metadata.

```
>>> from enum import unique, Enum
>>> from typing import Any, List, Optional, Sequence, Union
>>> from convclasses import structure, unstructure
>>> from dataclasses import dataclass
>>>
>>> @unique
... class CatBreed(Enum):
...     SIAMESE = "siamese"
...     MAINE_COON = "maine_coon"
...     SACRED_BIRMAN = "birman"
...
>>> @dataclass
... class Cat:
...     breed: CatBreed
...     names: Sequence[str]
...
>>> @dataclass
... class DogMicrochip:
...     chip_id: Any
...     time_chipped: float
...
>>> @dataclass
... class Dog:
...     cuteness: int
...     chip: Optional[DogMicrochip]
...
>>> p = unstructure([Dog(cuteness=1, chip=DogMicrochip(chip_id=1, time_chipped=10.0)),
...                  Cat(breed=CatBreed.MAINE_COON, names=('Fluffly', 'Fluffer'))])
...
>>> print(p)
[{'cuteness': 1, 'chip': {'chip_id': 1, 'time_chipped': 10.0}, {'breed': 'maine_coon',
→ 'names': ('Fluffly', 'Fluffer')}]
>>> print(structure(p, List[Union[Dog, Cat]]))
[Dog(cuteness=1, chip=DogMicrochip(chip_id=1, time_chipped=10.0)), Cat(breed=
→ CatBreed.MAINE_COON: 'maine_coon', names=['Fluffly', 'Fluffer'])]
```

Consider unstructured data a low-level representation that needs to be converted to structured data to be handled, and use `structure`. When you're done, `unstructure` the data to its unstructured form and pass it along to another library or module. Use `dataclasses` `type metadata` to add type metadata to attributes, so `convclasses` will know how to structure and destructure them.

- Free software: MIT license
- Documentation: <https://convclasses.readthedocs.io>.
- Python versions supported: 3.7 and up.

1.1 Features

- Converts structured data into unstructured data, recursively:
 - `dataclasses` classes are converted into dictionaries in a way similar to `dataclasses.asdict`, or into tuples in a way similar to `dataclasses.astuple`.
 - Enumeration instances are converted to their values.
 - Other types are let through without conversion. This includes types such as integers, dictionaries, lists and instances of non-`dataclasses` classes.
 - Custom converters for any type can be registered using `register_unstructure_hook`.
- Converts unstructured data into structured data, recursively, according to your specification given as a type. The following types are supported:
 - `typing.Optional[T]`.
 - `typing.List[T]`, `typing.MutableSequence[T]`, `typing.Sequence[T]` (converts to a list).
 - `typing.Tuple` (both variants, `Tuple[T, ...]` and `Tuple[X, Y, Z]`).
 - `typing.MutableSet[T]`, `typing.Set[T]` (converts to a set).
 - `typing.FrozenSet[T]` (converts to a frozenset).
 - `typing.Dict[K, V]`, `typing.MutableMapping[K, V]`, `typing.Mapping[K, V]` (converts to a dict).
 - `dataclasses` classes with simple attributes and the usual `__init__`.
 - * Simple attributes are attributes that can be assigned unstructured data, like numbers, strings, and collections of unstructured data.
 - All `dataclasses` classes with the usual `__init__`, if their complex attributes have type metadata.
 - `typing.Union`s of supported `dataclasses` classes, given that all of the classes have a unique field.
 - `typing.Union`s of anything, given that you provide a disambiguation function for it.
 - Custom converters for any type can be registered using `register_structure_hook`.

1.2 Credits

Major credits and best wishes for the original creator of this concept - [Tinche](#), he developed [cattrs](#) which this project is fork of.

Major credits to Hynek Schlawack for creating [attrs](#) and its predecessor, [characteristic](#).

`convclasses` is tested with [Hypothesis](#), by David R. MacIver.

`convclasses` is benchmarked using [perf](#), by Victor Stinner.

INSTALLATION

2.1 Stable release

To install convclasses, run this command in your terminal:

```
$ pip install convclasses
```

This is the preferred method to install convclasses, as it will always install the most recent stable release.

If you don't have `pip` installed, this [Python installation guide](#) can guide you through the process.

2.2 From sources

The sources for convclasses can be downloaded from the [Github repo](#).

You can either clone the public repository:

```
$ git clone git://github.com/zeburek/convclasses
```

Or download the [tarball](#):

```
$ curl -OL https://github.com/zeburek/convclasses/tarball/master
```

Once you have a copy of the source, you can install it with:

```
$ pip install -e .
```


COMMON USAGE EXAMPLES

This section covers common use examples of convclasses features.

3.1 Using Pendulum for Dates and Time

To use the excellent [Arrow](#) library for datetimes, we need to register structuring and unstructuring hooks for it.

First, we need to decide on the unstructured representation of a datetime instance. Since all our datetimes will use the UTC time zone, we decide to use the UNIX epoch timestamp as our unstructured representation.

Define a class using Arrow's `DateTime`:

```
import arrow
from arrow import Arrow

@dataclass
class MyRecord:
    a_string: str
    a_datetime: Arrow
```

Next, we register hooks for the `Arrow` class on a new `Converter` instance.

```
converter = Converter()

converter.register_unstructure_hook(Arrow, lambda ar: ar.float_timestamp)

converter.register_structure_hook(Arrow, lambda ts, _: arrow.get(ts))
```

And we can proceed with unstructuring and structuring instances of `MyRecord`.

```
>>> my_record = MyRecord('test', arrow.Arrow(2018, 7, 28, 18, 24))
>>> my_record
MyRecord(a_string='test', a_datetime=<Arrow [2018-07-28T18:24:00+00:00]>)

>>> converter.unstructure(my_record)
{'a_string': 'test', 'a_datetime': 1532802240.0}

>>> converter.structure({'a_string': 'test', 'a_datetime': 1532802240.0}, MyRecord)
MyRecord(a_string='test', a_datetime=<Arrow [2018-07-28T18:24:00+00:00]>)
```

After a while, we realize we *will* need our datetimes to have timezone information. We decide to switch to using the ISO 8601 format for our unstructured datetime instances.

```
>>> converter = convclasses.Converter()
>>> converter.register_unstructure_hook(Arrow, lambda dt: dt.isoformat())
>>> converter.register_structure_hook(Arrow, lambda isostring, _: arrow.
↳ get(isostring))

>>> my_record = MyRecord('test', arrow.Arrow(2018, 7, 28, 18, 24, tzinfo='Europe/Paris
↳ '))
>>> my_record
MyRecord(a_string='test', a_datetime=<Arrow [2018-07-28T18:24:00+02:00]>)

>>> converter.unstructure(my_record)
{'a_string': 'test', 'a_datetime': '2018-07-28T18:24:00+02:00'}

>>> converter.structure({'a_string': 'test', 'a_datetime': '2018-07-28T18:24:00+02:00
↳ '}, MyRecord)
MyRecord(a_string='test', a_datetime=<Arrow [2018-07-28T18:24:00+02:00]>)
```

CONVERTERS

All `convclasses` functionality is exposed through a `convclasses.Converter` object. Global `convclasses` functions, such as `convclasses.unstructure()`, use a single global converter. Changes done to this global converter, such as registering new `structure` and `unstructure` hooks, affect all code using the global functions.

4.1 Global converter

A global converter is provided for convenience as `convclasses.global_converter`. The following functions implicitly use this global converter:

- `convclasses.structure`
- `convclasses.unstructure`
- `convclasses.structure_dataclass_fromtuple`
- `convclasses.structure_dataclass_fromdict`

Changes made to the global converter will affect the behavior of these functions.

Larger applications are strongly encouraged to create and customize a different, private instance of `Converter`.

4.2 Converter objects

To create a private converter, simply instantiate a `convclasses.Converter`. Currently, a converter contains the following state:

- a registry of `unstructure` hooks, backed by a `singledispatch` and a `function_dispatch`.
- a registry of `structure` hooks, backed by a different `singledispatch` and `function_dispatch`.
- a LRU cache of union disambiguation functions.
- a reference to an unstructuring strategy (either `AS_DICT` or `AS_TUPLE`).
- a `dict_factory` callable, used for creating `dicts` when dumping `dataclasses` classes using `AS_DICT`.

WHAT YOU CAN STRUCTURE AND HOW

The philosophy of `convclasses` structuring is simple: give it an instance of Python built-in types and collections, and a type describing the data you want out. `convclasses` will convert the input data into the type you want, or throw an exception.

All loading conversions are composable, where applicable. This is demonstrated further in the examples.

5.1 Primitive values

5.1.1 `typing.Any`

Use `typing.Any` to avoid applying any conversions to the object you're loading; it will simply be passed through.

```
>>> convclasses.structure(1, Any)
1
>>> d = {1: 1}
>>> convclasses.structure(d, Any) is d
True
```

5.1.2 `int`, `float`, `str`, `bytes`

Use any of these primitive types to convert the object to the type.

```
>>> convclasses.structure(1, str)
'1'
>>> convclasses.structure("1", float)
1.0
```

In case the conversion isn't possible, the expected exceptions will be propagated out. The particular exceptions are the same as if you'd tried to do the conversion yourself, directly.

```
>>> convclasses.structure("not-an-int", int)
Traceback (most recent call last):
...
ValueError: invalid literal for int() with base 10: 'not-an-int'
```

5.1.3 Enums

Enums will be structured by their values. This works even for complex values, like tuples.

```
>>> @unique
... class CatBreed(Enum):
...     SIAMESE = "siamese"
...     MAINE_COON = "maine_coon"
...     SACRED_BIRMAN = "birman"
...
>>> convclasses.structure("siamese", CatBreed)
<CatBreed.SIAMESE: 'siamese'>
```

Again, in case of errors, the expected exceptions will fly out.

```
>>> convclasses.structure("alsatian", CatBreed)
Traceback (most recent call last):
...
ValueError: 'alsatian' is not a valid CatBreed
```

5.2 Collections and other generics

5.2.1 Optionals

Optional primitives and collections are supported out of the box.

```
>>> convclasses.structure(None, int)
Traceback (most recent call last):
...
TypeError: int() argument must be a string, a bytes-like object or a number, not
↳ 'NoneType'
>>> convclasses.structure(None, Optional[int])
>>> # None was returned.
```

Bare Optional s (non-parameterized, just Optional, as opposed to Optional[str]) aren't supported, use Optional[Any] instead.

This generic type is composable with all other converters.

```
>>> convclasses.structure(1, Optional[float])
1.0
```

5.2.2 Lists

Lists can be produced from any iterable object. Types converting to lists are:

- Sequence[T]
- MutableSequence[T]
- List[T]

In all cases, a new list will be returned, so this operation can be used to copy an iterable into a list. A bare type, for example Sequence instead of Sequence[int], is equivalent to Sequence[Any].

```
>>> convclasses.structure([1, 2, 3], MutableSequence[int])
[1, 2, 3]
```

These generic types are composable with all other converters.

```
>>> convclasses.structure([1, None, 3], List[Optional[str]])
['1', None, '3']
```

5.2.3 Sets and frozensets

Sets and frozensets can be produced from any iterable object. Types converting to sets are:

- `Set[T]`
- `MutableSet[T]`

Types converting to frozensets are:

- `FrozenSet[T]`

In all cases, a new set or frozenset will be returned, so this operation can be used to copy an iterable into a set. A bare type, for example `MutableSet` instead of `MutableSet[int]`, is equivalent to `MutableSet[Any]`.

```
>>> convclasses.structure([1, 2, 3, 4], Set)
{1, 2, 3, 4}
```

These generic types are composable with all other converters.

```
>>> convclasses.structure([[1, 2], [3, 4]], Set[FrozenSet[str]])
{frozenset({'1', '2'}), frozenset({'4', '3'})}
```

5.2.4 Dictionaries

Dicts can be produced from other mapping objects. To be more precise, the object being converted must expose an `items()` method producing an iterable key-value tuples, and be able to be passed to the `dict` constructor as an argument. Types converting to dictionaries are:

- `Dict[K, V]`
- `MutableMapping[K, V]`
- `Mapping[K, V]`

In all cases, a new dict will be returned, so this operation can be used to copy a mapping into a dict. Any type parameters set to `typing.Any` will be passed through unconverted. If both type parameters are absent, they will be treated as `Any` too.

```
>>> from collections import OrderedDict
>>> convclasses.structure(OrderedDict([(1, 2), (3, 4)]), Dict)
{1: 2, 3: 4}
```

These generic types are composable with all other converters. Note both keys and values can be converted.

```
>>> convclasses.structure({1: None, 2: 2.0}, Dict[str, Optional[int]])
{'1': None, '2': 2}
```

5.2.5 Homogeneous and heterogeneous tuples

Homogeneous and heterogeneous tuples can be produced from iterable objects. Heterogeneous tuples require an iterable with the number of elements matching the number of type parameters exactly. Use:

- `Tuple[A, B, C, D]`

Homogeneous tuples use:

- `Tuple[T, ...]`

In all cases a tuple will be returned. Any type parameters set to `typing.Any` will be passed through unconverted.

```
>>> convclasses.structure([1, 2, 3], Tuple[int, str, float])
(1, '2', 3.0)
```

The tuple conversion is composable with all other converters.

```
>>> convclasses.structure([{'1': 1}, {'2': 2}], Tuple[Dict[str, float], ...])
({'1': 1.0}, {'2': 2.0})
```

5.2.6 Unions

Unions of `NoneType` and a single other type are supported (also known as `Optional`s). All other unions require a disambiguation function.

Automatic Disambiguation

In the case of a union consisting exclusively of `dataclasses` classes, `convclasses` will attempt to generate a disambiguation function automatically; this will succeed only if each class has a unique field. Given the following classes:

```
>>> @dataclass
... class A:
...     a: Any
...     x: Any
...
>>> @dataclass
... class B:
...     a: Any
...     y: Any
...
>>> @dataclass
... class C:
...     a: Any
...     z: Any
...
```

`convclasses` can deduce only instances of `A` will contain `x`, only instances of `B` will contain `y`, etc. A disambiguation function using this information will then be generated and cached. This will happen automatically, the first time an appropriate union is structured.

Manual Disambiguation

To support arbitrary unions, register a custom structuring hook for the union (see *Registering custom structuring hooks*).

5.3 dataclasses classes

5.3.1 Simple dataclasses classes

dataclasses classes using primitives, collections of primitives and their own converters work out of the box. Given a mapping `d` and class `A`, convclasses will simply instantiate `A` with `d` unpacked.

```
>>> @dataclass
... class A:
...     a: Any
...     b: int
...
>>> convclasses.structure({'a': 1, 'b': '2'}, A)
A(a=1, b=2)
```

dataclasses classes deconstructed into tuples can be structured using convclasses.structure_attrs_fromtuple (fromtuple as in the opposite of dataclasses.astuple and converter.unstructure_attrs_astuple).

```
>>> # Loading from tuples can be made the default by creating a new
... @dataclass
... class A:
...     a: Any
...     b: int
...
>>> convclasses.structure_dataclass_fromtuple(['string', '2'], A)
A(a='string', b=2)
```

Loading from tuples can be made the default by creating a new Converter with unstruct_strat=convclasses.UnstructureStrategy.AS_TUPLE.

```
>>> converter = convclasses.Converter(unstruct_strat=convclasses.UnstructureStrategy.
↪AS_TUPLE)
>>> @dataclass
... class A:
...     a: Any
...     b: int
...
>>> converter.structure(['string', '2'], A)
A(a='string', b=2)
```

Structuring from tuples can also be made the default for specific classes only; see registering custom structure hooks below.

5.3.2 Complex dataclasses classes

Complex `dataclasses` classes are classes with type information available for some or all attributes. These classes support almost arbitrary nesting.

Type information can be set using type annotations when using Python 3.6+.

```
>>> @dataclass
... class A:
...     a: int
...
>>> fields(A)
(Field(name='a', type=<class 'int'>, default=<dataclasses._MISSING_TYPE object at 0x...>
↪, default_factory=<dataclasses._MISSING_TYPE object at 0x...>, init=True, repr=True,
↪hash=None, compare=True, metadata=mappingproxy({}), _field_type=_FIELD),)
```

Type information, when provided, can be used for all attribute types, not only attributes holding `dataclasses` classes.

```
>>> @dataclass
... class A:
...     a: int = field(default=0)
...
>>> @dataclass
... class B:
...     b: A
...
>>> convclasses.structure({'b': {'a': '1'}}, B)
B(b=A(a=1))
```

5.4 Registering custom structuring hooks

`convclasses` doesn't know how to structure non-`dataclasses` classes by default, so it has to be taught. This can be done by registering structuring hooks on a converter instance (including the global converter).

Here's an example involving a simple, classic (i.e. non-`dataclasses`) Python class.

```
>>> class C(object):
...     def __init__(self, a):
...         self.a = a
...     def __repr__(self):
...         return f'C(a={self.a})'
>>> convclasses.structure({'a': 1}, C)
Traceback (most recent call last):
...
ValueError: Unsupported type: <class '__main__.C'>. Register a structure hook for it.
>>> convclasses.register_structure_hook(C, lambda d, t: C(**d))
>>> convclasses.structure({'a': 1}, C)
C(a=1)
```

The structuring hooks are callables that take two arguments: the object to convert to the desired class and the type to convert to. The type may seem redundant but is useful when dealing with generic types.

When using `convclasses.register_structure_hook`, the hook will be registered on the global converter. If you want to avoid changing the global converter, create an instance of `convclasses.Converter` and register the hook on that.

In some situations, it is not possible to decide on the converter using typing mechanisms alone (such as with dataclasses classes). In these situations, convclasses provides a `register_structure_func_hook` instead, which accepts a function to determine whether that type can be handled instead.

The function-based hooks are evaluated after the class-based hooks. In the case where both a class-based hook and a function-based hook are present, the class-based hook will be used.

```
>>> class D(object):
...     custom = True
...     def __init__(self, a):
...         self.a = a
...     def __repr__(self):
...         return f'D(a={self.a})'
...     @classmethod
...     def deserialize(cls, data):
...         return cls(data["a"])
>>> convclasses.register_structure_hook_func(lambda cls: getattr(cls, "custom",
↪False), lambda d, t: t.deserialize(d))
>>> convclasses.structure({'a': 2}, D)
D(a=2)
```


WHAT YOU CAN UNSTRUCTURE AND HOW

Unstructuring is intended to convert high-level, structured Python data (like instances of complex classes) into simple, unstructured data (like dictionaries).

Unstructuring is simpler than structuring in that no target types are required. Simply provide an argument to `unstructure` and `convclasses` will produce a result based on the registered unstructuring hooks. A number of default unstructuring hooks are documented [here](#).

Unstructuring is primarily done using `Converter.unstructure`.

6.1 Primitive types and collections

Primitive types (integers, floats, strings...) are simply passed through. Collections are copied. There's relatively little value in unstructuring these types directly as they are already unstructured and third-party libraries tend to support them directly.

A useful use case for unstructuring collections is to create a deep copy of a complex or recursive collection.

```
>>> # A dictionary of strings to lists of tuples of floats.
>>> data = {'a': [(1.0, 2.0), (3.0, 4.0)]}
>>>
>>> copy = convclasses.unstructure(data)
>>> data == copy
True
>>> data is copy
False
```

6.2 dataclasses classes

`dataclasses` classes are supported out of the box. `Converter`s support two unstructuring strategies:

- `UnstructureStrategy.AS_DICT` - similar to `dataclasses.asdict`, unstructures `dataclasses` instances into dictionaries. This is the default.
- `UnstructureStrategy.AS_TUPLE` - similar to `dataclasses.astuple`, unstructures `dataclasses` instances into tuples.

```
>>> @dataclass
... class C:
...     a: Any
...     b: Any
```

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```
...
>>> inst = C(1, 'a')
>>> converter = convclasses.Converter(unstruct_strat=convclasses.UnstructureStrategy.
↳AS_TUPLE)
>>> converter.unstructure(inst)
(1, 'a')
```

6.2.1 Mixing and matching strategies

Converters publicly expose two helper methods, `Converter.unstructure_attrs_asdict()` and `Converter.unstructure_attrs_astuple()`. These methods can be used with custom unstructuring hooks to selectively apply one strategy to instances of particular classes.

Assume two nested dataclasses classes, `Inner` and `Outer`; instances of `Outer` contain instances of `Inner`. Instances of `Outer` should be unstructured as dictionaries, and instances of `Inner` as tuples. Here's how to do this.

```
>>> @dataclass
... class Inner:
...     a: int
...
>>> @dataclass
... class Outer:
...     i: Inner
...
>>> inst = Outer(i=Inner(a=1))
>>>
>>> converter = convclasses.Converter()
>>> converter.register_unstructure_hook(Inner, converter.unstructure_dataclass_
↳astuple)
>>>
>>> converter.unstructure(inst)
{'i': (1,)}
```

Of course, these methods can be used directly as well, without changing the converter strategy.

```
>>> @dataclass
... class C:
...     a: int
...     b: str
...
>>> inst = C(1, 'a')
>>>
>>> converter = convclasses.Converter()
>>>
>>> converter.unstructure_dataclass_astuple(inst) # Default is AS_DICT.
(1, 'a')
```

MODIFIERS

7.1 Change structuring/unstructuring name in dict

Sometimes you may face a problem, when you need to structure data, that has fields that could not be created in python:

```
>>> @dataclass
... class TestModel:
...     user_agent: str
...
>>> convclasses.structure({"User-Agent": "curl/7.71.1"}, TestModel)
Traceback (most recent call last):
...
TypeError: __init__() missing 1 required positional argument: 'user_agent'
```

Now you can avoid this, and also use convclasses with such structures:

```
>>> from convclasses import mod
>>> @dataclass
... class TestModel:
...     user_agent: str = mod.name("User-Agent")
...     other_field: str = mod.name("other field", field(default=None))
...
>>> obj = convclasses.structure({"User-Agent": "curl/7.71.1"}, TestModel)
>>> obj
TestModel(user_agent='curl/7.71.1', other_field=None)
>>> convclasses.unstructure(obj)
{'User-Agent': 'curl/7.71.1', 'other field': None}
```


CONTRIBUTING

Contributions are welcome, and they are greatly appreciated! Every little bit helps, and credit will always be given. You can contribute in many ways:

8.1 Types of Contributions

8.1.1 Report Bugs

Report bugs at <https://github.com/zeburek/convclasses/issues>.

If you are reporting a bug, please include:

- Your operating system name and version.
- Any details about your local setup that might be helpful in troubleshooting.
- Detailed steps to reproduce the bug.

8.1.2 Fix Bugs

Look through the GitHub issues for bugs. Anything tagged with “bug” and “help wanted” is open to whoever wants to implement it.

8.1.3 Implement Features

Look through the GitHub issues for features. Anything tagged with “enhancement” and “help wanted” is open to whoever wants to implement it.

8.1.4 Write Documentation

convclasses could always use more documentation, whether as part of the official convclasses docs, in docstrings, or even on the web in blog posts, articles, and such.

8.1.5 Submit Feedback

The best way to send feedback is to file an issue at <https://github.com/zeburek/convclasses/issues>.

If you are proposing a feature:

- Explain in detail how it would work.
- Keep the scope as narrow as possible, to make it easier to implement.
- Remember that this is a volunteer-driven project, and that contributions are welcome :)

8.2 Get Started!

Ready to contribute? Here's how to set up *convclasses* for local development.

1. Fork the *convclasses* repo on GitHub.
2. Clone your fork locally:

```
$ git clone git@github.com:your_name_here/convclasses.git
```

3. Install your local copy into a virtualenv. Assuming you have virtualenvwrapper installed, this is how you set up your fork for local development:

```
$ mkvirtualenv convclasses
$ cd convclasses/
$ pip install -e .\[dev\]
```

4. Create a branch for local development:

```
$ git checkout -b name-of-your-bugfix-or-feature
```

Now you can make your changes locally.

5. When you're done making changes, check that your changes pass flake8 and the tests, including testing other Python versions with tox:

```
$ flake8 convclasses tests
$ pytest
$ tox
```

To get flake8 and tox, just pip install them into your virtualenv.

6. Commit your changes and push your branch to GitHub:

```
$ git add .
$ git commit -m "Your detailed description of your changes."
$ git push origin name-of-your-bugfix-or-feature
```

7. Submit a pull request through the GitHub website.

8.3 Pull Request Guidelines

Before you submit a pull request, check that it meets these guidelines:

1. The pull request should include tests.
2. If the pull request adds functionality, the docs should be updated. Put your new functionality into a function with a docstring, and add the feature to the list in README.rst.
3. The pull request should work for all supported Python versions. Make sure that the tests pass for all supported Python versions.

8.4 Tips

To run a subset of tests:

```
$ pytest tests.test_unstructure
```


HISTORY

9.1 2.0.0

- **Add support for modifiers**
 - Add dataclass field name modifier
- Add support for generic types (ported from `cattrs`)

9.2 1.1.0

- Removed Python 3.6 support
- Added Python 3.9 support

9.3 1.0.0

- Rename `cattrs` into `conclasses`
- Move `conclasses` from `attrs` usage onto `dataclasses`
- Fix incorrect structuring/unstructuring of private fields
- Change `pendulum` in docs onto `arrow`

9.3.1 `cattrs` history

9.4 0.9.1 (2019-10-26)

- Python 3.8 support.

9.5 0.9.0 (2018-07-22)

- Python 3.7 support.

9.6 0.8.1 (2018-06-19)

- The disambiguation function generator now supports unions of `attrs` classes and `NoneType`.

9.7 0.8.0 (2018-04-14)

- Distribution fix.

9.8 0.7.0 (2018-04-12)

- Removed the undocumented `Converter.unstruct_strat` property setter.
- Removed the ability to set the `Converter.structure_attrs` instance field. As an alternative, create a new `Converter`:

```
>>> converter = cattr.Converter(unstruct_strat=cattr.UnstructureStrategy.AS_TUPLE)
```

- Some micro-optimizations were applied; a `structure(unstructure(obj))` roundtrip is now up to 2 times faster.

9.9 0.6.0 (2017-12-25)

- Packaging fixes. (#17)

9.10 0.5.0 (2017-12-11)

- `structure/unstructure` now supports using functions as well as classes for deciding the appropriate function.
- added `Converter.register_structure_hook_func`, to register a function instead of a class for determining handler func.
- added `Converter.register_unstructure_hook_func`, to register a function instead of a class for determining handler func.
- vendored typing is no longer needed, nor provided.
- Attributes with default values can now be structured if they are missing in the input. (#15)
- *Optional* attributes can no longer be structured if they are missing in the input.

In other words, this no longer works:

```
@attr.s
class A:
    a: Optional[int] = attr.ib()

>>> cattr.structure({}, A)
```

- `cattr.typed` removed since the functionality is now present in `attrs` itself. Replace instances of `cattr.typed(type)` with `attr.ib(type=type)`.

9.11 0.4.0 (2017-07-17)

- *Converter.loads* is now *Converter.structure*, and *Converter.dumps* is now *Converter.unstructure*.
- Python 2.7 is supported.
- Moved `cattr.typing` to `cattr.vendor.typing` to support different vendored versions of `typing.py` for Python 2 and Python 3.
- Type metadata can be added to `attrs` classes using `cattr.typed`.

9.12 0.3.0 (2017-03-18)

- Python 3.4 is no longer supported.
- Introduced `cattr.typing` for use with Python versions 3.5.2 and 3.6.0.
- Minor changes to work with newer versions of `typing`.
 - Bare Optionals are not supported any more (use `Optional[Any]`).
- Attempting to load unrecognized classes will result in a `ValueError`, and a helpful message to register a loads hook.
- Loading `attrs` classes is now documented.
- The global converter is now documented.
- `cattr.loads_attrs_fromtuple` and `cattr.loads_attrs_fromdict` are now exposed.

9.13 0.2.0 (2016-10-02)

- Tests and documentation.

9.14 0.1.0 (2016-08-13)

- First release on PyPI.

INDICES AND TABLES

- `genindex`
- `modindex`
- `search`