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# **fastavro Documentation**

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The current Python *avro* package is dog slow.

On a test case of about 10K records, it takes about 14sec to iterate over all of them. In comparison the JAVA *avro* SDK does it in about 1.9sec.

*fastavro* is an alternative implementation that is much faster. It iterates over the same 10K records in 2.9sec, and if you use it with PyPy it'll do it in 1.5sec (to be fair, the JAVA benchmark is doing some extra JSON encoding/decoding).

If the optional C extension (generated by [Cython](#)) is available, then *fastavro* will be even faster. For the same 10K records it'll run in about 1.7sec.



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## Supported Features

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- File Writer
- File Reader (iterating via records or blocks)
- Schemaless Writer
- Schemaless Reader
- JSON Writer
- JSON Reader
- Codecs (Snappy, Deflate, Zstandard, Bzip2, LZ4, XZ)
- Schema resolution
- Aliases
- Logical Types
- Parsing schemas into the canonical form
- Schema fingerprinting





## CHAPTER 2

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### Missing Features

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- Anything involving Avro's RPC features



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## Example

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```
from fastavro import writer, reader, parse_schema

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}
parsed_schema = parse_schema(schema)

# 'records' can be an iterable (including generator)
records = [
    {'station': u'011990-99999', 'temp': 0, 'time': 1433269388},
    {'station': u'011990-99999', 'temp': 22, 'time': 1433270389},
    {'station': u'011990-99999', 'temp': -11, 'time': 1433273379},
    {'station': u'012650-99999', 'temp': 111, 'time': 1433275478},
]

# Writing
with open('weather.avro', 'wb') as out:
    writer(out, parsed_schema, records)

# Reading
with open('weather.avro', 'rb') as fo:
    for record in reader(fo):
        print(record)
```



## CHAPTER 4

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### Documentation

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## 5.1 fastavro.read

**class reader** (*fo*: Union[IO, fastavro.io.json\_decoder.AvroJSONDecoder], *reader\_schema*: Union[str, List[T], Dict[KT, VT], None] = None, *return\_record\_name*: bool = False)  
Iterator over records in an avro file.

### Parameters

- **fo** – File-like object to read from
- **reader\_schema** – Reader schema
- **return\_record\_name** – If true, when reading a union of records, the result will be a tuple where the first value is the name of the record and the second value is the record itself

Example:

```
from fastavro import reader
with open('some-file.avro', 'rb') as fo:
    avro_reader = reader(fo)
    for record in avro_reader:
        process_record(record)
```

The *fo* argument is a file-like object so another common example usage would use an *io.BytesIO* object like so:

```
from io import BytesIO
from fastavro import writer, reader

fo = BytesIO()
writer(fo, schema, records)
fo.seek(0)
for record in reader(fo):
    process_record(record)
```

### metadata

Key-value pairs in the header metadata

**codec**

The codec used when writing

**writer\_schema**

The schema used when writing

**reader\_schema**

The schema used when reading (if provided)

**class block\_reader** (*fo: IO, reader\_schema: Union[str, List[T], Dict[KT, VT], None] = None, return\_record\_name: bool = False*)

Iterator over *Block* in an avro file.

**Parameters**

- **fo** – Input stream
- **reader\_schema** – Reader schema
- **return\_record\_name** – If true, when reading a union of records, the result will be a tuple where the first value is the name of the record and the second value is the record itself

Example:

```
from fastavro import block_reader
with open('some-file.avro', 'rb') as fo:
    avro_reader = block_reader(fo)
    for block in avro_reader:
        process_block(block)
```

**metadata**

Key-value pairs in the header metadata

**codec**

The codec used when writing

**writer\_schema**

The schema used when writing

**reader\_schema**

The schema used when reading (if provided)

**class Block** (*bytes\_, num\_records, codec, reader\_schema, writer\_schema, named\_schemas, offset, size, return\_record\_name=False*)

An avro block. Will yield records when iterated over

**num\_records**

Number of records in the block

**writer\_schema**

The schema used when writing

**reader\_schema**

The schema used when reading (if provided)

**offset**

Offset of the block from the beginning of the avro file

**size**

Size of the block in bytes



**schemaless\_reader** (*fo: IO, writer\_schema: Union[str, List[T], Dict[KT, VT]], reader\_schema: Union[str, List[T], Dict[KT, VT], None] = None, return\_record\_name: bool = False*) → Union[None, str, float, int, decimal.Decimal, bool, bytes, List[T], Dict[KT, VT]]

Reads a single record written using the `schemaless_writer()`

#### Parameters

- **fo** – Input stream
- **writer\_schema** – Schema used when calling `schemaless_writer`
- **reader\_schema** – If the schema has changed since being written then the new schema can be given to allow for schema migration
- **return\_record\_name** – If true, when reading a union of records, the result will be a tuple where the first value is the name of the record and the second value is the record itself

Example:

```
parsed_schema = fastavro.parse_schema(schema)
with open('file', 'rb') as fp:
    record = fastavro.schemaless_reader(fp, parsed_schema)
```

Note: The `schemaless_reader` can only read a single record.

**is\_avro** (*path\_or\_buffer: Union[str, IO]*) → bool

Return True if path (or buffer) points to an Avro file. This will only work for avro files that contain the normal avro schema header like those create from `writer()`. This function is not intended to be used with binary data created from `schemaless_writer()` since that does not include the avro header.

**Parameters** `path_or_buffer` – Path to file

## 5.2 fastavro.write

**writer** (*fo: Union[IO, fastavro.io.json\_encoder.AvroJSONEncoder], schema: Union[str, List[T], Dict[KT, VT]], records: Iterable[Any], codec: str = 'null', sync\_interval: int = 16000, metadata: Optional[Dict[str, str]] = None, validator: bool = False, sync\_marker: bytes = b'', codec\_compression\_level: Optional[int] = None*)

Write records to fo (stream) according to schema

#### Parameters

- **fo** – Output stream
- **schema** – Writer schema
- **records** – Records to write. This is commonly a list of the dictionary representation of the records, but it can be any iterable
- **codec** – Compression codec, can be 'null', 'deflate' or 'snappy' (if installed)
- **sync\_interval** – Size of sync interval
- **metadata** – Header metadata
- **validator** – If true, validation will be done on the records
- **sync\_marker** – A byte string used as the avro sync marker. If not provided, a random byte string will be used.

- `codec_compression_level` – Compression level to use with the specified codec (if the codec supports it)

Example:

```
from fastavro import writer, parse_schema

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}

parsed_schema = parse_schema(schema)

records = [
    {'station': u'011990-99999', 'temp': 0, 'time': 1433269388},
    {'station': u'011990-99999', 'temp': 22, 'time': 1433270389},
    {'station': u'011990-99999', 'temp': -11, 'time': 1433273379},
    {'station': u'012650-99999', 'temp': 111, 'time': 1433275478},
]

with open('weather.avro', 'wb') as out:
    writer(out, parsed_schema, records)
```

The `fo` argument is a file-like object so another common example usage would use an `io.BytesIO` object like so:

```
from io import BytesIO
from fastavro import writer

fo = BytesIO()
writer(fo, schema, records)
```

Given an existing avro file, it's possible to append to it by re-opening the file in `a+b` mode. If the file is only opened in `ab` mode, we aren't able to read some of the existing header information and an error will be raised. For example:

```
# Write initial records
with open('weather.avro', 'wb') as out:
    writer(out, parsed_schema, records)

# Write some more records
with open('weather.avro', 'a+b') as out:
    writer(out, None, more_records)
```

Note: When appending, any schema provided will be ignored since the schema in the avro file will be re-used. Therefore it is convenient to just use `None` as the schema.

**schemaless\_writer** (*fo: IO, schema: Union[str, List[T], Dict[KT, VT]], record: Any*)

Write a single record without the schema or header information

#### Parameters

- `fo` – Output file

- **schema** – Schema
- **record** – Record to write

Example:

```
parsed_schema = fastavro.parse_schema(schema)
with open('file', 'wb') as fp:
    fastavro.schemaless_writer(fp, parsed_schema, record)
```

Note: The `schemaless_writer` can only write a single record.

## 5.2.1 Using the tuple notation to specify which branch of a union to take

Since this library uses plain dictionaries to represent a record, it is possible for that dictionary to fit the definition of two different records.

For example, given a dictionary like this:

```
{"name": "My Name"}
```

It would be valid against both of these records:

```
child_schema = {
    "name": "Child",
    "type": "record",
    "fields": [
        {"name": "name", "type": "string"},
        {"name": "favorite_color", "type": ["null", "string"]},
    ]
}

pet_schema = {
    "name": "Pet",
    "type": "record",
    "fields": [
        {"name": "name", "type": "string"},
        {"name": "favorite_toy", "type": ["null", "string"]},
    ]
}
```

This becomes a problem when a schema contains a union of these two similar records as it is not clear which record the dictionary represents. For example, if you used the previous dictionary with the following schema, it wouldn't be clear if the record should be serialized as a *Child* or a *Pet*:

```
household_schema = {
    "name": "Household",
    "type": "record",
    "fields": [
        {"name": "address", "type": "string"},
        {
            "name": "family_members",
            "type": {
                "type": "array", "items": [
                    {
                        "name": "Child",
                        "type": "record",

```

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```

        "fields": [
            {"name": "name", "type": "string"},
            {"name": "favorite_color", "type": ["null", "string"]},
        ]
    }, {
        "name": "Pet",
        "type": "record",
        "fields": [
            {"name": "name", "type": "string"},
            {"name": "favorite_toy", "type": ["null", "string"]},
        ]
    }
]
},
]
}

```

To resolve this, you can use a tuple notation where the first value of the tuple is the fully namespaced record name and the second value is the dictionary. For example:

```

records = [
    {
        "address": "123 Drive Street",
        "family_members": [
            ("Child", {"name": "Son"}),
            ("Child", {"name": "Daughter"}),
            ("Pet", {"name": "Dog"}),
        ]
    }
]

```

## 5.2.2 Using the record hint to specify which branch of a union to take

In addition to the tuple notation for specifying the name of a record, you can also include a special *-type* attribute (note that this attribute is *-type*, not *type*) on a record to do the same thing. So the example above which looked like this:

```

records = [
    {
        "address": "123 Drive Street",
        "family_members": [
            ("Child", {"name": "Son"}),
            ("Child", {"name": "Daughter"}),
            ("Pet", {"name": "Dog"}),
        ]
    }
]

```

Would now look like this:

```

records = [
    {
        "address": "123 Drive Street",
        "family_members": [
            {"-type": "Child", "name": "Son"},

```

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```

        {"-type": "Child", "name": "Daughter"},
        {"-type": "Pet", "name": "Dog"},
    ]
}
]

```

Unlike the tuple notation which can be used with any avro type in a union, this *-type* hint can only be used with records. However, this can be useful if you want to make a single record dictionary that can be used both in and out of unions.

## 5.3 fastavro.json\_read

**json\_reader** (*fo*: IO, *schema*: Union[str, List[T], Dict[KT, VT]], \*, *decoder*=<class 'fastavro.io.json\_decoder.AvroJSONDecoder'>) → fastavro.\_read\_py.reader  
 Iterator over records in an avro json file.

### Parameters

- **fo** – File-like object to read from
- **schema** – Reader schema
- **decoder** – By default the standard AvroJSONDecoder will be used, but a custom one could be passed here

Example:

```

from fastavro import json_reader

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ]
}

with open('some-file', 'r') as fo:
    avro_reader = json_reader(fo, schema)
    for record in avro_reader:
        print(record)

```

## 5.4 fastavro.json\_write

**json\_writer** (*fo*: IO, *schema*: Union[str, List[T], Dict[KT, VT]], *records*: Iterable[Any], \*, *write\_union\_type*: bool = True, *validator*: bool = False, *encoder*=<class 'fastavro.io.json\_encoder.AvroJSONEncoder'>) → None  
 Write records to fo (stream) according to schema

### Parameters

- **fo** – File-like object to write to

- **schema** – Writer schema
- **records** – Records to write. This is commonly a list of the dictionary representation of the records, but it can be any iterable
- **write\_union\_type** – Determine whether to write the union type in the json message. If this is set to False the output will be clear json. It may however not be decodable back to avro record by *json\_read*.
- **validator** – If true, validation will be done on the records
- **encoder** – By default the standard AvroJSONEncoder will be used, but a custom one could be passed here

Example:

```
from fastavro import json_writer, parse_schema

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}

parsed_schema = parse_schema(schema)

records = [
    {'station': u'011990-99999', 'temp': 0, 'time': 1433269388},
    {'station': u'011990-99999', 'temp': 22, 'time': 1433270389},
    {'station': u'011990-99999', 'temp': -11, 'time': 1433273379},
    {'station': u'012650-99999', 'temp': 111, 'time': 1433275478},
]

with open('some-file', 'w') as out:
    json_writer(out, parsed_schema, records)
```

## 5.5 fastavro.schema

**parse\_schema** (*schema*: Union[str, List[T], Dict[KT, VT]], *named\_schemas*: Optional[Dict[str, Dict[KT, VT]]] = None, *expand*: bool = False, *\_write\_hint*: bool = True, *\_force*: bool = False) → Union[str, List[T], Dict[KT, VT]]

Returns a parsed avro schema

It is not necessary to call parse\_schema but doing so and saving the parsed schema for use later will make future operations faster as the schema will not need to be reparsed.

### Parameters

- **schema** – Input schema
- **named\_schemas** – Dictionary of named schemas to their schema definition
- **expand** – If true, named schemas will be fully expanded to their true schemas rather than being represented as just the name. This format should be considered an output only and

not passed in to other reader/writer functions as it does not conform to the avro specification and will likely cause an exception

- **`_write_hint`** – Internal API argument specifying whether or not the `__fastavro_parsed` marker should be added to the schema
- **`_force`** – Internal API argument. If True, the schema will always be parsed even if it has been parsed and has the `__fastavro_parsed` marker

Example:

```
from fastavro import parse_schema
from fastavro import writer

parsed_schema = parse_schema(original_schema)
with open('weather.avro', 'wb') as out:
    writer(out, parsed_schema, records)
```

Sometimes you might have two schemas where one schema references another. For the sake of example, let's assume you have a *Parent* schema that references a *Child* schema. If you were to try to parse the parent schema on its own, you would get an exception because the child schema isn't defined. To accommodate this, we can use the `named_schemas` argument to pass a shared dictionary when parsing both of the schemas. The dictionary will get populated with the necessary schema references to make parsing possible. For example:

```
from fastavro import parse_schema

named_schemas = {}
parsed_child = parse_schema(child_schema, named_schemas)
parsed_parent = parse_schema(parent_schema, named_schemas)
```

**fullname** (*schema: Dict[KT, VT]*) → str

Returns the fullname of a schema

**Parameters** `schema` – Input schema

Example:

```
from fastavro.schema import fullname

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}

fname = fullname(schema)
assert fname == "test.Weather"
```

**expand\_schema** (*schema: Union[str, List[T], Dict[KT, VT]]*) → Union[str, List[T], Dict[KT, VT]]

Returns a schema where all named types are expanded to their real schema

NOTE: The output of this function produces a schema that can include multiple definitions of the same named type (as per design) which are not valid per the avro specification. Therefore, the output of this should not be passed to the normal *writer/reader* functions as it will likely result in an error.

**Parameters** `schema` (*dict*) – Input schema

Example:

```

from fastavro.schema import expand_schema

original_schema = {
    "name": "MasterSchema",
    "namespace": "com.namespace.master",
    "type": "record",
    "fields": [{
        "name": "field_1",
        "type": {
            "name": "Dependency",
            "namespace": "com.namespace.dependencies",
            "type": "record",
            "fields": [
                {"name": "sub_field_1", "type": "string"}
            ]
        }
    ], {
        "name": "field_2",
        "type": "com.namespace.dependencies.Dependency"
    }]
}

expanded_schema = expand_schema(original_schema)

assert expanded_schema == {
    "name": "com.namespace.master.MasterSchema",
    "type": "record",
    "fields": [{
        "name": "field_1",
        "type": {
            "name": "com.namespace.dependencies.Dependency",
            "type": "record",
            "fields": [
                {"name": "sub_field_1", "type": "string"}
            ]
        }
    ], {
        "name": "field_2",
        "type": {
            "name": "com.namespace.dependencies.Dependency",
            "type": "record",
            "fields": [
                {"name": "sub_field_1", "type": "string"}
            ]
        }
    }]
}

```

**load\_schema** (*schema\_path*: str, \*, *repo*: Optional[fastavro.repository.base.AbstractSchemaRepository] = None, *named\_schemas*: Optional[Dict[str, Dict[KT, VT]]] = None, *\_write\_hint*: bool = True, *\_injected\_schemas*: Set[str] = None) → Union[str, List[T], Dict[KT, VT]]

Returns a schema loaded from repository.

Will recursively load referenced schemas attempting to load them from same repository, using *schema\_path* as schema name.



If *repo* is not provided, *FlatDictRepository* is used. *FlatDictRepository* will try to load schemas from the same directory assuming files are named with the convention `<full_name>.avsc`.

### Parameters

- **schema\_path** – Full schema name, or path to schema file if default repo is used.
- **repo** – Schema repository instance.
- **named\_schemas** – Dictionary of named schemas to their schema definition
- **\_write\_hint** – Internal API argument specifying whether or not the `__fastavro_parsed` marker should be added to the schema
- **\_injected\_schemas** – Internal API argument. Set of names that have been injected

Consider the following example with default *FlatDictRepository*...

namespace.Parent.avsc:

```
{
  "type": "record",
  "name": "Parent",
  "namespace": "namespace",
  "fields": [
    {
      "name": "child",
      "type": "Child"
    }
  ]
}
```

namespace.Child.avsc:

```
{
  "type": "record",
  "namespace": "namespace",
  "name": "Child",
  "fields": []
}
```

Code:

```
from fastavro.schema import load_schema

parsed_schema = load_schema("namespace.Parent.avsc")
```

**load\_schema\_ordered** (*ordered\_schemas*: List[str], \*, *\_write\_hint*: bool = True) → Union[str, List[T], Dict[KT, VT]]

Returns a schema loaded from a list of schemas.

The list of schemas should be ordered such that any dependencies are listed before any other schemas that use those dependencies. For example, if schema *A* depends on schema *B* and schema *B* depends on schema *C*, then the list of schemas should be [C, B, A].

### Parameters

- **ordered\_schemas** – List of paths to schemas
- **\_write\_hint** – Internal API argument specifying whether or not the `__fastavro_parsed` marker should be added to the schema

Consider the following example...

Parent.avsc:

```
{
  "type": "record",
  "name": "Parent",
  "namespace": "namespace",
  "fields": [
    {
      "name": "child",
      "type": "Child"
    }
  ]
}
```

namespace.Child.avsc:

```
{
  "type": "record",
  "namespace": "namespace",
  "name": "Child",
  "fields": []
}
```

Code:

```
from fastavro.schema import load_schema_ordered

parsed_schema = load_schema_ordered(
    ["path/to/namespace.Child.avsc", "path/to/Parent.avsc"]
)
```

**to\_parsing\_canonical\_form** (*schema: Union[str, List[T], Dict[KT, VT]]*) → str

Returns a string representing the parsing canonical form of the schema.

For more details on the parsing canonical form, see here: <https://avro.apache.org/docs/current/spec.html#Parsing+Canonical+Form+for+Schemas>

**Parameters** *schema* – Schema to transform

**fingerprint** (*parsing\_canonical\_form: str, algorithm: str*) → str

Returns a string representing a fingerprint/hash of the parsing canonical form of a schema.

For more details on the fingerprint, see here: [https://avro.apache.org/docs/current/spec.html#schema\\_fingerprints](https://avro.apache.org/docs/current/spec.html#schema_fingerprints)

**Parameters**

- **parsing\_canonical\_form** – The parsing canonical form of a schema
- **algorithm** – The hashing algorithm

## 5.6 fastavro.validation

**validate** (*datum: Any, schema: Union[str, List[T], Dict[KT, VT]], field: str = "", raise\_errors: bool = True*) → bool

Determine if a python datum is an instance of a schema.

**Parameters**

- **datum** – Data being validated
- **schema** – Schema
- **field** – Record field being validated
- **raise\_errors** – If true, errors are raised for invalid data. If false, a simple True (valid) or False (invalid) result is returned

Example:

```
from fastavro.validation import validate
schema = {...}
record = {...}
validate(record, schema)
```

**validate\_many** (*records: Iterable[Any], schema: Union[str, List[T], Dict[KT, VT]], raise\_errors: bool = True*) → bool  
Validate a list of data!

#### Parameters

- **records** – List of records to validate
- **schema** – Schema
- **raise\_errors** – If true, errors are raised for invalid data. If false, a simple True (valid) or False (invalid) result is returned

Example:

```
from fastavro.validation import validate_many
schema = {...}
records = [{...}, {...}, ...]
validate_many(records, schema)
```

## 5.7 fastavro.utils

**generate\_one** (*schema: Union[str, List[T], Dict[KT, VT]]*) → Any  
Returns a single instance of arbitrary data that conforms to the schema.

**Parameters** **schema** – Schema that data should conform to

Example:

```
from fastavro import schemaless_writer
from fastavro.utils import generate_one

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}
```

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```
with open('weather.avro', 'wb') as out:
    schemaless_writer(out, schema, generate_one(schema))
```

**generate\_many** (*schema: Union[str, List[T], Dict[KT, VT]]*, *count: int*) → Iterator[Any]

A generator that yields arbitrary data that conforms to the schema. It will yield a number of data structures equal to what is given in the count

#### Parameters

- **schema** – Schema that data should conform to
- **count** – Number of objects to generate

Example:

```
from fastavro import writer
from fastavro.utils import generate_many

schema = {
    'doc': 'A weather reading.',
    'name': 'Weather',
    'namespace': 'test',
    'type': 'record',
    'fields': [
        {'name': 'station', 'type': 'string'},
        {'name': 'time', 'type': 'long'},
        {'name': 'temp', 'type': 'int'},
    ],
}

with open('weather.avro', 'wb') as out:
    writer(out, schema, generate_many(schema, 5))
```

**anonymize\_schema** (*schema: Union[str, List[T], Dict[KT, VT]]*) → Union[str, List[T], Dict[KT, VT]]

Returns an anonymized schema

**Parameters** **schema** – Schema to anonymize

Example:

```
from fastavro.utils import anonymize_schema

anonymized_schema = anonymize_schema(original_schema)
```

## 5.8 Logical Types

Fastavro supports the following official logical types:

- decimal
- uuid
- date
- time-millis
- time-micros

- timestamp-millis
- timestamp-micros
- local-timestamp-millis
- local-timestamp-micros

Fastavro is missing support for the following official logical types:

- duration

### 5.8.1 How to specify logical types in your schemas

The docs say that when you want to make something a logical type, you just need to add a *logicalType* key. Unfortunately, this means that a common confusion is that people will take a pre-existing schema like this:

```
schema = {
  "type": "record",
  "name": "root",
  "fields": [
    {
      "name": "id",
      "type": "string",
    },
  ]
}
```

And then add the uuid logical type like this:

```
schema = {
  "type": "record",
  "name": "root",
  "fields": [
    {
      "name": "id",
      "type": "string",
      "logicalType": "uuid", # This is the wrong place to add this key
    },
  ]
}
```

However, that adds the *logicalType* key to the *field* schema which is not correct. Instead, we want to group it with the *string* like so:

```
schema = {
  "type": "record",
  "name": "root",
  "fields": [
    {
      "name": "id",
      "type": {
        "type": "string",
        "logicalType": "uuid", # This is the correct place to add this key
      },
    },
  ]
}
```

## 5.8.2 Custom Logical Types

The Avro specification defines a handful of logical types that most implementations support. For example, one of the defined logical types is a microsecond precision timestamp. The specification states that this value will get encoded as an avro *long* type.

For the sake of an example, let's say you want to create a new logical type for a microsecond precision timestamp that uses a string as the underlying avro type.

To do this, there are a few functions that need to be defined. First, we need an encoder function that will encode a datetime object as a string. The encoder function is called with two arguments: the data and the schema. So we could define this like so:

```
def encode_datetime_as_string(data, schema):
    return datetime.isoformat(data)

# or

def encode_datetime_as_string(data, *args):
    return datetime.isoformat(data)
```

Then, we need a decoder function that will transform the string back into a datetime object. The decoder function is called with three arguments: the data, the writer schema, and the reader schema. So we could define this like so:

```
def decode_string_as_datetime(data, writer_schema, reader_schema):
    return datetime.fromisoformat(data)

# or

def decode_string_as_datetime(data, *args):
    return datetime.fromisoformat(data)
```

Finally, we need to tell *fastavro* to use these functions. The schema for this custom logical type will use the type *string* and can use whatever name you would like as the *logicalType*. In this example, let's suppose we call the logicalType *datetime2*. To have the library actually use the custom logical type, we use the name of `<avro_type>-<logical_type>`, so in this example that name would be *string-datetime2* and then we add those functions like so:

```
fastavro.write.LOGICAL_WRITERS["string-datetime2"] = encode_datetime_as_string
fastavro.read.LOGICAL_READERS["string-datetime2"] = decode_string_as_datetime
```

And you are done. Now if the library comes across a schema with a logical type of *datetime2* and an avro type of *string*, it will use the custom functions. For a complete example, see here:

```
import io
from datetime import datetime

import fastavro
from fastavro import writer, reader

def encode_datetime_as_string(data, *args):
    return datetime.isoformat(data)

def decode_string_as_datetime(data, *args):
    return datetime.fromisoformat(data)

fastavro.write.LOGICAL_WRITERS["string-datetime2"] = encode_datetime_as_string
```

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```

fastavro.read.LOGICAL_READERS["string-datetime2"] = decode_string_as_datetime

writer_schema = fastavro.parse_schema({
    "type": "record",
    "name": "root",
    "fields": [
        {
            "name": "some_date",
            "type": [
                "null",
                {
                    "type": "string",
                    "logicalType": "datetime2",
                },
            ],
        },
    ],
})

records = [
    {"some_date": datetime.now()}
]

bio = io.BytesIO()

writer(bio, writer_schema, records)

bio.seek(0)

for record in reader(bio):
    print(record)

```

## 5.9 fastavro command line script

A command line script is installed with the library that can be used to dump the contents of avro file(s) to the standard output.

Usage:

```

usage: fastavro [-h] [--schema] [--codecs] [--version] [-p] [file [file ...]]

iter over avro file, emit records as JSON

positional arguments:
  file                file(s) to parse

optional arguments:
  -h, --help          show this help message and exit
  --schema            dump schema instead of records
  --codecs            print supported codecs
  --version           show program's version number and exit
  -p, --pretty       pretty print json

```

## 5.9.1 Examples

Read an avro file:

```
$ fastavro weather.avro

{"temp": 0, "station": "011990-99999", "time": -619524000000}
{"temp": 22, "station": "011990-99999", "time": -619506000000}
{"temp": -11, "station": "011990-99999", "time": -619484400000}
{"temp": 111, "station": "012650-99999", "time": -655531200000}
{"temp": 78, "station": "012650-99999", "time": -655509600000}
```

Show the schema:

```
$ fastavro --schema weather.avro

{
  "type": "record",
  "namespace": "test",
  "doc": "A weather reading.",
  "fields": [
    {
      "type": "string",
      "name": "station"
    },
    {
      "type": "long",
      "name": "time"
    },
    {
      "type": "int",
      "name": "temp"
    }
  ],
  "name": "Weather"
}
```



## 6.1 fastavro.io

### 6.1.1 fastavro.io.json\_decoder

**class AvroJSONDecoder** (*fo: IO*)  
Decoder for the avro JSON format.

NOTE: All attributes and methods on this class should be considered private.

**Parameters** **fo** – File-like object to reader from

### 6.1.2 fastavro.io.json\_encoder

**class AvroJSONEncoder** (*fo: IO, \*, write\_union\_type: bool = True*)  
Encoder for the avro JSON format.

NOTE: All attributes and methods on this class should be considered private.

**Parameters**

- **fo** – Input stream
- **write\_union\_type** – Determine whether to write the union type in the json message.



## 7.1 fastavro.repository

### 7.1.1 fastavro.repository.base

**class AbstractSchemaRepository**

- genindex
- modindex
- search



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