# **Python and C++ Containers**

Release 0.4.0

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## INTRODUCTION

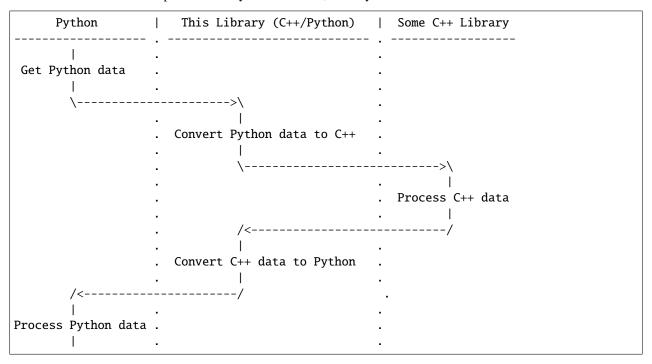
Python is well known for it's ability to handle heterogeneous data in containers such as lists like:

```
>>> 1 = [1, 2.0, "some string", ]
```

But what if you need to interact with C++ containers such as std::vector<T> that require *homogeneous* data types?

This project is about converting Python containers such as list, tuple, dict, set, frozenset containing homogeneous types such as bool, int, float, complex, bytes, str or user defined types to and from their C++ equivalent.

Here is a general example of the use of this library where Python data needs to be passed to and from a C++ library and those results need to be presented in Python. Like this, visually:



Here is a, problematic, example of how to do this:

# 1.1 A Problematic Example

Suppose that you have a Python list of floats and need to pass it to a C++ library that expects a std::vector<double>. If the result of that call modifies the C++ vector, or creates a new one, you need to return a Python list of floats from the result.

Your C++ code might look like this:

```
PyObject *example(PyObject *op) {
    std::vector<double> vec;
    // Populate the vector, function to be defined...
    write_to_vector(op, vec);
    // Do something in C++ with the vector
    // ...
    // Convert the vector back to a Python list.
    // Function to be defined...
    return read_from_vector(vec);
}
```

What should the implementation of write\_to\_vector() and read\_from\_vector() look like?

The answer seems fairly simple; firstly write\_to\_vector converting a Python list to a C++ std::vector<double> with Pythons C-API:

```
void write_to_vector(PyObject *op, std::vector<double> &vec) {
   vec.clear();
   for (Py_ssize_t i = 0; i < PyList_Size(op); ++i) {
      vec.push_back(PyFloat_AsDouble(PyList_GET_ITEM(op, i)));
   }
}</pre>
```

And the inverse, read\_from\_vector creating a new Python list from a C++ std::vector<double>:

```
PyObject *read_from_vector(const std::vector<double> &vec) {
    PyObject *ret = PyList_New(vec.size());
    for (size_t i = 0; i < vec.size(); ++i) {
        PyList_SET_ITEM(ret, i, PyFloat_FromDouble(vec[i]));
    }
    return ret;
}</pre>
```

There is no error handling shown here, and all errors would be runtime errors.

However if you need to support other object types, say lists of int, str, bytes then each one needs a pair of hand written functions; Python to C++ and C++ to Python. It gets worse when you want to support other containers such as tuple, list, set, frozenset, dict. You end up with hundreds of functions, all individually named, to handle all the combinations. Then you have to write individual conversion functions, and their tests, for all the combinations of object types *and* containers.

This is tedious and error prone and hard to extend in the general case.

# 1.2 Why This Project

This project simplifies the problem of converting data from Python to C++ and vice versa in general.

The project makes extensive use of C++ templates, partial template specialisation and code generation to dramatically reduce the amount of hand maintained code. It also converts many runtime errors to compile time errors.

The types and containers this library supports are:

Table 1: Supported Object types.

C++ Type	Python Type	Notes
bool	True, False	
long	int	
double	float	
std::complex <double></double>	> complex	
std::vector <char></char>	bytes	bytearray is not supported as we need hashable types for set and
		dict containers.
std::string	str	Specifically a PyUnicode_1BYTE_KIND <sup>1</sup> . Python documentation
std::u16string	str	Specifically a PyUnicode_2BYTE_KIND. Python documentation
std::u32string	str	Specifically a PyUnicode_4BYTE_KIND. Python documentation

Used in these containers:

Table 2: Supported Containers.

C++ Container	Python Equivalent
std::vector	Either a tuple or list
std::list	Either a tuple or list
std::unordered_set	Either a set or frozenset
std::unordered_map	dict
std::map	dict

The number of possible conversion functions is worse than the cartesian product of the types and containers as in the case of a dict the types can appear as either a key or a value.

Supporting all these conversions would normally require 352 conversion functions to be written, tested and documented<sup>2</sup>.

This project simplifies this by using a mix of C++ templates and code generators to reduce this number to just **six** hand written templates for all 352 cases.

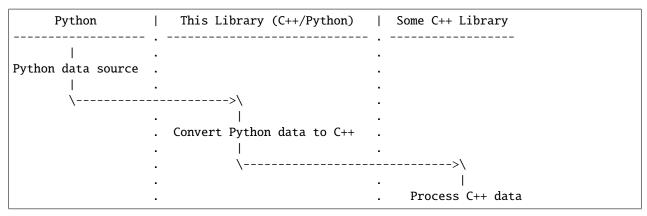
<sup>&</sup>lt;sup>1</sup> We are currently targeting C++14 so we use std::string which is defined as std::basic\_string<char>. C++20 allows a stricter, and more desirable, definition std::basic\_string<char8\_t> that we could use here. See C++ reference for std::string

<sup>&</sup>lt;sup>2</sup> There are six unary container pairings (tuple <-> std::list, tuple <-> std::vector, list <-> std::list, list <-> std::vector, set <-> std::unordered\_set, frozenset <-> std::unordered\_set) with eight types (bool, int, float, complex, bytes, str[1], str[2], str[4]). Each container/type combination requires two functions to give two way conversion from Python to C++ and back. Thus 6 (container pairings) \* 8 (types) \* 2 (way conversion) = 96 required functions. For dict there are two container pairings (dict <-> std::map, dict <-> std::unordered\_map) with the eight types either of which can be the key or the value so 64 (8\*\*2) possible variations. Thus 2 (container pairings) \* 64 (type pairs) \* 2 (way conversion) = 256 required functions. Thus is a total of 96 + 256 = 352 functions.

# 1.3 Using This Library

# 1.3.1 Python to C++

Using the library is as simple as this, suppose you have data in Python that needs to be passed to a C++ library:



The C++ code using this library looks like this:

#### C++ Code

```
#include "python_convert.h"

// Create a Python list of floats: [21.0, 42.0, 3.0]
PyObject *op = Py_BuildValue("[ddd]", 21.0, 42.0, 3.0);

// Create the C++ vector that we want to convert this data to...
std::vector<double> cpp_vector;

// The template specialisation will automatically invoke the appropriate
// function call.
// It will be a compile time error if the container/type function
// is not supported.
// At run time this will return zero on success, non-zero on failure,
// for example if op is not a Python tuple or members of op can not be
// converted to C++ doubles.
int err = Python_Cpp_Containers::py_list_to_cpp_std_list_like(op, cpp_vector);
// Handle error checking if err is non-zero...
```

**Note:** If you were to change the C++ container to a std::list<double> the function call py\_list\_to\_cpp\_std\_list\_like() would be the same. Of course py\_list\_to\_cpp\_std\_list\_like() would then dispatch to code handling a std::list<double>.

Another example, suppose the Python data source is a typing.Dict[int, str] and this needs to be converted to a C++ std::map<long, std::string>> then a function using the conversion code using this library is as simple as this:

```
#include "python_convert.h"

void convert_py_data_to_cpp(Py0bject *arg) {
    std::unordered_map<long, std::string> map;
    if (Python_Cpp_Containers::py_dict_to_cpp_std_map_like(arg, map)) {
        // Handle error...
    } else {
        // Use the map...
    }
}
```

## 1.3.2 C++ to Python

Suppose that you have data from a C++ library and this data needs to be represented in Python:

The C++ code using this library looks like this:

```
#include "python_convert.h"

std::vector<double> cpp_vector;
// Populate the C++ vector...
cpp_vector.push_back(21.0);
cpp_vector.push_back(42.0);
cpp_vector.push_back(3.0);

// Now convert to Python.
// This will be a compile time error if the C++ type is not supported.
Py0bject *op = Python_Cpp_Containers::cpp_std_list_like_to_py_list(cpp_vector);
// op is a Python list of floats: [21.0, 42.0, 3.0]
// op will be null on failure and a Python exception will have been set.
```

**Note:** If you were to change the C++ container to a std::list<double> the function call cpp\_std\_list\_like\_to\_py\_list() would be the same. Of course cpp\_std\_list\_like\_to\_py\_list() would then dispatch to code handling a std::list<double>.

Another example, suppose the C++ data source is a std::map<long, std::string>> and we need this a Python dict typing.Dict[int, str] then the conversion code in this library is as simple as this:

```
#include "python_convert.h"

PyObject *convert_cpp_data_to_py() {
    std::map<long, std::string> map;
    // Populate map from the C++ data source
    // ...
    // Now convert to a Python dict:
    return Python_Cpp_Containers::cpp_std_map_like_to_py_dict(map);
}
```

### 1.4 The Hand Written Functions

At the heart off this library here are only six non-trivial hand written functions along with a much larger of generated functions that successively specialise these handwritten functions. They are defined as templates in src/cpy/python\_object\_convert.h.

- Two C++ templates for Python tuple / list to and from std::list or std::vector for all types.
- Two C++ templates for Python set / frozenset to and from std::unordered\_set for all types.
- Two C++ templates for Python dict to and from std::map or std::unordered\_map for all type pairs.

These six handwritten templates are short, fairly simple and comprehensible. Then, for simplicity, a Python script is used to create the final, instantiated, 352 functions.

As an example, here how the function is developed that converts a Python list of float to and from a C++ std::vector<double> or std::list<double>.

First C++ to Python.

### 1.4.1 Converting a C++ std::vector<T> or std::list<T> to a Python tuple or list

The generic function signature looks like this:

```
template<
    template<typename ...> class ListLike,
    typename T,
    PyObject *(*ConvertCppToPy)(const T &),
    PyObject *(*PyUnaryContainer_New)(size_t),
    int(*PyUnaryContainer_Set)(PyObject *, size_t, PyObject *)
>
PyObject *
very_generic_cpp_std_list_like_to_py_unary(const ListLike<T> &list_like) {
    // Handwritten code, see "C++ to Python Implementation" below.
    // ...
}
```

Table 3: very\_generic\_cpp\_std\_list\_like\_to\_py\_unary() template parameters.

Template Parame-	Notes		
ter			
ListLike	The C++ container type, either a std::vector <t> or std::list<t>.</t></t>		
T	The C++ type of the objects in the target C++ container.		
ConvertCppToPy	A pointer to a function that converts any C++ T to a PyObject *, for example from double		
	-> float. The function signature is PyObject *ConvertCppToPy(const T&). This		
	returns NULL on failure.		
PyUnaryContainer_N	ext pointer to a function that creates a new Python container, for exam-		
	ple a list, of a particular length. The function signature is PyObject		
	*PyUnaryContainer_New(Py_ssize_t). This returns NULL on failure.		
PyUnaryContainer_S	PyUnaryContainer_\$eA pointer to a function that sets a PyObject * in the Python container at a given index. The		
	<pre>function signature is int PyUnaryContainer_Set(PyObject *container, size_t</pre>		
	pos, PyObject *value)). This returns 0 on success.		

And the function has the following parameters.

Table 4: very\_generic\_cpp\_std\_list\_like\_to\_py\_unary() parameters.

Туре	Name	Notes
ListLike <t> &amp;</t>	list_like	The C++ list like container to read from to.

The return value is non-NULL on success or NULL if there is a runtime error. These errors could be:

- PyObject \* container can not be created.
- A member of the Python container can not be created from the C++ type T.
- The PyObject \* can not be inserted into the Python container.

## 1.4.2 C++ to Python Implementation

The implementation is fairly straightforward in src/cpy/python\_object\_convert.h (lightly edited):

(continues on next page)

```
// be done when decrefing the container.
                // e.g. tupledealloc():
                // https://github.com/python/cpython/blob/main/Objects/tupleobject.c
                PyErr_Format(PyExc_ValueError, "C++ value of can not be converted.");
                goto except;
            // PyUnaryContainer_Set usually wraps a void function, always succeeds
            // returning non-zero.
            if (PyUnaryContainer_Set(ret, i++, op)) { // Stolen reference.
                PyErr_Format(PyExc_RuntimeError, "Can not set unary value.");
                goto except;
            }
        }
    } else {
       PyErr_Format(
            PyExc_ValueError,
            "Can not create Python container of size %ld",
            list_like.size()
        );
        goto except;
   assert(!PyErr_Occurred());
    assert(ret);
   goto finally;
except:
   Py_XDECREF(ret);
   assert(PyErr_Occurred());
   ret = NULL;
finally:
   return ret;
```

# 1.4.3 Partial Specialisation to Convert a C++ std::vector<T> or std::list<T> to a Python list`

As an example this is specialised for a C++ std::vector and a Python list with a handwritten oneliner:

```
template<
    typename T,
    PyObject *(*ConvertCppToPy)(const T &)
>
PyObject *
generic_cpp_std_list_like_to_py_list(const std::vector<T> &container) {
    return very_generic_cpp_std_list_like_to_py_unary<
        std::vector, T, ConvertCppToPy, &py_list_new, &py_list_set
    >(container);
}
```

**Note:** The use of the function pointers to py\_list\_new, and py\_list\_set that are defined in this project namespace.

These are thin wrappers around existing functions or macros in "Python.h".

There is a similar partial specialisation for a Python tuple:

```
template<
    typename T,
    PyObject *(*ConvertCppToPy)(const T &)
>
PyObject *
generic_cpp_std_list_like_to_py_list(const std::vector<T> &container) {
    return very_generic_cpp_std_list_like_to_py_unary<
        std::vector, T, ConvertCppToPy, &py_tuple_new, &py_tuple_set
    >(container);
}
```

## 1.4.4 Converting a Python tuple or list to a C++ std::vector<T> or std::list<T>

The reverse is converting Python to C++. This generic function that converts unary Python indexed containers (tuple and list) to a C++ std::vector<T> or std::list<T> for any type has this signature:

This template has these parameters:

plate paran	neters.	
Template Parameter	Notes	
ListLike	The C++ container type, either a std::vector <t> or std::list<t>.</t></t>	
T	The C++ type of the objects in the target C++ container.	
PyObject_Check	A pointer to a function that checks that any PyObject * in the Python container	
	is the correct type, for example that it is a bytes object. The function signature	
	is int PyObject_Check(PyObject *). This returns non-zero if the Python	
	object is as expected.	
PyObject_Convert	A pointer to a function that converts any PyObject * in the Python container to	
	the C++ type, for example from bytes -> std::vector <char>. The function</char>	
	signature is T PyObject_Convert(PyObject *).	
PyUnaryContainer_Check	A pointer to a function that checks that the PyObject * argument is the cor-	

rect container type, for example a tuple. The function signature is int PyUnaryContainer\_Check(PyObject \*). This returns non-zero if the Python

A pointer to a function that returns the size of the Python container. The function signature is Py\_ssize\_t PyUnaryContainer\_Size(PyObject \*op). This

A pointer to a function that gets a PyObject \* from the Python

The function signature is PyObject

Table 5: very\_generic\_py\_unary\_to\_cpp\_std\_list\_like() template parameters.

And the function has the following parameters.

PyUnaryContainer\_Size

PyUnaryContainer\_Get

Table 6: generic\_py\_unary\_to\_cpp\_std\_list\_like() parameters.

\*PyUnaryContainer\_Get(PyObject \*, size\_t).

returns the size of the the Python container.

Туре	Name	Notes
PyObject *	op	The Python container to read from.
ListLike <t> &amp;</t>	list_like	The C++ list like container to write to.

The return value is zero on success or non-zero if there is a runtime error. These errors could be:

container is not as expected.

container at a given index.

- PyObject \*op is not a container of the required type.
- A member of the Python container can not be converted to the C++ type T (Py0bject\_Check fails).

## 1.4.5 Python to C++ Implementation

The implementation is fairly straightforward in src/cpy/python\_object\_convert.h (lightly edited):

(continues on next page)

```
int ret = 0:
   list_like.clear();
   Py_INCREF(op); // Increment borrowed reference
   if (!PyUnaryContainer_Check(op)) {
        PyErr_Format(
            PyExc_ValueError,
            "Can not convert Python container of type %s",
            op->ob_type->tp_name
        );
        ret = -1;
        goto except;
    for (Py_ssize_t i = 0; i < PyUnaryContainer_Size(op); ++i) {</pre>
        PyObject *value = PyUnaryContainer_Get(op, i);
        if (!value) {
            ret = -2;
            goto except;
        if (!(*PyObject_Check)(value)) {
            list_like.clear();
            PyErr_Format(
                    PyExc_ValueError,
                    "Python value of type %s can not be converted",
                    value->ob_type->tp_name
            );
            ret = -3;
            goto except;
        list_like.push_back((*PyObject_Convert)(value));
        // Check !PyErr_Occurred() which could never happen as we check first.
   }
   assert(!PyErr_Occurred());
   goto finally;
except:
    assert(PyErr_Occurred());
    list_like.clear();
   Py_DECREF(op); // Decrement borrowed reference
   return ret:
```

# 1.4.6 Partial Specialisation to Convert a Python list to a C++ std::vector<T> or std::list<T>

This template can be partially specialised for converting Python *lists* of any type to C++ std::vector<T> or std::list<T>. This is hand written code but it is trivial by wrapping a single function call.

In the particular case of a std::vector we can use .reserve() as an optimisations to avoid excessive re-allocations.

```
template<
  typename T,</pre>
```

(continues on next page)

```
int (*PyObject_Check)(PyObject *),
   T (*PyObject_Convert)(PyObject *)
>
int generic_py_list_to_cpp_std_list_like(
    PyObject *op, std::vector<T> &container
) {
        // Reserve the vector, but only if it is a list.
        // If it is any other Python object then ignore it as py_list_len()
        // may give undefined behaviour.
        // Leave it to very_generic_py_unary_to_cpp_std_list_like() to error
        if (py_list_check(op)) {
            container.reserve(py_list_len(op));
        }
        return very_generic_py_unary_to_cpp_std_list_like<
            std::vector, T, PyObject_Check, PyObject_Convert,
            &py_list_check, &py_list_len, &py_list_get
        >(op, container);
}
```

**Note:** The use of the function pointers to py\_list\_check, py\_list\_len and py\_list\_get that are defined in this project namespace. These are thin wrappers around existing functions or macros in "Python.h".

There is a similar partial specialisation for the Python tuple:

# 1.5 Generated Functions

The particular function specialisations are created by a Python script that takes the cartesian product of object types and container types and creates functions for each container/object.

## 1.5.1 C++ to Python

For example, to convert a C++ std::vector<double> to a Python list of float the following are created:

A base declaration in *auto\_py\_convert\_internal.h*:

```
template<typename T>
PyObject *
cpp_std_list_like_to_py_list(const std::vector<T> &container);
```

And a concrete declaration for each C++ target type T in *auto\_py\_convert\_internal.h*:

```
template <>
Py0bject *
cpp_std_list_like_to_py_list<double>(const std::vector<double> &container);
```

And the concrete definition is in *auto\_py\_convert\_internal.cpp*, this simply calls the generic function:

Here is the function hierarchy for converting lists to C++ std::vector<T> or std::list<T>: This is the function hierarchy for the code that converts C++ std::vector<T> or std::list<T> to Python list and tuple for all supported object types.

## 1.5.2 Python to C++

For example, to convert a Python list of float to a C++ std::vector<double> the following are generated:

A base declaration in *auto\_py\_convert\_internal.h*:

```
template<typename T>
int
py_list_to_cpp_std_list_like(PyObject *op, std::list<T> &container);
```

And a concrete declaration for each C++ target type T in *auto\_py\_convert\_internal.h*:

```
template <>
int
py_list_to_cpp_std_list_like<double>(PyObject *op, std::list<double> &container);
```

And the concrete definition is in *auto\_py\_convert\_internal.cpp*:

This is the function hierarchy for the code that converts Python list and tuple to C++ std::vector<T> or std::list<T> for all supported object types.

**CHAPTER** 

**TWO** 

# **USING THIS C++ LIBRARY**

# 2.1 The Basics

# 2.1.1 Get the Project from git

```
$ git pull https://github.com/paulross/PyCppContainers.git
```

#### 2.1.2 Code Generation

If necessary run the code generator:

```
cd src/py
python3 code_gen.py
```

Which should give you something like:

```
Version: 0.4.0

Target directory "src/cpy"

Writing declarations to "src/cpy/auto_py_convert_internal.h"

Wrote 4125 lines of code with 356 declarations.

Writing definitions to "src/cpy/auto_py_convert_internal.cpp"

Wrote 3971 lines of code with 352 definitions.

Process finished with exit code 0
```

## 2.1.3 Build Configuration

You need to compile the following C++ files by adding them to your makefile or CMakeLists.txt: This project has a CMakeLists.txt as an example.

```
src/cpy/auto_py_convert_internal.cpp
src/cpy/python_container_convert.cpp
src/cpy/python_object_convert.cpp
```

#### 2.1.4 Source Inclusion

Your pre-processor needs access to the header files with the compiler flag -I src/cpy.

Then in your C++ code include:

```
#include "python_convert.h"
```

Which gives you access to the whole C++API in the namespace Python\_Cpp\_Containers.

#### **2.1.5 Errors**

If using this library in C++ there will be a linker error if you specify a template type that is not supported. For example here is some code that tries to copy a Python list of unsigned integers. The two conversion functions are not defined for unsigned int.

```
static PyObject *
new_list_unsigned_int(PyObject *Py_UNUSED(module), PyObject *arg) {
    std::vector<unsigned int> vec;
    if (!py_list_to_cpp_std_list_like(arg, vec)) {
        return cpp_std_list_like_to_py_list(vec);
    }
    return NULL;
}
```

A C++ tool chain will complain with a linker error such as:

```
Undefined symbols for architecture x86_64:

"_object* Python_Cpp_Containers::cpp_std_list_like_to_py_list<unsigned int>(std::__

-1::vector<unsigned int, std::__1::allocator<unsigned int> > const&)", referenced from:
    new_list_unsigned_int(_object*, _object*) in cPyCppContainers.cpp.o

"int Python_Cpp_Containers::py_list_to_cpp_std_list_like<unsigned int>(_object*, std::_
-1::vector<unsigned int, std::__1::allocator<unsigned int> >&)", referenced from:
    new_list_unsigned_int(_object*, _object*) in cPyCppContainers.cpp.o

ld: symbol(s) not found for architecture x86_64
```

If you are building a Python extension this will, most likely, build but importing the extension will fail immediately with something like:

# 2.2 Examples

There are some examples of using this library in *src/ext/cPyCppContainers.cpp*. This extension is built by *setup.py* and tested with *tests/unit/test\_cPyCppContainers.py*.

To build this extension:

```
$ python setup.py develop
```

And to use it:

```
import cPyCppContainer
```

# 2.2.1 Using C++ to Double the Values in a Python List of float

Here is one of those examples in detail; doubling the values of a Python list of floats.

At the beginning of the extension C/C++ code we have:

```
#include "python_convert.h"
```

For convenience we use the namespace that the conversion code is within:

```
using namespace Python_Cpp_Containers;
```

Here is the C++ function that we want to call that multiplies the values of a std::vector<double> in-place by 2.0:

```
/** Double the values of a vector in-place. */
static void
vector_double_x2(std::vector<double> &vec) {
    for (size_t i = 0; i < vec.size(); ++i) {
        vec[i] *= 2.0;
    }
}</pre>
```

And here is the code that takes a Python list of floats, then calls the C++ function and finally converts the C++ std::vector<double> back to a new Python list of floats:

```
/** Create a new list of floats with doubled values. */
static PyObject *
list_x2(PyObject *Py_UNUSED(module), PyObject *arg) {
    std::vector<double> vec;
    // py_list_to_cpp_std_list_like() will return non-zero if the Python
    // argument can not be converted to a std::vector<double>
    // and a Python exception will be set.
    if (!py_list_to_cpp_std_list_like(arg, vec)) {
        // Double the values in pure C++ code.
        vector_double_x2(vec);
        // cpp_std_list_like_to_py_list() returns NULL on failure
        // and a Python exception will be set.
        return cpp_std_list_like_to_py_list(vec);
    }
    return NULL;
}
```

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The vital piece of code is the declaration std::vector<double> vec; and that means:

- If a py\_list\_to\_cpp\_std\_list\_like() implementation does not exist for double there will be a compile time error.
- Giving py\_list\_to\_cpp\_std\_list\_like() anything other than a list of floats will create a Python runtime error.
- If cpp\_std\_list\_like\_to\_py\_list() fails for any reason there will be a Python runtime error.

### **Using the Extension**

Once the extension is built you can use it thus:

```
>>> import cPyCppContainers
>>> cPyCppContainers.list_x2([1.0, 2.0, 4.0])
[2.0, 4.0, 8.0]
```

You can verify that the returned list is a new one rather than modifying the input in-place:

```
>>> a = [1.0, 2.0, 4.0]
>>> b = cPyCppContainers.list_x2(a)
>>> hex(id(a))
'0x1017150c0'
>>> hex(id(b))
'0x101810dc0'
```

If the values are not floats or the container is not a list a ValueError is raised:

```
>>> cPyCppContainers.list_x2([1, 2, 4])
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
ValueError: Python value of type int can not be converted
>>> cPyCppContainers.list_x2((1.0, 2.0, 4.0))
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
ValueError: Can not convert Python container of type tuple
```

## 2.2.2 Reversing a tuple of bytes in C++

Here is another example, suppose that we have a function to to reverse a tuple of bytes in C++:

```
/** Returns a new vector reversed. */
template<typename T>
static std::vector<T>
reverse_vector(const std::vector<T> &input){
    std::vector<T> output;
    for (size_t i = input.size(); i-- > 0;) {
        output.push_back(input[i]);
    }
    return output;
}
```

Here is the extension code that call this:

```
/** Reverse a tuple of bytes in C++. */
static PyObject *
tuple_reverse(PyObject *Py_UNUSED(module), PyObject *arg) {
    std::vector<std::vector<char>> vec;
    if (!py_tuple_to_cpp_std_vector(arg, vec)) {
        return cpp_std_vector_to_py_tuple(reverse_vector(vec));
    }
    return NULL;
}
```

Once again the declaration std::vector<std::vector<char>> vec; ensures that the correct instantiations of the conversion functions are called.

When the extension is built it can be used like this:

```
>>> import cPyCppContainers
>>> cPyCppContainers.tuple_reverse((b'ABC', b'XYZ'))
(b'XYZ', b'ABC')
```

# 2.2.3 Incrementing dict values in C++

Here is an example of taking a Python dict of [bytes, int] and creating a new dict with the values increased by one. The C++ code in the extension is this:

```
/** Creates a new dict[bytes, int] with the values incremented by 1 in C++ */
static PyObject *
dict_inc(PyObject *Py_UNUSED(module), PyObject *arg) {
    std::unordered_map<std::vector<char>, long> dict;
    /* Copy the Python structure to the C++ one. */
    if (!py_dict_to_cpp_std_unordered_map(arg, dict)) {
        /* Increment. */
        for(auto &key_value: dict) {
            key_value.second += 1;
        }
        /* Copy the C++ structure to a new Python dict. */
        return cpp_std_unordered_map_to_py_dict(dict);
    }
    return NULL;
}
```

Once the extension is built this can be used thus:

```
>>> import cPyCppContainers
>>> cPyCppContainers.dict_inc({b'A' : 65, b'Z' : 90})
{b'Z': 91, b'A': 66}
```

There are several other examples in *src/ext/cPyCppContainers.cpp* with tests in *tests/unit/test\_cPyCppContainers.py*.

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**CHAPTER** 

THREE

# **BUILDING AND TESTING**

This chapter describes how to build and test this library. It assumes that you are in your chosen directory and have done:

```
$ git pull https://github.com/paulross/PyCppContainers.git
```

# 3.1 Building and Testing C++ Code

### 3.1.1 Code Generation

If necessary run the code generator:

```
cd src/py
python3 code_gen.py
```

Which should give you something like:

```
Version: 0.4.0

Target directory "src/cpy"
Writing declarations to "src/cpy/auto_py_convert_internal.h"
Wrote 4125 lines of code with 356 declarations.
Writing definitions to "src/cpy/auto_py_convert_internal.cpp"
Wrote 3971 lines of code with 352 definitions.

Process finished with exit code 0
```

# 3.1.2 Building C++ Code

Here are some examples of cleaning and building this project, both debug and release builds:

```
echo "---> C++ clean debug"

cmake --build cmake-build-debug --target clean -- -j 6

echo "---> C++ build debug"

cmake --build cmake-build-debug --target PyCppContainers -- -j 6

echo "---> C++ clean release"

cmake --build cmake-build-release --target clean -- -j 6

echo "---> C++ build release"

cmake --build cmake-build-release --target PyCppContainers -- -j 6
```

## 3.1.3 Testing C++ Code

```
echo "---> C++ debug tests"
cmake-build-debug/PyCppContainers
echo "---> C++ release tests"
cmake-build-release/PyCppContainers
```

The debug tests only test the basic functionality but they do so with some expensive integrity tests. Release tests are much more extensive testing the performance of the library.

For example running the debug tests:

```
time cmake-build-debug/PyCppContainers
Hello, World!
Python version: 3.12.1
test_functional_all START
RSS(Mb): was:
                  17.555 now:
                                   17.613 diff:
                                                     +0.059 Peak was:
                                                                           17.555 now:
→17.613 diff:
                   +0.059 test_vector_to_py_tuple<bool>
RSS(Mb): was:
                  17.617 now:
                                   17.660 diff:
                                                     +0.043 Peak was:
                                                                           17.617 now:
→17.660 diff:
                   +0.043 test_vector_to_py_tuple<long>
RSS(Mb): was:
                  17.660 now:
                                   17.664 diff:
                                                     +0.004 Peak was:
                                                                           17.660 now:
→17.664 diff:
                   +0.004 test_vector_to_py_tuple<double>
8<---- Snip ---->8
                                                                           16.762 now:
                  16.762 now:
                                   16.805 diff:
                                                     +0.043 Peak was:
RSS(Mb): was:
\rightarrow16.805 diff:
                   +0.043 test_cpp_std_map_to_py_dict_string
RSS(Mb): was:
                  16.805 now:
                                   16.828 diff:
                                                     +0.023 Peak was:
                                                                           16.805 now:
\rightarrow16.828 diff:
                   +0.023 test_py_dict_to_cpp_std_map_like_string
RSS(Mb): was:
                  16.805 now:
                                   16.828 diff:
                                                     +0.023 Peak was:
                                                                           16.805 now:
\rightarrow16.828 diff:
                   +0.023 test_py_dict_to_cpp_std_map_string
==== RSS(Mb): was:
                         7.203 now:
                                        16.828 diff:
                                                          +9.625 Peak was:
                                                                                 7.203 now:
      16.828 diff:
                       +9.625 ==== test_functional.cpp
test_functional_all FINISH
```

This is followed by the test results:

```
Number of tests: 3333
\hookrightarrowS+)\s+(\S+)"
\hookrightarrow S+([0-9+-.]+)\S+(\d+)\S+([0-9+-.]+)\S+(\S+)"
REGEX_TAIL: "TAIL:\s+(.+)"
HEAD: Fail
            Scale Repeat
                                             Std.Dev.(s)
                                                                Min.(s)
                                                                               Max.
                                 Mean(s)
\hookrightarrow (s)
                    Rate(/s) Name
         Count
TEST:
        0 1048576
                             0.001676959
                                                    N/A
                                                                    N/A
                                                                                   N/
⊶A
           1
                     596.3 test_functional_tuple_setitem():[1048576]
TEST:
        0 1048576
                       1
                             0.000595792
                                                                    N/A
                                                                                   N/
                                                    N/A
⊶A
           1
                    1678.4 test_functional_list_setitem():[1048576]
TEST:
        0 1048576
                       1
                             0.000947000
                                                                                   N/
                                                    N/A
                                                                    N/A
                    1056.0 test_functional_set_add():[1048576]
                                                                8<---- Snip ---->8
\hookrightarrow A
           1
TEST:
           32768
                       1
                             0.117341292
                                                    N/A
                                                                    N/A
                                                                                   N/
\hookrightarrow A
           1
                       8.5 test_py_tuple_str32_to_vector std::string[2048]>():[32768]
            65536
TEST:
                       1
                             0.249899958
                                                    N/A
                                                                    N/A
                                                                                   N/
\hookrightarrow A
                       4.0 test_py_tuple_str32_to_vector std::string[2048]>():[65536]
TEST:
                                                    N/A
             4096
                       1
                             4.806965084
                                                                    N/A
                                                                                   N/
                       0.2 test_unordered_set_bytes_to_py_set std::string [dom:bides 6] next page)
\rightarrow (): [4096]
```

```
TAIL: Passed=3333/3333 Failed=0/3333
All tests pass.

===RSS(Mb): was: 10.012 now: 180.496 diff: +170.484 Peak was: 10.012 now: 3250.059 diff: +3240.047 main.cpp
Total execution time: 2370.715 (s)
Count of unique strings created: 5895690
Bye, bye! Returning 0
2371.63 real 2324.66 user 32.06 sys
```

This takes, typically, 40 minutes. A return code of 0 is success. If there are any failing tests then the return code will be the number of failing tests.

The release tests are similar but they include all the performance tests which take a long while. Run time is around six hours.

# 3.2 Building and Testing Python Code

## 3.2.1 Building Python Code

To build all the Python code create a virtual environment then:

```
$ pip install -r requirements-dev.txt
$ python setup.py develop
```

This takes a minute or so.

# 3.2.2 Testing Python Code

The Python tests check these things:

- Functional testing for Python C extensions.
- Performance testing for Python C extensions that exercise the C++ library. Usually round tripping Python structures to C++ and back again.
- Memory usage testing for Python C extensions that use this C++ library.

As a basic, from your virtual environment:

(continues on next page)

By default this only does the functional tests and skips the others such as performance and memory tests. To run these tests you need to add the arguments --runslow and -pymemtrace respectively, see below. This takes about 40 minutes.

#### **Testing Performance**

To include all the performance tests, this takes about 25 minutes:

```
$ pytest tests/ --runslow
```

#### Example:

```
$ time pytest tests --runslow
platform darwin -- Python 3.12.1, pytest-8.3.3, pluggy-1.5.0
rootdir: PythonCppHomogeneousContainers
configfile: pytest.ini
collected 128 items
.....
                                                    [ 57%]
tests/unit/test_cUserDefined.py ......
                                                    「 64%]
tests/unit/test_perf_cPyCppContainers.py ..... [ 91%]
tests/unit/test_with_pymemtrace.py ssssssssss
                                                    [100\%]
======= 116 passed, 11 skipped, 1 xfailed in 1595.44s (0:26:35) ===========
pytest tests --runslow 1122.24s user 376.19s system 93% cpu 26:35.99 total
```

#### **Testing Memory Usage**

To include all the memory tests, this takes about 20 minutes:

```
$ time pytest tests --pymemtrace
============= test session starts ===========================
platform darwin -- Python 3.12.1, pytest-8.3.3, pluggy-1.5.0
rootdir: PythonCppHomogeneousContainers
configfile: pytest.ini
collected 128 items
tests/unit/test_cPyCppContainers.py .....x..... [ 35%]
.....
                                                   [ 57%]
tests/unit/test_cUserDefined.py ......
                                                   [ 64%]
tests/unit/test_with_pymemtrace.py .....
                                                   [100\%]
Opening log file 20241010_102101_11927.log
Opening log file 20241010_102106_11927.log
```

(continues on next page)

```
Opening log file 20241010_102115_11927.log
Opening log file 20241010_102136_11927.log
Opening log file 20241010_102202_11927.log
Opening log file 20241010_102214_11927.log
Opening log file 20241010_103049_11927.log
Opening log file 20241010_103453_11927.log
Opening log file 20241010_103925_11927.log
Opening log file 20241010_104003_11927.log
Opening log file 20241010_1040051_11927.log
Opening log file 20241010_104051_11927.log
Opening log file 20241010_104051_11927.log
Opening log file 20241010_104051_11927.log
```

Running with both --runslow and --pymentrace takes about about 40 minutes.

# 3.3 Building the Documentation

This describes how create the documentation with gnuplot, Sphinx or doxygen.

## 3.3.1 gnuplot Plots

#### **Recreating Plot Data**

If required the performance data can recreated. Firstly the C++ performance, from the project directory:

```
$ # Pipe the results of the C++ tests to a specific file.
$ cmake-build-release/PyCppContainers > perf_notes/cpp_test_results.txt
$ cd perf_notes
$ # Run this script that will take the C++ output and split it into .dat files
$ # in perf_notes/dat.
$ python write_dat_files_for_cpp_test_results.py
```

Now for the Python tests, from the project directory, this takes about 40 minutes:

```
$ # Pipe the results of the Python tests to a specific file.
$ pytest tests --runslow --pymemtrace -vs > perf_notes/python_test_results.txt
$ cd perf_notes
$ # Run this script that will take the Python output and split it into .dat files
$ # in perf_notes/dat.
$ python write_dat_files_for_python_test_results.py
```

```
$ # Copy the .dat files to the documentation ready for gnuplot, from project directory.
$ cp perf_notes/dat/*.dat docs/sphinx/source/plots/dat
```

**Note:** Memory plots

The .dat files for memory plots are not (yet) automated and have to be done by hand by copying the pymemtrace log files.

#### **Recreating Plot Images**

To recreate the gnuplot plot images that are used by the documentation from the project directory:

```
$ cd docs/sphinx/source/plots
$ gnuplot -p *.plt
```

## 3.3.2 Sphinx

To build the HTML and PDF documentation from the project directory:

```
$ cd docs/sphinx
$ make clean
$ make html latexpdf
$ cp build/latex/PyCppContainers.pdf ..
$ open build/html/index.html
$ open .../PyCppContainers.pdf
```

# 3.3.3 Doxygen

To build the HTML Doxygen documentation from the project directory:

```
$ cd docs
$ doxygen PyCppContainers.dox
$ open doxygen/html/index.html
```

The Doxygen PDF:

```
$ cd docs/doxygen/latex
$ make pdf
$ cp refman.pdf ../../PyCppContainers_Doxygen.pdf
```

# 3.4 Building and Testing Everything for Multiple Python Versions

The script build\_all.sh will execute:

- C++ clean and build debug and release versions.
- Run C++ debug build and the associated tests (this omits C++ performance tests).
- Run C++ release build and the all the tests including C++ performance tests.
- For each Python version (currently 3.8, 3.9, 3.10, 3.11, 3.12, 3.13) it:
  - Creates a new virtual environment.
  - Runs pip install -r requirements-dev.txt.
  - Runs python setup.py develop.
  - Runs pytest tests/ -x to catch any functional errors.
  - Runs pytest tests/ -vs --runslow --pymemtrace to run all tests.
  - Runs python setup.py bdist\_wheel to create the wheels.

- Runs python setup.py sdist to create the source distribution.

If any of these fail the script will halt with a failure indication.

The output is verbose typically 30,000 lines.

The typical time breakdown is:

- C++ debug and release builds: 5 minutes.
- C++ debug tests (3,000+): 40 minutes.
- C++ release tests (around 25,000): about 40 minutes.
- Python: create environment and run all tests, including slow and memory tests (127): around 45 minutes per Python version.

For all Python versions (6 currently) this takes about eight hours.

This does not build the documentation.

**CHAPTER** 

**FOUR** 

C++ API

This chapter describes the C++ API to this library.

# 4.1 Include File and Namespace

```
#include "python_convert.h"
```

All these APIs are in the namespace Python\_Cpp\_Containers.

# 4.2 Python Containers to C++

#### 4.2.1 Error Indication

All of the conversion functions from Python to C++ return an integer which is zero on success, non-zero otherwise. Reasons for failure can be:

- The PyObject \* is not the expected Python container, for example passing a Python tuple when a list is expected.
- A member of the Python container can not be converted to C++ type <T>.

In the error case a PyErr\_... will be set and the given container cleared.

## 4.2.2 Python tuple to std::vector or std::list

#### **API**

```
template<typename T>
int
py_tuple_to_cpp_std_list_like(PyObject *op, std::vector<T> &container);

template<typename T>
int
py_tuple_to_cpp_std_list_like(PyObject *op, std::list<T> &container);
```

## **Arguments**

Argument op	Argument	Return value
	container	
A Python tuple containing val-	The std::vector	0 on success, non-zero on failure in which case the con-
ues convertable to type <t>.</t>	or std::list to	tainer will be empty. The causes of failure can be; op is
	write to.	not a tuple or a member of the op can not be converted
		to type <t>.</t>

## **Example**

Process a tuple of Python float:

```
void tuple_float_to_cpp(PyObject *arg) {
    std::vector<double> vec;
    if (py_tuple_to_cpp_std_list_like(arg, vec)) {
        // Handle error...
    } else {
        // Use vec...
    }
}
```

# 4.2.3 Python list to std::vector or std::list

#### API

```
template<typename T>
int
py_list_to_cpp_std_list_like(PyObject *op, std::vector<T> &container);

template<typename T>
int
py_list_to_cpp_std_list_like(PyObject *op, std::list<T> &container);
```

### **Arguments**

Argument op	Argument	Return value
	container	
A Python list containing val-	The std::vector	0 on success, non-zero on failure in which case the con-
ues convertable to type <t>.</t>	or std::list to	tainer will be empty. The causes of failure can be; op is
	write to.	not a list or a member of the op can not be converted to
		type <t>.</t>

### **Example**

Process a list of Python float:

```
void list_float_to_cpp(Py0bject *arg) {
    std::list<double> list;
    if (py_list_to_cpp_std_list_like(arg, list)) {
        // Handle error...
    } else {
        // Use vec...
    }
}
```

## 4.2.4 Python set to std::unordered\_set

### API

```
template<typename T>
int
py_set_to_cpp_std_unordered_set(PyObject *op, std::unordered_set<T> &container);
```

### **Arguments**

Argument op	Argument container	Return value
A Python set containing values convertable to	The std::unordered_set	0 on success, non-zero on
type <t>.</t>	to write to.	failure.

## **Example**

Process a set of Python float:

```
void set_float_to_cpp(PyObject *arg) {
    std::unordered_set<double> set;
    if (py_set_to_cpp_std_unordered_set(arg, set)) {
        // Handle error...
    } else {
        // Use set...
    }
}
```

## 4.2.5 Python frozenset to std::unordered\_set

#### API

```
template<typename T>
int
py_frozenset_to_cpp_std_unordered_set(PyObject *op, std::unordered_set<T> &container);
```

### **Arguments**

Argument op	Argument container	Return value
A Python frozenset containing values con-	The std::unordered_set	0 on success, non-zero on
vertable to type <t>.</t>	to write to.	failure.

### **Example**

Process a frozenset of Python float:

```
void frozenset_float_to_cpp(PyObject *arg) {
    std::unordered_set<double> frozenset;
    if (py_frozenset_to_cpp_std_unordered_set(arg, frozenset)) {
        // Handle error...
    } else {
        // Use frozenset...
    }
}
```

### 4.2.6 Python dict to std::unordered\_map or std::map

#### API

```
template<typename K, typename V>
int
py_dict_to_cpp_std_map_like(PyObject *op, std::unordered_map<K, V> &container);

template<typename K, typename V>
int
py_dict_to_cpp_std_map_like(PyObject *op, std::map<K, V> &container);
```

### **Arguments**

Argument op	Argument container	Return value
A Python dict containing keys convertable to type	The std::unordered_map	0 on success, non-zero on
<k> and values convertable to type <v>.</v></k>	or std::map to write to.	failure.

#### **Example**

Process a dict of Python [int, float]:

```
void dict_int_float_to_cpp(PyObject *arg) {
    std::unordered_map<long, double> map;
    if (py_dict_to_cpp_std_map_like(arg, map)) {
        // Handle error...
    }
    // Use map...
}
```

# 4.3 C++ Containers to Python

### 4.3.1 Error Indication

All of the conversion functions from C++ to Python return an PyObject \*. If this is non-NULL it is a *new reference* and it is the responsibility of the caller to dispose off it.

On failure these functions will return NULL Reasons for failure can be:

- The new Python container can not be created with the CPython API, perhaps for memory reasons.
- A C++ object can not be converted to a Python object. I can not imagine how this would be the case.
- The converted C++ object can not be can not be inserted into the Python container. I can not imagine how this would be the case.

In the failure case a PyErr\_... will be set.

## 4.3.2 C++ std::vector or std::list to Python tuple

#### **API**

To convert to a Python tuple:

```
template<typename T>
Py0bject *
cpp_std_list_like_to_py_tuple(const std::vector<T> &container);

template<typename T>
Py0bject *
cpp_std_list_like_to_py_tuple(const std::list<T> &container);
```

### **Arguments**

Argument container	Return value
A std::vector or std::list of type <t> convertable</t>	The new Python container, NULL on failure in which case
to an appropriate Python type.	a PyErr will be set.

#### **Example**

Create a tuple of Python float:

```
PyObject *vector_double_to_tuple() {
    std::vector<double> vec;
    // Populate vec
    // ...
    return cpp_std_list_like_to_py_tuple(vec);
}
```

## 4.3.3 C++ std::vector or std::list to Python list

### API

To convert to a Python list:

```
template<typename T>
Py0bject *
cpp_std_list_like_to_py_list(const std::vector<T> &container);

template<typename T>
Py0bject *
cpp_std_list_like_to_py_list(const std::list<T> &container);
```

#### **Arguments**

Argument container	Return value
A std::vector or std::list of type <t> convertable</t>	The new Python container, NULL on failure in which case
to an appropriate Python type.	a PyErr will be set.

### **Example**

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Create a list of Python float:

```
PyObject *vector_double_to_list() {
    std::vector<double> vec;
    // Populate vec
    // ...
    return cpp_std_list_like_to_py_list(vec);
}
```

# 4.3.4 C++ std::unordered\_set to Python set

### API

```
template<typename T>
Py0bject *
cpp_std_unordered_set_to_py_set(const std::unordered_set<T> &container);
```

### **Arguments**

Argument container	Return value
A std::unordered_set of type <t> convertable to an</t>	The new Python container, NULL on failure in which case
appropriate Python type.	a PyErr will be set.

### **Example**

Create a set of Python float:

```
PyObject *vector_double_to_list() {
    std::unordered_set<double> set;
    // Populate set
    // ...
    return cpp_std_unordered_set_to_py_set(set);
}
```

# 4.3.5 C++ std::unordered\_set to Python frozenset

### API

```
template<typename T>
PyObject *
cpp_std_unordered_set_to_py_frozenset(const std::unordered_set<T> &container);
```

### **Arguments**

Argument container	Return value
A std::unordered_set of type <t> convertable to an</t>	The new Python container, NULL on failure in which case
appropriate Python type.	a PyErr will be set.

#### **Example**

Create a frozenset of Python float:

```
PyObject *vector_double_to_list() {
    std::unordered_set<double> set;
    // Populate set
    // ...
    return cpp_std_unordered_set_to_py_frozenset(set);
}
```

## 4.3.6 C++ std::unordered\_map or std::map to a Python dict

### API

```
template<typename K, typename V>
PyObject *
cpp_std_map_like_to_py_dict(const std::unordered_map<K, V> &container);

template<typename K, typename V>
PyObject *
cpp_std_map_like_to_py_dict(const std::map<K, V> &container);
```

#### **Arguments**

Argument container	Return value
A std::unordered_map or std::map of type <k, v=""></k,>	The new Python container, NULL on failure in which case
convertable to appropriate Python types.	a PyErr will be set.

#### **Example**

Create a dict of Python [int, float] from a std::unordered\_map<long, double>:

```
PyObject *map_double_to_list() {
    std::unordered_map<long, double> map;
    // Populate map
    // ...
    return cpp_std_map_like_to_py_dict(map);
}
```

Create a dict of Python [int, str] from a std::map<long, std::string>:

```
PyObject *map_double_to_list() {
    std::map<long, std::string> map;
    // Populate map
    // ...
    return cpp_std_map_like_to_py_dict(map);
}
```

**CHAPTER** 

**FIVE** 

## DESIGN

This library uses C++ templates but not in a particularly complex way. There are six essential C++ templates and a Python script is used to to auto-generate the partial template specialisations and their instantiations.

As described in the previous chapter new types can be added pretty easily, alternatively the code generator can be manipulated to do this.

Essentially there are functions to make these conversions between Python and C++:

- Objects that can be members of containers. Examples are float/double, strings and user defined types. See *User Defined Types* for more information on the latter.
- Containers. Examples are Python lists to and from C++ std::list<T>.

# **5.1 Object Conversion Source Files**

The are in python\_object\_convert.h and python\_object\_convert.cpp.

There are hand written files that contains implementations of functions to convert Python types to and from their C++ equivalent. There are three functions to each type:

- Check that a Python object is of the expected type.
- Convert a Python object to a C++ value.
- Convert a C++ value to a new Python object.

For example here are the three functions for Python int and C++ long:

```
int py_long_check(PyObject *op);
long py_long_to_cpp_long(PyObject *op);
PyObject *cpp_long_to_py_long(const long &1);
```

The implementations of these are just one line wrappers around functions or macros in the Python C API:

```
#include "Python.h"

int py_long_check(Py0bject *op) {
    return PyLong_Check(op);
}

long py_long_to_cpp_long(Py0bject *op) {
    assert(py_long_check(op));
    return PyLong_AsLong(op);
```

```
PyObject *cpp_long_to_py_long(const long &l) {
    return PyLong_FromLong(l);
}
```

The implementations for more complex types, such as string conversion, are a little more complicated but not greatly so.

### 5.2 Container Conversion Source Files

These are in python\_container\_convert.h and python\_container\_convert.cpp.

This is a hand written file that contains implementations of functions to create and access Python unary containers (such as list, tuple, set). There are a number of functions to each container, for example a list:

- Check that each Python container is of the expected type.
- Create a new Python container.
- Find the length of a Python container.
- Set a value in a Python container.
- Get a value from a Python container.

For example here are the three functions for Python lists:

```
int py_list_check(Py0bject *op);
Py0bject *py_list_new(size_t len);
Py_ssize_t py_list_len(Py0bject *list_p);
int py_list_set(Py0bject *list_p, size_t pos, Py0bject *op);
Py0bject *py_list_get(Py0bject *list_p, size_t pos);
```

The implementations of these are just one line wrappers around functions or macros in the Python C API. For example:

```
// List wrappers around PyList_Check, PyList_New, PyList_SET_ITEM, PyList_GET_ITEM
int py_list_check(PyObject *op) {
    return PyList_Check(op);
}
PyObject *py_list_new(size_t len) {
    return PyList_New(len);
}
Py_ssize_t py_list_len(PyObject *op) {
    return PyList_Size(op);
}
int py_list_set(PyObject *list_p, size_t pos, PyObject *op) {
        PyList_SET_ITEM(list_p, pos, op);
        return 0;
}
PyObject *py_list_get(PyObject *list_p, size_t pos) {
        return PyList_GET_ITEM(list_p, pos);
}
```

# 5.3 python\_convert.h

This is the top level file that gives access to the whole library. Is is a hand written file that contains templates that convert containers to and fro between Python and C++. It includes python\_object\_convert.h and python\_container\_convert.h, that declares the templates then includes auto\_py\_convert\_internal.h the auto generated file of template specialisations.

# 5.4 Python list and tuple

There are several levels of specialisation here as we want to support conversion from Python list and tuple to and from std::vector and std::list.

These functions are described in detail and, for brevity, the functions that handle sets and dicts that follow the same pattern are describe in less detail.

## 5.4.1 Conversion From C++ to Python

This provides conversion From a std::vector<T> or a std::list<T> to a Python List or Tuple. Firstly there is a highly generic handwritten function in python\_convert.h:

The template types are:

Table 1: Convert a std::vector or std::list to a Python tuple or list.

Туре	Description
ListLike	The C++ container, for example a std::vector or a
	std::list.
typename T	The C++ type of each object in the container.
<pre>PyObject *(*ConvertCppToPy)(const T &amp;)</pre>	A pointer to a function that takes a type T and returns a
	new Python PyObject*.
PyObject *(*PyUnaryContainer_New)(size_t)	A pointer to a function that returns a new Python con-
	tainer of the given length.
<pre>int(*PyUnaryContainer_Set)(PyObject *,</pre>	Sets a Python object in the Python container at the given
size_t, PyObject *)>	position.

And the parameters are:

Table 2: Function to convert a std::vector or std::list to a Python tuple or list.

Parameter	Description
list_like	The C++ container.

This returns a new PyObject or NULL on failure.

The hand written implementation looks like this:

```
template<
        template<typename ...> class ListLike,
        typename T,
        PyObject *(*ConvertCppToPy)(const T &),
        PyObject *(*PyUnaryContainer_New)(size_t),
        int(*PyUnaryContainer_Set)(PyObject *, size_t, PyObject *)>
PyObject 3
very_generic_cpp_std_list_like_to_py_unary(const ListLike<T> &list_like) {
    assert(!PyErr_Occurred());
   PyObject *ret = PyUnaryContainer_New(list_like.size());
   if (ret) {
        size_t i = 0;
        for (const auto &val: list_like) {
            PyObject *op = (*ConvertCppToPy)(val);
            if (!op) {
                // Failure, do not need to decref the contents as that will be done
                // when decref'ing the container. e.g. tupledealloc():
                // https://github.com/python/cpython/blob/main/Objects/tupleobject.c#L268
                PyErr_Format(PyExc_ValueError, "C++ value of can not be converted.");
                goto except;
#ifndef NDEBUG
            // Refcount may well be >> 1 for interned objects.
            Py_ssize_t op_ob_refcnt = op->ob_refcnt;
#endif
            // PyUnaryContainer_Set usually wraps a void function,
            // always succeeds returning non-zero.
            if (PyUnaryContainer_Set(ret, i++, op)) { // Stolen reference.
                PyErr_Format(PyExc_RuntimeError, "Can not set unary value.");
                goto except;
#ifndef NDEBUG
            assert(op->ob_refcnt == op_ob_refcnt
                   && "Reference count incremented instead of stolen.");
#endif
        }
    } else {
       PyErr_Format(
            PyExc_ValueError,
            "Can not create Python container of size %ld",
            list_like.size()
        );
        goto except;
   }
```

```
assert(!PyErr_Occurred());
assert(ret);
goto finally;
except:
    Py_XDECREF(ret);
assert(PyErr_Occurred());
    ret = NULL;
finally:
    return ret;
}
```

This template is then partially specified four ways for both Python tuple and list from both C++ std::vector<T> and std::list<T>. This is handwritten code in python\_convert.h but they are, effectively, just one-liners:

```
// C++ std::vector<T> to a Python tuple
template<typename T, PyObject *(*ConvertCppToPy)(const T &)>
PyObject *
generic_cpp_std_list_like_to_py_tuple(const std::vector<T> &container) {
    return very_generic_cpp_std_list_like_to_py_unary<</pre>
        std::vector, T, ConvertCppToPy, &py_tuple_new, &py_tuple_set
   >(container);
}
// C++ std::list<T> to a Python tuple
template<typename T, PyObject *(*ConvertCppToPy)(const T &)>
PyObject *
generic_cpp_std_list_like_to_py_tuple(const std::list<T> &container) {
   return very_generic_cpp_std_list_like_to_py_unary<</pre>
        std::list, T, ConvertCppToPy, &py_tuple_new, &py_tuple_set
   >(container);
}
// C++ std::vector<T> to a Python list
template<typename T, PyObject *(*ConvertCppToPy)(const T &)>
PvObiect *
generic_cpp_std_list_like_to_py_list(const std::vector<T> &container) {
   return very_generic_cpp_std_list_like_to_py_unary<</pre>
        std::vector, T, ConvertCppToPy, &py_list_new, &py_list_set
   >(container);
}
// C++ std::list<T> to a Python list
template<typename T, PyObject *(*ConvertCppToPy)(const T &)>
generic_cpp_std_list_like_to_py_list(const std::list<T> &container) {
   return very_generic_cpp_std_list_like_to_py_unary<</pre>
        std::list, T, ConvertCppToPy, &py_list_new, &py_list_set
   >(container);
}
```

Then these are specialised by auto-generated code in auto\_py\_convert\_internal.h for the specific types bool, long, double, std::vector<char>, std::string and so on.

For brevity only the declarations and definitions are shown for the type long. For example to create a Python tuple from a C++ std::vector the base declaration for any type T is:

```
// Base declaration
template<typename T>
PyObject *
cpp_std_list_like_to_py_tuple(const std::vector<T> &container);
```

And the declaration for type long in auto\_py\_convert\_internal.h is:

```
// Instantiations
template <>
Py0bject *
cpp_std_list_like_to_py_tuple<long>(const std::vector<long> &container);
```

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, for example for C++ type long. These are just one-liners:

```
template <>
PyObject *
cpp_std_list_like_to_py_tuple<long>(const std::vector<long> &container) {
    return generic_cpp_std_list_like_to_py_tuple<long, &cpp_long_to_py_long>(container);
}
```

That is for std::vector, for std::list the declarations and definitions are very similar. Firstly in auto\_py\_convert\_internal.h, again just showing for long:

```
// Base declaration
template<typename T>
PyObject *
cpp_std_list_like_to_py_tuple(const std::list<T> &container);

// Instantiations
template <>
PyObject *
cpp_std_list_like_to_py_tuple<long>(const std::list<long> &container);

// And so on...
```

And the declarations auto-generated in auto\_py\_convert\_internal.cpp:

```
template <>
PyObject *
cpp_std_list_like_to_py_tuple<long>(const std::list<long> &container) {
    return generic_cpp_std_list_like_to_py_tuple<long, &cpp_long_to_py_long>(container);
}
// And so on...
```

## 5.4.2 Conversion From Python to C++

The conversion from a Python list or tuple to a C++ std::vector<T> or std::list<T> follows a similar pattern as described above.

Firstly there is a highly generic handwritten function in python\_convert.h:

Template parameters are:

Table 3: Template to convert a Python tuple or list to a std::vector or std::list.

Туре	Description
ListLike	The C++ container, for example a std::vector or a
	std::list.
typename T	The C++ type of the object.
<pre>int (*Py0bject_Check)(Py0bject *)</pre>	A pointer to a function returns true if Python object can
	be converted to a C++ object of type T.
T (*PyObject_Convert)(PyObject *)	A pointer to a function to convert a Python object to a
	C++ object of type T.
<pre>int(*PyUnaryContainer_Check)(PyObject *)</pre>	A pointer to a function returns true if the Python con-
	tainer is of the relevant type (list or tuple in this case).
Py_ssize_t(*PyUnaryContainer_Size)(PyObject	A pointer to a function that returns the size of the Python
*)	container.
PyObject *(*PyUnaryContainer_Get)(PyObject	Gets a Python object in the Python container at the given
*, size_t)	position.

Parameters are:

Table 4: Function to convert a std::vector or std::list to a Python tuple or list.

Parameter	Description
op	The Python container.
list_like	The C++ container. This will be empty on failure.

This returns zero on success, non-zero on failure. Failure reasons can be:

- The Python object is not the expected container type.
- A Python object in the container is NULL.
- A Python object in the container can not be converted to a C++ type T.

The implementation looks like this:

```
template<
        template<typename ...> class ListLike,
        typename T,
        int (*PyObject_Check)(PyObject *),
        T (*PyObject_Convert)(PyObject *),
        int(*PyUnaryContainer_Check)(PyObject *),
        Py_ssize_t(*PyUnaryContainer_Size)(PyObject *),
        PyObject *(*PyUnaryContainer_Get)(PyObject *, size_t)>
int very_generic_py_unary_to_cpp_std_list_like(Py0bject *op, ListLike<T> &list_like) {
   assert(!PyErr_Occurred());
   int ret = 0;
   list_like.clear();
   Py_INCREF(op); // Increment borrowed reference
   if (!PyUnaryContainer_Check(op)) {
        PyErr_Format(
            PyExc_ValueError,
            "Can not convert Python container of type %s",
            op->ob_type->tp_name
        );
        ret = -1;
        goto except;
    for (Py_ssize_t i = 0; i < PyUnaryContainer_Size(op); ++i) {</pre>
        PyObject *value = PyUnaryContainer_Get(op, i);
        if (!value) {
            ret = -2;
            goto except;
        }
        if (!(*PyObject_Check)(value)) {
            list_like.clear();
            PyErr_Format(
                    PyExc_ValueError,
                    "Python value of type %s can not be converted",
                    value->ob_type->tp_name
            );
            ret = -3;
            goto except;
        list_like.push_back((*PyObject_Convert)(value));
        // Check !PyErr_Occurred() which could never happen as we check first.
    assert(!PyErr_Occurred());
   goto finally;
except:
    assert(PyErr_Occurred());
    list_like.clear();
finally:
   Py_DECREF(op); // Decrement borrowed reference
   return ret;
}
```

This template is then partially specified with handwritten code. Here is the handwritten code in python\_convert.h for Python tuple to a C++ std::vector or std::list. They are basically one-liners, the interesting variation is for

the std::vector where we exploit .reserve() to reduce reallocation.

```
template<typename T, int (*PyObject_Check)(PyObject *), T (*PyObject_Convert)(PyObject_
*)>
int generic_py_tuple_to_cpp_std_list_like(PyObject *op, std::vector<T> &container) {
   // Reserve the vector, but only if it is a tuple. If not then ignore it as
    // very_generic_py_unary_to_cpp_std_list_like() will error
   if (py_tuple_check(op)) {
        container.reserve(py_tuple_len(op));
   return very_generic_py_unary_to_cpp_std_list_like<</pre>
        std::vector, T, PyObject_Check, PyObject_Convert,
        &py_tuple_check, &py_tuple_len, &py_tuple_get
   >(op, container);
}
template<typename T, int (*PyObject_Check)(PyObject *), T (*PyObject_Convert)(PyObject_
*)>
int generic_py_tuple_to_cpp_std_list_like(PyObject *op, std::list<T> &container) {
   return very_generic_py_unary_to_cpp_std_list_like<</pre>
        std::list, T, PyObject_Check, PyObject_Convert,
        &py_tuple_check, &py_tuple_len, &py_tuple_get
   >(op, container);
}
```

The declarations for Python tuple to a C++ std::vector are auto-generated in auto\_py\_convert\_internal.h. Here shown just for long:

```
// Base declaration
template<typename T>
int
py_tuple_to_cpp_std_list_like(Py0bject *op, std::vector<T> &container);

// Instantiations
template <>
int
py_tuple_to_cpp_std_list_like<long>(Py0bject *op, std::vector<long> &container);
```

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, here shown just for long:

```
template <>
int

py_tuple_to_cpp_std_list_like<long>(Py0bject *op, std::list<long> &container) {
    return generic_py_tuple_to_cpp_std_list_like<
        long, &py_long_check, &py_long_to_cpp_long
    >(op, container);
}
```

# 5.5 Python set and frozenset

Here is the handwritten code in python\_convert.h supports the conversion too and from a Python set or frozenset to and from a a C++ std::unordered\_set.

## 5.5.1 Conversion From C++ to Python

The template types are:

Table 5: Convert a std::unordered\_set to a Python set or frozenset.

Туре	Description
typename T	The C++ type of each object in the container.
<pre>PyObject *(*ConvertCppToPy)(const T &amp;)</pre>	A pointer to a function that takes a type T and returns a new Python PyObject*.
PyObject *(*PyUnaryContainer_New)(size_t)	A pointer to a function that returns a new Python con-
	tainer.

And the parameters are:

Table 6: Function to convert a std::vector or std::list to a Python tuple or list.

Parameter	Description
set	The C++ container.

This returns a new PyObject or NULL on failure.

Here is the handwritten specialisations in python\_convert.h supports the conversion too and from a Python set and frozenset. They are basically one-liners.

(continues on next page)

```
generic_cpp_std_unordered_set_to_py_frozenset(const std::unordered_set<T> &set) {
    return generic_cpp_std_unordered_set_to_py_set_or_frozenset<
        T, ConvertCppToPy, &PyFrozenSet_New
        >(set);
}
```

Then declarations are auto-generated in auto\_py\_convert\_internal.h, here shown just for a Python set containing long:

```
// Base declaration
template<typename T>
PyObject *
cpp_std_unordered_set_to_py_set(const std::unordered_set<T> &container);

// Instantiations
template <>
PyObject *
cpp_std_unordered_set_to_py_set<long>(const std::unordered_set<long> &container);

// And so on..
```

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, here shown just for a Python set containing long:

## 5.5.2 Conversion From Python to C++

Template parameters are:

Table 7: Template to convert a Python set or frozenset to a std::unordered)set.

Туре	Description
typename T	The C++ type of the object.
<pre>int(*PyUnaryContainer_Check)(PyObject *)</pre>	A pointer to a function returns true if the Python con-
	tainer is of the relevant type (set or frozenset in this
	case).
<pre>int (*PyObject_Check)(PyObject *)</pre>	A pointer to a function returns true if Python object can
	be converted to a C++ object of type T.
T (*PyObject_Convert)(PyObject *)	A pointer to a function to convert a Python object to a
	C++ object of type T.

#### Parameters are:

Table 8: Function to convert a Python set or frozenset to a std::unordered)set.

Parameter	Description
op	The Python container.
set	The C++ container. This will be empty on failure.

This returns zero on success, non-zero on failure. Failure reasons can be:

- The Python object is not the expected container type.
- A Python object in the container is NULL.
- A Python object in the container can not be converted to a C++ type T.

Here are the specialisations for set and frozenset:

The declarations are auto-generated in auto\_py\_convert\_internal.h, here shown just for a Python set containing long:

```
// Base declaration
template<typename T>
int
```

(continues on next page)

```
py_set_to_cpp_std_unordered_set(
    Py0bject *op, std::unordered_set<T> &container
);

// Instantiations
template <>
int

py_set_to_cpp_std_unordered_set<long>(
    Py0bject *op, std::unordered_set<long> &container
);

// And so on..
```

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, here shown just for a Python set containing long:

```
template <>
int

py_set_to_cpp_std_unordered_set<long>(
    Py0bject *op, std::unordered_set<long> &container
) {
    return generic_py_set_to_cpp_std_unordered_set<
        long, &py_long_check, &py_long_to_cpp_long
    >(op, container);
}

// And so on..
```

# 5.6 Python dict

This supports the two-way conversion from a Python dict to and from a C++ std::unordered\_map or a std::map.

## 5.6.1 Conversion From C++ to Python

A handwritten function in python\_convert.h provides the basis for specialisation:

The template types are:

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Table 9: Convert a std::map or std::unordered\_map to a Python dict.

Туре	Description
Map	The C++ container, for example a std::map or a
	std::unordered_map.
typename K	The C++ type of the keys in the container.
typename V	The C++ type of the values in the container.
PyObject *(*Convert_K)(const T &)	A pointer to a function that takes a type K and returns a
	new Python PyObject *.
<pre>PyObject *(*Convert_V)(const T &amp;)</pre>	A pointer to a function that takes a type V and returns a
	new Python PyObject *.

And the parameters are:

Table 10: Function to convert a std::map or std::unordered\_map to a Python dict.

Parameter	Description
map	The C++ container.

This returns a new PyObject or NULL on failure.

The specialised declarations are auto-generated in auto\_py\_convert\_internal.h, here shown just for a Python dict from a std::unordered\_map or a std::map containing long, long:

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, here shown just for a Python dict from a std::unordered\_map containing long, long:

```
template <>
PyObject *
cpp_std_map_like_to_py_dict<std::unordered_map, long, long>(
    const std::unordered_map<long, long> &map
) {
    return generic_cpp_std_map_like_to_py_dict<
        std::unordered_map,</pre>
```

(continues on next page)

## 5.6.2 Conversion From Python to C++

The reverse, converting from Python to C++, is accomplished by a single handwritten template in python\_convert.h:

Template parameters are:

Table 11: Template to convert a Python dict to a std::map or a std::unordered\_map.

Туре	Description
Map	The C++ container, for example a std::map or a
	std::unordered_map.
typename K	The C++ type of the keys.
typename V	The C++ type of the values.
<pre>int (*Check_K)(PyObject *)</pre>	A pointer to a function returns true if Python object can
	be converted to a C++ object of type K.
<pre>int (*Check_V)(PyObject *)</pre>	A pointer to a function returns true if Python object can
	be converted to a C++ object of type V.
T (*Convert_K)(PyObject *)	A pointer to a function to convert a Python object to a
	C++ object of type K.
T (*Convert_V)(PyObject *)	A pointer to a function to convert a Python object to a
	C++ object of type V.

Parameters are:

Table 12: Function to convert a std::vector or std::list to a Python tuple or list.

Parameter	Description
dict	The Python container.
map	The C++ container. This will be empty on failure.

This returns zero on success, non-zero on failure. Failure reasons can be:

5.6. Python dict 51

- The Python object is not the expected container type.
- A Python object in the container is NULL.
- A Python object in the container can not be converted to a C++ type K or V.

The declarations are auto-generated in auto\_py\_convert\_internal.h, here shown just for a Python dict from a std::unordered\_map or std::map containing long, long:

```
// Base declaration
template<template<typename ...> class Map, typename K, typename V>
int
py_dict_to_cpp_std_map_like(PyObject *op, Map<K, V> &map);

// Instantiations
template <>
int
py_dict_to_cpp_std_map_like<std::unordered_map, long, long>(
    PyObject* op, std::unordered_map<long, long> &map
);

template <>
int
py_dict_to_cpp_std_map_like<std::map, long, long>(
    PyObject* op, std::map<long, long> &map
);
```

The definitions are auto-generated in auto\_py\_convert\_internal.cpp, here shown just for a Python dict from a std::unordered\_map containing long, long:

```
template <>
int

py_dict_to_cpp_std_map_like<std::unordered_map, long, long>(
    PyObject* op, std::unordered_map<long, long> &map
) {
    return generic_py_dict_to_cpp_std_map_like<
        std::unordered_map,
        long, long,
        &py_long_check, &py_long_check,
        &py_long_to_cpp_long, &py_long_to_cpp_long
    >(op, map);
}
```

# 5.7 Code Generation

If necessary run the code generator:

```
cd src/py
python code_gen.py
```

Which should give you something like:

```
venv/bin/python src/py/code_gen.py
Version: 0.4.0
Target directory "src/cpy"
Writing declarations to "src/cpy/auto_py_convert_internal.h"
Wrote 4125 lines of code with 356 declarations.
Writing definitions to "src/cpy/auto_py_convert_internal.cpp"
Wrote 3971 lines of code with 352 definitions.
Process finished with exit code 0
```

5.7. Code Generation 53

**CHAPTER** 

SIX

## **USER DEFINED TYPES**

This shows how to support conversion of containers of user defined types between Python and C++ and back.

This is probably best done by example. In this case we take an existing Python object defined in a CPython extension and develop its equivalent in C++. Of course, the opposite, having an existing C++ class and needing to develop a Python equivelent in a CPython extension, might be the use case. The principles are the same.

This example will demonstrate supporting the conversion of a list of user defined Python to and from a std::vector C++ equivalents.

There are several steps:

- Have the definitions of both the CPython and C++ equivalent objects. See the examples *A Python Class* and *The C++ Class*.
- Define the function to check the Python object type. An example is Checking the Python Type.
- Define the two conversion functions from CPython to C++ and the reverse. See the examples *From Python to C++* and *From Python to C++*.

These steps only has to be done once regardless of how many containers are to be supported.

Finally for each container conversion declare the two way template specialisation and definitions. These will be simple one-liner calls to this project's generic functions. See *From Python to C++* and *From C++ to Python*.

All this code is in the project directory in src/ext/cUserDefined.h and src/ext/cUserDefined.cpp.

# 6.1 User Defined Types in a C Extension

## 6.1.1 A Python Class

This is based on the example in the Python documentation That is varied slightly for this example:

- The module name is cUserDefined (rather than custom in the original example).
- The code for the C extension is in cUserDefined.cpp.

Otherwise it is identical to the example in the Python documentation.

In this example the CustomObject class is created in this project in src/ext/cUserDefined.cpp. It looks like this:

```
typedef struct {
    Py0bject_HEAD
    Py0bject *first; /* first name */
    Py0bject *last; /* last name */
```

```
int number;
} CustomObject;
```

This also has a method name() that combines the first and last names such as this:

```
static PyObject *
Custom_name(CustomObject *self, PyObject *Py_UNUSED(ignored)) {
   return PyUnicode_FromFormat("%S %S", self->first, self->last);
}
```

**Note:** For clarity this is equivalent to this Python code:

```
import dataclasses

@dataclasses.dataclass
class CustomObject:
    first: str
    last: str
    number: int

def name(self) -> str:
    return f'{self.first} {self.last}'
```

The setup.py file would include this Extension definition:

```
Extension(
   'cUserDefined',
   sources=[
        'src/ext/cUserDefined.cpp',
        # Include this libraries source files.
        'src/cpy/auto_py_convert_internal.cpp',
        'src/cpy/python_container_convert.cpp',
        'src/cpy/python_object_convert.cpp',
    ],
   include_dirs=[
        'src',
   ],
   extra_compile_args=extra_compile_args,
),
```

Once the Python cUserDefined extension is built it can be used in Python like this:

```
>>> import cUserDefined
>>> custom_object = cUserDefined.Custom('François', 'Truffaut', 21468)
>>> custom_object.name()
'François Truffaut'
```

So much for the CPython example, now for the equivalent code in C++.

### 6.1.2 The C++ Class

Here is the user defined pure C++ class that contains a first name, second name and a number which mirrors the CPython code above. It is declared in the File cUserDefined.h:

```
#include <string>
class CppCustomObject {
public:
   CppCustomObject(
        const std::string &first,
        const std::string &last,
        long number) : m_first(first), m_last(last), m_number(number) {}
    // Accessors
   const std::string &first() const { return m_first; }
   const std::string &last() const { return m_last; }
   long number() const { return m_number; }
    std::string name() { return m_first + " " + m_last; }
   // Other methods here...
private:
    std::string m_first;
    std::string m_last;
   long m_number;
};
```

# 6.1.3 Checking the Python Type

We need to know that any PyObject is really a well formed CustomObject. Here is the code to verify the Python type and its contents in cUserDefined.cpp. It returns 1 on success, 0 otherwise:

```
#include "cUserDefined.h"
#include "cpy/python_object_convert.h"

int py_custom_object_check(PyObject *op) {
    if (Py_TYPE(op) != &CustomType) {
        return 0;
    }
    CustomObject *p = (CustomObject *) op;
    if (!Python_Cpp_Containers::py_unicode_check(p->first)) {
        return 0;
    }
    if (!Python_Cpp_Containers::py_unicode_check(p->last)) {
        return 0;
    }
    return 1;
}
```

Now add some conversion code from the CPython CustomObject to the C++ CppCustomObject:

### 6.1.4 Conversion Code

Now, in the Python C extension add the verification and conversion code between the Python CustomObject and the C++ CppCustomObject.

This code is in cUserDefined.cpp and include the necessary files, this ensures that we have access to the C++ CppCustomObject class definition and this library's conversion machinery:

### From Python to C++

The code to convert from a Python CustomObject to a new C++ CppCustomObject:

### From C++ to Python

The code to convert from a C++ CppCustomObject to a new Python CustomObject (error checking omitted for clarity):

```
#include "cUserDefined.h"
#include "cpy/python_object_convert.h"

PyObject *
cpp_custom_object_to_py_custom_object(const CppCustomObject &obj) {
    CustomObject *op = (CustomObject *) Custom_new(&CustomType, NULL, NULL);
    if (op) {
        op->first = Python_Cpp_Containers::cpp_string_to_py_unicode8(obj.first());
        op->last = Python_Cpp_Containers::cpp_string_to_py_unicode8(obj.last());
        op->number = obj.number();
    }
    return (PyObject *) op;
}
```

# 6.1.5 Template Specialisation

Now in the file, cUserDefined.h, include this project's header file and then in this project's namespace declare the specialisations to call this library's generic functions to convert to and from containers. Specifically a std::vector of these objects. These are basically one-liners:

### From Python to C++

```
#include "python_convert.h"

// Specialised declaration in cUserDefined.h

namespace Python_Cpp_Containers {

   template<>
    int
    py_list_to_cpp_std_list_like<CppCustomObject>(
        PyObject *op, std::vector<CppCustomObject> &container
   );

} // namespace Python_Cpp_Containers
```

In the file cUserDefined.cpp implement the specialisation, this is just a one-liner calling the generic conversion code in this library with the types and functions we have created.

### From C++ to Python

And for the reverse:

```
#include "python_convert.h"

// Specialised declaration in cUserDefined.h

namespace Python_Cpp_Containers {

    // C++ to Python
    template<>
        Py0bject *
        cpp_std_list_like_to_py_list<CppCustomObject>(
            const std::vector<CppCustomObject> &container
    );
} // namespace Python_Cpp_Containers
```

In the file cUserDefined.cpp implement the specialisation, this is just a one-liner calling the generic conversion code in this library.

Note: If you wanted to support Python lists to and from C++ std::list<CppCustomObject>

Then create new specialisations of the templates with std::list<CppCustomObject> Instead of std::vector<CppCustomObject>.

**Note:** If you also wanted to support Python tuples to and from C++ std::vector<T> then specialise the templates with generic\_py\_tuple\_to\_cpp\_std\_list\_like and generic\_cpp\_std\_list\_like\_to\_py\_tuple.

Now you have all the code needed to convert sequences of these objects between C++ and Python.

# 6.2 Using the C++ Conversion Functions

## 6.2.1 From C++ to Python

Here is an example of converting a C++ std::vector<CppCustomObject> to a Python list of CustomObject:

```
std::vector<CppCustomObject> vec_cpp_custom_object;
// Populate the C++ vector
// ...
// Convert to a new Python list of Python CustomObject. This will return NULL on failure.
return Python_Cpp_Containers::cpp_std_list_like_to_py_list(vec_cpp_custom_object);
```

## 6.2.2 From Python to C++

Here is an example of converting a Python list of CustomObject to a C++ std::vector<CppCustomObject>:

```
// op is a PyObject* which is a list of Python CustomObject
// Convert to C++
std::vector<CppCustomObject> vec_cpp_custom_object;
// Populate this C++ vector from the Python list
if (! Python_Cpp_Containers::py_list_to_cpp_std_list_like(op, vec_cpp_custom_object)) {
    // Converted successfully, use the vec_cpp_custom_object
    // ...
} else {
    // Handle error condition
    // ...
}
```

## 6.2.3 Example of Round-trip Conversion

Here is a complete example that takes a list of Python CustomObject and creates a C++ std::vector<CppCustomObject> with the first name and last name reversed in C++. Then it converts that C++ std::vector<CppCustomObject> back to a new list of of Python CustomObject.

In cUserDefined.cpp:

```
}
return NULL;
}
```

Add this function to the module, in cUserDefined.cpp:

Build the cUserDefined module and try it out:

Now reverse the names in C++, the objects returned are new objects (compare with above):

And the names are reversed:

```
>>> [v.name() for v in result]
['Last First', 'Truffaut François']
```

### 6.2.4 Supporting dict[int, cUserDefined.Custom]

Now it takes very little additional work to support conversion between a Python dict[int, cUserDefined.Custom] to and from a C++ std::map<long, CppCustomObject> or, indeed, any other container.

First add two specialised declarations in cUserDefined.h:

```
namespace Python_Cpp_Containers {
    // Specialised declarations

    // Python to C++
    template <>
    int
    py_dict_to_cpp_std_map_like<std::map, long, CppCustomObject>(
```

```
PyObject* op, std::map<long, CppCustomObject> &map
);

// C++ to Python
template<>
PyObject *
cpp_std_map_like_to_py_dict<std::map, long, CppCustomObject>(
    const std::map<long, CppCustomObject> &map
);

} // namespace Python_Cpp_Containers
```

And their definitions in cUserDefined.cpp. Again these are just one-liners to this project's generic functions (expanded for clarity).

### From Python to C++

```
namespace Python_Cpp_Containers {
    // Python to C++
   template <>
   py_dict_to_cpp_std_map_like<std::map, long, CppCustomObject>(
       PyObject* op, std::map<long, CppCustomObject> &map
   ) {
        return generic_py_dict_to_cpp_std_map_like<
            std::map,
            long,
            CppCustomObject,
            &py_long_check,
            &py_custom_object_check,
            &py_long_to_cpp_long,
            &py_custom_object_to_cpp_custom_object
       >(op, map);
   }
} // namespace Python_Cpp_Containers
```

## From C++ to Python

```
namespace Python_Cpp_Containers {

    // Specialised definitions
    // C++ to Python
    template<>
        Py0bject *
        cpp_std_map_like_to_py_dict<std::map, long, CppCustomObject>(
            const std::map<long, CppCustomObject> &map
        ) {
```

#### **Example Code**

Here is an example of using both of them in the cUserDefined extension a similar way to above by creating a new dict with the names reversed in C++.

In cUserDefined.cpp:

```
static PyObject *
reverse_dict_names(PyObject *Py_UNUSED(module), PyObject *arg) {
    std::map<long, CppCustomObject> input;
   if (! Python_Cpp_Containers::py_dict_to_cpp_std_map_like(arg, input)) {
        std::map<long, CppCustomObject> output;
        for (const auto &iter: input) {
            output.emplace(
                std::make_pair(
                    iter.first,
                    CppCustomObject(
                        iter.second.last(), iter.second.first(), iter.second.number()
                )
            );
        return Python_Cpp_Containers::cpp_std_map_like_to_py_dict(output);
   }
   return NULL;
}
```

Add this in to the module methods so they look now like this:

Rebuild the module and try it:

Create a new dict with the names reversed in C++ code. The IDs show that we have new objects:

Check that the names have been reversed:

```
>>> {k: v.name() for k, v in e.items()}
{0: 'Last First', 1: 'Truffaut François'}
```

# 6.3 User Defined Types From Pure Python Types

**Todo:** User Defined Types From Pure Python Types: Add in version 0.5.0

# 6.4 Interoperation with numpy ND Arrays

**Todo:** Interoperation with numpy ND Arrays: Add the existing example code in version 0.5.0.

**CHAPTER** 

SEVEN

## **PERFORMANCE**

Here are some benchmarks for converting Python containers to and from their C++ equivalents.

The C++ code was compiled with -03 and run on Apple M1 (2020) CPU running Mac OS X 13.5.1 (Ventura) with 16 GB RAM. These test results are built with Python 3.12.

There is a lot of formal engineering data in this chapter but it starts with a summary:

# 7.1 Summary

- Fundamental types (bool, int, float, complex) can be converted at around 100m objects/sec. Converting these from Python to C++ is often around 4x faster than the reverse.
- Sequences of bytes or 8 bit character strings are converted at a memory rate of around 30 GB/sec.
- Sequences 8 bit character strings are 3x slower than bytes achieving a memory rate of around 10 GB/sec.
- 16 and 32 bit character strings are much slower than 8 bit strings (100x), around 100 MB/s.
- Dicts and sets are about 3-10x slower than lists and tuples. This can be explained by, whilst both list and dict operations are O(1), the list insert is much faster as an insert into a dict/set involves hashing.
- Memory usage matches the object data sizes.
- There are no memory leaks in this library.

## 7.2 C++ Performance Tests

## 7.2.1 Test Procedure

The main entry point to the PyCppContainers project is in src/main.cpp and runs the functional, performance and memory tests.

The performance tests are in src/cpy/tests/test\_performance.h and src/cpy/tests/test\_performance.cpp. There are a number of macros TEST\_PERFORMANCE\_\* there that control which tests are run. Running all tests takes about 6.5 hours.

The tests can be run by building and running the C++ binary from the project root:

```
cmake --build cmake-build-release --target clean -- -j 6
cmake --build cmake-build-release --target PyCppContainers -- -j 6
cmake-build-release/PyCppContainers
```

**Note:** The debug build includes more exhaustive internal tests (using assert()) but excludes the performance tests as they take a *very* long time for a debug build.

The output is large and looks like this:

	16.523 now: 16.531 diff: +0.008 test_	16.531 now: 16.535 diff: +0.004 test_	16.535 now: 16.539 diff: +0.004 test_		N/A 1 488.4 test_py_	N/A 1 249.8 test_py_	N/A 1 122.2 test_py_	N/A 1 25.6 test_py_	N/A 1 22.7 test_py_	N/A 1 0.2 test_		9.262 now: 3593.207 diff: +3583.945 main.
					N/A	N/A	N/A	N/A	N/A	N/A		
	+0.008 Peak was:	+0.004 Peak was:	+0.004 Peak was:		N/A	N/A	N/A	N/A	N/A	N/A	>():[4096]	+835.621 Peak was:
ontainers	16.531 diff:	16.535 diff:	16.539 diff:		0.002047584	:string[2048]>():[4096] 0.004002917 string[2048]>()·[8192]	1 0.008183250	<pre>std::string[2048]&gt;():[16384] 1     0.039068668</pre>	:string[2048]>():[32768] 0.044092626	:string[2048]>():[65536] 4.745317500	set std::string[1048576] led=0/24192	23880.011 (s) created: 131724750
<pre>\$ cmake-build-release/PyCppContainers&gt; C++ release tests Hello, World! Python version: 3.12.1 test_functional_all START</pre>	RSS(Mb): was: 16.523 now: -vector_to_pv_tuple <bool></bool>	RSS(Mb): was: 16.531 now: -vector_to_py_tuple <long></long>	RSS(Mb): was: 16.535 now: -vector_to_py_tuple <double></double>	8< Snip>8	TEST: 0 4096 1	tuple_str32_to_vector std::string[2048]>():[4096]   TEST: 0 8192   0.004002917   timle str32 to vector std::string[2048]>():[4192]	TEST: 0 16384 1	tuple_str32_to_vector std:: TEST: 0 32768 1	→tuple_str32_to_vector std::string[2048]>():[32768] TEST: 0 65536 1 0.044092626	tuple_str32_to_vector std::string[2048]>():[65536]   TEST:	<pre>-unordered_set_bytes_to_py_set std::string[1048576]&gt;():[4096] TAIL: Passed=24192/24192 Failed=0/24192 All tests pass.</pre>	===RSS(Mb): was: 9.262 now:  cpp  Total execution time: 23880.01  Count of unique strings created:  Bye, bye! Returning 0

The complete output can be captured to perf\_notes/cpp\_test\_results.txt with this command:

```
$ time cmake-build-release/PyCppContainers > perf_notes/cpp_test_results.txt
```

Then there is a Python script perf\_notes/write\_dat\_files\_for\_cpp\_test\_results.py that will extract all the performance data into perf\_notes/dat suitable for gnuplot. Copy those \*.dat files into docs/sphinx/source/plots/dat then cd into docs/sphinx/source/plots and run gnuplot -p \*.plt to update all the performance plots referenced in the documentation.

**Note:** See *Round-trip Python to C++ and back to Python* for the Python plots which can be built by gnuplot at the same time.

## 7.2.2 Fundamental Types

These C++ functions test the cost of converting ints, floats and bytes objects between Python and C++. These test are executed if the macro TEST\_PERFORMANCE\_FUNDAMENTAL\_TYPES is defined.

#### **Numeric Types**

complex

Type C++/Py C++ to Py Py to C++ Ratio Notes bool, bool 1.56 1.46 1.07x The mean is around 660 million/s 4.16 long, int 18.4 4.42xThe mean is around 88 million/s. 14.2 The mean is around 100 million/s. double, float 5.56 2.55xcomplex<double>, 21.2 6.42 3.30x The mean is around 72 million/s.

Table 1: Fundamental Type Conversion Time. Times in nanoseconds.

Converting from C++ to Python is always slower than from Python to C++. Presumably this reflects to cost of 'boxing' a Python object is higher that the cost of extracting ('unboxing') the object

The actual tests in src/cpy/tests/test\_performance.cpp are:

Table 2: Fundamental Type Conversion Time Test Code.

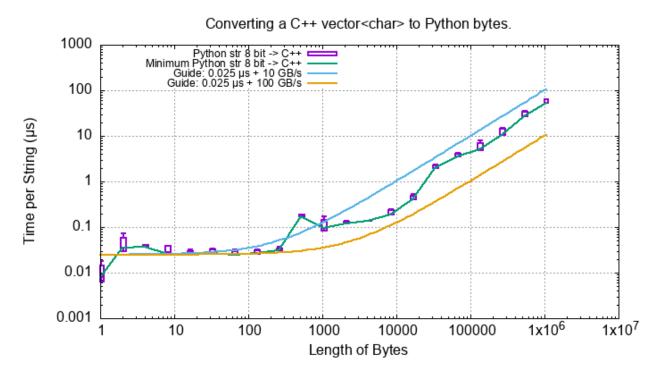
C++ to Py Test

Py to C++ Test

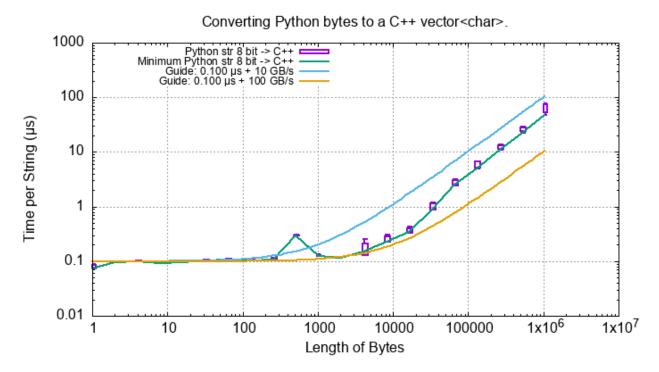
Type C++/Py	C++ to Py Test	Py to C++ Test
bool, bool	<pre>test_bool_to_py_bool_multiple()</pre>	test_py_bool_to_bool_multiple()
	calls cpp_bool_to_py_bool().	calls py_bool_to_cpp_bool().
long, int	<pre>test_long_to_py_int_multiple()</pre>	<pre>test_py_int_to_cpp_long_multiple()</pre>
	calls cpp_long_to_py_long().	calls py_long_to_cpp_long().
double, float	<pre>test_double_to_py_float_multiple()</pre>	<pre>test_py_float_to_cpp_double_multiple()</pre>
	calls cpp_double_to_py_float().	calls cpp_double_to_py_float().
complex <double>,</double>	test_complex_to_py_complex_multiple	<pre>(Dest_py_complex_to_cpp_complex_multiple()</pre>
complex	<pre>calls cpp_complex_to_py_complex().</pre>	calls py_complex_to_cpp_complex().

bytes

For a single C++ std::vector<char> to and from Python bytes of different lengths:



This shows a linear rate asymptotic to around 30 GB/s.



This is symmetric with the performance of Python to C++.

The tests are in src/cpy/tests/test\_performance.cpp:

Table 3: Bytes Conversion Time Test Code.

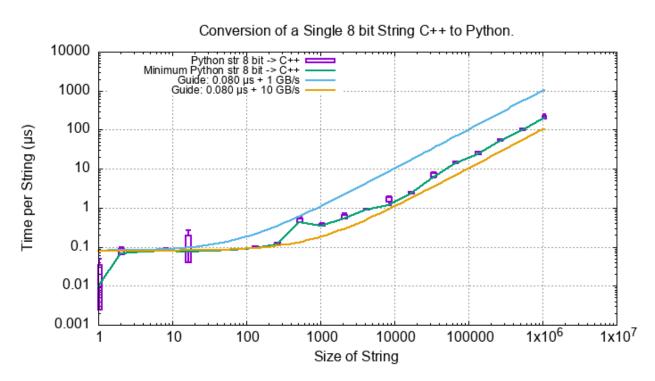
Conversion	Test Function	Calls
C++ to Python	test_cpp_vector_char_to_py_bytes_multiple	(pp_vector_char_to_py_bytes().
Python to C++	test_py_bytes_to_cpp_vector_char_multiple	(py_bytes_to_cpp_vector_char().

### **Strings**

For a single C++ std::string, std::u32string and std::u32string to and from Python str of different lengths and different word sizes.

Table 4: String Conversion Time Test Code, C++ to Python.

Type C++	Test Function	Calls
std::string	test_cpp_string_to_py_str_multiple(	()cpp_string_to_py_unicode8().
std::u16string	test_cpp_u16string_to_py_str16_mult	i <b>cplpe(i)</b> 16string_to_py_unicode16().
std::u32string	test_cpp_u32string_to_py_str32_mult	i <b>cplpe(i)</b> 32string_to_py_unicode32().



Firstly the 8 bit Unicode converts consistently at a rate of around 10 GB/s. This compares with the conversion of std::vector<char> to bytes objects at 30 GB/s (above). The threefold increase can be possibly explained by having more internal checks on unicode objects.

#### Intermezzo: Creating Python 16/32 bit Unicode strings

Initially this code was 100x slower than for 8 bit string and this section describes why. An explanation might be the way the Python Unicode C-API works. There are several ways of creating Unicode strings which are UCS1, UCS2 or UCS4 in CPython. The function PyUnicode\_FromKindAndData() is the recommended way. However if a PyUnicode\_2BYTE\_KIND or a PyUnicode\_4BYTE\_KIND this function inspects the multibyte data and if there are no code points above 0xFF then a PyUnicode\_1BYTE\_KIND is created which is not what we want.

Instead we use PyUnicode\_New with a suitable maxchar to ensure that we get the correct word size. Then we copy each character into the Unicode string in a loop.

Here is an example from this library using 16 bit unicode characters:

```
PyObject *cpp_u16string_to_py_unicode16(const std::u16string &s) {
    assert(! PyErr_Occurred());
    PyObject *ret = PyUnicode_New(s.size(), 65535);
    assert(py_unicode16_check(ret));
    for (std::u16string::size_type i = 0; i < s.size(); ++i) {</pre>
        int result = PyUnicode_WriteChar(ret, i, s[i]);
        if (result) {
            PyErr_Format(
                PyExc_SystemError,
                "PyUnicode_WriteChar() failed to write at [%ld] returning %d.",
                i, result
            );
            Py_DECREF(ret);
            return NULL;
        }
    assert(py_unicode16_check(ret));
    assert(! PyErr_Occurred());
    return ret;
}
```

This code make the conversion of std::u16string and std::u32string to Python str around 100 times slower than for 8 bit strings. This loop, the type conversions and the PyUnicode\_WriteChar internal checks is probably what is causing the slowdown.

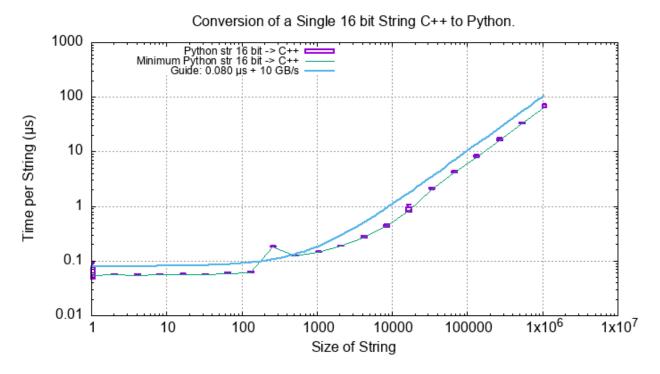
To overcome this the memory was copied directly:

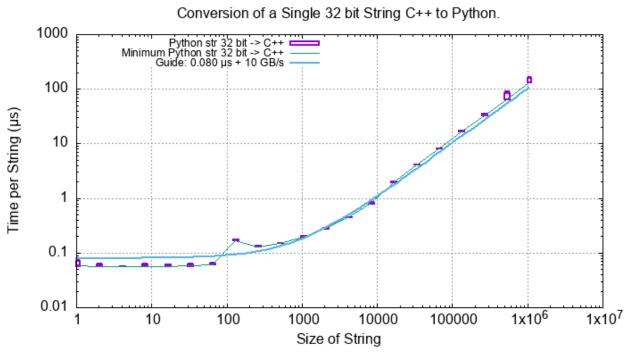
This gives 16/32 bit conversion the same performance as 8 bit conversion.

See the notes on cpp\_u16string\_to\_py\_unicode16() and cpp\_u32string\_to\_py\_unicode32() in src/cpy/python\_object\_convert.cpp for more information.

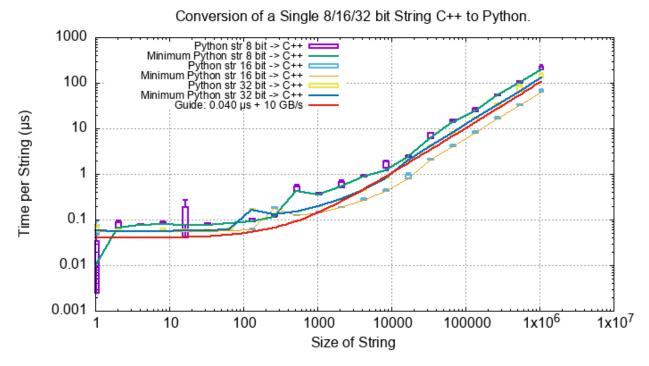
## Back to 8/16/32 bit C++ to Python Strings

So now the 16/32 bit word strings are the same speed as 8 bit strings:





Or all three 8/16/32 bit plotted together:

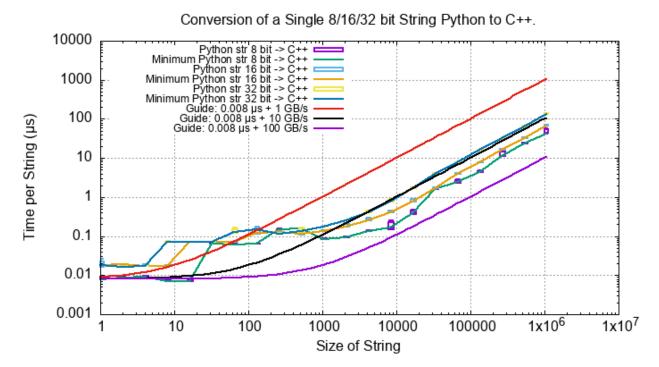


Python to C++:

Table 5: String Conversion Time Test Code, Python to C++.

Type C++	Test Function	Calls
std::string	test_py_str_to_cpp_string_multiple	()py_unicode8_to_cpp_string().
std::u16string	test_py_str16_to_cpp_u16string_mult	ippleu(n)icode16_to_cpp_u16string().
std::u32string	test_py_str32_to_cpp_u32string_mult	ippleu(n)icode32_to_cpp_u32string().

And the plot of Python str to C++ std::string, std::u16string and std::u32string:



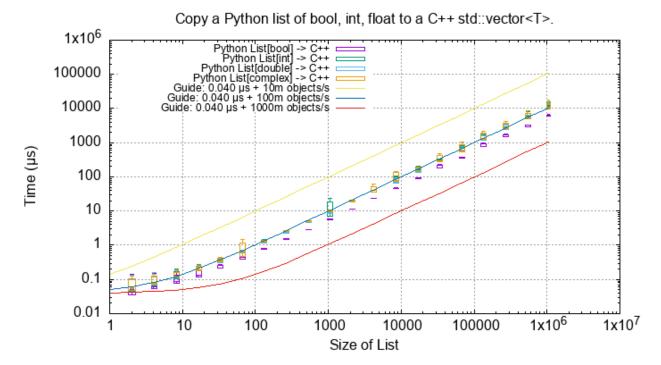
This is much more consistent, typically asymptotic to 10 GB/s. The conversion code does involve memcpy() (presumably). Here is an example from this library using 16 bit unicode characters:

The conversion time of 10 GB/s is about thrice the time for bytes to an from a std::vector<char>. Presumably this is because of the complexities of the Unicode implementation.

# 7.2.3 Python List to and from a C++ std::vector<T>

This as an extensive example of the methodology used for performance tests. Each container test is repeated 5 times and the min/mean/max/std. dev. is recorded. The min value is regarded as the most consistent one as other results may be affected by arbitrary context switching. The tests are run on containers of lengths up to 1m items.

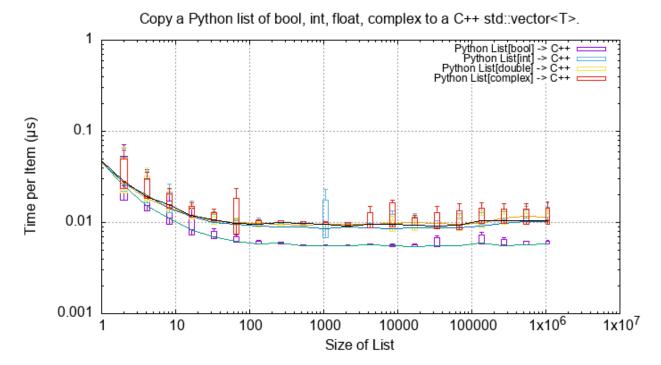
For example here is the total time to convert a list of bool, int, float and complex Python values to C++ for various list lengths:



This time plot is not that informative apart from showing linear behaviour. More useful are *rate* plots that show the total time for the test divided by the container length. These rate plots have the following design features:

- For consistency a rate scale of µs/item is used.
- The extreme whiskers show the minimum and maximum test values.
- The box shows the mean time  $\pm$ the standard deviation, this is asymmetric as it is plotted on a log scale.
- The box will often extend beyond a minimum value where the minimum is close to the mean and the maximum large.
- The line shows the minimum time per object in µs.

Here is the same data plotted as a *rate of conversion* of a list of bool, int, float and complex Python values to C++ for various list lengths:



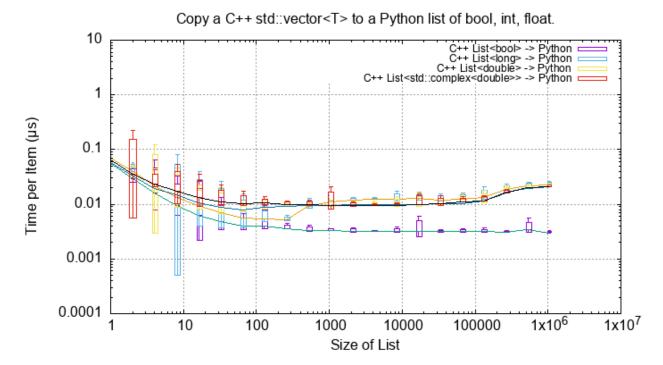
These rate plots are used for the rest of this section.

## Lists of bool, int, float and complex

The rate plot is shown above, it shows that:

- int, float and complex take 0.01 µs per object to convert from C++ to Python.
- bool objects take around 0.007 µs per object.

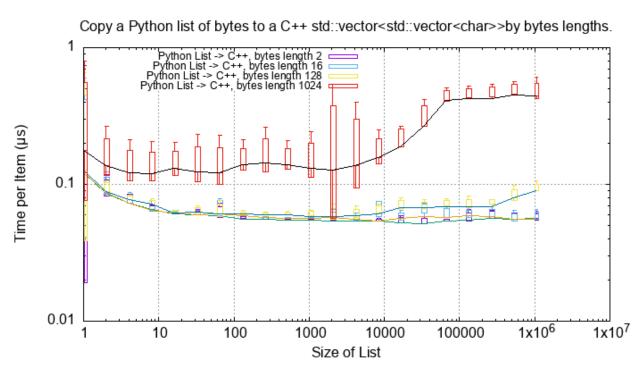
And the reverse converting a list of bool, int, float and complex from C++ to Python:



This is broadly symmetric with the Python to C++ performance except that bool values are twice as quick (typically 0.003 µs per object) compared with Python to C++.

## Lists of bytes

Another area of interest is the conversion of a list of bytes or str between Python and C++. In these tests a list of of bytes or str objects of lengths 2, 16, 128 and 1024 are used to convert from Python to C++.

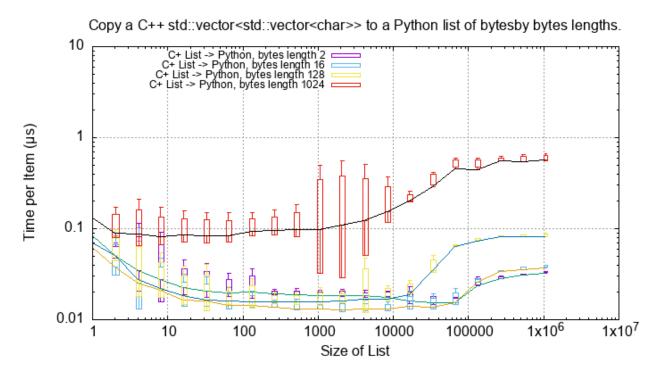


This graph shows a characteristic rise in rate for larger list lengths of larger objects. This is most likely because of memory contention issues with the larger, up to 1GB, containers. This characteristic is observed on most of the following plots, particularly with containers of bytes and str.

In summary:

Object	~Time per object (μs)	Rate Mb/s	Notes
bytes[2]	0.06	30	
bytes[16]	0.06	270	
bytes[128]	0.06	2,000	
bytes[1024]	0.15 to 0.4	2,500 to 6,800	

This is the inverse, converting a C++ std::vector<std::vector<char>> to a Python list of bytes:

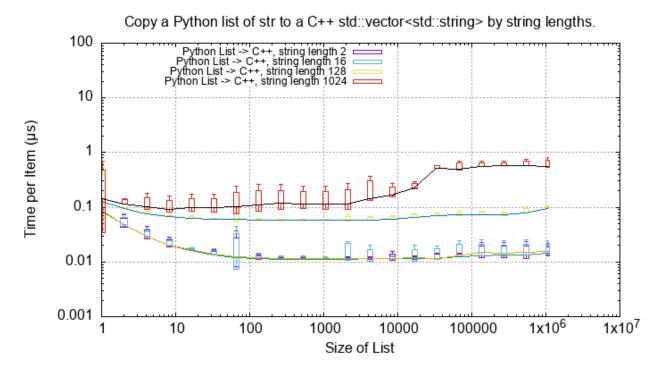


Object	~Time per object (µs)	Rate Mb/s	Notes
bytes[2]	0.015 to 0.03	67 to 133	
bytes[16]	0.015 to 0.04	400 to 133	
bytes[128]	0.02 to 0.09	1,400 to 6,400	
bytes[1024]	0.1 to 0.6	1,600 to 10,000	

This shows that converting C++ to Python is about twice as fast as the other way around. This is in line with the performance of conversion of fundamental types described above.

## Lists of str [8 bit]

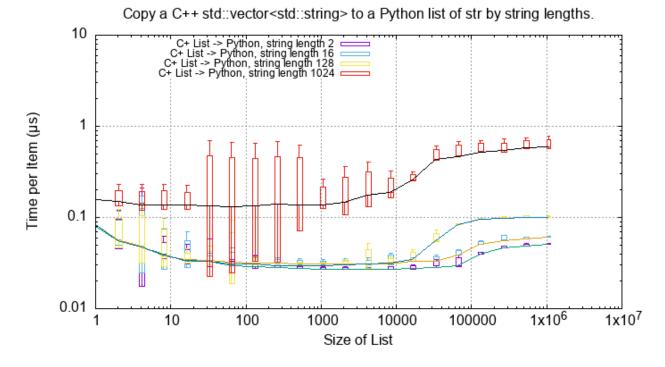
Similarly for converting a a Python list of str to and from a C++ std::vector<std::string>. First Python -> C++:



Notably with small strings (2 and 16 long) are about eight times faster that for bytes. For larger strings this perfformance is very similar to Python bytes to a C++ std::vector<std::vector<char>>:

Object	~Time per object (µs)	Rate Mb/s	Notes
str[2]	0.01	200	
str[16]	0.01	1600	
str[128]	0.08	1,600	
str[1024]	0.1 to 0.8	1,300 to 10,000	

And C++ -> Python:

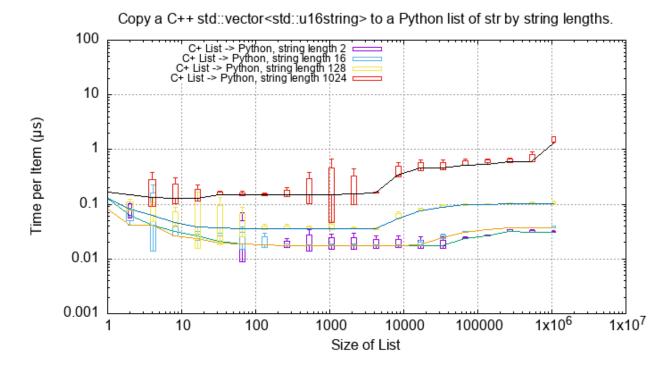


Object ~Time per object (µs) Rate Mb/s Notes str[2] 0.03 70 0.03 500 str[16] str[128] 0.03 to 0.1 1,300 to 4,000 0.15 to 0.8 1,300 to 6,800 str[1024]

Slightly slower than the twice the time for converting bytes especially for small strings this is abut twice the time for converting bytes but otherwise very similar to Python bytes to a C++ std::vector<std::vector<char>>

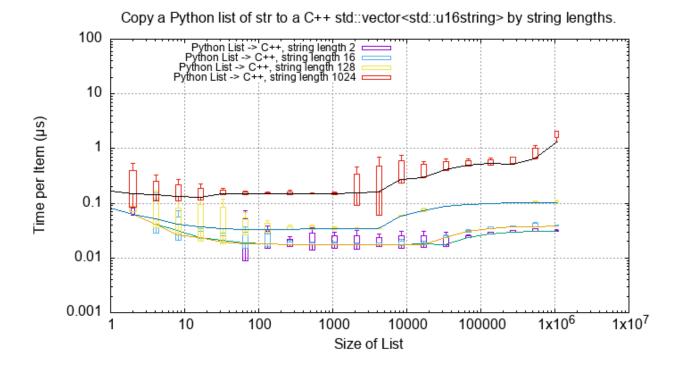
### Lists of str [16 bit]

C++ to Python:



Object ~Time per object (µs) Rate Mb/s Notes 0.03 70 str[2] 0.1 160 str[16] str[128] 0.9 140 7 145 str[1024]

This is about 100x slower than that for 8 bit strings which is aligned with the performance of *Strings*. Python to C++:



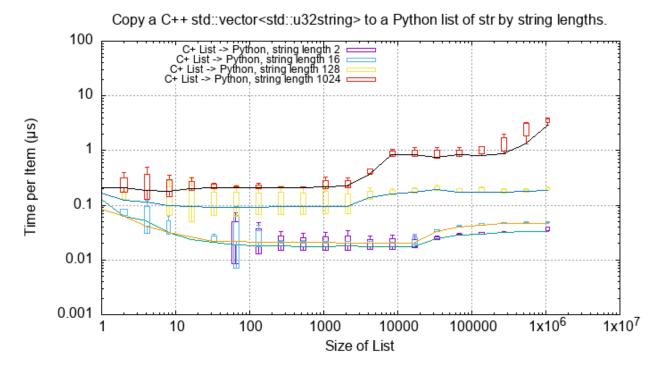
Object	~Time per object (μs)	Rate Mb/s	Notes
str[2]	0.03	70	
str[16]	0.1	160	
str[128]	0.9	140	
str[1024]	7	145	

This is symmetric performance with the C++ to Python conversion code.

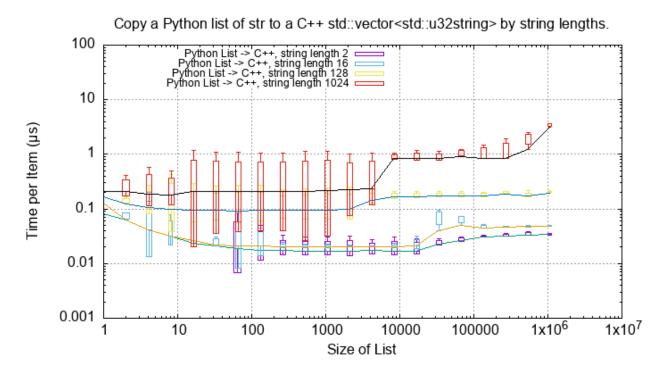
# Lists of str [32 bit]

The performance is very close to that of 16 bit strings:

C++ to Python:



Python to C++:



# 7.2.4 Python Tuple to and from a C++ std::vector<T>

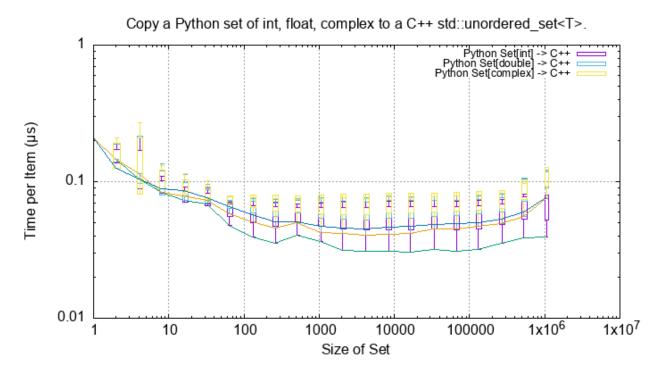
This is near identical to the performance of a list for:

- The conversion of bool, int, float and complex for Python to C++ and C++ to Python.
- The conversion of bytes for Python to C++ and C++ to Python.
- The conversion of str for Python to C++ and C++ to Python.

# 7.2.5 Python Set to and from a C++ std::unordered\_set<T>

#### Set of int, float and complex

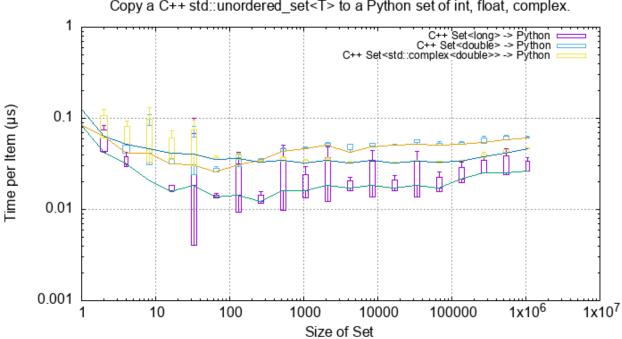
Here is the rate graph for converting a Python set to C++ std::unordered\_set<T> for Python int, float and complex objects:



Object	set (µs)	list (µs)	Ratio	Notes
int	0.03	0.01	3x	
double	0.05	0.01	5x	
complex	0.05	0.01	5x	

The cost of insertion is O(N) for both list and set but due to the hashing heeded for the set it is about 3x to 5x slower.

And the reverse, converting a C++ std::unordered\_set<T> to a Python set to for Python int, float and complex objects:



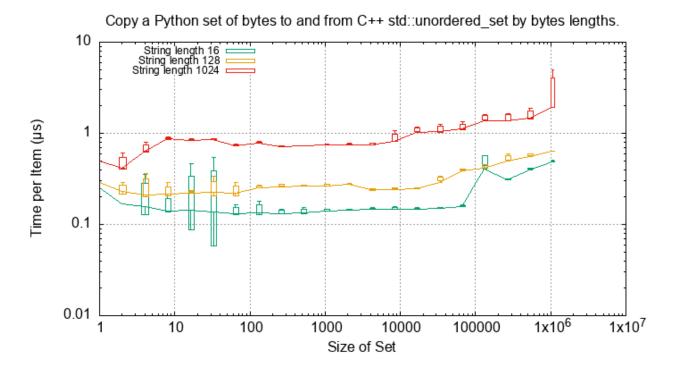
Copy a C++ std::unordered\_set<T> to a Python set of int, float, complex.

The conversion and insertion of C++ to Python is significantly faster that from Python to C++. Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
int	0.02	0.01	2x	
double	0.06	0.01	6x	
complex	0.04	0.01	4x	

## Set of bytes

Here is the rate graph for converting a Python set of bytes to C++ std::unordered\_set<std::vector<char>>:



 Object
 ~Time per object (μs)
 Rate Mb/s
 Notes

 bytes[16]
 0.2
 80

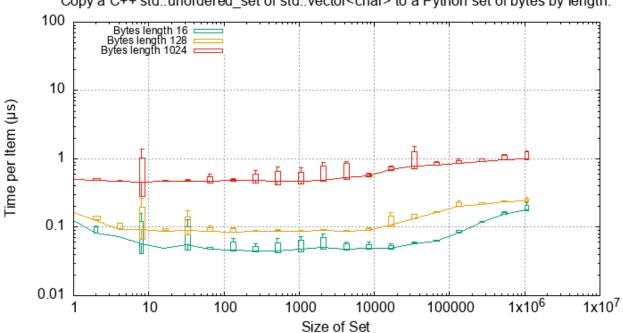
 bytes[128]
 0.3
 400

 bytes[1024]
 1.0
 1,000

Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
bytes[16]	0.2	0.06	3x	
bytes[128]	0.3	0.06	5x	
bytes[1024]	1.0	0.15 to 0.4	x2.5 to 6x	

And the reverse, converting a C++ std::unordered\_set<std::vector<char>> to a Python set of bytes:



Copy a C++ std::unordered\_set of std::vector<char> to a Python set of bytes by length.

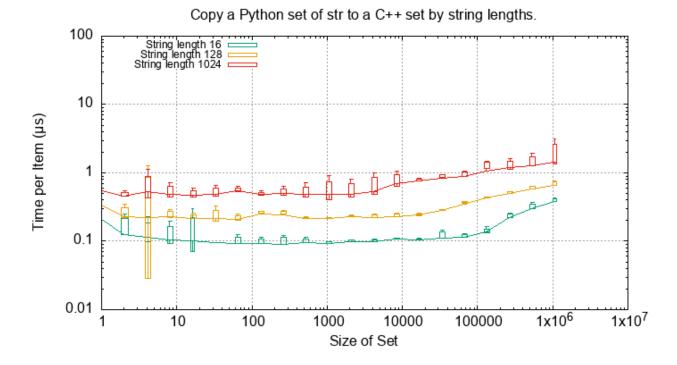
Object	~Time per object (μs)	Rate Mb/s	Notes
bytes[16]	0.05	320	
bytes[128]	0.1	1,280	
bytes[1024]	0.8	1,300	

Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
bytes[16]	0.05	0.015 to 0.04	1x to 3x	
bytes[128]	0.1	0.02 to 0.09	1x to 5x	
bytes[1024]	0.8	0.1 to 0.6	1.25x to x8	

## Set of str (8 bit)

Here is the rate graph for converting a Python set of str to C++ std::unordered\_set<std::string>:

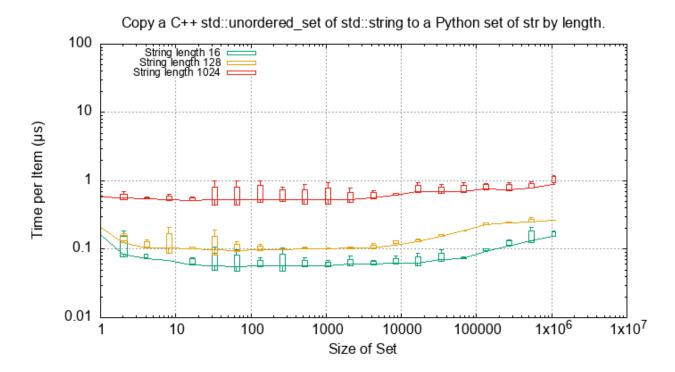


Object	~Time per object (μs)	Rate Mb/s	Notes
string[16]	0.1	160	
string[128]	0.2	640	
string[1024]	0.7 to 1.0	1,000 to 1,500	

Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
string[16]	0.1	0.01	10x	
string[128]	0.2	0.08	2.5x	
string[1024]	0.7 to 1.0	0.1 to 0.8	~8x	

And the reverse, converting a C++ std::unordered\_set<std::string> to a Python set of str:



 Object
 ~Time per object (μs)
 Rate Mb/s
 Notes

 string[16]
 0.08
 200

 string[128]
 0.1
 1,300

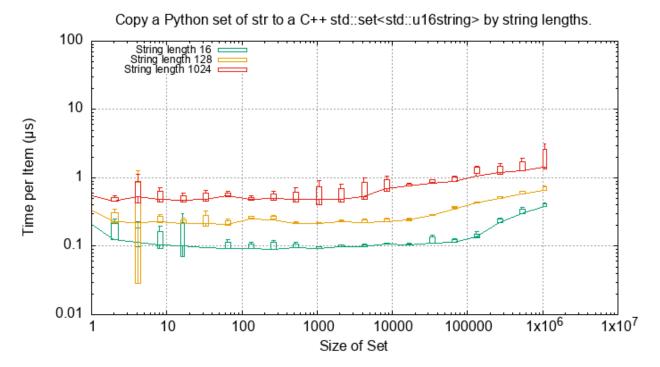
 string[1024]
 0.8
 1,300

Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
string[16]	0.08	0.03	3x	
string[128]	0.1	0.03 to 0.1	1x to 3x	
string[1024]	0.8	0.15 to 0.8	1x to 5x	

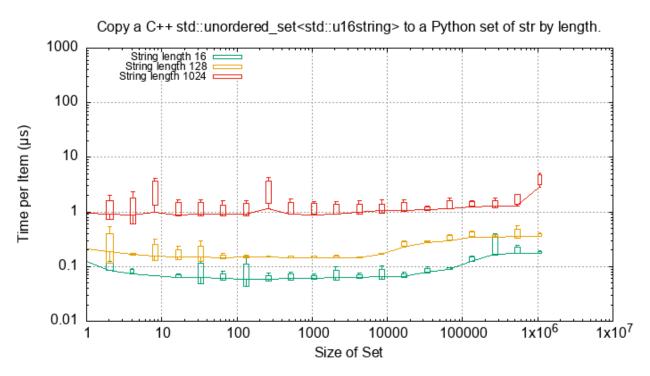
## Set of str (16 bit)

Here is the rate graph for converting a Python set of str to C++ std::unordered\_set<std::u16string>: Python to C++:



This is pretty much comparable with the 8 bit string conversion from Python to C++.

And the reverse, from C++ to Python:



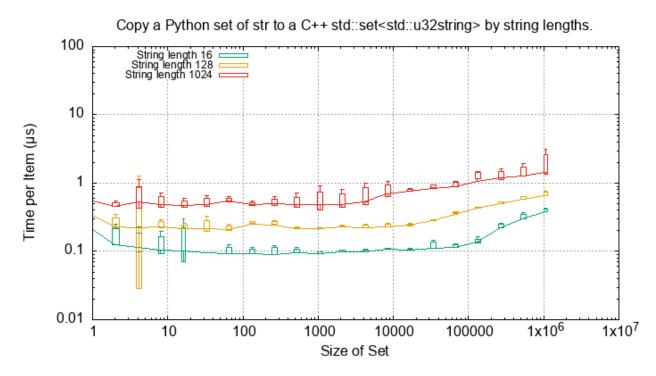
Because of the issues identified in *Strings* string16 conversion of std::u16string and std::u32string to Python str is around 100 times slower than for 8 bit strings.

Here is the comparison with 8 bit strings:

Object	8 bit (µs)	16 bit (μs)	Ratio	Notes
string[16]	0.08	0.15	2x	
string[128]	0.1	1	10x	
string[1024]	0.8	8	10x	

# Set of str (32 bit)

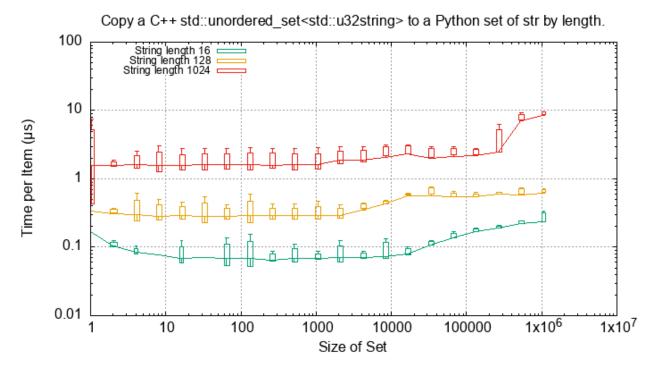
Here is the rate graph for converting a Python set of str to C++ std::unordered\_set<std::u32string>: Python to C++:



This is pretty much comparable with the 8 and 16 bit string conversion from Python to C++.

And the reverse, from C++ to Python:

C++ to Python:



This is essentially the same as for 16 bit strings.

## 7.2.6 Python Dict to and from a C++ std::unordered\_map<K, V>

Since dictionaries operate in much the same way as sets the performance is rather similar. For brevity the full results of dictionaries are not reproduced here, instead here is a summary of the performance of a dictionary compared to a set.

Object	Python to C++	C++ to Python	Notes
int, float, complex	Same as a set	Twice that of a set	
bytes	Slightly slower than a set	Twice that of a set	
str	Same as a set	Twice that of a set	

# **7.2.7 Summary**

### **Converting Individual Objects**

- bool, int, float, complex from C++ to Python is around two to three times faster than from Python to C++.
- Converting bytes from C++ to Python is the same as from Python to C++. This is memory bound at around 50 Gb/s.
- With str then Python to C++ is about twice as fast as C++ to Python. With the former performance is twice as fast as bytes, for the latter it is broadly similar to bytes conversion.
- Converting C++ to Python strings with word sizes of 16 and 32 bits is typically 10x to 100x than for 8 bits because of the Python C-API. Conversion from Python to C++ is pretty much identical for all string word sizes of 8/16/32 bits.

#### **Converting Containers of Objects**

- The performance of Python lists and tuple is the same.
- For Python list containers converting C++ to Python may be 2x faster in some cases compared to Python to C++.
- For Python list containing bytes and str objects are converted at a rate of 2 to 5 Gib/s, with some latency.
- Python set <-> C++ std::unordered\_set and Python dict <-> C++ std::unordered\_map conversion is typically x3 to x10 times slower than for lists and tuples.

# 7.3 Round-trip Python to C++ and back to Python

# 7.3.1 Python Objects

This shows the performance of converting a Python single object to its C++ equivalent and back again.

### **Python Bytes**

The Python test creates a Python bytes object of specified length(s), converts it to a C++ std::vector<char> and then back to a Python bytes object all using this C++ library code.

The test code looks like this:

```
@pytest.mark.slow
def test_new_bytes():
   results = []
   proc = psutil.Process()
   rss = proc.memory_info().rss
    for size in SIZE_DOUBLING:
        original = b' ' * size
        timer = TimedResults()
        for _r in range(REPEAT):
            time_start = time.perf_counter()
            # See below for this function.
            cPyCppContainers_new_bytes(original)
            time_exec = time.perf_counter() - time_start
            timer.add(time_exec)
        results.append((size, timer))
    # pprint.pprint(results)
   print()
   print('test_new_bytes()')
   rss_new = proc.memory_info().rss
   print(f'RSS was {rss:,d} now {rss_new:,d} diff: {rss_new - rss:+,d}')
   print(f'{"Size":<8s} {results[0][1].str_header():s} {"Min/Size e9":>12s}')
    for s, t in results:
        print(f'{s:<8d} {t} {1e9 * t.min() / s:12.1f}')</pre>
```

And the C/C++ code that provides the function cPyCppContainers.new\_bytes():

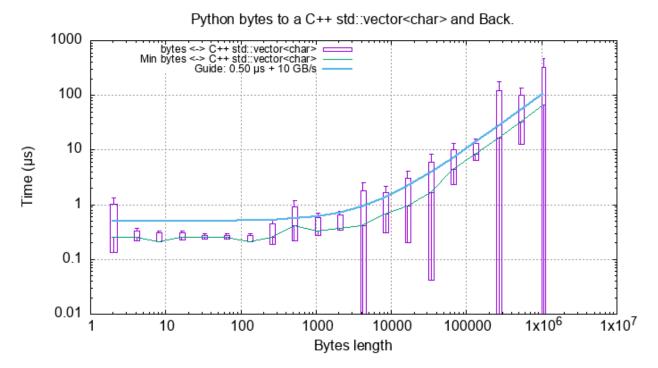
```
* Take a Python \c bytes object, convert it to a \c std::vector<char> then convert that
→back to a Python \c bytes
* object.
static PyObject *
new_bytes(PyObject *Py_UNUSED(module), PyObject *arg) {
    assert(! PyErr_Occurred());
    PyObject *ret = NULL;
    if (py_bytes_check(arg)) {
        std::vector<char> vec = py_bytes_to_cpp_vector_char(arg);
        ret = cpp_vector_char_to_py_bytes(vec);
    } else {
       PyErr_Format(
           PyExc_ValueError,
            "new_bytes() argument is not bytes but type %s", arg->ob_type->tp_name
       );
       ret = NULL;
    assert(ret || PyErr_Occurred());
    return ret;
```

And the output looks like this:

Max/Min	37.757	2.427	1.403	1.263	1.220	2.133	2.822	3.265	3.981	3.840	6.019	2.351	2.235	10.236	15.034	3.025	2.294
Max Ma	0.000008873	0.000000614	0.000000355	0.000000317	0.000000310	0.000000544	0.000000714	0.000000839	0.000001035	0.000001774	0.000002197	75600000000	0.000001180	0.000007462	0.000016537	0.000007598	0.000011073
Std.Dev.	0.000003841	0.000000168	0.000000044	0.000000028	0.000000024	0.000000128	0.000000199	0.000000255	0.000000344	0.000000563	0.000000783	0.000000225	0.000000267	0.000002982	0.000006875	0.000002208	0.000002674
Median	0.000000322	0.000000277	0.000000258	0.000000255	0.000000256	0.000000258	0.000000270	0.000000271	0.000000264	0.000000492	0.000000455	0.000000569	0.000000620	0.000000794	0.000001217	0.000002617	0.000005193
+2,703,360 Mean	0.000002002	0.000000381	0.000000276	0.000000269	0.000000267	0.000000314	0.000000359	0.000000385	0.000000421	0.000000788	0.00000003	0.000000593	0.000000708	0.000002129	0.000004238	0.000003666	0.000006307
58,531,840 diff: +2	0.000000235	0.000000253	0.000000253	0.000000251	0.000000254	0.000000255	0.000000253	0.000000257	0.000000260	0.000000462	0.000000365	0.000000407	0.000000528	0.000000729	0.000001100	0.000002512	0.000004827
180 now Count	5	5	2	2	2	2	2	2	2	2	2	2	2	5	2	2	2
RSS was 55,828,4 Size	d)		8 05.5	16 31.0 16 1.7	32 15.7	6.4	4.0 128	256	1.0 512	<ul><li>0.5</li><li>1024</li><li>0.5</li></ul>	0.5 2048	↓ 0.2 4096	↔ 0.1 8192	↓ 0.1 16384	⇒ 0.0 32768	65536 65536	→ 0.0 131072

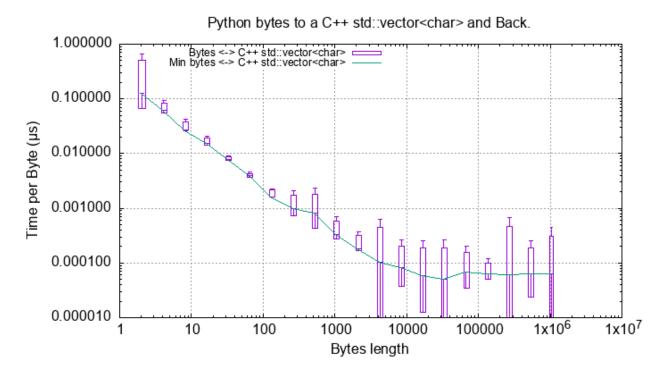
5 52144 5							
•	0.000009739	0.000045346	0.000009934	0.000077901	0.000184683	18.963	1
ა.ა							
524288 5	0.000017699	0.000066024	0.000017893	0.000107562	0.000258436	14.602	3
0.0							
1048576 5	0.000034801	0.000130087	0.000035806	0.000207265	0.000500793	14.390	3
0.0							

When plotted in time the performance looks like this:



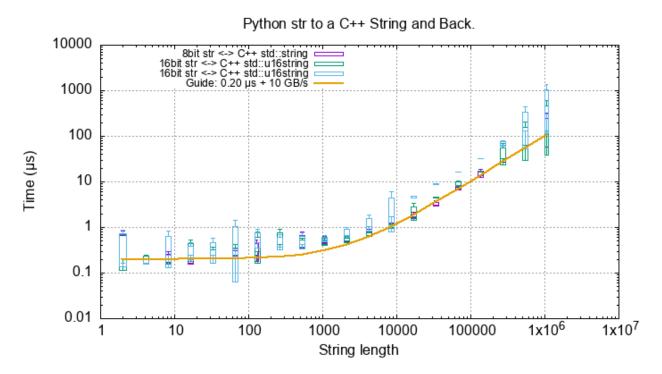
This is asymptotic to around 5 GB/s round trip conversion time.

The rate plot, that is the time value divided by the length of the bytes is:

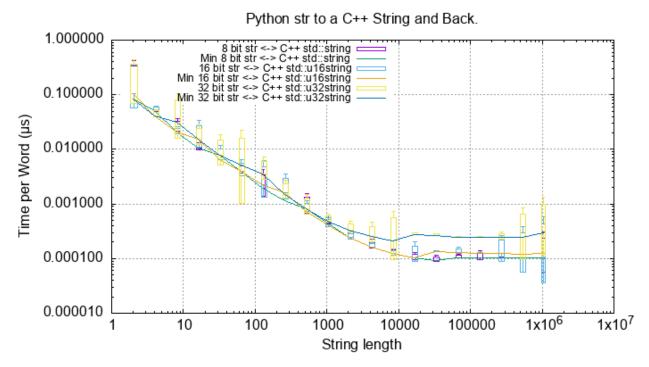


### **Python Strings**

For strings of 8/16/32 bit ord sizes the roundtrip time plot looks like this:



Again we are getting around 10 GB/s roundtrip conversion. The rate plot is rather more revealing.



This shows the 16 bit word size takes about twice the 8 bit word size and the 32 bit word size takes about four time the 8 bit word size which is exactly as expected.

# 7.3.2 Python Containers Code

This uses some methods in the cPyCppContainers module that takes a Python container, converts it to a new C++ container and then converts that to a new Python container. Timing is done in the Python interpreter.

This template converts a Python list to C++ and back:

```
#include "python_convert.h"

using namespace Python_Cpp_Containers;

template<typename T>
static Py0bject *
new_list(Py0bject *arg) {
    std::vector<T> vec;
    if (!py_list_to_cpp_std_vector(arg, vec)) {
        return cpp_std_vector_to_py_list(vec);
    }
    return NULL;
}
```

Then the extension has the following instantiations for bool, int, float, complex, bytes and str:

```
static PyObject *
new_list_bool(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<bool>(arg);
static PyObject *
new_list_float(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<double>(arg);
}
static PyObject *
new_list_int(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<long>(arg);
}
static PyObject *
new_list_complex(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<std::complex<double>>(arg);
}
static PyObject *
new_list_bytes(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<std::vector<char>>(arg);
}
static PyObject *
new_list_str(PyObject *Py_UNUSED(module), PyObject *arg) {
   return new_list<std::string>(arg);
}
static PyObject *
```

(continues on next page)

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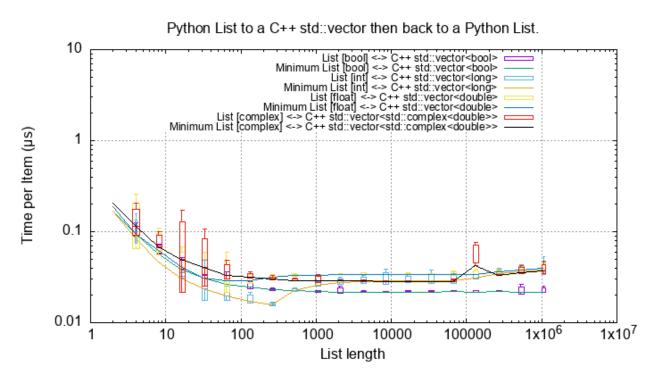
```
new_list_str16(Py0bject *Py_UNUSED(module), Py0bject *arg) {
    return new_list<std::u16string>(arg);
}
static Py0bject *
new_list_str32(Py0bject *Py_UNUSED(module), Py0bject *arg) {
    return new_list<std::u32string>(arg);
}
```

Similar code exists for Python sets and dicts of specific types. Since the tuple conversion C++ code is essentially identical to the list conversion code no performance tests are done on tuples. It might be that the Python C API for tuples is significantly different than for list but this is considered unlikely.

# 7.3.3 Python Lists

#### Python List of bool, int, float and complex

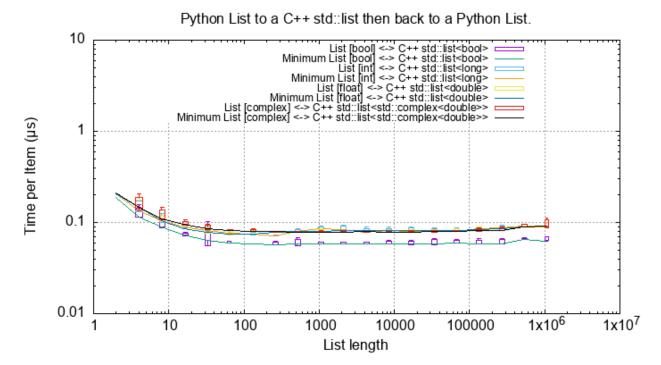
Here is the *round trip* performance of a Python list of bool, int, float and complex numbers via a C++ std::vector:



These are typically *round trip* converted at:

- 0.02 µs per object for booleans, say 50m objects a second.
- 0.04 µs per object for int, float and complex, say 25m objects a second.

And the round trip performance of a Python list of bool, int, float and complex numbers via a C++ std::list:

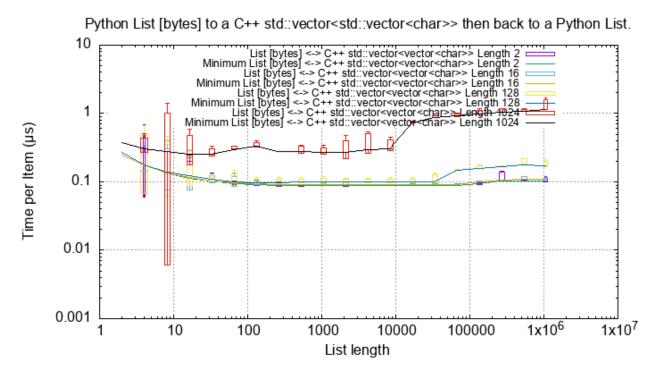


These are typically round trip converted at:

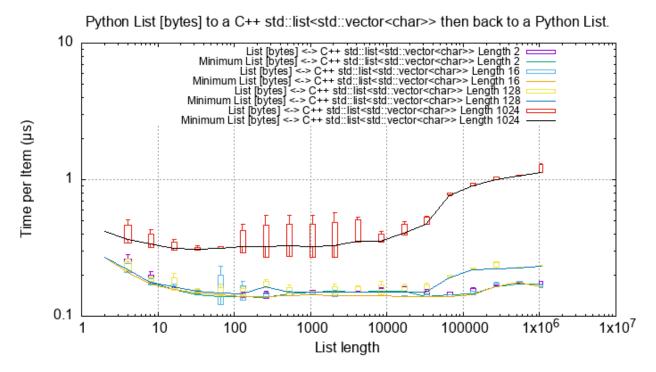
- 0.08 µs per object for booleans, say 120m objects a second. This is about 8x the cost of using a std::vector.
- 0.9 µs per object for other objects, say 10m objects a second. This is about 2x the cost of using a std::vector.

#### Python List of bytes

And a Python list of bytes for different lengths; 2, 16, 128 and 1024 bytes long via a C++ std::vector:



And a Python list of bytes for different lengths; 2, 16, 128 and 1024 bytes long via a C++ std::list:

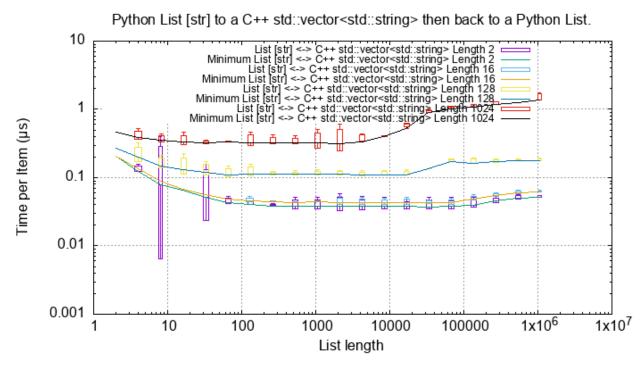


Given the size of each object this round trip time for lists can be summarised as:

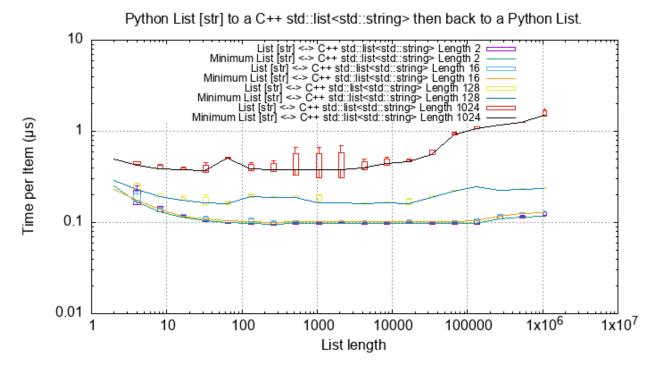
Object	Time per object (μs)	Rate (million/s)	Rate (Mb/s)	Notes
bytes[2]	0.1	10	20	
bytes[16]	0.1	10	160	
bytes[128]	0.1	10	1280	
bytes[1024]	0.4 to 2.0	0.5 to 2.5	500 to 2500	

### Python List of str

And a Python list of str for different lengths; 2, 16, 128 and 1024 via a C++ std::vector:



And via a C++ std::list:



Given the size of each object this  $round\ trip$  time for lists can be summarised as:

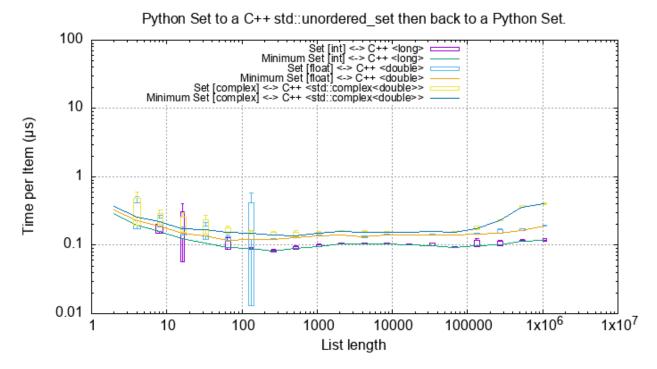
Object	Time per object (µs)	Rate (million/s)	Rate (Mb/s)	Notes
str[2]	0.05 to 0.1	10 to 20	20 to 40	
str[16]	0.05 to 0.1	10 to 20	160 to 320	
str[128]	0.2 to 0.4	2.5 to 5	320 to 640	
str[1024]	0.4 to 1.5	0.7 to 2.5	700 to 2500	

Lists of str has, essentially, the same performance as a list of bytes.

## 7.3.4 Python Sets

#### Python Set of int, float and complex

Here is the *round trip* performance of a Python set of int, float and complex numbers:



These are typically *round trip* converted at (for sets < 100,000 long):

- 0.1 µs per object for int, say 10m objects a second.
- 0.2 µs per object for float, say 5m objects a second.
- 0.3 µs per object for complex, say 3m objects a second.

The *round trip* time for a list takes 0.025 µs for int, float and complex so a set takes:

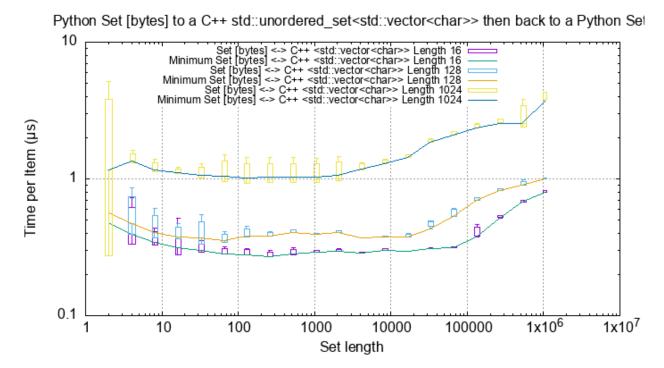
- 2.5x longer for an int
- 5x longer for a float.
- 8x longer for a complex number.

An explanation would be that the cost of hashing and insertion (and possible re-hashing the container) dominates the performance compared to the cost of object conversion.

The rise in rate towards larger sets also suggests that re-hashing becomes dominant with larger sets.

#### Python Set of bytes

And a Python set of bytes for different lengths; 16, 128 and 1024 bytes long:



Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
bytes[16]	~0.6	0.1	6x	
bytes[128]	0.6 to 1.5	0.1	6x to 15x	
bytes[1024]	1.0 to 5.0	0.4 to 2	2.5x	

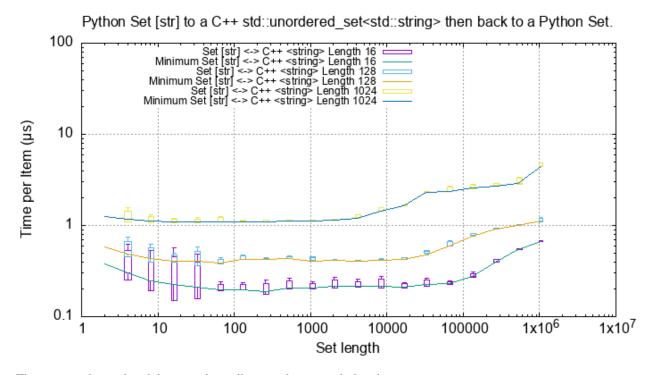
Again, the cost of hashing and insertion explains the difference.

Given the size of each object this round trip time for sets can be summarised as:

Object	Time per object (µs)	Rate (million/s)	Rate (Mb/s)	Notes
bytes[16]	~0.6	1.7	27	
bytes[128]	0.6 to 1.5	0.7 to 1.7	90 to 220	
bytes[1024]	1.0 to 5.0	0.2 to 1	200 to 1000	

#### Python Set of str

And a Python set of str (8 bit) for different lengths; 16, 128 and 1024 bytes long:



This is near identical with bytes with small strings having a slight edge.

Here is the time per object compared with a list:

Object	set (µs)	list (µs)	Ratio	Notes
str[16]	0.3	0.05 to 0.1	3x to 6x	
str[128]	0.8	0.2 to 0.4	2x to 44	
str[1024]	1.0 to 5.0	0.4 to 1.5	1x to 10x	

Again, the cost of hashing and insertion explains the difference.

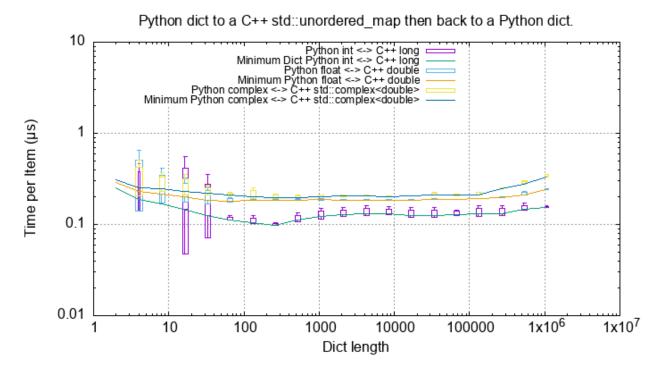
Given the size of each object this *round trip* time for sets can be summarised as:

Object	Time per object (μs)	Rate (million/s)	Rate (Mb/s)	Notes
bytes[16]	0.3	3	48	
bytes[128]	0.8	1.25	160	
bytes[1024]	1.0 to 5.0	0.2 to 1	200 to 1000	

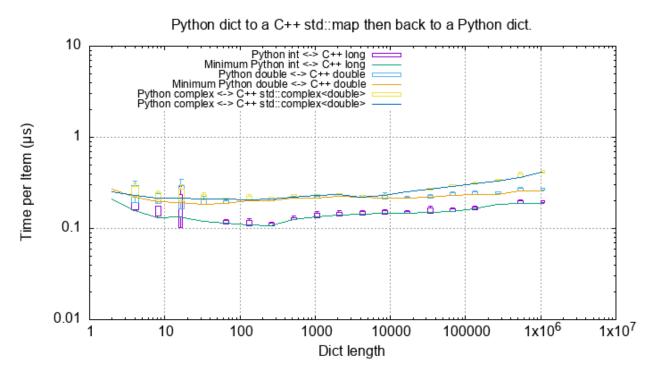
## 7.3.5 Python Dictionaries

### Python Dict of int, float and complex

Here is the round trip time for a Python dict to and from a C++ std::unordered\_map for integers, floats and complex numbers. The keys and values are the same type. This plots the *round trip* cost *per key/value pair* against dict size.



And for conversion via a C++ std::map:



The results for std::unordered\_map and std::map are nearly identical.

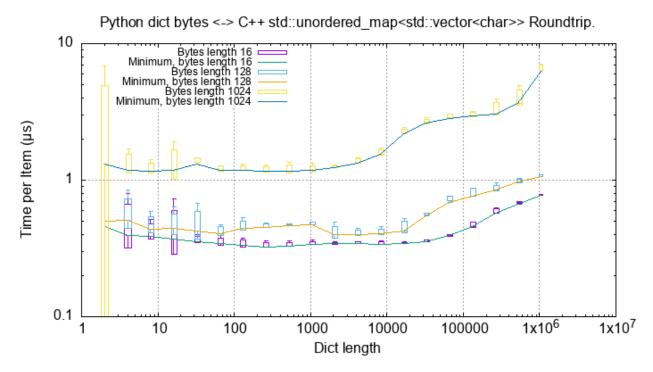
These are typically *round trip* converted at:

- 0.2 µs per object for an int or float, say 5m objects a second.
- 0.25 µs per object for a complex number, say 4m objects a second.

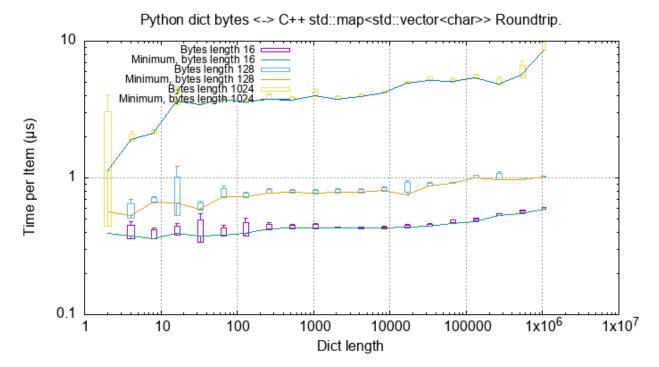
This is identical to the values for the set but includes the conversion time for both key and value. The hashing, insertion and potential re-hashing dominate the performance.

### **Python Dict of bytes**

Here is the *round trip* time for a Python dict [bytes, bytes] to and from a C++ std::unordered\_map<std::vector<char>, std::vector<char>> for different lengths; 16, 128 and 1024 bytes long. The key and the value are the same length.



And via a C++ std::map:



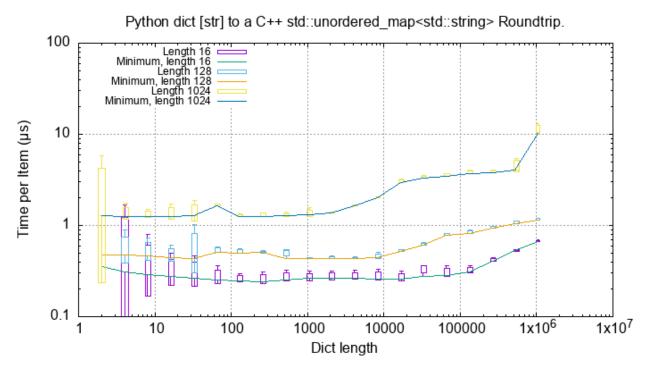
The std::map takes a shade longer than the std::unordered\_map.

This round trip time for both keys and values for dicts can be summarised as:

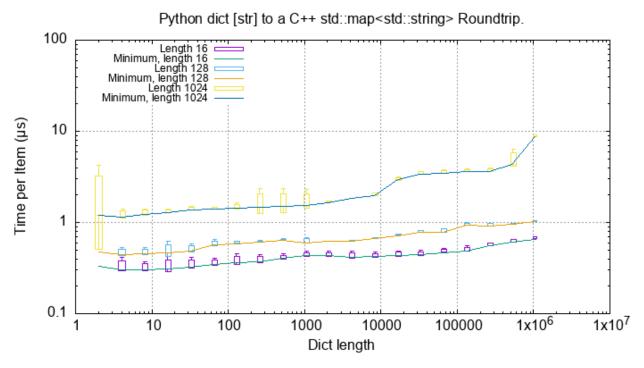
Object	Time per object (μs)	Rate (million/s)	Rate (Mb/s)	Notes
bytes[16]	0.5	2	64	
bytes[128]	0.6 to 1	1 to 1.5	256 to 425	
bytes[1024]	1 to 6	0.15 to 1.0	300 to 2000	

#### **Python Dict of str**

Here is the *round trip* time for a Python dict [str, str] (8 bit strings) to and from a C++ std::unordered\_map<std::string, std::string> for different lengths; 16, 128 and 1024 bytes long. The key and the value are the same length.



And via a C++ std::map:



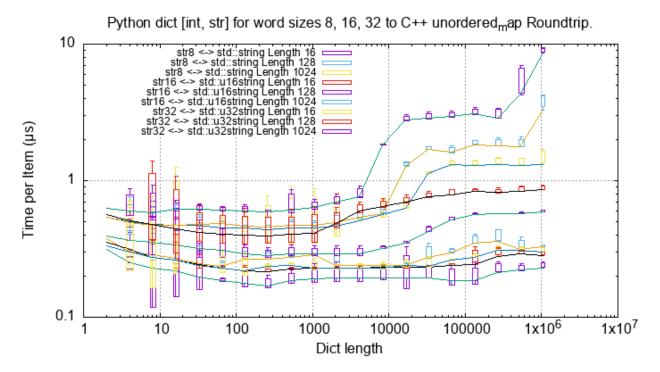
The std::map is near identical to the std::unordered\_map.

This round trip time for both keys and values for dicts can be summarised as:

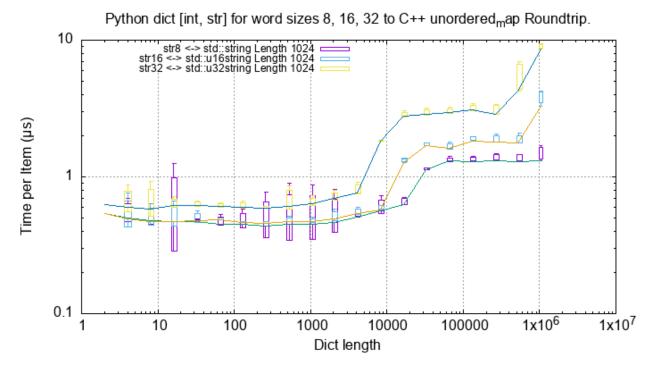
Object	Time per object (µs)	Rate (million/s)	Rate (Mb/s)	Notes
str[16]	0.3	3	96	
str[128]	0.6 to 1	1 to 1.7	256 to 440	
str[1024]	1 to 10	0.1 to 1	200 to 2000	

## **Unicode Strings of Different Codepoint Sizes**

Here is a plot of round tripping a dict of [int, str] for unicode sizes of 8 bit, 16 bit and 32 bit to a C++ std::unordered\_map and back:

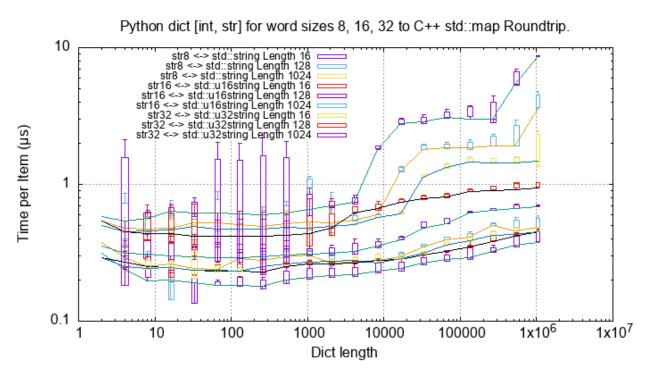


This is probably best shown in simplified form for 1024 length strings only:

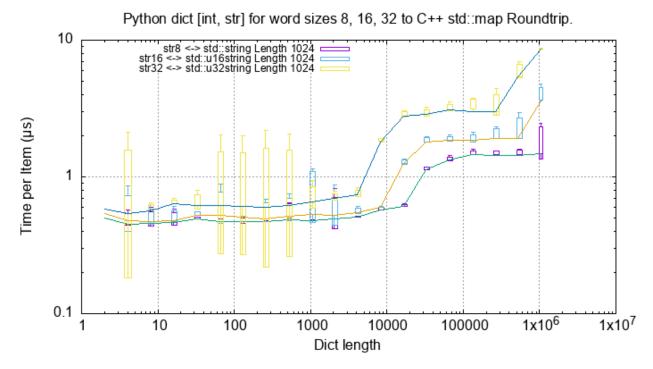


As expected, the 16 bit strings take around twice as long to convert as the 8 bit strings and the 32 bit strings take around four times as long to convert as the 8 bit strings.

And similar plots for converting to a std::map:



And, simplified for 1024 length strings.



The std::map is near identical to the std::unordered\_map.

## **7.3.6 Summary**

The fairly simple summary is that the round trip performance, as measured by the Python interpreter, agrees very closely with the total cost Python -> C++ and C++ -> Python. In some cases the performance is twice that figure but no more.

# 7.4 Memory Use

## 7.4.1 Single Bytes and Strings

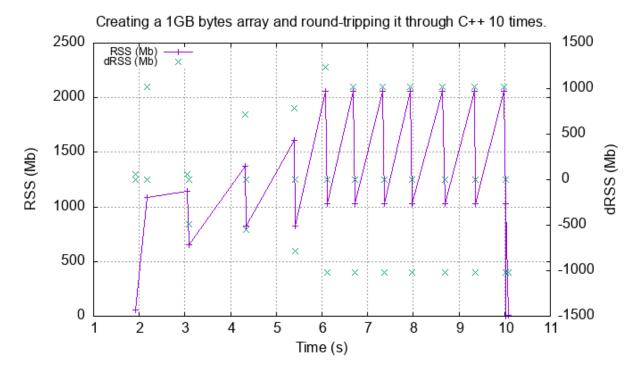
Here is a typical test that takes a single 1GB bytes object, passes it into cPyCppContainers.new\_bytes() which creates a new C++ std::vector<char> then converts that back into a new Python bytes object. This is repeated 10 times under the watchful eye of cPyMemTrace.Profile(). The test is from tests/unit/test\_with\_pymemtrace.py:

```
import cPyCppContainers
import cPyMemTrace

def test_new_bytes():
    with cPyMemTrace.Profile(4096 * 16):
        original = b' ' * (2 << 30)
        results = []
        for _r in range(10):
            cPyCppContainers.new_bytes(original)
        // This helps cPyMemTrace complete.
        gc.collect()</pre>
```

#### **Bytes**

Creating a 1GB bytes object and round-tripping it between a C++ std::vector<char> and back to a new Python bytes object.

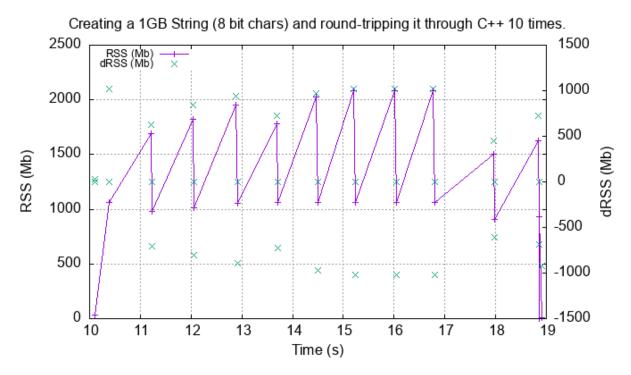


This demonstrates that there is no memory leak in converting these objects.

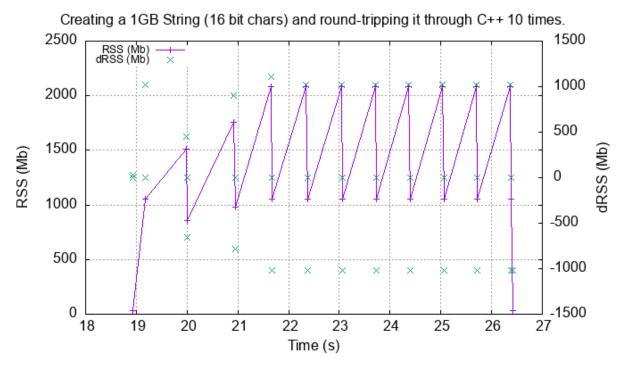
### **Strings**

Strings with 8 bit characters, this 1GB of 8 bit characters:

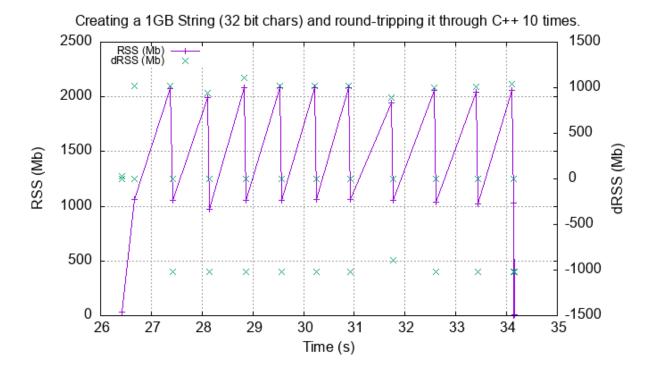
7.4. Memory Use 117



Strings with 16 bit characters, this 0.5GB of 16 bit characters:



Strings with 32 bit characters, this 0.25GB of 32 bit characters:



## 7.4.2 Containers

To examine the typical memory use a round-trip was made between Python to C++ and back to Python with a container (list, set or dict) of bytes. The container was 1m long and each member was 1k bytes, so a total of 1Gb to convert to C++ and back to a new Python container.

These tests were made using Python 3.12.

The creation/destruction was repeated 10 times and the memory profiled using pymemtrace.

The code to do this for a list is something like:

```
from pymemtrace import cPyMemTrace
import cPyCppContainers
with cPyMemTrace.Profile():
    for _r in range(10):
        original = [b' ' * 1024 for _i in range(1024 * 1024)]
        new_list = cPyCppContainers.new_list_bytes(original)
```

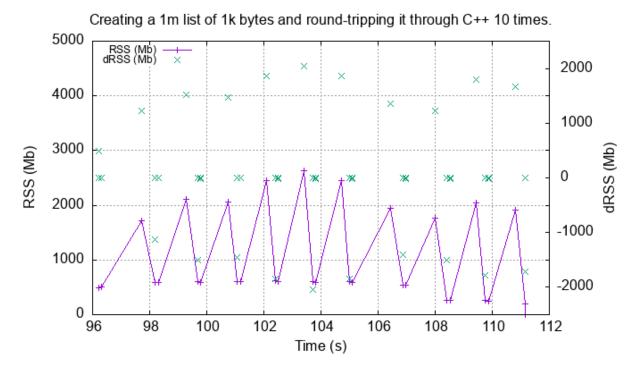
pymemtrace produces a log file of memory usage such as (not the actual data that created the plot below):

7.4. Memory Use 119

TFXT. 0	Event dEvent Clock		What	File #1	ine Function	RSS	dRSS	
•	<b>0</b> +		CALL	CALL test_with_pymemtrace.py# 15 _test_new_list_bytes	15 _test_new_list_bytes	29384704	29384704	
'REV: 83	+83		CALL	test_with_pymemtrace.py#	26 <listcomp></listcomp>	29384704	0	
VEXT: 84	+84	1.268744	RETURN	test_with_pymemtrace.py#	26 <listcomp></listcomp>	29544448	159744	
PREV: 87	+3		$C\_CALL$	test_with_pymemtrace.py#	28 new_list_bytes	29544448	0	
NEXT: 88	+4		C_RETURN	test_with_pymemtrace.py#	28 new_list_bytes	1175990272	1146445824	
NEXT: 89	+1		$C\_CALL$	test_with_pymemtrace.py#	29 perf_counter	34713600	-1141276672	
PREV: 93	+4		CALL	test_with_pymemtrace.py#	26 <listcomp></listcomp>	34713600	0	
NEXT: 94	+5		RETURN	test_with_pymemtrace.py#	26 <listcomp></listcomp>	34844672	131072	
VEXT: 95	+1		$C\_CALL$	test_with_pymemtrace.py#	27 perf_counter	34775040	-69632	
PREV: 97	+2		$C\_CALL$	test_with_pymemtrace.py#	28 new_list_bytes	34775040	0	
VEXT: 98	+3		C_RETURN	test_with_pymemtrace.py#	28 new_list_bytes	1176018944	1141243904	
VEXT: 99	+1		$C\_CALL$	test_with_pymemtrace.py#	29 perf_counter	34713600	-1141305344	

#### **Python List of bytes**

The following is a plot of RSS and change of RSS over time:



This result is unsurprising. The maximum RSS should reflect that at some point the following are held in memory:

- · Basic Python, say 30Mb
- The original Python list of bytes, 1024Mb.
- The C++ std::vector<std::string>, 1024Mb.
- The new Python list of bytes, 1024Mb.

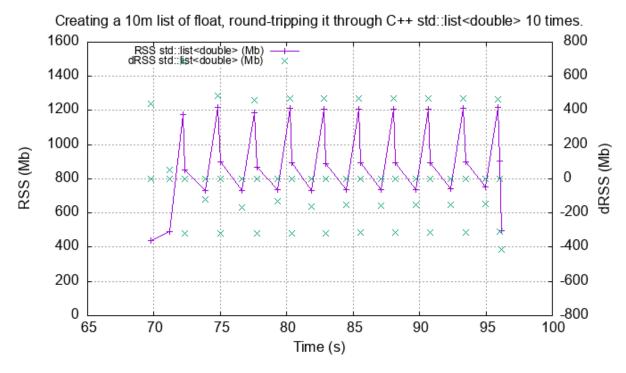
This would be a total of 3102Mb which is, broadly speaking the maximum RSS that we are seeing.

**Note:** Earlier Python versions with different memory managers displayed significantly lower maximum RSS of around 2200 MB.

#### **Python List of floats**

For comparison here is the time/memory plot of round-tripping a list of Python float as a C++ std::vector or std::list:

7.4. Memory Use 121



The memory usage is not significantly different but using a std::list takes about twice as long.

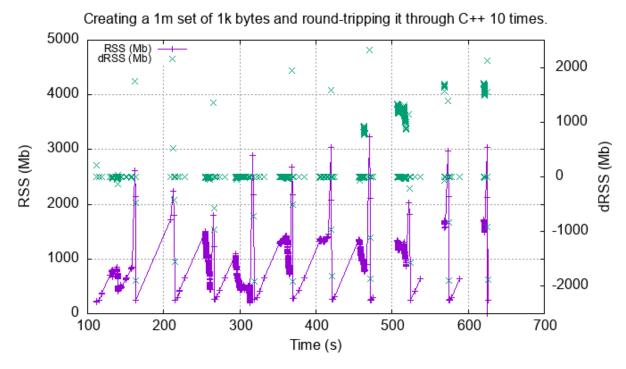
#### Python Set of bytes

A similar test was made of a gigabyte sized Python set of bytes. Each key and value were 1024 bytes long and the set was 1m long. The Python set was round-tripped to a C++ std::unordered\_set<std::string> and back to a new Python set.

The code looks like this:

```
with cPyMemTrace.Profile(4096 * 16):
    total_bytes = 2**20 * 2**10
    byte_length = 1024
    set_length = total_bytes // byte_length // 2
    random_bytes = [random.randint(0, 255) for _i in range(byte_length)]
    for _r in range(10):
        original = set()
        for i in range(set_length):
            k = bytes(random_bytes)
            original.add(k)
            # Shuffle is quite expensive. Try something simpler:
            # chose a random value and increment it with roll over.
            index = random.randint(0, byte_length - 1)
            random_bytes[index] = (random_bytes[index] + 1) % 256
            cPyCppContainers.new_set_bytes(original)
```

The following is a plot of RSS and change of RSS over time:



In the set case constructing the original set takes around 1500Mb. So on entry to new\_set\_bytes the RSS is typically 1700Mb. Constructing the std::unordered\_set<std::string> and a new Python set takes an extra 1000Mb taking the total memory to around 2500MB. On exit from new\_set\_bytes the RSS decreases back down to 200Mb.

In theory the maximum RSS use should be:

- · Basic Python, say 30Mb
- The original Python set, 1024Mb.
- The C++ std::unordered\_set<std::string>, 1024Mb.
- The new Python dict, 1024Mb.

This would be a total of 3102Mb.

#### Python Dictionary of bytes or str

A similar test was made of a gigabyte sized Python dict of bytes. Each key and value were 1024 bytes long and the dictionary was 0.5m long. The Python dict was round-tripped to a C++ std::unordered\_map<std::vector<char>, std::vector<char>> and back to a new Python dict.

The code looks like this:

```
with cPyMemTrace.Profile(4096 * 16):
    total_bytes = 2**20 * 2**10
    byte_length = 1024
    dict_length = total_bytes // byte_length // 2
    random_bytes = [random.randint(0, 255) for _i in range(byte_length)]
    for _r in range(10):
        original = {}
        for i in range(dict_length):
            k = bytes(random_bytes)
```

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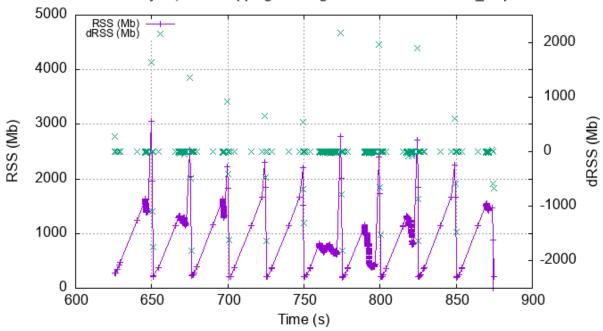
7.4. Memory Use 123

(continued from previous page)

```
original[k] = b' ' * byte_length
    # Shuffle is quite expensive. Try something simpler:
    # chose a random value and increment it with roll over.
    index = random.randint(0, byte_length - 1)
    random_bytes[index] = (random_bytes[index] + 1) % 256
cPyCppContainers.new_dict_bytes_bytes(original)
```

The following is a plot of RSS and change of RSS over time:





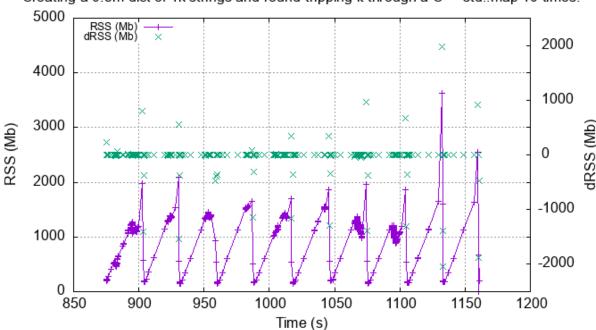
In the dictionary case constructing the original dict takes around 1500Mb. So on entry to new\_dict\_bytes\_bytes the RSS is typically 1700Mb. Constructing the std::unordered\_map<std::vector<char>, std::vector<char>> and a new Python dict takes an extra 2500Mb taking the total memory to around 4200MB. On exit from new\_dict\_bytes\_bytes the RSS decreases in two stages, destroying the std::unordered\_map<std::string, std::string> frees 2000Mb then freeing the original gives back another 2000Mb. This brings the total RSS back down to 200Mb.

In theory the maximum RSS use should be:

- Basic Python, say 30Mb
- The original Python dict, 1024Mb.
- The C++ std::unordered\_map<std::vector<char>, std::vector<char>>, 1024Mb.
- The new Python dict, 1024Mb.

This would be a total of 3102Mb. The fact that we are seeing around 4200Mb, 35% more, is probably due to over-allocation either any or all of the Python dict or bytes allocators or the C++ std::unordered\_map<T> or std::vector<char> allocators.

Similar results are obtained for a Python dict was round-tripped to a C++ std::map<std::string, std::string> and back to a new Python dict.



Creating a 0.5m dict of 1k strings and round-tripping it through a C++ std::map 10 times.

This is broadly similar to the results for std::unordered\_map<std::vector<char>, std::vector<char>>.

All these graphs show that there are no memory leaks.

### **Containers of Just One Object**

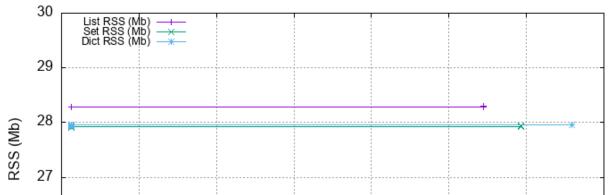
This test was to create a list, set or dict with one entry of 1024 bytes and then convert it 10,000,000 times to a C++ container and then back to Python. The memory was monitiored with pymemtrace set up to spot and changes in RSS of >=4096 bytes.

For example here is the code for a list:

```
original = [b' ' * 1024]
with cPyMemTrace.Profile():
    for _r in range(10_000_000):
        cPyCppContainers.new_list_bytes(original)
    # Tends to force an event in pymemtrace.
    gc.collect()
```

The following is a plot of RSS and change of RSS over time for list, set, dict:

7.4. Memory Use 125



Time (s)

ry usage of creating containers with one item of 1k bytes and round-tripping it through C++ 10,000

This graph shows that there are no memory leaks on container construction.

#### **CHAPTER**

# **EIGHT**

## **HISTORY**

## 8.1 Version 0.4.0

- Support std::u16string and std:u32string.
- Renaming to make more consistent use of "PyCppContainers".
- Fairly major documentation and test changes.

# 8.2 Version 0.3.2

• Support Python versions 3.8 to 3.13.

## 8.3 Version 0.3.1

• First release.

# 8.4 Version 0.3.0

• Migrate to CLion

## 8.5 Version 0.2.0

- Unpublished.
- Date: Mon Mar 15 10:04:48 2021 +0000

# 8.6 Version 0.1.0

- Unpublished.
- Date: Thu Nov 22 10:54:39 2018 +0000

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## **CHAPTER**

# **NINE**

# **TODO**

**Todo:** User Defined Types From Pure Python Types: Add in version 0.5.0

original entry

**Todo:** Interoperation with numpy ND Arrays: Add the existing example code in version 0.5.0.

original entry

130 Chapter 9. TODO

# **CHAPTER**

# **TEN**

# **INDICES AND TABLES**

- genindex
- modindex
- search