Nuitka Developer Manual

Contents

The purpose of this developer manual is to present the current design of Nuitka, the coding rules, and the motivations for choices made. It is intended to be a guide to the source code, and to give explanations that don't fit into the source code in comments form.

It should be used as a reference for the process of planning and documenting decisions we made. Therefore we are e.g. presenting here the type inference plans before implementing them. And we update them as we proceed.

It grows out of discussions and presentations made at conferences as well as private conversations or discussions on the mailing list or bug tracker.

Milestones

1. Feature parity with CPython, understand all the language construct and behave absolutely compatible.

   Feature parity has been reached for CPython 2.6 and 2.7. We do not target any older CPython release. For CPython 3.2 it also has been reached. We do not target older CPython 3.1 and 3.0 releases.

   This milestone was reached.

2. Create the most efficient native code from this. This means to be fast with the basic Python object handling.

   This milestone was reached.

3. Then do constant propagation, determine as many values and useful constraints as possible at compile time and create more efficient code.

   This milestone is considered almost reached.

4. Type inference, detect and special case the handling of strings, integers, lists in the program.

   This milestone is considered in progress

5. Add interfacing to C code, so Nuitka can turn a `ctypes` binding into an efficient binding as written with C.

   This milestone is planned only.

6. Add hints module with a useful Python implementation that the compiler can use to learn about types from the programmer.

   This milestone is planned only.

Version Numbers

For Nuitka we use a defensive version numbering system to indicate that it is not yet ready for everything. We have defined milestones and the version numbers should express which of these, we consider done.

- So far:

  Before milestone 1, we used "0.1.x" version numbers. After reaching it, we used "0.2.x" version numbers.

  Before milestone 2 and 3, we used "0.3.x" version numbers. After almost reaching 3, and beginning with 4, we use "0.4.x" version numbers.

- Future:

  When we start to have sufficient amount of type inference in a stable release, that will be "0.5.x" version numbers. With `ctypes` bindings in a sufficient state it will be "0.6.x".

- Final:

  We will then round it up and call it "Nuitka 1.0" when this works as expected for a bunch of people. The plan is to reach this goal during 2014. This is based on lots of assumptions that may not hold up though.

Of course, this may be subject to change.

Current State

Nuitka top level works like this:

- `nuitka.tree.Building` outputs node tree
- `nuitka.optimization` enhances it as best as it can
- `nuitka.finalization` marks the tree for code generation
- `nuitka.codegen.CodeGeneration` creates identifier objects and code snippets
- `nuitka.codegen.Generator` knows how identifiers and code is constructed
- `nuitka.MainControl` keeps it all together

This design is intended to last.

Regarding Types, the state is:

- Types are always `PyObject *`, implicitly.
- The only more specific use of type is "compile time constant", which can be used to predict some operations, conditions, etc.
- Every operation is expected to have `PyObject *` as result, if it is not a constant, then we know nothing about it.

The limitation to only `PyObject *` will go away.

Coding Rules

These rules should generally be adhered when working on Nuitka code. It's not library code and it's optimized for readability, and avoids all performance optimization for itself.

Line Length

No more than 120 characters. Screens are wider these days, but most of the code aims at keeping the lines below 100.

Indentation

No tabs, 4 spaces, no trailing white space.

Identifiers

Classes are camel case with leading upper case. Methods are with leading verb in lower case, but also camel case. Around braces, and after comma, there is spaces for better readability. Variables and parameters are lower case with "_" as a separator.

```python
class SomeClass:
```

```
    def doSomething(some_parameter):
        some_var = ( "foo", "bar" )
```

Base classes that are abstract end in `Base`, so that a meta class can use that convention.

Function calls use keyword argument preferably. These are slower in CPython, but more readable:

```
return Generator.getSequenceCreationCode(
    sequence_kind      = sequence_kind,
    element_identifiers = identifiers,
    context            = context
)
```

The `=` are all aligned to the longest parameter names without extra spaces for it.

When the names don't add much value, sequential calls should be done, but ideally with one value per line:

```
return Identifier(
    "TO_BOOL( %s )" % identifier.getCodeTemporaryRef(),
    0
)
```

Here, `Identifier` will be so well known that the reader is expected to know the argument names and their meaning, but it would be still better to add them.

Contractions should span across multiple lines for increased readability:

```
result = [
    "PyObject *decorator_%d" % ( d + 1 )
    for d in
    range( decorator_count )
]
```

Module/Package Names

Normal modules are named in camel case with leading upper case, because their of role as singleton classes. The difference between a module and a class is small enough and in the source code they are also used similarly.

For the packages, no real code is allowed in them and they must be lower case, like e.g. `nuitka` or `codegen`. This is to distinguish them from the modules.

Packages shall only be used to group packages. In `nuitka.codegen` the code generation packages are located, while the main interface is `nuitka.codegen.CodeGeneration` and may then use most of the entries as local imports.

The use of a global package `nuitka`, originally introduced by Nicolas, makes the packaging of Nuitka with `distutils` etc. easier and lowers the requirements on changes to the `sys.path` if necessary.

> ### *Note*
>
> There are not yet enough packages inside Nuitka, feel free to propose changes as you see fit.

Names of modules should be plurals if they contain classes. Example is `Nodes` contains `Node` classes.

Prefer list contractions over built-ins

This concerns `map`, `filter`, and `apply`. Usage of these built-ins is highly discouraged within Nuitka source code. Using them is considered worth a warning by "PyLint" e.g. "Used builtin function 'map'". We should use list comprehensions instead, because they are more readable.

List contractions are a generalization for all of them. We love readable and with Nuitka as a compiler will there won't be any performance difference at all.

I can imagine that there are cases where list comprehensions are faster because you can avoid to make a function call. And there may be cases, where map is faster, if a function must be called. These calls can be very expensive, and if you introduce a function, just for `map`, then it might be slower.

But of course, Nuitka is the project to free us from what is faster and to allow us to use what is more readable, so whatever is faster, we don't care. We make all options equally fast and let people choose.

For Nuitka the choice is list contractions as these are more easily changed and readable.

Look at this code examples from Python:

```python
class A:
    def getX(self):
        return 1
    x = property( getX )

class B(A):
   def getX(self):
      return 2



A().x == 1 # True
B().x == 1 # True (!)
```

This pretty much is what makes properties bad. One would hope `B().x` to be `2`, but instead it's not changed. Because of the way properties take the functions and not members, and because they then are not part of the class, they cannot be overloaded without re-declaring them.

Overloading is then not at all obvious anymore. Now imagine having a setter and only overloading the getter. How to you easily update the property?

So, that's not likable about them. And then we are also for clarity in these internal APIs too. Properties try and hide the fact that code needs to run and may do things. So lets not use them.

For an external API you may exactly want to hide things, but internally that has no use, and in Nuitka, every API is internal API. One exception may be the `hints` module, which will gladly use such tricks for easier write syntax.

The "git flow" model

- The flow was is used for releases and occasionally subsequent hot fixes.

  A few feature branches were used so far. It allows for quick delivery of fixes to both the stable and the development version, supported by a git plugin, that can be installed via "apt-get install git-flow" on latest Debian Testing at least.

- Stable (master branch)

The stable version, is expected to pass all the tests at all times and is fully supported. As soon as bugs are discovered, they are fixed as hot fixes, and then merged to develop by the "git flow" automatically.

• Development (develop branch)

The future release, supposedly in almost ready for release state at nearly all times, but this is as strict. It is not officially supported, and may have problems and at times inconsistencies.

• Feature Branches

On these long lived developments that extend for multiple release cycles or contain changes that break Nuitka temporarily. They need not be functional at all.

Current Feature branches:

  • `feature/ctypes_annotation`: Achieve the inlining of ctypes calls, so they become executed at no speed penalty compared to direct calls via extension modules. This being fully CPython compatible and pure Python, is considered the "Nuitka" way of creating extension modules that provide bindings.

Checking the Source

The checking for errors is currently done with "PyLint". In the future, Nuitka will gain the ability to present its findings in a similar way, but this is not a priority, and not there yet.

So, we currently use "PyLint" with options defined in a script.

```
./misc/check-with-pylint --hide-todos
```

Ideally the above command gives no warnings. This has not yet been reached. The existing warnings often still serve as a kind of "TODO" items. We are not white listing them, because they indicate a problem that should be solved.

If you submit a patch, it would be good if you checked that it doesn't introduce new warnings, but that is not strictly required. it will happen before release, and that is considered enough. You probably are already aware of the beneficial effects.

Running the Tests

This section describes how to run Nuitka tests.

Running all Tests

The top level access to the tests is as simple as this:

```
./misc/check-release
```

For fine grained control, it has the following options:

```
-h, --help            show this help message and exit
--skip-basic-tests    The basic tests, execute these to check if Nuitka is
                      healthy. Default is True.
--skip-syntax-tests   The syntax tests, execute these to check if Nuitka
                      handles Syntax errors fine. Default is True.
--skip-program-tests  The programs tests, execute these to check if Nuitka
                      handles programs, e.g. import recursions, etc. fine.
```

```
                            Default is True.
--skip-reflection-test
                    The reflection test compiles Nuitka with Nuitka, and
                    then Nuitka with the compile Nuitka and compares the
                    outputs. Default is True.
--skip-cpython26    The standard CPython2.6 test suite. Execute this for
                    all corner cases to be covered. With Python 2.7 this
                    covers exception behavior quite well. Default is True.
--skip-cpython27    The standard CPython2.7 test suite. Execute this for
                    all corner cases to be covered. With Python 2.6 these
                    are not run. Default is True.
--skip-cpython32    The standard CPython3.2 test suite. Execute this for all
                    corner cases to be covered. With Python 2.x these are not
                    run. Default is True.
```

You will only run the CPython test suites, if you have the submodules of the Nuitka git repository checked out. Otherwise, these will be skipped automatically with a warning that they are not available.

The policy is generally, that `./misc/check-release` running and passing all tests shall be considered sufficient for a release.

Basic Tests

You can run the "basic" tests like this:

```
./tests/basics/run_all.py search
```

These tests normally give sufficient coverage to assume that a change is correct, if these tests pass. To control the Python version used for testing, you can set the `PYTHON` environment variable to e.g. "python3.2", or execute the "run_all.py" with the intended version, it is portable across all supported Python versions.

Syntax Tests

Then there are "syntax" tests, i.e. language constructs that need to give a syntax error.

It sometimes happens that Nuitka must do this itself, because the `ast.parse` don't see the problem. Using `global` on a function argument is an example of this. These tests make sure that the errors of Nuitka and CPython are totally the same for this:

```
./tests/syntax/run_all.py search
```

Program Tests

Then there are small programs tests, that exercise all kinds of import tricks and problems with inter-module behavior. These can be run like this:

```
./tests/programs/run_all.py search
```

Compile Nuitka with Nuitka

And there is the "compile itself" or "reflected" test. This test makes Nuitka compile itself and compare the resulting C++, which helps to find in-determinism. The test compiles every module of Nuitka into an extension module and all of Nuitka into a single binary.

That test case also gives good coverage of the `import` mechanisms, because Nuitka uses a lot of packages.

```
./tests/reflected/compile_itself.py
```

Design Descriptions

These should be a lot more and contain graphics from presentations given. It will be filled in, but not now.

Nuitka Logo

The logo was submitted by "dr. Equivalent". It's source is contained in `misc/Logo` where 3 variants of the logo in SVG are placed.

- Symbol only (symbol)

- Text next to symbol (horizontal)

- Text beneath symbol (vertical)

From these logos, PNG images, "favicons", and a BMP file for the Windows installer are derived.

The exact ImageMagick commands are in `misc/make-doc.py`, but are now executed each time, the commands are also replicated here:

```
convert -background none misc/Logo/Nuitka-Logo-Symbol.svg images/Nuitka-Logo-Symbol.png
convert -background none misc/Logo/Nuitka-Logo-Vertical.svg images/Nuitka-Logo-Vertical.
convert -background none misc/Logo/Nuitka-Logo-Horizontal.svg images/Nuitka-Logo-Horizon

optipng -o2 images/Nuitka-Logo-Symbol.png
optipng -o2 images/Nuitka-Logo-Vertical.png
optipng -o2 images/Nuitka-Logo-Horizontal.png

convert -background grey -resize 152x261 misc/Logo/Nuitka-Logo-Vertical.svg -alpha backg
```

Choice of the Target Language

- Choosing the target language, is an important decision

  - The portability of Nuitka is decided here
- Other factors:

  - How difficult is it to generate the code?

  - Does the Python C-API have bindings?

  - Is that language known?

  - Does the language aid to find bugs?
- These candidates were considered

  - C++03, C++11, Ada

Requirement to Language matrix:

Requirement\Language	C++03	C++11	Ada
Portable	Yes	No [1]	Yes
Knowledge	Yes	No [2]	Yes
Python C-API	Yes	Yes	No [3]
Runtime checks	No	No	Yes [4]
Code Generation	Hard	Easy	Harder

1:: C++11 is not fully supported from any compiler (temporary problem)

2:: Not a whole lot of people have C++11 knowledge. My *only* C++11 code was that in Nuitka.

3:: The Python C-API for Ada would have to be created by us, possible just big project by itself.

4:: Runtime checks exist only for Ada in that quality. I miss automatic `CONSTRAINT_ERROR` exceptions, for data structures with validity indicators, where in other languages, I need to check myself.

The *decision for C++03* is ultimately:

- for portability

- for language knowledge

All of these are important advantages.

For C++11 initially spoke easy code generation:

- variadic templates

- raw strings

Yet, as it turns out, variadic templates do not help at all with evaluation order, so that code that used it, needed to be changed to generating instances of their code. And raw strings turned out to be not as perfect as one wants to be, and solving the problem with C++03 is feasible too, even if not pretty.

For Ada would have spoken the time savings through run time checks, which would have shortened some debugging sessions quite some. But building the Python C-API bindings on our own, and potentially incorrectly, would have eaten that up.

Use of Scons internally

Nuitka does not make its users interface with Scons at all, it's purely used internally. Nuitka itself, being pure Python, will run without any build process just fine.

For interfacing, there is the module `nuitka.build.SconsInterface` that will support calling scons - potentially from an inline copy on Windows or when using source releases - and passing arguments to it. These arguments are passed as `key=value`, and decoded in the scons file of Nuitka.

The scons file is named `SingleExe.scons` for lack of better name. It's really wrong now, but we have yet to find a better name. It once expressed the intention to be used to create executables, but the same works for modules too.

It supports operation in multiple modes, and modules is just one of them. It runs outside of Nuitka process scope, even with a different Python version potentially, so all the information must be passed on the command line.

What follows is the (lengthy) list of arguments that the scons file processes:

- `source_dir`

  Where is the generated C++ source code. Scons will just compile everything it finds there. No list of files is passed.

- `nuitka_src`

  Where do the include files and static C++ parts of Nuitka live. These provide e.g. the implementation of compiled function, generators, and other helper codes, this will point to where `nuitka.build` package lives normally.

- `result_base`

  This is not a full name, merely the basename for the result to be produced, but with path included, and the suffix comes from module or executable mode.

- `module_mode`

  Build a module instead of a program.

- `debug_mode`

  Enable debug mode, which is a mode, where Nuitka tries to help identify errors in itself, and will generate less optimal code. This also asks for warnings, and makes the build fail if there are any.

- `python_debug`

  Compile and link against Python debug mode, which does assertions and extra checks, to identify errors, mostly related to reference counting. May make the build fail, if no debug build library of CPython is available. On Windows it typically is not installed.

- `optimize_mode`

  Optimization mode, enable as much as currently possible. This refers to building the binary.

- `full_compat_mode`

  Full compatibility, even where it's stupid, i.e. do not provide information, even if available, in order to assert maximum compatibility. Intended to control level of compatability to absurd.

- `experimental_mode`

  Do things that are not yet accepted to be safe.

- `lto_mode`

  Make use of link time optimization of g++ compiler if available and known good with the compiler in question. So far, this was not found to make major differences.

- `win_disable_console`

  Windows subsystem mode: Disable console for windows builds.

- `unstriped_mode`

  Unstriped mode: Do not remove debug symbols.

- `clang_mode`

  Clang compiler mode, default on MacOS X and FreeBSD, optional on Linux.

- `mingw_mode`

  MinGW compiler mode, optional and interesting to Windows only.

- `standalone_mode`

  Building a standalone distribution for the binary.

- `show_scons`

  Show scons mode, output information about Scons operation. This will e.g. also output the actual compiler used, output from compilation process, and generally debug information relating to be build process.

- `python_prefix`

  Home of Python to be compiled against, used to locate headers and libraries.

- `target_arch`

  Target architecture to build.

- `icon_path`

  The icon to use for Windows programs if given.

Locating Modules and Packages

The search for of modules used is driven by `nuitka.Importing` module.

- From the module documentation

  The actual import of a module may already execute code that changes things. Imagine a module that does `os.system()`, it will be done. People often connect to databases, and these kind of things, at import time. Not a good style, but it's being done.

  Therefore CPython exhibits the interfaces in an `imp` module in standard library, which one can use those to know ahead of time, what file import would load. For us unfortunately there is nothing in CPython that is easily accessible and gives us this functionality for packages and search paths exactly like CPython does, so we implement here a multi step search process that is compatible.

  This approach is much safer of course and there is no loss. To determine if it's from the standard library, one can abuse the attribute `__file__` of the `os` module like it's done in `isStandardLibraryPath` of this module.

- Role

  This module serves the recursion into modules and analysis if a module is a known one. It will give warnings for modules attempted to be located, but not found. These warnings are controlled by a while list inside the module.

Hooking for module *`import`* process

Currently, in created code, for every `import` variable a normal `__import__()` call is executed. The "ModuleUnfreezer.cpp" (located in "nuitka/build/static_src") provides the implementation of a `sys.meta_path` hook.

This one allows us to have the Nuitka provided module imported even when imported by non-compiled code.

> ### *Note*
>
> Of course it would make sense to compile time detect which module it is that is being imported and then to make it directly. At this time, we don't have this inter-module optimization yet, mid-term it should become easy to add.

Supporting __*class*__ of Python3

In Python3 the handling of `__class__` and `super` is different from Python2. It used to be a normal variable, and now the following things have changed.

- The use of the `super` variable name triggers the addition of a closure variable `__class__`, as can be witnessed by the following code:

```python
class X:
    def f1(self):
        print( locals() )

    def f2(self):
        print( locals() )
        super

x = X()
x.f1()
x.f2()
```

```
{'self': <__main__.X object at 0x7f1773762390>}
{'self': <__main__.X object at 0x7f1773762390>, '__class__': <class '__main__.X'>}
```

- This value of `__class__` is also available in the child functions.

- The parser marks up code objects usage of "super". It doesn't have to be a call, it can also be a local variable. If the `super` builtin is assigned to another name and that is used without arguments, it won't work unless `__class__` is taken as a closure variable.

- As can be seen in the CPython3.2 code, the closure value is added after the class creation is performed.

- It appears, that only functions locally defined to the class are affected and take the closure.

This left Nuitka with the strange problem, of how to emulate that.

The solution is this:

- Under Python3, usage of `__class__` as a reference in a function body that is not a class dictionary creation, marks it up via `markAsClassClosureTaker`.

- Functions that are marked up, will be forced to reference variable to `__class__`.

> ### *Note*
>
> This one should be optimized away later if not used. Currently we have "no unused closure variable" detection, but it would cover it.

- When recognizing calls to `super` without arguments, make the arguments into variable reference to `__class__` and potentially `self` (actually first argument name).

- Class dictionary definitions are added.

  These are special direct function calls, ready to propagate also "bases" and "metaclass" values, which need to be calculated outside.

  The function bodies used for classes will automatically store `__class__` as a shared local variable, if anything uses it. And if it's not assigned by user code, it doesn't show up in the "locals()" used for dictionary creation.

  Existing `__class__` local variable values are in fact provided as closure, and overridden with the built class , but they should be used for the closure giving, before the class is finished.

  So `__class__` will be local variable of the class body, until the class is built, then it will be the `__class__` itself.

Frame Stack

In Python, every function, class, and module has a frame. It creates created when the scope it entered, and there is a stack of these at run time, which becomes visible in tracebacks in case of exceptions.

The choice of Nuitka is to make this non-static elements of the node tree, that are as such subject to optimization. In cases, where they are not needed, they may be removed.

Consider the following code.

```python
def f():
    if someNotRaisingCall():
        return somePotentiallyRaisingCall()
    else:
        return None
```

In this example, the frame is not needed for all the code, because the condition checked wouldn't possibly raise at all. The idea is the make the frame guard explicit and then to move it downwards in the tree, whenever possible.

So we start out with code like this one:

```python
def f():
    with frame_guard( "f" ):
        if someNotRaisingCall():
            return somePotentiallyRaisingCall()
        else:
            return None
```

This is to be optimized into:

```
def f():
    if someNotRaisingCall():
        with frame_guard( "f" ):
            return somePotentiallyRaisingCall()
    else:
        return None
```

Notice how the frame guard taking is limited and may be avoided, or in best cases, it might be removed completely. Also this will play a role when in-lining function, it will not be lost or need any extra care.

Parameter Parsing

The parsing of parameters is very convoluted in Python, and doing it in an compatible way is not that easy. This is a description of the required process, for easier overview.

Input

The input is an argument tuple (type is fixed), which contains the positional arguments, and potentially an argument dictionary (type is fixed, but could also be NULL, indicating no keyword arguments.

Keyword dictionary

The keyword argument dictionary is checked first. Anything in there, that cannot be associated, either raises an error, or is added to a potentially given star dict argument. So there are two major cases.

- No star dict argument: Iterate over dictionary, and assign or raise errors.

   This check covers extra arguments given.

- With star dict argument: Iterate over dictionary, and assign or raise errors.

   Interesting case for optimization are no positional arguments, then no check is needed, and the keyword argument dictionary could be used as the star argument. Should it change, a copy is needed though.

What's noteworthy here, is that in comparison of the keywords, we can hope that they are the same value as we use. The interning of strings increases chances for non-compiled code to do that, esp. for short names.

We then can do a simple == comparison and only fall back to real string comparisons, after all of these failed. That means more code, but also a lot faster code in the positive case.

Argument tuple

After this completed, the argument tuple is up for processing. The first thing it needs to do is to check if it's too many of them, and then to complain.

For arguments in Python2, there is the possibility of them being nested, in which case they cannot be provided in the keyword dictionary, and merely should get picked from the argument tuple.

Otherwise, the length of the argument tuple should be checked against its position and if possible, values should be taken from there. If it's already set (from the keyword dictionary), raise an error instead.

Language Conversions to make things simpler

There are some cases, where the Python language has things that can in fact be expressed in a simpler or more general way, and where we choose to do that at either tree building or optimization time.

The `assert` statement

The `assert` statement is a special statement in Python, allowed by the syntax. It has two forms, with and without a second argument. The later is probably less known, as is the fact that raise statements can have multiple arguments too.

The handling in Nuitka is:

```python
assert value
# Absolutely the same as:
if not value:
    raise AssertionError
```

```python
assert value, raise_arg
# Absolutely the same as:
if not value:
    raise AssertionError, raise_arg
```

This makes assertions absolutely the same as a raise exception in a conditional statement.

This transformation is performed at tree building already, so Nuitka never knows about `assert` as an element and standard optimizations apply. If e.g. the truth value of the assertion can be predicted, the conditional statement will have the branch statically executed or removed.

The "comparison chain" expressions

```python
a < b > c < d
# With "temp variables" and "assignment expressions", absolutely
# the same as:
a < ( tmp_b = b ) and tmp_b > ( tmp_c = c ) and ( tmp_c < d )
```

This transformation is performed at tree building already. The temporary variables keep the value for the potential read in the same expression. The syntax is not Python, and only pseudo language to expression the internal structure of the node tree after the transformation.

This useful "keeper" variables that enable this transformation and allow to express the short circuit nature of comparison chains by using `and` operations.

The `execfile` built-in

Handling is:

```python
execfile( filename )
# Basically the same as:
exec( compile( open( filename ).read() ), filename, "exec" )
```

Note

This allows optimizations to discover the file opening nature easily and apply file embedding or whatever we will have there one day.

This transformation is performed when the `execfile` builtin is detected as such during optimization.

Generator expressions with `yield`

These are converted at tree building time into a generator function body that yields the iterator given, which is the put into a for loop to iterate, created a lambda function of and then called with the first iterator.

That eliminates the generator expression for this case. It's a bizarre construct and with this trick needs no special code generation.

Function Decorators

When one learns about decorators, you see that:

```python
@decorator
def function():
    pass
# Is basically the same as:
def function():
    pass
function = decorator( function )
```

The only difference is the assignment to function. In the `@decorator` case, if the decorator fails with an exception, the name `function` is not assigned.

Therefore in Nuitka this assignment is from a "function body expression" and only the last decorator returned value is assigned to the function name.

This removes the need for optimization and code generation to support decorators at all. And it should make the two variants optimize equally well.

In-place Assignments

In-place assignments are re-formulated to an expression using temporary variables.

These are not as much a reformulation of `+=` to `+`, but instead one which makes it explicit that the assign target may change its value.

```python
a += b
```

```python
_tmp = a.__iadd__( b )

if a is not _tmp:
    a = _tmp
```

Using `__iadd__` here to express that not the `+`, but the in-place variant `iadd` is used instead. The `is` check may be optimized away depending on type and value knowledge later on.

Complex Assignments

Complex assignments are defined as those with multiple targets to assign from a single source and are re-formulated to such using a temporary variable and multiple simple assignments instead.

```python
a = b = c
```

```
_tmp = c
b = _tmp
a = _tmp
del _tmp
```

This is possible, because in Python, if one assignment fails, it can just be interrupted, so in fact, they are sequential, and all that is required is to not calculate `c` twice, which the temporary variable takes care of.

Unpacking Assignments

Unpacking assignments are re-formulated to use temporary variables as well.

```
a, b.attr, c[ind] = d = e, f, g = h()
```

Becomes this:

```
_tmp = h()

_iter1 = iter( _tmp )
_tmp1 = unpack( _iter1, 3 )
_tmp2 = unpack( _iter1, 3 )
_tmp3 = unpack( _iter1, 3 )
unpack_check( _iter1 )
a = _tmp1
b.attr = _tmp2
c[ind] = _tmp3
d = _tmp
_iter2 = iter( _tmp )
_tmp4 = unpack( _iter2, 3 )
_tmp5 = unpack( _iter2, 3 )
_tmp6 = unpack( _iter2, 3 )
unpack_check( _iter1 )
e = _tmp4
f = _tmp5
g = _tmp6
```

That way, the unpacking is decomposed into multiple simple statementy. It will be the job of optimizations to try and remove unnecessary unpacking, in case e.g. the source is a known tuple or list creation.

Note

The `unpack` is a special node which is a form of `next` that will raise a `ValueError` when it cannot get the next value, rather than a `StopIteration`. The message text contains the number of values to unpack, therefore the integer argument.

> ### *Note*
>
> The `unpack_check` is a special node that raises a `ValueError` exception if the iterator is not finished, i.e. there are more values to unpack.

With Statements

The `with` statements are re-formulated to use temporary variables as well. The taking and calling of `__enter__` and `__exit__` with arguments, is presented with standard operations instead. The promise to call `__exit__` is fulfilled by `try`/`except` clause instead.

```python
with some_context as x:
    something( x )
```

```python
tmp_source = some_context

# Actually it needs to be "special look-up" for Python2.7, so attribute
# look-up won't be exactly what is there.
tmp_exit = tmp_source.__exit__

# This one must be held for the whole with statement, it may be assigned
# or not, in our example it is. If an exception occurs when calling
# ``__enter__``, the ``__exit__`` should not be called.
tmp_enter_result = tmp_source.__enter__()

# Indicator variable to know if "tmp_exit" has been called.
tmp_indicator = False

try:
    # Now the assignment is to be done, if there is any name for the
    # manager given, this may become multiple assignment statements and
    # even unpacking ones.
    x = tmp_enter_result

    # Then the code of the "with" block.
    something( x )
except Exception:
    # Note: This part of the code must not set line numbers, which we
    # indicate with special source code references, which we call "internal".
    # Otherwise the line of the frame would get corrupted.

    tmp_indicator = True

    if not tmp_exit( *sys.exc_info() ):
        raise
finally:
    if not tmp_indicator:
        # Call the exit if no exception occurred with all arguments
        # as "None".
        tmp_exit( None, None, None )
```

> ### *Note*
>
> We don't refer really to `sys.exc_info()` at all, instead, we have references to the current exception type, value and trace, taken directory from the caught exception object on the C++ level.
>
> If we had the ability to optimize `sys.exc_info()` to do that, we could use the same transformation, but right now we don't have it.

For Loops

The for loops use normal assignments and handle the iterator that is implicit in the code explicitly.

```python
for x,y in iterable:
    if something( x ):
        break
else:
    otherwise()
```

This is roughly equivalent to the following code:

```python
_iter = iter( iterable )
_no_break_indicator = False

while True:
    try:
        _tmp_value = next( _iter )
    except StopIteration:
        # Set the indicator that the else branch may be executed.
        _no_break_indicator = True

        # Optimization should be able to tell that the else branch is run
        # only once.
        break

    # Normal assignment re-formulation applies to this assignment of course.
    x, y = _tmp_value
    del _tmp_value

    if something( x ):
        break

if _no_break_indicator:
    otherwise()
```

> ### *Note*
>
> The `_iter` temporary variable is of course in a temp block and the `x, y` assignment is the normal is of course re-formulation of an assignment that cannot fail.
>
> The `try/except` is detected to allow to use a variant of `next` that throws no C++ exception, but instead to use `ITERATOR_NEXT` and which returns NULL in that case, so that the code doesn't really have any Python level exception handling going on.

While Loops

Loops in Nuitka have no condition attached anymore, so while loops are re-formulated like this:

```python
while condition:
    something()
```

```python
while True:
    if not condition:
        break

    something()
```

This is to totally remove the specialization of loops, with the condition moved to the loop body in a conditional statement, which contains a break statement.

That makes it clear, that only break statements exit the loop, and allow for optimization to remove always true loop conditions, without concerning code generation about it, and to detect such a situation, consider e.g. endless loops.

> ### *Note*
>
> Loop analysis can therefore work on a reduced problem (which `break` statements are executed under which conditions) and be very general, but it cannot take advantage of the knowledge encoded directly anymore. The fact that the loop body may not be entered at all, if the condition is not met, is something harder to discover.

Exception Handler Values

Exception handlers in Python may assign the caught exception value to a variable in the handler definition.

```python
try:
    something()
except Exception as e:
    handle_it()
```

That is equivalent to the following:

```python
try:
    something()
except Exception:
    e = sys.exc_info()[1]
    handle_it()
```

Of course, the value of the current exception, use special references for assignments, that access the C++ and don't go via `sys.exc_info` at all, these are called `CaughtExceptionValueRef`.

Statement `try/except` with `else`

Much like `else` branches of loops, an indicator variable is used to indicate the entry into any of the exception handlers.

Therefore, the `else` becomes a real conditional statement in the node tree, checking the indicator variable and guarding the execution of the `else` branch.

Class Creation (Python2)

Classes in Python2 have a body that only serves to build the class dictionary and is a normal function otherwise. This is expressed with the following re-formulation:

```python
# in module "SomeModule"
# ...

class SomeClass(SomeBase, AnotherBase)
    """ This is the class documentation. """

    some_member = 3
```

```python
def _makeSomeClass:
    # The module name becomes a normal local variable too.
    __module__ = "SomeModule"

    # The doc string becomes a normal local variable.
    __doc__ = """ This is the class documentation. """

    some_member = 3

    return locals()

    # force locals to be a writable dictionary, will be optimized away, but
    # that property will stick. This is only to express, that locals(), where
    # used will be writable to.
    exec ""

SomeClass = make_class("SomeClass", (SomeBase, AnotherBase), _makeSomeClass())
```

That is roughly the same, except that `_makeSomeClass` is *not* visible to its child functions when it comes to closure taking, which we cannot express in Python language at all.

Therefore, class bodies are just special function bodies that create a dictionary for use in class creation. They don't really appear after the tree building stage anymore. The type inference will of course have to become able to understand `make_class` quite well, so it can recognize the created class again.

Class Creation (Python3)

In Python3, classes are a complicated way to write a function call, that can interact with its body. The body starts with a dictionary provided by the metaclass, so that is different, because it can __prepare__ a non-empty locals for it, which is hidden away in "prepare_class_dict" below.

What's noteworthy, is that this dictionary, could e.g. be a OrderDict. I am not sure, what __prepare__ is allowed to return.

```python
# in module "SomeModule"
# ...

class SomeClass(SomeBase, AnotherBase, metaclass = SomeMetaClass)
    """ This is the class documentation. """

    some_member = 3
```

```python
# Non-keyword arguments, need to be evaluated first.
tmp_bases = ( SomeBase, AnotherBase )

# Keyword arguments go next, __metaclass__ is just one of them. In principle
# we need to forward the others as well, but this is ignored for the sake of
# brevity.
tmp_metaclass = select_metaclass(tmp_bases, SomeMetaClass )

tmp_prepared = tmp_metaclass.__prepare__("SomeClass", tmp_bases)

# The function that creates the class dictionary. Receives temporary variables
# to work with.
def _makeSomeClass:
    # This has effect, currently I don't know how to force that in Python3
    # syntax, but we will use something that ensures it.
    locals() = tmp_prepared

    # The module name becomes a normal local variable too.
    __module__ = "SomeModule"

    # The doc string becomes a normal local variable.
    __doc__ = """ This is the class documentation. """

    some_member = 3

    # Create the class, share the potential closure variable "__class__"
    # with others.
    __class__ = tmp_metaclass("SomeClass", tmp_bases, locals())

    return __class__

# Build and assign the class.
SomeClass = _makeSomeClass()
```

Generator Expressions

There are re-formulated as functions.

Generally they are turned into calls of function bodies with (potentially nested) for loops:

```python
gen = ( x*2 for x in range(8) if cond() )
```

```python
def _gen_helper(__iterator):
    for x in __iterator:
        if cond():
            yield x*2

gen = _gen_helper( range(8 ) )
```

List Contractions

The list contractions of Python2 are different from those of Python3, in that they don't actually do any closure variable taking, and that no function object ever exists.

```python
list_value = [ x*2 for x in range(8) if cond() ]
```

```python
def _listcontr_helper(__iterator):
    result = []

    for x in __iterator:
        if cond():
            result.append( x*2 )

    return result

list_value = listcontr_helper( range(8) )
```

The difference is that with Python3, the function "_listcontr_helper" is real and named `<listcomp>`, whereas with Python2 the function must be considered in-lined.

This in-inlining in case of Python2 causes difficulties, because it's statements that occur inside an expression, which means a lot of side effects, that may or may not be possible to unroll to outside.

Set Contractions

TODO.

Dict Contractions

TODO.

Boolean expressions `and` and `or`

The short circuit operators `or` and `and` tend to be only less general that the `if/else` expressions and are therefore re-formulated as such:

```python
expr1() or expr2()
```

```python
_tmp if ( _tmp = expr1() ) else expr2()
```

```
expr1() and expr2()
```

```
expr2() if ( _tmp = expr1() ) else _tmp
```

In this form, the differences between these two operators becomes very apparent, the operands are simply switching sides.

With this the branch that the "short-circuit" expresses, becomes obvious, at the expense of having the assignment expression to the temporary variable, that one needs to create anyway.

Simple Calls

As seen below, even complex calls are simple calls. In simple calls of Python there is still some hidden semantic going on, that we expose.

```
func( arg1, arg2, named1 = arg3, named2 = arg4 )
```

On the C-API level there is a tuple and dictionary built. This one is exposed:

```
func( *( arg1, arg2 ), **{ "named1" : arg3, "named2" : arg4 } )
```

A called function will access this tuple and the dictionary to parse the arguments, once that is also re-formulated (argument parsing), it can then lead to simple inlining. This way calls only have 2 arguments with constant semantics, that fits perfectly with the C-API where it is the same, so it is actually easier for code generation.

Although the above looks like a complex call, it actually is not. No checks are needed for the types of the star arguments and it's directly translated to `PyObject_Call`.

Complex Calls

The call operator in Python allows to provide arguments in 4 forms.

- Positional (or normal) arguments
- Named (or keyword) arguments
- Star list arguments
- Star dictionary arguments

The evaluation order is precisely that. An example would be:

```
something( pos1, pos2, name1 = named1, name2 = named2, *star_list, **star_dict )
```

The task here is that first all the arguments are evaluated, left to right, and then they are merged into only two, that is positional and named arguments only. for this, the star list argument and the star dict arguments, are merged with the positional and named arguments.

What's peculiar, is that if both the star list and dict arguments are present, the merging is first done for star dict, and only after that for the star list argument. This makes a difference, because in case of an error, the star argument raises first.

```
something( *1, **2 )
```

This raises "TypeError: something() argument after ** must be a mapping, not int" as opposed to a possibly more expected "TypeError: something() argument after * must be a sequence, not int."

That doesn't matter much though, because the value is to be evaluated first anyway, and the check is only performed afterwards. If the star list argument calculation gives an error, this one is raised before checking the star dict argument.

So, what we do, is we convert complex calls by the way of special functions, which handle the dirty work for us. The optimization is then tasked to do the difficult stuff. Our example becomes this:

```python
def _complex_call(called, pos, kw, star_list_arg, star_dict_arg):
    # Raises errors in case of duplicate arguments or tmp_star_dict not
    # being a mapping.
    tmp_merged_dict = merge_star_dict_arguments( called, tmp_named, mapping_check( calle

    # Raises an error if tmp_star_list is not a sequence.
    tmp_pos_merged = merge_pos_arguments( called, tmp_pos, tmp_star_list )

    # On the C-API level, this is what it looks like.
    return called( *tmp_pos_merged, **tmp_merged_dict )

returned = _complex_call(
    called        = something,
    pos           = (pos1, pos2),
    named         = {
        "name1" : named1,
        "name2" = named2
    },
    star_list_arg = star_list,
    star_list_arg = star_dict
)
```

The call to `_complex_call` is be a direct function call with no parameter parsing overhead. And the call in its end, is a special call operation, which relates to the "PyObject_Call" C-API.

Print statements

The `print` statement exists only in Python2. It implicitly coverts its arguments to strings before printing them. In order to make this accessible and compile time optimized, this is made visible in the node tree.

```python
print arg1, "1", 1
```

```python
print str(arg1), "1", str(1)
```

Only string objects are spared from the `str` built-in wrapper, because that would only cause noise in optimization stage.

Nodes that serve special purposes

Side Effects

When an exception is bound to occur, and this can be determined at compile time, Nuitka will not generate the code the leads to the exception, but directly just raise it. But not in all cases, this is the full thing.

Consider this code:

```
f( a(), 1 / 0 )
```

The second argument will create a `ZeroDivisionError` exception, but before that `a()` must be executed, but the call to `f` will never happen and no code is needed for that, but the name look-up must still succeed. This then leads to code that is internally like this:

```
f( a(), raise ZeroDivisionError )
```

which is then modeled as:

```
side_effect( a(), f, raise ZeroDivisionError )
```

where you can consider side_effect a function that returns the last expression. Of course, if this is not part of another expression, but close to statement level, side effects, can be converted to multiple statements simply.

Another use case, is that the value of an expression can be predicted, but that the language still requires things to happen, consider this:

```
a = len( ( f(), g() ) )
```

We can tell that `a` will be 2, but the call to `f` and `g` must still be performed, so it becomes:

```
a = side_effects( f(), g(), 2 )
```

Modelling side effects explicitly has the advantage of recognizing them easily and allowing to drop the call to the tuple building and checking its length, only to release it.

Caught Exception Type/Value References

When catching an exception, in C++, an exception object is used. Exception handler code is being re-formulated to assign the caught exception to a name, to check its type for values, etc.

For these, not `sys.exc_info()` is used, instead there are special nodes dedicated to these values: `CaughtExceptionTypeRef` and `CaughtExceptionValueRef`.

Plan to replace "python-qt" for the GUI

Porting the tree inspector available with `--dump-gui` to "wxWindows" is very much welcome as the "python-qt4" bindings are severely under documented.

Plan to add "ctypes" support

Add interfacing to C code, so Nuitka can turn a `ctypes` binding into an efficient binding as if it were written manually with Python C-API or better.

Goals/Allowances to the task

1. Goal: Must not use any pre-existing C/C++ language file headers, only generate declarations in generated C++ code ourselves. We would rather write a C header to `ctypes` declarations convert if it needs to be, but not mix and use declarations from existing header code.

2. Allowance: May use `ctypes` module at compile time to ask things about `ctypes` and its types.

3. Goal: Should make use of `ctypes`, to e.g. not hard code what `ctypes.c_int()` gives on the current platform, unless there is a specific benefit.

4. Allowance: Not all `ctypes` usages must be supported immediately.

5. Goal: Try and be as general as possible. For the compiler, `ctypes` support should be hidden behind a generic interface of some sort. Supporting `math` module should be the same thing.

Type Inference - The Discussion

Main goal is to forward value knowledge. When you have `a = b`, that means that a and b now "alias". And if you know the value of `b` you can assume to know the value of `a`. This is called "Aliasing".

When assigning `a` to something new, that won't change `b` at all. But when an attribute is set, a method called of it, that impacts both, or actually the value. We need to understand mutable vs. immutable though.

```
a = 3
b = 3

b += 4 # a is not changed

a = [ 3 ]
b = [ 3 ]

b += [ 4 ] # a is changed
```

If we cannot tell, we must assume that `a` might be changed. It's either `b` or what `a` was before. If the type is not mutable, we can assume the aliasing to be broken up, and if it is, we can assume both to be the same value still.

When that value is a compile time constant, we will want to push it forward, because storing such a constant under a variable name has a cost and loading it back from the variable as well. So, you want to be able collapse such code:

```
a = 3
b = 7
c = a / b
```

to:

```
c = 3 / 7
```

and that obviously to:

```
c = 0
```

This may be called "(Constant) Value Propagation". But we are aiming for even more. We want to forward propagate abstract properties of the values.

> ### *Note*
>
> Built-in exceptions, and built-in names are also compile time constants.

In order to fully benefit from type knowledge, the new type system must be able to be fully friends with existing built-in types. The behavior of a type `long`, `str`, etc. ought to be implemented as far as possible with the builtin `long`, `str` as well.

> ### *Note*
>
> This "use the real thing" concept extends beyond builtin types, e.g. `ctypes.c_int()` should also be used, but we must be aware of platform dependencies. The maximum size of `ctypes.c_int` values would be an example of that. Of course that may not be possible for everything.
>
> This approach has well proven itself with built-in functions already, where we use real built-ins where possible to make computations. We have the problem though that built-ins may have problems to execute everything with reasonable compile time cost.

Another example, consider the following code:

```
len( "a" * 1000000000000 )
```

To predict this code, calculating it at compile time using constant operations, while feasible, puts an unacceptable burden on the compilation.

Esp. we wouldn't want to produce such a huge constant and stream it, the C++ code would become too huge. So, we need to stop the `*` operator from being used at compile time and cope with reduced knowledge, already here:

```
"a" * 10000000000000
```

Instead, we would probably say that for this expression:

- The result is a `str` or `PyStringObject`.
- We know its length exactly, it's `10000000000000`.
- Can predict every of its elements when sub-scripted, sliced, etc., if need be, with a function we may create.

Similar is true for this horrible thing:

```
range( 10000000000000 )
```

So it's a rather general problem, this time we know:

- The result is a `list` or `PyListObject`
- We know its length exactly, `10000000000000`
- Can predict every of its elements when index, sliced, etc., if need be, with a function.

Again, we wouldn't want to create the list. Therefore Nuitka avoids executing these calculation, when they result in constants larger than a threshold of e.g. 256. This concept has to be also applied to integers and more CPU and memory traps.

Now lets look at a more common use case:

```python
for x in range( 10000000000000 ):
    doSomething()
```

Looking at this example, one traditional way to look at it, would be to turn `range` into `xrange`, and to note that `x` is unused. That would already perform better. But really better is to notice that `range()` generated values are not used at all, but only the length of the expression matters.

And even if `x` were used, only the ability to predict the value from a function would be interesting, so we would use that computation function instead of having an iteration source. Being able to predict from a function could mean to have Python code to do it, as well as C++ code to do it. Then code for the loop can be generated without any CPython library usage at all.

> ### *Note*
>
> Of course, it would only make sense where such calculations are "O(1)" complexity, i.e. do not require recursion like "n!" does.

The other thing is that CPython appears to at - run time - take length hints from objects for some operations, and there it would help too, to track length of objects, and provide it, to outside code.

Back to the original example:

```python
len( "a" * 1000000000000 )
```

The theme here, is that when we can't compute all intermediate expressions, and we sure can't do it in the general case. But we can still, predict some of properties of an expression result, more or less.

Here we have `len` to look at an argument that we know the size of. Great. We need to ask if there are any side effects, and if there are, we need to maintain them of course. This is already done by existing optimization if an operation generates an exception.

> ### *Note*
>
> The optimization of `len` has been implemented and works for all kinds of container creation and ranges.

Applying this to "ctypes"

The not so specific problem to be solved to understand `ctypes` declarations is maybe as follows:

```python
import ctypes
```

This leads to Nuitka in its tree to have an assignment from a `__import__` expression to the variable `ctypes`. It can be predicted by default to be a module object, and even better, it can be known as `ctypes` from standard library with more or less certainty. See the section about "Importing".

So that part is "easy", and it's what will happen. During optimization, when the module `__import__` expression is examined, it should say:

- `ctypes` is a module
- `ctypes` is from standard library (if it is, may not be true)
- `ctypes` has a `ModuleFriend` that knows things about it attributes, that should be asked.

The later is the generic interface, and the optimization should connect the two, of course via package and module full names. It will need a `ModuleFriendRegistry`, from which it can be pulled. It would be nice if we can avoid `ctypes` to be loaded into Nuitka unless necessary, so these need to be more like a plug-in, loaded only if necessary, i.e. the user code actually uses `ctypes`.

Coming back to the original expression, it also contains an assignment expression, because it is more like this:

```
ctypes = __import__( "ctypes" )
```

The assigned to object, simply gets the type inferred propagated as part of an SSA form. Ideally, we could be sure that nothing in the program changes the variable, and therefore have only one version of that variable.

For module variables, when the execution leaves the module to unknown code, or unclear code, it might change the variable. Therefore, likely we will often only assume that it could still be ctypes, or something else.

Depending on how well we control module variable assignment, we can decide this more of less quickly. With "compiled modules" types, the expectation is that it's merely a quick C++ == comparison check. The module friend should offer code to allow a check if it applies, for uncertain cases.

Then when we come to uses of it:

```
ctypes.c_int()
```

At this point, using SSA, we are more of less sure, that `ctypes` is at that point the module, and that we know what it's `c_int` attribute is, at comile time, and what it's call result is. We will use the module friend to help with that. It will attach knowledge about the result of that expression during the SSA collection process.

This is more like a value forward propagation than anything else. In fact, constant propagation should only be the special case of it, and one design goal of Nuitka was always to cover these two cases with the same code.

Excursion to Functions

In order to decide what this means to functions and their call boundaries, if we propagate forward, how to handle this:

```
def my_append(a, b):
    a.append( b )

    return a
```

We would notate that `a` is first a "unknown but defined parameter object", then later on something that definitely has an `append` attribute, when returned. Otherwise an exception occurs.

The type of `a` changes to that after `a.append` look-up succeeds. It might be many kinds of an object, but e.g. it could have a higher probability of being a `PyListObject`. And we would know it cannot be a `PyStringObject`, as that one has no "append".

> ### *Note*
>
> If classes, i.e. other types in the program, have an `append` attribute, it should play a role too, there needs to be a way to plug-in to this decisions.

> ### *Note*
>
> On the other hand, types without `append` attribute could be eliminated.

It would be great, if functions provided some sort of analysis on their return type, or a quick way to predict return value properties, based on input value knowledge.

So this could work:

```python
b = my_append( [], 3 )

assert b == [3] # Could be decided now
```

Goal: The structure we use makes it easy to tell what `my_append` may be. So, there should be a means to ask it about call results with given type/value information. We need to be able to tell, if evaluating `my_append` makes sense with given parameters or not, if it does impact the return value.

We should e.g. be able to make `my_append` tell, one or more of these:

- Returns the first parameter value as return value (unless it raises an exception).
- The return value has the same type as `a` (unless it raises an exception).
- The return value has an `append` attribute.
- The return value might be a `list` object.
- The return value may not be a `str` object.
- The function will raise if first argument has no `append` attribute.

The exactness of statements may vary. But some things may be more interesting. If e.g. the aliasing of a parameter value to the return value is known exactly, then information about it need to all be given up, but some can survive.

It would be nice, if `my_append` had sufficient information, so we could specialize with `list` and `int` from the parameters, and then e.g. know at least some things that it does in that case. Such specialization would have to be decided if it makes sense. In the alternative, it could be done for each variant anyway, as there won't be that many of them.

Doing this "forward" analysis appears to be best suited for functions and therefore long term. We will try it that way.

Excursion to Loops

```python
a = 1

while 1:    # think loop: here
    b = a + 1
    a = b

    if cond():
        break

print a
```

The handling of loops (both "for" and "while" are re-formulated to loops with breaks) has its own problem. The loop start and may have an assumption from before it started, that "a" is constant, but that is only true for the first iteration. So, we can't pass knowledge from outside loop forward directly into the for loop body.

So the collection for loops needs to be two pass. First, to collect assignments, and merge these into the start state, before entering the loop body. The need to make two passes is special to loops.

For a start, it could be done like this though: At loop entry, all knowledge is removed about everything, and so is at loop exit. That way, only the loop inner working is optimized, and before and after the loop are separate things. The optimal handling of "a" in the example code will take a while.

For a general solution, it would be sweet to trace different exit paths differently. One loop exit may be good enough, as it will be the common case.

Excursion to Conditions

```python
if cond:
    x = 1
else:
    x = 2

b = x < 3
```

The above code contains a condition, and these have the problem, that when exiting the conditional block, a merge must be done, of the "x" versions. It could be either one. The merge may trace the condition under which a choice is taken. That way, we could decide pairs of traces under the same condition.

These merges of SSA variable versions, represent alternatives. They pose difficulties, and might have to be reduced to commonality. In the above example, the "<" operator will have to check for each version, and then to decide that both indeed give the same result.

The constraint collection tracks variable changes in conditional branches, and then merges the existing state at conditional statement exits.

> ### *Note*
>
> A branch is considered "exiting" if it is not abortive. Should it end in a `raise`, `break`, `continue`, or `return`, there is no need to merge that branch, as execution of that branch is terminated.

> Should both branches be abortive, that makes things really simple, as there is no need to even continue.
>
> Should only one branch exist, but be abortive, then no merge is needed, and the collection can assume after the conditional statement, that the branch was not taken, and continue.

When exiting both the branches, these branches must both be merged, with their new information.

In the above case:

- The "yes" branch knows variable `x` is an `int` of constant value `1`
- The "no" branch knows variable `x` is an `int` of constant value `2`

That might be collapsed to:

- The variable `x` is an integer of value in `(1,2)`

Given this, we then should be able to precompute the value of this:

```
b = x < 3
```

The comparison operator can therefore decide and tell:

- The variable `b` is a boolean of constant value `True`.

Were it unable to decide, it would still be able to say:

- The variable `b` is a boolean.

For conditional statements optimization, it's also noteworthy, that the condition is known to pass or not pass the truth check, inside branches, and in the case of non-exiting single branches, after the statement it's not true.

We may want to take advantage of it. Consider e.g.

```python
if type( a ) is list:
    a.append( x )
else:
    a += ( x, )
```

In this case, the knowledge that `a` is a list, could be used to generate better code and with the definite knowledge that `a` is of type list. With that knowledge the `append` attribute call will become the `list` built-in type operation.

Excursion to *return* statements

The `return` statement (like `break`, `continue`, `raise`) is "aborting" to control flow. It is always the last statement of inspected block. Were there statements to follow it, optimization will remove it as dead code.

If all branches of a conditional statement are "aborting", the statement is decided "aborting" too. If a loop doesn't break, it should be considered "aborting" too.

> ### *Note*
>
> The removal of statements following "aborting" statements is implemented, and so is the discovery of abortive conditional statements. It's not yet done for loops, temp blocks, etc. though.

So, `return` statements are easy for local optimization. In the general picture, it would be sweet to collect all return statements, and analyze the commonality of them. The goal to predict function results, might be solvable by looking at their traces.

Excursion to *`yield`* expressions

The `yield` expression can be treated like a normal function call, and as such invalidates some known constraints just as much as they do. It executes outside code for an unknown amount of time, and then returns, with little about the outside world known anymore.

Mixed Types

Consider the following inside a function or module:

```python
if cond is not None:
    a = [ x for x in something() if cond(x) ]
else:
    a = ()
```

A programmer will often not make a difference between `list` and `tuple`. In fact, using a `tuple` is a good way to express that something won't be changed later, as these are mutable.

> ### *Note*
>
> Better programming style, would be to use this:
>
> ```python
> if cond is not None:
>     a = tuple( x for x in something() if cond(x) )
> else:
>     a = ()
> ```
>
> People don't do it, because they dislike the performance hit encountered by the generator expression being used to initialize the tuple. But it would be more consistent, and so Nuitka is using it, and of course one day Nuitka ought to be able to make no difference in performance for it.

To Nuitka though this means, that if `cond` is not predictable, after the conditional statement we may either have a `tuple` or a `list` type object in `a`. In order to represent that without resorting to "I know nothing about it", we need a kind of `min/max` operating mechanism that is capable of say what is common with multiple alternative values.

> **Note**
>
> At this time, we don't really have that mechanism to find the commonality between values.

Back to "ctypes"

```
v = ctypes.c_int()
```

Coming back to this example, we needed to propagate `ctypes`, then we can propagate "something" from `ctypes.int` and then known what this gives with a call and no arguments, so the walk of the nodes, and diverse operations should be addressed by a module friend.

In case a module friend doesn't know what to do, it needs to say so by default. This should be enforced by a base class and give a warning or note.

Now to the interface

The following is the intended interface:

- Iteration with node methods `computeStatement` and `computeNode`.

  These traverse modules and functions (i.e. scopes) and visit everything in the order that Python executes it. The visiting object is `ConstraintCollection` and pass forward. Some node types, e.g. `StatementConditional` new create child constraint collections and handle the SSA merging at exit.

- Replacing nodes during the visit.

  Both `computeStatement` and `computeNode` are tasked to return potential replacements of themselves, together with "tags" (meaningless now), and a "message", used for verbose tracing.

  The replacement node of "+" operator, may e.g. the pre-computed result, wrapped in side effects of the node.

- Assignments and references affect SSA.

  The SSA tree is initialized every time a scope is visited. Then during traversal, traces are built up. Every assignment and merge starts a new trace for that matter. References to a given variable version are traced that way.

- Value escapes are traced too.

  When an operation hands over a value to outside code, it indicates so to the constraint collection. This is for it to know, when e.g. a constant value, might be mutated meanwhile.

- Nodes can be queried about their properties.

  The node base classes offers methods that allow to check if certain operations are supported or not. These can always return `True` (yes), `False` (no), and `None` (cannot decide). In the case of the later, optimizations may not be able do much about it. Lets call these values "tri-state".

  The default implementation will be very pessimistic. Specific node types may then declare, that they e.g. have no side effects, do no raise, have a know truth value, have a known iteration length, can predict their iteration values, etc.

- Nodes are linked to certain states.

During the collect, a variable reference, is linked to a certain trace state, and that can be used by parent operations.

```
a = 1
b = a + a
```

In this example, the references to "a", can look-up the "1" in the trace, and base their responses to "+" on it. It will ask "isCompileTimeConstant()" and both nodes will respond "True", then "getCompileTimeConstant()" will return "1", which will be computed. Then "extractSideEffects()" will return "()" and therefore, the result "2" will not be wrapped.

• Class for module import expression `ExpressionImportModule`.

This one just knows that something is imported, but not how or what it is assigned to. It will be able in a recursive compile, to provide the module as an assignment source, or the module variables or submodules as an attribute source when referenced from a variable trace or in an expression.

• Base class for module friend `ModuleFriendBase`.

This is intended to provide something to overload, which e.g. can handle `math` in a better way.

• Module `ModuleFriendRegistry`

Provides a register function with `name` and instances of `ValueFriendModuleBase` to be registered. Recursed to modules should integrate with that too. The registry could well be done with a metaclass approach.

• The module friends should each live in a module of their own.

With a naming policy to be determined. These modules should add themselves via above mechanism to `ModuleFriendRegistry` and all shall be imported and register. Importing of e.g. `ctypes` should be delayed to when the friend is actually used. A meta class should aid this task.

The delay will avoid unnecessary blot of the compiler at run time, if no such module is used. For "qt" and other complex stuff, this will be a must.

• The walk should initially be single pass, and not maintain history.

Instead optimization that needs to look at multiple things, e.g. "unused assignment", will look at the whole SSA collection afterwards.

Discussing with examples

The following examples:

```python
# Assignment, the source decides the type of the assigned expression
a = b

# Operator "attribute look-up", the looked up expression "ctypes" decides
# via its trace.
ctypes.c_int

# Call operator, the called expressions decides with help of arguments,
# which have been walked, before the call itself.
called_expression_of_any_complexity()

# import gives a module any case, and the "ModuleRegistry" may say more.
import ctypes
```

```
# From import need not give module, "x" decides what it is.
from x import y

# Operations are decided by arguments, and CPython operator rules between
# argument states.
a + b
```

The optimization is mostly performed by walking of the tree and performing constraint collection. When it encounters assignments and references to them, it considers current state of traces and uses it for `computeExpression`.

> **Note**
>
> Assignments to attributes, indexes, slices, etc. will also need to follow the flow of `append`, so it cannot escape attention that a list may be modified. Usages of `append` that we cannot be sure about, must be traced to exist, and disallow the list to be considered known value again.

Code Generation Impact

Right now, code generation assumes that everything is a `PyObject *`, i.e. a Python object, and does not take knowledge of `int` or other types into consideration at all, and it should remain like that for some time to come.

Instead, `ctypes` value friend will be asked give `Identifiers`, like other codes do too. And these need to be able to convert themselves to objects to work with the other things.

But Code Generation should no longer require that operations must be performed on that level. Imagine e.g. the following calls:

```
c_call( other_c_call() )
```

Value returned by "other_c_call()" of say `c_int` type, should be possible to be fed directly into another call. That should be easy by having a `asIntC()` in the identifier classes, which the `ctypes` Identifiers handle without conversions.

Code Generation should one day also become able to tell that all uses of a variable have only `c_int` value, and use `int` instead of `PyObjectLocalVariable` more or less directly. We could consider `PyIntLocalVariable` of similar complexity as `int` after the C++ compiler performed its in-lining.

Such decisions would be prepared by finalization, which then would track the history of values throughout a function or part of it.

Initial Implementation

The basic interface will be added to *all* expressions and a node may override it, potentially using constraint collection state, as attached during "computeExpression".

Goal 1

Initially most things will only be able to give up on about anything. And it will be little more than a tool to do simple look-ups in a general form. It will then be the first goal to turn the following code into better performing one:

```
a = 3
b = 7
c = a / b
return c
```

to:

```
a = 3
b = 7
c = 3 / 7
return c
```

and then:

```
a = 3
b = 7
c = 0
return c
```

and then:

```
a = 3
b = 7
c = 0
return 0
```

This depends on SSA form to be able to tell us the values of `a`, `b`, and `c` to be written to by constants, which can be forward propagated at no cost.

Goal 2

The assignments to `a`, `b`, and `c` shall all become prey to "unused" assignment analysis in the next step. They are all only assigned to, and the assignment source has no effect, so they can be simply dropped.

```
return 0
```

In the SSA form, these are then assignments without references. These assignments, can be removed if the assignment source has no side effect. Or at least they could be made "anonymous", i.e. use a temporary variable instead of the named one. That would have to take into account though, that the old version still needs a release.

The most general form would first merely remove assignments that have no impact, and leave the value as a side effect, so we arrive at this first:

```
3
7
0
return 0
```

When applying the removal of expression only statements without effect, this gives us:

```
return 0
```

which is the perfect result. Doing it in one step would only be an optimization.

In order to be able to manipulate nodes related to a variable trace, we need to attach the nodes that did it. Consider this:

```python
if cond():
    x = 1
elif other():
    x = 3

# Not using "x".
return 0
```

In the above case, the merge of the value friends, should say that `x` may be undefined, or one of `1` or `3`, but since `x` is not used, apply the "dead value" trick to each branch.

The removal of the "merge" of the 3 `x` versions, should exhibit that the other versions are also only assigned to, and can be removed. These merges of course appear as usages of the `x` versions.

Goal 3

Then third goal is to understand all of this:

```python
def f():
    a = []

    print a

    for i in range(1000):
        print a

        a.append( i )

    return len( a )
```

> ### Note
>
> There are many operations in this, and all of them should be properly handled, or at least ignored in safe way.

The first goal code gave us that the `list` has an annotation from the assignment of `[]` and that it will be copied to `a` until the for loop in encountered. Then it must be removed, because the `for` loop somehow says so.

The `a` may change its value, due to the unknown attribute look-up of it already, not even the call. The for loop must be able to say "may change value" due to that, of course also due to the call of that attribute too.

The code should therefore become equivalent to:

```python
def f():
    a = []
```

```
    print []

    for i in range(1000):
        print a

        a.append( i )

    return len( a )
```

But no other changes must occur, especially not to the "return" statement, it must not assume "a" to be constant "[]" but an unknown "a" instead.

With that, we would handle this code correctly and have some form constant value propagation in place, handle loops at least correctly, and while it is not much, it is important demonstration of the concept.

Goal 4

The fourth goal is to understand the following:

```
def f(cond):
    y = 3

    if cond:
        x = 1
    else:
        x = 2

return x < y
```

In this we have a branch, and we will be required to keep track of both the branches separately, and then to merge with the original knowledge. After the conditional statement we will know that "x" is an "int" with possible values in "(1,2)", which can be used to predict that the return value is always "True".

The forth goal will therefore be that the "ValueFriendConstantList" knows that append changes "a" value, but it remains a list, and that the size increases by one. It should provide an other value friend "ValueFriendList" for "a" due to that.

In order to do that, such code must be considered:

```
a = []

a.append( 1 )
a.append( 2 )

print len( a )
```

It will be good, if "len" still knows that "a" is a list, but not the constant list anymore.

From here, work should be done to demonstrate the correctness of it with the basic tests applied to discover undetected issues.

Fifth and optional goal: Extra bonus points for being able to track and predict "append" to update the constant list in a known way. Using "list.append" that should be done and lead to a constant result of "len" being used.

The sixth and challenging goal will be to make the code generation be impacted by the value friends types. It should have a knowledge that "PyList_Append" does the job of append and use "PyList_Size" for "len". The "ValueFriends" should aid the code generation too.

Last and right now optional goal will be to make "range" have a value friend, that can interact with iteration of the for loop, and "append" of the "list" value friend, so it knows it's possible to iterate 5000 times, and that "a" has then after the "loop" this size, so "len(a)" could be predicted. For during the loop, about a the range of its length should be known to be less than 5000. That would make the code of goal 2 completely analyzed at compile time.

Limitations for now

- Aim only for limited examples. For `ctypes` that means to compile time evaluate:

```python
print ctypes.c_int( 17 ) + ctypes.c_long( 19 )
```

  Later then call to "libc" or something else universally available, e.g. "strlen()" or "strcmp()" from full blown declarations of the callable.

- We won't have the ability to test that optimization are actually performed, we will check the generated code by hand.

  With time, we will add XML based checks with "xpath" queries, expressed as hints, but that is some work that will be based on this work here. The "hints" fits into the "ValueFriends" concept nicely or so the hope is.

- No inter-function optimization functions yet

  Of course, once in place, it will make the `ctypes` annotation even more usable. Using `ctypes` objects inside functions, while creating them on the module level, is therefore not immediately going to work.

- No loops yet

  Loops break value propagation. For the `ctypes` use case, this won't be much of a difficulty. Due to the strangeness of the task, it should be tackled later on at a higher priority.

- Not too much.

  Try and get simple things to work now. We shall see, what kinds of constraints really make the most sense. Understanding `list` subscript/slice values e.g. is not strictly useful for much code and should not block us.

> ### *Note*
>
> This design is not likely to be the final one.

Idea Bin

This an area where to drop random ideas on our minds, to later sort it out, and out it into action, which could be code changes, plan changes, issues created, etc.

- Make "SELECT_METACLASS" meta class selection transparent.

  Looking at the "SELECT_METACLASS" it should become an anonymous helper function. In that way, the optimization process can remove choices at compile time, and e.g. inline the effect of a meta class, if it is known.

  This of course makes most sense, if we have the optimizations in place that will allow this to actually happen.

- Keeping track of iterations

  The constraint collection trace should become the place, where variables or values track their use state. The iterator should keep track of the "next()" calls made to it, so it can tell which value to given in that case.

  That would solve the "iteration of constants" as a side effect and it would allow to tell that they can be removed.

  That would mean to go back in the tree and modify it long after.

  ```
  a = iter( ( 2, 3 ) )
  b = next( a )
  c = next( a )
  del a
  ```

  It would be sweet if we could recognize that:

  ```
  a = iter( ( 2, 3 ) )
  b = side_effect( next( a ), 2 )
  c = side_effect( next( a ), 3 )
  del a
  ```

  That trivially becomes:

  ```
  a = iter( ( 2, 3 ) )
  next( a )
  b = 2
  next( a )
  c = 3
  del a
  ```

  When the "del a" is examined at the end of scope, or due to another assignment to the same variable, ending the trace, we would have to consider of the "next" uses, and retrofit the information that they had no effect.

  ```
  a = iter( ( 2, 3 ) )
  b = 2
  b = 3
  del a
  ```

- Aliasing

  Each time an assignment is made, an alias is created. A value may have different names.

  ```
  a = iter( range(9 ))
  b = a
  c = next(b)
  d = next(a)
  ```

  If we fail to detect the aliasing nature, we will calculate "d" wrongly. We may incref and decref values to trace it.

  Aliasing is automatically traced already in SSA form. The "b" is assigned to version of "a". So, that should allow to replace it with this:

  ```
  a = iter( range(9 ))
  c = next(a)
  d = next(a)
  ```

  Which then will be properly handled.

- Shelve for caching

  If we ever came to the conclusion to want and cache complex results of analysis, we could do so with the shelve module. We would have to implement `__deepcopy__` and then could store in there optimized node structures from start values after parsing.

- Tail recursion optimization.

  Functions that return the results of calls, can be optimized. The Stackless Python does it already.

- Integrate with "upx" compression.

  Calling "upx" on the created binaries, would be easy.

- The timing of `__del__` calls.

  When you do a(b(c())) in Python, it deletes the argument value, i.e. return value of c() immediately after calling b().

  Currently we translate that to C++ roughly like this: a(b(c())) as well. Only that in C++, b returns an object, that has a scope. It appears, the d-tor is executed at the end of the statement. In C++ the ";" is a sequence point, i.e. things must be done by then.

  Unfortunately C++ loves temporaries so much, it won't immediately delete them after use, but only after full expression, which means ")" or ";", and attempts with fake sequence points all failed.

  But, there may be another way. Right now, `PyObject *` is the interface for about everything passed around. And "PyObjectTemporary" releases values that are needed by the interface to have a reference, and deleted afterwards.

  But it could, and should be different. All helper functions should be template functions that accept `PyObjectRef1` and `PyObjectRef0`, and know about the reference, and then manage `PyObjectRef1` instances to release their reference as soon as they are not needed. With `PyObjectRef0` that would be a no-op.

  This is a lot of work. The good news, is that it's work that will be needed, to support types other than `PyObject *` efficiently. Them being converted to `PyObject *` and releasing that reference, it would be transparent to all code.

- In-lining constant "exec" and "eval".

  It should be possible to re-formulate at least cases without "locals" or "globals" given.

```python
def f():
    a = 1
    b = 2

    exec( """a+=b;c=1""" )

    return a, c
```

Should become this here:

```python
def f():
    a = 1
    b = 2

    a+=b  #
    c=1    # MaybeLocalVariables for everything except known local ones.

    return a, c
```

If this holds up, inlining `exec` should be relatively easy.

- Original and overloaded built-ins

  This is about making things visible in the node tree. In Nuitka things that are not visible in the node tree tend to be wrong. We already pushed around information to the node tree a lot.

  Later versions, Nuitka will become able to determine it has to be the original built-in at compilt time, then a condition that checks will be optimized away, together with the slow path. Or the other path, if it won't be. Then it will be optimized away, or if doubt exists, it will be correct. That is the goal.

  Right now, the change would mean to effectively disable all built-in call optimization, which is why we don't immediately do it.

  Making the compatible version, will also require a full listing of all built-ins, which is typing work merely, but not needed now. And a way to stop built-in optimization from optimizing builtin calls that it used in a wrap. Probably just some flag to indicate it when it visits it to skip it. That's for later.

  But should we have that both, I figure, we could not raise a `RuntimeError` error, but just do the correct thing, in all cases. An earlier step may raise `RuntimeError` error, when built-in module values are written to, that we don't support.

- SSA form for Nuitka nodes

  - Assignments collect a counter from the variable, which becomes the variable version. This happens during tree building phase.

  - References need to back track to the last assignment on their path, which may be a merge. Constraint collection can do that.

  - Data structures

    Every constraint collection has these:

    - variable_actives

      Dictionary, where per "variable" the currently used version is. Used to track situations changes in branches. This is the main input for merge process.

    - variable_traces

      Dictionary, where "variable" and "version" form the key. The values are objects with or without an assignment, and a list of usages, which starts out empty.

      These objects have usages appended to them. In "onVariableSet", a new version is allocated, which gives a new object for the dictionary, with an empty usages list, because each write starts a new version. In "onVariableUsage" the version is detected from the current version. It may be not set yet, which means, it's a read of an undefined value (local variable, not a parameter name), or unknown in case of global variable.

      These objects may be told that their value has escaped. This should influence the value friend they attached to the initial assignment. Each usage may have a current value friend state that is different.

  - When merging branches of conditional statements, the merge shall apply as follows.

    - Branches have their own collection, with deviating sets of "variable_actives". These are children of an outer collections

    - Case a) One branch only.

      For that branch a collection is performed. As usual new assignments generate a new version making it "active", references then related to these "active" versions.

      Then, when the branch is merged, for all "active" variables, it is considered, if that is a change related to before the branch. If it's not the same, a merge trace with the branch condition is created with the one active in the collection before that statement.

    - Case b) Two branches.

      When there are two branches, they both as are treated as above, except for the merge.

      When merging, a difference in active variables between the two branches creates the merge trace.

      ### *Note*

      For conditional expressions, there are always only two branches. Even if you think you have more than one branch, you do not. It's always nested branches, already when it comes out of the parser.

  - Trace structure

    - Initial write of the version

There may be a initial write for each version. It can only occur at the start of it, but not later, and there is only one. The "value friend" of it.

- Merge of other one or two other versions

One could be empty, i.e. the variable would not be assigned. This is kind of the initial write, and the merge references one or multiple "value friends", which are optional.

- Bunch of read usages. They may allow escape of the value or not. When they do, it's a change. The value friend must be informed of it. If it's a real escape, usage is not known. If it's merely an alias, e.g. the value is now in another variable trace, they could be linked. Otherwise the "value friend" must be demoted immediately to one that gives more vague information.

This should be reflected in a class "VariableTrace".

- Recursion checks are expensive.

If the "caller" or the "called" can declare that it cannot be called by itself, we could leave it out.

TODO: Are they really that expensive? Unnecessary yes, but expensive may not be true.

- References

Currently Nuitka has "Variable" objects. Every variable reference node type refers to a "VariableReference" node and there are multiple of them. Every variable traces the reference objects created.

The idea of references started out with closure references and has expanded from there. It's now used to decide that a variable is shared. You can ask it, and because it knows its references, it can tell.

The thing is, this is not updated, so should a closure variable reference go away, it's still shared, as the reference remains. The thing with replaced and removed nodes, is that currently they do not remove themselves, there is no `__del__` being called. I consider this too unreliable.

That makes the detection of "shared" unreliable and with false positives, that so far do not harm much. There is an issue with Python3 not compiling with debug mode that might be a cause of it.

Anyway, the problem is increased by the scope of code in use in each optimization pass is only ever increasing, but starts out small. That a variable is shared or merely used elsewhere, might be discovered late. By starting from scratch again, over and over, we might discover this only later.

That may mean, we should do trace based optimization only after it's all complete, and not before. During the collection, information about the sharing should be reset at the start, and the built up and judged at the end.

The task to maintain this would be near ModuleRegistry.

- Outline functions

The list contractions of Python2, and potentially other contractions or in-lined functions too, in case they don't need any closure from it, could be considered part of the surrounding function.

These would have function bodies, with proper return, and generate code as a function would, but with the closure and local variables shared from arguments in what is considered a direct call.

The outline functions would not be considered closure takers, nor closure givers. They should be visited when they are used, almost like a statement sequences, and returns would define their value.

Updates for this Manual

This document is written in REST. That is an ASCII format which is readable as ASCII, but used to generate PDF or HTML documents.

You will find the current source under: http://nuitka.net/gitweb/?p=Nuitka.git;a=blob_plain;f=Developer_Manual.rst

And the current PDF under: http://nuitka.net/doc/Developer_Manual.pdf