OMNeT++ Goes Python

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Introduction

- After finishing OMNeT++ 6.0...
- We started playing around with Python because we saw huge potentials there
 - enthused by the Python-powered Analysis Tool
 - during INET development and maintenance, a Python REPL with a home-grown lib has grown to be an indispensable tool
 - discovered the cppyy C++ FFI Python package

(Snapshot from the office)



Areas where Python can be useful

- Setting up models (e.g. topology building)
- Extending the capabilities of NED expressions
- Implementing simulation-specific ad-hoc components (e.g. scenario managers, custom traffic generators, etc.)
- Simulation control (e.g. custom stop conditions)
- Managing and running simulations and simulation campaigns (workflow automation)
- Result analysis (inside and outside of the IDE)
- Disadvantages of Python in OMNeT++
 - speed (or lack thereof)
 - you want to keep Python code out of the "hot" parts of the model
 - the word "module" has become ambiguous
 - do you mean "importable Python file", or "OMNeT++ simulation component"? ;-)

Upcoming Python-related tools

- Python in NED expressions
 - pyeval() and pycode() NED functions
- Modules implemented in Python
 - @pythonClass NED property
 - modules can be written from scratch, or subclassed from existing (e.g. INET) modules
- Simulation library as a Python package
 - from omnetpp.runtime import *
 - based on cppyy
- Python package for processing simulation results
 - from omnetpp.scave import results, ...
- Python package for managing and running simulations
 - from omnetpp.simulation import *



The pyeval(), pycode() NED functions

- pyeval(<expr>,...) evaluates a Python expression string
 - pyeval("2*3")
 pyeval("x: 2*x", 3) --> 6
- pycode(<block>,...) evaluates a Python statement block
 - block must end in a "return" statement (like a function body)
 - pycode("import math; return math.factorial(15)")

More details about pyeval()/pycode(), @pythonClass and cppyy in Attila Török's Summit presentation: *Using Python within Simulations*

The @pythonClass NED property

Denotes that module is implemented by a Python class

```
// sink.ned
simple Sink {
    parameters:
        @pythonClass;
        ...
}

# sink.py
from omnetpp.runtime import *

class Sink(omnetpp.cSimpleModule):
    def handleMessage(self, msg):
        ...
```

The cppyy package

- Dynamic runtime Python bindings for C++
 - allows cross-inheritance and callbacks, template instantiation and more
 - simple example:

```
import cppyy
cppyy.include("iostream")
cppyy.cppdef("""class A { public: void sayHello() { std::cout << "Hello" << std::endl; } };""")
A = cppyy.gbl.A
a = A()
a.sayHello() # prints "Hello"</pre>
```

- Foundations:
 - Cling, the interactive C++ interpreter from CERN
 - Cling itself builds on Clang and LLVM
 - (at least this is what they want you to believe, but actually it relies on magic)



The omnetpp.runtime package

- Based on cppyy, it exposes the simulation library as a Python package
 - both simulation kernel and the "envir" infrastructure
 - its essence:

```
import cppyy
cppyy.include("omnetpp.h")
```

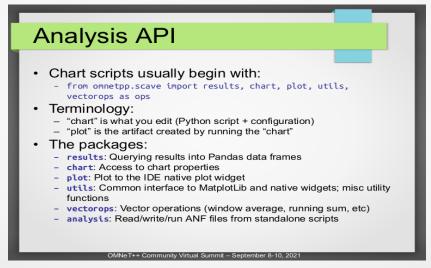
- OMNeT++ is undergoing extensive refactoring:
 - Reusability of "envir" part improved
 - Python readiness
 - Multi-thread support
 - Allow multiple threads to be used for simulation (but simulations are still <u>single-threaded!</u>)

```
./aloha -u Cmdenv --cmdenv-num-threads=8 ...
```

Global variables became thread_local or simulation-scope (csimulation::getSharedVariable<T>(name) - new)

The omnetpp.scave package

Part of OMNeT++ 6.0



Slide from last year's Summit

The omnetpp.simulation package

- Library and toolset for managing and running simulations and campaigns in various ways
 - in-process / local / distributed (e.g. ssh cluster using Dask)
 - for results, for regression testing (e.g. fingerprints), etc.
- Grown from the needs of INET development and maintenance
 - "run all simulations, utilizing all computing resources I have access to"
 - "refresh list of simulations to be fingerprint-tested"
 - "re-run failing fingerprint tests"
 - "re-run simulations that contain WiFi" (after WiFi model change)
 - "compare results to those with INET version X"
- Human time is expensive:
 - automate/assist as much as possible (high-level tools, REPL, etc)
 - store instead of recompute (fingerprints, simulation results, etc)
 - utilize available computing resources (multi-core, ssh cluster, etc)

DEMO



Recap of the Demo

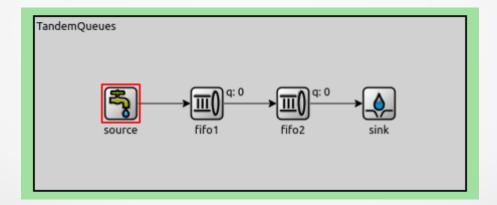
Python added to several OMNeT++ sample simulations:

- pyfifo: pure Python simple modules
- routing: various examples for making use of Python
- aloha: examples for running simulations from Python



samples/pyfifo

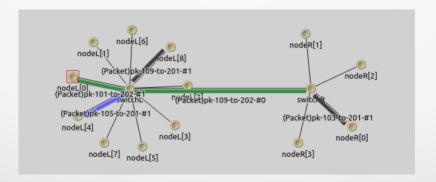
- A fairly verbatim re-implementation of the fifo example
- With all modules written in Python



samples/routing

Configurations in omnetpp.ini:

- [FromCsv]: Setting up a network specified in CSV, using Python
- [RandomTree]: Setting up a network generated with NetworkX Python package
- [DumbbellFaultyLink]: Shows a scenario manager written in Python
- [AppExt]: Extending a C++ simple module from Python
- [App2]: Parameter values produced using pyeval()/pycode()



samples/aloha

Python scripts:

- example1.py: instantiating a simulation
 simulation = omnetpp.cSimulation(...)
- example2.py: all simulations in an omnetpp.ini parameter study ini.getNumRunsInConfig("PureAlohaExperiment")
- example3.py: manually organizing a parameter study for numHosts in [10,15,20]: for iaMean in [1,2,3,4,5,7,9]:

example4.py: utilizing multiple CPU cores

```
with multiprocessing.Pool() as p:
    p.map(alohaJob, taskList)
```

- example5.py: simulation-based optimization
 scipy.optimize.minimize(...)
- example5a.py: preserve and also plot all simulations

