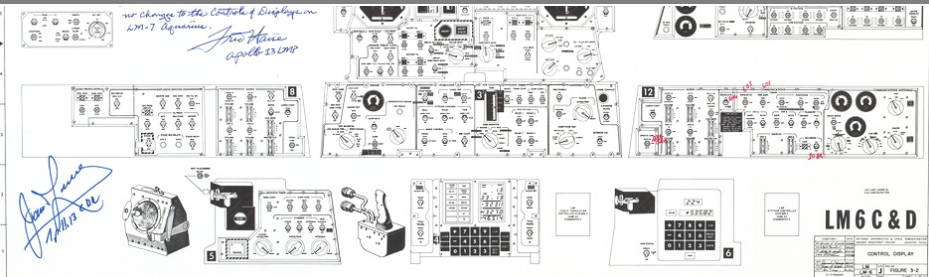


Experiment Control for High-Speed Tomography

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Motivation

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We have *strong* X-ray light sources , *fast* detectors, *distributed* device access via TANGO and *huge* processing capacity.

- Let's do awesome stuff with that!
- Process data and monitor changes on-line
- Build feedback-based control algorithms



Collaborative effort to

- Build hardware and software for high-speed tomography experiments
- Develop fast 2D detector and library for direct access
- Implement GPU-based data processing framework
 - Running on heterogenous compute systems
 - \approx 10 to 100 times *faster* reconstruction
- Do on-line reconstruction and data analysis
- Provide image-based feedback control

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That's what *Concert* will be for.

What it is about

- A Python framework for conducting high-speed experiments
- Local instead of a distributed system
- Standard procedures for common tomography tasks
- *Prototype* for high-speed tomography experiments at ANKA

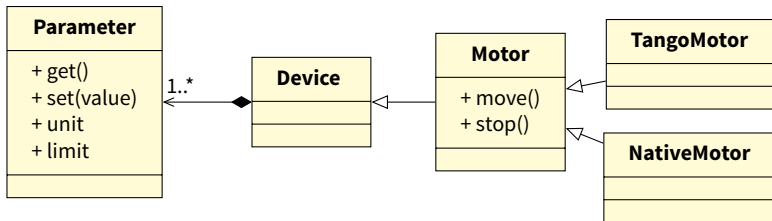
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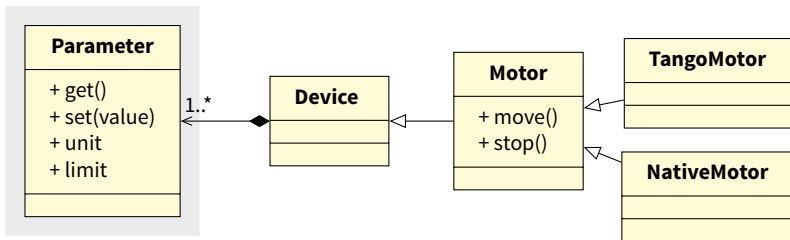
What it is not

- A general solution for all beamline problems
- Data archival system (e.g. meta data)
- Providing a GUI (Taurus?)

Partial Class Hierarchy Overview



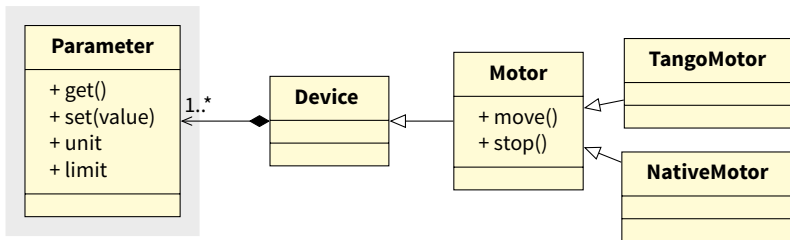
Partial Class Hierarchy Overview



A parameter

- Controls *one* aspect
- Has device specific getters & setters,
- optional SI units (via quantities),
- limits and descriptive doc string

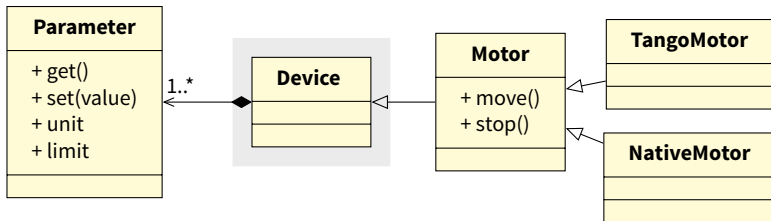
Partial Class Hierarchy Overview



Benefits:

- Validation of user input units
- Automatic access logging

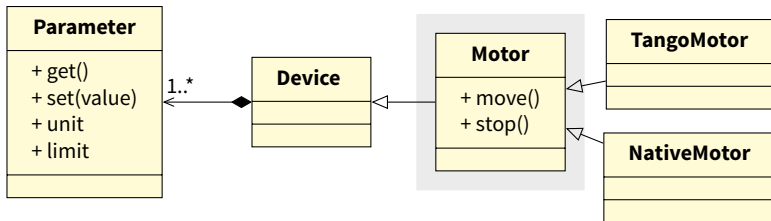
Partial Class Hierarchy Overview



A device consists of

- One or more parameters and
- Auxiliary methods

Partial Class Hierarchy Overview



Base device class provides

- Type-safe device distinction
- Common interface and methods

■ Enumerate parameters

```
motor = TangoMotor()
```

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for param in motor:  
    print(param)    # prints parameters value and unit
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■ Attribute access for setting/getting values

```
print(motor.position)  
x = motor.position
```

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■ Attribute access for setting/getting values

```
print(motor.position)  
x = motor.position
```

■ Invalid assignment fails gracefully with an exception

```
>>> motor.position = 1 * q.keV  
Sorry, 'position' can only receive values of unit 1 m (meter) but got 1.0 keV
```

- Careless synchronization can lead to excessive latencies

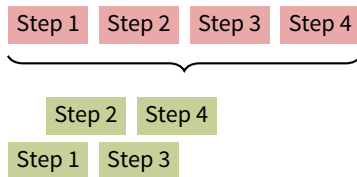
Step 1

Step 2

Step 3

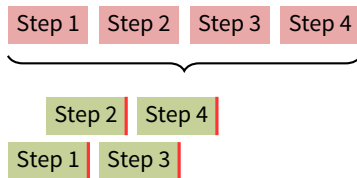
Step 4

- Careless synchronization can lead to excessive latencies
- Latencies are reduced by executing tasks in parallel



Asynchronous Device Access

- Careless synchronization can lead to excessive latencies
- Latencies are reduced by executing tasks in parallel
- We *must* be notified when a task is finished



Asynchronous execution

- Futures instead of raw threads
 - A future promises to return the result of a task at some point in the future
- Callbacks are called, no matter *when* they are attached
- Synchronization via device locks

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Monitoring and notification

- Messaging bus for process-wide notification
- Subscribers sign up for messages and are notified upon message arrival
- Light-weight monitoring mechanism

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m.position = 1.5 * q.mm           # Blocks until finished
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f2 = m.get_position()
f3 = m['position'].get()
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- Query futures and add callbacks

```
print("Done yet? {}".format(f1.done()))
f1.add_done_callback(do_something)
```

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- Query futures and add callbacks

```
print("Done yet? {}".format(f1.done()))
f1.add_done_callback(do_something)
```

- Wait for the result synchronously and do something with it

```
future = m.get_position()
result = future.result()
print(result)
```

- @async decorator turns any method into an asynchronous one

```
class Motor(Device):  
    @async  
    def move(self, delta):  
        self.position += delta
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- Usage is the same as for the parameter access:

```
m = MotorImpl()  
f = m.move(-5 * q.cm)  
print("Still running? {0}".format(f.running()))
```


- Single message *dispatcher* is used for subscription

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- Caller provides a callback handler ...

```
def alert(sender):  
    msg = "We ran into a limit, current position is {0}"  
    print(msg.format(sender.position))
```

- Single message *dispatcher* is used for subscription
- Caller provides a callback handler ...

```
def alert(sender):  
    msg = "We ran into a limit, current position is {0}"  
    print(msg.format(sender.position))
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- ...and subscribes on the bus

```
m = MotorImpl()  
dispatcher.subscribe(m, m.LIMIT, alert)
```

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Solution

- Provide *abstract* skeletons for recurring tasks
- Let the scientist compose complete processes

- Correlate scan parameter and feedback values
- Feedback can be of any complexity
- For example, a detector calibration procedure calculates the sensitivity over a range of exposure times

Simplified EMV1288 Sensitivity Procedure

```
detector = UcaCamera('pco')
shutter = Shutter()

def compute_parameters():
    shutter.close().wait()
    mean_dark = np.mean(detector.grab())
    shutter.open().wait()
    mean_bright = np.mean(detector.grab())
    return (mean_bright - mean_dark)

scanner = Scanner(detector['exposure-time'], compute_parameters)
scanner.minimum = 10 * q.microsecond
scanner.maximum = 1 * q.second

# Wait for the scan to complete and resolve the future
exp_times, sensitivity = scanner.run().result()
plt.plot(exp_times, sensitivity)
```

Simplified EMV1288 Sensitivity Procedure

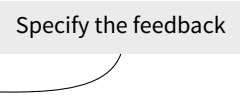
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Specify the feedback



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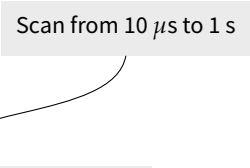
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Scan from 10 μ s to 1 s



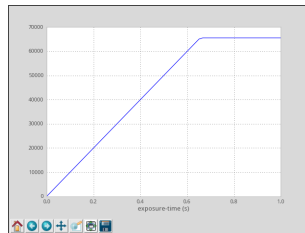
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- ...or scan the exposed parameter

```
scanner = Scanner(process['axis-pos'], do_something)
```

- Encapsulate experiment types into pre-defined sessions
- Combine sessions via import

```
import tomography

rot_motor.set_velocity(10 * q.deg / q.second)
shutter.open().wait()
pco_dimax.start_record()
...
```

- Starting a session launches an IPython shell (for now)

- Started to think about saving NeXus data sets
- Prototype stores tomographic scan data sets using Nexpy

```
def do_nothing():  
    pass  
  
tomo_scanner = StepTomoScanner(detector, rotary_stage)  
dataset = get_tomo_scan_result(tomo_scanner).result()  
dataset.nxsave('scan.hdf5')
```

- We are currently investigating DESY's pni-libraries as a backend

- Continuous integration with Jenkins
- 75 unit tests
- flake8 (pep8 + pyflakes) & pylint checks
- Sphinx documentation at `concert.readthedocs.org`
- Usable with pip and virtualenv

Summary

- We built an open prototype to integrate control and data processing
 - github.com/ufo-kit/concert
 - pypi.python.org/pypi/concert
- Interoperability with TANGO, UFO framework, NeXus, ...
- Parallel execution with defined synchronization points and messaging

Next steps

- Provide stable control loops based on python-control
- Use `IPython.traits` for unit-less parameters

Thanks for your attention! Questions?

Title image (“Control Display from Apollo 13”) courtesy of Steve Jurvetson under CC-BY 2.0.

Implementation Details

- Runs on Python 2.6+
- Data processing with the UFO framework
- General device access via Tango
- Detectors accessed with libuca
- quantities, logbook, PyTango and IPython

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- Devices implement the context manager protocol and keep a lock when used
- Multiple devices are locked with multicontext object or Python 2.7's enhanced with statement

```
# In-process safe device access
with motor, detector:
    motor.set_position(0.5 * q.mm)
    frame = detector.grab()
```


Requirements

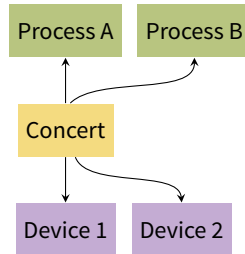
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Concert

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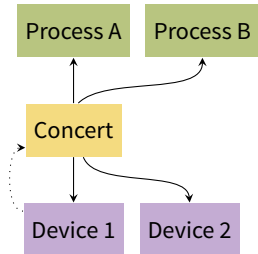
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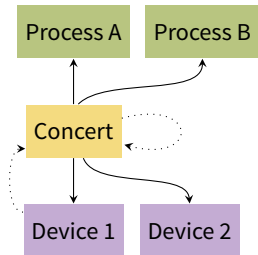
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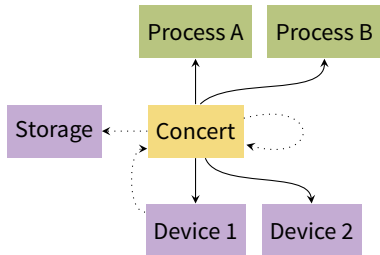
- Controls devices and processes under study
- Acquires data
- Reacts on data analysis results



Requirements

Such an approach requires a system that

- Controls devices and processes under study
- Acquires data
- Reacts on data analysis results
- Stores data



1. Focus on usage and favor
 - User before instrument
 - Scientist before developer
2. Local over distributed processing
3. Small, high quality core
4. Code re-use wherever, whenever possible