

PySPH: Smoothed Particles Hydrodynamics in Python

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SciPy.in 2010

Outline

- 1 Introduction
- 2 PySPH Architecture
- 3 Summary

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Smoothed Particle Hydrodynamics

SPH

- Mesh-free Lagrangian particle method
- Started with applications to astrophysics
- Fluid mechanics, free surface flows, fracture, porous media, explosions
- Games and videos animations
- Major advantage lies in simulation of complex problems with moving geometries

SPH Basics

Interpolating Integrals

$$f(\mathbf{r}) = \int f(\mathbf{r}') \delta(\mathbf{r} - \mathbf{r}') d\mathbf{r}'$$

Kernel Approximation

$$f(\mathbf{r}) \approx \int f(\mathbf{r}') w(\mathbf{r} - \mathbf{r}', h) d\mathbf{r}'$$

Particle Approximation

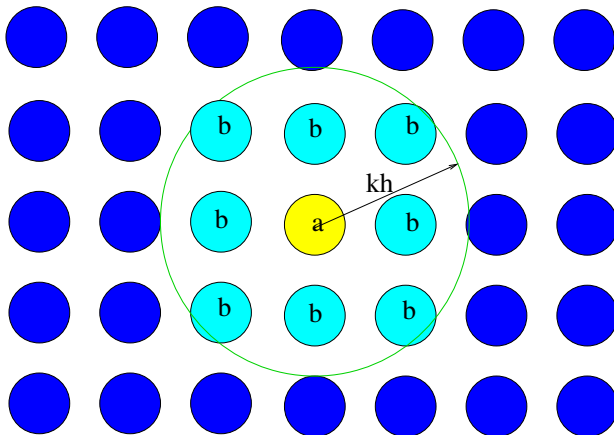
$$\langle f(\mathbf{r}_i) \rangle = \sum_j f(\mathbf{r}_j) \frac{m_j}{\rho_j} w(\mathbf{r}_i - \mathbf{r}_j, h_j)$$

SPH Basics

- Smoothed particles diffused in space by kernel function
- Material properties are weighted sum of particle properties
- Particle properties updated using Lagrangian form of governing equations
- Derivatives transferred to the kernel function
- PDE \rightarrow ODE

SPH Basics

SPH Approximation



PySPH

- Parallel extensible SPH framework written in Python/Cython
- <http://code.google.com/p/pysph>
- Open source (BSD licensed)

Why PySPH?

- Many SPH codes exist, including few open source
- Most in Fortran/C/C++
 - Learning and extension difficult, slow development, memory management complicated
- Needed tool for experimentation in SPH methods, trying out new methods and ideas quickly
- Choice of Python as programming language for PySPH

Why Python?

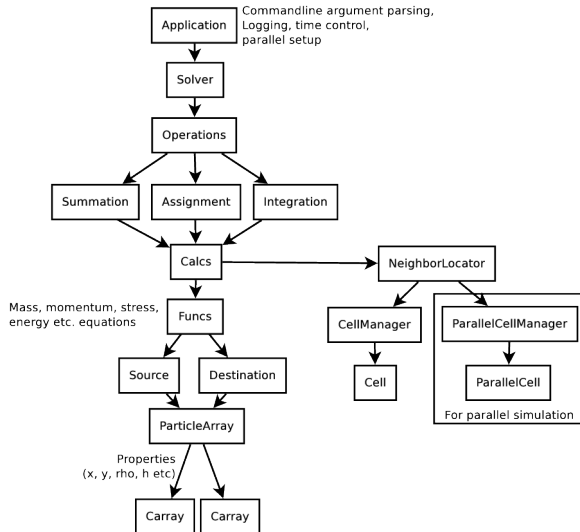
Python

- easier and faster to learn and code
- easier for new contributors to start contributing
- self-documenting (doc-strings)
- builtin high level data structures
- lots of available scientific/other libraries
- performance critical sections written in Cython (compiles python-like code to C extension)

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PySPH Architecture Overview



Particle Storage

- Particle properties (mass, x, y etc) stored in carrays (resizable typed c arrays like numpy 1d arrays)
- Different arrays stored in a ParticleArray (Python allows adding/removing arrays as named attributes)
- Different ParticleArrays for different entities
- All SPH operations on ParticleArrays

```
pa =  
pysph.base.get_particle_array(name='fluid',  
type=pysph.base.Fluid, x=x, m=m, ...)
```

Solver

Solver has all the operations to be performed in a simulation. Steps for creating a new solver are:

- Define relevant operations (subclasses of **SPHFunctionParticle**) in the *sph* module (e.g. Hooke's law in case of solid mechanics problems)
- Add relevant operations to instance of the *Solver* class (or a subclass to make the solver reusable) in appropriate order (e.g. add Hooke's law before the stress-momentum equation)
- Add relevant properties to the particles (e.g. stress σ_{ij}) while creating the particles for the problem to solve

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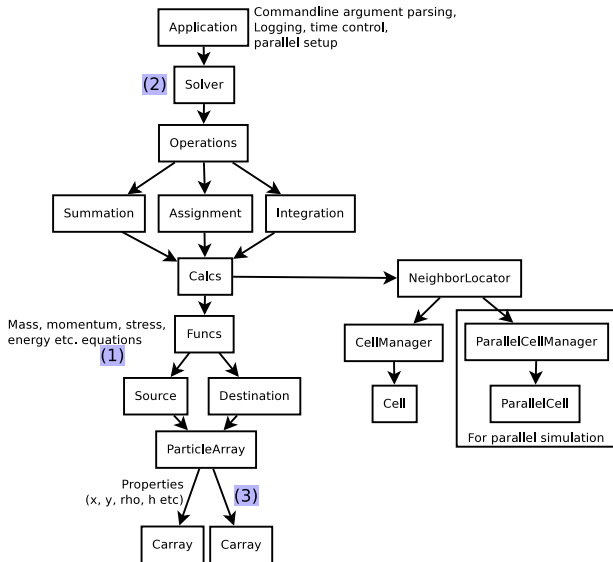
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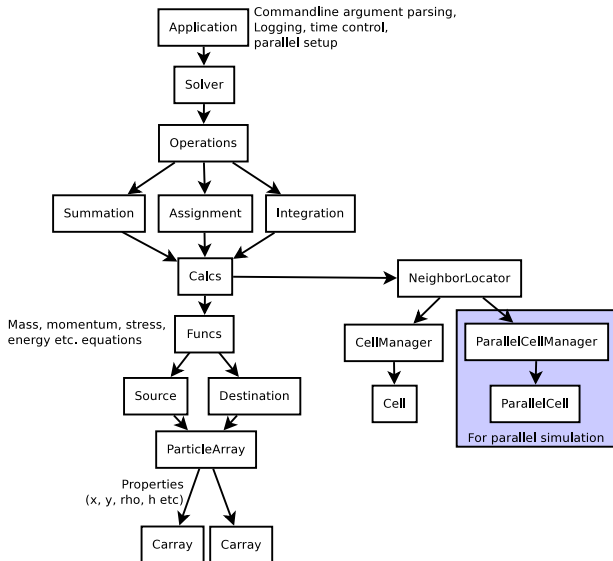
Extension



Parallelization

- Parallelization done using mpi4py bindings
- User requires no knowledge of parallelization
- Automatic load balancing among different processes
- *Application* class implements various switches to handle runs for parallel/serial cases
- command-line option parsing, load distribution, dumping output files

RunSnakeRun



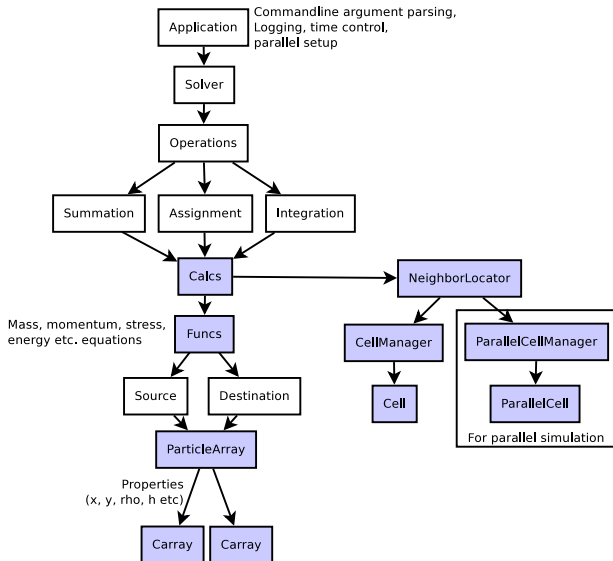
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Cython

- Statically typed C-like code in python-like language
- Compiled to fast native-code
- Mix-n-match C and Python data types
- Performance critical code written in Cython

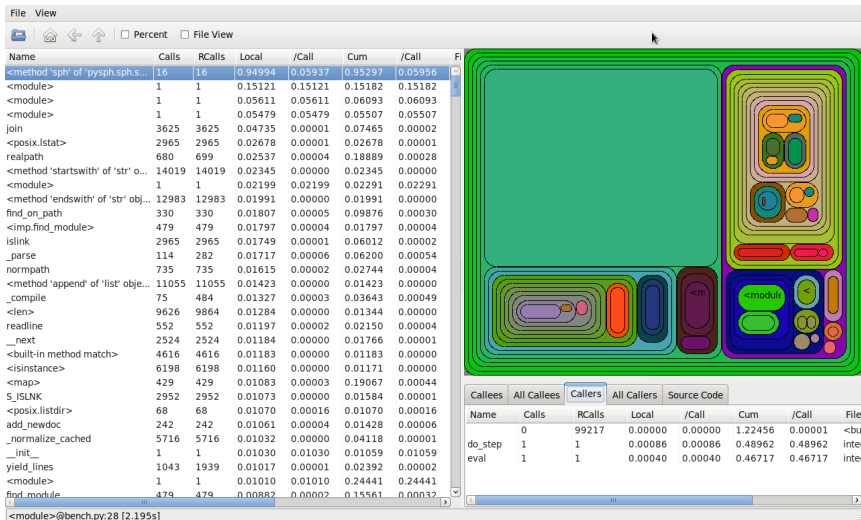
RunSnakeRun



Profiling

- Premature optimization root of all evil
- Never optimize w/o knowing what to optimize
- Works with cython if compiled using `profile=True` directive
- Visualize profiling data using `runsnakerun/KCacheGrind`
- `Valgrind/SystemTap(DTrace)` for tracing python/cython functions
- benchmark critical code sections to keep track of progress

RunSnakeRun



Testing

- Verify correct functioning, refactoring, regressions
- Python unittest module for writing tests
- Nose test collector and runner
- Ned Batchelder's coverage: custom monkey patch for cython function coverage

Coverage report: 45%

<i>Module</i>	<i>statements</i>	<i>missing</i>	<i>excluded</i>	<i>coverage</i>
/mnt/data/CourseWare/ddp/pysph/pyx_coverage	92	69	0	25%
main	20	2	0	90%
test	13	1	0	92%
test2	7	1	0	86%
Total	132	73	0	45%

coverage.py v3.5a1

Coverage for **main** : 90%

20 statements 18 run 2 missing 0 excluded

```

1 | import test
2 | import test2
3 |
4 | def uncovered():
5 |     print 'yo'
6 |     a = 1+2
7 |
8 |
9 | def covered1():
10 |
11 |     a = test.A()
12 |     a.func_b()
13 |     a.func_c()
14 |
15 | def covered2():
16 |     test.func_f()
17 |     test.func_e()
18 |
19 | def covered3():
20 |     test2.main()
21 |
22 | def main():
23 |     covered1()
24 |     covered2()
25 |     covered3()
26 |
27 | if __name__ == '__main__':
28 |     main()
29 |

```

« index coverage.py v3.5a1

Coverage for **test** : 92%

13 statements 12 run 1 missing 0 excluded

```

1 | #cython:profile=True
2 | cdef class A:
3 |     cdef public int a
4 |     cdef public str s
5 |     def __cinit__(self):
6 |         print 'A.__cinit__'
7 |
8 |     cdef str func_a(self, int a=1):
9 |         print 'A.func_a'
10 |         self.a += a
11 |         return self.s
12 |
13 |     cpdef str func_b(self, int a=2):
14 |         print 'A.func_b'
15 |         self.a += a
16 |         return self.s
17 |
18 |     def func_c(self, a=3):
19 |         print 'A.func_c'
20 |         self.a += a
21 |         return self.s, self.a
22 |
23 |     cdef uncovered_a(self):
24 |         print 'A.uncovered_a'
25 |
26 | cdef class B(object):
27 |     cdef public int a
28 |     cdef public str s
29 |     def __cinit__(self):
30 |         print 'B.__cinit__'
31 |
32 |     cdef str func_a(self, int a=1):
33 |         print 'B.func_a'

```

Python for scientific computing

- Fast prototyping, interactive interpreter
- Simplicity, less code, easier contribution
- Many scientific libraries available
- Full-fledged programming language, no restrictions on future ideas
- Great plotting and visualization packages available

Thank You !!!