

# PySPH: Smoothed Particles Hydrodynamics in Python

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# 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

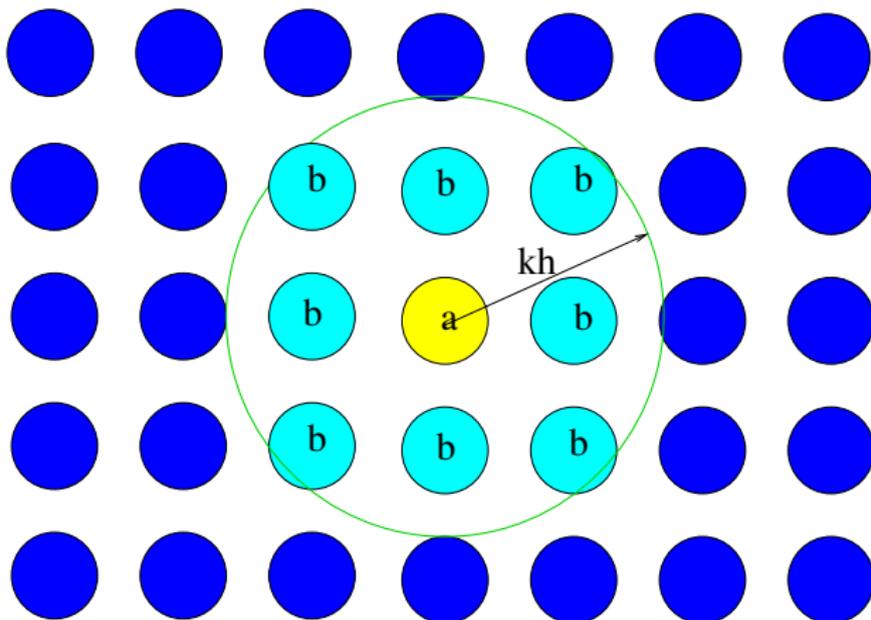
$$\langle f(\mathbf{r}_i) \rangle = \sum_j f(\mathbf{r}_j) \frac{m_j}{\rho_j} w(\mathbf{r}_i - \mathbf{r}_j, h_j)$$

# 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?

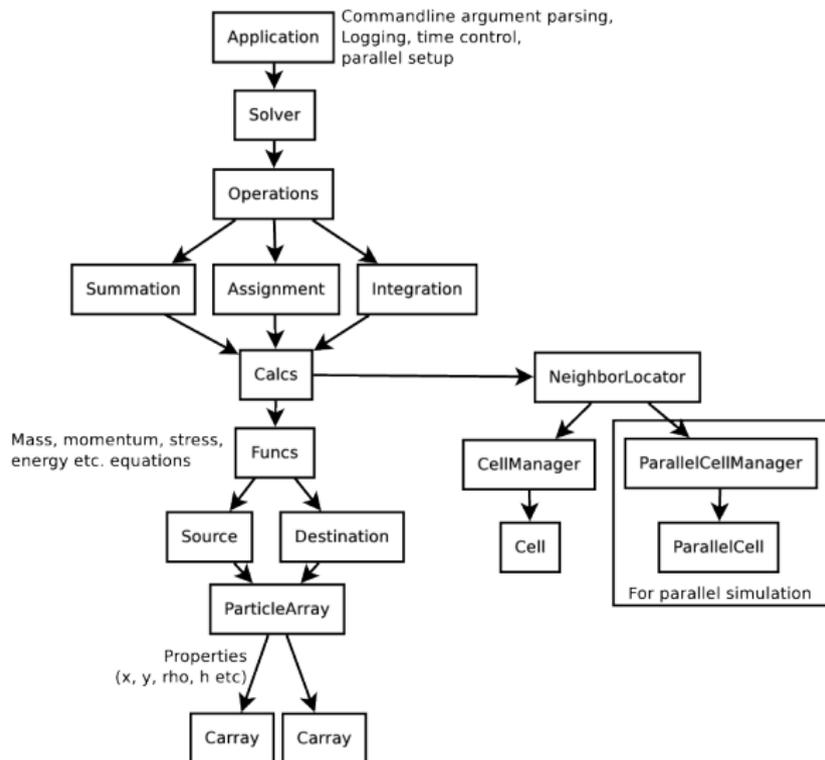
## Python

- 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  $\sigma_{ij}$ ) while creating the particles for the an problem to solve

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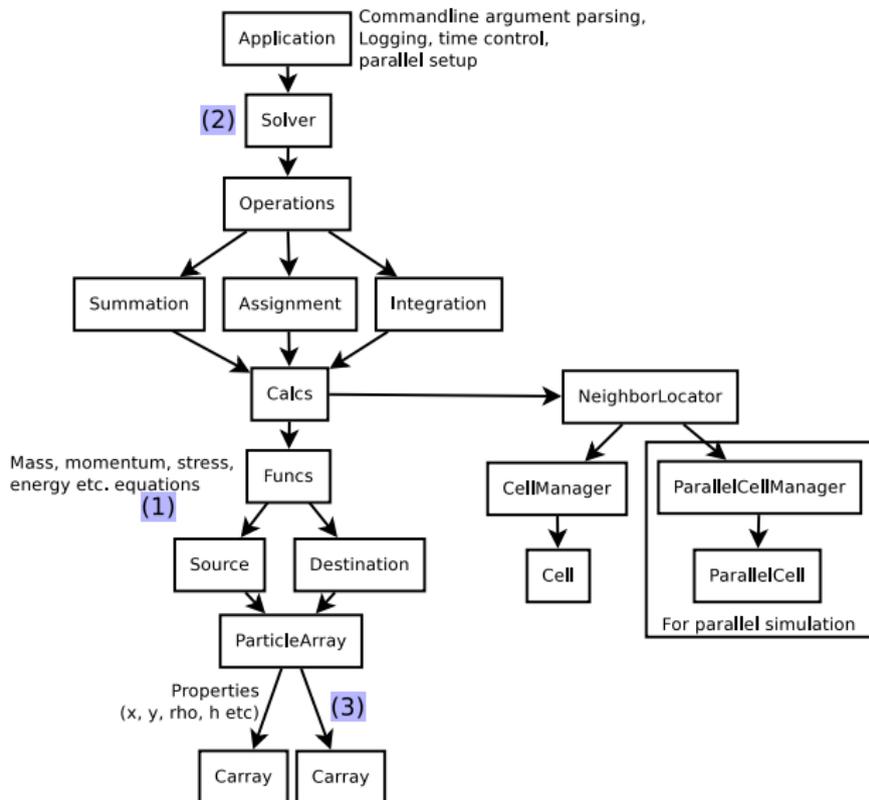
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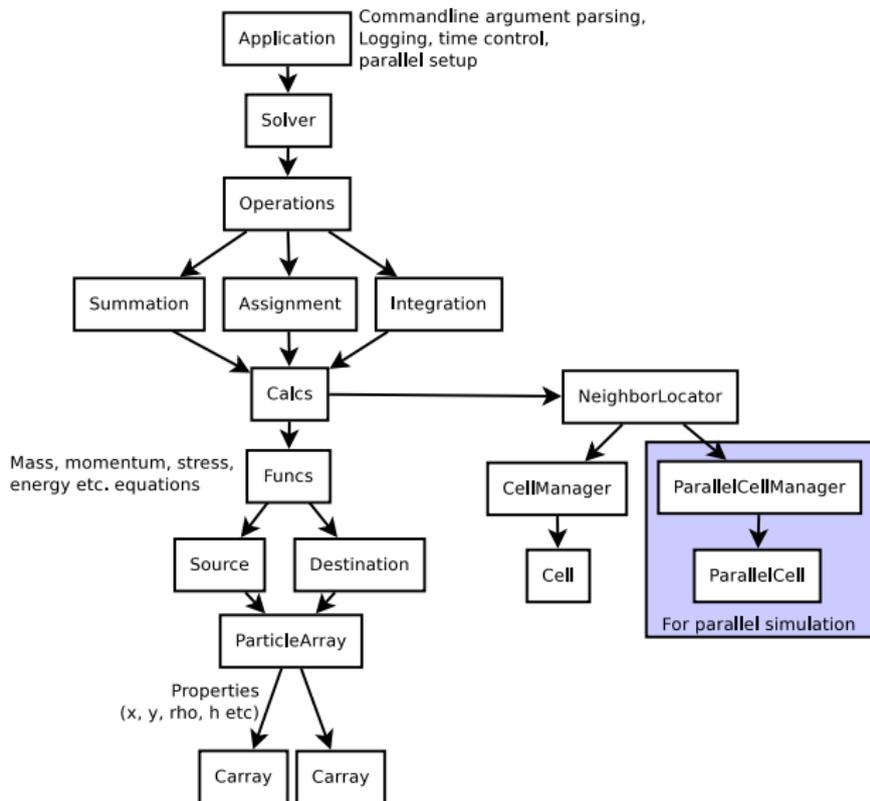
# Extension



# 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



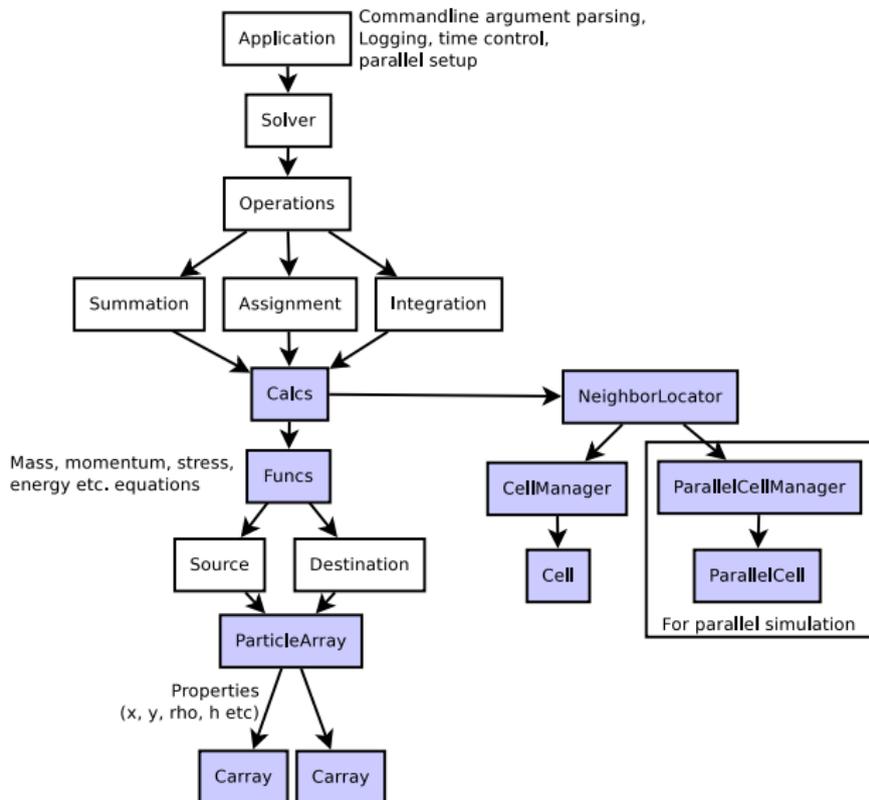
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# Cython

- 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

File View

Percent File View

Name	Calls	RCalls	Local	/Call	Cum	/Call	File
<method 'sph' of 'pysph.sphs...	16	16	0.94994	0.05937	0.95297	0.05956	
<module>	1	1	0.15121	0.15121	0.15182	0.15182	
<module>	1	1	0.05611	0.05611	0.06093	0.06093	
<module>	1	1	0.05479	0.05479	0.05507	0.05507	
join	3625	3625	0.04735	0.00001	0.07465	0.00002	
<posix.lstat>	2965	2965	0.02678	0.00001	0.02678	0.00001	
realpath	680	699	0.02537	0.00004	0.18889	0.00028	
<method 'startswith' of 'str' o...	14019	14019	0.02345	0.00000	0.02345	0.00000	
<module>	1	1	0.02199	0.02199	0.02291	0.02291	
<method 'endswith' of 'str' obj...	12983	12983	0.01991	0.00000	0.01991	0.00000	
find_on_path	330	330	0.01807	0.00005	0.09876	0.00030	
<imp.find_module>	479	479	0.01797	0.00004	0.01797	0.00004	
islink	2965	2965	0.01749	0.00001	0.06012	0.00002	
_parse	114	282	0.01717	0.00006	0.06200	0.00054	
normpath	735	735	0.01615	0.00002	0.02744	0.00004	
<method 'append' of 'list' obje...	11055	11055	0.01423	0.00000	0.01423	0.00000	
_compile	75	484	0.01327	0.00003	0.03643	0.00049	
<len>	9626	9864	0.01284	0.00000	0.01344	0.00000	
readline	552	552	0.01197	0.00002	0.02150	0.00004	
__next	2524	2524	0.01184	0.00000	0.01766	0.00001	
<built-in method match>	4616	4616	0.01183	0.00000	0.01183	0.00000	
<isinstance>	6198	6198	0.01160	0.00000	0.01171	0.00000	
<map>	429	429	0.01083	0.00003	0.19067	0.00044	
S_ISLNK	2952	2952	0.01073	0.00000	0.01584	0.00001	
<posix.listdir>	68	68	0.01070	0.00016	0.01070	0.00016	
add_newdoc	242	242	0.01061	0.00004	0.01428	0.00006	
_normalize_cached	5716	5716	0.01032	0.00000	0.04118	0.00001	
__init__	1	1	0.01030	0.01030	0.01059	0.01059	
yield_lines	1043	1939	0.01017	0.00001	0.02392	0.00002	
<module>	1	1	0.01010	0.01010	0.24441	0.24441	
find_module	479	479	0.00882	0.00002	0.15561	0.00032	

Callees All Callees **Callers** All Callers Source Code

Name	Calls	RCalls	Local	/Call	Cum	/Call	File
	0	99217	0.00000	0.00000	1.22456	0.00001	<bu
do_step	1	1	0.00086	0.00086	0.48962	0.48962	intet
eval	1	1	0.00040	0.00040	0.46717	0.46717	intet

<module>@bench.py:28 [2.195s]

# 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%
<b>Total</b>	<b>132</b>	<b>73</b>	<b>0</b>	<b>45%</b>

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

```

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 !!!