



A Pythonic interface to a particle-resolved Monte Carlo aerosol simulation framework

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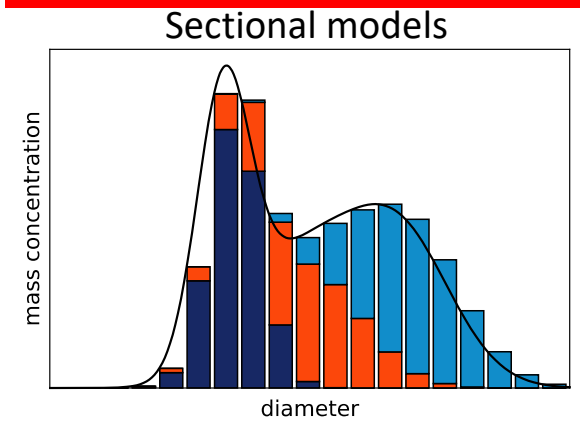
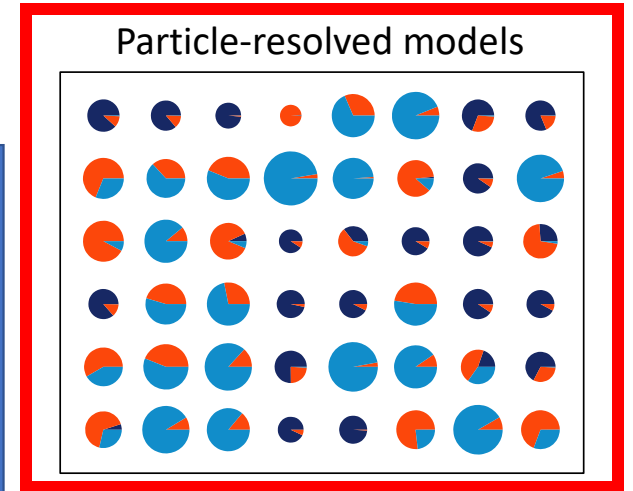
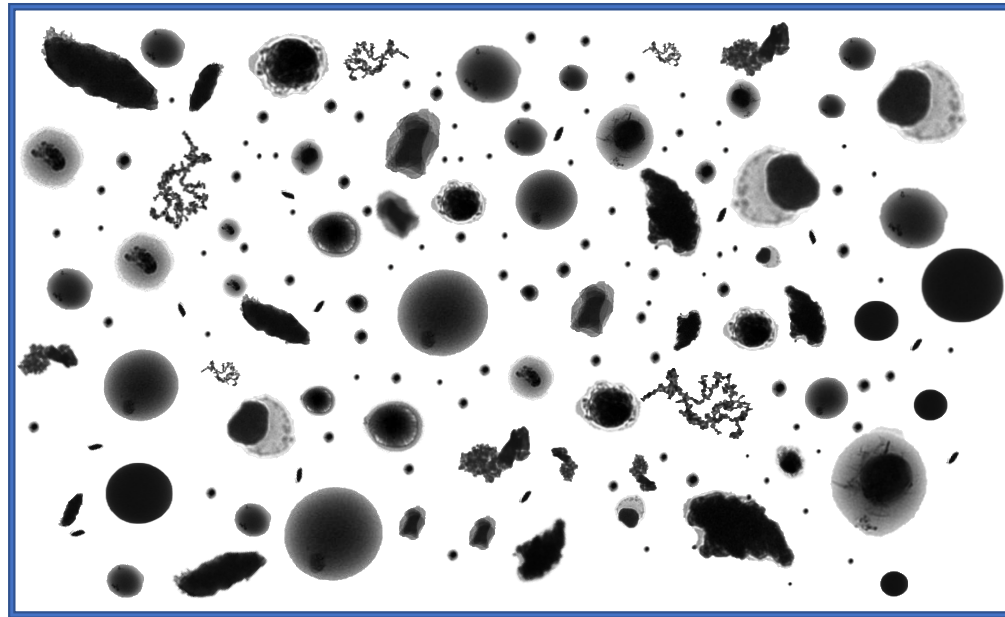
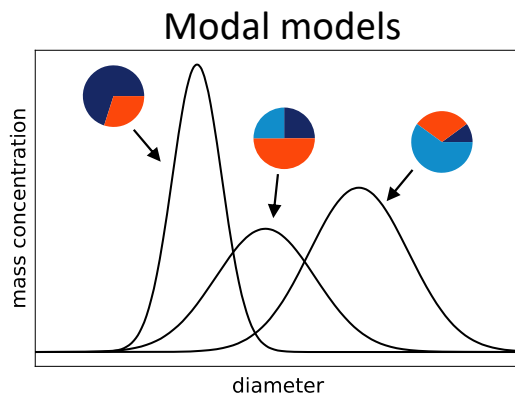
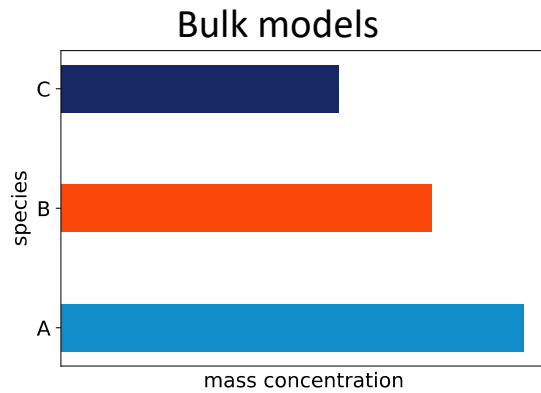
December 7th, 2023



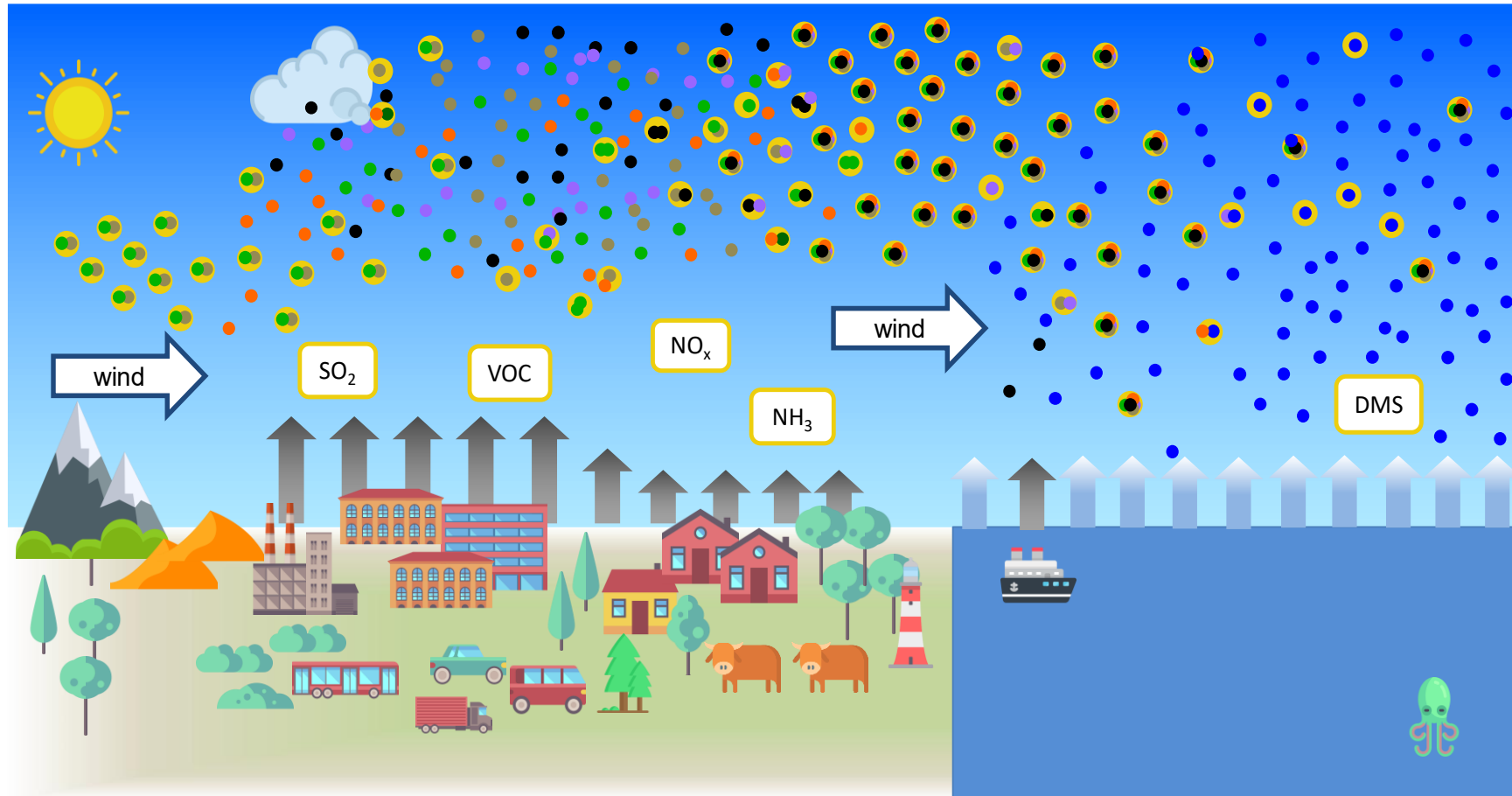
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Particle-resolved modeling

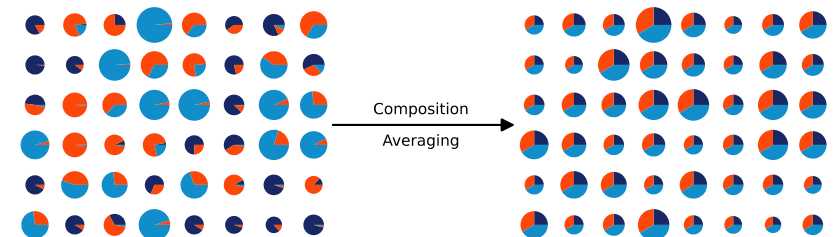
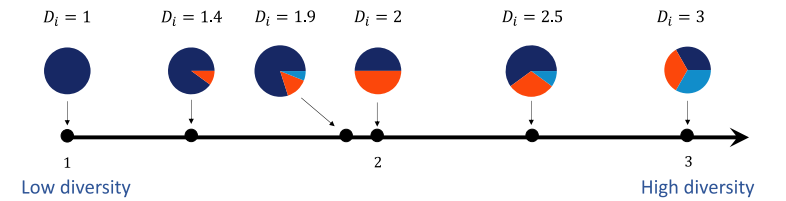
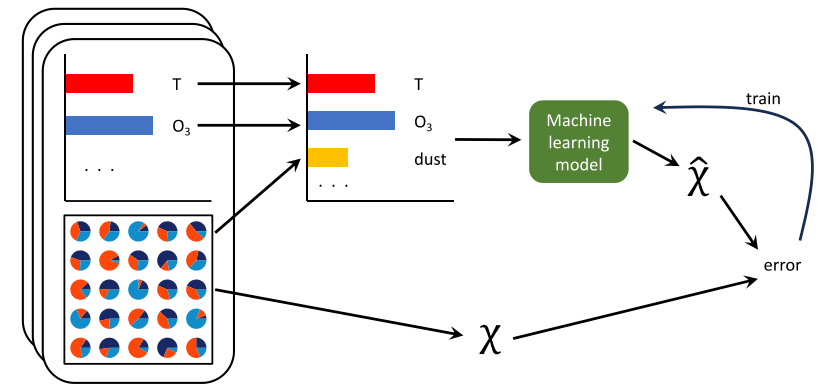
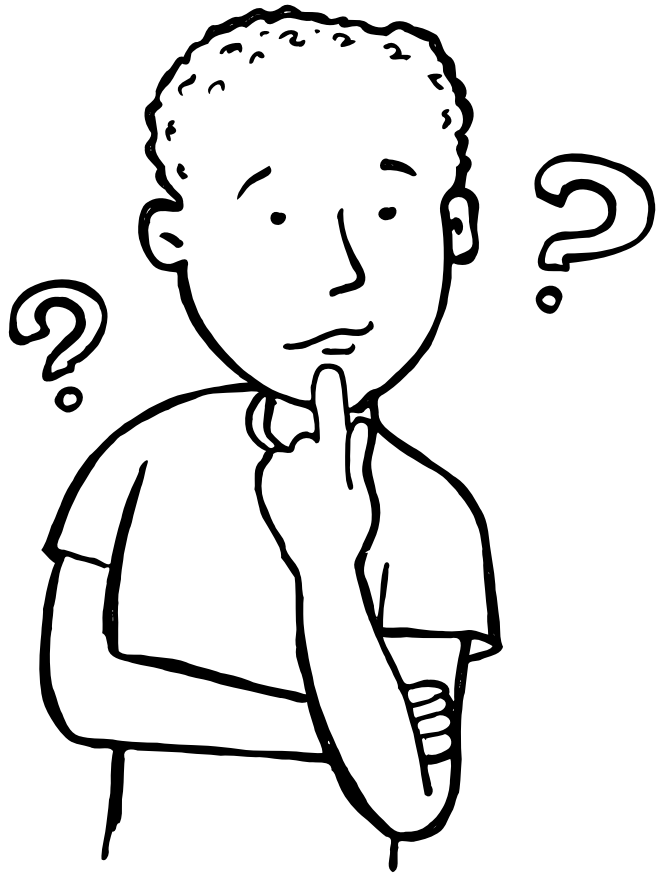


PartMC

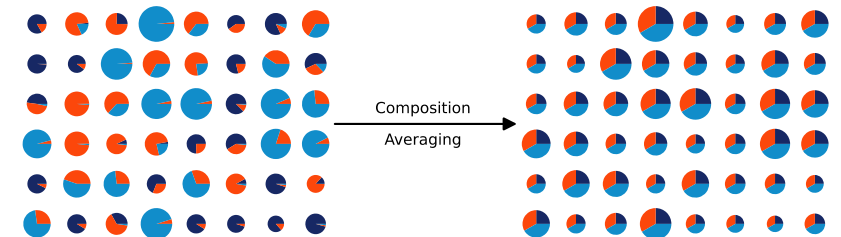
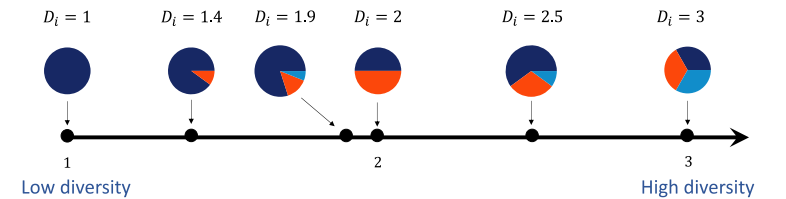
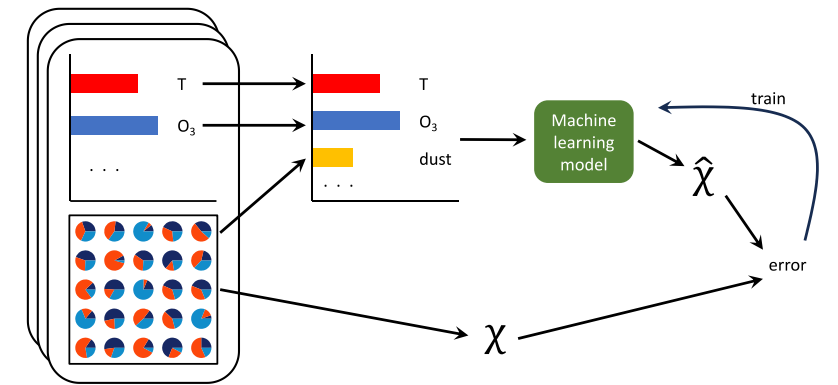
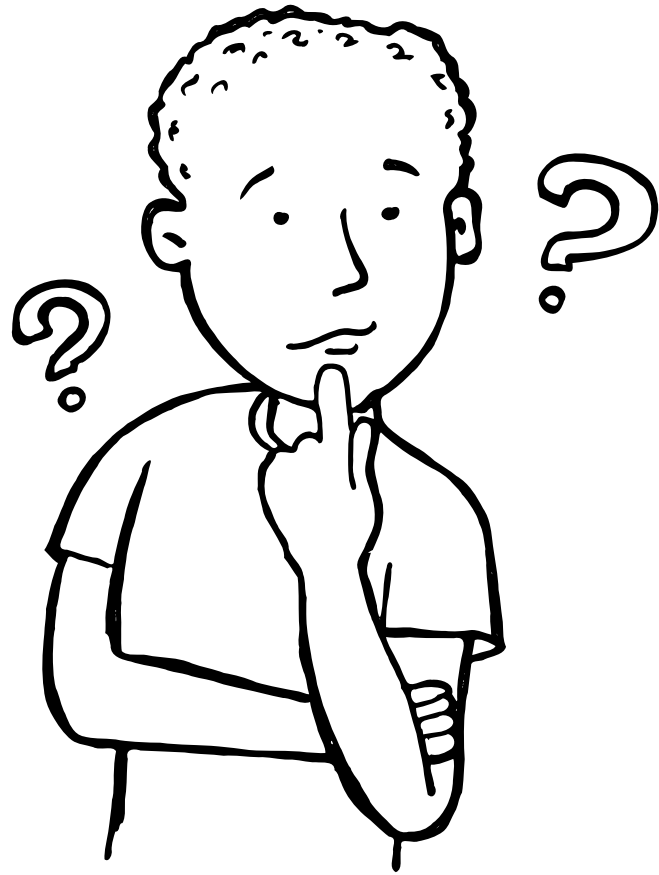


- Emission from primary sources
- Brownian coagulation
- Nucleation scavenging
- Dry deposition

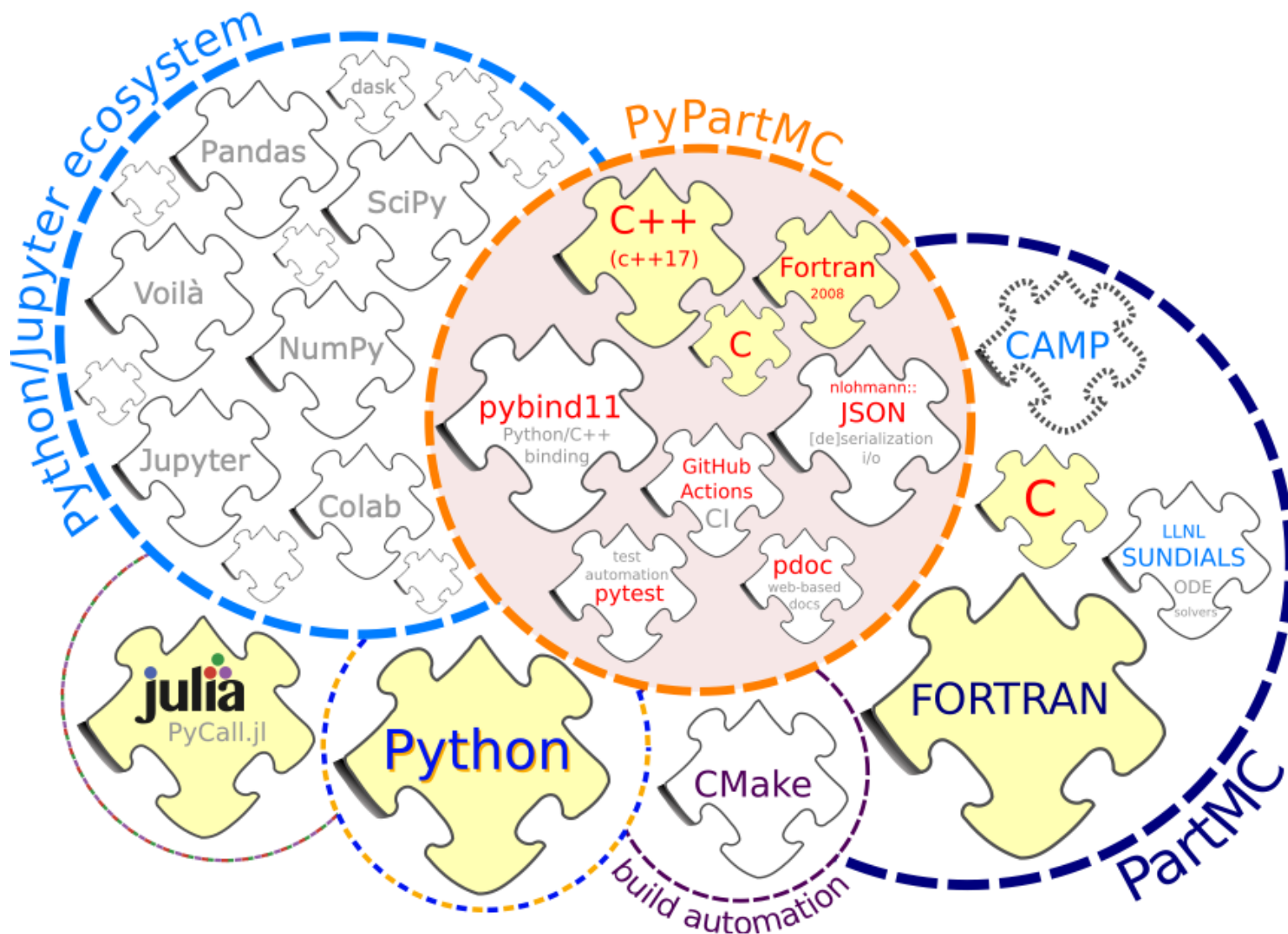
Why do we need PyPartMC?



Why do we need PyPartMC?



Key technologies



Key technologies

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Classes

- **`AeroData`**
 - `densities`
 - `density`

Package PyPartMC

► [EXPAND SOURCE CODE](#)

Functions

```
def condense_equilib_particle(arg0: _PyPartMC.EnvState, arg1: _PyPartMC.AeroData, arg2: _PyPartMC.AeroParticle)
```

Determine the water equilibrium state of a single particle.

```
def condense_equilib_particles(arg0: _PyPartMC.EnvState, arg1: _PyPartMC.AeroData, arg2: _PyPartMC.AeroState)
```

Call `condense_equilib_particle()` on each particle in the aerosol to ensure that every particle has its water content in equilibrium.

```
def diam2rad(arg0: float) -> float
```

Convert diameter (m) to radius (m).

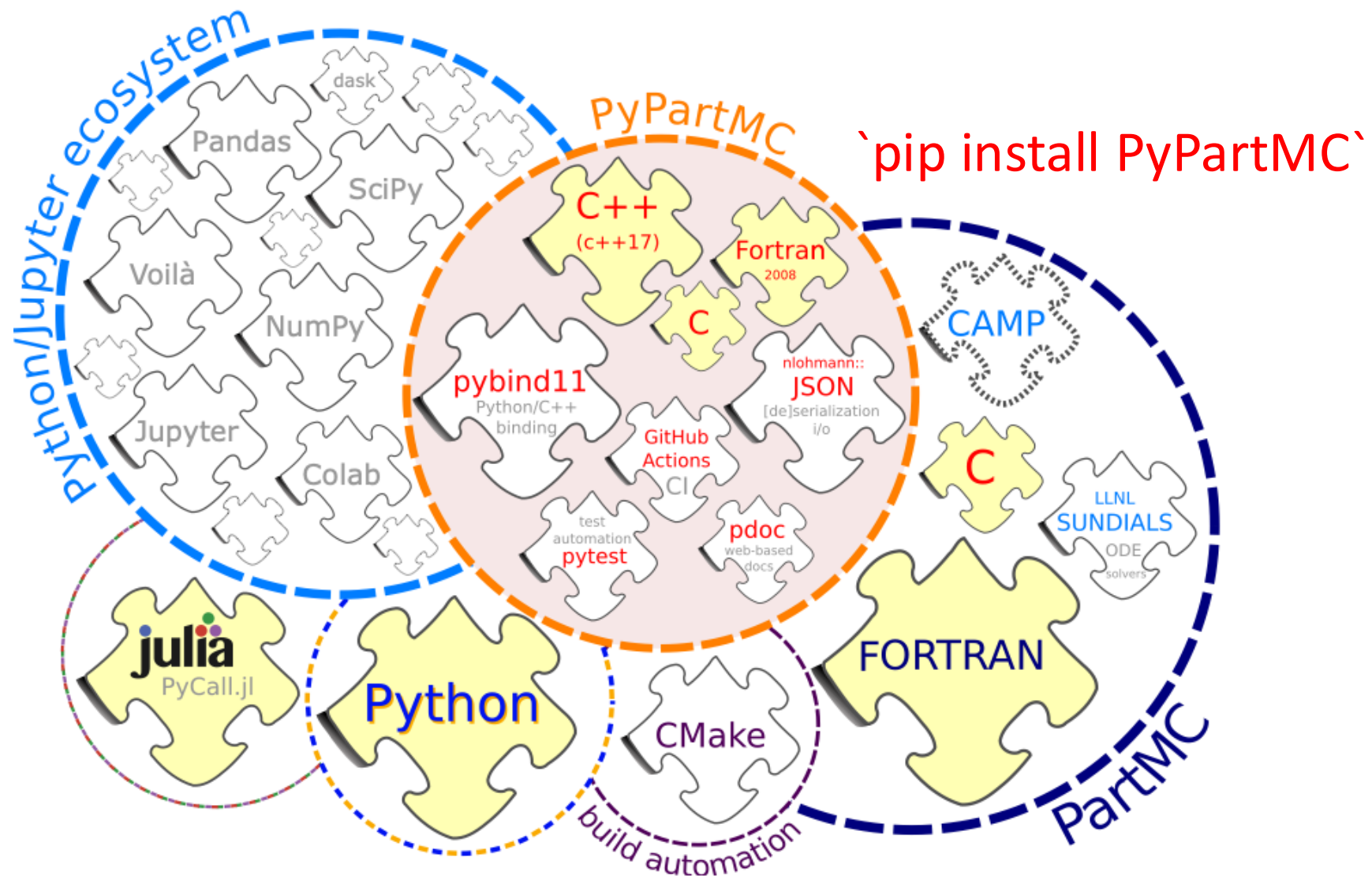
```
def histogram_1d(arg0: _PyPartMC.BinGrid, arg1: List[float], arg2: List[float]) -> List[float]
```

Return a 1D histogram with of the given weighted data, scaled by the bin sizes.

```
def histogram_2d(arg0: _PyPartMC.BinGrid, arg1: List[float], arg2: _PyPartMC.BinGrid, arg3: List[float], arg4: List[float]) -> List[List[float]]
```

Return a 2D histogram with of the given weighted data, scaled by the bin sizes.

Key technologies





```

module pmc_aero_particle

  type aero_particle_t
    !> Constituent species volumes [length # of species] (m^3).
    eal(kind=dp), allocatable :: vol(:)
  end type aero_particle_t

contains

  !> Total volume of the particle (m^3).
  elemental real(kind=dp) function aero_particle_volume(aero_particle)

  !> Particle.
  type(aero_particle_t), intent(in) :: aero_particle

  aero_particle_volume = sum(aero_particle%vol)

end function aero_particle_volume

end module

```



- PyPartMC operates with unmodified PartMC
- Data remains in underlying PartMC Fortran types
- Modularity of PartMC allows wrapping structure to work
- C++ wrappers allow pybind11 to automate API generation



```

module PyPartMC_aero_particle
  use iso_c_binding
  use pmc_aero_particle
  implicit none

contains

  subroutine f_aero_particle_volume(ptr_c, vol) bind(C)
    type(aero_particle_t), pointer :: ptr_f => null()
    type(c_ptr), intent(in) :: ptr_c
    real(c_double), intent(out) :: vol

    call c_f_pointer(ptr_c, ptr_f)

    vol = aero_particle_volume(ptr_f)
  end subroutine

end module

```



```

extern "C" void f_aero_particle_volume(const void *ptr, double *vol) noexcept;

namespace py = pybind11;
struct AeroParticle {
  static auto volume(const AeroParticle &self) {
    double vol;
    f_aero_particle_volume(
      self.ptr.f_arg(),
      &vol
    );
    return vol;
  }
}

```



```

PYBIND11_MODULE(_PyPartMC, m) {
  pybind11::class_<AeroParticle>(m, "AeroParticle",
    R"pbdoc(
      Single aerosol particle data structure.

      The \c vol array stores the total volumes of the different
      species that make up the particle.
    )pbdoc"
  )
  .def_property_readonly("volume", AeroParticle::volume,
    "Total volume of the particle (m^3).")
}

```



```

import PyPartMC as ppmc
from PyPartMC import si
import gc

class TestAeroParticle:

  @staticmethod
  def test_volume():
    aero_data_arg = (
      {"H2O": [1000 * si.kg / si.m**3,
        1, 18e-3 * si.kg / si.mol, 0]}
    )
    aero_data = ppmc.AeroData(aero_data_arg)
    volumes = [1e-6]
    sut = ppmc.AeroParticle(aero_data, volumes)
    aero_data = None
    gc.collect()

    vol = sut.volume

    assert vol == sum(volumes)

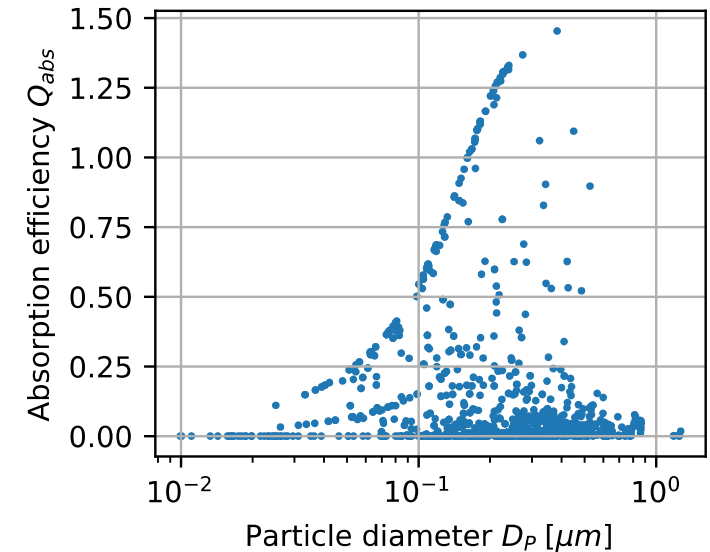
```

Example using external package

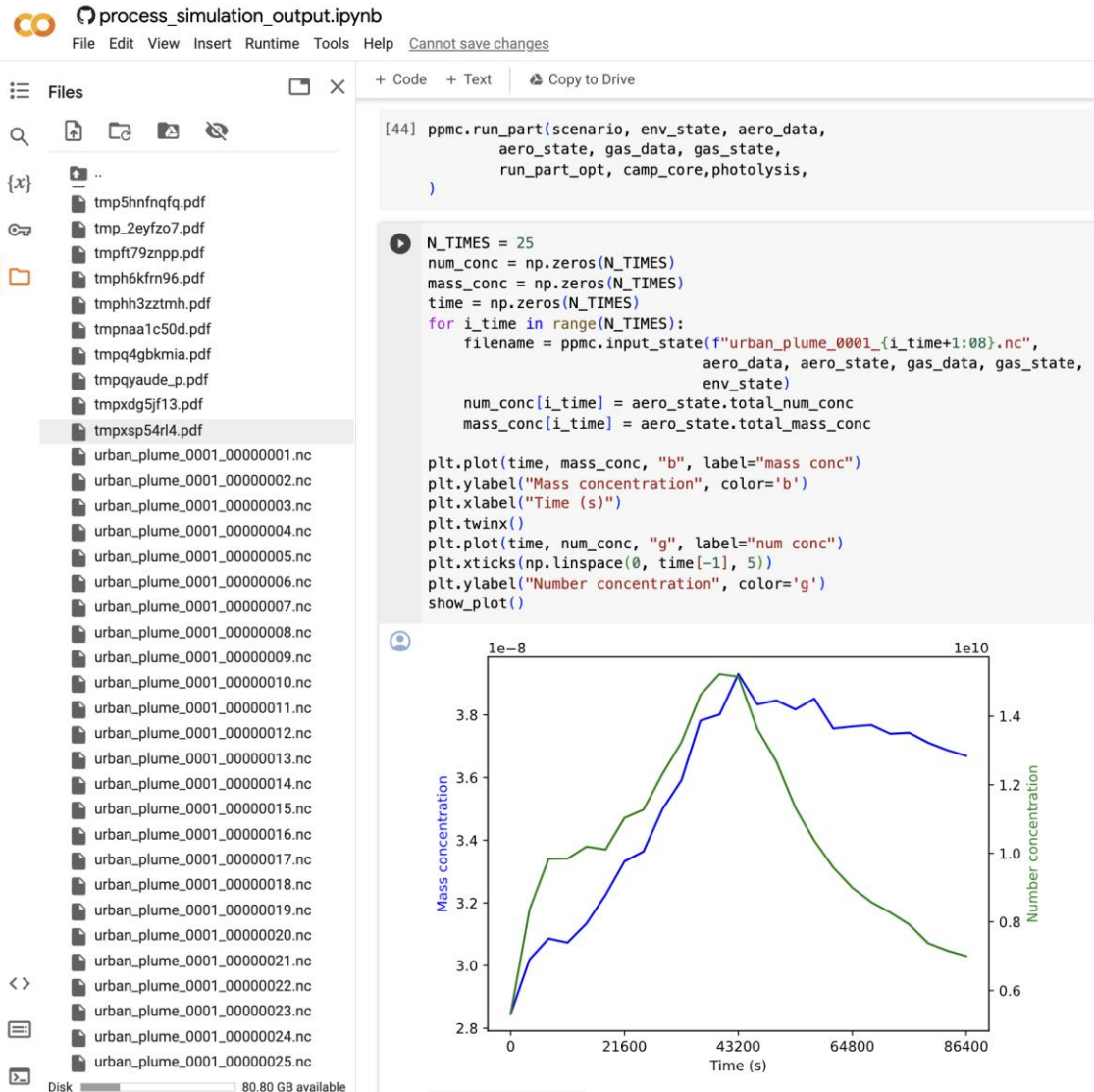
User-defined function employs the **PyMieScatt** package to compute optical properties at each timestep

```
def aero_state_compute_optical(aero_state_optical):  
    wl = 550.0 # unit: nm  
    refr_shell = 1.52+0*1j  
    refr_core = 1.82+0.74*1j  
    diams_core = np.array(aero_state_optical.diameters(include=["BC"])) * 1e9 # unit: nm  
    diams_total = np.array(aero_state_optical.diameters()) * 1e9 # unit: nm  
  
    qsca_part = np.zeros(len(aero_state_optical))  
    qabs_part = np.zeros(len(aero_state_optical))  
    for i_part in range(len(aero_state_optical)):  
        val = PyMieScatt.MieQCoreShell(refr_core,  
                                       refr_shell,  
                                       wl,  
                                       diams_core[i_part],  
                                       diams_total[i_part],  
                                       asDict=True)  
        qsca_part[i_part] = val['Qsca']  
        qabs_part[i_part] = val['Qabs']  
  
    cross_section = np.pi * (diams_total / 2 / 1e9)**2  
  
    num_concs = aero_state_optical.num_concs  
    B_abs = np.sum(qabs_part * cross_section * num_concs)  
    B_sca = np.sum(qsca_part * cross_section * num_concs)  
  
    return (diams_total, qsca_part, qabs_part, B_sca, B_abs)
```

```
for i_block in range(1,N_BLOCKS+1):  
    i_next = int(N_STEPS_PER_BLOCK * i_block)  
    ppmc.run_part_timeblock(  
        scenario,  
        env_state,  
        aero_data,  
        aero_state,  
        gas_data,  
        gas_state,  
        run_part_opt,  
        camp_core,  
        photolysis,  
        i_time,  
        i_next,  
        0,  
    )  
    time[i_block] = env_state.elapsed_time  
    diameters, qsca, qabs, Bsca[i_block], Babs[i_block] = aero_state_compute_optical(aero_state)  
    i_time = i_next + 1
```



Time stepping in Fortran



The screenshot shows a Jupyter Notebook titled "process_simulation_output.ipynb". The left sidebar displays a file explorer with a list of files, including several PDF files and a series of netCDF files named "urban_plume_0001_00000001.nc" through "urban_plume_0001_00000025.nc". The main notebook area contains the following code:

```
[44] ppmc.run_part(scenario, env_state, aero_data,
                  aero_state, gas_data, gas_state,
                  run_part_opt, camp_core, photolysis,
                  )

N_TIMES = 25
num_conc = np.zeros(N_TIMES)
mass_conc = np.zeros(N_TIMES)
time = np.zeros(N_TIMES)
for i_time in range(N_TIMES):
    filename = ppmc.input_state(f"urban_plume_0001_{i_time+1:08}.nc",
                              aero_data, aero_state, gas_data, gas_state,
                              env_state)

    num_conc[i_time] = aero_state.total_num_conc
    mass_conc[i_time] = aero_state.total_mass_conc

plt.plot(time, mass_conc, "b", label="mass conc")
plt.ylabel("Mass concentration", color='b')
plt.xlabel("Time (s)")
plt.twinx()
plt.plot(time, num_conc, "g", label="num conc")
plt.xticks(np.linspace(0, time[-1], 5))
plt.ylabel("Number concentration", color='g')
show_plot()
```

Below the code, a plot is displayed with two y-axes. The x-axis is labeled "Time (s)" and ranges from 0 to 86400. The left y-axis is labeled "Mass concentration" and ranges from 2.8 to 3.8 (scaled by 1e-8). The right y-axis is labeled "Number concentration" and ranges from 0.6 to 1.4 (scaled by 1e10). A blue line represents "mass conc" and a green line represents "num conc". Both lines show a similar trend, increasing to a peak around 43200 seconds and then decreasing.

← Time stepping in Fortran

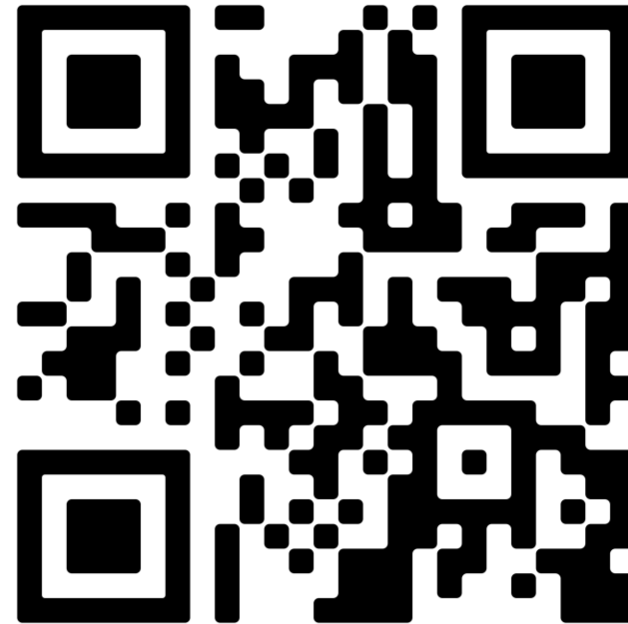
Processes data from outputted netCDF files and use matplotlib to visualize

- Entire simulation runs in Google Colab
- Choice in time stepping control
- Data generation and processing in a single notebook

More than just a wrapper!



GitHub



Documentation

