

Improving the continuum modelling within BioDynaMo

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Continuum modelling with BioDynaMo

Slides overview

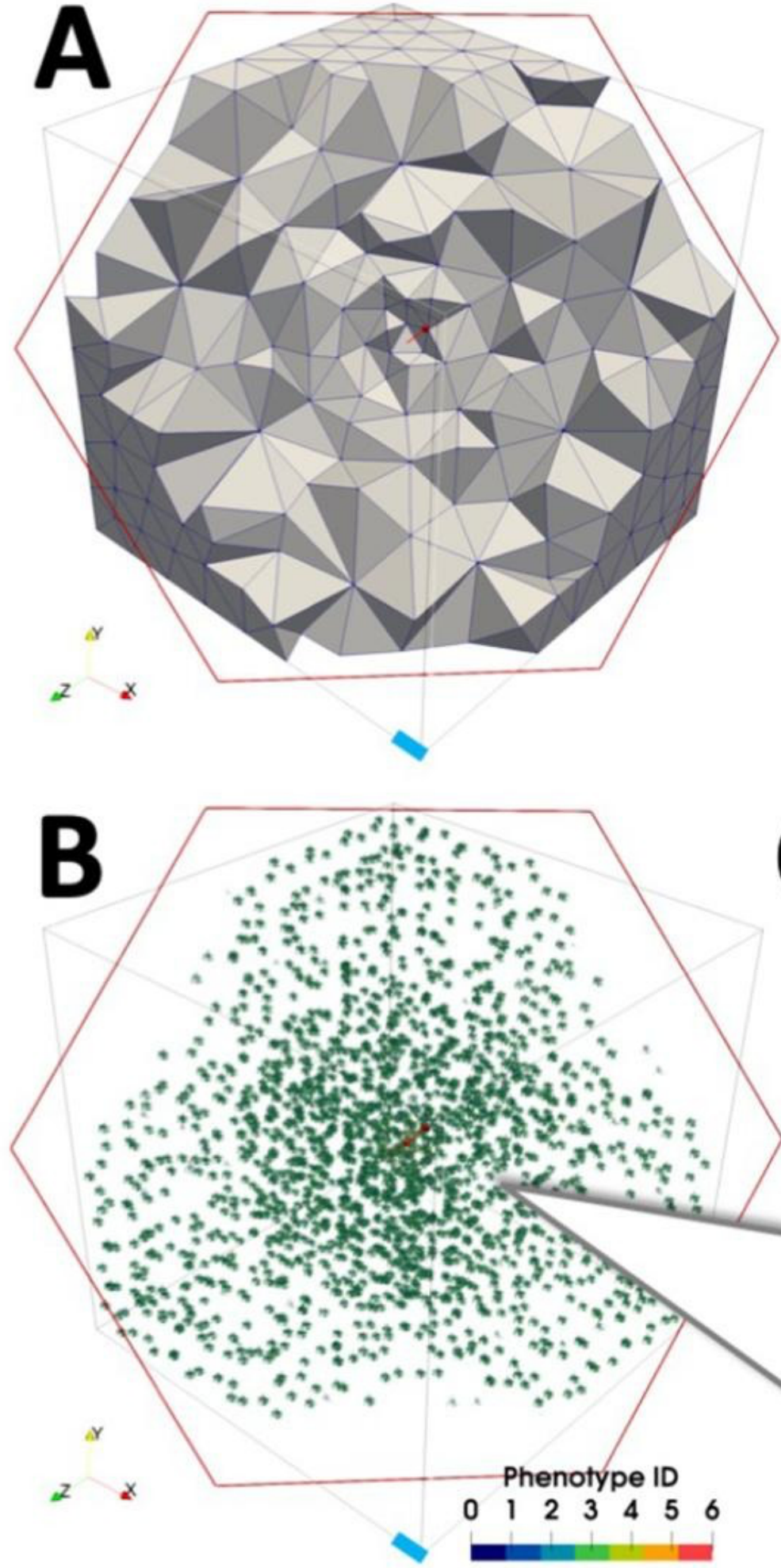
1. Motivation
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3. MFEM Integration
4. A Conceptual Example: Diffusion

Motivation

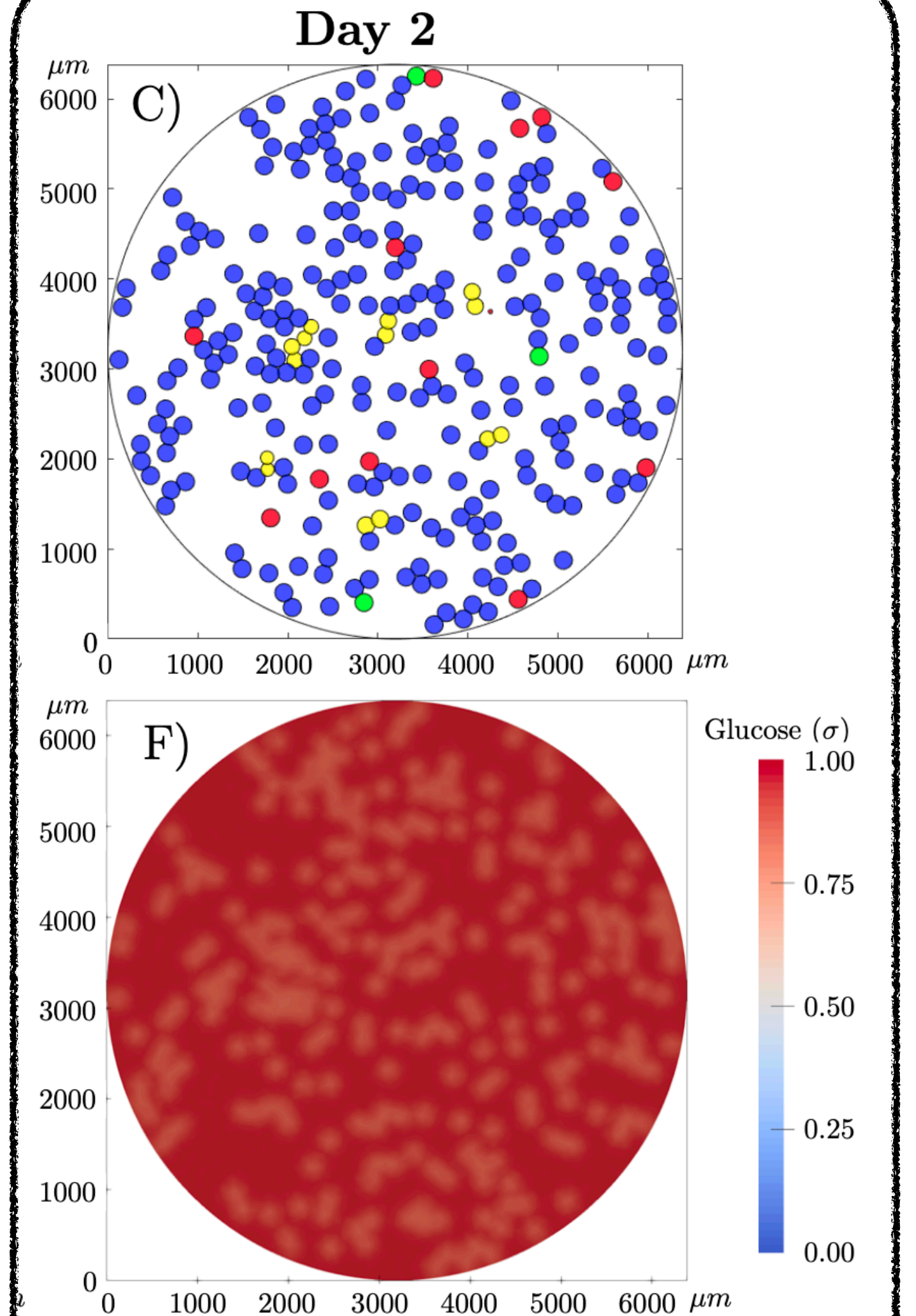


Hybrid Modelling

ABM + continuum model



de Montigny, Methods, 2021
10.1016/j.ymeth.2020.01.006



Lima, bioRxiv, 2021
10.1101/2021.03.03.433731

Current State



Current State

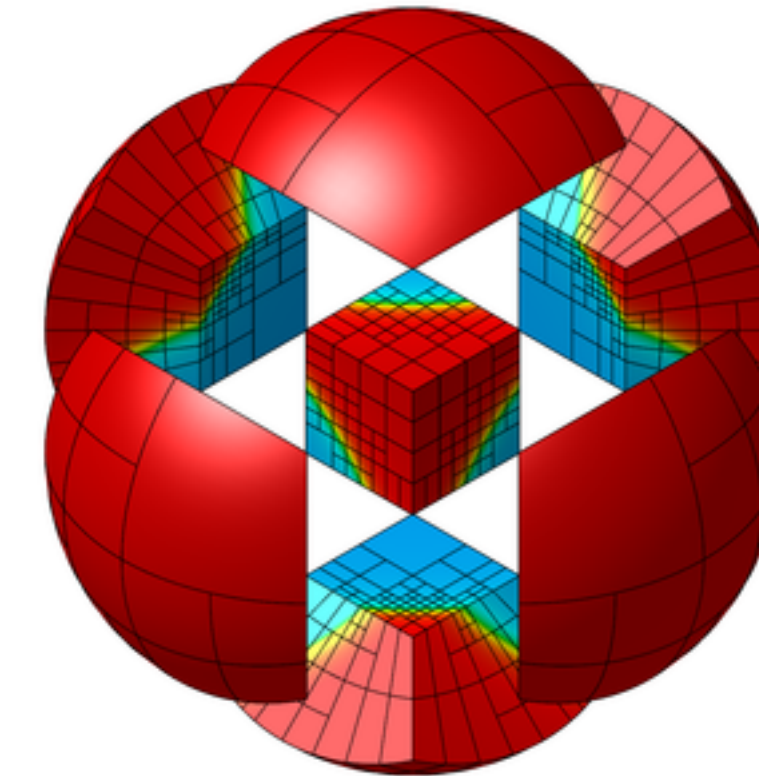
	Currently	Target
Equation	Heat / Diffusion equation	Arbitrary
Time discretization	Forward Difference	Explicit & implicit ODE solvers
Space discretization	Central difference	Finite elements
Geometry	Cube	Arbitrary but convex
ABM - Continuum interactions	Read of values, modify values	Read of values, density function

Conclusion

**It's unfeasible to support a variety of use cases with the current approach.
We need an interface to a sophisticated library handling
different geometries, solvers, and equations.**

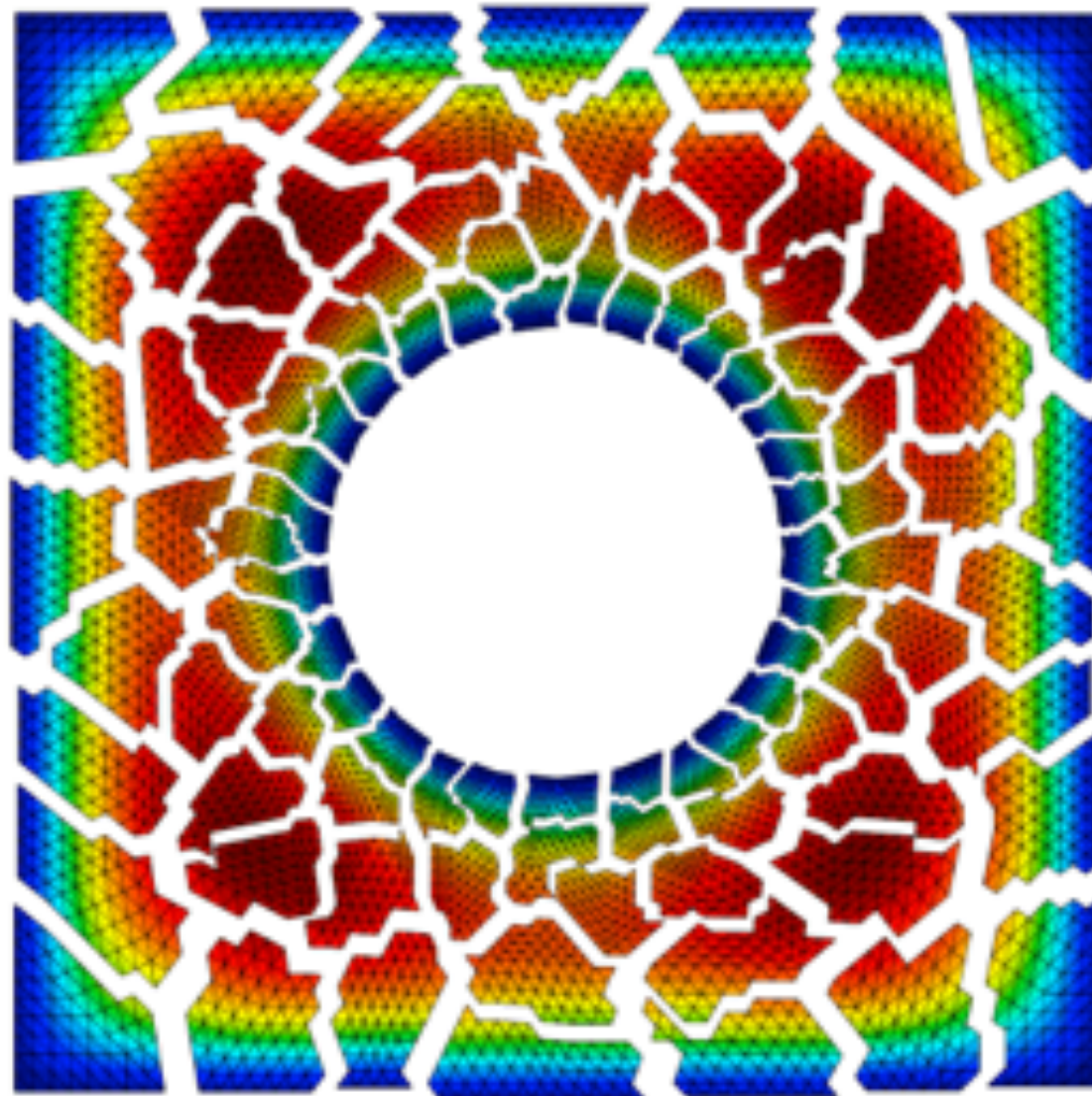
MFEM integration

What is MFEM? Why MFEM?

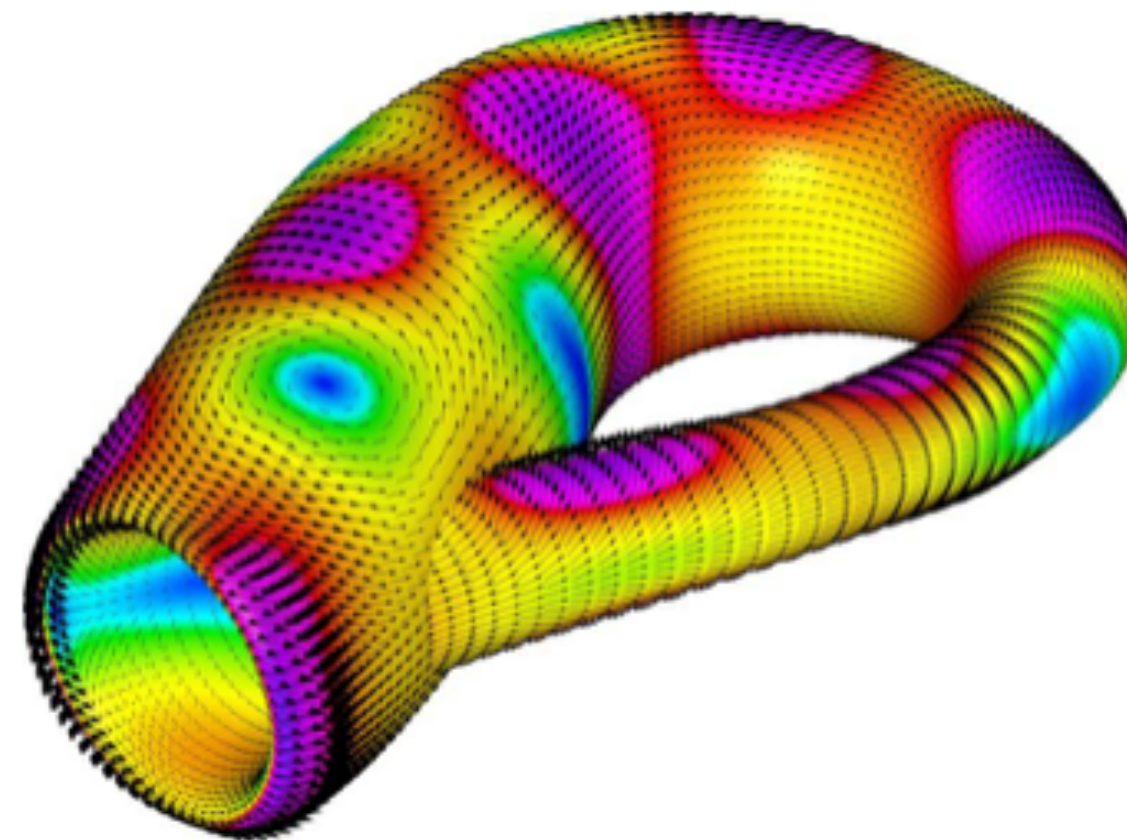


- “A finite element to linear algebra translator”
- Modern, object-oriented software design, C++
- Many options for parallelism
- Cmake compatible
- Active community & responsive developers
- Backed by Lawrence Livermore NL
- Licences compatible (BSD 3-Clause)
- 300K (cloc)

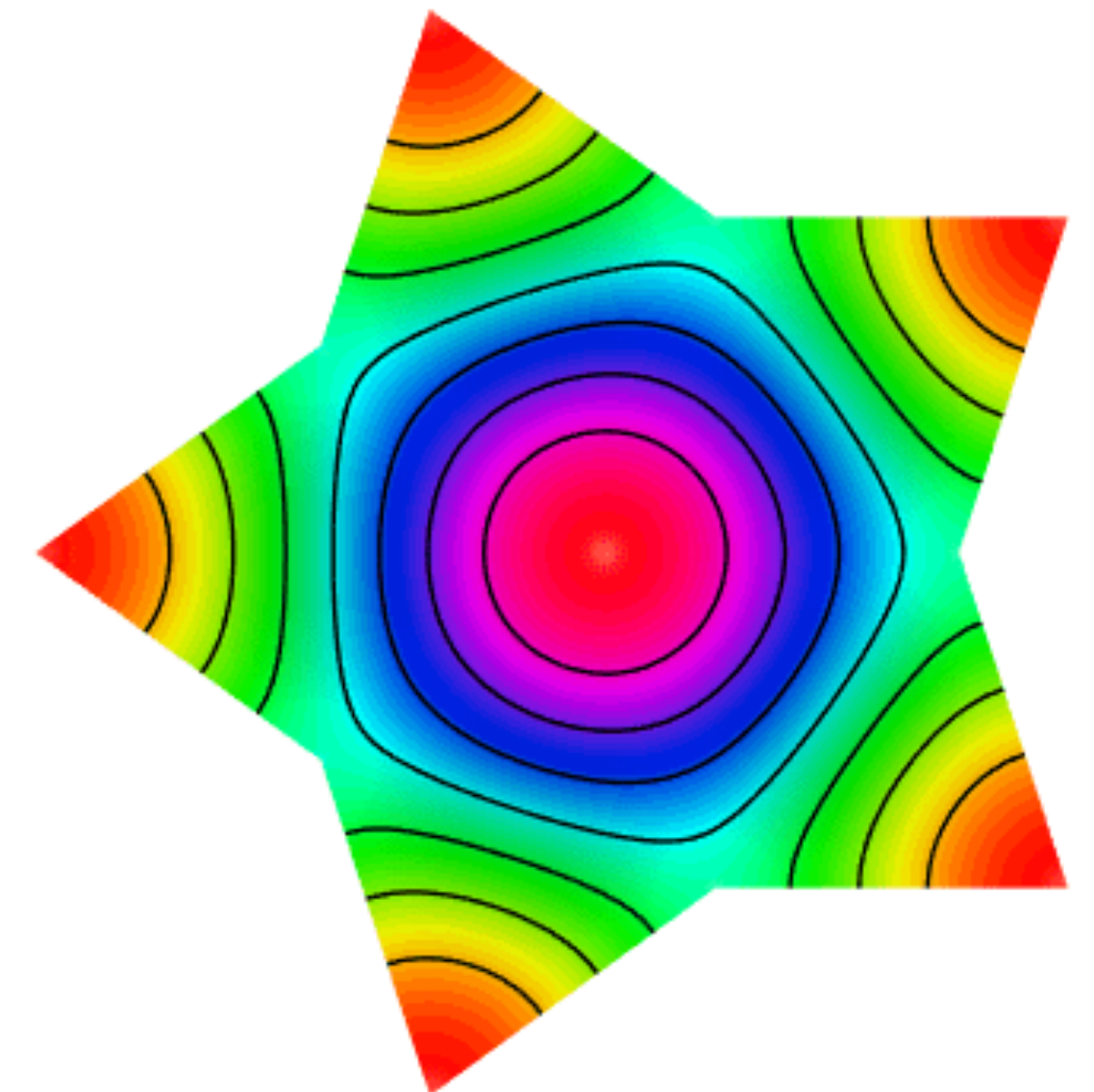
MFEM examples



$$-\Delta \Phi = 1$$



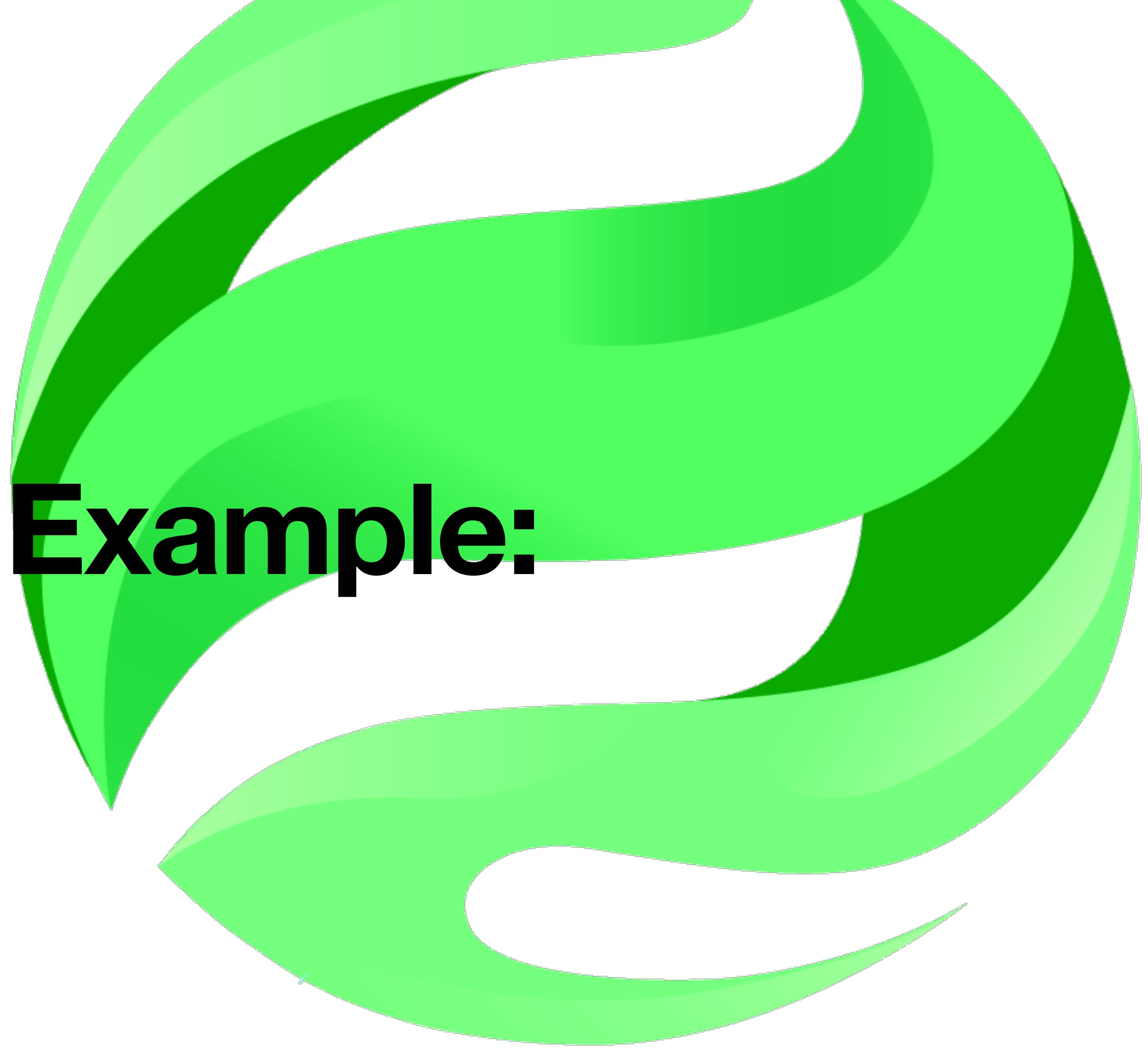
$$\nabla \times \nabla \times E + E = f$$



$$\frac{\partial \Phi}{\partial t} = \nabla (\kappa + \alpha \Phi) \nabla \Phi$$

**Target: support time dependent
diffusion processes based on
the Method of Lines.**

A Conceptual Example: Diffusion



Equation:

$$\partial_t \Phi - \nabla D \nabla \Phi = f$$

FE expansion:

$$\Phi(x, t) = \sum_{j=1}^n u_j(t) \Psi_j(x)$$

Derive ODE system: Insert the expansion and apply $\int_{\Omega} d\Omega \Psi_i(x)$

Result:

$$M \frac{\partial \vec{u}}{\partial t} = D \vec{u} + \vec{f}$$

$$M \in \mathbb{R}^{n \times n}, M_{ij} = \int_{\Omega} \Psi_i \Psi_j d\Omega$$

$$D \in \mathbb{R}^{n \times n}, D_{ij} = \int_{\Omega} \Psi_i \left(\nabla D \nabla \Psi_j \right) d\Omega$$

$$\vec{f} \in \mathbb{R}^n, f_i = \int_{\Omega} \Psi_i f d\Omega$$

MFEM example for diffusion

$f=0$

$$M \in \mathbb{R}^{n \times n}, M_{ij} = \int_{\Omega} \Psi_i \Psi_j d\Omega$$

$$D \in \mathbb{R}^{n \times n}, D_{ij} = \int_{\Omega} \Psi_i \left(\nabla D \nabla \Psi_j \right) d\Omega$$

```
M = new BilinearForm(&fespace);
M->AddDomainIntegrator(new MassIntegrator());
M->Assemble();
M->FormSystemMatrix(ess_tdof_list, M_matrix);

D = new BilinearForm(&fespace);
ConstantCoefficient d_coeff(d_value);
D->AddDomainIntegrator(new DiffusionIntegrator(d_coeff));
D->Assemble();
D->FormSystemMatrix(ess_tdof_list, D_matrix);
```

Usage example

```
ModelInitializer::DefineMFEMSubstanceOnMesh(  
    mesh, kSubstance1, "kSubstance1", order, dimension,  
    bdm::experimental::MFEMODESolver::kRK4Solver,  
    bdm::experimental::PDEOperator::kDiffusion,  
    InitializeGridValues, parameters);
```

Q & A



Backup



Notation:

$$x \in \Omega \subset \mathbb{R}^3 \quad \Phi : (\Omega \times \bar{T}) \rightarrow \mathbb{R}$$

$$t \in \bar{T} = (0, T] \quad f : (\Omega \times \bar{T}) \rightarrow \mathbb{R}$$

Equation:

$$\partial_t \Phi - \nabla D \nabla \Phi = f$$

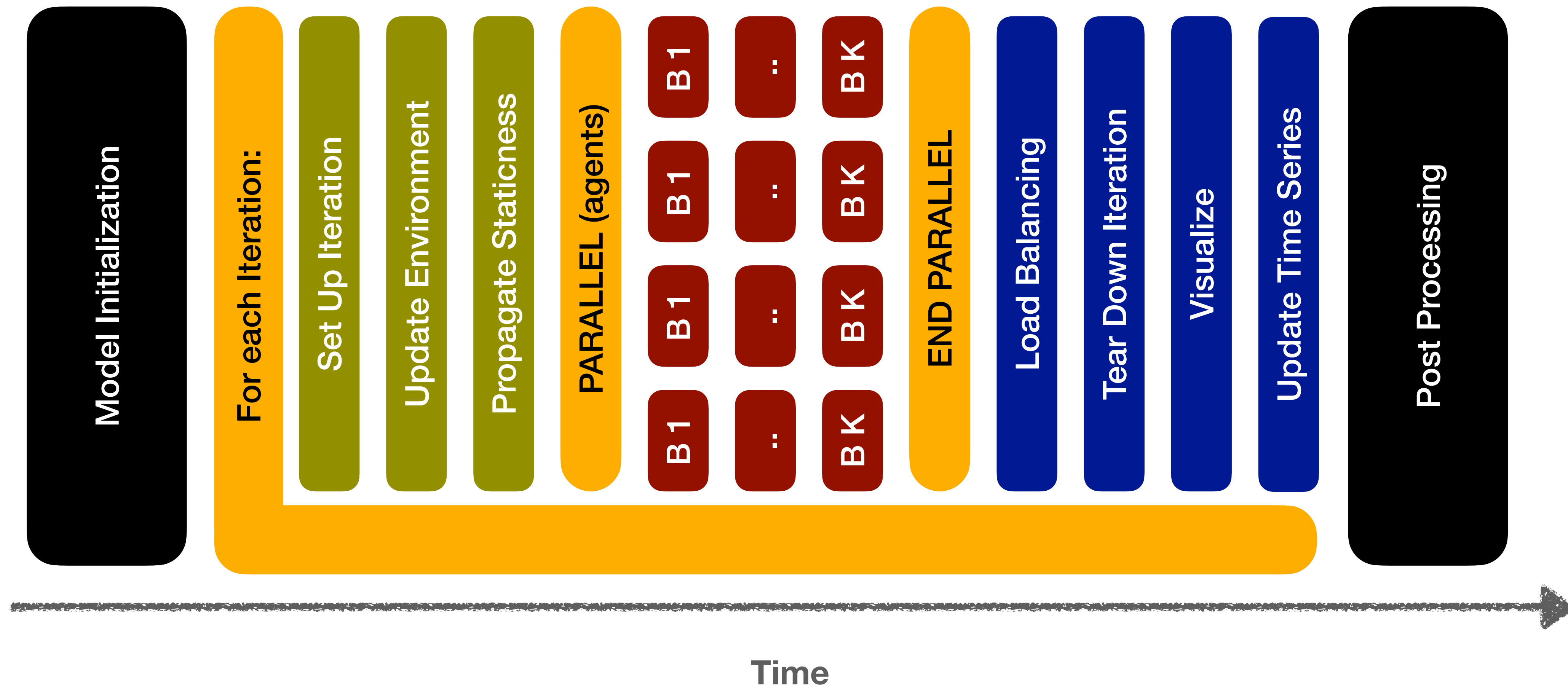
Boundaries:

$$\Phi(x, 0) = g(x) \quad \forall x \in \Omega$$

$$\frac{\partial \phi}{\partial n} = 0 \quad \forall x \in \partial\Omega \quad \wedge \quad \forall t \in \bar{T}$$

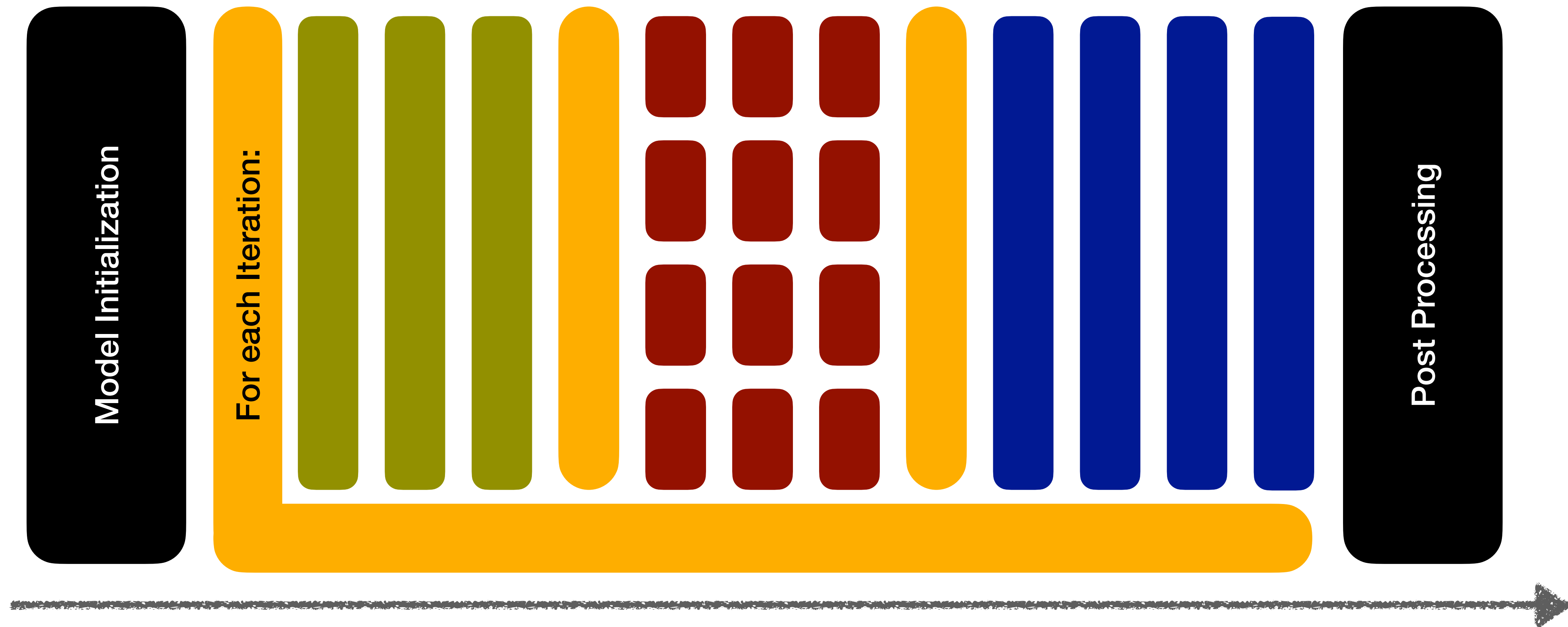
BioDynaMo's ABM engine

(Without user-defined operations)



BioDynaMo's ABM engine

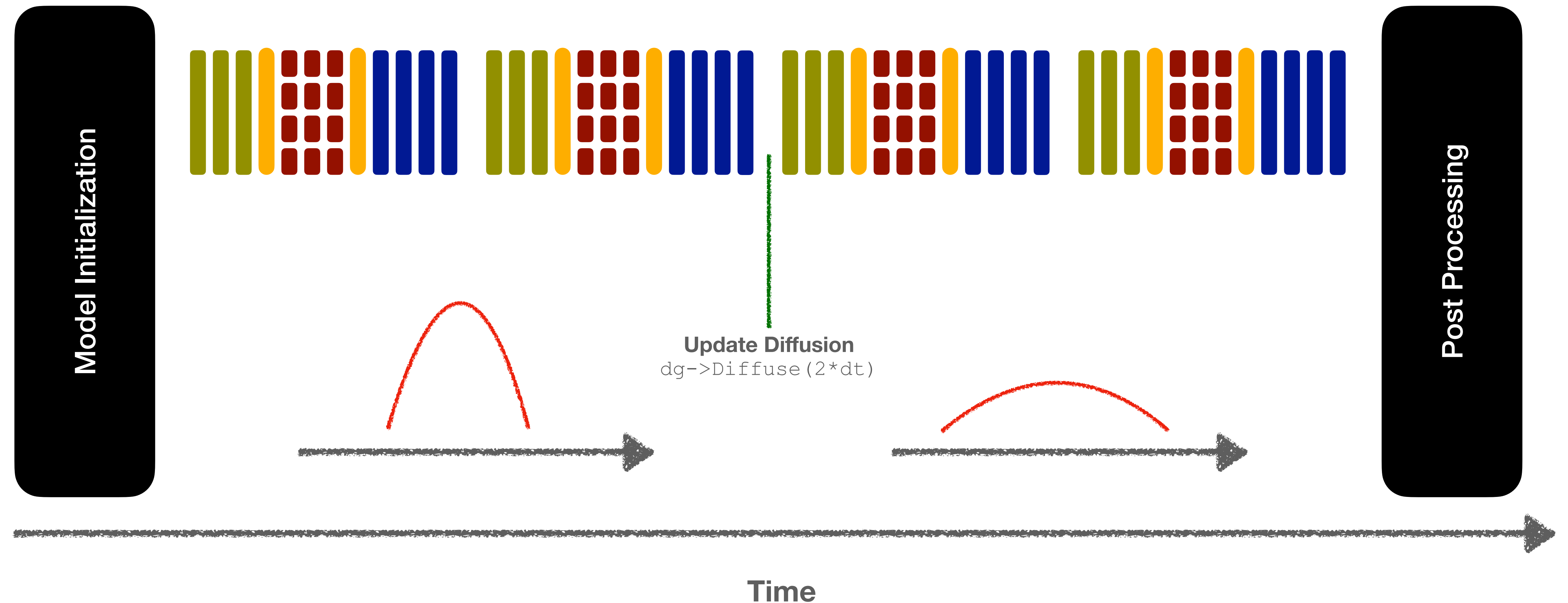
(Without user-defined operations)



Time

BioDynaMo's ABM engine

(Without user-defined operations)



Method of Lines

Main idea

1. Get a clear understanding of the physics that you want to model and write down the corresponding PDE (incl. boundaries)
2. Discretise all but one dimension with FE, FV, or FD
(Practically: discretise space, leave time continuous)
3. Derive the ODE system & construct matrices
4. Solve ODE system with established time-integrators
5. Critically reflect all steps and ideally discuss with colleagues

Spatial discretisation

MFEM - Finite Element Discretization Library

- Use a FE approach for spatial discretisation
- Use MFEM library to construct matrices and time integration
- MFEM offers various interfaces towards highly efficient solvers
- MFEM offers multiple options for parallelism

$$M \frac{\partial \vec{u}}{\partial t} = D \vec{u} + \vec{f}$$

$$M \in \mathbb{R}^{n \times n}, M_{ij} = \int_{\Omega} \Psi_i \Psi_j d\Omega$$

$$D \in \mathbb{R}^{n \times n}, D_{ij} = \int_{\Omega} \Psi_i \left(\nabla D \nabla \Psi_j \right) d\Omega$$

$$\vec{f} \in \mathbb{R}^n, f_i = \int_{\Omega} \Psi_i f d\Omega$$

Software structure

`bdm::TimeDependentScalarField3d`

- Handles mfem related objects
- MFEM mesh
- ODE solver
- PDE operator
- Continuum values

`mfem::TimeDependentOperator`

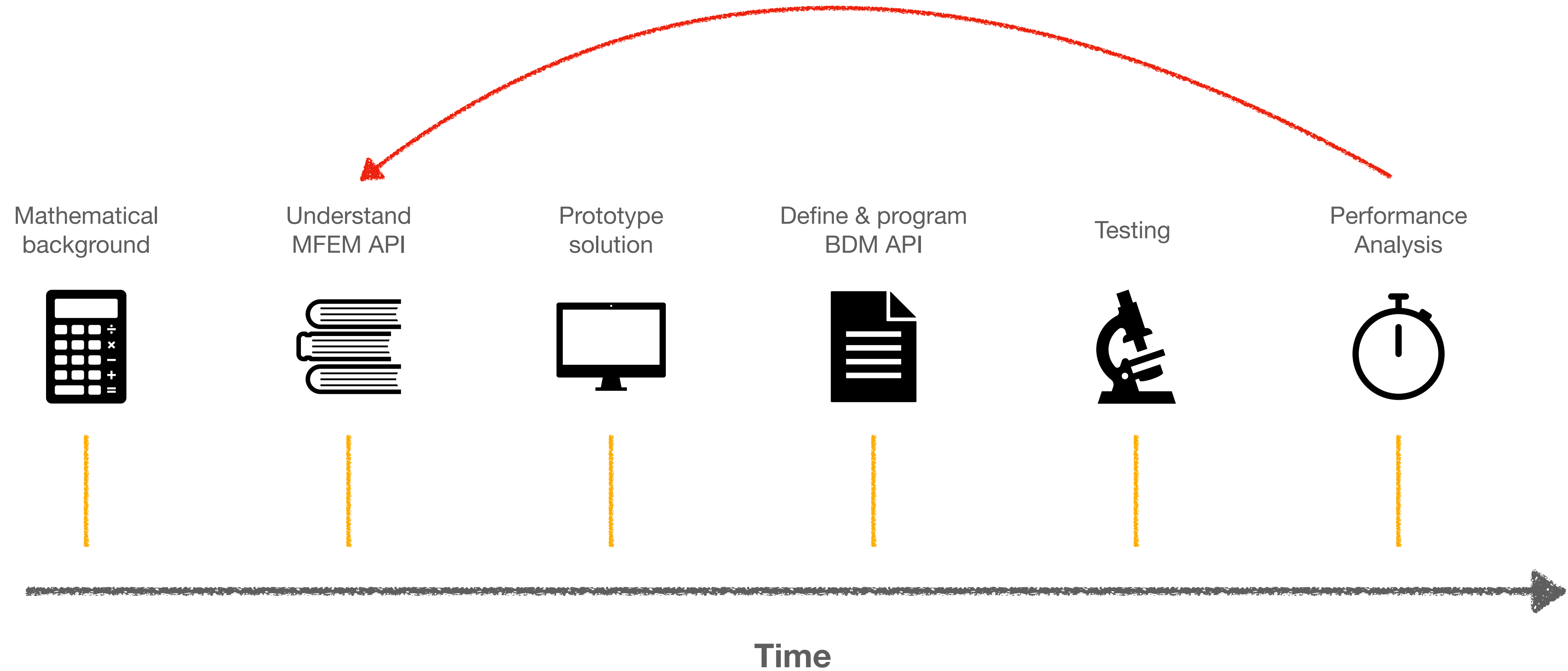
`bdm::MolOperator`

`bdm::DiffusionOperator`

`bdm::ConductionOperator`

`bdm::DiffusionOperatorPerformance`

Development cycle



Coupling - Retrieving continuum information

Cycle 1

```
elem,ip = mfem::Mesh.FindPoints(agent->pos);  
value = mfem::GridFunction.GetValue(elem, ip);
```

Runtimes:

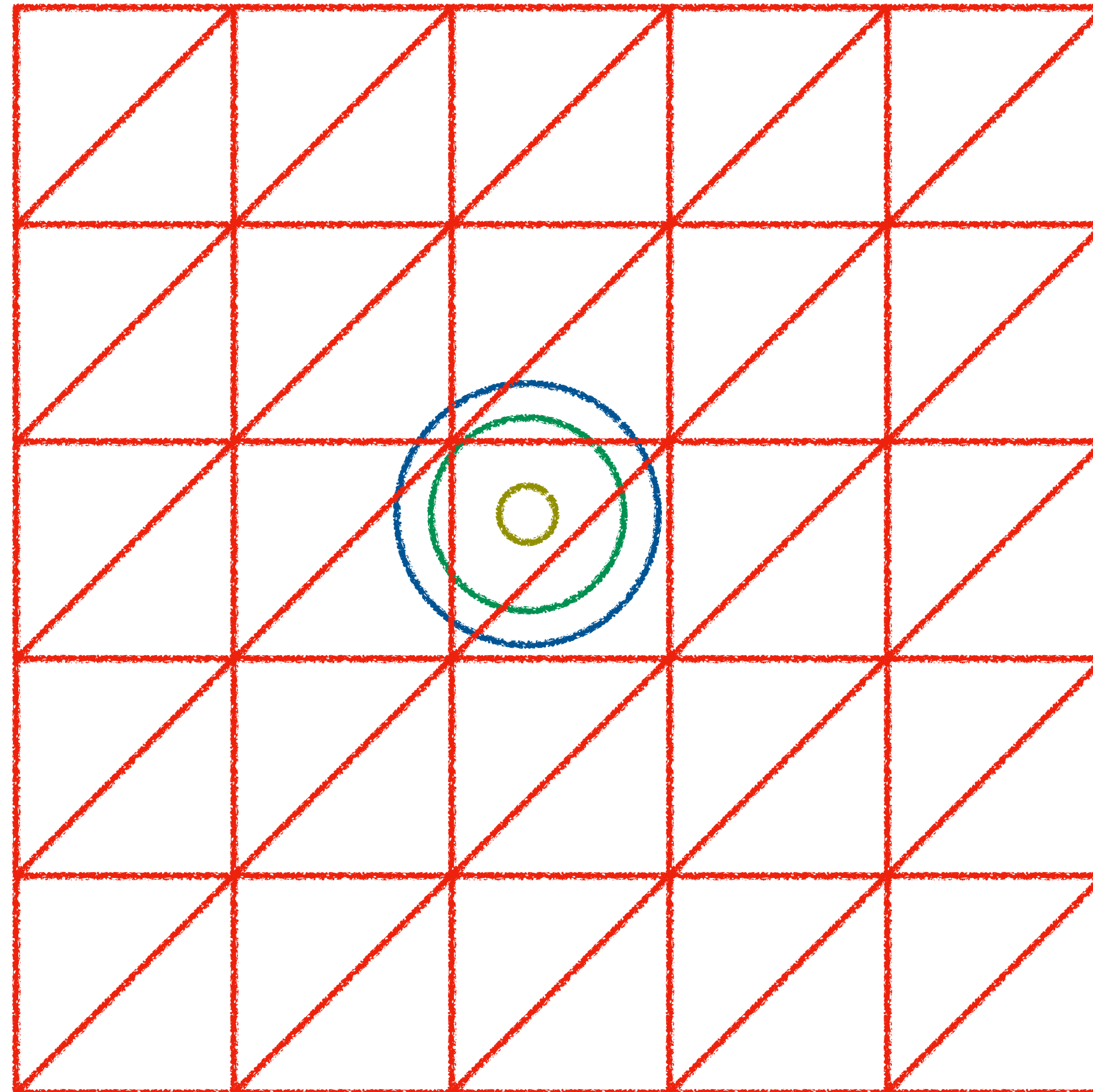
71 ns * n_elements * n_agents

Cycle 3

```
elem,ip = ElementFinder.Octree(agent->pos);  
value = mfem::GridFunction.GetValue(elem, ip);
```

Runtimes:

Static agents: 1 micro second
Moving agents: 8 micro seconds



Cycle2

```
If agent->HasElement():  
    bdm::Continuum.VerifyElement(agent);  
If not agent->HasElement():  
    elem,ip = mfem::Mesh.FindPoints(agent->pos);  
    agent->SaveElement(elem);  
    value = mfem::GridFunction.GetValue(elem, ip);
```

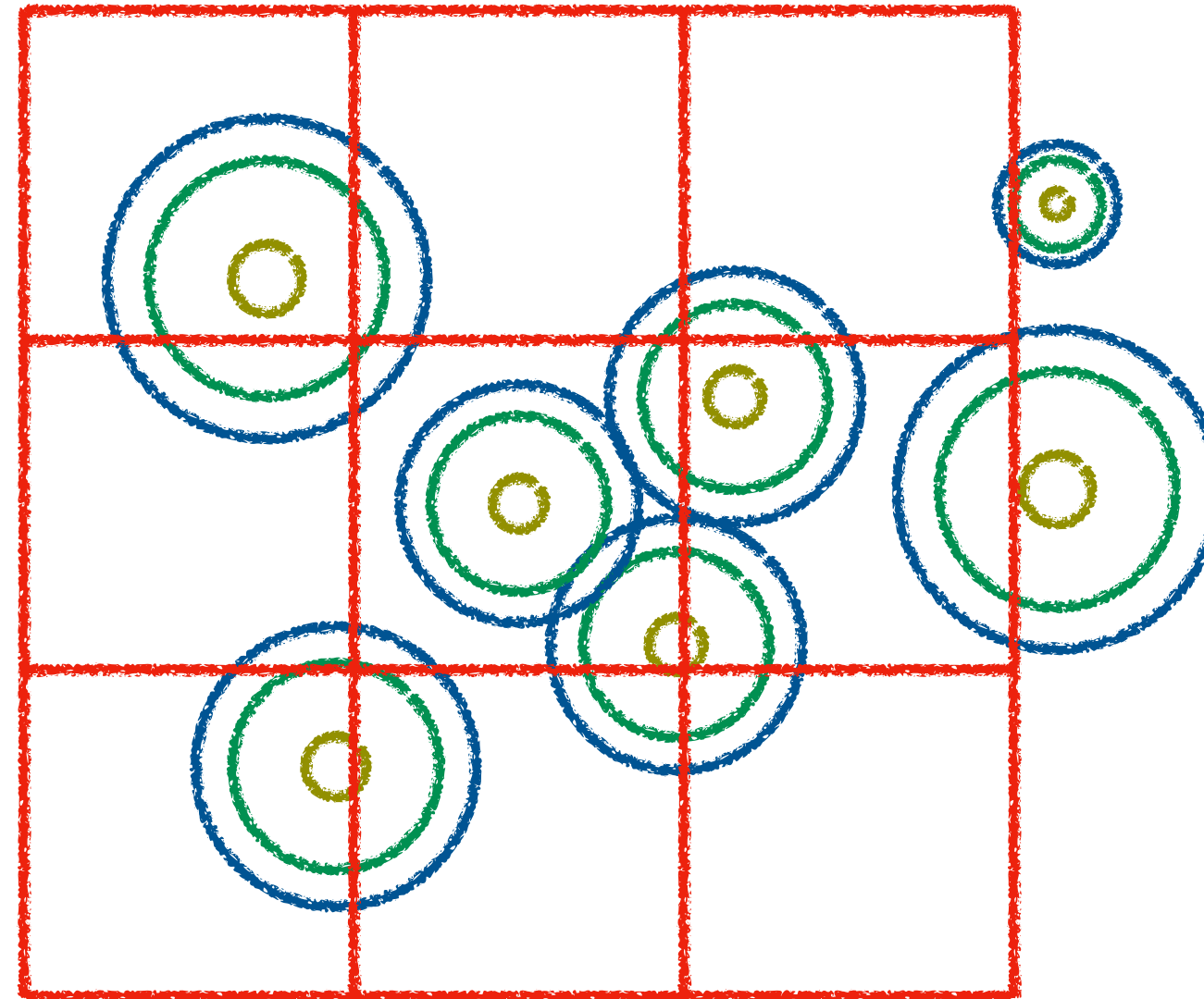
Runtimes:

Static agent:
1 micro second
Moving agent:
55 micro seconds

Coupling - Retrieving agent information

Idea 1

- 1) Use uniform grid for efficient computation of a density function.
- 2) Pseudo Code:
sum = 0;
auto treat_neighbor = [&](agent* a){
 sum += Gaussian3D(x_in - a->pos);
};
ForAllNeighbors(treat_neighbor);
// + Normalisation



Idea 2

- 1) Use uniform grid for efficient computation of a density function.
- 2) Pseudo Code:
sum = 0;
auto treat_neighbor = [&](agent* a){
 sum += 1;
};
ForAllNeighbors(treat_neighbor);
// + Normalisation

Performance considerations

- How expensive is the treatment of ABMs and the PDE?
- Are all mesh resolutions a good choice?
- What kind of backends are available? BDM uses OMP at the moment.
- Possibly advanced splitting with MPI?

