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HPC Training Series

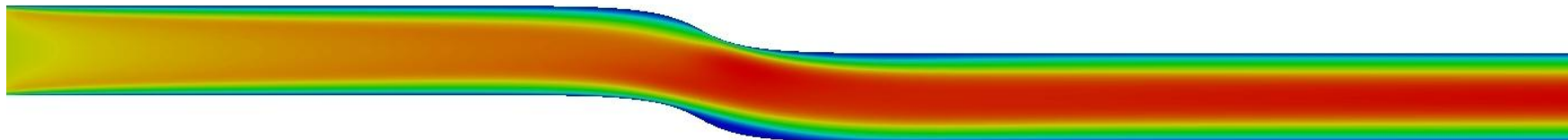
Setting up and running an optimization case in OpenFOAM.

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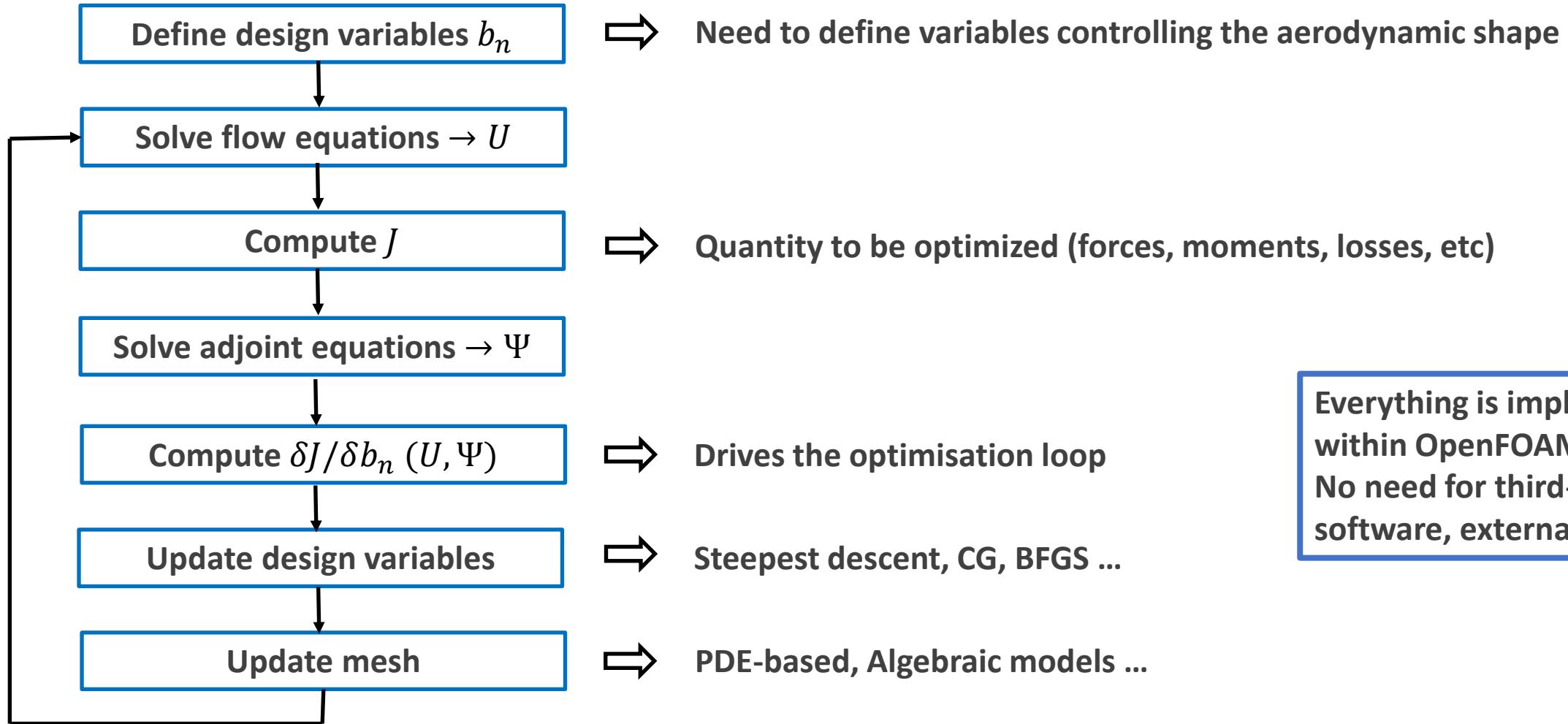
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The tutorial case

- Case is derived from `$FOAM_TUTORIALS/incompressible/adjointOptimisationFoam/shapeOptimisation/sbend/laminar/opt/unconstrained/BFGS/` but with a smaller mesh to get results faster
- Laminar flow within an S-bend 2D duct, mesh is provided
- $Re = 1000$
- Objective: minimize volume-weighted total pressure losses $J = - \int_{S_{I,O}} \left(p + \frac{1}{2} v_k^2 \right) v_i n_i dS$

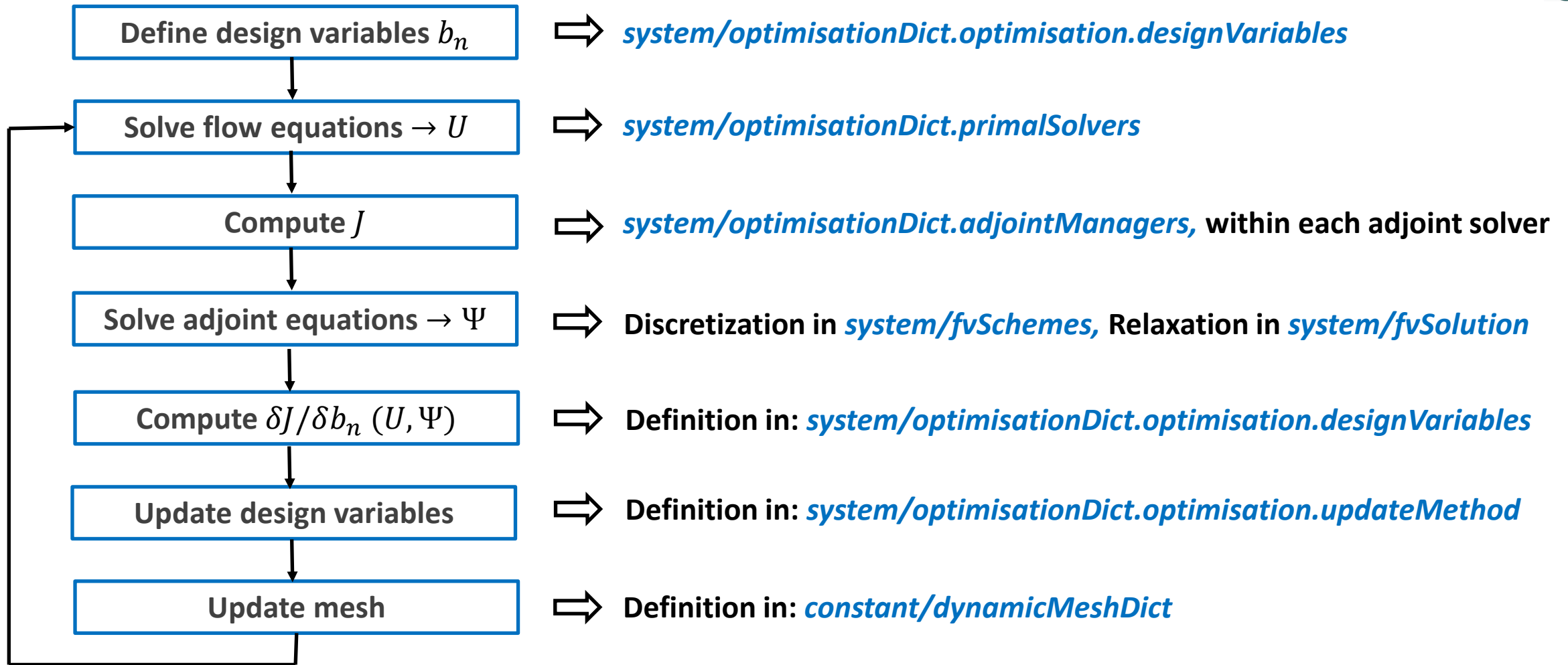


Gradient-based Shape Optimisation Loop

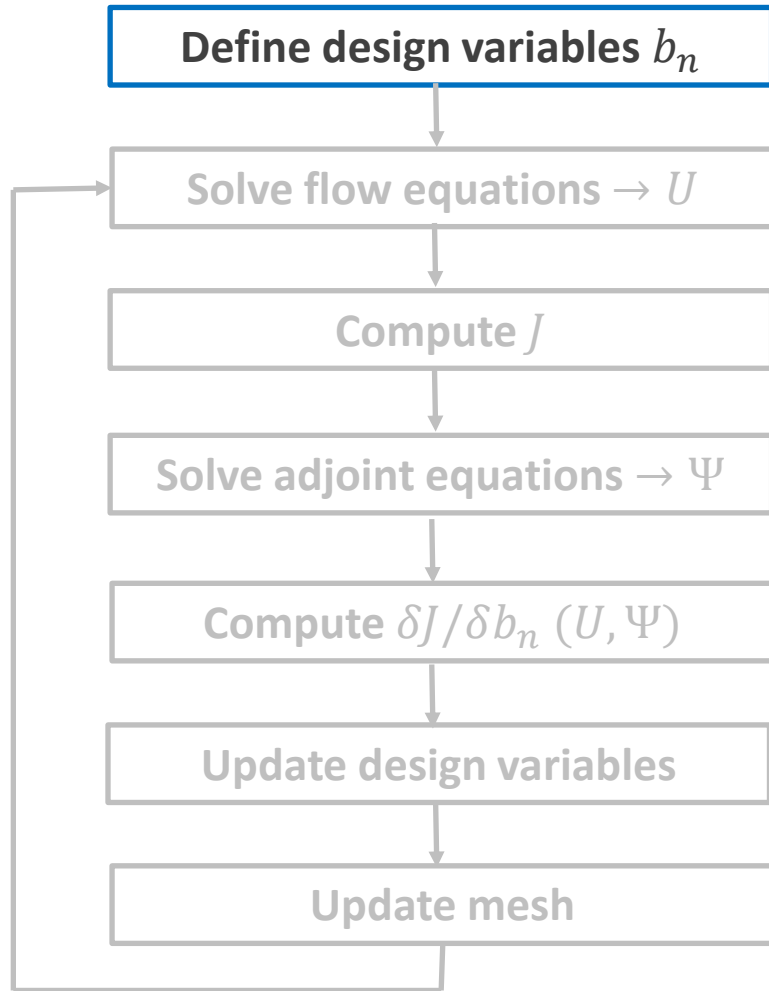


Everything is implemented within OpenFOAM:
No need for third-party software, external scripts, etc

Gradient-based Shape Optimisation Loop

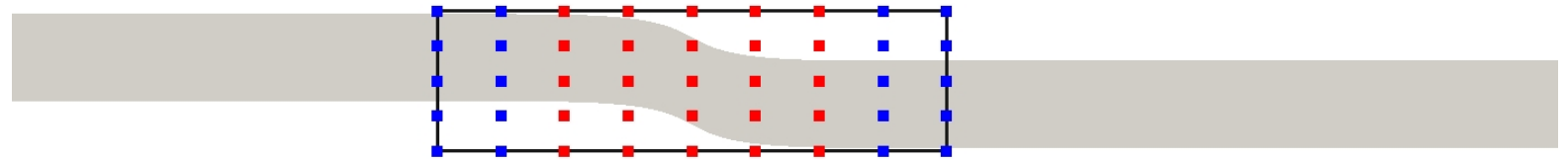


Gradient-based Shape optimisation Loop



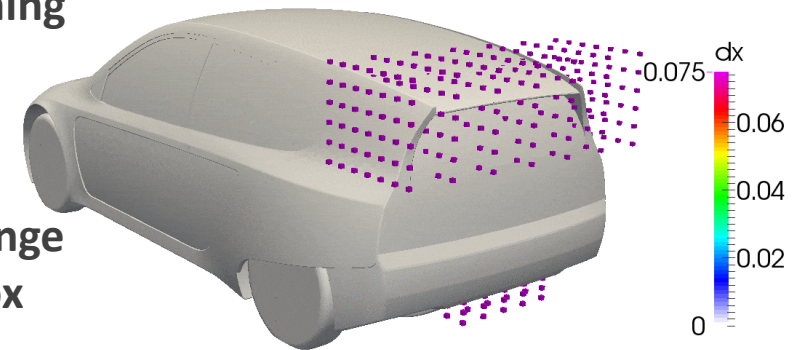
Parameterization for shape optimisation:

- NURBS Curves (2D) and Surfaces (3D)
- All of the wall nodes
- **Volumetric B-Splines (Free Form Deformation, FFD)**



Volumetric B-Splines:

- Maps all CFD grid points within the morphing boxes from the Cartesian to a parametric space $(x, y, z) \rightarrow (u, v, w)$
- Mapping has to be done only once
- Then, changing the control points will change all CFD grid nodes within the morphing box (boundary and internal)
- Update is done through an algebraic relation: very fast!



Gradient-based Shape optimisation Loop

Define design variables b_n

Defined in: *constant/dynamicMeshDict*

```

solver volumetricBSplinesMotionSolver;

volumetricBSplinesMotionSolverCoeffs
{
  duct
  {
    type cartesian;
    nCPsU 9;
    nCPsV 5;
    nCPsW 3;
    degreeU 3; // max: nCPsU - 1
    degreeV 3; // max: nCPsV - 1
    degreeW 2; // max: nCPsW - 1

    controlPointsDefinition axisAligned;
    lowerCpBounds (-1.1 -0.21 -0.05);
    upperCpBounds ( 1.1 0.39 0.15);

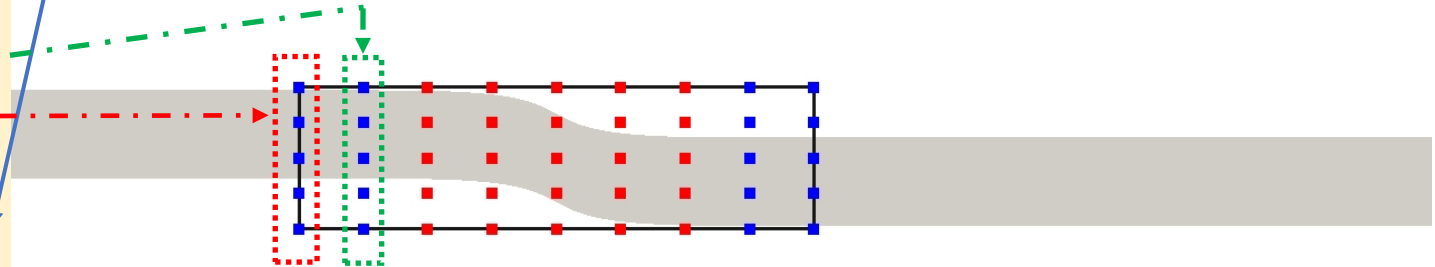
    confineUMovement false;
    confineVMovement false;
    confineWMovement true;
    confineBoundaryControlPoints false;

    confineUMinCps ((true true true) (true true true));
    confineUMaxCps ( (true true true) (true true true) );
    confineWMinCps ( (true true true) );
    confineWMaxCps ( (true true true) );
  }
}

```

Basic entries:

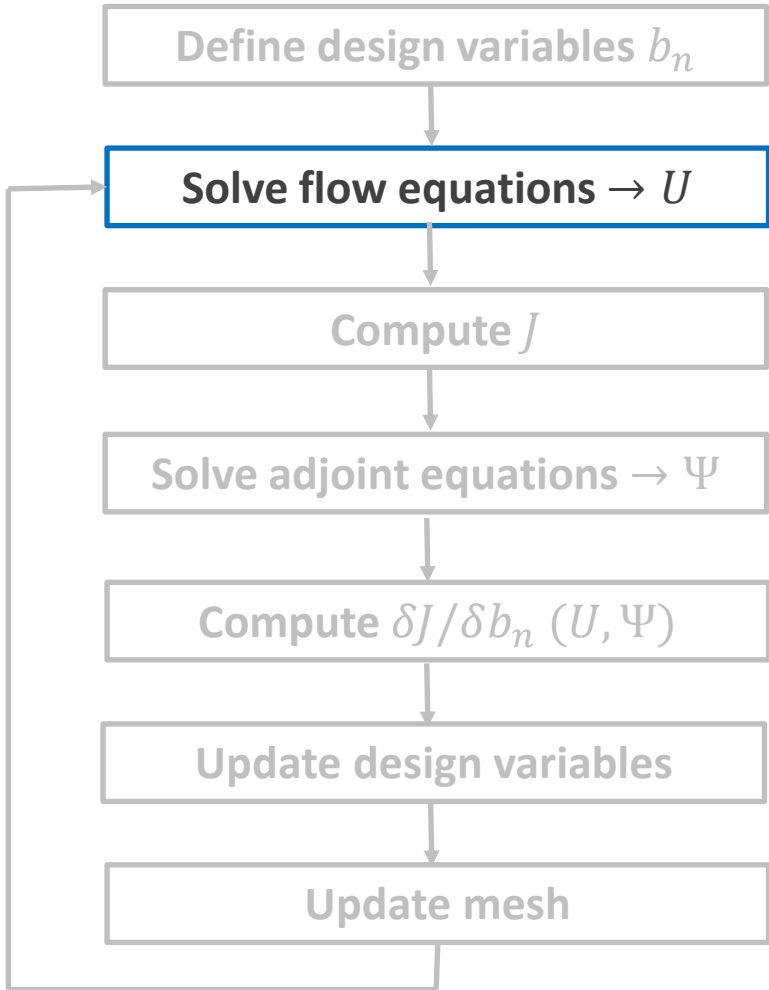
- Number of control points (CPs) per box direction
- Degree per direction (smaller degree → more local support)
- CPs defined either aligned with a coordinate system (Cartesian, cylindrical) or given manually through a dictionary
- Possible to confine the movement of (some of) the CPs in certain directions
- Continuity with the stationary part of the mesh must be preserved! Keeping the boundary CPs constant



>> writeMorpherCps

Writes the control points in a Paraview-readable format

Gradient-based Shape optimisation Loop



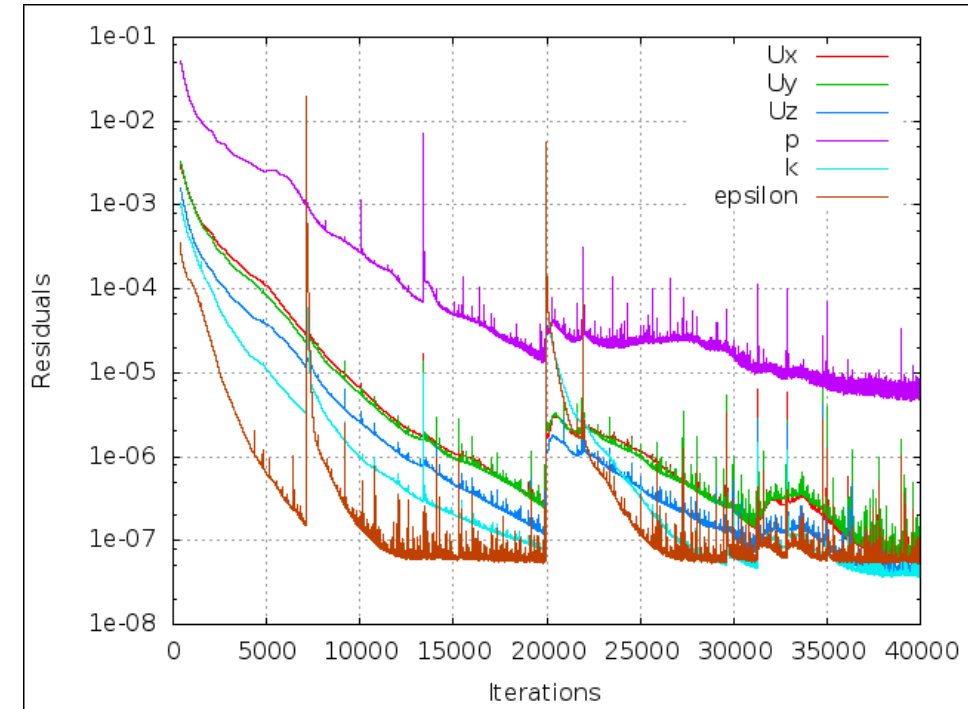
Defined in `system/optimisationDict.primalSolvers`

Incompressible, steady-state flows

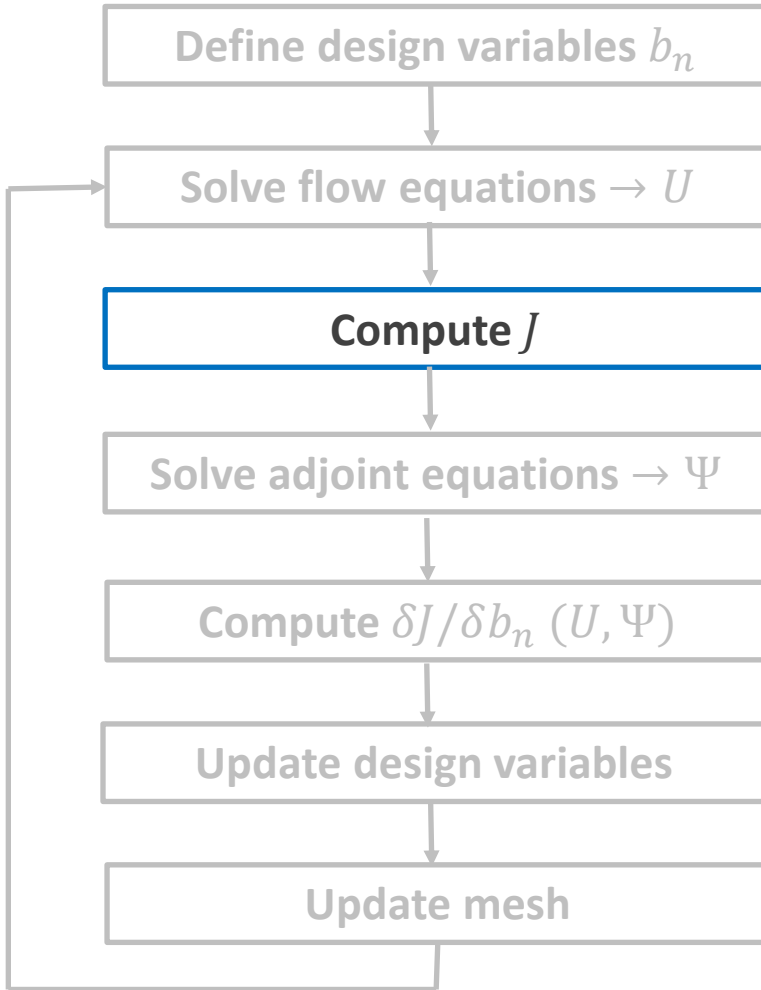
- SIMPLE is incorporated into `adjointOptimisationFoam`
- Multi-point optimisation supported; can define more than one primal solvers

Desired for optimisation, if possible

- Well converged solution (e.g. residuals of $\sim 1.e-05$, $1.e-06$)
- Non-oscillating residuals



Gradient-based Shape optimisation Loop



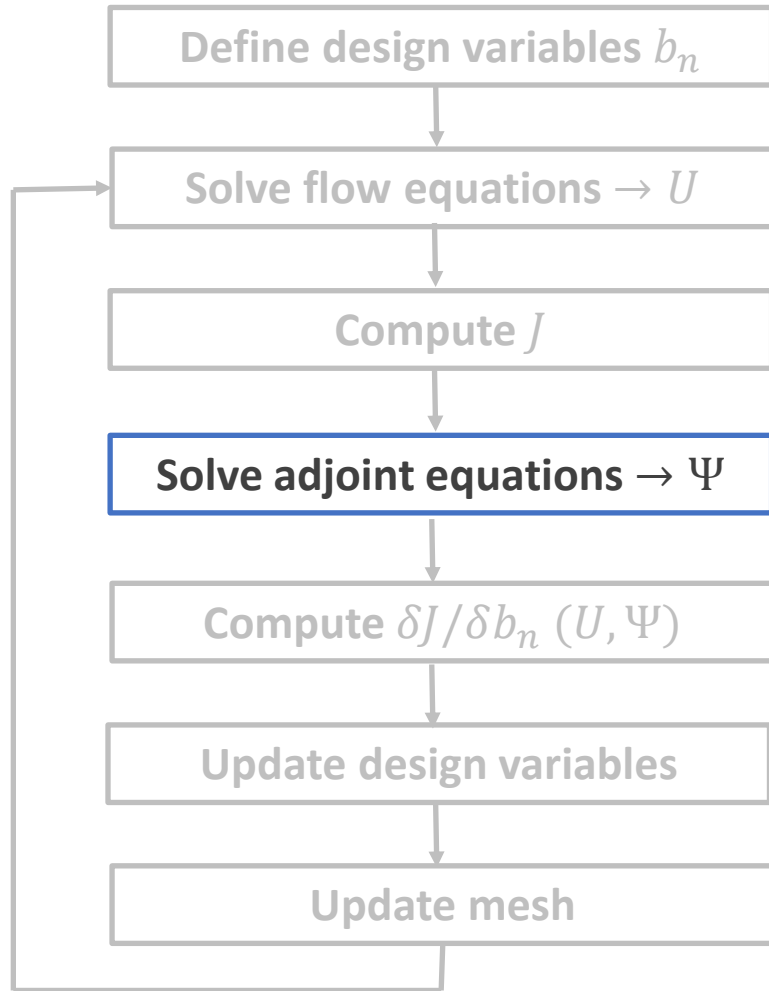
Defined in system/optimisationDict.adjointManagers, within each adjoint solver

Quantity to be optimised

- *adjointOptimisationFoam* always assumes minimization
- Objectives can be defined as (surface or volume) integral quantities
- A number of objective functions are available: Forces, moments, total pressure losses etc ...
- Multiple objective functions can be tackled by concatenating them into a single one using appropriate weights

$$J = w_1 J_1 + w_2 J_2$$

Gradient-based Shape optimisation Loop



Discretization in system/fvSchemes, Relaxation in system/fvSolution

$$R^q = -\frac{\partial u_j}{\partial x_j} + \frac{\partial J_{\Omega'}}{\partial p} = 0$$

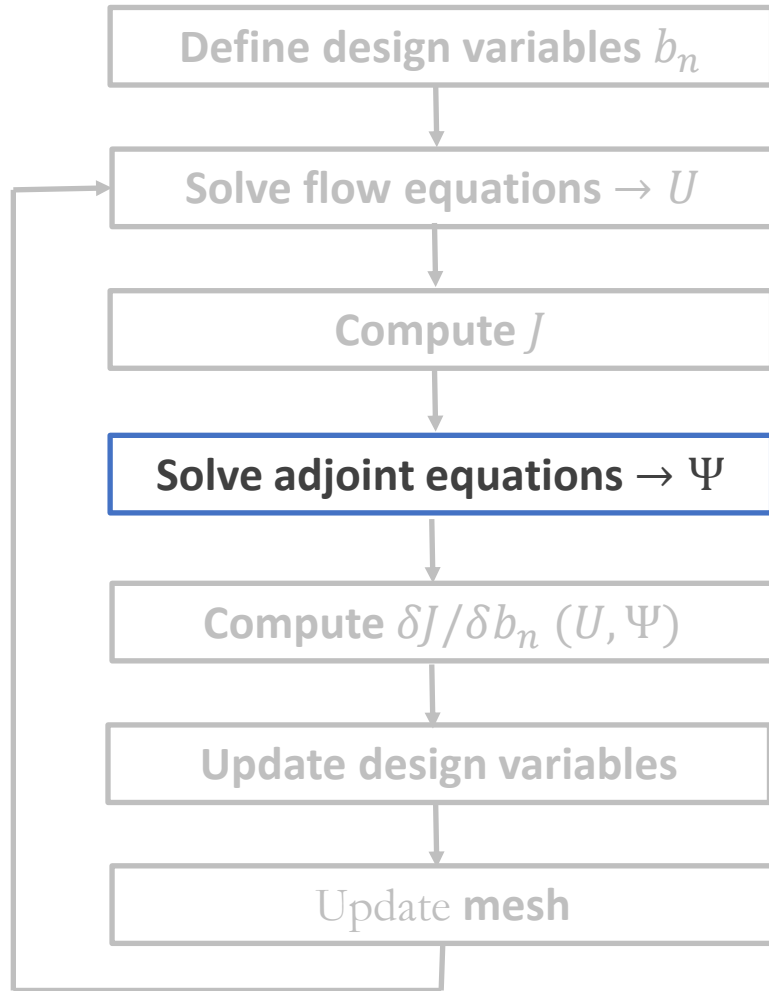
$$R_i^u = \underbrace{u_j \frac{\partial v_j}{\partial x_i}}_{\text{ATC}} - \underbrace{\frac{\partial (u_i v_j)}{\partial x_j}}_{\text{AC}} - \frac{\partial \tau_{ij}^a}{\partial x_j} + \frac{\partial q}{\partial x_i} + \frac{\partial J_{\Omega'}}{\partial v_i} = 0, \quad i = 1, 2, 3$$

Adjoint PDEs (laminar flows):

- Similar form with the Navier-Stokes equations. A few noticeable differences
- Adjoint convection (AC): adjoint velocity is convected by the (minus) primal velocity. Linear equations!
- Adjoint Transpose Convection (ATC): Non-conservative term. Numerically tricky in real-world applications.
- Source terms if the objective function includes volume integrals containing p or v_i

Additional terms and equations when dealing with turbulent flows

Gradient-based Shape optimisation Loop



Defined in 0/pa and 0/Ua

$$u_{\langle n \rangle} = u_j n_j = - \frac{\partial J_{S_{I-W},i}}{\partial p} n_i$$

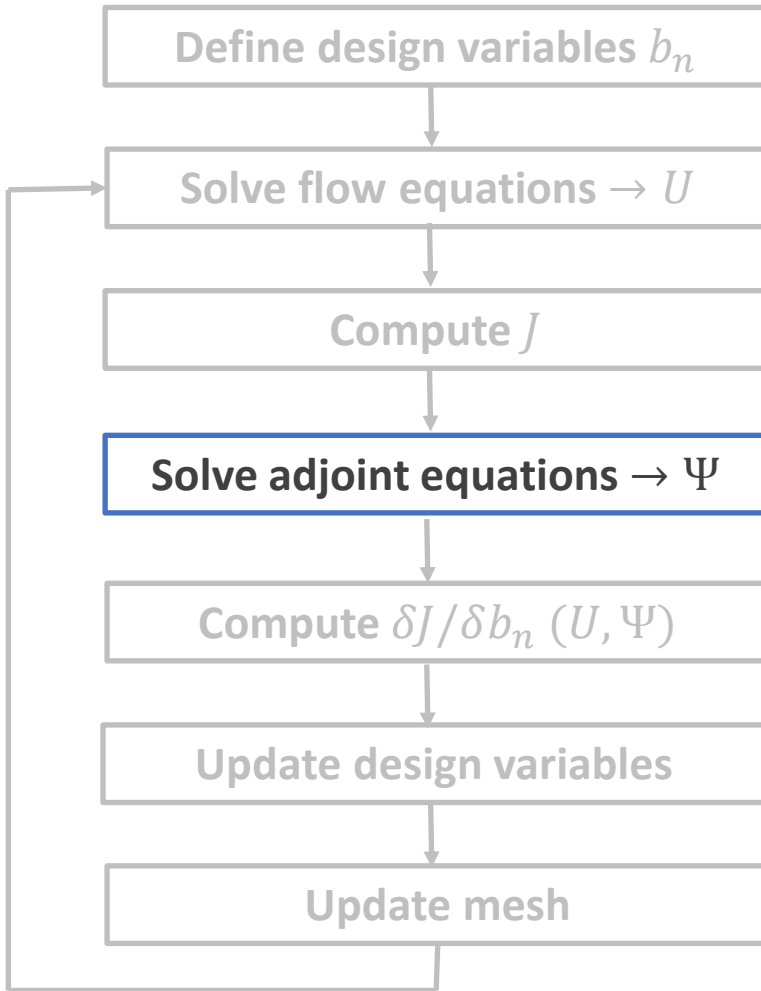
$$u_{\langle t \rangle}^I = u_i t_i^I = \frac{\partial J_{S_{I-W},k}}{\partial \tau_{ij}} n_k t_i^I n_j + \frac{\partial J_{S_{I-W},k}}{\partial \tau_{ij}} n_k t_j^I n_i$$

$$u_{\langle t \rangle}^{II} = u_i t_i^{II} = \frac{\partial J_{S_{I-W},k}}{\partial \tau_{ij}} n_k t_i^{II} n_j + \frac{\partial J_{S_{I-W},k}}{\partial \tau_{ij}} n_k t_j^{II} n_i$$

Adjoint Boundary conditions:

- Depend on the type (**not value!**) of primal boundary conditions!
- Most common for incompressible flows: Dirichlet Inlet \vec{v} , Dirichlet Outlet p
- Depend on the derivatives of J w.r.t. the pressure, velocity and stress tensor

Gradient-based Shape optimisation Loop

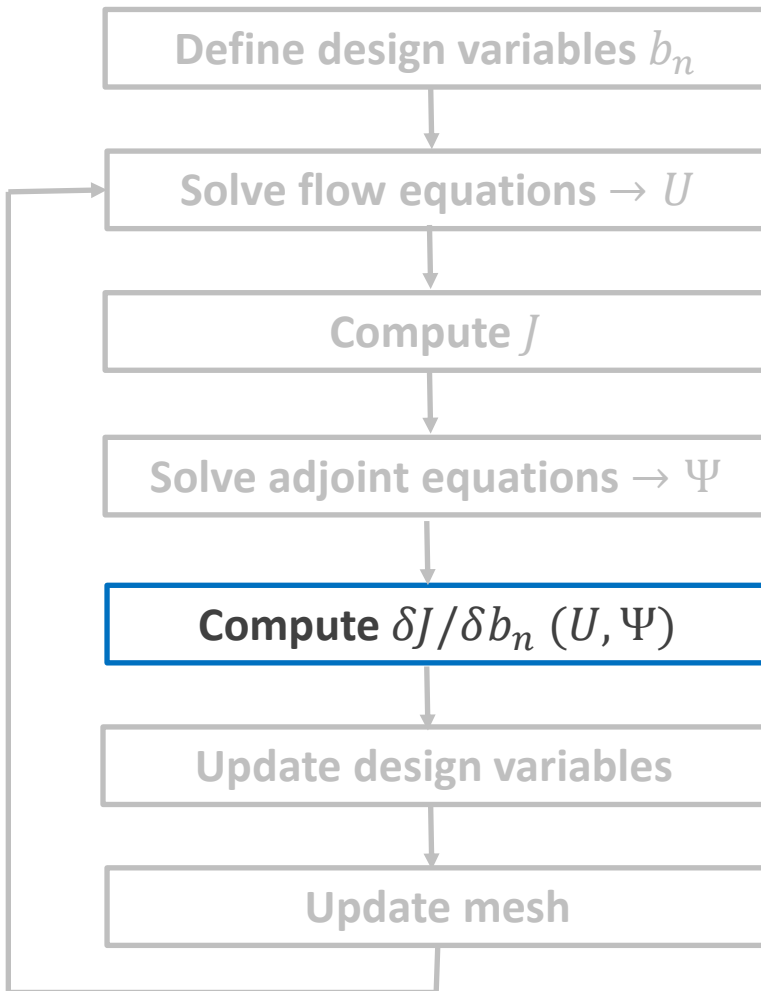


Defined in `system/optimisationDict.adjointManagers`

How many adjoint equations do we have to solve?

- One for each objective for which we need the gradient
 - Gradients of linear combinations of functions defined at a single operating point can be computed with one adjoint solution!
 - Advanced methods dealing with constraints (e.g. SQP, constraint projection) need the gradient of the constraint function separately
- (At least) One for each operating point solved

Gradient-based Shape optimisation Loop



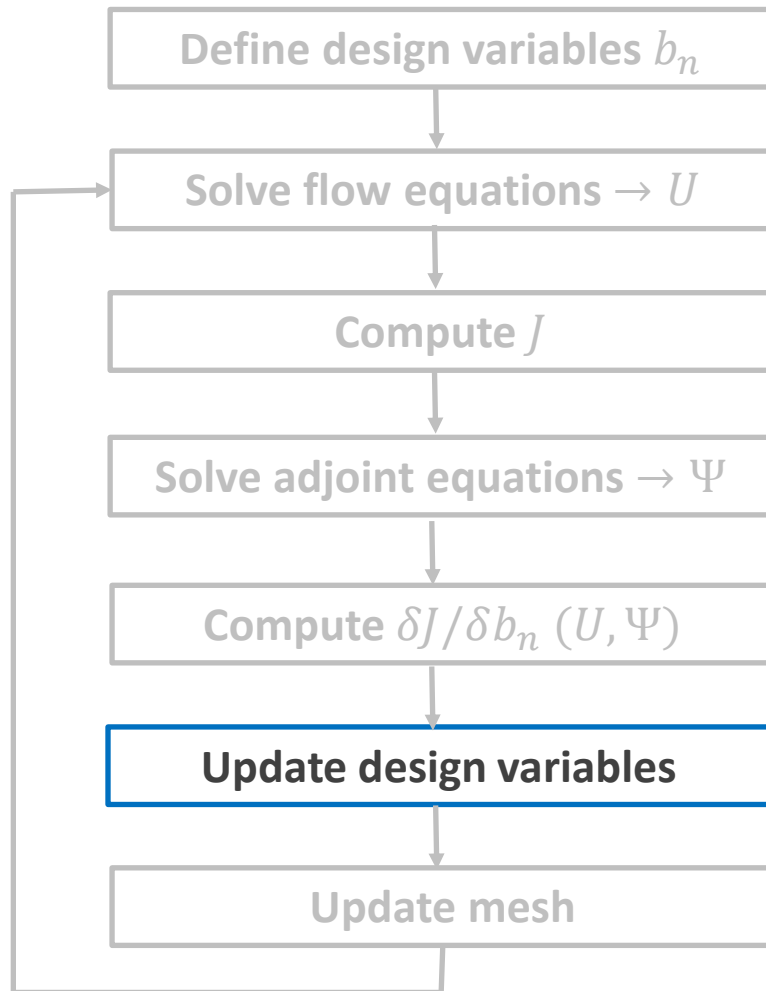
Defined in `system/optimisationDict.optimisation.designVariables`

Two mathematical formulations for shape optimisation

- Based on Surface Integrals, (E)-SI
 - Need to solve an additional adjoint grid displacement PDE for m_i^a
 - Boundary conditions are created automatically
 - Need to define a linear solver in `fvSolution`
 - No relaxation is required
 - Solved at a post-processing level, i.e. after the solution of the adjoint mean flow equations

- Based on Field Integrals, FI
 - Need to compute the grid sensitivities fields, i.e. $\frac{\delta x_k}{\delta b_n}$
 - Depending on the grid displacement model this might be computed by
 - solving additional PDEs (e.g. PDE-based grid displacement)
 - Analytically (e.g. Volumetric B-Splines)

Gradient-based Shape optimisation Loop



Defined in

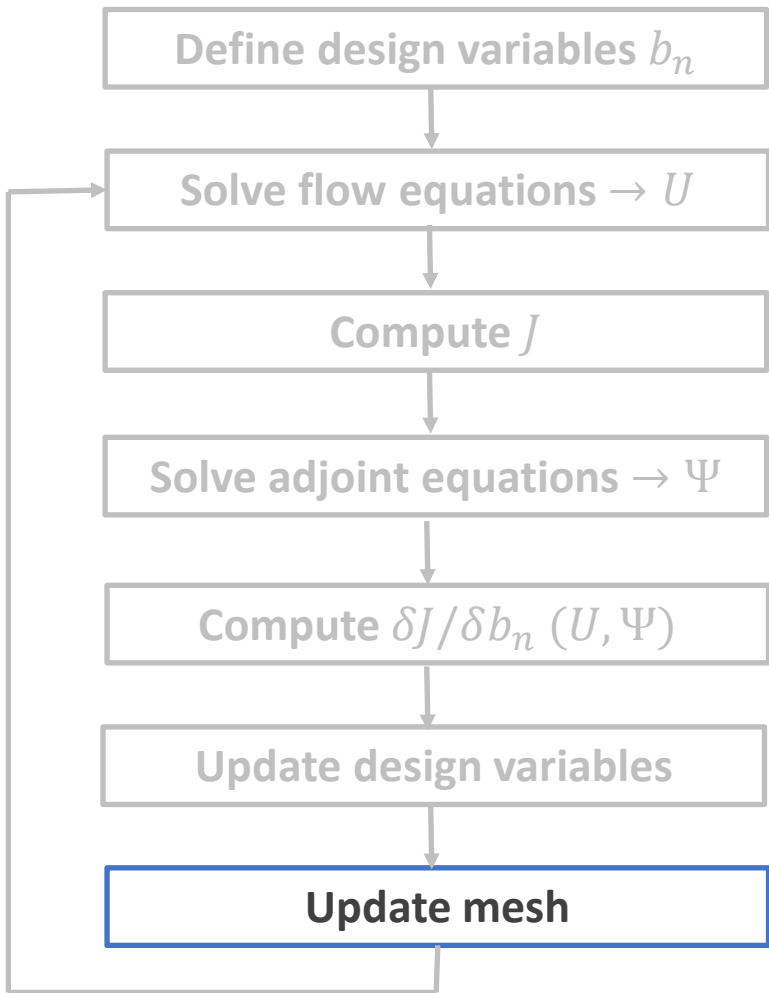
`system/optimisationDict.optimisation.updateMethod`

Compute the update of the design variables based on $\frac{\delta J}{\delta b_n}$ through

$$b_n^{new} = b_n^{old} + \eta s_n$$

- **Unconstrained optimisation**
 - Steepest descent
 - Conjugate Gradient
 - Quasi-Newton methods: BFGS, SR1
- **Constrained optimisation**
 - Constraint projection (exceptional for linear constraints)
 - SQP
- **Step (η) definition**
 - Direct (usually not practical)
 - Through a max. desired deformation in the initial opt. cycle

Gradient-based Shape optimisation Loop



Defined in
constant/dynamicMeshDict

- Need to translate Δb_n into a new geometry and computational mesh
- Remeshing can be costly and possibly result to inconsistent sensitivity derivatives. Grid displacement is preferable
- Depends on the parameterization and chosen grid displacement method
 - Usually, one tool for parameterization (e.g. NURBS), a different one for grid displacement (e.g. Laplace PDEs)
 - Volumetric B-Splines handles both simultaneously
- checkMesh ran after each update to check mesh quality

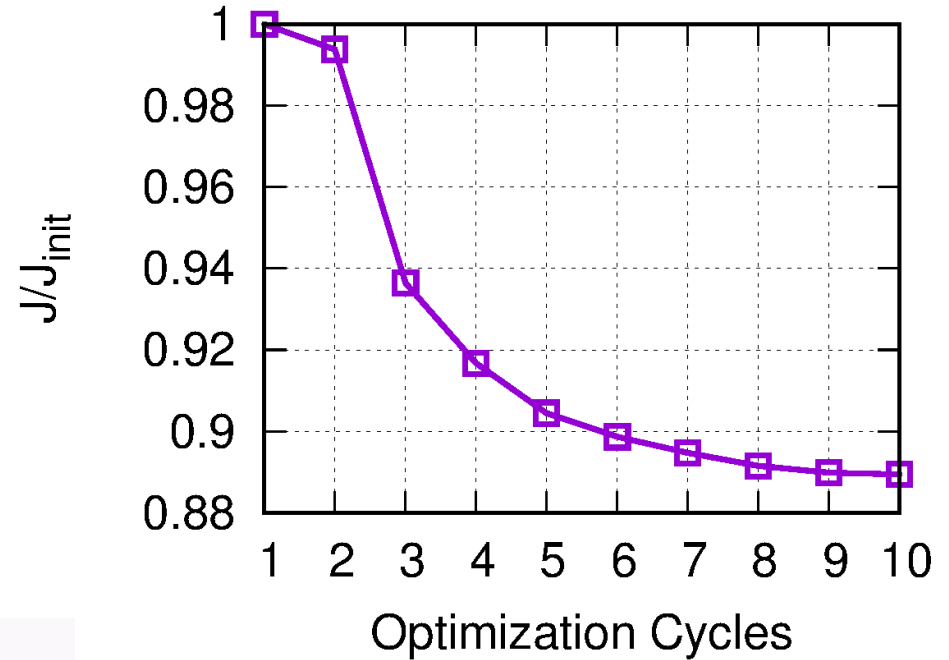
S-bend: optimisation results

Run the optimization loop

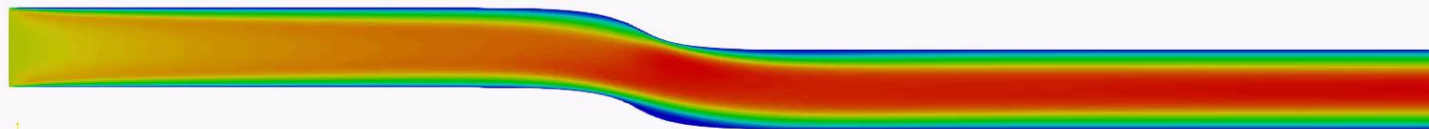
```
>> ./Allrun.parallel > log 2> err & (~2.5 min/4 procs)
```

What to examine:

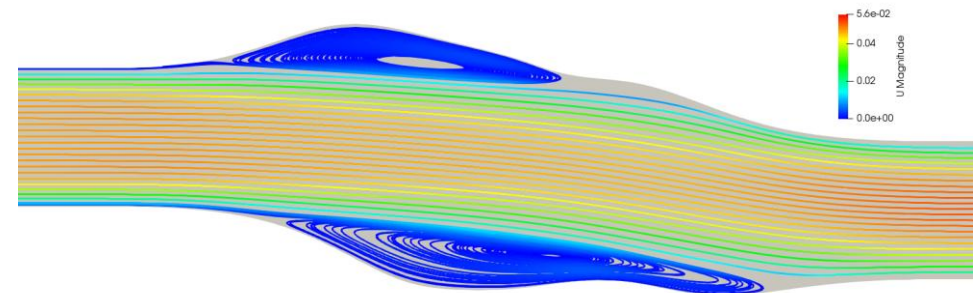
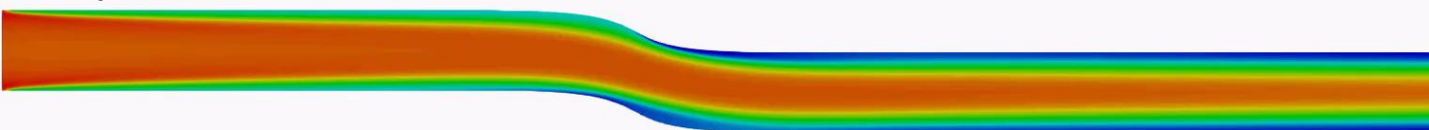
- Is J reduced?
- Is J converged? (history in *optimisation/objective* folder)
- Have the flow equations converged? (check log file)
- Is the mesh valid at the optimised solutions? (check log file or checkMesh)
- What is the mechanism behind the reduction in J ?
- **Don't be afraid of exotic solutions!**



$|\vec{v}|$

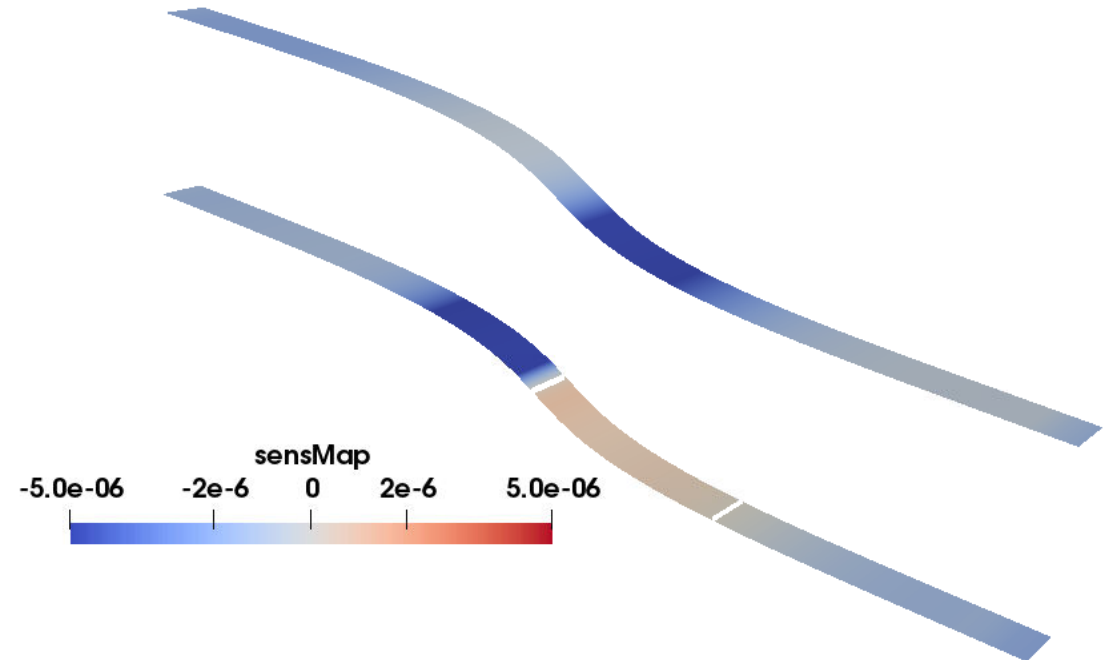


p_t



S-bend: Computing sensitivity maps

- Compute $\frac{\delta J}{\delta x_i} n_i$
- A few changes in *optimisationDict* and *controlDict*
- Tells us how each boundary node has to move to reduce J
 - **Red**: move against the surface normal (inwards)
 - **Blue**: move towards the surface normal (outwards)
 - **White-ish**: insignificant
- Computed on the initial geometry: does not mean that the optimised geometry will follow this ! ...
- Good feedback towards the designer
- Useful in placing morhing boxes

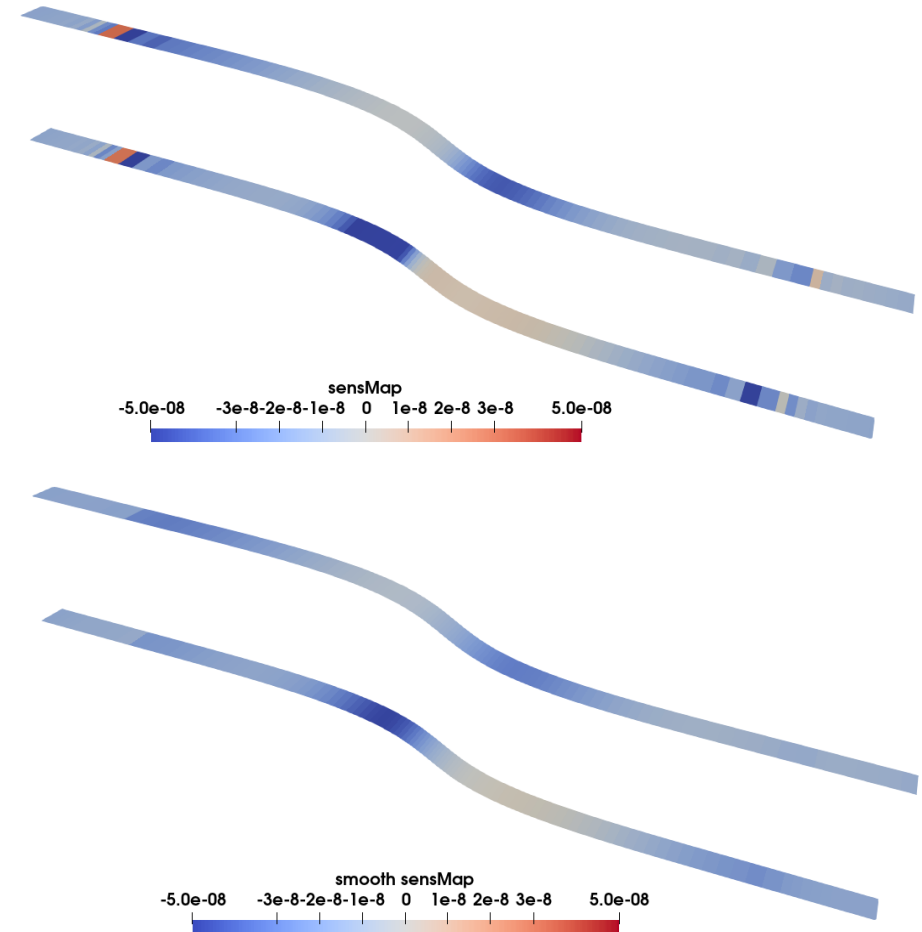


S-bend: Smoothing the sensitivity map

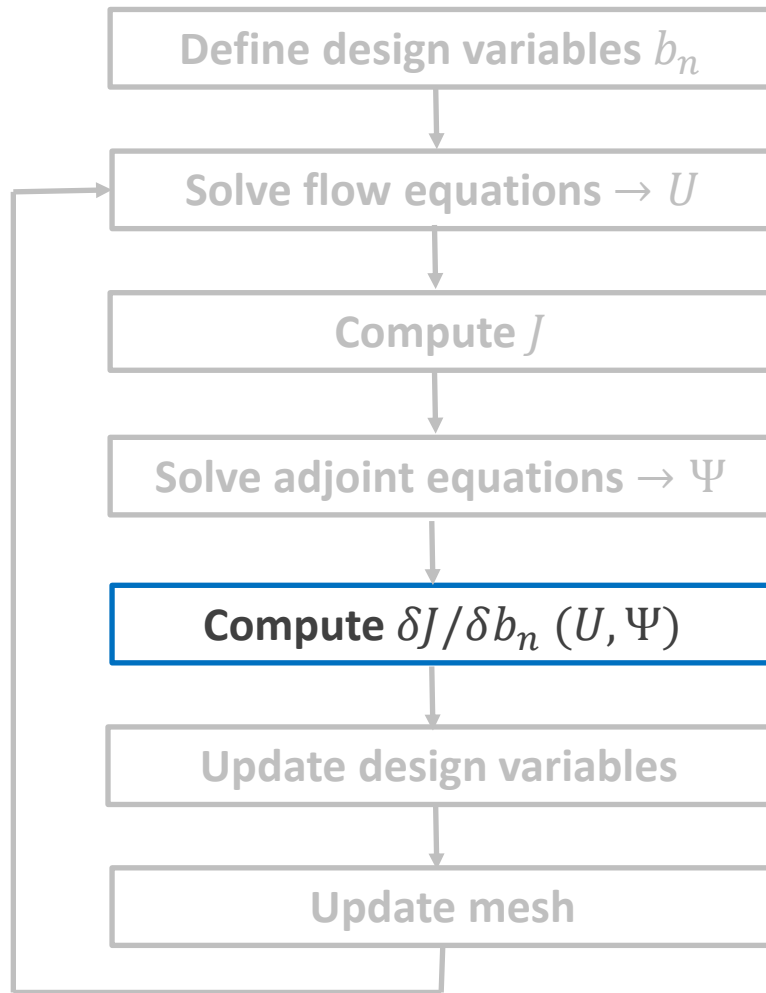
- In more complex/industrial cases, checkerboards occur in the computed sensitivity maps.
- This problem becomes pronounced in meshes built with snappyHexMesh!
- The direction of favorable surface displacement becomes ambiguous...
- Smooth the sensitivity-map, G , by solving

$$-R^2 \frac{\partial^2 \hat{G}}{\partial x_j^2} + \hat{G} = G$$

on a *finiteArea* mesh.



S-bend: Smoothing the sensitivity map – Additional Entries



- Additional entries in *system/optimisationDict.optimisation.designVariables* related to the Laplace-Beltrami equation

$$-R^2 \frac{\partial^2 \hat{G}}{\partial x_j^2} + \hat{G} = G$$

- The smoothing radius is either specified explicitly, or computed as a multiple of the average surface edges' length.
- Boundary conditions for the smooth sensitivity field are created automatically.
- For the creation of the *faMesh*, an *faMeshDefinition* dictionary can be optionally provided in the *system* folder.
- *faSchemes* & *faSolution* should be present in the *system* directory.

Takeaway messages:

- **Adjoint supports optimisation loops at a small CPU cost (~ 20 cycles $\rightarrow \sim 40$ flow solutions)**
- **Ideal for both early-stage development and refinement**
- **More optimisation types available**
 - **Active flow control (jet-based optimisation)**
 - **A Posteriori Error Analysis (optimally refine your mesh to compute an accurate objective)**
 - **Design under uncertainties**
- **Optimisation (like CFD) is not magic. Take care when defining your problem**
- **Before accepting (or discarding) an optimised geometry**
 - **Check the convergence of the flow equations**
 - **Check the mesh quality**
- **Try to understand the mechanisms behind the objective reduction**
 - **Often leads to better designs and/or better-defined optimisation problems!**

Additional topics covered through the tutorials under [\\$FOAM_TUTORIALS/incompressible/adjointOptimisationFoam](#)

- Effect of the update method
[shapeOptimisation/sbend/laminar/opt/unconstrained](#)
- Constrained optimisation
[shapeOptimisation/naca0012/lift/opt/constraintProjection](#)
- 3D, industrial-like cases
[shapeOptimisation/motorbike](#)



When in doubt about the case settings, you can consult the manual