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**RESEARCH
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Abstract: We present a method to simulate the flow past bioinspired swimmers starting from pictures of an actual fish. The overall approach requires i) a skeleton graph generation to get a level-set function from pictures; ii) optimal transportation to obtain the velocity on the body surface; iii) flow simulations realized thanks to a cartesian method based on penalization. We illustrate this paradigm by simulating the swimming of a mackerel fish.

Key-words: Bioinspired Swimming, Level Set, Skeleton, Optimal transportation, Cartesian mesh, Penalization

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Bioinspired Swimming Simulations

Résumé : Un ensemble de méthodes pour simuler l'écoulement généré par un nageur construit à partir de photographies est présenté. Cette approche nécessite i) une méthode de squelettisation pour générer les fonctions level-set ; ii) le transport optimal pour obtenir la vitesse de déformation entre deux level set successives ; iii) les simulations numériques réalisées sur maillages cartésiens avec des méthodes level set et de pénalisation. Les résultats sont illustrés pour la nage d'un poisson de type maquereau.

Mots-clés : Nage bio-inspirée, fonctions level set, squelettisation, transport optimal, maillage Cartésien, pénalisation

1 Introduction

Autonomous swimming robots can address in the future needs such as search and rescue, environmental monitoring and emergency response in rapidly unfolding scenarios. Fishes have the ability to evolve in complex and unpredictable environments that may abruptly change. They are capable of maneuvering in ways that can hardly be achieved by present engineering devices. In this sense, the design of such devices can significantly benefit from bioinspired principles.

However, quantitative evaluation of the geometrical and dynamical characteristics of swimming in experiments is challenging [33, 37]. In particular, precise control of the kinematics and dynamics of the experimental model is difficult; harder still is the accurate measurement of forces and power of a freely swimming model. In this regard, numerical models are well suited for such investigation.

Three-dimensional numerical simulation of swimming has only become viable in the last decade [23, 30, 12, 17, 16, 7] thanks to the progress of both numerical schemes and computers architectures. Nonetheless, many of the studies presented in the literature typically deal with prescribed swimmer geometries. Only few start from actual fish geometries, like for example Patankar's group [30, 12, 8] and Mittal's group [23, 9, 14].

The aim of this paper is to model a three-dimensional swimmer starting from actual fish pictures. The first step is to build the three-dimensional fish profile starting from two-dimensional data retrieved from one or several subsequent pictures of a fish at rest. To this end, a skeleton technique [15] is used to reconstruct the three-dimensional level set function describing the body surface. Then the skeleton can be deformed using an appropriate swimming law to obtain a sequence of level set functions corresponding to snapshots of the body surface uniformly taken at different instants. Otherwise, if a sequence of pictures of swimming is available, the body surface can be reconstructed with the same technique for each snapshot. The level set functions give at each instant an implicit description of the surface which is quite different from usual swimming simulations where the deforming body is fully described by in a Lagrangian framework.

Thanks to skeleton deformation or to a sequence of actual pictures we typically can reconstruct only a small portion of the snapshots necessary to simulate a swimming stroke, since the time scale of the simulation is significantly smaller than the time step between two subsequent reconstructed three-dimensional snapshots. Moreover, the surface deformation velocity is required to set the boundary conditions of the flow problem. For this reason it is necessary to build intermediate level set functions and to compute the deformation velocity field between subsequent reconstructed fish snapshots. Optimal transportation [36] is well suited to achieve this goal providing an objective model to compute intermediate geometries and deformation velocities.

Optimal transportation is an old topic first introduced by Monge in 1781 [24]. It consists in a distance minimization problem that allows to define a mapping between an initial and a final mass distribution. This problem has attracted considerable attention these last years especially from a numerical point of view [4, 19, 27, 13, 26]. Here we exploit the scheme introduced in [1] to define a non-linear interpolation mapping between geometries at different instants and to compute the corresponding deformation velocities.

In the following we describe the numerical scheme, the swimmer kinematic modeling and show validations with examples of applications.

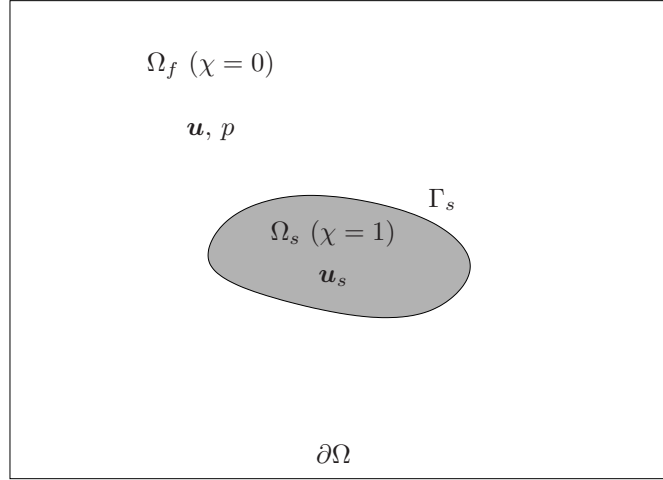


Figure 1: Sketch of the general flow configuration.

2 Numerical modeling

2.1 Flow problem

The flow configuration is given in figure 1 where Ω_s is the domain defined by the body, Ω_f is the domain filled by the fluid, and $\Omega = \Omega_f \cup \Omega_s$ is the whole domain. The boundary between Ω_f and Ω_s is noted Γ_s .

The flow is modeled using the incompressible Navier-Stokes equations written in the fluid domain Ω_f as follows

$$\rho \left(\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right) = -\nabla p + \nabla \cdot 2\mu D(\mathbf{u}) + \rho \mathbf{g} \text{ in } \Omega_f, \quad (1a)$$

$$\nabla \cdot \mathbf{u} = 0 \text{ in } \Omega_f. \quad (1b)$$

where $D(\mathbf{u}) = \frac{\nabla \mathbf{u} + \nabla^T \mathbf{u}}{2}$, \mathbf{u} is the velocity field, p is the pressure field, μ is the dynamic viscosity, ρ is the density and \mathbf{g} is the gravity. Unsteady boundary conditions are imposed on the body boundary

$$\mathbf{u}(\mathbf{x}, t) = \mathbf{u}_s(\mathbf{x}, t) \text{ on } \Gamma_s, \quad (1c)$$

and on the external boundary

$$\mathbf{u}(\mathbf{x}, t) = \mathbf{u}_{ext}(\mathbf{x}, t) \text{ on } \partial\Omega. \quad (1d)$$

The dynamics of the boundary Γ_s that represents the swimmer will be described later.

Classically, system (1) is discretized using a body fitted mesh such that the boundary condition on the body boundary can be easily and accurately taken into account. However, these kind of meshes are difficult to generate in complex unsteady geometries in three-dimensions as large mesh deformations are required when the body is displaced or is changing topology.

In this study we will use a fixed cartesian grid. The body boundary is defined and tracked thanks to a level set function [28] and the boundary condition on Γ_s is taken in account in the Navier-Stokes equation using a penalty term [2].

We thus consider the penalized Navier-Stokes equations in the whole domain Ω (fluid and body):

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = -\frac{1}{\rho} \nabla p + \frac{1}{\rho} \nabla \cdot 2\mu D(\mathbf{u}) + \mathbf{g} + \chi \lambda (\mathbf{u}_s - \mathbf{u}) \quad \text{in } \Omega, \quad (2a)$$

$$\nabla \cdot \mathbf{u} = 0 \quad \text{in } \Omega, \quad (2b)$$

where boundary conditions on the body boundary are implicitly imposed through the penalty term $\chi \lambda (\mathbf{u}_s - \mathbf{u})$. In the penalty term χ is the characteristic function ($\chi(\mathbf{x}) = 1$ if $\mathbf{x} \in \Omega_s$ and $\chi(\mathbf{x}) = 0$ elsewhere) and λ is a large penalty parameter (we chose $\lambda = 10^8$). In what follows, the characteristic function will be computed from the signed distance function φ , so that $\chi = H(\varphi)$ where H denotes the Heaviside function. A second order penalization method can easily be obtained thanks to the signed distance function [5].

It has been proven [38] that the solution of the system (2) tends to the solution of the system (1) as $1/\sqrt{\lambda}$.

The system (2) is solved in time using a projection method [11, 32] and is discretized in space on a uniform cartesian mesh using second order centered finite differences for the diffusive term and third-order upwind finite differences for the convective term.

The resolution of system (2) requires knowledge of both the unsteady characteristic function χ and the body velocity \mathbf{u}_s . The following sections will explain how to compute these quantities.

For a self propelled body the body velocity can be decomposed such that

$$\mathbf{u}_s(\mathbf{x}, t) = \bar{\mathbf{u}}(\mathbf{x}, t) + \mathbf{u}^\theta(\mathbf{x}, t) + \tilde{\mathbf{u}}(\mathbf{x}, t) \quad (3)$$

where $\tilde{\mathbf{u}}(\mathbf{x}, t)$, $\bar{\mathbf{u}}(\mathbf{x}, t)$ and $\mathbf{u}^\theta(\mathbf{x}, t)$ are the deformation, the translation and the rotation velocities, respectively. While the deformation velocity is imposed (by muscles), the translation and rotation velocities are the results of the hydrodynamic forces and torques applied by the fluid onto the body. These latter are computed integrating the Newtons laws. Once the body geometry is obtained (or imposed), we can deform the backbone (the midline) following swimming laws observed in nature. Considering that the backbone of the steady body is $0 \leq x \leq \ell$, $y = 0$ and $z = 0$, we can apply a swimming law $y(x, t)$ that remains the backbone length to be constant.

The swimming law observed in nature for most of fishes is [3]:

$$y(x, t) = a(x) \sin(2\pi(x/\lambda + ft)), \quad (4a)$$

where the envelop is

$$a(x) = c_0 + c_1 x + c_2 x^2. \quad (4b)$$

This swimming law is defined with a constant phase velocity $c_p = \lambda f$, where λ and f are the wave length and the frequency of the oscillations. The parameters c_0 , c_1 and c_2 are adjusted to reach a maximal tail amplitude $A/2$ that is an important parameter for the swimming [20]. It has been shown [34, 39] that carangiform swimming mode is close to $c_0 = 0.02$, $c_1 = -0.12$ and $c_2 = 0.2$ for a unit length fish with $0 \leq x \leq 1$. These parameters can be rescaled when modifying the fish length.

3 Generation of level set functions from pictures: the skeleton approach

Thanks to the scheme presented above it was possible to perform numerical simulations of fish-like swimmers [6, 7] or jellyfishes [5]. The swimmer geometries were prescribed to mimic realistic fishes or jellyfishes. The motion of these swimmers was described in a purely Lagrangian way by tracking markers distributed on the swimmer boundary. The objective is now to simulate bioinspired swimmers using geometries and possibly deformation laws obtained by pictures or sequences of pictures. As an example we consider the definition of a swimmer starting from the mackerel fish presented in figure 2 (left column) [18].

As mentioned in §2, our numerical approach requires the knowledge of the level set function (distance function) associated with a body to get a second order scheme in space. A characteristic function (mask function) is sufficient for a first order penalty method [6].

The process is synthesized in figure 2. We start with pictures of a mackerel fish. Top, bottom and side views are necessary. To rebuild fish in 3D from only three pictures (front, side, top), an underlying assumption is made: the whole cross-section of the fish along length is geometrically similar to the fish on front view (front picture). A dilatation takes place in width and another takes place in height. In the tail region of the fish, two connected components can appear on the cross section, each connected component will be geometrically similar to the front view. The scaling factors will be extracted from the thickness of the fish (from the side view) and the width (from the top view), according to the longitudinal position of the transverse cut.

Segmentations are first performed to extract the boundaries (black lines in center column). The distance function (right column) can be computed by solving Eikonal equation. Then the skeleton is computed (red lines in the center column) as being the center points of the inscribed circles. The natural tool to identify this information is the centerline or skeleton of an object as recalled

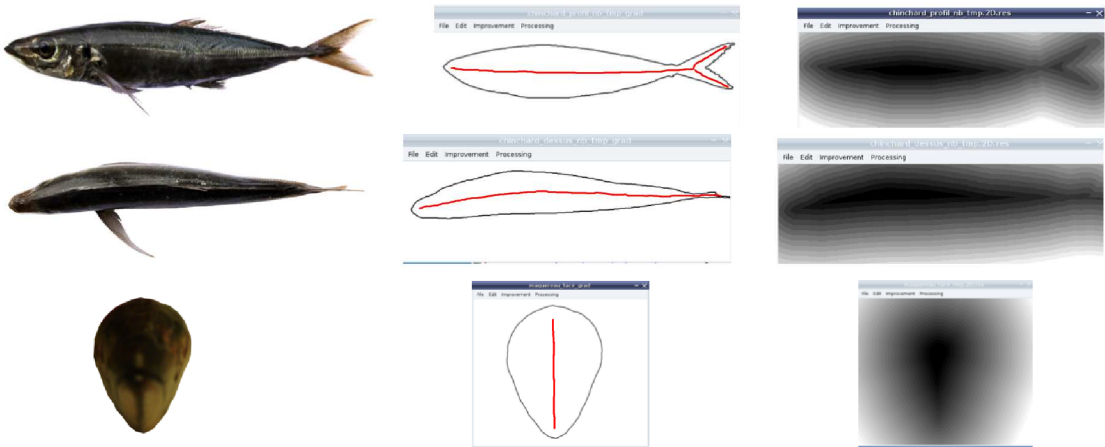


Figure 2: Skeleton steps: pictures (left), segmentation with skeleton (center) and associated level set functions (right).

in [15]. It identifies the locus of points equidistant from further points of the boundary. The special feature here is that the relevant skeleton is not constructed from the Euclidean distance. In order to identify the thicknesses and the widths of the fish, the distance is measured in a fixed direction (depending on the height or along the width) of the side view or top view. The skeleton

is then simplified compared to the skeleton associated with the Euclidean distance. Thus, the top view will have a skeleton made of a line from the head to the tail of the fish, with a single point of the skeleton on each transverse cut. The skeleton of the profile view for the vertical distance will consist of a central line splitting in the fish's tail. Finally, for each transverse cut of the profile, the number of connected components of the profile view is the number (1 or 2) of points of the skeleton of this cut. This property is true thanks to the choice of the distance and is not true with Euclidian distance. For a given cut, let us denote a a point of the skeleton of the profile view and let b be the point of the skeleton of the top view. Denote d_a the vertical distance from a to the boundary of the fish and d_b the horizontal distance from b to the boundary of the fish. The connected component of the cut fish associated to (a, b) can be rebuilt by a scaling (based on factors d_a and d_b) of the front view and translated according to the position (a, b) . It also becomes possible to mirror the top view of the fish (or front view) by rectifying the skeleton in a line to correct the defects of shooting fish in the picture, see figure 3. Initial rescaling of



Figure 3: Re-centering (top view).

pictures has been performed so that the fish length is the same for top and side views, and height is the same for top and side views. A simple 3D reconstruction of the fish could then take place, but the goal is to deform the fish, which will have the effect of losing the notion of directional distance, more complex work is to be implemented. First, the skeleton of the top view (for the vertical distance) S_v is distorted by law (ref eq: swimlaw). the concept of directional distance is no longer suitable for reconstruction, then the chosen approach is to distort the classic skeleton associated with Euclidean distance, relying on distorted S_v . This new skeleton allows to reconstruct, for the Euclidean distance, the deformed fish seen from top view. Then, the 3D reconstruction of the fish can be implemented by harnessing both concepts skeleton as shown in figure 4 (right).

It is thus possible to generate a series of level set functions corresponding to different time

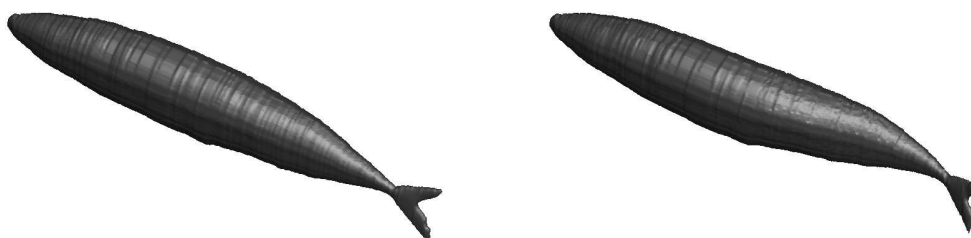


Figure 4: Three-dimensional profiles. Undeformed (left) and deformed (right).

t for the swimming law (4). In what follows, we considered the periodical swimming law (4) and uniformly extract N snapshots for the level set function over one period (one swimming stroke). Finally, the level set functions have been modified (transported) with a rigid velocity so

that we do not add any linear or angular translations during deformation. Linear and angular translations have only to be the results of the hydrodynamic forces and torques.

Note that in what follows, the level set are chosen to be the sign distance functions (positive inside the body).

4 Computation of intermediate level-sets and deformation velocities: optimal transportation

As mentioned, we have access to N (here $N = 20$) snapshots $\{\varphi^k\}_{k=0,\dots,N-1}$ of the level set function over a swimming period T . Here, k denotes the time $t = kT i/N$ for the i^{th} swimming period. The number of available snapshots is of course inessential, but it is typically small compared to the number of snapshots necessary for a numerical simulation. It is then necessary to build missing level set function corresponding to the Naiver-Stokes time discretization. Moreover, it is also necessary to compute the fish velocity for the penalty method. This velocity can be decomposed into a deformation velocity, to be computed from the subsequent N level set functions, and a rigid velocity computed from hydrodynamics effects. The deformation velocity is derived from an optimal transportation model.

Optimal transportation is a model to compute a mapping between two densities. The model assumes that the mass is transported along straight lines with constant initial velocity. The initial velocity is found such that to overall distance measured in a given norm is minimum. The formulation of the L^2 optimal transportation problem is illustrated with respect to generic densities ϱ_0 and ϱ_1 . The main step is to compute the velocity field transforming ϱ_0 into ϱ_1 . Then it is possible to compute any desired intermediate densities for time $t = rT i/N$ with $0 < r < 1$.

Let $\varrho_0(\xi), \varrho_1(x)$ be two smooth enough non-negative scalar density functions with compact support Ω_0 and Ω_1 , where $\xi, x \in \mathbb{R}^d$ and d is the space dimension. We assume that

$$\int_{\Omega_0} \varrho_0(\xi) d\xi = \int_{\Omega_1} \varrho_1(x) dx.$$

Let $X : \Omega_0 \rightarrow \Omega_1$ a smooth one-to-one map such that $X(\xi)$ realizes the transfer of ϱ_0 onto ϱ_1 , i.e., a map that satisfies the following Jacobian equation:

$$\varrho_0(\xi) = \det(\nabla X(\xi)) \varrho_1(X(\xi)).$$

This equation is under determined with respect to $X(\xi)$ and a solution is selected among all possible maps by introducing the following L^2 Kantorovich-Wasserstein distance:

$$\inf_X \int_{\Omega_0} \varrho_0(\xi) |X(\xi) - \xi|^2 d\xi.$$

The L^2 Monge-Kantorovich problem (MKP) corresponds to finding a map X_o such that this infimum is achieved. It has been proved that this problem admits a unique solution [10, 35, 36], which is the gradient of a.e. convex function $\Psi_o : \Omega_0 \rightarrow \mathbb{R}$:

$$X_o(\xi) = \nabla \Psi_o(\xi).$$

The main idea of solution relies on the assumption that we dispose of a mapping at iteration n that is a perturbation of the optimal mapping $X_o(\xi)$ [1]. We derive a linear perturbation equation that is used to iteratively improve the initial guess. Let us assume that the mapping obtained at iteration n is

$$X^n(\xi) = \nabla_\xi \Psi_o + \nabla_\xi \Psi_\epsilon^n.$$

where the error Ψ_ϵ^n satisfies $\|\Psi_\epsilon^n\|_2 \approx \epsilon$. We define $\varrho_0^n(\xi)$ as the initial density at iteration n that mapped by $X^n(\xi)$ gives the exact final density $\varrho_1(x)$. Then, taking a first-order Taylor expansion, we have

$$\begin{aligned} \varrho_0^n(\xi) &:= \varrho_1(X^n(\xi)) \det(\nabla_\xi X^n(\xi)) \\ &= \underbrace{\varrho_1(X_o) \det(\nabla_\xi X_o)}_{=\varrho_0(\xi)} + \det(\nabla_\xi X_o) \varrho_1(X_o) \text{Tr}((\nabla_\xi X_o)^{-1} \nabla_\xi^2 \Psi_\epsilon^n) \\ &\quad + \det(\nabla_\xi X_o) \nabla_x \varrho_1(X_o) \cdot \nabla_\xi \Psi_\epsilon^n + o(\epsilon), \end{aligned}$$

where Tr denotes the matrix trace operator.

At first order in ϵ , we have

$$\frac{\varrho_0^n(\xi) - \varrho_0(\xi)}{\det(\nabla_\xi X_o)} \approx \varrho_1(X_o) \text{Tr}((\nabla_\xi X_o)^{-1} \nabla_\xi^2 \Psi_\epsilon^n) + \nabla_x \varrho_1(X_o) \cdot \nabla_\xi \Psi_\epsilon^n \quad (5)$$

The mapping update is then found by the iteration

$$X^{n+1} = X^n - \alpha \nabla \Psi_\epsilon^n,$$

that converges to X_o as a geometric series, for $\alpha \in [0, 1]$, in the linearized approximation [21, 27].

If the optimal solution and the approximated mapping are a perturbation of identity, then equation (5) reduces to

$$\varrho_0^n(\xi) - \varrho_0(\xi) \approx \nabla_\xi \cdot (\varrho_1(X^n(\xi)) \nabla_\xi \Psi_\epsilon^n) \quad (6)$$

that is equivalent to a semidiscretization in time of the continuity equation written in Eulerian form.

4.1 Choice of the density functions

Since the level set functions corresponding to the N fish snapshots are not strictly positive, they are not admissible candidates for optimal transportation. A regularized mask function (Heaviside of the level set function) is also not satisfying since the weight of the thin caudal fin in the mapping is very limited and the mapping tends to smear it in comparison to the whole body.

We thus considered Gaussian density functions defined from the level sets $\{\varphi^k\}_{k=0, \dots, N-1}$, with a support that is approximatively equal to $\ell/10$ (ℓ being the length of the fish) with maximum near the interface but inside the body. In particular, we take $\varrho_0 = G(\xi, k) = \exp\left(\frac{(\varphi^k(\xi, t) - \varphi_b)^2}{\gamma}\right) + \epsilon$, with $\gamma = 2 \cdot 10^{-5}$, $\varphi_b = 1.5 \cdot 10^{-3}$ and the regularization parameter $\epsilon = 10^{-3}$. In a similar way, we take $\varrho_1 = G(\xi, k+1)$. Using these definitions of the densities to be transported, each point of the body surface has a comparable weight in the transportation. We denote by G_0 the value of the Gaussian function corresponding to the swimmer surface. The density $G(\xi, k)$ is presented in figure 5 where the external white line is the body boundary G_0 .

4.2 Computation of intermediate images

Starting from the optimal mapping, the computation of intermediate images is straightforward. Let $X(\xi, \tau) = \xi + \tau(X_o - \xi)$ and

$$\varrho(\xi, \tau) = \frac{\varrho_0(\xi)}{\det(\nabla_\xi X(\xi, \tau))}$$

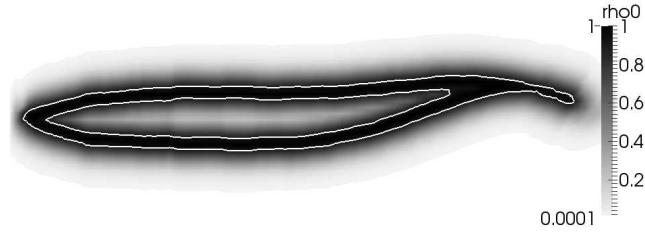


Figure 5: Gaussian density used in the plan $z = 0$. The external white contour corresponds to the fish boundary.

We have $\varrho(\xi, 0) = \varrho_0(\xi)$, $\varrho(\xi, 1) = \varrho_1(X_o(\xi))$ and for $0 < \tau < 1$ we obtain the intermediate images.

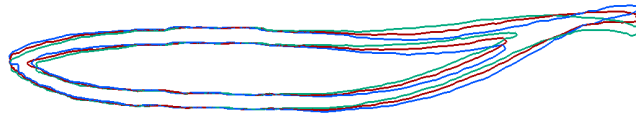


Figure 6: Construction of an image for $\tau = 0.5\Delta t_s$ (red) between ϱ_0 (green) and ϱ_1 (blue) in the plan $z = 0$.

As for the deformation velocity, given the time step Δt_s between two images (here $\Delta t_s = T/20$), the deformation velocity from ϱ_0 to ϱ_1 is

$$\tilde{\mathbf{u}}^0 = X_o/\Delta t_s \quad (7)$$

The index 0 highlights the fact that this velocity has to be mapped to the actual position of the swimmer in space, since the level set snapshots are taken with respect to a center of mass fixed in space. Figure 6 shows the reconstruction the intermediate density for $\tau = 0.5$ between densities ϱ_0 and ϱ_1 .

The same approach can be applied to the whole series of level sets $\{\varphi_i\}_{i=0}^N$, thus computing the corresponding deformation velocities and intermediate densities. Starting from these densities we can compute a characteristic function and a sign distance function thanks to the fast marching approach [29, 28] or by solving an Hamilton Jacobi equation [31, 22]. In what follows, $\varphi^0(\xi, t)$ are the level set function of the deformed profiles with fixed center of mass and $\tilde{\mathbf{u}}^0(\xi, t)$ is the deformation velocity.

5 Dynamics of the swimmer

5.1 Computation of the swimmer velocity and position

Since we are interested in a self propelled swimmer, the position of the swimmer is determined thanks to the level set function $\varphi_s(\xi, t)$ that is an image of the level set function $\varphi^0(\xi, t)$ obtained by the transformation characterizing the rigid motion (translation plus rotation), $X^r(\xi, t)$. We have

$$\varphi_s(\xi, t) = \varphi^0(X^r(\xi, t)).$$

The characteristic function χ required in the penalty method is

$$\chi(\xi, t) = H(\varphi_s(\xi, t) - G_0)$$

see §4.1 for the definition of the constant G_0 . We can then impose $\chi(\xi, t) = 1$ inside the whole body, and compute the signed distance function for second order penalty method [5].

In the same way, the velocity of the swimmer is $\mathbf{u}^0(\xi, t) = \bar{\mathbf{u}}(\xi, t) + \mathbf{u}^\theta(\xi, t) + \tilde{\mathbf{u}}^0(\xi, t)$, where $\tilde{\mathbf{u}}^0(\xi, t)$ is computed from (7). Since the position of the swimmer is modified, the velocity has to act onto the modified position and we have:

$$\mathbf{u}_s(\xi, t) = \mathbf{u}^0(X^r(\xi, t)).$$

The transformation $X^r(\xi, t)$ as well as the translation velocity $\bar{\mathbf{u}}(\xi, t)$ and rotation velocity $\mathbf{u}^\theta(\xi, t)$ are evaluated using the Newton's laws computing the forces and the torques exerted by the fluid onto the body.

5.2 Computation of the forces and the torques

The forces and the torques exerted by the fluid onto the body could be computed integrating the stress tensor $\mathbb{T}(\mathbf{u}, p) = -p\mathbf{I} + \mu(\nabla\mathbf{u} + \nabla\mathbf{u}^T)$ on the body boundary. Let \mathbf{n} be the unit outward normal vector to the interface Γ , and $\mathbf{r} = \mathbf{x} - \mathbf{x}_G$ with \mathbf{x}_G the mass center, the force and the torques write:

$$\mathbf{F} = - \int_{\Gamma_s} \mathbb{T}(\mathbf{u}, p) \mathbf{n} \, d\mathbf{x}, \quad (8a)$$

$$\mathcal{M} = - \int_{\Gamma_s} \mathbf{r} \wedge \mathbb{T}(\mathbf{u}, p) \mathbf{n} \, d\mathbf{x}. \quad (8b)$$

However, since the body boundary is defined in an implicit way as being the zero level of the sign distance function, integration of equations (8) is thus not straightforward. Thus we turn to moment balance as follows.

We consider an arbitrarily subdomain $\tilde{\Omega}_f$ delimited by the boundaries Γ_s and $\tilde{\Gamma}_f$ surrounding the body with domain Ω_s such that the forces and torques write [25]:

$$\begin{aligned} \mathbf{F} = & - \frac{d}{dt} \int_{\tilde{\Omega}_f(t)} \mathbf{u} \, dV + \int_{\tilde{\Gamma}_f(t)} (\mathbb{T} + (\mathbf{u} - \mathbf{u}_f) \otimes \mathbf{u}) \mathbf{n} \, dS \\ & - \int_{\Gamma_s(t)} ((\mathbf{u} - \mathbf{u}_s) \otimes \mathbf{u}) \mathbf{n} \, dS. \end{aligned} \quad (9a)$$

$$\begin{aligned} \mathcal{M} = & - \frac{d}{dt} \int_{\tilde{\Omega}_f(t)} \mathbf{r}_i \wedge \mathbf{u} \, dV + \int_{\tilde{\Gamma}_f(t)} \mathbf{r} \wedge (\mathbb{T} + (\mathbf{u} - \mathbf{u}_f) \otimes \mathbf{u}) \mathbf{n} \, dS \\ & - \int_{\Gamma_s(t)} \mathbf{r} \wedge ((\mathbf{u} - \mathbf{u}_s) \otimes \mathbf{u}) \mathbf{n} \, dS. \end{aligned} \quad (9b)$$

where \mathbf{u}_s and \mathbf{u}_f are respectively the velocities of boundaries Γ_s (thus, the body velocity) and $\tilde{\Gamma}_f$. Note that in many applications (like for swimmers), the integral over the Γ_s boundary vanish due to the fact that we have $\mathbf{u} = \mathbf{u}_s$.

To simplify the computation, it is possible to chose $\tilde{\Omega}_f = \Omega_f$ and thus $\tilde{\Gamma}_f = \partial\Omega$ or smaller regular boxes that fit the mesh grid.

6 Validation and Numerical results

6.1 Validation

The numerical solver has been already intensively validated for several bodies where the geometry is defined by markers and the body followed in a Lagrangian way [6, 5, 7]. Our goal is then to validate the proposed method using optimal transportation comparing it with a well defined test case. All the following numerical simulations are performed using cartesian mesh with $\Delta x = \Delta y = \Delta z = \ell/150$ on 256 CPUs.

In this section we do not consider an actual sequence of fish snapshots. We consider a fish model where the geometry is known and defined by markers. The geometry is however chosen to fit as best as possible the mackerel fish, except the lunate tail. Figure 7 presents the fish geometry for this test case as well as the markers (note that we actually use two times more markers in each directions).

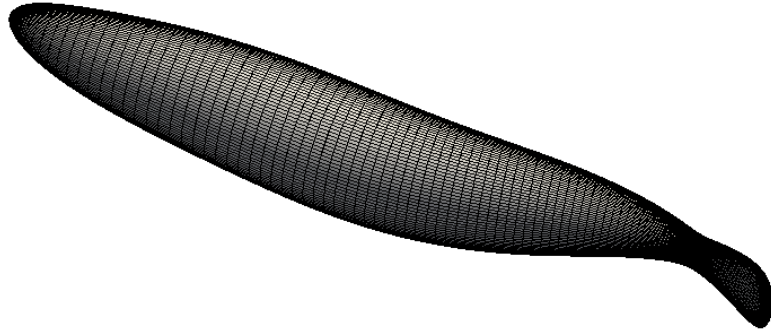


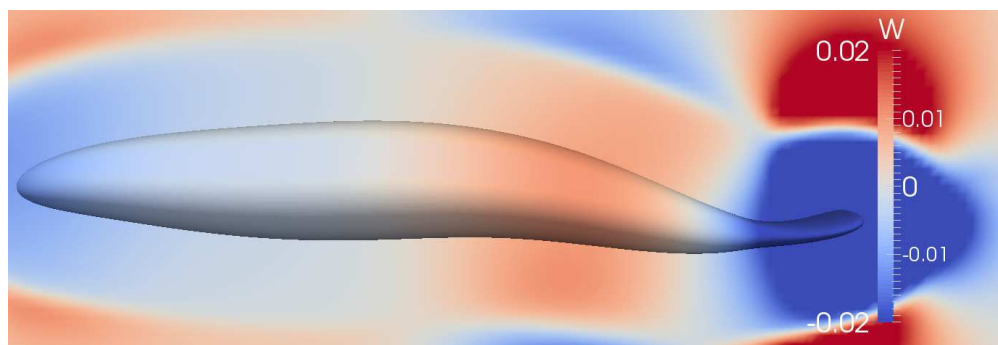
Figure 7: Geometry profile used to validate the method.

Our goal is then to compare the two numerical models for the fish just defined:

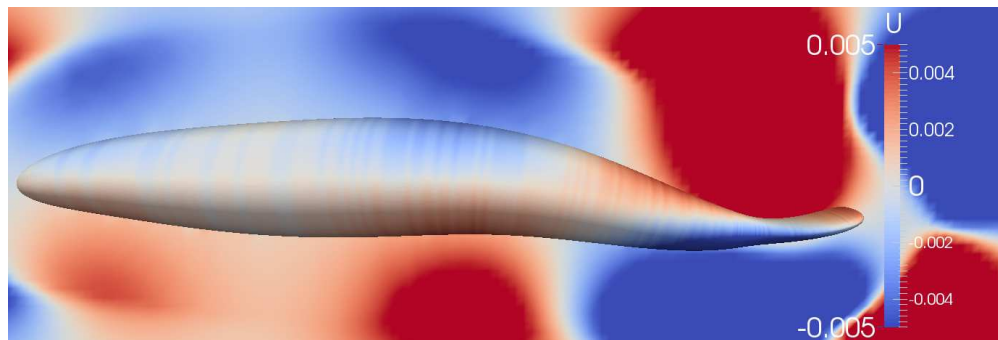
- Case *I* ("Lagrangian"): the level set functions and the deformation velocities are given at each Navier-Stokes time step by following the Lagrangian markers on the body surface;
- Case *II* ("Eulerian"): we suppose that we only know 20 snapshots of the level set function over one stroke period. We have thus to compute all the missing level set as well as the deformation velocity using optimal transportation.

In both cases the length fish is $\ell = 10\text{ cm}$ and the swimming law (4) is defined with frequency $f = 4\text{ Hz}$ and $c_0 = -0.002$, $c_1 = -0.12$, $c_2 = 2$ and $\lambda = \ell$. These parameters are close to

those observed in nature [39]. A comparison of the deformation velocity fields obtained in cases *I* ("exact") and *II* (approximated using optimal transportation) is presented in figure 8. The swimming law is defined in the plan (X, Z) , and we present the velocities U and W where the velocity U is the velocity in the swimming direction. A very good agreement is obtained for the velocity W : the Eulerian deformation velocity field and the Lagrangian one continuously match at the body surface. Some difference are however visible for the U component. Due to the fish configuration, the velocity U can be considered to be the velocity that is tangential to the body. Since optimal transportation is irrotational by definition, it is not able to model any skin rotation. The tangential velocity can not be computed if no other constraints are taken into account, such as done for example in [13, 26] using a rigid motion constraint or a divergence-free constraint.



(a) Velocity in the swimming direction



(b) Velocity normal to the swimming direction

Figure 8: comparison of the deformation velocity fields obtained in cases *I* ("exact", presented inside the fish) and *II* (approximated using optimal transportation, represented outside the fish).

These Lagrangian and the Eulerian velocities are then used to simulate the self propelled swimming in the both cases. These self propelled velocities are presented in figure 9. As expected, the swimming velocity W shows good agreement between cases *I* and *II*. The velocity U is influenced by the differences observed for the deformation velocity U in figure 8. However, considering all the modeling hypothesis, the Eulerian forward swimming velocity correctly approaches the actual Lagrangian forward velocity from which it has been derived.

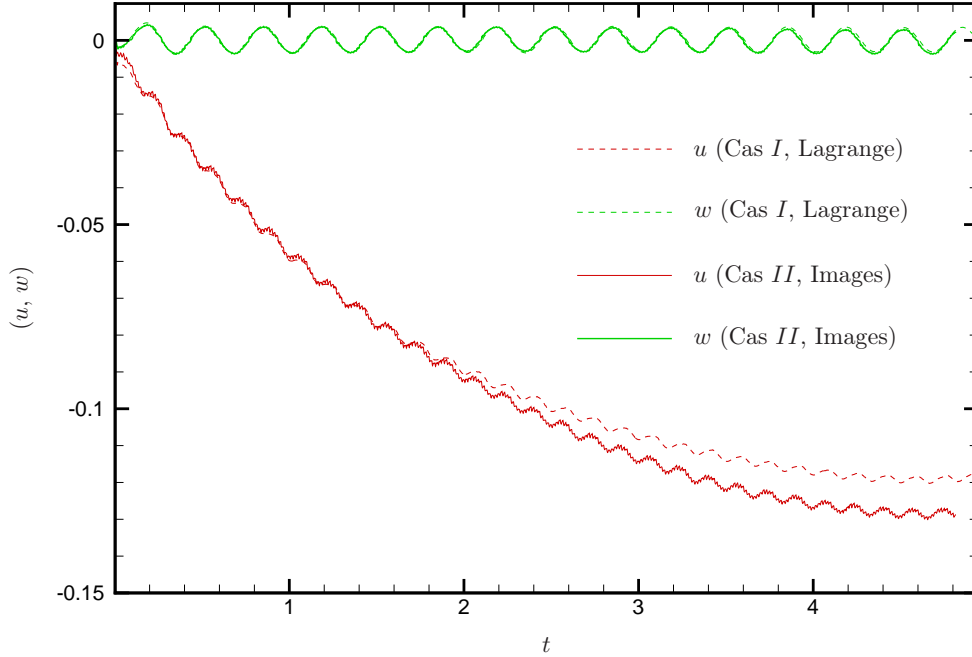


Figure 9: Comparison of the self propelled velocities obtained in cases *I* and *II*.

6.2 Mackerel fish swimming from pictures

We have applied the method described above to recover the geometry of a Mackerel, to deform its skeleton by a swimming law and to simulate swimming. Figure 10 shows the wake flow generating by the fish built from pictures. We chose the same parameters as before for the swimming law, *i.e.* the fish length is 10 cm and the swimming frequency is $f = 4\text{ Hz}$. The swimming velocity of the mackerel fish is slightly higher than the previous test cases ($u \approx -0.15\text{ m/s}$). This difference can be explained by a different shape, and moreover by the lunate tail for the real fish.

7 Conclusions

In this work we have described a procedure to define a bioinspired geometry for a swimmer starting either from one image of a fish or from a sequence of images. Thanks to this procedure it is possible to recover an Eulerian deformation velocity on the base of a limited number of modeling hypothesis, namely those at the base of L^2 optimal transportation. We have shown an *in silico* example where we compare the swimming velocity of a swimmer whose geometry and deformation law is completely determined a priori, to that obtained by the identification method we propose. The results in terms of deformation velocity show that the irrotational hypothesis at the base of optimal transportation induces an error in the tangential velocity of the body. This error coupled to the discretization inaccuracies due to the fully Eulerian treatment of the geometry obtained from Lagrangian data reflects in an overestimation of the actual velocity of the swimmer of less than 10%. As an example of an actual application we have modeled a

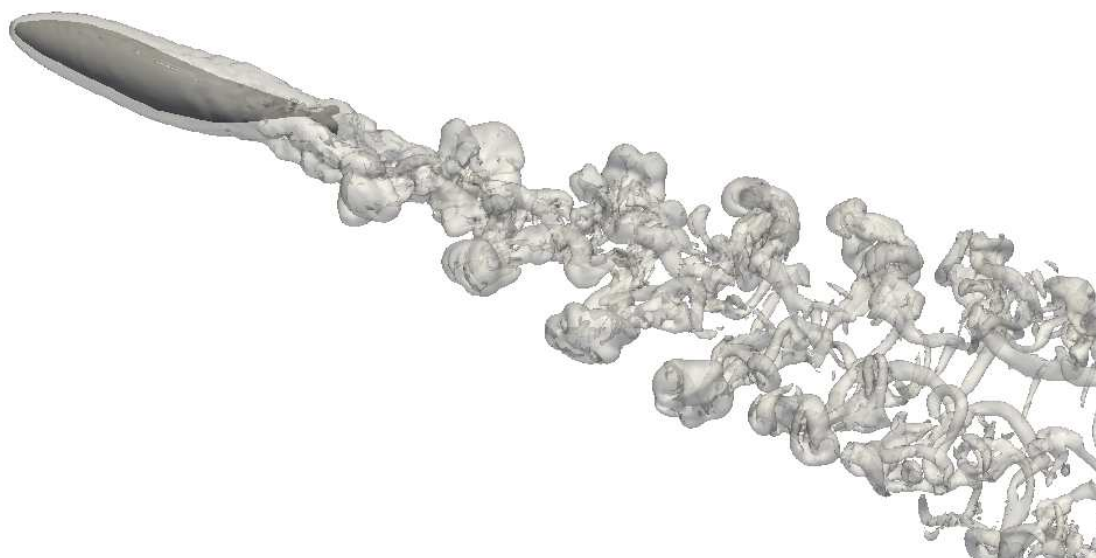


Figure 10: Swimming simulation of the mackerel fish starting from pictures.

swimmer using pictures of a Mackerel. Future improvements of the results may be obtained by constrained optimal transportation.

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