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A numerical algorithm coupling a bifurcating indicator and a direct method for the computation of Hopf bifurcation points in fluid mechanics

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This paper deals with the computation of Hopf bifurcation points in fluid mechanics. This computation is done by coupling a bifurcation indicator proposed recently (Cadou et al., 2006) [1] and a direct method (Jackson, 1987; Jepson, 1981) [2,3] which consists in solving an augmented system whose solutions are Hopf bifurcation points. The bifurcation indicator gives initial critical values (Reynolds number, Strouhal frequency) for the direct method iterations. Some classical numerical examples from fluid mechanics, in two dimensions, are studied to demonstrate the efficiency and the reliability of such an algorithm.

1. Introduction

In fluid mechanics, flow can change when the Reynolds number reaches critical values. In this paper, we focus on the transition between a steady flow to a time-periodic solution. This modification characterizes an instability which is called Hopf's bifurcation. The present study concerns the numerical computations of these bifurcation points. However, to be precise, we do not discuss, in this paper, the stability of the flow after these instabilities. With the numerical methods presented in this paper, it is then impossible to give information on the kind of instability (subcritical or supercritical bifurcation for example) found. From a mathematical point of view, Hopf bifurcations correspond to two conjugated complex eigenvalues of the Jacobian matrix crossing the imaginary axis. The method to compute numerically this kind of instability can be classified into two families.

The first one is the family of indirect methods which consist in following a test function which has the property to be null in a Hopf bifurcation point [1,4,5]. This test function can be the smallest real part of one or more eigenvalues of the Jacobian matrix [6]. When this real part is equal to zero, then a Hopf bifurcation point is found. Nevertheless, as numerical fluid mechanics requires fine spatial discretization, the eigenvalue computation can lead to long CPU time. Moreover, due to non symmetric matrices, specific algorithms have to be used, for example the Arnoldi method [7].

In a recent work [1], the authors proposed an indirect method based on a bifurcation indicator initially introduced in solid mechanics [8,9]. These indicators are scalar functions which have the property to be null in a bifurcation point. In Ref. [1], two indicators have been proposed. The first one allows for the characterization of stationary instabilities, such as the loss of symmetry in the flow and has been used to study the Coanda effect [10]. The second one leads to determining precisely Hopf bifurcation points. This indicator was first introduced in Refs. [11,12], respectively for models with small numbers of degrees of freedom and solid mechanics involving following loads. This bifurcation indicator is a function of the Reynolds number and the frequency. In Ref. [1], the bifurcation indicator is computed for a fixed value of the Reynolds number. It means that for a given value of the Reynolds number, the evolution of this indicator is evaluated versus the frequency. As the computation cost of this indicator can be great, an Asymptotic Numerical Method is used [13]. The latter permits the user to obtain analytic solutions. Nevertheless, with this method, the user has to follow the minimum value of the indicator versus the Reynolds and finally isolate the critical values of the flow indicating a Hopf bifurcation. This method is not an automatic ones and needs a great deal of post-analysis work for each computation. Moreover, precisely computing the critical values of the flow can lead to a great amount of indicator calculus and therefore to long CPU time.

The second family of methods used to characterize Hopf bifurcation points is the direct methods. They consist in the resolution of a nonlinear augmented system whose solutions are the Hopf

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bifurcation points. One can cite the methods proposed by Jepson [3] and Griewank and Reddien [14] which are solved by a Newton method [2]. These methods need initial values close to the bifurcation point to assure convergence. Usually [2,15], initial values are obtained by the computation of some eigenvalues of the Jacobian matrix. In fact, one must isolate or at least have an idea of what the eigenvalues which can cross the imaginary axis are. Unfortunately, as this operator depends on the Reynolds number, the eigenvalues also do. One might note that their evolution with this number is not linear [16]. Consequently, it can be difficult to estimate precisely the eigenvalues which can cross the imaginary axis. Moreover, the numerical resolution of fluid mechanics requires a fine spatial discretization which can lead to an extensive matrix operator. The use of eigenvalues as initial values can then be quite difficult.

To avoid these eigenvalue computations, some authors [17] propose the quantities coming from an unsteady Navier–Stokes simulation as initial values. If for a given Reynolds number, the flow is time periodic, a Fourier transformation to the time series is applied to determine the critical frequency. Additional relations help give an approximation of the complex eigenvector [17].

In this work, we propose to use the quantities coming from the bifurcation indicator proposed in Ref. [1] as initial approximations for the augmented system proposed by Jepson [3] and Griewank and Reddien [14]. The basic idea is to associate the advantages of both methods to obtain a simple and efficient algorithm for the computation of Hopf bifurcation points. Indeed, the latter method can easily give good initial guesses but can demand long CPU time if precise values of the critical flow parameters are required. The direct method can compute precise critical numbers in few iterations if initial approximations are relatively close to the bifurcation point.

This paper is organized as follows. In the first part, the governing equations to be solved are recalled. The second part is concerned by the presentation of the two methods: firstly the direct method and secondly the bifurcation indicator. The numerical scheme is introduced during the third part. Finally, numerical examples, concerning classical tests in fluid mechanics are reported in the last part of this paper.

2. Governing equations

The Navier–Stokes equations for a Newtonian and incompressible fluid are the following:

$$\begin{cases} \frac{\partial u_i}{\partial t} - \nu u_{i,jj} + u_j u_{i,j} + \frac{1}{\rho} p_{,i} = 0 & \text{in } \Omega \\ u_{i,i} = 0 & \text{in } \Omega \\ u = \lambda u_d & \text{on } \partial_u \Omega \end{cases} \quad (1)$$

where u and p are respectively the velocity and the pressure, u_d is the imposed velocity in the boundary $\partial_u \Omega$ and ρ is the density. In Eq. (1), λ designates the intensity of the imposed velocity and can be identified by the Reynolds number, denoted Re , which is defined by using a reference length and the kinetic viscosity ν . The previous equations are written in the following operator form:

$$M(\dot{U}) + L(U) + Q(U, U) - \lambda F = 0 \text{ in } \Omega \quad (2)$$

where M is the mass matrix, $L(U)$ and $Q(U, U)$ are respectively the linear and quadratic operators and U is a mixed unknown vector defined by:

$$U = \begin{Bmatrix} u \\ p \end{Bmatrix} \quad (3)$$

In Eq. (2), a load vector F is introduced which is identical to the velocity imposed in the boundary $\partial_u \Omega$. For a more precise definition of these operators, the reader can refer to Ref. [18].

To study the stability of the Navier–Stokes equations (1), a perturbation vector $\mathcal{V}$ from the steady solution U^S is introduced in Eq. (2) and by neglecting the second order terms in $\mathcal{V}$, the latter equation becomes:

$$M(\dot{\mathcal{V}}) + L(\mathcal{V}) + Q(\mathcal{V}, U^S) + Q(U^S, \mathcal{V}) = 0 \text{ in } \Omega \quad (4)$$

As U^S is a steady solution of the Navier–Stokes equations, U^S verifies:

$$L(U^S) + Q(U^S, U^S) - \lambda^S F = 0 \text{ in } \Omega \quad (5)$$

where λ^S is equivalent to the Reynolds number at the steady solution U^S .

A Hopf bifurcation occurs when two eigenvalues of the Jacobian matrix cross the imaginary axis. So the velocity perturbation is sought under the form:

$$\mathcal{V}(\mathbf{x}, t) = V(\mathbf{x})e^{i\omega t} \quad (6)$$

where ω is the frequency, V the complex eigenvector and $\mathbf{x} = (x, y)$ is the position vector of any point in the flow domain Ω . The previous expression is introduced into Eq. (4) which becomes, vector $\mathbf{x}$ being omitted for the sake of simplicity:

$$\begin{cases} i\omega M(V) + L(V) + Q(V, U^S) + Q(U^S, V) = 0 & \text{in } \Omega \\ V_u = 0 & \text{on } \partial_u \Omega \end{cases} \quad (7)$$

where V_u designates the velocity part of the mixed unknown vector V . The latter is given by:

$$V = V^a + iV^b \quad (8)$$

In the previous expression, V^a and V^b are respectively the real and imaginary part of V . Finally, by introducing relation (8) into Eq. (7), a Hopf bifurcation point is the solution of the following system:

$$\begin{cases} L(U^S) + Q(U^S, U^S) - \lambda^S F = 0 & \text{in } \Omega \\ L(V^a) + Q(V^a, U^S) + Q(U^S, V^a) - \omega M(V^b) = 0 & \text{in } \Omega \\ L(V^b) + Q(V^b, U^S) + Q(U^S, V^b) + \omega M(V^a) = 0 & \text{in } \Omega \\ V_u^a = 0 & \text{on } \partial_u \Omega \\ V_u^b = 0 & \text{on } \partial_u \Omega \end{cases} \quad (9)$$

The previous nonlinear system is rewritten in the following condensed form:

$$\begin{cases} \mathcal{L}(\Lambda) = 0 & \text{in } \Omega \\ V_u^a = 0 & \text{on } \partial_u \Omega \\ V_u^b = 0 & \text{on } \partial_u \Omega \end{cases} \quad (10)$$

where the unknown Λ is defined by $\Lambda = \{U^S, V^a, V^b, \lambda^S, \omega\}$. Additional normalization conditions have to be added to the previous system to obtain a well-posed problem. These conditions are specified in the following sections where the methods used to compute the unknown vector Λ are presented.

3. Methods for computing Hopf bifurcation points

To make this paper self-contained, let us recall the two methods used to compute bifurcation points. The coupling between these methods is also presented.

3.1. Direct method

The method proposed by Jepson [3] consists in solving problem (9) with the help of the Newton method. The unknown vector Λ is then sought under the following expression:

$$\Lambda = \Lambda_0 + \Delta \Lambda \quad (11)$$

where A_0 is a known initial approximation and ΔA is the unknown increment. The previous relation is introduced into Eq. (10) and by neglecting the second order terms in ΔA , one obtains the equations to be solved at each iteration of the Newton method:

$$\begin{cases} \mathcal{L}_t^{A_0}(\Delta A) = \mathcal{R}^{A_0} \text{ in } \Omega \\ \Delta V_u^a = 0 \text{ on } \partial_u \Omega \\ \Delta V_u^b = 0 \text{ on } \partial_u \Omega \end{cases} \quad (12)$$

where $\mathcal{L}_t^{A_0}(\Delta A)$ and $\mathcal{R}^{A_0}$ are respectively the tangent operator and the residual vector computed at the initial point A_0 . These quantities are precisely defined in Appendix A.

To have a well-posed problem (25), two additional conditions are introduced [2,3,14]:

$$\begin{cases} \langle l, V_u^a \rangle = 0 \\ \langle l, V_u^b \rangle = 1 \end{cases} \quad (13)$$

where l is a given vector (in our case an arbitrary chosen vector) and $\langle \bullet, \bullet \rangle$ stands for the euclidian scalar product. Although these normalization conditions can lead to turning points (i.e. $\omega = 0$ and $V^a = 0$) [16,17], they are linear expressions and their numerical treatment is easy to do. Discussions about such a choice are given in Refs. [3,17].

3.2. Bifurcation indicator

In Ref. [1], the authors have proposed an indicator, the property of which is to be null at a Hopf bifurcation point. Instead of solving the augmented system (9), they have fixed the Reynolds number (i.e. the steady solution U^S is known) and they have computed the unknowns (V^a, V^b, ω) which satisfied system (9) with the help of an indicator. The steady solution is then computed independently by using an Asymptotic Numerical Method [18]. Eq. (9) are then replaced by the following ones:

$$\begin{cases} L(V^a) + Q(V^a, U^S) + Q(U^S, V^a) - (\omega_0 + \hat{\omega})M(V^b) = \Phi f \text{ in } \Omega \\ L(V^b) + Q(V^b, U^S) + Q(U^S, V^b) + (\omega_0 + \hat{\omega})M(V^a) = 0 \text{ in } \Omega \\ V_u^a = 0 \text{ on } \partial_u \Omega \\ V_u^b = 0 \text{ on } \partial_u \Omega \end{cases} \quad (14)$$

where f is a known random vector (chosen real), Φ is an unknown scalar, ω_0 designates the initial angular frequency. In the numerical examples presented in this paper, ω_0 is equal to zero. This is not a rule and ω_0 can be chosen close to a critical value of the angular frequency.

The scalar Φ is the bifurcation indicator. Determining a Hopf bifurcation point consists in finding, for the steady solutions, the values $\omega = \omega_0 + \hat{\omega}$ where Φ is equal to zero. The previous system can be solved by using a Newton method for each value of the angular frequency ω . Nevertheless as the resolution of system (14) requires the triangulation of a great size operator, the total CPU time for the computation of Φ can be great. To avoid this, a perturbation method is proposed to solve system (14). The unknown vector $X = (\Phi, V^a, V^b)$ is then sought as an integro-power series with respect to the angular frequency $\hat{\omega}$:

$$X = X_0 + \hat{\omega}X_1 + \hat{\omega}^2X_2 + \dots + \hat{\omega}^pX_p \quad (15)$$

To obtain a well-formulated problem, the following addition condition is required:

$$\|V_u\|^2 = \|W_{u_0}\|^2 \quad (16)$$

where $\|\bullet\|$ designates the euclidian norm of the vector ' $\bullet$ '. The additional condition (16) is chosen to avoid numerical instabilities (see Ref. [1] for more details). The vector W_{u_0} is the solution to problem

(14) where $\omega = 0$ and Φ_0 equal to 1. The asymptotic expansions (15) are then inserted in Eq. (14) and with balancing terms with identical powers of $\hat{\omega}$, we obtain the following set of linear equations:

Order 0 in $\hat{\omega}$

$$\begin{cases} L_t(V_0^a) - \omega_0 M(V_0^b) = \Phi_0 f \text{ in } \Omega \\ L_t(V_0^b) + \omega_0 M(V_0^a) = 0 \text{ in } \Omega \\ V_{u_0}^a = 0 \text{ on } \partial_u \Omega \\ V_{u_0}^b = 0 \text{ on } \partial_u \Omega \\ \|V_{u_0}\|^2 = \|W_{u_0}\|^2 \end{cases} \quad (17)$$

Order p in $\hat{\omega}$

$$\begin{cases} L_t(V_p^a) - \omega_0 M(V_p^b) = \Phi_p f + M(V_{p-1}^b) \text{ in } \Omega \\ L_t(V_p^b) + \omega_0 M(V_p^a) = -M(V_{p-1}^a) \text{ in } \Omega \\ V_{u_p}^a = 0 \text{ on } \partial_u \Omega \\ V_{u_p}^b = 0 \text{ on } \partial_u \Omega \\ \text{For } p = 1, \quad \langle V_{u_1}, V_{u_0} \rangle = 0 \\ \text{For } p \geq 2, \quad \langle V_{u_p}, V_{u_0} \rangle + \sum_{r=1}^{p-1} \langle V_{u_{p-r}}, V_{u_r} \rangle = 0 \end{cases} \quad (18)$$

The linear systems (17) and (18) all have the same operator. So only one triangulation and $(p+1)$ forward/backward substitutions permit the computation of the solution (15). In the previous systems, the tangent operator L_t is the one which is defined in Eq. (27) in Appendix A. The validity range of these asymptotic expansions, which means the maximum value of $\hat{\omega}$, is determined by a simple criterion introduced in Ref. [13] and recalled in Appendix B.

From this maximum value of $\hat{\omega}$, one can compute a new initial value ω_0 and by applying this scheme in a step-by-step manner, the indicator Φ can be determined for any value of ω . To increase the validity range of asymptotic expansions and then decrease the number of operator triangulations, the polynomial expansions (15) are replaced by equivalent rational ones, called Padé approximants [19]:

$$X_{\text{Padé},p}(\hat{\omega}) - X_0 = \sum_{k=1}^{p-1} \frac{P_{(p-1-k)}(\hat{\omega})}{Q_{(p-1)}(\hat{\omega})} \hat{\omega}^k X_k \quad (19)$$

where P_k and Q_k are polynoms of degree k . One can remark that all the fractions in (19) have the same denominator Q_k . This representation has been tested and evaluated in Refs. [20,21] and is generally preferred to limit the number of roots of the Padé's denominator, the so-called "poles" of Padé approximants.

To compute the entire response indicator versus frequency, denoted by $\mathcal{G}_{\text{Re}}(\Phi, \omega)$, a continuation method based on the Padé approximants and first proposed by Elhage-Hussein et al. [22] is used. As for the polynomial expansions, the computation of the validity range of these rational fractions is recalled in Appendix B. The latter is the only difference between the method proposed in Ref. [1] and the indicator used in this work. Numerical comparisons between these two representations are done in the first part of the numerical section (see Section 5.2).

With the previous relations, the curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ are computed for a fixed value of the Reynolds number. The determination of Hopf bifurcation points consists in finding where ω leads to a null value of the scalar Φ on these curves. In practice, it consists in

following the minimas of the function $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ and then estimates the critical numbers of the flow where a Hopf bifurcation appears.

3.3. The proposed algorithm

The computation of a Hopf bifurcation point with the presented direct method in Section 3.1, requires an initial approximation of A_0 which is relatively close to the solution to ensure the convergence of the Newton method. The bifurcation indicator, introduced in the previous section, allows one to obtain an estimation of the critical values of the flow. Thus we propose the following algorithm for the computation of the Hopf bifurcating point:

0. Set the parameters of the method:
 - (a) The area of interest for the Reynolds number: $\text{Re} \in [\text{Re}^1, \text{Re}^2]$.
 - (b) The order of truncation of the asymptotic expansions: p (expression (15)). In practice p is between 15 and 20.
 - (c) The number of steps, ns , in the continuation technique [22] (the range of interest for the perturbation parameter $\hat{\omega}$).
 - (d) The maximum number of iterations for the Newton method (relations (25)).
1. Compute the steady solution U_0^S and the corresponding Reynolds number, Re . In this work, it is done with the help of ANM [18].
2. If $\text{Re} \in [\text{Re}^1, \text{Re}^2]$, then:
 - Indicator phase, for step $j = 1, \dots, \text{ns}$.
 - 2.1 Computation of the Φ_p^j and V_p^a and V_p^b with the expressions (18).
 - 2.2 Evaluation of the validity range of the Padé approximants with the criterion defined in Ref. [22] and Appendix B. This means computing the maximum value of the perturbation parameter $\hat{\omega}$, denoted $\hat{\omega}_{\text{max}}$.
 - 2.3 Build the Padé approximants with the expressions (19) and $\hat{\omega}_{\text{max}}$.
 - 2.4 Continuation technique: $\omega_0^j = \omega_0^{j-1} + \hat{\omega}_{\text{max}}$, $V_0^a = V_{\text{Padé}}^a(\hat{\omega}_{\text{max}})$, $V_0^b = V_{\text{Padé}}^b(\hat{\omega}_{\text{max}})$.
 - Direct method phase
 - 3.1 Iterations of the Newton method (relations (25) and (13)), with U_0^S , V_0^a , V_0^b and ω_0 coming from steps 2.1 to 2.5 until convergence or the maximum number of iterations reached.

To summarize, the bifurcation indicator gives an initial approximation A_0 to the direct method during steps 2.1–2.5 of the previous algorithm. We will show in the numerical section how these initial values can be chosen in the best manner.

4. Spatial discretization

Spatial discretization of the previous equations are performed by using the classical finite element method. The chosen finite element is a quadrilateral element, with 9 nodes for the velocity (bi-quadratic interpolation) and 3 for the pressure (linear interpolation) [23]. The continuity equation is solved by using a penalty method [23]. In the following and for the sake of simplicity, discrete and continuous quantities have the same names.

The finite element formulation applied to expressions (12) and (13) leads to a linear system with a matrix which has a size $(\mathbb{R}^{3n+2} \times \mathbb{R}^{3n+2})$ (where n is the number of degrees of freedom). The triangulation of such an operator leads to a very long computational times. Therefore to avoid a large amount of computational time, the system (12) is solved in a two-step method [2]. Firstly,

the stationary solution, ΔU^S , is looked for under the following form:

$$\begin{cases} \Delta U^S = \alpha + \Delta \lambda \beta \\ \text{with } \alpha = K_t^{-1} \cdot R^S \text{ and } \beta = K_t^{-1} \cdot F \end{cases} \quad (20)$$

The previous relation is the discrete form of the linearized stationary Navier–Stokes Eq. (5) and K_t is the corresponding tangent matrix. Let us precise that this tangent matrix is the discrete form of the tangent operator defined in Eq. (27) in Appendix A.

Secondly, vectors α and β are substituted in the discrete expression of Eq. (12) which can be written:

$$[\mathcal{K}_t(A_0)]\{\Delta \Gamma\} = \{\mathcal{R}^{A_0}\} \quad (21)$$

where $[\mathcal{K}_t(A_0)]$, $\{\Delta \Gamma\}$ and $\{\mathcal{R}^{A_0}\}$ represent respectively the tangent matrix, the unknown vector and the residual computed at the initial value A_0 . The tangent matrix $[\mathcal{K}_t(A_0)]$ is approximatively the discrete form of the linear sytem (25) defined in Appendix A. In fact, the first equation (the stationary part) in Eq. (25) is removed and the stationary unknown, ΔU^S , is modified according to the expression (20).

The unknown vector $\{\Delta \Gamma\}$ is defined by the following expression:

$$\{\Delta \Gamma\} = \{\Delta \lambda, \Delta V^a, \Delta V^b, \Delta \omega\} \quad (22)$$

The computation of the vector $\{\Delta \Gamma\}$ is performed by the factorization of the operator $[\mathcal{K}_t(A_0)]$. As the size of this operator is $(\mathbb{R}^{2n+2} \times \mathbb{R}^{2n+2})$, the computation cost is cheaper than the initial ones. Finally, the stationary part, ΔU^S , of the vector ΔA is determined by introducing $\Delta \lambda$ in the Eq. (20).

For the bifurcation indicator, the discrete linear system at the truncature order p (18) is written:

$$\begin{bmatrix} K_t & -\omega_0 M \\ \omega_0 M & K_t \end{bmatrix} \begin{Bmatrix} V_p^a \\ V_p^b \end{Bmatrix} = \begin{Bmatrix} \Phi_p f + M V_{p-1}^b \\ -M V_{p-1}^a \end{Bmatrix} \quad (23)$$

where M is the discrete mass matrix. Solving system (23), associated with the additional condition (16) written at the order p , gives all the unknowns (V_p^a, V_p^b, Φ_p) . For all these linear systems, only one matrix triangulation is needed and 'p' backward and forward substitutions. As the size of this matrix is approximatively the same as that of the Newton method, this means that the computation of one step of the indicator requires the same computational times as a single iteration of the direct method. By using relation (19), one can build an analytical part of the curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$. The entire curves, $\mathcal{G}_{\text{Re}}(\Phi, \omega)$, are computed by using a continuation method based on the Padé approximants [22].

5. Numerical results

5.1. Numerical examples

The numerical tests presented in this paper are classical ones but have the advantage of presenting Hopf bifurcations which are relatively well-known. Hence, one considers two examples. The first one is the flow around a cylinder. This example is the same as in Ref. [2] excepted for the boundary conditions. Indeed, in this work, the outflow condition is a Dirichlet condition whereas in Ref. [2] a stress boundary condition is imposed. In fact, this Dirichlet condition can lead to spurious oscillations of the solutions. Nevertheless, as the considered Reynolds are relatively small (between 23 and 54), these spurious oscillations have not been observed in all our numerical computations. Let us remark that for higher Reynolds numbers, a stress boundary condition has to be imposed in the outflow. The second one is the lid-driven cavity. For this problem, several geometrical cases or aspect ratios A are studied in this paper (see Table 1).

Table 1
Mesh sizes used for the numerical tests.

Example	Aspect ratio	Number of d.o.f.
Flow around a cylinder	–	6112
Lid-driven cavity	$A = 0.8$	20.402
	$A = 1.0$	13.122
	$A = 1.5$	13.122

The geometrical description of these examples and the boundary conditions are shown in Figs. 1 and 2 respectively for the flow around a cylinder and the lid-driven cavity. As discussed previously, the finite element used in the numerical tests is a quadrangle with nine velocity nodes and three for the pressure. The number of d.o.f. for the numerical tests is given in Table 1.

We give in Tables 2 and 3 the critical Reynolds and Strouhal numbers reported by various researchers, respectively for the flow around a cylinder and the lid-driven cavity with an aspect ratio A equal to 1. We also indicate in these tables, for all the references, the numerical method used to characterize a Hopf bifurcation point. One can remark in these tables that, most of the works are based on an eigenvalue computation (EC in the Tables 2 and 3). In the latter and concerning the eigenvalue computation we do not distinguish the kind of eigensolution method used in the different references. For example, in Ref. [28] a subspace iteration is used, in Ref. [25] it is a classical QR algorithm

Fig. 1. Flow around a cylinder, description of the geometry.

Fig. 2. Geometry for the 2-D lid-driven cavity flow.

Table 2
Comparison of the critical parameters for the first Hopf bifurcation from the literature. The following acronyms EC, DNS, NM respectively stand for eigenvalue computation, direct numerical simulations and Newton method. The considered example is the flow around a cylinder.

Authors	Re_c	St_c	Method
Jackson [2]	46.184	0.137	NM
Cossu and Morino [25]	45.63	0.112	EC
Kumar and Mittal [26]	46.88	0.116	DNS and EC
Ding and Kawahara [27]	46.389	0.126	EC
Morzynski et al. [28]	47	0.132	EC
This study	45.98	0.137	

Table 3
Comparison of the critical parameters for the first Hopf bifurcation from the literature. The following acronyms EC, POD, DNS, NM, LE respectively stand for eigenvalue computation, proper orthogonal decomposition (associated to the computation of the eigenvalues of the reduced system), direct numerical simulations, Newton method, Lyapunov exponent. The considered example is the lid-driven cavity with $A = 1$.

Authors	Re_c	St_c	Method
Fortin et al. [5]	8000	0.45	EC
Boppana and Gajjar [6]	8026.6	0.4497	EC
Polishenko and Aidun [17]	7763.4	0.45	NM
Bruneau and Saad [29]	8000–8050	–	LE
Peng et al. [30]	7402	0.59	DNS
Abouhamza and Pierre [31]	8004.5	–	EC
Gervais et al. [32]	7960	0.45	EC
	8040	0.45	EC
Tiesinga et al. [33]	8375	0.4399	EC
Cazemier et al. [34]	7819	0.61	POD
	7972	0.45	DNS
Auteri et al. [35]	≈ 8018	0.4496	DNS
This study	7890	0.44	

and in Ref. [6] the eigenvalue problems is solved with an Arnoldi package. In Ref. [34], the authors have also computed the eigensolutions of the linearized perturbed Navier–Stokes equations. Nevertheless, a low order model (POD) is used to reduce the size of the Jacobian matrix.

One should remark that in Ref. [2], an eigenvalue computation is also realized (inverse iteration method) to give an initial guess for the Newton method. In Ref. [17], the Hopf bifurcation point is also computed with the help of a Newton method, but this time, the starting values for the iterative method are obtained with a time dependant Navier–Stokes simulation. From these tables, one can note that in a few works, a direct numerical simulation of the time dependant Navier–Stokes equations has been done. This is mainly due to the fact that this kind of technique requires a great amount of CPU time to localize precisely the bifurcation point. Finally, in Ref. [29], the stability of the flow is studied by computing the first Lyapunov exponent of the linearized problem.

For these two classical benchmarks in the numerical study of flow stability, the literature's results presented in Tables 2 and 3 show that all the studies give approximatively the same critical Reynolds and Strouhal numbers. Hence, for the flow around a cylinder, the Hopf bifurcation occurs for a Reynolds and a Strouhal number respectively between [45–47] and [0.112–0.137]. For the lid-driven cavity, the Hopf bifurcation point is found for a Reynolds number included in the range [7400–8400] and the corresponding critical frequency is close to 0.45. In Ref. [34], the found critical Strouhal number is greater than the other works reported in Table 3. This difference is explained by the fact that the small dynamical system (from which an eigenvalue computation is done) is obtained with a time dependant simulation realized at a Reynolds number equal to 22,000.

In the following, we first propose to give some explanations on how the bifurcation indicator [1] works. Next, we compare the computing performances obtained with the first version of the indicator [1] and the one based on the Padé approximants. In the following section, we show how this indicator is linked to the Newton method (see Section 3.1) to give an efficient algorithm for the computation of Hopf bifurcation points. A section is also devoted to the computational times of the proposed method. The proposed algorithm is then applied to the lid-driven cavity with geometrical aspect ratio equal to 0.8 and 1.5. In the last part of this numerical section, we show that the coupled method can easily give some supplementary Hopf bifurcation points.

5.2. Bifurcation indicator

We first propose to give some explanations about the bifurcation indicator presented in Section 3.2 and in Ref. [1]. We then consider the flow around the cylinder. The critical numbers for this flow are $Re_c = 45.98$, $\omega_c = 3.97$ [1] and the corresponding Strouhal

number is equal to $St_c = 0.137$. Jackson [2] has found, for this example, the following values: $Re_c = 46.184$ and $St_c = 0.138$. The Strouhal number is computed from the following expression:

$$St = \frac{D \cdot \omega}{2\pi u} \quad (24)$$

where u is the imposed velocity (see Fig. 1) and D is the cylinder diameter. To understand how the indicator, Φ , works we have computed it for a Reynolds number $Re \in [20, 60]$. The curves $\mathcal{G}_{Re}(\Phi, \omega)$ for four values in the range of Reynolds numbers are shown in Fig. 3. We indicate for each curve, $\mathcal{G}_{Re}(\Phi, \omega)$, the minimum value of the indicator Φ . One can see in these curves that the minimum of the indicator decreases when we get closer to the critical Reynolds and increases when we get farther from this value. Indeed the minimum is equal to 0.42 for $Re = 23.05$, 0.19 for $Re = 32.41$, 0.0029 for $Re = 45.60$ and 0.04 for $Re = 54.92$, which means that, according to the conclusions presented in Ref. [1] there is a Hopf bifurcation point between the Reynolds number 32.41 and 54.92. So to determine precisely this critical value one has to compute

Fig. 3. Indicator versus angular frequency, curves $\mathcal{G}_{Re}(\Phi, \omega)$. The dashed lines indicate the range in ω where the Newton algorithm converges. The considered example is the flow around a cylinder.

supplementary curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ for some Reynolds numbers in this range. From these computations, one has to determine a new range of Reynolds number where the indicator is minimum and so on until this minimum reaches a fixed user given tolerance.

In Table 4, we give, for each Reynolds number considered in Fig. 3, the number of steps and the final value of the parameter ω reached. We consider the polynomial approximation (15) and the Padé approximants (19). The truncature order p is fixed and equal to 20 according to the conclusions of previous studies about ANM [24]. In fact, in Table 4 we have fixed the number of steps of the Padé approximants to 20 and we give the number of steps demanded with the polynomial representation to reach the same final value of ω . In this table, one can remark that the number of steps demanded with the Padé approximants is always smaller than with the polynomial representations.

To illustrate greater step lengths obtained with the Padé approximants, we have plotted in Figs. 4 and 5 the evolution of the indicator versus the angular frequency (for a Reynolds number equal to 39.01). In these figures, the square symbol indicates the starting point X_0 of the asymptotic expansions (15). These starting points are determined by using expression (30). In these figures, the reference curve is the one obtained with the continuation method. In these curves, Fig. 5 being an enlarged view of Fig. 4, we have plotted for the second continuation point the evolutions

Table 4

Number of steps to compute the indicator by using the polynomial representation (15) or with the Padé approximants (19). The order of truncature p is equal to 20. The considered example is the flow around a cylinder.

	Re	Number of steps	ω reached
Padé	23.05	20	10.16
Poly		26	10.17
Padé	32.41	20	8.71
Poly		24	8.74
Padé	39.01	20	6.75
Poly		23	7.21
Padé	45.60	20	4.00
Poly		24	4.00
Padé	54.92	20	8.20
Poly		23	8.23

Fig. 4. Bifurcation indicator, comparison of the solutions obtained with Padé and polynomial approximation. Re = 39.01. Square symbols indicate the beginning of the step. The considered example is the flow around a cylinder.

Fig. 5. Bifurcation indicator, comparison of the solutions obtained with Padé and polynomial approximation. Re = 39.01. Zoom of the Fig. 4. The considered example is the flow around a cylinder.

of the polynomial approximations (15) and of the Padé approximants (19).

One can clearly see in these figures that the solution obtained with the Padé approximants is closer to the reference curve for greatest values of the angular frequency than with the polynomial approximations. It shows that the range of validity of the Padé approximants is generally greater than the one obtained with the polynomial approximations. Each step of the continuation method needs a single matrix triangulation (Eq. (23)) and p backward/forward substitution. The points where a matrix triangulation is realized are indicated by square symbols in Figs. 4 and 5. We have reported in Table 5, the CPU time needed for the computation of one step of the indicator. One considers the CPU time for the matrix triangulation, for the computation of the solution of the p linear systems (18) and for the construction of the Padé approximants (19). These times are given for the flow around a cylinder and the lid-driven cavity, respectively denoted by Examples 1 and 2 in Table 5. In the latter, one can see that the largest CPU time of one step of the perturbation method is for the matrix triangulation. The computation of all the unknowns X_p requires approximatively 30% and 14% of the total CPU cost respectively for Examples 1 and 2. This short CPU time is due to the fact that the matrix triangulation is done only once, for the order of truncature equal to 0, and for the supplementary unknowns (for $p \geq 1$), only forward and backward substitutions are required. Concerning the Padé approximants, Table 5 illustrates the fact that the CPU time for the construction of the rational fraction (19) is insignificant compared to what it needed for a matrix triangulation. Finally

Table 5

CPU time (in second) for one step of the indicator computation. The truncature order is equal to 20. Examples 1 and 2 designate respectively the flow around a cylinder and the 2-D lid-driven cavity. Computations are realized on a workstation DELL Precision 490.

	Example 1 (6112 d.o.f.)	Example 2 (13 122 d.o.f.)
Matrix triang	7.3	58.67
Computations of all the X_p	2.83	9.64
Padé approximants	0.04	0.07
Total	10.13	68.31

as the Padé approximants lead to larger step sizes than with the polynomial approximations, the CPU computing times to determine the indicator curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ is then reduced. This is an improvement of the bifurcation indicator compared to the ones previously presented in Ref. [1]. Comparisons of the computational cost between the indicator proposed in Ref. [1] and the method proposed in this work are realized in the following section.

Another important feature of the indicator curves is that, for a fixed number of steps, the maximum value of ω decreases when a bifurcating point is close. For example, in Table 4, one can see that for a Reynolds number equal to 23.05 with 20 steps and the Padé approximants, we get a maximum value of ω equal to 10.17, whereas for a Reynolds number equal to 45.60, the curve $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ is computed only up to $\omega = 4$. In fact, the nonlinear curve $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ is less regular when a bifurcation point is close to the considered Reynolds number. This is mainly due to the definition (16). Let us recall that this definition has been introduced in Ref. [1] to avoid numerical instabilities. In the following results, the Padé approximants are always the representation used for the computation of the curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$.

We now propose to associate the direct method presented in Section 3.1 by taking the initial values from the curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$.

5.3. Results with the proposed algorithm

We consider the same example as before (flow around a cylinder). For each curve $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ shown in Fig. 3, we choose initial values for the augmented system (12). It is quite easy to define these initial values. Indeed, as the perturbation method gives analytic solutions, one has to choose a value ω on the curves $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ and by introducing the chosen value in expressions (19), one obtains initial values for the Newton method. This has been done for the four values of the Reynolds number in the case of the flow around a cylinder. For a Reynolds number equal to 23.05 (Fig. 3a, no chosen value ω on this curve leads to a convergence of the Newton algorithm. For the other values of the Reynolds number, the area in ω for which the Newton method converges is plotted in dashed lines in Fig. 3. The closer the critical Reynolds number is, the wider the convergence area is.

In Fig. 6, we plot the number of necessary iterations to get the required accuracy with the Newton algorithm for the values of ω

Fig. 6. Number of iterations of the Newton method (system (12)) versus pulsation for several Reynolds numbers. The convergence is reached when the following criteria are satisfied: $\|R^S\| < 10^{-4}$, $\|R^a\| < 10^{-8}$ and $\|R^b\| < 10^{-8}$. The considered example is the flow around a cylinder.

highlighted in Fig. 3 by dashed lines. Obviously the minimum number of iterations is obtained when the Reynolds number is equal to 45.60. Nevertheless, one can remark that each curve $\mathcal{G}_{\text{Re}}(\Phi, \omega)$ gives initial values leading to the convergence of the Newton method. Moreover, if on each curve $\mathcal{G}_{\text{Re}}(\Phi, \omega)$, one chooses as an initial guess the value of ω for which the indicator Φ is minimum, the Newton method converges to the critical numbers of the flow which are $\text{Re}_c = 45.98$ and $\omega_c = 3.972$ ($\text{St}_c = 0.137$). These critical values are very similar to those found in the literature and reported in Table 2.

We now consider the 2-D lid-driven cavity with an aspect ratio A equal to 1. We have computed the indicator curves, $\mathcal{G}_{\text{Re}}(\Phi, \omega)$, for Reynolds numbers between 6000 and 11,000. The evolution of the indicator versus the Strouhal number for four values of the Reynolds number included in the studied range is plotted in Fig. 7. This example is interesting because the curves, $\mathcal{G}_{\text{Re}}(\Phi, \text{St})$, show a lot of minima of the indicator, Φ , for St_c in $[0; 0.8]$. Moreover, these minima change with the Reynolds number. For example, when $\text{Re} = 7257$ the minimum of the indicator is reached for St_c close to 0.27, whereas when $\text{Re} = 8089$ it is for $\text{St}_c = 0.44$. These values are summarized in Table 6. For each minimum value of the curves $\mathcal{G}_{\text{Re}}(\Phi, \text{St})$, the direct method is used to determine precisely the critical values of the flow. The number of iterations of the Newton method and the found critical numbers are given in Table 6.

For this example, the first Hopf bifurcation is found for a Reynolds number and a Strouhal number respectively equal to 7890 and 0.44. These critical values are close to the ones found in the literature, see for example Table 3 or the Ref. [6] where a great number of works on this numerical test are summarized. To prove that our results correlate well with those of the literature, we plot in Fig. 8 the streamlines of the stationary solution (at $\text{Re} = 7890$, Fig. 8a) and of the real and imaginary parts of the critical eigenvector (respectively Fig. 8b and c). These figures are quite similar to those presented in Refs. [5,6,31].

5.4. Computational costs

This section is devoted to the comparison of the computational cost needed by the proposed method and the indicator presented in Ref. [1]. We consider the two previous examples: the flow around a cylinder (denoted by Example 1) and the lid-driven cavity (denoted by Example 2). These computational cost are given in Table 7. In this table, we give the number of matrix triangulations, N_1 , (which is the most time consuming operation in both methods) needed for each method and the corresponding CPU times. For the proposed method, the estimation of the CPU times is easy to do. For example, let us consider the flow around a cylinder (Example 1) with an initial Reynolds number equal to 39.01 ($\text{Re}_{\text{ini}} = 39.01$ in Table 7). To compute the initial guesses for the Newton method, the indicator phase requires 20 steps of the perturbation method (according to the results presented in Table 4. From this computation, the number of Newton iterations to converge (with the following required accuracy $\|R^S\| < 10^{-4}$, $\|R^a\| < 10^{-8}$ and $\|R^b\| < 10^{-8}$) towards the Hopf bifurcation point is equal to 5 (see Fig. 6). The total number of triangulated matrix to determine precisely the critical values of the flow is then equal to 25 (20 for the indicator phase and 5 for the Newton method). By taking into account this number of matrix triangulations and the CPU times given in Table 5 for each step of the proposed method, the total computational cost can be estimated. Let us point out that, as the tangent matrix of the Newton method has approximatively the same size as the one of the indicator phase, we consider the same computational times for both methods.

The computational times needed to compute the bifurcation point when only indicator calculi are realized (as initially proposed in Ref. [1]) are now considered. These computational times are

Fig. 7. Indicator versus angular Strouhal number. The considered example is the 2-D lid-driven cavity with $A = 1$.

Table 6

Critical parameter values obtained with several initial values. Convergence of the Newton method for the following accuracy: $\|R^S\| < 10^{-4}$, $\|R^e\| < 10^{-8}$ and $\|R^b\| < 10^{-8}$. 'NC' means no convergence of the Newton method after 20 iterations. The considered example is the lid-driven cavity, $A = 1$.

Re_{ini}	St_{ini}	Φ_{mini}	Nb iterations	Re_c	St_c
6332.42	0.27	0.65	NC		
7257.45	0.27	0.508	NC		
7672.70	0.44	0.45	5	7890.29	0.44
8089.12	0.44	0.39	4	7890.28	0.44
8488.05	0.52	0.29	5	8829.59	0.52
8888.08	0.52	0.04	4	8829.59	0.52
9296.59	0.34	0.26	6	11286.55	0.33
9706.36	0.34	0.206	7	11286.55	0.33
10147.27	0.34	0.14	6	11286.55	0.33
10589.87	0.33	0.088	6	11286.55	0.33

more difficult to estimate than with the proposed method. As explained in Section 5.2, for Example 1, the minimum of the indicator decreases when we get closer to the critical Reynolds number and

increases when we get farther from this value. This indicates that a Hopf bifurcation occurs between the tested Reynolds numbers. It means that the indicator needs at least three computations of the curve $\mathcal{G}_{Re}(\Phi, St)$ (for three Reynolds numbers) to suggest that a Hopf bifurcation is in this range of Reynolds numbers. With these three computations, the indicator does not give a precise value for the critical values of the flow (especially for the Reynolds number). Indeed, these critical values are $(Re_c = 45.6, St_c = 0.137)$ and $(Re_c = 8089, St_c = 0.44)$ respectively for the flow around and the lid driven cavity. To compute more precise values of these critical values, supplementary indicator calculi have to be done in this range of Reynolds numbers. Finally, if with the indicator, the same accuracy as with the proposed method is required this can lead to a great number of computed indicator curves and consequently to a large computational cost. So in Table 7, the minimum computational cost is given that permits one to prove with the indicator that a Hopf bifurcation exists in the studied range of Reynolds numbers. These computational costs are established by taking into account the results presented in Figs. 3 and 7 respectively for the Examples 1 and 2. With the results presented in Table 7, one can

Fig. 8. Streamlines for the lid-driven cavity, $A = 1$, at $Re = 7890$. The Strouhal Number is equal to 0.44.

clearly see that the proposed method leads to lower CPU times than the indicator proposed in Ref. [1]. So the proposed hybrid method is a real improvement of the indicator presented in Ref. [1].

5.5. The lid-driven cavity with aspect ratio, $A = 0.8$ and $A = 1.5$

We now consider the lid-driven cavity with a geometric parameter A equal to 0.8 and 1.5. The number of references concerning

these two examples is relatively small compared to case $A = 1$. The found critical parameters for these values of parameter A are summarized in Table 9. The computations are done in the same manner as before. First the curves $\mathcal{G}_{Re}(\Phi, St)$ are computed. Next the minimum of the indicator is automatically detected on this curve and is finally used as an initial guess for the Newton method. One considers first that case A is equal to 0.8. For this example, the following critical numbers corresponding to the first Hopf

Table 7

Comparison of the estimated CPU time to compute the first bifurcation point with the indicator[1] and the proposed method according to the CPU time given in the Table 5 and the number of Newton iterations shown in Fig. 6 and in Table 6. N1 is the number of matrix triangulation needed and T1 is the CPU time. Examples 1 and 2 are respectively the flow around a cylinder and the lid-driven cavity with $A = 1$.

Method			N1	T1 (s)
Example 1	Indicator [1]	$Re \in [39.01, 54.92]$	3×20	610
	This study	$Re_{ini} = 39.01$	$20 + 5$	254
		$Re_{ini} = 45.6$	$20 + 4$	244
		$Re_{ini} = 54.92$	$20 + 5$	254
Example 2	Indicator [1]	$Re \in [7257, 8089]$	$63 + 66 + 68$	13,470
	This study	$Re_{ini} = 7672.7$	$66 + 5$	4806
		$Re_{ini} = 8089$	$68 + 4$	4885

Table 8

Comparison of the critical parameters for the first Hopf bifurcation from the literature. The following acronyms EC and NM respectively stand for eigenvalues computation and Newton method. The considered examples are the lid-driven cavity with $A = 0.8$ and $A = 1.5$.

Aspect ratio	Authors	Re_c	St_c	Method
$A = 0.8$	Polishenko and Aidun [17]	5225	0.35	NM
	This study	5698	0.31	
$A = 1.5$	Boppana and Gajjar [6]	5326	0.34	EC
	Polishenko and Aidun [17]	7216	0.44	NM
	Abouhamza and Pierre [31]	5674	0.49	EC
	This study	7769	0.239	

bifurcation, $Re_c = 5698$ and $St_c = 0.31$ (see Table 9), are found. In Ref. [17] (see also results in Table 8), the authors have mentioned

a critical Reynolds number equal to 5225 and an adimensional frequency close to 0.35. These results are quite different from those found in our study. However, by comparing the streamlines of

(a) Steady solution

(b) Real part of the eigenvector

(c) Imaginary part of the eigenvector

Fig. 9. Streamlines for the lid-driven cavity, $A = 0.8$ plotted at $Re = 5698$ and $St_c = 0.31$.

the stationary solution (at $Re_c = 5698$) for both real and imaginary parts of the eigenvector found in this work (see Fig. 9) and those of Ref. [17], one can clearly say that it is the same Hopf bifurcation point.

Concerning $A = 1.5$, the present algorithm gives a critical Reynolds number equal to 7769 and a corresponding Strouhal number equal to 0.239 (Table 9). For this example, Polianshenko and Aidun

[17] have found $Re_c = 7216.9$ and $St_c = 0.44$ whereas Boppana and Gajjar [6] have mentioned critical Reynolds and Strouhal numbers equal respectively to 5326.9 and 0.34 (see also results in Table 8). Once more, if the stationary solution, the real and the imaginary parts of the eigenvector at this critical Reynolds number are compared (see Fig. 10) to those shown in Refs. [6,17], one comes to the same Hopf bifurcation points.

Fig. 10. Streamlines for the lid-driven cavity, $A = 1.5$ plotted at $Re = 7769$ and $St_c = 0.24$.

Nevertheless, Boppana and Gajjar [6] also reported a difference between the critical numbers they have found and the other works of the literature. The previous authors explain these differences by the fact that the number of computations done for these examples is relatively small (the case $A = 1.5$). We have performed supplementary computations with other grid sizes (smaller and larger) and the critical numbers are quite similar to those reported in this paper. In conclusion, the values of the critical numbers for these two examples seem to remain an open question.

The results presented in Tables 6 and 9 show an interesting task when using the proposed algorithm to determine Hopf points. Let us consider, for example, case $A = 1.5$ (Table 9). For this test, six initial values give the same bifurcation point ($Re_c = 7769$ and $St_c \approx 0.24$). These six initial guesses have approximatively the same value for the frequency, but are quite different for the Reynolds number. It shows that the bifurcation indicator can give good initial values on the stationary solution, even if the initial Reynolds numbers are far from the critical values. Moreover, results presented in Table 9 show that this is true for initial values before and after the critical Reynolds number. This is very interesting, because it means that with the indicator, we do not have to check any sign change of a function before performing the Newton method.

5.6. Additional bifurcation points

Another important feature concerning this work is the ability of the proposed algorithm to detect several bifurcation points. Indeed as the proposed algorithm is entirely automatic, the minimum on the curves $\mathcal{G}_{Re}(\Phi, St)$ is computed automatically and evolves with the Reynolds number. So in this section, we present for the previous considered examples the second and the third (if it is found) Hopf bifurcation points. For the sake of simplicity, we characterize them as Hopf bifurcation points because these points are solutions of the nonlinear system (9). Few works give critical numbers for the second and following Hopf bifurcations.

We consider the lid-driven cavity with the three previous aspect ratios A . The results concerning $A = 1$ are given in Table 6 and in Table 9 for $A = 0.8$ and $A = 1.5$. In Table 6, one can see that the second and the third bifurcations are respectively found for $Re_c = 8829$, $St_c = 0.52$ and $Re_c = 11,287$, $St_c = 0.33$. If these results are compared with those found in the literature (see Table 10) one can remark that few results correlate well with our results or with each other. Sometimes, the critical Reynolds and frequency numbers are quite similar but the bifurcation is not the same (see results of Tiesanga et al., Table 10, for $8875 < Re_c < 9375$ and $St_c = 0.52$ for the third bifurcation point). At other times, the Reynolds number and the frequency do not correspond. Even if the physical signification of these bifurcation points can be dis-

Table 9

Critical parameter values obtained with several initial values. Convergence of the Newton method for the following accuracy: $\|R^5\| < 10^{-4}$, $\|R^a\| < 10^{-8}$ and $\|R^b\| < 10^{-8}$. 'NC' means no convergence of the Newton method after 20 iterations. The considered example is the lid-driven cavity.

	Re_{ini}	St_{ini}	Nb iterations	Re_c	St_c
$A = 0.8$	6023.72	0.31058	4	5698.57	0.31239
	6549.52	0.41938	4	6308.61	0.420738
	7079.31	0.52311	4	6707.58	0.524798
$A = 1.5$	3077.77	0.00157	NC		
	3658.37	0.26426	9	7769.67	0.23941
	4246.35	0.26141	11	7769.67	0.23941
	5070.06	0.25644	7	7769.67	0.23941
	5902.49	0.25098	8	7769.67	0.23941
	7025.93	0.24352	7	7769.67	0.23941
	8158.78	0.23768	6	7769.67	0.23941
	9547.93	0.31275	6	10335.39	0.30919

Table 10

Results from the literature for the second and the third Hopf bifurcation points. The considered example is the lid-driven cavity with $A = 1$.

Authors	Re_c	St_c
Tiesanga et al. [33]	8375–8875	0.6
	8875–9375	0.52
	8875–9375	0.7
Cazemier et al. [34]	8214	1.16
	10,116	–
	11,175	–
Auteri et al. [35]	9765	0.27

cussed, the present method can easily and automatically give more than one bifurcation point.

Concerning the two other aspect ratios A , no results are reported in the literature. Thus, we just give, for information, the critical values found for the second and the third Hopf bifurcations in Table 9.

6. Conclusion

In this paper, we propose an algorithm to compute Hopf bifurcation points in fluid mechanics. This algorithm is based on the coupling of an indirect method (a bifurcation indicator presented in Ref. [1]) and a direct one (a nonlinear augmented system solved by the well-known Newton method). The bifurcation indicator presented in this paper is improved by using the Padé approximants which lead to less matrix triangulations and consequently saving computational times.

The bifurcation indicator gives initial values for the direct method. As these initial values are the minima values of the indicator Φ , they can be computed automatically and are easily taken as initial guesses for the Newton method. Several numerical examples emanating from classical benchmarks in fluid mechanics have shown the efficiency of the proposed algorithm. Indeed, the bifurcation indicator gives initial conditions which permit a fast convergence of the Newton method. As the minimum on the curves $\mathcal{G}_{Re}(\Phi, St)$ evolves with the Reynolds number, several bifurcation points can be easily computed. The advantage of the present algorithm is that these Hopf bifurcation points are automatically determined. In comparison with other methods, with the proposed algorithm the user does not have to check if a test function becomes null, the computation of the minima on the curves $\mathcal{G}_{Re}(\Phi, St)$ is automatically done and finally the Newton method starts with good initial values to determine the singular points. The drawback lies in the fact that the method is based on the detection of the minimum of the curves $\mathcal{G}_{Re}(\Phi, \omega)$ in a range $[0, \omega_{max}]$. So the user has to define, from the beginning, the range of interest in ω for which a bifurcation can appear. This value of ω_{max} is the only user parameter which seems to be difficult to define to avoid missing a bifurcation point.

Moreover, with the proposed method one can not certify, without doing a lot of computations, that the found critical values correspond to the first bifurcation point. Whereas, by computing the eigenvalues of the Jacobian matrix, this can be asserted.

The proposed method lies on the computation of at least one indicator curve, $\mathcal{G}_{Re}(\Phi, St)$. The computation cost of this curve can be great (see the results in Table 7). This is a drawback of the proposed method if, for example, 3D flow are investigated. Nevertheless, this computational cost can be decreased by using either reduced order models for example or a linear solver which is well-adapted to repeated right-hand side problems [38] such as the ones defined in Eqs. (17) and (18). Application of these techniques to the proposed method is a subject of current studied.

Future works also concern the use of a high order iterative method [36,37] instead of the Newton one. The main objective will be to decrease the number of matrix triangulations to compute precisely the bifurcation points.

Appendix A. Details on the direct method

In this appendix, we present the direct method used in this work. During one iteration of the Newton method, one has to determine the unknown increment, ΔA , which is the solution of the linearized system (12). This system written in a condensed form in the core of this paper is precisely defined by the following relations:

$$\begin{cases} L_t(\Delta U^S) - \Delta \lambda F = R^S \text{ in } \Omega \\ L_t(\Delta V^a) + Q(V_0^a, \Delta U^S) + Q(\Delta U^S, V_0^a) - \omega_0 M(\Delta V^b) \\ \quad - \Delta \omega M(V_0^b) = R^a \text{ in } \Omega \\ L_t(\Delta V^b) + Q(V_0^b, \Delta U^S) + Q(\Delta U^S, V_0^b) + \omega_0 M(\Delta V^a) \\ \quad + \Delta \omega M(V_0^a) = R^b \text{ in } \Omega \\ \Delta V_u^a = 0 \text{ on } \partial_u \Omega \\ \Delta V_u^b = 0 \text{ on } \partial_u \Omega \end{cases} \quad (25)$$

where R^S , R^a and R^b are respectively the steady residual, the real and the imaginary “dynamic” residual. L_t is a tangent operator. The residual quantities are defined by the following expressions:

$$\begin{cases} R^S = -L(U_0^S) - Q(U_0^S, U_0^S) + \lambda_0^S F \text{ in } \Omega \\ R^a = -L(V_0^a) - Q(V_0^a, U_0^S) - Q(U_0^S, V_0^a) + \omega_0 M(V_0^b) \text{ in } \Omega \\ R^b = -L(V_0^b) - Q(V_0^b, U_0^S) - Q(U_0^S, V_0^b) - \omega_0 M(V_0^a) \text{ in } \Omega \end{cases} \quad (26)$$

In the relations (25), the tangent operator L_t has been introduced. This operator is defined by:

$$L_t(\bullet) = L(\bullet) + Q(\bullet, U_0^S) + Q(U_0^S, \bullet) \quad (27)$$

Appendix B. Range of validity of asymptotic expansion and Padé approximants

Once the polynomial approximation (15) is computed by solving the linear systems (17) and (18), one has to define its range of validity. This forwards defining a new starting point inside this domain of validity and finally by reapplying the perturbation method from this point to determine the following part of the solution.

In the case of polynomial approximation, a very simple criterion has been proposed in Ref. [13]. The idea is to require that the difference between two consecutive order solutions is smaller than a chosen small parameter ϵ (typically $10^{-3} \leq \epsilon \leq 10^{-6}$):

$$\frac{\|V^a(N) - V^a(N-1)\|}{\|V^a(N) - V_0^a\|} = \frac{\|\hat{\omega}^N V_N^a\|}{\|\hat{\omega} V_1^a + \dots + \hat{\omega}^N V_N^a\|} \leq \epsilon \quad (28)$$

where $V^a(N)$ is the series expansions (15) computed at the order of truncature $p = N$. By considering that the denominator of the previous relation can be estimated by $\|\hat{\omega} V_1^a\|$, this leads to the following definition of the maximum value, $\hat{\omega}_{\max \text{ Poly}}$, of the perturbation parameter $\hat{\omega}$:

$$\hat{\omega}_{\max \text{ Poly}} = \left(\epsilon \frac{\|V_1^a\|}{\|V_N^a\|} \right)^{1/(N-1)} \quad (29)$$

In Ref. [22], the authors have extended the criterion (28) to the Padé approximants (19). By requiring that the difference between two rational solutions (19) at consecutive orders remains smaller at the

end of the step than a chosen parameter, δ , the maximum value, $\hat{\omega}_{\max \text{ Padé}}$, of the perturbation parameter $\hat{\omega}$ can be defined:

$$\delta = \frac{\|V_{\text{Padé},N}^a(\hat{\omega}_{\max \text{ Padé}}) - V_{\text{Padé},(N-1)}^a(\hat{\omega}_{\max \text{ Padé}})\|}{\|V_{\text{Padé},N}^a(\hat{\omega}_{\max \text{ Padé}}) - V_0^a\|} \quad (30)$$

One can remark that with the previous relation, we do not have an explicit relation of the validity range of the Padé approximants, contrariwise to polynomial approximations with the relation (29). In fact, the maximum value of the perturbation parameter, $\hat{\omega}_{\max \text{ Padé}}$, has to be evaluated with expression (30) in the range $[\hat{\omega}_{\max \text{ Poly}}, r_1]$, where r_1 designates the first pole of the Padé approximants.

Finally, the relations (29) and (30) permit defining an automatic continuation procedure by successively reapplying the perturbation method in a step-by-step manner. Let us remark that the two parameters ϵ and δ play the same role: they govern the step length and the accuracy, respectively of the asymptotic expansions and the Padé approximants.

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