

Sensitivity analysis of limit cycles of Navier–Stokes equations by the Harmonic–Balance method

An application case with PETSc/SLEPC

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Motivation

Motivation

Time-periodic evolutionary PDE

$$\begin{cases} \mathbf{M} \frac{\partial \mathbf{q}}{\partial t} = \mathbf{F}(\mathbf{q}, \nu) \\ \mathbf{q}(t + T) = \mathbf{q}(t) \end{cases} \quad (1)$$

Two cases of interest:

- Incompressible Navier–Stokes equations: $\mathbf{q} = [\mathbf{u}, p]$
- Compressible Navier–Stokes equations: $\mathbf{q} = [\rho, \mathbf{u}, T, p]$

p' at the far-field

p' at the near-field

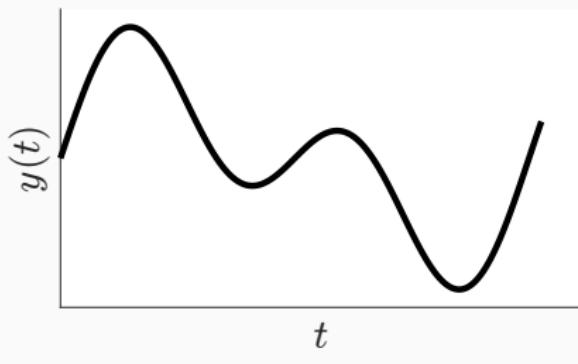
Motivation (II) – A dynamical system detour

Time-periodic evolutionary PDE

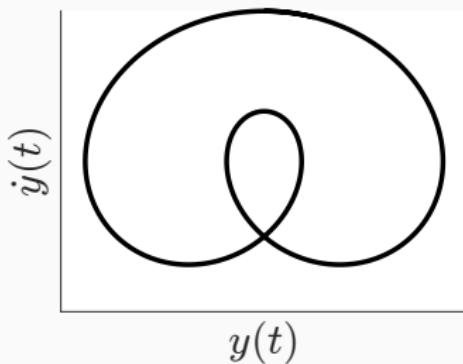
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The periodic solution \mathbf{q} can be
stable or *unstable*

Time evolution of the solution



Phase portrait



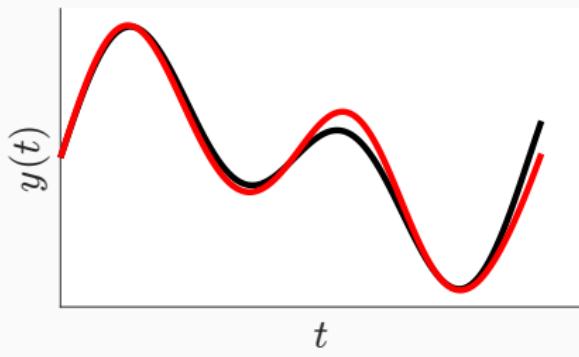
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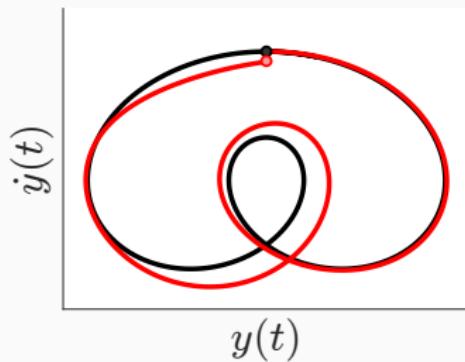
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Motivation (III) – Interests in (unstable) periodic solutions

- Continuation past *self-sustained* instabilities.
- Detection and continuation past instabilities of limit cycles.
- Determination of the *sensitivity quantities*, e.g. the gradient of a property of the system with respect to a control parameter ¹.
- Identification of properties of chaotic systems, e.g. *cycle expansions*², *Lyapunov exponents*³.

¹(Giannetti et al. 2019).

²(Cvitanovic, 1988, Artuso, 1990a,b).

³(Ott 1993).

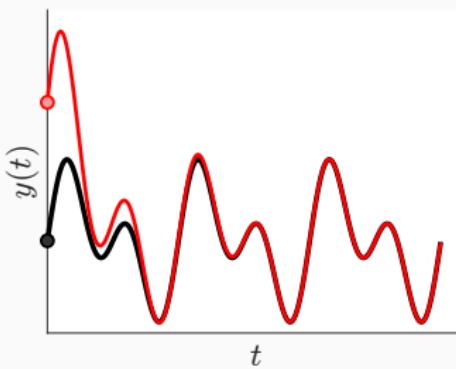
Motivation (IV) – Computing (unstable) periodic solutions

Methods for the computation of
the periodic solution

- **Explicit**
 - Time marching
 - Shooting
 - Multi-shooting
- **Implicit**

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$$\mathbf{q}(t) = \mathcal{S}^t \mathbf{q}(t_0) \quad \text{stable if } ||\mathcal{S}^t|| < 1$$

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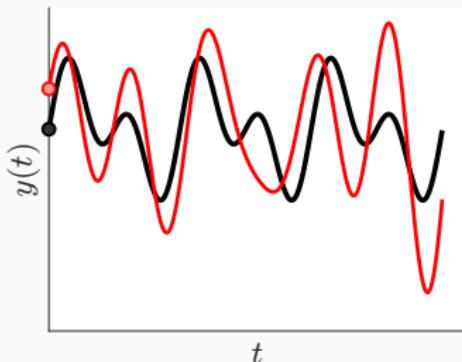
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unstable → does not converge
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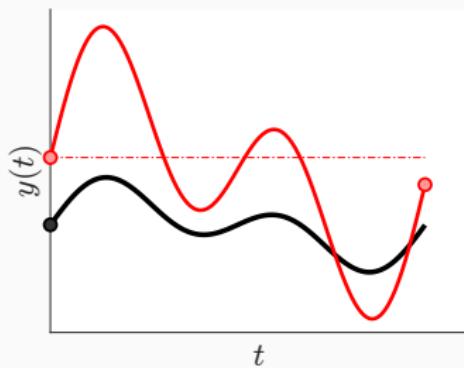
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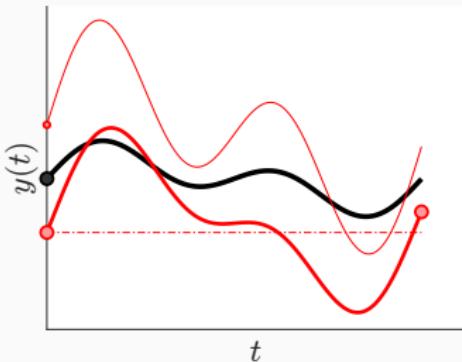
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A dichotomy process until
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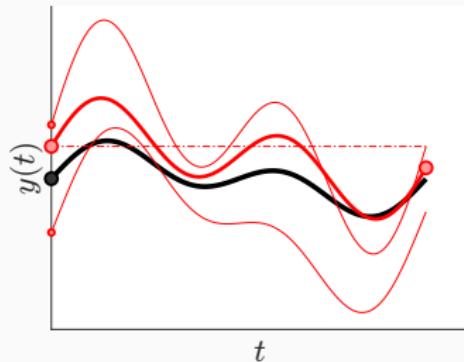
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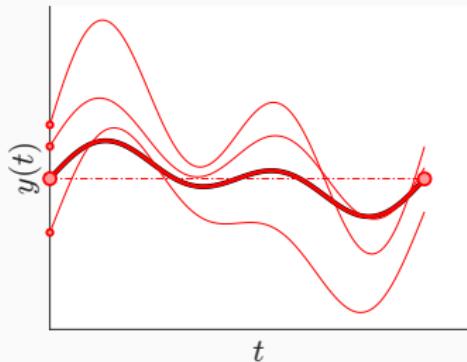
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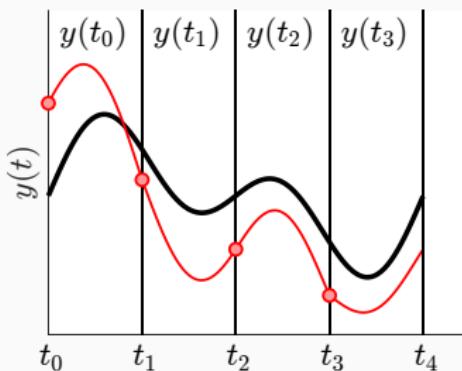
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- **Implicit**

- Collocation, finite
differences, orthogonal
collocation

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Divide and conquer strategy. The
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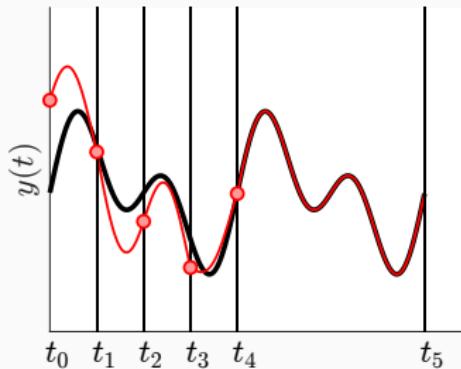
- Time marching
- Shooting
- Multi-shooting . It requires to stabilize the system if *unstable*

- **Implicit**

- Collocation, e.g. finite differences, orthogonal collocation
- Spectral differentiation in the frequency or time domain

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Subdivide the interval $[t_0, t_0 + T_0]$ into N intervals:

$$t_0 < \tau_1 < \tau_2 < \cdots < \tau_N = t_0 + T_0$$

Solve the system:

$$\begin{cases} \mathbf{M}(\mathbf{q}^{j+1} - \mathbf{q}^{j-1}) - \Delta\tau_j T_0 \mathbf{F}(\tilde{\mathbf{q}}^j) = 0, \\ \mathbf{q}^N - \mathbf{q}^0 = 0, \\ \Psi[\mathbf{q}^0, \mathbf{q}^1, \dots, \mathbf{q}^N] = 0 \end{cases}$$

$$\Delta\tau_j = (\tau_{j+1} - \tau_{j-1}) \text{ and} \\ \tilde{\mathbf{q}}^j = \frac{1}{2}(\mathbf{q}^{j+1} + \mathbf{q}^j)$$

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Methods for the computation of
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- **Explicit**

– Direct time integration

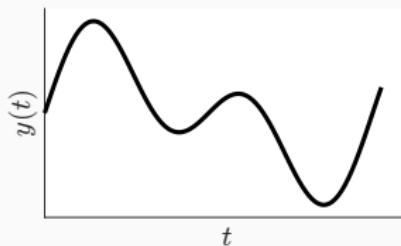
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- Collocation, e.g. finite differences, orthogonal collocation

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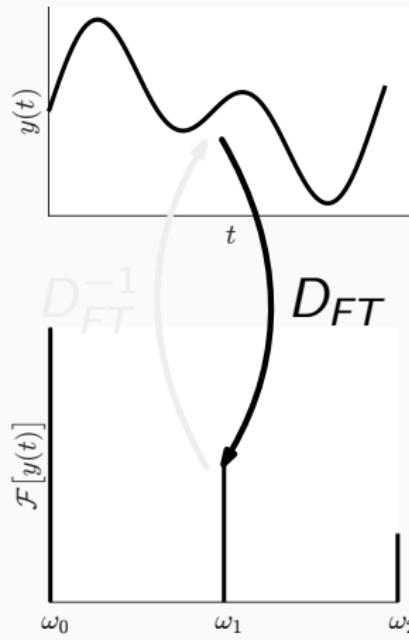
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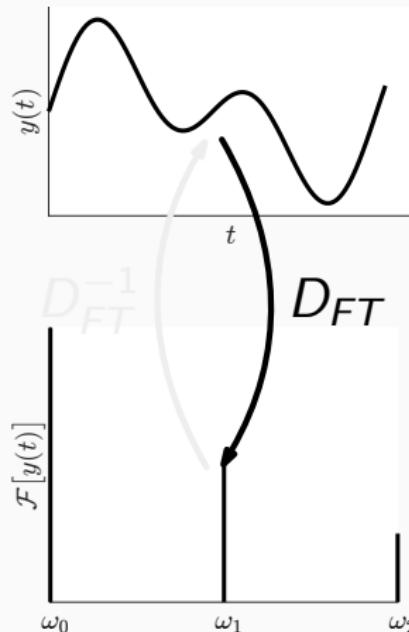
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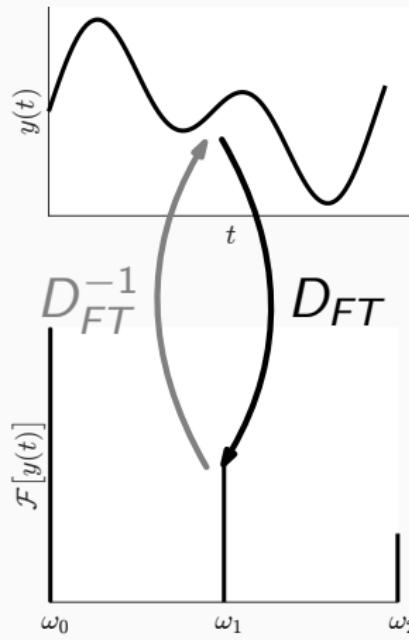
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Motivation (V) – Once computed, can we control it?

- Add a perturbation term into the governing equations, e.g. passive control.⁴

$$\begin{cases} \mathbf{M} \frac{\partial(\mathbf{q} + \delta\mathbf{q})}{\partial t} = \mathbf{F}(\mathbf{q} + \delta\mathbf{q}, \nu) + \delta\mathbf{H}(\mathbf{q} + \delta\mathbf{q}) \\ \mathbf{q}(t + T) = \mathbf{q}(t). \end{cases}$$

- If $\delta\mathbf{q} = \epsilon \hat{\mathbf{q}} e^{\sigma t}$, with $0 < \text{Re}(\sigma)$. The control is determined by
 - *Structural sensitivity* \bar{S}_s , i.e. sensitivity to perturbations whose Floquet exponent is λ .
 - *Baseflow sensitivity* \bar{S}_b , i.e. sensitivity to variations of the periodic orbit itself.
 - *Frequency sensitivity* \bar{S}_ω , i.e. sensitivity to variations of the frequency ω (period T).
- Unstable solutions far from equilibrium can be also stabilized, but that is another story . . . , e.g. shadowing sensitivity from UPOs⁵

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Overview

1. Motivation
2. Computation of periodic solutions with spectral methods
3. Stability of periodic solutions
4. Some insights on sensitivity of periodic solutions
5. A FreeFem example
6. Conclusions

Computation of periodic solutions with spectral methods

Spectral methods for limit cycles (I)

The residual is defined as

$$\begin{cases} \mathbf{R} : X \times X \rightarrow X, \quad \text{with } X \text{ a suitable Hilbert space,} \\ \mathbf{R}(\mathbf{q}, \frac{\partial \mathbf{q}}{\partial t}) = \mathbf{M} \frac{\partial \mathbf{q}}{\partial t} - \mathbf{F}(\mathbf{q}). \end{cases}$$

And the solution vector \mathbf{q} is projected onto a periodic basis,

$$\begin{aligned} \pi_N : X \times \mathbb{R} &\rightarrow X \times (\mathbb{Z}/(2N+1)\mathbb{Z}) \\ \pi_N(\mathbf{q}) &= \mathbf{q}_h(t) = \mathbf{q}_0 + \sum_{n=1}^N [\mathbf{q}_{c,n} \cos(n\omega t) + \mathbf{q}_{s,n} \sin(n\omega t)] \\ &= \underbrace{[\mathbf{q}_0, \mathbf{q}_{1,c}, \mathbf{q}_{1,s}, \dots, \mathbf{q}_{N,c}, \mathbf{q}_{N,s}]}_{(\check{\mathbf{Q}}^{(\tau,N)})^T} \underbrace{[1, \cos(\omega t), \sin(\omega t), \dots, \cos(N\omega t), \sin(N\omega t)]}_{\mathcal{F}_N} \end{aligned} \tag{2}$$

A key difference with *collocation methods* is the fact that periodicity is imposed to the basis and not as a constraint.

Spectral methods for limit cycles (II)

Subdivide the interval $[t_0, t_0 + T_0]$ into M intervals with $\tau_m = \frac{2\pi m}{M}$ for $m = 1, \dots, M-1$, $t_0 = \tau_1 < \tau_2 < \dots < \tau_M < t_0 + T_0$

$$\check{\mathbf{R}}^{(\tau, N)} \approx \frac{2}{M} \underbrace{\begin{pmatrix} \frac{1}{2} & \dots & \frac{1}{2} \\ \cos\left(\frac{2\pi(1)(0)}{M}\right) & \dots & \cos\left(\frac{2\pi(1)(M-1)}{M}\right) \\ \vdots & & \vdots \\ \sin\left(\frac{2\pi(N)(0)}{M}\right) & \dots & \sin\left(\frac{2\pi(N)(M-1)}{M}\right) \end{pmatrix}}_{D_{FT}} \otimes \mathbf{I} \begin{pmatrix} \mathbf{R}|t = \tau_1 \\ \vdots \\ \mathbf{R}|t = \tau_{M-1} \end{pmatrix}$$

$$\mathbf{R}^{(\tau, M)} \approx \underbrace{\begin{pmatrix} 1 & \cos\left(\frac{2\pi(1)(0)}{M}\right) & \dots & \sin\left(\frac{2\pi(0)(N)}{M}\right) \\ \vdots & \vdots & & \vdots \\ 1 & \cos\left(\frac{2\pi(M-1)(1)}{M}\right) & \dots & \sin\left(\frac{2\pi(M-1)(N)}{M}\right) \end{pmatrix}}_{D_{FT}^{-1}} \otimes \mathbf{I} \begin{pmatrix} \check{\mathbf{R}}_0 \\ \check{\mathbf{R}}_{1,c} \\ \vdots \\ \check{\mathbf{R}}_{N,s} \end{pmatrix}$$

Also note that the time derivative is easy to evaluate in the frequency domain:

$$\begin{aligned}
 \mathbf{M} \frac{\partial \mathbf{q}}{\partial t} &\approx \mathbf{M} \frac{\partial \pi_N(\mathbf{q})}{\partial t} \\
 &= \sum_{n=1}^N n\omega \mathbf{M} [-\mathbf{q}_{c,n} \sin(n\omega t) + \mathbf{q}_{s,n} \cos(n\omega t)] \\
 &= \boldsymbol{\Omega} \check{\mathbf{Q}}^{(\tau, N)}
 \end{aligned}$$

Four kinds of spectral methods

Fourier-Galerkin -- HB

$$\check{\mathbf{R}}(\check{\mathbf{Q}}^{\tau, N}, \boldsymbol{\Omega} \check{\mathbf{Q}}^{\tau, N}) = 0$$

AFT Fourier-Galerkin

$$D_{FT} \mathbf{R}(D_{FT}^{-1} \check{\mathbf{Q}}^{\tau, N}, D_{FT}^{-1} \boldsymbol{\Omega} \check{\mathbf{Q}}^{\tau, N}) = 0$$

Trigonometric collocation

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Time spectral

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Four kinds of spectral methods

Fourier-Galerkin -- HB

$$\begin{aligned}
 \check{\mathbf{R}}(\check{\mathbf{Q}}^{\tau, N}, \boldsymbol{\Omega} \check{\mathbf{M}} \check{\mathbf{Q}}^{\tau, N}) &= 0 \\
 \text{AFT Fourier-Galerkin} \\
 D_{FT} \mathbf{R}(D_{FT}^{-1} \check{\mathbf{Q}}^{\tau, N}, D_{FT}^{-1} \boldsymbol{\Omega} \check{\mathbf{M}} \check{\mathbf{Q}}^{\tau, N}) &= 0
 \end{aligned}$$

Trigonometric collocation

$$\mathbf{R}(D_{FT}^{-1} \check{\mathbf{Q}}^{\tau, N}, D_{FT}^{-1} \boldsymbol{\Omega} \check{\mathbf{M}} \check{\mathbf{Q}}^{\tau, N}) = 0$$

*Frequency
unknowns*

Time-Fourier

$$D_{FT} \mathbf{R}(\mathbf{Q}^{\tau, M}, \mathbf{D}_{FT}^{-1} \boldsymbol{\Omega} \mathbf{M} D_{FT} \mathbf{Q}^{\tau, M}) = 0$$

Frequency residual

Time spectral

$$\mathbf{R}(\mathbf{Q}^{\tau, M}, \mathbf{D}_{FT}^{-1} \boldsymbol{\Omega} \mathbf{M} D_{FT} \mathbf{Q}^{\tau, M}) = 0$$

*Time
unknowns*

Four kinds of spectral methods

- Free of aliasing ✓
- Low CPU time ✓
- Intensive memory consumption ✗
- Evaluation of analytic residual ⚙

Fourier-Galerkin -- HB

$$\check{\mathbf{R}}(\check{\mathbf{Q}}^{\tau,N}, \Omega \mathbf{M} \check{\mathbf{Q}}^{\tau,N}) = 0$$

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Time spectral

$$\mathbf{R}(\mathbf{Q}^{\tau,M}, \mathbf{D}_{FT}^{-1} \Omega \mathbf{M} D_{FT} \mathbf{Q}^{\tau,M}) = 0$$

Frequency residual

Time residual

- Inherent aliasing ✗ *It can be reduced increasing the number of snapshots M, but higher memory consumption ...*
- Low CPU time ✓
- Intensive memory consumption ✗
- Evaluation of analytical time-derivative ✓

Frequency unknowns

Time unknowns



Spectral methods for limit cycles (III)

Fourier–Galerkin or Harmonic–Balance is a Galerkin method:

$$\begin{aligned}\check{\mathbf{R}} : X \times (\mathbb{Z}/(2N+1)\mathbb{Z}) &\rightarrow X \times (\mathbb{Z}/(2N+1)\mathbb{Z}) \\ \check{\mathbf{R}}(\check{\mathbf{Q}}^{(\tau, N)}, \omega) &= \int_0^{\frac{2\pi}{\omega}} \check{\mathbf{R}}(\mathbf{q}_h, \frac{\partial \mathbf{q}_h}{\partial t}, t)^T \mathcal{F}_N dt = \mathbf{0},\end{aligned}\tag{3}$$

⁶Fourier coefficients of the non-linear operator detailed in (Sierra et al. 2021).

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if $\mathbf{F}(\mathbf{q}) = \mathbf{L}\mathbf{q} + \mathbf{N}(\mathbf{q})$ where \mathbf{L} is a linear operator and \mathbf{N} is a non-linear operator, then *Harmonic–Balance equations* ($\check{\mathbf{R}}(\check{\mathbf{Q}}^{(\tau,N)}, \omega) = 0$)⁶

$$\begin{aligned}0 &= \mathbf{L}\mathbf{q}_0 + \mathbf{N}_0 \\ n\omega \mathbf{M}\mathbf{q}_{n,s} &= \mathbf{L}\mathbf{q}_{n,c} + \mathbf{N}_{n,c}, \quad n = 1, \dots, N \\ -n\omega \mathbf{M}\mathbf{q}_{n,c} &= \mathbf{L}\mathbf{q}_{n,s} + \mathbf{N}_{n,s}, \quad n = 1, \dots, N \\ \Psi(\mathbf{q}_h) &= 0.\end{aligned}\tag{4}$$

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Resolution of an implicit non-linear problem \longrightarrow *Newton–Krylov* method

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Spectral methods for limit cycles (IV)

Newton–Krylov iteration

At each step of the procedure, starting from an estimate $\check{\mathbf{Q}}_n^{(\tau, N)}, \omega_n$ of the solution, we look for an improved estimate defined as $\check{\mathbf{Q}}_{n+1}^{(\tau, N)} = \check{\mathbf{Q}}_n^{(\tau, N)} + \delta \check{\mathbf{Q}}_n^{(\tau, N)}$, $\omega_{n+1} = \omega_n + \delta \omega_n$, from the linear system

$$D\check{\mathbf{R}} \left[\delta \check{\mathbf{Q}}_n^{(\tau, N)}, \delta \omega \right]^T = - \left[\check{\mathbf{R}}(\check{\mathbf{Q}}_n^{(\tau, N)}), \Psi(\check{\mathbf{Q}}_n^{(\tau, N)}) \right]^T$$

Solved with a Krylov-method + preconditioning,

$$\mathbf{P}^{-1} D\check{\mathbf{R}} \left[\delta \check{\mathbf{Q}}_n^{(\tau, N)}, \delta \omega \right]^T = - \mathbf{P}^{-1} \left[\check{\mathbf{R}}(\check{\mathbf{Q}}_n^{(\tau, N)}), \Psi(\check{\mathbf{Q}}_n^{(\tau, N)}) \right]^T$$

Choices for \mathbf{P} : Block Jacobi, Gauss Seidel or block–circulant (*not explored yet*). Inner blocks solved by a direct method or preconditioning again, e.g. ASM, Augmented Lagrangian⁷

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Stability of periodic solutions

Some insights in stability of limit cycles

Consider the periodic solution $\mathbf{q}^*(t + T) = \mathbf{q}^*(t)$,

Add a small perturbation $\delta\mathbf{q}(t)$

What is the response of the system? *stable* or *unstable*?

Answer: Floquet theory,

$$\delta\mathbf{q}(t) = \sum_{n=1}^N c_n \delta\mathbf{q}_n(t) \text{ where } \delta\mathbf{q}_n(t) = e^{\lambda_n t} \mathbf{v}_n(t),$$

here \mathbf{v}_n is a T -periodic vector and λ_n are called the *Floquet exponents*.

Thus, the stability of the system is determined by the following T -periodic linear problem,

$$\lambda_n \mathbf{M}\mathbf{v}_n = \left[-\mathbf{M} \frac{\partial}{\partial t} + \frac{\partial \mathbf{F}(\mathbf{q}^*, \lambda)}{\partial \mathbf{q}} \right] \mathbf{v}_n.$$

and the periodic solution \mathbf{q} is *stable* if $\operatorname{Re}(\lambda_n) < 0$ and *unstable* if $\operatorname{Re}(\lambda_n) > 0$.

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Floquet exponents λ_n and *Floquet vectors* \mathbf{v}_n are determined by solving a generalized eigenvalue problem:

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here $\mathbf{V}^{(\tau, N)} = [\mathbf{v}_0, \mathbf{v}_{1,c}, \mathbf{v}_{1,s}, \dots, \mathbf{v}_{N,c}, \mathbf{v}_{N,s}]^T$.

The same preconditioning techniques are adopted.

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Some insights on sensitivity of periodic solutions

Sensitivity of periodic solutions (I)

Consider the case where $|\text{Re}(\lambda)| \ll 1$, it is said that the system is near an *instability*. In these cases we may wonder if we can control the system

$$\mathbf{M} \frac{\partial(\mathbf{q}^* + \delta\mathbf{q})}{\partial t} = \mathbf{F}(\mathbf{q}^* + \delta\mathbf{q}, \nu) + \delta\mathbf{H}(\mathbf{q}^* + \delta\mathbf{q}), \quad \mathbf{q}^*(t + T) = \mathbf{q}^*(t).$$

by the introduction of a force-feedback perturbation $\delta\mathbf{H}(\mathbf{q}^* + \delta\mathbf{q})$ ⁸.

Consider a localised feedback $\mathbf{H}(\mathbf{q}^* + \delta\mathbf{q}) = \delta(\mathbf{x} - \mathbf{x}_0)\mathbf{C} \cdot \mathbf{B}(\mathbf{q}^* + \delta\mathbf{q})$ where \mathbf{B} is a input/output matrix and \mathbf{C} is a gain matrix.

The variation of the *Floquet exponent* is determined by

$$\delta\lambda = \mathbf{C} : \mathbf{S}_{\text{tot}}$$

here $\mathbf{S}_{\text{tot}} = \mathbf{S}_b + \mathbf{S}_s$, the *baseflow* and *structural sensitivities*.

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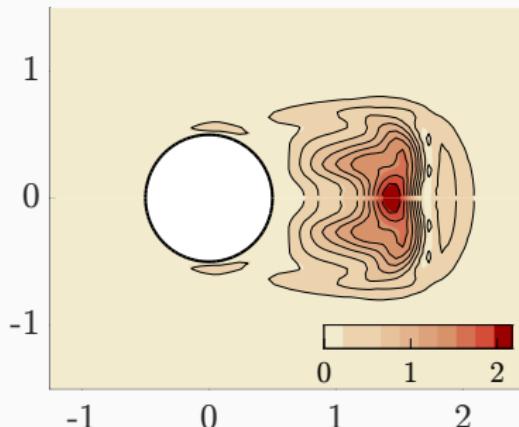
Sensitivity of periodic solutions (II)

Sensitivity maps are determined from the periodic solution \mathbf{q} , direct and adjoint Floquet modes \mathbf{v} and \mathbf{v}^\dagger ,

$$S_s(\mathbf{x}, t) = \frac{\mathbf{Bv} \otimes \mathbf{Bv}^\dagger}{\int_t^{T+t} \int_{\Omega} \mathbf{Bv} \cdot \mathbf{Bv}^\dagger d\mathbf{x} dt}, \quad \bar{S}_s(\mathbf{x}) = \frac{\int_t^{T+t} \mathbf{Bv} \otimes \mathbf{Bv}^\dagger}{\int_t^{T+t} \int_{\Omega} \mathbf{Bv} \cdot \mathbf{Bv}^\dagger d\mathbf{x}}.$$

Instantaneous structural sensitivity

Averaged structural sensitivity



A FreeFem example

Computation of periodic solutions with FreeFem(I)

Example: Harmonic-Balance of Navier-Stokes equations with $N = 1$

The governing equations are quadratic in the state variables $\mathbf{q} = [\mathbf{u}, p]$,

$$\mathbf{M} \frac{\partial \mathbf{q}}{\partial t} = \mathbf{L}\mathbf{q} + \mathbf{N}(\mathbf{q}, \mathbf{q})$$

with the linear operator

$$\begin{aligned} \mathbf{L}\mathbf{q} \equiv & \frac{\partial \mathbf{u}}{\partial t} + \frac{1}{2} [\mathbf{u}_0 \cdot \nabla \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u}_0] + \nabla p - \nabla \cdot \tau(\mathbf{u}) \\ & \nabla \cdot \mathbf{u}, \end{aligned}$$

and the non-linear quadratic operator

$$\mathbf{N}(\mathbf{q}_1, \mathbf{q}_2) \equiv \mathbf{u}_1 \cdot \nabla \mathbf{u}_2 + \mathbf{u}_1 \cdot \nabla \mathbf{u}_2$$

Thus, the governing HB equations are as follows,

$$\begin{aligned} 0 &= \mathbf{L}\mathbf{q}_0 + \frac{1}{4} [\mathbf{N}(\mathbf{q}_{1,c}, \mathbf{q}_{1,c}) + \mathbf{N}(\mathbf{q}_{1,s}, \mathbf{q}_{1,s})] \\ \omega \mathbf{M}\mathbf{q}_{1,s} &= \mathbf{L}\mathbf{q}_{1,c}, \\ -\omega \mathbf{M}\mathbf{q}_{1,c} &= \mathbf{L}\mathbf{q}_{1,s}, \\ \Psi(\mathbf{q}_h) &= 0. \end{aligned}$$

Computation of periodic solutions with FreeFem (II)

```
fespace XXMh(th,[P2,P2,P1]);  
XXMh defQ(u0),defQ(u1c),defQ(u1s);
```

```
macro NSL(nu,u,v) (-2*nu*(D(u):D(v)) + u#p*div(v)  
+ v#p*div(u) - .5*Conv(U,u,v) ) //EOM. Linear operator  
macro F20(u,v,w) Conv(u,v,w) // EOM. Nonlinear operator
```

```
varf vDROx0(defQ(du),defQ(v)) =  
int2d(th)( NSL(nu,u0,du,v) ) + BCO(du,v);  
varf vDROx1c(defQ(du),defQ(v))=  
int2d(th)( -0.5*F20(u1c,du,v) );  
varf vDROx1s(defU(du),defU(v))=  
int2d(th)( -0.5*F20(u1s,du,v) );
```

```
matrix<real> ffDROx0 = vDROx0(XXMh,XXMh);  
matrix<real> ffDROx1c = vDROx1c(XXMh,XXMh);  
matrix<real> ffDROx1s = vDROx1s(XXMh,XXMh);  
Mat<real> DROx0(0,ffDROx0,intersection,DDD, bs=1);  
Mat<real> DROx1; // MatNest  
DROx1 = [ [ffDROx1c, ffDROx1s, matNull] ];
```

```
varf vRHS0(defQ(du),defQ(v))= int2d(th)(NSL(nu,u0,u0,v))  
+ BCO(du,v)+ int2d(th)(-0.25*F20(u1c,u1c,v)  
-0.25*F20(u1s,u1s,v) );  
real[int] RHS0(u0x[].n); RHS0 = vRHS0(0,XXMh);  
real[int] RHS = [RHSFO, RHFic, RHFis, RHSphase];
```

```
KSPSSolve(funcMatVec,RHS,dSol,  
precon=funcGS,  
sparams = "-ksp_type gmres -ksp_monitor  
-ksp_view_final_residual -ksp_converged_reason  
-ksp_gmres_restart 1000 -pc_type shell");
```

```
func real[int] funcMatVec(real[int] & x)  
{ int n = x.n;  
real[int] u(n);u=0;  
u(0:(n-1)) += DROx0*x(0:(n-1));  
u(0:(n-1)) += DROx1* x(n:(3*n));  
u(n:(3*n)) += DR1x0* x(0:(n-1));  
u(n:(3*n)) += DR1x1*x(n:(3*n));  
return u; }
```

```
func real[int] funcGS(real[int] & x)  
{ int n = x.n;  
real[int] u(x.n); u=0;  
real[int] v(x.n); v=0;  
real[int] w(x.n); w=0;  
v(n:(3*n)) = DR1x1^-1*VecIN(n:(3*n));  
u(n:(3*n)) += v(nDOF:(3*nDOF));  
v(0:(n-1)) = DROx1* u(nDOF:(3*nDOF));  
w(0:(n-1)) = -v(0:(n-1));  
w(0:(n-1)) += x(0:(n-1));  
v(0:(n-1)) = DROx0^-1*w(0:(n-1));  
u(0:(n-1)) += v(0:(n-1));  
return u; }
```

Stability of periodic solutions with FreeFem

```
string DRParams = "-ksp_type gmres -ksp_view_final_residual  
-ksp_converged_reason -ksp_gmres_restart 1000 -pc_type shell ";  
set(DR, sparams = DRParams, precon = funcGS);  
  
ssparams =  
" -eps_nev " + nEig /* Number of eigenvalues */  
+" -eps_type krylovSchur /* Type of Eigen Problem Solver */  
+ " -eps_target " + shiftValue /* Shift value */  
+" -eps_gen_non_hermitian /* Not hermitian ( $A^* \neq A$ ) by default */  
+" -st_type sinvert /* Spectral Transformation */  
+" -st_ksp_type gmres /* Iterative lin. solver */  
+" -eps_monitor "  
+" -eps_view ";  
  
nComputed = EPSSolve  
(DR, /* matrix OP = DR - shift M */  
M, /* M matrix */  
array = EigenVEC, /* Array to store the FEM-EigenFunctions */  
values = EigenVAL, /* Array to store the EigenValues */  
sparams = ssparams /* Parameters for the distributed EigenValue solver */  
);
```

Conclusions

Conclusions and perspectives

• Conclusions

- Set of *spectral* tools for the computation of periodic solutions and the determination of its *stability* and the *sensitivity* to force-feedback terms. Some tools available in *StabFem*⁹.
- This strategy was implemented in FreeFem using the linear algebra brackends PETSc/SLEPc¹⁰.
- Implemented for *incompressible* and *compressible* flows. Large speed-up in CPU time for *low Mach* flows (acoustics).
- Large speed-up for the computation of *sensitivity* maps in comparison with classical time marching methods.

• Perspectives

- Block Toeplitz preconditioning to boost GMRES convergence with high number of Fourier modes.
- Extension to other flow configurations: three-dimensional flows in cylindrical coordinates, free-surface flows, fluid-structure ...

⁹Sierra et al., Journal of Computational Physics, 2021 (under review)

¹⁰<https://stabfem.gitlab.io/StabFem/>

Questions?

Adjoint-based sensitivity analysis of periodic orbits by
the Fourier–Galerkin method

J. Sierra^{a,b}, P. Jolivet^c, F. Giannetti^b, V. Citro^b

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^b*Dipartimento di Ingegneria (DIIN), Università degli Studi di Salerno, Fisciano 84084, Italy*

^c*CNRS-IRIT, 2 rue Charles Camichel, 31071 Toulouse Cedex 7, France*

Efficient computation of periodic compressible flows
with spectral techniques

Javier Sierra^{a,b}, Vincenzo Citro^a, Flavio Giannetti^a, David Fabre^b

^a*DIIN, Via Giovanni Paolo II, 132, 84084 Fisciano SA, Italy*

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Quadratic non-linear Fourier coefficients

$$\begin{aligned}\mathbf{N}_0 &= \mathbf{N}(\mathbf{q}_0, \mathbf{q}_0) + \frac{1}{2} \sum_{i=1}^N \mathbf{N}(\mathbf{q}_{i,s}, \mathbf{q}_{i,s}) + \mathbf{N}(\mathbf{q}_{i,c}, \mathbf{q}_{i,c}) \\ \mathbf{N}_{i,c} &= [\mathbf{N}(\mathbf{q}_{i,c}, \mathbf{q}_0) + \mathbf{N}(\mathbf{q}_0, \mathbf{q}_{i,c})] \\ &\quad + \frac{1}{2} \sum_{j=1}^{i-1} [\mathbf{N}(\mathbf{q}_{j,c}, \mathbf{q}_{i-j,c}) - \mathbf{N}(\mathbf{q}_{j,s}, \mathbf{q}_{i-j,s})] \\ &\quad + \frac{1}{2} \sum_{j=i+1}^N [\mathbf{N}(\mathbf{q}_{j,c}, \mathbf{q}_{j-i,c}) + \mathbf{N}(\mathbf{q}_{j-i,s}, \mathbf{q}_{j,s})] \\ &\quad + \frac{1}{2} \sum_{j=i+1}^N [\mathbf{N}(\mathbf{q}_{j-i,c}, \mathbf{q}_{j,c}) + \mathbf{N}(\mathbf{q}_{j,s}, \mathbf{q}_{j-i,s})] \\ \mathbf{N}_{i,s} &= [\mathbf{N}(\mathbf{q}_{i,s}, \mathbf{q}_0) + \mathbf{N}(\mathbf{q}_0, \mathbf{q}_{i,s})] \\ &\quad + \frac{1}{2} \sum_{j=1}^{i-1} [\mathbf{N}(\mathbf{q}_{j,c}, \mathbf{q}_{i-j,s}) + \mathbf{N}(\mathbf{q}_{j,s}, \mathbf{q}_{i-j,c})] \\ &\quad - \frac{1}{2} \sum_{j=i+1}^N [\mathbf{N}(\mathbf{q}_{j,c}, \mathbf{q}_{j-i,s}) + \mathbf{N}(\mathbf{q}_{j-i,s}, \mathbf{q}_{j,c})] \\ &\quad + \frac{1}{2} \sum_{j=i+1}^N [\mathbf{N}(\mathbf{q}_{j-i,c}, \mathbf{q}_{j,s}) + \mathbf{N}(\mathbf{q}_{j,s}, \mathbf{q}_{j-i,c})].\end{aligned}$$

Matrices of Fourier–Galerkin

$$\tilde{\mathbf{B}}\mathbf{Q}_N = \begin{bmatrix} \mathbf{0} & & & \\ & \mathbf{B}_1 & & \\ & & \ddots & \\ & & & \mathbf{B}_N \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \vdots \\ \mathbf{q}_N \end{bmatrix} \text{ with } \mathbf{B}_n = \begin{bmatrix} \mathbf{0} & n\mathbf{B} \\ -n\mathbf{B} & \mathbf{0} \end{bmatrix} \text{ for } n = 1, \dots, N \quad (5)$$

$$\tilde{\mathbf{L}}\mathbf{Q}_N = \begin{bmatrix} \mathbf{L} & & & \\ & \mathbf{L} & & \\ & & \ddots & \\ & & & \mathbf{L} \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \vdots \\ \mathbf{q}_N \end{bmatrix} \text{ with } \mathbf{q}_n = \begin{bmatrix} \mathbf{q}_{n,c} \\ \mathbf{q}_{n,s} \end{bmatrix} \text{ for } n = 1, \dots, N. \quad (6)$$

$$\tilde{\mathbf{N}}(\mathbf{Q}_N, \mathbf{Q}_N) = \begin{bmatrix} \mathbf{N}_0 \\ \mathbf{N}_1 \\ \vdots \\ \mathbf{N}_N \end{bmatrix} \text{ with } \mathbf{N}_n = \begin{bmatrix} \mathbf{N}_{n,c} \\ \mathbf{N}_{n,s} \end{bmatrix} \text{ for } n = 1, \dots, N. \quad (7)$$