

A PARAMETERIZATION METHOD FOR  
COMPUTING NORMALLY HYPERBOLIC  
INVARIANT TORI  
SOME NUMERICAL EXAMPLES

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# Outline

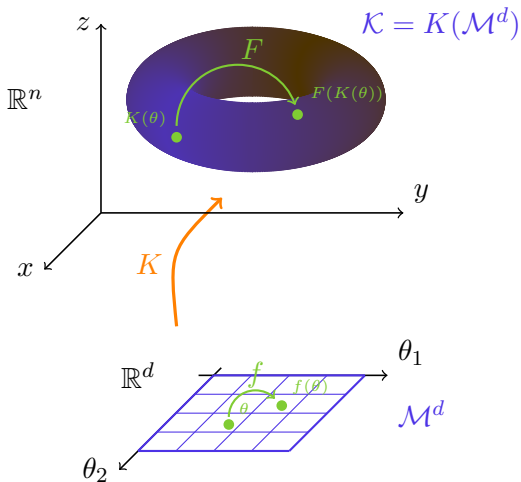
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  - STEP 2: Solve  $DF(K(\theta)N(\theta) - N(f(\theta)))\Lambda^n(\theta) = 0$
  - Improvements in the method
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# DEFINITIONS AND INTRODUCING THE PROBLEM

## The system model

- Consider a map in  $\mathbb{R}^n : F : \mathbb{R}^n \rightarrow \mathbb{R}^n$  (a discrete dynamical system).
- Consider a  $d$ -manifold  $\mathcal{M}^d = \{\mathbb{T}^d, \mathbb{S}^d, \dots\} \subset \mathbb{R}^n$ .  
A **parameterization** of it will be an immersion  $K : \mathcal{M}^d \rightarrow \mathbb{R}^n$  ( $DK$  has maximum rank  $d$ ),  $d < n$ ,  $\mathcal{K} = K(\mathcal{M}^d)$ .
- Let  $f : \mathcal{M}^d \rightarrow \mathcal{M}^d$  be the dynamics of  $F$  restricted over the manifold  $\mathcal{M}^d$ .

# The system model



# The Invariance equation

## Definition

We say that the manifold parameterized by  $K, \mathcal{K}$ , is invariant under  $F$  with internal dynamics  $f$  if  $K$  and  $f$  meets the invariance equation:

$$F \circ K = K \circ f$$

ie: if for each  $\theta \in \mathcal{M}^d$  we have  $F(K(\theta)) = K(f(\theta))$ .

We want to find these functions  $K$  and  $f$ .

# Vector Bundles, hyperbolicity and Normal Hyperbolicity

Consider  $M \subset \mathbb{R}^n$  a manifold.

For one point of the manifold  $M \rightsquigarrow$  **vector spaces**:

- $T_p M$ , tangent space to  $M$  at  $p$ .  
Particulary,  $T_p \mathbb{R}^n = \mathbb{R}^n$  the tangent space to  $\mathbb{R}^n$  at  $p$ .
- $N_p = \{v \in \mathbb{R}^n | v \perp T_p M\}$ , normal space to  $M$  at  $p$ .
- $T_p \mathbb{R}^n = T_p M + N_p$ , as a vectorial spaces sum.

## Vector Bundles, hyperbolicity and Normal Hyperbolicity

For all points (together) of the manifold  $M \rightsquigarrow$  **vector bundles**:  
 a manifold such which have a linear vector space associated to  
 every point of it:

- $TM = \{(p, v) \in M \times \mathbb{R}^n | v \in T_p M\}$ , tangent bundle of  $M$ .
- $N = \{(p, v) \in M \times \mathbb{R}^n | v \perp T_p M\}$ , normal bundle of  $M$ .
- $T\mathbb{R}^n_M = TM \oplus N$ , as a Whitney sum of the vector bundles.

For each  $p \in M$  we have  $T_p M \cong \mathbb{R}^d$  and  $N_p \cong \mathbb{R}^{n-d}$ , called the  
*fibers* of the vector space.



Suppose we have found  $\mathcal{K} = K(\mathcal{M}^d)$  and  $f$ .

- $DK$  generates the tangent space to each point of the invariant manifold,  $T_{K(\theta)}\mathcal{K}$  :

$$(DK)_\theta : T_\theta\mathcal{M}^d \rightarrow T_{K(\theta)}\mathcal{K}$$

for each  $\theta \in \mathcal{M}^d$ ,  $DK(\theta)$  is represented as a  $n \times d$  matrix, a fiber.

Considering all  $\theta \in \mathcal{M}^d$ , we have a parameterization of the tangent bundle,  $T\mathcal{K}$ .

- if  $N(\theta)$  is a  $n \times n - d$  matrix composed by  $n - d$  vectors linearly independent to the vectors of  $DK(\theta)$ , then it generates the normal space  $\mathcal{N}_{K(\theta)}$ , complementary to  $T_{K(\theta)}\mathcal{K}$ . Considering all  $\theta \in \mathcal{M}^d$ , we have a parameterization of the normal bundle,  $\mathcal{N}(\mathcal{K})$ .

So, we could write:

$T_{K(\theta)}\mathbb{R}^n = T_{K(\theta)}\mathcal{K} + \mathcal{N}_{K(\theta)} \cong \mathbb{R}^n$  as a sums of vector spaces.

$T\mathbb{R}^n|_{\mathcal{K}} = T\mathcal{K} \oplus \mathcal{N}(\mathcal{K})$  as a sums of vector bundles.

We could define  $P(\theta) := (DK(\theta)|N(\theta))_{n \times n}$ , a vector bundle which generates the total space. So, it is an adapted frame of if

$$P : \mathcal{M}^d \rightarrow T_{K(\theta)}\mathcal{K} + \mathcal{N}_{K(\theta)}$$

We could also define the matrix map  $\Lambda : \mathcal{M}^d \rightarrow M_{n \times n}$  as the dynamics over the tangent and normal bundles, the linearized dynamics on this frame.

Then,  $\Lambda$  and  $P$  must satisfy the invariance of the splitting on the bundles:

$$P(f(\theta))^{-1}DF(K(\theta))P(\theta) = \Lambda(\theta)$$

### Remark

*The dynamics of the bundles on the model manifold will be of the form*

$$\Lambda(\theta) = \begin{pmatrix} \Lambda^t(\theta) & B(\theta) \\ O & \Lambda^n(\theta) \end{pmatrix}$$

*where  $\Lambda^t$  is the linearized dynamics on the tangent space (a  $d \times d$  matrix),  $\Lambda^n$  is the linearized dynamics on the normal space (a  $n - d \times n - d$  matrix) and  $B(\theta)$  is a  $d \times n - d$  matrix.*

As

$$DF(K(\theta))(DK(\theta)|N(\theta)) = (DK(f(\theta))|N(f(\theta))) \begin{pmatrix} \Lambda^t(\theta) & B(\theta) \\ O & \Lambda_n(\theta) \end{pmatrix} :$$

- If  $B(\theta) = 0$ , then the normal bundle is invariant: the invariance equation is satisfied on the normal subspace

$$DF(K(\theta))N(\theta) = N(f(\theta))\Lambda^n(\theta)$$

- If  $\Lambda^t(\theta) = Df(\theta)$ , then the tangent bundle is invariant: the invariance equation is satisfied on the tangent subspace

$$DF(K(\theta))DK(\theta) = DK(f(\theta))Df(\theta)$$

This is always true, as this equation is just the derivative of invariance equation.

So, we will consider  $B(\theta) = 0$  and  $\Lambda^t(\theta) = Df(\theta)$  to have last two conditions true and have both bundles invariant. In this way, the invariant manifold will be normally hyperbolic.

# THE METHOD

- The algorithm is inspired in the **parameterization method** (*Cabré, Fontich, de la Llave*) for finding a parameterization of the invariant manifold and a dynamics on it.
- The framework leads to **solving invariance equations**, for which one uses a **Newton method** adapted to the dynamics and the geometry of the (invariant) manifold, normally hyperbolic invariant tori.

## Summary of the problem

If we want to find a normally hyperbolic invariant manifold, we are looking for  $K$ ,  $f$ ,  $P$  and  $\Lambda$  such that:

- $F(K(\theta) - K(f(\theta))) = 0$
- $DF(K(\theta)P(\theta) - P(f(\theta))\Lambda(\theta)) = 0$



$$DF(K(\theta)N(\theta) - N(f(\theta))\Lambda^n(\theta)) = 0$$

But as tangent bundle and its dynamics are well defined directly using derivatives of known values  $K$  and  $f$ , so we only have to solve the second equation for the normal part.

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## A newton method to compute $K, f, P, \Lambda$

Given an approximate normally hyperbolic invariant torus  $(K_0, f_0)$  and its bundles  $(P_0, \Lambda_0)$  with error:

- $F(K_0(\theta) - K_0(f_0(\theta))) = R(\theta)_{n \times 1}$
- $DF(K_0(\theta))P_0(\theta) - P_0(f_0(\theta))\Lambda_0(\theta) = S(\theta)_{n \times n}$

We look for  $(H, h, Q, \Delta)$  satisfying

- $-\tilde{R}(\theta) = \Lambda(\theta)H(\theta) - \underbrace{\begin{pmatrix} h(\theta) \\ 0 \end{pmatrix}}_{n \times d} - H(f(\theta))$
- $-\tilde{S}^n(\theta) = \Lambda(\theta)Q(\theta) - Q(f(\theta))\Lambda^n(\theta) - \Delta(\theta)$

And obtain the improved torus:

$$\begin{aligned} K(\theta) &= K_0(\theta) + P_0(\theta)H(\theta), & f(\theta) &= f_0(\theta) + h(\theta), \\ N(\theta) &= N_0(\theta) + P_0(\theta)Q(\theta), & \Lambda^n(\theta) &= \Lambda_0^n(\theta) + \Delta(\theta) \end{aligned}$$

Consider  $K(\theta) = K_0(\theta) + P_0(\theta)H(\theta)$  and  $f(\theta) = f_0(\theta) + h(\theta)$  as new approximations of  $K$  and  $f$ .

We want to solve  $F(K(\theta) - K(f(\theta))) = 0$ , which in terms of the new approximation means:

$$0 = F(K_0(\theta) + P_0(\theta)H(\theta)) \\ - (K_0(f_0(\theta) + h(\theta)) + P_0(f_0(\theta) + h(\theta))H(f_0(\theta) + h(\theta)))$$

Doing computations, and neglecting quadratically small terms, it becomes:

$$-\tilde{R}(\theta) = \Lambda(\theta)H(\theta) - \begin{pmatrix} h(\theta) \\ 0 \end{pmatrix} - H(f(\theta))$$

where  $\tilde{R}(\theta)$  is the projection over the invariant subspaces  
 $\tilde{R}(\theta) := P_0^{-1}(f_0(\theta))R(\theta)$ .

As the dynamics  $\Lambda_0$  splits into tangent and normal part

$$\Lambda_0(\theta) = \begin{pmatrix} \Lambda_0^t(\theta) & 0 \\ 0 & \Lambda_0^n(\theta) \end{pmatrix} = \begin{pmatrix} Df_0(\theta) & 0 \\ 0 & \Lambda_0^n(\theta) \end{pmatrix}$$

last equation also splits into tangent and normal part:

$$\begin{aligned} -\tilde{R}^t(\theta) &= Df_0(\theta)H^t(\theta) - h(\theta) - H^t(f_0(\theta)) \\ -\tilde{R}^n(\theta) &= \Lambda_0^n(\theta)H^n(\theta) - H^n(f_0(\theta)) \end{aligned}$$

and can be solved separately.

## Normal component

Also, as we suppose that the invariant manifold is hyperbolic, the linearized normal dynamics is expressed as

$$\Lambda_0^n(\theta) = \begin{pmatrix} \Lambda_0^u(\theta) & 0 \\ 0 & \Lambda_0^s(\theta) \end{pmatrix}$$

and the normal part equation also splits into stable and unstable components :

$$\begin{aligned} -\tilde{R}^s(\theta) &= \Lambda_0^s(\theta)H^s(\theta) - H^s(f_0(\theta)) \\ -\tilde{R}^u(\theta) &= \Lambda_0^u(\theta)H^u(\theta) - H^u(f_0(\theta)) \end{aligned}$$

and can be solved separately.

## Normal component

We could solve both equation by simple iteration using the contracting principle, which will converge to the wanted solutions  $H^s, H^u$ .

$\rightsquigarrow$  STABLE COMPONENT : For  $H^s$  the equation is already a contraction (linearized stable dynamics  $\Lambda_0^s$  contracting). Doing some arrangements on it we obtain the iterating equation:

$$-\tilde{R}^s(\theta) = \Lambda_0^s(\theta)H^s(\theta) - H^s(f_0(\theta)) \rightsquigarrow H_{N+1}^s(\theta) = \Lambda_0^s(f_0^{-1}(\theta))H_N^s(f_0^{-1}(\theta)) + \tilde{R}^s(f_0^{-1}(\theta))$$

$\rightsquigarrow$  UNSTABLE COMPONENT : For  $H^u$  we have the equation as an expansion (linearized unstable dynamics  $\Lambda_0^u$  expanding). Doing arrangements to get  $\Lambda_0^u$  inverted and so the equation as a contraction, the iterating equation is:

$$-\tilde{R}^u(\theta) = \Lambda_0^u(\theta)H^u(\theta) - H^u(f_0(\theta)) \rightsquigarrow H_{N+1}^u(\theta) = (\Lambda_0^u(\theta))^{-1} [H_N^u(f_0(\theta)) - \tilde{R}^u(\theta)]$$

## Tangent component

To solve the tangent part

$$-\tilde{R}^t(\theta) = Df_0(\theta)H^t(\theta) - h(\theta) - H^t(f_0(\theta))$$

we have to solve an overdetermined system: have one equation with two unknowns (  $H^t$  and  $h$  ).

So, we have to chose some condition over the system to be able to solve it.

- Election:  $H^t(\theta) = 0$ , which means we don't modify the tangent part.

Doing this election, we obtain the uniqueness of the solution of the equation, which will be:

$$h(\theta) = \tilde{R}^t(\theta)$$

## Solution of $F(K(\theta) - K(f(\theta))) = 0$

The improved torus and its dynamics becomes of the form

$$\begin{aligned}\hat{K}(\theta) &= K_0(\theta) + N_0(\theta)H_N^n(\theta) \\ \hat{f}(\theta) &= f_0(\theta) + \tilde{R}^t(\theta)\end{aligned}$$

with a new error  $\hat{R}(\theta) = F(\hat{K}(\theta) - \hat{K}(\hat{f}(\theta)))$  quadratically small than  $R(\theta)$ .

Suppose  $\hat{K}$  and  $\hat{f}$  had already been computed. The error on the bundles changes to:

$$DF(\hat{K}(\theta))P_0(\theta) - P_0(\hat{f}(\theta))\Lambda_0(\theta) = \hat{S}(\theta)_{n \times n}$$

and taking into account only the normal part of it, we consider:

$$DF(\hat{K}(\theta))N_0(\theta) - N_0(\hat{f}(\theta))\Lambda_0^n(\theta) = \hat{S}^n(\theta)_{n \times n-d}$$

Consider  $N(\theta) = N_0(\theta) + P_0(\theta)Q(\theta)$  and  $\Lambda^n(\theta) = \Lambda_0^n(\theta) + \Delta(\theta)$  as new approximations of bundles and bundles dynamics.

We want to solve  $DF(K(\theta))N(\theta) - N(f(\theta))\Lambda^n(\theta) = 0$ , which in terms of the new approximation means:

$$0 = DF(\hat{K}(\theta)) (N_0(\theta) + N_0(\theta)Q(\theta)) \\ - \left( N_0(\hat{f}(\theta)) + N_0(\hat{f}(\theta))Q(\hat{f}(\theta)) \right) - (\Lambda_0^n(\theta) + \Delta^n(\theta))$$



Doing computations, and neglecting quadratically small terms, it becomes

$$-\tilde{S}^n(\theta) = \Lambda_0(\theta)Q(\theta) - Q(\hat{f}(\theta))\Lambda_0^n(\hat{f}(\theta)) - \begin{pmatrix} 0_{d \times n-d} \\ \Delta^n(\theta)_{n-d \times n-d} \end{pmatrix}$$

where  $\tilde{S}^n(\theta)$  is the projection over the invariant subspaces,  
 $\tilde{S}^n(\theta) := P_0^{-1}(\hat{f}(\theta))\hat{S}^n(\theta)$ .

## Matrix notation

For  $Q$  and  $S$  we use the notation:

$$M(\theta) = \begin{pmatrix} M^{tt}(\theta) & M^{ts}(\theta) & M^{tu}(\theta) \\ M^{st}(\theta) & M^{ss}(\theta) & M^{su}(\theta) \\ M^{ut}(\theta) & M^{us}(\theta) & M^{uu}(\theta) \end{pmatrix}_{n \times n}$$

where  $t, s, u$  means *tangent*, *stable* and *unstable* direction projections respectively.

In our case we do not modify the tangent space, so we only need last  $n - d$  columns corresponding to the normal space:

$$M(\theta) = \begin{pmatrix} M^{ts}(\theta) & M^{tu}(\theta) \\ M^{ss}(\theta) & M^{su}(\theta) \\ M^{us}(\theta) & M^{uu}(\theta) \end{pmatrix}_{n \times n-d}$$

Using this notation, we could split the last matrix equation into 6 equations as:

$$\begin{aligned}
 -\tilde{S}^{ts}(\theta) &= \Lambda_0^t(\theta)Q^{ts}(\theta) - Q^{ts}(\hat{f}(\theta))\Lambda_0^s(\hat{f}(\theta)) \\
 -\tilde{S}^{ss}(\theta) &= \Lambda_0^s(\theta)Q^{ss}(\theta) - Q^{ss}(\hat{f}(\theta))\Lambda_0^s(\hat{f}(\theta)) - \Delta^s(\theta) \\
 -\tilde{S}^{rus}(\theta) &= \Lambda_0^u(\theta)Q^{us}(\theta) - Q^{us}(\hat{f}(\theta))\Lambda_0^s(\hat{f}(\theta)) \\
 -\tilde{S}^{tu}(\theta) &= \Lambda_0^t(\theta)Q^{tu}(\theta) - Q^{tu}(\hat{f}(\theta))\Lambda_0^u(\hat{f}(\theta)) \\
 -\tilde{S}^{su}(\theta) &= \Lambda_0^s(\theta)Q^{su}(\theta) - Q^{su}(\hat{f}(\theta))\Lambda_0^u(\hat{f}(\theta)) \\
 -\tilde{S}^{uu}(\theta) &= \Lambda_0^u(\theta)Q^{uu}(\theta) - Q^{uu}(\hat{f}(\theta))\Lambda_0^u(\hat{f}(\theta)) - \Delta^u(\theta)
 \end{aligned}$$

# Projection Method

## Remark

*To simplify the algorithm we use the projection method: make projections into the normal direction, making zero the components projected to itself:  $Q^{ss} = Q^{uu} = 0$ .*

Doing it, we obtain directly the correction of the linearized normal dynamics  $\Lambda^n(\theta)$ :

$$\tilde{S}^{ss}(\theta) = \Delta^s(\theta)$$

$$\tilde{S}^{uu}(\theta) = \Delta^u(\theta)$$

The other 4 equations can be solved iterating by the contraction principle: all equations are contractions or expansions by NHIM definition (normal dynamics always dominates the character of  $\Lambda^t(\theta)$ ).

Doing arrangements to obtain  $Q^{**}(\theta)$ ,  $* = \{t, s, u\}$ , conveniently isolated, these 4 iterating equations become:

$$Q_{N+1}^{ts}(\theta) = (Q_N^{ts}(\hat{f}(\theta))\Lambda_0^s(\theta) - \tilde{S}^{ts}(\theta)) (\Lambda_0^t)^{-1}(\theta)$$

$$Q_{N+1}^{us}(\theta) = (Q_N^{us}(\hat{f}(\theta))\Lambda_0^t(\theta) - \tilde{S}^{us}(\theta)) (\Lambda_0^u)^{-1}(\theta)$$

$$Q_{N+1}^{tu}(\theta) = (\Lambda_0^t(\hat{f}^{-1}(\theta))Q_N^{tu}(\hat{f}^{-1}(\theta)) + \tilde{S}^{tu}(\hat{f}^{-1}(\theta))) (\Lambda_0^u)^{-1}(\hat{f}^{-1}(\theta))$$

$$Q_{N+1}^{su}(\theta) = (\Lambda_0^s(\hat{f}^{-1}(\theta))Q_N^{su}(\hat{f}^{-1}(\theta)) + \tilde{S}^{su}(\hat{f}^{-1}(\theta))) (\Lambda_0^u)^{-1}(\hat{f}^{-1}(\theta))$$

and they converge to the solution when  $N \rightarrow \infty$ .

## Solution of $DF(K(\theta)N(\theta) - N(f(\theta)))\Lambda^n(\theta) = 0$

The improved bundles and its dynamics becomes of the form

- $\hat{P}(\theta) = (\hat{T}(\theta) | \hat{N}^s(\theta) | \hat{N}^u(\theta)):$

$$\begin{aligned}\hat{N}^s(\theta) &= N_0^s(\theta) + DK_0(\theta)Q^{ts}(\theta) + N_0^u(\theta)Q^{us}(\theta) \\ \hat{N}^u(\theta) &= N_0^u(\theta) + DK_0(\theta)Q^{tu}(\theta) + N_0^s(\theta)Q^{su}(\theta) \\ \hat{T}(\theta) &= D\hat{K}(\theta)\end{aligned}$$

- $\hat{\Lambda}(\theta) = \begin{pmatrix} \hat{\Lambda}^t(\theta) & 0 & 0 \\ 0 & \hat{\Lambda}^s(\theta) & 0 \\ 0 & 0 & \hat{\Lambda}^u(\theta) \end{pmatrix} : \begin{aligned} \hat{\Lambda}^s(\theta) &= S^{ss}(\theta) \\ \hat{\Lambda}^u(\theta) &= S^{uu}(\theta) \\ \hat{\Lambda}^t(\theta) &= D\hat{f}(\theta) \end{aligned}$

with a new error  $\hat{S}(\theta) = DF(\hat{K}(\theta))\hat{P}(\theta) - \hat{P}(\hat{f}(\theta))\hat{\Lambda}(\theta)$   
 quadratically small than  $S(\theta)$ .

We have to repeat **STEP 1** and **STEP 2** until new errors  $R$  and  $S$  achieve the desired error-tolerance.

## The inverse of $P(\theta)$ as an unknown

Adding the inverse of  $P(\theta)$  as another unknown  $\rightsquigarrow$  make faster the method

Let  $PI_0(\theta)$  as an approximation of  $P_0^{-1}(\theta)$ , with an error  $EPI(\theta) = PI_0(\theta)P(\theta) - Id_{n \times n}$  small.

If the new approximation is  $\hat{P}I(\theta) = PI_0(\theta) + QI(\theta)PI_0(\theta)$  with a modification  $QI(\theta) = -E\hat{P}I(\theta)$ , the improved inverse becomes:

$$\hat{P}I(\theta) = PI_0(\theta) - E\hat{P}I(\theta)PI_0(\theta)$$

which is computed each time after **STEP 2**.



# Discretization

To apply this method numerically, we need the torus, bundles and dynamics discretized.

We use the following discretization methods

- Interpolation
- Fourier expansions

This kind of discretization works correctly with this method as long as the invariant tori remains as a graph.

We consider the model manifold 1-dimensional,  $\mathcal{M}^d = \mathbb{T}^1$ , and we use functions  $h(\theta)$  and  $M(\theta)$  to represent the problem:

$$\begin{aligned}h &: \mathbb{T}^1 \rightarrow \mathbb{R}^n \\M &: \mathbb{T}^1 \rightarrow M_{n \times n}\end{aligned}$$

## Discretization by interpolation

We will use functions  $h$  represented as a Lagrange Interpolating Polynomial of degree  $r$ :

$$h(\tau) = \sum_{i=0}^r h(\theta_i) L_i(\tau)$$

So we need to storage a finite mesh of values over the function.

Let  $h$  discretized by function values  $h(\theta_j)$  at  $N$  equidistant meshing points  $\theta_j \in \mathbb{T}^1$ ,  $j = 1, \dots, N$ ,  $N$  large enough to well approximate  $h$ .

When we need the value  $h(\tau)$ , for some  $\tau \neq \theta_j$ ,  $j = 1, \dots, N$  we have to:

- 1 Look for  $\theta_j$  and  $\theta_{j+1}$  such that:  $\dots < \theta_j \leq \tau < \theta_{j+1} < \dots$
- 2 Compute the *Lagrange basis polynomials* of this value  $\tau$

$$L_i(\theta) = \frac{\prod_{j=0, j \neq i}^r \theta - \theta_j}{\prod_{j=0, j \neq i} \theta_i - \theta_j}, \quad i = 0, \dots, r$$

- 3 Compute *Lagrange Polynomial* of  $h$  on  $\tau$ :

$$h(\tau) = \left( \sum_{i=0}^r h(\theta_i) L_i(\tau) \right) \text{ mod } 2\pi$$

## Lagrange basis polynomials modulus $2\pi$

We are interpolating periodic functions modulus  $2\pi$  instead of real functions

$\rightsquigarrow$

TAKE CARE in the computation of Lagrange basis polynomials modulus  $2\pi$

We use the next convention (interpolation cubic case) to consider the nodes  $\theta_{j-1} < \theta_j \leq \tau < \theta_{j+1} < \theta_{j+1}$ :

$$\begin{array}{llllll}
 j = 0 & \Rightarrow & \theta_{j-1} = \theta_{N-1} - 2\pi, & \theta_j = \theta_0, & \theta_{j+1} = \theta_1, & \theta_{j+1} = \theta_2 \\
 j = N - 2 & \Rightarrow & \theta_{j-1} = \theta_{N-3}, & \theta_j = \theta_{N-2}, & \theta_{j+1} = \theta_{N-1}, & \theta_{j+1} = \theta_0 + 2\pi \\
 j = N - 1 & \Rightarrow & \theta_{j-1} = \theta_{N-2}, & \theta_j = \theta_{N-1}, & \theta_{j+1} = \theta_0 + 2\pi, & \theta_{j+1} = \theta_1 + 2\pi
 \end{array}$$

Otherwise, we have all values inside  $(0, 2\pi)$  and we don't have any modulus conflicting problem.

To compute the tangent bundle and tangent linearized dynamics, as we need to interpolate derivatives of  $K$  and  $f$ , we may compute the *derived Lagrange basis polynomials* instead of the Lagrange basis polynomials, and other computations are identically the same.

## Discretization by Fourier expansion

We will use functions  $h$  represented as Fourier expansions

$$h(\tau) = c_0 + \sum_{k=0}^r c_k \cos(k\tau) + s_k \sin(k\tau)$$

Then, we need to storage a finite number  $r$  of Fourier coefficients  $s_k, c_k$  for each function we need in the method.

With this discretization method, we could add a condition to check that the approximations we found are good approximated by inspecting the tails of the Fourier series.

# IMPLEMENTATION

## Continue $K$ NHIT with respect one parameter

- The algorithm is applied to continuation of invariant curves, saddle or attractor ones.
- We continue the torus w.r.t. ONE parameter regardless its dynamics, crossing resonances.
- We explore the mechanism of breakdown of the saddle invariant curve.



## Continuation steps

- Let our discrete dynamical system  $F$  depends on one perturbation parameter  $\epsilon$ .
- For some  $\epsilon_0$ , suppose known (or could had been computed explicitly) a  $K$  NHIT with his dynamics and bundles.
- Varying  $\epsilon$ , we apply the method to compute a new  $K$ ,  $f$ ,  $P$ ,  $\Lambda$  for each new  $\epsilon$ .
- We can follow incrementing  $\epsilon$  while the method converges for new epsilon: while we reach the prefixed tolerance in less than 15 steps. At this moment, the invariant torus breakdown (it ceases to be normally hyperbolic).

## 3D Fattened Arnold Family

The Fattened Arnold Family 3-dimensional (3D-FAF) is a map  $F : \mathbb{T}^1 \times \mathbb{R}^2 \rightarrow \mathbb{T}^1 \times \mathbb{R}^2$  defined by :

$$F \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x + a + \epsilon(\sin(x) + y + z/2) \\ b(\sin(x) + y) \\ c(\sin(x) + y + z) \end{pmatrix}$$

where  $a \in \mathbb{T}^1$ ,  $b, c \in \mathbb{R}$  are fixed parameters ( $b < 1$ ,  $c > 1$ ) and  $\epsilon \in \mathbb{R}$  is the perturbation parameter.

We apply our method with  $\mathcal{M}^1 = \mathbb{T}^1$  and  $n = 3$ .

We use the fixed parameters fixed as:

$$a = 0.1$$

$$b = 0.3$$

$$c = 2.4$$

## Unperturbed case: $\epsilon = 0$

$\rightsquigarrow a = 0$ : we have an explicit invariant circle given by the expression:

$$x \in \mathbb{T}^1 : K_0(x) = \left( x, \frac{b}{1-b} \sin(x), \frac{c}{(1-b)(1-c)} \sin(x) \right) \quad (2)$$

which is a graph.

$\rightsquigarrow a > 0$ : an invariant circle already exists and it is an approximation of (2), of the form:

$$K_0(x) = (x, \varphi(x), \psi(x)), \quad x \in \mathbb{T}^1$$

By invariance, it has to meet the invariance equation:

$$F((x, \varphi(x), \psi(x))) = (f_0(x), \varphi(f_0(x)), \psi(f_0(x)))$$

from where we obtain the initial dynamics:  $f_0(x) = x + a$ .

## Unperturbed case: $\epsilon = 0$

- Second coordinate equality of the equation is a contraction (as  $b = 0.3 < 1$ )
- Third coordinate equality of the equation is an expansion ( as  $c = 2.4 > 1$ )

So, we could solve each equality by iteration obtaining the initial approximation of the invariant circle:

$$K_0(x) = (x, \varphi(x), \psi(x)), \text{ where } \begin{cases} x \in \mathbb{T}^1 \\ \varphi(x) = \sum_{k=1}^{\infty} b^k \sin(x - ka) \\ \psi(x) = - \sum_{k=1}^{\infty} \frac{\varphi(x+ka) + \sin(x+ka)}{c^k} \end{cases}$$

## Perturbed case: $\epsilon > 0$

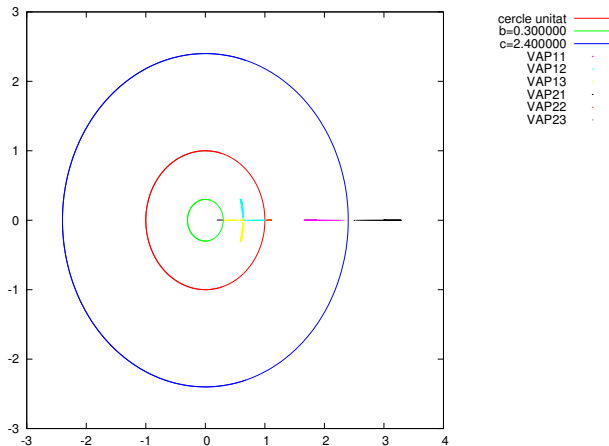
- It hasn't fixed points while  $\epsilon \in [0, \epsilon_1)$ ,  $\epsilon_1 := \left| \frac{-2a(1-b)(1-c)}{2-c} \right|$
- At  $\epsilon_1$  a fixed point appears (have a **FOLD**) with eigenvalues:

$$\begin{aligned}\lambda_1 &= c > 1 \\ \lambda_2 &= 1 \\ \lambda_3 &= b < 1\end{aligned}$$

- For  $\epsilon > \epsilon_1$ , this fixed point splits into two saddles, with eigenvalues:

$$\left\{ \begin{array}{c|ccc} & \text{SADDLE 1} & & \text{SADDLE 2} \\ \hline & 1 < \lambda_1 < c & & 1 < c < \mu_1 \\ & b < \lambda_2 < 1 & & 1 < \mu_2 < c \\ & b < \lambda_3 < 1 & & \mu_3 < b < 1 \end{array} \right\}$$

We can increase  $\epsilon$  until  $\epsilon_2 := \approx 0.776177304$ , where  $\lambda_2$  and  $\lambda_3$  collide (with  $\lambda \approx 0.624$ ) and become two complex conjugate eigenvalues. At this point, the invariant circle loses its normal hyperbolicity.



**Figure:** Evolution of all eigenvalues for the two fixed points of 3D-FAF from fold  $\epsilon_1 = 0.49$  until  $\epsilon = 1$ .

# First initial approximation

We use as a initial approximation:

- $K_0(\theta) = \left( \theta, \sum_{k=1}^{10^5} b^k \sin(\theta - ka), - \sum_{k=1}^{10^5} \frac{\varphi(\theta+ka) + \sin(\theta+ka)}{c^k} \right)$

- $f_0(\theta) = \theta + a$

- $\Lambda_0(\theta) = \begin{pmatrix} \Lambda_0^t(\theta) & 0 & 0 \\ 0 & \Lambda_0^s(\theta) & 0 \\ 0 & 0 & \Lambda_0^u(\theta) \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{pmatrix}$

the three eigenvalues of  $DF_0$ .

- $P_0(\theta) = \left( \frac{\partial K_0}{\partial \theta}(\theta) \mid N_0^s(\theta) \mid N_0^u(\theta) \right)$ , where

$$N_0^s(\theta) = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \text{ and } N_0^u(\theta) = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

We use the method with:

- INTERPOLATION DISCRETIZATION CASE: initial mesh of the parameterization of torus: 200 points.
- FOURIER DISCRETIZATION CASE: initial number of Fourier coefficients: 50 coefficients .
- As a tolerance of the errors:  $\|R\|, \|S\|, \|EPI\| < 10^{-10}$ .

At the end, we obtain:

↪ INTERPOLATION DISCRETIZATION CASE:

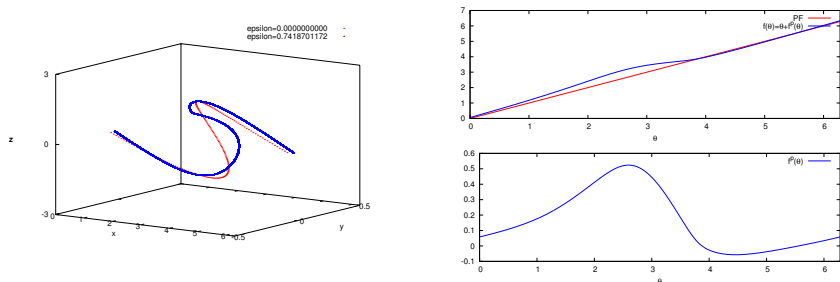
- final mesh of the parameterization of torus: 409.600 points.
- final epsilon value: 0.7418701172.

↪ FOURIER DISCRETIZATION CASE:

- final number of Fourier coefficients: 10.000 coefficients.
- final epsilon value: 0.7166513681.

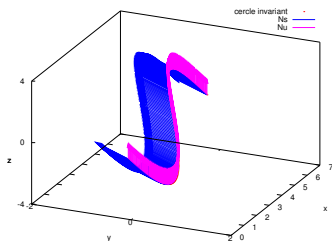


# INTERPOLATION DISCRETIZATION CASE

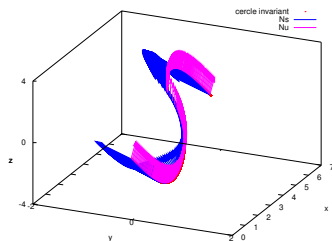


**Figure:** Evolution of the invariant tori (in red there the corresponding to the unperturbed system and in blue the last NHIM we can compute, for  $\epsilon = 0.7418701172$ ) and the dynamics over our last tori.

# INTERPOLATION DISCRETIZATION CASE



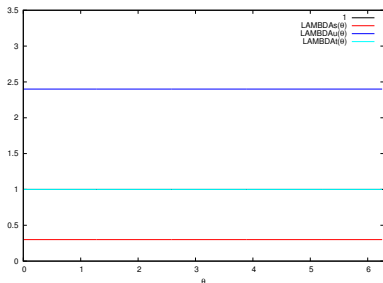
(a)



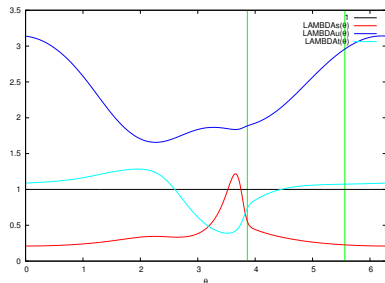
(b)

**Figure:** Variation of normal fibers from the unperturbed system (a) until the loss of normal hyperbolicity, at  $\epsilon = 0.7418701172$  (b)

# INTERPOLATION DISCRETIZATION CASE



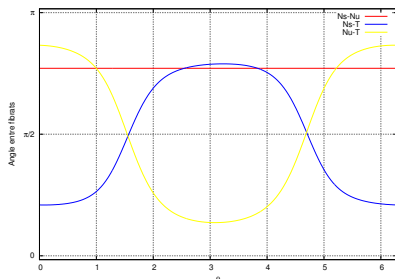
(a)



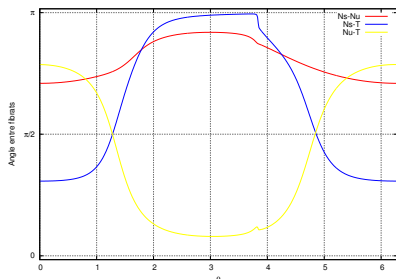
(b)

**Figure:** Evolution of the dynamics of tangent and normal bundle, from the unperturbed system (a) until  $\epsilon = 0.7418701172$  (b).

# INTERPOLATION DISCRETIZATION CASE



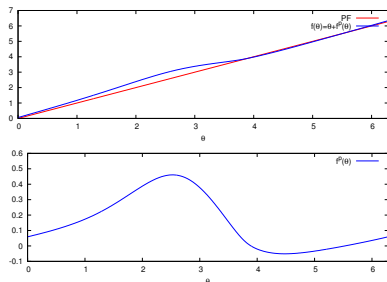
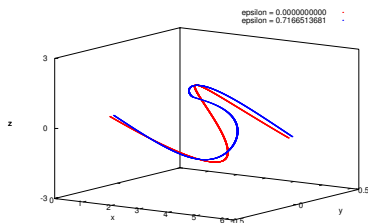
(a)



(b)

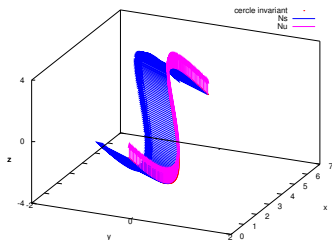
**Figure:** Angles between all bundles, from the unperturbed system (a) until  $\epsilon = 0.7418701172$  (b).

# FOURIER DISCRETIZATION CASE

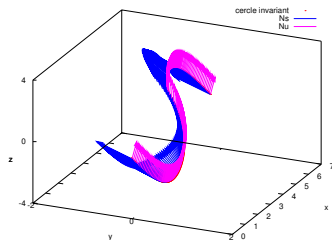


**Figure:** Evolution of the invariant tori (in red there the corresponding to the unperturbed system and in blue the last NHIM we can compute, for  $\epsilon = 0.7166513681$ ) and the dynamics over our last tori.

# FOURIER DISCRETIZATION CASE



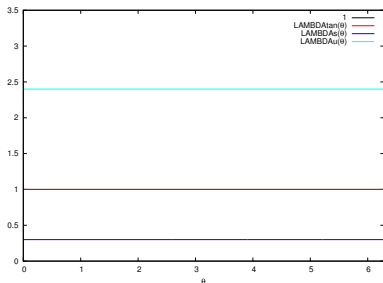
(a)



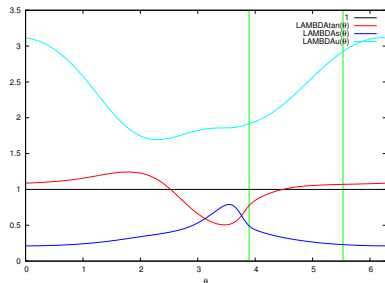
(b)

**Figure:** Variation of normal fibers from the unperturbed system (a) until the loss of normal hyperbolicity, at  $\epsilon = 0.7166513681$  (b)

# FOURIER DISCRETIZATION CASE



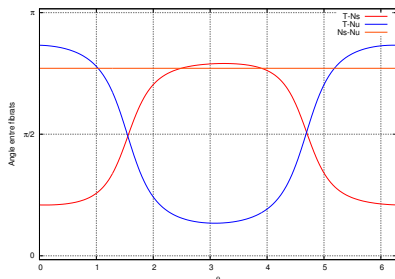
(a)



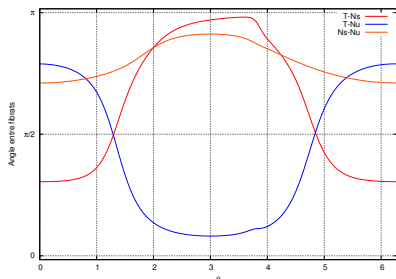
(b)

**Figure:** Evolution of the dynamics of tangent and normal bundle, from the unperturbed system (a) until  $\epsilon = 0.7166513681$  (b).

# FOURIER DISCRETIZATION CASE



(a)



(b)

**Figure:** Angles between all bundles, from the unperturbed system (a) until  $\epsilon = 0.7166513681$  (b).



# FOURIER DISCRETIZATION CASE

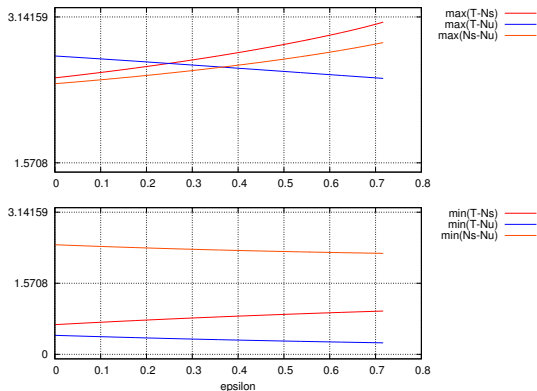


Figure: Maximum and minimum angles between all bundles from  $\epsilon = 0$  until  $\epsilon = 0.7166513681$ .