

# Intra-disciplinary bridges for multi-dimensional patterns: Part II

Priya Subramanian

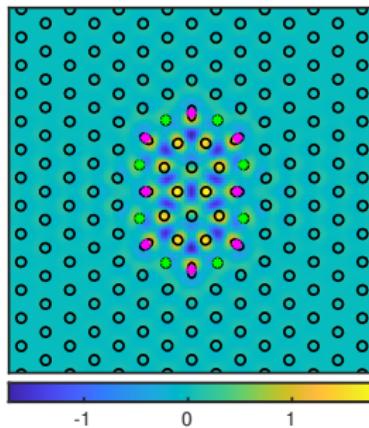
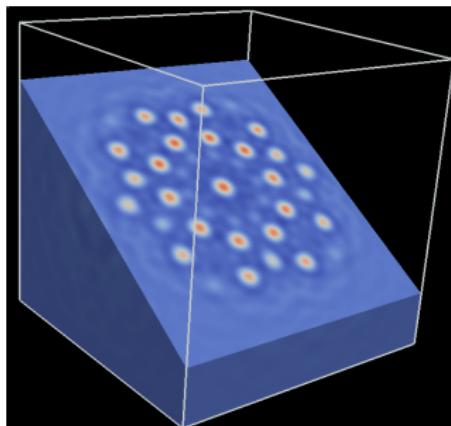


By Frits Ahlefeldt

with many collaborators: AMR, AJA, EK, VR, CL, MC, etc.

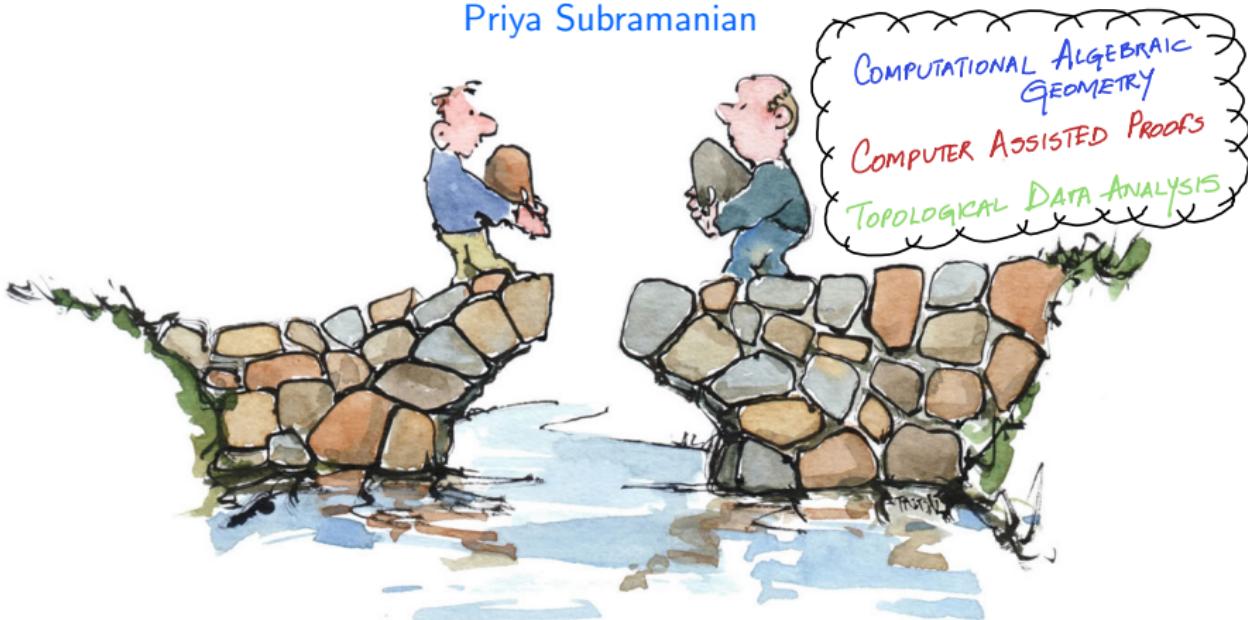
## A small recap...

- ▶ Pattern formation often requires techniques to simulate/analyse nonlinear PDEs
- ▶ Things we can do: numerical simulations (IVP), numerical continuation (root finding), **weakly nonlinear analysis**
- ▶ Things we can prove in 1D: Existence proofs for specific families of patterns, fronts and waves, spatially localisation of patterns and pinned defects



# Intra-disciplinary bridges for multi-dimensional patterns: Part II

Priya Subramanian



By Frits Ahlefeldt

with many collaborators: AMR, AJA, EK, VR, CL, MC, etc.

## Computer assisted proofs (CAPs) in nonlinear analysis

We want to construct algorithms that provide an approximate solution to a problem together with precise and possibly efficient bounds within which a rigorous exact solution is guaranteed to exist.

This area uses ideas from

- ▶ scientific computing
- ▶ functional analysis
- ▶ approximation theory
- ▶ numerical analysis
- ▶ topological methods

We will use a contraction mapping argument on a Newton-like operator to identify closed balls in a Banach space with bounds on error and on uniqueness.

## Preliminaries 1: existence theorem

Let  $X, Y$  are Banach spaces together with a smooth  $F : X \rightarrow Y$ . We want solutions  $x \in X$  such that

$$F(x) = 0$$

- ▶ Solutions can be an equilibrium, periodic solution, a bifurcating solution, a connecting orbit, etc.
- ▶ Often we have multiple efficient numerical methods to obtain a finite approximation of a solution as  $\bar{x} \in X$  with  $F(\bar{x}) \approx 0$
- ▶ *a-posteriori existence theorem*: We want to prove the existence of a unique true solution  $x \in X$  'near' a 'good' approximate solution  $\bar{x} \in X$  that we already know

## Preliminaries 2: Newton's method

- ▶ Let a scalar function  $f(x)$  that is  $C^2$  have a solution  $x^*$  such that  $f(x^*) = 0$  and  $f'(x^*) \neq 0$ .
- ▶ Then there exists  $\epsilon > 0$  such that for all  $a \in (x^* - \epsilon, x^* + \epsilon)$ , Newton iterations

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)},$$

initialised at  $x = a$  will converge to  $x^*$  with contraction property

$$|x_{n+1} - x^*| \leq \alpha |x_n - x^*|^2$$

- ▶ Every iterate gets closer to the true solution  $\rightarrow$  contraction
- ▶ What if we do not know  $x^*$  and do not know about properties of  $f$ , but only have an approximate solution  $x_0$  with a small  $f(x_0)$ ?

## Preliminaries 2: Newton's method

- ▶ Let a scalar function  $f(x)$  that is  $C^2$  have a solution  $x^*$  such that  $f(x^*) = 0$  and  $f'(x^*) \neq 0$ .
- ▶ Then there exists  $\epsilon > 0$  such that for all  $a \in (x^* - \epsilon, x^* + \epsilon)$ , Newton iterations

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)},$$

initialised at  $x = a$  will converge to  $x^*$  with contraction property

$$|x_{n+1} - x^*| \leq \alpha |x_n - x^*|^2$$

- ▶ Every iterate gets closer to the true solution  $\rightarrow$  contraction
- ▶ What if we do not know  $x^*$  and do not know about properties of  $f$ , but only have an approximate solution  $x_0$  with a small  $f(x_0)$ ?
- ▶ Hand wave!

## Preliminaries 3: extending to infinite dimensions

- ▶ Given  $X$  a Banach space,  $F : X \rightarrow Y$  a smooth map and  $B \subset X$  is closed
- ▶  $B$  is a complete metric space - inherits norm from  $X$
- ▶  $F : X \rightarrow Y$  is Fréchet differentiable at  $x_0 \in X$ , if there exists a bounded linear operator  $A : X \rightarrow Y$  having

$$\lim_{h \rightarrow 0} \frac{\|F(x_0 + h) - F(x_0) - Ah\|_Y}{\|h\|_X} = 0$$

- ▶  $A$  if it exists, is unique
- ▶ Therefore we define,

$$DF(x_0) := A$$

and call  $A$  the derivative of  $F$  at  $x_0$

## Preliminaries 4:

- ▶ Every point  $x \in X$  has an associated  $DF(x)$  in a Banach space  $G(X, Y)$ . If this correspondence is continuous, we say  $F \in C^1(X, Y)$
- ▶ If  $F \in C^1(X, Y)$ ,  $M \geq 0$  and has

$$\sup_{w \in B} \|DF(w)\|_{G(X, Y)} \leq M \quad B \subset X, B \text{ is convex}$$

then for  $x_1, x_2 \in B$ ,

$$\|F(x_1) - F(x_2)\|_Y \leq M \|x_1 - x_2\|_X$$

- ▶ Given  $B \subset X$  and  $B$  is convex, if

$$\sup_{w \in B} \|DF(w)\|_{G(X, Y)} < 1$$

then  $F$  is a contraction on the complete metric space  $B$

- ▶ *Fixed point theorem*: If  $B$  is a complete metric space and  $F$  is a contraction mapping, then  $F$  has a unique fixed point within  $B$

## a-posteriori existence theorem setup

- ▶ Let  $X, Y$  be Banach spaces and  $F = C^1(X, Y)$
- ▶ We have an approximate solution  $\bar{x} \in X$  with  $\|F(\bar{x})\|_X \ll 1$
- ▶ NOTE: there is no generic reason why  $F$  should be a contraction near  $\bar{x}$
- ▶ For the Newton method in Banach space,  $x_{n+1} = x_n + h_n$  where  $h_n$  is a solution to the linear equation

$$DF(x_n)h_n = -F(x_n)$$

- ▶ This suggests that we look for fixed points of near  $\bar{x}$  for

$$T(x) = x - DF(x)^{-1}F(x) \quad (\text{Newton operator})$$

- ▶  $T$  might be a contraction near  $\bar{x}$ !

## a-posteriori existence theorem setup

- ▶ **Problem 1:** Computing  $DF(x)^{-1}$  might be problematic
- ▶ **Approximation 1:** Replace inverse of derivative at  $x$  with a quantity independent of  $x$ , say the inverse of derivative at  $\bar{x}$  to define

$$T(x) = x - DF(\bar{x})^{-1}F(x) \quad (\text{Newton-like operator})$$

- ▶ **Problem 2:**  $DF(\bar{x})$  might still be hard to invert
- ▶ **Approximation 2:** Replace  $DF(\bar{x})$  with any  $A^\dagger \in G(X, Y)$  and  $DF(\bar{x})^{-1}$  with  $A \in G(Y, X)$
- ▶ This lets us define  $T = x - AF(x)$  as the Newton-like operator
- ▶ If  $A$  is injective, then fixed points of  $T$  correspond to zeros of  $F$
- ▶ **Approximation 3:** Actually we would like it better if  $A^\dagger \approx DF(\bar{x})$  and if  $A \approx \text{inv}(A^\dagger)$

## Newton-Kantorovich theorem

Suppose  $F : X \rightarrow Y$  is continuously differentiable and  $\bar{x} \in X$ ,  
 $A^\dagger \in G(X, Y)$  and  $A \in G(Y, X)$  and  $A$  is one-to-one

We want scalar bounds for

- ▶  $\|AF(\bar{x})\|_X \leq Y_0$  a-posteriori error
- ▶  $\|I - AA^\dagger\|_{G(X)} \leq Z_0$  approximate inverse
- ▶  $\|A(A^\dagger - DF(\bar{x}))\|_{G(X)} \leq Z_1$  approximate derivative
- ▶  $\sup_{x \in B_r(\bar{x})} \|A(DF(\bar{x}) - DF(x))\|_{G(X)} \leq Z_2(r)$   
Lipschitz bound for first derivative
- ▶ We require  $Y_0, Z_0, Z_1 > 0$  and  $Z_2(r) : [0, \infty) \rightarrow [0, \infty)$  to define the function  
$$p(r) = Z_2(r)r^2 - (1 - Z_0 - Z_1)r + Y_0$$

- ▶ Newton-Kantorovich theorem states that if there is an  $r > 0$  such that  $p(r) < 0$ , then there exists a unique  $x \in B_r(\bar{x})$  with  $F(x) = 0$

## Remarks

- ▶ In many applications, it can be arranged that  $Z_2(r)$  and hence  $p(r)$  are polynomial
- ▶ Therefore, this method is called the method of **radii polynomials**
- ▶ NOTE: This reduces infinite dimension zero finding problem to a one-dimensional zero finding problem
- ▶ We normally look for intervals  $I = [r_-, r_+]$  such that any  $r \in I$  implies  $p(r) < 0$
- ▶ NOTE: Both limits of interval have to be finite ( $p(r) \rightarrow \infty$  as  $r \rightarrow \infty$ ) and non-zero ( $p(0) = Y_0 > 0$ )
- ▶  $r_-$  gives the sharpest error bounds and  $r_+$  gives the lower bound on isolation

Proof: Assume  $r > 0$  has  $b(r) < 0$

We can define operator  $T: \overline{B_r(\bar{x})} \subset X \rightarrow X$  by  
$$T(x) = x - AF(x), \quad x \in \overline{B_r(\bar{x})}$$

Step 1: Show  $T$  maps into  $\overline{B_r(\bar{x})}$

Step 2: Show  $T$  is a contraction on  $\overline{B_r(\bar{x})}$

Step 3: Use the Banach fixed point theorem to  
conclude  $T$  has a unique fixed point  
 $x \in \overline{B_r(\bar{x})}$ .

Step 4: Since  $A$  is one-to-one, zeros of  $T$  are  
zeros of  $F$  in  $\overline{B_r(\bar{x})}$ .

Step 1:  $T$  is continuously differentiable on  $\overline{B_r(\bar{x})}$  as  $F$  is.  
 We want to bound the norm

$$\|DT(\bar{x})\|_{G(\bar{x})} = \|I - ADF(\bar{x})\|_{G(\bar{x})}$$

[Add & subtract  $AA^\dagger$  and  $ADF(\bar{x})$ ]

$$\begin{aligned} \Rightarrow \|DT(\bar{x})\|_{G(\bar{x})} &= \|I - AA^\dagger\|_{G(\bar{x})} \\ &\quad + \|A(A^\dagger - DF(\bar{x}))\|_{G(\bar{x})} \\ &\quad + \|A(DF(\bar{x}) - DF(\bar{x}))\|_{G(\bar{x})} \\ &\leq Z_0 + Z_1 + Z_2(r)r \end{aligned}$$

$$\begin{aligned} \text{But } p(r) &= Z_2(r)r^2 - (1 - Z_1 - Z_0)r + Y_0 < 0 \\ \Rightarrow Z_2(r)r^2 + (Z_1 + Z_0)r + Y_0 &< r \end{aligned}$$

$$\text{or } Z_2(r)r + (Z_1 + Z_0) + \frac{Y_0}{r} < 1$$

But  $Y_0, r > 0$

$$\Rightarrow \|DT(z)\|_{G(X)} \leq Z_2(r)r + Z_1 + Z_0 < 1, z \in \overline{B_r(\bar{z})}$$

$\Rightarrow T$  maps into  $\overline{B_r(\bar{z})}$ . Step 1 proved.

Step 2:

$$\begin{aligned} \text{Consider } \| \bar{z} - T(z) \|_X &= \| \bar{z} - T(\bar{z}) \|_X \\ &\quad + \| T(\bar{z}) - T(z) \|_X \\ &\leq \| AF(\bar{z}) \|_X + \sup_{w \in B_r(\bar{z})} \| DT(w) \|_{G(X)} \| \bar{z} - w \|_X \\ &\leq Y_0 + (Z_2(r)r + Z_1 + Z_0) \| \bar{z} - \bar{z} \|_X \\ &\leq Y_0 + (Z_2(r)r + Z_1 + Z_0) r \\ &\leq Y_0 + Z_2 r^2 + (Z_1 + Z_0) r < r \quad (\mu(r) < 0) \end{aligned}$$

Then

$$T(\overline{B_r(\bar{x})}) \subset B_r(\bar{x}) \subset \overline{B_r(\bar{x})}$$

Since  $X$  is a Banach space,  $\overline{B_r(\bar{x})}$  is a complete metric space.

Choose  $x, y \in \overline{B_r(\bar{x})}$  and consider

$$\|T(y) - T(x)\|_X \leq \sup_{w \in \overline{B_r(\bar{x})}} \|DT(w)\|_{B_r(\bar{x})} \|x - y\|_X$$

$$\leq (Z_2(r)r + Z_1 + Z_0) \|x - y\|_X$$

$$\text{with } Z_2(r)r + Z_1 + Z_0 < 1$$

Then  $T$  is a strict contraction on the complete metric space  $\overline{B_r(\bar{x})}$ . Step 2 proved.

Given Step 1 & 2, we can use the fixed point theorem to conclude that  $T$  has a unique fixed point  $x \in B_r(\bar{x})$ , not on the boundary.

Again since  $A$  is one to one,  $T(\bar{x}) = \bar{x}$  implies  $F(\bar{x}) = 0$ .

//

In practice, we need...

- ▶ A numerical approximate solution  $\bar{x} \in X$
- ▶ A numerical approximation  $A^\dagger$  of  $DF(\bar{x})$
- ▶ A numerical approximation  $A$  of  $inv(A^\dagger)$
- ▶ Use some knowledge of the asymptotics of the derivative to define  $A^\dagger$  and  $A$
- ▶ An ability to check bounds – interval arithmetic

## Swift-Hohenberg equation in 1D

$$\frac{\partial U}{\partial t} = \mu U - (1 + \partial_x^2)^2 U + \nu U^2 - U^3.$$

Goal: Develop CAPS of existence of  $2d$  spatially periodic even equilibrium solutions on the interval  $\Omega_0 := (-d, d)$  for some  $d > 0$

We want  $F(U) = 0$  where  $U : \Omega_0 \rightarrow \mathbb{R}$ ,  $U(x) = U(-x)$  and  $U$  is  $2d$  periodic

So we can write

$$U(x) = \sum_{n \geq 0} \omega_n U_n \cos(2\pi(n/2d)x)$$

where  $\omega_0 = 1$  and  $\omega_n = 2$  for  $n \geq 1$

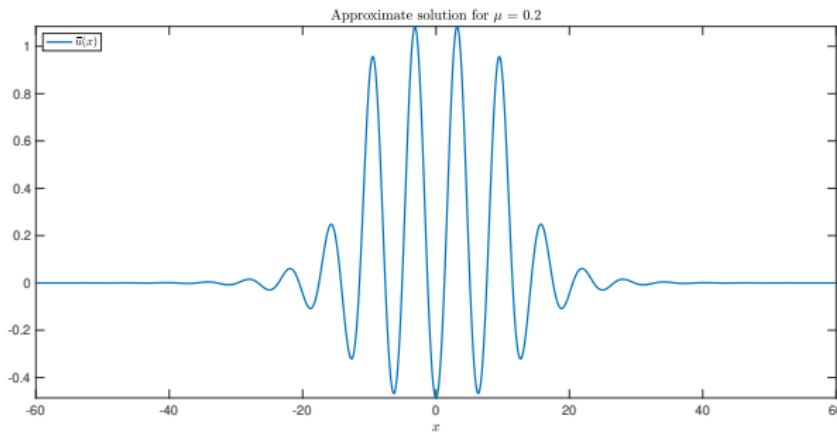
# Numerical approximation

We fix  $\mu \stackrel{\text{def}}{=} 0.2$ ,  $\nu \stackrel{\text{def}}{=} 1.6$ ,  $d = 60$  and  $N = 300$ .

The paper [Burke & Knobloch. Phys. Rev. E, 2006] provides the following ansatz  $u_{ini}$  to initialize Newton's method

$$u_{ini}(x) \stackrel{\text{def}}{=} 2\sqrt{\frac{2\mu}{\gamma}} \operatorname{sech}\left(\frac{x\sqrt{\mu}}{2}\right) \cos(x), \quad \gamma \stackrel{\text{def}}{=} \frac{38\nu^2}{9} - 3$$

Applying the FFT to  $u_{ini}$  on  $\Omega_0 = (-d, d)$  gave  $U_{ini} \in \mathbb{R}^{N+1}$ , which served as the initial guess for Newton's method on  $F^N$ , yielding the numerical approximation  $\bar{U}$



## Fourier series representation

Define  $\ell : \mathbb{R} \rightarrow \mathbb{R}$ , the Fourier transform of the differential operator as

$$\ell(\xi) := ((2\pi\xi)^2 - 1)^2 + \mu$$

for all  $\xi \in \mathbb{R}$

Plugging in the Fourier series expansion in SH equation, we get

$$F(U) := \mathcal{L}U + U * U * U - \nu U * U = 0$$

with  $\mathcal{L}U = ((\mathcal{L}U)_n)_{n \geq 0}$ ,  $U * U = ((U * U)_n)_{n \geq 0}$  given component-wise by

$$(\mathcal{L}U)_n := \ell(\tilde{n})U_n \quad \text{and} \quad (U * U)_n := \sum_{k \in \mathbb{Z}} U_{|k|}U_{|n-k|}$$

We define a bounded linear map

$$F : X \rightarrow \ell^2, \quad \text{where} \quad \|U\|_X = \|\mathcal{L}U\|_2$$

## Finite numerical representation of a solution

- ▶ Numerical methods allow us to recover a finite number of Fourier coefficients of  $\bar{U}$ , in the vicinity of  $F$  close to a true zero  $U \in X$
- ▶ We use projection operators to represent the finite dimensional objects

$$(\Pi^{\leq N} U)_n = \begin{cases} U_n, & |n| \leq N \\ 0, & |n| \leq N \end{cases}$$

and

$$(\Pi^{\geq N} U)_n = \begin{cases} 0, & |n| \leq N \\ U_n, & |n| \leq N \end{cases}$$

- ▶ We can identify elements in  $\Pi^{\leq N}$  with vectors in  $\mathbb{R}^{N+1}$
- ▶ The approximate solution  $\bar{U}$  we assume  $\bar{U} = \Pi^{\leq N} \bar{U}$  such that  $\bar{U}$  has at most  $N + 1$  non-zero Fourier coefficients

## Constructing an approximate inverse

We define  $M_U$  such that

$$M_U : X \rightarrow \ell^2 : V \rightarrow M_U W := U * W$$

it is the linear discrete convolution multiplication operator associated to  $U$ .

Given  $\mathcal{N}(U) = U^3 - \nu U^2$ , so

$$DF(\bar{U}) = \mathcal{L} + D\mathcal{N}[\bar{U}] = \mathcal{L} + 3M_{\bar{U}^2} - 2\nu M_{\bar{U}} = L + M_{\bar{V}}$$

$M_{\bar{V}}\mathcal{L}^{-1} : \ell^2 \rightarrow \ell^2$  is compact

This means that given a  $N$  big enough, we have

$$M_{\bar{V}}\mathcal{L}^{-1} \approx \Pi^{\leq N} M_{\bar{V}}\mathcal{L}^{-1} \Pi^{\leq N}$$

as a matrix

Computation of bounds - not shown here

In practice, we need...

- ▶ A numerical approximate solution  $\bar{x} \in X$
- ▶ A numerical approximation  $A^\dagger$  of  $DF(\bar{x})$
- ▶ A numerical approximation  $A$  of  $inv(A^\dagger)$
- ▶ Use some knowledge of the asymptotics of the derivative to define  $A^\dagger$  and  $A$
- ▶ An ability to check bounds – interval arithmetic

## Introduction to interval arithmetic & rounding

- ▶ It is a technique designed to rigorously account for rounding errors by representing numbers as intervals rather than points
- ▶ The endpoints of the intervals are chosen so that they can be stored exactly on the computer
- ▶ Arithmetic operations on intervals produce new intervals that rigorously enclose all outcomes of the corresponding real-number operations, including propagation of numerical error
- ▶ If  $\mathbb{F}$  is the set of representable real numbers that the computer can represent with a given precision, then  $\Delta$  and  $\nabla$  are functions  $\mathbb{R} \rightarrow \mathbb{F}$ , the corresponding round-up and round-down operators
- ▶ For every  $x \in \mathbb{R}$ ,

$$\Delta = \min\{y \in \mathbb{F}, y \leq x\}$$

$$\nabla = \max\{y \in \mathbb{F}, y \geq x\}$$

- ▶ NOTE: The equality  $\Delta(x) \geq x \geq \nabla(x)$  iff  $x \in \mathbb{F}$

## Rigorous enclosures and elementary operations

- ▶ Standing-point arithmetic replaces a number  $x$  by the closest floating point approximation, i.e., either  $\Delta(x)$  or  $\nabla(x)$
- ▶ Interval arithmetic replaces a real number  $x$  by the interval

$$[x] := [\nabla(x), \Delta(x)]$$

- ▶ This interval contains the real number  $x$  even if it is not representable at the current precision
- ▶ We can extend basic operations to intervals by combining their definitions with suitable rounding to ensure that the outcome always encloses the value
- ▶ If  $I = [a, b]$  and  $J = [c, d]$  with  $a, b, c, d \in \mathbb{F}$ , then

$$I + J = [\nabla(a + c), \Delta(b + d)]$$

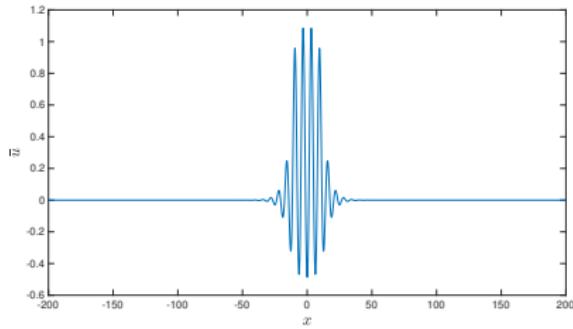
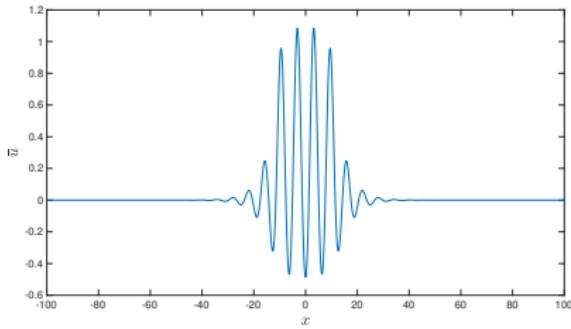
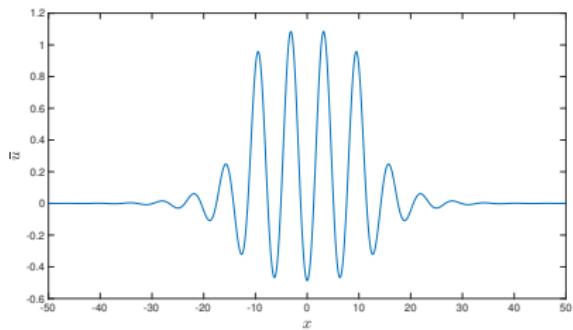
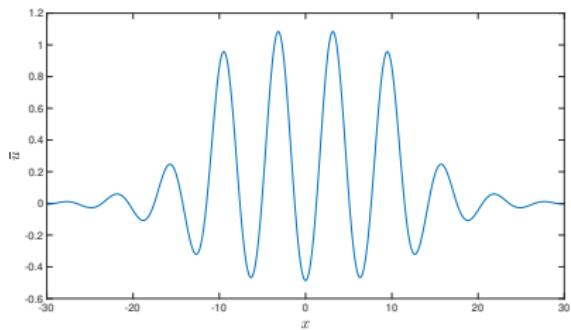
and contains by construction  $\{x + y, x \in I, y \in J\}$

- ▶ Many implementations exist: Intlab in Matlab, IntervalArithmetic.jl, etc.

$d$	N	$Y$	$Z_1$	$Z_2(r)$	$r$
30	200	$3.52 \times 10^{-11}$	$7.84 \times 10^{-6}$	249	$3.56 \times 10^{-11}$
60	300	$1.89 \times 10^{-11}$	$2.42 \times 10^{-5}$	250	$1.99 \times 10^{-11}$
100	500	$3.24 \times 10^{-11}$	$2.43 \times 10^{-5}$	251	$3.34 \times 10^{-11}$
200	800	$6.50 \times 10^{-11}$	$5.91 \times 10^{-5}$	252	$6.74 \times 10^{-11}$

Table 2.1: Computer-assisted proofs of periodic solutions for different half periods  $d$ .

# Results for the Swift-Hohenberg Equation



In summary, CAP allows us to reduce an infinite dimensional zero finding problem to a one dimensional root finding problem

The way to do this involved proving contraction of a Newton-like operator and use the Banach fixed point theorem

- ▶ Let  $\bar{x}$  be a numerical approximation to  $\mathcal{F}(x) = 0$  using a finite dimensional reduction
- ▶ Construct a linear operator  $\mathcal{A}$  that is the approximate inverse of  $\mathcal{D}\mathcal{F}(x)$
- ▶ Verify that  $\mathcal{A}$  is an injective linear operator
- ▶ Define a Newton-like operator  $T(x) = x - \mathcal{A}\mathcal{F}(x)$  about the numerical approximation  $\bar{x}$
- ▶ Consider  $\mathcal{B}_{\bar{x}}(r) \in X$ , the closed ball of radius  $r$  centered at  $\bar{x}$
- ▶ Find a radius  $r > 0$  such that the operator  $T$  is a contraction mapping

How can we extend a similar method to analyse spatial localisation in 2D/3D?  
Or to more complicated PDEs?



By Frits Ahlefeldt

CAP is agnostic to dimensions and might be a viable bridge to answer this question