

# Dynamics in Bose-Einstein Condensates

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## Outline

- Motivation
- Mathematical Models
- Vortices
- Ground State & Bound States

# 1 Motivation

- What's Bose-Einstein Condensate (BEC)?



The image features a blue background with four photographs of different states of matter. In the top left, a hand pours a white substance from a container, labeled "gas". In the top middle, a blue liquid surface is shown, labeled "liquid". In the top right, a large white ice formation is shown, labeled "solid". In the bottom left, a glowing orange and red celestial body is shown, labeled "plasma". In the center, the text "Phases of matter" is written in white. Below this text, it says "A new form of matter at the coldest temperatures in the universe...". At the bottom right, a purple rectangular box contains the text "BEC" in white.

gas

liquid

solid

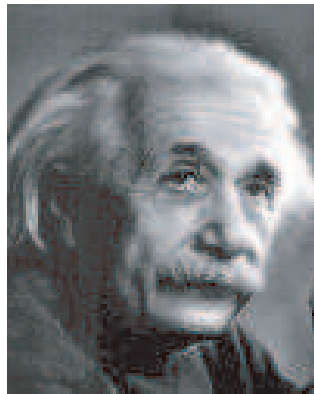
plasma

Phases of matter

A new form of matter at the coldest temperatures in the universe...

BEC

- Theoretical prediction 1924 ...
  - S. Bose: derived Planck's black body radiation law from considering the cavity radiation as an ideal photon gas and worked out Bose statistics for photons.
  - A. Einstein: generalized Bose statistics to other Bosonic particles and atoms (Bose-Einstein statistics) and predicted if the atoms were cold enough, almost all of the particles would congregate in the ground states (BEC).
  - Since 1924, BEC is the Holy Grail in physics.

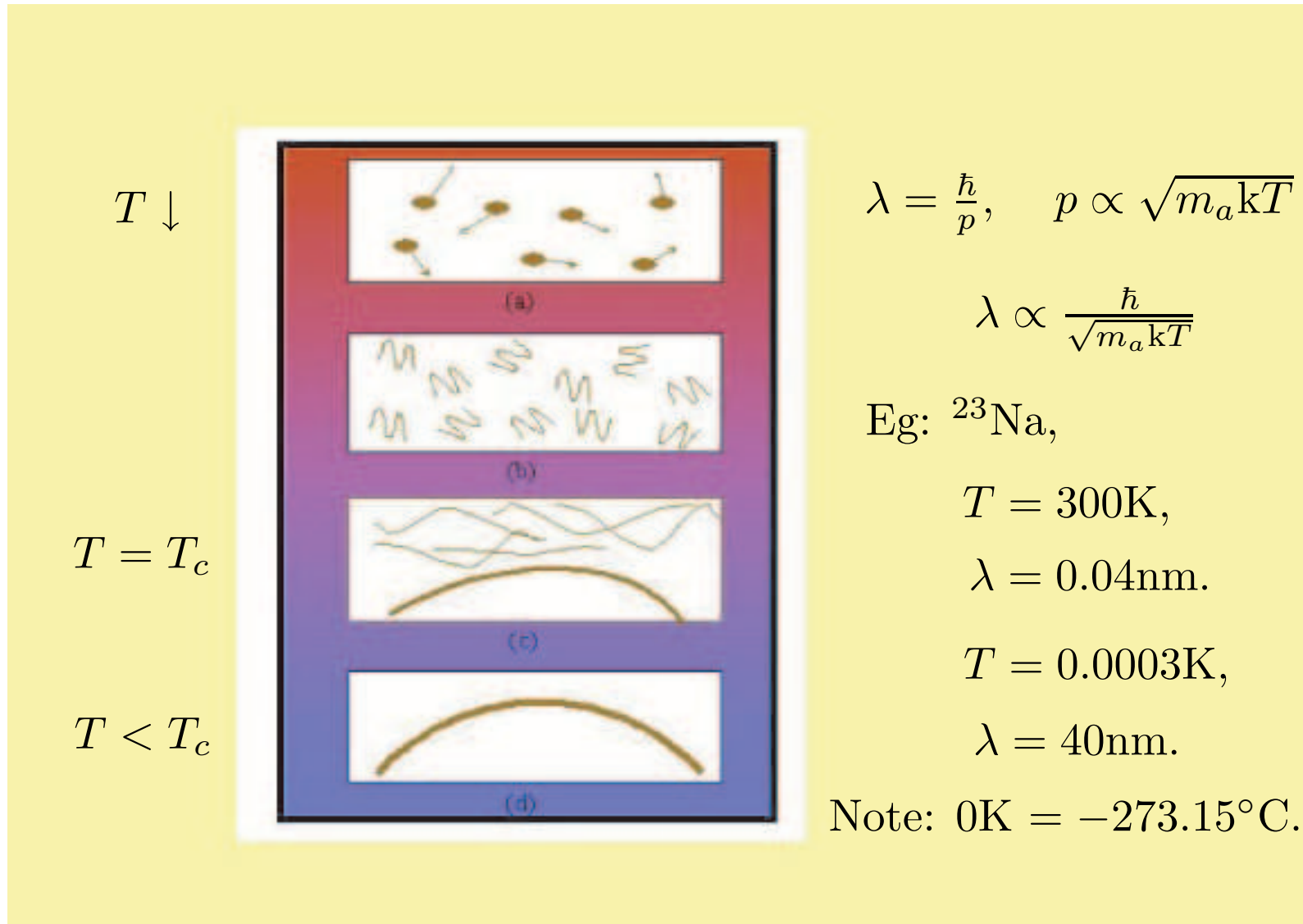


A. Einstein (1879 ~ 1955)



S. Bose (1894 ~ 1974)

- How does BEC happen?



$$\lambda = \frac{\hbar}{p}, \quad p \propto \sqrt{m_a kT}$$

$$\lambda \propto \frac{\hbar}{\sqrt{m_a kT}}$$

Eg:  $^{23}\text{Na}$ ,

$$T = 300\text{K},$$

$$\lambda = 0.04\text{nm}.$$

$$T = 0.0003\text{K},$$

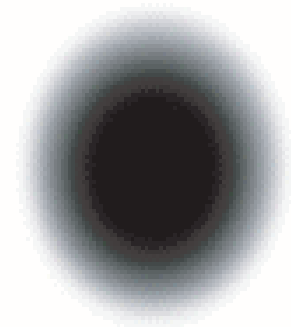
$$\lambda = 40\text{nm}.$$

Note:  $0\text{K} = -273.15^\circ\text{C}$ .

- (a) Cold atom: atoms in the lowest energy level spread out a little, so they look like very small fuzzy balls.
- (b) Super atom: at the special incredibly low temperatures (needed for BEC) they lose their individual identities and coalesce into a single blob.



(a)



(b)

- Physical experiments

- Superfluid He<sup>4</sup> 1938:

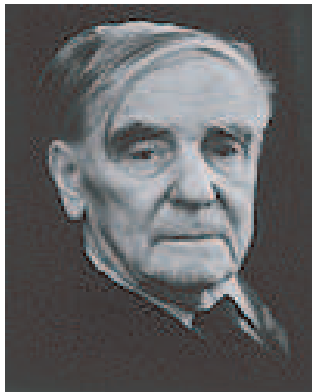
- P. L. Kapitza, Allen and Misener: discovered the superfluidity of liquid helium.

- F. London: proposed that the superfluid fraction consisting of those atoms which have “condensed” to the ground state.

- Difficulties

- \* Low temperature  $\approx$  absolutely zeros

- \* Dilute Bose gas

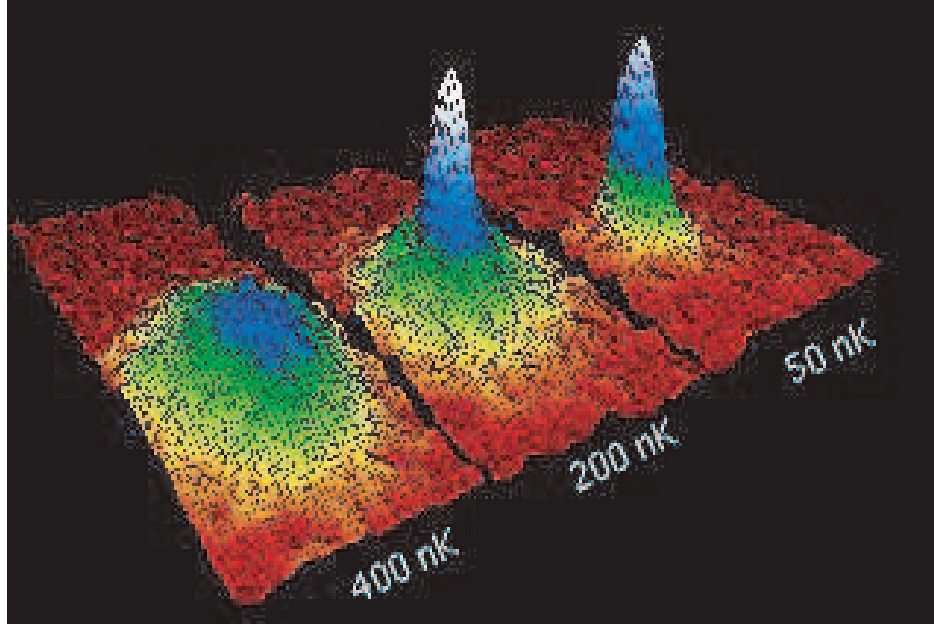
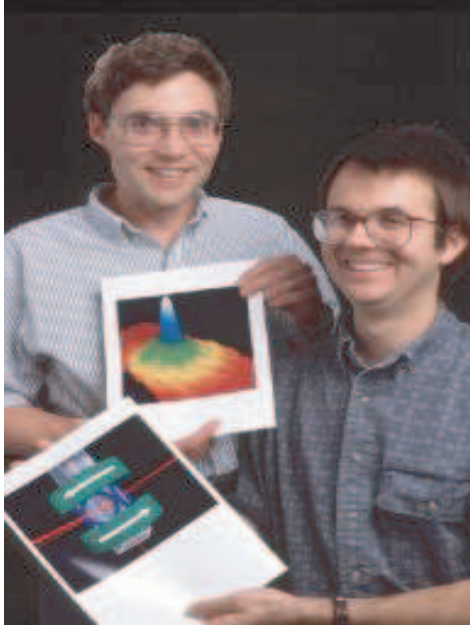


P. L. Kapitza  
(1894 ~ 1984)



F. London  
(1900 ~ 1954)

- – E. A. Cornell & C. E. Wieman (JILA, 1995):  
first observed BEC of rubidium ( $^{87}\text{Rb}$ ) atoms at 20 nK, i.e.  
0.000 000 02 K.



C. E. Wieman & E. A. Cornell

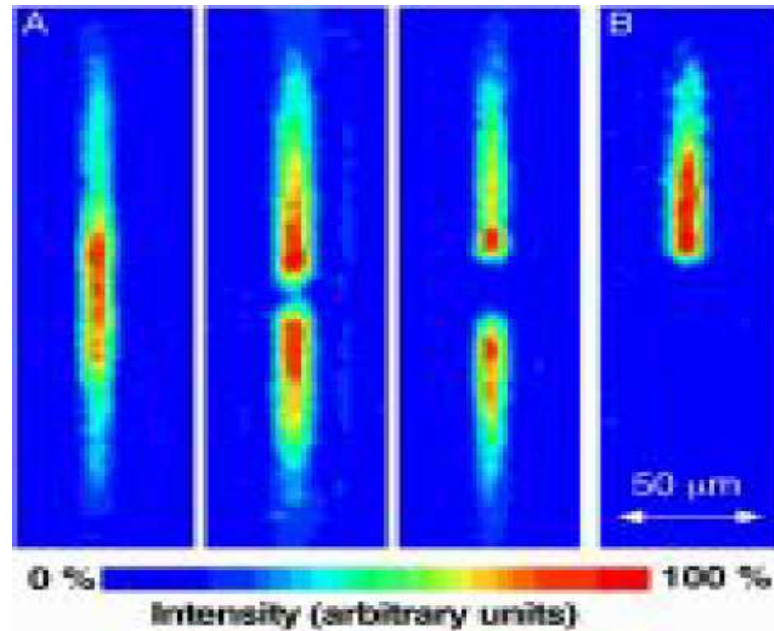
BEC at 400, 200, and 50 nK



- W. Ketterle (MIT, 1995):  
observed BECs of sodium ( $^{23}\text{Na}$ ) atoms.



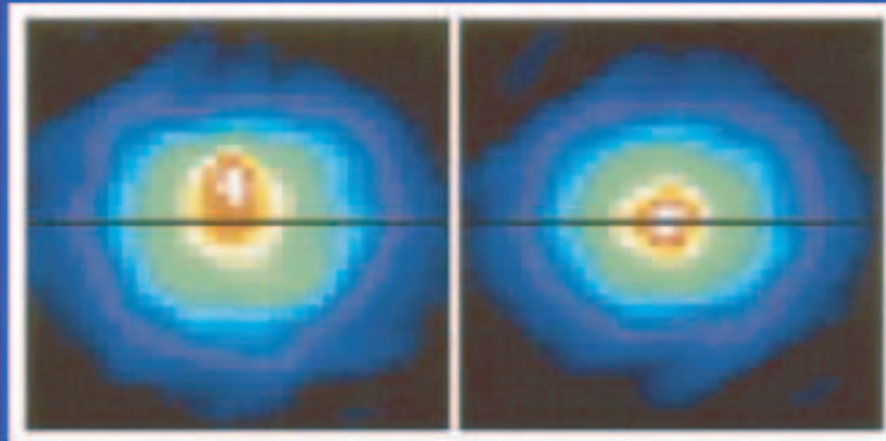
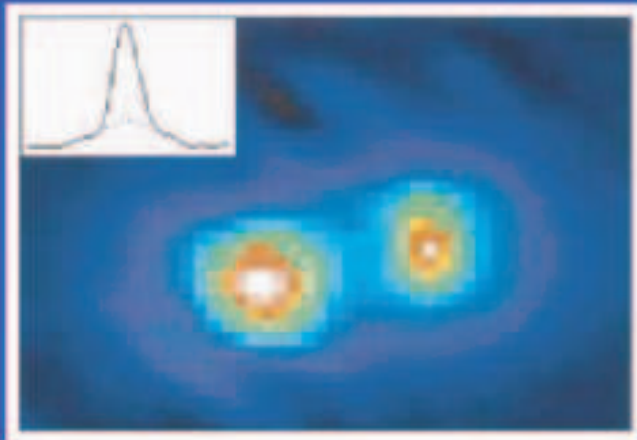
W. Ketterle



Two-Component BECs

## Two-Component Condensates

JILA, 1997



- Experimental implementation
  - The BEC named Science Magazine's "Molecule of the Year 1995"!
  - Nobel Prize in Physics (2001), E. A. Cornell, C. E. Wieman (JILA), W. Ketterle (MIT):  
for the achievement of BEC in dilute gases of alkali atoms, and for early fundamental studies of the properties of the condensates.
- Applications of BEC: atom laser, quantum computer, MEMS.
- Mathematical model: nonlinear Schrödinger equations, Gross-Pitaevskii equations (GPE), coupled nonlinear Schrödinger equations, coupled Gross-Pitaevskii equations (CGPE).
- Numerical simulation: method, guide for experiment etc.

## 2 Mathematical Models

(i) Time-dependent Gross-Pitaevskii equation

$$i u_t = -\Delta u + V_\epsilon(x, y) u + \frac{1}{\epsilon^2} (|u|^2 - 1) u, \quad t > 0, \quad (2.1)$$

with the initial data  $u|_{t=0} = u_0(x, y)$  and  $(x, y) \in \mathbb{R}^2$ .

$u$ : a complex-valued order parameter,

$\epsilon > 0$ : a small parameter,

$V_\epsilon(x, y) = \alpha_\epsilon x^2 + \beta_\epsilon y^2$ : a harmonic trap potential,

$\alpha_\epsilon, \beta_\epsilon > 0$ : depending on  $\epsilon$ .

This time-dependent Gross-Pitaevskii equation was introduced as a phenomenological equation for the order parameter in superfluids.

(ii) Coupled Gross-Pitaevskii equations

$$\begin{cases} \iota\hbar\frac{\partial\psi_1(\mathbf{x},t)}{\partial t} = -\frac{\hbar^2}{2m_a}\nabla^2\psi_1 + V_1\psi_1 + \mu_{11}|\psi_1|^2\psi_1 + \mu_{12}|\psi_2|^2\psi_1, \\ \iota\hbar\frac{\partial\psi_2(\mathbf{x},t)}{\partial t} = -\frac{\hbar^2}{2m_a}\nabla^2\psi_2 + V_2\psi_2 + \mu_{22}|\psi_2|^2\psi_2 + \mu_{21}|\psi_1|^2\psi_2. \end{cases} \quad (2.2)$$

$$\mathbf{x} \in \Omega \in \mathbb{R}^{2,3}, \quad \psi_j(\mathbf{x}, t) = 0, \quad \mathbf{x} \in \partial\Omega, \quad j = 1, 2.$$

$\psi_j$ : macroscopic wave fts,  $V_j$ : trap potential,

$\mu_{jj}$ : intra-comp.,  $\mu_{ij}$  ( $i \neq j$ ): inter-comp. scattering lengths.

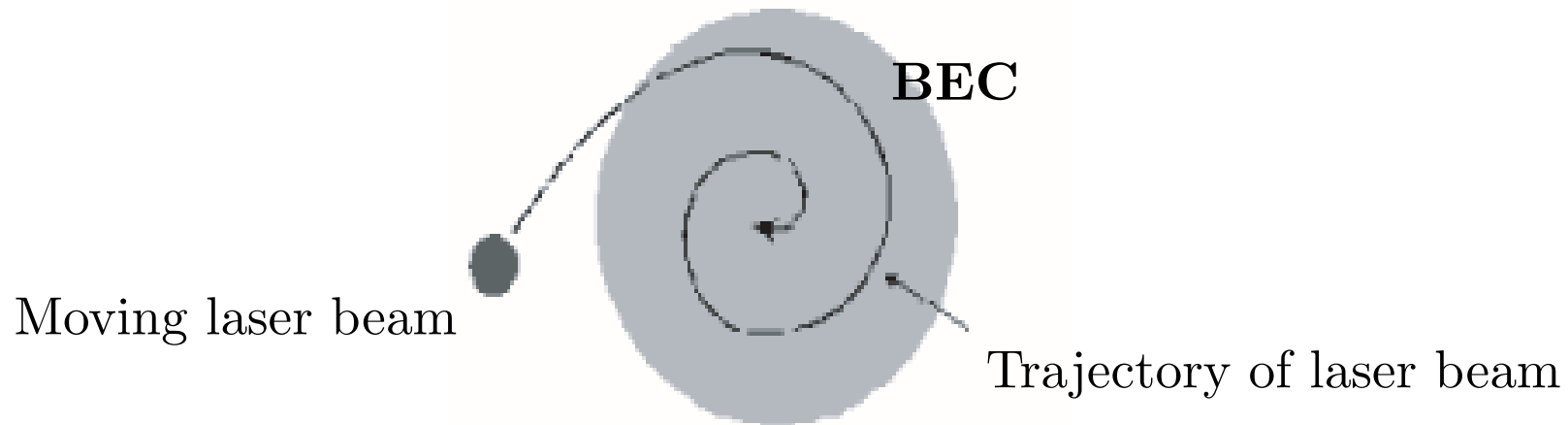
## Goal

- (i) Time-dependent Gross-Pitaevskii equation in BEC:  
to observe the motion of vortices.
  
- (ii) Coupled Gross-Pitaevskii equations in BECs:  
to solve Ground state & bound states.

### 3 Vortices in BEC

- How do vortices happen?
  - Idea 1: rotation (standard way in fluid mechanics).
  - Idea 2: laser beam moving slowly through the condensate (without rotation), by B. Jackson et al. (1998, theoretical); K. Staliunas (1999, experiment).

- Idea of K. Staliunas – stirred BEC:
  - (1) Create one component BEC.
  - (2) The laser beam enters the condensate spiraling clockwise.
  - (3) Reaching the center of the condensate it is switched off.





## Numerical Simulation

- Make a study of vortices's behavior in a two-dimensional trapped BEC.
  - PDE: time-dependent Gross-Pitaevskii equation.
  - ODE: asymptotic motion equations of vortices.

- Dynamics of vortices in trapped BEC

Suppose  $u_0$  has  $d$  vortex centers at  $q_j(0) = (q_{jx}(0), q_{jy}(0))^\top$ .

Under some specific assumptions on  $u_0$ , we obtain the asymptotic motion equations of  $d$  vortices  $q_j$ 's in the following:  
(T. C. Lin done)

$$\left\{ \begin{array}{l} \dot{q}_{jx} = - \sum_{\substack{k=1 \\ k \neq j}}^d n_k \frac{q_{jy} - q_{ky}}{|q_j - q_k|^2} - \omega_1 q_{jy} , \\ \dot{q}_{jy} = \sum_{\substack{k=1 \\ k \neq j}}^d n_k \frac{q_{jx} - q_{kx}}{|q_j - q_k|^2} + \omega_2 q_{jx} , \end{array} \right. \quad (3.1)$$

where  $q_j = q_j(t) = (q_{jx}(t), q_{jy}(t))$ ,  $n_j$ : winding numbers and  $\omega_1 = -\omega + 2\beta_0$ ,  $\omega_2 = -\omega + 2\alpha_0$ . For the stability of the vortex structure in  $u$ , we require  $n_j \in \{\pm 1\}$ ,  $j = 1, \dots, d$ .

- Characterize the motion:
  - Lyapunov exponent,
  - Poincaré map,
  - Spectrums of waveforms.
- Indicator for ratio topologically synchronized chaotic regimes (Afraimovich et al. (1999, 2000)):
  - the Poincaré dimension for Poincaré recurrences.

## Numerical Results

We consider  $d = 3$ , then obtain

- (1) the bounded and collisionless trajectories of three vortices form chaotic, quasi 2- or quasi 3-periodic orbits,
- (2) a new phenomenon of 1 : 1-topological synchronization is observed in the chaotic trajectories of vortices with the same sign of winding numbers..

Case  $(n_1, n_2, n_3) = (1, -1, -1)$

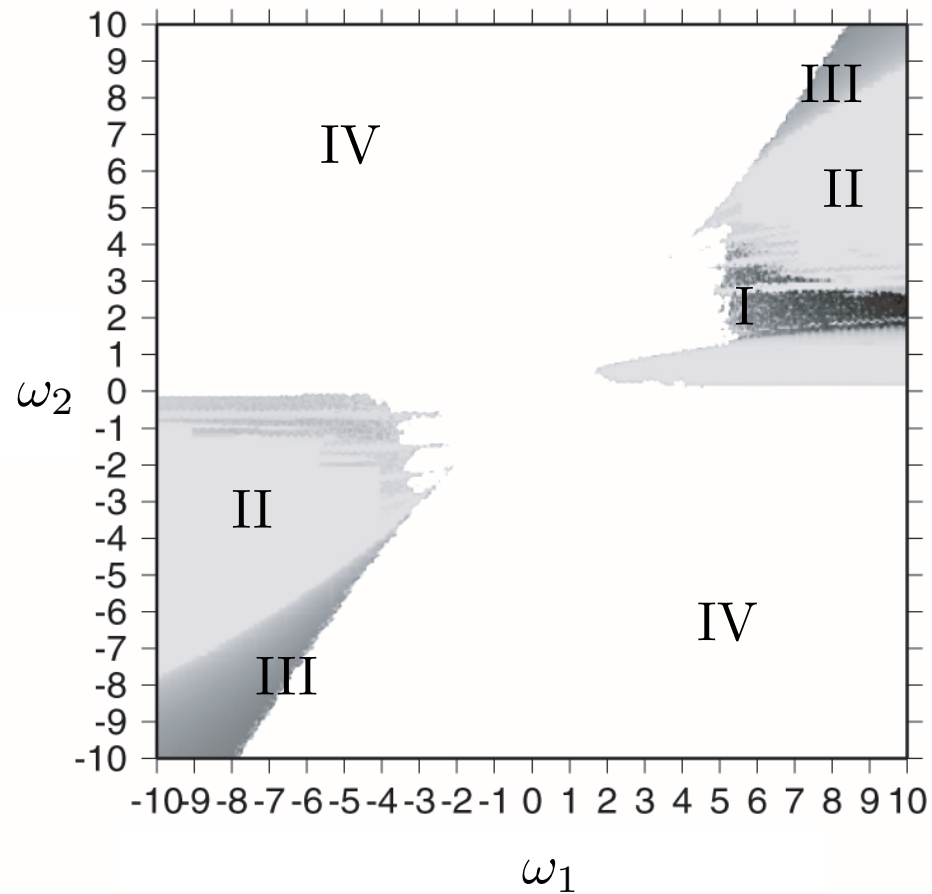


Figure 3.1: The first Lyapunov exponent.

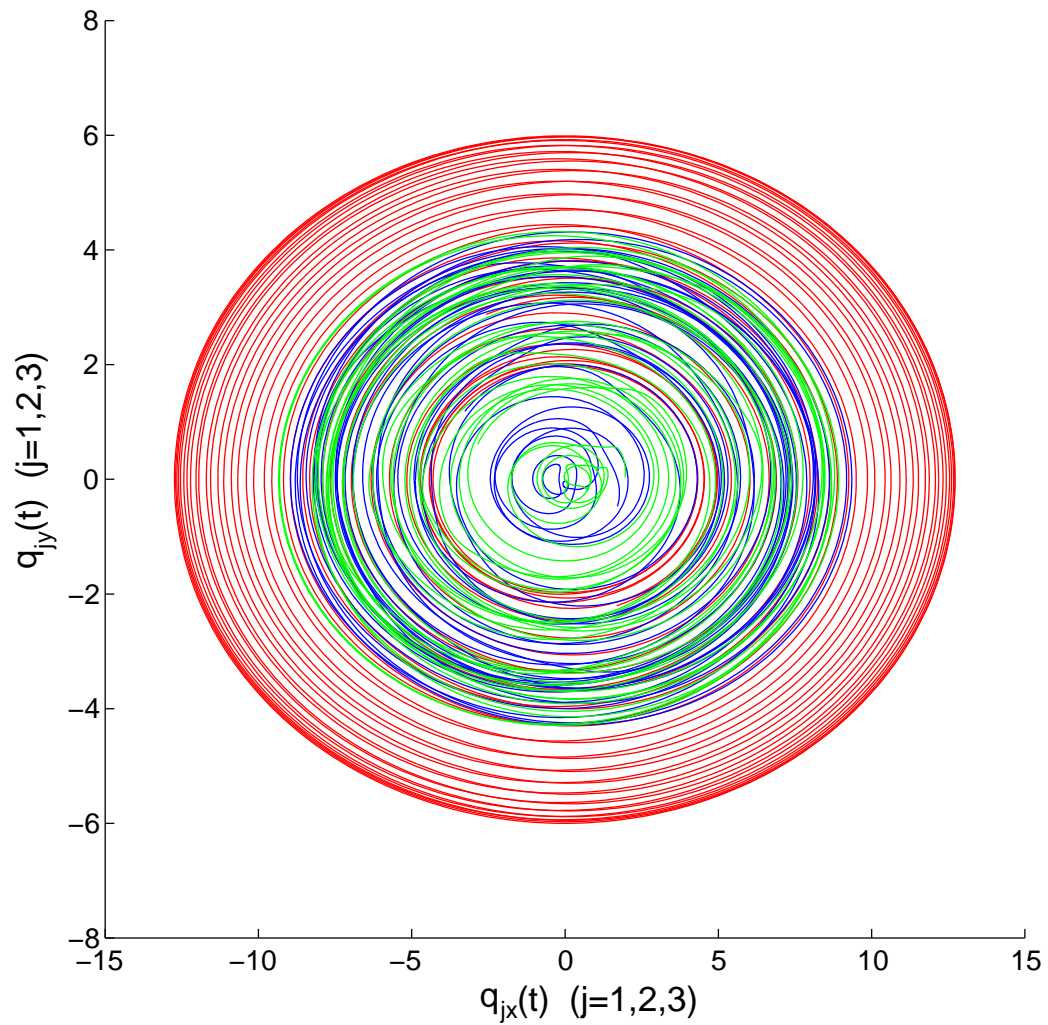


Figure 3.2: Chaotic trajectories:  $(\omega_1, \omega_2) = (9.88, 2.24)$ ,  $t = 25,050 \sim 25,100$  sec.

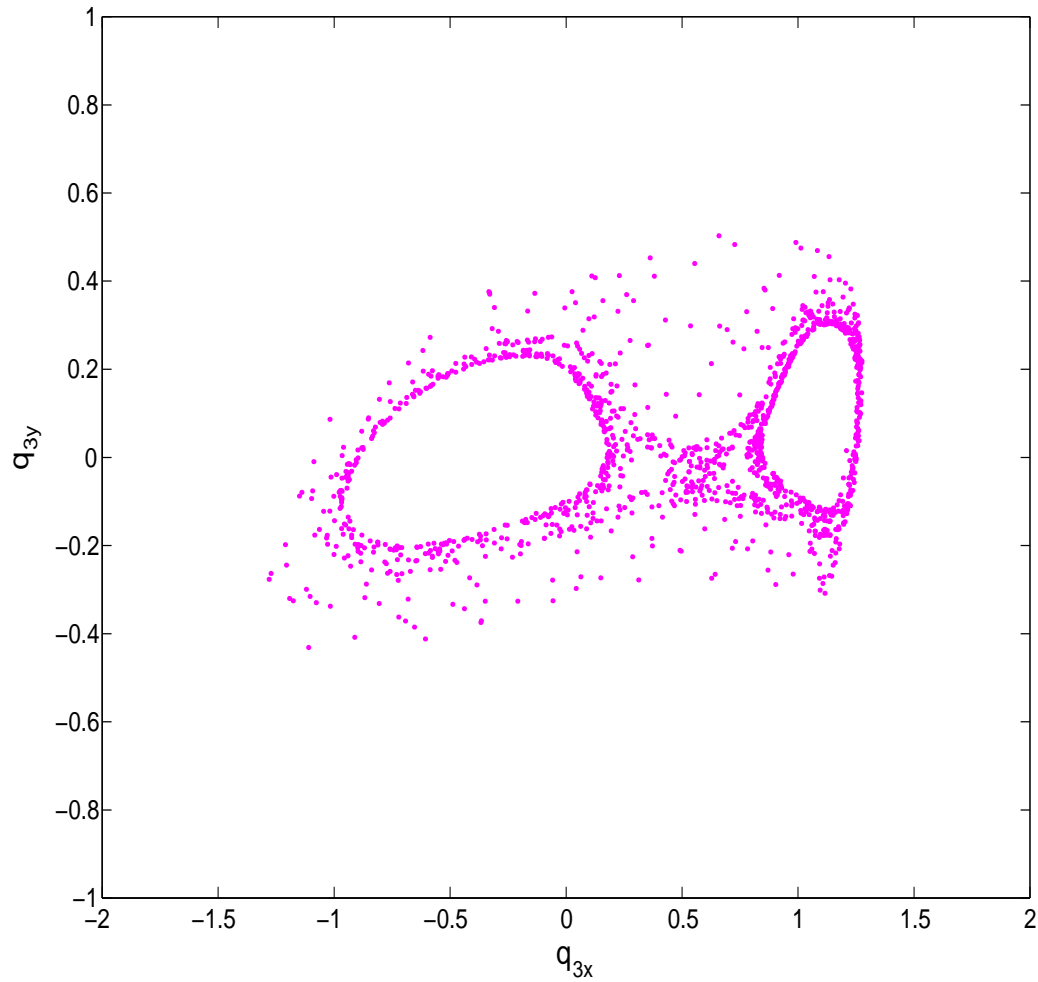


Figure 3.3: Chaotic second-Poincaré maps (4 dim.),  $t = 1,393 \sim 5,000,000$  sec.

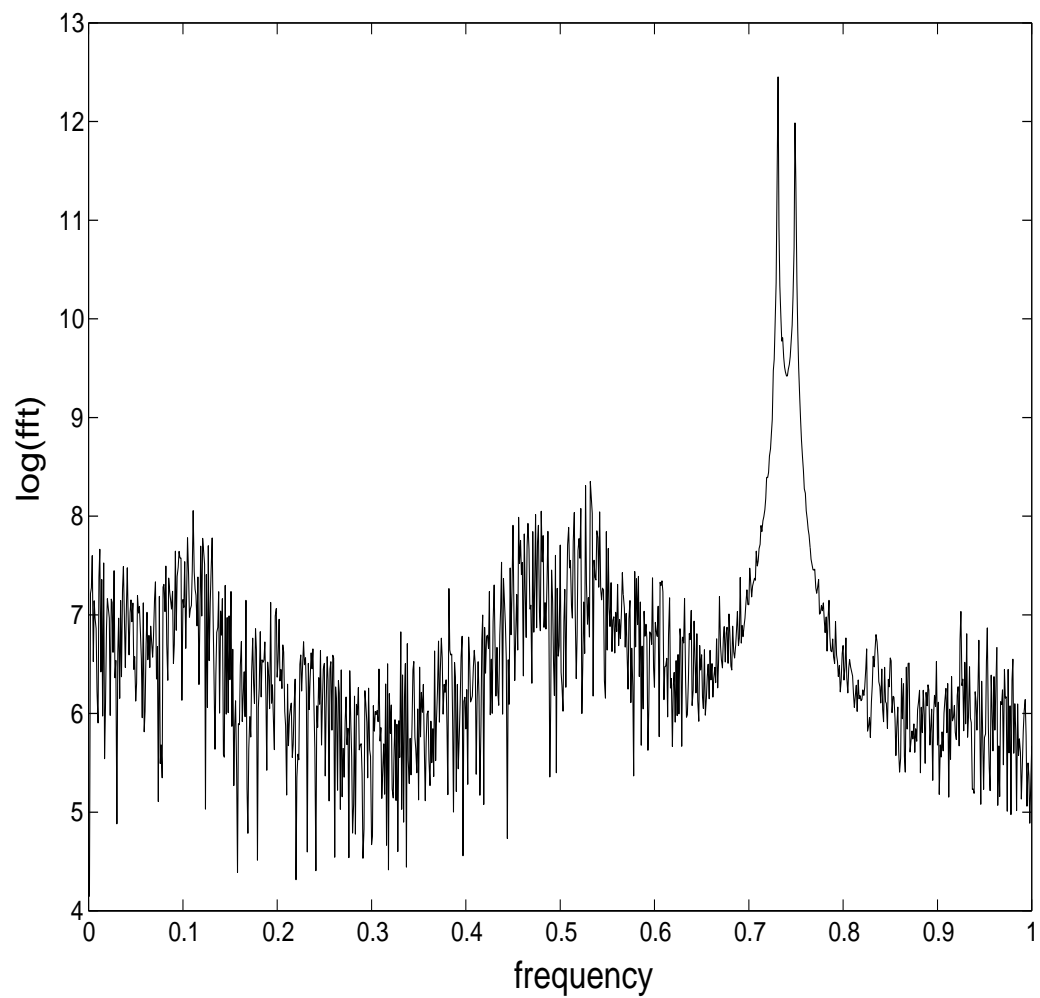


Figure 3.4: Chaotic spectrum of waveforms,  $t = 1,050 \sim 25,500$  sec.



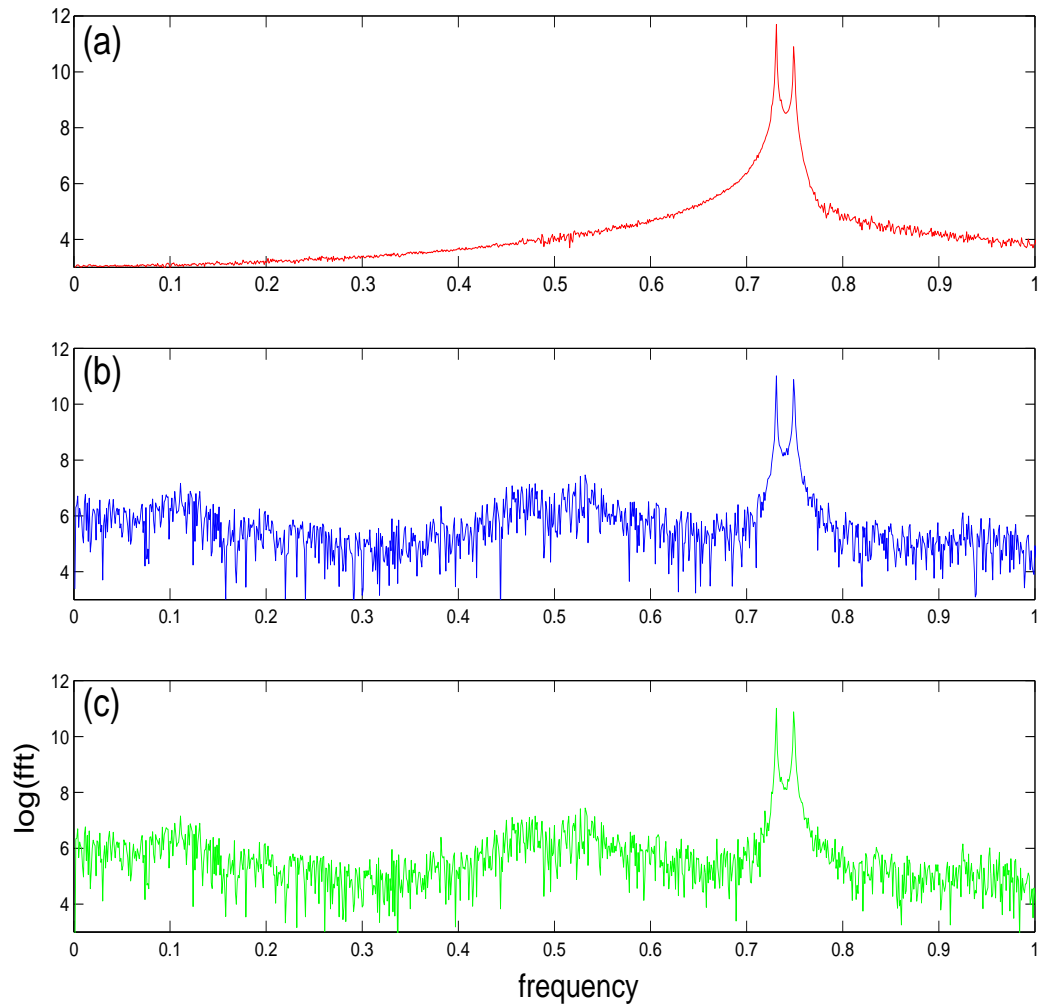


Figure 3.5: Chaotic individual spectrum,  $t = 1,000 \sim 25,500$  sec.

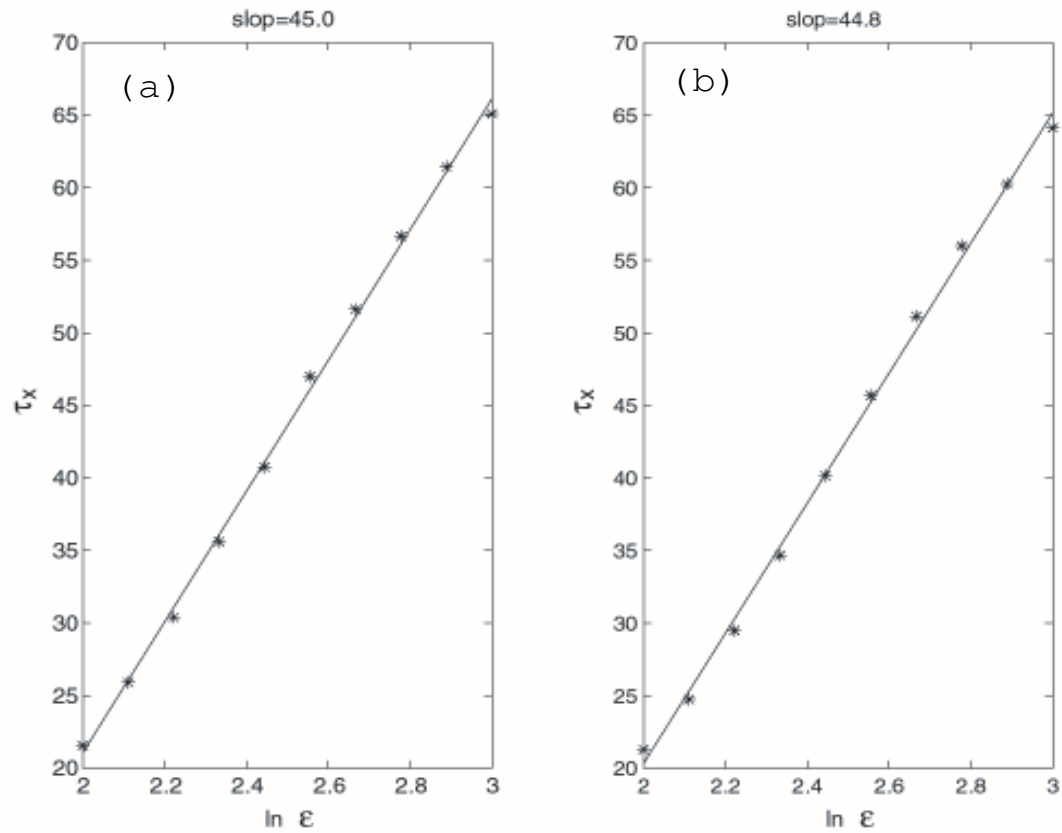


Figure 3.6: The ratio of slopes =  $45.0/44.8 \approx 1.006$

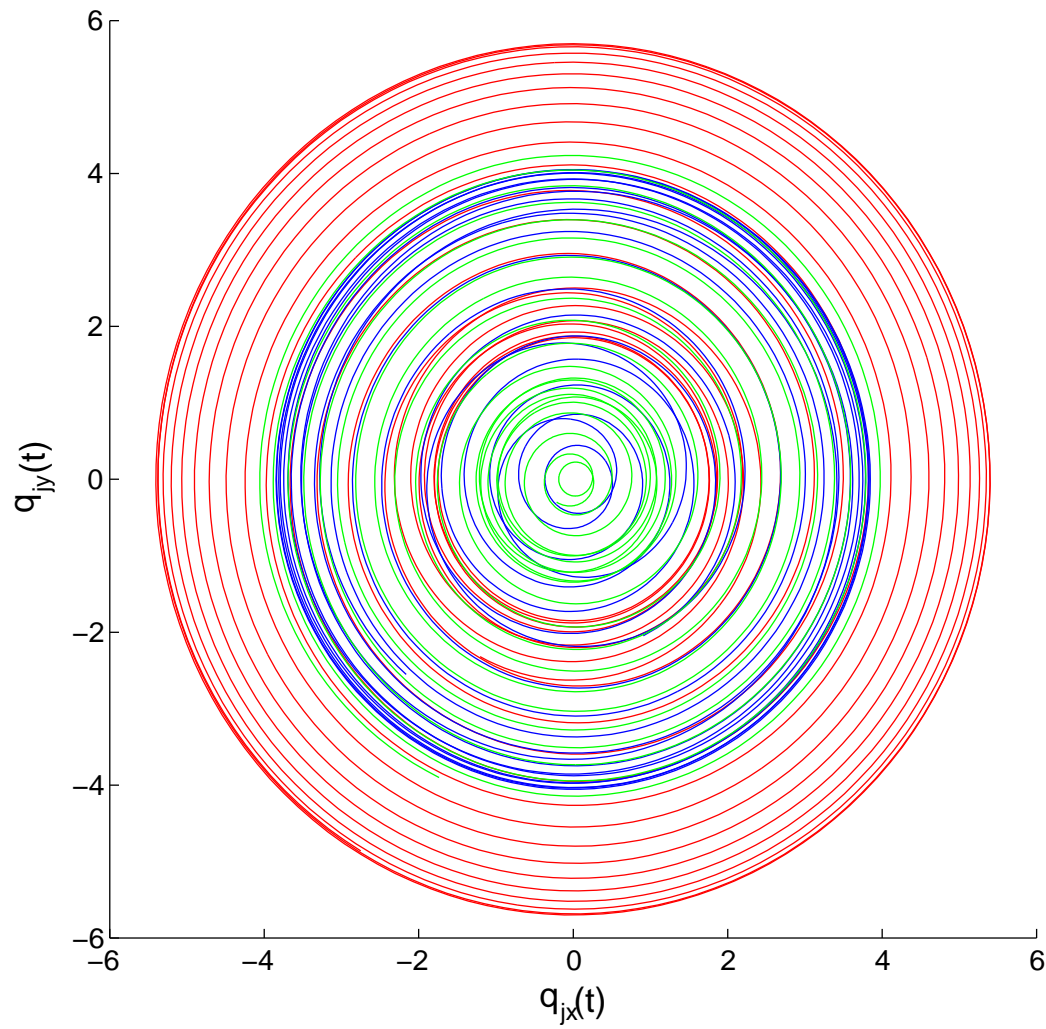


Figure 3.7: Quasi 3-periodic trajectories:  $(\omega_1, \omega_2) = (9, 10), t = 25,080 \sim 25,095$  sec.

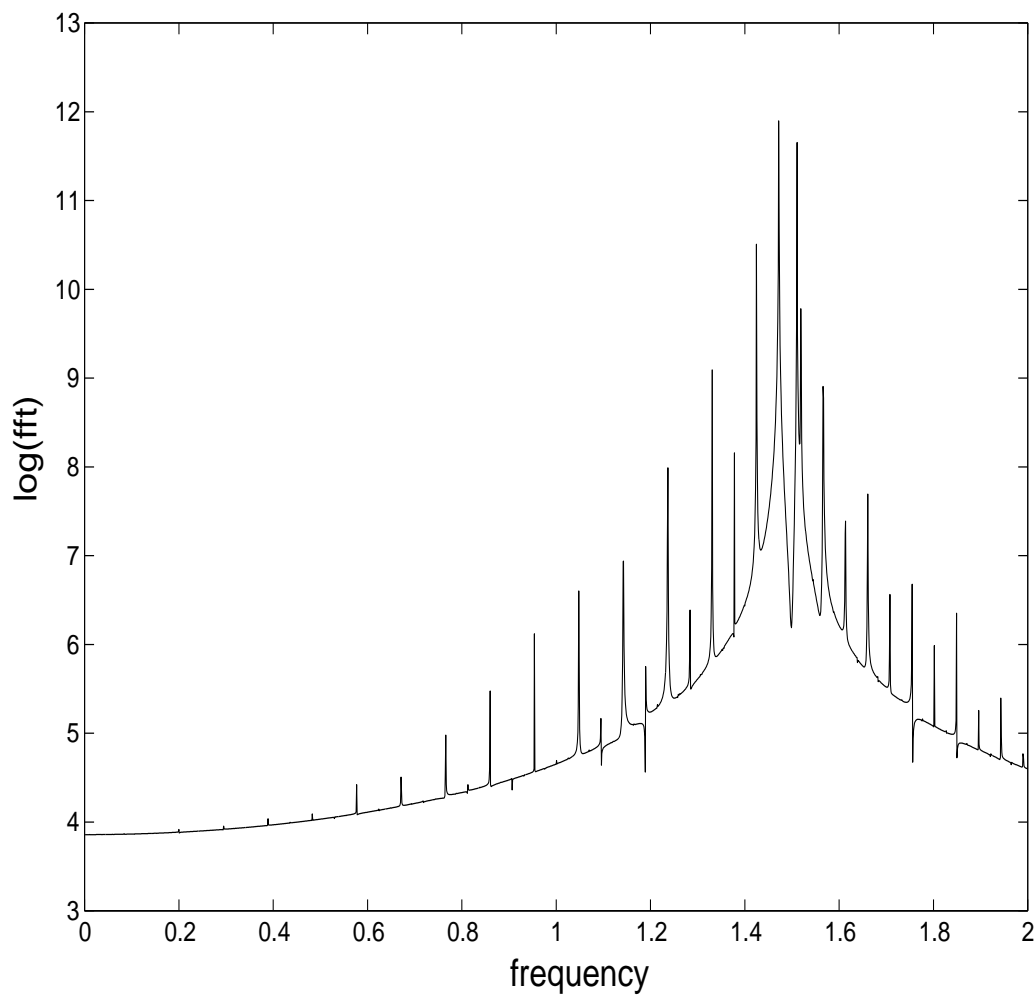


Figure 3.8: Quasi 3-periodic spectrum,  $t = 1,000 \sim 25,500$  sec.

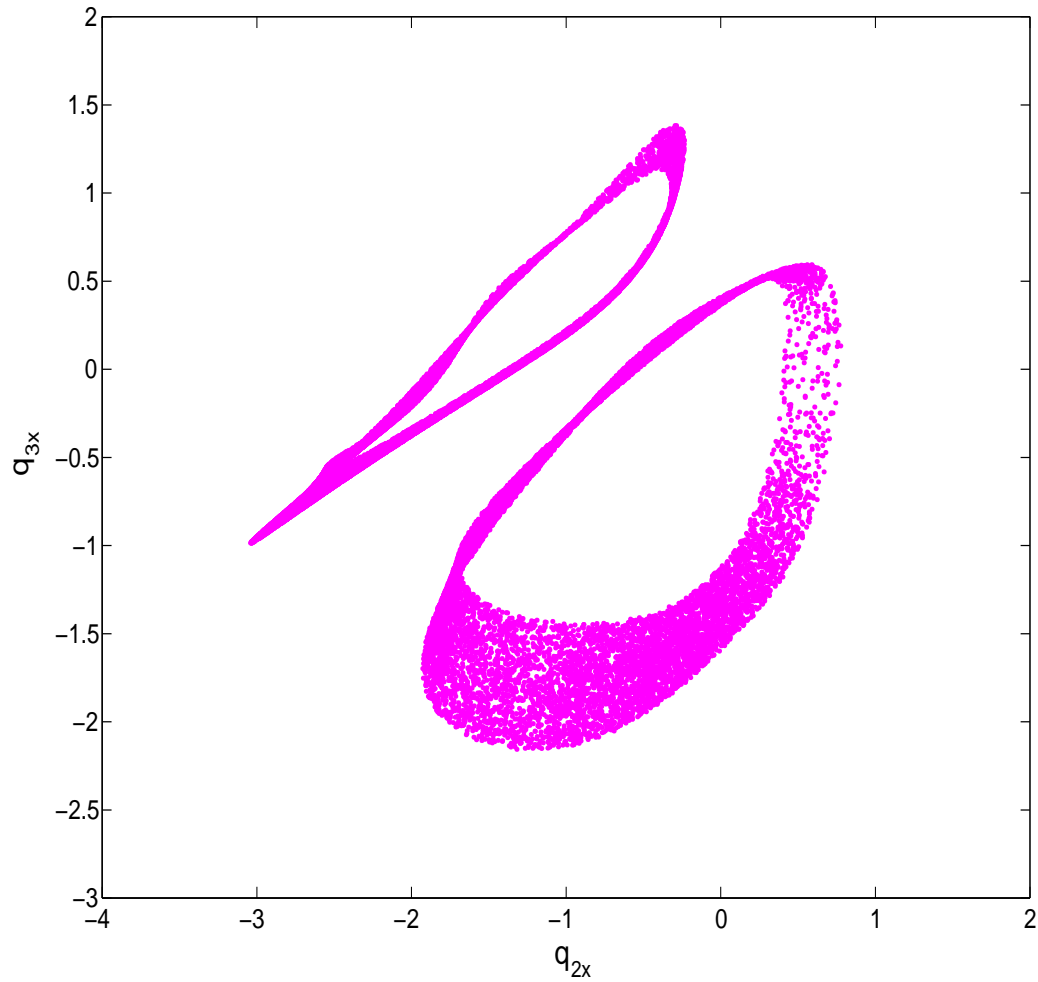


Figure 3.9: Quasi 3-periodic second-order Poincaré maps (4 dim.),  
 $t = 41,179 \sim 4,000,000$  sec.

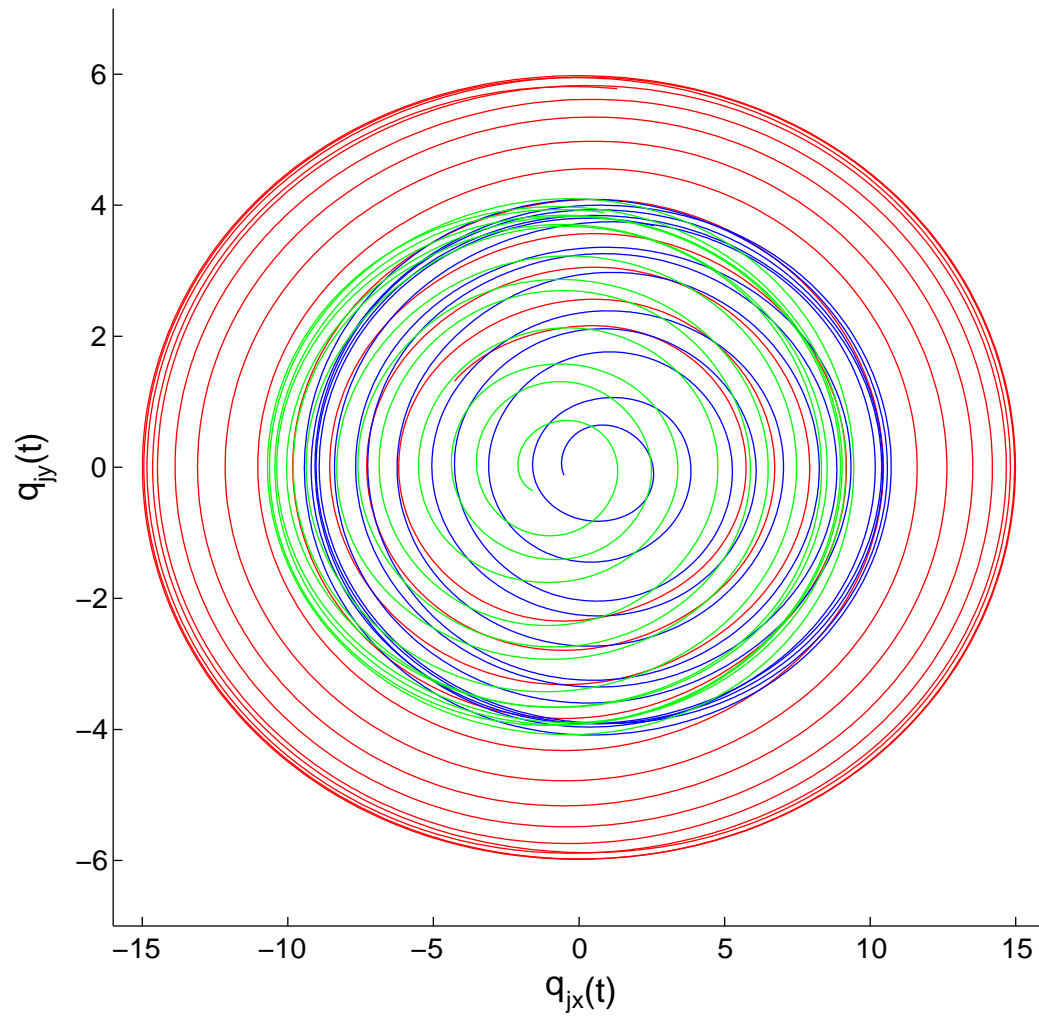


Figure 3.10: Quasi 2-periodic trajectories:  $(\omega_1, \omega_2) = (6, 1), t = 25, 155 \sim 25, 190$  sec.

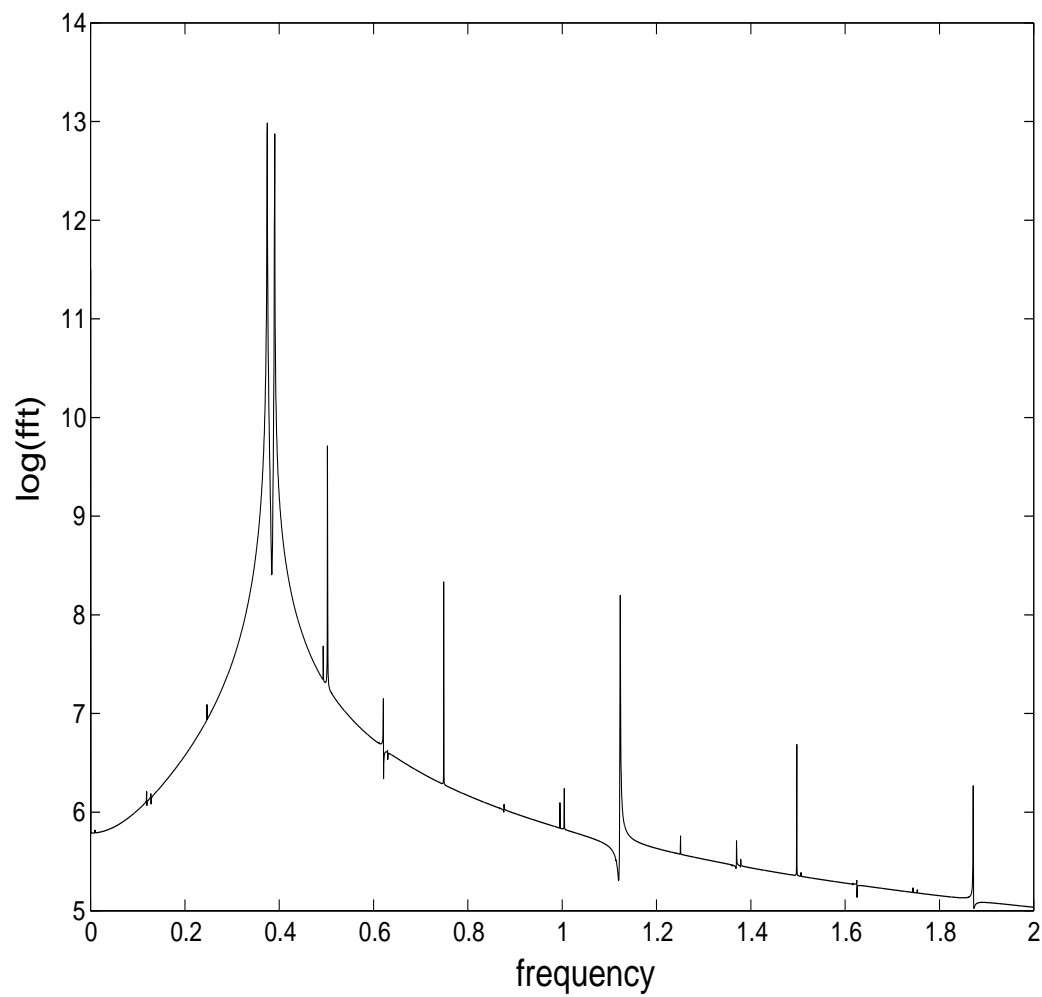


Figure 3.11: Quasi 2-periodic spectrum,  $t = 2,000 \sim 25,500$  sec.

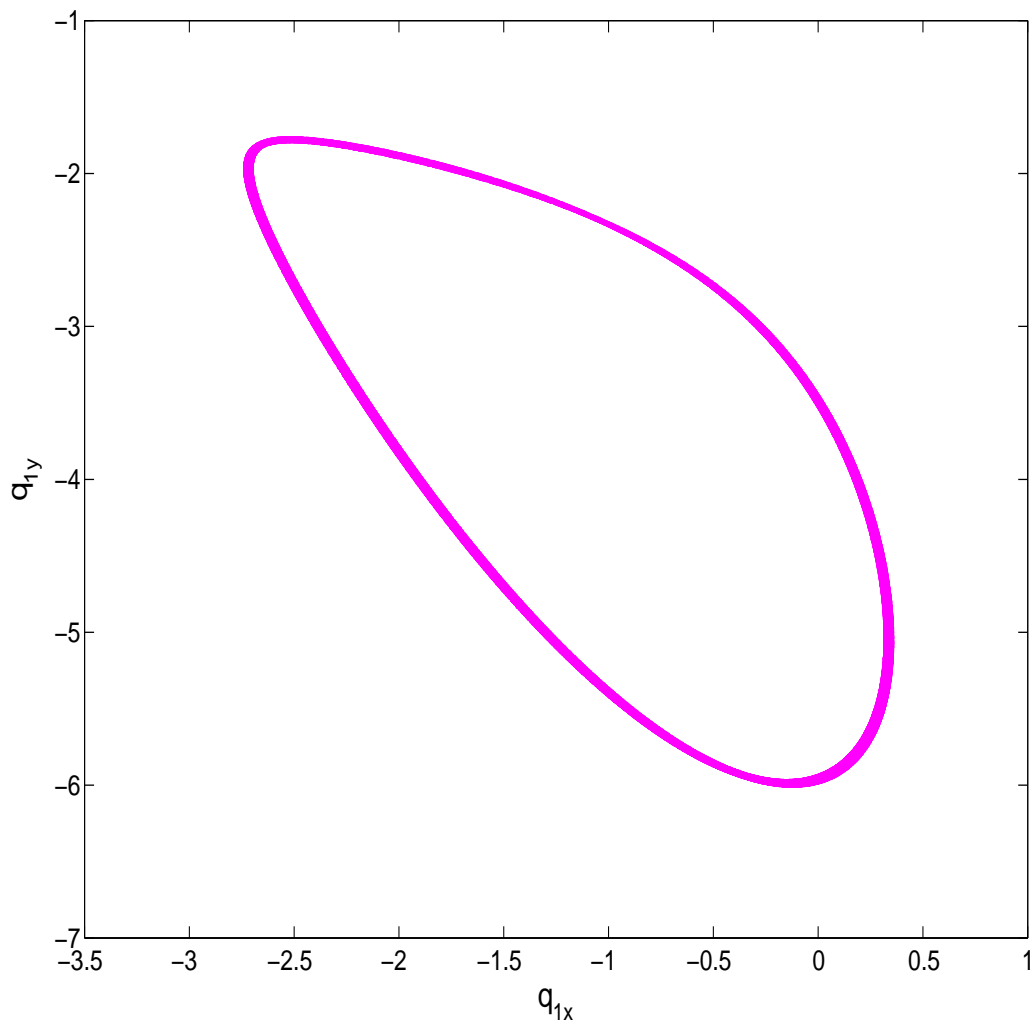


Figure 3.12: Quasi 2-periodic first-order Poincaré maps (5 dim.),  
 $t = 37,193 \sim 1,000,000$  sec.



Case  $(n_1, n_2, n_3) = (1, 1, 1)$

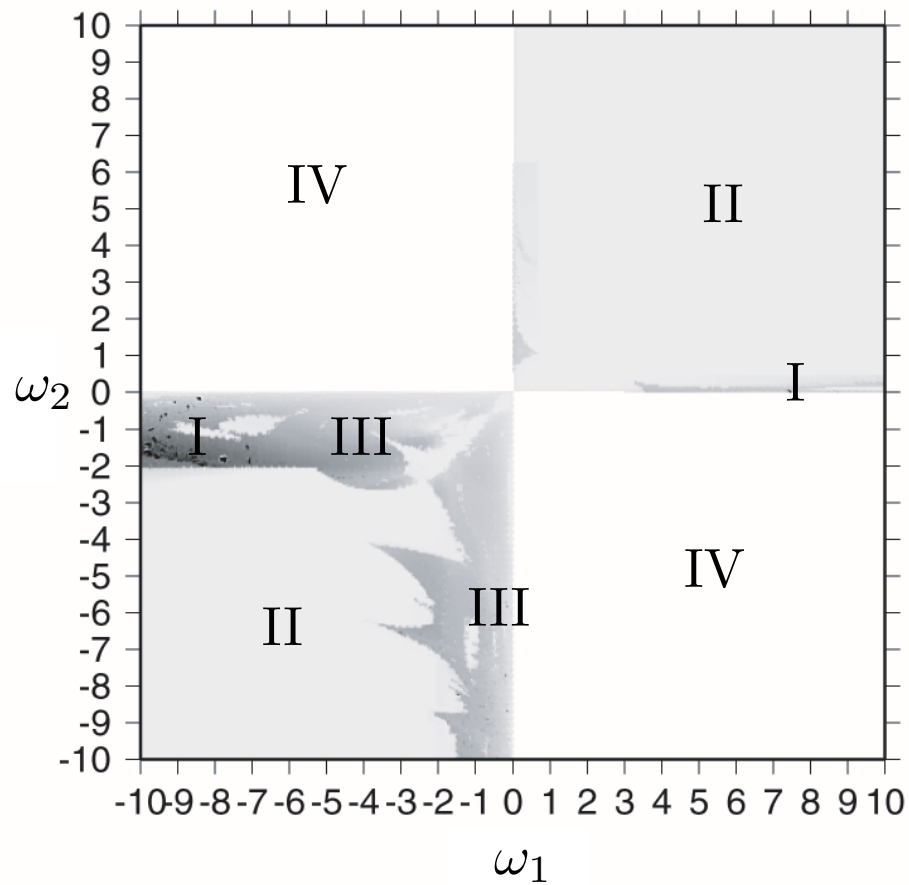


Figure 3.13: The first Lyapunov exponent.

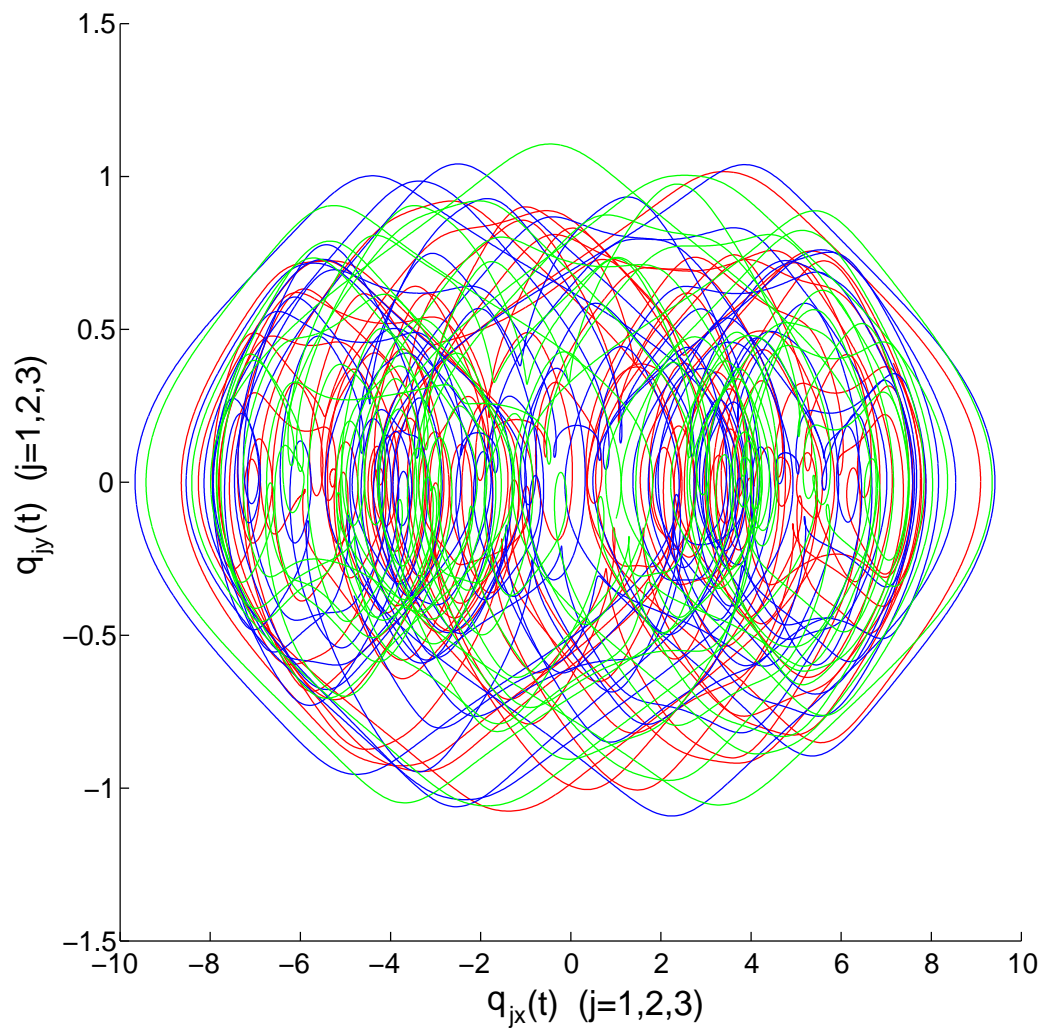


Figure 3.14: Chaotic trajectories:  $(\omega_1, \omega_2) = (7.4, 0.025)$ ,  $t = 25,050 \sim 25,070$  sec.

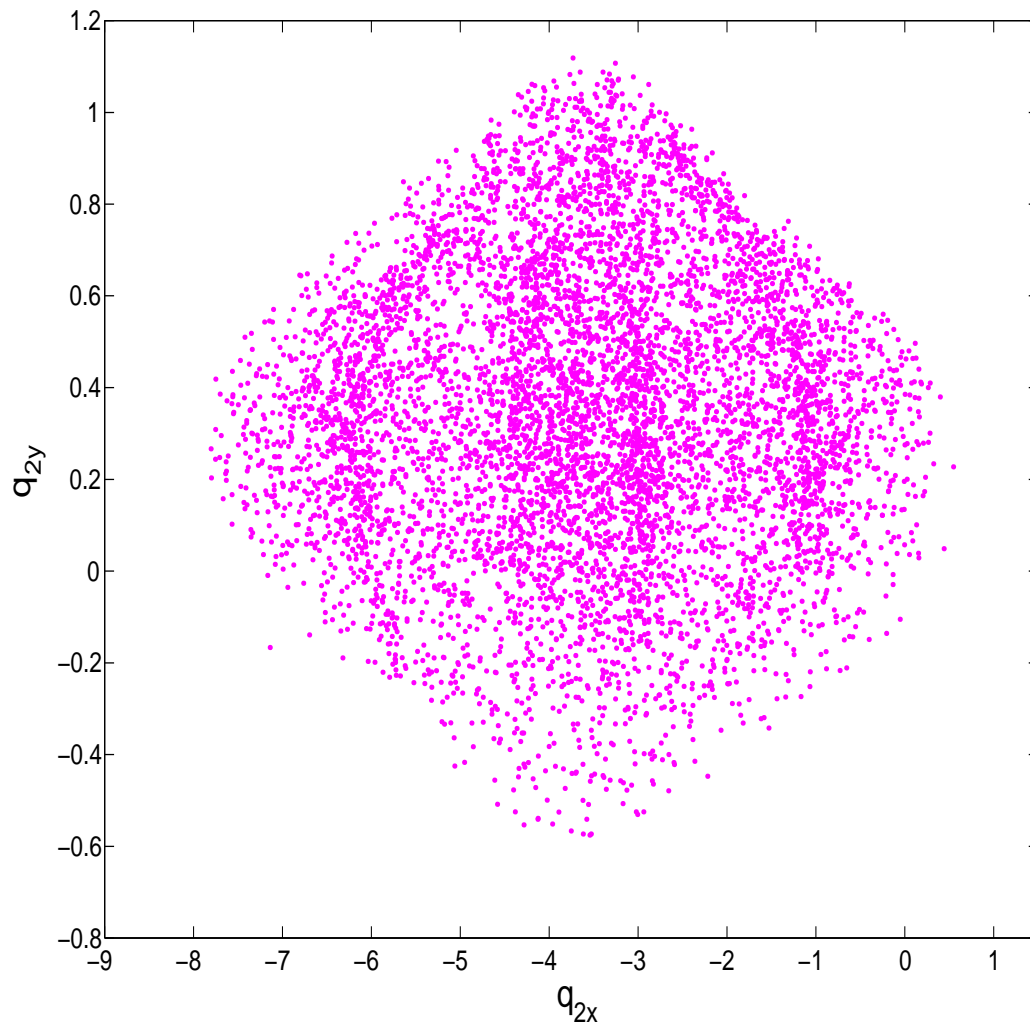


Figure 3.15: Chaotic first-order Poincaré maps (5 dim.),  $t = 1,000 \sim 100,000$  sec.

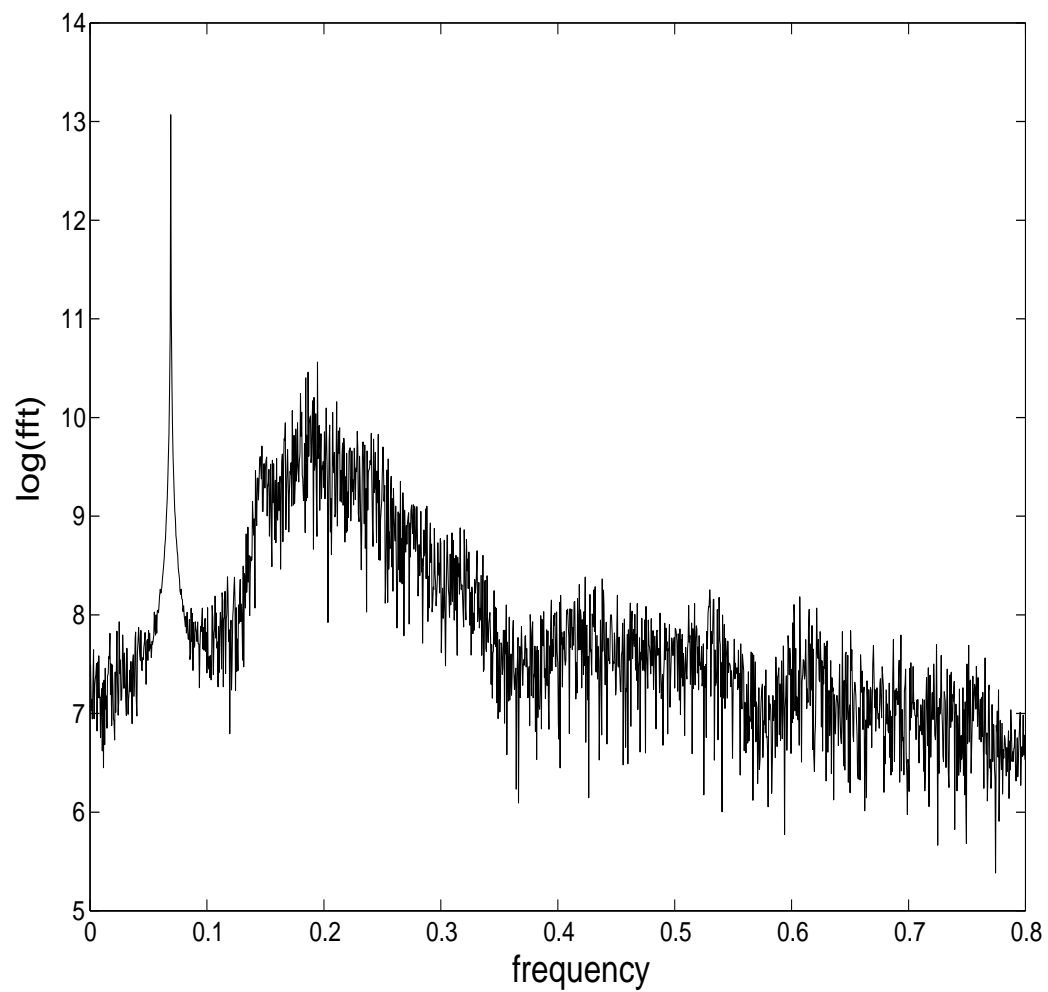


Figure 3.16: Chaotic spectrum,  $t = 2,000 \sim 25,500$  sec.

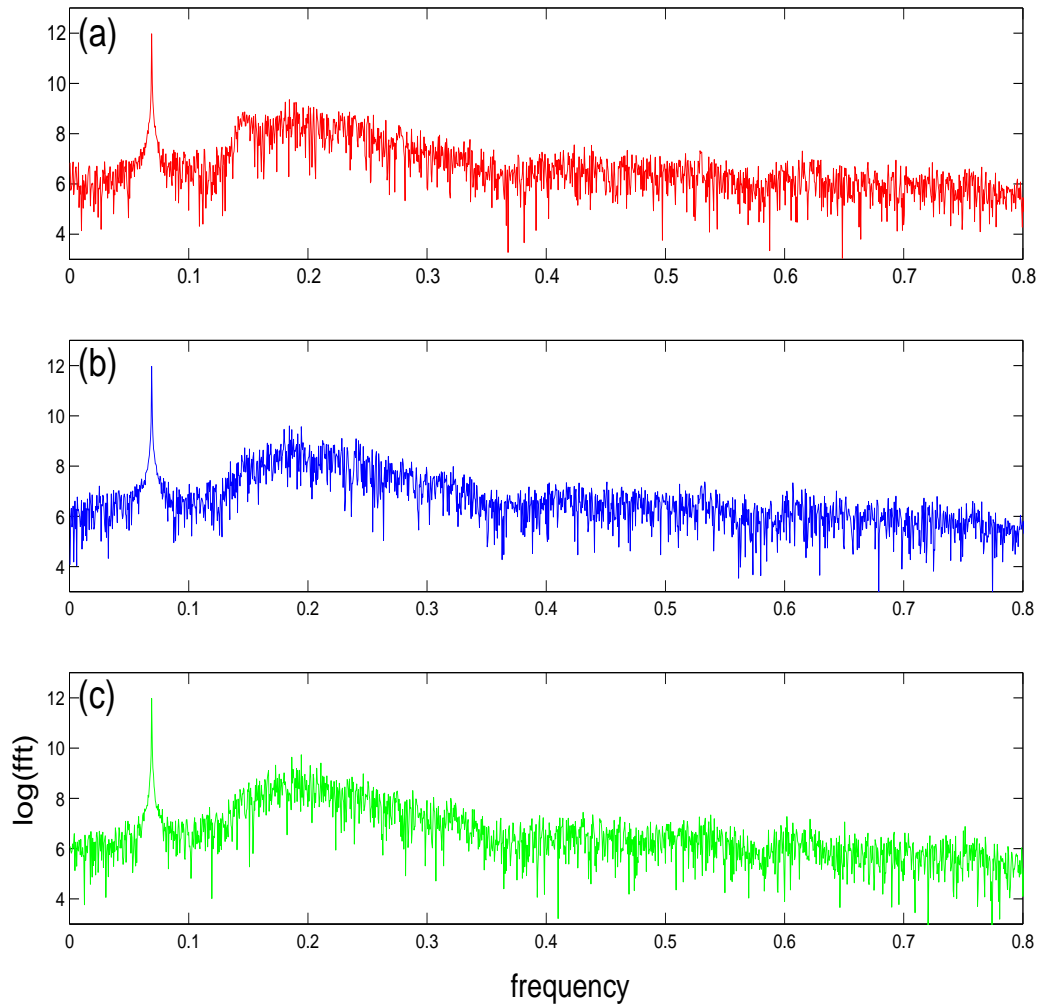


Figure 3.17: Chaotic individual spectrum,  $t = 2,000 \sim 25,500$  sec.

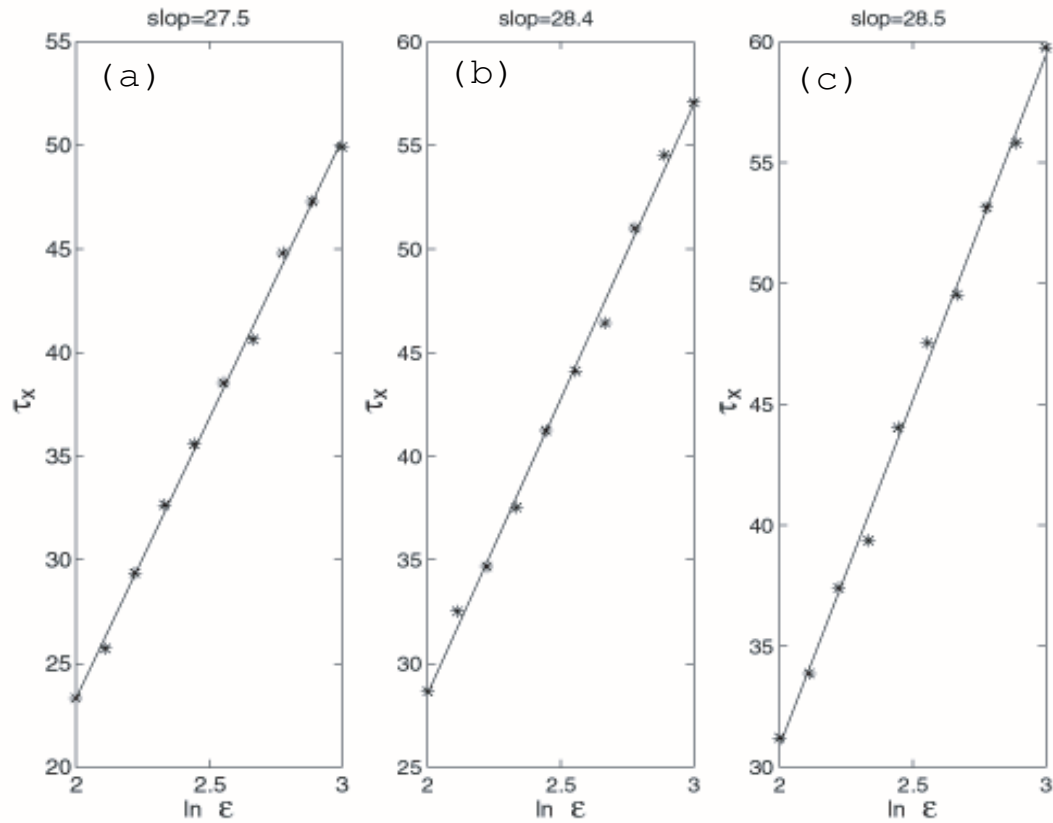


Figure 3.18: The ratio of slopes = 27.5 : 28.4 : 28.5  $\approx$  0.97 : 0.996 : 1.

## 4 Ground State & Bound States in BECs

- Dimensionless CGPE

$$\begin{cases} \iota \frac{\partial \psi_1(\mathbf{x}, t)}{\partial t} = -\nabla^2 \psi_1 + V_1 \psi_1 + \hat{\mu}_{11} |\psi_1|^2 \psi_1 + \hat{\mu}_{12} |\psi_2|^2 \psi_1, \\ \iota \frac{\partial \psi_2(\mathbf{x}, t)}{\partial t} = -\nabla^2 \psi_2 + V_2 \psi_2 + \hat{\mu}_{22} |\psi_2|^2 \psi_2 + \hat{\mu}_{21} |\psi_1|^2 \psi_2. \end{cases} \quad (4.1)$$

$$\mathbf{x} \in \Omega \in \mathbb{R}^{2,3}, \quad \psi_j(\mathbf{x}, t) = 0, \quad \mathbf{x} \in \partial\Omega, \quad j = 1, 2.$$

CGPE (4.1) conserve the normalization

$$n(\psi_j) := \int_{\mathbb{D}} |\psi_j(\mathbf{x}, t)|^2 d\mathbf{x} = 1, \quad j = 1, 2,$$

as well as the energy.

## Energy

$$E(\boldsymbol{\psi}) = \sum_{j=1}^2 \frac{N_j^0}{N_0} E_j(\boldsymbol{\psi}),$$

where  $N_j^0 > 0$  is the number of particles with  $N_1^0 + N_2^0 = N^0$  and

$$E_j(\boldsymbol{\psi}) = \int_{\mathbb{D}} \left[ \frac{1}{2} |\nabla \psi_j|^2 + V_j |\psi_j|^2 + \frac{1}{2} \sum_{k=1}^2 \hat{\mu}_{j,k} |\psi_j|^2 |\psi_k|^2 \right] d\mathbf{x},$$

for  $j = 1, 2$ .



Let  $\psi_j(\mathbf{x}, t) = e^{-\iota\lambda_j t}\phi_j(\mathbf{x})$ ,  $j = 1, 2$ . Substituting  $\psi_j$  into CGPE gives the time-indep. CGPE or NEP:

$$\begin{cases} -\nabla^2\phi_1(\mathbf{x}) + V_1(\mathbf{x})\phi_1(\mathbf{x}) + \hat{\alpha}_1|\phi_1|^2\phi_1(\mathbf{x}) + \hat{\beta}_1|\phi_2|^2\phi_1(\mathbf{x}) = \lambda_1\phi_1(\mathbf{x}), \\ -\nabla^2\phi_2(\mathbf{x}) + V_2(\mathbf{x})\phi_2(\mathbf{x}) + \hat{\alpha}_2|\phi_2|^2\phi_2(\mathbf{x}) + \hat{\beta}_2|\phi_1|^2\phi_2(\mathbf{x}) = \lambda_2\phi_2(\mathbf{x}), \end{cases} \quad (4.2a)$$

for  $\mathbf{x} \in \Omega \subseteq \mathbb{R}^2$  or  $\mathbb{R}^3$  with

$$\int_{\Omega} |\phi_j(\mathbf{x})|^2 d\mathbf{x} = 1, \quad \phi_j(\mathbf{x}) = 0, \quad \mathbf{x} \in \partial\Omega, \quad j = 1, 2, \quad (4.2b)$$

where  $\hat{\alpha}_1 = \alpha_{11}N_1^0$ ,  $\hat{\alpha}_2 = \alpha_{22}N_2^0$ ,  $\hat{\beta}_1 = \beta_{12}N_2^0$ ,  $\hat{\beta}_2 = \beta_{21}N_1^0$ , with  $\beta_{12} = \beta_{21} > 0$ ,

$\phi_j(\mathbf{x})$  : the corres. condensate solitary wave functions

$V_j(\mathbf{x})$  : magnetic trap potentials

$\hat{\alpha}_1 = \alpha_{11}N_1^0$ ,  $\hat{\alpha}_2 = \alpha_{22}N_2^0$  and

$\hat{\beta}_1 = \beta_{12}N_2^0$ ,  $\hat{\beta}_2 = \beta_{21}N_1^0$ , with  $\beta_{12} = \beta_{21} > 0$ ,

$N_j^0$  : the number of particles of the  $j$ -th component

$\alpha_{11}, \alpha_{22}$  : the intra-component scattering lengths,

$\beta_{12}, \beta_{21}$  : inter-component (repulsive) scattering lengths.

$$\begin{aligned}
& \text{Minimize } E(\boldsymbol{\phi}) \\
& \boldsymbol{\phi} = (\phi_1, \phi_2) \\
& \text{subject to } \int_{\Omega} |\phi_j(\mathbf{x})|^2 d\mathbf{x} = 1, \quad \phi_j(\mathbf{x}) = 0, \quad \mathbf{x} \in \partial\Omega, \\
& \phi_j(\mathbf{x}) > 0, \quad \mathbf{x} \in \Omega, \quad j = 1, 2,
\end{aligned} \tag{4.3}$$

where

$$E(\boldsymbol{\phi}) = 2 \sum_{j=1}^2 \frac{N_j^0}{N^0} E_j(\boldsymbol{\phi}).$$

with  $N^0 = N_1^0 + N_2^0$ ,

$$E_j(\boldsymbol{\phi}) = \int_{\Omega} \left( \frac{1}{2} |\nabla \phi_j|^2 + \frac{1}{2} V_j |\phi_j|^2 + \frac{\hat{\alpha}_j}{4} |\phi_j|^4 \right) + \frac{\hat{\beta}_j}{4} \int_{\Omega} |\phi_j|^2 |\phi_k|^2,$$

$k \neq j$ ,

for  $j, k = 1, 2$ .

# Nonlinear Algebraic Eigenvalue Problems (NAEP)

For the study of bifurcation and computation, we derive the discretization of NEP and the associated opt. problem. We consider  $\Omega \subseteq \mathbb{R}^2$  a bounded domain.

The central finite difference discretizes  $-\nabla^2 \phi_j(\mathbf{x})$  into

$$\mathbf{A}\mathbf{u}_j = \mathbf{A}[u_{j1}, \dots, u_{jl}, \dots, u_{jN}]^\top, \quad \mathbf{A} \in \mathbb{R}^{N \times N},$$

where  $\mathbf{u}_j$  is an approx. of the  $j$ -th wave ft.  $\phi_j(\mathbf{x})$ .

- Parametrization

$$0 < \hat{\alpha}_1 := \alpha_1, \hat{\alpha}_2 := \alpha_2 \leq K \text{ (bounded),}$$

$$\hat{\beta}_1 := \beta \rho_1, \hat{\beta}_2 := \beta \rho_2 \quad (\beta \text{ sufficiently large})$$

with  $\rho_1/\rho_2 = N_2^0/N_1^0$ .

- Discretization

$$-\nabla^2 + V(\mathbf{x}) \rightarrow \mathbf{A} \in \mathbb{R}^{N \times N} \text{ (an irreducible M-matrix)}$$

$$\phi_j(\mathbf{x}) \rightarrow \frac{1}{h} \mathbf{u}_j, \quad \alpha_j \rightarrow h^2 \alpha_j, \quad \beta \rightarrow h^2 \beta$$

## NAEP & FOP

- Nonlinear algebraic eigenvalue problem (NAEP)

$$\mathbf{A}\mathbf{u}_1 + \alpha_1 \mathbf{u}_1^{(3)} + \beta \rho_1 \mathbf{u}_2^{(2)} \circ \mathbf{u}_1 = \lambda_1 \mathbf{u}_1, \quad \mathbf{u}_1^\top \mathbf{u}_1 = 1, \quad (4.1a)$$

$$\mathbf{A}\mathbf{u}_2 + \alpha_2 \mathbf{u}_2^{(3)} + \beta \rho_2 \mathbf{u}_1^{(2)} \circ \mathbf{u}_2 = \lambda_2 \mathbf{u}_2, \quad \mathbf{u}_2^\top \mathbf{u}_2 = 1. \quad (4.1b)$$

- Finite-dim. opt. problem (FOP):

$$\begin{aligned} & \min_{\mathbf{u}=(\mathbf{u}_1, \mathbf{u}_2)} E(\mathbf{u}) \\ & \text{subject to } \mathbf{u}_j^\top \mathbf{u}_j = 1, \quad \mathbf{u}_j > 0, \quad j = 1, 2, \end{aligned} \quad (4.2)$$

where

$$E(\mathbf{u}) = \sum_{j,k=1, k \neq j}^2 \rho_k \left( \frac{1}{2} \mathbf{u}_j^\top \mathbf{A} \mathbf{u}_j + \frac{\alpha_j}{4} \mathbf{u}_j^{(2)\top} \mathbf{u}_j^{(2)} \right) + \frac{\beta \rho_1 \rho_2}{2} \mathbf{u}_1^{(2)\top} \mathbf{u}_2^{(2)}.$$

**Notation:**  $\mathbf{u} \circ \mathbf{v} = (u_1 v_1, \dots, u_N v_N)$ ,  $\mathbf{u}^{(r)} = \mathbf{u} \circ \dots \circ \mathbf{u}$ .

# Gauss-Seidel Type Iteration for NAEP

Define

$$\mathcal{M} = \{\mathbf{v} \in \mathbb{R}^N \mid \mathbf{v}^\top \mathbf{v} = 1, \mathbf{v} \geq 0\}, \quad \overset{\circ}{\mathcal{M}} = \text{interior of } \mathcal{M}.$$

Recall NAEP:

$$\mathbf{A}\mathbf{u}_j + \mathbf{V}_j \circ \mathbf{u}_j + \sum_{k=1}^m \beta_{jk} \mathbf{u}_k^{(2)} \circ \mathbf{u}_j = \lambda_j \mathbf{u}_j, \quad \mathbf{u}_j^\top \mathbf{u}_j = 1, \quad j, k = 1, \dots, m.$$

$\mathbf{A}$  is diagonal dominant and  $\mathbf{A}\mathbf{e} \gneq 0$ , where  $\mathbf{e} = (1, \dots, 1)^\top$ . For  $\mathbf{V}_j \geq 0$  and  $(\mathbf{u}_1, \dots, \mathbf{u}_m) \in \prod_{j=1}^m \mathcal{M}$ , the matrix

$$\bar{\mathbf{A}}_j \equiv \mathbf{A}_j + \sum_{k=1}^m \llbracket \beta_{jk} \mathbf{u}_k^{(2)} \rrbracket,$$

with  $\mathbf{A}_j = \mathbf{A} + \llbracket \mathbf{V}_j \rrbracket$  is an irreducible  $M$ -matrix. Then  $\bar{\mathbf{A}}_j^{-1} \geq 0$  is an irreducible and nonnegative matrix.

By Perron-Frobenius Theorem,  $\exists!$  positive eigenvector  $\bar{\mathbf{u}}_j > 0$  with  $\bar{\mathbf{u}}_j^\top \bar{\mathbf{u}}_j = 1$  corr. to the max. eigenvalue  $\mu_j^{\max}$  of  $\bar{\mathbf{A}}_j^{-1}$ . i.e.,  $\bar{\mathbf{u}}_j > 0$  is uniquely determined by  $(\mathbf{u}_1, \dots, \mathbf{u}_m)$  and satisfies

$$\bar{\mathbf{A}}_j \bar{\mathbf{u}}_j \equiv \left( \mathbf{A}_j + \sum_{k=1}^m \llbracket \beta_{jk} \mathbf{u}_k^{\textcircled{2}} \rrbracket \right) \bar{\mathbf{u}}_j = \lambda_j^{\min} \bar{\mathbf{u}}_j,$$

where  $\lambda_j^{\min} = 1/\mu_j^{\max}$  and  $\bar{\mathbf{u}}_j^\top \bar{\mathbf{u}}_j = 1$ , for  $j = 1, \dots, m$ .



We now define a function  $\mathbf{f} : \prod_{j=1}^m \mathcal{M} \rightarrow \prod_{j=1}^m \mathcal{M}$  by

$$\mathbf{f}(\mathbf{u}_1, \dots, \mathbf{u}_m) = (\bar{\mathbf{u}}_1, \dots, \bar{\mathbf{u}}_m),$$

where  $\bar{\mathbf{u}}_j > 0$  is well-defined,  $j = 1, \dots, m$ .

**Theorem 4.1** *The function  $\mathbf{f}$  has a fixed point in  $\prod_{j=1}^m \overset{\circ}{\mathcal{M}}$ . In other words, there is a point  $(\mathbf{u}_1^*, \dots, \mathbf{u}_m^*) \in \prod_{j=1}^m \overset{\circ}{\mathcal{M}}$  and  $\boldsymbol{\lambda} = (\lambda_1^*, \dots, \lambda_m^*)$  which solve the NAEP, that is,*

$$\mathbf{A}_j \mathbf{u}_j^* + \sum_{k=1}^m \beta_{jk} \mathbf{u}_k^{*\textcircled{2}} \circ \mathbf{u}_j^* = \lambda_j^* \mathbf{u}_j^*, \quad j = 1, \dots, m.$$

Recall FOP:

$$\begin{aligned} \min \quad & E(\mathbf{u}) \\ \text{s.t.} \quad & \mathbf{u}_j^\top \mathbf{u}_j = 1, \quad j = 1, \dots, m, \end{aligned}$$

where

$$E(\mathbf{u}) \equiv \frac{1}{2} \sum_{j=1}^m \mathbf{u}_j^\top \mathbf{A}_j \mathbf{u}_j + \frac{1}{2} \sum_{1 \leq j < k \leq m} \beta_{jk} \mathbf{u}_k^{(2)\top} \mathbf{u}_j^{(2)}.$$

We define the restricted Lagrangian function of the opt. problem by

$$L(\mathbf{u}) = E(\mathbf{u}) - \frac{1}{2} \sum_{j=1}^m \lambda_j (\mathbf{u}_j^\top \mathbf{u}_j - 1).$$

**Theorem 4.2** Let  $\mathbf{u}^* = (\mathbf{u}_1^*, \dots, \mathbf{u}_m^*)$  be a KKT point of the opt. problem assoc. with the Lagrangian multipliers  $(\lambda_1^*, \dots, \lambda_m^*)$ .

Denote the Hessian of  $L(\mathbf{u})$  at  $\mathbf{u}^*$  by  $\nabla^2 L(\mathbf{u}^*) = [\nabla^2 L(\mathbf{u}^*)_{ij}]_{i,j=1}^m$ , where

$$\nabla^2 L(\mathbf{u}^*)_{jj} = \left( \mathbf{A}_j + \sum_{k=1}^m \llbracket \beta_{jk} \mathbf{u}_k^{*\circledast} \rrbracket - \lambda_j^* \mathbf{I}_N \right)$$

and

$$\nabla^2 L(\mathbf{u}^*)_{ij} = \nabla^2 L(\mathbf{u}^*)_{ji} = 2 \llbracket \beta_{ji} \mathbf{u}_i^* \circ \mathbf{u}_j^* \rrbracket, \quad j \neq i,$$

The positivity condition

$$\mathbf{d}^\top (\nabla^2 L(\mathbf{u}^*)) \mathbf{d} > 0$$

holds, for all  $\mathbf{d} = (\mathbf{d}_1^\top, \dots, \mathbf{d}_m^\top)^\top$  with  $\mathbf{u}_j^{*\top} \mathbf{d}_j = 0$ ,  $j = 1, \dots, m$ , if and only if  $\mathbf{u}^*$  is a strictly local minimum of the opt. problem.

## Jacobi Iteration (JI)

Define  $\mathbf{f} : \prod_{j=1}^m \mathcal{M} \rightarrow \prod_{j=1}^m \mathcal{M}$  by

$$\mathbf{f}(\mathbf{u}_1, \dots, \mathbf{u}_m) = (\bar{\mathbf{u}}_1, \dots, \bar{\mathbf{u}}_m),$$

where  $\bar{\mathbf{u}}_j > 0$  is well-defined,  $j = 1, \dots, m$ .

**Theorem 4.3** *Let  $(\boldsymbol{\lambda}^*, \mathbf{u}^*) = ((\lambda_1^*, \dots, \lambda_m^*), (\mathbf{u}_1^*, \dots, \mathbf{u}_m^*))$  be a fixed point of NAEP. If the JI converges to  $(\boldsymbol{\lambda}^*, \mathbf{u}^*)$  locally and linearly with an initial in  $\overset{\circ}{\prod}_{j=1}^m \mathcal{M}$ , then  $\mathbf{u}^* = (\mathbf{u}_1^*, \dots, \mathbf{u}_m^*)$  is a strictly local min. of the opt. problem.*

## Gauss-Seidel Iteration (GSI)

Define  $\mathbf{g} : \prod_{j=1}^m \mathcal{M} \rightarrow \prod_{j=1}^m \mathcal{M}$  by

$$\mathbf{g}(\mathbf{u}_1, \dots, \mathbf{u}_m) = (\bar{\mathbf{u}}_1, \dots, \bar{\mathbf{u}}_m),$$

where

$$\bar{\mathbf{u}}_1 = \mathbf{g}_1(\mathbf{u}_1, \dots, \mathbf{u}_m) = \mathbf{f}_1(\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m),$$

$$\bar{\mathbf{u}}_2 = \mathbf{g}_2(\mathbf{u}_1, \dots, \mathbf{u}_m) = \mathbf{f}_2(\bar{\mathbf{u}}_1, \mathbf{u}_2, \mathbf{u}_3, \dots, \mathbf{u}_m),$$

$$\vdots \qquad \qquad \qquad \vdots$$

$$\bar{\mathbf{u}}_m = \mathbf{g}_m(\mathbf{u}_1, \dots, \mathbf{u}_m) = \mathbf{f}_m(\bar{\mathbf{u}}_1, \bar{\mathbf{u}}_2, \dots, \bar{\mathbf{u}}_{m-1}, \mathbf{u}_m),$$

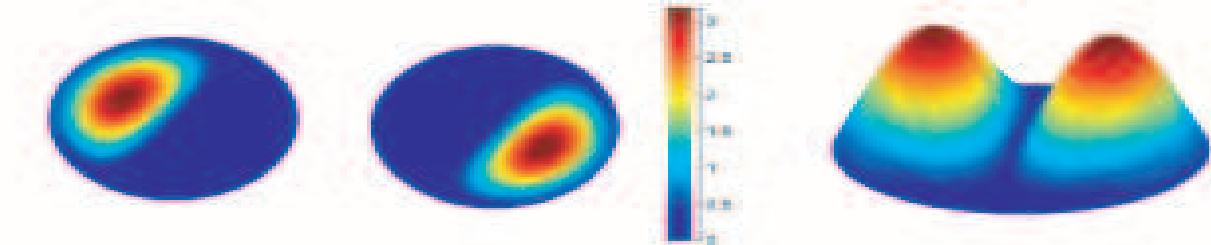
in which  $\{\mathbf{f}_j\}_{j=1}^m$  are given in JI. The ft.  $\mathbf{g}$  defines a Gauss-Seidel type iteration (GSI).

**Theorem 4.4** *Let  $(\boldsymbol{\lambda}^*, \mathbf{u}^*) = ((\boldsymbol{\lambda}_1^*, \dots, \boldsymbol{\lambda}_m^*), (\mathbf{u}_1^*, \dots, \mathbf{u}_m^*))$  be a fixed point of the NAEP. Suppose the matrix  $\mathbf{Z}^\top \nabla^2 L(\mathbf{u}^*) \mathbf{Z}$  is nonsingular. The GSI converges to  $(\boldsymbol{\lambda}^*, \mathbf{u}^*)$  locally and linearly with an initial in  $\prod_{j=1}^m \overset{\circ}{\mathcal{M}}$  iff  $\mathbf{u}^* = (\mathbf{u}_1^*, \dots, \mathbf{u}_m^*)$  is a strictly local min. of the opt. problem, provided  $\beta_{jj} > 0$  suff. small,  $j = 1, \dots, m$ .*

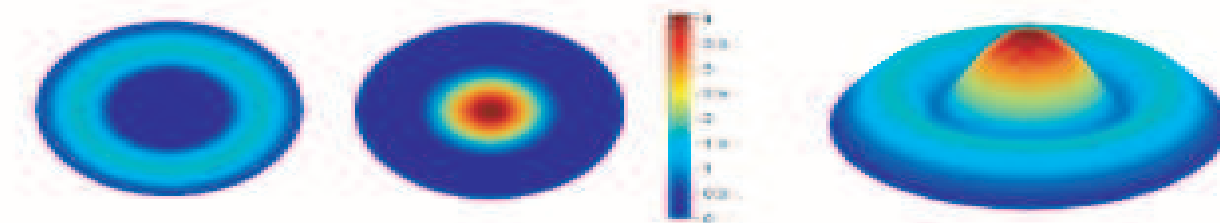
## Gauss-Seidel Iteration (GSI( $m$ ))

- (i) Given  $\mathbf{A}_j = \mathbf{A} + \llbracket \mathbf{V}_j \rrbracket + \beta_{jj} \llbracket \mathbf{u}_j^{(0)\textcircled{2}} \rrbracket$ ,  $\beta_{jj} \ll 0$ ,  $\beta_{jk} = \beta_{kj} \geq 0$  ( $j \neq k$ ),  $j, k = 1, \dots, m$  and  $\mathbf{u}_j^{(0)} > 0$  with  $\|\mathbf{u}_j^{(0)}\|_2 = 1$ ,  $n = 0$ ,
- (ii) Repeat  $n$ : until convergence,  
 For  $j = 1, \dots, m$ ,  
 Use e.g., the Jacobi-Davidson alg. to solve the min. pos. EW.  $\lambda_j^{(n+1)}$  of  $\mathbf{A}_j^{(n+1)}$  and the assoc. EV  $\mathbf{u}_j^{(n+1)}$  with  $\|\mathbf{u}_j^{(n+1)}\|_2 = 1$ , where
- $$\mathbf{A}_j^{(n+1)} := \mathbf{A}_j + \sum_{k < j} \llbracket \beta_{jk} \mathbf{u}_j^{(n+1)} \rrbracket + \sum_{k \geq j} \llbracket \beta_{jk} \mathbf{u}_j^{(n)} \rrbracket,$$
- Endfor  $j$ ;
- (iii) Compute  $\text{res}_j^{(n+1)} = \mathbf{A}_j^{(n+1)} \mathbf{u}_j^{(n+1)} - \lambda_j^{(n+1)} \mathbf{u}_j^{(n+1)}$ ,  $j = 1, \dots, m$ .
- (iv) If  $\|\text{res}_j^{(n+1)}\|_2 < \text{Tol}$ ,  $j = 1, \dots, m$ , then stop, else  $n \leftarrow n + 1$  go to repeat.

# Numerical Results



(a) green:  $\beta^* = 1000$ ,  $\lambda_1^* = \lambda_2^* = 7.07$ ,  $E(\mathbf{u}^*) = 7.02$



(b) red:  $\beta^* = 1000$ ,  $\lambda_1^* = 10.34$ ,  $\lambda_2^* = 14.54$ ,  $E(\mathbf{u}^*) = 12.43$



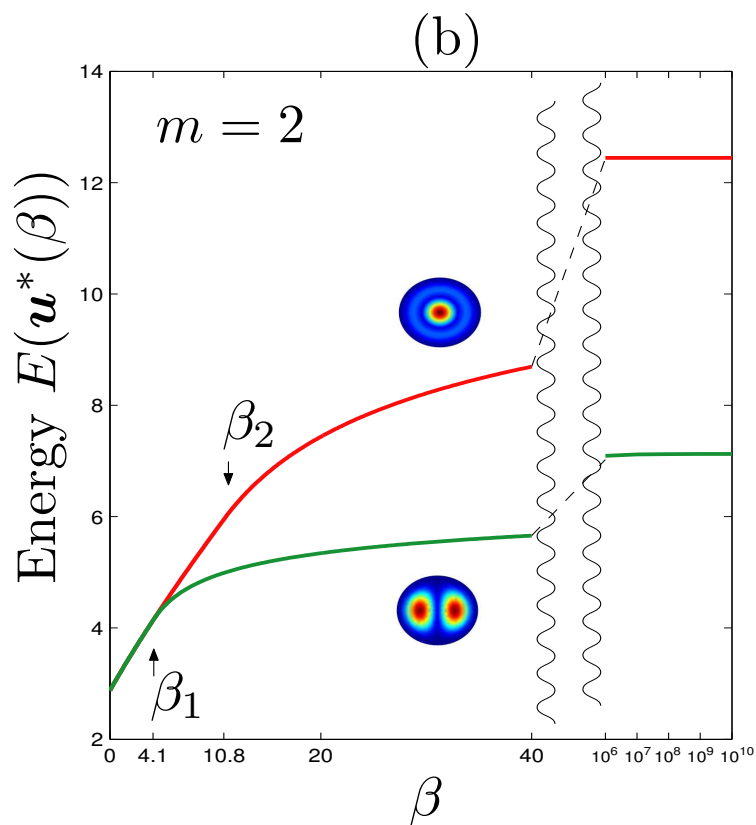
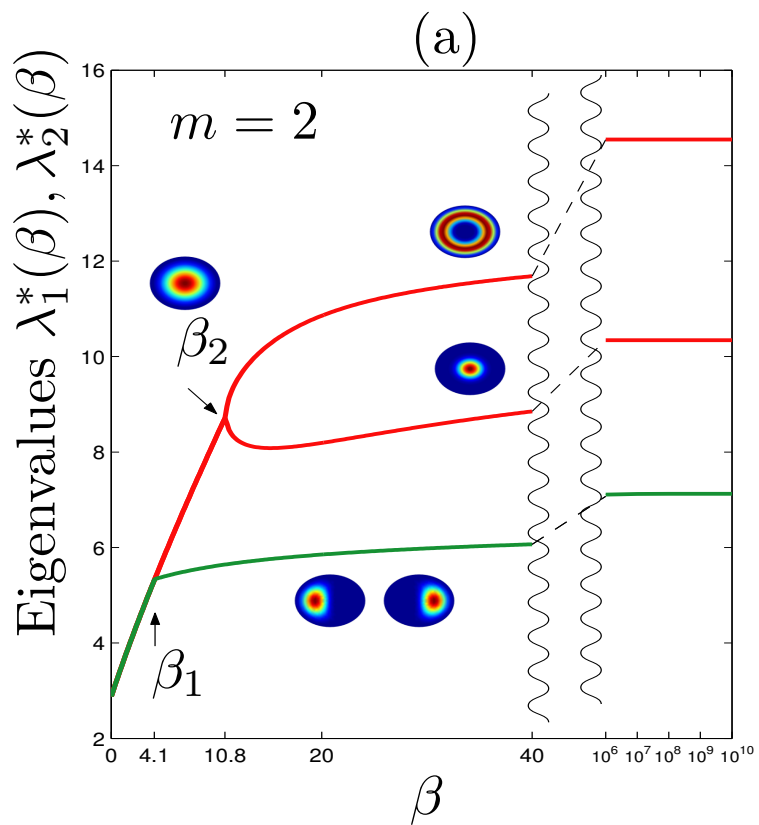
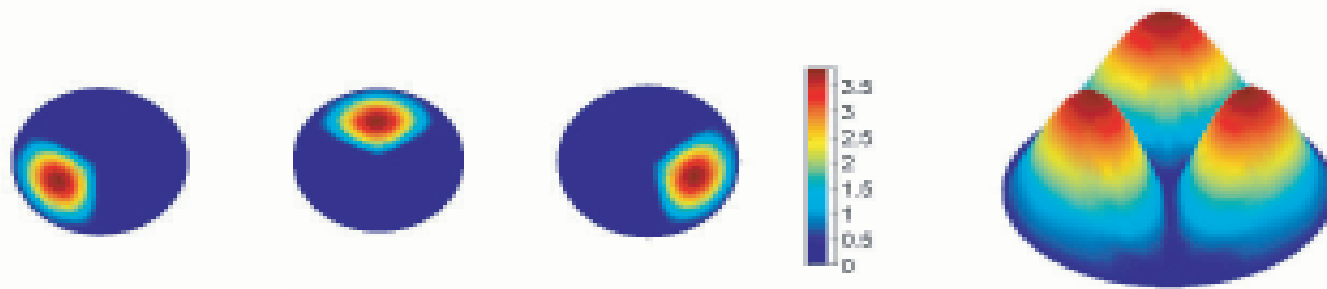
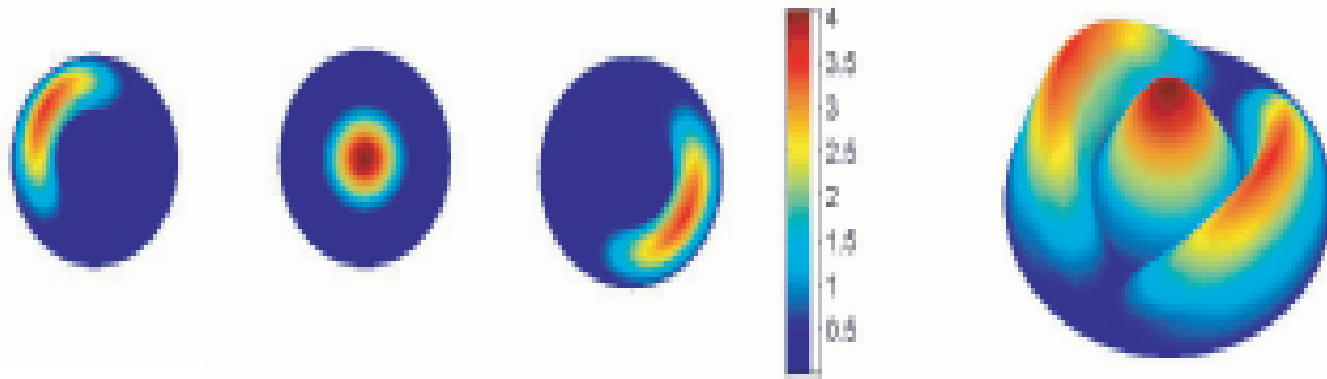


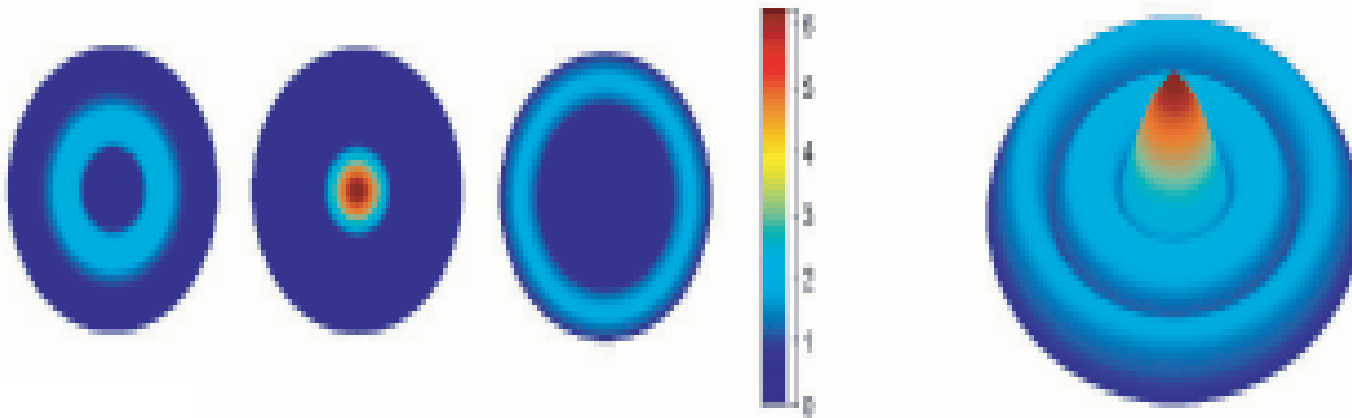
Figure 4.19: (a): Eigenvalue curves, (b): energy curves, vs  $\beta$ .



(a) green:  $\beta^* = 1000$ ,  $\lambda_1^* = \lambda_2^* = \lambda_3^* = 9.57$ ,  $E(\mathbf{u}^*) = 9.52$



(b) red:  $\beta^* = 1000$ ,  $\lambda_1^* = \lambda_3^* = 18.36$ ,  $\lambda_2^* = 20.85$ ,  $E(\mathbf{u}^*) = 19.09$



(c) blue:  $\beta^* = 1000$ ,  $\lambda_1^* = 20.84$ ,  $\lambda_2^* = 24.84$ ,  $\lambda_3^* = 32.14$ ,  
 $E(\mathbf{u}^*) = 25.85$

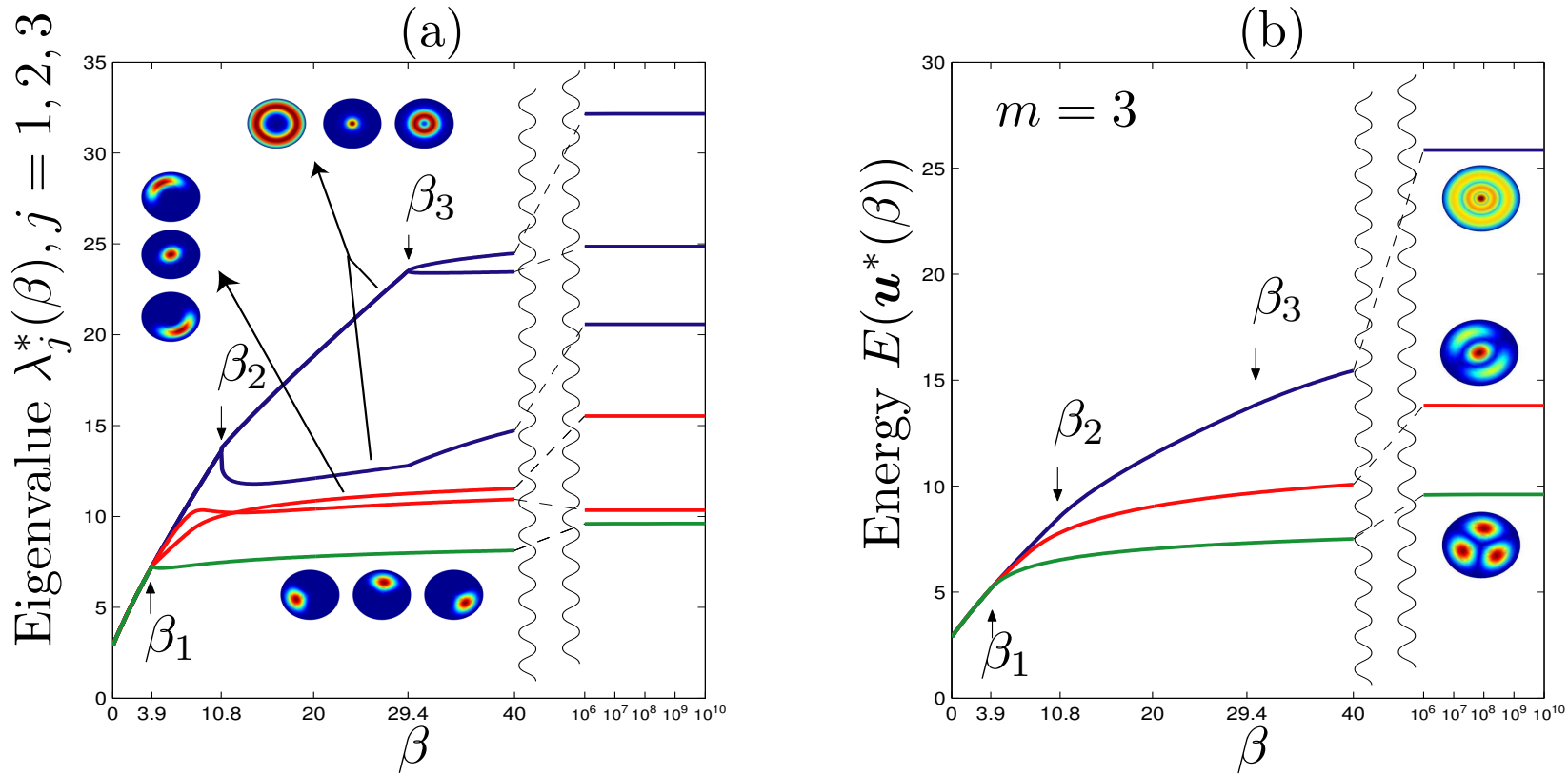
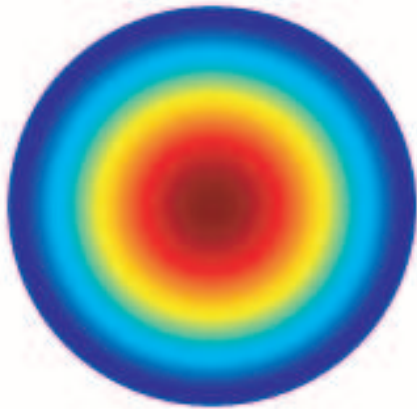


Figure 4.20: (a): Eigenvalue curves, (b): energy curves, vs  $\beta$ .

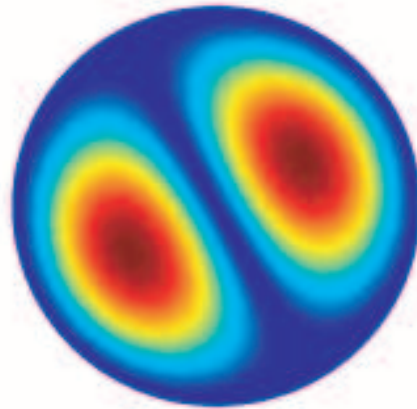
## Verticillate Structures

- How to distribute in multi-component BEC when the scattering length is sufficiently large?
- All positive bound state solutions may repel each other and form finitely segregated nodal domains when scattering length approaches to infinity. (C.S. Lin and T.C. Lin, 2003)
- Verticillate: [Botany] leaf, arranged in verticils.

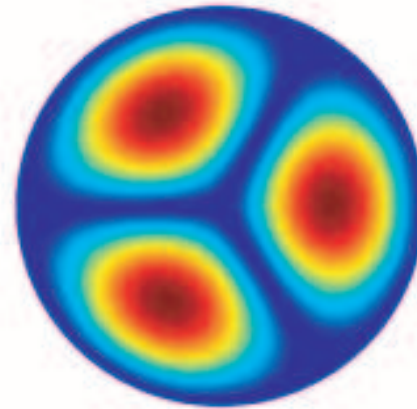




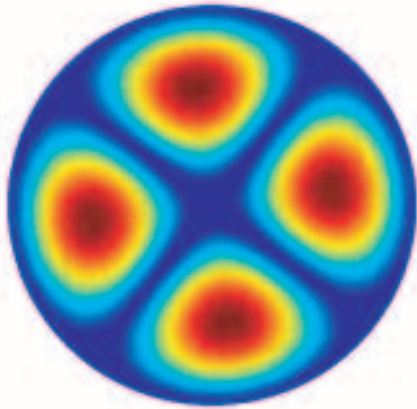
$m=1, E=2.8877$



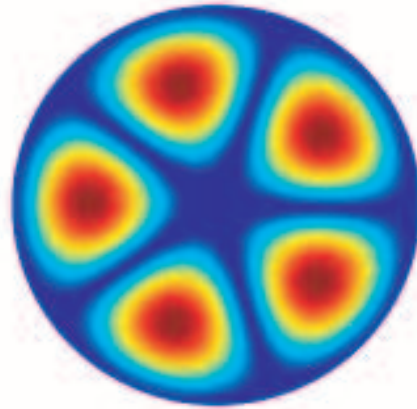
$m=2, E=7.1796$



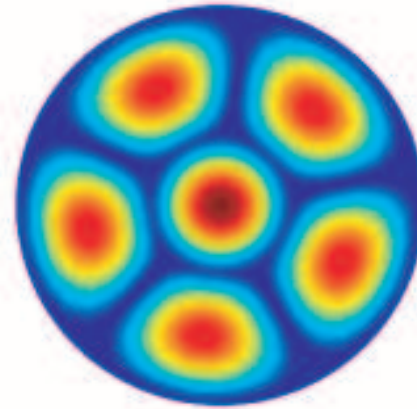
$m=3, E=9.8067$



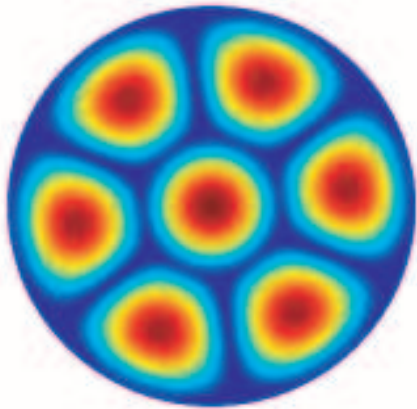
$m=4, E=12.8001$



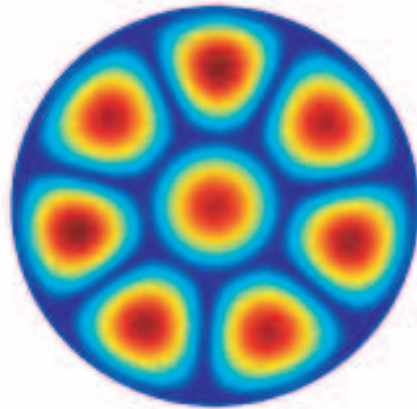
$m=5, E=16.2239$



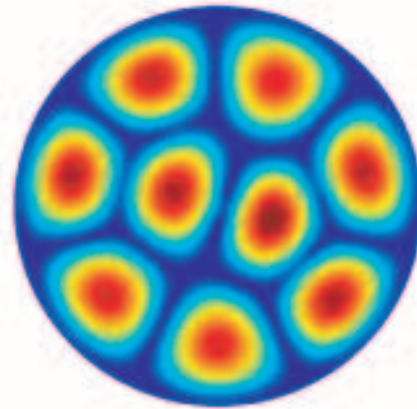
$m=6, E=19.0031$



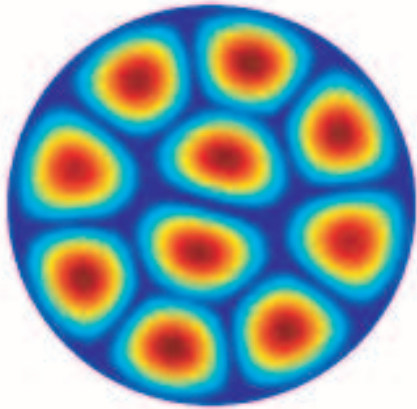
$m = 7, E = 20.4094$



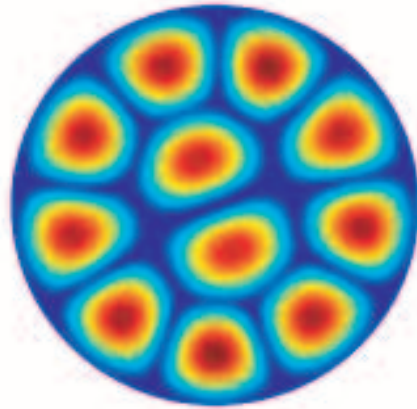
$m = 8, E = 23.2431$



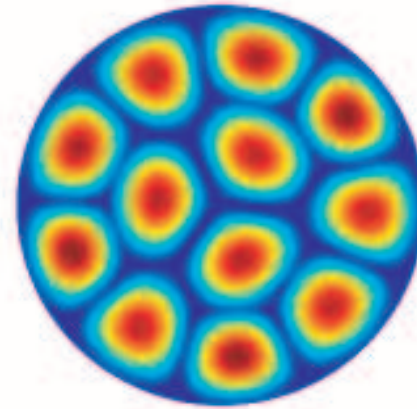
$m = 9, E = 26.0214$



$m = 10, E = 28.128$

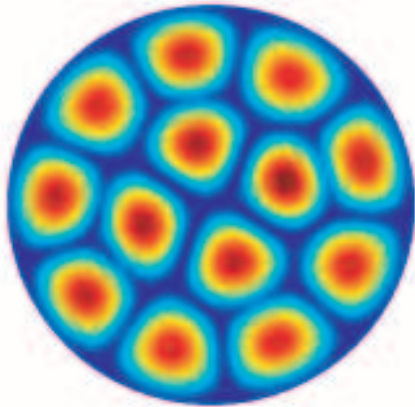


$m = 11, E = 31.0852$

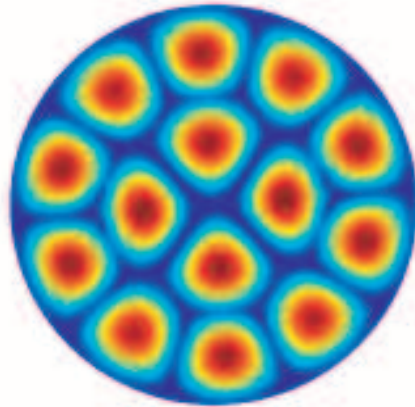


$m = 12, E = 34.2095$

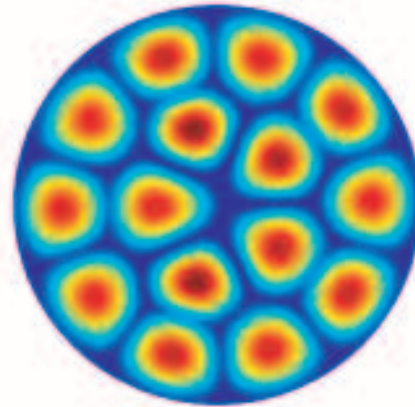




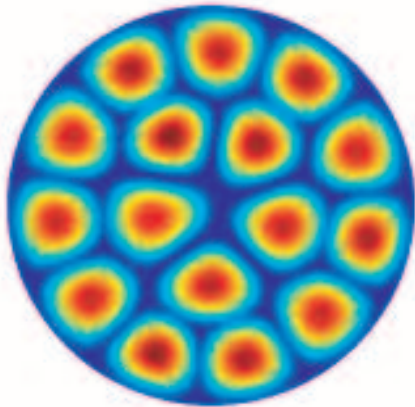
$n = 13, E = 37.0091$



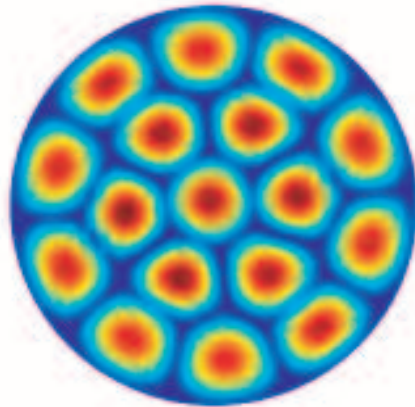
$n = 14, E = 39.3709$



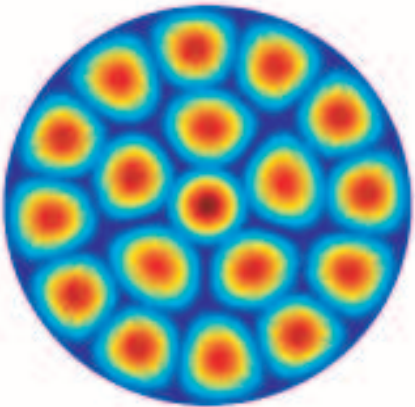
$n = 15, E = 42.1987$



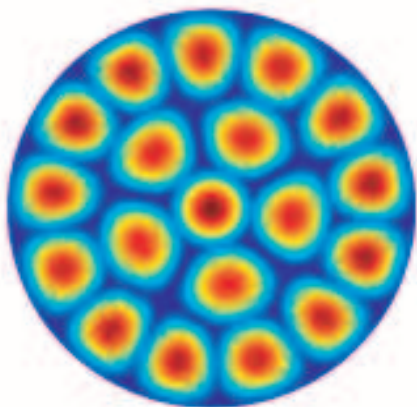
$n = 16, E = 46.0042$



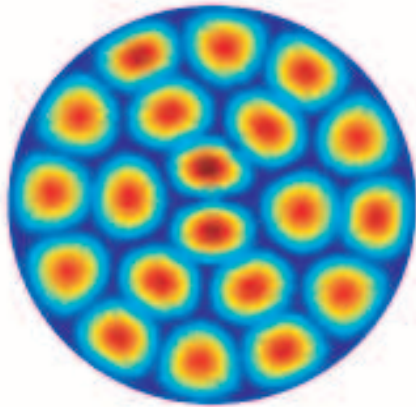
$n = 17, E = 47.0058$



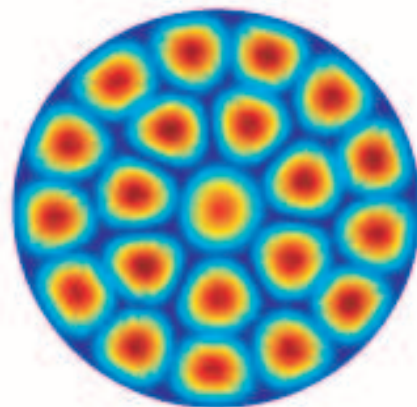
$n = 18, E = 48.0001$



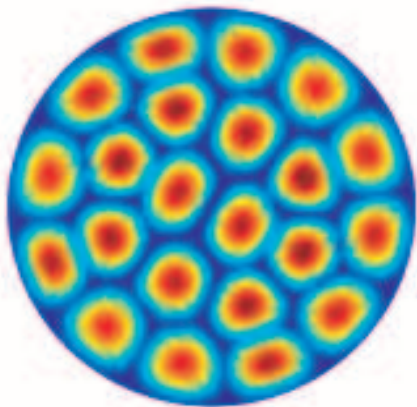
$n = 16, E = 97.2810$



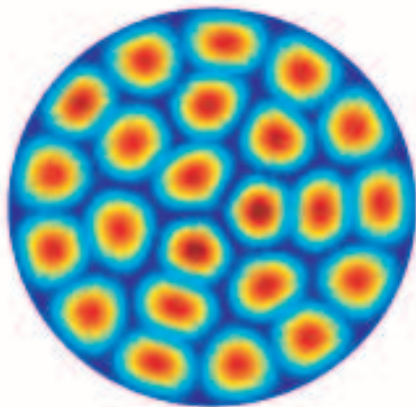
$n = 20, E = 94.4707$



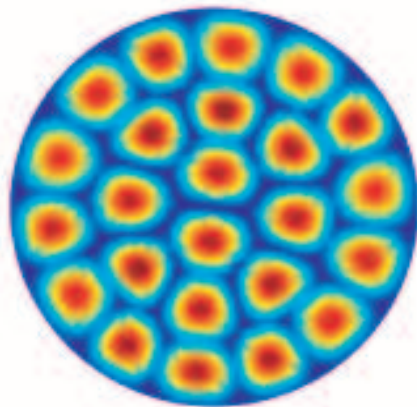
$n = 24, E = 92.9918$



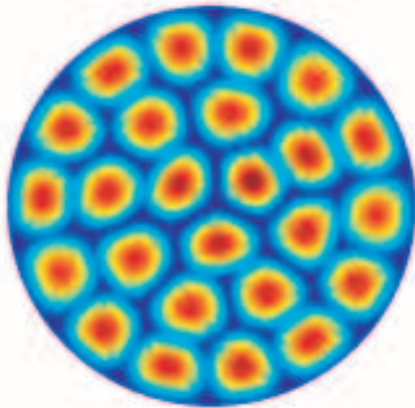
$n = 28, E = 90.9670$



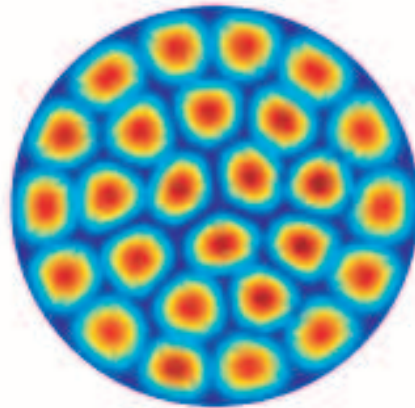
$n = 32, E = 89.6132$



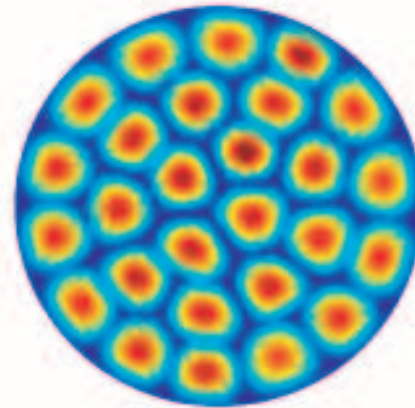
$n = 36, E = 88.8389$



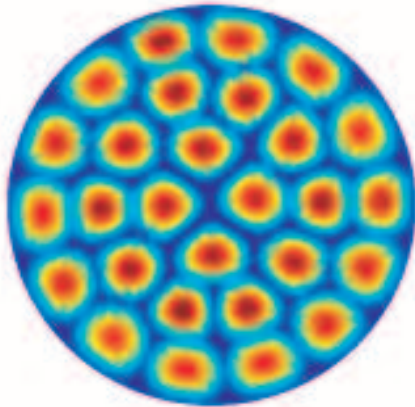
$n = 25, E = 66.6673$



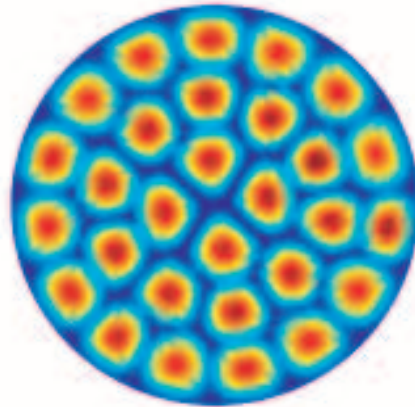
$n = 26, E = 68.5401$



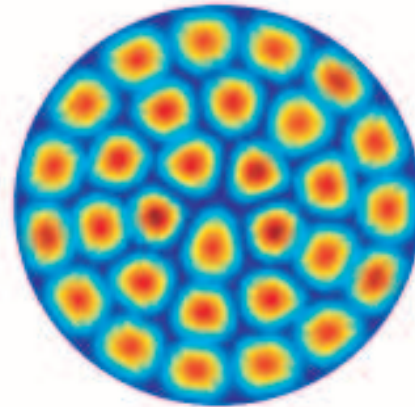
$n = 27, E = 71.0000$



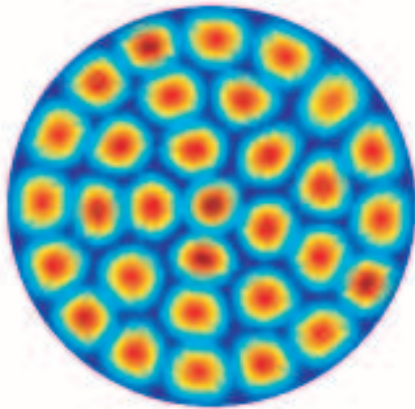
$n = 28, E = 73.7594$



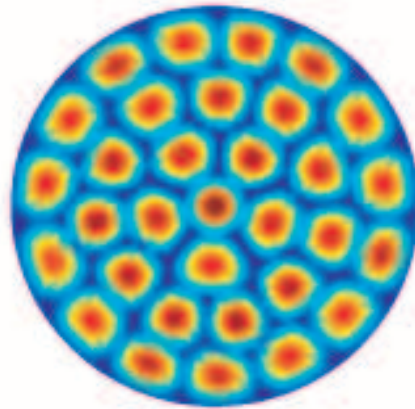
$n = 29, E = 76.6070$



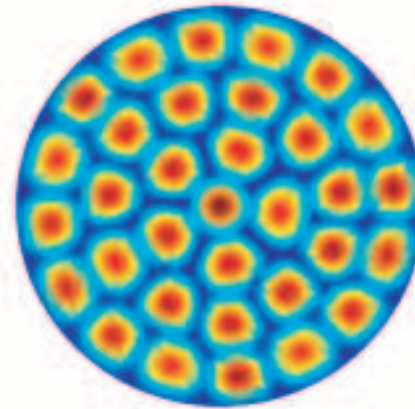
$n = 30, E = 79.0004$



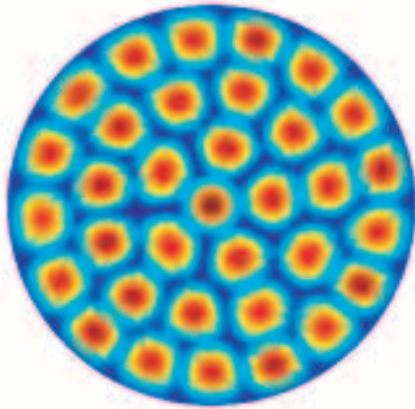
$m = 31, E = 79.0010$



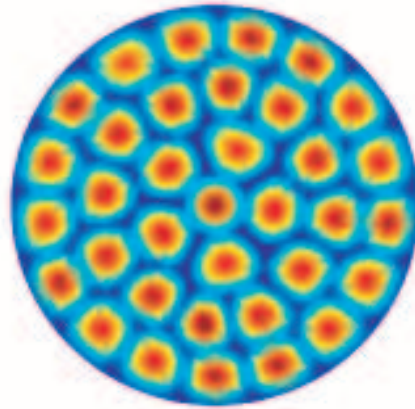
$m = 32, E = 81.8564$



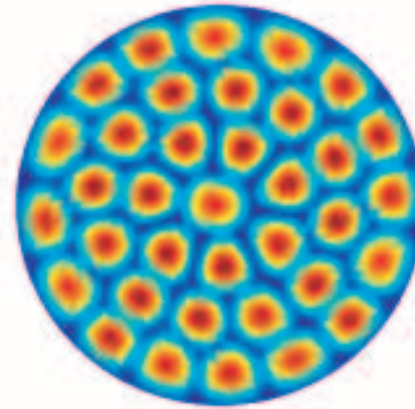
$m = 33, E = 83.8817$



$m = 34, E = 85.0214$



$m = 35, E = 87.0615$



$m = 36, E = 91.3400$

We observe that verticillate or multiple verticillate structure

(i)  $(n_1, \dots, n_\gamma)$  depends on  $m$  and  $\sum_{i=1}^{\gamma} n_i = m$  ( $\beta \gg 1$ ),

(Single, Double, Triple, Quadruple verticillate, ...)

(ii)  $1 \leq n_1 \leq 5$ .

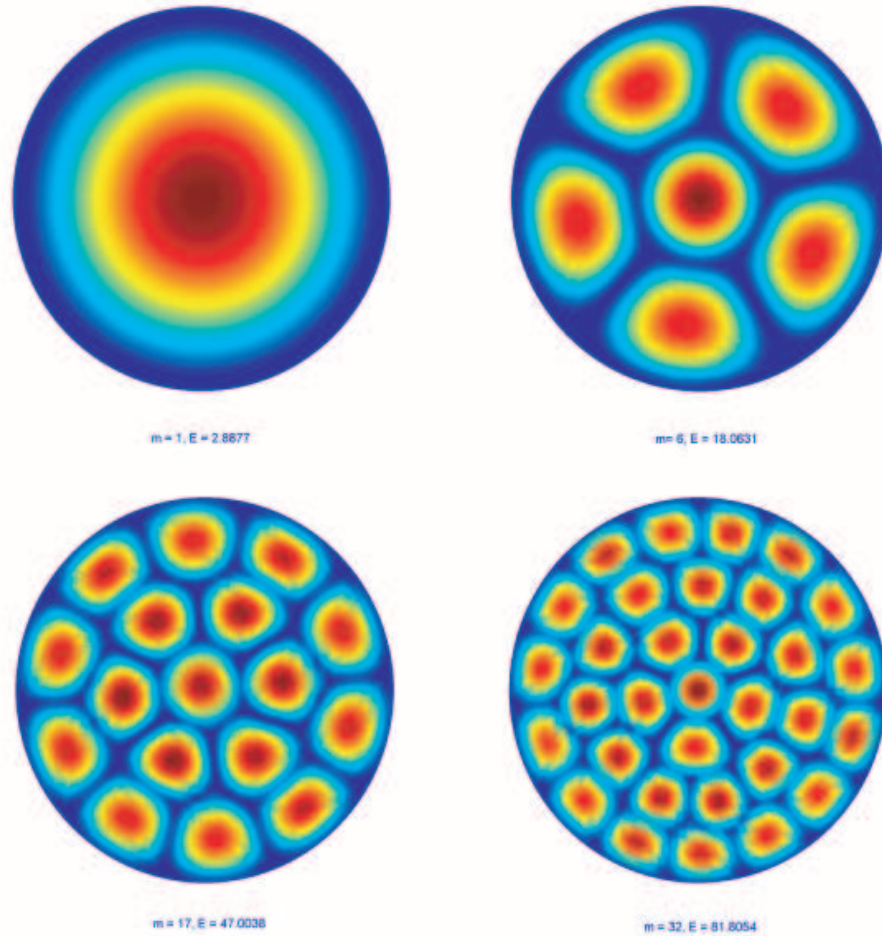
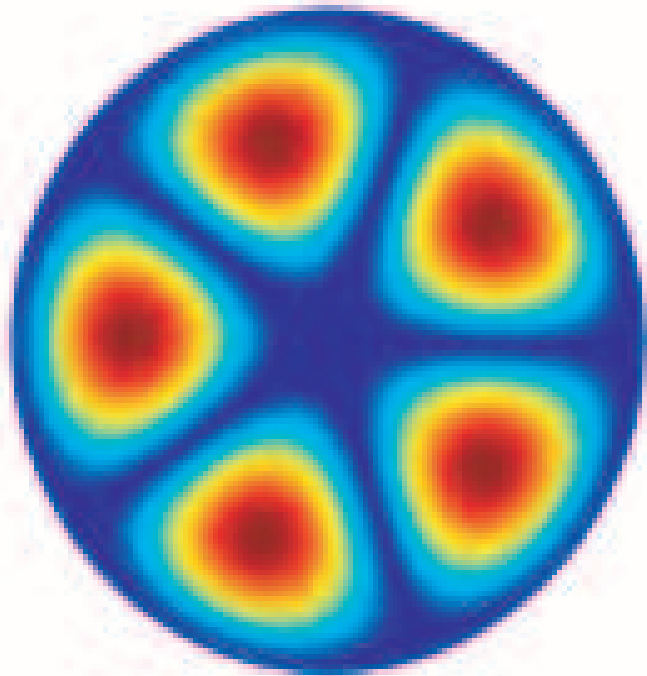


Figure 4.21: Single, Double, Triple, Quadruple verticillate:  
 (1), (1,5), (1,6,10), (1,5,11,15).

(a)



(b)

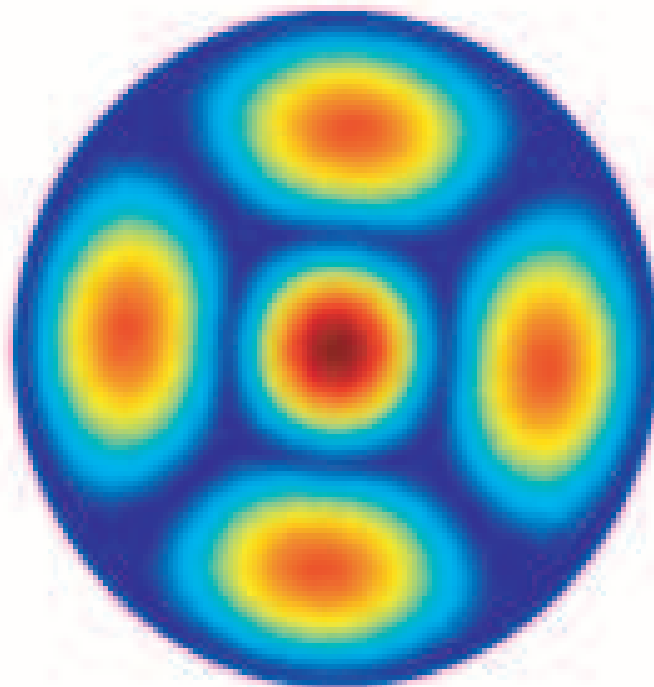


Figure 4.22:  $m = 5$ : (a) Ground state solutions, (b) bound state solutions.

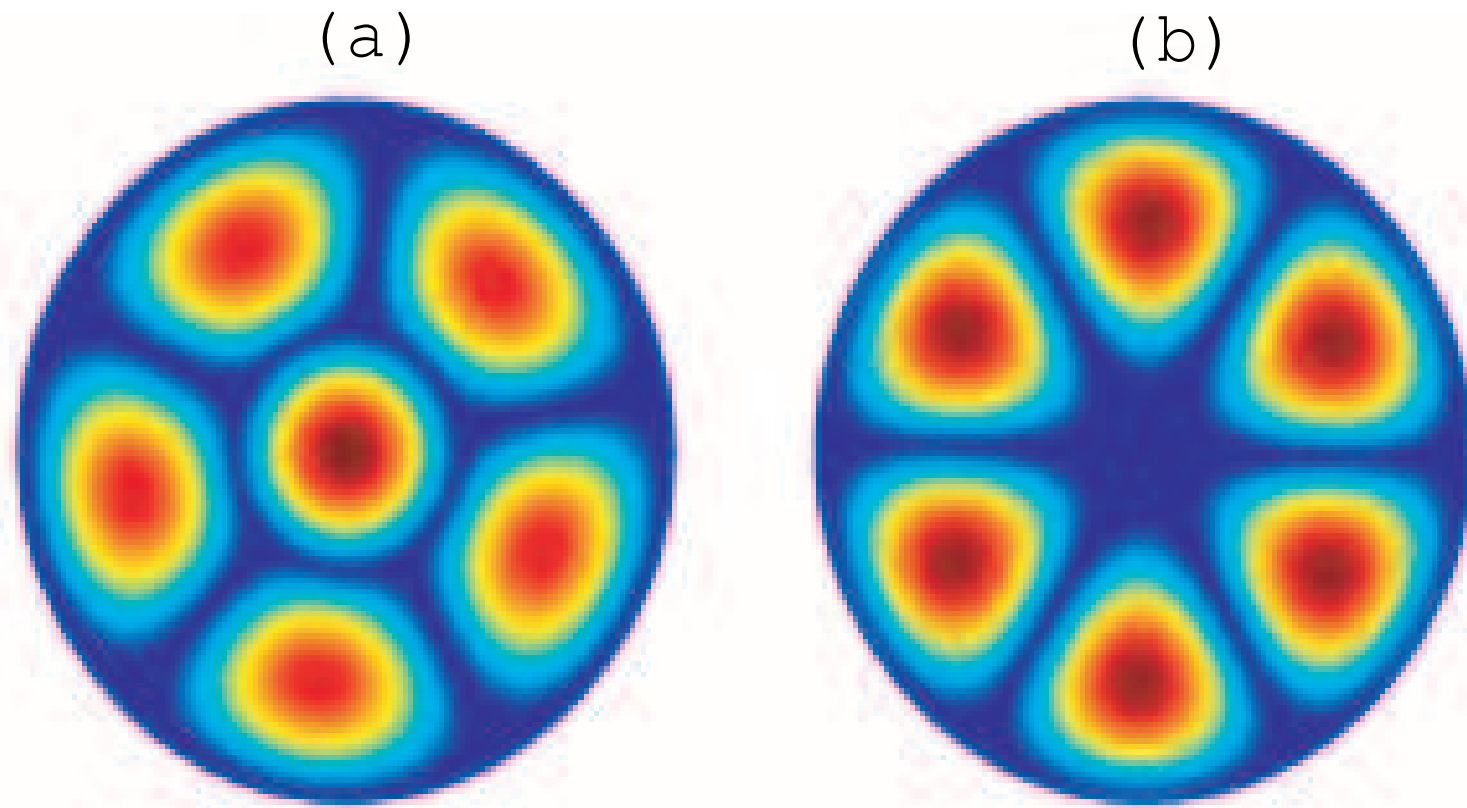


Figure 4.23:  $m = 6$ : (a) Ground state solutions, (b) bound state solutions.



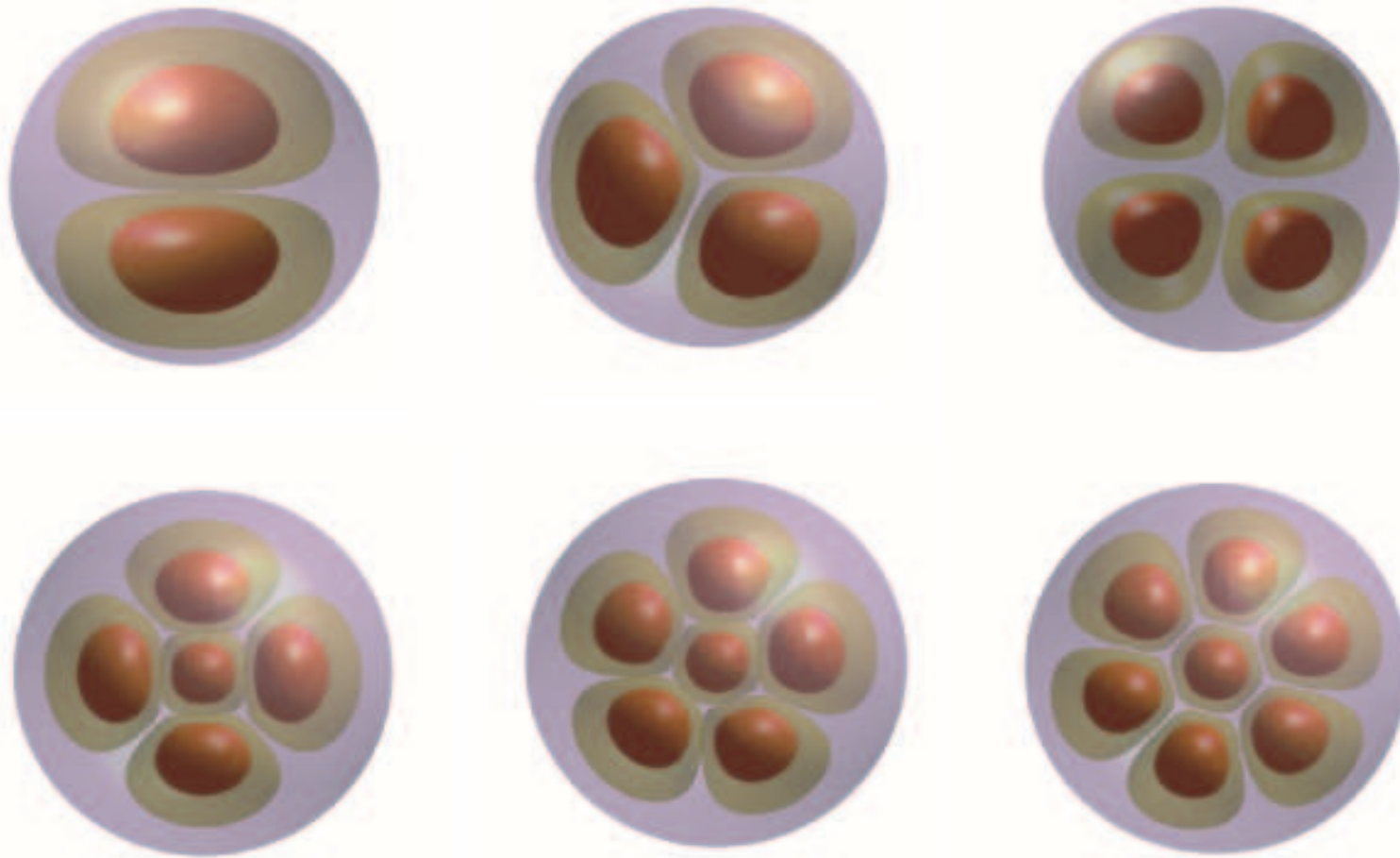


Figure 4.24: In three-dimensional domain from  $m = 2$  to  $m = 7$ .

# Continuation BSOR-Lanczos-Galerkin (BSOR-LG) Method

Nonlinear algebraic eigenvalue problems (NAEP):

$$\mathbf{A}\mathbf{u}_j + \mathbf{V}_j \circ \mathbf{u}_j + \alpha_j \mathbf{u}_j^{(2)} \circ \mathbf{u}_j + \sum_{k \neq j, k=1}^m \beta_{kj} \mathbf{u}_k^{(2)} \circ \mathbf{u}_j = \lambda_j \mathbf{u}_j,$$
$$\mathbf{u}_j^\top \mathbf{u}_j = 1, \quad j = 1, \dots, m,$$

where  $\mathbf{A} \in \mathbb{R}^{N \times N}$ ,  $\mathbf{u}_j \in \mathbb{R}^N$  for  $j = 1, \dots, m$ .

Assume that

$$\beta_{kj} = \beta_{jk} = \beta > 0, \quad k \neq j, \quad k, j = 1, \dots, m,$$

as a parameter.

Let

$$\mathbf{x} = (\mathbf{u}_1^\top, \lambda_1, \dots, \mathbf{u}_m^\top, \lambda_m)^\top.$$

Then the NAEP can be rewritten by

$$\mathbf{G}(\mathbf{x}, \beta) = 0,$$

where  $\mathbf{G} \equiv (\mathbf{G}_1, g_1, \dots, \mathbf{G}_m, g_m)$  is a smooth ft. with

$$\mathbf{G}_j(\mathbf{x}, \beta) = \mathbf{A}\mathbf{u}_j + \mathbf{V}_j \circ \mathbf{u}_j + \alpha_j \mathbf{u}_j^{(2)} \circ \mathbf{u}_j + \beta \sum_{k \neq j}^m \mathbf{u}_k^{(2)} \circ \mathbf{u}_j - \lambda_j \mathbf{u}_j,$$

$$g_j(\mathbf{x}, \beta) = \frac{1}{2}(\mathbf{u}_j^\top \mathbf{u}_j - 1),$$

for  $j = 1, \dots, m$ .

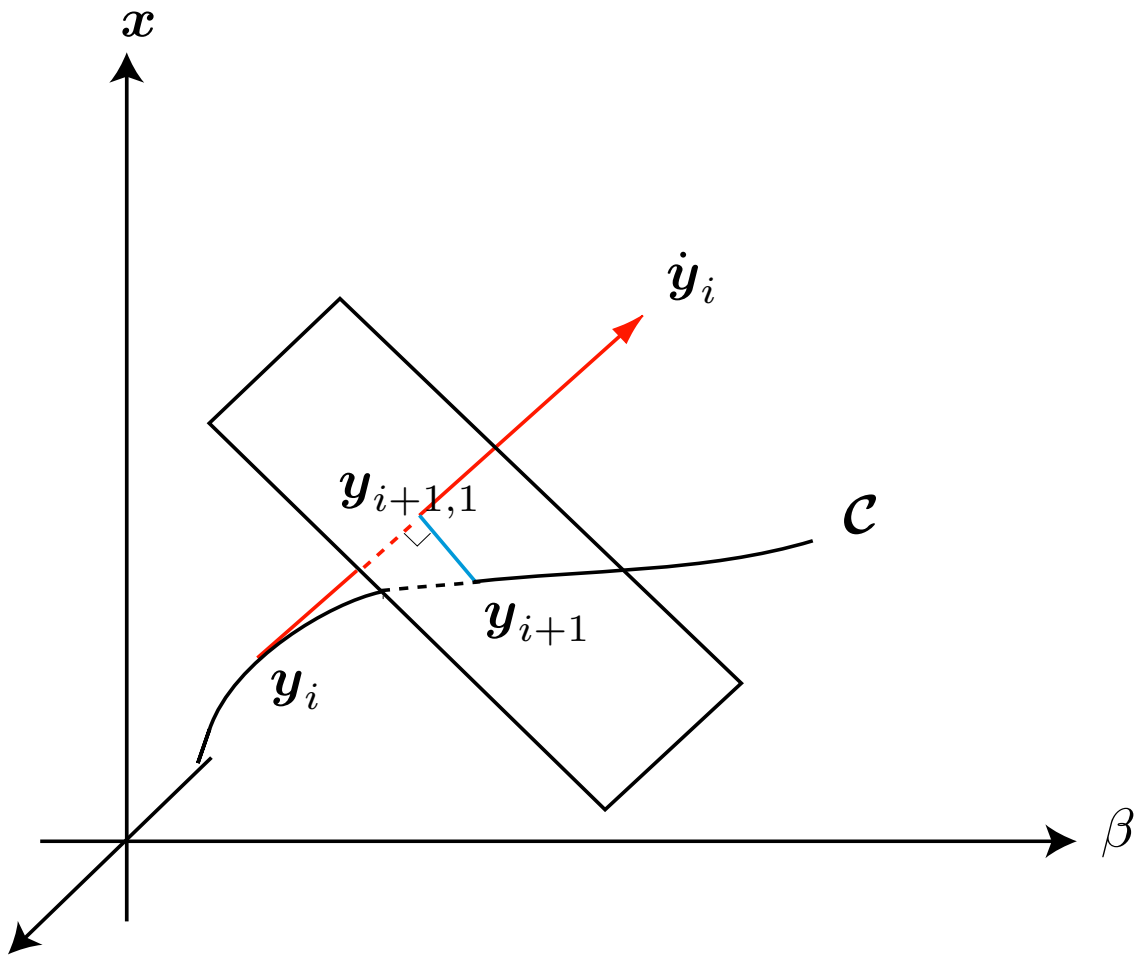
We denote the Jacobian of  $\mathbf{G}$  by  $\mathcal{D}\mathbf{G} = [\mathbf{G}_{\mathbf{x}}, \mathbf{G}_{\beta}] \in \mathbb{R}^{M \times (M+1)}$  with  $M = (N + 1)m$ , and the solution curve  $\mathcal{C}$  of  $\mathbf{G}(\mathbf{x}, \beta) = 0$  by

$$\mathcal{C} = \{\mathbf{y}(s) = (\mathbf{x}(s)^\top, \beta(s)^\top)^\top \mid \mathbf{G}(\mathbf{y}(s)) = 0, s \in \mathbf{J} \subseteq \mathbb{R}\}.$$

Assume  $s$  is a parametrization via arc length is available on  $\mathcal{C}$ . By differentiating with  $s$  we have

$$\mathcal{D}\mathbf{G}(\mathbf{y}(s))\dot{\mathbf{y}}(s) = 0,$$

where  $\dot{\mathbf{y}}(s) = (\dot{\mathbf{x}}(s)^\top, \dot{\beta}(s)^\top)^\top$  is a tangent vector to  $\mathcal{C}$  at  $\mathbf{y}(s)$ .



## Prediction

Let  $\mathbf{y}_i = (\mathbf{x}_i^\top, \beta_i)^\top \in \mathbb{R}^{M+1}$  be an approx. point for  $\mathcal{C}$ . Suppose  $\mathbf{y}_{i+1,1} = \mathbf{y}_i + h_i \dot{\mathbf{y}}_i$  is used to predict a new  $\mathbf{y}_{i+1,1}$ , where  $\dot{\mathbf{y}}_i$  is the unit tangent vector by solving

$$\left[ \begin{array}{c|c} \mathbf{G}_x(\mathbf{y}_i) & \mathbf{G}_\beta(\mathbf{y}_i) \\ \hline \mathbf{c}_i^\top & \end{array} \right] \dot{\mathbf{y}}_i = \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} \quad (4.1)$$

with some constant vector  $\mathbf{c}_i \in \mathbb{R}^{M+1}$ .

## Correction

$$\begin{cases} \mathbf{G}(\mathbf{y}) = 0 \\ \dot{\mathbf{y}}_i^\top \mathbf{y} = \dot{\mathbf{y}}_i^\top \mathbf{y}_{i+1,1} \end{cases}$$

Newton's method is chosen as a corrector,

$$\left[ \begin{array}{c|c} \mathbf{G}_x(\mathbf{y}_{i+1,l}) & \mathbf{G}_\beta(\mathbf{y}_{i+1,l}) \\ \hline \dot{\mathbf{y}}_i^\top & \end{array} \right] \delta_l = \begin{bmatrix} -\mathbf{G}(\mathbf{y}_{i+1,l}) \\ -\rho_l \end{bmatrix}, \quad l = 1, 2, \dots, \quad (4.2)$$

with  $\rho_l = \dot{\mathbf{y}}_i^\top (\mathbf{y}_{i+1,l} - \mathbf{y}_{i+1,1})$ , is solved by  $\mathbf{y}_{i+1,l+1} = \mathbf{y}_{i+1,l} + \delta_l$ . If  $\{\mathbf{y}_{i+1,l}\}$  converges until  $l = l_\infty$ , we accept  $\mathbf{y}_{i+1} = \mathbf{y}_{i+1,l_\infty}$  as an approx to  $\mathcal{C}$ .

- **BSOR-Lanczos-Galerkin algorithm**

Linear systems (4.1) and (4.2) can be rewritten in

$$\begin{bmatrix} \mathbf{B} & \mathbf{f} \\ \mathbf{g}^\top & \gamma \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \beta \end{bmatrix} = \begin{bmatrix} \mathbf{p} \\ \rho \end{bmatrix}, \quad (4.3)$$

where  $\mathbf{B} \in \mathbb{R}^{M \times M}$ ,  $\mathbf{f}, \mathbf{g}$  and  $\mathbf{p} \in \mathbb{R}^M$ , and solved by the block elimination algorithm.



### Algorithm 1: Block Elimination

- (i) Solve  $B\xi = \mathbf{f}$  and  $B\eta = \mathbf{p}$ ,
- (ii) Compute  $\beta = (\rho - \mathbf{g}^\top \eta) / (\gamma - \mathbf{g}^\top \xi)$ ,
- (iii) Compute  $\mathbf{x} = \eta - \beta\xi$ .

The main step in (4.1) or in (4.2) is to solve a linear system of the form  $\mathbf{G}_x(\mathbf{y})\xi = \mathbf{f}$ , that can be formulated in

$$B\xi \equiv \begin{bmatrix} B_{11} & B_{12} & \cdots & B_{1m} \\ B_{21} & B_{22} & \cdots & B_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ B_{m1} & B_{m2} & \cdots & B_{mm} \end{bmatrix} \begin{bmatrix} \xi_1 \\ \xi_2 \\ \vdots \\ \xi_m \end{bmatrix} = \begin{bmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \vdots \\ \mathbf{f}_m \end{bmatrix},$$

where

$$\begin{aligned} \mathbf{B}_{jj} &= \mathcal{D}_{(\mathbf{u}_j, \lambda_j)} \begin{bmatrix} \mathbf{G}_j(\mathbf{y}) \\ \mathbf{g}_j(\mathbf{y}) \end{bmatrix} \\ &= \left[ \begin{array}{c|c} \mathbf{A} + \llbracket \mathbf{V}_j + 3\alpha_j \mathbf{u}_j^{(2)} + \beta \sum_{k \neq j} \mathbf{u}_k^{(2)} \rrbracket - \lambda_j I & \mathbf{u}_j \\ \hline \mathbf{u}_j^\top & 0 \end{array} \right] \equiv \left[ \begin{array}{c|c} \mathbf{A}_j & \mathbf{u}_j \\ \hline \mathbf{u}_j^\top & 0 \end{array} \right] \end{aligned}$$

and

$$\mathbf{B}_{kj} = \mathcal{D}_{(\mathbf{u}_k, \lambda_k)} \begin{bmatrix} \mathbf{G}_j(\mathbf{y}) \\ \mathbf{g}_j(\mathbf{y}) \end{bmatrix} = \left[ \begin{array}{c|c} 2\beta \llbracket \mathbf{u}_k \circ \mathbf{u}_j \rrbracket & 0 \\ \hline 0 & 0 \end{array} \right], \quad k \neq j,$$

$$k, j = 1, \dots, m.$$

## Algorithm 2: Block SOR (BSOR)

- (i) Choose a parameter  $\omega \in (0, 2)$  and initials  $\{\xi_j^{(0)}\}_{j=1}^m$ ,  $i = 0$ ;  
(ii) Repeat  $i$  : until convergence,  
For  $j = 1, \dots, m$ ,

solve the linear system for  $\xi_j^{(i+1)}$

$$B_{jj}\xi_j^{(i+1)} = \omega \left[ \mathbf{f}_j - \sum_{k>j} B_{jk}\xi_k^{(i)} - \sum_{k<j} B_{jk}\xi_k^{(i+1)} \right] + (1 - \omega)B_{jj}\xi_j^{(i)}, \quad (4.4)$$

end for  $j$ ;

- (iii) If converges, then  $\xi_j \leftarrow \xi_j^{(i+1)}$  ( $j = 1, \dots, m$ ), stop;  
else  $i \leftarrow i + 1$ , Goto Repeat (ii).

In Algorithm 2.1 the linear system in (4.4) is

$$\left[ \begin{array}{c|c} \mathbf{A}_j & \mathbf{u}_j \\ \hline \mathbf{u}_j^\top & 0 \end{array} \right] \begin{bmatrix} \boldsymbol{\xi}_{j,1}^{(i)} \\ \xi_{j,2}^{(i)} \end{bmatrix} = \begin{bmatrix} \mathbf{b}^{(i)} \\ \rho^{(i)} \end{bmatrix}$$

We reduce the system to solving several linear systems of the form

$$\mathbf{A}_j \boldsymbol{\xi}_{j,1}^{(i)} = \mathbf{b}^{(i)}, \quad i = 1, \dots, r,$$

involving the same  $N \times N$  matrix  $\mathbf{A}_j$  but different right-hand sides  $\mathbf{b}^{(i)}$ .

### Algorithm 3: Lanczos-Galerkin Projection Method

(i) First pass.

Solve  $\mathbf{A}_j \boldsymbol{\xi}^{(1)} = \mathbf{b}^{(1)}$  by  $q$ -step Lanczos algorithm;

Let  $\mathbf{V}_q = [\mathbf{v}_1, \dots, \mathbf{v}_q]$  be the orthog. Lanczos basis spanning the Krylov subsp. with  $\mathbf{v}_1 = (\mathbf{b}^{(1)} - \mathbf{A}_j \boldsymbol{\xi}_0^{(1)}) / \|\mathbf{b}^{(1)} - \mathbf{A}_j \boldsymbol{\xi}_0^{(1)}\|$  and  $\mathbf{T}_q$  be the corr.  $q \times q$  tridiagonal matrix;

(ii) Second pass.

For  $i = 2, \dots, r$ ,

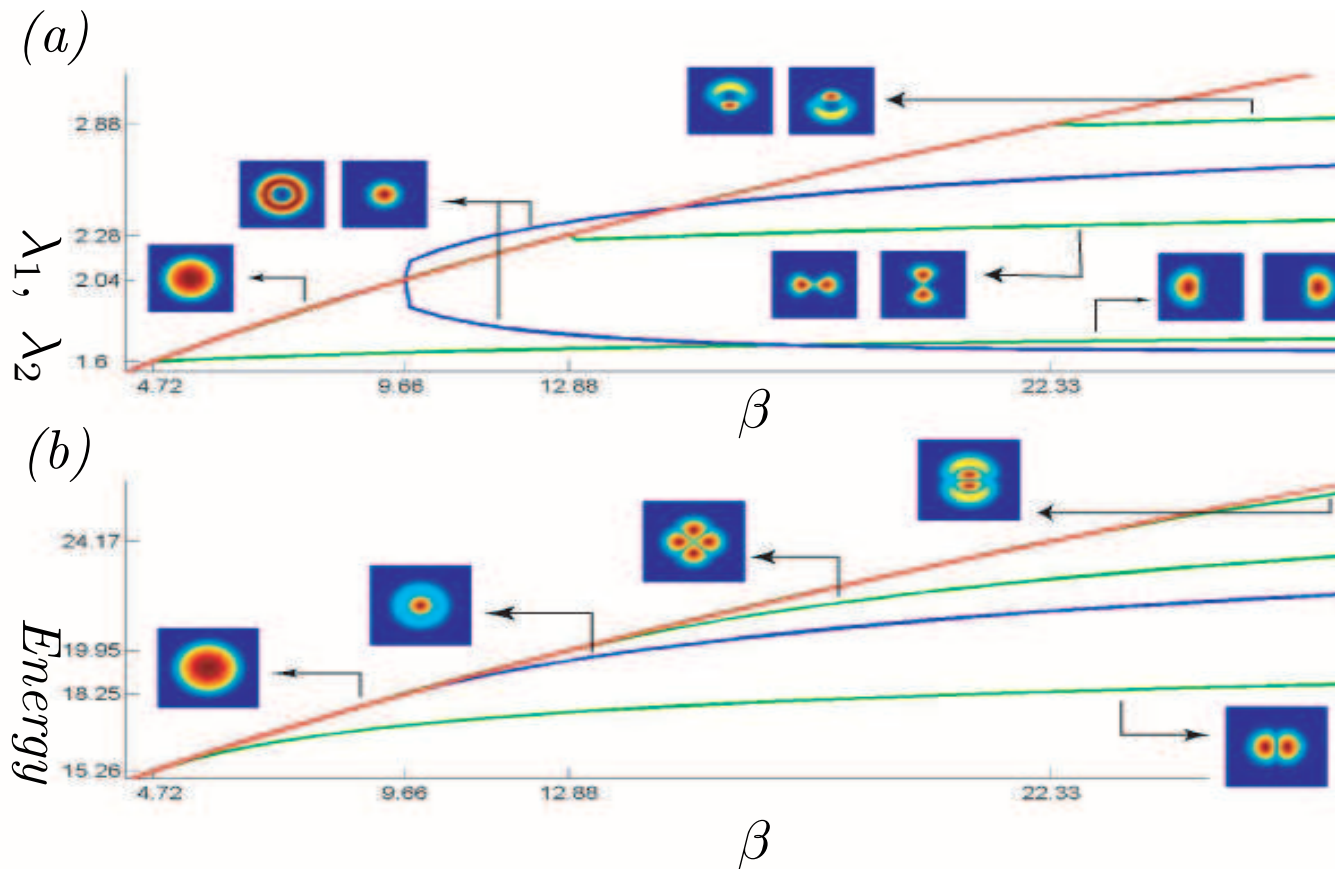
Compute  $\mathbf{r}_0^{(i)} = \mathbf{b}^{(i)} - \mathbf{A}_j \boldsymbol{\xi}_0^{(i)}$  with an initial  $\boldsymbol{\xi}_0^{(i)}$ ,

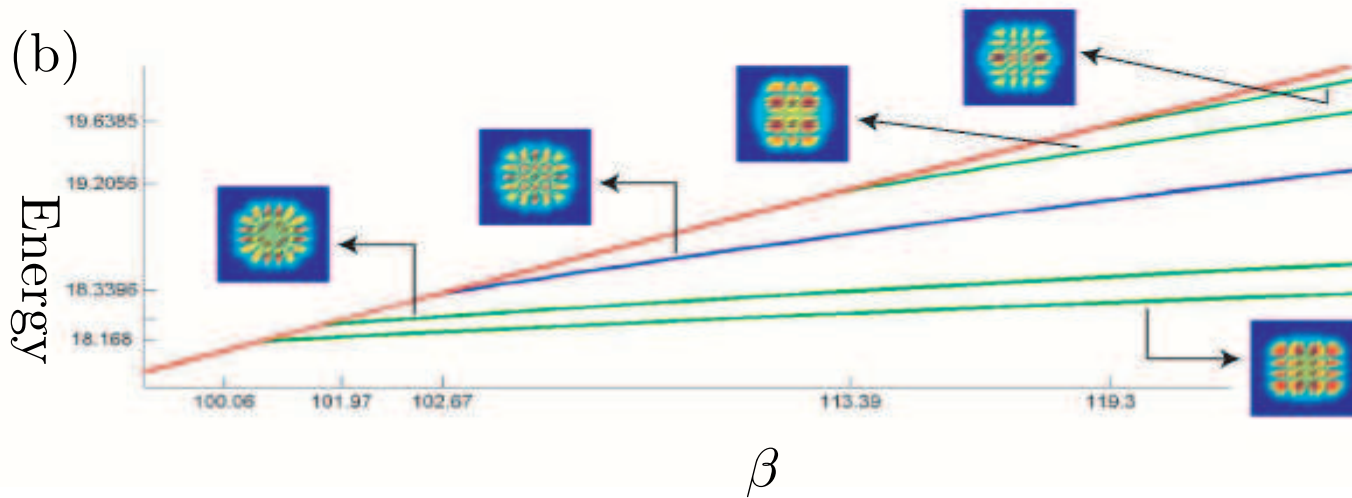
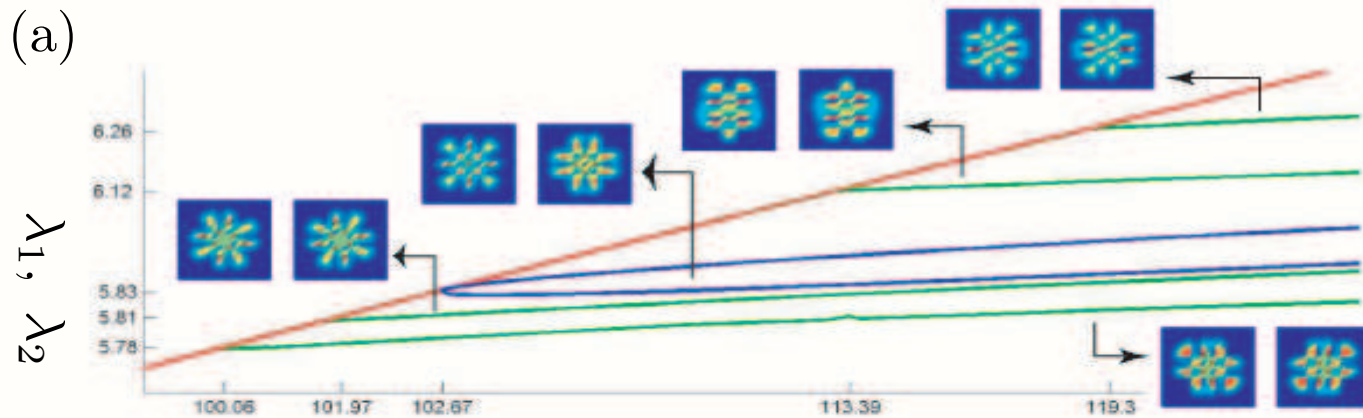
Compute  $\boldsymbol{\xi}^{(i)} = \boldsymbol{\xi}_0^{(i)} + \mathbf{V}_q \mathbf{T}_q^{-1} \mathbf{V}_q^\top \mathbf{r}_0^{(i)}$ ,

If the accuracy of  $\boldsymbol{\xi}^{(i)}$  is not satisfactory, perform a refinement (restarted) Lanczos-Galerkin process,

end for  $i$ .

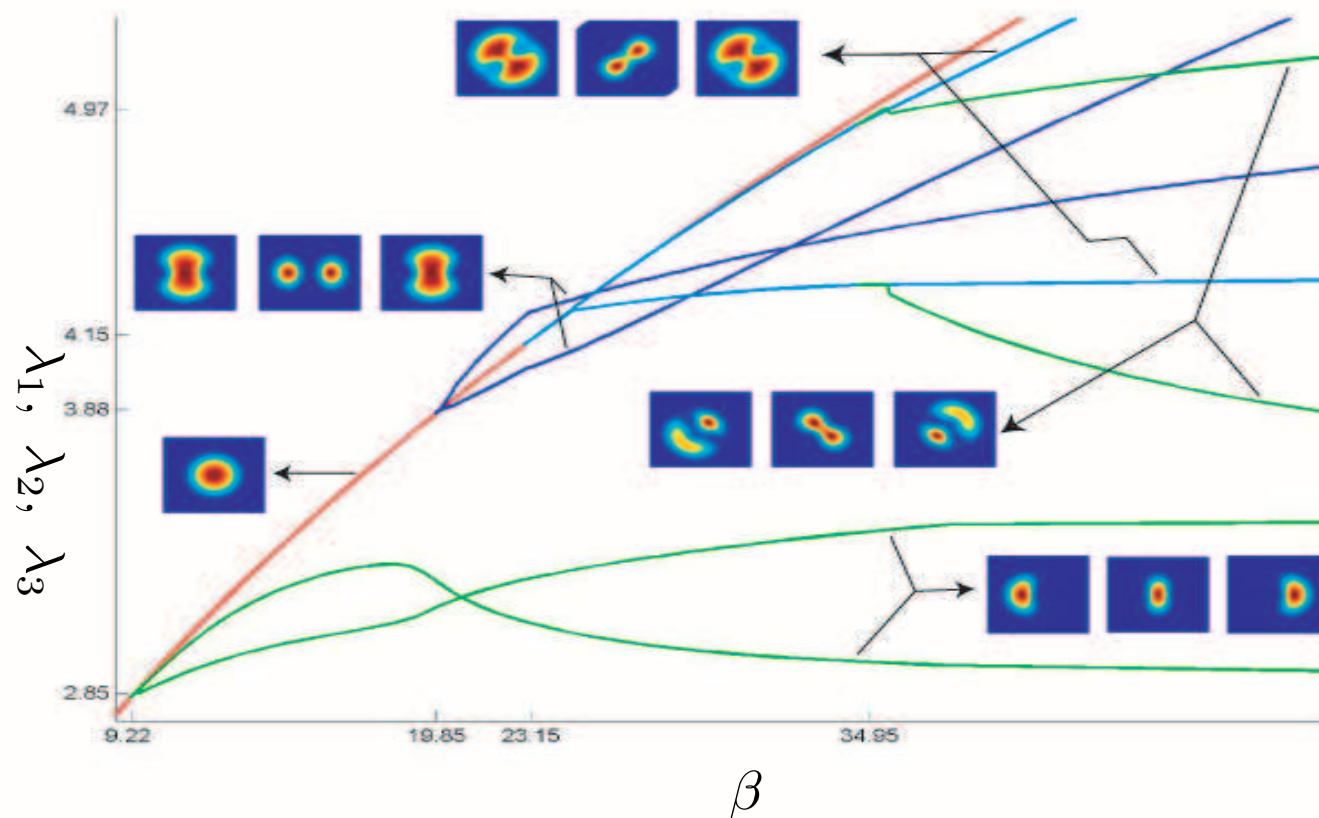
**Example 4.1** For  $m = 2$ :  $\Omega = [-5, 5] \times [-4.8, 4.8]$ ,  
 $V_1 = V_2 = x^2 + y^2$ ,  $\alpha_1 = \alpha_2 = 0.1$ ,  $\beta_{12} = \beta_{21} = \beta > 0$ .



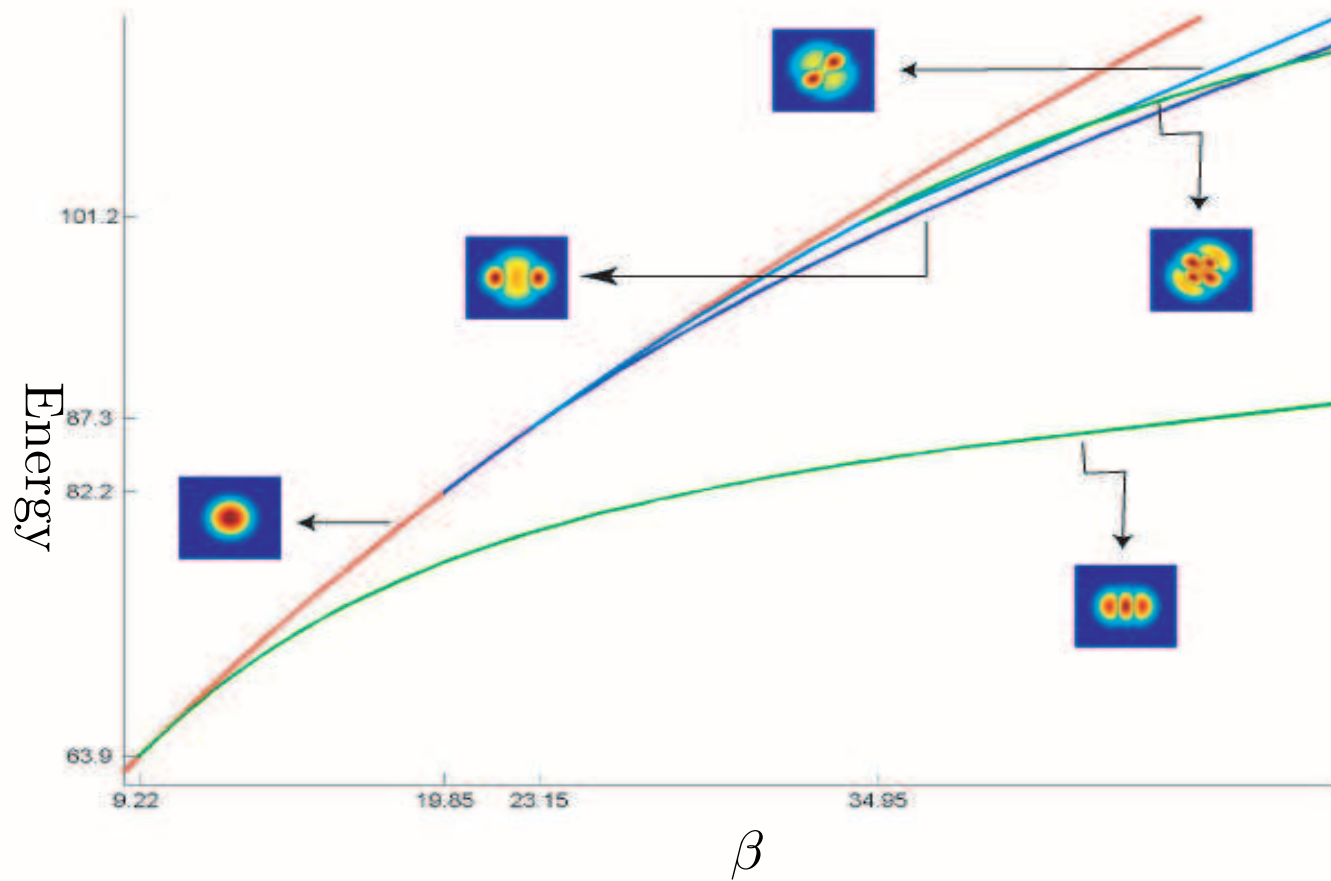


$m = 2$ . Solution curve of eigenvalues and energy versus  $\beta$ , for  $\beta \in (98, 125)$ .

**Example 4.2** For  $m = 3$ :  $\Omega = [-5, 5] \times [-4.8, 4.8]$ ,  
 $V_1 = V_2 = V_3 = x^2 + y^2$ ,  $\alpha_1 = \alpha_2 = \alpha_3 = 0.1$ ,  $\beta_{kj} = \beta$ ,  $k \neq j$ ,  
 $k, j = 1, 2, 3$ .







$m = 3$ . Solution curve of energy versus  $\beta$ , for  $\beta \in (8.7, 51)$ .