

# A Cartesian-grid integrated-RBF Galerkin technique

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**Abstract:** This paper describes a high-order Galerkin technique, which is based on indirect/integrated radial-basis-function networks (IRBFNs) and Cartesian grids, for the discretisation of elliptic problems in two dimensions. The field variable is approximated by high-order IRBFNs that can work on uniform grids without suffering from Runge's phenomenon. Unlike conventional Galerkin techniques, derivative boundary values are incorporated into the approximations and their imposition is conducted in an exact manner. The Galerkin formulation is then applied to force IRBFNs to satisfy the governing equation. The present technique is verified numerically through the solution natural convection in a square slot - a benchmark problem in CFD. Highly accurate solutions are obtained using relatively coarse grids, which show the effectiveness of using RBFs as trial functions in the Galerkin formulation.

**Keywords:** Integrated RBFNs, Galerkin approach, Cartesian grids, elliptic problems.

## 1 Introduction

Radial-basis-function networks (RBFNs) have been shown to be a powerful numerical tool for the solution of partial-differential equations (PDEs). The first report on this subject was made by Kansa (1990). For Kansa's method, a function is first represented by an RBFN which is then differentiated to obtain approximate expressions for its derivative functions. On the other hand, to avoid the reduction in convergence rate caused by differentiation, Mai-Duy and Tran-Cong (2001) proposed an indirect/integrated RBFN (IRBFN) approach in which the highest-order derivatives in the PDE are first decomposed into RBFs, and their lower-order derivatives and the function itself are then obtained through integration. Previous studies (e.g. [Mai-Duy and Tran-Cong (2001)]) showed that IRBFN collocation methods yield better accuracy than differentiated RBFN (DRBFN) ones for both the representation of functions and the solution of PDEs.

Since the global RBF interpolation matrix is fully populated and its condition number grows rapidly with respect to the increase of RBF centres and/or widths [Schaback (1995)], several RBF techniques based on local approximations have been proposed. In the context of IRBFNs, collocation schemes, based on one-dimensional (1D) IRBFNs and Cartesian grids, for the solution of 2D

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elliptic PDEs were reported in, e.g. [Mai-Duy and Tran-Cong (2007)]. The RBF approximations at a grid node involve only points that lie on the grid lines intersected at that point rather than the whole set of nodes. As a result, the construction process is conducted for a series of small matrices rather than for a large single matrix (“local” approximation).

Along with the point-collocation approach, a Galerkin-type approach, based on 1D-IRBFNs, has been developed in [Mai-Duy and Tran-Cong (2009)]. In this approach, the boundary conditions are satisfied in a local sense using the point collocation formulation, and the solution to the problem is satisfied in a global sense using the Galerkin formulation. The use of integration to construct the approximations generates some additional coefficients (i.e. the constants of integration) that can be exploited for the effective implementation of Neumann and multiple boundary conditions. The resultant system of algebraic equations is often symmetric and has a relatively-low condition number, which facilitate the employment of much larger numbers of nodes. Numerical results showed that this technique yields accurate results, high rates of convergence, and especially similar levels of accuracy for both types of problems (i.e. Dirichlet only and Dirichlet-Neumann boundary conditions).

In this paper, the high-order Galerkin technique, which is based on 1D-IRBFNs and Cartesian grids, is applied to simulate natural convection in a square slot. It will be shown that convergent solutions are achieved for very high values of the Rayleigh number (i.e. up to  $10^8$ ). Numerical results obtained are compared with those by other techniques available in the literature.

The remainder of this paper is organised as follows. Section 2 presents the proposed integrated-RBF Galerkin method, including the Galerkin formulation, 1D-IRBFN representations of the field variables and imposition of boundary conditions. A CFD benchmark test problem is simulated in section 3. Section 4 concludes the paper.

## 2 The integrated-RBF Galerkin technique

### 2.1 Galerkin formulation

Let

$$u(\mathbf{x}) = \sum_{i=1}^N \alpha_i \phi_i(\mathbf{x}) \approx \bar{u}(\mathbf{x}), \quad (1)$$

be an approximate solution to a differential equation of the form

$$L(\bar{u}) = 0 \quad \mathbf{x} \in \Omega, \quad (2)$$

with its boundary conditions of the form

$$B(\bar{u}) = 0 \quad \mathbf{x} \in \Gamma. \quad (3)$$

In Eq. 1, Eq. 2 and Eq. 3,  $u$  is the field/dependent variable, the overbar denotes the exact solution,  $L$  the prescribed differential operator,  $\Omega$  the problem domain,  $\Gamma$  the boundary of the domain  $\Omega$ ,  $\{\alpha_i\}_{i=1}^N$  the set of unknown coefficients and  $\{\phi_i(\mathbf{x})\}_{i=1}^N$  the set of linearly-independent functions. The terms  $\phi_i$  are usually referred to as the trial/basis/approximating functions.

The unknown coefficients  $\alpha_i$  can then be found by constructing a scheme to minimise the following residuals

$$R_s = L(u) \quad (4)$$

and

$$R_b = B(u). \quad (5)$$

This process can be stated mathematically as

$$\int_{\Omega} WR_s d\Omega + \int_{\Gamma} \tilde{W} R_b d\Gamma = 0 \quad (6)$$

where  $W$  and  $\tilde{W}$  are weighting functions to be chosen. The Galerkin formulation chooses the weighting function from the set of trial functions, i.e.  $W(\mathbf{x}) = \phi_i(\mathbf{x})$ . The above volume integrals can be evaluated numerically using Gaussian quadrature.

## 2.2 One-dimensional IRBFN representations of the field variables

In this work, the system of PDEs is of second order. Consider a  $x$  grid line. Applying the integral RBF scheme of second order, a function  $f$  and its derivatives with respect to  $x$  can be represented as follows

$$\frac{d^2 f(x)}{dx^2} = \sum_{i=1}^{N_x} w_i g_i(x) = \sum_{i=1}^{N_x} w_i I_i^{(2)}(x), \quad (7)$$

$$\frac{df(x)}{dx} = \sum_{i=1}^{N_x} w_i I_i^{(1)}(x) + c_1, \quad (8)$$

$$f(x) = \sum_{i=1}^{N_x} w_i I_i^{(0)}(x) + c_1 x + c_2, \quad (9)$$

where  $N_x$  is the number of nodes on the grid line,  $\{w_i\}_{i=1}^{N_x}$  the set of network weights, and  $\{g_i(x)\}_{i=1}^{N_x} \equiv \{I_i^{(2)}(x)\}_{i=1}^{N_x}$  the set of RBFs,  $I_i^{(1)}(x) = \int I_i^{(2)}(x) dx$ ,  $I_i^{(0)}(x) = \int I_i^{(1)}(x) dx$ , and  $c_1$  and  $c_2$  are the constants of integration.

Evaluation of Eq. 7 - Eq. 9 at the grid nodes leads to

$$\widehat{\frac{d^2 f}{dx^2}} = \widehat{\mathcal{F}}^{(2)} \widehat{\alpha}, \quad (10)$$

$$\widehat{\frac{df}{dx}} = \widehat{\mathcal{F}}^{(1)} \widehat{\alpha}, \quad (11)$$

$$\widehat{f} = \widehat{\mathcal{F}}^{(0)} \widehat{\alpha}, \quad (12)$$

where the superscript  $(\cdot)$  is used to denote the order of the corresponding derivative function;

$$\widehat{\mathcal{F}}^{(2)} = \begin{bmatrix} I_1^{(2)}(x_1), & I_2^{(2)}(x_1), & \cdots, & I_{N_x}^{(2)}(x_1), & 0, & 0 \\ I_1^{(2)}(x_2), & I_2^{(2)}(x_2), & \cdots, & I_{N_x}^{(2)}(x_2), & 0, & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ I_1^{(2)}(x_{N_x}), & I_2^{(2)}(x_{N_x}), & \cdots, & I_{N_x}^{(2)}(x_{N_x}), & 0, & 0 \end{bmatrix};$$

$$\widehat{\mathcal{F}}^{(1)} = \begin{bmatrix} I_1^{(1)}(x_1), & I_2^{(1)}(x_1), & \cdots, & I_{N_x}^{(1)}(x_1), & 1, & 0 \\ I_1^{(1)}(x_2), & I_2^{(1)}(x_2), & \cdots, & I_{N_x}^{(1)}(x_2), & 1, & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ I_1^{(1)}(x_{N_x}), & I_2^{(1)}(x_{N_x}), & \cdots, & I_{N_x}^{(1)}(x_{N_x}), & 1, & 0 \end{bmatrix};$$

$$\widehat{\mathcal{F}}^{(0)} = \begin{bmatrix} I_1^{(0)}(x_1), & I_2^{(0)}(x_1), & \cdots, & I_{N_x}^{(0)}(x_1), & x_1, & 1 \\ I_1^{(0)}(x_2), & I_2^{(0)}(x_2), & \cdots, & I_{N_x}^{(0)}(x_2), & x_2, & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ I_1^{(0)}(x_{N_x}), & I_2^{(0)}(x_{N_x}), & \cdots, & I_{N_x}^{(0)}(x_{N_x}), & x_{N_x}, & 1 \end{bmatrix};$$

$$\widehat{\alpha} = (w_1, w_2, \cdots, w_{N_x}, c_1, c_2)^T;$$

and

$$\widehat{\frac{d^k f}{dx^k}} = \left( \frac{d^k f_1}{dx^k}, \frac{d^k f_2}{dx^k}, \cdots, \frac{d^k f_{N_x}}{dx^k} \right)^T, \quad k = \{1, 2\},$$

$$\widehat{f} = (f_1, f_2, \cdots, f_{N_x})^T,$$

in which  $d^k f_j / dx^k = d^k f(x_j) / dx^k$  and  $f_j = f(x_j)$  with  $j = \{1, 2, \cdots, N_x\}$ .

The relations between the RBF-coefficient space  $\widehat{\alpha}$  and the physical space  $\widehat{f}$  are given by

$$\begin{pmatrix} \widehat{f} \\ \widehat{e} \end{pmatrix} = \begin{bmatrix} \widehat{\mathcal{F}}^{(0)} \\ \widehat{\mathcal{K}} \end{bmatrix} \widehat{\alpha} = \widehat{\mathcal{C}} \widehat{\alpha}, \quad (13)$$

$$\widehat{\alpha} = \widehat{\mathcal{C}}^{-1} \begin{pmatrix} \widehat{f} \\ \widehat{e} \end{pmatrix}, \quad (14)$$

where  $\widehat{e} = \widehat{\mathcal{K}} \widehat{\alpha}$  represents the extra information (e.g. normal derivative values at the two end-points) and  $\widehat{\mathcal{C}}$  the conversion matrix.

Making use of Eq. 14, the values of  $f$  and its derivatives at an arbitrary point  $x$  on the grid line will be computed by

$$f(x) = \left( I_1^{(0)}(x), I_2^{(0)}(x), \dots, I_{N_x}^{(0)}(x), x, 1 \right) \widehat{\mathcal{C}}^{-1} \begin{pmatrix} \widehat{f} \\ \widehat{e} \end{pmatrix}, \quad (15)$$

$$\frac{\partial f(x)}{\partial x} = \left( I_1^{(1)}(x), I_2^{(1)}(x), \dots, I_{N_x}^{(1)}(x), 1, 0 \right) \widehat{\mathcal{C}}^{-1} \begin{pmatrix} \widehat{f} \\ \widehat{e} \end{pmatrix}, \quad (16)$$

$$\frac{\partial^2 f(x)}{\partial x^2} = \left( I_1^{(2)}(x), I_2^{(2)}(x), \dots, I_{N_x}^{(2)}(x), 0, 0 \right) \widehat{\mathcal{C}}^{-1} \begin{pmatrix} \widehat{f} \\ \widehat{e} \end{pmatrix}. \quad (17)$$

They can be rewritten in compact form

$$f(x) = \sum_{i=1}^{N_x} \varphi_i(x) f_i + \varphi_{N_x+1}(x) e_1 + \varphi_{N_x+2}(x) e_2, \quad (18)$$

$$\frac{\partial f(x)}{\partial x} = \sum_{i=1}^{N_x} \frac{\partial \varphi_i(x)}{\partial x} f_i + \frac{\partial \varphi_{N_x+1}(x)}{\partial x} e_1 + \frac{\partial \varphi_{N_x+2}(x)}{\partial x} e_2, \quad (19)$$

$$\frac{\partial^2 f(x)}{\partial x^2} = \sum_{i=1}^{N_x} \frac{\partial^2 \varphi_i(x)}{\partial x^2} f_i + \frac{\partial^2 \varphi_{N_x+1}(x)}{\partial x^2} e_1 + \frac{\partial^2 \varphi_{N_x+2}(x)}{\partial x^2} e_2, \quad (20)$$

where  $\{\varphi_i\}_{i=1}^{N_x+2}$  is the set of IRBFN basis functions in the physical space.

One can take products of integrated RBFs in each direction as basis functions for the interpolation of  $f$  over the entire 2D domain. The IRBFN approximation is defined everywhere in the domain. It is easy to get the value of  $f$  at any point in the domain.

### 2.3 Imposition on boundary conditions

If PDEs are subject to Dirichlet boundary conditions only, the matrix  $\widehat{\mathcal{K}}$  and the vector  $\widehat{e}$  in Eq. 13 are simply set to null.

In the case of Cartesian coordinate system, approximate expressions for the field variable  $u$  will take the form

$$u(x,y) = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} \varphi_i^{(x)}(x) \varphi_j^{(y)}(y) u_{i,j}, \quad (21)$$

where  $N_x$  and  $N_y$  are the numbers of grid lines in the  $x$  and  $y$  directions, respectively.

Consider PDE subjected to both types of boundary conditions. Assume that Dirichlet and Neumann boundary conditions are prescribed on the two vertical and two horizontal walls, respectively. The integral approach allows one to incorporate Neumann boundary conditions into the IRBFN approximations through the integration constants. For each  $y$  grid line, the matrix  $\widehat{\mathcal{K}}$  and the vector  $\widehat{e}$  in Eq. 13 will become

$$\widehat{\mathcal{K}} = \begin{bmatrix} I_1^{(1)}(y_1), & I_2^{(1)}(y_1), & \cdots, & I_{N_y}^{(1)}(y_1), & 1, & 0 \\ I_1^{(1)}(y_{N_y}), & I_2^{(1)}(y_{N_y}), & \cdots, & I_{N_y}^{(1)}(y_{N_y}), & 1, & 0 \end{bmatrix},$$

$$\widehat{e} = \begin{pmatrix} \frac{\partial u_1}{\partial y} \\ \frac{\partial u_{N_y}}{\partial y} \end{pmatrix},$$

leading to

$$u(x,y) = \sum_{i=1}^{N_x} \varphi_i^{(x)}(x) \left( \sum_{j=1}^{N_y} \varphi_j^{(y)}(y) u_{i,j} + \varphi_{N_y+1}^{(y)}(y) \frac{\partial u_{i,1}}{\partial y} + \varphi_{N_y+2}^{(y)}(y) \frac{\partial u_{i,N_y}}{\partial y} \right). \quad (22)$$

In Eq. 21 - Eq. 22,  $u_{i,j}$  is the values of the  $u$  variable at the intersection of the  $i$ th horizontal grid line and  $j$ th vertical grid line; the products  $\varphi_i^{(x)} \varphi_j^{(y)}$  are usually referred to as the trial/basis/approximating functions; and  $\partial u_{i,1}/\partial y$  and  $\partial u_{i,N_y}/\partial y$  are nodal boundary derivative values.

### 3 Application of the proposed technique

Simulation of natural convection in a square slot is conducted here to further validate the proposed technique. For this benchmark test problem, uniform grids are used to represent the computational domain, and 1D-IRBFNs are implemented with the multiquadric (MQ) function

$$g_i(x) = \sqrt{(x - c_i)^2 + a_i^2},$$

where  $c_i$  and  $a_i$  are the centre and the width/shape-parameter of the  $i$ th MQ-RBF. The MQ width is simply chosen to be the grid size.

### 3.1 Problem definition

The governing equations which are obtained from the streamfunction-vorticity-temperature formulation can be written as

$$\frac{\partial^2 \psi}{\partial x^2} + \frac{\partial^2 \psi}{\partial y^2} = -\omega, \quad (23)$$

$$\frac{\partial \omega}{\partial t} + u \frac{\partial \omega}{\partial x} + v \frac{\partial \omega}{\partial y} = \sqrt{\frac{Pr}{Ra}} \left( \frac{\partial^2 \omega}{\partial x^2} + \frac{\partial^2 \omega}{\partial y^2} \right) + \frac{\partial T}{\partial x}, \quad (24)$$

$$\frac{\partial T}{\partial t} + u \frac{\partial T}{\partial x} + v \frac{\partial T}{\partial y} = \frac{1}{\sqrt{RaPr}} \left( \frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} \right), \quad (25)$$

where

$$u = \frac{\partial \psi}{\partial y}, \quad v = -\frac{\partial \psi}{\partial x},$$

In Eq. 25,  $Pr$  and  $Ra$  are the Prandtl and Rayleigh numbers defined as  $Pr = \nu/\alpha$  and  $Ra = \beta g \Delta T L^3 / \alpha \nu$ , respectively in which  $\nu$  is the kinematic viscosity,  $\alpha$  the thermal diffusivity,  $\beta$  the thermal expansion coefficient,  $g$  the gravity, and  $L$  and  $\Delta T$  the characteristic length and temperature difference, respectively. In this dimensionless scheme, the velocity scale is taken as  $U = \sqrt{gL\beta\Delta T}$  for the purpose of balancing the buoyancy and inertial forces.

The domain of interest is an enclosed square slot with all stationary walls, leading to  $\psi = \partial\psi/\partial n = 0$  on the boundaries. The two horizontal walls are adiabatic (i.e.  $\partial T/\partial y = 0$ ), while the two vertical walls are maintained at constant temperatures (i.e.  $T = +0.5$  (left wall) and  $T = -0.5$  (right wall)).

### 3.2 Computation boundary conditions for the vorticity

The computation boundary conditions for the vorticity can be derived from the streamfunction equation. The process is similar to that in [Ho-Minh, Mai-Duy, and Tran-Cong (2009)].

Taking into account the streamfunction boundary values (i.e.  $\psi = 0$ ), expressions for the vorticity on the boundaries will reduce to

$$\omega = \frac{\partial^2 \psi}{\partial n^2}, \quad (26)$$

where  $n$  is the local direction normal to the wall.

Consider a  $x$  grid line. Owing to the fact that the present coefficient vector is larger, one can add two extra equations representing  $\partial\psi_1/\partial x$  and  $\partial\psi_{N_x}/\partial x$  to the conversion process

$$\begin{pmatrix} \widehat{\psi} \\ \frac{\partial \psi_1}{\partial x} \\ \frac{\partial \psi_{N_x}}{\partial x} \end{pmatrix} = \begin{bmatrix} \widehat{\mathcal{F}}^{(0)} \\ \widehat{\mathcal{K}} \end{bmatrix} \widehat{\alpha} = \widehat{\mathcal{C}} \widehat{\alpha}, \quad (27)$$

in which  $\widehat{\mathcal{K}}$  is the matrix made up of the first and last rows of  $\widehat{\mathcal{F}}^{(1)}$ , i.e.

$$\widehat{\mathcal{K}} = \begin{bmatrix} I_1^{(1)}(x_1), & I_2^{(1)}(x_1), & \cdots, & I_{N_x}^{(1)}(x_1), & 1, & 0 \\ I_1^{(1)}(x_{N_x}), & I_2^{(1)}(x_{N_x}), & \cdots, & I_{N_x}^{(1)}(x_{N_x}), & 1, & 0 \end{bmatrix}.$$

It can be seen from Eq. 27 that, despite the presence of nodal derivative values, the approximate solution  $\psi$  is collocated at the whole set of centres on the grid line.

The second derivatives of  $\psi$  at the two boundary points can now be expressed in terms of the values of  $\psi$  at every point on the grid line and the values of  $\partial\psi/\partial x$  at the two boundary points ( $x_1, x_{N_x}$ )

$$\begin{pmatrix} \frac{\partial^2 \psi_1}{\partial x^2} \\ \frac{\partial^2 \psi_{N_x}}{\partial x^2} \end{pmatrix} = \widehat{\mathcal{D}} \widehat{\mathcal{C}}^{-1} \begin{pmatrix} \widehat{\psi} \\ \frac{\partial \psi_1}{\partial x} \\ \frac{\partial \psi_{N_x}}{\partial x} \end{pmatrix}, \quad (28)$$

where  $\widehat{\mathcal{D}}$  is the sub-matrix of  $\widehat{\mathcal{F}}^{(2)}$  (i.e. the first and last rows)

$$\widehat{\mathcal{D}} = \begin{bmatrix} I_1^{(2)}(x_1), & I_2^{(2)}(x_1), & \cdots, & I_N^{(2)}(x_1), & 0, & 0 \\ I_1^{(2)}(x_{N_x}), & I_2^{(2)}(x_{N_x}), & \cdots, & I_N^{(2)}(x_{N_x}), & 0, & 0 \end{bmatrix},$$

and  $\widehat{\mathcal{C}}$  is defined in Eq. 27.

It can be seen that the IRBFN approximations for  $\partial^2\psi/\partial x^2$  at the boundaries satisfy exactly the prescribed derivative boundary values. With Eq. 28, we can obtain the computational boundary conditions for the vorticity. On a  $y$  grid line, the process can be taken in a similar fashion.

### 3.3 Solution procedure

Using the 1D-IRBFN scheme introduced in section 2.2 and 2.3, the approximation of  $\psi, \omega$  is taken the form of Eq. 21. Because the energy equation Eq. 25 is subject to both types of boundary conditions, the variable  $T$  can be approximated by Eq. 22.

Follow the Galerkin discretisations of the PDEs outlined in 2.1, one will obtain the three sets of discretised form of governing equations Eq. 23 - Eq. 25. Note that the IRBFNs approximations satisfy a priori not only the Dirichlet boundary conditions but also the Neumann boundary conditions. As a result, the boundary-integral terms in the Galerkin weighting process can be eliminated.

In this paper, we will adopt a time-marching approach, where the diffusion and convection terms are treated implicitly and explicitly, respectively. All equations involve the Laplacian term and their discrete form remains unchanged during the solution process. At each time level, the three equations are solved separately for efficiency purposes. The solution procedure can be summarised as follows.



1. Guess values of  $T$ ,  $\psi$ ,  $\omega$  and their first-order spatial derivatives at time  $t = 0$
2. Discretise spatial derivatives using 1D-IRBFNs, resulting in a high-order approximation scheme in space
3. Discretise time derivatives using Euler (forward difference) method, resulting in a first-order accurate scheme in time
4. Compute the convective terms and the boundary values for  $\omega$  with the process given in section 3.2
5. Solve the energy equation Eq. 25 for  $T$ , subject to Dirichlet and Neumann conditions  
Solve the vorticity equation Eq. 24 for  $\omega$ , subject to Dirichlet conditions  
Solve the streamfunction equation Eq. 23 for  $\psi$ , subject to Dirichlet conditions
6. Check to see whether the solution has reached a steady state

$$\frac{\sqrt{\sum_{i=1}^N \left( T_i^{(k)} - T_i^{(k-1)} \right)^2}}{\sqrt{\sum_{i=1}^N \left( T_i^{(k)} \right)^2}} < \varepsilon, \quad (29)$$

where  $k$  is the time level and  $\varepsilon$  is a prescribed tolerance

7. If it is not satisfied, advance time step and repeat from step 3. Otherwise, stop the computation and output the results.

### 3.4 Results and discussion

Numerical results for this problem are extensive. A range of  $Ra$  from  $10^3$  to  $10^6$  has been widely used for the validation of new numerical schemes. Davis (1983) provided finite-difference results which have been then often cited in the literature for comparison purposes. Later on, there are increased levels of interest for higher values of  $Ra$ , namely  $10^7$  and  $10^8$ . Works reported include [Qu  r   (1991)] (the pseudo-spectral method), [Wan, Patnail, and Wei (2001)] (FEM), [Wan, Patnail, and Wei (2001)] (discrete singular convolution (DSC) method), [Sadat and Couturier (2000)] (meshless diffuse approximation method (DAM)), and [Kosec and Sarler (2007)] (mesh-free local RBF collocation method (RBFCM)). For this higher range, it has been generally observed that (i) the strength of boundary layers is significantly increased, (ii) convergence becomes much more difficult, and (iii) significant discrepancies in the Nusselt number occur in some cases (e.g. between the pseudo-spectral technique [Qu  r   (1991)] and the DSC method [Wan, Patnail, and Wei (2001)]).

The Galerkin-IRBFN method is employed to study this problem for  $Ra = 10^7$  and  $Ra = 10^8$ . Results are presented in the form of contour plots for  $\psi$ ,  $\omega$  and  $T$  and through the values of the following quantities

- The average Nusselt number on the vertical plane at  $x = 1/2$  (middle cross-section), which is defined by

$$Nu_{1/2} = Nu(x = 1/2, y),$$

in which

$$Nu(x, y) = \int_0^1 \left( uT - \frac{\partial T}{\partial x} \right) dy. \quad (30)$$

- The average Nusselt number throughout the cavity, which is defined by

$$\overline{Nu} = \int_0^1 Nu(x, y) dx. \quad (31)$$

It is noted that integrals Eq. 30 and Eq. 31 are computed here using Simpson's rule. Results for  $Ra$  from  $10^7$  to  $10^8$  are presented in Tab. 1 and Fig. 1. Tab. 1 shows a comparison of the average Nusselt numbers between the present method and several other methods. It can be seen that there are significant discrepancies among various numerical techniques. For the case of  $Ra = 10^7$ , the DSC [Wan, Patnail, and Wei (2001)] and FEM [Manzari (1999)] produced the values of 13.86 and 13.99 for the average Nusselt number, while the pseudo-spectral [Quéré (1991)], FE [Wan, Patnail, and Wei (2001)], DA [Sadat and Couturier (2000)] and RBFCM [Kosec and Sarler (2007)] techniques yielded the following values: 16.523, 16.656, 16.59 and 16.92. The differences between the two groups are much wider for the case of  $Ra = 10^8$ : 23.67 for the DSC method, and (30.225, 31.486, 30.94, 32.12) for the second group. The Galerkin-IRBFN results are in close agreement with the second group, particularly with the pseudo-spectral technique [Quéré (1991)]. Variations of the local Nusselt number on the left and right walls are presented in Fig. 2. It is clearly shown that the proposed technique is able to capture very stiff changes of the local Nusselt number in the region close to the boundary. It can be seen from Fig. 1, the present contour plots for the streamfunction, vorticity and temperature variables look feasible when compared with those of the pseudo-spectral technique [Quéré (1991)]. Very thin boundary layers are formed at these high values of  $Ra$ . It is noted that iso-values used in these plots are the same as those used in [Quéré (1991)].

#### 4 Concluding remarks

The Galerkin-IRBFN method is successfully applied to simulate natural convection governed by the streamfunction-vorticity-temperature formulation in two dimensions. Numerical experiments (e.g. the achievement of very high  $Ra$  solutions) confirm attractive features of our method: (i) easy implementation, (ii) effective treatment of the vorticity boundary condition, (iii) effective handling

Table 1: Natural convection flow in a square slot: Comparison of the Galerkin-IRBFN results with those of other techniques at  $Ra = 10^7$  and  $Ra = 10^8$  ( $Pr = 0.71$ )

$Ra$	Technique	$\overline{Nu}$	$Nu_{1/2}$
$10^7$	Present study (Grid size: $91 \times 91$ )	16.661	16.661
	[Quéré (1991)]	16.523	16.523
	[Manzari (1999)]	13.99	
	[Sadat and Couturier (2000)]	16.59	
	[Wan, Patnail, and Wei (2001)] (FEM)	16.656	
	[Wan, Patnail, and Wei (2001)] (DSC)	13.86	
	[Kosec and Sarler (2007)]	16.92	
$10^8$	Present study (Grid size: $91 \times 91$ )	30.548	30.525
	[Quéré (1991)]	30.225	30.225
	[Sadat and Couturier (2000)]	30.94	
	[Wan, Patnail, and Wei (2001)] (FEM)	31.486	
	[Wan, Patnail, and Wei (2001)] (DSC)	23.67	
	[Kosec and Sarler (2007)]	32.12	

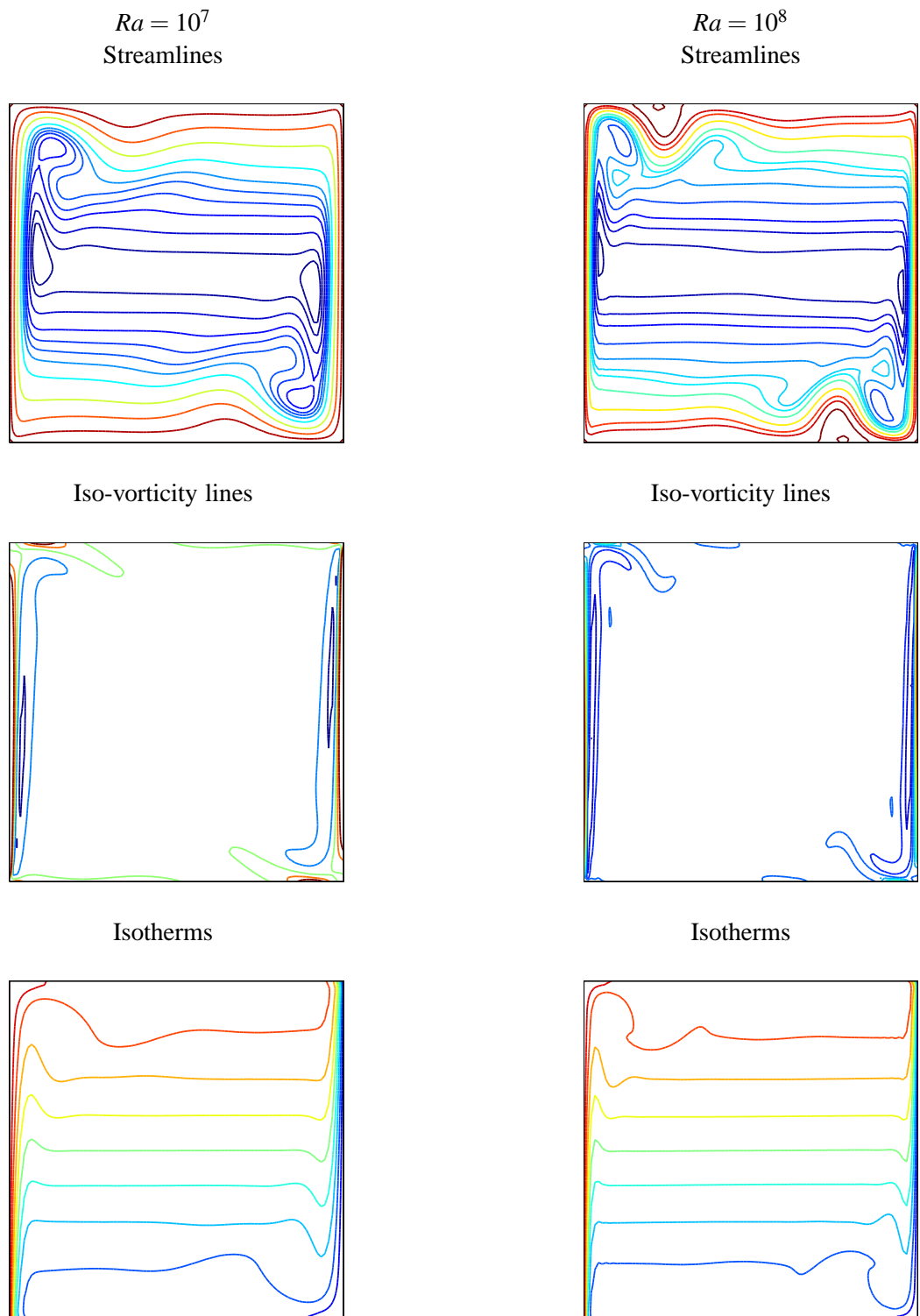


Figure 1: Natural convection flow in a square slot: Contour plots for the  $\psi$ ,  $\omega$  and  $T$  variables at  $Ra = 10^7$  and  $Ra = 10^8$  using a grid of  $91 \times 91$ . Iso-values used in these plots are the same as those in [Quéré (1991)].

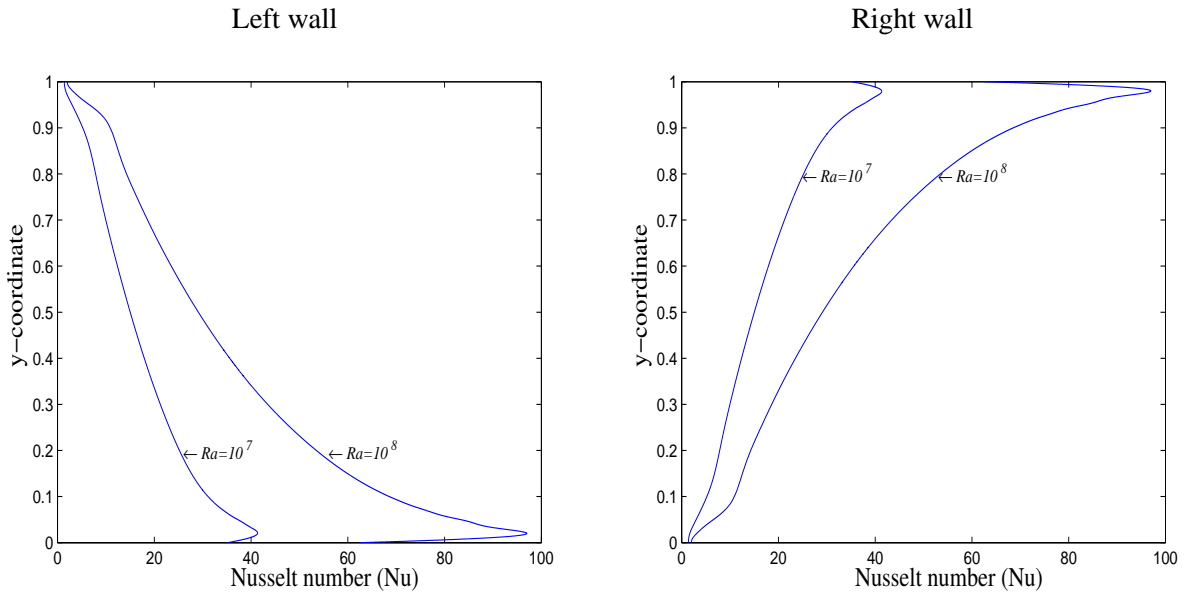


Figure 2: Natural convection flow in a square slot: Variations of the local Nusselt number along the left and right walls.

of the Neumann boundary condition, and (iv) ability to capture very thin boundary layers using relatively-coarse grids.

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