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Citation for published version:
Hirol, H, Mohd Noor, MA, Abd Jamil, MZ, Mokhtaram, MH, Kasiman, EH & Mohd Yassin, AY 2020, 'Performance of radial point interpolation method in solving kinematic wave equation for hydrologic modelling', Jurnal Teknologi, vol. 82, no. 3, pp. 61-68. https://doi.org/10.11113/jt.v82.14066

Digital Object Identifier (DOI):
10.11113/jt.v82.14066

Link:
Link to publication record in Heriot-Watt Research Portal

Document Version:
Publisher's PDF, also known as Version of record

Published In:
Jurnal Teknologi

Publisher Rights Statement:
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PERFORMANCE OF RADIAL POINT INTERPOLATION METHOD IN SOLVING KINEMATIC WAVE EQUATION FOR HYDROLOGIC MODELLING

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Abstract

This paper presents the solution of the kinematic wave equation using a meshless radial point interpolation method (RPIM). The partial differential equation is discretized using a Galerkin weighted residual method employing RPIM shape functions. A forward difference scheme is used for temporal discretization, while the direct substitution method is employed to solve the nonlinear system at each time step. The formulation is validated against solutions from conventional numerical techniques and physical observation. In all cases, excellent agreements are achieved and hence the validation of the proposed formulation. Optimum values of the multi-quadrics shape parameters were then determined before the assessment of the performance of the method. Based on the convergence rate, it has been shown that the proposed method performs better than the finite difference method and equivalent to the finite element method. This highlights the potential of RPIM as an alternative method for hydrologic modeling.

Keywords: Meshless method, radial point interpolation method, saint venant kinematic wave, hydrologic modeling, finite element method, finite difference method

Abstrak

Kertas kerja ini membentangkan penyelesaian persamaan gelombang kinematik dengan menggunakan kaedah Interpolasi Titik Radial Tanpa Mesh (RPIM). Persamaan pembezaan separa dibincangkan dengan menggunakan kaedah pemberat residual Galerkin yang menggunakan fungsi bentuk RPIM. Skema perbezaan hadapan digunakan untuk pengantian sementara manakala kaedah penggantian langsung digunakan untuk menyelesaikan sistem tak linear pada setiap langkah masa. Perumusan ini disahkan dengan membuat perbandingan dengan penyelesaian dari teknik numerik konvensional dan pemerhatian fizikal. Dalam semua kes, keputusan yang sangat baik dicapai dan dengan itu menjadi pengesahan kepada rumusan yang dicadangkan. Nilai optimum parameter bentuk multi-quadric kemudian ditentukan sebelum penilaian prestasi kaedah tersebut. Berdasarkan kadar konvergen, telah ditunjukkan bahawa kaedah yang dicadangkan...
1.0 INTRODUCTION

A combination of one-dimensional continuity and momentum equations, also known as Saint Venant equations are a set of nonlinear unsteady partial differential equations. These equations are typically used in the hydrological modelling of surface runoff and channel flow. The full Saint Venant equations, commonly known as the dynamic wave equations, are often used to obtain the complete description of the flow. However, in most cases, where no backwater effect is expected and when the bed and frictional slope can be assumed as similar, the momentum equation can be simplified and represented by the Manning equation. The combination of continuity and Manning equations yields what is known as Kinematic Wave equation. While dynamic wave equation provides better description of the flow, the numerical solutions are expensive and usually unstable. Therefore, unless necessary, the kinematic wave has been the preferred method to solve many hydrological related problems [1]. However, despite its simplicity, the equation is still unsteady and nonlinear in nature. Due to the nonlinearity, an exact solution of the equations is difficult to be obtained. Nevertheless, a limited number of exact solutions for simplified cases are available in [2-5]. For other cases, the kinematic wave equation is typically solved numerically using finite difference method (FDM) [6] or finite element method (FEM) [7, 8].

1.1 Radial Point Interpolation Method (RPIM): A Meshless Method

A typical numerical method such as FDM or FEM requires the construction of grid or mesh to represent the physical domain of the problem. In most engineering problems where the physical domain is complex, geometrical meshing is difficult and a time consuming process. Meshless methods are some recent attempts to remedy this shortcoming as it requires no meshing of the domain [9,10].

RPIM can be considered as an enhancement of an earlier method known as Point Interpolation Method (PIM). The latter is a meshless method that uses polynomial functions to approximate the field variables. The application of PIM in hydrological modelling of flood routing has been detailed in Hirol, (2016) and Hirol, et. al. (2017) [11, 12]. The use of PIM might lead to matrix singularity during the derivation of the shape functions due to the inappropriate choice of polynomial function, leading to an ill-conditioned matrix.

Alternatively, instead of polynomial function, a radial basis function can be used as an interpolation functions, the latter becomes the basis for RPIM. Liu and Gu (1999) [13] is the first to propose RPIM and the method has since been extended in [14-16]. Unlike PIM, RPIM is free from ill-conditioned matrix and singularity problem [17, 18]. The delta function property is also preserved in RPIM allowing boundary conditions to be imposed conveniently. These attractive features of RPIM motivate extension of the method to solve kinematic wave equation for hydrologic modelling.

2.0 METHODOLOGY

Governing Equations

Saint Venant equations [19] are time dependent partial differential equations which describe the distribution of flow rate, Q and flow cross-sectional area, A as functions of distance, x along the channel and time t. The equations can be given as:

Equation of Mass

$$\frac{\partial A}{\partial t} + \frac{\partial Q}{\partial x} = q(x)$$

(1)

where $A$ is the cross-sectional area of the flow, $Q$ is the flow rate and $q(x)$ is the forcing term (i.e. precipitation, lateral flow).

Equation of Momentum

$$\frac{1}{A} \frac{\partial Q}{\partial t} + \frac{1}{A} \frac{\partial}{\partial x} \left( \frac{Q^2}{A} \right) + g \frac{\partial y}{\partial x} - g(S_b - S_f) = 0$$

(2)

where $S_b$ is the bed slope and $S_f$ is the frictional slope, whilst $y$ and $g$ are the depth of water and gravitational pull, respectively. The complete form of Equation (2) is termed as full dynamics equation. However, Equation (2) can be further simplified if it is assumed that $S_b = S_f$. This is known as the kinematic
wave assumption. This condition can be equivalently expressed in Manning form as

$$A = aQ^b$$  \hspace{1cm} (3)$$

Equation (1) and (3) are the Saint Venant kinematic wave equations. By combining Equation (1) and Equation (3), the following equation can be obtained.

$$\frac{\partial Q}{\partial t} + \frac{1}{a^bQ^b} \frac{\partial Q}{\partial x} = q$$  \hspace{1cm} (4)$$

Weak Form of the Kinematic Wave Equation

The kinematic wave equation given by Equation (4) can be solved numerically by converting the equation into weak form. This can be done by employing Galerkin weighted residual method. We first discretize Equation (4) in time by forward difference to obtain

$$\frac{Q^{t+1} - Q^t}{\Delta t} + \frac{1}{a^bQ^b} \frac{\partial Q^{t+1}}{\partial x} = q^{t+1}$$  \hspace{1cm} (5)$$

where \(t + 1\) and \(t\) refer to present and previous time-step, respectively. Rearranging gives

$$Q^{t+1} + \frac{\Delta t}{a^bQ^b} \frac{\partial Q^{t+1}}{\partial x} = Q^t + q^{t+1}$$  \hspace{1cm} (6)$$

By weighting Equation (6) using shape functions, \(N_i\) and expressing the flow rate as \(Q = N_i \hat{Q}_j\), where \(\hat{Q}_j\) are the nodal values of \(Q\), the following is obtained.

$$\int_L N_i \left( \frac{\Delta t}{a^bQ^b} \frac{\partial Q^{t+1}}{\partial x} \right) dx = \int_L N_i q dx$$  \hspace{1cm} (7)$$

To note, in Equation (7), superscript \(t + 1\) is omitted for ease of notation. By collecting the nodal values, \(\hat{Q}_j\) and shifting known terms to the right hand side of the equation, Equation (7) can now be given as

$$\left( \int_L N_i dx + \int_L \frac{\Delta t}{a^bQ^b} \frac{\partial N_i}{\partial x} dx \right) \hat{Q}_j = \int_L N_i q dx$$  \hspace{1cm} (8)$$

Equation (8) can be represented in matrix form as

$$[M + K(\hat{Q})]\{\hat{Q}\} = \{-F\}$$  \hspace{1cm} (9)$$

where \([K]\) is the coefficient matrix, \([M]\) is the mass matrix whilst \(\{\hat{Q}\}\) and \(\{F\}\) are the vector of nodal values and nodal loads, respectively.

Nonlinear Solver

Equation (9) is nonlinear thus requires a nonlinear solver. In this work, direct substitution scheme is employed. For mild nonlinearity the scheme works well.

Derivation of RPIM Shape Functions

RPIM interpolation function contains both radial and polynomials basis given as

$$Q = \{R\}\{a\}^T + \{P\}\{b\}^T$$  \hspace{1cm} (10)$$

where \(\{R\}\) is the vector of radial basis function (RBF), \(\{a\}\) and \(\{b\}\) are the vectors of the coefficient of RBF and the monomials, \(\{P\}\) respectively. There are several forms of radial basis functions that can be used as detailed in [16]. This study uses multi-quadrics (MQ) interpolation which, for one-dimensional problem can be given as

$$\{R\} = ((r)^2 + (\alpha_c d_c)^2)$$  \hspace{1cm} (11)$$

where \(\alpha_c\) and \(q\) are the shape parameters which values can be obtained through numerical test whilst \(r\) is the vector of radial distance of point of interest, \(x_{pi}\) given as

$$\{r\} = \sqrt{(x - x_{pi})^2}$$  \hspace{1cm} (12)$$

Coefficients in \(\{a\}\) and \(\{b\}\) of Equation (10) are constants which values can be determined by forcing the interpolation to be equal to nodal value of the field variable at the node’s location. In evaluated form, this can be given as

$$\{\hat{Q}\}^T = [R|_n]\{a\}^T + [P|_n]\{b\}^T$$  \hspace{1cm} (13)$$

where \(\{\hat{Q}\}\) is the vector of the nodal values of the flow rate, \(\{R|_n\}\) and \(\{P|_n\}\) contain the evaluated values of radial basis function and the monomials evaluated at the location of the nodes respectively. Due to the additional polynomial terms, the simultaneous equation of Equation (13) can be expanded into

$$\{\hat{Q}\}^T = [R|_n] \begin{bmatrix} a \end{bmatrix} + [P|_n] \begin{bmatrix} b \end{bmatrix}$$  \hspace{1cm} (14)$$

where
\[
\{ \tilde{Q} \} = \begin{bmatrix} Q_1 & Q_2 & \ldots & Q_n & 0 & \ldots & 0 \end{bmatrix}
\]

(15)

\[
\begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} a_1 & a_2 & \ldots & a_n & b_1 & b_2 & \ldots & b_m \end{bmatrix}
\]

(16)

\[ [G]_{m} \] is termed as moment matrix. \( n \) is the number of the field node in the support domain and \( m \) is the number of polynomial terms used which can be zero or \( m \ll n \). Using Equation (14), we can solve the corresponding coefficients as

\[
\begin{pmatrix} a \\ b \end{pmatrix}^T = [G]_{m}^{-1} \{ \tilde{Q} \}^T
\]

(17)

By inserting Equation (17) into Equation (10) we obtain

\[
Q = (R \cdot P) [G]_{m}^{-1} \{ \tilde{Q} \}^T
\]

(18)

In the same time, the flow rate can also be expressed in terms of shape functions and nodal values as

\[
Q = (N) \{ \tilde{Q} \}^T
\]

(19)

Since Equation (18) and Equation (19) represent the same distribution of the flow rate, evaluating the equations at a point of interest, \( X_{pi} \) would give

\[
\{ N \}_{pi} \{ \tilde{Q} \}^T = \{ R \}_{pi} \{ P \}_{pi} [G]_{m}^{-1} \{ \tilde{Q} \}^T
\]

(20)

After rearrangement, Equation (20) can be given as

\[
[G]_{m}^T \{ N \}_{pi} \{ \tilde{Q} \}^T = \{ R \}_{pi} \{ P \}_{pi}^T
\]

(21)

From Equation (21), shape functions of RPIM at a point of interest i.e. \( \{ N \}_{pi} \) can be solved by employing simultaneous solver such as Gauss elimination thus avoiding the need for direct inversion.

**Derivation of the First Derivative of RPIM Shape Functions**

Differentiating Equation (10) and Equation (19) and evaluating at a point of interest, \( X_{pi} \), give

\[
\begin{pmatrix} \frac{\partial N_1}{\partial x_{pi}} & \ldots & \frac{\partial N_n}{\partial x_{pi}} \end{pmatrix} \begin{pmatrix} \tilde{Q}_1 \\ \vdots \\ \tilde{Q}_n \end{pmatrix}
\]

(22)

\[
= \begin{pmatrix} \frac{\partial R_1}{\partial x_{pi}} & \ldots & \frac{\partial R_n}{\partial x_{pi}} \\ \vdots & \vdots & \vdots \\ \frac{\partial P_1}{\partial x_{pi}} & \ldots & \frac{\partial P_m}{\partial x_{pi}} \end{pmatrix} \begin{pmatrix} a_1 \\ \vdots \\ b_m \end{pmatrix}
\]

Equation (22) can be compactly expressed as

\[
\begin{pmatrix} \frac{\partial N_1}{\partial x} \\ \vdots \\ \frac{\partial N_n}{\partial x} \end{pmatrix} \{ \tilde{Q} \}^T = \begin{pmatrix} \frac{\partial R_1}{\partial x} \\ \vdots \\ \frac{\partial P_m}{\partial x} \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix}^T
\]

(23)

Inserting Equation (14) into Equation (23) gives

\[
\begin{pmatrix} \frac{\partial N_1}{\partial x} \\ \vdots \\ \frac{\partial N_n}{\partial x} \end{pmatrix} \{ G \}_{m} \{ a \\ b \}^T = \begin{pmatrix} \frac{\partial R_1}{\partial x} \\ \vdots \\ \frac{\partial P_m}{\partial x} \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix}^T
\]

(24)

After some cancellations and rearrangement, we obtain

\[
[ G ]_{m}^T \left\{ \frac{\partial N_1}{\partial x} \right\}^T = \left\{ \frac{\partial R_1}{\partial x} \right\}^T + \left\{ \frac{\partial P_m}{\partial x} \right\}^T
\]

(25)

By solving \( \{ \frac{\partial N_1}{\partial x} \} \), we then obtain the evaluated values of the derivative of the RPIM shape functions at the point of interest.

**Final Form of Kinematic Wave for RPIM Formulation**

Inserting \( \{ N \}_{pi} \) of Equation (21) and \( \{ \frac{\partial N_1}{\partial x} \} \) of Equation (25) into the discretized weak formulation of Equation (8), the final form of the RPIM formulation for the kinematic wave can be given as (in numerical integration forms)

\[
[M] = M_{ij} = \sum_{f=1}^{BC} \sum_{g=1}^{GP} \tilde{W}_g N_{i|g} N_{j|g} \left| J_f \right|
\]

(26)

\[
[K] = K_{ij} = \sum_{f=1}^{BC} \sum_{g=1}^{GP} \frac{\Delta t}{\alpha B} N_{i|g} (N_{k|g} Q_{k|g})^{(1-\beta)} \frac{\partial N_{i|g}}{\partial x} \left| J_f \right|
\]

(27)

\[
[F] = F_{i} = \sum_{f=1}^{BC} \sum_{g=1}^{GP} \tilde{W}_g N_{i|g} N_{j|g} Q_{j} + N_{i|g} \left| J_f \right|
\]

(28)

\[
[T] = T_{ij} = \sum_{f=1}^{BC} \sum_{g=1}^{GP} \tilde{W}_g \left[ N_{i|g} N_{j|g} \right.
\]

(29)

\[ + (2-\beta) \frac{\Delta t}{\alpha B} N_{i|g} (N_{k|g} Q_{k|g})^{(1-\beta)\times-1} \frac{\partial N_{i|g}}{\partial x} \left| J_f \right| \]

where \( \left| J_f \right| \) is the Jacobian for \( f \)th background cell and \( \tilde{W}_g \) is the Gauss weighting factor for the \( g \)th Gauss point. GP is the total number of Gauss points.
and BC is the background cell of the Gauss quadrature.

### 3.0 RESULTS AND DISCUSSION

Prior to the determination of the optimum value of the shape parameters of the radial basis function, the derived formulation is verified herein against two of benchmark cases.

**Case 1: Verification against Chow et al (1988) [6]**

The first verification is made against the numerical solution obtained from FDM as detailed in [6]. The flow is driven by a time-varying inflow as given in Table 1. The hypothetical channel, discretized into a finite grid system, is shown in Figure 1. The channel has a bed slope of one percent and a Manning’s roughness factor of 0.035. There is no lateral flow or rainfall. The initial condition is a uniform flow of 2000 cfs along the channel.

![Figure 1 Uniform discretization of the channel [6]](image)

**Table 1 Inlet data [6]**

| Inflow Time (min) | Inflow Rate (cfs) |
|-------------------|-------------------|
| 0                 | 2000              |
| 12                | 2000              |
| 24                | 3000              |
| 36                | 4000              |
| 48                | 5000              |
| 60                | 6000              |
| 72                | 5000              |
| 84                | 4000              |
| 96                | 3000              |
| 108               | 2000              |
| 120               | 2000              |

Figure 2 (a) shows the plot of flow rate, Q calculated at various locations whilst Figure 2 (b) shows the plot of Q a distance of 6000 ft from upstream. In the latter, close agreement can be observed between the results given by RPIM and and the one given by Chow (1988). It can also be seen that the RPIM solution converges with the increase in the number of node.

In the latter, close agreement can be observed between the results given by RPIM and and the one given by Chow (1988). It can also be seen that the RPIM solution converges with the increase in the number of node.

**Figure 2 Flow rate validation for Case 1**

**Case 2: Verification against gauged (real) data (Litrico et al. 2010) [8]**

In contrast to the previous case, Litrico et al. (2010) [8] dealt with real data, gauged from Jacui River in Brazil. The flow was driven by time-varying upstream boundary conditions (varying inflow) as shown in Figure 3. The data consisted of propagation of dam release on the Jacui River in Brazil between Itauba and Volta Grande, recorded at a time step of 30 minutes. Table 2 gives the data for the river.
Multi-quadrics (MQ) as the radial basis function, the specification of two shape parameters, $a_c$ and $q$, is required which values are best determined by conducting a numerical test. Optimum values are taken as those that provide the lowest error norms which can be calculated as

$$L_2\text{-norm error}, e = \frac{1}{N} \sum_{j=1}^{N} \left| \frac{Q(x_j)^{RPIM} - Q(x_j)^{FEM}}{Q(x_j)^{FEM}} \right|^2$$  \hspace{1cm} (30)$$

where $N$ is the number of results considered whilst $Q(x_j)^{RPIM}$ and $Q(x_j)^{FEM}$ are the values of flow rates obtained from RPIM and FEM respectively. Equation (30) is a $L_2$-norm loss function formulation. To note, since there is no closed-form solution available for the two benchmark cases, the “accurate” solution herein, i.e. $Q(x_j)^{FEM}$ is taken as the converged value from FEM formulation.

Since there are two parameters to be determined, one is set constant at a time. Herein, the optimum value of $q$ is first sought by setting the value of $a_c$ to unity. Figure 5 shows the error norms of the flow rates from the benchmark cases for varying values of shape parameter $q$ and numbers of nodes (i.e., 11, 21 and 41). From the plots, while the results seem to be insensitive to $q$ for Case 1, there is a slight reduction in the error norms for Case 2. Narrowing the range of the reduction in between 0.6 and 0.8, the optimum value for $q$ is chosen as 0.7.

Having determined the optimum value of $q$, the optimum value for $a_c$ is then determined. Similar trend is observed where there is a slight reduction in error norms for Case 2 as shown in Figure 6. Narrowing the range of the reduction in between 0.5 and 1.5, the optimum value for $a_c$ is chosen as unity.

Numerical Test to Determine the Optimum Values of Shape Parameters

Selection of shape parameters value to be used in the analysis is crucial as it governs the performance and the accuracy of a formulation [20]. By adopting

Figure 3 Upstream flow [8]

Table 2 Data of Jacui River [8]

| Data            | Value      |
|-----------------|------------|
| Channel Length (L) | 29600 m    |
| Width (W)       | 55.6 m     |
| Manning coefficient (n) | 0.07     |
| Slope (Sb)      | 0.00089    |

Figure 4 Flow rate validation for Case 2

Figure 5 Optimum value determination for shape parameter $q$

Figure 6 L2 norm error $e$ for Case 2
Convergence Study and Assessment on the Numerical Performance

In numerical formulation related works, convergence study is an important stage as it allows the assessment of the performance of a newly derived formulation. The performance is measured in terms of the convergence rate; the rate at which the solution approaches the "correct" solution. Herein, the numerical performance of RPIM is assessed by comparing its convergence rate against FEM and FDM. To ensure that RPIM performs at its best, optimum values from previous tests are used.

By referring to Figures 7 and 8, the different performance between the weighted-residual based formulations (FEM and RPIM) and the collocation-based FDM is apparent where the formers converge faster than the latter despite all being of 1st order accuracy. It is well known that the averaging of the error inherent in the weighted residual formulation would lead to a better performance. In this context, RPIM performs identical to FEM and better than FDM. This highlights the potential of RPIM as an alternative numerical method to FEM for hydrologic modelling as the method, whilst has equal performance, does not require the formation of mesh and assembly process.

4.0 CONCLUSION

This paper details the RPIM formulation in solving Saint Venant Kinematic Wave equation. The formulation is verified against a couple of benchmark problems. For best performance, the optimum value of the shape parameters is determined through a numerical test. Employing the optimum values, the performance of the RPIM formulation is assessed against the conventional FEM and FDM in a convergence study. It is shown that RPIM performs similar to FEM and better than FDM. This highlights the potential of the RPIM as an alternative numerical method for hydrologic modelling as it does not require the formation of mesh and assembly process.
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