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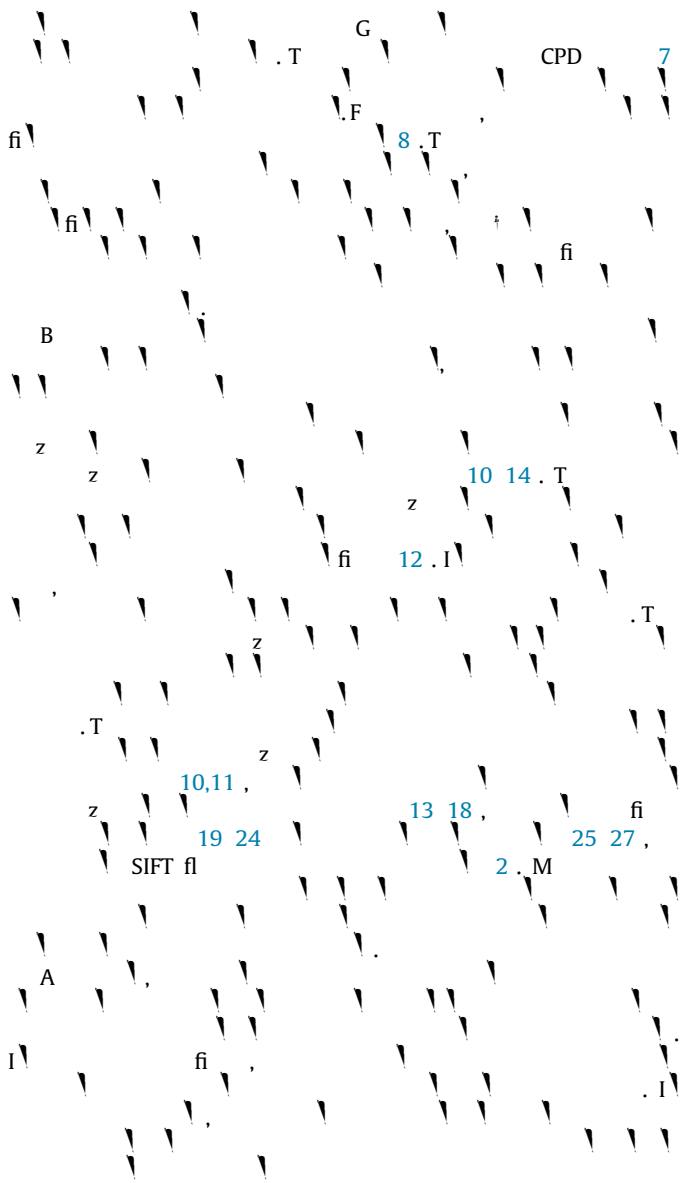
# Applied Soft Computing

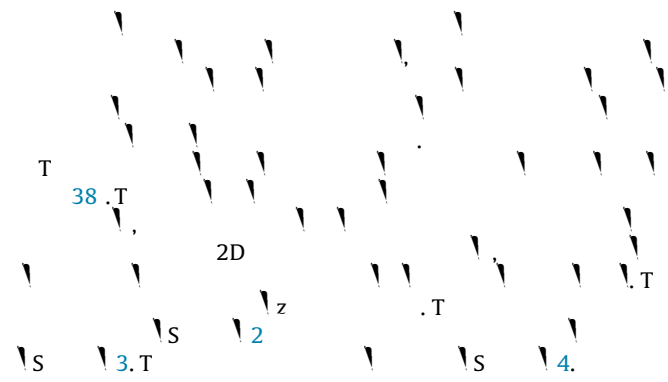
journal homepage: [www.elsevier.com/locate/asoc](http://www.elsevier.com/locate/asoc)



S







2. Methods

2.1. Case file: c, e, a chi-g ba.ed - b c, a chi-g

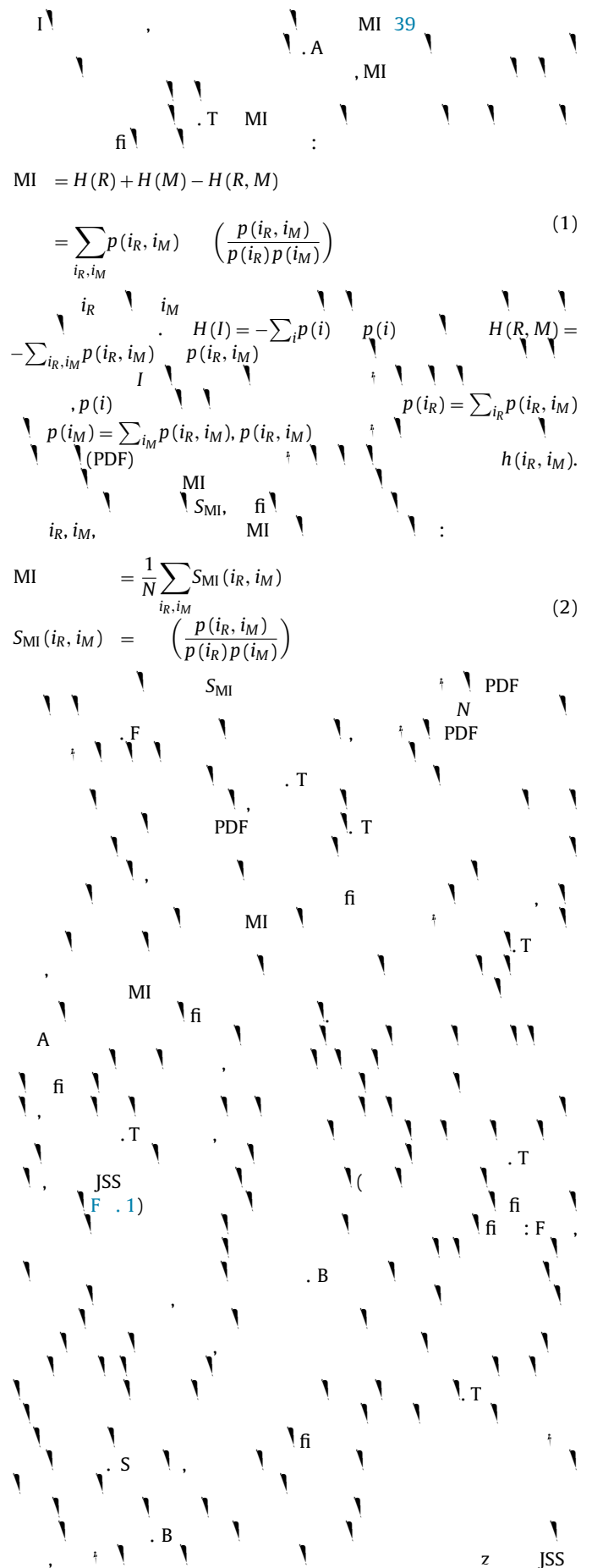
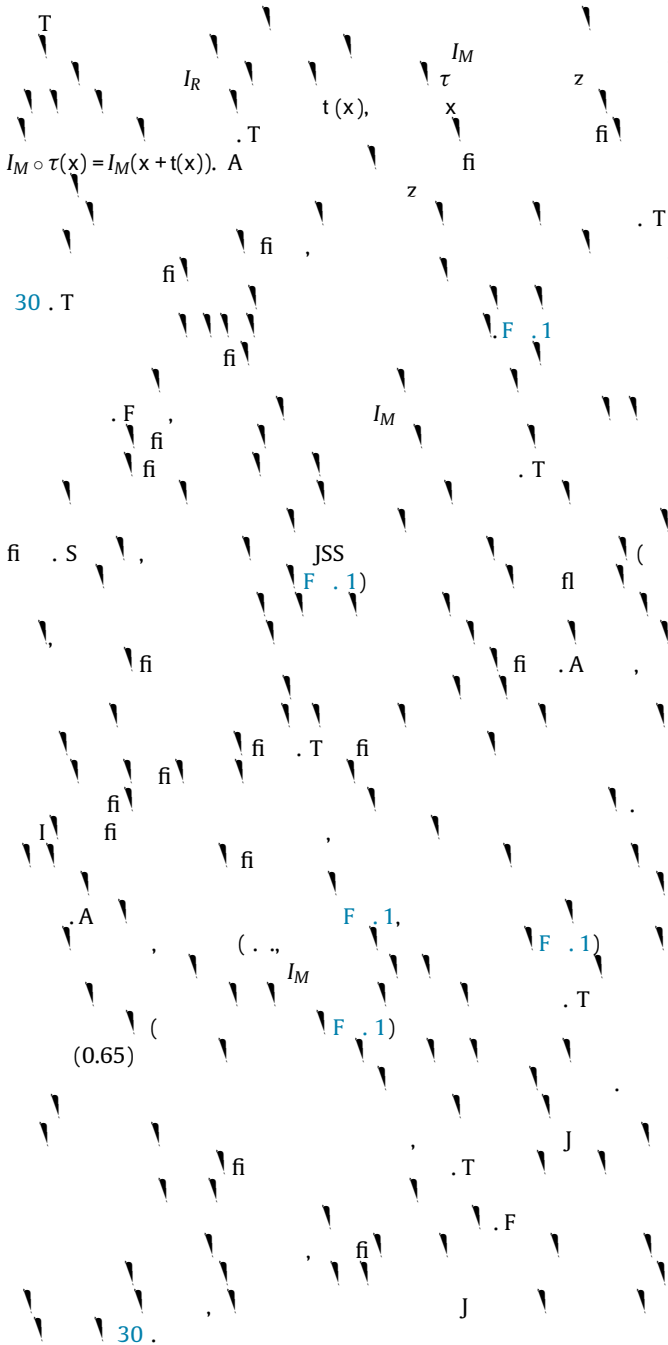
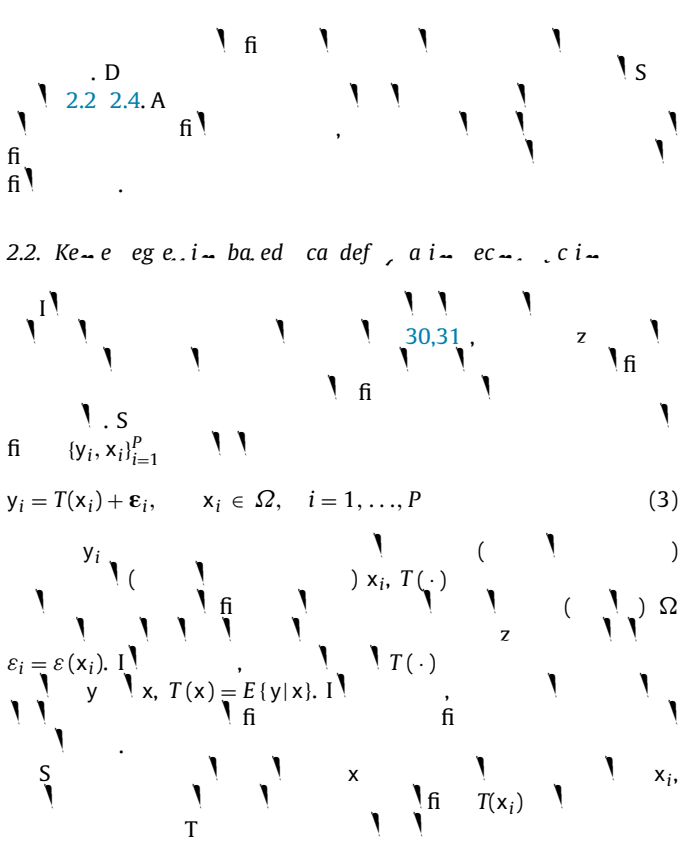


Fig. 1. Fuzzy inference process for the fuzzy rule-based system.



$$T(x_i) \approx T(x) + \{\nabla T(x)\}^T(x_i - x) + \frac{1}{2}(x_i - x)^T \{H(x)\} T(x) + \dots \approx \beta_0 + \beta_1^T(x_i - x) + \beta_2^T \{(x_i - x)(x_i - x)^T\} + \dots \quad (4)$$

$$\{\beta_0, \beta_1, \beta_2, \dots, \beta_N\} \in \mathbb{R}^{N+1}, \quad x = [x_1, x_2]^T$$

$$\beta_0 = T(x)$$

$$\beta_1 = \left[ \frac{\partial T(x)}{\partial x_1}, \frac{\partial T(x)}{\partial x_2} \right]^T \quad (5)$$

$$\beta_2 = \frac{1}{2} \left[ \frac{\partial^2 T(x)}{\partial x_1^2}, \frac{\partial^2 T(x)}{\partial x_1 \partial x_2}, \frac{\partial^2 T(x)}{\partial x_2^2} \right]^T$$

...

$$\{\beta_n\}_{n=0}^N$$

$$\sum_{i=1}^P y_i - \beta_0 - \beta_1^T(x_i - x) - \dots - \beta_2^T K_H(x_i - x) \quad (6)$$

$$K_H(\cdot)$$

$$y = y_1, y_2, \dots, y_p^T, \quad b = \begin{pmatrix} K_H(x_1 - x) \\ K_H(x_2 - x) \\ \vdots \end{pmatrix}$$

$$b = (y - Xb)^T K (y - Xb) \quad (7)$$

$$X = \begin{pmatrix} 1 & (x_1 - x) & T_{\{(x_1 - x)(x_1 - x)^T\}} & \dots \\ 1 & (x_2 - x) & T_{\{(x_2 - x)(x_2 - x)^T\}} & \dots \\ \vdots & \vdots & \vdots & \vdots \\ 1 & (x_p - x) & T_{\{(x_p - x)(x_p - x)^T\}} & \dots \end{pmatrix}$$

$$b = (X^T K X)^{-1} X^T K y \quad (8)$$

$$T(x) = \beta_0 = \frac{\sum_{i=1}^P K_H(x_i - x) y_i}{\sum_{i=1}^P K_H(x_i - x)}$$

$$T(x) = \beta_0 = \frac{\sum_{i=1}^P K_H(x_i - x) \cdot (y_i \cdot c_i)}{\sum_{i=1}^P K_H(x_i - x) \cdot c_i} \quad (9)$$

$$T(x) = \beta_0 = \frac{\sum_{i=1}^P K_H(x_i - x) \cdot (y_i \cdot c_i)}{\sum_{i=1}^P K_H(x_i - x) \cdot c_i} = \frac{K \otimes (y \cdot c)}{K \otimes c} \quad (10)$$

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2.3. Local coefficients

$$A = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad B = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad C = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad D = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad E = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad F = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad G = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}, \quad H = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}$$

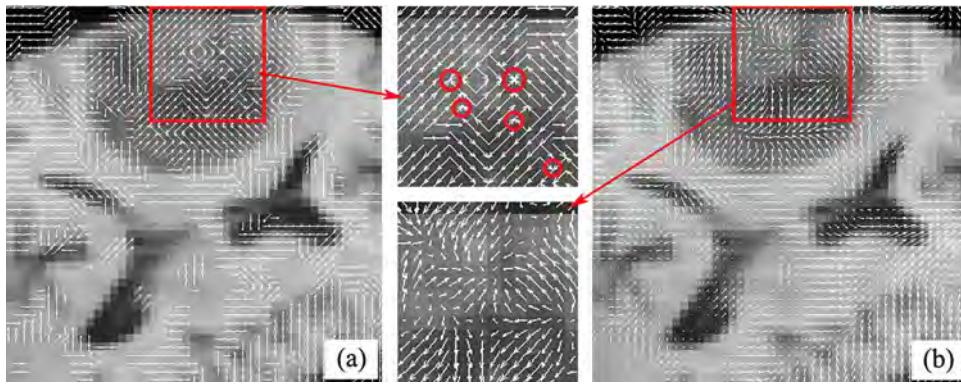


Fig. 2. A

$\{ \sigma_u, \sigma_v \}$   
 $G$   
 $2D$

$$a(\mathbf{x}, \mathbf{x}_0) = \frac{1}{2\pi\sigma_u\sigma_v} \left[ -\left( \frac{d_u^2}{2\sigma_u} + \frac{d_v^2}{2\sigma_v} \right) \right] \quad (13)$$

$d_u = \langle d, u \rangle, \quad d_v = \langle d, v \rangle, \quad d = \mathbf{x} - \mathbf{x}_0$

$\{ \sigma_u, \sigma_v \}$   
 $A$   
 $G$

$$\sigma_u = \frac{\alpha}{\alpha + A} \sigma_c, \quad \sigma_v = \frac{\alpha + A}{\alpha} \sigma_c \quad (14)$$

$A = (\lambda_u - \lambda_v) / (\lambda_u + \lambda_v)$   
 $\alpha > 0$

$\sigma_c$   
 $\sigma_c$

$\alpha = 0.5$   
 $\sigma_c = 1.5$   
 $z = 3 \times 3$

$T$

$F.3,$

$F.3,$

$F.3()$   
 $F.3()$

$F.4$

$T$

$(\sigma_u, \sigma_v)$   
 $5$

$T$

$F.4()$

$F.4().C$

$F.4()$

$F.4()$   
 $F.4()$

$F.4()$   
 $F.4()$

$F.4()$   
 $F.4()$

$F.4()$   
 $F.4()$

$(\sigma_u, \sigma_v)$   
 $G$   
 $2D$

$$H$$

$\{ \sigma_u, \sigma_v \}$   
 $A$   
 $G$

$F.4()$

$2.4. R$

$1,$   
 $\sigma_c(x),$   
 $0$

$z$

$28,29$   
 $S$   
 $z$

$30$

$T$

$S$   
 $z$

$T$

$T$

$T$

$JSM$   
 $JSS$

$T$

$JSS$

$T$

$T$

$T$

$T$

$T$

$T$

Fig. 3. G

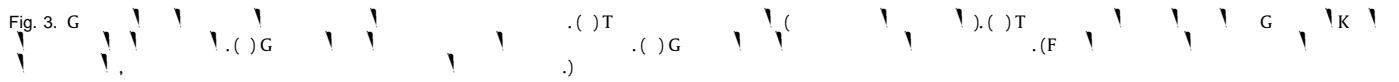
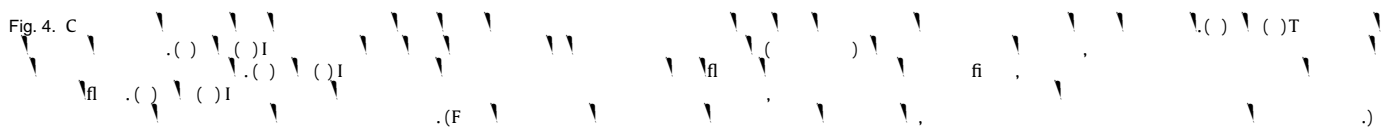


Fig. 4. C



$$S(x_0) = \text{avg} \sum_{x \in \Omega} \|LST(x) - LST(x_0)\|_D \quad (15)$$

$$\|T_1 - T_2\|_D = \sqrt{\frac{8\pi}{15} (\|T_1 - T_2\|_C^2 - \frac{1}{3} \text{Tr}^2(T_1 - T_2))} \quad (16)$$

$$\|T_1 - T_2\|_C = \sqrt{\text{Tr}(T_1 - T_2)^2} / \{\text{Tr}(T_1), \text{Tr}(T_2)\}$$

Fig. 5. JSM E

$$JS(x_R, x_M) = \frac{A \cdot B}{B + \|LST(x_R) - LST(x_M)\|_D} \quad (17)$$

$$\{S_R(\cdot), S_M(\cdot)\} \quad A=10 \quad B=\frac{1}{2}$$

$$LST(x_R) \quad LST(x_M)$$

Fig. 5

Fig. 5. JSM E



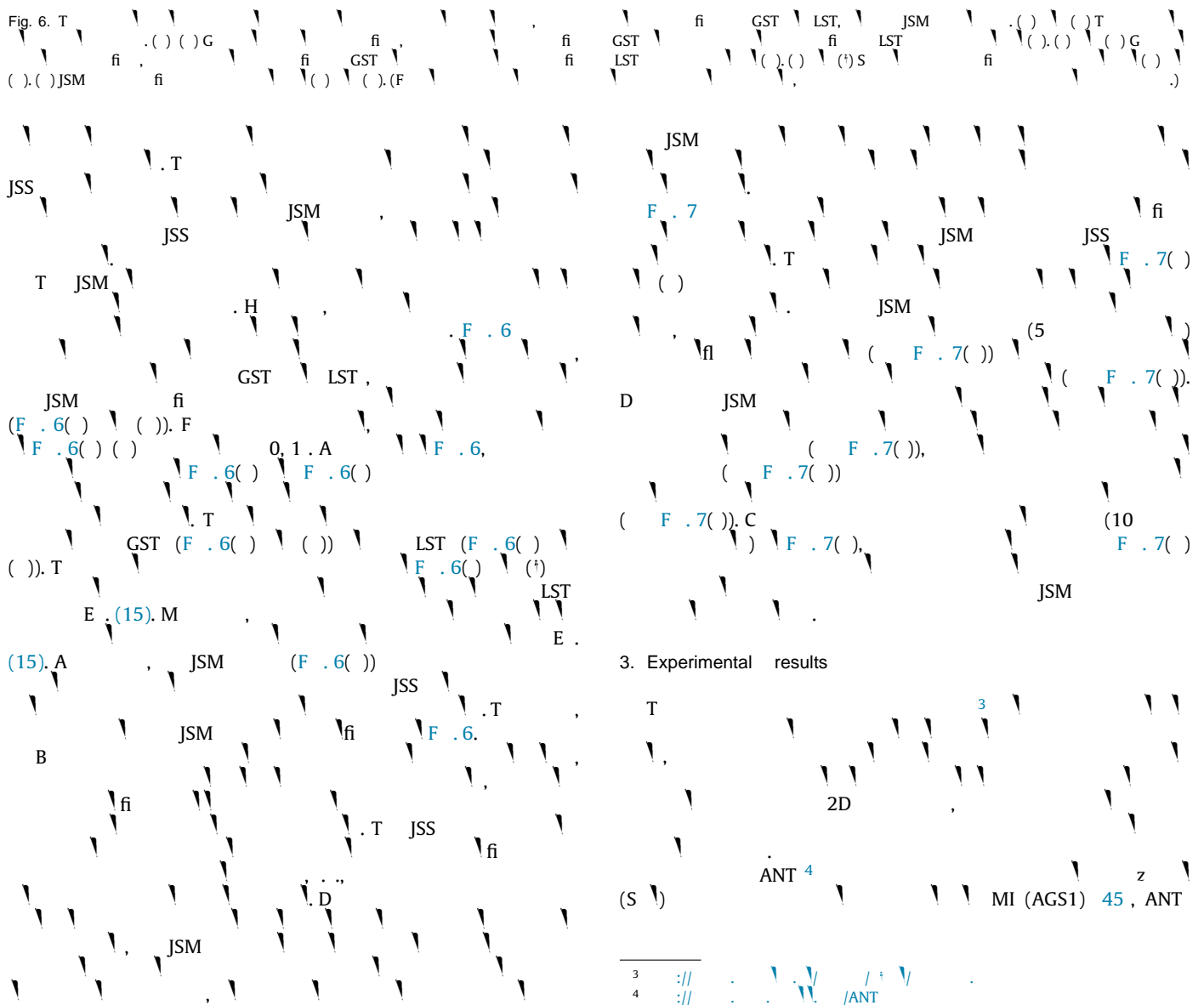


Fig. 7. M

0.571Ti524prea1\_0 1converg1 Tf 0.0002 0 0 -0.0002 194.6897 533.5539 Tm212.354/GS24prea1\_0 1 Tf 6.3761 0 0 6.3761 182.0849 533.5539 Tm215539 67524

Fig. 8. Comparison of the edge detection results of the proposed method with other methods. (a) Original image, (b) AGS1, (c) AGS2, (d) AMI, (e) DDD, (f) BMI, (g) AMM, (h) EPPM, (i) LDOF, (j) F-NL. (F-NL is the method proposed in [10]).

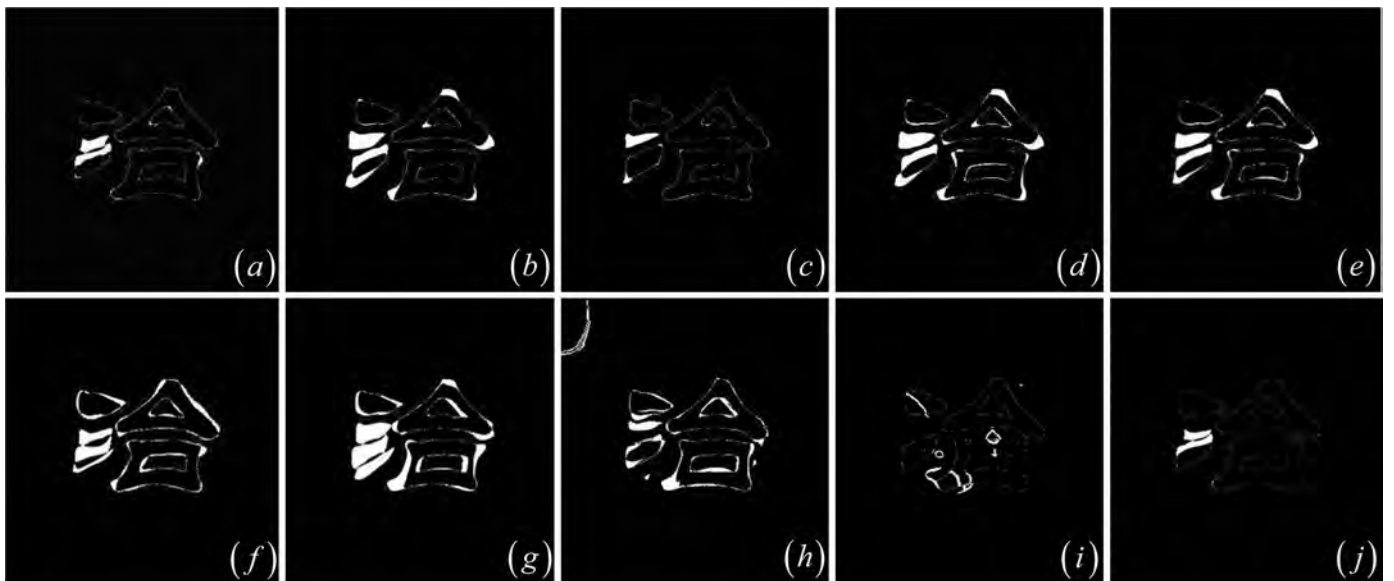


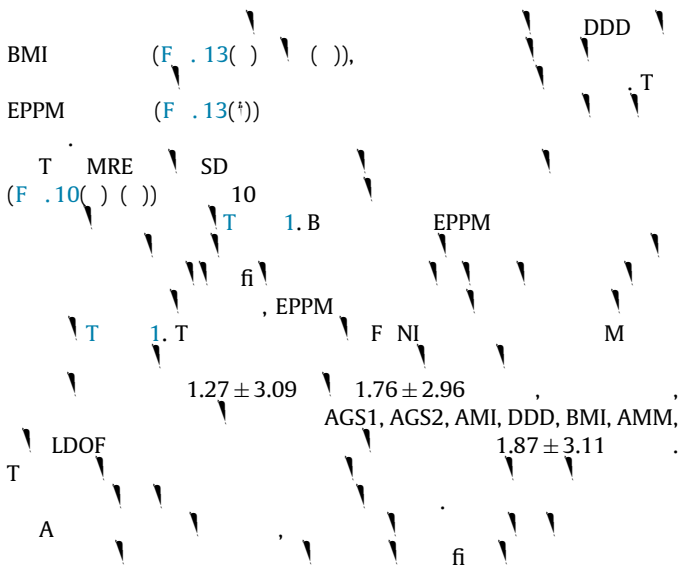
Fig. 9. Comparison of the edge detection results of the proposed method with other methods. (a) Original image, (b) AGS1, (c) AGS2, (d) AMI, (e) DDD, (f) BMI, (g) AMM, (h) EPPM, (i) LDOF, (j) F-NL. (F-NL is the method proposed in [10]).



Fig. 11. M... T... F NI... ( ) T... ( ) AGS1, ( ) AGS2, ( ) AMI, ( ) DDD, ( ) BMI, ( ) AMM, ( ) EPPM, ( ) LDOF, ( ) F NI. (F...)



Fig. 12. B ( ) T AMI AGS2, ( ) AMI, ( ) DDD, ( ) BMI, ( ) AMM, (†) EPPM, ( ) LDOF, ( ) F NI. (F ( ) ( ) T ( ) ( ) AGS1, ( )



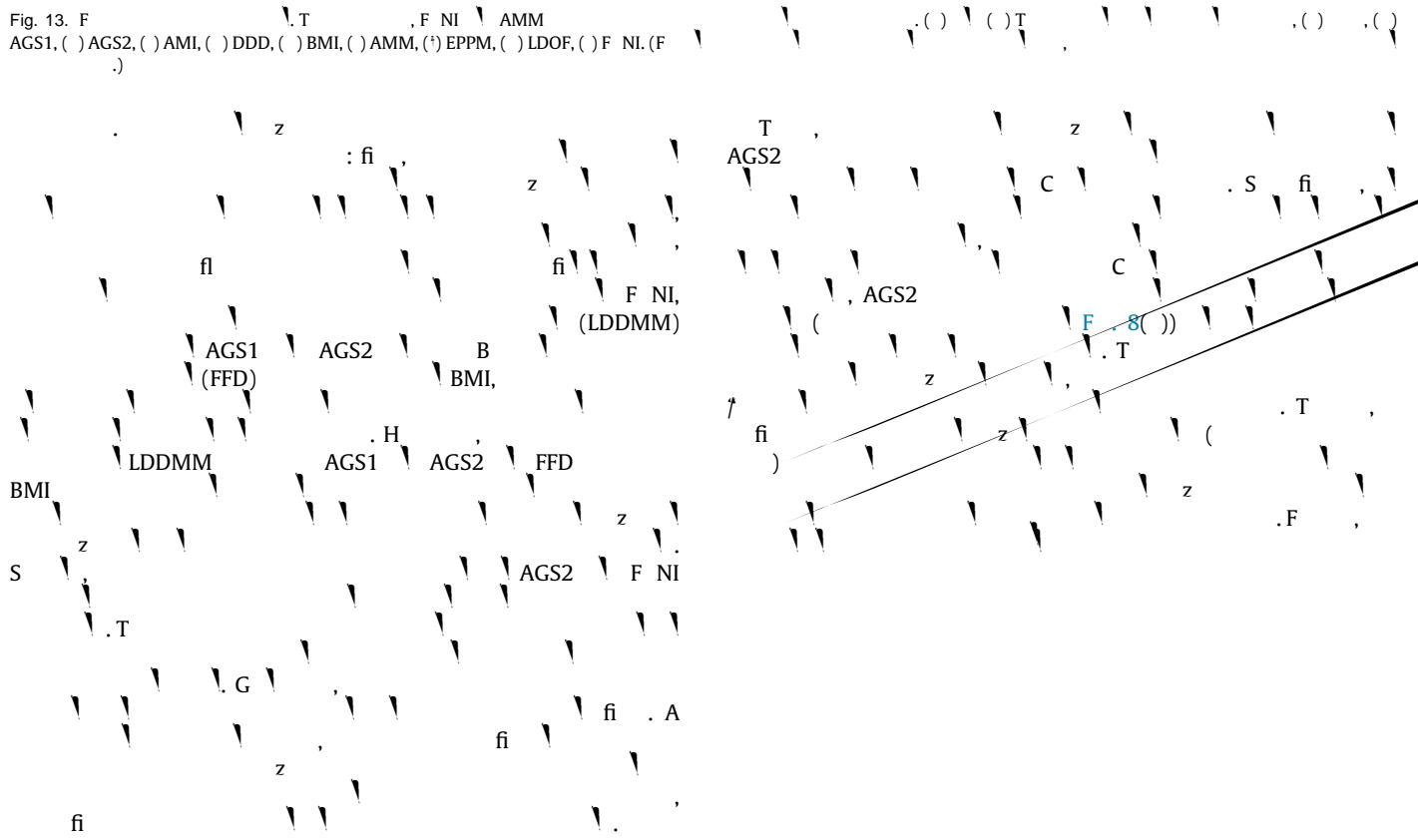


Table 2  
Comparison of the proposed method with other methods.

Method	Time (s)	AGS1	AGS2	AMI	DDD	BMI	AMM	EPPM	LDOF	FN
1	17.67	3563	30.67	13.99	4.42	11.41	13.99	0.79	41.23	4.02
2	18.83	3521.5	54.64	21.52	4.99	16.14	17.29	0.90	55	2.99
3	35.86	3761.5	33.88	15.69	8.97	32	14.94	0.90	84.43	4.53
4	14.67	3271.3	26.16	8.56	4.08	30.57	2.85	0.74	68.48	2.82

The proposed method is implemented on a PC with the following configuration: Intel Core i5-4460Q CPU (3.2 GHz), 4 GB RAM, and Windows 7 operating system. The proposed method is compared with other methods in terms of execution time, accuracy, and the number of iterations. The proposed method shows superior performance in terms of execution time and accuracy compared to other methods. The proposed method also shows a significant reduction in the number of iterations compared to other methods. The proposed method is implemented on a PC with the following configuration: Intel Core i5-4460Q CPU (3.2 GHz), 4 GB RAM, and Windows 7 operating system.

4. Conclusion and discussion

In this paper, a new method for solving the problem of image denoising is proposed. The proposed method is based on the combination of the AMM and the LDOF. The proposed method shows superior performance in terms of execution time and accuracy compared to other methods. The proposed method also shows a significant reduction in the number of iterations compared to other methods. The proposed method is implemented on a PC with the following configuration: Intel Core i5-4460Q CPU (3.2 GHz), 4 GB RAM, and Windows 7 operating system. The proposed method is compared with other methods in terms of execution time, accuracy, and the number of iterations. The proposed method shows superior performance in terms of execution time and accuracy compared to other methods. The proposed method also shows a significant reduction in the number of iterations compared to other methods. The proposed method is implemented on a PC with the following configuration: Intel Core i5-4460Q CPU (3.2 GHz), 4 GB RAM, and Windows 7 operating system.

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References

- A.S., C.D., N.P., D. IEEE T. M. I. 32 (7) (2013) 1153–1190.
- T.B., J.M., L. IEEE T. P. A. M. I. 33 (3) (2011) 500–513.
- H.C., A.R., I.A. IEEE T. M. I. 32 (3) (2003) 114–141.
- W., E.D., A.A.H., R., J.G., H.L., M.G.M., D.H.B., J.C.G., L. IEEE T. M. I. 33 (2014) 903–913.
- W., J.P., J.S., I.B., E.O. IEEE T. M. I. 32 (2) (2014) S2.
- A.M., S.P. IEEE T. M. I. 32 (2010) 2262–2275.
- S.S., G.K., P.G. IEEE T. M. I. 32 (2015) 368–379.
- Q.S., J., R. IEEE T. M. I. 34 (2013) 1557–1565.
- W.O., C.D., D. DRAMMS: D. IPMI 2009 LNCS, 5636, 2009, 50–62.

<sup>10</sup> // ...  
<sup>11</sup> // ...



10 S. P. ... H. F. ... M. ... M. ...  
 I. A. ... 10 (2006) 452–464.

11 J. M. ... J.C. ... C. ... F. C. ... A. ...  
 E. O. ... R. S. ... 5 (2012) 110–124.  
 IEEE J. S. T. A. ...

12 B. ... M. S. ... R. D. ... B. F. ... P. G. ... E. ...  
 (2008) 603–615.  
 M. I. A. ... 12

13 R. S. ... O. C. ... G. M. ... P. ... B. ... N. A. ... P. ...  
 B. ... N. ... R. ... A. ... S. ... R. ... P. ... C. ... P. ...  
 MICCAI 2004 LNCS, ... 3216, 2004, ... 704–711.

14 F.F. B. ... A.N.T.J. K. ... A.A.C. ... L. ... I.M. J. ... S. ... M.A.  
 J.P. ... P. ... R. ... MRI

15 P. R. ... E. S. ... I. T. ... M. ... B. ... 59 (2014) 4033–4045.  
 A. ... R. ... R. ... IPMI 2009 LNCS, ... 5636, 2009,  
 ... 447–458.

16 L. T. ... G. H. ... R. A. ... R. ... D. ... S. ... A.  
 R. ... D. ... R. ... BIR 2010 LNCS, ... 6204, 2010,  
 ... 173–185.

17 N. C. ... K.P. ... J.S. D. ... R. ... B. ... R. ... MRI  
 I. ... L. ... P. ... : 2011 IEEE I ... S  
 B. ... I. ... F. ... N. ... M. ... 2011, ... 1520–1523.

18 I.J.A.S. ... J.A.S. ... A.R.G. ... J.L.R.A. ... M. ... N. ... I. ... P. ...  
 (2012) 2438–2451.

19 M.F. ... B.D. ... L.G. ... S.C. ... E.C. ... N.S. ... S.T. ... J.R. ...  
 S. J. ... L. ... 3D

20 S. G. ... L. ... H. ... R. ... C. ... D.D. K. ... R. M. ... L. D. ... A.  
 M. P. ... 33 (9) (2006) 3304–3312.

21 M. B. ... C. ... M. D. C. ... D. ... B. M. ... C. P. ... J. P. T. ... D. ... MR  
 C. ... M. ... P. ... B. ... 84 (2006) 66–75.

22 E.I. ... C.S. H. ... D. S. ... G. B. ... C. D. ... z. ... N. ...  
 N. ... I. ... 46 (2009) 762–774.

23 A.M. ... S.B. ... A.T. ... T.B. z. ... C. ... M.  
 I. ... : 2010: I ... P. ... P. ... SPIE 7623 (2010), 76230C 1–12.

24 M. S. ... D. P. ... N. ... H. ... B. ... M. ...  
 30 (4) (2009) 1060–1067.

25 M. B. ... A.P. L. ... C. R. ... J. A. ... S. ... N. ... I. ... z. ... 14 (2) (2001)  
 486–500.

26 S.M.A. ... S. ... R. ... P.M.B. ... C. ... N. ... I. ... 53 (1) (2010)  
 z. ... 78–84.

27 P. R. ... J. M. ... P. ... R. ... D. ... B. ... J. M. ... M. F. ... M.  
 J. ... F. R. ... A. R. ... z. F. ... A. ... N. ... I. ... 60 (2) (2012)  
 1296–1306.

28 K. ... A. F. ... K. E. ... z. ... J. A. ... F. ...  
 I. ... J.C. ...