

# Learning Accurate Representation to Nonstandard Tensors via a Mode-Aware Tucker Network

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## I. INTRODUCTION

**T**HIS is the supplementary file for paper entitled *Learning Accurate Representation to Nonstandard Tensors via a Mode-Aware Tucker Network*. Supplementary equations and experimental results are put into this file.

## II. SUPPLEMENTARY PROOF OF CONVERGENCE

First, we review the formula for MSGD and rewrite it to separate the deterministic gradient part from the random noise part.

$$\begin{aligned} v_{n+1} &= \gamma v_n + \eta \nabla_{\omega_n} J(\omega_n) + \eta (\nabla_{\omega_n} J(\omega_n, \xi_n) - \nabla_{\omega_n} J(\omega_n)), \\ \omega_{n+1} &= \omega_n - v_{n+1}. \end{aligned} \quad (S1)$$

This transformation borrows the idea of random approximation. Then calculate the following Taylor expansion and do some processing to get:

$$J(\omega_{n+1}) - J(\omega_n) = -\nabla_{\omega} J(\omega_n)^T v_n + \frac{1}{2} v_n^T H_{\omega\omega}(\zeta_n) v_n. \quad (S2)$$

Using the above Taylor formula, we can make the following calculation:

$$\begin{aligned} &\nabla_{\omega_n} J(\omega_n)^T v_n \\ &= (\nabla_{\omega_{n-1}-v_{n-1}} J(\omega_{n-1} - v_{n-1}))^T (\gamma v_{n-1} + \eta_n \nabla_{\omega_n} J(\omega_n, \xi_n)) \\ &= \gamma (\nabla_{\omega_{n-1}} J(\omega_{n-1}) - H_{\omega\omega}(\zeta_{n-1}) v_{n-1})^T v_{n-1} + \eta_n \nabla_{\omega_n} J(\omega_n) \nabla_{\omega_n} J(\omega_n, \xi_n) \\ &= \gamma \nabla_{\omega_{n-1}} J(\omega_{n-1})^T v_{n-1} - \gamma v_{n-1}^T H_{\omega\omega}(\zeta_{n-1}) v_{n-1} + \eta_n \nabla_{\omega_n} J(\omega_n) \nabla_{\omega_n} J(\omega_n, \xi_n). \end{aligned} \quad (S3)$$

Next, processing the above recursive formula, we get:

$$\nabla_{\omega_n} J(\omega_n)^T v_n = \gamma \nabla_{\omega_1} J(\omega_1)^T v_1 - \sum_{i=1}^{n-1} \gamma v_i^T H_{\omega\omega}(\zeta_i) v_i + \sum_{i=2}^t \gamma \eta_i \nabla_{\omega_i} J(\omega_i) \nabla_{\omega_i} J(\omega_i, \xi_i). \quad (S4)$$

Then, bring (S4) back to (S2),

$$J(\omega_{n+1}) - J(\omega_n) = -\gamma \nabla_{\omega_1} J(\omega_1)^T v_1 - \sum_{i=2}^t \gamma \eta_i \nabla_{\omega_i} J(\omega_i) \nabla_{\omega_i} J(\omega_i, \xi_i) + \sum_{i=1}^{n-1} \gamma v_i^T H_{\omega\omega}(\zeta_i) v_i + \frac{1}{2} v_n^T H_{\omega\omega}(\zeta_n) v_n. \quad (S5)$$

Next, we will show that  $J(\omega_n + 1)$  is almost certainly convergent. We do this by recursing on (S5) and dividing it into three parts:

$$\begin{aligned}
J(\omega_{t+1}) = & \underbrace{J(\omega_1) - \frac{\gamma - \gamma^{t+1}}{1 - \gamma} \nabla_{\omega_1} J(\omega_1)^T v_1 + \frac{1 - \gamma^t}{1 - \gamma} \eta_1 \nabla_{\omega_1} J(\omega_1) \nabla_{\omega_1} J(\omega_1, \xi_1)}_A \\
& - \underbrace{\sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \nabla_{\omega_n} J(\omega_n)^T \nabla_{\omega_n} J(\omega_n, \xi_n)}_B \\
& - \underbrace{\frac{1}{2} \sum_{n=1}^t v_n^T H_{\omega\omega}(\zeta_n) v_n + \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} v_n^T H_{\omega\omega}(\zeta_n) v_n}_C.
\end{aligned} \tag{S6}$$

Because  $a < 1$ ,  $a$  is convergent, which ensures that part A is convergent. Then, Lemma 3 ensures that part C converges almost everywhere. For Part B, we have the following proof:

$$\begin{aligned}
B &= \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \nabla_{\omega_n} J(\omega_n)^T \nabla_{\omega_n} J(\omega_n, \xi_n) \\
&= \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \|\nabla_{\omega_n} J(\omega_n)\|^2 \\
&+ \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \nabla_{\omega_n} J(\omega_n)^T (\nabla_{\omega_n} J(\omega_n, \xi_n) - \nabla_{\omega_n} J(\omega_n, \xi_n)).
\end{aligned} \tag{S7}$$

By Lemma 4, we have:

$$\sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \|\nabla_{\omega_n} J(\omega_n)\|^2 < +\infty. \tag{S8}$$

According to Lemma 1 and Lemma 4, we can conclude

$$\sum_{n=1}^{t-1} \frac{1 - \gamma^{t-n}}{1 - \gamma} \eta_{n+1} \nabla_{\omega_n} J(\omega_n)^T (\nabla_{\omega_n} J(\omega_n, \xi_n) - \nabla_{\omega_n} J(\omega_n, \xi_n)). \tag{S9}$$

The above formula is convergent. So B is convergent, and  $g(\omega_t + 1)$  is also convergent. Substituting (S7) into (S6) we get:

$$J(\omega_{t+1}) = \zeta_t - \sum_{n=1}^t \eta_n \|\nabla_{\omega_n} J(\omega_n)\|^2, \tag{S10}$$

where  $\{\zeta_t\}$  is defined as follows:

$$\begin{aligned}
\zeta_t = & J(\omega_1) - \frac{\gamma - \gamma^{t+1}}{1 - \gamma} \nabla_{\omega_1} J(\omega_1)^T v_1 + \frac{1 - \gamma^t}{1 - \gamma} \eta_1 \nabla_{\omega_1} J(\omega_1) \nabla_{\omega_1} J(\omega_1, \xi_1) \\
& + \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} \eta_n \nabla_{\omega_n} J(\omega_n)^T (\nabla_{\omega_n} J(\omega_n, \xi_n) - \nabla_{\omega_n} J(\omega_n)) \\
& - \frac{1}{2} \sum_{n=1}^t v_n^T H_{\omega\omega}(\zeta_n) v_n + \sum_{n=1}^t \frac{1 - \gamma^{t-n+1}}{1 - \gamma} v_n^T H_{\omega\omega}(\zeta_n) v_n.
\end{aligned} \tag{S11}$$

Since  $\{\zeta_t\}$  almost certainly converges, Lemma 2 shows that,

$$\omega_t \rightarrow \omega^*. \tag{S12}$$

## III. SUPPLEMENTARY TABLE

Here are some supplementary tables in the Experiments section.

TABLE S1  
HYPER-PARAMETERS SETTINGS OF M1-10 ON D1-8

Datasets	Hyper-parameter settings				
D1	M1: Adaptive M6: $\lambda_1=2^{-10}$ , $\lambda_2=2^{-9}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-6}$ M7: $\lambda_1=2^{-6}$ , $\lambda_2=2^{-2}$	M3: $\eta=2^{-10}$ , $\lambda=2^{-8}$ M8: $\lambda_1=2^{-3}$ , $\lambda_2=2^{-4}$	M4: $\eta=2^{-7}$ , $\lambda=2^{-8}$ M9: $\eta=2^{-9}$ , $\lambda=2^{-10}$	M5: $P=1.9$ , $\lambda=2^{-6}$ M10: $\alpha=2^{-0}$ , $\beta=2^{-8}$ , $\lambda=2^{-7}$
D2	M1: Adaptive M6: $\lambda_1=2^{-8}$ , $\lambda_2=2^{-11}$	M2: $\eta=2^{-6}$ , $\lambda=2^{-6}$ M7: $\lambda_1=2^{-3}$ , $\lambda_2=2^{-2}$	M3: $\eta=2^{-10}$ , $\lambda=2^{-10}$ M8: $\lambda_1=2^{-1}$ , $\lambda_2=2^{-3}$	M4: $\eta=2^{-8}$ , $\lambda=2^{-6}$ M9: $\eta=2^{-8}$ , $\lambda=2^{-6}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-0}$ , $\beta=2^{-9}$ , $\lambda=2^{-6}$
D3	M1: Adaptive M6: $\lambda_1=2^{-7}$ , $\lambda_2=2^{-10}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-6}$ M7: $\lambda_1=2^{-7}$ , $\lambda_2=2^{-1}$	M3: $\eta=2^{-9}$ , $\lambda=2^{-12}$ M8: $\lambda_1=2^{-9}$ , $\lambda_2=2^{-3}$	M4: $\eta=2^{-7}$ , $\lambda=2^{-8}$ M9: $\eta=2^{-8}$ , $\lambda=2^{-2}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-1}$ , $\beta=2^{-7}$ , $\lambda=2^{-6}$
D4	M1: Adaptive M6: $\lambda_1=2^{-6}$ , $\lambda_2=2^{-8}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-6}$ M7: $\lambda_1=2^{-5}$ , $\lambda_2=2^{-2}$	M3: $\eta=2^{-9}$ , $\lambda=2^{-12}$ M8: $\lambda_1=2^{-0}$ , $\lambda_2=2^{-4}$	M4: $\eta=2^{-8}$ , $\lambda=2^{-5}$ M9: $\eta=2^{-7}$ , $\lambda=2^{-7}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-0}$ , $\beta=2^{-4}$ , $\lambda=2^{-5}$
D5	M1: Adaptive M6: $\lambda_1=2^{-7}$ , $\lambda_2=2^{-9}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-6}$ M7: $\lambda_1=2^{-12}$ , $\lambda_2=2^{-6}$	M3: $\eta=2^{-10}$ , $\lambda=2^{-11}$ M8: $\lambda_1=2^{-2}$ , $\lambda_2=2^{-4}$	M4: $\eta=2^{-6}$ , $\lambda=2^{-9}$ M9: $\eta=2^{-9}$ , $\lambda=2^{-11}$	M5: $P=1.7$ , $\lambda=2^{-6}$ M10: $\alpha=2^{-0}$ , $\beta=2^{-3}$ , $\lambda=2^{-8}$
D6	M1: Adaptive M6: $\lambda_1=2^{-8}$ , $\lambda_2=2^{-11}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-7}$ M7: $\lambda_1=2^{-9}$ , $\lambda_2=2^{-6}$	M3: $\eta=2^{-9}$ , $\lambda=2^{-12}$ M8: $\lambda_1=2^{-1}$ , $\lambda_2=2^{-4}$	M4: $\eta=2^{-5}$ , $\lambda=2^{-10}$ M9: $\eta=2^{-9}$ , $\lambda=2^{-10}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-2}$ , $\beta=2^{-1}$ , $\lambda=2^{-8}$
D7	M1: Adaptive M6: $\lambda_1=2^{-9}$ , $\lambda_2=2^{-12}$	M2: $\eta=2^{-10}$ , $\lambda=2^{-7}$ M7: $\lambda_1=2^{-12}$ , $\lambda_2=2^{-6}$	M3: $\eta=2^{-10}$ , $\lambda=2^{-12}$ M8: $\lambda_1=2^{-0}$ , $\lambda_2=2^{-5}$	M4: $\eta=2^{-5}$ , $\lambda=2^{-7}$ M9: $\eta=2^{-9}$ , $\lambda=2^{-9}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-2}$ , $\beta=2^{-2}$ , $\lambda=2^{-8}$
D8	M1: Adaptive M6: $\lambda_1=2^{-8}$ , $\lambda_2=2^{-11}$	M2: $\eta=2^{-7}$ , $\lambda=2^{-7}$ M7: $\lambda_1=2^{-5}$ , $\lambda_2=2^{-6}$	M3: $\eta=2^{-8}$ , $\lambda=2^{-12}$ M8: $\lambda_1=2^{-0}$ , $\lambda_2=2^{-4}$	M4: $\eta=2^{-5}$ , $\lambda=2^{-9}$ M9: $\eta=2^{-8}$ , $\lambda=2^{-10}$	M5: $P=1.9$ , $\lambda=2^{-5}$ M10: $\alpha=2^{-2}$ , $\beta=2^{-3}$ , $\lambda=2^{-11}$

TABLE S2

TOTAL TIME OF ALL TEST MODELS ON D1-8. (UNIT: MINUTES. NOTE THAT THE TIME INCLUDES THE TIME TO ADJUST THE HYPER-PARAMETERS.)

Models	D1	D2	D3	D4	D5	D6	D7	D8	Win/Loss
Time cost in RMSE↓									
M1	5.8 $\pm$ 3.3E-01	<b>3.3</b> $\pm$ 6.2E-01	2.0 $\pm$ 2.4E-01	<b>0.7</b> $\pm$ 6.1E-02	28.0 $\pm$ 3.1E-00	16.4 $\pm$ 7.2E-01	8.1 $\pm$ 4.5E-01	5.6 $\pm$ 1.0E-00	—
M2	308.1 $\pm$ 1.3E+01	71.7 $\pm$ 1.0E+00	99.0 $\pm$ 2.5E+00	32.0 $\pm$ 1.0E+00	74.2 $\pm$ 1.1E+00	74.4 $\pm$ 1.9E+01	295.0 $\pm$ 2.6E+00	22.3 $\pm$ 7.4E-01	8/0
M3	306.7 $\pm$ 2.1E+01	82.7 $\pm$ 3.7E-01	23.7 $\pm$ 1.7E-01	11.0 $\pm$ 2.1E-01	28.8 $\pm$ 1.5E-00	<b>13.3</b> $\pm$ 7.3E-02	18.6 $\pm$ 1.7E-01	1.8 $\pm$ 5.4E-02	6/2
M4	50.5 $\pm$ 4.5E-00	80.6 $\pm$ 5.6E-00	22.8 $\pm$ 5.6E-01	20.5 $\pm$ 3.9E-01	48.9 $\pm$ 7.8E-01	38.1 $\pm$ 2.1E-00	5.2 $\pm$ 1.8E-00	11.1 $\pm$ 5.3E-01	7/1
M5	40.3 $\pm$ 5.2E-00	43.0 $\pm$ 4.7E-00	33.2 $\pm$ 3.2E-01	11.5 $\pm$ 4.8E-01	267.4 $\pm$ 6.3E-00	125.2 $\pm$ 1.0E-01	115.5 $\pm$ 5.6E-01	86.1 $\pm$ 6.5E-00	8/0
M6	<b>2.2</b> $\pm$ 6.8E-01	5.9 $\pm$ 6.4E-01	<b>1.3</b> $\pm$ 4.8E-02	0.8 $\pm$ 7.2E-02	<b>3.4</b> $\pm$ 2.7E-02	17.6 $\pm$ 3.4E-00	38.6 $\pm$ 2.2E-01	10.5 $\pm$ 3.4E-01	5/3
M7	67.4 $\pm$ 1.4E+01	56.2 $\pm$ 6.4E-00	17.4 $\pm$ 6.8E-00	20.3 $\pm$ 5.5E-00	139.9 $\pm$ 1.1E+01	224.2 $\pm$ 6.6E-00	123.9 $\pm$ 8.5E-00	246.6 $\pm$ 2.1E-00	8/0
M8	5.1 $\pm$ 1.3E-00	16.6 $\pm$ 2.6E-00	10.9 $\pm$ 6.0E-02	1.3 $\pm$ 1.9E-01	55.1 $\pm$ 1.2E+01	39.0 $\pm$ 6.4E-00	5.2 $\pm$ 1.3E-01	<b>1.6</b> $\pm$ 1.0E-01	5/3
M9	131.5 $\pm$ 2.8E+01	45.5 $\pm$ 1.2E-00	36.6 $\pm$ 9.3E-01	9.7 $\pm$ 4.3E-01	19.1 $\pm$ 5.1E-01	16.3 $\pm$ 4.8E-01	63.9 $\pm$ 1.6E-00	8.8 $\pm$ 5.9E-01	6/2
M10	222.1 $\pm$ 8.8E-00	202.2 $\pm$ 5.9E-00	42.8 $\pm$ 4.3E-00	1.9 $\pm$ 7.6E-02	127.9 $\pm$ 3.1E-00	109.7 $\pm$ 5.2E-01	95.2 $\pm$ 1.6E-00	32.0 $\pm$ 9.7E-01	8/0
Time cost in MAE↓									
M1	8.1 $\pm$ 2.0E-01	<b>4.1</b> $\pm$ 7.8E-01	2.8 $\pm$ 8.9E-02	<b>0.6</b> $\pm$ 5.8E-02	48.1 $\pm$ 1.2E-00	26.7 $\pm$ 4.2E-00	<b>15.0</b> $\pm$ 2.2E-00	8.3 $\pm$ 1.0E-00	—
M2	515.8 $\pm$ 2.2E+01	133.6 $\pm$ 4.6E+00	190.6 $\pm$ 5.5E+00	34.7 $\pm$ 1.5E+00	252.7 $\pm$ 1.4E+00	192.6 $\pm$ 3.5E+01	294.7 $\pm$ 3.2E+00	46.3 $\pm$ 1.4E+00	8/0
M3	328.2 $\pm$ 2.0E+01	106.2 $\pm$ 1.8E-00	41.9 $\pm$ 4.2E-01	13.2 $\pm$ 1.9E-01	117.3 $\pm$ 2.9E-00	63.5 $\pm$ 1.6E-00	62.7 $\pm$ 1.0E-00	<b>4.0</b> $\pm$ 4.4E-02	7/1
M4	70.2 $\pm$ 6.4E-00	110.8 $\pm$ 4.6E-00	34.0 $\pm$ 1.2E-00	18.4 $\pm$ 4.5E-01	117.4 $\pm$ 2.6E-00	94.1 $\pm$ 3.1E-00	22.9 $\pm$ 7.3E-01	18.7 $\pm$ 6.4E-01	8/0
M5	40.0 $\pm$ 3.7E-00	35.3 $\pm$ 1.9E-00	27.5 $\pm$ 8.1E-01	10.9 $\pm$ 4.8E-01	197.8 $\pm$ 9.0E-00	91.8 $\pm$ 4.7E-00	82.1 $\pm$ 5.5E-01	53.2 $\pm$ 3.6E-00	8/0
M6	<b>2.2</b> $\pm$ 6.8E-01	5.3 $\pm$ 2.6E-02	<b>1.2</b> $\pm$ 1.5E-01	0.9 $\pm$ 7.2E-02	<b>3.5</b> $\pm$ 3.2E-02	17.9 $\pm$ 3.8E-00	36.8 $\pm$ 7.4E-01	10.6 $\pm$ 2.2E-01	4/4
M7	153.2 $\pm$ 3.6E-00	71.2 $\pm$ 2.9E-00	29.7 $\pm$ 1.1E-00	20.3 $\pm$ 2.4E-00	145.6 $\pm$ 1.0E+01	205.9 $\pm$ 1.1E+01	146.9 $\pm$ 1.0E+01	235.0 $\pm$ 5.5E-00	8/0
M8	46.8 $\pm$ 6.6E-00	19.2 $\pm$ 1.1E-00	11.8 $\pm$ 5.0E-02	7.3 $\pm$ 2.7E-01	55.8 $\pm$ 7.5E-00	41.6 $\pm$ 3.1E-00	40.8 $\pm$ 9.0E-00	12.0 $\pm$ 2.7E-00	8/0
M9	131.0 $\pm$ 2.8E+01	52.8 $\pm$ 2.6E-00	50.0 $\pm$ 1.9E-00	11.0 $\pm$ 3.8E-01	19.1 $\pm$ 5.1E-01	<b>16.3</b> $\pm$ 4.8E-01	64.7 $\pm$ 1.4E-00	12.9 $\pm$ 7.1E-01	6/2
M10	389.5 $\pm$ 2.0E+01	242.0 $\pm$ 2.7E-00	66.7 $\pm$ 3.2E-00	3.6 $\pm$ 3.0E-01	127.6 $\pm$ 3.1E-00	109.7 $\pm$ 5.2E-01	95.2 $\pm$ 1.6E-00	32.0 $\pm$ 9.7E-01	8/0
Time cost in R <sup>2</sup> ↓									
M1	5.8 $\pm$ 3.3E-01	<b>3.2</b> $\pm$ 8.2E-01	2.3 $\pm$ 1.5E-01	<b>0.7</b> $\pm$ 6.1E-02	33.0 $\pm$ 3.1E-00	16.6 $\pm$ 5.1E-01	8.1 $\pm$ 3.6E-01	5.6 $\pm$ 1.0E-00	—
M2	366.2 $\pm$ 1.8E+01	78.2 $\pm$ 1.1E+00	123.1 $\pm$ 2.0E+00	34.6 $\pm$ 1.2E+00	78.7 $\pm$ 1.4E+00	83.2 $\pm$ 2.3E+01	294.7 $\pm$ 3.2E+00	22.5 $\pm$ 8.1E-01	8/0
M3	322.1 $\pm$ 2.3E+01	101.6 $\pm$ 1.1E-00	25.5 $\pm$ 2.9E-01	11.6 $\pm$ 2.1E-01	30.6 $\pm$ 1.3E-00	<b>13.9</b> $\pm$ 3.1E-01	20.3 $\pm$ 1.7E-01	1.8 $\pm$ 6.6E-02	5/3
M4	53.0 $\pm$ 3.3E-00	110.7 $\pm$ 6.2E-00	24.5 $\pm$ 5.6E-01	20.7 $\pm$ 3.0E-01	65.8 $\pm$ 7.9E-01	55.0 $\pm$ 2.1E-00	<b>5.1</b> $\pm$ 2.8E-01	18.5 $\pm$ 6.6E-01	7/1
M5	40.3 $\pm$ 1.2E-00	42.1 $\pm$ 3.7E-00	33.6 $\pm$ 6.6E-01	12.0 $\pm$ 9.7E-01	266.5 $\pm$ 4.7E-00	124.7 $\pm$ 8.1E-00	117.3 $\pm$ 3.1E-00	84.4 $\pm$ 5.4E-00	8/0
M6	<b>2.2</b> $\pm$ 6.8E-01	5.9 $\pm$ 6.4E-01	<b>1.3</b> $\pm$ 4.8E-02	0.8 $\pm$ 7.2E-02	<b>3.4</b> $\pm$ 4.4E-02	17.6 $\pm$ 3.4E-00	38.6 $\pm$ 1.8E-01	10.4 $\pm$ 1.7E-01	5/3
M7	72.2 $\pm$ 2.4E+01	73.3 $\pm$ 4.9E-00	21.0 $\pm$ 9.3E-00	23.8 $\pm$ 3.6E-00	139.9 $\pm$ 1.1E+01	229.8 $\pm$ 8.2E-00	131.8 $\pm$ 7.2E-00	249.1 $\pm$ 2.8E-00	8/0
M8	5.1 $\pm$ 1.3E-00	16.6 $\pm$ 2.6E-00	10.9 $\pm$ 6.0E-02	1.3 $\pm$ 1.9E-01	56.9 $\pm$ 1.0E+01	39.0 $\pm$ 6.4E-00	5.2 $\pm$ 6.4E-02	<b>1.6</b> $\pm$ 1.0E-01	5/3
M9	131.5 $\pm$ 2.8E+01	46.8 $\pm$ 1.1E-00	39.3 $\pm$ 3.4E-00	9.9 $\pm$ 3.5E-01	19.1 $\pm$ 5.1E-01	16.3 $\pm$ 4.8E-01	64.7 $\pm$ 1.4E-00	9.3 $\pm$ 5.9E-01	6/2
M10	233.1 $\pm$ 4.3E-01	203.7 $\pm$ 5.7E-00	43.5 $\pm$ 4.7E-00	1.9 $\pm$ 7.6E-02	127.9 $\pm$ 3.1E-00	109.7 $\pm$ 5.2E-01	95.2 $\pm$ 1.6E-00	32.0 $\pm$ 9.7E-01	8/0

TABLE S3

FRIEDMAN TEST RESULTS ON ACCURACY (RMSE, MAE, R<sup>2</sup>) AND EFFICIENCY (TIME COST IN RMSE, TIME COST IN MAE, TIME COST IN R<sup>2</sup>)

	M1	M2	M3	M4	M5	M6	M7	M8	M9	M10
Accuracy	<b>1.00</b>	2.42	4.96	3.04	6.46	8.17	5.67	6.08	7.63	9.58
Efficiency	<b>2.25</b>	8.83	5.21	5.77	6.79	2.38	7.88	3.31	4.88	7.71