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In MapReduce mode, a job is divided into a series of *map tasks* and *reduce tasks*. The execution of the job is coordinated by one or more nodes, which are called *map reducers*. To finish the job, a mapper and a reducer, called *Map Reduce checker*, are placed on each node of the cluster. However, had a *Map Reduce checker* call to detect a connection between the checker and the node of a *checker* (Hadoop homepage <http://hadoop.apache.org/>, 2011; Zaharia et al. 8th USENIX Conference on System Administration, ACM, New York, pp. 29–42, 2008). To observe, the checker is called *Hadoop-based Auto-Tuning (HAT) Map Reduce checker*, which can detect the node of a *checker* and data of the cluster to generate a *checker*. HAT checker is the *checker* of each *checker* of a *checker* and a *checker* of a *checker* to detect the *checker* of a *checker*.

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he acc^t a e e gh of he ha e o ca c a e he vo^tg'e of c^tWe a . Ba ed
 o he acc^t a e-cac a ed vo^tg'e t of a , HAT e a e he^te a t g t e of
 a t acc^t a e a d f^t h e a che bac a f o t he t a t ha ha e t he t o ge t
 t e a g t e. E e t e a e ho ha HAT ca t g t ca t vo e h e
 e f o a ce of Ma Red ce a ca o t 37% co a ed h Hadoo a d
 t o 16% co a ed h LATE ched e .

H -ba ed a o- g . Sched g a g o h . He e oge eo
 e vo e t . Ma Red ce

C We da a e ea ca o eed o voce a g e a d a g e da a e h he
 e o o of f o a o . D e o he a g e da a e of a ca o , e v e f e w o t e
 o e a d o e voce o o co t e oge h e t o e v e h e e c o t e of he
 a ca o v e a o a b e a d a c c e a b e . Th eed ha vo t o ed he de t o e t of
 Ma Red ce, h ch o e of he t o o a vo g a g a d ched g ode
 o voce a d g e e a e a g e da a e t [8]. Ma Red ce e a b e e o e c f a a
 f c o ha voce e a e / a e a t o g e e a e a e of e t e d a e e / a e
 a t , a d a t e d c e f c o ha e g e a t he e t e d a e a e a o c a e d h he
 a e e e d a e e [8]. Ma Red ce t e d t C o d C o g he beg t g
 [2, 3, 7, 23, 24]. I a e d b Goog e, oge h e h GFS [12] a d B g Tab e [4]
 co t g bac bo e of Goog e' C o d C o t g a f o . A a f o he C o d
 C o g a f o , Ma Red ce a o o e d o o GPU a d t voce o .
 I add t o , t a o e t e d e d o o e o e o e co g vo b e [6, 9, 10, 18,
 20, 22, 27].

F o a Ma Red ce ob (.e., a a ca o ha e e e d ba ed o he
 Ma Red ce vo g a g ode), t da a e t d d e d o a t a da a e .
 Whe a Ma Red ce e a t o e e c t e a Ma Red ce ob, he Ma Red ce
 ched e¹ he e t a che a a a t f o each of he a da a e , a d
 a che a g o o f t e d ce a o co e c h e v e of a t he a a t . A t e he
 d o , he Ma Red ce ched e d t b e he e a o o d f f e o d e t a c c o
 d g o he oca o of he a ' da a e . I h t a , a t he ode (ca ed a
 workers) e t e t e he a h ch a e a t g e d o he a a e . S ce e e ode
 eed o e e c t e t a t a Ma Red ce ched e a che a task scheduler f o
 each ode o t a age a t . F v h e o e, ce a Ma Red ce ob o co e e d
 a he da a voce ed co e e , he e c o t e of he Ma Red ce ob
 t d e d e d b he a t h e d a (.e., he e a e t t e f f e c t) .

I o a e o t vo b e t ho oge eo e t vo e t , ce ho oge eo
 ode e e c t e a h he a e da a e e a e . I he e oge eo
 e vo e t , o he o h e ha d, he e c t o t e of a Ma Red ce ob t vo o g e d
 b t he a t h e d a (ca ed a straggler tasks) e o ce o e e v e

¹A Ma Red ce ched e a ched e t h a ched e a a d e d ce a .

ã o e acco h g e e he a e a d e o he d ff e ce , ch a ca ac e of co a o a d co t ca o , a ch ec t e , a d e õ e .

O e of he o t t o a o o of h v ob e Ma Red ce a ch g bac a t fõ t agg e a o fa ode. If a Ma Red ce ched e a che a bac t a g b fõ t a agg e a g , he a da a e of g v o ce ed co e e he e t h e g b õ g t he . I t h ca e , f g b t he befõ e g , t he e ec t o t e of t he ob v ed ced.

A ho gh c We Ma Red ce ched e v o a ch bac a fõ v agg e a t t he fa o de ec v agg e a cõ v ec t d e o he v o g e t a ed t e a - t g t e of a t he a t [13, 28]. The v o g de ec ed v agg e a t ca e a ea t o t v ob e . Fõ t , a ch g bac a fõ t h e e v o g v agg e a ca o t t v o e he e fõ a ce of he Ma Red ce ob t ce he v ea v agg e a v o o g he e ec t o e . Seco d , he bac a h ch a e a ched fõ h e v o g v agg e a t a t e e v e o v ce . The co e v o o he e v e o v ce e e deg ade t he o e a e fõ a ce of he Ma Red ce ob .

The a g of e v e o v ce t o e of he a v ob e of he bac v a - eg . C We t , Ma Red ce ched e ca f t ode o fa ode a d o ode , o ha bac t a ca be a ched o fa ode . Ho e e t , o ode ca be fõ h e ca ed t o map slow nodes a d reduce slow nodes a v e a e , ce t e o be ha a ode v o ce e a a fa b v o ce e v ed ce a o a d ce e a . We e a x ed ce o t ode o t e e e he ode ha e ec e a x ed ce a o ha o of o h e v ode . The d t t g h g b e t e e a o ode a d v ed ce o ode t a e e v e o v ce . Le a e a v ed ce t a g ha eed a bac a fõ e a e . C We t Ma Red ce ched e t o a t ch he bac a o a o ode N s . Ho e e t , f N s o a map slow node , a ch g he bac a of g v o N s ca v e v e o v ce o N s ef ce t a d v o e he o e a e fõ a ce , ce N s ca t v o ce v ed ce a fa .

I õ d e t o de ec v agg e a a d a ch bac a ef ce t , e v o o e a History-based Auto-Tuning (HAT) Ma Red ce ched e . HAT cõ õ a e h õ - ca fõ a o v ecõ ded o each ode o e a a e e a d de ec v agg e a d a ca t . HAT f v h e ca e o t ode o a o ode a d v ed ce o ode . I h a , HAT ca a ch bac a t fõ a v agg e a o v ed ce o ode a d a ch bac a fõ v ed ce v agg e a t o a t o ode .

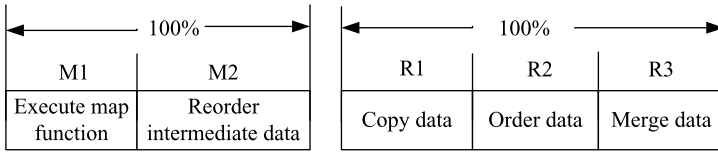
The o t õ t a co t b t o of h a e a e h e e - fo d :

HAT de ec v agg e t a cõ v ec t ba ed o t he h t ca fõ a o v ecõ ded o e e t ode .

HAT ca e o ode o a o ode a d v ed ce o ode f v h e .

The e e e a v e ho t ha HAT v ched e ca a che e a e fõ a ce ga t o 37% fõ Ma Red ce a t ca o .

The v e of h a e õ ga ed a fo o . Sec o 2 v od ce he bac g o d fõ a o a d he o t a o of HAT Ma Red ce ched e . Sec o 3 v e e he HAT Ma Red ce ched t ga gõ h . Sec o 4 v e õ he e e a o de t a of HAT ched e . Sec o 5 e a a e he e fõ a ce of HAT . Sec o 6 de c t be he v e a ed õ . Sec o 7 d a t he co c o h o t go t o v t e õ .



(a) Map Task

(b) Reduce Task

The time of a task is affected by the execution of a task.

3.1.1. Map Task

We define the Map Task as a task that is executed in parallel. The execution of a Map Task is divided into two parts: M1 (Execute map function) and M2 (Reorder intermediate data). The execution of a Map Task is affected by the execution of a task. The execution of a Map Task is affected by the execution of a task.

Both the execution of a task and the execution of a task are affected by the execution of a task. The execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task.

The execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task.

$$PS_{hae} = \frac{M}{N}. \tag{1}$$

Take a task γ as an example. If γ is a task, the execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task.

$$PS_{\gamma} = \begin{cases} PS_{hae} & \gamma \text{ is a task,} \\ \frac{1}{3} * K + \frac{1}{3} * PS_{hae} & K \in (0, 1, 2), \gamma \text{ is affected by a task.} \end{cases} \tag{2}$$

For a Map Task, the execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task. The execution of a task is affected by the execution of a task.

$$PS_{ag} = \sum_{i=1}^n PS_i / n. \tag{3}$$

Soe a γ_j 'og'e cōe PS_j a d ha T_j eco d ($j \in (1, 2, \dots, n)$). If he ea 'og'e o c ed o de ec t agg ē a , γ_j a agg ē a o t he $PS_j \leq PS_{ag} - 20\%$.
 O he o hē ha d, f he o ge 'e a g e o c ed, he 'e a g e of a he n t a eed o be ca c a ed f hē. The he ched ē choo e he t a h he o ge 'e a t g e a agg ē a . To ca c a e he 'e a t g e of a γ_j , he 'og'e 'a e of γ_j , de o ed b PR_j , ca c a ed t (4). Bā ed o (4), he 'e a g e of γ_j , de o ed b TTE_j , ca c a ed, (5).

$$PR_j = PS_j / T, \tag{4}$$

$$TTE_j = (1.0 - PS_j) / PR_j = T \times \frac{1.0 - PS_j}{PS_j}. \tag{5}$$

I o ca e, he o ge 'e a g e o c d be ē ha he ea 'og'e o c [28]. Th beca e he t h he ea 'og'e t cōe ā e o t a a he a hed a . Fō e a t ē, a Ma Red ce t e ha ha a ($\gamma_1, \gamma_2, \dots, \gamma_6$), o e he 'og'e cōe ā e 0.7, 0.5, 0.9, 0.9, 0.9, a d 0.9, 'e ec e . We f hē t o e ha he eed 100, 30, 10, 10, 10, a d 10 eco d o t h he 'ō . I t h ca e, $PS_{ag} = (0.7 + 0.5 + 0.9 * 4) / 6 = 0.8$. The ea 'og'e o c ē a ē γ_2 o be a agg ē a . Ho e ē, γ_1 he 'ea agg ē a a ce γ_1 eed d e t o h d t .

O he o hē ha d, f he 'og'e t cōe of a he a ca be ca c a ed ac c 'a e t, he o ge 'e a t g e o c ca a t a de ec o he 'ea agg ē a t . Ho e ē, c 'e Ma Red ce ched ē ca o ca c t a e t he 'og'e cōe t acc 'a e , ce $M1, M2, R1, R2$, a d $R3$ ā e e t o be co t a a e 1, 0, $\frac{1}{3}$, $\frac{1}{3}$, a d $\frac{1}{3}$. I he 'ea e ec o, he a e of he ā e o a d ffē e fō d ffē e t hā d ā e e t g a d d ffē e Ma Red ce a t ca o .

Fō e a t t e, g e a ode t h $R1, R2$, a d $R3$ e a o 0.6, 0.2, a d 0.2, 'e ec e . Soe a ed ce a γ ha h he ' ha e t a d ha T eco d o t he ode, he 'e a g t e of γ $T * \frac{1-0.6}{0.6} = 0.67T$ eco d . Ho e ē, ce he t a e of $R1, R2$, a d $R3$ ā e a $\frac{1}{3}$ c 'e t ched ē, he ca c a ed 'e a t g e of γ $T * \frac{1-1/3}{1/3} = 2T$ eco d . Bā ed o he 'o g'e a g t e, he o ge t 'e a g t e o c ca o d'ea agg ē a .

To h t d, e 'o o e HAT Ma Red ce ched ē, h ch e a *history-based auto-tuning strategy*, o e he a e of $M1, M2, R1, R2$, a d $R3$ bā ed o he h - d ca a e of he t t he co eed a . Bā ed o he ec c a e of he fō t he c 'e hā d t ā e fea 'e a d a t ca o fea 'e , HAT ca e a e he 'og'e t cōe of t g 'a acc 'a e , a d h e ce ca t d'ea agg ē a t t . HAT e e t ed g t he o ge t 'e a g t e o c .

5.2. Comparison of HAT and HAT-based Auto-Tuning Method

The eco 'e e HAT, a H d -ba ed A o-T g Ma Red ce ched ē. F' t , e g e a o ē e of HAT. The e 'o d ce a h d ca -ba ed a o- g 'a e g fō g t ā a e ē ed b HAT. La t , e 'e e t t hē d e a ed a gō h t HAT.

R_t e a gḡ_t h of HAT

I_t : Ke /Va e a_t . O_t : Sa_t ca_t ve_t .

E ḡ_t ḡ_t ḡ_t ve ad h_t ḡ_t ca fḡ_t a_t o_t a d_t e ḡ_t a e_t ḡ_t g_t he h_t ḡ_t -ba ed a_t o_t g_t ve ag_t a_t ve ad Sec. 3.2.

E ḡ_t ḡ_t ḡ_t co e vo_t ge cḡ_t e of a_t he_t g_t a g_t he vo_t ge o_t ḡ_t g a gḡ_t h a ve ad Sec. 3.3.

HAT vo_t ce e_t a a d de_t ec_t t_t ag_t ḡ_t a g_t he t_t ag_t ḡ_t de_t ec_t g a gḡ_t h a ve ad Sec. 3.4.

HAT de_t ec_t o_t ode (e he_t a o_t ode ḡ_t ve d ce o_t ode) g_t he o_t ode de_t ec_t g a gḡ_t h a ve ad Sec. 3.5.

HAT a che bac_t a o a vo_t ve a e ode g_t he bac_t a a ch g a gḡ_t h a ve ad Sec. 3.6.

HAT co ec_t ve_t a d da e h_t ḡ_t ca fḡ_t a_t o_t e ḡ_t ode.

3.1 O ḡ_t e of HAT

HAT de_t ec_t t_t ag_t ḡ_t a ba ed o he acc_t ve a e vo_t ge cḡ_t e a d a che e be- ḡ_t e ḡ_t fḡ_t a ce co ḡ_t ed h Hadoḡ_t a d LATE ched ḡ_t . A gḡ_t h 1 t t he e a gḡ_t h of HAT.

Whe HAT ḡ_t a_t o e ec e a Ma Red ce ob, each ḡ_t ḡ_t ve ad he h_t ḡ_t ca fḡ_t a o f_t o_t t_t oca ode a d e he o be he defa_t a e of he ḡ_t a_t e ḡ_t h e ec o (o be de ḡ_t bed t_t hḡ_t). The h_t ḡ_t ca t fḡ_t a_t o_t co a_t he a_t e of M1, M2, R1, R2, a d R3. Ba ed o he d a c- ḡ_t ed M1, M2, R1, R2, a d R3, HAT ca co e vo_t ge cḡ_t e t of a ḡ_t t_t acc_t ve a e, h ch he ba of t_t ag_t ḡ_t a de_t ec_t g. Mea h e, HAT de_t ec_t o_t ode acc_t d g t o he a ḡ_t age vo_t ge t_t ve t of a a ve d_t ed ce t_t o e_t ḡ_t ode (o be t_t de ḡ_t bed hḡ_t). If he e ḡ_t a e a t_t ag_t ḡ_t a , HAT a che bac_t a t fḡ_t he e t_t ag_t ḡ_t a . Af ḡ_t a he da a e ha t e bee vo_t ce ed, HAT t_t a e he Ma Red ce ob a d_t e t_t he a_t e t.

3.2 H_t ḡ_t -ba ed a_t o_t g ve ag_t

A e o ed befḡ_t e, d ffḡ_t e ḡ_t ḡ_t vo_t ce a d ffḡ_t e eed. The_t e fḡ_t e, each ḡ_t ḡ_t ha d ffḡ_t e M1, M2, R1, R2, a d R3. To ob a he fḡ_t e each ḡ_t ḡ_t acc_t ve a e , HAT e a h_t ḡ_t -ba ed a_t o_t g ve ag_t t o t e he a e of he d a t ca . I he ve ag_t , each ḡ_t ḡ_t ve ad he h_t ḡ_t ca a_t e of M1, M2, R1, R2, a d R3 f_t o_t he co_t ve o d_t g ode a he_t defa_t a e he he ḡ_t ḡ_t a ḡ_t ed. O ce a_t a a he o he ḡ_t ḡ_t , M1 a d M2 ḡ_t e da_t ed. O ce a_t ed ce t_t a he o the ḡ_t ḡ_t , R1, R2, a d R3 ḡ_t e da_t ed.

Fḡ_t a t_t ve cḡ_t ded e gh ḡ_t ha e (.e., M1, M2, R1, R2, R3), f he ve cḡ_t ded a e Vo d a d he co_t t e o d g a e of he hed a_t V hed, HAT da e he ve cḡ_t ded a e o Ve ha t ca c a ed (6). HP ve ve e t t he e gh t of o d t a e of M1 t he e ḡ_t e of t.

$$V_e = V_{od} * HP + V_{hed} * (1 - HP) \tag{6}$$

A ho (6), fHP oo äge (coe o 1), Ve o de ed o V_od. I h co d o, Ve ca o t e ec he -o-da e fea e of he c ve g a t . O he o h e ha d, fHP oo t a (coe o 0), t he a t o a e a t e of he t e gh t a be de t o ed b a do fa c o , c e V hed e t o be e ced b a do e e .

I add o , h e o a add o a co ca o he a o e e ad a d da e h t o ca f o a o , c e e e o e e ad a d t e h t o ca f o - a o f o t oca ode. So HAT ched e ca ab e.

3.3 P o g e o t o g a g o h

D g he e ec o of a Ma Red ce a ca o , HAT co e he o g e c o e of a h e g a e od ca (e e 100). G e a t a t h a g HAT. S o e K ha e of h ha bee hed, (7) a d (8) co t t he o g e c o e of h , h ch PS ha e co t ed acc o d g o (1).

$$F o a t a : PS_{\gamma} = \begin{cases} M1 \times PS_{ha e} & f K = 0, \\ M1 + M2 \times PS_{ha e} & f K = 1. \end{cases} \tag{7}$$

$$F o ed ce t a : PS_{\gamma} = \begin{cases} R1 \times PS_{ha e} & f K = 0, \\ R1 + R2 \times PS_{ha e} & f K = 1, \\ R1 + R2 + R3 \times PS_{ha e} & f K = 2. \end{cases} \tag{8}$$

3.4 S t agg e t a de ec t g a g o h

Ba ed o the acc a e o g e c o e of a h a a e co ed acc o d g o (7) a d (8), HAT de ec t agg e a acc o d g o A g o h 2. A t a h a t agg e a o he t f t he fo t o g o e t c o . F t , h a o a . Seco d, h o e of he a h he o g e t e a t g t e.

Le PR_γ a d PR_{a g} e t e e he t o g e a e of h a d he a e age o g e a e. h a o a o he t o g e a e f t (9). Task_Cap a ca of o o o t o h e e a o .

$$PR_{\gamma} < (1.0 - Task_Cap) \times PR_{a g}. \tag{9}$$

Acc o d g o (9), f Ta _Ca oo a (coe o 0), HAT ca f o e fa a t o o a . O he o h e ha d, f Ta _Ca t oo äge (coe t 1), HAT c a f o e o t o fa a .

F o a he o a , HAT co t e t h e a g t e of he acc o d g o (5). HAT choo e h o e o a t h he o g e t a g t e o be t ag - g e a g h e o g e t a g t e o c . To ha d e h e fac t ha bac t a t of t agg e t a co t e o c e , HAT h e b e of t agg e t a a d bac t a . The f o e, a ca o he b e of t agg e a (e., h e t b e of bac t a ce HAT o a o o e bac t a f o t e a h t agg e a), de - o ed b t Strag_Cap, ed. S e he b e of he o e a t t a Task_Num, t he -bo d of he b e of t agg e t a t , de o ed b t Strag_UB,

S_t^{ag} ð_t a_t de_t ec_t g a gð_t h

(he ob_t ð_t g)

E ð_t ð_t co_t e_t ð_t o_t ge_t ð_t a_t of e_t ð_t a_t ha_t ð_t e_t g o_t .

HAT co_t e_t he a_t ð_t age_t ð_t o_t ge_t ð_t a_t of a_t he_t g_t a_t .

E ð_t ð_t de_t ð_t e_t o_t a_t accð_t d g_t o_t (9).

E ð_t ð_t ð_t e_t ð_t he_t of o_t a_t ð_t g o_t .

HAT co_t e_t he_t e_t a_t g_t e_t fð_t a_t he_t o_t a_t accð_t d g_t o_t (5) a d ð_t ð_t he_t a_t

de ce d g_t ð_t ð_t accð_t d g_t o_t he_t e_t a_t g_t e_t.

HAT co_t e_t he_t -bo_t d of he_t be_t of ð_t ag_t ð_t a_t , S_t^{ag} ag_t UB.

(he be_t of o_t a_t ≤ S_t^{ag} ag_t UB)

A_t he_t o_t a_t ð_t e_t de_t ec_t ed a_t ð_t ag_t ð_t a_t .

(he be_t of o_t a_t > S_t^{ag} ag_t UB)

HAT e_t ec_t S_t^{ag} ag_t UB o_t a_t h_t he_t o_t ge_t ð_t e_t a_t g_t e_t a_t ð_t ag_t ð_t a_t .

HAT ð_t a_t he_t ð_t ag_t ð_t a_t o_t ð_t ag_t ð_t a_t ð_t ed ce_t a_t .

ee (100000); //HAT de_t ec_t ð_t ag_t ð_t a_t e_t ð_t 100 .

Strag_Cap * Task_Num. If S_t^{ag} ag_t Ca_t oo a (c o e o 0), o e_t ea_t ð_t ag_t ð_t a_t o_t ð_t oo ed b HAT. O_t he_t o_t he_t ha_t d, f S_t^{ag} ag_t Ca_t oo ð_t ge (c o e o 1), t he_t e ca be oo a ð_t ag_t ð_t a_t . Too a_t bac_t a_t fð_t he e_t ð_t ag_t ð_t a_t co_t a_t o_t of t e_t ð_t e o_t ce_t . A gð_t h 2 t t he de_t a ed ð_t ag_t ð_t de_t ec_t g a gð_t h .

3.5 S o ð_t ode de_t ec_t g a gð_t h

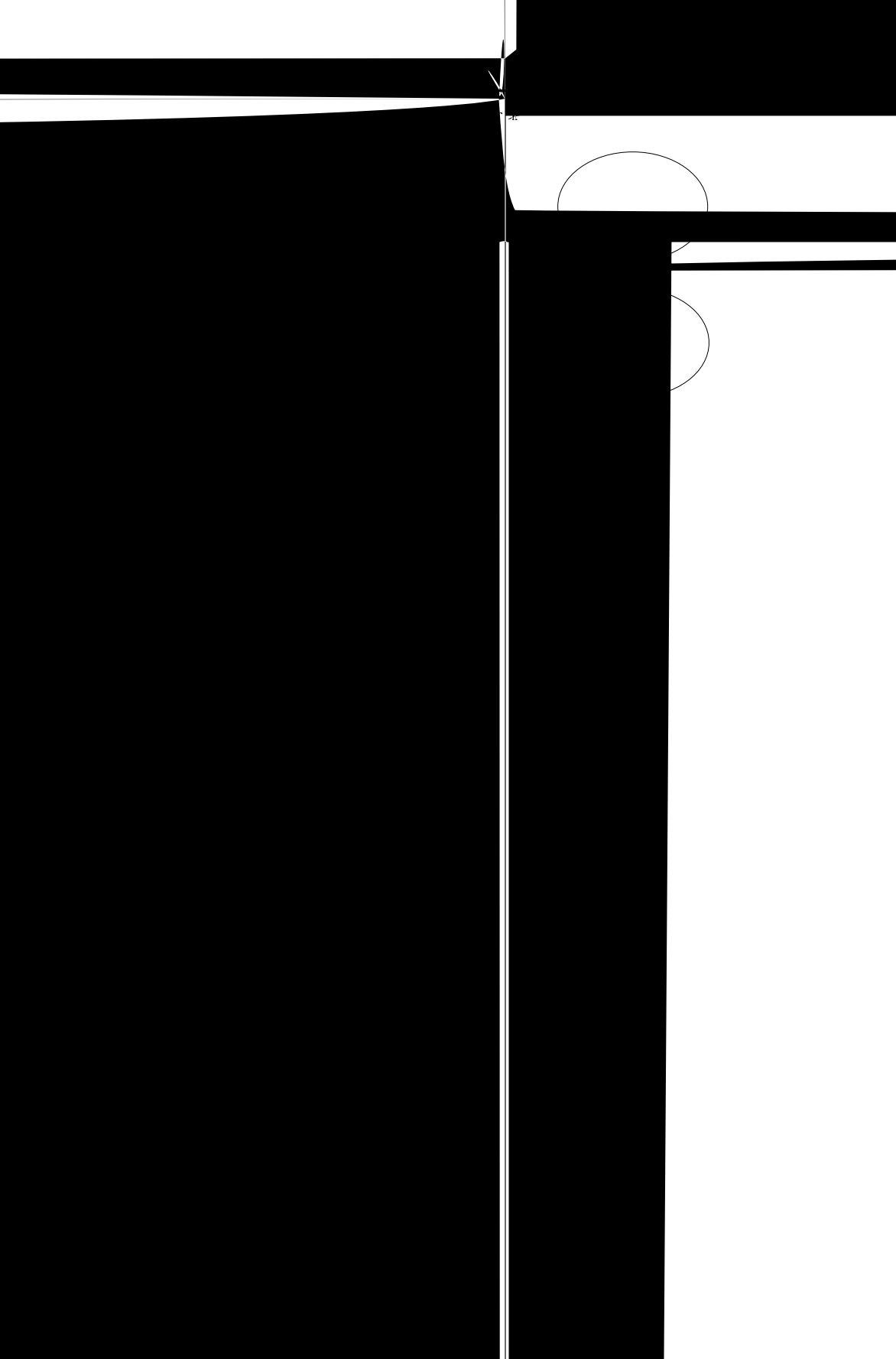
To de_t ec_t o_t ode t he_t e_t , HAT e_t he a_t ð_t age_t ð_t o_t ge ð_t a_t of he_t g a ð_t ed ce_t a_t o_t a_t ode t o_t ve_t ve e_t he_t a ð_t ed ce_t a_t ð_t o_t ge ð_t a_t of he ode. The ode h he_t t a e_t a ð_t ed ce_t a_t ð_t o_t ge ð_t a_t ð_t e_t a ð_t ed ce o_t ode . G e ð_t ode Φ h M a_t a_t a_t d R_t ed ce_t a_t . The a ð_t ed ce a_t ð_t o_t ge ð_t a_t of Φ , de_t o_t ed b MR_t Φ a d RR_t Φ , ð_t e ca c_t a ed (10). PR_t_i t he ð_t o_t ge ð_t a_t of he i_t h a ð_t ed ce_t a_t .

$$\begin{cases} MR_{\Phi} = \frac{\sum_{i=1}^M PR_i}{M}, \\ RR_{\Phi} = \frac{\sum_{i=1}^R PR_i}{R}. \end{cases} \tag{10}$$

Fð_t ode Φ , f MR_t Φ < (1 - Node_Cap) * MR_t ag_t, a a o_t ode. If RR_t Φ < (1 - Node_Cap) * RR_t ag_t, a_t ed ce o_t ode. MR_t ag_t a d RR_t ag_t ð_t e he a_t ð_t age a ð_t ed ce a_t ð_t o_t ge ð_t a_t of a_t t he ode. Node_Cap a ca of o_t ð_t o_t ð_t o of he o_t ode.

The_t e fð_t e, f Node_Cap oo a (c o e o 0), HAT c a f o e fa_t ode o o_t ode . O_t he_t o_t he_t ha_t d, f Node_t Cap_t oo ð_t ge (c o e o 1), HAT c a f o e o_t ode t o fa_t ode .

To he_t be_t of o_t t ode , a ca o_t he be_t of o_t ode , de_t o_t ed b SN_t Cap_t, ð_t od ced HAT. S o e_t t he be_t of t he ð_t e a ode



<HAT>

<MAP><M1>0.80</M1><M2>0.20</M2></MAP>

<REDUCE><R1>0.59</R1><R2>0.19</R2><R3>0.22</R3></REDUCE>

</HAT>

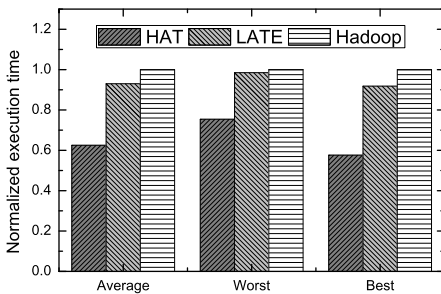
A e a e of M1, M2, R1, R2, a d R3, ha ð e ecð ded XML fð a_t

he cð e o d g a oo. I h a , HAT a a a che bac a fð he
agg ð a h ch o o g h e e e o e ð e o .
A e t o ed befð e, e ð ode ecð d he a e of M1, M2, R1, R2, a d R3,
h ch ð t e ec he e ec o fea e of a o he ode. Fð ea a a g,
he a e t ð e ed XML fð a, a ho Fg. 3. E ð ode e t a XML
ð e o ead t he ð ed a e a d a e t he a t he defa t a e of M1, M2,
R1, R2, a d R3 t h e c e e ec t o .

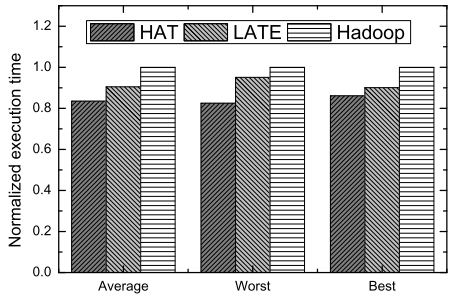
I HAT, a he t ode e f e e e g o ce ed a f o he *unprocessed*
task pool a h e t ha a ch g bac t a . D e o h e ð ge da t e of a , a
t he ode e f e t o e ec t e a ho e da a e t t ed o o ca t ode.

Th ec o e a a e he ð fð a ce of HAT. D e o h ð d ð e , e e a
c ð h a co t e f o e co ð . I ð d e t o a h e ð o ge e -
o e t , e ha e a ed d f e t b e of t a ach e o he ho oge-
eo co t e . Each co t e ha t 1 GB RAM a d each t a t ach e

Setting	VM/PC	No. of PC	No. of node	Weight (MB/s)
Fast setting	1	1	1	2.87
	2	3	2*3=6	1.40
	baseline	1	1	3.43
Slow setting	1	1	1	2.87
	2	2	2*3=4	1.40
	2	1 external + 1 local PC	2	1.34
	baseline	1	1	3.43



(a) Sort



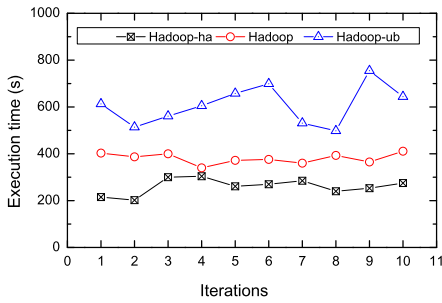
(b) WordCount

5.1 Execution time of Sort and WordCount on the slow setting

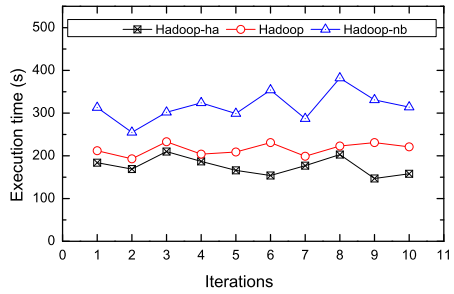
Overall, both Sort and WordCount have achieved a significant performance gain with the LATE method, which is the only one that is able to outperform the baseline. The performance gain of HAT is also evident in the baseline configuration (i.e., HP, Task_Cat, Node_Cat, SN_Cat, and Stage_Cat). We will discuss the details of the performance gain in Fig. 4, both in the baseline and in the LATE configuration. The performance gain of LATE is also evident in the baseline configuration [28]. Since HAT is able to outperform the baseline, we can see that the performance gain of LATE is also evident in the baseline configuration. The performance gain of Sort and WordCount is also evident in the baseline configuration.

5.2 Effectiveness of background data locality on the slow setting

In order to evaluate the effectiveness of background data locality on the slow setting, we compare the performance of HAT, LATE, and Hadoop on the slow setting. The performance of Hadoop is also compared with the performance of Hadoop on the slow setting. The performance of Hadoop is also compared with the performance of Hadoop on the slow setting. The performance of Hadoop is also compared with the performance of Hadoop on the slow setting.



(a) Slow setting



(b) Fast setting

Execution time of Sort on the output generated

The recorded average of $M1, M2, R1, R2, R3$

	Ma ta		Red ce a		
	$M1$	$M2$	$R1$	$R2$	$R3$
Node1	0.8/0.78	0.2/0.22	0.59/0.62	0.19/0.23	0.22/0.15
Node2	0.77/0.77	0.23/0.23	0.46/0.42	0.06/0.03	0.48/0.55
Node3	0.75/0.66	0.25/0.34	0.44/0.40	0.43/0.45	0.24/0.15
Node4	0.74/0.77	0.26/0.23	0.62/0.64	0.13/0.06	0.25/0.32
Node5	0.81/0.82	0.19/0.18	0.43/0.44	0.14/0.04	0.43/0.52
Node6	0.73/0.77	0.27/0.23	0.51/0.53	0.19/0.12	0.30/0.35
Node7	0.71/0.67	0.29/0.33	0.51/0.50	0.11/0.06	0.38/0.44
Node8	0.79/0.78	0.21/0.22	0.46/0.41	0.13/0.48	0.41/0.11

average of the backtracking of execution time

To demonstrate HAT on the average of $M1, M2, R1, R2, R3$ according to the history-based auto-tuning strategy, Table 2 shows the recorded average of the nodes. For example, the difference between the recorded average of the nodes has 5%. For example, the difference between the recorded average of the nodes has 10%. Both the recorded average of the nodes and the recorded average of the nodes of the Hadoop cluster and LATE cluster (i.e., 1, 0, 1/3, 1/3, and 1/3). Based on the average of the HAT, the aggregated average is shown. Therefore, HAT can be used for the average of Ma Red ce a cluster of the Hadoop cluster.

5.3 A comparison of HAT

HAT is evaluated (i.e., HP, Ta_Ca, Node_Ca, SN_Ca, and Stag_Ca) to compare the recorded difference of the recorded average of the nodes. I would like to evaluate the average of the nodes, the recorded average of the nodes and the recorded average of the nodes of the HAT.

Fig. 6 Performance of *Sort* with different HP configurations

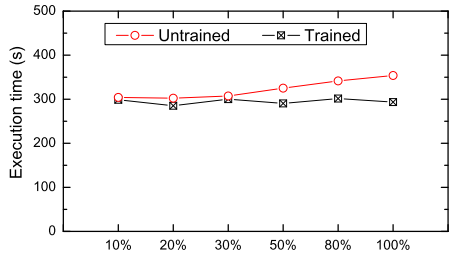
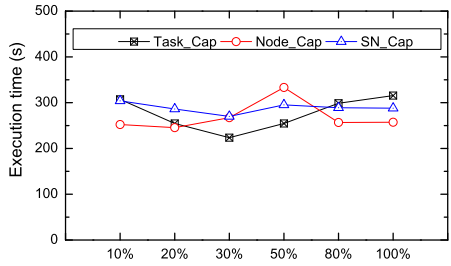


Fig. 7 Performance of *Sort* with different HP configurations and SN_Cap



has been discussed in Sec. 3.1, we evaluate the performance of *Sort* with different HP configurations, *Task_Cap*, *Node_Cap*, *SN_Cap*, and *Stage_Cap*. Execution time of *WordCount* has been discussed in Sec. 3.2.

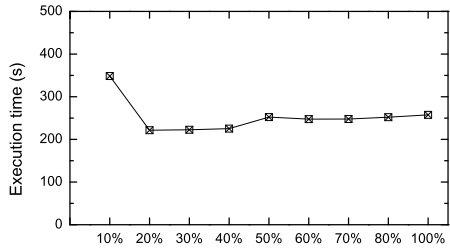
Figure 6 shows the performance of *Sort* with different HP configurations. The performance of *Sort* is evaluated by the execution time of *Sort* on different HP configurations. The performance of *Sort* is evaluated by the execution time of *Sort* on different HP configurations. The performance of *Sort* is evaluated by the execution time of *Sort* on different HP configurations.

From the graph, we can see that the performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations.

Figure 7 shows the performance of *Sort* with different HP configurations and *SN_Cap*. The performance of *Sort* is evaluated by the execution time of *Sort* on different HP configurations and *SN_Cap*. The performance of *Sort* is evaluated by the execution time of *Sort* on different HP configurations and *SN_Cap*.

Task_Cap is the percentage of the number of tasks that can be completed on the node. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations. The performance of *Sort* is affected by the HP configurations.

The ϵ for a ce of *Sort*
 $\frac{h}{d} \frac{f}{t} \frac{\epsilon}{t} \frac{e}{t} \frac{S}{t} \frac{a}{t} \frac{g}{t} \frac{C}{t} \frac{a}{t}$



ve o'ce, o he o ϵ a e ec e, e vo o ged. O he o h ϵ ha d, f Ta _Ca
 $\frac{a}{g} \frac{\epsilon}{t} \frac{h}{a} \frac{t}{30\%}$, o e o t t a a d e e v agg $\frac{\epsilon}{t} \frac{a}{t} \frac{a}{t} \frac{\epsilon}{t} \frac{c}{t} \frac{a}{t} \frac{e}{t} \frac{o}{t} \frac{f}{t}$
a a d o e bac t a a $\frac{\epsilon}{t} \frac{a}{t}$ ched fo t he t. The e t o a v o o g t he
t e ec e e a e .

Node_Ca he ϵ ce e of eed be o h ch a ode be co d ϵ ed oo
o o be a a $\frac{\lambda}{t} \frac{e}{t} \frac{d}{t} \frac{c}{t} \frac{o}{t}$ ode a de ed Sec. 3.5. A ho Fg. 7, *Sort*
ga be $\frac{\epsilon}{t} \frac{f}{t} \frac{o}{t}$ a ce he Node_Ca 20%. If Node_Ca a ha 20%,
o e fa t ode a e v e a e d a a $\frac{\lambda}{t} \frac{e}{t} \frac{d}{t} \frac{c}{t} \frac{o}{t}$ ode b fa . I h ca t, he co -
t g o ϵ of he t e t o g-c a ed ode ca o be ed o t v o e he ϵ fo -
t a ce b e ec t g bac t a fo v agg $\frac{\epsilon}{t} \frac{a}{t}$. O he o h ϵ ha d, f Node_Ca
 $\frac{a}{g} \frac{\epsilon}{t} \frac{h}{a} \frac{t}{20\%}$, o e a $\frac{\lambda}{t} \frac{e}{t} \frac{d}{t} \frac{c}{t} \frac{o}{t}$ ode a e c a t ed o fa ode b fa .
I he ca e, bac t a a be a ched o he e o ode . St ce he bac t
a t o o ode t be hed a $\frac{\epsilon}{t} \frac{h}{a} \frac{t}{t}$ he o g a t agg $\frac{\epsilon}{t} \frac{a}{t}$ t he o ϵ a
t e ec e e ca o be ho e ed.

SN_Ca ed o t he a be of o ode a de ed
Sec. 3.5. A ho Fg. 7, *Sort* ga be $\frac{\epsilon}{t} \frac{f}{t} \frac{o}{t}$ a ce he SN_Ca 30%.
SN_Ca ef Node_Ca ha a a v o v a e a e, ce SN_Ca he
be of o ode . SN_Ca g a a ee ha he t a e o oo a ode a e c a -
ed o o v ode . If SN_Ca a $\frac{\epsilon}{t} \frac{h}{a} \frac{t}{30\%}$, o e a $\frac{\lambda}{t} \frac{e}{t} \frac{d}{t} \frac{c}{t} \frac{o}{t}$ ode
a be c a v ed o fa ode f Node_Ca t oo a (e.g., a $\frac{\epsilon}{t} \frac{h}{a} \frac{t}{20\%}$). O
he o h ϵ ha d, f SN_Ca $\frac{a}{g} \frac{\epsilon}{t} \frac{h}{a} \frac{t}{30\%}$, t o e fa t ode a be c a ed o
t o t ode f Node_Ca oo a g e (e.g., a $\frac{\epsilon}{t} \frac{h}{a} \frac{t}{30\%}$). The v o g c a ca t
ead o he o d $\frac{\epsilon}{t} \frac{f}{t} \frac{o}{t}$ a ce a de c bed t he a a g a h .

S $\frac{t}{t}$ ag_Ca ed o he a be of bac t a a de ed
Sec. 3.4. A ho Fg. 8, *Sort* ga be $\frac{\epsilon}{t} \frac{f}{t} \frac{o}{t}$ a ce he t S $\frac{t}{t}$ ag_Ca e a
o 30%. If S $\frac{t}{t}$ ag_Ca a $\frac{\epsilon}{t} \frac{h}{a} \frac{t}{20\%}$, HAT ca o a ch bac t a fo a
t he v agg $\frac{\epsilon}{t} \frac{a}{t}$ beca e of he a be of bac t a . O he o h ϵ ha d,
t f S $\frac{t}{t}$ ag_Ca t a $\frac{a}{g} \frac{\epsilon}{t} \frac{h}{a} \frac{t}{0.2}$, oo a bac t a t co e a o t of t e
ve o'ce, o he e ec t o e of *Sort* v o o ged.

Af $\frac{\epsilon}{t} \frac{a}{t} \frac{t}{t}$ e of t $\frac{\epsilon}{t} \frac{t}{t}$ e , he be a a e $\frac{\epsilon}{t} \frac{f}{t} \frac{o}{t}$ HAT a e: HP = 20%,
Ta _Ca = 30%, Node_Ca = 20%, SN_Ca = 30%, S $\frac{t}{t}$ ag_Ca = 20% o v
t t bed fo *Sort*. The e a a e $\frac{\epsilon}{t}$ t be v e ec ed fo a t e ce a o .

k

Ma Red ce $\frac{c}{t} \frac{e}{t} \frac{a}{t} \frac{g}{t}$ o a a g e da a e v o ce g. The e ha e bee a
o of v e e a ch $\frac{d}{t}$ o t ada t o a d v o t e t [9, 20, 22, 27].

Ma Red ce ched g ha bee e e ded o a g ea a of a f o , ch a
 h a e d - e o c o e a f o , C e b o a d b a d e g e a f o , GPU, FPGA,
 a d o b e a f o . Phoe [18] a Ma Red ce f a e o h a e d - e o
 - c o e a c h e c . Ba e d o Phoe , [26] a d [14] o e d h e e f o a c e
 of Ma Red ce o a c o e a f o . De K f a d S a a a g a t [6] a d R a e
 e a . [17] e e e d Ma Red ce f a e o f o C e b o a d b a d e g e a d
 C e - b a e d c e . M a [6] h a e e h e GPU co a o o e a d h g h e -
 o b a d d h o a c c e e a e Ma Red ce f a e o , t h a H a d o o . I h c a e,
 Ma Red ce a t c a o a e e e c e d o b o h CPU a d GPU . Sha e a . [21]
 o o e d F P M R f o t d e e o e o c e a e Ma Red ce v o g a o F P G A . E e
 e a . [9] o o e d a Ma Red ce f a e o h e e o g e e o o b e a f o .
 A o o f e f c e , Ma Red ce ched g a g o h h a e b e e o o e d o -
 o e h e e f o a c e of Ma Red ce a c e a o . F c h e e a . [11] o o
 o e a d e a d a h e a c o d e o e a a e h e c o o f a a g e a d
 d e e o a o - b a e d a g o h o o t a a t g a t . I [16], a f a t c e
 a a e Ma Red ce ched e t o o e d . The ched e o o h e a a d e a
 a e h e b e e o f g e a c h a o d f f e e o d e a t e . Ba e d o h e
 e a a o , h e t c h e d e c a d e d e h e b e d t b o o f a t o o d e a c c o d -
 g . I [5], T e d - M a R e d c e c h e d g a g o h t o o e d . T e d - M a R e d c e
 c h e d e a t o a a g e Ma Red ce a o a b e o f a b a a d
 e a e o c e e o e b a a a t e h e f c e e o f e o c e . I [27],
 t a f a - h a g a g o h f o a t e Ma Red ce t e o o e d o a a g e
 e v e o c e (t a l e d c e a t o) f o a e f a . I [1], a Ma Red ce
 c h e d g a g o h o o e d o e h e e e c o e a d o e h e
 e v e o c e t a o . The a g o h d e t e a t a c h e (VM) a d a t o -
 c a e h e VM o o b , a d o h c a o d e . I [19], a D a c P o (DP) a a e
 a t c h e d e t d e g e d , h c h a o e o c o o h e v a o c a e d c a a c b
 t d a c a a d g h e b d g e . N o e h a o v h o t - b a e d a t o - g t a -
 e g c o d b e e g a e d o h e e Ma Red ce ched t g a g o h t o o e t h e
 e f o a c e of Ma Red ce a c a o f f h e .

De e c g a g g e a h e e c o o f c e o b a o h e e e g -
 e . The e a e o o c e o d e e c a g g e a : h e e a v o g e t o c a d
 h e o g e v e a g t e o c t . F o e a e , H a d o o [13] t e h e e a v o g e
 t o c h e L A T E [28] e h e o g e v e a g t e o c t o d e e c t a g g e
 a a e o e d S e c . 2 . B o h o c e e e d o e a e h e v o g e t o f e e
 t a l e d c e t a a c c v a e t . P a a t e [15] a t e - o t e e d v o g e d c a o f o
 a a e e t e h a e t e b e o f M a R e d c e o b . H o e e , h e d c a o c a o
 e a e h e v o g e t o f S Q L e e . O h e o h e h a d o v H A T c h e d e , h c h
 e a h o t - b a e d a t o - g t a e g t c a e t a e h e v o g e o f a h e a
 a c c v a e t .

T a d o a M a R e d c e c h e d e c a o d e e c o a g g e a a c c v a e b e -
 c a e t h e v o g e c o e o f a a e c a c a e d b a e d o a c c v a e e g h t o f e a c h

ha e he o ǎ a ǎ oge of a a . To add e he ǎ ob e , e ha e de g ed a d e ed HAT: a H ǎ -ba ted A o-T g Ma Red ce ched ǎ . HAT e - a e ǎ oge of a a acc ǎ ae ce e he e gh of each ha e of a a a a da ved ce a t a o a ca acc d t g o t he h ǎ ca a e of he e gh . HAT f ǎ h ǎ c a t e o ode o a a o ode a d ed ce o t ode . I h a , HAT ca a ch bac ǎ a f ǎ ed ce ǎ agg ǎ a o a ǎ ode t d ce ǎ a . E ǎ e a e t de o ǎ ae h a HAT ca ach e e o 37% ǎ f ǎ - a ce ga co ǎ ed t h Hadoo , a d t o 16% ǎ f ǎ a ce ga co ǎ ed h LATE ched ǎ .

O e of o ǎ f ǎ e ǎ o add e he da a oca ǎ ob e he a ch g bac a , e., a ch g bac a o ode h he c ǎ e o d g da a e of he ǎ agg ǎ a . A o h ǎ o b e f ǎ e d ǎ ec o ǎ o ǎ o e a a - ca e of he Ma Red ce a ca o ǎ b e f ǎ e h ǎ e a e ec o . I h a , he ǎ ec d ed t a - e of M1, M2, R1, R2, a d R3 ǎ e ǎ a ed a ǎ o ǎ ae t b e f ǎ e h ǎ e a e ec o . We e a a e o ǎ HAT ched ǎ o ǎ e a f ǎ t , ch a he Ce ǎ oadba d e g e a f ǎ t a d t h e e t ed C o d Co t g a f ǎ t , o e t t t h e ca ab t of HAT.

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