# Dual-Entangled Polynomial Code: Three-Dimensional Coding for Distributed Matrix Multiplication

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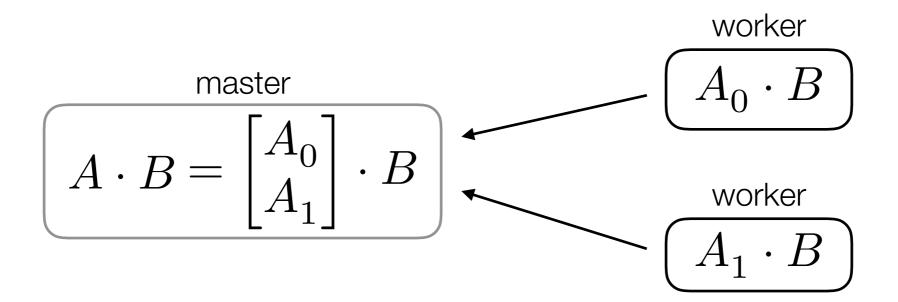
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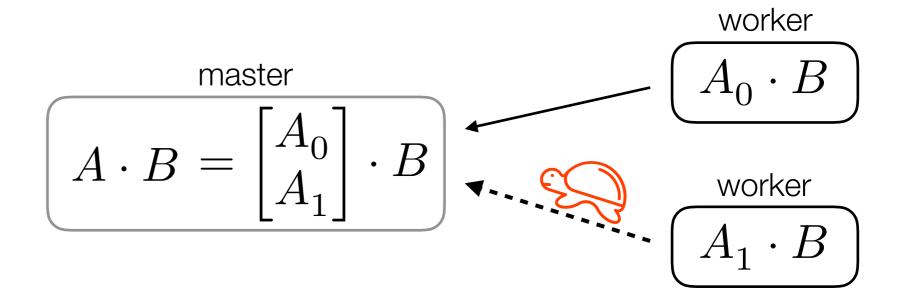
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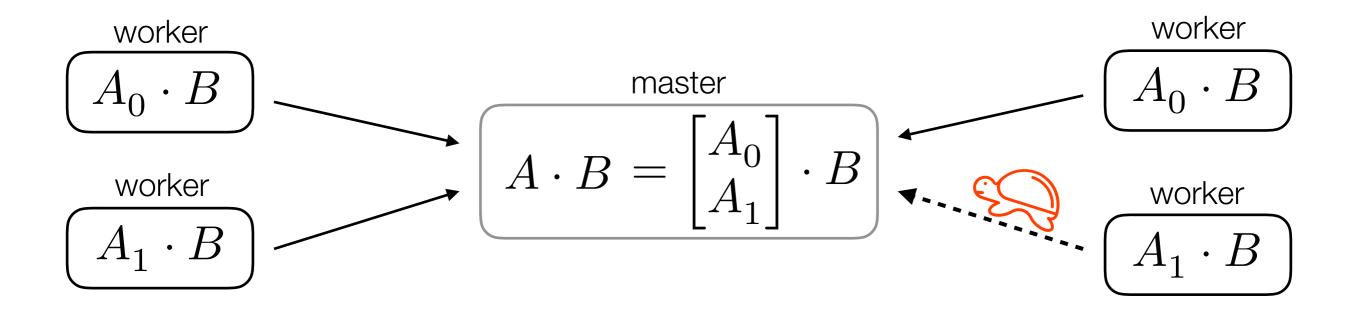
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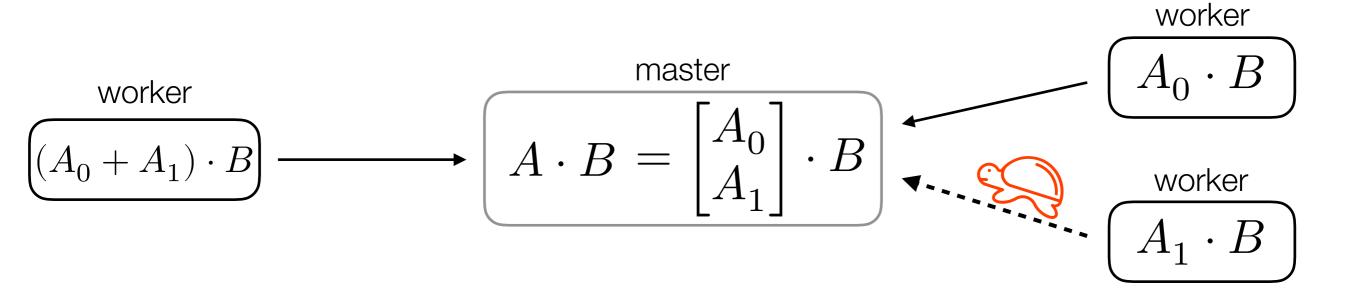
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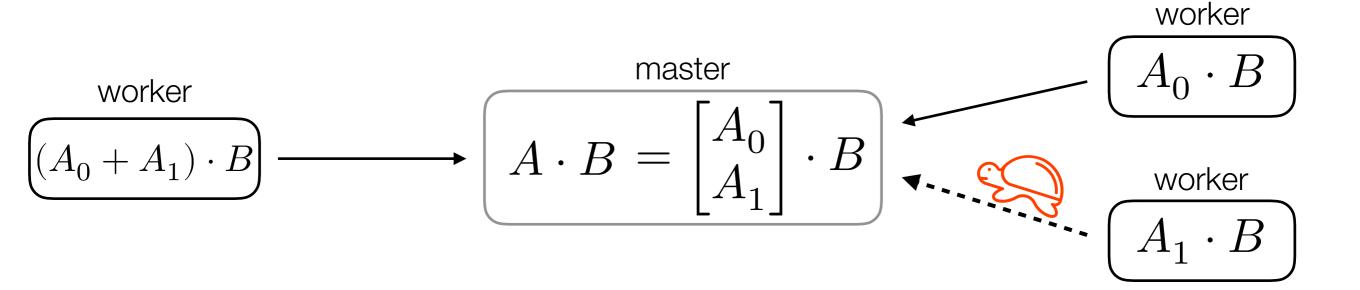








Stragglers are common in distributed systems, due to load imbalance, resource contention, etc.



Coded matrix multiplication can tolerate stragglers with fewer tasks.

#### Related Works

coding	matrix partition	recovery threshold
1D [Lee et al., Trans. IT, 2018]	$\begin{bmatrix} A_0 \\ \vdots \\ A_{x-1} \end{bmatrix} \cdot B$	X
2D [Yu et al., NIPS 2017]	$\begin{bmatrix} A_0 \\ \vdots \\ A_{x-1} \end{bmatrix} \cdot \begin{bmatrix} B_0 & \cdots & B_{y-1} \end{bmatrix}$	Xy
3D [Yu et al., ISIT 2018]	$\begin{bmatrix} A_{0,0} & \cdots & A_{0,z-1} \\ \vdots & \ddots & \vdots \\ A_{x-1,0} & \cdots & A_{x-1,z-1} \end{bmatrix} \cdot \begin{bmatrix} B_{0,0} & \cdots & B_{0,y-1} \\ \vdots & \ddots & \vdots \\ B_{z-1,0} & \cdots & B_{z-1,y-1} \end{bmatrix}$	xyz+z-1

- ▶ Entangled Polynomial (EP) code [Yu et al., ISIT 2018] is the state-of-the-art three-dimensional coding.
- For example, if  $A = \begin{bmatrix} A_{0,0} & A_{0,1} \\ A_{1,0} & A_{1,1} \end{bmatrix}$  and  $B = \begin{bmatrix} B_{0,0} & B_{0,1} \\ B_{1,0} & B_{1,1} \end{bmatrix}$ , i.e., x=y=z=2, a task will be:

$$((A_{0,0}\delta^0 + A_{0,1}\delta^1)\delta^0 + (A_{1,0}\delta^0 + A_{1,1}\delta^1)\delta^4) \times ((B_{0,0}\delta^1 + B_{1,0}\delta^0)\delta^0 + (B_{0,1}\delta^1 + B_{1,1}\delta^0)\delta^2)$$

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$$\tilde{A_0} \qquad \tilde{A_1} \qquad \tilde{B_0} \qquad \tilde{B_1} \\ ((A_{0,0}\delta^0 + A_{0,1}\delta^1)\delta^0 + (A_{1,0}\delta^0 + A_{1,1}\delta^1)\delta^4) \times ((B_{0,0}\delta^1 + B_{1,0}\delta^0)\delta^0 + (B_{0,1}\delta^1 + B_{1,1}\delta^0)\delta^2)$$

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$$\underbrace{\tilde{A}_{0}}_{([A_{0,0}\delta^{0}+A_{0,1}\delta^{1})]}\!\delta^{0} + \underbrace{\left(A_{1,0}\delta^{0}+A_{1,1}\delta^{1}\right)}\!\delta^{4}) \times \underbrace{\left(\left(B_{0,0}\delta^{1}+B_{1,0}\delta^{0}\right)}\!\delta^{0} + \underbrace{\left(B_{0,1}\delta^{1}+B_{1,1}\delta^{0}\right)}\!\delta^{2}\right)}_{\bullet}$$

#### Motivation

- As the input matrices are divided smaller and smaller, the number of results uploaded and decoded at the master also increase.
  - the master's incoming traffic becomes congested, or
  - the master is overwhelmed by the decoding complexity
- Dual entangled polynomial codes allows a tradeoff between computation and communication/decoding overhead.

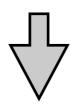
▶ Dual Entangled Polynomial code (DEP) doubles the computational complexity with two multiplications, lowering the recovery threshold to  $\frac{3}{4}xyz + \frac{1}{2}z - 1$ . For example, when x=y=z=2, a task will be

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$$\begin{split} (A_{0,0}\delta^0 + A_{1,1}\delta^1) \times (B_{0,0}\delta^0 + B_{1,1}\delta^1 + B_{0,1}\delta^2 + B_{1,0}\delta^3) \\ + \\ (A_{1,0}\delta^0 + A_{0,1}\delta^{-1})(B_{0,0}\delta^0 + B_{1,1}\delta^{-1} + B_{0,1}\delta^{-2} + B_{1,0}\delta^{-3})\delta^4 \end{split}$$

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$\delta^0$	$\delta^1$	$\delta^2$	$\delta^3$	$\delta^4$	$\delta^5$
$A_{0,0}B_{0,0}$		$A_{1,1}B_{1,1}$	$A_{0,0}B_{0,1}$		$A_{1,1}B_{1,0}$
+	noise	+	+	noise	+
$A_{0,1}B_{1,0}$		$A_{1,0}B_{0,1}$	$A_{0,1}B_{1,1}$		$A_{1,0}B_{0,0}$

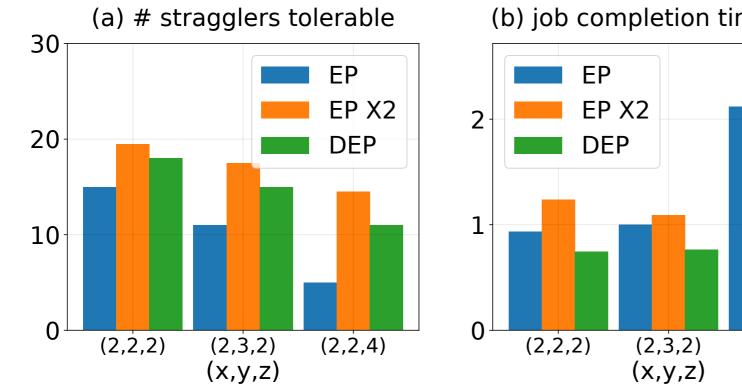
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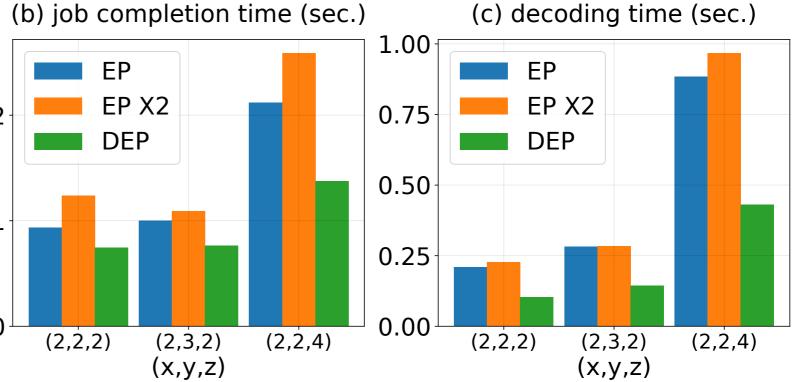
Compared to EP codes, the recovery threshold is reduced from **9** tasks to **6** tasks.

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$A_{0,1}B_{1,0}$		$A_{1,0}B_{0,1}$	$A_{0,1}B_{1,1}$		$A_{1,0}B_{0,0}$

#### Evaluation

- We implemented DEP codes with Open MPI, and ran the evaluation on 24 workers hosted on Microsoft Azure.
  - ▶ EP X2 runs two tasks on each worker.
  - ▶ DEP tolerates similar stragglers, while significantly saving job completion time.





#### Conclusion

- We propose the dual entangled polynomial code, another three-dimensional coding scheme, for distributed matrix multiplication.
- The extra computation at each node allows DEP codes to have a significantly lower recovery threshold than EP codes, leading to a lower communication overhead and decoding complexity.
- ▶ Future work: explore more flexible tradeoff between computation and communication/decoding overhead.

# Thank you.