

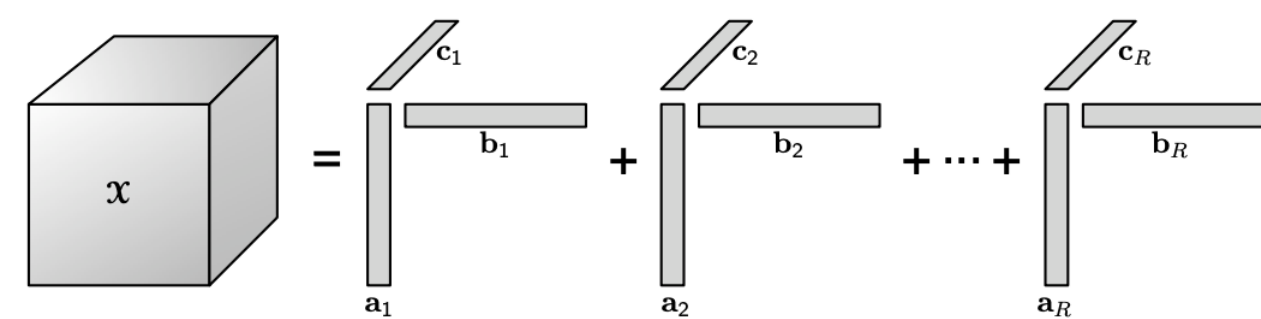
# Error Analysis of Pairwise Perturbation for Tensor Decomposition

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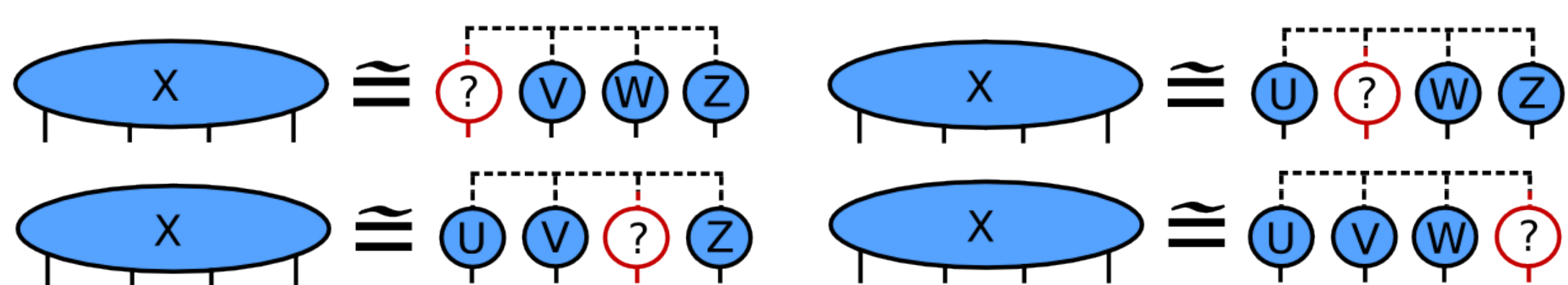
## CP Decomposition with ALS



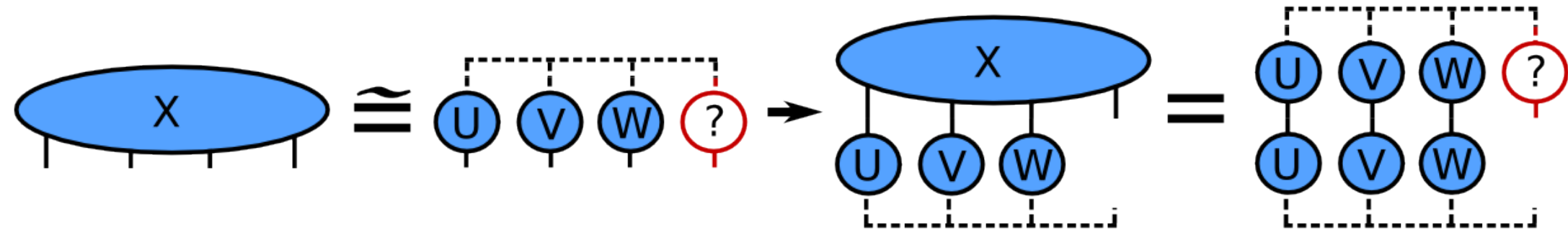
Consider rank  $R$  CP decomposition of an  $s \times s \times s \times s$  tensor

$$x_{ijkl} \approx \sum_{r=1}^R u_{ir} v_{jr} w_{kr} z_{lr}$$

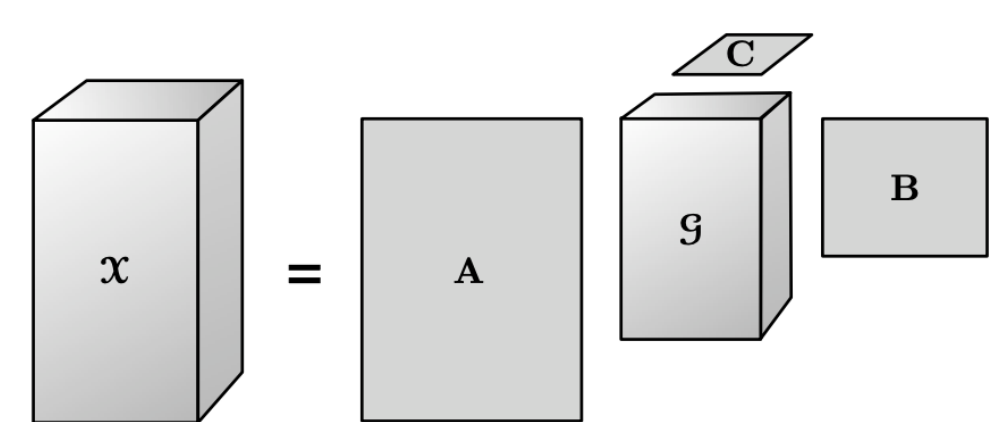
ALS updates factor matrices in an alternating manner



Each quadratic subproblem is typically solved via normal equations



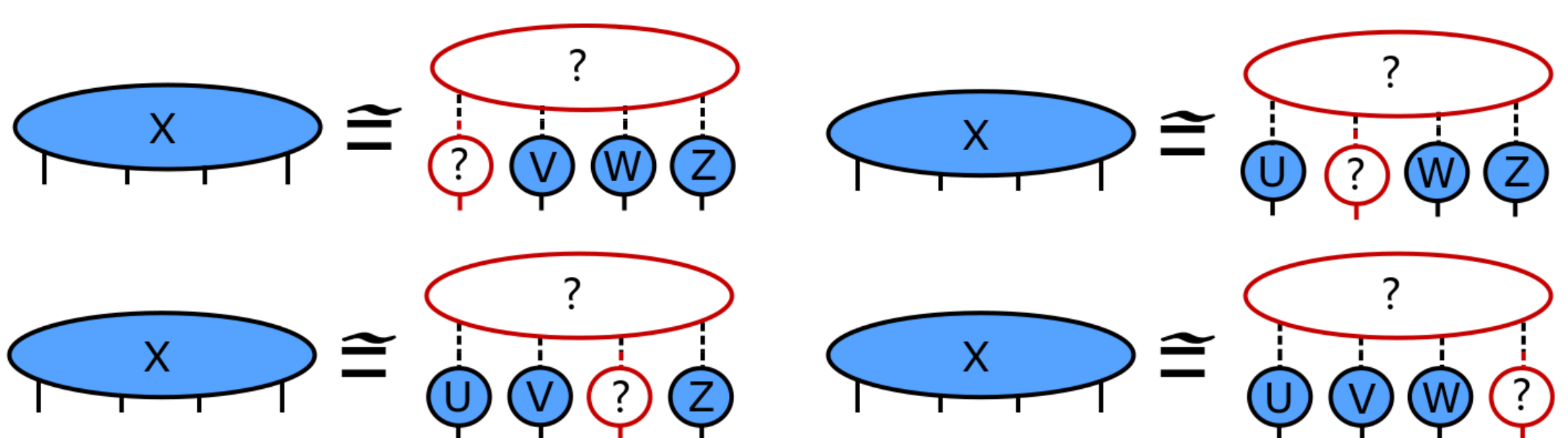
## Tucker Decomposition with ALS



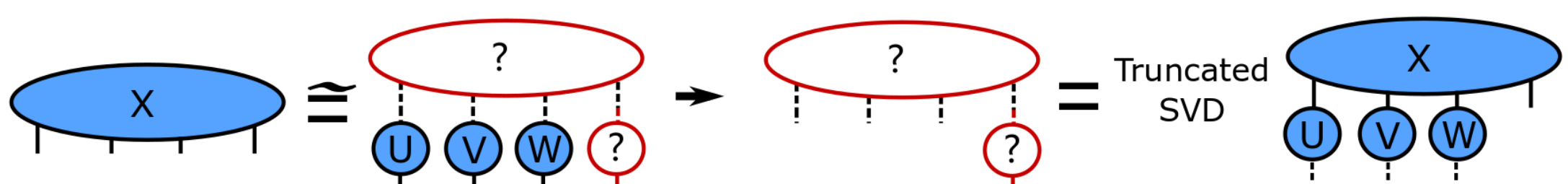
Consider rank  $R$  Tucker decomposition of an  $s \times s \times s \times s$  tensor

$$x_{ijkl} \approx \sum_{a,b,c,d} g_{abcd} u_{ia} v_{jb} w_{kc} z_{ld}$$

ALS updates factor matrices in an alternating manner



Tucker-ALS is usually solved with HOOI (Higher-Order Orthogonal Iteration)



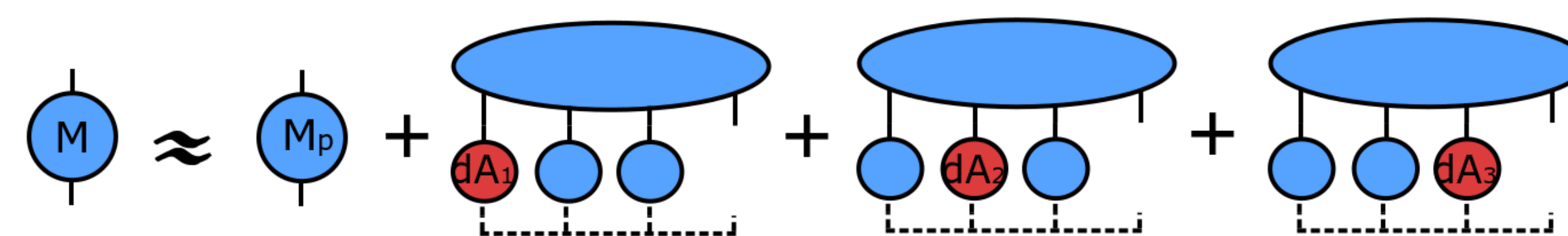
## Pairwise Perturbation

For CP, Pairwise perturbation (PP) approximates MTTKRP result  $M^{(n)} \approx \tilde{M}^{(n)}$  using pairwise perturbation operators  $\mathcal{M}_p^{(i,n)}$

- Write  $A^{(n)} = A_p^{(n)} + dA^{(n)} \rightarrow M^{(n)} = X_{(n)} \odot_{i=1, i \neq n}^N (A_p^{(i)} + dA^{(i)})$
- Element-wise,

$$M^{(n)}(y, k) = M_p^{(n)}(y, k) + \sum_{i=1, i \neq n}^N \sum_{x=1}^{s_i} \mathcal{M}_p^{(i,n)}(x, y, k) dA^{(i)}(x, k) + \dots$$

$$\sum_{i=1, i \neq n}^N \sum_{j=i+1, j \neq n}^N \sum_{x=1}^{s_i} \sum_{z=1}^{s_j} \mathcal{M}_p^{(i,j,n)}(x, z, y, k) dA^{(i)}(x, k) dA^{(j)}(z, k) + \dots$$



	DT ALS	PP initialization step	PP approximate step
CP	$4s^N R$	$6s^N R$	$2N^2 s^2 R$
Tucker	$4s^N R$	$6s^N R$	$2N^2 s^2 R^{N-1}$

## Error Analysis for Tucker

Consider order  $N = 3$  tensor  $\mathcal{X}$ , let  $\mathbf{y}^{(3)}$  be the HOOI result needed to form the third factor matrix  $A^{(3)}$

- Bound relative error of  $\tilde{\mathbf{y}}^{(3)}$  computed by PP middle step
- The  $i$ th factor matrix changed by  $dA^{(i)}$  since the first step of PP

- The spectral norm of the tensor corresponds to  $\|\mathcal{X}\|_2 = \sup\{\|\mathbf{f}\mathcal{X}\|_2\}$ , where  $\mathbf{f}\mathcal{X} \in \mathbb{R}^s \times \mathbb{R}^s \rightarrow \mathbb{R}^s$ ,
- $$z = \mathbf{f}\mathcal{X}(u, v) \Rightarrow z_k = \sum_{i,j} x_{ijk} u_i v_j$$

**Theorem 0.1** (Error Bound with Bounded Residual). If  $\|dA^{(l)}\|_2 \leq \epsilon \ll 1$  for  $l \in \{1, 2, 3\}$  and residual spectral norm  $\leq \frac{1}{3}\|\mathcal{X}\|_2$ ,

$$\frac{\|\tilde{\mathbf{y}}^{(3)} - \mathbf{y}^{(3)}\|_2}{\|\mathbf{y}^{(3)}\|_2} = O(\epsilon^2).$$

**Theorem 0.2** (Error Bound when Tucker starts with interlaced HOSVD). If  $\|dA^{(l)}\|_F \leq \epsilon \ll 1$  for  $l \in \{1, 2, 3\}$  and

- interlaced HOSVD is used to initialize Tucker-ALS
- the decomposition residual is no higher than that attained by HOSVD

$$\frac{\|\tilde{\mathbf{y}}^{(n)} - \mathbf{y}^{(n)}\|_F}{\|\mathbf{y}^{(n)}\|_F} = O\left(\epsilon^2 \left(\frac{s}{R}\right)^{N/2}\right).$$

The error bound is independent of the input tensor conditioning

## Error Analysis for CP

Consider order  $N = 3$  tensor  $\mathcal{X}$ , let  $M^{(3)}$  be the MTTKRP result needed to form the third factor matrix  $A^{(3)}$

- Bound columnwise error of  $\tilde{M}^{(3)}$  computed by PP middle step
- The  $i$ th factor matrix changed by  $dA^{(i)}$  since the first step of PP
- Error bound based on conditioning bound of  $\mathbf{f}\mathcal{X}$

**Theorem 0.3** (Columnwise Error Bound from Tensor Conditioning). If  $\|da_k^{(l)}\|_2 / \|a_k^{(l)}\|_2 \leq \epsilon$  for  $l \in \{1, 2, 3\}$ ,

$$\frac{\|\tilde{m}_k^{(3)} - m_k^{(3)}\|_2}{\|m_k^{(3)}\|_2} \leq \frac{\max_{u,v \in \mathbb{S}^{s-1}} \|\mathbf{f}\mathcal{X}(u, v)\|_2}{\min_{y,z \in \mathbb{S}^{s-1}} \|\mathbf{f}\mathcal{X}(y, z)\|_2} O(\epsilon^2).$$

- If  $\min_{u,v \in \mathbb{S}^{s-1}} \|\mathbf{f}\mathcal{X}(u, v)\|_2 = 0$ , bound is trivial
- There exist  $2 \times 2 \times 2$ ,  $4 \times 4 \times 4$ , and  $8 \times 8 \times 8$  tensors for which  $\|\mathbf{f}\mathcal{X}(u, v)\|_2 = 1$  for all  $u, v \in \mathbb{S}^{s-1}$
- However, for any  $s \notin \{1, 2, 4, 8\}$ , any  $s \times s \times s$  tensor  $\mathcal{X}$  has  $\min_{u,v \in \mathbb{S}^{s-1}} \|\mathbf{f}\mathcal{X}(u, v)\|_2 = 0$

We can bound columnwise error of approximate update  $\tilde{H}^{(1,3)}$  to  $\tilde{M}^{(3)}$  computed by PP middle step due to change in  $A^{(1)}$

- For simplicity, assume  $N = 3$  and  $R = 1$ , so that

$$\mathcal{X} \approx a^{(1)} \circ a^{(2)} \circ a^{(3)}$$

- Updating  $a^{(n)}$  by  $\delta a^{(n)}$  yields update  $h^{(m,n)}$  to  $m$ th factor matrix
- $h^{(1,3)} = R^{(3)} a^{(2)}$  where  $R^{(3)} \in \mathbb{R}^{s \times s}$  is  $\mathcal{T}$  contracted along the last mode with  $\delta a^{(3)}$

**Theorem 0.4** (Columnwise Error Bound from Matricization Conditioning). PP approximate step performs update  $\tilde{h}^{(1,3)}$  with relative error

$$\frac{\|h^{(1,3)} - \tilde{h}^{(1,3)}\|_2}{\|h^{(1,3)}\|_2} \leq \kappa(R^{(3)}) \|da^{(2)}\|_2,$$

where  $da^{(2)}$  is the change to  $a^{(2)}$  since PP initialization

## References

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- T. G. Kolda and B. W. Bader. Tensor decompositions and applications. SIAM review, 51(3):455–500, 2009.
- <https://github.com/cyclops-community/ctf>