

# Quick solvers for endomorphisms of modules

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Joint work with James B. Wilson and Josh Maglione

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# Simultaneous Sylvester System

Given  $A_i \in K^{s \times b}$ ,  $B_i \in K^{a \times t}$ ,  $C_i \in K^{a \times b}$

Find  $X, Y$  satisfying  $\forall i \in \{1, \dots, c\}$

$$\boxed{X} \boxed{A_i} + \boxed{B_i} \boxed{Y} = \boxed{C_i}$$

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$$\begin{aligned}\text{End}_{\Omega}(M) &= \{X \in \text{End}_K(M) \mid \forall A \in \Omega, XA = AX\} \\ &= \{X \in K^{d \times d} \mid \forall i \in [n], XM_i = M_iX\}\end{aligned}$$

## ► Adjoint

$t : K^n \times K^n \rightarrow K^n$  bilinear,  $t(u, v) = (u^\top T_1 v, \dots, u^\top T_n v)$

$$\begin{aligned}\text{Adj}(t) &= \{(\varphi, \psi) \in \text{End}(K^n) \times \text{End}(K^n)^{\text{op}} \mid t(\varphi(\cdot), \cdot) = t(\cdot, \psi(\cdot))\} \\ &= \{(X, Y) \in K^{n \times n} \times K^{n \times n} \mid \forall i \in [n], T_i X = Y^\top T_i\}\end{aligned}$$

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## ► Wedderburn complements

$$\left\langle \begin{bmatrix} B_i & C_i \\ & A_i \end{bmatrix} \right\rangle \sim \left\langle \begin{bmatrix} B_i & \\ & A_i \end{bmatrix} \right\rangle$$

if there exists  $X$  such that  $XA_i - B_iX = C_i$

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## ► Engineering applications in control theory, PDEs, and robotics



$$A \in K^{s \times b \times c}, B \in K^{a \times t \times c}, C \in K^{a \times b \times c}$$

$$(\forall i) \ XA_i + B_iY = C_i$$

$A_{*1*}^\top$	$B_{1**}^\top$	$X_{11}$	=	.
$\ddots$	$\vdots$	$\vdots$		
$A_{*1*}^\top$	$B_{a**}^\top$	$X_{1s}$		
$A_{*2*}^\top$	$B_{1**}^\top$	$\vdots$		
$\ddots$	$\vdots$	$X_{a1}$		
$A_{*2*}^\top$	$B_{a**}^\top$	$\vdots$		
$\vdots$	$\ddots$	$X_{as}$		
$A_{*b*}^\top$	$B_{1**}^\top$	$Y_{11}$		
$\ddots$	$\vdots$	$\vdots$		
$A_{*b*}^\top$	$B_{a**}^\top$	$Y_{t1}$		
		$\vdots$		
		$Y_{1b}$		
		$\vdots$		
		$Y_{tb}$		

$C_{111}$
$\vdots$
$C_{11c}$
$\vdots$
$C_{a11}$
$\vdots$
$C_{a1c}$
$\vdots$
$C_{ab1}$
$\vdots$
$C_{abc}$

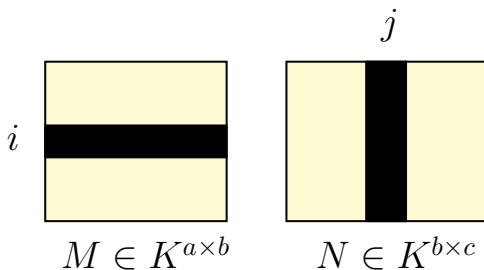
- ▶  $O(n^6)$ , breaks on my laptop after  $n = 50$
- ▶ Schneider conjectures can do no better than  $O(n^6)$
- ▶  $n = 1000$  means  $10^{18}$  FLOPs, exascale computing

Squeezing 5 indices into a single matrix should be a crime.

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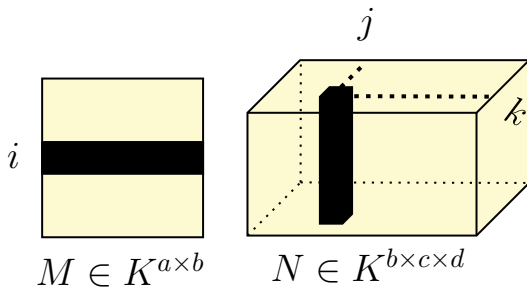
Came from a tensor problem, discovered as a tensor solution, implement as a tensor algorithm.

# Tensor Notation



$$(MN)_{ij} = (M \times_b N)_{ij} = \sum_{k=1}^b M_{ik} N_{kj}$$

# Tensor Notation



$$(M \times_b N)_{ijk} = \sum_{l=1}^b M_{il} N_{ljk}$$

Simultaneous Sylvester System:  $X \times_s A + B \times_t Y = C$

# Tensor Network Diagrams

$$\text{---} \bigcirc M \text{---} \bullet^x = \text{---} \bullet^b$$

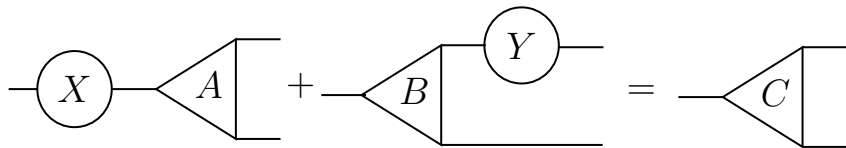
# Tensor Network Diagrams

$$\text{---} \bigcirc M \text{---} \bullet x = \text{---} \bullet b$$

$$\text{---} \bigcirc M \text{---}^b \triangle N \text{---} \quad \text{---} \triangle P \text{---}^b \text{---}^c \text{---} \square Q \text{---}$$

$M \times_b N$   $P \times_{bc} Q$





## Sylvester Equation (1884)

$$XA + BX = C$$

# Sylvester Equation (1884)

$a_{11} + b_{11}$ $\cdots$ $b_{1n}$ $\vdots$ $\ddots$ $\vdots$ $b_{n1}$ $\cdots$ $a_{11} + b_{nn}$	$\cdots$	$a_{1n}$ $\ddots$ $a_{1n}$	$X_{11}$ $\vdots$ $X_{n1}$	$=$	$C_{11}$ $\vdots$ $C_{n1}$
	$\ddots$		$\vdots$		$\vdots$
$a_{n1}$ $\ddots$ $a_{n1}$	$\cdots$	$a_{nn} + b_{11}$ $\cdots$ $b_{1n}$ $\vdots$ $\ddots$ $\vdots$ $b_{n1}$ $\cdots$ $a_{nn} + b_{nn}$	$X_{1n}$ $\vdots$ $X_{nn}$		$C_{1n}$ $\vdots$ $C_{nn}$

$n^2$  variables in  $n^2$  equations. Naively,  $O(n^6)$  as well

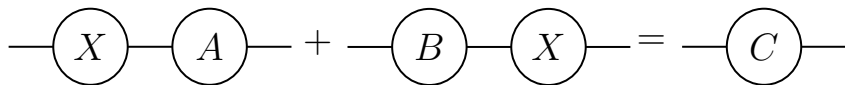
Row operations on this matrix



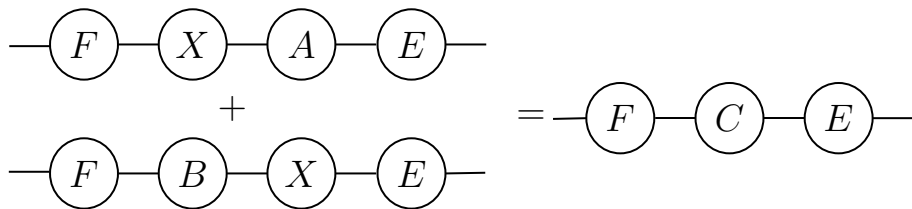
Act on left and right



# Bartels-Stewart (1972)



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$$E^*AE = \begin{bmatrix} * & & \\ \vdots & \ddots & \\ * & \dots & * \end{bmatrix}, F^*BF = \begin{bmatrix} * & \dots & * \\ & \ddots & \vdots \\ & & * \end{bmatrix}$$

Computing  $E, F$  is  $O(n^3)$

# Simultaneous Sylvester System

$$X \times_s A + B \times_t Y = C$$

$A_{*1*}^\top$	$B_{1**}^\top$
$\ddots$	$\vdots$
$A_{*1*}^\top$	$B_{a**}^\top$

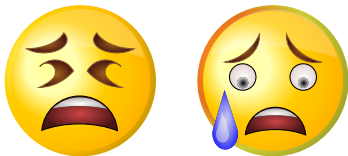
$\vdots$
$X_{1s}$
$\vdots$
$X_{a1}$
$\vdots$
$X_{as}$
$Y_{11}$
$\vdots$
$Y_{t1}$
$\vdots$
$Y_{1b}$
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$\vdots$
$C_{a1c}$
$\vdots$
$C_{ab1}$
$\vdots$
$C_{abc}$

$$=$$



Row operations on this matrix

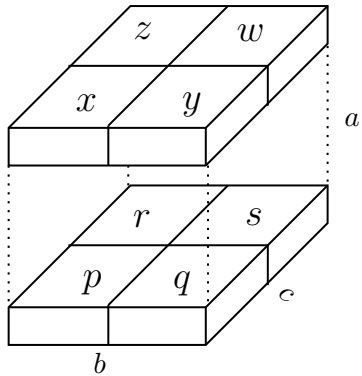


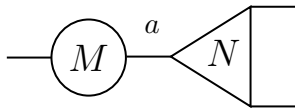
Slice operations on  $A$  and  $B$

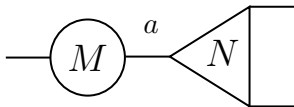
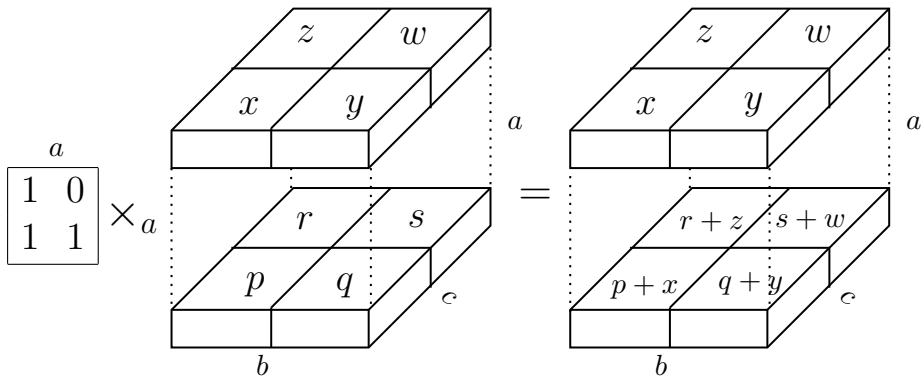


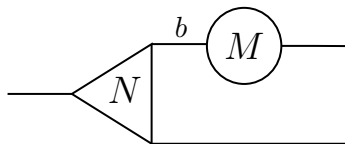
$$M = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

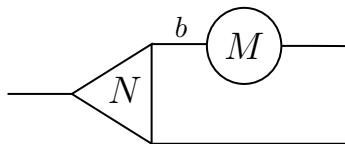
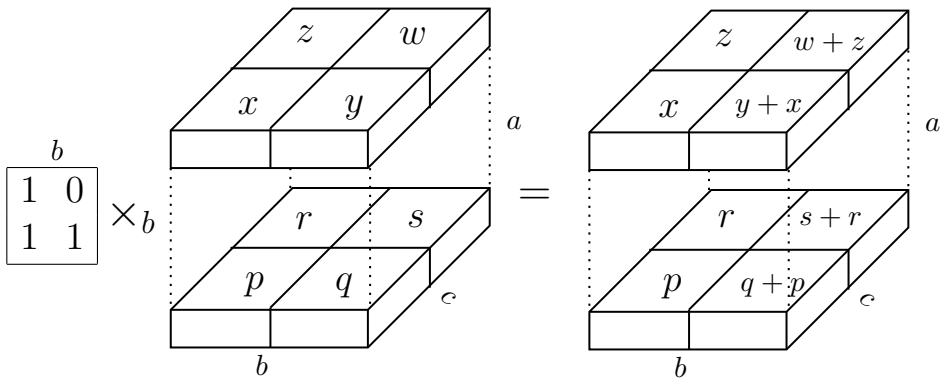
$$N =$$

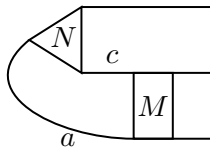




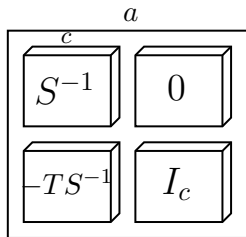




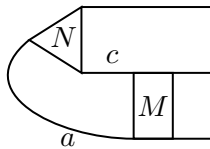


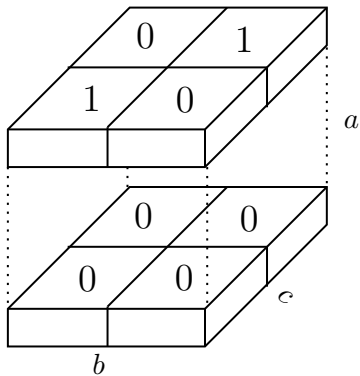




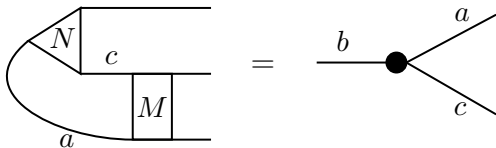


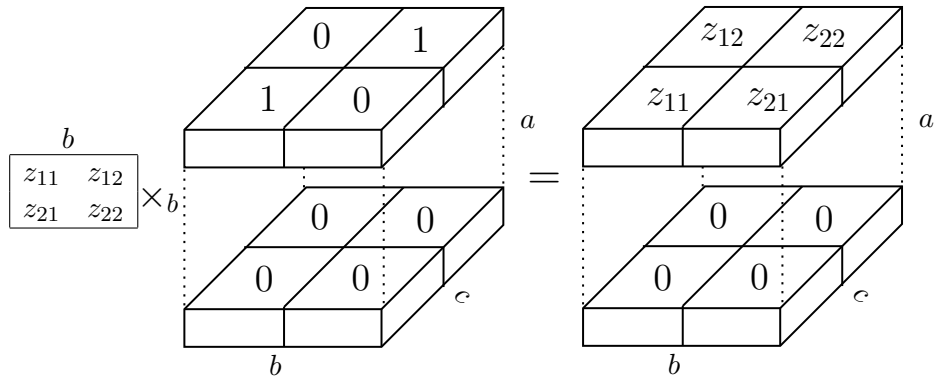
A diagram showing the multiplication of two 2x2 grids of blocks. The first grid has blocks labeled  $S$  and  $T$ . The second grid has blocks labeled  $0$ ,  $1$ ,  $1$ , and  $0$ . The result is a 2x2 grid of blocks labeled  $0$ ,  $0$ ,  $0$ , and  $0$ . The dimensions  $a$ ,  $b$ , and  $c$  are indicated. The operation is labeled  $\times_{ac}$ .

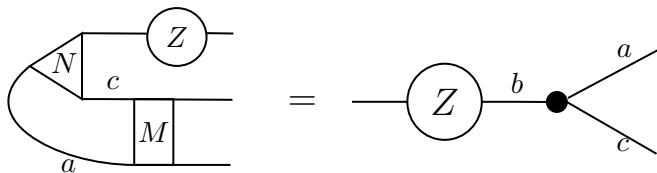
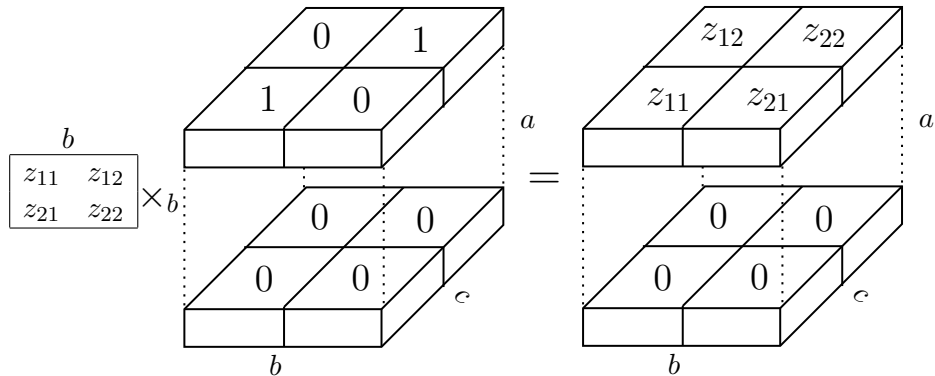


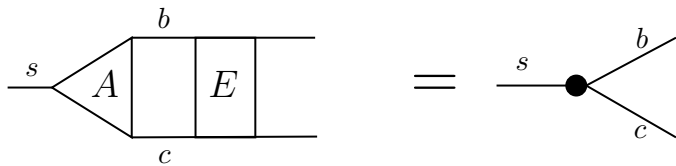


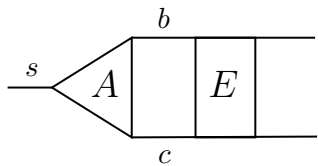
Inert



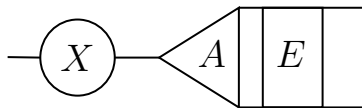
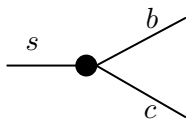




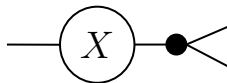


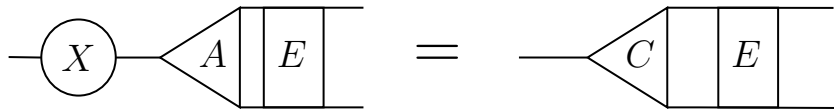
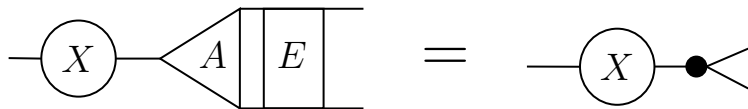
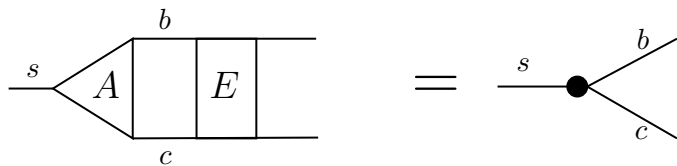


$=$

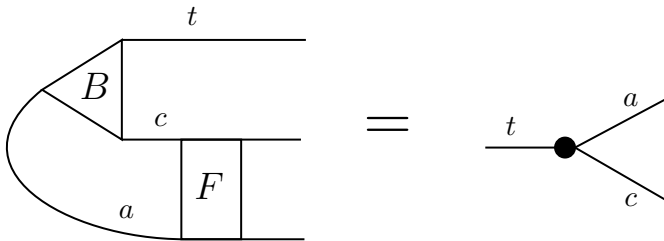


$=$

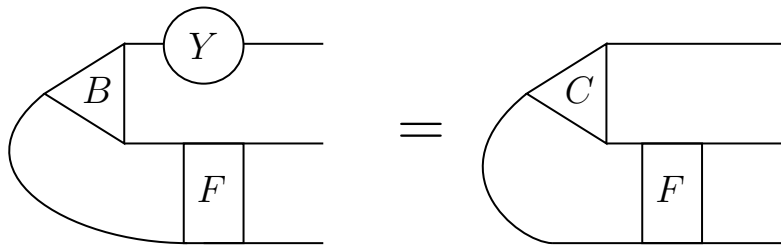




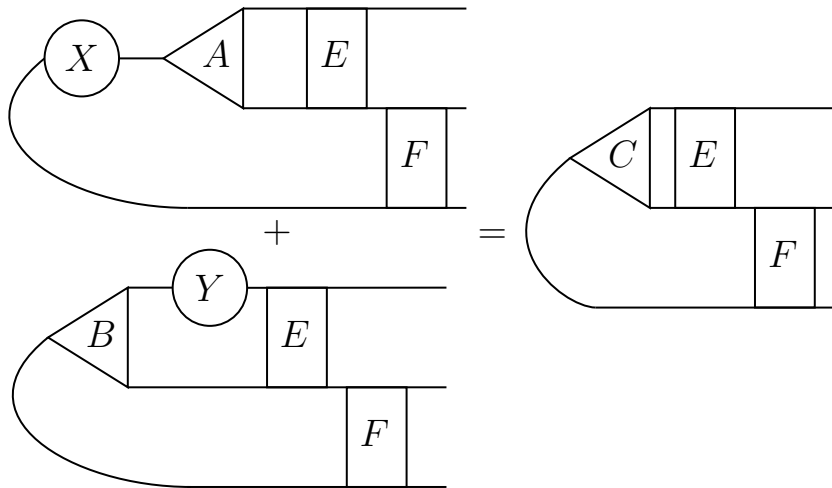
$B = 0$  case solved

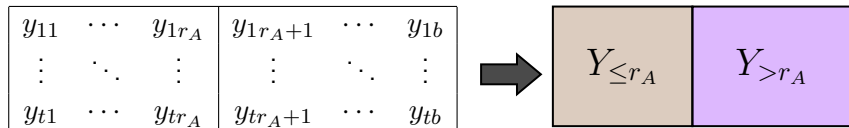
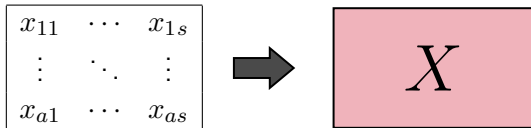






$A = 0$  case solved

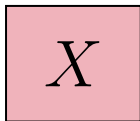


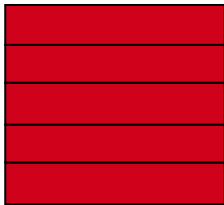
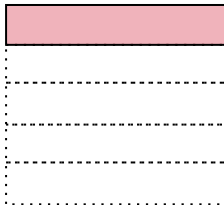


$X \times_s A$ 
 $+$ 
 $B \times_t Y$ 
 $=$ 
 $C$

$$E(\text{stack of 8 red blocks}) = \text{single pink block}$$

same data as



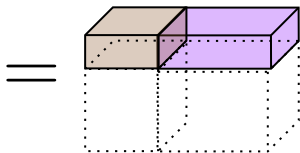
$E^b$  $=$ 

$$\mathcal{A} = [A_{*1*}^\top \ \cdots \ A_{*r_A*}^\top]^\top, \mathcal{A}^\# \mathcal{A} = I_s, \mathcal{A}^\perp \mathcal{A} = 0$$

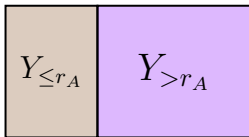
$$\begin{array}{|c|c|} \hline \mathcal{A}^\# \\ \mathcal{A}^\perp \\ \hline -A_{*r_A+1*}^\top \mathcal{A}^\# \\ \vdots \\ -A_{*b*}^\top \mathcal{A}^\# \\ \hline \end{array} \begin{array}{|c|} \hline \\ \hline I_c \\ \vdots \\ I_c \\ \hline \end{array} \left( \begin{array}{|c|} \hline A_{*1*}^\top \\ \vdots \\ A_{*r_A*}^\top \\ \vdots \\ A_{*b*}^\top \\ \hline \end{array} X^\top \right)$$

$$= \begin{array}{|c|} \hline \mathcal{A}^\# \mathcal{A} = I_s \\ \mathcal{A}^\perp \mathcal{A} = 0 \\ (-A_{*r_A+1*}^\top \mathcal{A}^\# \mathcal{A} + A_{*r_A+1*}^\top) = 0 \\ \vdots \\ (-A_{*b*}^\top \mathcal{A}^\# \mathcal{A} + A_{*b*}^\top) = 0 \\ \hline \end{array} X^\top = \begin{array}{|c|} \hline X^\top \\ 0 \\ 0 \\ \vdots \\ 0 \\ \hline \end{array}$$

$$F\left( \begin{array}{|c|c|} \hline \text{brown} & \text{purple} \\ \hline \end{array} \right)$$

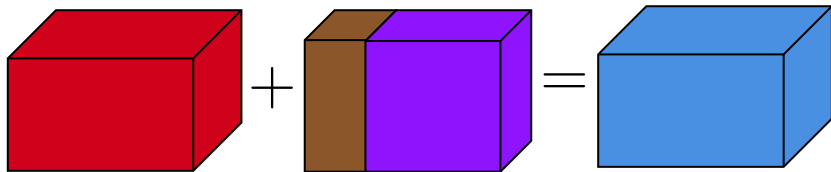


same data as





Instead of solving



The diagram illustrates a tensor addition operation. On the left, a red 3D cube represents the tensor  $X \times_s A$ . This is followed by a plus sign. Next is a 3D cube divided vertically into two parts: a brown left half representing  $B$  and a purple right half representing  $Y$ , together forming the tensor  $B \times_t Y$ . This is followed by an equals sign. On the right is a solid blue 3D cube representing the result tensor  $C$ .

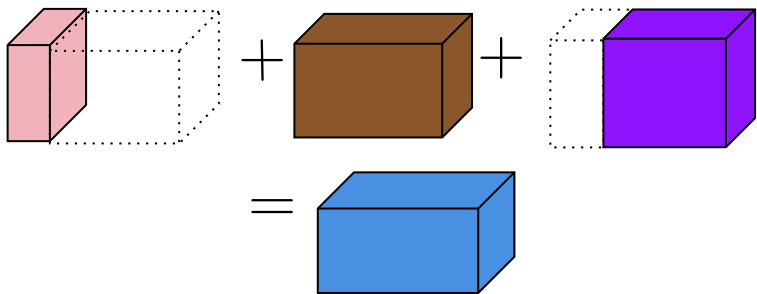
$$X \times_s A + B \times_t Y = C$$

Solve this instead

$$F(E(\text{red cube} + \text{brown and purple cube})) \\ = F(E(\text{blue cube}))$$

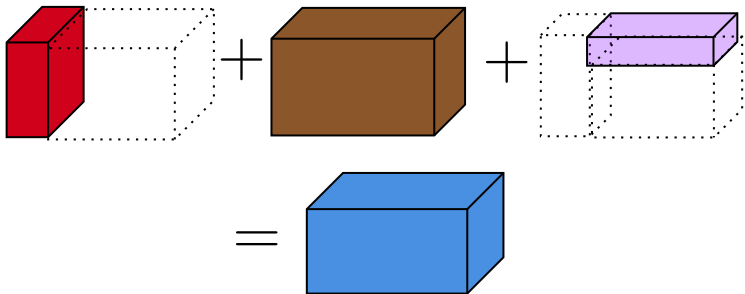
$$E(\text{red cube} + \text{brown cube} + \text{purple cube}) = E(\text{blue cube})$$

The diagram illustrates the principle of superposition in quantum mechanics. The top line shows the expectation value  $E$  of the sum of three cubes: a red cube, a brown cube, and a purple cube. The bottom line shows this is equal to the expectation value  $E$  of a single blue cube. The brown and purple cubes are shown with dotted lines, indicating they are not simultaneously present in a single state.



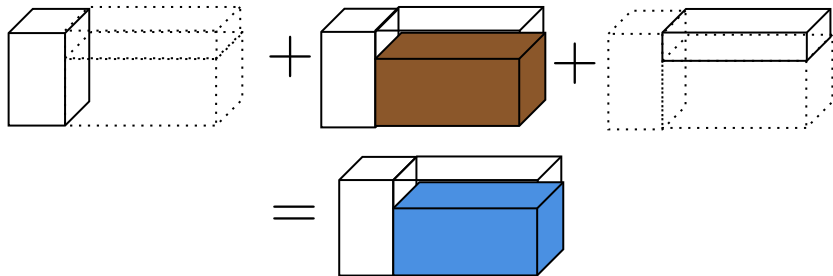
$$E \times_{bc} (X \times_s A + B \times_t Y) = E \times_{bc} C$$

$$F(\text{red cube} + \text{brown cube} + \text{purple cube}) = F(\text{blue cube})$$



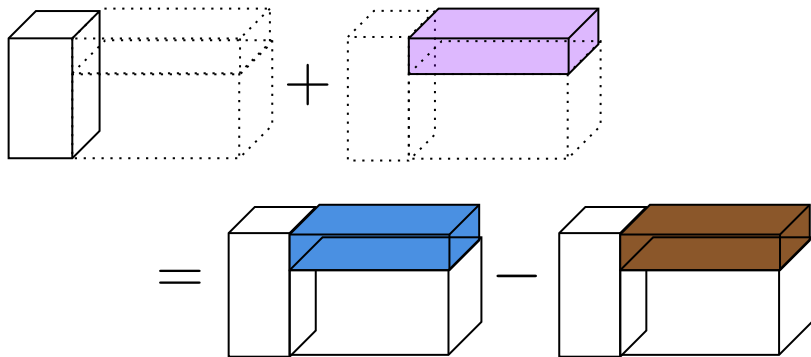
$$F \times_{ac} E \times_{bc} (X \times_s A + B \times_t Y) = F \times_{ac} E \times_{bc} C$$

# Solve linear system for $Y_{\leq r_A}$



$abc - sa - b(t - r_A)$  equations in  $tr_A$  variables. If  $r_A \approx s/c$  and  $s, c$  are  $O(n)$ , equation of  $O(n)$  variables, solved in  $O(n^3)$ .

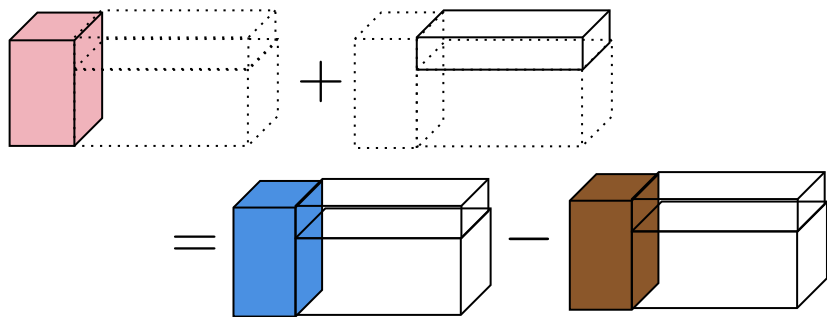
## Back-substitute for $Y_{>r_A}$



Total of  $t(b - r_A)$  coefficients to calculate, each costing  $2(r_A + r_B)c$ , is  $O(n^3)$  under same assumptions.

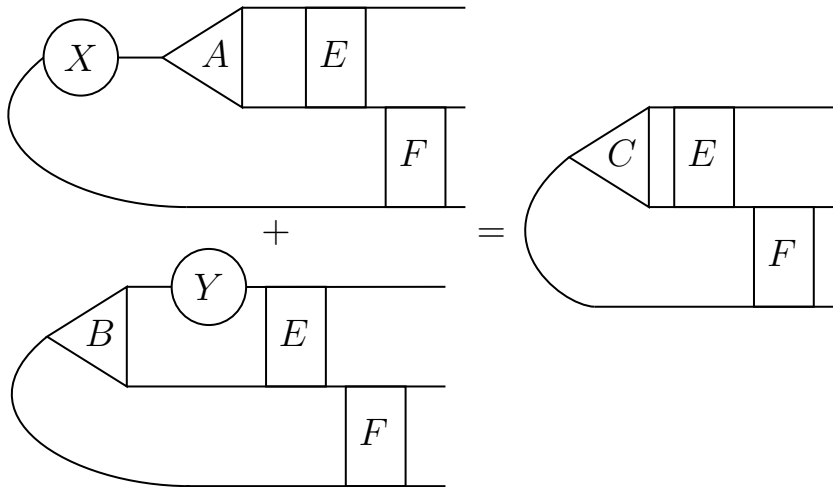


## Back-substitute for $X$

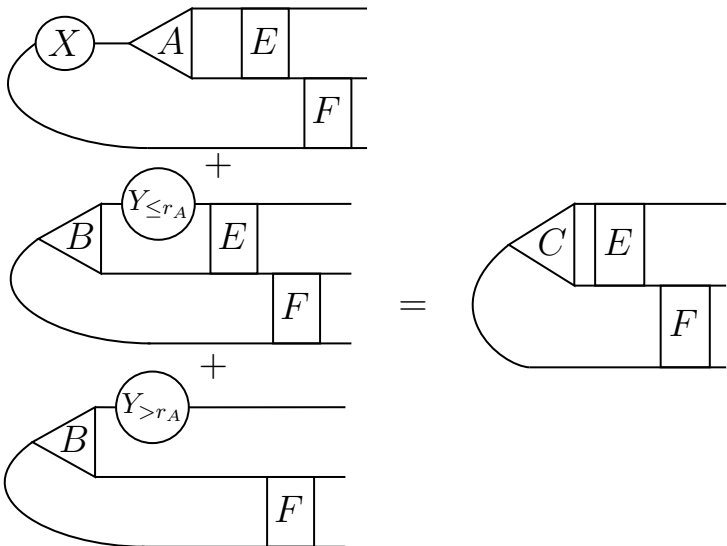


Total of  $sa$  coefficients to calculate, each costing  $2(r_A + r_B)c$ , is  $O(n^3)$  under same assumptions.

# QuickSylver



# QuickSylver - What actually happens



# A zoo of tensor software

ID	Package Name	Functionality					Tensor Type	Platform	Language
		DatM	EWOps	SpecCon	Con	Decomp			
0	Acrotensor [28]	–	–	✓	✓	–	D	C, G	C++
1	AdaTM [54]	–	–	✓	–	✓	S	C	C
2	Boost.uBlas.Tensor [6]	✓	✓	✓	✓	–	D	C	C++
3	BTAS [73]	✓	✓	✓	✓	✓	nan	C	C++
4	COGENT [48]	–	–	✓	✓	–	D	G	Python → CUDA
5	COMET [86]	–	–	✓	✓	–	S	C	C++ → C++
6	CoTenGra [34]	–	–	✓	✓	–	D	C, D, G	Python
7	CP-CALS [75]	–	–	✓	–	✓	D	C, G	C++, Mat <sup>i</sup>
8	CSTF [9]	–	–	–	–	✓	S	D	Scala
9	CuTensor [64]	✓	✓	✓	✓	–	D	G	C, CUDA
10	cuTT [41]	✓	–	–	–	–	D	G	C++, CUDA
11	Cyclops [81]	✓	✓	✓	✓	–	S, D	C, D, G	C++
12	D-Tucker [43]	–	–	–	–	✓	D	C	Matlab
13	DFacTo [15]	–	–	–	–	✓	S	C, D	C++
14	Eigen Tensor [16]	✓	✓	✓	✓	–	D	C, G	C++
15	ExaTN [60]	✓	✓	✓	✓	✓	D	C, D, G	C++, Py <sup>i</sup>
16	Fastor [74]	✓	✓	✓	✓	–	D	C	C++
17	FTensor [52]	✓	✓	✓	✓	–	D	C	C++
18	Genten [72]	–	–	–	–	✓	D, S	C, G	C++
19	GigaTensor [45]	–	–	–	–	✓	S	C, D	Unknown
20	HPTT [85]	✓	–	–	–	–	D	C	C++, Python <sup>i</sup> , C <sup>i</sup>
21	ITensor [29]	–	✓	✓	✓	✓	D, BS	C, G <sup>x</sup>	C++, Julia
22	libtensor [42]	–	–	✓	✓	–	D, BS	C	C++
23	Ltensor [2]	–	–	✓	✓	–	D	C	C++
24	MATLAB [58]	✓	✓	✓	✓	–	D	C	Matlab
25	MultiArray [30]	✓	–	–	–	–	D	C	C++
26	multiway [38]	–	–	–	–	✓	D	C	R
27	N-way toolbox [4]	–	–	–	–	✓	D	C	Matlab
28	NCON [69]	–	–	✓	✓	–	D	C	Matlab
29	netcon [70]	–	–	✓	✓	–	D	C	Matlab

# With ITensor

```
1 FEB = contract(F, E, B)
2 FEA = contract(F, E, A)
3 FEC = contract(F, E, C)
4
5 Y_dense = solve(dense_part(FEB), dense_part(FEC))
6 Y_backsub = rest_of_Y_part(FEC) -
7     contract(rest_of_Y_part(FEB), Y_dense)
8 X_backsub = contract(E, C) - contract(E, B, Y)
```

# Results

- ▶  $A, B, C \in K^{500 \times 500 \times 500}$  solved on my laptop in about 20 seconds
- ▶ Thinner tensor requires solving bigger dense system. (i.e  $A, B, C \in K^{400 \times 400 \times 20}$  solved on my laptop in about 80 seconds)

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- ▶ Features for free: contraction sequence optimization, concurrency, GPUs
- ▶ Caveat: ITensor do not support finite fields, so below are done with Float64.

## Next steps

- ▶ Implement bits of the tensor network abstractions in Magma
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# Next steps

- ▶ Implement bits of the tensor network abstractions in Magma
- ▶ QuickSylver in Magma for module endomorphisms
- ▶ Generalize to higher valence and derivations
- ▶ Investigate into tensor computation software ([survey](#) paper lists 79 packages!!)

Tensor problems deserve tensor solutions.

