

# On fast multiplication of a matrix by its transpose

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# Outline

- 1 Strassen-Winograd fast multiplication algorithm
- 2 Fast computation of  $A \cdot A^T$
- 3 Skew orthogonal matrices
- 4 Cost bounds for block algorithms
- 5 Space and time efficient implementation
- 6 Minimality

## $2 \times 2$ matrix multiplication

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} (A_{11}B_{11} + A_{12}B_{21}) & (A_{11}B_{12} + A_{12}B_{22}) \\ (A_{21}B_{11} + A_{22}B_{21}) & (A_{21}B_{12} + A_{22}B_{22}) \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

Classical Algorithm

8 multiplications, 4 additions

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*[Winograd 1973? 1977]*

7 multiplications, 15 additions

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[Strassen 1969]

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[Winograd 1973? 1977]

7 multiplications, 15 additions



[Hopcroft-Kerr 1969]: 7 multiplications minimum



[Bshouty 1995]: 15 additions minimum  
(for a bilin. alg. with 7 mult.)

# Matrix multiplication by its transpose $A \cdot A^T$

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} A_{11}^T & A_{21}^T \\ A_{12}^T & A_{22}^T \end{bmatrix} = \begin{bmatrix} (A_{11}A_{11}^T + A_{12}A_{12}^T) & (A_{21}A_{21}^T + A_{22}A_{22}^T) \\ (A_{21}A_{11}^T + A_{22}A_{12}^T) & (A_{21}A_{21}^T + A_{22}A_{22}^T) \end{bmatrix} = \begin{bmatrix} C_{11} & C_{21}^T \\ C_{21} & C_{22} \end{bmatrix}$$

Divide & Conquer Algorithm

6 multiplications, 3 additions

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Divide & Conquer Algorithm

6 multiplications, 3 additions

here (over  $\mathbb{C}$  & over any field  
in  $> 0$  characteristic)

5 multiplications, 7.5 additions



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# From Strassen-Winograd multiplication algorithm

**Require:**  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  and  $B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$ ;

**Ensure:**  $C = A \cdot B$

① 8 additions:

$$\begin{aligned} s_1 &\leftarrow a_{11} - a_{21}, & s_2 &\leftarrow a_{21} + a_{22}, & s_3 &\leftarrow s_2 - a_{11}, & s_4 &\leftarrow a_{12} - s_3, \\ t_1 &\leftarrow b_{22} - b_{12}, & t_2 &\leftarrow b_{12} - b_{11}, & t_3 &\leftarrow b_{11} + t_1, & t_4 &\leftarrow b_{21} - t_3. \end{aligned}$$

② 7 recursive multiplications:

$$\begin{aligned} p_1 &\leftarrow a_{11} \cdot b_{11}, & p_2 &\leftarrow a_{12} \cdot b_{21}, & p_3 &\leftarrow a_{22} \cdot t_4, & p_4 &\leftarrow s_1 \cdot t_1, \\ p_5 &\leftarrow s_3 \cdot t_3, & p_6 &\leftarrow s_4 \cdot b_{22}, & p_7 &\leftarrow s_2 \cdot t_2. \end{aligned}$$

③ 7 final additions:

$$\begin{aligned} c_1 &\leftarrow p_1 + p_5, & c_2 &\leftarrow c_1 + p_4, & c_3 &\leftarrow p_1 + p_2, & c_4 &\leftarrow c_2 + p_3, \\ c_5 &\leftarrow c_2 + p_7, & c_6 &\leftarrow c_1 + p_7, & c_7 &\leftarrow c_6 + p_6. \end{aligned}$$

④ **return**  $C = \begin{bmatrix} c_3 & c_7 \\ c_4 & c_5 \end{bmatrix}$ .

... to its specialization to  $A \cdot A^T$  ...

**Require:**  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  with  $A^T = \begin{bmatrix} a_{11}^T & a_{21}^T \\ a_{12}^T & a_{22}^T \end{bmatrix}$ ;

**Ensure:**  $C = A \cdot A^T$

① 6 additions:

$$\begin{aligned} s_1 &\leftarrow a_{11} - a_{21}, & s_2 &\leftarrow a_{21} + a_{22}, & s_3 &\leftarrow s_2 - a_{11}, & \cancel{s_4 \leftarrow a_{12} - s_3}, \\ t_1 &\leftarrow a_{22}^T - a_{21}^T, & \cancel{t_2 \leftarrow a_{21}^T - a_{11}^T}, & t_3 &\leftarrow a_{11}^T + t_1, & t_4 &\leftarrow a_{12}^T - t_3. \end{aligned}$$

② 6 multiplications (2 recursive, 4 general):

$$\begin{aligned} p_1 &\leftarrow a_{11} \cdot a_{11}^T, & p_2 &\leftarrow a_{12} \cdot a_{12}^T, & p_3 &\leftarrow a_{22} \cdot t_4, & p_4 &\leftarrow s_1 \cdot t_1, \\ p_5 &\leftarrow s_3 \cdot t_3, & \cancel{p_6 \leftarrow s_4 \cdot a_{22}^T}, & p_7 &\leftarrow s_2 \cdot \boxed{s_1^T}. \end{aligned}$$

③ 5 final additions:

$$\begin{aligned} c_1 &\leftarrow p_1 + p_5, & c_2 &\leftarrow c_1 + p_4, & c_3 &\leftarrow p_1 + p_2, & c_4 &\leftarrow c_2 + p_3, \\ c_5 &\leftarrow c_2 - p_7, & \cancel{c_6 \leftarrow c_1 - p_7}, & \cancel{c_7 \leftarrow c_6 + p_6}. \end{aligned}$$

④ **return**  $C = \begin{bmatrix} c_3 & \cancel{c_4} \\ c_4 & c_5 \end{bmatrix}$ .

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**Require:**  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  with  $A^T = \begin{bmatrix} a_{11}^T & a_{21}^T \\ a_{12}^T & a_{22}^T \end{bmatrix}$ ;

**Ensure:**  $C = A \cdot A^T$

☹ all variants have sign discrepancies

① 6 additions:

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## ... with a parameterized algorithm

**Require:**  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  and  $Y$  s.t.  $YY^T = -I_{n/2}$ ;

**Ensure:**  $C = A \cdot A^T$

- ① 4 additions and 2 multiplications by  $Y$ :

$$\begin{array}{l}
 s_1 \leftarrow (a_{21} - a_{11}) \boxed{Y}, \quad s_2 \leftarrow a_{22} - a_{21} \boxed{Y}, \quad s_3 \leftarrow s_1 - a_{22}, \quad s_4 \leftarrow s_3 + a_{12}, \\
 \del{t_2 \leftarrow \boxed{Y^T} (a_{21}^T - a_{11}^T)} \quad \del{t_3 \leftarrow t_2 - a_{22}^T} \quad \del{t_4 \leftarrow t_3 + a_{12}^T}
 \end{array}$$

- ② 5 multiplications (3 recursive, 2 general):

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 p_1 \leftarrow a_{11} \cdot a_{11}^T, \quad p_2 \leftarrow a_{12} \cdot a_{12}^T, \quad p_3 \leftarrow a_{22} \cdot \boxed{s_4^T}, \quad p_4 \leftarrow s_1 \cdot \boxed{s_2^T}, \\
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 \end{array}$$

- ③ 5 final additions:

$$\begin{array}{l}
 c_1 \leftarrow p_1 + p_5, \quad c_2 \leftarrow c_1 + p_4, \quad c_3 \leftarrow p_1 + p_2, \quad c_4 \leftarrow c_2 + p_3, \\
 c_5 \leftarrow c_2 + \boxed{p_4^T}.
 \end{array}$$

- ④ **return**  $C = \begin{bmatrix} c_3 & \\ c_4 & c_5 \end{bmatrix}$ .

# Fast algorithm for $A \cdot A^T$ exploiting symmetries

**Require:**  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  and  $Y$  s.t.  $YY^T = -I_{n/2}$ ;

**Ensure:**  $C = A \cdot A^T$

- ① 4 additions and 2 multiplications by  $Y$ :

$$s_1 \leftarrow (a_{21} - a_{11})Y, \quad s_2 \leftarrow a_{22} - a_{21}Y, \quad s_3 \leftarrow s_1 - a_{22}, \quad s_4 \leftarrow s_3 + a_{12},$$

- ② 5 multiplications (3 recursive, 2 general):

$$p_1 \leftarrow a_{11} \cdot a_{11}^T, \quad p_2 \leftarrow a_{12} \cdot a_{12}^T, \quad p_3 \leftarrow a_{22} \cdot s_4^T, \quad p_4 \leftarrow s_1 \cdot s_2^T, \\ p_5 \leftarrow s_3 \cdot s_3^T.$$

- ③ 2 complete and 3 symmetric additions:

$$\boxed{Low(c_1) \leftarrow Low(p_1) + Low(p_5)}, \quad c_2 \leftarrow c_1 + p_4, \quad \boxed{Low(c_3) \leftarrow Low(p_1) + Low(p_2)},$$

$$\boxed{Low(c_5) \leftarrow Low(c_2) + Low(p_4^T)}, \quad c_4 \leftarrow c_2 + p_3.$$

- ④ **return**  $C = \begin{bmatrix} c_3 & \\ c_4 & c_5 \end{bmatrix}$ .

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Definition (Skew orthogonal matrix)

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➡ no-op: swap real/imaginary & 1 sign

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✓ Other fields with positive characteristic,  $n \geq 4$ :  $Y = \begin{pmatrix} a & b \\ -b & a \end{pmatrix} \otimes I_{n/2} = \begin{pmatrix} aI_{n/2} & bI_{n/2} \\ -bI_{n/2} & aI_{n/2} \end{pmatrix} \in \mathbb{F}_q^{n \times n}$

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## Proof.

$$\begin{pmatrix} a & b \\ -b & a \end{pmatrix} \begin{pmatrix} a & -b \\ b & a \end{pmatrix} = \begin{pmatrix} a^2 + b^2 & 0 \\ 0 & a^2 + b^2 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \text{ if } \boxed{a^2 + b^2 = -1} \quad \square$$

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# Leading terms of the cost bounds for matrix multiplication

Definition (Leading term of a cost bound for matrix multiplication)

$$\text{MM}_\omega(n) = C_\omega n^\omega$$

denotes the leading term in the arithmetic cost of an algorithm multiplying two  $n \times n$  matrices (in  $\text{MM}_\omega(n) + o(\text{MM}_\omega(n))$ ).

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## Example

Classical method	$\text{MM}_3(n)$	$= 2n^3$
Strassen's algorithm	$\text{MM}_{\log_2 7}(n)$	$= 7n^{\log_2 7}$ with $\log_2(7) \approx 2.807$
Strassen-Winograd algorithm	$\text{MM}_{\log_2 7}(n)$	$= 6n^{\log_2 7}$ with $\log_2(7) \approx 2.807$
Strassen-Winograd with base case at a threshold	$\begin{cases} C(n) \leq 7C(\lceil n/2 \rceil) + 15\lceil n/2 \rceil^2 \\ C(k) = 2k^3 \text{ for } k \lesssim 1000 \end{cases}$	
⋮		

😊 [LeGall 2014]

$$\omega < 2.3728639$$

# Leading terms of the arithmetic cost bounds for $A \cdot A^T$

- Classical method:  $\frac{1}{2}MM_3(n) = n^3$

- Our algorithm:

$$\rightarrow S(n) \leq \boxed{3}S(\lceil n/2 \rceil) + \boxed{2}MM_\omega(\lceil n/2 \rceil) + \left( 7.5 + \left\{ \begin{array}{c} 0 \\ 2 \\ 4 \\ 6 \end{array} \right\} \right) \lceil n/2 \rceil^2 \quad \text{or} \quad S(k) = k^3 \text{ for } k \lesssim 2000$$

$$\rightarrow \frac{2}{2^\omega - 3} MM_\omega(n)$$

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Problem	Algorithm	$\omega = 3$
$A \cdot A^T \in \mathbb{F}^{n \times n}$	D & C	$n^3$
	<b>here</b>	$0.8n^3$

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Problem	Algorithm	$\omega = 3$	$\omega = \log_2(7)$
$A \cdot A^T \in \mathbb{F}^{n \times n}$	D & C	$n^3$	$\frac{2}{3} MM_{\log_2(7)}(n)$
	<b>here</b>	$0.8n^3$	$\frac{1}{2} MM_{\log_2(7)}(n)$

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	<b>here</b>	$0.8n^3$	$\frac{1}{2} MM_{\log_2(7)}(n)$	$\frac{2}{2^\omega - 3} MM_\omega(n)$

# Leading terms of the cost bounds over $\mathbb{C}$

Counting field ops over  $\mathbb{R}$

Problem	Algorithm	$\omega = 3$	$\omega = \log_2(7)$	$\omega$
$A \cdot B \in \mathbb{C}^{n \times n}$	naive	$8n^3$	$4 \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$4 \text{MM}_{\omega}^{\mathbb{R}}(n)$
$(U + iV) \cdot (W + iX)$	Karatsuba: 3M	$6n^3$	$3 \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$3 \text{MM}_{\omega}^{\mathbb{R}}(n)$

\*

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$(U + iV) \cdot (W + iX)$	Karatsuba: 3M	$6n^3$	$3 \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$3 \text{MM}_{\omega}^{\mathbb{R}}(n)$
$A \cdot A^T \in \mathbb{C}^{n \times n}$	2M	$4n^3$	$2 \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$2 \text{MM}_{\omega}^{\mathbb{R}}(n)^*$
	D & C	$3n^3$	$2 \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$\frac{6}{2^{\omega} - 4} \text{MM}_{\omega}^{\mathbb{R}}(n)$
	<b>here</b>	$2.4n^3$	$\frac{3}{2} \text{MM}_{\log_2(7)}^{\mathbb{R}}(n)$	$\frac{6}{2^{\omega} - 3} \text{MM}_{\omega}^{\mathbb{R}}(n)$

\*If  $\omega < \log_2(6) \approx 2.585$ , then  $2 < \frac{6}{2^{\omega} - 3}$

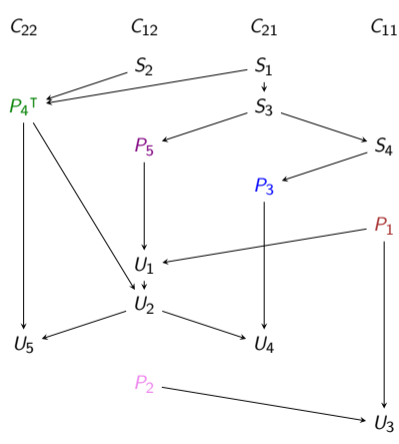


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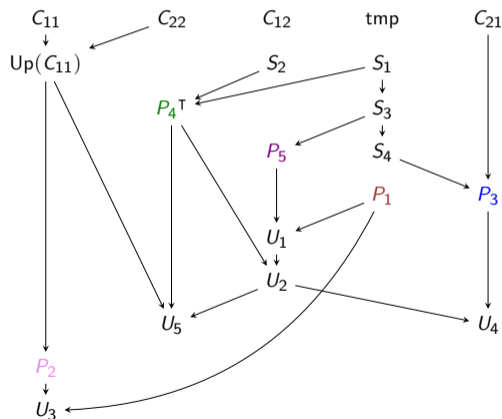
## Scheduling

$$C \leftarrow \alpha A \cdot A^T$$



#	operation	location
1	$S_1 = (A_{21} - A_{11}) \cdot Y$	$C_{21}$
2	$S_2 = A_{22} - A_{21} \cdot Y$	$C_{12}$
3	$P_4^T = S_2 \cdot S_1^T$	$C_{22}$
4	$S_3 = S_1 - A_{22}$	$C_{21}$
5	$P_5 = S_3 \cdot S_3^T$	$C_{12}$
6	$S_4 = S_3 + A_{12}$	$C_{11}$
7	$P_3 = A_{22} \cdot S_4^T$	$C_{21}$
8	$P_1 = A_{11} \cdot A_{11}^T$	$C_{11}$
9	$U_1 = P_1 + P_5$ $\text{Up}(U_1) = \text{Low}(U_1)^T$	$C_{12}$
10	$U_2 = U_1 + P_4^T$	$C_{12}$
11	$U_4 = U_2 + P_3$	$C_{21}$
12	$U_5 = U_2 + P_4^T$	$C_{22}$
13	$P_2 = A_{12} \cdot A_{12}^T$	$C_{12}$
14	$U_3 = P_1 + P_2$	$C_{11}$

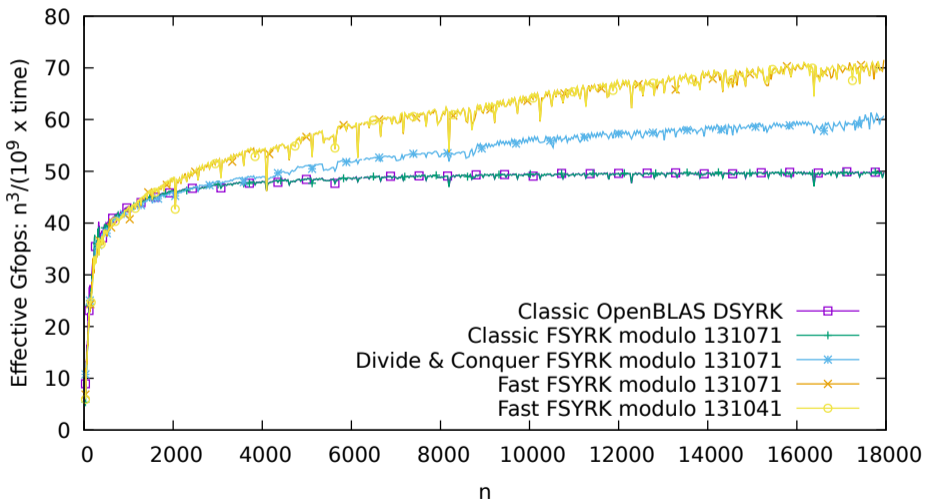
## Scheduling

$$C \leftarrow \alpha A \cdot A^T + \beta C$$
 with one temporary block


#	operation	location
1	$S_1 = (A_{21} - A_{11}) \cdot Y$	tmp ( $\frac{n}{2} \times \frac{n}{2}$ )
2	$S_2 = A_{22} - A_{21} \cdot Y$	$C_{12}$
	$Up(C_{11}) = Low(C_{22})^T$	$C_{11}$
3	$P_4^T = \alpha S_2 \cdot S_1^T$	$C_{22}$
4	$S_3 = S_1 - A_{22}$	tmp
5	$P_5 = \alpha S_3 \cdot S_3^T$	$C_{12}$
6	$S_4 = S_3 + A_{12}$	tmp
7	$P_3 = \alpha A_{22} \cdot S_4^T + \beta C_{21}$	$C_{21}$
8	$P_1 = \alpha A_{11} \cdot A_{11}^T$	tmp
9	$U_1 = P_1 + P_5$	$C_{12}$
	$Up(U_1) = Low(U_1)^T$	$C_{12}$
10	$U_2 = U_1 + P_4$	$C_{12}$
11	$U_4 = U_2 + P_3$	$C_{21}$
12	$U_5 = U_2 + P_4^T + \beta Up(C_{11})^T$	$C_{22}$
13	$P_2 = \alpha A_{12} \cdot A_{12}^T + \beta C_{11}$	$C_{11}$
14	$U_3 = P_1 + P_2$	$C_{11}$

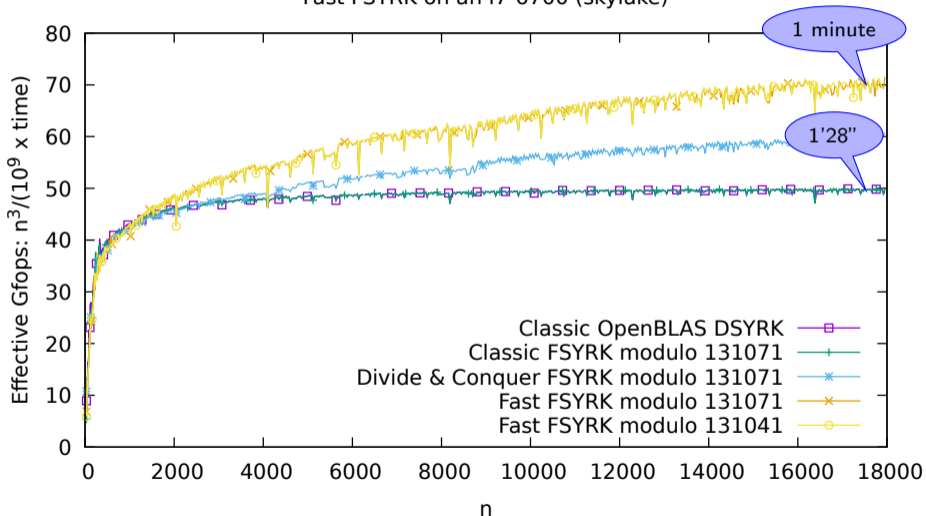
# Speed: LINBOX/FFLAS-FFPACK

Fast FSYRK on an i7-6700 (skylake)



# Speed: LINBOX/FFLAS-FFPACK

Fast FSYRK on an i7-6700 (skylake)



# Outline

- 1 Strassen-Winograd fast multiplication algorithm
- 2 Fast computation of  $A \cdot A^T$
- 3 Skew orthogonal matrices
- 4 Cost bounds for block algorithms
- 5 Space and time efficient implementation
- 6 Minimality**

# On minimality of bilinear algorithms computing $A \cdot A^T$

## Conjecture (5 multiplications)

- 1 *5 multiplications required*
- 2  *$5 = 3$  recursive calls + 2 general multiplications*

## Conjecture (Skew orthogonal)

*All variants with 5 multiplications require a skew orthogonal matrix*

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## Theorem (non-commutative)

Non-commutative  $2 \times 2$  variants with 5 multiplications require at least 11 additions

## Conjecture (9 additions)

All block variants with 5 multiplications require at least 9 block additions



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## Conjecture (9 additions)

All block variants with 5 multiplications require at least 9 block additions

➔ With symmetries of the blocks this is reduced to  $\left(4 + 2 + \frac{3}{2}\right) n^2 = 7.5n^2$  additions

# Strassen tensor

Example: Winograd's variant

$$p_1 = a_{11} \cdot b_{11}$$

$$p_2 = a_{12} \cdot b_{21}$$

$$p_3 = (a_{11} - a_{21}) \cdot (b_{22} - b_{12})$$

$$p_4 = (a_{21} + a_{22}) \cdot (b_{12} - b_{11})$$

$$p_5 = (a_{11} + a_{12} - a_{21} - a_{22}) \cdot b_{22}$$

$$p_6 = a_{22}(b_{12} - b_{11} + b_{21} - b_{22})$$

$$p_7 = (a_{21} - a_{11} + a_{22}) \cdot (b_{11} - b_{12} + b_{22})$$

$$c_{11} = p_1 + p_2$$

$$c_{12} = p_1 + p_4 + p_5 + p_7$$

$$c_{21} = p_1 + p_3 + p_6 + p_7$$

$$c_{22} = p_1 + p_3 + p_4 + p_7$$

$$\begin{aligned} \sum_{i=1}^7 S_{i1} \otimes S_{i2} \otimes S_{i3} &= \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} + \\ & \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & -1 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} + \\ & \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} \otimes \begin{pmatrix} -1 & 1 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} 1 & 1 \\ -1 & -1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + \\ & \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + \begin{pmatrix} -1 & 0 \\ 1 & 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \end{aligned}$$

# Strassen tensor

Example: Winograd's variant

$$p_1 = a_{11} \cdot b_{11}$$

$$p_2 = a_{12} \cdot b_{21}$$

$$p_3 = (a_{11} - a_{21}) \cdot (b_{22} - b_{12})$$

$$p_4 = (a_{21} + a_{22}) \cdot (b_{12} - b_{11})$$

$$p_5 = (a_{11} + a_{12} - a_{21} - a_{22}) \cdot b_{22}$$

$$p_6 = a_{22}(b_{12} - b_{11} + b_{21} - b_{22})$$

$$p_7 = (a_{21} - a_{11} + a_{22}) \cdot (b_{11} - b_{12} + b_{22})$$

$$c_{11} = p_1 + p_2$$

$$c_{12} = p_1 + p_4 + p_5 + p_7$$

$$c_{21} = p_1 + p_3 + p_6 + p_7$$

$$c_{22} = p_1 + p_3 + p_4 + p_7$$

$$\sum_{i=1}^7 S_{i1} \otimes S_{i2} \otimes S_{i3} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} +$$

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & -1 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} +$$

$$\begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} \otimes \begin{pmatrix} -1 & 1 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} 1 & 1 \\ -1 & -1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} +$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + \begin{pmatrix} -1 & 0 \\ 1 & 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$$

Theorem (📄 [de Groot 1978])

All the 7 mult. variants ( $U, V, W$  unimodular) are given by:

$$\left( U^{-1} \cdot S_{i1} \cdot V \right) \otimes \left( V^{-1} \cdot S_{i2} \cdot W \right) \otimes \left( W^{-1} \cdot S_{i3} \cdot U \right)$$

# Isotropy group and Gröbner basis: multiplications


5 multiplications for  $A \cdot A^T$ ?

- **12 unknowns:**  $U = \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \end{bmatrix}$ ,  $V = \begin{bmatrix} v_{11} & v_{12} \\ v_{21} & v_{22} \end{bmatrix}$ ,  $W = \begin{bmatrix} w_{11} & w_{12} \\ w_{21} & w_{22} \end{bmatrix}$ .
- **31 equations of degree 4**
  - $3 * 4$  equations: recursive calls  $p_j = X \cdot X^T$  and
  - $2 * 2 * 4$  equations: two symmetries  $p_j = X \cdot Y^T = (Y \cdot X^T)^T = p_k^T$
  - 3 equations: unimodularity of  $U, V, W$

# Isotropy group and Gröbner basis: multiplications

5 multiplications for  $A \cdot A^T$ ?

$$\text{Gröbner basis via FGB} \left\{ \begin{array}{l} z = v_{11} - v_{21} \\ v_{11} = z + v_{21} \\ v_{22} = (2v_{21}(v_{21} + z) - 1)v_{21} + z^3 \\ v_{12} = -(v_{21}^2 + (v_{21} + z)^2 + 1)v_{21} - z \\ u_{11} = -((z + v_{21})^2 + v_{21}^2)(w_{21} + w_{22}) \\ u_{21} = -((z + v_{21})^2 + v_{21}^2)(w_{11} + w_{12}) \\ u_{12} = -((z + v_{21})^2 + v_{21}^2)w_{22} \\ u_{22} = ((z + v_{21})^2 + v_{21}^2)w_{12} \\ 1 = w_{11}w_{22} - w_{12}w_{21} \\ 0 = ((z + v_{21})^2 + v_{21}^2)^2 + 1^* \end{array} \right.$$


  $z^2 = \sqrt{-1}$ ,  $v_{11} = 1$ ,  $v_{21} = 0$ ,  $w_{11} = 1$ ,  $w_{12} = 0$ ,  $w_{21} = 0$ ; then scale by  $1/z$   $\rightarrow$  our algorithm

\* $0 = ((z + v_{21})^2 + v_{21}^2)^2 + 1$  imposes a square root of  $-1$


# Perspective

- Minimality of multiplications conjectures: reduce polynomial system?
  - ✓ 5 multiplications **are** required
  - ✓ except, 4 multiplications for: ( $p=2$ ) & ( $2 \times 2$ ) & (commutative)
  - 😊 conjectures verified  $\forall$  isotropies of Strassen-Winograd ...

# Perspective

- Minimality of multiplications conjectures: reduce polynomial system?
-  [Schwartz et al. 2017, 2019] for  $A \cdot A^T$ ?
  - 😊  $O(n^2 \log(n))$  non matrix based pre-computations → only 12 additions
    - 😊  $\geq 5$  recursive levels: fewer operations than Strassen-Winograd for  $A \cdot B$
    - 😞 Probably for matrices larger than  $64k \dots$

# Perspective

- Minimality of multiplications conjectures: reduce polynomial system?
-  [Schwartz et al. 2017, 2019] for  $A \cdot A^T$ ?
- Accelerating  $LDL^T$  in practice for arbitrary rank profile?