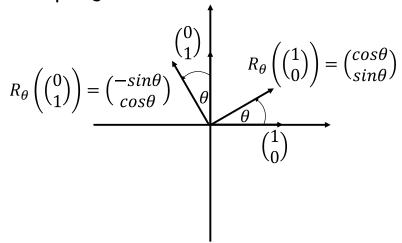
Matrix-Matrix Multiplication

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Opening Remarks

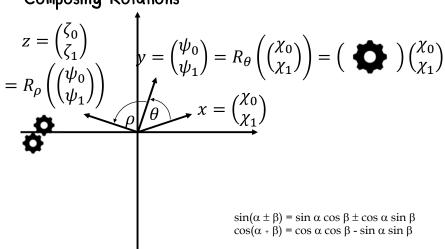
Composing Rotations



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Opening Remarks

Composing Rotations



Observations

- Partitioned Matrix-Matrix Multiplication
- Properties Transposing a Product of Matrices
- Matrix-Matrix Multiplication with Special Matrices

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Observations

Partitioned Matrix-Matrix Multiplication

$$C = AB$$

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Observations

Partitioned Matrix-Matrix Multiplication

$$(C_0 \mid C_1) = A(B_0 \mid B_1)$$

$$\begin{array}{c} \bullet \quad \begin{pmatrix} \gamma_{0,0} & \gamma_{0,1} \\ \gamma_{1,0} & \gamma_{1,1} \\ \gamma_{2,0} & \gamma_{2,1} \\ \gamma_{3,0} & \gamma_{3,1} \end{pmatrix} \begin{pmatrix} \gamma_{0,2} & \gamma_{0,3} \\ \overline{\gamma_{1,2}} & \gamma_{1,3} \\ \gamma_{2,2} & \gamma_{2,3} \\ \gamma_{3,2} & \gamma_{3,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \alpha_{0,2} & \alpha_{0,3} \\ \overline{\alpha_{1,0}} & \alpha_{1,1} & \alpha_{1,2} & \alpha_{1,3} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \beta_{3,0} & \beta_{3,1} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \alpha_{0,2} & \alpha_{0,3} \\ \alpha_{1,0} & \alpha_{1,1} & \alpha_{1,2} & \alpha_{1,3} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,2} & \beta_{0,3} \\ \beta_{1,2} & \beta_{1,3} \\ \beta_{1,2} & \beta_{1,3} \\ \beta_{2,2} & \beta_{2,2} \\ \beta_{3,2} & \beta_{3,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \alpha_{0,2} & \alpha_{0,3} \\ \alpha_{1,0} & \alpha_{1,1} & \alpha_{1,2} & \alpha_{1,3} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \beta_{3,2} & \beta_{3,2} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \alpha_{1,0} & \alpha_{1,1} & \alpha_{1,2} & \alpha_{1,3} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \beta_{3,2} & \beta_{3,2} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \beta_{3,2} & \beta_{3,2} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \alpha_{2,0} & \alpha_{2,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{1,0} & \beta_{1,1} \\ \beta_{2,0} & \beta_{2,1} \\ \beta_{3,2} & \beta_{3,2} \end{pmatrix} \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \alpha_{0,2} & \alpha_{0,3} \\ \alpha_{0,0} &$$

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Observations

Partitioned Matrix-Matrix Multiplication

$$\left(\frac{C_0}{C_1}\right) = \left(\frac{A_0}{A_1}\right)B$$

$$\bullet \quad \left(\frac{\gamma_{0,0} \quad \gamma_{0,1} \quad \gamma_{0,2} \quad \gamma_{0,3}}{\gamma_{1,0} \quad \gamma_{1,1} \quad | \gamma_{1,2} \quad \gamma_{1,3}}}{\gamma_{2,0} \quad \gamma_{2,1} \quad \gamma_{2,2} \quad \gamma_{2,3}}}{\gamma_{3,0} \quad \gamma_{3,1} \quad \gamma_{3,2} \quad \gamma_{3,3}} \right) = \frac{\begin{pmatrix} \alpha_{0,0} \quad \alpha_{0,1} \quad \alpha_{0,2} \quad \alpha_{0,3} \\ \overline{\alpha_{1,0} \quad \alpha_{1,1} \quad \alpha_{1,2} \quad \alpha_{1,3}} \\ \alpha_{2,0} \quad \alpha_{2,1} \quad \alpha_{2,2} \quad \alpha_{2,3} \\ \alpha_{3,0} \quad \alpha_{3,1} \quad \alpha_{3,2} \quad \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{1,0} \quad \beta_{1,1} \\ \beta_{2,0} \quad \beta_{2,1} \\ \beta_{3,0} \quad \beta_{3,1} \end{pmatrix} \begin{pmatrix} \beta_{0,3} \\ \beta_{1,3} \\ \beta_{2,2} \\ \beta_{3,0} \quad \beta_{3,1} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0,1} \\ \beta_{1,0} \quad \beta_{1,1} \\ \beta_{2,0} \quad \beta_{2,1} \\ \beta_{2,2} \\ \beta_{3,0} \quad \beta_{3,1} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{1,0} \quad \beta_{1,1} \\ \beta_{2,0} \quad \beta_{2,1} \\ \beta_{2,2} \\ \beta_{3,0} \quad \beta_{3,1} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{1,0} \quad \beta_{1,1} \\ \beta_{2,0} \quad \beta_{2,1} \\ \beta_{2,2} \\ \beta_{2,3} \\ \beta_{3,2} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{1,2} \quad \beta_{1,3} \\ \beta_{2,2} \quad \beta_{2,3} \\ \beta_{2,3} \\ \beta_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0,1} \\ \beta_{2,1} \quad \beta_{2,2} \\ \beta_{2,3} \\ \beta_{3,0} \quad \beta_{3,1} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0,1} \\ \beta_{2,2} \quad \beta_{2,3} \\ \beta_{2,3} \\ \beta_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} \quad \beta_{0,1} \\ \beta_{0,0} \quad \beta_{0$$

Observations

Partitioned Matrix-Matrix Multiplication

$$C = (A_0 \mid A_1) \left(\frac{B_0}{B_1} \right) = A_0 B_0 + A_1 B_1$$

$$\bullet \quad \begin{pmatrix} \gamma_{0,0} & \gamma_{0,1} & \gamma_{0,2} & \gamma_{0,3} \\ \gamma_{1,0} & \gamma_{1,1} & \boxed{\gamma_{1,2}} & \gamma_{1,3} \\ \gamma_{2,0} & \gamma_{2,1} & \gamma_{2,2} & \gamma_{2,3} \\ \gamma_{3,0} & \gamma_{3,1} & \gamma_{3,2} & \gamma_{3,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} \\ \boxed{\alpha_{1,0} & \alpha_{1,1}} \\ \alpha_{2,0} & \alpha_{2,1} \\ \boxed{\alpha_{3,0} & \alpha_{3,1}} \\ \boxed{\alpha_{3,0} & \alpha_{3,1}} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} & \boxed{\beta_{0,2}} \\ \beta_{1,0} & \beta_{1,1} & \boxed{\beta_{1,2}} & \beta_{2,3} \\ \boxed{\beta_{1,0} & \beta_{1,1}} & \boxed{\beta_{2,2}} & \beta_{2,3} \\ \boxed{\beta_{3,0} & \beta_{3,1}} & \boxed{\beta_{2,2}} & \beta_{2,3} \\ \boxed{\beta_{3,0} & \beta_{3,1}} & \boxed{\beta_{3,2}} & \beta_{3,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \boxed{\beta_{0,2}} \\ \boxed{\alpha_{1,0} & \alpha_{1,1}} \\ \boxed{\alpha_{2,0} & \alpha_{2,1}} \\ \boxed{\alpha_{2,0} & \alpha_{2,1}} \\ \boxed{\alpha_{3,0} & \alpha_{3,1}} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} & \boxed{\beta_{0,2}} \\ \boxed{\beta_{1,2}} & \beta_{1,2} & \beta_{1,3} \end{pmatrix} + \begin{pmatrix} \alpha_{0,2} & \alpha_{0,3} \\ \boxed{\alpha_{1,2} & \alpha_{1,3}} \\ \boxed{\alpha_{2,2} & \alpha_{2,3}} \\ \boxed{\alpha_{3,0} & \beta_{3,1}} & \boxed{\beta_{2,2}} & \beta_{2,3} \\ \boxed{\beta_{3,0} & \beta_{3,1}} & \boxed{\beta_{2,2}} & \beta_{2,3} \\ \boxed{\beta_{3,0} & \beta_{3,1}} & \boxed{\beta_{3,2}} & \beta_{3,3} \end{pmatrix}$$

Observations

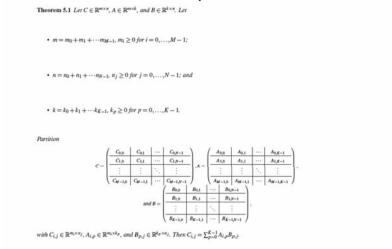
• Partitioned Matrix-Matrix Multiplication $\begin{pmatrix} C_{00} & C_{01} \\ C_{10} & C_{11} \end{pmatrix} = \begin{pmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{pmatrix} \begin{pmatrix} B_{00} & B_{01} \\ B_{10} & B_{11} \end{pmatrix} = \begin{pmatrix} A_{00}B_{00} + A_{01}B_{10} & A_{01}B_{01} + A_{01}B_{11} \\ A_{10}B_{00} + A_{11}B_{10} & A_{10}B_{01} + A_{11}B_{11} \end{pmatrix}$

 $\begin{pmatrix} \gamma_{0,0} & \gamma_{0,1} & \gamma_{0,2} & \gamma_{0,3} \\ \gamma_{1,0} & \gamma_{1,1} & \gamma_{1,2} & \gamma_{1,3} \\ \gamma_{2,0} & \gamma_{2,1} & \gamma_{2,2} & \gamma_{2,3} \\ \gamma_{3,0} & \gamma_{3,1} & \gamma_{2,2} & \gamma_{3,3} \end{pmatrix} = \begin{pmatrix} \frac{\alpha_{0,0}}{\alpha_{1,0}} & \alpha_{0,1} & \alpha_{0,2} & \alpha_{0,3} \\ \frac{\alpha_{1,0}}{\alpha_{2,0}} & \alpha_{2,1} & \alpha_{1,2} & \alpha_{1,3} \\ \alpha_{2,0} & \alpha_{3,1} & \alpha_{2,2} & \alpha_{2,3} \\ \alpha_{3,0} & \alpha_{3,1} & \alpha_{3,2} & \alpha_{3,3} \end{pmatrix} \begin{pmatrix} \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,3} \\ \beta_{0,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \beta_{0,0} & \beta_{0,1} & \beta_{0,0} & \beta_{0,1} \\ \alpha_{0,0} & \alpha_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,2} & \beta_{2,3} & \beta_{2,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \beta_{0,0} & \beta_{0,1} & \beta_{0,0} & \beta_{0,1} & \beta_{0,0} & \beta_{0,1} \\ \beta_{0,0} & \beta_{2,1} & \beta_{2,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \\ \beta_{2,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \beta_{0,0} & \beta_{0,1} & \beta_{0,0} & \beta_{0,1} & \beta_{0,0} & \beta_{0,1} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,1} & \beta_{0,2} & \beta_{0,3} \\ \beta_{0,0} & \beta_{2,1} & \beta_{2,2} & \beta_{2,3} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \beta_{0,1} & \beta_{0,1} & \beta_{0,1} & \beta_{0,1} & \beta_{0,1} & \beta_{0,1} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,2} & \beta_{0,2} \\ \beta_{0,0} & \beta_{0,1} & \beta_{0,$

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Observations

• Partitioned Matrix-Matrix Multiplication



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Observations

• Partitioned Matrix-Matrix Multiplication

$$A = \begin{pmatrix} -1 & 2 & 4 & 1 \\ 1 & 0 & -1 & -2 \\ 2 & -1 & 3 & 1 \\ 1 & 2 & 3 & 4 \end{pmatrix}, B = \begin{pmatrix} -2 & 2 & -3 \\ 0 & 1 & -1 \\ \hline -2 & -1 & 0 \\ 4 & 0 & 1 \end{pmatrix}, \text{ and } AB = \begin{pmatrix} -2 & -4 & 2 \\ -8 & 3 & -5 \\ -6 & 0 & -4 \\ 8 & 1 & -1 \end{pmatrix};$$

If
$$A_0 = \begin{pmatrix} -1 & 2 \\ 1 & 0 \\ 2 & -1 \\ 1 & 2 \end{pmatrix}, A_1 = \begin{pmatrix} 4 & 1 \\ -1 & -2 \\ 3 & 1 \\ 3 & 4 \end{pmatrix}, B_0 \begin{pmatrix} -2 & 2 & -3 \\ 0 & 1 & -1 \end{pmatrix}, \text{ and } B_1 = \begin{pmatrix} -2 & -1 & 0 \\ 4 & 0 & 1 \end{pmatrix}.$$

Observations

• Partitioned Matrix-Matrix Multiplication

$$AB = \begin{pmatrix} A_0 & A_1 \end{pmatrix} \begin{pmatrix} B_0 \\ B_1 \end{pmatrix} = A_0 B_0 + A_1 B_1 :$$

$$\begin{pmatrix} -1 & 2 & 4 & 1 \\ 1 & 0 & -1 & -2 \\ 2 & -1 & 3 & 1 \\ 1 & 2 & 3 & 4 \end{pmatrix} \begin{pmatrix} -2 & 2 & -3 \\ 0 & 1 & -1 \\ -2 & -1 & 0 \\ 4 & 0 & 1 \end{pmatrix}$$

$$B$$

$$= \begin{pmatrix} -1 & 2 \\ 1 & 0 \\ 2 & -1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} -2 & 2 & -3 \\ 0 & 1 & -1 \\ 4 & 0 & 1 \end{pmatrix}$$

$$B$$

$$= \begin{pmatrix} 2 & 1 \\ -2 & 2 & -3 \\ 1 & 2 & 3 & 4 \end{pmatrix} + \begin{pmatrix} 4 & 1 \\ -1 & -2 \\ 3 & 1 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} -2 & -1 & 0 \\ 4 & 0 & 1 \end{pmatrix}$$

$$A_1$$

$$= \begin{pmatrix} 2 & 0 & 1 \\ -2 & 2 & -3 \\ -4 & 3 & -5 \\ -2 & 4 & -5 \end{pmatrix} + \begin{pmatrix} -4 & -4 & 1 \\ -6 & 1 & -2 \\ -2 & -3 & 1 \\ 10 & -3 & 4 \end{pmatrix} = \begin{pmatrix} -2 & -4 & 2 \\ -8 & 3 & -5 \\ -6 & 0 & -4 \\ 8 & 1 & -1 \end{pmatrix}.$$

$$AB$$

Robert v

Observations

• Properties Transposing a Product of Matrices

Homework 5.2.3.1 Let
$$A = \begin{pmatrix} 2 & 0 & 1 \\ -1 & 1 & 0 \\ 1 & 3 & 1 \\ -1 & 1 & 1 \end{pmatrix}$$
 and $B = \begin{pmatrix} 2 & 1 & 2 & 1 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{pmatrix}$. Compute
$$\bullet A^T A = \bullet AA^T = \bullet (AB)^T = \bullet A^T B^T = \bullet B^T A^T =$$

Homework 5.2.3.3 Let A, B, and C be conformal matrices so that ABC is well-defined. Then $(ABC)^T = C^T B^T A^T$.

Always/Sometimes/Never

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Homework 5.2.3.2 Let $A \in \mathbb{R}^{m \times k}$ and $B \in \mathbb{R}^{k \times n}$. $(AB)^T = B^T A^T$.

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Always/Sometimes/Never

Observations

• Matrix-Matrix Multiplication with Special Matrices

Homework 5.2.4.14 Evaluate
$$\begin{pmatrix} 1 & -1 & | & -2 \\ 0 & 2 & | & 3 \\ 0 & 0 & | & 1 \end{pmatrix} \begin{pmatrix} -2 & 1 & | & -1 \\ 0 & 1 & | & 2 \\ 0 & 0 & | & 1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 0 & 2 & | & 0 \\ 0 & 0 & | & 1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 0 & 2 & | & 0 \\ 0 & 0 & | & 1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 1 & 0 & 2 \\ 2 & | & -1 \end{pmatrix} \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 2 \\ 2 & 0 & -1 \\ 2 & | & 1 \end{pmatrix} \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 2 & 2 \\ 2 & 0 & -1 \\ 2 & -1 & | & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 2 & 2 \\ 2 & 0 & -1 \\ 2 & -1 & | & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 2 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 2 \\ 2 & 0 & -1 \\ 2 & 0 &$$

Algorithms for Computing Matrix-Matrix Multiplication

- Lots of Loops
- Matrix-Matrix Multiplication by Columns
- Matrix-Matrix Multiplication by Rows
- Matrix-Matrix Multiplication with Rank-1 Updates

Algorithms for Computing Matrix-Matrix Multiplication

Lots of Loops

$$C = AB$$

$$\gamma_{i,j} = \sum_{p=0}^{\kappa-1} \alpha_{i,p} \beta_{p,j}$$

Consider the MATLAB function

Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Columns

In Theorem 5.1 let us partition C and B by columns and not partition A. In other words, let M = 1, $m_0 = m$; N = n, $n_j = 1$, j = 0, ..., n-1; and K = 1, $k_0 = k$. Then

$$C = \begin{pmatrix} c_0 & c_1 & \cdots & c_{n-1} \end{pmatrix}$$
 and $B = \begin{pmatrix} b_0 & b_1 & \cdots & b_{n-1} \end{pmatrix}$

so tha

$$\left(\begin{array}{c|c}c_0&c_1&\cdots&c_{n-1}\end{array}\right)=C=AB=A\left(\begin{array}{c|c}b_0&b_1&\cdots&b_{n-1}\end{array}\right)=\left(\begin{array}{c|c}Ab_0&Ab_1&\cdots&Ab_{n-1}\end{array}\right).$$

Homework 5.3.2.3

•
$$\begin{pmatrix} 1 & -2 & 2 \\ -1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & -2 & 2 \\ -1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 2 & 1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} 1 & -2 & 2 \\ -1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 2 & 1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} 1 & -2 & 2 \\ -1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} -1 & 0 & 1 \\ 2 & 1 & -1 \\ 1 & -1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 1 \\ 1 & -1 & 2 \end{pmatrix}$$

Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Columns

By moving the loop indexed by j to the outside in the algorithm for computing C = AB + C we observe that

Exchanging the order of the two inner-most loops merely means we are using a different algorithm (dot product vs. AXPY) for the matrix-vector multiplication $c_i := Ab_i + c_i$.

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Algorithms for Computing Matrix-Matrix Multiplication

Matrix-Matrix Multiplication by Columns

Algorithm:
$$C := \text{GEMM_UNB_VAR1}(A, B, C)$$

Partition $B \to \begin{pmatrix} B_L & B_R \end{pmatrix}$, $C \to \begin{pmatrix} C_L & C_R \end{pmatrix}$ where B_L has 0 columns, C_L has 0 columns

while $n(B_L) < n(B)$ do

Repartition
$$\begin{pmatrix} B_L & B_R \end{pmatrix} \to \begin{pmatrix} B_0 & b_1 & B_2 \end{pmatrix}$$
, $\begin{pmatrix} C_L & C_R \end{pmatrix} \to \begin{pmatrix} C_0 & c_1 & C_2 \end{pmatrix}$
where b_1 has 1 column, c_1 has 1 column
$$c_1 := Ab_1 + c_1$$

Continue with
$$\begin{pmatrix} B_L & B_R \end{pmatrix} \to \begin{pmatrix} B_0 & b_1 & B_2 \end{pmatrix}$$
, $\begin{pmatrix} C_L & C_R \end{pmatrix} \to \begin{pmatrix} C_0 & c_1 & C_2 \end{pmatrix}$
endwhile

Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Rows

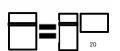
In Theorem 5.1 partition C and A by rows and do not partition B. In other words, let $M=m, m_i=1, i=0,\ldots,m-1; N=1, n_0=n$; and $K=1, k_0=k$. Then

$$C = \begin{pmatrix} \frac{\mathcal{C}_0^T}{\mathcal{C}_1^T} \\ \vdots \\ \overline{\mathcal{C}_{m-1}^T} \end{pmatrix} \quad \text{and} \quad A = \begin{pmatrix} \frac{\mathcal{C}_0^T}{\mathcal{C}_1^T} \\ \vdots \\ \overline{\mathcal{C}_{m-1}^T} \end{pmatrix}$$

so that

$$\begin{bmatrix}
\frac{\bar{c}_{0}^{T}}{c_{1}^{T}} \\
\vdots \\
\bar{c}_{-}^{T}
\end{bmatrix} = C = AB = \begin{bmatrix}
\frac{\bar{a}_{0}^{T}}{\bar{a}_{1}^{T}} \\
\vdots \\
\bar{a}_{-}^{T}
\end{bmatrix} B = \begin{bmatrix}
\frac{\bar{a}_{0}^{T}B}{\bar{a}_{1}^{T}B} \\
\vdots \\
\bar{d}_{-}^{T}
\end{bmatrix} B$$

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Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Rows

Homework 5.3.3.2

•
$$\left(\begin{array}{c|cc} 1 & -2 & 2 \\ \hline \end{array}\right) \left(\begin{array}{ccc} -1 & 0 & 1 \\ 2 & 1 & -1 \\ 1 & -1 & 2 \end{array}\right) =$$

• $\left(\begin{array}{ccc} 1 & -2 & 2 \\ \hline -1 & 2 & 1 \\ \hline \end{array}\right) \left(\begin{array}{ccc} -1 & 0 & 1 \\ 2 & 1 & -1 \\ 1 & -1 & 2 \end{array}\right) =$

• $\left(\begin{array}{ccc} 1 & -2 & 2 \\ \hline -1 & 2 & 1 \\ \hline 0 & 1 & 2 \end{array}\right) \left(\begin{array}{ccc} -1 & 0 & 1 \\ 2 & 1 & -1 \\ 1 & -1 & 2 \end{array}\right) =$

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Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Rows

In the algorithm for computing C = AB + C the loop indexed by i can be moved to the outside so that

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Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication by Rows

Algorithm:
$$C := \text{GEMM_UNB_VAR2}(A, B, C)$$

Partition $A \to \left(\frac{A_T}{A_B}\right)$, $C \to \left(\frac{C_T}{C_B}\right)$
where A_T has 0 rows, C_T has 0 rows
while $m(A_T) < m(A)$ do

Repartition
$$\left(\frac{A_T}{A_B}\right) \to \left(\frac{A_0}{\frac{a_1^T}{A_2}}\right)$$
, $\left(\frac{C_T}{C_B}\right) \to \left(\frac{C_0}{\frac{c_1^T}{C_2}}\right)$
where a_1 has 1 row, c_1 has 1 row
$$\overline{c_1^T := a_1^T B + c_1^T}$$

Continue with
$$\left(\frac{A_T}{A_B}\right) \leftarrow \left(\frac{A_0}{\frac{a_1^T}{A_2}}\right)$$
, $\left(\frac{C_T}{C_B}\right) \leftarrow \left(\frac{C_0}{\frac{c_1^T}{C_2}}\right)$
endwhile

Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication with Rank-1 Updates

In Theorem 5.1 partition A and B by columns and rows, respectively, and do not partition C. In other words, let M=1, $m_0=m$, N=1, $n_0=n$; and K=k, $k_p=1$, $p=0,\ldots,k-1$. Then

$$A = \begin{pmatrix} a_0 \mid a_1 \mid \cdots \mid a_{k-1} \end{pmatrix}$$
 and $B = \begin{pmatrix} \frac{b_0^T}{b_1^T} \\ \vdots \\ \frac{b_{k-1}^T}{b_{k-1}^T} \end{pmatrix}$

so that

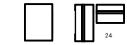
$$C = AB = \begin{pmatrix} a_0 & a_1 & \cdots & a_{k-1} \end{pmatrix} \begin{pmatrix} \frac{b_0^T}{b_1^T} \\ \vdots \\ \vdots \\ b_{k-1}^T \end{pmatrix} = a_0 b_0^T + a_1 b_1^T + \cdots + a_{k-1} b_{k-1}^T.$$

Notice that each term $a_p \vec{b_p}$ is an outer product of a_p and $\vec{b_p}$. Thus, if we start with C := 0, the zero matrix, then we can compute C := AB + C as

$$C := a_{k-1}b_{k-1}^T + (\cdots + (a_pb_p^T + (\cdots + (a_1b_1^T + (a_0b_0^T + C))\cdots))\cdots),$$

which illustrates that C := AB can be computed by first setting C to zero, and then repeatedly updating it with rank-1 updates.

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Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication with Rank-1 Updates

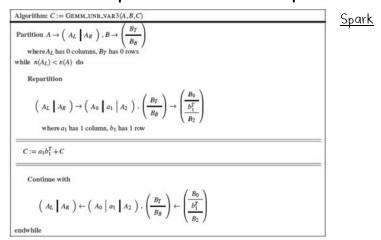
Homework 5.3.4.1

•
$$\begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$$
 $\begin{pmatrix} -1 & 0 & 1 \\ \hline & & \\ & &$

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Algorithms for Computing Matrix-Matrix Multiplication

• Matrix-Matrix Multiplication with Rank-1 Updates



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Dot product

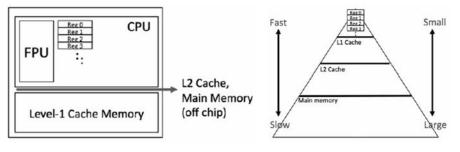
Enrichment

• Slicing and Dicing for Performance

-CPU: Central Processing Unit

• FPU: Floating Point Unit

• Registers

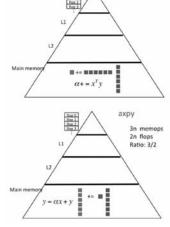


Enrichment

• Slicing and Dicing for Performance

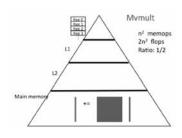
- Vector-Vector Computations

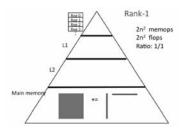
- Matrix-Vector Computations



Enrichment

- Slicing and Dicing for Performance
 - Matrix-Vector Computations



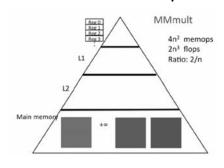


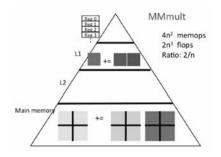
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Enrichment

- Slicing and Dicing for Performance
 - Matrix-Matrix Computations





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Questions and Answers

