

# Interval Matrix Multiplication Algorithms Implementation and Accuracy

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# Content

Algorithms for Matrix-Matrix Multiplication

Implementation Issues, Parallelism, and Multicores

Accuracy: the Relative Radius Error Metric

Conclusion

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# Infsup Arithmetic Formulas

**Definition** An interval  $\mathbf{x}$  is a closed convex subset of  $\mathbb{R}$ .

**Infsup Representation**  $\mathbf{x} = [\underline{x}, \bar{x}] = \{y \mid \underline{x} \leq y \leq \bar{x}\}$

- ▶ addition

$$\mathbf{x} + \mathbf{y} = [\underline{x} + \underline{y}, \bar{x} + \bar{y}]$$

- ▶ subtraction

$$\mathbf{x} - \mathbf{y} = [\underline{x} - \bar{y}, \bar{x} - \underline{y}]$$

- ▶ multiplication

$$\mathbf{x} \times \mathbf{y} = [\min(\underline{x} \times \underline{y}, \underline{x} \times \bar{y}, \bar{x} \times \underline{y}, \bar{x} \times \bar{y}), \max(\underline{x} \times \underline{y}, \underline{x} \times \bar{y}, \bar{x} \times \underline{y}, \bar{x} \times \bar{y})]$$

# Classical Algorithm for Matrix Product

**Input:**  $\mathbf{A} = [\underline{\mathbf{A}}, \overline{\mathbf{A}}] \in \mathbb{IF}^{m \times k}$ ,  $\mathbf{B} = [\underline{\mathbf{B}}, \overline{\mathbf{B}}] \in \mathbb{IF}^{k \times n}$

**Output:**  $\mathbf{C} \in \mathbb{IF}^{m \times n}$ ,  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

```
1: for  $i = 1$  to  $m$  do
2:   for  $j = 1$  to  $n$  do
3:      $\underline{\mathbf{C}}_{ij} \leftarrow 0$ ;  $\overline{\mathbf{C}}_{ij} \leftarrow 0$ 
4:     for  $l = 1$  to  $k$  do
5:        $\underline{\mathbf{C}}_{ij} \leftarrow$ 
          $fl_{\nabla}(\underline{\mathbf{C}}_{ij} + \min\{\underline{\mathbf{A}}_{il} \times \underline{\mathbf{B}}_{lj}, \underline{\mathbf{A}}_{il} \times \overline{\mathbf{B}}_{lj}, \overline{\mathbf{A}}_{il} \times \underline{\mathbf{B}}_{lj}, \overline{\mathbf{A}}_{il} \times \overline{\mathbf{B}}_{lj}\})$ 
6:        $\overline{\mathbf{C}}_{ij} \leftarrow$ 
          $fl_{\Delta}(\overline{\mathbf{C}}_{ij} + \max\{\underline{\mathbf{A}}_{il} \times \underline{\mathbf{B}}_{lj}, \underline{\mathbf{A}}_{il} \times \overline{\mathbf{B}}_{lj}, \overline{\mathbf{A}}_{il} \times \underline{\mathbf{B}}_{lj}, \overline{\mathbf{A}}_{il} \times \overline{\mathbf{B}}_{lj}\})$ 
7:     end for
8:   end for
9: end for
10: return  $[\underline{\mathbf{C}}, \overline{\mathbf{C}}]$ 
```

# Classical Algorithm for Matrix Product

**Input:**  $\mathbf{A} = [\underline{A}, \overline{A}] \in \mathbb{IF}^{m \times k}$ ,  $\mathbf{B} = [\underline{B}, \overline{B}] \in \mathbb{IF}^{k \times n}$

**Output:**  $\mathbf{C} \in \mathbb{IF}^{m \times n}$ ,  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

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 $fl_{\nabla}(\underline{C}_{ij} + \min\{\underline{A}_{il} \times \underline{B}_{lj}, \underline{A}_{il} \times \overline{B}_{lj}, \overline{A}_{il} \times \underline{B}_{lj}, \overline{A}_{il} \times \overline{B}_{lj}\})$ 
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7:     end for
8:   end for
9: end for
10: return  $[\underline{\mathbf{C}}, \overline{\mathbf{C}}]$ 
```

# Midrad Addition/Subtraction Formulas

Neumaier, *Interval methods for systems of equations*, 1990

Midrad Representation  $\mathbf{x} = \langle \text{mid } \mathbf{x}, \text{rad } \mathbf{x} \rangle = \left\langle \frac{\bar{x} + \underline{x}}{2}, \frac{\bar{x} - \underline{x}}{2} \right\rangle$

- ▶ addition

$$\mathbf{x} + \mathbf{y} = \langle \text{mid } \mathbf{x} + \text{mid } \mathbf{y}, \text{rad } \mathbf{x} + \text{rad } \mathbf{y} \rangle$$

- ▶ subtraction

$$\mathbf{x} - \mathbf{y} = \langle \text{mid } \mathbf{x} - \text{mid } \mathbf{y}, \text{rad } \mathbf{x} + \text{rad } \mathbf{y} \rangle$$



# Midrad Multiplication Formulas

Neumaier 1990, Rump 1999, Nguyen 2011

- multiplication  $\mathbf{x} \times \mathbf{y} = \mathbf{z} \subseteq \mathbf{z}_1 \subseteq \mathbf{z}_2$

	mid $\mathbf{z}$	rad $\mathbf{z}$
$\mathbf{z}$	$\alpha + \text{sign}(\alpha) \times \min\{\beta, \gamma, \delta\}$	$\max\{\beta, \gamma\} + \max\{\min\{\beta, \gamma\}, \delta\}$
$\mathbf{z}_1$	$\alpha + \text{sign}(\alpha) \times \min\{\alpha', \beta, \gamma, \delta\}$	$\beta + \gamma + \delta - \min\{\alpha', \beta, \gamma, \delta\}$
$\mathbf{z}_2$	$\alpha$	$\beta + \gamma + \delta$

where

$$\alpha = \text{mid } \mathbf{x} \times \text{mid } \mathbf{y}$$

$$\alpha' = |\text{mid } \mathbf{x}| \times |\text{mid } \mathbf{y}|$$

$$\beta = |\text{mid } \mathbf{x}| \times \text{rad } \mathbf{y}$$

$$\gamma = \text{rad } \mathbf{x} \times |\text{mid } \mathbf{y}|$$

$$\delta = \text{rad } \mathbf{x} \times \text{rad } \mathbf{y}$$

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$\mathbf{z}_2$	$\alpha$	$\beta + \gamma + \delta$

where

$$\alpha = \text{mid } \mathbf{x} \times \text{mid } \mathbf{y}$$

$$\alpha' = |\text{mid } \mathbf{x}| \times |\text{mid } \mathbf{y}|$$

$$\beta = |\text{mid } \mathbf{x}| \times \text{rad } \mathbf{y}$$

$$\gamma = \text{rad } \mathbf{x} \times |\text{mid } \mathbf{y}|$$

$$\delta = \text{rad } \mathbf{x} \times \text{rad } \mathbf{y}$$

# IIMu14 Algorithm

Rump, *Fast and Parallel Interval Arithmetic*, 1999

**Input:**  $\mathbf{A} = [\underline{A}, \overline{A}]$ ,  $\mathbf{B} = [\underline{B}, \overline{B}]$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

- 1:  $\langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{A})$
- 2:  $\langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{B})$
- 3:  $R_{\mathbf{C}} \leftarrow fl_{\Delta}(|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}))$
- 4:  $\overline{C} \leftarrow fl_{\Delta}(M_{\mathbf{A}} \times M_{\mathbf{B}} + R_{\mathbf{C}})$
- 5:  $\underline{C} \leftarrow fl_{\nabla}(M_{\mathbf{A}} \times M_{\mathbf{B}} - R_{\mathbf{C}})$
- 6: **return**  $[\underline{C}, \overline{C}]$

4 Matrix-Matrix Multiplications

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Rump, *Fast and Parallel Interval Arithmetic*, 1999

**Input:**  $\mathbf{A} = [\underline{A}, \overline{A}]$ ,  $\mathbf{B} = [\underline{B}, \overline{B}]$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

1:  $\langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{A})$

2:  $\langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{B})$

3:  $R_{\mathbf{C}} \leftarrow fl_{\Delta}(|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}))$

4:  $\overline{C} \leftarrow fl_{\Delta}(M_{\mathbf{A}} \times M_{\mathbf{B}} + R_{\mathbf{C}})$

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Rump, *Fast and Parallel Interval Arithmetic*, 1999

**Input:**  $A = [\underline{A}, \overline{A}]$ ,  $B = [\underline{B}, \overline{B}]$

**Output:**  $C \supseteq A \times B$

1:  $\langle M_A, R_A \rangle \leftarrow \text{InfsupToMidrad}(A)$

2:  $\langle M_B, R_B \rangle \leftarrow \text{InfsupToMidrad}(B)$

3:  $R_C \leftarrow fl_{\Delta}(|M_A| \times R_B + R_A \times (|M_B| + R_B))$

4:  $\overline{C} \leftarrow fl_{\Delta}(M_A \times M_B + R_C)$

5:  $\underline{C} \leftarrow fl_{\nabla}(M_A \times M_B - R_C)$

6: **return**  $[\underline{C}, \overline{C}]$

## 4 Matrix-Matrix Multiplications

# IIMu17 Algorithm

Nguyen, *Efficient algorithms for verified scientific computing: numerical linear algebra using interval arithmetic*, 2011

**Input:**  $\mathbf{A} = [\underline{\mathbf{A}}, \overline{\mathbf{A}}]$ ,  $\mathbf{B} = [\underline{\mathbf{B}}, \overline{\mathbf{B}}]$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

1:  $\langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{A})$

2:  $\langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{B})$

3:  $\rho_{\mathbf{A}} \leftarrow \text{sign}(M_{\mathbf{A}}) \cdot \min(|M_{\mathbf{A}}|, R_{\mathbf{A}})$

4:  $\rho_{\mathbf{B}} \leftarrow \text{sign}(M_{\mathbf{B}}) \cdot \min(|M_{\mathbf{B}}|, R_{\mathbf{B}})$

5:  $R_{\mathbf{C}} \leftarrow$

$$fl_{\Delta} (|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}) + (-|\rho_{\mathbf{A}}|) \times |\rho_{\mathbf{B}}|)$$

6:  $\overline{\mathbf{C}} \leftarrow fl_{\Delta} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} + R_{\mathbf{C}})$

7:  $\underline{\mathbf{C}} \leftarrow fl_{\nabla} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} - R_{\mathbf{C}})$

8: **return**  $[\underline{\mathbf{C}}, \overline{\mathbf{C}}]$

7 Matrix-Matrix Multiplications



# IIMu17 Algorithm

Nguyen, *Efficient algorithms for verified scientific computing: numerical linear algebra using interval arithmetic*, 2011

**Input:**  $\mathbf{A} = [\underline{\mathbf{A}}, \overline{\mathbf{A}}]$ ,  $\mathbf{B} = [\underline{\mathbf{B}}, \overline{\mathbf{B}}]$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

1:  $\langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{A})$

2:  $\langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{B})$

3:  $\rho_{\mathbf{A}} \leftarrow \text{sign}(M_{\mathbf{A}}) \cdot \min(|M_{\mathbf{A}}|, R_{\mathbf{A}})$

4:  $\rho_{\mathbf{B}} \leftarrow \text{sign}(M_{\mathbf{B}}) \cdot \min(|M_{\mathbf{B}}|, R_{\mathbf{B}})$

5:  $R_{\mathbf{C}} \leftarrow$

$$fl_{\Delta} (|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}) + (-|\rho_{\mathbf{A}}|) \times |\rho_{\mathbf{B}}|)$$

6:  $\overline{\mathbf{C}} \leftarrow fl_{\Delta} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} + R_{\mathbf{C}})$

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5:  $R_{\mathbf{C}} \leftarrow$

$$fl_{\Delta} (|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}) + (-|\rho_{\mathbf{A}}|) \times |\rho_{\mathbf{B}}|)$$

6:  $\overline{\mathbf{C}} \leftarrow fl_{\Delta} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} + R_{\mathbf{C}})$

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## 7 Matrix-Matrix Multiplications

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1:  $\langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{A})$

2:  $\langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \leftarrow \text{InfsupToMidrad}(\mathbf{B})$

3:  $\rho_{\mathbf{A}} \leftarrow \text{sign}(M_{\mathbf{A}}) \cdot \min(|M_{\mathbf{A}}|, R_{\mathbf{A}})$

4:  $\rho_{\mathbf{B}} \leftarrow \text{sign}(M_{\mathbf{B}}) \cdot \min(|M_{\mathbf{B}}|, R_{\mathbf{B}})$

5:  $R_{\mathbf{C}} \leftarrow$

$$fl_{\Delta} (|M_{\mathbf{A}}| \times R_{\mathbf{B}} + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}) + (-|\rho_{\mathbf{A}}|) \times |\rho_{\mathbf{B}}|)$$

6:  $\overline{\mathbf{C}} \leftarrow fl_{\Delta} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} + R_{\mathbf{C}})$

7:  $\underline{\mathbf{C}} \leftarrow fl_{\nabla} (M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}} - R_{\mathbf{C}})$

8: **return**  $[\underline{\mathbf{C}}, \overline{\mathbf{C}}]$

The same operations are computed twice but with different rounding modes.

# Bound on the Error of a Matrix Product in Rounding to Nearest

Rump, *Error estimation of floating-point summation and dot product*, 2011

## Theorem

Let  $A \in \mathbb{F}^{m \times k}$  and  $B \in \mathbb{F}^{k \times n}$  with  $2(k+2)u \leq 1$  be given, and let  $C = fl_{\square}(A \times B)$  and  $\Gamma = fl_{\square}(|A| \times |B|)$ . Here  $C$  may be computed in any order, and we assume that  $\Gamma$  is computed in the same order.

Then

$$|fl_{\square}(A \times B) - A \times B| \leq fl_{\square} \left( \frac{k+2}{2} \text{ulp}(\Gamma) + \frac{1}{2} u^{-1} \eta \right)$$

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$$|fl_{\square}(A \times B) - A \times B| \leq fl_{\square} \left( \frac{k+2}{2} \text{ulp}(\Gamma) + \frac{1}{2} u^{-1} \eta \right)$$

# MMu13 Algorithm

Rump, *Fast Interval Matrix Multiplication*, 2011

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{IF}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{IF}^{k \times n}$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

1:  $M_{\mathbf{C}} \leftarrow fl_{\square}(M_{\mathbf{A}} \times M_{\mathbf{B}})$

2:  $R'_{\mathbf{B}} \leftarrow fl_{\Delta}((k+2)u|M_{\mathbf{B}}| + R_{\mathbf{B}})$

3:  $R_{\mathbf{C}} \leftarrow fl_{\Delta}(|M_{\mathbf{A}}| \times R'_{\mathbf{B}} + u^{-1}\eta + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}))$

4: **return**  $\langle M_{\mathbf{C}}, R_{\mathbf{C}} \rangle$

3 Matrix-Matrix Multiplications

# MMu13 Algorithm

Rump, *Fast Interval Matrix Multiplication*, 2011

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{IF}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{IF}^{k \times n}$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

- 1:  $M_{\mathbf{C}} \leftarrow fl_{\square}(M_{\mathbf{A}} \times M_{\mathbf{B}})$
- 2:  $R'_{\mathbf{B}} \leftarrow fl_{\Delta}((k+2)u|M_{\mathbf{B}}| + R_{\mathbf{B}})$
- 3:  $R_{\mathbf{C}} \leftarrow fl_{\Delta}(|M_{\mathbf{A}}| \times R'_{\mathbf{B}} + u^{-1}\eta + R_{\mathbf{A}} \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}))$
- 4: **return**  $\langle M_{\mathbf{C}}, R_{\mathbf{C}} \rangle$

3 Matrix-Matrix Multiplications

# MMu15 Algorithm

Rump, *Fast Interval Matrix Multiplication*, 2011

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{IF}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{IF}^{k \times n}$

**Output:**  $\mathbf{C} \supseteq \mathbf{A} \times \mathbf{B}$

- 1:  $\rho_{\mathbf{A}} \leftarrow \text{sign}(M_{\mathbf{A}}) \cdot \min(|M_{\mathbf{A}}|, R_{\mathbf{A}})$
- 2:  $\rho_{\mathbf{B}} \leftarrow \text{sign}(M_{\mathbf{B}}) \cdot \min(|M_{\mathbf{B}}|, R_{\mathbf{B}})$
- 3:  $M_{\mathbf{C}} \leftarrow fl_{\square}(M_{\mathbf{A}} \times M_{\mathbf{B}} + \rho_{\mathbf{A}} \times \rho_{\mathbf{B}})$
- 4:  $\Gamma \leftarrow fl_{\square}(|M_{\mathbf{A}}| \times |M_{\mathbf{B}}| + |\rho_{\mathbf{A}}| \times |\rho_{\mathbf{B}}|)$
- 5:  $\gamma \leftarrow fl_{\Delta}((k+1)\text{ulp}(\Gamma) + \frac{1}{2}u^{-1}\eta)$
- 6:  $R_{\mathbf{C}} \leftarrow fl_{\Delta}((|M_{\mathbf{A}}| + R_{\mathbf{A}}) \times (|M_{\mathbf{B}}| + R_{\mathbf{B}}) - \Gamma + 2\gamma)$
- 7: **return**  $\langle M_{\mathbf{C}}, R_{\mathbf{C}} \rangle$

5 Matrix-Matrix Multiplications



# MMu15 Algorithm

Rump, *Fast Interval Matrix Multiplication*, 2011

**Input:**  $A = \langle M_A, R_A \rangle \in \mathbb{IF}^{m \times k}$ ,  $B = \langle M_B, R_B \rangle \in \mathbb{IF}^{k \times n}$

**Output:**  $C \supseteq A \times B$

- 1:  $\rho_A \leftarrow \text{sign}(M_A) \cdot \min(|M_A|, R_A)$
- 2:  $\rho_B \leftarrow \text{sign}(M_B) \cdot \min(|M_B|, R_B)$
- 3:  $M_C \leftarrow fl_{\square} (M_A \times M_B + \rho_A \times \rho_B)$
- 4:  $\Gamma \leftarrow fl_{\square} (|M_A| \times |M_B| + |\rho_A| \times |\rho_B|)$
- 5:  $\gamma \leftarrow fl_{\Delta} \left( (k+1)\text{ulp}(\Gamma) + \frac{1}{2}u^{-1}\eta \right)$
- 6:  $R_C \leftarrow fl_{\Delta} \left( (|M_A| + R_A) \times (|M_B| + R_B) - \Gamma + 2\gamma \right)$
- 7: **return**  $\langle M_C, R_C \rangle$

## 5 Matrix-Matrix Multiplications

# Content

Algorithms for Matrix-Matrix Multiplication

Implementation Issues, Parallelism, and Multicores

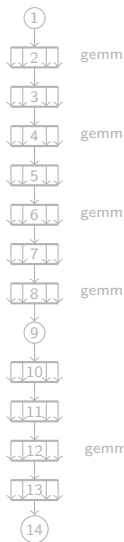
Accuracy: the Relative Radius Error Metric

Conclusion

# MMU15 Implementation with Level-3 BLAS

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{F}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{F}^{k \times n}$

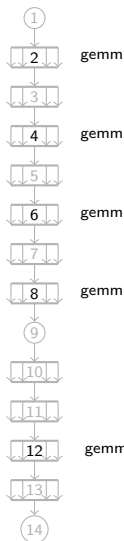
- 1: set rounding mode to nearest
- 2:  $M_{\mathbf{C}} \leftarrow M_{\mathbf{A}} \times M_{\mathbf{B}}$
- 3:  $T_1 \leftarrow |M_{\mathbf{A}}|$ ;  $T_2 \leftarrow |M_{\mathbf{B}}|$
- 4:  $R_{\mathbf{C}} \leftarrow T_1 \times T_2$
- 5:  $T_3 \leftarrow \text{Compute}(\rho_{\mathbf{A}})$ ;  $T_4 \leftarrow \text{Compute}(\rho_{\mathbf{B}})$
- 6:  $M_{\mathbf{C}} \leftarrow T_3 \times T_4 + M_{\mathbf{C}}$
- 7:  $T_5 \leftarrow |T_3|$ ;  $T_6 \leftarrow |T_4|$
- 8:  $R_{\mathbf{C}} \leftarrow T_5 \times T_6 + R_{\mathbf{C}}$
- 9: set rounding mode towards  $+\infty$
- 10:  $T_7 \leftarrow (k + 1)\text{ulp}(R_{\mathbf{C}}) + 0.5u^{-1}\eta$
- 11:  $T_8 \leftarrow T_1 + R_{\mathbf{A}}$ ;  $T_9 \leftarrow T_2 + R_{\mathbf{B}}$
- 12:  $R_{\mathbf{C}} \leftarrow T_8 \times T_9 - R_{\mathbf{C}}$
- 13:  $R_{\mathbf{C}} \leftarrow R_{\mathbf{C}} + 2T_7$
- 14: **return**  $\langle M_{\mathbf{C}}, R_{\mathbf{C}} \rangle$



# MMU15 Implementation with Level-3 BLAS

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{F}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{F}^{k \times n}$

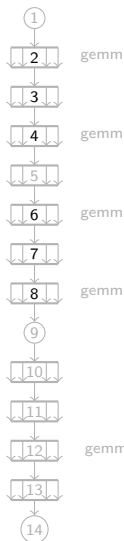
- 1: set rounding mode to nearest
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- 3:  $T_1 \leftarrow |M_{\mathbf{A}}|$ ;  $T_2 \leftarrow |M_{\mathbf{B}}|$
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**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{F}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{F}^{k \times n}$

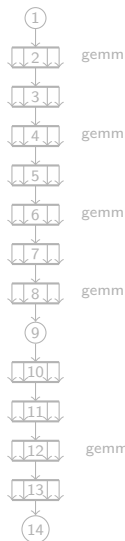
- 1: set rounding mode to nearest
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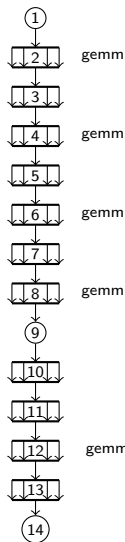
- 1: set rounding mode to nearest
- 2:  $M_{\mathbf{C}} \leftarrow M_{\mathbf{A}} \times M_{\mathbf{B}}$
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# MMu15 Implementation with Level-3 BLAS

**Input:**  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{F}^{m \times k}$ ,  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{F}^{k \times n}$

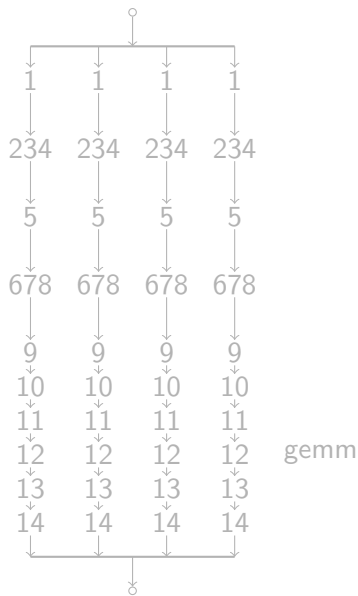
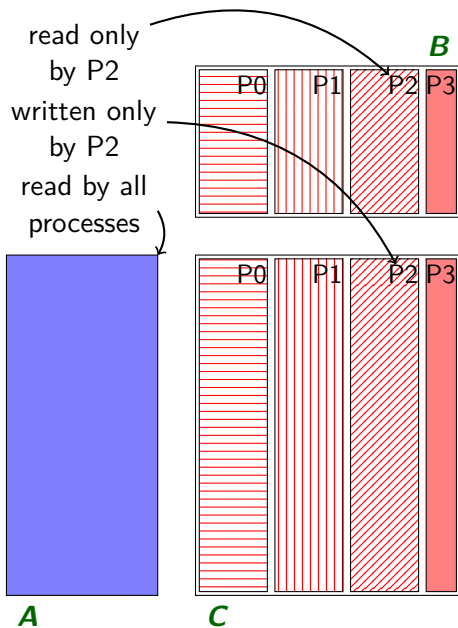
- 1: set rounding mode to nearest
- 2:  $M_{\mathbf{C}} \leftarrow M_{\mathbf{A}} \times M_{\mathbf{B}}$
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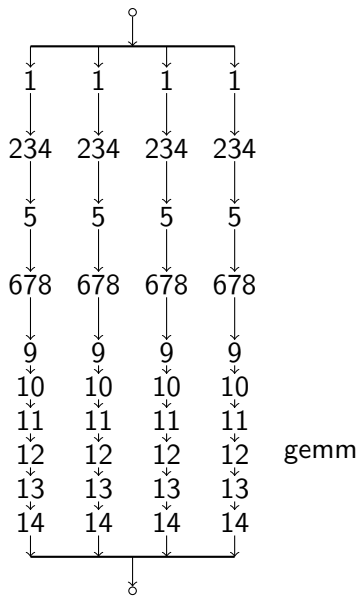
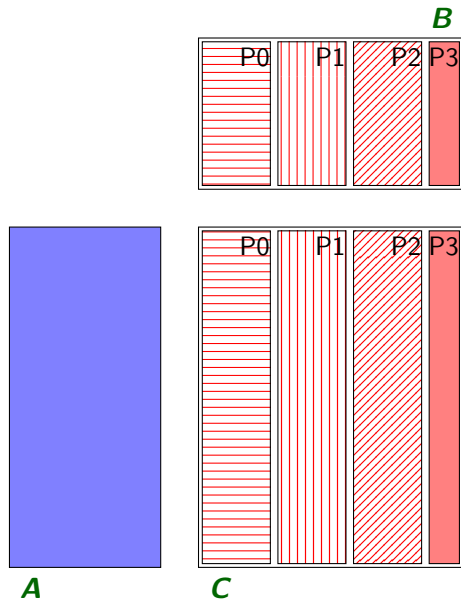




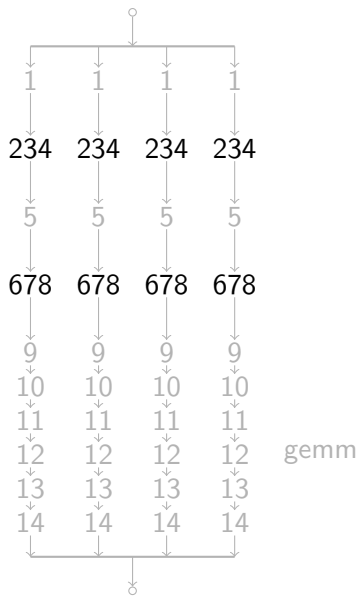
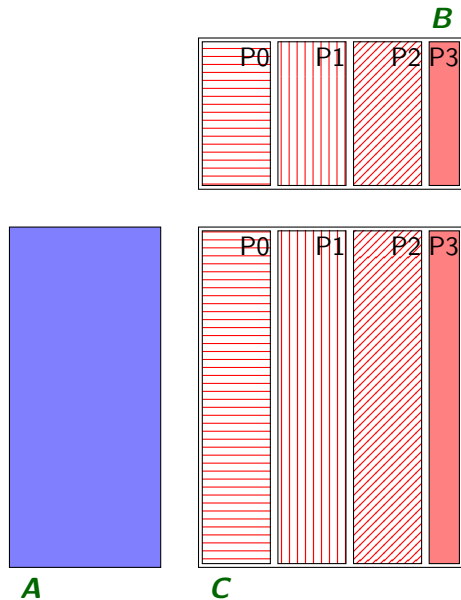
# MMMu15: Blocked Version



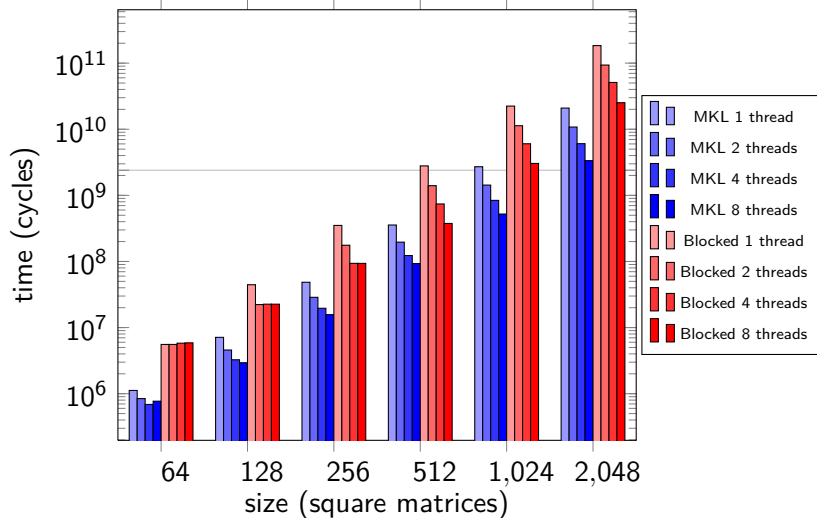
# MMU15: Blocked Version



# MM<sub>u</sub>15: Blocked Version



## Execution Time – MMMu15 BLAS vs blocked versions



Minimum time for product of two square matrices – Intel Bi-Xeon E5620, 2 x Quad-cores 2,40 GHz – Intel Math Kernel Library version 10.3 – gcc version 4.7.1

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# Notations

Let  $\mathbf{A} = \langle M_{\mathbf{A}}, R_{\mathbf{A}} \rangle \in \mathbb{IR}^{m \times k}$ , and  $\mathbf{B} = \langle M_{\mathbf{B}}, R_{\mathbf{B}} \rangle \in \mathbb{IR}^{k \times n}$  two interval matrices.

We note

- ▶ the exact  $\mathbf{A} \times \mathbf{B}$ :  $\mathbf{C} = \langle M_{\mathbf{C}}, R_{\mathbf{C}} \rangle$
- ▶  $\mathbf{A} \times \mathbf{B}$  computed with MMMu15:  $\mathbf{C}_5 = \langle M_{\mathbf{C}_5}, R_{\mathbf{C}_5} \rangle$
- ▶  $\mathbf{A} \times \mathbf{B}$  computed with MMMu13:  $\mathbf{C}_3 = \langle M_{\mathbf{C}_3}, R_{\mathbf{C}_3} \rangle$

As  $\mathbf{C}_5 \supseteq \mathbf{C}$ , we have

$$\left| M_{\mathbf{C}_5} - M_{\mathbf{C}} \right| \leq R_{\mathbf{C}_5} - R_{\mathbf{C}}.$$

Thus, the radius error accounts for the midpoint drift error.

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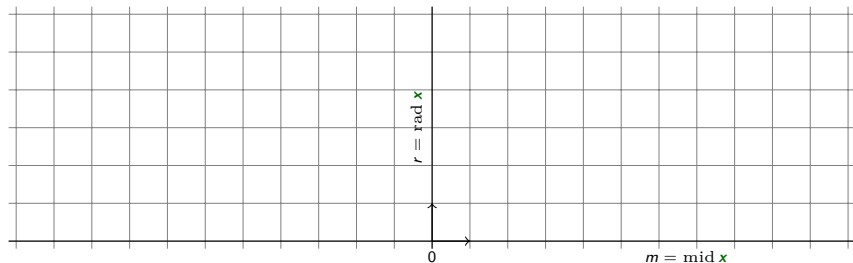
# Relative Precision

**Definition** An interval  $x = \langle m, r \rangle$  is said to be of *relative precision*  $e$  if  $r \leq e \cdot |m|$ .



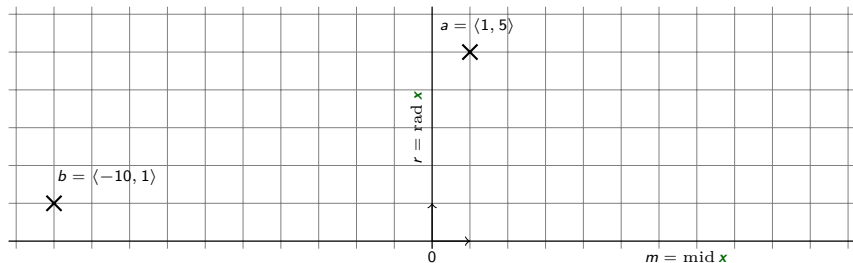
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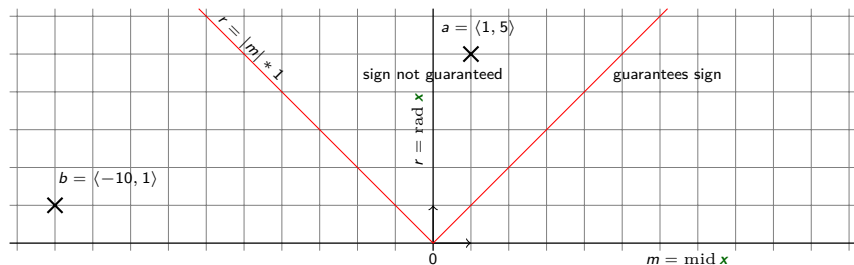
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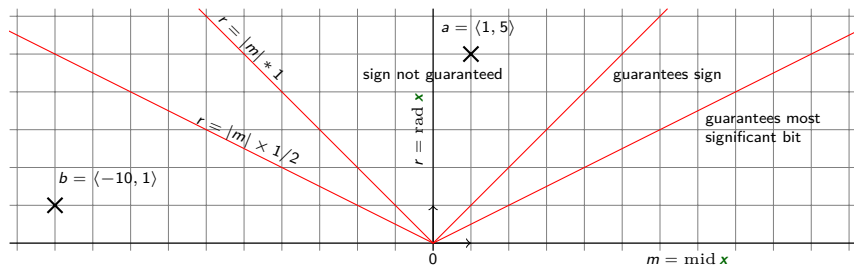
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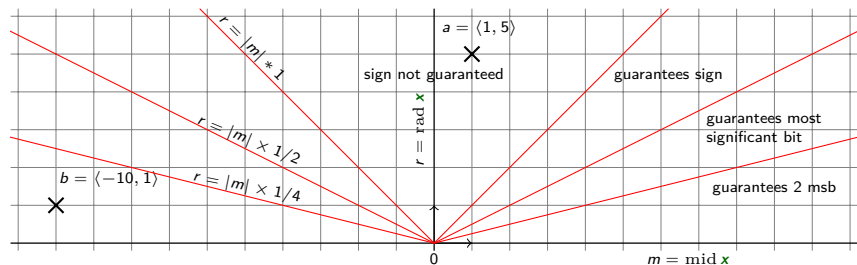
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# Radius of the Product of Matrices of Fixed Relative Precisions

## Proposition

Let  $\mathbf{A}$  with fixed relative precision  $e$  and  $\mathbf{B}$  with fixed relative precision  $f$ .

Then the radii of the approximations of the product  $\mathbf{A} \times \mathbf{B}$  verify

$$\begin{aligned} R_{\mathbf{C}} &= (\max_1 + \max_2) && |M_{\mathbf{A}}| \times |M_{\mathbf{B}}| \\ R_{\mathbf{C}_5} &= (\max_1 + \max_2 + \min_3 - \min_4) && |M_{\mathbf{A}}| \times |M_{\mathbf{B}}| \\ R_{\mathbf{C}_3} &= (\max_1 + \max_2 + \min_3) && |M_{\mathbf{A}}| \times |M_{\mathbf{B}}| \end{aligned}$$

where

$$\begin{aligned} \max_1 &= \max\{e, f\} \\ \max_2 &= \max\{\min\{e, f\}, ef\} \\ \min_3 &= \min\{e, f, ef\} \\ \min_4 &= \min\{1, e, f, ef\} \end{aligned}$$

## Relative Radius Errors

The relative radius error for  $C_5$  is

$$E_5 = \frac{R_{C_5} - R_C}{R_C} = \begin{cases} 0 & \text{if } e \leq 1 \text{ or } f \leq 1 \\ \frac{\min\{e,f\}-1}{\max\{e,f\}+ef} & \text{otherwise} \end{cases}$$

So,  $E_5 \leq 3 - 2\sqrt{2} \leq 0.18$  and maximum is reached for  $e = f = 1 + \sqrt{2}$ .

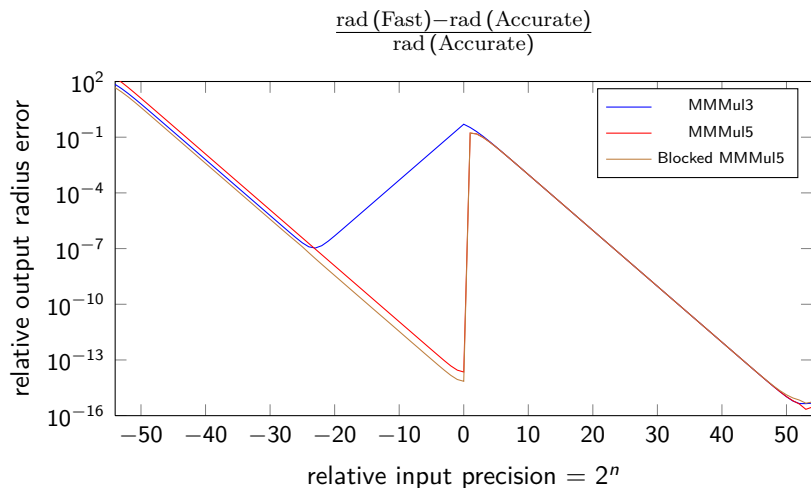
The relative radius error for  $C_3$  is

$$E_3 = \frac{R_{C_3} - R_C}{R_C} = \frac{\min_3}{\max_1 + \max_2}$$

So,  $E_3 \leq 0.5$  and maximum is reached for  $e = f = 1$ .



# Relative Radius Error – Numerical Experiment



Median relative precision of 10 products of  $100 \times 100$  matrices with random midpoint (standard normal distribution) and fixed relative precision  $2^{-k}$ . Floating-point numbers are in double precision.

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# Conclusion: trading off efficiency for accuracy, still being efficient

**A panel of formulas exist** for operations in interval arithmetic using mid-rad representation:

- ▶ trade off accuracy for efficiency, but not always;
- ▶ reduce drastically the number of rounding mode changes;
- ▶ allow to resort to Level-3 BLAS.

## **Accuracy:**

Floating-point error dominates for input intervals with small relative precision.

Rounding modes: beware your BLAS library, beware your compiler.

## **Efficiency:**

Parallelization of algorithms may invalidate some hypotheses.

Are blocked algorithms the best choice to exploit multi-cores (more than 10 cores)?