

**CS 4552**  
**Assignment 1**  
**Matrix Multiplication Performance**

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# Table of Content

<b>Table of Content</b>	<b>1</b>
<b>1.1 Results of the Experiments</b>	<b>2</b>
N = 10	2
N = 100	2
N = 500	3
N = 1000	3
Valgrind Run	4
<b>1.2 Results Analysis</b>	<b>5</b>
N = 10	5
N = 100	5
N = 500	6
N = 1000	6
<b>1.3 Best Loop Orderings</b>	<b>7</b>
<b>1.4 Worst Loop Orderings</b>	<b>7</b>
<b>1.5 Available CPU Instruction Set Extensions</b>	<b>8</b>
<b>1.6 Alternative Further Optimizations</b>	<b>9</b>
<b>1.7 Summary</b>	<b>10</b>

## 1.1 Results of the Experiments

The experiment rounds were run 20 times for each loop order and for each N (10, 100, 500, 1000). Finally, the average execution time was calculated with its variance.

Since the three matrices A, B, and C represent most of the memory utilization, estimated memory utilization was calculated as follows.

$$\text{Estimated Memory Utilization} = \text{No. of matrices} * N * N * \text{sizeof(int or float)}$$

*No. of matrices = 3*

*sizeof(int) = 4 bytes*

*sizeof(float) = 4 bytes*

**N = 10**

Floats			Integers		
Main Mem Available	Est Mem Utilization		Main Mem Available	Est Mem Utilization	
127039784 bytes	3*10*10*4 = 1200 bytes		127039280 bytes	3*10*10*4 = 1200 bytes	
Order	Avg Time	Variance	Order	Avg Time	Variance
IJK	0.000017 s	0.000000 s	IJK	0.000016 s	0.000000 s
IKJ	0.000016 s	0.000000 s	IKJ	0.000016 s	0.000000 s
JKI	0.000017 s	0.000000 s	JKI	0.000016 s	0.000000 s
KJI	0.000016 s	0.000000 s	KJI	0.000016 s	0.000000 s
KIJ	0.000016 s	0.000000 s	KIJ	0.000017 s	0.000000 s
KJI	0.000016 s	0.000000 s	KJI	0.000016 s	0.000000 s

**N = 100**

Floats			Integers		
Main Mem Available	Est Mem Utilization		Main Mem Available	Est Mem Utilization	
127039280 bytes	3*100*100*4 = 120000 bytes		127039280 bytes	3*100*100*4 = 120000 bytes	
Order	Avg Time	Variance	Order	Avg Time	Variance
IJK	0.006380 s	0.000003 s	IJK	0.006237 s	0.000003 s
IKJ	0.005541 s	0.000000 s	IKJ	0.005571 s	0.000000 s

JKI	0.005821 s	0.000000 s	JKI	0.005617 s	0.000000 s
KJI	0.005494 s	0.000000 s	KJI	0.005511 s	0.000000 s
KIJ	0.005529 s	0.000000 s	KIJ	0.005558 s	0.000000 s
KJI	0.005496 s	0.000000 s	KJI	0.005513 s	0.000000 s

**N = 500**

Floats			Integers		
Main Mem Available		Est Mem Utilization	Main Mem Available		Est Mem Utilization
127038208 bytes		$3*500*500*4 = 3000000$ bytes	127037748 bytes		$3*500*500*4 = 3000000$ bytes
Order	Avg Time	Variance	Order	Avg Time	Variance
IJK	0.477692 s	0.004017 s	IJK	0.459795 s	0.000000 s
IKJ	0.578760 s	0.000001 s	IKJ	0.577875 s	0.000001 s
JKI	0.463647 s	0.000000 s	JKI	0.466737 s	0.000000 s
KJI	0.422671 s	0.000000 s	KJI	0.422956 s	0.000000 s
KIJ	0.577101 s	0.000000 s	KIJ	0.578653 s	0.000000 s
KJI	0.423421 s	0.000000 s	KJI	0.423311 s	0.000000 s

**N = 1000**

Floats			Integers		
Main Mem Available		Est Mem Utilization	Main Mem Available		Est Mem Utilization
127037812 bytes		$3*1000*1000*4 = 12000000$ bytes	127035528 bytes		$3*1000*1000*4 = 12000000$ bytes
Order	Avg Time	Variance	Order	Avg Time	Variance
IJK	3.961426 s	0.000003 s	IJK	3.958236 s	0.000025 s
IKJ	4.954249 s	0.006279 s	IKJ	4.904521 s	0.000691 s
JKI	4.177223 s	0.000025 s	JKI	4.170228 s	0.000001 s
KJI	3.379006 s	0.000063 s	KJI	3.374859 s	0.000000 s
KIJ	5.023502 s	0.000011 s	KIJ	4.846281 s	0.000030 s
KJI	3.393048 s	0.000055 s	KJI	3.393101 s	0.000102 s

As mentioned above, the experiment was conducted 20 times for each iteration. With the given results, it is clear that the variance tends to increase with the value of N. This can be the case for several reasons.

- When the matrix size increases, it's more likely to get cache misses and the number of cache misses and cache misses penalties can introduce more variability in execution time.
- When the matrix size increases, the processors can be fully utilized. At full utilization, there can be competition for resources such as the memory bus or arithmetic logic unit (ALU), which can lead to contention and cause more variability in the execution time.

## Valgrind Run

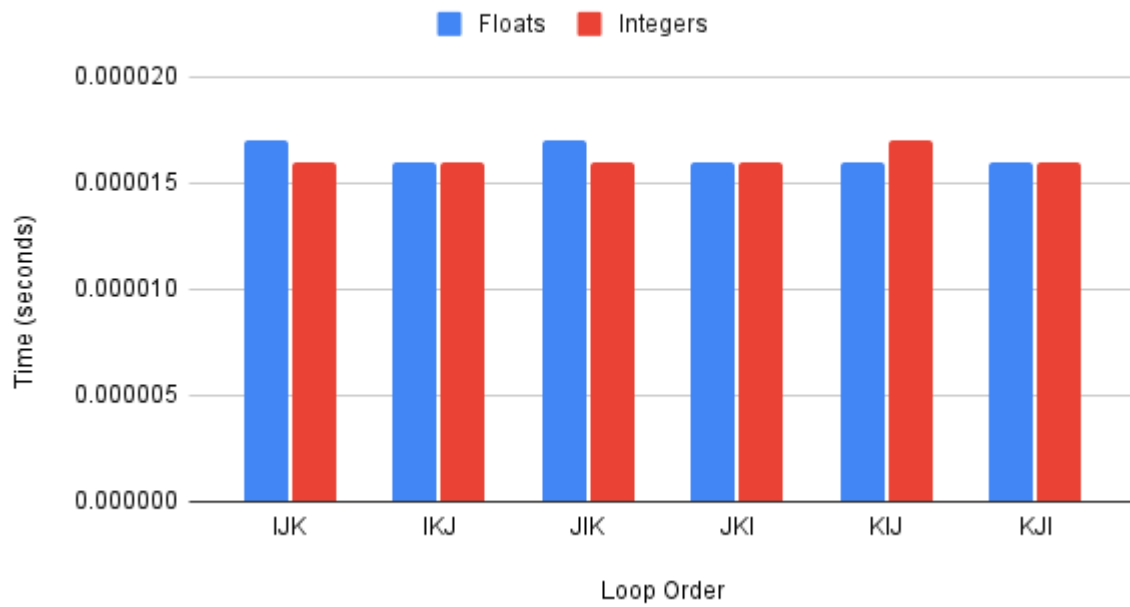
The written code for matrix multiplication was verified with the '*valgrind*' tool for memory leaks and it was found that there aren't any leaks possible. A sample log of a valgrind run is as follows.

```
==3552083== Memcheck, a memory error detector
==3552083== Copyright (C) 2002-2017, and GNU GPL'd, by Julian
Seward et al.
==3552083== Using Valgrind-3.18.1 and LibVEX; rerun with -h for
copyright info
==3552083== Command: ./matMulFloat 100 20
==3552083== Parent PID: 3551992
==3552083==
==3552083==
==3552083== HEAP SUMMARY:
==3552083==       in use at exit: 0 bytes in 0 blocks
==3552083==   total heap usage: 5 allocs, 5 frees, 121,184 bytes
allocated
==3552083==
==3552083== All heap blocks were freed -- no leaks are possible
==3552083==
==3552083== For lists of detected and suppressed errors, rerun
with: -s
==3552083== ERROR SUMMARY: 0 errors from 0 contexts (suppressed: 0
from 0)
```

## 1.2 Results Analysis

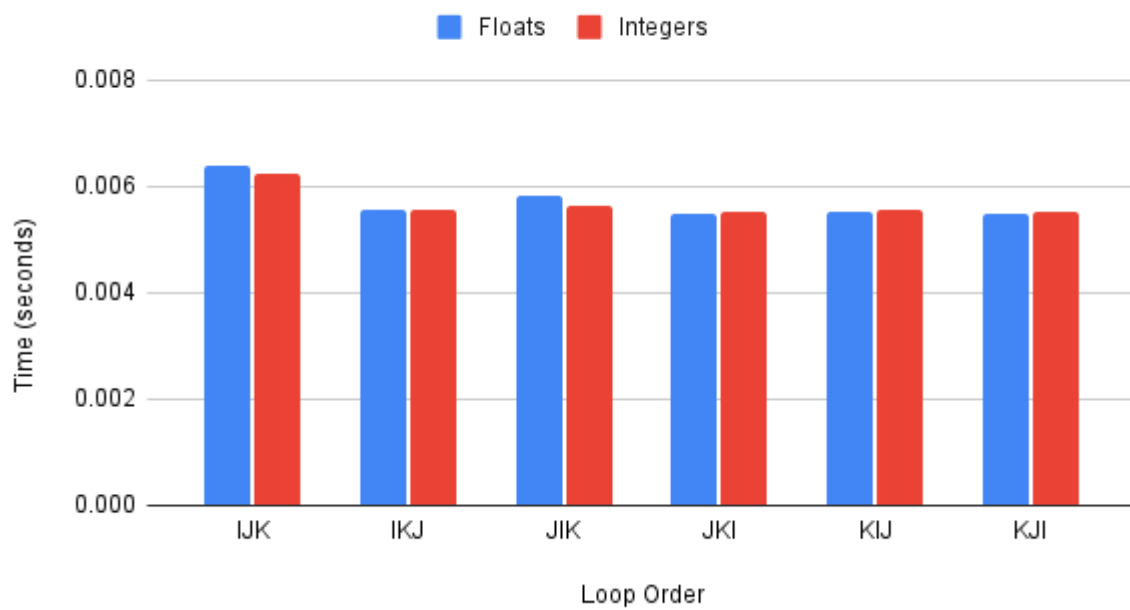
**N = 10**

N = 10



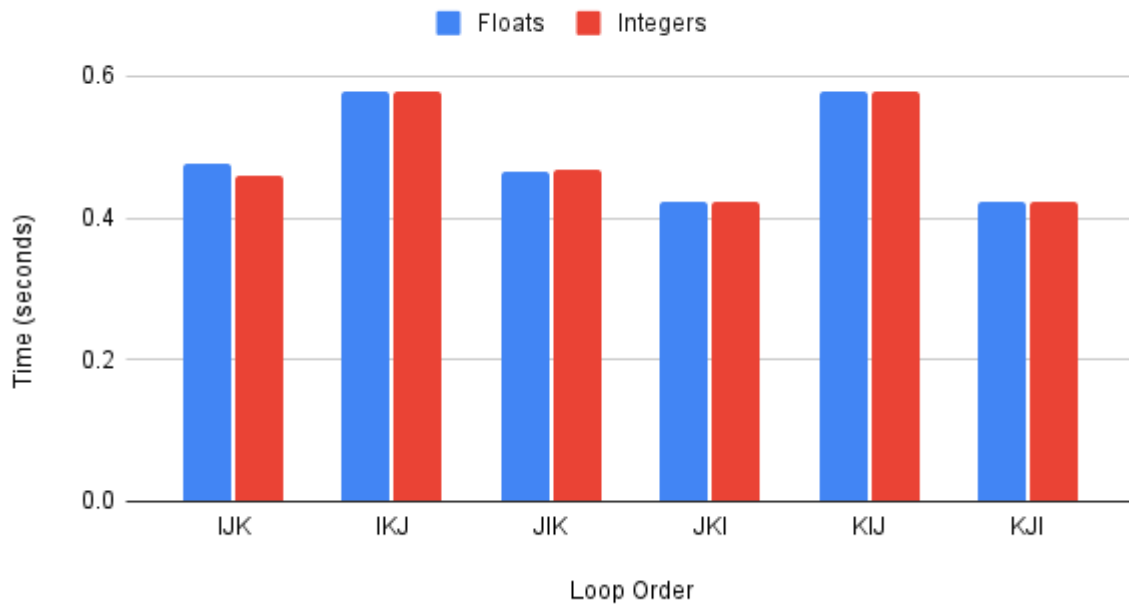
**N = 100**

N = 100



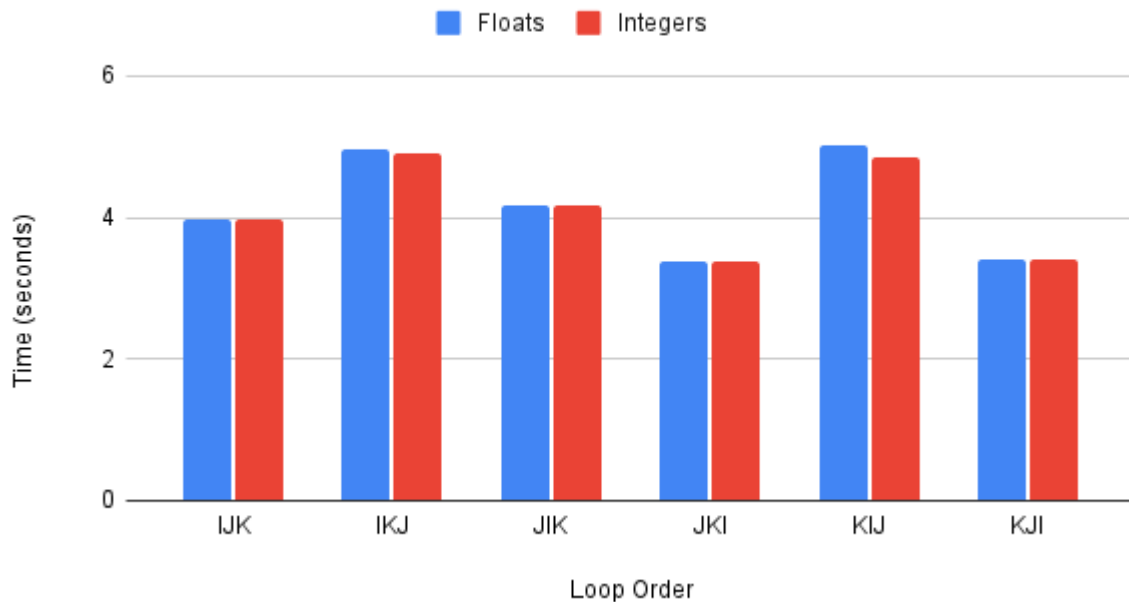
**N = 500**

N = 500



**N = 1000**

N = 1000



For small N (10, 100) values, it is clear that the difference between loop orders is negligible and every loop order performs the same for all the loop orders. But for large N (500, 1000) values, we can see that **JKI** and **KJI** significantly outperform all the other loop orderings.

Furthermore, the execution times for integers are a bit lesser than of floats but it is almost negligible. Even though, the size of floats and ints are the same in C, this time difference can be caused due to the fact that the floating-point arithmetic, which is used to perform operations on float data types, is typically more computationally expensive than integer arithmetic, which is used to perform operations on int data types.

### 1.3 Best Loop Orderings

**JKI** and **KJI** loop orders performed best out of all 6 six loop orders for  $N = 1000$ . Furthermore, JKI is slightly better than KJI and this is the case for both floats and integers. Thus, it is clear that loop orderings with '*i*' within the inner loop give the best performance overall. The way the matrices are implemented on memory and how they are cached while executing the operations are the two factors that affect the performance of different loop orderings.

The JKI and KJI loop orderings are generally faster because, for each iteration of the inner loop, the memory access point of the A iterates one by one  $[A[i+k*n]]$  (Since matrix A is represented by a contiguous block of 1D memory and it's in column-major). Thus, accessing a whole column of A at once, make it faster because the said memory locations are already cached due to being closer in the same contiguous memory block.

### 1.4 Worst Loop Orderings

**IKJ** and **KIJ** loop orders performed worst out of all 6 six loop orders for  $N = 1000$ . Furthermore, IKJ is slightly better than KJI and this is the case for both floats and integers. Thus, it is clear that loop orderings with '*j*' within the inner loop give the worst performance overall.

The IKJ and KIJ loop orderings are generally slower because, for each iteration of the inner loop, the memory access point of the B iterates n by n  $[B[k+j*n]]$  (Therefore, it jumps n by n and the cached memory block is useless. So, the memory should be loaded again). Thus, accessing a whole row of B at once, make it slower.



## 1.5 Available CPU Instruction Set Extensions

Using the `'lscpu'`, the following CPU instruction set extensions can be found in the “Flags” field.

- sse
- sse2
- pclmulqdq
- vmx
- smx
- ssse3
- fma
- sse4\_1
- sse4\_2
- popcnt
- aes
- xsave
- avx
- f16c
- sha
- bmi1
- avx2
- bmi2
- avx512f
- avx512dq
- rdseed
- adx
- avx512ifma
- avx512cd
- sha
- avx512bw
- avx512vl
- xsaveopt
- xsave
- xsave
- avx512vbmi
- avx
- bmi2
- aes
- pclmulqdq
- avx
- avx
- avx
- popcnt

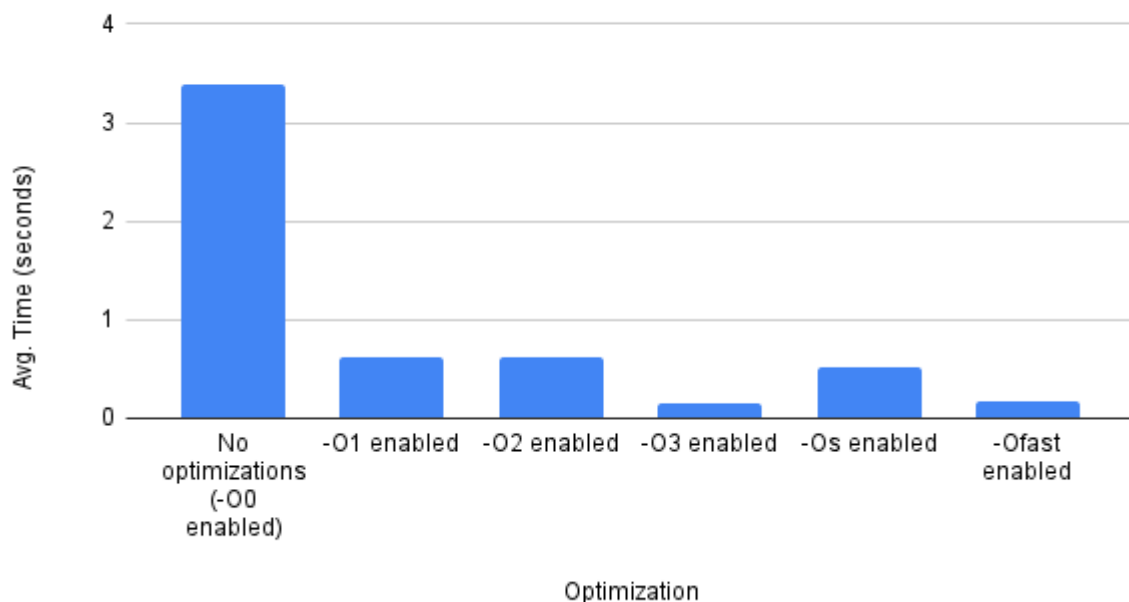
## 1.6 Alternative Further Optimizations

The JKI loop was run for  $N = 1000$  for 20 times for each of the following optimizations. (Only the matrix multiplication of floats was considered.)

- *No optimizations (-O0 enabled)* - default
- *-O1 enabled* - tries to reduce code size and execution time
- *-O2 enabled* - all supported optimizations that do not involve a space-speed tradeoff
- *-O3 enabled* - turns on all optimizations specified by -O2 and a few more
- *-Os enabled* - Optimize for size
- *-Ofast enabled* - enables all -O3 optimizations. It also enables optimizations that are not valid for all standard-compliant programs

Optimization	Avg. Time (seconds)	Variance (seconds)
No optimizations (-O0 enabled)	3.379006	0.000063
-O1 enabled	0.62465	0.007686
-O2 enabled	0.623017	0.007565
-O3 enabled	0.154659	0.000860
-Os enabled	0.526042	0.007237
-Ofast enabled	0.171403	0.000453

### Optimization Analysis



According to the above graph, it is clear that **-O3 enabled** and **-Ofast** gives the best performance and it is almost **20** times better than -O0 enabled (no optimizations). This is due to O3 enabling all the standard optimization parameters for a given compilation and Ofast enabling even non-standard optimizations.

## 1.7 Summary

The experiment was conducted to evaluate the performance of six different loop orderings for matrix multiplication, using matrices of different sizes and data types. The performance was evaluated in terms of execution time and estimated memory utilization.

The results indicated that the choice of loop ordering significantly impacted performance, with loop orderings that included 'i' in the inner loop (JKI and KJI) providing the best performance overall. In contrast, loop orderings with 'j' in the inner loop (IKJ and KIJ) provided the worst performance overall.

The experiment also revealed that the performance variability tended to increase with the value of N. This was likely due to factors such as cache misses, cache misses penalties, and contention for resources at full processor utilization.

In addition to evaluating loop orderings, the experiment also tested the impact of different optimization settings on performance. It was found that -O3 and -Ofast optimizations provided the best performance, resulting in a nearly 20x improvement over the default -O0 setting. These optimization settings enable a range of standard and non-standard optimizations, respectively, to be applied to the compilation process.

Overall, the experiment highlights the importance of carefully selecting loop orderings and optimization settings to achieve optimal performance in matrix multiplication.