

CPSC 131 – Data Structures

Analysis of Algorithms

Professor T. L. Bettens
Spring 2023

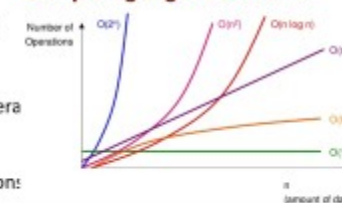
Key terms

- Asymptotic analysis
- Asymptotic efficiency class
- Worst-case analysis
- Big-Oh notation
- Constant time operations $O(1)$
- Logarithmic time operations $O(\log_2 n)$
- Linear time operations $O(n)$
- Quadratic time operations $O(n^2)$

Common classes of Big-Oh functions

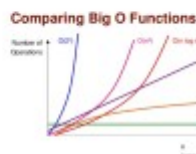
- $O(1)$: Constant time operations
– Does **not** depend on n
- $O(n)$: Linear time operations
– Proportional to n
- $O(\log n)$: Logarithmic time operations
– Bounded by \log_2 of n
- $O(n^2)$: Quadratic time operation:
– Unbounded

Comparing Big O Functions



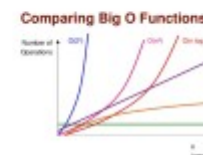
Efficiency Class Examples

- $O(1)$: Constant time operations
- $O(n)$: Linear time operations
- $O(\log n)$: Logarithmic time operations
- $O(n^2)$: Quadratic time operations



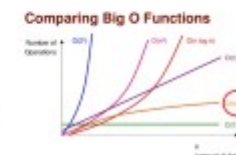
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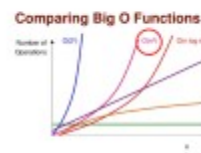
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What about memory requirement?

- So far, we have been speaking of running **time**
- What about how much memory/**space** is occupied by a data structure?
- Can also use $O(n)$ concept:
– Does memory usage go up proportionately to number of elements?



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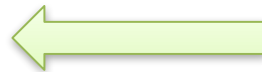
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Key terms

- **Asymptotic analysis**
- **Asymptotic efficiency class**
- **Worst-case analysis**
- **Big-Oh notation**
- **Constant time operations $O(1)$**
- **Logarithmic time operations $O(\log_2 n)$**
- **Linear time operations $O(n)$**
- **Quadratic time operations $O(n^2)$**

Comparing data structures

- Is a doubly linked list better than a singly linked list?
- What does it mean to be better? Under what conditions? What does “good” mean?
 - Running time? Memory used?
 - Usually means a trade-off
- Two approaches to answering this question:
 - Option 1: Experimental analysis
 - Option 2: Asymptotic analysis



Comparing

Data Structure

- Arrays
- Singly linked lists
- Doubly linked lists
- Binary Trees
- Hash Tables

Operations

- Create empty
- Get front element
- Add or remove front element
- Get back element
- Add or remove back element
- Clear data structure
- Get/add/remove i^{th} element



Comparing

- This course's goal: the design of “good” data structures and algorithms
- Data structures: systematic way of organizing and accessing data
- Algorithms: Step-by-step procedure for performing a task in a finite time.
- What does “good” mean?
 - Running time—fast
 - Space usage—small
- Usually means a trade-off
 - Faster often requires more memory for extra pointers
 - Smaller often requires more complex algorithms
- Essential point: **Running time increases with input size**

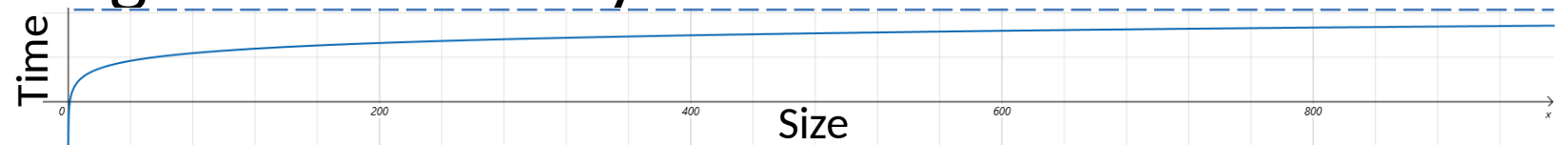


Option 2: Asymptotic Analysis

- Analysis without actually running any code
 - Done at the desktop looking at source code

Look for the loops!

- Asymptotic: approaching a value closely



- Key idea:
 - We are interested in running time for large data sets
 - How fast will running time increase as we increase data size?
 - Rate of increase

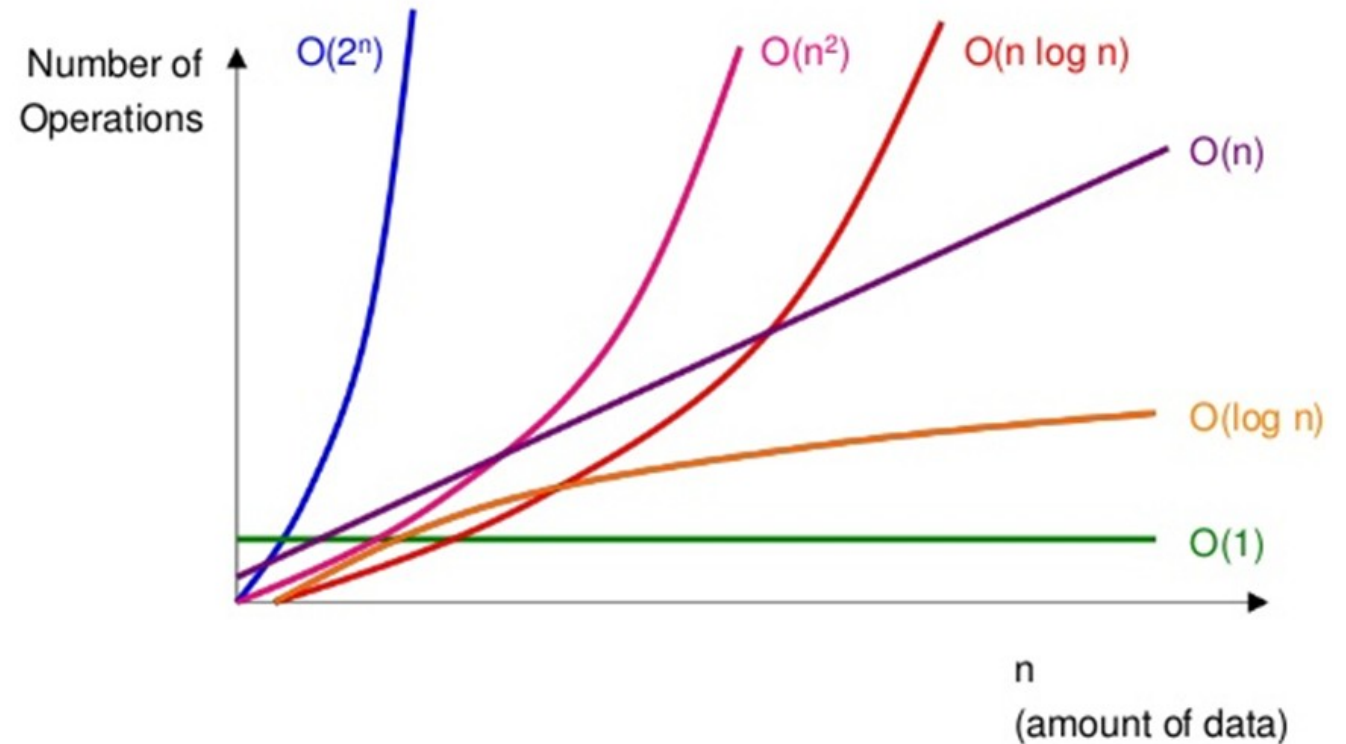
Option 2: Asymptotic Analysis

- Most important factor?
 - Number of elements in the data structure (i.e., size)
- Represent this by n
- Analysis
 - How does running time increase in terms of n ?

Common classes of Big-Oh functions

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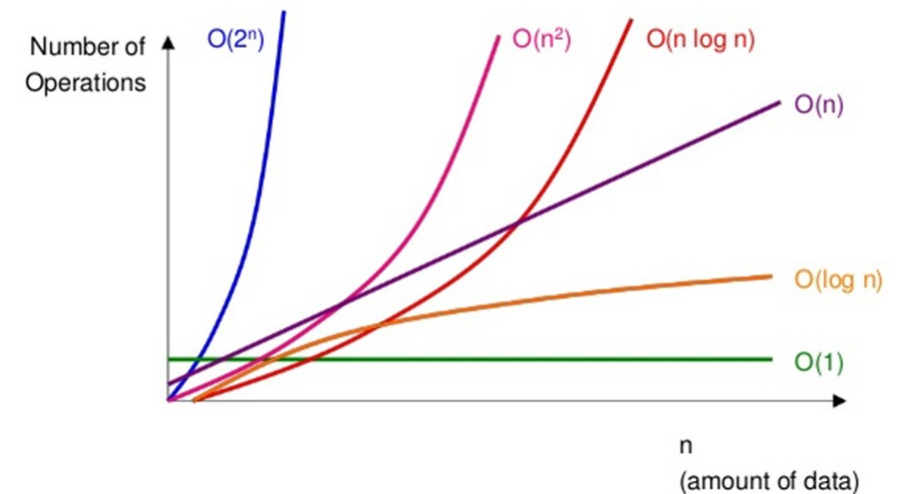
Comparing Big O Functions



Efficiency Class Examples

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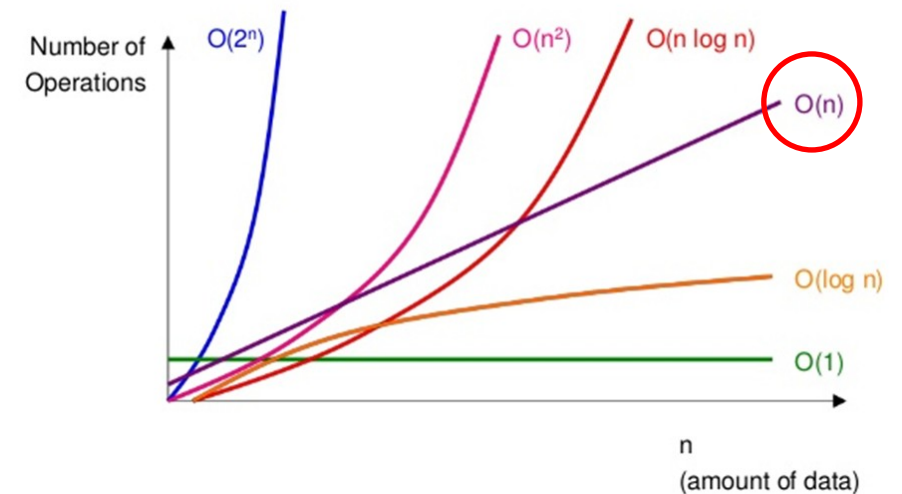
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Comparing Big O Functions



Efficiency Class Example: $O(n)$

Printing a list

- Consider a list of n elements

```
for (const auto & value : list )  
{  
    cout << value << endl;  
}
```

Look for the
loops!

- How many steps did we have to do?
 - Too complex!
- Do the number of steps increase in proportion to n ?



Efficiency Class Example: $O(n)$

Printing a list

- Yes! number of operations increase in proportion to n
- “Printing all elements in a list takes **on the order of n** ”
- Written as **$O(n)$**
- Also commonly spoken as “**Oh of n** ”

Efficiency Class Example: $O(n)$

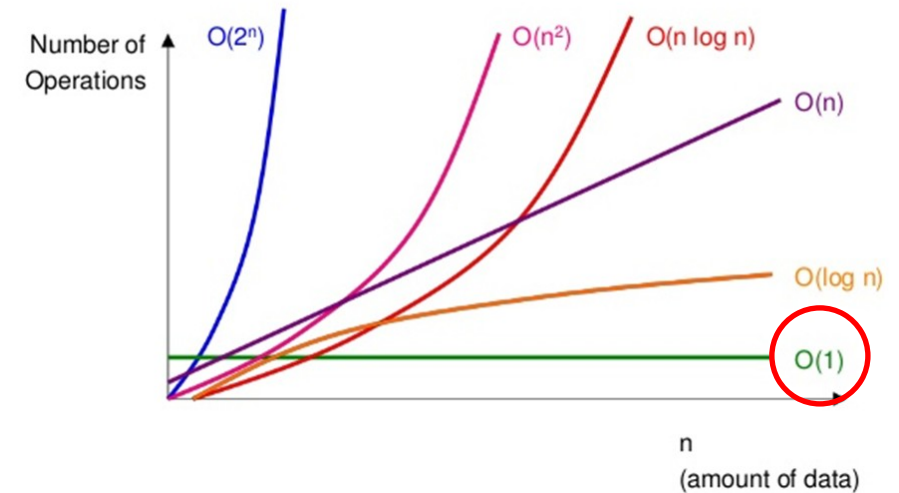
Printing a list

- But what about the fact that we had to
 - initialize the loop
 - Printing required cout
 - We also printed endl
 - ...
- Don't care about constant factors
 - Focus on the big picture
 - Not details like initialization

Efficiency Class Examples

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Comparing Big O Functions



Efficiency Class Example: $O(1)$ Algorithm Independent of “n”

- Return the value of an attribute (e.g., getters)

```
std::size_t size() { return _size; }
```

Look for the
loops!

- Fixed length loops

```
for (unsigned int i = 0; i < 10; ++i )  
{  
    cout << i << '\n';  
}
```

But make sure the
loops are dependent
on 'n'

- Math expressions

```
limit = 2 * sin(x+y) + 3 * z;
```



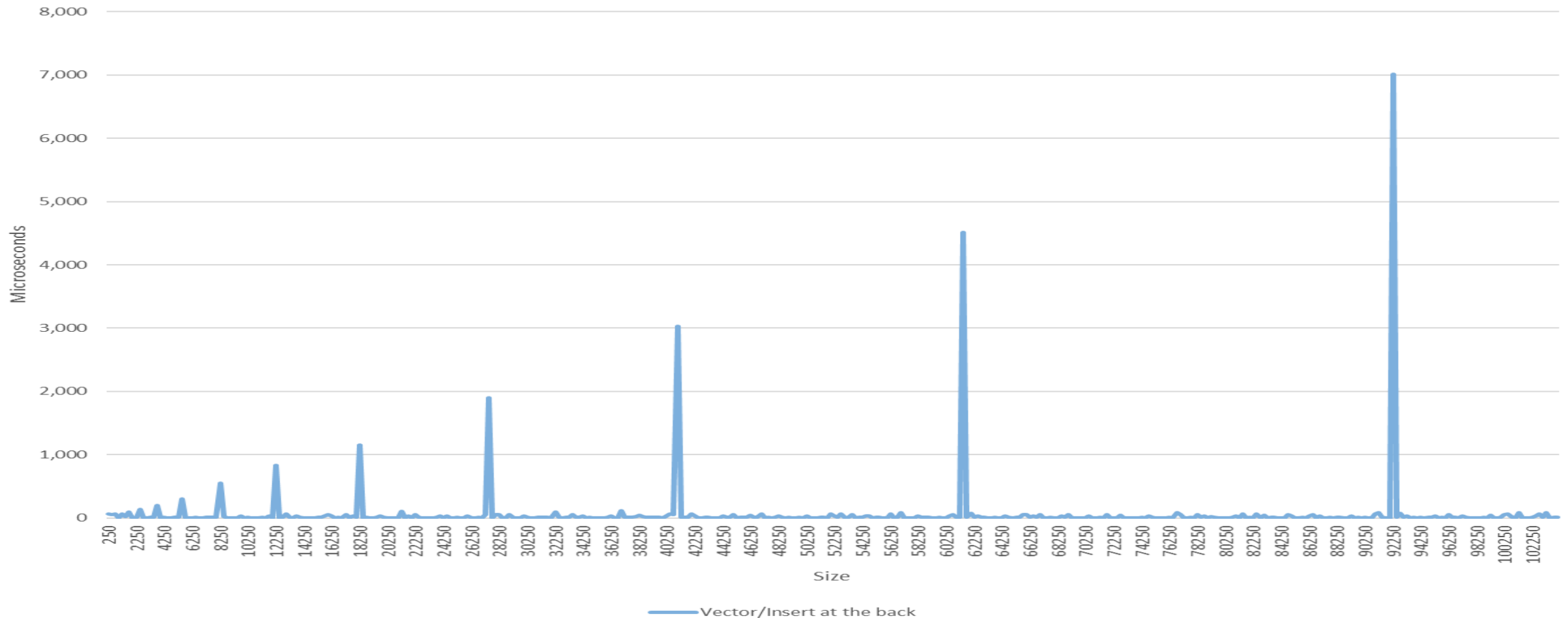
Efficiency Class Example: $O(1)$

Amortization

- Financial: Amortization is paying off an amount owed over time by making planned, incremental payments of principal and interest.
- Computer Science: To even out the costs of running an algorithm over many iterations, so that high-cost iterations are much less frequent than low-cost iterations, which lowers the average running time per iteration.

Efficiency Class Example: Amortized $O(1)$

Time to perform an operation vs Size of container



Efficiency Class Example: Amortized $O(1)$

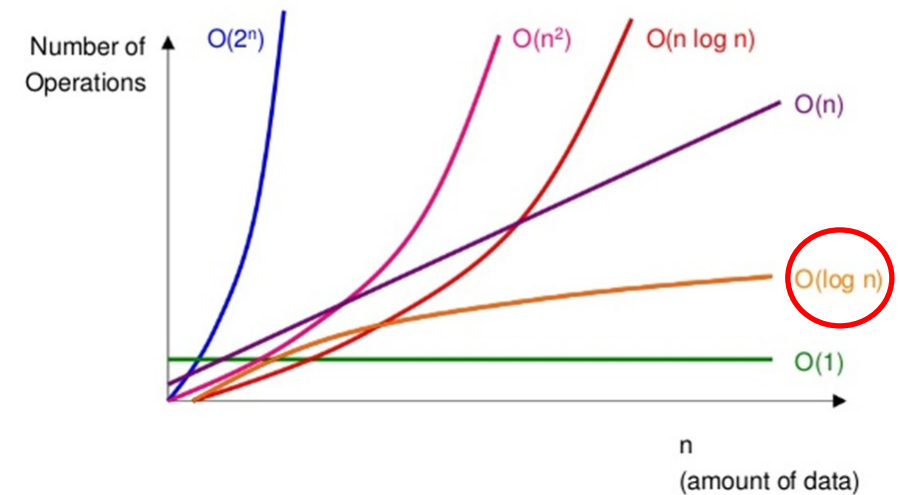
Increase the Extension Size, Reduce the Number of Copy Operations

Insert #	Extend by 1			Double Each Extension		
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1	1	1	0	1	1	
2	2	2	1	2	2	1
3	3	3	2	3	4	2
4	4	4	3	4	4	
5	5	5	4	5	8	4
6	6	6	5	6	8	
7	7	7	6	7	8	
8	8	8	7	8	8	
9	9	9	8	9	16	8
10	10	10	9	10	16	
11	11	11	10	11	16	
12	12	12	11	12	16	
13	13	13	12	13	16	
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Efficiency Class Example: $O(\log n)$

Binary Search

Search for 75

Round 1

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
lwr		mid				upr		

$N = 9$

$75 == 50? (1)$

Round 2

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
				lwr		mid		upr

$75 == 70? (2)$

Round 3

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
						lwr mid		upr

$75 == 80? (3)$

Round 4

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
						upr	lwr	
range is empty								

(4)

- $\log_2(9) = 3.17 \leq 4$
- Any n between 9 and 16 will take at most 4 operations
- Binary search of sorted data: $O(\log n)$



Log₂(n) Example: Binary Search

Search for 75

Round 1

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
lwr		mid				upr		

Round 2

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
					lwr		mid	upr

Round 3

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
						lwr		upr
						mid		

Round 4

0	1	2	3	4	5	6	7	8
10	20	30	40	50	60	70	80	90
						upr	lwr	
range is empty								

```
size_t search( const array<int, 9> & c, int
v )
{
    size_t s      = c.size();
    size_t current = s / 2;

    while( s != 0 )
    {
        if( v == c[current] ) return current;
        if( s == 1             ) break;

        s = ( s + 1 ) / 2;

        if( v < c[current] ) current -= s / 2;
        else                 current += s / 2;
    }

    return numeric_limits<size_t>::max();
}
```

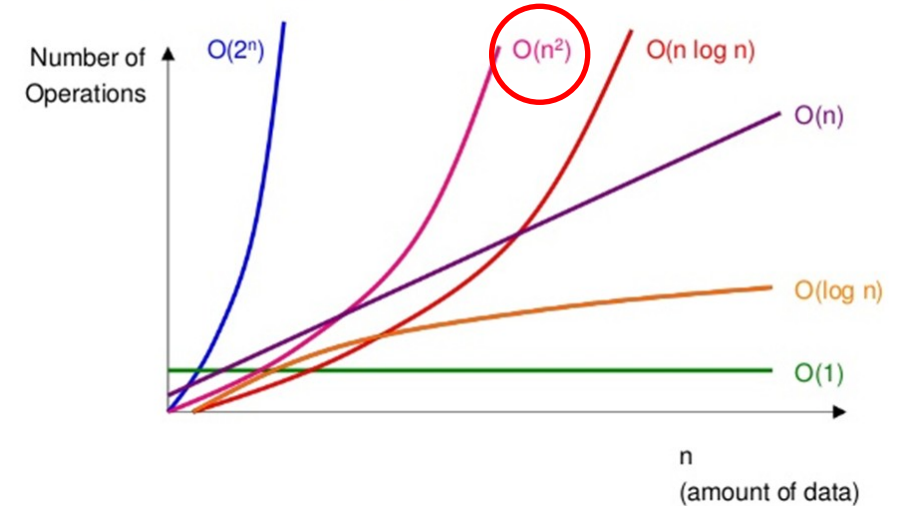
Look for the loops!



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Comparing Big O Functions



Efficiency Class Example: $O(n^2)$

Visit each cell of a Matrix

$N = 4$

4 X 4 Matrix

For each row r visit each column c

	0	1	2	3
0	(0, 0)	(0, 1)	(0, 2)	(0, 3)
1	(1, 0)	(1, 1)	(1, 2)	(1, 3)
2	(2, 0)	(2, 1)	(2, 2)	(2, 3)
3	(3, 0)	(3, 1)	(3, 2)	(3, 3)

Efficiency Class Example: $O(n^2)$

Visit each cell of a Matrix

```
// Defines an NxN matrix containing elements of type E
template<typename E, std::size_t N>
using Matrix = std::array<std::array<E, N>, N>;

// Determine if an NxN matrix is symmetrical along its major axis
template<typename E, std::size_t N>
bool isSymmetrical( const Matrix<E, N> & matrix )
{
    for( std::size_t row = 0; row < matrix.size(); ++row )
        for( std::size_t col = 0; col < matrix[row].size(); ++col )
        {
            if( ( row != col ) && ( matrix[row][col] != matrix[col][row] ) ) return false;
        }

    return true;
}
```

Look for the loops!

In this case,
Nested loops!

What about memory requirement?

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