



Exercise Session 02 – Containers, Templates

Data Structures and Algorithms

These slides are based on those of the lecture, but were adapted and extended by the teaching assistant Adel Gavranović

Today's Schedule

Intro
Follow-up
Feedback for **code expert**
Learning Objectives
C++ Container Library
Templates Recap
Repetition theory: Induction
Subarray Sum Problem
Code Example
Programming Exercise
Tips for **code expert**
Outro



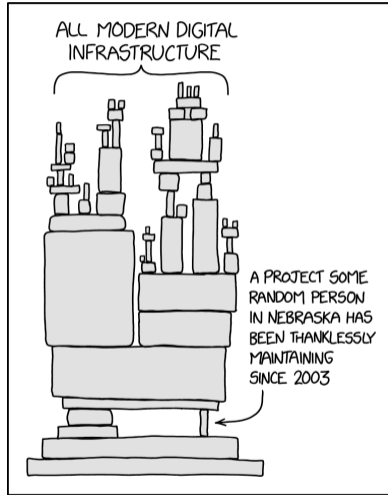
`n.ethz.ch/~agavranovic`

▶ Exercise Session Material

▶ Adel's Webpage

▶ Mail to Adel

Comic of the Week



1. Intro

Intro

- Welcome Back!

- Welcome Back!
- There was a miscommunication regarding exercise sessions in first week – Sorry for that!

2. Follow-up

Follow-up from last exercise session

- There's a **code expert** sandbox¹ now! (To try out code outside of exercises)

¹Can be found under "code examples" at the top

Follow-up from last exercise session

- There's a **code expert** sandbox¹ now! (To try out code outside of exercises)
- That one confusing Runtime-Slide

¹Can be found under "code examples" at the top

Slide from last session "A good strategy?"

If today I can solve a problem of size n (in some fixed time), then with a 10 or 100 times faster machine I can solve ... ²

Complexity of Algorithm	(speed $\times 10$)	(speed $\times 100$)
$\log_2 n$	$n \rightarrow n^{10}$	$n \rightarrow n^{100}$
n	$n \rightarrow 10 \cdot n$	$n \rightarrow 100 \cdot n$
n^2	$n \rightarrow 3.16 \cdot n$	$n \rightarrow 10 \cdot n$
2^n	$n \rightarrow n + 3.32$	$n \rightarrow n + 6.64$

²To see this, you set $f(n') = c \cdot f(n)$ ($c = 10$ or $c = 100$) and solve for n'

Main Takeaway

- Faster computers won't be able to compensate for inefficient algorithms, since the increase in problem size that a significantly faster computer allows is uselessly small
 - e.g. from $n = 4$ to $n' \approx 7$ (per unit of time) in case of an algorithm of complexity $\mathcal{O}(2^n)$ if the new computer runs 10-times faster than the old

Main Takeaway

- Faster computers won't be able to compensate for inefficient algorithms, since the increase in problem size that a significantly faster computer allows is uselessly small
 - e.g. from $n = 4$ to $n' \approx 7$ (per unit of time) in case of an algorithm of complexity $\mathcal{O}(2^n)$ if the new computer runs 10-times faster than the old
- Seriously, just write efficient code

3. Feedback for **code** expert

General things regarding **code** expert

General things regarding **code expert**

- Nothing yet since the deadline for the current is tonight 23:59

Questions regarding **code expert** from your side?

4. Learning Objectives

Learning Objectives

- Understand what Containers are
- Understand what benefits Containers bring
- Understand what Templates are
- Understand what benefits Templates bring
- Understand how to do Induction Proofs in this course
- Be prepared to solve the next **code expert** exercises

5. C++ Container Library

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- Each Container has its use cases

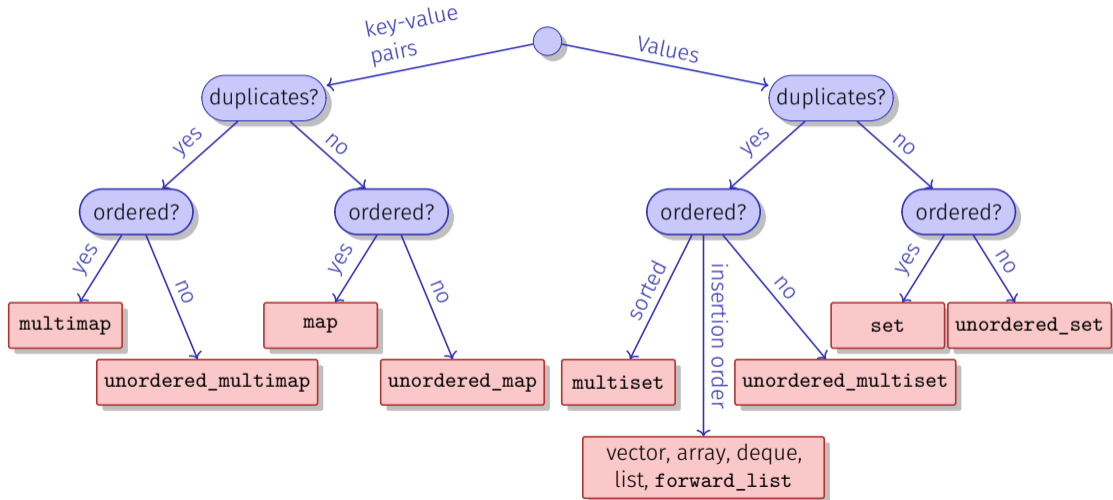
What are containers *abstractly*?

- Essentially, a container is some sort of organized collection of things
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- ...since you will be familiar with many of them by the end of this course because you will study some of them very closely

What are containers *abstractly*?

- Essentially, a container is some sort of organized collection of things
- Each Container has its benefits and drawbacks
- Each Container has its use cases
- Don't bother learning them by heart, ...
 - ...since you will be familiar with many of them by the end of this course because you will study some of them very closely
- Each Container comes with its own cool helper-functions!
 - e.g. `.push_back()` for our beloved `std::vector`

C++ Containers



Sequence-Container

<code>vector</code>	<code>array</code>	<code>deque</code>	<code>list</code>	<code>forward_list</code>
contiguous dynamic memory	contiguous static memory	Non-contig. dynamic memory	Non-contig. dynamic memory	Non-contig. dynamic memory
random access	random access	random access		
fast push/pop back		fast push/pop front/back	fast push/pop front/back	fast push/pop front
bidirectional iteration	bidirectional iteration	bidirectional iteration	bidirectional iteration	forward iteration

dynamic: size can change during runtime, **static:** size fixed at compile-time, **random access:** direct, immediate access to any element by its *index* (e.g. `vec[42]`), **bidirectional:** backward and forwards iterable

Sets and Multisets

- `std::set<E>` contains unique elements
- `std::multiset<E>` allows duplicate elements
 - Iteration yields all elements in decreasing order (in non-deterministic order if `unordered_multiset`)
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Example of `std::multiset`

```
Content: Xanten Xenon Xenon Xenon Xerografie Xerophil Xylose  
count("Xenon") = 3  
count("Xylose") = 1
```

Maps and Multimaps

- `std::map<K,V>` contains pairs (key, value), where a key maps to at most one value
- `std::multimap<K,V>` allows duplicate pairs
 - Iteration yields all pairs in descending key order (in non-deterministic order, if `unordered_multimap`)
 - `std::multimap<K,V>::count(key)` returns the number of occurrences of a given key
 - `std::multimap<K,V>::equal_range(key)` returns all values (in non-det. order) for a given key

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Example of `std::multimap<K,V>`

```
Content: {2, er} {2, du} {2, es} {3, Axt} {3, sie} {4, Igel}
```

```
count(2) = 3
```

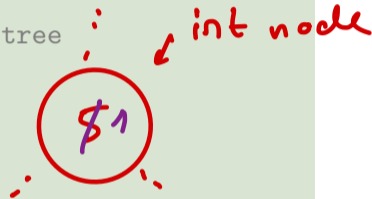
```
Values for key 2: er du es
```

6. Templates Recap

Motivation

Goal: generic binary tree without duplicating code

```
class Node { ... }; // Node of a binary search tree
auto n1 = Node<int>(5);
auto n2 = Node<std::string>("Zürich");
n1.insert(1);
n2.contains(2); // Compiler error
```



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```

Idea:

- Make classes and functions parametric in types (= template parameters) ...
- ... just as they are already parametric in values (= function parameters)

Types as Template Parameters

1. In the concrete implementation of a class replace the type that should become generic (e.g. `int`) by a representative element, e.g. `T`.
2. Put in front of the class the construct `template<typename T>` (Replace `T` by the representative name).

The construct `template<typename T>` can be understood as **“for all types T”**.

Class template

```
template <typename K>
class Node {
    K key;
    Node* left, right;
public:
    Node(K k, Node* l, Node* r): key(k), left(l), right(r) {}

    bool contains(K search_key) const {
        return search_key == key
            || left != nullptr && left->contains(search_key)
            || right != nullptr && right->contains(search_key)
    }
    ...
};
```



Function Template: Analogous Approach

1. To make a concrete implementation generic, replace the specific type (e.g. `int`) with a name, e.g. `T`,
2. Put in front of the function the construct `template<typename T>` (Replace `T` by the chosen name)

Examples

- For free functions

```
template <typename T>
void swap(T& x, T& y) {
    T temp = x;
    x = y;
    y = temp;
}
```

- For operators

```
template <typename T>
ostream& operator<<(ostream& out, const Node<T> root) {
    ...
}
```

```
template <typename Iter>
void is_sorted(Iter begin, Iter end){
    ... ←
}
```

*// map rec --
... is_sorted(rec.begin, rec.end)*

Semantics (Code-Generation)

For each template instance, the compiler creates a corresponding instantiated class (or function) → static code generation

```
Node<int> n1 = ...;  
Node<std::string> n2 = ...;  
Node<Student> n3 = ...;
```

n1

```
class Node_int {  
    int key;  
    ...  
    bool contains(int k) {...}  
    int max() {...}  
};
```

n2

```
class Node_string {  
    std::string key;  
    ...  
};
```

n3

```
class Node_Student {  
    Student key;  
    ...  
};
```

Semantics (Code-Generation)

For each template instance, the compiler creates a corresponding instantiated class (or function) → static code generation

Question: what does this imply for separate compilation?

- Should templates go into .h (declarations) or .cpp (definitions) files?
- Is it possible to ship the compiled implementation (binary file compiled from .cpp) alongside the header file?

Generalizing Code using Templates

```
→  
class Vector {  
public:  
    Vector() {...}  
    float& operator [] (int i) { return data[i]; }  
private:  
    float data[3];  
};
```

```
→  
float scalar_product(Vector a, Vector b) {  
    float result = 0;  
    for (int i=0; i<3; ++i)  
        result += a[i] * b[i];  
    return result;  
}
```

Generalizing Code using Templates

```
template <typename T>
class Vector {
public:
    Vector() {...}
    T& operator [] (int i) { return data[i]; }
private:
    T data[3];
};
```

```
template <typename T>, T s
T scalar_product (Vector<T> a, Vector<T> b) {
    T result = 0;
    for (int i=0; i<3; ++i) s
        result += a[i] * b[i];
    return result;
}
```

Generalizing Code using Templates

```
template <unsigned N, typename T>
class Vector {
public:
    Vector() {...}
    T& operator [] (int i) { return data[i]; }
private:
    T data[N];
};
```

```
template <unsigned N, typename T>
T scalar_product(Vector<N, T> a, Vector<N, T> b) {
    T result = 0;
    for (int i=0; i<N; ++i)
        result += a[i] * b[i];
    return result;
}
```

Type testing

- Templates: syntactic checks
- Instances: checks as usual

```
template <typename T>
T abs(T v) {
    return 0 <= v ? v : -v;
}
// main
abs(8); // OK
```

```
template <typename T>
void swap(T& x, T& y) {
    ...
}
// main
double a = 1.0;
double b = 7;
swap(a, b); // OK
```

```
template <typename T>
T abs(T v) {
    return 0 <= v ? v : -v; // Error
}
// main
abs("hi"); // Error
```

```
template <typename T>
void swap(T& x, T& y) {
    ...
}
// main
double a = 1.0;
string b = "seven";
swap(a, b); // Error
```

Other Languages

All languages try to foster code reuse but chose different solutions.

- C++, Rust:
 - static code generation
 - no runtime overhead
 - difficult to integrate into OOP
- C#, Scala (, Java)
 - type parameters are turned into runtime values
 - well-suited for OOP
 - minor runtime overhead
- Python, JavaScript:
 - dynamic typing (duck typing)
 - no syntactic overhead
 - potentially significant runtime overhead

6.1 auto vs templates

auto

■ Placeholder type specifier

- Must be uniquely determined by direct context: initialiser code, or returns
- User could write type themselves, but leave it to the compiler

```
std::vector<int> vec = ...;  
auto it = vec.cbegin();   
// placeholder for std::vector<int>::const_iterator
```

*changes it
but not elem
(read-only)*

■ Failing examples:

```
auto x; // x has no initializer  
x = 0.0;  
auto first_or_else(std::vector<int> data, unsigned int or_else) {  
    if (data.size() == 0) return or_else;  
    else return data[0];  
}
```

Templates

- Parameters are unknown until instantiated

```
template <typename N>
char sign(N v) {
    if (0 <= v) return '+';
    else return '-';
}
```

```
template <typename T1, typename T2>
struct Pair {
    T1 fst;
    T2 snd;
};
```

- Instantiation may happen anywhere

```
Pair<int, double> p1 = Pair{1, 0.1};
auto p2 = Pair<std::string, bool>{"Brazil", true};
```


Combining templates and auto

auto inside template must be determined after instantiation

```
template <typename C>
void print(C container) {
    for (auto& e : container)
        std::cout << e << ' ';
}
```



```
std::vector<int> numbers = {1, 2, 3};
print(numbers); // now auto can be determined
```

```
std::vector<std::string> airports = {"LAX", "LDN", "ZHR"};
print(airports); // now auto can be determined
```

Combining templates and auto

auto inside template must be determined after instantiation

```
template <typename C>
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Question: Is it possible to not use auto here?

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auto inside template must be determined after instantiation

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template <typename C>
void print(C container) {
    for (auto& e : container)
        std::cout << e << ' ';
}
```

Question: Is it possible to not use auto here?

Answer: Yes, for example by replacing auto with an additional template parameter E

From auto to templates

- Before C++20 auto function parameters are forbidden

```
void print(auto x) {...} // Compiler error
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void print(auto x) {...} // ok
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Clearly, it is still not possible to determine what auto stands for.

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Question: What could be the meaning of auto in this case?

From auto to templates

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void print(auto x) {...} // Compiler error
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Question: Why do you think that is?

Answer: Cannot determine type from context

- Since C++20 auto function parameters are allowed

```
void print(auto x) {...} // ok
```

Clearly, it is still not possible to determine what auto stands for.

Question: What could be the meaning of auto in this case?

Answer: It is a shorthand for a template parameter!

```
template <typename T>  
void Print(T x){ ... }
```

7. Repetition theory: Induction

Induction: what is required?

- Prove statements, for example $\sum_{i=1}^n i = \frac{n(n+1)}{2}$.

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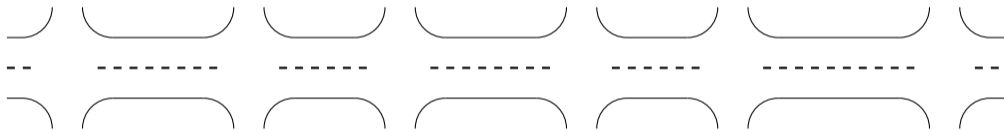
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 - e.g. $\sum_{i=1}^1 i = 1 = \frac{1(1+1)}{2}$.
- Induction hypothesis: we assume that the statement holds for some n
- Induction step ($n \rightarrow n + 1$):
 - From the validity of the statement for n (induction hypothesis) it follows the one for $n + 1$.
 - e.g.: $\sum_{i=1}^{n+1} i = n + 1 + \sum_{i=1}^n i = n + 1 + \frac{n(n+1)}{2} = \frac{(n+2)(n+1)}{2}$.

8. Subarray Sum Problem

Naïve Solution, prefix sums, binary search, Sliding Window

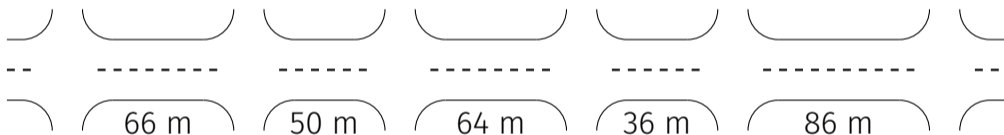
Street section of a given length

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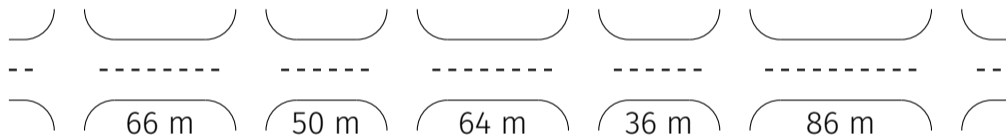
Street section of a given length

Given: distances between all crossroads on a street



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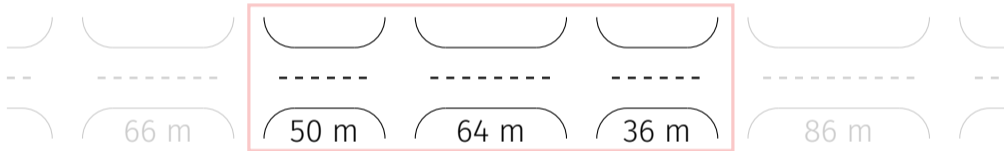
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Subarray Sum Problem

Subarray Sum Problem

Given: a sequence $a[0], \dots, a[n - 1]$ of non-negative integers

Wanted: a subsequence with sum k :

pair (l, r) with $0 \leq l \leq r \leq n - 1$ such that $\sum_{i=l}^r a[i] = k$

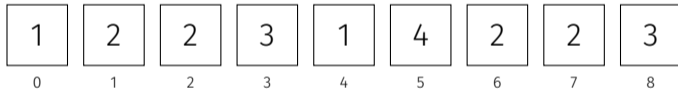
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Example: $n = 9, k = 7$



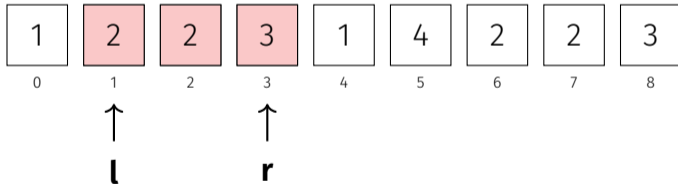
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Example: $n = 9, k = 7$ **Solution:** $l = 1, r = 3$.



Strategies?

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Strategies

$\Theta(n^3)$	Three loops
$\Theta(n^2)$?
$\Theta(n \log n)$?
$\Theta(n)$?

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$\Theta(n^3)$	Three loops
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Strategies

$\Theta(n^3)$	Three loops
$\Theta(n^2)$	Prefix Sums
$\Theta(n \log n)$	Binary Search
$\Theta(n)$	Sliding Window

Subarray Sum Problem: Sliding Window

Sliding Window Idea

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Subarray Sum Problem: Sliding Window

Sliding Window Idea

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Subarray Sum Problem: Sliding Window

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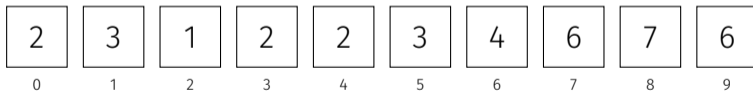
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 - window **too large** ($\text{sum} > k$) \Rightarrow increment left pointer
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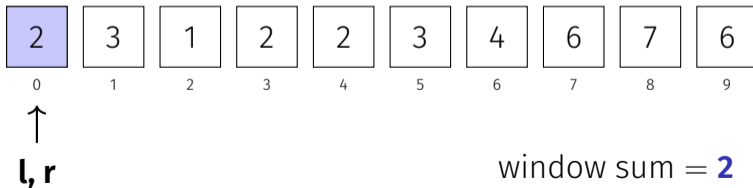


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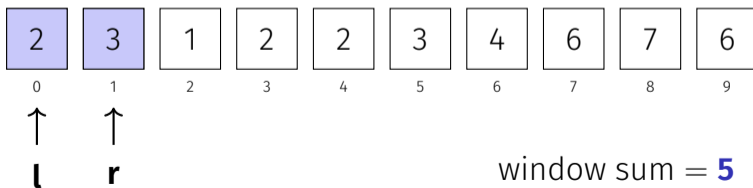


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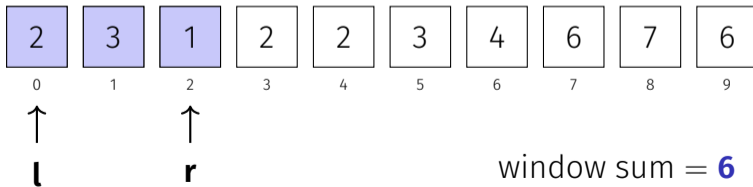


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Example: $k = 7$

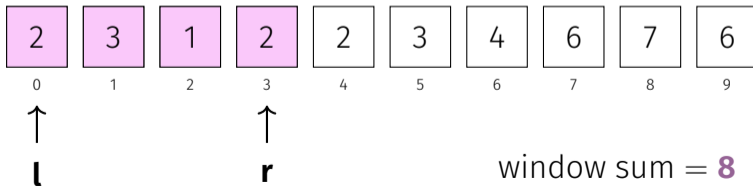


Subarray Sum Problem: Sliding Window

Sliding Window Idea

- start with left and right pointer at 0
- repeat until the end of the sequence:
 - window **too small** ($\text{sum} < k$) \Rightarrow increment right pointer
 - window **too large** ($\text{sum} > k$) \Rightarrow increment left pointer
 - window **as desired** ($\text{sum} = k$) \Rightarrow done!

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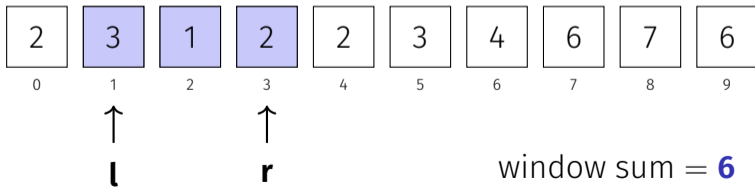


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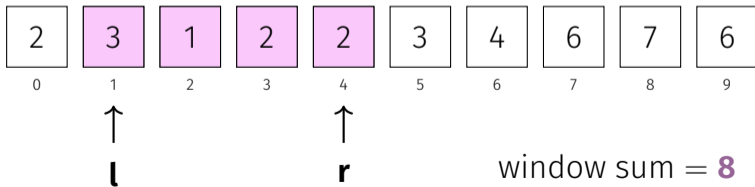


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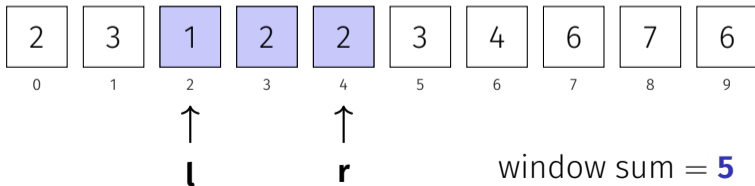


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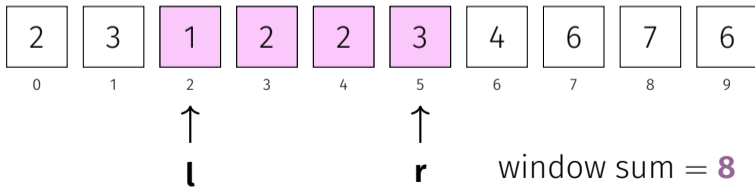


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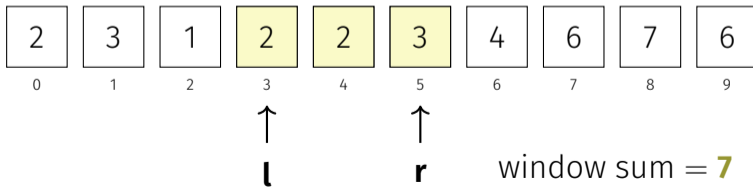


Subarray Sum Problem: Sliding Window

Sliding Window Idea

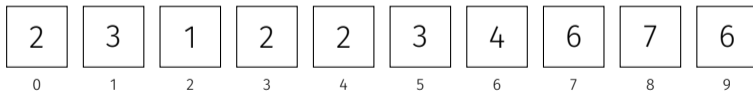
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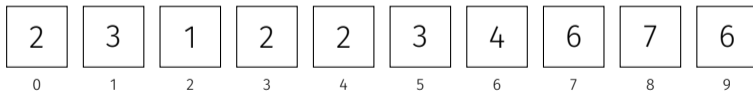
Subarray Sum Problem: Sliding Window Analysis

Subarray Sum Problem: Sliding Window Analysis



Subarray Sum Problem: Sliding Window Analysis

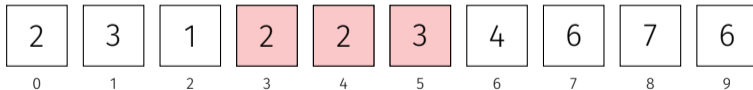
- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps



Subarray Sum Problem: Sliding Window Analysis

- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

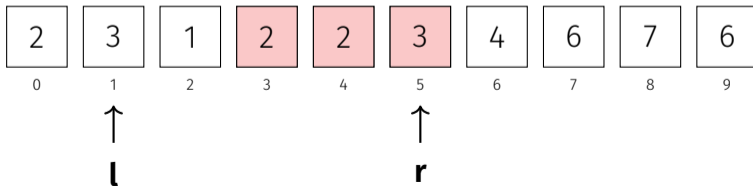


Subarray Sum Problem: Sliding Window Analysis

- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start

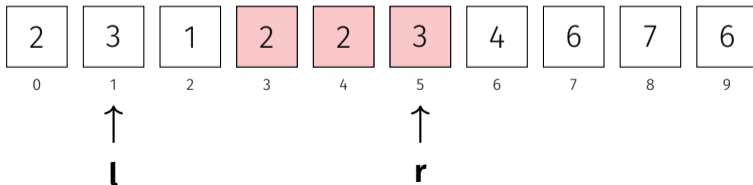


Subarray Sum Problem: Sliding Window Analysis

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⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start
⇒ sum too large

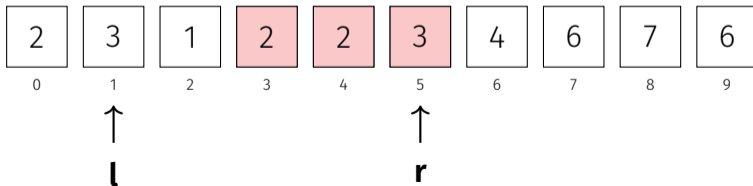


Subarray Sum Problem: Sliding Window Analysis

- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start
⇒ sum too large ⇒ l is increased until it reaches the start of the window

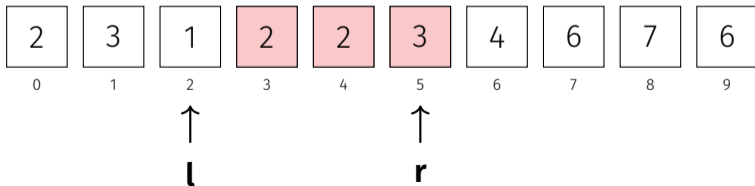


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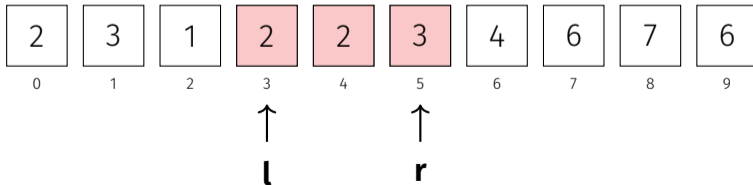


Subarray Sum Problem: Sliding Window Analysis

- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start
⇒ sum too large ⇒ l is increased until it reaches the start of the window

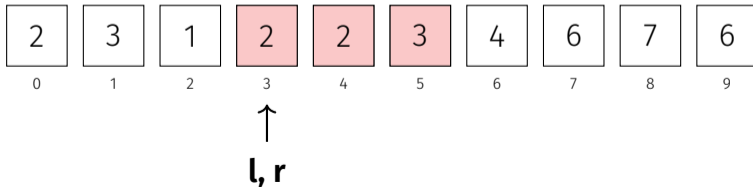


Subarray Sum Problem: Sliding Window Analysis

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⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start
⇒ sum too large ⇒ l is increased until it reaches the start of the window
- if l reaches the start before r reaches the end

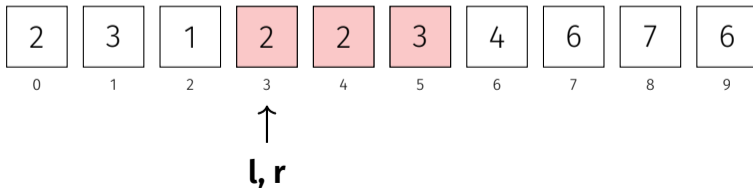


Subarray Sum Problem: Sliding Window Analysis

- in each step: either l or r is increased
⇒ algorithm terminates after a maximum of $2n$ steps

target window: lexicographically smallest (left-most) window with sum k

- if r reaches the end before l reaches the start
⇒ sum too large ⇒ l is increased until it reaches the start of the window
- if l reaches the start before r reaches the end
⇒ sum too small

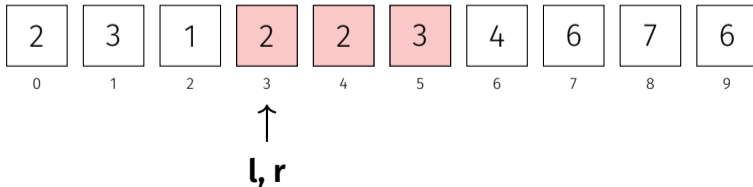


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- in each step: either l or r is increased
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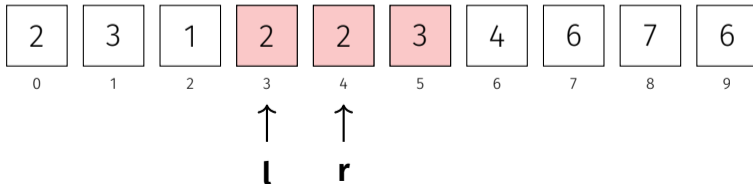


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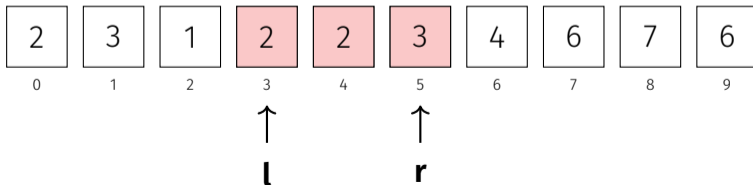


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⇒ sum too large ⇒ l is increased until it reaches the start of the window
- if l reaches the start before r reaches the end
⇒ sum too small ⇒ r is increased until it reaches the end of the window



Analysis

We consider the lexicographically smallest (left-most) window with sum k , called *target window*

- In each step of the algorithm either l or r is increased. The algorithm terminates after a maximum of $2n$ steps.
- Assume r reaches the end of the target window before l reaches the start of the target window, then l keeps increasing until it reaches the start of the window.
- Assume l reaches the start of the target window before r reaches the end of the target window, then r keeps increasing until it reaches the end of the window.

Exercise: window with sum closest to k

9. Code Example

10. Programming Exercise

Preparing remarks for the homework (Prefix Sum in 2D)

Sum in Subarray (naive algorithm)

Input: A sequence of n numbers $(a_0, a_1, \dots, a_{n-1})$ and a sub-interval

$$I = [x_0, x_1]$$

Output: $\sum_{i=x_0}^{x_1} a_i$.

$\mathcal{S} \leftarrow 0$

for $i \in \{x_0, \dots, x_1\}$ **do**

$\mathcal{S} \leftarrow \mathcal{S} + a_i$

return \mathcal{S}

Sum in Subarray (naive algorithm)

Input: A sequence of n numbers $(a_0, a_1, \dots, a_{n-1})$ and a sub-interval

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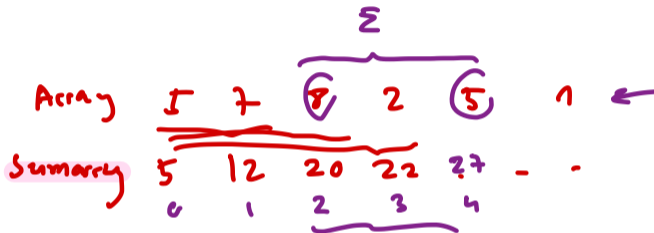
Output: $\sum_{i=x_0}^{x_1} a_i$.

$S \leftarrow 0$

for $i \in \{x_0, \dots, x_1\}$ **do**

$S \leftarrow S + a_i$

return S



Idea of the exercise

- Use the **prefix sum** to compute the sum of arbitrary sub-intervals with constant running time
- **Generalize** to two dimensions.

11. Tips for **code** expert

Tips for **code expert** Exercise 2

Task "Prefix Sum in 2D"

³There's an implementation in the code examples on **code expert**

Tips for **code expert** Exercise 2

Task "Prefix Sum in 2D"

- Study the Prefix Sum in 1D³ well and go from there
- Make sketches!

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- The binomial formula will be useful for the second one
- Please format it well or just scan a PDF and upload it

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- The binomial formula will be useful for the second one
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Task "Karatsuba Ofman"

- Just translate the math into code

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12. Outro

General Questions?

See you next time

Have a nice week!