

Lazy Employee

An office employee takes L minutes to serve a customer.
How slowly can he work?

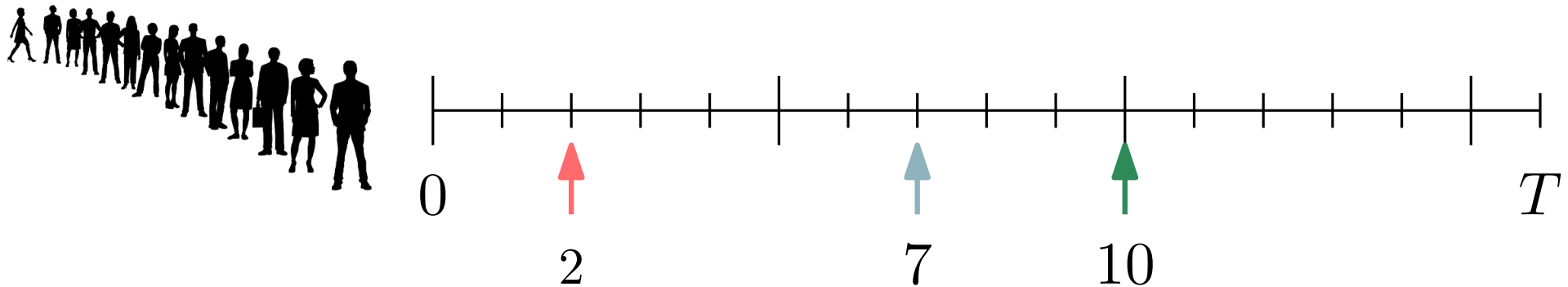
- n customers arrive at distinct times $t_1, t_2, \dots, t_n \in \mathbb{N}^+$
- All customers must leave by time T (closing time, $T > n$).
- The employee can only serve one customer at a time.
- The employee starts serving the next customer as soon as it finishes the current one. If no customer is available, he has to wait for one to arrive.

Goal: Maximize $L \in \mathbb{N}^+$.



Example

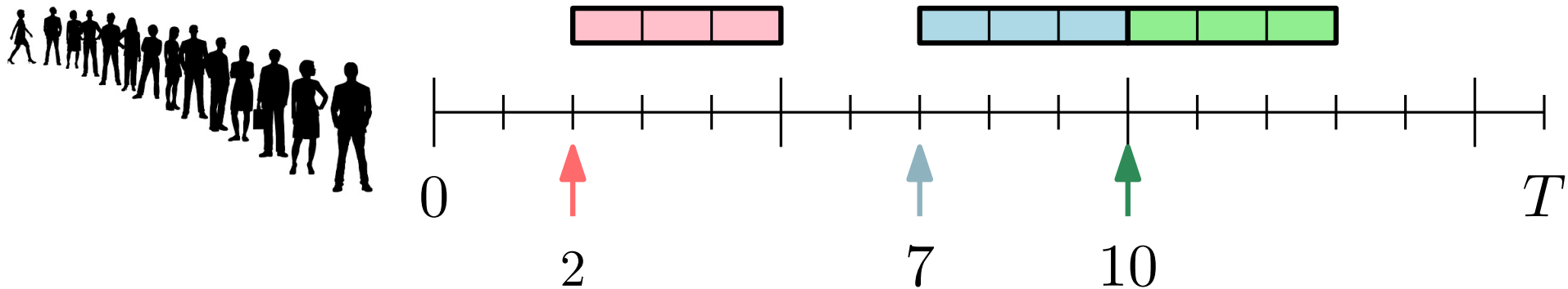
- 3 customers arrive at times $t_1 = 2$, $t_2 = 10$, $t_3 = 7$
- All customers must leave by time $T = 16$



Example

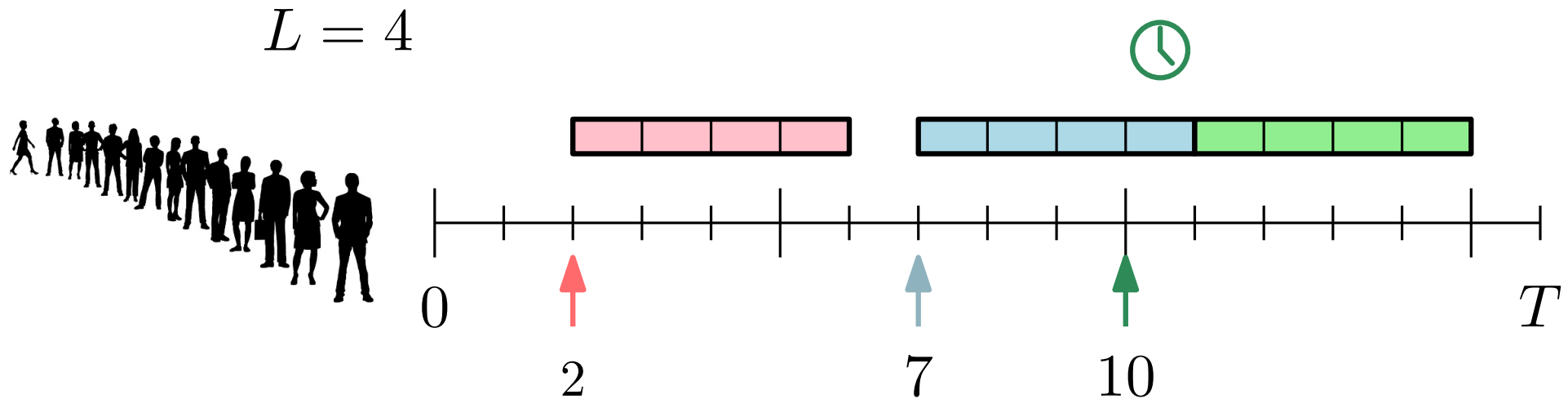
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$$L = 3$$



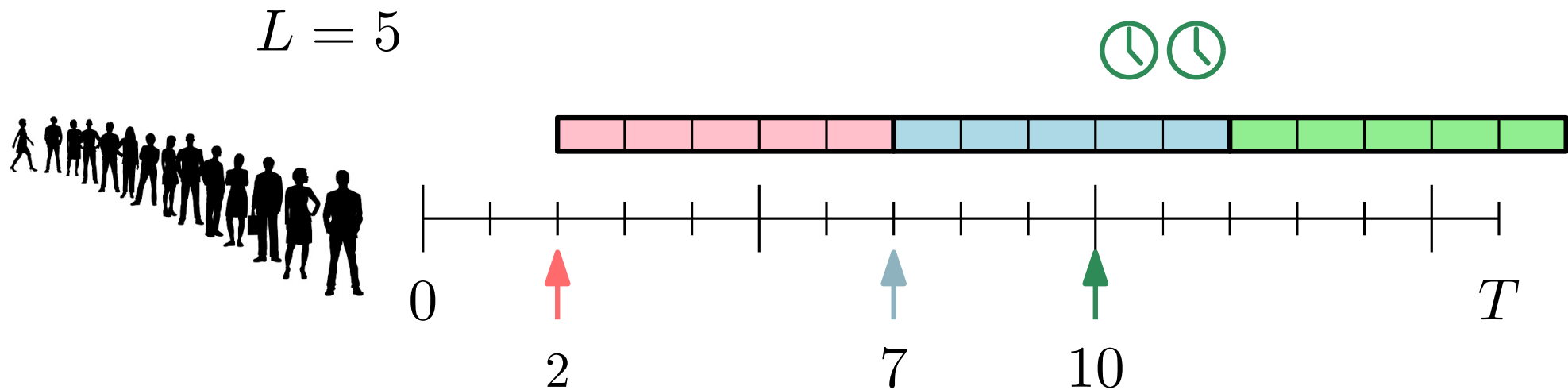
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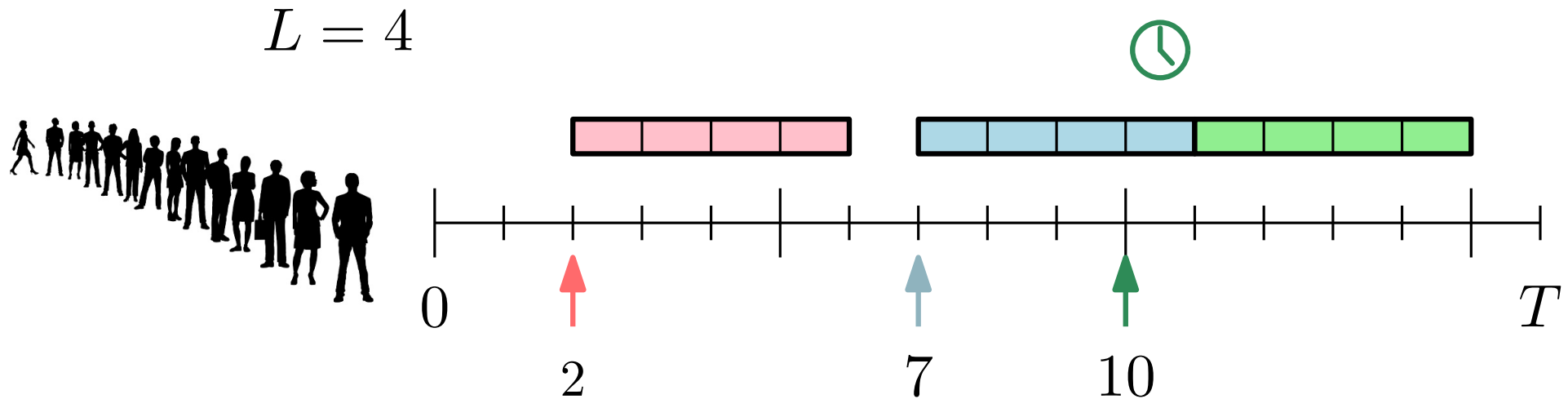
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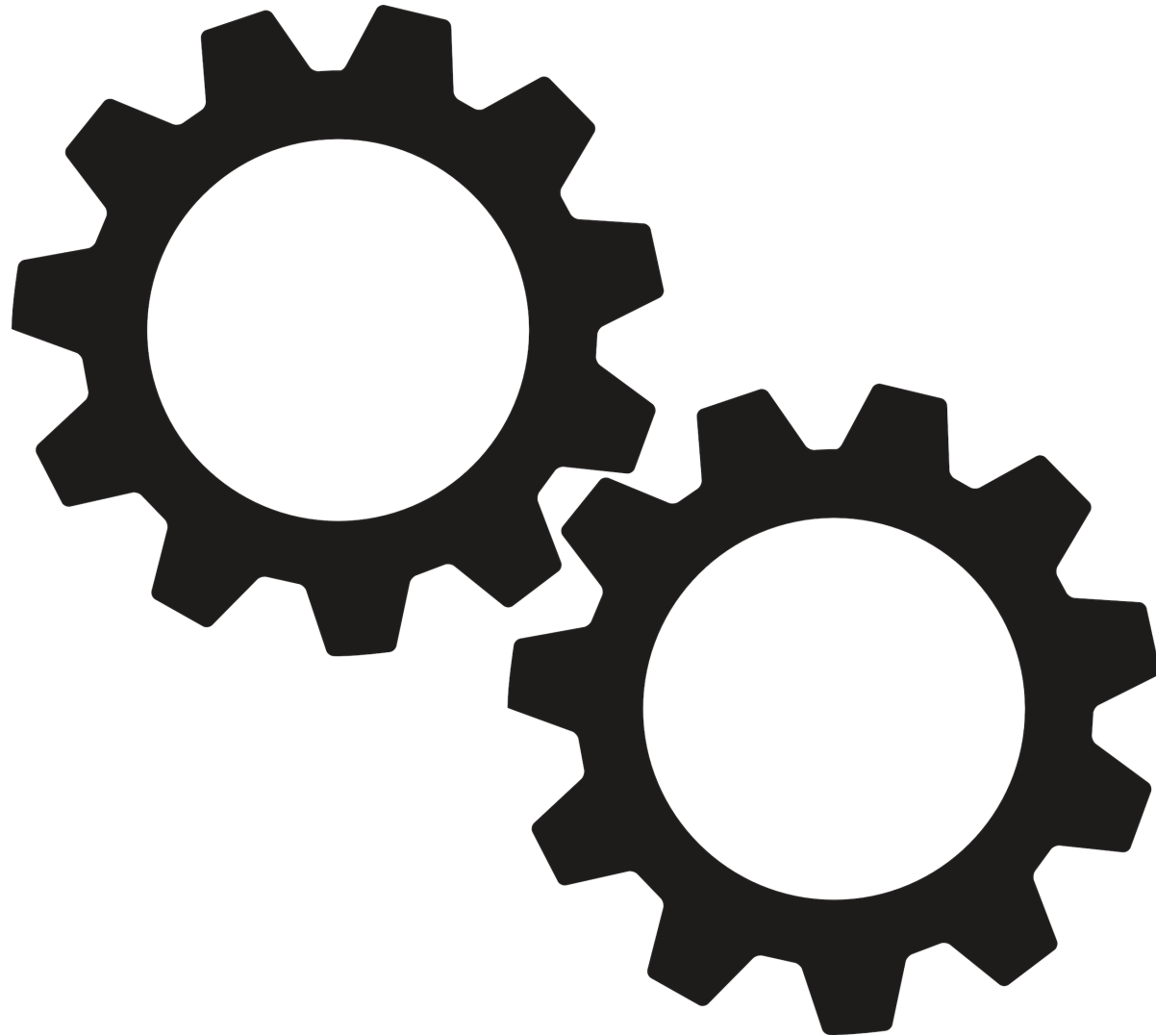


Example

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Solution: $L = 4$



A simple solution

- For each possible value of $L = 2, \dots, T$
 - // Check whether all clients can be served
 - While there are unserved clients
 - Find and serve next client
- Stop at the first value of L for which clients can't be served
- Return $L - 1$

A simple solution

- For each possible value of $L = 2, \dots, T$ $O(T)$
 - // Check whether all clients can be served
 - While there are unserved clients $O(n)$
 - Find and serve next client $O(n)$
- Stop at the first value of L for which clients can't be served
- Return $L - 1$


Time complexity: $O(T \cdot n^2)$

A simple solution

- For each possible value of $L = 2, \dots, \cancel{T} \lceil T/n \rceil$ $O(T/n)$
 - // Check whether all clients can be served
 - While there are unserved clients $O(n)$
 - Find and serve next client $O(n)$
- Stop at the first value of L for which clients can't be served
- Return $L - 1$

Time complexity: $O(T \cdot n)$

A simple solution

- **Preprocessing:** Sort t_1, \dots, t_n $O(n \log n)$
 - For each possible value of $L = 2, \dots, \cancel{T} \lceil T/n \rceil$ $O(T/n)$
 - // Check whether all clients can be served
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Time complexity: $O(T + n \log n)$

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Trick/Technique: Sorting

Sorting can be a powerful preprocessing step.

A Possible Implementation

```
std::sort(arrival_times.begin(), arrival_times.end());

int L;
for(L=2; L<=1+T/n; L++)
{
    int time = 1; //Next available time
    for(const int t : arrival_times)
        time = std::max(time, t) + L;

    if(time > T)
        break;
}

std::cout << L-1 << "\n";
```

Solutions so far

Naive algorithm

$$O(T \cdot n^2)$$

Naive algorithm (better analysis)

$$O(T \cdot n)$$

+ Preprocessing (sort arrival times)

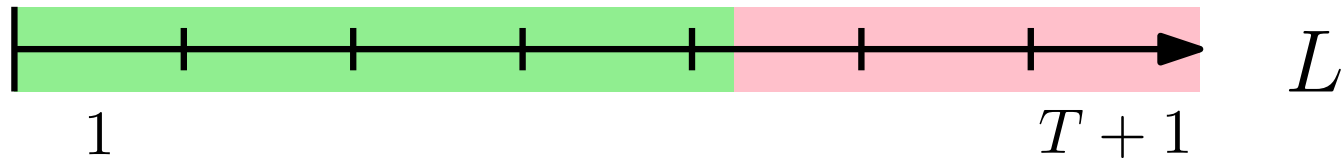
$$O(T + n \log n)$$

A Key Observation

Definition: We say that L is feasible if it allows to serve all customers by time T .

Observation: The property of being feasible is *monotone* w.r.t. L .

For $L > 1$, $\text{feasible}(L) \implies \text{feasible}(L - 1)$.
and $\neg \text{feasible}(L - 1) \implies \neg \text{feasible}(L)$.



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Proof sketch

Suppose w.l.o.g. that the t_i s are sorted. For a given L , let c_i^L the time after serving the i customers arriving at times t_1, \dots, t_i .

We prove by induction on $i = 1, \dots, n$, that $c_i^{L+1} > c_i^L$.

Base case ($i = 1$):

$$c_1^{L+1} = t_1 + L + 1 > t_1 + L = c_1^L.$$

Induction step ($i > 1$):

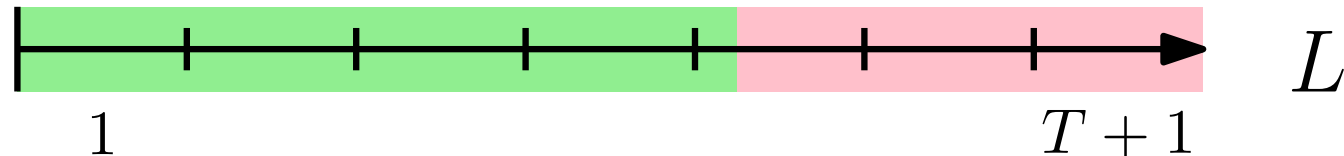
$$c_i^{L+1} = \max\{c_{i-1}^{L+1}, t_i\} + L + 1 > \max\{c_{i-1}^L, t_i\} + L = c_i^L. \quad \square$$

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For $L > 1$, $\text{feasible}(L) \implies \text{feasible}(L - 1)$.
and $\neg \text{feasible}(L - 1) \implies \neg \text{feasible}(L)$.



Trick/Technique: Binary Search

Use **binary search** to efficiently find the largest feasible value of a monotone property.

A Possible Implementation

```
//Returns the smallest index i in [min, max] such that
//p(i) is false.
//If no such index exists, returns max.
template<typename T>
    T binary_search(T min, T max, bool (*p)(T) )
{
    while(min < max)
    {
        if(T mid = (min+max)/2; p(mid))
            min = mid+1;
        else
            max = mid;
    }

    return max;
}
```

Time complexity: $O(\log(\max - \min))$

A Possible Implementation

```
bool feasible(int L)
{
    int time=1;
    for(const int t : arrival_times)
        time = std::max(time, t) + L;

    return time<=T;
}

std::sort(arrival_times.begin(), arrival_times.end());
std::cout << binary_search(2, T, feasible)-1 << "\n";
```

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bool feasible(int L)
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Time complexity: $O(n \log T)$

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$$O(T \cdot n^2)$$

Naive algorithm (better analysis)

$$O(T \cdot n)$$

+ Preprocessing (sort arrival times)

$$O(T + n \log n)$$

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+ Binary Search for L

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Naive algorithm $O(T \cdot n^2)$

Naive algorithm (better analysis) $O(T \cdot n)$

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+ Binary Search for L $O(n \log T)$

$O(n \log T)$ is never worse than $O(T + n \log n)$

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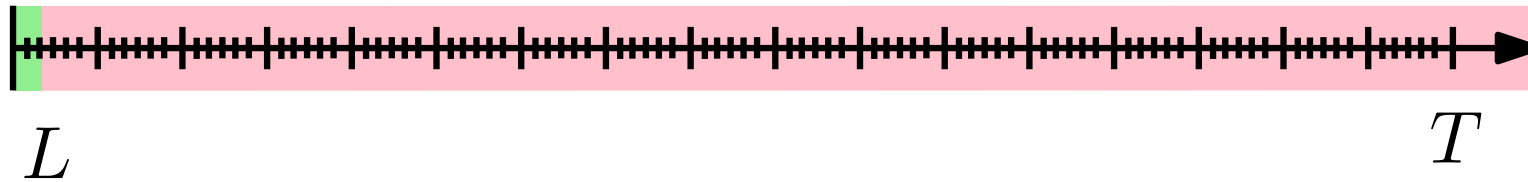
+ Binary Search for L $O(n \log T)$

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What if T is large (e.g., 2^n) but L is small?

Handling large values of L

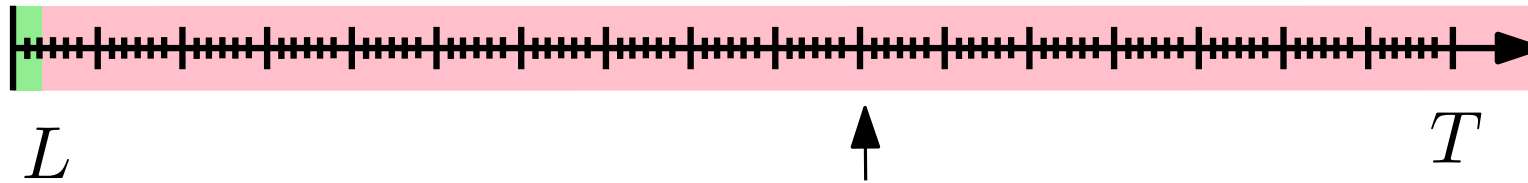
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(not to scale)

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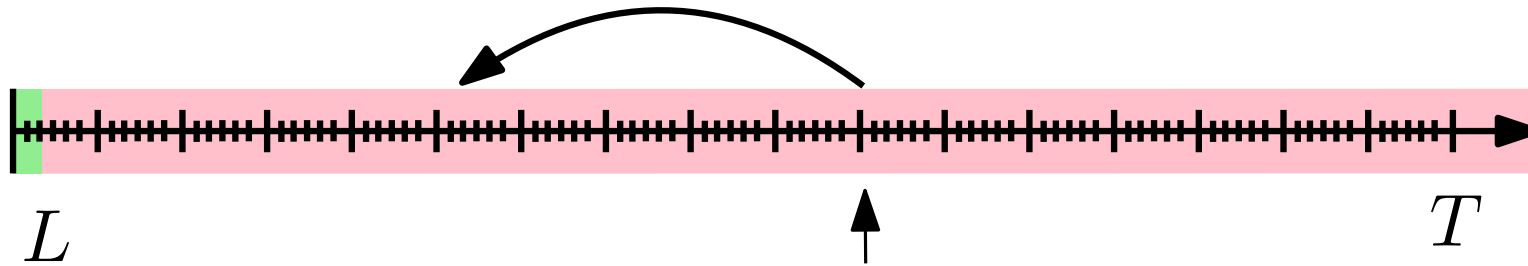
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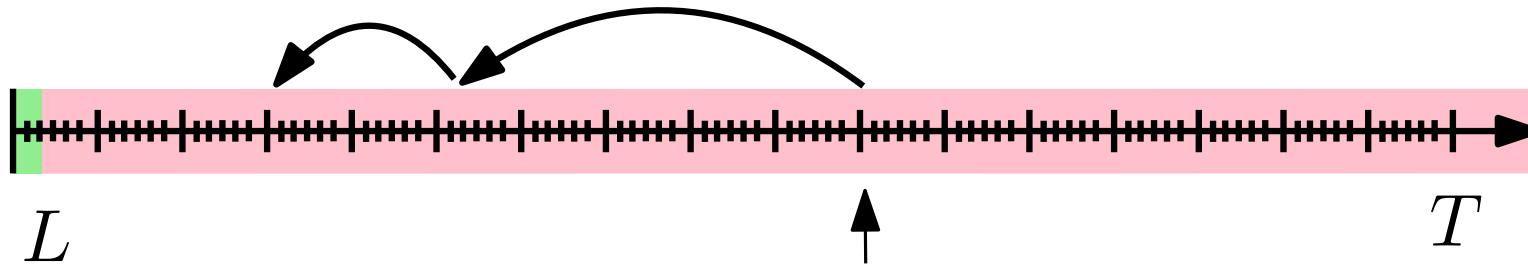
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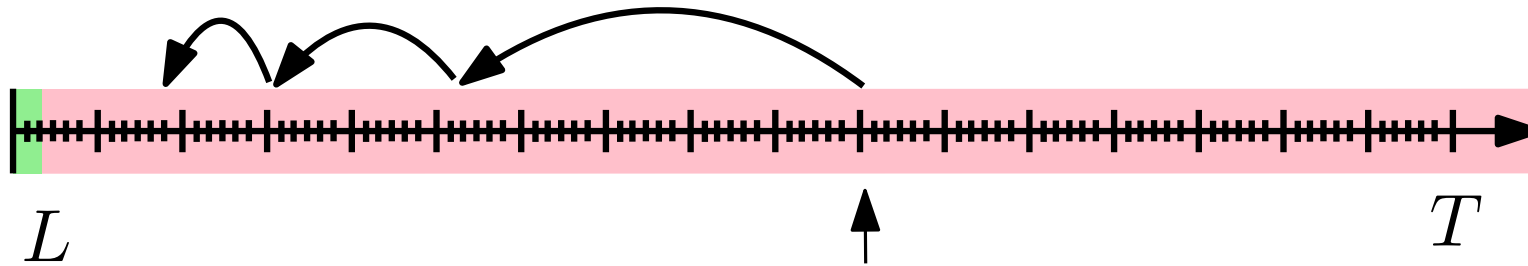
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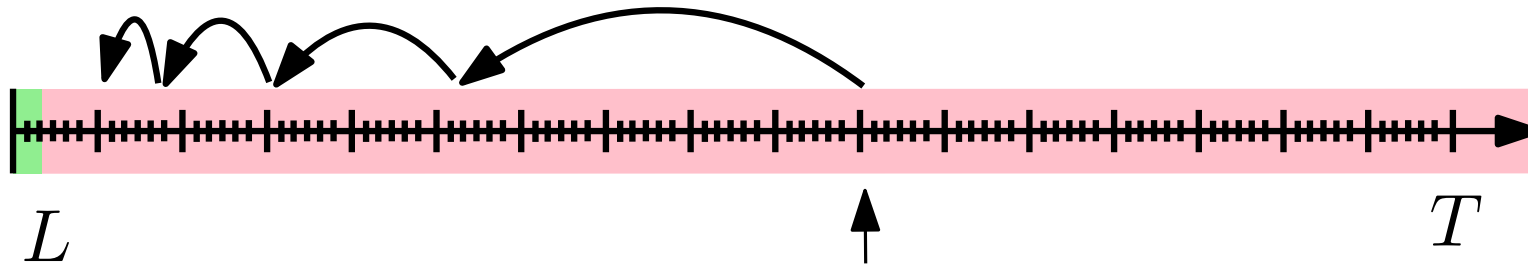
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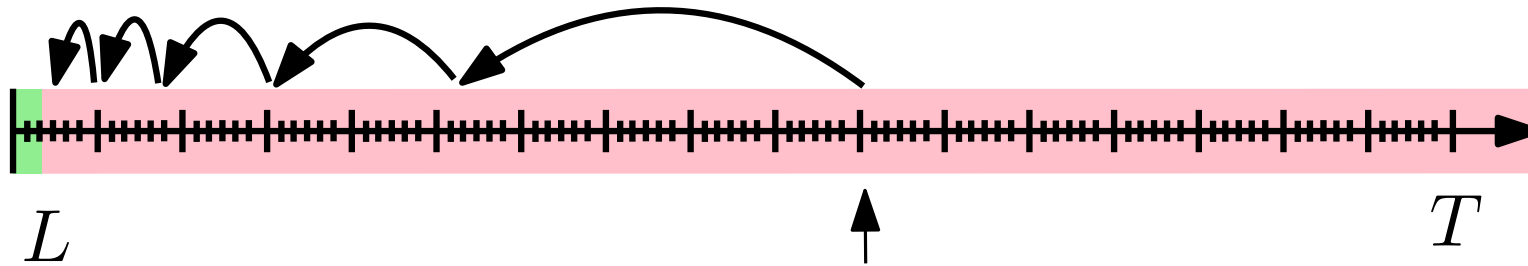
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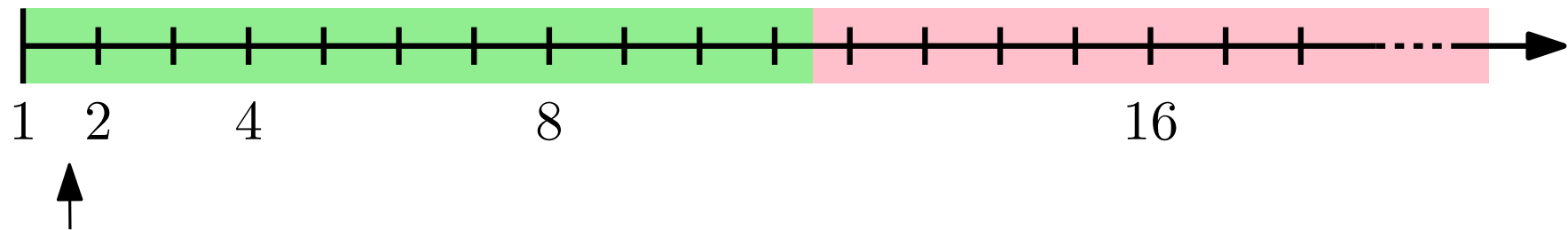


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Exponential Search

Idea: Check $U = 1, 2, 4, 8, \dots$, until $\text{feasible}(U)$ is false.

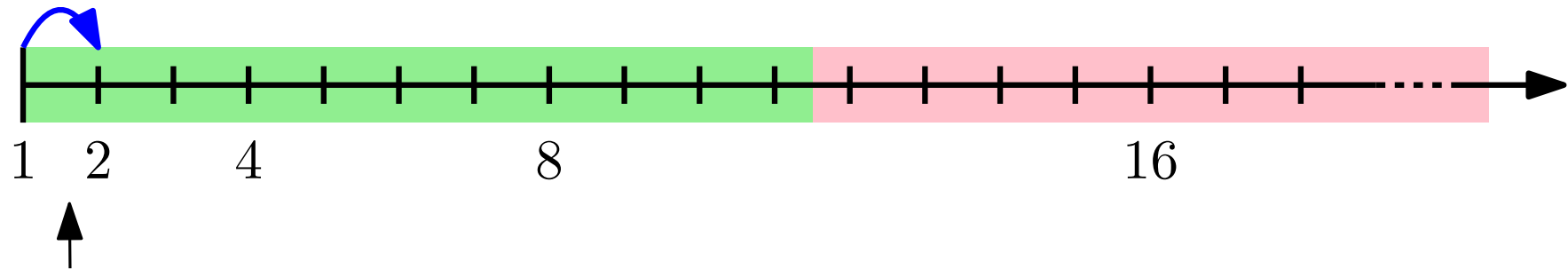
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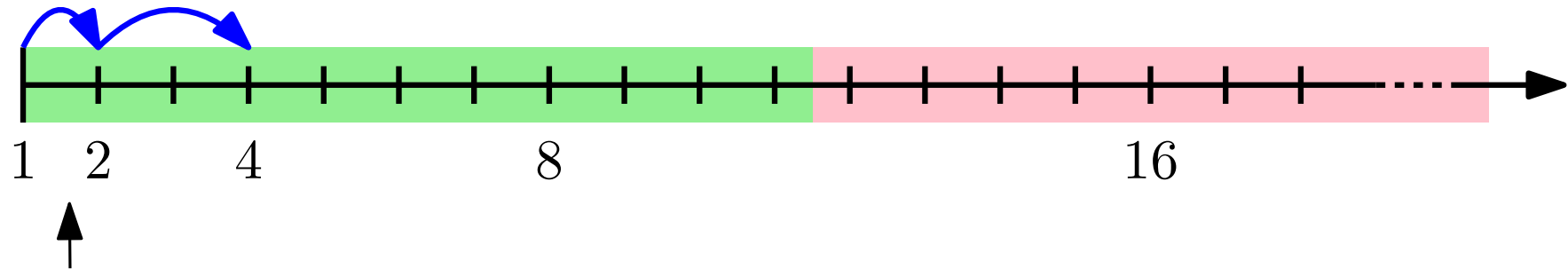
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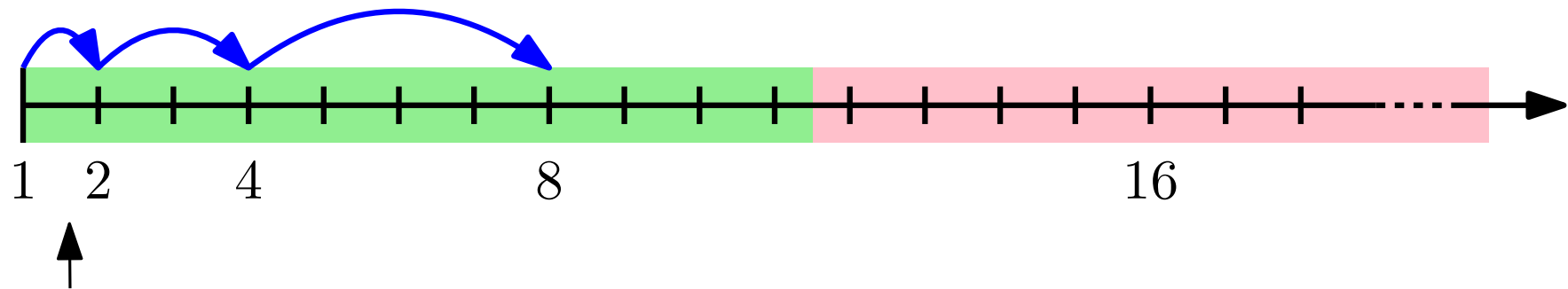
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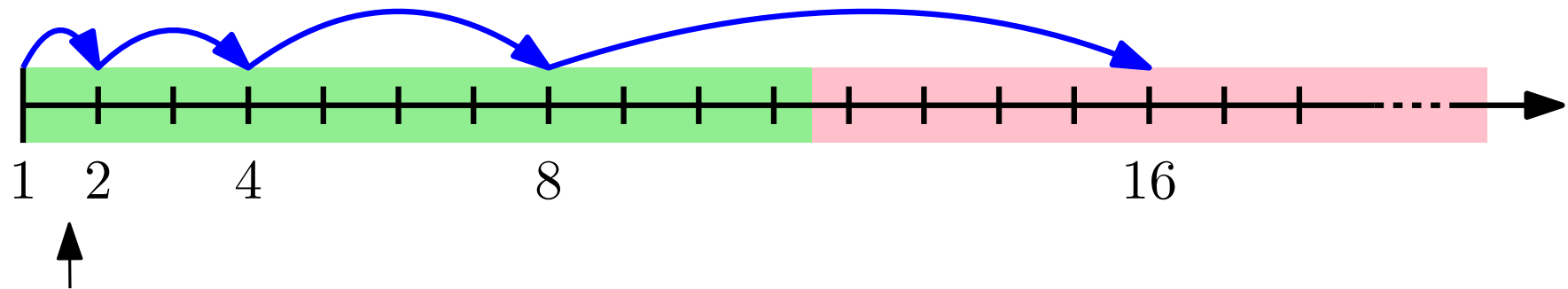
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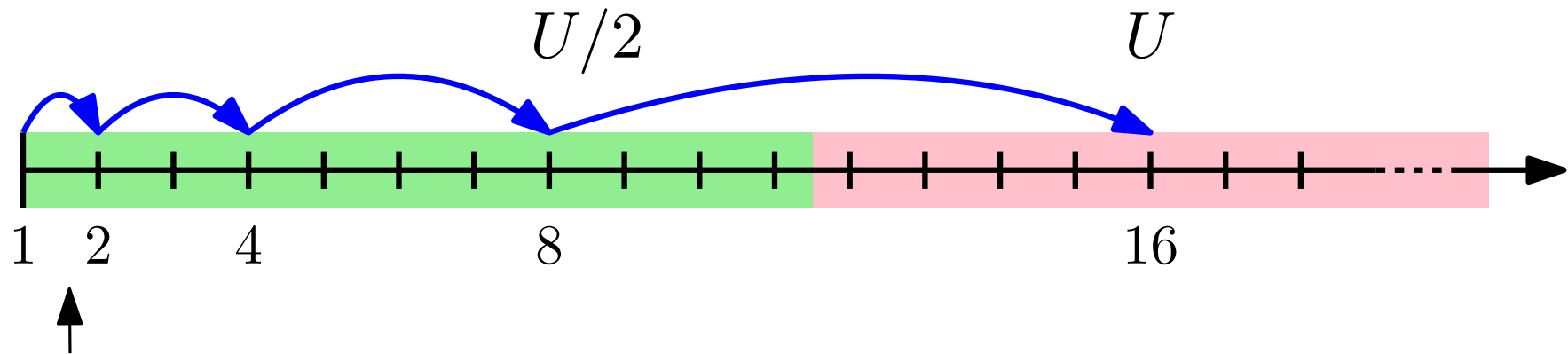
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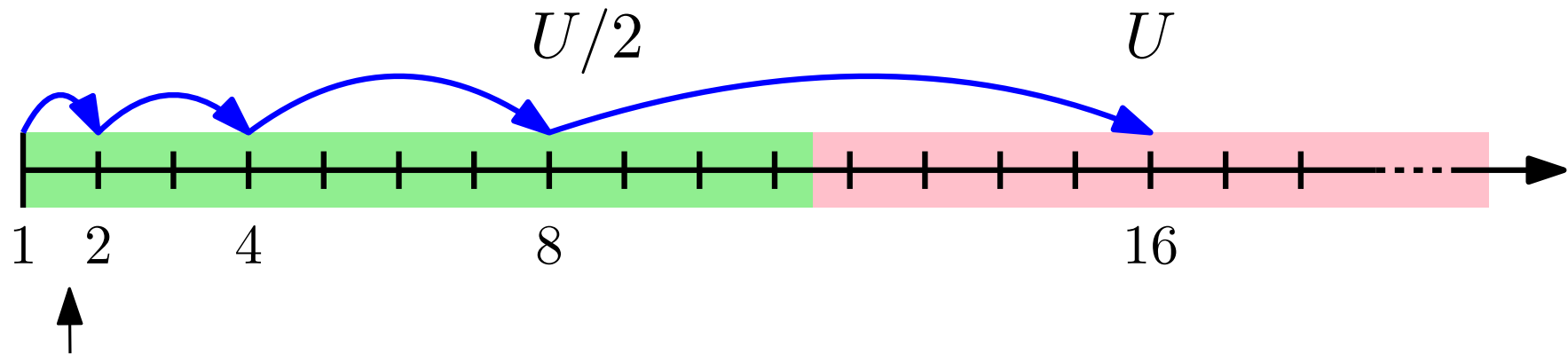
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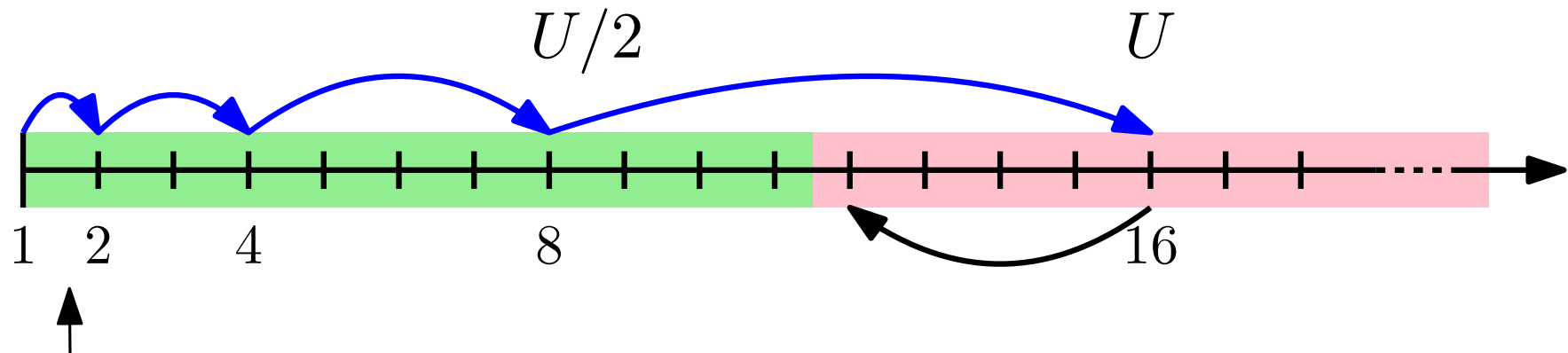
Trick/Technique: Exponential Search

Use **exponential search** to efficiently find an upper bound for applying **binary search**.

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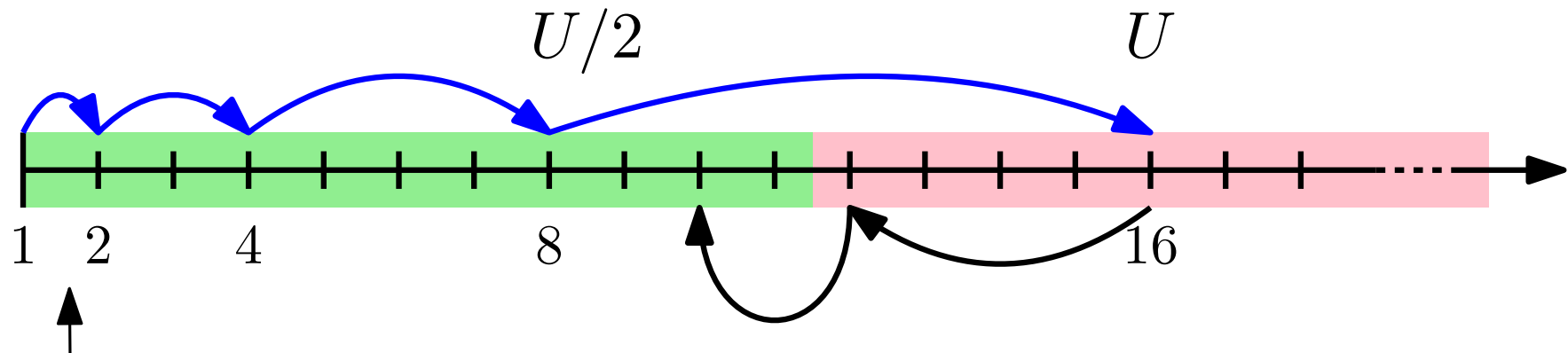
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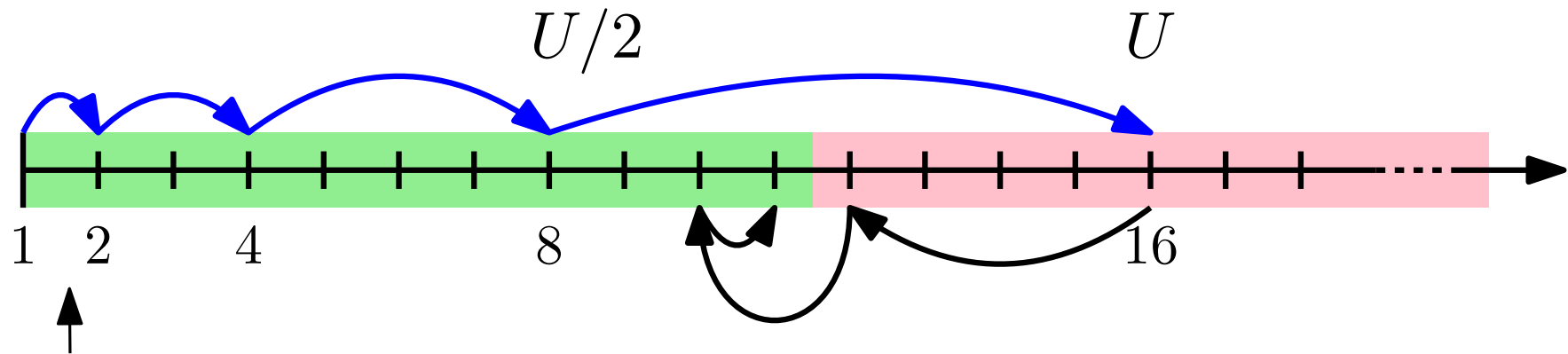
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Trick/Technique: Exponential Search

Use **exponential search** to efficiently find an upper bound for applying **binary search**.

A Possible Implementation

```
//If p(min) is false returns min
//Otherwise returns an index min+i in (min, max] such that
//p(min+i) is false and p(min+i/2) is true.
//If no such index exists, returns max.
template<typename T>
    T exponential_search(T min, T max, bool (*p)(T) )
{
    if(!p(min))
        return min;

    for(T i=1; min+i<max; i*=2)
        if(!p(min+i))
            return min+i;

    return max;
}
```

Time complexity: $O(\log(\max - \min))$

A Possible Implementation

```
std::sort(arrival_times.begin(), arrival_times.end());  
int U = exponential_search(2, T, feasible);  
std::cout << binary_search(U/2, U, feasible) - 1;
```

Time complexity: $O(n \log n + n \log L)$

Solutions so far

Naive algorithm

$$O(T \cdot n^2)$$

Naive algorithm (better analysis)

$$O(T \cdot n)$$

+ Preprocessing (sort arrival times)

$$O(T + n \log n)$$

+ Binary Search for L

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+ Exponential search $O(n \log n + n \log L)$

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