

Discrete Optimization

Local Search: Part VII

Goal of the Lecture

- ▶ Local search
 - more systematic presentation
 - beyond neighborhood
 - heuristics versus meta-heuristics
 - heuristics

Local Search

- ▶ **States**
 - either solutions or configurations

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 - $N(s)$: neighborhood of s

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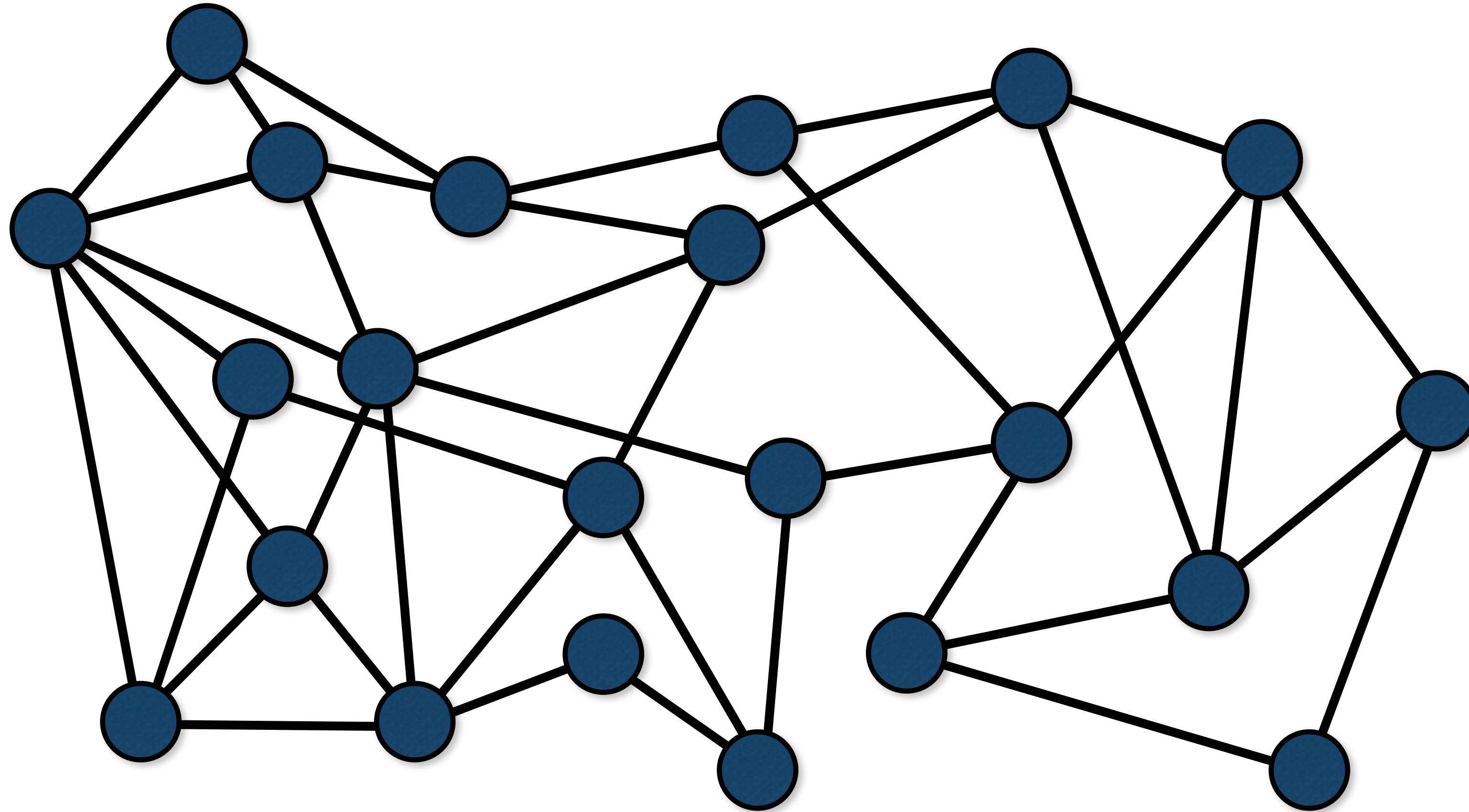
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 - $S(L(N(s),s),s)$; selection function
- ▶ Objective function
 - minimizing $f(s)$

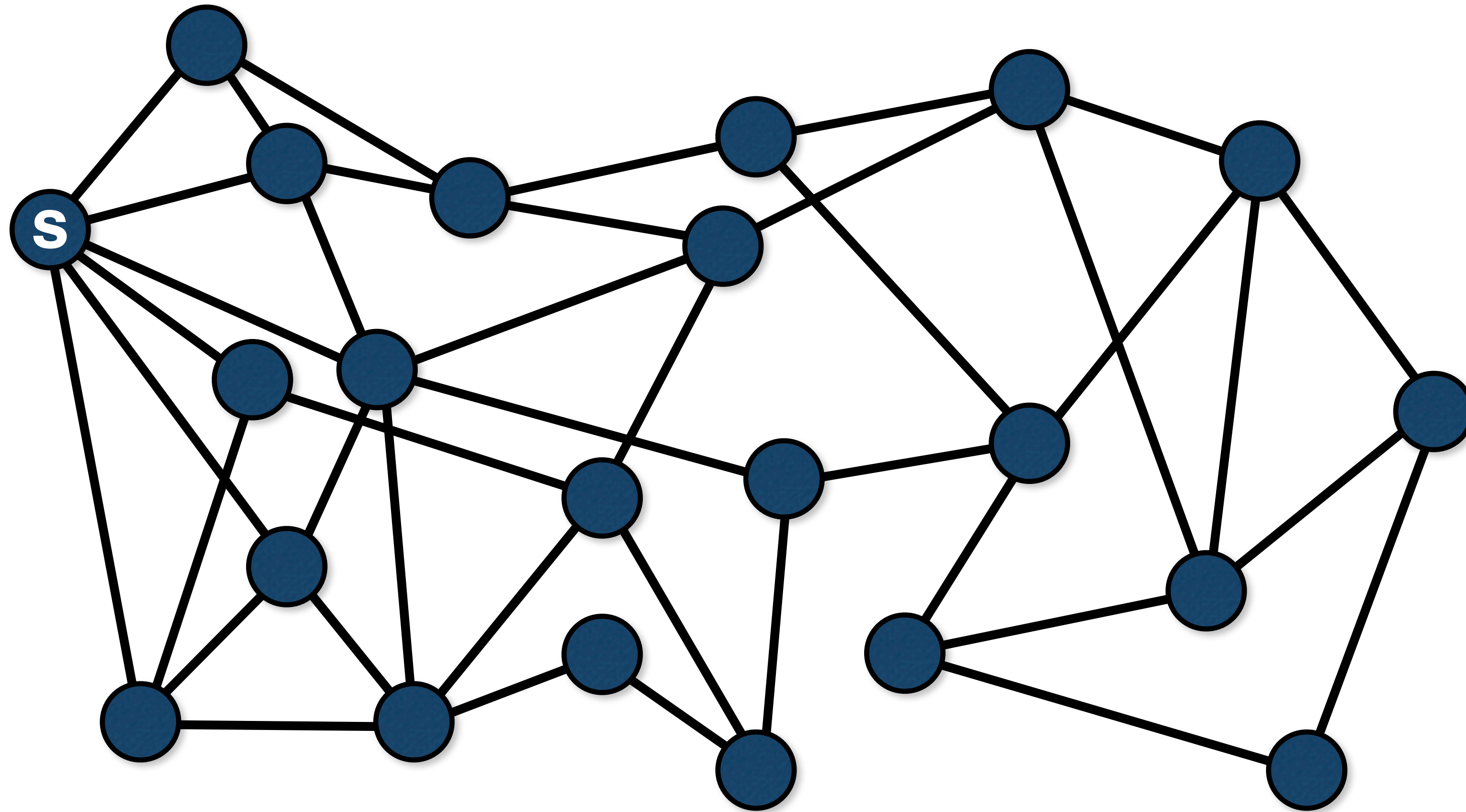
The Basic Local Search

```
1.  function LOCALSEARCH( $f, N, L, S$ ) {  
2.       $s := \text{GENERATEINITIALSOLUTION}()$ ;  
3.       $s^* := s$ ;  
4.      for  $k := 1$  to  $MaxTrials$  do  
5.          if  $satisfiable(s) \wedge f(s) < f(s^*)$  then  
6.               $s^* := s$ ;  
7.               $s := S(L(N(s), s), s)$ ;  
8.      return  $s^*$ ;  
9.  }
```

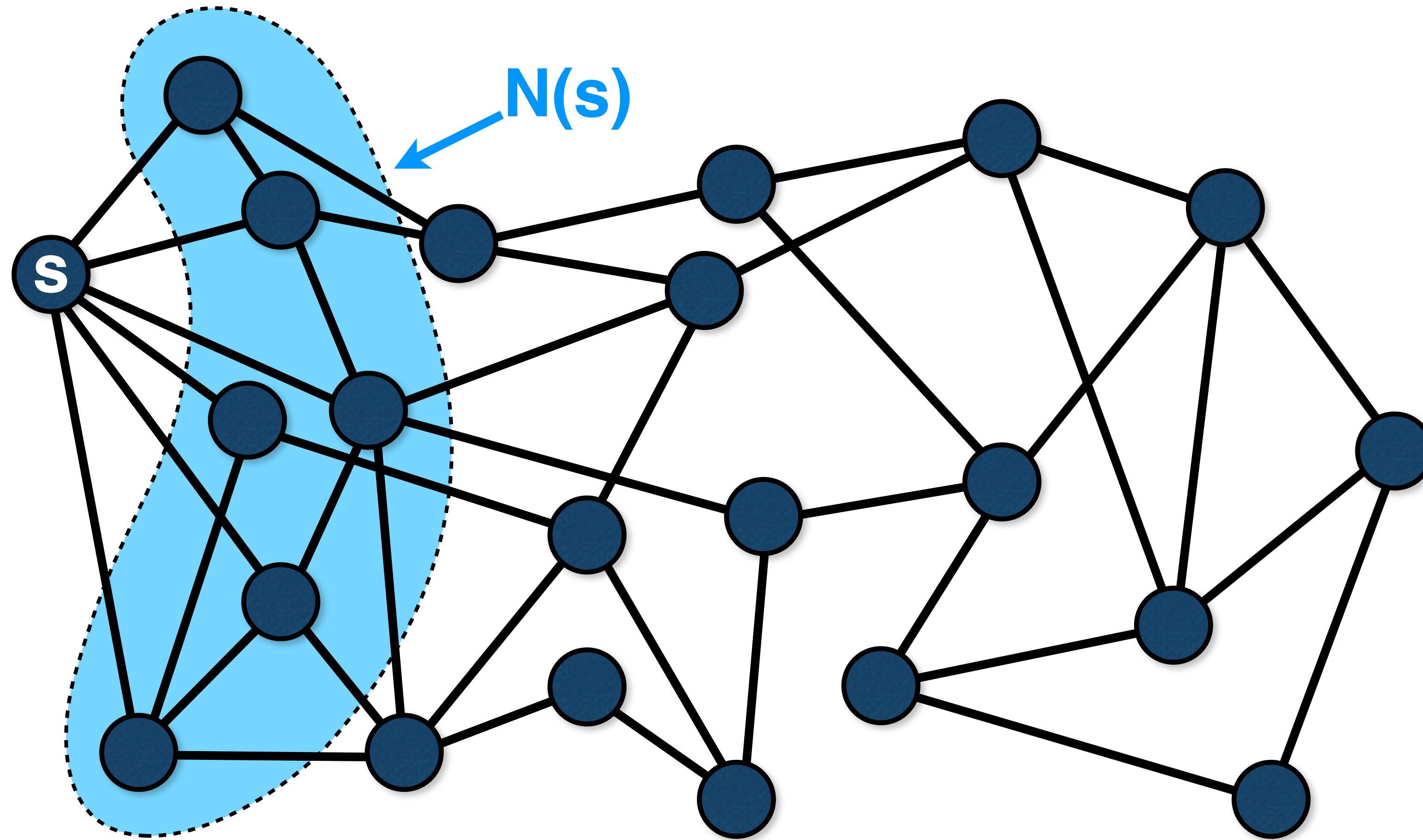

The Basic Local Search



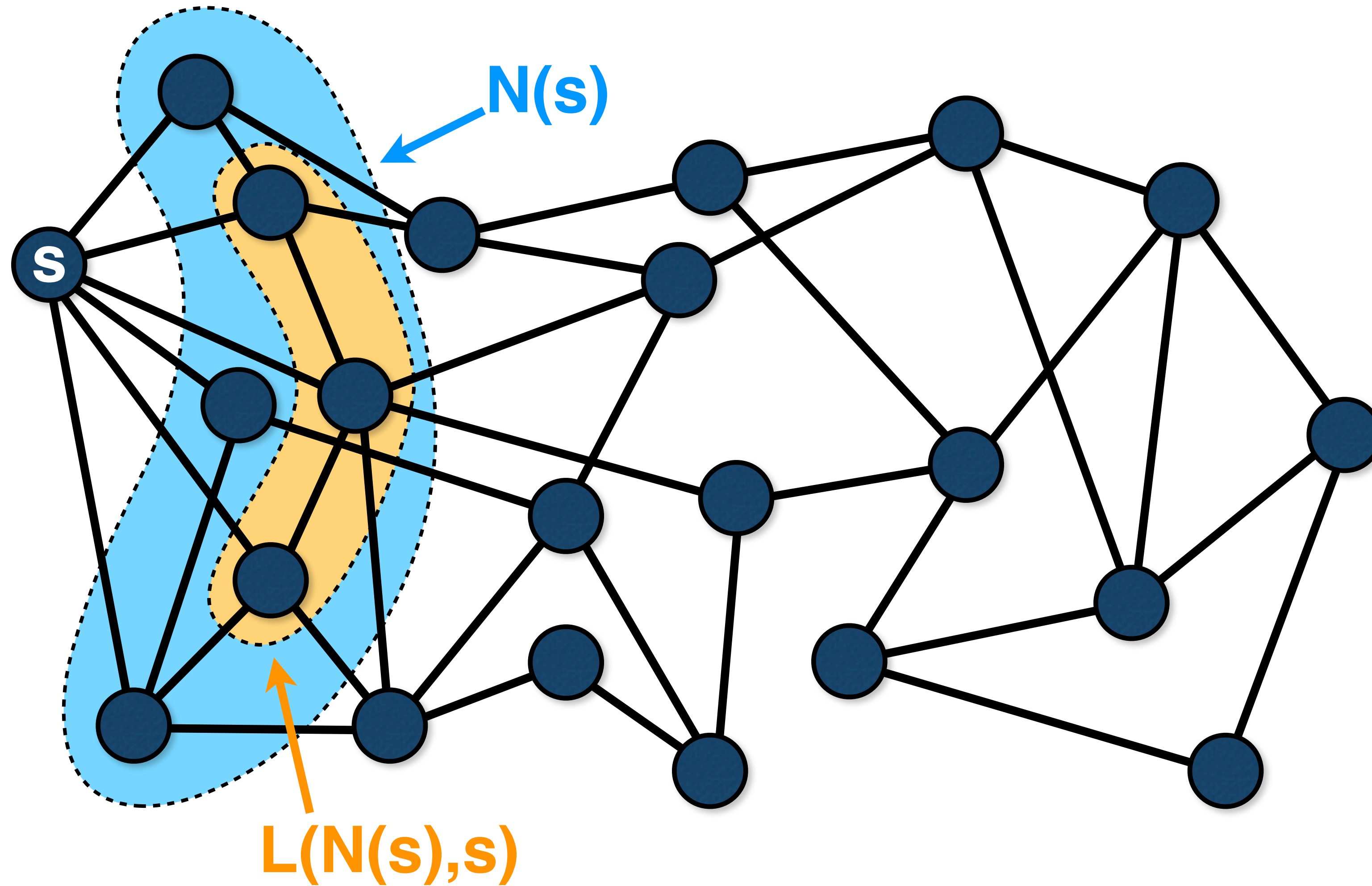
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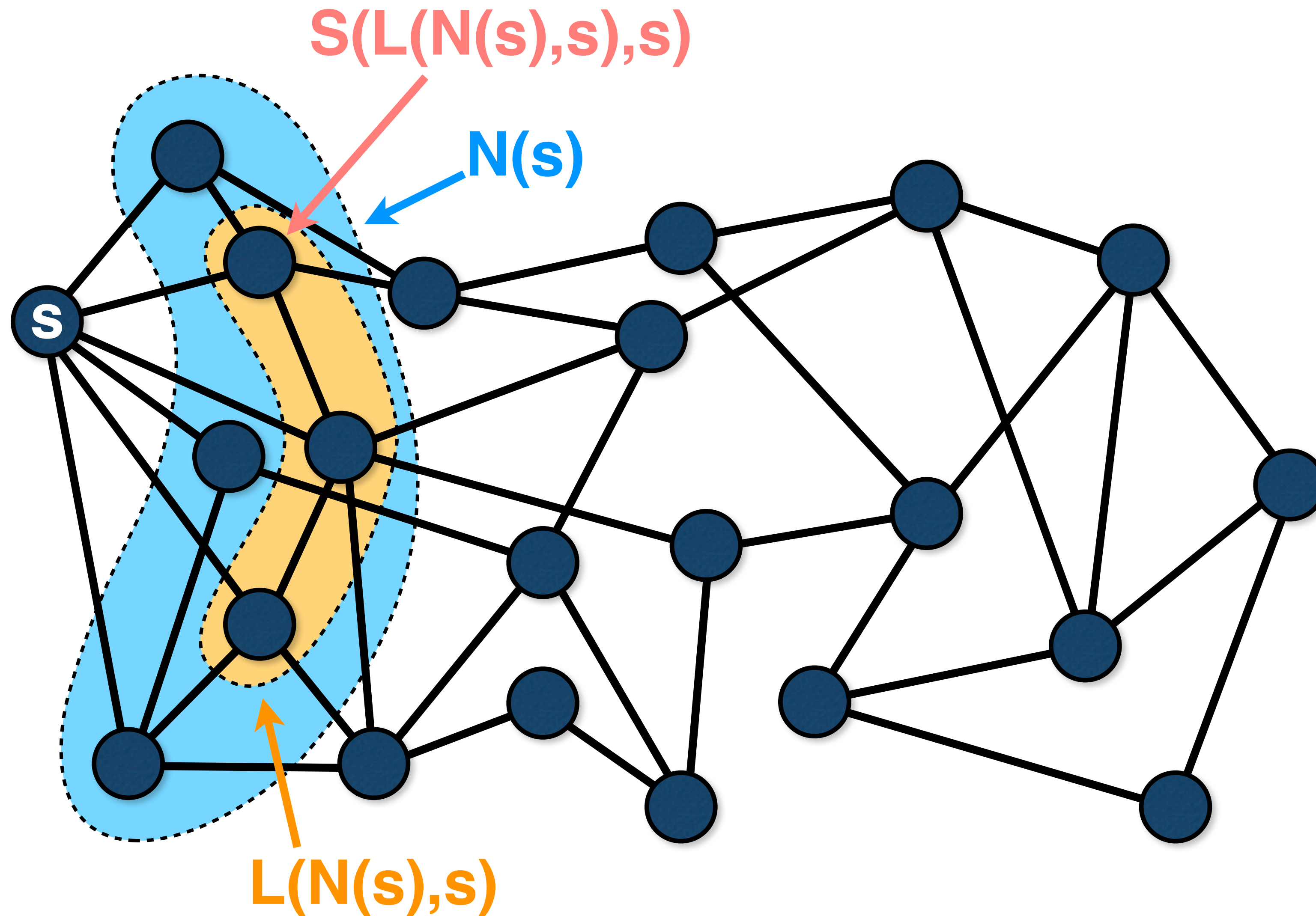
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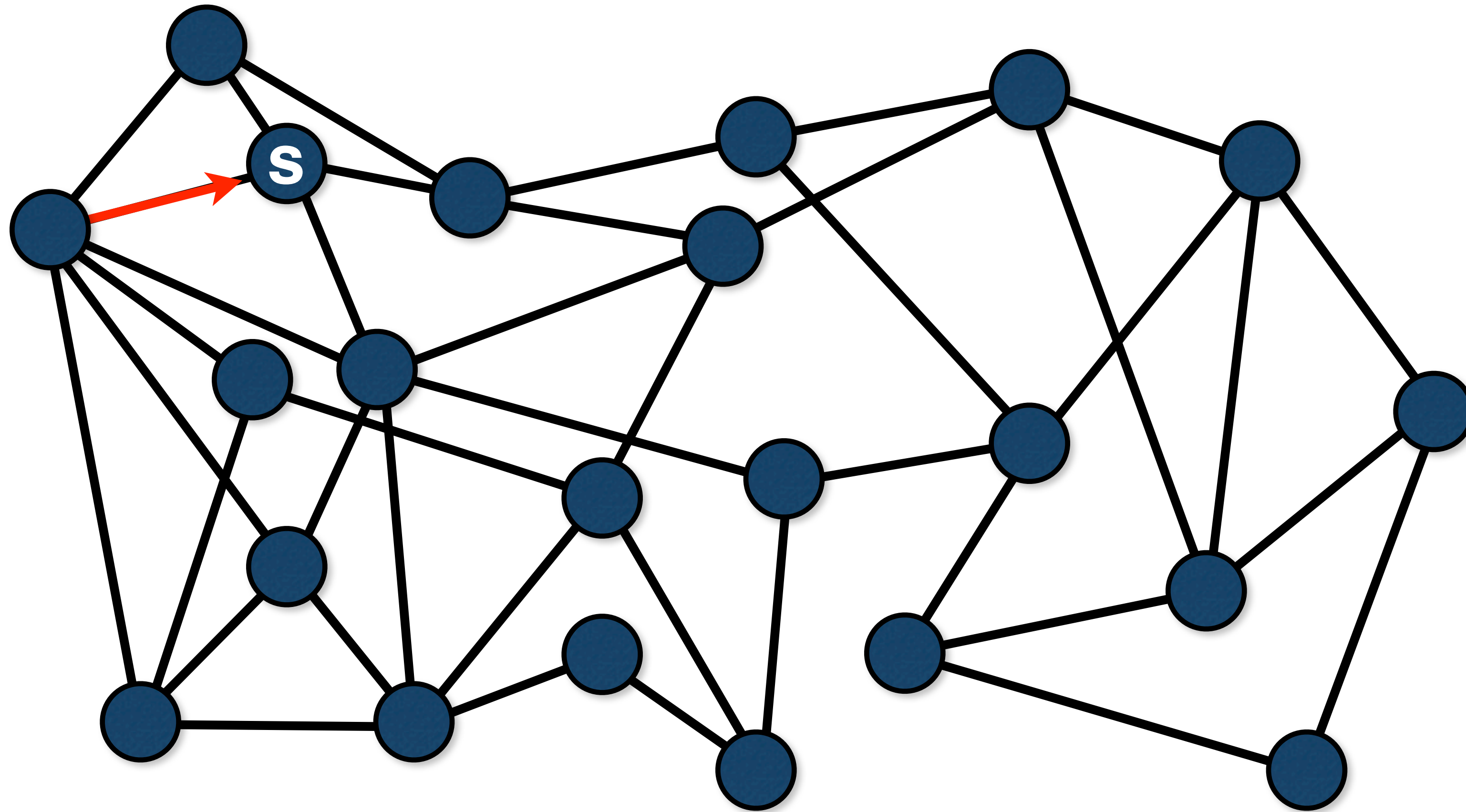
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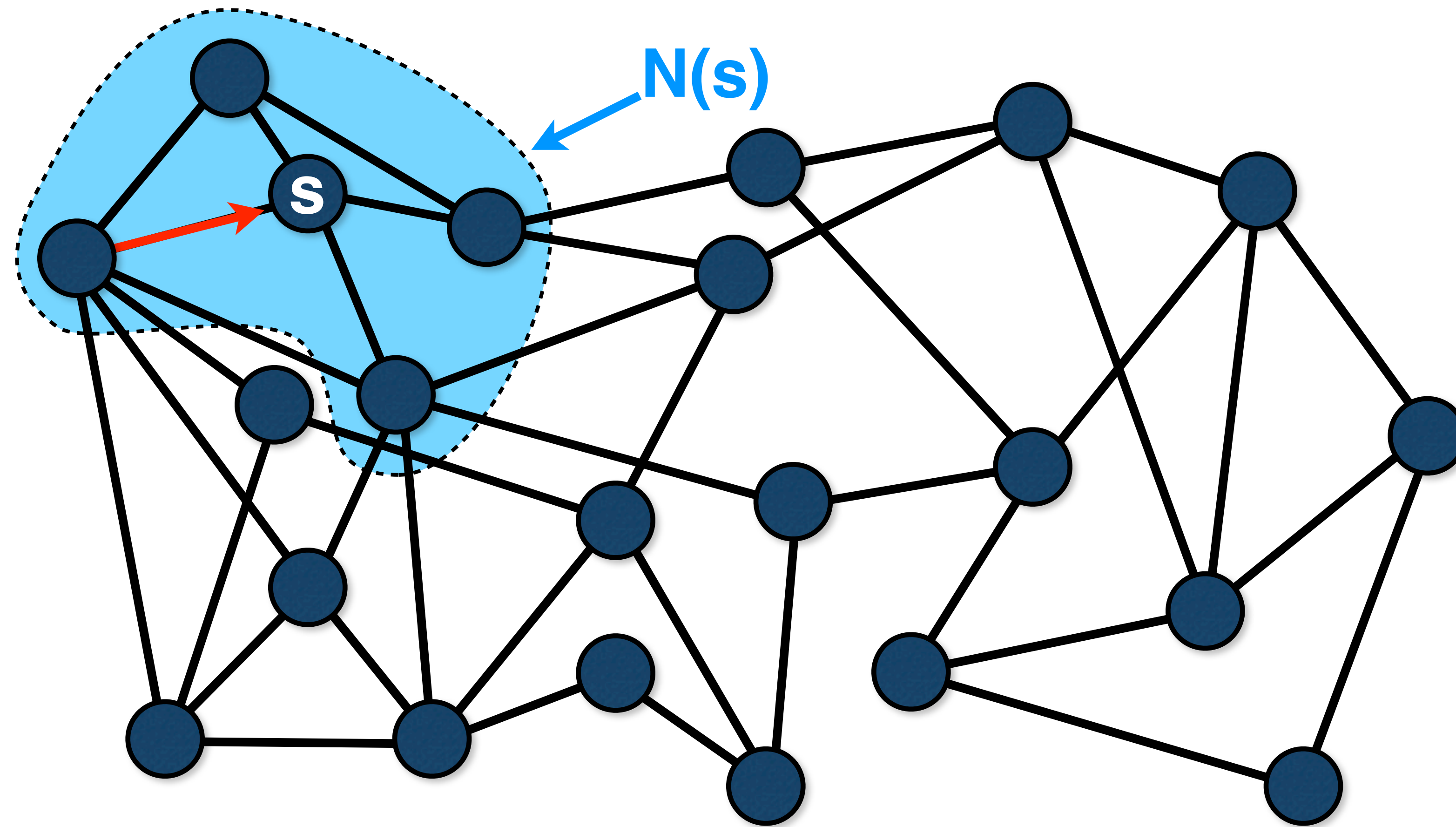
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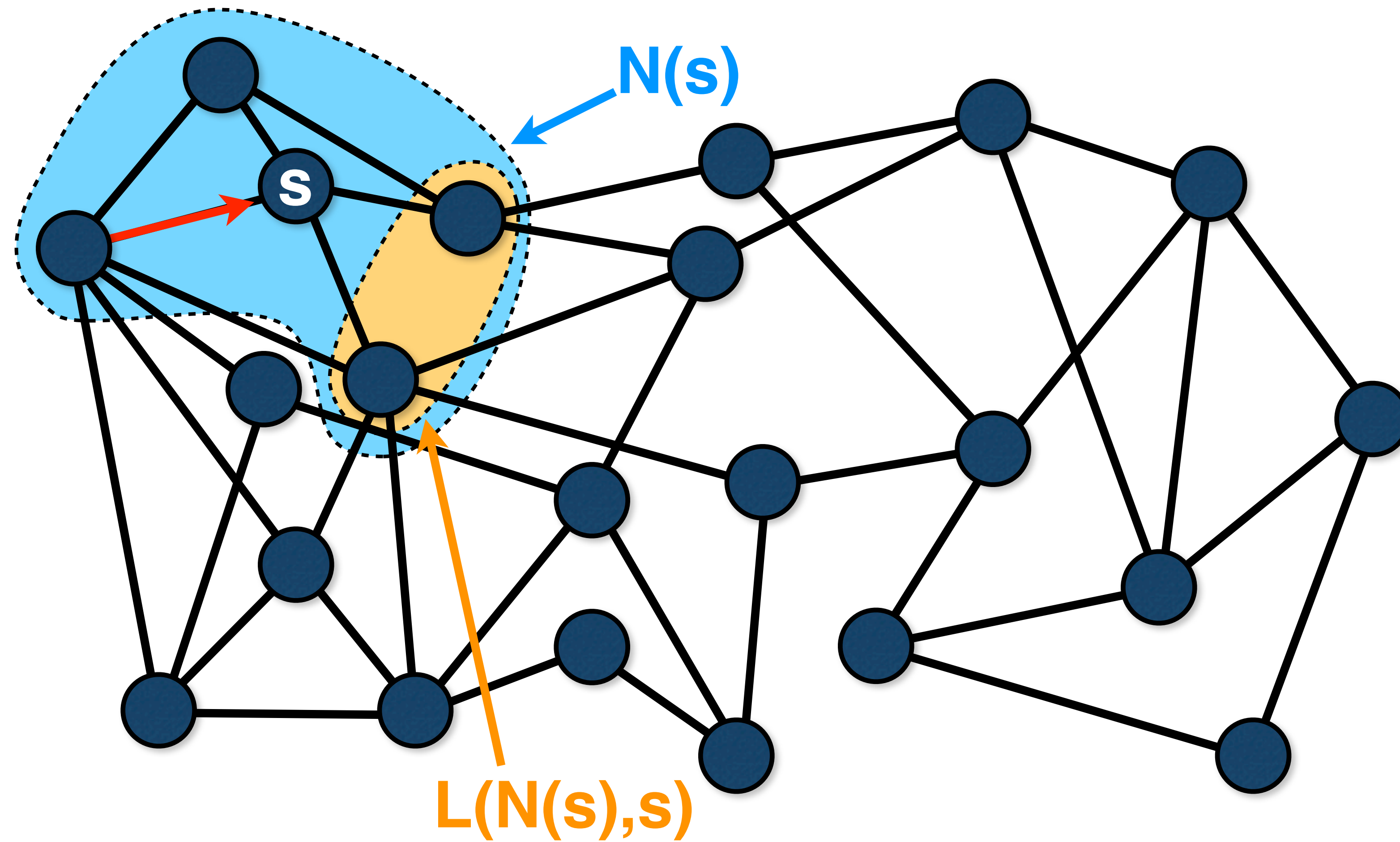
The Basic Local Search



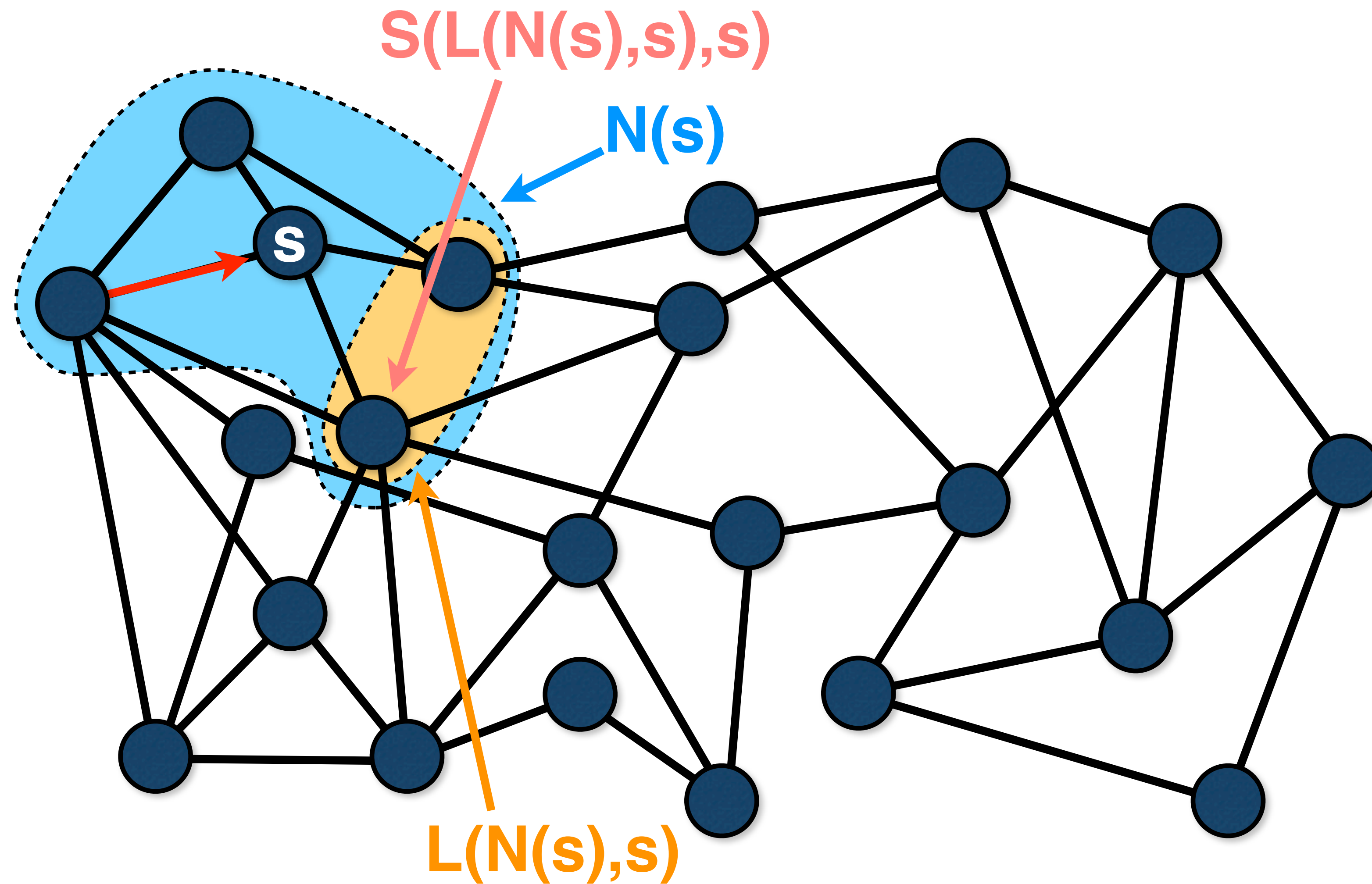
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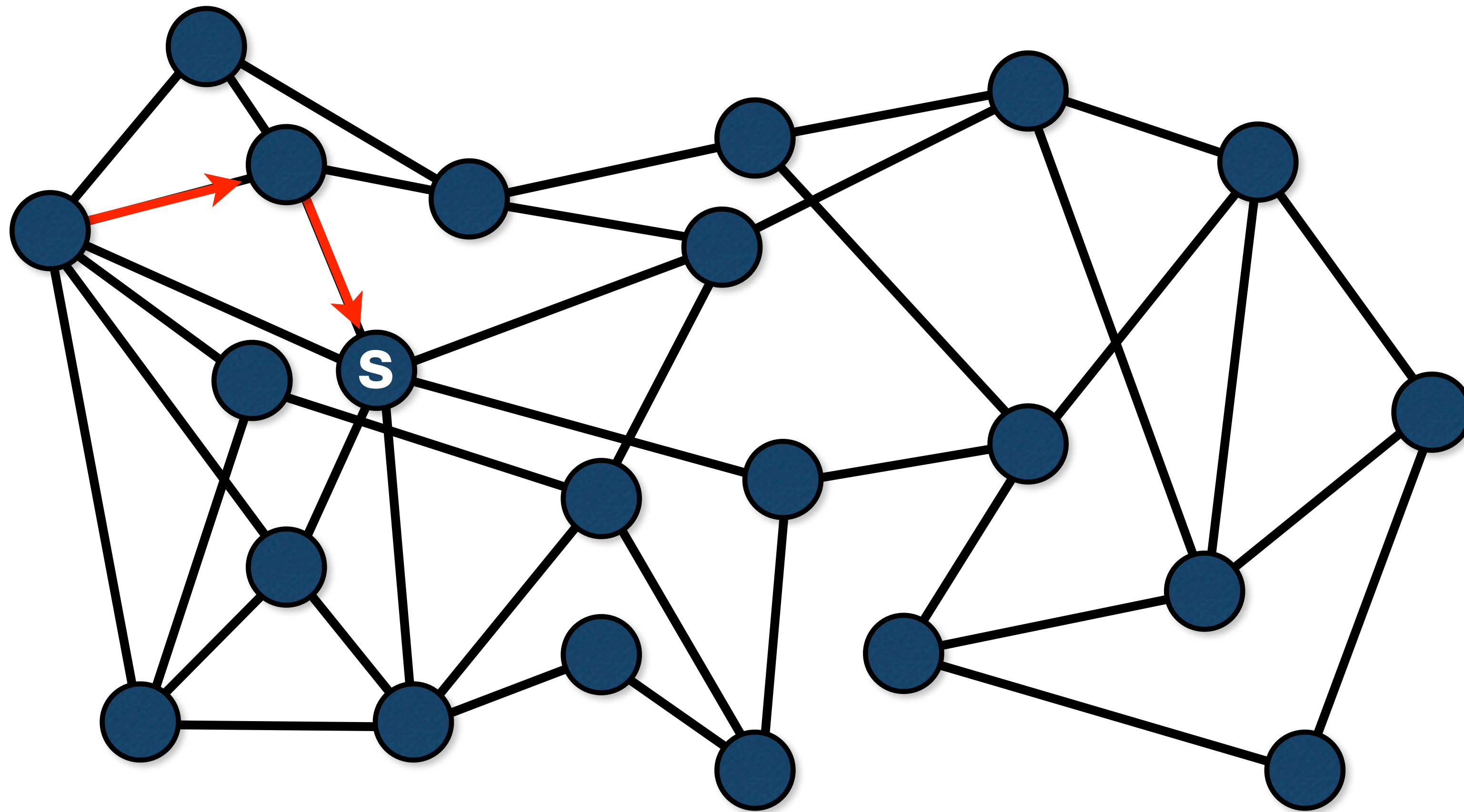
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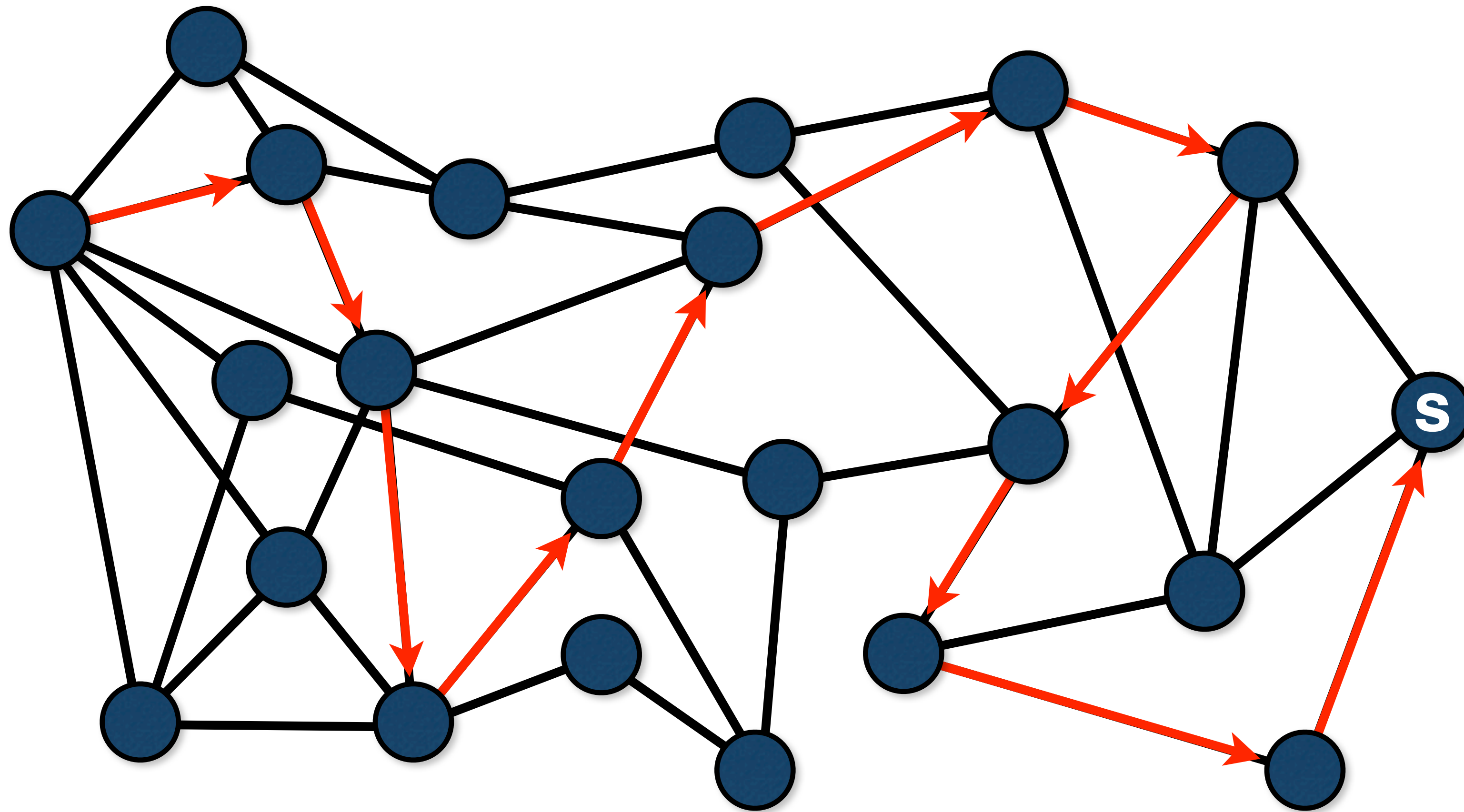
The Basic Local Search



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The Basic Local Search



A Example of Local Search

- ▶ Legal moves: local improvements
 - $L(N,s) = \{ n \text{ in } N \mid f(n) < f(s) \}$

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- ▶ Legal moves: local improvements
 - $L(N,s) = \{ n \text{ in } N \mid f(n) < f(s) \}$
- ▶ Selection function: greedy selection
 - $S(L,s) = \arg\text{-min}(n \text{ in } L) f(n)$

Heuristics and Metaheuristics

► Heuristics

- choose the next neighbor
- use local information:
 - the state s and its neighborhood
- drive the search towards a local minimum

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► Metaheuristics

- aim at escaping local minima
- drive the search towards a global minimum
- typically include some memory or learning

Properties of the Neighbors

- ▶ Legal neighbors
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 - $L(N,s) = \{ n \text{ in } N \mid f(n) < f(s) \}$
- ▶ No degradation
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 - $L(N,s) = \{ n \text{ in } N \mid f(n) < f(s) \}$
- ▶ No degradation
 - $L(N,s) = \{ n \text{ in } N \mid f(n) \leq f(s) \}$
- ▶ Potential degradation
 - $L(N,s) = N$

Selecting a Neighbor

- ▶ How to select the neighbor?
 - exploring the whole or part of the neighborhood

Selecting a Neighbor

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 - exploring the whole or part of the neighborhood
- ▶ Best neighbor
 - select “the” best neighbor in the neighborhood

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- ▶ First neighbor
 - select the first “legal” neighbor in the neighborhood
- ▶ Multi-stage selection
 - first select one “part” of neighbor
 - second select the remaining “part” of the neighbor

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 - more on this soon

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 1. **function** S-BEST(N, s)
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 3. **return** $n \in N^*$ with probability $1/\#N^*$;

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- ▶ Best improvement

1. **function** BESTIMPROVEMENT(s)
2. **return** LOCALSEARCH(f, N, L -IMPROVEMENT, S-BEST);

First Neighbor

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 - avoid scanning the entire neighborhood

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1. **function** S-FIRST(N, s)
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- ▶ First improvement

1. **function** FIRSTIMPROVEMENT(s)
2. **return** LOCALSEARCH(f, N, L -IMPROVEMENT, S-FIRST);

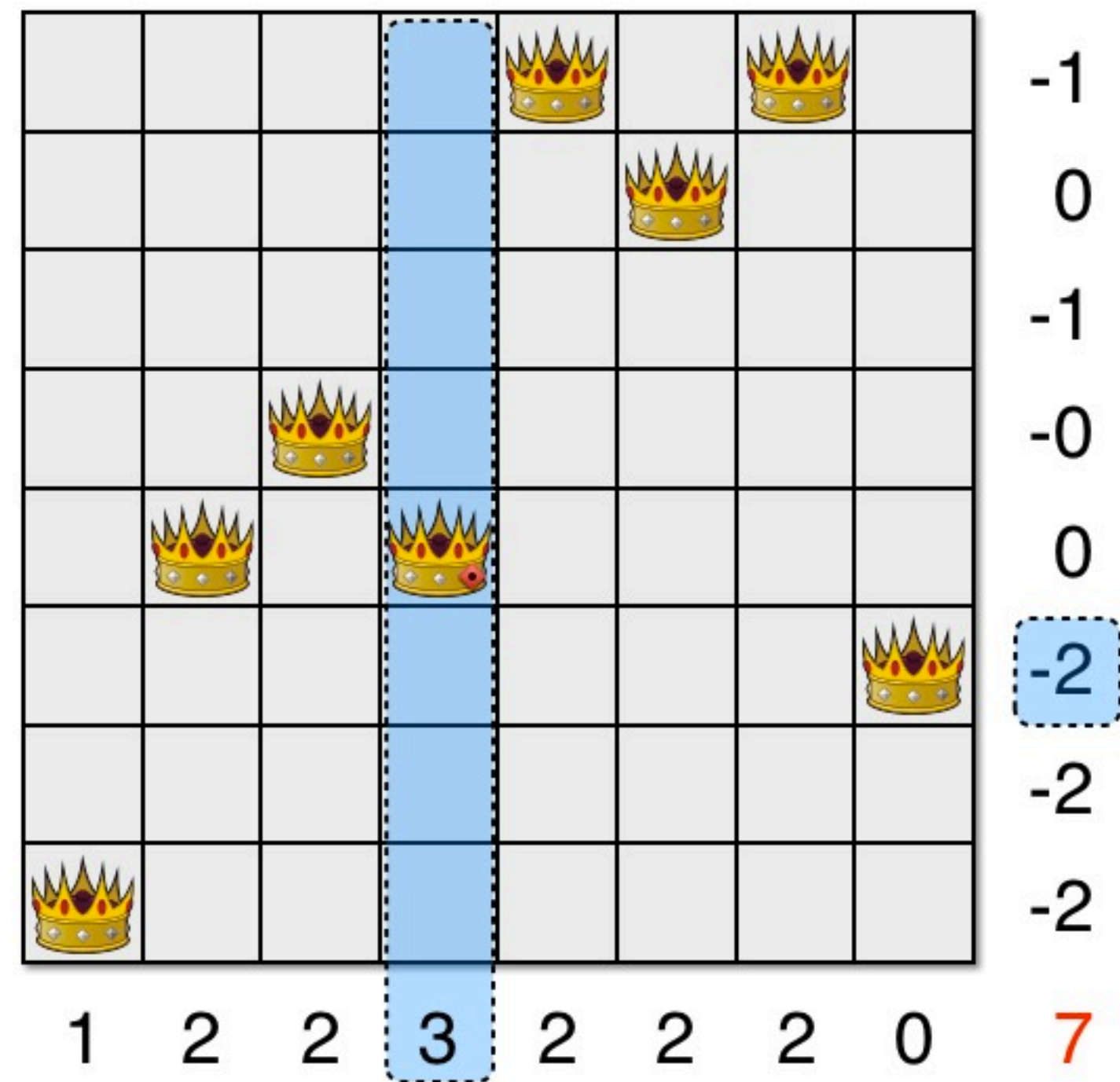
Multi-Stage Heuristics

- ▶ **Motivation**
 - avoid scanning the entire neighborhood
 - still keep a greedy flavor

Multi-Stage Heuristics

► Motivation

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- still keep a greedy flavor



Multi-Stage Heuristics

► Max/Min-Conflict

- select the variable with the most violations
 - first stage: greedy
- select the value with the fewest resulting violations
 - second stage: greedy

Multi-Stage Heuristics

► Max/Min-Conflict

- select the variable with the most violations
 - first stage: greedy
- select the value with the fewest resulting violations
 - second stage: greedy

► Min-conflict heuristic

- randomly select a variable with some violations
 - first-stage: randomized
- select the value with the fewest resulting violations
 - second stage: greedy

Multi-Stage Heuristics

- ▶ What was the alternative?
 - $N(s): \{ s[q \leftarrow v] \mid q \text{ in Queens \& } v \text{ in Rows} \}$
 - $s[q \leftarrow v]$ is the solution s where queens q is assigned to v ;

Multi-Stage Heuristics

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► Complexity

- quadratic: all pairs (c,r) where c is a column and r is a row
- $O(n^2)$ where n is the number of queens

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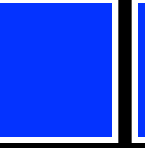

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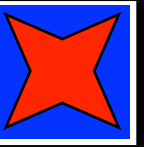
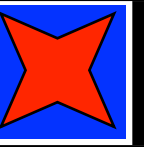
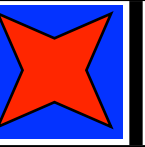
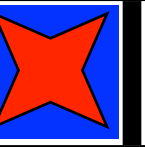
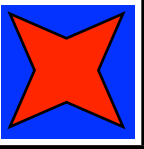
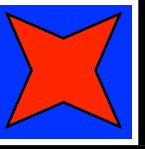
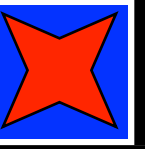
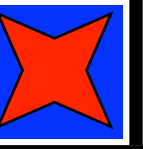



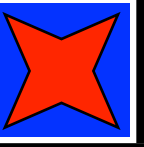



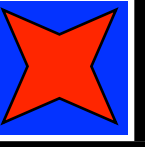
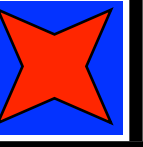
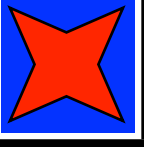
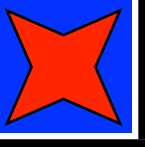
► Complexity of min-conflict

- $O(n)$ where n is the number of queens

Multi-Stage in Car Sequencing

Slots	1	2	3	4	5	6	7	8	9	10	Demand
Class 1											1
Class 2											1
Class 3											2
Class 4											2
Class 5											2
Class 6											2

Options	1	2	3	4	5	Demand
Class 1	yes		yes	yes		1
Class 2				yes		1
Class 3		yes			yes	2
Class 4		yes		yes		2
Class 5	yes		yes			2
Class 6	yes	yes				2
Capacity	1/2	2/3	1/3	2/5	1/5	

Setup	1	2	3	4	5	6	7	8	9	10	Capacity
Option 1											1/2
Option 2											2/3
Option 3											1/3
Option 4											2/5
Option 5											1/5

3
2
2
2
3

Multi-Stage in Car Sequencing

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Class 1											1
Class 2											1
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Class 5											2
Class 6											2

Options	1	2	3	4	5	Demand
Class 1	yes		yes	yes		1
Class 2				yes		1
Class 3		yes			yes	2
Class 4		yes		yes		2
Class 5	yes		yes			2
Class 6	yes	yes				2
Capacity	1/2	2/3	1/3	2/5	1/5	

Setup	1	2	3	4	5	6	7	8	9	10	Capacity
Option 1											1/2
Option 2											2/3
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Option 4											2/5
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3
2
2
2
3

Multi-Stage Heuristics

- ▶ Quadratic neighborhood
 - consider all possible swaps
 - quadratic in the size of the assembly line

Multi-Stage Heuristics

- ▶ Quadratic neighborhood
 - consider all possible swaps
 - quadratic in the size of the assembly line
- ▶ Multi-stage neighborhood
 - select a slot s whose car induces some violations
 - swap slot s with all other slots
 - linear in the size of the assembly line

Random Walks

- ▶ Randomization
 - select a neighbor at random

Random Walks

- ▶ Randomization
 - select a neighbor at random
- ▶ Decide whether to accept it
 - random improvement
 - Metropolis algorithm

Random Walks

- ▶ Randomization
 - select a neighbor at random
- ▶ Random improvement
- ▶ Random improvement search

Random Walks

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- select a neighbor at random

- ▶ Random improvement

1. **function** S-RANDOMIMPROVEMENT(N, s)
2. **select** $n \in N$ with probability $1/\#N$;
3. **if** $f(n) < f(s)$ **then**
4. **return** n ;
5. **else**
6. **return** s ;

- ▶ Random improvement search

Random Walks

► Randomization

– select a neighbor at random

► Random improvement

```
1.  function S-RANDOMIMPROVEMENT( $N, s$ )
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3.      if  $f(n) < f(s)$  then
4.          return  $n$ ;
5.      else
6.          return  $s$ ;
```

► Random improvement search

```
1.  function RANDOMIMPROVEMENT( $s$ )
2.      return LOCALSEARCH( $f, N, L\text{-ALL}, S\text{-RANDOMIMPROVEMENT}$ );
```

The Traveling Tournament

T-R	1	2	3	4	5	6	7	8	9	10
1	6	@2	4	3	@5	@4	@3	5	2	@6
2	5	1	@3	@6	4	3	6	@4	@1	@5
3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	@2	1	5	2	@6	@3
5	@2	@3	6	4	1	@6	@4	@1	3	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

The Traveling Tournament

T-R	1	2	3	4	5	6	7	8	9	10
1	6	@2	4	3	@5	@4	@3	5	2	@6
2	5	1	@3	@6	4	3	6	@4	@1	@5
3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	@2	1	5	2	@6	@3
5	@2	@3	6	4	1	@6	@4	@1	3	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

$$d_{12} + d_{21} + d_{15} + d_{54} + d_{43} + d_{31} + d_{16} + d_{61}$$

+ ... +

$$d_{61} + d_{14} + d_{45} + d_{56} + d_{63} + d_{36} + d_{62} + d_{26}$$

The Neighborhood

- ▶ A number of moves
 - swap homes
 - swap rounds
 - swap teams
 - partial swap rounds
 - partial swap teams

Swap Teams

T-R	1	2	3	4	5	6	7	8	9	10
1	6	@2	4	3	@5	@4	@3	5	2	@6
2	5	1	@3	@6	4	3	6	@4	@1	@5
3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	@2	1	5	2	@6	@3
5	@2	@3	6	4	1	@6	@4	@1	3	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

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2	5	@3	6	4	1	@6	@4	@1	3	@5
3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	2	1	5	@2	@6	@3
5	@2	1	@3	@6	4	3	6	@4	@1	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

Swap Teams

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4	3	6	@1	@5	2	1	5	@2	@6	@3
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Swap Teams

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3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	2	1	5	@2	@6	@3
5	@2	1	@3	@6	4	3	6	@4	@1	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

Swap Teams

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4	3	6	@1	@5	2	1	5	@2	@6	@3
5	@2	1	@3	@6	4	3	6	@4	@1	2
6	@1	@4	@5	2	@3	5	@2	3	4	1

Swap Teams

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3	@4	5	2	@1	6	@2	1	@6	@5	4
4	3	6	@1	@5	2	1	5	@2	@6	@3
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3	@4	2	5	@1	6	@5	1	@6	@2	4
4	3	6	@1	@2	@5	1	2	5	@6	@3
5	@2	1	@3	@6	4	3	6	@4	@1	2
6	@1	@4	@2	5	@3	2	@5	3	4	1

Until Next Time