

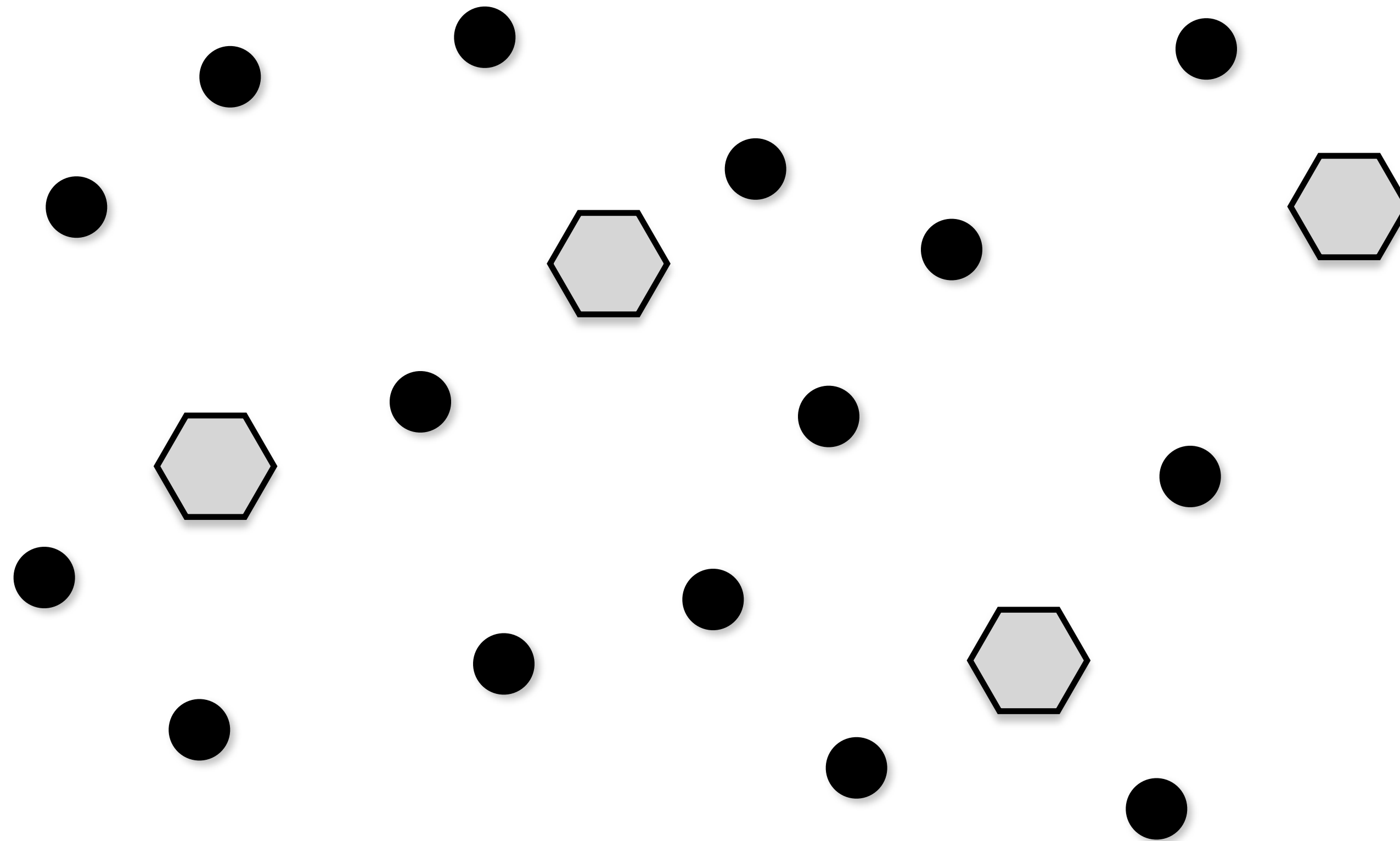
Discrete Optimization

Local Search: Part III

Goals of the Lecture

- ▶ Local search
 - optimization
 - warehouse location
 - traveling salesman problem

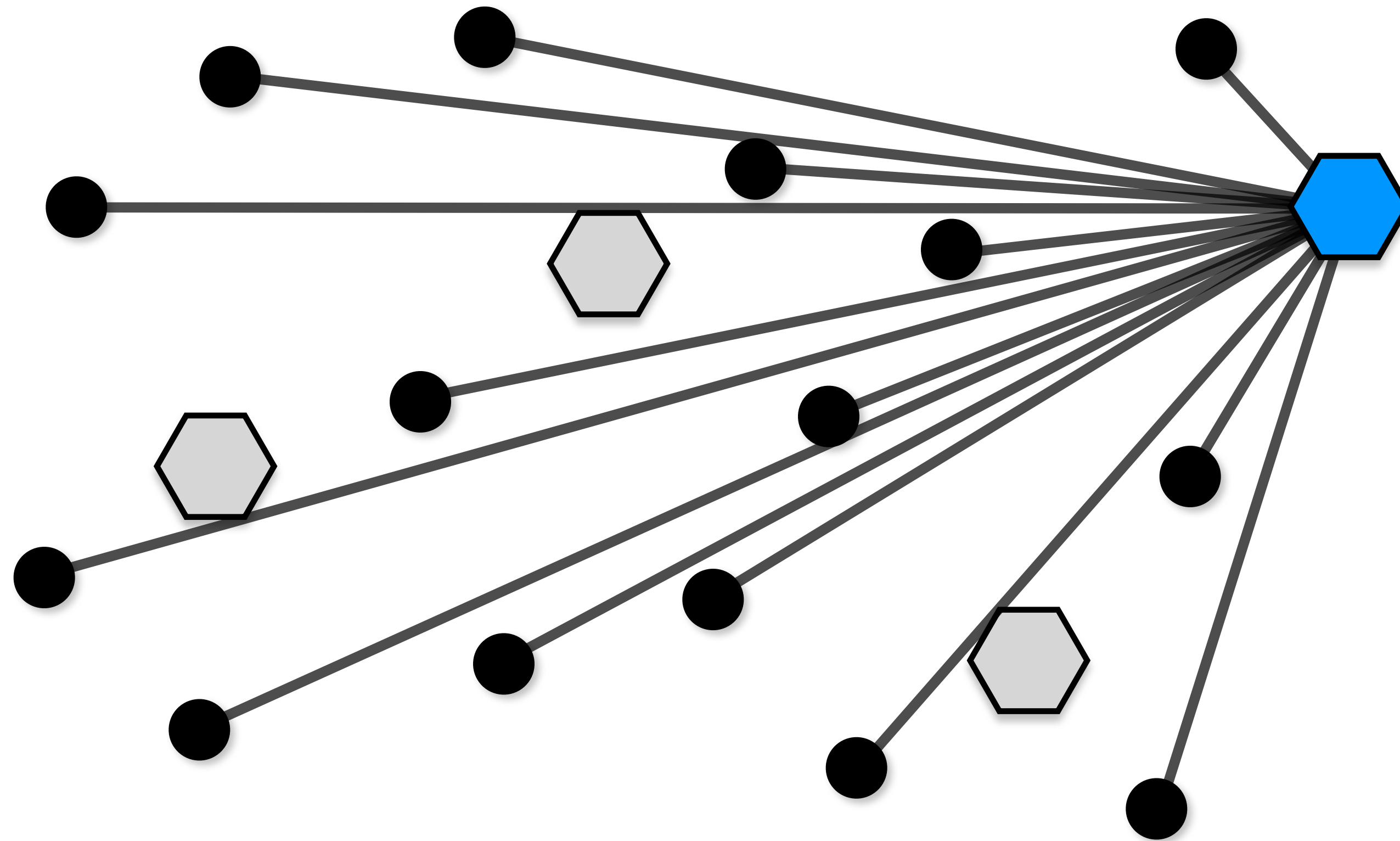
Warehouse Location



min warehouse setup cost + transport cost

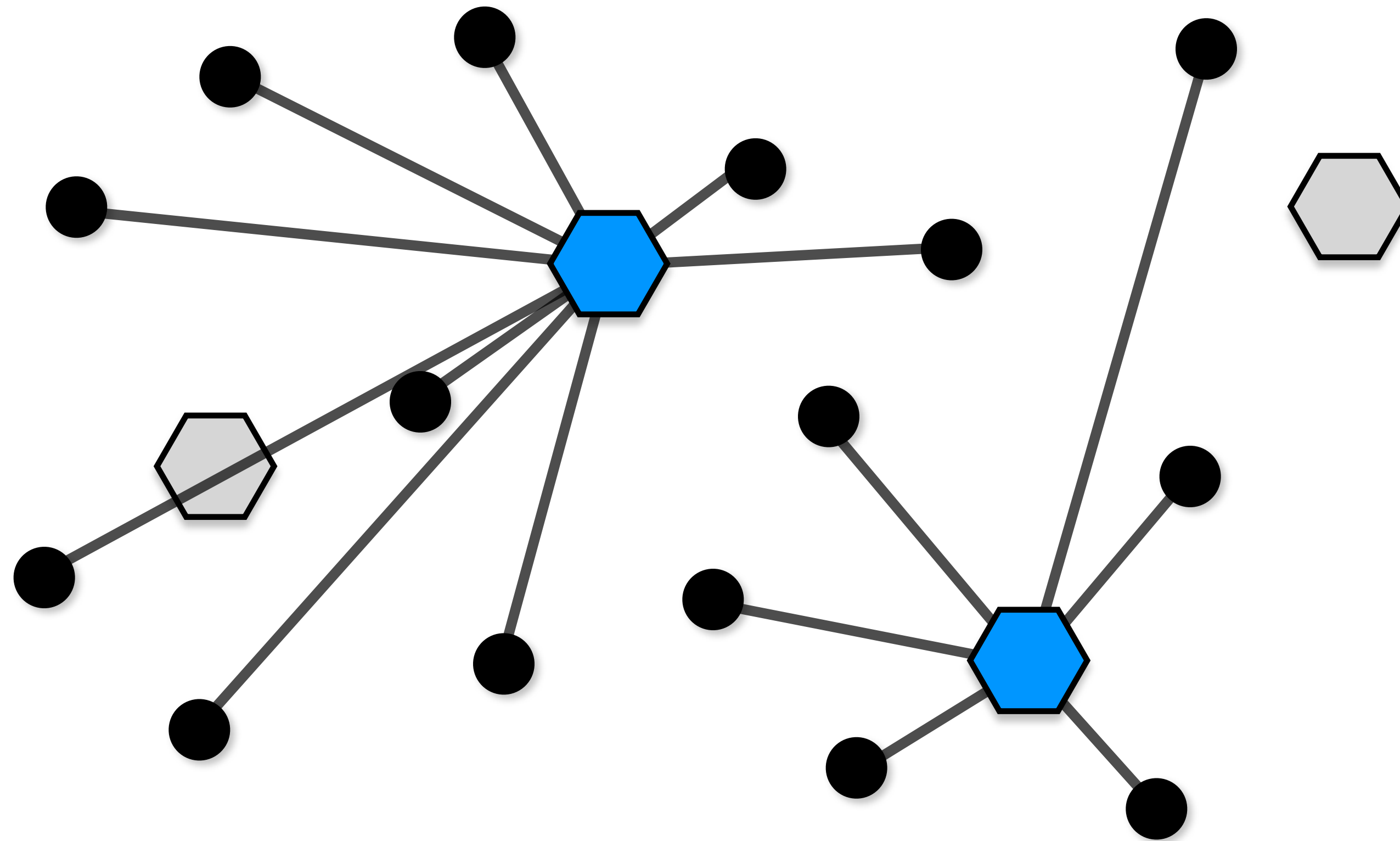
 - Warehouse  - Customer

Warehouse Location



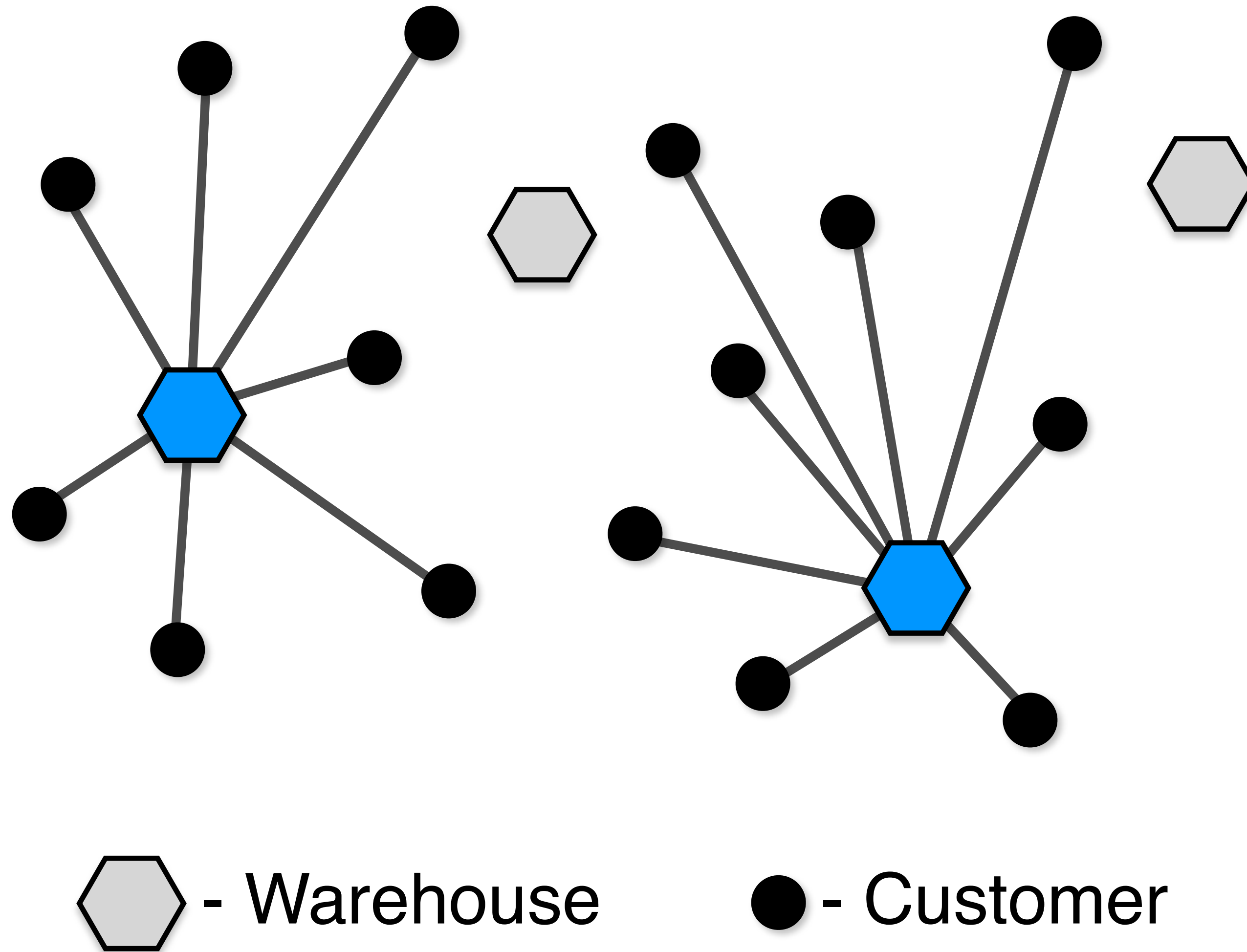
 - Warehouse  - Customer

Warehouse Location



 - Warehouse  - Customer

Warehouse Location



Warehouse Location

► Given

- a set of warehouses W , each warehouse with a fixed cost f_w
- a set of customers C
- a transportation cost $t_{w,c}$ from warehouse w to customer c

► Find which warehouses to open to minimize the fixed and transportation costs

Warehouse Location

- ▶ What are the decision variables?
 - o_w : whether warehouse w is open (0/1)
 - $a[c]$: the warehouse assigned to customer c

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Warehouse Location

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 - o_w : whether warehouse w is open (0/1)
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 - no constraints 😊
- ▶ What is the objective?

Warehouse Location

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 - o_w : whether warehouse w is open (0/1)
 - $a[c]$: the warehouse assigned to customer c
- ▶ What are the constraints?
 - no constraints 😊
- ▶ What is the objective?

$$\text{minimize } \sum_{w \in W} f_w o_w + \sum_{c \in C} t_{a[c], c}$$

Warehouse Location

- ▶ **Key observation**
 - once the warehouse locations have been chosen, the problem is easy
 - it suffices to assign a customer to the open warehouse minimizing its transportation cost

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Warehouse Location

- ▶ Key observation
 - once the warehouse locations have been chosen, the problem is easy
 - it suffices to assign a customer to the open warehouse minimizing its transportation cost
- ▶ What is the objective?

$$\text{minimize } \sum_{w \in W} f_w o_w + \sum_{c \in C} \min_{w \in W: o_w = 1} t_{w,c}$$

Warehouse Location

- ▶ Neighborhood
 - many possibilities

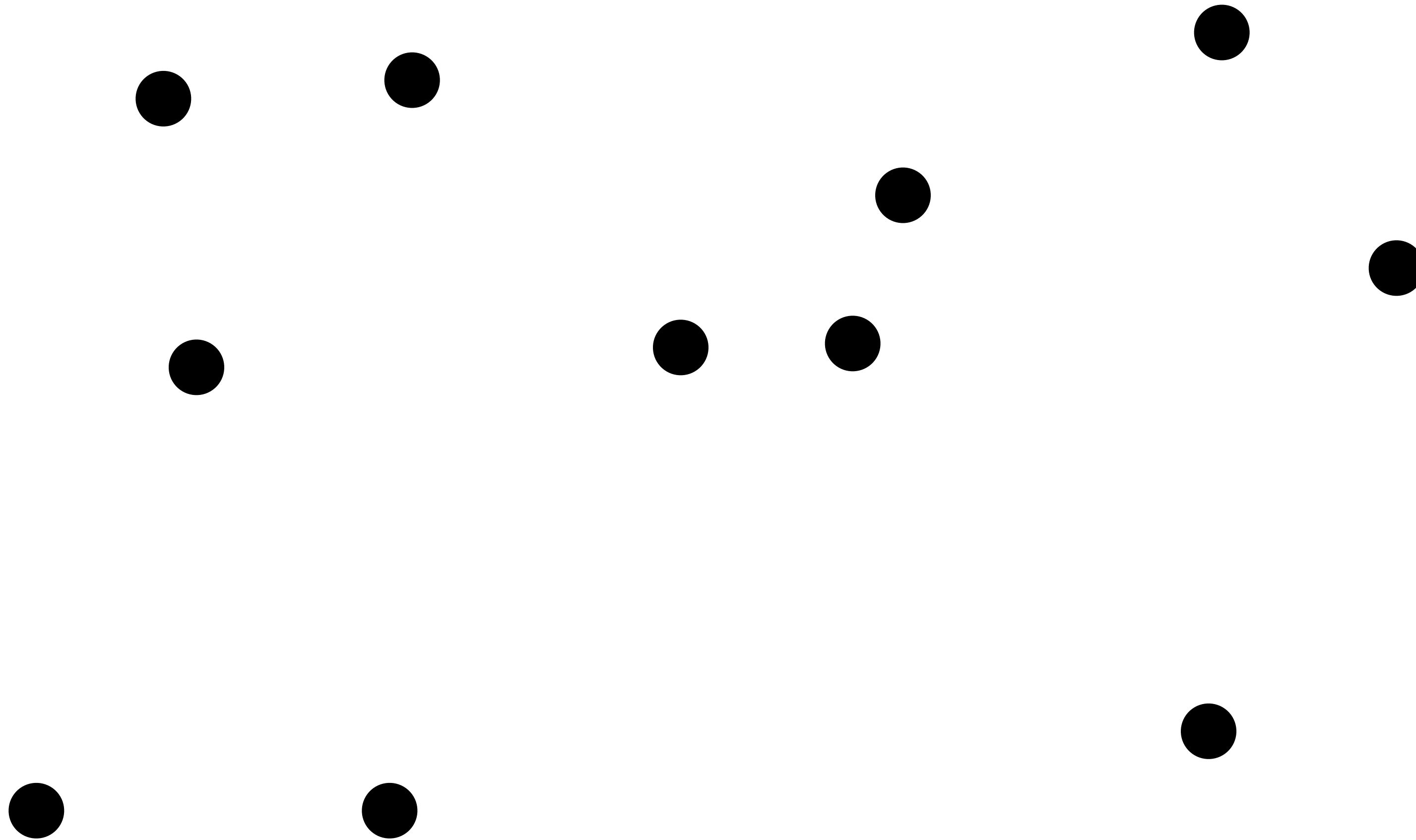
Warehouse Location

- ▶ Neighborhood
 - many possibilities
- ▶ Simplest neighborhood
 - open and close warehouses
 - that is, flip the value of o_w

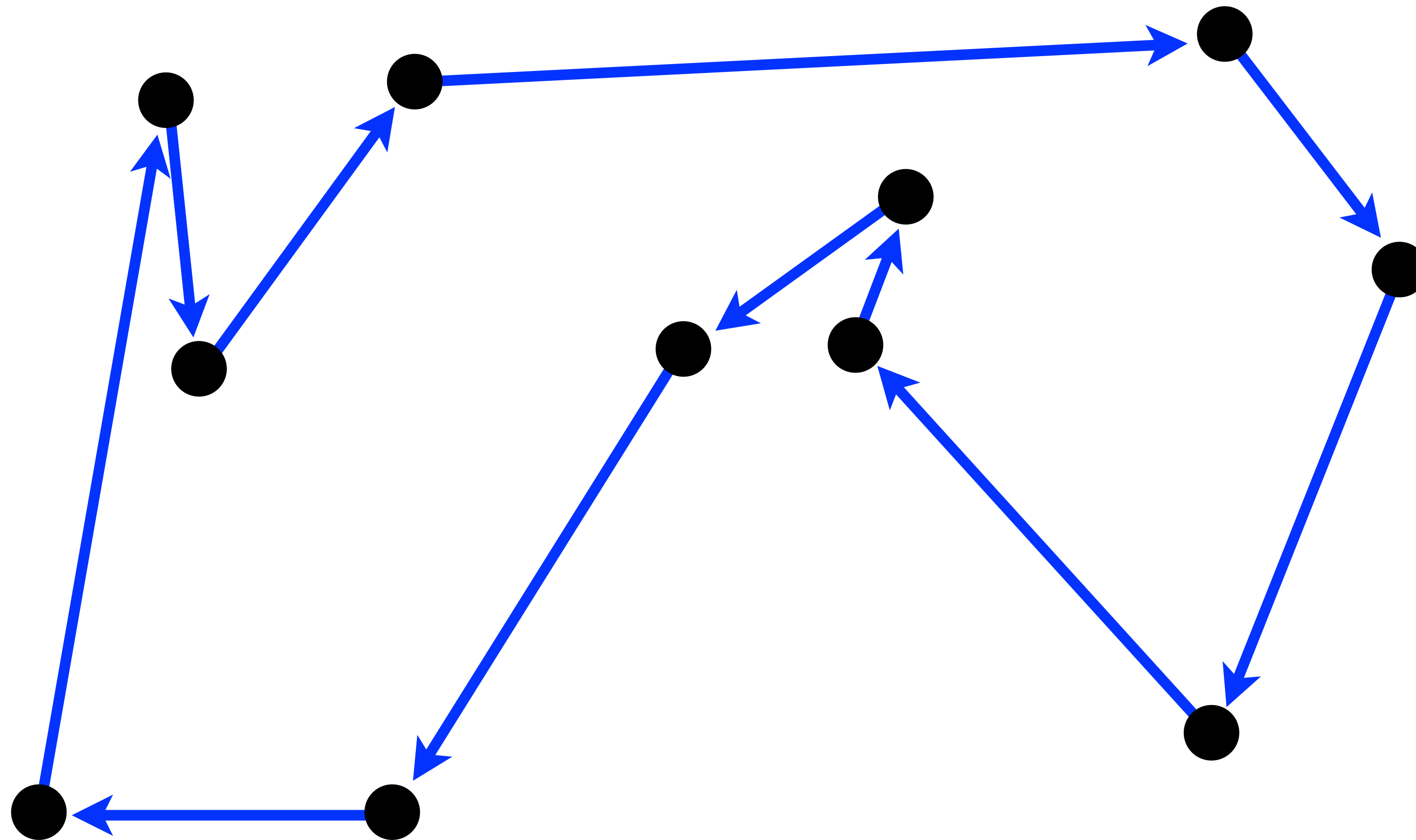
Warehouse Location

- ▶ Neighborhood
 - many possibilities
- ▶ Simplest neighborhood
 - open and close warehouses
 - that is, flip the value of o_w
- ▶ Union of neighborhoods
 - open and close a warehouse
 - swap two warehouses
 - close one and open the other

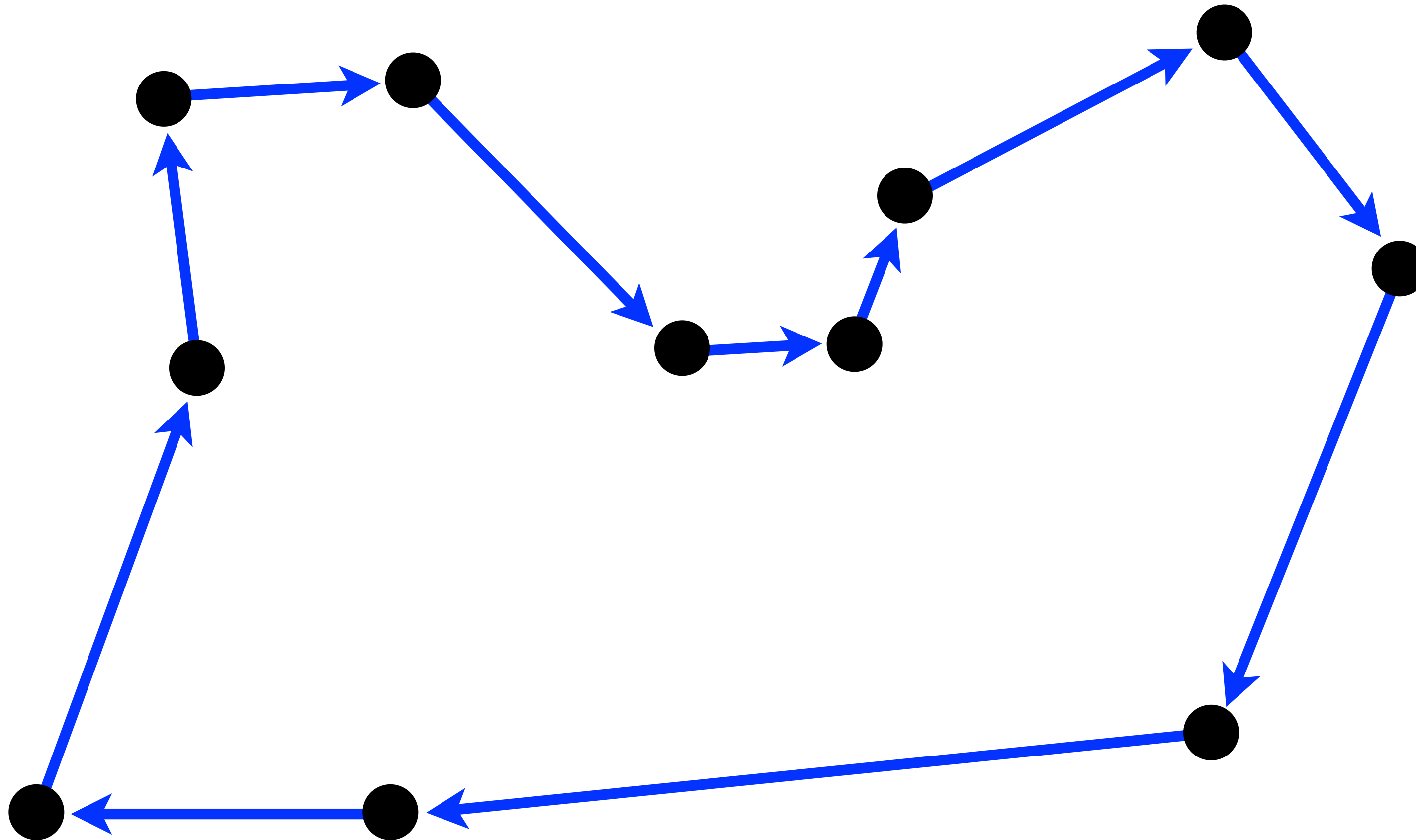
Traveling Salesman Problem



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Traveling Salesman Problem

► Given

- a set C of cities to visit
- a symmetric distance matrix d between every two cities

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► Find

- a tour of minimal cost visiting each city exactly once

Traveling Salesman Problem

- ▶ **Given**

- a set C of cities to visit
- a symmetric distance matrix d between every two cities

- ▶ **Find**

- a tour of minimal cost visiting each city exactly once

- ▶ **The traveling salesman problem (TSP) is probably the most studied combinatorial problem**

Traveling Salesman Problem

- ▶ Decision variables
 - like in the Euler tour
 - specify where to go next for every city

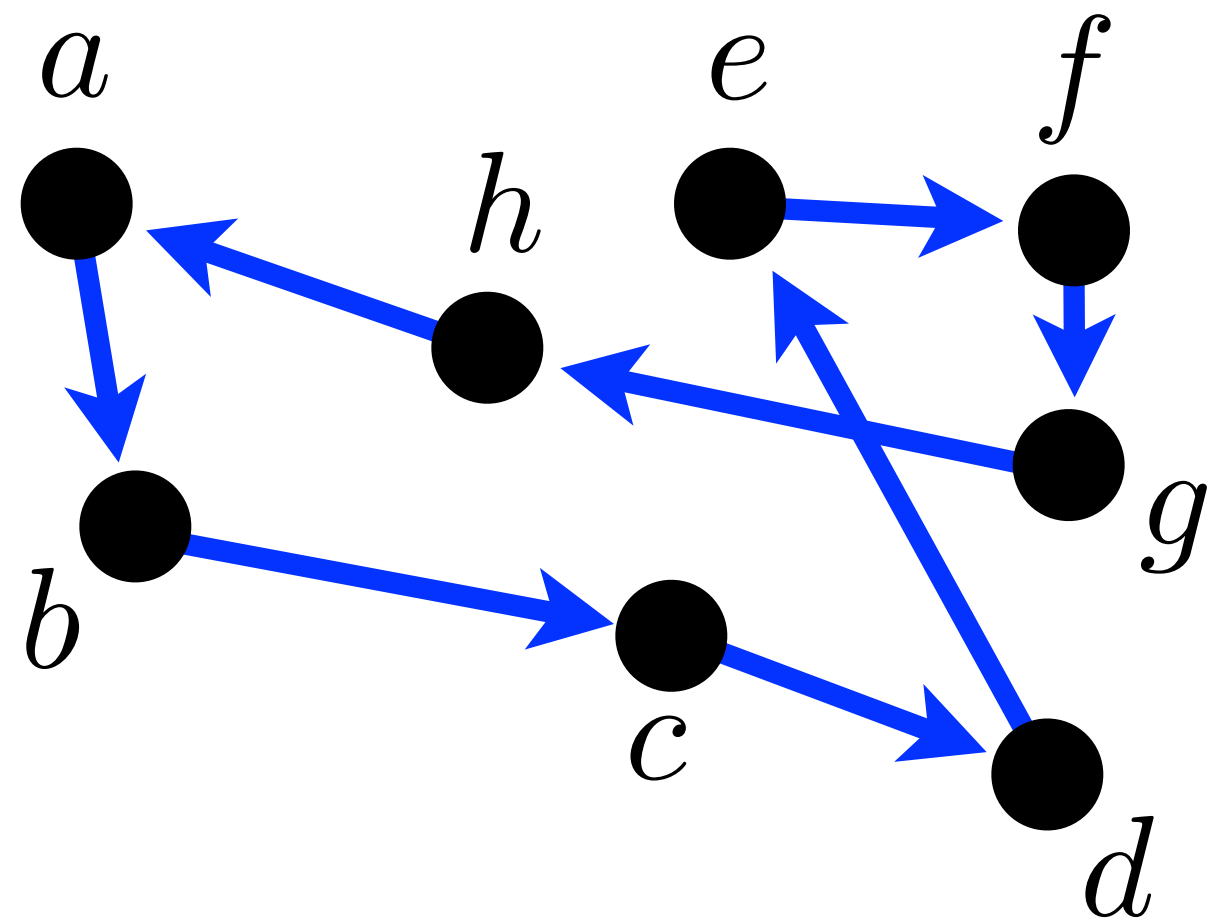
```
range Cities = 1..n;  
int distance[Cities,Cities] = ...;  
var{int} next[Cities] in Cities;  
minimize  
    sum(c in Cities) d[c,next[c]]  
subject to  
    circuit(next);
```

Local Search for the TSP

- ▶ 2-OPT neighborhood for the TSP
 - stay feasible, that is always maintain a tour
 - select two edges and replace them by two other edges

Local Search for the TSP

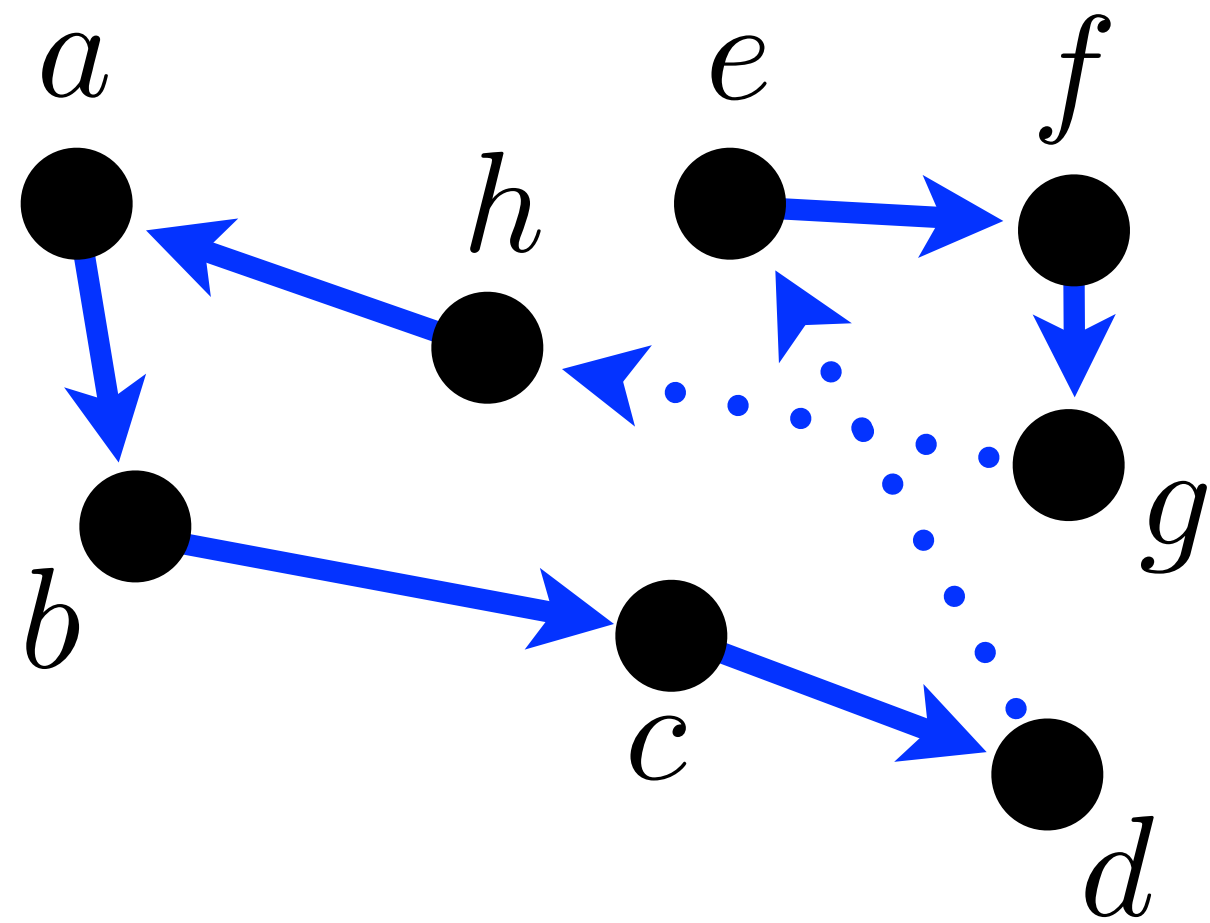
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$a \rightarrow b \rightarrow c \rightarrow d \rightarrow e \rightarrow f \rightarrow g \rightarrow h \rightarrow a$

Local Search for the TSP

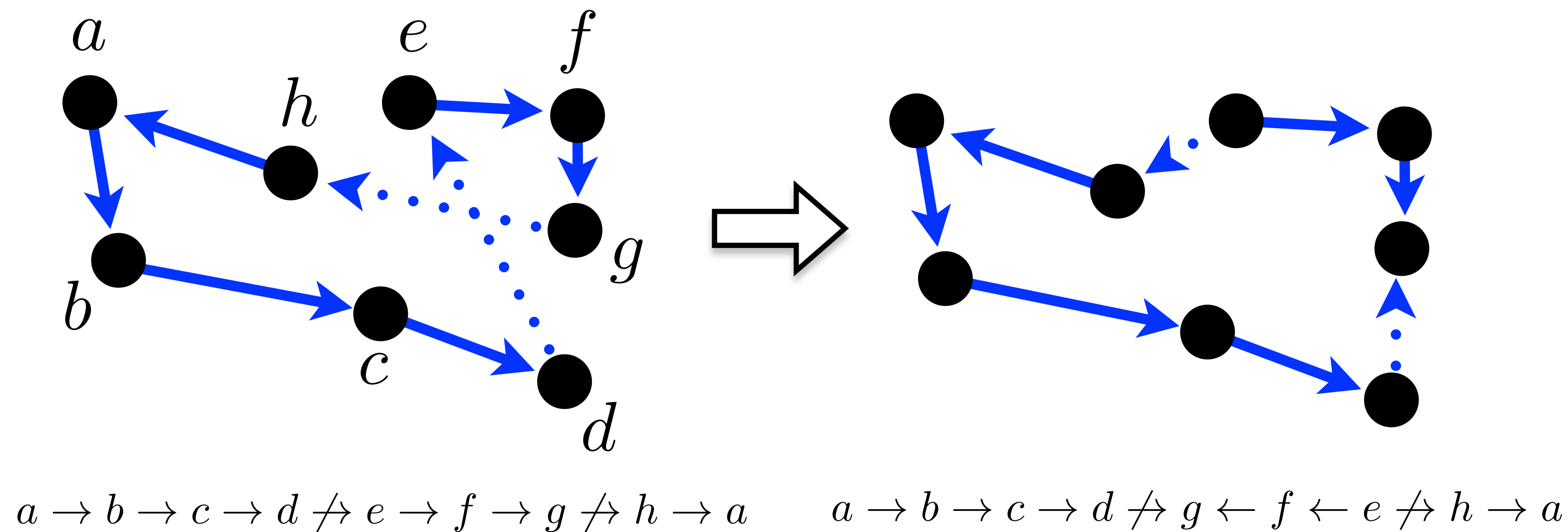
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$a \rightarrow b \rightarrow c \rightarrow d \not\rightarrow e \rightarrow f \rightarrow g \not\rightarrow h \rightarrow a$

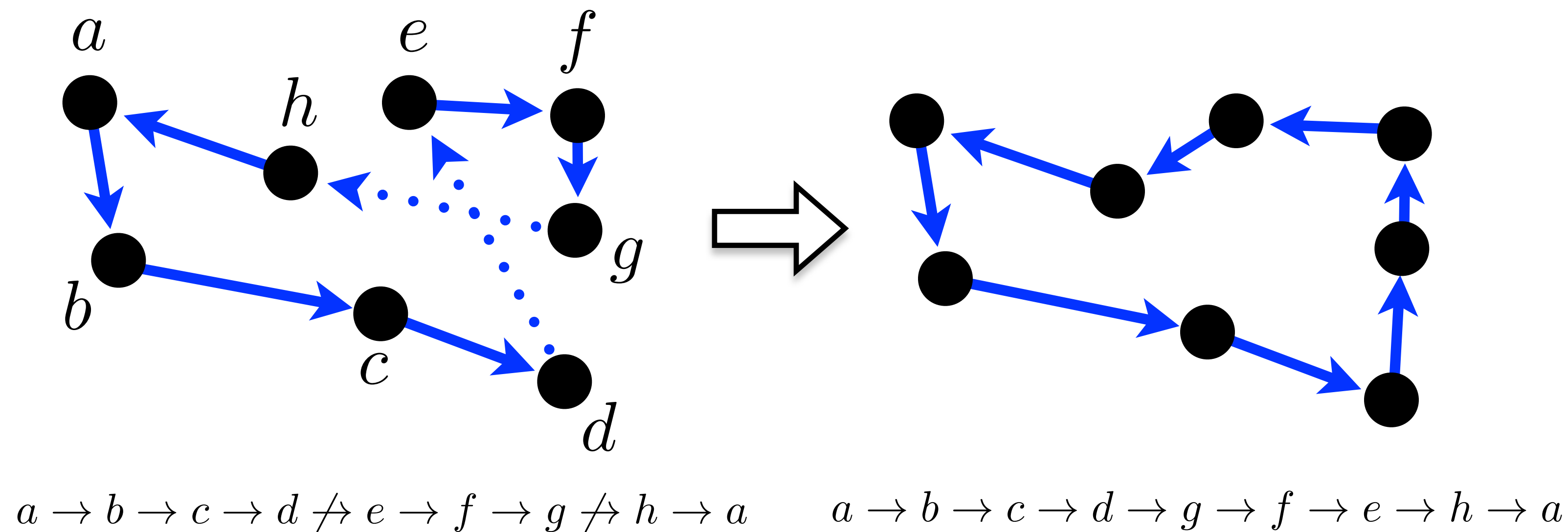
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Local Search for the TSP

► 2-OPT

- the neighborhood is the set of all tours that can be reached by swapping two edges
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Local Search for the TSP

► 2-OPT

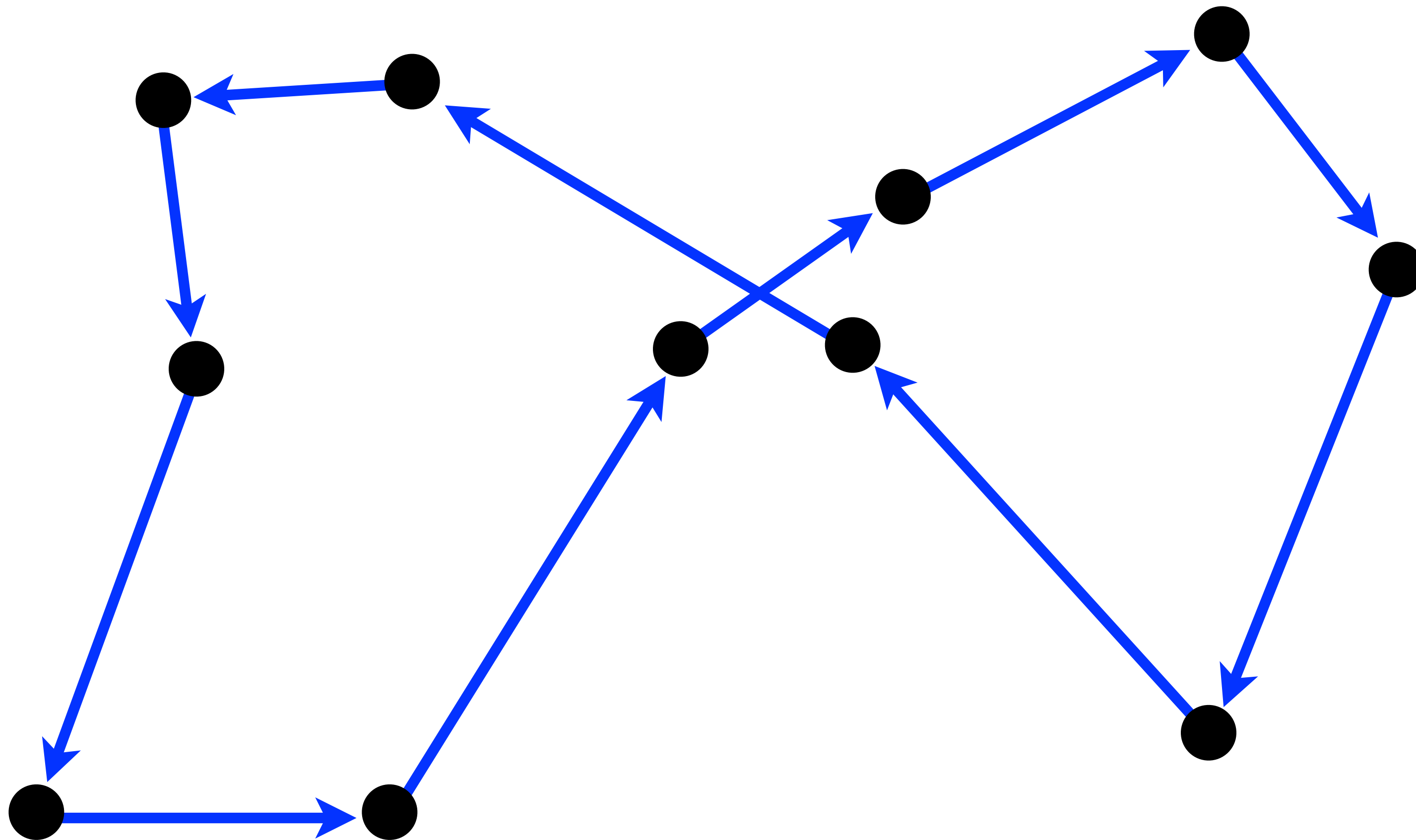
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► 3-OPT

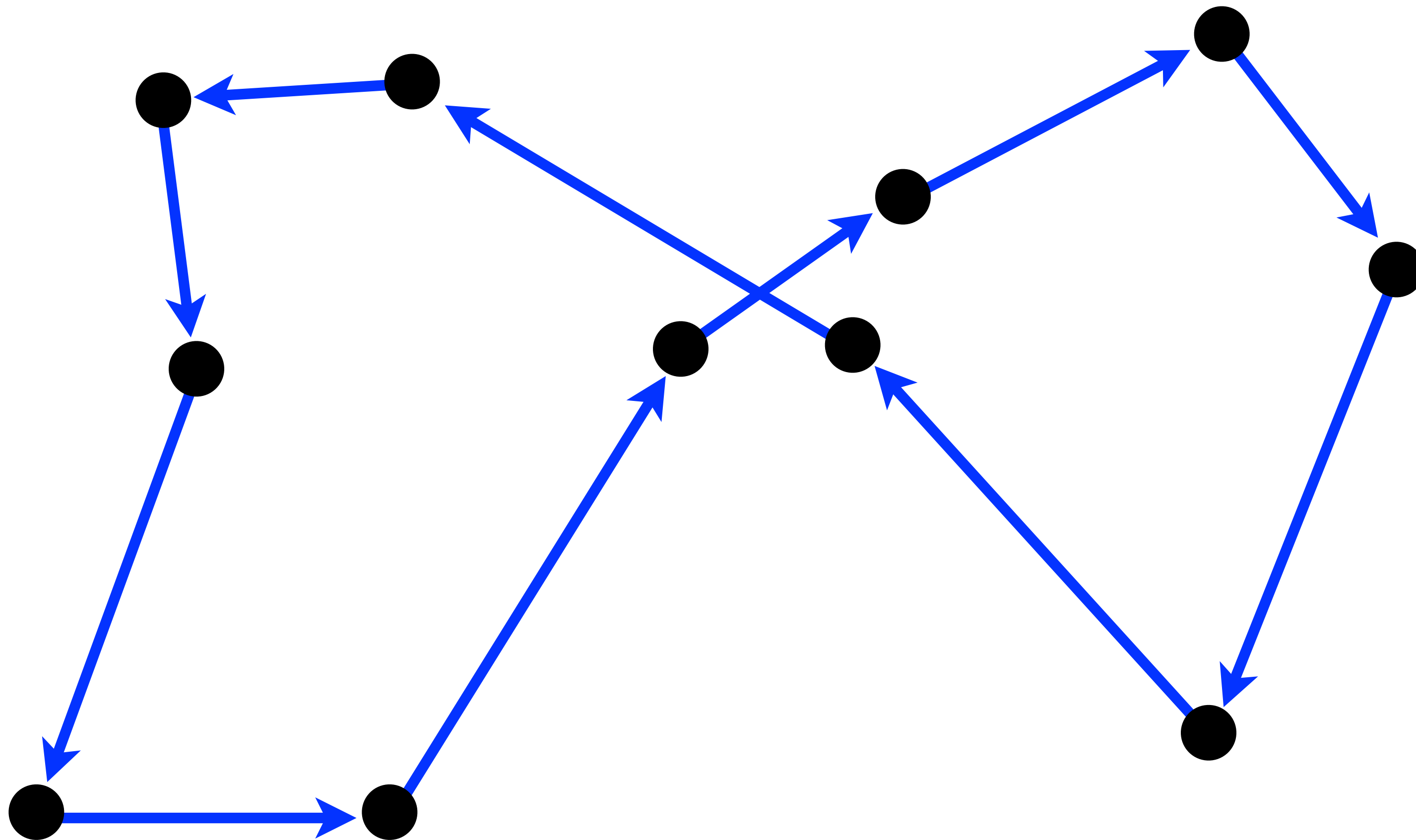
- the neighborhood is the set of all tours that can be reached by swapping three edges

better than 2-OPT in quality
expensive in computing

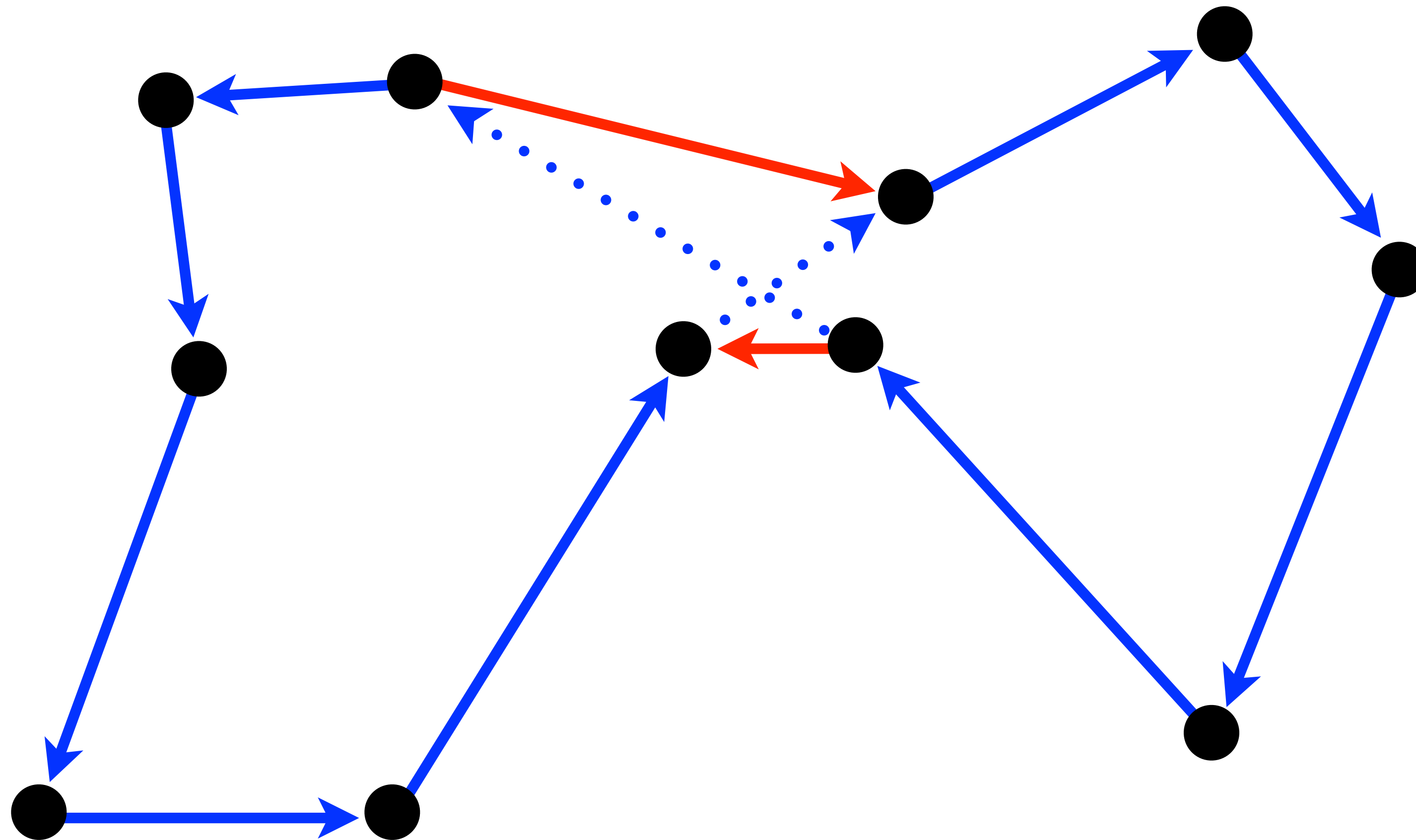
2-OPT



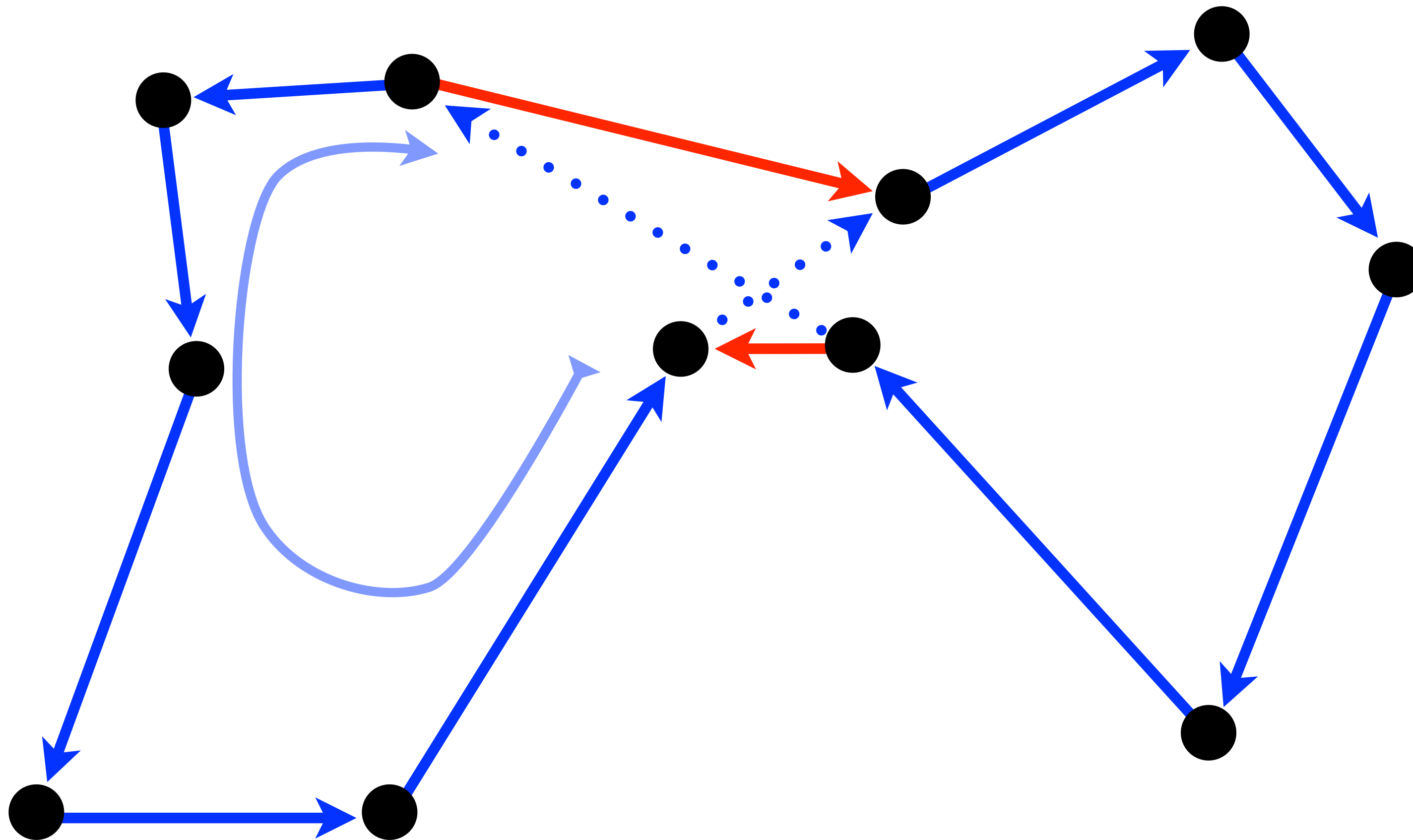
2-OPT



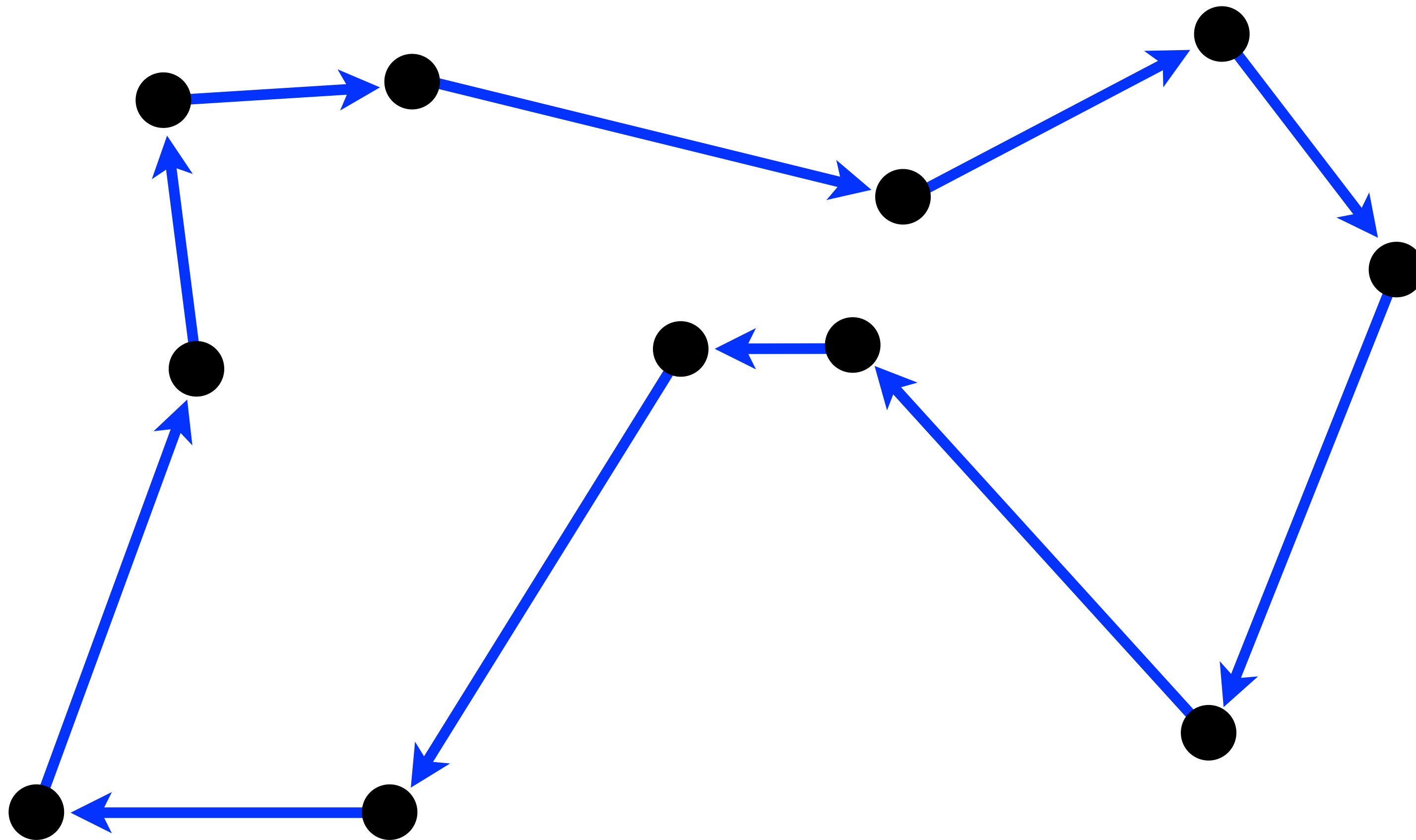
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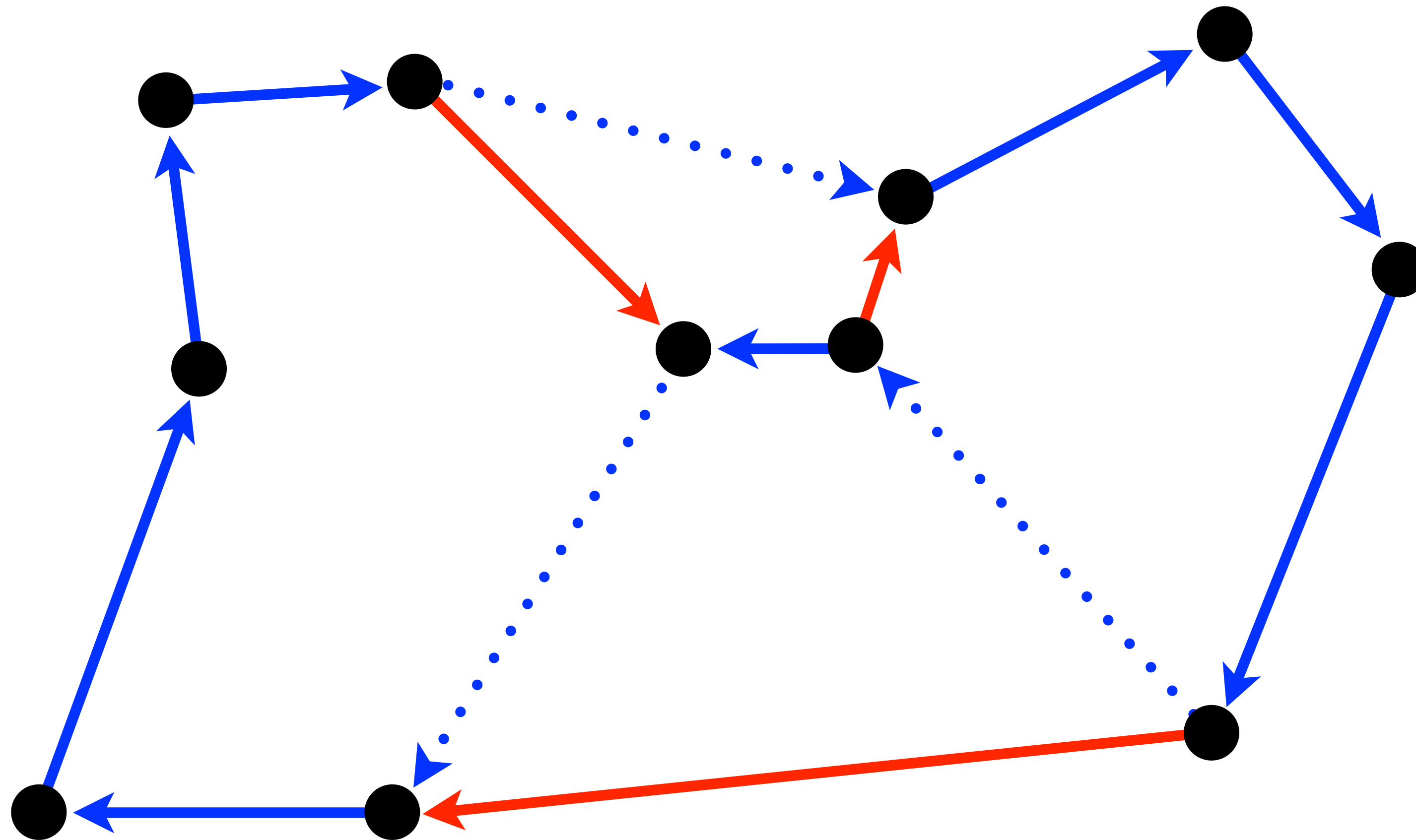
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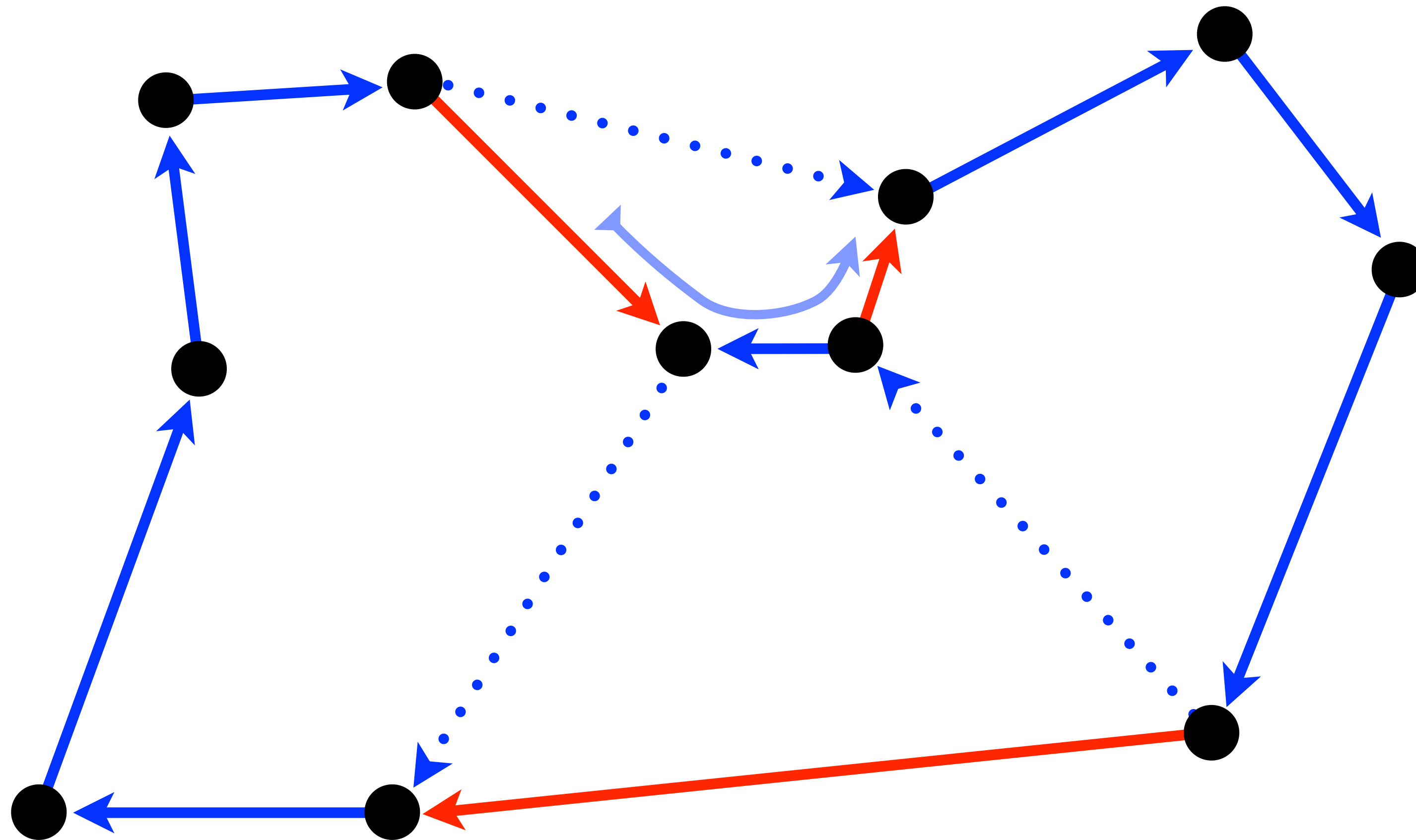
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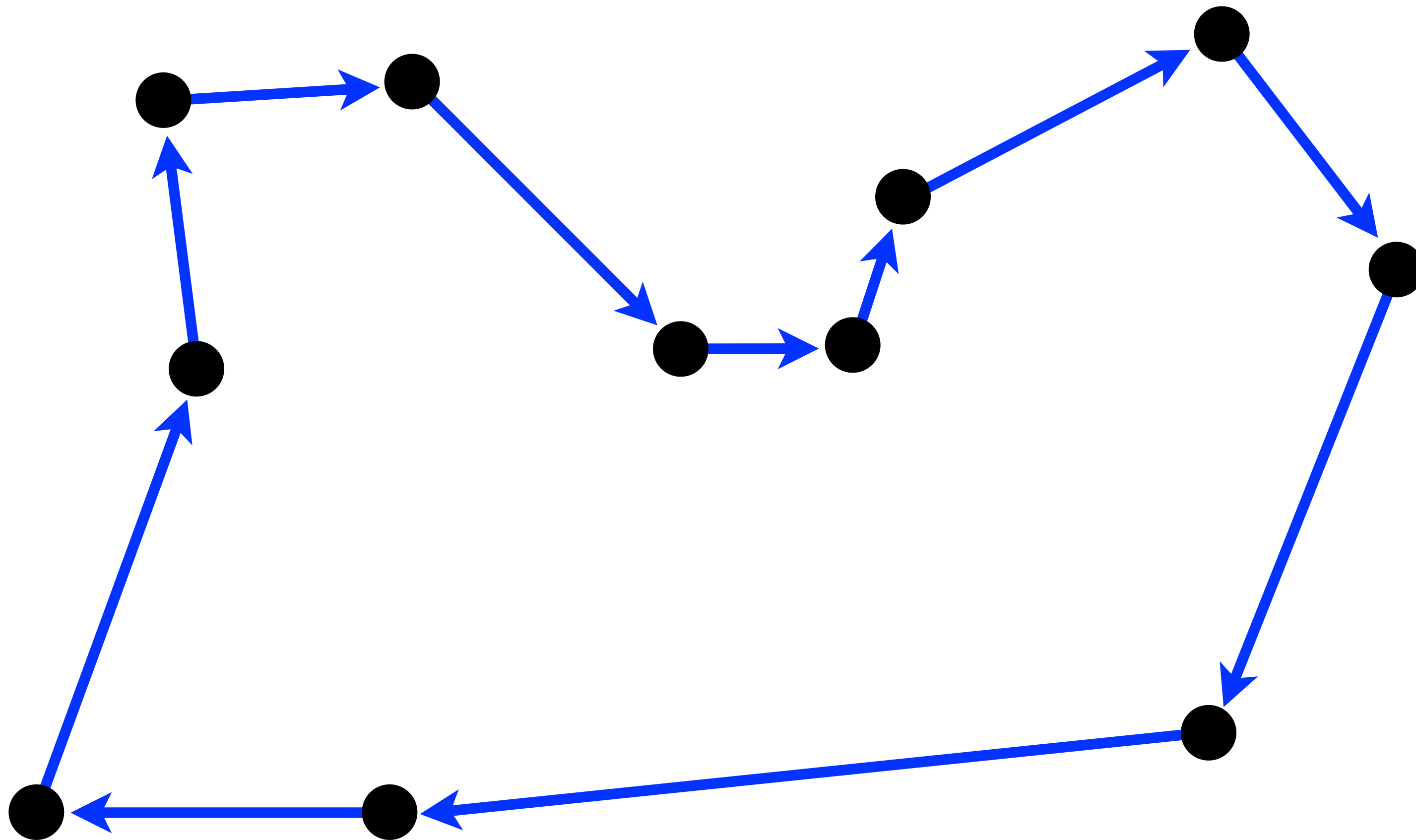
3-OPT



3-OPT



3-OPT



Local Search for the TSP

► 2-OPT

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- the neighborhood is the set of all tours that can be reached by swapping three edges
- much better than 2-OPT in quality but more expensive

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4-OPT

► 2-OPT

► 3-OPT

- the neighborhood is the set of all tours that can be reached by swapping three edges
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► 4-OPT

- often marginally better but much more expensive

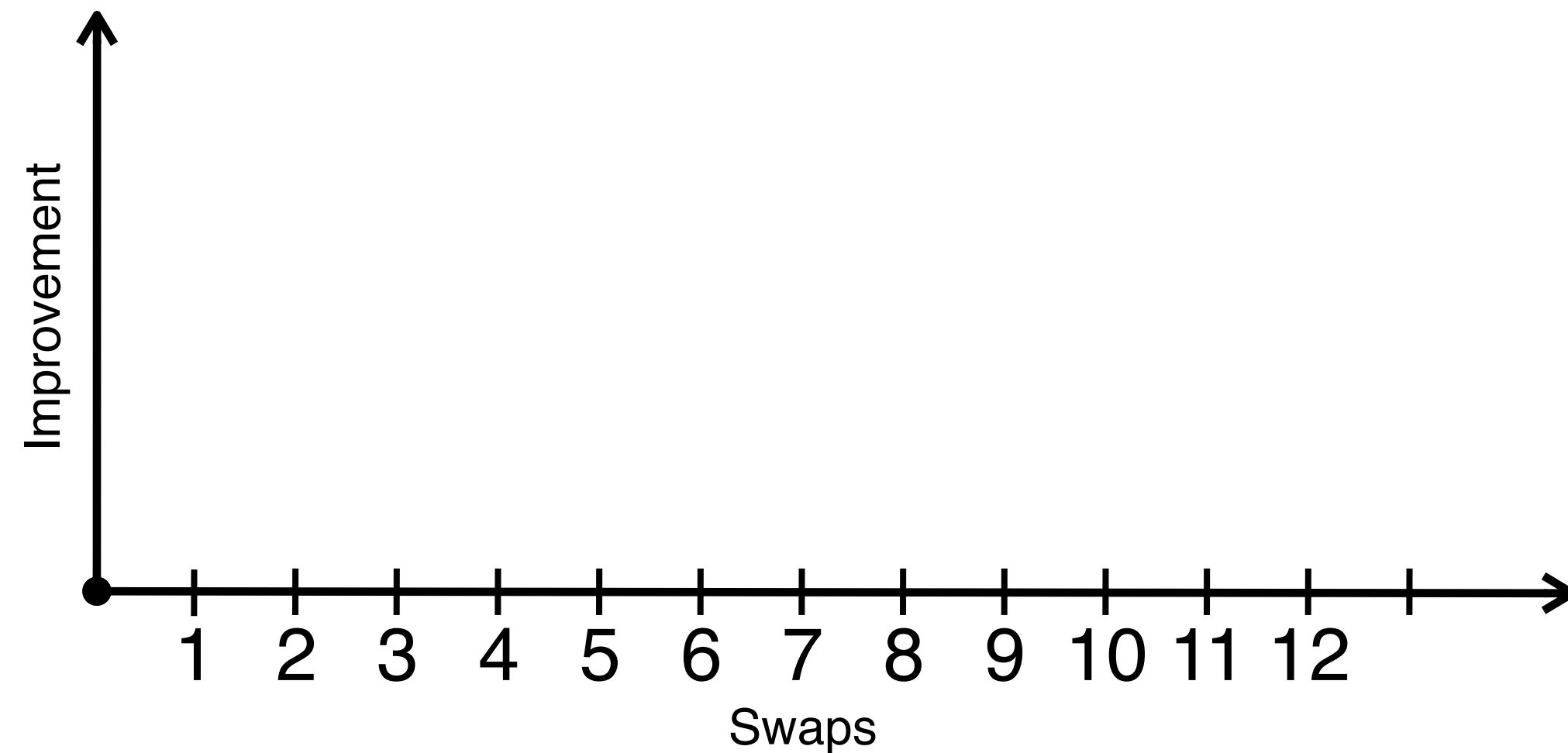
► K-OPT

- replace the notion of one favorable swaps by a search of a favorable sequence of swaps
- do not search for the entire set of sequences but build one incrementally

K-OPT

► K-OPT

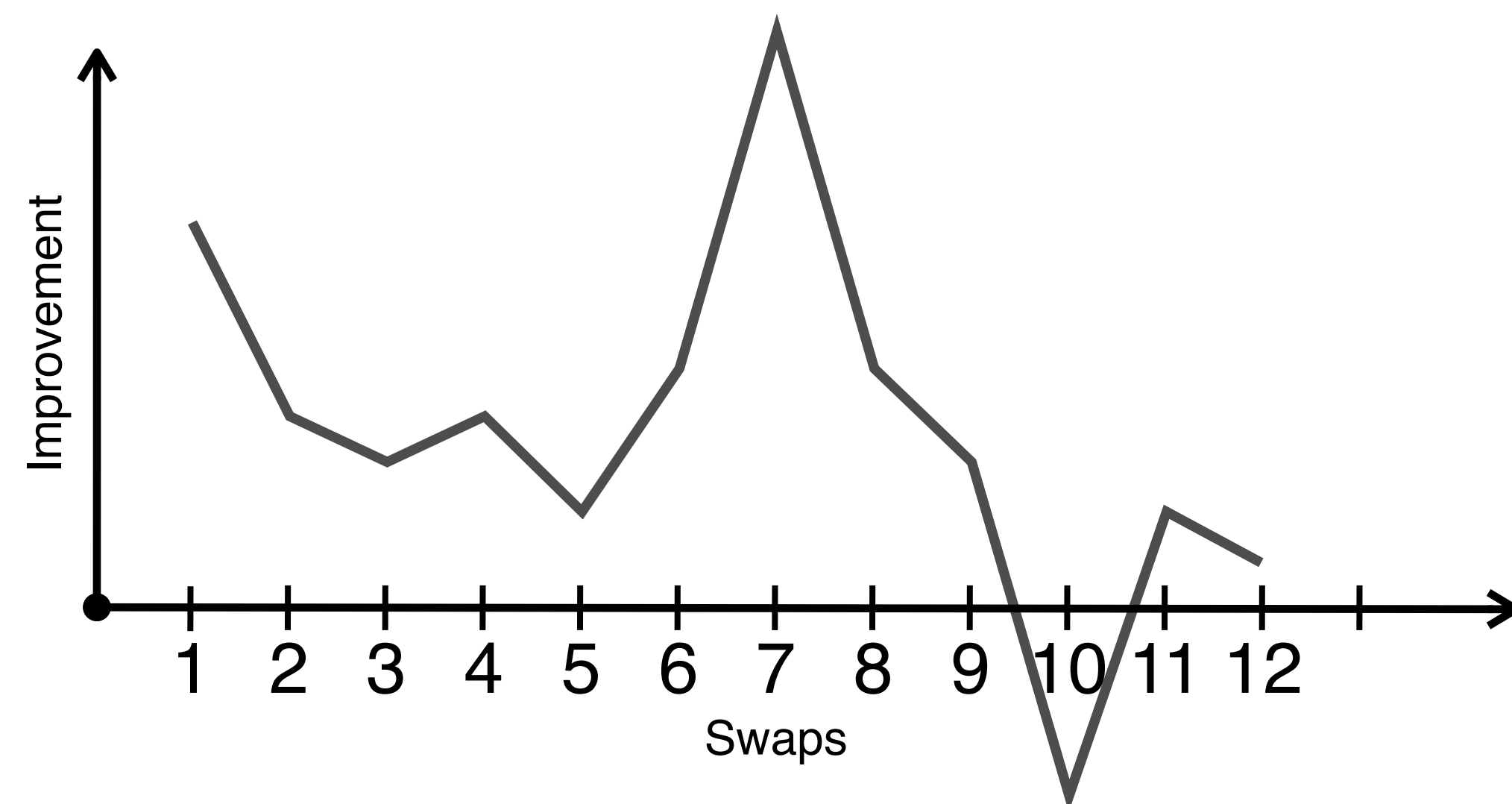
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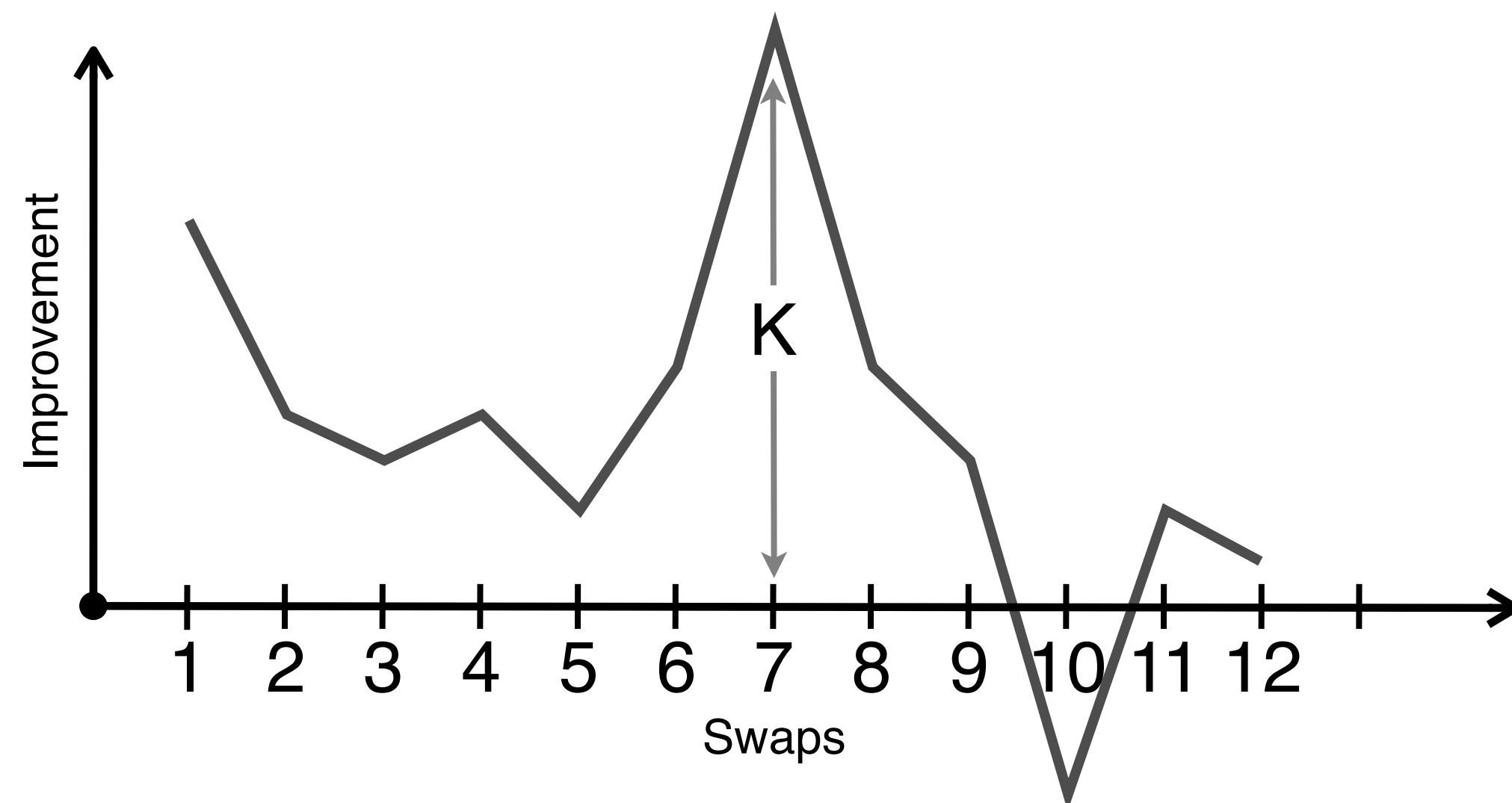
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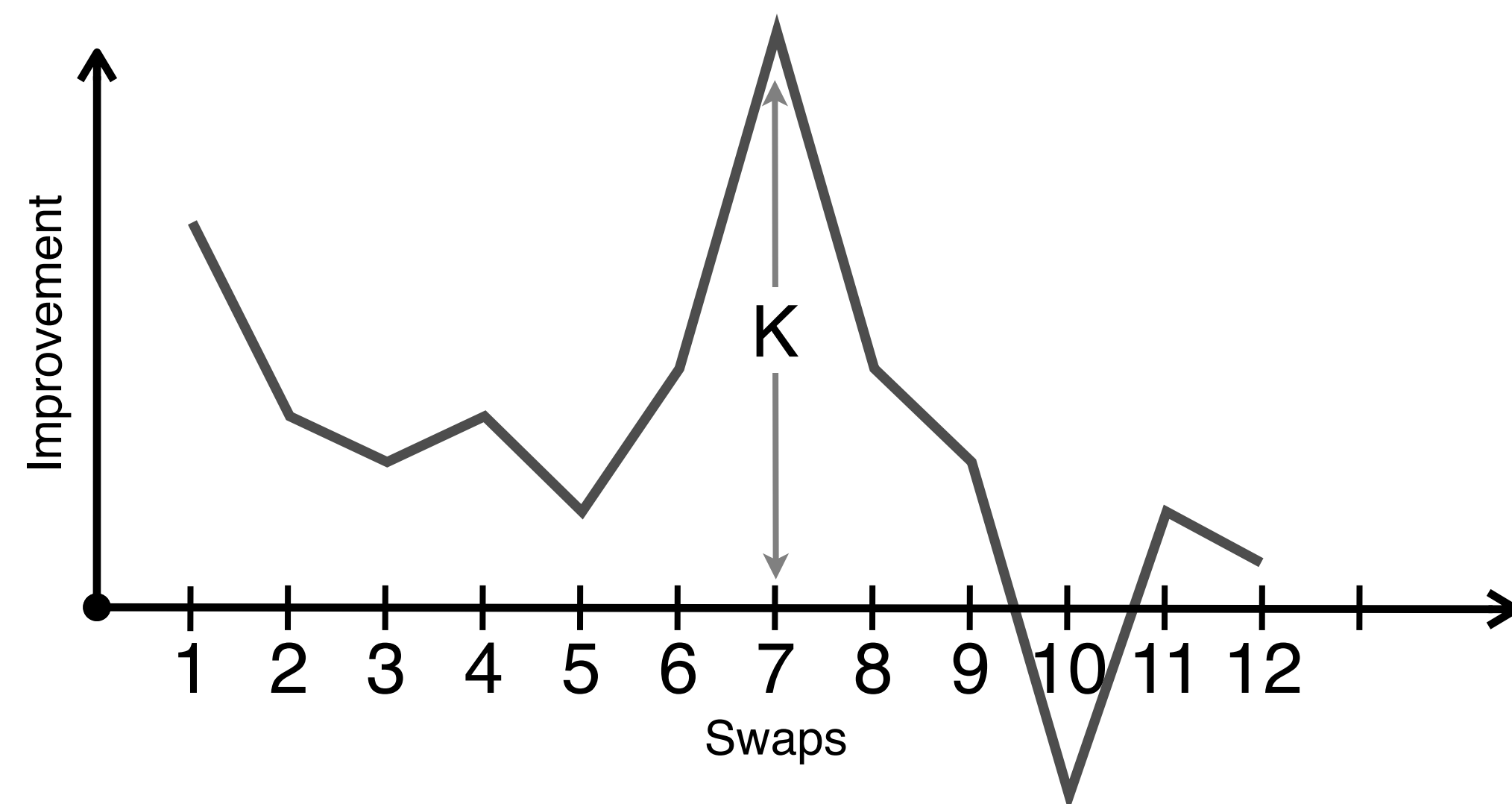
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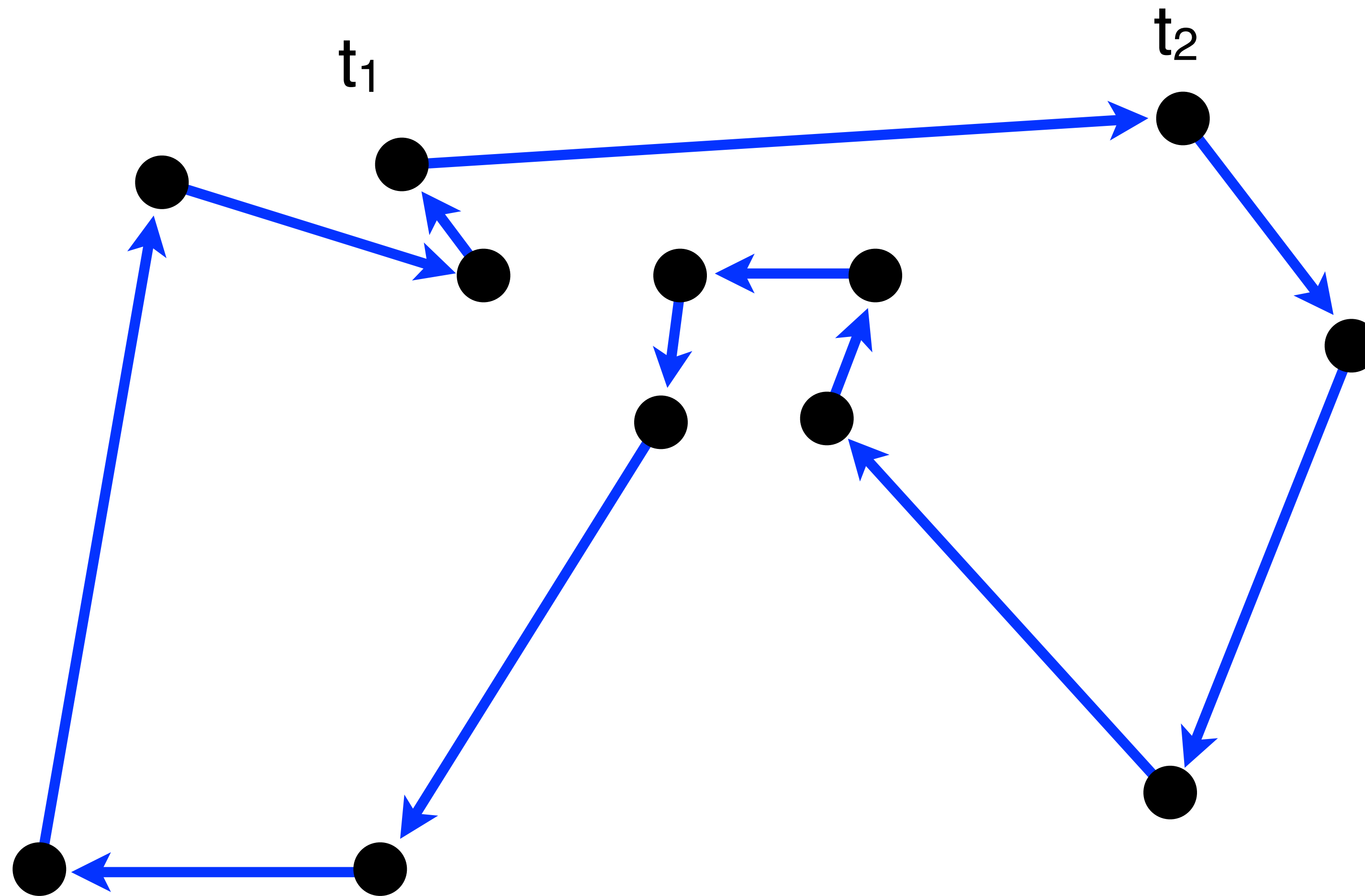
- find a good k dynamically at a fraction of the cost
- explore a sequence of swaps of increasing sizes



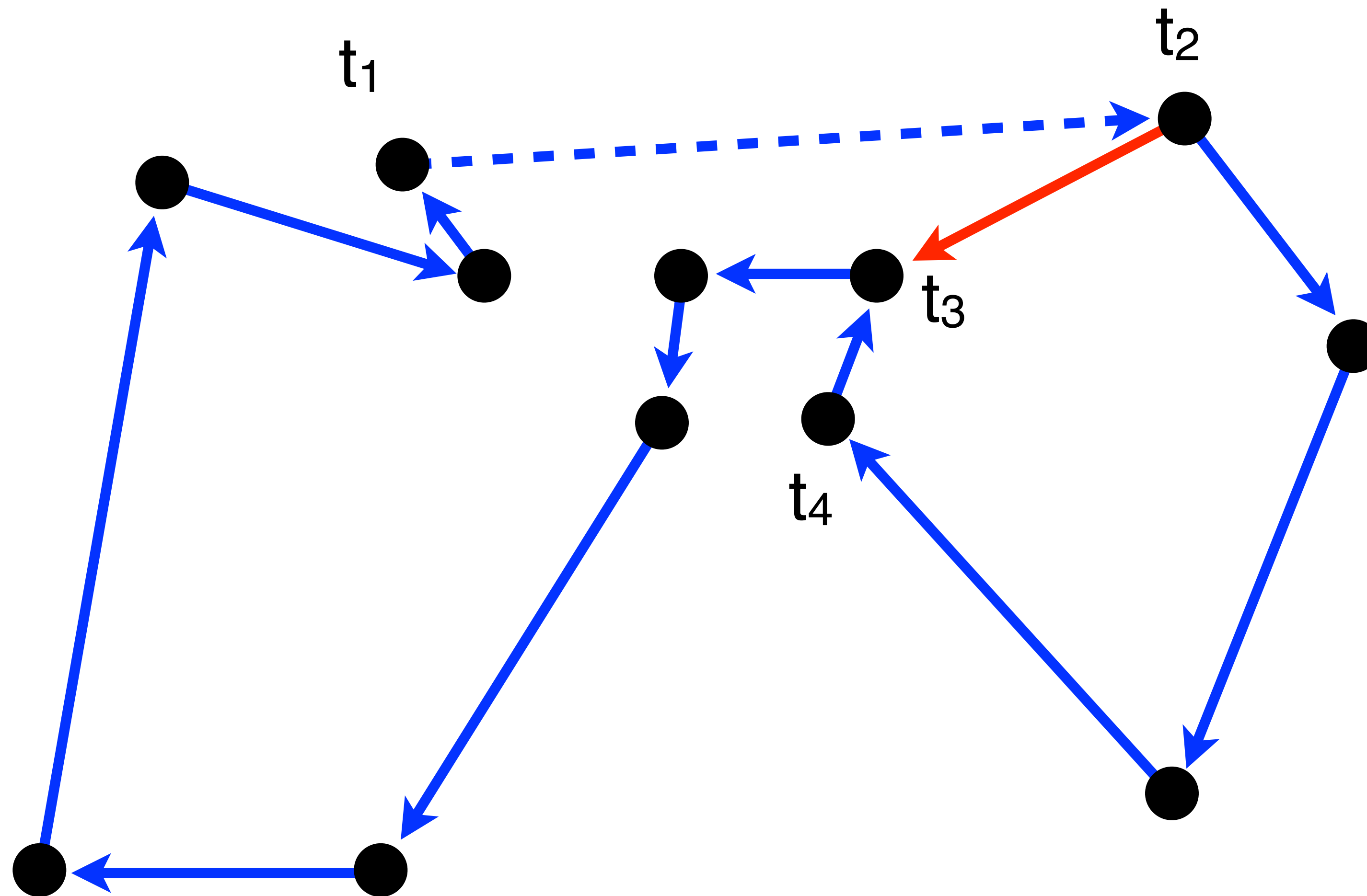
► K-OPT

- choose a vertex t_1 and its edge $x_1 = (t_1, t_2)$
- choose an edge $x_2 = (t_2, t_3)$ with $d(x_2) < d(x_1)$
- if none exist, restart with another vertex
- else we have a solution by removing the edge (t_4, t_3) and connecting (t_1, t_4)

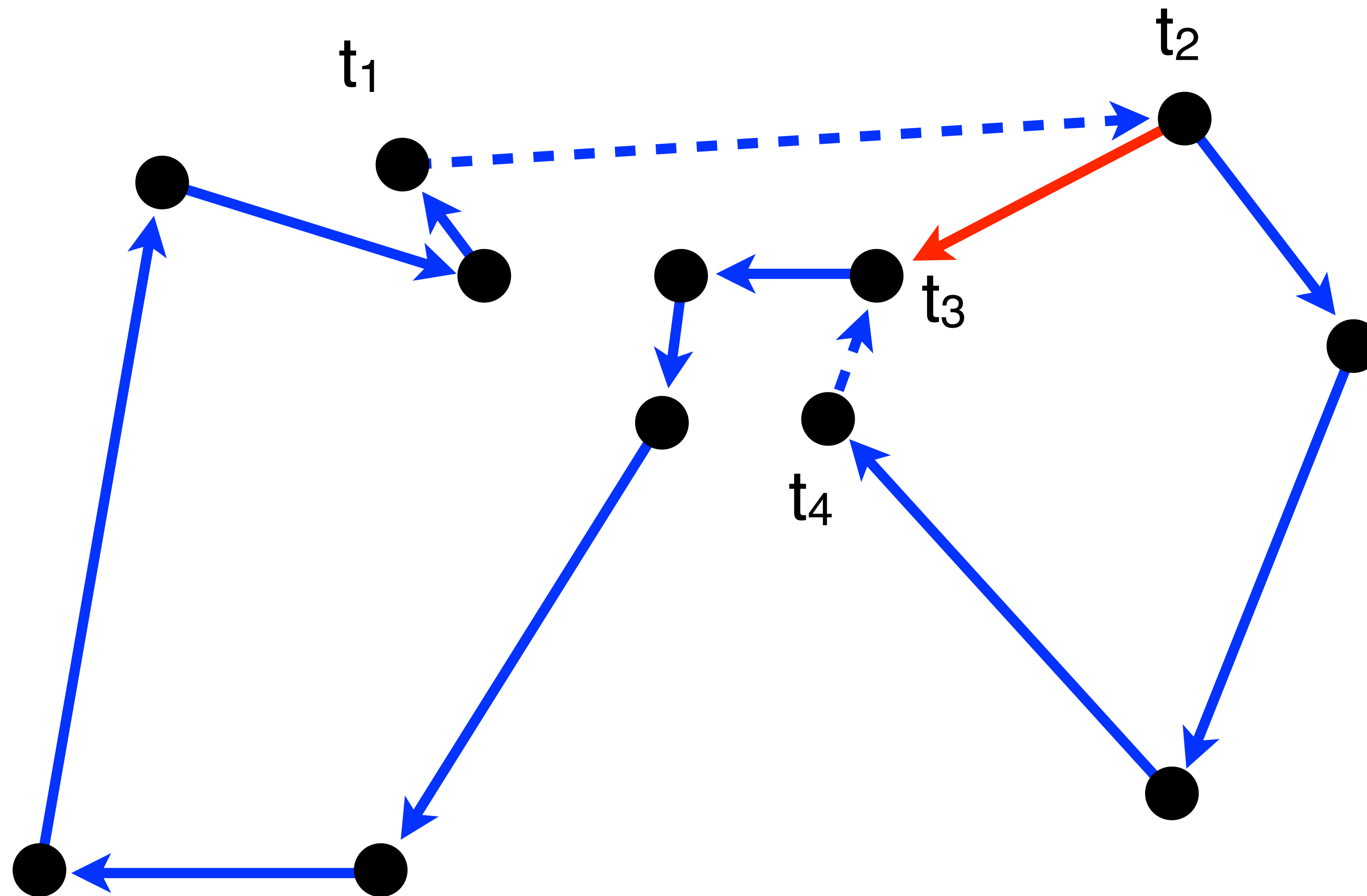
K-OPT (first iteration)



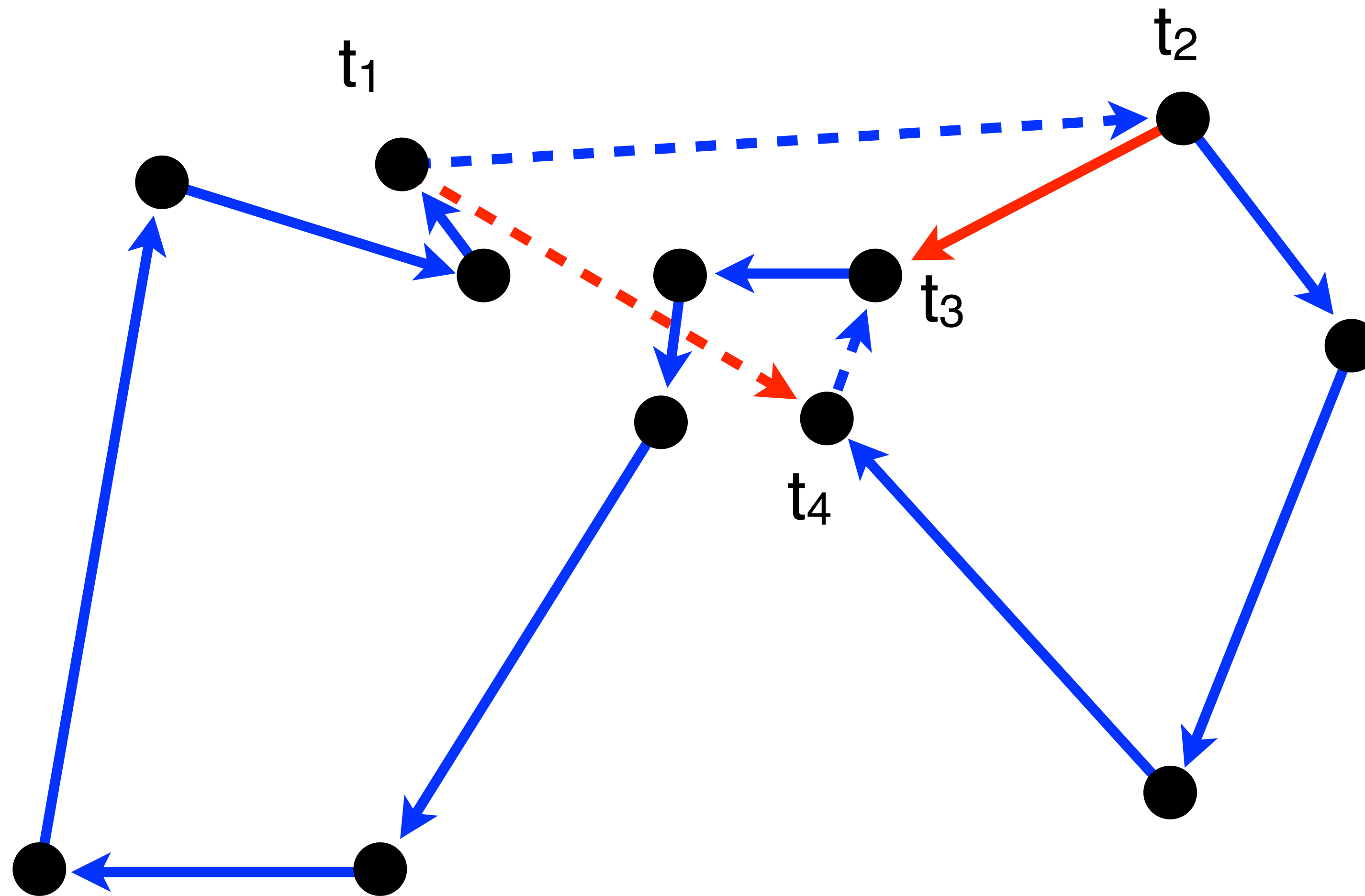
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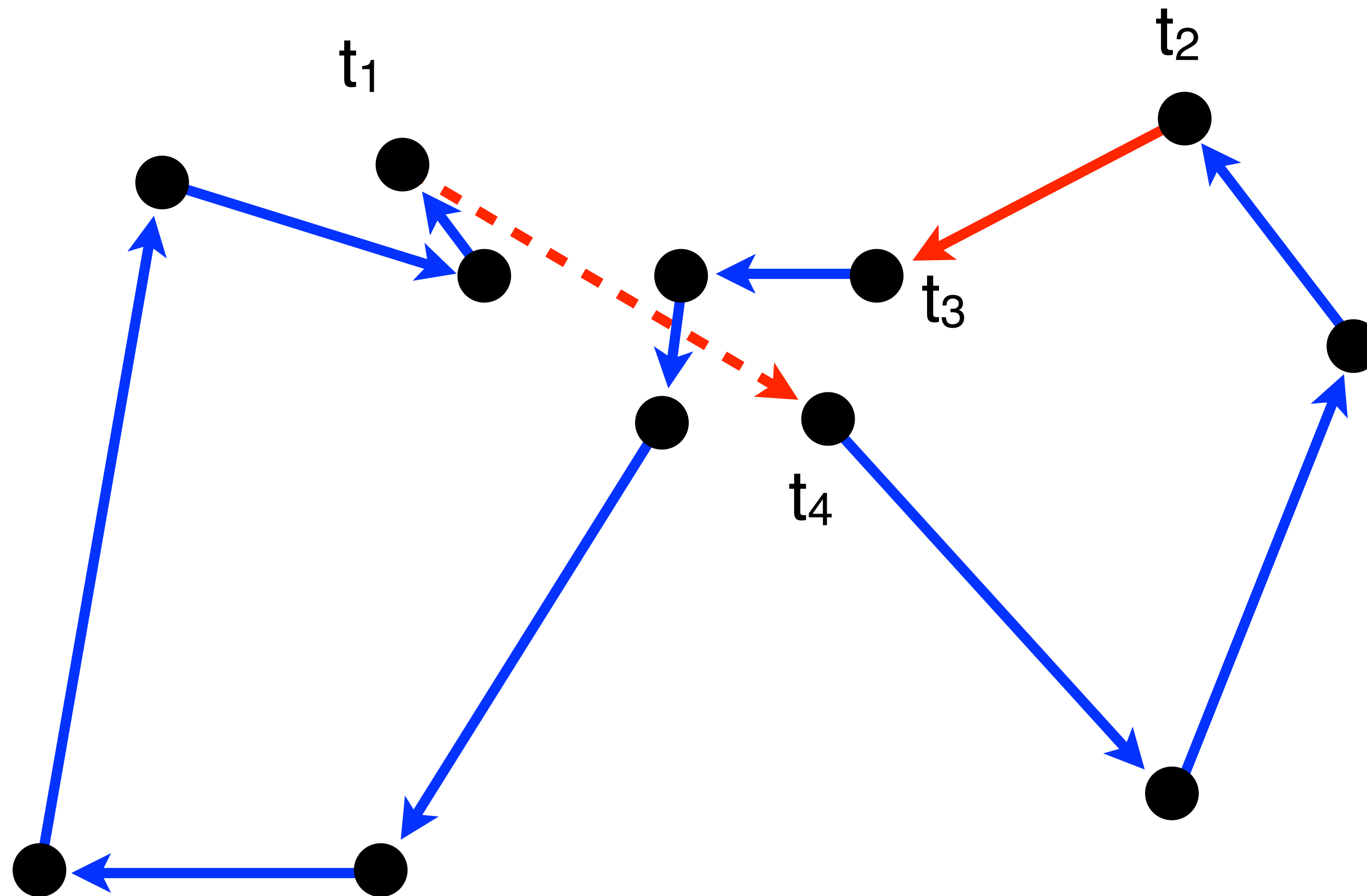
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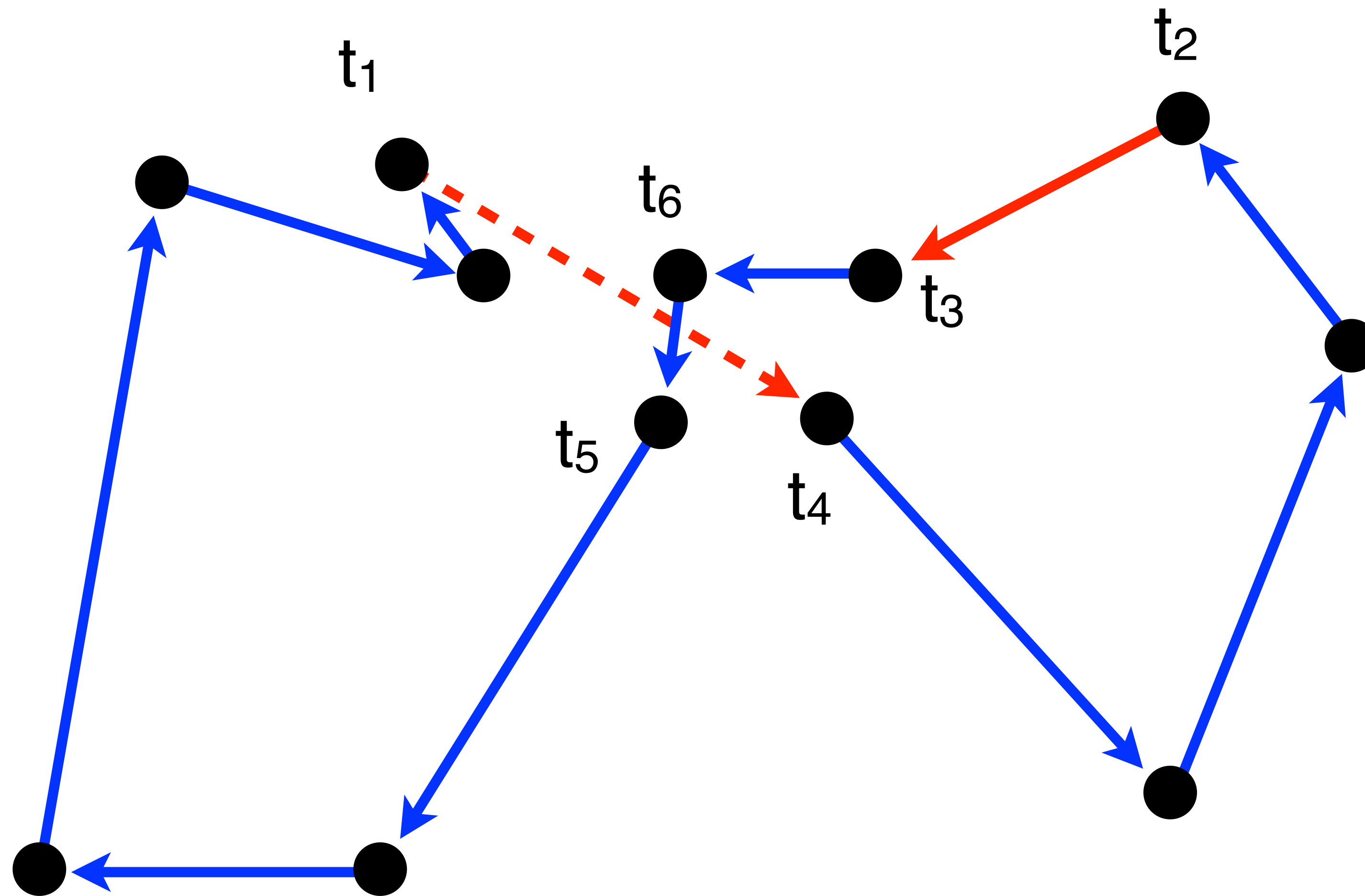
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- compute the cost but do not connect

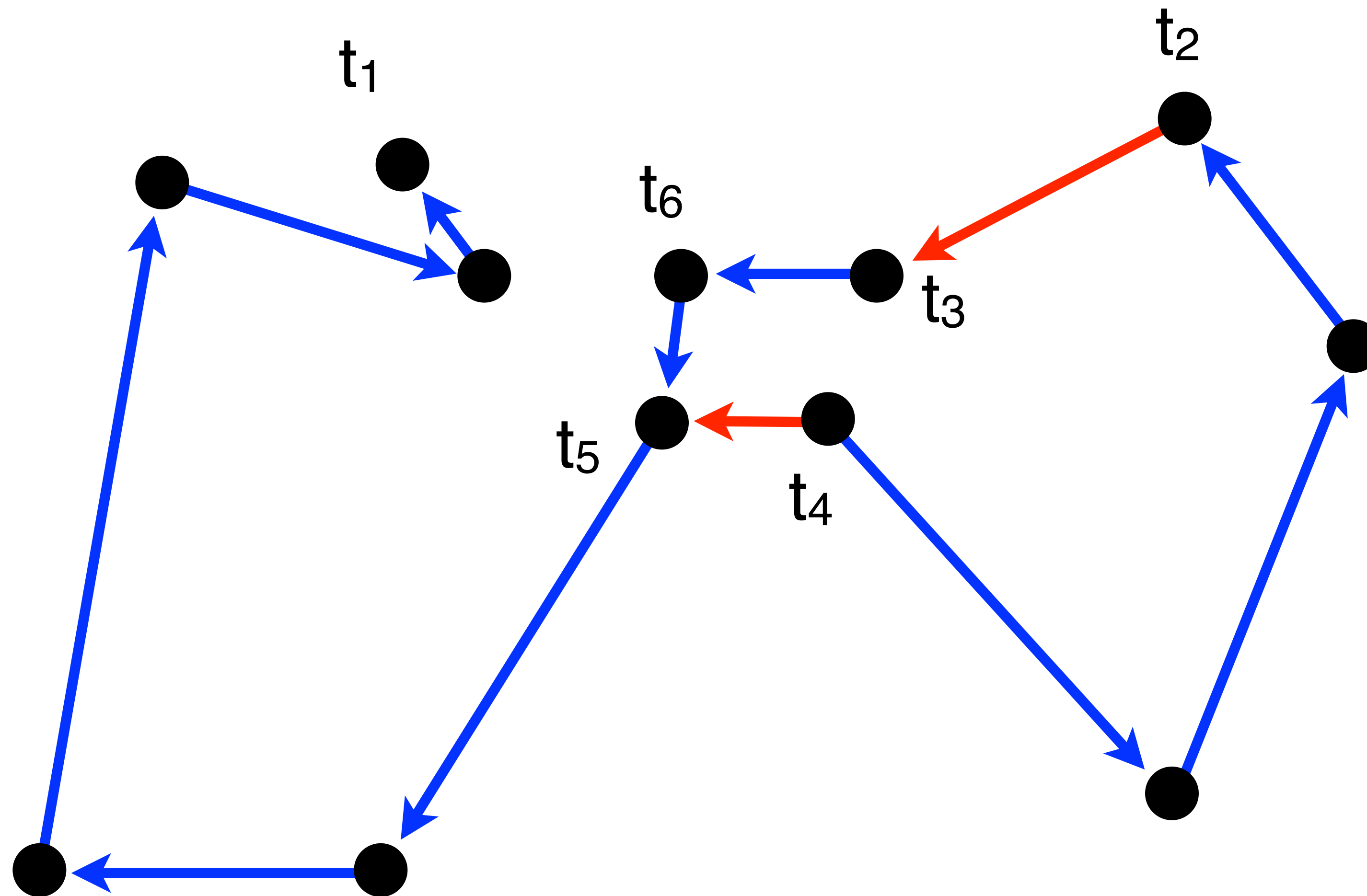
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- else we have a solution by removing the edge (t_4, t_3) and connecting (t_1, t_4)
- compute the cost but do not connect
- instead restart with t_1 and its (pretended) edge (t_1, t_4)

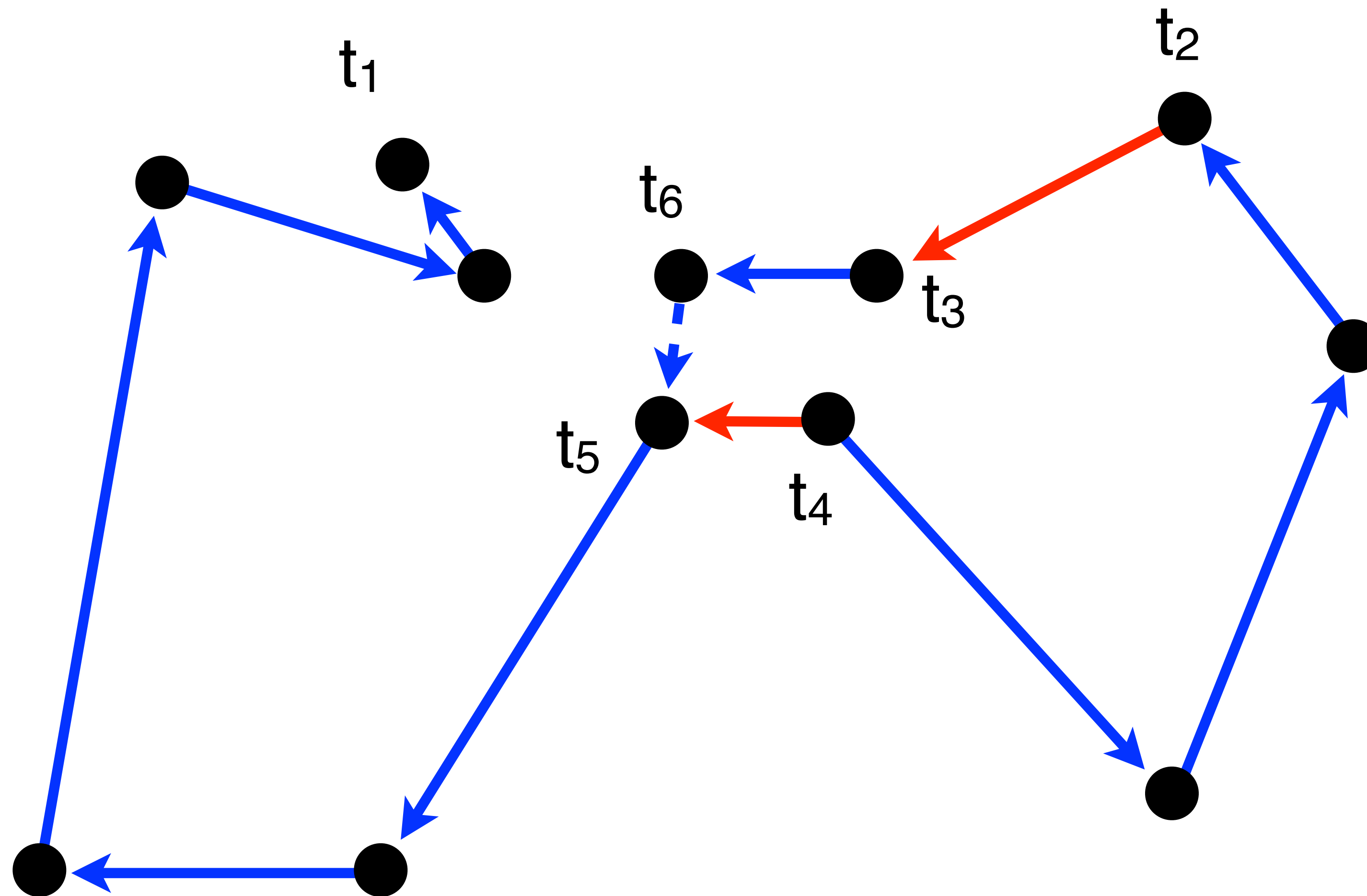
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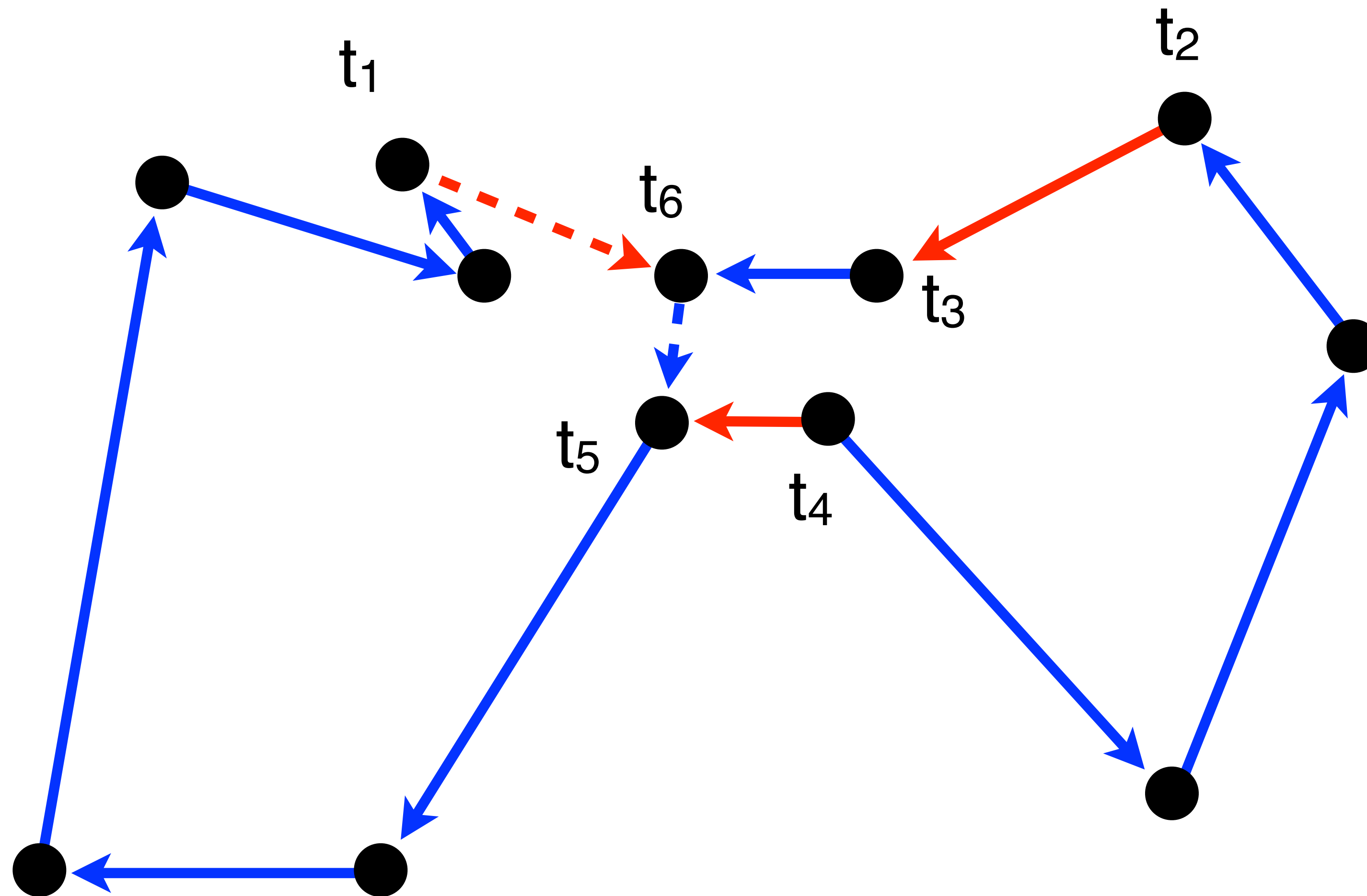
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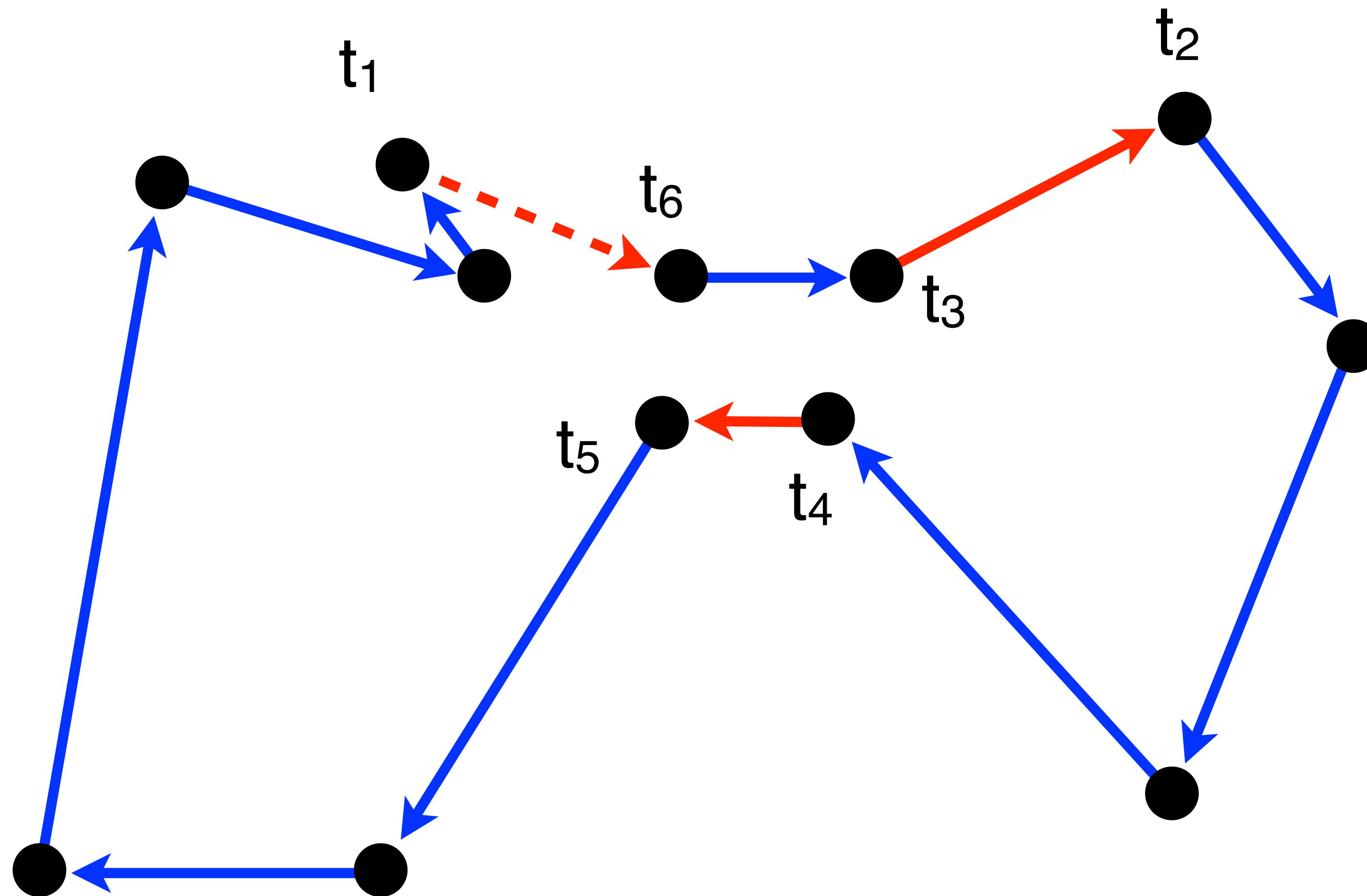
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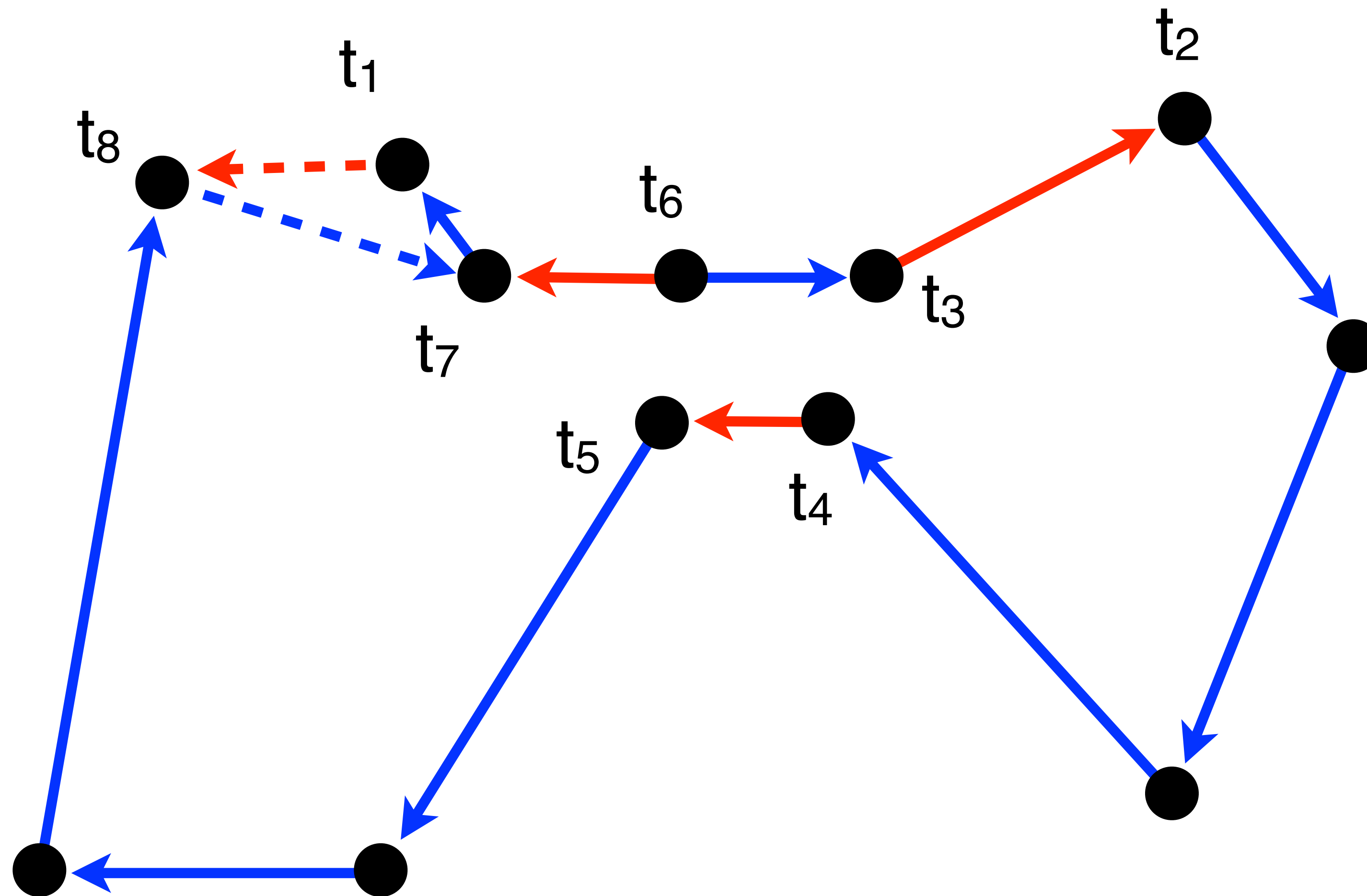
► K-OPT

- choose a vertex t_1 and its edge $y_1 = (t_1, t_4)$
- choose an edge $x_2 = (t_4, t_5)$ with $d(y_2) < d(y_1)$
- if none exist, restart with another vertex
- else we have a solution by removing the edge (t_6, t_5) and connecting (t_1, t_6)
- compute the cost but do not connect
- instead restart with t_1 and its (pretended) edge (t_1, t_6)

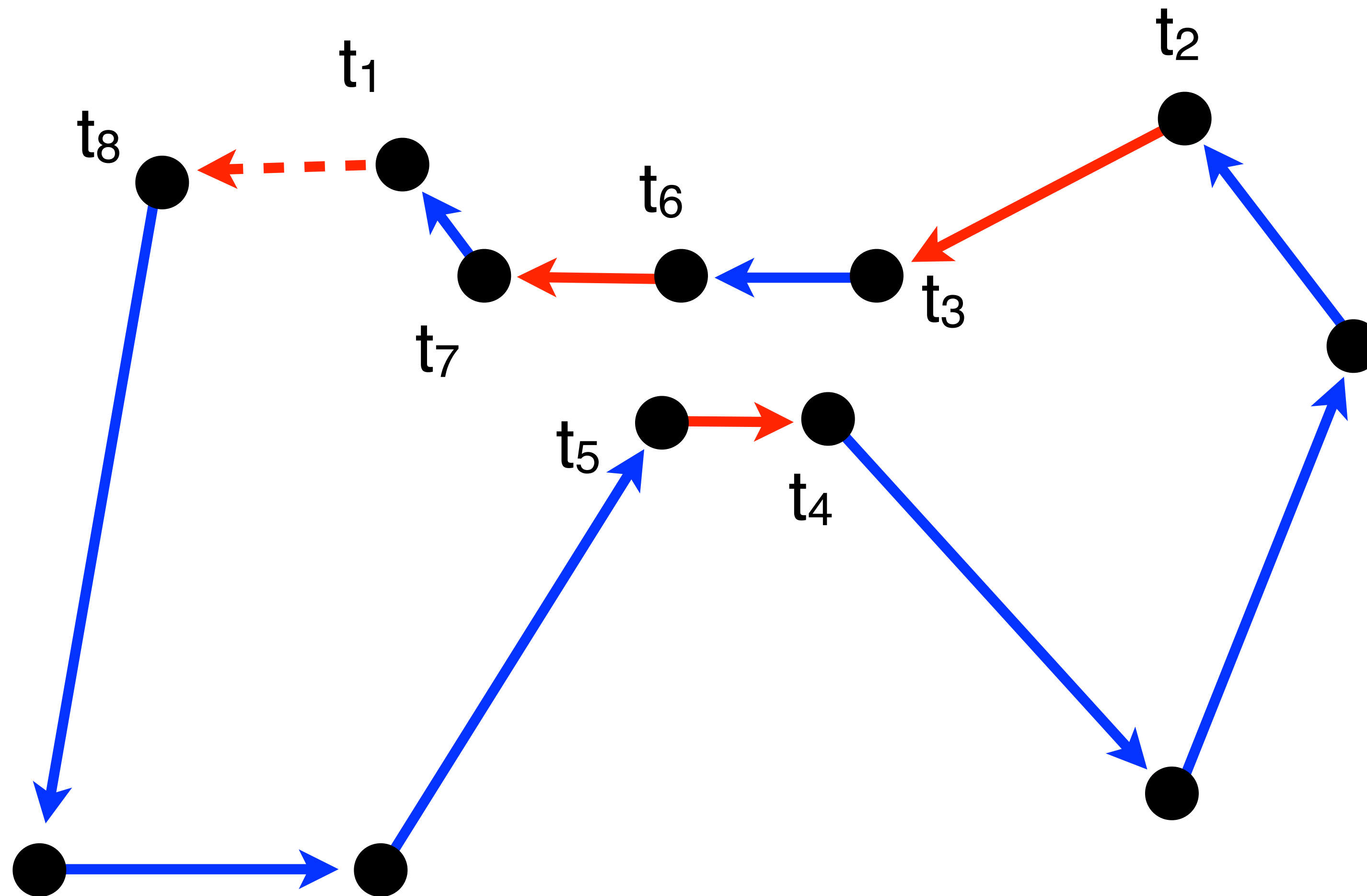
K-OPT (first iteration)



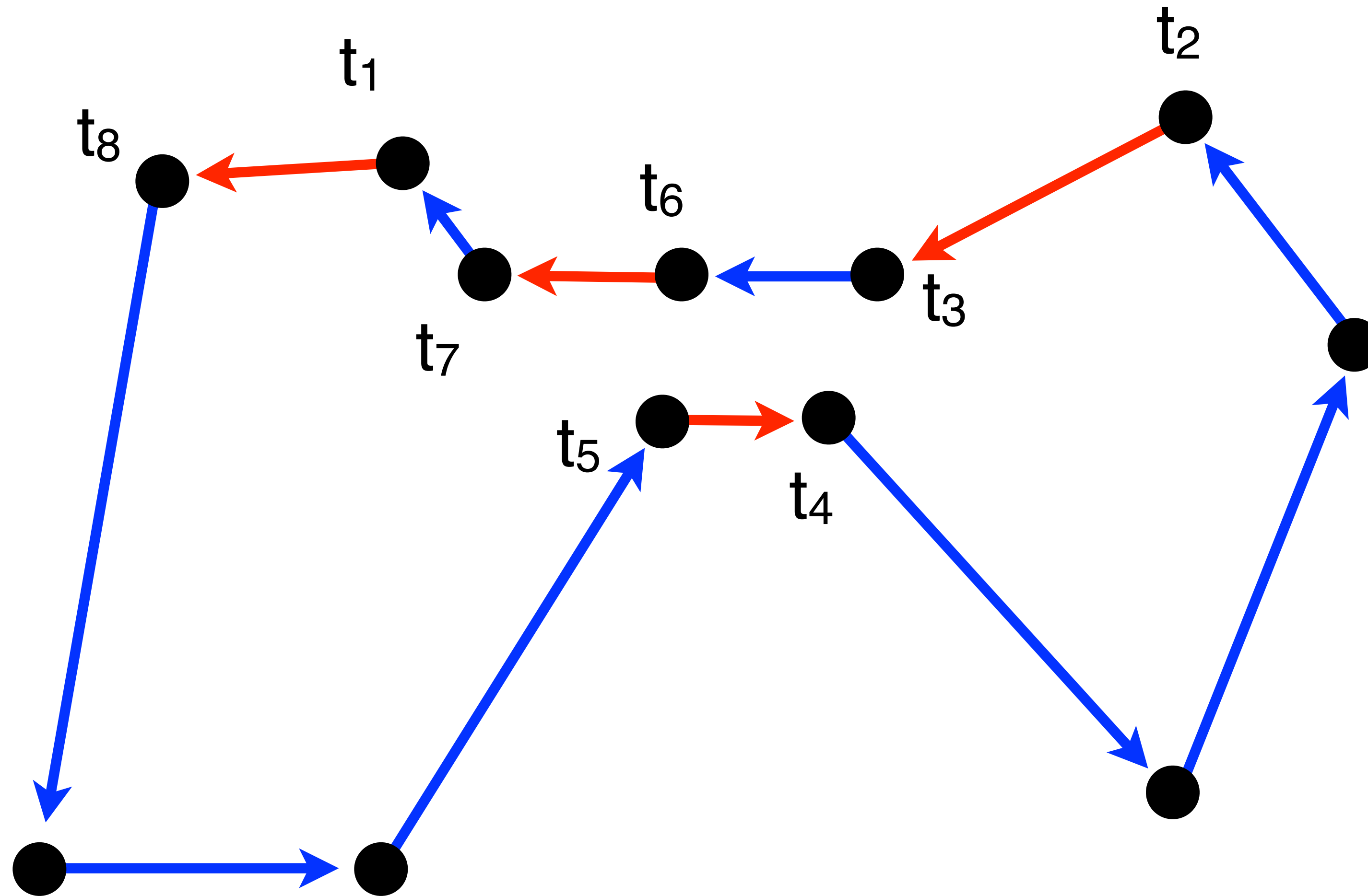
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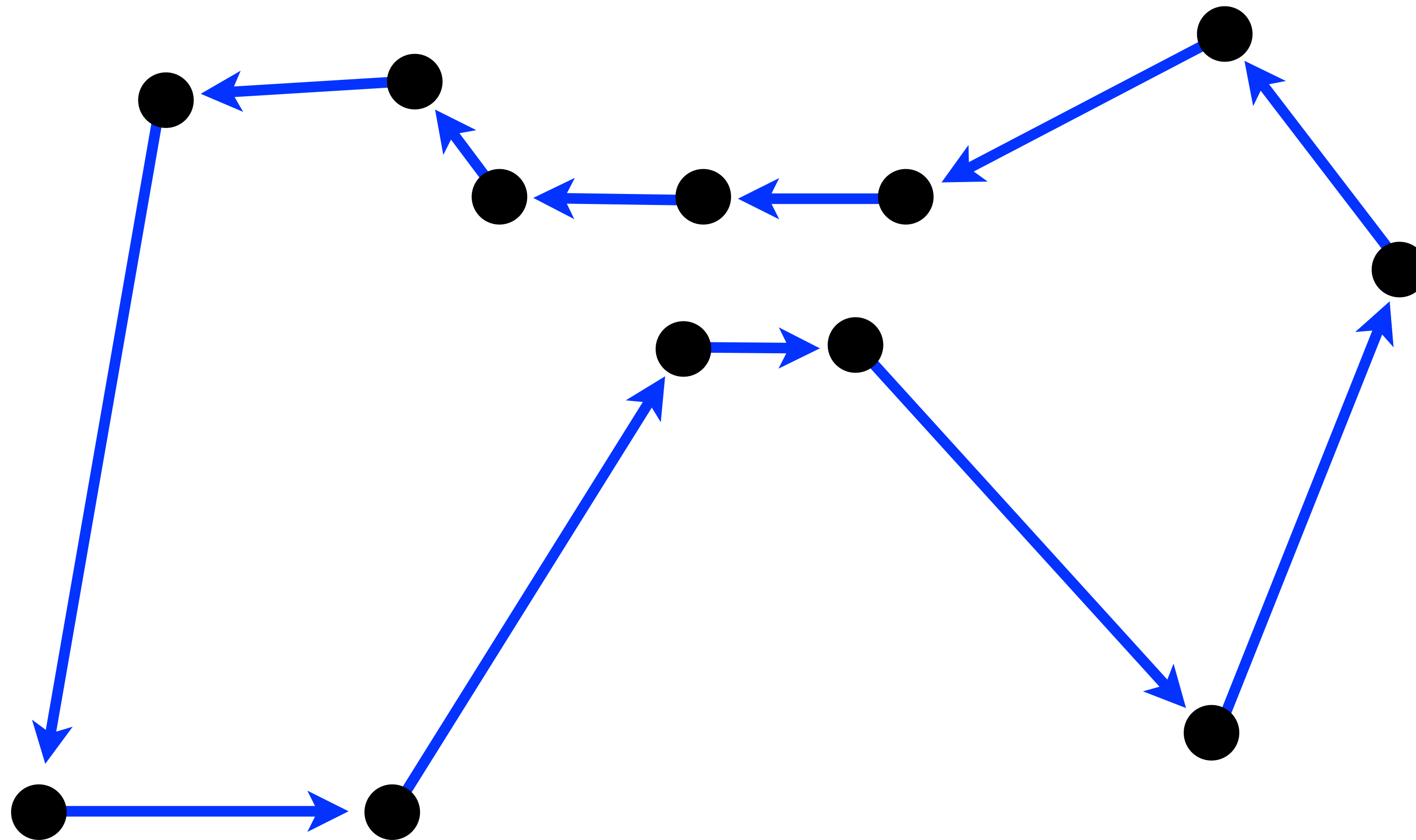
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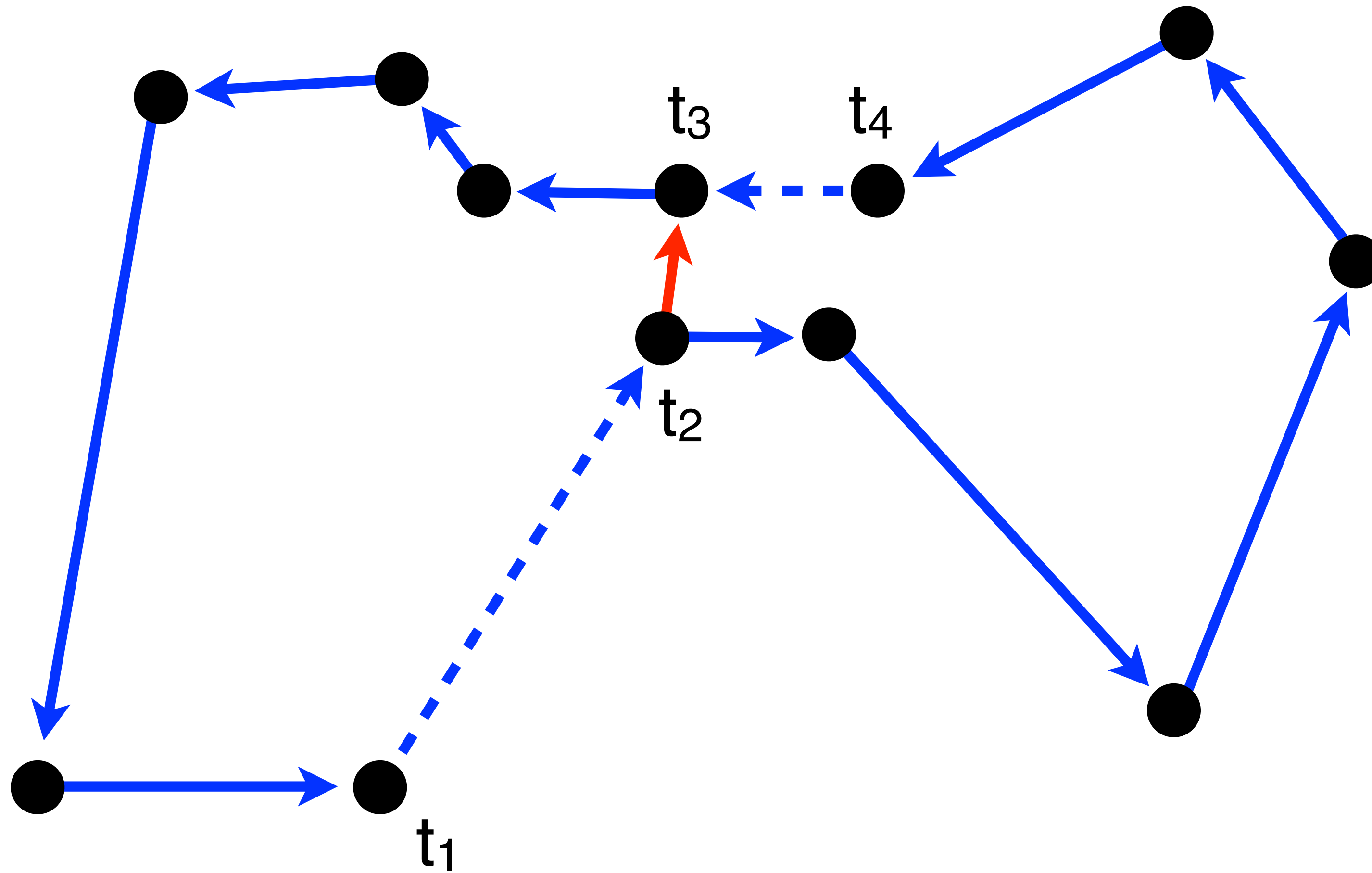
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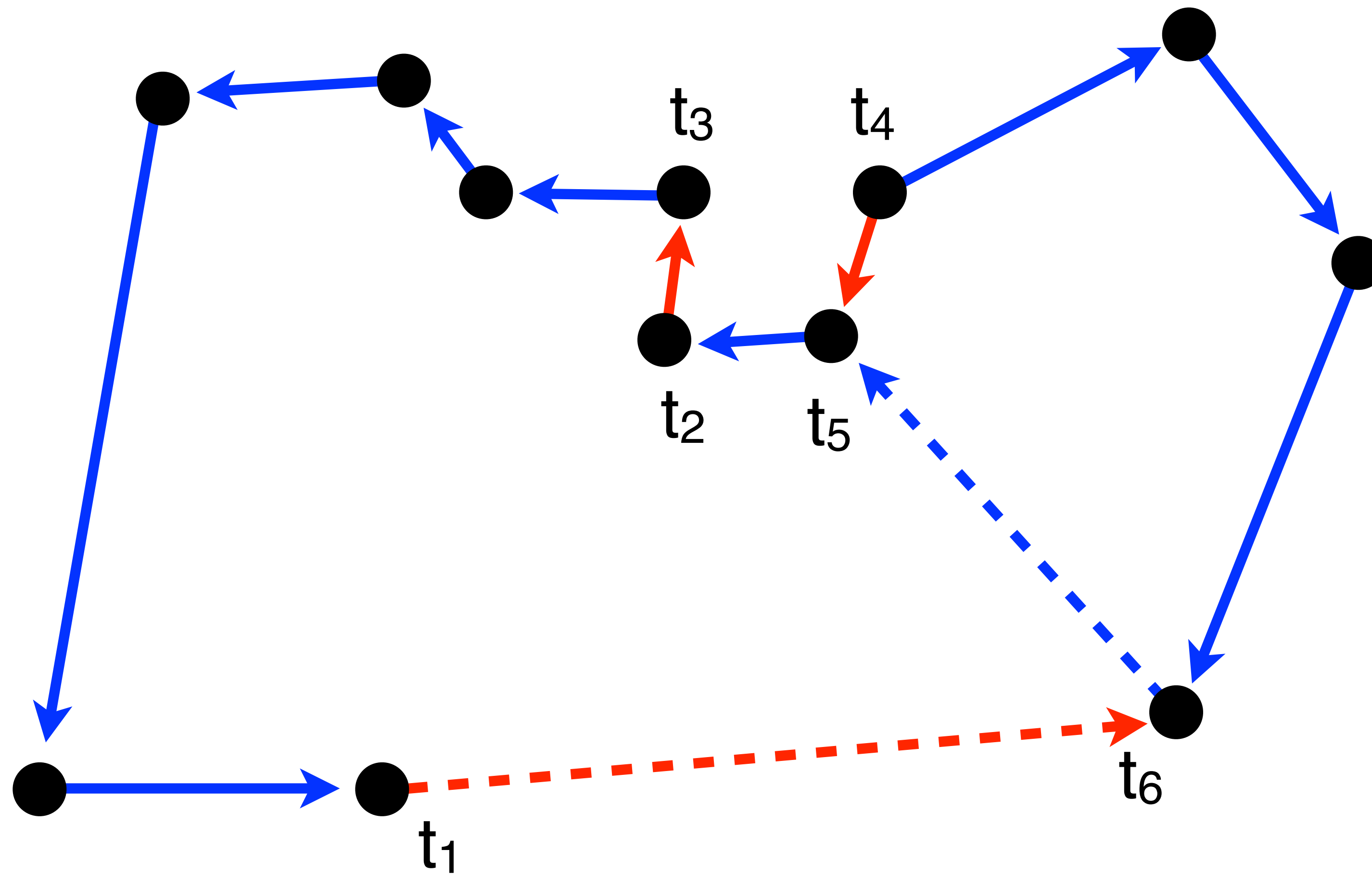
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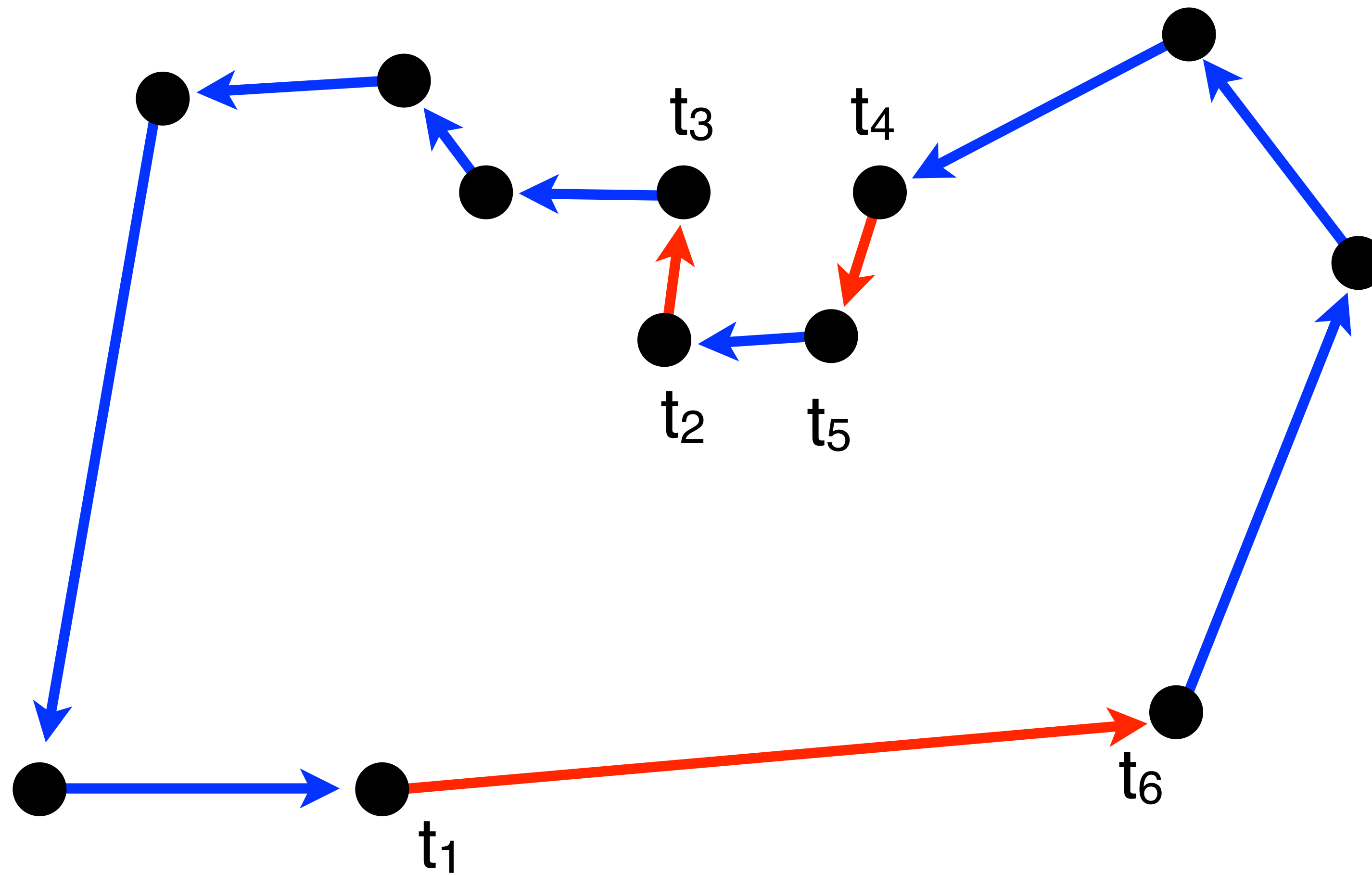
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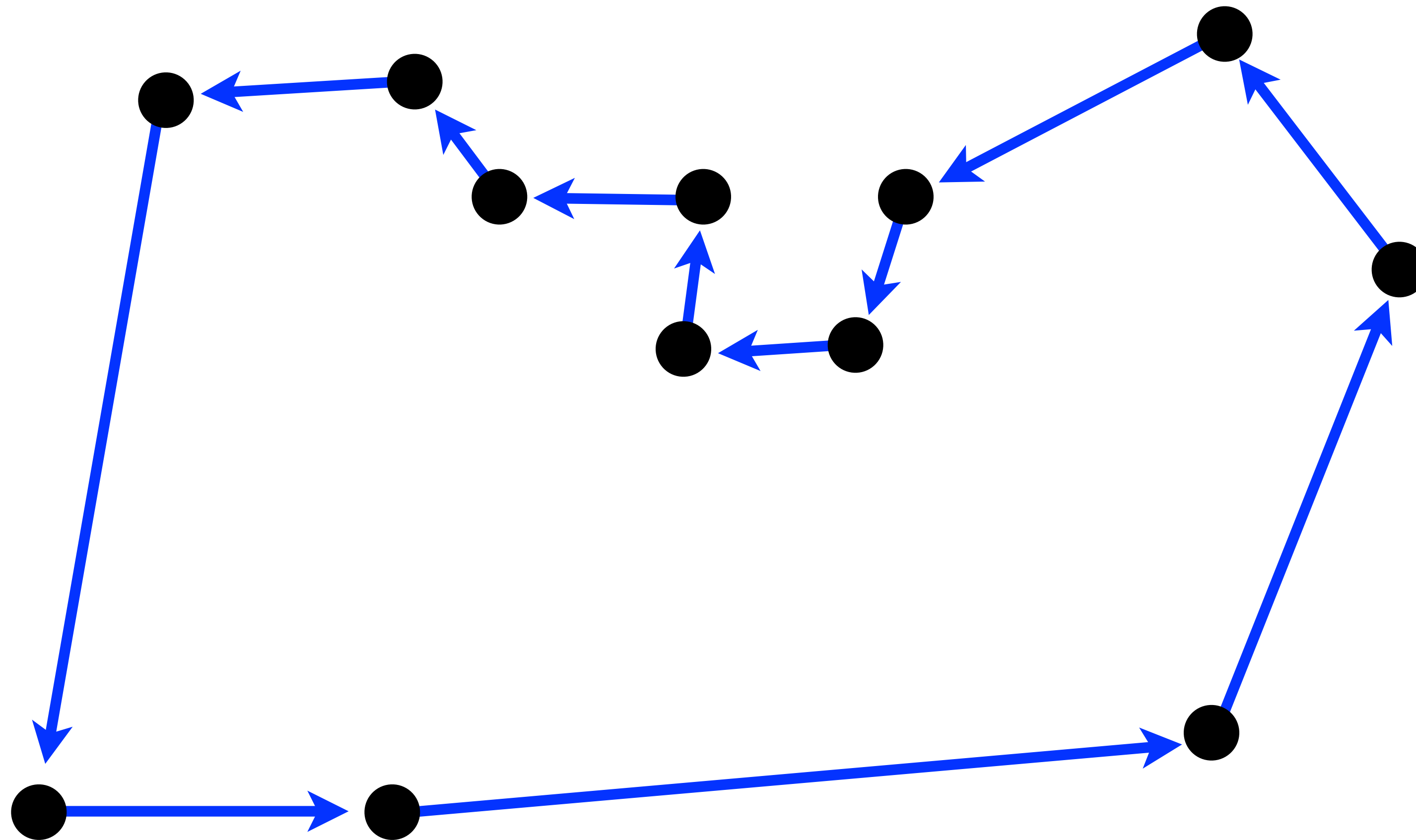
K-OPT (second iteration)



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K-OPT (second iteration)



Until Next Time