

# Автономная навигация мобильных роботов

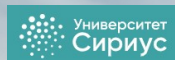
A. Matveev

almat1712@yahoo.com

Department of Mathematics and Mechanics,  
Saint Petersburg state University,



Scientific and Technological  
University "Sirius"



## Тема 3: Алгоритмы навигации на основе обширных данных о сцене

# Общие определения и примеры

## Map

An abstracted representation of the salient features of the environment in which the robot moves. Should identify

- the zones where the robot can move and
- the zones (obstacles) where the robot cannot do so

Decomposition of the environment

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  - rectangular worlds
  - spherical worlds
  - polygonal worlds
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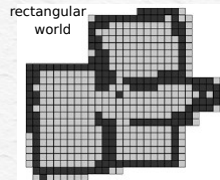
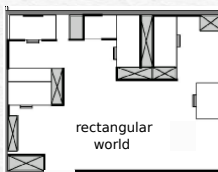
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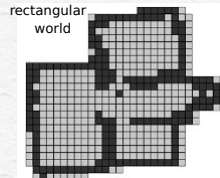
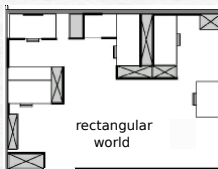
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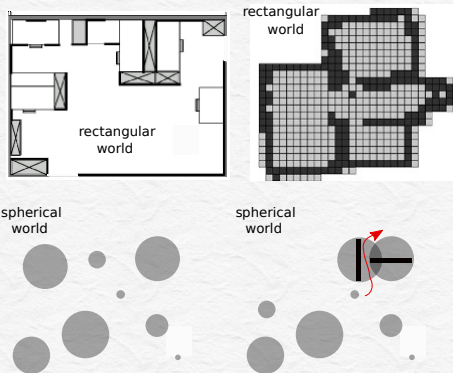
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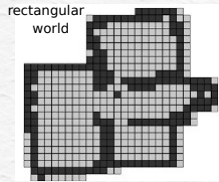
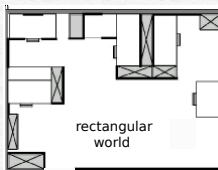
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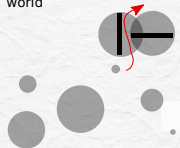
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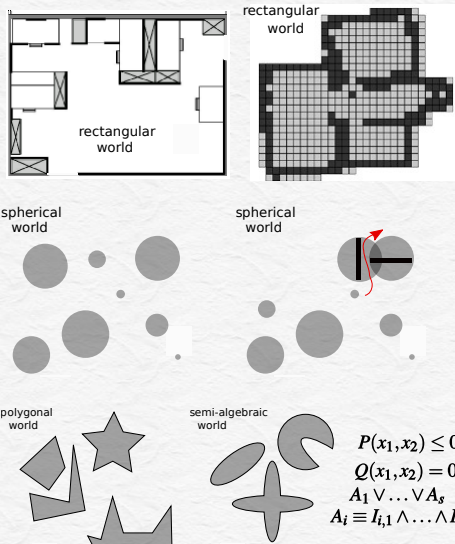
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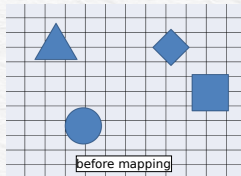
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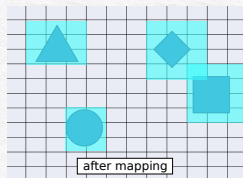
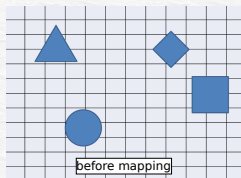
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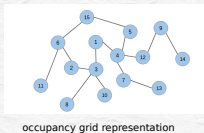
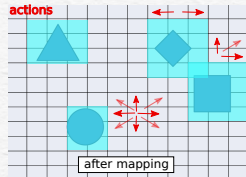
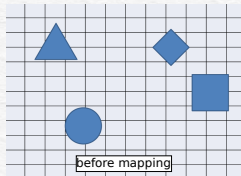
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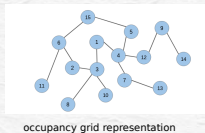
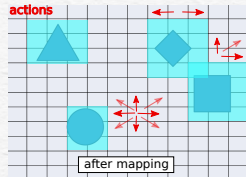
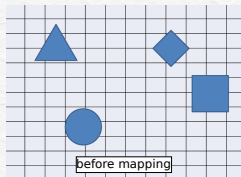
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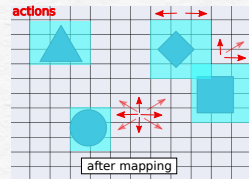
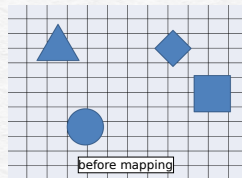
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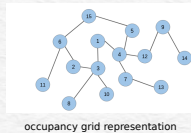
- Continuous geometric representations
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Nodes  $\sim$  free cells;  
two nodes are linked  $\sim$  the  
robot can immediately move  
from any of them to the other,  
and “knows” how

Target point  $\sim$  target node  
Path planning  $\sim$  finding a path  
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Sequence of way-points (“way-  
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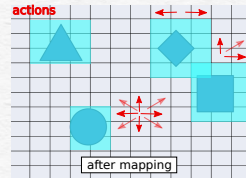
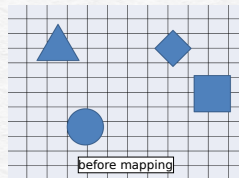
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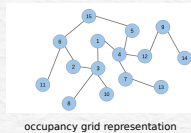
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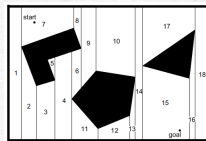
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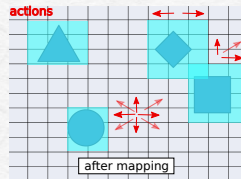
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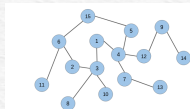
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occupancy grid representation

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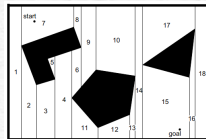
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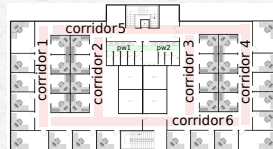
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cell decomposition

34 workplaces  
8 toilet cabins

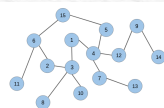


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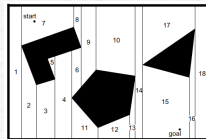
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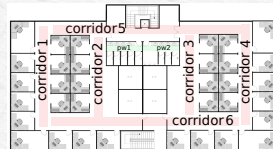
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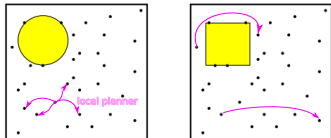
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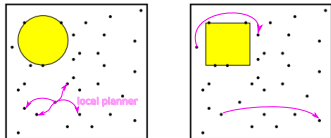
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Nodes  $\sim$  sample points;  
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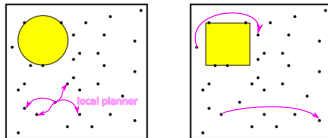
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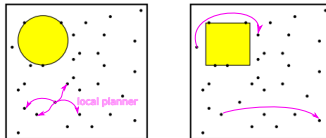
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# Разбиение (мозаика) Вороного

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## Definition

Let  $X$  be a metric space with the distance function  $d(\cdot, \cdot)$  and let  $\{x_1, \dots, x_n\} \subset X$  be a finite set of points. The Voronoi cell with the center  $x_i$  is defined to be  $C_i := \{x \in X : d(x, x_i) < d(x, x_j) \forall j \neq i\}$ .

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## Basic properties

- $C_i \cap C_j = \emptyset \forall i \neq j$
- $x \notin \bigcup_{i=1}^n C_i \Rightarrow x \in H_U := \bigcup_{i \neq j} H_{i,j}$ ,  
where  
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- $\{C_i\}_{i=1}^n$  partition  $X$  "up to"  $H_U$

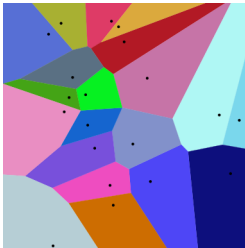
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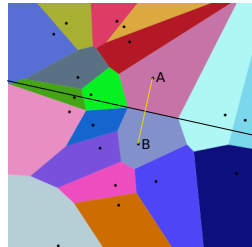
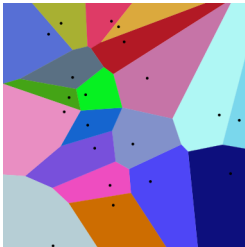
# Разбиение (мозаика) Вороного

## Definition

Let  $X$  be a metric space with the distance function  $d(\cdot, \cdot)$  and let  $\{x_1, \dots, x_n\} \subset X$  be a finite set of points. The **Voronoi cell with the center  $x_i$**  is defined to be  $C_i := \{x \in X : d(x, x_i) < d(x, x_j) \forall j \neq i\}$ .

## Basic properties

- $C_i \cap C_j = \emptyset \forall i \neq j$
- $x \notin \bigcup_{i=1}^n C_i \Rightarrow x \in H_U := \bigcup_{i \neq j} H_{i,j}$ ,  
where  
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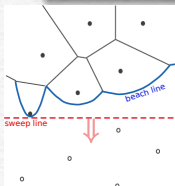
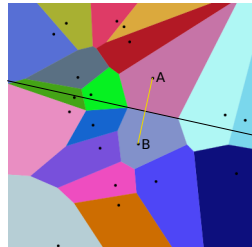
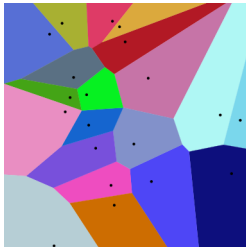
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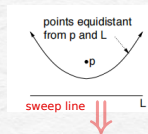
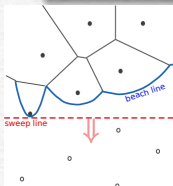
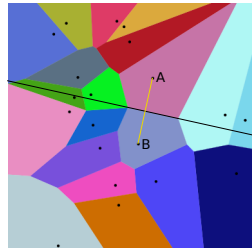
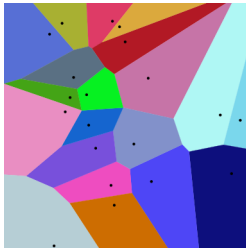
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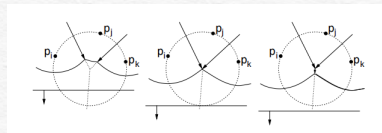
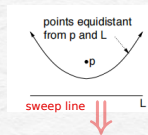
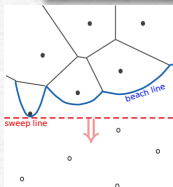
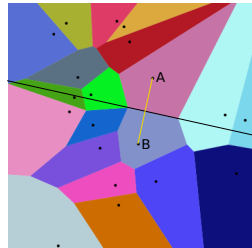
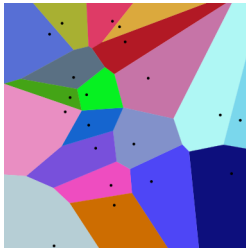
# Разбиение (мозаика) Вороного

## Definition

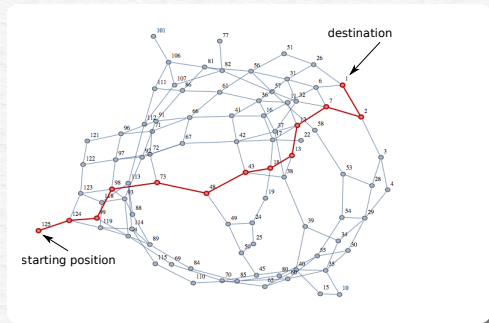
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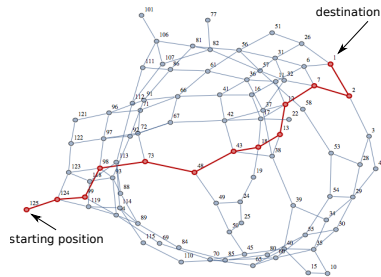


# Планирование: подходы, основанные на логическом представлении



# Планирование: подходы, основанные на логическом представлении

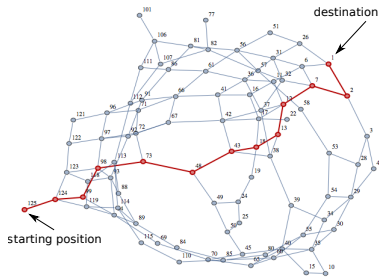
## Walk over nodes: terminology



# Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

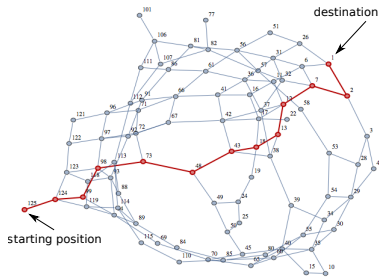
- Visited/unvisited node



# Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

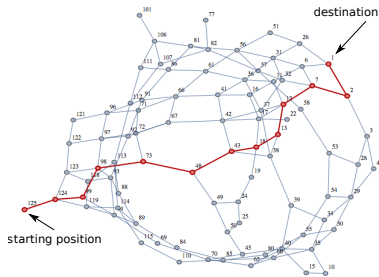
- Visited/unvisited node
- Visited dead/alive node



# Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

- Visited/unvisited node
- Visited dead/alive node
- Priority queue  $Q$  of the alive nodes, the dead ones are excluded

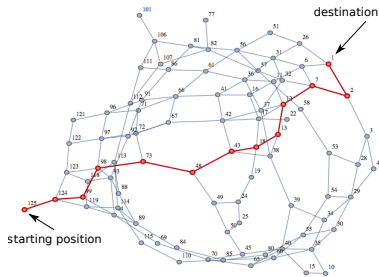




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## Walk over nodes: terminology

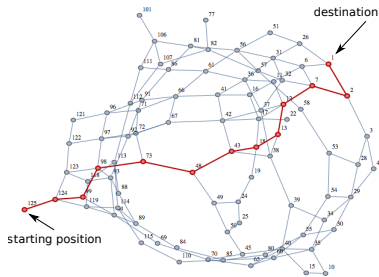
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## Walk over nodes: terminology

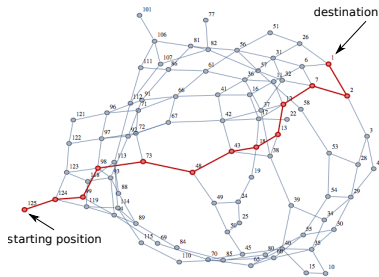
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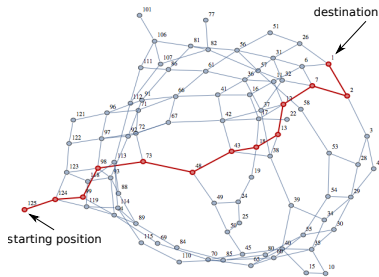
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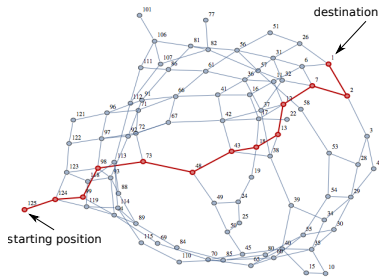
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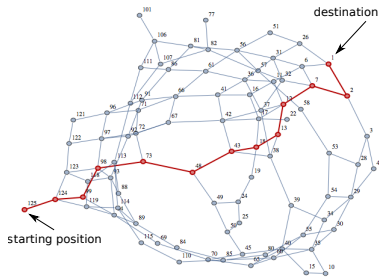
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## Breadth first (grass-fire front) method

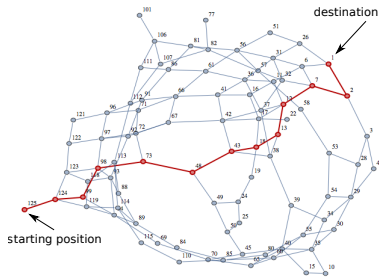
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## Breadth first (grass-fire front) method

- Start with the source, put in in  $Q$



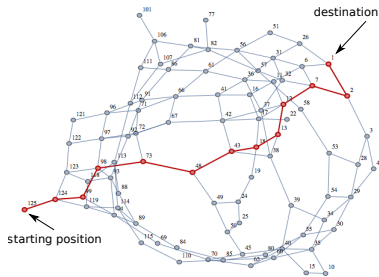
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## Breadth first (grass-fire front) method

- Start with the source, put in in  $Q$
- Sort  $Q$  according to FIFO (first-in-first-out) policy





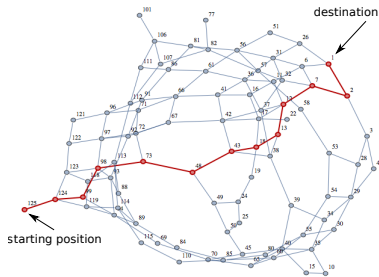
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## Breadth first (grass-fire front) method

- Start with the source, put in in  $Q$
- Sort  $Q$  according to FIFO (first-in-first-out) policy
- Test only unvisited neighbors



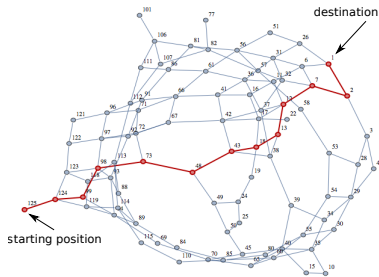
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## Breadth first (grass-fire front) method

- Start with the source, put in in  $Q$
- Sort  $Q$  according to FIFO (first-in-first-out) policy
- Test only unvisited neighbors
- When arriving at the destination, terminate processing nodes



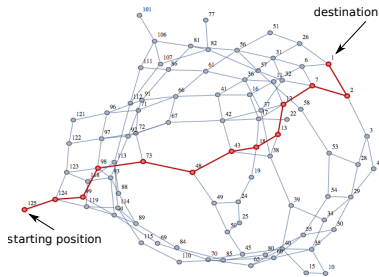
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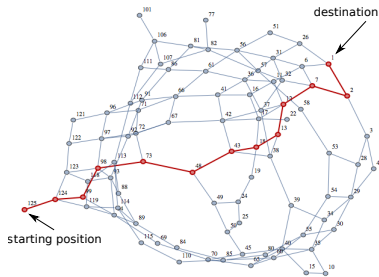
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## Properties

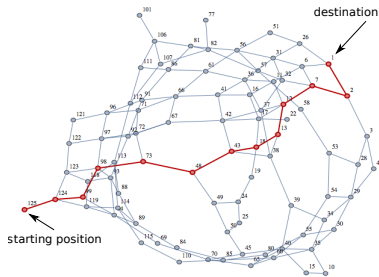
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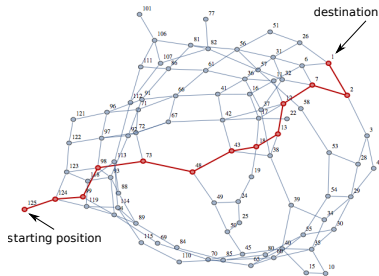
## Properties

- Degree of the node := the length of the shortest path from the source

# Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

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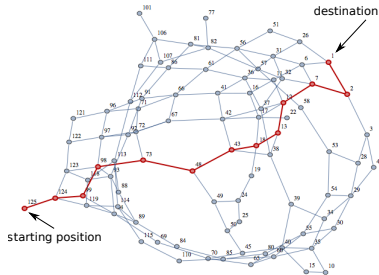
## Properties

- Degree of the node := the length of the shortest path from the source
- for any step of the search stage, there exists  $k \geq 0$  such that  $Q$  contains only nodes of degree  $k$  or  $k + 1$  and all nodes of degree  $< k$  are dead

# Планирование: подходы, основанные на логическом представлении

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- References highlight a shortest path



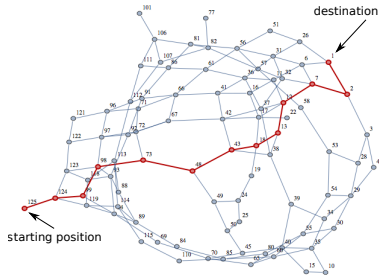




## Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

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## Depth first (aggressive deepening) method

- Start with the source, put in in  $Q$
- Sort  $Q$  according to LIFO (last-in-first-out) policy
- Test only unvisited neighbors
- When arriving at the destination, terminate processing nodes
- Use the references to build a path

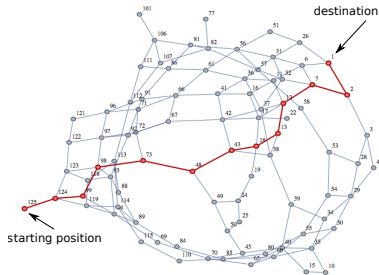
## Properties

- **Degree of the node** := the length of the shortest path from the source
- for any step of the search stage, there exists  $k \geq 0$  such that  $Q$  contains only nodes of degree  $k$  or  $k + 1$  and all nodes of degree  $< k$  are dead
- References highlight a shortest path
- $O(|V| + |E|)$

# Планирование: подходы, основанные на логическом представлении

## Walk over nodes: terminology

- Visited/unvisited node
- Visited dead/alive node
- Priority queue  $Q$  of the alive nodes, the dead ones are excluded
- Prioritization method  $\sim$  to sort  $Q$
- Direct walk (search) starts from the source and aims at the destination
- Reference  $\sim$  indication to a parent
- Mark "dead"
- Use the references to find the path (finalizing backward walk)



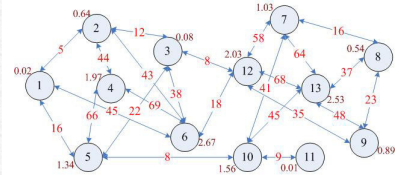
## Depth first (aggressive deepening) method

- Start with the source, put in in  $Q$
- Sort  $Q$  according to LIFO (last-in-first-out) policy
- Test only unvisited neighbors
- When arriving at the destination, terminate processing nodes
- Use the references to build a path

## Properties

- Degree of the node := the length of the shortest path from the source
- Finds the destination
- References highlight a path
- $O(|V| + |E|)$

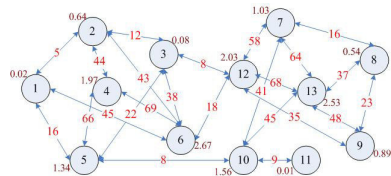
# Планирование: подходы, основанные на логическом представлении



$W(n \rightarrow n') > 0$  weight of the edge

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Dijkstra's algorithm

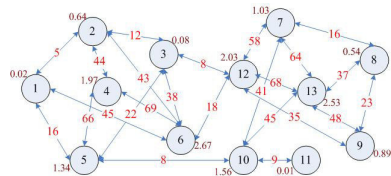


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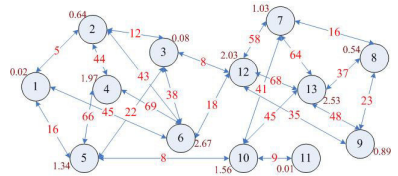


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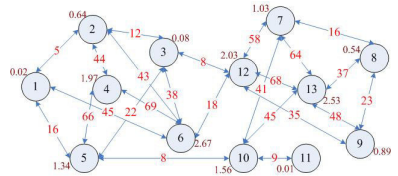


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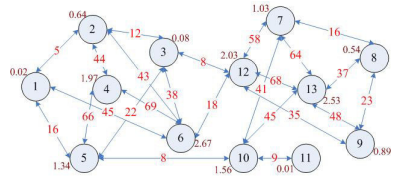

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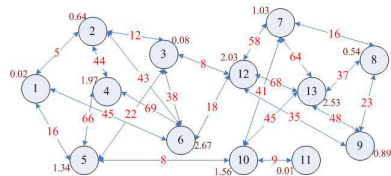


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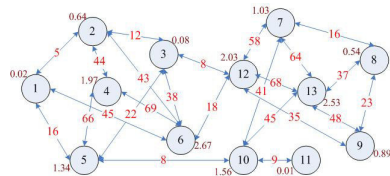


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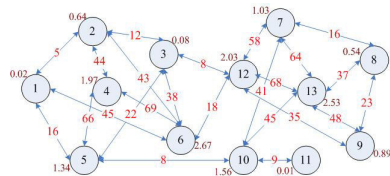


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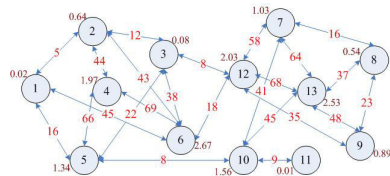


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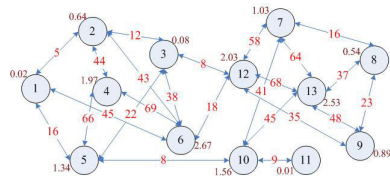


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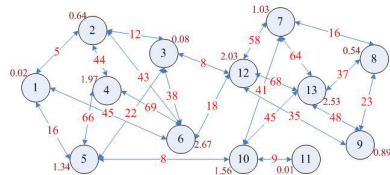


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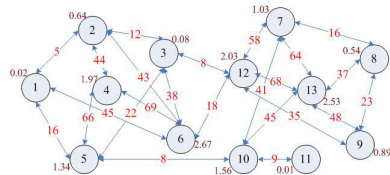
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Proof:

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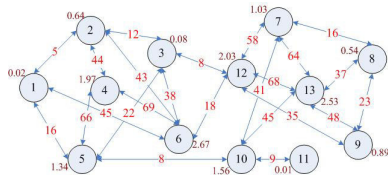
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**Proof:** Induction on the number  $N$  of visited nodes.

### Lemma

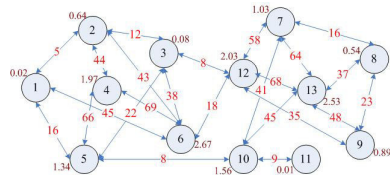
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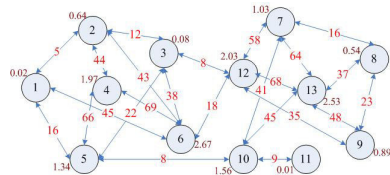
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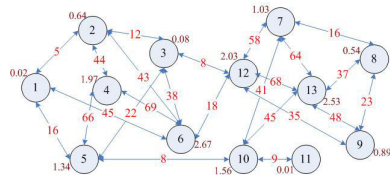
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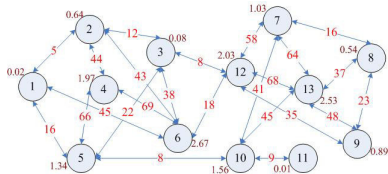
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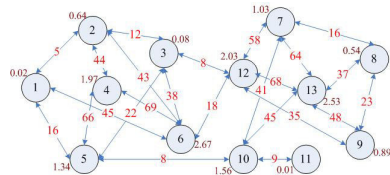
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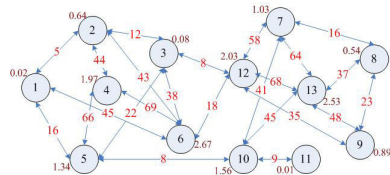
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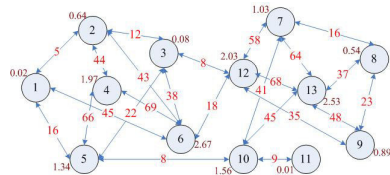
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- Initially,  $V := \emptyset$ ,  $U$  contains all nodes,  $L(s) := 0$ ,  $L(n) := \infty \forall n \neq s$
- At any step, do the following
  - find a minimizer of  $L(n)$  over  $n \in U$
  - run through the set of all unvisited nodes  $n'$  such that  $n \rightarrow n'$  is an edge (the set of nearest unvisited descendants)
  - for any such  $n'$ , put  $L(n') := \min \{L(n'); L(n) + W(n \rightarrow n')\}$
  - when the run is completed, remove  $n$  from  $U$  to  $V$
- terminate the algorithm whenever either  $U = \emptyset$  or  $L(n) = \infty \forall n \in U$

## Lemma

At any step and for the node visited at this step, its label  $L(n)$  is the shortest length  $d(n)$  of the path from the source  $s$  to  $n$ . The final set  $U$  is the set of the nodes unreachable from  $s$ .



$W(n \rightarrow n') > 0$  weight of the edge

**Proof:** Induction on the number  $N$  of visited nodes. For  $N = 0$  is trivial. Let the claim be true for some  $N$  and one more node  $n_*$  is visited. Then  $L(n_*) \geq d(n_*)$ . Consider the shortest path from the source  $s$  to this node  $n_*$ . Let  $n_u$  be the first unvisited node on this path. It is preceded by a visited node  $n_v$ .

$d(n_u) = d(n_v) + W(n_v \rightarrow n_u)$ . Meanwhile,  $L(n_v) = d(n_v)$  and so  $L(n_u) \leq L(n_v) + W(n_v \rightarrow n_u) = d(n_v) + W(n_v \rightarrow n_u) = d(n_u)$ .

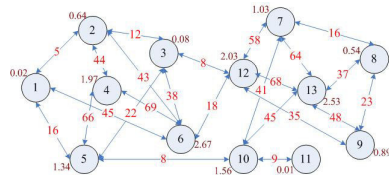
# Планирование: подходы, основанные на логическом представлении

## Dijkstra's algorithm

- At any step, forms a partition of the nodes into  $V$  (visited) and  $U$  (unvisited)
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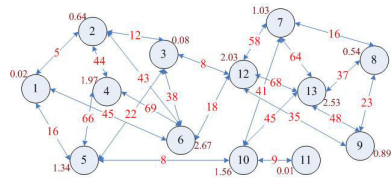


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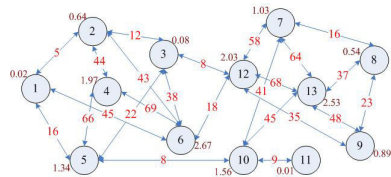
# Планирование: подходы, основанные на логическом представлении



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# Планирование: подходы, основанные на логическом представлении

$A^*$  -algorithm

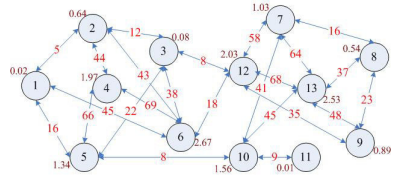


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# Планирование: подходы, основанные на логическом представлении

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●  $G(n)$  – the minimum weight of the path from  $n$  to the destination (cost-to-go)

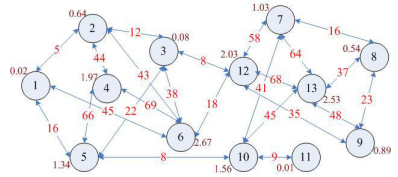


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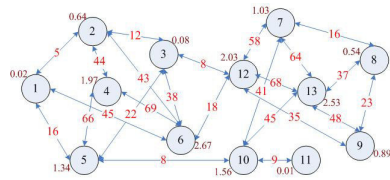


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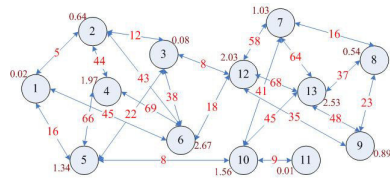


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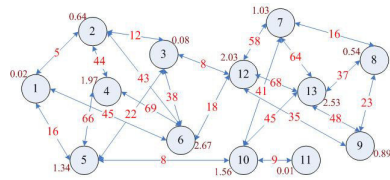
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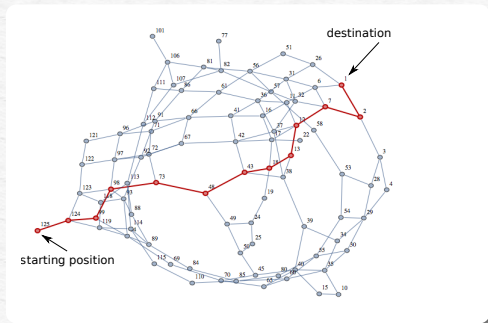


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## Lemma

At any step and for any node, its label  $L(n)$  is the shortest length  $d(n)$  of the path among those that go from the source  $s$  to  $n$  through  $V$  (except for the last node). (The minimum over the empty set is defined to be  $\infty$ .) The final set  $U$  is the set of the nodes unreachable from  $s$ .

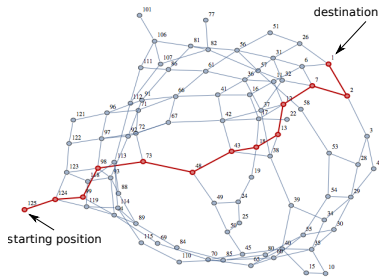
# Планирование: подходы, основанные на логическом представлении



## Планирование: подходы, основанные на логическом представлении

## Best first

- A numerical evaluation  $G_*(n)$  of “successfulness” of node  $n$  (e.g., cost-to-go)
  - Start with the source, put it in  $Q$
  - Sort  $Q$  in the decreasing value of  $G_*(n)$
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  - When arriving at the destination, terminate processing nodes
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# Планирование: подходы, основанные на логическом представлении

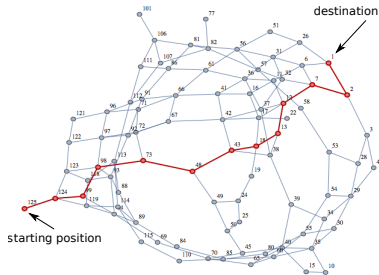
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## Iterative deepening

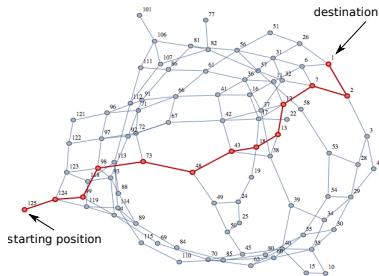
- 1 Put  $d := 1$
- 2 Perform the depth-first search, with limiting it to nodes of degree  $\leq d$
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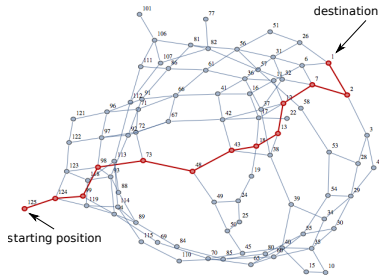
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## Bidirectional search

A forward and backward search methods are run simultaneously until meeting.





Search as evolution of a dynamical system



### Search as evolution of a dynamical system

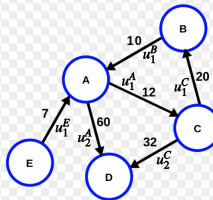
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- States  $x$ , nodes of the graph
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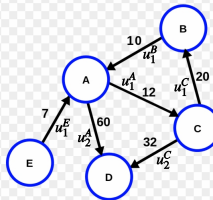
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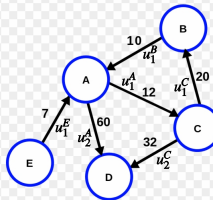
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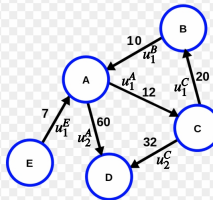
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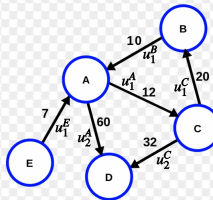
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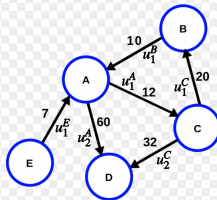
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$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

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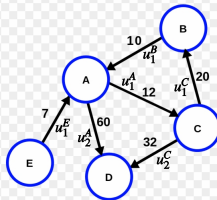
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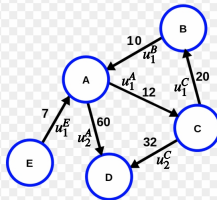
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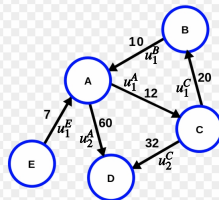
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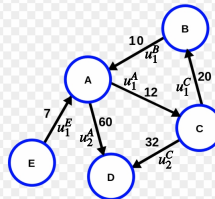
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It is needed to

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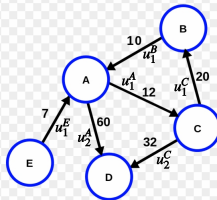
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- arrive at  $b$ :  $\eta(x) := \lambda \|x - b\|, \lambda \approx \infty$

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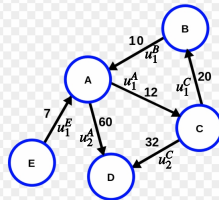
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### Search as evolution of a dynamical system

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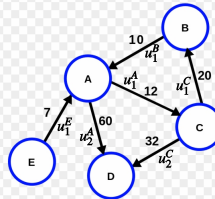
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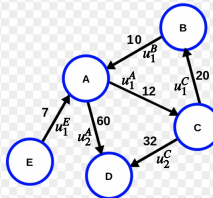
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- arrive at some of  $b_1, \dots, b_s$ :  $\eta(x) := \lambda \min_{i=1, \dots, s} \|x - b_i\|, \lambda \approx \infty$

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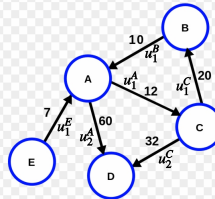
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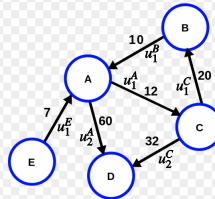
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Permits us to compute

$$V_k(\cdot) \curvearrowright V_{k-1}(\cdot) \curvearrowright \dots \curvearrowright V_0(\cdot), \quad V_t(a) \in [0, \infty);$$

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## Theorem

The following statements hold:

- The Bellman function is the unique solution of the Bellman equation
- $u_t \in \mathcal{U}_t(x_t)$  is the optimal control law (search rule)

# Proof of the theorem



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Given  $t_0$  and  $\mathbf{a} \in \mathbb{R}^n$ , we build a process from  $\mathfrak{M}(t_0, \mathbf{a})$  as follows:

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Process optimal for  $(t_0, \mathbf{a}) \xrightarrow{\text{disintegrate}} \text{the first term}(t_0, \mathbf{a}, u_{t_0}^0)$  and the remainder that corresponds to  $t \geq t_0 + 1$

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$$\text{Overall: } V_{t_0}(\mathbf{a}) = \min_{u \in U(\mathbf{a})} \left\{ \varphi(\mathbf{a}, u) + V_{t_0+1}[f(\mathbf{a}, u)] \right\}, \quad u_{t_0}^0 = \operatorname{argmin}_{u \in U(\mathbf{a})} \left\{ \dots \right\} \Rightarrow u_{t_0}^0 \in \mathcal{U}_{t_0}[x_{t_0}^0] \Rightarrow u_t^0 \in \mathcal{U}_{t_0}[x_t^0] \quad \forall t \geq t_0.$$



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$$V_{t+1}[\underbrace{f(x_t, u_t)}_{x_{t+1}}] + \varphi(x_t, u_t) = \min_{u \in U(x_t)} \{V_{t+1}[f(x_t, u)] + \varphi(x_t, u)\} = V_t(x_t)$$

$$\begin{aligned} I_{t_0}(p) &= \sum_{t \in [t_0:k)} \varphi[x_t, u_t] = \sum_{t=t_0}^{k-1} \varphi[x_t, u_t] + V_k[x_k] = \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \left\{ \varphi_{k-1}[x_{k-1}, u_{k-1}] + V_k[x_k] \right\} \\ &\stackrel{\tau:=k-1}{=} \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \left\{ \underbrace{\varphi[x_\tau]}_{=:a_*}, \underbrace{u_\tau}_{=:v \in U(a_*)} \right\} + V_{\tau+1}[f(a_*, v)] \end{aligned}$$

$$\text{Belmann equation} \stackrel{=}{=} \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \underbrace{V_\tau(a_*)}_{=:V_{k-1}(x_{k-1})} = \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + V_{k-1}(x_{k-1}) = \dots = V_{t_0}[x_{t_0}] = V_{t_0}(a)$$

# Proof of the theorem

Let the process  $p = [\{x_t\}_{t=t_0}^{t_1}, \{u_t\}_{t=t_0}^{t_1-1}] \in \mathfrak{M}(t_0, a)$  be generated by the dynamic programming regulator  $u_t \in \mathcal{U}_t(x_t) = \text{Arg min}_{u \in U(x_t)} \{V_{t+1}[f(x_t, u)] + \varphi(x_t, u)\}$

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$$I_{t_0}(p) = \sum_{t \in [t_0:k)} \varphi[x_t, u_t] = \sum_{t=t_0}^{k-1} \varphi[x_t, u_t] + V_k[x_k] = \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \left\{ \varphi_{k-1}[x_{k-1}, u_{k-1}] + V_k[x_k] \right\}$$

$$\stackrel{\tau:=k-1}{=} \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \left\{ \underbrace{\varphi[x_\tau]}_{=:a_*}, \underbrace{u_\tau}_{=:v \in U(a_*)} \right\} + V_{\tau+1}[f(a_*, v)]$$

$$\stackrel{\text{Bellmann equation}}{=} \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + \underbrace{V_\tau(a_*)}_{=:V_{k-1}(x_{k-1})} = \sum_{t=t_0}^{k-2} \varphi[x_t, u_t] + V_{k-1}(x_{k-1}) = \dots = V_{t_0}[x_{t_0}] = V_{t_0}(a)$$

$$V_{t_0}(a) = I_{t_0}(p)$$

# Оптимальное планирование с нефиксированной длиной пути



# Оптимальное планирование с нефиксированной длиной пути

## Problem statement

$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

$$x_0 = a, \quad k \text{ is freely manipulable} \quad (+)$$

$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\infty)$$

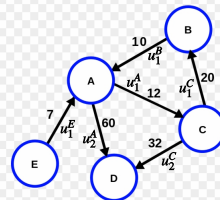
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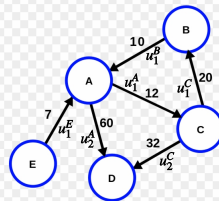
$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\odot)$$

## Bellman equation

$$V_t(a) = \min_{u \in U(a)} \{ V_{t+1}[f(a, u)] + \varphi(a, u) \} \quad \forall a$$

$$\text{Boundary condition: } V_k(\cdot) := \eta(\cdot)$$

$$\text{Regulator } \mathcal{U}_t(a) := \text{Arg min}_{u \in U(a)} \{ V_{t+1}[f(a, u)] + \varphi(a, u) \}$$



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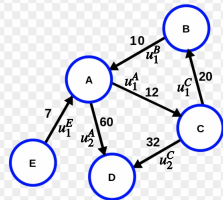
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## The same as an iteration

$$V(\cdot) \xrightarrow{\mathfrak{F}} V_-(\cdot), \quad V_-(a) := \min_{u \in U(a)} \{ V[f(a, u)] + \varphi(a, u) \}$$

$$V_t(\cdot) = \mathfrak{F}[V_{t+1}(\cdot)]$$



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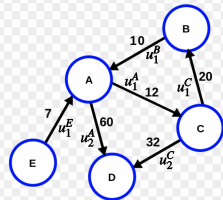
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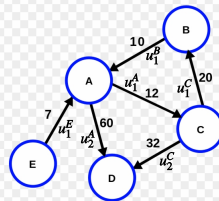
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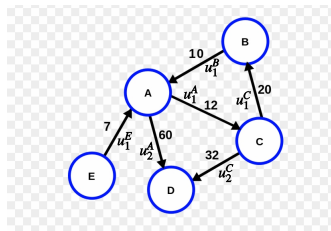
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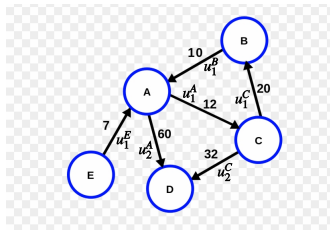
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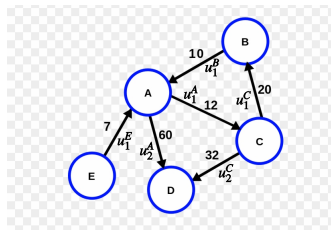
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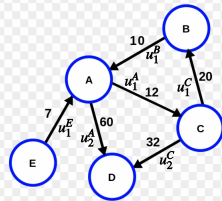
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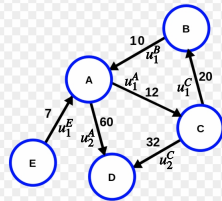
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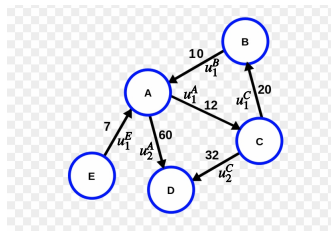
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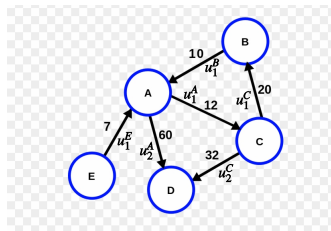
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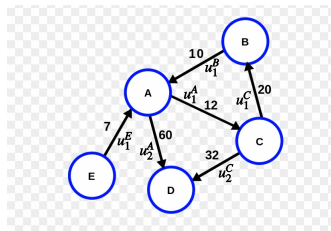
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$$V_t(\cdot) = \mathfrak{F}[V_{t+1}(\cdot)] \quad t = -1, \dots, -\infty$$



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$$\begin{aligned} V_{-k}^\downarrow(a) &= \min_{\theta \in [-k:0]} V_\theta(a) = \min \left\{ \min_{\theta \in [-k:-1]} V_\theta(a); \eta(a) \right\} \\ &= \min \left\{ \min_{u \in U(a)} \min_{\theta \in [-k:-1]} \left( V_{\theta+1}[f(a, u)] + \varphi(a, u) \right); \eta(a) \right\} \\ &= \min \left\{ \min_{u \in U(a)} \left[ \varphi(a, u) + \min_{\tau \in [-k+1:0]} V_\tau[f(a, u)] \right]; \eta(a) \right\} \\ &= \min \left\{ \min_{u \in U(a)} \left[ \varphi(a, u) + V_{-k+1}^\downarrow[f(a, u)] \right]; \eta(a) \right\} \end{aligned}$$

# Оптимальное планирование с нефиксированной длиной пути

## Problem statement

$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

$$x_0 = a, \quad k \text{ is freely manipulable} \quad (+)$$

$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\varphi(\cdot) > 0) \quad (\infty)$$

## Bellman equation

$$V_{-k}^{\downarrow}(a) = \min \left\{ \min_{u \in U(a)} \left[ \varphi(a, u) + V_{-k+1}^{\downarrow}[f(a, u)] \right]; \eta(a) \right\}, \quad V_0^{\downarrow}(\cdot) := \eta(\cdot)$$

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$$V_{-k}^\downarrow(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V_{-k+1}^\downarrow[f(a, u)]]; \eta(a) \right\}, \quad V_0^\downarrow(\cdot) := \eta(\cdot)$$

## Theorem

The sequence  $V_{-k}^\downarrow(\cdot)$  is monotonically decreasing  $V_{-k}^\downarrow(\cdot) \leq V_{-k+1}^\downarrow(\cdot)$  and there exists  $k \leq -1$  such that  $V_{-k}^\downarrow(\cdot) \equiv V_{-k+1}^\downarrow(\cdot)$ . For such  $k$ , we have  $V^\downarrow(\cdot) := V_{-k}^\downarrow(\cdot) = V_{-\theta}^\downarrow(\cdot) \forall \theta \geq k$  and  $V^\downarrow(\cdot)$  is the unique solution of the equation

$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$  and simultaneously the Bellman function of the optimization problem.



# Оптимальное планирование с нефиксированной длиной пути

## Problem statement

$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

$$x_0 = a, \quad k \text{ is freely manipulable} \quad (+)$$

$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\varphi(\cdot) > 0) \quad (\infty)$$

$$V\text{-closing state} \sim V(a) = \eta(a)$$

## Bellman equation

$$V_{-k}^\downarrow(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V_{-k+1}^\downarrow[f(a, u)]]; \eta(a) \right\}, \quad V_0^\downarrow(\cdot) := \eta(\cdot)$$

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# Оптимальное планирование с нефиксированной длиной пути

## Problem statement

$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

$$x_0 = a, \quad k \text{ is freely manipulable} \quad (+)$$

$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\varphi(\cdot) > 0) \quad (\infty)$$

$$V\text{-closing state} \sim V(a) = \eta(a) \\ \mathcal{U}_V(a) := \text{Arg min}_{u \in U(a)} \{ V[f(a, u)] + \varphi(a, u) \}$$

## Bellman equation

$$V_{-k}^\downarrow(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V_{-k+1}^\downarrow[f(a, u)]]; \eta(a) \right\}, \quad V_0^\downarrow(\cdot) := \eta(\cdot)$$

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$$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\} \text{ and simultaneously the Bellman function of the optimization problem.}$$

# Оптимальное планирование с нефиксированной длиной пути

## Problem statement

$$x_{t+1} = f(x_t, u_t), \quad u_t \in U[x_t] \quad \forall t \in [0 : k) \quad (\div)$$

$$x_0 = a, \quad k \text{ is freely manipulable} \quad (+)$$

$$I := \sum_{t \in [0:k)} \varphi(x_t, u_t) + \eta(x_k) \rightarrow \min \quad (\varphi(\cdot) > 0) \quad (\infty)$$

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## Theorem

The set of  $V^\downarrow$ -closing states is not empty. **Optimal process**  $\Leftrightarrow u_t \in \mathcal{U}_V(x_t)$  and the process terminates upon arrival at a closing state.

## Bellman equation

$$V_{-k}^\downarrow(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V_{-k+1}^\downarrow[f(a, u)]]; \eta(a) \right\}, \quad V_0^\downarrow(\cdot) := \eta(\cdot)$$

## Theorem

The sequence  $V_{-k}^\downarrow(\cdot)$  is monotonically decreasing  $V_{-k}^\downarrow(\cdot) \leq V_{-k+1}^\downarrow(\cdot)$  and there exists  $k \leq -1$  such that  $V_{-k}^\downarrow(\cdot) \equiv V_{-k+1}^\downarrow(\cdot)$ . For such  $k$ , we have  $V^\downarrow(\cdot) := V_{-k}^\downarrow(\cdot) = V_{-\theta}^\downarrow(\cdot) \forall \theta \geq k$  and  $V^\downarrow(\cdot)$  is the unique solution of the equation

$$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\} \text{ and simultaneously the Bellman function of the optimization problem.}$$



## Lemma

Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V^\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

## Lemma

Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V^\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

$$V^\downarrow(x_0) = \varphi(x_0, u_0) + V^\downarrow(x_1) = \varphi(a, u_0) + V^\downarrow[f(a, u_0)], \quad a := x_0$$

$$V^\downarrow(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V^\downarrow[f(a, u)]]; \eta(a) \right\}$$

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Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V^\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

## Lemma

Any solution of the equation

$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$  is a lower estimate of the Bellman function.

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$$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$$
$$I = \sum_{t=0}^{k-1} \varphi(x_t, u_t) + \eta[x_k] \geq \sum_{t=0}^{k-1} \varphi(x_t, u_t) + V[x_k]$$



## Lemma

Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V^\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

## Lemma

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$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$  is a lower estimate of the Bellman function.

$$\begin{aligned} V(a) &= \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\} \\ I &= \sum_{t=0}^{k-1} \varphi(x_t, u_t) + \eta[x_k] \geq \sum_{t=0}^{k-1} \varphi(x_t, u_t) + V[x_k] \\ &= \sum_{t=0}^{k-2} \varphi(x_t, u_t) + \varphi(x_{k-1}, u_{k-1}) + V[f(x_{k-1}, u_{k-1})] \end{aligned}$$

## Lemma

Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

## Lemma

Any solution of the equation

$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$  is a lower estimate of the Bellman function.

$$\begin{aligned} V(a) &= \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\} \\ I &= \sum_{t=0}^{k-1} \varphi(x_t, u_t) + \eta[x_k] \geq \sum_{t=0}^{k-1} \varphi(x_t, u_t) + V[x_k] \\ &= \sum_{t=0}^{k-2} \varphi(x_t, u_t) + \varphi(x_{k-1}, u_{k-1}) + V[f(x_{k-1}, u_{k-1})] \\ &\geq \sum_{t=0}^{k-2} \varphi(x_t, u_t) + V[x_{k-1}] \geq \dots \geq V[x_0] \end{aligned}$$

# Доказательства.

## Lemma

Let  $[\{x_t\}_{t=0}^k, \{u_t\}_{t=0}^{k-1}]$  be an optimal process. Then the process  $[\{x_t\}_{t=\theta}^k, \{u_t\}_{t=\theta}^{k-1}]$  is also optimal for any  $\theta \in [0 : k]$ , the state  $x_k$  is stopping,  $u_t \in \mathcal{U}_{V^\downarrow}(x_t)$ , and  $V^\downarrow(x_0) = \sum_{t=0}^{\theta-1} \varphi(x_t, u_t) + V^\downarrow(x_\theta)$ .

## Lemma

Any solution of the equation

$V(a) = \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\}$  is a lower estimate of the Bellman function.

## Lemma

Let  $u_t \in \mathcal{U}_V(x_t)$  and the process terminates upon arrival at the first stopping state. Then this process is optimal.

$$\begin{aligned} V(a) &= \min \left\{ \min_{u \in U(a)} [\varphi(a, u) + V[f(a, u)]]; \eta(a) \right\} \\ I &= \sum_{t=0}^{k-1} \varphi(x_t, u_t) + \eta[x_k] \geq \sum_{t=0}^{k-1} \varphi(x_t, u_t) + V[x_k] \\ &= \sum_{t=0}^{k-2} \varphi(x_t, u_t) + \varphi(x_{k-1}, u_{k-1}) + V[f(x_{k-1}, u_{k-1})] \\ &\geq \sum_{t=0}^{k-2} \varphi(x_t, u_t) + V[x_{k-1}] \geq \dots \geq V[x_0] \end{aligned}$$