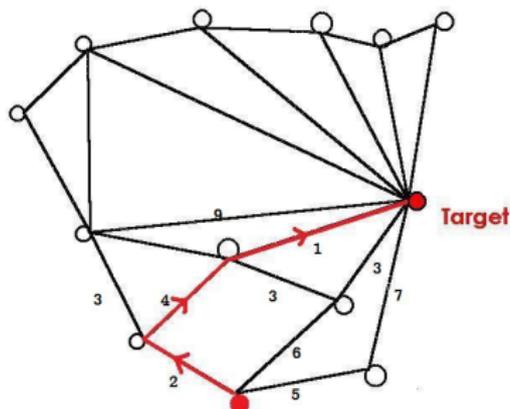


# Shortest paths and Hamilton-Jacobi equations on a network

joint work with Dirk Schieborn (Eberhard-Karls University, Tübingen)

11w5086 Advancing numerical methods for viscosity solutions and applications

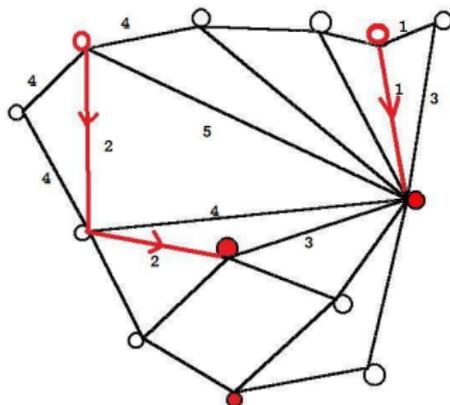
# The single target shortest path problem



## Problem

In a weighted graph, finding the distance of the vertices from a prescribed target vertex and detect the shortest path (Dijkstra's algorithm)

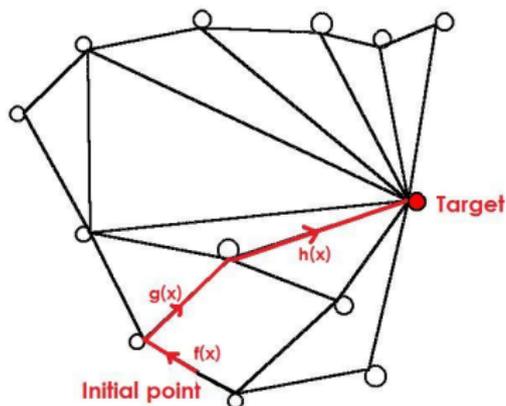
# The multiple targets shortest path problem



## Problem

In a weighted graph, finding the distance of the vertices from a prescribed target set and detect the shortest path.

# The multiples targets shortest path problem with continuous running cost



## Problem

Finding the distance of **any point in the graph** from a prescribed target set when the cost varies in a continuous way along the edges

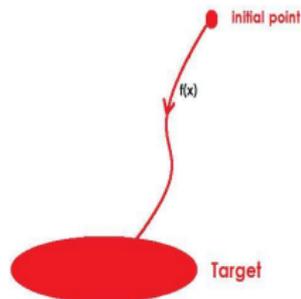
# Target problem in $\mathbb{R}^N$



In  $\mathbb{R}^N$ , finding the **weighted distance** from a given target set is equivalent to solve the Eikonal equation  $|Du(x)| = f(x)$  with  $u = 0$  on the target.

To solve the target problem with continuous running cost, introduce Eikonal equations of the form  $H(x, Du) = 0$  on a graph.

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# Literature about differential equations on networks

- Lumer, Nicaise, von Below: linear and semilinear problem on networks (maximum principle, spectral theory, etc.)
- Lagnese-Leugering: applications to wave equations (networks of vibrating strings)
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## Aim

Introduce a concept of viscosity solution which preserves the main features of viscosity theory: uniqueness, existence, and stability; sufficiently “**weak**” to yield existence, while sufficiently “**selective**” to ensure uniqueness and stability with respect to uniform convergence.

## Difficulties

- 1 How to modelize the differential structure of the network (which is not a regular manifold).
- 2 Which condition to impose at the vertices (**transition condition**). For second order linear equation, transition conditions are the key point to obtain the Maximum Principle.

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# The definition of Network

A **network**  $\Gamma$  is couple  $(V, E)$  where

- $V := \{v_i, i \in I\}$  is a finite collection of pairwise different points in  $\mathbb{R}^N$ ;
- $E := \{e_j : j \in J\}$  is a finite collection of *differentiable curves* in  $\mathbb{R}^N$  given by  $e_j := \pi_j((0, l_j))$  with  $\pi_j : [0, l_j] \subset \mathbb{R} \rightarrow \mathbb{R}^N, j \in J$ . Furthermore

- $\pi_j(0), \pi_j(l_j) \in V$  for all  $j \in J$  and  $\#(\bar{e}_j \cap V) = 2$  for all  $j \in J$
- $\bar{e}_j \cap \bar{e}_k \subset V$ , and  $\#(\bar{e}_j \cap \bar{e}_k) \leq 1$  for all  $j, k \in J, j \neq k$ .
- For all  $v, w \in V$  there is a path with end points  $v$  and  $w$  (i.e. a sequence of edges  $\{e_j\}_{j=1}^N$  such that  $\#(\bar{e}_j \cap \bar{e}_{j+1}) = 1$  and  $v \in \bar{e}_1, w \in \bar{e}_N$ ) (the graph is *connected*).

## Some definitions

- $Inc_i := \{j \in J : e_j \text{ incident } v_i\}$  is the set of arcs incident the vertex  $v_i$ .
- The parametrization of the arcs  $e_j$  induces an orientation on the edges, expressed by the signed incidence matrix  $A = \{a_{ij}\}_{i,j \in J}$

$$a_{ij} := \begin{cases} 1 & \text{if } v_i \in \bar{e}_j \text{ and } \pi_j(0) = v_i, \\ -1 & \text{if } v_i \in \bar{e}_j \text{ and } \pi_j(l_j) = v_i, \\ 0 & v_i \notin \bar{e}_j. \end{cases}$$

- Given a nonempty set  $I_B \subset I$ , we define  $\partial\Gamma := \{v_i, i \in I_B\}$  to be the set of boundaries vertices, while for  $I_T := I \setminus I_B$  is the set of transition vertices.

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## Continuity

Given  $u : \bar{\Gamma} \rightarrow \mathbb{R}$ ,  $u^j$  the restriction of  $u$  to  $\bar{e}_j$ , i.e.

$$u^j := u \circ \pi_j : [0, l_j] \rightarrow \mathbb{R}.$$

$u$  is continuous in  $\bar{\Gamma}$  if  $u^j \in C([0, l_j])$  for any  $j \in J$  and

$$u^j(\pi_j^{-1}(v_i)) = u^k(\pi_k^{-1}(v_i)) \quad \text{for any } i \in I, j, k \in \text{Inc}_i.$$

## Differentiation

We define differentiation along an edge  $e_j$  by

$$\partial_j u(x) := \partial_j u^j(\pi_j^{-1}(x)) = \frac{\partial}{\partial x} u^j(\pi_j^{-1}(x)), \quad \text{for all } x \in e_j,$$

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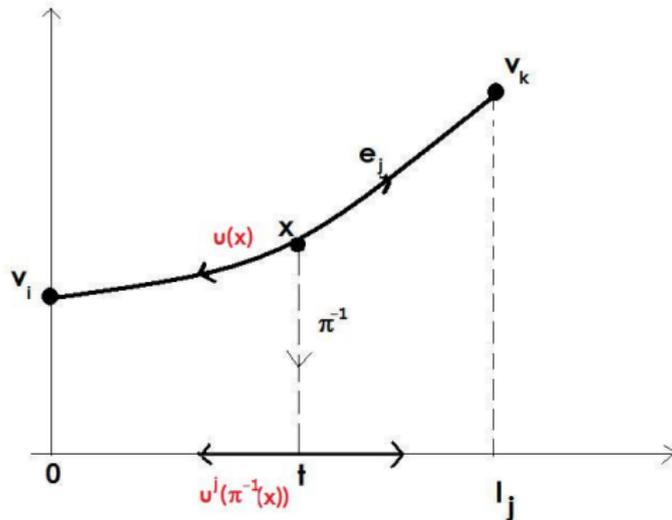


Figure: Differentiation along the edge

# Hamiltonian

A Hamiltonian  $H : \bar{\Gamma} \times \mathbb{R} \rightarrow \mathbb{R}$  of **eikonal type** is given by

$H(x, p) = H^j(\pi_j^{-1}(x), p)$  for  $x \in e_j$  where  $(H^j)_{j \in J}$  with

$H^j : [0, l_j] \times \mathbb{R} \rightarrow \mathbb{R}$

$$H^j \in C^0([0, l_j] \times \mathbb{R}), \quad j \in J, \quad (1)$$

$$H^j(x, p) \text{ is convex in } p \in \mathbb{R} \text{ for any } x \in [0, l_j], j \in J, \quad (2)$$

$$H^j(x, p) \rightarrow +\infty \text{ as } |p| \rightarrow \infty \text{ for any } x \in [0, l_j], j \in J, \quad (3)$$

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(1)–(3) are standard conditions. Assumptions (4)–(5) are compatibility conditions of  $H$  at the vertices of  $\bar{\Gamma}$ , i.e. continuity at the vertices and independence of the orientation of the incident arc (the network is not oriented). For example,  $H^j(x, p) := |p|^2 - f^j(x)$ ,  $j \in J$ , where  $f^j \in C^0([0, l_j])$ ,  $f^j(x) \geq 0$ ,  $f^j(\pi_j^{-1}(v_i)) = f^k(\pi_k^{-1}(v_i))$  for any  $i \in I$ ,  $j, k \in \text{Inc}_i$ .

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# Test Functions

## Definition

- i)  $\varphi$  is differentiable at  $x \in e_j$ , if  $\varphi^j := \varphi \circ \pi_j : [0, l_j] \rightarrow \mathbb{R}$  is differentiable at  $t = \pi_j^{-1}(x)$ .
- ii) Let  $x = v_i$ ,  $i \in I_T$ ,  $j, k \in Inc_i$ ,  $j \neq k$ .  $\varphi$  is  $(j, k)$ -differentiable at  $x$  if

$$\boxed{a_{ij} \partial_j \varphi_j(\pi_j^{-1}(x)) + a_{ik} \partial_k \varphi_k(\pi_k^{-1}(x)) = 0,} \quad (6)$$

where  $(a_{ij})$  as is the incidence matrix.

## Remark

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A function  $u$  is called a **viscosity subsolution** if

- i) If  $x \in e_j$ ,  $j \in J$ , and for any  $\varphi \in C(\Gamma)$  which is differentiable at  $x$  and for which  $u - \varphi$  attains a local maximum at  $x$

$$H^j(\pi_j^{-1}(x), \partial_j \varphi_j(\pi_j^{-1}(x))) \leq 0.$$

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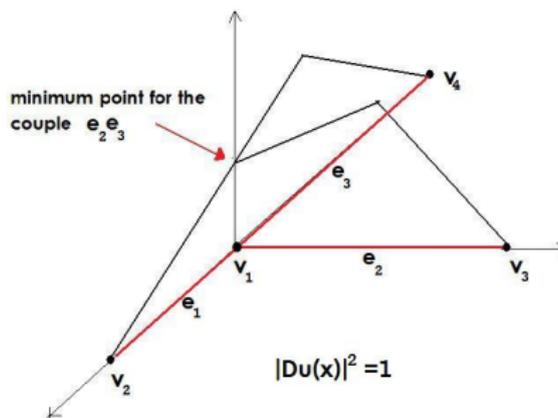
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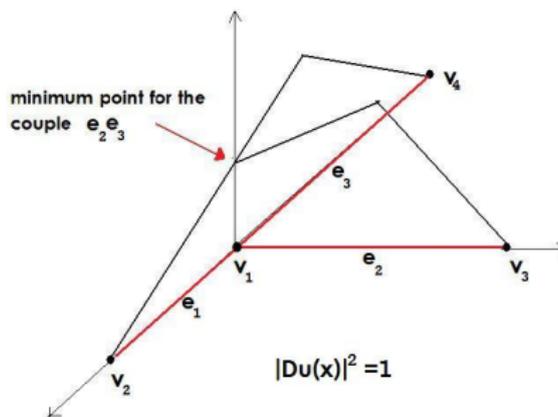
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## The distance function

$$S(y, x) = \inf \left\{ \int_0^t L(\gamma(s), \dot{\gamma}(s)) ds : \gamma \in B_{y,x}^t \right\}, \quad x, y \in \Gamma$$

where

- $B_{y,x}^t$  is the set of paths  $\gamma : [0, t] \rightarrow \Gamma$  connecting  $y$  to  $x$  and piecewise differentiable (i.e.  $t_0 := 0 < t_1 < \dots < t_{n+1} := t$  s.t. for any  $m = 0, \dots, n$ , we have  $\gamma([t_m, t_{m+1}]) \subset \bar{e}_{j_m}$  for some  $j_m \in J$ ,  $\pi_{j_m}^{-1} \circ \gamma \in C^1(t_m, t_{m+1})$ , and  $\dot{\gamma}(s) = \frac{d}{ds}(\pi_{j_m}^{-1} \circ \gamma)(s)$ ).
- The Lagrangian  $L(x, q)$  is defined by

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The **path distance**  $d(y, x)$  on the network coincides with  $S(y, x)$  for  $H(x, p) = |p|^2 - 1$ .

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- $B_{y,x}^t$  is the set of paths  $\gamma : [0, t] \rightarrow \Gamma$  connecting  $y$  to  $x$  and piecewise differentiable (i.e.  $t_0 := 0 < t_1 < \dots < t_{n+1} := t$  s.t. for any  $m = 0, \dots, n$ , we have  $\gamma([t_m, t_{m+1}]) \subset \bar{e}_{j_m}$  for some  $j_m \in J$ ,  $\pi_{j_m}^{-1} \circ \gamma \in C^1(t_m, t_{m+1})$ , and  $\dot{\gamma}(s) = \frac{d}{ds}(\pi_{j_m}^{-1} \circ \gamma)(s)$ ).
- The Lagrangian  $L(x, q)$  is defined by

$$L(x, q) = \sup_{p \in \mathbb{R}} \{p q - H^j(\pi_j^{-1}(x), p)\} \quad x \in \bar{e}_j$$

The **path distance**  $d(y, x)$  on the network coincides with  $S(y, x)$  for  $H(x, p) = |p|^2 - 1$ .

# Existence and uniqueness

## Theorem

Assume that there exists a differentiable function  $\psi$  such that  $H(x, D\psi) < 0$  in  $\Gamma$ . Let  $g : \partial\Gamma \rightarrow \mathbb{R}$  be a continuous function satisfying

$$g(x) - g(y) \leq S(y, x) \quad \text{for any } x, y \in \partial\Gamma = I_B,$$

Then

$$u(x) := \min\{g(y) + S(y, x) : y \in \partial\Gamma\}$$

is the **unique viscosity solution** of

$$\begin{cases} H(x, Du) = 0, & x \in \Gamma; \\ u(x) = g(x), & x \in \partial\Gamma. \end{cases}$$

# Sketch of the proof

## Uniqueness:

Classical doubling argument

$$\Phi_\varepsilon(x, y) := u(x) - v(y) - \varepsilon^{-1}d(x, y)^2$$

for Maximum Principle ( $d^2$  is an admissible test function) + Ishii's trick.

## Existence:

The function  $S(y, \cdot)$  is a subsolution in  $\Gamma$  and a supersolution in  $\Gamma \setminus \{y\}$ .

Moreover

$$S(y, x) = \max\{u(x) : u \text{ is a subsolution s.t. } u(y) = 0\}.$$

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# Stability

## Theorem

Assume  $H_n(x, p) \rightarrow H(x, p)$  uniformly for  $n \rightarrow \infty$  (i.e.  $H_n^j(\pi_j^{-1}(x), p) \rightarrow H^j(\pi_j^{-1}(x), p)$  uniformly for  $(x, p) \in \bar{e}_j \times \mathbb{R}$  for any  $j \in \mathcal{J}$ ). For any  $n \in \mathbb{N}$  let  $u_n$  be a solution of

$$H_n(x, Du) = 0, \quad x \in \Gamma,$$

and assume  $u_n \rightarrow u$  uniformly in  $\Gamma$  for  $n \rightarrow \infty$ . Then  $u$  is a solution of

$$H(x, Du) = 0.$$

# Classification of singularities

We consider the equation

$$\begin{cases} |Du|^2 - f(x) = 0, & x \in \Gamma; \\ u(x) = 0, & x \in I_B. \end{cases}$$

with  $f > 0$  in  $\Gamma$ . It is possible to prove that an edge contains at most one non-differentiability (singular) point. We define  $k^{edge} : E \rightarrow \{0, 1\}$  by

$$k^{edge}(e_j) := \begin{cases} 1, & \text{if } e_j \text{ contains a singular point;} \\ 0, & \text{if } e_j \text{ does not contain a singular point.} \end{cases}$$

For a vertex  $v_i$ , we

$$k^{vertex}(v_i) := \#(Inc_i^-)$$

where  $Inc_i^-$  are the edges entering “downhill” in  $v_i$  (the more incident edges lead “downhill”, the more  $v_i$  assumes the character of a local maximum and the higher it should be weighted when counting the singularities).

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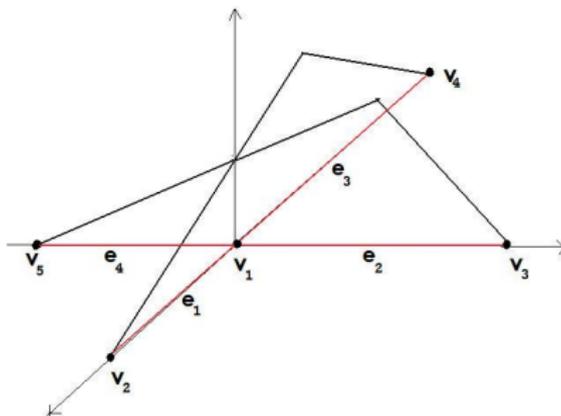


Figure:  $Inc_1^- = 2$

## Theorem

$$\sum_{i \in I} k^{vertex}(v_i) + \sum_{j \in J} k^{edge}(e_j) = \#(J)$$

i.e. the dimension of the singular set of the viscosity solution only depends on the number of edges of  $\Gamma$ .

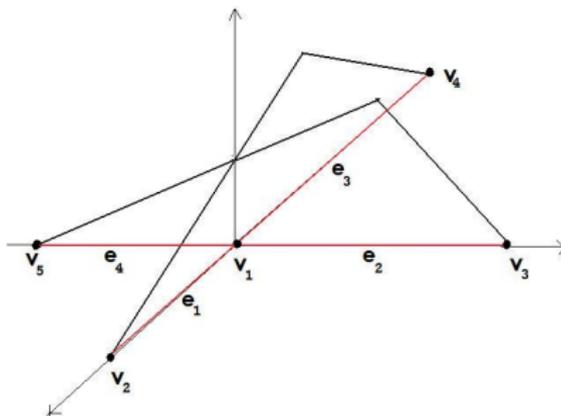


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# A semi-Lagrangian approximation scheme

(see Falcone-Ferretti, J. Comput. Phys. 175 (2002))

## Discretization in time

For  $h > 0$ , we define

- i) An admissible trajectory  $\gamma^h = \{\gamma_m^h\}_{m=0}^M \subset \Gamma$  is a finite number of points  $\gamma_m^h = \pi_{j_m}(t_m) \in \Gamma$  such that for any  $m = 0, \dots, M$ , the arc  $\widehat{\gamma_m^h \gamma_{m+1}^h} \subset \bar{e}_{j_m}$  for some  $j_m \in J$ .
- ii)  $B_{x,y}^h$  is the set of all such paths with  $\gamma_0^h = x$ ,  $\gamma_M^h = y$ .

We set

$$u_h(x) = \inf \left\{ \sum_{m=0}^{M-1} hL(\gamma_m^h, q_m) + g(y) : \gamma^h \in B_{x,y}^h, y \in \partial\Gamma \right\} \quad x \in \Gamma$$

Set  $x_{hq} := \pi_j(t - hq)$  (hence  $d(x, x_{hq}) = h|q|$ ). Then  $u_h$  is the unique Lipschitz-continuous solution of

- If  $x = \pi_j(t) \in e_j$

$$\begin{aligned} u_h^j(x) &= \inf_{q \in \mathbb{R}: x_{hq} \in \bar{e}_j} \{u_h(x_{hq}) + hL(x, q)\} \\ &= \inf_{q \in \mathbb{R}: x_{hq} \in \bar{e}_j} \{u_h^j(t - hq) + hL^j(t, q)\} \end{aligned}$$

- If  $x = v_i \in I_T$

$$\begin{aligned} u_h^j(v_i) &= \inf_{k \in Inc_j} \left[ \inf_{q \in \mathbb{R}: x_{hq} \in \bar{e}_k} \{u_h(x_{hq}) + hL(v_i, q)\} \right] \\ &= \inf_{k \in Inc_j} \left[ \inf_{q \in \mathbb{R}: x_{hq} \in \bar{e}_k} \{u_h^k(t - hq) + hL^k(t, q)\} \right] \end{aligned}$$

- If  $x \in I_B$ ,  $u_h^j(x) = g(x)$ .

## Discretization in space

For  $j \in J$ , consider a partition

$$P^j = \{t_0^j = 0 < \dots < t_m^j < \dots < t_{M_j}^j = l_j\}$$

of  $[0, l_j]$  such that  $|P^j| = \max_{1, \dots, M_j} (t_m^j - t_{m-1}^j) \leq k_j$ . Set  $x_m^j = \pi_j(t_m^j)$  and consider

$$W_{k_j}^j = \{w \in C(\bar{e}_j) : \partial_j w(x) \text{ is constant in } (x_{m-1}^j, x_m^j), m = 1, \dots, M_j\}.$$

Every element  $w$  in  $W_{k_j}^j$  can be expressed as

$$w(x) = \sum_{m=1}^{M_j} \bar{\beta}_m^j(x) w^j(x_m^j), \quad x \in e_j$$

for  $\bar{\beta}_j(x) = \beta_j(\pi_j^{-1}(x))$  and  $\beta_j$  tent functions for the partition  $P_j$ .

Set  $x_{hq}^{j,m} = \pi_j(t_m^j - hq)$  and  $k = \max_{j \in J} k_j$  and consider:

Find  $u_{hk} : \bar{\Gamma} \rightarrow \mathbb{R}$  such that  $u_{hk}^j \in W_{k_j}^j$  for  $j \in J$  and

- If  $x_m^j = \pi_j(t_m^j) \in e_j$

$$\begin{aligned} u_{hk}^j(x_m^j) &= \inf_{q \in \mathbb{R}: x_{hq}^{j,m} \in \bar{e}_j} \{u_{hk}(x_{hq}^{j,m}) + hL(x_{hq}^{j,m}, q)\} \\ &= \inf_{q \in \mathbb{R}: x_{hq}^{j,m} \in \bar{e}_j} \{u_{hk}^j(t_m^j - hq) + hL^j(t_m^j, q)\} \end{aligned}$$

- If  $x_m^j = v_i$

$$\begin{aligned} u_{hk}^j(v_i) &= \inf_{r \in \text{Inc}_i} \left[ \inf_{q \in \mathbb{R}: x_{hq}^{r,m} \in \bar{e}_r} \{u_{hk}(x_{hq}^{r,m}) + hL(v_i, q)\} \right] \\ &= \inf_{r \in \text{Inc}_i} \left[ \inf_{q \in \mathbb{R}: x_{hq}^{r,m} \in \bar{e}_r} \{u_{hk}^r(t_m^r - hq) + hL^r(t_m^r, q)\} \right] \end{aligned}$$

- If  $x_m^j = v_i \in I_B$ ,  $u_h^j(x) = g(x)$ .

For  $j \in J$ , we set  $U^j = \{u_{hk}^j(t_m^j)\}_{m=1}^{M_j}$ ,  $B^j(q) = \{\beta_l(t_m^j - hq)\}_{l,m=1}^{M_j}$  and  $\mathcal{L}^j(q) = \{L^j(t_m^j, q)\}_{m=1}^{M_j}$ , and we rewrite the previous as the finite-dimensional system

$$U_m^j = \inf_{q \in \mathbb{R}: \pi_j(t_m^j - hq) \in \bar{e}_j} \{B^j(q)U^j + h\mathcal{L}_m^j(q)\} \quad x_m^j = \pi_j(t_m^j) \in e_j$$

$$U_m^j = \inf_{r \in \text{Inc}_i} \inf_{q \in \mathbb{R}: \pi_r(t_m^j - hq) \in \bar{e}_r} \{B^r(q)U^j + h\mathcal{L}_m^r(q)\} \quad v_i = \pi_r(t_m^j) \in I_T$$

$$U_m^j = g(x_m^j) \quad x_m^j = \pi_j(t_m^j) \in I_B$$

## Theorem

Assume that  $h \rightarrow 0$  and  $\frac{k}{h} \rightarrow 0$ . Then  $u_{hk}$  converges uniformly to  $u$  solution of

$$\begin{cases} H(x, Du) = 0, & x \in \Gamma; \\ u(x) = g(x), & x \in \partial\Gamma. \end{cases}$$