

Visibility Optimizations

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Joint work with Li-Tien Cheng

Museum v.s. Tourists

Museum:

- Place exhibition items under surveillance cameras
- Place as few cameras as possible

Tourist:

- Shortest path to see Mona-Lisa (La Jaconde)
- Shortest path through the museum
- Not missing any exhibition item along the way

Complication:

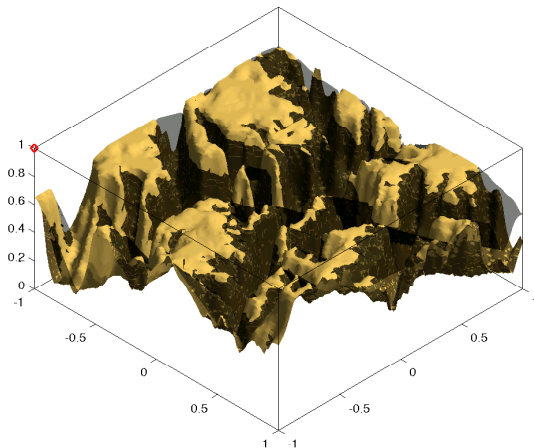
- The tourist does not have a map.
- The “tourist” wants to go from A to B, hiding from the cameras as much as possible.
- Guard following the “tourist”.

Problem 1: Find visibility

Given implicit surfaces (occluders) $\Gamma = \partial\Omega$.

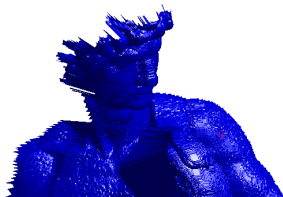
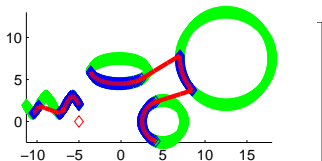
Light source (vantage point) at x_0 .

Construct: $u(y; x_0)$ s.t. $\{y : u(y; x_0) < 0\} =$ the invisible regions.



Construct visibility “on the fly”

Visibility from point clouds:



Problem 2: Optimizing for visibility

Known occluder configuration:

- 1 Place vantage point(s) maximizing visibility.
(vantage points could form a path.)
- 2 Path planning under visibility constraints: $H(\nabla v) = r(x, u, \phi)$.
(possibly with time dependence.)

Unknown occluder configuration:

- 1 Do the above using surfaces reconstructed from some remote sensing mechanism.
- 2 Place/Shape occluders optimizing certain visibility objective.

Problem 2: Optimizing for visibility

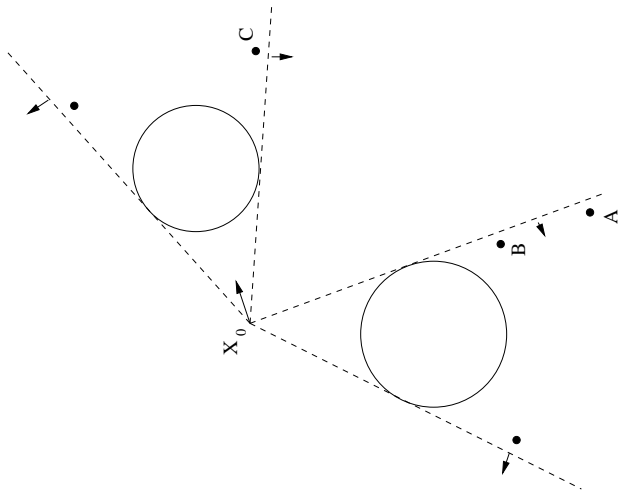
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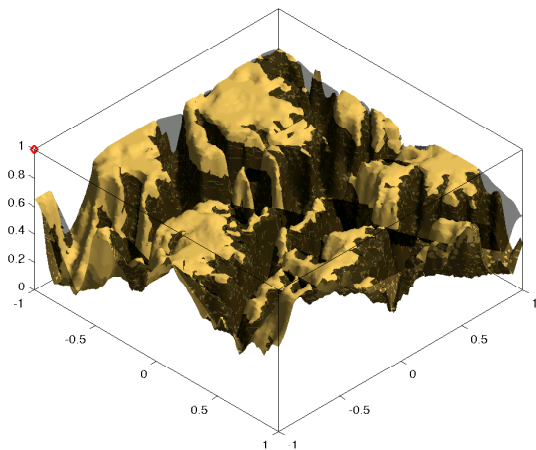
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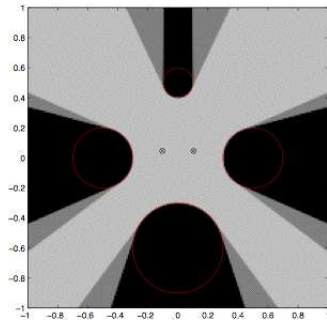
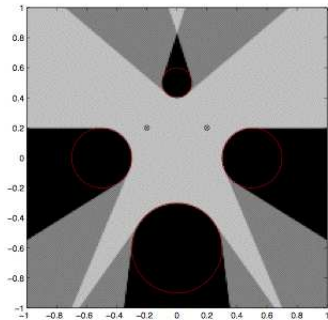
Maximize distance to occlusion



Maximize visible volume



Maximize “averaged exposure”



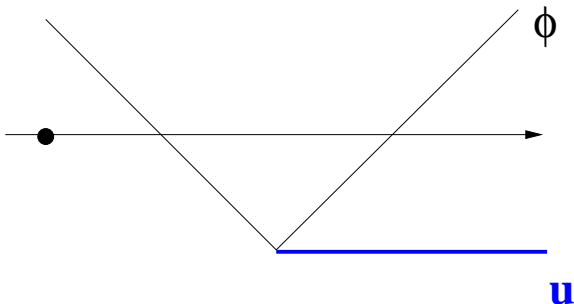
- Numerous work in the computer graphics community for computing visibility in polygonal environments.
E.g. The z-buffer, [Coorg et al.], [F. Durand] etc.
- From the robotics community: certain robotic path planning with visibility penalty in polygonal environments.
E.g. [Lavalle et al.], [Latombe et al]
Complexity scales with the number of edges
- Visibility of implicit surfaces: [Adalsteinson]

Our contribution:

- Continuous visibility representation and fast algorithm
- A variety of models of visibility optimization and the corresponding numerical algorithms.

Finding visibility

$$u(y, \mathbf{x}_0) := \min_{\xi \in L(\mathbf{x}_0, y)} \phi(\xi).$$



$$u(y, \mathbf{x}_0) := \min_{\xi \in L(\mathbf{x}_0, y)} \phi(\xi).$$

$$\frac{\partial}{\partial r} u = H(u - \phi) \left(\frac{\partial}{\partial r} \phi(\mathbf{x}_+) \right)^-$$

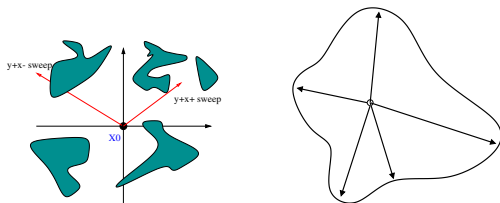
or

$$u(\mathbf{x}) = \phi(\mathbf{x}_0) + \int H(u - \phi) \left(\frac{\partial}{\partial r} \phi(\mathbf{x}_+) \right)^-.$$

[C.Y. Kao and Tsai, 2005]

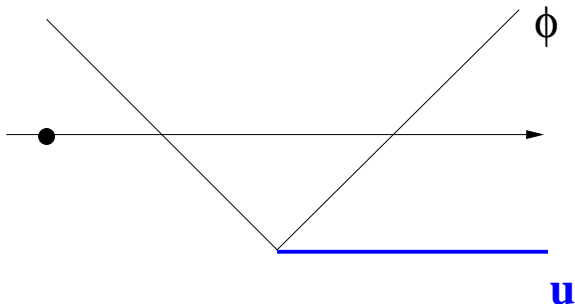
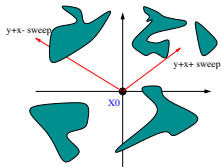
Grid-based algorithm

One-pass upwinding algorithm through each grid point in a predefined order (**independent of the occluders**): e.g.



Operations performed at each grid point:

- 1) Explicit solution of a 3 by 3 system
- 2) Taking minimum of two values

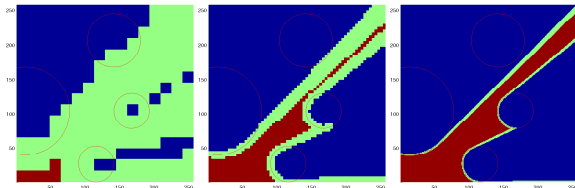


For each grid point x , $r(x) = (x - x_0) / |x - x_0|$:

- 1 Solve $\nabla u \cdot r(x) = 0$ at x ;
- 2 Update: $u(x) = \min(\phi(x), u(x))$.

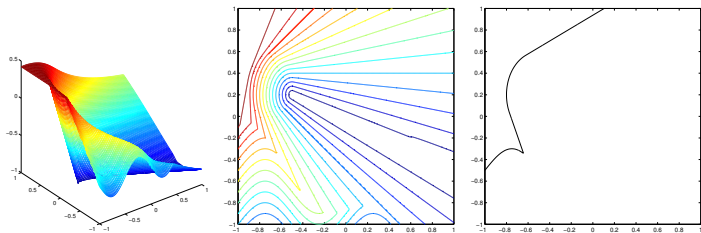
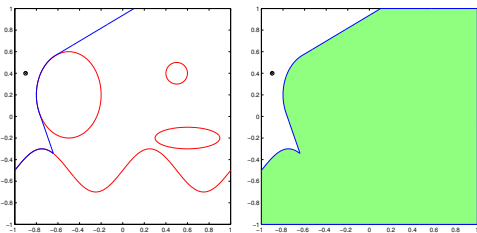
Variant: a optimal complexity version for DTED data.

Multi-level algorithm



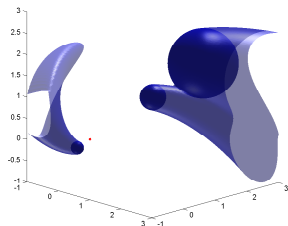
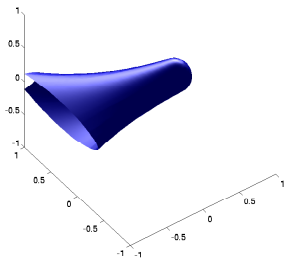
Skipping regions on which ψ does not change signs!
Based on Lipschitz estimates.

Visibility function



Visibility in more general ray fields

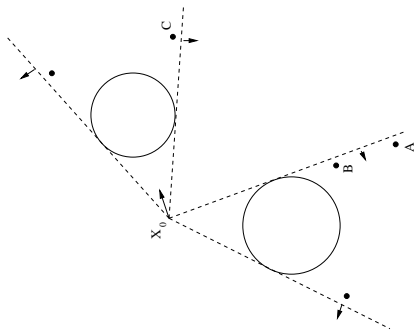
$$\nabla u \cdot r(x) = H(u - \phi) \min(\nabla \phi \cdot r(x), 0) + B.C.$$



Viscosity solution theory and discretizations. [C. Kao and Tsai]
Efficient algorithm.

Advantages

- Extremely easy algorithm
- Sharp, implicit, representation of occlusion boundaries
- Reciprocity of Solution: $u(x; y) = u(y; x)$.
- Continuity of solution: $\nabla_x u(y; x)$ and $\nabla_y u(y; x)$ make sense.
Notion of “how much occluded”?



- Algorithm independent of the topology and geometry of the occluders
- Extensions: many vantage points, moving surfaces.

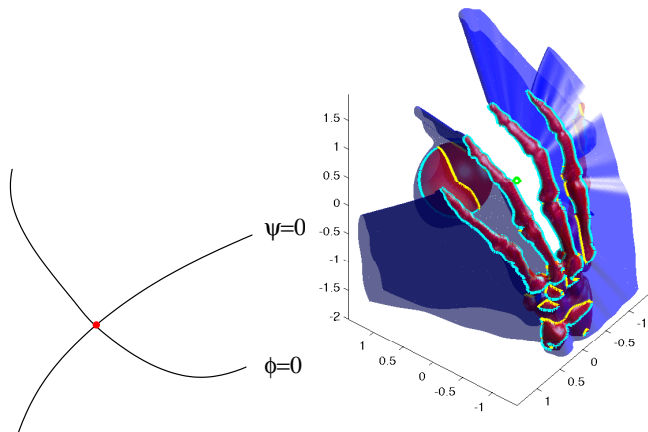
$$\max_{\alpha \in A} F(x, u, \nabla u, r_\alpha) = 0,$$

$$F(x, u, \nabla u, r_\alpha) = \nabla u \cdot r_\alpha - H(u - \phi) \min(\nabla \phi, 0).$$

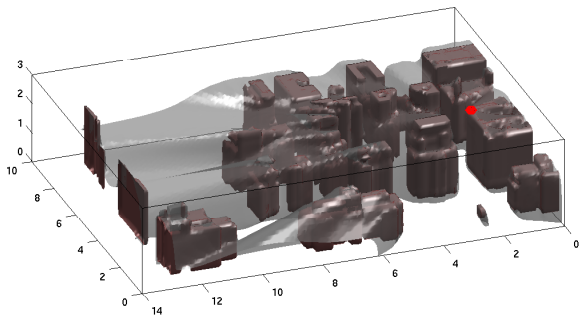
- Approximation of visibility from a path:
 $0 \leq u(x; \gamma) - u(x; \{\gamma_j\}) \leq \Delta \tau K \|\gamma\|_\infty.$

Characterizing horizons and terminators

We characterize the horizon and its terminator by the **intersections of level set functions**.



Visibility in DC.



Acknowledgement: DC grid data obtained from Flow Analysis, Inc.

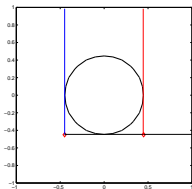
Joint visibility

Given two observers x_0 and x_1 :

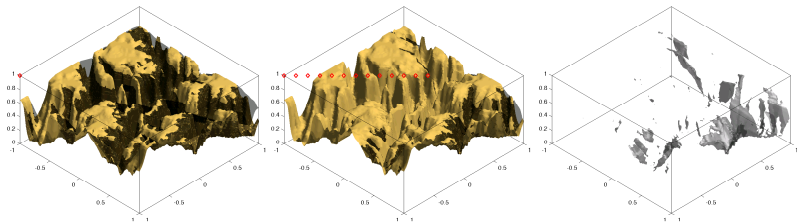
$$u(y; x_0, x_1) := \max(u(y; x_0), u(y; x_1)).$$

$\{u(y; x_0, x_1) < 0\}$ is the region occluded from both x_0 and x_1 .

Other Boolean operations possible.



Accumulative result (memory)



Joint visibility of a set of points \leftrightarrow visibility of a path \leftrightarrow
visibility with memory of a trajectory

Maximizing visibility

$$\min_{x_0} F(u; x_0) \quad \text{or} \quad \min_{\phi} F(\phi; u_{x_0})$$

Known occluder configuration:

- 1 Place vantage point(s) maximizing visibility:
 - Shortest path to see a given location
 - The “I want to see everything” problem (complete visibility)
 - Uniform viewing habit (exposure)
- 2 Path planning under visibility constraints: $H(\nabla v) = r(x, \phi, u)$. (with time dependence.)
 - The “I don’t want to be seen” problem.

Unknown occluder configuration:

- 1 Do the above.
- 2 Place/Shape occluders maximizing certain visibility objective.

$$\min_{x_0} F(u; x_0) \quad \text{or} \quad \min_{\phi} F(\phi; u_{x_0})$$

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An example with unknown occluder

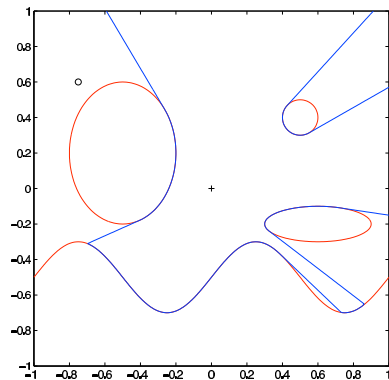
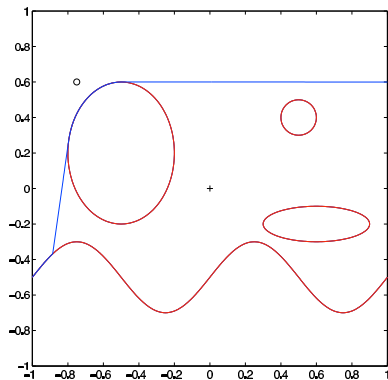
Shape reconstruction from gray scale images: $\min_{\phi} F(\phi; u_{x_0})$



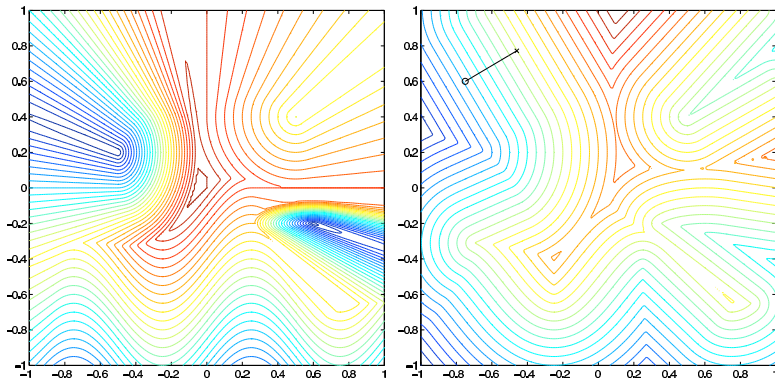
Joint work with Soatto et al.

Taking advantage of the reciprocity

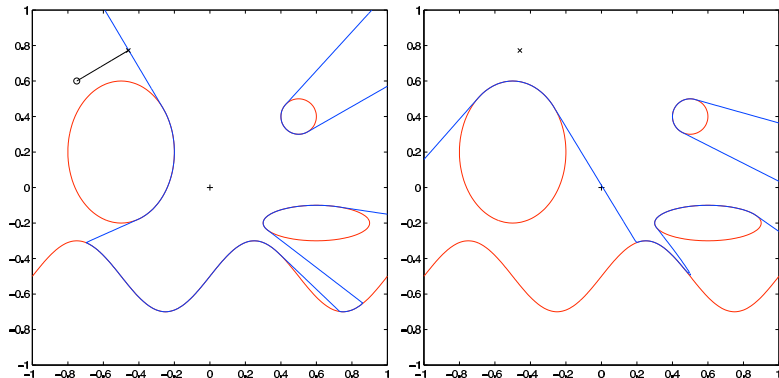
$$u(y, x_0) = u(x_0; y).$$



Taking advantage of the reciprocity

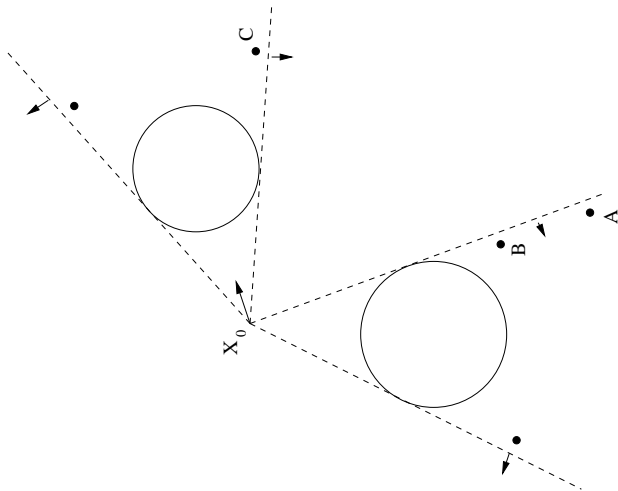


Taking advantage of the reciprocity



Same approach to non-trivial multiple targets.

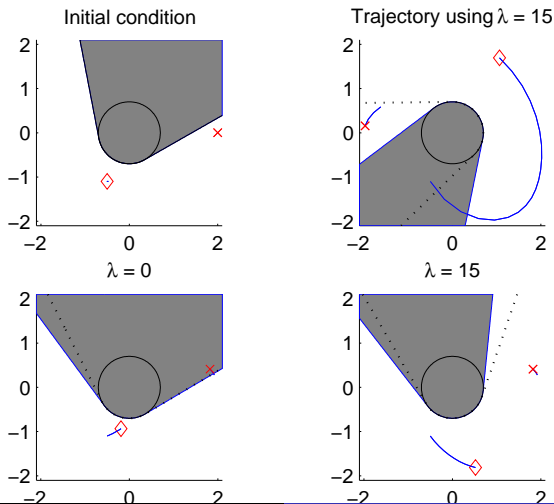
Taking advantage of the continuity



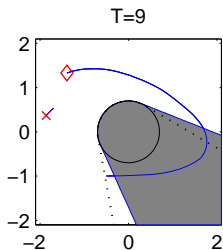
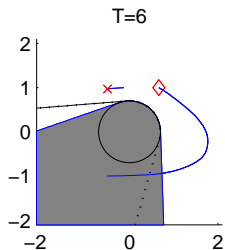
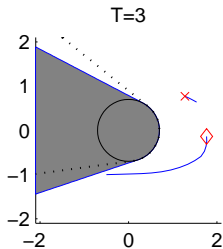
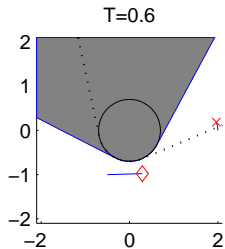
E.g. Move x_0 such that $u(B, x_0)$ is as “positive” as possible.

Keeping a moving object in sight

$$\max_{x(t;\alpha), \alpha \in \mathcal{A}} u(y(t); x(t)) \quad \text{s.t.} \quad |y - x| \sim d_0.$$



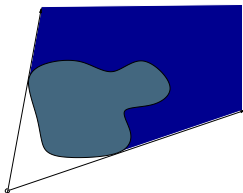
Keeping a moving object in sight



Volume based visibility optimization

Find x minimizing the area of invisible region outside of the occluders:

$$\min_x A(x) = \min_x \int_{\Omega} H(-u(y;x))H(\phi) dy$$



Continuity of u + compact $\Omega \implies$ continuity of A

Gradient descent:

$$\frac{dx}{dt} = -\nabla_x A(x) = \int_{\Omega} H(\phi) \frac{\nabla_x u(y;x)}{|\nabla_y u(y;x)|} \delta(u(y;x)) |\nabla_y u(y;x)| dy.$$

Generalization:

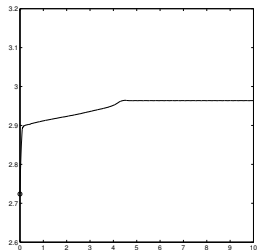
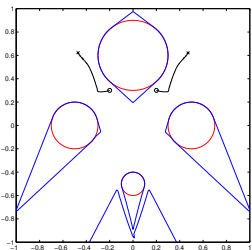
$$\min_{\{x_j\}} A(x) = \min_{\{x_j\}} \int_{\Omega} w(\phi, u, \Psi, y, \{x_j\}) H(-u(y; \{x_j\})) dy,$$

$w(\phi, u, \Psi, y, \{x_j\})$ is a weight function; e.g. near sighted — $w(d)$, where d = distance to the occluder.

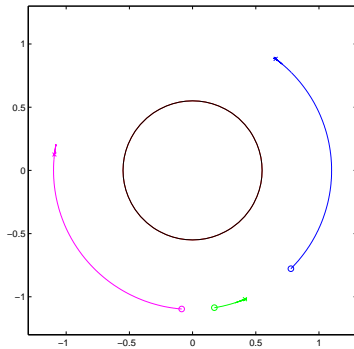
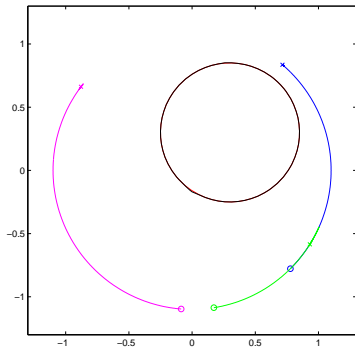
In particular, maximize “illumination” on Γ :

$$\min_x \int_{\Omega} \delta(\phi(y)) |\nabla \phi(y)| H(-u(y; x)) dy.$$

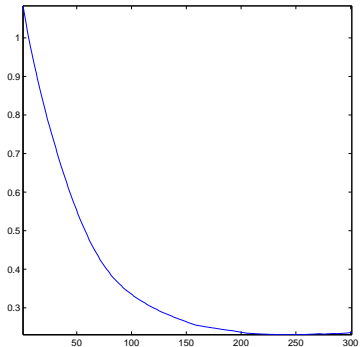
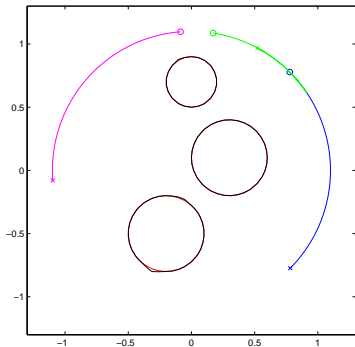
More examples



Guarding cameras



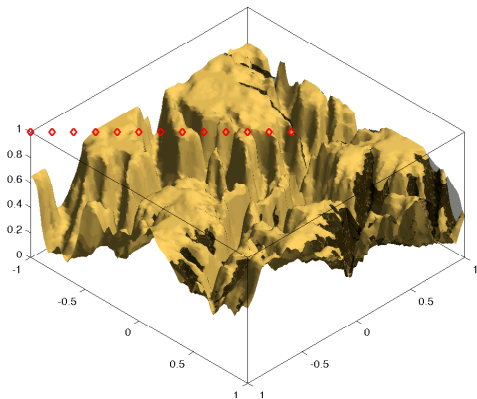
Guarding cameras



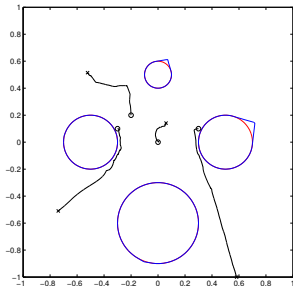
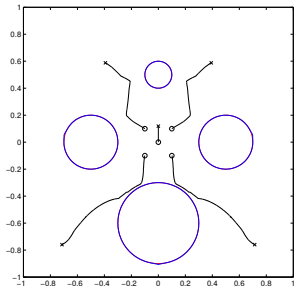
Visibility with memory

Consider the past trajectory as the vantage points:

$$\min_{x(t)} A(x) = \min_{x(t)} \int_{\Omega} \omega(\mathbf{x}, \phi, \dots) H(-\max_{s \in [0, t]} (u(y; x(s)))) H(\phi) dy.$$



Visibility with memory

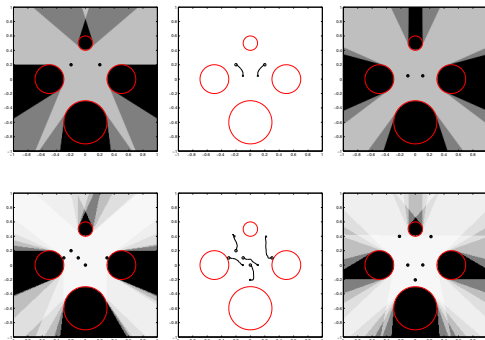


Average exposure

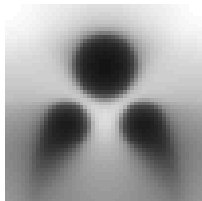
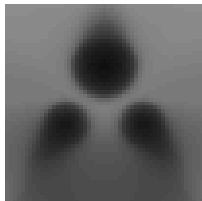
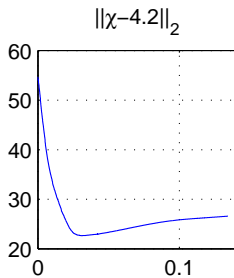
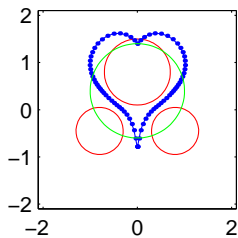
Exposure function: $\mathcal{X}(y; \{x_j\}) = \sum_j H(u(y; x_j))$.

$$\mathcal{X}(y; \gamma) = \int H(u(y; \gamma(s))) ds$$

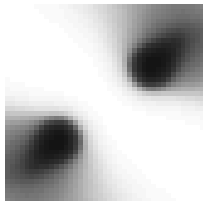
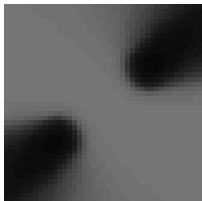
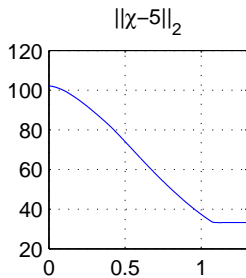
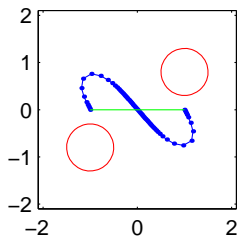
$$E(\gamma, \mathbf{C}) = \frac{1}{2} \|\mathcal{X}(\cdot; \gamma) - \mathbf{C}\|_{L^2}^2 + \lambda \text{length}(\gamma).$$



Path example 1



Path example 2



Spatial and temporal averaging

Optimal path $\gamma: [0, 1] \mapsto R^d, \gamma(0) = A, \gamma(1) = B$.

Introduce weighting:

$$\tilde{\chi}(t, \mathbf{x}; \gamma) = \int \mathcal{K}(\mathbf{x}, t, \tau) H(u(\mathbf{x}; \gamma(\tau))) |\dot{\gamma}(\tau)| d\tau,$$

and

$$\tilde{E}(\gamma, \mathbf{C}) = \frac{1}{2} \int \|\tilde{\chi}(t, \cdot, \gamma) - \mathbf{C}\|_{L^2}^2 dt + G(\gamma)$$

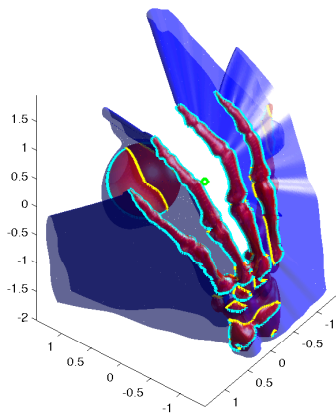
E.g. Temporal averaging = Memory:

$$\tilde{\chi}(t, \mathbf{x}; \gamma) = \int_{t-\delta t}^t H(u(\mathbf{x}; \gamma(\tau))) |\dot{\gamma}(\tau)| d\tau.$$

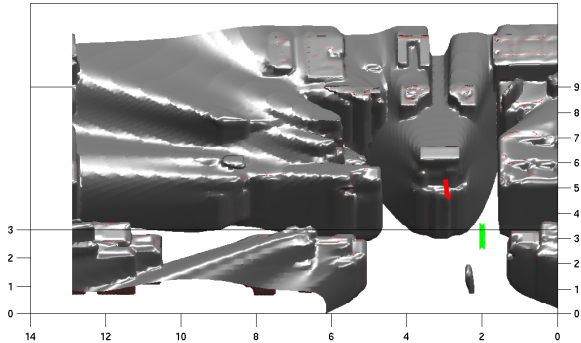
Visibility of occluder surface

$$\tilde{\chi}(t, x; \gamma) = \int K(x, t, \tau) H(u(x; \gamma(\tau))) |\gamma'(\tau)| d\tau,$$

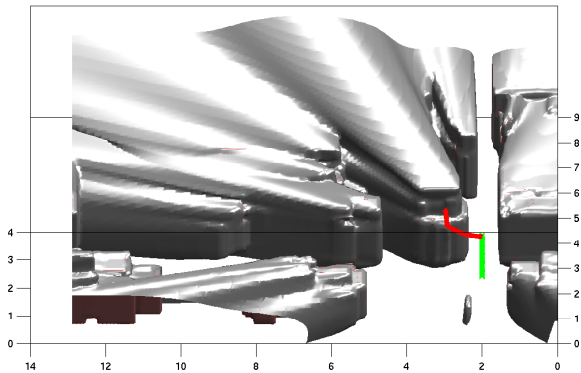
E.g. Visibility of occluder surface $K(x, t, \tau) = \delta(\phi(x)) |\nabla\phi|$.



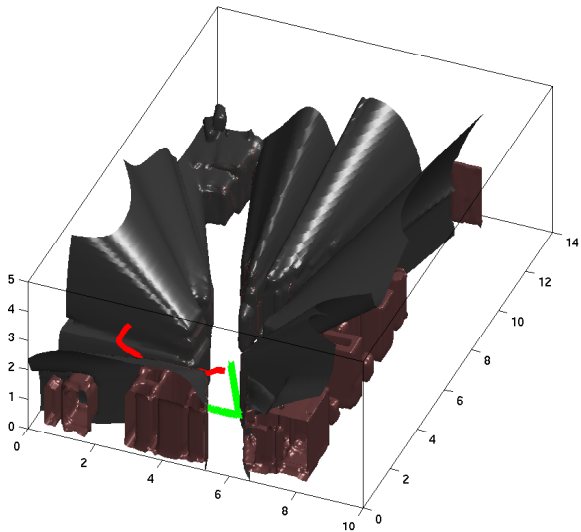
A Chase in DC



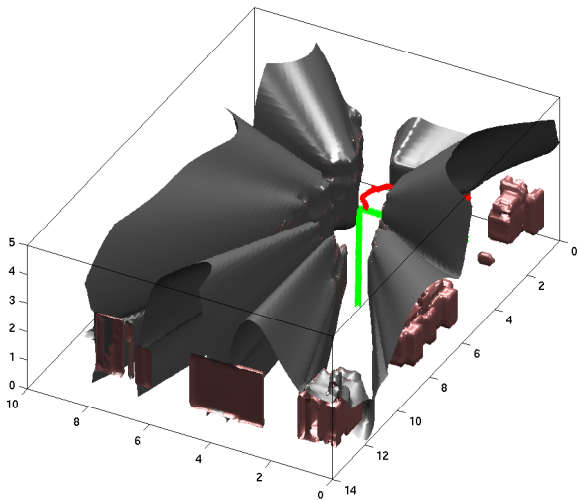
A Chase in DC



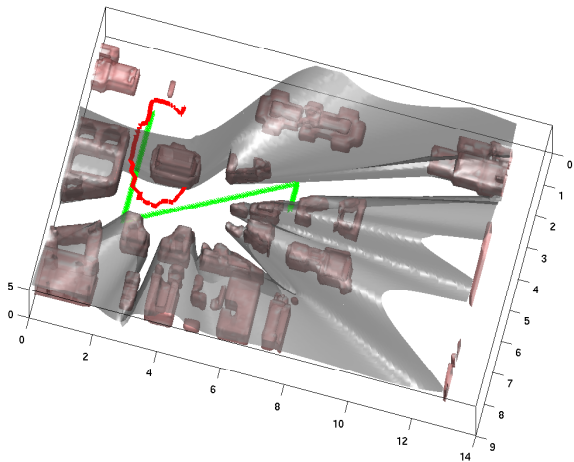
A Chase in DC



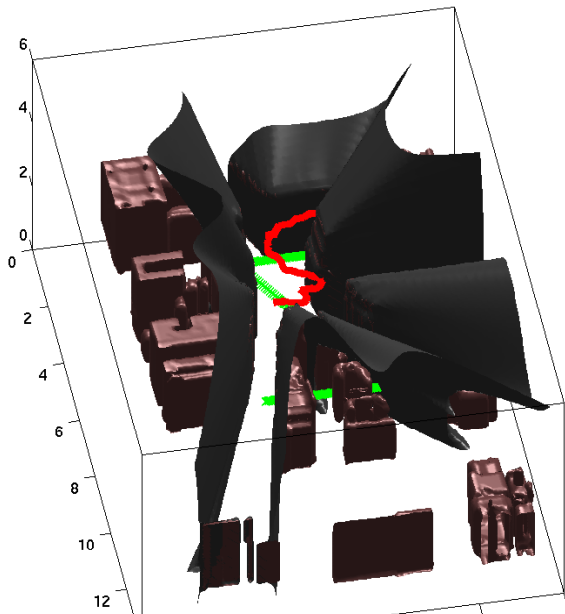
A Chase in DC



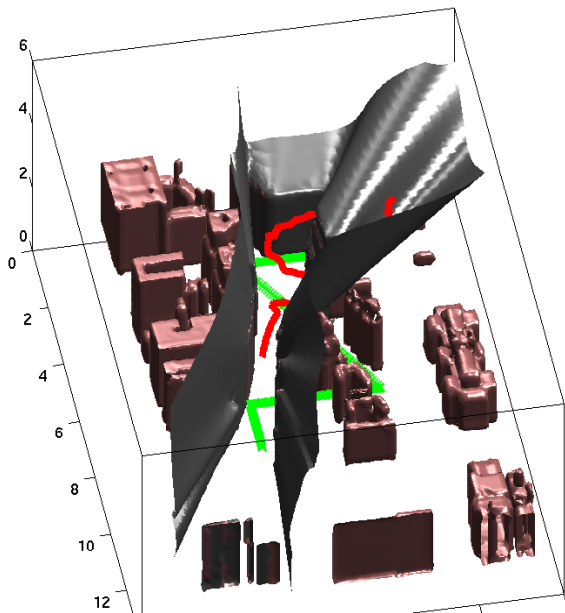
A Chase in DC



A Chase in DC



A Chase in DC



Under much simplified settings:

- A fast algorithm for constructing shadow boundaries (visibility)
- Adaptivity in the location of vantage points (sensors)

Challenges, current work, and wish list:

- Control and Game formulations [Kao-Navasca-Osher-Tsai]
- Bridging the shadow dynamics derived earlier with $\nabla_x u(y; x)$
- Analogues in other imaging mechanisms?
- Stochastic settings