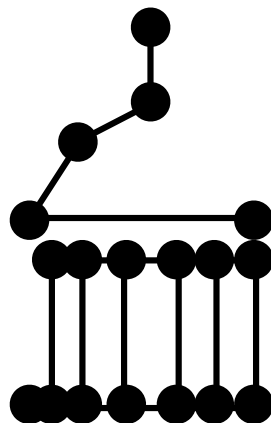


Use of 3D Graph-Theoretic Approaches in the Segmentation of Ophthalmic Structures

Mona K. Garvin, Ph.D.

Department of Electrical and Computer Engineering
The University of Iowa

VA Center for the Prevention and Treatment of Visual Loss





Collaborators involved in graph segmentation work

Faculty at Iowa:

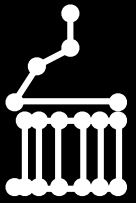
- Michael Abramoff
- Reinhard Beichel
- Mona Garvin (*me*)
- Andreas Wahle
- Milan Sonka
- Xiaodong Wu

Faculty at Notre Dame:

- Danny Chen

Students and post-docs (current/prior):

- Bhavna Antony
- Xin Dou
- Matt Gibson
- Dongfeng Han
- Zhihong Hu
- Kyungmoo Lee
- Qi Song
- Yin Yin
- Honghai Zhang
- Fei Zhao

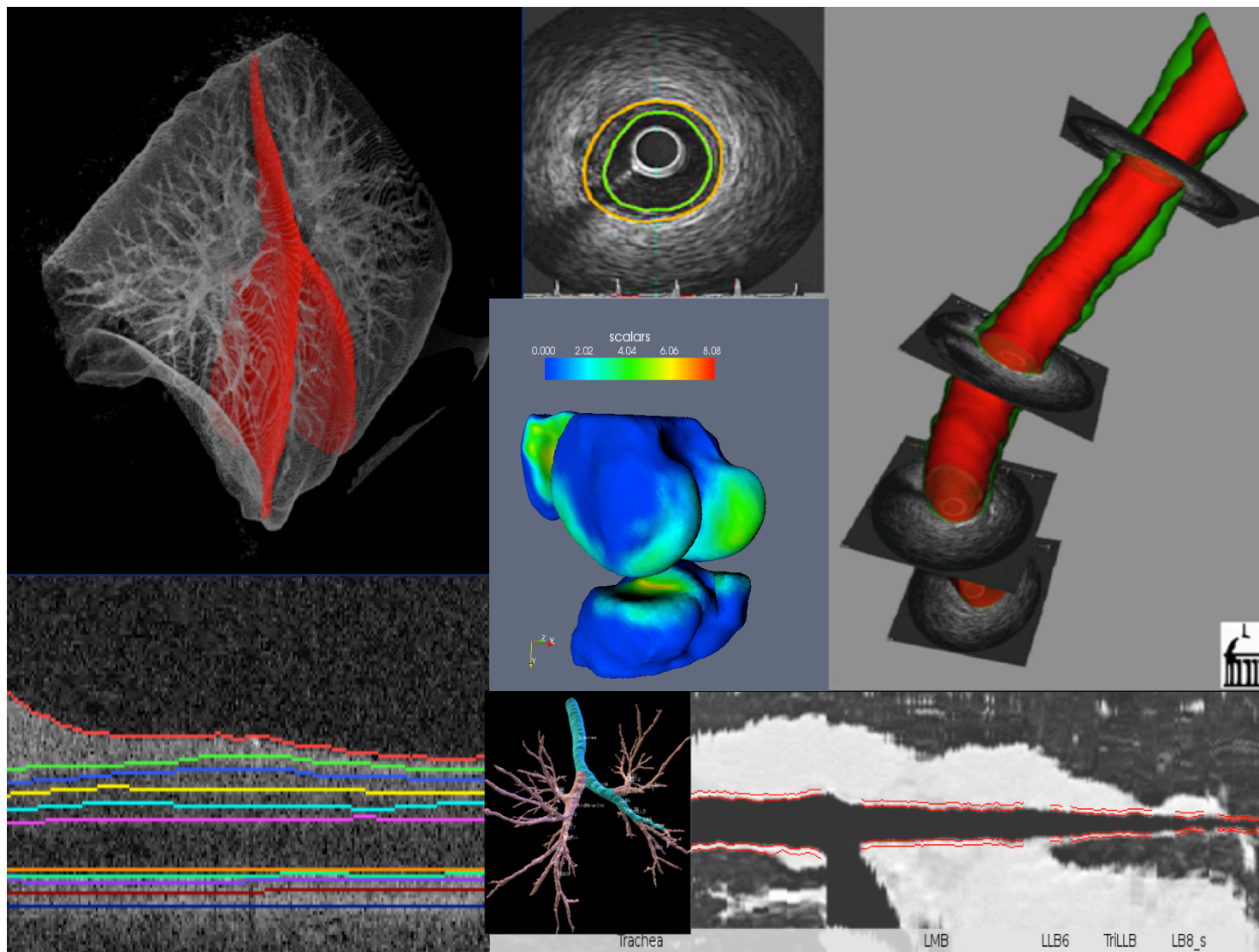


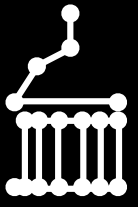
Financial support

- NIH grants NIBIB R01-EB004640, NEI R01-EY017066, NEI R01-EY018853
- U.S. Department of Veteran Affairs and the VA Center to Prevent and Treat Visual Loss
- Research to Prevent Blindness
- Marlene S. and Leonard A. Hadley Glaucoma Research Fund

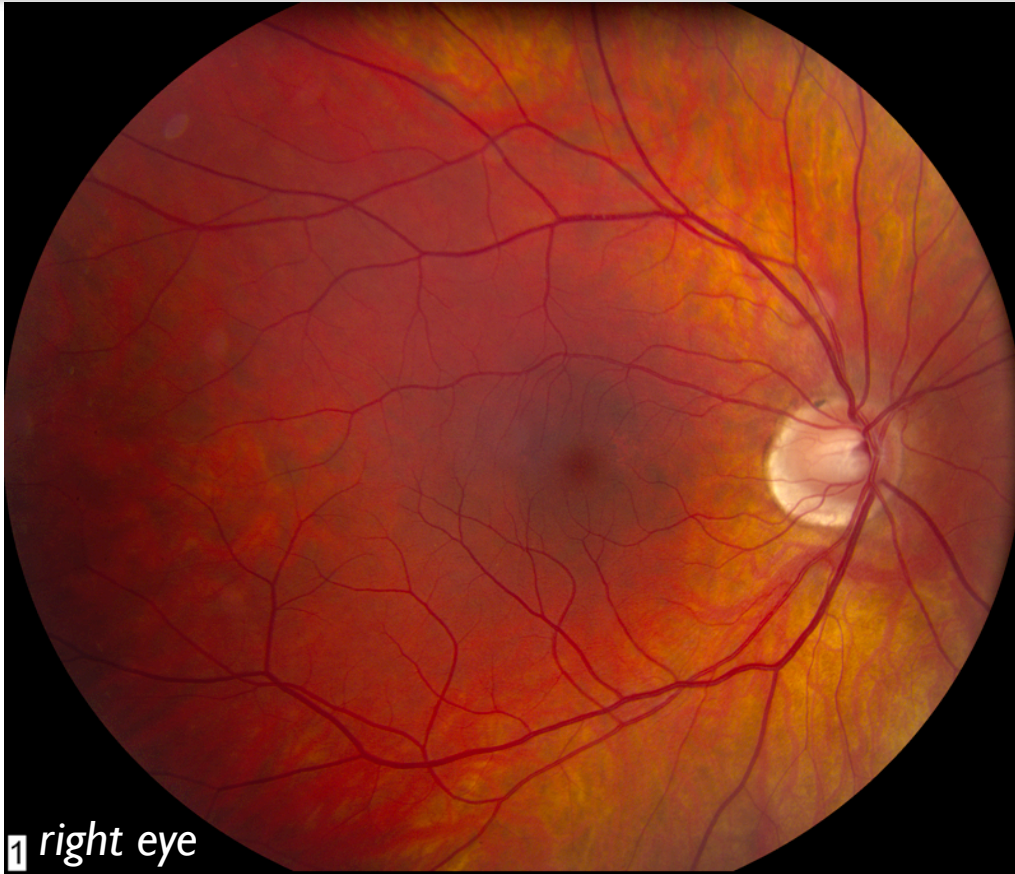


Motivation: need for 3D segmentation algorithms



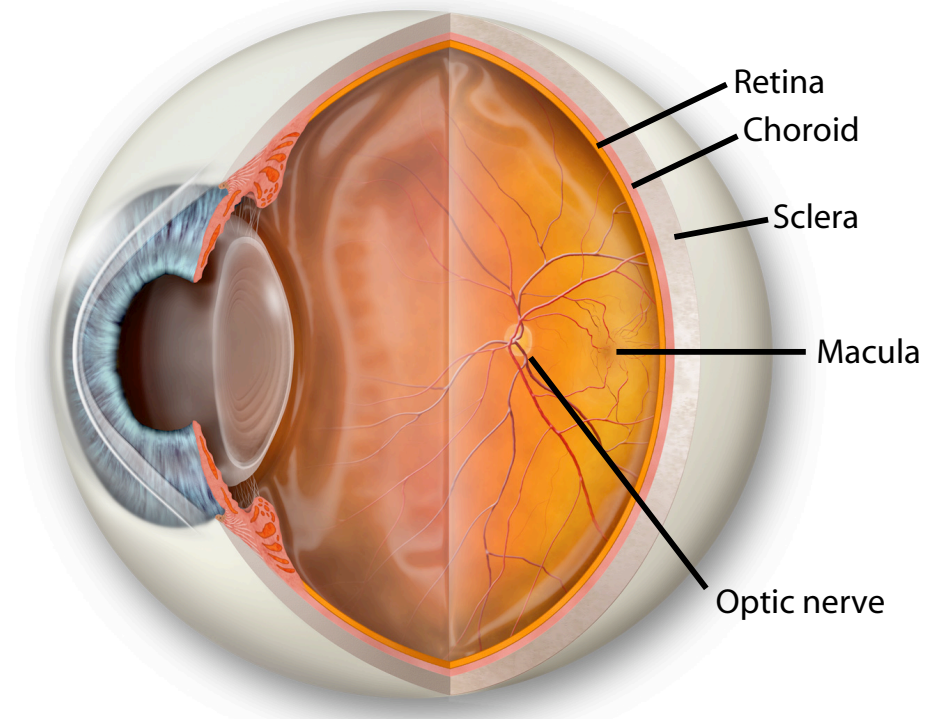


Example: ophthalmology is experiencing a recent shift towards use of 3D imaging modalities

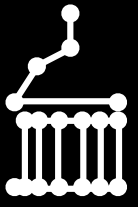


2D

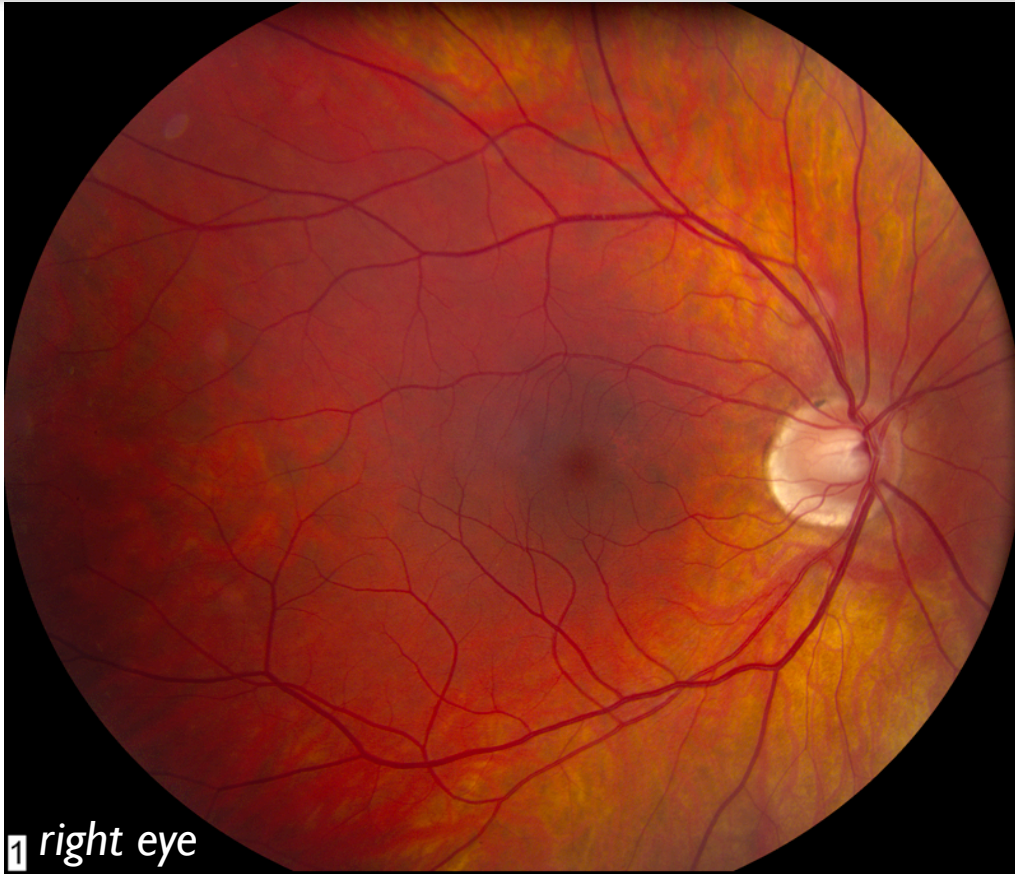
fundus photography



left eye



Example: ophthalmology is experiencing a recent shift towards use of 3D imaging modalities



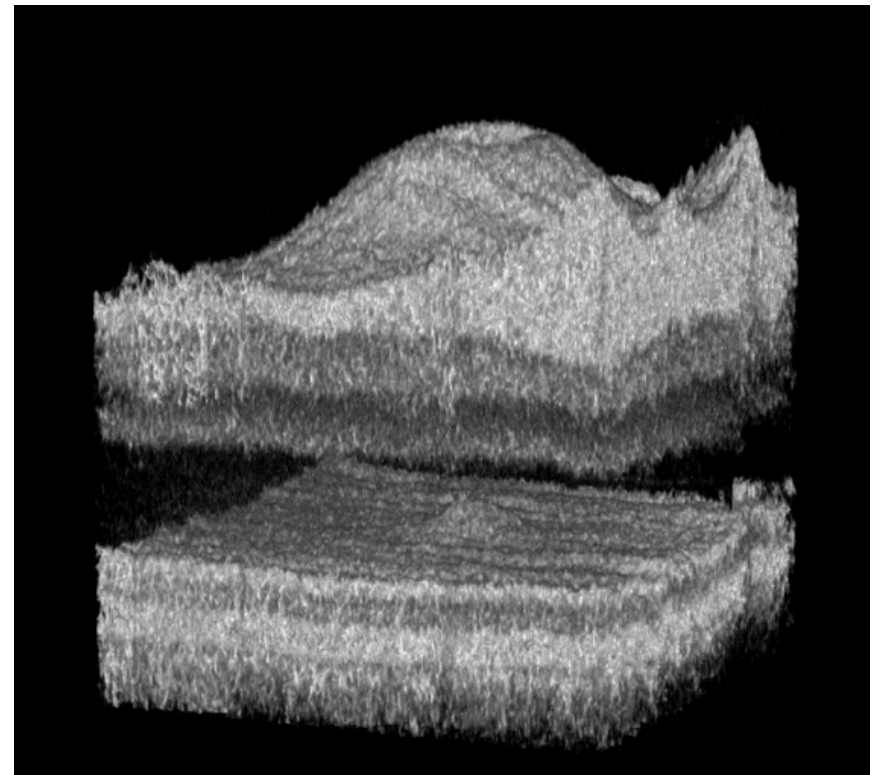
1 right eye

2D

fundus photography

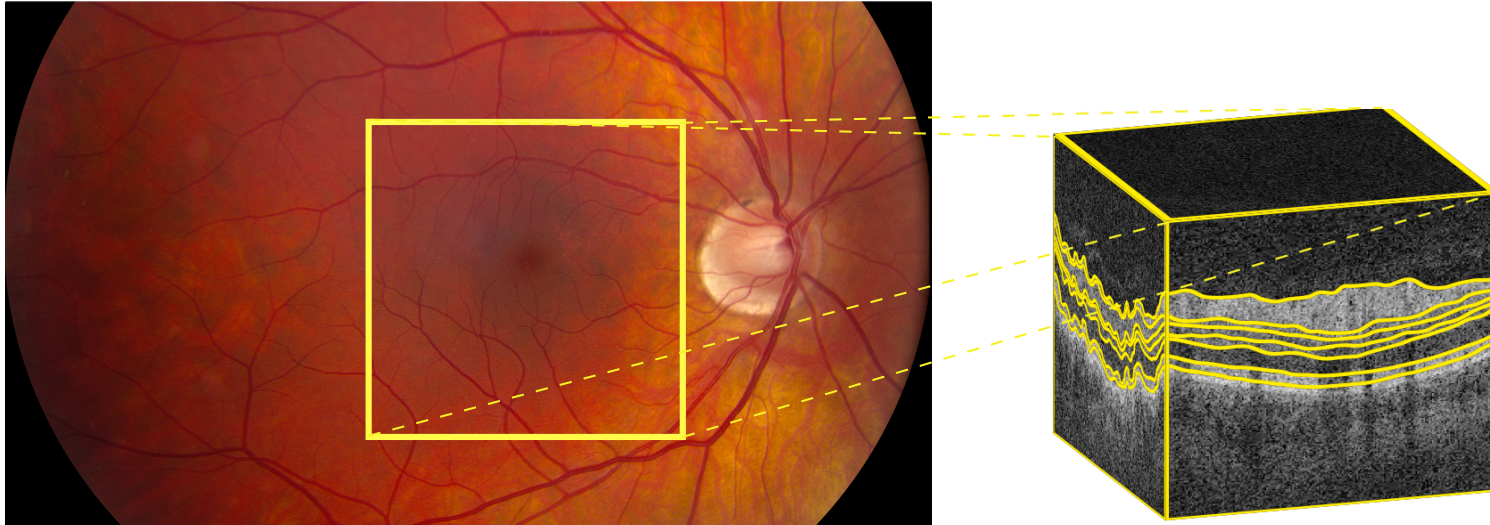
optical coherence tomography

3D

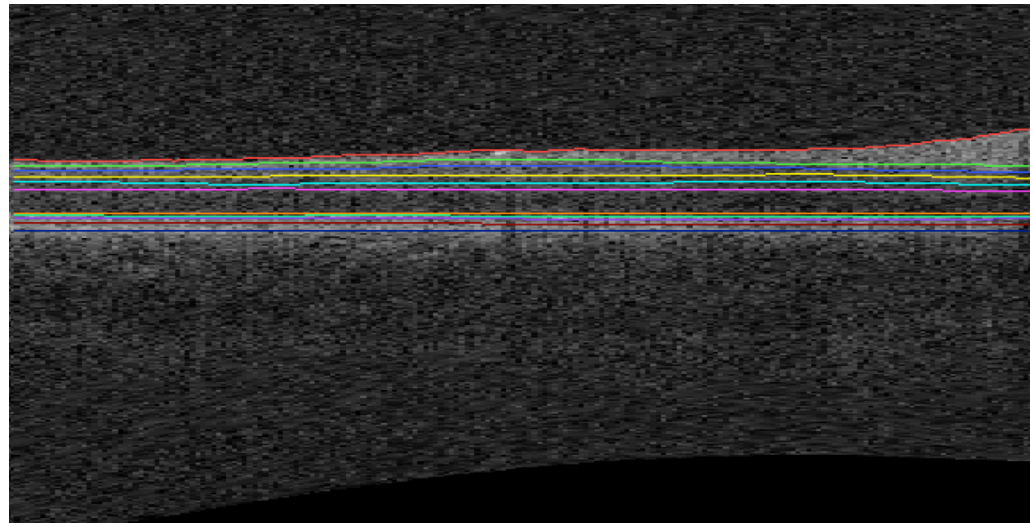




Segmenting the 3D retinal layers is important for diagnosing and managing a variety of diseases causing blindness, such as glaucoma

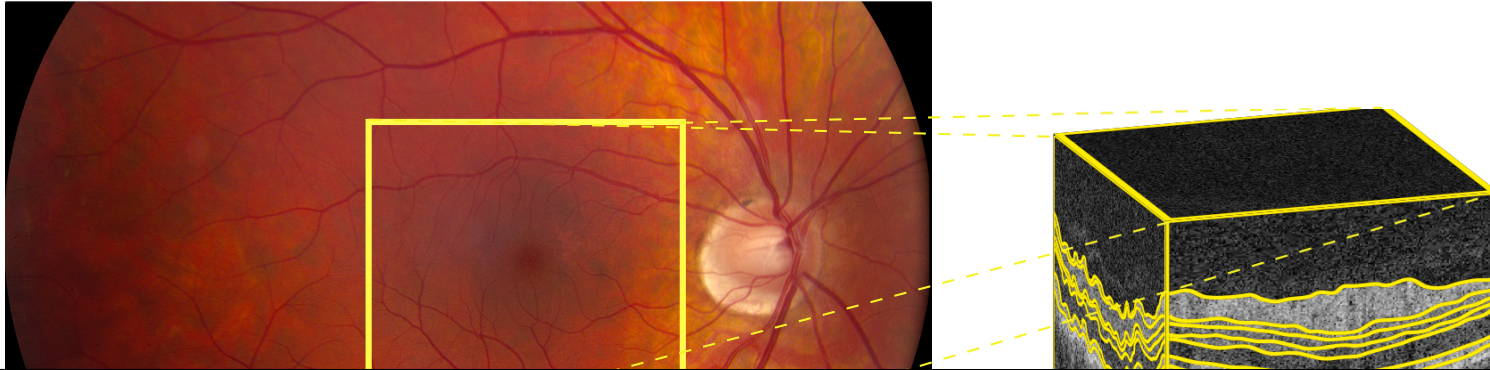


7-11
surfaces!



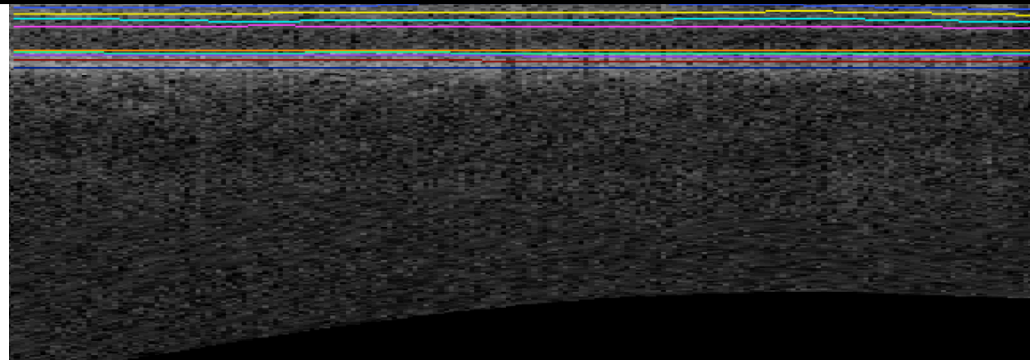


Segmenting the 3D retinal layers is important for diagnosing and managing a variety of diseases causing blindness, such as glaucoma



One challenge: developing algorithms that can **efficiently** take advantage of **3D** (or nD) contextual information and **simultaneously** find the surfaces of interacting objects

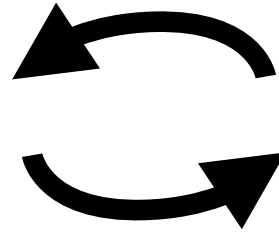
7-11
surfaces!





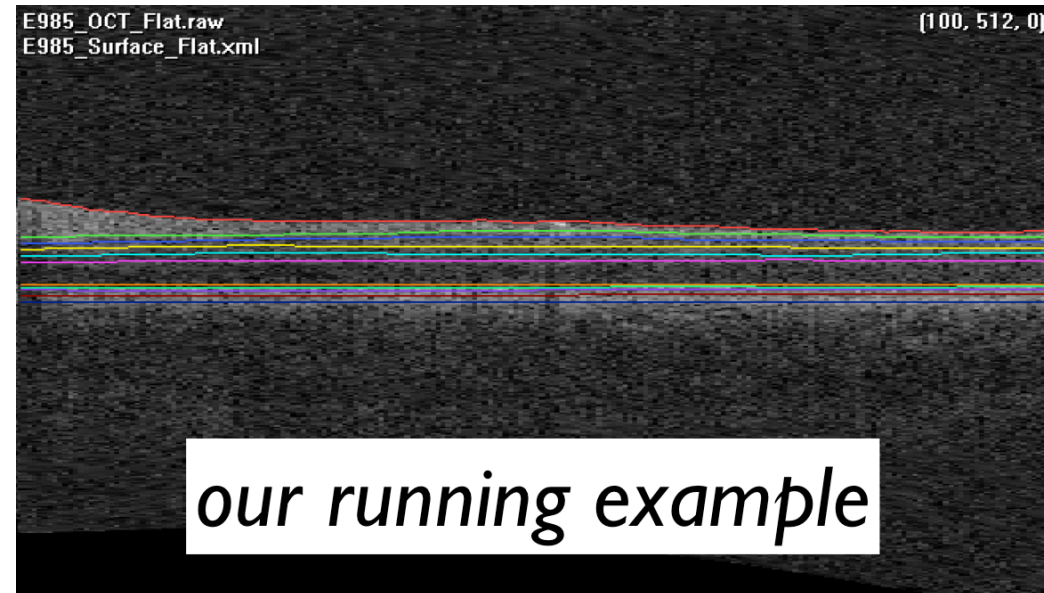
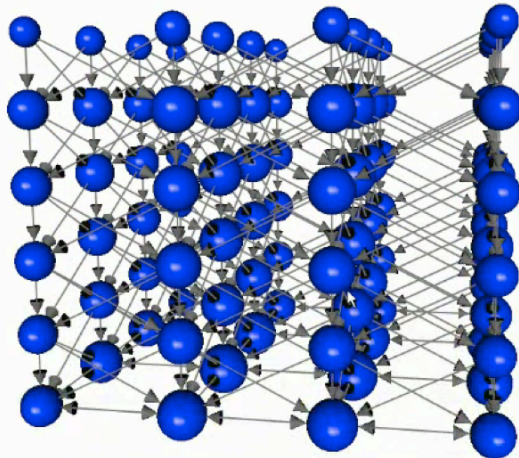
Outline

LOGISMOS
approach



Intraretinal layer
segmentation

*LOGISMOS = Layered Optimal Graph Image
Segmentation of Multiple Objects and Surfaces*

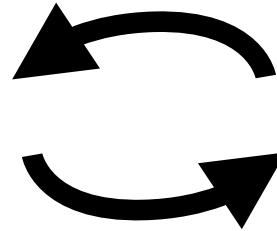


Other applications and future directions



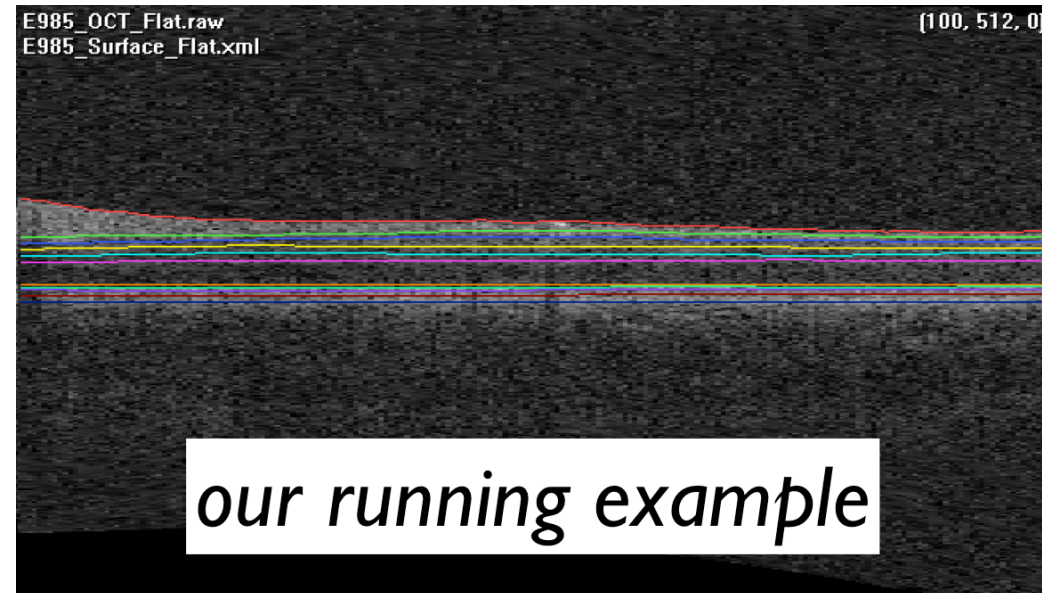
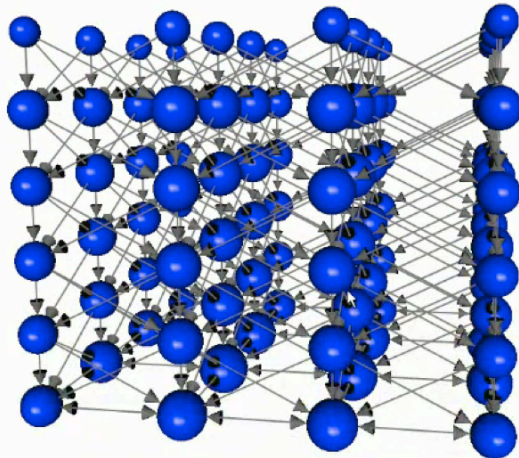
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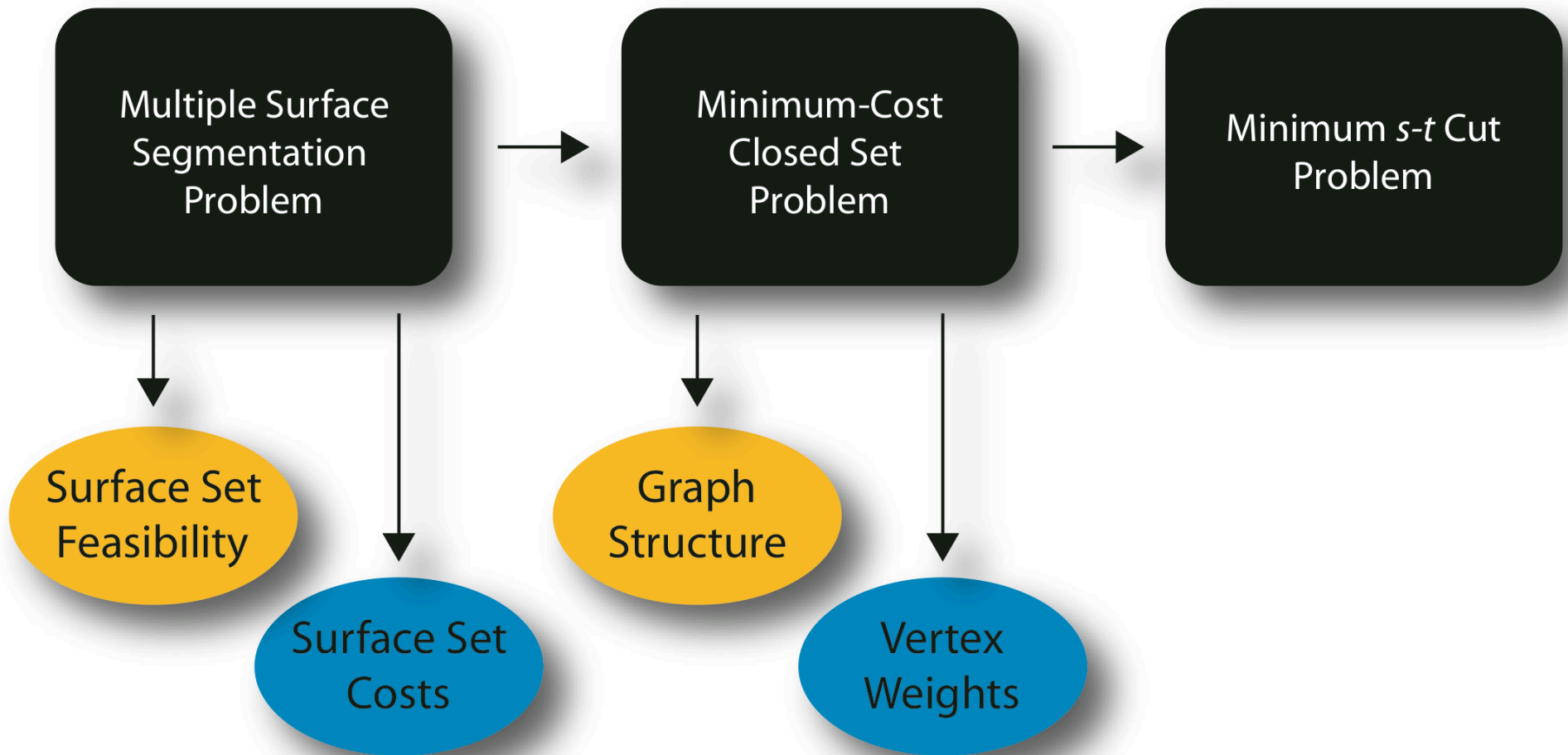


Other applications and future directions



LOGISMOS basics

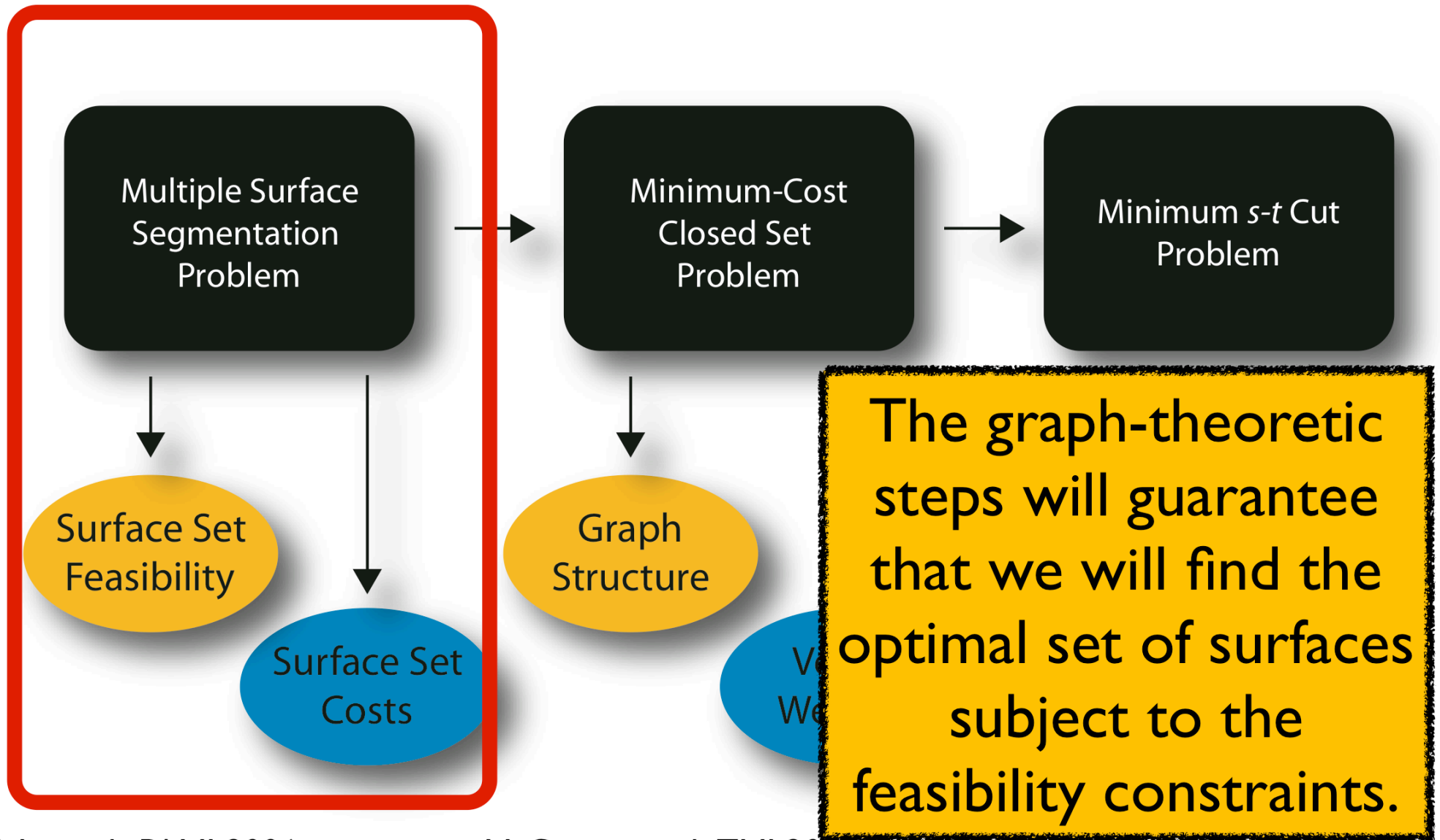
“Find the feasible set of surfaces that has the minimum cost.”



K. Li et al., PAMI 2006, extensions: M. Garvin et al., TMI 2009



The optimality of the LOGISMOS approach allows a user to focus on cost function design (without thinking about graphs)



K. Li et al., PAMI 2006, extensions: M. Garvin et al., TMI 2009

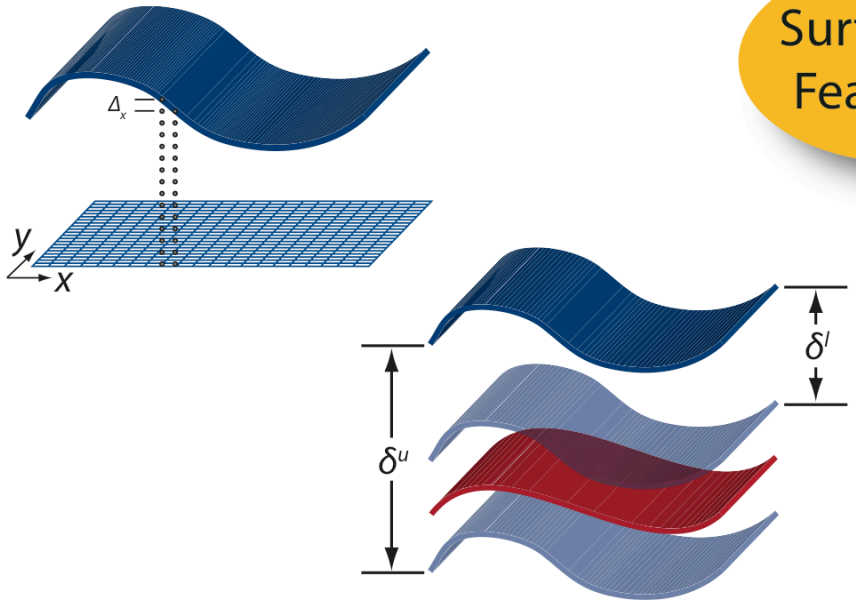


The multiple surface segmentation problem

Surface set feasibility

smoothness constraints

surface interaction constraints



Multiple Surface Segmentation Problem

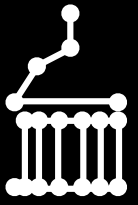
Surface Set Feasibility

Surface Set Costs

Surface set costs

on-surface costs

in-region costs

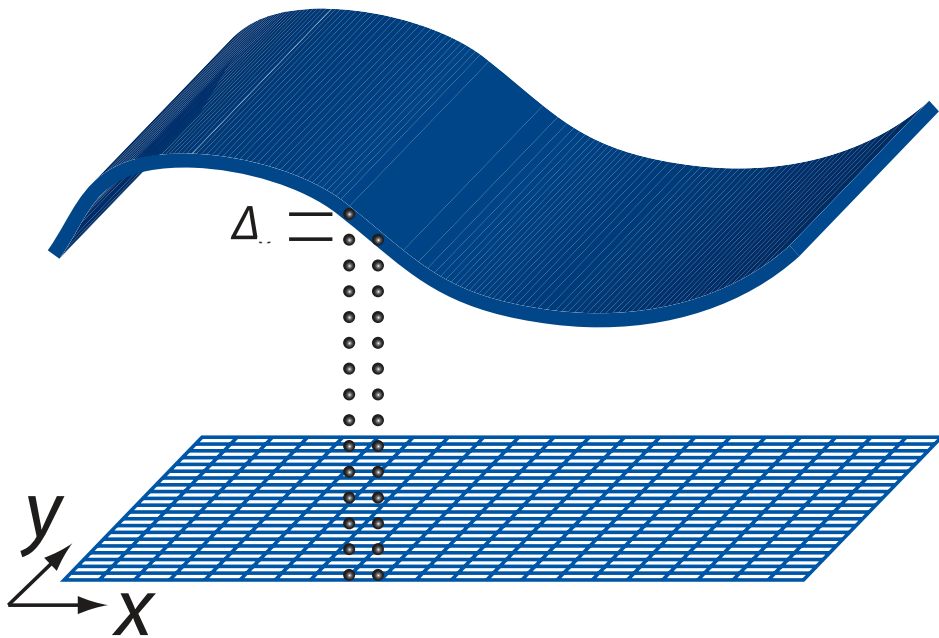


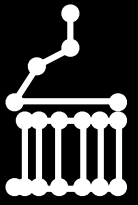
Surface set feasibility

Set of n surfaces: $\{f_1(x, y), \dots, f_n(x, y)\}$

Smoothness constraints

$$-\Delta_{\{(x_1, y_1), (x_2, y_2)\}}^u \leq f(x_1, y_1) - f(x_2, y_2) \leq \Delta_{\{(x_1, y_1), (x_2, y_2)\}}^l$$



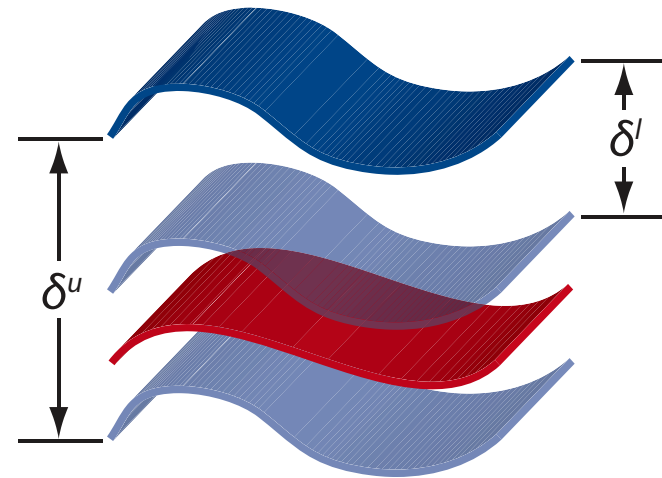
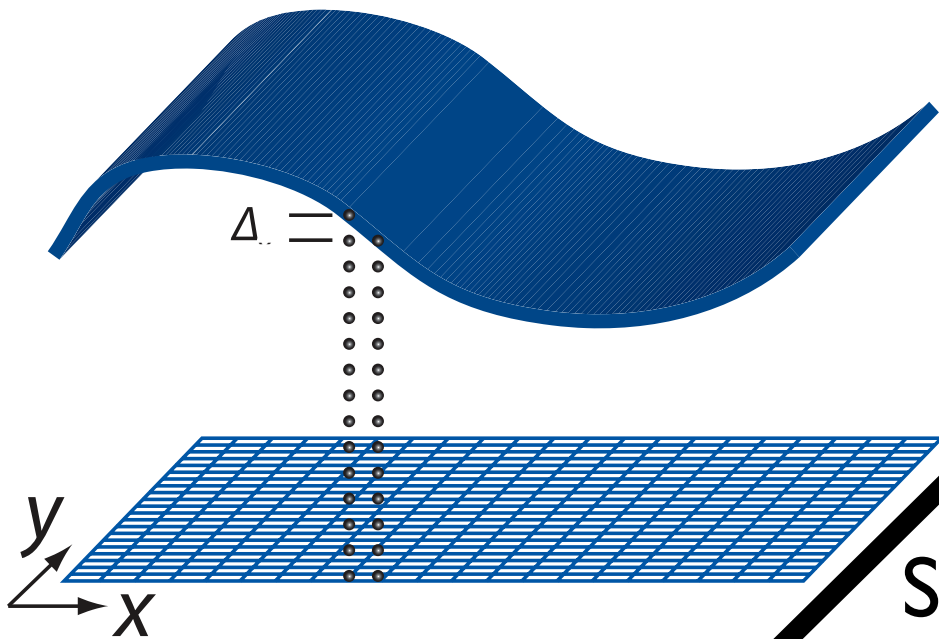


Surface set feasibility

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Smoothness constraints

$$-\Delta^u_{\{(x_1, y_1), (x_2, y_2)\}} \leq f(x_1, y_1) - f(x_2, y_2) \leq \Delta^l_{\{(x_1, y_1), (x_2, y_2)\}}$$



Surface interaction constraints

$$\delta^l(x, y) \leq f_i(x, y) - f_j(x, y) \leq \delta^u(x, y)$$

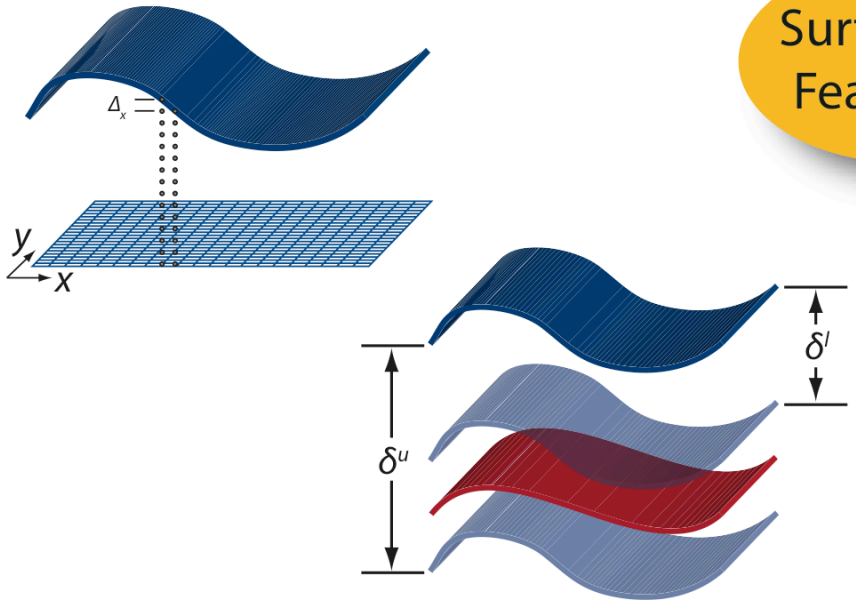


The multiple surface segmentation problem

Surface set feasibility

smoothness constraints

surface interaction constraints



Multiple Surface Segmentation Problem

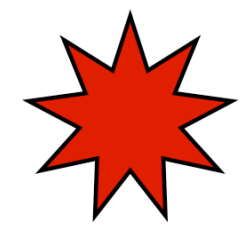
Surface Set Feasibility

Surface Set Costs

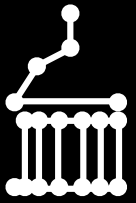
Surface set costs

on-surface costs

in-region costs



More on surface set costs...



Two categories of cost functions

n surfaces, $n+1$ regions

- **On-surface costs:** Each voxel has n on-surface costs corresponding to the unlikeliness of belonging to each surface.
- **In-region costs:** Each voxel has $n+1$ in-region costs corresponding to the unlikeliness of belonging to each region.



Cost function using both on-surface and in-region costs

Surface set cost function:

$$C_{\{f_1(x,y), f_2(x,y), \dots, f_n(x,y)\}} = \sum_{i=1}^n C_{f_i(x,y)} + \sum_{i=0}^n C_{R_i}$$



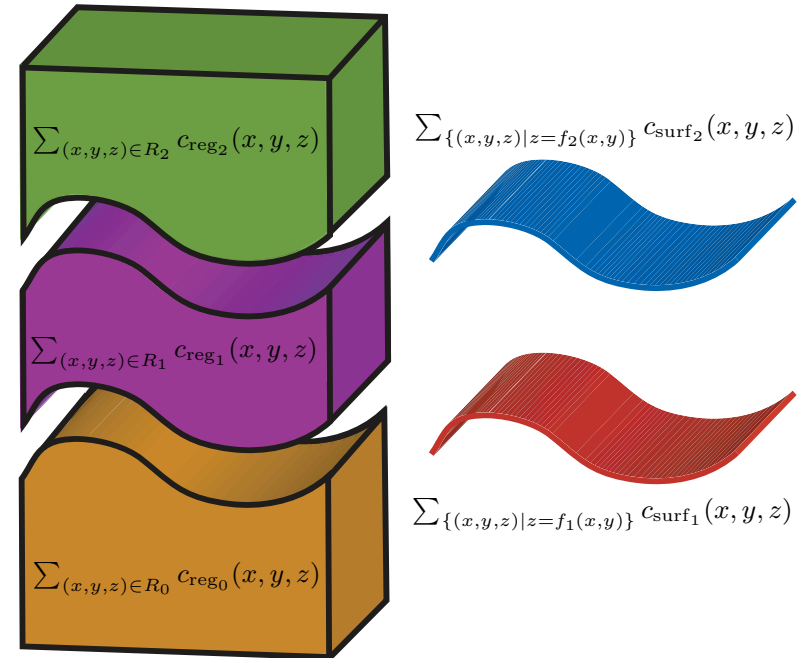
On-surface costs:

$$C_{f_i(x,y)} = \sum_{\{(x,y,z) | z=f_i(x,y)\}} c_{\text{surf}_i}(x,y,z)$$



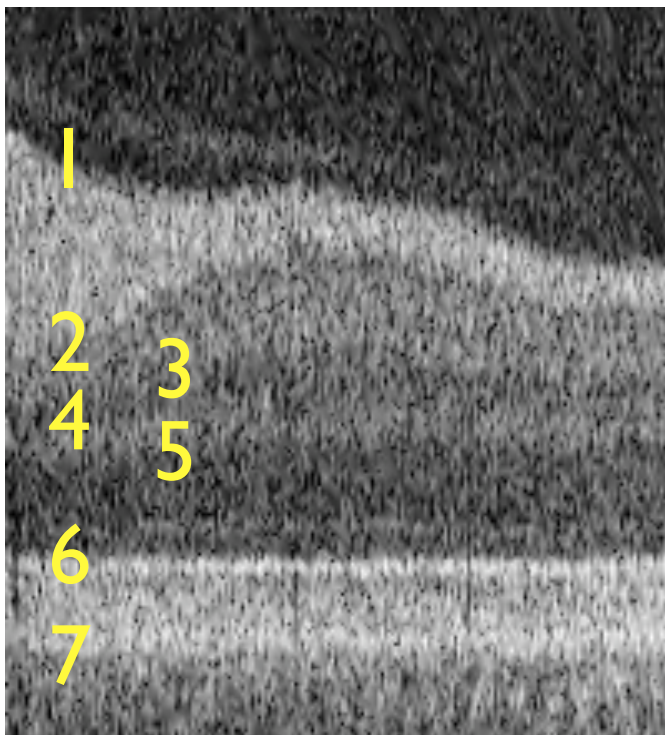
In-region costs:

$$C_{R_i} = \sum_{(x,y,z) \in R_i} c_{\text{reg}_i}(x,y,z)$$





Example using only on-surface costs

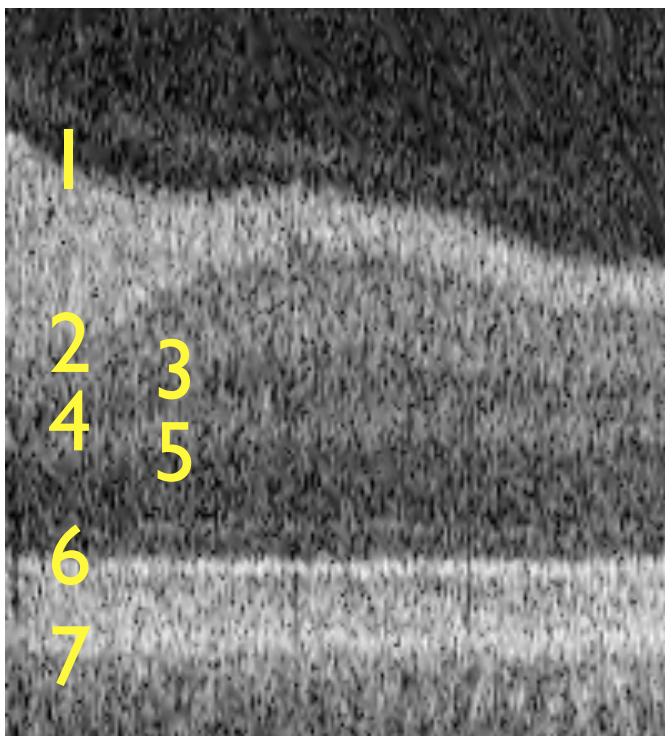


OCT Image

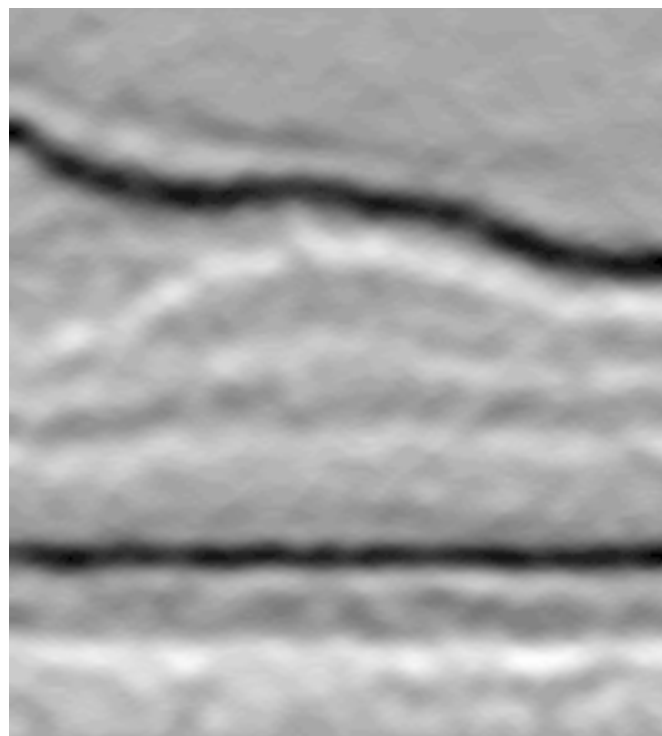
Find indicated 7 surfaces
in OCT image using only
on-surface costs.



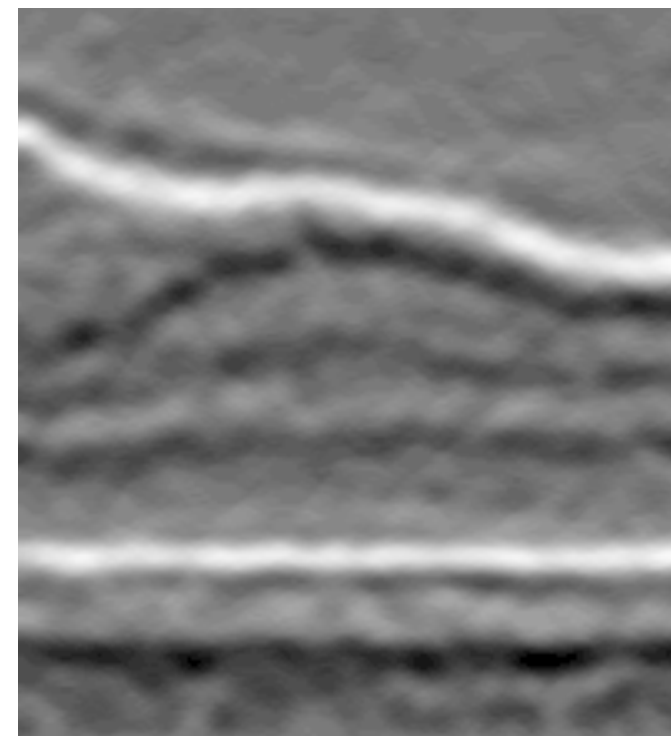
Example using only on-surface costs



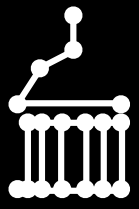
OCT Image



Encourage
dark-to-bright
transitions

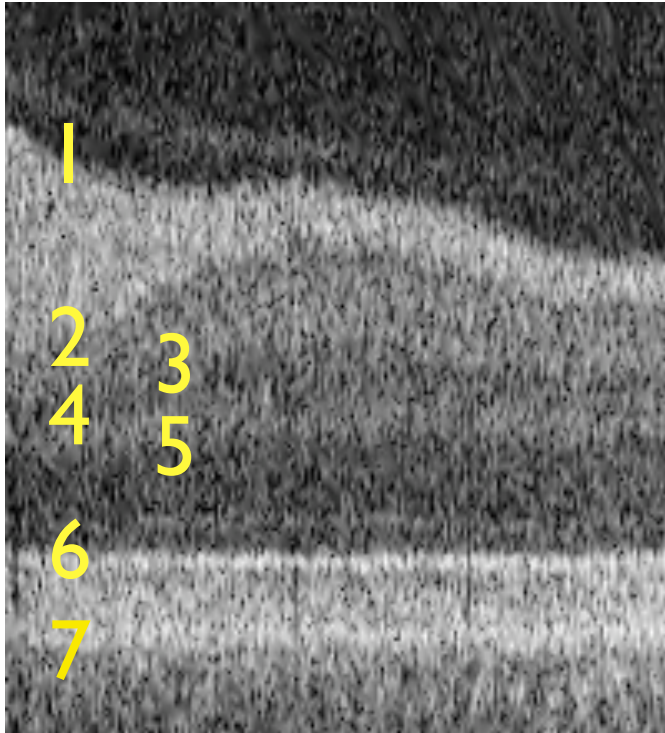


Encourage
bright-to-dark
transitions



Example using only on-surface costs

Seven cost images:

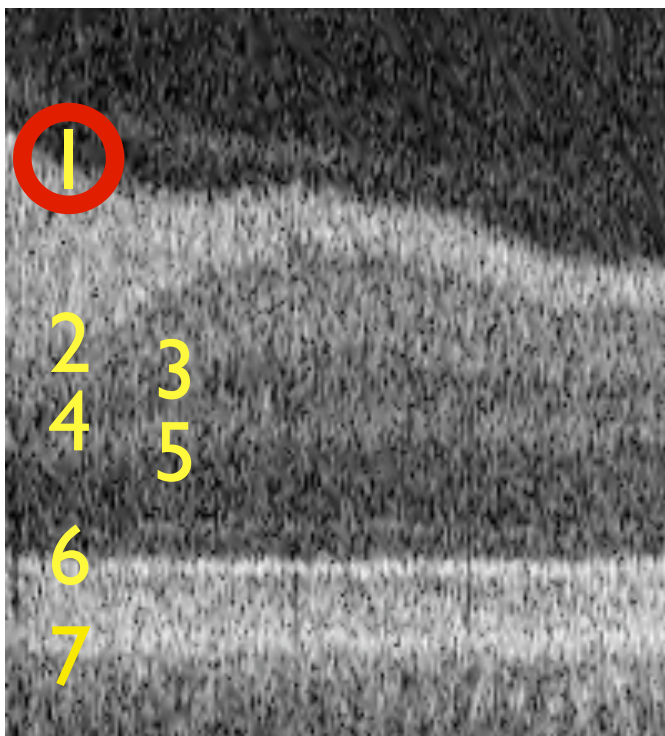


OCT Image

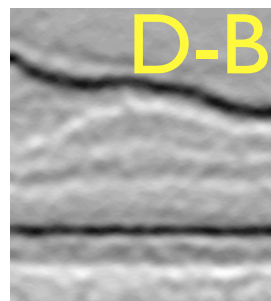


Example using only on-surface costs

Seven cost images:



OCT Image



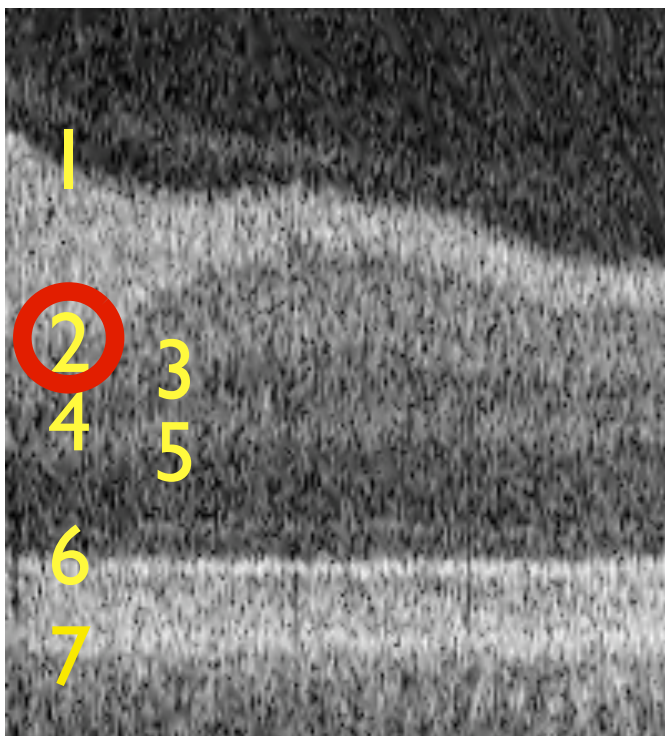
D-B



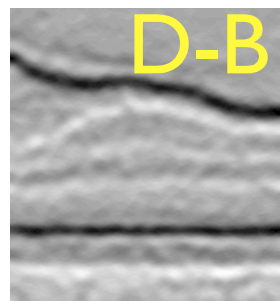


Example using only on-surface costs

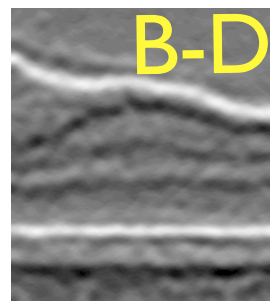
Seven cost images:



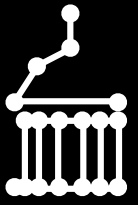
OCT Image



1

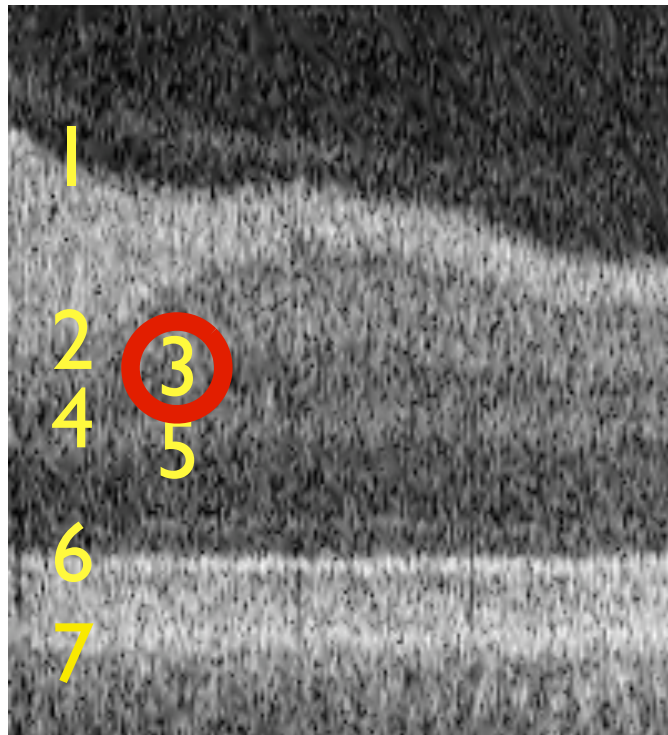


2

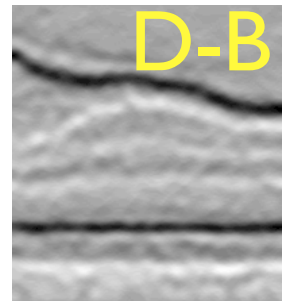


Example using only on-surface costs

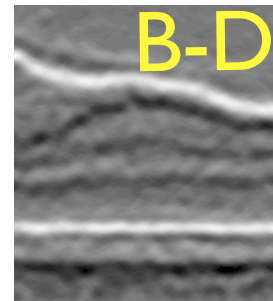
Seven cost images:



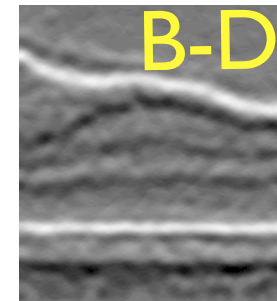
OCT Image



1



2

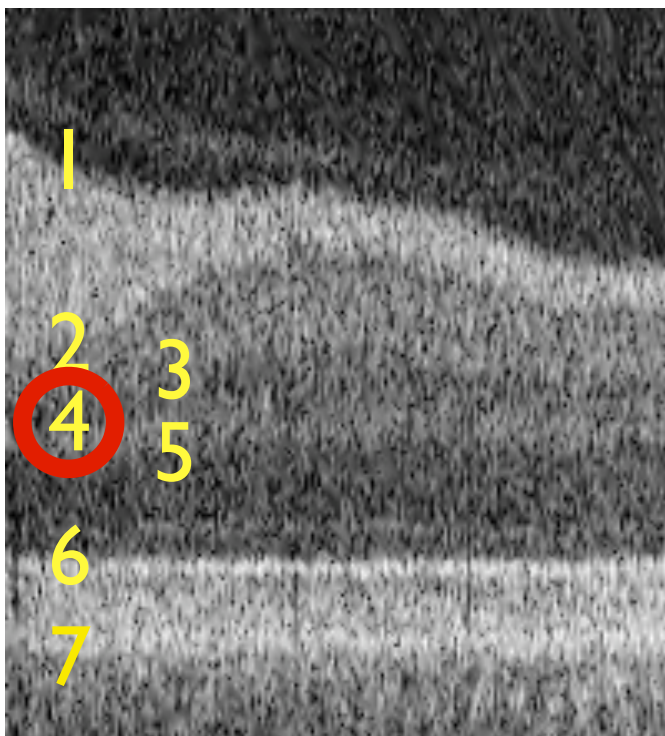


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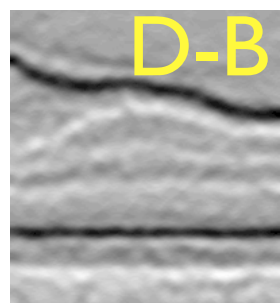


Example using only on-surface costs

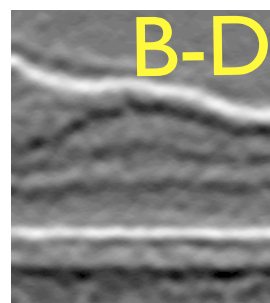
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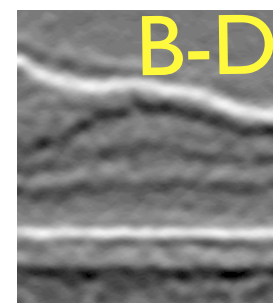
OCT Image



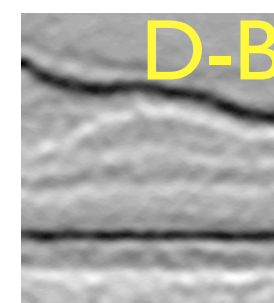
1



2



3

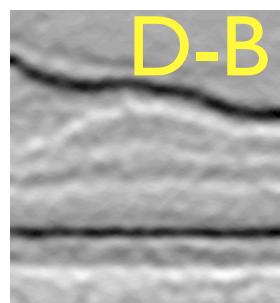
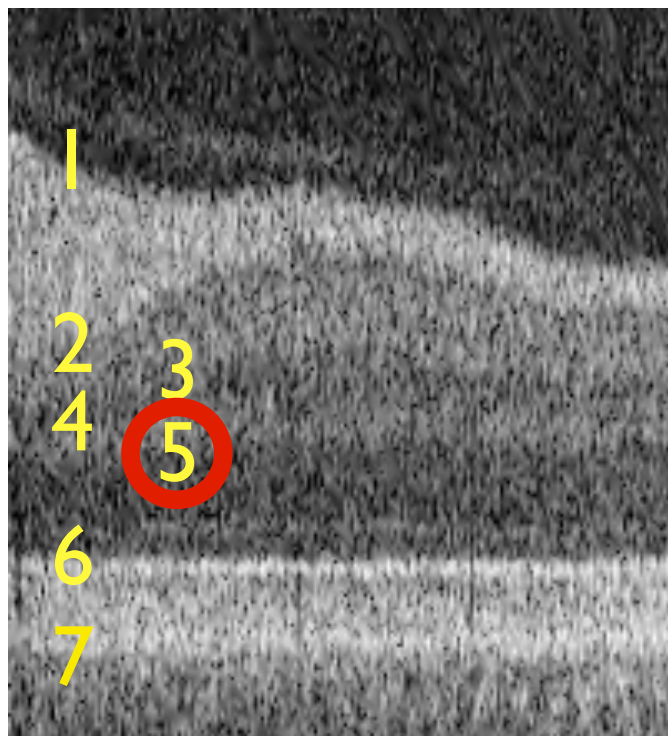


4

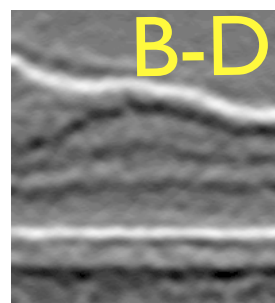


Example using only on-surface costs

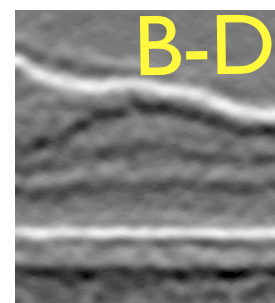
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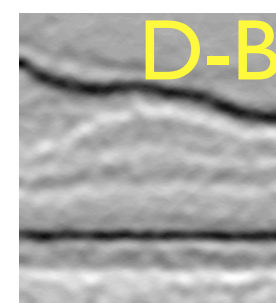
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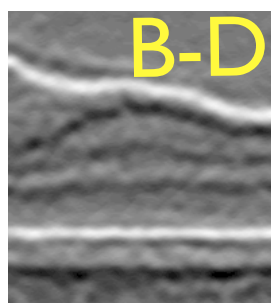
2



3



4

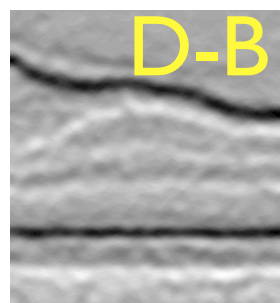
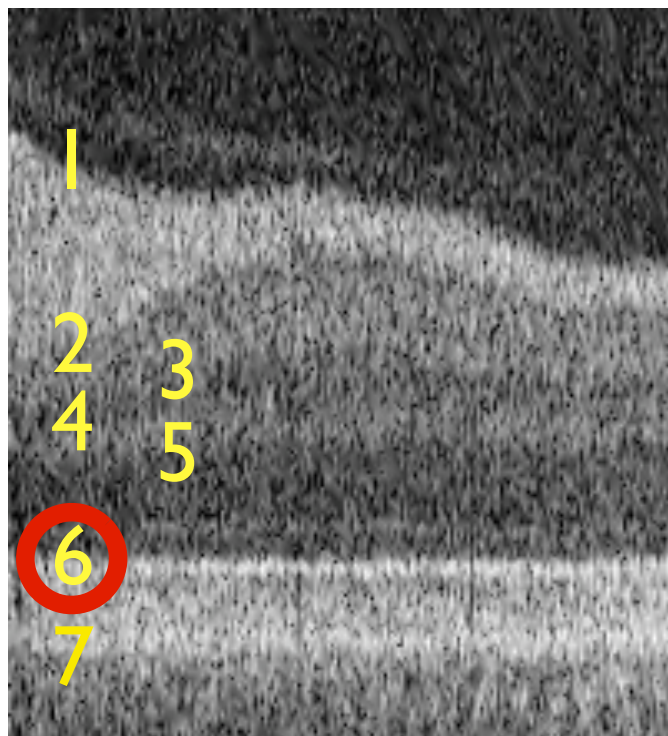


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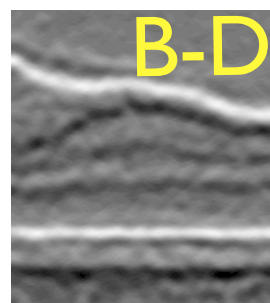


Example using only on-surface costs

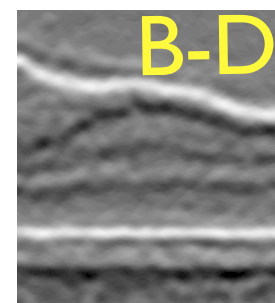
Seven cost images:



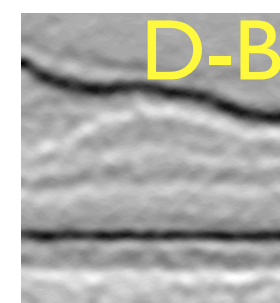
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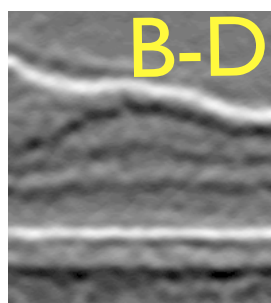
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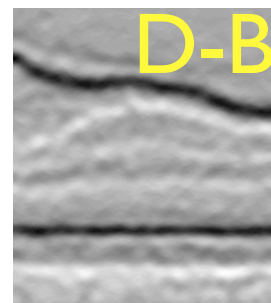
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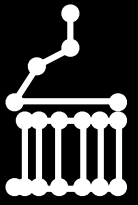
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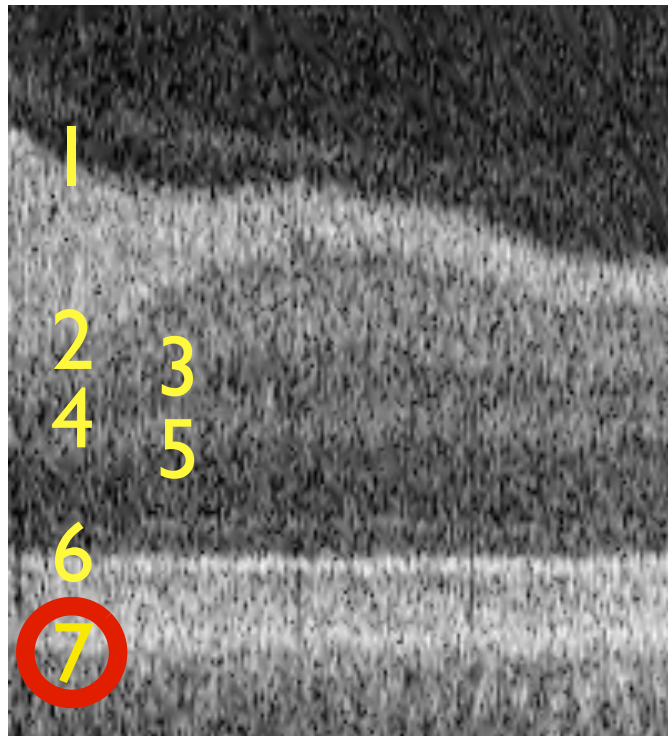


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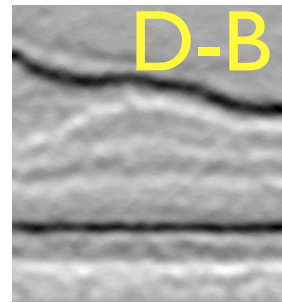


Example using only on-surface costs

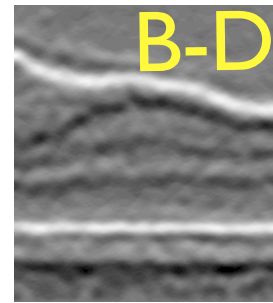
Seven cost images:



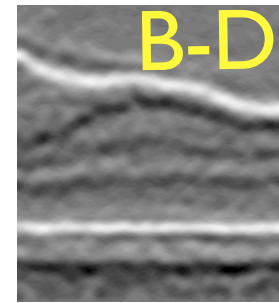
OCT Image



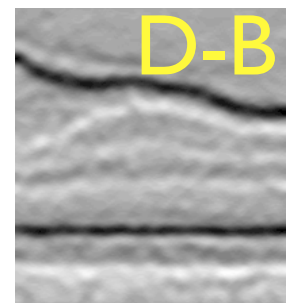
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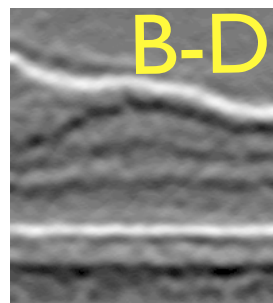
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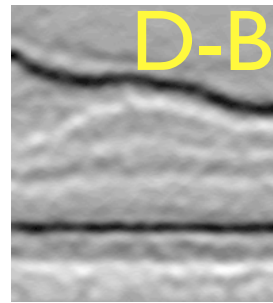
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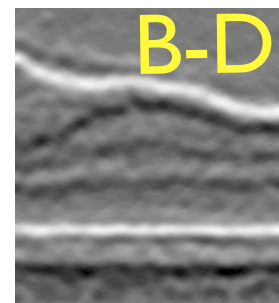
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5



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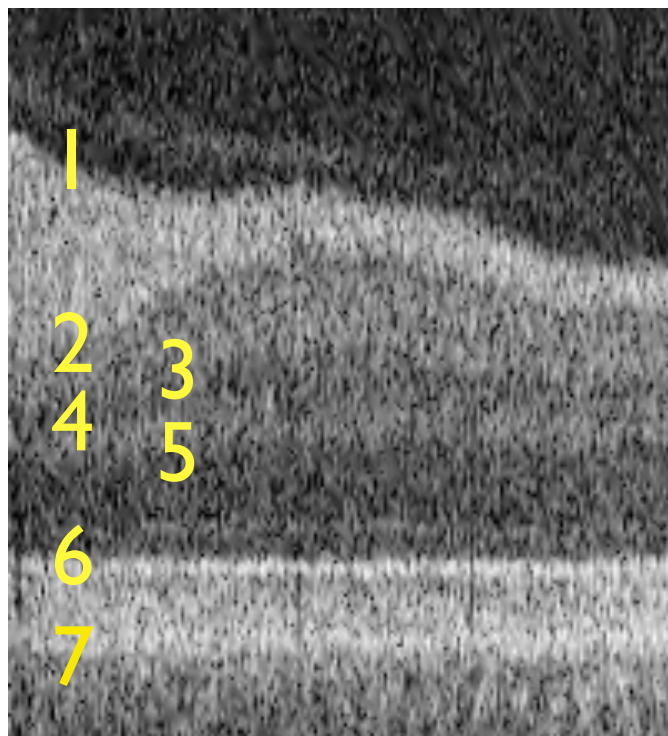


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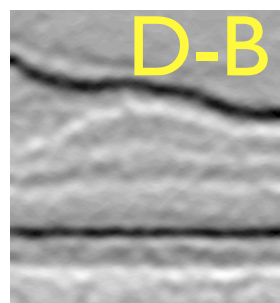


Example using only on-surface costs

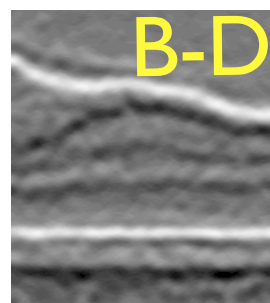
Seven cost images:



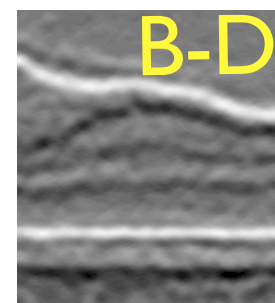
OCT Image



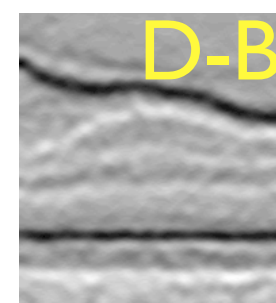
1



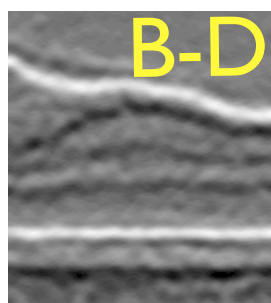
2



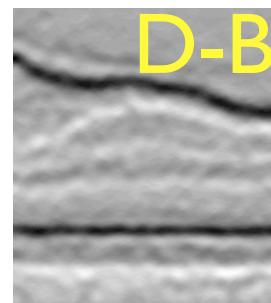
3



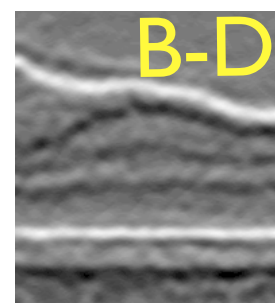
4



5

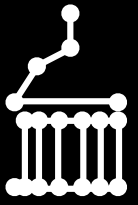


6



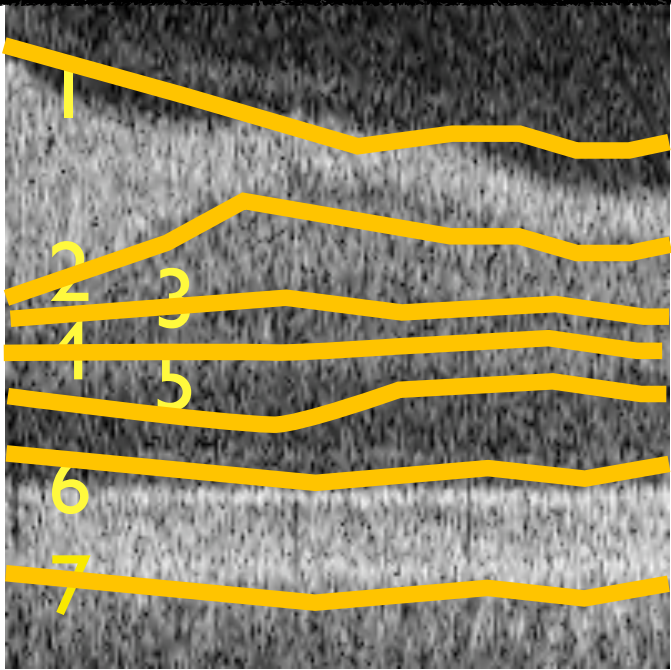
7

+ smoothness constraints
+ thickness constraints



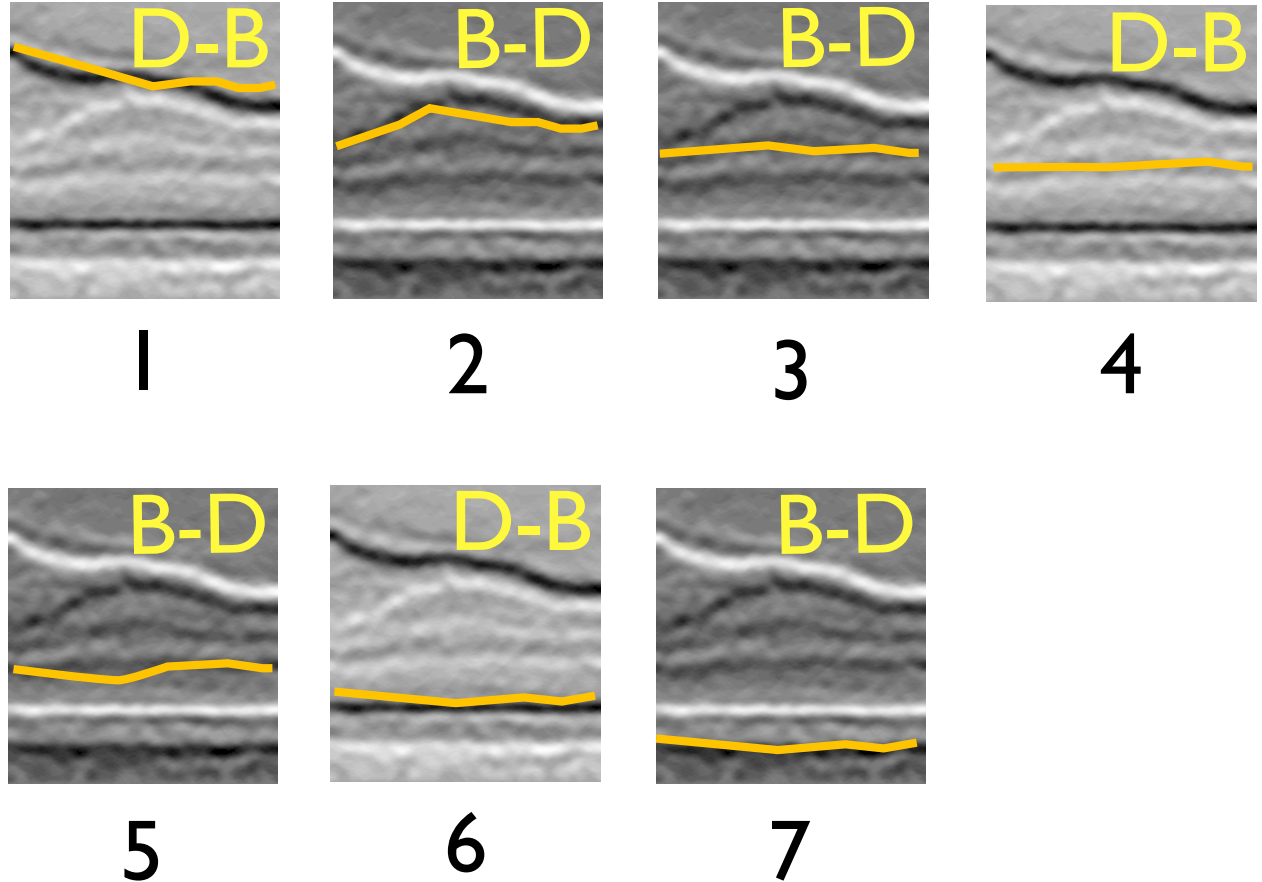
Example using only on-surface costs

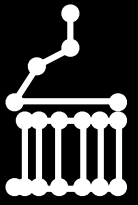
A non-optimal set of surfaces



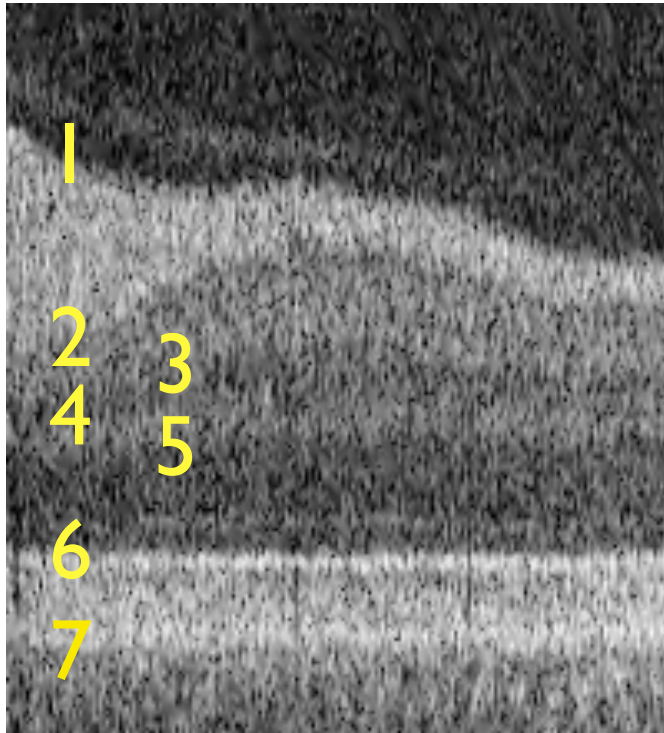
OCT Image

Seven cost images:

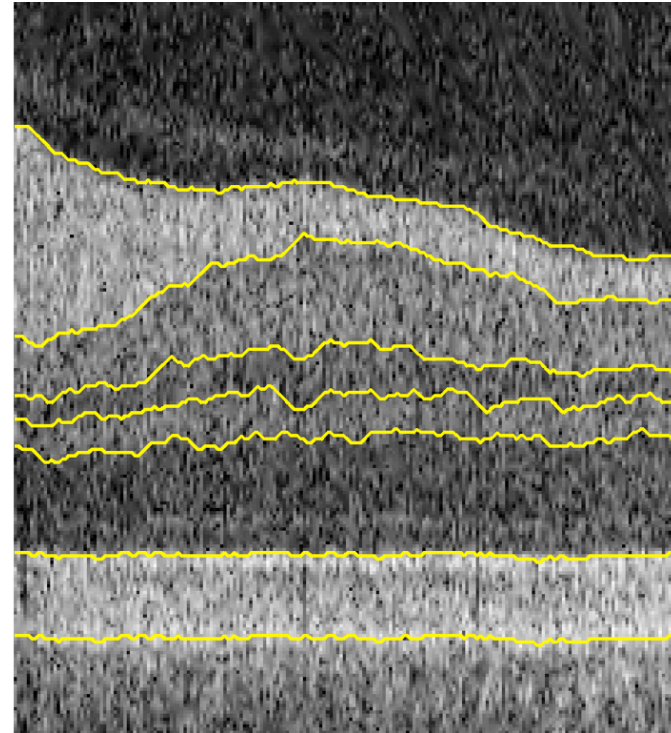




Example using only on-surface costs



OCT Image

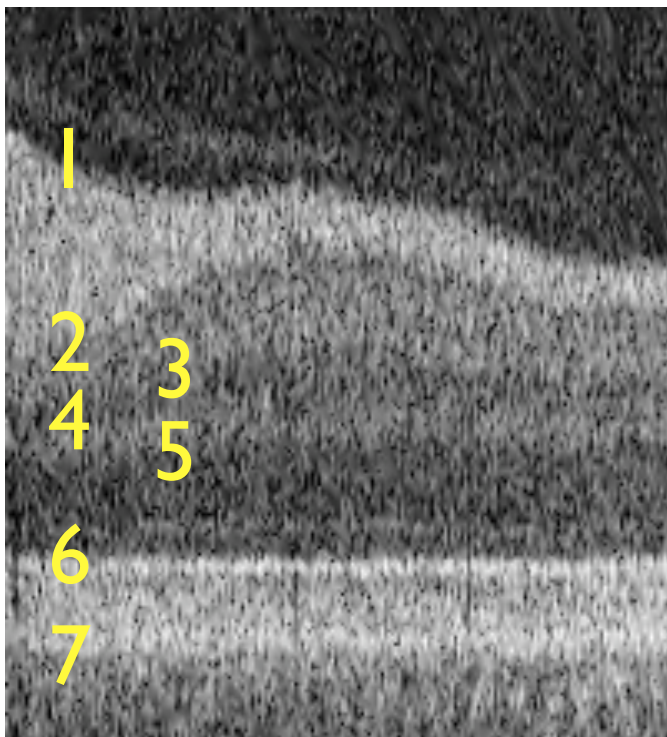


Result

The optimal set of surfaces

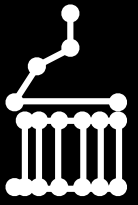


Example using only in-region costs



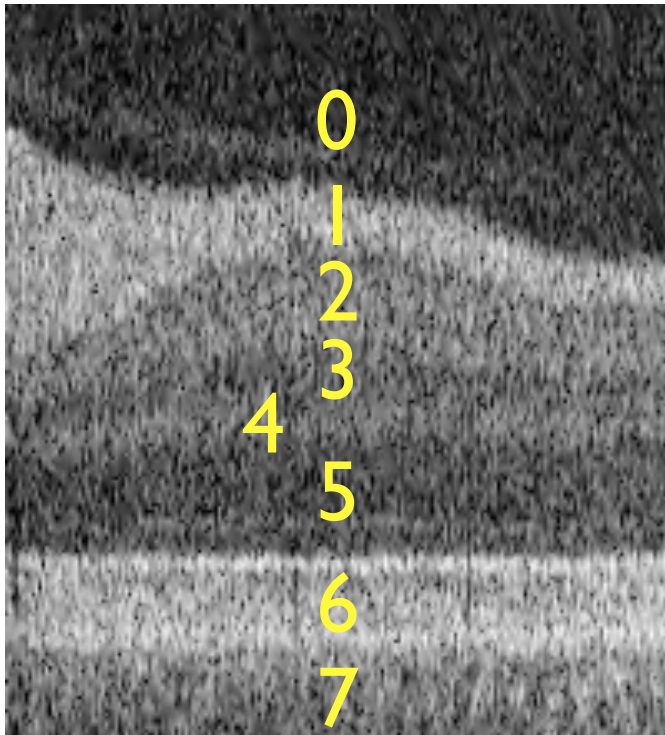
OCT Image

Find indicated 7 surfaces
in OCT image using only
in-region costs.

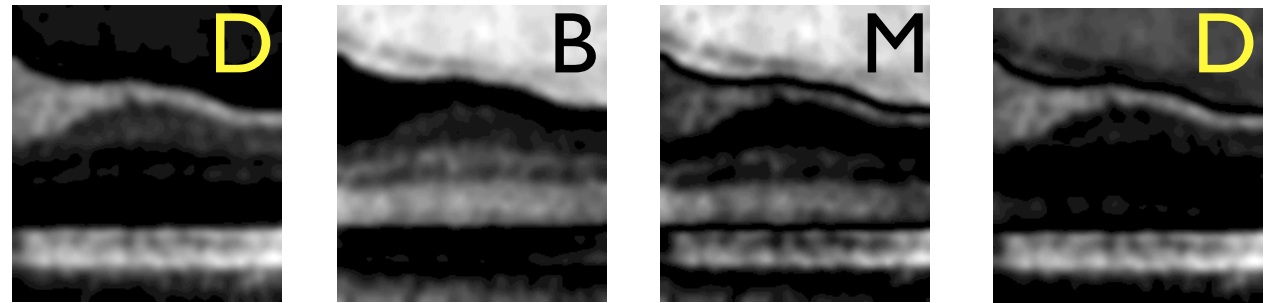


Example using only in-region costs

Eight cost images:



OCT Image
(labeled regions)

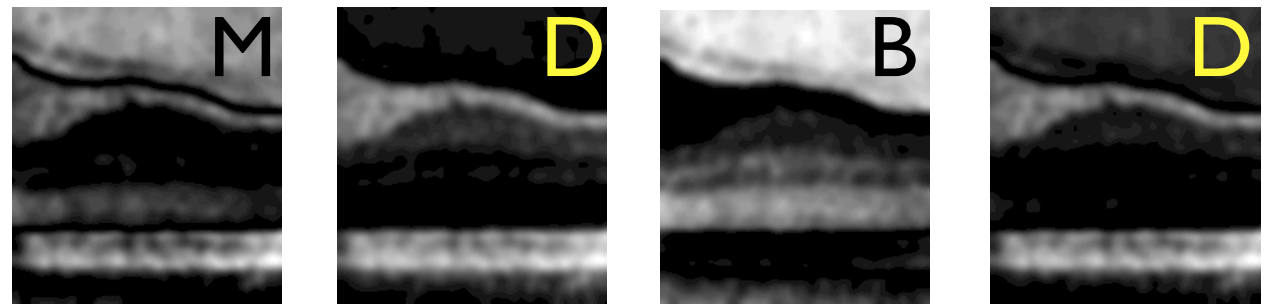


0

1

2

3



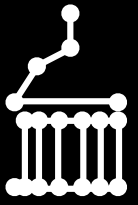
4

5

6

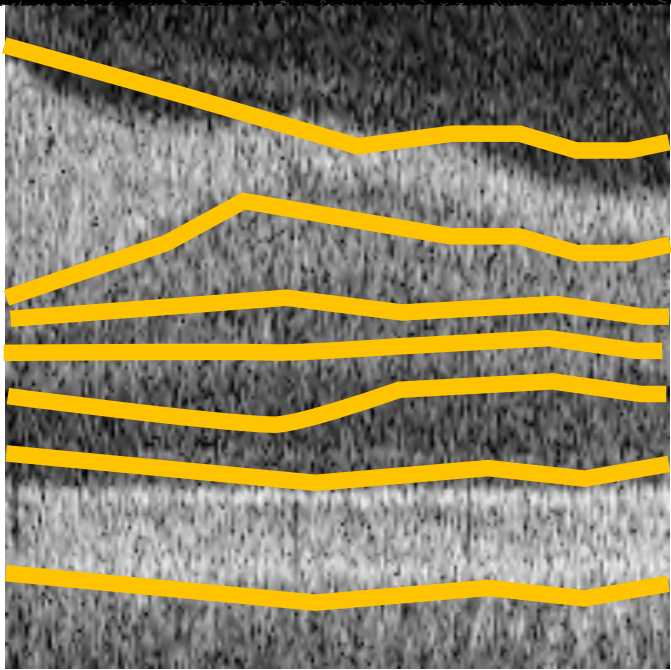
7

+ smoothness constraints
+ thickness constraints



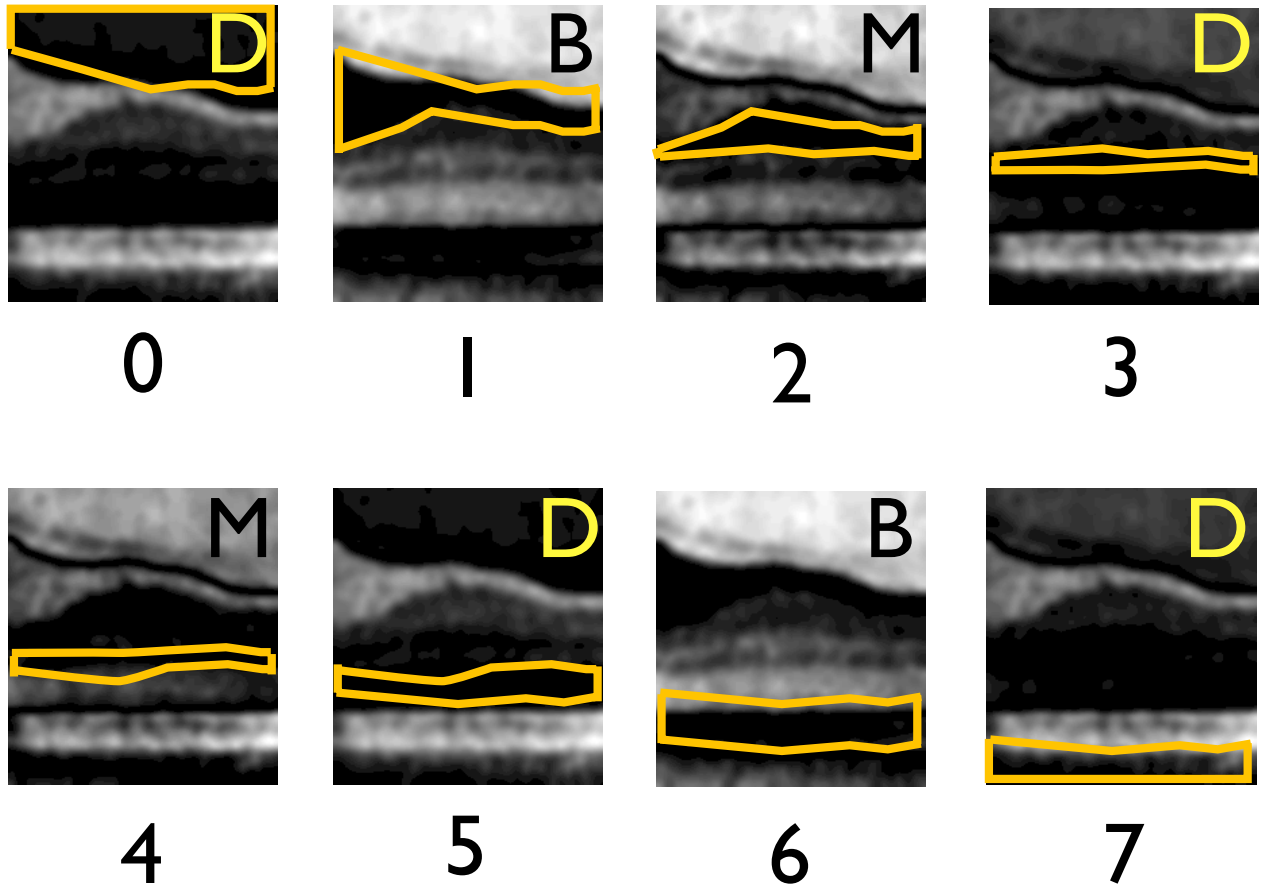
Example using only in-region costs

A non-optimal set of surfaces



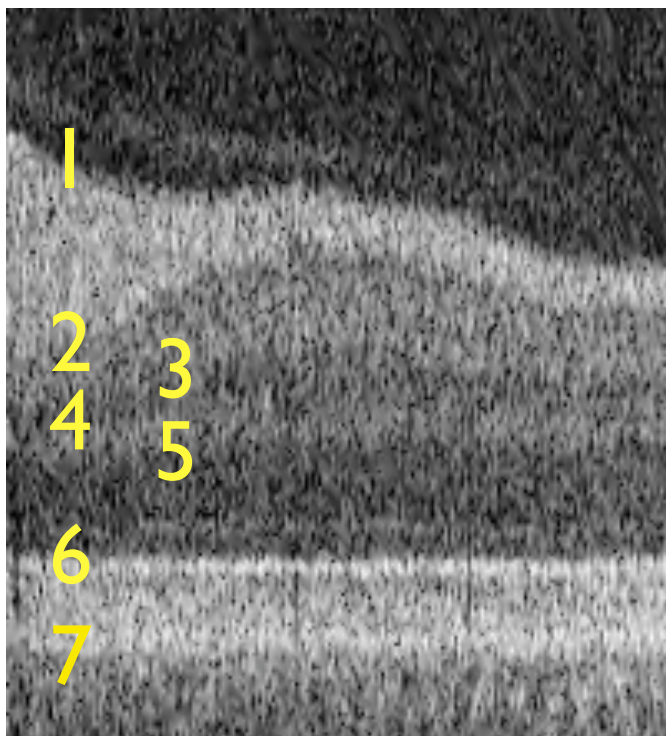
OCT Image

Eight cost images:

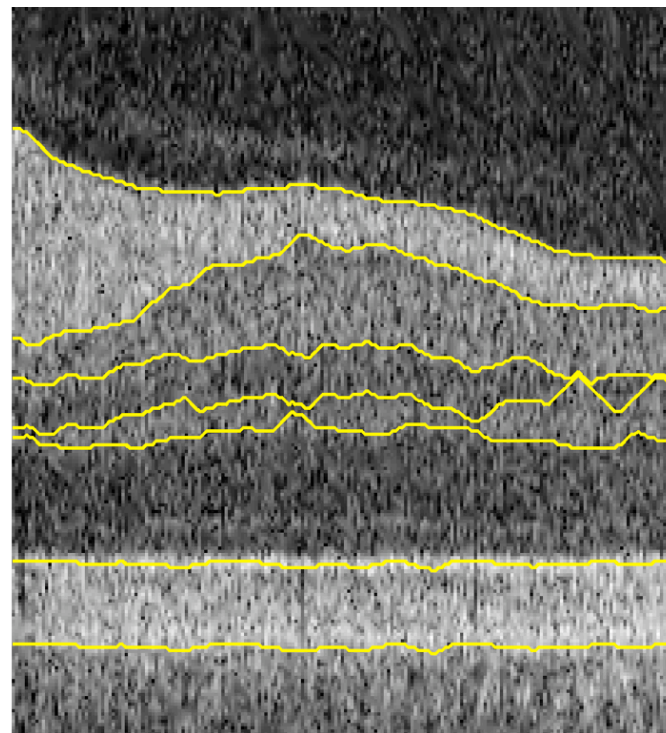




Example using only in-region costs

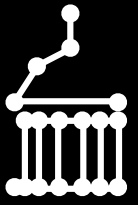


OCT Image

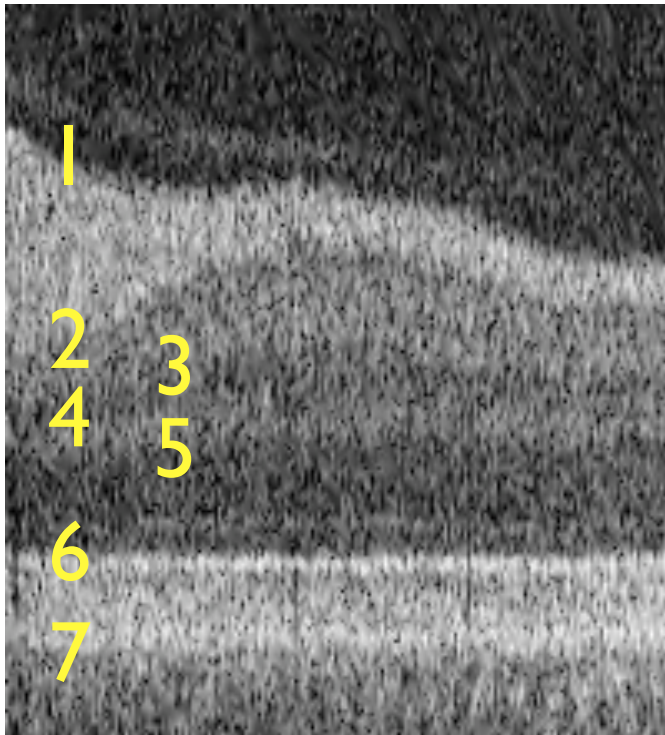


Result

The optimal set of surfaces

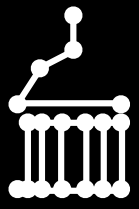


Example using both on-surface and in-region costs



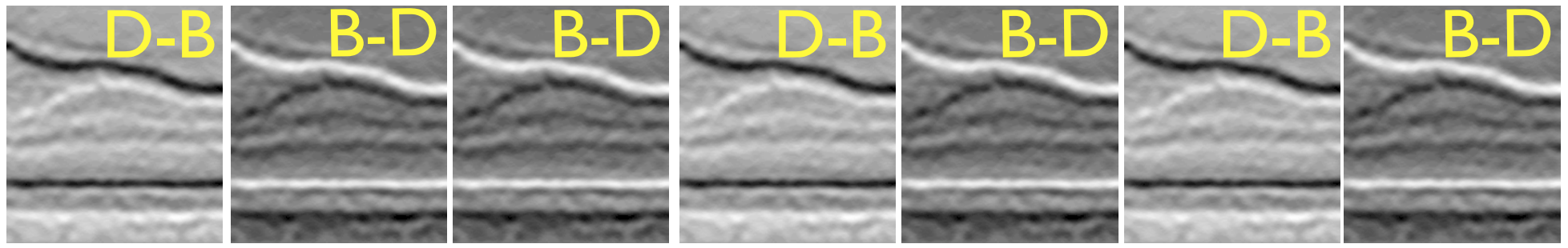
OCT Image

Find indicated 7 surfaces in OCT image using both on-surface and in-region costs.

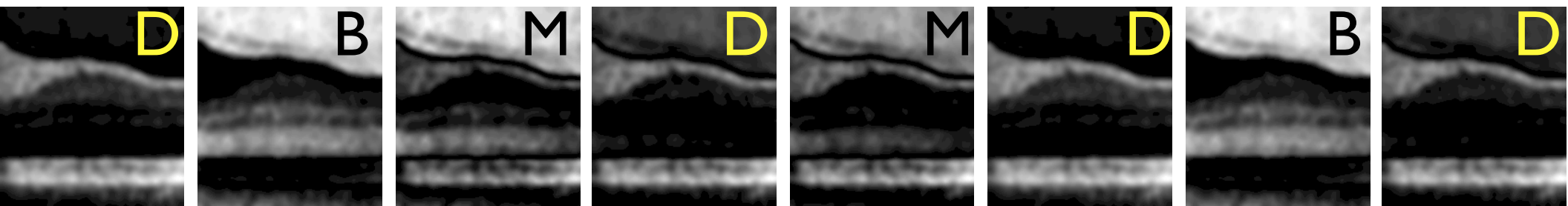


Example using both on-surface and in-region costs

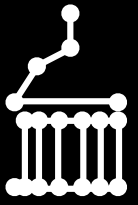
Seven on-surface and eight in-region cost images:



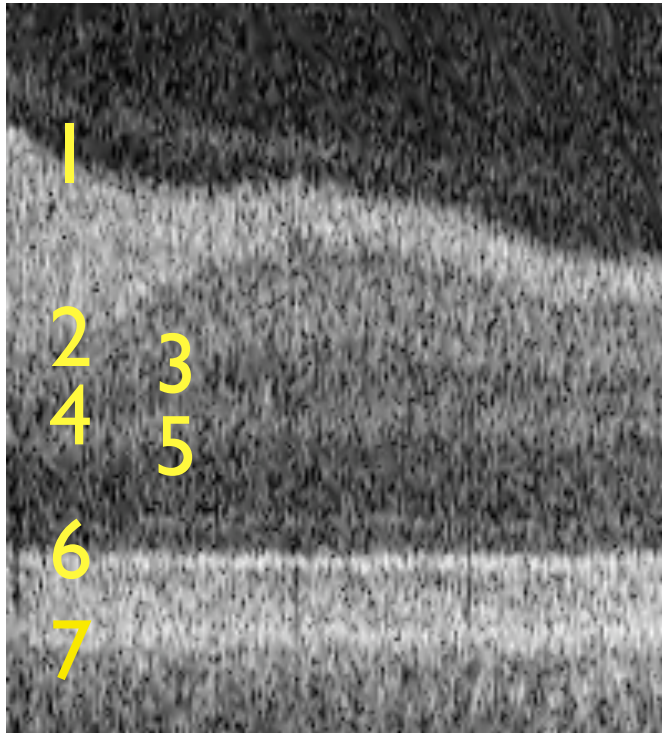
surf1 surf2 surf3 surf4 surf5 surf6 surf7



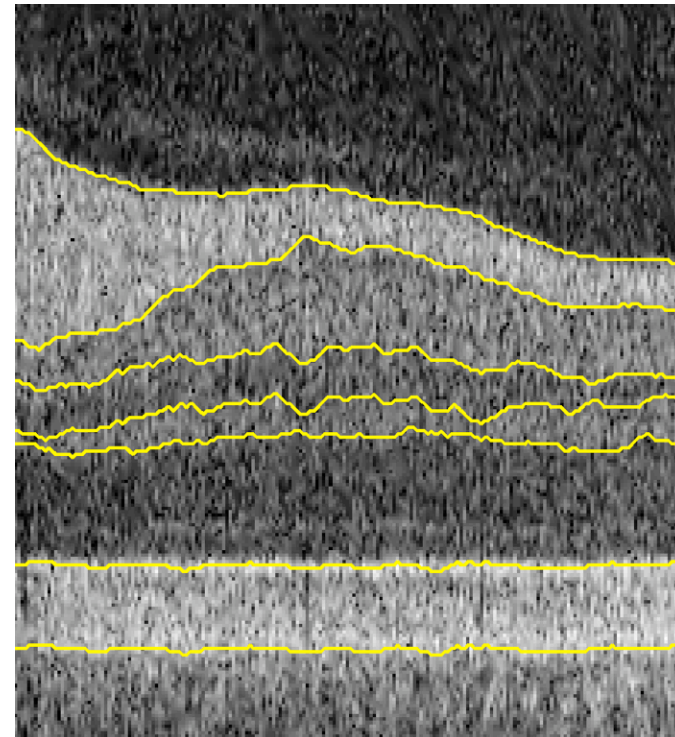
reg0 reg1 reg2 reg3 reg4 reg5 reg6 reg7



Example using both on-surface and in-region costs



OCT Image

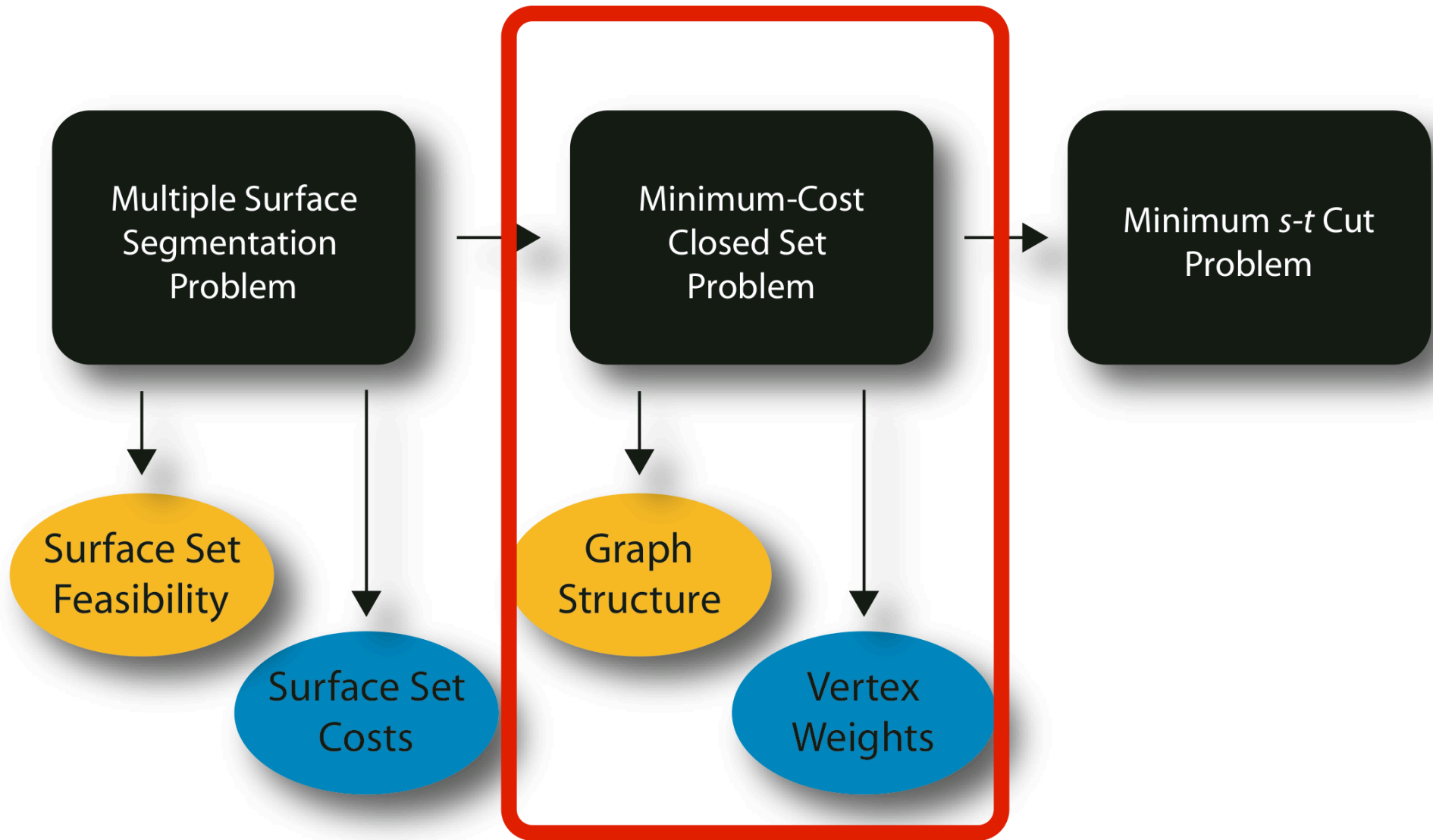


Result

The optimal set of surfaces



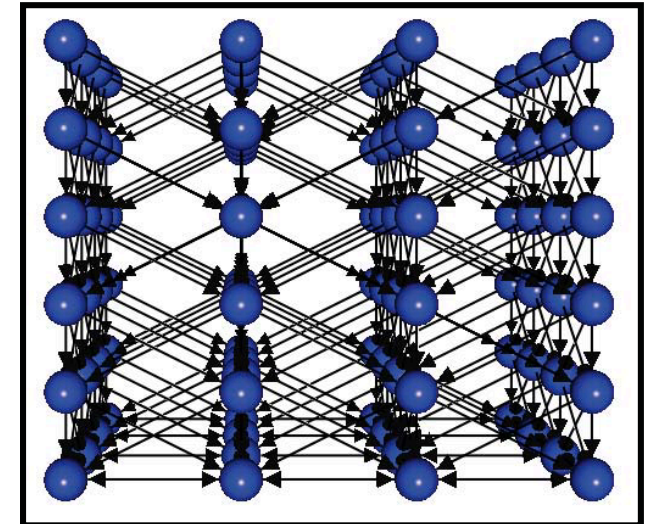
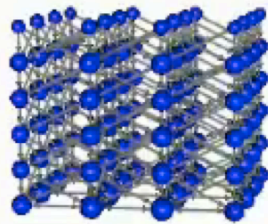
The graph structure ensures surface set feasibility. The assigned vertex weights ensure the optimal feasible surface set will be found.



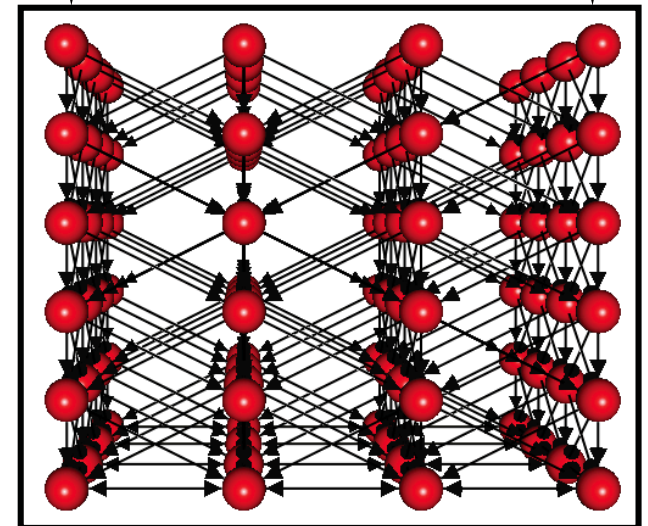
K. Li et al., PAMI 2006, extensions: M. Garvin et al., TMI 2009



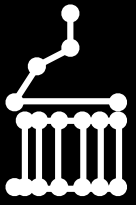
Each set of feasible surfaces corresponds to a non-empty closed set in the constructed graph (and vice versa)



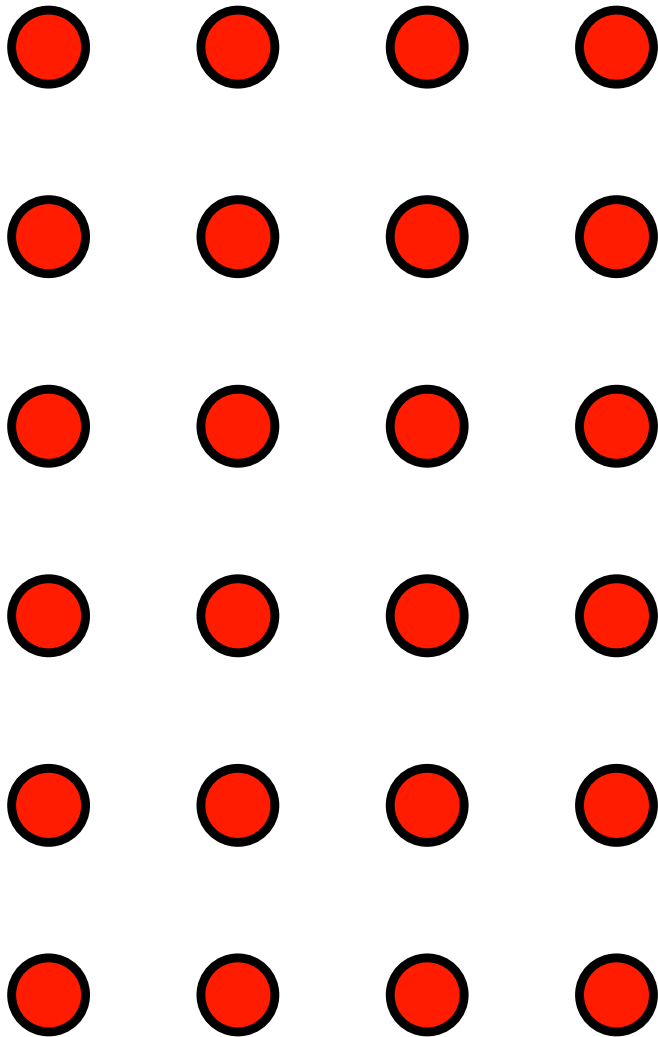
↕ Intersurface edges ↕



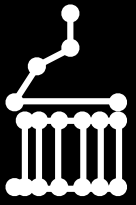
Intracolumn and intercolumn edges



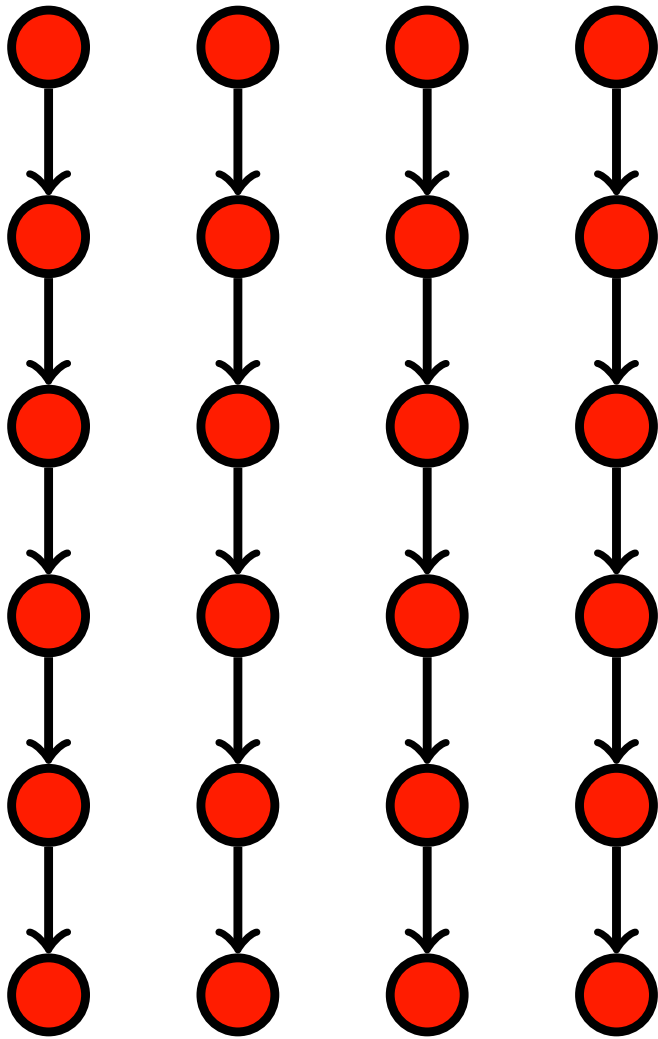
Intracolumn and intercolumn edges enforce smoothness constraints



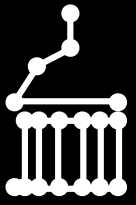
Graph nodes



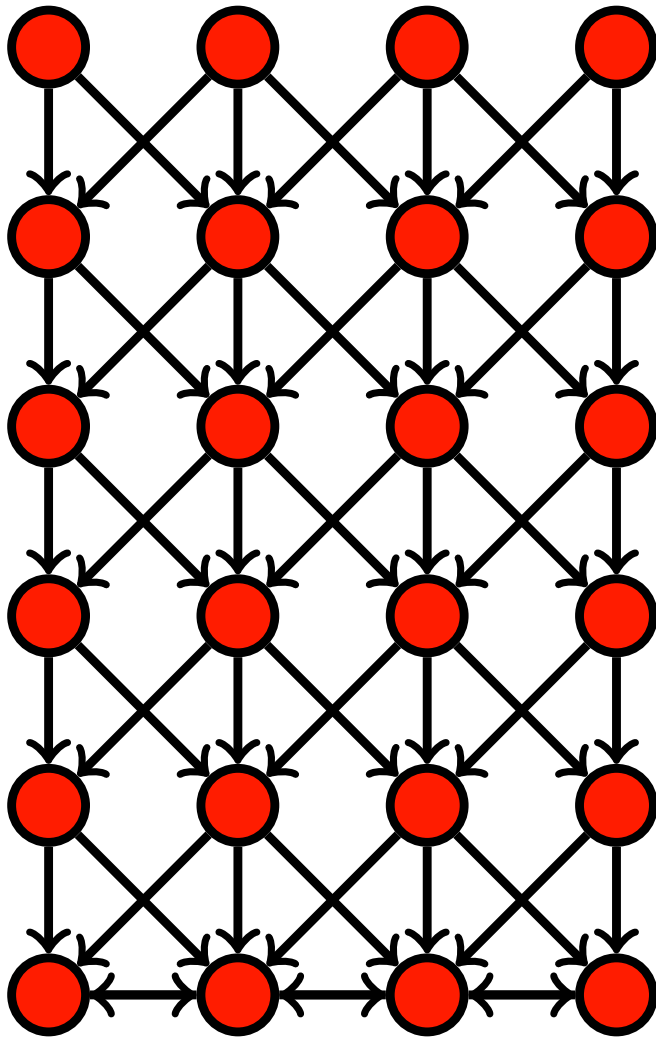
Intracolumn and intercolumn edges enforce smoothness constraints



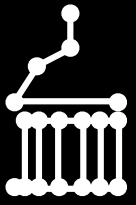
Add intracolumn edges



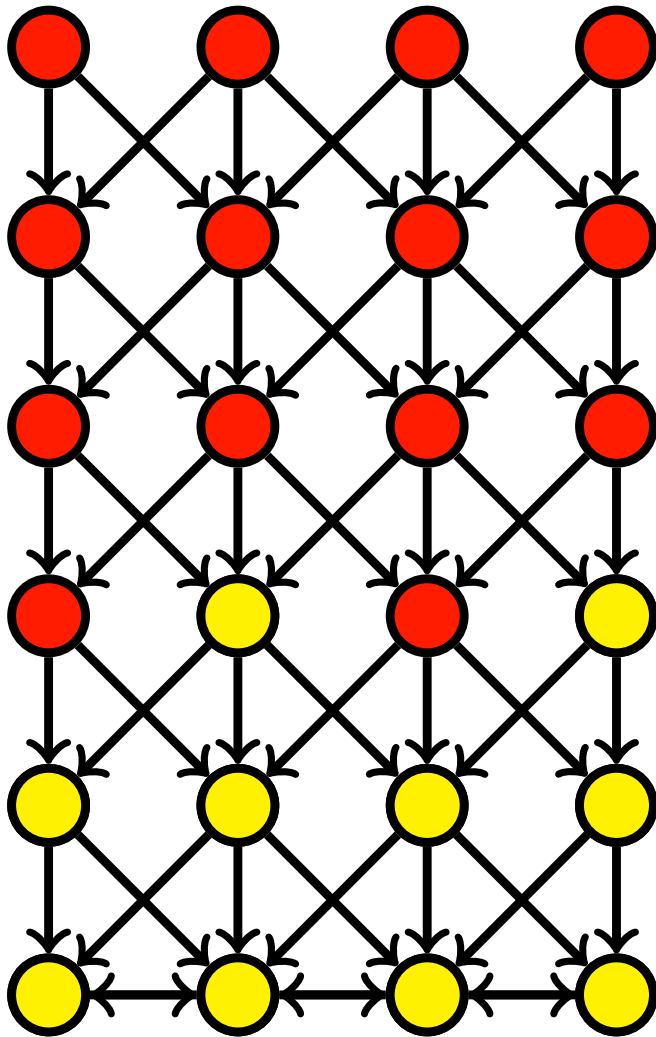
Intracolumn and intercolumn edges enforce smoothness constraints



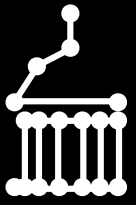
Add intercolumn edges



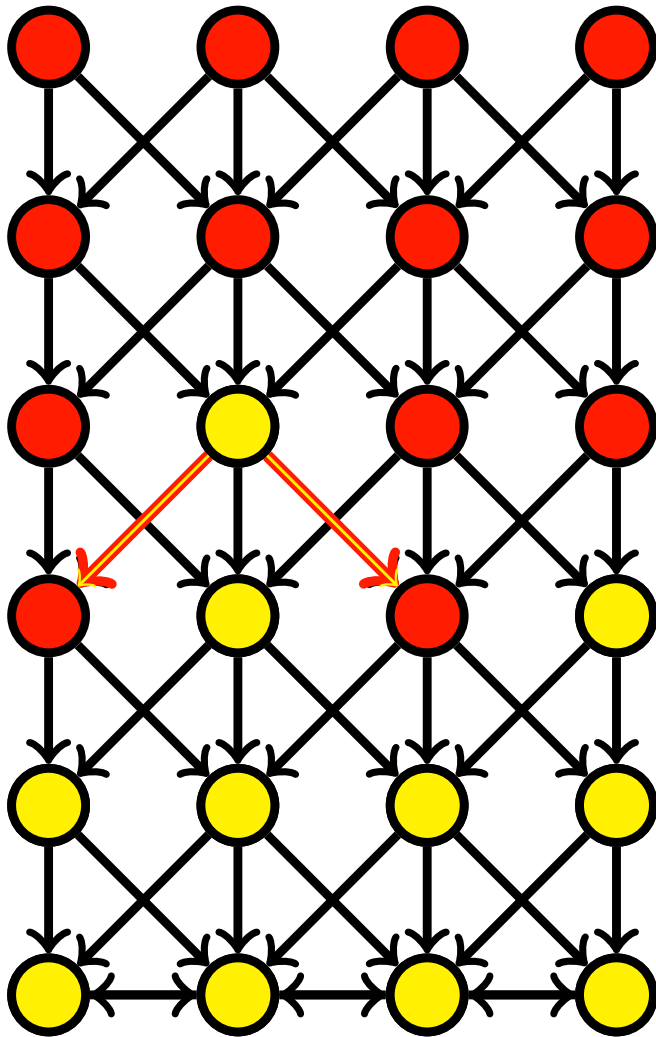
Intracolumn and intercolumn edges enforce smoothness constraints



A closed set



Intracolumn and intercolumn edges enforce smoothness constraints



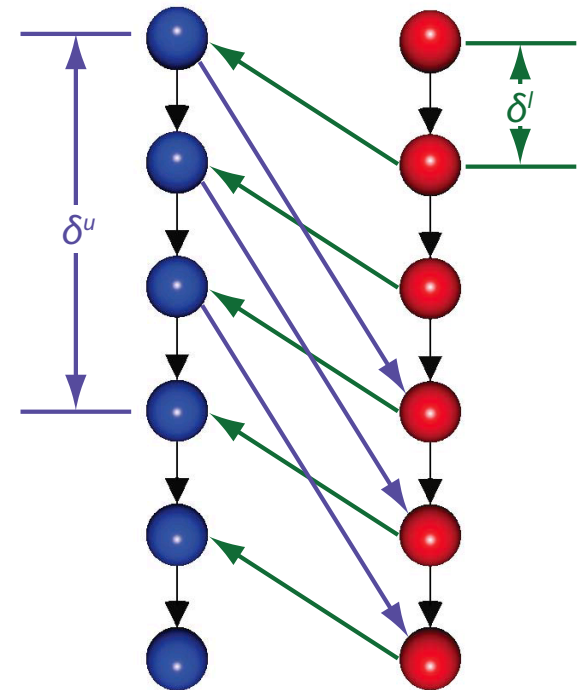
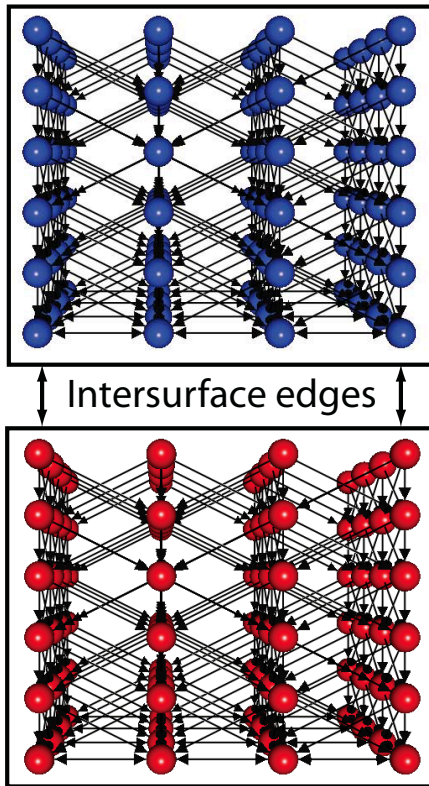
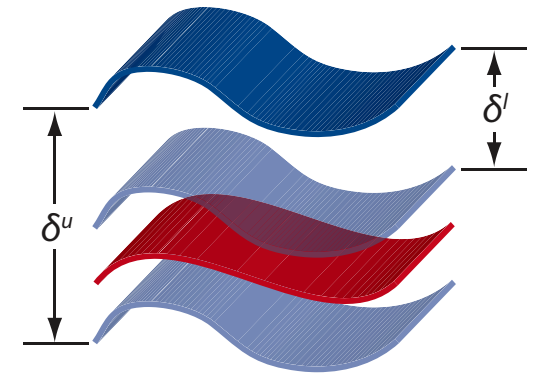
Not a closed set

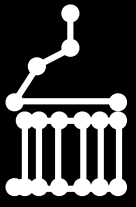


Intersurface edges enforce surface interaction constraints

Surface interaction constraints:

$$\delta^l(x, y) \leq f_i(x, y) - f_j(x, y) \leq \delta^u(x, y)$$





The cost of each surface set corresponds (within a constant) to the cost of the corresponding closed set

Surface set cost:

$$\sum_{i=1}^n C_{f_i(x,y)} + \sum_{i=0}^n C_{R_i}$$

Closed set cost:

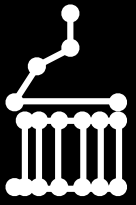
$$\sum_{i=1}^n C_{f_i(x,y)} + \sum_{i=0}^n C_{R_i} + K$$



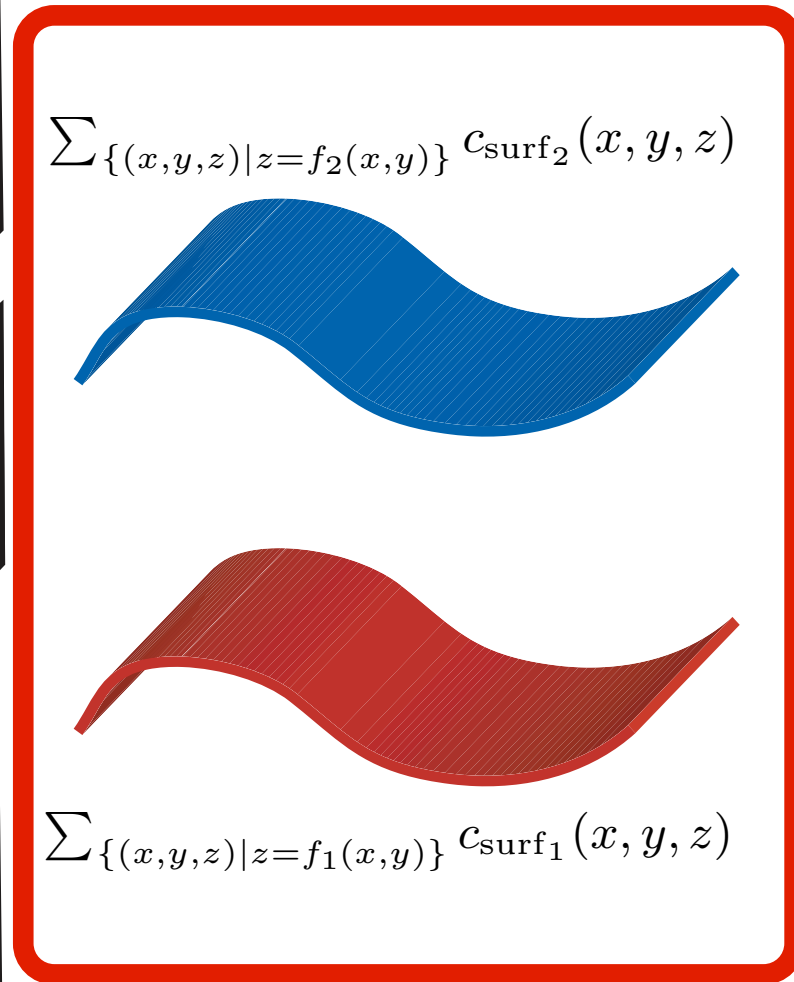
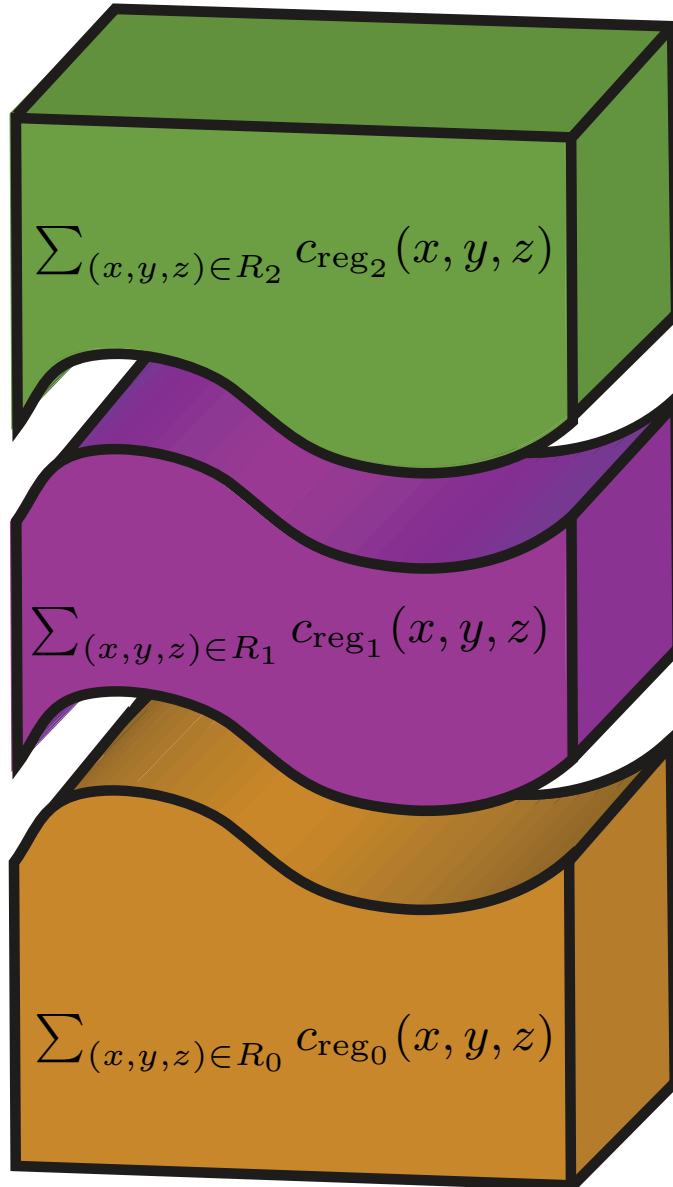
Minimum surface
set cost



Minimum closed set



On-surface cost representation





On-surface cost representation

node weight:

$$w_{\text{on-surf}_i}(x, y, z) = \begin{cases} c_{\text{surf}_i}(x, y, z) & \text{if } z = 0 \\ c_{\text{surf}_i}(x, y, z) - c_{\text{surf}_i}(x, y, z - 1) & \text{otherwise} \end{cases}$$

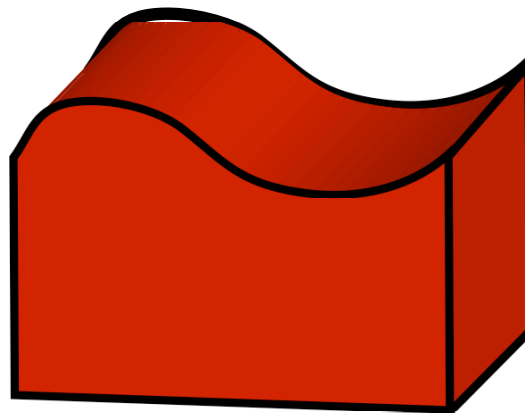
Surface Cost 2



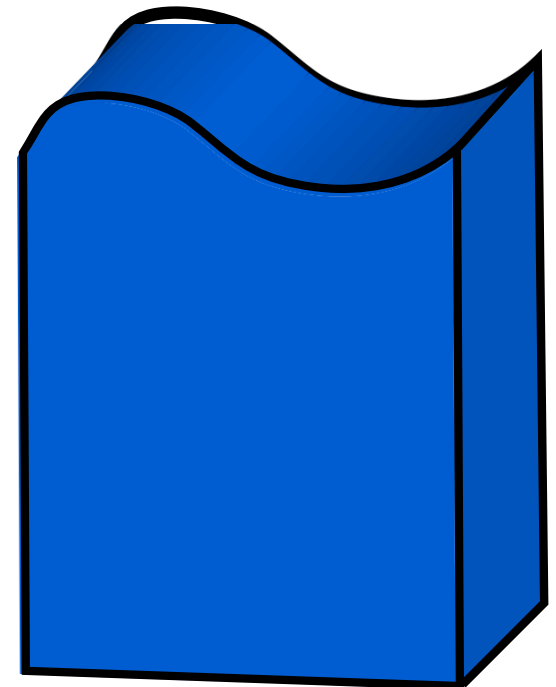
Surface Cost 1

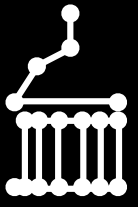


Closed Set Cost 1

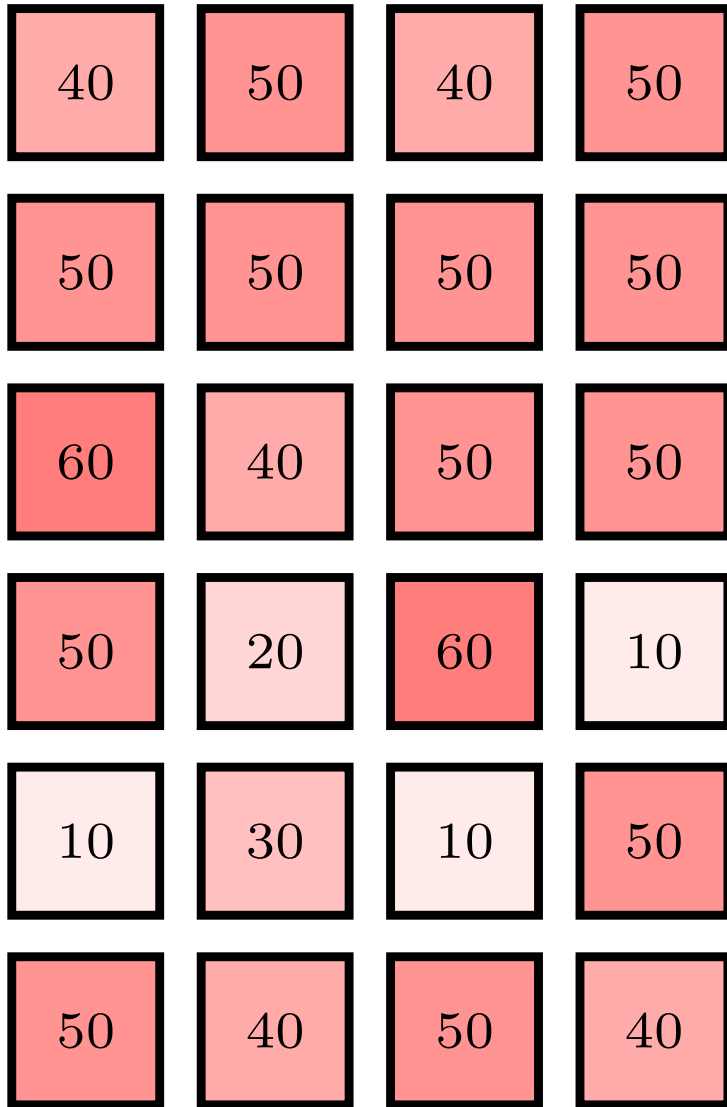


Closed Set Cost 2

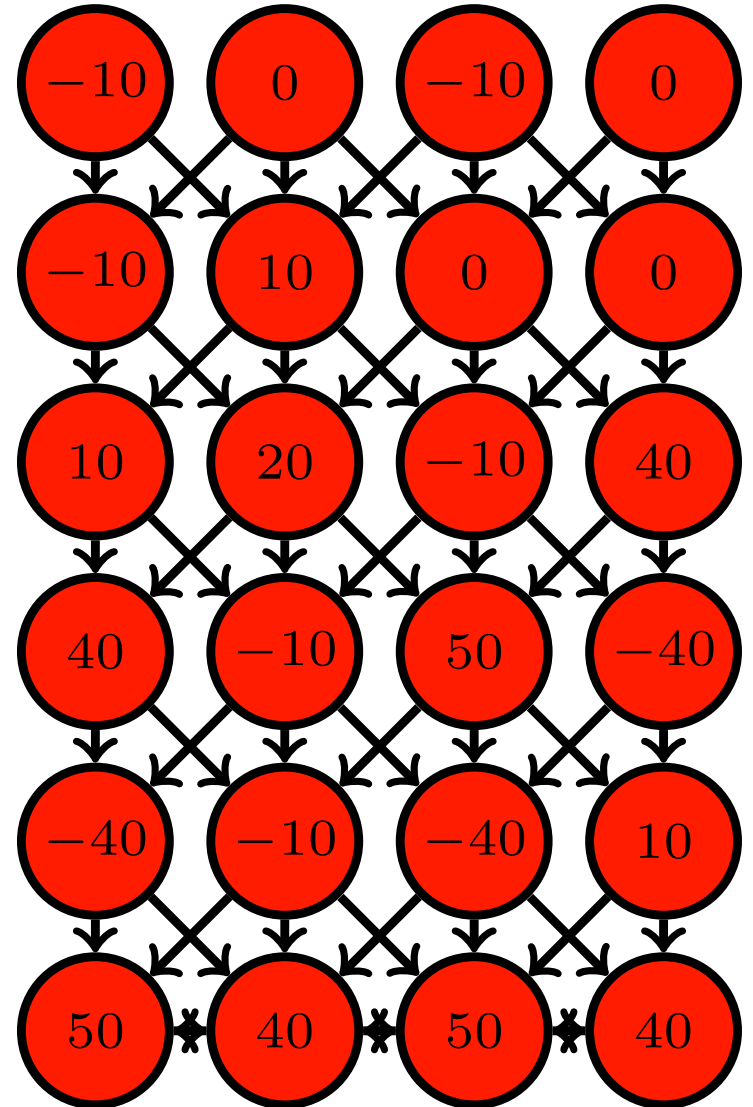




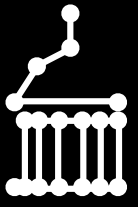
Graph representation of on-surface costs (toy example)



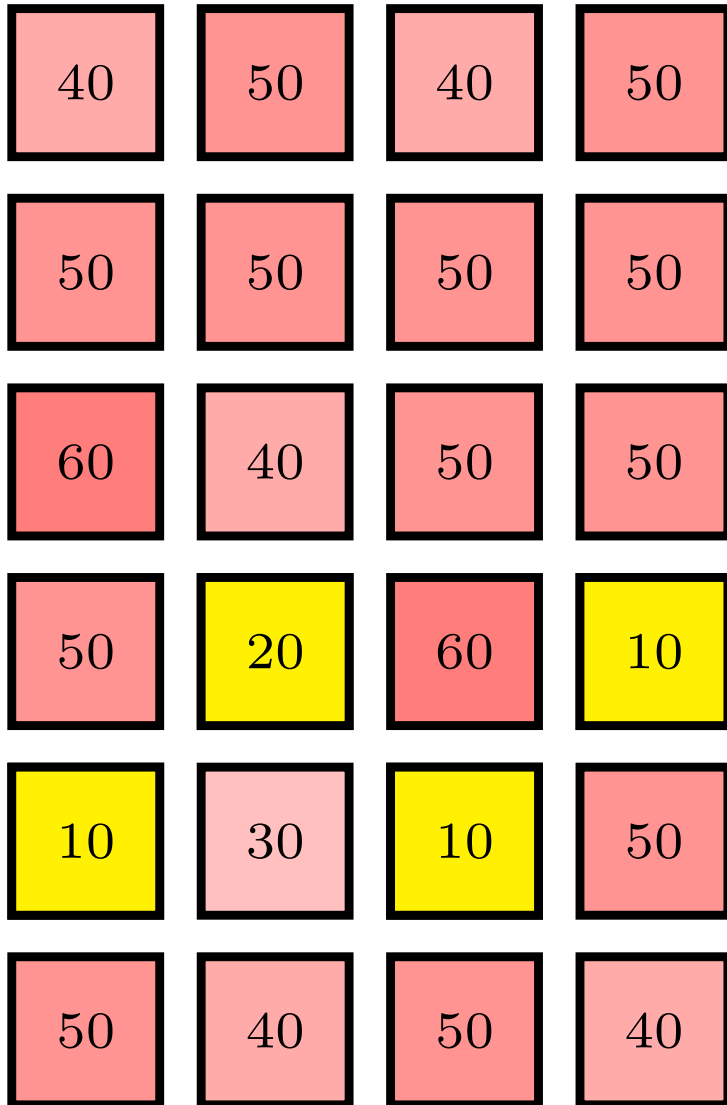
cost image



graph representation

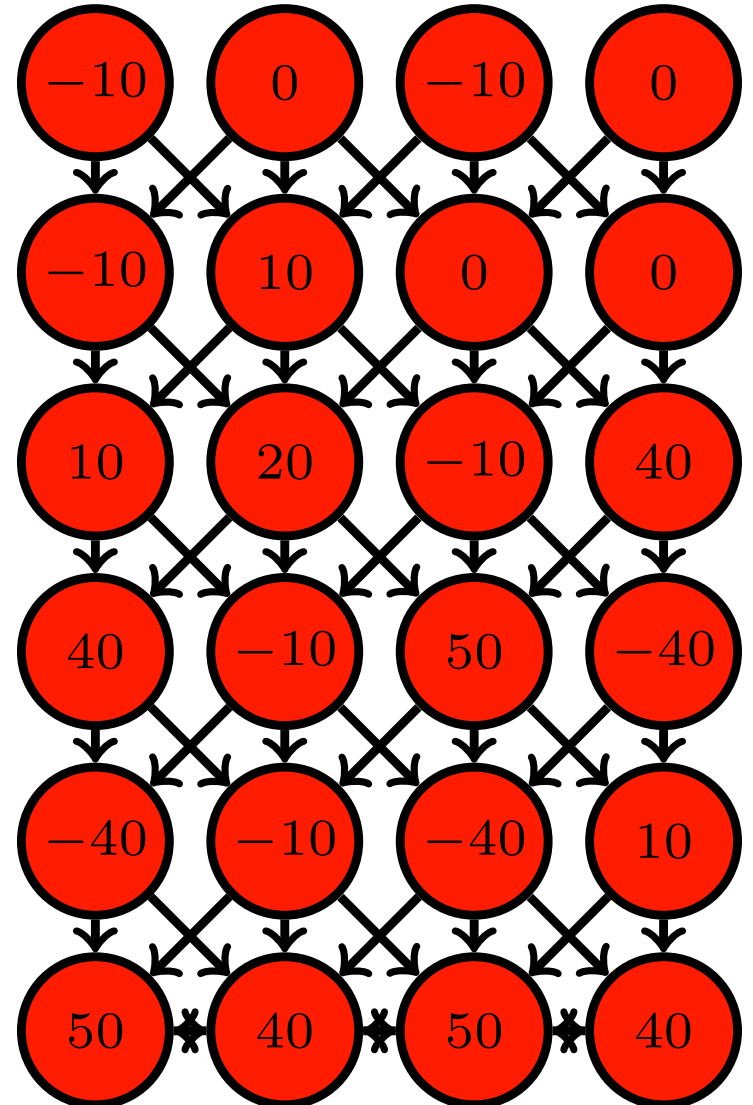


Graph representation of on-surface costs (toy example)

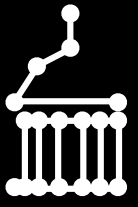


cost image

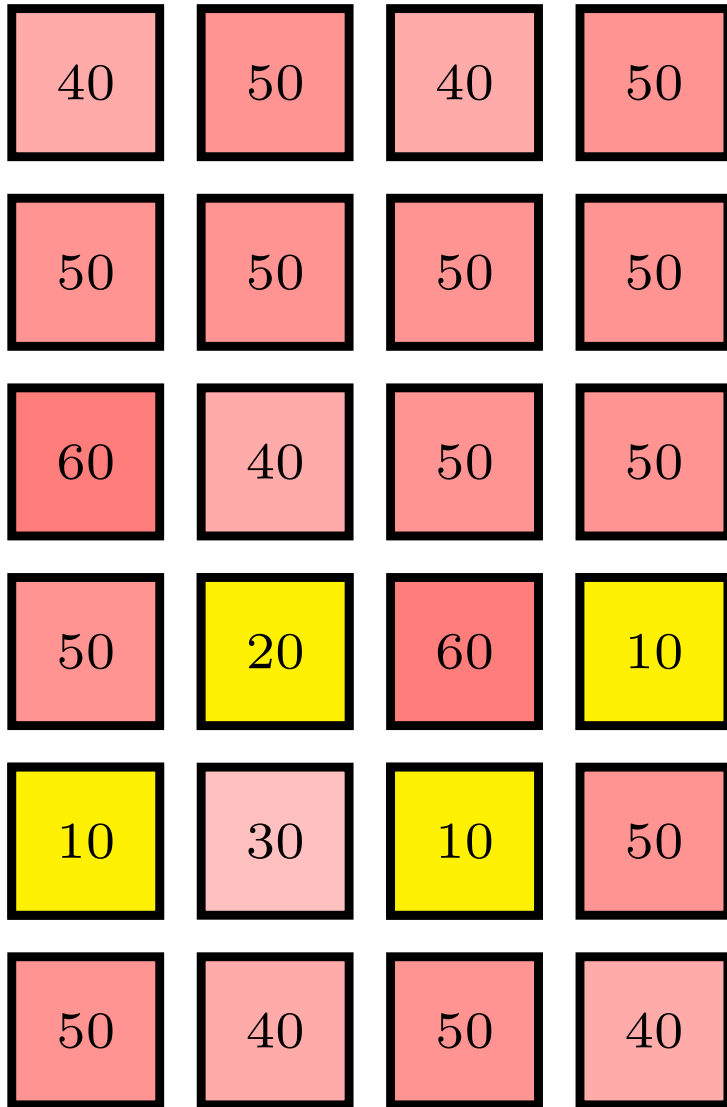
surf. cost=50



graph representation



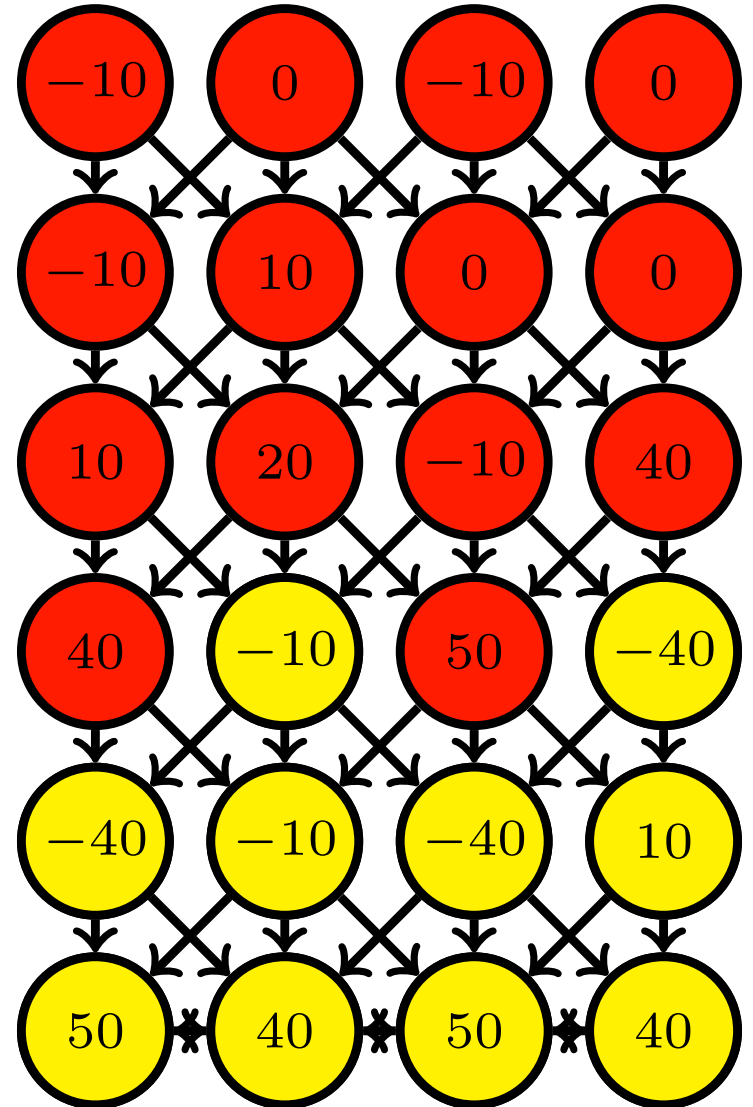
Graph representation of on-surface costs (toy example)



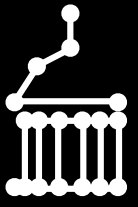
cost image

surf. cost=50

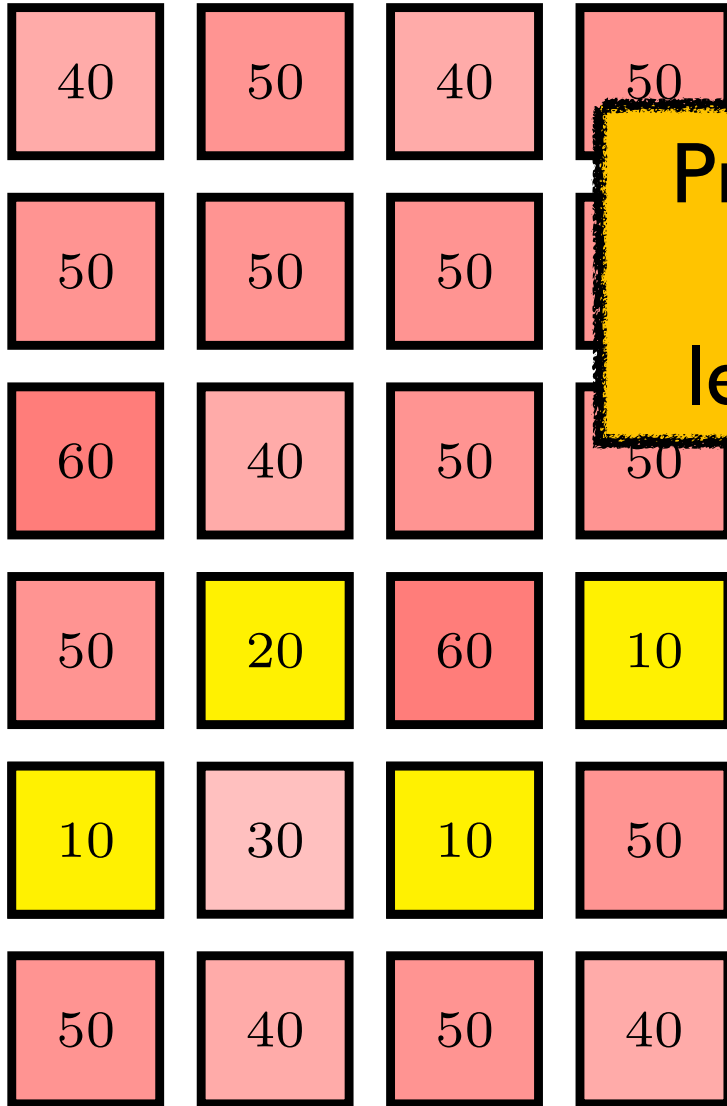
CS cost=50



graph representation



Graph representation of on-surface costs (toy example)

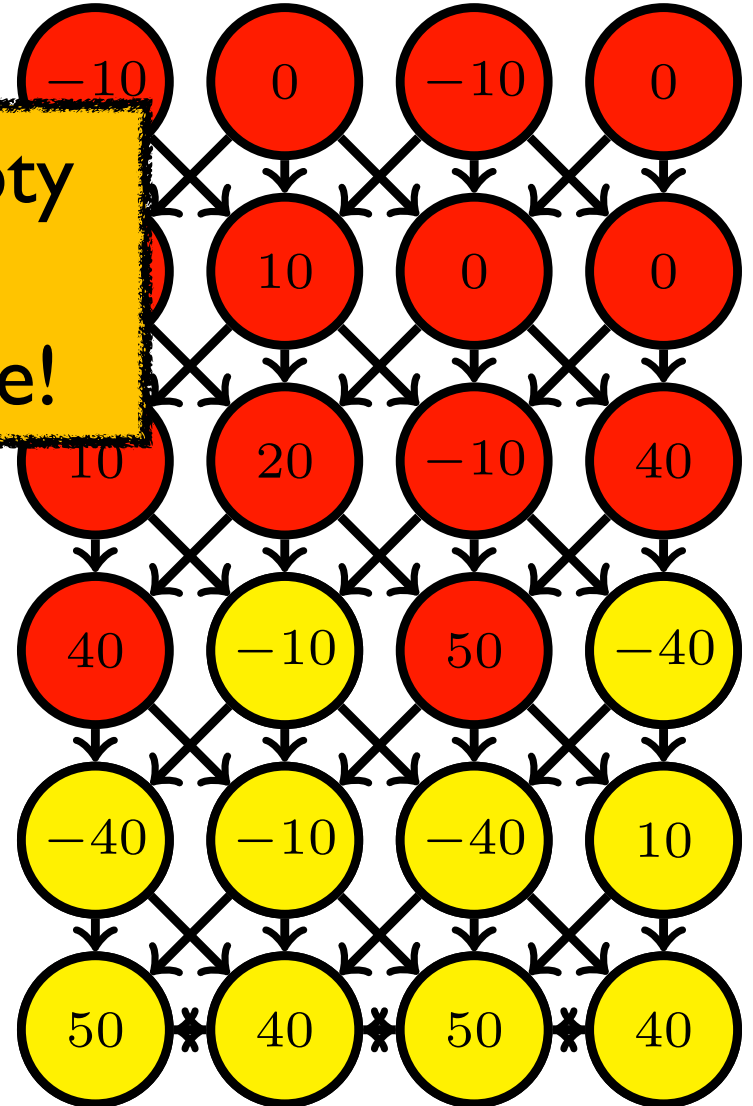


cost image

Problem: empty closed set less expensive!

surf. cost=50

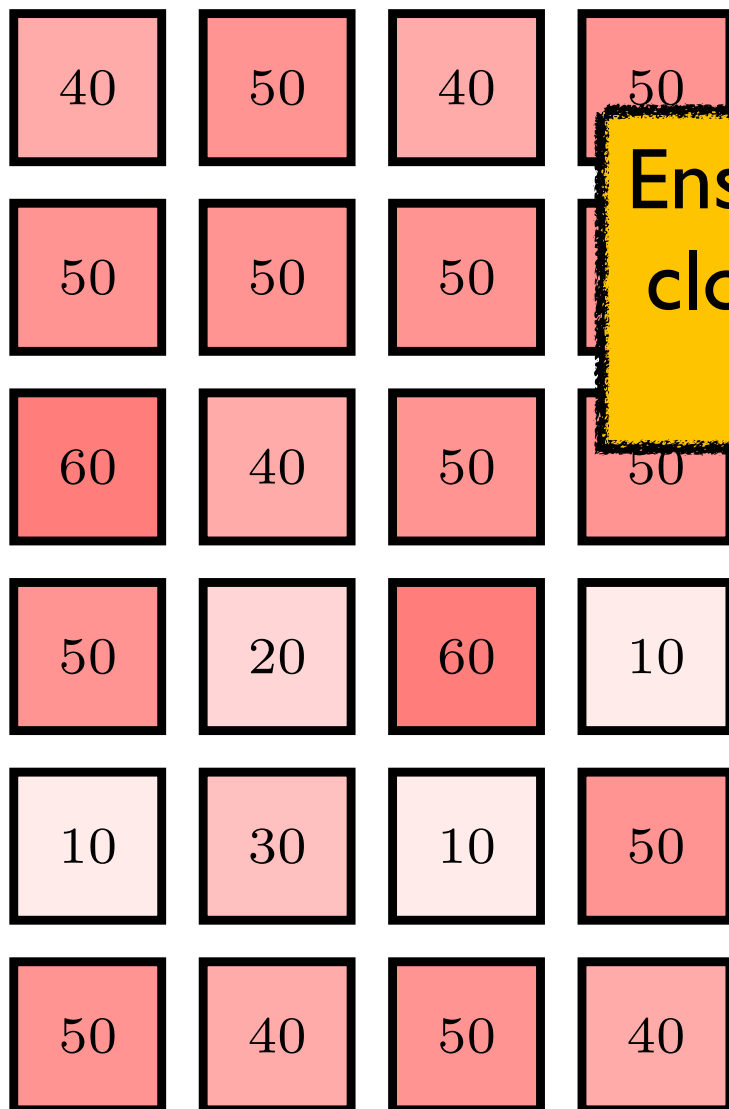
CS cost=50



graph representation



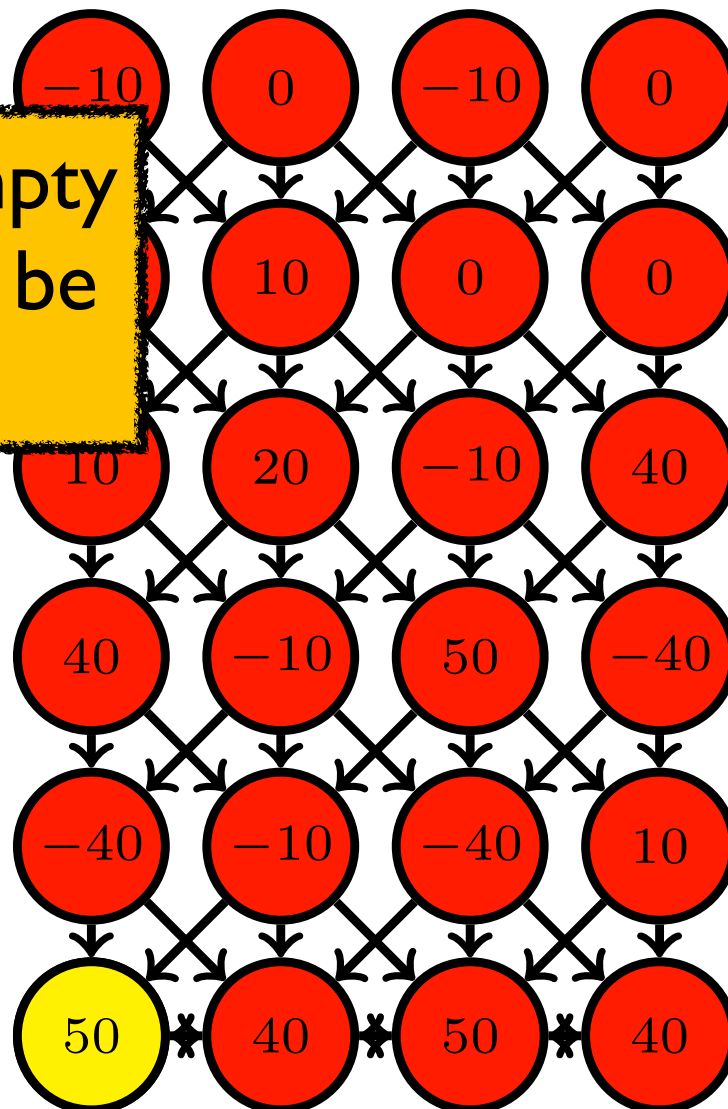
Graph representation of on-surface costs (toy example)



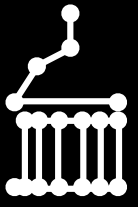
cost image

Ensure non-empty closed set will be minimum.

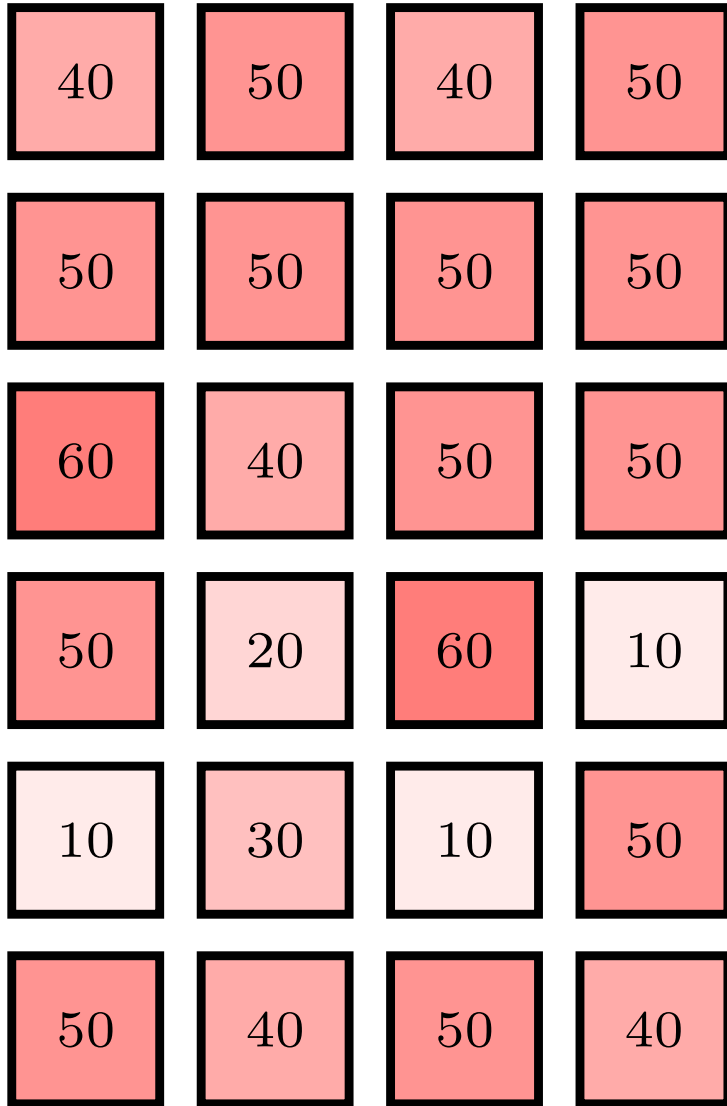
subtract
(sum of last row + 1) =
181



graph representation

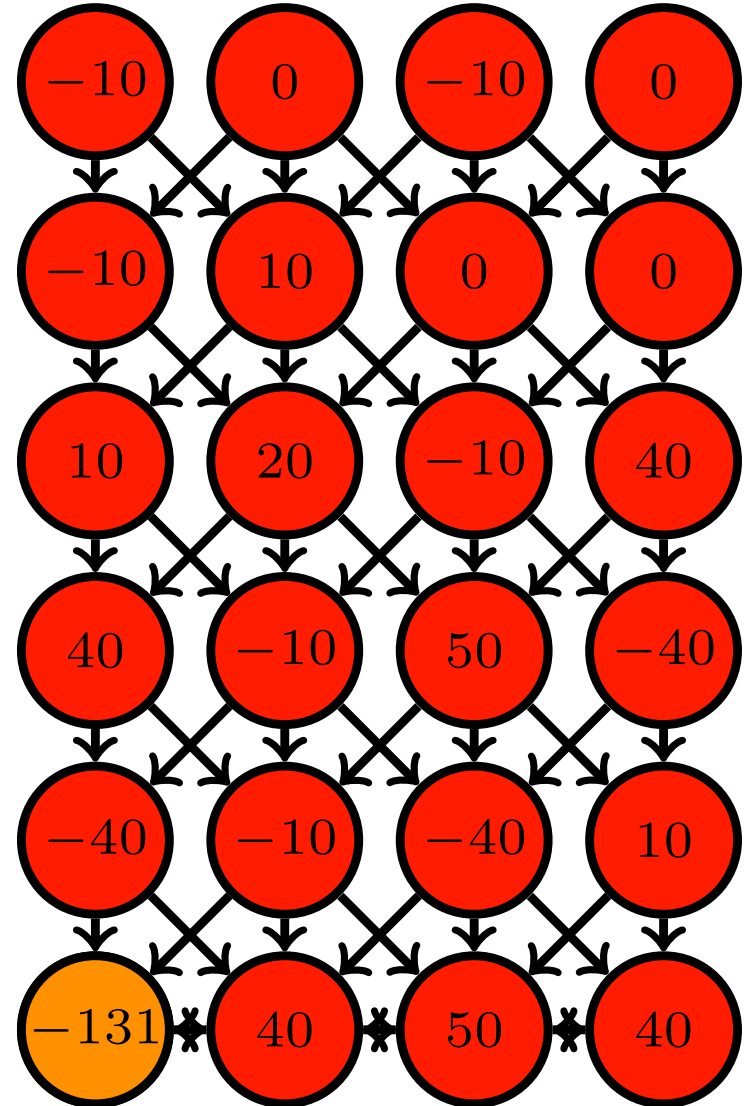


Graph representation of on-surface costs (toy example)

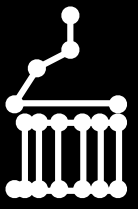


cost image

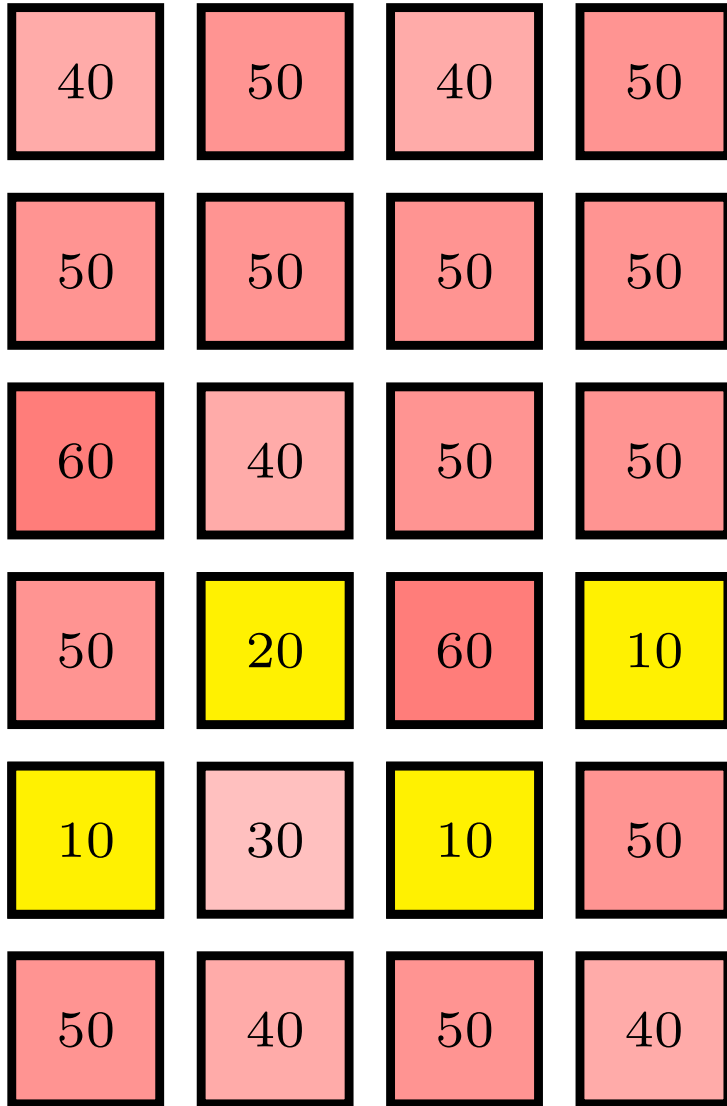
50 - 181



graph representation



Graph representation of on-surface costs (toy example)

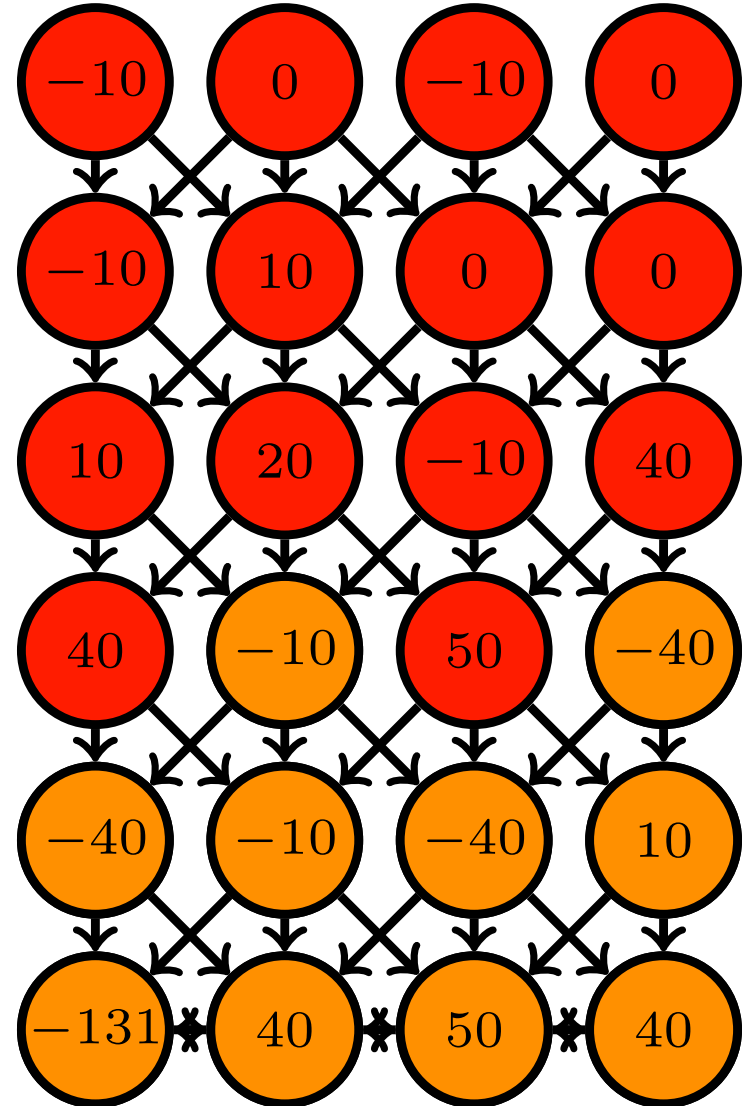


cost image

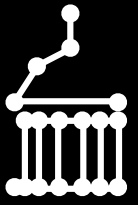
surf. cost=50

CS cost=
 $K + 50$

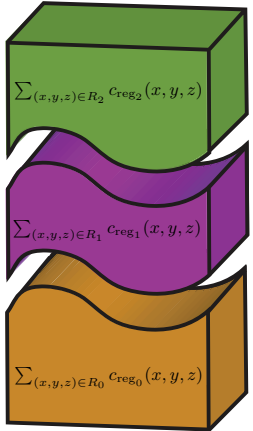
($K = -181$)



graph representation



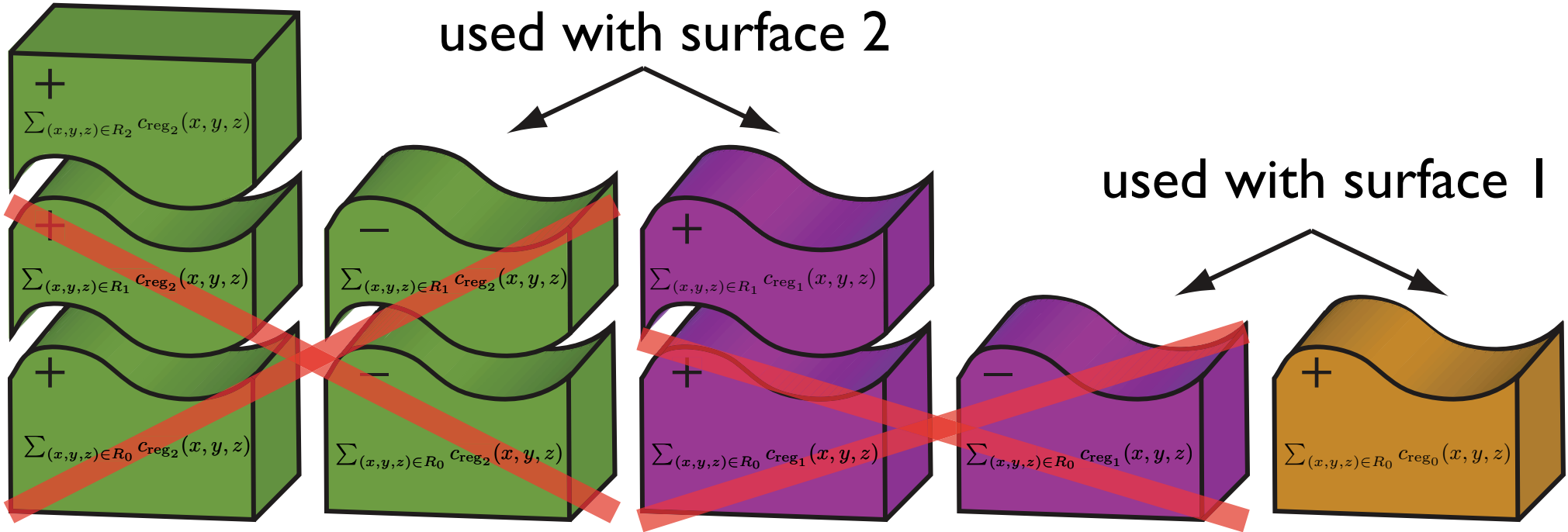
In-region cost representation



node weight (in subgraph associated with surface i)
 $w_{\text{in-reg}_i}(x,y,z) = c_{\text{reg}_{i-1}}(x,y,z) - c_{\text{reg}_i}(x,y,z)$
 (region below) (region above)

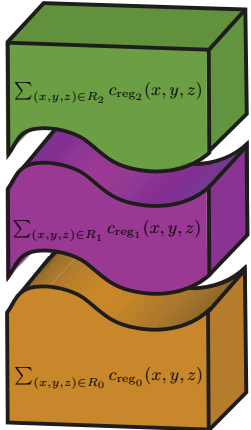
used with surface 2

used with surface 1

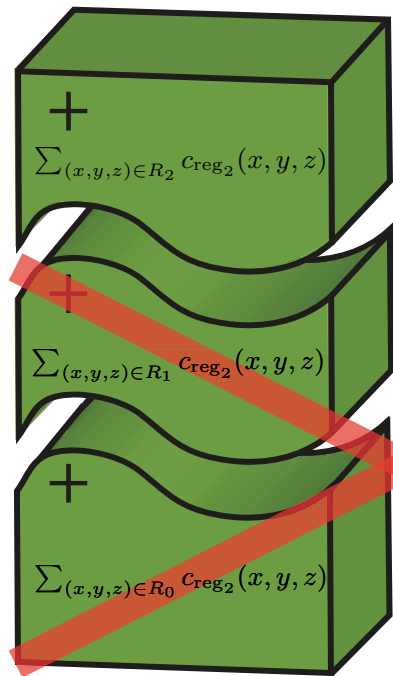




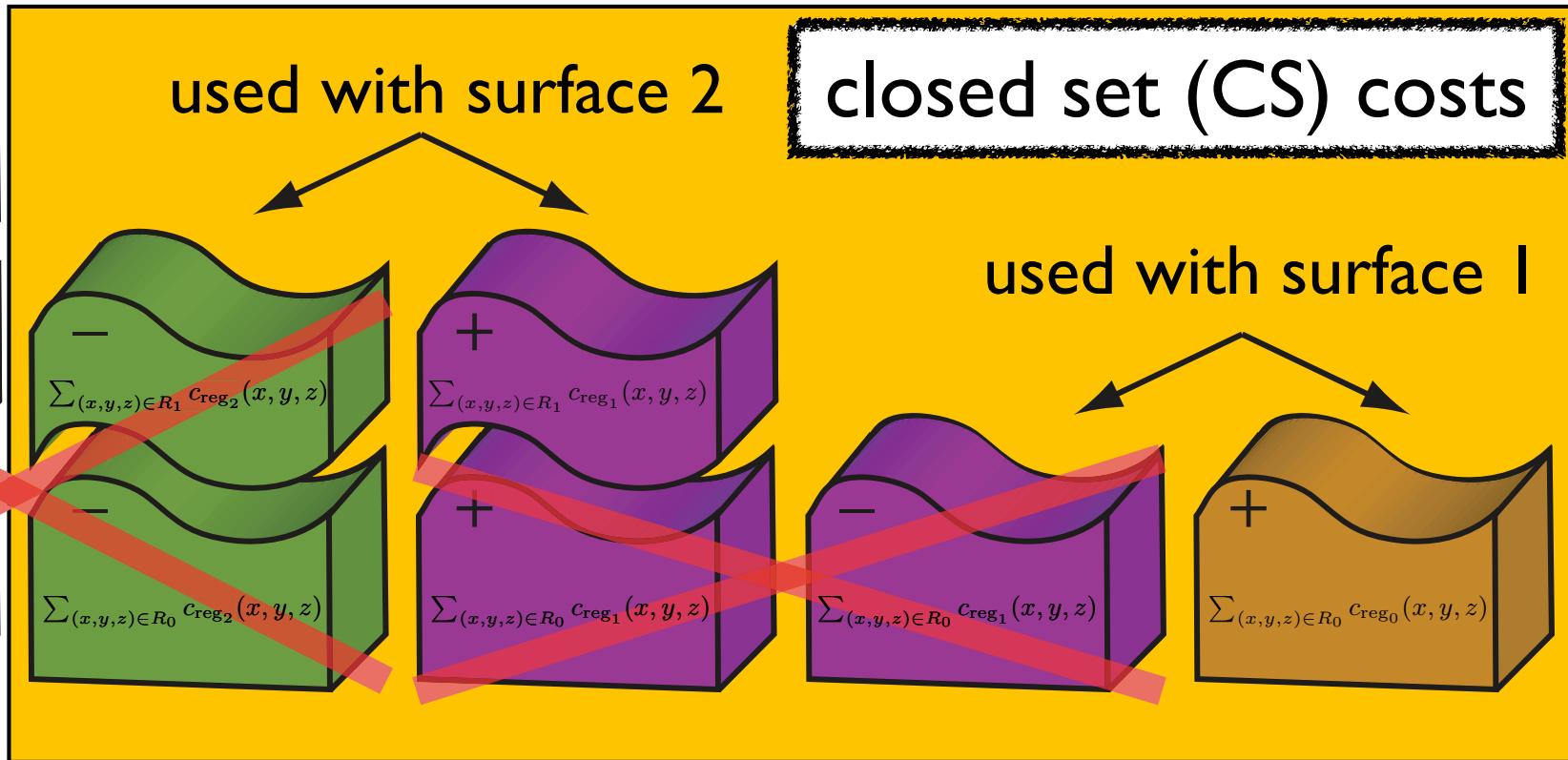
In-region cost representation



node weight (in subgraph associated with surface i)
 $w_{\text{in-reg}_i}(x,y,z) = c_{\text{reg}_{i-1}}(x,y,z) - c_{\text{reg}_i}(x,y,z)$
 (region below) (region above)

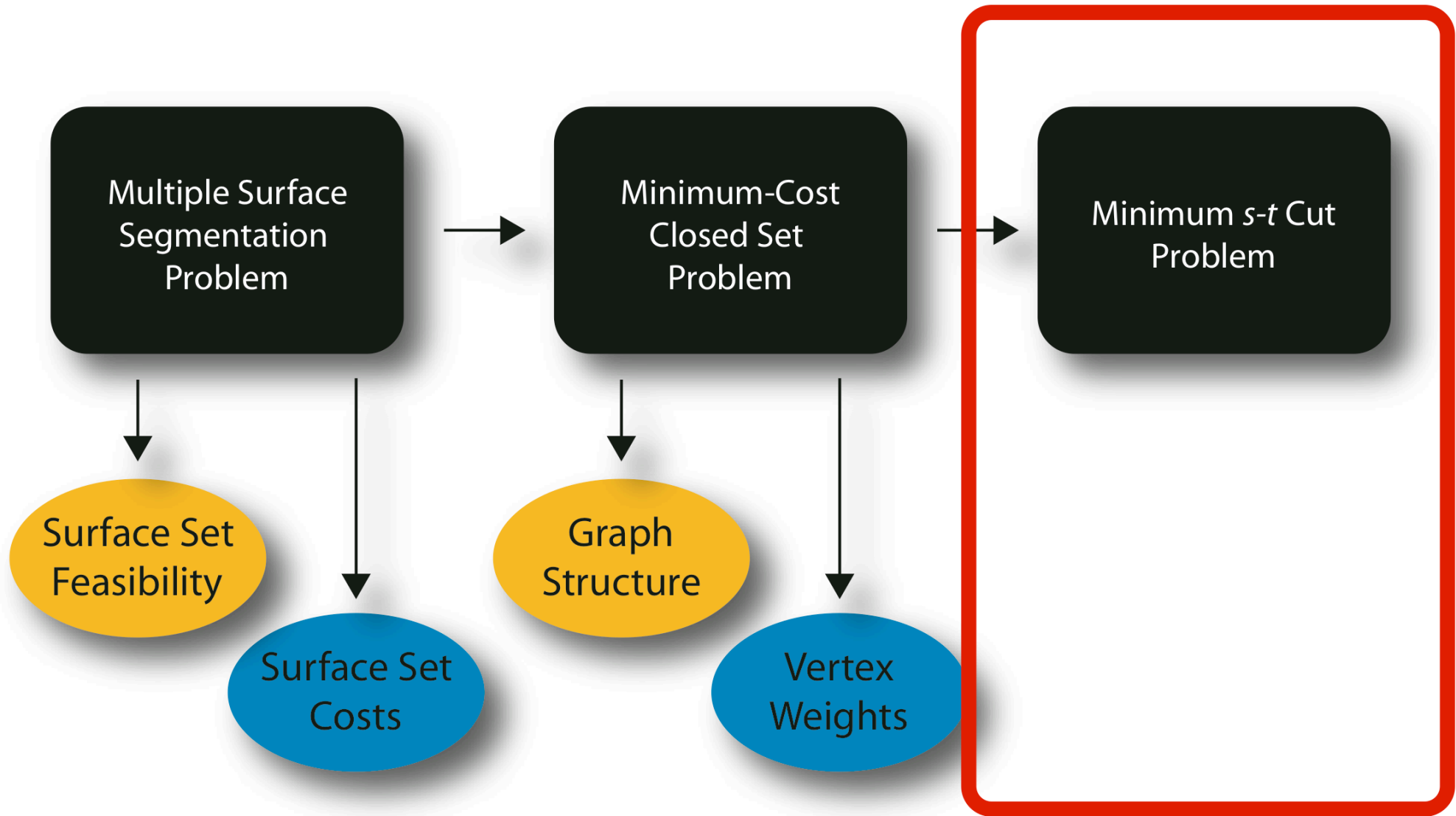


constant





The graph structure ensures surface set feasibility. The assigned vertex weights ensure the optimal feasible surface set will be found.

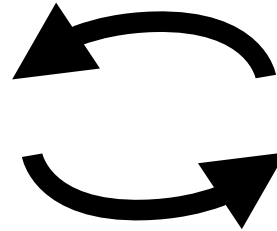


K. Li et al., PAMI 2006, extensions: M. Garvin et al., TMI 2009



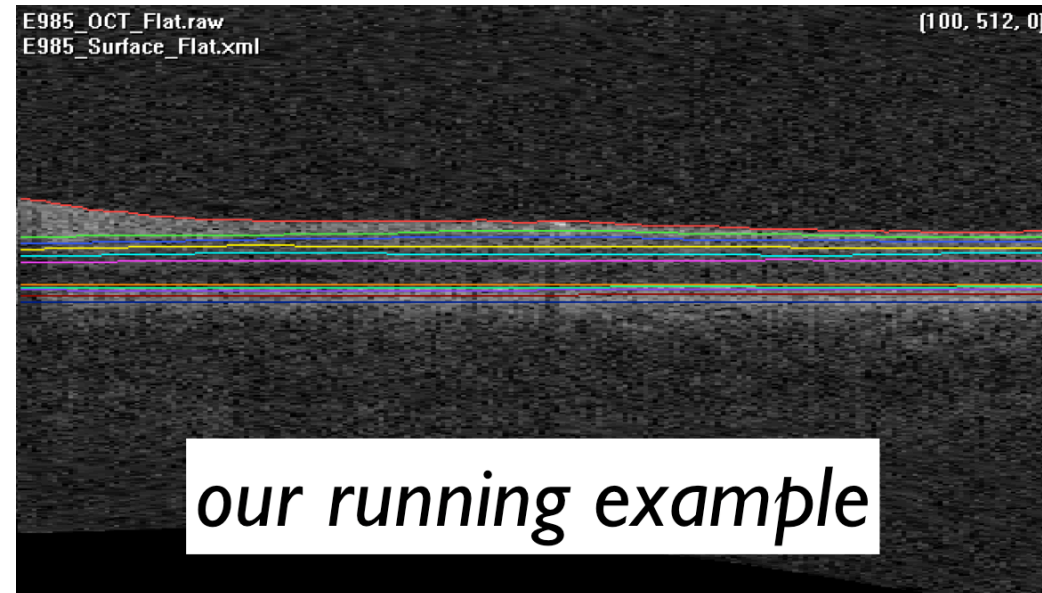
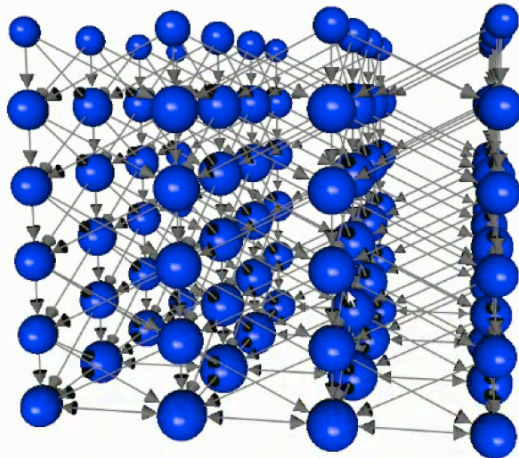
Outline

LOGISMOS
approach



Intraretinal layer
segmentation

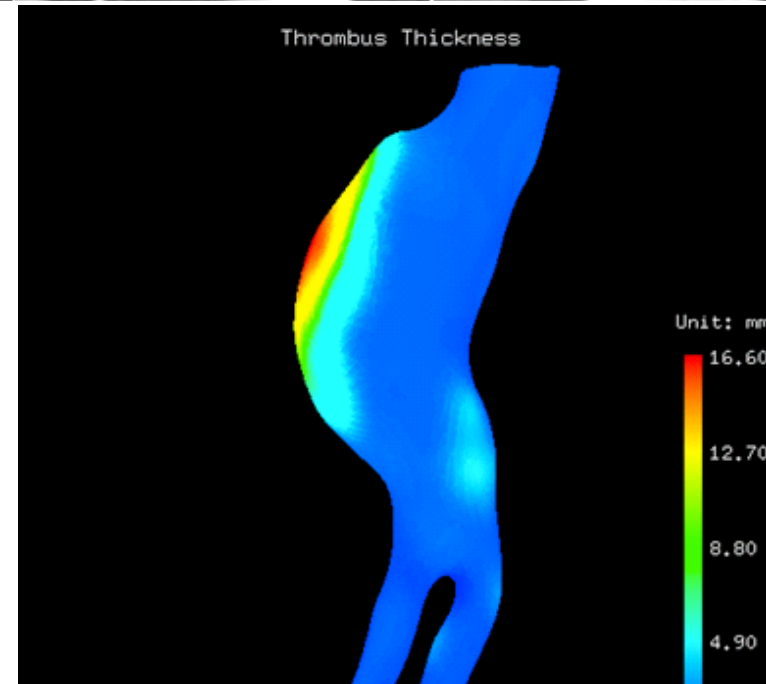
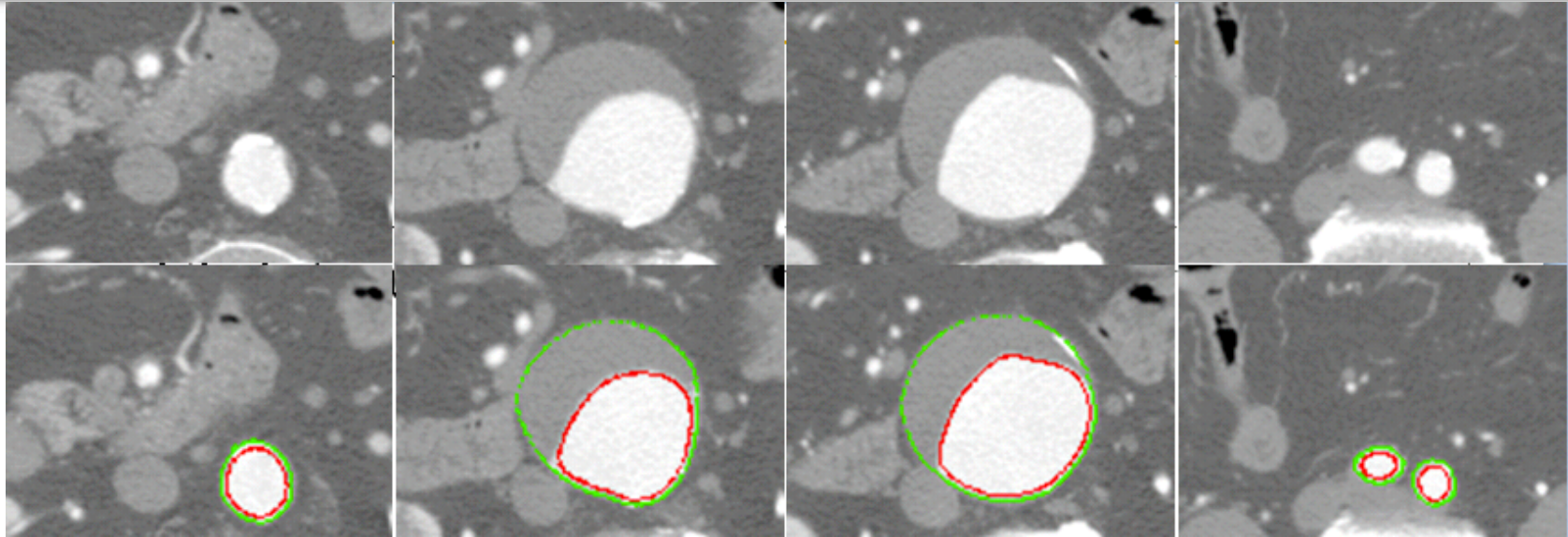
LOGISMOS = Layered Optimal Graph Image Segmentation of Multiple Objects and Surfaces



Other applications and future directions



Example other LOGISMOS applications: thrombus segmentation in abdominal aortic aneurysms

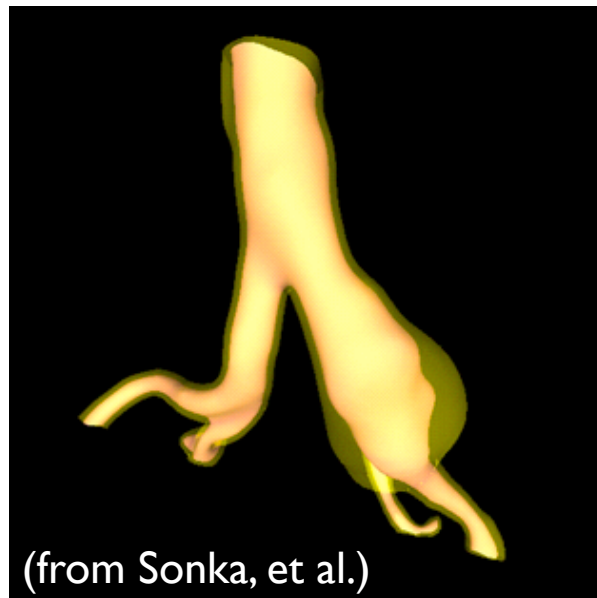
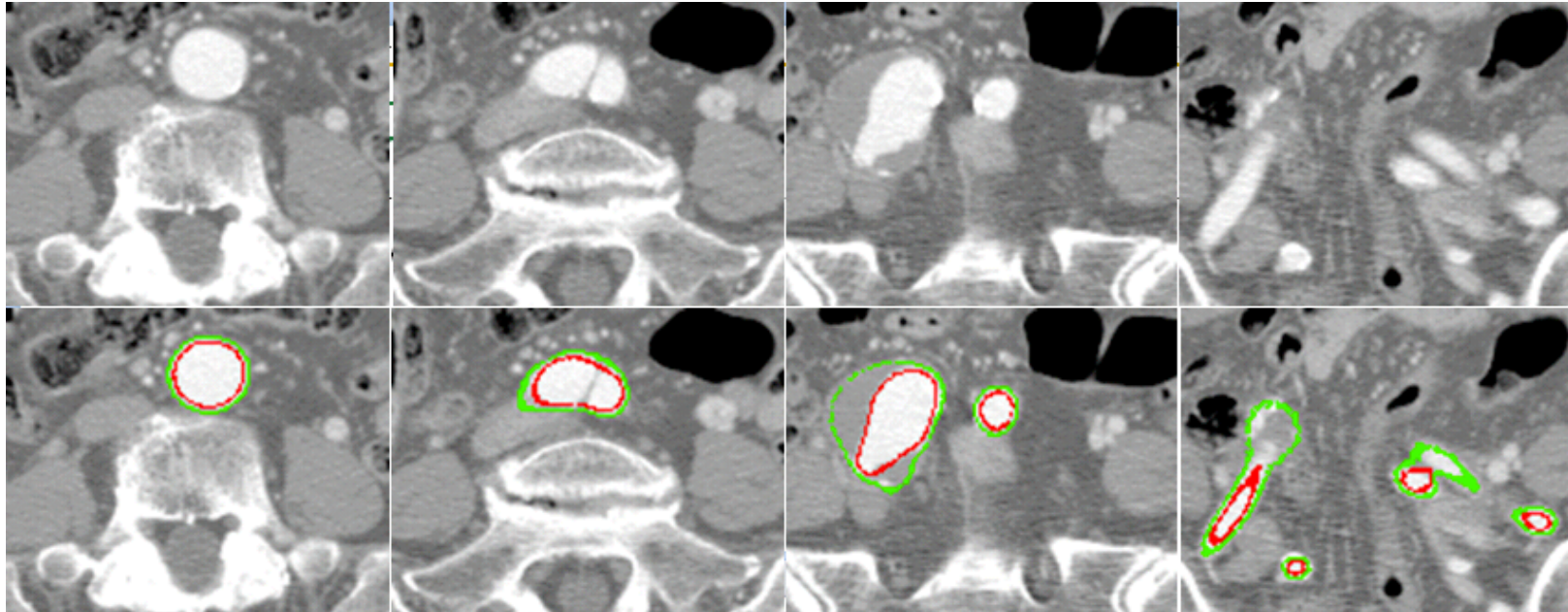


Lee, et al.,
Comput Biol
Med, 2010

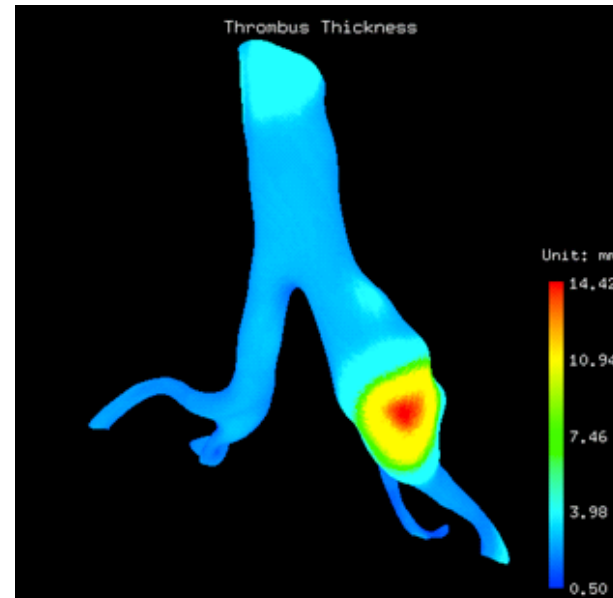
(from Sonka, et al.)



Example other LOGISMOS applications: thrombus segmentation in abdominal aortic aneurysms



(from Sonka, et al.)

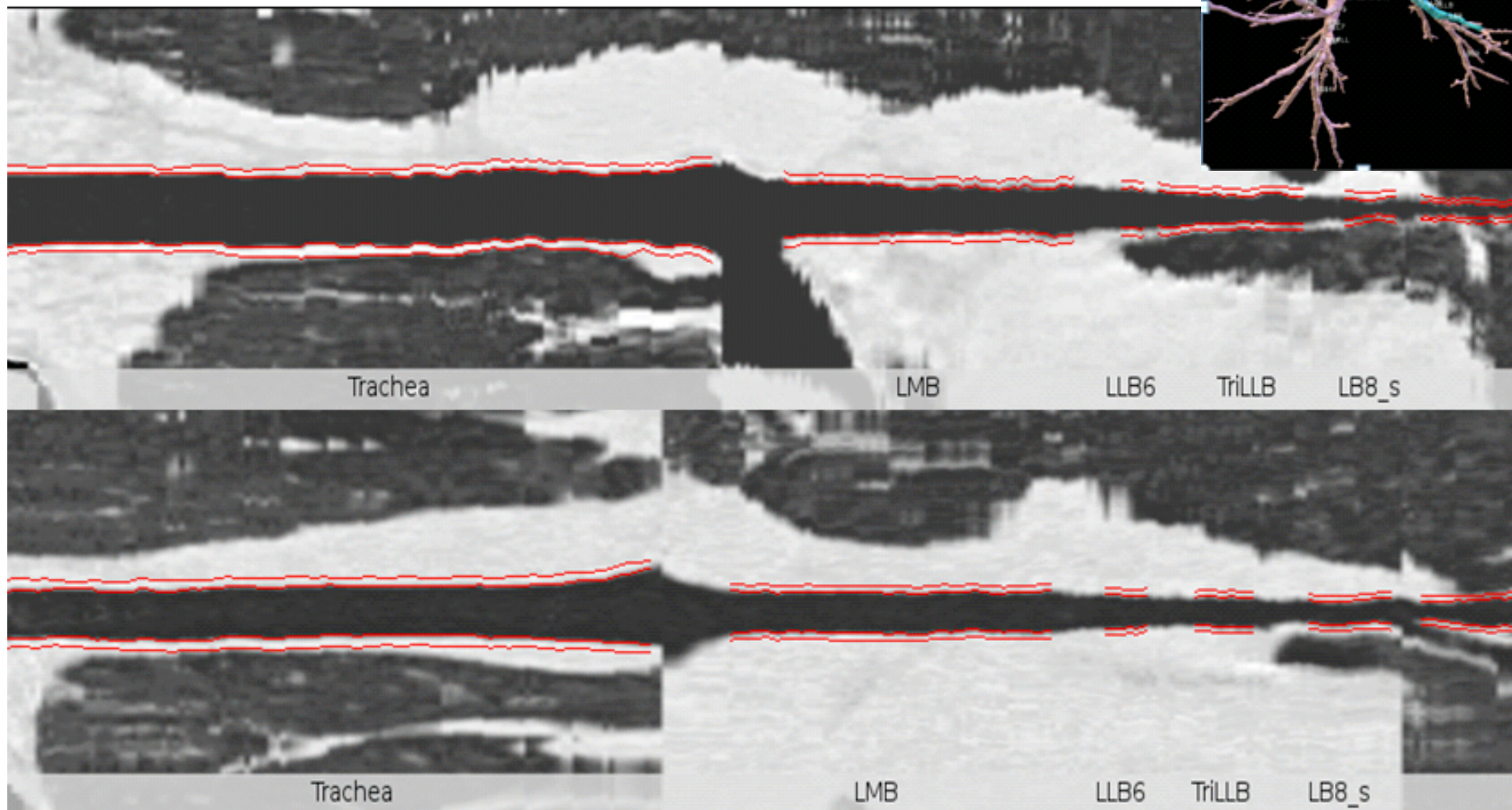


Lee, et al.,
Comput Biol
Med, 2010



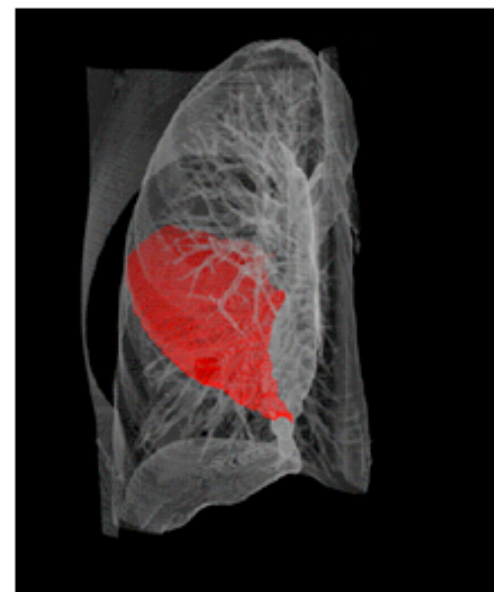
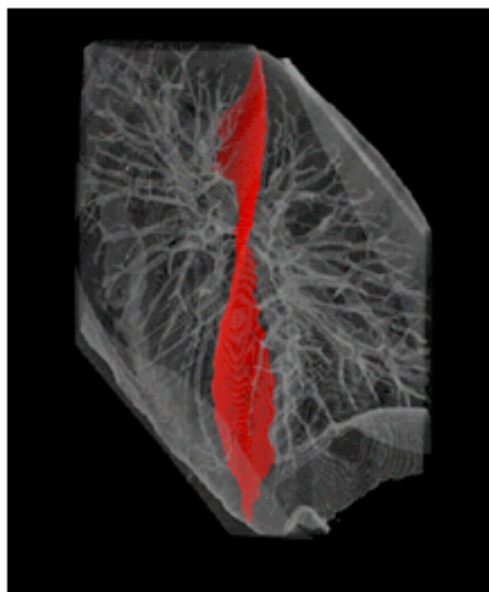
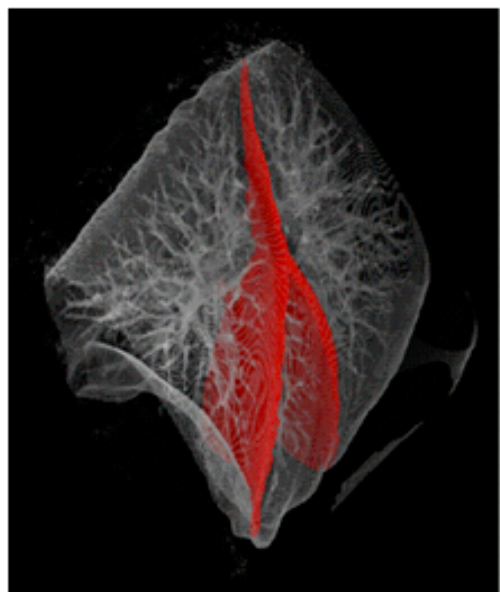
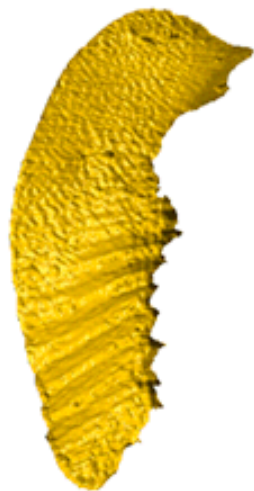
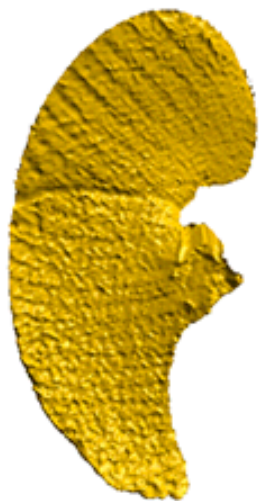
Example other LOGISMOS applications: airway tree

(from Sonka, et al.)





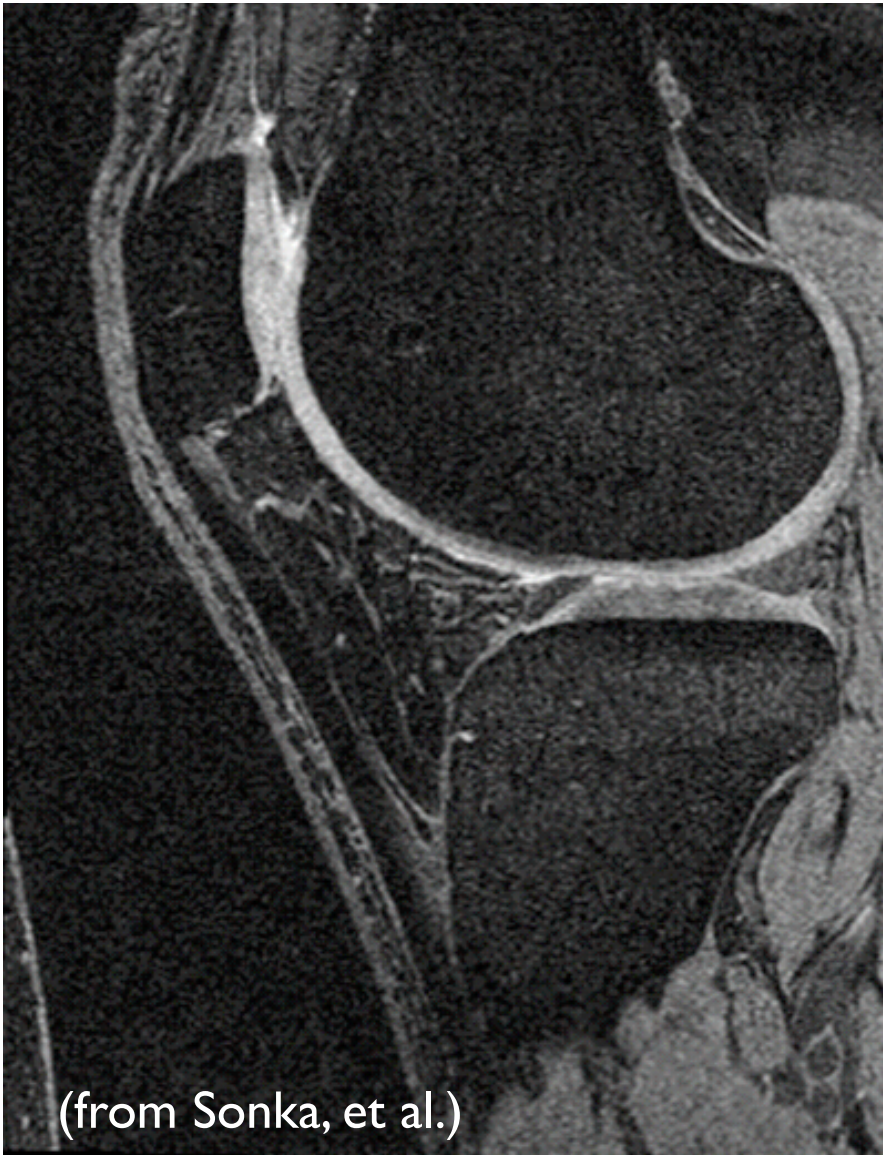
Example other LOGISMOS applications: pulmonary fissures



(from Reinhardt, et al.)



Example other LOGISMOS applications: determining cartilage thickness



(from Sonka, et al.)

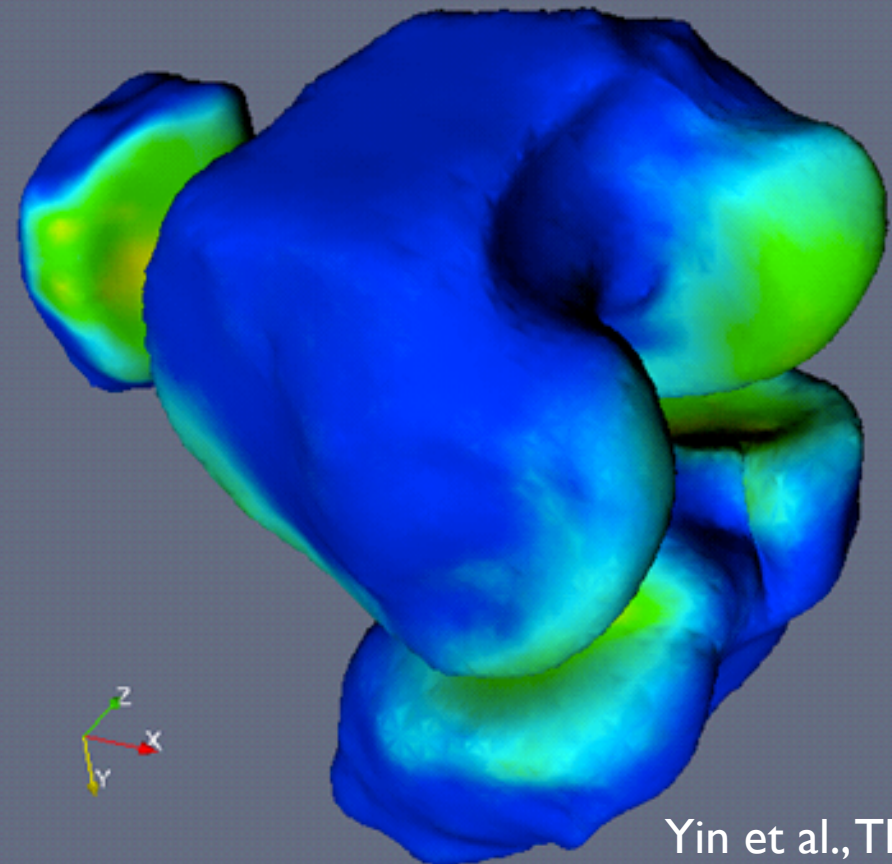
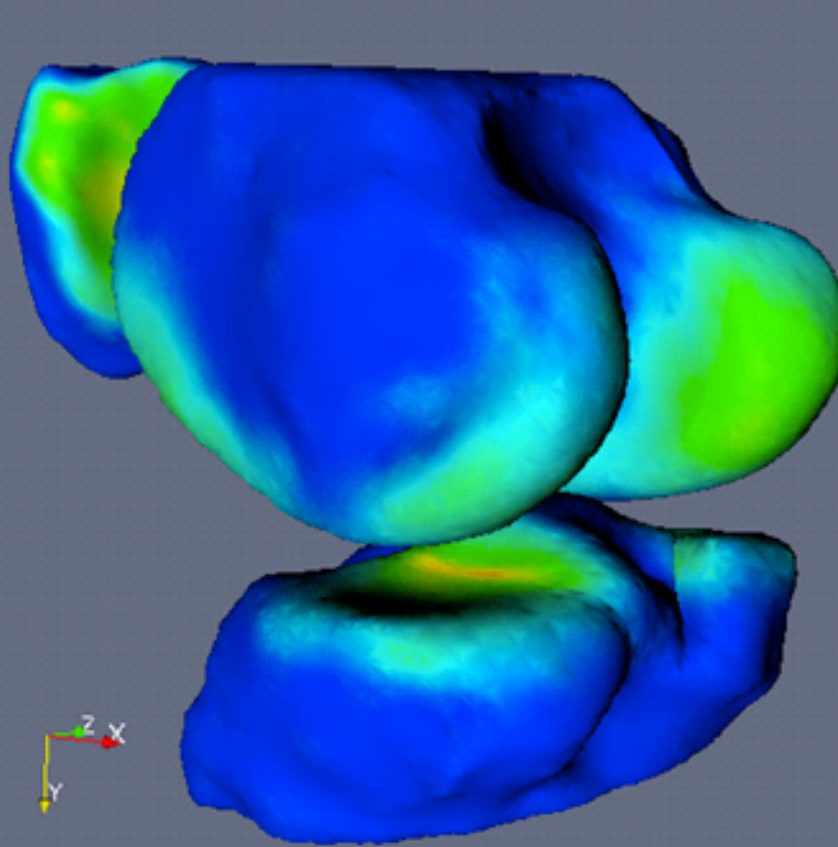


Yin et al., TMI, 2010



Example other LOGISMOS applications: determining cartilage thickness

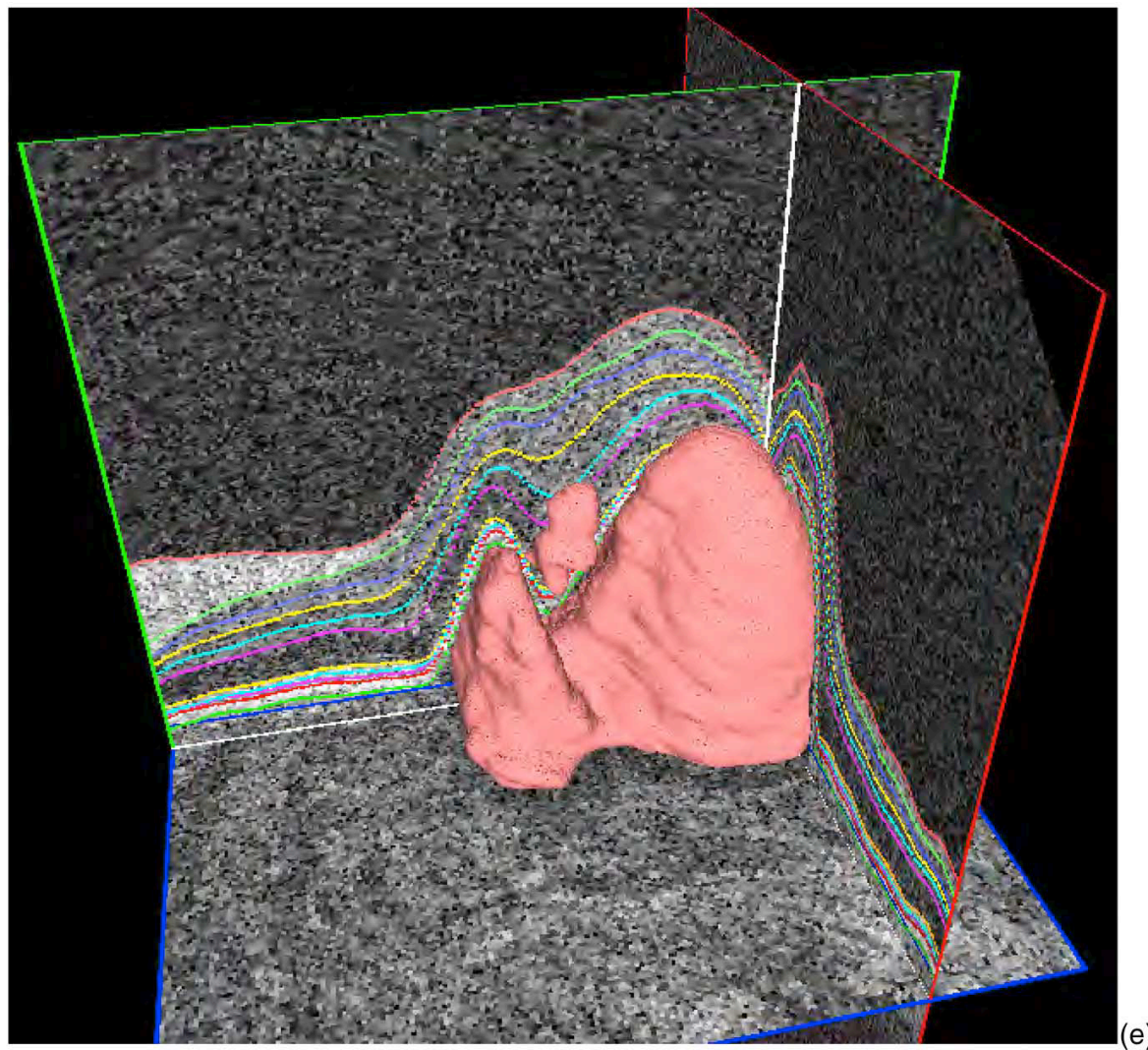
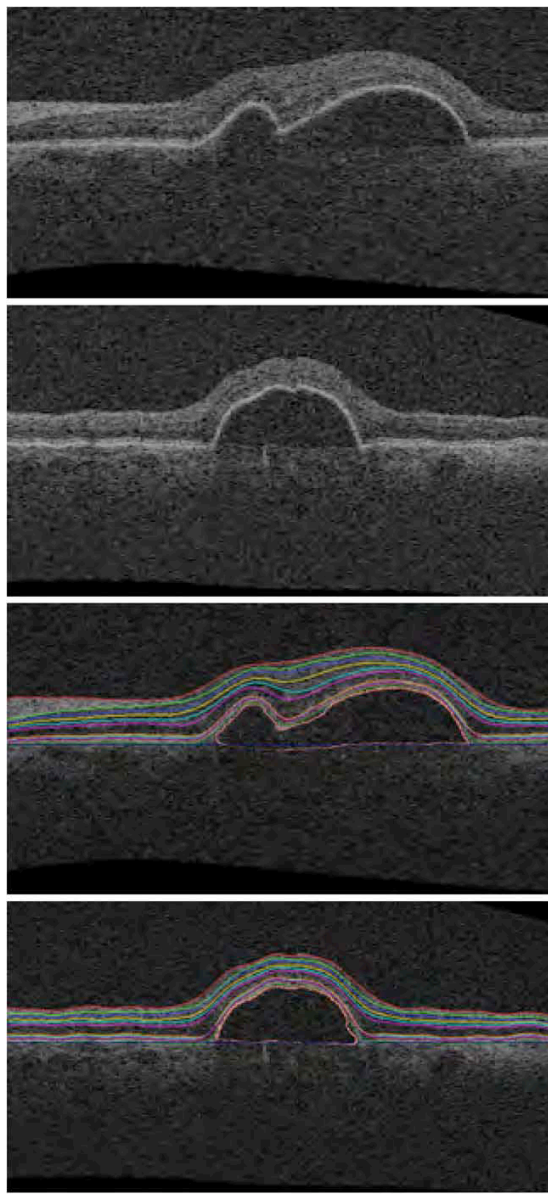
(from Sonka, et al.)



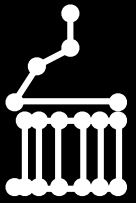
Yin et al., TMI, 2010



Challenge: layers with disruptions



Quellec et al., TMI 2010, Abramoff et al., R-BME, 2010



Summary

- The LOGISMOS approach enables the optimal and simultaneous segmentation of multiple surfaces and/or objects in polynomial time.
- Example applications include intraretinal layer segmentation, knee cartilage segmentation, vascular segmentation, airway segmentation, ...