

Segmentation

- Threshold Segmentation
 - Local thresholding
 - Edge thresholding
 - Threshold using averaging
 - Gradient Detectors
- Region Growing
- Split & Merge

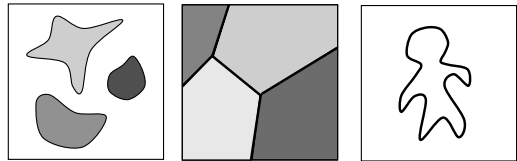
- Shape Matching
- Shape Representation



Segmentation

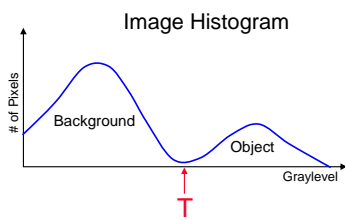
Image Segmentation = divide image into (continuous) regions or sets of pixels.

- 1) Region Based
- 2) Boundary Based
- 3) Edge Based

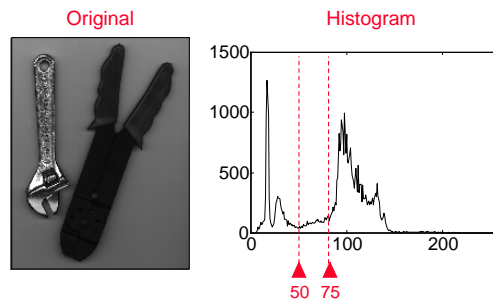


Thresholding

Global Thresholding = Choose threshold T that separates object from background.

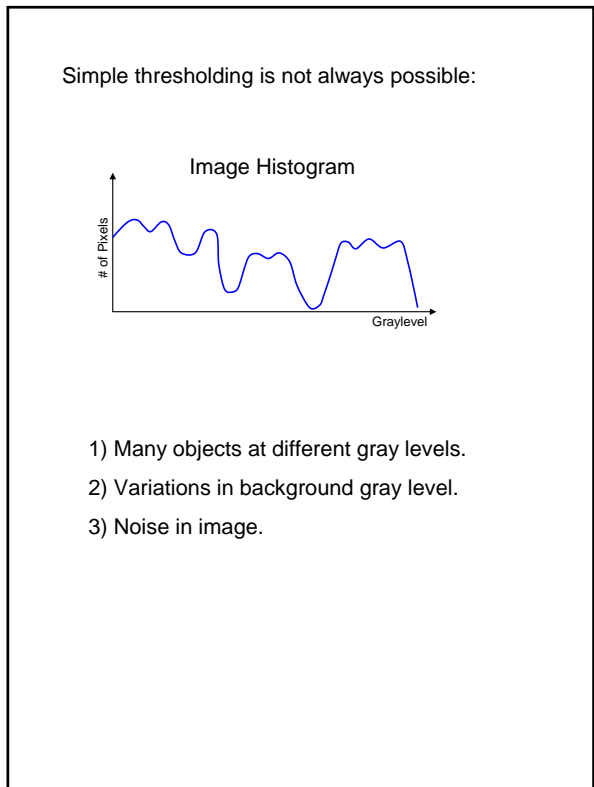
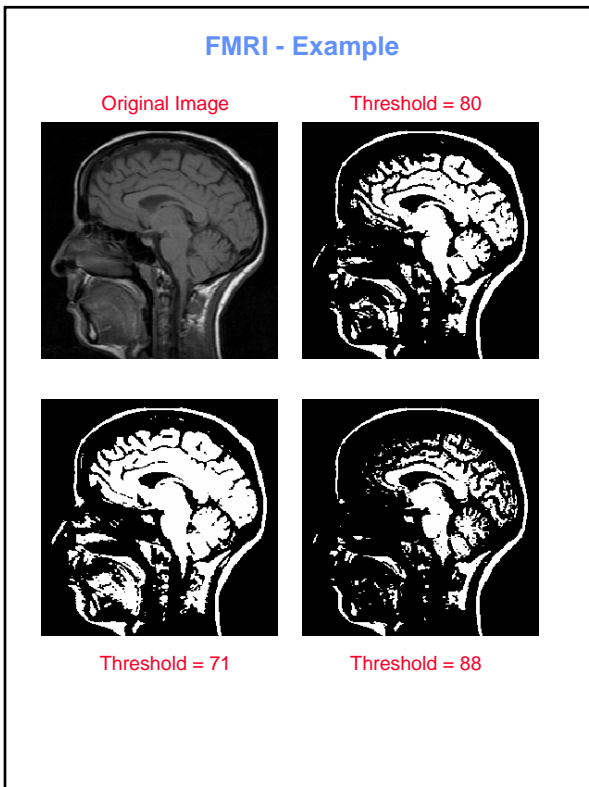
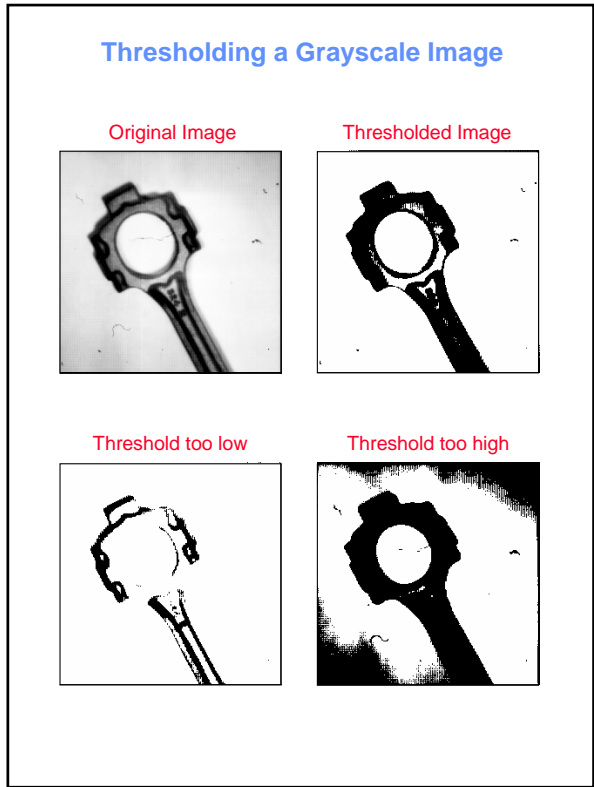
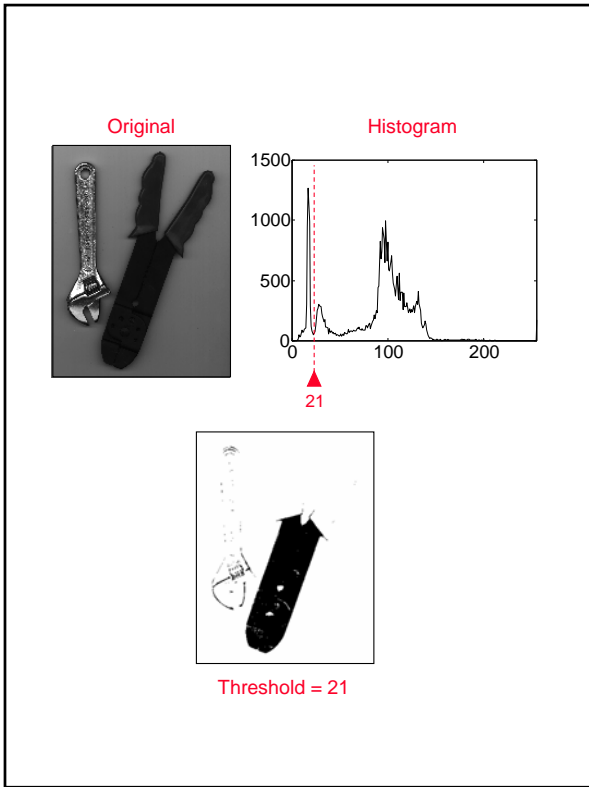


Segmentation using Thresholding



Threshold = 50

Threshold = 75

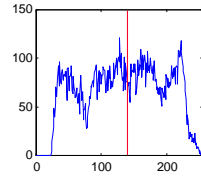


Thresholding Example

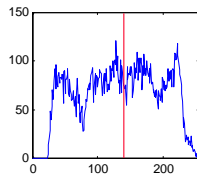
Original



Histogram



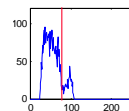
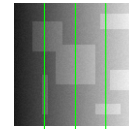
Single Global Threshold



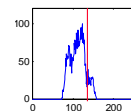
T = 128

Local Thresholding - 4 Thresholds

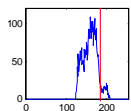
Divide image into regions. Perform thresholding independently in each region.



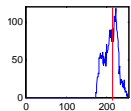
T = 80



T = 128



T = 188

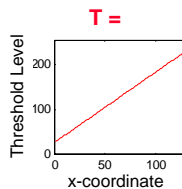


T = 226



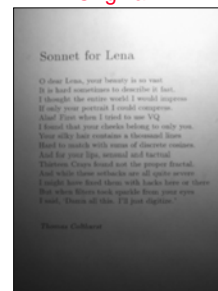
Adaptive Thresholding

Every pixel in image is thresholded according to the histogram of the pixel neighborhood.

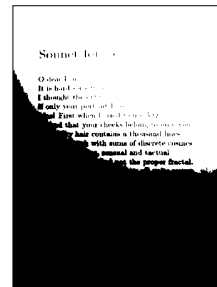


Adaptive Thresholding - Example

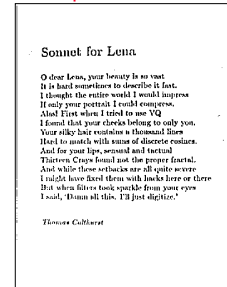
Original



Global Threshold

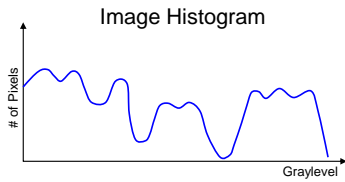


Adaptive Threshold

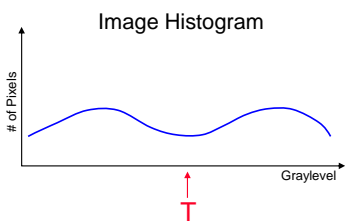


Threshold Segmentation of Noisy Images

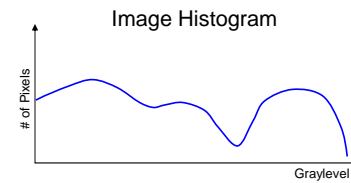
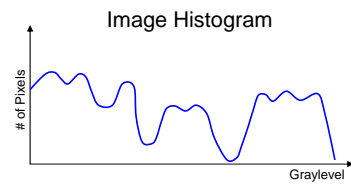
Noise inhibits localization of threshold.



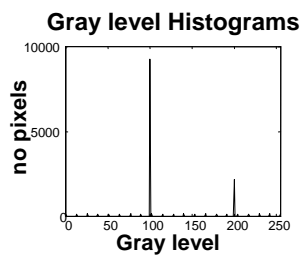
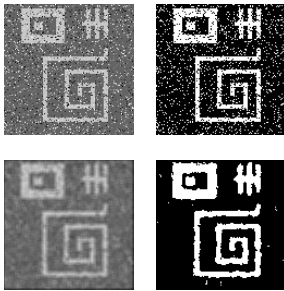
Smooth image and obtain a histogram for which threshold is easily determined.



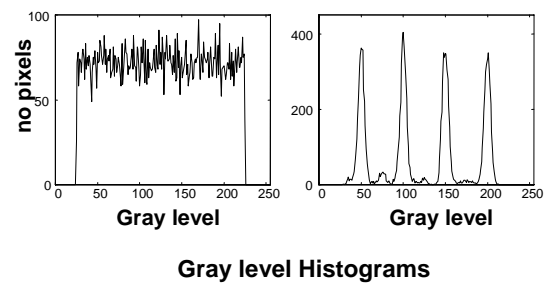
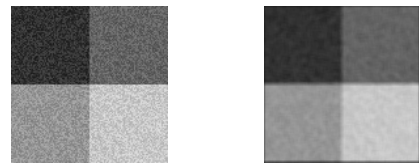
Note: Smooth the image, not the histogram...



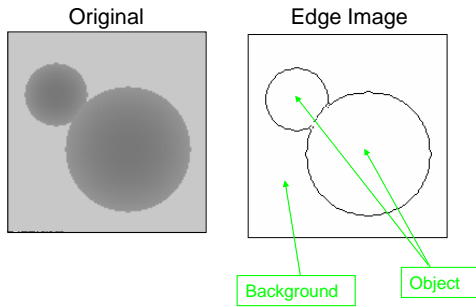
Threshold using Average



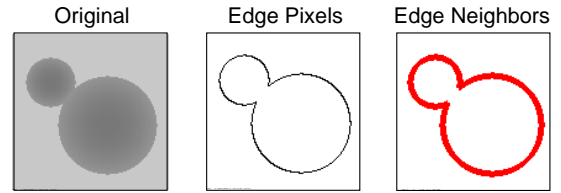
Threshold using Average



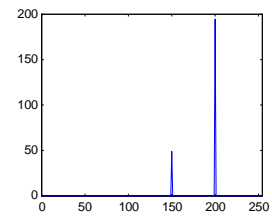
Edge Based Segmentation



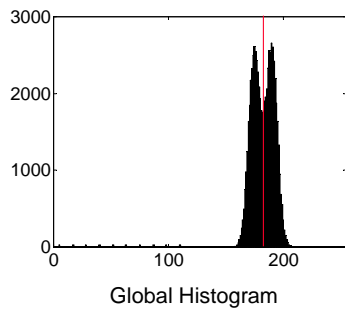
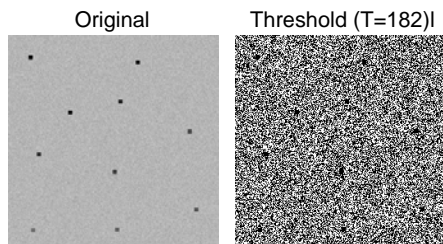
Edge Based Thresholding



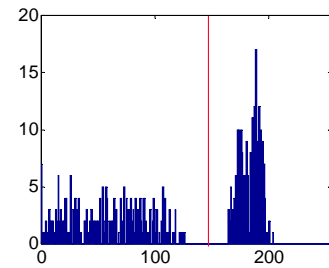
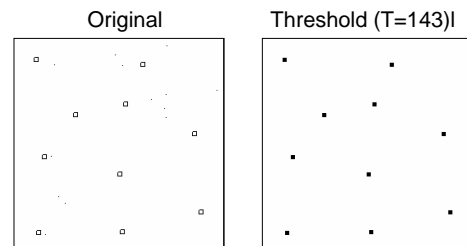
Edge Neighbors Histogram



Thresholding Based on Boundary Characteristics



Thresholding Based on Boundary Characteristics



Region Growing

Define:

S = the set of pixels inside the region.

Q = queue of pixels to be checked.

(x_0, y_0) = a pixel inside the region.

Algorithm:

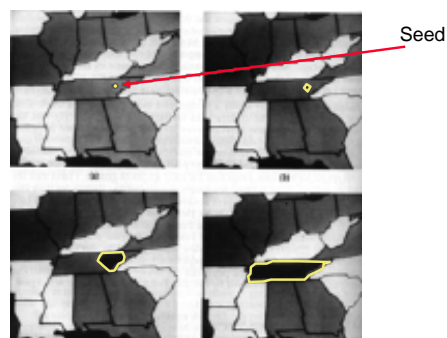
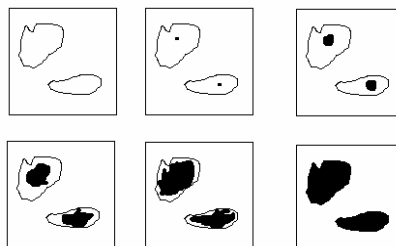
Initialize: $S = \emptyset$

$Q = \{ (x_0, y_0) \}$

- 1) Extract pixel P from queue Q
- 2) Add P to S .
- 3) For each neighbor P' of P :
 - if P' is "similar" to P and $P' \notin S$ then
 - add P' to Q .
- 4) If $Q = \emptyset$ then end, else return to 1.

S = the extracted pixels of the region.
 Define what "similar" means.
 Problematic in small gradient regions.

Region Growing - Example

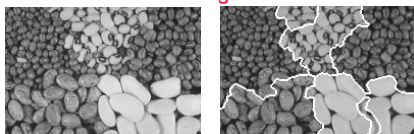


Region Growing - Examples

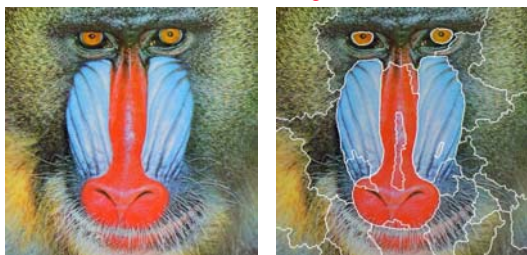
Color Segmentation



Texture Segmentation

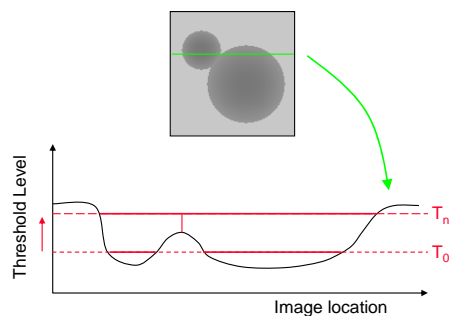
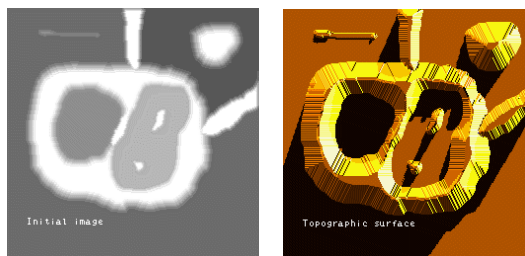


Color + Texture Segmentation

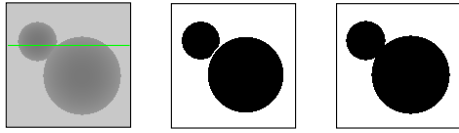


Watershed Threshold Algorithm

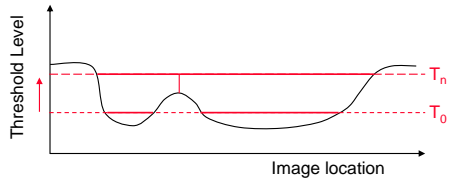
An Image can be viewed as a topographic map



Watershed Threshold Algorithm

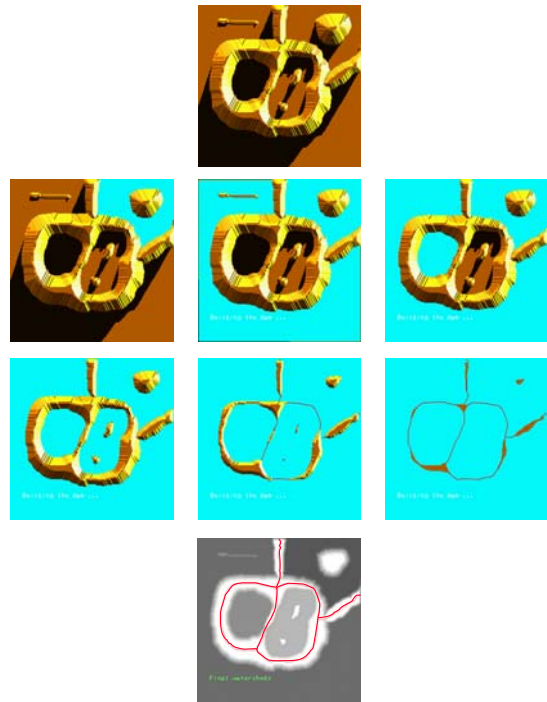


Original T = 149 T = 150

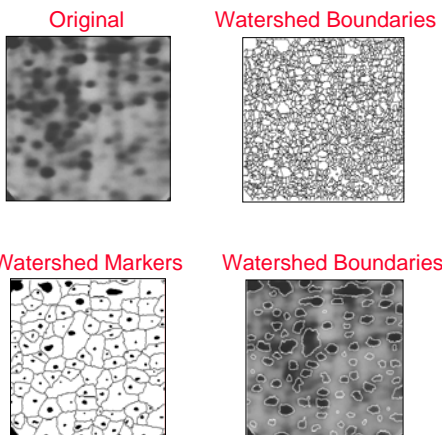


Initialize threshold at T_0 that separates objects well.
 Determine connected components.
 Raise threshold, and detect pixels that pass threshold and belong to more than one connected component.
 Do not let objects merge.
 Set these pixels as object boundaries.

Watershed Threshold Algorithm



Watershed Threshold Algorithm



Watershed Markers may be chosen manually or local global maximas (as above)

Split & Merge Segmentation

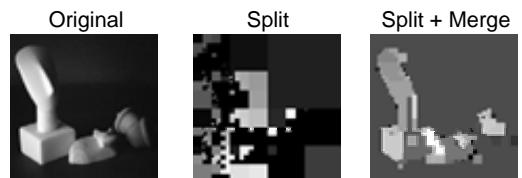
2 Stage Algorithm:

Stage 1: Split

Split image into regions using a Quad Tree representation.

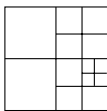
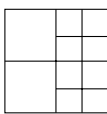
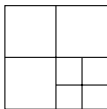
Stage 2: Merge

Merge "leaves" of the Quad Tree which are neighboring and "similar".

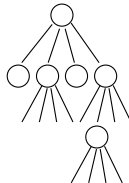
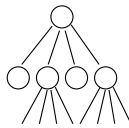
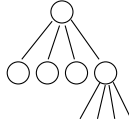
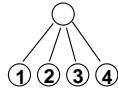


Quad Tree - Representation

Image



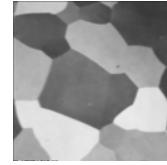
Quad Tree



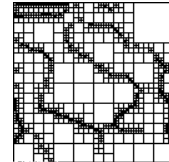
Demo

Quad Tree Representation

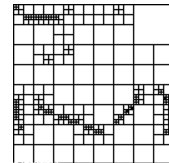
Original



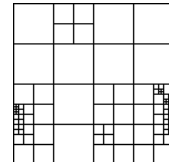
Thresh = 0.20



Thresh = 0.40

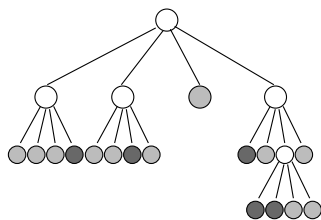
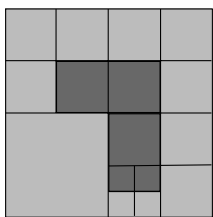


Thresh = 0.55

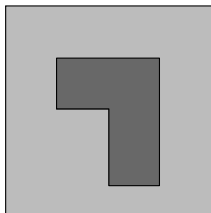


Split & Merge Example

Stage 1: Split



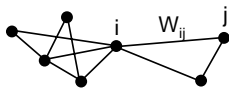
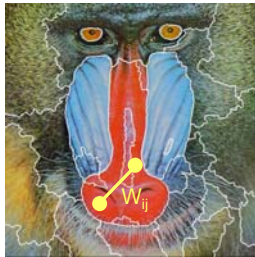
Stage 2: Merge



Split & Merge Example



Graph-Cut Segmentation

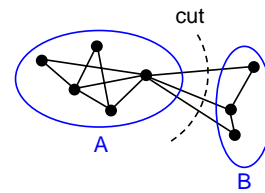


$$G = \{ V, E \}$$

V = vertices V = image pixels
 E = Edges E = pixel similarity

Segmentation = Graph Partitioning

Min-Cut Segmentation



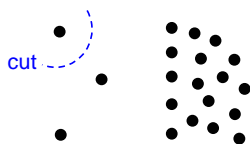
$$\text{cut}(A, B) = \sum_{i \in A, j \in B} W_{ij}$$

Segmentation by min-cut:

Find A, B such that $\text{cut}(A, B)$ is *minimal*.

(Wu and Leahy 1993)

Normalized-Cut Segmentation



Min-cut segmentation favors small segments.

Segmentation by normalized-cut:

Find A, B such that $\text{Ncut}(A, B)$ is *minimal*.

$$\text{Ncut}(A, B) = \sum_{i \in A, j \in B} W_{ij} \left(\frac{1}{\text{vol}(A)} + \frac{1}{\text{vol}(B)} \right)$$

where $\text{vol}(A) = \sum_{i \in A, j \in A} W_{ij}$

(Shi and Malik 2000)

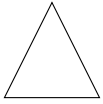
Normalized-Cut Segmentation - Examples



(from Cohen-Or 2005)

Shape Matching / Object Recognition

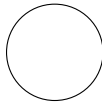
Model #1



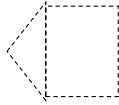
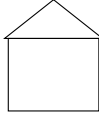
Model #2



Model #3



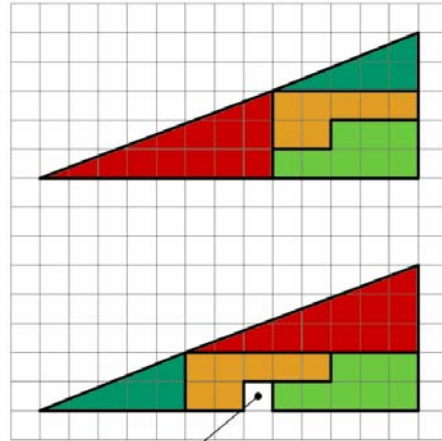
Model #4



Which Model matches the Measurement?

- Which Model
- What is the transformation from Model to Measurement (translation, rotation, scale,...)

HOW CAN THIS BE TRUE ?



Below the four parts are moved around

The partitions are exactly the same, as those used above

From where comes this "hole" ?