

Segmentation of Objects in Images and Videos

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Introduction

What is segmentation of images and videos?

- The segmentation of objects in images and videos aims at extracting an object of interest from the background.

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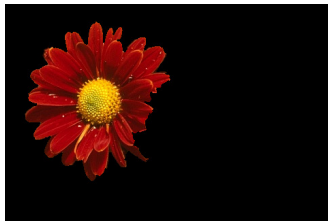
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- And what is the background? → Highly dependent on the context.

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- acquire and **model** object information;
- **enhance** regions wherein image properties are similar to those of the object, background, or their transition;
- **locate** the object and **delineate** its spatial extent in the image.

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- Computers are more precise than humans in object delineation.

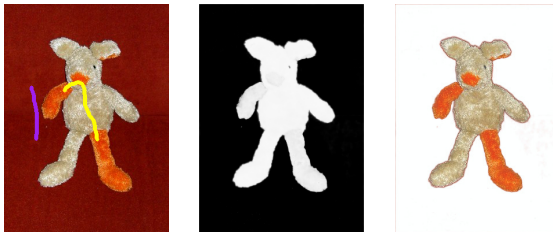
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In this example, delineation was solved by object enhancement followed by binarization (i.e., a spel classification), with no need for **connectivity**.

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Simple connectivity is needed for delineation when object and other disconnected parts of the background have similar properties.



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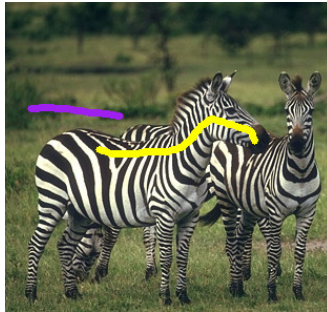
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However, when do we need **optimum connectivity**?

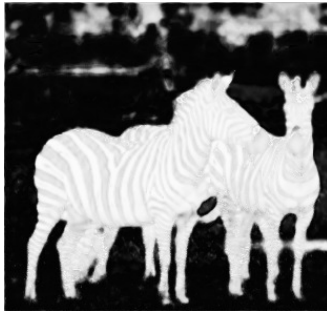
When do we need optimum connectivity?

Optimum connectivity is needed for delineation when object and parts of the background with similar properties are connected to each other.



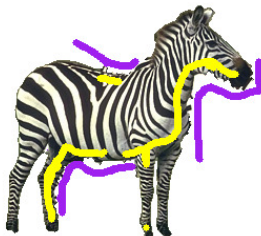
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In this case, however, some markers needed to disconnect the object are **not suitable** for object enhancement.

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- For automatic segmentation, we can exploit a **synergism** between object location by some **object model** and **computer** delineation.
- In both cases, delineation based on **optimum connectivity** can be used in the image domain and/or in the feature space by simple choice of the **adjacency relation**.

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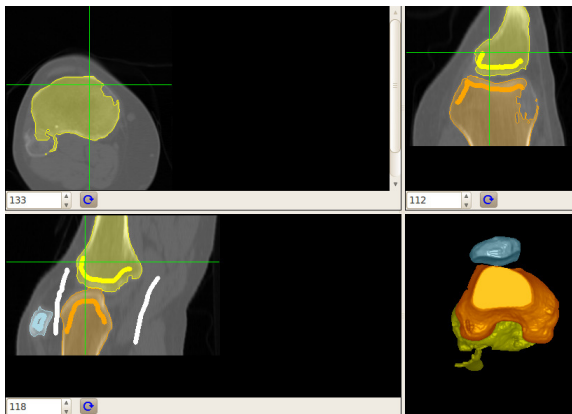
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We will focus on

- Object delineation using the image foresting transform: boundary-based, region-based, and hybrid approaches.
- A comparative analysis between the IFT and the min-cut/max-flow algorithms for region-based segmentation.
- Fuzzy object models and video segmentation.

Region-based object delineation

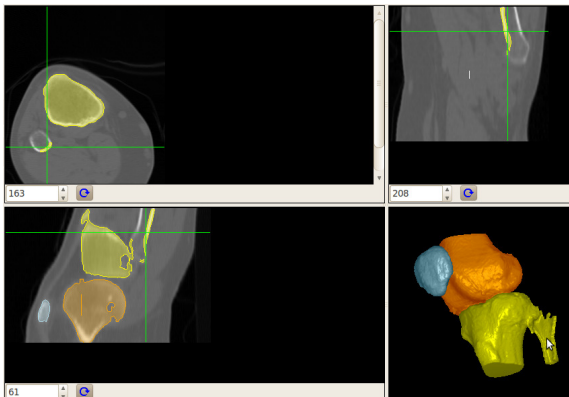
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- Interactive 3D visualization is crucial to help on object location and correction.

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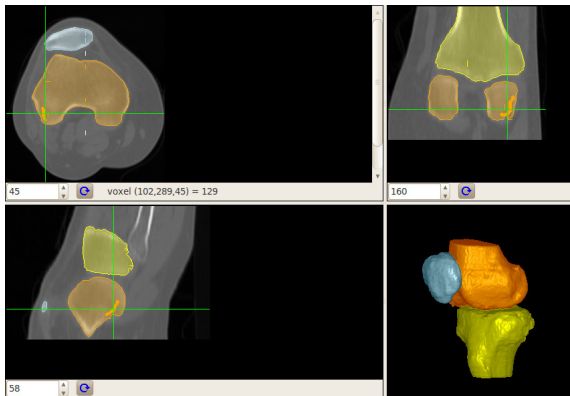
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Differential IFT with seed competition (IFTSC)

- In this case, the object is an optimum-path forest for f_{\max} rooted at internal seeds.

$$f_{\max}(\langle t \rangle) = \begin{cases} 0 & \text{if } t \in \mathcal{S} = \mathcal{S}_i \cup \mathcal{S}_e \\ +\infty & \text{otherwise} \end{cases}$$
$$f_{\max}(\pi_s \cdot \langle s, t \rangle) = \max\{f_{\max}(\pi_s), w(s, t)\},$$

where \mathcal{S}_i and \mathcal{S}_e are internal and external seed sets.

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- The dual formulation holds for fuzzy connected segmentation.

Differential IFT with f_{\max} and seed competition (IFTSC)

Algorithm

– DIFTSC - ALGORITHM

1. $(V, P, R, \mathcal{F}) \leftarrow \text{DIFT-FORESTREMOVAL}(V, P, R, \mathcal{A}, \mathcal{R}_M)$.
2. $\mathcal{F} \leftarrow \mathcal{F} \setminus S$.
3. **While** $S \neq \emptyset$, remove t from S , set $V(t) \leftarrow 0$,
4. \perp set $L(t) \leftarrow \lambda(t)$, $R(t) \leftarrow t$, $P(t) \leftarrow \text{nil}$, and $\mathcal{F} \leftarrow \mathcal{F} \cup \{t\}$.
5. **While** $\mathcal{F} \neq \emptyset$, remove t from \mathcal{F} and insert t in Q .
6. **While** Q is not empty **do**
7. Remove s from Q such that $V(s)$ is minimum.
8. **For each** $t \in \mathcal{A}(s)$, **do**
9. Compute $\text{tmp} \leftarrow \max\{V(s), w(s, t)\}$.
10. **If** $\text{tmp} < V(t)$ or $P(t) = s$, **then**
11. **If** $t \in Q$, **then** remove t from Q .
12. Set $P(t) \leftarrow s$, $V(t) \leftarrow \text{tmp}$, $R(t) \leftarrow R(s)$,
13. $L(t) \leftarrow L(s)$, and Insert t in Q .

Differential IFT with f_{\max} and seed competition (IFTSC)

Algorithm

– DIFT-FORESTREMOVAL

INPUT: *Maps V , P , R , adjacency \mathcal{A} , and set $\mathcal{R}_{\mathcal{M}}$ of selected roots.*

OUTPUT: *Maps V , P , and set \mathcal{F} of frontier pixels.*

AUXILIARY: *FIFO Queue T .*

1. Set $\mathcal{F} \leftarrow \emptyset$.
2. **For each** $t \in \mathcal{R}_{\mathcal{M}}$, **do** insert t in T , set $V(t) \leftarrow +\infty$ and $P(t) \leftarrow \text{nil}$.
3. **While** $T \neq \emptyset$, **do**
 4. Remove s from T .
 5. **For each** $t \in \mathcal{A}(s)$, **do**
 6. **If** $P(t) = s$, **then**
 7. Set $V(t) \leftarrow +\infty$, $P(t) \leftarrow \text{nil}$ and insert t in T .
 8. **Else If** $R(t) \notin \mathcal{R}_{\mathcal{M}}$, **then** $\mathcal{F} \leftarrow \mathcal{F} \cup \{t\}$.
9. Set $\mathcal{R}_{\mathcal{M}} \leftarrow \emptyset$.

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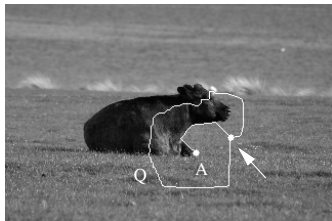
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They can be implemented by a sequence of IFTs using 4- or 8-adjacency relation and suitable connectivity function.

Live-wire-on-the-fly

In live-wire-on-the-fly (LWOF), optimum paths are incrementally computed from the moving wavefront Q .

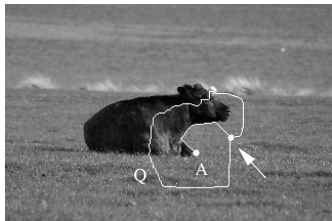
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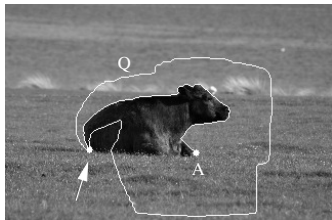
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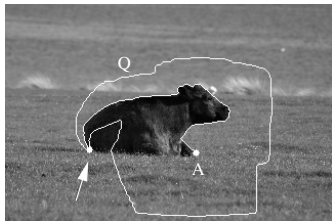
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- When the cursor is close to the boundary, the path snaps on to it.
- The user may accept it as a boundary segment, and
- the process is repeated from its terminus B until the user decides to close the contour.

Live-wire-on-the-fly (LWOF)

The IFT algorithm with **early termination** and function f_{sum} finds optimum paths from a starting point s^* on **counter-clockwise** oriented boundaries.

$$f_{sum}^{\circ}(\langle t \rangle) = \begin{cases} 0 & \text{if } t = s^* \\ +\infty & \text{otherwise} \end{cases}$$
$$f_{sum}^{\circ}(\pi_s \cdot \langle s, t \rangle) = \begin{cases} f_{sum}^{\circ}(\pi_s) + \bar{w}^{\beta}(s, t) & \text{if } O(l) \geq O(r) \\ f_{sum}^{\circ}(\pi_s) + K^{\beta} & \text{otherwise,} \end{cases}$$

where l and r are the spels at the left and right sides of arc $\langle s, t \rangle$. The weights $\bar{w}(s, t)$ are lower on the boundary than inside and outside it and $\beta \geq 1$ favors **longer segments**.

Path computation from s^* to u in LWOFF

Algorithm

– PATH COMPUTATION FROM s^* TO u IN LWOFF

1. **If $V(u) = +\infty$ or $u \in Q$, then**
2. Set $s \leftarrow nil$.
3. While $Q \neq \emptyset$ and $s \neq u$, do
4. Remove from Q a spel s such that $V(s)$ is *minimum*.
5. For each $t \in \mathcal{A}(s)$ such that $V(t) > V(s)$, do
6. **If $O(l) \geq O(r)$,**
7. **then** set $tmp \leftarrow V(s) + \bar{w}^\beta(s, t)$
8. **Else** set $tmp \leftarrow V(s) + K^\beta$.
9. **If $tmp < V(t)$, then**
10. **If $V(t) \neq +\infty$, remove t from Q .**
11. Set $P(t) \leftarrow s$ and $V(t) \leftarrow tmp$.
12. Insert t in Q .

Riverbed

Riverbed simulates the behavior of water flowing through a riverbed, always seeking lower ground levels, snaking through the river bends, instead of short-cutting the path as in live-wire. This leads to the following connectivity function for a starting seed point s^* :

$$f_w^\circ(\langle t \rangle) = \begin{cases} 0 & \text{if } t = s^* \\ +\infty & \text{otherwise} \end{cases}$$
$$f_w^\circ(\pi_s \cdot \langle s, t \rangle) = \begin{cases} \bar{w}(s, t) & \text{if } O(l) \geq O(r) \\ K & \text{otherwise.} \end{cases}$$

where l and r are the spels at the left and right sides of arc $\langle s, t \rangle$.

Riverbed

- Although f_w° is not smooth, it selects segments such that the **maximum** arc weight

$$\max_{\forall (l,r) \in \mathcal{A}', L(l)=1, L(r)=0} \bar{w}(l, r)$$

is **minimum**, considering all possible cuts in the dual graph $(\mathcal{N}, \mathcal{A}')$.

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- This implies that IFTSC and riverbed decide for the same optimum graph cut.
- Riverbed is more suitable than live-wire for more intricate shapes, but live-wire can jump weakly defined parts of the boundary.

Riverbed versus live-wire



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Riverbed versus live-wire



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- Riverbed asks for more user intervention on poorly defined parts.
- Their combination requires only two segments (live wire in cyan, riverbed in red).

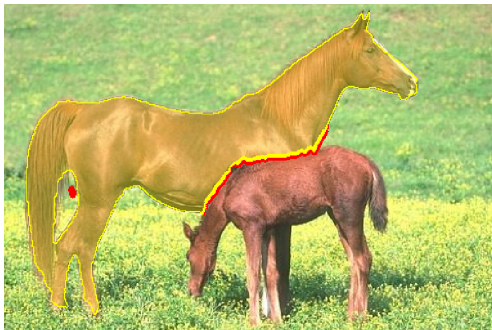
Live Markers

Live Markers is another hybrid approach that takes advantage from the superior ability of LWOFF on weakly defined segments and from IFTSC to handle complex 2D/3D shapes of multiple objects.



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The markers may be selected by the user or may come from the live-wire segments.

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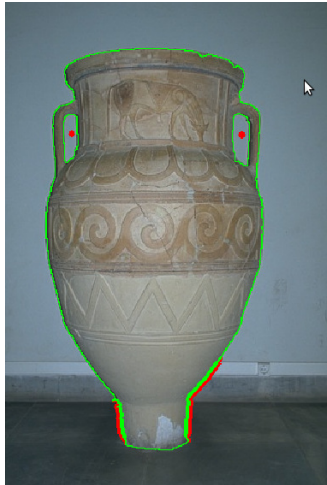
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- Object delineation using the image foresting transform.
- A comparison between IFTSC and object delineation using the min-cut/max-flow algorithm (GCMF).
- Fuzzy object models and video segmentation.

IFTSC and min-cut/max-flow algorithms

- It can be shown that the IFTSC computes the **graph cut** whose **minimum** arc weight

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- Similar region-based delineation could be obtained by GCMF as a **graph cut** whose sum of arc weights

$$\sqrt[\beta]{\sum_{\forall (s,t) \in \mathcal{A} \mid L(s)=1, L(t)=0} \bar{w}^\beta(s, t)}$$

is **minimum** for $\beta \geq 1$, with **lower** values favoring smaller cuts and **higher** values making both equivalent.

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- IFTSC is also **more robust** with respect to seed location than GCMF, but the latter provides **smoother** boundaries with **less leaking** than the former.
- Interestingly, GCMF and LWOFF are known to be **related** by dual graphs just as IFTSC and Riverbed.

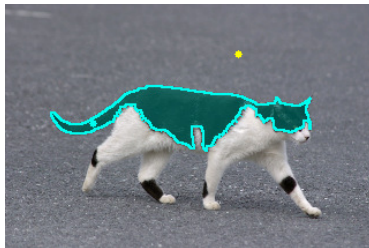
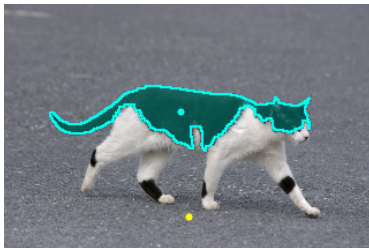
IFTSC and min-cut/max-flow algorithms

A lower β value allows GCMF (left) to obtain a smoother boundary, reducing the leaking of IFTSC (right).



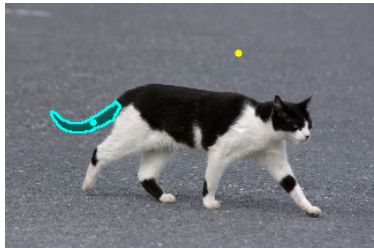
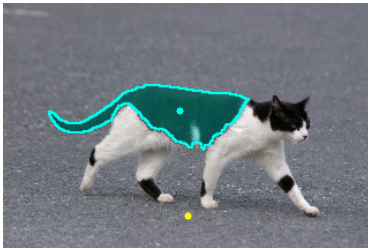
IFTSC and min-cut/max-flow algorithms

However, a same connected component is always obtained with IFTSC, independently of seed location.



IFTSC and min-cut/max-flow algorithms

The same **does not** happen in GCMF, when β is not **high enough**.



Graph Cuts using the Min-Cut/Max-Flow algorithm

Some GCMF approaches further change the graph topology by adding two virtual *source* and *sink* nodes, connected to every node in the graph

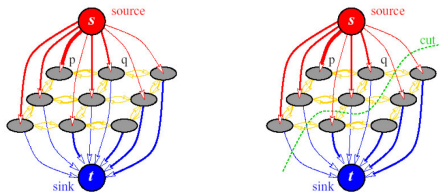


Image source: mathworks.com

and consider the following energy function in order to circumvent the drawbacks of smaller cuts

$$\sum_{\forall (s,t) \in \mathcal{A} | L(s)=1, L(t)=0} \tilde{w}(s, t) + \lambda \left(\sum_{\forall (s) \in \mathcal{I} | L(s)=1} \bar{P}_o(s) + \sum_{\forall (t) \in \mathcal{I} | L(t)=0} \bar{P}_b(t) \right),$$

where P is an object membership map (probability map).

Graph Cuts using the Min-Cut/Max-Flow algorithm

- However, this leads to a dependence on the **quality of the object membership map P** .

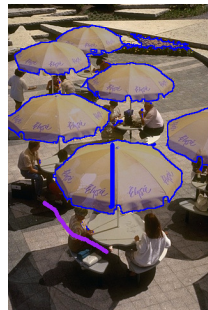
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- Moreover, the addition of the virtual nodes makes it **difficult to guarantee** that the resulting segmentation will be **spatially connected** to the user-drawn markers.



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- **Lower λ values** imply the **regular GCMF** minimum cut measure while **higher values** lead to simple **thresholding of P** .

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Registration is an **expensive** task that may **force** delineation to fit with the model irrespective to the local image information.

Fuzzy object models

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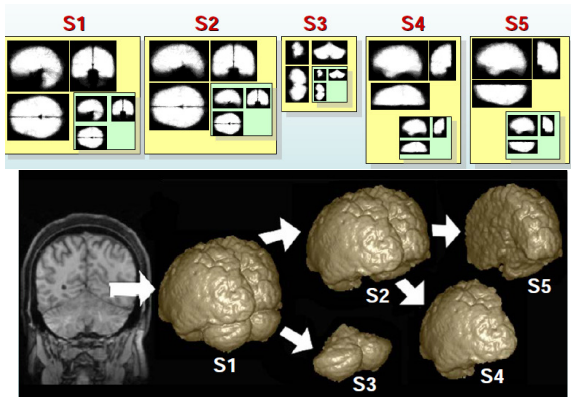
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- Segmentation is solved by translating the model and executing delineation at each location.
- It is possible to considerably **speed-up** this object search process by using multiple scales of the image and models.

Fuzzy object models

Examples of objects and their fuzzy models.

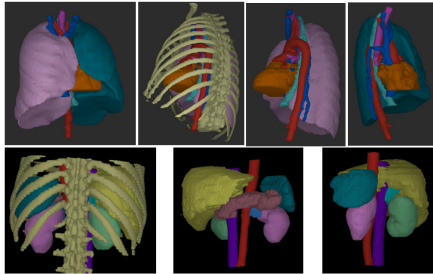
Medical imaging: Object modeling and image segmentation



This lecture will present only the first case, named **Cloud System Model (CSM)**, using IFTSC for delineation.

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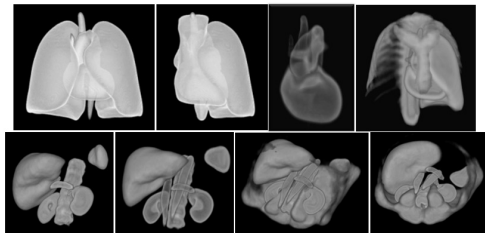
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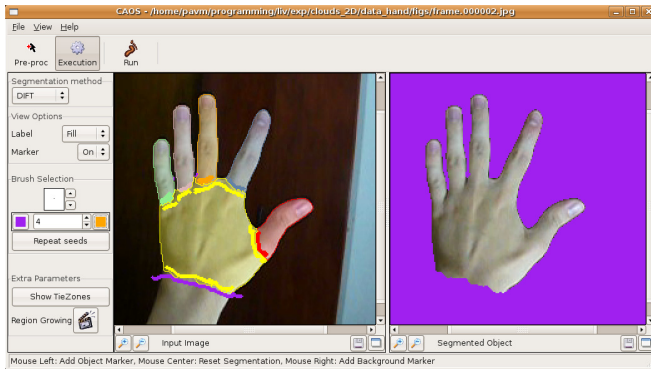
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The cloud system model

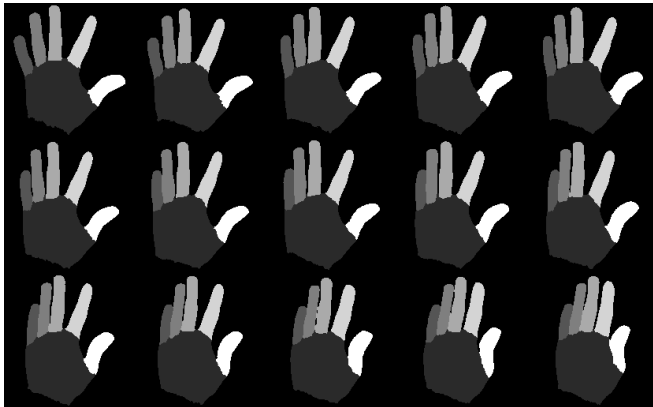
A set of training objects is first provided by interactive IFTSC segmentation.



Each image with multiple objects forms an **object system** with a common reference point (e.g., the geometric center of the objects).

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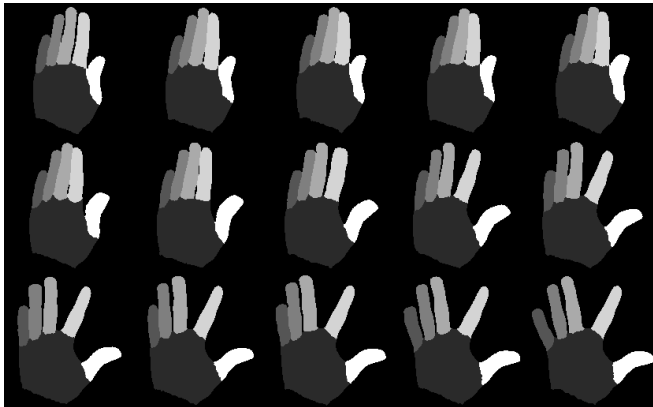
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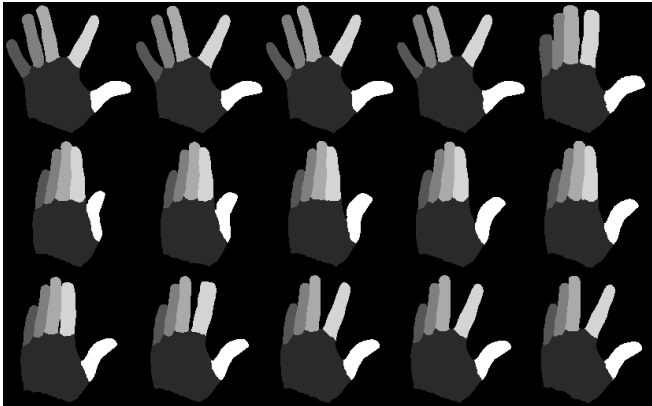
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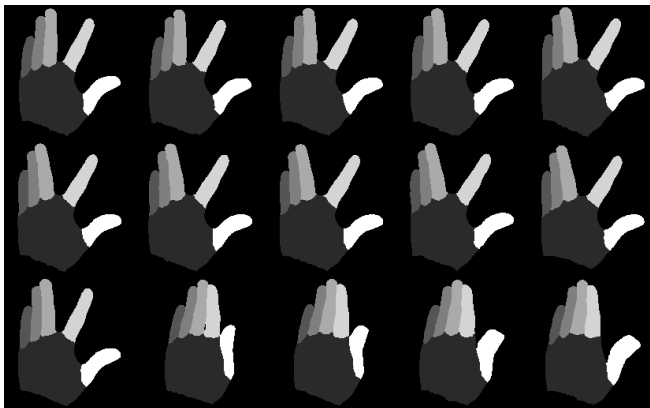
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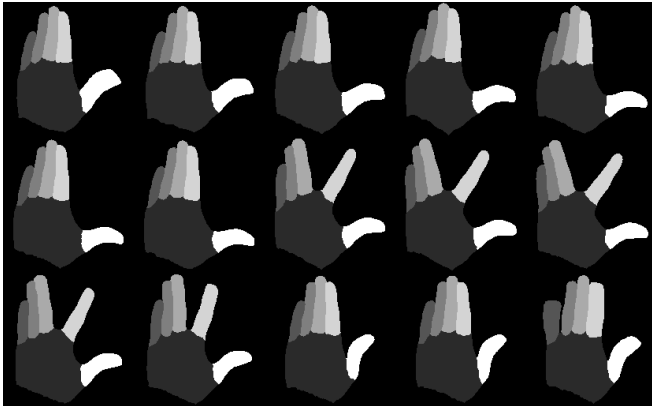
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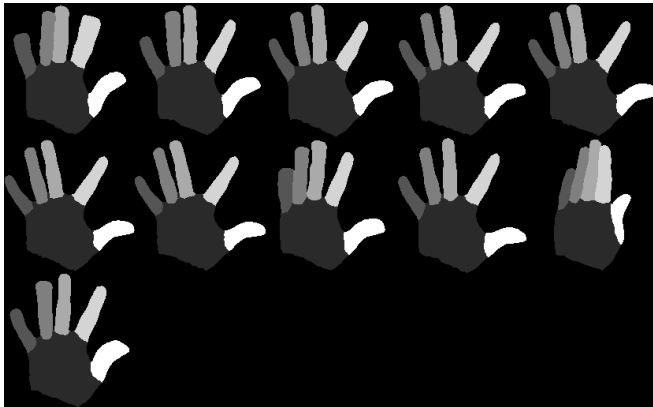
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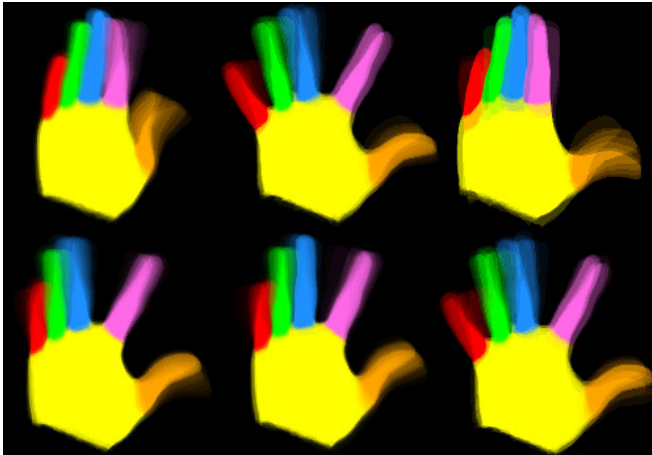
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- Each object system becomes a node of a complete graph, where the weight of each arc derives from the similarities between the corresponding objects in shape, size and position.
- The groups are found as **maximal cliques** in which all arc weights are higher than a threshold.

The cloud system model

The object systems in each group are finally translated to a same reference point and the corresponding object masks are averaged, forming a **set of cloud systems**.



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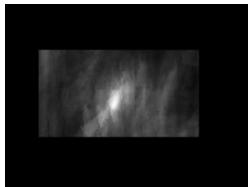
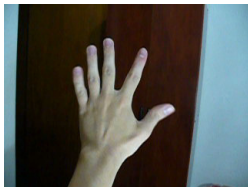
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- A fuzzy membership map (object clouds), which indicates an **object uncertainty region** with values strictly lower than 1 and higher than 0.
- A **delineation algorithm** (IFTSC), whose execution is constrained in the uncertainty region.
- A **criterion function**, which assigns a score to any set of delineated objects.

The cloud system model

Segmentation using CSM consists of a **search for the translation** to the image location which produces the highest score, when the reference point of the most suitable cloud system is at that position.



show video handsearch.mpg

The cloud system model in video

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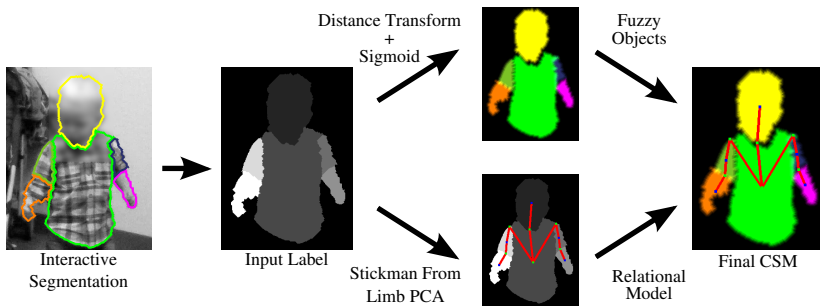
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- The remainder of the video is **automatically** segmented using CSM.

The cloud system model in video

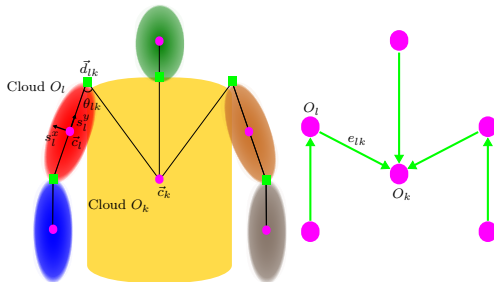
Creation of a CSM from a single segmentation mask in the first frame.



This process outputs a CSM with one cloud per body part.

The cloud system model in video

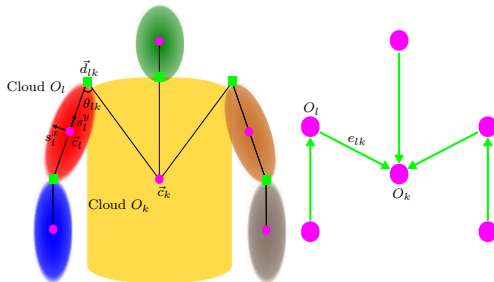
To deal with articulation, we have coupled the CSM with a **relational model** that is used to reposition the clouds during the search in a new test frame.



- The **torso translates** over the image and carries along the head and limbs.
- While the torso can be globally rotated, the remaining body part **joint angles are relative** between the connected parts.
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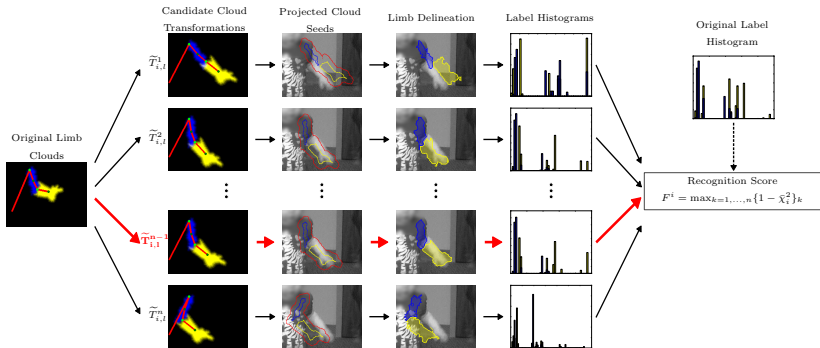
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- While the torso can be globally rotated, the remaining body part **joint angles are relative** between the connected parts.
- The scale is relative to the first frame for all body parts.
- The search for the body in a new frame consists of finding an optimal configuration for those parameters using **multi-scale parameter search (MSPS)**.

The cloud system model in video

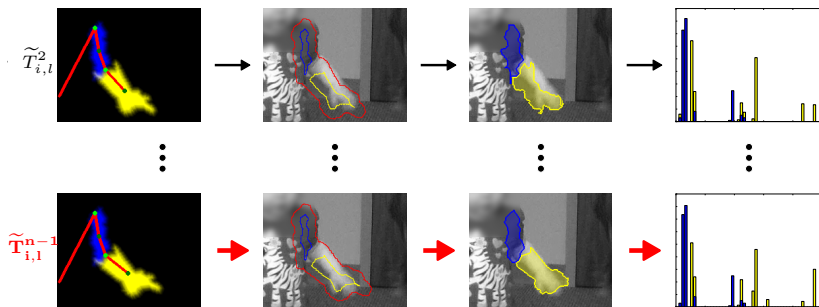
For every configuration of the relational model parameters chosen by MSPS, the CSM performs delineation and evaluates the criterion function.



In this case, the criterion function considers the χ^2 distance between color histograms extracted from both the first and the current frame.

The cloud system model in video

Delineation uses the regular IFTSC with the pixels surrounding the uncertainty region as seeds.



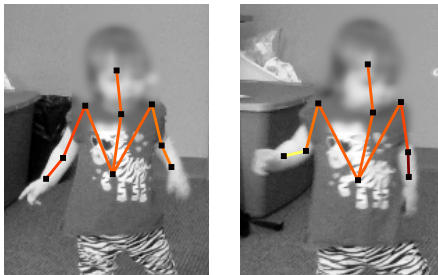
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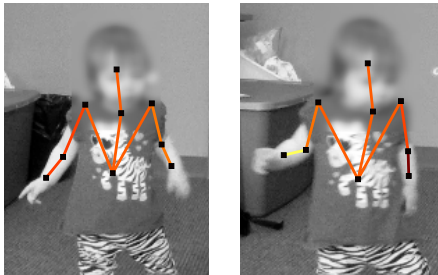
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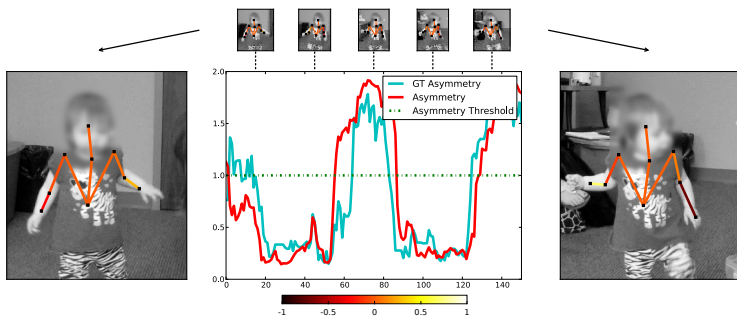
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- We use the 2D body pose to detect **arm asymmetry during unsupported gait**, a possible sign of autism in toddlers, from videos of real in-clinic ASD assessments.

But how does this aid autism assessment?

We provide quantifiable measures to aid the clinician in his/her assessment, which can be used for both research and diagnosis.



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- The synergistic combination between object models and delineation methods makes automatic segmentation feasible.
- When any automatic segmentation method fails, we have also developed solutions to reduce it into an optimum-path forest with minimum number of roots, so this facilitates corrections by the differential IFTSC algorithm.