

5. Object recognition in videos

Semantic video analysis

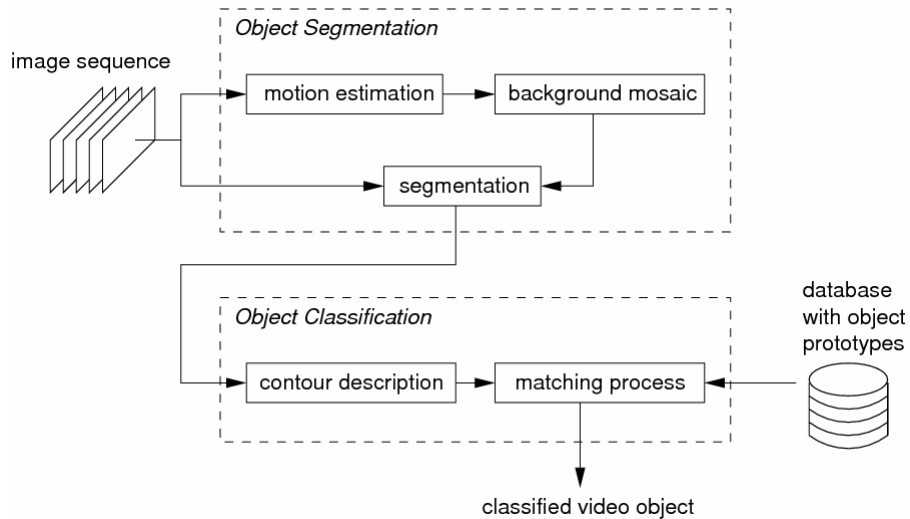
Stephan Kopf

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Outline

- General idea
- Object segmentation and parameterization
- Classification of objects
- Experimental results

General approach



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Segmentation (I)

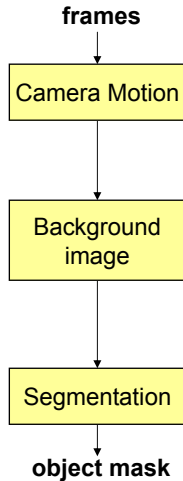
frames

Camera Motion

- Assumption: At least half of the visible area in each frame is background.
- Estimate the camera motion between consecutive frames.
- Calculate parameters of the camera model.



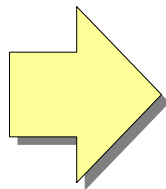
Segmentation (II)



- Assumption: At least half of the visible area in each frame is background.
- Estimate the camera motion between consecutive frames.
- Calculate parameters of the camera model.
- Apply a median filter on the transformed frames to construct the background image.
- Compare the background image with the transformed frame to get the object mask.

Object classification approach

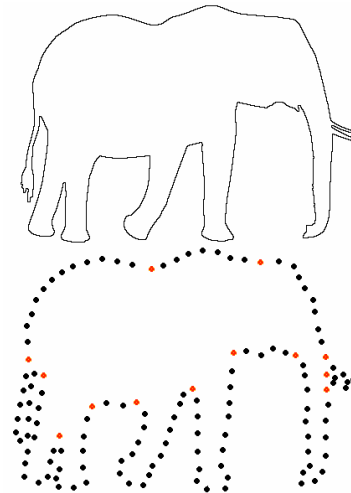
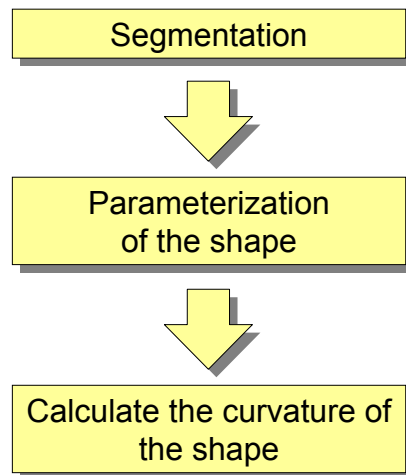
1. Sample the shape of an segmented object with a fixed number of sample points.
2. Identify feature points to describe the curvature of the shape.



Features are calculated based on the **Curvature Scale Space Image**

3. Compare the features of the unknown shape with features of objects that are stored in a database

Parameterization



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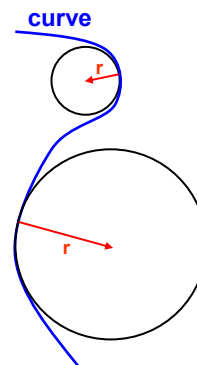
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Definition of the curvature (I)

- The curvature at a given point has a magnitude equal to the reciprocal of the radius of an osculating circle (the circle touches the curve):

$$K = \frac{1}{r}$$

- The curvature is a vector pointing to the direction of the circle's center.
- A small circle corresponds with a high curvature's magnitude, a straight line has a curvature of zero.



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Definition of the curvature (II)

- Consider a plane curve $\mathbf{u}(t)$ that lies completely within a 2D plane. $\mathbf{u}(t)$ is parameterized by the arc length t .
- The curve $\mathbf{u}(t)$ is parametrically defined by two functions $x(t)$ and $y(t)$:

$$\mathbf{u}(t) = (x(t), y(t)).$$

- Curvature K for a plane curve $\mathbf{u}(t)$:

$$K = \frac{\dot{x} \cdot \ddot{y} - \dot{y} \cdot \ddot{x}}{(\dot{x}^2 + \dot{y}^2)^{3/2}}$$

\dot{x} and \dot{y} define the first derivative (gradient),

\ddot{x} and \ddot{y} the second derivative (change of the gradient).

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Definition of the curvature (III)

- We can derive a less general definition of the curvature if we explicitly use plane curves defined by $y = f(x)$. We get the following definition of the curvature for each point $(x, f(x))$:

$$K = \frac{f''(x)}{(1 + (f'(x))^2)^{3/2}}$$

- This form is widely used in engineering:
 - to approximate the fluid flow around surfaces, e.g. in aerodynamics (gases) or hydrodynamics (liquids),
 - to derive the characteristic behavior when bending structural elements, e.g., put weight on a beam and analyze the flexure.

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Definition of the curvature (IV)

Example

- Consider the parameterized curve $\mathbf{u}(t) = (x(t), y(t)) = (t, t^2)$.
The explicit definition of this curve is: $y = f(x) = x^2$.

- Calculate curvature based on the **parameterized curve**:

First and second derivatives: $\dot{x} = 1$, $\ddot{x} = 0$, $\dot{y} = 2t$, $\ddot{y} = 2$.

$$K(t) = \frac{\dot{x} \cdot \ddot{y} - \dot{y} \cdot \ddot{x}}{(\dot{x}^2 + \dot{y}^2)^{3/2}} = \frac{1 \cdot 2 - 2t \cdot 0}{(1^2 + (2t)^2)^{3/2}} = \frac{2}{(1 + 4t^2)^{3/2}}$$

- Calculate curvature based on the **explicit definition**:

$$f'(x) = 2x, \quad f''(x) = 2 \quad K(x) = \frac{f''(x)}{(1 + (f'(x))^2)^{3/2}} = \frac{2}{(1 + 4x^2)^{3/2}}$$

Calculation of the curvature (V)

- Approximation of derivatives for discrete values
(parameterized shapes):

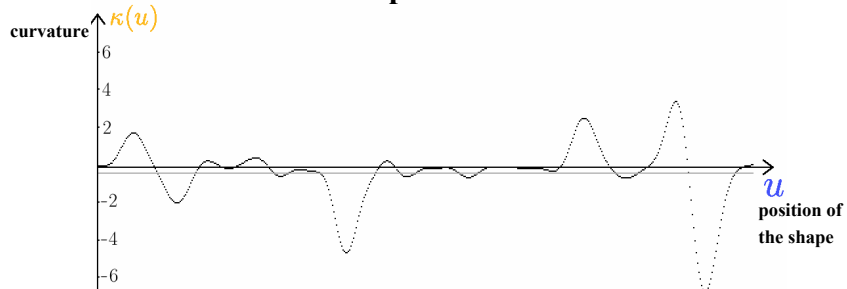
$$\dot{x}(t) = \frac{x(t+1) - x(t-1)}{2 \cdot hx}$$

$$\dot{y}(t) = \frac{y(t+1) - y(t-1)}{2 \cdot hy}$$

- Parameter t is defined for whole numbers ($t \in \mathbb{N}$).
- hx and hy normalize the derivatives depending on the distance between sample points.

Shape matching based on curvatures

Curvature function of a shape



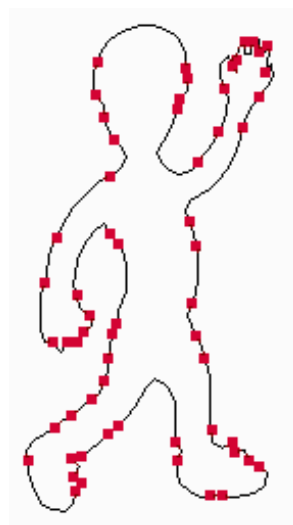
Problem: It is very difficult to match the curvature functions of two shapes.

→ Identify significant curvature features.

We use the curvature scale space technique.

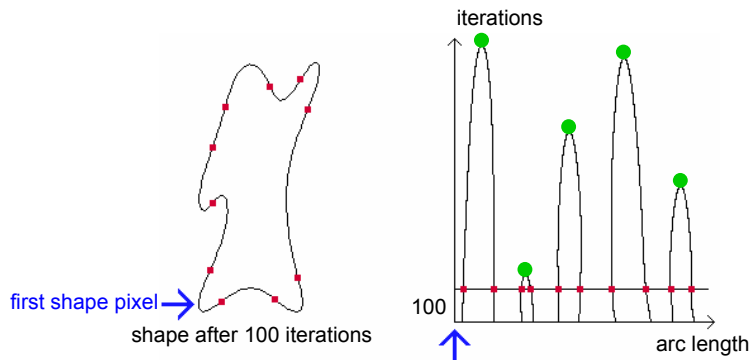
Curvature scale space (I)

- Analyze the outer shape of an object.
- Smooth the shape with a Gaussian kernel.
- The **inflection points** in each iteration are used as features to describe the object.



Curvature scale space (II)

- A **curvature scale space image** is a visual representation of the inflection points during the smoothing process.



- The **peaks** are used as features to describe the object.

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Curvature scale space (III)

Characteristics of CSS images

- The peaks in the curvature scale space image describe **concave segments** of a shape.
- The peaks are used as features to describe the shape.
- Each peak characterizes
 - a **position**
relative position compared to other peaks,
 - a **value**
strength of the concave segment.

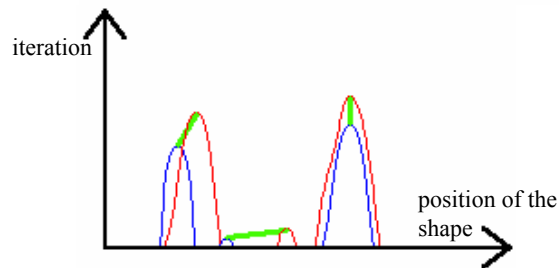
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Comparison of two shapes

1. Shift one curvature scale space image until the largest peaks match (makes the approach invariant to rotation)
2. Calculate the Euclidian distance between two peaks

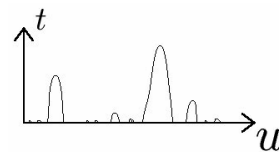
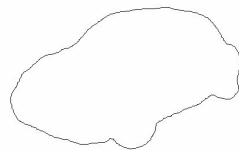
$$D = \sqrt{d_u \cdot d_u + d_t \cdot d_t}$$



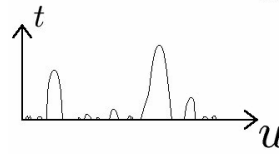
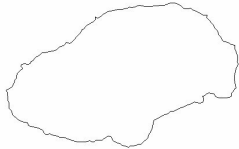
3. Summarize the distances

Features of CSS images (I)

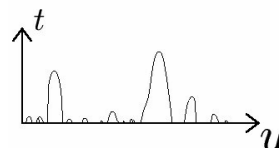
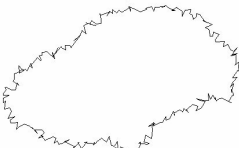
- Original shape of a car



- Shape with noise



- Shape with severe noise



→ CSS images are very similar

Features of CSS images (II)

Pros

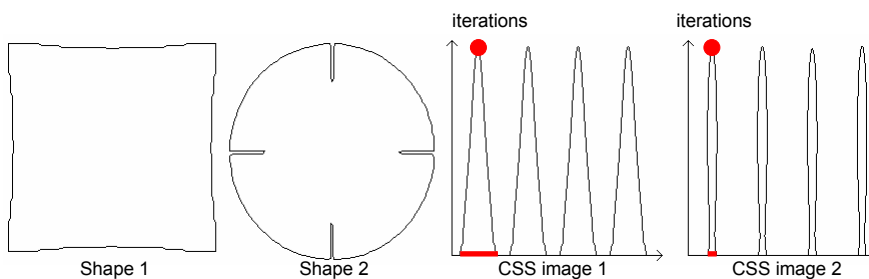
- Only a few values are required to describe complex objects.
- The approach is invariant to rotation or scaling.
- Low computation time.

Cons

- Bad classification results with some shapes:
 - shallow and deep concavities
 - convex regions

Ambiguities of CSS Images (I)

Shallow vs. deep concavities

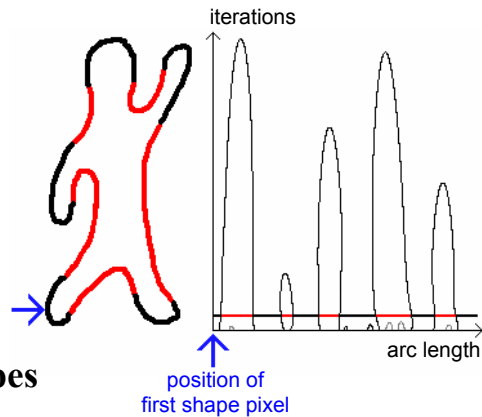


→ Solution: Use **position**, **height** and **width** of each peak as feature.

Ambiguities of CSS Images (II)

Convex regions

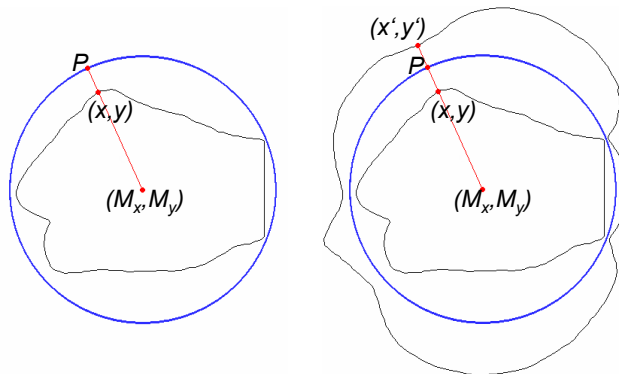
- Poor representation of convex regions of a shape.
- Convex objects cannot be represented at all.



→ **Solution: Mapped shapes**

Mapped Shapes (I)

- **Idea:** Reflect each shape pixel and create a new shape.
- Strong convex segments of the original shape become concave segments of the mapped shape.



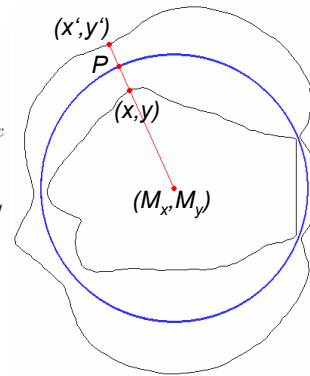
Mapped Shapes (II)

Calculation of mapped shapes

$$D_{x(u),y(u)} = \sqrt{(M_x - x(u))^2 + (M_y - y(u))^2}$$

$$x'(u) = \frac{2 \cdot R - D_{x(u),y(u)}}{D_{x(u),y(u)}} \cdot (x(u) - M_x) + M_x$$

$$y'(u) = \frac{2 \cdot R - D_{x(u),y(u)}}{D_{x(u),y(u)}} \cdot (y(u) - M_y) + M_y$$



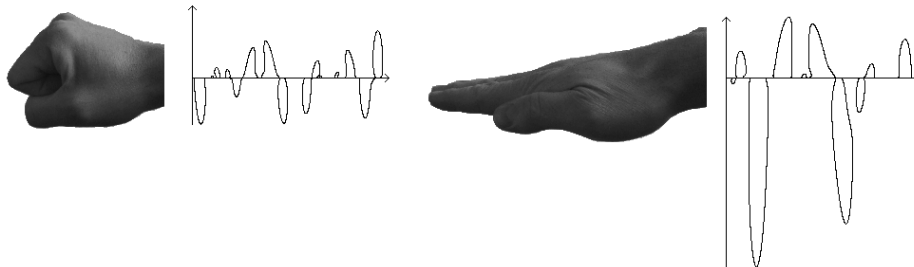
R : radius of the circle

D : distance between (M_x, M_y) and (x, y)

Mapped Shapes (III)

CSS images with mapped shapes

- Get standard curvature scale space features.
- Calculate features for the mapped shape.



Object recognition in videos (I)

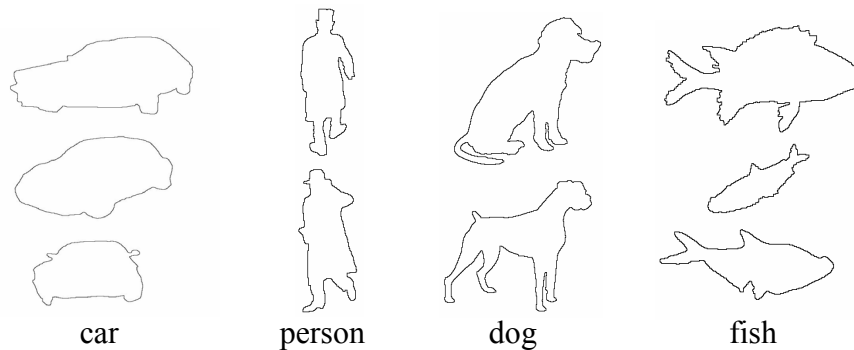
Approach

- Compare each object in the sequence with the objects in the database.
- Calculate average difference of the objects in the database to each object class.
- Display most similar objects.

Object recognition in videos (II)

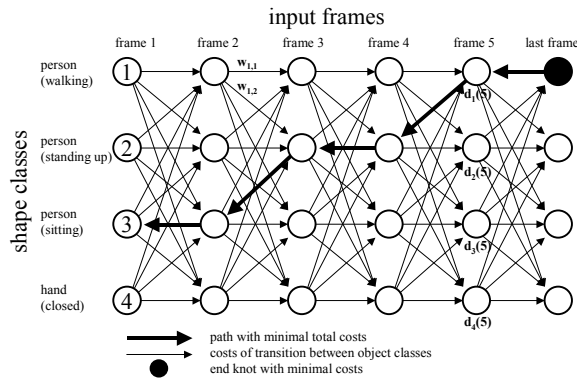
Objects in the database

- 300 objects are stored in the database.
- 13 object classes group similar objects.



Object recognition in videos (III)

Aggregation of object classification results



- Calculate distance $d_c(i)$ between input object i and object class c .
- Transition costs $w_{n,m}$ occur for each change of the object class.

- Solve minimization problem: $\min_c \sum_{i=1}^N (d_{c(i)}(i) + w_{c(i),c(i-1)})$

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Object recognition in videos (IV)

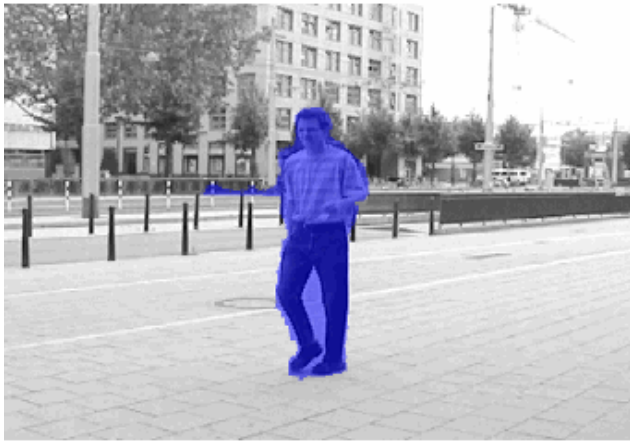
Recognition rates

- Recognition rates between 25-95 % (depends on the complexity of the object and the number of objects in the database).
- The recognition rates of rigid objects (e.g., a car) is much higher compared to deformable objects.
- The curvature scale space approach is invariant to scaling and rotation, and it is very robust to noise
- The comparison is very fast (smooth shape once and calculate Euclidean distances)

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Object recognition in videos (V)



standing
walking
turning around
sitting down
sitting
open hand
closed hand
fist
thumb

Questions ?