

# **5. Object recognition in videos**

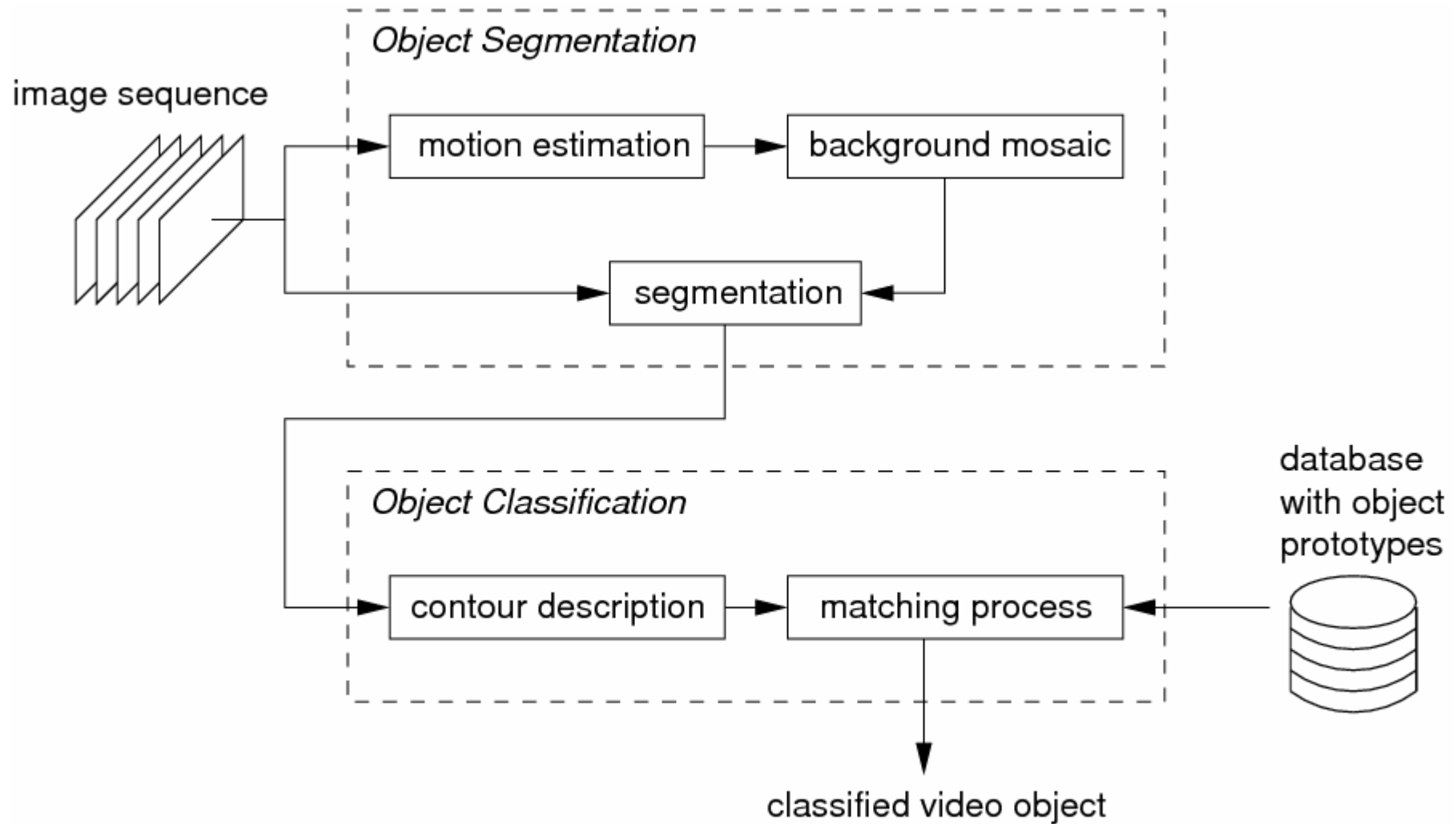
## **Semantic video analysis**

Stephan Kopf

# Outline

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

# General approach



# Segmentation (I)

frames

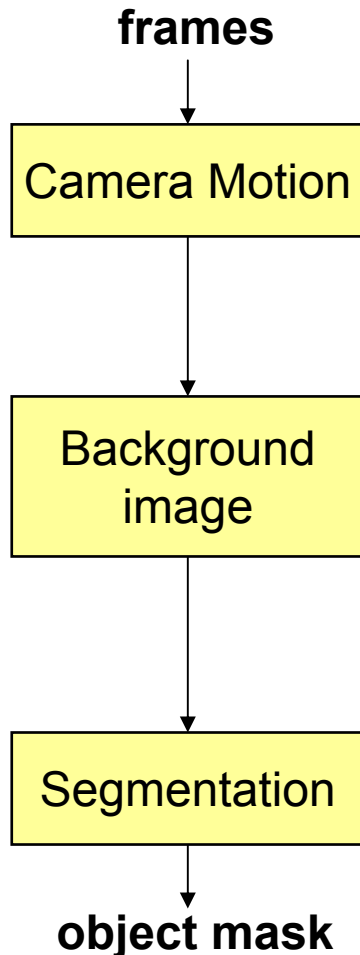


Camera Motion

- 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.



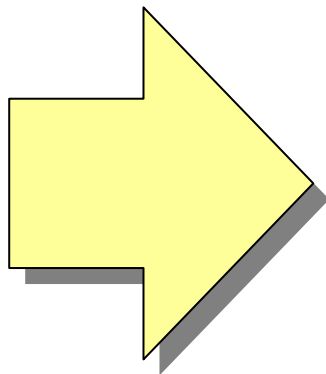
# Segmentation (II)



- 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.

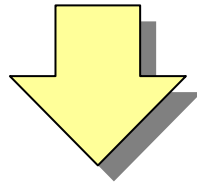


Features are calculated  
based on the  
***Curvature Scale Space Image***

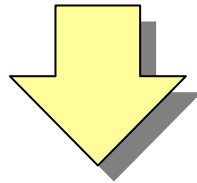
3. Compare the features of the unknown shape with features of objects that are stored in a database

# Parameterization

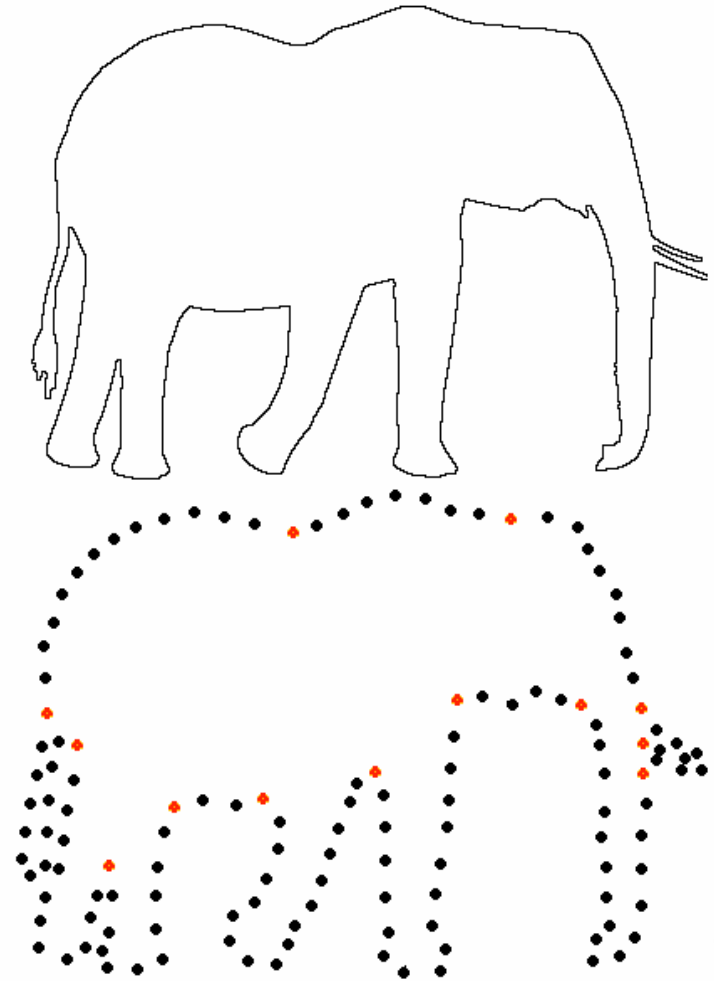
Segmentation



Parameterization  
of the shape



Calculate the curvature of  
the shape

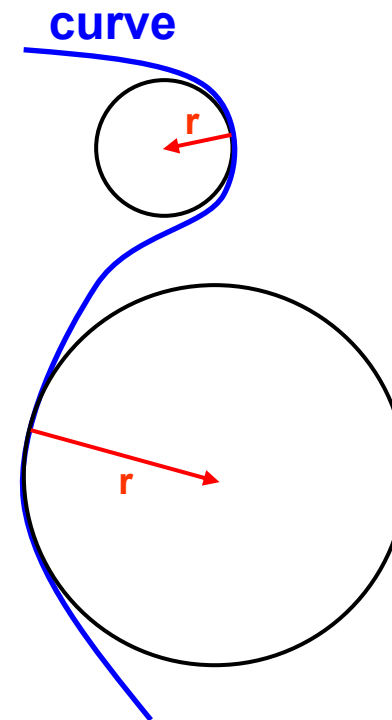


# Definition of the curvature (I)

- 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.





# 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).

# 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  $(x, f(x))$ :

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

- 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.

# Definition of the curvature (IV)

## Example

- Consider the parameterized curve  $\mathbf{u}(t) = (\mathbf{x}(t), \mathbf{y}(t)) = (t, t^2)$ .  
The explicit definition of this curve is:  $\mathbf{y} = \mathbf{f}(\mathbf{x}) = \mathbf{x}^2$ .
- Calculate curvature based on the **parameterized curve**:  
First and second derivatives:  $\dot{\mathbf{x}} = 1$ ,  $\ddot{\mathbf{x}} = 0$ ,  $\dot{\mathbf{y}} = 2t$ ,  $\ddot{\mathbf{y}} = 2$ .

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

- 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):

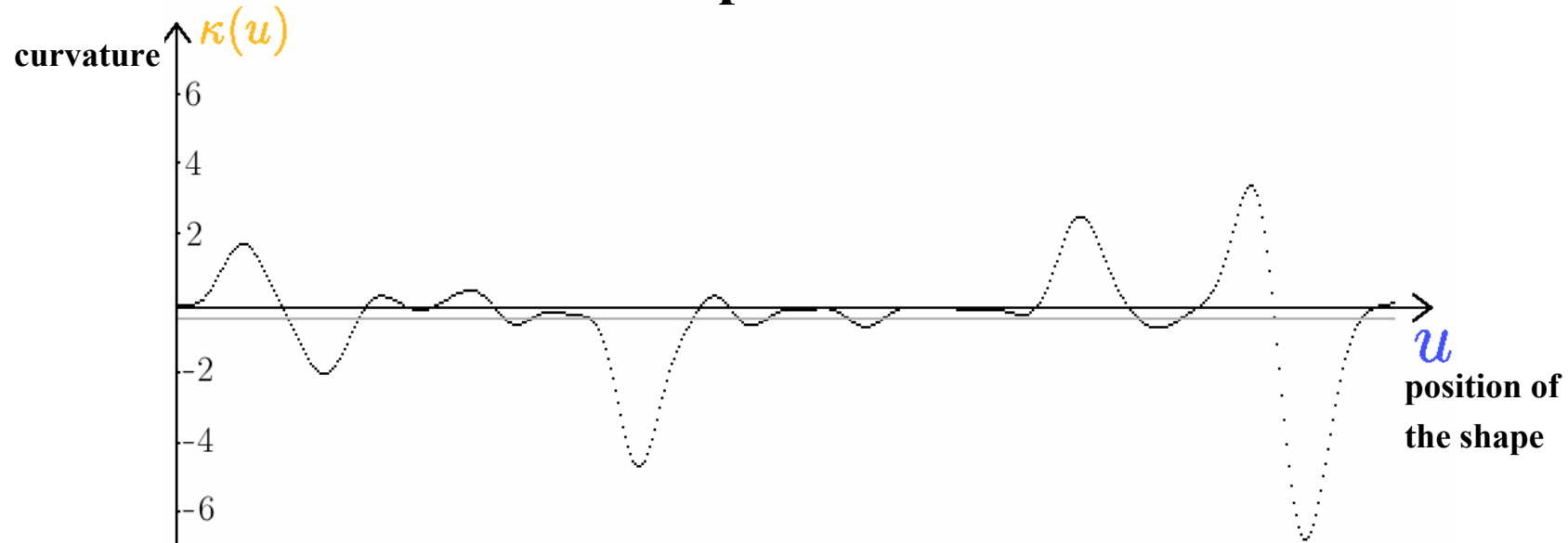
$$\dot{x}(t) = \frac{x(t+1) - x(t-1)}{2 \cdot hx}$$

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

- 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



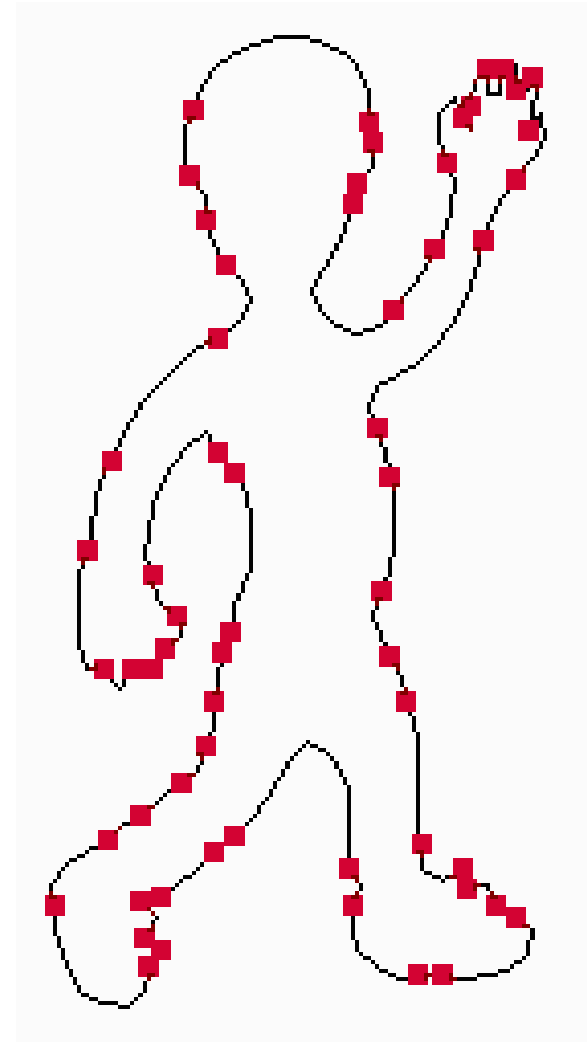
**Problem:** It is very difficult to match the curvature functions of two shapes.

→ Identify significant curvature features.

We use the curvature scale space technique.

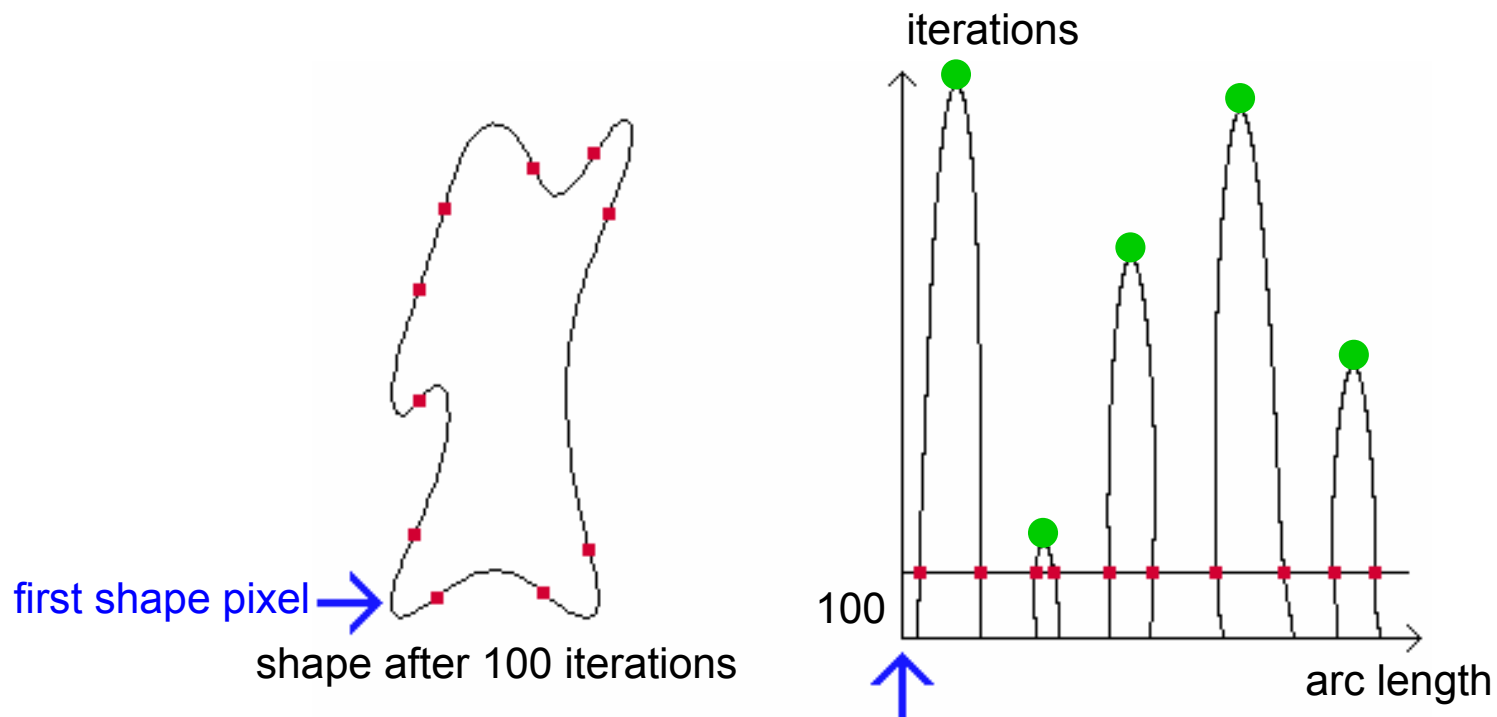
# Curvature scale space (I)

- 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.

# Curvature scale space (III)

## Characteristics of CSS images

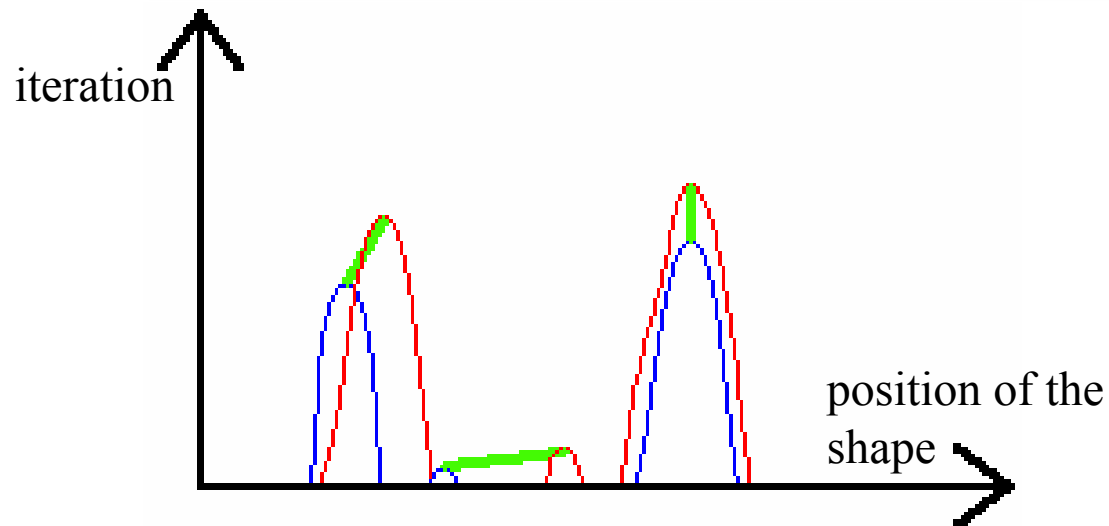
- The peaks in the curvature scale space image describe **concave segments** of a shape.
- The peaks are used as features to describe the shape.
- Each peak characterizes
  - a **position**  
relative position compared to other peaks,
  - a **value**  
strength of the concave segment.



# 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

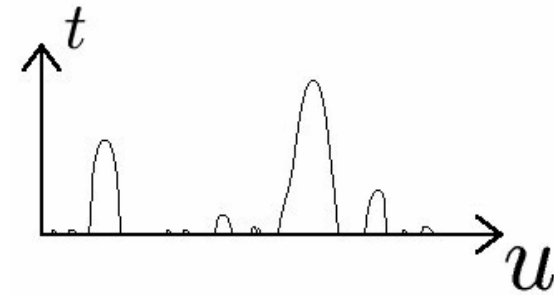
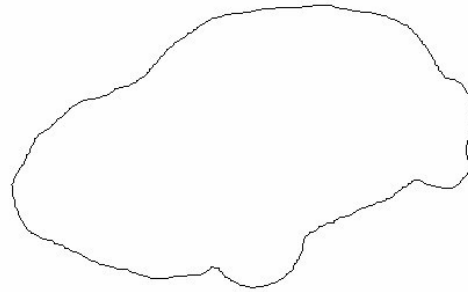
$$D = \sqrt{d_u \cdot d_u + d_t \cdot d_t}$$



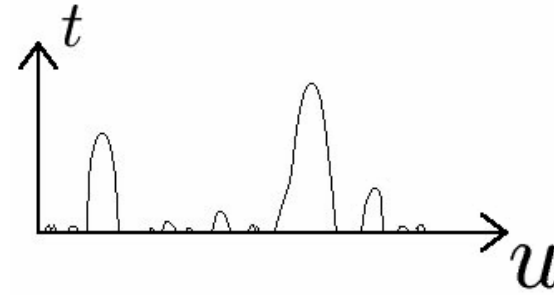
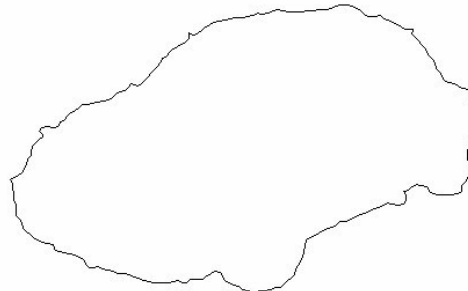
3. Summarize the distances

# Features of CSS images (I)

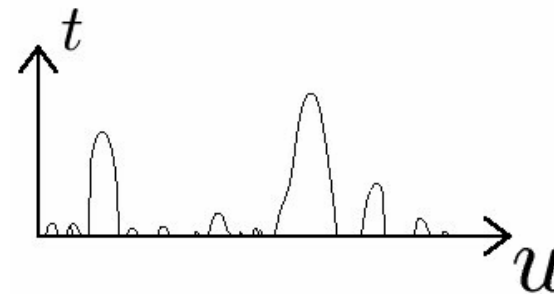
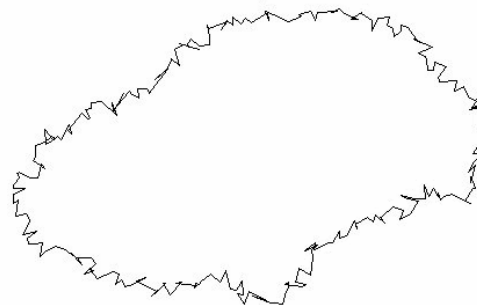
- Original shape of a car



- Shape with noise



- Shape with severe noise



→ CSS images are very similar

# Features of CSS images (II)

## Pros

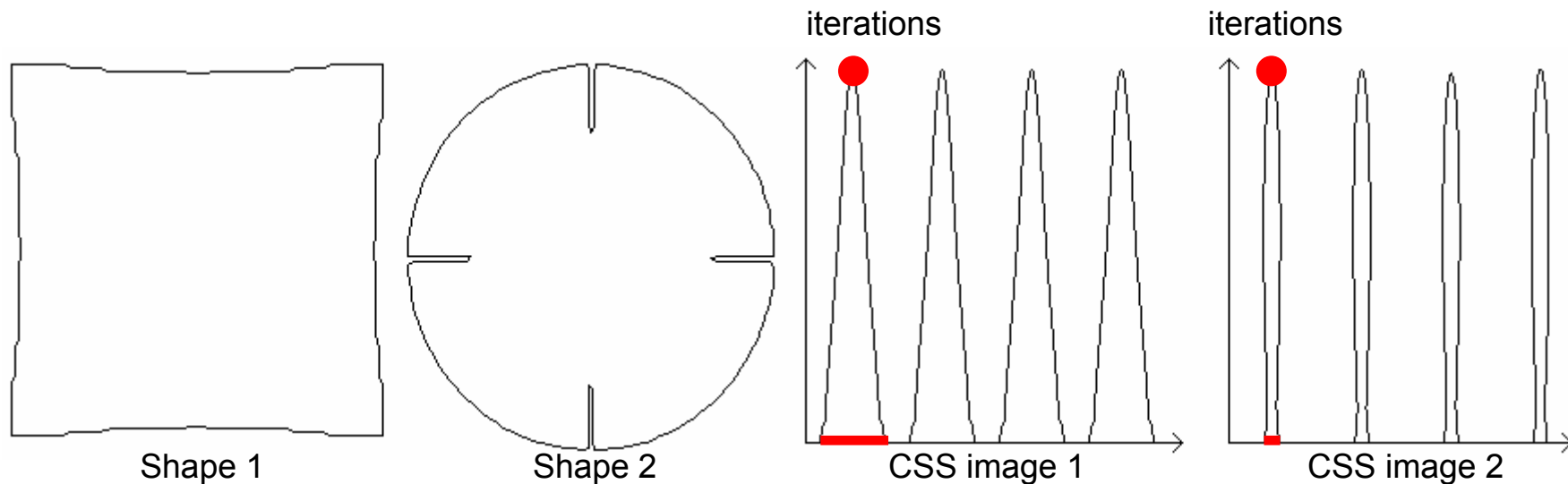
- Only a few values are required to describe complex objects.
- The approach is invariant to rotation or scaling.
- Low computation time.

## Cons

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

# Ambiguities of CSS Images (I)

## Shallow vs. deep concavities

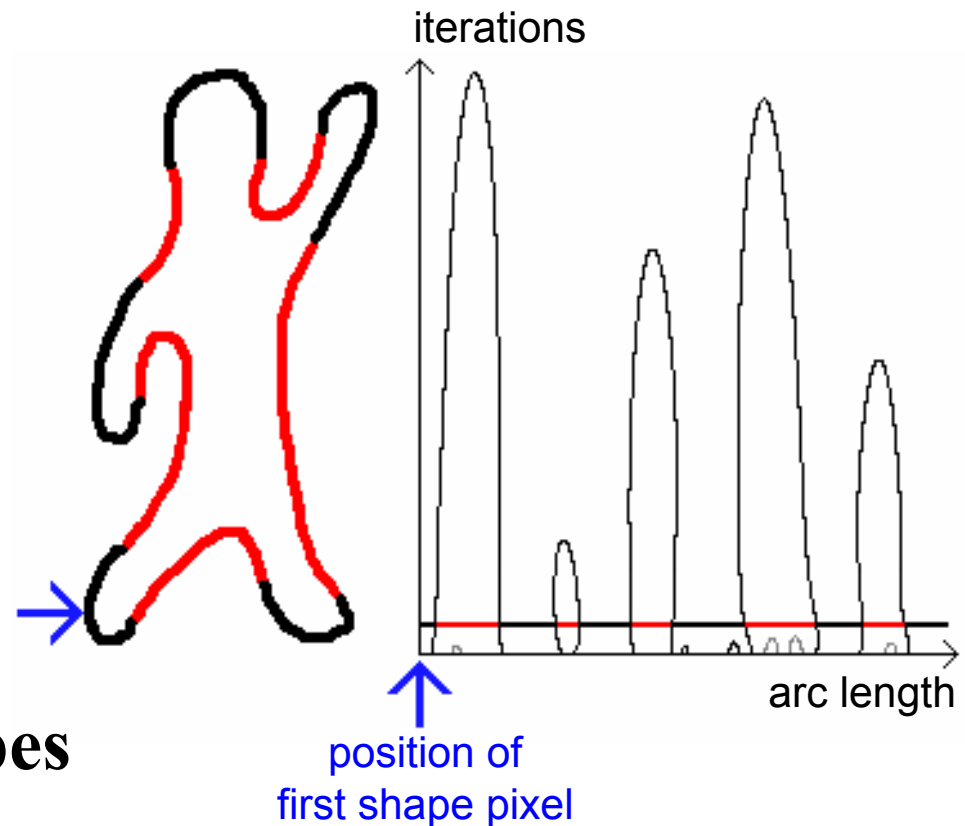


→ Solution: Use **position**, **height** and **width** of each peak as feature.

# Ambiguities of CSS Images (II)

## Convex regions

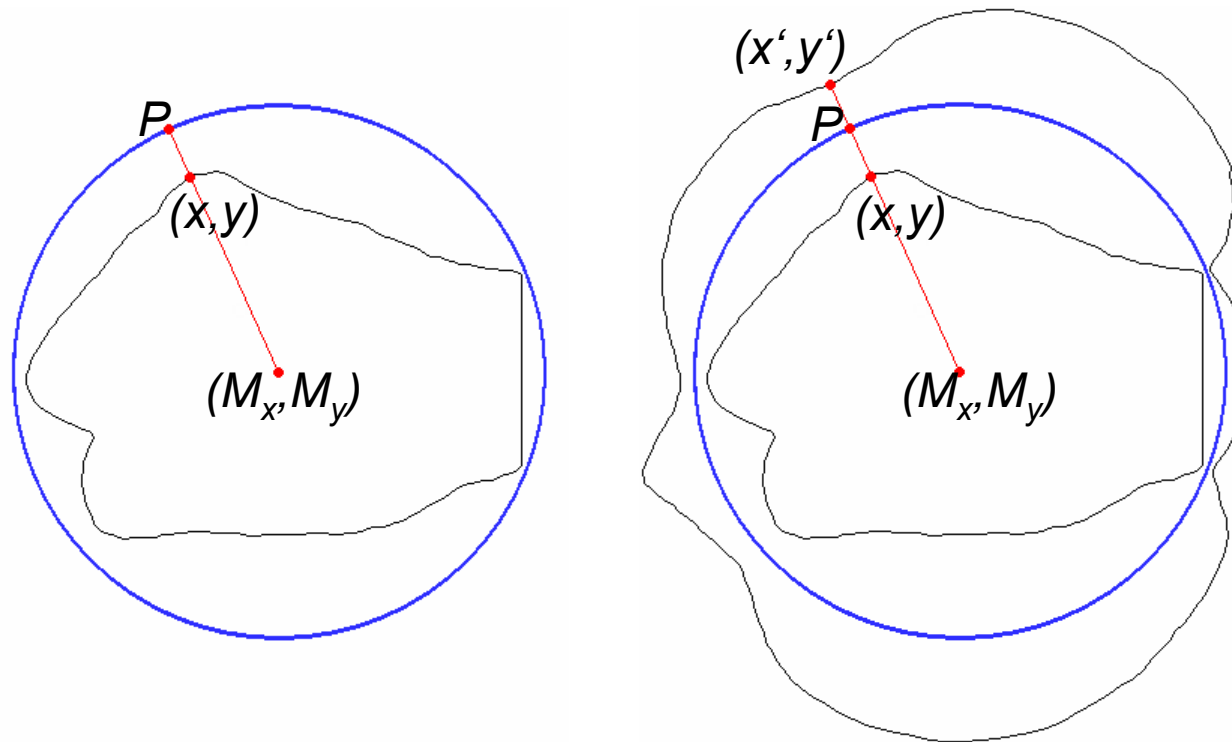
- Poor representation of convex regions of a shape.
- Convex objects cannot be represented at all.



→ **Solution: Mapped shapes**

# Mapped Shapes (I)

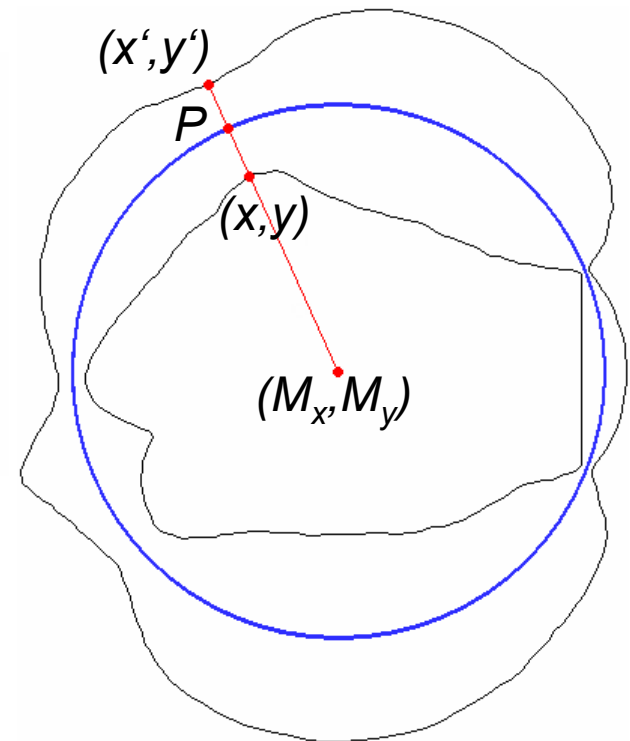
- **Idea:** Reflect each shape pixel and create a new shape.
- Strong convex segments of the original shape become concave segments of the mapped shape.



# Mapped Shapes (II)

## Calculation of mapped shapes

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



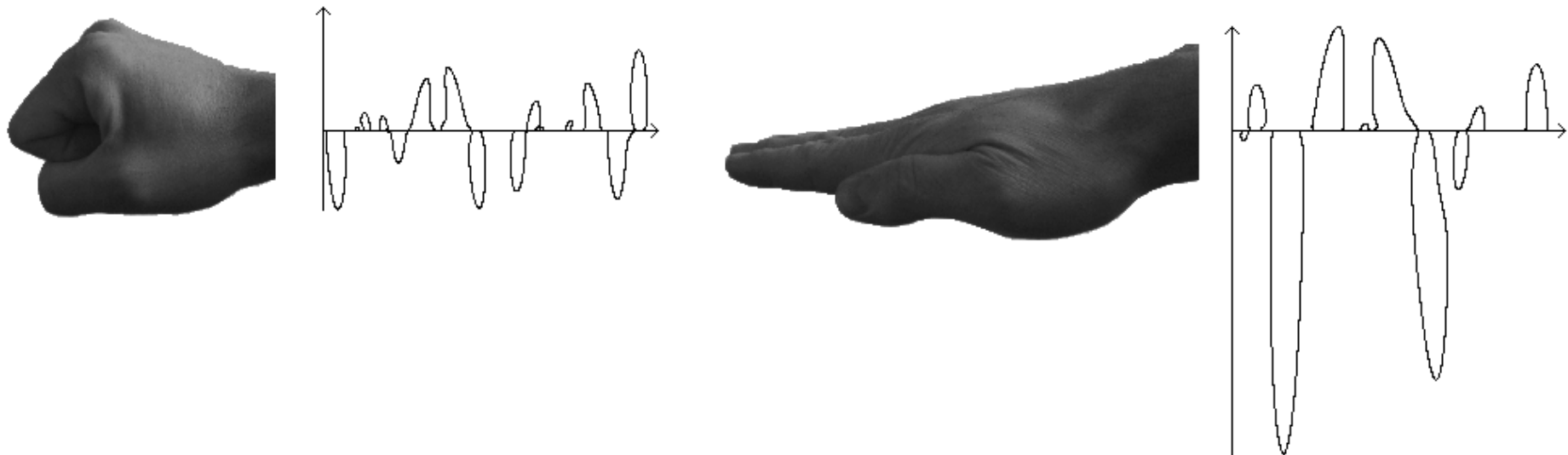
$R$ : radius of the circle

$D$ : 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)

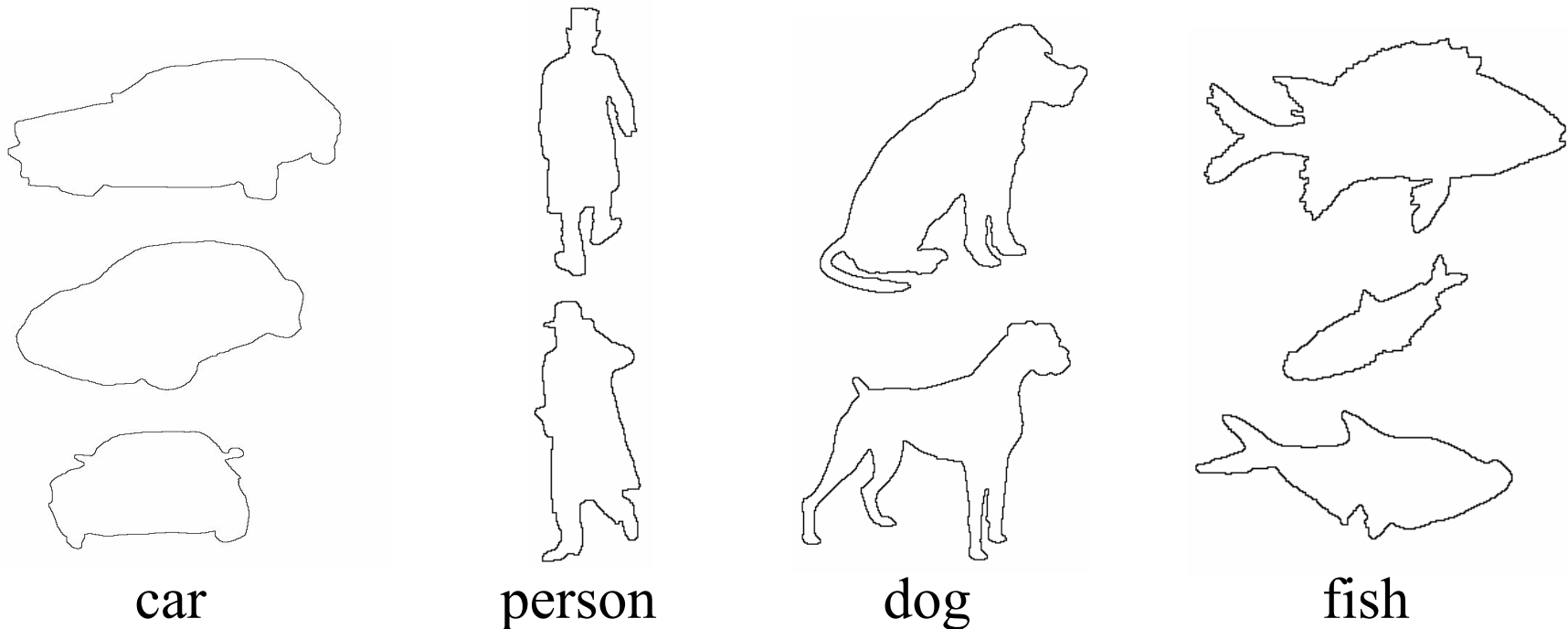
## Approach

- 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)

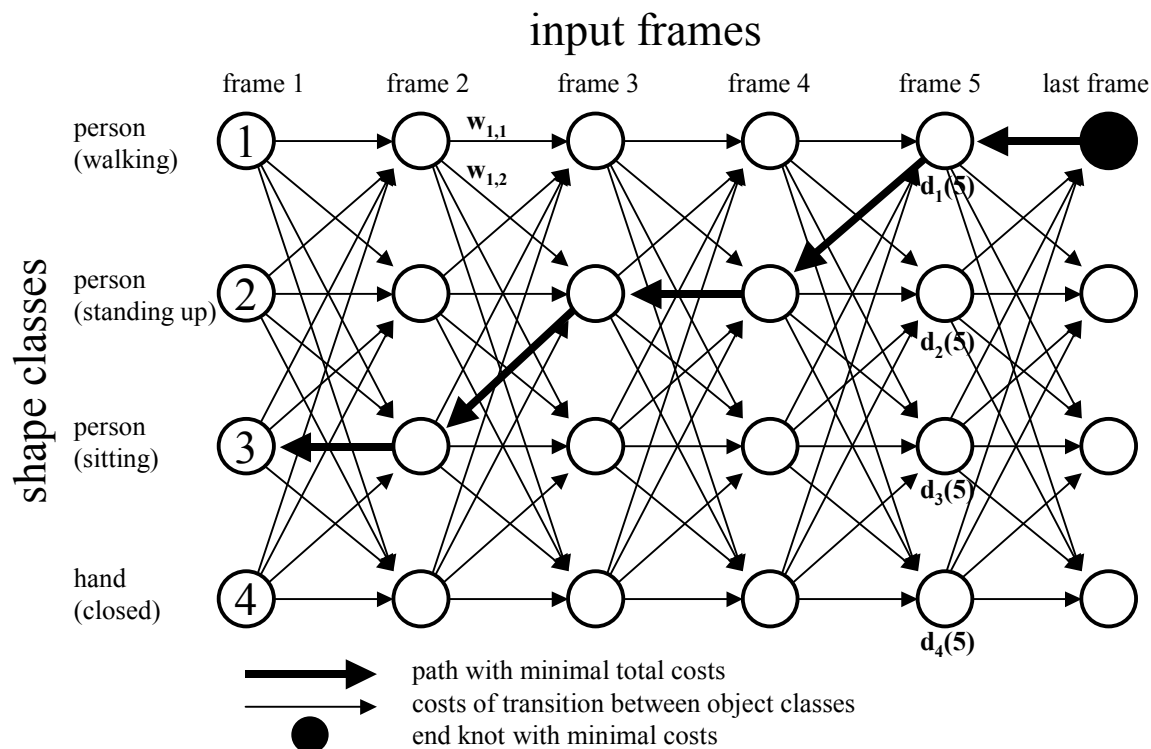
## Objects in the database

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

# 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)

# Object recognition in videos (V)



**standing**

**walking**

**turning around**

**sitting down**

**sitting**

**open hand**

**closed hand**

**fist**

**thumb**

# Questions ?