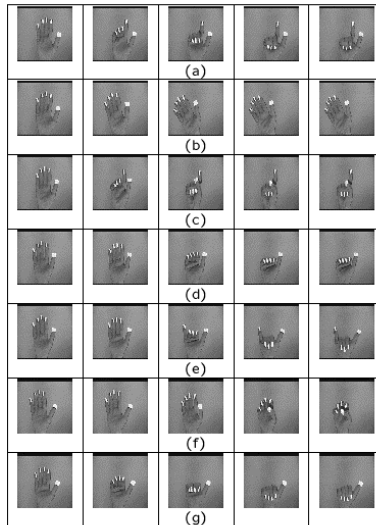


# Hand Gesture Recognition, Aerobic exercises, Events

Lecture-15

## Hand Gesture Recognition

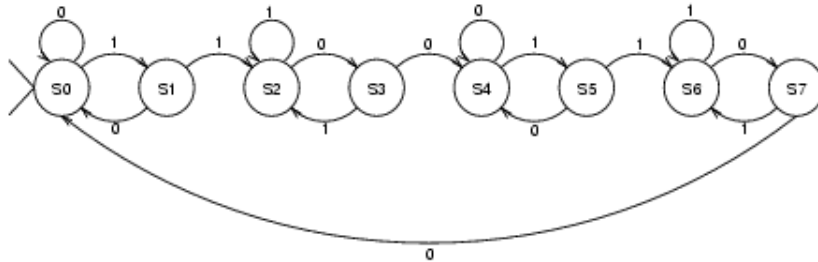
## Seven Gestures



## Gesture Phases

- Hand fixed in the **start position**.
- Fingers or hand move smoothly to **gesture position**.
- Hand fixed in **gesture position**.
- Fingers or hand return smoothly to **start position**.

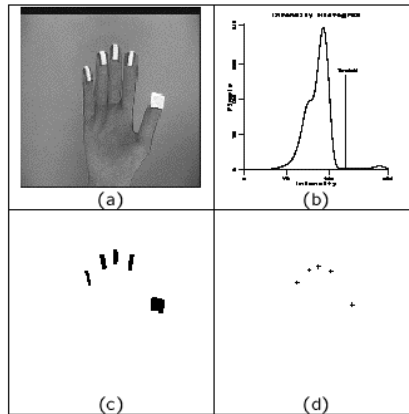
## Finite State Machine



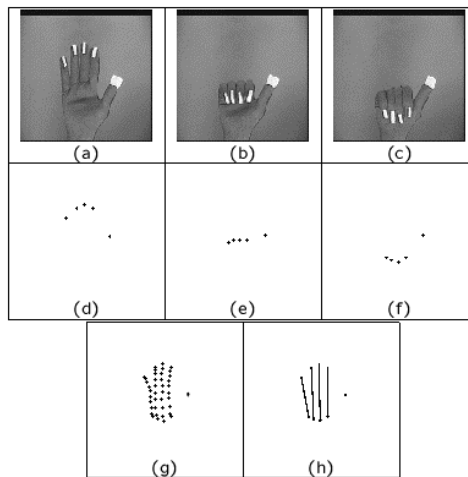
## Main Steps

- Detect fingertips.
- Create fingertip trajectories using motion correspondence of fingertip points.
- Fit vectors and assign motion code to unknown gesture.
- Match

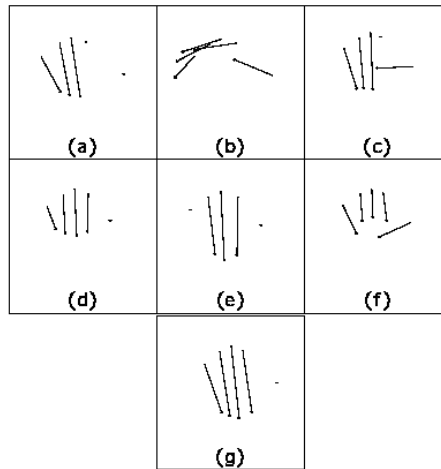
# Detecting Fingertips



# Vector Extraction



# Vector Representation of Gestures



# Results

## Results

Run	Frames	L	R	U	D	T	G	S
1	200	✓	✓	✓	✓	✓	✓	✓
2	250	✓	✓	✓	✓	✓	✓	✓
3	250	✓	✓	✓	X	✓	✓	✓
4	250	✓	✓	✓	✓	✓	✓	✓
5	300	✓	✓	✓	✓	✓	✓	✓
6	300	✓	✓	✓	✓	✓	✓	✓
7	300	✓	✓	✓	✓	✓	✓	✓
8	300	✓	✓	✓	✓	✓	✓	✓
9	300	✓	✓	✓	✓	*	*	*
10	300	✓	✓	✓	✓	✓	✓	✓

L = Left, R = Right, U = Up, D = Down, T = Rotate, G = Grab, S = Stop, ✓ - Recognized, X - Not Recognized, \* - Error in Sequence.

# Action Recognition Using Temporal Templates

Jim Davis and Aaron Bobick

## Main Points

- Compute a sequence of difference pictures from a sequence of images.
- Compute Motion Energy Images (MEI) and Motion History Images (MHI) from difference pictures.
- Compute Hu moments of MEI and MHI.
- Perform recognition using Hu moments.

# MEI and MHI

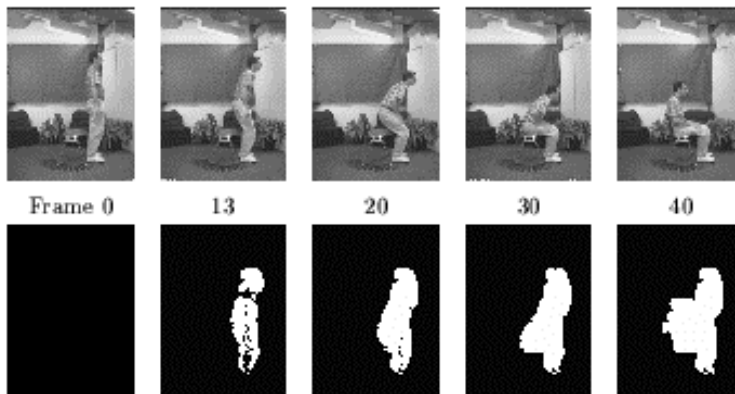
## Motion-Energy Images (MEI)

$$E_{\tau}(x, y, t) = \bigcup_{i=0}^{\tau-1} D(x, y, t-i)$$

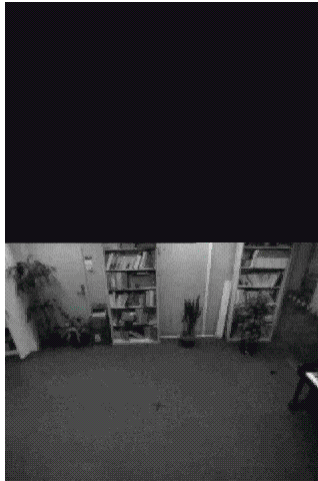
## Motion History Images (MHI) Change Detected Images

$$H_{\tau}(x, y, t) = \begin{cases} \tau & \text{if } D(x, y, t) = 1 \\ \max(0, H_{\tau}(x, y, t-1) - 1) & \text{otherwise} \end{cases}$$

## MEIs



## Color MHI Demo



## Summary

- Use seven Hu moments of MHI and MEI to recognize different exercises.
- Use seven views (-90 degrees to +90 degrees in increments of 30 degrees).
- For each exercise several samples are recorded using all seven views, and the mean and covariance matrices for the seven moments are computed as a model.
- During recognition, for an unknown exercise all seven moments are computed, and compared with all 18 exercises using Mahalanobis distance.
- The exercise with minimum distance is computed as the match.
- They present recognition results with one and two view sequences, as compared to seven view sequences used for model generation.



# Moments

Binary image

## General Moments

$$m_{pq} = \int \int x^p y^q \rho(x, y) dx dy$$

## Central Moments (Translation Invariant)

$$\mu_{pq} = \int \int (x - \bar{x})^p (y - \bar{y})^q \rho(x, y) d(x - \bar{x}) d(y - \bar{y})$$

$$\bar{x} = \frac{m_{10}}{m_{00}}, \bar{y} = \frac{m_{01}}{m_{00}} \quad \text{centroid}$$

# Central Moments

$$\mu_{00} = m_{00} \equiv \mu$$

$$\mu_{01} = 0$$

$$\mu_{10} = 0$$

$$\mu_{20} = m_{20} - \mu \bar{x}^2$$

$$\mu_{11} = m_{11} - \mu \bar{x} \bar{y}$$

$$\mu_{02} = m_{02} - \mu \bar{y}^2$$

$$\mu_{30} = m_{30} - 3m_{20} \bar{x} + 2\mu \bar{x}^3$$

$$\mu_{21} = m_{21} - m_{20} \bar{y} - 2m_{11} \bar{x} + 2\mu \bar{x}^2 \bar{y}$$

$$\mu_{12} = m_{12} - m_{02} \bar{x} - 2m_{11} \bar{y} + 2\mu \bar{x} \bar{y}^2$$

$$\mu_{03} = m_{03} - 3m_{02} \bar{y} + 2\mu \bar{y}^3$$

# Moments

**Hu Moments: translation, scaling and rotation invariant**

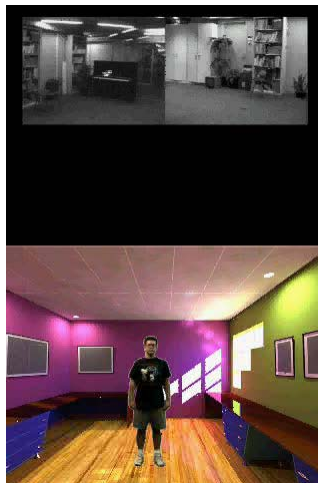
$$\nu_1 = \mu_{20} + \mu_{02}$$

$$\nu_2 = (\mu_{20} - \mu_{02})^2 + \mu_{11}^2$$

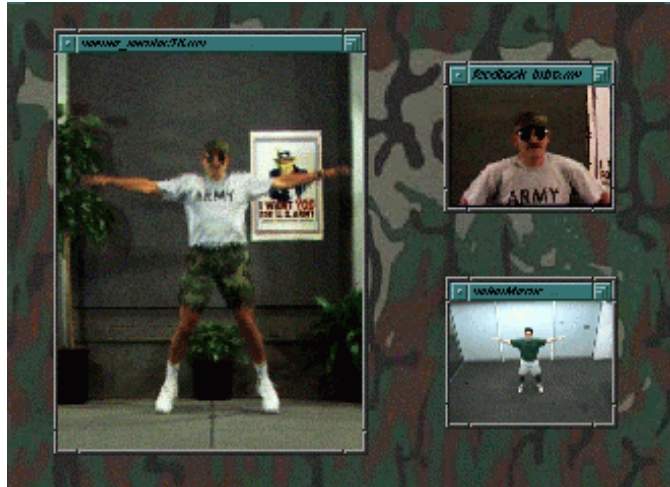
$$\nu_3 = (\mu_{30} - 3\mu_{12})^2 + (3\mu_{12} - \mu_{03})^2$$

$$\nu_4 = (\mu_{30} + \mu_{12})^2 + (\mu_{21} + \mu_{03})^2$$

⋮



## PAT (Personal Aerobic Trainer)



## PAT (Personal Aerobic Trainer)



[http://vismod.www.media.mit.edu/vismod/demos/actions/mhi\\_generation.mov](http://vismod.www.media.mit.edu/vismod/demos/actions/mhi_generation.mov)

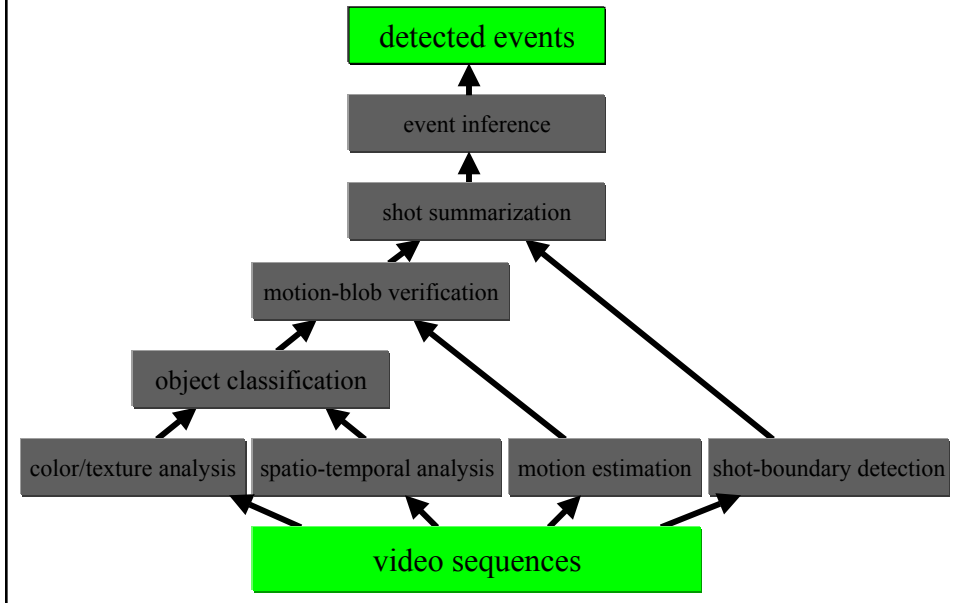
## PAT (Personal Aerobic Trainer)



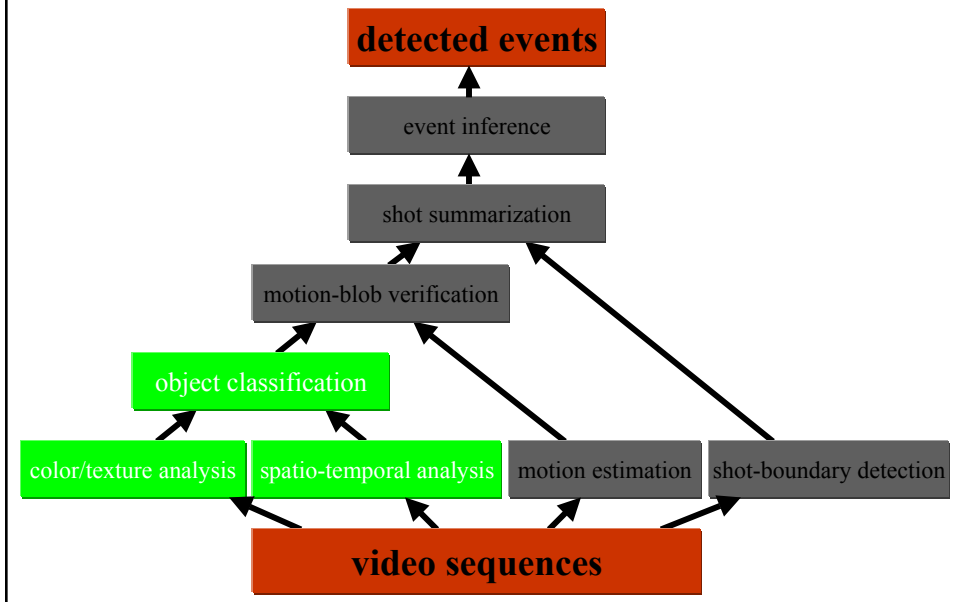
## A Framework for the Design of Visual Event Detectors

**Niels Haering**

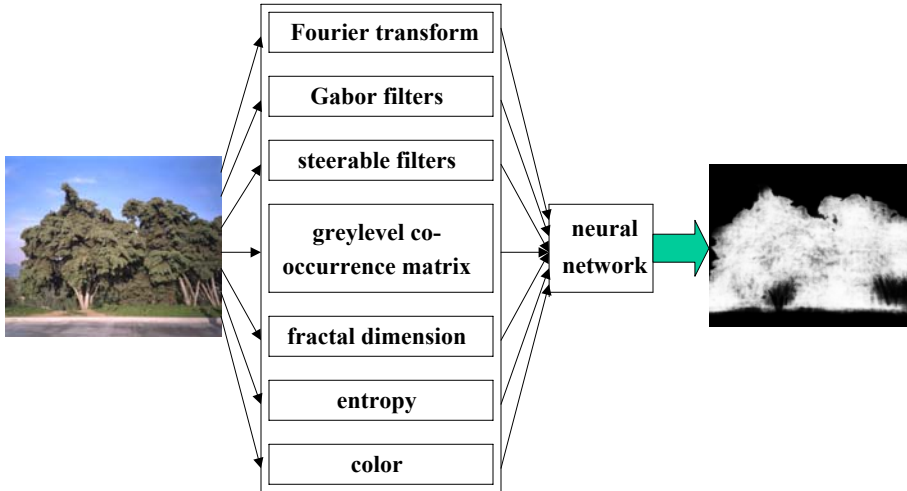
# Our Framework



# Object Classification



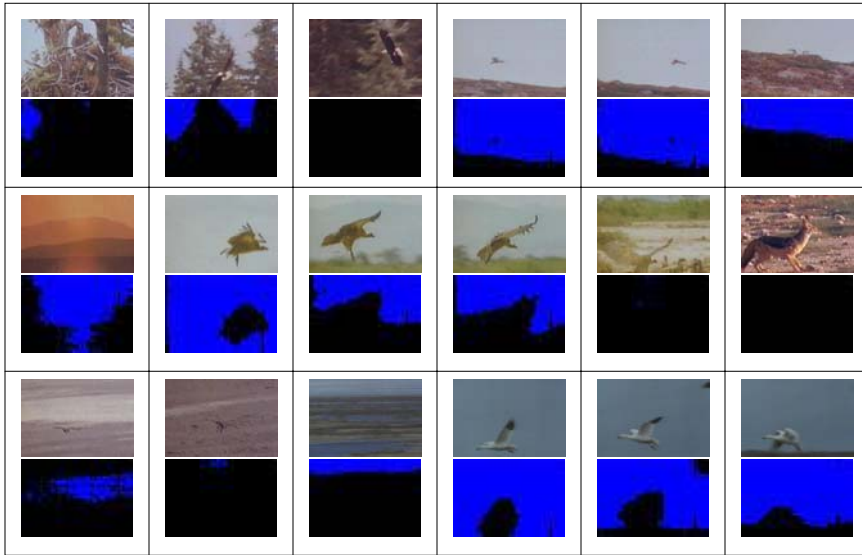
# Object Classification



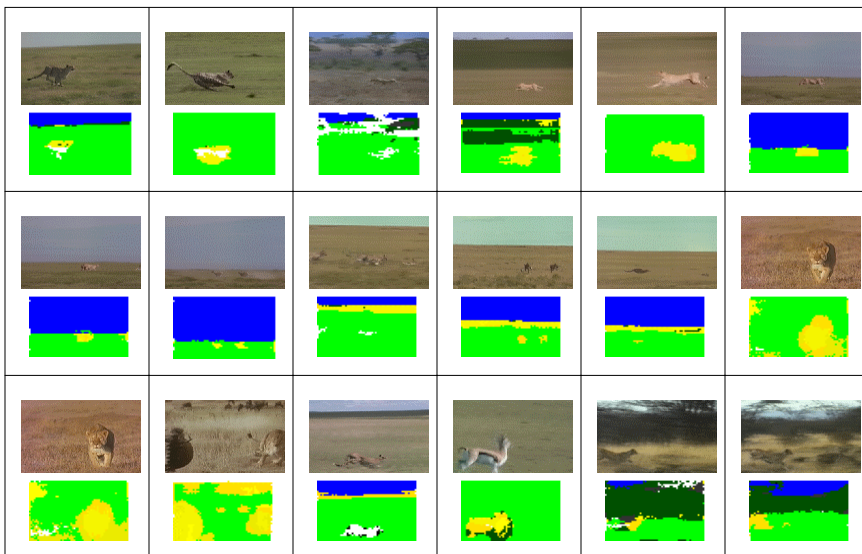
# Deciduous Trees



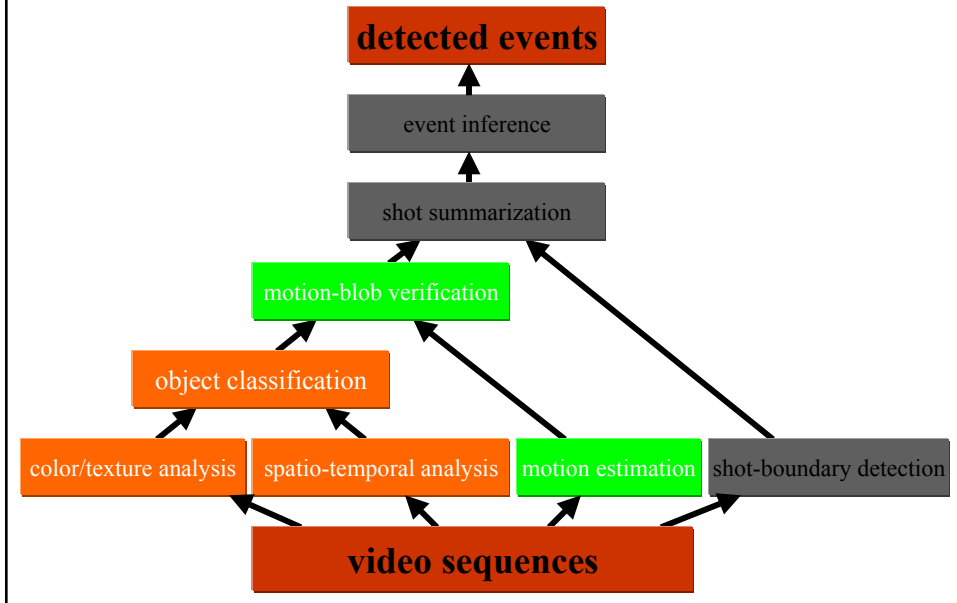
# Sky



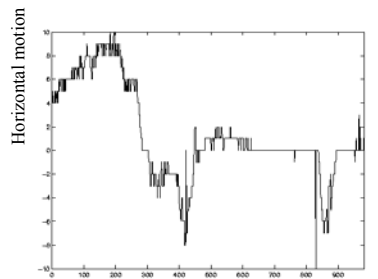
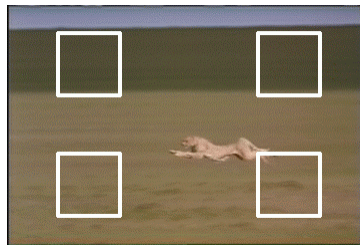
# Animals, Sky, Grass, Trees, Rock



# Motion-blob Verification



# Motion Estimation

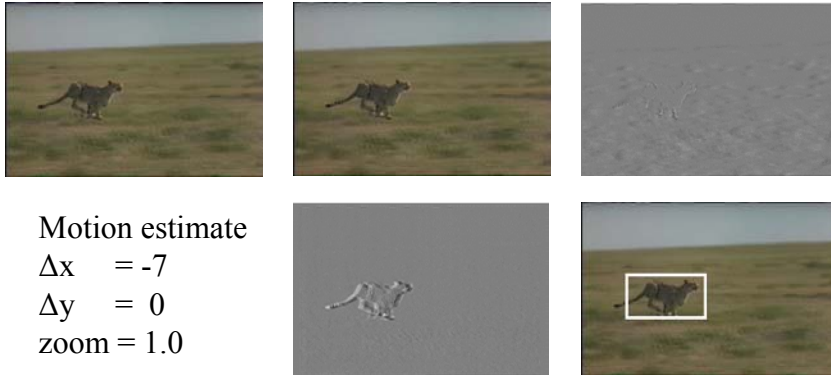


Frame number

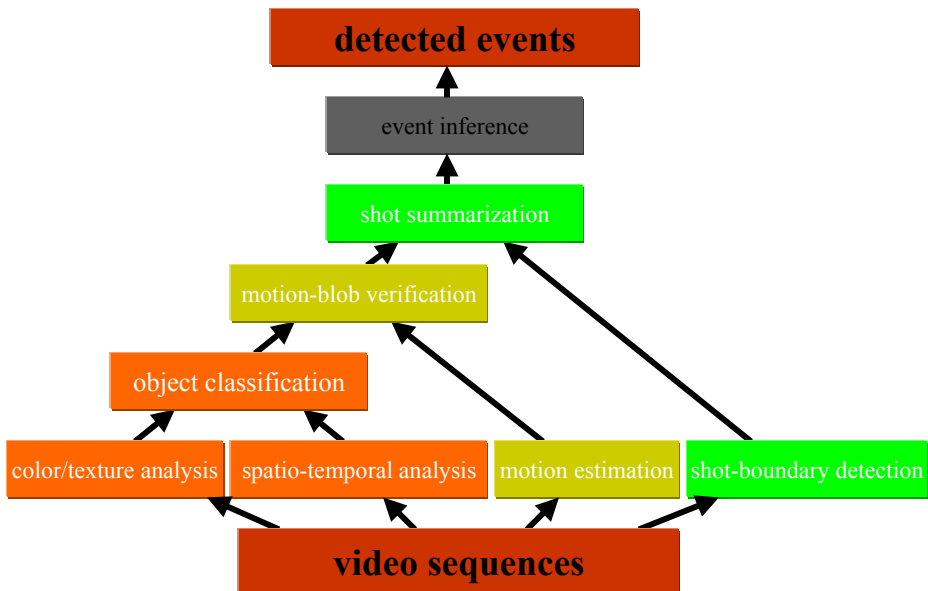
- three parameter system: x-, y-translation, and zoom,
- 4 motion estimates based on pyramid,
- 4 motion estimates based on previous best match,
- “texture” measure prevents ambiguous matches



# Motion-blob detection



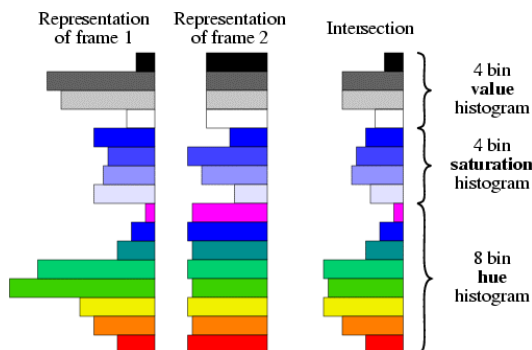
# Shot Summarization



# Shot Detection

## Characteristics of shot boundaries:

- Change of camera/viewpoint
- Change of color characteristics

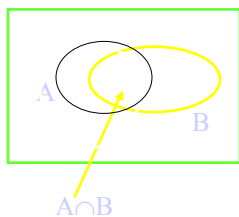
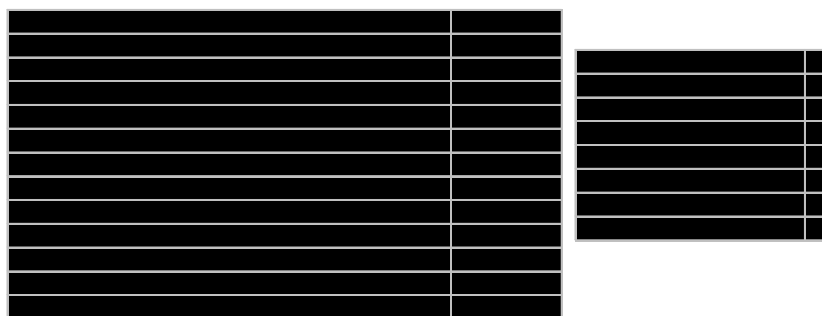


4 Bins for Value  
4 Bins for Saturation  
8 bins for hue

$$= 0.79$$

$$\cap = \frac{\sum_{n=0}^{15} \min(f_1(n), f_2(n))}{\min(\sum_{n=0}^{15} f_1(n), \sum_{n=0}^{15} f_2(n))}$$

# Shot Summaries

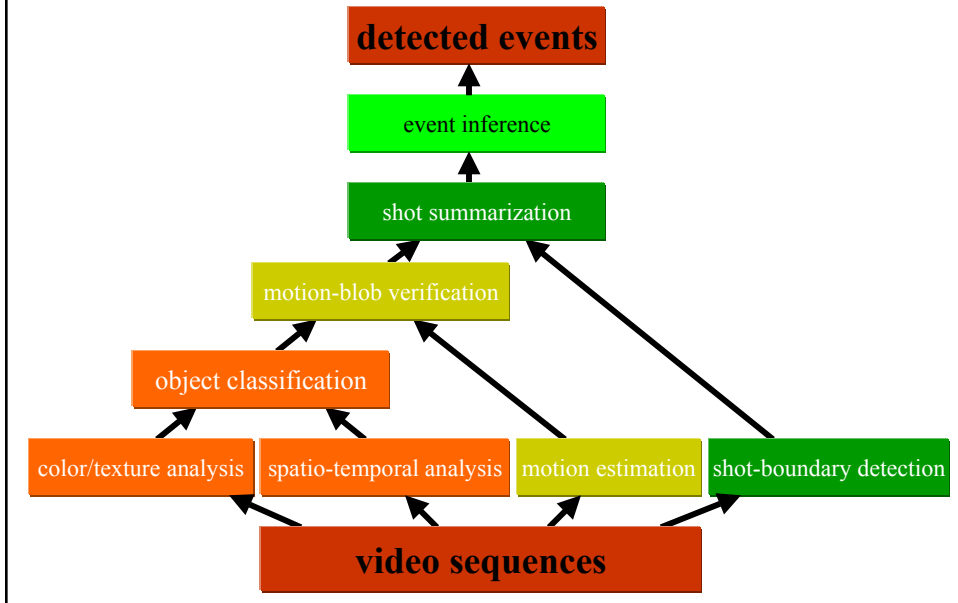


A = Ground Truth  
B = Result of Algorithm  
 $A \cap B$  = Correct detection

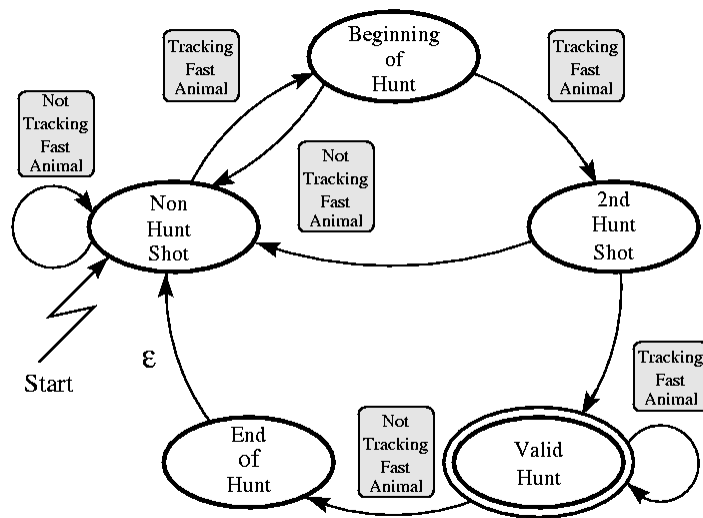
$$recall = \frac{A \cap B}{A}$$

$$precision = \frac{A \cap B}{B}$$

# Event Inference



# Hunt events



# Hunts

Hunt



Non-hunt



# Hunts

Non-hunt



Hunt



Non-hunt

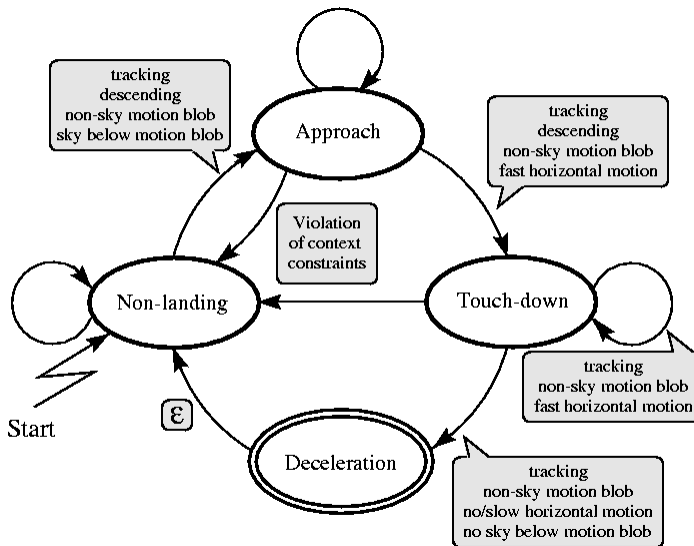
# Event Detection



# Landing Events



# Landing Events



# Landing Events

Non-landing



Approach



Touch-down



Deceleration



Non-landing



# Landing Events

Non-landing



Approach



Touch-down



Deceleration



Non-landing

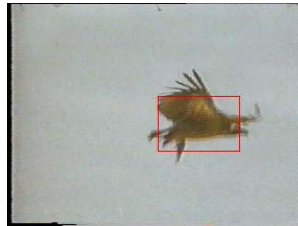


# Landing Events

Non-landing



Approach



Touch-down



Deceleration



Non-landing



## Conclusions

- Many natural objects are easily recognized by their color and texture signatures (shape is often not needed)
- Many events are easily detected and recognized by the classes of the comprising objects and their approximate motions
- The proposed visual event detection is robust to changes in scale, color, shape, occlusion, lighting conditions, view points and distances, and image compression