

#### Feel the beat:

using cross-modal rhythm to integrate perception of objects, others, and self

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#### Modal and amodal features

mode-specific

color

pitch

temperature

4.14

amodal

timing

location

intensity

shape

texture

#### Modal and amodal features

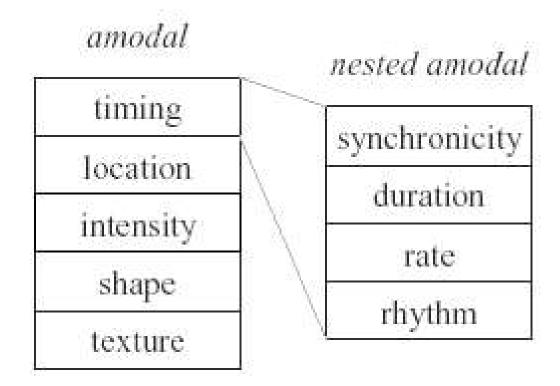
mode-specific

color

pitch

temperature

222



(following Lewkowicz)

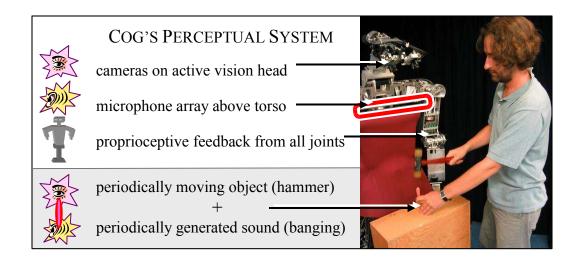
### Motivation

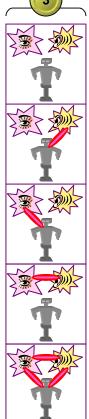
- Tools and toys are often used in a manner that is composed of some repeated motion consider hammers, saws, brushes, files, ...
- Rhythmic information across the visual and acoustic sensory modalities have complementary properties
- Features extracted from visual and acoustic processing are what is needed to build an object recognition system

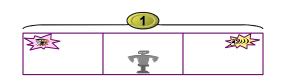
#### Talk Outline

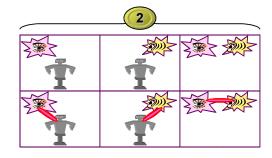
- Hardware
- Matching sound and vision
- Priming for attention
- Differentiation
- Integration
- The self and others

## Cog's Perceptual System









## Interacting with the robot



### Making sense of the senses...

### Bang, Bang!



#### Who is he?



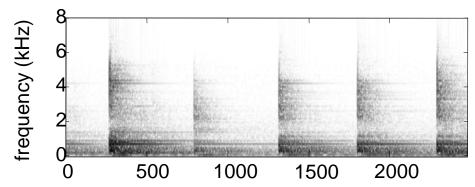
#### Talk Outline

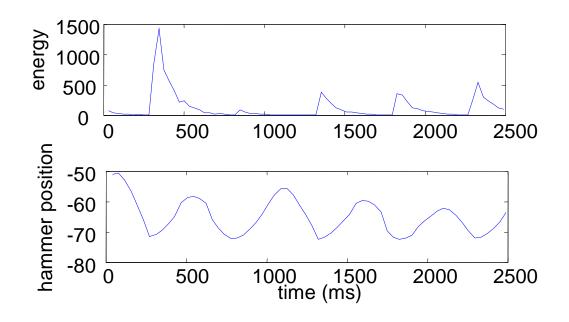
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## Matching sound and vision



The sound intensity peaks once per visual period of the hammer (CIRAS 2003)





## Matching algorithm

 Estimate signal period (histogram technique from CIRAS 2003)

 Cluster rising and falling intervals, guided by the scale of estimated period

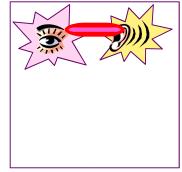
Merge sufficiently close clusters

Segment full periods in the signal

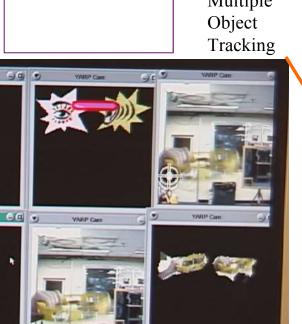


#### Playing a tambourine

Appearance and sound of tambourine are bound together



Multiple



Sound Segmentation (window divided in 4x4 images)

Object

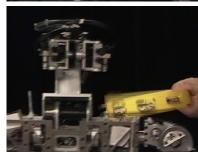
Segmentation

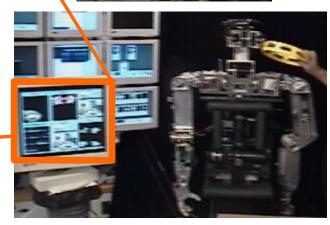
Cog's view

Object Recognition (window divided in 2x2 images)

robot sees and hears a tambourine shaking







tambourine segmentations



### Robustness

to random visual disturbances



#### to auditory disturbances



Person talks – sound not matched to object!

#### Talk Outline

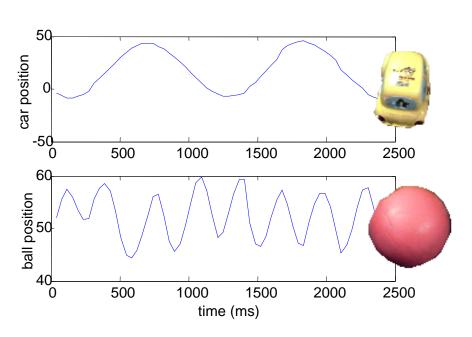
- Hardware
- Matching sound and vision
- Priming for attention
  - Priming visual foreground with sound
  - Priming acoustic foreground with vision
  - Matching multiple sources
- Differentiation
- Integration
- The self and others

# Priming visual foreground with sound

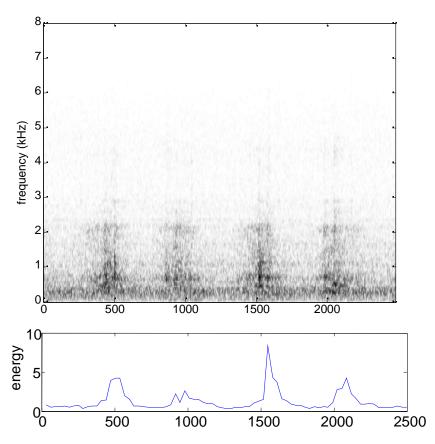


- One object (the car) making noise
- Another object (the ball) in view
  - Problem: which object goes with the sound?
  - Solution: Match periods of motion and sound

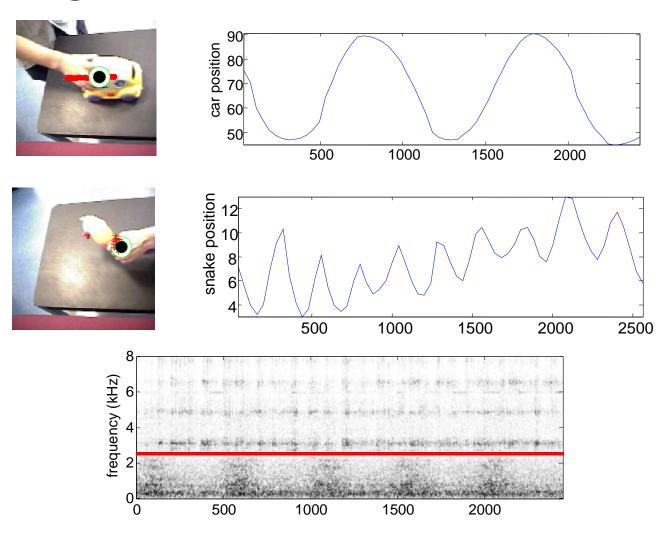
## Comparing periods



•The sound intensity peaks twice per visual period of the car



## Matching with acoustic distraction



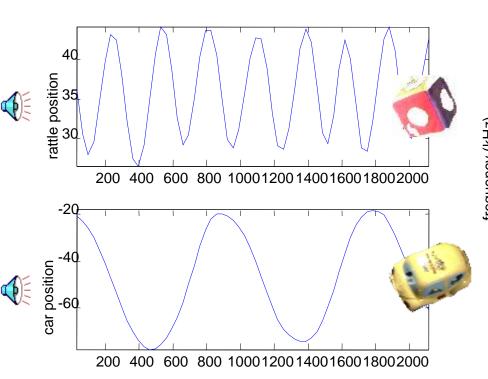
## Matching multiple sources

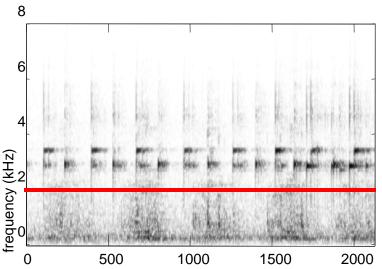




- Two objects making sounds with distinct spectrums
  - Problem: which object goes with which sound?
  - Solution: Match periods of motion and sound

## Binding periodicity features





•The sound intensity peaks twice per visual period of the car. For the cube rattle, the sound/visual signals have different ratios according to the frequency bands

# Cross-modal association - errors

Experiment	visual period found	sound period found	bind sound, vision	candidate binds	correct binds	incorrect binds
hammer	8	8	8	8	8	0
car and ball plane & mouse/remote	14 18	6 3	6 3	15 20	5 3	1 0
car (snake in backg'd) snake (car in backg'd)	5 8	1 6	1 6	20 8	1 6	0
car & cube $\begin{cases} car \\ cube \end{cases}$	9 10	3 8	3 8	11 11	3 8	0
car & snake $\begin{cases} car \\ snake \end{cases}$	8	0 5	0 5	8 8	0 5	0

#### Talk Outline

- Hardware
- Matching sound and vision
- Priming for attention
- Differentiation
  - Visual Recognition
  - Sound Recognition
- Integration
- The self and others

# Visual Object Segmentation/Recognition

Object Segmentation



Object Recognition

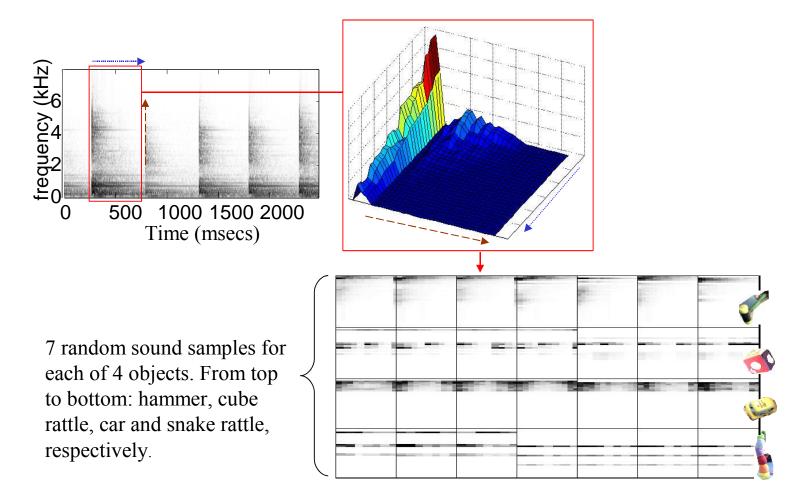
see Arsenio, MIT PhD thesis, 2004 for visual objectsegmentation/recognition

## Sound Segmentation

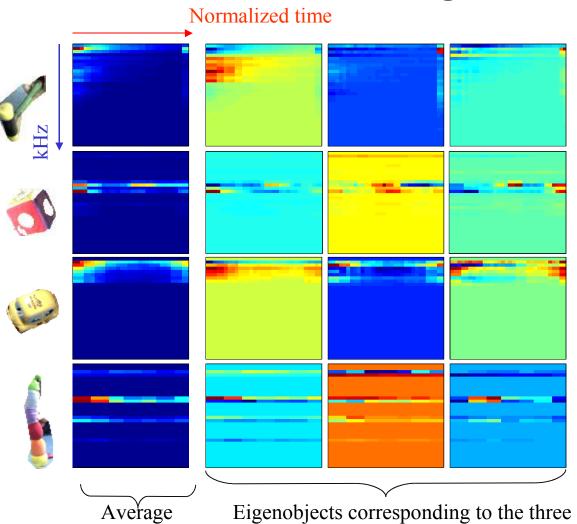
Goal: Extract acoustic signatures from repetitive data

Problem: STFTs applied for spectral analysis, but not ideal for irregular signals

Solution: Build histograms of hypothesized periods



## Sound Recognition



sound images

Recognition rate: 82%

Eigenobjects corresponding to the three highest eigenvalues

#### Talk Outline

- Hardware
- Matching sound and vision
- Priming for attention
- Differentiation
- Integration
  - Cross-modal segmentation/recognition
  - Cross-modal enhancement of detection
- The self and others

## Cross-modal object recognition

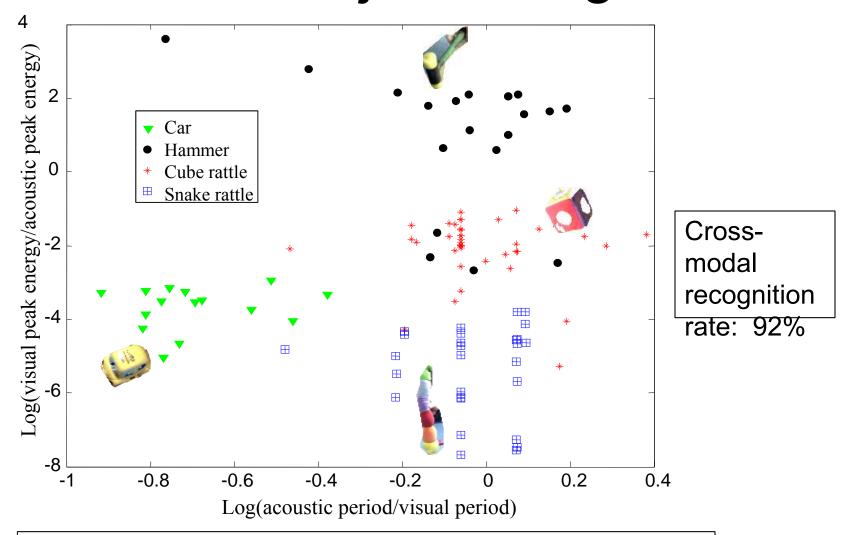






Causes sound when changing direction after striking object; quiet when changing direction to strike again Causes sound while moving rapidly with wheels spinning; quiet when changing direction Causes sound when changing direction, often quiet during remainder of trajectory (although bells vary)

## Cross-modal object recognition



#### Dynamic Programming

is applied to match previously segmented sensory signals: visual trajectories to the sound energy signal

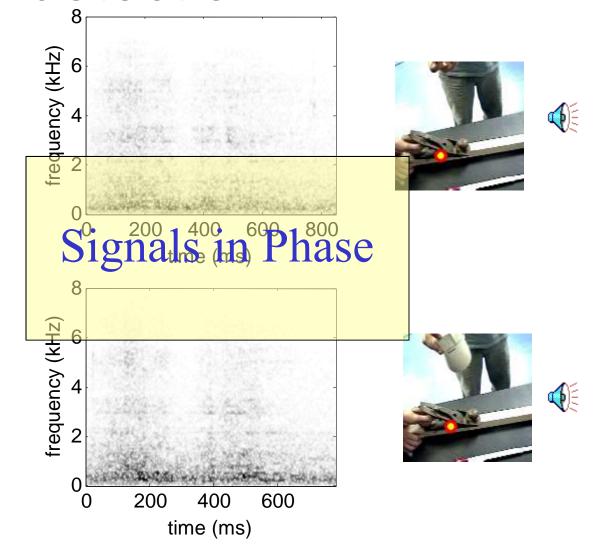
# Cross-modal recognition – confusion table

Confusion matrix	car	cube	$\operatorname{snake}$	hammer
car	30	0	0	0
cube	0	52	7	1
$\operatorname{snake}$	0	0	45	О
hammer	0	5	0	25

# Cross-modal enhancement of detection

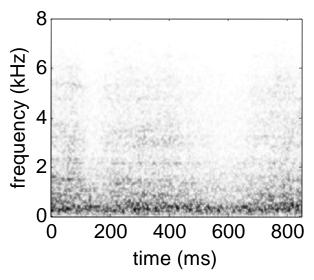






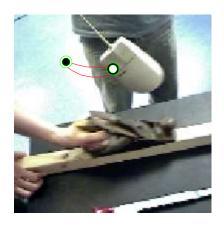
## Signals out of phase!

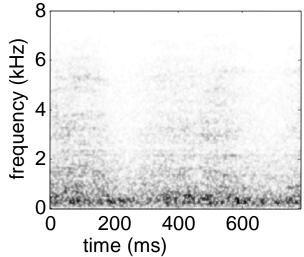
















#### Talk Outline

- Hardware
- Matching sound and vision
- Priming for attention
- Differentiation
- Integration
- The self and others
  - Learning about people
  - Learning about the self

# Cross-modal rhythm to integrate perception of

#### Control Experiment Others

the robot sees a person shaking head – no periodic sound





#### **Experiment 2**

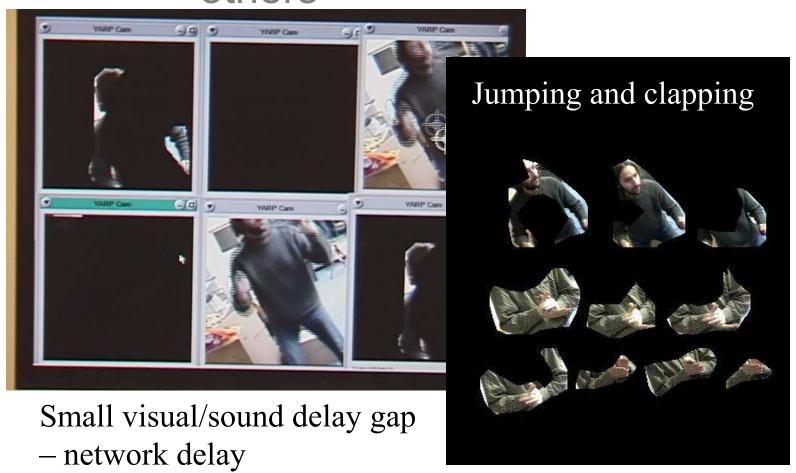
the robot sees a person shaking head and saying "no"





### Cross-modal rhythm

to integrate perception of others

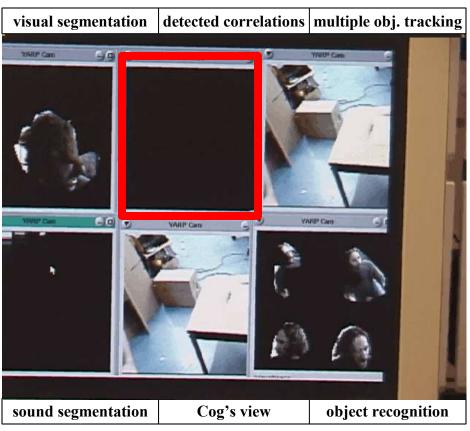


## Binding

#### Sound and Proprioceptive Data



Detecting ones'own rhythms



## Binding Vision, Sound and Proprioceptive Data

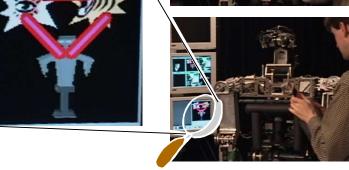
Visual image segmented, sound detected, and all bounded to the motion of the arm



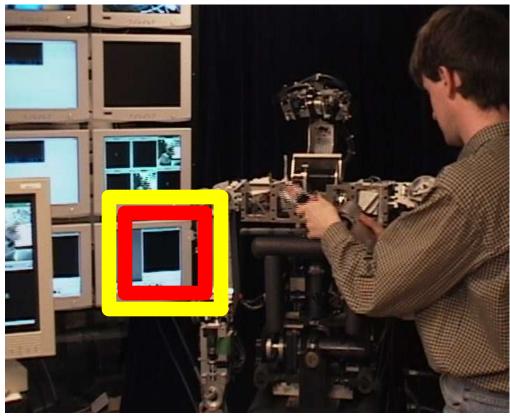
robot is looking





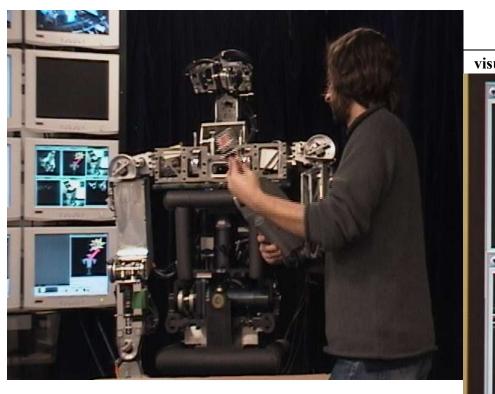


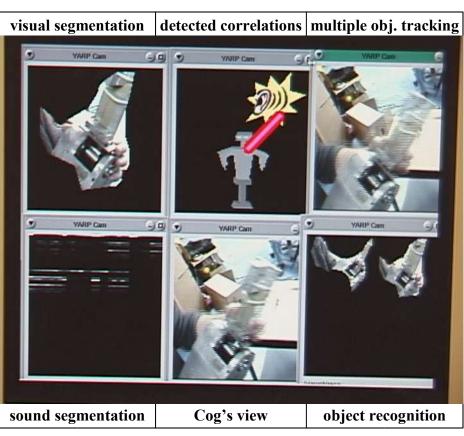




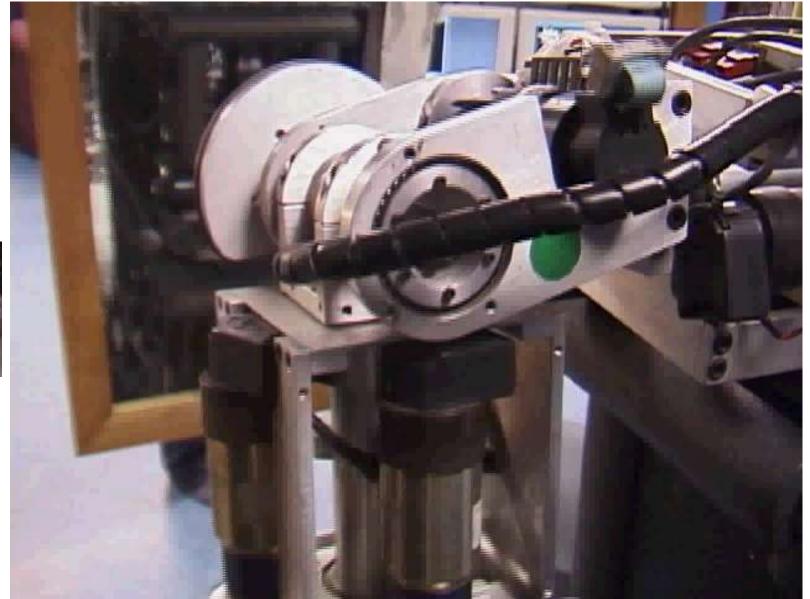
## Binding

#### Vision, Sound and Proprioceptive Data



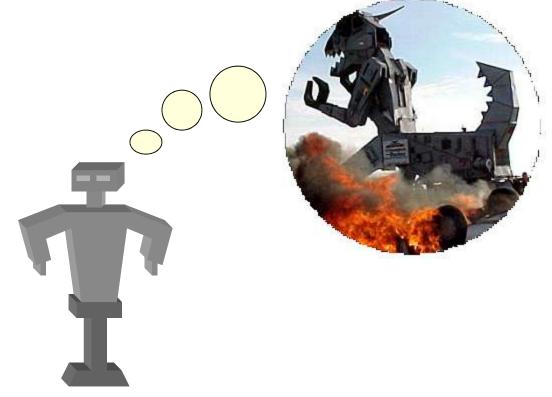


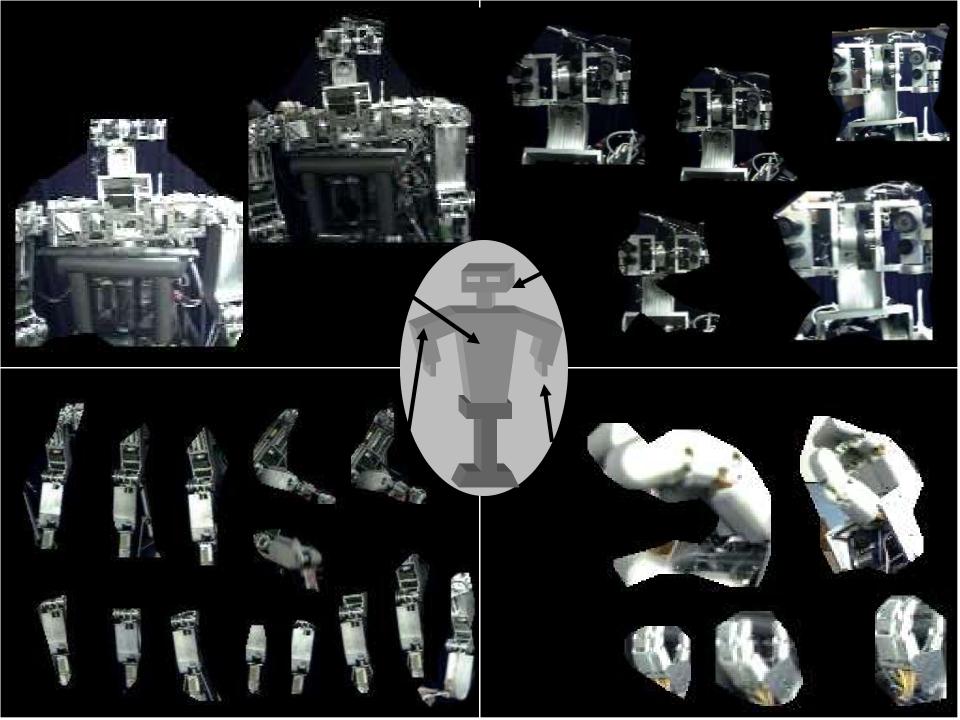
## Cog's mirror image



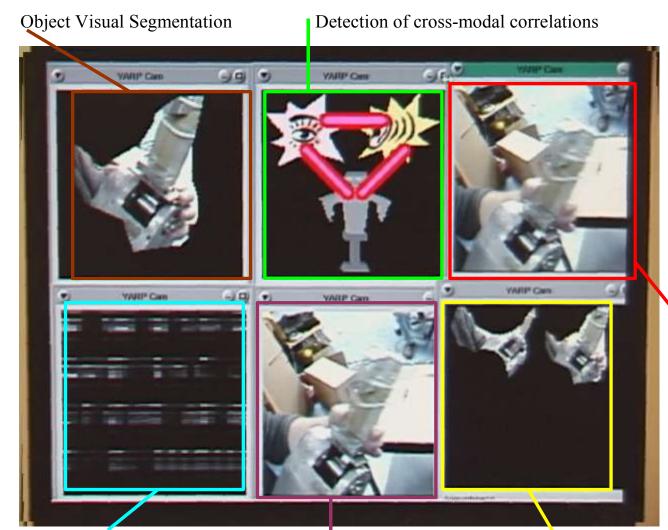


So, how does Cog perceive himself?





## The robot's experience of an event



Multiple Object Tracking

#### Sound Segmentation

(window separated on a set of 4x4 images - each image contains spectogram over 1 period of the signal)

Cog's view

#### Object Recognition

(window divided on a set of 2x2 images – downloaded from the class assigned to the object)

#### Conclusions

- Amodal features are key to detecting relationships across senses
- Useful for learning to recognize an object in different senses (e.g. by its appearance or its sound)
- There are features for object recognition that exist only in relationships across senses and do not exist in any one sense
- Useful both for perception of external objects and robot's own body, by incorporating proprioception as another sense