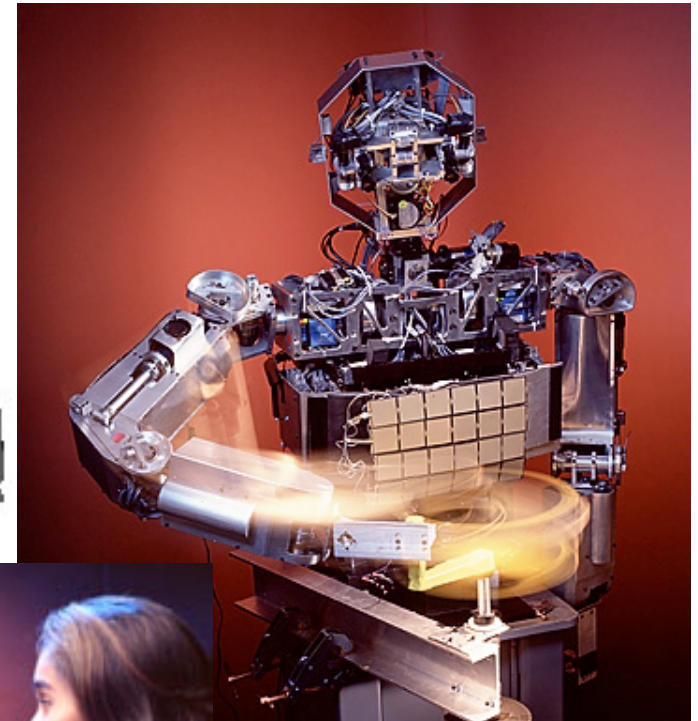
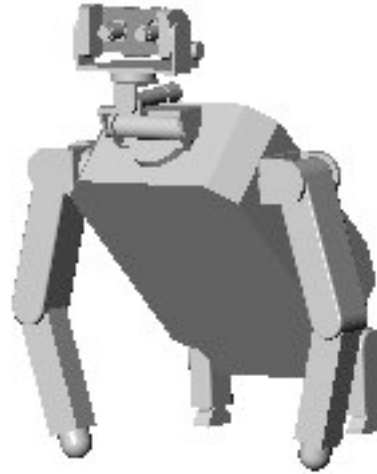




Natural Tasking of Robots Based on Human Cues -- MIT AI Lab

- **Challenge:** Make commanding robots as intuitive and natural as commanding soldiers
- We will build software systems (on 4 robot platforms) that
 - interact with people using intuitive social cues
 - are self-adapting through developmental learning
 - exploit their physical embodiment to simplify computation
 - use multi-modal sensory integration to maximize effectiveness





Research Issues: Learning by Imitating Humans

- **Knowing what to imitate**
 - what sensory signals are relevant to the current task?
- **Mapping between bodies**
 - how to convert sensory view of one agent to self-mapping?
- **Correcting failures and recognizing success**
 - how can the robot automatically develop success measures?
- **Chaining actions together**
 - how can the learned pieces be chained together into complete task sequences?
- **Generalizing to more complex tasks**
 - how to use invariants and common-sense knowledge to adapt the learned sequences to new situations
- **Intuitive interactions for the human**
 - how to use natural social signals to make it intuitive for humans and robots to communicate with each other



Our Approaches Lead to Specific Progress on Each Research Issue

Research Issues

Our Approaches		Knowing what to imitate	Mapping between bodies	Correcting failures and recognizing success	Chaining actions together	Generalizing to more complex tasks	Intuitive Interactions
	Social Interaction	Uses attentional cues to recognize task relevant events		Identifies and displays emotional states for reinforcement			Allows humans and robots to share similar social cues
	Development	Limits search space by incremental refinement of perception	Simplifies mapping by incremental refinement of motor skills		Provides natural decomposition of complex tasks	Exploits incremental learning to build new skills	Provides methods for building social skills
	Embodiment	Constrains movement which assists directed perception	Simplifies body mapping between human and robot	Enables simple but robust low-level behaviors	Places physical limits on successive actions		Allows the human to observe natural social feedback cues
	Integration	Allows robot to recognize cues in multiple modalities		Increases robustness through redundancy		Assists in transfer to new modalities	Allows human to use natural modalities (voice, gesture, facial expressions)



The Research Team

Paul Fitzpatrick

Brian Scassellati



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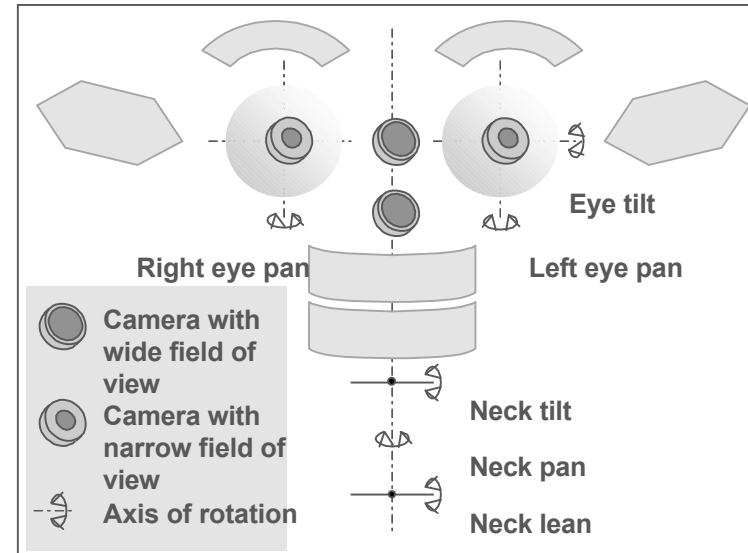
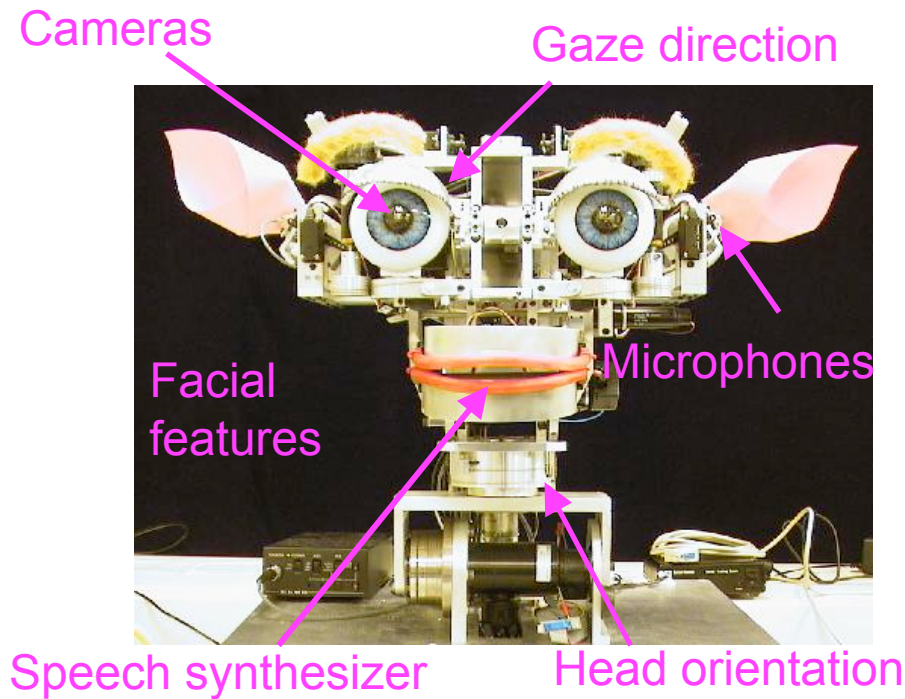
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Designing for the Human In the Loop



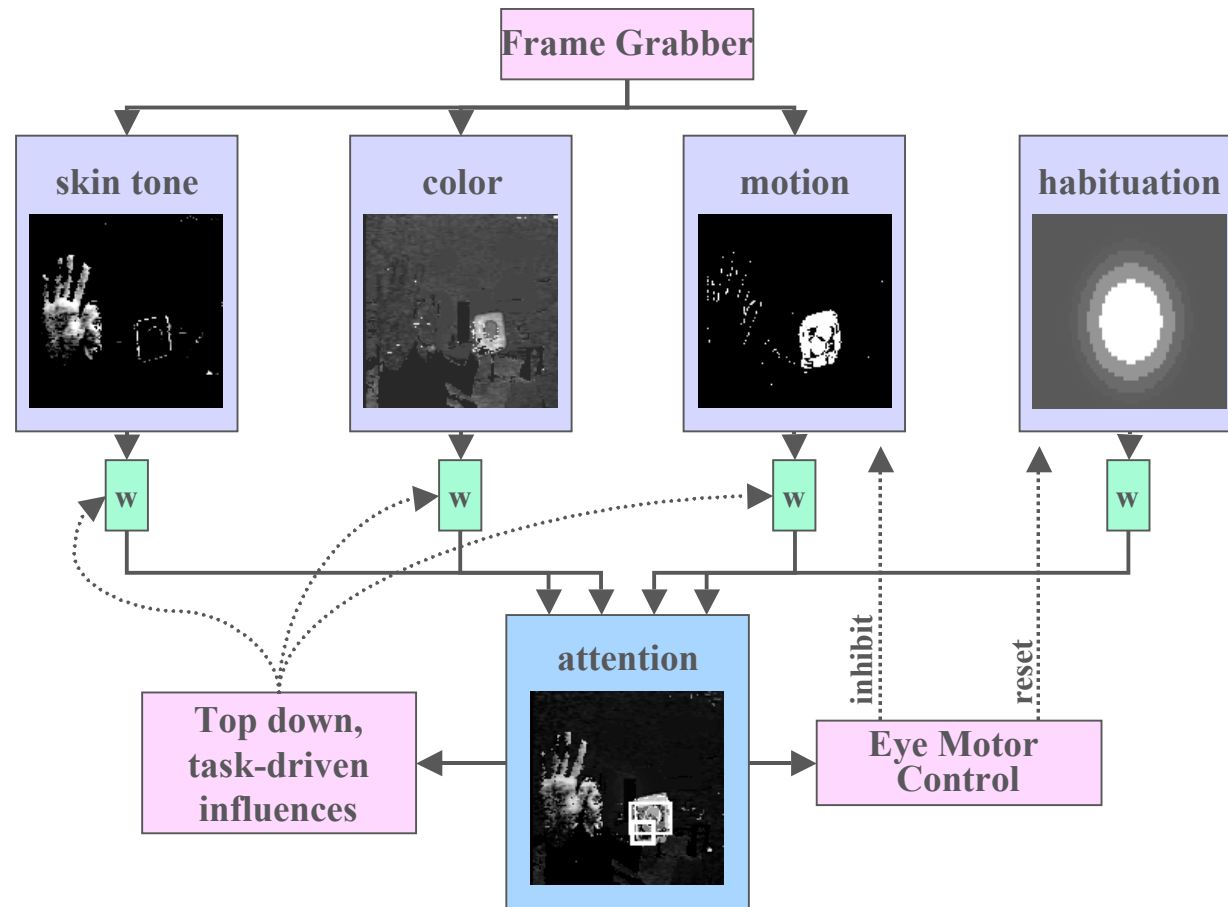


Evaluation Metrics

- **Must consider human response as well as robot's behavior**
 - Subjective but quantifiable
- **Four Criteria used to evaluate Kismet:**
 - Can people intuitively read and do they naturally respond to Kismet's social cues?
 - Does Kismet readily elicit social cues (scaffolding) from the human that could be used to benefit learning?
 - Can Kismet perceive and appropriately respond to these naturally offered social cues?
 - Does the human adapt to the robot, and robot adapt to the human, in a way that benefits the interaction? (entrainment)

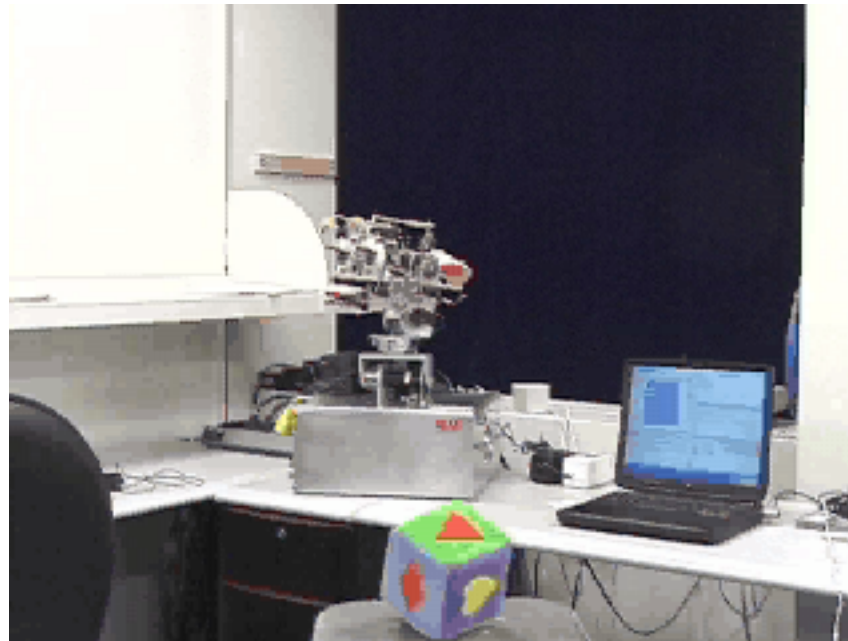


Visual Attention-Source of Coherence



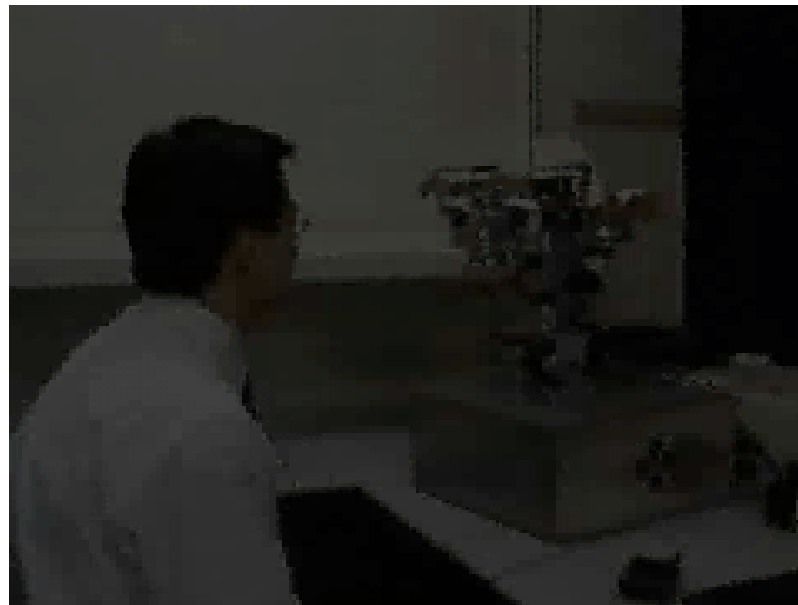


Video of Visual Search for Toy





Turn-Taking and Proto-Dialog





Our Approaches Lead to Specific Progress on Each Research Issue

Research Issues

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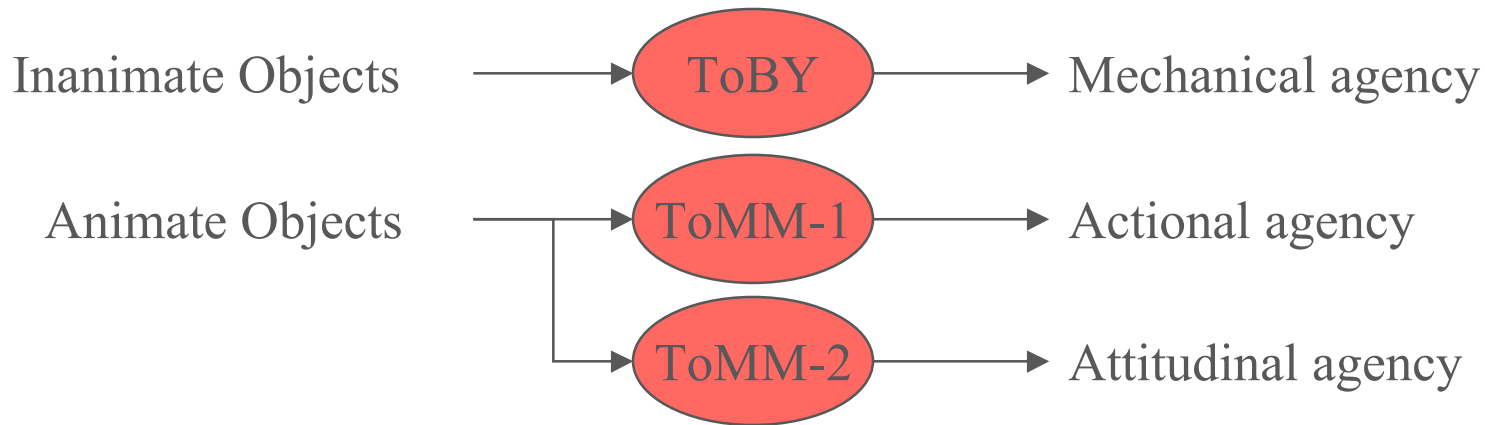


Theory of Mind

- For complex learning tasks the robot needs to
 - understand what it is the human knows
 - » about the world
 - » about the robot's current state
 - understand what the human is doing and why
 - understand what is the human



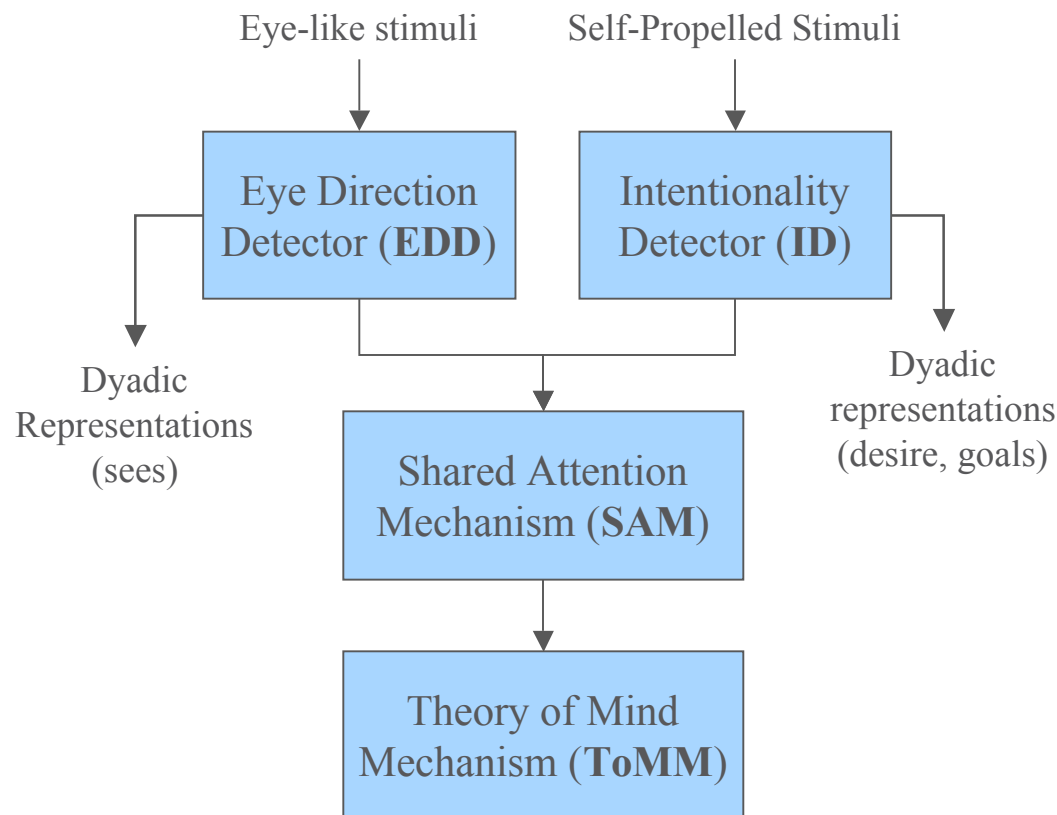
Leslie's Model



- Divides the world into three spheres of causation, each of which is processed by a separate module:
 - Theory of Body (ToBY) – applies naïve physics to objects
 - Theory of Mind Mechanism (ToMM) – applies folk psychology to agents
 - » System 1 applies rules of goals and desires
 - » System 2 applies rules of belief and knowledge



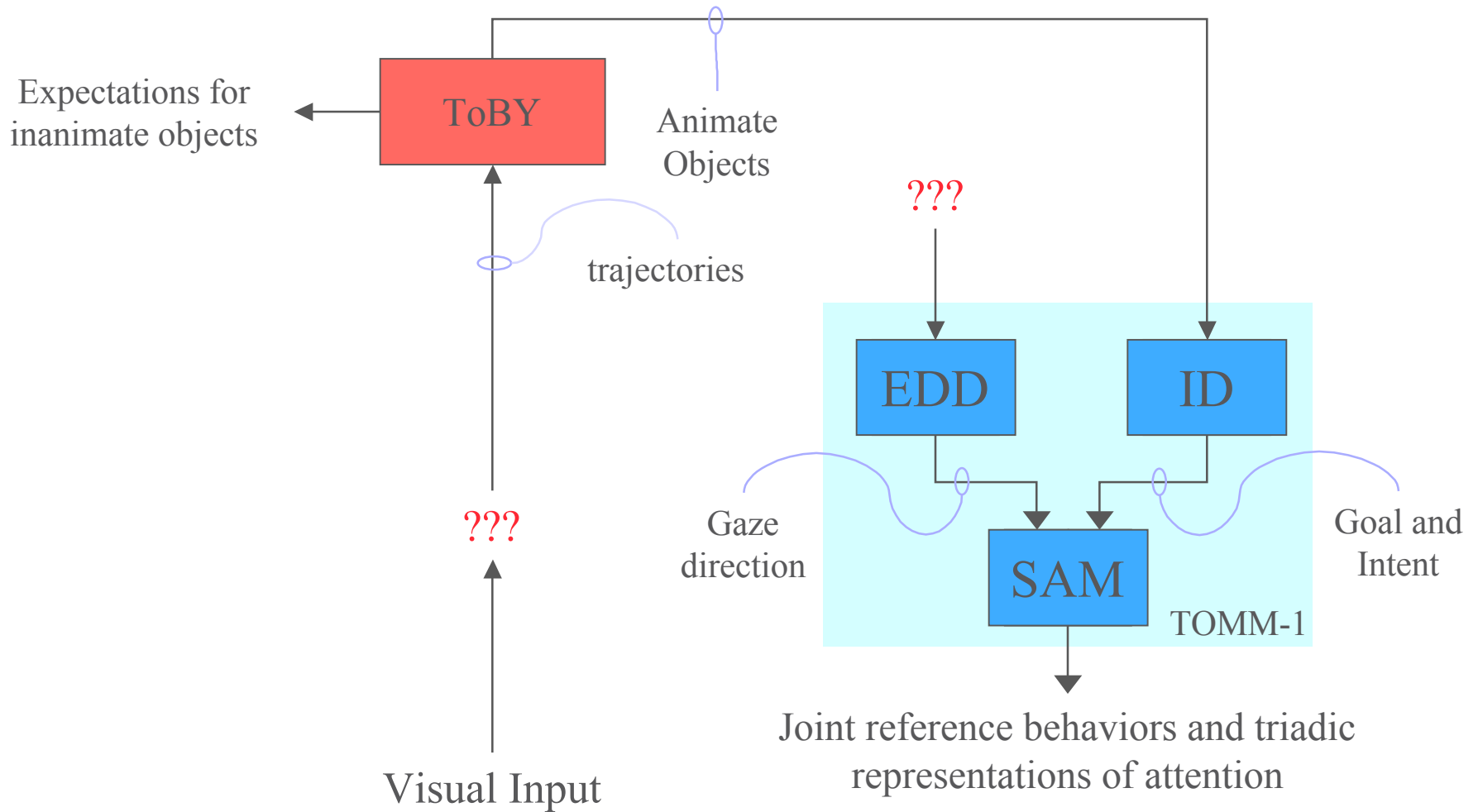
Baron-Cohen's Model



- **Requires two types of input stimuli:**
 - Eye-like stimuli
 - Self-propelled (animate) stimuli
- **Provides greater detail on the inner workings of Leslie's ToMM-1 module**
- **Proposes that autism is an impairment of either SAM (subgroup A) or ToMM (subgroup B)**

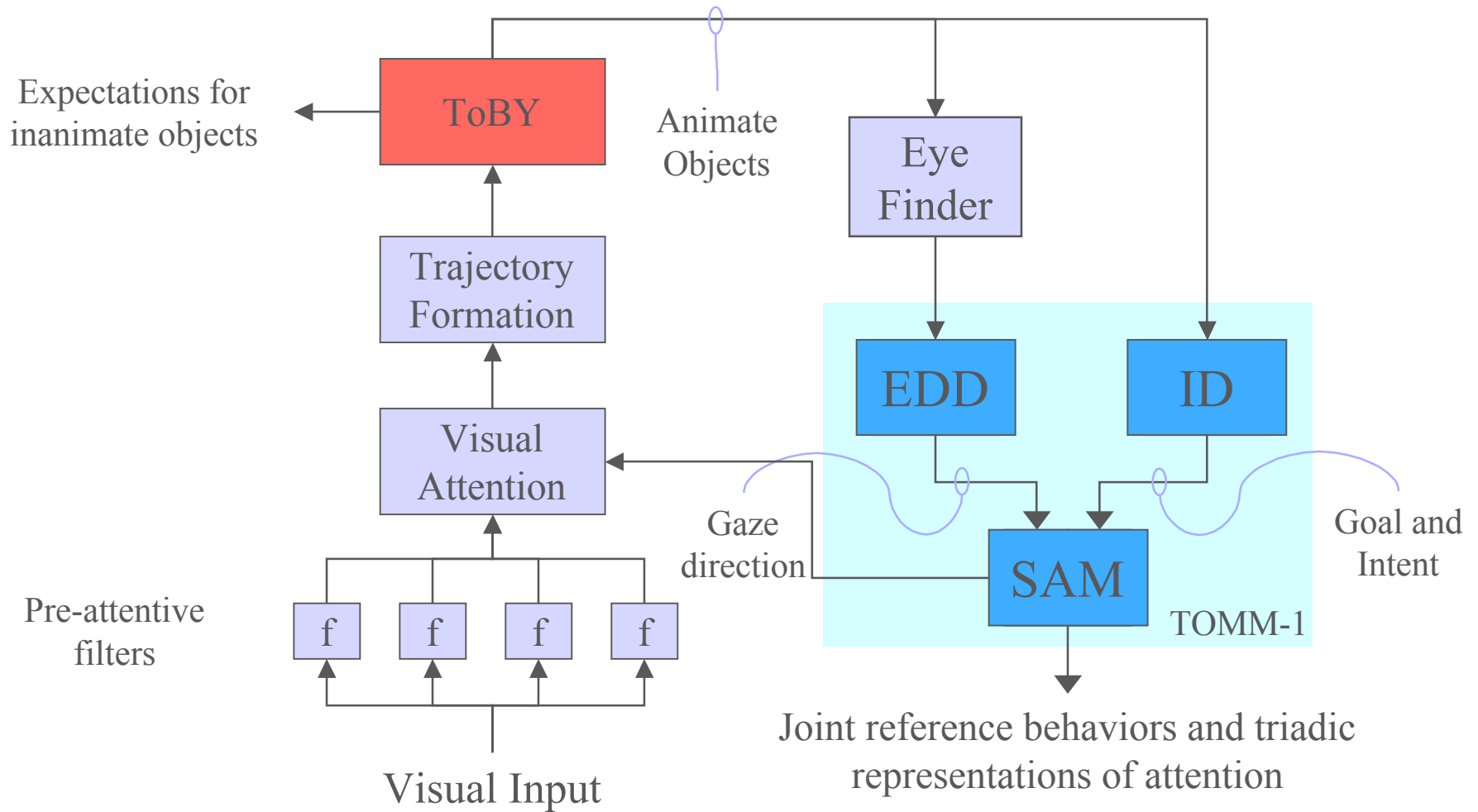


Scassellati's Model



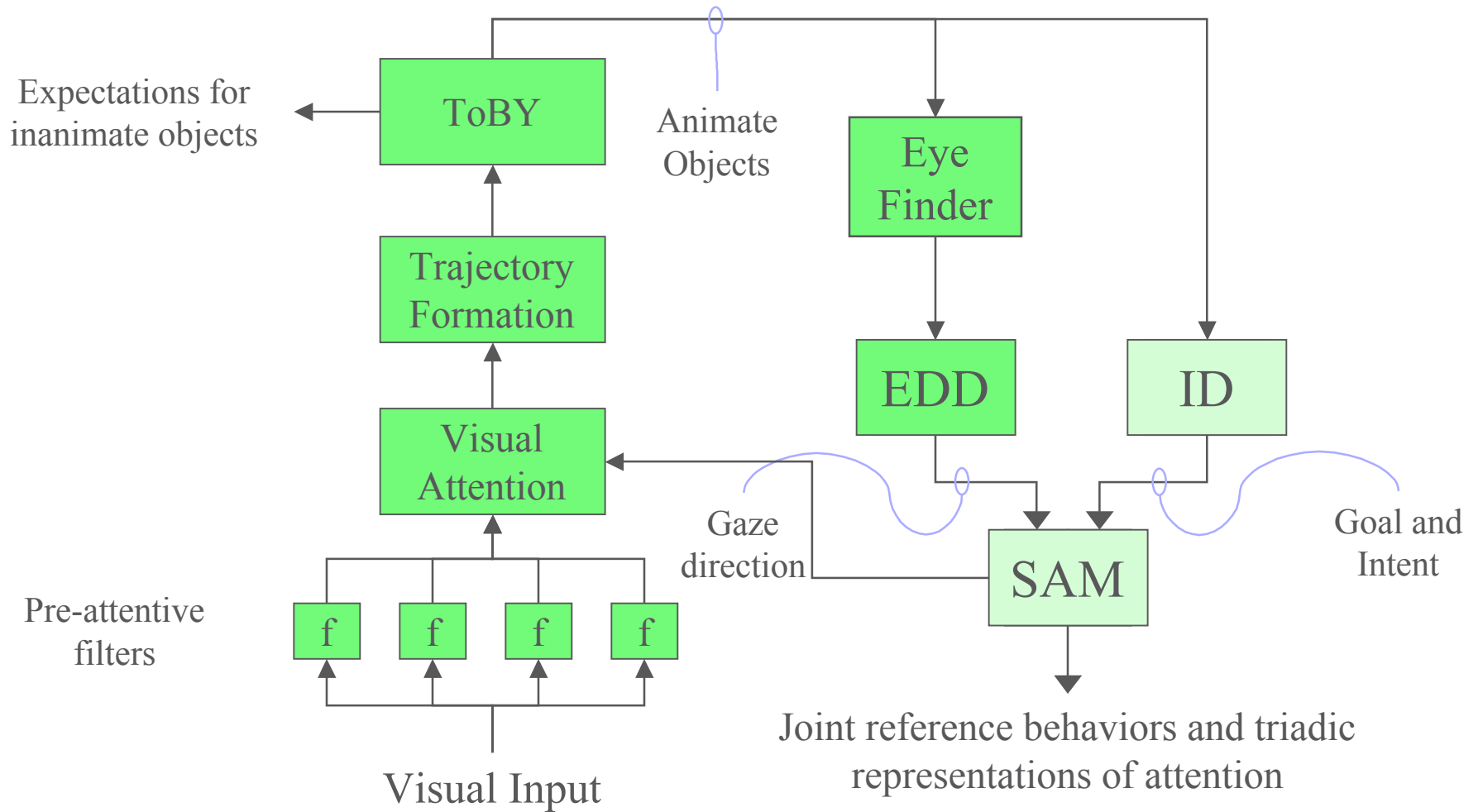


Scassellati's Model



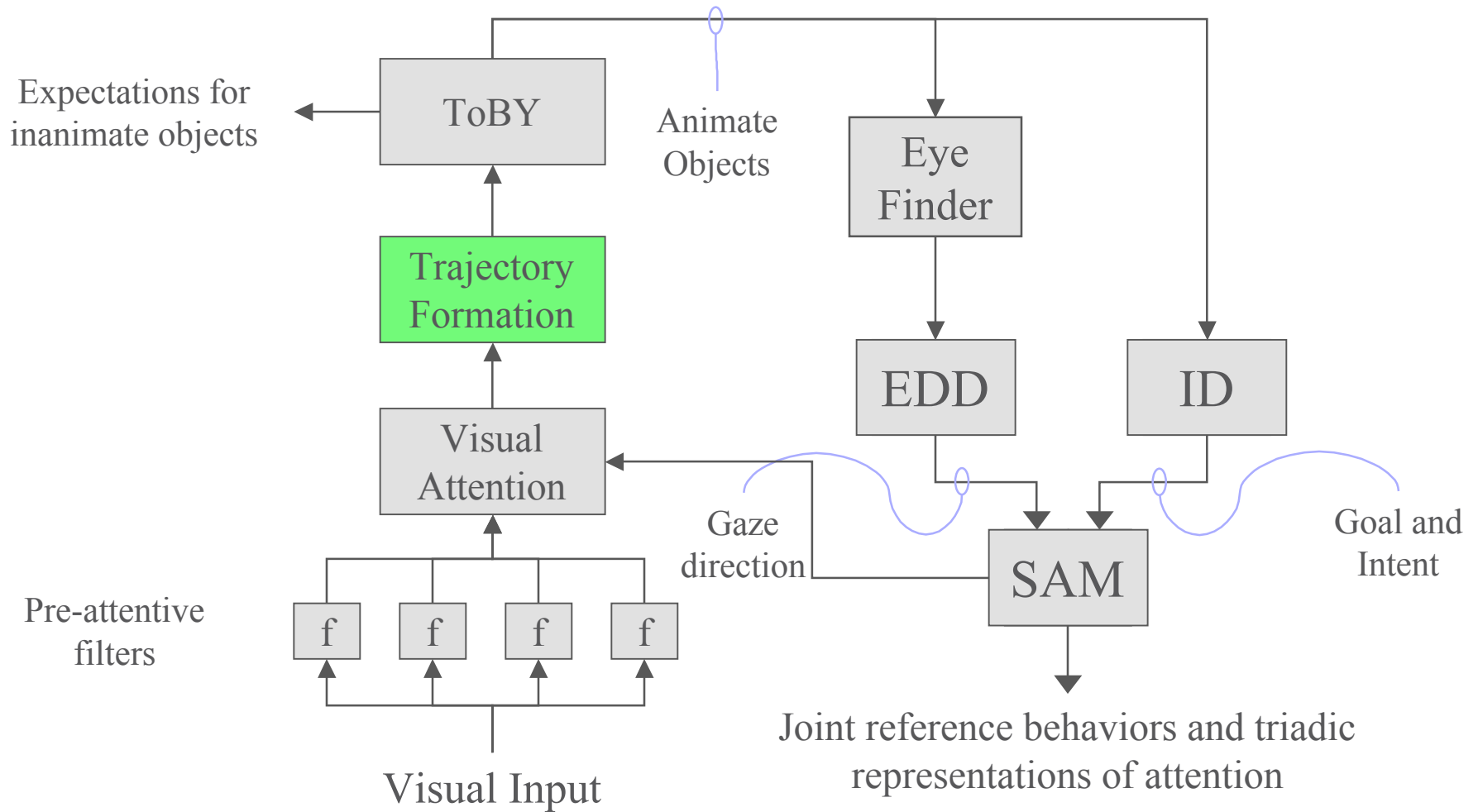


Where We Are Now



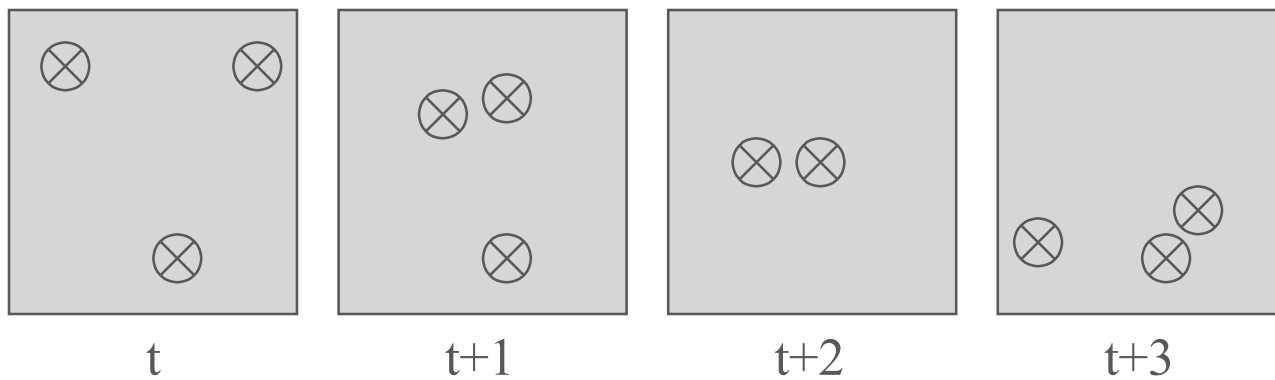


Roadmap





Trajectory Formation

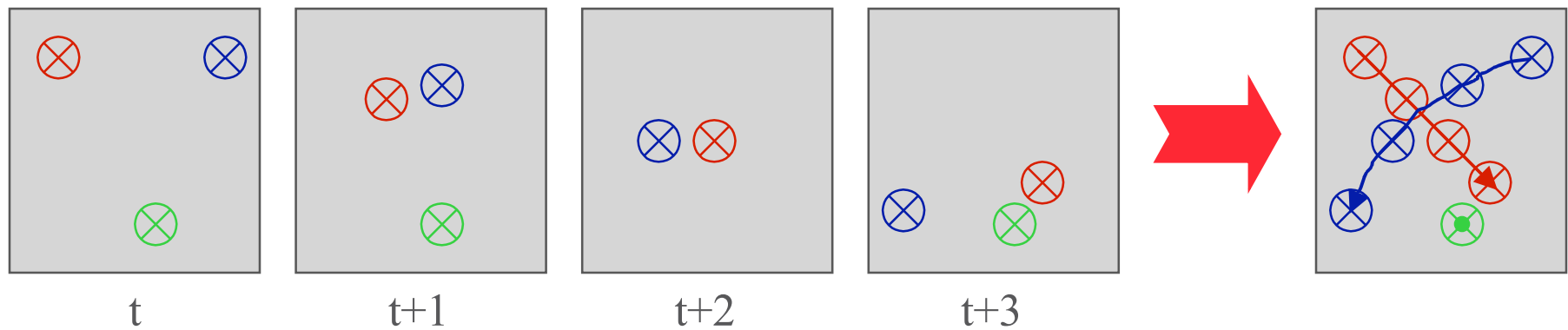


- Each frame produces a set of target points

$$\{P_t^1, P_t^2 \dots P_t^n\}, \{P_{t+1}^1, P_{t+1}^2 \dots P_{t+1}^n\}, \dots \{P_{t+k}^1, P_{t+k}^2 \dots P_{t+k}^n\}$$



Motion Correspondence



- Each frame produces a set of target points

$$\{P_t^1, P_t^2 \dots P_t^n\}, \{P_{t+1}^1, P_{t+1}^2 \dots P_{t+1}^n\}, \dots \{P_{t+k}^1, P_{t+k}^2 \dots P_{t+k}^n\}$$

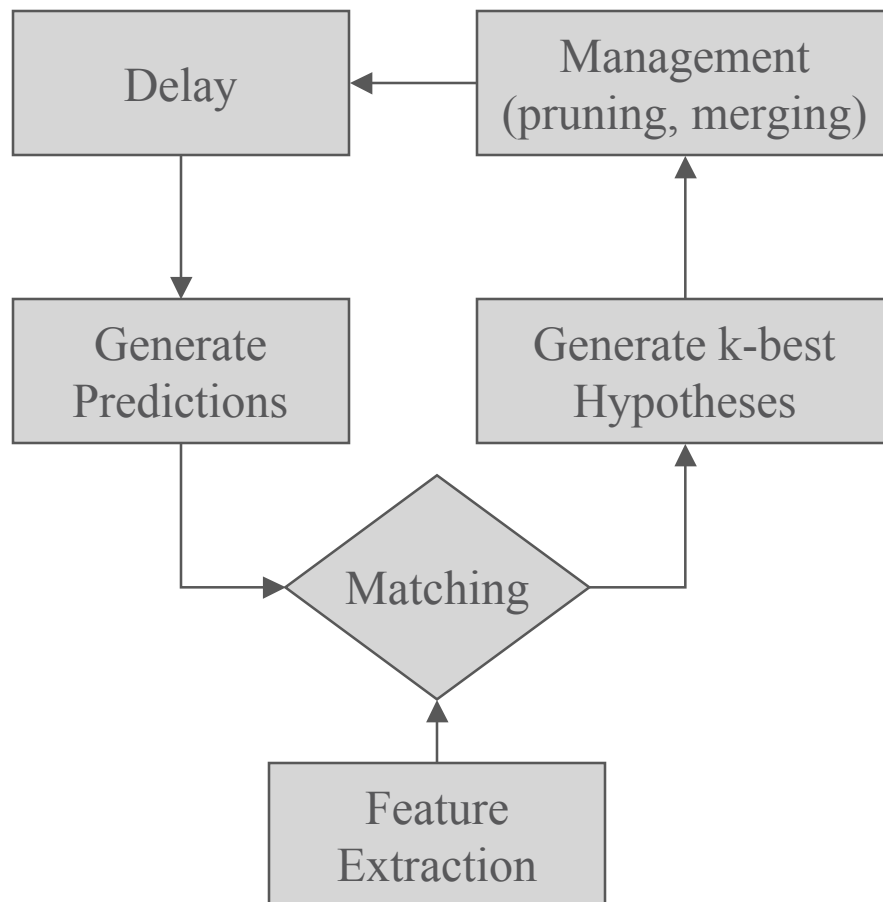
- Objective is to identify sequences through a subset of the frames

$$\{P_t^{\alpha_{11}}, P_{t+1}^{\alpha_{12}} \dots P_{t+k}^{\alpha_{1k}}\}, \{P_{t+1}^{\alpha_{21}}, P_{t+1}^{\alpha_{22}} \dots P_{t+1}^{\alpha_{2k}}\}, \dots \{P_{t+k}^{\alpha_{k1}}, P_{t+k}^{\alpha_{k2}} \dots P_{t+k}^{\alpha_{kk}}\}$$



Multiple Hypothesis Tracking

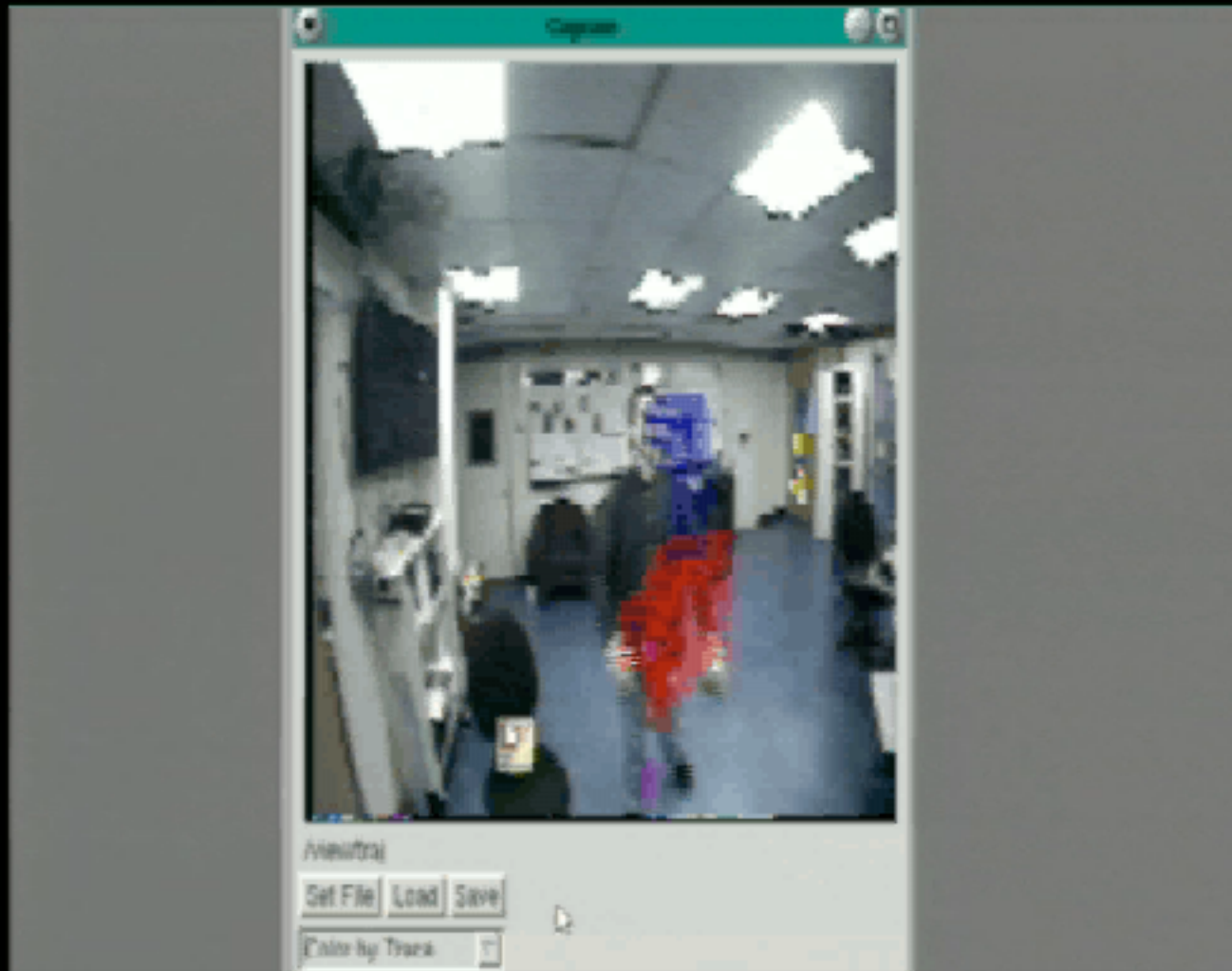
(Reid, 1979)(Cox and Hingorani, 1996)



- **Allows for**
 - Trajectory Initiation
 - Trajectory Termination
 - Minor occlusion
- **Matching based on**
 - Mahalanobis distance
 - Cross-correlation of features including
 - » Area
 - » Overall Saliency
 - » Saliency among the individual feature channels

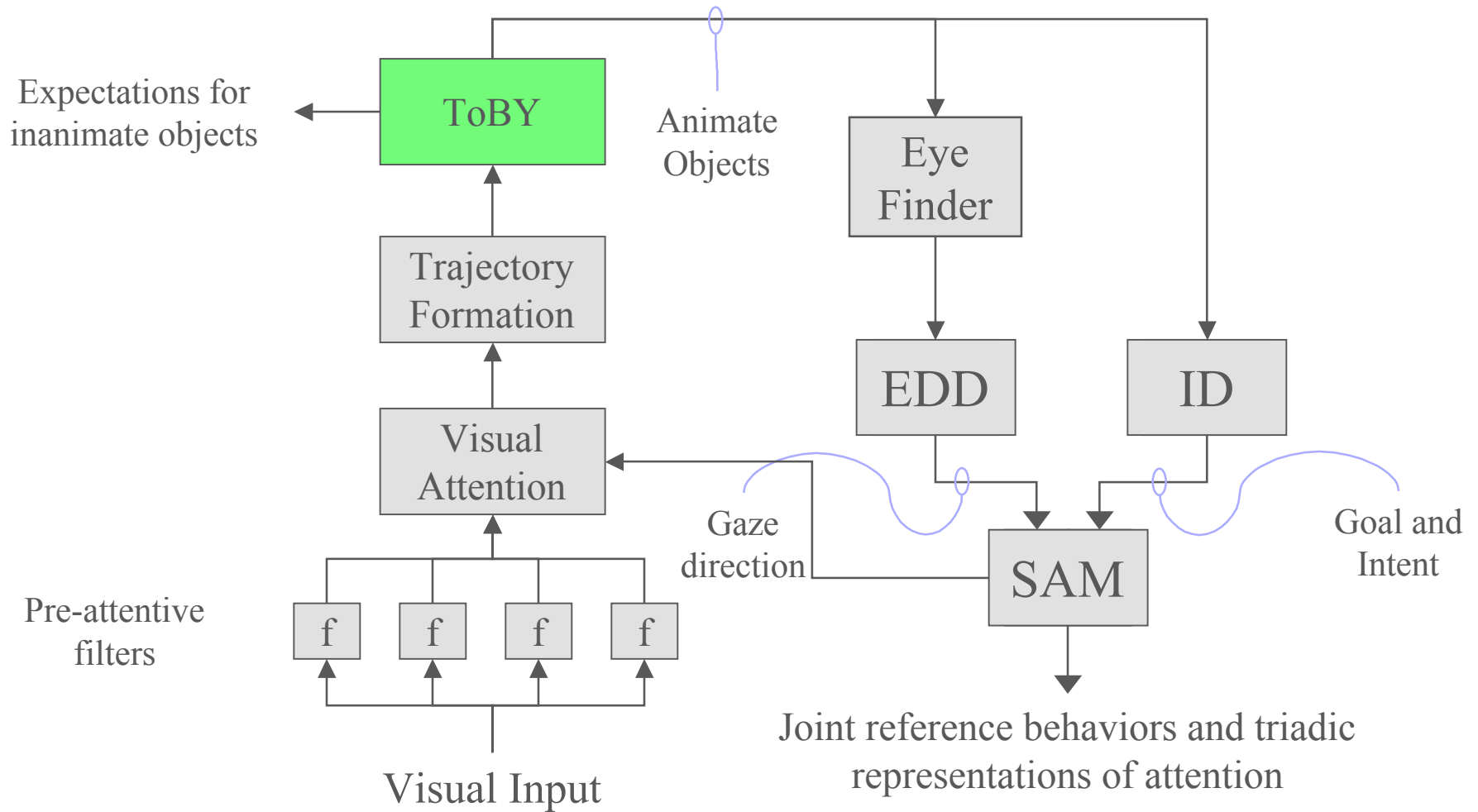


Trajectory Example



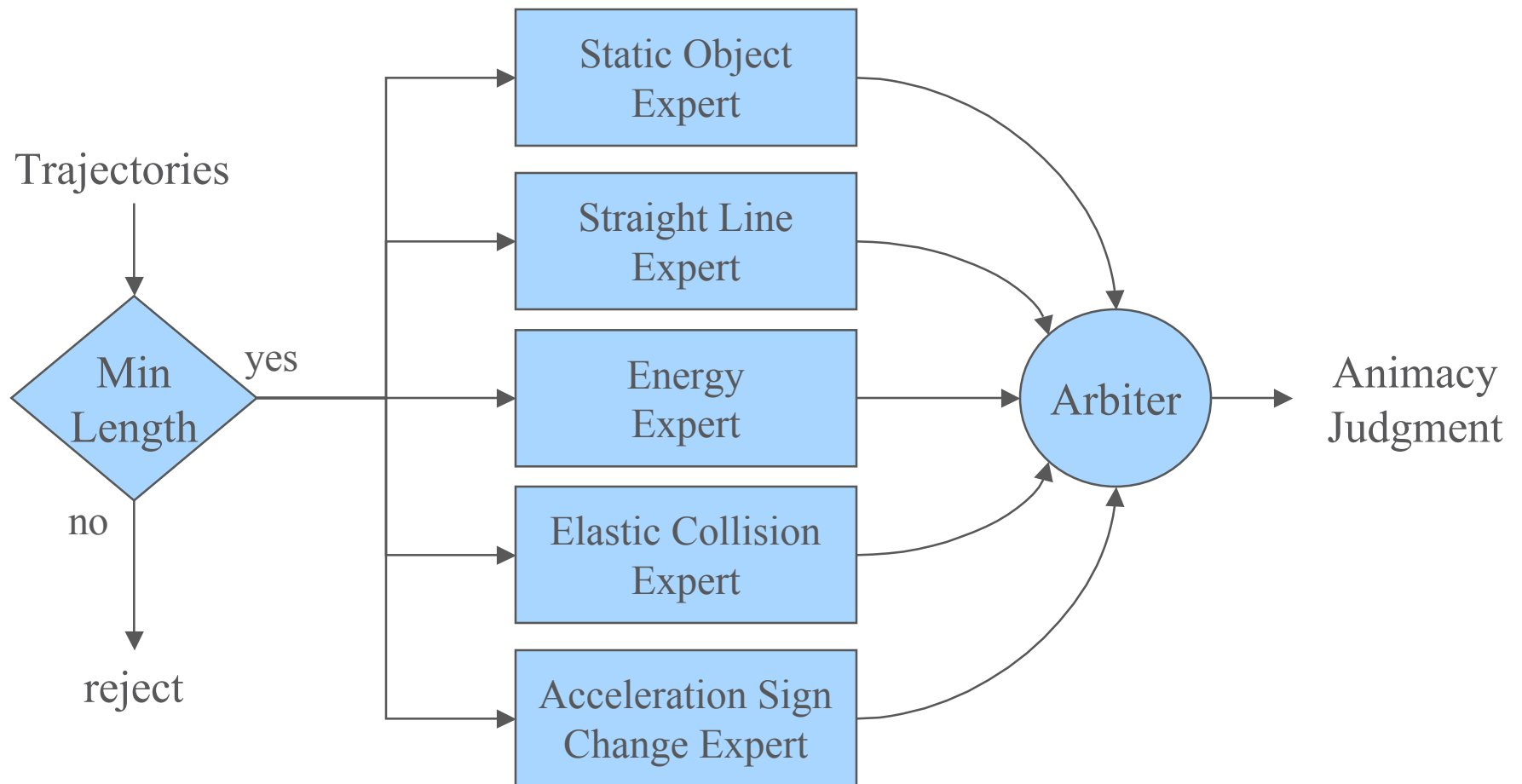


Roadmap





ToBY Architecture





ToBY Agents

Straight Line Expert

$$\min \sum \sqrt{(v - \bar{v})^2}$$

Minimize the sum of the deviations from the mean velocity

Energy Expert

$$\frac{1}{2} m v_y^2 + mgy$$

Constant mass and the inertial system provides the gravity vector

Elastic Collision Expert

$$v_{t-1}^1 + v_{t-1}^2 = v_{t+1}^1 + v_{t+1}^2$$

Look for transfer in velocities before and after collision

Acceleration Sign Change Expert

$$\text{sgn}(v_t) \neq \text{sgn}(v_{t+1})$$

Look for multiple sign changes in the acceleration



Animacy Results



Animate



Inanimate

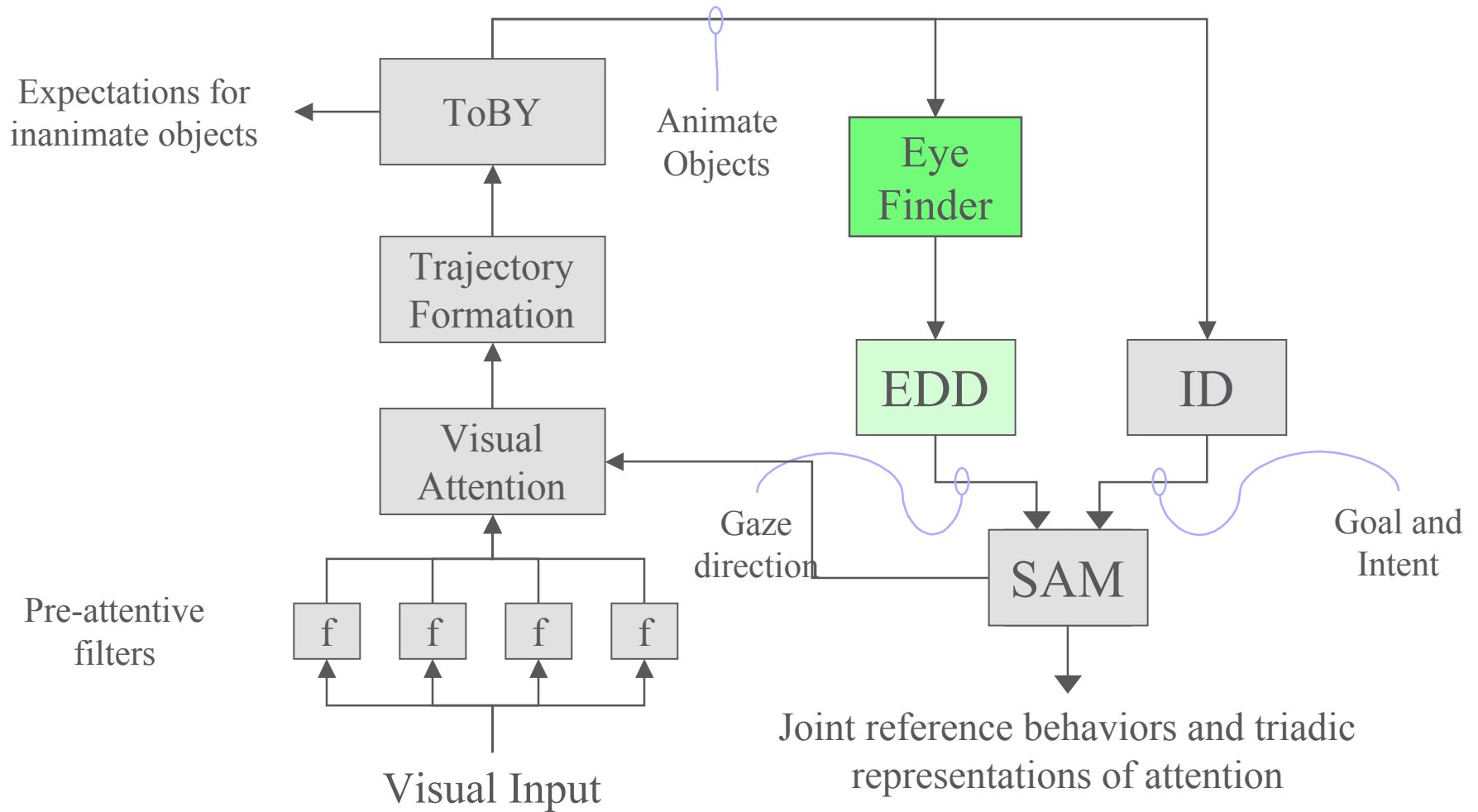


Animate

- Arbitration methods
 - Weighted sum
 - Winner-take-all



Roadmap





Post-Attentive Visual Processing: Finding Faces and Eyes

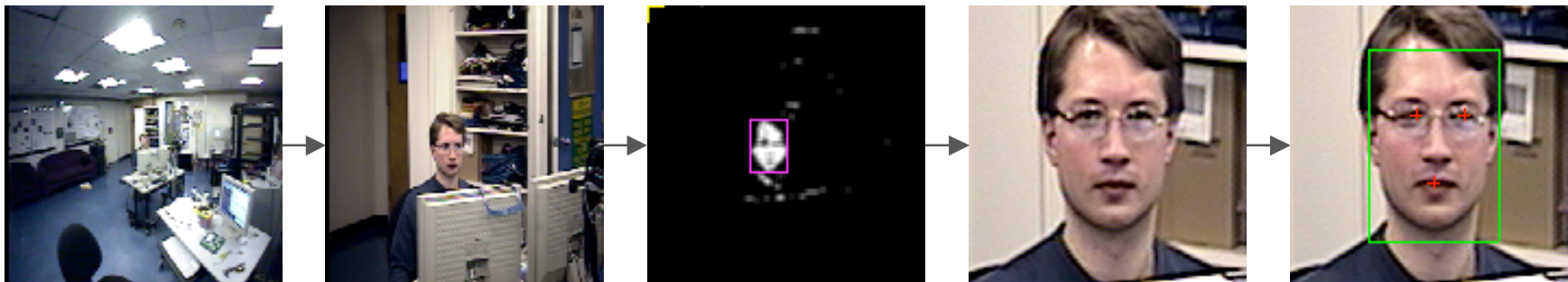
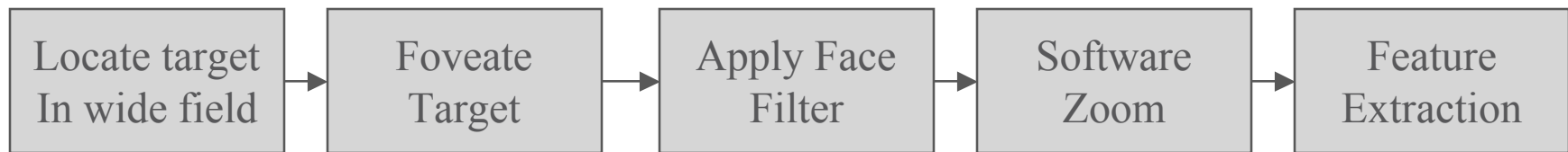


Can you find the face in this image?

Can you tell where Scaz is looking?



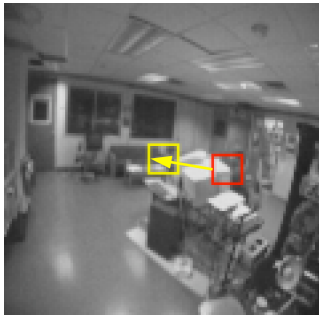
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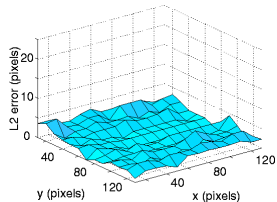
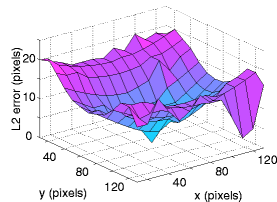


Two Sensory-Motor Mappings

Saccade Map

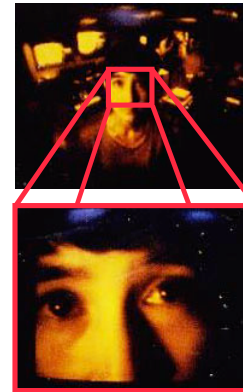


Maps image positions to motor commands necessary to center that location in the visual image.

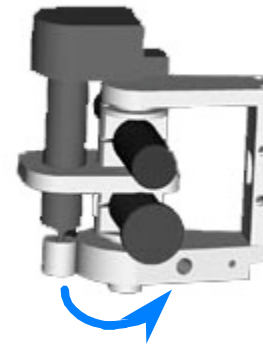


Learned using standard self-supervised techniques (lookup tables, neural nets, etc.)

Peripheral-Foveal Map



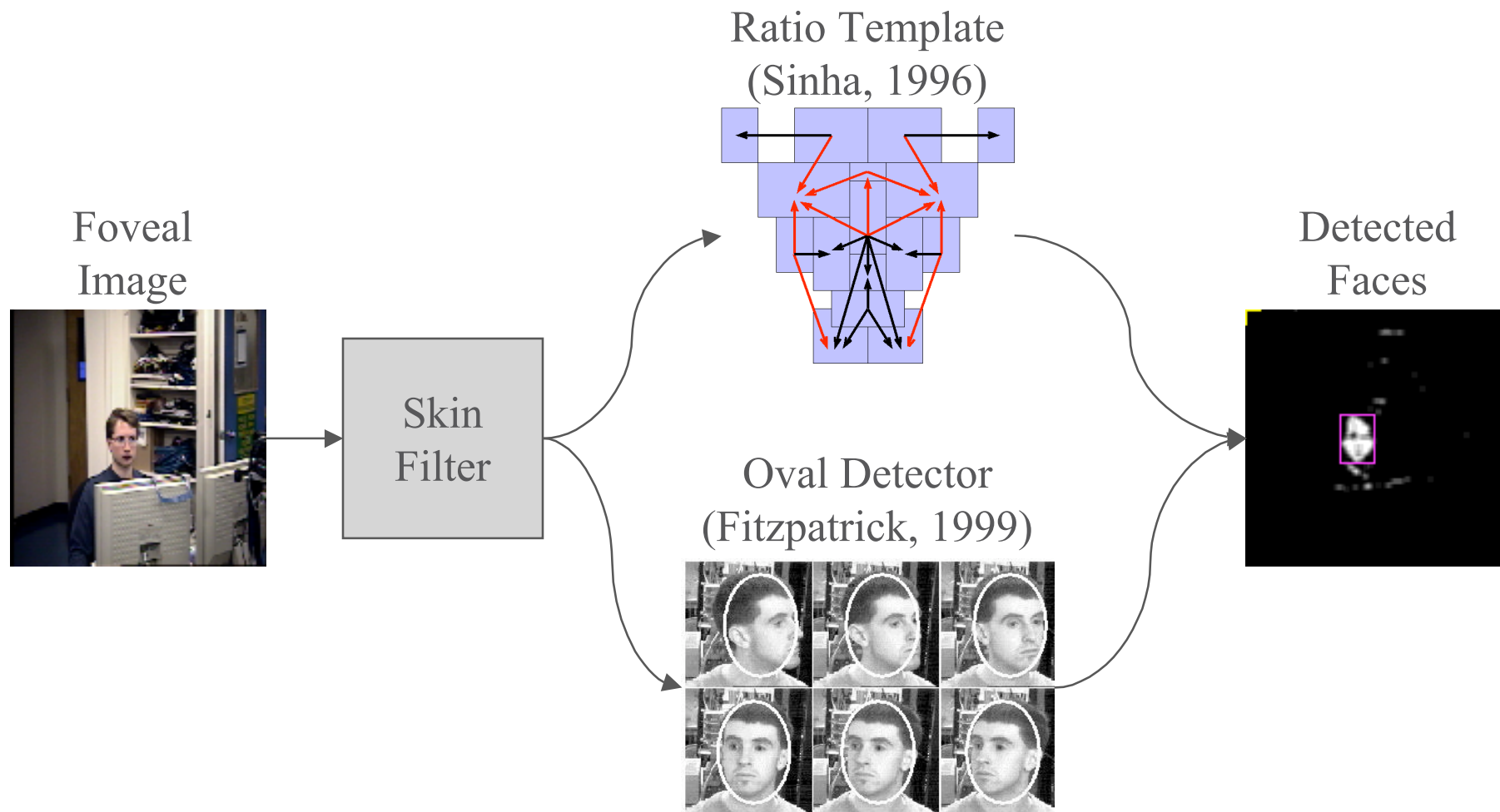
Maps pixels in the peripheral image to pixels in the foveal image.



Scale is learned using optic flow rates obtained while the cameras are moving. Position is learned using correlation.

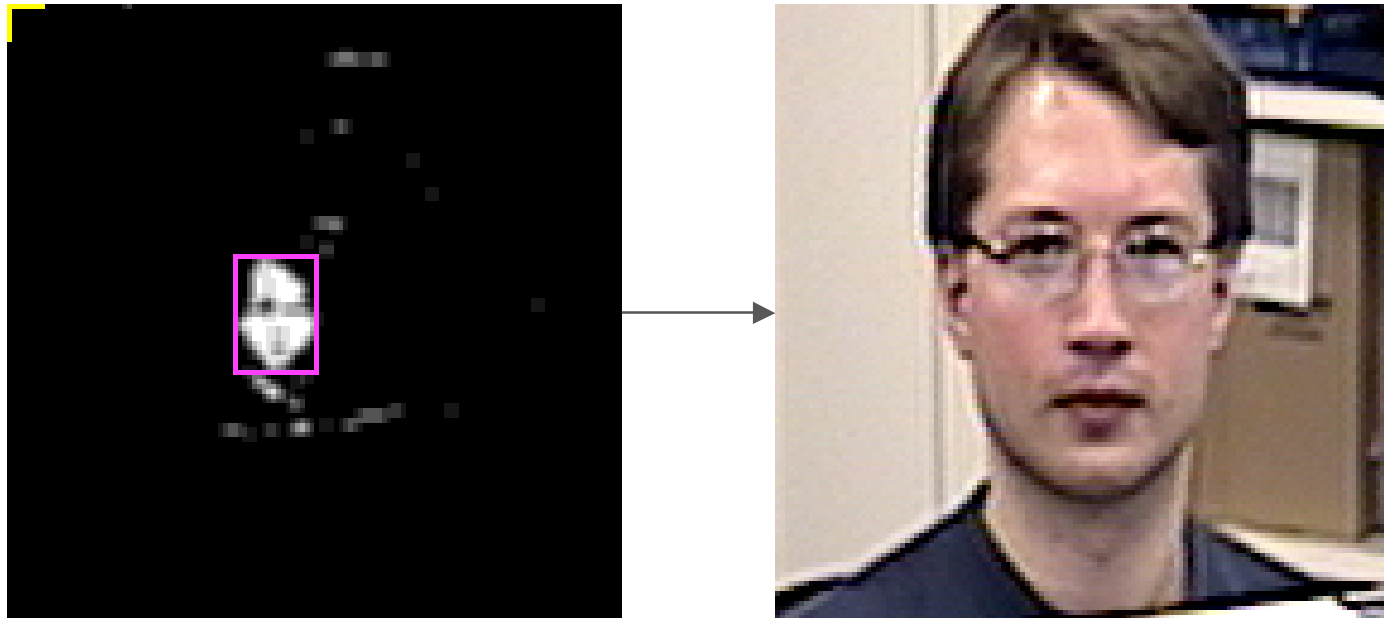


Face Finding



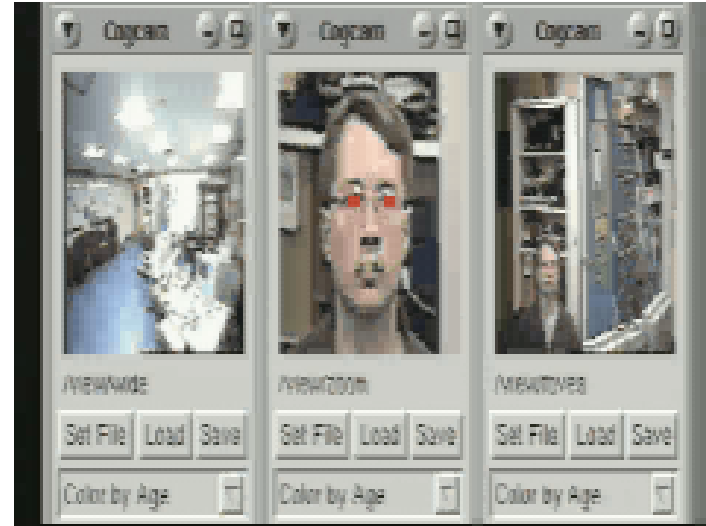
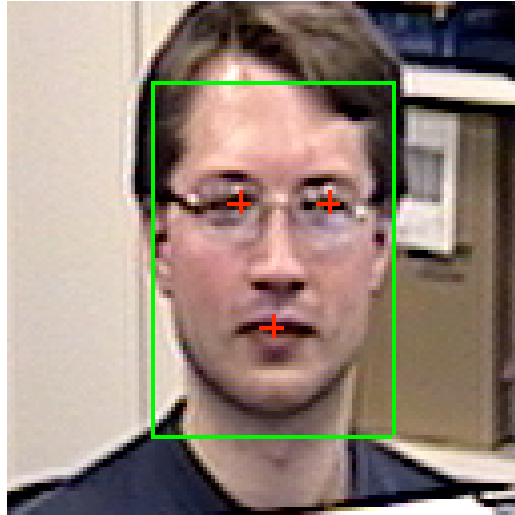


Software Zoom



- From full 640x480 image, extract the relevant 128x128 sub-image
- Introduces the majority of the system delay

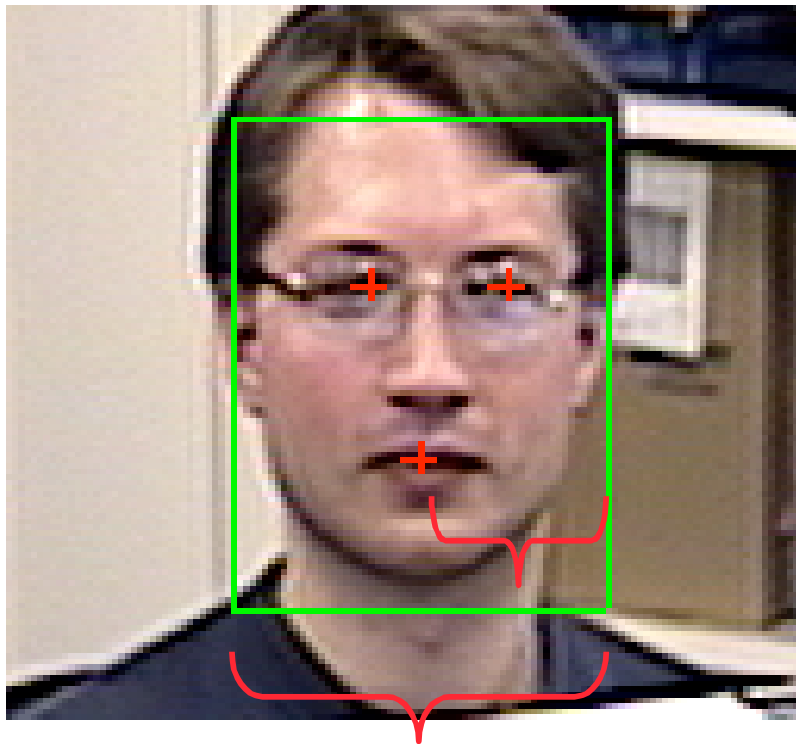
Feature Finding



- **Locate eyes and mouth by looking for centroid of luminance minima**
- **Mouth:**
 - Iterative algorithm with adaptive regions provides performance similar to simulated annealing
- **Eyes:**
 - Add symmetry requirement



Head Pose Derivation



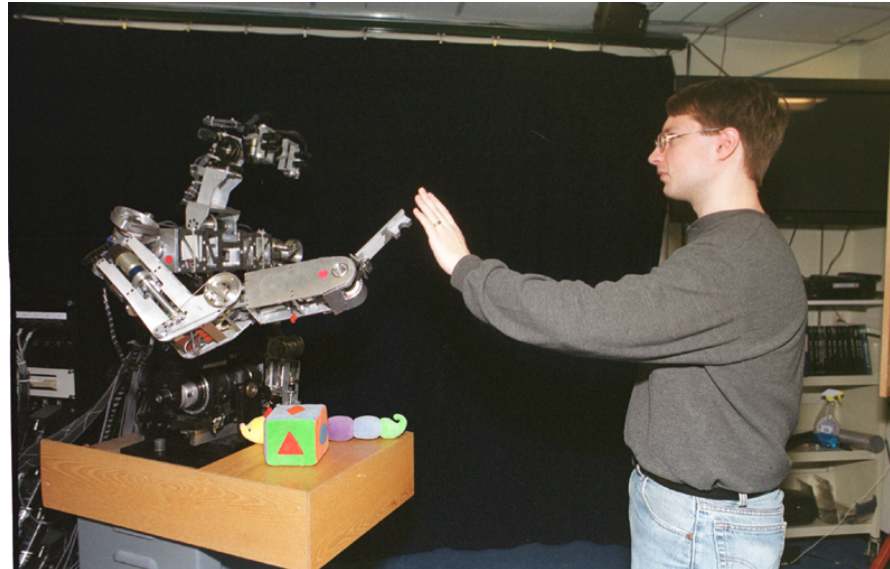
- Obtain estimate of head pose from the positions of the mouth, nostrils, and eyes

$$\frac{\text{Mouth} - \text{to} - \text{Left Edge}}{\text{Face Width}}$$

- Failure modes
 - Match to nostril
 - Vertical Lengthening
- Accuracy of +/- 5 degrees at a distance of 6 meters



An Application to Mimicry



- **Imitation of arm trajectories**
 - No body model
 - Imitation based around detected faces
 - Imitation only of animate motion trajectories



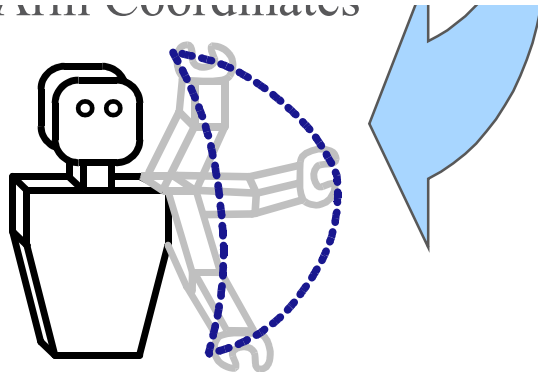
Mapping Visual Trajectories to Arm Movements

Visual Coordinates

- Postural primitives define a sub-

Current demos suffer from
low stiffness as we rebuild
our arm control system

ARM COORDINATES



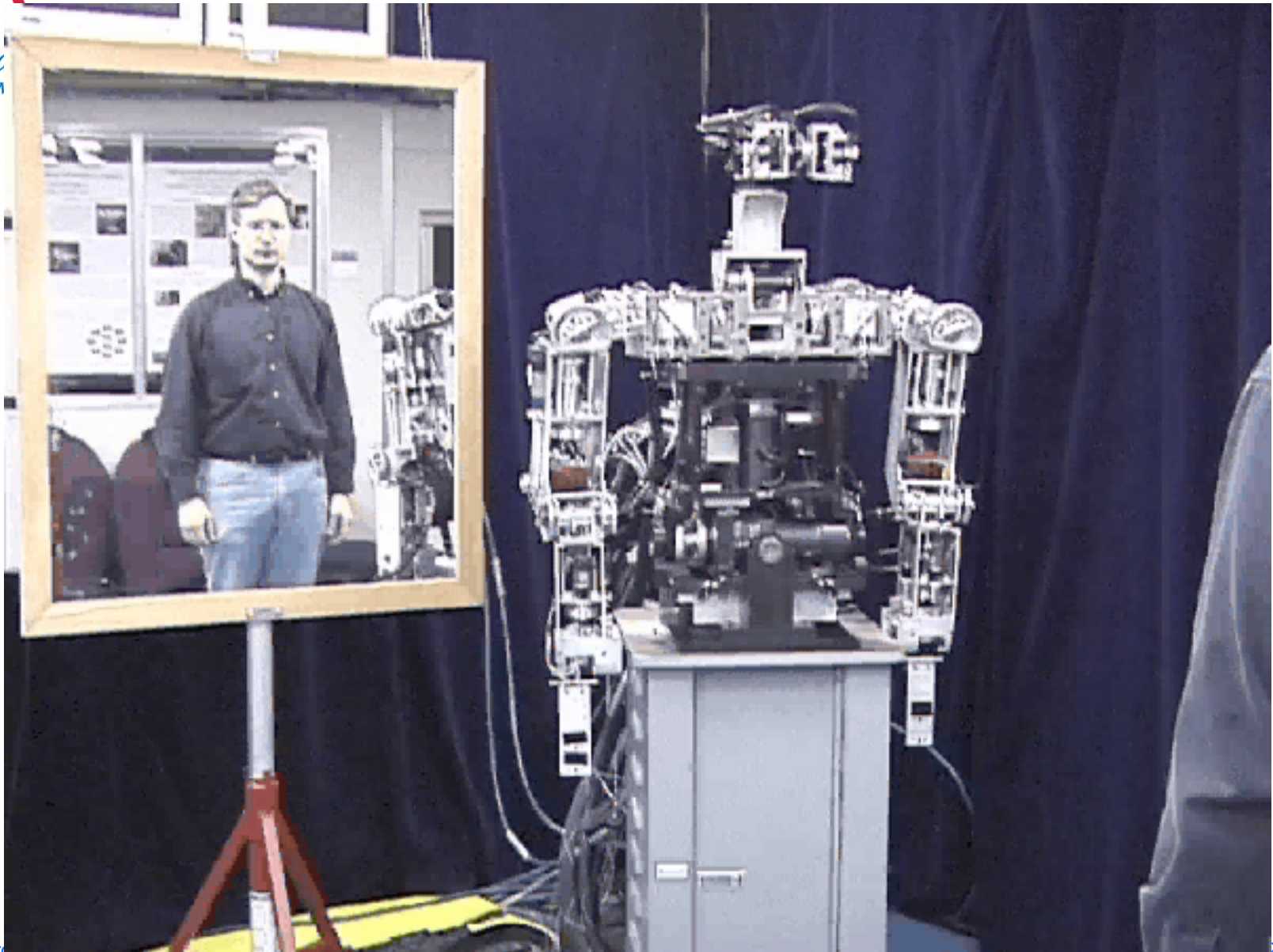
- Mapping is based on perceived head position and robot's own symmetry axis

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Imitation Based on Social Cues

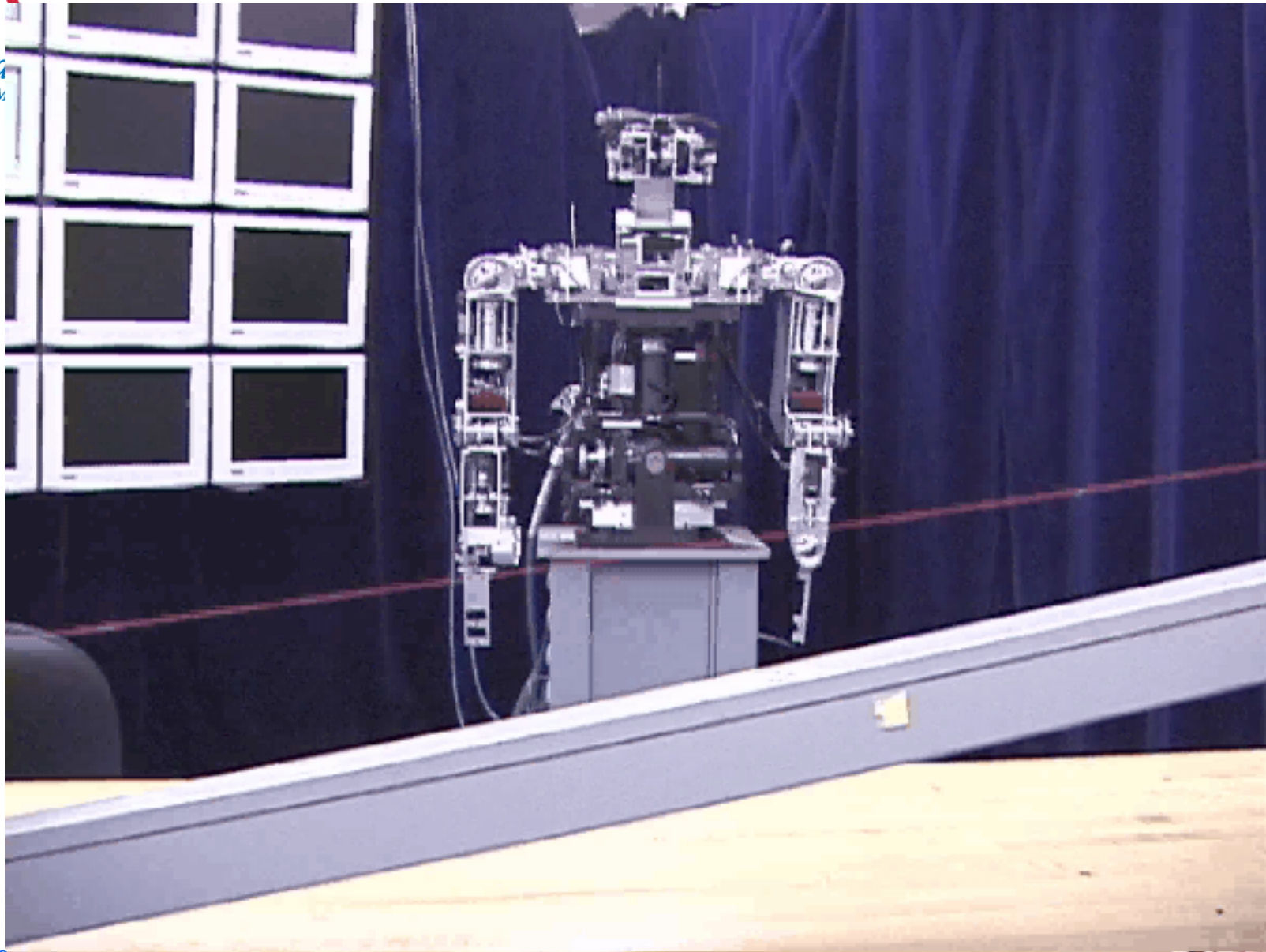


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Imitating an Object



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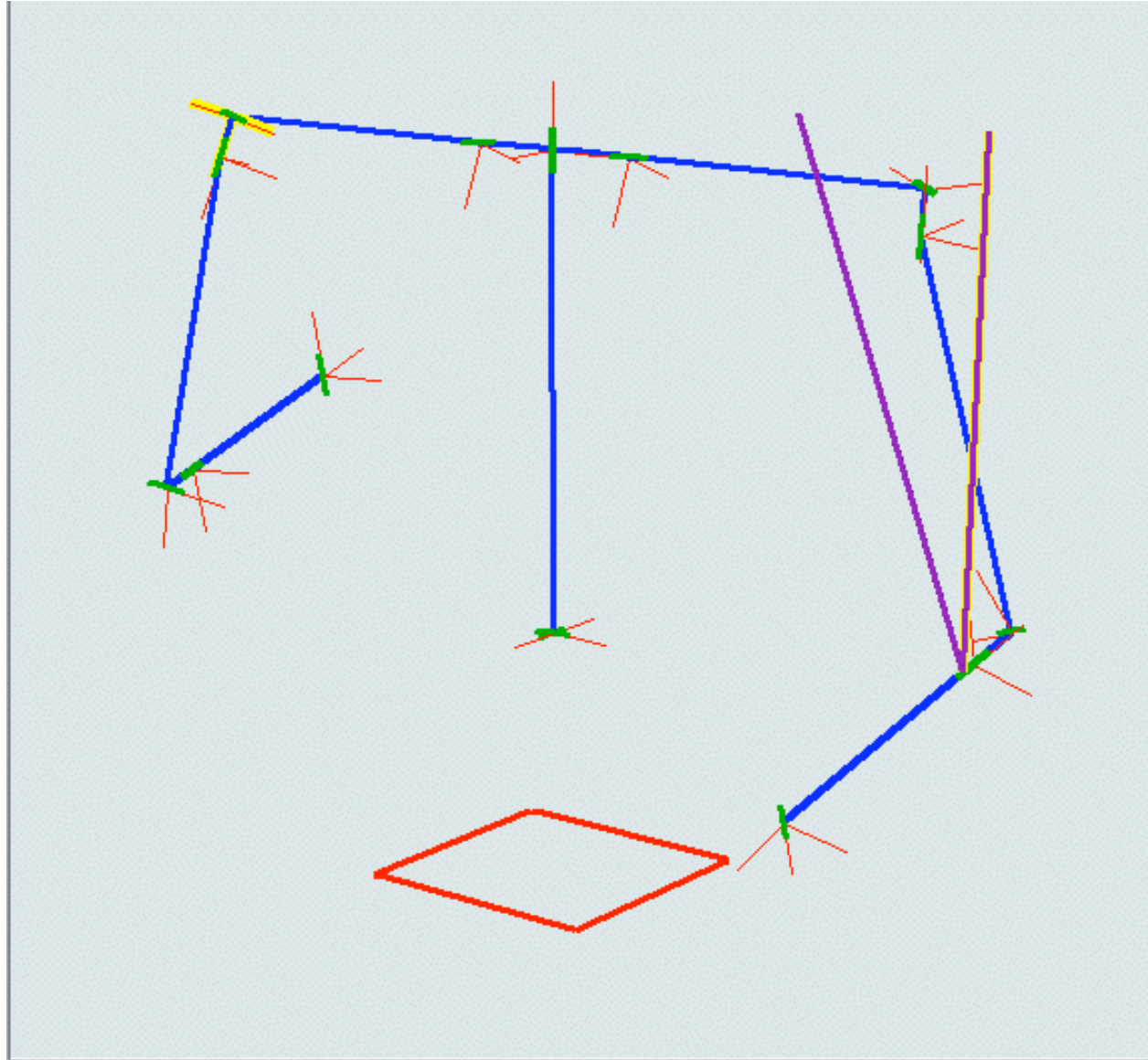
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Work in Progress: Muscle Groups



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Work in Progress: Sequence Learning

