MAPI – Computer Vision

Active Vision

- Most computer vision systems still use a static camera.
 - Some have a camera mounted on a vehicle or a moving arm, but usually there is little control over the camera's movements.
- In biological systems, the eye is extremely active.
 - Rapid, frequent saccades (fast changes in direction) and tracking movements are made.
 - Information pickup is active, not passive. As well as eye movements, body movements are made to change the position of the viewpoint.

 Rapid eye movements recorded during driving (in this case by M.F. Land of BIOLS) provide a good example.



- The term **active vision** is used in at least two senses.
 - "active operation in the world in order to change the images that are being collected in a way which enhances task achievement"
 - This recognises that vision is part of a perception-action cycle.
 - "active autonomous processes (e.g. snakes) which exploit the coherence of images in a sequence in order to efficiently and reliably track aspects of interest over time"
 - This recognises that vision takes place in a temporal continuum.

- Active vision requires new:
 - Hardware
 - active heads, mobile cameras;
 - Software
 - algorithms that integrate information over time;
 - Philosophy
 - the goal of vision is to enable useful interaction with the environment. The goal is not to find the 3-D layout, and is not to make a symbolic description.

Active heads

- Recently, there has been much research on developing active heads:
 - mountings for cameras that allow very rapid movements for tracking purposes.
 - High angular velocity is necessary for tracking whilst high angular acceleration is important for capture.
 - Maximum angular velocities of several hundred degrees per second are achieved, with maximum accelerations of thousands of degrees per second
- Such systems are built by research groups, not commercially.
- Motor and controller technology are very important.

Active heads

The device on the right was built as part of the *COG* project at MIT.

See http:// www.ai.mit.edu/people/ scaz/3heads.html

It has two pairs of cameras, with independent *pan* (sideways) movements but coupled *tilt* (up and down) movements.



Tracking in Active Vision

- Current tracking methods concentrate on active contours, deformable templates, dynamic models and probabilistic predictors such as Kalman filters.
- These methods are related to optic flow methods, but make more use of the **history** of the motion information is integrated over time.

• Active contours

 An approach which combines high-level feature tracking with a model of the motion is the snake. This is a parametrised curve in the image (often a spline curve). Snakes have particular advantages for nonrigid motion.

Tracking in Active Vision

- **Simulated dynamics** of a physical system are used to give the contour **momentum**, smoothness, elasticity, and other properties.
 - The equations of motion of the contour combine forces from these intrinsic properties with forces from the image to make the contour track objects through frames.



Tracking in Active Vision

• Tracking using anticipation

- An key to successful tracking is to focus computational resources on important parts of the image.
- Using anticipation allows this to occur. Knowledge of the properties of the tracked object — geometrical and dynamic — is central to this.





- The distribution of receptor cells (particularly cones) on the human retina is extremely nonuniform.
 - The greatest concentration is at the **fovea**.
 - This contrasts with the uniform representation usually used in computer vision.



- we believe that we perceive a scene completely and sharply, only a small subpart is sharp (the one on the fovea).
- The complete perceived sharp image is created by *saccading* over the scene.



- Two parameters are of high importance in visual sensor design: **resolution** and **field of view**:
 - resolution determines the scale of the smallest details that can be detected in the images;
 - field of view determines how far from the central view point objects can be detected.
- Computer vision systems usually control efficiency by controlling image size, either by
 - reducing resolution or the field of view:
 - image resolution is reduced uniformly according to some desired criteria;
 - field of view is reduced by defining windows of interest around objects, where further processing is preformed.

- These strategies have been applied successfully in some structured environments, where good models for object sizes and motions exist.
- However, they are too rigid to be applied in more unstructured situations.
 - In the first case, resolution may not be enough to detect objects whose scale changes along time.
 - In the second case, moving objects easily move away from the windows of interest.

- Foveation deals with methods to represent images efficiently
 - preserving both the field of view
 - maximum resolution,
 - at the expense of reducing resolution at some parts of the image.
- With such a representation, the sensory strategy can allocate high resolution areas to objects of interest as they are detected in a wide field of view.

- Active vision has greatly increased the interest in foveal vision.
 - Active heads can be used to fixate objects of interest ensure that their image remains in the densely sampled region. Meanwhile the periphery can continue to pick up important events that might signal a new region of interest.

- Foveal vision offers
 - concentration of computational resources on important parts of the scene
 - task-based visual processes
 - good algorithms for template matching and optic flow estimation.

 Some systems uses two cameras for each eye, one to give a broad view and one to give a detailed central ('foveal') view.



- It is more elegant to use a nonuniform distribution of receptors in a single camera.
 - superpixel methods



Face recognition using foveal vision Silviu Minut *et al*



- It is more elegant to use a nonuniform distribution of receptors in a single camera.
 - One common receptor layout for foveal vision is the log-polar sampled image (LSI).
 - Positions are represented by ring and wedge rather than by row and column.





Face recognition using foveal vision Silviu Minut *et al*



- In the LSI, rotation and expansion are very simple transformations.
- Translation is complex, so active vision is needed to fixate and track points of interest.
- LSI cameras have been implemented in hardware, though much work uses software simulations.

- The Logpolar Transformation
- Is based on the complex logarithmic function log(z), which is used to approximate the retinocortical mapping of primates.
- Let us consider the complex retinal and cortical planes, represented by the variables z = x + jy and w = u + jv, respectively, where j is the complex imaginary unit:
- $w = \log(z) = \log(|z|) + j \arg(z) = \log(\rho) + j\theta$ ρ is the input eccentricity and θ is the input angle.

- The Logpolar Transformation
- Image rotations and scalings in the center of the retinal plane become simple translations along the *jv* and *ju* axes in the cortical plane, respectively.







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Reduct Image - Uniform RFs





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Control Image - Uniform RFs



radius. Cartical Image - Uniform RFs



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