

Computer Vision

- **Computer vision:** processing data from any modality that uses the electromagnetic spectrum which produces an image
- **Image:**
 - *A way of representing data in a picture-like format where there is a direct physical correspondence to the scene being imaged*
 - *Results in a 2D array or grid of readings*
 - *Every element in array maps onto a small region of space*
 - *Elements in image array are called pixels*
- **Modality** determines what image measures:
 - *Visible light → measures value of light (e.g. color or gray level)*
 - *Thermal → measures heat in the given region*
- **Image function:** converts signal into a pixel value



Pan-Tilt-Zoom camera



CMU Cam
(for color blob tracking)



Omnidirectional
Camera



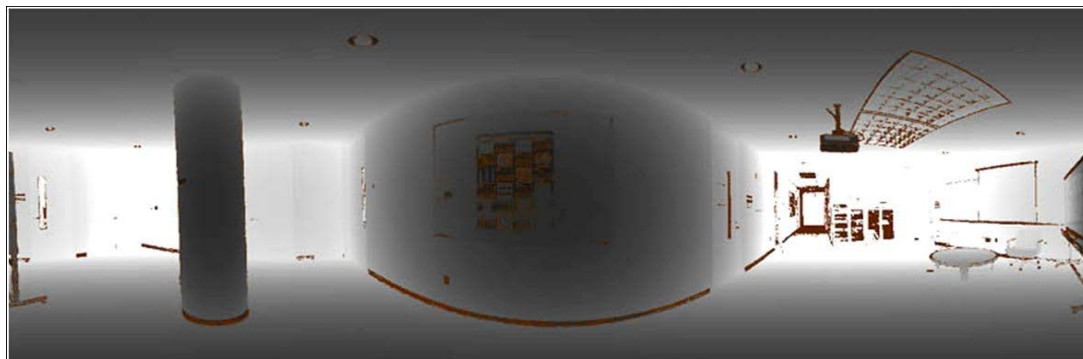
Stereo vision

Types of Computer Vision

- Computer vision includes:
 - *Cameras (produce images over same electromagnetic spectrum that humans see)*
 - *Thermal sensors*
 - *X-rays*
 - *Laser range finders*
 - *Synthetic aperture radar (SAR)*



Thermal image



3D Laser scanner image



SAR image (of U.S. capitol building)

Computer Vision is a Field of Study on its Own

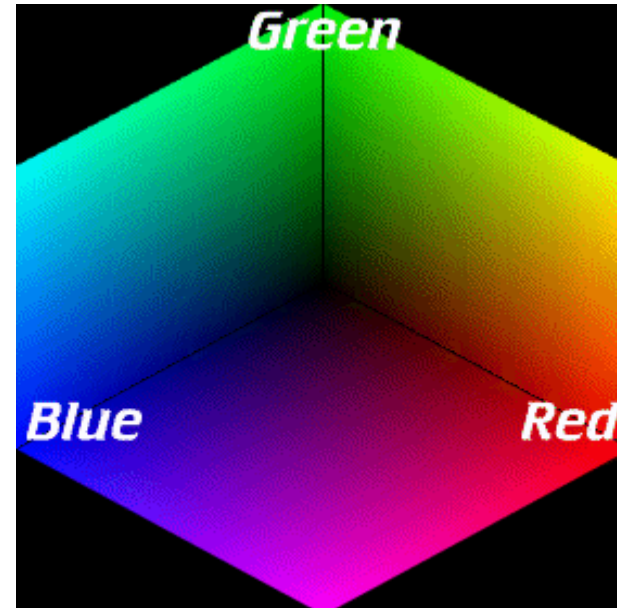
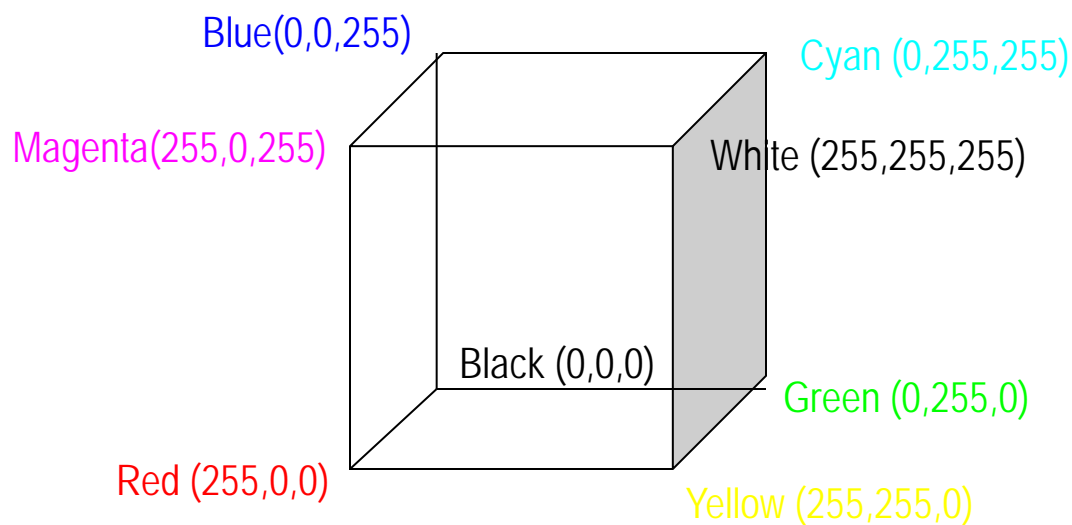
- Computer vision field has developed algorithms for:
 - *Noise filtering*
 - *Compensating for illumination problems*
 - *Enhancing images*
 - *Finding lines*
 - *Matching lines to models*
 - *Extracting shapes and building 3D representations*
- *However, autonomous mobile robots operating in dynamic environments must use computationally efficient algorithms; not all vision algorithms can operate in real-time*

CCD (Charge Coupled Device) Cameras

- **CCD technology:** Typically, computer vision on autonomous mobile robots is from a video camera, which uses CCD technology to detect visible light
- **Output of most cameras:** analog; therefore, must be digitized for computer use
- **Framegrabber:**
 - *Card that is used by the computer, which accepts an analog camera signal and outputs the digitized results*
 - *Can produce gray-scale or color digital image*
 - *Have become fairly cheap – color framegrabbers cost about \$200-\$500.*

Representation of Color

- Color measurements expressed as three color planes – red, green, blue (abbreviated RGB)
- RGB usually represented as axes of 3D cube, with values ranging from 0 to 255 for each axis



Software Representation

1. Interleaved: colors are stored together (most common representation)
 - *Order: usually red, then green, then blue*

Example code:

```
#define RED 0
#define GREEN 1
#define BLUE 2

int image[ROW][COLUMN][COLOR_PLANE];
...
red = image[row][col][RED];
green = image[row][col][GREEN];
blue = image[row][col][BLUE];
display_color(red, green, blue);
```

Software Representation (con't.)

2. Separate: colors are stored as 3 separate 2D arrays

Example code:

```
int  image_red[ROW][COLUMN];
int  image_green[ROW][COLUMN];
int  image_blue[ROW][COLUMN];

...

red = image_red[row][col];
green = image_green[row][col];
blue = image_blue[row][col];
display_color(red, green, blue);
```

Challenges Using RGB for Robotics

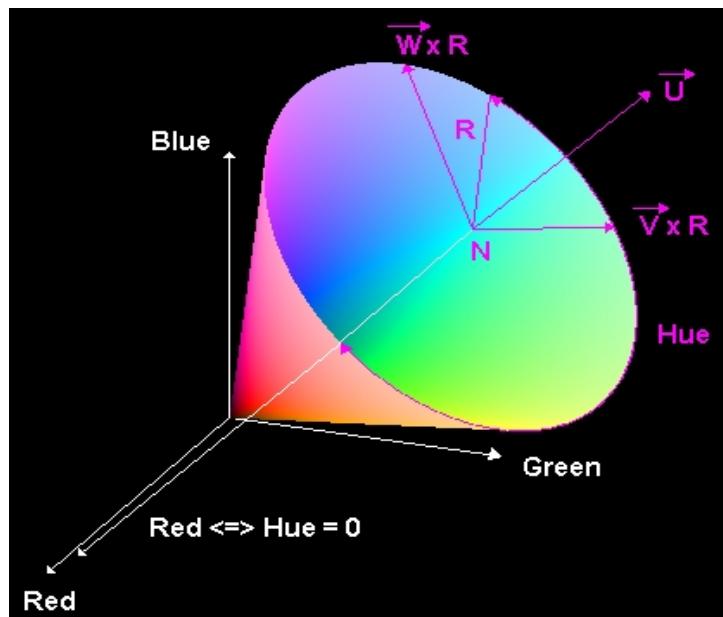
- Color is function of:
 - *Wavelength of light source*
 - *Surface reflectance*
 - *Sensitivity of sensor*
- → Color is not absolute;
 - *Object may appear to be at different color values at different distances to due intensity of reflected light*

Better: Device which is sensitive to absolute wavelength

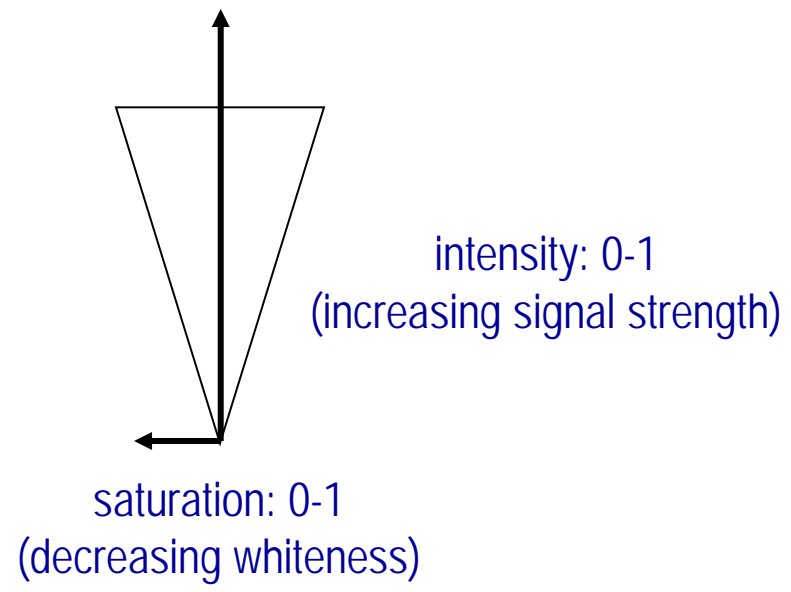
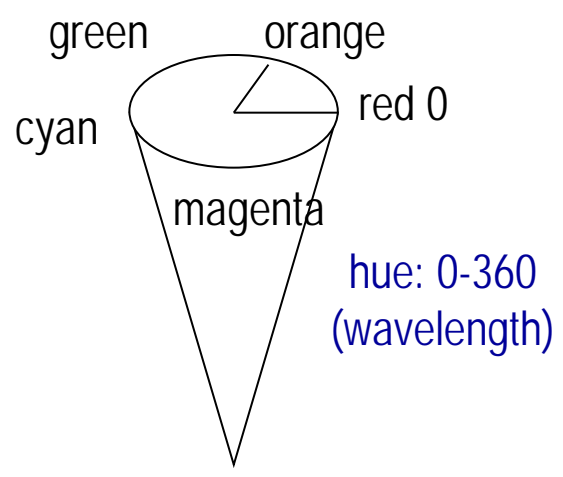
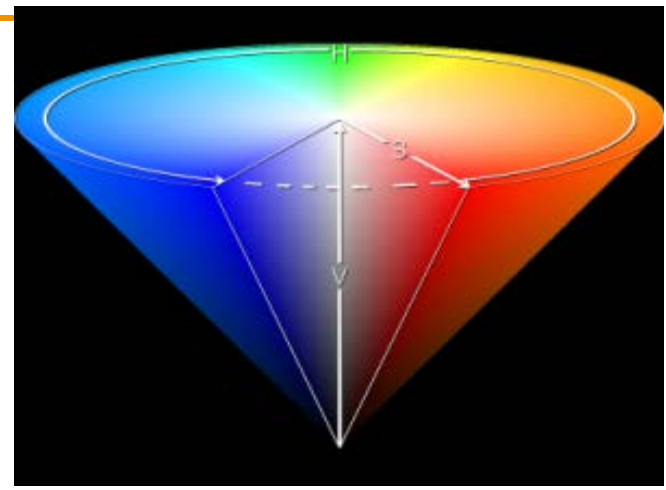
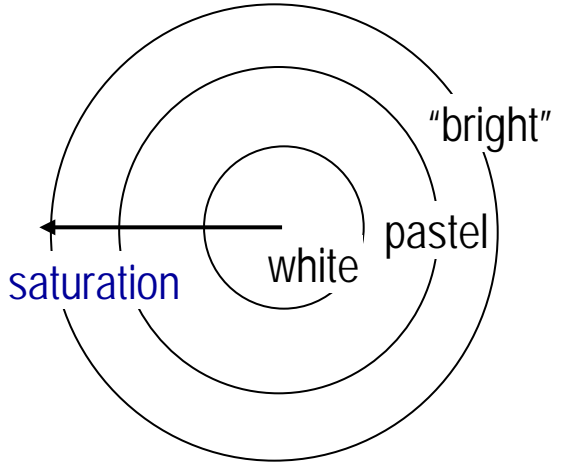
Better: Hue, saturation, intensity (or value) (HSV) representation of color

- **Hue:** dominant wavelength, does not change with robot's relative position or object's shape
- **Saturation:** lack of whiteness in the color (e.g., red is saturated, pink is less saturated)
- **Intensity/Value:** quantity of light received by the sensor

Transforming RGB to HSV



Representation of HSV

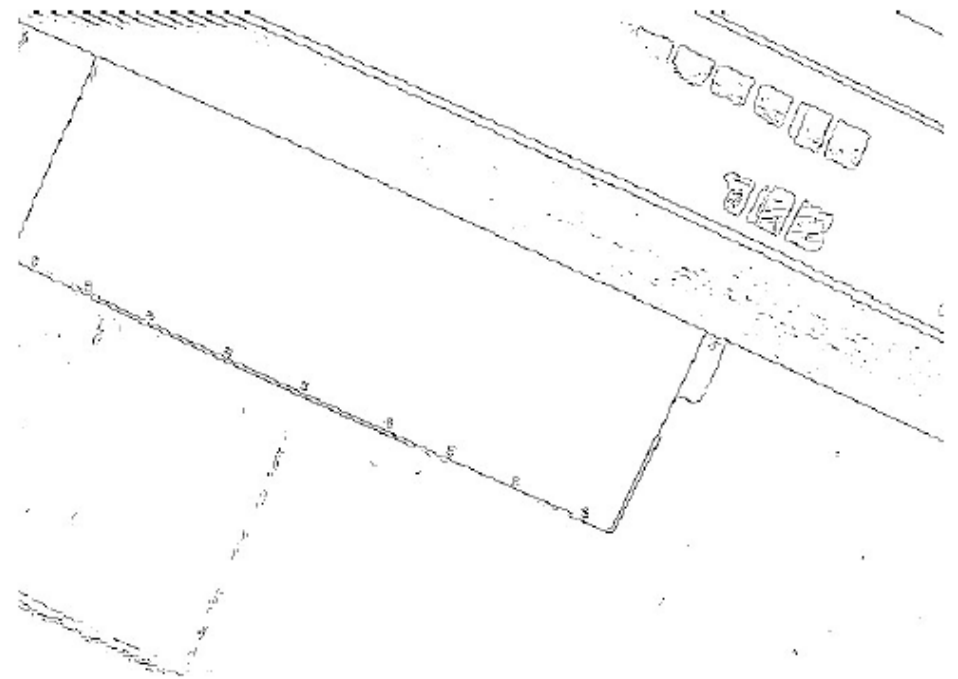
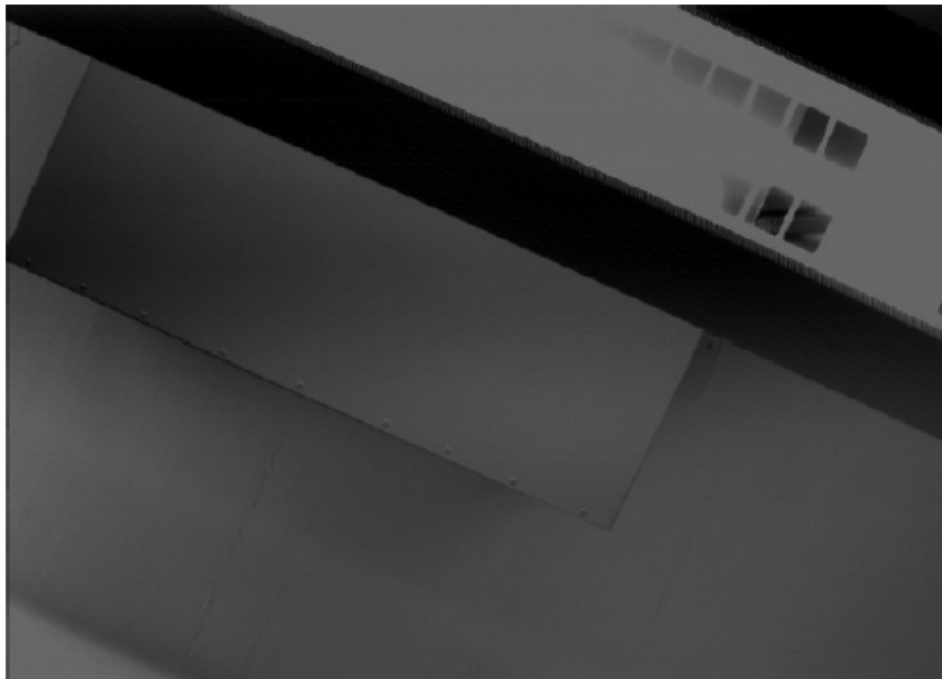


HSV Challenges for Robotics

- Requires special cameras and framegrabbers
- Expensive equipment
- Alternative: Use algorithm to convert -- Spherical Coordinate Transform (SCT)
 - *Transforms RGB data to a color space that more closely duplicates response of human eye*
 - *Used in biomedical imaging, but not widely used for robotics*
 - *Much more insensitive to lighting changes*

Edge Detection

- Ultimate goal of edge detection
 - *an idealized line drawing.*
- Edge contours in the image correspond to important scene contours.



Region Segmentation

- **Region Segmentation:** most common use of computer vision in robotics, with goal to identify region in image with a particular color
- Basic concept: identify all pixels in image which are part of the region, then navigate to the region's centroid
- Steps:
 - *Threshold all pixels which share same color (thresholding)*
 - *Group those together, throwing out any that don't seem to be in same area as majority of the pixels (region growing)*

Example Code for Region Segmentation

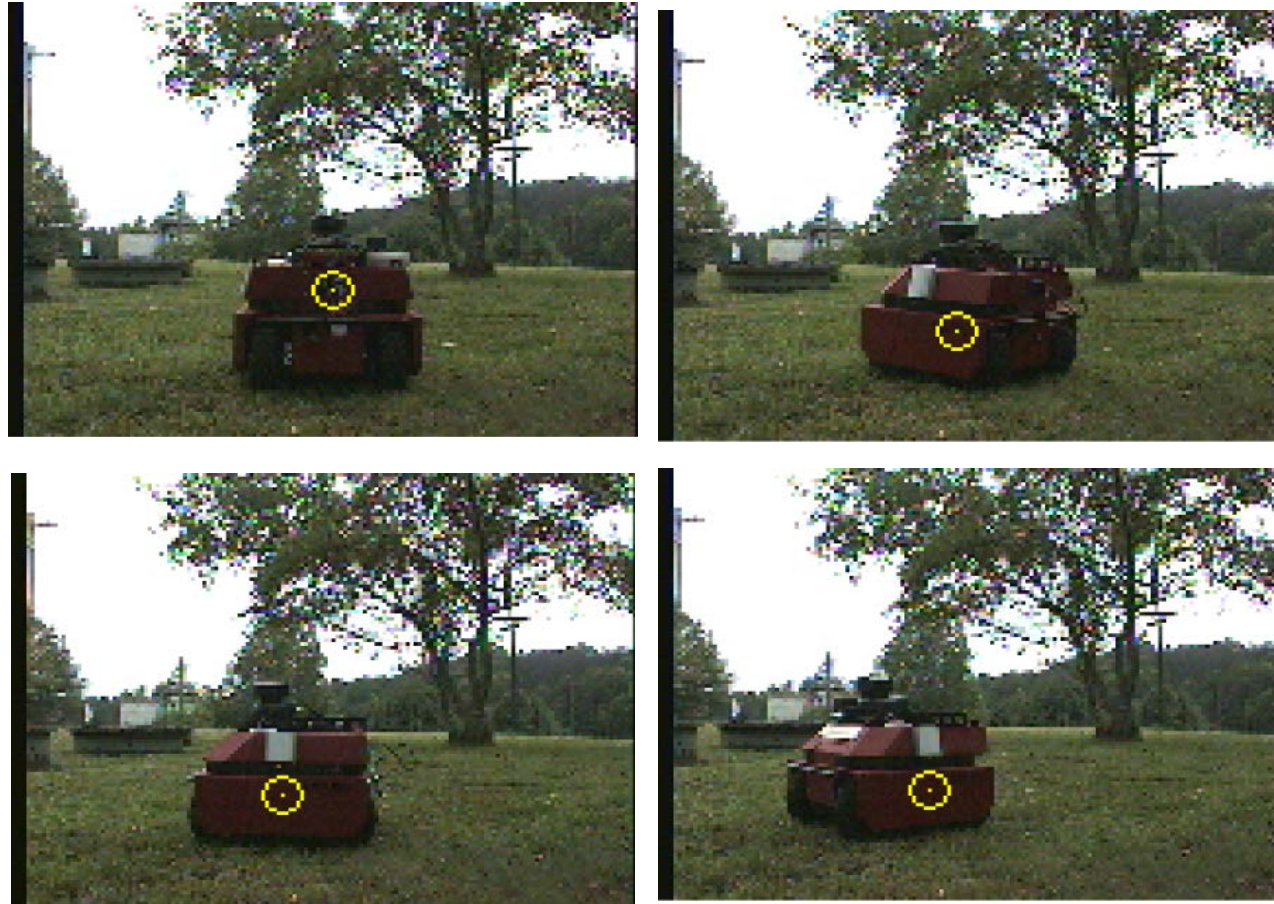
```
for (i=0; i<numberRows; i++)
  for (j=0; j<numberColumns; j++)
    { if (((ImageIn[i][j][RED] >= redValueLow)
          && (ImageIn[i][j][RED] <= redValueHigh))
        && ((ImageIn[i][j][GREEN] >= greenValueLow)
          && (ImageIn[i][j][GREEN] <= greenValueHigh))
        && ((ImageIn[i][j][BLUE] >= blueValueLow)
          && (ImageIn[i][j][BLUE] <= blueValueHigh)))
      ImageOUT[i][j] = 255;
    else
      ImageOut[i][j] = 0;
    }
```

Note range of readings required due to non-absolute color values

Example of Region-Based Robotic Tracking using Vision

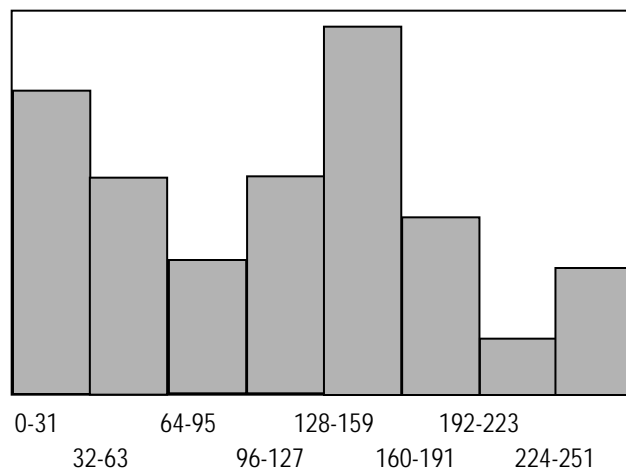


Another Example of Vision-Based Robot Detection Using Region Segmentation



Color Histogramming

- Color histogramming:
 - *Used to identify a region with several colors*
 - *Way of matching proportion of colors in a region*
- Histogram:
 - *Bar chart of data*
 - *User specifies range of values for each bar (called buckets)*
 - *Size of bar is number of data points whose value falls into the range for that bucket*
- Example:

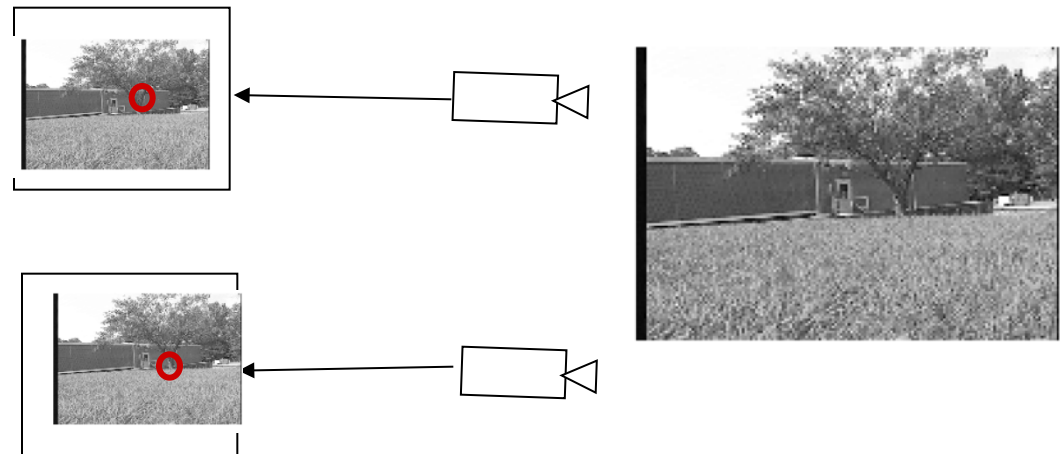


Color Histograms (con't.)

- Advantage for behavior-based/reactive robots: **Histogram Intersection**
 - *Color histograms can be subtracted from each other to determine if current image matches a previously constructed histogram*
 - *Subtract histograms bucket by bucket; difference indicates # of pixels that didn't match*
 - *Number of mismatched pixels divided by number of pixels in image gives percentage match = **Histogram Intersection***
- This is example of local, behavior-specific representation that can be directly extracted from environment

Range from Vision

- Perception of depth from stereo image pairs, or from optic flow
- Stereo camera pairs: range from stereo
- Key challenge: how does a robot know it is looking at the same point in two images?
 - This is the *correspondence problem*.



Simplified Approach for Stereo Vision

- Given scene and two images
- Find interest points in one image
- Compute matching between images (**correspondence**)
- Distance between points of interest in image is called **disparity**
- Distance of point from the cameras is inversely proportional to disparity
- Use **triangulation and standard geometry** to compute depth map

- Issue: **camera calibration**: need known information on relative alignment between cameras for stereo vision to work properly