

Computer Vision Feature Detectors

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What is a feature?

It's interesting.

-Wikipedia puts it well

(http://en.wikipedia.org/wiki/Feature_detection_%28computer_vision%29):

-“There is no universal or exact definition of what constitutes a feature, and the exact definition often depends on the problem or the type of application. Given that, a feature is defined as an "interesting" part of an [image](#), and features are used as a starting point for many computer vision algorithms.”

For example...

Edges

Blobs

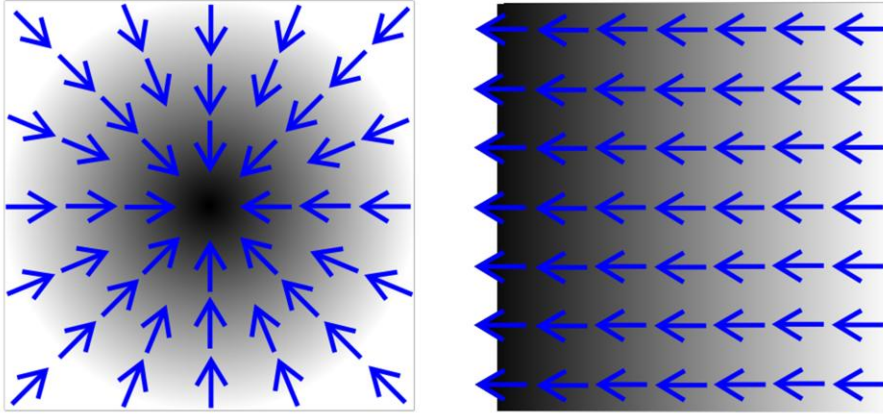
Corners

Ridges

-Edges:

- Boundary between two image regions
- Points with a strong gradient magnitude

Image Gradient



-The gradient of a single point in the image is a 2D vector whose components are given by the derivatives in the horizontal and vertical directions

-Image from Wikipedia

-<http://en.wikipedia.org/wiki/File:Gradient2.svg>



- We find an approximate/discrete version of the derivate of the image intensity
- Use a kernel convolution over the image (for example, the Sobel operator)
- If this is the before image...

-Images from Wikipedia

-<http://en.wikipedia.org/wiki/File:Bikesgray.jpg>



- ...this is the normalized Sobel gradient image
- The whitest values can be taken as edges

-Image from Wikipedia:

-<http://en.wikipedia.org/wiki/File:Bikesgraysobel.jpg>

For example...

Edges

Blobs

Corners

Ridges

-Corners:

- Term used interchangeably with 'interest points'
- Points detected aren't always on corners as we understand them
- Harris corners, shown in previous talk, a good example of 'true' corners



- Example of one interest point detector is SIFT
- First step is Difference of Gaussians (DoG), an image enhancement algorithm that works on greyscale images
 - (so above image would be converted to greyscale first)
- Convolve the image with two Gaussian kernels and subtract them....
- Image:
 - http://en.wikipedia.org/wiki/File:Flowers_before_difference_of_gaussians.jpg

Difference of Gaussians



-....to get this result

-For SIFT, various DoG results obtained over several successive scales

-Local minima/maxima of these results that occur at several scales become the keypoints

-Further processing is done, but SIFT will be discussed in the next talk so we'll leave it at that

-Image:

-http://en.wikipedia.org/wiki/File:Flowers_after_difference_of_gaussians_grayscale.jpg

For example...

Edges Blobs
Corners Ridges

-Blob detection

- We look for regions (or points) that are darker or brighter than the surrounding regions
- Closely related to corners, since they often contain a preferred point
- Some methods in corner detection, including DoG, are also used here



-Maximally Stable Extremal Regions (MSER)

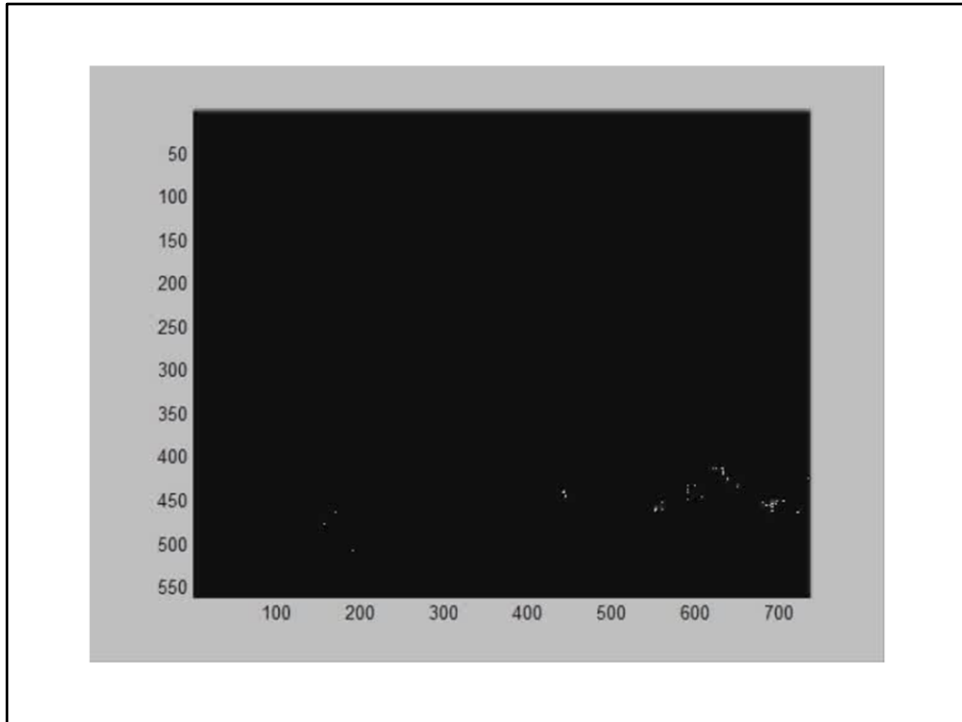
-An example of a blob detector that addresses the issue of changing viewpoints

-Imagine your image is actually a topographic map

-In other words, you are looking at a terrain from a bird's eye view

-The image intensities represent how high a point on the terrain is (with, say, white being the highest)

-Now imagine filling up the terrain with water...



- ...the lowest regions fill with water first
- Some regions keep the same shape for some time
- It is these regions that will become MSERs
- (Note: The above is a movie that you may not be able to view online. Basically, it was drawing all pixels with 0 intensity white in the first frame, adding those with intensity 1 in the next frame, intensity 2 in the next, and so on, until the whole image is white.)



-The MSER regions found in the detection process are all superimposed in this image as white regions



- Here's the same example with the bounding ellipses of the MSER regions shown in blue, and centers of gravity of the regions shown as red crosses

For example...

Edges Blobs
Corners Ridges

-Ridges

-“From a practical viewpoint, a ridge can be thought of as a one-dimensional curve that represents an axis of symmetry, and in addition has an attribute of local ridge width associated with each ridge point.” (Wikipedia)

-Harder to compute

-Not used a lot, but can be seen in road and blood vessel detection

Now what?

Feature description.

- Once you have detected your features, you need a unique way to describe them.
- Then you can match them between images, use them to build 3D scenes, find similar objects from a database, etc...
- But that's another story. 😊