Charlie's notes

December 1, 2015

- 1 CS31310 AGILE
- 2 CS36110 Machine Learning
- 3 CS34110 COMPUTER VISION
- 3.1 November 20: Motion Models

Modelling Change & Tracking

MOTION:

- Background Subtraction
- Optical Flow

MIXTURE OF GAUSSIANS (MoG):

- Robust to noise
- Handles shadows ok
- Common first step

3.1.1 Tracking: Modelling Change

VIDEO:

- detections in each frame
- \bullet detections are noisy & computationally expensive
- ullet tracking mitigates both issues

Noise can occur if the camera on a robot/car is moving up/down

3.1.2 A GENERAL FRAMEWORK FOR TRACKING

RECURSIVELY:

- An idea about how something will change (Model)
- Make a prediction (Predict)
- See what happens (Measure)
- Update model (*Update*)

Advantages:

- Smooths the data
 - estimate location upon predictions & the measurement
- Constrains search
 - start looking for target in the location it was last seen

3.1.3 Kalman Filter

- Like predict, measure, update from earlier
- Useful for tracking
- Copes well with missing information (occlusions)

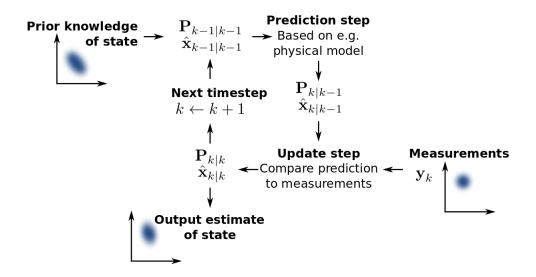


Figure 3.1: Sourced from Wikipedia

| Background subtraction | \rightarrow | Pixels grouped into objects | \rightarrow | |
|------------------------|---------------|-------------------------------|---------------|---------|
| Sparse Optical Flow | \rightarrow | Features grouped into objects | \rightarrow | Tracker |
| Face Detection | \rightarrow | \rightarrow | \rightarrow | |

USE KALMAN TO SMOOTH ANY MEASUREMENT

- X,Y location
- size
- colour

See also: Particle filtering: works with combining and splitting objects (e.g. people holding hands, then letting go)

Hannah's video: https://www.youtube.com/watch?v=NYdwpX1a7-Y

3.1.4 Mean Shift

Computer the mean of the data within the window

Shift the window to the mean every time

Notes:

- Changes size can use CAM-Shift to mitigate?
- Lighting change not really, gradually changes mean over time
- If it picks up something you're not looking for, will slowly drift off

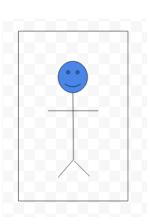
3.1.5 Problems with Tracking

- Initialisation (what are you tracking?)
- Having more than 1 item to track
- Losing target due to motion / occlusion
- Losing target due to appearance change

Usually initialise from a detector of some sort Useful speed up for detectors & accuracy Look into: TLD: Tracking Learning Description

3.1.6 Closing Note

Vision Systems tend to have multiple layers Tracking is extremely common in anything which deals with change.



HOG FOR EXAMPLE, HAS SEVERAL LAYERS:

- 2D filters
- Tangent
- Histogram
- Superimpose grid
- \bullet SVM

May be more, however couldn't write quick enough... You get the idea...

3.2 November 24: Shape from Shading

3.2.1 Shape

3D STRUCTURE OF:

- Object
- Scenes

WHY?

- Science (surface of Mars)
- Graphical (3D model of heritage)

There is a table in the slides of the topics that will be covered in the 'Shape' series

3.2.2 Shape from Shading

Brightness depends on:

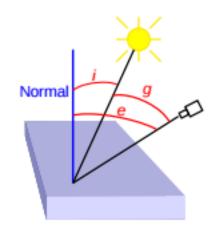
- where light source is
- where viewer is
- local orientation of surface
- properties of surface (matt vs glossy)

зеотнесту.

• i: incident angle

e: emittance angle

g: phase angle



BRDF:

• Bidirectional Reflectance Distribution Function

• fraction of light reflected in the direction of the viewer

• Lambertian: $\Phi(i, e, g) = \cos i$

• Specular: $\Phi(i, e, g) = 1$ when i = eand g = i + e

• Largely property of the surface

• Used in graphics a lot

• Difficult to establish wavelength of light & properties of surface

• Determined experimentally

3D Surface information Look up - 'Normal Mapping' wikipedia page Orientation of surface described by a normal vector (x,y,z):

• Change in z when x changes (p)

• Change in z when y changes (q)

3.2.3 Gradient Space

$$p = \frac{\delta z}{\delta x}$$
$$q = \frac{\delta z}{\delta y}$$

REFLECTANCE MAP:

• For surface of 'x' orientation, we expect this reflection back

Light Change: 1 = lightest, 0 = darkest

- Global solution found by integration in gradient and image space + smoothness assumption:
 - Maths in confusing!
 - Don't need to know the maths for the exam

3.2.4 Photometric Stereo

We can reduce assumptions if we have multiple lighting conditions. Example: 2 overlapping reflectance maps on sphere

- Use same point in image plane under different lighting
- You know where the lights are (calibrated)
- Nothing moves just the lights on/off
- Calibration with sphere of similar reflectance
- There is always noise
- Reasonably robust

This is at the heart of light-stage imaging

- Illuminate with 1 source at a time
- Calibrate & results in look-up table

Have to reduce inter-reflections!

X,Y,Z MAPPED TO RGB FIGURES:

• gives diagram of 3D structure

Fewer images \rightarrow More assumptions

* Look up ISO-contours

3.3 NOVEMBER 27: SHAPE FROM X

3.3.1 Shape from Texture

1 + images, with a lot of assumptions.

Marr 1982:

- ullet 2 textured gradients
- Limited information
- Circles turn into ellipses (illusion of depth)

Types of textures:

Isotropic: rotation perpendicular to texture plane does not 'change' the texture

Homogeneous: texture looks 'the same' everywhere

Texel: basic texture elements, the repetition of which creates the texture

LOCAL BINARY PATTERNS

- 3x3 filters
- Capture changes in greyscale

| 1 | 1 | 0 |
|---|---|---|
| 0 | Т | 1 |
| 0 | 1 | 0 |

| 0 | 1 | 1 |
|---|---|---|
| 0 | Т | 1 |
| 0 | 0 | 1 |

Figure 3.2: Top: Standard output. Bottom: Could be used to detect edges

0 = darker than T 1 = lighter than T

GLOBAL PLANES

- Find the plane, uphill, downhill etc
- Gradients in texture
- Isotropic texture:
 - circles \rightarrow ellipses

- Major axes of ellipses to work out slant & tilt
- Homogeneous texture:
 - textures made of dots
 - look for a density change

Generally you fit methods locally rather than globally

3.3.2 Shape from Motion

Look up Marr (again)

Can also do structure from Motion: this is next week!

The examples on BB are really good: biomotionlab!

Things further away move more slowly - can infer things from this

Assumes:

- rigidity
- parallel projection

Works both long and short range...

- Initial guess on 3D model of world (flat screen)
- \bullet + new image
- update on those which disagree with the initial model

Very popular!

3.3.3 Shape from Occlusions

Infer shapes from which bits are in front of other parts

OCCLUSION CONTOURS:

- discontinuity in depth signal
- corresponds to silhouette (shadow)

Assumptions (1 image):

- each point on contour corresponds to a single point on the object (except when a 'T' junction)
- nearby points on the contour correspond to nearby points on the object
- points on contour correspond to planar points on the object

When you have 2+ images:

- Space Carving
- Volumetric reconstruction
- Gives where the object is isn't
- Object is intersection of several generalised cones

Watch the Youtube video on BB - Greenvision

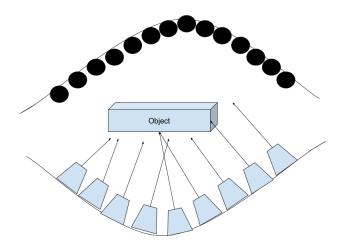


Figure 3.3: Can the cameras see the points on the other side of the object? Make a map of what it can / can't see to build a shape of the object using multiple cameras (bottom row).

FROM A VIDEO SEQUENCE:

- inferred from motion & occlusion
- what occludes people in the scene?

Things at the front = black, as it's never occluded from you Lots of motion = white

3.3.4 Shape from focal length

Depth from defocus

Things are blurry past a depth of focus Take 10 pictures, at different lengths, with different focuses

- Texture really helps focus
- Assumptions about smoothness

Markov-Random Field

How to tell if in focus?

- • Only low-frequency components = $\pmb{\varkappa}$
 - 4 CS32310 ADVANCED COMPUTER GRAPHICS
 - 5 SE31520 Internet-based Applications
 - 6 OTHER