Leow Wee Kheng CS4243 Computer Vision and Pattern Recognition

Background Removal

Here's an image...

• We often just want the eagle



Background Removal

- Related to tracking and segmentation
 - Tracking
 - Tracks location of moving object in video.
 - Segmentation
 - Separate object and background in single image.
 - Background removal
 - Separate object and background given > 1 image.

Background Removal

- Two general approaches:
 - With known background, also called clean plate.

• Without known background.

With Clean Plate

• Clean plate: background only image



• Subtract clean plate *P* from image *I*

$$D(x, y) = |I(x, y) - P(x, y)|$$

absolute difference

Colour image has 3 components
 R: red, G: green, B: blue
 So, get 3 sets of differences

$$D_R(x, y) = |I_R(x, y) - P_R(x, y)|$$
$$D_G(x, y) = |I_G(x, y) - P_G(x, y)|$$
$$D_B(x, y) = |I_B(x, y) - P_B(x, y)|$$

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• Combine 3 sets of differences into 1 set

$$D(x, y) = \alpha_R D_R(x, y) + \alpha_G D_G(x, y) + \alpha_B D_B(x, y)$$

 $\circ \alpha_R$, α_G , α_B are constant weights.

• Usually, $\alpha_R + \alpha_G + \alpha_B = 1$.

• In the case of equal weights, $\alpha_R = \alpha_G = \alpha_B = 1/3$.

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• Finally, fill in foreground object colour

$$F(x, y) = \begin{cases} I(x, y) & \text{if } D(x, y) > \Gamma \\ B & \text{otherwise} \end{cases}$$

 $\circ \Gamma$ is threshold.

O If $D(x, y) > \Gamma$, pixel at (x, y) is foreground pixel.

○ *B* is constant background colour, e.g., black.

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Notice

 Some parts of the eagle's tail are missing. Why?

Dynamic Clean Plate

- Stationary camera
 - Stationary background.
 - O Need only one image as clean plate.
- Moving camera
 - Moving background.
 - Need a video clean plate.
 - With motion-controlled camera, controlled lighting
 - Shoot clean plate video.
 - Shoot target video with same camera motion.

Remove background with corresponding clean plate.

clean plate



scene video



background removed



Without Clean Plate

- Background removal without clean plate is more difficult.
- Possible if moving objects do not occupy the same position all the time.
- 3 cases
 - Stationary camera, fixed lighting.
 - Stationary camera, varying lighting.
 - Moving camera.

Stationary Camera, Fixed Lighting

• Consider these video frames:



Moving object occupies a small area.
Moving object does not occupy the same position.
What if we average the video frames?

Averaging

Mean of video frame

$$M(x, y) = \frac{1}{n} \sum_{i} I_i(x, y)$$

- *i* : frame number
- \circ *n* : number of frames

Notes:

• The above direct formula can lead to overflow error.

• Refer to colour.pdf for a better formula.

Case 1: average over whole video



Averaging gives mostly background colours. Some faint foreground colours remain.

Case 2: average over first 3 seconds



Foreground colours are more localised in one region.
 Foreground colours are stronger.

Subtract background from video frame



Case 1



Copy foreground colours to foreground pixels



Case 1

Case 2

Background colours are removed: true rejection.
 Some foreground colours are missing: false rejection.

Use lower thresholds



Case 1

Case 2

More foreground colours are found: true acceptance.
 Background colours are also found: false acceptance.

Another example



Averaging video frames



Case 1: over whole video



Case 2: over first 3 seconds

Subtract background from video frame



Case 1

Case 2

Copy foreground colours to foreground pixels



Case 1

Case 2

Background colours are removed: true rejection.
 Some foreground colours are missing: false rejection.

Use lower thresholds



Case 1

Case 2

More foreground colours are found: true acceptance.
 Background colours are also found: false acceptance.

Background Modelling

- Averaging is simple and fast but not perfect.
- Better than average: colour distribution.
 For each pixel location, compute distribution of colours over whole video.

• For a background pixel:



Single cluster of colours (due to random variation).
Peak: most frequent colour.

• For a pixel that is background most of the time:



k-means clustering

- A method for grouping data points into clusters.
- Represent each cluster C_i by a cluster centre \mathbf{w}_i .
- Repeatedly distribute data points and update cluster centres.

k-means clustering

- 1. Choose k initial cluster centres $\mathbf{w}_1(0), \dots, \mathbf{w}_k(0)$.
- 2. Repeat until convergence
 - O Distribute each colour **x** to the nearest cluster $C_i(t)$

 $\mathbf{x} \in C_i(t)$ if $\|\mathbf{x} - \mathbf{w}_i\| < \|\mathbf{x} - \mathbf{w}_j\| \quad \forall j \neq i$

Update cluster centres:
 Compute mean of colours in cluster

t is iteration number

$$\mathbf{w}_i(t+1) = \frac{1}{|C_i(t)|} \sum_{\mathbf{x} \in C_i(t)} \mathbf{x}$$

- For background removal, can choose k = 2
 One for foreground, one for background.
- Initial cluster centres
 - Get from foreground and background in video.
- Possible termination criteria
 - Very few colours change clusters.
 - Fixed number of iterations.
- After running clustering
 - If foreground area is small, then smaller cluster is foreground.

Background removed



Most background colours are removed.

- A bit of shadow remains.
- Most foreground colours are found.

Stationary Camera, Varying Lighting

• Basic ideas

- Multiple background clusters for different lighting conditions.
- Apply *k*-means clustering with k > 2.

Example from [Stauffer98]





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Moving Camera

Basic ideas

- Track and recover camera motion [Bergen92].
- Stabilise video by removing camera motion [Matsushita05].
- Do stationary camera background removal.

• Put back camera motion.

Summary

- With clean plate
 - Subtract clean plate from video frames.
- Without clean plate
 - Estimate background
 - Average video frame
 - Cluster pixel colours

Subtract estimated background from video frames.

Moving camera

• Stabilise video, then perform background removal.

Further Reading

- Code book method
 OpenCV [Bradski08] chapter 9.
- Varying lighting condition
 O [Stauffer98]
- Motion estimation
 - O [Bergen92]
- Video stabilization
 - O [Matsushita05]

References

- G. Bradski and A. Kaebler, *Learning OpenCV*, O'Reilly, 2008.
- J. R. Bergen, P. Anandan, K. J. Hanna, and R. Hingorani. Hierarchical model-based motion estimation. In *Proc. ECCV*, pages 237–252, 1992.
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