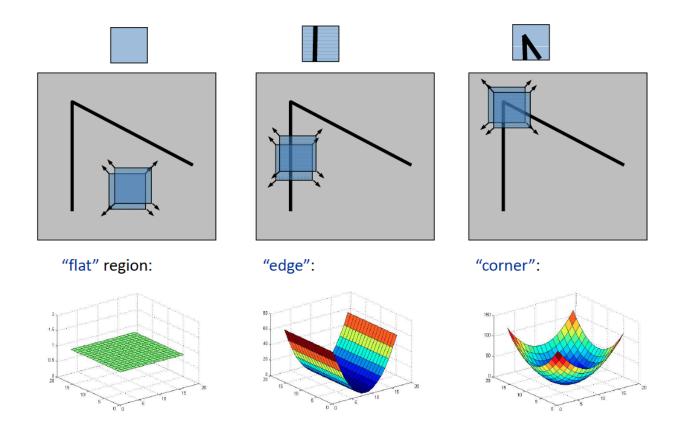


# COMP 9517 Computer Vision

**Feature Detection** 

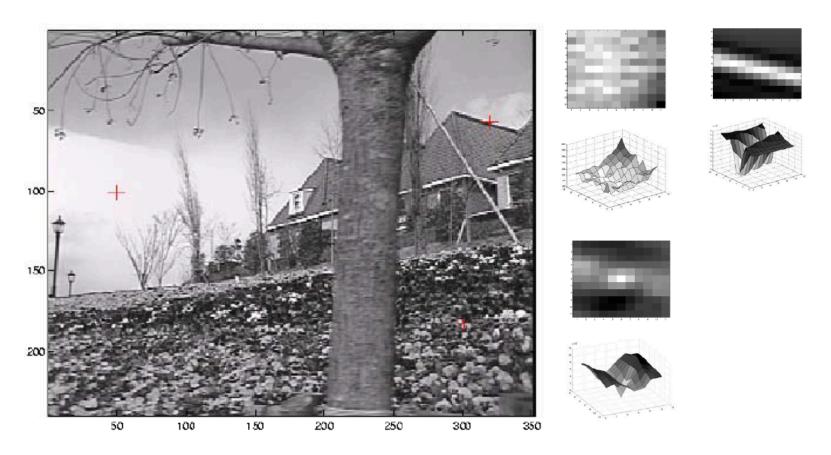
## Properties of Feature Points

What are good feature points?

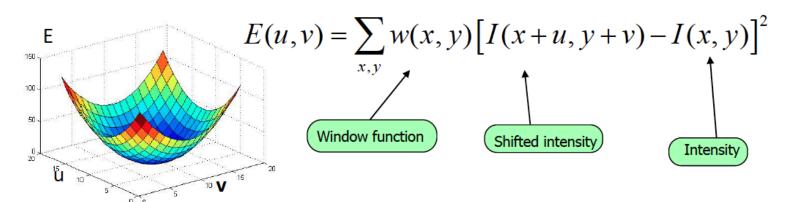


## **Properties of Feature Points**

What are good feature points?



- "interestingness" ≈ High change of intensity when shift
- Change of intensity for the shift [u, v]:



Window function W(x,y) = 01 in window, 0 outside Gaussian

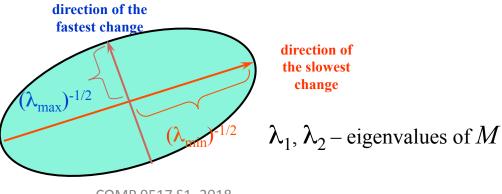
• For small shifts [u, v] we have a bilinear approximation:

$$E(u,v) \cong \begin{bmatrix} u,v \end{bmatrix} \ M \quad \begin{bmatrix} u \\ v \end{bmatrix}$$

where M is a 2×2 matrix computed from image derivatives

$$M = \sum_{x,y} w(x,y) \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix}$$

Eigen analysis



21/3/18

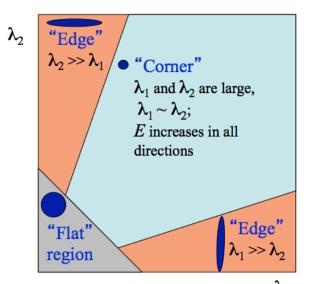
Measure of corner response

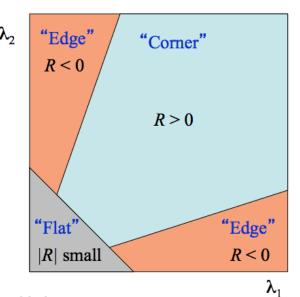
$$R = \det M - k \left( \operatorname{trace} M \right)^2$$

$$\det M = \lambda_1 \lambda_2$$

$$\operatorname{trace} M = \lambda_1 + \lambda_2$$

(k - empirical constant, k = 0.04-0.06)





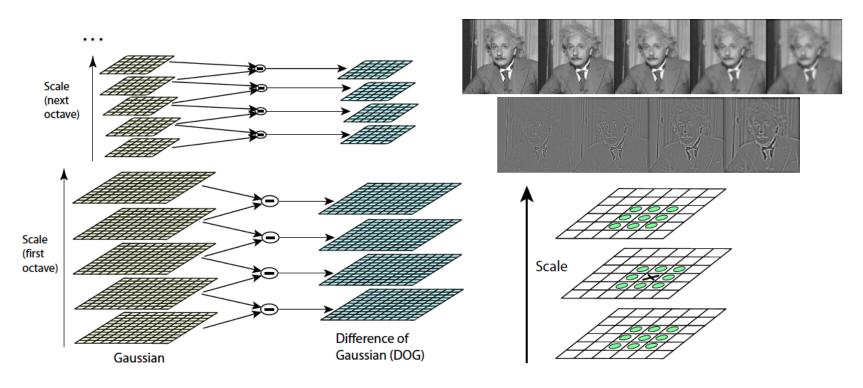
**ሊ** COMP 9517 S1, 2018

#### The algorithm

- 1. Compute the horizontal and vertical derivatives of the image  $I_x$  and  $I_y$  by convolving the original image with derivatives of Gaussians (Section 3.2.3).
- 2. Compute the three images corresponding to the outer products of these gradients. (The matrix A is symmetric, so only three entries are needed.)
- 3. Convolve each of these images with a larger Gaussian.
- 4. Compute a scalar interest measure using one of the formulas discussed above.
- 5. Find local maxima above a certain threshold and report them as detected feature point locations.

### Scale Invariant Detectors

Scale Invariant Feature Transform – SIFT



 Extremes (maxima & minima) in the resulting 3D volume are detected by comparing a pixel to its 26 neighbours

### References and Acknowledgements

- Szeliski, Chapter 4
- Some content are extracted from the above resource and James Hays slides