Speeded-Up Robust Features (SURF)

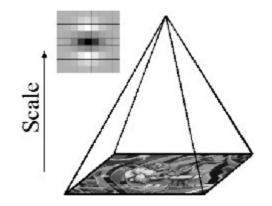
- Expected Properties of good Feature Point Descriptor
 - Repeatability
 - Distinctive to different objects
 - Speed -computation time to detect and describe, and also to match
 - Robust to
 - Noise
 - Geometric deformations (Skew, anisotropic scaling, and perspective)
 - Photometric deformations
- Drawback of SIFT:
 - High dimensionality of the descriptor
 - Time taken for convolution

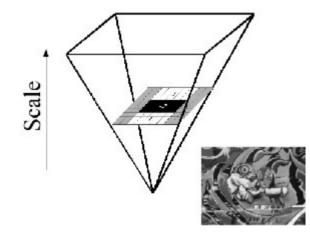
Introduction to Speeded-Up Robust Features

- (Bay et al. ECCV, 2006)
- Box filters
- Filtering is done using integral images to speed up the computation.
- Use of Hessian operator for key point detection __ Local maxima of det(H).
- Haar wavelets are used to find gradient.

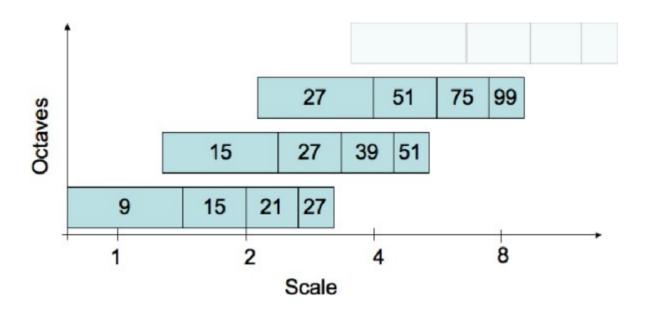
Steps in Feature Descriptor

- Key point detection
 - Extrema in Scale space-Similar to SIFT
- Feature Description
- Feature Matching

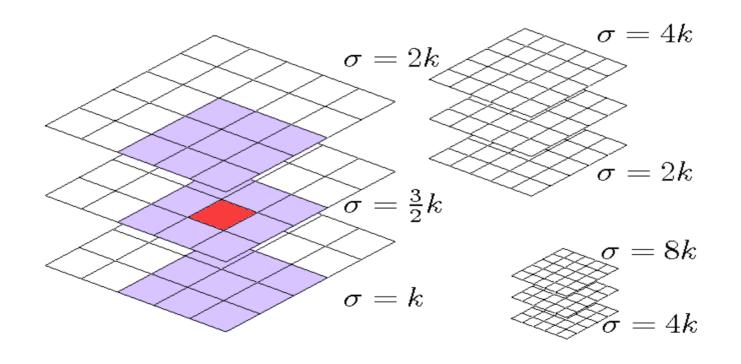




Scale Space Organization



Key point detection in scale space



Hessian Operator

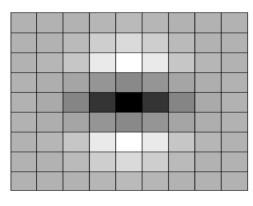
Convolution with the Gaussian second order derivative with image.

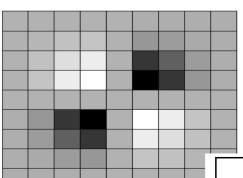
$$\mathcal{H}(\mathbf{x},\,\sigma) = \begin{bmatrix} L_{xx}(\mathbf{x},\,\sigma) & L_{xy}(\mathbf{x},\,\sigma) \\ L_{xy}(\mathbf{x},\,\sigma) & L_{yy}(\mathbf{x},\,\sigma) \end{bmatrix},$$

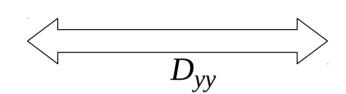
where $L_{xx}(\mathbf{x}, \sigma)$ is the convolution of the Gaussian second order derivative $\frac{\partial^2}{\partial x^2}g(\sigma)$ with the image I in point \mathbf{x} , and similarly for $L_{xy}(\mathbf{x}, \sigma)$ and $L_{yy}(\mathbf{x}, \sigma)$.

Keypoint: Maximum of det(H(.)) over space and scale.

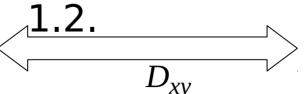
Approximation of Gaussian by Box filters

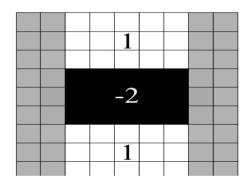


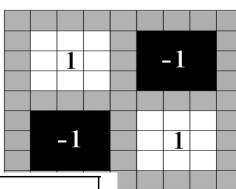




9x9 Box-filters are approximation of Gaussian width





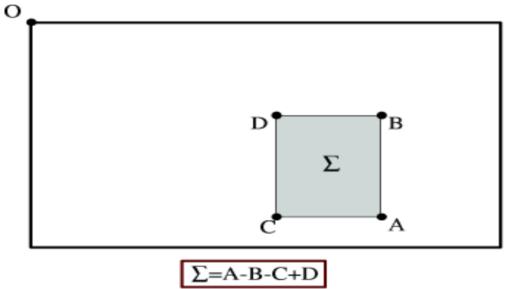


 $\det(\mathcal{H}_{\text{approx}}) = D_{xx}D_{yy} - (wD_{xy})^2.$

 $w \sim 0.9$

Fast computation using the integral image

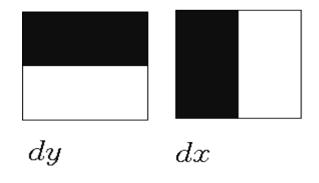
$$I_{\Sigma}(\mathbf{x}) = \sum_{i=0}^{i \leq x} \sum_{j=0}^{j \leq y} I(i, j)$$



Only 3 add/sub and four memory access.

Haar Filter Responses

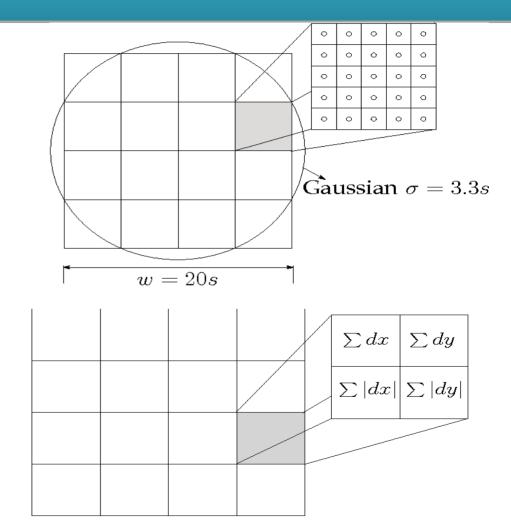
- Dominant orientation
 - The longest vector provides the dominant direction.
- Haar Filters -Box filter implementation.



6 operations needed for computing each filter response using integral image.

SURF Desc

- Partitioned into 4x4 square sub-regions.
- Haar wavelet responses at regularly spaced
 5x5 sample patches in each sub-region.
- Each sub-region has 4D vector.
- Concatenate them to 64 D vector.



Matching

- Representation of a key-point by a feature vec
 - e.g. $[f_0 f_1 ... f_n]^T$
- Use distance functions / similarity measures.
 - L₁ norm

orm
$$L_1(\vec{f}, \vec{g}) = \sum_{i=0}^n |f_i - g_i|$$

- $L_2 \text{ norm}$ $L_2(\vec{f}, \vec{g}) = \left(\sum_{i=0}^n |f_i g_i|^2\right)^{\frac{1}{2}}$
- L_p norm $L_p(\vec{f}, \vec{g}) = \left(\sum_{i=0}^{n} |f_i g_i|^p\right)^{\frac{1}{p}}$