# Analysis of Image Tranforms for Sketch-based Retrieval

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Introduction and Background

#### Introduction and Background

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### Motivation



# Challenges of CBIR

#### The Semantic Gap

"The semantic gap is the **lack of coincidence** between the information that one can extract from the **visual data** and the **interpretation** that the same data have for a user in a given situation." – Smeulders et al.

#### The Sensory Gap

"The sensory gap is the gap between the **object in the** world and the information in a (computational) description derived from a **recording of that scene**." – Smeulders et al.



Introduction and Background

### Prior Work on Human Recognition

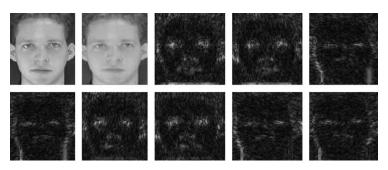


Figure: "Face recognition using curvelet based PCA.", T. Mandal and Q. M.J Wu, ICPR 2008

### Prior Work on Human Recognition

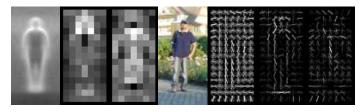
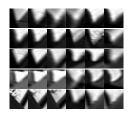


Figure: "Histograms of oriented gradients for human detection", Dalal and Triggs, CVPR 2005

#### Prior Work on Visual Codebooks







Results

Figure: "Video Google: A text retrieval approach to object matching in videos", Sivic and Zisserman, ICCV 2003



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#### Prior Work on Scene Classification

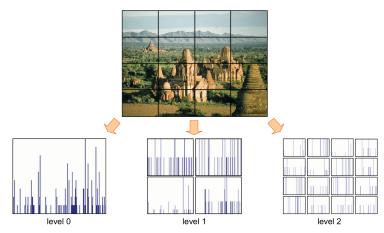


Figure: "Spatial pyramid matching", Lazebnik et al., 2009

# Anatomy of a CBIR System

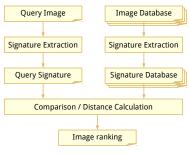


Figure: Global Descriptors

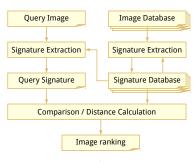


Figure: Local Descriptors

# Proposed Retrieval Pipelines (Global)



# Proposed Retrieval Pipelines (Local)



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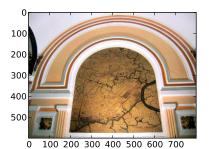


Figure: Original Image



Figure: Luma Conversion

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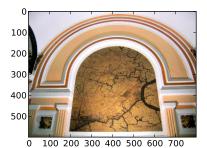


Figure: Original Image

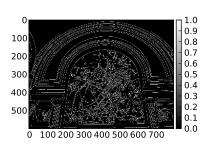


Figure: Canny Operator

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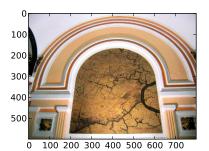


Figure: Original Image

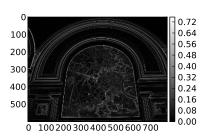


Figure: Sobel Operator

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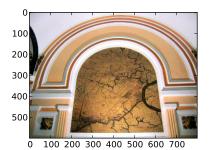


Figure: Original Image

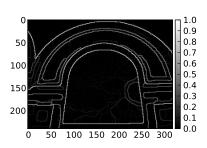


Figure: gPb-owt-ucm Transform

### Properties of the Curvelet Transform

- ► An extension of the wavelet transform
- Localized in position, scale and orientation
- ► Curvelets obey parabolic scaling:  $width \approx length^2$
- ▶ Approximation error along edges using m largest coefficients decays with  $\frac{log(m)^3}{m^2}$  (compare  $\frac{1}{m}$  for wavelets)
- Defined in frequency domain using



### Constructing the Curvelets

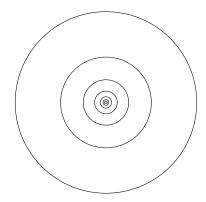


Figure: Frequency Domain

Figure: Spatial Domain



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### Constructing the Curvelets

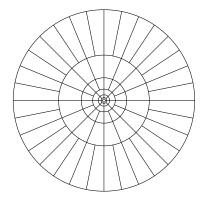


Figure: Frequency Domain

Figure: Spatial Domain



### Constructing the Curvelets

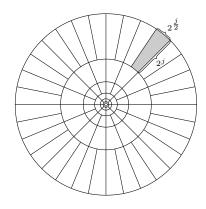


Figure: Frequency Domain

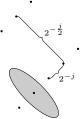


Figure: Spatial Domain

# Constructing the Curvelets

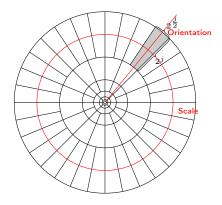


Figure: Frequency Domain

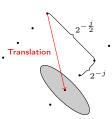


Figure: Spatial Domain

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# **Example Curvelets**

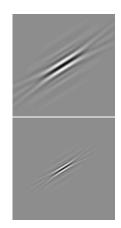


Figure: Frequency Domain

Figure: Spatial Domain



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#### The Fast Discrete Curvelet Transform

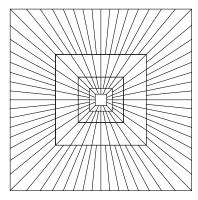


Figure: Frequency Domain

Figure: Parallelogram Support



#### The Fast Discrete Curvelet Transform

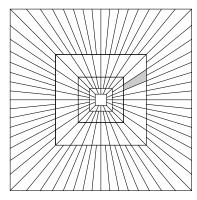


Figure: Frequency Domain



Figure: Parallelogram Support

#### The Fast Discrete Curvelet Transform

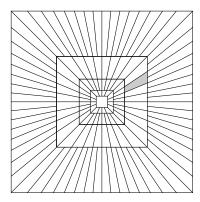


Figure: Frequency Domain

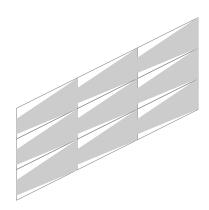


Figure: Parallelogram Support

#### The Fast Discrete Curvelet Transform

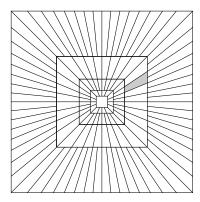


Figure: Frequency Domain

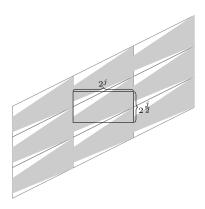


Figure: Parallelogram Support

#### Global Feature Extraction

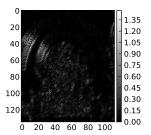


Figure: Curvelet coefficients at a specific scale and angle

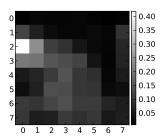


Figure: Mean values on an  $8 \times 8$  grid

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### Local Feature Extraction (Sampling)

PMEAN Collect  $(n-m+1)^2$  sample vectors of length  $N_s \cdot N_{\theta_s} \cdot m^2$  by concatenating across scales and angles

PMEAN2 Collect  $N_s \cdot (n-m+1)^2$  sample vectors of length  $N_{\theta_a} \cdot m^2$  by concatenating across angles

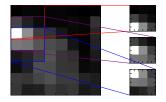


Figure:  $8 \times 8$  mean coefficient grid sampled using  $3 \times 3$  window

n image width and height

- m window width and height
- $N_s$  Number of scales
- $N_{\theta}$ . Number of angles at scale s

# Local Feature Extraction (Clustering)

- k-means clustering
- ightharpoonup Codebook size k = 1000
- ▶ Each sample vector is assigned to the closest cluster  $S_i$ , i = 1, ..., k

Proposed Solution

► Image signature is the number of occurences of each "visual word" in the image:

$$\tilde{I} = [|S_1|, |S_2|, \dots, |S_k|]$$



# Ranking



# Benchmarking Method



### Cross-Domain Results



Results ○○●

### Intra-Domain Results



