

Detecting Large Repetitive Structures with Salient Boundaries



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Repetition Detection

In urban photos, regions of repeating elements are often **large**.

- Large appearance variations between elements due to occlusions, reflections, depth changes, or actual structural differences, etc



Key Observations

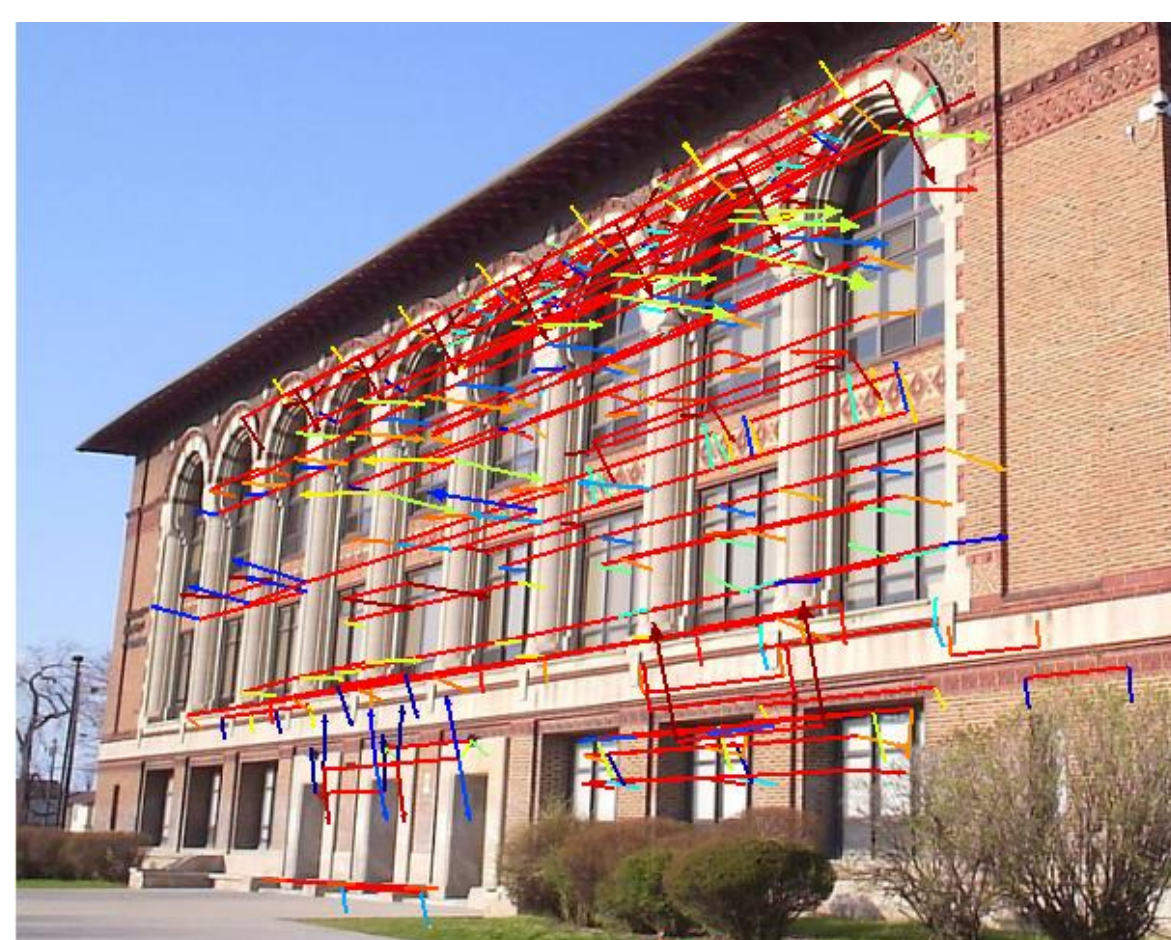
- Repetitions happen along Vanishing Point (VP) direction(s)
 - Allows our repetition-based VP refinement
- Reflective symmetry along the horizontal direction
 - Enables our boundary selection deploying the coexistence of repetition and reflective symmetry
- Possible lack of repetition/symmetry in the vertical direction
 - Determine the boundaries according to the repetition in the perpendicular direction with a novel repetition quality measurement

Algorithm Outline

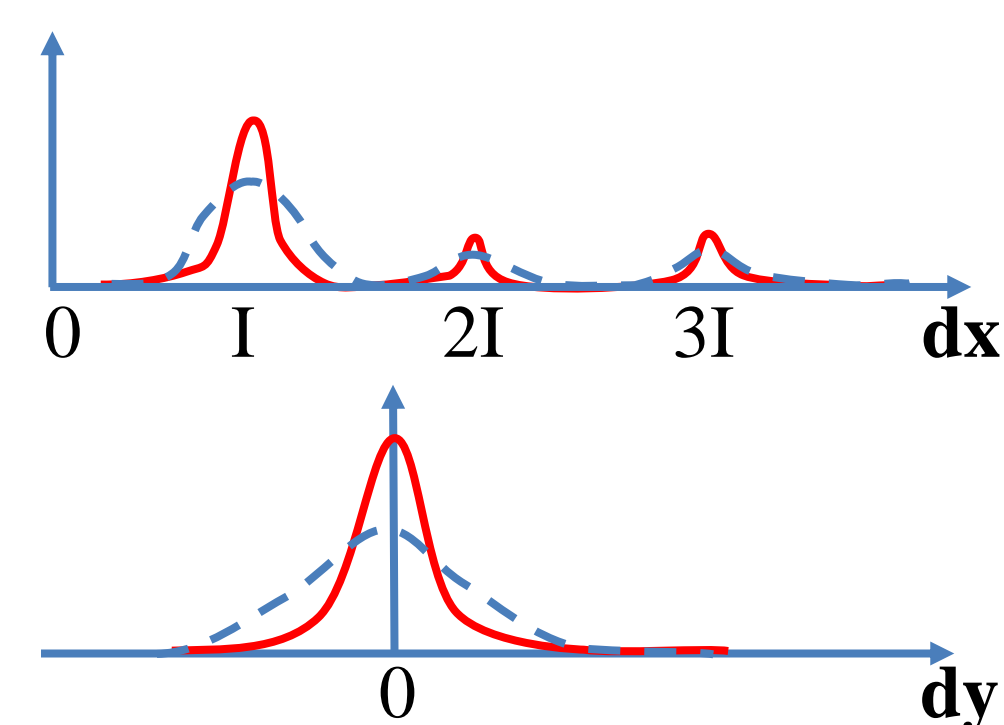
- Vanishing point detection and image rectification
- Finding possible repetition intervals and symmetry axes
- Determining regions for each new repetition interval

Repetition-based VP Refinement

- The horizontal/vertical differences of the matched feature pairs should have histograms with only a few strong peaks.
- Similar for symmetrical feature pairs and edge segments
 - Entropy**-based optimization of vanishing point



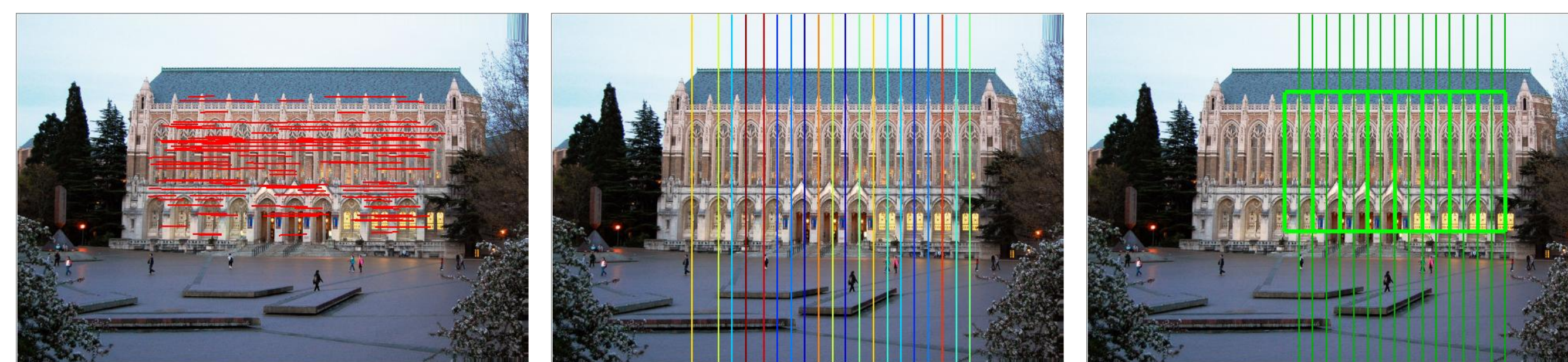
Example of matched feature pairs along VP directions



Histogram of coordinate differences (**dx** and **dy**) in rectified image; The optimization seeks strong narrow peaks (red vs. blue)

Repetition Intervals and Symmetry Axes

- The distances between repeating symmetry axes are equal to half of the repetition interval (width of repeating element)



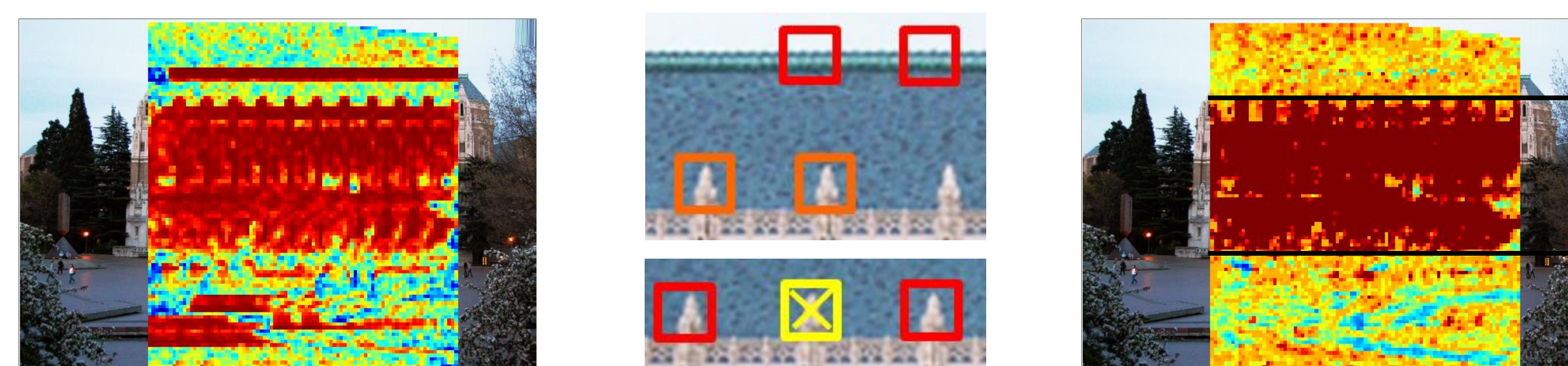
- Initialize repetition regions from groups of repeating axes

Determining Boundaries of Repetition Region

- Evaluate patch similarity for repetition
 - Adaptive patch size proportional to the repetition interval
 - Robust comparison by using upright SIFT descriptor (GPU)
 - $D(x, I)$ denotes the distance at location x for interval I
 - Propagate regions according to the similarity measurements
- Distinguish with integer multiples of repetition
 - Choose a set of offsets $T = \{0, \pm I/2, \pm I/3, \dots\}$
 - Compute a set of distances $V = \{D(x, I + t) \mid t \in T\}$
 - Repetition quality measurement based on distance ratio

$$f(x, I) = \min\left(\alpha \frac{V_{(2)} + \sigma}{D(x, I) + \sigma}, 1\right)$$

where α is a ratio parameter, and σ is the variance parameter



Similarity Ambiguity Repetition Evaluation

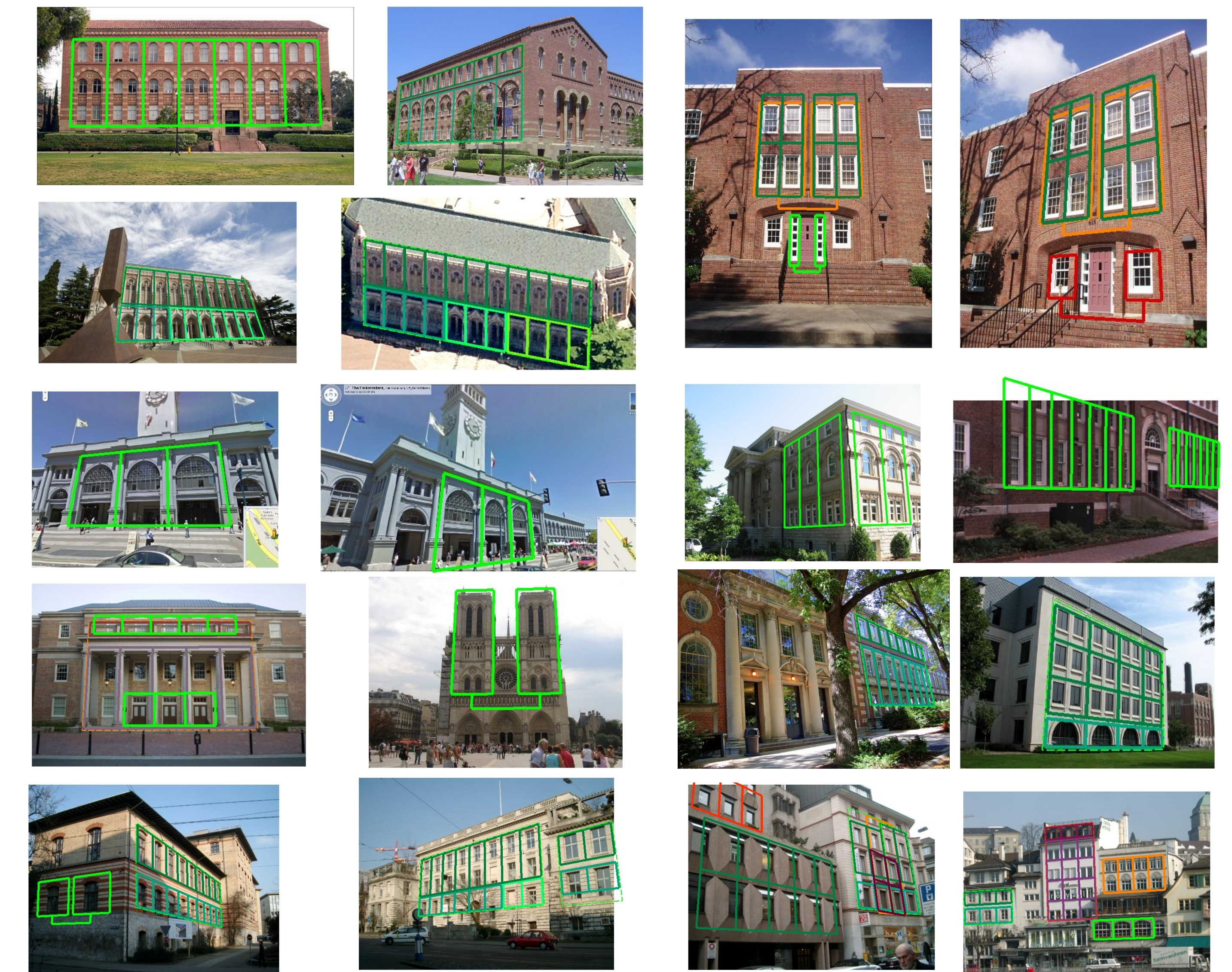
- Determine vertical boundary for non-repeating direction

Further decomposition of repetition regions

- Divide regions at multiple vertical locations
- Find repetition along the vertical repetitions
- Decompose regions according to continuation of repetition



Qualitative Results



Quantitative Evaluation

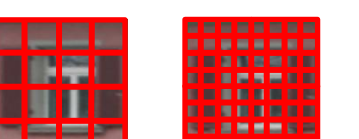
- Evaluation of our repetition detection on the Zürich Building Database (ZubuD): 1005 images; 5 images per building.
<http://www.vision.ee.ethz.ch/showroom/zubud/>

Category (excluding 282 manually discarded images)	#	Percentage
No detection due to VP detection failure	25	4%
No correct detection due to other algorithmic limitations	34	5%
Partial Detection; Missing major repetitions	88	12%
Full detection of all major repetitions; Some boundaries errors	67	9%
Full detection of all major repetitions; Good boundaries	509	70%

- Single feature image retrieval on ZubuD.



- Describe repeating elements by 4x4 or 8x8 gradient histogram



- Query closest image for every element (Consider correct if a retrieval belong to the other 4 images of the same building)

- Use the standard SIFT feature as a baseline for comparison

