

Adaptive Parameter Tuning for Morphological Segmentation of Building Facade Images

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Pattern Spectrum (PS)

8x10⁵

6x10⁵

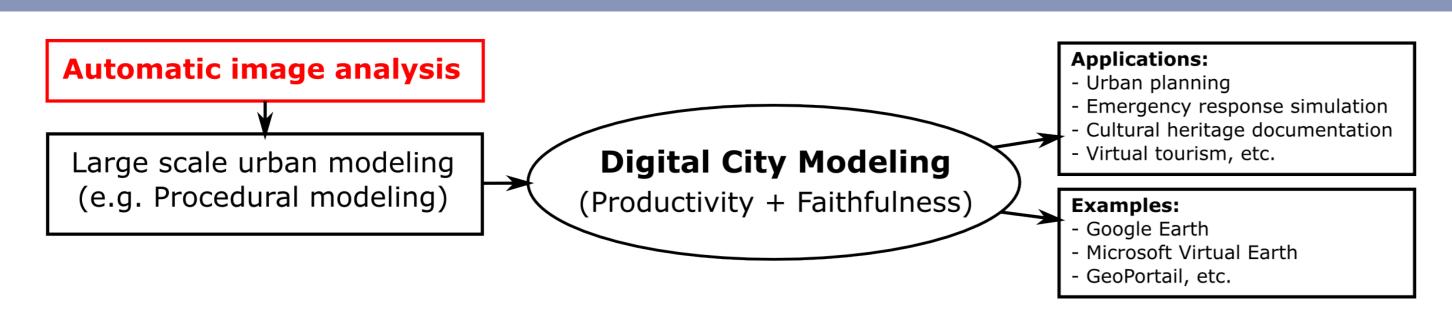
 $4x10^{5}$

 $2x10^{5}$

monge_55

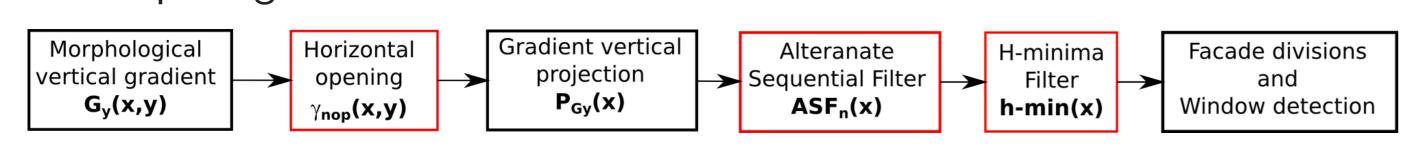
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1. Introduction



► Hypothesis:

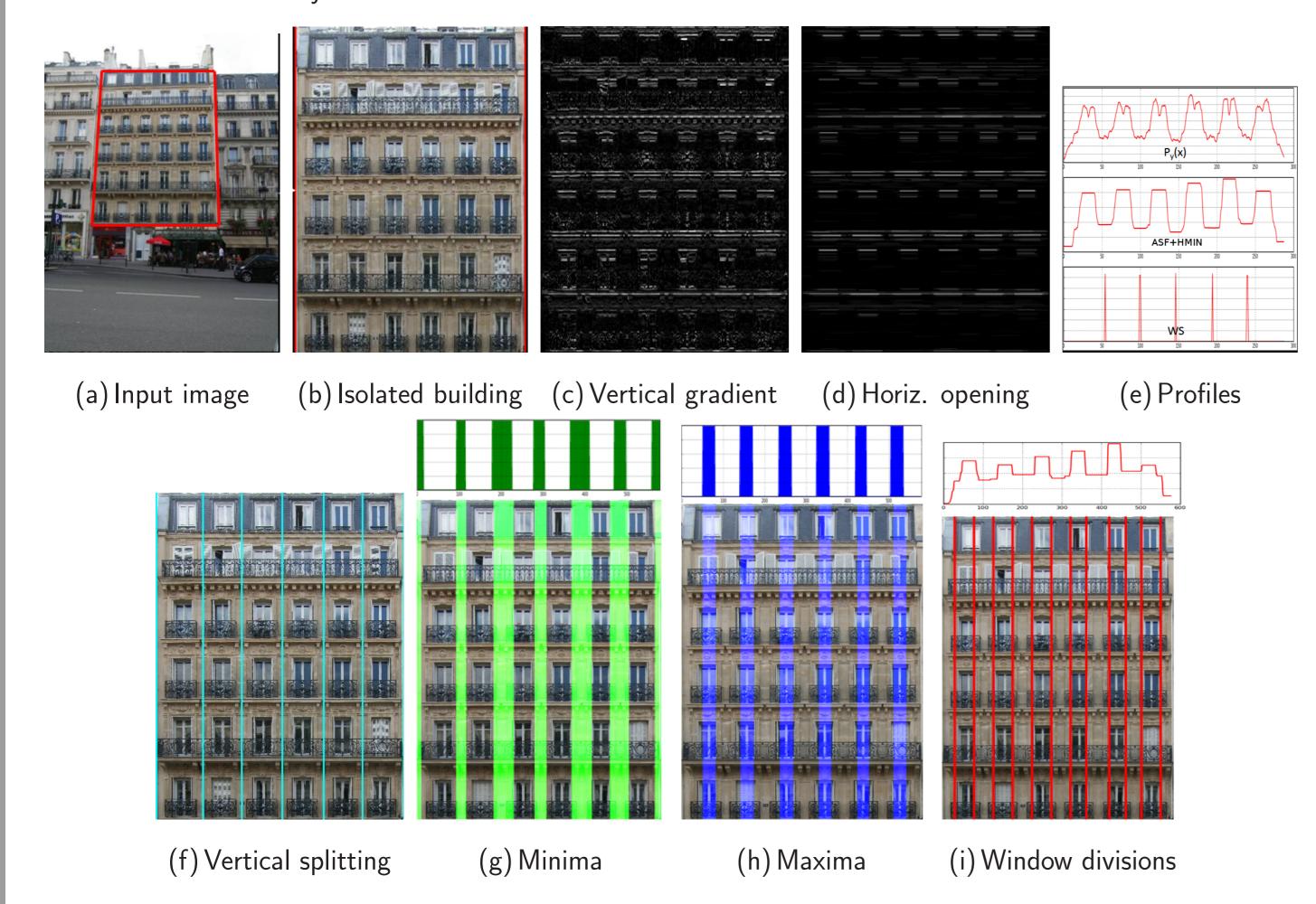
- ▶ Images are rectified and cropped.
- ▶ Facade elements are pseudo-regularly aligned (Haussmannian buildings).
- ► Our method is based on:
 - ▶ Accumulation of directional color gradients.
 - ▶ Morphological filters.



► We propose the adaptive tuning of morphological filters.

2. Method

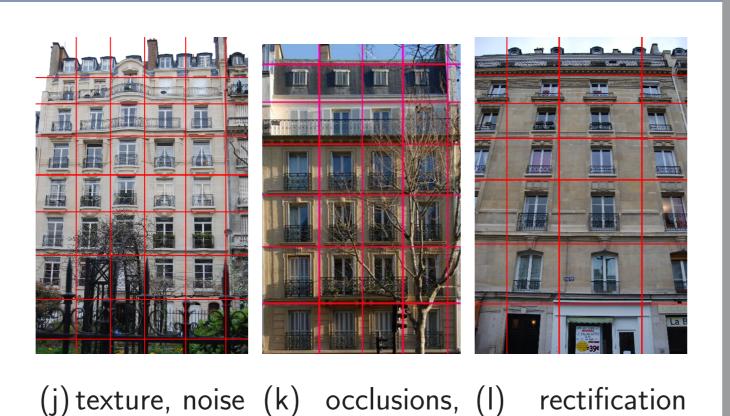
- ▶ Input images (Fig. a) are assumed rectified and cropped (Fig. b).
- \blacktriangleright Morphological vertical gradient G_y detects horizontal contours (Fig. c).
- ▶ Opening $\gamma_{n_{op}}$ is applied to eliminate undesirable details (Fig. d).
- ightharpoonup Accumulated vertical gradient P_{G_y} (Fig. e) contains peaks at window locations and valleys between them.
- ightharpoonup ASF_n and h-min filters get a single maximum for each window (Fig. e).
- ► Watershed on the inverted profile computes facade divisions (Figs. e,f).
- In the filtered profile $\hat{\mathbf{P}}_{\mathbf{G}_y}$, minima pass through the wall (Fig. g) while maxima pass through the windows (Fig. h).
- ightharpoonup Constrained watershed on the projected horizontal gradient \mathbf{P}_{G_x} , taking the extrema of $\tilde{\mathbf{P}}_{G_v}$ as markers, determines the window line divisions (Fig. i).



► Although a vertical splitting is explained, the same technique applies to horizontal splitting, just changing vertical by horizontal and vice versa.

3. Some Segmentation Examples

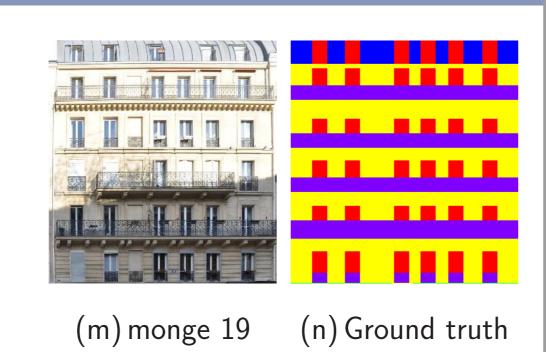
- ➤ Our approach applies to pseudo-periodic structures and it is robust to different issues (Figs. j,k,l).
- ➤ Good results are obtained, but they depend on a correct parameter tuning.



shadows

4. Database

▶ 100 rectified and annotated images from the public database: http://vision.mas.ecp. fr/Personnel/teboul/data.php.



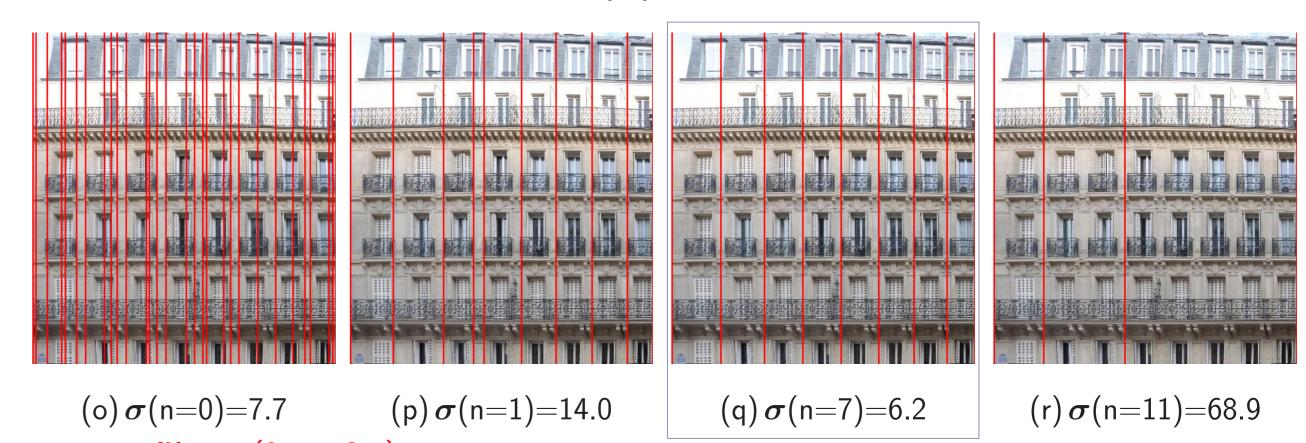
5. Adaptive filter tuning

Opening filter $(\gamma_{n_{op}})$

- $ightharpoonup \gamma_{n_{op}}$ gets rid of noise from gradient images.
- Size distribution of the noise is characterized using $PS_i = \sum_{\forall pixel} (\gamma_{i-1} \gamma_i)$.
- ► n_{op} is chosen as the value i for which **PS** falls down under 25% of its maximum.
- ► Robustness to image resolution changes.

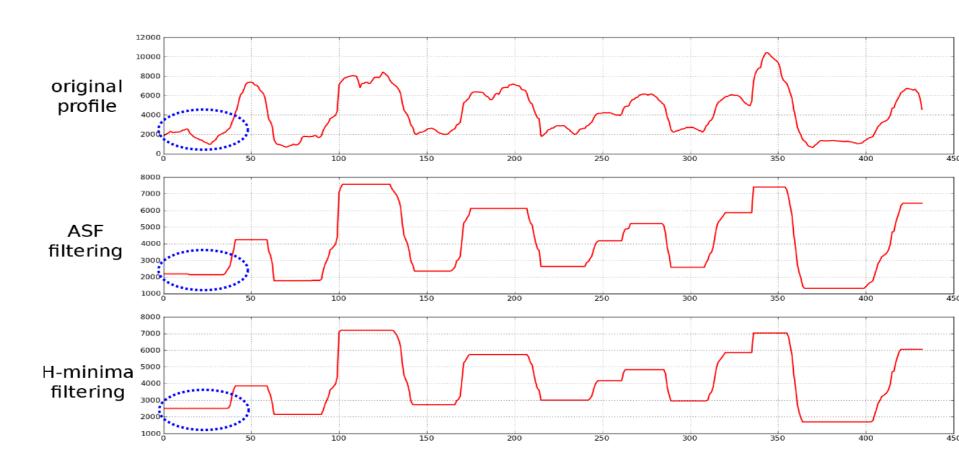
Alternate Sequential Filter (ASF_n)

- ► ASF_n is a sequence of openings (γ_i) and closings (φ_i) of increasing sizes: ASF_n(P_{G_y}) = $\gamma_n \varphi_n ... \gamma_2 \varphi_2 \gamma_1 \varphi_1$ (P_{G_y}).
- ► The **n** leading to the most regular result is selected.
- ightharpoonup Regularity: standard deviation $\sigma(n)$ of the segmented facade division sizes.



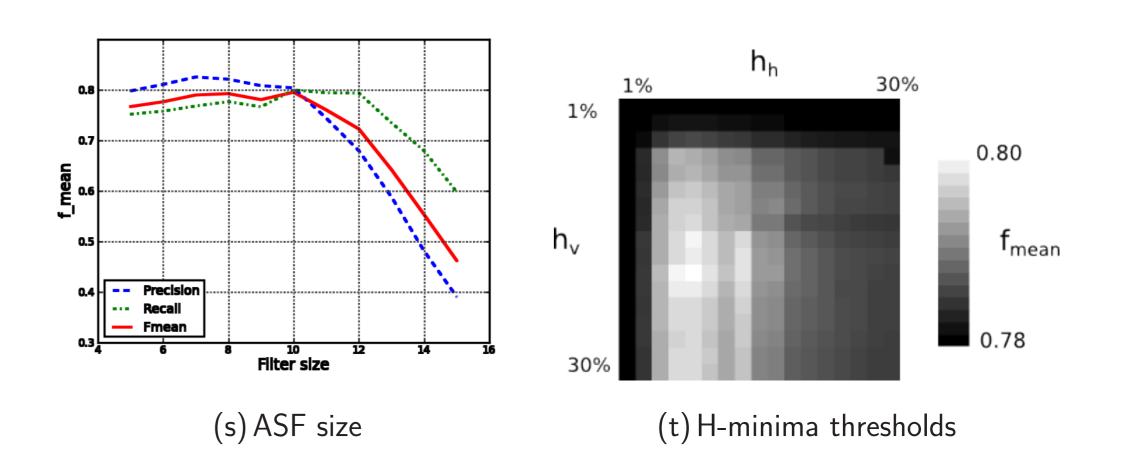
H-minima filter (h-min)

- ▶ h-min removes remaining low contrasted extrema (lower than h).
- ▶ h is chosen depending on the extrema dynamic: $h \propto range(ASF_n(P_{G_v}))$



6. Results

- ► We evaluate window localization using Precision, Recall, f_{mean}.
- ► An exhaustive test for **n** shows that our tuning method offers the best possible score \mathbf{f}_{mean} =0.79 (Fig. s).
- ► For **h**, the best possible f_{mean} corresponds to $h_v=14\%$ and $h_h=2.5\%$ in vertical and horizontal profiles, respectively (Fig. t).



► Overall performances: Precision=0.82, Recall=0.79 and f_{mean}=0.80



7. Conclusions

- ➤ Our adaptive segmentation method offers robustness to texture, noise, image resolution changes, occlusions, shadows and rectification errors.
- ► It fails if pseudo-periodic hypothesis is not fulfilled.
- Good performances on Teboul's public database.
- ▶ In the future, ultimate opening will be studied. This operator automatically adapts its size to image structures based on a contrast criterion.