13th International Conference on Computer Vision (ICCV 2011), November 6-13, 2011, Barcelona, Spain





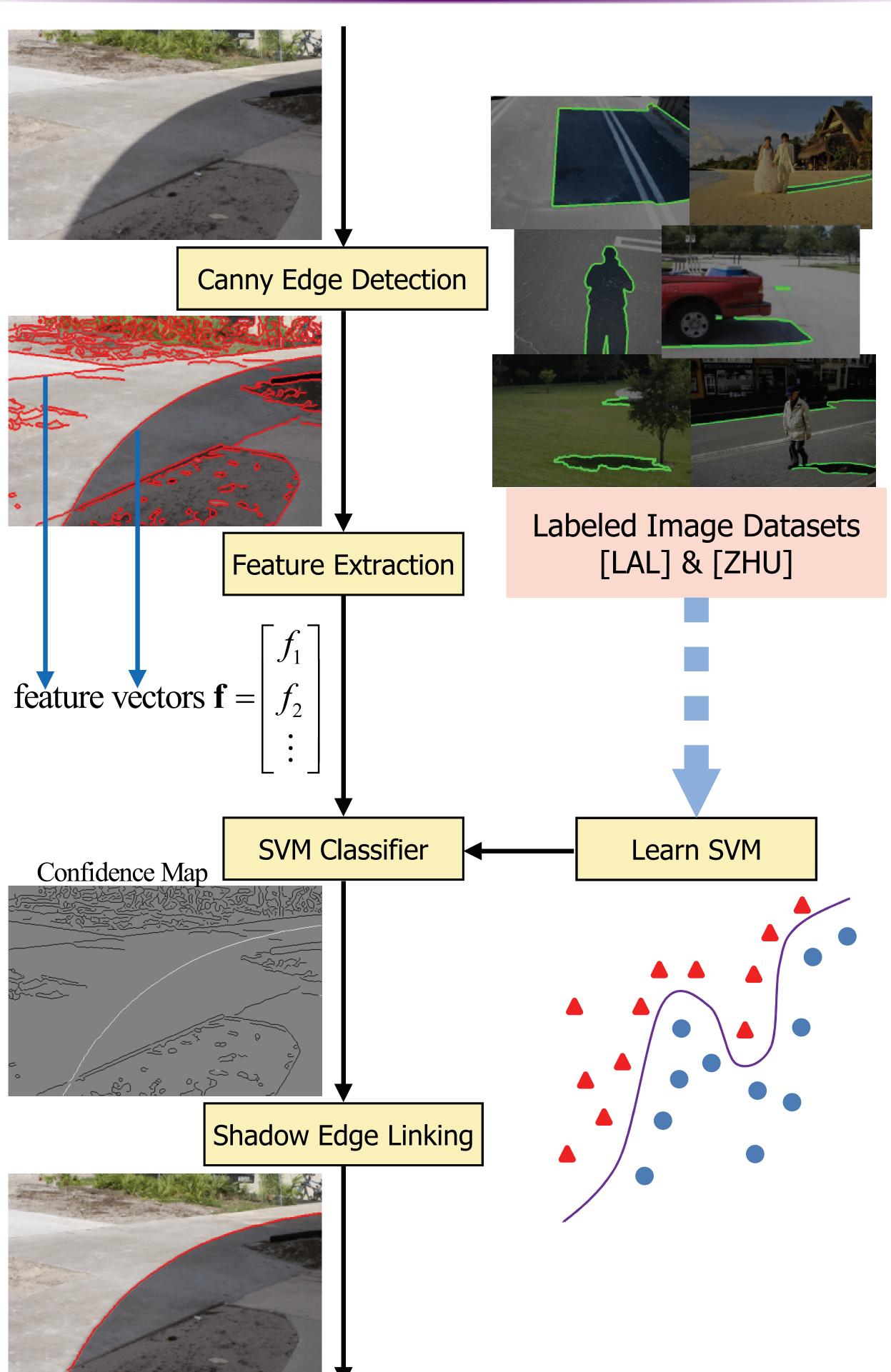


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Key Contributions

► A careful analysis of the physical models of shadows under the sun and the sky; ► A compact set of robust visual features motivated from these physical models for shadow boundary detection; A end-to-end shadow boundary detection system built upon these features, which outperforms previous methods.

Our Shadow Detection System



What Characterizes a Shadow Boundary under the Sun and Sky?

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New Features: Outdoor Shadow Physical Models

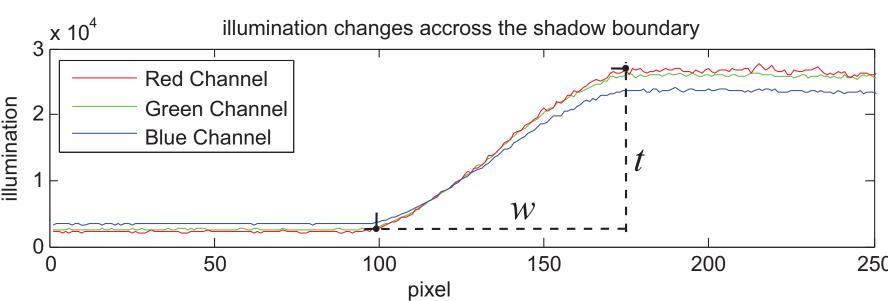
enlarge

Corresponding Penumbra Image

- w ——>

Compute Penumbra Width, Shape and Color

- Penumbra width proportional to occluder height: $\frac{h}{cos^2\theta}$
- Similar shape for all penumbra widths (soft or sharp shadow): illumination change rate proportional to $\sqrt{1-(\frac{2x}{w})^2-\frac{w}{2}} \le x \le \frac{w}{2}$.
- Color shifts from bluish sky to reddish sun



► Visual features Motivated from Physical Models

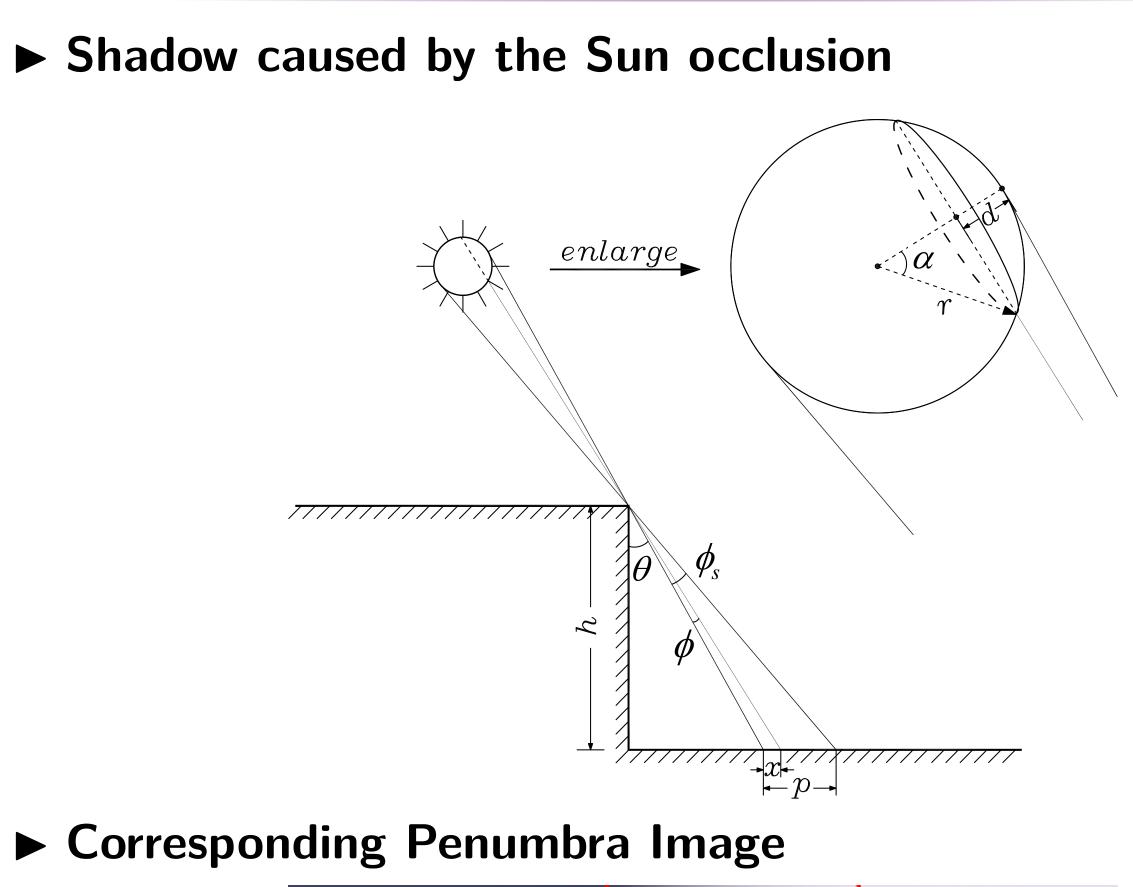
- Shadow sharpness w in RGB channels
- Dark-to-bright slope
- Dark-to-bright ratio: Sky vs. Sun
- Dark-to-bright gradient direction mismatch

Figure: Gradient of shadow edges: same direction in R, G, B in (a). Reflectance, silhouette & other edges: any direction in (a) (b) (c).

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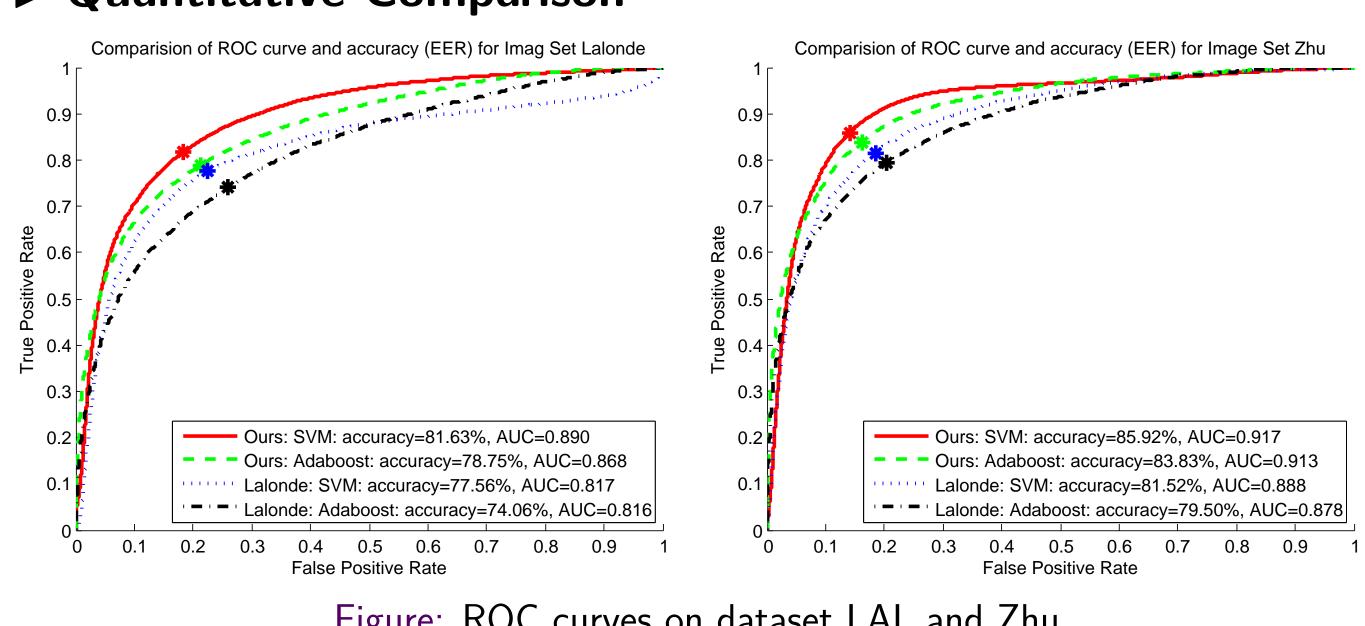
lancejwilliams@gmail.com





► Dataset and Settings:

Quantitative Comparison



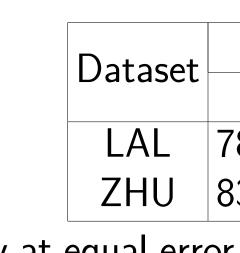


Table: Accuracy at equal error

► Qualitative Comparison



consumer photographs," In *Proc. ECCV*, 2010, pp. 1–14. Monochromatic Natural Images," In Proc. CVPR, 2010, pp. 223–230.

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Experiments

• Two datasets: [LAL] (135 images, 100 training, 35 testing) and [ZHU] (162 images, 100 training, 62 testing)

• Feature Vector: 12 features, each at three scales

Classifier: SVM with RBF kernel (also compared with AdaBoost)

Figure: ROC curves on dataset LAL and Zhu.

AdaBoost		SVM				
Ours	Lalonde's	Ours	Lalonde's			
78.75%	74.06%	81.63%	77.56%			
33.83%	79.50%	85.92%	81.52%			
r rate, c	compared v	vith [LAI] in datase	et [LAL]	and	[ZHU

- Figure: Shadow labeled with red color. Compare ours (bottom) with [LAL] (top).
- [LAL] J.-F. Lalonde, A. A. Efros, and S. G. Narasimhan, "Detecting ground shadows in outdoor [ZHU] J. Zhu, K. Samuel, S. Masood, and M. Tappen, "Learning to Recognize Shadows in