

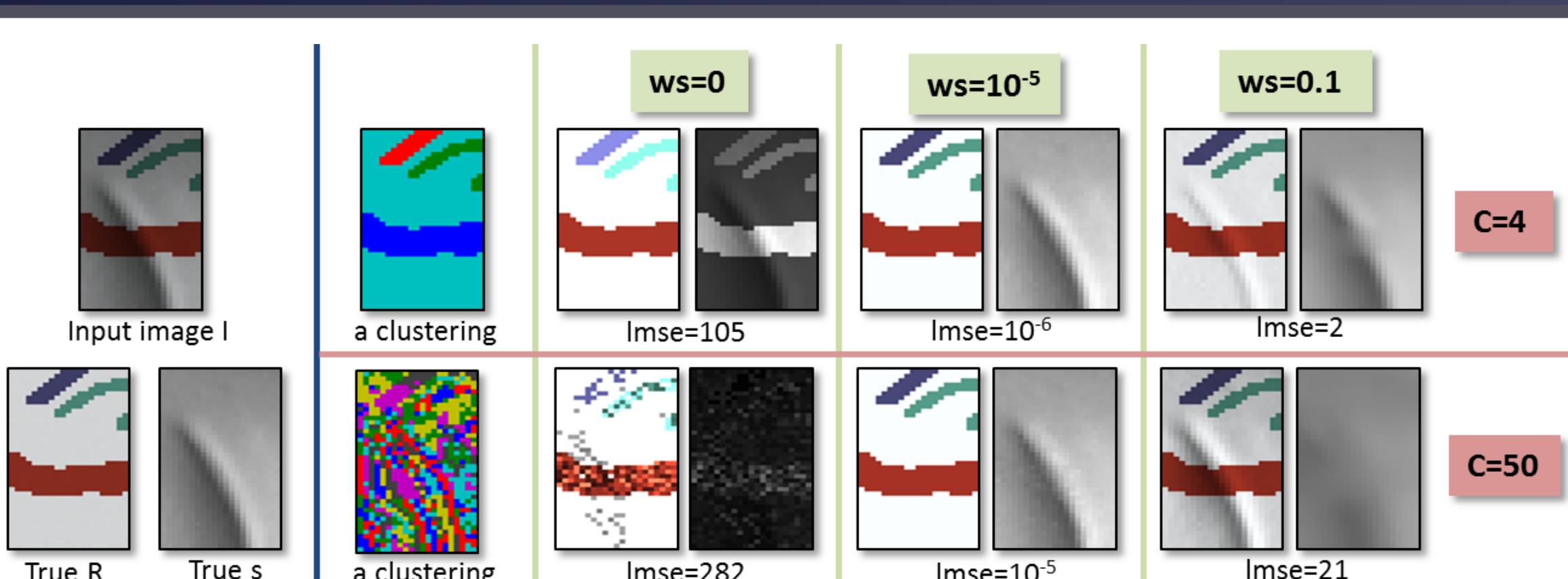
PROBLEM

Decompose a single image $I \in [0, 255]^{m \times n \times 3}$ into separate shading $S \in [0, 255]^{m \times n}$ and reflectance $R \in [0, 255]^{m \times n \times 3}$ layers.

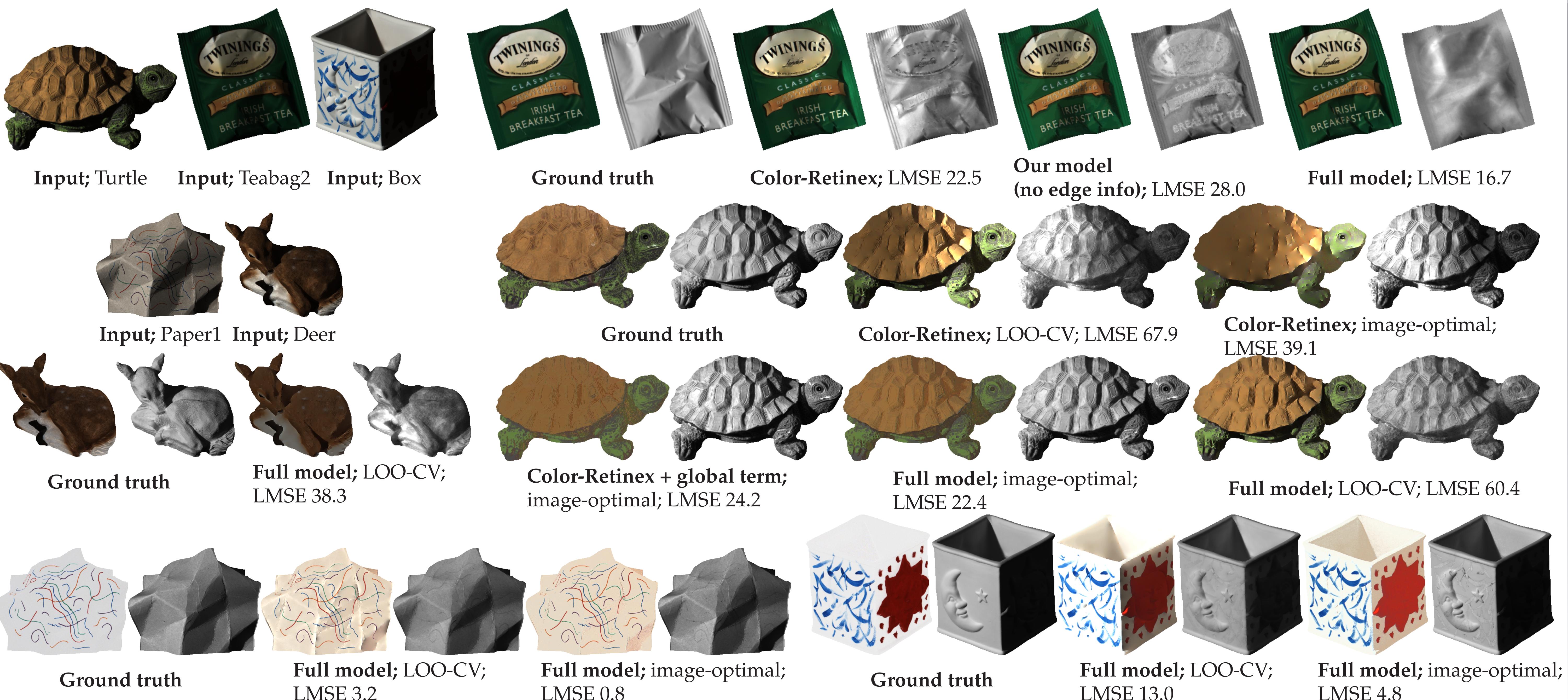


$$I_i = R_i s_i, \quad I_i^c = s_i R_i^c, \quad R_i = r_i \vec{R}_i$$

EFFECT OF E_{cl}



RESULTS



NUMERICAL RESULTS

comment	E_s	E_{cl}	E_{ret}	LOO-CV	best single	image opt.
Color-Retinex	-	-	✓	29.5	29.5	25.5
no edge information	✓	✓	-	30.0	30.6	18.2
Col-Ret+ global term	-	✓	✓	27.2	24.4	18.1
full model	✓	✓	✓	27.4	24.4	16.1

Comparing the effect of including different terms.

	LOO-CV	rank	best single	im. opt.
Tappen et al., 2005, PAMI	56*	-	-	-
Tappen et al., 2006, CVPR	39*	-	-	-
Shen et al., 2008, CVPR	n/a	n/a	56.2	n/a
Shen&Yeo, 2011, ICCV [2]	n/a	n/a	(20.4)	-
Baseline: Grosse, 2011, ICCV [1]	72.6	5.1	60.3	36.6
Gray-Retinex [1]	40.7	4.9	40.7	28.9
Color-Retinex [1]	29.5	3.7	29.5	25.5
full model	27.4	3.0	24.4	16.1
Weiss, 2001, ICCV (sequences)	21.5	2.7	21.5	21.5
Weiss+Retinex, [1](sequences)	16.4	1.7	16.4	15.0

Comparison to literature.

SUMMARY

- A new latent variable MRF model for intrinsic images
- A global sparse reflectance prior in pixel domain
- Extends of Color-Retinex
- More training data is needed for progress in this field!

REFERENCES & SOURCE CODE

- R. Grosse, M. Johnson, E. Adelson, and W.T. Freeman. Ground truth dataset and baseline evaluations for intrinsic image algorithms In *ICCV '09*
 - L. Shen and C. Yeo. Intrinsic images decomposition using a local and global sparse representation of reflectance In *CVPR '11*
- The source code is available at <http://tinyurl.com/cql2dh1>

