

Ramen Spoon Eraser : CNN-based photo transformation for improving attractiveness of ramen photos

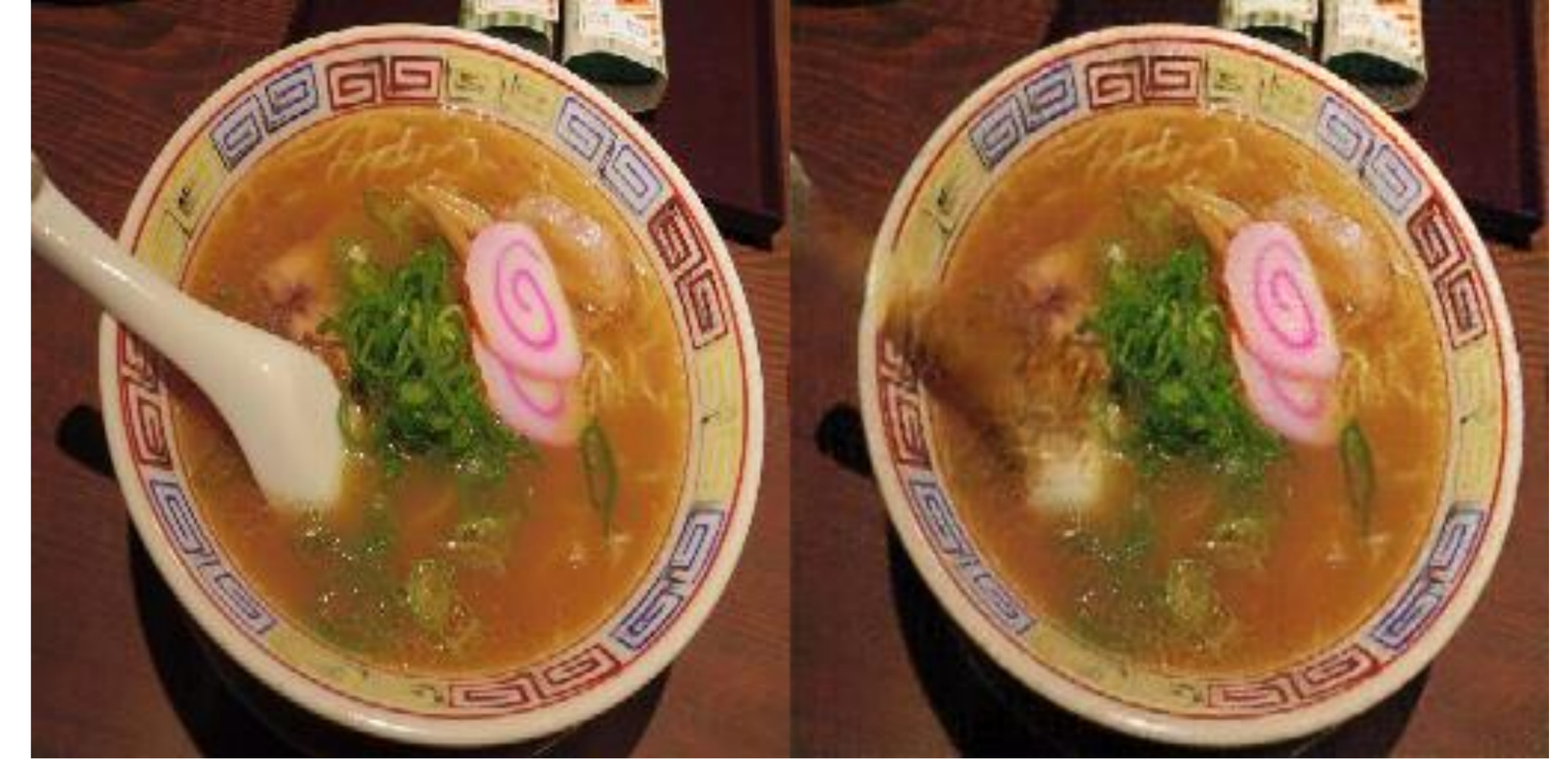
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Introduction

- A ramen spoon in a bowl spoils its appearance greatly as shown in the right figure.
- To improve attractiveness of ramen photos, erasing ramen spoon is indispensable.
- High-quality image transformation is possible with GAN-based image-to-image transformation (Pix2pix) [3], which potentially erases ramen spoons from photos.



Proposed Method

- We adopt a U-Net[2] as a image transformation network.
- We train it in two ways: MSE loss and Pix2pix loss[3]
- Mean Squared Error (MSE) Loss

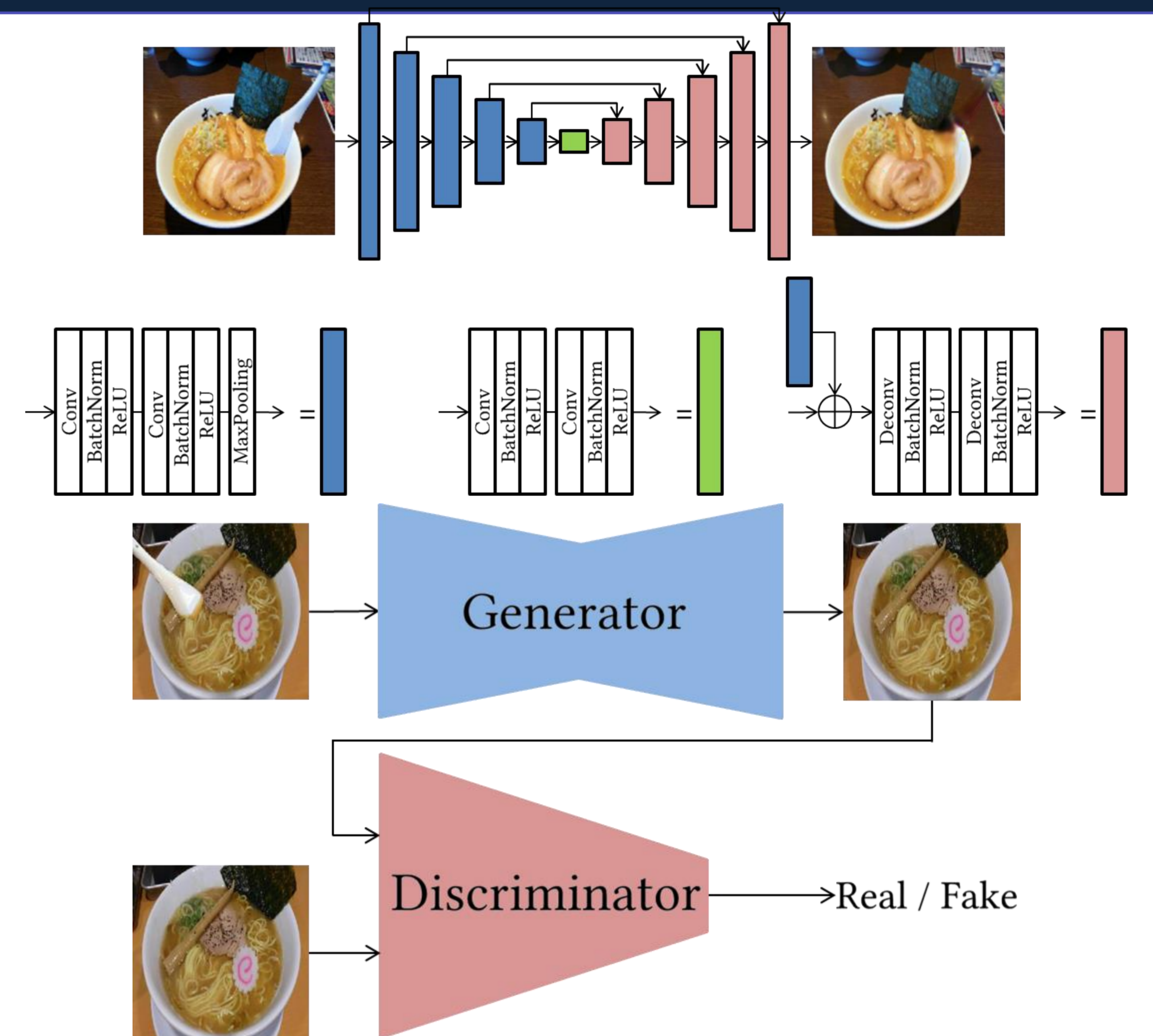
$$\ell(x, y) = L = \{l_1, \dots, l_N\}^T, \quad l_n = (x_n - y_n)^2,$$

- Pix2pix [3] (with adversarial training)

$$\mathcal{L}_{cGAN}(G, D) = \mathbb{E}_{x, y} [\log D(x, y)] + \mathbb{E}_{x, y} [\log (1 - D(x, G(x, z)))] .$$

$$\mathcal{L}_{L_1}(G) = \mathbb{E}_{x, y, z} [\|y - G(x, z)\|_1] .$$

$$G^* = \arg \min_G \max_D \mathcal{L}_{GAN}(G, D) + \lambda \mathcal{L}_{L_1}(G) .$$



Experimental Results

- [Training data] 6000 pairs of ramen-without/with-spoon photos
 - We generated ramen-with-spoon photos by overlaying ramen spoon images on the ramen-without-spoon photos randomly.
- We trained two model using the MSE loss and the Pix2pix loss.
- We compare two models :
 - **MSE loss** erased spoons. However, the results are blurred.
 - **Pix2pix** erased spoons more naturally than MSE loss.
- In the third row, both models failed to remove spoons, because the training data did not contain enough images with red ramen spoons.
- We have not evaluated which one seems to be more attractive by subjects.
 - So, we would like you to attach a sticker to more attractive images !!

(a) Original (b) MSE loss (c) pix2pix



MSE Loss

Pix2pix

Conclusions

- We proposed an application which erases a spoon from a ramen-with-spoon photo by CNN-based image transformation.
- We plan to extend our model for diverse utensils such as chopsticks, forks and knives to make food photos more attractive.

[1] I. Goodfellow, J. Pouget-Abadie, M. Mirza, B. Xu, D. Warde-Farley, S. Ozair, A. Courville, and Y. Bengio. Generative adversarial nets. In Advances in Neural Information Processing Systems (NIPS), pp.2672–2680. 2014.

[2] O. Ronneberger, P. Fischer and T. Brox. U-Net: Convolutional Networks for Biomedical Image Segmentation. In Proceedings of the International Conference on Learning Representation (ICLR), 2015.

[3] P. Isola, J.-Y. Zhu, T. Zhou, and A. A. Efros. Image-to-image translation with conditional adversarial networks. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2017.