# Magical Rice Bowl: A Real-Time Food Category Changer

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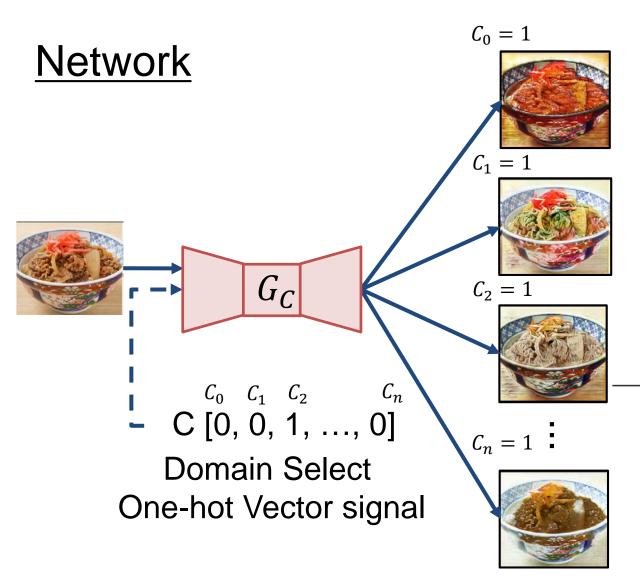
**Datasets** 



### **Overview**

# Multi-Domain Image-to-Image Translation only to food area

-Based on a StarGAN[1] with a large-scale food image data collected from the Twitter Stream.



-We have gathered 230k food images which consist of 10 kind of food categories from Twitter stream.

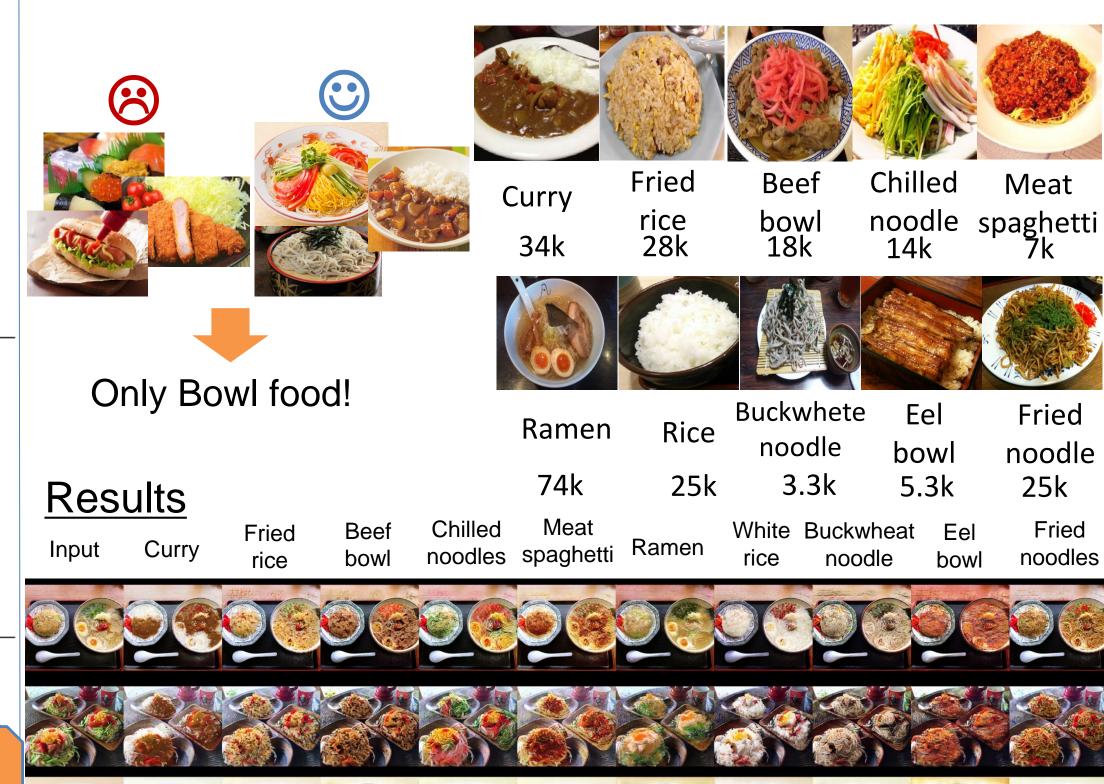
Table 1: training data

category	image number
chilled noodles	13,499
meat spaghetti	7,138
buckwheat noodle	3,530
ramen	74,007
fried noodles	24,760
white rice	21,324
curry rice	34,216
beef bowl	18,396
eel bowl	5,329
fried rice	27,854
TOTAL	230,053
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## **Experiments**

In the case of the dataset has diversity geometric structure, it is difficult to translate.

So we set the constraint of bowl only.



## Background

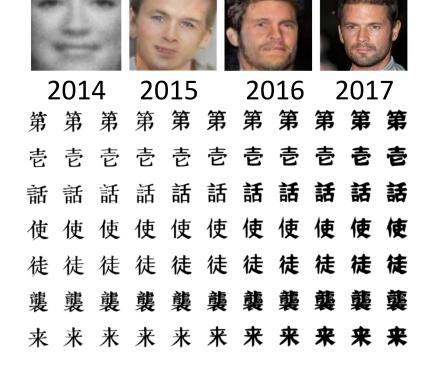
There are many cases in **letters** and **faces** are well generated and translated by GANs.

However, there are few cases of **food**. So we challenged food transformation.

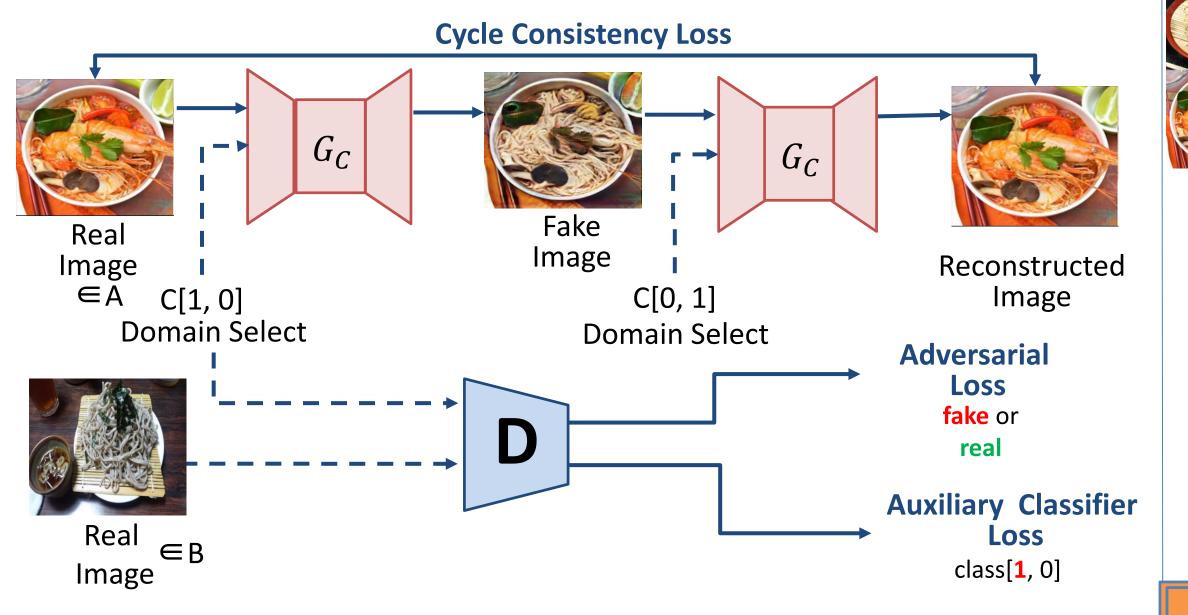








#### **Methods**



 $L_D = L_{adversarial} + \lambda_{classifier} L_{classifier}$ 

 $L_G = L_{adversarial} + \lambda_{classifier} L_{classifier} + \lambda_{cycle} L_{cycle}$ 

# **Demo Food**

Changes in quality due to differences of the number of

The number of images and the quality are proportional

**FULL** 

10k

100k

**FULL** 



Input 10k / 230k 100k

total images.



### References

[1] Y.Choi, M.Choi, M.Kim, J.W.Ha, S.Kim and J.Choo, StarGAN: Unified Generative Adversarial Networks for Multi-Domain Image-to-Image Translation. In Proc. of CVPR2018.

[2]J. Y. Zhu, T.Park, P. Isola, A.A. Efros, Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks. In Proc. of ICCV2017.

[3] A. Odena, C. Olah, and J. Shlens. Conditional Image Synthesis With Auxiliary Classifier GANs. In Proc. of ICML2017.