

AR DeepCalorieCam: An iOS App for Food Calorie Estimation

with Augmented Reality

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Background & Objective

Meal management apps enable us to record food calories. Some of them need human help for calorie estimation.

Image-based food calorie estimation using recipe information with **Augmented Reality**



[DeepCalorieCam]



[AR DeepCalorieCam]

Method: Multi-task CNN for calorie estimation

Related work ① : Miyazaki et al.[1] 2011

Search-based food calorie estimation with conventional features.
Similar image retrieval with SURF and color histograms and so on.
Calculate food calories from retrieved images' calories.

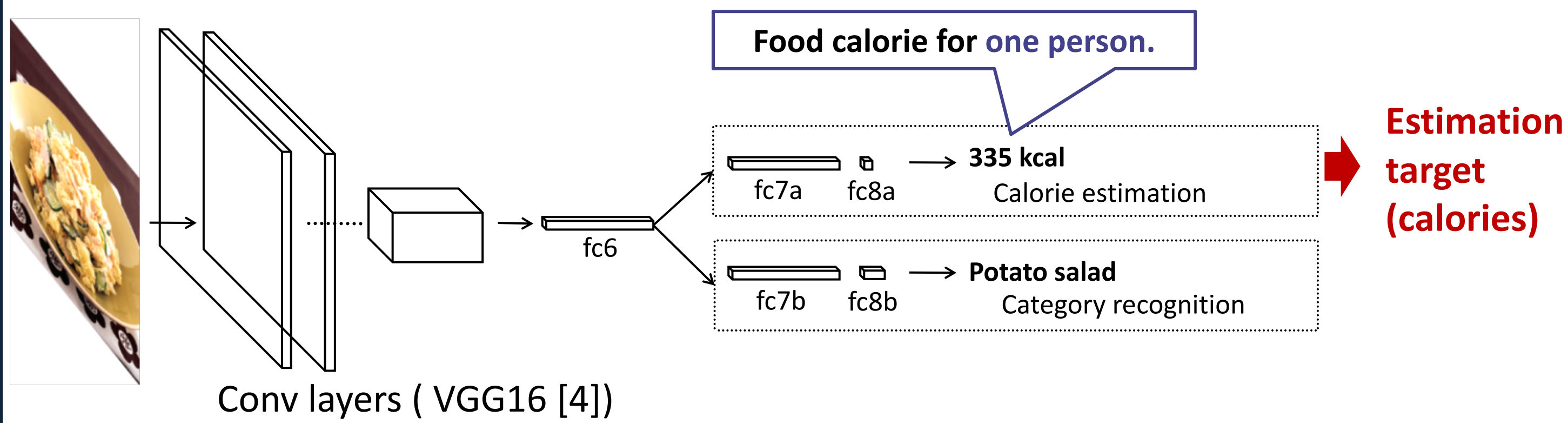
→ We propose regression-based method using **CNN**.

Related work ② : Chen and Ngo[2] 2016

Multi-task estimation of food categories and food ingredients.
Multi-task CNN[3] of food categories and food ingredients
Multi-task learning improve both task's performance.

→ We use **multi-task CNN** for calorie estimation.

[Our network]



→ It is expected to improve the accuracy of each task.

[Loss function]

(1) **Calorie estimation loss:** $L_{cat} = \lambda_{re}L_{re} + \lambda_{ab}L_{ab}$ (λ is the weight on the loss.)

(2) **Food Category loss:** relative err. loss $L_{re} = \frac{|y - g|}{g}$ absolute err. loss $L_{ab} = |y - g|$
 $L_{cat} = - \sum_{k=1}^n g_k \log y_k$ (softmax cross entropy) (y is an estimated food calorie. g is ground-truth.)

[DATASET]

For training our network, we collected calorie-annotated food photos from the online cooking recipe sites.

Food 15 categories.
A total of 4877 images.



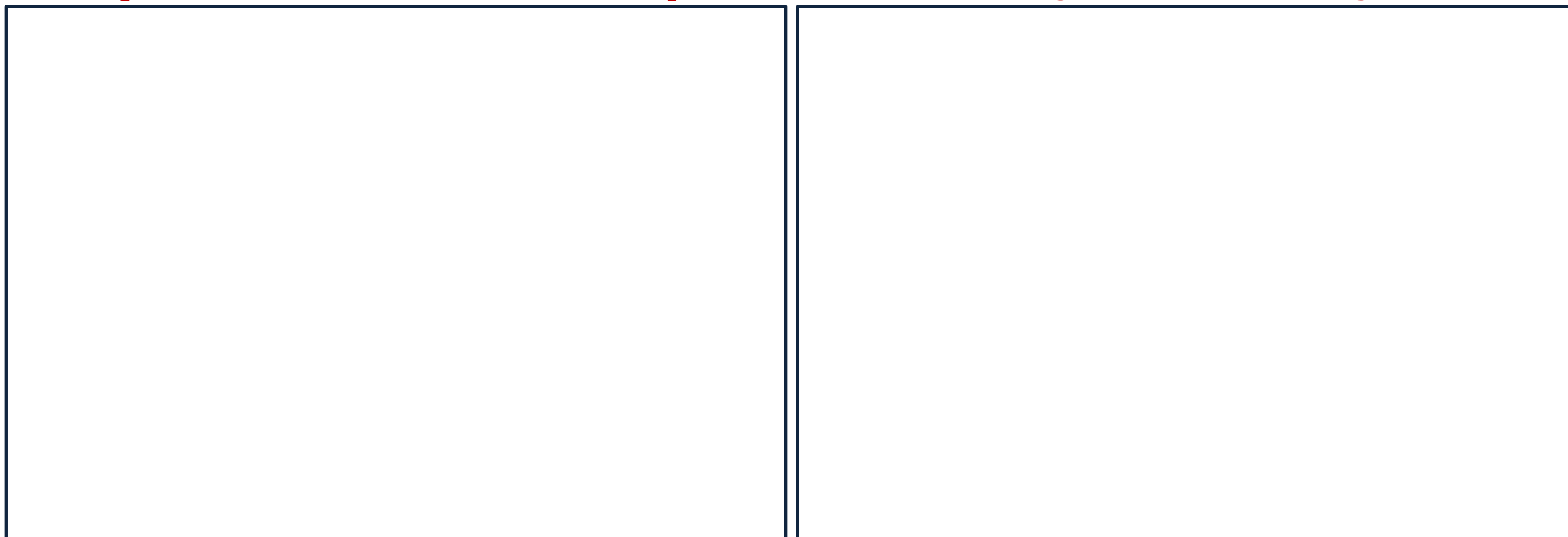
[Implementation]

- Train with **Keras**(backend **TensorFlow**) framework
- Convert **Keras** model to **CoreML** model for iOS deployment
- Display calorie estimation result using **Apple ARKit** framework

Application Demo:

[Calorie Estimation with AR]

[Food Image Transfer using GANs]



My Pet Project Introduction:

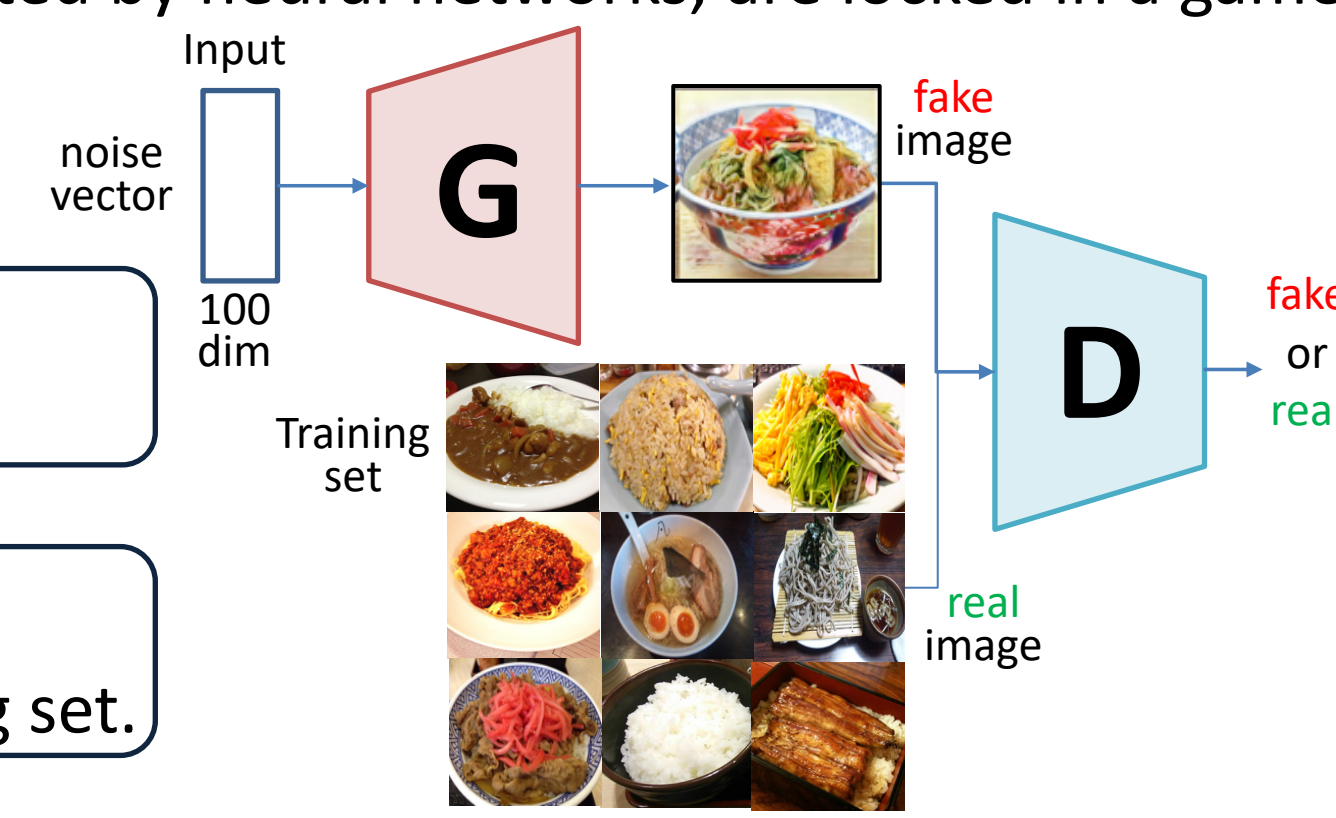
Background: **Food Image Transfer using Generative Adversarial Networks(GANs)**

GANs are a kind generative models designed by Goodfellow et al in 2014.

In a GAN setup, two differentiable functions, represented by neural networks, are locked in a game. The two players, the **Generator** and the **Discriminator**, have different roles in this framework.

Generator Player 1
The **Generator** tries to produce data that come from some probability distribution.

Discriminator Player 2
The **Discriminator**, acts like a judge. It gets to decide if its input comes from the **G** or from the true training set.



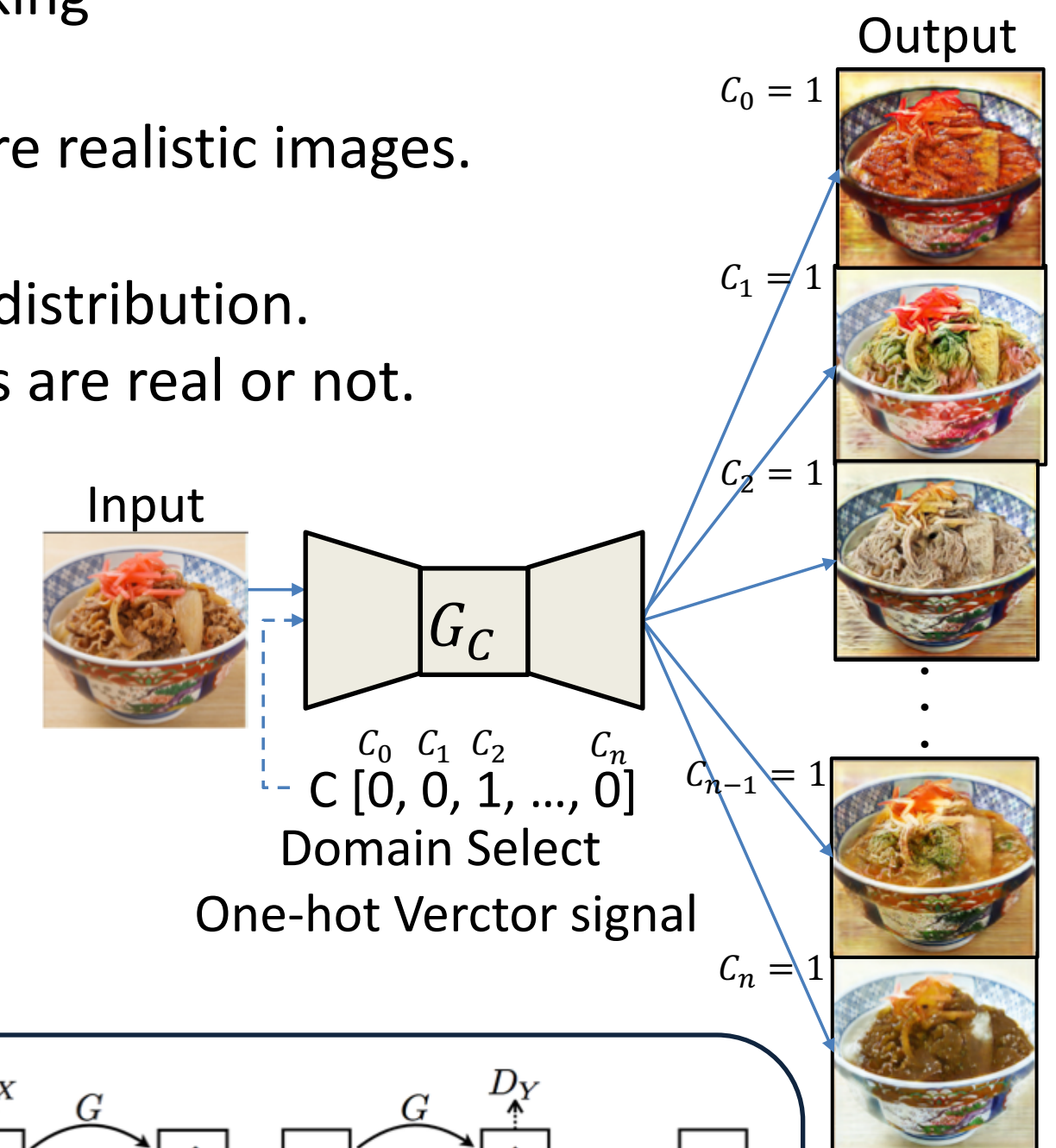
In summary, the game follows with:

- The **Generator** trying to maximize the probability of making the discriminator mistakes its inputs as real.
- The **Discriminator** guiding the generator to produce more realistic images.
- The **Generator** would capture the general training data distribution.
- The **Discriminator** is always unsure of whether its inputs are real or not.

In the perfect equilibrium, as a result,

Objective:

Transfer food images to multiple domains with high quality using the GAN method for dietary images



Method: conditional CycleGAN

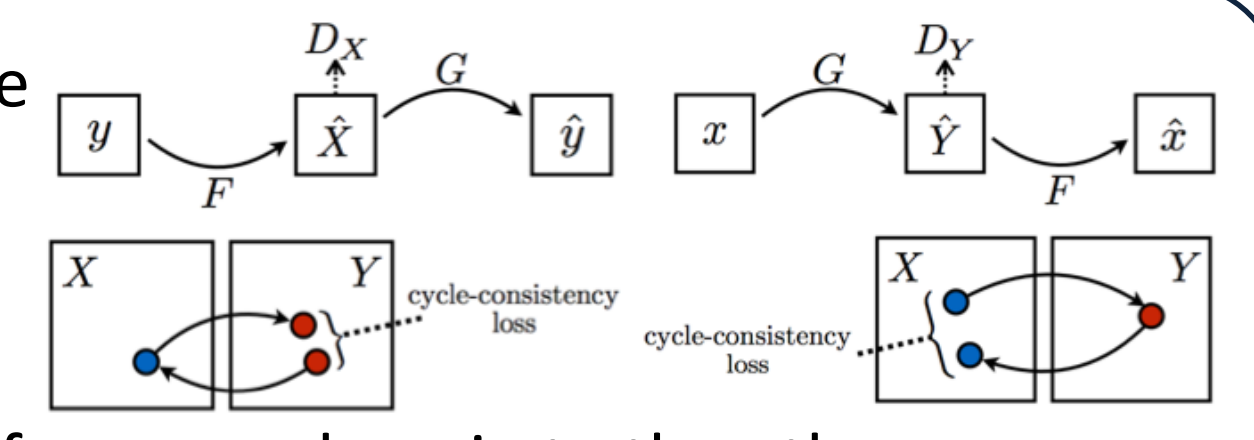
Related work ① : CycleGAN[3] ICCV2017

An approach for learning to translate an image from a source domain X to a target domain Y.

→ They introduce a **Cycle Consistency Loss** to push $F(G(X)) \approx X$ (and vice versa).

Cycle Consistency Loss(CCL):

Capture the intuition that if they translate from one domain to the other and back again they should arrive at where they started.



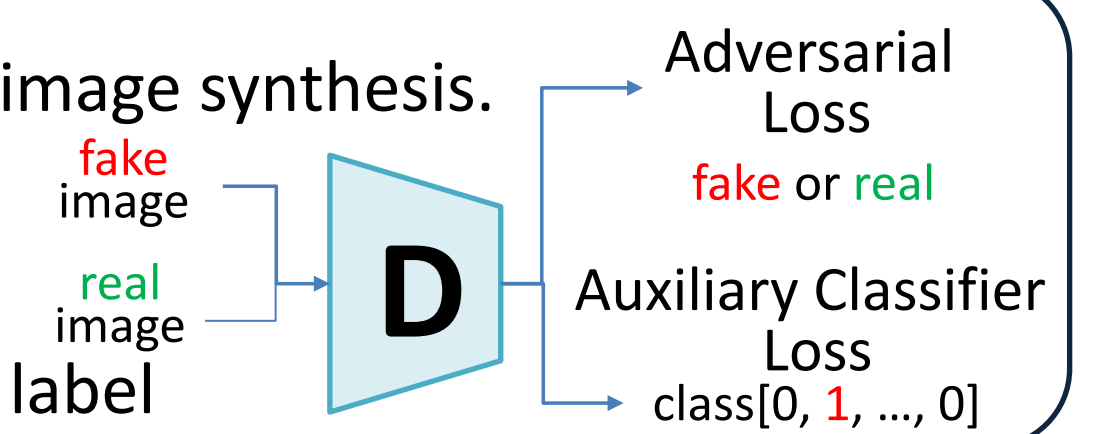
Related work ② : ACGAN[4] ICML2017

New methods for the improved training of GANs for image synthesis.

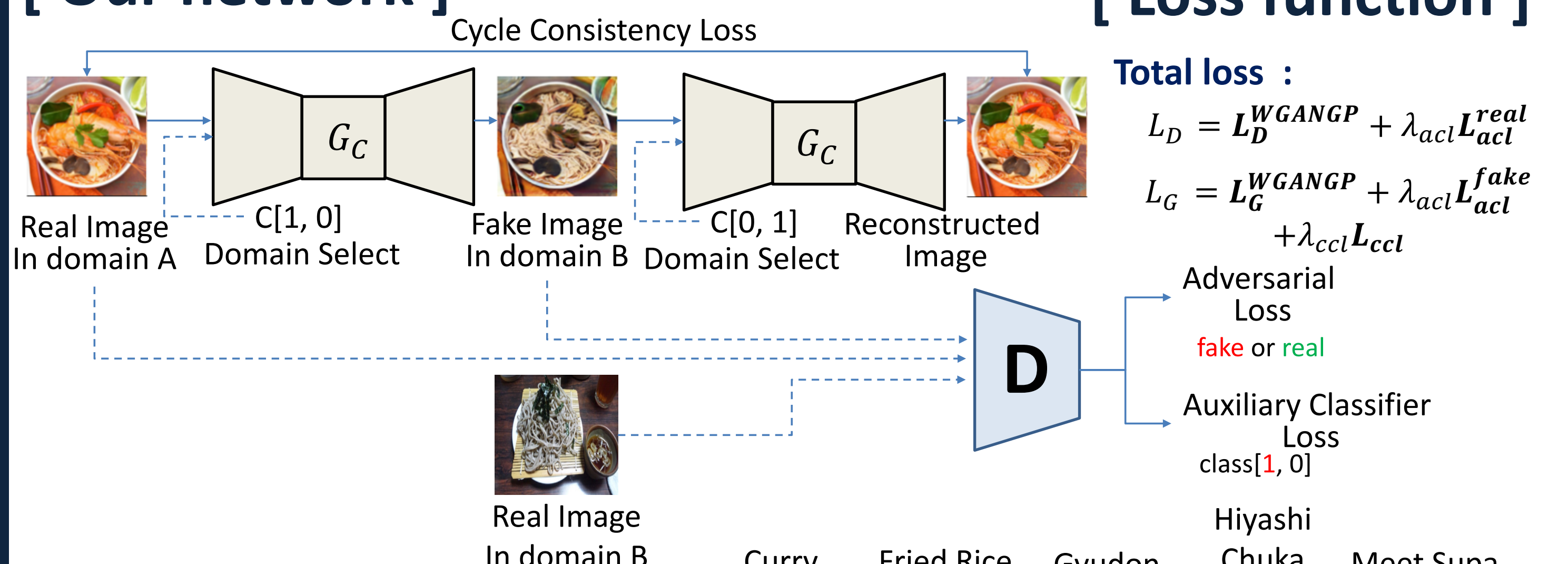
→ They introduce a **Auxiliary Classifier Loss** to make high quality image.

Auxiliary Classifier Loss(ACL):

Every generated sample has a corresponding class label



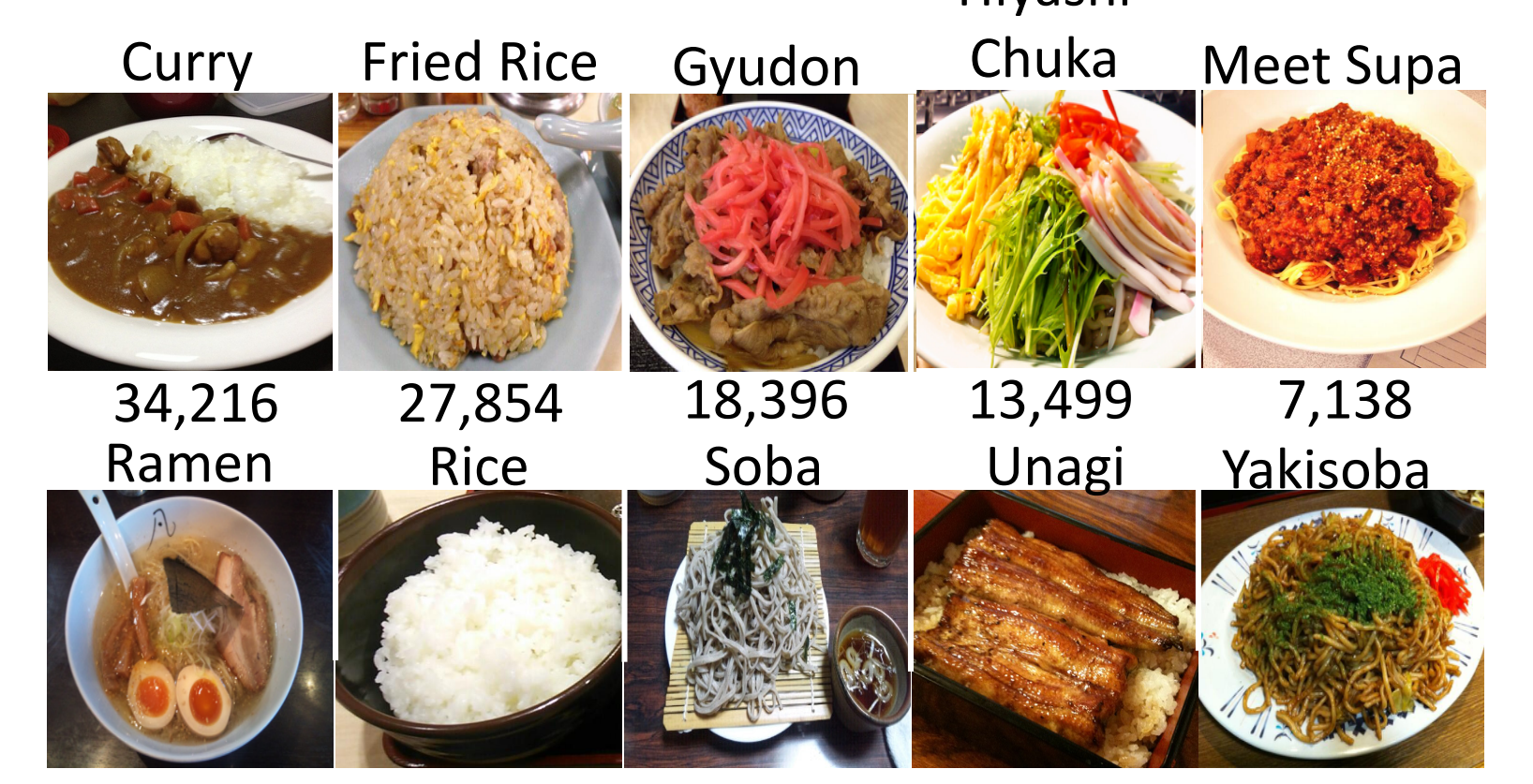
[Our network]



[DATASET]

For training our network, we collected 10 category food photos from the Twitter.

Food 10 categories.
A total of 230,053 images.



Result:



Conclusions

[Calorie Estimation with AR]

- We proposed **food calorie estimation app** with a multi-task CNN using **Augmented Reality**.
- Multi-task learning improved both food calorie and category estimation.

[Food Image Transfer using GANs]

- We proposed **food image transfer using conditional CycleGAN**.
- Conditional CycleGan can **convert multiple domains** while keeping the shape of the food.

[1] T. Miyazaki, G. Chaminda, D. Silva, and K. Aizawa. Image - based calorie content estimation for dietary assessment. In Proc. of IEEE ISM Workshop on Multimedia for Cooking and Eating Activities, 2011.
 [2] J. J. Chen and C. W. Ngo. Deep-based ingredient recognition for cooking recipe retrieval. In Proc. of ACM International Conference Multimedia, 2016.
 [3] J. Y. Zhu, T. Park, P. Isola, A. A. Efros. Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks. In Proc. of IEEE International Conference on Computer Vision, 2017.
 [4] A. Odena, C. Olah, and J. Shlens. Conditional Image Synthesis With Auxiliary Classifier GANs. In Proc. of the 34th International Conference on Machine Learning, 2017.
 [5] I. Gulrajani, F. Ahmed, M. Arjovsky, V. Dumoulin, and A. Courville. Improved Training of Wasserstein GANs. Advances in Neural Information Processing Systems, 2017.