

# Image-Based Food Calorie Estimation Using Knowledge on Food Categories, Ingredients and Cooking Directions

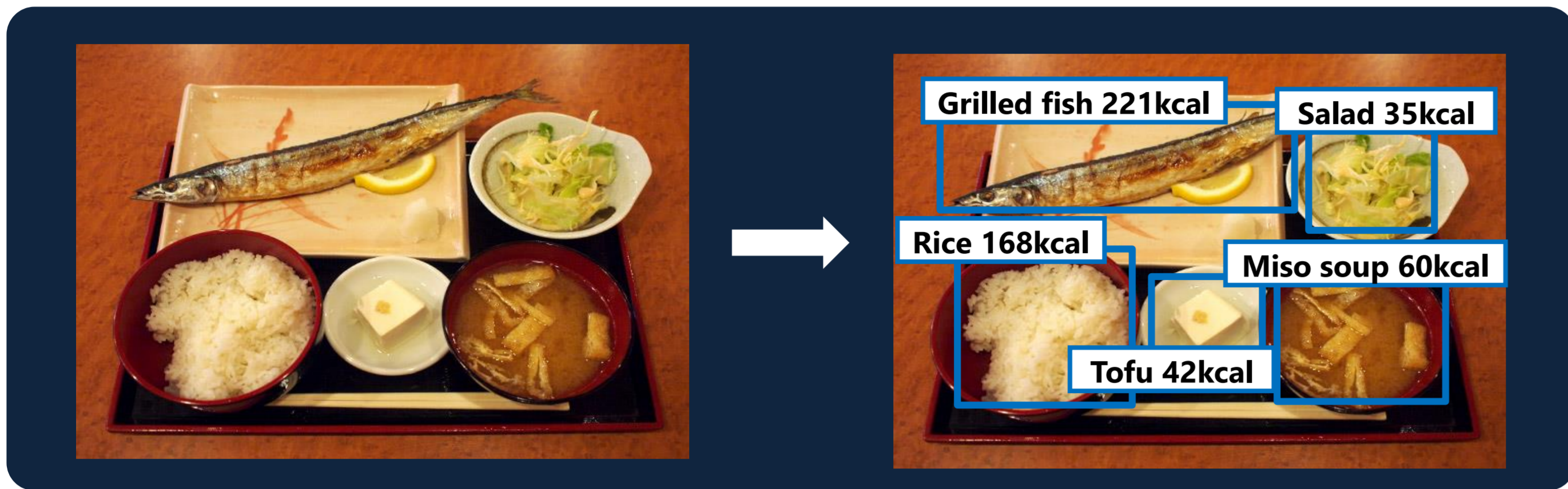
Takumi Ege and Keiji Yanai (The University of Electro-Communications, Tokyo)



## 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



## Method: Multi-task CNN with recipe info.

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

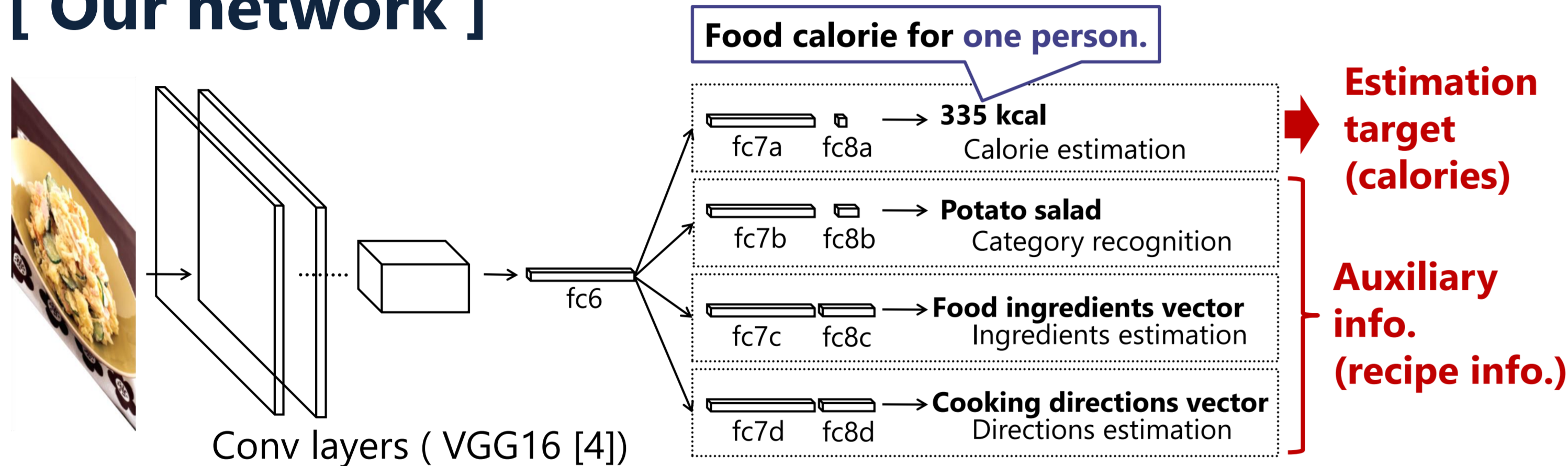
Multi-task estimation of food categories and food ingredients.

Multi-task CNN[3] of food categories and food ingredients

**Multi-task learning improve both task's performance.**

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

## [ Our network ]



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

## [ Loss function ]

Total loss (combining four tasks) :

$$L = \frac{1}{N} \sum_{n=0}^N (L_{cal} + \lambda_{cat} L_{cat} + \lambda_{ing} L_{ing} + \lambda_{dir} L_{dir}) \quad (\lambda \text{ is the weight on the loss.})$$

$L_{cal}$  Calorie estimation loss
 $L_{cat}$  Category recognition loss
 $L_{ing}$  Ingredients estimation loss
 $L_{dir}$  Direction estimation loss

(1) Calorie estimation loss:

$$L_{cal} = \lambda_{re} L_{re} + \lambda_{ab} L_{ab}$$

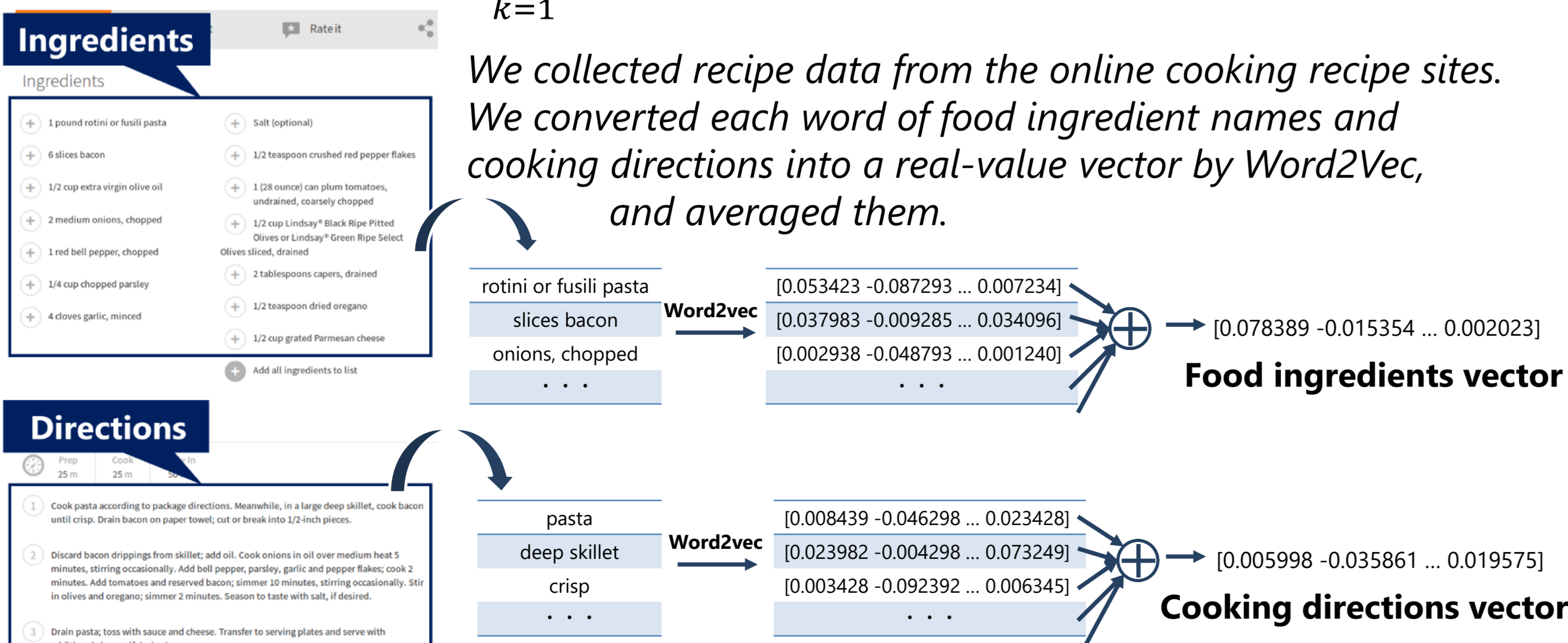
$$\text{relative err. loss } L_{re} = \frac{|y - g|}{g} \quad \text{absolute err. loss } L_{ab} = |y - g| \quad \left( \begin{array}{l} y \text{ is an estimated food calorie.} \\ g \text{ is ground-truth.} \end{array} \right)$$

(2) Food Category loss:

$$L_{cat} = - \sum_{k=1}^n g_k \log y_k \quad (\text{softmax cross entropy})$$

(3) Food Ingredient loss, (4) Cooking Direction loss: (mean squared error)

$$L_{ing, dir} = \frac{1}{2} \sum_{k=1}^n (y_k - g_k)^2 \quad (g \text{ is a ingredient/cooking direction vector.})$$



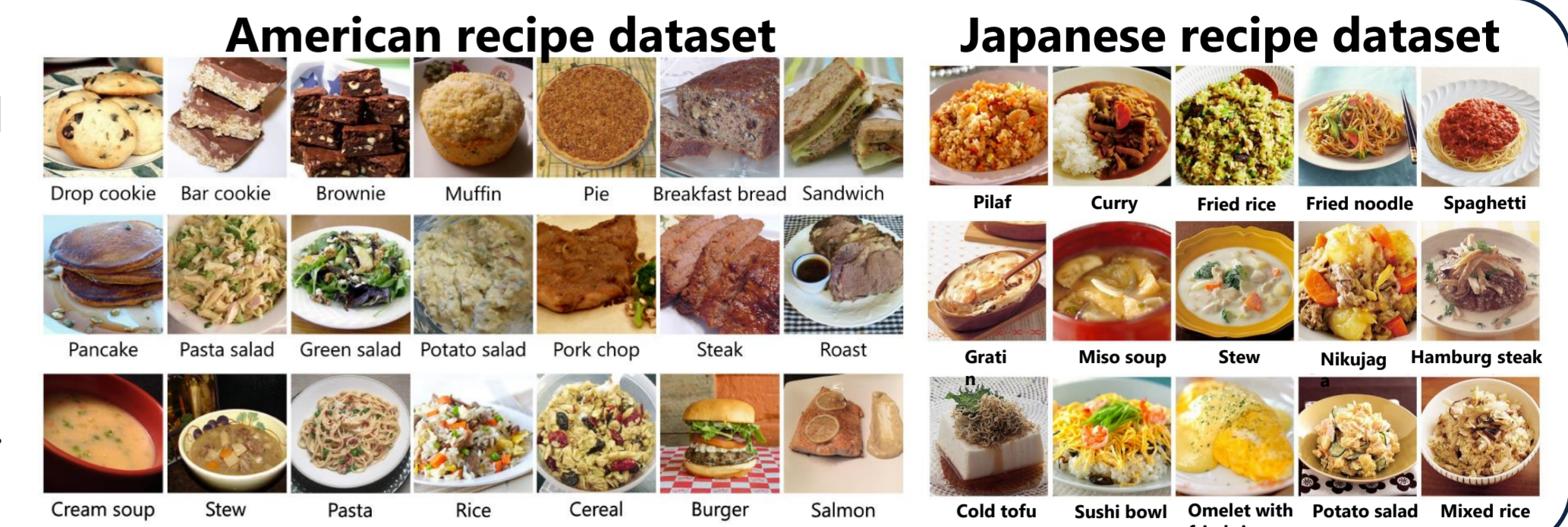
## Experiment : Comparison of single & multi-task

### DATASET

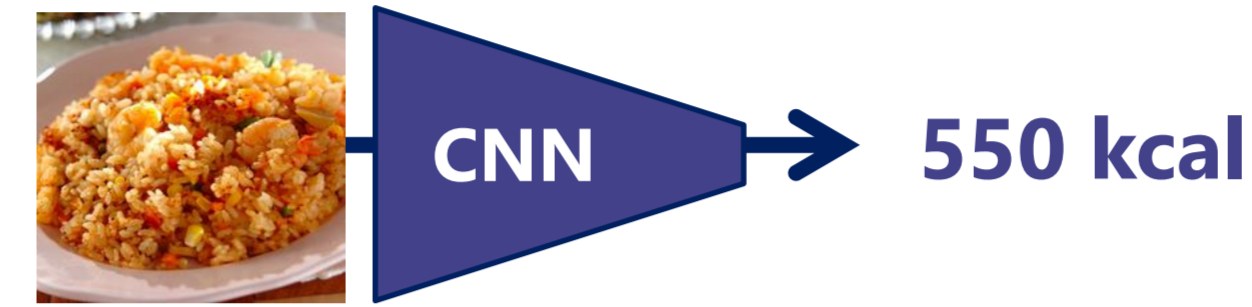
Two kinds of calorie-annotated food photo datasets.

**American recipe datasets**  
**Japanese recipe datasets**

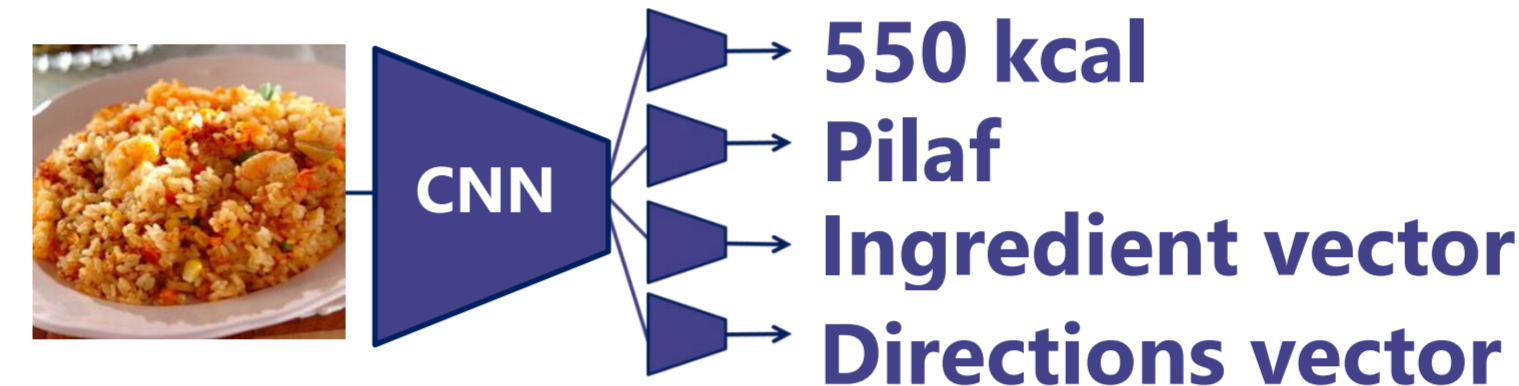
We collected recipe data from the online cooking recipe sites.



### [Single-task CNN (baseline)]

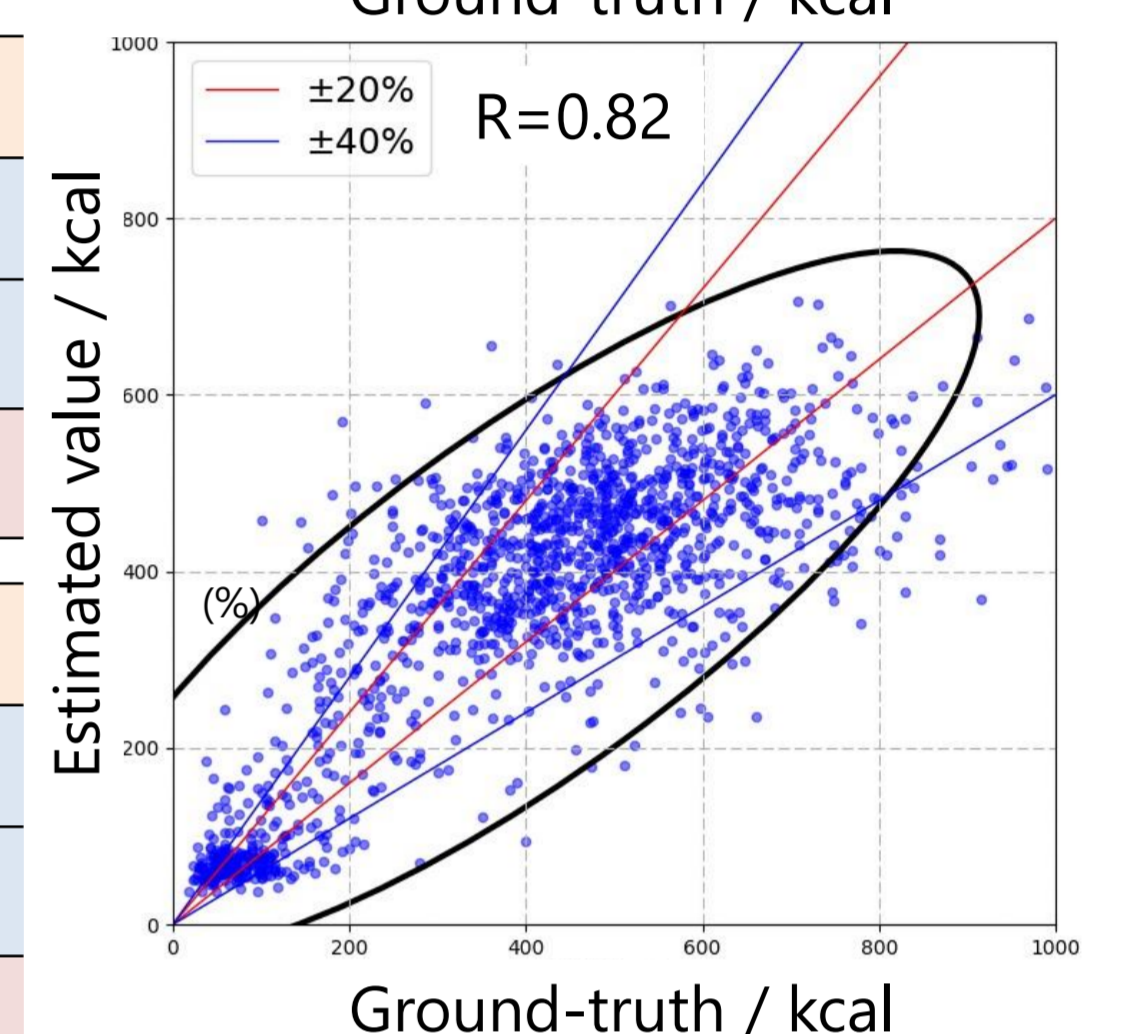
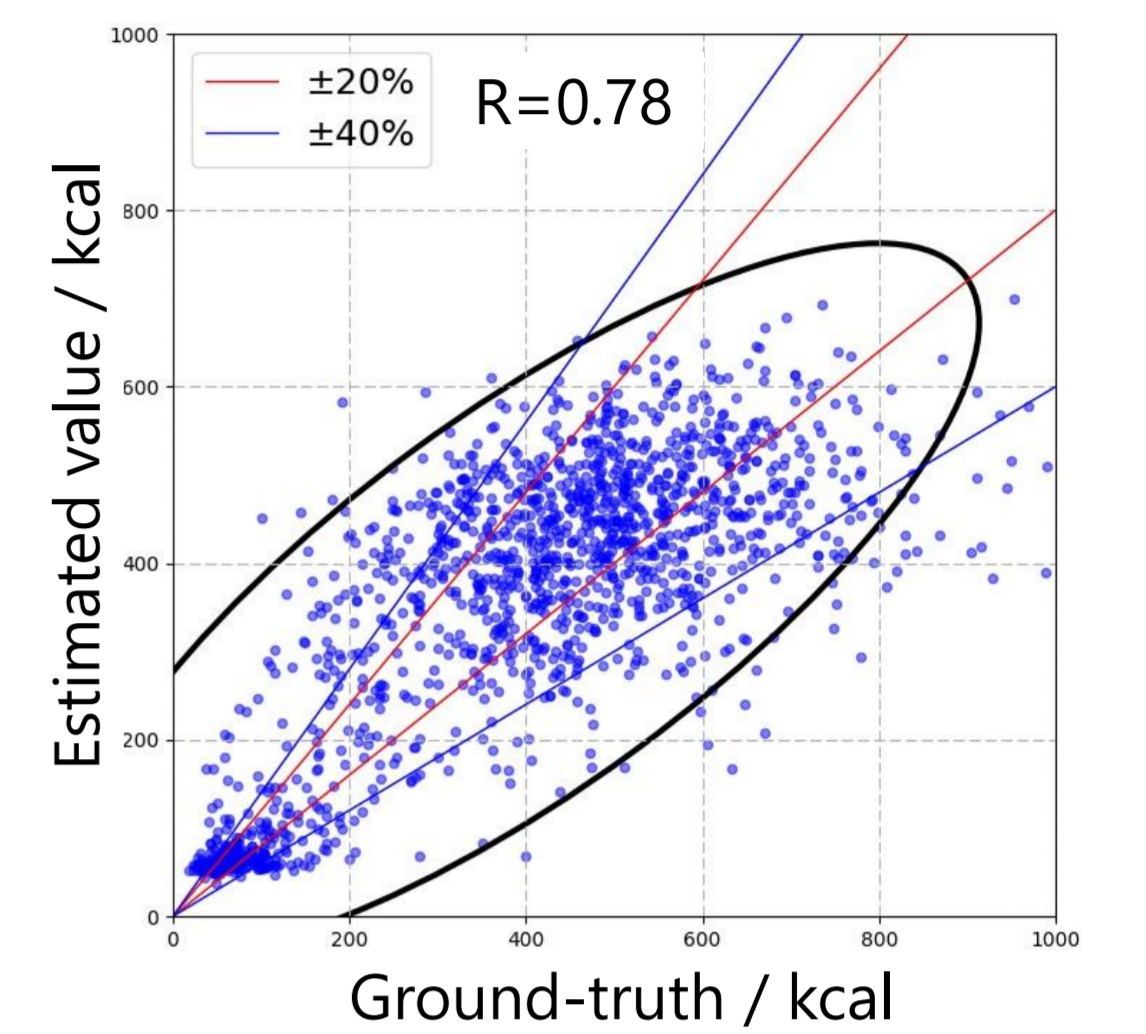


### [Multi-task CNN (proposal method)]

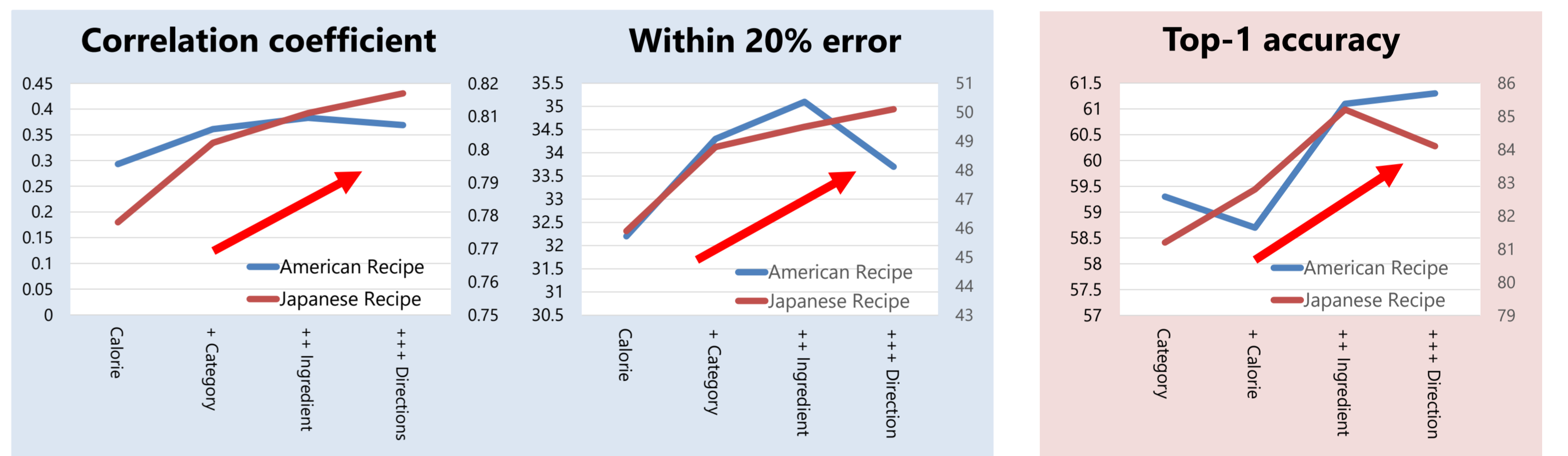


Japanese Recipe	Single	Multi	
Correlation coefficient	0.78	<b>0.82</b>	<b>+0.04</b>
Within 20% error	46 %	<b>50 %</b>	<b>+4 %</b>
Top-1 accuracy (categorization)	81 %	<b>84 %</b>	<b>+3 %</b>

American Recipe	Single	Multi	
Correlation coefficient	0.29	<b>0.37</b>	<b>+0.08</b>
Within 20% error	32 %	<b>34 %</b>	<b>+2 %</b>
Top-1 accuracy (categorization)	54 %	<b>61 %</b>	<b>+7 %</b>



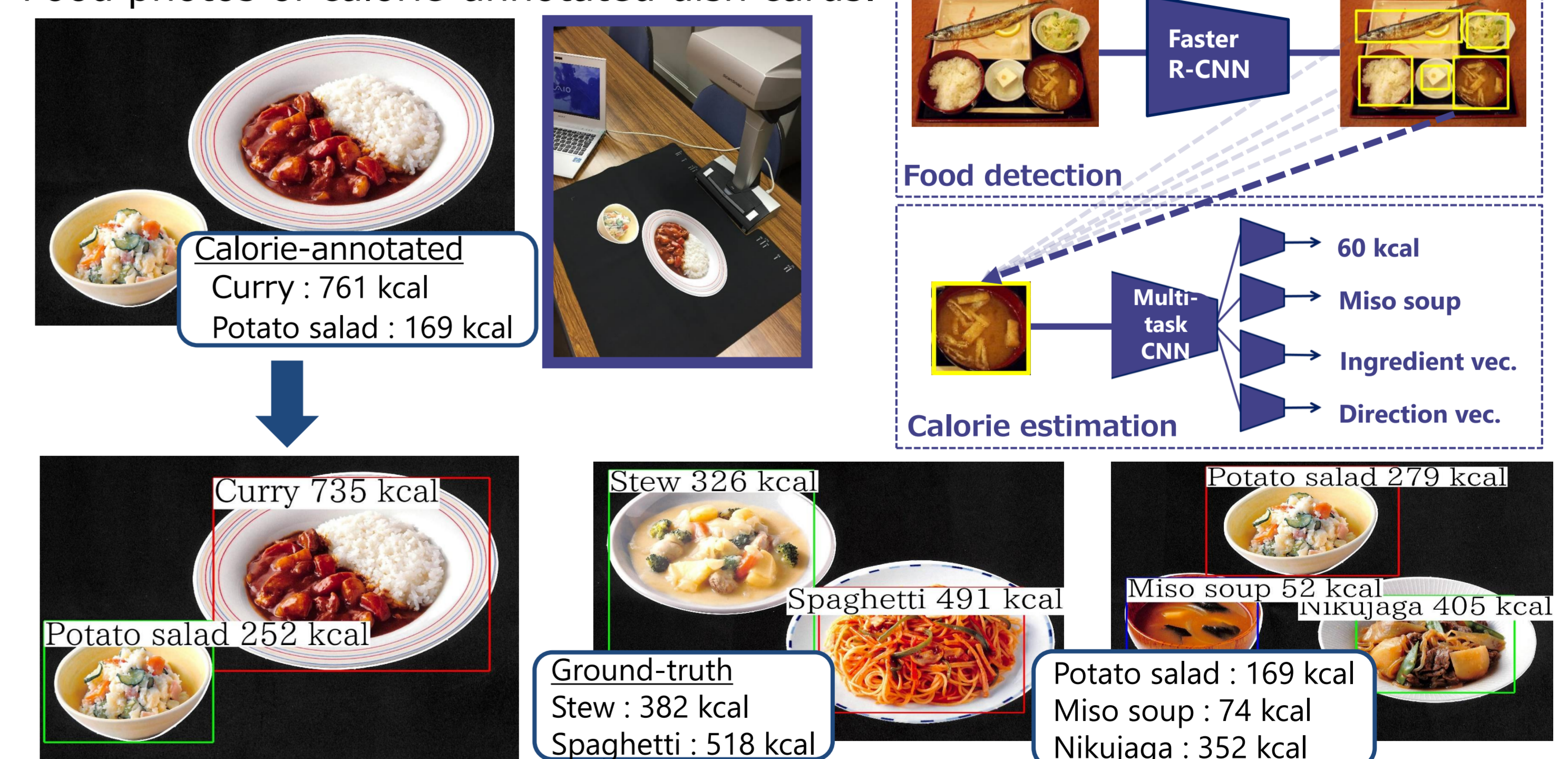
→ **Multi-task learning improves the performance of both tasks.**



	Good	Miso soup	Categorization failure	Bad
Estimated value	432 kcal	35 kcal	243 kcal	486 kcal
Ground-truth	429 kcal	32 kcal	575 kcal	944 kcal
Error	+3 kcal	+3 kcal	-332 kcal	-458 kcal

## Extension : Combining with Food Detector

Food photos of calorie-annotated dish cards.



## Conclusions & Future work

- We proposed **food calorie estimation with a multi-task CNN.**
- Multi-task learning improved both food calorie and category estimation.
- We constructed a **new calorie-annotated food photo dataset.**
- *Food calorie estimation considering the volume of foods.*
- *Construction of large-scale calorie-annotated food photo dataset.*

[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] H. A. Abrar, W. Gang, L. Jiwen, and J. Kui. Multi-task CNN model for attribute prediction. IEEE Transactions on Multimedia, Vol. 17, No. 11, pp. 1949-1959, 2015.

[4] K. Simonyan and A. Zisserman. Very deep convolutional networks for large-scale image recognition. In arXiv preprint arXiv:1409.1556, 2014.

[5] Y. Matsuda, H. Hajime, and K. Yanai. Recognition of multiple-food images by detecting candidate regions. In Proc. of IEEE International Conference on Multimedia and Expo, 2012