Image Recognition of 85 Food Categories by Feature Fusion University of electro-Communications, Tokyo, Japan Hajime Hoashi, Taichi Joutou and Keiji Yanai

Background & Objective

Healthful eating habit is important to avoid obesity and diseases.

If there is a food recommendation system, it is work to keep people in good health.

A food recognition engine is needed to build a automatic food recommendation system.

We propose a food recognition engine using MKL in this paper

It can recognize many kinds of food with a high classification rate.

Food Image Dataset

We built a food image set.

- Includes 85 kinds of food categories.
 - They are common in Japan.
- Has 100 images for each category.



Features Bag-of-Features(BOF) (1) Sample points in three ways

Experiments

Classification rate for 85 categories

Classification by one-vs-rest

Table I **RESULTS FROM SINGLE FEATURES AND FUSION BY MKL**

image features	classification rate
BoF (dog1000)	33.47%
BoF (dog2000)	33.42%
BoF (grid1000)	30.73%
BoF (grid2000)	32.21%
BoF (random1000)	29.61%
BoF (random2000)	30.36%
Color	27.08%
Gabor (3×3)	23.60%
Gabor (4×4)	25.35%
Gradient (180, 1×1)	3.87%
Gradient (180, 2×2)	10.12%
Gradient (180, 4×4)	17.04%
Gradient (180, 8×8)	19.44%
Gradient (360, 1×1)	5.67%
Gradient (360, 2×2)	13.15%
Gradient (360, 4×4)	20.87%
Gradient (360, 8×8)	21.84%
SVM (uniform)	60.87%
MKL (mean- χ^2 distance)	62.52%

***85** food categories

Using MKL(Multiple Kernel Learning)

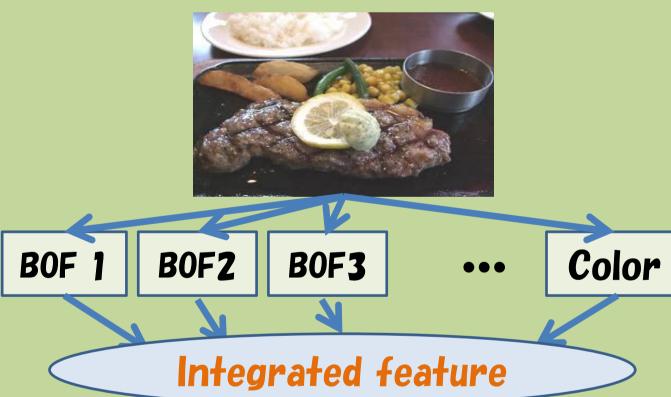
Achieved the highest performance for Caltech-101/256 [Varma et al. ICCV2007] Integrate many kinds of image features.

Method

♦ Feature fusion

by Multiple Kernel Learning (MKL)

- Bag-of-Features (BOF) : 6 kinds
- : 2 kinds Gabor Features
- Gradient Features : 8 kinds
- Color histogram : 1 kinds



(DoG Keypoints, Random sampling, Grid sampling)

- (2) Describe local patterns around the sampled points with SIFT [Lowe 2004]
- (3) Generate codebooks by K-means (size of a codebook: k=1000, 2000)

(4) Convert images into BoF vectors by voting to nearest codewords







Gabor Features

6 directions * **4** frequencies Extract from 3x3 and 4x4 grids Totally 216dim. & 384dim.

Color histogram

64 colors Extract from 2x2 grids Totally **256**dim.

Gradient Features

(1)Extract edges by Canny edge detector (2)Calculate gradient on edges (3) Build histograms by voting gradient orientation Extract from 1x1, 2x2, 4x4, 8x8 grids



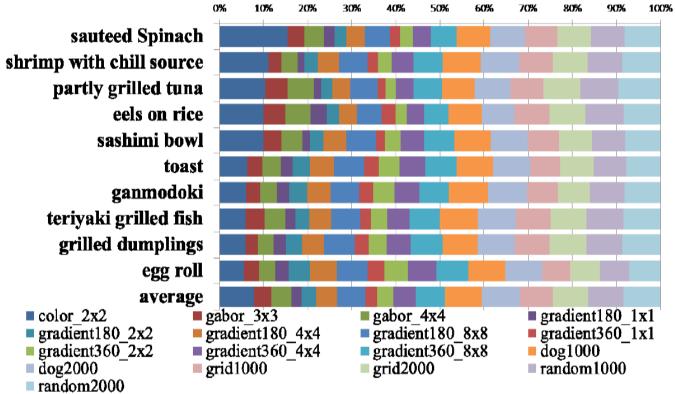
The best five and the worst five

top 5	category	recall	worst 5	category	recall
1	soba noodle	95%	1	Ganmodoki	17%
2	eels on rice	94%	2	sandwich	24%
3	sauteed spinach	93%	3	toast	30%
4	miso soup	92%	4	grilled eggplant	30%
5	rice	90%	5	simmered pork	31%





Weights estimated by MKL



Classification rate within top N



DoG Keypoints Random sampling

Grid sampling

ultiple Kernel Learning

Is an extension of a SVM.

Can handle "a combined kernel" which is a linear combination of kernels.

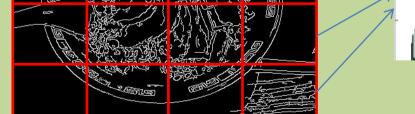
Can estimate Kernel weights and SVM model parameters simultaneously.

Can integrate features by assigning one feature to one kernel.

Combined Kernel

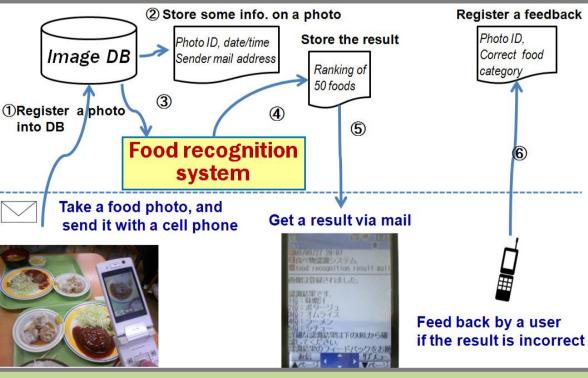
BoF1 kernel..... Gabor1 kernel..... Color kernel

Combined Kernel $\mathbf{k}(\mathbf{x}_i, \mathbf{x}_i) =$ Kernel weight (to be estimated by MKL)





Prototype System



You can try it !

Please Send a food photo to food@mm.cs.uec.ac.jp

- And you will get a recognition results in a Japanese mail.
- Sorry !! But you can see a result with photos by clicking a URL in the mail !)

0.6 + 0.5 +			nk 3			
	\sum		66 %			
.3 —	Rank					
.2 —	62.5	2%		 	 	
0.1						

Conclusions

Propose a food recognition engine with MKL-based future fusion Achieved 62.52% classification rate More than 80.66% when allowing three candidates

Future work

More than 100 categories More features (e.g. shape context, CSIFT)

Other features (e.g. date/time, GPS info.) Implement a food recommendation system

Multi-object recognition