

Twitter Food Photo Mining and Analysis for One Hundred Kinds of Foods



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Twitter Realtime Food Photo Mining System (mm.cs.uec.ac.jp/tw/)

- What kinds of foods are being eaten in Japan ?

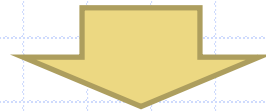
Real-time Geo-Tweet Food Photo Mapping System



Background



- *Various kinds of photos are posted to microblogs (Twitter) every minutes*
- *Twitter photos are uploaded with text messages (Tweet messages)*



- *Microblogs such as Twitter can be regarded as another tagged photo source than Flickr.*

Twitter(TW) vs Flickr(FL)

- **Same points:**

- Photos with texts
- Easy to collect data via WebAPI

- **Different points:**

- On-the-spot \Leftrightarrow at home (from mobile vs from PC)
- TW:30million/day vs FL:3million/day
- TW msg is not tag. Free msg.
 - FL: Tags describes contents of photos in general
 - TW: Attached msgs do not always describe the content of the photos \Rightarrow needs for image analysis

Twitter Food Photo Mining

- **Twitter Photos represent the current state of the world !**
- **Mining food photos from Twitter is the best way to get to know what people are eating now !!!**
 - **What kinds of ramen is being eaten at this moment in Japan / over the world?**
- **We propose**

Twitter Food Photo Mining



VS



Ramen

Curry

- **Which food is the most popular in Japan?**
 - **“Ramen vs Curry” problem ⇒ very controversial**
 - **I would like to put a period to this controversy by Twitter food photo mining !!!**

Related work on Twitter photo

- **Event photo mining from geotagged Tweet photos [ICME WS 2013]**
 - **Text-based event detection + photo sele.**



The results of detected event photos in 2012

Related works on TW photos(2)

- Classifying “visual” / “non-visual” tweets by generic methods [Chen et al. MM13]

Visual
70.5 %

陈建斌怎么看怎么还是曹操的样子啊! (No matter how I look at it, Chen Jianbing looks like Cao Cao!)



Non-visual

可恶的蚊子，我要杀了你!
(Horrible mosquitoes, I will kill you!)



- Brand image mining [Gao et al. ICMR 2014]
 - Supervised logo detector



Approach for food photo mining

- **Two-step food photo selection**
 - [1] **Keyword-based tweet selection**
 - [2] **Image-based photo selection**
 - **Generic food/non-food classification**
 - **Specific food classifiers (100 kinds)**



Targets: 100 kinds of foods in the UEC-Food100 data set

- Includes common foods in Japan
- Has more than 100 images / category



Ramen



Curry



[1] Keyword-based selection

- **Select the photo tweets the messages of which include any of 100 kinds of food names**
 - **In the experiments, we used Japanese food names.**
 - **We tried query expansion as well.**

e.g.) I came to eat **ramen** noodle.
Very delicious **ramen** !!!
Ramen is my life.

[2] Two-step image-based selection

[2-1] Food/non-food classification

- Remove non-food photos and select only food photos

[2-2] Specific food classifiers

- Extended version of FoodCam recognition engine. 1-vs-rest 100-class classification
- Select the food photo if the corresponding food category is ranked within the top five.



I ate **sushi** !

100-class
food classification

[top-5] Pizza, ramen,
curry, **sushi**, tempura

Sushi
Photo

Food recognition method

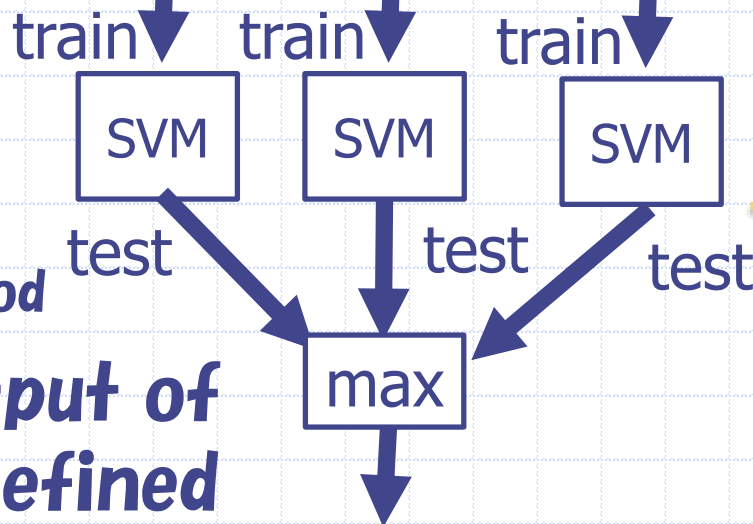
- **Local patch + Fisher Vector + linear SVM**
 - **Color patch, HOG patch**
 - **Color: 24 dim HOG: 32 dim**
 - **dense sampling**
 - **GMM: K=64**
 - **Spatial Pyramid: 1x1 + 2x2**
 - **Improved Fisher Vector [Perronnin et al.2010]**
 - **Color: 15360 dim, HOG: 20480 dim**
 - **Classifier**
 - **Linear SVM**

[2-1] Food / non-food Classifier

- Train 13 linear SVMs
 - Pos.: UEC-FOOD 100
 - Neg.: typical irrelevant photos
 - Inside / outside restaurant
 - Menu, people eating, ...



13 groups




- Classify of food / non-food
 - The maximum value of output of 13 classifiers with pre-defined threshold values

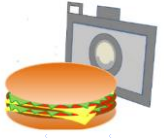
Foodness score



[2-2] 100-class specific food category recognition

- **100-class food classification engine**
 - **Extended version of FoodCam**
[Kawano et al, MTA14]
 - **Very fast (0.025 sec. / image)**
 - **Multi-threaded implementation optimized for quad core CPU**
 **Suitable for big data recognition**
 - **HOG-FV + Color-FV + 1-vs-rest linear SVM**

<http://foodcam.mobi/>

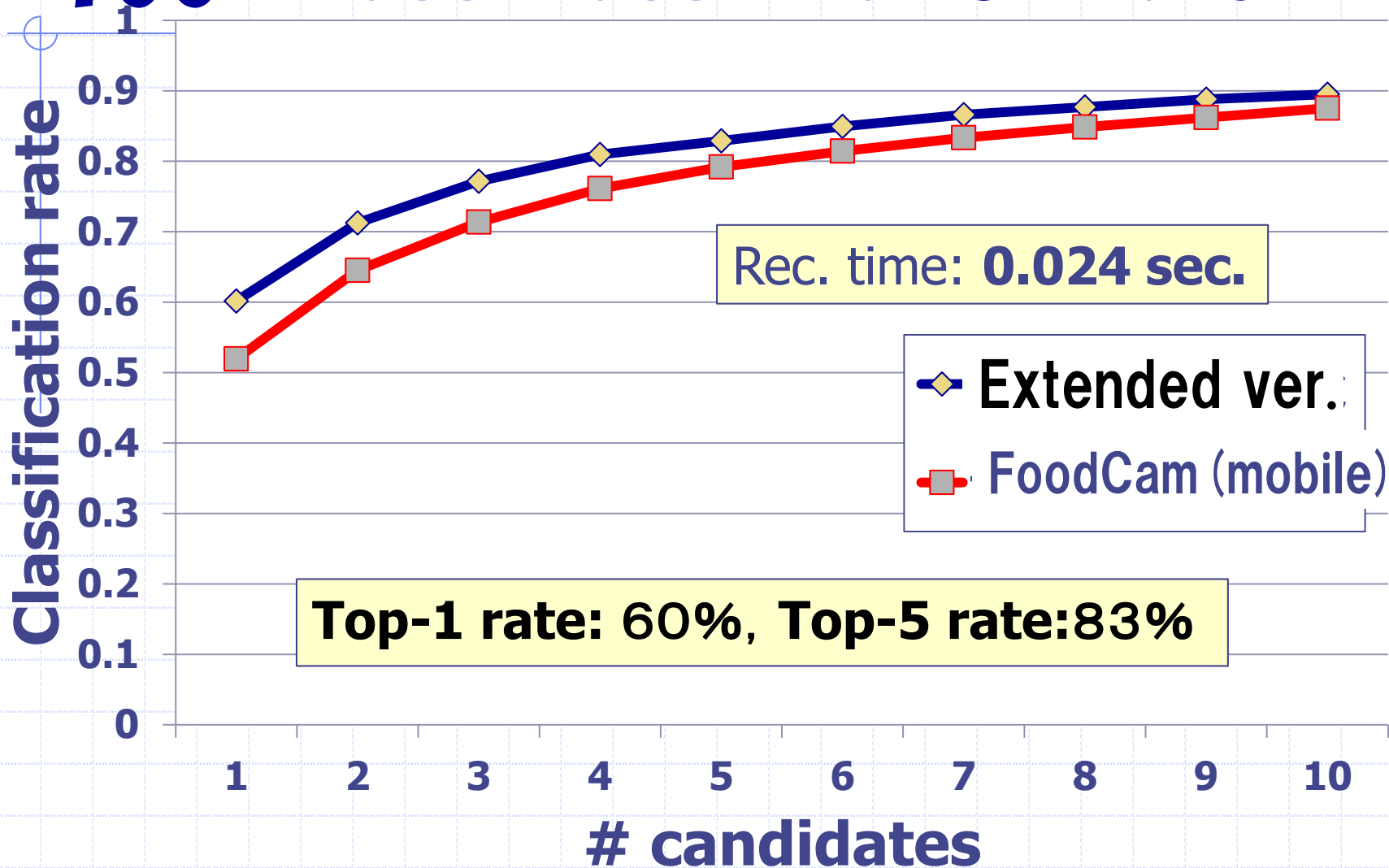


FoodCam : [Kawano et al. MTA13]

- Real-time mobile food recognition
Android application



100-Class classification rate



Experiments

- **Collect photo tweets via Twitter Streaming API**
 - **From 2011/5 to 2013/8**
 - **About one billion tweets**
- **Search for the tweets including any of 100-food names (in Japanese)**
 - **1.7 million** ⇐ **Apply food image analysis**
- **Food / non-food classifier**
+ 100-food classifier
 - **470,335 food photos**

Evaluations on five kinds of representative food

- Num. of obtained food images
- Precision (random sampling of 300 images)
 - (1) Only keyword search
 - (2) Keyword + food/non-food classifier
 - (3) Keyword + specific food classifier
 - (4) All (kw+food/non-food+specific) **proposed**
- Geographic analysis with geotagged photos
 - Ramen vs Curry

Twitter food photo ranking

rank

foods

#photos



Ramen noodle is the most popular food in Japan.
I have solved “ramen vs curry” problem !!!

Precision of the top 5 foods

Food	(1) KW	(2) f/n	(3) spec.	(4) ALL
ramen	275,652 72.0%	200,173 92.7%	84,189 95.0%	80,021 99.7%
curry	224,685 75.0%	163,047 95.0%	62,824 97.0%	59,264 99.3%
sushi	86,509 69.0%	43,536 86.0%	48,019 72.3%	25,898 92.7%
tsukemen	33,165 88.7%	24,896 96.3%	28,846 93.7%	22,158 99.0%
omelet	34,125 90.0%	28,887 96.3%	18,370 98.0%	17,520 99.0%

Only keyword search (Ramen noodle) (72.0%)



After applying food/non-food classifier (92.7%)



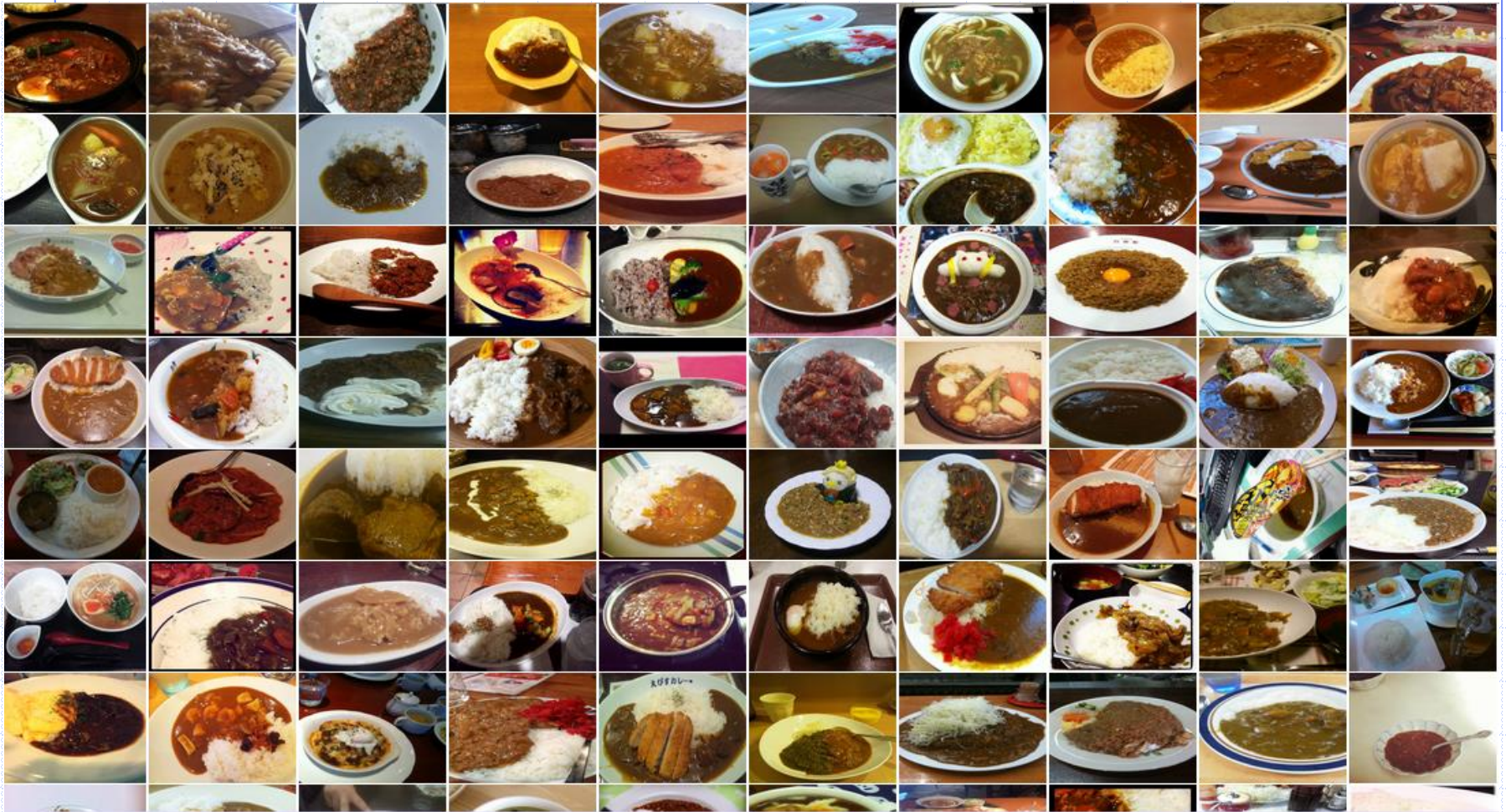
After applying 100-class food classifier (final) (99.7%)



Only keyword search (curry) (75.0%)



Final results (curry) (99.3%)



Some interesting findings

- Letters or drawings are sometimes drawn on omelets with Ketchup



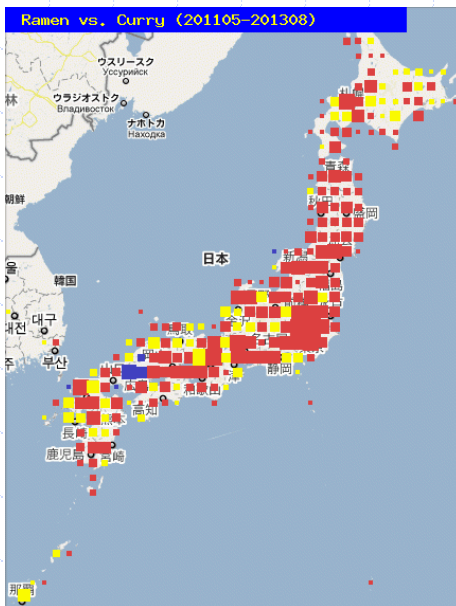
- Fast-foods such as hamburger (rank 30th) and beef bowl (rank 27th) are ranked lower, since their appearance is always the same.



Not worth posting fastfood photos to Twitter

Geographical analysis on ramen vs curry

12.6% of the obtained food photos have geotag.



Whole year



Dec. (winter)
Ramen is popular.



Aug. (summer)

Curry gets more popular
than ramen in many areas.

● **Ramen**

● **Curry**

Real-time Food Collection

- **Monitor the Twitter stream**
 - **Photo Tweet**
 - **Text including any of 100 food names**
 - **13 candidate photo tweets / minute on avg.**
 - **Download: 2~3sec. , recognition: ~1sec.**
 - **Single machine is enough !**
- **Recognize 20,000 photos and find 5,000 food photos from the TW stream everyday in our lab**

Demo visualization system

- **Map each food photo on an online map with online clustering [Yanai ICMR2012]**
 - **Geotagged Tweets**
 - **Non-geotagged Tweets for which GeoNLP can assign locations based on text msg.**
- **Overlay a food photo on the Streetview**
 - **Finding “ramen noodle shop” game !**

Additional work for more Ramen

Photos: **Finding** “**Koike-san**”

Koike-san is a Fujiko-Fujio comic's character who loves “ramen noodle”.

*He is **always** eating “ramen noodle” when he appears in the comic.* (Wikipedia)



Finding “Koike-san” on Twitter

-query expansion based on user, loc & word-

- Pick up the top- k users who frequently post ramen photos.

- ($k=30$ in the experiments)

- Apply food classifiers to all the photos “Koike-sans” posted.

[raumenbot](#),657,0
[toyamamenrui \(map\)](#),128,128
[oishii_bot2](#),107,0
[Ramen_Bot](#),90,0
[daikoubutsu_bot](#),88,0
[shibumen](#),65,0
[vientM535i \(map\)](#),50,36
[oishii_bot](#),45,0
[kido_maru](#),42,0
[ishikawamenrui \(map\)](#),42,42
[shomax96 \(map\)](#),36,33
[rAsAmAya \(map\)](#),36,28

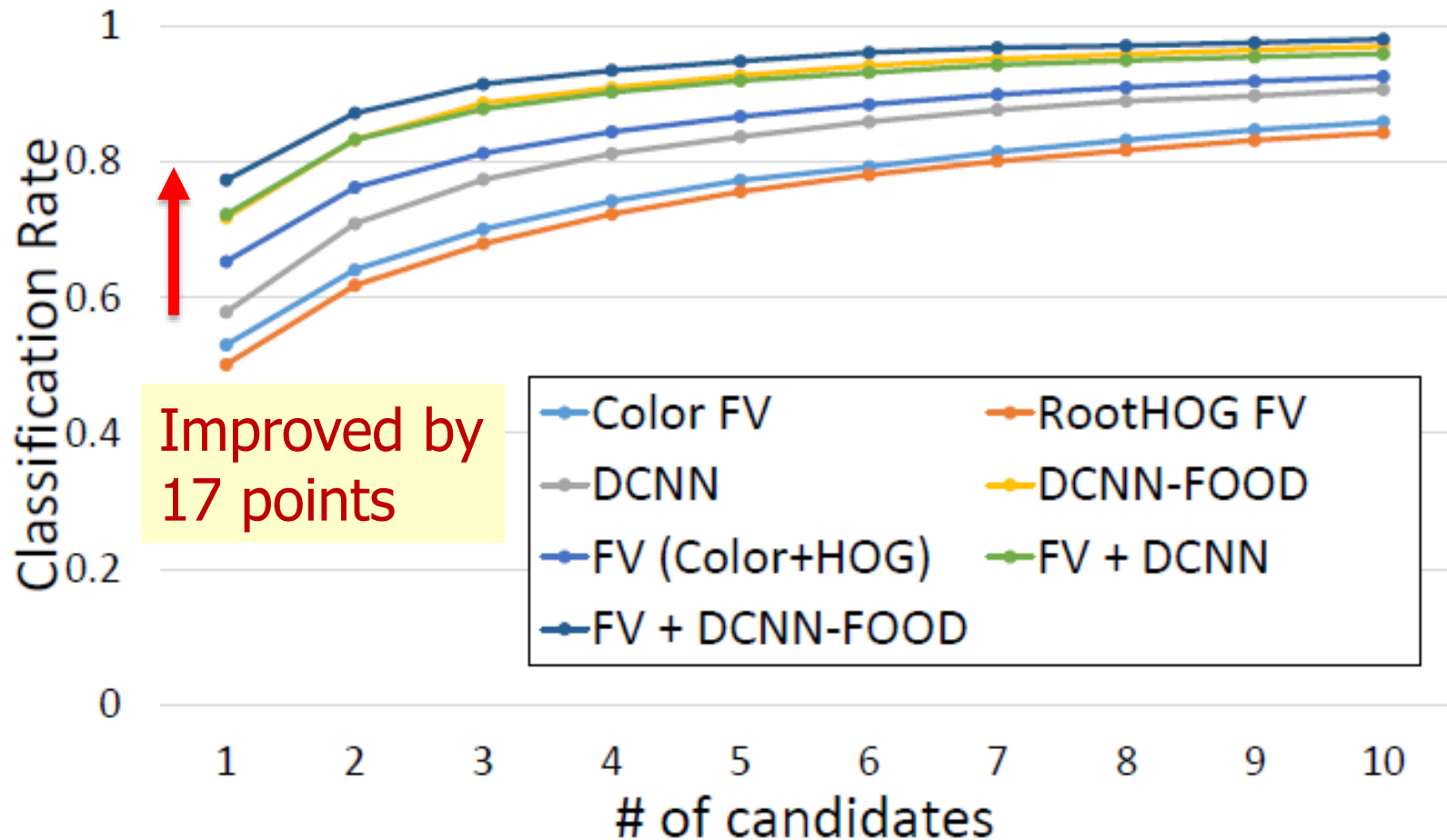
- Other methods: Finding “Koike-san” places , Finding co-occurrence words

More “ramen / curry / sushi”

- Precision is not as good as the results by keyword-based candidate selection.

	Frequent user “Koike-sans”	Frequent co-occurrence word	Frequent places
Ramen	6050 58.0 %	5851 68.5 %	594 44.0 %
Curry	3163 23.5 %	2806 49.0 %	313 25.5 %
Sushi	2474 13.5 %	1591 41.5 %	991 17.0 %

State-of-the-art (DCNN-based) (presented at MDBA WS)



Top-1 rate 77.4%, Top-5 rate 94.8%

Conclusions

- **Food Photo Mining from Twitter Photo data / the Twitter stream.**
- **Have completely solved the “ramen vs. curry” problem.**
 - **Note that only in summer searson, Curry becomes more popular than Ramen.**
- **Real-time system (demo)**

Future work

- **One million “ramen noodle photo dataset”**
 - **For all the “Ramen” fans over the world.**
- **Methods for collecting more Ramen !**
 - **Use DCNN-based classifier**
 - **Improve “Koike-san” methods**
- **Extension to World-wide foods**

Thank you for your attention !

Real-time Geo-Tweet Food Photo Mapping System

豊島区, 東京都
近辺

地図 航空写真

RT @byassist: 昭和歌謡ショー@庚申塚の中華そば。煮干し出汁に慣れた舌には物足りなくもあるが、これはこれで昔から馴染みの正統派無化調の醤油ラーメン。素直に美味しい♪ [nlp #sg]

Google
豊島区, 東京都 Mon Dec 01 12:59:47 2014 JST

韓国 豊島区, 東京都 和歌山 高松 高知 大分 松山

[0]

[1]

[2]

[3]

[4]

[5]

1	ramen noodle	80021
2	curry	59264
3	sushi	25898
4	dipping noodle	22158
5	omelet with fried rice	17520
6	pizza	16921
7	jiaozi	16014
8	Japanese-style pancake	15234
9	steamed rice	14264
10	sashimi	13927
11	hambarg steak	11583
12	beef stake	9503
13	takoyaki	9004
14	fried rice	8383
15	fried noodle	7905
16	oden	7453
17	toast	6350
18	cutlet curry	6339
19	tempura	5905
20	rice ball	5462
21	gratin	5223
22	croquette	4837
23	stew	4797
24	sashimi bowl	4730
25	chicken-'n'-egg on rice	4513
26	tempura bowl	4464
27	beef bowl	4285
28	spicy chili-flavored tofu	4081
29	yakitori	3829
30	hamburger	3662
31	chilled noodle	3473
32	sukiyaki	3408
33	miso soup	3295



34	fish-shaped pancake with bean jam	3281
35	pork cutlet on rice	3188
36	omelet with grilled minced meat	2592
37	bibimbap	2368
38	spaghetti	2171
39	lightly roasted fish	2162
40	seasoned beef with potatoes	2129
41	natto	2094
42	spaghetti with meat source	1994
43	steamed egg hotchpotch	1843
44	egg sunny-side up	1635
45	croissant	1579
46	udon noodle	1500
47	simmered pork	1443
48	mixed sushi	1371
49	pork miso soup	1229
50	ginger-fried pork	1158
51	potato salad	1150
52	egg omelet	1146
53	eels on rice	1071
54	egg roll	1058
55	sweet and sour pork	1049
56	fried shrimp	1049
57	sauteed vegetables	1040
58	shrimp with chill source	1003
59	cabbage roll	965
60	mixed rice	901
61	pilaf	891
62	soba noodle	880
63	potage	816
64	hot dog	795
65	chicken rice	736
66	wiener sausage	577

67	dried fish	563
68	steamed meat dumpling	561
69	french fries	561
70	beef ramen noodle	555
71	sandwiches	551
72	cold tofu	517
73	boiled chicken and vegetables	352
74	sirloin cutlet	331
75	nanbanzuke	323
76	fried chicken	314
77	stir-fried beef and peppers	312
78	roll bread	288
79	roast chicken	263
80	macaroni salad	239
81	boiled fish	228
82	kinpira-style sauteed burdock	225
83	tempura udon	213
84	raisins bread	205
85	goya chanpuru	198
86	green salad	145
87	chinese soup	141
88	Japanese tofu and vegetable chowder	137
89	salmon meuniere	96
90	grilled pacific saury	84
91	chip butty	76
92	fried fish	72
93	begitable tempura	71
94	tensin noodle	69
95	ganmodoki	34
96	grilled salmon	25
97	sauteed spinach	12
98	teriyaki grilled fish	3
99	grilled eggplant	2
100	pizza toast	0

2.1 group and some food categories




noodles	udon noodles, dipping noodles, ramen
yellow color	omlet, potage, steamed egg hotchpotch
soup	miso soup, pork miso soup, japanese tofu and vegetable chowder
fried	takoyaki, japanese-style pancake, fried noodle
deep fried	croquette, sirloin cutlet, fried chicken
salad	green salad, sauteed vegetables, vegetable tempura
bread	sandwiches, raisin bread, roll bread
seafood	sashimi, sashimi bowl, sushi
rice	rice, pilaf, fried rice
fish	grilled salmon, grilled pacific saury, dried fish
boiled and seasoned	seasoned beef with potatoes simmered ganmodoki seasoned beef with potatoes
sauteed	sauteed vegetables, goya chanpuru, kinpira-style sauteed burdock
sauce	stew, curry, stir-fried shrimp in chili sauce

2.1 group and some food categories




noodles	udon noodles, dipping noodles, ramen
yellow color	potato, potato, steamed egg hotpot
	
boiled and seasoned	seasoned beef with potatoes simmered ganmodoki seasoned beef with potatoes
sauteed	sauteed vegetables, goya chanpuru, kinpira-style sauteed burdock
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2.1 group and some food categories

noodles	udon nooles, dipping noodles, ramen
yellow color	omlet, potage, steamed egg hotchpotch
	
ushi	 grilled salmon, grilled pacific saury, dried fish
boiled and seasoned	sesoned beef with potatoes simmered ganmodoki sesoned beef with potatoes
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