



Automated food log

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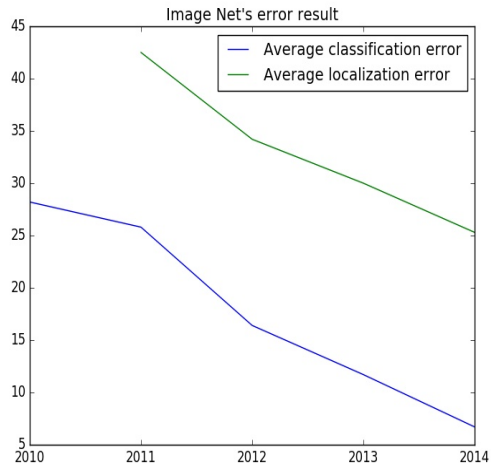
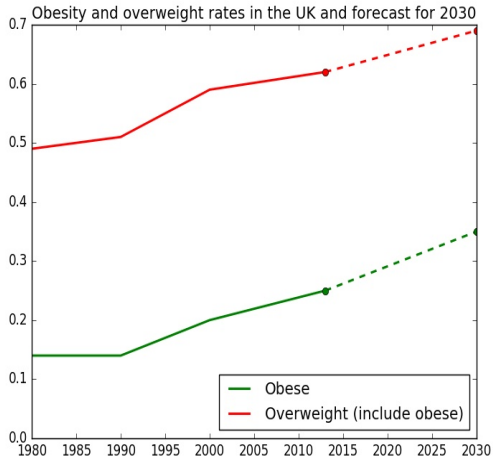
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Why a food log analysis system?





Overall process

Roughly copying the procedure of FoodLog [1, 2, 3, 4, 5]:

- Generate a relevant dataset
- Extract characteristic
- Machine learning
- For a new picture from a user, classify and estimate intake

Focus on localisation and classification



Dataset

Name	Release date	Number of pictures	Type of food	Number of classes	Multiple food items
PFID [6]	2009	4545	American fast-food	101	No
UEC FOOD 100 [7]	2012	14361	Japanese	100	Yes
FIDS 30 [8]	2013	971	Fruit	30	No
ETHZ Food-101 [9]	2014	101 000	European	100	No
FooDD [10]	2015	3000	Fruit	23	Yes
UEC FOOD 256 [11]	2015	31395	World	256	Yes

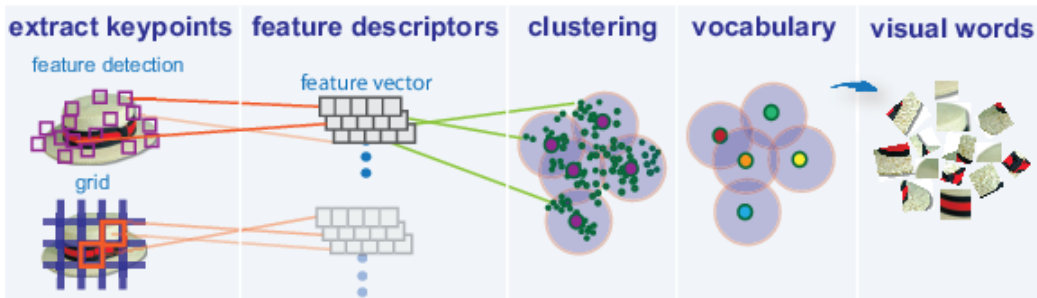
Example of multi-items



Feature description

Bag of visual words

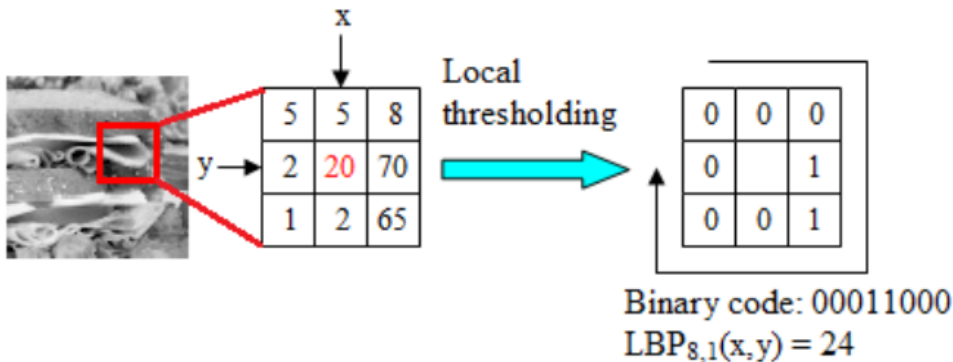
Common feature descriptor, use in [6, 12, 13]



Feature description

Local binary pattern

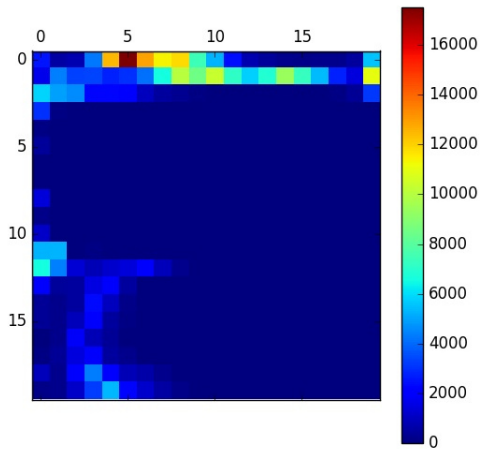
Use in [14, 15] for texture description





Feature description

Color moments and histograms



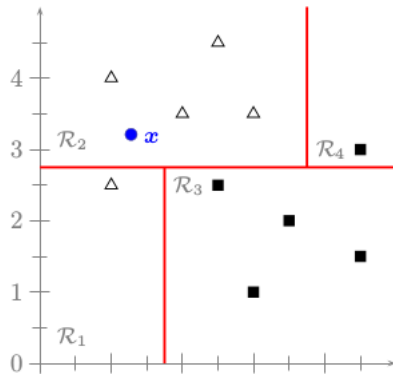
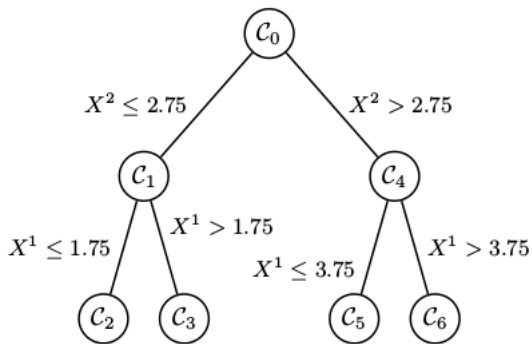
Mean:

$$\mu = \frac{1}{n} \sum_{i=1}^n x_i$$

Variance:

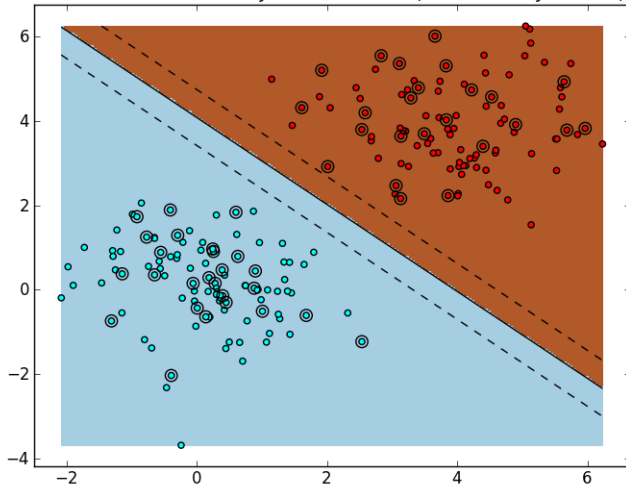
$$\text{Var}(X) = \sum_{i=1}^n p_i \cdot (x_i - \mu)^2$$

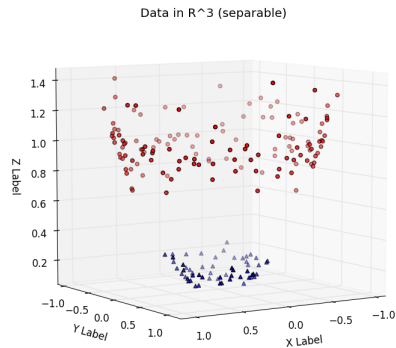
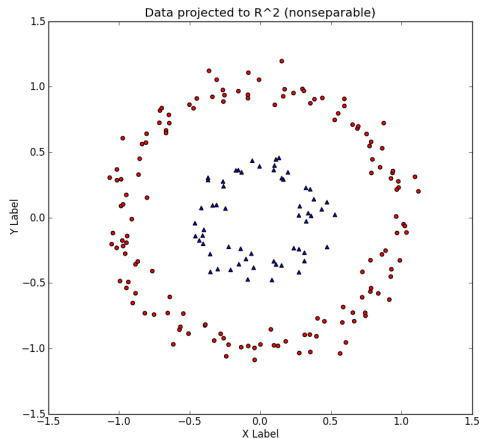
Decision tree and random forest



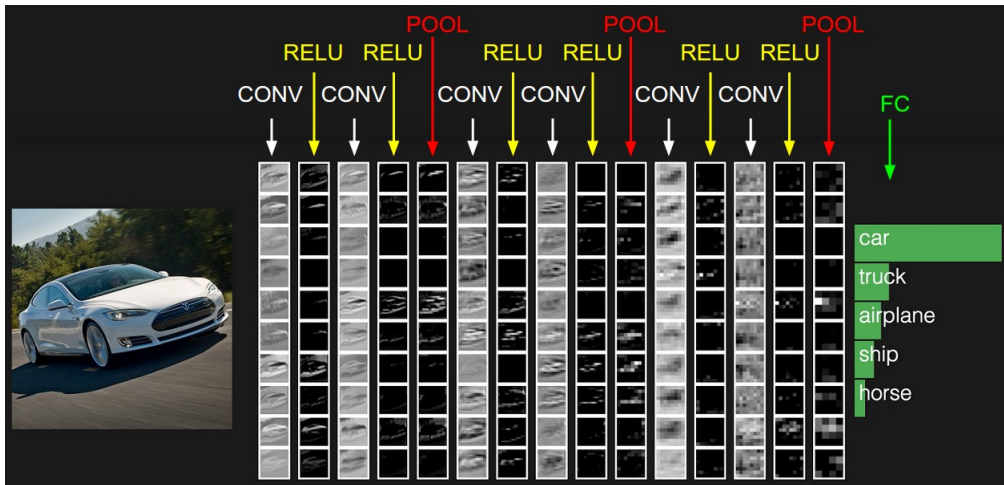
Classifiers

SVM Decision Boundary, Linear Kernel (1.0 accuracy, $C=1.0$)

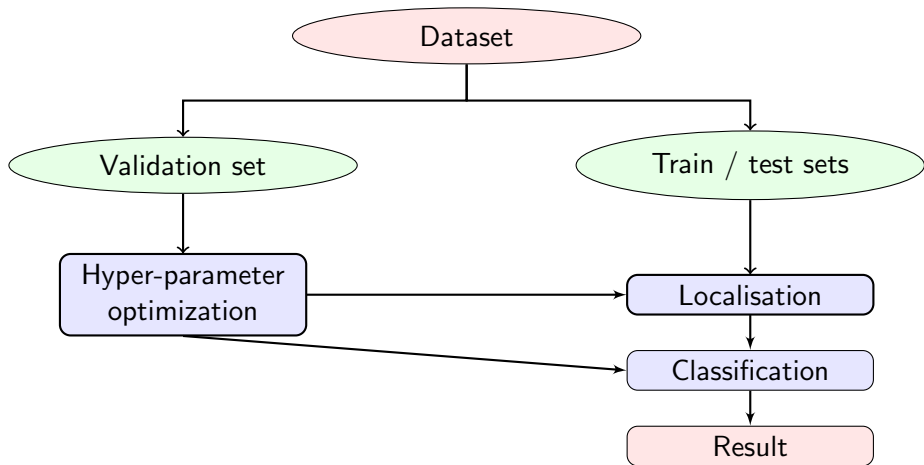




Classifiers



Methodology





Results

Localisation

Using a DCNN pre-trained on [16] to detect saliency object

Correctness metric: As describe in [17], must have an intersection over union greater than 50 %

$$IoU = \frac{area(B_p \cap B_{gt})}{area(B_p \cup B_{gt})}$$

UEC FOOD 256:

Metric	My method	DCNN from [18]
Accuracy	73 %	60 %
Recall	74 %	80 %
Precision	79 %	70 %

Method	Average accuracy
CNN as descriptor + RF	40 %
BoW (1000 words)+ SVM with χ^2	10 %
LBP + color historams and moments + Decision tree	5 %
LBP + color historams and moments + SVM	11 %
LBP + color historams and moments + RF	16 %
DCNN from [18]	63 %
DCNN from [19]	67 %



Results

Localisation and classification

Localisation: pre-trained DCNN followed by the classification: CNN as a descriptor and RF

UEC FOOD 256

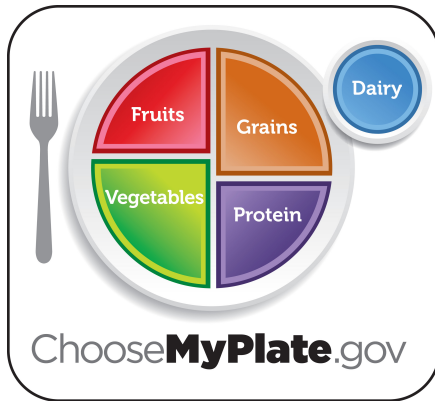
Accuracy	My method	DCNN from [18]
Overall	28 %	36 %
Localisation	74 %	60 %
Classification	38 %	60 %

UEC FOOD 100

Accuracy	My method	[20]	[21]
Overall	33 %	-	-
Localisation	67 %	60 %	-
Classification	50 %	-	72 %

Future work

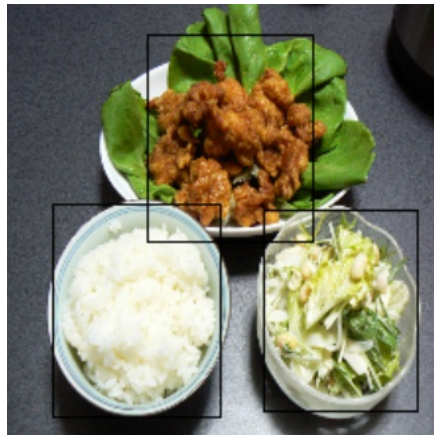
- Use a better feature descriptor / classifier
- Regroup the classification in 5 big categories for food intake as in [4]



Segmentation and classification applied on two reference datasets

Obtain great localisation results

Overall accuracy of 28 % (to date, the best result is 36 % in [18]).





References I



Keigo Kitamura, Toshihiko Yamasaki, and Kiyoharu Aizawa. “FoodLog: Capture, Analysis and Retrieval of Personal Food Images via Web”. In: **Proceedings of the ACM multimedia 2009 workshop on Multimedia for cooking and eating activities - CEA '09** (2009), p. 23.



Keigo Kitamura, Toshihiko Yamasaki, and Kiyoharu Aizawa. “Food log by analyzing food images”. In: **ACM international conference on Multimedia** (2008), p. 999.



Gamhewage C. De Silva and Kiyoharu Aizawa. “Clustering meal images in a web-based dietary management system”. In: **Proceedings - IEEE International Conference on Multimedia and Expo**. IEEE, 2011, pp. 1–6.



References II



Kiyoharu Aizawa et al. “Food balance estimation by using personal dietary tendencies in a multimedia food log”. In: **IEEE Transactions on Multimedia** 15.8 (2013), pp. 2176–2185.



Hokuto Kagaya, Kiyoharu Aizawa, and Makoto Ogawa. “Food Detection and Recognition Using Convolutional Neural Network”. In: **ACM Multimedia**. 2. 2014, pp. 1085–1088.



Mei Chen et al. “PFID: Pittsburgh Fast-food Image Dataset”. In: **Proceedings - International Conference on Image Processing, ICIP** (2009), pp. 289–292.



Yuji Matsuda, Hajime Hoashi, and Keiji Yanai. “Recognition of multiple-food images by detecting candidate regions”. In: **Proceedings - IEEE International Conference on Multimedia and Expo**. IEEE, 2012, pp. 25–30.



References III



Škrjanec Marko. “Automatic fruit recognition using computer vision”. Mentor: Matej Kristan. Bsc thesis. Faculty of Computer and Information Science, University of Ljubljana, 2013.



Lukas Bossard, Matthieu Guillaumin, and Luc Van Gool. “Food-101 - Mining discriminative components with random forests”. In: **Lecture Notes in Computer Science**. Vol. 8694 LNCS. PART 6. 2014, pp. 446–461.



Parisa Pouladzadeh Abdulsalam Yassine and Shervin Shirmohammadi. “FooDD: Food Detection Dataset for Calorie Measurement Using Food Images”. In: **New Trends in Image Analysis and Processing – ICIAP 2015 Workshops** 9281 (2015), pp. 441–448.



References IV



Yoshiyuki Kawano and Keiji Yanai. “Automatic expansion of a food image dataset leveraging existing categories with domain adaptation”. In: **Lecture Notes in Computer Science** 8927 (2015), pp. 3–17.



Hajime Hoashi, Taichi Joutou, and Keiji Yanai. “Image recognition of 85 food categories by feature fusion”. In: **Proceedings - 2010 IEEE International Symposium on Multimedia, ISM 2010**. IEEE, 2010, pp. 296–301.



Vinay Bettadapura et al. “Leveraging context to support automated food recognition in restaurants”. In: **Proceedings - 2015 IEEE Winter Conference on Applications of Computer Vision, WACV 2015**. 2015, pp. 580–587.



References V



Zhimin Zong et al. “On the combination of local texture and global structure for food classification”. In: **Proceedings - 2010 IEEE International Symposium on Multimedia, ISM 2010**. IEEE, 2010, pp. 204–211.



Duc Thanh Nguyen et al. “Food image classification using local appearance and global structural information”. In: **Neurocomputing** 140 (2014), pp. 242–251.



Jianming Zhang et al. “Unconstrained Salient Object Detection via Proposal Subset Optimization”. In: **IEEE Conference on Computer Vision and Pattern Recognition(CVPR)** (2016).



M. Everingham et al. **The PASCAL Visual Object Classes Challenge 2012 (VOC2012) Results**.



References VI



Marc Bolaños and Petia Radeva. “Simultaneous Food Localization and Recognition”. In: (2016), pp. 2–7.



Keiji Yanai and Yoshiyuki Kawano. “Food image recognition using deep convolutional network with pre-training and fine-tuning”. In: **2015 IEEE International Conference on Multimedia & Expo Workshops (ICMEW)**. IEEE, 2015, pp. 1–6.



Wataru Shimoda and Keiji Yanai. “CNN-based food image segmentation without pixel-wise annotation”. In: **New Trends in Image Analysis and Processing – ICIAP 2015 Workshops**. Vol. 9281. 2015, pp. 449–457.



Yoshiyuki Kawano and Keiji Yanai. “Food Image Recognition with Deep Convolutional Features”. In: **ACM International Joint Conference on Pervasive and Ubiquitous Computing (UbiComp)** (2014), pp. 589–593.