



# Automated food log

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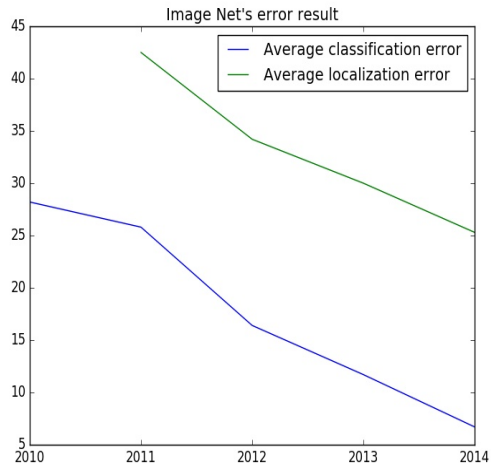
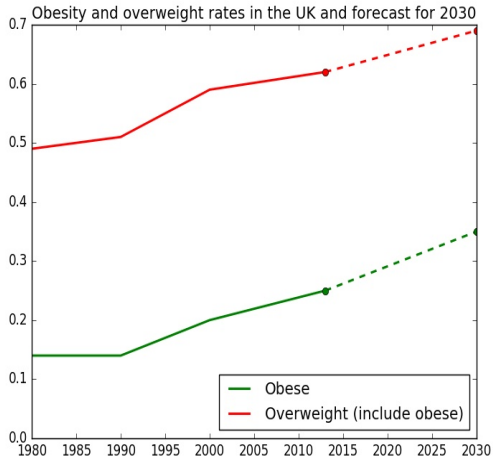
[www.cranfield.ac.uk](http://www.cranfield.ac.uk)



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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 localization 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
<b>UEC FOOD 256 [11]</b>	<b>2015</b>	<b>31395</b>	<b>World</b>	<b>256</b>	<b>Yes</b>

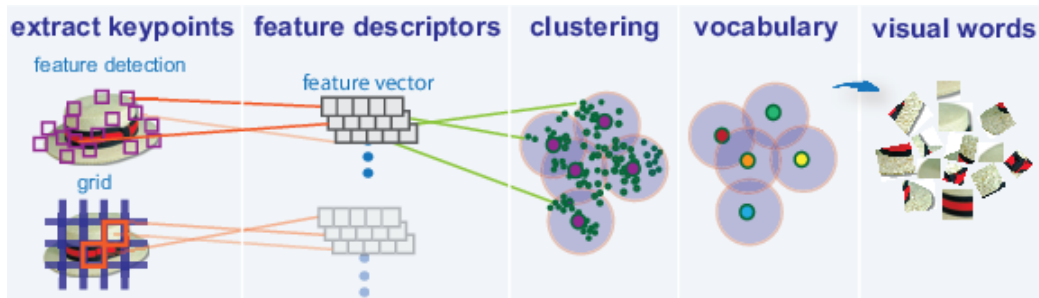
## Example of multi-items



# Feature description

## Bag of visual words

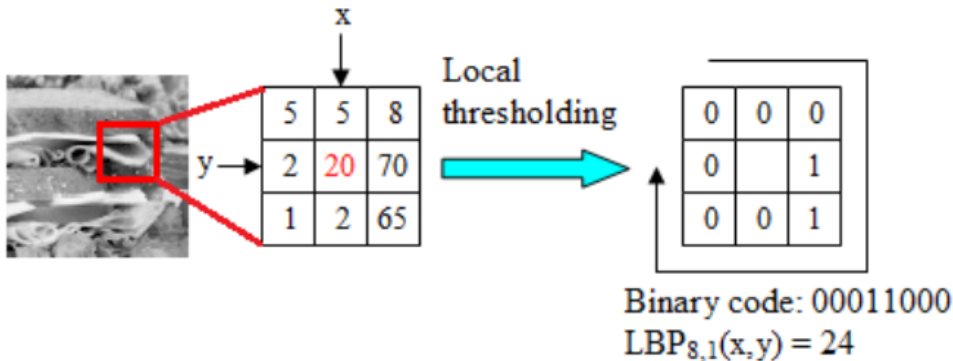
Common feature descriptor, use in [6, 12, 13]



# Feature description

## Local binary pattern

Use in [14, 15] for texture description

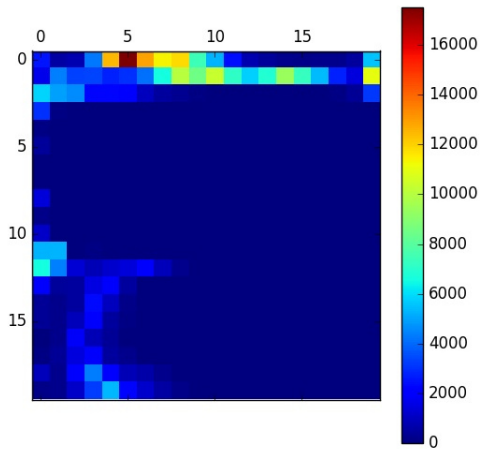






# Feature description

## Color moments and histograms



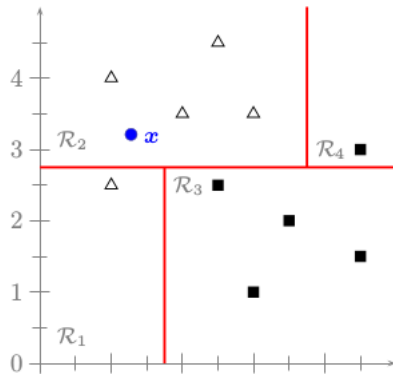
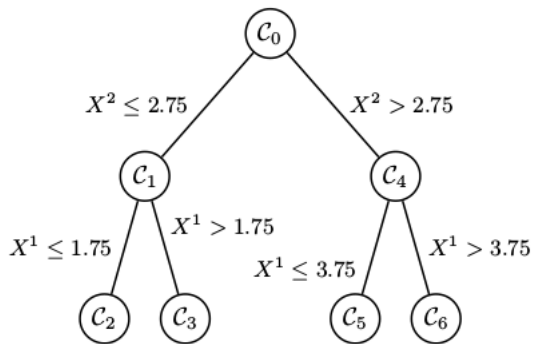
Mean:

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

Variance:

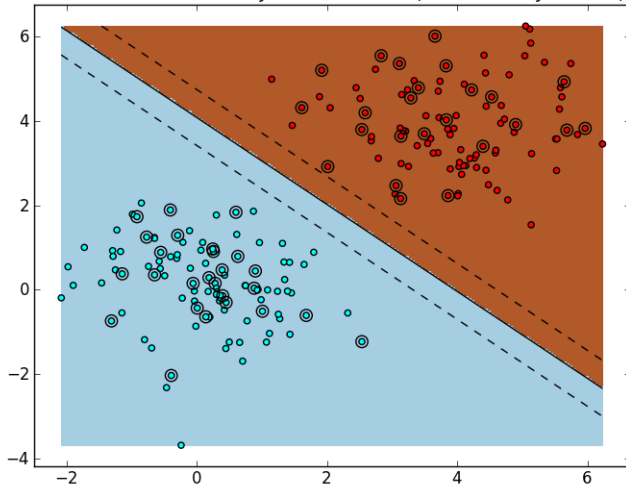
$$\text{Var}(X) = \sum_{i=1}^n p_i \cdot (x_i - \mu)^2$$

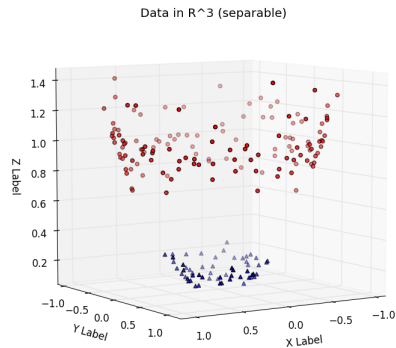
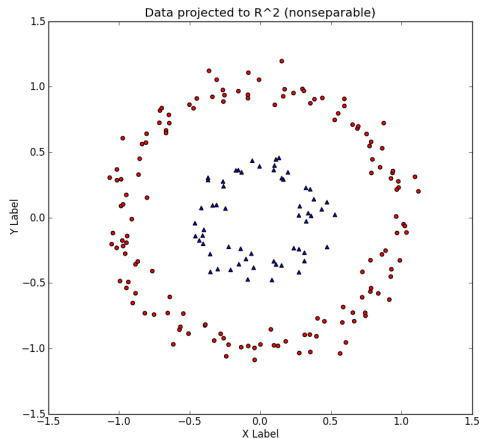
## Decision tree and random forest



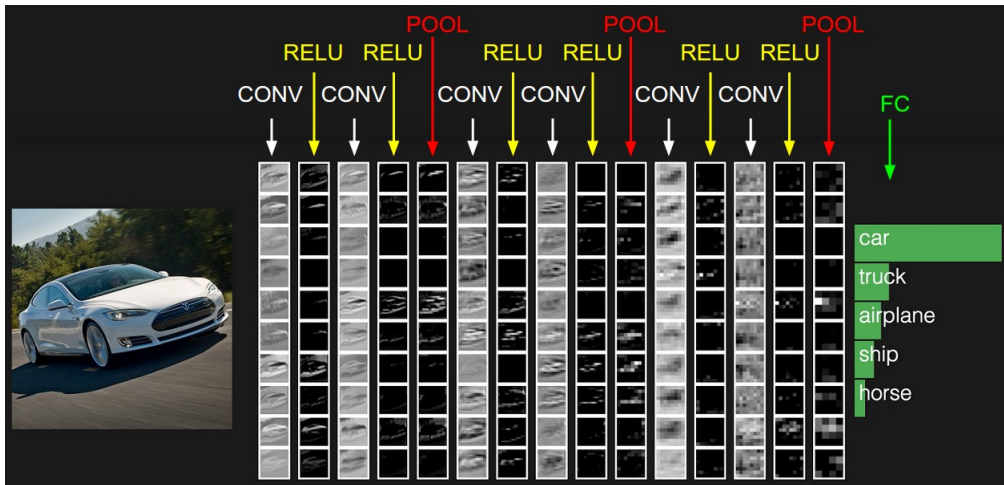
# Classifiers

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

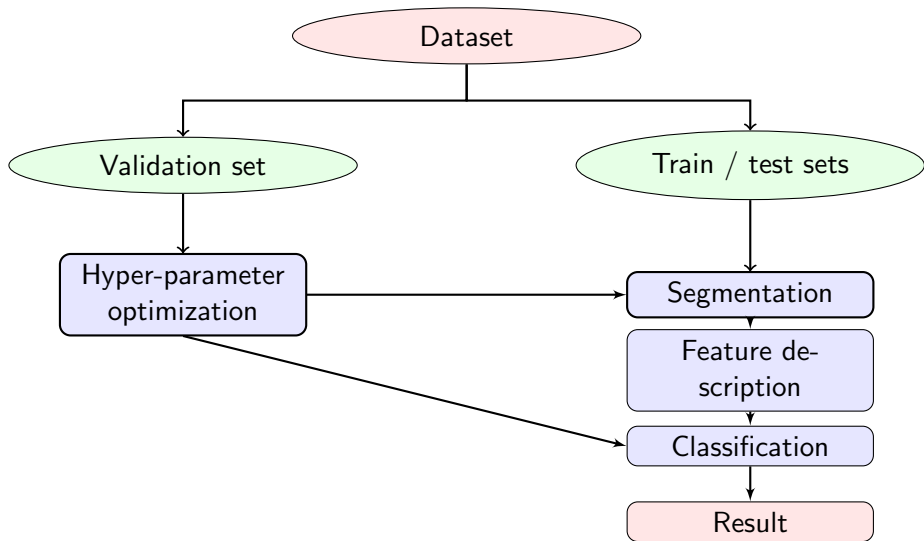




# Classifiers



# Structure





# Results

## Localization

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 %

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

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

Method	Average accuracy
CNN as descriptor + RF	<b>40 %</b>
BoW (1000 words)+ SVM with $\chi^2$	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

## Segmentation and classification

Segmentation: DCNN followed by the classification: CNN as a descriptor and RF

UEC FOOD 256

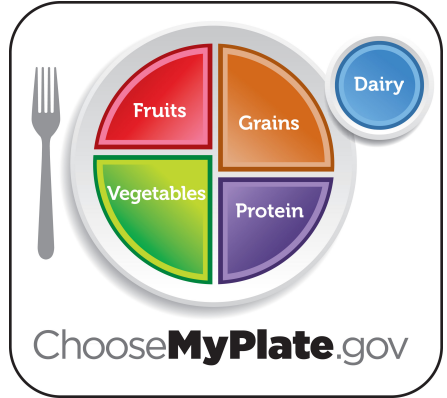
Accuracy	My method	DCNN from [18]
Overall	<b>28 %</b>	36 %
Segmentation	<b>74 %</b>	60 %
Classification	<b>38 %</b>	60 %

UEC FOOD 100

Accuracy	My method	[20]	[21]
Overall	<b>33 %</b>	-	-
Segmentation	<b>67 %</b>	60 %	-
Classification	<b>50 %</b>	-	72 %

## Future work and comment

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



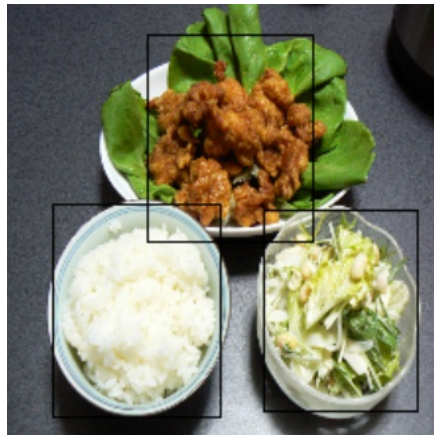


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Segmentation and classification applied on two reference datasets

Obtain great localization results

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





## References I



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





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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.



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