

Dish Detection and Segmentation for Dietary Assessment on Mobile Phones

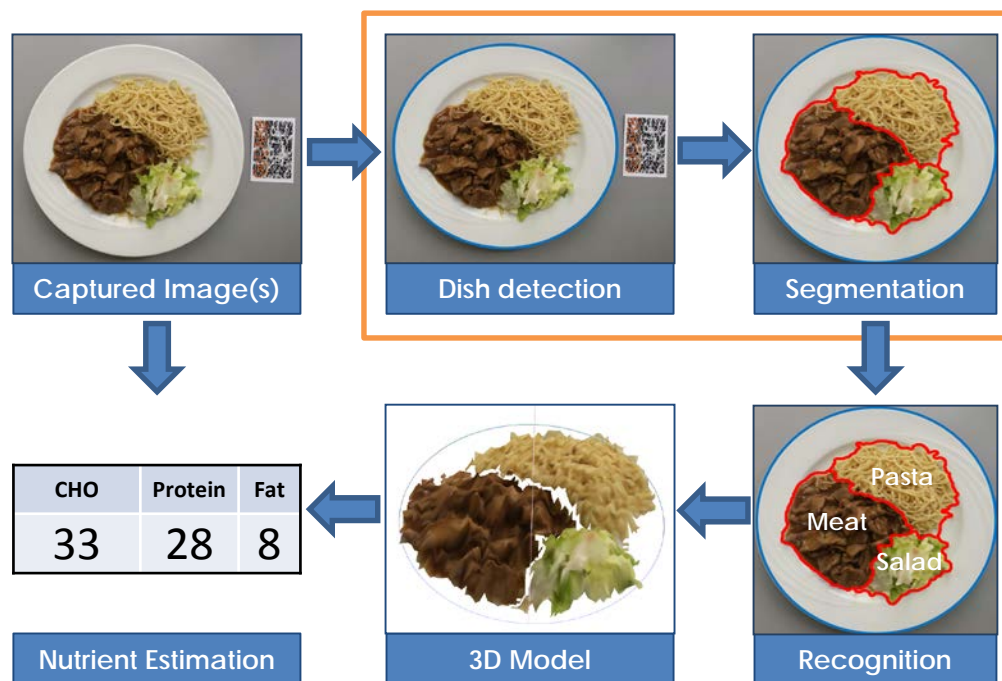
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GoCARB System



Dish detection

Input

An image of the meal inside an elliptic dish

Process

1. Preprocessing and Canny edge detection
2. Filtering of edge graph
3. Incremental RANSAC on edge segments

Output

A map with label codes for background and foreground
Ellipse equation



Segmentation

Seeded region growing

Input

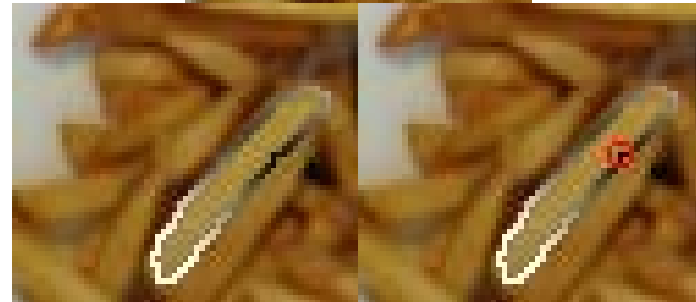
“Seed regions”: pixel sets

Process

1. Luminosity-dampened CIE94 colour distance
2. Iterative growth of the regions:
Add the border pixel with the smallest distance to its region

Output

A map with label codes for background and food items



Segmentation

Statistical region merging

Input

Segmentation map

Process

1. Luminosity-dampened CIE94 colour distance
2. Iterative merging of the regions:
Merge the two regions with the smallest distance between them

Output

A map with label codes for background and food items



Segmentation

Automatic

1. Seeds



2. Region growing



3. Region Merging



Semi-automatic



Evaluation - Procedure

Dataset:

- 1600 Manually annotated images
- Single large round plate
- Food from the local hospital restaurant

Scoring:

- Average overlap between ground truth and result
- $$NI_{sum}(G \Rightarrow R) = \frac{\sum_i \max_j (|R \cap G_j|)}{\sum_i |R_i|}$$
- $$F_{sum} = 2 * NI_{sum}(G \Rightarrow R) * NI_{sum}(R \Rightarrow G) / (NI_{sum}(G \Rightarrow R) + NI_{sum}(R \Rightarrow G))$$

Evaluation

Automatic	Average F_{sum} (%)	Time (s/image)
Proposed	88.2	0.45
Mean-shift [1]	87.5	2.1
Local Variation [2]	82.6	2.8
Ultrametric contours [3]	69.2	19
Semi-Automatic	Average F_{sum} (%)	Time (s/image)
Proposed	90.8	0.49
Flood fill	89.9	0.52

[1] Anthimopoulos, M.; Dehais, J.; Diem, P.; Mougiakakou, S., "Segmentation and recognition of multi-food meal images for carbohydrate counting," IEEE 13th International Conference on Bioinformatics and Bioengineering (BIBE), (2013)

[2] Felzenszwalb, P. F., Huttenlocher, D. P.: Image segmentation using local variation. In: IEEE Conference on Computer Vision and Pattern Recognition, pp. 98-104 (1998)

[3] Arbelaez P., Maire M., Fowlkes C., and Malik J.: Contour detection and hierarchical image segmentation. IEEE Trans. Pattern Anal. Mach. Intell., vol.33, no.5, pp.898–916 (2011.)

Thank you for your attention.



(Photo credit: Kem Sypher)

Dish detection

Incremental RANSAC

1. Each segment contains s edge points, $s \gg 5$
2. A conic can be generated from $k \leq 5$ segments

Incremental sampling:

1. Set $k=1$
2. Apply RANSAC
Max consensus set size: n_k
3. If $n_k > n_{k-1}$ and $k < 5$
 1. Set $k=k+1$
 2. Repeat 2-3

Average accuracy: 99.1%



Segmentation

Colour Distance

- $\text{CIE94}(L, a, b, L', a', b') = \sqrt{\Delta L_N^2 + \Delta C_N^2 + \Delta H_N^2}$, with
- $\Delta L_N = (L - L')$,
- $\Delta C_N = (\sqrt{a^2 + b^2} - \sqrt{a'^2 + b'^2}) / (1 + 0.045 \sqrt{a^2 + b^2})$,
- $\Delta H_N = \sqrt{(a - a')^2 + (b - b')^2 - \Delta C_N^2} / (1 + 0.015 \sqrt{a^2 + b^2})$
- $\text{dist}(L, a, b, L', a', b') = \sqrt{|\Delta L_N| + \Delta C_N^2 + \Delta H_N^2}$

Automatic	Av. F _{sum} (%)
RGB - Euclidian	61.1
CIE94	80.2
Proposed	88.2
Semi-Automatic	Av. F _{sum} (%)
RGB - Euclidian	72.3
CIE94	78.2
Proposed distance	90.8

Segmentation

Merging Distance

- $\text{Dist}(R_i, R_j) = \frac{\text{dist}(L_i, a_i, b_i, L_j, a_j, b_j)}{\sqrt{|\text{edge}_{i,j}|}}$, where
 - L_i : Median L_p , p in R_i
 - a_i, b_i : Average a_p, b_p , p in R_i

Merging Cost	Av. F _{sum} (%)
Color distance	85.8
Proposed	88.2