

Vision Intelligence III: Visual Object Computations

Wei Qi Yan

Auckland University of Technology

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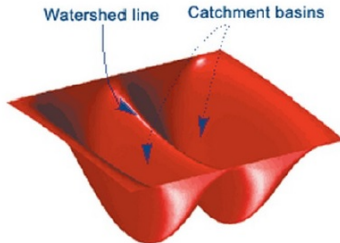
Visual Features

- **Pixel colour:** Binary, pseudo, real colour, etc.
- **Histogram:** Color histogram, vector histogram, etc.
- **Texture:** Gabor texture, Tamura texture, GLCM, etc.
- **Edge:** Canny(5×5), Sobel (3×3), Robert (2×2), etc.
- **Gradient:** Horizontal, vertical, both, etc.
- **Local features:** SIFT, HOG, etc.
-

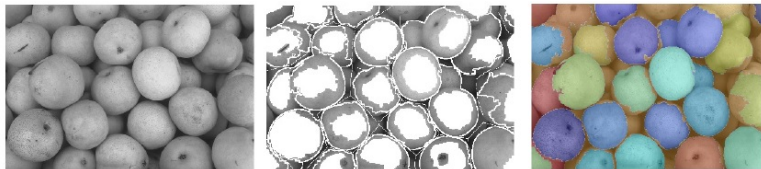
Visual Object Segmentation

- Thresholding
- Clustering methods
- Histogram-based methods
- Edge detection
- Region growing methods
- Watershed segmentation
-

MALAB: Watershed Segmentation

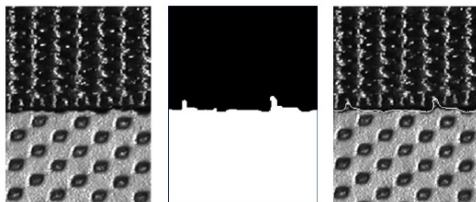


MATLAB: Watershed Segmentation



- Use gradient magnitude as segmentation function;
- Mark foreground objects;
- Compute background markers.

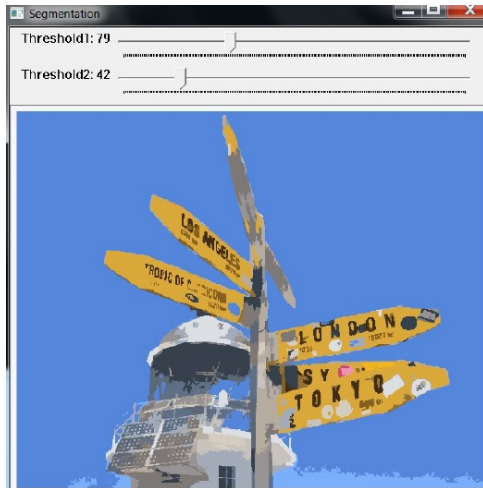
MATLAB: Texture Image Segmentation



- Create a rough mask for the texture;
- Use the rough mask to segment the top texture;
- Display segmentation result.

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OpenCV: Image Segmentation



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Questions?



Questions?

In image segmentation, gradient includes:

- ❶ Vertical gradient
- ❷ Horizontal gradient
- ❸ Vertical and horizontal gradients
- ❹ None of the given options

The wrong answer is:---

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Questions?



Visual Object Detection

- Given positive and negative samples (well selected)
- Supervised learning provides a natural framework for studying object recognition.
- Train a function that can map novel images to one of labels (e.g., face or non-face).

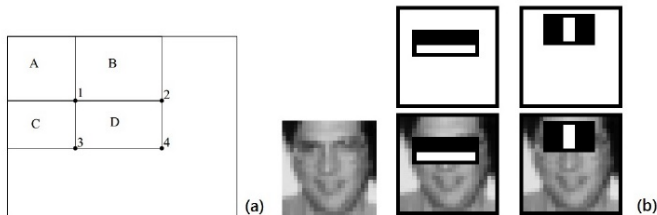
OpenCV: Face Detection - Training Set



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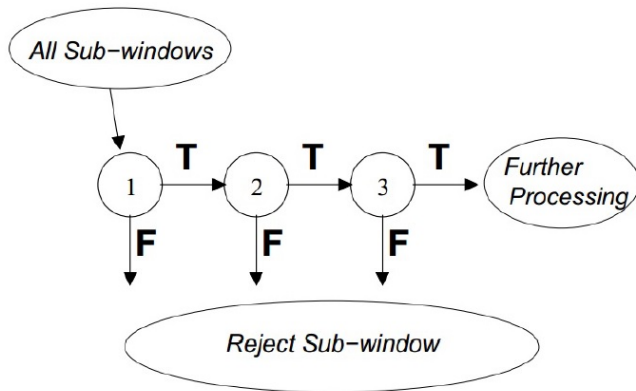
OpenCV: Viola & Jones Face Detection

- Haar Feature Selection
- Creating Integral Image
- AdaBoost Training algorithm
- Cascaded Classifiers



Paul Viola and Michael Jones (2001) Rapid Object Detection using a Boosted Cascade of Simple Features. CVPR 2001.

OpenCV: Viola & Jones Face Detection (Cascade Classifier)



Paul Viola and Michael Jones (2001) Rapid Object Detection using a Boosted Cascade of Simple Features. CVPR 2001.

OpenCV: Viola & Jones Face Detection (Advantage)

- Extremely fast feature computation
- Efficient feature selection
- Scale and location invariant detector
- Detect other types of objects (e.g., cars, hands, etc.).

OpenCV: Viola & Jones Face Detection (Disadvantage)

- Detector is most effective only on frontal images of a face.
- It can hardly cope with 45° face rotation both around vertical and horizontal axes.
- Sensitive to lighting conditions

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Questions?



Questions?

The V&J algorithm for human face detection in OpenCV is,

- ❶ Extremely fast.
- ❷ Efficient.
- ❸ Scale and location invariant.
- ❹ None of the given options.

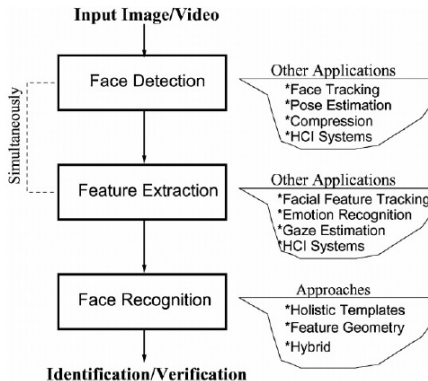
The wrong answer is: ---

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Questions?



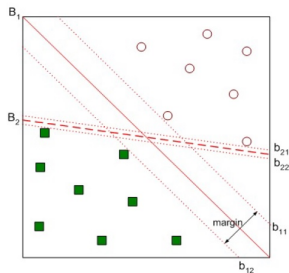
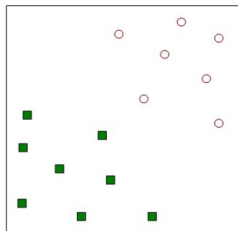
OpenCV: Face Detection and Recognition System



W. Zhao, et al. Face recognition: a literature survey, ACM Computing Surveys, Vol. 35, No. 4, 2003, pp. 399–458.

Support Vector Machine

- *Problem:* Given a vector dataset, find a linear hyperplane (decision boundary) that will separate the data.
- *Solution:* Find a hyperplane maximizes the margin (B_1 is better than B_2),
- *SVM Readings:*
<http://www.csie.ntu.edu.tw/~cjlin/libsvm/>



WEKA: Computational Tool



Machine Learning Group at the University of Waikato

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[Software](#)

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Weka 3: Data Mining Software in Java

Weka is a collection of machine learning algorithms for data mining tasks. The algorithms can either be applied directly to a dataset or called from your own Java code. Weka contains tools for data pre-processing, classification, regression, clustering, association rules, and visualization. It is also well-suited for developing new machine learning schemes.

Found only on the islands of New Zealand, the Weka is a flightless bird with an inquisitive nature. The name is pronounced like **this**, and the bird sounds like **this**.

Weka is open source software issued under the **GNU General Public License**.

We have put together several free online courses that teach machine learning and data mining using Weka. Check out the **website for the courses** for details on when and how to enrol. The videos for the courses are available **on Youtube**.

Yes, It is possible to apply Weka to **big data**!

Getting started

- [Requirements](#)
- [Download](#)
- [Documentation](#)
- [FAQ](#)
- [Getting Help](#)

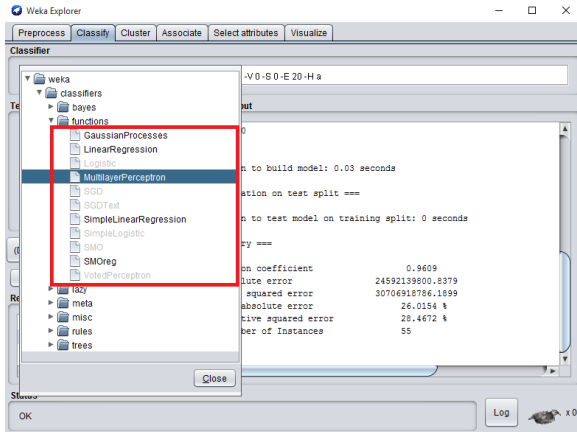
Further Information

- [Citing Weka](#)
- [Datasets](#)
- [Related Projects](#)
- [Miscellaneous Code](#)
- [Other Literature](#)

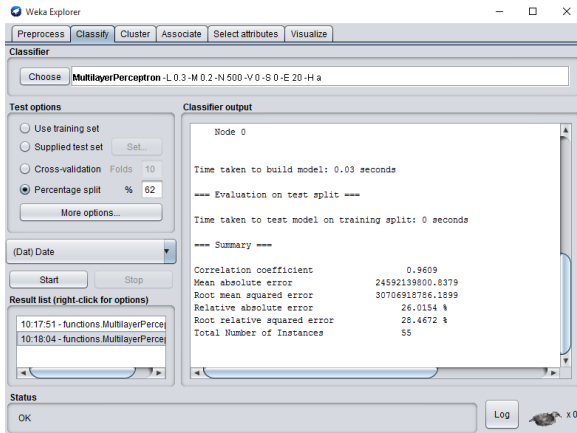
Developers

- [Development](#)
- [History](#)
- [Subversion](#)
- [Contributors](#)
- [Commercial licenses](#)

WEKA: Computational Tool



WEKA: Computational Tool



WEKA: Computational Tool

```
% 5. Number of Instances: 150 (50 in each of three classes)
% 6. Number of Attributes: 4 numeric, predictive attributes and the class
% 7. Attribute Information:
%   1. sepal length in cm
%   2. sepal width in cm
%   3. petal length in cm
%   4. petal width in cm
%   5. class:
%       -- Iris Setosa
%       -- Iris Versicolour
%       -- Iris Virginica
% 8. Missing Attribute Values: None
% Summary Statistics:
%      Min  Max   Mean    SD   Class Correlation
%  sepal length: 4.3  7.9   5.84  0.83    0.7826
%  sepal width: 2.0  4.4   3.05  0.43   -0.4194
%  petal length: 1.0  6.9   3.76  1.76    0.9490 (high!)
%  petal width: 0.1  2.5   1.20  0.76    0.9565 (high!)
% 9. Class Distribution: 33.3% for each of 3 classes.
```

```
@RELATION iris
@ATTRIBUTE sepallength REAL
@ATTRIBUTE sepalwidth REAL
@ATTRIBUTE petallength REAL
@ATTRIBUTE petalwidth REAL
@ATTRIBUTE class {Iris-setosa,Iris-versicolor,Iris-virginica}
@DATA
5.1,3.5,1.4,0.2,Iris-setosa
4.9,3.0,1.4,0.2,Iris-setosa
```

Evaluations of Empirical Algorithms

- Training set
- Test set
- Ground truth
- Precision: $p = \frac{TP}{TP+FP}$
- Recall: $r = \frac{TP}{TP+FN}$
- F-measure: $F = \frac{2 \cdot p \cdot r}{p+r}$
- G-measure: $G = \sqrt{p \cdot r}$

Note: F-measure is the harmonic mean (average) of recall and precision, G-measure is the geometric mean (average).

Precision?



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Questions?



Questions?

In evaluation of empirical algorithms, if $TP=60$, $FP=20$, what is the precision $p = \frac{TP}{TP+FP}$?

- ❶ 20%
- ❷ 60%
- ❸ 75%
- ❹ 80%

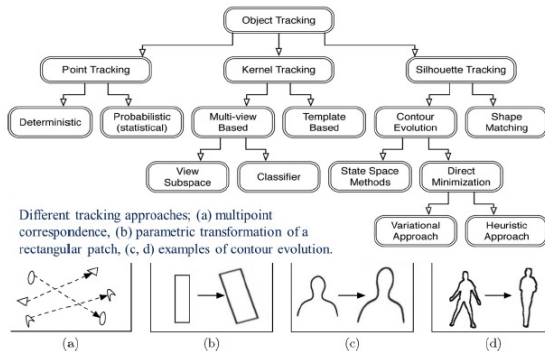
The correct answer is:---

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Questions?



Visual Object Tracking (Demo)



A. Yilmaz, Object Tracking: A Survey. ACM Computing Surveys, Vol. 38, No. 4, Article 13, 2006.

Visual Object Tracking

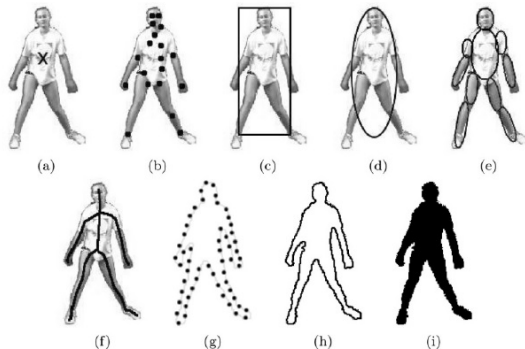


Fig. 1. Object representations. (a) Centroid, (b) multiple points, (c) rectangular patch, (d) elliptical patch, (e) part-based multiple patches, (f) object skeleton, (g) complete object contour, (h) control points on object contour, (i) object silhouette.

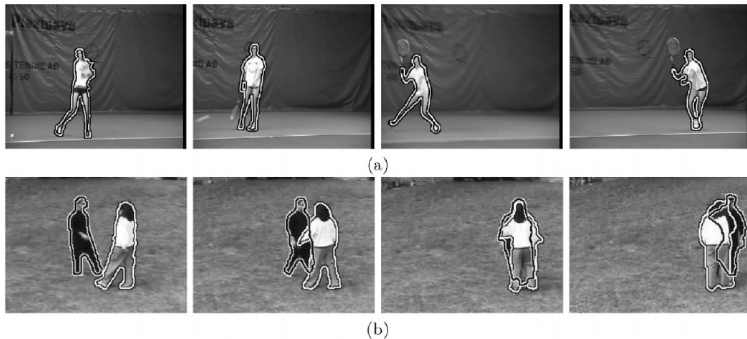
Car Tracking Using Level Set



Car tracking using the level sets method

A. Yilmaz, et al. (2006) Object Tracking: A Survey. ACM Computing Surveys, Vol. 38, No. 4, Article 13.

Contour Tracking



Contour tracking results. (a) tracking of a tennis player, (b) tracking in presence of occlusion

A. Yilmaz, et al. (2006) Object Tracking: A Survey. ACM Computing Surveys, Vol. 38, No. 4, Article 13.

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Questions?



Learning Objectives

- Explain how AI theories could be used in visual analytics.
- Evaluate the complexities involved in the automatic extraction of knowledge encoded in free texts and images.