MLRF Lecture 06

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More definitions

Lecture 06 part 05

Computer Vision

Computer Vision: a definition

Computer Vision: the automation of visual tasks with the goal of producing results directly or indirectly usable by humans. \Leftarrow Engineer definition

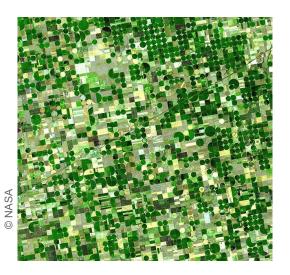
Input: image(s) in machine format (image acquisition is a subpart of CV)

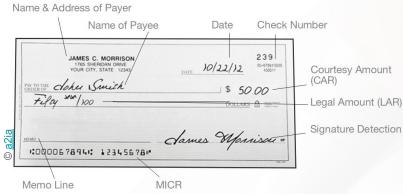
Output: some pieces of information about the image, new image(s)...

Computer Vision: some examples (1/4)

How would you process image pixels to get those results?







Input: still picture of insect Output: insect name Application: farming, ...

Classification

Input: satellite image (near visible

range)

Output: crop maturity

Application: farming, trading...

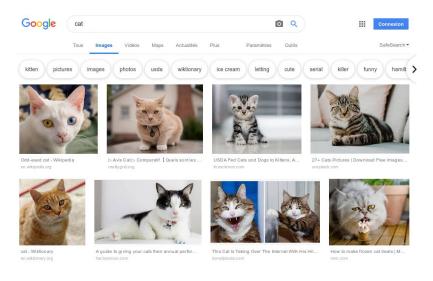
Regression

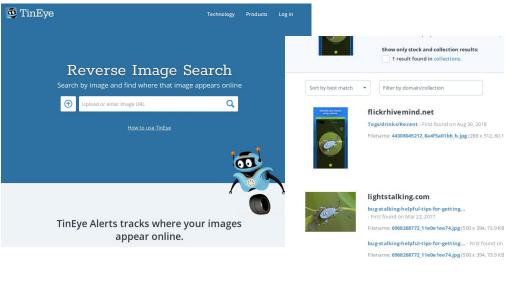
Input: bank check image (greylevel) **Output:** account number, amount...

Application: banks



Computer Vision: some examples (2/4)





Input: text

Output: list of relevant images **Application:** look for cats, ...

Indexing, ...

Input: image

Output: similar images

Application: duplicate detection, copyright, ...

Content-based

Image retrieval, ...

COLLECTION

COMMONS.Wikimedia.org

wiki/File:D1E791-1_(6968268772.).jpg - First found on Mai

Filename: D1E791-1_(6968268772.).jpg (3516 x 2771, 8.4 Mi

Computer Vision: some examples (3/4)

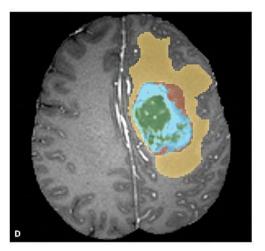




Input: two images of face
Output: same/different person
Application: authentication, ...

VS





Input: brain 3D scan
Output: regions with high tumor probability
Application: assisted diagnosis, ...



Computer Vision: some examples (4/4)

Some applications are direct (like the insect recognition app): a human reads and uses the output

Some applications are indirect (like bank check reading): the output is fed to a business system

Some applications extend what humans can naturally do:
either by extending our range of visible colors (satellite example)
or by simply being more efficient (face verification)

And there are many many more examples...

Pattern Recognition

Pattern Recognition: a definition

Pattern Recognition: The field of pattern recognition is concerned with the automatic discovery of regularities in data through the use of computer algorithms and with the use of these regularities to take actions such as **classifying** the data into different categories. — Bishop 2006



The International Association for Pattern Recognition (IAPR) is an international association [...] concerned with <u>pattern recognition</u>, <u>computer vision</u>, and <u>image processing</u> in a broad sense.

Pattern Recognition: examples

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OCR computer Vision

Pedestrian detection Computer Vision

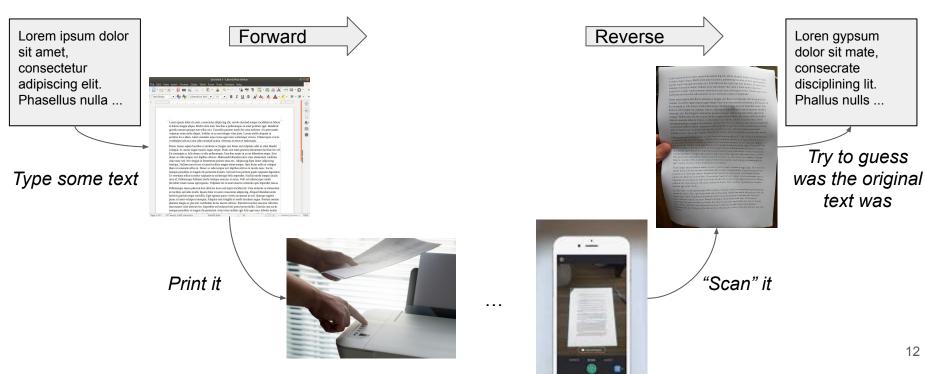
Credit card fraud detection Computer Vision

...
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 \Rightarrow CV \cap PR $\neq \emptyset$, "recognition" often means "classification"

Pattern Recognition is an inverse problem

OCR example – Why Pattern Recognition is hard



Sometimes used to describe "visual percepts" (image patterns) which exhibit a "large [deviation] from randomness". See *Cao et al. 2008*.

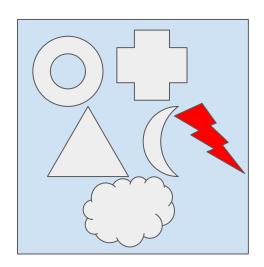
A way to designate meaningful visual patterns.

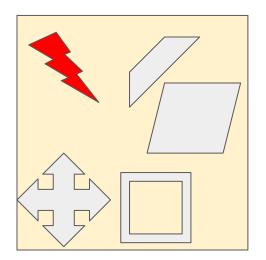
An interesting mathematical foundation to compare them:

Let S and S' be two shapes observed in two different images and which happen to be similar. Denote their (small) Hausdorff distance after registration by $\delta = d(S, S')$. Assume we know enough of the background model to compute the probability $Pr(S, \delta) = Pr(d(S, \Sigma) \le \delta)$ that some shape in the background, Σ be as similar to S as S' is. If this probability is very small one can deduce that S' does not look like S just by chance. Then S and S' will be identified as the same shape.

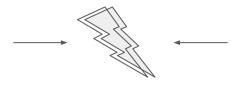
We can check whether two shapes are <u>significantly</u> close.

Let S and S' be two shapes observed in two different images and which happen to be similar.



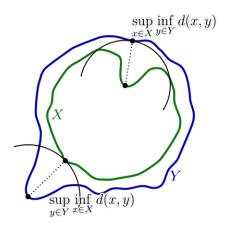


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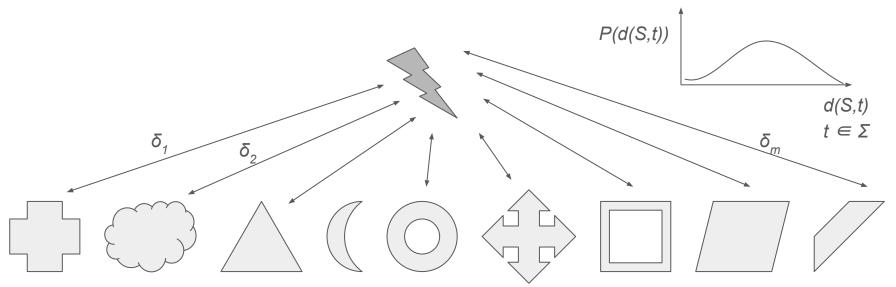


Hausdorff distance = max of min distances between points on the contours of two shapes.

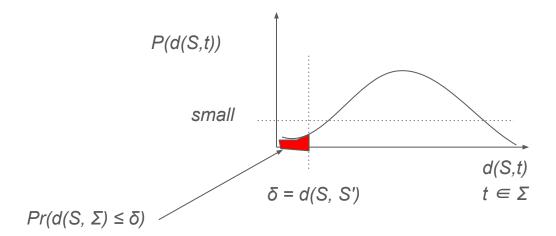
$$d_{\mathrm{H}}(X,Y) = \max \set{\sup_{x \in X} \inf_{y \in Y} d(x,y), \, \sup_{y \in Y} \inf_{x \in X} d(x,y)}.$$



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So, some **statistics** can help us making better decisions...

Idea: **learn** the distance threshold under which shapes can be deemed identical.

A contrario approach.

Benefits of ML for CV/PR

A duck example

How to filter the grass to keep only the duck shape, using thresholds in the color domain?







Did you try it during practice session?

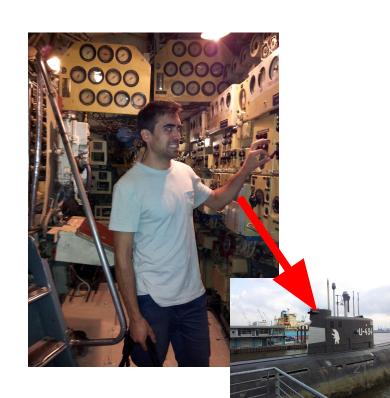
How boring was it to tune the parameters by hand?

Why using Machine Learning in Computer Vision?

To avoid knob tuning. It's complex. It's unsafe.



Photo by jc.winkler

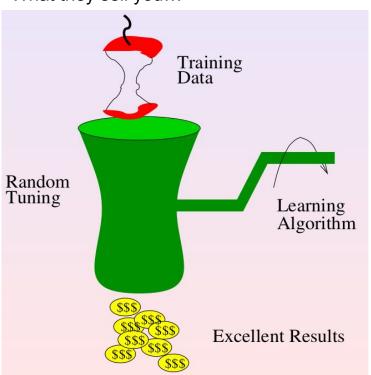


Computer Vision / Pattern Recognition



But Beware of the Machine Learning Magic

What they sell you...



But most often...

