MLRF Lecture 02 J. Chazalon, LRDE/EPITA, 2021

Agenda for lecture 2

- 1. Introduction
- 2. Global image descriptors
- 3. Clustering
- 4. Local feature detectors

Introduction

Lecture 02 part 01

Previously, in MLRF...

Summary of last lecture

Machine learning

- Machine learning = searching for the best model in a hypothesis space
- Inductive machine learning, optimization-based
- Inductive bias, biais/variance compromise
- Supervised, reinforcement, unsupervised learning
- Regression, classification, density estimation
- Model validation: test generalisation, separate/decorrelate test & training sets

Template matching

- Sum of squared differences (T-I)², or correlation-based methods (T×I)
- Normalization needed for correlation-based methods
- Tolerates translation and small noise, but not rotation, intensity shift, ...

Debriefing of practice session 1

PS1 content

- 1. Jupyter tricks
- 2. NumPy reminders
- 3. Intro to image manipulations
- 4. Twin it! part 1: Template matching
- 5. (Bonus level: segmentation)

Discussion

- Who completed part 1? 2? 3? 4? 5?
- Any remarks, comments, questions?
- Things to keep, change, remove?

Practice session 1: Take home messages (1/2)



How annoying was it to <u>manually adjust color thresholds</u> to select the duck?

How could have we <u>automated</u> it?

Practice session 1: Take home messages (2/2)

Results with method SQDIFF_NORMED → (lower is better)

Strengths and weaknesses of <u>template matching</u> for the Twin it! case?

Effects of <u>normalization</u>?



query images

result images (closest to query according to method)

Next practice session(s)

Next practice session

Twin it!, again, with a slightly more elaborated approach:

1. **Pre-select bubbles based on their colors ⇒ Color histograms**



1.1. Color quantization: reduce the colors of the bubbles.









Recolored









1.2. Compute the color histogram of each bubble.



1.3. Compute the distance matrix between each bubble, using its color histogram.



1.4. Visualize the bubbles in an interesting way using hierarchical clustering.



Next practice session

Twin it!, again, with a slightly more elaborated approach:

- 1. Pre-select bubbles based on their colors \Rightarrow Color histograms
- For the pre-selected bubbles, check their content is similar
 ⇒ Detect stable points and extract the patches around them





Next practice session

Twin it!, again, with a slightly more elaborated approach:

- 1. Pre-select bubbles based on their colors \Rightarrow Color histograms
- For the pre-selected bubbles, check their content is similar
 ⇒ Detect stable points and extract the patches around them
 - ⇒ Compare (match) those patches





Image descriptors

Issues with methods based on pixel comparison

What is important? What do they consider? **Raw pixels!** ⇒ We want to be able to make use of **domain knowledge**! *Like sensitivity to shape, or dominant color information.*

They are terribly **slow** and works **only for small images**. ⇒ We want to **summarize an image** to a much smaller vector.

They are **sensible to rotation, scaling**, and many other perturbations. ⇒ We want to adjust sensitivity/invariance to perturbations. *Do we tolerate translation? Rotation? Intensity shift?*

How can we compare different pairs of images? Metric issues.

 \Rightarrow We want to be able to achieve **more than 1 vs all comparisons**.

Image descriptors: Overview

Different sizes and contents ⇒ Different kind of descriptors



 \approx 500×500 px and more

 \approx 20×20 px and less

Image descriptors: Overview

Different sizes and contents ⇒ Different kind of descriptors

Different problems ⇒ Different choices

- Computation / memory constraints
- Which perturbations to we have to tolerate? *rotation, translation...*
- What is the expected output?

classification, detection, ranking, segmentation...

Many, many approaches ⇒ Impossible to list them all

- Examples of several categories
- Focus on very useful or instructive ones