

1, general architecture

1.1, computing features:

I choose HOG as the feature. I compute the HOG feature by `cv2.HOGDescriptor.compute()`.

1.2, performing detection:

I implement OCL algorithm for the connected component labeling. The algorithm is from the paper "Optimizing Two-Pass Connected-Component Labeling Algorithms" by Kesheng Wu.

I have also referred to the paper "The connected-component labeling problem: A review of state-of-the-art algorithms" and webpage

["https://aishack.in/tutorials/connected-component-labelling/"](https://aishack.in/tutorials/connected-component-labelling/).

OCL is a two-pass algorithm. It marks labels in the first pass and combines the connected labels in the second pass. It is based on decision tree and union-find data structure.

About the union-find data structure, I implement it with dictionary.

1.3, performing recognition:

I use L2 distance for recognition.

For every character block c waiting for recognition, for every template character t , I compute $\text{score}_c = L2(c, t)$.

If $\min(\text{score}_c) > 3$, I think no template character t can match c , so c is "UNKNOWN".

Otherwise, I choose the template character t with the minimum score.

2, Output

features folder is `./feature`

`results.json`

3, Code structure:

extractor: features extractor for enrollment step.

ccl: connected component labeling for detection step.

tools: several tools like `encoder.py` for json encoding, `data_print.py` for image print function and several config variables.

data: all data

In order to match the requirement, I also add other directories.

feature: feature for enrollment step

character: input character templates