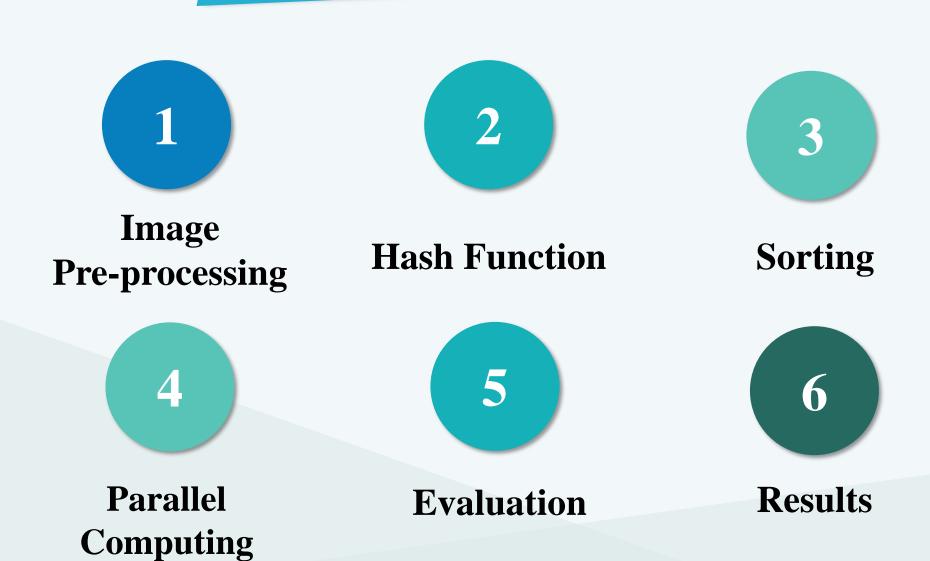
# Final Project

111318093 黄俊傑

111C52005 吳翊睿

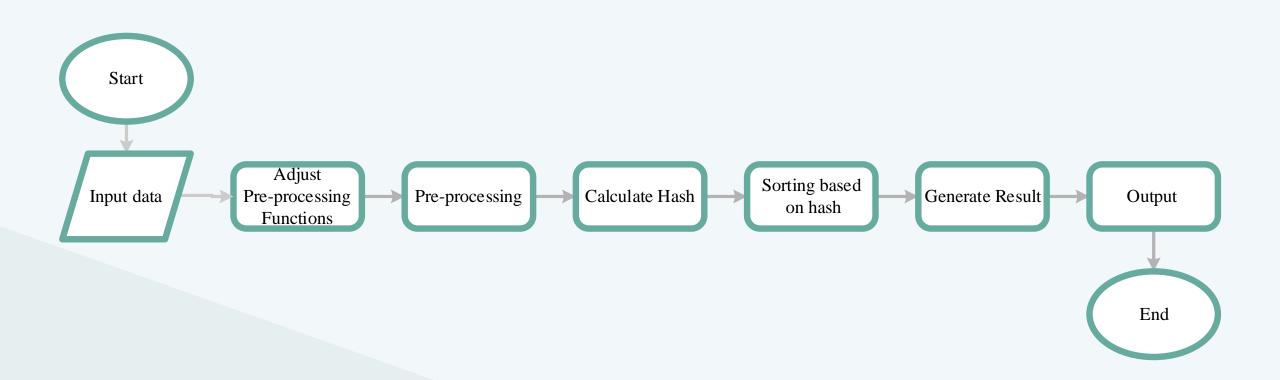
111C52016 吳佩霖

# Content

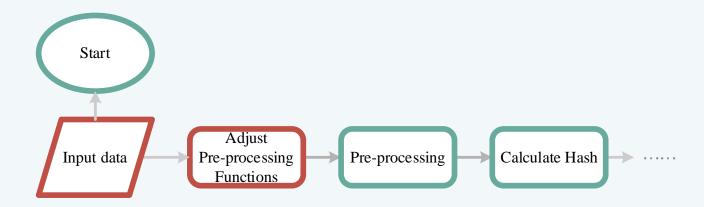


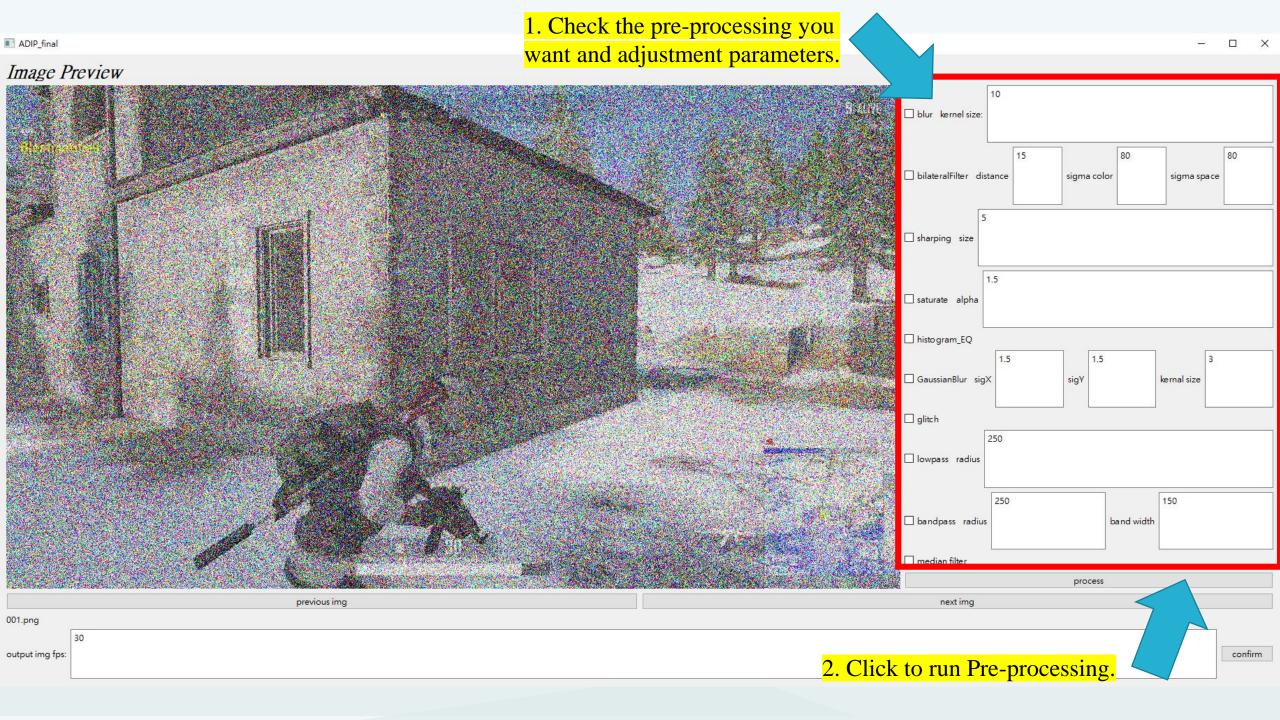


#### **Flow Chart**

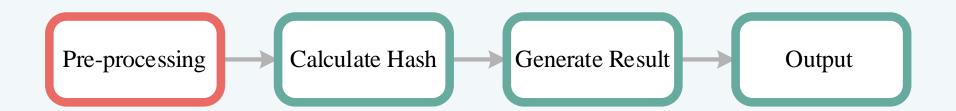


# **Choose Pre-process**





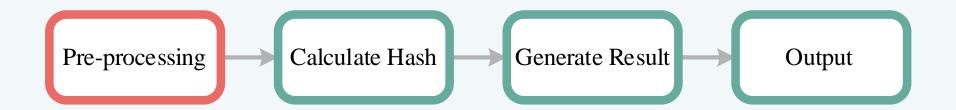
#### **Choose Pre-process**



- 1. Mean Filter
- 2. Bilateral Filter
- 3. Sharping
- 4. Saturate
- 5. Contrast Limited AHE

- 6. Gaussian Blur
- 7. Glitch (Preset)
- 8. Lowpass Filter
- 9. Bandpass Filter
- 10. Median Filter (Preset)

#### **Pre-processing - Preset Process**





(a) Original image

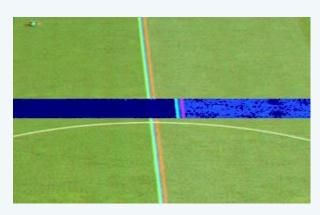


(b) Directly apply median filter

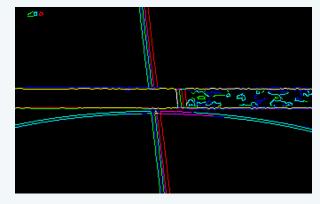


(c) Apply our median filter

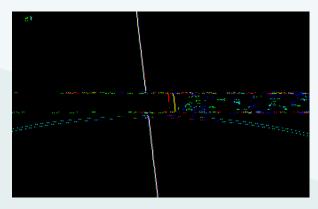
#### Pre-processing - Glitch (Color Plane Alignment)



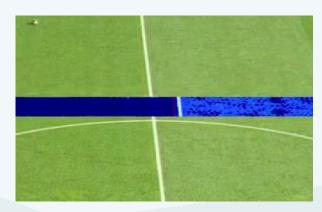
(a) Original image



(b) Edge of each color plate

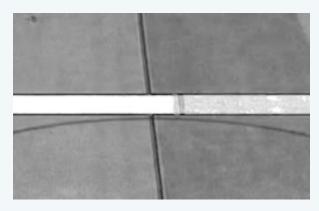


(c) Alignment results for each color plate



(d) Color plane alignment result

# **Pre-processing - Glitch**



(a) s channel

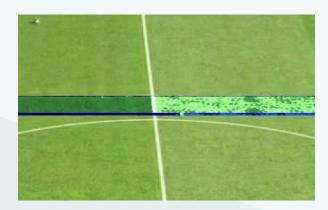


(c) After morphological processing





(b) Binarized image



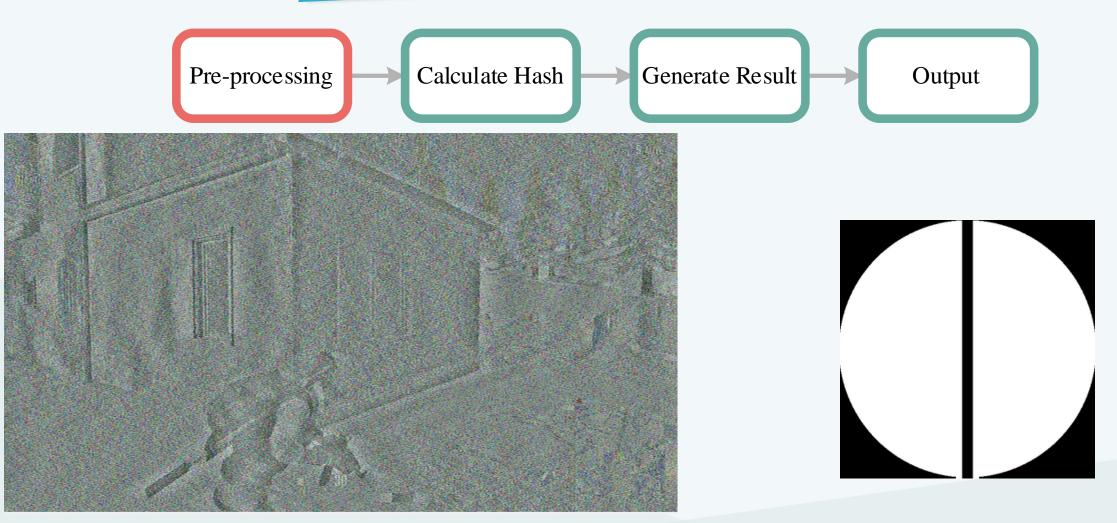
(d) Results of Glitch

0	0	1	1	1	0	0
0	1	1	1	1	1	0
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
0	1	1	1	1	1	0
0	0	1	1	1	0	0

(A) kernel 1

(B) kernel 2

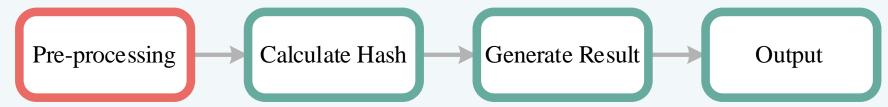
# Pre-processing - Lowpass Filter

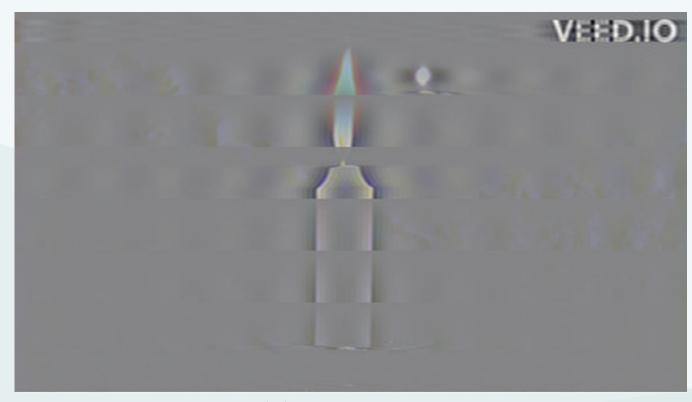


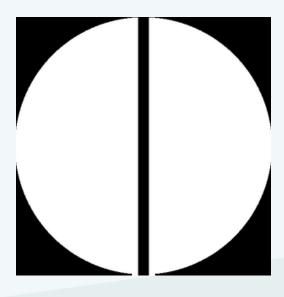
(a) Processed image

(b) Filter schematic

### **Pre-processing - Lowpass Filter**



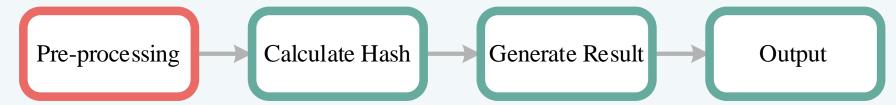


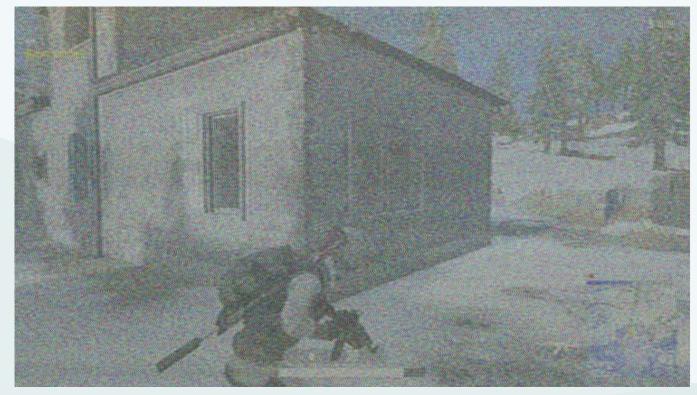


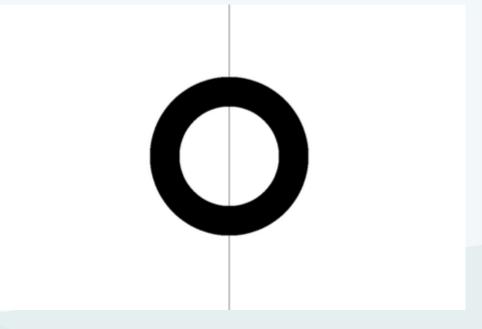
(a) Processed image

(b) Filter schematic

### **Pre-processing - Bandpass Filter**



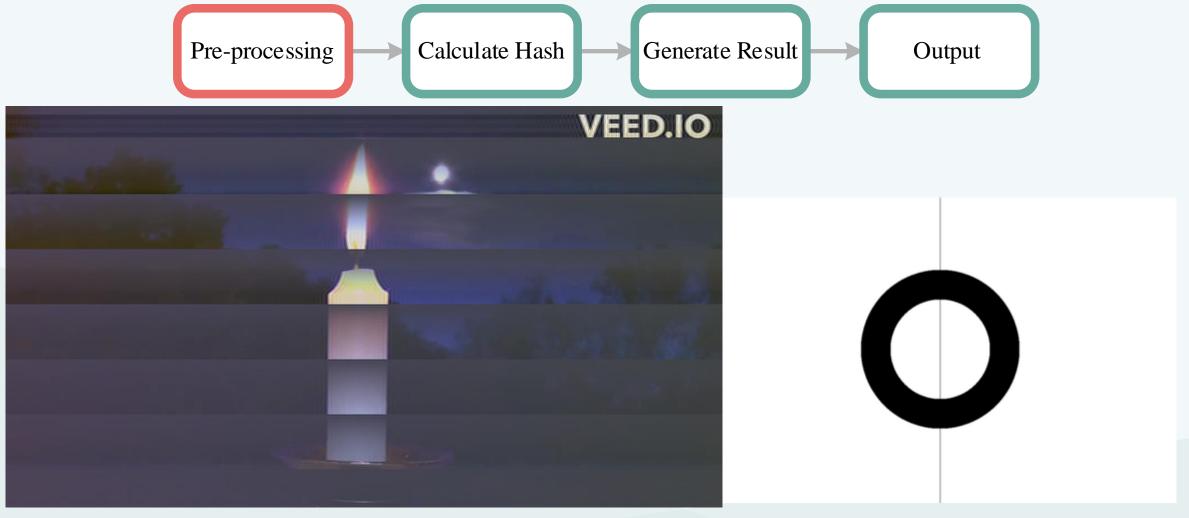




(a) Processed image

(b) Filter schematic

### Pre-processing - Bandpass Filter

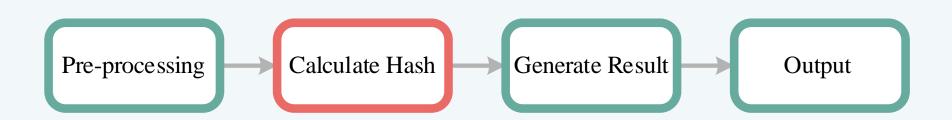


(a) Processed image

(b) Filter schematic



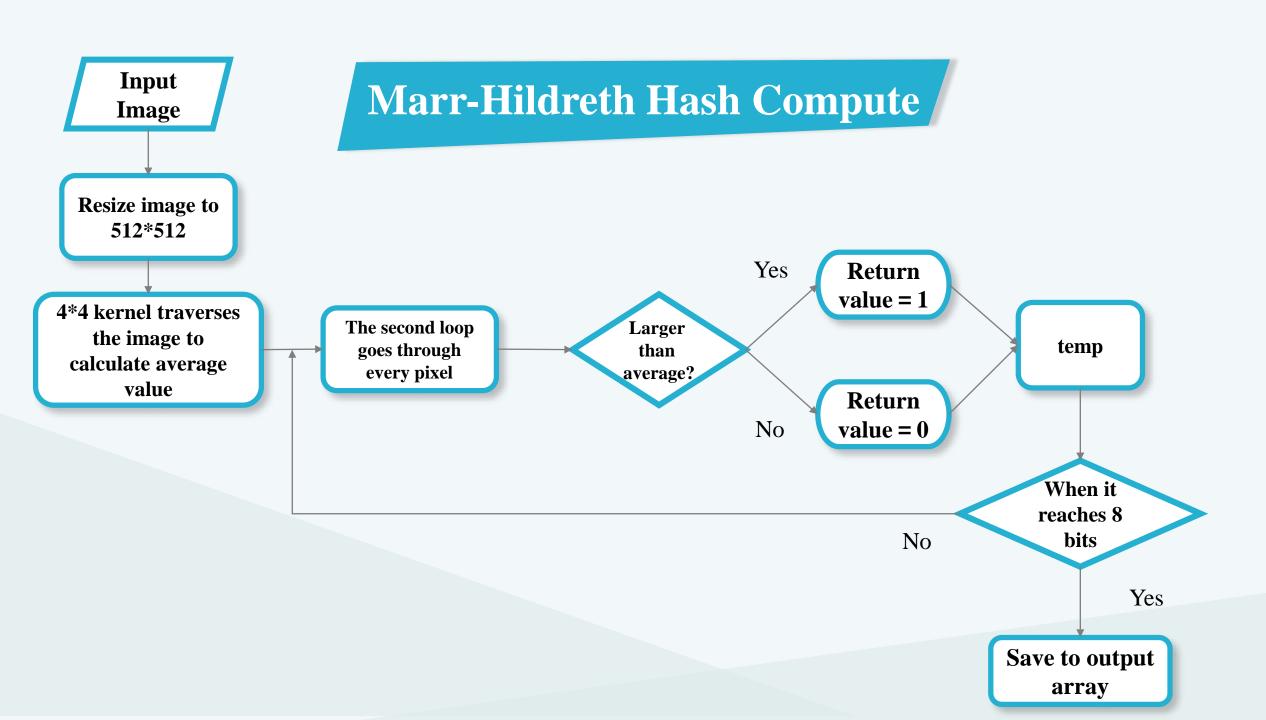
#### Marr-Hildreth Hash



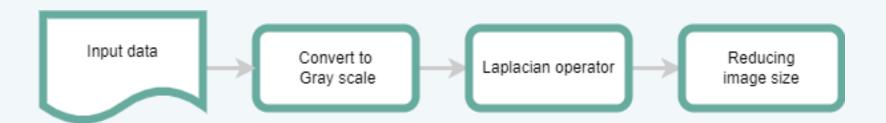
The basic idea of the Marr-Hildreth algorithm is to smooth the image using a Gaussian smoothing filter and then use the Laplace operator to calculate the second-order derivatives of the image. Based on the magnitude and direction of the second-order derivative, the edges of the image are calculated.

Advantages of the Marr-Hildreth algorithm:

- 1. Ability to detect fine edges in images.
- 2. Better performance when handling noisy images.



### Marr-Hildreth Hash Compute





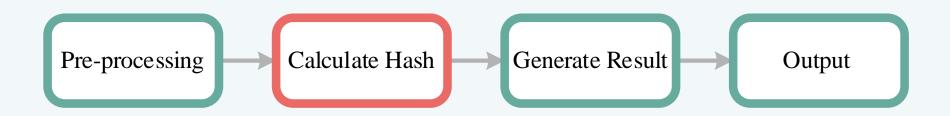


(b)

Original image size: 1280 \* 720

Hash result size: 72 \* 1

### **Marr-Hildreth Hash Compare**



In image processing, we use the Hamming distance to measure the difference between two images to compare the similarity of two images.

The Hamming distance is calculated as follows.

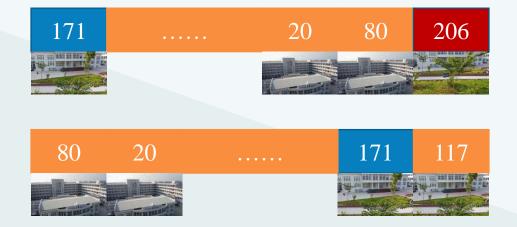
- 1. Obtain the pixel values of the binary strings of the two images.
- 2. Compare the two binary strings by bit and calculate the different digits.
- 3. Add up the different digits to get the Hamming distance.

The final result of 0 means the same picture, and the larger the number means the more dissimilar the two pictures are.

### **Sorting**







1. Find the image which with the largest hash difference.

2. Find the image with the smallest hash difference as next frame.

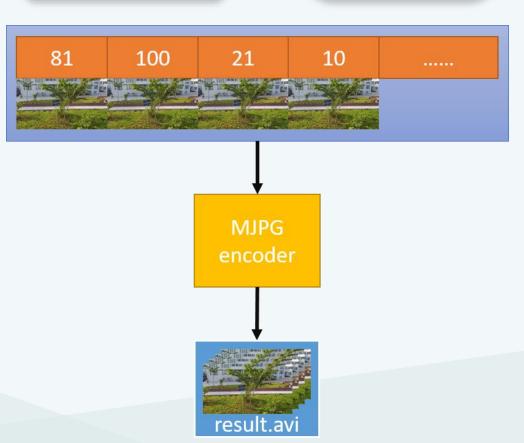
3. When the hash difference is greater than the average of the hash change, the whole sequence is reversed and then continued.

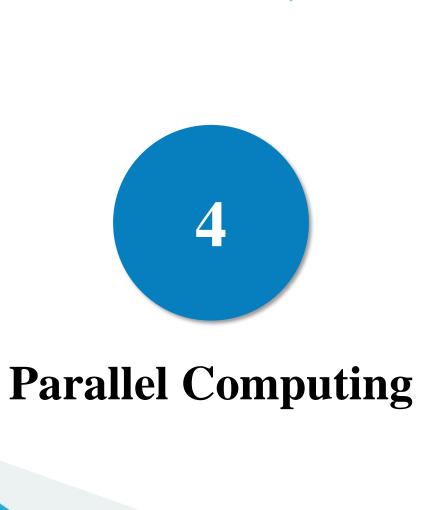


#### **Generate Result**



The final project uses the VideoWriter function in OpenCV to output avi video files using the MJPG encoder.





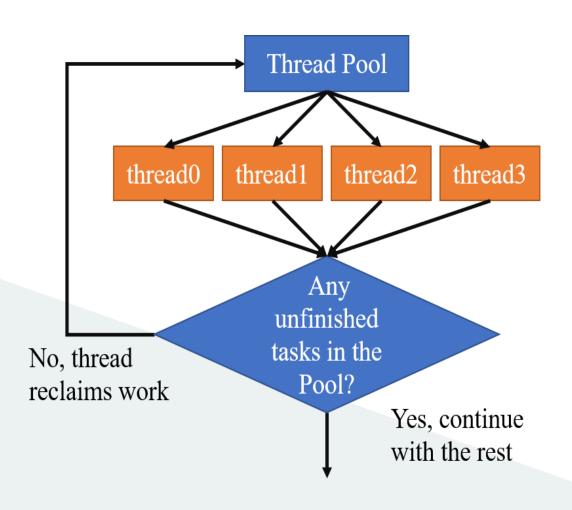
#### **Parallel Computing**



Since image pre-processing and hash calculation takes a lot of time, and due to the nature of C language, only one thread is used by default, the waiting time will be prolonged when multiple images need to be pre-processing and hash calculated, and the computer hardware cannot be used effectively.

However, if we can submit all the tasks to the Thread Pool first, and wait for the Thread Pool to finish all the tasks before running the rest of the programs, we can save a lot of time.

### **Parallel Computing**



Thread Pool: Assigning tasks.

Thread: Process the assigned tasks.

Thread: Complete the task and submit the results.



#### **Evaluation**

$$SRCC = \left| 1 - \frac{6\sum d_i^2}{n(n^2 - 1)} \right|$$

MSE = 
$$\frac{1}{mn} \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} [I(i,j) - K(i,j)]^2$$

MSE for evaluating grayscale images.

MSE = 
$$\frac{1}{mnc} \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} \sum_{c=0}^{2} [I(i,j,c) - K(i,j,c)]^{2}$$

MSE for evaluating channel color images.

MSE = 
$$\frac{1}{N} \sum_{x=0}^{N-1} \frac{1}{mnc} \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} \sum_{c=0}^{n-1} [I_x(i,j,c) - K_x(i,j,c)]^2$$
 MSE for evaluating videos.

 $i \cdot j$ : location of the pixel

c: image channel

x: video frame

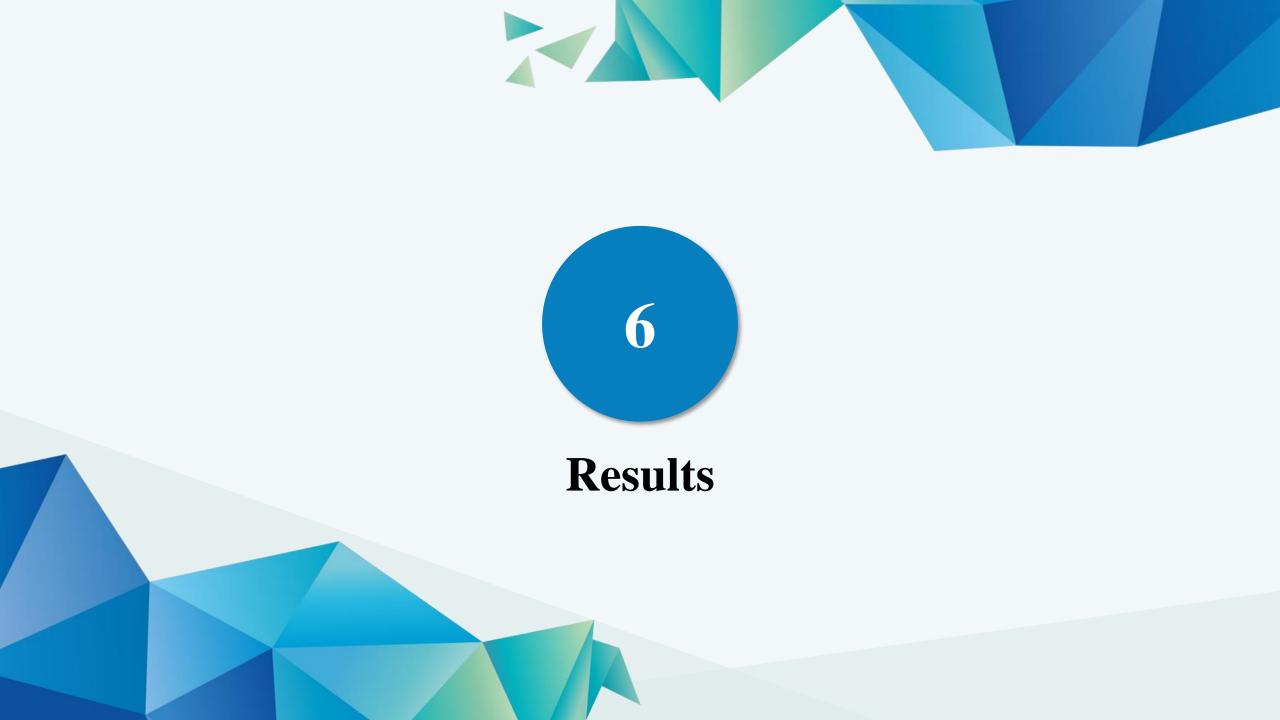
m: image height

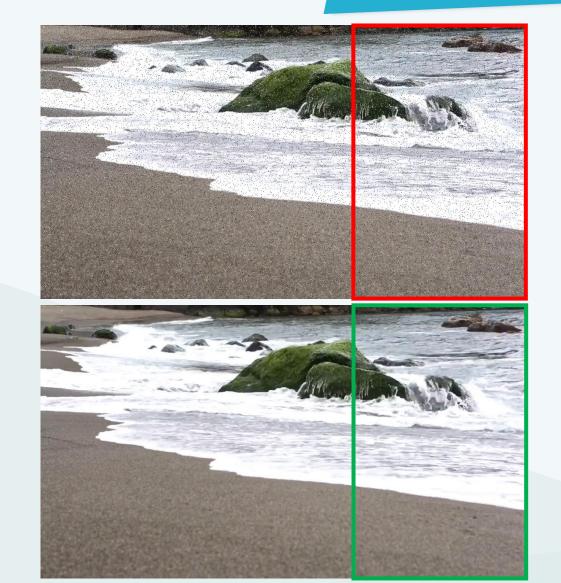
*n* : image width

N: video frame count

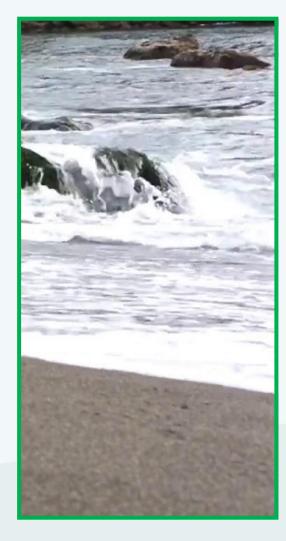
# **Evaluation**

資料集	圖片張數	前處理時 間 (sec)	SRCC	MSE	雜湊計算 與排序時 間 (sec)	處理時長 總計(sec)
beach	270	6.718	0.9995	1477.7	0.271	7.522
Boat_style	338	8.063	0.6948	2827.6	0.357	8.980
candle_style	340	11.31	0.0242	7211.3	0.365	12.26
CCTV	162	10.29	0.2582	1553.2	0.303	11.33
coastline	165	4.183	0.4650	15834	0.158	4.681
Desert	142	1.750	0.8145	18181	0.095	1.999
DMC	600	14.38	0.0037	2138.5	0.742	16.36
flyout	450	11.08	0.9860	1102.2	0.493	12.50
helltaker	300	21.46	0.5877	1323.3	0.584	23.16
PUBG	43	1.425	0.0219	47041	0.042	1.573
RushPixar	300	7.498	0.2731	17160	0.287	8.407
School	230	5.942	0.9908	3130.2	0.219	6.637
soccer_style	150	4.537	0.4749	3408.2	0.135	5.000
TKUC	153	3.765	0.5500	15613	0.147	4.220
typing	720	16.78	0.1162	1939.3	0.926	19.070
全體平均	290.866	8.612	0.484	9329.366	0.341	9.579





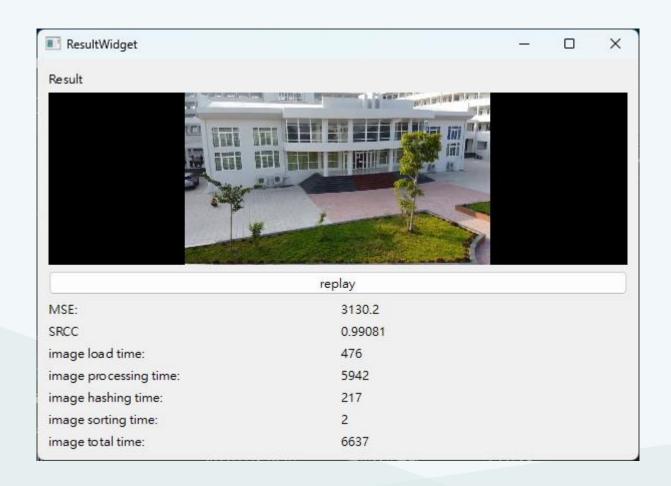












# **Work Distribution Chart**

_		前處理 Case 1	前處理 Case 2	前處理 Case 3	雜湊計算 排序	評估與輸出	優化執行 程式整合	報告書 與簡報製 作	總計
	黄俊傑	<b>✓</b> 100%			<b>√</b> 50%	<b>✓</b> 100%	<b>✓</b> 15%	<b>✓</b> 15%	34%
	吳翊睿		<b>✓</b> 100%				<b>√</b> 85%	<b>✓</b> 15%	34%
	吳佩霖			<b>✓</b> 100%	<b>√</b> 50%			<b>✓</b> 70%	32%

#### Reference

- [1] <a href="https://github.com/opencv/opencv\_contrib">https://github.com/opencv/opencv\_contrib</a>
- [2] https://github.com/vit-vit/CTPL
- [3] https://ppfocus.com/0/mic6b8c41.html

# Thank you for listening