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RST Template Matching-Based PCB Alignment System on Embedded GPU Board

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Abstract

The present study proposed an embedded printed circuit board (PCB) alignment system, in which a rotation, scale and translation (RST) template-matching algorithm was employed to locate the marks on the PCB surface. The coordinates and angles of the detected marks were then compared with the reference values which were set by users, and the difference between them was used to adjust the position of the vision system accordingly. To improve the positioning accuracy, the angle and location matching process was performed in refinement processes. To overcome the matching time, in the present study we accelerated the rotation matching by eliminating the weak features in the scanning process and converting the normalized cross correlation (NCC) formula to a sum of products. Moreover, the scanning time was reduced by implementing the entire RST process in parallel on threads of a graphics processing unit (GPU) by applying hash functions to find refined positions in the refinement matching process. The experimental results showed that the resulting matching time was around 32x faster than that achieved on a conventional central processing unit (CPU) for a test image size of 1280 × 960 pixels.

Overview System

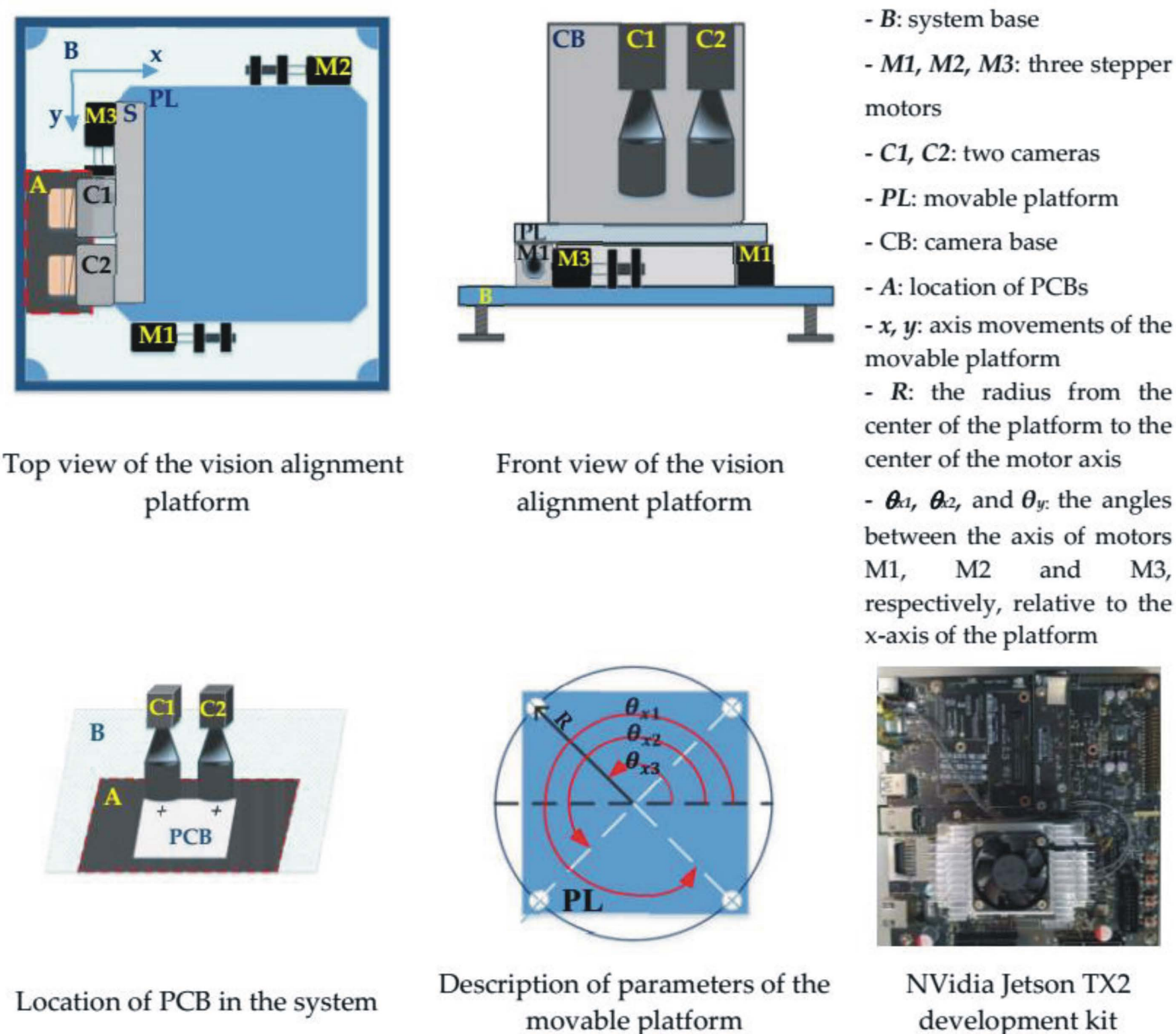


Figure 1. Overview of the hardware of the proposed embedded printed circuit board (PCB) alignment system.

RST Refinement Template Matching Algorithm

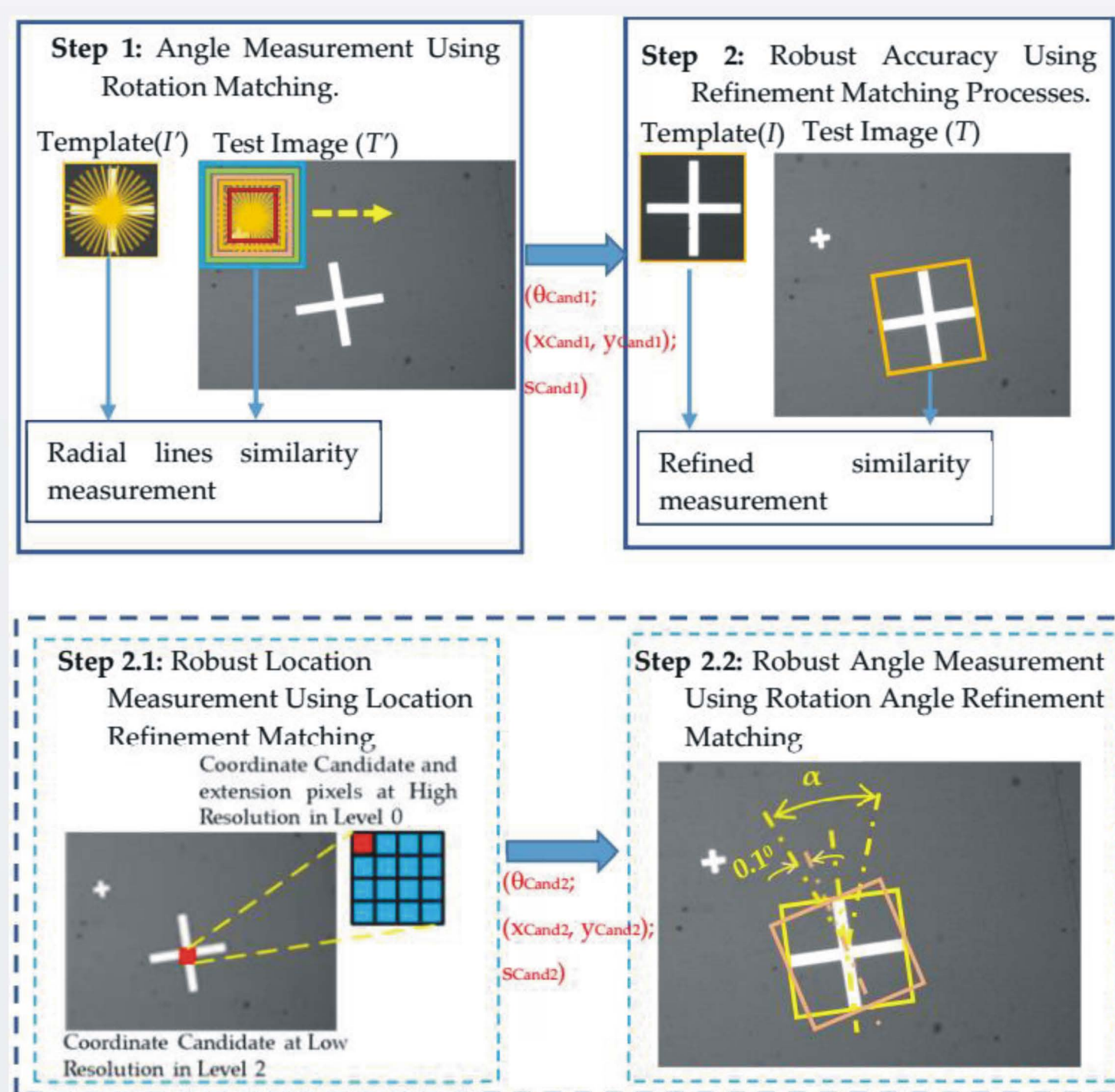


Figure 2. Global framework of the RST refinement template matching algorithm.

Acceleration of the RST Refinement Template Matching Algorithm

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Algorithm 1 Pseudo-code of Location Refinement Matching using CUDA
1: Inputs: Test image I, size of template w × h, coordinate candidates (xCand1, yCand1), angle candidates θCand1, scale candidates sCand1, pyramid level Np, number of candidate NCand1
2: Outputs: Correlation coefficient ηNCC
3: X index: idxX ← blockDim.x + blockIdx.x + threadIdx.x //Number of extension pixels
4: Y index: idxY ← blockDim.y + blockIdx.y + threadIdx.y //Number of candidates
5: if (idxX < 2Np) and (idxY < NCand1) then
6:   Coordinate X: xRefined ← (idxX mod 2Np) + (xCand1 + 2Np) //Refinement coordinate x
7:   Coordinate Y: yRefined ← (idxY div 2Np) + (yCand1 + 2Np) //Refinement coordinate y
8:   Scale s: s ← sCand1[idxY]
9:   Angle θ: θ ← θCand1[idxY]
10:  for j in h do
11:    for i in w do
12:      Collect intensity pixel values inside a search window with a center point at (xRefined, yRefined), an orientation: θ, and a scale: s
13:    end
14:  end
15: Calculate the NCC score ηNCC between the template and the search window
16: end
17: Return ηNCC

Algorithm 2 Pseudo-code of Rotation Angle Refinement Matching using CUDA
1: Inputs: Test image I, size of template w × h, coordinate candidates (xCand2, yCand2), angle candidates θCand2, scale candidates sCand2, angular resolution α, number of candidate NCand2
2: Outputs: Correlation coefficient ηNCC
3: X index: idxX ← blockDim.x + blockIdx.x + threadIdx.x //Angular resolution
4: Y index: idxY ← blockDim.y + blockIdx.y + threadIdx.y //Number of candidates
5: if (idxX < α + 10) and (idxY < NCand2) then
6:   Angle θ: θRefined ← (idxX / 10.0) + θCand2[idxY] //Refined angle
7:   Coordinate X: x ← xCand2[idxY]
8:   Coordinate Y: y ← yCand2[idxY]
9:   Scale s: s ← sCand2[idxY]
10:  for j in h do
11:    for i in w do
12:      Collect intensity pixel values inside a search window using a bilinear interpolation approach with a center point at (x, y), an orientation: θRefined, and a scale: s
13:    end
14:  end
15: Calculate the NCC score ηNCC between the template and the search window
16: end
17: Return ηNCC

```

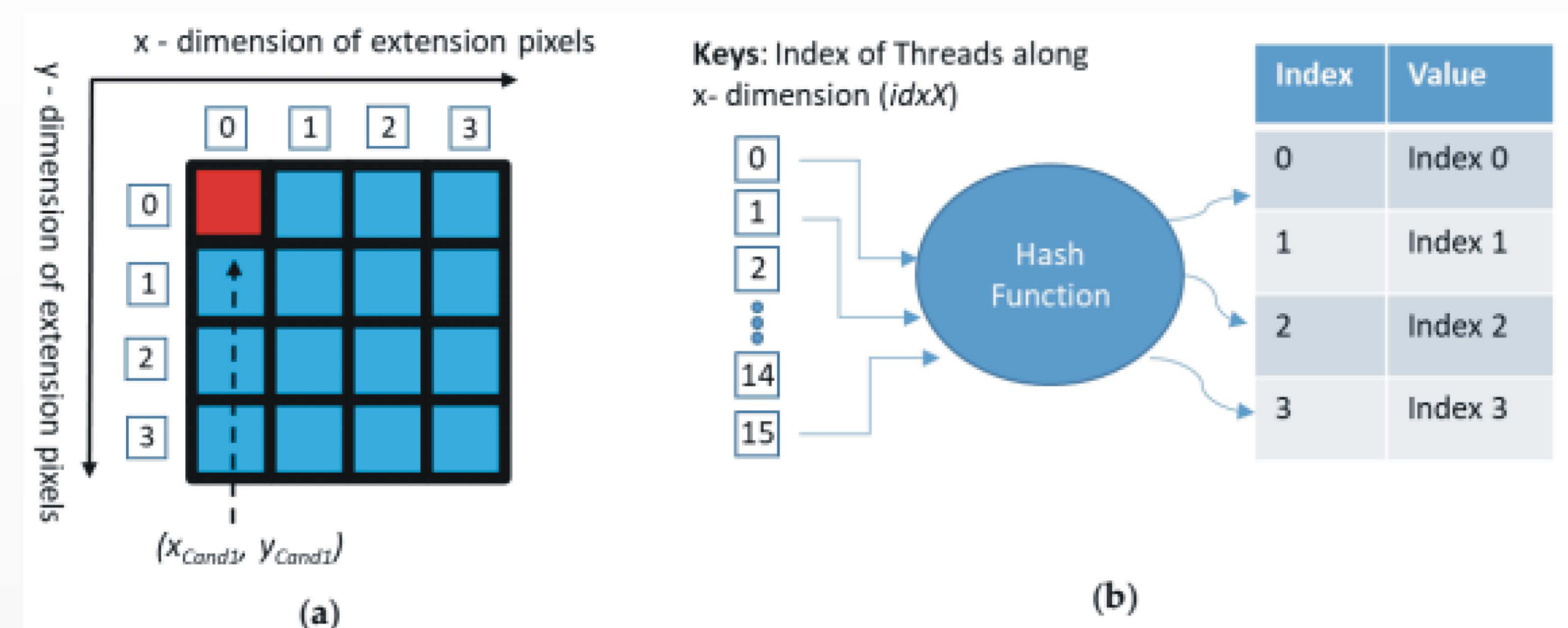


Figure 3. Reducing matching time by using the hash table for CUDA coding. (a) Description of coordinate candidate (x_{Cand1} , y_{Cand1}) and extension pixels; (b) description of finding x and y indexes of the extension pixels.

Experimental Results

Table 1. Comparison of average matching time and accuracy of RST template matching on different platforms and with different image sizes.

Methods	Image Size					
	640x480 (pixels)		800x600 (pixels)		1280x960 (pixels)	
	Time(s)	Accuracy	Time(s)	Accuracy	Time(s)	Accuracy
PC-based Platform						
PC-RST[1]	0.618	98.0%	1.075	98.0%	1.326	97.5%
FAsT-Match[2]	0.1	99.9%	-	-	0.4	99.8%
PC-Improved RST	0.099	97.5%	0.186	95.5%	0.291	97.0%
Embedded System-based Platform						
em-RST	1.914	97.0%	5.668	96.5%	9.517	95.0%
emCPU-Improved RST	0.568	92.0%	1.633	95.0%	3.664	98.0%
emGPU-Improved RST	0.197	96.5%	0.342	95.5%	0.301	96.0%

The results show that the performance of the proposed method is faster than others methods on both PC platform (PC-Improved RST) and embedded platform (emGPU-Improved RST), while having no significant effect on the accuracy of the matching results.

References

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- [2] Liu, B.; Shu, X.; Wu, X. *Fast screening algorithm for rotation invariant template matching*. In Proceedings of 25th IEEE Int. Conf. Image Processing (ICIP), 2018; pp. 3708-3712.

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