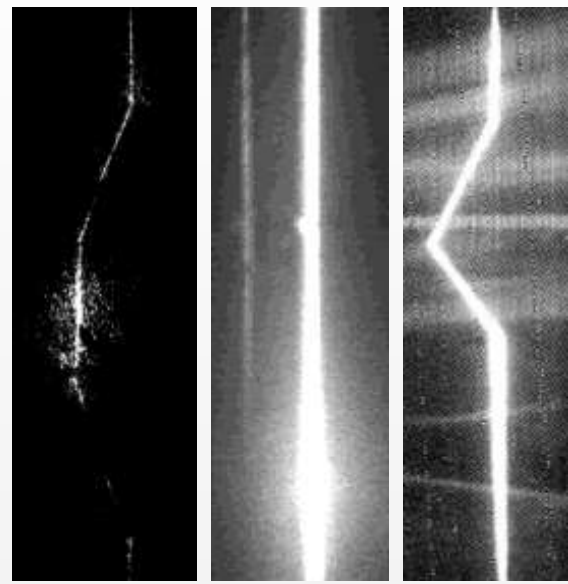
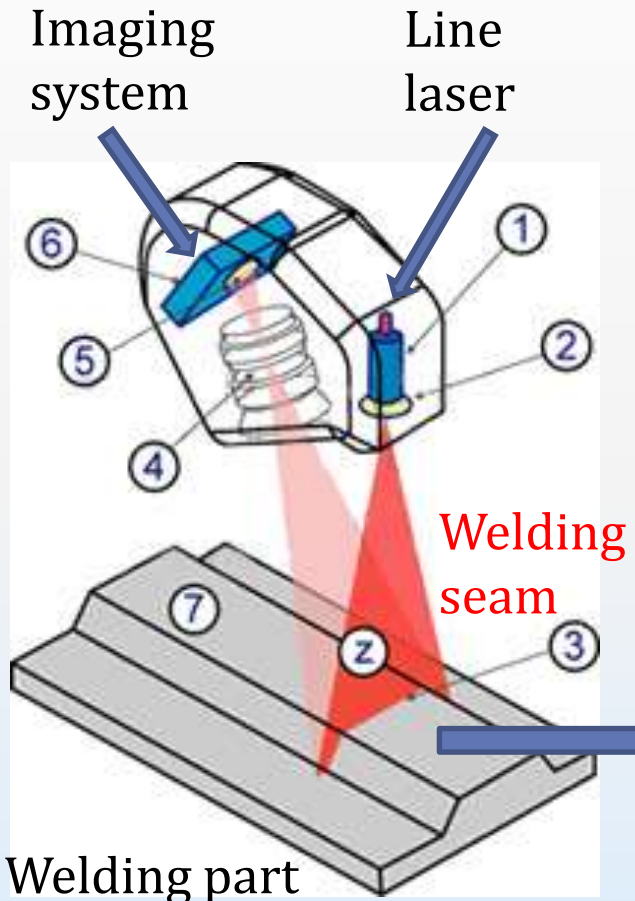


Briefing

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- ◆ Research Interests (computer vision)
 - ◆ **Low complexity image processing algorithms**
 - ◆ **Image based personal authentication**

Low complexity image processing algorithms

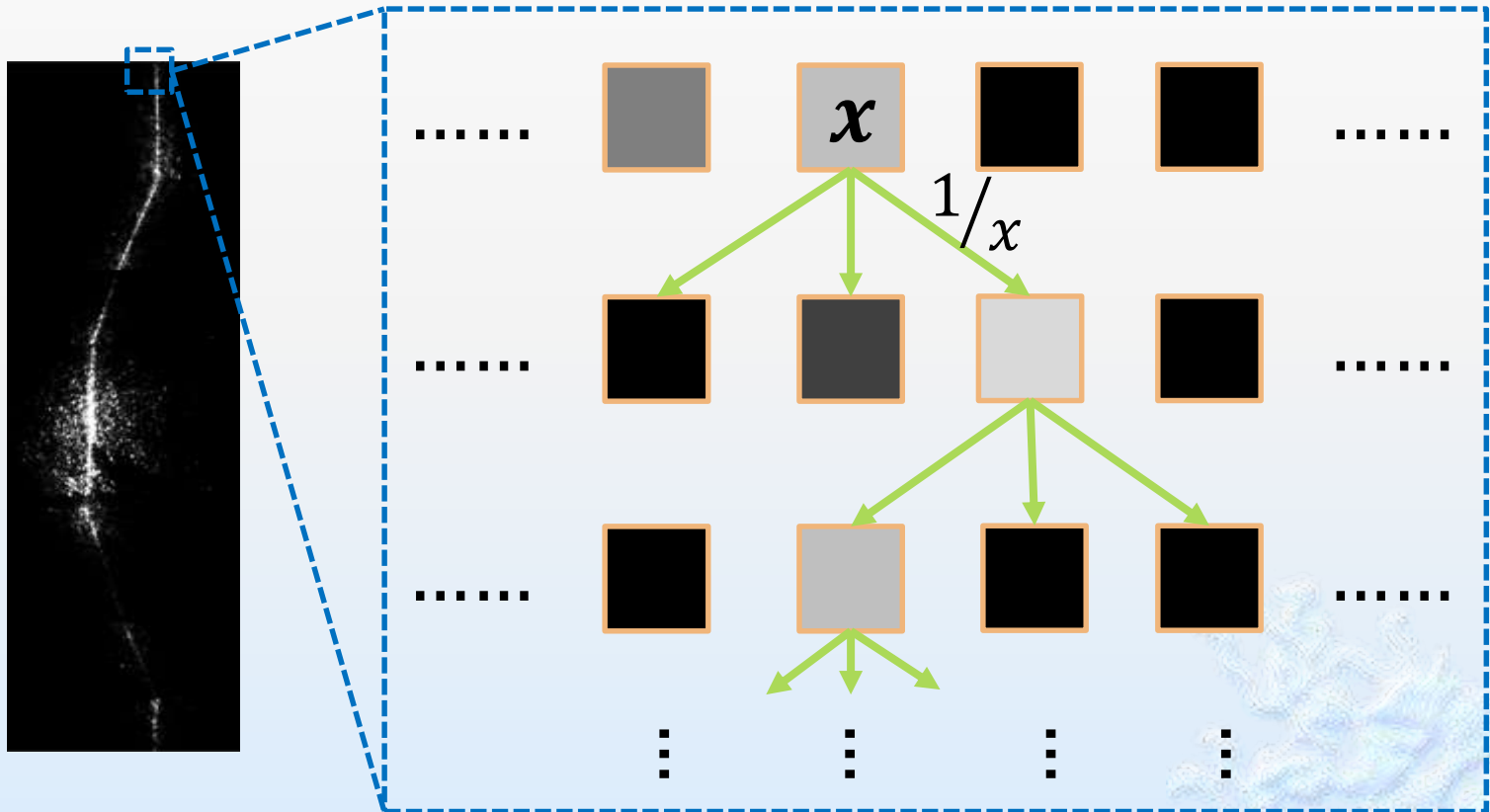
- ◆ *Efficient welding seam tracking*



Low complexity image processing algorithms

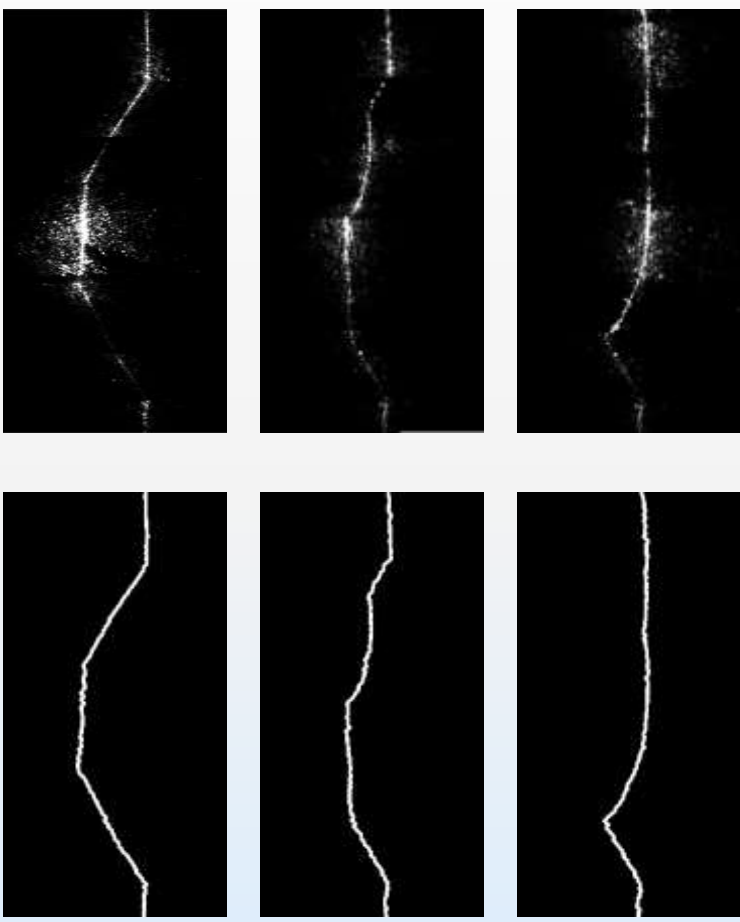
- ◆ *Efficient welding seam tracking*

Shortest path on DAG
Complexity: **$O(\text{pixels})$**

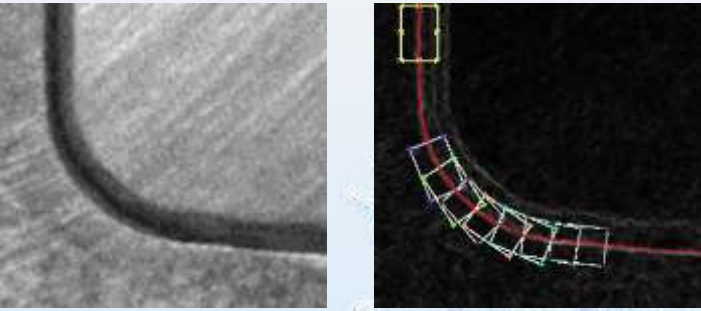
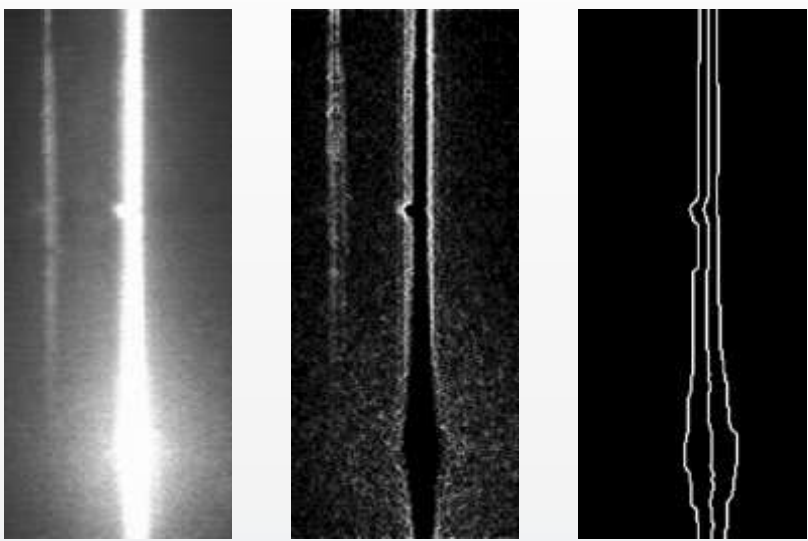


Low complexity image processing algorithms

- ◆ *Efficient welding seam tracking*

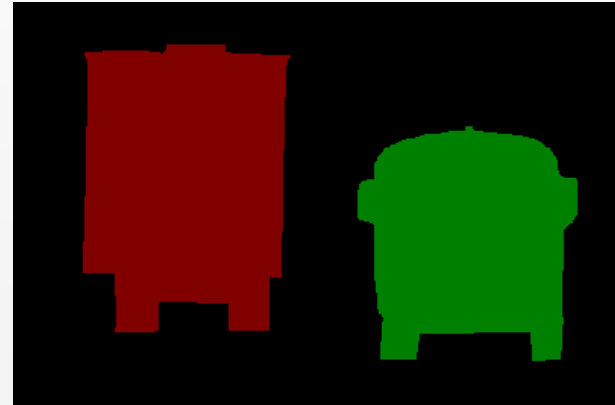
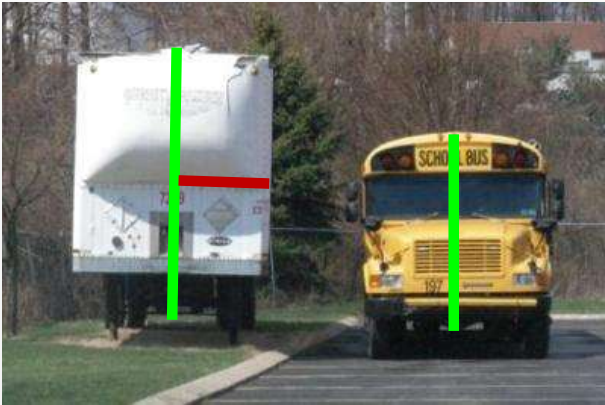


gradient



Low complexity image processing algorithms

- ◆ *Image symmetry detection using longest palindromic substring (LPS) detection*



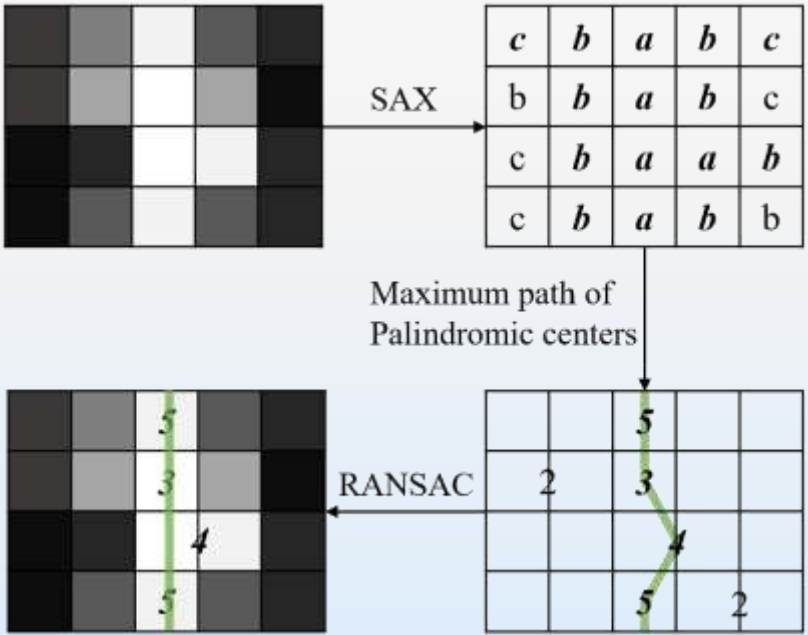
LPS: E F **C B D B C** A E

SVEUGJVABEQ**TQLOHTER**RETHOLQTQEBAVJG**BPQT**

Manacher Algorithm('94) : $O(n)$!

Low complexity image processing algorithms

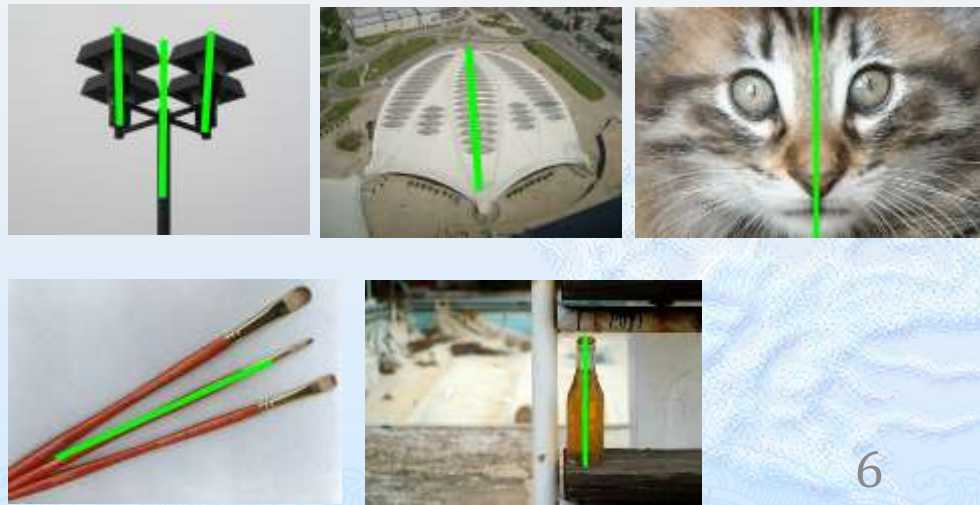
◆ *Image symmetry detection using LPS detection*



Complexity: ***O(pixels)***

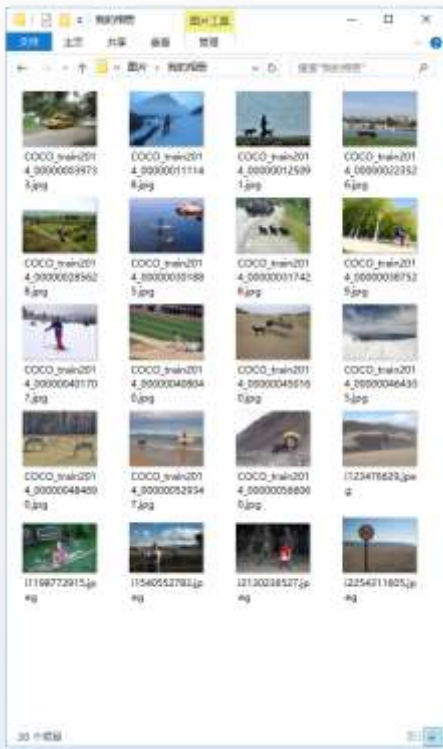
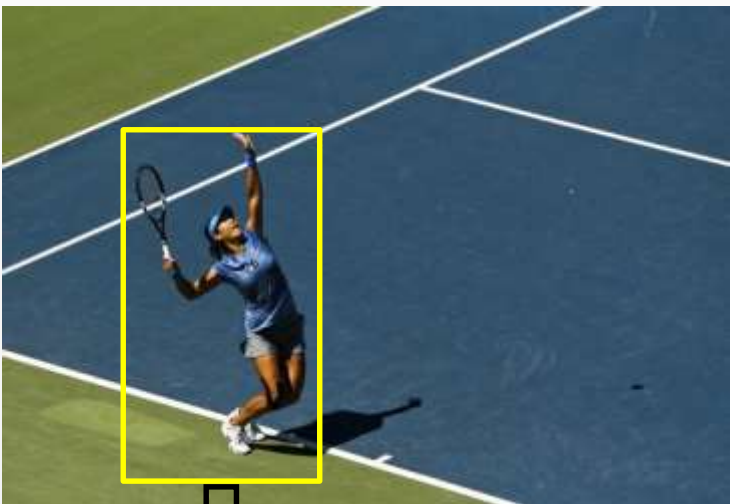
Two orders of magnitude faster (~0.1 sec)

Comparable accuracy(92% vs 92%, 79% vs 84%)



Low complexity image processing algorithms

Automatic image cropping/thumbnailing



Content aware thumbnailing



Low complexity image processing algorithms

Automatic image cropping/thumbnailing

Matrix minimum coverage problem: locating the smallest area rectangle with enough summed value

Algorithm 4 *Fixed_AspRatio_Rectangle*(G, τ, r)

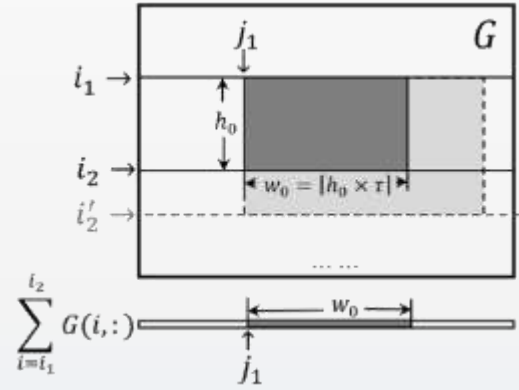
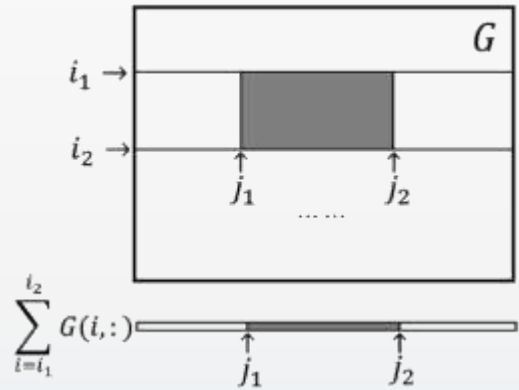
Input: G is a non-negative attention map with size $m \times n$; τ is the percentage of total attention to be preserved; r is the aspect ratio of cropping rectangle; suppose integral maps G^+ and G_c^+ are calculated.

Output: The smallest rectangle \hat{R} with aspect ratio r that satisfies (1); four values to define \hat{R} : i and j are the upper left corner coordinates, w and h are the width and height.

```

1:  $i \leftarrow 0, j \leftarrow 0, w \leftarrow \infty, h \leftarrow \infty$ 
2:  $i_1 \leftarrow 1, i_2 \leftarrow 1, T \leftarrow \tau G^+(m, n), S_{min} \leftarrow -1$ 
3: repeat
4:    $h_0 \leftarrow i_2 - i_1 + 1, w_0 \leftarrow \lceil h_0 \times r \rceil$ 
5:   if  $w_0 > n$  then
6:      $i_1 \leftarrow i_1 + 1$ 
7:   else
8:      $\hat{a} = G_c^+(i_2, :) - G_c^+(i_1 - 1, :)$ 
9:      $j_1, S_0 \leftarrow \text{maxSubarrayFL}(\hat{a}, w_0, T)$ 
10:    if  $j_1 > 0$  then
11:      if  $w_0 h_0 < wh \vee (w_0 h_0 = wh \wedge S_0 > S_{min})$  then
12:         $i \leftarrow i_1, j \leftarrow j_1, w \leftarrow w_0, h \leftarrow h_0$ 
13:         $S_{min} \leftarrow S_0$ 
14:      end if
15:       $i_1 \leftarrow i_1 + 1$ 
16:    else
17:       $i_2 \leftarrow i_2 + 1$ 
18:    end if
19:  end if
20: until  $i_2 > m \wedge i_1 \geq m$ 
21: return  $i, j, w, h$ 

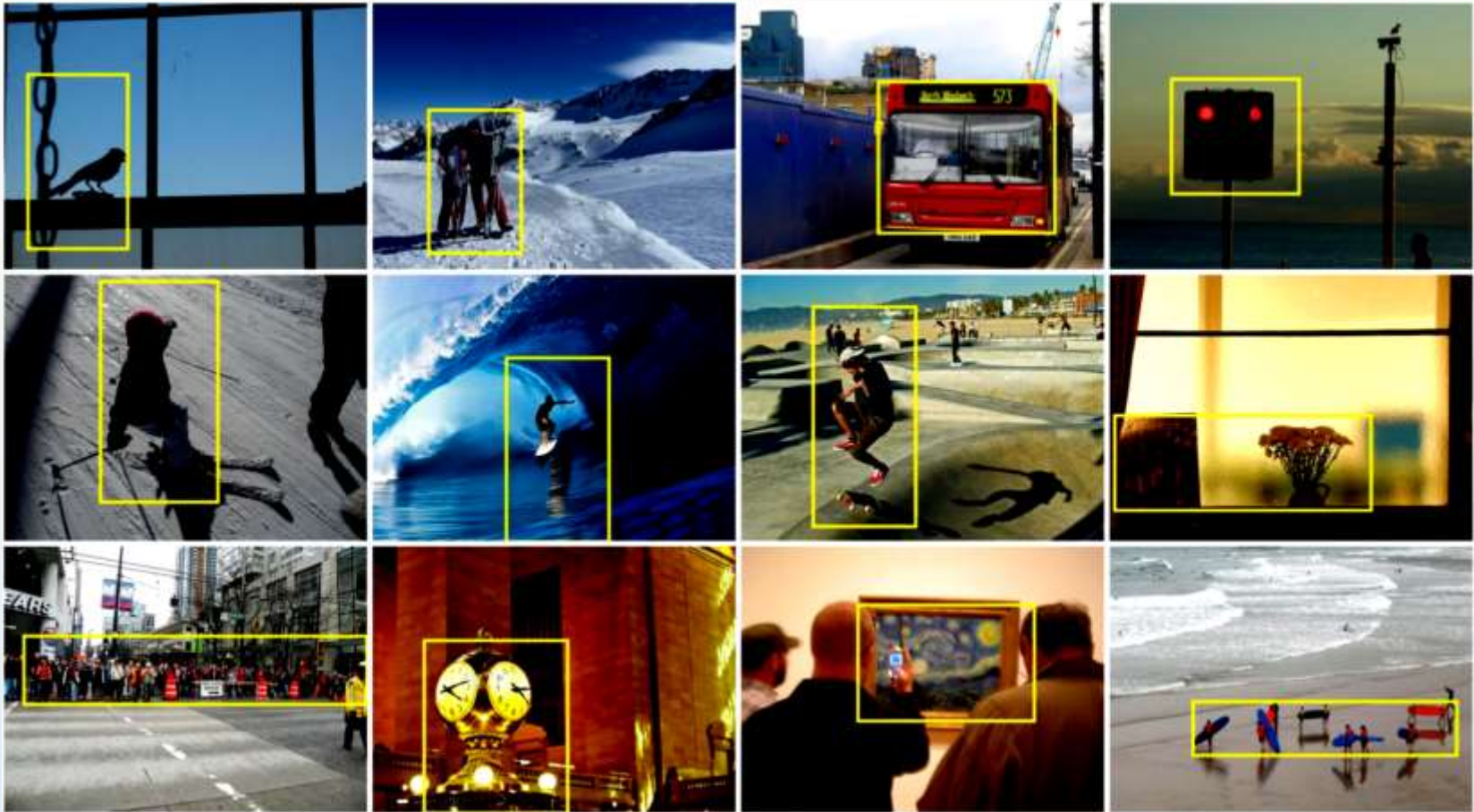
```



The computational complexity can be as low as **$O(\text{pixels})$** given the target aspect ratio !

Low complexity image processing algorithms

- ◆ *Automatic image cropping/thumbnailing*



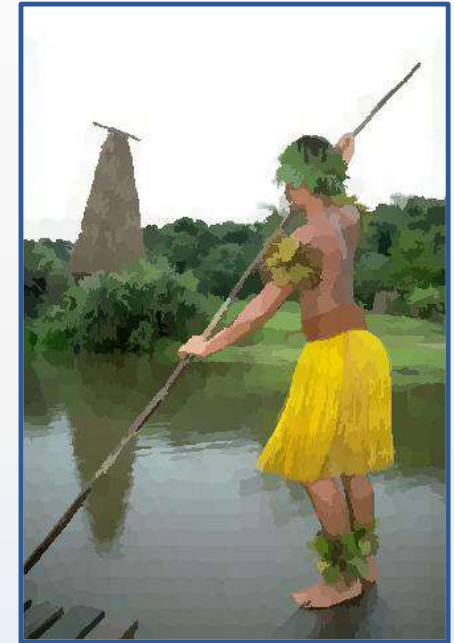
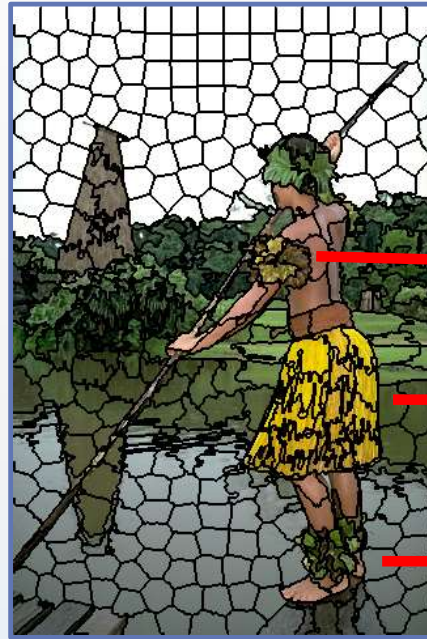
Low complexity image processing algorithms

- ◆ *Automatic image cropping/thumbnailing*



Low complexity image processing algorithms

- ◆ Linear spectral clustering superpixel



Low complexity image processing algorithms



$$\sum_{k=1}^K \frac{\sum_{p \in \pi_k} \sum_{q \in \pi_k} w(p)\phi(p) \cdot w(q)\phi(q)}{\sum_{p \in \pi_k} w(p)}$$

Global methods
 Preserving global image structure,
 high superpixel compactness;
 high complexity $O(\text{pixels}^{1.5})$, low
 boundary adherence.

Local methods
 High boundary adherence, low
 complexity $O(\text{pixels})$;
 Poor global structure, low
 superpixel compactness

$$\sum_{k=1}^K \sum_{p \in \pi_k} w(p) \left\| \phi(p) - \frac{\sum_{q \in \pi_k} w(q)\phi(q)}{\sum_{q \in \pi_k} w(q)} \right\|^2$$



Low complexity image processing algorithms

◆ Linear spectral clustering superpixel

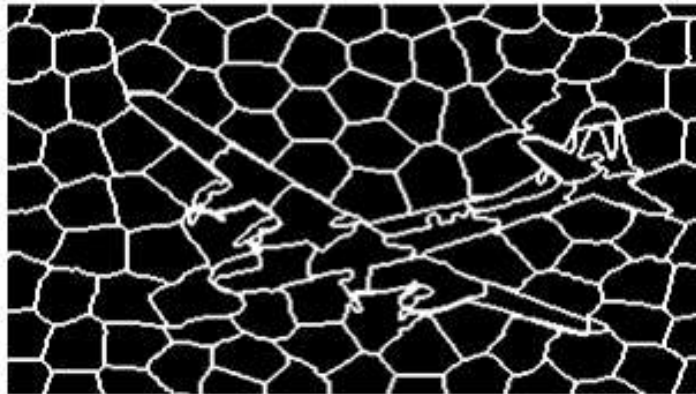
Global methods and **local methods** actually may share the same optimization objection!

$$\begin{aligned} & \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \left\| \phi(\mathbf{p}) - \frac{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q})} \right\|^2 \\ &= \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \left(\|\phi(\mathbf{p})\|^2 - 2\phi(\mathbf{p}) \cdot \frac{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q})} + \left\| \frac{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q})} \right\|^2 \right) \\ &= \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \|\phi(\mathbf{p})\|^2 - \sum_{k=1}^K \left(2 \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \phi(\mathbf{p}) \cdot \frac{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q})} - \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \left\| \frac{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{q} \in \pi_k} w(\mathbf{q})} \right\|^2 \right) \\ &= \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \|\phi(\mathbf{p})\|^2 - \sum_{k=1}^K \left(2 \frac{\|\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \phi(\mathbf{p})\|^2}{\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p})} - \frac{\|\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \phi(\mathbf{p})\|^2}{\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p})} \right) \\ &= \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \|\phi(\mathbf{p})\|^2 - \sum_{k=1}^K \frac{\|\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \phi(\mathbf{p})\|^2}{\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p})} \\ &= \sum_{k=1}^K \sum_{\mathbf{p} \in \pi_k} w(\mathbf{p}) \|\phi(\mathbf{p})\|^2 - \sum_{k=1}^K \frac{\sum_{\mathbf{p} \in \pi_k} \sum_{\mathbf{q} \in \pi_k} w(\mathbf{p}) \phi(\mathbf{p}) \cdot w(\mathbf{q}) \phi(\mathbf{q})}{\sum_{\mathbf{p} \in \pi_k} w(\mathbf{p})} \end{aligned}$$

Low complexity image processing algorithms

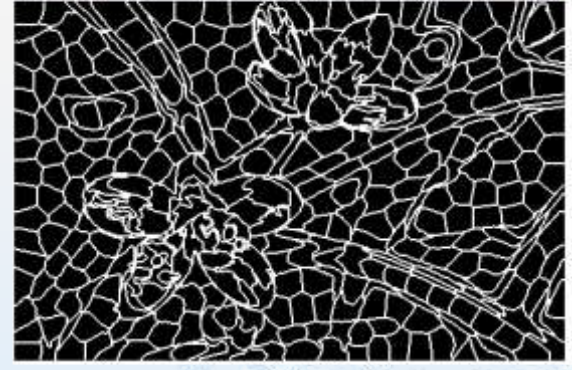
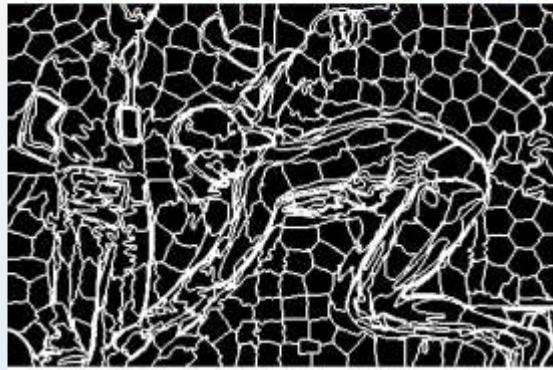
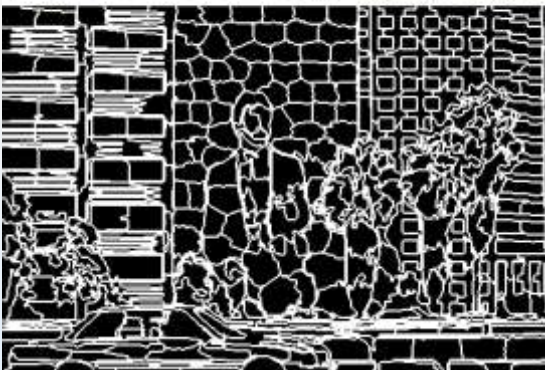
- ◆ Linear spectral clustering superpixel

LSC: an $O(\text{pixels})$ algorithm capable of generating superpixels with both high boundary adherence and satisfactory global structure preservance.



Low complexity image processing algorithms

- ◆ Linear spectral clustering superpixel



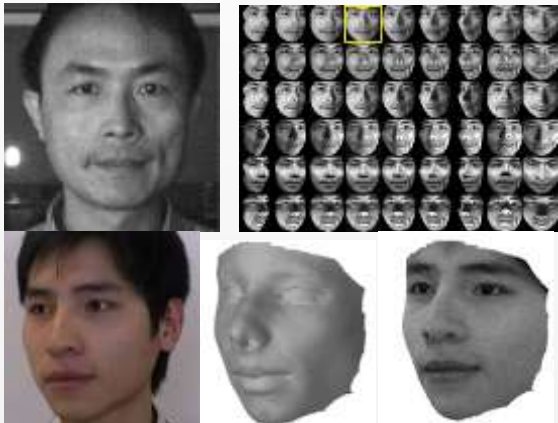
Z.Q. Li & J.S. Chen, Superpixel Segmentation using Linear Spectral Clustering, CVPR'15

Image based personal authentication

- ◆ Face recognition for open environments



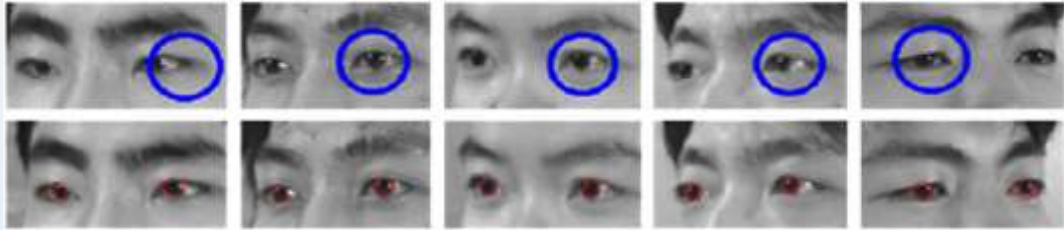
Illumination (ECCV'10)



3D Modeling (ICIP'13,CVPRW'14)



Face quality (SPL'15)



Eye tracking (PR'13)



Facial symmetry (SPL'14)



Super-resolution (ACCV'12)

Image based personal authentication

- ◆ Face illumination transformation



Locally constrained global optimization



$$\arg \min_{\alpha, I_t^B} \sum_i \left(\underbrace{\sum_{j \in \omega_i} \left(I_t^B(j) - \alpha_i I_t^A(j) \right)^2}_{\text{local}} + \underbrace{\lambda (\alpha_i - \tau_i)^2}_{\text{global}} \right)$$

Image based personal authentication

◆ Face 3D modeling

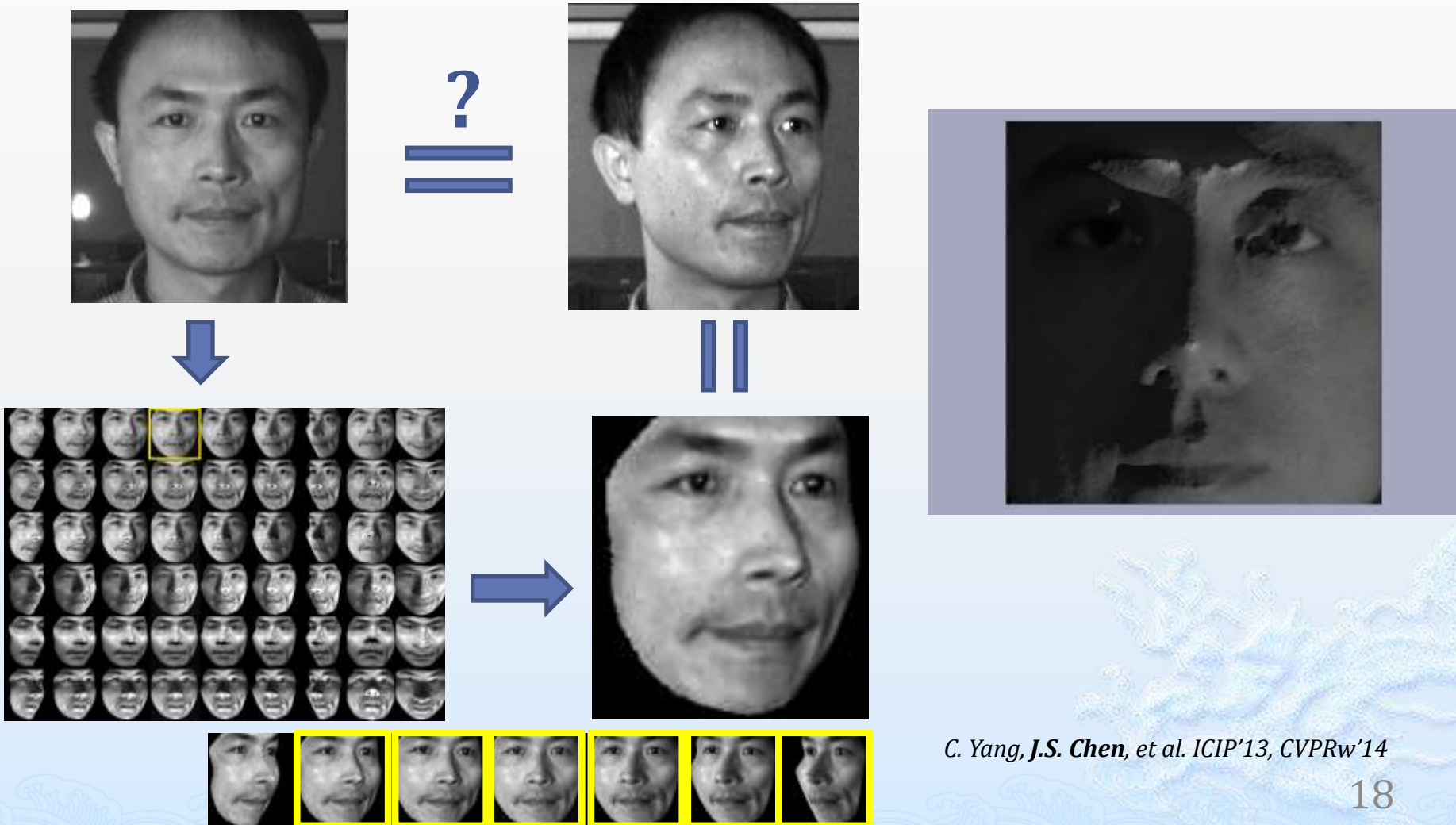


Image based personal authentication

- ◆ Face super-resolution based on manifold learning



LR Recont. HR

Frequency domain manifold learning

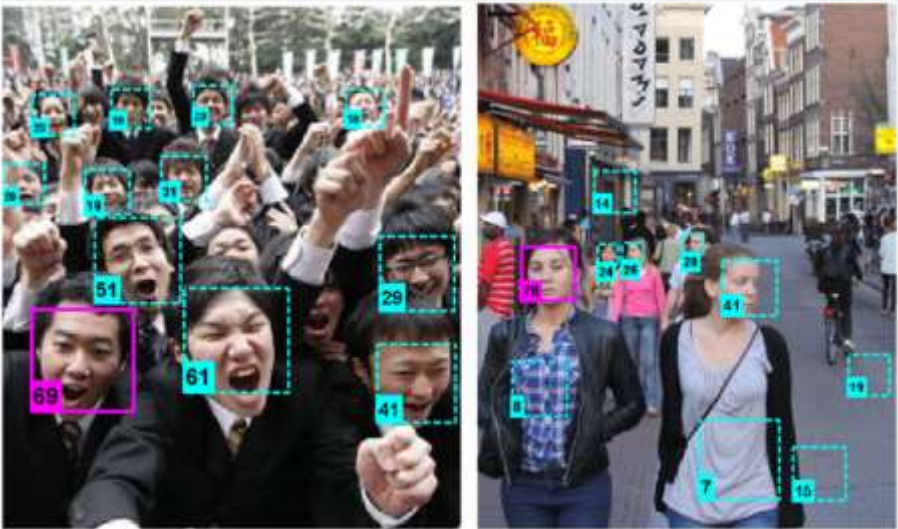
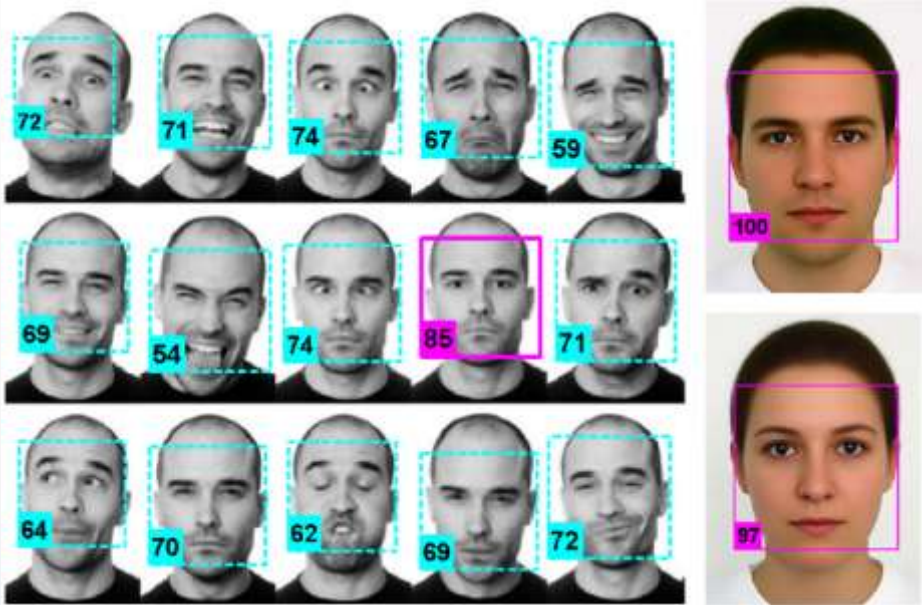
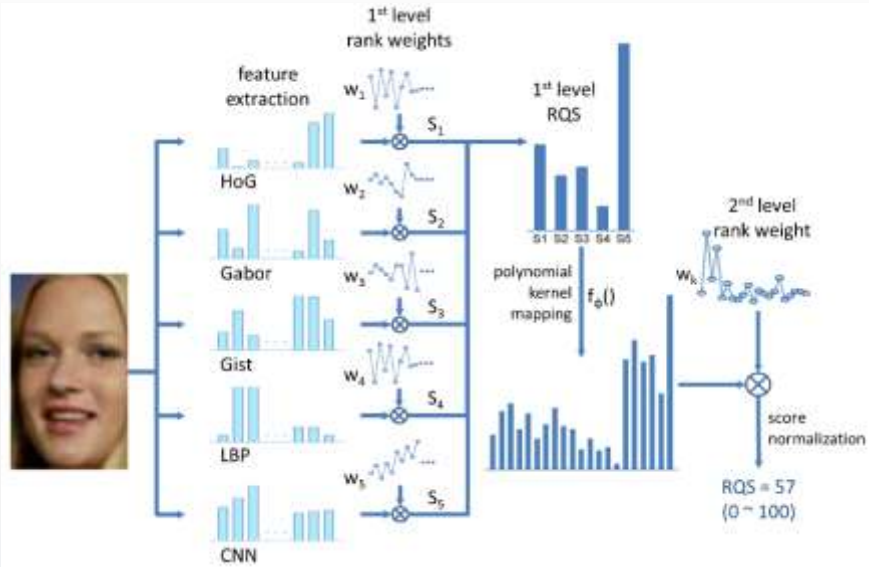


SSIM 93.9% → 95.4%
PSNR 31.9dB → 33.7dB

IEICE Trans 2012
Science China 2012
ACCV 2012

Image based personal authentication

- ◆ Face quality assessment based on rank learning



- ◆ A framework for face quality assessment
- ◆ Using rank learning to expand training set space
- ◆ Improving video based face recognition robustness in real life

Thanks !