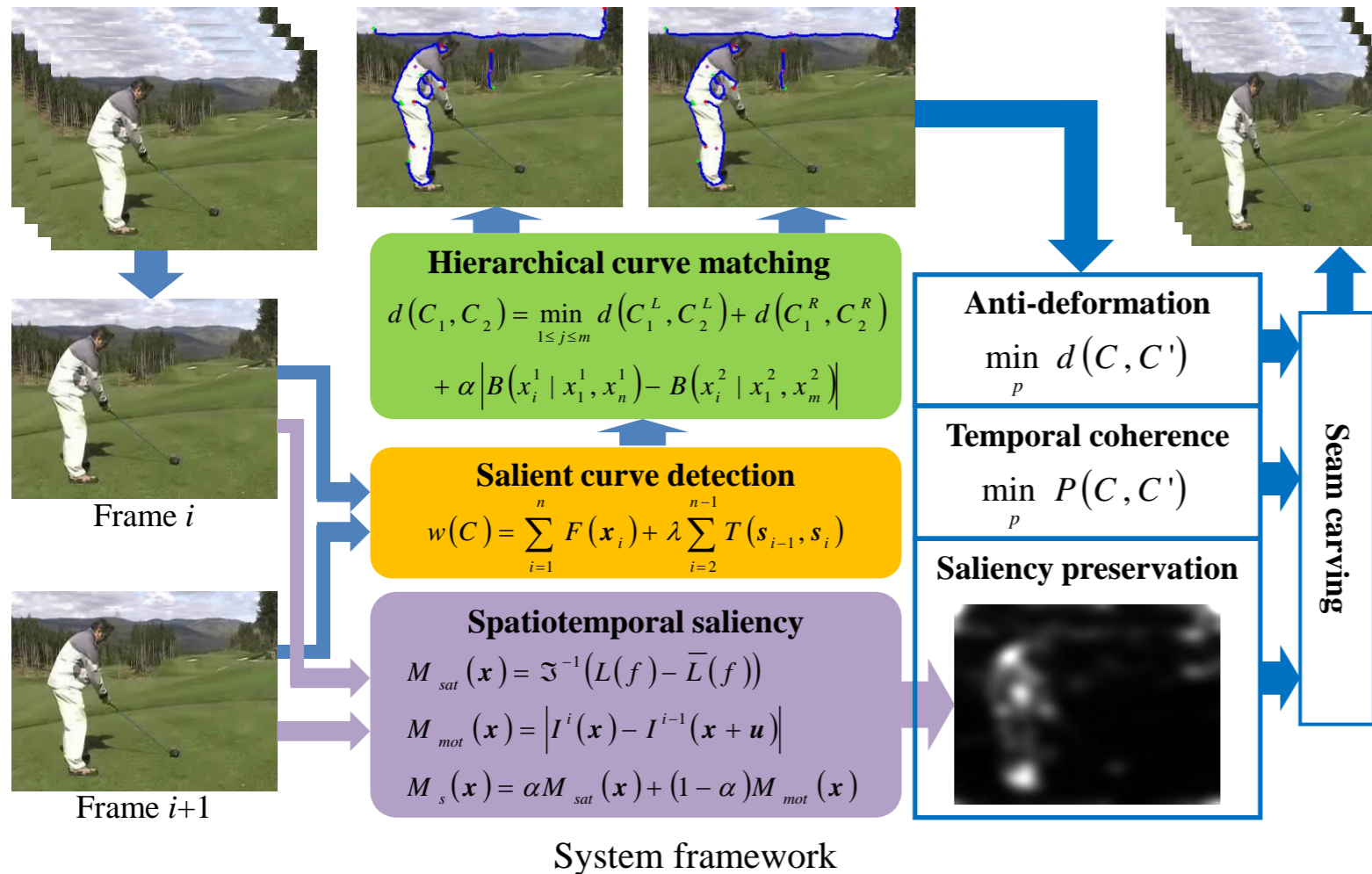


Visual Preserving Video Retargeting with Deformable Shape Consistency

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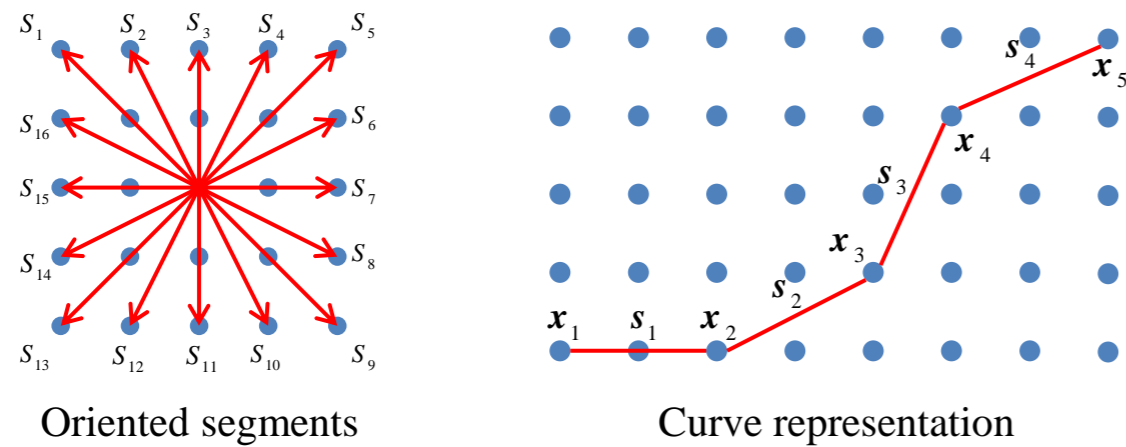
1. Introduction

Video retargeting is a technique that resizes the aspect ratio and resolution of a video in a content aware manner. In this paper, we propose a novel shape preserving video retargeting method to reduce deformations and maintain temporal consistency via matching salient curves in the video frames. Our method transforms the anti-deformation and temporal consistency problems in video retargeting into a curve matching cost minimization problem. By incorporating a deformation cost and temporal inconsistency cost into the seam carving framework, the quality of the retargeted videos can be significantly improved.



2. Salient curve detection

A curve is represented as a sequence of oriented segments $c = (x_1, x_2, \dots, x_n)$, where $s_i = x_{i+1} - x_i \in S$



The weight of a curve is defined as

$$W(c) = \sum_{i=1}^n F(x_i) + \lambda \sum_{i=2}^{n-1} T(s_{i-1}, s_i)$$

where $F(x)$ is the saliency map, and $T(u, v)$ is a smoothness term that encourages consistent orientations of consecutive fragments u and v . Salient curves can be extracted by detecting curves with large weights and sufficient lengths.

3. Hierarchical shape matching

A curve is represented by a binary shape tree. Each node in the shape tree stores the relative location of the midpoint with respect to the two endpoints of a sub-curve. The left child of the node stores the first half sub-curve and the right child stores the second half sub-curve in the same manner.

Given two endpoints $x_1 = (x_1, y_1)$ and $x_n = (x_n, y_n)$ of a curve, the relative location of the midpoint $x_i = (x_i, y_i)$ is represented in the Bookstein coordinate

$$x'_i = \frac{(x_n - x_1)(x_i - x_1) + (y_n - y_1)(y_i - y_1)}{(x_n - x_1)^2 + (y_n - y_1)^2} - \frac{1}{2}, y'_i = \frac{(x_n - x_1)(y_i - y_1) + (y_n - y_1)(x_i - x_1)}{(x_n - x_1)^2 + (y_n - y_1)^2}$$

Matching two curves in the Bookstein coordinate is invariant to translation, rotation and scaling.

To match curve $c_1 = (x_k^1)_{k=1}^n$ and $c_2 = (x_k^2)_{k=1}^m$, we fix the shape tree of c_1 , and look for a mapping from the points in c_2 to the points in c_1 , which minimizes the deformation cost

$$d(c_1, c_2) = \min_{1 \leq j \leq m} \{d(c_1^L, c_2^L) + d(c_1^R, c_2^R) + \alpha |B(x_i^1 | x_1^1, x_n^1) - B(x_j^2 | x_1^2, x_m^2)|\}$$

where x_i^1 is the midpoint of c_1 , x_j^2 is the midpoint of c_2 , $d(c_1^L, c_2^L)$ is the matching cost of the left subtrees of c_1 and c_2 , and $d(c_1^R, c_2^R)$ is the matching cost of the right subtrees of c_1 and c_2 .

3. Shape preserving seam carving

The energy of a seam is

$$E(p) = E_s(p) + \beta E_d(p) + \gamma E_t(p)$$

where E_s is the visual saliency term, E_d is the deformation term and E_t is the temporal term.

Visual saliency term E_s is the sum of the saliency intensities of seam p in the saliency map M_s :

$$E_s(p) = \sum_{j=1}^H M_s(p_j, j)$$

The deformation term E_d is the sum of deformation costs of corresponding curves after removing seam p :

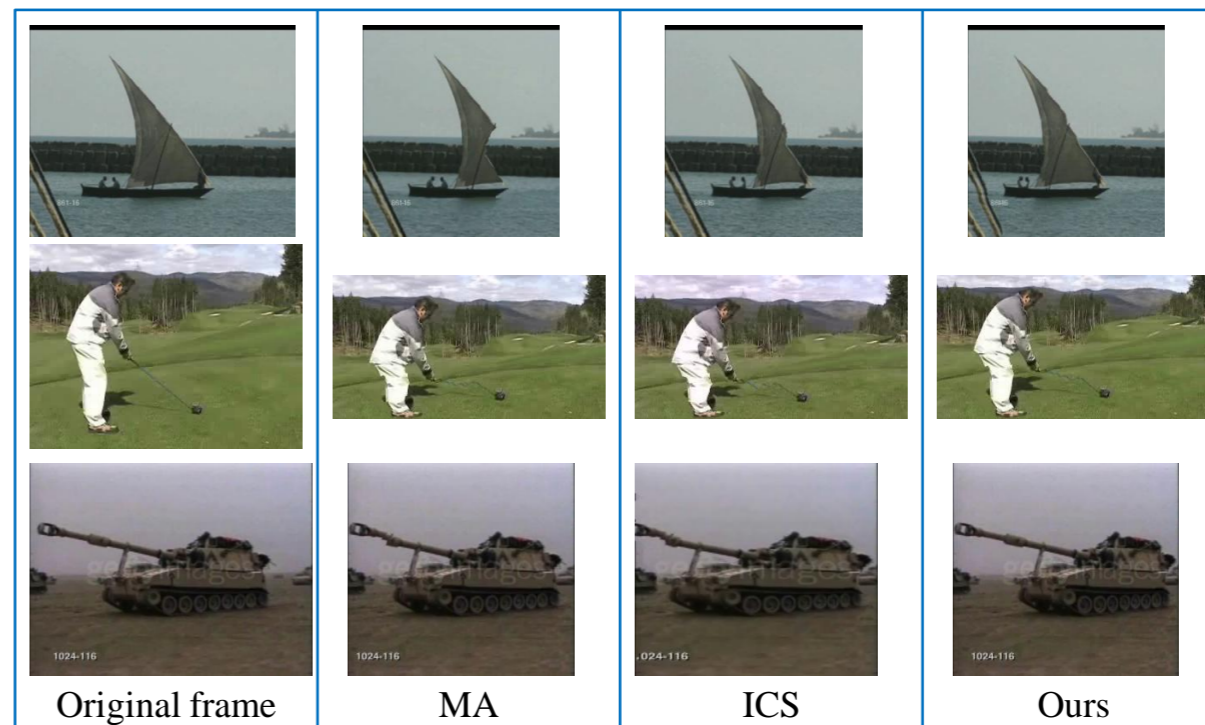
$$E_d(p) = \sum_{i=1}^N d(c_i, c'_i)$$

where c'_i is the deformed curve of c_i . The temporal term E_t penalizes the variation of relative positions of curves.

The optimal seam is obtained via dynamic programming.

$$p^* = \operatorname{argmin}_p E_s(p) + \beta E_d(p) + \gamma E_t(p)$$

4. Experimental results



Psycho-visual evaluation result

	Ours	ICS	MA
Ours	-	17	15
ICS	59	-	31
MA	61	45	-
Total	120	62	46
Prefer	78.9%	40.8%	30.3%

ICS: M. Rubinstein et al., "Improved seam carving for video retargeting," SIGGRAPH, 2008.
MA: Y. Bo et al., "Matching area based seam carving for video retargeting," TCSVT, 2008.