



## Efficient Image morphing algorithm using Laplacian Pyramids

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**Abstract:** Image morphing has been the difficulty of a whole lot interest in latest years. It has tested to be a powerful visible consequences device in film and television, depicting the transformation of one virtual photo into any other. While this method is used successfully, the photo may be transformed into whatever a person needs with dramatic results. Morphing can be described as an animated transformation of one photo into any other photo. Morphing entails photo processing techniques like warping and move dissolving, pass dissolving manner that one photo fades to every other photograph the usage of linear interpolation. This technique is visually negative because the functions of both pictures are not aligned, and as a way to result in double exposure in misaligned areas. So as to overcome this hassle, warping is used to align the two photos earlier than cross dissolving. Warping determines the way pixels from one photo are correlated with corresponding pixels from the alternative picture. It's far needed to map the vital pixels, else warping doesn't paintings. Shifting other pixels is obtained by means of extrapolating the data distinctive for the manipulate pixels. In this project we implemented Laplacian Pyramid Transformation primarily based techniques of image morphing and as compared them. We tried to automate the procedure as much as possible.

**Keywords:** image, morphing, laplacian, pyramids

### I. INTRODUCTION

Morphing is a unique effect in movement pics and animations that modifications (or morphs) one image or shape into another through a seamless transition. Most often it is used to depict one individual becoming some other through technological means or as part of a fable or surreal series. Traditionally the sort of depiction would be accomplished through pass-fading techniques on movie. Morphing algorithms preserve to develop and applications can robotically morph pictures that correspond carefully enough with exceedingly little education from the user. This has led to using morphing techniques to create convincing gradual-motion results where none existed in the authentic movie or video pictures by way of morphing among each individual frame using optical flow era. Morphing has additionally regarded as a transition technique between one scene and some other in TV suggests, although the contents of the two pics are entirely unrelated. The set of rules in this situation tries to find corresponding factors between the pics and warp one into the opposite as they crossfade at the same time as possibly less obvious as in the beyond morphing is used closely today. Whereas the effect changed into to start with a novelty, today, morphing results are most customarily designed to be seamless and invisible to the attention. Selected use for morphing consequences is modern-day virtual font design. the use of morphing era, called interpolation or more than one grasp technology, a clothier can create an intermediate between patterns, as an example producing a semi-bold font through compromising among a bold and regular style, or make bigger a fashion to create an extremely-light or extremely-bold. The technique is commonly utilized by font design studios.

### II. MORPHING PRINCIPLE

The primary principle behind picture morphing is explained on this phase. "Morphing" refers back to the mixture of generalized photo warping with a move-dissolve among

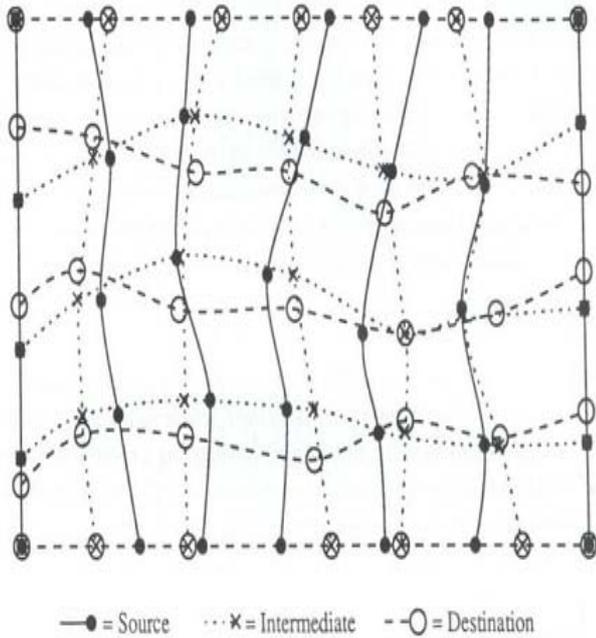
image factors. On the way to morph among pixels we outline corresponding control pixels in source photo I0 and vacation spot image I1. We then outline each intermediate frame I of the metamorphosis with the aid of creating a new set of control pixels via interpolating the manage pixels from their positions in I0 to the positions in I1. Each snap shots I0 and I1 are then warped closer to the placement of the control pixels in I. those warped photos are cross-dissolved during the metamorphosis. Consequently the distinct morphing techniques differ especially inside the way in which they carry out warping. Within the following section we can describe various image warping strategies.

### III. MORPHING TECHNIQUES

#### 1. Mesh warping

This is an algorithm shaped in steps that accepts a source image and two 2d arrays of coordinates S and D. The S coordinates represent the control pixels in the supply photograph, and the D coordinates are the locations where the S coordinates will match. The very last image is the preliminary photo warped by means of mesh S and mesh D. The 2d arrays wherein the manipulate points are stored, impose a rectangular topology to the mesh. The only constraint is that the meshes defined through each arrays be topologically. Consequently the D coordinates are coordinates which can pass as a long way from S as important, so long as they do no longer intersect with themselves. Step one manner re-sampling each row independently. An intermediate array of points I, whose x coordinates are same as those in D and whose y coordinates are similar to those in S, is created. Vertical splines are generated to match every column of information in S and in I. The facts for every place in a row is interpolated to create intermediate picture I. The second one step consists in re-sampling every column independently. Horizontal splines are then generated to match each row of information in arrays I and D. The statistics for each place in a column is

interpolated from intermediate picture I to create destination image D.



The collection of vertical splines geared up through Sand I in the first step and with the horizontal splinesfitted throughI and D within the 2d step, are shown in the figure 1.

**2. Feature based photograph Warping**

This is a technique that gives an excessive stage of manage over the method. The corresponding function lines in the photographs that are being morphed, are interactivelydecided on. The set of rules makes use of strains to relate features inside the source image to capabilities inside the final photo. This algorithm is based totally upon fields of affect surrounding the function traces selected. It makes use of opposite mappingfor warping the image. A pair of strains (one defined relative to the source photograph, the other defined relative to the destination picture) defines a mapping from one photograph to the other (see fig. 2).

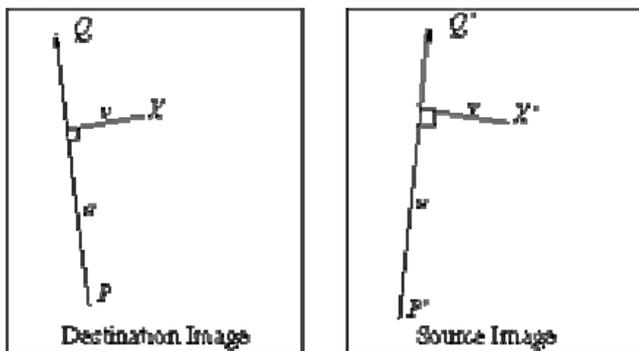


Fig. 2 Mapping from one image to the other.

**3. Thin Plate Spline based totally picture Warping**

Thin-plate Spline is a conventional tool for floor interpolation over scattered records. It is an interpolation technique that unearths a "minimally bended" easy floor that passes through all given factors. The call "thin Plate" comes from the truth that a TPS extra or less simulates how a skinny metallic plate could behave if it became compelled

via the identical manipulate points. Let us denote the target function values  $v_i$  at places  $(x_i, y_i)$  inside the plane, with  $i=1,2,...,p$ , in which  $p$  is the number of function points. Specially, we will set  $v_i$  identical to the coordinates  $(x_i', y_i')$  in flip to gain one non-stop transformation for each coordinate. An assumption is made that the places  $(x_i, y_i)$  are all distinct and are not collinear.

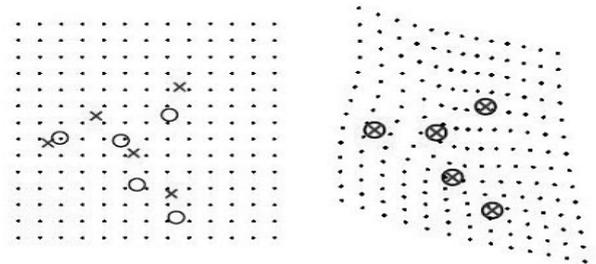


Fig. 3: Example of coordinate transformation using TPS.

The fig3 determines 3 an easy example of coordinate transformation the use of TPS. It begins from units of factors for which it's far assumed that the correspondences are known (a). The TPS warping permits an alignment of the points and the bending of the grid suggests the deformation needed to deliver the 2 units on pinnacle of every different (b). Inside the case of TPS implemented to coordinate transformation we truly use two splines, one forthe displacement within the x coordinate and one for the displacement in the y direction. The 2 resulting alterations are blended right into a single mapping.

**IV. IMAGE MORPHINGALGORITHM**

A top level view of our morphing set of rules upon conformal welding is provided in Fig. 1, wherein we use the form con- tour with unmarried factor to demonstrate the set of rules go with the flow. Given contours (may with a couple of-connected additives), we aim to deform the enter one into the goal even as preserving the shape consistency. We generate the signature for the shape contour of the input cup, as proven in the crimson curve in Fig. 1. in addition, the signature of the goal form (the green curve in Fig. 1) can be also obtained through conformal mapping. The signatures are actual-valued periodic features of to itself, and represent the form information as given by Theorem 1. We're able to deform the input signature to the goal one in the circle domain by using the use of the Laplace constraint, yielding the signature in blue of Fig. 1. Eventually, we reconstruct the desired form from the signature via conformal mapping. Theoretical analysis inside the preceding segment suggests that the proposed set of rules also works for shape contours with multiple-linked additives, although illustrated with the aid of pretty less difficult contours in Fig. 1. We gift a narrative of all the modules inside the manner beneath.

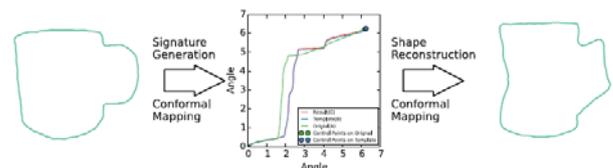


Fig. 1 Flowchart of our method

I. Preprocessingandconformalmapping

As a preprocessing step of our approach, we should extract the form contour of the object of interest from a given image. Contour extraction is this type of warm topic in pc vision that several algorithms are to be had. One may additionally resort to the classical active contour (Snake) set of rules for frequent images. For pictures of rather established objects (e.g., human faces), the present day regression-based algorithms are capable of seasoned- vide fast and correct facial landmarks [9, 22, 38]. These landmarks decide the form contours to which our morphing algorithm applies. We normalize the corresponding shapes beneath translations and scaling so that we are able to express the landmark coordinates within the identical reference body. Inside the relaxation of the paper, we find the warping among contour whilst retaining shape consistence in this reference body in any other case said.

We take the cup contour with complex shape configurations (more than one-nested components) as an instance to demonstrate the mappings that map all of the contours to a circle area  $\Omega$  inside the unit disk as shown in Fig. 2. The circle domain in Fig. 2 has 3 levels and 4 components  $\Omega_i (i=0, \dots, 3)$ . These segments  $\Gamma_i$  partition the whole domain into the planar domains  $\Omega_i (i=0, \dots, 3)$ . Each segment  $\Gamma_i$  can be denoted as the outer contour  $\Gamma_i^-$  and the inner  $\Gamma_i^+$  in different components except for the outermost one  $\Gamma_0$ . We triangulate all  $\Omega_i$  by the Delaunay triangulation given the landmarks in the contour components, and introduce the angle structure in triangular meshes in order to perform conformal mapping. In the mesh, we denote  $v_i$  as a vertex,  $[v_i, v_j]$  as an edge, and  $[v_i, v_j, v_k]$  as a face. The angle at vertex  $v_i$  in the triangle  $[v_i, v_j, v_k]$  is denoted as  $\theta_{ik}^i$ . The corner angle of the mesh is defined as the set

$$A(\Omega) = \theta_{jk}^i, \theta_{ij}^k, \theta_{ki}^j \mid [v_i, v_j, v_k] \in \Omega$$

We calculate the conformal mapping that preserves corner angles in mesh structures as detailed in [19]. Consequently, we conformally map the planar domains,  $\Omega_0$  and  $\Omega_1$ , to the annuli  $D_0$  with a concentric hole and  $D_1$  with multiple holes, respectively. The domains  $\Omega_2$  and  $\Omega_3$  are mapped to unit disks  $D_2$  and  $D_3$ , respectively.  $\Gamma_i^-$  and  $\Gamma_i^+$  ( $i = 1, \dots, 3$ ) turn into  $\gamma_i^-$  and  $\gamma_i^+$  by conformal mappings, and  $\Gamma_0^-$  corresponds to  $\gamma_0$ . The results of these conformal mappings for multiple components are shown in Fig. 2.

II. Shape signatures

The important thing to our method is the derivation of shape signatures (given in Theorem 1) within the mapped area. The signatures include conformal modules and diffeomorphisms.

The conformal module of a site is determined with the aid of the configuration of the circle domain, i.e., the middle and radius of the inner circle. For the cup having four additives in Fig. 2, the middle and radius of each hollow of  $D_i$  gives the conformal module of each component. Those facilities and radii of the holes at the plane are conformal invariants so that we can adjust picture shapes by using changing the corresponding conformal modules.

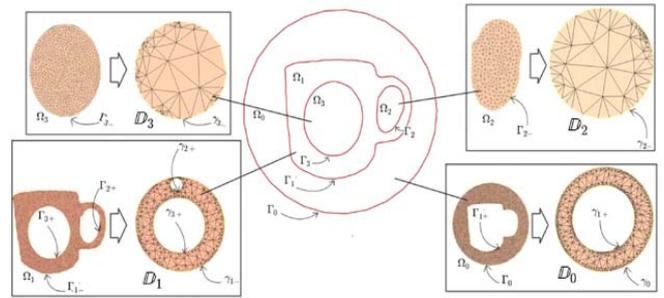


Fig. 2 Translate and scale shape components, and then apply a Riemann mapping for each domain given by shape components

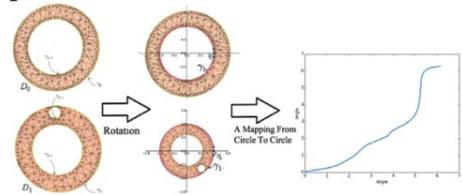


Fig. 3 Construction of shape signatures

We use the outer contour of the cup  $\Gamma_1$  as an instance to demonstrate the computation of diffeomorphisms. An identical computational procedure, detailed in Alg. 1, applies to every contour additives. Figure 2 shows that  $\Gamma_1$  is mapped to  $\gamma_1^-$  and  $\gamma_1^+$ . We firstly build the correspondence between points on  $\gamma_1^+$  and  $\gamma_1^-$ , finding the corresponding pairs  $(p_r, p_s)$ , where  $p_r$  and  $p_s$  are the points on  $\gamma_1^+$  and  $\gamma_1^-$  shown in Fig. three, respectively. Sooner or later, we select a corresponding factor pair  $p_r$  and  $p_s$  on  $\gamma_1^+$  and  $\gamma_1^-$  as the origin points so that any point on  $\gamma_1^+$  and  $\gamma_1^-$  flip to perspective representations on circles. Sooner or later, we construct the desired diffeomorphism among  $\gamma_1^+$  and  $\gamma_1^-$ , a periodic real-valued function from  $\theta$  as illustrated in Fig.3.

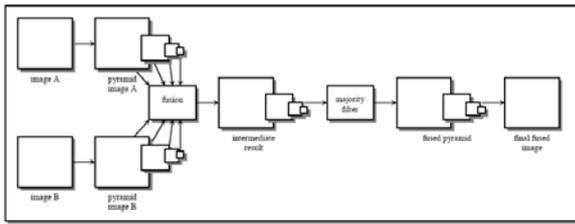
The conformal module together with diffeomorphisms form the signature  $s_i (i = 1, \dots, \text{three})$ , which generates a easy and intuitive shape illustration (centers and radii) while stopping undesired mesh systems (e.g., folding and twisting) for deformations. Hence, we're able to really exchange form contours within the parametric area via adjusting the signatures. Additionally, we can apply shape constraint to the deformation at the signature as mentioned underneath.

V. LAPLACIAN PYRAMID

Photograph pyramids were defined for a multiresolution image evaluation as a model for the binocular fusion for human imaginative and prescient. A photo pyramid can be defined as collection of low or band pass copies of an authentic photograph wherein both the band restriction and sample density are reduced in ordinary steps.

The Laplacian Pyramid implements a "pattern selective" method to picture fusion, so that the composite picture is built now not a pixel at a time. The simple concept is to carry out a pyramid decomposition on every source picture, then integrate a lot of these decompositions to form a composite illustration, and ultimately reconstruct the morphed picture by way of performing an inverse pyramid transform.

Schematic diagram of the Laplacian Pyramid fusion technique is shown in parent 3.



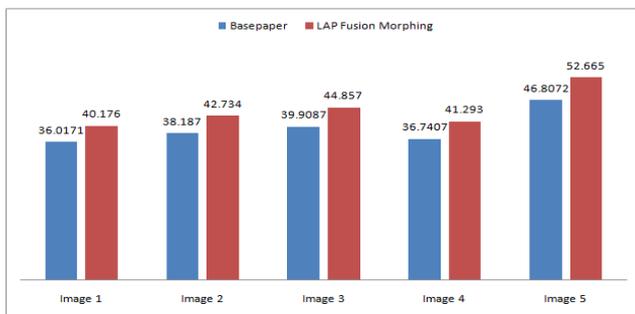
**Figure 3: Schematic diagram of the Laplacian Pyramid fusion method**

Laplacian Pyramid used several modes of combination, including choice or averaging. Inside the first one, the aggregate approaches selects the issue sample from the supply and copy it to the composite pyramid, while discarding the less pattern. Within the 2nd one, the technique averages the resources styles. This averaging reduces noise and gives balance in which supply photos comprise the identical sample information.

**VI.RESULT AND ANALYSIS**

**Table 1 Comparison of Peak Signal To Noise ratio for various images using basepaper and Laplacian fusion pyramids**

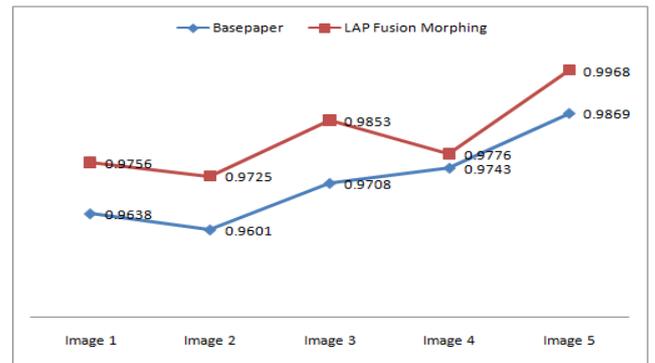
Input Images	Basepaper	LAP Morphing
Image 1	36.0171	40.176
Image 2	38.187	42.734
Image 3	39.9087	44.857
Image 4	36.7407	41.293
Image 5	46.8072	52.665



**fig 1 Comparison of Peak Signal To Noise ratio for various images using basepaper and Laplacian fusion pyramids**

**Table 2 Comparison of SSIM for various images using basepaper and Laplacian fusion pyramids**

Input Images	Basepaper	LAP Morphing
Image 1	0.9638	0.9756
Image 2	0.9601	0.9725
Image 3	0.9708	0.9853
Image 4	0.9743	0.9776
Image 5	0.9869	0.9968



**Fig 5.2 Comparison of SSIM for various images using basepaper and Laplacian fusion pyramids**

**Table 5.3 Comparison of Time taken (ms) for various images using base paper and Laplacian pyramids.**

Input Images	Base paper	LAP Morphing
Image 1	14.2479	12.82311
Image 2	13.7133	12.34197
Image 3	13.8123	12.43107
Image 4	14.319	12.8871
Image 5	22.6656	20.39904

**VII. CONCLUSION AND FUTURE WORK**

Photo morphing is the system of sluggish transition among images, and has obtained a great deal interest in digital multimedia fields over decades. Morphing between two or more photos over time is a useful visual approach that generates computer graphics for educational or enjoyment purposes. Moreover, morphing can be also implemented to other fields like scientific image processing and reputation. The morphing trouble includes problems, i.e., correspondence establishments (or warping) and transition path among pix (or blending). On this paper, we focus on the correspondence issue among shapes in photos, which typically needs guide intervention. This difficulty is also essential to photograph mixing as latest tactics typically treat a photograph as a fixed of embedded curves as opposed to a grid of pixels. Specially, we address complex shapes a couple of-nested additives (e.g., human faces) displaying remarkable demanding situations to hold form consistency in morphing

As future work, further to handling extra complex oclusions, we would like to explore the complete space of color and movement paths, perhaps even allowing express consumer manage with a stroke-primarily based interface. We'd also want to bear in mind extending this technique to morphing between video sequences.

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