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# Effective Video Inpainting Based on Super-Resolution Algorithm

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**ABSTRACT**: Image Inpainting is recover missing part of image as Image is save memories of life's important moments. A Novel framework is one of the model of inpainting in which image inpainting work on scratchy version of image inpainting. Inpainting of the low resolution image is simple than high quality image. It will display high and complex image. Using different image inpainting techniques create high quality of image from low resolution image and collect the high level images. For this purpose our system uses the super resolution algorithm which is responsible for inpainting of single image.

**KEYWORDS:** Examplar based inpainting, single image super Resolution.

### I. INTRODUCTION

Nowadays, Image in painting gives some methods that contains in missing and in filling- regions (holes) in an image. Previous method can be describe in main two categories. The first type is diffusion based which propagated linear structure or level lines via diffusion based on partially different equation and various method. Unfortunately, the diffusion based method tend to introduce some blur when the hole to be filled in is large. The approach is using appropriate model which copy best matching texture patches from clear match patches. These methods have been inspired from techniques like texture synthesis and are known to work well in cases of regular or repeatable textures. The first attempt to use exemplar based techniques for object removal. Diffusion and exemplar based methods combined effectively by these structure use to given priority to patches for filled. In previous techniques having some difficulties exist when have to large fill and critical term is the high computational time in general required. Hierarchical approaches considerable lower resolution of input image and inpainting using K-NN (K Nearest Neighbors) exemplar based method. Correspondences between K-NN low- resolution and from input image high resolution patches first learn and store in dictionary. Image Inpainting create modification in image which will not observed. In these studies we proposed the novel technique for clear the image and for digital video inpainting. This technology work as hotspot in computer graphic. Many application like some object remove from the digital photo, renovate the old movie, super resolution, red eye alteration, image coding work in image inpainting. Also remove and clear unwanted part in form image based on the background information in visually possible way. The purpose of image inpainting not recover original image but make close to the original image. One of the unwanted object remove from images is an image manipulation technique. The purpose is main image differ from remove unwanted object to improve quality of image.to make high quality image remove undesirable object hence start the mask out unwanted object, making area where this object previous occupy the gap. Then the gap will be filled by using graphical techniques. Among the graphical techniques which are used to fill the gap after object removal, two most commonly used are as image inpainting and texture synthesis.



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(a) Before Inpainting

(b) After Inpainting

Fig.1.Image Inpainting

### **II.** LITERATURE REVIEW

### A. Overview Of The Algorithm

There are many techniques exists which can be used for inpainting of the image. These techniques can be the diffusion based or the examplar based techniques. To develop the hierarchical approach of high level image inpainting having some limitations in above approches.

### B. Traditional Image Inpainting

Image inpainting is only work for the make high quality image and clear the image. But this approach is not suitable for high quality images. It uses patch based inpainting. Here this area is marked as the sigma notation. The sigma means masking done on the image and it is removed by using Efros and leungs algorithm. In the selected area apply inpainting algorithms by manualy.

#### C. Examplar-Based Inpainting

In [4] author presented a technique which introduce a novel examplar based Image Inpainting Algorithm with an improved priority term which defines the filling order of patches in the image. This algorithm is based on propagation of patch by propagating the image patches from the source region into the interior of the target region patch by patch. The block effects are removed by the PDE. Examplar-based technique not used for complex geometrics structure completion because the novel method used for restore original image with large target region and complex geometrics structure. There are two types of exemplar-based method is as follows:

1) The Filling Order and

### 2) Texture Synthesis

1) Filling Order by Patch Priority: To define the structures from textures the filling order computation defines a measure of priority for each patch in order.

2) Texture Synthesis: The filling process starts with the patch having the highest priority. The similarity metric is used for this purpose.

### D. Image Regularization Using PDE's

In [2] author gave an overview of a method uses vector valued algorithm for elaborate the diffusion. It is mainly based on the following approaches.

- 1. Functional minimization.
- 2. Divergence expression.
- 3. Oriented laplaciouns [2].

### E. Fragment Based Inpainting

In [5] author described the principles of simplicity and familiarity of image inpainting techniques. Hence, apply simple and smoother process in undesirable object area. These process can generate iterative simple reconstruction to help for the completion process which is based on the given image.



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#### **III. IMPLEMENTATION DETAILS**

The System architecture of project is shown in Fig 2. It works in following steps:

- 1. From the original image first extract low resolution area;
- 2. An in-painting algorithm is applied to fill-in the holes of the low-resolution picture;
- 3. The quality of the in-painted regions is improved by using a single-image SR method.



Fig. 2.System architecture

#### A. Image Inpainting

In painting is the process of reconstructing lost or deteriorated parts of images and videos. For example, in the museum world, in case of a valuable painting, this work would be carried out by a skilled art conservator or art restorer. In digital world, in painting refers to the application of sophisticated algorithms to replace lost or corrupted parts of the image data.

### B. Image Restoration

Image restoration is the operation of taking a corrupted/noisy image and estimating the clean original image. Corruption can come in many forms such as <u>motion blur</u>, <u>noise</u>, camera miss focus' etc.

#### C. Super Resolution

Super resolution (SR) is a class of techniques that enhance the <u>resolution</u> of an <u>imaging</u> system. In some SR techniques—termed as <u>optical SR</u>—the <u>diffraction limit</u> of systems is transcended, while in others termed as geometrical SR—the resolution of digital <u>imaging sensors</u> is enhanced.

#### 1) Super Resolution Algorithm:

Once the in painting of the low-resolution picture is completed, a single image super resolution approach is used to reconstruct the high resolution of the image. The idea is to use low resolution in painted areas in order to guide the texture synthesis at the higher resolution. The problem is to find higher resolution patch from a database of examples.



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Dictionnary

Fig.3. Flowchart of super resolution algorithm

### 1. Dictionary building:

It consists of the correspondences between low and high resolution image patches. The unique constraint is that the high-resolution patches have to be valid which is entirely composed of known pixels. In the proposed approach, high-resolution and valid patches are evenly extracted from the known part of the image. The size of dictionary is a user-parameter. An array is used to store the spatial coordinates of HR patches (DHR). Those of LR patches are simply deduced by using the decimation factor.

### 2. Filling order of the HR picture:

The computation of the filling order is defines a measure of priority for each patch. It is computed on the HR picture with the sparsity-based method. This improves the quality of the in painted picture compared to a raster-scan filling order.

### 3. For the LR patch corresponding to the HR patch :

Which patch having the highest priority, according to its K-NN in the in painted images of lower resolution are sought. Only best candidate is kept.

### **IV. EXPECTED RESULT**

For comparing results we are tested on variety of natural images and compared to five state-of-art inpainting methods.

### a) Parameter Analysis:

Suppose two versions of the proposed method are evaluated. One uses down sampling factor of 4 in both directions (the patch size is equal to  $5 \times 5$ ) whereas this factor is set to 2 for the second version (the patch size is equal to  $7 \times 7$ ). For both versions, the size of the dictionary is the same and can contain at most 6000 patches evenly distributed over the picture. The LR patch size is  $3 \times 3$  and the HR patch size is  $15 \times 15$ .

### *b) Comparison with the state of the art method:*

Rather than using similar images, it will be possible to use state of art inpainted pictures and to combine them.by using this method we can improve the performance.We can easily choose the low resolution picture to redesigned.



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Picture	Resolution	Missing Areas	Inpainting and LBP	SR	Total
Elephant	$480 \times 320$	17%	32s	1m58s	2m30s
SnowBall	$480 \times 320$	13%	22s	1m36s	1m58s
Tiger	$480 \times 320$	28%	59s	2m36s	3m35s
Soldier	$320 \times 480$	30%	51s	2m39s	3m30s
Cow	$600 \times 400$	13%	40s	2m11s	2m51s

Table 1. Running time for low resolution in painting and super resolution

From results it is found that running time results with low resolution inpainting and super resolution gives better result. Also here we are using state-of-the-art method for filling missing areas so it gives better result.

## V. EXPERIMENTAL RESULT

	Registration Frame
Name:	rohit panchwagh
Address	Pune
User Name	rohit
Password	
Re-enter Password	****
Email Id	ohit.panchwagh@gmail.com
Mobile Number	9890430022

Fig.4. Registration

User should be completed Registration process before using this software, to get OTP password for secure login.

User	Authentication For In	age Inpainting	1	
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After successfully Registration User get OTP Password on Registered Email ID.



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Fig.6 Select Image

Further successfully login user have to selected Video for the On which Apply Inpainting method. In the left corner yellow mark has been reflected.



Fig.7.Apply Inpainting Method

Click on Run button to apply inpainting Method.



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### Fig.8. Inpainting completed successfully



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### VII. CONCLUSION

A novel in-painting method is used for recovering the damage images or filling missing regions. The low-resolution in-painted pictures are combined by globally minimizing an energy term. Once combination is completed, a hierarchical single image super resolution method is applied to recover details at the native resolution.



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