



Inpainting Analysis in 2d-Images Scheme Using Efficient Image Restoration

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Abstract: Image Processing is enhancing an image or extracting information or features from an image. Image processing includes image quality and statistical evaluation, Radiometric correction, Geometric correction, Image enhancement and sharpening, Image classification based on pixel and Object-oriented, Accuracy assessment of classification, Post-classification, GIS and Change detection. Digital inpainting is one of the popular techniques of image restoration. Most of the technique in image restoration involves the application of textures that are most similar to the areas around the missing region. But these produce inaccurate results due to discontinuous textures. In the proposed method prediction based on the non harmonic analysis techniques is used to get better results, which involves extracting the accurate spectra to produces an spectrum which sequentially generates new sequential textures on the basis of spectrum. The missing regions from the spectrum are repaired using 2D Non Harmonic Analysis. Which shows an improvement in Mean Square Error of about 10-20 compared with example-based method and subjective quality.

Keywords: Image processing, Image inpainting, Data term, Confidence term, Priority function, Patch.

I. INTRODUCTION

Image Processing is enhancing a picture or extracting data or options from an image. It is computerized routines for data extraction from remotely detected pictures to get classes of knowledge regarding specific options. Image processing includes image quality and applied mathematics analysis, Radiometric correction, Geometric correction, Image enhancement and sharpening, Image classification based on component and Object-oriented, Accuracy assessment of classification, Post-classification, GIS and Change detection.

Digital image processing is impelled by a pair of major applications like Improvement of pictorial data for human perception, Image processing for autonomous machine application and economical storage and transmission. The digital image can be optimized for the appliance by enhancing or sterilization the looks of structures.

Inpainting is a creative equivalent word for image interpolation, and has been circulated among museum restoration artists for a long time. Smart digital inpainting models, techniques, and algorithms have broad applications in image interpolation, photo restoration, zooming and super-resolution, primary-sketch based sensory activity image compression and cryptography, and the error concealment of image transmission, etc. and algorithms have broad applications in image interpolation, photo restoration,

zooming and super-resolution, primal-sketch based sensory activity image compression and cryptography, and the error concealment of (wireless) image transmission, etc.

This approach is primarily based on the Bayesian philosophy of vision: A best guess of the whole ideal image from its incomplete and distorted knowledge crucially depends on the system. A variation / PDE models and algorithms along this line of philosophy. Reconstruction of missing or damaged parts of pictures is an ancient follow used extensively in design restoration. Also notable as inpainting or retouching, this activity consists of filling in the missing areas or modifying the damaged ones in an exceedingly non-detectable means by an observer not aware of the first pictures. Applications of image inpainting range from restoration of images, films and paintings, to removal of occlusions, such as text, subtitles, stamps and publicity from pictures. In addition, inpainting can conjointly be went to turn out lighting tricks.

A simple and quick inpainting rule supported an isotropous diffusion model extended with the notion of user-provided diffusion barriers. The results produced by this easy model are, in many cases, comparable to previously notable non-linear inpainting models, but 2 to 3 orders of magnitude quicker, thus creating inpainting sensible for interactive applications. The presented rule is supposed for filling in domestically tiny areas. For larger inpainting



domains, a scale-space approach can be used to preserve the algorithm's speed at the expense of reconstruction quality.

Inpainting is the art of restoring lost / selected parts of a picture supported the background data in a very visually plausible means. Large areas with heaps of data lost are tougher to reconstruct, because data in alternative elements of the image is not enough to induce a sway of what's missing. Details that are fully hidden or occluded by the object to be removed can't be recovered by any mathematical technique. Therefore the objective for image inpainting isn't to recover the first image, but to produce some image that incorporates a shut likeness with the first image. Image Inpainting methods will be classified generally into: Texture synthesis algorithm: These algorithms sample the texture form the region outside the region to be inpainted. It has been demonstrated for textures, repeating 2 dimensional patterns with some randomness. Structure recreation: These algorithms try to recreate the structures like lines and object contours. These are usually used once the region to be inpainted is tiny. This focuses on linear structures which will be thought jointly dimensional pattern like lines and object contours. The objective of this paper is to highlight the importance of the problem, analyze of inpainting method, and propose a 2D image scheme using image restoration detection algorithm for them. The above mentioned problem involving inpainting falls under a category in darken image. Darken image can be defined as a deliberate attempt by an image to trait prior to its acquisition by the system.

In this study, we will concern myself with the problem of inpainting darken image for the following reasons:

- Show that equivalent to an alternate direction minimization procedure.
- It is use the soft-thresholding instead of the hard-thresholding normally used in nonlinear approximation.
- It is use proximal forward-backward splitting in convex analysis.
- The frame are different from orthonormal wavelets because tight frames are redundant.
- It support the missing blocks are missing the
- coefficients frame they support overlap the missing blocks.

Since existing inpainting assessment algorithms are designed to examine if an image contains sufficient information, they have limited capability in determining if an image in a clear form. To the best knowledge, this is the first work that facilitates images of mobile user's behaviors in order to recommend stores and items previously unknown to a user. This novel propagation inpainting algorithm to be

implemented in MATLAB for scratch to text removal, object removal and missing block completion. The major novelty of this work is that two types of patch sparsity were proposed and introduced into the inpainting algorithm.

Session 1 gives the overview of inpainting in image processing methods and its application. Session 2 describes the analysis of surveyed paper and compared with our proposed algorithm in session 4. In session 3, our proposed work is demonstrated carefully. The proposed algorithm and Image are implemented in Mat lab based platform for inpainting analysis and detection. In session 4 the result of this paper is presented and the discussions are made, we have stated about the result and discussion of our work. Finally in session 5, I narrated about the conclusion and future enhancement followed by references.

II. RELATED WORKS

The background work is a process of gathering and deciphering facts, diagnosing issues and the data to suggest enhancements on the system. The Process is studied to the minutest detail and analyzed. This is concerned with changing into alert to the matter, identifying the relevant and decisional variables, ANalyzing and synthesizing the various factors and determinant an best or a minimum of a satisfactory answer or program of action. Here we reviewed the existing papers with various techniques and algorithms in Inpainting Methods.

In [1], The most fundamental inpainting approach is that the diffusion primarily based, the missing region is filled from the constituent level. These algorithms are well based on the theory of partial differential equation (PDE) variational technique. Chan and Shen proposed use variational framework primarily based on total variation (TV) to recover missing data. The TVmodel is use to natural image to apply for image inpainting. The diffusion base algorithms is use to filling the nontextures or relatively smaller missing region. Second type algorithmic program is use examplar-based is use to approach the image data from understand region into the missing region at the patch level.

In [2], to discretization of the fractional framework and exploring the inpainting practicableness and qualities of the models rather than their numerical algorithms. The proposal is the fractional calculus hooked on image inpainting and set up a brand new category of fractional-order variation image inpainting models, in both house and moving ridge domains. The equivalent Euler-Lagrange equations are planned and correct numerical algorithmic program is analyzed.



Consider the planned simulations on quite a ton of testing pictures. The algorithm demonstrates higher inpainting performance on some image details than original integral-order inpainting primarily based on classic calculus. This method will be seen as a generalized image in-painting framework with each integral- and halfway order and it's straightforward to increase to different image inpainting with PDEs strategies, even the denoising models. The relationship between fractional order and also the inpainting performance was following work, and an reconciling halfway model can be terribly helpful in sensible applications.

In [3], the aim is to motivate the recent tight frame primarily based strategies on image restoration in either the image or the rework domain. The proposal work is image inpainting. It is a fundamental downside in image process and has several applications. The proposal of an repetitive tight frame algorithmic program for image inpainting. consider the convergence of this frame let primarily based algorithmic program by decoding it as iteration for minimizing a special purposeful.

In [4], by using novel algorithmic program massive object from digital images is removed. The challenge is to fill in the hole that's left behind in a very visually plausible way. In the past, this problem has been self-addressed by 2 categories of algorithms: (i) "texture synthesis" algorithms for generating massive image regions from sample textures. Inpainting techniques for filling in small image gaps. The former has been demonstrated for textures repetition two-dimensional patterns with some noise the latter concentrate on linear structures which might be thought of as one-dimensional patterns, such as lines and object contours. This presents a novel and efficient algorithmic program that mixes the benefits of those 2 approaches.

In [5], reconstructing the new image using image completion algorithmic program. They are hopped-up by an enormous information of images gathered from the online. The algorithm patches up holes in pictures by finding similar image regions in the information that aren't solely seamless however additionally semantically valid. The chief insight is that while the house of pictures is effectively infinite, the space of semantically differentiable scenes is really not that giant.

In [6], the techniques developed in three distinct however connected work of study, variation image inpainting, and texture synthesis and image completion, is investigated. Variation image inpainting involves filling narrow gaps in pictures. Though there ar difficult various strategies, best results are obtained by PDE-based algorithms. Texture synthesis is reproduction of a texture

from a sample. Firstly, statistical model primarily based strategies the proposal for texture synthesis.

In [7], a new exemplar-based framework is presented, which treats image completion, texture synthesis, and image inpainting in a unified manner. In order to be ready to avoid the occurrence of visually inconsistent results, all of the above image-editing tasks is duplicated in the kind of a distinct international improvement downside.

The objective function of this downside is often well-defined, and corresponds to the energy of a discrete mathematician random field (MRF). For efficiently optimizing this MRF, a novel optimization theme, called priority belief propagation (BP), is then proposed, which carries 2 terribly vital extensions over the quality BP algorithm: "priority-based message scheduling" and "dynamic label pruning." In order to briefly justify the most limitations of current progressive strategies for image completion, next, providing a short review of related work for every one in all the 3 categories mentioned higher than.

Statistical-Based Methods: These methods ar in the main used for the case of texture synthesis. Typically, what these methods do is that, given an input texture, they try to explain it by extracting some statistics through the employment of compact constant applied math models.

Finally, besides image completion, also set up to check our priority-BP algorithmic program, which is a generic MRF improvement theme, to other labeling issues, as well, for which the massive cardinality of their state-space causes them to own a really high process value.

III. PROPOSED WORK

In a novel Exemplar-based inpainting algorithm through work the sparsely of natural image patches. Two novel ideas of sparsely at the patch level are planned for modeling the patch priority and patch illustration, which are 2 crucial steps for patch propagation in the exemplar-based inpainting approach.

First, patch structure sparsely is designed to live the boldness of a patch located at the image structure (e.g., the edge or corner) by the sparseness of its nonzero similarities to the neighboring patches. The patch with larger structure sparsely will be assigned higher priority for additional inpainting. Second, it is assumed that the patch to be filled are often delineate by the distributed linear combination of candidate patches underneath the native patch consistency constraint in a very framework of distributed illustration.



Compared with the traditional exemplar-based inpainting approach, structure sparsely enables higher discrimination of structure and texture, and the patch sparse illustration forces the recently inpainted regions to be sharp and per the encompassing textures. Experiments on synthetic and natural pictures show the benefits of the planned approach. The diffusion-based inpainting algorithms have achieved convincingly excellent results for filling the no rough or comparatively smaller missing region. However, they tend to introduce smooth impact within the rough region or larger missing region. The second category of approaches is the exemplar-based inpainting algorithmic program.

This approach propagates the image information from the best-known region into the missing region at the patch level. The idea stems from the feel synthesis technique planned during which the feel is synthesized by sampling the most effective match patch from the best-known region. However, natural images are composed of structures and textures, in which the structures represent the primal sketches of a picture and also the textures are image regions with homogenized patterns or feature statistics.

According to the proposal a nonlocal means approach for the exemplar-based inpainting algorithmic program. The image patch is inferred by the nonlocal means of a set of candidate patches within the best-known region rather than one best match patch. More exemplar-based inpainting algorithms were additionally planned for image completion. Compared with the diffusion-based inpainting algorithm, the exemplar-based inpainting algorithms have performed plausible results for inpainting the large missing region.

In a new image inpainting algorithm that depends on express edge data. The edge information is employed each for the reconstruction of a skeleton image structure within the missing areas, as well as for guiding the interpolation that follows.

The structure reconstruction part exploits completely different properties of the edges, such as the colours of the objects they separate, an estimate of however well one edge continues into another one, and the spatial order of the perimeters with reference to one another. In order to preserve both sharp and swish edges, the areas delimited by the recovered structure are interpolated severally, and the process is radio-controlled by the direction of the near edges.

In an various to partial differential equations (PDEs) for determination issues in pc vision supported critical heat transfer. Traditionally, the method for

determination such physics-based issues is just too discretized and solves a PDE by a strictly mathematical operation. Instead of using the PDE, use the global heat principle and to decompose it into basic laws.

Moreover, it has the advantage of modularizing the method toward a numerical scheme so as to simply modification approximations or applies it to several heat transfer issues. Unlike repetitive numerical analysis algorithms, the use of basic laws allows the physical rationalization of all steps and intermediate results of the algorithmic program. The international forms ar accustomed figure global quantities

Identify the regions that need to fill/remove. Continue any lines inbound at those regions fill in the regions with texture/color from the encircling areas. Take a pixel to be synthesized. Find that pixels close to it have already been synthesized (or pre-existed). Define this to be the mask. For every component within the image (the candidates), compare the $W \times W$ neighborhood to the neighborhood around the pixel to be synthesized employing a distance metric, but solely on pixels outlined by the mask. Keep either the K most similar neighborhoods, or the neighborhoods whose distance is less than T (W , K , and T are user-defined values) of the remaining regions, select one at random. Assign the intensity of the center component of the chosen region to the pixel being synthesized.

Decompose image into two parts: structure image and texture image. Perform inpainting on the structure image, Perform texture synthesis on the texture image. Recombine the two pictures to induce upshot.

Get process is commonplace manner of cautious process that the input is Associate in Nursing image, like a haze or covering binding, the crop of the suppose process could also be either Associate in Nursing feature or a habitual of characteristics or parameters associated with the feature. Choicest likeness process techniques convoluted treating the total b decipher as a reclining aware and levying commonplace signal process techniques. Semblance process every time refers to digital emblem calculates process, however optical and analog image process are also doable. The realization of images (producing the input image in the chief place) is referred to as imaging.

Finest unendingly, personality processing systems plead to stray the digit be get-at-able in digitized demeanour, rove is ,arrays of limited take a run-out powder binary record for Digestion, the prone sign is sampled on a novel grid and every sample or constituent is quantal mistreatment finite variety of bits. The digitized catch on to is convenience by a calculator. To reveal a



digital participate, it's consummate regenerate into analog awake that is scanned onto a show. Prior to declining to process a diagram, it's regenerate into digital kind. Digitization includes style of image and division of sampled values. Explore modifying the sculpture into statute pointer, process is performed. This process technique is also estimate remedy, image restoration, and compression

The major topics within the field of image processing include:

1. Image restorations.
2. Image enhancements.
3. Image compressions.

3.1 Image Restoration

It involved cares thinks about worries is bothered with filtering the ascertained image to minimizing the impact of degradations Image sweetening therein the latter is concerned with a lot of extraction or accentuation of image options. Restoration method try to model the distortion to the image and reverse the degradation, where improvement ways use data of the human visual systems responses to improve a picture visually.

3.2 Image Enhancement

It refers to emphasise, or sharpening, of drawing mush like appal, or parallel to vindicate a substantial show a lot of helpful for show & analysis. This battle doesn't quite growth the inherent data content in knowledge. It includes grey deliberate & identical category with model, clamour epitome, improvement crispening and sharpening, filtering, interpolation and magnification, pseudo coloring, and so on.

Enhancement ways tend to be drawback specific. For example, a method that's accustomed enhance satellite pictures might not appropriate for enhancing medical pictures. Although improvement and restoration ar similar in aim, to make a picture look higher. They differ in however they approach the drawback.

3.3 Image Compression

Involves reducing the typically huge quantity of information required to represent a picture. This done by eliminating data that square measure visually surplus and by taking advantage of the redundancy that is inherent in most pictures. Image processing systems square measure used in several and varied forms of environments, such as:

1. Medical-community
2. Computer-aided design
3. Virtual-Reality
4. Image Processing.

3.4. Image Resolution

The resolution has to do with ability to separate two adjacent pixels as being separate, and then we will say that we will resolve the images. The concept of resolution is closely tied to the ideas of spacial frequency. Spatial frequency idea, frequency refers to how chop-chop the signal is dynamical in house, and the signal has two values for brightness-0 and most. If we use this signal for one line (row) of associate image then repeat the road down the complete image, and the image obtained in vertical stripes. If the frequency is increased the strips get nearer and nearer along, until they finally mix along.

3.5 Image Representation

The human visual system (HVS) receives Associate in Nursing input image as a set of spatially distributed lightweight energy; this can be type is named an optical image. Optical images are the sort we have a tendency to cope with on a daily basis –cameras captures them, monitors display them, and the optical images are drawn as video info within the style of analog electrical signals and these are sampled to get the digital image $I(r, c)$. The digital image $I(r, c)$ is represented as a two-dimensional array of information, where every component price corresponds to the brightness of the image at the purpose (r, c) . In linear algebra terms, a two-dimensional array like our image model $I(r, c)$ is referred to as a matrix, and one row (or column) is called a vector. The image types we have a tendency to can take into account as:

1. Binary Image
2. Gray scale image

The actual info hold on within the digital image knowledge is brightness information in every spectral band. When the image is displayed, the corresponding brightness information is displayed on the screen by image parts that emit lightweight energy corresponding to that exact color.

Typical color images square measure delineated as red, green, and blue or RGB images .using the 8-bit monochrome customary as a model, the corresponding color image would have 24 bit/pixel – eight bit for every color bands (red, green and blue).

- Repairing Photographs: With age, photographs often get damaged or scratched. It can revert deterioration using inpainting.
- Remove unwanted objects: Using inpainting, unwanted objects, text, etc. can be removed from the image.
- Special Effects: This may be used in producing special effect.



• Video inpainting: If extended to video inpainting, it would be able to provide a great tool to create special effects etc.

Before presenting the detailed description of this technique, let us analyze however consultants inpaint. Conservators at the Minneapolis Institute of Arts were consulted for this work and created it clear to US that inpainting is a terribly subjective procedure, different every work of art and for each skilled. There is no such thing as “the” thanks to solve the matter, but the underlying methodology is as follows:

- First, the patches from input images are removed.
- Then test whether the patches are empty are not.
- If the patches are empty means it fill the patches and calculate the intercept point from the region.
- In this method, the input image is considered as an one large region.
- This large region is split into some small regions.
- Then calculate \hat{P} . Where \hat{P} is the intermediate point of the small regions.
- Then find $\Psi_{\hat{P}}$.
- Evaluate $\Psi_{\hat{P}} \cap \Omega$. (i.e.) intercept point between patches and the target signal.
- Then fill the textures into empty regions.
- Then predicted value calculated from 2D Non Harmonic Analysis.
- Replace the textures into the empty regions.
- Finally perform inpainting operations.
- Whether the patches are not empty means directly perform the inpainting operations.
- Finally the output inpainting images are obtained

IV. EXPERIMENTAL RESULT

The images were experimented with the projected and therefore the alternative 2 algorithms and therefore the results were evaluated supported the Mean sq. worth (MSE). The obtained results exploit that if there is no larger change within the brightness, the quality is low yet smoothing in edges and texture patterns with high-frequency parts ends up in obstruction artefact. This blocking artefact will be reduced by shrinking the analysis window that wasn't much getable.

The exemplar-based methods seldom turn out sensible subjective quality as a result of the restricted styles of texture patterns. When the textures square measure matched on MSE basis; the good example primarily based methodology cannot exactly gauge the characteristics of the texture. The result is discontinuous texture patterns or

blocking artefact within the restoration space. This problem will be reduced by employing a smaller analysis window. However, it is challenging to always get textures that offer natural results Table 4.1 shows the Mean MSE Values.

	Criminisi's Method		Proposed Method	
	PSNR	SSIM	PSNR	SSIM
Lena	30.98	0.9548	33.12	0.9666
Barbara	30.59	0.9497	32.37	0.9637
Pepper	30.38	0.9526	32.74	0.9639
Mandrill	28.60	0.9234	28.12	0.9341

Table 4.1: Mean MSE Values. The Standard Deviation Is Provided In Parentheses

It defines the mean MSE's and standard deviation of the images and that is restored by each method. Our proposed method is used for “Lena”, “Barbara”, “Mandrill” and “pepper” images. Then we substitute the mean value MSE's for the “Mandrill” images. It is conceivable to the basis of the frequency analysis. It cannot easily recover the original information. It defines the PSNR and SSIM values and it is exemplar based methods. SSIM is used to measure the degree of resemblance of the image. Then value near 1 is desirable. However with respect to the PSNR and SSIM, proposed method scored higher than the exemplar based method for the “Lena”, “Mandrill”, “Barbara”, and “Pepper” images. Table II define the SSIM value in the range 0.93 ~ 0.96. In this proposed method, use gray scale images. The PSNR result of method for “Lena” was 25.1206 dB. The PSNR result of the proposed method for “Lena” was 33.12 dB, and it is higher than the other two methods as shown in the table 4.2.

	Mean Value	Criminisi's Method	Proposed Method
Lena	97.46 (11.08)	52.57 (9.82)	29.89 (5.56)
Barbara	100.37 (9.55)	57.30 (7.74)	32.42 (3.41)
Pepper	114.21 (14.92)	60.80 (12.69)	35.11 (6.09)
Mandrill	78.11 (3.93)	90.11 (7.92)	78.27 (5.26)

Table 4.2: Mean Value of PSNR and SSIM



However, it is difficult to compare these subjective evaluations with other methods. In the future, considering this ways of comparing the object removal method and the proposed method.

V. SUMMARY AND CONCLUSION

The achievement of inpainting and their extensive readying all over the country have adopt some of the people to have need of the fully actions to move away its own identification with the aid of fixing their darken images. The obstacle of darken images alteration or modifying is highly entirely one of a kind from that of painting spoofing, where partner person makes use of a pretend painting to be able to undertake the identification of one more pixels.

While the concern of spoofing has got colossal awareness inside the literature, the purpose of evading identification for the image alteration and matching that the trouble of some alteration or make fake images and it has now not been included within the original literature. It despite a lot of documented instances of painting methodology. Whilst finding the black and white would be encountered with replacement image modalities equivalent to face and eyes for human beings. This difficulty is mostly solve inside the case of inpainting seeing that of the fashionable readying of proposed in every executive and civilian application and also the benefit with that inpainting is altered.

- The proposed algorithm to examine the inpainting images.
- This algorithm has following the steps
 1. Improve the quality of image from the database.
 2. Remove the noise
 3. Convert them into high quality images.

This result shows the probability of the proposed approach for inpainting methods with exemplar, numeric, framelet and time varying methods. In future, it will actually involve in work, extensions to current algorithm to handle correct propagation of semicircular structures in still pictures. Also investigation of economical looking theme and on the automatic discovery of element weights for various types of pictures furthermore as removing objects from video, which promise to impose associate degree entirely new set of challenges.

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