

PRIORITIZATION OF DARK CHANNELS IN IMAGE DENOISING AND DEHAZING ALGORITHMS

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Abstract

To dehaze and denoise simultaneously, we present a new fast variant technique in this paper. In the proposed method, the transmission map is first estimated using an adaptive approach based on the well-known darkish channel before the proposed method is applied. In the end, this transmission map reduces the threshold artefact and improves estimation precision. Intensity maps can then be used to build a new version model to search for the final haze- and noise-free image. Also noted is that the proposed variant model has a strong point in the form of a minimizer. Based on the Chambolle–Pock set of rules, a numerical process is laid out that ensures that the set of regulations is consistent. The results of large-scale experiments on real-world scenes show that our method can successfully restore incredible and contrasted images that are haze and noise free.

Keywords: Dehaze, Denoise, Adaptive, Chambolle–Pock algorithm

I. Introduction

Together with haziness, foggy and smokey degradation of the outdoor scene due to bad weather conditions. It's a headache for photographers because it alters the colours and reduces the contrast of everyday photos, it diminishes the visibility of the scenes and it's a risk to the reliability of many programmes like outside surveillance, item detection, it also decreases the clarity of the satellite tv for laptop pix and underwater snaps. This is why the removal of haze from images is an essential and widely sought-after area of image processing. There are a lot of these suspended particles around, and they cause light to be scattered before it reaches the camera, resulting in a pixelated picture. A thick layer of haze obscures the meditative light in the scene and blends it with the ambient light. This pondered light (i.e. scene colorings) from mixed light tends to be improved by haze removal techniques. The visual system's stability and power can also be boosted by using this effective haze removal of photograph. You can use polarisation independent issue analysis and the dark channel earlier to remove haze from a picture. It is common for the light from a situation to dissipate before it reaches the camera's lens, and the light gathered through any digital camera lens is usually mixed with the airlight. Because of this, there will be an increase in noise, a loss of contrast between light and dark, and a lack of colour constancy. When the weather is bad, such as when aerosols, haze, fog, rain, dirt, or fumes are present, this type of degradation can be particularly damaging. Fog, for example, may also produce an albedo effect, which results in ambiguity and noise, as a common climate phenomenon.

Certain aspects of these phenomena are detrimental to comprehension and the extraction of information from images. Real-world packages urgently need effective haze removal (or dehazing) and denoising methods.

There has been a lot of interest in image dehazing and denoise in imaging technology recently. One of the great advantages of these operations is that they are completely clean. These images are visually more

appealing, as well as more suitable for many important applications, such as image segmentation, feature extraction and photo fusion, because of the absence of haze and noise. Dehazing photos can be difficult because the haze relies so heavily on unknown intensity facts. If all that's provided as input is a single photograph, the situation may be made even worse. This interdisciplinary effort includes not only machine vision but meteorology, optics, and a few aspects of PC images. Visual diversity in the ecosystem is limited by haze and fog, both of which have a significant impact on the evaluation of target scenes.

A primary goal of image evaluation is to improve visibility, restore the hues, and all other photo parameters as if the picture had been taken or received in a more favourable environment.

Computer and human vision systems can benefit from the improved and refined images provided by picture dehazing for a wide range of purposes, as the middle benefit. Most computer vision software packages, from low-level image evaluation schemes to high-level item recognition, typically rely on the input photo as the final and most reliable source of the scene's brightness.

To put it another way, regardless of how high-level an algorithm's performance, it is strongly dependent on how well and reliably the input image is processed. As long as there is haze or fog in the target scene, these algorithms will be plagued by biased and corrupted input images.

For a long time now, researchers have been working on ways to improve photo dehazing. The modern brand new can be loosely institutionalized into the most important companies. As a first step, we have schemes that don't take into account the physical modelling of a photograph and how it's formed. Such schemes only aim to enhance the visual appeal of a photograph in order to please the viewer. Dehazing schemes based on such enhancements have spread across the board.

II Block Diagram

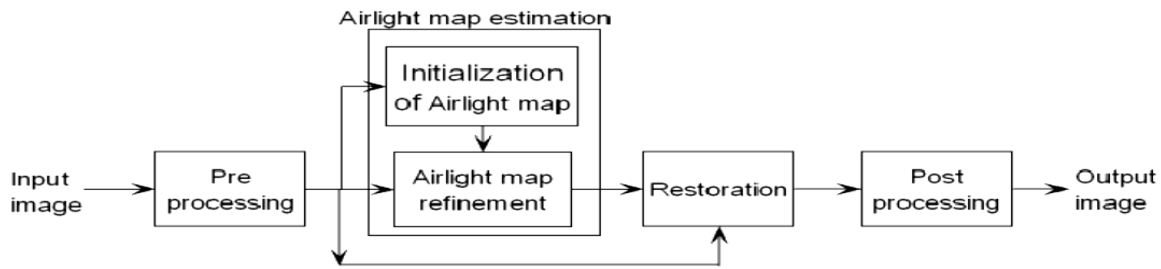


Fig 1 Block diagram of Proposed Haze removal model

The main Objectives will be:

- We present a windows adaptive method to estimate the transmission map;
- We propose a new energy model for dehazing and denoising simultaneously;
- We describe the existence and uniqueness of the minimizer of the proposed energy functional;
- To the best of our knowledge, the framework of the weighted vectorial total variation introduced here is somewhat new and could be applied elsewhere;
- This project is comprised of several components and thus has a number of key objectives
- Take as input any user-specified RGB source image that is polluted with haze.
- Accurately determine which areas are polluted with haze.
- Dehaze the image using the dark channel prior.
- Complete all computation in a reasonable amount of time (under 30 seconds for an 800x600 pixel image, if possible)

III Research Methodology/Planning of Work

The dark channel prior method has become a well-adopted. The algorithm has been used in numerous research projects to improve hazy images. A video application, for example, uses the dark channel prior to identify fog in hazy weather based on the traffic scene. The dark channel has also been extended in the past to improve underwater film and video quality. Additionally, efforts have been made to improve the current method of dark channel prior. Contrast enhancement, for example, was used to improve the contrast of the dark channel prior method, resulting in less colour distortion. Still, little effort has been made to speed up the computation time of the dark channel prior method. In the dark channel prior algorithm, there is a soft matting function. To improve the time it takes to clear up hazy images, we design an optimization algorithm that balances between a system of three bilateral filters and the darkish channel before. Using experimental results, it can be concluded that the proposed method correctly identifies hazy and clear areas in an image. Our primary goal was to improve the performance of the dark channel in the previous method, rather than to improve the quality of the final image. When compared to the traditional darkish channel prior approach, the results are significantly quicker with

speeds running at around 12 seconds for a 800x600 pixel image now.

Last but not least, techniques previously claimed their methods worked on photos that were polluted with smoke and other haze, but those methods never tested experimental results the use of photos other than hazy images. We used images of fog/haze as well as images that had been contaminated by smoke and steam in our approach. Prior to writing MATLAB code, we used a dark channel to conduct our experiments. To apply the guided joint filter in the bilateral filtering algorithm, we used MATLAB code from For the time being, we'll be running the code in MATLAB because it's a good choice for computational photography. Dehazing method as a whole and algorithm for the three-step bilateral filtering process are required. In order to produce haze-free images, the following outlines the steps and their associated functions. In our experiments, we used images taken in the field before the project as our test images. Smoke, steam, or haze can be seen in images. However, the results would be the same for a hazy or steamy image as the one used in this section.

Various computer vision-based applications now require haze removal algorithms. However, there are many aspects that have been overlooked in current approaches, such as the fact that no technique is accurate in every situation.

The results of a survey have revealed the overlooked aspects of the methods presented, such as noise reduction techniques.

Dehazing methods also face the problem of uneven and excessive illumination. Existing methods must therefore be revised in order that they function more effectively. An algorithm that incorporates a dark channel prior, CLAHE, and bilateral filter can yield better results.

IV Design and Implementation of Adaptive Filters

The adaptive filtering component of the algorithm begins after the dark channel priors of the image have been successfully computed. The dark channel priori-based image is greatly enhanced by this component, which also aids in increasing the efficiency of the algorithm's subsequent components. The term "adaptable" refers to filters that are able to change their filtering parameters (coefficients) over time in order to adapt to changing image dynamics. Adaptive filters must be able to learn on their own. The adaptive filter coefficients are able to adjust themselves as the input image arrives at the filter in order to achieve an optimal outcome, such as identifying an unknown filter

component or cancelling out noise in the input image. Some filter properties must be taken into account when designing an adaptive filter in order to get the best results. The following are a few of the benchmark properties. As a result, the rate at which a filter achieves its final state is determined by the filter's Convergence Rate. An adaptive system's ability to quickly adapt to changes in the environment is typically sought after. However, the rate of convergence is not independent of all other performance characteristics. Convergence rate will be improved at the expense of other performance metrics; this means that other performance metrics may suffer as a result.

Convergence can be sped up, but this reduces the system's stability, increasing the likelihood of divergence instead of convergence. Likewise, a decrease in convergence rate can cause the system to become more stable. This shows that the convergence rate can only be considered in relation to the other performance metrics, not by itself with no regards to the rest of the system.

- **Minimum Mean Square Error:**

As an indicator of system adaptability, the minimum mean square error (MSE) is used. Small minimum MSE indicates that the adaptive system accurately modelled, predicted and/or adjusted to the system's needs. In general, a large MSE indicates that the adaptive filter is unable to accurately model the given system or that the adaptive filter's initial state is insufficient to cause the adaptive filter to converge. Quantization noise, adaptive system order, measurement noise, and the error of the gradient due to the finite step size all play a role in determining the minimum MSE.

- **Computational Complexity**

Computational complexity is critical in real-time adaptive filtering systems. Hardware limitations can affect the performance of a real-time system when it is implemented. It will take a lot more computing power to run a complex algorithm than a simpler one.

- **Stability**

The adaptive system's most important performance metric is stability. There are very few adaptive systems that are completely asymptotically stable, due to the

nature of the adaptive system itself. The initial conditions, the transfer function, and the step size of the input determine the stability of most implemented systems.

- **Robustness**

In order to have a stable system, you must have a system that is robust. The system's ability to withstand both input and quantization noise is a measure of its robustness..

- **Filter Length**

Other performance metrics are directly related to the length of the adaptive system's filter. An adaptive filter's ability to accurately model a given system is determined by the filter's length. As a result, the filter length has an impact on the convergence rate, the stability of the system, and the minimum MSE when applied to certain step sizes. The maximum convergence rate will be reduced if the filter length of the system is increased. To put it another way, the number of computations will decrease if the filter length is reduced, thereby increasing the maximum convergence rate. You can add additional poles or zeroes to a filter in order to maintain stability when the filter length increases for a given system. To maintain stability, the maximum step size or convergence rate must be reduced. Because there aren't enough poles and zeros to model the system, the mean square error will eventually reach a nonzero constant. If the system is over specified, meaning it has too many poles and/or zeroes for the system model, it will have the potential to converge to zero, but increased calculations will affect the maximum convergence rate possible.

VI Conclusion

Algorithms based on the Dark Channel for dehazing In this paper, a major advancement in the field of picture dehazing was examined. As a result, the entire process of photograph haze removal had to be slowed down and complicated by Dark Channel's use of complex post-processing mechanisms. According to this paper's findings, improvements to the virtual image's statistics like Mean, Variance, and Entropy can be expected.

References

1. S.-R. Kuang and J.-P. Wang, "Design of power-efficient configurable booth multiplier," *IEEE Trans. Circuits Syst. I, Reg. Papers*, vol. 57, no. 3, pp. 568–580, Mar. 2010.
2. W. Ling and Y. Savaria, "Variable-precision multiplier for equalizer with adaptive modulation," in *Proc. 47th Midwest Symp. Circuits Syst.*, vol. 1, Jul. 2004, pp. I 553–I556.
3. Xilinx Inc.: XAPP 290: Two flows for Partial Reconfiguration: Module Based or Difference Based. www.xilinx.com, Sept. (2004).
4. O.A. Pfander, R. Hacker, and H.-J. Pfliederer, "A multiplexer-based concept for reconfigurable multiplier arrays," in *Proc. Int. Conf. Field Program. Logic Appl.*, vol. 3203, Sep. 2004, pp. 938–942.
5. H. Lee, "A power-aware scalable pipelined booth multiplier," in *Proc. IEEE Int. SOC Conf.*, Sep. 2004, pp. 123–126.
6. Mesquita, D., Moraes, F., Palma, J., Moller, L., Calazanas, N.: *Remote and Partial Reconfiguration of FPGAs: tools and trends. International Parallel and Distributed Processing Symposium*, (2003).
7. Meyer-Baese, U.: *Digital Signal Processing with Field Programmable Gate Arrays*. Springer, (2001).
8. Xilinx Inc.: *Development System Reference Guide*. www.xilinx.com.
9. A. Bermak, D. Martinez, and J.-L. Noullet, "High density 16/8/4-bit configurable multiplier," *Proc. Inst. Electr. Eng. Circuits Devices Syst.*, vol. 144, no. 5, pp. 272–276, Oct. 1997.

10. M. Hatamian and G. L. Cash, "A 70 MHz 8 bit x 8 bit parallel pipelined multiplier in 2.5 μ m CMOS," IEEE Journal of Solid-State Circuits, vol. 21, no. 4, pp. 505–513, 1986.
11. Yeong-Jae Oh, Hanho Lee, Chong-Ho Lee, "A Reconfigurable FIR Filter Design Using Dynamic Partial Reconfiguration", IEEE, vol-06, pp. 4851–4854, ISCAS 2006.
12. S.Karthick, Dr. S. Valarmathy and E.Prabhu," Reconfigurable Fir Filter With Radix-4 Array Multiplier" , jatit, Vol. 57 No.3, pp.326-336 , Nov.2013.
13. K.Anandan and N.S.Yogaanath, "VLSI Implementation of Reconfigurable Low Power Fir Filter Architecture" , IJIRCCE, Vol.2, Special Issue 1, pp no 3514-. 3520, March 2014.
14. Martin Kumm, Konrad M'oller and Peter Zipf "Dynamically Reconfigurable FIR Filter Architectures with Fast Reconfiguration", ieee journal of solid-state circuits, vol. 41, no. 4, april 2006.
15. Pramod Kumar Meher, Shrutisagar Chandrasekaran and Abbas Amira , "FPGA Realization of FIR Filters by Efficient and Flexible Systolization Using Distributed Arithmetic", ieee transactions on signal processing ,pp no-1-9.
16. Xiaoxiao Zhang, Farid Boussaid and Amine Bermak, "32 Bit \times 32 Bit Multiprecision Razor-Based Dynamic Voltage Scaling Multiplier With Operands Scheduler", ieee transactions on very large scale integration (vlsi) systems, vol. 22, no. 4, april 2014.
17. K.Gunasekaran and M.Manikandan," High Speed Reconfigurable FIR Filter using Russian Peasant Multiplier with Sklansky Adder", Research Journal of Applied Sciences, Engineering and Technology 8(24): 2451-2456, 2014.
18. Shidhartha Das, David Roberts, Seokwoo Lee, Sanjay Pant, David Blaau, Todd Austin, Krisztián Flautner and Trevor Mudge , "Self-Tuning DVS Processor Using Delay-Error Detection and Correction", ieee journal of solid-state circuits, vol. 41, no. 4, april 2006.
19. J Britto Pari ,et al., "Reconfigurable Architecture Of RNS Based High Speed FIR Filter" , Indian Journal Of Engineering And Material Sciences, pp. 230-240, vol.21, April 2014.