

Restoration of Lost Pixels Using Generalized Octagonal Fuzzy Numbers

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ABSTRACT

There are many methods that are used to restore the lost pixel in a digital image. The use of fuzzy numbers gives a simple method for restoring the lost pixels in a digital image. In this paper we use generalized octagonal fuzzy numbers obtained from the neighboring pixels of the lost pixel to restore it.

Keywords: lost pixel restoration, octagonal fuzzy numbers

INTRODUCTION

Application of Fuzzy numbers has dominated the research world owing to the ease, versatility and reliability of such numbers. They provide a simple means by which certain unknown but predictable quantities can be constructed based on them. When uncertainty is the issue, fuzzy numbers give a method by which the required number can be guessed to certain degree of certainty. Fuzzy numbers are not single numbers, but are rather a collection of numbers which have a 'look alike' behavior to a certain number to a certain degree of similarity. Depending upon how many numbers make up a fuzzy number, they are categorized as triangular fuzzy numbers, trapezoidal fuzzy numbers, pentagonal fuzzy numbers, hexagonal fuzzy numbers and so on. Each of these numbers can be used to construct a single number by a process called defuzzification. Initially the fuzzy numbers can be generated using a process called fuzzification. This pair of processes, the fuzzification – defuzzification is at the center of all application that use fuzzy numbers. Many researchers have successfully used fuzzy numbers in several application domains and addressed the problem of uncertainty [3, 6].

The recent development of digital technology has enabled the images to be captured and stored with ease and in little space without compromising on the visible quality to the human eye. However, owing to the way the images are processed in the digital form renders the image lose quality and sometimes parts of the image itself. These problems have been addressed by researchers who have designed several algorithms to retrieve the important aspects of the image so that the appeal to the viewer's eyes is not compromised. Amongst the many works done by the researchers are the following [2, 8]. Lost pixel restoration is being attempted by many researchers and have suggested many useful and workable methods. To name a few, we have [4, 9].

RESTORATION OF LOST PIXELS USING GENERALIZED OCTAGONAL FUZZY NUMBERS

We use generalized octagonal fuzzy numbers to retrieve the lost pixel. Octagonal fuzzy numbers have been used widely in several applications [5, 1]. This method is suitable when the lost pixel is at the center of the 5×5 sub-matrix of the pixel value matrix associated with the digital image. Assume that the 5×5 sub-matrix with the lost pixel at its center is as follows. In this method, we try to restore the lost pixel value using the pixel values in its neighboring pixels.

a_{11}	a_{12}	<i>a</i> ₁₃	a_{14}	a_{15}
a_{21}	a_{22}	a_{23}	a_{24}	a_{25}
a_{31}	a_{32}	X	<i>a</i> ₃₄	a_{35}
a_{41}	a_{42}	a_{43}	a_{44}	a_{45}
a_{51}	a_{52}	<i>a</i> ₅₃	<i>a</i> ₅₄	<i>a</i> ₅₅



To restore the lost pixel, we consider the following generalized octagonal fuzzy numbers namely,

 $(a_{22}, a_{23}, a_{24}, a_{32}, a_{34}, a_{42}, a_{43}, a_{44}),$ $(a_{12}, a_{13}, a_{14}, a_{21}, a_{22}, a_{23}, a_{24}, a_{25}),$ $(a_{21}, a_{31}, a_{41}, a_{12}, a_{22}, a_{32}, a_{42}, a_{52}),$ $(a_{25}, a_{35}, a_{45}, a_{14}, a_{24}, a_{34}, a_{44}, a_{54}) and$ $(a_{52}, a_{52}, a_{53}, a_{54}, a_{55}, a_{42}, a_{43}, a_{44}).$

Considering these sets after ordering as \tilde{A} , \tilde{B} , \tilde{C} , \tilde{D} , and \tilde{E} , with respective weights 1, 0.5, 0.5, 0.5 and 0.5, we rank these generalized octagonal fuzzy numbers following the ranking method given in [7]. The maximum rank determines the fuzzy set which is the most associated with the lost pixel. Select this set as the one that most represents X. By Defuzzifying the selected set using simple average and finding the value of the floor of this value, we get the lost pixel value.

The following illustration attempts to demonstrate the method discussed above for the restoration of lost pixels in a digital image.

Illustration 1

We shall demonstrate this with an example. Consider the following portion of a digital image, with a lost pixel *X*, which has to be restored.

131 138 140 136 110 141 150 149 132 155 140 142 129 145 152 142 130 139 145 155 138 144 148 150 138 156 159 144 137 138 143 152 155 150 160 162 130 152 175 X132 151 142 154 158 161 165 168 140 162 180 149 162 140 145 158 160 166 172 175 156 170 175 162 182 165 169 168 170 160 159 172 163 176 185 180 196 171 177 184 187 166 179 188 186 188 182 200 208 182 189 202 190 175 182 203 199 192 201 208 213 195 199 198 186 189 186 198 204 203 200 198 189 207 200 198 192 217 221 208 205 211 204 221 228 217 210 214 211 205 197 218 216 224 212 231 238 228 217 216 213 217 209 216

The 5×5 sub-matrix with *X* at its center is given below.

We consider the fuzzy sets (159, 144, 137, 132, 162, 149, 180, 175), (152, 142, 130, 156, 159, 144, 137, 138),(156, 152, 162, 152, 159, 175, 189, 175),(138, 151, 140, 130, 137, 132, 162, 182) and (175, 162, 182, 162, 180, 149, 162, 140).

Upon ordering, the sets are \tilde{A} = (132, 137, 144, 149,159, 162, 175, 180), \tilde{B} = (130, 137, 138, 142, 144,152, 156, 159), \tilde{C} = (152, 152,156, 159,162, 175, 175, 189), \tilde{D} = (130, 132, 137, 138, 140, 151, 162, 182) and \tilde{E} = (140, 149, 162,162, 162, 175, 180, 182). The corresponding generalized octagonal fuzzy sets are, (132, 137, 144, 149,159, 162, 175, 180; 1), (130, 137, 138, 142, 144, 152, 156, 159; 1), (152, 152, 156, 159, 162, 175, 175, 189; 1), (130, 132, 137, 138, 140, 151, 162, 182; 1) and (140, 149, 162,162, 162, 175, 180, 182; 1).

Now, using the ranking method suggested in [7], we get, $R(\tilde{A}) = 154.4286 = 154$, $R(\tilde{B}) = [144.5357] = 144$, $R(\tilde{C}) = [164] = 164$, $R(\tilde{D}) = [144.5714] = 144$, $R(\tilde{E}) = [164.25] = 164$. While the rank of the generalized fuzzy sets \tilde{C} and \tilde{E} are equal, they are equal only upon taking the floor values. However, $R(\tilde{E})$ is the highest amongst all the ranks. Hence, \tilde{E} is the fuzzy set most associated with the lost pixel X. Defuzzifying, using the simple average, the most likely value of X is 164.



CONCLUSION

The above suggested method gives a better approximation of the lost pixel compared to the restoration method that uses the generalized trapezoidal fuzzy numbers. It is apparent that, more the number of values used for restoration, better will be the approximate value of the lost pixel value.

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