

IMAGE QUALITY ASSESSMENT BASED ON NIQE, PIQE, GLCM, AND LBP USING SVM

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Abstract: Image Quality would be a characteristics of an image which measures the discern image humiliation. It plays a vital role in various image processing applications. Measuring Image Quality becomes more necessary as a result of diverse applications involving digital imaging and communication. The main aim of Image Quality assessment is to produce quality measures which will predict discern image quality mechanically. various techniques are planned for measuring the quality of the image. In this paper, we proposed one of the technique to predict and to evaluate visual quality of an image by Support Vector Machine (SVM) which classifies the images along with image quality descriptors which are NIQE, PIQE, GLCM, LBP used for feature extractions of an images. After the classification, we can predict that whether the image is quality or not based on quality score produced.

Index Terms-SVM, image quality descriptors, Naturalness image quality evaluator(NIQE), perception image quality evaluator (PIQE), Gray-level co-occurrence matrix(GLCM), Local binary pattern(LBP), feature extractions.

INTRODUCTION

Image quality assessment(IQA) can be referred as the evaluation of the quality of a distorted image with respect to the original image. In image quality assessment we can encountered with 2 categories of IQA; namely subjective methods and objective methods [1]. Subjective methods are difficult and time-consuming for real-time assessments needs. This method requires human intervention and are very expensive. While the objective methods refer to the algorithmic models to estimate the image quality. For the real-time assessments we require automated or objective methods. There are three quality metrics for objective IQA, namely Full Reference FR, Reduced reference RR and No-reference NR. FR IQA methods requires the whole information of the reference image, While NR IQA methods determine the distorted image be assessed without any information of the reference image. RR methods are a compromise between FR and NR methods that needs partial information of the reference image.

Machine learning is the scientific study of statistical models and algorithms that computer systems use to effectively perform a specific task without any direct instructions. Machine learning algorithms develop a mathematical model of sample data, known as “training data”, in order to make predictions or decisions without being explicitly programmed to accomplish specific task [2].

Machine learning algorithms are used in various applications such as email filtering, detection of network intruders, images classification and computer vision. Machine learning tasks are classified into several wide categories. In supervised learning, the algorithm builds a mathematical model from a set of data that consists both inputs and the desired outputs. In semi-supervised learning algorithms progress mathematical models from incomplete training data, where a part of the sample input doesn't have labels. While, in unsupervised learning, the algorithm develops a mathematical model from a set of data which contains only inputs and no desired output labels.

In supervised learning we prefer classification algorithms. As Classification algorithms is one of the most important task for various application such as text categorization, tone recognition, image classification etc. This paper refers to Support Vector Machine(SVM) classifier as it is capable of delivering higher performance in terms of classification accuracy than the other data classification algorithms [3] and this process is commonly referred to linear SVM known as optimized linear regression which are often applied to linearly separable problems. In order to classify the images, we need to extract features of the images for that we can use descriptors.

In this paper to extract features of an images we prefer Naturalness image quality evaluator (NIQE), Perception image quality evaluator (PIQE), Gray-level co-occurrence matrix (GLCM), Local Binary pattern (LBP).NIQE is a completely blind quality analyzer that only makes use of measurable derivations from statistical regularities perceive in natural images without training on human-rated distorted images, and, certainly without any exposure to distorted images and PIQE also extracts quality score for an image and there are somewhat similar to NIQE.GLCM and its function characterize the texture of an image by calculating how often pairs of pixel with specific values occur in an image and by creating GLCM, we then extract statistical measures from this matrix and that statistics will provide information about the texture of an image like contrast, energy etc. while the LBP features encode localtexture information that are obtained from GLCM. The results from these descriptors are given to SVM to trained the machine and the accuracy of the result is depending upon the how well the machine is trained.

II. LITERATURE SURVEY:

In previous several techniques have implied for image quality assessment [4]. In that methods we have seen in how many ways we can assess the image quality in objective method. [1] clearly explains the methods and databases required for image quality assessments. In that we can see various approaching for assessment. In [2] we can see overview of machine learning and its applications. In [3] they have discussed about the classification of data using support vector machine which plays vital role in this paper. While in [4] that paper they have discussed about the various assessments approaches used to evaluate the quality of an image but their experimental results demonstrate that the MSE and PSNR methods are simple and are easy to implement but it does not correlate highly with human awareness. In [5] they have proposed a new approach to design FR-IQA algorithm. This approach is depending on a classification process such as the human being is supposed to proceed to judge the quality of an object and to apply the classification process, a vector of features has been generated. In that paper selected features are chosen from FR image HVS-based features and FR image features, for both of them a reference image is needed. The obtained results show that LMIQM gives better results and yields a significant improvement of the correlation coefficients with human judgements. In [6] they proposed about the NIQE descriptor. In previous reference we have seen how to assess the image quality using FR but what about with NR and this paper will give response to the above statement. [7] proposes performance analysis of No reference image quality based on Human Perception. [8] provides the in detail information about database related to image quality assessment. [9] provides the information about the LBP.

III. PROPOSED METHOD:

In this paper, proposed method refers to image quality assessment based on NIQE, PIQE, GLCM and LBP using SVM. In this

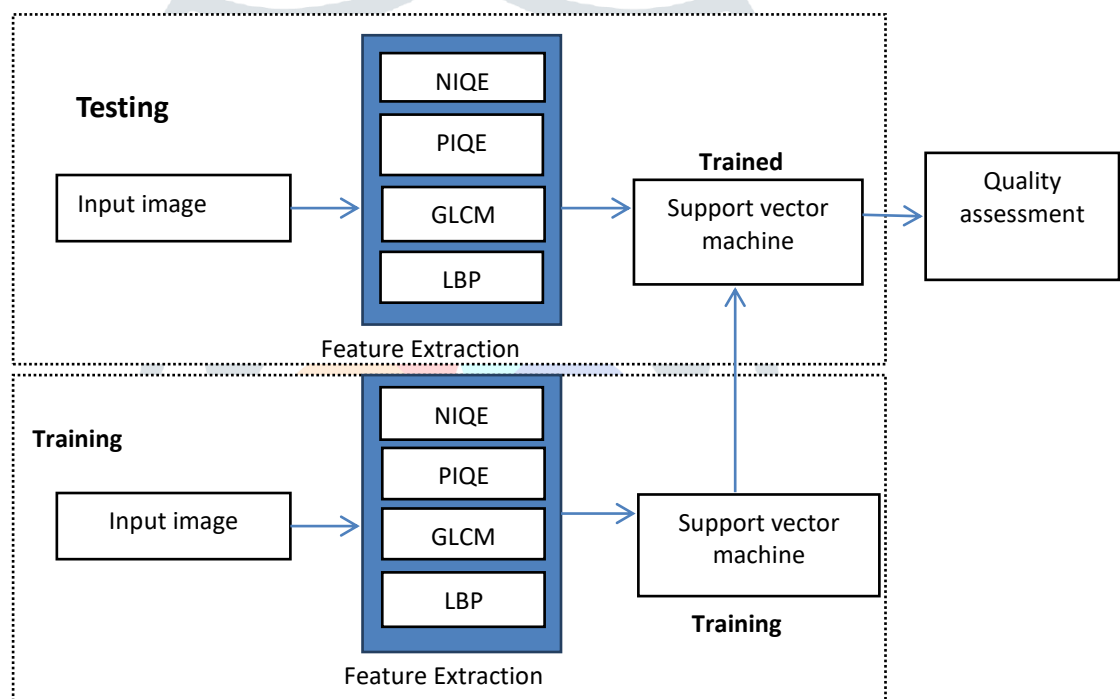


Fig: Block diagram of image quality assessment based on NIQE, PIQE, GLCM and LBP using SVM

process, a set of images are given for the descriptors which are NIQE, PIQE, GLCM, and LBP, along with briquise are used for the feature extractions of an images. For this process required database is collected from Waterloo university with 100 pristine images. By using 100 pristine images four distortion types with five levels each are chosen to alter the source images. All distorted images are generated using MATLAB functions as follows:

JPEG compression: The quality factor that parameterizes the DCT quantization matrix is set to be $\text{jpeg_level} = [43, 12, 7, 4, 0]$; for five levels, respectively.

JPEG2000 compression: The compression ratio is set to be $\text{jp2k_level} = [0.46, 0.16, 0.07, 0.04, 0.02]$ for five levels, respectively.

Gaussian blur: 2D circularly symmetric Gaussian blur kernels with standard deviations (std) of $\text{gblur_level} = [7, 15, 39, 91, 199]$ for five levels are used to blur the source images.

White Gaussian noise: white Gaussian noise is added to the source images, where variances are set to be $\text{wn_level} = [-10, -7.5, -5.5, -3.5, 0]$ for five levels, respectively.

The above four distortion types are the most common ones in existing IQA databases, and many IQA models are claimed to excel at handling these distortions [8]. For this process, we have generated 2nd level of distorted images.

TYPES OF DESCRIPTORS:**NIQE:**

NIQE is a completely blind quality analyzer that only makes use of measurable derivations from statistical regularities perceived in natural images without training on human-rated distorted images, and, certainly without any exposure to distorted images [6]. It is based on the construction of a quality aware collection of statistical features depends on simple and successful space domain natural scene statistic (NSS) model. These features are derived from an entity of natural, undistorted images. The NIQE descriptor extracts quality score for an image and smaller score of it indicates better perceptual Quality.

NIQE score for original image is 2.8867 NIQE score for noisy image is 11.0894 NIQE score for blur image is 5.5879.



Original image

noise image

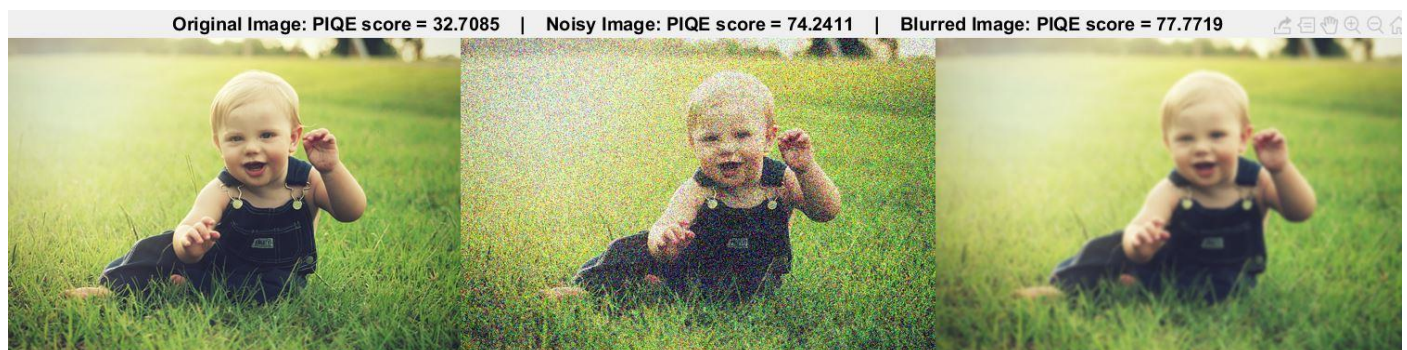
blur image

The above images and niqe values can be generated by the below matlab commands:

```
i = imread('00306.bmp');
inoise = imnoise(i, 'salt& pepper', 0.02);
iblur = imgaussfilt(i, 2);
niqeI = niqe(i);
fprintf('NIQE score for original image is %0.4f.\n', niqeI)
niqeInoise = niqe(inoise);
fprintf('NIQE score for noisy image is %0.4f.\n', niqeInoise)
niqeIblur = niqe(iblur);
fprintf('NIQE score for blurry image is %0.4f.\n', niqeIblur)
```

PIQE and BRISQUE:

PIQE calculates the no-reference image quality score for image A using a perception based image quality evaluator and brisque is also used for to get better results but it has less preference in this process. BRISQUE evaluates the no-reference image quality score for image A using the Blind/Referenceless Image Spatial Quality Evaluator (BRISQUE). Brisque compare A to a default model computed from images of natural scenes with similar distortions. A smaller score indicates better perceptual quality for both PIQE and BRISQUE.



The above images and piqe values can be generated by the below matlab commands:

```
A = imread('00306.bmp');
Anoise = imnoise(A, 'Gaussian', 0, 0.05);
Ablur = imgaussfilt(A, 2);
score = piqe(A);
score_noise = piqe(Anoise);
score_blur = piqe(Ablur);
figure
```

```
montage({A,Anoise,Ablur},'Size',[1 3])
title(['Original Image: PIQE score = ', num2str(score),' | Noisy Image: PIQE score = ', num2str(score_noise),' ' ...
' | Blurred Image: PIQE score = ', num2str(score_blur)]), 'FontSize',12)
```

GLCM and LBP:

GLCM could be a method inspects texture that considers the spatial relationship of pixels is that the gray-level co-occurrence matrix (GLCM), additionally referred to as the gray-level spatial dependence matrix. The functions of GLCM characterize the feel of a picture by hard however usually pairs of component with specific values and in a very such that spatial relationship occur in an image,making a GLCM, and then extracting applied mathematics measures from this matrix.

By victimization graycomatrix command we are able to produce GLCM and that we can derive many statistics from them using graycoprops command . These statistics give into concerning the feel of a picture. The co-occurrence matrices offers into concerning the homogeneity of a picture, its distinction, linearity, etc. we are able to select five efficient measures extracted from the GLCMs: energy, entropy, correlation, homogeneity and inertia [9].

The LBP descriptor operator is outlined as a gray-scale invariant texture live, derived from a general definition of texture in a very native neighborhood. It is created invariant against the rotation of the image domain, and supplemented with a rotation invariant live of native distinction. The LBP is referred as a unifying texture model that describes the formation of a texture with small textons and their applied mathematics placement rules. The fundamental LBP is extended to facilitate the analysis of textures with multiple scales by combining neighborhoods with varied sizes. The MATLAB command for LBP is extractLBPFeatures () [9].

The commands to extract options from LBP are given below:

features = extractLBPFeatures(I) returns extracted uniform native binary pattern (LBP) from a grayscale image.

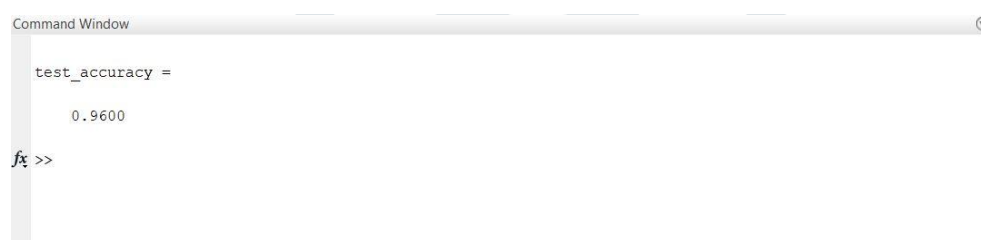
The LBP options inscribe native texture information.

features = extractLBPFeatures(I,Name,Value) uses extra choices such that by one or additional Name, price combinearguments.

After the options extractions associate in nursing pictures from the descriptors we will trained the SVM per the values of the extractions. Once the coaching is completed, the support vector machine will classify any image that are given as an output. In testing a group of Ima- gesor a picture is given as associate in nursing input then feature extractions are do for the specific images/image. After that, the featuresof associate in nursing input are given to support vector machine and because the machine is already trained it will the image and manufacture the output which determine whether or not the image quality is nice or not.

IV.RESULTS:

Experimental results of this process have been showing that NR based image quality assessments can also be used for IQA. This results are completely based upon the database that we have preferred for this process. The below image shows the experimental results of this process. By discern the results we can determine whether the image quality or not.



```
Command Window
test_accuracy =
    0.9600
fx >>
```

Fig: quality score of the image

V.CONCLUSION AND FUTURE SCOPE:

In this paper, we proposed image quality assessment method which deal determination of image quality. Features extractions of the Images captured with different types of descriptors were the level of accuracy of an image and classification of an images is done by using Support Vector Machine(SVM). The proposed method relies on four types of descriptors: NIQE, PIQE, GLCM, and LBP descriptors. The NIQE descriptor extracts quality score for an image and smaller score of it indicates better perceptual Quality. Similarly, PIQE and briques are also extracts quality score for an image and there are somewhat similar to NIQE. GLCM and its function characterize the texture of an image by calculating how often pairs of pixel with specific values occur in an image and by creating GLCM, we then extract statistical measures from this matrix and that statistics will provide information about the texture of an image like contrast, energy etc. while the.LBP features encode local texture information. The scores obtained from

descriptors were given to Support Vector Machine(SVM) Which classify the images and gets trained and produce accuracy score which determine image quality. These descriptors are selected by performing trial and error method on my database and these given a good accuracy among all others. Intensive experiments were conducted to evaluate the quality of an image using database which is certified by the Waterloo university, Canada. The results indicate that the proposed method significantly performed these methods.

Although the Proposed method significantly perform image quality assessment with present database, we have check its performance with another database too. It can be improved by determining how much quality that the image has so that we can select or predict the image which has to be displayed/selected by using different level of descriptors.

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