

DESIGN THINKING APPROACH FOR DEEP VISUAL SALIENCY ON STEREOSCOPIC USING IMAGE PROCESSING

Dr.M.PRAVEENA¹, Associate Professor,
praveenamamarannan@gmail.com

GOKILA PRIYA. S², GOWSIK. M², GOWTHAM. K²
Department of Computer Science,

Dr.SNS Rajalakshmi College of Arts and Science (Autonomous), Coimbatore - 49.

ABSTRACT

Design thinking of Detecting visually salient areas in pix is necessary troubles Salient object areas is a gentle decomposition of foreground and heritage photo elements. To realize salient areas in an picture in phrases of saliency map. To create saliency map by way of the usage of linear aggregate of colorations in High dimensional colour space. To enhance the overall performance of saliency estimation, make use of the relative area and coloration distinction between brilliant pixels. To get to the bottom of the saliency estimation from tramp with the aid of the usage of gaining knowledge of primarily based algorithm. To create three bench mark datasets it is environment friendly in assessment with preceding country of artwork saliency estimation methods. Image processing very regularly exists as a technique for visible inspection in industry. Automated structures for visible inspection are very vital section of the high-quality manipulate in manufacturing line. Quality manipulate has been usually carried out in difficult work conditions. Using automatic visible structures nice manipulate turns into easier. The perceptual fantastic of stereoscopic snap shots performs an indispensable function in the human grasp of visible information. However, most accessible stereoscopic picture exceptional assessment (SIQA) strategies consider 3D visible trip the use of homemade points or shallow architectures, which can't mannequin the visible homes of stereo snap shots well. In this paper, we use convolution neural networks (CNNs) to analyze deeper neighborhood quality-aware buildings for stereo images. With one of a kind inputs, two CNN fashions are designed for no-reference SIQA tasks. The one-column CNN mannequin without delay accepts a cyclopean view as the input, and the three-column CNN mannequin collectively considers the cyclopean, left and proper

Identifying visually salient areas is beneficial in purposes such as object primarily based photograph retrieval, adaptive content material transport adaptive region-of-interest primarily based photograph compression, and clever photograph

views as CNN inputs. The two SIQA frameworks share the identical implementation approach. First, to overcome the impediment of constrained SIQA datasets, we receive picture patches that have been cropped from corresponding stereo pairs as inputs for neighborhood quality-sensitive function extraction. Next, a nearby function choice algorithm is used to get rid of associated facets on non-salient patches, which ought to motive massive prediction errors. Finally, the reserved nearby visible constructions of salient areas are aggregated into a remaining best rating in an end-to-end manner. Experimental effects on three public SIQA databases show that our technique outperforms most state-of-the-artno-reference (NR) SIQA methods. The outcomes of a cross-database test additionally exhibit the robustness and generality of theproposed method. The photograph acquisition module and the photograph processing software program graph module are used to figuring out visually salient areas is beneficial in purposes such as object primarily based photo retrieval, adaptive content material transport adaptive region-of-interest based totally photo compression, and clever picture resizing. We discover salient areas as these areas of an photograph that are visually extra conspicuous with the aid of advantage of their distinction with appreciate to surrounding regions. Similar definitions of saliency exist in literature the place saliency in pix is referred to as nearby contrast.

Keywords—Design thinking, empathize, ideate, prototype. Feature extraction, Visualization, Distortion, Saliency detection, Two dimensional displays, Image color analysis.

INTRODUCTION

SALIENT REGION DETECTION

resizing. We pick out salient areas as these areas of an picture that are visually extra conspicuous with the aid of advantage of their distinction with appreciate to surrounding regions. Similar definitions of saliency exist in literature the place

saliency in photographs is referred to as nearby contrast.

Method for discovering salient areas makes use of a distinction willpower filter that operates at more than a few scales to generate saliency maps containing “saliency values” per pixel. Combined, these character maps end result in our ultimate saliency map. We reveal the use of the last saliency map in segmenting total objects with the resource of a pretty easy segmentation technique. The novelty of our method lies in discovering excessive best saliency maps of the equal dimension and decision as the enter picture and their use in segmenting total objects. The approach is positive on a broad vary of photographs consisting of these of paintings, video frames, and photographs containing noise

Salient vicinity detection is essential in picture perception and analysis. Its intention is to observe salient areas in an picture in phrases of a saliency map, the place the detected areas would draw humans’ attention. Many preceding research have proven that salient area detection is useful, and it has been utilized to many purposes inclusive of segmentation, object recognition, photograph retargetting, picture rearrangement, photo best assessment, photo thumbnailing and video compression. The principal techniques as comply with as:

- Salient place decision is modeled as the facility vicinity problem, which is solved with the aid of maximizing a submodular goal function. This offers a new standpoint the use of submodularity for salient place detection, and it achieves state-of-art overall performance on two public saliency detection benchmarks.

- The similarities between hypothesized place facilities and their place factors are formulated as a labeling hassle on the vertices of a graph. It is solved by using discovering a harmonic feature on the graph, which has a closed-form solution.

- We current an environment friendly grasping algorithm with the aid of the use of the submodularity property of the goal function.

- We naturally combine high-level priors with low-level saliency into a unified framework for salient location detection.

SUPERPIXEL

Superpixel map has many favored properties:

- It is computationally efficient: it reduces the complexity of pictures from thousands of heaps of pixels to solely a few hundred superpixels.

- It is additionally representationally efficient: pairwise constraints between units, whilst solely for adjoining pixels on the pixel-grid, can

now mannequin a great deal longer-range interactions between superpixels.

- The superpixels are perceptually meaningful: every superpixel is a perceptually constant unit, i.e. all pixels in a superpixel are most probably uniform in, say, coloration and texture.

- It is near-complete: due to the fact superpixels are effects of an over segmentation, most buildings in the picture is conserved. There is very little loss in shifting from the pixel-grid to the superpixel map.

It is virtually no longer novel to use superpixels or atomic areas to velocity up later-stage visible processing; the concept has been round the neighborhood for a while.

(1) To empirically validate the completeness of superpixel maps

(2) To follow it to resolve difficult imaginative and prescient issues such as discovering humans in static images.

TRIMAP SEGMENTATION

Digital matting consists in extracting a foreground aspect from the background. Standard strategies are initialized with a trimap, a partition of the photograph into three regions: a particular foreground, a precise background, and a blended location the place pixels are viewed as a combination of foreground and heritage colors. Recovering these colours and the share of combination between each is an under-constrained inverse problem, touchy to its initialization: one has to specify an correct trimap, leaving undetermined as few pixels as possible. First, we suggest a new segmentation scheme to extract an correct trimap from simply a coarse indication of some historical past and/or foreground pixels.

Standard statistical fashions are used for the foreground and the background, whilst a precise one is designed for the blended region. The segmentation of the three areas is carried out concurrently through an iterative Graph Cut based totally optimization scheme. This uncomplicated trimap is comparable to cautiously hand unique ones. As a 2nd step, we take benefit of our blended vicinity mannequin to layout an extended matting technique coherent. Based on international records alternatively than on nearby ones, our approach is a whole lot quicker than popular Bayesian matting, barring first-rate loss, and additionally usable with guide trimaps.

RANDOM FORESTS

A random woodland multi-way classifier consists of a variety of trees, with every tree grown

the usage of some structure of randomization. The leaf nodes of every tree are labeled by means of estimates of the posterior distribution over the picture classes. Each inner node incorporates a check that excellent splits the area of facts to be classified. An picture is categorised by way of sending it down each tree and aggregating the reached leaf distributions. Randomness can be injected at two factors in the course of training: in subsampling the coaching records so that every tree is grown the use of a exceptional subset; and in deciding on the node tests

An photograph is represented the use of the spatial pyramid scheme proposed via Lazebnik et al, which is based totally on spatial pyramid matching, however right here utilized to each look and shape. The illustration is illustrated.

Appearance. SIFT descriptors are computed at factors on a normal grid with spacing M pixels. At every grid factor the descriptors are computed over 4 round guide patches with special radii, hence every factor is represented by way of 4 SIFT descriptors. Multiple descriptors are computed to enable for scale variant between images. The dense points are vector quantized into V visible phrases the usage of K-means clustering.

Shape. Local structure is represented via a histogram of part orientations gradients inside an photo subregion quantized into K bins. Each bin in the histogram represents the quantity of edges that have orientations inside a positive angular range. This illustration can be in contrast to the usual “bag of (visual) words”, the place right here every visible phrase is a quantization on part orientations.

DECISION TREES

In order to apprehend how random forests works it is essential to end up acquainted with choice trees. Decision timber are predictive fashions that use a set of binary policies to calculate a goal value. Two sorts of selection timber are classification timber and regression trees. Classification bushes are used to create specific statistics units such as land cowl classification and regression timber are used to create non-stop records units such as biomass and percentage tree cover.

HOW RANDOM FORESTS WORKS

Random forests, like selection trees, can be used to remedy classification and regression issues however it is in a position to overcome the drawbacks related with single choice bushes whilst keeping the benefits. The random forests mannequin calculates a response variable (e.g., land cover, percentage tree cover) by way of developing many (usually a number of hundred) distinct selection timber (the woodland of trees) and then

placing every object to be modeled (in our case the object is a multi-layered pixel) down every of the selection trees. The response is then decided by means of evaluating the responses from all of the trees. In the case of classification the category that is expected most is the category that is assigned for that object (Leo Breiman & Cutler A.).

In different words, if five hundred bushes are grown and four hundred of them predict that a unique pixel is woodland and a hundred predict it is grass the estimated output for that pixel will be forest. In the case of regression the ensuing price for an object is the suggest of all of the predictions. Since predictions from random forests are derived the use of a wooded area of timber it is now not viable to without difficulty illustrate how the predictions are made. To illustrate the system it would be imperative to draw all of the bushes for every prediction which would end result in heaps of choice tree diagrams for every model.

The key to the success of random forests is how it creates every of the selection bushes that make up the forest. There are two steps involving random choice that are used when forming the timber in the forest. The first step entails randomly selecting, with replacement, facts from furnished education areas to construct every tree. For every tree a one of a kind subset of the coaching facts are used to boost the choice tree mannequin and the closing one-third of the education records are used to check the accuracy of the model. The pattern statistics used for trying out are regularly referred to as the “out-of-bag” samples. The 2nd random sampling step is used to decide the break up prerequisites for every node in the tree. At every node in the tree a subset of the predictor variables is randomly chosen to create the binary rule.

When jogging random forests there are a variety of parameters that want to be specified. The most frequent parameters:

- Input education facts which include predictor variables such as photo bands and digital elevation fashions and response variables such as land cowl kind and biomass
- The quantity of timber that have to be built
- The range of predictor variables to be used to create the binary rule for every split
- Parameters to calculate data associated to error and variable significance.

LITERATURE REVIEW

SLIC SUPERPIXELS

R. Achanta, A. Shaji, al., [1] has proposed superpixels are turning into more and more famous for use in laptop imaginative and prescient

applications. However, there are few algorithms that output a preferred range of regular, compact superpixels with a low computational overhead. We introduce a novel algorithm that clusters pixels in the mixed 5 dimensional coloration and photograph airplane house to effectively generate compact, almost uniform superpixels. The simplicity of our strategy makes it extraordinarily effortless to use a lone parameter specifies the wide variety of superpixels and the effectivity of the algorithm makes it very practical. Experiments exhibit that our strategy produces superpixels at a decrease computational price whilst reaching a segmentation first-rate equal to or larger than 4 present day methods, as measured by way of boundary recall and under-segmentation error. We also reveal the advantages of our superpixel strategy in distinction to current strategies for two duties in which superpixels have already been proven to extend overall performance over pixel-based methods.

Our strategy generates superpixels by means of clustering pixels based totally on their colour similarity and proximity in the picture plane. This is executed in the five-dimensional [labxy] space, the place [lab] is the pixel colour vector in CIELAB colour space, which is extensively regarded as perceptually uniform for small shade distances, and xy is the pixel position. While the most feasible distance between two shades in the CIELAB area (assuming s RGB enter images) is limited, the spatial distance in the xy airplane relies upon on the photograph size. It is now not viable to absolutely use the Euclidean distance in this 5D house barring normalizing the spatial distances. In order to cluster pixels in this 5D space, we consequently introduce a new distance measure that considers superpixel size. Using it, we put into effect coloration similarity as properly as pixel proximity in this 5D area such that the predicted cluster sizes and their spatial extent are about equal.

Superpixels supply a handy primitive from which to compute nearby photo features. They seize redundancy in the picture and radically minimize the complexity of subsequent photo processing tasks. They have proved more and more beneficial for functions such as depth estimation, photo segmentation, skeletonization, physique mannequin estimation and object localization. For superpixels to be beneficial they ought to be fast, handy to use, and produce excessive great segmentations. Unfortunately, most today's superpixel techniques do no longer meet all these requirements. As we will demonstrate, they frequently splendid from a excessive computational cost, negative fantastic segmentation, inconsistent measurement and shape,

or incorporate a couple of difficult-to-tune parameters.

HIGH-DIMENSIONAL COLOR TRANSFORM FOR SALIENCY DETECTION

J. Kim, D. Han, al., [2] has proposed to introduce a novel approach to routinely notice salient areas of an photograph through highdimensional colour transform. Our predominant concept is to signify a saliency map of an photo as a linear aggregate of high-dimensional coloration house the place salient areas and backgrounds can be distinctively separated. This is primarily based on an commentary that salient areas regularly have exclusive colorings in contrast to the heritage in human perception, but human appreciation is frequently problematic and tremendously nonlinear. By mapping a low dimensional RGB coloration to a function vector in a high-dimensional coloration space, we exhibit that we can linearly separate the salient areas from the historical past by way of discovering an most excellent linear mixture of colour coefficients in the high-dimensional colour space. Our excessive dimensional coloration area contains a couple of coloration representations together with RGB, CIELab, HSV and with gamma corrections to enrich its consultant power. Our experimental consequences on three benchmark datasets exhibit that our approach is effective, and it is computationally environment friendly in evaluation to preceding today's techniques.

In this paper, exploring the strength of special shade house representations, we suggest high-dimensional colour radically change which maps a low dimensional RGB coloration tuple into a high-dimensional characteristic vector. Our excessive dimensional shade seriously change combines various consultant colour areas such as RGB, CIELab, HSV, collectively with extraordinary gamma corrections to enrich the consultant electricity of our high-dimensional coloration seriously change space. Starting from a few preliminary shade examples of detected salient areas and backgrounds, our method estimates an most beneficial linear aggregate of shade values in the high-dimensional shade radically change house that consequences in a per-pixel saliency map. As validated in our experimental results, our per-pixel saliency map represents how specific the shade of salient areas is in contrast to the colour of the background. Note that a easy linear mixture or transformation of the coloration area can't gain consequences comparable to ours.

Assumptions Since our method makes use of solely colour statistics to separate salient areas from the background, our approach shares a hassle when identically-colored objects are current in each the

salient areas and the background. In such cases, making use of high-level features, such as texture, is the solely way to get to the bottom of this ambiguity. Nevertheless, we exhibit that many salient areas can truly be detected the usage of solely coloration statistics through our highdimensional shade radically change space, and we attain excessive detection accuracy and higher overall performance in contrast with many preceding techniques that makes use of a couple of high-level features.

SALIENT OBJECT DETECTION

A. Borji, M.-M. Cheng al.,[3] has proposed detecting and segmenting salient objects in herbal scenes, frequently referred to as salient object detection, has attracted a lot of pastime in laptop vision. While many fashions have been proposed and quite a few functions have emerged, but a deep perception of achievements and troubles is lacking. We goal to furnish a complete evaluation of the latest growth in salient object detection and situate this subject amongst different intently associated areas such as regularly occurring scene segmentation, object notion generation, and saliency for fixation prediction. Covering 228 publications, we survey i) roots, key concepts, and tasks, ii) core strategies and foremost modeling trends, and iii) datasets and comparison metrics in salient object detection. We additionally talk about open issues such as contrast metrics and dataset bias in mannequin overall performance and recommend future lookup directions.

Humans are capable to observe visually distinctive, so known as salient, scene areas without difficulty and swiftly (i.e., pre-attentive stage). These filtered areas are then perceived and processed in finer important points for the extraction of richer high-level data (i.e., attentive stage). This functionality has lengthy been studied with the aid of cognitive scientists and has lately attracted a lot of pastime in the pc imaginative and prescient neighborhood mostly due to the fact it helps locate the objects or areas that correctly characterize a scene and consequently harness complicated imaginative and prescient troubles such as scene understanding. Some matters that are carefully or remotely associated to visible saliency include: salient object detection, fixation prediction, object importance, memorability, scene clutter, video interestingness, surprise, picture first-rate assessment, scene typicality, aesthetic and attributes. Given house limitations, this paper can't totally discover all the aforementioned lookup directions. Instead, we solely focal point on salient object detection, a lookup region that has been substantially developed

in the previous twenty years in precise considering the fact that 2007.

OBJECT DETECTION: A BENCHMARK

A. Borji, M.-M. Cheng al.,[4] has proposed countless salient object detection strategies have been posted which have been assessed the usage of special comparison rankings and datasets ensuing in discrepancy in mannequin comparison. This calls for a methodological framework to examine present fashions and consider their execs and cons. We analyze benchmark datasets and scoring methods and, for the first time, supply a quantitative contrast of 35 country of the artwork saliency detection models. We locate that some fashions function persistently higher than the others. Saliency fashions that intend to predict eye fixations operate lower on segmentation datasets in contrast to salient object detection algorithms. Further, we suggest mixed fashions which exhibit that integration of the few exceptional models outperforms all fashions over different datasets. By examining the consistency amongst the nice fashions and amongst human beings for every scene, we discover the scenes the place fashions or people fail to notice the most salient object. We spotlight the cutting-edge problems and recommend future lookup directions. Recently, salient object detection has attracted a lot of hobby in laptop imaginative and prescient as it presents quick options to various complicated processes. Firstly, it detects the most salient and attention-grabbing object in a scene, and then it segments the complete extent of that object. The output commonly is a map the place the depth of every pixel represents the likelihood of that pixel belonging to the salient object. This trouble in its essence is a segmentation hassle however barely differs from the regular generic photo segmentation. While salient object detection fashions phase solely the salient foreground object from the background, frequent segmentation algorithms partition an photograph into areas of coherent properties. Saliency object detection strategies additionally range from different saliency fashions that intention to predict scene areas the place a human observer might also fixate. Since saliency models, whether or not they tackle segmentation or fixation prediction, each generate saliency maps; they are interchangeably applicable.

To the authors' high-quality knowledge, such try for benchmarking salient object segmentation techniques has no longer been reported. Unfortunately, these strategies have regularly been evaluated on special datasets, which in some instances are small and no longer effortlessly accessible. The lack of posted benchmarks reasons

discrepancy in quantitative evaluation of competing models. Not solely does a benchmark permit researchers to evaluate their fashions with different algorithms, however it additionally helps perceive the chief elements affecting performance. This ought to end result in an even quicker overall performance improvement.

Existing System

Saliency detection ambitions to simulate the human visible gadget for detecting the most beautiful regions. Some complicated saliency photos that incorporate uneven illumination, preceding strategies are unable to absolutely look at entire saliency regions, in particular when the saliency objects stumble on in the photo shadow region. Such strategies make use of low-level processing to decide the distinction of picture areas to their surroundings, and use function attributes such as intensity, color, and edges. We commonly classify the algorithms into neighborhood and international schemes. Note that the classification is no longer strict as some of the lookup efforts can be listed beneath each categories. Local distinction based totally strategies look at the rarity of photo areas with appreciate to (small) neighborhood neighborhoods.

Methods can't concurrently realize all sorts of floor saliency's, such as permitting the detection of each giant saliency objects and small saliency objects. Affected via the uneven illumination, present saliency detection strategies can't successfully section the correct saliency vicinity from the saliency images. The current algorithms have terrible adaptability to saliency's with miscellaneous patterns or low contrast, and fail to discover the total saliency regions. In order to understand the on line detection of product quality, the saliency detection algorithm wants to meet the real-time requirements. The spatial smoothness of the shading and piecewise continuity of the reflectance and employed a quickly computation approach to remedy the minimization problem. Methods have efficiently addressed this trouble through incorporating constrains on the decomposition mode. The reflectance layer stays uneven to some extent and can't be employed for saliency detection. These strategies have executed incredible overall performance for uneven illumination elimination in the nature scene, however there are nevertheless now not fantastic for floor saliency images.

Disadvantages

- Less Color code density.
- Artifacts may appear.

- Segmentation accuracy not proper
- Edges not clear.
 - Inaccurate results in extraction high density images

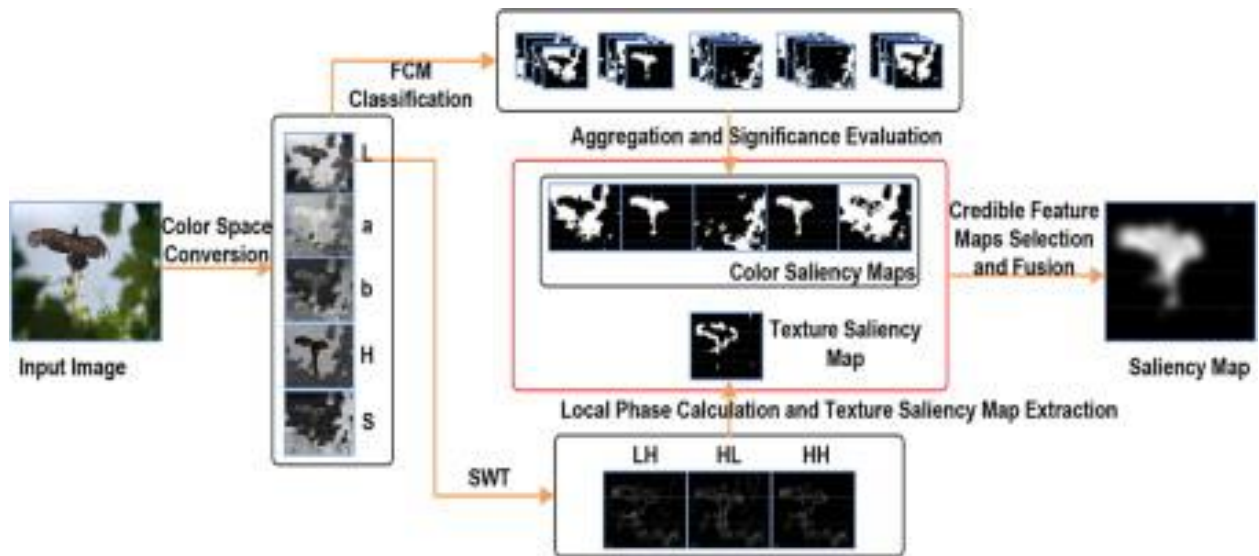
Proposed System

A technique to ranks the product is worked to present a product ranking model that applies weights to product review factors to calculate a product ranking score. In this proposed system, the sentences that are not related with the quality of a product such as customer service or sentence related to the. In this paper the pre processing is done by Support Vector Machine (SVM). First of all, it removes the comments which neither is nor related with the quality of the product. Second stage describes the weights of the reviews based on the votes. The final stage calculates the overall ranking of the product. The ranking score is calculated by the relevance of the review with quality of the product, review content, and posting date of the review. They use 10-fold cross validation on the training set. In the evaluation process they use two measures to quantify effectiveness of the ranking model which are as following: correlation between the ranking method and the Amazon's rank and second is the Mean Average Precision (MAP), which is a very commonly used technique for evaluating ranking accuracy. As this system is finding the fake reviews by using the only two properties of the reviews but as per the future work describes in the paper more properties can be used to find out the fake reviews more accurately.

Spam reviews detection by using Temporal Pattern Discovery is proposed to observe the reviews related to the normal reviewers arrival pattern and fake reviewers arrival pattern and they observe that the normal reviewer arrival pattern is stable and uncorrelated to their rating pattern temporally. On the opposite side the spam attacks are usually bursty and either positively or negatively correlated with the rating pattern. The data set which they have taken is snapshot of a review Website on October 6, 2010. It includes 408469 reviews which are written by 343629 reviewers, which are written for 25034 stores of a Website. For each review they collect the following information like rating, postdate and whether it is a Spammer Review (SR) or not. In the evaluation process they select 53 stores each of which has more than 1000 reviews. Human evaluators make decision about the stores to be SR spam attack or not if two or more evaluators declared a store as SR spam attack than system considers the store a dishonest in its selling. Out of 53 stores 34 are

suspicious one and the remaining are normal ones. Out of 34 stores 22 stores have at least two votes for being suspicious. The recall related to the system is 75.86% which shows that the system detects most of the stores having SR spam attack. The precision related to the proposed approach is 61.11%. This proposed system is good in terms of the training of their model for finding the relation as the model is trained by using the large number of reviews contains in their dataset.

Proposed diagram



MODULE&MODULE DESCRIPTION

Module:

- Image Preprocessing
- Saliency Detection By Mssg
- Intrinsic Image Decomposition By Bsiid
- Optimization Of The Decomposition Model
- Image Postprocessing

Module Description

IMAGE PREPROCESSING

The aim of the image preprocessing is to enhance the contrast between the saliency regions and background regions. We first apply the histogram equalization method to normalize the image and then adopt the adaptive gamma correction operation (AGC) to enhance the images. In this way, the contrast of the original images can be improved after image preprocessing. The training sets, validation sets and testing sets of the database are split according to the principle of reference stereo pairs. In this module the pre-processing is a

Advantages

- High Dimensional Color Transform Algorithm
- Learning Based Algorithm
- High Accuracy.
- Better Sharpened Edges.
- Better Segmentation result and promising to achieve automated image segmentation.

common name for operations with images at the lowest level of abstraction both input and output are intensity images. These iconic images are of the same kind as the original data captured. The aim of pre-processing is an improvement of the image data that suppresses unwilling distortions or enhances some image features important for further processing, although geometric transformations of images classified among pre-processing methods here since similar techniques are used. The position in the image, but this assumption is not valid in many practical cases

SALIENCY DETECTION BY MSSG

In this module saliency of a region than comparable contrast to far-away regions. Since directly introducing spatial relationships when computing pixel-level contrast is computationally expensive, we introduce a contrast analysis method, region contrast (RC), so as to integrate spatial relationships into

region-level contrast computation. In RC sub pixel and super pixel, we first segment the input image into regions, then compute color contrast at the region level, and finally define saliency for each region as the weighted sum of the region's contrasts to all other regions in the image. The weights are set according to the spatial distances with farther regions being assigned smaller weights. This section details the saliency detection. By investigating sample images from saliency datasets, we determine that all the saliency images generally consist of two parts: background and region.

The background region contains no strong structure information. Meanwhile, the saliency region can be regarded as the saliency region, which corresponds to the background; so we utilized the saliency detection method to approximately locate saliency. As discussed in the DoG response can reflect the local image structure at the current scale and is hardly affected by the uneven illumination. By taking full advantage of the characteristics of the image, we apply the DoG to locate the saliency region. Among the DoG image, the saliency intensity is larger than the background; so we utilize it to extract the saliency region. The DoG operator is defined as the difference between the two Gaussian responses with different standard deviation.

INTRINSIC IMAGE DECOMPOSITION BY BSIID

This section aims to decompose the enhanced Stereoscopic image into shading and reflectance layers. The reflectance layer is employed for the next accurate saliency detection. This decomposition process can not only remove the effect of the uneven illumination but also enhance the stereoscopic contrast. The shading layer calculation is key to the decomposition process. Different from the traditional methods that simultaneously calculate the shading and reflectance layer, we propose the BSIID method that only estimates the shading with some constraints. The reflectance can be functioned as $R = I/S$, where the division is element-wise. This method can not only shrink the solution space but also reduce the computational cost to reach the desired result. The first term is an L2-norm fidelity term that ensures the similarity between the shading

layer and the original image in the background region. The second term is an L1-norm regularization term that guarantees the smoothness constrain of the shading image.

OPTIMIZATION OF THE DECOMPOSITION MODEL

• EXACT SOLUTION

Problem is difficult to solve directly because of its non-convex properties. Therefore, we utilize the alternating direction minimization (ADM) technique to transform the energy function into a series of convex optimization problems. First, the two auxiliary variables T1 and T2 are introduced where F and F⁻¹ denote the discrete Fourier transform and inverse Fourier transform, respectively, and ∇ are the Toeplitz matrices from the discrete gradient operators with forward difference. Directly calculating the matrices $F \nabla T \nabla$ is computationally expensive.

• SPEED-UP SOLUTION

To satisfy the real-time requirement of saliency detection, we proposed two speed-up strategies to reduce the time cost. The details are presented as follows: Strategy I: Approximate Solver Since the L1-norm in the regularization term is non-convex, solving Problem is difficult. To mitigate this problem, we obtain the approximation of L1-no

IMAGE POSTPROCESSING

After the saliency image decomposition, the reflectance image still contains noise or other background texture information, which will cause many inspection errors. Thus, we need to process the reflectance image to reduce the influence of noise and background texture and then improve the robustness to localization error. To effectively solve this issue, we applied the guided filter in the project. According to the experimental results, we set the local radius $\gamma = 3$ and $\epsilon = 0.1$, which can achieve the best result. The prediction accuracy depends on the value of the saliency threshold. To identify a suitable saliency threshold for the realization of high accuracy, we analyze the relationship between the saliency threshold and the prediction performance, where the saliency threshold is set to 0, 0.10, 0.25, 0.5 and 0.75..

RESULT AND SCREENSHOTS

In this section, we first introduce four image datasets, which are typical saliency datasets with uneven illumination. Then we adopt three metrics to evaluate the performance of the surface saliency inspection methods. Three types of distortions, namely, JPEG, WN and GB, with four levels are applied to the left and right views of the reference stereo pairs in the database symmetrically and asymmetrically. Last, we compare our proposed method with some state-of-the-art methods and illustrate in detail how these methods perform. In this section, we compare our approach with 5 recent models for salient region detection FT, GC,

MR, GS, wCtr. As baseline, we also show the results achieved by 2 bottom-up saliency models Itti, GB. We use either the pre-computed saliency maps or codes provided by other authors to generate their saliency maps. Notice that several other recent works also present excellent results. However, they use part of the datasets for supervised training. The models based on deep neural networks rely on very large datasets for deep learning. Therefore, these methods are not compared in this paper due to fairness consideration.

The PR curves are shown in and the F-measure scores are given in Table I.

Table 1 F-Measures Of The Benchmarking Methods On Two Datasets

Data Set	FT	GC	wCtr	GS	MR	Our
SRCC	0.7097	0.8037	0.8468	0.8200	0.8406	0.8627
PASCAL-S	0.4154	0.5824	0.6379	0.6169	0.6425	0.6846

CONCLUSION

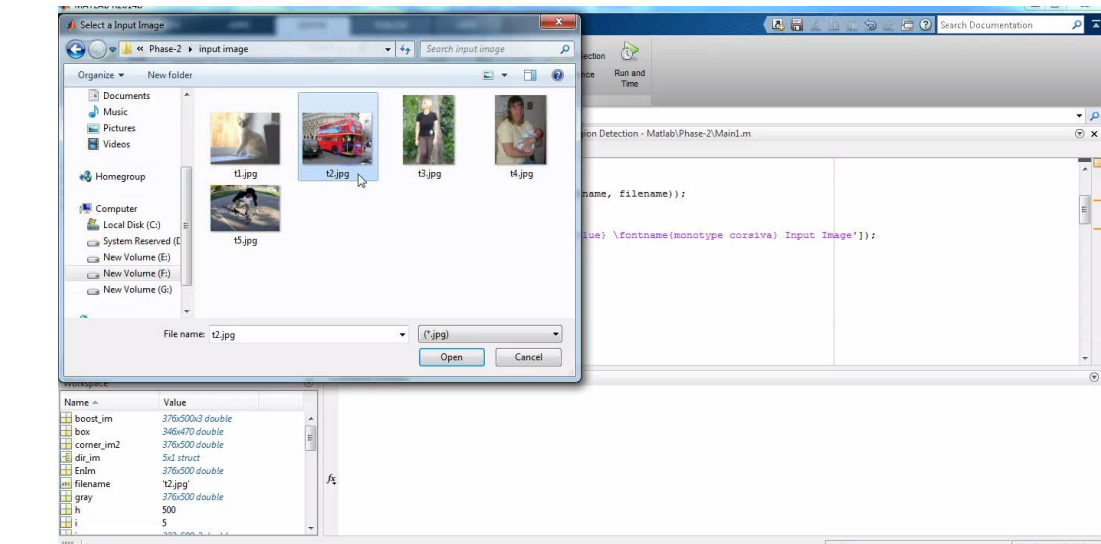
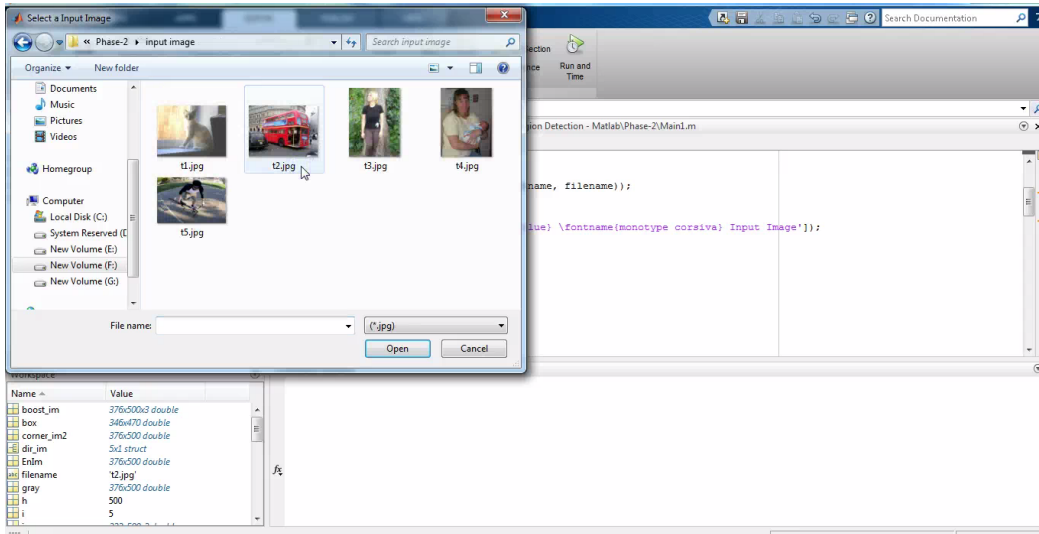
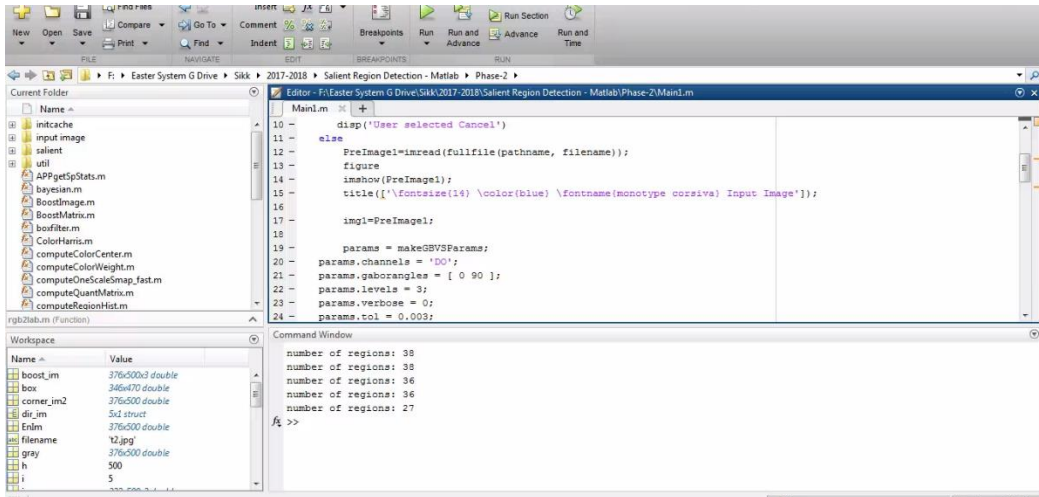
In this paper, we present two CNN-based SIQA frameworks that differ in terms of input. Both adopt a patch wise strategy and a saliency-guided local feature extraction method. Through the training procedure of local-to-global quality-aware feature aggregation, a final objective score can be obtained from the designed CNN model in an end-to-end manner. The results of a series of experiments demonstrate the superior performance of the proposed CNN model compared to most state-of-the-art FR and NR SIQA methods by significant margins, as well indicate that CNN realizes strong visual feature learning performance in the evaluation of stereo image quality.

FUTURE ENHANCEMENT

Detection of salient regions in images is useful for object based image retrieval and browsing

SAMPLE SCREEN SHOTS:

applications. This task can be done using methods based on the human visual attention model, where feature maps corresponding to color, intensity and orientation capture the corresponding salient regions. In this project, we propose a strategy for combining the salient regions from the individual feature maps based on a new Composite Saliency Indicator (CSI) which measures the contribution of each feature map to saliency. The method also carries out a dynamic weighting of individual feature maps. The experiment results indicate that this combination strategy reflects the salient regions in an image more accurately.



Current Folder: F:\Easter System G Drive\Sikk\2017-2018\Salient Region Detection - Matlab\Phase-2

```

10 - disp('User selected Cancel')
11 - else
12 -     PreImage1=imread(fullfile(pathname, filename));
13 -     figure
14 -     imshow(PreImage1);
15 -     title(['\fontsize(14) \color(blue) \fontname(monospace corsiva) Input Image']);
16 -
17 -     img1=PreImage1;
18 -
19 -     params = makeGBVSParams;
20 -     params.channels = 'DO';
21 -     params.gaborangles = [ 0 90 ];
22 -     params.levels = 3;
23 -     params.verbose = 0;
24 -     params.tol = 0.003;

```

Workspace:

Name	Value
boost_img	376x500x3 double
box	346x470 double
corner_img2	376x500 double
dir_img	5x1 struct
EnIm	376x500 double
filename	't2.jpg'
gray	376x500 double
h	500
i	5

Command Window:

```

number of regions: 38
number of regions: 38
number of regions: 36

```

Current Folder: F:\Easter System G Drive\Sikk\2017-2018\Salient Region Detection - Matlab\Phase-2

```

10 - disp('User selected Cancel')
11 - else
12 -     PreImage1=imread(fullfile(pathname, filename));
13 -     figure
14 -     imshow(PreImage1);
15 -     title(['\fontsize(14) \color(blue) \fontname(monospace corsiva) Input Image']);
16 -
17 -     img1=PreImage1;
18 -
19 -     params = makeGBVSParams;
20 -     params.channels = 'DO';
21 -     params.gaborangles = [ 0 90 ];
22 -     params.levels = 3;
23 -     params.verbose = 0;
24 -     params.tol = 0.003;

```

Workspace:

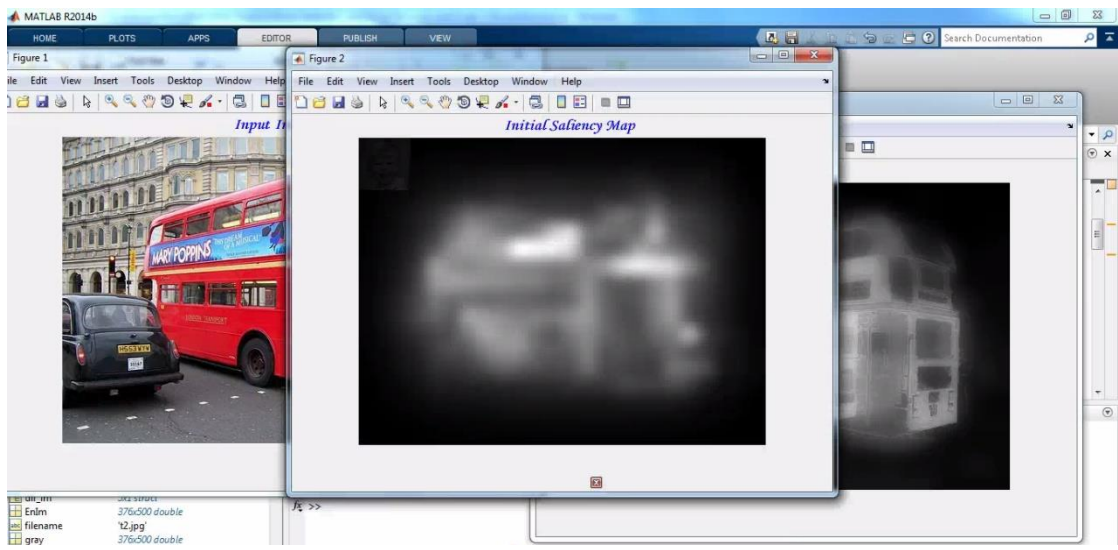
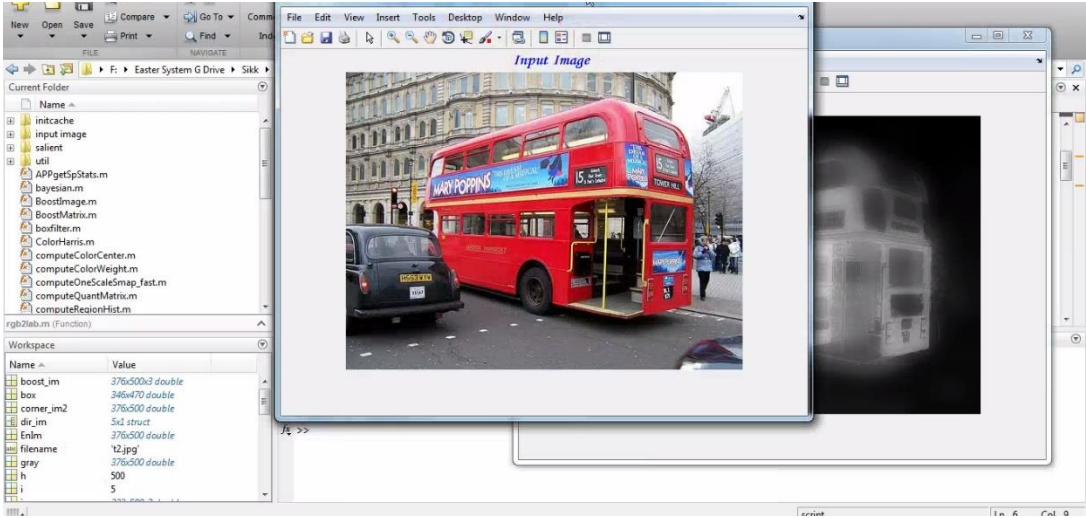
Name	Value
boost_img	376x500x3 double
box	346x470 double
corner_img2	376x500 double
dir_img	5x1 struct
EnIm	376x500 double
filename	't2.jpg'
gray	376x500 double
h	500
i	5

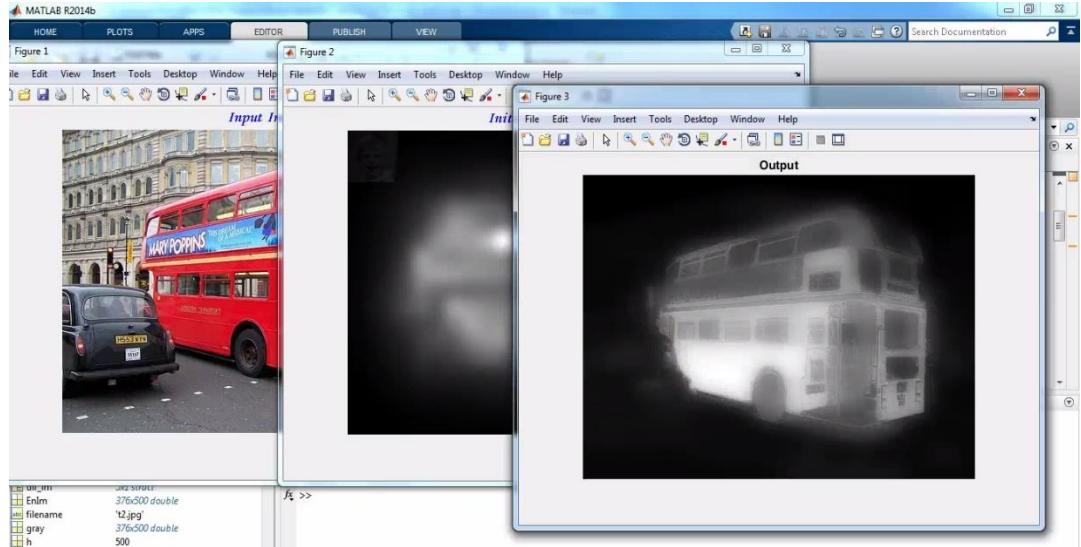
Command Window:

```

number of regions: 38
number of regions: 38
number of regions: 36
number of regions: 36
number of regions: 27

```





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