

IMAGE PROCESSING SYSTEM FOR LEATHER DEFECT IDENTIFICATION

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ABSTRACT

The defects on a leather surface may be caused by the poor material handling process during the production and manufacturing stages. It is essential to eliminate the natural variations and artificial injuries on the leather surfaces, in order to control the quality of the products and achieve customer satisfaction. To date, the visual inspection of the leather defects is performed manually by human operators. Thus, this study aims to introduce an automatic defect detection technique by employing a deep learning method. Specifically, the proposed method consists of two stages: classification and instance segmentation. The former stage distinguishes whether the piece of the leather sample contains a defective part or not, whereas the latter is to localize the precise defective location. To accomplish the tasks, the dataset is first collected under a proper laboratory environment.

The purpose of this self study is to use the image processing to perform some operations on an image, in order to get an enhanced image or to extract some useful information from it. It is a type of signal processing in which input is an image and output may be image or characteristics/features associated with that image.

INTRODUCTION

Leather defects are imperfections occurring in the grain surface or surface of the hide or skin, which ultimately result in the unsightly appearance and weakness of the resultant leather. Defects that appear in the finished leather are the result of damage to the raw skin, poor storage or improper processing during manufacturing. Dye transfer from leather is the result of insufficient colour fixation. For example, it can be caused by excessively concentrated dyes or insufficient washing after the dyeing stage. If the colour concentration is too high, bronzing can also occur.

In the liming process, excessive contact with fresh air can cause liming stains on the grain side of the leather. Irregular blood stains are caused by lack of drainage of the salt liquor after the preservation of the hides by salting. Colour defects can occur if the applied colouration does not cover uniformly, especially on already damaged areas. Fat marks on leather occurs where leathers have a high fat content, such as sheep leather. Tannin stains are caused by uneven concentration of the tanning agent in the leather production. Basically leather is the skin of animals after it is treated to many processes that refine it and beautify it. Below are some examples, which are typical in practice and make the leather partially unusable and cause higher waste.

Many of the skin damages are not detectable on the rawhide. They only become apparent after the first processing stages in the tannery.

Barbed wire scars and scratches:- Animals can incur a whole range of scars and scratches caused by contact with sharp objects, such as thorns or branches. The number of scars, their length, width and spread across the skin are all factors that determine the extent of the use of such leather.

Pitchfork injuries:- Pitchfork injuries, , are caused by prodding the animals with sharp objects. They are also known as goad marks.

Warts:- Cattle get warts, just like humans. warts deteriorate the skin quality.

Horn scratches and blows:- Injuries from horn occur during bullfight and leave significant scar .

Surgical scar:- Animals can also have surgical scars which are visible on the leather.

Neck wrinkles –

Creases:-They occur in cattle in the neck and abdomen, where the connective fibres are longer.

Insect bites:- Tick bites and stings of other insects appear as small surface damages on the skin.

Image processing which is a low level process reduce noise , does contrast , enhancement , image sharpening .Then image analysis done which is a mid level process . It does Segmentation & Classification . Then there is vision derived which is a high level process . It is making sense of an ensemble of recognized objects .

EFFECTIVE MATLAB APPLICABILITY:-

Color-Based Segmentation Using K-Means Clustering:-

K-means clustering aims to partition data into k clusters in a way that data points in the same cluster are similar and data points in the different clusters are farther apart. Similarity of two points is determined by the distance between them. The similarity measure is at the core of k-means clustering. The k-means clustering algorithm attempts to split a given anonymous data set (a set containing no information as to class identity) into a fixed number

(k) of clusters. The resulting classifier is used to classify (using k = 1) the data and thereby produce an initial randomized set of clusters. The problem is computationally difficult (NP-hard); however, efficient heuristic algorithms converge quickly to a local optimum. K-Means clustering tends to find clusters of comparable spatial extent, while the Gaussian mixture model allows clusters to have different shapes.

PROCEDURE:- First the image is read from graphic file & then RGB is converted into CIE 1976 L*a*b* & then the image is converted into single precision colors. The Clustering thus made are repeated three times to avoid local minima. After it, the K-Means Clustering based image segmentation is done in which image is labelled by Cluster Index and then masking is done for each cluster. L*a*b* provides sufficient precision for most computation task & then scaling the range of array elements & binarizing them to 2-D grey scale image to finally obtain the final image.

CASE 1:- WRINKLES ON LEATHER

PROGRAM

```
A = imread('MM4.jpg'); %Read image from
                        %graphic file

imshow(A); pause(20);

lab_A = rgb2lab(A); %Convert RGB to CIE
                  1976 L*a*b*

ab = lab_A(:,:,2:3);

ab = im2single(ab); %Convert image to
                  single precision

nColors = 3; % repeat the
            clustering 3 times to
            avoid local minima

pixel_labels =
imsegkmeans(ab,nColors,'NumAttempts',3);
            %K-means clustering
            based image
            Segmentation

imshow(pixel_labels,[]);
pause(20)
title('Image Labeled by Cluster Index');
mask1 = pixel_labels==1;% Gives pattern with
                  specified display
                  name

cluster1 = A .* uint8(mask1);
imshow(cluster1);pause(20);
title('Objects in Cluster 1');

mask2 = pixel_labels==2; % Gives pattern
                  with specified
                  display name

cluster2 = A .* uint8(mask2);
imshow(cluster2);
pause(20);
title('Objects in Cluster 2');
mask3 = pixel_labels==3; % Gives pattern
```

with specified
display name

```
cluster3 = A .* uint8(mask3);
imshow(cluster3);pause(20);
title('Objects in Cluster 3');
L = lab_A(:,:,1);
L_blue = L .* double(mask3); % Provides
                        Sufficient
                        precision for most
                        computation task

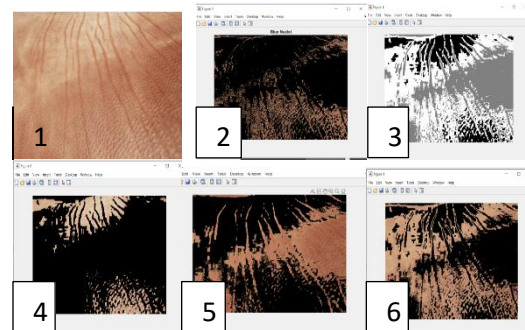
L_blue = rescale(L_blue); % Scale range
                        of array
                        elements

idx_light_blue =
imbinarize(nonzeros(L_blue)); % Binarize to
                        2-D gray scale
                        image

blue_idx = find(mask3);
mask_dark_blue = mask3;
mask_dark_blue(blue_idx(idx_light_blue)) = 0;
blue_nuclei = A .* uint8(mask_dark_blue);

imshow(blue_nuclei);pause(20);
title('Blue Nuclei')
```

RESULTS:-



(1) IMAGE SHOWING WRINKLES ON LEATHER (2) IMAGE SHOWING PIXEL LABELS ON CLUSTER INDEX (3&4&5) IMAGE SHOWING CLUSTER 1, CLUSTER 2 & CLUSTER 3 (6) FINAL IMAGE AFTER MASK AND BINARY OPERATION

MATLAB ENABLING TEXTURE SEGMENTATION USING GABOR FILTERS:-

Texture segmentation is done to identify regions based on their texture. The goal is to segment the different or irregular pattern from the background or required pattern. The segmentation is visually obvious because of the difference in texture between the two. K-means iterative clustering algorithm applied to image texture segmentation:-The K-means algorithm is a robust, unsupervised clustering algorithm applied here to the problem of image texture segmentation. The K-Means algorithm involves two steps. First, K-means is applied. Second, the K-means class assignments are used to estimate parameters required to improve the solution. Gabor filters and co-occurrence probabilities are used as texture features.

PROCEDURE:- First the image is read from graph & then Returns a row vector whose elements are the lengths of the

corresponding dimensions of image . Then resizing image & converting RGB image to gray scale . An RGB image is a three dimensional byte array that explicitly stores a color value for each pixel . RGB images are made up of width , height & three channels of color information . Scanned photographs are commonly stored as RGB images .Then Start image segmentation row wise & columnwise . After it Sample wavelength in increasing powers of two starting from 4/sqrt upto the hypotenuse length of the input image & then Rounding it towards the negative infinity of the value . Then using Gabor filters which Provide gabor magnitude in terms of wavelength & energy . Then choosing sigma wavelength which is matched to gabor filter, extracting each feature . A smoothing term K equal to 3 is introduced whose value determining how much smoothing is applied to the gabor magnitude response . Then creating a mesh plot which is a three dimensional surface having solid edge colors & no face colors & then extracting high level feature from data of concatenate arrays .Then applying element wise operation to two arrays with implicit expansion enabled and does principle component analysis of raw data . By K- Means clustering , partitioning n-by-p matrix into K clusters obtaining n-by-1 vector containing cluster indices of each observation .After it piecewise distribution segmentation over copies of arrays is done . After doing K-Means clustering based image segmentation and overlaying label matrix region on 2-d image & obtaining labeled image with additional pixel information.

CASE 2- PITCH FORK INJURIES

PROGRAM

```
A = imread('PF4.png'); % Read image from
                        % graphic file
size(A)
A = imresize(A,0.75); % Returns image that
                     % is scale times the prior image

A=A(1:100,1:100, 1:3);
Agray = rgb2gray(A); % Convert RGB image
                    % or color map to gray scale

figure(1)
imshow(A);pause(20);
imageSize = size(A);
figure(2)
imshow(Agray);pause(20);
numRows = imageSize(1); % Read rows of
                        % binary table column
numCols = imageSize(2); % Read columns of
                        % binary table column

figure
wavelengthMin = 4/sqrt(2); % Sample wavelength
                          % in increasing
                          % powers of two
                          % starting from
                          % 4/sqrt(2)
wavelengthMax = hypot(numRows,numCols);
                          % Sample
                          % wavelength in
                          % increasing
                          % powers of two
                          % starting from
                          % 4/sqrt upto
                          % hypotenuse
                          % length of the
                          % input image
```

```
n =
floor(log2(wavelengthMax/wavelengthMin));
%Round towards the
%negative infinity
%of the value
wavelength = 2.^(0:(n-2)) * wavelengthMin;
deltaTheta = 45;
orientation = 0:deltaTheta:(180-deltaTheta);

g = gabor(wavelength,orientation); % Round
%towards the negative
%infinity of the
%value
gabormag = imgaborfilt(Agray,g); % provides
%gabor magnitude
%response of the gray
%image after gabor
%filter operated upon

for i = 1:length(g)
sigma = 0.5*g(i).Wavelength;%chosen sigma
%wavelength is
%matched to gabor
%filter,extracting
%each feature

K = 3; % A smoothing
%term K is
%introduced. It's
%value determining
%how much smoothing
%is applied to the
%gabor magnitude
%response
```

```
gabormag(:,:,i) =
imgaussfilt(gabormag(:,:,i),K*sigma);
end
X = 1:numCols;
Y = 1:numRows;
[X,Y] = meshgrid(X,Y);
%Creates a mesh plot
%which is a three
%dimensional surface
%having solid edge
%colors & no face
%colors

featureSet = cat(3,gabormag,X);
%Extracting high level
%feature from data of
%concatenate arrays

featureSet = cat(3,featureSet,Y);
numPoints = numRows*numCols;
X = reshape(featureSet,numRows*numCols,[]);
X = bsxfun(@minus, X, mean(X)); %Apply
%element-wise
%operation to
%two arrays
%with
%implicit
%expansion
%enabled .
```

```

X = bsxfun(@rdivide,X,std(X)); %Apply
                                element-wise operation
                                to two arrays with implicit
                                expansion enabled.

coeff = pca(X); % Does
                principle component
                analysis of raw data

feature2DImage =
reshape(X*coeff(:,1),numRows,numCols);
figure (3)
L =imshow(feature2DImage,[]);pause(20);
kmeans(X,2,'Replicates',5); %does K
                             means clustering to
                             partition the
                             operation of
                             the n-by-p data
                             matrix X
                             into K clusters ,
                             returning an n-by-1
                             vector(ldx)
                             containing cluster
                             indices of each
                             observation

% L = reshape(L,[numRows numCols]);
figure (4)
%imshow(label2rgb(L));pause(20); %Show the
                                conversion of the
                                label matrix into
                                RGB image

Aseg1 = zeros(size(A),'like',A);
Aseg2 = zeros(size(A),'like',A);
BW = L == 2;
BW = repmat(BW,[1 1 3]); %Repeat
                           copies of arrays

Aseg1(BW) = A(BW); %Does
                   piecewise
                   distribution
                   segmentation

Aseg2(~BW) = A(~BW); %Does
                     piecewise
                     distribution
                     segmentation

figure(5)

imshowpair(Aseg1,Aseg2,'montage');pause(20);
%Display
multiple image
frames as
rectangular
montages

RGB=imread('PF4.png');
RGB=imresize(RGB,0.5)
imshow(RGB);pause(20);
L = imsegkmeans(RGB,2);pause(20); %Kmeans
                                clustering based
                                image

segmentation

B = labeloverlay(RGB,L); %Overlay
                        label matrix
                        regions on
                        2-D image

imshow(B);pause(20);
title('Labeled Image')
wavelength = 2.^(0:5) * 3;
orientation = 0:45:135;
g = gabor(wavelength,orientation);
I = rgb2gray(im2single(RGB)); %
                              Convert RGB image
                              to gray scale

gabormag = imgaborfilt(I,g); % gives
                              gabor magnitude
                              of the output of
                              the set of
                              filters on 2-d
                              image

montage(gabormag,'Size',[4 6]);pause(20);
for i = 1:length(g)
    sigma = 0.5*g(i).Wavelength; %
                                  chosen sigma
                                  wavelength is
                                  matched to gabor
                                  filter ,
                                  extracting each
                                  feature

gabormag(:, :, i) =
imgaussfilt(gabormag(:, :, i), 3*sigma);
end
montage(gabormag,'Size',[4 6]);pause(20);
%Makes image frame as
Rectangular montage
of the gabor
magnitude response

nrows = size(RGB,1);
ncols = size(RGB,2);
[X,Y] = meshgrid(1:ncols,1:nrows);
%Creates a mesh plot
Which is a three
Dimensional surface
Having solid edge
Colors & no face
Colors .

featureSet = cat(3,I,gabormag,X,Y);
%Extracting high level
Feature from data

L2 =
imsegkmeans(featureSet,2,'NormalizeInput',true);
C = labeloverlay(RGB,L2); % Overlay label matrix
                        Regions on 2-D image

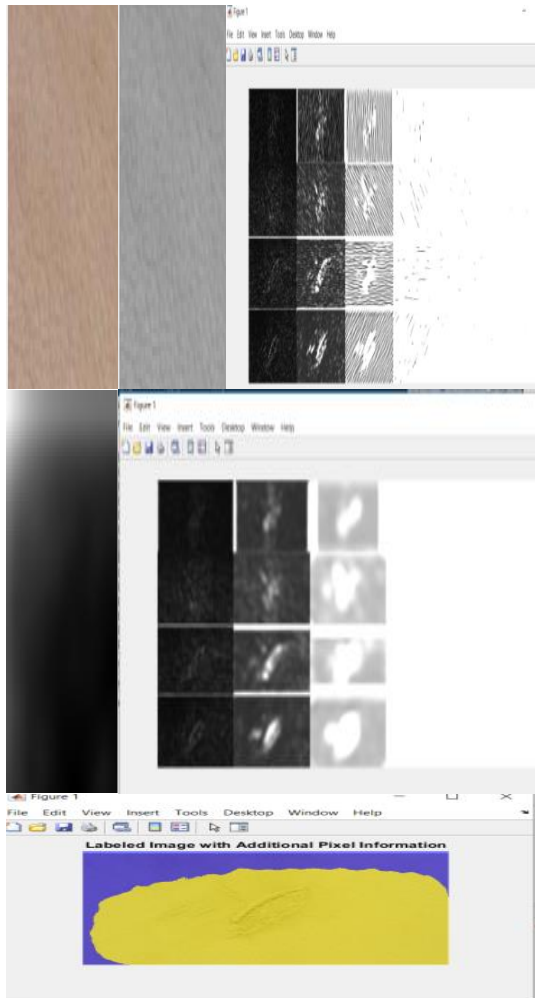
regions on 2-D
image

imshow(C);pause(5)

```

```
title('Labeled Image with Additional Pixel Information')
```

RESULTS:-



1) IMAGE OF PITCHFORK INJURIES ON LEATHER 2) GREY PICTURE OF THE IMAGE 3) K-MEANS OUTPUT OF THE IMAGE 4) MONTAGE OF GABOR MAGNITUDE 5) MONTAGE OF GABOR MAGNITUDE AFTER DETERMINING SIGMA 6) IMAGE WITH ADDITIONAL PIXEL INFORMATION

RESULT & DISCUSSION:-

In first attempt using the $L^*a^*b^*$ color space & K-means clustering which use Euclidean distance metric to segment colors & converting into clusters obtaining blue object in one of them. After extracting brightness level of pixel then thresholding with global threshold getting dark blue cell nucleus end result. The Wrinkles & Horn scratches on leather which are major defect are clearly identified and marked using this method but for the small defects like Pitch fork injuries, this method is not found practically generous. For this, using Gabor filter followed by K-means clustering & then choosing sigma & determining Gabor magnitude & then obtaining additional pixel information after segmentation

proved effective. Thus second method is more applicable for fine defects.

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