




**Medical Image Analysis**  
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**Lecture 07**

**Contrast Enhancement**

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### Overview-Week 2



- Contrast enhancement
- Histogram Equalization/Matching
- Edge enhancement- Laplacian
- Edge enhancement - Derivatives
- Linear Filtering
- Median Filtering
- Diffusion Filtering
- Bilateral Filtering
- Bayesian Image Restoration

Hello and welcome back. So, in this video we are going to look at contrast enhancement technique

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### Image Enhancement

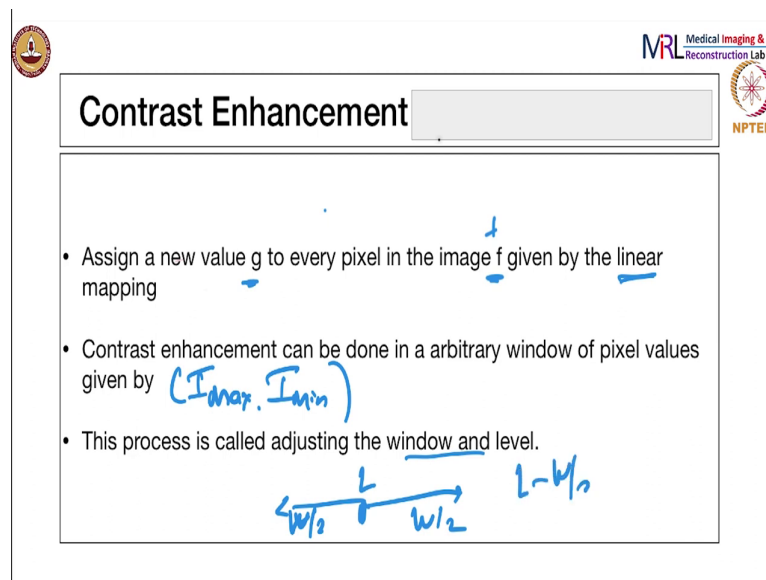
- One of the simple operations on medical images is image enhancement
- Done to improve identification of features in an image
- Typically, this is contrast enhancement and noise reduction and emphasis of difference between different structures in the image

So, contrast enhancement is basically a form of innate image enhancement and it is one of the most simple operations that is used in medical image analysis. And this is primarily done to

improve the identification of a particular feature in an image, basically to make something appear more bright or more distinctive from the background. So, you call it contrast enhancement because contrast if you want to define contrast it is basically how different residues from the background.

And it is also of contrast enhancement noise enhancement noise reduction together and with in combination with like you mentioned earlier emphasis of different structures on the image. So, we like we also like we had also like had also mentioned earlier we will just look at the formulation of these techniques and I will show them through the results in a separate class.

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The slide is titled "Contrast Enhancement" and features a logo for "MIRL Medical Imaging & Reconstruction Lab" and "NPTEL" in the top right corner. The main content consists of three bullet points with handwritten annotations:

- Assign a new value  $g$  to every pixel in the image  $f$  given by the linear mapping
- Contrast enhancement can be done in an arbitrary window of pixel values given by  $(I_{max}, I_{min})$
- This process is called adjusting the window and level.

Below the text, a diagram illustrates a window of pixel values. It shows a horizontal line with a central point labeled  $L$ . To the left of  $L$  is a point labeled  $L - w/2$ , and to the right is a point labeled  $L + w/2$ . The distance between  $L$  and  $L - w/2$  is labeled  $w/2$ , and the distance between  $L$  and  $L + w/2$  is also labeled  $w/2$ .

So, the idea behind this method is to have for every pixel in an image if you have to find a new value  $g$  through a linear mapping. So, the linear transformation is a linear transformation I will say linear mapping and typically what you do is you want to map all the pixel values into a range between  $I_{max}$  and  $I_{min}$ .

So, this  $I_{max}$  and  $I_{min}$  chosen by the user typically a clinician or radiologist who in his experience knows the best  $I_{max}$  and  $I_{min}$  this process is also called adjusting the window and level primarily because you can choose a level  $L$  and from there you can go  $w/2$  to the right  $w/2$  to the left. So, which means you will be mapping between  $L - w/2$  to  $L + w/2$ .

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$$g(f) = (f - f_{\min}) \left\{ \frac{I_{\max} - I_{\min}}{f_{\max} - f_{\min}} \right\} + I_{\min}$$

Transfer function

$f \in [f_{\min}, f_{\max}]$   
 $g \in [I_{\min}, I_{\max}]$

$f_{\min} = W_{\min}$   
 $f_{\max} = W_{\max}$

CT Images  $\rightarrow$  0-255  
 12 bits  $\rightarrow$  0-4095  $\rightarrow$  8-bit



So, let us see how it is actually done. So, the formula write this down is  $g(f)$  that is the transformed intensity is  $g(f) = (f - f_{\min}) \frac{I_{\max} - I_{\min}}{f_{\max} - f_{\min}} + I_{\min}$ . So, this  $g$  is called the transfer function. So, you can actually see that if you plug in the appropriate values you will get the desired result.

So, for instance when  $f$  you can try to see when  $f = f_{\min}$  and  $f = f_{\max}$  then this the  $g$  then goes to  $I_{\min}$  and in this case  $g$  goes to  $I_{\max}$  you can just plug the values in here and then verify that. Now, here what you are doing is taking every pixel value in  $f$  of course we are identifying the minimum and the maximum values in  $f$  and denote them as  $f_{\max}$  and  $f_{\min}$  and then you are trying to do a linear mapping.

So, think of this expression here as estimating the slope. So, doing a linear fit is estimating a slope here and then you can just use a straight line fit to go from  $f$  to  $g$ . So, this process is often done with CT images CT images are typically 12 bit images 12 bits. So, which means that they go from having 4096 values like 0 to 4095 this might be a very wide dynamic range wherein you might not be able to see very good contrast between let us say lesion or tumor and background small tumor and background.

So, you might want to map this into a slightly smaller range. Maybe you want to go to 0 to 255. So, this is an 8 bit range where maybe you get better contrast. So, this is one way of doing the transformation now once again knowing that  $f_{\min}$  and  $f_{\max}$  do not necessarily need to come from the maximum and minimum values in the image.

So, the maximum and minimum values in a given image might not be meaningful. So, you are also allowed to set your own instead of  $f_{\min}$  and  $f_{\max}$  instead of  $f_{\min}$  you can go to  $w_{\min}$  and instead of  $f_{\max}$  you can go to  $w_{\max}$  and because in the original image of anything less than  $w_{\min}$  might not be meaningful it could be some artifact some error anything greater than  $w_{\max}$  might not be meaningful also.

So, when you do this kind of mapping then you have to make sure you do a thresholding operation also find out all the pixel values less than  $w_{\min}$  and then turn them into  $I_{\min}$  which is what is here then find out all the pixel values greater than  $w_{\max}$  and find out and set it to  $I_{\max}$  but of course there is some conditions here which is basically you want to make sure that  $w_{\min}$  and  $w_{\max}$  are within the range  $I_{\min}$  and  $I_{\max}$  that is one other condition from which then you can map easily.

So, again there are variations of this so this is the simplest version and often done. So, the window and level are often done on CT images. So, this is one of the simplest contrast enhancements and you will end up doing this often. So, for instance even if you are let us say doing deep learning and you have a bunch of CT volumes to process. So, it will be nice before you provide these CT images as inputs to a neural network a CNN for instance in order to have an outcome you might want to do the window and level on this image.

So, one of the strategies that people often use is to have an ensemble of networks wherein each network has an input adjusted to a different window and level. So, that it highlights different anatomies or structures of interest. So, this is a commonly used trick. So, that is all we have for contrast enhancement. So, in the next video we will look at other techniques. Thank you.