## AIM:

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## THEORY:

1. **Region Growing**

The objective of segmentation is to partition an image into regions. One way to do is to by finding the regions directly. When we are segmenting based on regions, we are essentially finding similarities. In edge detection, we find differences. The basic approach is to start with a set of “seed” points, and from these grow regions by appending to each seed those neighboring pixels that have predefined properties similar to the seed. Properties can be: ranges of intensity, color, texture, shape, model etc.

Region growth should stop when no more pixels satisfy the criteria for inclusion in that region. Criteria such as intensity values, texture, and color are local in nature and do not take into account the “history” of region growth. Additional criteria that can increase the power of a region-growing algorithm utilize the concept of size,likeness between a candidate pixel and the pixels grown so far (such as a comparison of the intensity of a candidate and the average intensity of the grown region), and the shape of the region being grown.

Let:

*f(x,y): input image*

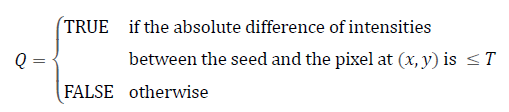
*S(x,y): seed array containing 1’s at the locations of seed points and 0’s elsewhere*

*Q: a predicate to be applied at each location (x, y)*

*Arrays f and S are assumed to be of the same size*

1. Find all connected components in *S*(*x*, *y*) and reduce each connected component to one pixel; label all such pixels found as 1. All other pixels in *S* are labeled 0.
2. Form an image *fQ* such that, at each point (*x*, *y*), *fQ (x,y) = 1* if the input image satisfies a given predicate, *Q*, at those coordinates, *fQ (x,y) = 0* and otherwise.

The predicate can use a threshold *T* for the same:



1. Let *g* be an image formed by appending to each seed point in *S* all the 1-valued points in *fQ* that are L-connected to that seed point. (L could be 4, 8 or m)
2. Label each connected component in *g* with a different region label (e.g.,integers or letters). This is the segmented image obtained by region growing
3. **Region Splitting and Merging**

An alternative is to subdivide an image initially into a set of disjoint regions and then merge and/or split the regions in an attempt to satisfy the conditions of segmentation stated in region growing

Let *R* represent the entire image region and select a predicate *Q*. One approach for segmenting *R* is to subdivide it successively into smaller and smaller quadrant regions so that, for any region. We start with the entire region, *R*. If we divide the image into quadrants. If *Q* is FALSE for any quadrant, we subdivide that quadrant into sub-quadrants, and so on. This splitting technique has a convenient representation in the form of so-called *quadtrees*; that is, trees in which each node has exactly four descendants

See a partitioned image and its corresponding quadtree:



1. Split into four disjoint quadrants any region Ri for which Q(Ri) = FALSE

2. When no further splitting is possible, merge any adjacent regions Rj and Rk and for which 

3. Stop when no further merging is possible.

**Lab Assignments to complete in this session**

**Problem Statement:** Develop a Python program utilizing the OpenCV library to manipulate images from the Fashion MNIST digits dataset. The program should address the following tasks:

1. Importing libraries
2. Read random image(s) from the MNIST fashion dataset.
3. **Dataset Link:** Digit MNIST Dataset
4. Extract a region of the input image depending on a start position and a stop condition.
5. The input should be a single channel 8 bits image and the seed a pixel position (x, y).
6. The threshold corresponds to the difference between outside pixel intensity and mean intensity of region.
7. In case no new pixel is found, the growing stops.
8. Output a single channel 8 bits binary (0 or 255) image. Extracted region is highlighted in white – REGION GROWING
9. For REGION MERGING AND SPLITTING, divide the whole image into regions if the threshold predicate does not match
10. When no further division is possible, merge regions if their unions satisfy the predicate.
11. Provide outputs of both #8 and #10

The solution to the operations performed must be produced by scratch coding without the use of built in OpenCV methods.

Code:

import tensorflow as tf

import numpy as np

import cv2

import random

from google.colab.patches import cv2\_imshow

# Function to read random image from the MNIST fashion dataset

def read\_random\_image(dataset):

    # Select random index

    idx = random.randint(0, len(dataset) - 1)

    # Convert image to numpy array

    image = np.array(dataset[idx], dtype=np.uint8)

    return image

# Function to perform region growing

def region\_growing(image, seed, threshold):

    height, width = image.shape

    visited = np.zeros\_like(image)

    region = np.zeros\_like(image)

    stack = [seed]

    while stack:

        x, y = stack.pop()

        if visited[x, y] == 1:

            continue

        visited[x, y] = 1

        region[x, y] = 255

        neighbors = [

            (x + 1, y),

            (x - 1, y),

            (x, y + 1),

            (x, y - 1)

        ]

        for nx, ny in neighbors:

            if 0 <= nx < height and 0 <= ny < width:

                diff = abs(int(image[x, y]) - int(image[nx, ny]))

                if diff <= threshold:

                    stack.append((nx, ny))

    return region

# Function to divide image into regions based on threshold predicate

def divide\_regions(image, threshold):

    height, width = image.shape

    regions = []

    visited = np.zeros\_like(image)

    for i in range(height):

        for j in range(width):

            if visited[i, j] == 0:

                seed = (i, j)

                region = region\_growing(image, seed, threshold)

                regions.append(region)

                visited |= (region > 0)

    return regions

# Function to merge regions if their unions satisfy the predicate

def merge\_regions(regions, threshold):

    merged\_regions = []

    num\_regions = len(regions)

    merged = [False] \* num\_regions

    for i in range(num\_regions):

        if merged[i]:

            continue

        merged\_region = regions[i]

        for j in range(i + 1, num\_regions):

            if not merged[j]:

                union = np.logical\_or(merged\_region, regions[j])

                if np.count\_nonzero(union) <= threshold:

                    merged\_region = np.logical\_or(merged\_region, regions[j])

                    merged[j] = True

        merged\_regions.append(merged\_region)

    return merged\_regions

# Main function

def main():

    # Load Fashion MNIST dataset

    fashion\_mnist = tf.keras.datasets.fashion\_mnist

    (train\_images, \_), (\_, \_) = fashion\_mnist.load\_data()

    # Read random image from the dataset

    image = read\_random\_image(train\_images)

    # Display original image

    print("Original Image")

    cv2\_imshow(cv2.resize(image, (400, 400)))

    # Define threshold for region growing

    threshold = 20

    # Perform region growing

    grown\_region = region\_growing(image, (10, 10), threshold)

    # Display region grown image

    print("Grown Region")

    cv2\_imshow(cv2.resize(grown\_region.astype(np.uint8), (400, 400)))

    # Divide the image into regions based on threshold predicate

    regions = divide\_regions(image, threshold)

    # Merge regions if their unions satisfy the predicate

    merged\_regions = merge\_regions(regions, threshold)

    # Display merged regions

    for i, merged\_region in enumerate(merged\_regions):

        cv2\_imshow(cv2.resize(merged\_region.astype(np.uint8), (400, 400)))

        print(f"Merged Region {i+1}")

if \_\_name\_\_ == "\_\_main\_\_":

    main()

Output:



