

The background of the slide features a close-up, artistic shot of a liquid, likely oil or honey, being poured. The liquid flows from the top right towards the bottom left, creating smooth, undulating waves. The color transitions from a pale, milky white on the left to a vibrant, warm orange on the right. The lighting is soft and diffused, highlighting the glossy texture of the liquid's surface.

Motion Localization

Compare Frames

Compare Frames

Compare image matrices and mark the pixel location that have two significantly different values



Result



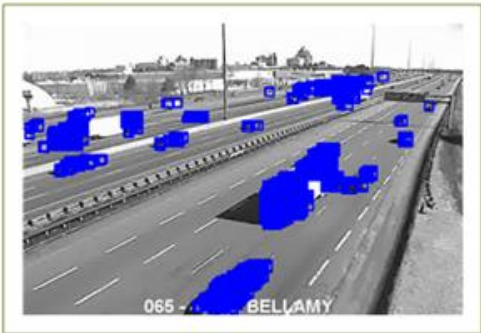
Check Diagonals

Check Diagonals

Set coordinates to bound the clusters



Result



Cluster Find

Cluster Find

Checks for overlapping rectangles and chooses the largest



Result



Pseudocode - Pixel Compare

```

Input Frame_Library //Images in exercise folder
Input pixel_rows    //Pixels in x-axis in image
Input pixel_columns //Pixels in y-axis
Input threshold     //Amount of pixel difference to determine change
Output Frame2
// width = W | height = H | kernel Diameter = D | W*H = n

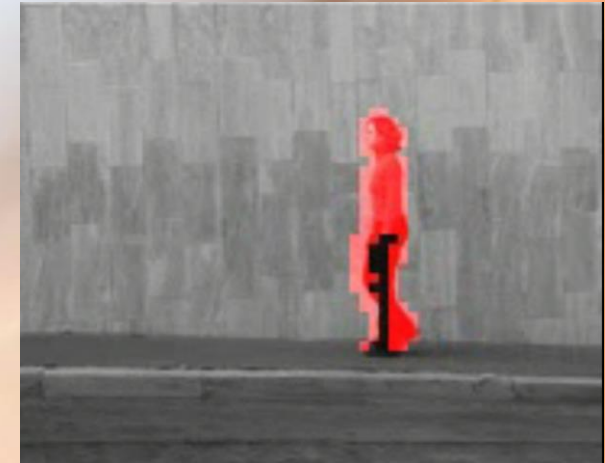
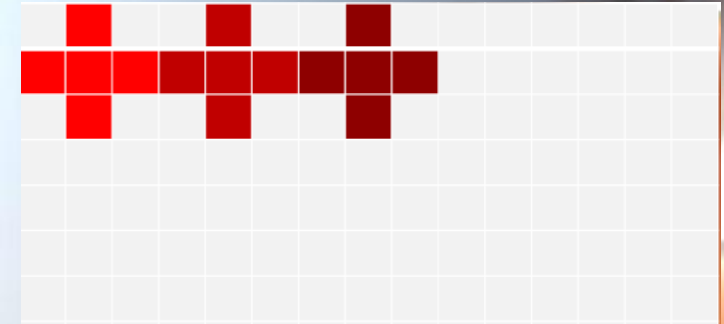
Frame1 = frame_library[f]           // 1
Frame2 = frame_library[f+1]         // 1
  For i ← 0 in pixel_rows ; i += D   // W*1/D
    For j ← 0 in pixel_columns; j += D // W*H*(1/D)*(1/D)
      mean_condition ← (absolute(frame2(i - 1, j) - frame1(i - 1, j) +
        absolute( frame2(i, j + 1) - frame1(i, j + 1)) +
        absolute(frame2 ( i , j - 1) - frame1(i, j-1)) +
        absolute(frame2( i + 1 , j ) - frame1( i + 1 , j )) +
        absolute(frame2( i , j ) - frame1( i , j)))/5
                                     // W*H*(1/D)*(1/D)*10
                                     // (10 assignments)

      if mean_condition > threshold   // W*H*(1/D)*(1/D)
        for x ← 0 in kernel_rows     // W*H*(1/D)*(1/D)*D
          for y ← 0 in kernel_columns // W*H*(1/D)*(1/D)*D*D
            frame2( i+x , j+y) = 255

    // ( L-1)*W*H*(1/D)*(1/D)*D*D
    // (L-1)

Return Frame2

```



```

// 3 + W*(1/D) + 12*W*H*(1/D)*(1/D)
//+ W*H*(1/D)*(1/D)*R + 2* W*H*(1/D)*(1/D)*D*D
//Taking the biggest term:
//2* W*H*(1/D)*(1/D)*D*D = 2* W*H -2*W*H
//2* W*H = 2*n => O(n)

```

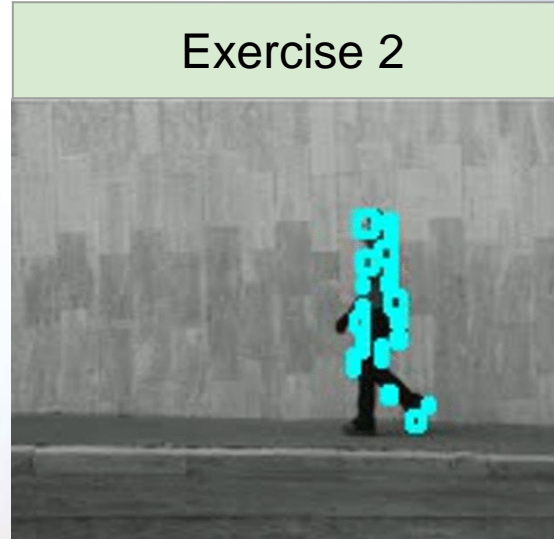
Pseudocode - Check Diagonals

```
Input Frame_Changes           // Image with significant changes highlighted in red
Input threshold                // Area of pixels to ignore
Input total_rows , total_columns // width and height of the image
Output rectangle_list         //list of rectangles
//width = W | height = H | W*H = n

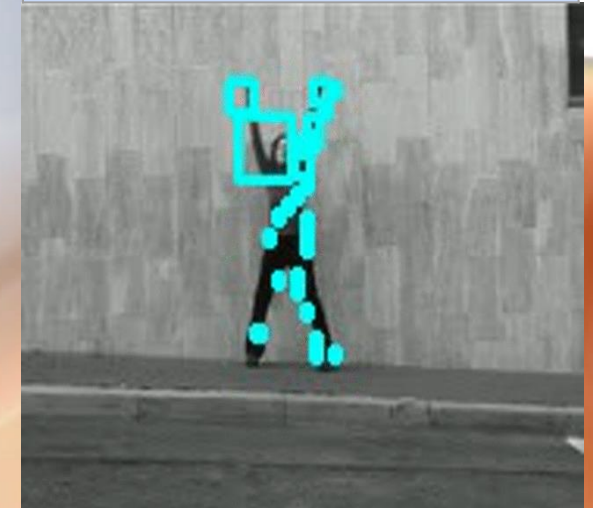
CheckDiagonals(Frame_Changes ,total_rows ,total_column, threshold):

rectangle_list = []           //1
For i <- 0 in total_rows      // H
    For j <- 0 in total_columns //H*W
        width = 0             // H*W
        height = 0            // H*W
        no_repeat = True      // H*W
        coord_stack = ([ j , i ]) // H*W
        while (len(coord_stack) > 0 and x < total_columns and y < total_rows): // H*W (W+H)
            x,y = coord_stack.pop() // H*W (W+H)
            if frame[x+1,y,2] != 255 and frame[x,y+1,2] !=255: // H*W*(2*(W+H))
                continue //H*W
            if frame[x+1,y,2] ==255 and no_repeat: //2*(H*W)*(W+H)
                no_repeat = False //W*(H*W)
                height+=1 //W*(H*W)
                coord_stack.push(x+1,y) //W*(H*W)
            else: //W*(H*W)
                no_repeat = True //H*(H*W)
                width +=1 //H*(H*W)
                coord_stack.push(x,y+1) //H*(H*W)
            if (width + height) > threshold // H*W
                rectangle = j, i , j+ width, i + height // H*W/threshold
                rectangle_list.append(rectangle) // H*W/threshold
return rectangle_list //1
```

Exercise 2



Exercise 1



```
//2 + H + 7*H*W + 2*H*W/threshold + 9*(H*W)*(H+W)
//Taking the biggest term
//9*n*(H+W) → 9*n*log(n) → O(n*log(n))
```


Pseudocode - Cluster Bounding

```
Input Rectangle_List      //list of rectangle points
Input Overlap_Threshold  // Overlap allowance (area allowed)
Output Boxes             // Remaining list of rectangle points

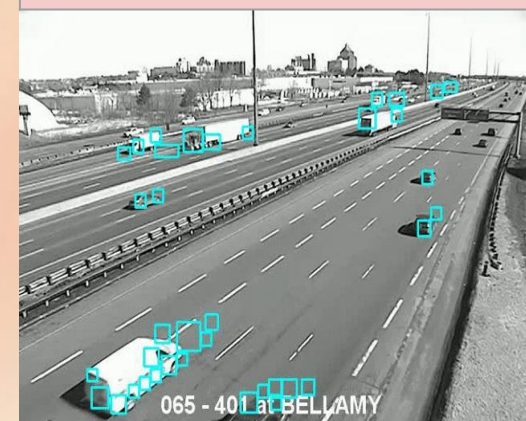
Cluster_find(Rectangle_list , Overlap_threshold):
    if len(Rectangle_list) = 0                                //1
        return Rectangle_list                                // 0
    x1 = rectangle_List[for each x,0], x2 = rectangle_List[for each x,1] , y1 = rectangle_List[ for each y, 2] ,
    y2 = rectangle_List[ foreach,3]                            //4n
    Index = QuickSortByIndex(y2)                               //n^2 (quick sort)
    area = (x2-x1) * (y2-y1)                                    //1
    while len(index) > 0:                                       //n+1
        last = len(index) -1                                    //n
        i=index[last]                                           //n
        pick.append(i)                                           //n
        big_x1 = Min(x1[i], x1[index[for each last]])           //n+(n-1)+ ...+1
        big_x2 = Max(x2[i], x2[index[for each last]])           // n+(n-1)+ ...+1
        big_y1 = Min(y1[i], y1[index[for each last]])           // n+(n-1)+ ...+1
        big_y2 = Max(y2[i], y2[index[for each last]])           // n+(n-1)+ ...+1
        width = Max(0,big_x2 - big_x1)                           //n
        height = Max(0,big_y2 - big_y1)                           // n
        overlap = (w*h)/area[index[for each last]]              //n + (n-1)+ ...+1
        index = delete(index, concatenate([last], where ( overlap > Overlap_Threshold )) //n +(n-1) + ... +1
    return  rectangle_list[pick]                                //1

//7 + n^2 + 7*n + 5*n(n-1)/2 → 5*n(n-1)/2 → O(n^2)
```

Exercise 3



Exercise 4

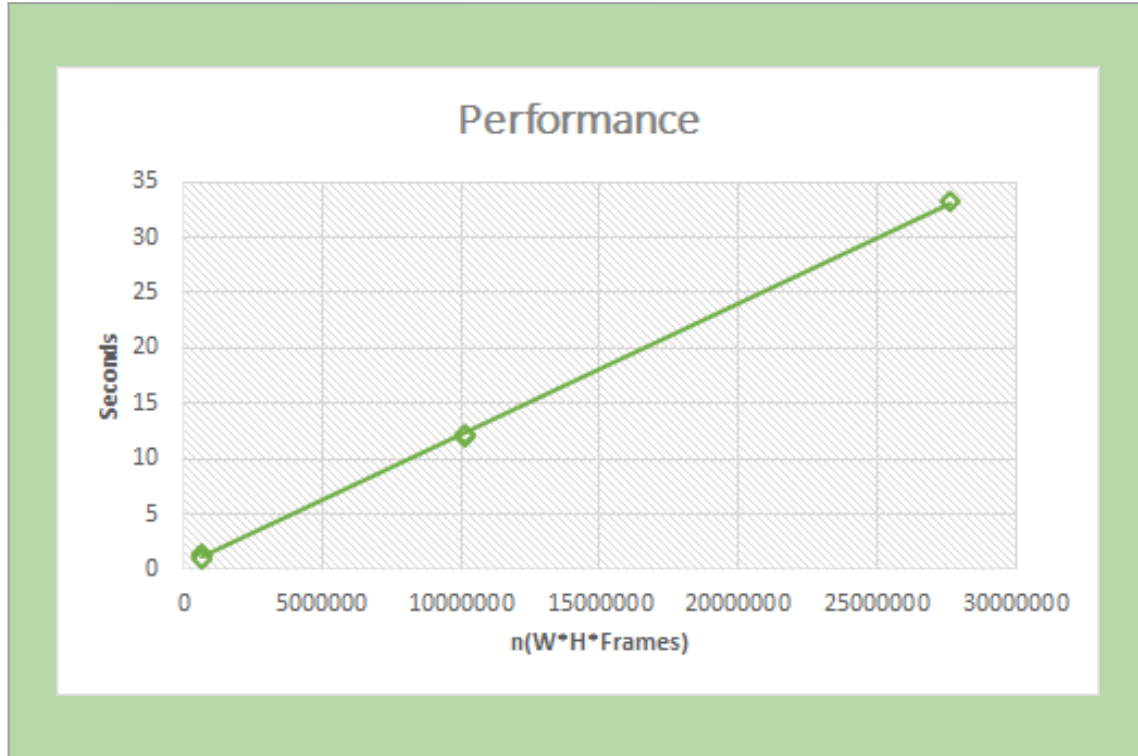


Timing Breakdown - After Grouping

Exercise	Resolution	Frames	Time	FPS
Exercise 1	180 x 144	25	1.42s	17.6
Exercise 2	180 x 144	25	1.35s	23.7
Exercise 3	180 x 144	25	0.938s	26.7
Exercise 4	704 x 480	30	11.99s	2.5
Exercise 5	704 x 480	30	12.16s	2.5
Exercise 6	640 x 480	90	33.23s	2.7

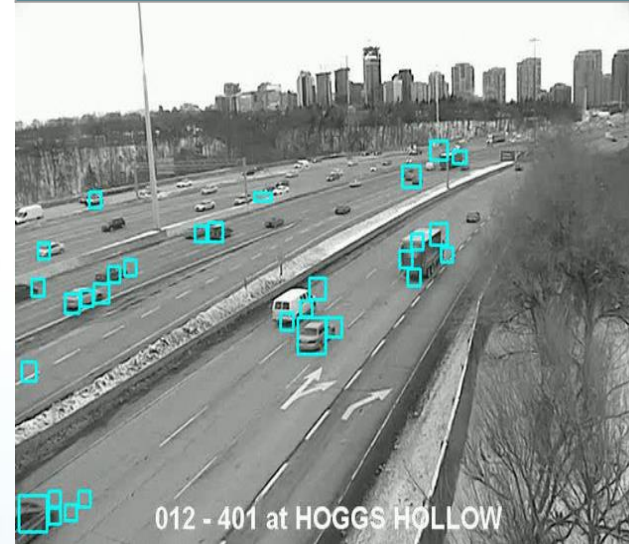
- Time illustrated is time it takes to process all frames in the exercise
- FPS = Frames per Second

Performance Plot

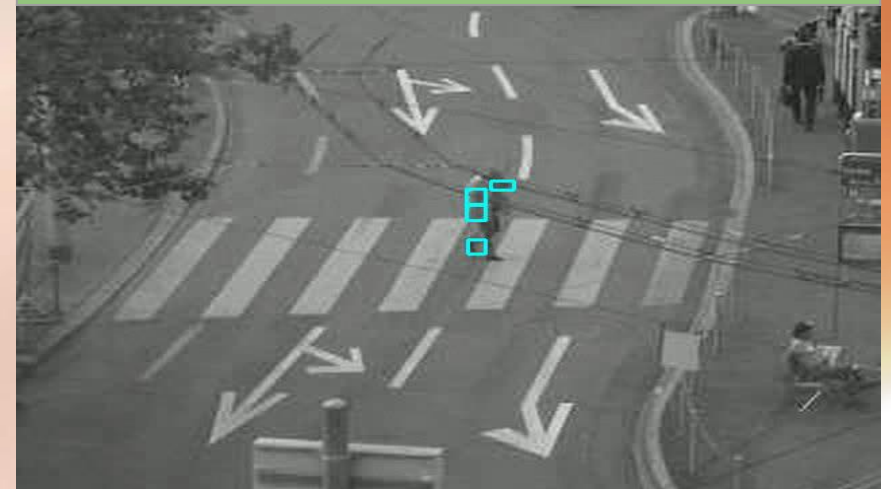


- While the Big O of our program is n^2 , it seems that in practice it appears linear

Exercise 5



Exercise 6



Pseudocode - Connected Components

Connected components with Union-Find

This function utilizes previously created functions to create an image of clusters with each cluster having its own unique label

INPUTS

B is the original binary image

LB will be the labeled connected component image, initially matrix same size as **B** but filled with zeros

OUTPUT

LB labeled connected component image

Procedure classical_with_union-find(**B**, **LB**)

Initialize()

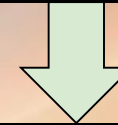
for **L** <- 0 to MaxRow

```
        //x
        for P <- 0 to MaxCol
            //y
            if B[L,P] == 1 then
                //(x*y)=n
                A <- prior_neighbors(L,P)

                //n
                if isempty(A) then
                    //n
                    M <- label
                    //n
                    label <- label +1
                    //n
                else
                    //n
                    M <- min(labels(A))

                // 4*n
                LB[L,P] <- M
                //n
                for X in labels(A)
                    //4*n
```

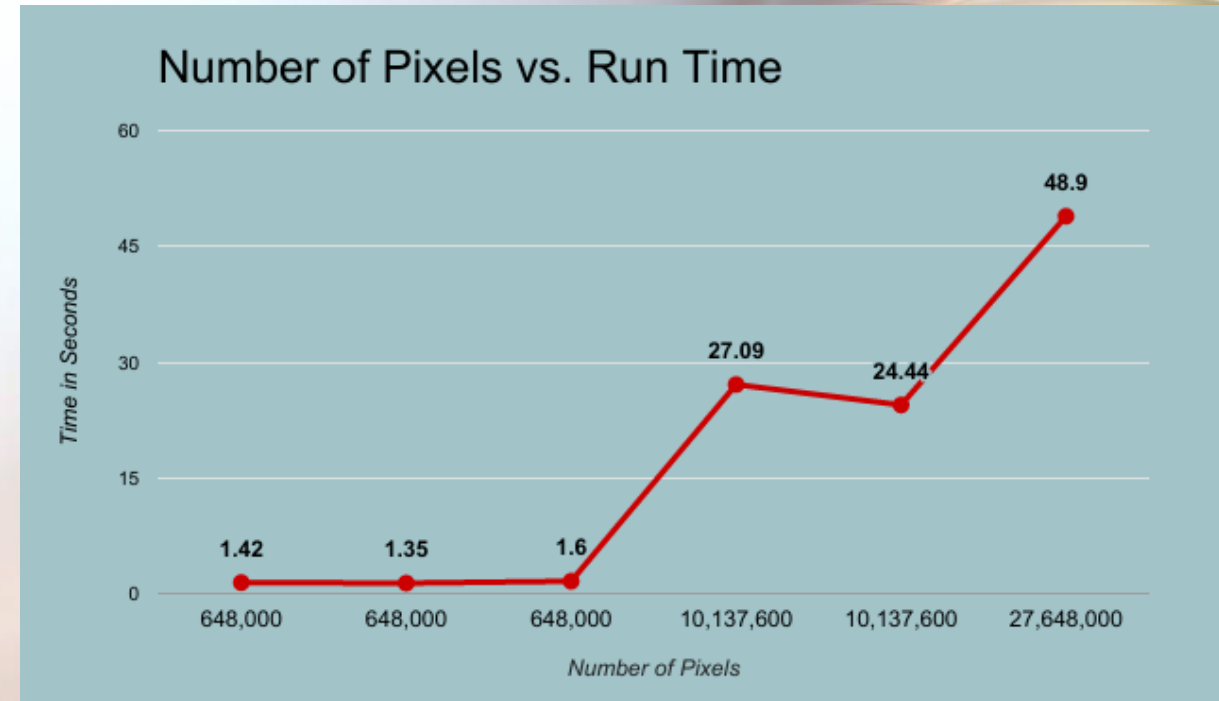
```
            if X != M
```



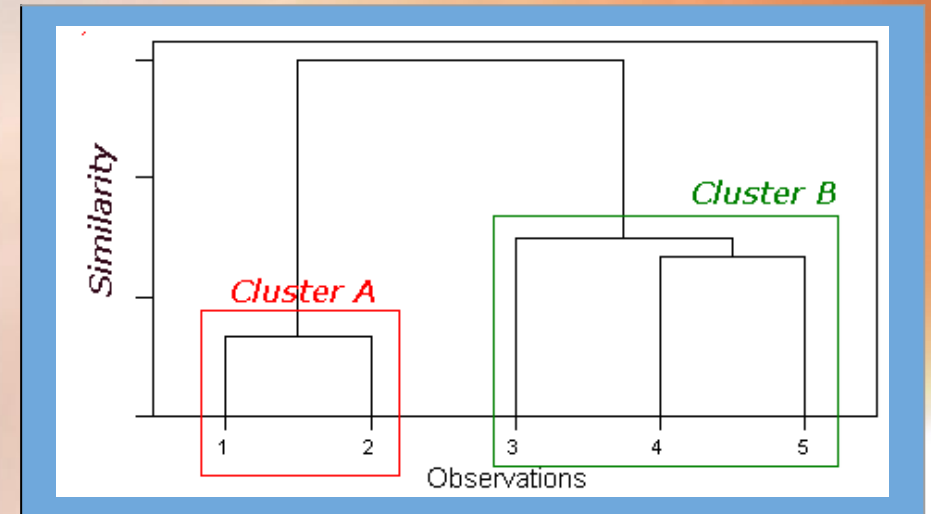
Timing Breakdown - Connected Components

Exercise	Resolution WxHxF	Input Pixels	Time	Cars
Exercise 1	180 x 144 x 25	648,000	1.97s	N/A
Exercise 2	180 x 144 x 25	648,000	1.986s	N/A
Exercise 3	180 x 144 x 30	648,000	1.604s	N/A
Exercise 4	704 x 480 x 30	10,137,600	27.1s	36
Exercise 5	704 x 480 x 30	10,137,160	24.44s	40
Exercise 6	640 x 480 x 90	27,648,000	48.9s	N/A

W = Width
H = Height
F = Frames



Component Grouping Algorithm



More on Image Processing

Gaussian Filtering



Histogram Equalization



Gaussian Filter

- Removes noise
- Leaves useful data
- Emphasizes large objects

Histogram Equalization

- Increases the contrast between pixels
- Easier to detect changes