# ARRAY DETECTION AND TRACKING

Submitted by
Aldrin Wilfred Arokiasamy

# Description

To share the particular vehicles information to the following vehicles, vehicle to vehicle communication is needed. The primary vehicles data were collected from its sensors and transmitted to the subordinate vehicles. For communicating huge data between vehicles, we need Multiple Input Multiple Output Radio Frequency (MIMO-RF) transceiver that is complex and has increased design cost. The drawback can be overcome with the use of Visual Light Communication (VLC), for the reason it is easily accessible and has high probability of seamless communication. In this simulation, a Light Emitting Array (LEA) arrangement was placed. The LEA acts as the transmitter, which was the required Region of Interest (ROI). The ROI was located using the Harris edge-corner detection method. The cameras placed in the secondary vehicles were the receivers in this case, where the ROI has to be extracted. To reduce the transmission error rate at the receiver processing unit, the Kalman filter tracking was used to locate and track the LEA. In this MATLAB simulation, for our convenience, we assume that the LEA moves horizontally or vertically, and the idea of Kalman Filter methodology was proposed along with geometry techniques to improve the locating and tracking of the LEA. The expected simulation results will show the tracking plots for both the schemes of Kalman filter with and without geometric properties.

### LEA arrangement:

In this simulation model, the LEA plays a main role for communication. The LEA arrangement has the array of LEDs that switches ON and OFF according to the output data from various sensors of the primary vehicle. For our convenience, the LEA arrangement has black outer boundary and white inner boundary as shown in figure 2.

# LEA locating using Harris edge-corner detection:

The Harris corner detection method was used to obtain the sharp intensity changes, thus locating the corners of the LEA. In this simulation model, the LEA

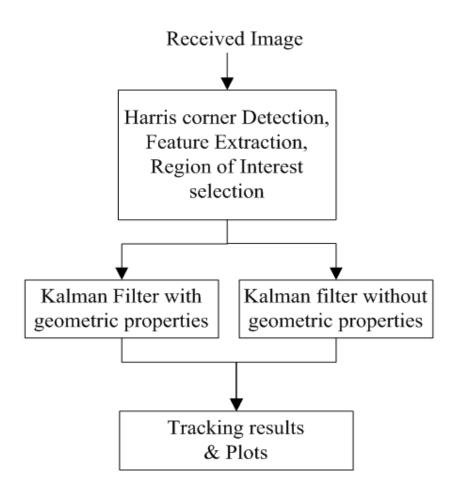


Figure 1: Implemented Algorithm.

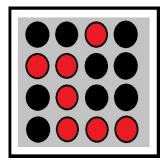


Figure 2: LEA arrangement with black and white boundary.



Figure 3: Background utilized for LEA detection and tracking method.



Figure 4: LEA with black and white boundary in a complex background.

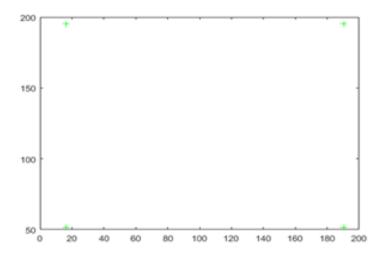


Figure 5: Four corners of the LEA found using Harris corner detection.

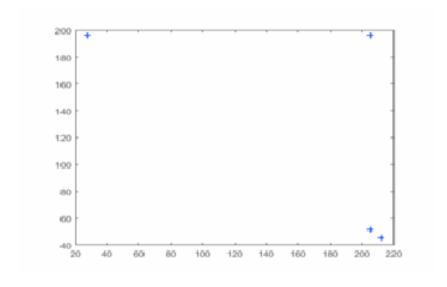


Figure 6: Harris detector finding difficult to find four corner points.

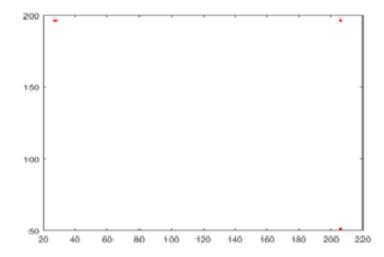


Figure 7: Combining two points which is nearer using tolerance.

arrangement was placed in the form of rectangle with black outer boundary and white inner boundary, thus improving the performance of the Harris edge-corner detection. For denoting the corner value the aggregate of the variations that was doubled is characterized as in (1).

$$E(u,v) = \sum_{(x,y)} w(x,y) [I(x+u,y+v) - I(x,y)]^2$$
 (1)

- E is the difference along the actual and the dispalced window.
- $\bullet$  u is the window's displacement in the x direction
- v is the window's displacement in the y direction
- w(x,y) is the window at position (x,y). This acts like a mask. Ensuring that only the desired window is used.
- I is the intensity of the image at a position (x,y)
- I(x+u,y+v) is the intensity of the moved window
- I(x,y) is the intensity of the original

Then, Using the Taylor series, expanding the above equation.

$$E(u,v) = B(\sum A)B'$$
(2)

where B is  $\begin{pmatrix} u & v \end{pmatrix}$  and A is  $\begin{pmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{pmatrix}$ 

$$M = \sum w(x, y)A \tag{3}$$

$$E(u,v) = B * M * B' \tag{4}$$

$$R = det(M) - k * (traceM)^2$$
(5)

where  $det(M) = \lambda_1 \lambda_2$  and  $traceM = \lambda_1 + \lambda_2$ 

If R is greater, then the obtained values are corners

#### MATLAB function used:

corners = detectHarrisFeatures (rgb2gray (INPUTIMAGE));

### Problem Statement and the Solution provided:

After Harris corner detection algorithm, the four corners of the rectangle array have to be obtained and the mid-point of the rectangle array has to be calculated. With the help of the mid-point, the rectangle array can be tracked using the Kalman filter, which was termed as the eye of the subordinate vehicles.

The four corner points and the midpoint of the rectangle were found. Sometimes the corner detection algorithm finds it difficult to find the four corner points and where the tolerance is used to make the closest points as the whole, thus helps us to get the fourth point, and then the LEA was found using the basic rectangle geometric properties. For instance, (x1, y1) were the top left most co-ordinates of the rectangle and (x3, y3) were the bottom right most co-ordinates of the same rectangle, then the bottom left most co-ordinates will be (x1, y3) and the top right most co-ordinates will be (x3, y1).

Error in detection of the corners was a problem that arises during the corner detection phase, because the algorithm detects the corners which have the sharp change over the pixels, so in-order to solve this problem, the geometric properties were used to find the corner that was not detected. Finally through the Kalman filter, tracking results for the cases with and without the geometric properties were plotted accordingly to show the best tracking.

# LEA tracking using Kalman filter:

The Kalman filter utilizes the estimation of object of the LEA in the obtained picture when the comparative assumption between the LEA and camera displaces,

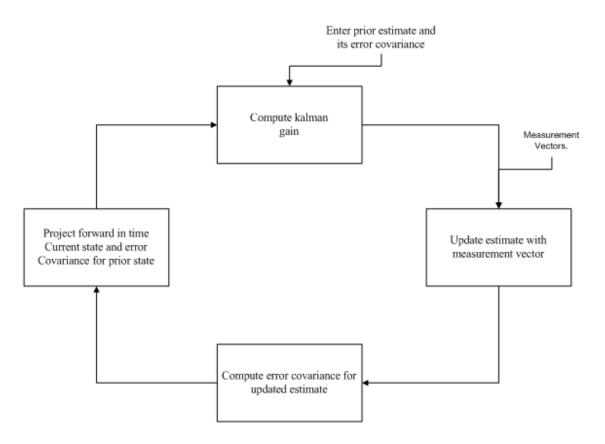


Figure 8: Kalman filter loop.

camera being the receiver. In addition to that the Kalman filter is used to track the position of the LEA in the complex background. The working of the Kalman filter is totally based on the four corner points obtained using the Harris corner detection and the midpoint of the LEA using the four corner points as per findings.

To apply Kalman filter to indicate the movement of the LEA in the obtained picture plane, we provide a midpoint of the LEA as the state vector  $x_k$  and the measure variable quantity  $y_k$  at time step k are defined as in the below Equation 6. Figure 8 shows the detailed summary of the Kalman filter operation.

$$x_k = \begin{pmatrix} x & y & v_x & v_y \end{pmatrix} y_k = \begin{pmatrix} x & y \end{pmatrix} \tag{6}$$

# Tracking results of moving LEA:

The tracking results of each and every co-ordinates for each cases, for the first 20 iterations were shown below from figure 9 to figure 12.

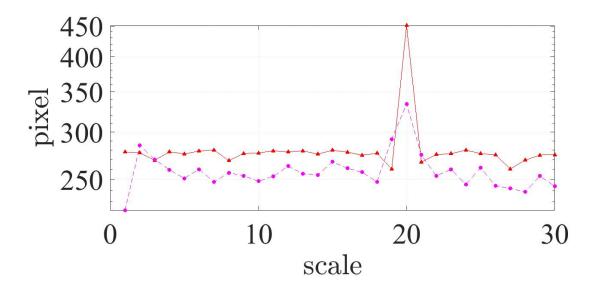


Figure 9: Center co-ordinate (x): Kalman filter tracking for case 1.

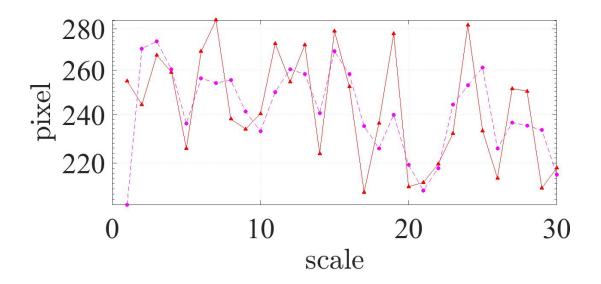


Figure 10: Center co-ordinate (y): Kalman filter tracking for case 1.

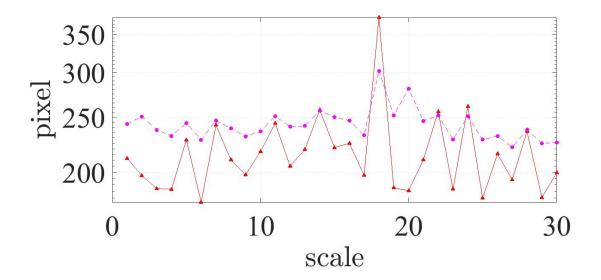


Figure 11: Center co-ordinate (x): Kalman filter tracking for case 2.

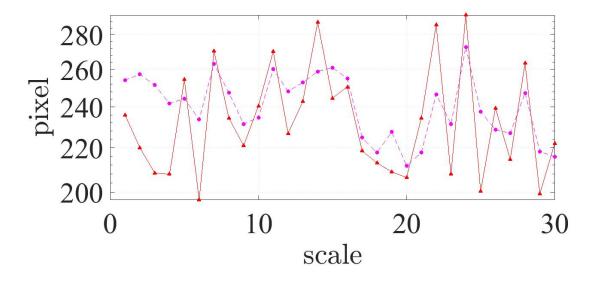


Figure 12: Center co-ordinate (y): Kalman filter tracking for case 2.

# Conclusion

The project is devised using a LEA location and pursual technique for the visual-MIMO framework using image analysing system that can be utilized for different applications. For building a reliable quality of LEA location, we chose the ROI that is the LEA, utilizing Harris corner detection for locating and kalman filter for tracking, with a composite foundation and background.

This project is mainly focusing on obtaining the LEA location and tracking the same for the seamless communications. Moreover the system is using LEDs and optical communications so the cost of design was simple and effective rather compared with RF wireless communications.

#### MATLAB CODE

```
clc;
close all;
clear all;
%------
No_Of_Imgs = 20;
%-----iterations used------
iteration_1=1;
i_1 = 1;
for s = 1:No_Of_Imgs
   var = 0.5;
%%-----Implementation with geomentry properties-----Implementation with geomentry
   Im=imread('canv.jpg');
   Imbg=imread('sasas.jpg');
   Imbgsize=size(Imbg);
   %-----scaling factor-----
   scaleK(s)=0.8+var*rand;
   position =[20 20];
   %-----implementing scaling in the image-----implementing scaling in the image-----
   Imfg=imresize(Im,scaleK(s));
   Imfgsize=size(Imfg);
   Imbg(position(1):position(1)+Imfgsize(1)-1,position(2):position(2)+Imfgsize(2)-1,:)=Imfg;
   figure;
   imshow(Imbg);
   %-----Harris corner implementation------
   corners = detectHarrisFeatures(rgb2gray(Imbg));
   a = corners.selectStrongest(4);
   %-----locating the strong points-----
   b = a.Location;
   hold on;
   figure,
   plot(a);
   hold on;
%----calculating the tolerence and proceeding to the geometry properties---
   to1=0.5;
   C = uniquetol(b,tol,'ByRows',true)
   figure;
   plot(C(:,1), C(:,2), '.r', 'MarkerSize', 10)
  [rows cols] = size(C);
%----calculation to find the center of the four strong points-----
   cdata = (C(:,[1 1]));
```

```
cdata = cdata';
    cdata
    c1 = min(min(cdata));
    c2 = c1;
    c4 = max(max(cdata));
    c3 = c4;
    ddata = (C(:,[2 2]));
    ddata = ddata';
    ddata
    d1 = min(min(ddata));
    d2 = d1;
    d4 = max(max(ddata));
    d3 = d4;
    %-----detecting the center point-----
    xc = ((d2/2)+(c1/2));
    yc = ((d1/2) + (c4/2));
   detected_center_point = [xc yc];
   detected_center_point
%------kalman filter implementation------kalman filter
 [xh, yh] = TrackKalman(xc, yc)
 figure,
 hold on;
 plot(xc, yc, 'y*')
 hold on;
 plot(xh, yh, 'bs')
 pause(0);
 Xcsaved(:, s) = [xc yc];
 Xhsaved(:, s) = [xh yh];
 xcsaved(:, s) = [xc yc]';
 Xhsaved(:, s) = [xh yh]';
%%-----properties-----Implementation without geomentry properties----------------
I=imread('canv.jpg');
Isize=size(I);
Ibg=imread('sasas.jpg');
Ibgsize=size(Ibg);
%-----scaling factor-----
scaleK(s)=0.8+var*rand;
position =[20 20];
%-----implementing scaling in the image-----implementing scaling in the
Ifg=imresize(I,scaleK(s));
Ifgsize=size(Ifg);
Ibg(position(1):position(1)+Ifgsize(1)-1,position(2):position(2)+Ifgsize(2)-1,:)=Ifg;
figure;
imshow(Ibg);
%------Harris corner implementation-------
corners = detectHarrisFeatures(rgb2gray(Ibg));
%[features, valid_corners] = extractFeatures(recovered, corners);
%-----locating the strong points-----
```

```
q = corners.selectStrongest(4);
%-----locating the strong points-----
 r = q.Location;
hold on;
figure,
plot(q);
hold on;
%-----calculation to find the center of the four strong points-----
    xwdata = (r(:,[1 1]));
    xwdata = xwdata';
    xwdata
    xw1 = min(min(xwdata));
    xw2 = max(max(xwdata));
    ywdata = (r(:,[2 2]));
    ywdata = ywdata';
    ywdata
    yw1 = min(min(ywdata));
    yw2 = max(max(ywdata));
    %I=imcrop(Ibg,[x1 y1 width height]);
%-----detecting the center point------
    xwc = ((yw2/2)+(xw1/2));
    ywc = ((yw1/2) + (xw2/2));
    detected_center_pointw = [xwc ywc];
    detected_center_pointw
%-----kalman filter implementation------kalman filter
  [xl, yl] = TrackKalman(xwc, ywc)
  figure;
  hold on;
  plot(xwc, ywc, 'y*')
  hold on;
  plot(x1, y1, 'bs')
  pause(0);
 Xpwsaved(:, s) = [xwc ywc];
 Xlsaved(:, s) = [xl yl];
 Xpwsaved(:, s) = [xwc ywc]';
 xlsaved(:, s) = [xl yl]';
 iteration_1=iteration_1+1;
end
%-----plotting the track points of the kalman filter for both cases------
figure,
hold on;
plot(Xcsaved(1,:), Xcsaved(2,:), '*');
hold on;
plot(Xhsaved(1,:), Xhsaved(2,:), 's');
hold on;
figure,
hold on;
```

```
plot(Xpwsaved(1,:), Xpwsaved(2,:), '*');
hold on;
plot(Xlsaved(1,:), Xlsaved(2,:), 's');
hold on;
%close all;
%-----comparing the tracking plots for both cases-----
scale = 1:1:No_Of_Imgs;
figure;
posi1_Plot=semilogy(scale,Xcsaved(1,:),'marker','^','markersize',6,'markerfacecolor','r','lin
estyle','-','color','r');
hold on;
esti1_Plot=semilogy(scale,Xhsaved(1,:),'marker','o','markersize',6,'markerfacecolor','m','lin
estyle','--','color','m');
hold on;
grid on
xlabel('$$\mathrm{scale}$$','FontName','Times New
Roman', 'FontSize', 50, 'Interpreter', 'latex');
ylabel('$$\mathrm{pixel}$$','FontName','Times New
Roman','FontSize',50,'Interpreter','latex');
%legend([posi1_Plot esti1_Plot], 'actual position', 'estimated position', +2);
set(gca,'FontSize',50,'XGrid','on','YGrid','on','GridLineStyle',':','MinorGridLineStyle','non
e','FontName','Times New Roman');
scale = 1:1:No_Of_Imgs;
posi2_Plot=semilogy(scale,Xcsaved(2,:),'marker','^','markersize',6,'markerfacecolor','r','lin
estyle','-','color','r');
hold on;
esti2_Plot=semilogy(scale,Xhsaved(2,:),'marker','o','markersize',6,'markerfacecolor','m','lin
estyle','--','color','m');
hold on;
grid on
xlabel('$$\mathrm{scale}$$','FontName','Times New
Roman','FontSize',50,'Interpreter','latex');
ylabel('$$\mathrm{pixel}$$','FontName','Times New
Roman', 'FontSize', 50, 'Interpreter', 'latex');
%legend([posi2_Plot esti1_Plot], 'actual position', 'estimated position', +2);
set(gca,'FontSize',50,'XGrid','on','YGrid','on','GridLineStyle',':','MinorGridLineStyle','non
e','FontName','Times New Roman');
scale = 1:1:No_Of_Imgs;
figure;
posi3_Plot=semilogy(scale,Xpwsaved(1,:),'marker','^','markersize',6,'markerfacecolor','r','li
nestyle','-','color','r');
hold on;
esti3_Plot=semilogy(scale,Xlsaved(1,:),'marker','o','markersize',6,'markerfacecolor','m','lin
estyle','--','color','m');
```

```
hold on;
grid on
xlabel('$$\mathrm{scale}$$','FontName','Times New
Roman','FontSize',50,'Interpreter','latex');
ylabel('$$\mathrm{pixel}$$','FontName','Times New
Roman','FontSize',50,'Interpreter','latex');
%legend([posi1_Plot esti1_Plot], 'actual position', 'estimated position',+2);
set(gca,'FontSize',50,'XGrid','on','YGrid','on','GridLineStyle',':','MinorGridLineStyle','non
e','FontName','Times New Roman');
scale = 1:1:No_Of_Imgs;
figure;
posi4_Plot=semilogy(scale,Xpwsaved(2,:),'marker','^','markersize',6,'markerfacecolor','r','li
nestyle','-','color','r');
hold on;
esti4_Plot=semilogy(scale,Xlsaved(2,:),'marker','o','markersize',6,'markerfacecolor','m','lin
estyle','--','color','m');
hold on;
grid on
xlabel('$$\mathrm{scale}$$','FontName','Times New
Roman','FontSize',50,'Interpreter','latex');
ylabel('$$\mathrm{pixel}$$','FontName','Times New
Roman', 'FontSize',50, 'Interpreter', 'latex');
% legend ([posi2\_Plot\ esti1\_Plot], 'actual\ position', 'estimated\ position', +2);
set(gca,'FontSize',50,'XGrid','on','YGrid','on','GridLineStyle',':','MinorGridLineStyle','non
e','FontName','Times New Roman');
Warning: Image is too big to fit on screen;
displaying at 67%
b =
  36.4001 37.4323
  447.7591 37.5271
  36.5300 402.0515
  468.4582 20.5533
C =
  36.4001 37.4323
   36.5300 402.0515
  447.7591 37.5271
cdata =
   36.4001 36.5300 447.7591
   36.4001 36.5300 447.7591
ddata =
```

```
37.4323 402.0515 37.5271
  37.4323 402.0515 37.5271
detected_center_point =
 276.9162 242.5957
xh =
 221.7536
yh =
 194.2699
Warning: Image is too big to fit on screen;
displaying at 67%
r =
  38.4546 39.5095
 497.8367 39.5334
  38.4410 446.6477
  521.4609 20.5807
xwdata =
  38.4546 497.8367 38.4410 521.4609
  38.4546 497.8367 38.4410 521.4609
ywdata =
  39.5095 39.5334 446.6477 20.5807
  39.5095 39.5334 446.6477 20.5807
detected_center_pointw =
 242.5443 271.0208
x1 =
 265.8558
y1 =
 276.2022
```

Warning: Image is too big to fit on screen;

displaying at 67%

```
b =
  39.5033 474.5107
  39.5417
          40.7330
  529.0895 474.5603
  553.6090 20.4496
C =
  39.5033 474.5107
          40.7330
  39.5417
  529.0895 474.5603
  553.6090 20.4496
cdata =
  39.5033 39.5417 529.0895 553.6090
  39.5033 39.5417 529.0895 553.6090
ddata =
 474.5107 40.7330 474.5603 20.4496
 474.5107 40.7330 474.5603 20.4496
detected_center_point =
 269.9764 287.0293
xh =
 291.6052
yh =
 311.5100
Warning: Image is too big to fit on screen;
displaying at 67%
r =
  33.5500 34.4568
  33.5169 329.7871
  384.4556 20.5040
  367.0783 34.4527
xwdata =
```

33.5500 33.5169 384.4556 367.0783

```
33.5500 33.5169 384.4556 367.0783
```

ywdata =

 34.4568
 329.7871
 20.5040
 34.4527

 34.4568
 329.7871
 20.5040
 34.4527

detected\_center\_pointw =

181.6520 202.4798

x1 =

243.7142

y1 =

269.1625

Warning: Image is too big to fit on screen; displaying at 67%

b =

455.4810 408.7960 455.4872 37.7879 476.4535 20.5546 20.5564 425.4581

C =

20.5564 425.4581 455.4810 408.7960 455.4872 37.7879

cdata =

20.5564 455.4810 455.4872 20.5564 455.4810 455.4872

ddata =

 425.4581
 408.7960
 37.7879

 425.4581
 408.7960
 37.7879

detected\_center\_point =

269.1721 246.6376

```
xh =
 264.5354
yh =
 267.8958
Warning: Image is too big to fit on screen;
displaying at 67%
r =
 435.4496 390.9664
 435.4542 37.1106
 455.4540 20.5472
  20.5454 407.4589
xwdata =
 435.4496 435.4542 455.4540 20.5454
 435.4496 435.4542 455.4540 20.5454
ywdata =
 390.9664 37.1106 20.5472 407.4589
 390.9664 37.1106 20.5472 407.4589
detected_center_pointw =
 214.0022 238.0006
x1 =
 247.0860
y1 =
 261.0396
```

The results for the first three iterations were shown below:



