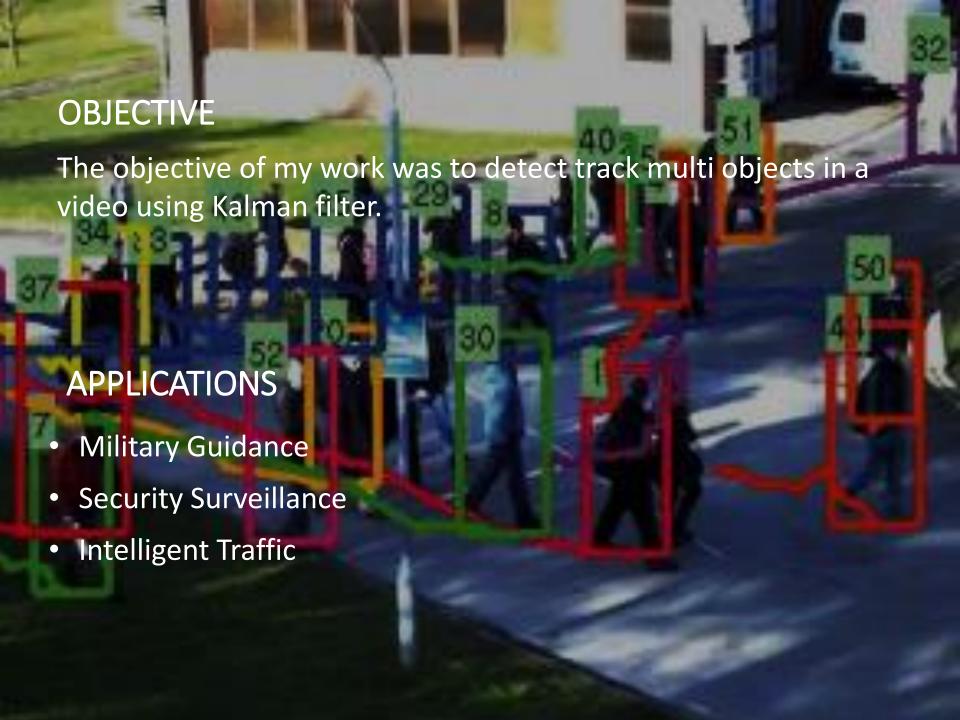
Multi-object Detection and Tracking Using Kalman Filter

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TWO PROBLEMS

First: Detect Objects from video

Image Processing Techniques

Second: Track Detected Objects

- Kalman Filter
- Assignment Problem

OBJECT DETECTION: BACKGROUND SUBTRACTION

How to do Background subtraction

- Estimate background at time t
- Subtract estimated background from current input frame
- Apply a threshold to the difference to get the foreground mask.



OBJECT DETECTION: BACKGROUND SUBTRACTION

If background estimated to be previous frame:

$$B(x,y,t) = I(x,y,t-1)$$

Background subtraction:

$$|I(x,y,t) - I(x,y,t-1)| > Th$$



OBJECT DETECTION: BACKGROUND SUBTRACTION

Use mean filtering for background estimation:

B(x,y,t) =
$$\frac{1}{n} \sum_{i=1}^{n} I(x, y, t - i)$$

Background subtraction:

$$|I(x,y,t) - \frac{1}{n} \sum_{i=1}^{n} I(x,y,t-i)| > Th$$

Time window n = 10

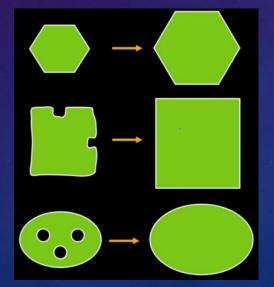


Foreground mask after thresholding to t: I(x,y,t)

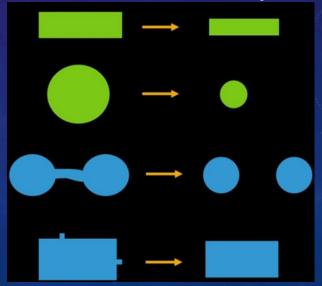
OBJECT DETECTION: MORPHOLOGICAL OPERATIONS

- Non-ideal Background estimation-> Noise in foreground mask,
 Objects are not enhanced
- Use Dilation and Erosion

Dilation technique



Erosion technique



SUMMARY OF OBJECT DETECTION



Original frame at time t



Foreground Mask at time t

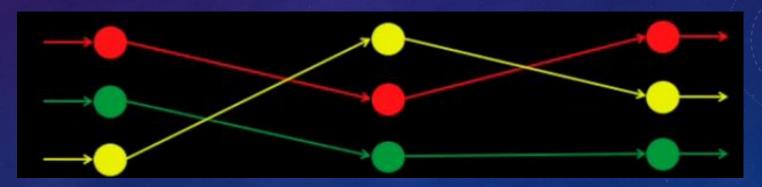


Using erosion followed by dilation at time *t*

Why do we even need prediction algorithm like Kalman filter?



Objects detected in consecutive frames



Assigning node

What is Kalman Filter?

- -Combination of two Ideas
- Linear system
- Linear recursive estimation

The kalman filter is an efficient algorithm for estimating the state of a system from noisy measurement.

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The kalman filter is an <u>efficient algorithm</u> for <u>estimating</u> the <u>state</u> of a <u>system</u> from <u>noisy measurement</u>.

A discrete time linear time-invariant dynamical system is a set of matrix of the formula

$$x_i = Ax_{i-1} + w_i$$
 w_i : represents process noise $\sim N(0,Q)$
 $z_i = Hx_i + v_i$ v_i : represents measurement noise $\sim N(0,R)$

- x_i is a state variable at *ith* instance: coordinates of centroid of detected object
- A is the transition matrix
- H is the measurement matrix
- $\bullet \quad E[w_i v_j^T] = 0$

Kalman filter estimates states in two steps:

Time Update Equation

$$\hat{x}_{i}^{-} = A\hat{x}_{i-1} + w_{i}$$
 $P_{i}^{-} = AP_{i-1}A^{T} + Q$

- Estimate future state variable
- Estimate for error covariance

Prediction of state

Kalman Gain

Measure update equation

$$K_{i} = P_{i}^{-}H^{T}(HP_{i}^{-}H^{T} + R)^{-1}$$

$$\hat{x}_{i} = \hat{x}_{i}^{-} + K_{i}(z_{i} - H\hat{x}_{i}^{-})$$

$$P_{i} = (1 - K_{i}H)P_{i}^{-}$$

- Compute Kalman Gain
- Update state variable *x* based on measurement *z*.
- Update error covariance

The basic programme process is as follows:

Processing the background



Initialize the Kalman filter

Show the labelled tracks

1

Compute time-update and measurement update equations and estimate coordinates

Read in one frame

Assign detected coordinates to previous predicted coordinates

Detect and extract object

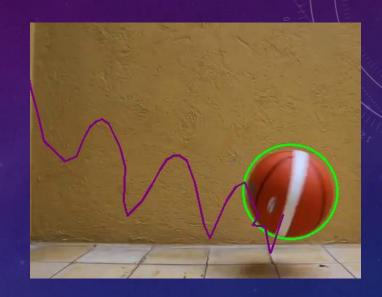


Find out the centroid

RESULTS









RESULTS

REFERENCES

- 1. Xin Li Kejun Wang, Wei Wang and Yang Li. A Multiple Object Tracking Method Using Kalman Filter.
- 2. Sumit Kumar Pal, Sohan Ghorai. Moving Object Tracking System In Video With Kalman Filter.
- 3. Jonathan Owensa, Andrew Hunter. A Fast Model-Free Morphology Based Object Tracking Algorithm.

THANK YOU