

Jaikrishna Neelkamal Rangeesh Riya



Motivation

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Motivation | Hawk-eye

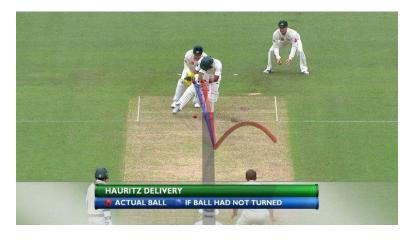
Crucial Decisions in Sports

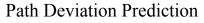


Ball Tracking



Future Path Prediction







Motivation | Hawk-eye

- Multiple cameras
 - Camera calibration
- Image Processing
 - o Ball Recognition
 - Ball Characteristics
- Filtering
 - State Model
 - Measurement Model
 - UKF parameters
 - o LSE
- Results





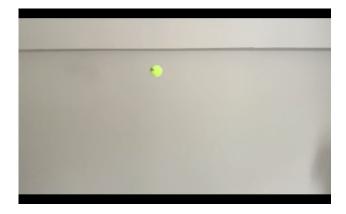
Computer Vision



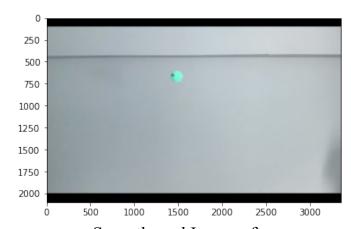
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Image Processing | Ball Recognition

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Imported Picture

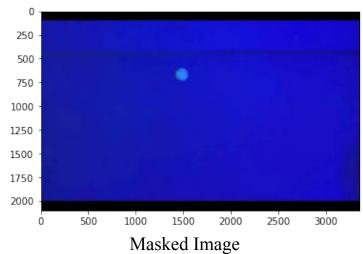


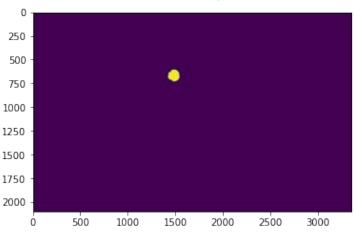
Smoothened Image after Gaussian Blur Filtering



Image Processing | Ball Recognition

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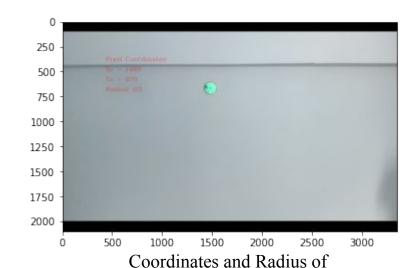


White Noise Cancellation

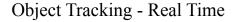


Image Processing | Ball Characteristics

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the Ball





Camera | Calibration

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Camera Options

- Phone Camera
 - Real-Time & Offline
- Laptop Camera
 - Real-Time & Offline
- DSLR
 - Offline
- USB Cam
 - Real-Time & Offline



Camera | Calibration

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Origin Determination and Focal Length Calibration







Image Processing | Real-time Ball Tracking

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Filtering

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Filtering | State Model

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Input to Model



2D Projectile motion



Filtering | State Model

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State : (x, y, z)
Ball coordinates in world frame

State Equations

$$x_{k+1} = x_k + \frac{(x_k - x_0)}{\Delta t} dt$$

$$y_{k+1} = y_k + V_{yo}dt - g * \Delta t * dt$$

$$z_{k+1} = z_k + \frac{(z_k - z_0)}{\Delta t} dt$$



Filtering | Measurement Model

- Single camera
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 - o Identify the ball
 - Calculate characteristics of ball
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 - **o** Measurement Model
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Measurements : (u, v, r)
Ball coordinates in image
frame and Ball Radius

Measurement Equations

$$u = 617 - \frac{fx}{z}$$

$$v = 360 - \frac{f}{z}(y + \frac{z}{20} - 18 * 2.54)$$

$$r = \frac{fR}{z}$$



Filtering | UKF Parameters

- Single camera
 - o Camera calibration

 $Q := \begin{bmatrix} 0.1 & 0 & 0 \\ 0 & 0.1 & 0 \\ 0 & 0 & 0.1 \end{bmatrix} \quad R := \begin{bmatrix} 10 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

- Image Processing
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Noise Q and R matrix

- The states x,y,z are uncorrelated. So the Q matrix is diagonal.
- Tennis ball is dense enough for its projectile to be assumed perfect. Hence we picked low process variance.
- The measurements u,v,r are uncorrelated. So the R matrix is diagonal.
- The state y in the model heavily depends on the initial conditions. So measurement v was entrusted with higher confidence (low variance)
- Variation in r is much smaller as distance from the screen is fixed. Low Variance.

Filtering | UKF Parameters

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Unscented Transform

$$\lambda = \alpha^2(n+\kappa) - n$$

$$\begin{cases} \mathcal{X}_0 = \mu \\ \mathcal{X}_i = \mu + \left[\sqrt{(n+\lambda)\Sigma}\right]_i, & fori = 1..n \\ \mathcal{X}_i = \mu - \left[\sqrt{(n+\lambda)\Sigma}\right]_{i-n} & fori = (n+1)..2n \end{cases}$$

$$W_0^m = \frac{\lambda}{n+\lambda}$$

$$W_0^c = \frac{\lambda}{n+\lambda} + 1 - \alpha^2 + \beta$$

$$W_i^m = W_i^c = \frac{1}{2(n+\lambda)}$$
 $i = 1..2n$

- α , β and κ are the parameters to be chosen
- Larger α spreads the sigma points further
- $\alpha = 0.1$ was a good choice (trial and error)
- $\beta = 2$ is advisable for gaussian problems
- $\kappa = 3$ -n is advisable in general



Filtering | LSE

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- The general trajectory of a projectile with velocity components along x and y as parameters was curve fit onto the measurements using least squares approach.
- Assumption: 2D projectile
- Susceptible to errors due to outliers but reasonable approach in the absence of extreme outliers.

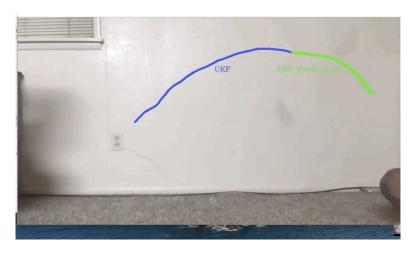
Cost Function

$$\sum_{i=1}^{N} \left[y - \frac{V_{yi}(x - x_o)}{V_{xi}} - y_o + \frac{g}{2} \left(\frac{x - x_o}{V_{xi}} \right)^2 \right]^2$$

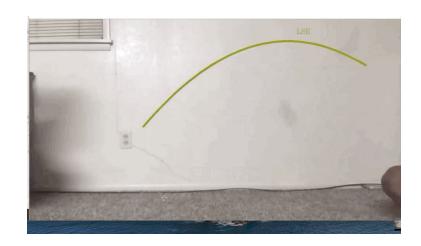


Filtering | Results

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UKF



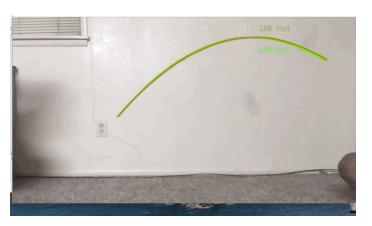




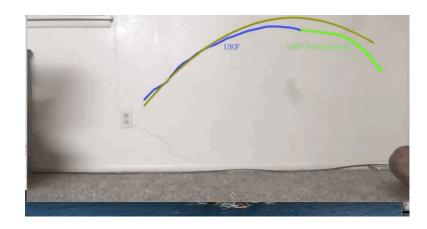
Filtering | Results - Comparison



LSE vs UKF Full



LSE Part vs LSE Full



UKF Part vs LSE



Thank You.

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