



## 1.0 Introduction

- The proposed system is aimed to continuously track moving object in random motion.
- Applications: robotic vision, security surveillance, traffic control and analysis of sports video.
- Some of the current tracking systems are less robust as focused in tracking object in predefined trajectory.
- An algorithm to track the random motion object is necessary.
- Multiple hypotheses property in particle filter enhanced with gradient optimization property of mean shift to estimate the target moving direction.



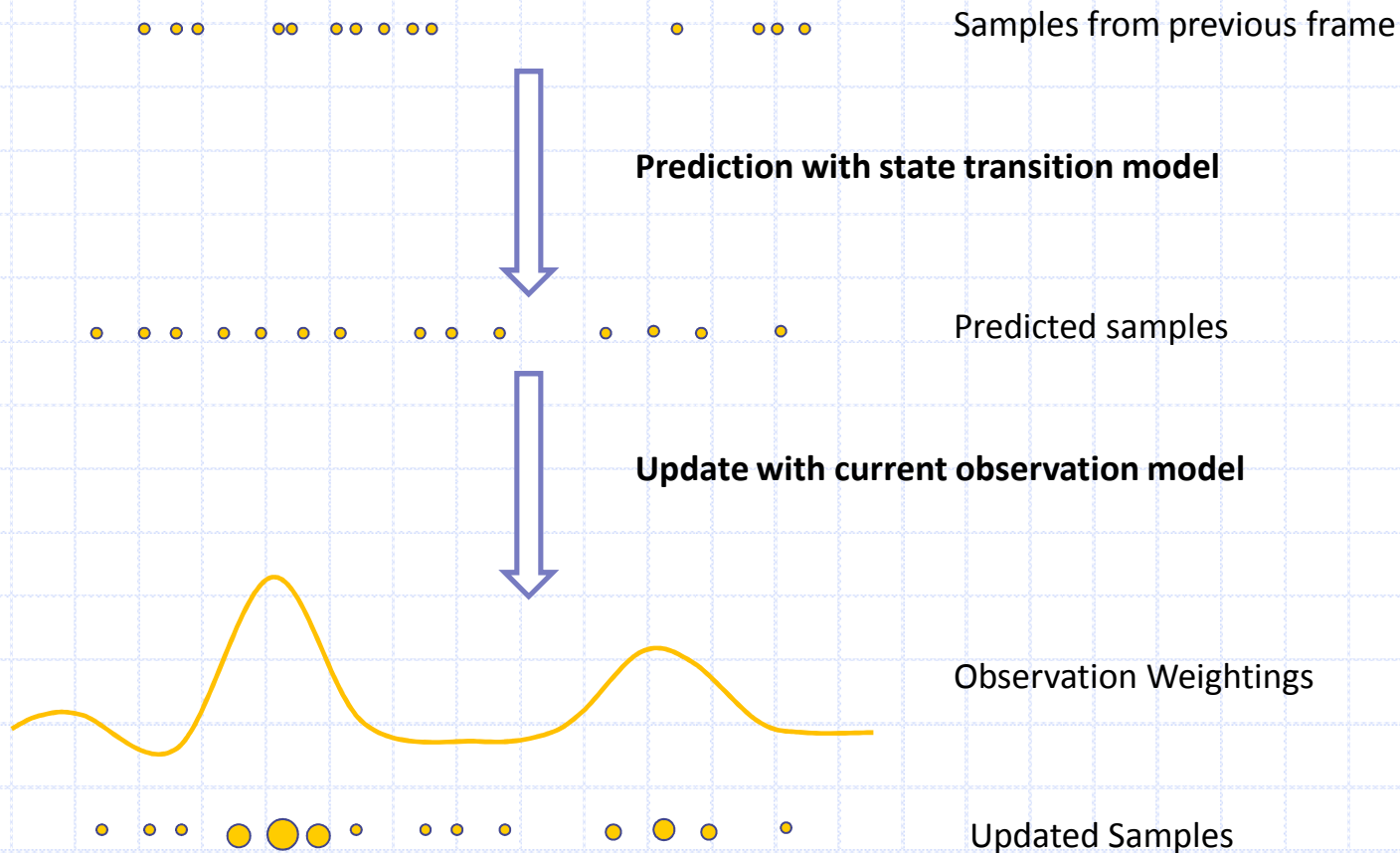
## 2.0 Objective

- To continuously track a random motion object under varying video conditions using the hybrid of particle filtering algorithm with mean shift approach.
- Capable to track object more efficient compared to particle filter tracker and mean shift tracker alone.



### 3.0 Methodology

- Particle Filter Framework





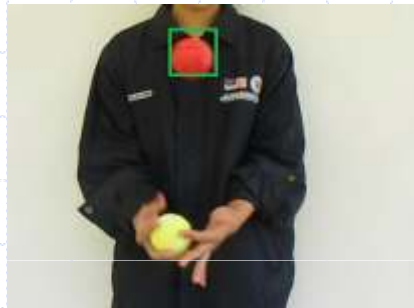
### 3.0 Methodology

- Mean Shift applied in Particle Filter
  - The direction of the movement for each particle is given by gradient optimization of the kernel masking.
  - The new resulting particle set represents the modes of the distribution, results in fewer particles are needed to maintain the multi-modal distribution.



## 4.0 Results and Discussions

### TRACKING RESULT ON ABRUPT DIRECTION AND VELOCITY CHANGING TARGET



Frame 104



Frame 117



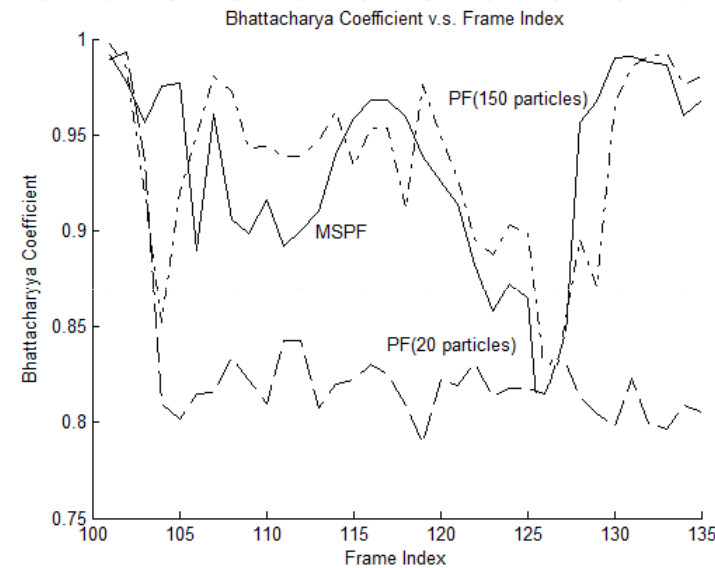
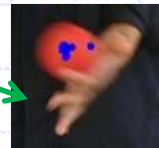
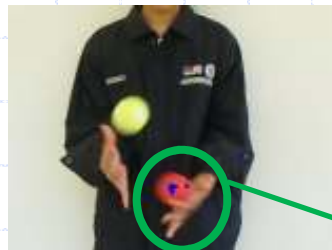
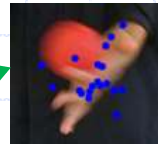
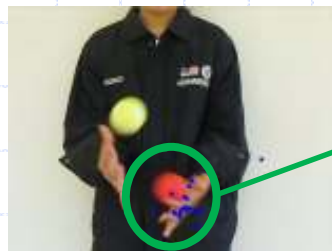
Frame 126



Frame 130

## 4.0 Results and Discussions

### EFFECT OF MEAN SHIFT ON PARTICLE SAMPLING



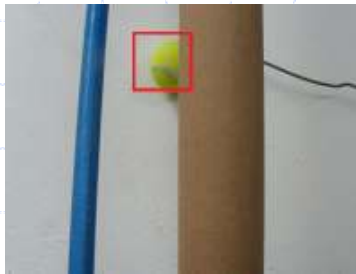
- After mean shift is applied to each particle, these particles move to the more likely position and concentrated in the local mode due to the gradient optimization property.
- PF tracker alone can only achieve sufficiently high coefficient when the number of particles is around 150, as compared to the MS embedded PF tracker where its sample size is only 20.





## 4.0 Results and Discussions

### Complete Occlusion



Frame 535



Frame 544



Frame 547



Frame 550

Mean shift embedded  
particle filter tracker



Frame 535



Frame 544



Frame 547



Frame 550

Mean shift tracker

## 4.0 Results and Discussion

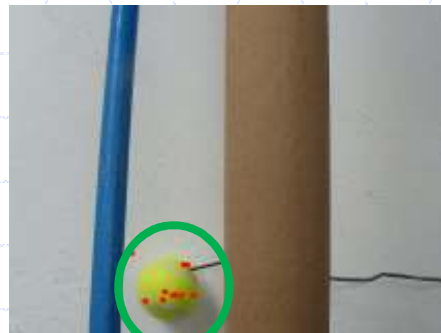
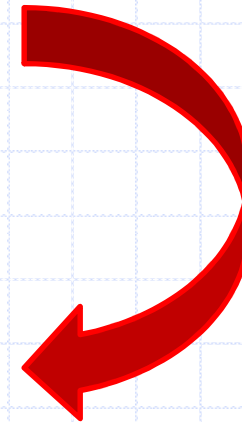
### Occlusion Detection and Recovery Process



Before occlusion



Completely occluded



Keep in track



Target reappears

- The uniform distribution covers larger search area around to predict the location of the target when it is reappears again.



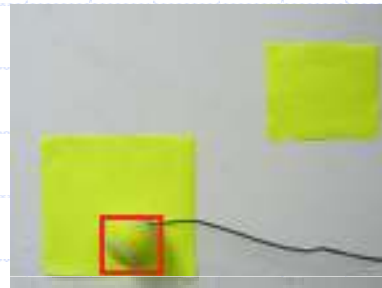


## 4.0 Results and Discussions

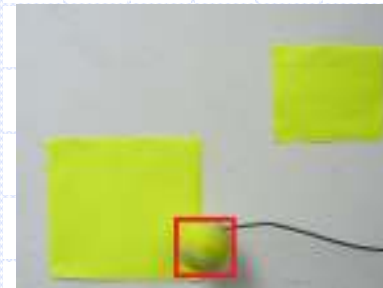
### Like-Colored Background



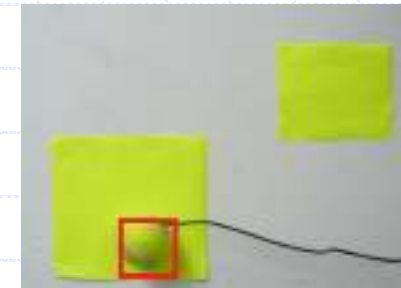
Frame 303



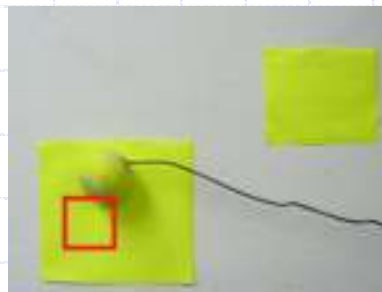
Frame 305



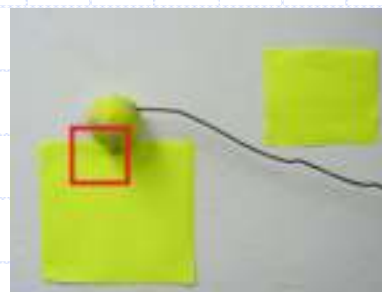
Frame 303



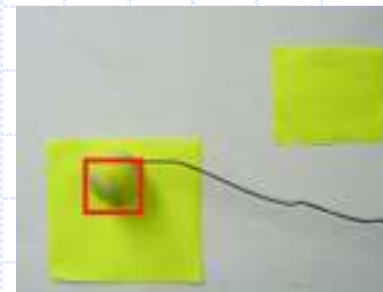
Frame 305



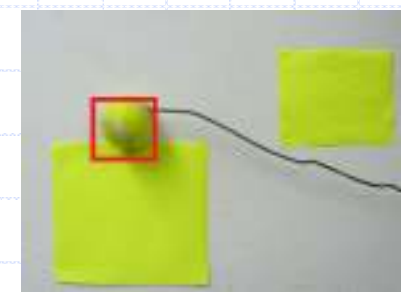
Frame 308



Frame 310



Frame 308



Frame 310

The tracking result of MS embedded  
PF tracker without edge feature

The tracking result of MS embedded  
PF tracker with edge feature



## 5.0 Conclusion

- Enhancement of particle filter sampling by mean shift algorithm.
- The developed tracker is capable to track abrupt direction and velocity changing object.
- The ability of maintaining multiple hypotheses on PF combined with the gradient optimization property of MS enable the tracker to deal with occlusion and clutter situations.