



1.0 Introduction

- Earlier vehicle tracking systems are normally based on colour feature because it is a promising feature that can be used to overcome partially occlusion and scale invariant incidents.
- Unfortunately, colour feature will lead to inaccurate results when the background colour is complex or too similar with the target vehicle.
- Hence, shape feature is introduced to the tracking algorithm to enhance the accuracy of the tracking performance.
- Although, shape feature will increase the accuracy of tracking rigid vehicle but it consume of more computation time during the tracking process.
- An enhanced particle filter with adaptive multiple cues overlapping vehicle tracking algorithm is proposed to continuously track the occluded vehicle effectively.



2.0 Objective

- Continuously tracking vehicle under various occlusion incidents using enhanced particle filter algorithm with multiple cues.
- Robustly tracking the vehicles, and significantly improved the accuracy in tracking the occluded vehicles without compromising the computational time.



3.0 Methodology

Colour Feature and Distribution Model

- Colour histogram of target vehicle is generated by using 8 x 8 x 8 bins RGB colour space.
- Bhattacharyya coefficient,

$$\rho[p, q] = \sum_{u=1}^{N_c} \sqrt{p_u q_u}$$

$$p = \{p_u\}_{u=1 \dots N_c}$$

$$q = \{q_u\}_{u=1 \dots N_c}$$

- From the coefficient obtained, the Bhattacharyya distance can be calculated.

$$d_c = \sqrt{1 - \rho[p, q]}$$

- Based on Bhattacharyya distance, the weight of the particles will calculate using

$$\varphi_c = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{d_c^2}{2\sigma^2}}$$



3.0 Methodology

Shape Feature and Distribution Model

- After the shape of the vehicle being extracted, the weight of the particles based on shape feature will be computed using Hausdorff distance.

- where $A = \{a_u\}_{u=1\dots N_c}$ is the set of points obtained from reference vehicle
 $B = \{b_u\}_{u=1\dots N_c}$ is the set of points obtained from target vehicle

- Hausdorff distance between two points set can be calculated based on

$$H_{dist}(A, B) = \max(h(A, B), h(B, A))$$

$$h(B, A) = K^{th}_{b \in B} \min_{a \in A} \|a - b\|$$

- where $K^{th}_{b \in B}$ denotes the K^{th} ranked value in the set of distance computed.

$$\varphi_s = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{H_{dist}^2}{2\sigma^2}}$$



3.0 Methodology

Proposed Particle Filter

- With the purpose of the shorten the computational time while increasing the accuracy of the tracking results, the proposed particle filter algorithm is need to be adaptive.
- Shape feature and colour feature will be calculated separately.
- In this adaptive particle filter algorithm, the shape feature is used to differentiate the target vehicle from obstacles.
- Resampling using colour feature to stronger the estimated position of the target vehicle.



4.0 Results & Discussions (Before Occluded)



a) Particle filter by using colour feature only



b) Particle filter by using shape feature only



c) Particle filter by using proposed multiple cues

- The single feature and multiple features can both track the moving vehicle accurately.
- Before occlusion occur, the information of the moving vehicle can be clearly obtained and without influenced by others obstacles.



4.0 Results & Discussions (Partially Occluded)



a) Particle filter by using colour feature only

b) Particle filter by using shape feature only

c) Particle filter by using proposed multiple cues

- The accuracy of the tracking by using single feature has been decreased due to the information of the moving vehicle is influenced by the static vehicle.
- With the proposed tracking algorithm, the moving vehicle can be accurately located.



4.0 Results & Discussions (After Occluded)



a) Particle filter by using colour feature only

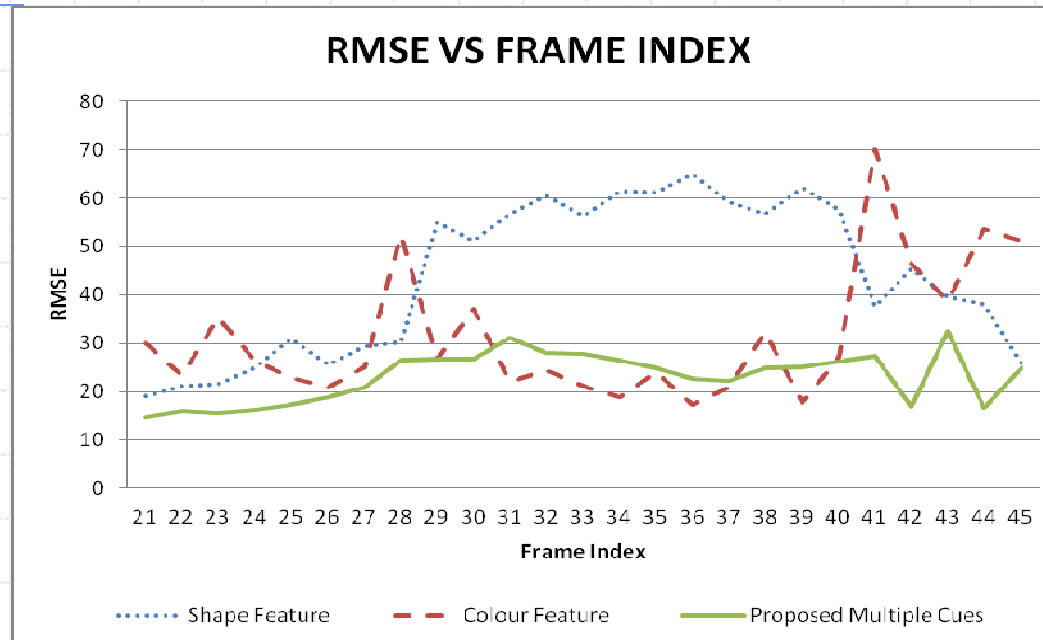
b) Particle filter by using shape feature only

c) Particle filter by using proposed multiple cues

- When occlusion occur, the information of the moving vehicle will be influenced by the obstacle or static vehicle.
- With single feature algorithm, more time will be needed to recover the tracking of the moving vehicle while it is totally lost track.
- With proposed algorithm, the moving vehicle can be located immediately after occlusion occurred.



4.0 Results & Discussions



Graph of root mean square (RMSE) versus frame index

- From the graph shown, the proposed multiple cues algorithm has a more accurate tracking results compared to single feature used.



5.0 Conclusions

- Tracking algorithm with multiple cues shown a promising result since more information describe the vehicle can be obtained from the experiments.
- Besides, the proposed algorithm with multiple cues is capable of dealing with the efficiency and effectiveness of the tracking performance.