

# Localisation and Mapping for QUT Motorsport

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## **QUT Motorsport**

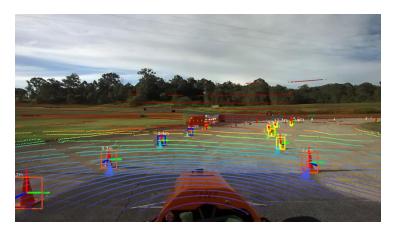


What can we learn from how a human driver balances high speed and safety to optimise performance?







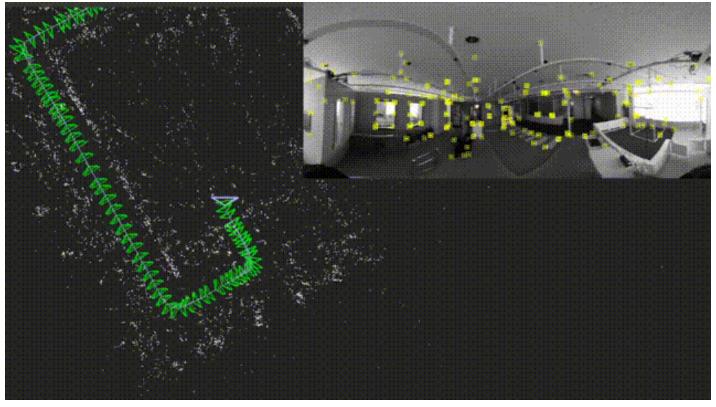






## **Background Context**

VSLAM: Visual Simultaneous Localisation and Mapping



CRICOS No.00213J OpenVSLAM [1]

## Aims & Objectives

#### **Challenges**

- Landmark detection in a sparse scene
- Loop closure with a lack of texture and repetitive features

## ObjectivesEvaluate sModify a s

- Evaluate state-of-the-art SLAM systems
- Modify a system for a racetrack scene
- Test loop closure and relocalisation





ORB SLAM 2<sup>[1]</sup> on a suburban sequence (left) and racetrack sequence (right).





## How do existing state-of-the-art SLAM methods perform comparatively?

## Research Questions

How can existing methods be modified to improve the QUT Motorsport driverless perception pipeline?

How can the viability of a SLAM method be verified for high-speed applications using a racetrack scene?

RQ1. Comparison of SoTA SLAM

### Related Work

#### Feature-**Based SLAM**

• Loop closure detection.[1]

Direct Methods Visual odometry in environments with little texture.[4]

- Rely on landmark density.<sup>[2]</sup>
- Feature loss in fast environments.[3]
- Cannot achieve loop closure.<sup>[5]</sup>
- Rely on high-quality and slow-moving video frames.<sup>[5]</sup>

#### DIRECT





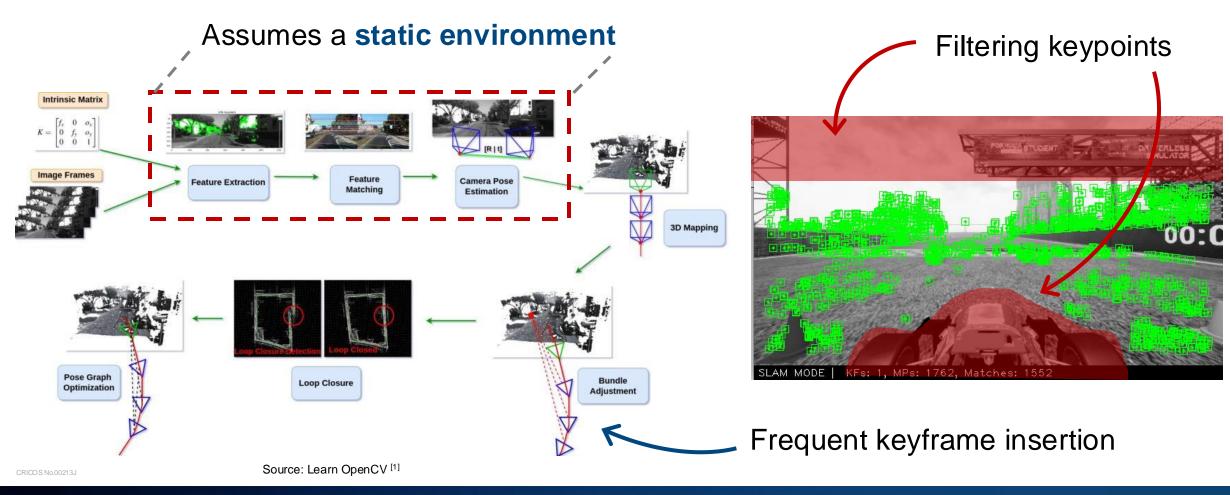
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Q1. Comparison of SoTA SLAM RQ3. Viability of SLAM using motors,

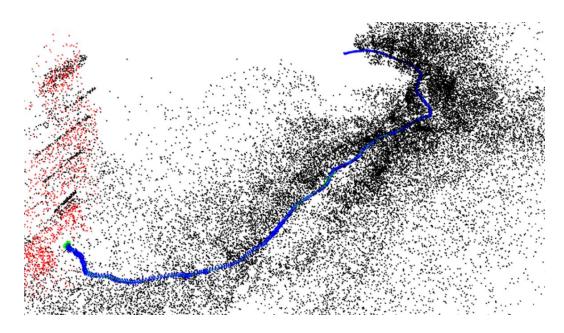
## Methodology

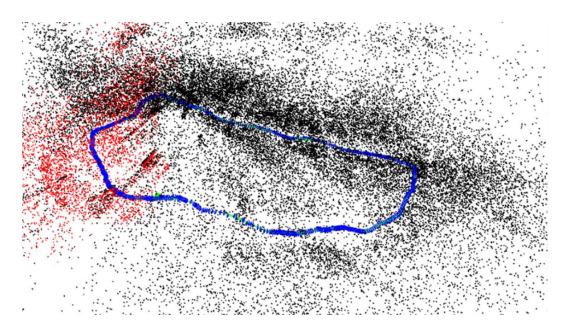






## **Loop Closure**





The trajectory before (left) and after (right) performing loop closure.

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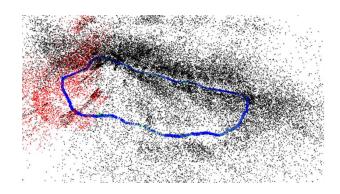


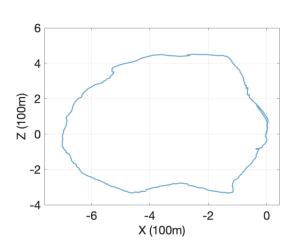


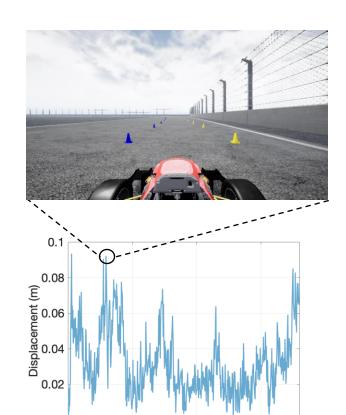


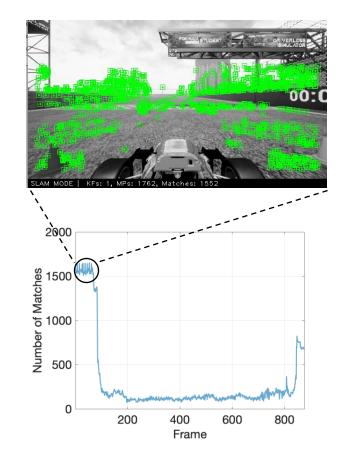
RQ1. Comparison of SoTA SLAM RQ3. Viability of SLAM using motorspor

### **Evaluation**









ORB SLAM2 map viewer (top left), frame 118 corresponding to the highest peak in displacement per frame (top middle), starting frame corresponding to stationary matches (top right), key frame trajectory (bottom left), displacement per frame (bottom middle), number of matches per frame (bottom right).

400

Frame

600

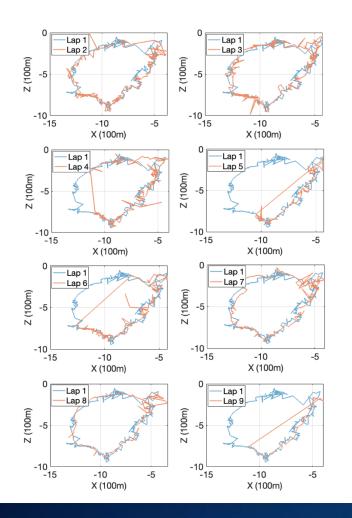
200

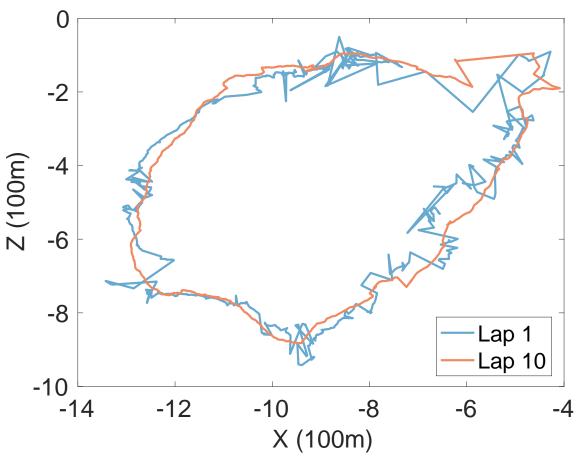




### **Evaluation**

Lap 1 vs:	ATE (m)
2	0.465
3	0.478
4	0.535
5	0.508
6	0.504
7	0.531
8	0.436
9	0.462
10	0.416





Keyframe trajectory of lap 1 compared against laps 2-10.





## Research Significance & Future Work

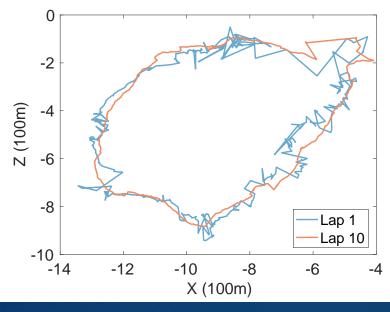
#### **Contributions**

- Modified feature-based SLAM system for a racetrack environment
- Verified viability of SLAM for high-speed challenging environments using motorsport
  - Map reuse, loop closure, relocalisation

#### **Future Work**

- Integrated SLAM methods
- Ground truth reference
- Constrain system using the vehicle's dynamics









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