Motion Detection and Tracking: Algorithms and Applications

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Introduction

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Introduction

Introduction

- Motion detection and tracking (MDT) are crucial in digital image processing (DIP).
- Applications include video surveillance, human-computer interaction, and autonomous navigation.
- This review provides an overview of state-of-the-art MDT algorithms and applications.

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- Early techniques: Frame differencing and basic background subtraction.
- Advanced methods: Gaussian Mixture Models (GMM), codebook-based methods.
- Optical flow methods: Lucas-Kanade, Horn-Schunck.
- Deep learning revolution: CNNs adapted for object detection and tracking.

- Occlusions: Multi-object tracking and detection-based approaches.
- Complex backgrounds: Adaptive background modeling techniques.
- Illumination changes: Histogram equalization and adaptive lighting compensation.
- Scale variations: Scale-invariant methods and models.
- Object appearances: Robust feature extraction and deformable models.

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Feature Extraction and Object Segmentation

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Feature Extraction and Object Segmentation

Feature Extraction and Object Segmentation

Key Paper: Yilmaz et al.

- Proposed a framework combining feature-based methods with mean-shift algorithms.
- Demonstrated robust performance in handling occlusions and abrupt movements.
- **Gap:** Limited performance in highly dynamic scenes with fast-moving objects.

Feature Extraction and Object Segmentation

Key Paper: Kalal et al.

- Developed the Tracking-Learning-Detection (TLD) algorithm.
- Integrated feature-based tracking with online object detection.
- Gap: Struggles with long-term occlusions and rapid appearance changes.

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Tracking Algorithms

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- Widely used for motion tracking.
- Handles noise and uncertainty in measurements.
- Suited for scenarios with simple motion patterns.
- Gap: Limited by linear assumptions and inadequate for complex motions.

Mean-shift Algorithms

- Based on iteratively shifting a window towards the mean of the feature distribution.
- Effective in tracking objects with complex, non-linear motion patterns.
- Gap: Sensitive to initialization and can drift in the presence of noise.

Graph-cut Optimization and Energy Minimization

- Handle complex tracking scenarios like multi-object tracking and occlusion.
- Formulate tracking as an optimization problem.
- Gap: Computationally expensive and may not be suitable for real-time applications.

Key Paper: Benfold and Reid

- Proposed a real-time head tracking algorithm.
- Combined detection and tracking for long-term trajectories in crowded scenes.
- **Gap:** Performance degrades in highly dense and cluttered environments.

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Feature Extraction and Object Segmentation Tracking Algorithms

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Deep Association Metric Learning

- Developed by Wojke et al.
- Learns a deep feature representation for object association across frames.
- Demonstrated excellent performance on challenging benchmarks.
- **Gap:** Requires large annotated datasets for effective training.

Deep Learning-Based Approaches

Joint Object Detection and Tracking

- Developed by Fang et al.
- Leverages object detection and tracking strengths.
- Applied to autonomous driving and video surveillance.
- Gap: High computational cost and complexity in real-time scenarios.

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Key Findings

- Robust feature extraction and object segmentation are crucial.
- Advanced tracking algorithms handle complex scenarios.
- Deep learning provides significant advancements but has challenges.

Discussion

Future Directions

- Real-time performance: Essential for many applications.
- Robustness to occlusion and lighting changes: Important for reliable tracking.
- Integration with other vision tasks: Enhances comprehensive visual understanding.
- Comprehensive benchmarking: Facilitates progress and innovation.

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Conclusion

- Significant progress has been made in MDT algorithms and applications.
- Future research should focus on real-time performance, robustness, and integration.
- This review serves as a valuable resource for researchers and practitioners.

Thank You!

Conclusion