Cell-Tracking and Classification with Transient Fluorescent Signals

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Introduction

- Time-lapse **Cell Tracking** is essential for studying cell division and death in biomedical research
- Critical for evaluating cancer cell response to treatments
- Unique challenges in comparison to traditional **Multi-Object Tracking (MOT)**:











- State-of-the-art methods from the Cell Tracking Challenge [1] are benchmarked on time-lapse videos with single, constant fluorescent signals, overlooking cell death events
- We present a hybrid tracking method that classifies **living** and **dead** cells and detects cell division events
- Overcoming the issues of transient signals by interpolating tracks with Kalman Filter [2] after re-identification and the usage of Low-Confidence detections

Method



Assume division if more than one possible candidate is found



Limitations

- CTC Metrics only consider continuous tracks
 - GT cells re-entering the frame require new ID assignments
- Silver-Truth bounding box annotations

Conclusion

- Detection remains the primary bottleneck
- Proposed methods mitigate challenges from transient signals
 - Incorporating a Kalman Filter and Low-Confidence detections improves performance
 - Kalman Filter alone achieves comparable MOT metric results to combining it with Low-Confidence detections, but lower CTC results

References

[1] https://celltrackingchallenge.net/

[2] Kálmán, R. E., A new approach to linear filtering and prediction problems, Journal of Basic Engineering (1960) [3] Zhu et al., Deformable DETR: Deformable Transformers for End-to-End Object Detection, ICLR (2021) [4] Zhang et al., ByteTrack: Multi-Object Tracking by Associating Every Detection Box, ECCV (2022) [5] Wang et al., SMILEtrack: SiMIlarity LEarning for Occlusion-Aware Multiple Object Tracking, AAAI (2024)