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Automatic Cell Tracking in Noisy Images for Microscopic Image Analysis

by

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September 2014

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Submitted in part fulfilment of the requirements for the
MSc degree in Computer Science (Artificial Intelligence) of Imperial College London

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6 Experimental results NEW

This feature has not yet been implemented

Explain the software used

Explain the computer used for the evaluation

Separation of experimental method and results

in experimental methods they explain the data and the datasets

See the cell population tracking and linear construction with spatiotemporal octet by Kang et al for a good results section

- tracking examples
- detection and tracking accuracy
- computation time

great Metrics: Research Article, Evaluating Multiple Object Tracking Performance: The CLEAR MOT Metrics

Think about which experiments to do, analyses, comparison with other methods if possible

6.1 Cell detector NEW

Figure 6.1 displays a temporal view of the detected cells. The vertical axis represents the frame of the sequence. The figure clearly shows that “cell tracks” are clearly discernible, even if the number of outliers is significant. For the tracking module it is better to have a higher recall than precision, as outliers can be much more easily discarded than segmented tracks linked.

6.1.1 Speed NEW

Measure the speed of detection in images of different sizes, and different number of cells

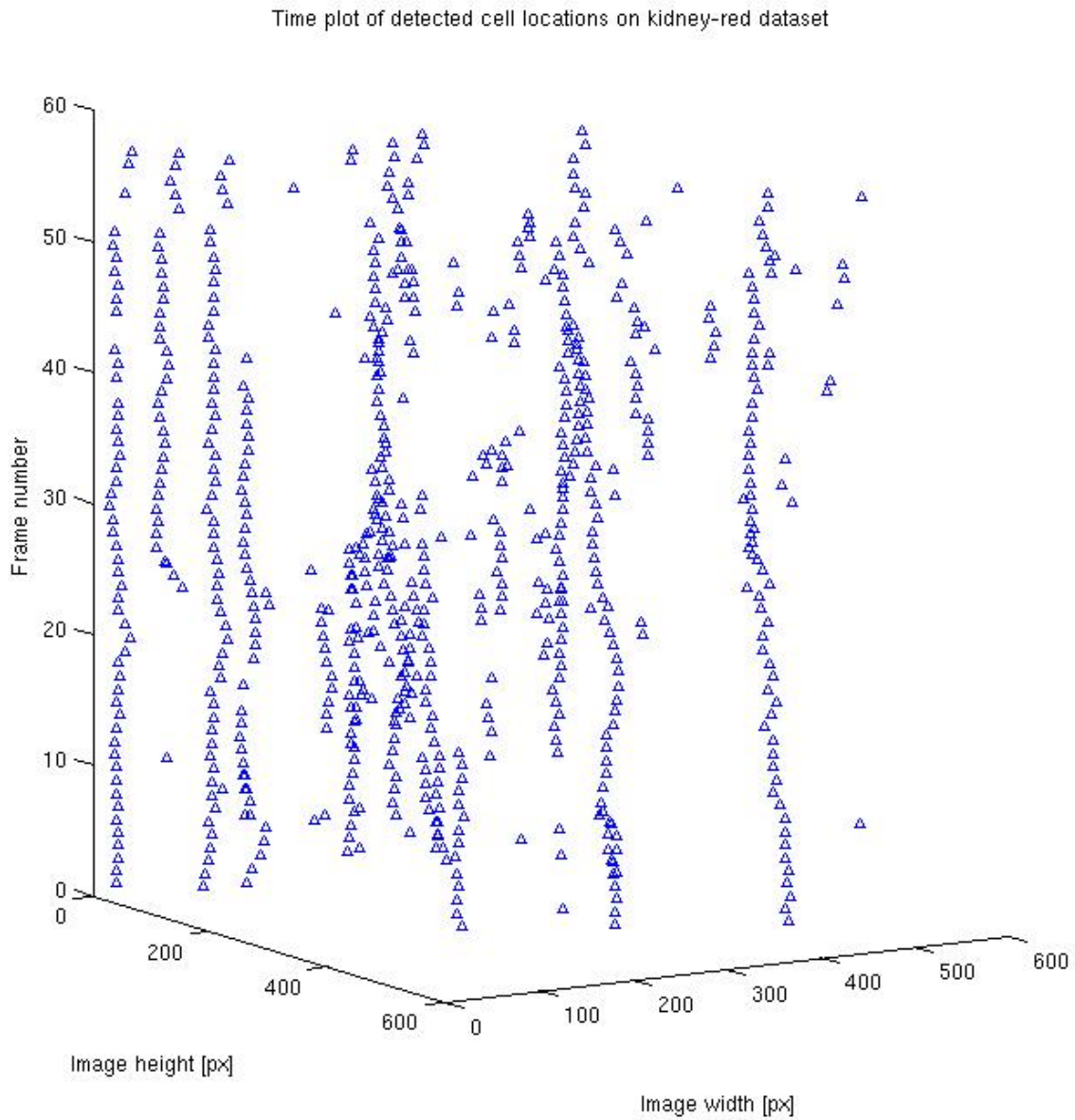


Figure 6.1: Cells detected over 60 consecutive frames are visualized as a time series. The vertical axis corresponds to the frames. Even in this raw detection data, it is possible to see the tracks of some of these cells.

6.1.2 Detection accuracy NEW

6.2 Cell tracker NEW

Define the different measures of accuracy

6.2.1 Performance metrics NEW

6.2.2 Speed NEW

Measure the speed of generating tracks, as a measure of per 1, 100, 1000 frames, depending on the number of tracks

6.2.3 Tracking accuracy

6.3 Limitations and areas of improvement NEW

Answer: what, why, how to improve in future

- detection training: only first few frames of datasets, not random – expect to detect later frames worse
- testing on only long datasets: no data on short datasets. difficult to train (what to link?), difficult to annotate
- speed of detector. Reduce number of hypothesis

6.4 Summary NEW

Appendices

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