Research Proposal

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1. What is the problem you would like to solve?

   Recently increase use of location-aware devices (such as GPS, Smart-phones and RFID tags) has allowed the collection of a vast amount of data with a spatial component linked to them. Different studies have focused in analyzing and mining this kind of collections [Leung, 2010][Miller and Han, 2001]. In this area, trajectory datasets have emerged as an interesting field where diverse kind of patterns can be identified [Zheng and Zhou, 2011][Vieira and Tsotras, 2013]. For instance, authors have proposed techniques to discover motion spatial patterns such as moving clusters[Kalnis et al., 2005], convoys[Jeung et al., 2008] and flocks [Benkert et al., 2006][Gudmundsson and van Kreveld, 2006]. In particular, [Vieira et al., 2009] and [Turdukulov et al., 2014] propose two novel algorithms to find moving flock patterns in very large spatio-temporal datasets.

2. Why is it important?

   A flock pattern is defined as a group of entities which move together for a defined lapse of time [Benkert et al., 2006]. Applications to this kind of patterns are diverse and range from surveillance to integrated transport systems. For example, [Turdukulov et al., 2014] explore the finding of this class of patterns to discover similarities between tropical cyclone paths. Also, [Calderon Romero, 2011] finds moving flock patterns in iceberg trajectories to understand their movement behavior and how they related to changes in ocean's currents.

3. What is your plan/outline of your solution?

   The algorithms proposed by [Vieira et al., 2009] and [Turdukulov et al., 2014] share the same initial strategy to detect flock patterns. In that, first they find clusters of points which could be close enough to initiate a flock for each time interval. This is a costly operation due to the large number of points and intervals to be analyzed. The technique uses a grid-based index and a stencil (see figure 1) to speed up the process but the complexity is still high. My plan is to allow individual threads to compute each of the stencils in the grid in parallel.
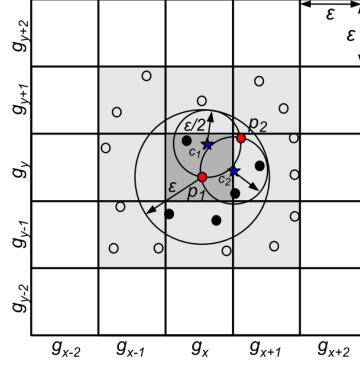
Figure 1: Grid-based index used in [Vieira et al., 2009].

4. How do you propose to study it?

   I already have access to sequential implementations of the algorithms in Java and Python, so it would not be difficult to code a port in C. Then, I plan to use a convolution kernel to find clusters of points as is required. Finally, I hope to compare both implementations to find some improvements.

5. What do you expect to find out?

   Certainly, the computation of each stencil in parallel should improve the overall performance of the algorithm.

References

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