MOVEMINE Mining Moving Object Databases

IBM Almaden

New York State Museum

Director of MoveBank.org

University of Illinois at Urbana-Champaign

Motivation With the maturity of tracking technologies, increasing amounts of movement data become widely vailable, such as vehicles, animals, climate, and human movement data. MoveMine, is designed for sophisticated moving object data mining by integrating several useful functions including pattern mining and trajectory mining using the state-of-the-art techniques. Our system is tested on various real movement data sets, such as those provided by MoveBank.org (an international organization of biologists). It will benefit people to carry out versatile analysis on these kinds of data.





Step 1: Detect the regions that are frequently Step 2: Detect periods for each reference visited (reference spot).







"This bald eagle stays in New York area (i.e., reference spot # 1) from December to March. In March, it flies to Great Lakes area (i.e., reference spot # 2) and stays there until the end of May. It flies to Quebec area (i.e., reference spot # 3) in the summer and stays there until late September. Then it is a back to rest. Laber at this decident of the decident lies back to Great Lake again staying there from mid October o mid November and goes back to New York in Dec

Step 3: Summarize periodic behavior.

Trajectory clustering focuses more on the geometric information of movements. It discovers the clusters of sub-trajectories.



arm Pattern (submitted) v.s. Convov Pattern

Both swarm and convoy patterns are trying to find the moving objects that move together. Two moving objects are considered "together" at one timestamp if they belong to the same cluster at that timestamp. The goal of convoy is to find objects that move together for at least k consecutive times. Swarm, which is more practical in real application, tries to discover the objects that are close for min_t non-consecutive times. Technique challenge for swarm mining is the exponential search space up to 2^{# of objects}*2^{# of timestamps}.

Convoy can be solved in polynomial takes 109 millisecond(s) to Dataset: Swainsoni ind all the convoys time. Swarm is an exhaustive search here are 1 convoys. mber of individuals: 43 Convoy#1 · Objects problem which requires strong pruning 'ime span: /29/1995 12:00:00 AM "Caroline ∘ "Kiki' rules. . 24/1998 12:56:00 AN $\begin{array}{c} \textcircled{\texttt{S}} & & & & \\ \mathcal{O}: \{\alpha_{i}\} & & & \\ \mathcal{O}: \{\alpha_{i}\} & & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}, t_{1}\} & & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}, t_{1}\} & & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}, t_{1}\} & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}, t_{1}\} & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}, t_{1}\} & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{1}\} & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}, t_{2}\} & & \\ \mathcal{T}_{max}(\mathcal{O}): \{t_{1}, t_{2}\} & & \\ \mathcal{T}_{max}(\mathcal{O$ It could be huge search min_t=0.2.Limited $O : \{o_i, o_k, o_l\}$ $O : \{o_2, o_k, o_l\}$ $T : \{o_2, o_k, o_l\}$ $\begin{array}{c} O: \{o_i, o_2, o_3\} \\ T_{mit}(O): \{\} \end{array} \qquad \qquad \begin{array}{c} O: \{o_i, o_2, o_3\} \\ T_{mit}(O): \{l_1, l_2, l_3\} \end{array}$ 2^43*2^1000. convoy are found. The node and its subtree are pruned by Apriori Pruning Rule gle 20 mi min_t = 0.5. More swarms are discovered. Timestamps are not consecutive. The Zhenhui Li, Bolin Ding, Jiawei Han, and Roland Kays, "Swarm: Mining Moving Object Clusters", In Submission. Hoyoung Jeung, Man Lung Yiu, Xiaofang Zhou, Christian S. Jensen, Heng Tao Shen, "Discovery of Convoys in Trajectory Databases", PVLDB, 2008.

Jae-Gil Lee, Jiawei Han, and Kyu-Young Whang, "Trajectory Clustering: A Partition-and-Group Framework", **SIGMOD**, 2007

rajectory Outlier (ICDE'08

This function discovers the sub-trajectories that do not follow the general trend of the movements.

> Two phases: partitioning and detection TR_3 TR_4 TR_7 TR_7 Partition A set of trajectory partitions Detect TR₃ Outlying trajectory partitions

Jae-Gil Lee, Jiawei Han, and Xiaolei Li, "Trajectory Outlier Detection: A Partition-and-Detect Framework", ICDE, 2008.