



PATH OVERLAP DETECTION AND PROPERTY GRAPH CONSTRUCTION FROM TCX DATA FOR ADVANCED ANALYSIS

IEEE 23rd World Symposium on Applied Machine Intelligence and Informatics (SAMI 2025)

Authors:

Tilen Hliš (<u>tilen.hlis@um.si</u>), Iztok Fister (<u>iztok.fister@proton.me</u>), and Iztok Fister Jr. (<u>iztok.fister1@um.si</u>)

PRESENTATION AGENDA

- 1. Motivation and Purpose of the Article
- 2. TCX Files Reading
- 3. Property Graphs
- 4. Path overlap Detection Algorithm
- 5. Experimental Results
- 6. Discussion and Future Work
- 7. Conclusion



MOTIVATION

Cycling Performance Optimization:
 Cyclists heavily rely on GPS data (speed, heart rate, distance) for training analysis and performance improvement.

Challenge of Data Complexity:
 TCX files store large amounts of unstructured data (GPS, heart rate, cadence), making advanced analysis difficult.

Need for Deeper Insights:
 Detecting overlapping paths in cycling sessions
 can reveal behavioral patterns and performance
 trends.



PURPOSE

- Efficient Data Analysis:

Develop a method to convert TCX data into property graphs for advanced analysis.

Path Overlap Detection:

Detect overlapping cycling paths using KDTreebased algorithms with tolerance for GPS inaccuracies.

Performance Insights:

Enrich detected segments with performance metrics (speed, heart rate, power, cadence,...) for better analysis.

Future Expansion:

Enable integration with Numerical Association Rule Mining (NARM) for predictive behavior modeling.



TCX DATA STRUCTURE

What is a TCX File?

- Training Center XML (TCX):
 Developed to store data from GPS-enabled fitness devices.
- Stores detailed workout data:
 - GPS Coordinates (latitude, longitude, altitude).
 - Performance Metrics (speed, heart rate, cadence, power).
 - Time-based session details.

TCX File Hierarchy

• Activity:

Overall session data (sport type, session metadata).

- Lap:
 - Segments of the session (time, distance, heart rate).
- Trackpoint:

Individual data points (timestamp, GPS, metrics).



TCX FILE



```
<TrainingCenterDatabase>
 <Activities>
   <Activity Sport="Biking">
    <Td>2023-10-19T10:38:47.0007</Td>
      <Lap StartTime="2023-10-19T10:38:47.000Z">
      <TotalTimeSeconds>5050.412</TotalTimeSeconds>
      <DistanceMeters>29305.5/DistanceMeters>
      <MaximumSpeed>10.039999961853027/MaximumSpeed>
      <Calories>571</Calories>
      <AverageHeartRateBpm>
        <Value>121</Value>
      </AverageHeartRateBpm>
      <MaximumHeartRateBpm>
        <Value>160</Value>
      </MaximumHeartRateBpm>
      <Intensity>Active</Intensity>
      <TriggerMethod>Manual</TriggerMethod>
      <Track>
        <Trackpoint>
             <Time>2023-10-19T10:38:47.000Z</Time>
             <Position>
                <LatitudeDegrees>46.3886827044189</LatitudeDegrees>
                <LongitudeDegrees>15.728210052475333
             </Position>
             <AltitudeMeters>237.60000610351562</AltitudeMeters>
             <DistanceMeters>0.0</DistanceMeters>
             <HeartRateBpm>
                <Value>91</Value>
             </HeartRateBpm>
             <Extensions>
                <ns3:TPX>
                  <ns3:Speed>0.0</ns3:Speed>
                </ns3:TPX>
             </Extensions>
           </Trackpoint>
```

TCX FILES READING

TCXReader.jl Library (https://github.com/firefly-cpp/TCXReader.jl)

- Custom-built Julia library for efficient reading and processing of TCX files.
- Optimized for handling large datasets with **high performance**.
- Converts raw TCX data into structured formats for further analysis.

struct TCXTrackPoint

time::DateTime

latitude::Union{Nothing, Float64}

longitude::Union{Nothing, Float64}

altitude_meters::Union{Nothing, Float64}

distance_meters::Union{Nothing, Float64}

heart_rate_bpm::Union{Nothing, Int}

cadence::Union{Nothing, Int}

speed::Union{Nothing, Float64}

watts::Union{Nothing, Int}

end



CONSTRUCTING THE PROPERTY GRAPH

Purpose of Property Graphs

- Provides a structured representation of TCX data for efficient storage and retrieval.
- Organizes GPS track points and related performance metrics for analysis.
- Facilitates the enrichment of detected overlapping segments with performance data (after detection using KDTree).

What is a Property Graph?

- A Directed Multigraph with nodes and edges storing rich data.
- Nodes (Vertices): Represent GPS track points from TCX files.
- Edges (Arcs): Connect consecutive points, forming the athlete's path.
- **Properties:** Store key-value pairs (e.g., speed, heart rate, cadence).

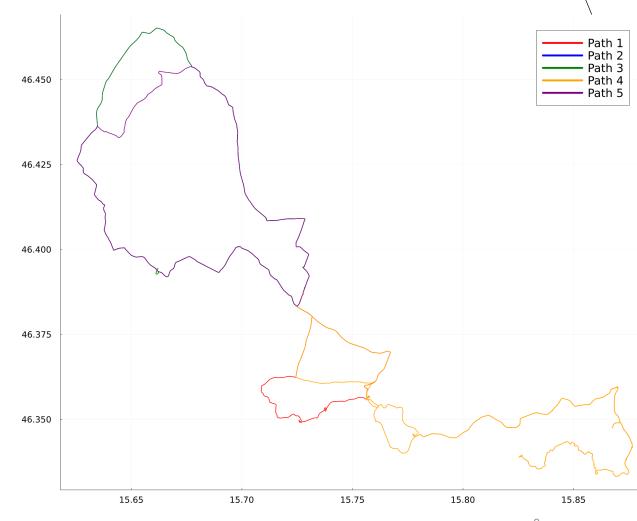


Formal Definition: $G=\langle N,A,K,V,\alpha,\kappa,\pi \rangle$

CONSTRUCTING THE PROPERTY GRAPH

Graph Construction Workflow

- 1. Parse TCX Files: using TCXReader.jl.
- **2. Add Nodes:** Each GPS track point becomes a graph vertex.
- **3. Connect Nodes with Edges:** Sequential points are linked to form a path.
- **4. Attach Properties:** Key metrics (speed, heart rate, cadence,...) are stored as properties of nodes and edges.



PATH OVERLAP DETECTION

Motivation for Path Overlap Detection

- Cyclists often repeat routes, leading to overlapping segments in TCX data.
- Detecting these overlaps helps analyze how performance changes across sessions.
- Understanding repeated segments provides insights into **training behavior** and **physiological responses**.

Challenges in Detection

- GPS Inaccuracies: Slight deviations in recorded data due to device errors.
- Path Variations: Small changes in route due to cyclist behavior or environmental factors.
- Efficient Processing: Large datasets require optimized algorithms for real-time analysis.

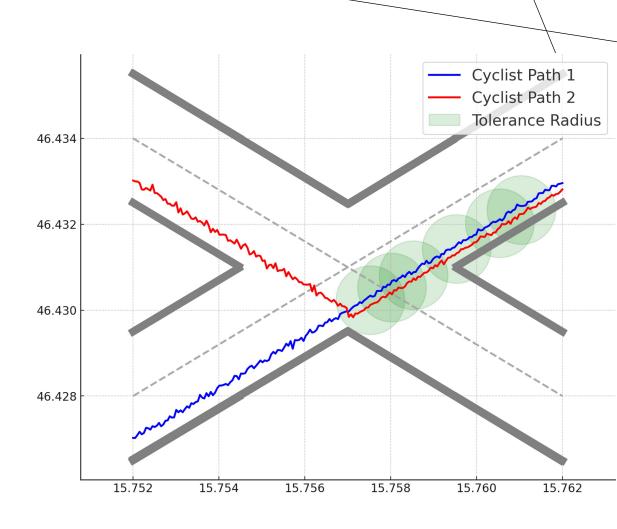
PATH OVERLAP DETECTION

Algorithm Overview:

- 1. Parse GPS data and build the KDTree.
- **2. Iterate through paths** to check for nearby points in other sessions.
- **3. Apply tolerance** to identify consistent overlaps.
- 4. Aggregate overlapping segments.

Advantages of the Method

- **Efficient Spatial Search** with KDTree for large datasets.
- Accurate Detection using a customizable tolerance to handle GPS noise.
- Scalable Solution for analyzing multiple sessions.



EXPERIMENTAL SETUP

Objective

- Evaluate the performance of the **Path Overlap Detection Algorithm**.
- Analyze how overlapping segments relate to athlete performance.

Dataset

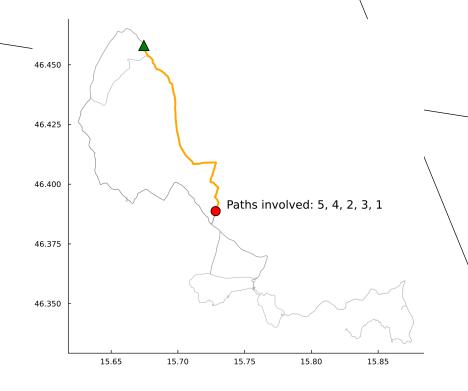
- Total Files: 462 TCX files.
- Selected for Analysis: 5 files from the same cyclist.
- Detected Overlapping Segments: 39 overlapping segments.
- Detailed Analysis: Segment 30 (5 rides) and Segment 18 (3 rides).



RESULTS: SEGMENT 30 (5 RIDES)

Observations

- **Higher variance** in speed and heart rate.
- External factors (e.g., wind) influenced performance (confirmed via Strava data).
- Consistent cadence indicates stable pedaling mechanics.

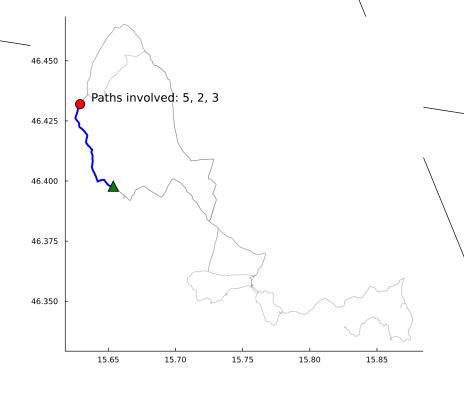


Metric	Min	Max	Mean	Median	Stdev
Speed (km/h)	22.5	25.2	24.16	24.8	1.19
Heart Rate (bpm)	143	165	151.6	147	9.48
Cadence (rpm)	79	86	83.4	85	2.88
Time (sec)	1,539	1,724	1,607.6	1,563	81.43

RESULTS: SEGMENT 18 (3 RIDES)

Observations

- Consistent performance across sessions.
- Low variance suggests minimal external impact.
- Stable heart rate and cadence throughout rides.



Metric	Min	Max	Mean	Median	Stdev
Speed (km/h)	21.5	22.3	21.93	22.0	0.40
Heart Rate (bpm)	162	166	163.67	163	2.08
Cadence (rpm)	75	81	78.33	79	3.06
Time (sec)	910	945	926.67	925	17.56

DISCUSSION

Effectiveness of the Algorithm:

The KDTree-based method effectively detected all overlapping segments with zero false positives/negatives.

Tolerance Factor:

Implementing a tolerance threshold allowed for **accurate detection** despite GPS inaccuracies.

• Performance Insights:

Analysis of segments provided insights into how external factors (e.g., wind) and physical condition influenced performance.

Property Graph Utilization:

The property graph effectively organized TCX data, enabling easy retrieval of performance metrics for detected segments.



FUTURE WORK

1. Integration with ARM:

Incorporate Association Rule Mining (ARM) for pattern discovery across multiple cyclists.

2. Evolutionary Algorithms:

Integrate **Differential Evolution (DE)** and **Particle Swarm Optimization (PSO)** from NiaARM.jl for rule identification.

3. Handling Larger Datasets:

Optimize the algorithm for large-scale, multi-user data analysis.

4. External Factors Analysis:

Include environmental data (e.g., weather, terrain) to assess external impacts on performance.

5. Expansion Beyond Sports Applications:

Adapt the algorithm for more **general use cases** beyond cycling and sports, enabling applications in areas like **transportation**, **logistics**, and **urban planning**.



THANK YOU FOR YOUR ATTENTION

