

Robust Timestamp Correction for Multi-Camera Trajectory Reconstruction in Road Safety Analysis

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Why This Matters: Road Safety & Vision Zero



Vision Zero

Prevent fatalities through near-miss detection.



Multi-Camera Systems

Fuse views for better trajectory reconstruction.



Surrogate Safety Measures

Measure conflict risk before crashes happen.

TTC (Time-to-Collision): time remaining until two users collide at current speeds

PET (Post-Encroachment Time): time gap after one user leaves a conflict zone before the next enters

Problem Statement: Timestamp Anomalies in Multi-Camera Data

Expected

$t_1 < t_2 < t_3 < \dots$ Strictly increasing, uniquely indexed

Observed in MIA STRIKE data

Timestamp freezes

Duplicated metadata

Variable metadata

Origin unclear

Downstream Impact

Fragmented trajectories

Stalled timestamps break cross-camera track links

Corrupted TTC & PET

$\Delta t = 0$ causes division by zero or ambiguous ordering

Silent failure

No error is raised - analysis continues

Case Study: MIA STRIKE Real-World Dataset

Dataset at a Glance

Project: MIA STRIKE (Mobiliteit Innovatief Aanpakken)

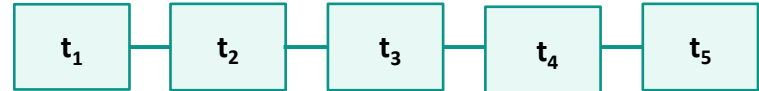
Location: Urban roadside intersection, Belgium

Setup: Two-camera roadside system

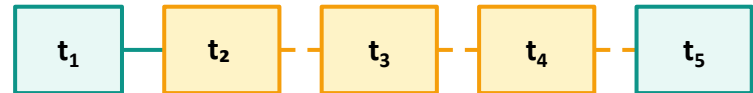
Goal: Proactive traffic conflict & near-miss detection

Anomaly: Recurring timestamp freezes (100–400 ms)

Expected: Strictly increasing timestamps



Observed: Frozen / Duplicated timestamps



Frames 2–4: visually different content but timestamp frozen

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Anomaly:	Recurring timestamp freezes (100–400 ms)

Same timestamp - visually different frames

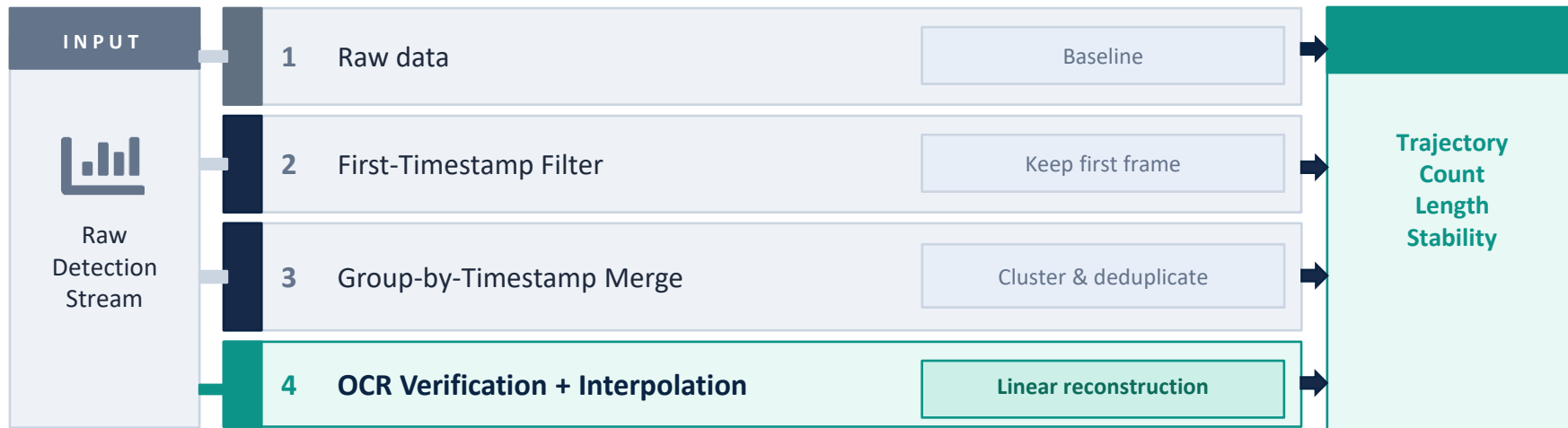
Frame A | t = 14:47:42.273128



Frame B | t = 14:47:42.273128 (car has moved!)



Methodology: Timestamp Correction Pipeline



← All four processed streams evaluated in parallel →

Trajectory Count

No. of unique trajectory IDs reconstructed

Avg Trajectory Length

Mean frames per tracked object

Motion Stability

Speed coefficient of variation (CV)

Four Correction Strategies



Raw Baseline

No correction applied

Fragmented trajectories & speed spikes

RAW



First-Timestamp Filter

Keep first frame per duplicate timestamp

Discards valid new detections

FT

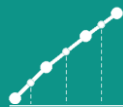


Group-by-Timestamp Merge

Cluster & deduplicate frames at same t

Timestamp error remains uncorrected

GT



Timestamp Interpolation

Linearly reconstruct frozen t — restores valid timeline

Best result · restrict to gaps < 1 s

IP

Each strategy applied to the same raw stream → results compared on trajectory count, length, and speed stability

OCR Verification



Root Cause Diagnosis

1

Extract embedded timecodes via OCR

2

Compare against system timestamps

3

Both frozen frames still differ

4

Fault: camera clock, not detection

Timestamp Interpolation

$$\hat{t} = t_a + [(j - a + 1)/(b - a + 2)] \cdot (t_{b+1} - t_a)$$

Result

Strictly Increasing
 $\hat{t}_a < \hat{t}_{a+1} < \dots < \hat{t}_b$

Range

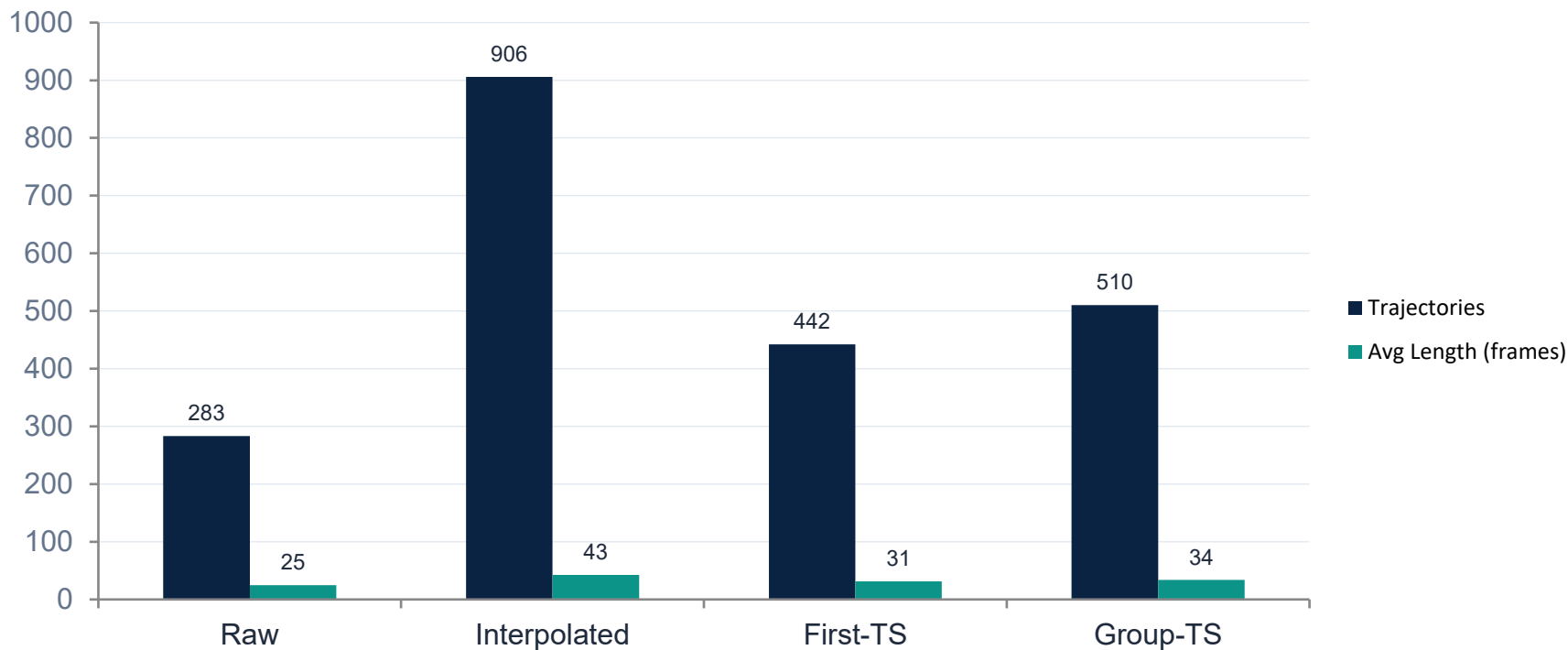
Freeze intervals 100 – 400 ms

Threshold

Short gaps only (< 1 s)

Results: Trajectory Count & Average Length

▲ +220% more trajectories | +73% longer average length vs. Raw



Results: Trajectory Count & Average Length

906

trajectories
(Interpolated)

42.5

avg frames/object
(Interpolated)

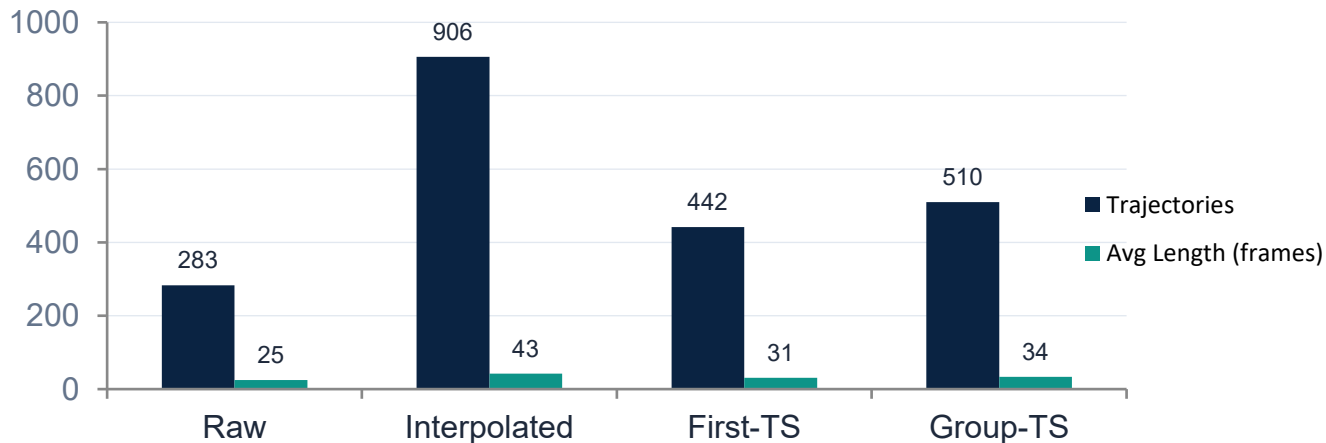
283

trajectories
(Raw)

24.6

avg frames/object
(Raw)

▲ +220% more trajectories | +73% longer average length vs. Raw



Results: Trajectory Count & Average Length

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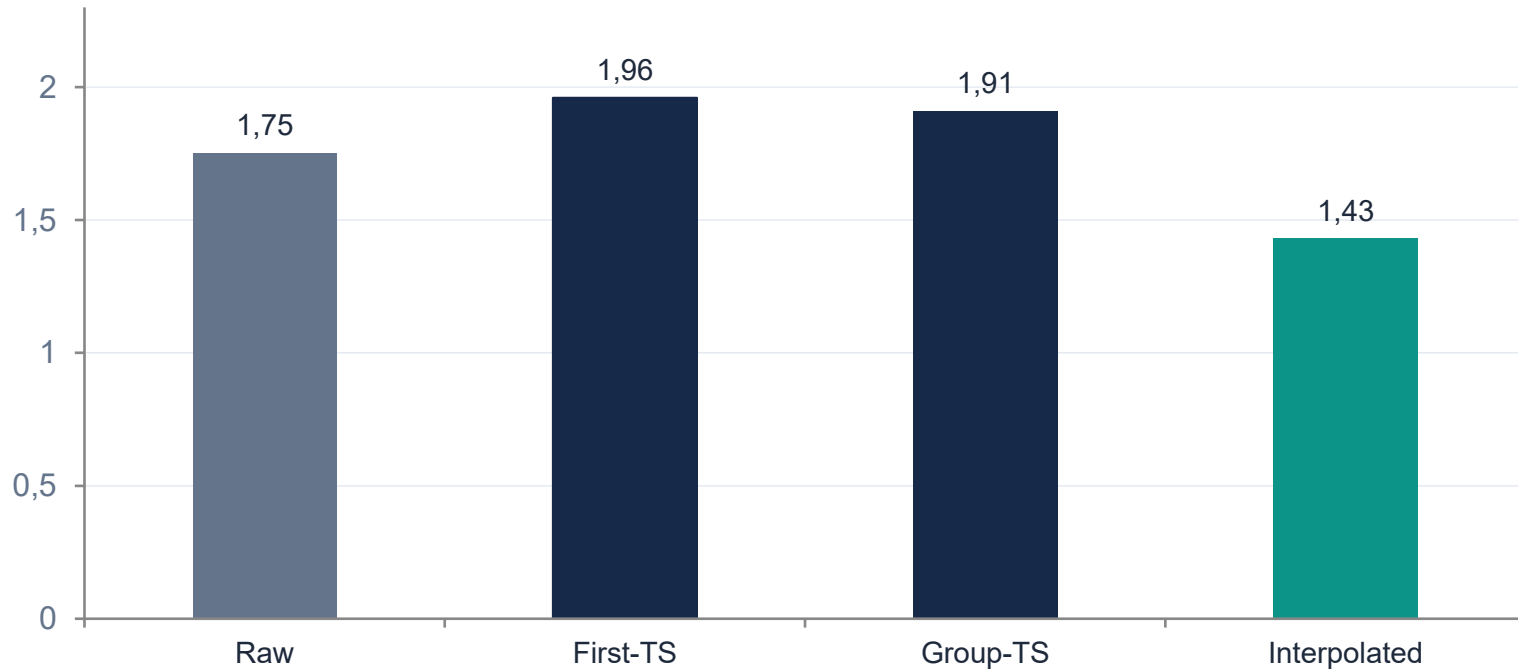
avg frames/object
(Raw)

Interpolated vs Raw trajectories - real MIA-STRIKE footage



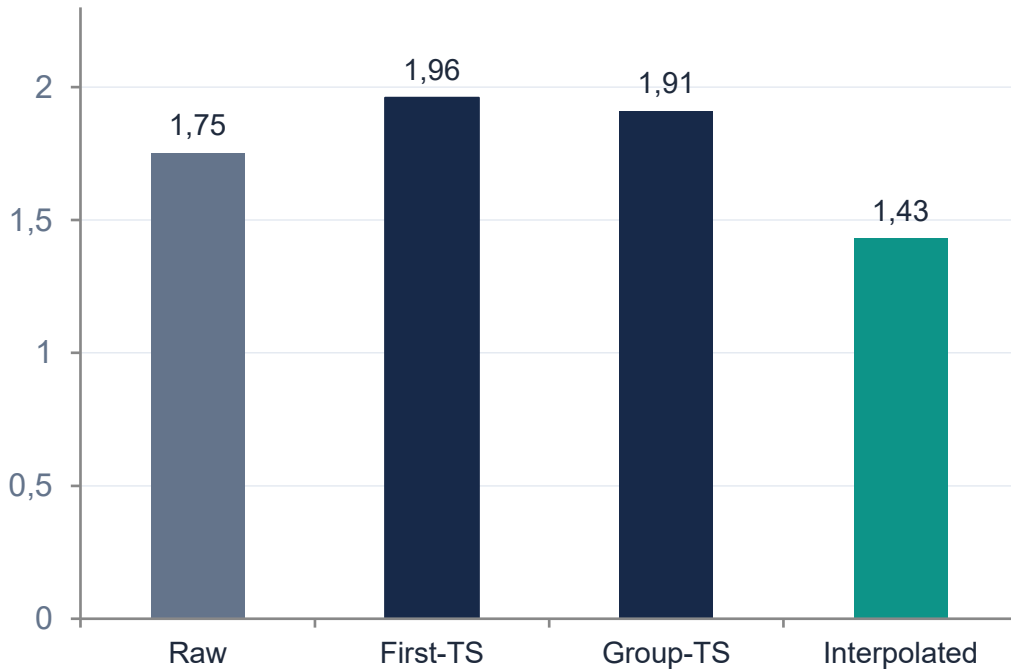
Results: Motion Stability & Speed Consistency

Lower CV = smoother, more reliable speed estimates



Results: Motion Stability & Speed Consistency

Lower CV = smoother, more reliable speed estimates



1.43

Interpolated - lowest CV

Smoother speed, fewer discontinuities

1.75

Raw - baseline

Speed resets from stalled timestamps

1.96

First-TS - worst

Information loss amplifies noise

Safety Impact: How Timestamp Faults Corrupt TTC & PET

Timestamp faults propagate silently into both safety measures - invalidating results without raising any error.

TTC — Time-to-Collision

Requires

Instantaneous velocity from consecutive Δt

Fault

Frozen $\Delta t = 0 \rightarrow$ division by zero \rightarrow meaningless TTC

Fixed by

Interpolation restores non-zero Δt between every frame pair

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PET — Post-Encroachment Time

Requires

Accurate temporal ordering across camera streams

Fault

Duplicate timestamps \rightarrow ambiguous ordering \rightarrow incorrect PET

Fixed by

Interpolation enforces strict ordering, resolving cross-camera ambiguity

Practical Implications & Future Work

Recommendations for Agencies

1

Validate timestamps before any analysis

2

Use OCR to detect embedded clock freezes

3

Interpolate only for gaps < 1 s

4

Flag corrected intervals in output metadata

Future Work

Formal interpolation constraints

Establish formal constraints on interpolation - define acceptable temporal-gap thresholds relative to frame rate and motion dynamics.

Conclusion

Temporal recalibration is essential for trustworthy trajectory-based road safety analysis.

01

Root Cause Found

OCR verification confirmed timestamp faults originate in the camera clock, not the detection pipeline.

02

Best Method: Interpolation

906 vs 283 trajectories (+220%). 42.5 vs 24.6 avg length. Speed CV 1.43 vs 1.75.

03

Practical Guidance

Validate timestamps before any safety analysis. Interpolate only for short gaps (< 1 s).

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