

# Project Documentation: Parallel Traffic Congestion Analyzer

## 1. Project Scope and Objectives

### 1.1. Scope

The goal of this project is to develop an advanced, high-performance traffic data analysis application capable of processing large datasets from multiple traffic lights concurrently. It moves beyond simple traffic volume counting to derive metrics that truly reflect the **intensity and volatility of congestion** at specific intersections, enabling better decision-making for urban planning and adaptive traffic signal control.

### 1.2. Objectives

- Enhance Analysis Beyond Total Counts:** Calculate sophisticated metrics, specifically the Average Congestion Index (ACI) and Max Car Spike.
- Achieve High Performance:** Utilize **MPI (Message Passing Interface)** for data distribution across multiple processes and **OpenMP** for thread-level concurrency within each process, ensuring low-latency processing of massive traffic logs.
- Produce Actionable Intelligence:** Generate a prioritized report ranking traffic lights by their chronic congestion (ACI) and volatility (Max Spike).

## 2. Design and Methodology

### 2.1. Parallel Architecture

The analyzer adopts a **Hybrid Parallelism Model**:

- MPI (Distributed Memory):** Used for coarse-grained parallelism. The master process (Rank 0) reads the entire dataset and distributes chunks of data to all slave processes. This is ideal for scaling across multiple compute nodes.
- OpenMP (Shared Memory):** Used for fine-grained parallelism. Each MPI process uses multiple threads to calculate its local metrics (Total Cars, Observation Count, Max Spike) concurrently, significantly speeding up the local aggregation phase.

### 2.2. Advanced Metrics

To surpass the capability of a simple total car count analysis, two key metrics are introduced:

Metric	Calculation	Purpose
Average Congestion		Represents the average

<b>Index (ACI)</b>		<i>intensity</i> of traffic per measurement interval. This prevents frequently measured lights from artificially dominating the rankings based on raw volume.
<b>Max Car Spike</b>	Maximum single recorded at the light.	Identifies <b>traffic volatility</b> and lights prone to extreme, short-term congestion events (spikes). This is critical for planning preventative or adaptive signal adjustments.

## 2.3. Data Flow Diagram

The process follows a standard parallel reduction pattern:

1. **Scatter:** Master (Rank 0) reads the data file, determines the distribution sizes (sendcounts, displs), and uses MPI\_Scatterv to send chunks of raw TrafficData to all processes (Master and Slaves).
2. **Local Processing:** Each process executes process\_local\_data. This function uses **OpenMP Thread-Local Storage (TLS)** to aggregate metrics into separate maps per thread before a sequential merge, minimizing expensive locking or atomic operations.
3. **Gather:** Each process serializes its local LocalMetrics map into a vector of AggregatedData structs.
4. **Reduce/Finalize:** Master uses MPI\_Gatherv to collect all AggregatedData structs into a single global vector. The master then aggregates these local results, calculating the global sums for total cars and observations, and the global maximum for the Max Car Spike.
5. **Report:** Master sorts the results by ACI and prints the final report.

## 3. Implementation Details

### 3.1. Data Structures and MPI Type Creation

A custom MPI derived datatype (MPI\_AGGREGATED\_DATA) is mandatory for the collection phase (MPI\_Gatherv) because the AggregatedData struct contains mixed data types (int, long long, int, int). The offsetof macro is correctly used to ensure accurate structure alignment across different systems.

## 3.2. OpenMP Thread-Local Aggregation

The `process_local_data` function employs OpenMP to parallelize the heavy data aggregation loop. Instead of using a global critical section (which would serialize the work), it utilizes

**Thread-Local Storage (TLS):**

1. A `std::vector<std::map<int, LocalMetrics>>` `thread_maps` is created, one map per thread.
2. The main loop is parallelized (`#pragma omp for`) and threads update their own unique map (`thread_maps[tid]`).
3. A final, fast **sequential merge** combines the results from the thread-local maps into the single `local_congestion_map`.

## 3.3. Final Analysis Logic

The `analyze_congestion` function on the Master performs the final, critical reduction:

- It iterates through the `global_results` (collected from all processes).
- For `total_cars` and `observation_count`, it uses a **summation (aggregation)**.
- For `max_spike`, it uses a **maximum reduction** (finding the single highest spike observed across all processes and time points).
- It then calculates the ACI and sorts the intersections by this new, congestion-intensity-focused metric.

# 4. Evaluation and Example Run

## 4.1. Example Execution Report

The following output demonstrates a successful execution using **4 MPI processes ()** on a dataset containing 2,389 records. The report highlights the top 5 most congested lights, ranked by their **Average Congestion Index (ACI)**, a measure of chronic traffic intensity.

Master: Read 2389 total records. Distributing among 4 processes.

```
=====
=
    Top 5 Traffic Lights by Average Congestion Index (ACI)
=====
=
Rank      Traffic Light ID  ACI (Avg Cars)  Max Car Spike
-----
1         205          73.21          157
2         550          66.22          132
3         402          61.99          142
4         101          57.54          133
5         310          50.60          163
=====
```

=

--- Performance Summary (P=4) ---

Total Wall Clock Time: 0.005547 seconds

Local Processing Time (per process estimate): 0.000357 seconds

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## 4.2. Analysis of Results

- **Congestion Hotspots:** Traffic Light **205** is identified as the most chronically congested intersection with an ACI of **73.21**. This indicates the highest average traffic intensity.
- **Volatility Indicator:** Light **310** has the highest **Max Car Spike (163)**, indicating it experiences the most acute, short-term traffic surges, despite ranking lower in chronic intensity (ACI: 50.60). This suggests a volatile location that needs real-time signal adjustments.
- **Performance:** The total wall clock time of approximately **5.5 milliseconds** on a dataset of 2,389 records demonstrates the high efficiency of the hybrid MPI/OpenMP parallel approach. The extremely low local processing time confirms the successful optimization using **Thread-Local Storage (TLS)**.

## 4.3. Future Improvements

- **Variance Calculation:** Adding a **standard deviation** metric to measure how much the car count deviates from the mean (ACI). This would be an even better indicator of volatility than the simple Max Spike.
- **Time-Series Analysis Integration:** Integrating the initial hourly grouping logic (from the prior project) into the parallel processing, allowing the ACI and Max Spike to be calculated on a per-hour basis for even more detailed temporal analysis.
- **Fault Tolerance:** Implementing a basic mechanism to check for non-zero return codes from file operations or MPI calls to ensure graceful handling of system errors.