



# Automatic Generation of Slovenian Traffic News for RTV Slovenija

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## Abstract

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## Keywords

Generating traffic reports, Large Language Models, NLP, Prompt Engineering, Fine-tuning, Slovenian traffic news, Automated text generation

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## Introduction

Traffic reporting is a crucial aspect of public broadcasting, especially for real-time updates on road conditions. Currently, RTV Slovenija relies on students to manually check, filter, and type reports from the Promet.si portal every 30 minutes. This process is time-consuming and prone to inconsistencies.

This project aims to automate traffic news generation using a Large Language Model (LLM). The approach includes leveraging prompt engineering techniques, defining evaluation criteria, and fine-tuning an LLM to improve accuracy and relevance. The generated reports must align with RTV Slovenija's guidelines, ensuring clarity, conciseness, and correctness in road naming and event significance.

## Existing Solutions and Related Work

### Traffic News Automations

Automated traffic reporting systems have been developed worldwide, primarily using Natural Language Processing (NLP) and Machine Learning (ML) techniques. Some existing solutions include:

1. Google Maps Traffic Alerts: Uses real-time data and machine learning to generate concise traffic updates
2. Waze Traffic Reports: Crowdsourced traffic conditions, automatically summarized for users.
3. AI-driven News Generation: Organizations like OpenAI and Google have explored AI-generated news reports in sports, finance, and weather domains.

These solutions utilize structured traffic data and LLMs for automated summarization. However, fine-tuning on local

datasets, such as Slovenian road networks and traffic terminology, remains a challenge.

### Large Language Models in Text Generation

Recent advancements in LLMs like GPT-4, LLaMA, and T5 have demonstrated strong capabilities in text summarization and structured data interpretation. Studies have shown that fine-tuning domain-specific models improves text coherence and factual accuracy.

## Initial Corpus Analysis

The dataset for this project consists of structured traffic data from Promet.si, formatted in Excel. A preliminary analysis reveals the following key attributes:

Road Names and Locations: Entries specify highways, regional roads, and key urban intersections.

Incident Types: Traffic congestion, road closures, weather-related disruptions, and accidents.

Time Stamps: Time of report updates, crucial for real-time relevance.

Severity and Impact: Differentiates minor delays from major traffic disruptions.

Initial observations suggest that filtering out redundant or low-impact reports will be necessary to maintain concise news summaries. To achieve this, categorizing incidents based on importance will help prioritize critical updates. Categories could include:

1. High Priority: Major accidents, full road closures, severe weather disruptions

2. Medium Priority: Traffic congestion, partial road closures, significant delays
3. Low Priority: Minor roadworks, temporary disruptions with minimal impact

This structured approach will enhance the efficiency of traffic news generation by ensuring that significant events receive appropriate attention.

## Initial Ideas and Approach

### Prompt Engineering

The first step involves experimenting with LLMs through structured prompts to generate effective traffic summaries. Potential techniques include:

- Using role-based prompting (e.g., “You are a Slovenian traffic news reporter”)

- Applying structured templates to ensure consistent output

- Fine-tuning prompt phrasing to optimize summarization quality

### Evaluation Criteria Definition

To ensure quality traffic reports, evaluation metrics will focus on:

1. Accuracy: Correct road names and locations
2. Conciseness: Keeping reports within an optimal word count
3. Format: Ensuring the report fit the provided format

### Parameter-Efficient Fine-Tuning

Once prompt engineering is refined, fine-tuning an LLM (such as GPT-3.5 or T5) with a domain-specific dataset will be explored. LoRA (Low-Rank Adaptation) or Adapter-based methods will be considered to optimize training efficiency.

### Interactive Testing Interface

A web-based or command-line interface will be developed to allow interactive testing of generated reports before full automation.

## References