Enhancing Communication with Motorists through

GogoGuru

Your Personalised Co-Pilot



Driving in Singapore: A Daily Challenge

Two major sources of frustration:

1. Fragmentation of information

Real-time traffic information, such as traffic updates, public transit schedules, parking availability, and route planning services, are often fragmented across different sources and maintained by multiple entities.

As a result, users often need to use multiple apps/ websites to look for the information they need, resulting in inconvenience.

2. Unpredictability of traffic conditions

Accidents, road closures, and unpredictable events can disrupt even the most meticulously planned journeys. Users often become aware of these disruptions only when they are already on their journey.

Users may also lack visibility on impending bad weather conditions, such as heavy rain, which could impact driving safety.

The lack of advance warning can lead to frustration, delays, and sometimes even safety hazards.

Overview of Flow

Backend AI Algorithm

User starts
Telegram bot using
/start command

User input:

"How do I go from Changi Airport to the Zoo? I want the cheapest route"

LLM for Intent Recognition

LLM is able to understand the user's intended destination, current location, and any user preferences.

Telegram bot
automatically prompts the
user for destination and/or
location, if it is not
detected

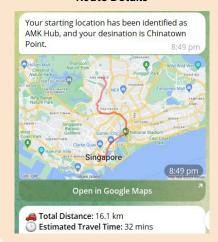
Triggers Google Maps API

to search for the best route and return route details (e.g. distance, estimated travel time)

Triggers LTA and NEA API

to search for real time traffic updates and weather conditions **specific to the** routes.

Route Details



Real Time Advisory

Real-Time Traffic Conditions:

There are potential issues along the route. Please exercise caution and prepare for an increased journey time.

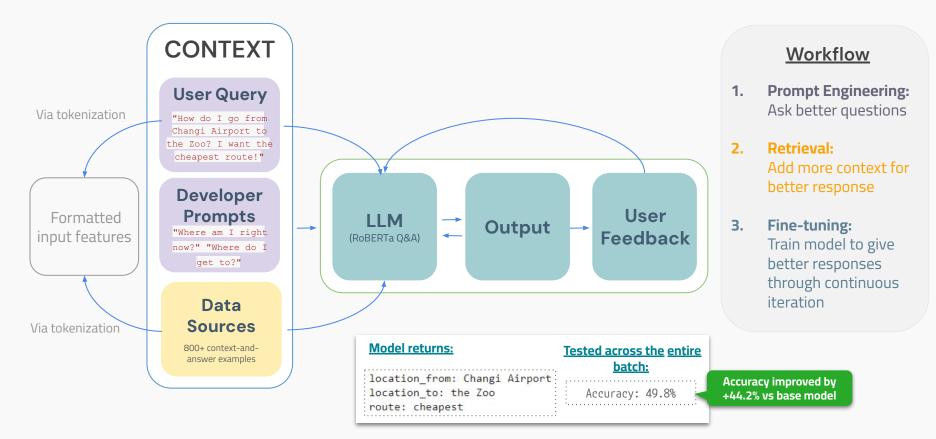
(20/3)20:12 Obstacle on PIE (towards Tuas) after CTE(City) Exit.

Real-Time Weather Updates:

There is current heavy rain along Bishan Road. Please drive carefully.

Tech Stack 1: LLM workflow

Continuously iterate with better prompts, better context, and/or adapt model with better responses



Tech Stack 1: LLM fine-tuning

Fine tuning model using local Singaporean context and greater diversity of queries

Initial baseline model using pre-trained HuggingFace QnA to test overall workflow

Did not adapt well to complicated prompts and local phrasing

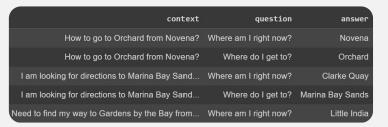
```
def qna(question, context):
    QA_input = {
        'question': question,
        'context': context
}

res = qa_model(QA_input)
return res

def from_to_qna(context):
    FROM = 'Where is the user coming from?'
    TO = 'Where is the user destination?'

from_resp = qna(FROM, context)
to_resp = qna(TO, context)
return from_resp['answer'], to_resp['answer']
```

Trained model using custom dataset (100+ samples) inspired from common queries and popular Singapore locations



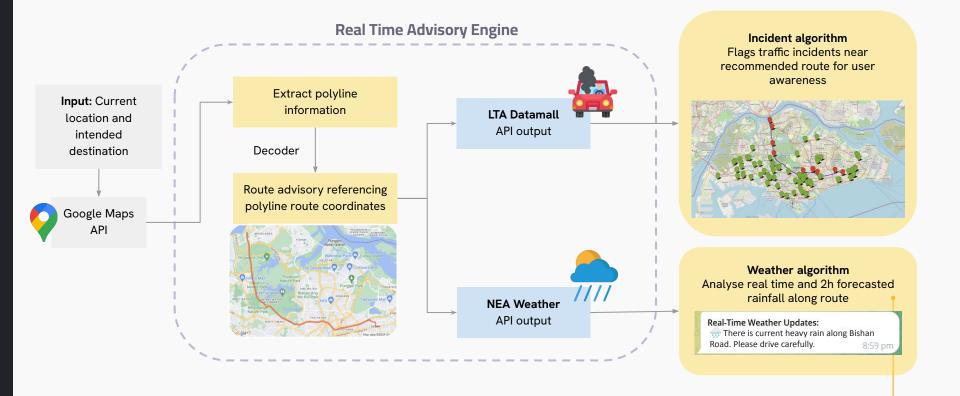
Increased dataset complexity (800+ samples) with common Singaporean phrasing and more dataset variety



Accuracy 5.6%

Accuracy 49.8% Improvement of +44.2% against base model

Tech Stack 3: Real-Time Analysis



Product Demo: Helping the "Busy Bee"





