

• For now, we are focusing on Case Study 1. So, you can skip next few slides and concentrate from <a href="Slides 08">Slides 08</a>-14

# Optimizing last-mile delivery

## **Research Outline**

- Introduction
  - Background
  - RQ
- Method
  - Proposed methodology
  - Solution algorithm
- Computational Result
- Discussion

## Background & Motivation

- Last-mile delivery is the most time-consuming and costly part of the supply chain, making its optimization crucial for competitive advantage.
- To meet rising e-commerce demands, logistics companies are turning to innovative solutions like drone delivery to improve efficiency.
- This paper addresses the challenges of optimizing large-scale last-mile drone delivery by using a column generation approach to solve the Vehicle Routing Problem with Drone (VRP-D).
- Extensive experiments show that the proposed method improves delivery efficiency and reduces costs, offering valuable insights for optimizing lastmile logistics with advanced technologies.



## **Research Questions**

#### **Model itself**

- What novel mathematical models can be developed to incorporate multi-modal transportation (e.g., combining trucks, drones, bikes) into the VRP for last-mile delivery?
- 2. How can models for drones be adapted to handle dynamic, real-time conditions, such as sudden changes in delivery locations, or unexpected drone failures?

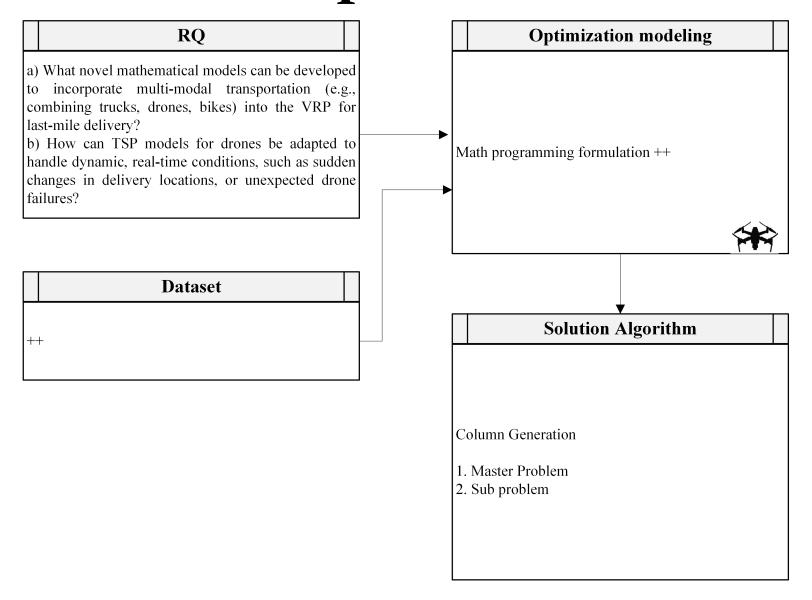
#### Solution algorithm

- 3. How can the pricing problem in column generation be optimized to generate routes more effectively for last-mile delivery?
- 4. How can heuristics or metaheuristics be integrated with column generation to improve solution quality for last-mile delivery optimization?

#### **Sensitivity**

- 5. What are the impacts of various problem parameters (e.g., number of customers, geographic distribution) on the performance of VRP models for last-mile delivery?
- 6. What are the trade-offs between solution quality and computational time when using heuristic-based column generation?
- 7. What are the impacts of varying problem parameters (e.g., number of deliveries, geographic distribution) on the performance of VRP algorithms in drone delivery?

## **Proposed Method**



# Solution algorithm

- Column Generation ++
- Relevance to large-scale optimization??

Case study 1
Optimizing emergency
responses: developing
operations research tools for
Flood response in
Bangladesh

## Problem Statement

- How we can maximize the coverage of shelter points given:
  - Disrupted road network
  - Resource constraints
  - Limited capacity of vehicles

 Coverage: delivering reliefs from warehouse to shelters

## Data

- Quick intro (of data collected so far)
- Add data table

- Warehouse: need to assume
- Shelters: capacity, geo-location
- Road network: national& regional highway, other roads, disrupted (can we get flood elevation or other info++ Riad)
- Demand: resource needed, Location of people,
- Resource/relief: Quantity and warehouse location

# Data Exploration

# Optimization model

### **Conceptual Model**

## Math model

- 1. Objective: maximizing coverage of shelter (demand) points
- 2. DV: no. of vehicles (# of drone, helicopter, truck), capacity of these vehicle
  - a. Vehicle level: x ijk, from where to where
  - b. Routing level: road selection (disrupted, non-disrupted routes)
- 3. Input parameters: packages.
  - a. Disrupted vs non-disrupted networks
- 4. Constraint
  - a. Travel each demand points only once
  - b. Capacity constraints: for each vehicle
  - c. Travelling distance: helicopter (max coverage), every vehicle (truck refuel), other practical limits
  - d. Time-window

## Results

# Results (Continued)

## Computational Result: Dataset

• Dataset 1 ( Wu et al., 2024):

 Dataset 2 (US Arctic emergency response related VRP): summary of the dataset+

## Implementation

- Modeling: Python/Pyomo
- Solver: CPLEX/Gurobi
- Simulation: AnyLogic
- Versioning: Github
- Packaging: Docker & Streamlit