

Happy Chinese New Year



Public transit planning and scheduling based on AVL data in China

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OUTLINE

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Background – Public transport in China

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Comprehensive framework of transit planning

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**An integrated solution for
public transport scheduling (iPTS)**

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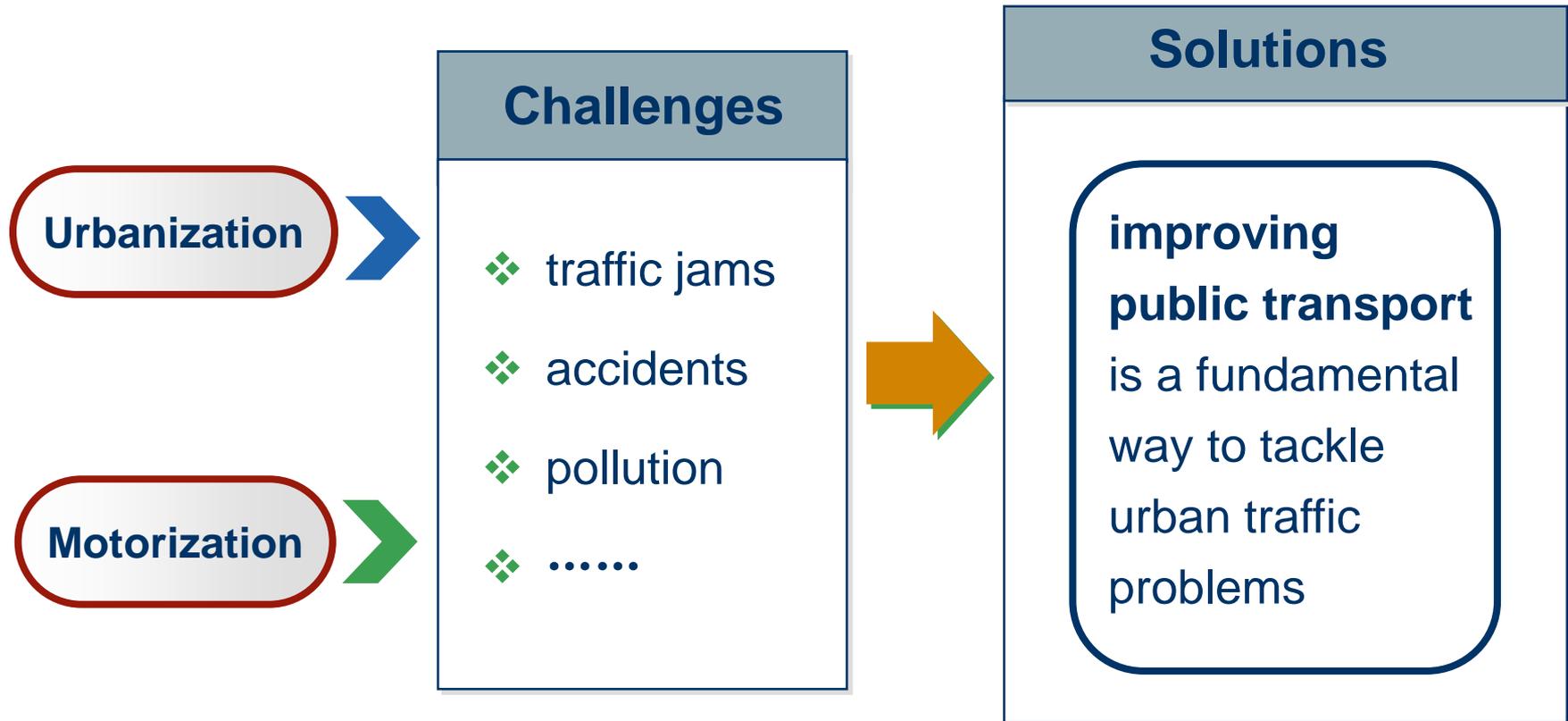
Vehicle scheduling based on AVL data

China



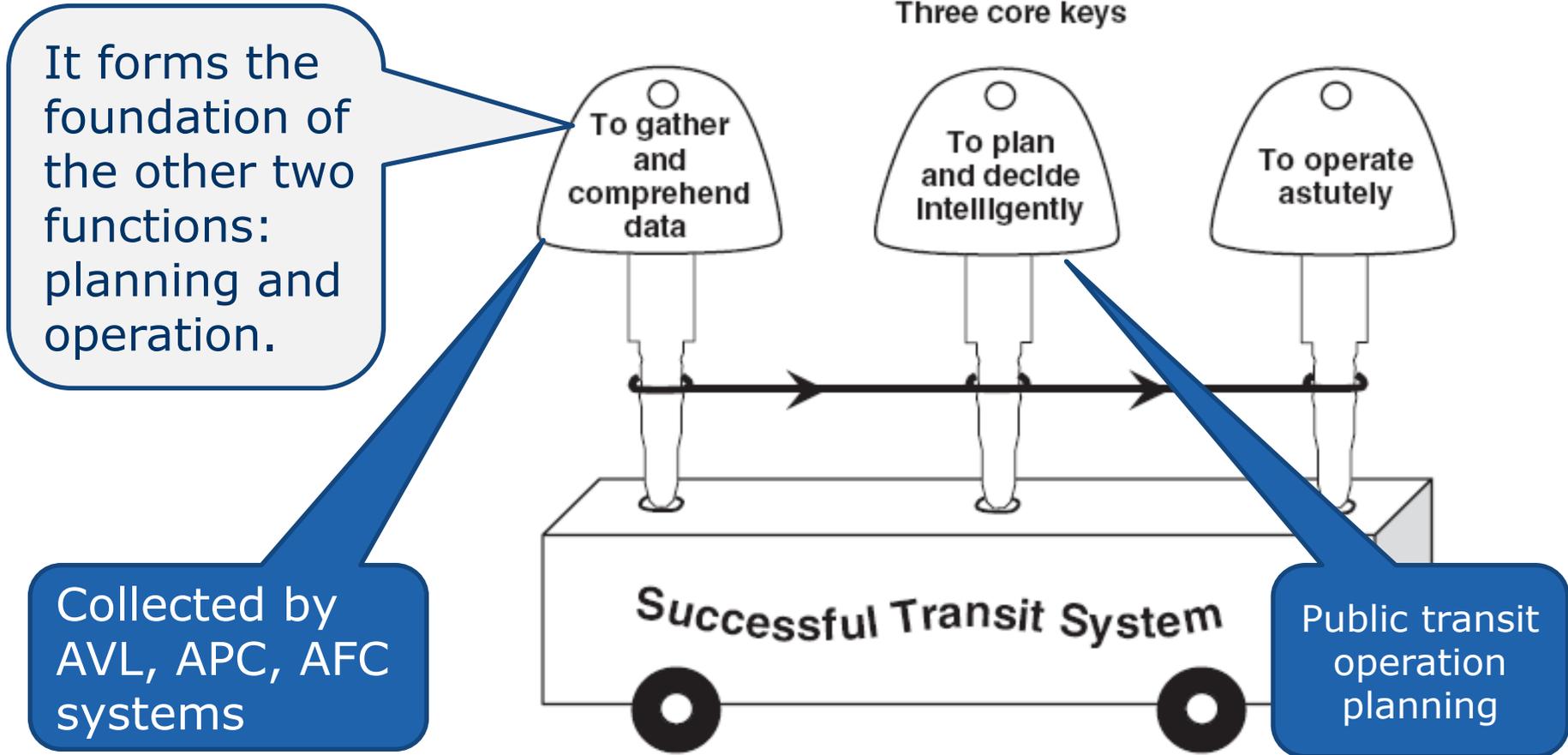
- ❖ Population: 1.328 billion by the end of 2008
- ❖ Area: 9.6 million km²
- ❖ Provinces (23) & autonomous regions (5+2) & Municipality (4): 34
- ❖ Capital: Beijing

Severe challenges to urban transportation



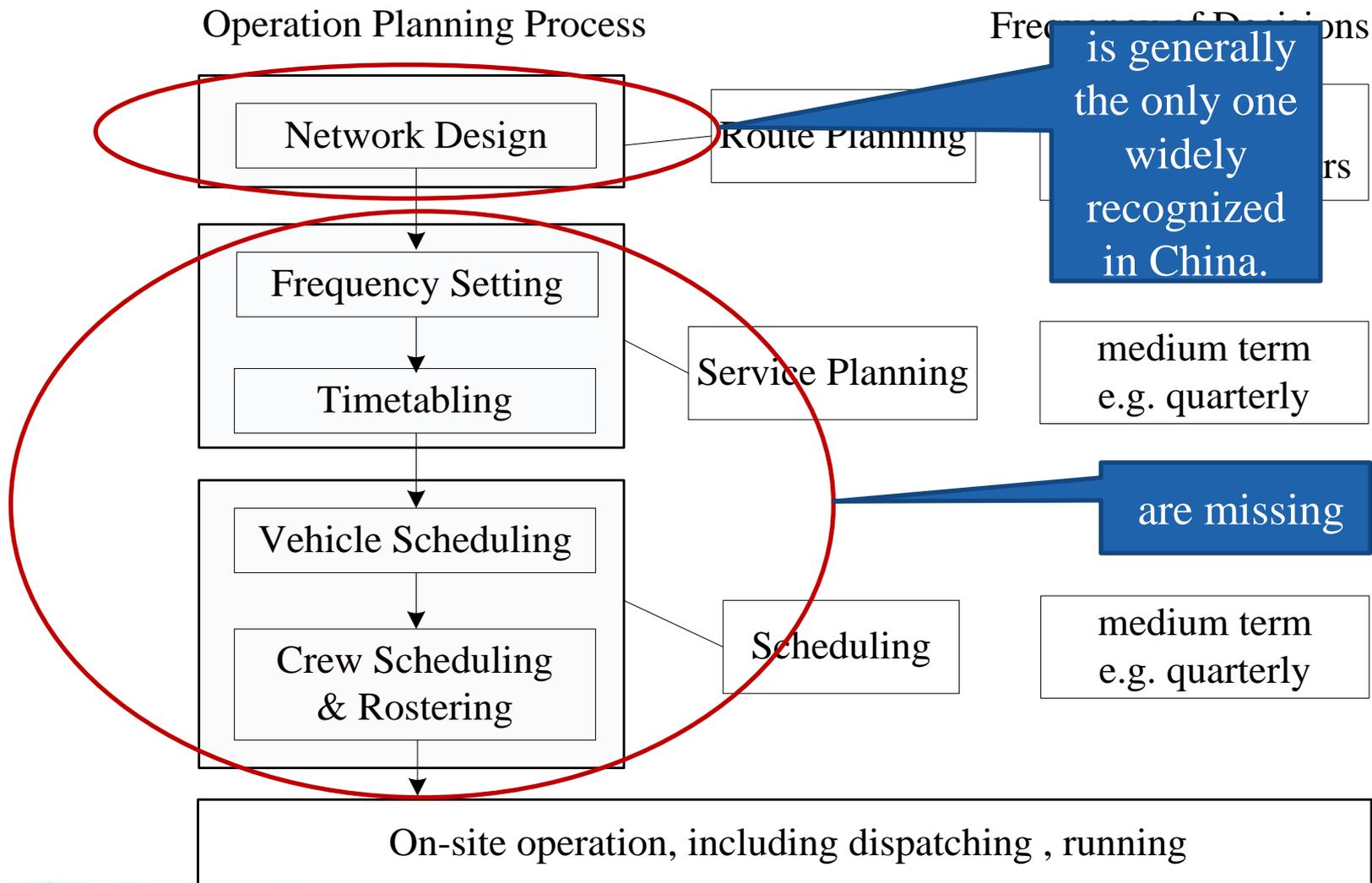
How to achieve it?

Components of successful transit system

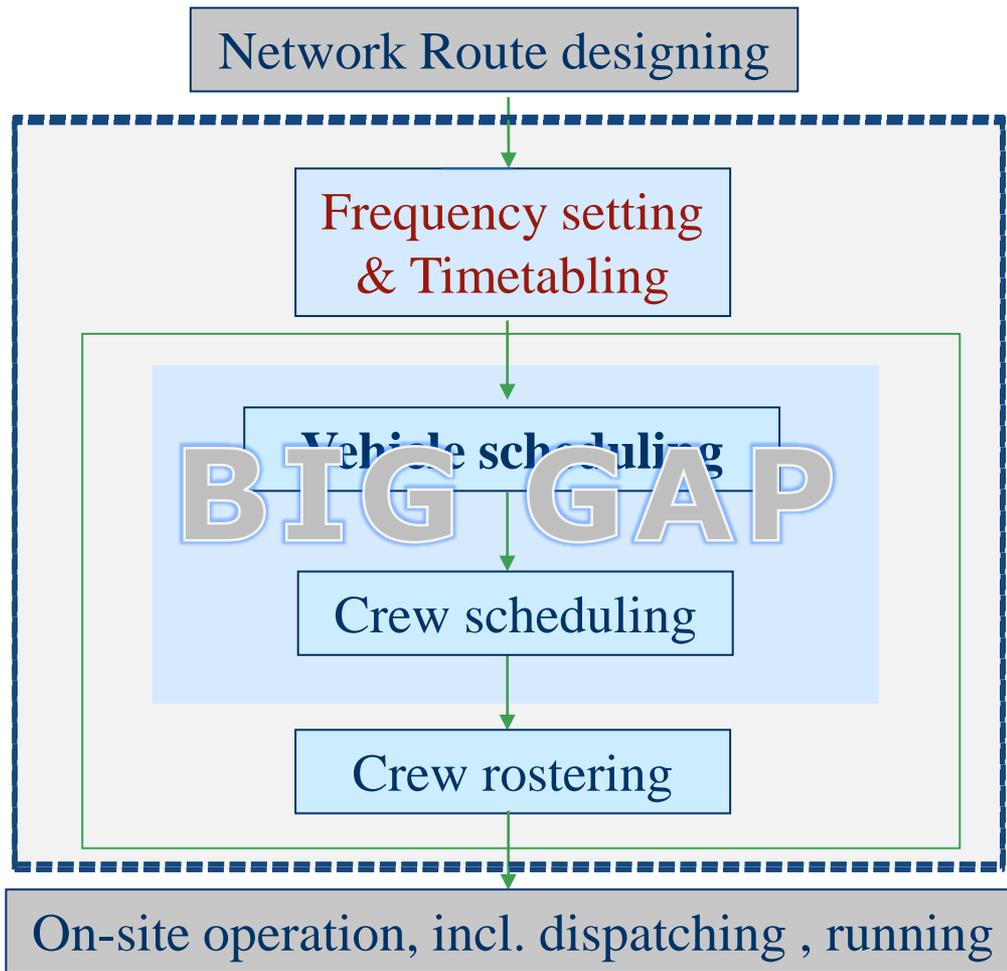


Ceder A (2007). Public transit planning and operation: theory, modeling and practice. Elsevier, Butterworth-Heinemann

Public transit operation planning process



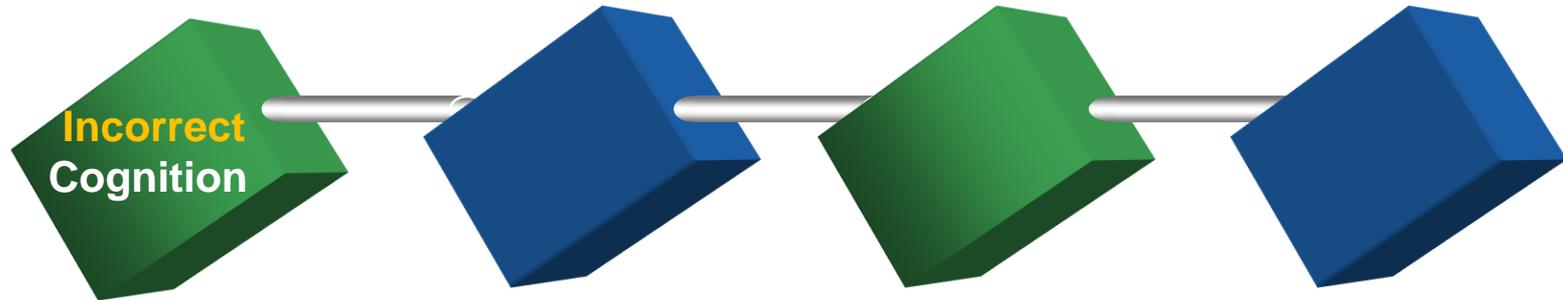
Gap between route design and operation



Why ?

❖ **Dispatching** is generally paid more attention than **scheduling** in China

Major Problems related to transit planning



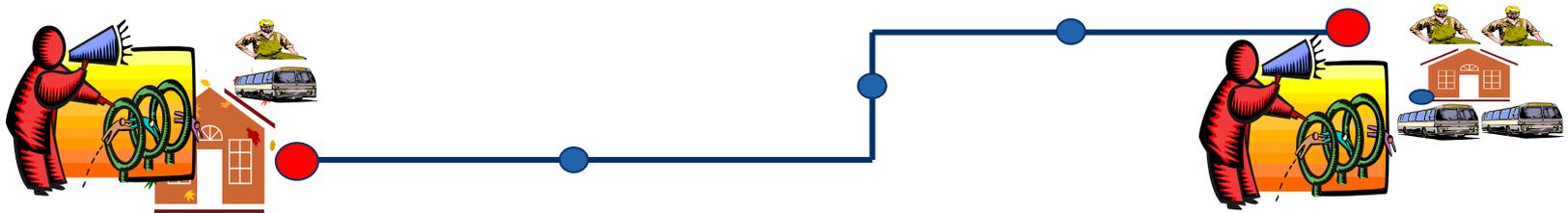
Misunderstanding the public transport planning as network route design.

Scheduling is often confused with **dispatching**.

- The **schedule compilation** is generally underrated, but
- **dispatching**, by contrast, plays a much more important role in bus operation.

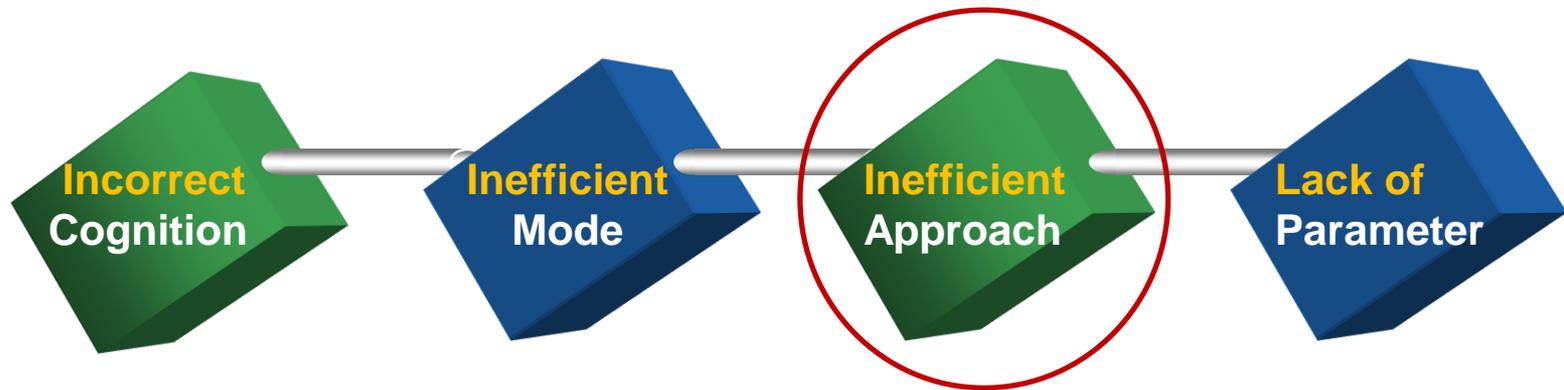
Pay more attention on dispatching

- ❖ Each **terminus** is usually equipped with at least a **dispatcher**, who plays part of the roles of schedulers by deciding the departure time of each service trip according to experiences.
- ❖ The efficiency of dispatching depends greatly on the experience, responsibility and personal authority over drivers of an individual dispatcher.



Line-by-Line Mode

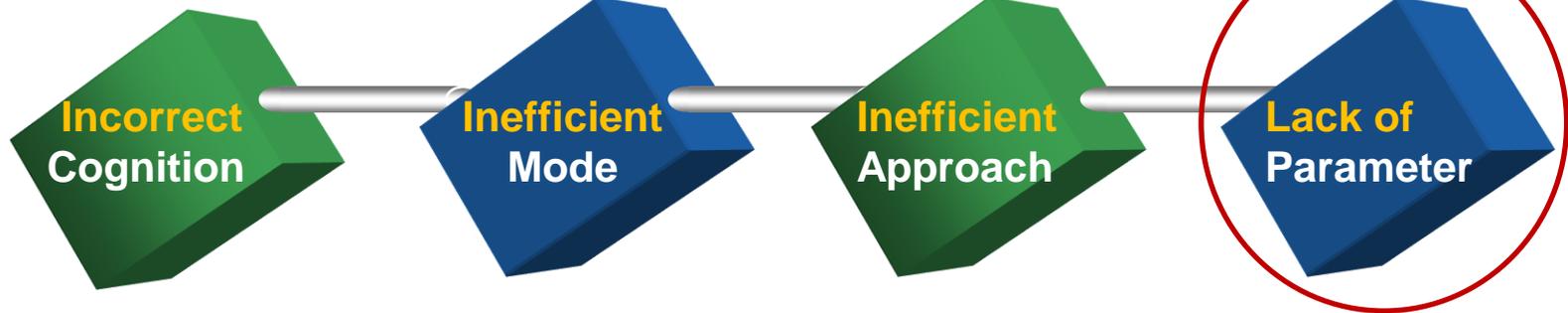
Major Problems related to transit planning



❖ The scheduling approaches successful in developed countries cannot be directly applied in China due to **some Chinese specific rules.**

The practical scheduling approach or **system** is missing.

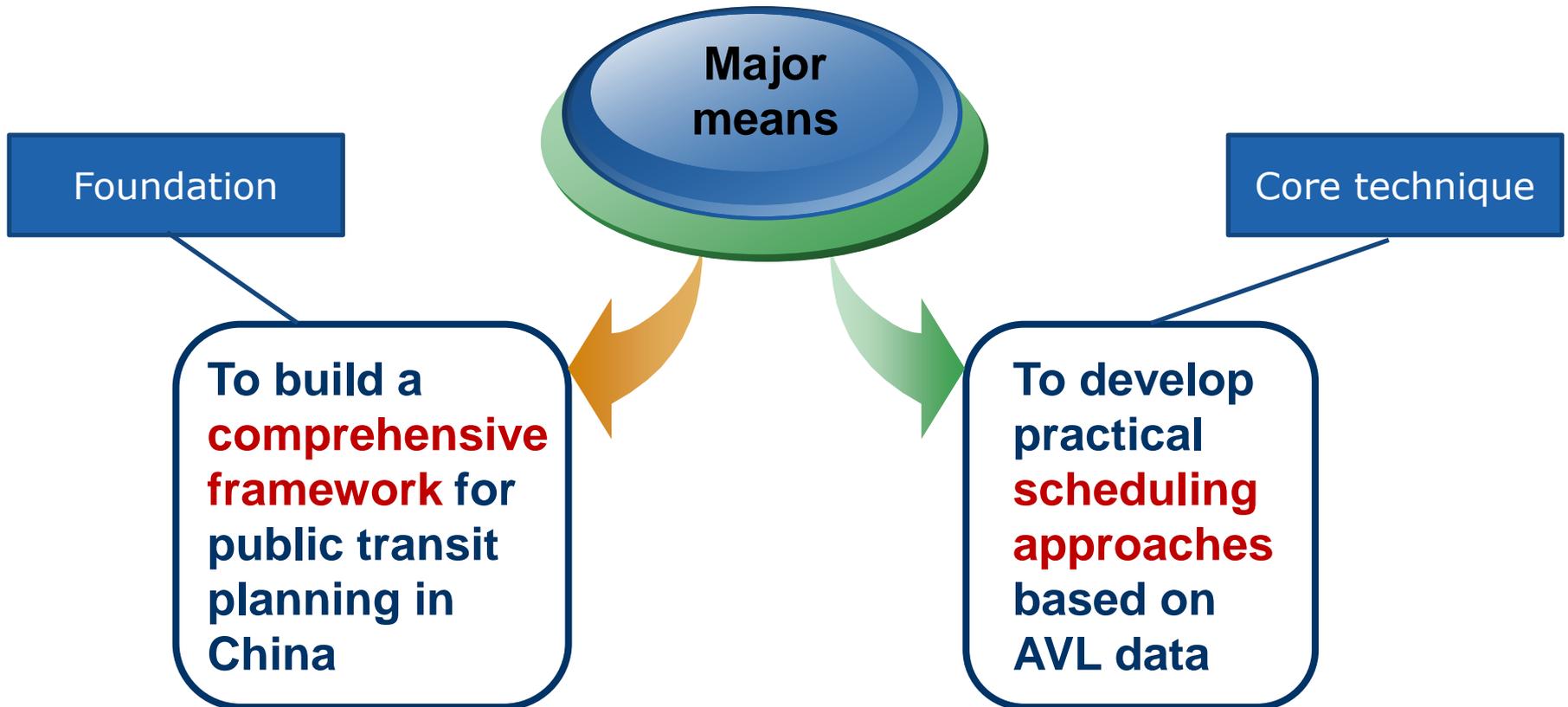
Major Problems related to transit planning



The service planning and scheduling are missing, or carried out roughly and arbitrarily

Lack of data or parameters has acted as a disincentive to the widespread use of scheduling systems.

Our major efforts



How to fulfil them?

Shortcut to achieve the goal

To achieve the goal

First of all

we should let transit authority and operators

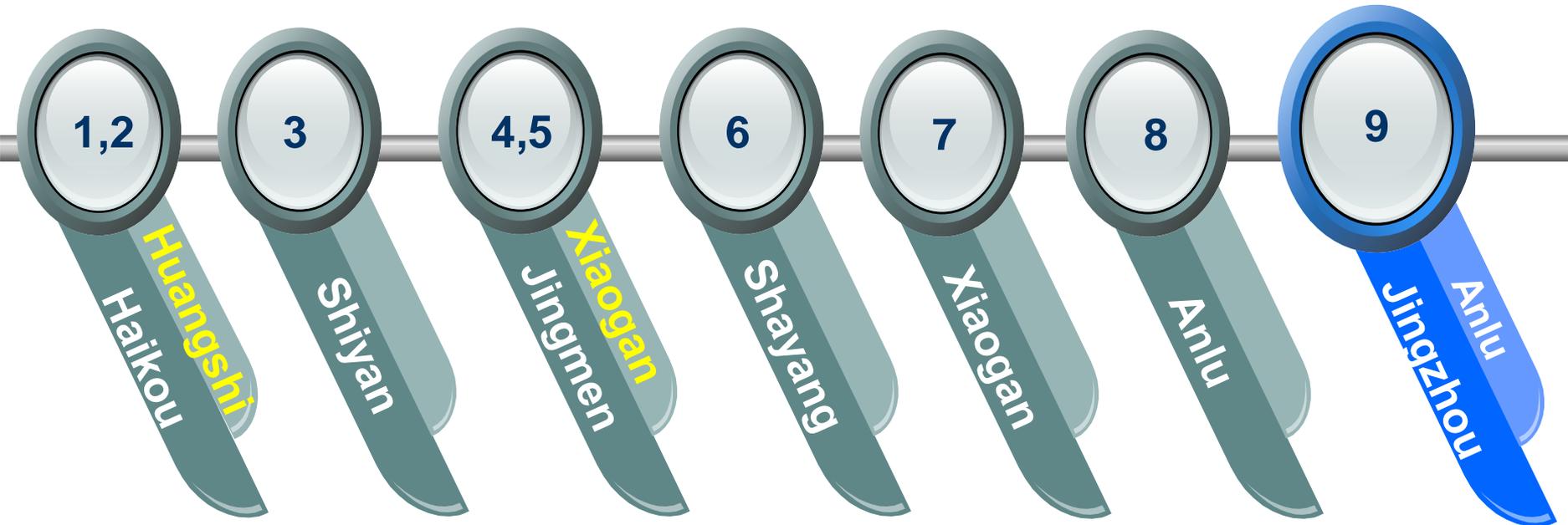
have a **comprehensive understanding** of public transit operation planning.

We believe

being involved in traditional transit planning projects would be a **shortcut**.

Projects on public transit planning

2010 → 2012 → 2014 → 2016 → 2018 → 2020 → **2022**
2011 → 2013 → 2015 → 2017 → 2019 → 2021

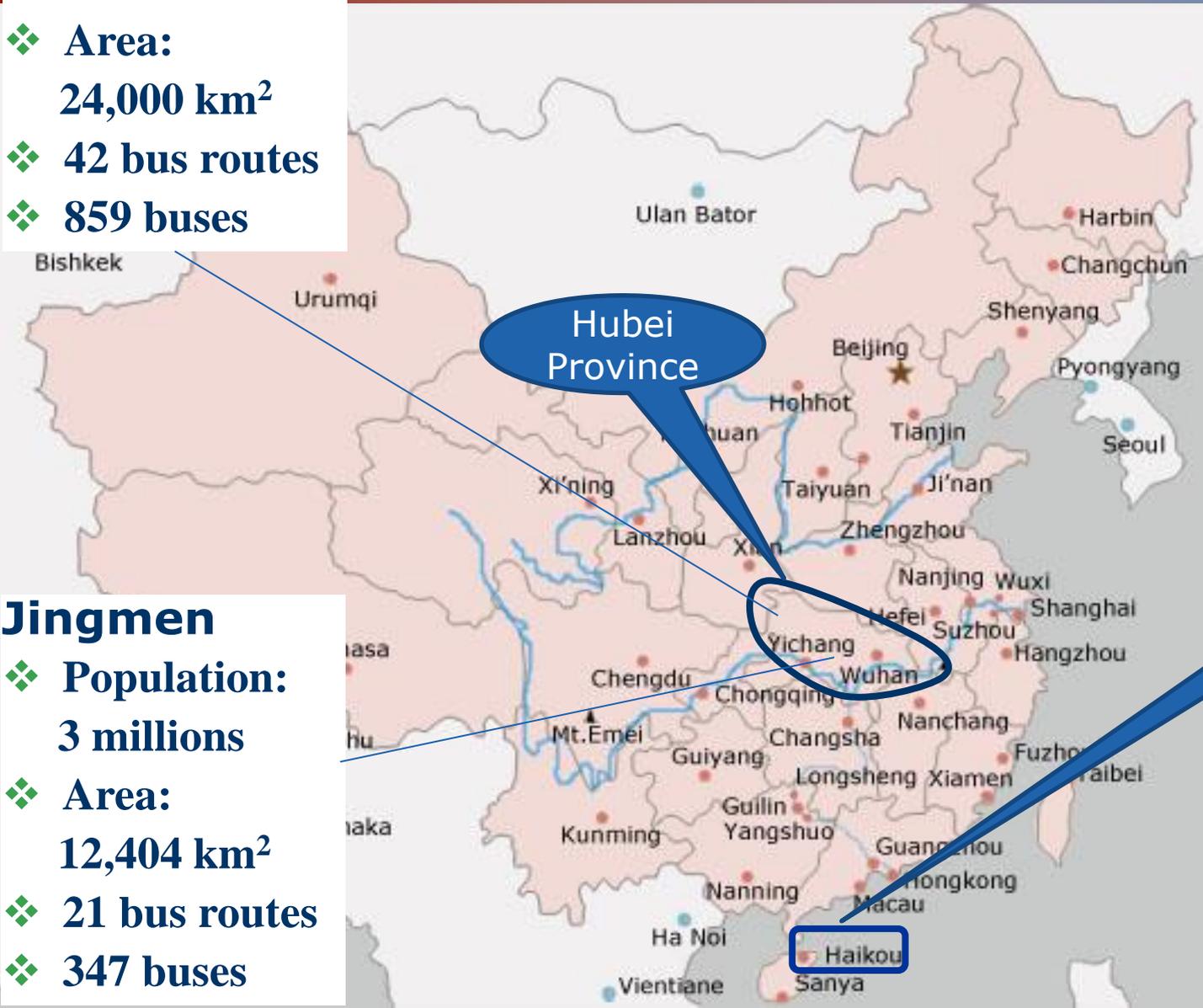


Shiyan

- ❖ Population: 3.5 millions
- ❖ Area: 24,000 km²
- ❖ 42 bus routes
- ❖ 859 buses

Some projects

sponsored by transit authorities in the cities and provinces.



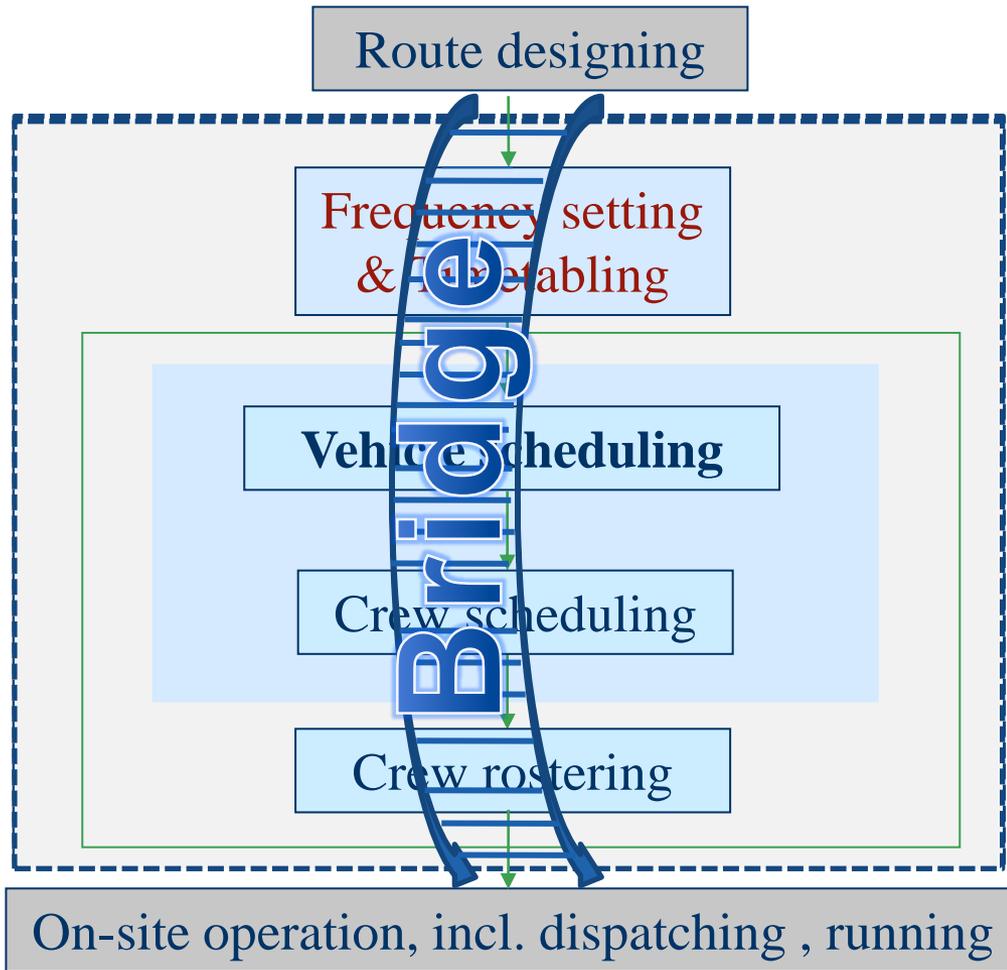
Jingmen

- ❖ Population: 3 millions
- ❖ Area: 12,404 km²
- ❖ 21 bus routes
- ❖ 347 buses

Haikou

- ❖ Population: 2.05 millions
- ❖ Area: 2304.84 km²
- ❖ 60 bus routes
- ❖ 1135 buses

Bridge between route design and operation

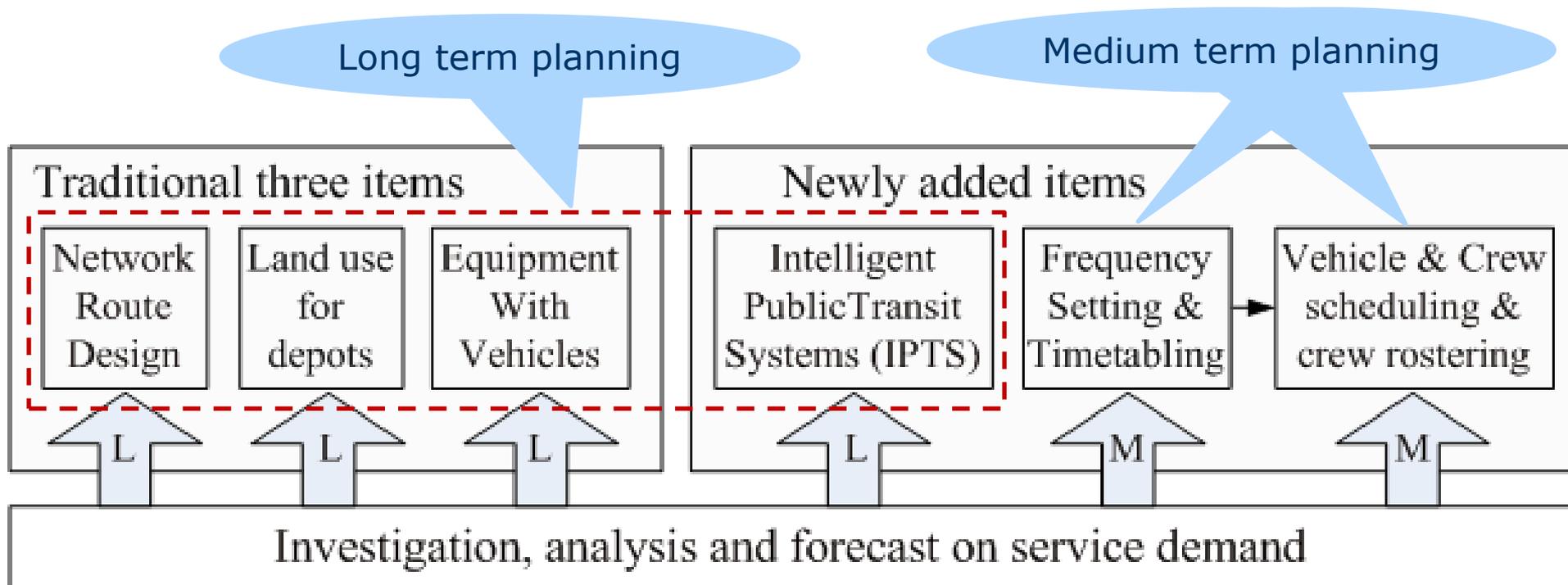


Structure design

- ❖ Set up a comprehensive framework of transit operation planning

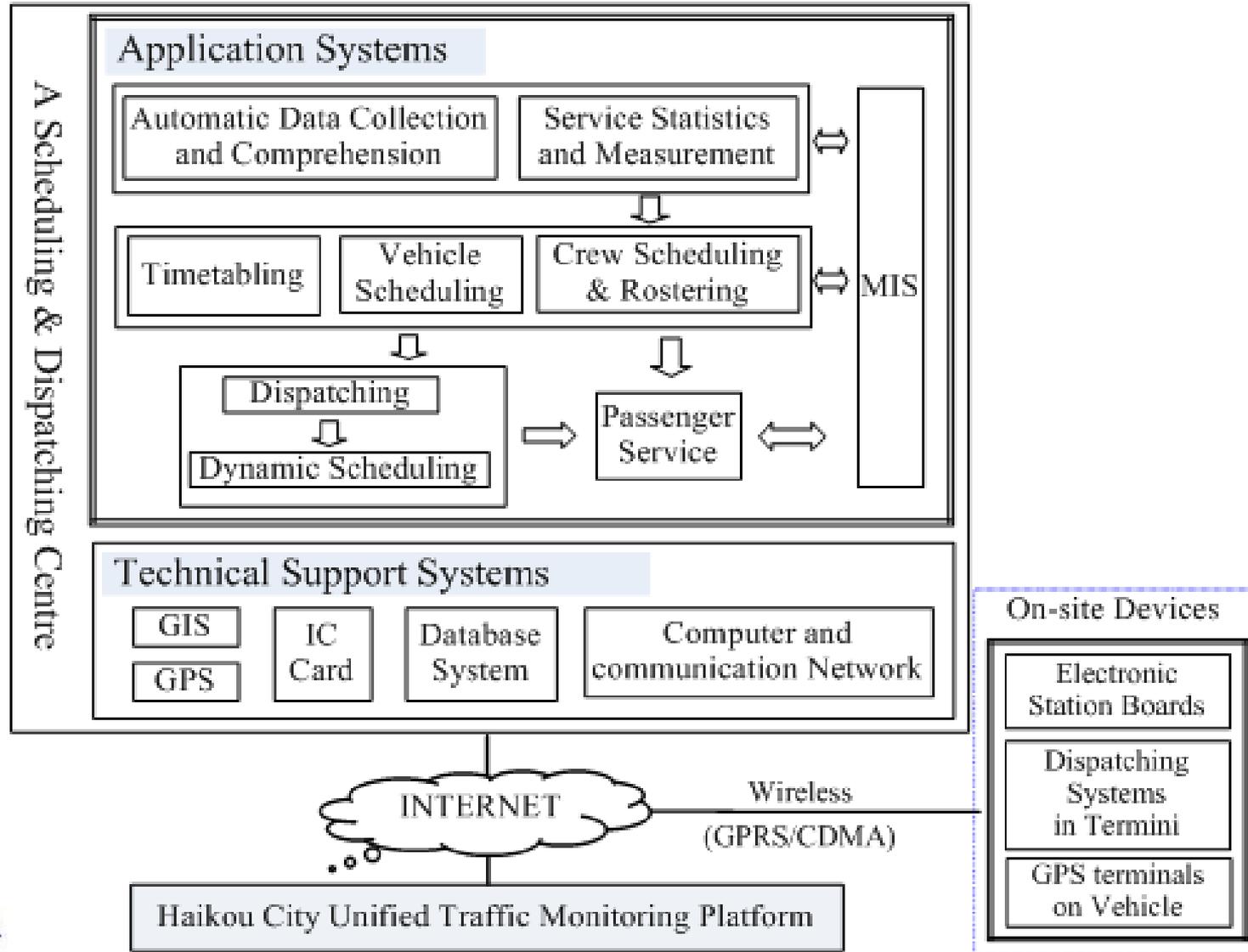
❖ **Dispatching** is generally paid more attention than **scheduling** is in China

Our proposed comprehensive framework

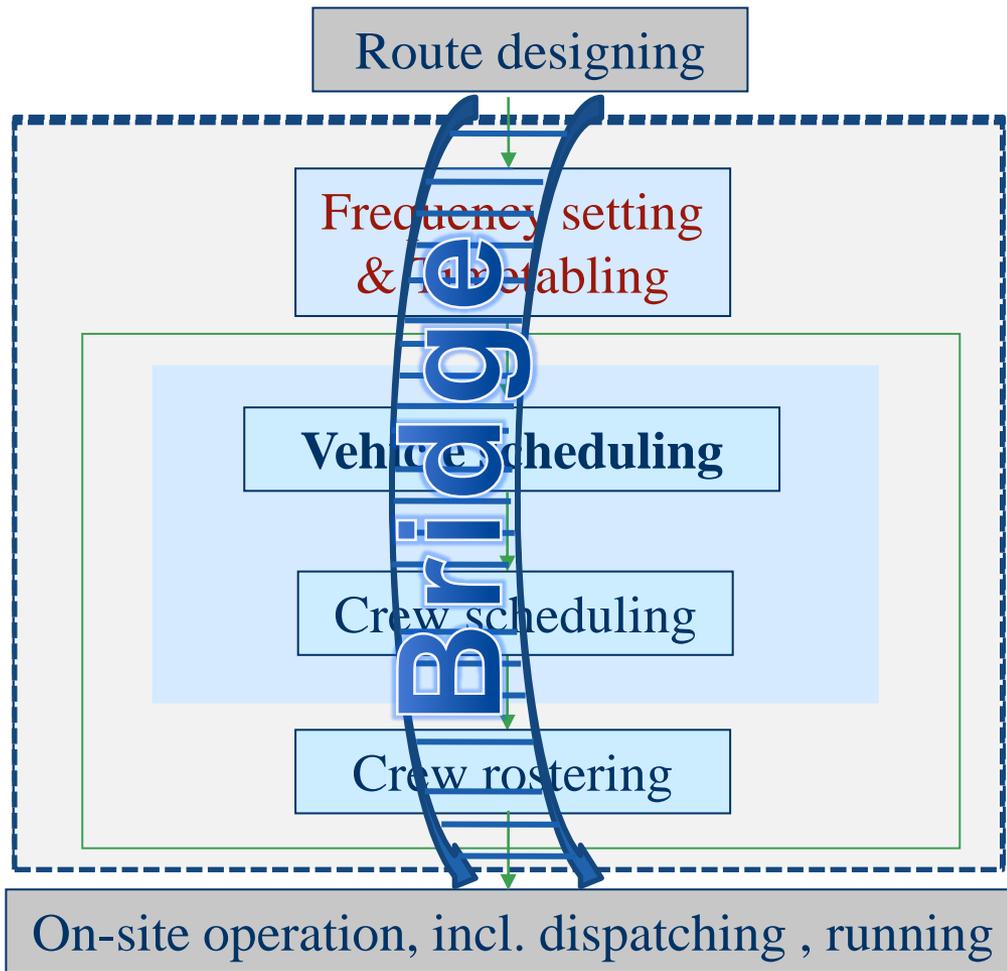


❖ The comprehensive framework of public transit planning in China encompasses all the contents in the public transit planning.

A general framework of IPTS plan of Haikou



Bridge between route design and operation



Structure design

- ❖ Set up a **comprehensive framework** of transit operation planning, in which,
- ❖ **IPTS plans** were successfully accepted in China.



Technical design

An Integrated Solution for Public Transport Scheduling

SCHEDULING

- ⊙ Vehicle scheduling
- ⊙ Crew scheduling
- ⊙ Crew rostering

GOVERNMENT SERVICE

- ⊙ Subsidies Accounting
- ⊙ Service Policy Setting

DATA ANALYSIS

- ⊙ Running Time Analysis
- ⊙ Ridership Analysis
- ⊙ Timetable Development
- ⊙ Comparison of Schedules

INTEGRATED TOOLS

- ⊙ Interfaces
- ⊙ External Systems

ERP

- ⊙ Infrastructure and Equipment Management
- ⊙ Operations and Safety Management
- ⊙ Logistics and Maintenance Management
- ⊙ Party Affairs and Administration
- ⊙ Financial Management

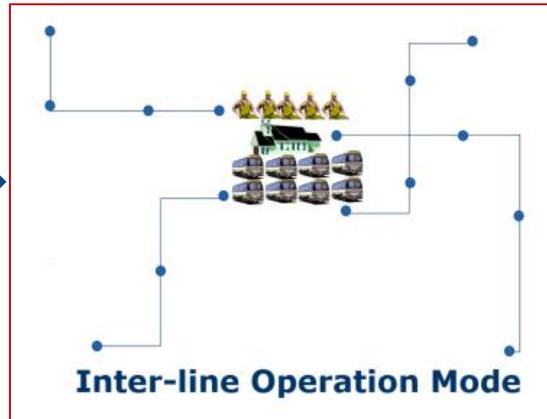
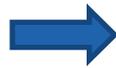
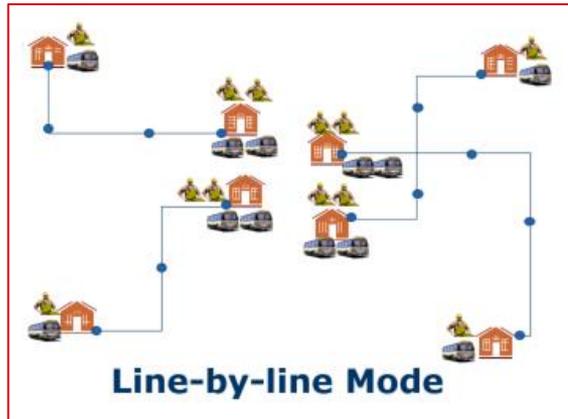
RELATED SERVICES

- ⊙ Network Route Planning
- ⊙ Feasibility Consultation

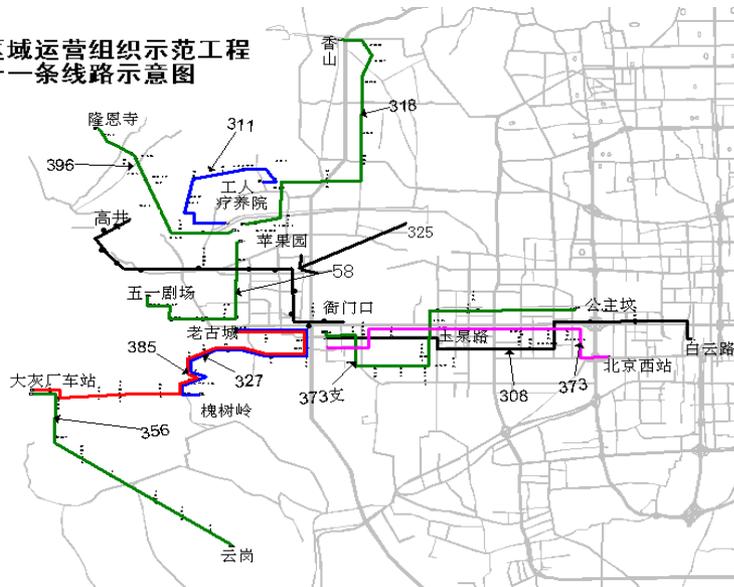
iPTS



Case Study: in Beijing Bus Group(BJBUS)



区域运营组织示范工程
十一条线路示意图



A B C D E F G H I										J K L M N O P					Q		
客六示范区区域行车方案总站行车计划表										苹果网站					第1页共10页		
到达车辆										开出车辆							
线路名称	本路车次	车号	到达时间	实际行驶时间	准点情况	准	早	晚	不准点原因	线路名称	本路车次	车号	开出时间	间隔时间	停站时间	停站过久及间隔不准的原因	备注及报告事项
										苹果园场发	B311路(上行)	6	5:10		0		
										苹果园场发	B58路(上行)	8	5:10		8		
										苹果园场发	B58路(上行)	12	5:18		4		
										苹果园场发	B311路(上行)	14	5:22		4		
										苹果园场发	B58路(上行)	17	5:26		8		
										苹果园场发	B311路(上行)	27	5:34		0		
										苹果园场发	B58路(上行)	28	5:34		8		
										苹果园场发	B58路(上行)	37	5:42		2		
										苹果园场发	B311路(上行)	38	5:44		6		
										苹果园场发	B58路(上行)	50	5:50		4		
										B58路(下行)		8	5:50		4	4	
										B311路(下行)		6	5:50		8	6	
										B58路(下行)		12	6:00		4	1	
										苹果园场发	B58路(上行)	86	6:05		6		
										B58路(下行)		17	6:11		3	3	
										B311路(下行)		14	6:14		4	4	
										苹果园场发	B58路(上行)	67	6:18		6		
										B58路(下行)		28	6:24		8	0	
										苹果园场发	B311路(上行)	31	6:24		7		
										B311路(下行)		27	6:31		9	2	
										苹果园场发	B311路(上行)	77	6:33		4		
										B311路(下行)		38	6:37		5	5	
客六示范区区域行车方案总站行车计划表										苹果网站					第2页共10页		
到达车辆										开出车辆							
线路名称	本路车次	车号	到达时间	实际行驶时间	准点情况	准	早	晚	不准点原因	线路名称	本路车次	车号	开出时间	间隔时间	停站时间	停站过久及间隔不准的原因	备注及报告事项
										苹果园场发	B311路(上行)	6	5:10		0		
										苹果园场发	B58路(上行)	8	5:10		8		
										苹果园场发	B58路(上行)	12	5:18		4		
										苹果园场发	B311路(上行)	14	5:22		4		
										苹果园场发	B58路(上行)	17	5:26		8		
										苹果园场发	B311路(上行)	27	5:34		0		
										苹果园场发	B58路(上行)	28	5:34		8		
										苹果园场发	B58路(上行)	37	5:42		2		
										苹果园场发	B311路(上行)	38	5:44		6		
										苹果园场发	B58路(上行)	50	5:50		4		
										B58路(下行)		8	5:50		4	4	
										B311路(下行)		6	5:50		8	6	
										B58路(下行)		12	6:00		4	1	
										苹果园场发	B58路(上行)	86	6:05		6		
										B58路(下行)		17	6:11		3	3	
										B311路(下行)		14	6:14		4	4	
										苹果园场发	B58路(上行)	67	6:18		6		
										B58路(下行)		28	6:24		8	0	
										苹果园场发	B311路(上行)	31	6:24		7		
										B311路(下行)		27	6:31		9	2	
										苹果园场发	B311路(上行)	77	6:33		4		
										B311路(下行)		38	6:37		5	5	

Case Study: in Beijing Bus Group(BJBUS)

Items	Manual Schedule	iPTS System	PRD
Num of Trips	1971	1971	0
Num of Buses	116	105	9.5%
Ratio (%)	16.99 trips/bus	18.77 trips/bus	10.4%
Num of Crews	182	166	8.8%

- ❖ The iPTS system produced a better solution by saving 11 from 116 buses and 16 from 182 crews.
- ❖ The computational time was very short (the elapsed time less than 4 minutes on a Pentium III 1GHz PC).
- ❖ The schedules generated have meet all the requirements, especially the particular ones of BJBUS .

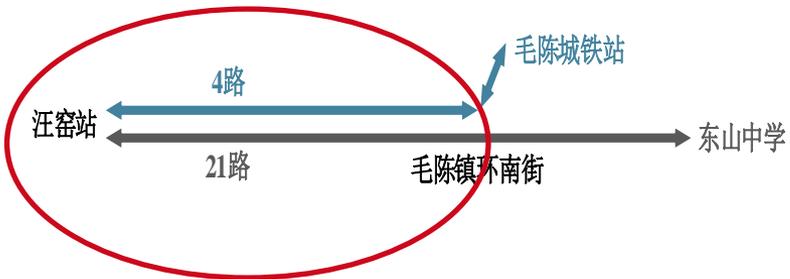
AWARDS

International Federation of Operational Research Societies (IFORS) “IFORS Prize for Operational Research in Development”, Runner-up

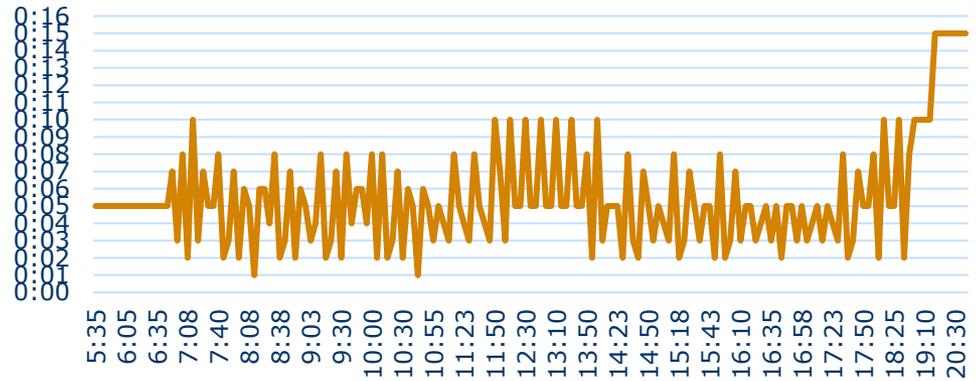
Case Study: in Xiaogan Bus Group

Even Headways

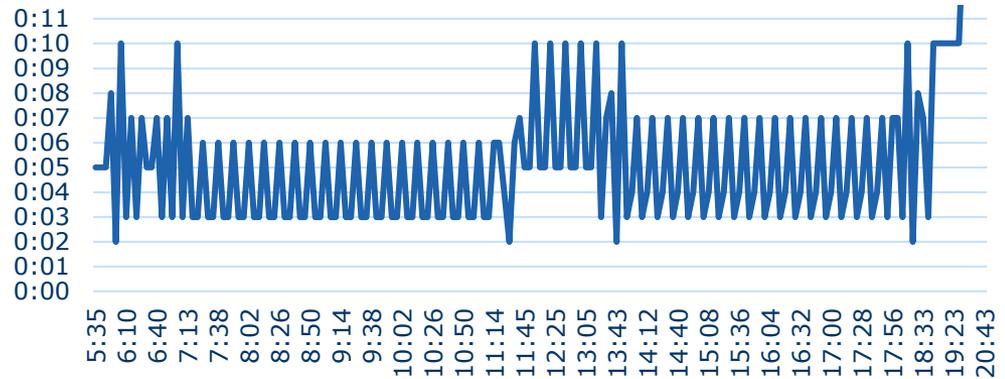
For the common route segment



Headways in Wangyao Terminal (Ori)



Headways in Wangyao Terminal (Ori)



Case Study: in Jingmen – timetable



8

周一至周五

热电厂 — 李宁工业园 (每4-7分钟一班)

往“热电厂”：首班：6:00；末班：20:50

往“李宁工业园”：首班：6:00；末班：20:50

热电厂	洪源市场	客运中心	职教集团	交通小学	李宁工业	李宁工业	交通小学	职教集团	客运中心	洪源市场	热电厂
①	②	③	④	⑤	⑥	⑥	⑤	④	③	②	①

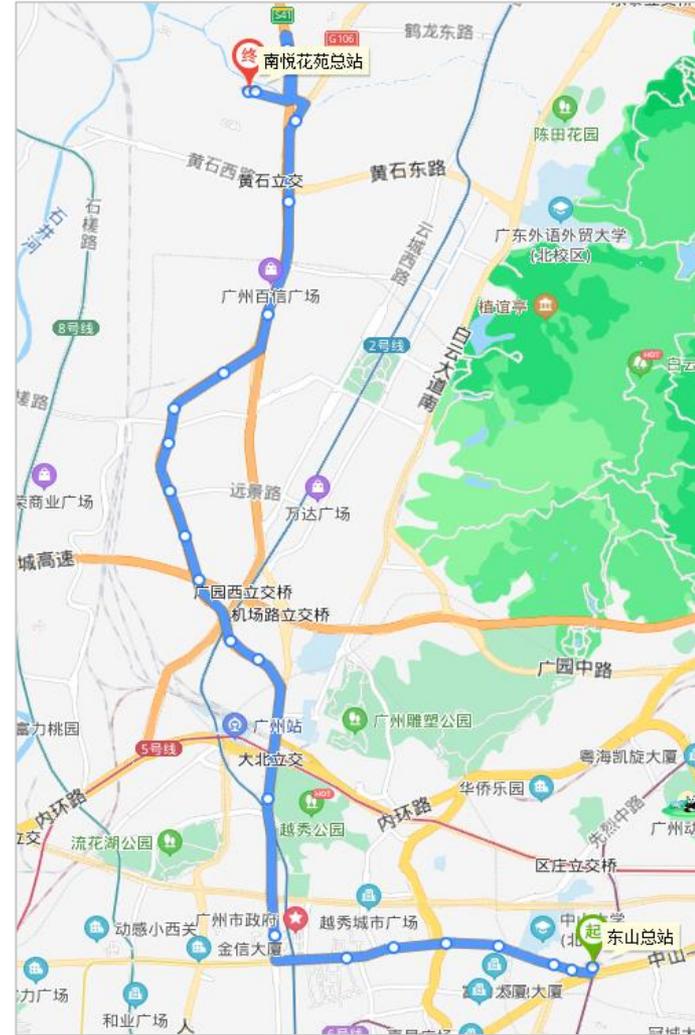
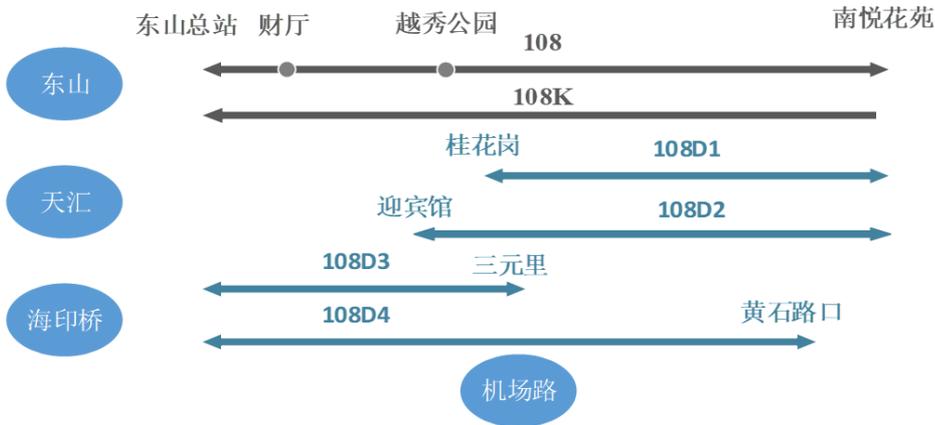
上行：热电厂-李宁工业园 单程时间：0:56:26 下行：李宁工业园-热电厂 单程时间：0:56:18

	6:00	6:09	6:16	6:31	6:46	6:52	6:00	6:07	6:12	6:26	6:39	6:46
	6:06	6:14	6:20	6:36	6:51	7:00	6:06	6:12	6:17	6:31	6:44	6:50
	6:10	6:19	6:26	6:41	6:56	7:05	6:10	6:17	6:22	6:36	6:49	6:55
	6:16	6:24	6:30	6:46	7:01	7:10	6:16	6:22	6:27	6:41	6:54	7:01
	6:20	6:29	6:36	6:51	7:06	7:15	6:20	6:27	6:32	6:46	6:59	7:06
	6:26	6:34	6:40	6:56	7:11	7:20	6:26	6:32	6:37	6:51	7:04	7:10
	6:30	6:39	6:46	7:01	7:16	7:25	6:30	6:37	6:42	6:56	7:09	7:15
	6:36	6:44	6:51	7:06	7:21	7:30	6:36	6:42	6:47	7:01	7:14	7:20
	6:40	6:49	6:56	7:11	7:26	7:35	6:40	6:47	6:52	7:06	7:19	7:25
	6:46	6:54	7:04	7:19	7:34	7:43	6:46	6:52	6:57	7:11	7:24	7:30
	6:49	6:58	7:08	7:23	7:38	7:47	6:49	6:56	7:01	7:15	7:28	7:34
	6:55	7:04	7:14	7:29	7:44	7:53	6:55	7:01	7:06	7:20	7:33	7:39
	6:57	7:08	7:18	7:33	7:48	7:57	6:57	7:04	7:09	7:23	7:36	7:42
	7:01	7:12	7:22	7:37	7:52	8:01	7:01	7:08	7:13	7:27	7:40	7:46
	7:06	7:16	7:26	7:41	7:56	8:05	7:06	7:13	7:18	7:32	7:45	7:51
	7:09	7:20	7:30	7:45	8:00	8:09	7:09	7:16	7:21	7:35	7:48	7:54
	7:13	7:24	7:34	7:49	8:04	8:13	7:13	7:20	7:25	7:39	7:52	7:58
	7:17	7:28	7:38	7:53	8:08	8:17	7:17	7:24	7:29	7:43	7:56	8:02
	7:21	7:32	7:42	8:01	8:16	8:25	7:21	7:28	7:33	7:47	7:60	8:06
	7:26	7:36	7:46	8:05	8:20	8:29	7:26	7:33	7:38	7:52	8:05	8:11
	7:29	7:40	7:50	8:07	8:22	8:31	7:29	7:36	7:41	7:55	8:08	8:14
	7:33	7:44	7:54	8:13	8:28	8:37	7:33	7:40	7:45	7:59	8:12	8:18
	7:37	7:48	7:58	8:17	8:32	8:41	7:37	7:44	7:49	8:03	8:16	8:22

工作日(周一至周五)

Case Study: in Guangzhou

A complicated bus route (108) with several short lines



Case Study: in Guangzhou

Running time analysis

To increase the on-time performance

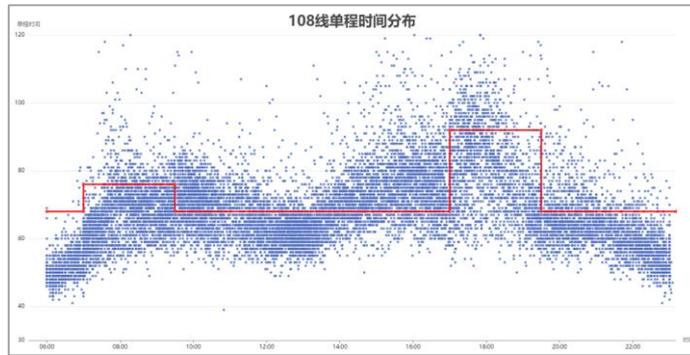


Fig. 1 Running time (in the manual schedule)

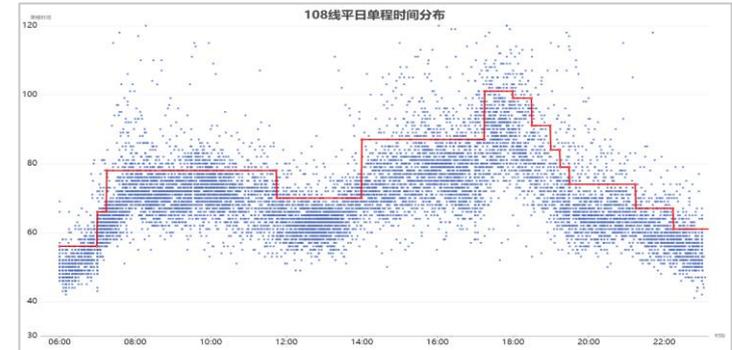


Fig. 2 Running time (Weekdays)

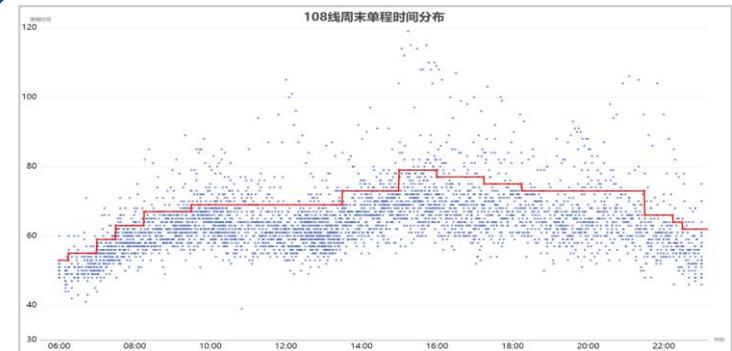


Fig. 3 Running time (Weekends)

Case Study: in Guangzhou

Headway optimization

- To meet passenger demands
- & even headways

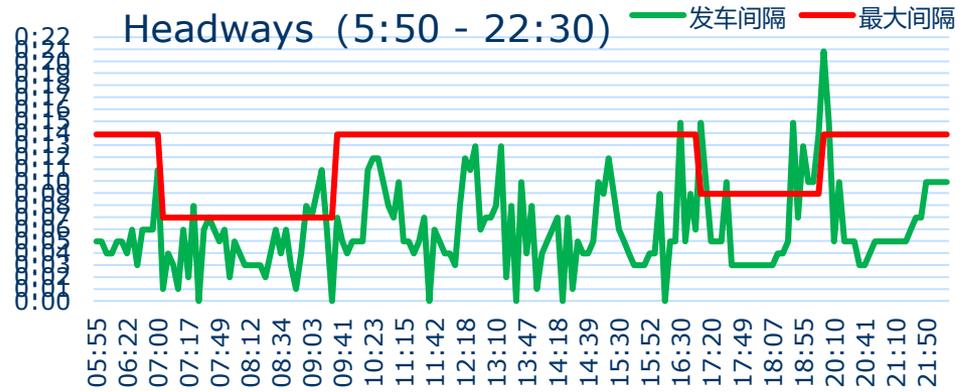


Fig. 1 Headways in 108 (NY-DS) in the manual schedule

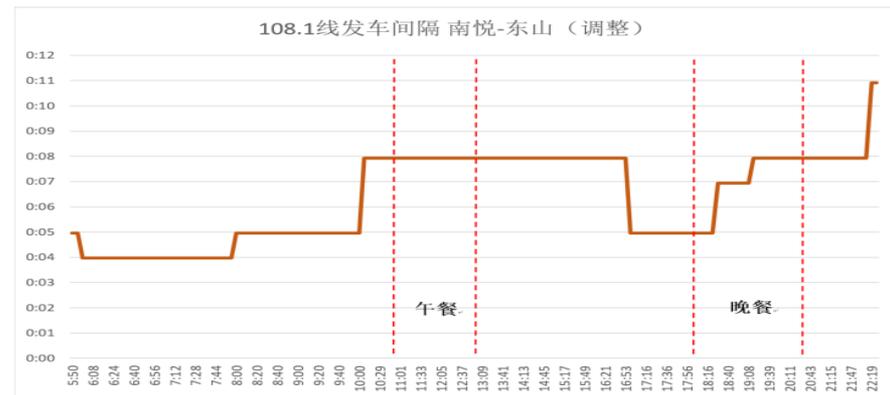
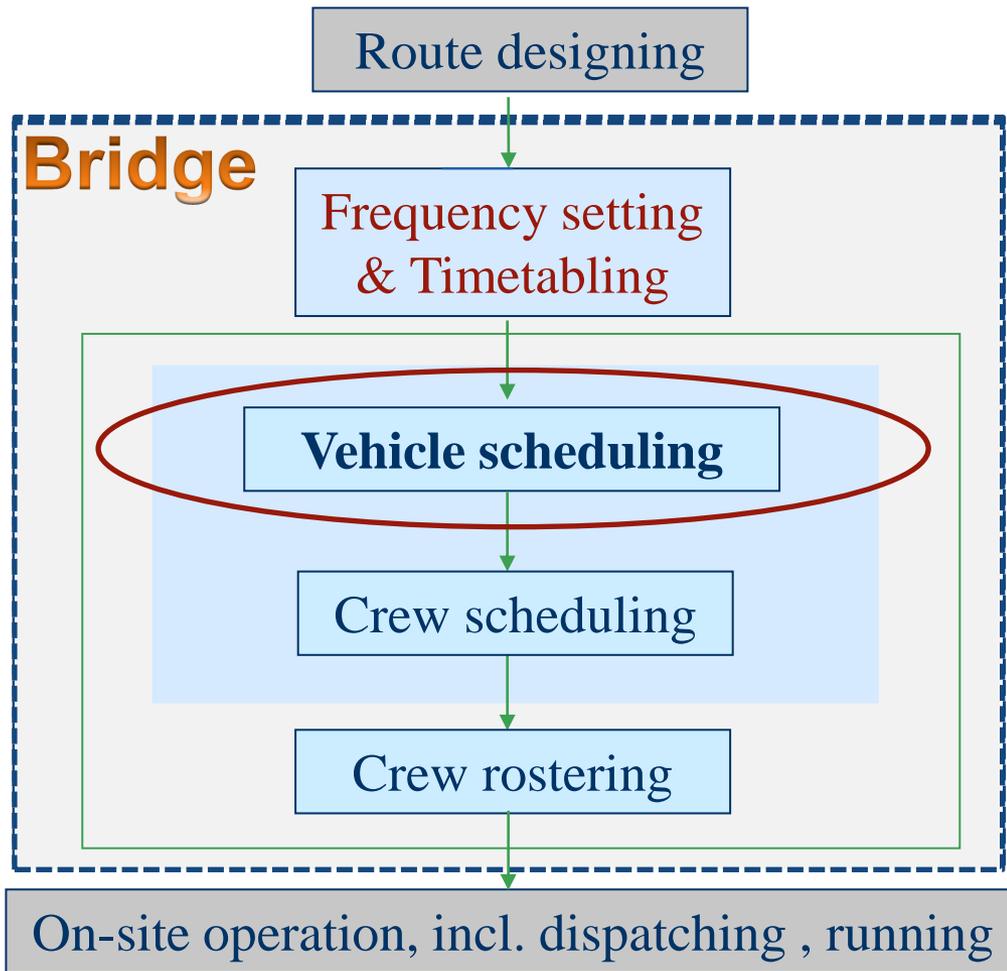


Fig. 2 Optimized headways in 108 (NY-DS)

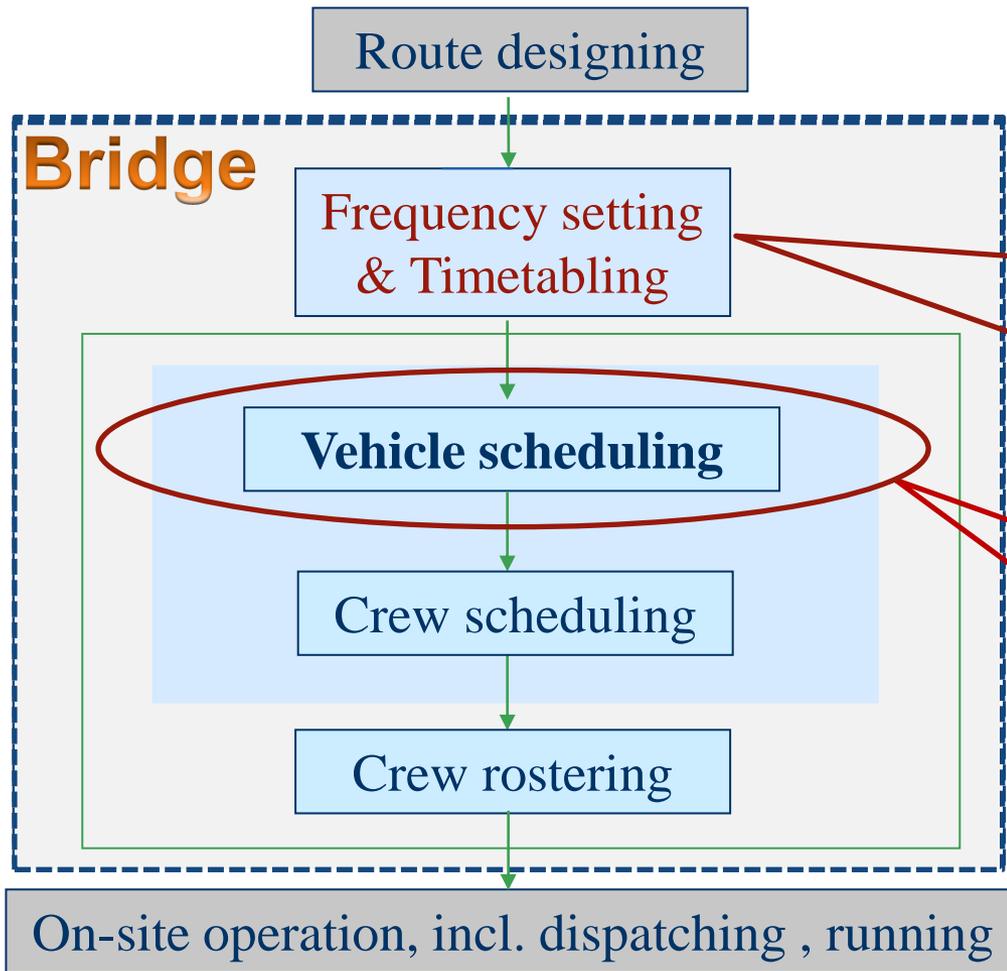
Vehicle scheduling based on AVL data



- ❖ Each sub-problem in the 'bridge' is individually hard,
- ❖ it is infeasible to develop a global solution approach.
- ❖ A **simple** but **applicable** bridge is more acceptable in China.

❖ **Dispatching** is generally paid more attention than **scheduling** is in China

Vehicle scheduling based on AVL data

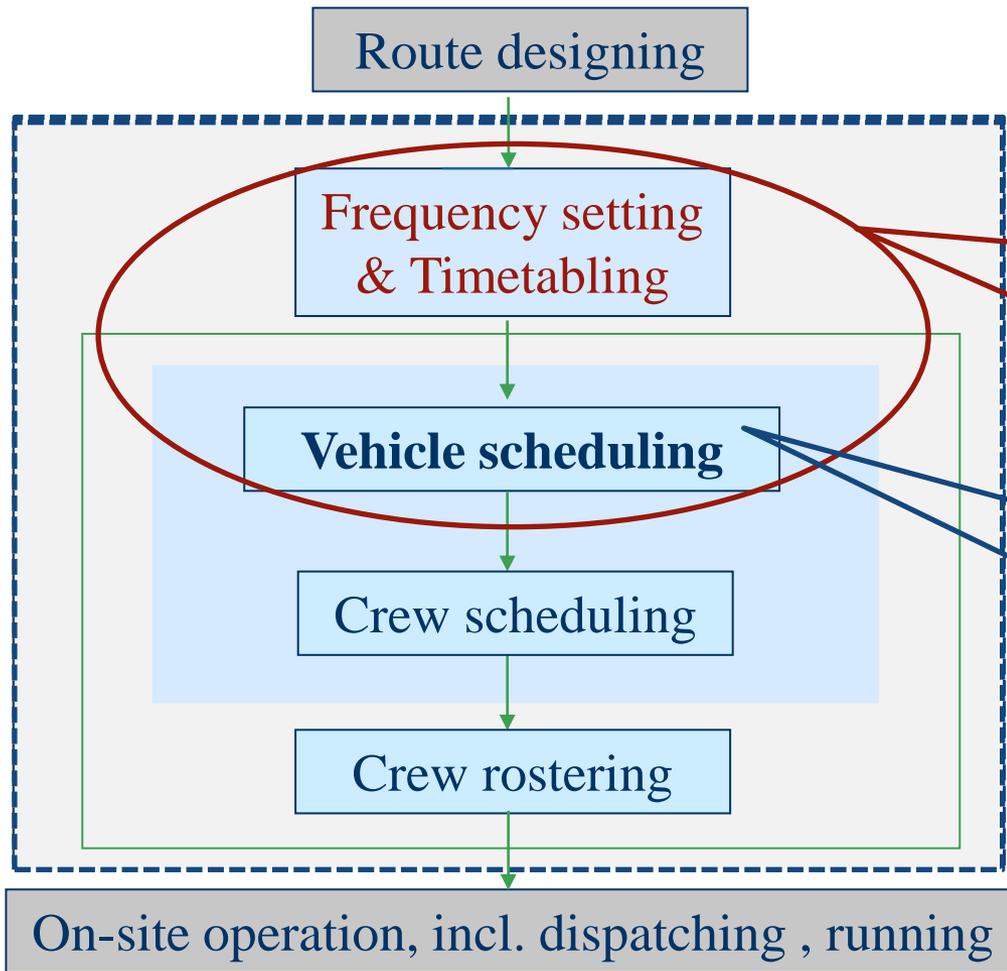


In China (lack of scheduling parameters -- trips)

- No precompiled timetables
- No well-designated trip times and headways

Traditional Vehicle scheduling is based on a **given timetable**, which consists of a set of **timetabled trips**.

Vehicle scheduling based on AVL data

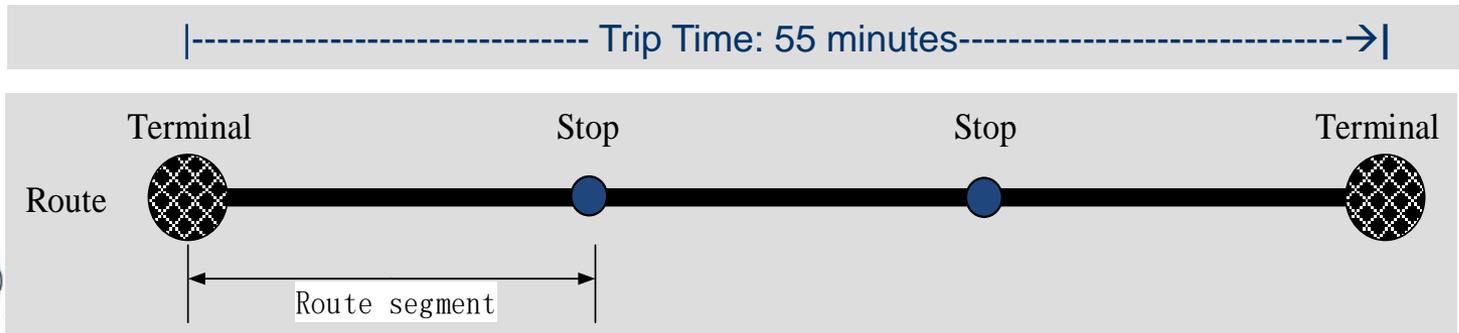


The proposed vehicle scheduling is based on AVL data, which **integrates** the service planning and vehicle scheduling.

Traditional Vehicle scheduling is based on a given timetable, which consists of a set of **timetabled trips**.

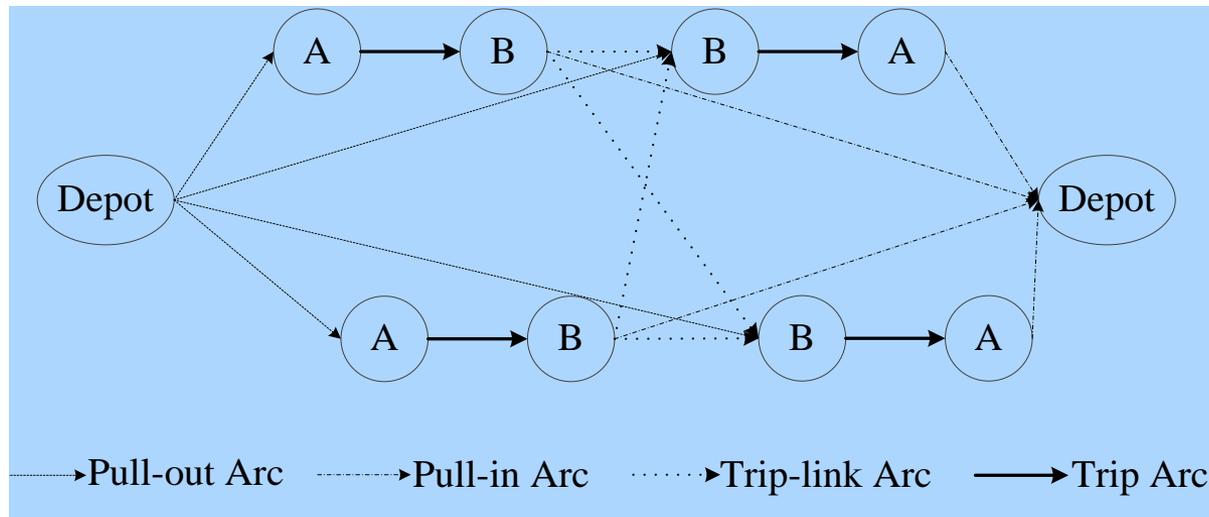
Vehicle Scheduling Problem (VSP)

- ❖ **Vehicle Scheduling Problem (VSP)** is concerned with the allocation of a set of **trips** in a predetermined timetable to a fleet of vehicles, in such a way that
 - the total **number of vehicles** and **operating cost** are minimized .
- ❖ An efficient schedule can bring transit operators a considerable saving on property and operating cost.
- ❖ **Each trip** contains a start time at its departure point and an end time at its arrival point.

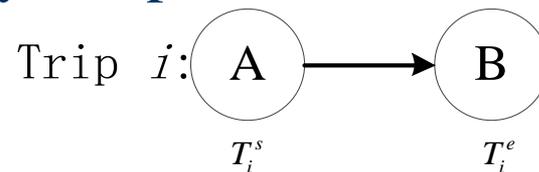


The network flow representation of VSP

- ❖ The VSP can be represented as a network flow

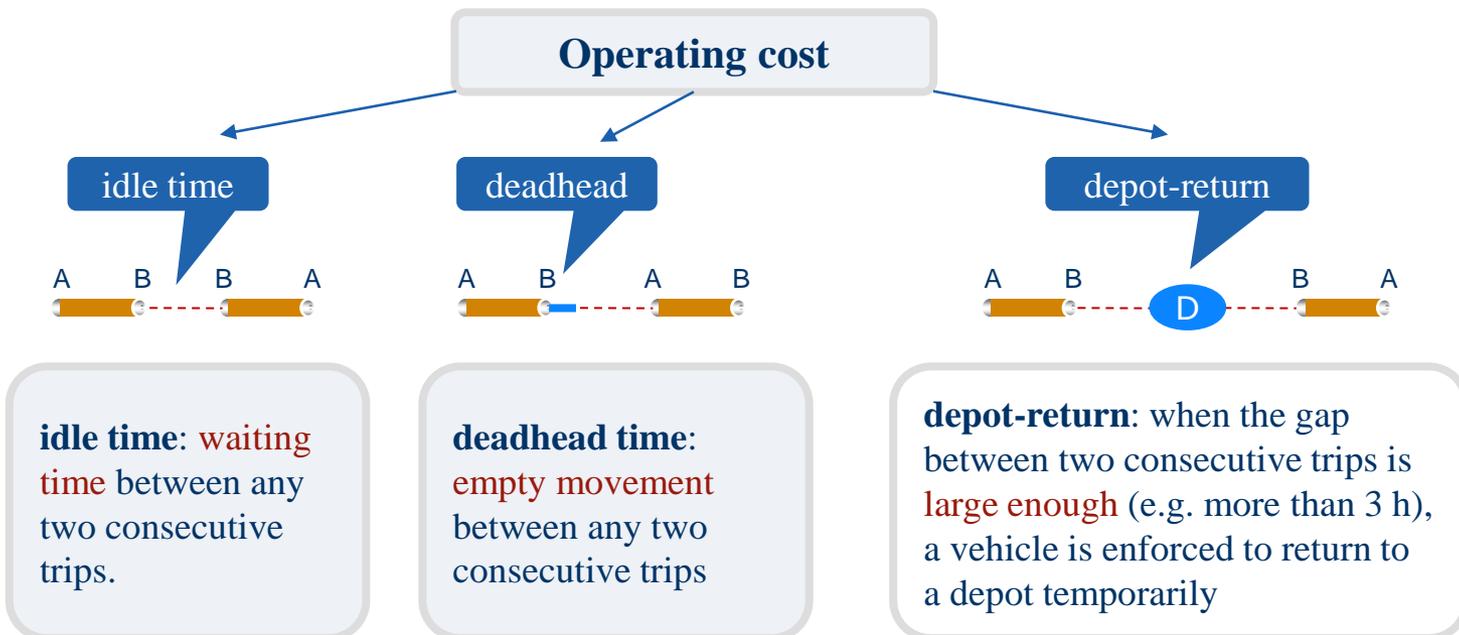
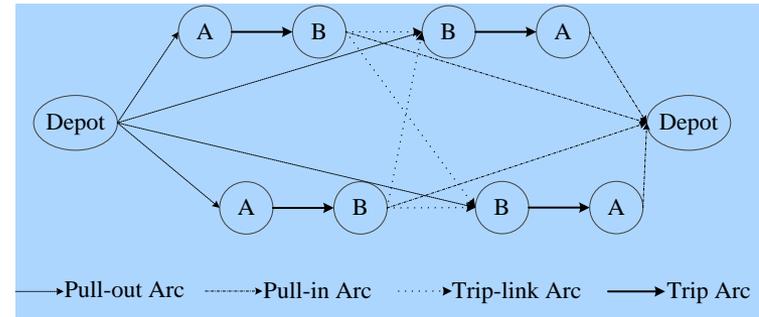


- ❖ Each trip is represented as a departure node and an arrival node, connected by a trip arc:



The objectives of VSP

1. To minimize the fleet size
2. To minimize the total operating cost



Formulations of traditional VSPs

- ❖ Let P denotes the set of pull-out arcs, Q denotes the set of pull-out arcs, R denotes the set of trip-link arcs,
 $A = P \cup Q \cup R$ denotes the set of all arcs

- ❖ Given a depot d and a set of trips T , the **VSP with single depot** can be modelled as follows.



$$\begin{aligned} \min \sum_{(i,j) \in A} c_{ij} x_{ij} & \quad (1) \\ \text{st. } \sum_{i:(i,j) \in A} x_{ij} - \sum_{i:(j,i) \in A} x_{ji} = 0, \quad \forall j \in T & \quad (2) \\ \sum_{j:(d,j) \in P} x_{dj} - \sum_{i:(i,d) \in Q} x_{id} = 0 & \quad (3) \\ \sum_{i:(i,j) \in A} x_{ij} = 1, \quad \forall j \in T & \quad (4) \\ x_{ij} \in \{0,1\}, \quad \forall (i,j) \in A & \quad (5) \end{aligned}$$

- ❖ Given a set of depots D and a set of trips T , the **VSP with multi depots (MDVSP)** can be modelled as follows.

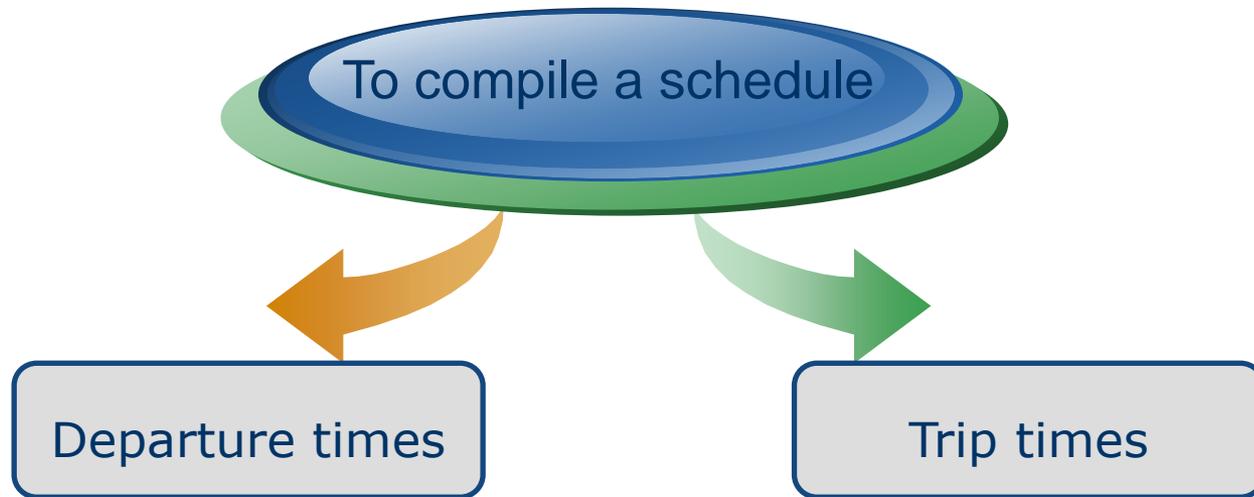
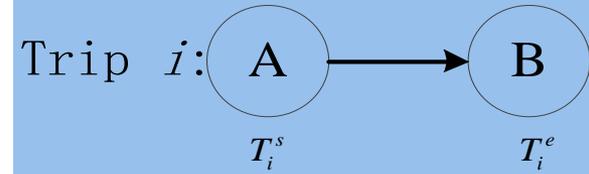


$$\begin{aligned} \min \sum_{d \in D} \sum_{(i,j) \in A^d} c_{ij}^d \cdot x_{ij}^d & \quad (1) \\ \text{st. } \sum_{i:(i,j) \in A^d} x_{ij} - \sum_{i:(j,i) \in A^d} x_{ji} = 0, \quad \forall j \in T, \forall d \in D & \quad (2) \\ \sum_{j:(d,j) \in P} x_{dj} - \sum_{i:(i,d) \in Q} x_{id} = 0, \quad \forall d \in D & \quad (3) \\ \sum_{d \in D} \sum_{i:(i,j) \in A^d} x_{ij} = 1, \quad \forall j \in T & \quad (4) \\ x_{ij} \in \{0,1\}, \quad \forall (i,j) \in A & \quad (5) \end{aligned}$$

VSP based on fixed trip times

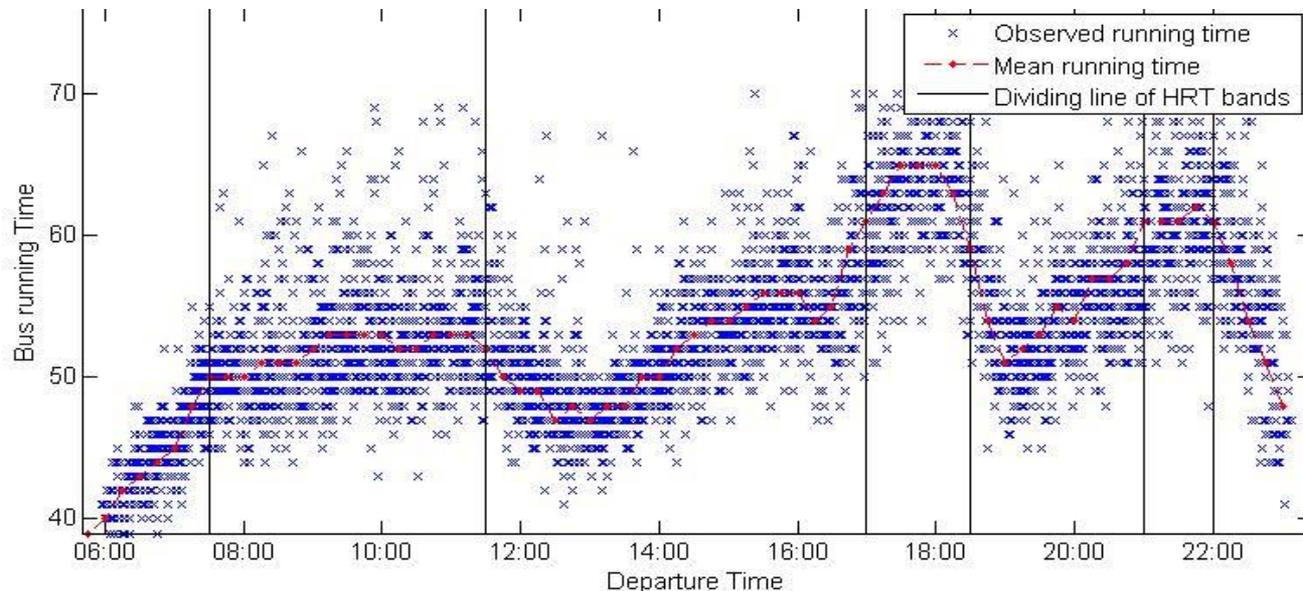
❖ Each trip in the timetable has

- a fixed departure time
- a fixed trip time –
called **scheduled trip time (ST)**



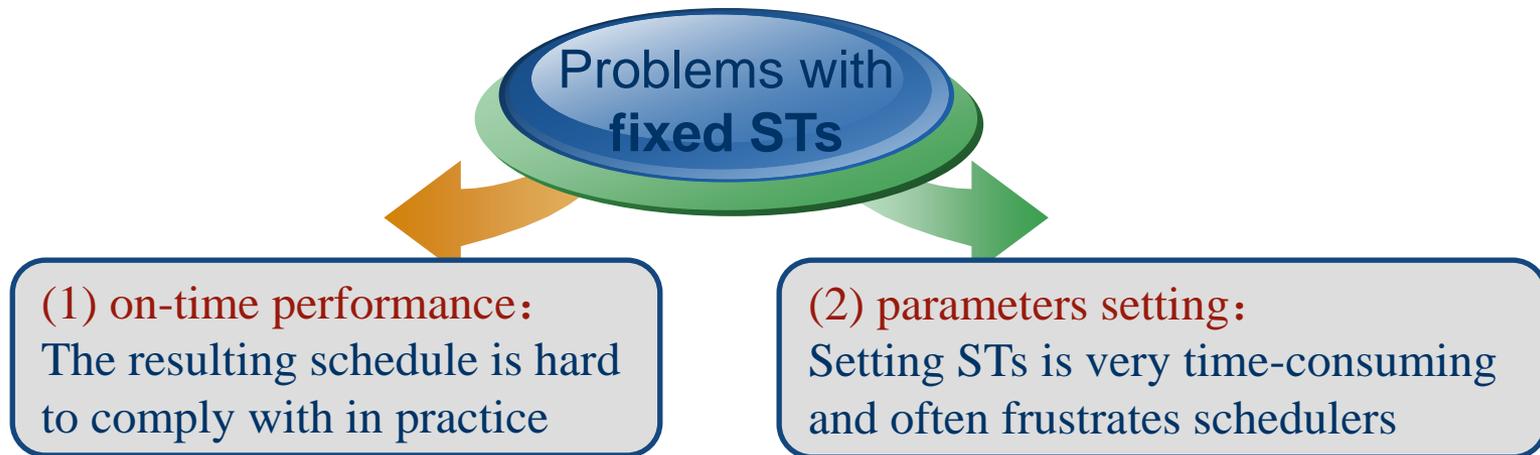
Variation of trip times

- ❖ The travel time at each route segment usually **varies dramatically** in different (e.g. peak- or off-peak) periods through a day due to fickle traffic, uncertain passenger demands and vehicle malfunction, etc.
- ❖ It is well known that the travel time is **hard to be precisely measured and predicted** (Chakroborty and Kikuchi, 2004; Vu and Khan, 2010; Ng et al., 2011).



Two main problems raised in the traditional VSP

- ❖ Due to trips times vary dramatically during a day, therefore, based on fixed **scheduled trip times (STs)**,

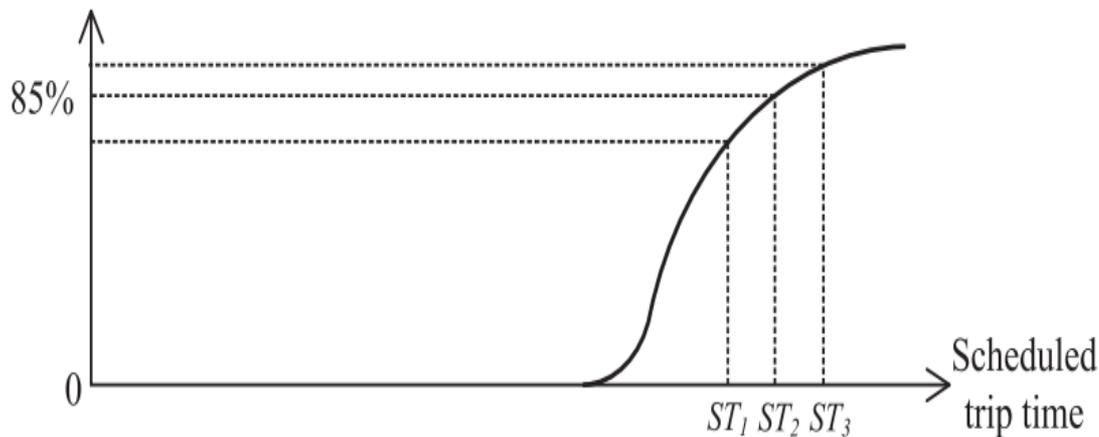


- ❖ To **increase** the on-time performance of a schedule, the common practice is to increase the scheduled trip times(STs).

The influence of increasing STs

- ❖ With enlarged scheduled trip times, the on-time performance of the resulting schedule will increase.

On-time performance



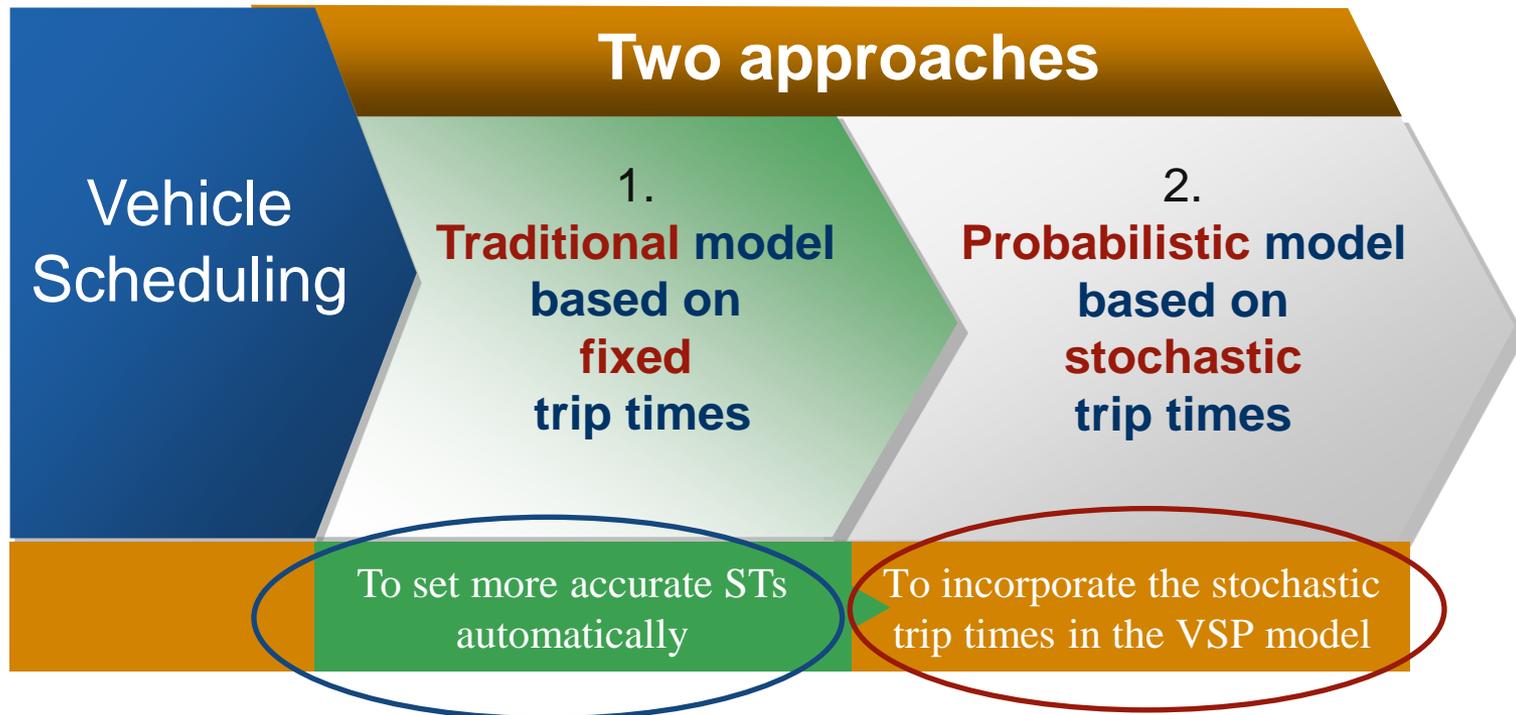
However, enlarged STs may lead to the **increase of the schedule's operating cost**, even more vehicles are needed.

- ❖ STs affect service reliability
 - Longer STs: high cost
 - Shorter STs: low on-time performance

❖ Therefore, it is non-trivial to set suitable scheduled trip times.

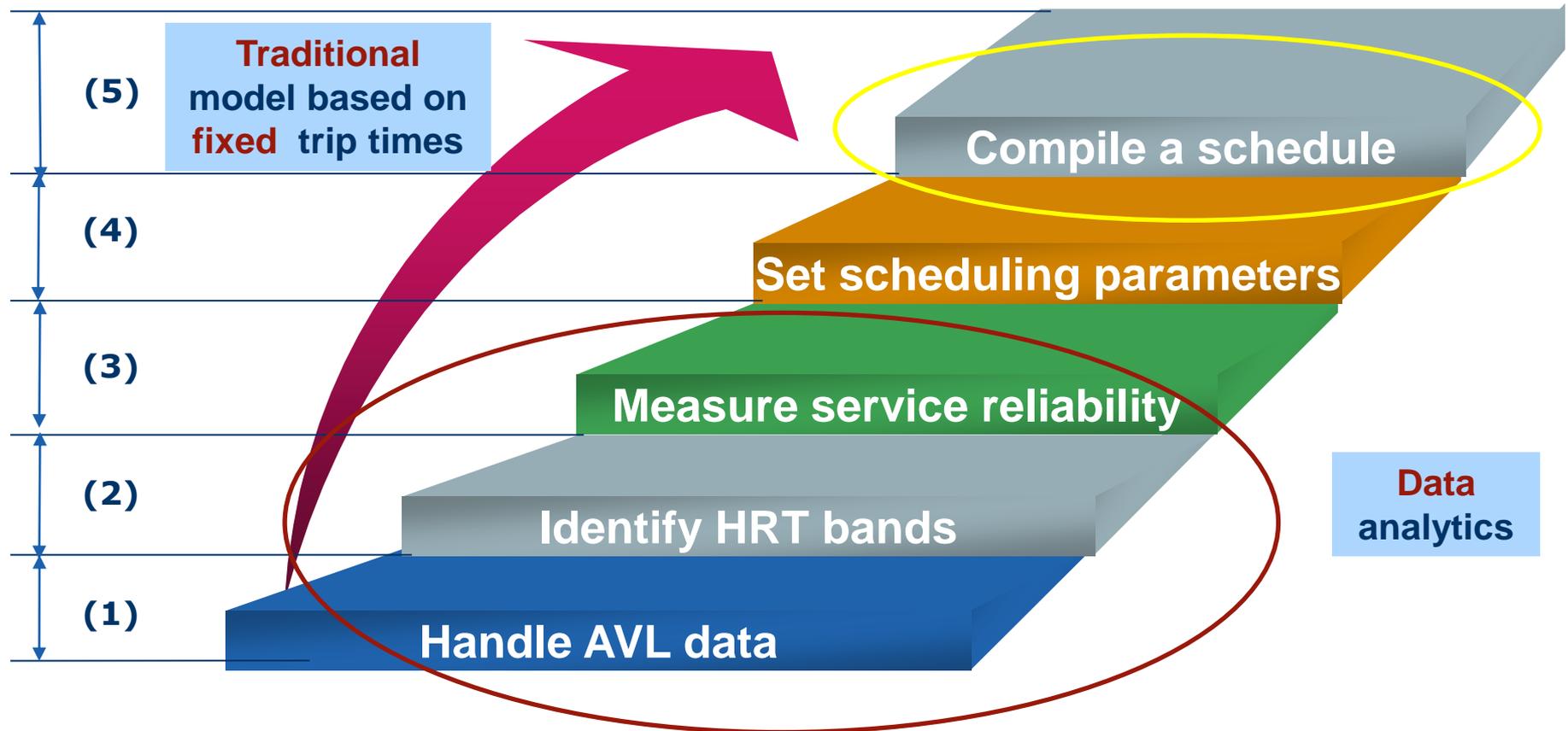
Vehicle scheduling based on AVL data

- ❖ **Vehicle scheduling approaches based on AVL data**, which aim to
 - increase the **on-time performance** of compiled schedules
 - reduce the **pressure** of human schedulers



Vehicle scheduling based on AVL data

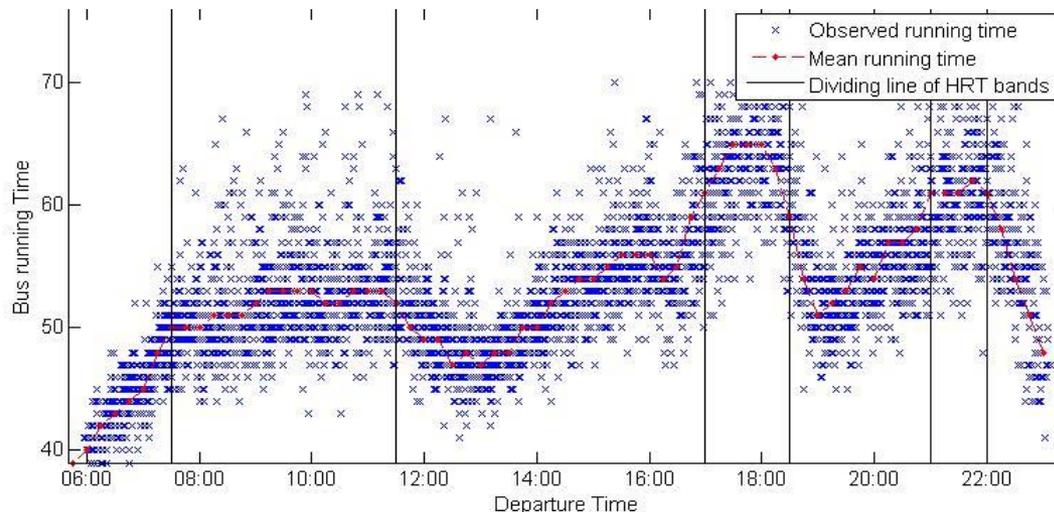
❖ The approach is composed of the following steps:



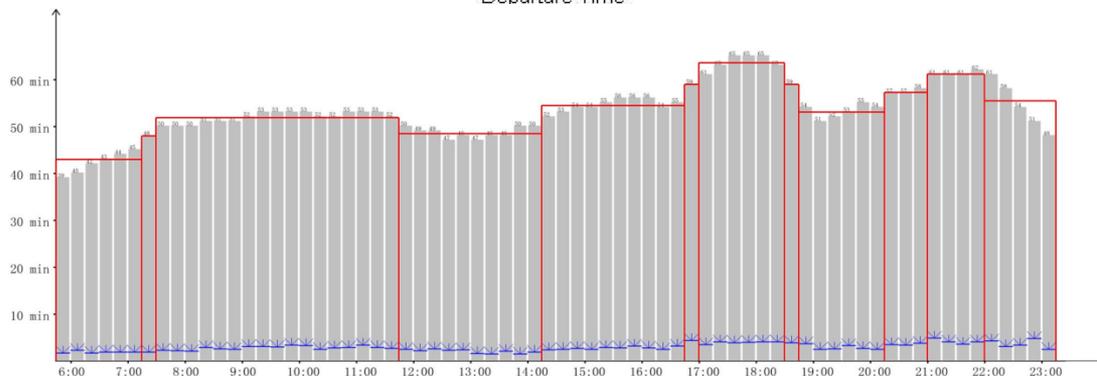
❖ To set more accurate STs automatically

Vehicle scheduling with stochastic trip times

- ❖ Handling AVL data, partitioning service span into HRT periods based on observed travel times



7 HRT bands/periods

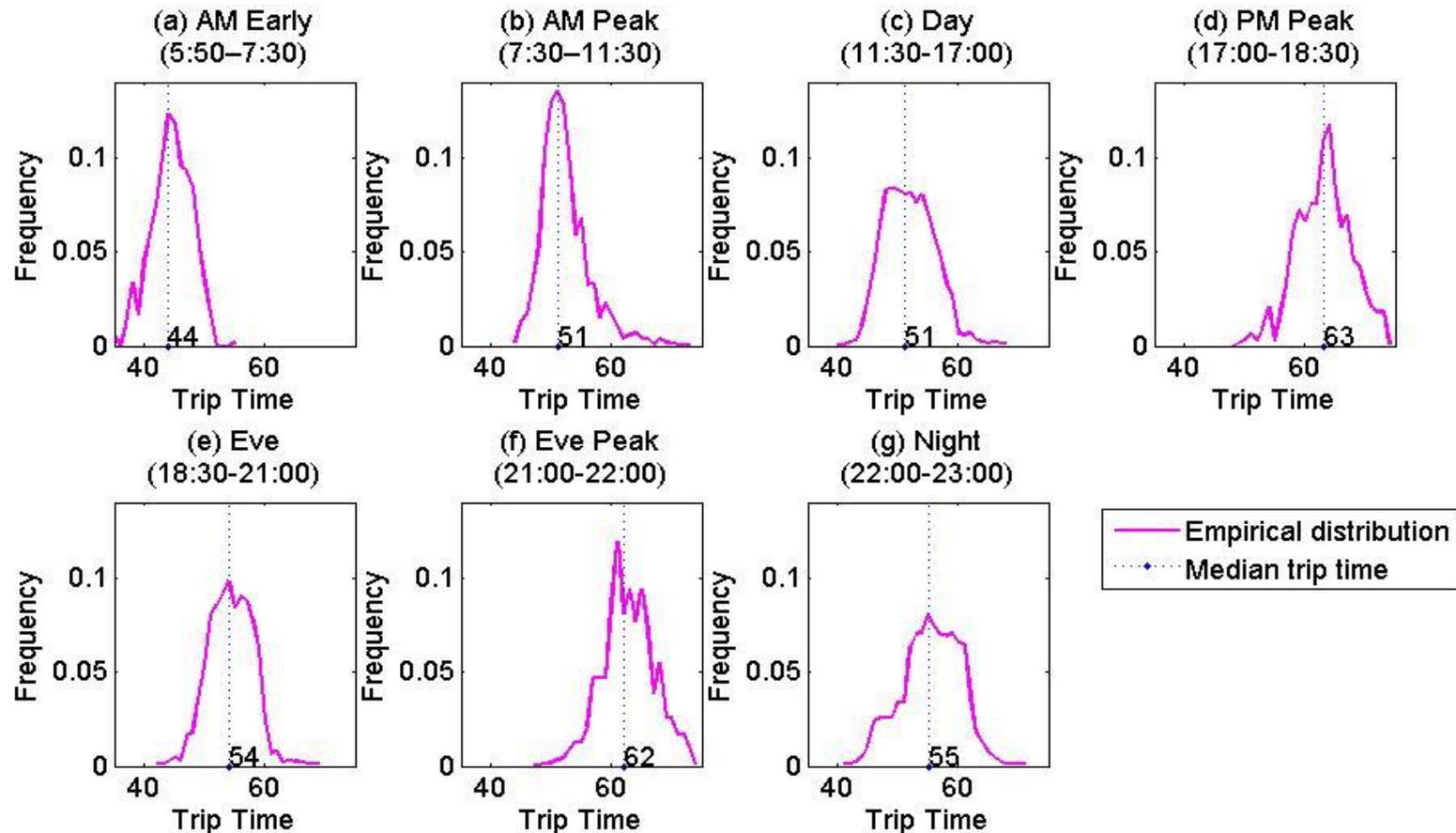


Mean trip time in an initial time band
 Standard deviation in an initial time band
 Mean trip time in a HRT band

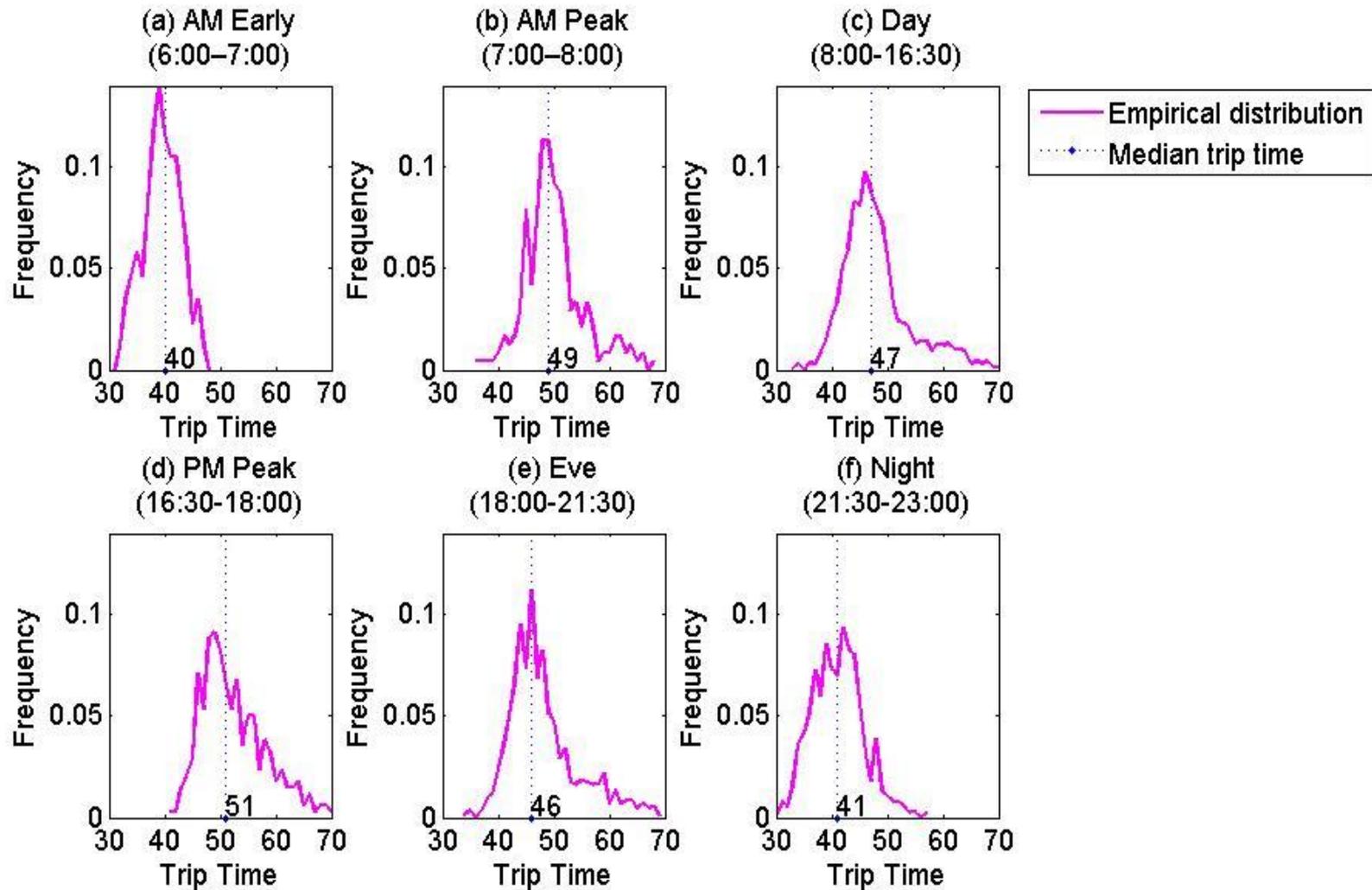


Empirical distributions of HKB4 (outbound)

- ❖ The **trip time distribution** for each HRT period is abstracted based on the AVL data

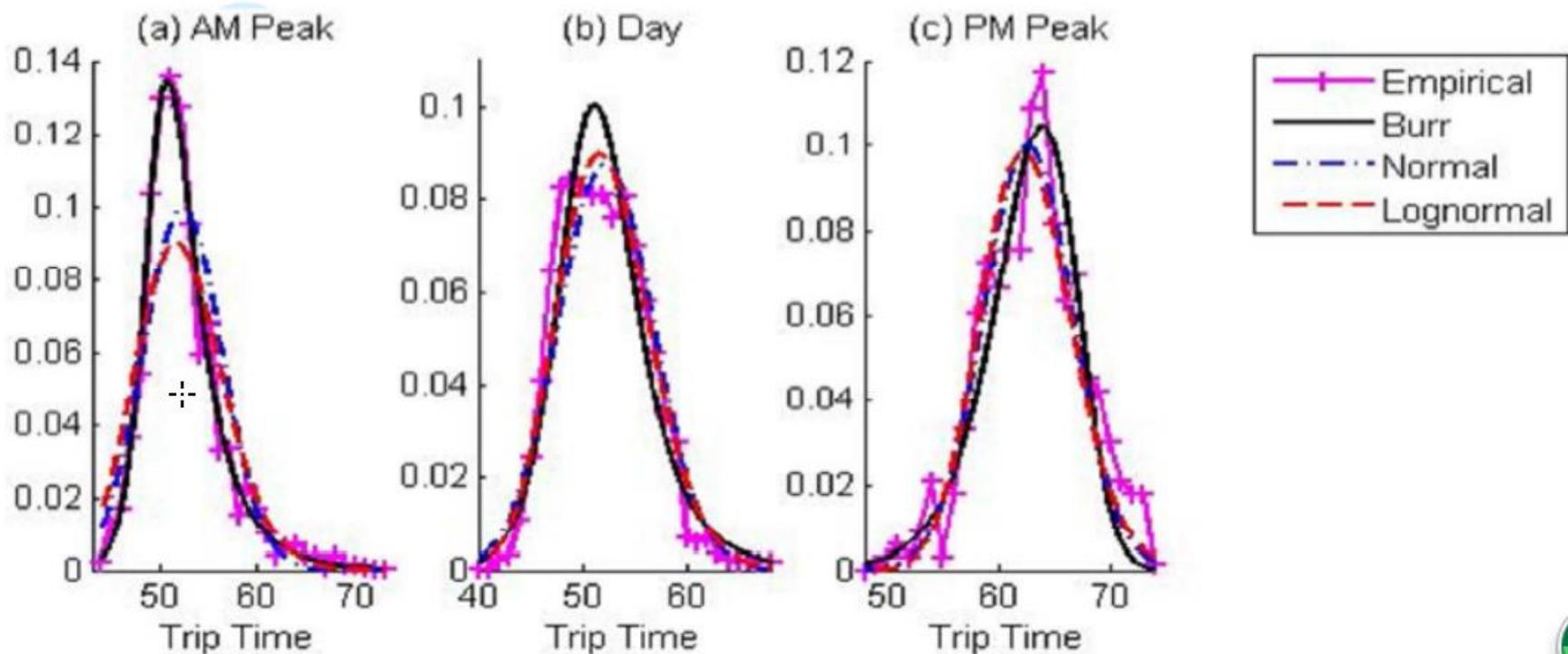


Empirical distributions of HKB4 (inbound)



Demonstrate the fitting of the distribution models to the empirical distributions

- ❖ It can be seen that the Burr distribution can match better the data



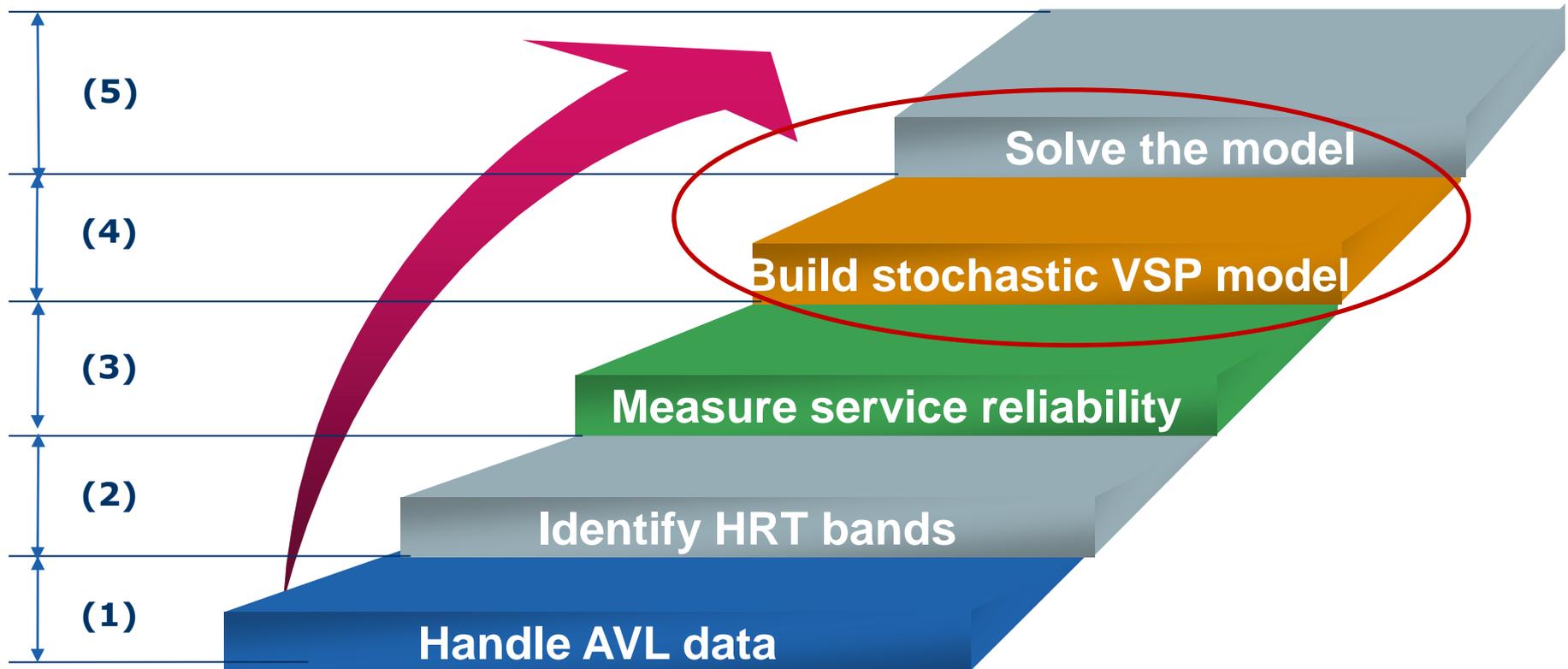
A compound travel time model

❖ Data from Bus Line 4 in Haikou

HRT	样本数	AIC 指数				最优分布模型
		正态	Lognormal	Gamma	Burr	
AM early	357	1922.3	1929.9	1926.5	1924.6	Normal
AM peak	1019	5729.2	5652.7	5677.2	5523.4	Burr
Day	1367	8020.1	7964.5	7980.4	7989.7	Lognormal
上行 PM peak	332	1864.5	1873.4	1870.1	1863.1	Burr
Eve	620	3576.8	3554.9	3560.9	3556.0	Lognormal
Eve peak	235	1337.6	1339.8	1338.7	1344.5	Normal
Night	244	1538.0	1543.7	1541.1	1540.9	Normal
AM early	86	495.7	479.9	492.1	474.7	Burr
AM peak	239	1450.5	1440.6	1443.8	1430.6	Burr
下行 Day	2090	13485.5	13269.1	13335.8	13071.4	Burr
PM peak	338	2162.6	2136.6	2144.6	2123.8	Burr
Eve	857	5638.4	5536.4	8499.6	5431.6	Burr
Night	385	2413.0	2368.7	2381.8	2350.3	Burr

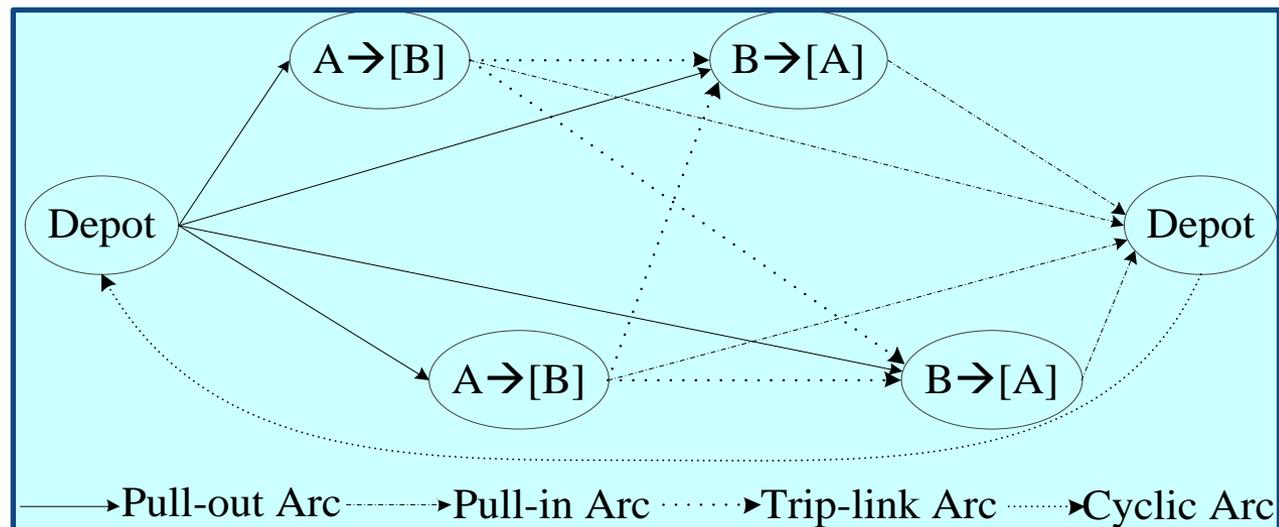
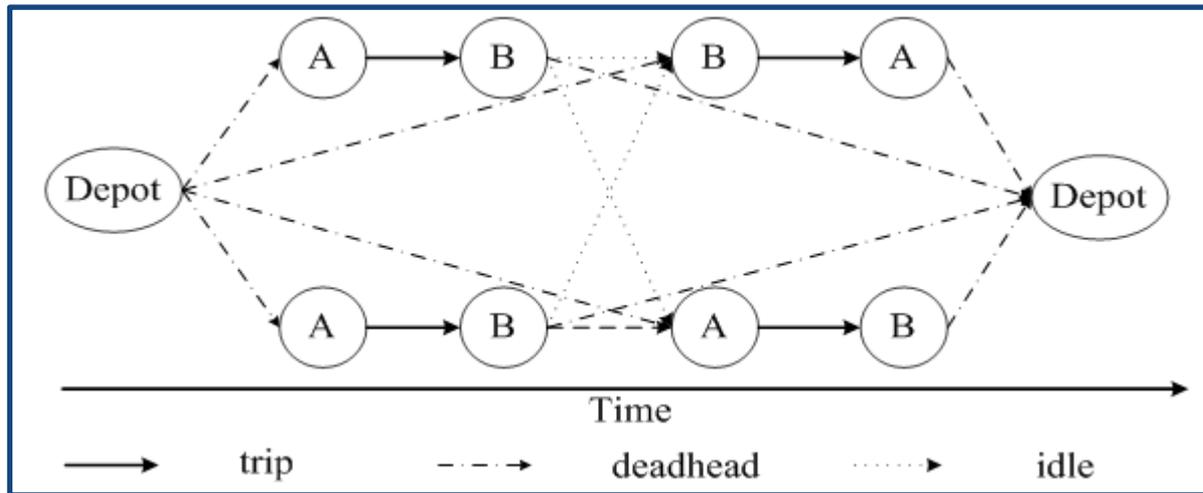
Stochastic Vehicle scheduling

- To incorporate the stochastic trip times in the VSP model
(in the light of AVL data)

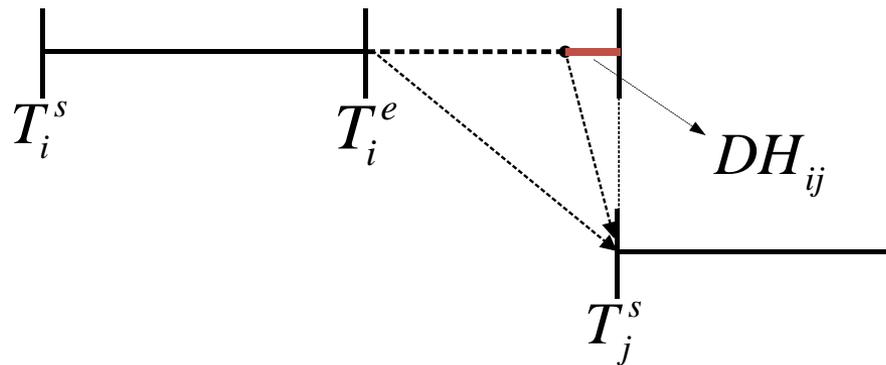


Compile a schedule based on stochastic trip times

❖ Diverting from the traditional vehicle scheduling,

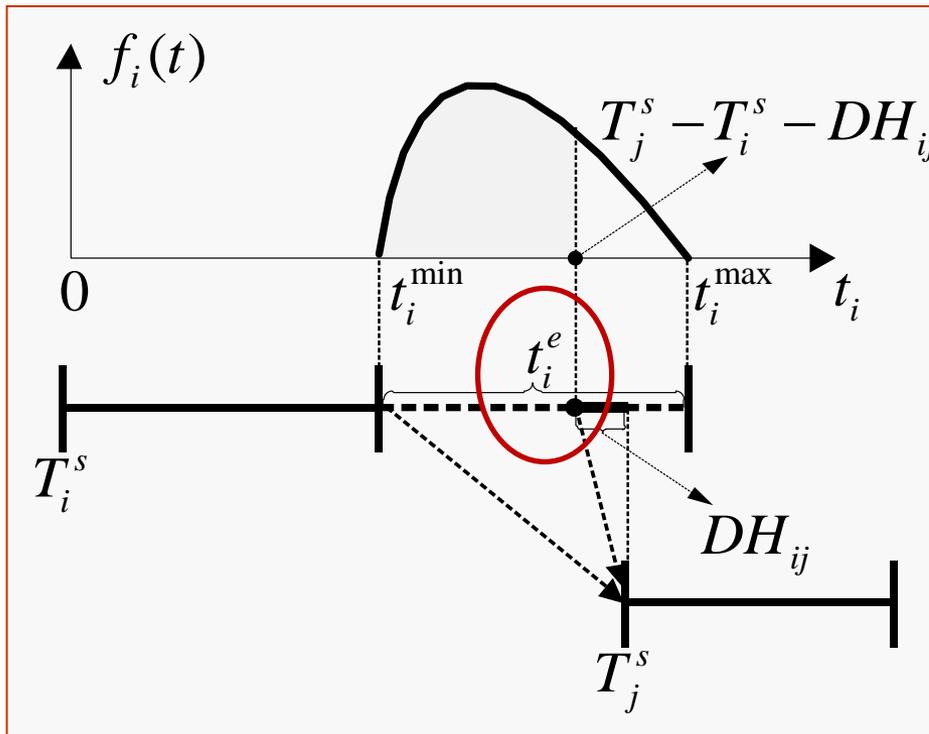


Compatibility of a pair of trips



❖ Bertossi et al.(1987) defined the **compatibility** of any two trips i and j as

$$T_j^s - T_i^e \geq DH_{ij}$$



❖ The **compatibility probability** can be expressed as

$$P\{i \Theta j\} = \int_0^{T_j^s - T_i^e - DH_{ij}} f_i(t) dt$$

Compatibility and incompatibility probability

- ❖ Between the trips i and j , an arc (i, j) is defined if $P\{i \ominus j\} > 0$; otherwise, no arc exists.
- ❖ Each arc (i, j) is also associated with an **incompatibility probability**

$$P\{i \overline{\ominus} j\} = 1 - P\{i \ominus j\}$$

where, $i \overline{\ominus} j$ denotes the incompatibility of trips i and j .

A probabilistic model for VSP (PVSP)

$$\min \sum_{(d,j) \in P} C_{veh} x_{dj} + \sum_{(i,j) \in A} C_{ij} x_{ij} + \alpha \cdot \sum_{(i,j) \in R} P_{ij} x_{ij} \quad (1)$$

$$\text{s.t.} \quad \sum_{i:(i,j) \in A} x_{ij} - \sum_{i:(j,i) \in A} x_{ji} = 0, \quad \forall j \in T \quad (2)$$

$$\sum_{j:(d,j) \in P} x_{dj} - \sum_{i:(i,d) \in Q} x_{id} = 0, \quad \forall d \in D \quad (3)$$

$$\sum_{i:(i,j) \in A} x_{ij} = 1, \quad \forall j \in T \quad (4)$$

$$x_{ij} \in \{0,1\}, \quad \forall (i,j) \in A \quad (5)$$

❖ The cost and penalty for the trip-link arcs:

- Cost (deadhead & idle time)

$$C_{ij} = DH_{ij} + E(ID_{ij}) = DH_{ij} + \int_0^{T_j^S - T_i^S - DH_{ij}} (T_j^S - T_i^S - t - DH_{ij}) f_i(t) dt$$

- Penalty (infeasible time --- When the trips i and j are incompatible, $IF_{ij} \geq 0$; otherwise, let $IF_{ij} = 0$.)

$$P_{ij} = E(IF_{ij}^2) = \int_{T_j^S - T_i^S - DH_{ij}}^{+\infty} (T_i^S + t - T_j^S + DH_{ij})^2 f_i(t) dt$$

Enhancement of PVSP considering delay propagation

- ❖ In practice, the delay can propagate between the consecutive trips operated by the same vehicle.
- ❖ Such **delay propagation** can cause more delays to the pre-compiled schedule.
- ❖ Considering delay propagation between two trips i and j :
 - The departure time of trip j is no-longer deterministic and depends on the compatible probability of trips i and j .
 - When the trips i and j are compatible, trip j can either depart on-time, or, experience a departure delay.
 - **The arrival time of trip j is affected by both its departure time and its trip time.**

An enhanced probabilistic model for VSP (PVSP-DP)

- ❖ Formulate the probabilistic model of VSP with delay propagation (PVSP-DP)

1. The expected idle time $E(ID_{ij})$ in the cost of a trip-link arc, is expressed as:

$$E(ID_{ij}) = \int_0^{T_j^s - DH_{ij}} (T_j^s - t_i^e - DH_{ij}) f_i^e(t) dt \quad (1)$$

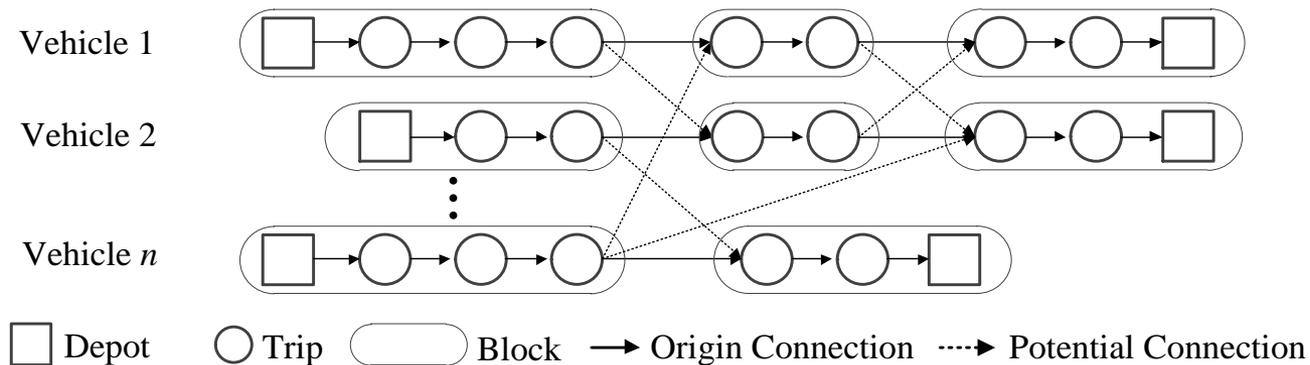
2. The penalty of trip-link arc, is expressed as:

$$E(DD_{ij}^2) = \int_{T_j^s - DH_{ij}}^{+\infty} (t_i^e - T_j^s + DH_{ij})^2 f_i^e(t) dt \quad (2)$$

3. The other parts of the model remains unchanged.

A heuristic approach for the PVSP-DP model

- ❖ **Stage 1: Matching based heuristic to form an initial schedule:**
 - (1) **The tier partitioning** is to partition all the trips into different tiers;
 - (2) **Link all the tiers** into a initial schedule by solving a sequence of matching problem.
- ❖ **Stage 2: Refinement.** **The iterative greedy local search method** is to firstly break part of the arcs in the current schedule, and then to re-link the nodes by solving the sub-problem containing a small subset of nodes.



Benchmark schedules with DVSP model

❖ Compile benchmark schedules with DVSP model:

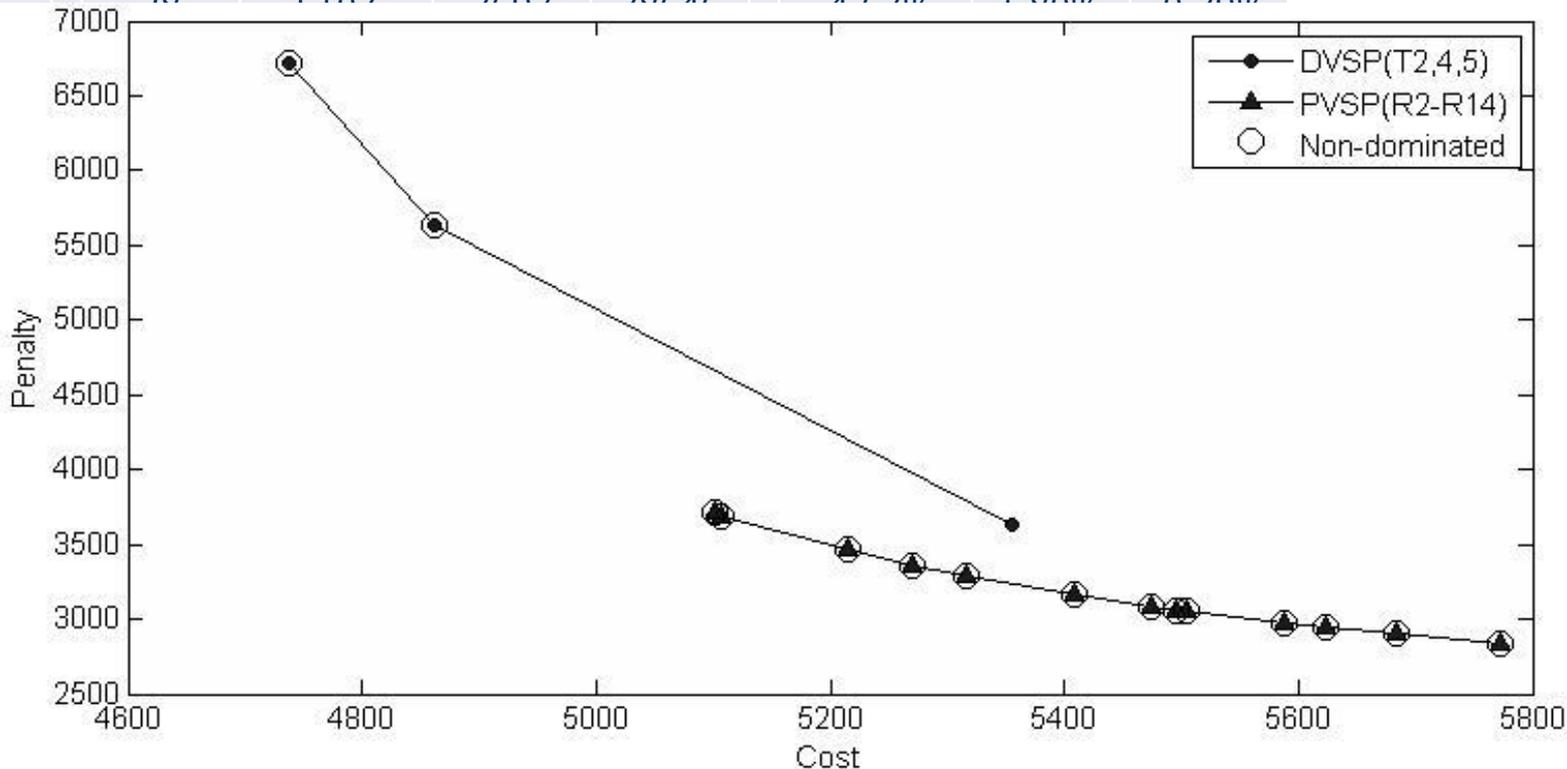
- To rules-of-thumb for setting the scheduled trip times are adopted: the 85th percentile of trip time; the sum of mean and standard deviation of trip time.
- Adjust by +/- 1 minute based on the two rules-of-thumb, 6 groups of scheduled trip times forms 6 problems of vehicle scheduling. Solved by CPLEX, 6 vehicle schedules are produced.
- From the results, along the increment of ST, the operating cost of schedule is increased, while the penalty is decreased(better on-time performance), however, the fleet size is greatly influenced by the ST.
- The best benchmark schedule is selected with respect to the best on-time performance of 28 vehicles

Problem	ST setting method	ST adjustment	Fleet size	Trip time	E(ID)	Deadhead	Operating cost	Penalty	Objective value
T1	Rule-of-thumb 1	-1	27	22816	2313	1980	4293	9577	45659
T2		0	28	23270	2698	2040	4738	6713	42807
T3		1	29	23724	3372	2100	5472	3497	39718
T4	Rule-of-thumb 2	-1	28	23290	2762	2100	4862	5637	41317
T5#		0	28	<u>23744</u>	<u>3315</u>	<u>2040</u>	<u>5355</u>	<u>3627</u>	<u>38796</u>
T6		1	29	24198	3611	2250	5861	2769	39015

Schedules compiled with PVSP model

❖ Schedules produced by PVSP model:

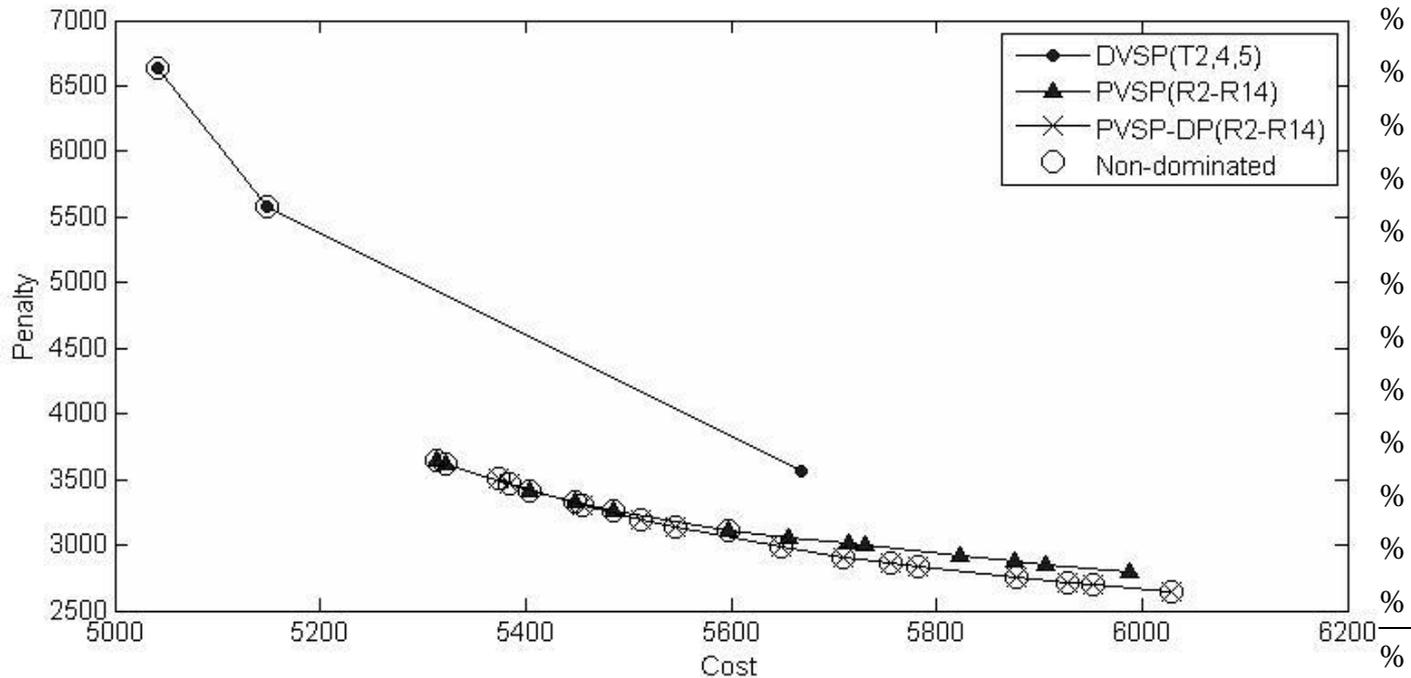
Problem	α	Results produced by PSVP model				RPD over the benchmark		
		Fleet size	Operating cost	Penalty	Objective value	Operating cost	Penalty	Objective value
R1	0.2	27	5394	6098	41541	0.73%	68.11%	7.08%
R2	0.3	28	5102	3708	38664	-4.72%	2.22%	-0.34%
R3	0.4	28	5107	3603	38646	-4.63%	1.80%	0.30%
R4	0.5	28	5107	3603	38646	-4.63%	1.80%	0.30%
R5	0.6	28	5107	3603	38646	-4.63%	1.80%	0.30%
R6	0.7	28	5107	3603	38646	-4.63%	1.80%	0.30%
R7	0.8	28	5107	3603	38646	-4.63%	1.80%	0.30%
R8	0.9	28	5107	3603	38646	-4.63%	1.80%	0.30%
R9	1	28	5107	3603	38646	-4.63%	1.80%	0.30%
R10	1.1	28	5107	3603	38646	-4.63%	1.80%	0.30%
R11	1.2	28	5107	3603	38646	-4.63%	1.80%	0.30%
R12	1.3	28	5107	3603	38646	-4.63%	1.80%	0.30%
R13	1.4	28	5107	3603	38646	-4.63%	1.80%	0.30%
R14	1.5	28	5107	3603	38646	-4.63%	1.80%	0.30%
R15	1.6	28	5107	3603	38646	-4.63%	1.80%	0.30%
AVG								



Schedules produced by the PVSP-DP model

❖ Solution with PVSP-DP model:

Index α	Fleet size		Cost				Penalty				Objective value				
	PVSP	PVSP -DP	PVSP	RPD	PVSP -DP	RPD	PVSP	RPD	PVSP -DP	RPD	PVSP	RPD	PVSP -DP	RPD	
R1	0.2	27	27	5618	4.91%	5670	5.88%	5997	65.34%	5857	61.48%	41614	7.26%	41456	6.86%
R2	0.3	28	28	5314	-0.77%	5373	0.34%	3641	0.37%	3507	-3.31%	38775	-0.05%	38634	-0.42%
R3	0.4	28	28	5323	-0.60%	5385	0.56%	3614	-0.37%	3474	-4.23%	38744	-0.13%	38596	-0.52%
R4	0.5														
R5	0.6														
R6	0.7														
R7	0.8														
R8	0.9														
R9	1														
R10	1.1														
R11	1.2														
R12	1.3														
R13	1.4														
R14	1.5														
R15	1.6														



Conclusions

- (1) Experimental results show that both of the probabilistic models can produce the schedules with the same fleet size but considerably higher on-time performance than the best known schedule generated under DVSP model.
- (2) Comparing the two probabilistic models, the PVSP-DP model may further increase the on-time performance, while the fleet size remains the same but with a little compromise in the operating cost.
- (3) Moreover, with the aid of the probabilistic models, human schedulers can be saved from the work of determining scheduled trip times, which is non-trivial and often frustrates schedulers.

我喜欢公交!

Thank You!

Questions?

