Time-dependent ambulance allocation considering data-driven empirically required coverage

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Basic models Optimization model A real world problem

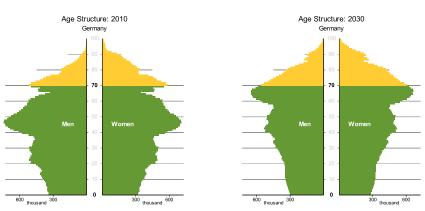
Outline

- 1. Motivation
- 2. Basic idea and models
- 3. Data driven optimization model for tactical ambulance planning
- 4. A real world EMS planning problem in the city of Bochum
- 5. Conclusions and Outlook

Motivation Basic models Optimization model A real world problem Outloo

Motivation

Aging population



- ▶ Age class of 70+ causes 50 % of ambulance operations
- ► EMS demand increases

source: Statistisches Bundesamt 2013

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Motivation

Aging population

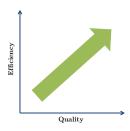


- ▶ Age class of 70+ causes 50 % of ambulance operations
- ► EMS demand increases

source: Statistisches Bundesamt 2013

Motivation

- Access to emergency medical services (EMS) is crucial
- Ensuring optimal supply quality
 - Short arrival time
 - Coverage of entire demand area
 - High degree of achievement
 - ⇒ Dependent on location and number of ambulances



Efficiency

- Reduction of fixed costs (avoid overcapacity)
- Utilization of ambulances

Quality criteria and objectives

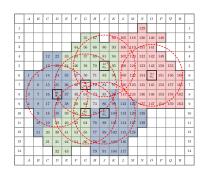
Evaluation of emergency medical services (objective of EMS provider):

▶ Degree of achievement (ex post):

```
# operations within a time standard T
# total number of operations
```

Objectives in literature (ex ante):

- single coverage models
- double coverage models
- busy fraction models
- queue models
- hypercube models



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Optimal quality in emergency medical services

Research project: 2 years (work in progress)





Focus:

- Analysis of required resources for EMS
- Analysis of time-dependent demand and speed fluctuations for EMS
- ► Tactical and strategic planning horizon
- Stochastic influences, uncertain parameters (demand, speed)

Goal:

- Dynamic (and robust) optimization model
- ⇒ IT-based decision support tool for local EMS providers

Decision support tool

Decision support

Optimization

location, allocation and resource planning

empirical demand, required/adequate coverage

Simulation

degree of achievement, resources, utilization

evaluations

Forecast

trends, development

population, experts

Decision support tool

Decision support

Optimization

location, allocation and resource planning

empirical demand, required/adequate coverage

Simulation

degree of achievement, resources, utilization

evaluations

Forecast

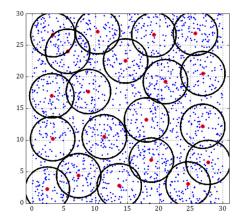
trends, development

population, experts

Basic models

Basic covering location models

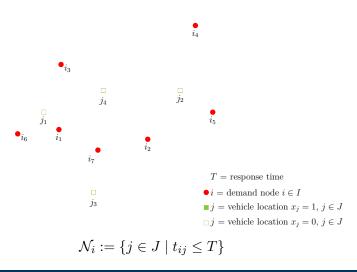
Idea: Demand nodes i have to be covered within a time standard T



source: Zarandi et al. 2011-The large scale maximal covering location problem; page 1565

Set Covering Problem (SCP)

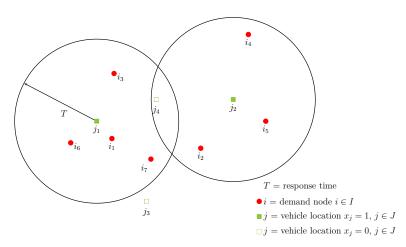
Toregas et al. 1971



Lara Wiesche (RUB)

Set Covering Problem (SCP)

Toregas et al. 1971

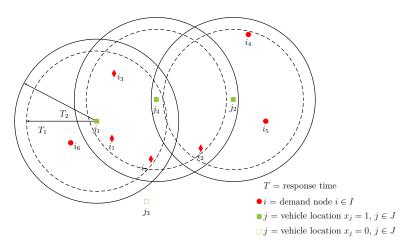


 $\mathcal{N}_i := \{ j \in J \mid t_{ij} \le T \}$

Lara Wiesche (RUB)

Double Standard Model (DSM)

Gendreau et al. 1997



$$\mathcal{N}_{i}^{T_{\ell}} := \{ j \in J \mid t_{ij} \leq T_{\ell}, \ \ell \in \{1, 2\} \}$$

Lara Wiesche (RUB)

Double Standard Model (DSM)

Gendreau et al. 1997

$$\max \sum_{i \in I} d_i x_i^2$$
s. t.
$$\sum_{j \in \mathcal{N}_i^{T_2}} y_j \ge 1 \qquad \forall i \in I$$

$$\sum_{i \in I} d_i x_i^1 \ge \alpha \sum_{i \in I} d_i$$

$$x_i^1 \ge x_i^2 \qquad \forall i \in I$$

$$\sum_{j \in \mathcal{N}_i^{T_1}} y_j \ge x_i^1 + x_i^2 \qquad \forall i \in I$$

$$\sum_{j \in J} y_j = p$$

$$ightharpoonup d_i$$
: demand at node i

$$\mathcal{N}_i^{T_\ell} := \{ j \in J \mid t_{ij} \le T_\ell \}$$

$$T_1 < T_2$$

p: number of ambulances (fleet size)

$$x_i^k = \left\{ \begin{array}{ll} 1, & \text{if demand node } i \text{ is} \\ & \text{covered } k \in \{1,2\} \text{ times} \\ 0, & \text{else.} \end{array} \right.$$

 y_i : number of ambulances at node j

 $y_i \in \mathbb{N}_0$

 $x_i^1, x_i^2 \in \{0, 1\}$

 $\forall i \in I$

 $\forall j \in J$

Double Standard Model (DSM)

Gendreau et al. 1997 — Limitations: (1) static consideration (2) fixed double coverage

$$\begin{aligned} & \max & & \sum_{i \in I} \boldsymbol{d}_i \boldsymbol{x}_i^2 \\ & \text{s. t.} & & \sum_{j \in \mathcal{N}_i^{T_2}} y_j \geq 1 & \forall i \in I \\ & & \sum_{i \in I} \boldsymbol{d}_i \boldsymbol{x}_i^1 \geq \alpha \sum_{i \in I} \boldsymbol{d}_i \\ & & \boldsymbol{x}_i^1 \geq \boldsymbol{x}_i^2 & \forall i \in I \\ & & \sum_{j \in \mathcal{N}_i^{T_1}} y_j \geq \boldsymbol{x}_i^1 + \boldsymbol{x}_i^2 & \forall i \in I \\ & & \sum_{j \in \mathcal{N}_i^{T_1}} y_j = \boldsymbol{p} \\ & & & \boldsymbol{x}_i^1, \boldsymbol{x}_i^2 \in \{0, 1\} & \forall i \in I \end{aligned}$$

- $ightharpoonup d_i$: demand at node i
- $\mathcal{N}_i^{T_\ell} := \{ j \in J \mid \mathbf{t}_{ij} \le T_\ell \}$ $T_1 < T_2$
- p: number of ambulances (fleet size)

$$x_i^k = \left\{ \begin{array}{ll} 1, & \text{if demand node } i \text{ is} \\ & \text{covered } k \! \in \! \{1, \textcolor{red}{\textbf{2}}\} \text{ times} \\ 0, & \text{else.} \end{array} \right.$$

 y_j : number of ambulances at node j

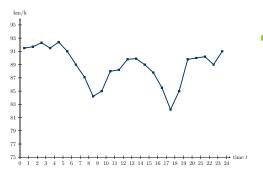
 $y_i \in \mathbb{N}_0$

 $\forall j \in J$

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Extensions of the Double Standard Model

Time-dependent parameter: speed

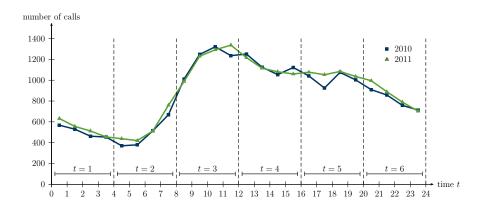


 Time-dependent speed on (city-)motorways (example of the city of Vienna)

Source: Kritzinger et al. (2011)

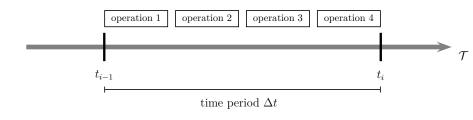
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Time-dependent parameter: demand



⇒ Dynamic considerations required

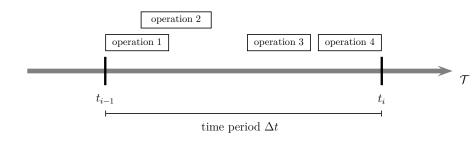
Time-dependent parameter: degree of required coverage



 ξ : number of ambulances, X : number of parallel operations

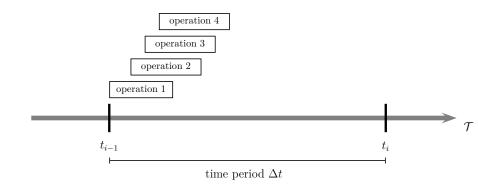
$$P({X \le \xi}) \ge \beta \quad (= 0.95)$$

Time-dependent parameter: degree of required coverage



 ξ : number of ambulances, X : number of parallel operations

$$P({X \le \xi}) \ge \beta \quad (= 0.95)$$



 ξ : number of ambulances, X : number of parallel operations

$$P({X \le \xi}) \ge \beta \quad (= 0.95)$$

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Extensions of the Double Standard Model

Existing:

- Speed is time-dependent and location-dependent
- ▶ Empirical investigation (e.g. Schmid/Doerner 2010; Degel et al. 2014)

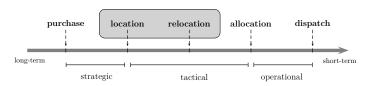
Extensions of the Double Standard Model

Existing:

- Speed is time-dependent and location-dependent
- ► Empirical investigation (e.g. Schmid/Doerner 2010; Degel et al. 2014)

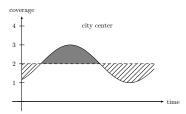
Additional:

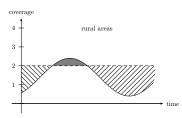
- Time- and location-dependent demand
- Empirically required coverage for each demand node/period
- Additional and flexible ambulance stations
- Dynamic and flexible allocation of ambulances to stations



New optimization approach

Objectives (1)





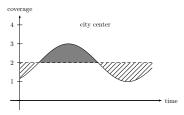
- Observed: differences between required and actual coverage
- ▶ Maximize the empirically required coverage:

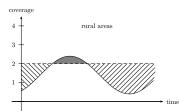
$$x_{it}^k := \begin{cases} 1, & \text{if demand node } i \text{ is coverd } k\text{-times in period } t, \\ 0, & \text{overwise.} \end{cases}$$

 $ightharpoonup k = \mathbf{e}(it)$ is determined empirically (\leftarrow parallel operations)

New optimization approach

Objectives (1)





- Observed: differences between required and actual coverage
- Maximize the empirically required coverage: $x_{it}^{e(it)} \in \{0,1\}$

$$\max \quad \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} d_{it} x_{it}^{e(it)}$$

k = e(it) is determined empirically (\leftarrow parallel operations)

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New optimization approach

Objectives (2)

- Relocations due to variations of demand/speed (penalty costs)
- Utilization of flexible ambulance stations (penalty costs)

$$\begin{aligned} & \min \quad \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{I}} u_{ijt} \\ & \min \quad \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{F}} y_{jt} \\ & y_{Dt} = p - p_t \\ & y_{jt} + \sum_{i \in \mathcal{J} \cup \{D\}} u_{ij(t+1)} - \sum_{i \in \mathcal{J} \cup \{D\}} u_{ji(t+1)} = y_{j(t+1)} \\ & y_{jT} + \sum_{i \in \mathcal{J} \cup \{D\}} u_{ij1} - \sum_{i \in \mathcal{J} \cup \{D\}} u_{ji1} = y_{j1} \\ & u_{ijt} \in \{0, 1\} \\ & y_{jt} \in \mathbb{N}_0 \end{aligned} \qquad \forall t \in \mathcal{T} \setminus \{T\}$$

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New optimization approach

Constraints

- covering constraints
- ▶ location constraints
- relocation constraints
- allocation constraints
- demand constraints

Bochum



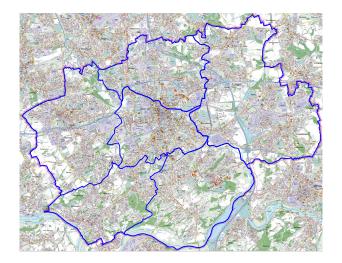
Coordinates $51^{\circ} 28' 55''N \quad 7^{\circ} 12' 57''E$



- 16th biggest city in Germany
- ightharpoonup Area: 145.4 km^2 (56.1 sq mi)
- Population: ca. 375,000
- Population density: $2,577/km^2$
- Services: about 21,000 operations per year
- Services per 1,000 inhabitants per year: about 56 operations
- ▶ 14 ambulances

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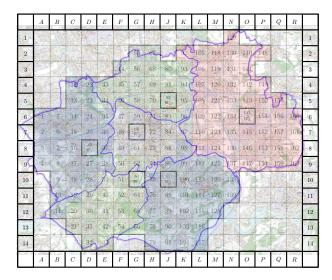
The city area of Bochum





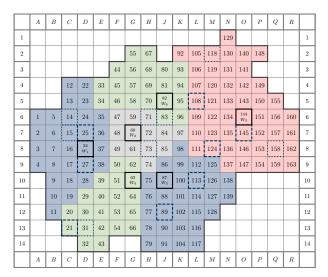
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The city area of Bochum





The city area of Bochum

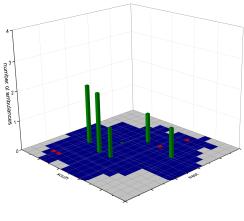


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Model results

Usage of flexible ambulance stations and relocations

Period t=1, number of ambulances 7

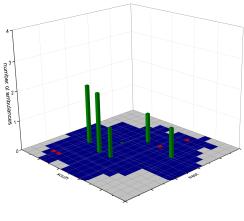


A real world problem

Model results

Usage of flexible ambulance stations and relocations

Period t = 2, number of ambulances 7

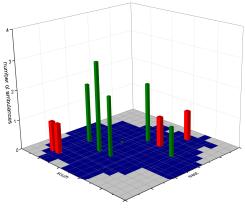


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Model results

Usage of flexible ambulance stations and relocations

Period t = 3, number of ambulances 14

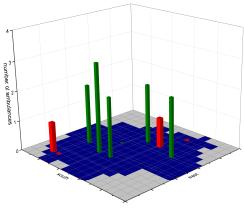


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Model results

Usage of flexible ambulance stations and relocations

Period t = 4, number of ambulances 13

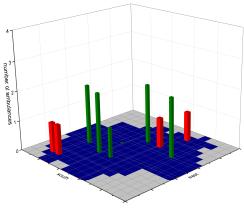


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Model results

Usage of flexible ambulance stations and relocations

Period t = 5, number of ambulances 13



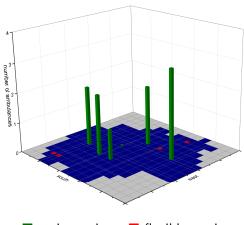


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Model results

Usage of flexible ambulance stations and relocations

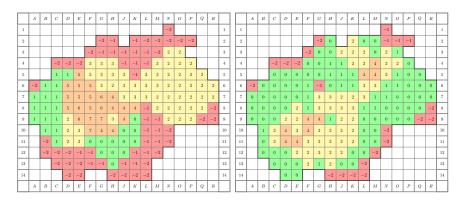
Period t = 6, number of ambulances 10



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Model results

Difference between resulting and required degree of coverage (period 8-12 a.m.)



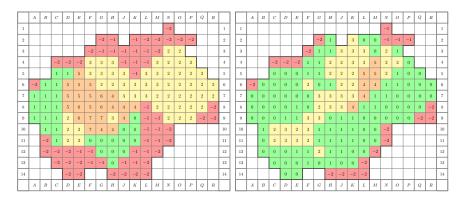
status quo

max double coverage

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Model results

Difference between resulting and required degree of coverage (period 8-12 a.m.)



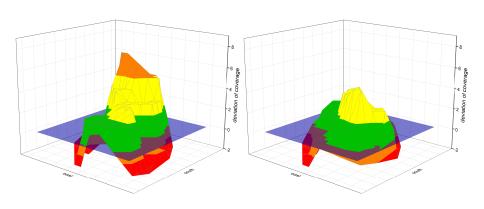
status quo

max empirically required coverage

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Coverage level

Difference to empirically required coverage (period 8-12 a.m.)



current allocation

optimal allocation

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Integration in a Decision Support Tool

Main Menu



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Integration in a Decision Support Tool

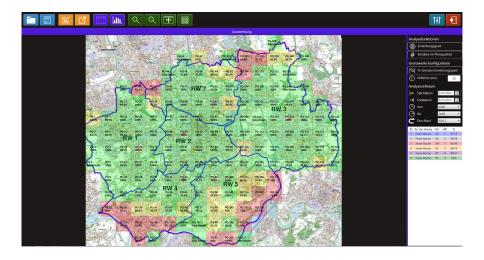
Input Parameters



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Integration in a Decision Support Tool

Extensive analyses – real-world performance measures



vation Basic models Optimization model A real world problem **Outlook**

Conclusions and outlook

Improvements & further research

- Quality improvements:
 - Consideration of flexible ambulance stations is recommended
 - More suitable coverage according to empirical demands
- Efficiency of resource utilization:
 - ► Reduced number of ambulances
 - Constant quality with cost reduction
 - Higher quality with equal costs
- Integration of uncertainties (demand, driving speed)
 - Robust Uncertain Set Covering Problem (considers uncertainty in driving speed)
- Evaluation of the solution with a discrete event simulation

Outlook

Contact

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Data Driven Ambulance Optimization







Literature — Selection

- Brotcorne, Luce; Laporte, Gilbert; Semet, Frédéric (2003): Ambulance location and relocation models, in: European Journal of Operational Research, Vol. 147, Nr. 3, S. 451-463.
- Degel, Dirk; Wiesche, Lara; Rachuba, Sebastian; Werners, Brigitte (2014): Time-dependent ambulance allocation considering data-driven empirically required coverage, in: Health Care Management Science (online first).
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- Kritzinger, Stefanie et al. (2011): Variable neighborhood search for the time-dependent vehicle routing problem with soft time windows, in: Coello Coello, C.(Hrsg.): Learning and Intelligent Optimization. 5th International Conference, LION 5, Berlin 2011, S. 61-75.
- Schmid, V.; Doerner, K. (2010): Ambulance location and relocation problems with time-dependent travel times, in: European Journal of Operational Research, Vol. 207, No. 3, 2010, S. 1293-1303.
- Statistisches Bundesamt (2013): Bevoelkerungsvorausberechnung.
- Zarandi, Fazel M.H.; Davan, S.; Sisakht, Haddad (2011): The large scale maximal covering location problem, in: Scientia Iranica, Vol. 18, No. 6, 2011, S. 1564-1570.

Objectives

maximize the demand which get the required coverage

$$\max \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} d_{it} x_{it}^{e(it)}$$

avoid relocations

$$\min \quad \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{J}} \sum_{j \in \mathcal{J}} u_{ijt}$$

avoid the use of flexible EMS stations

$$\min \quad \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{F}} y_{jt}$$

location/covering constraints

$$\sum_{j \in \mathcal{N}_{it}^{2}} y_{jt} \geq 1 \qquad \forall i \in \mathcal{I}, \forall t \in \mathcal{T}$$

$$\sum_{j \in \mathcal{N}_{it}^{1}} y_{jt} \geq \sum_{k=1}^{p} x_{it}^{k} \qquad \forall i \in \mathcal{I}, \forall t \in \mathcal{T}$$

$$x_{it}^{(k-1)} \geq x_{it}^{k} \qquad \forall i \in \mathcal{I}, \forall t \in \mathcal{T}, k = 2, \dots, p$$

$$\sum_{j \in \mathcal{J}} y_{jt} \leq p_{t} \qquad \forall t \in \mathcal{T}$$

relocation constraints

$$\begin{aligned} y_{Dt} &= p - p_t & \forall t \in \mathcal{T} \\ y_{jt} + \sum_{i \in \mathcal{J} \cup \{D\}} u_{ij(t+1)} - \sum_{i \in \mathcal{J} \cup \{D\}} u_{ji(t+1)} &= y_{j(t+1)} & \forall j \in \mathcal{J} \cup \{D\}, \ \forall t \in \mathcal{T} \setminus \{T\} \\ y_{jT} + \sum_{i \in \mathcal{J} \cup \{D\}} u_{ij1} - \sum_{i \in \mathcal{J} \cup \{D\}} u_{ji1} &= y_{j1} & \forall j \in \mathcal{J} \cup \{D\} \\ \sum_{i \in \mathcal{J}} u_{iDt} \leq |p_{(t-1)} - p_t| & \forall t \in \mathcal{T} \setminus \{1\} \\ \sum_{i \in \mathcal{J}} u_{Djt} \leq |p_{(t-1)} - p_t| & \forall t \in \mathcal{T} \setminus \{1\} \\ \sum_{i \in \mathcal{J}} u_{iD1} \leq |p_{T} - p_{1}| & \sum_{i \in \mathcal{J}} u_{Dj1} \leq |p_{T} - p_{1}| \end{aligned}$$

demand constraints

$$\sum_{i \in \mathcal{I}} d_{it} x_{it}^{1} \ge \alpha \sum_{i \in \mathcal{I}} d_{it} \qquad \forall t \in \mathcal{T}$$

$$\sum_{i \in \mathcal{I}} d_{it} \le c \sum_{i \in \mathcal{I}} y_{jt} \qquad \forall t \in \mathcal{T}$$

site capacity constraints

$$y_{jt} \le \rho_j$$
 $\forall j \in \mathcal{J}, \forall t \in \mathcal{T}$
 $y_{jT} = 0$ $\forall j \in \mathcal{F}$

variable domain

$$x_{it}^{k} \in \{0, 1\}$$

$$\forall i \in \mathcal{I}, k = 1, \dots, p, \forall t \in \mathcal{T}$$

$$y_{jt} \in \{0, 1, \dots, p\} \subset \mathbb{N}_{0}$$

$$\forall j \in \mathcal{J}, \forall t \in \mathcal{T}$$

$$u_{ijt} \in \{0, 1, \dots, p\} \subset \mathbb{N}_{0}$$

$$\forall i \in \mathcal{I}, \forall j \in \mathcal{J}, \forall t \in \mathcal{T}$$