



Federal Ministry
of Economics
and Technology

Models for Railway Track Allocation

Thomas Schlechte

Joint work with

Ralf Borndörfer

Martin Grötschel

16.11.2007

ATMOS 2007 Sevilla



Thomas Schlechte ▪ Konrad-Zuse-Zentrum für Informationstechnik Berlin (ZIB)

schlechte@zib.de

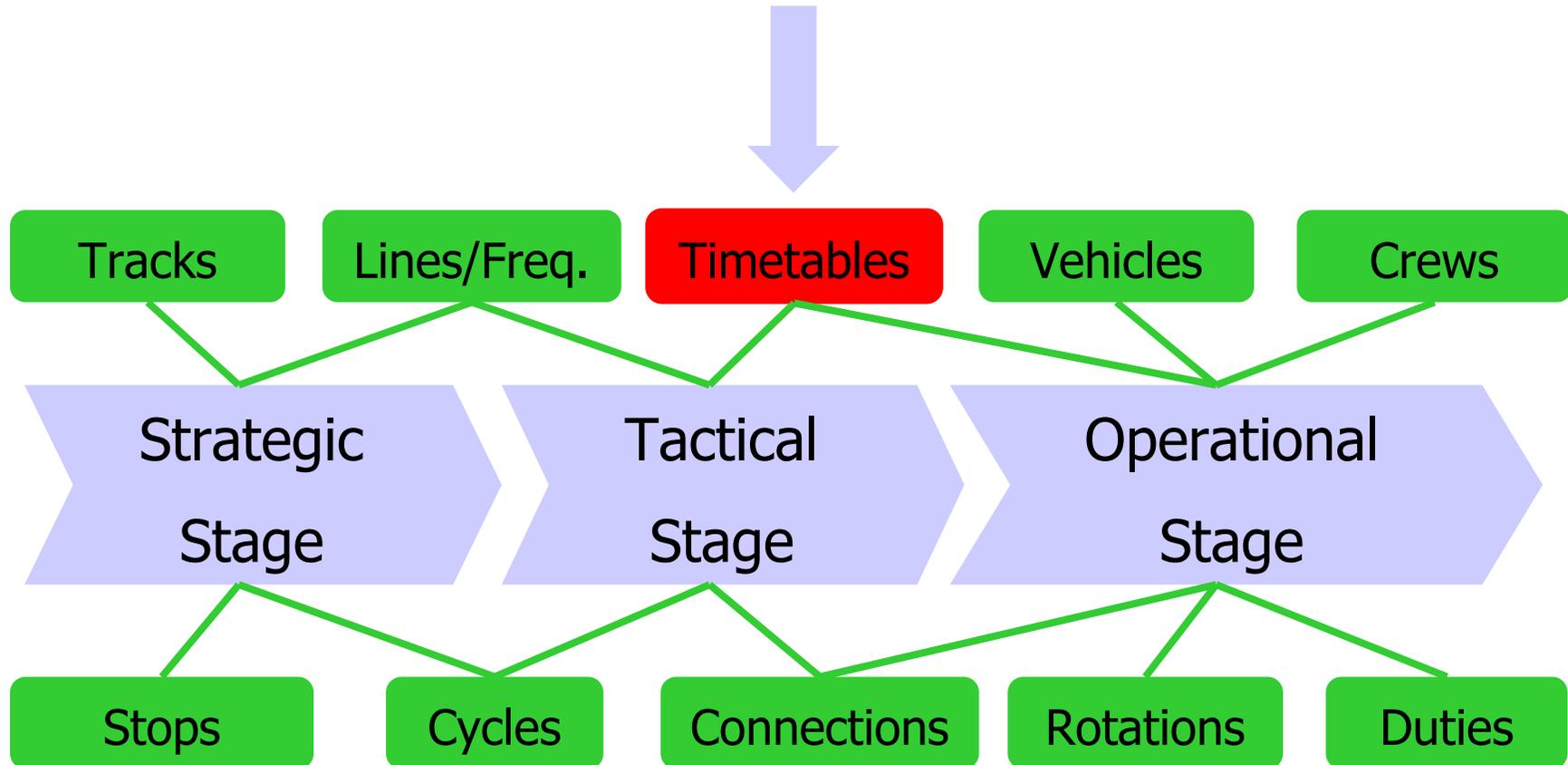
<http://www.zib.de/schlechte>

Overview

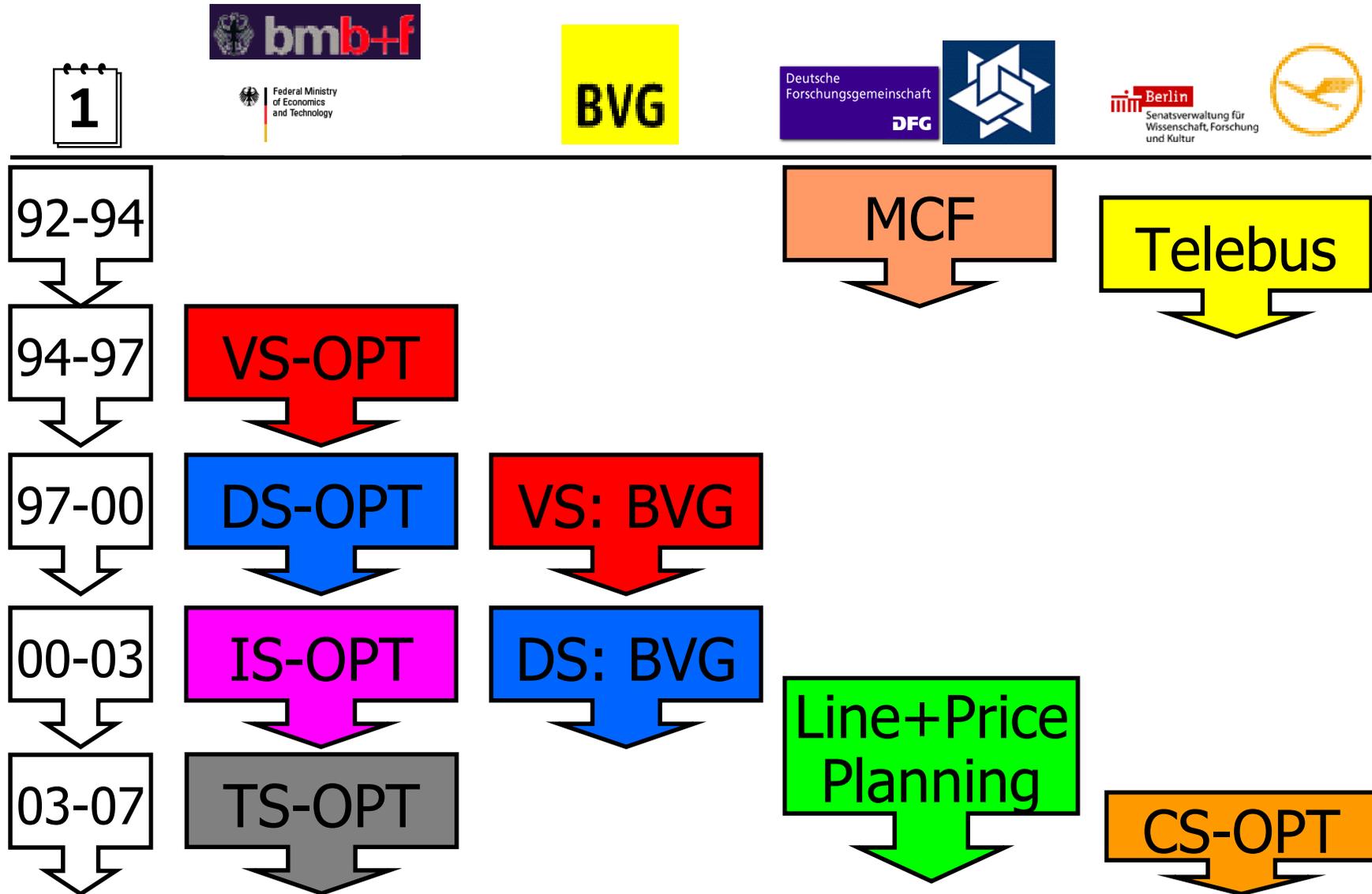
1. Problem Introduction
2. Model Discussion
3. Column Generation Approach
4. Computational Results



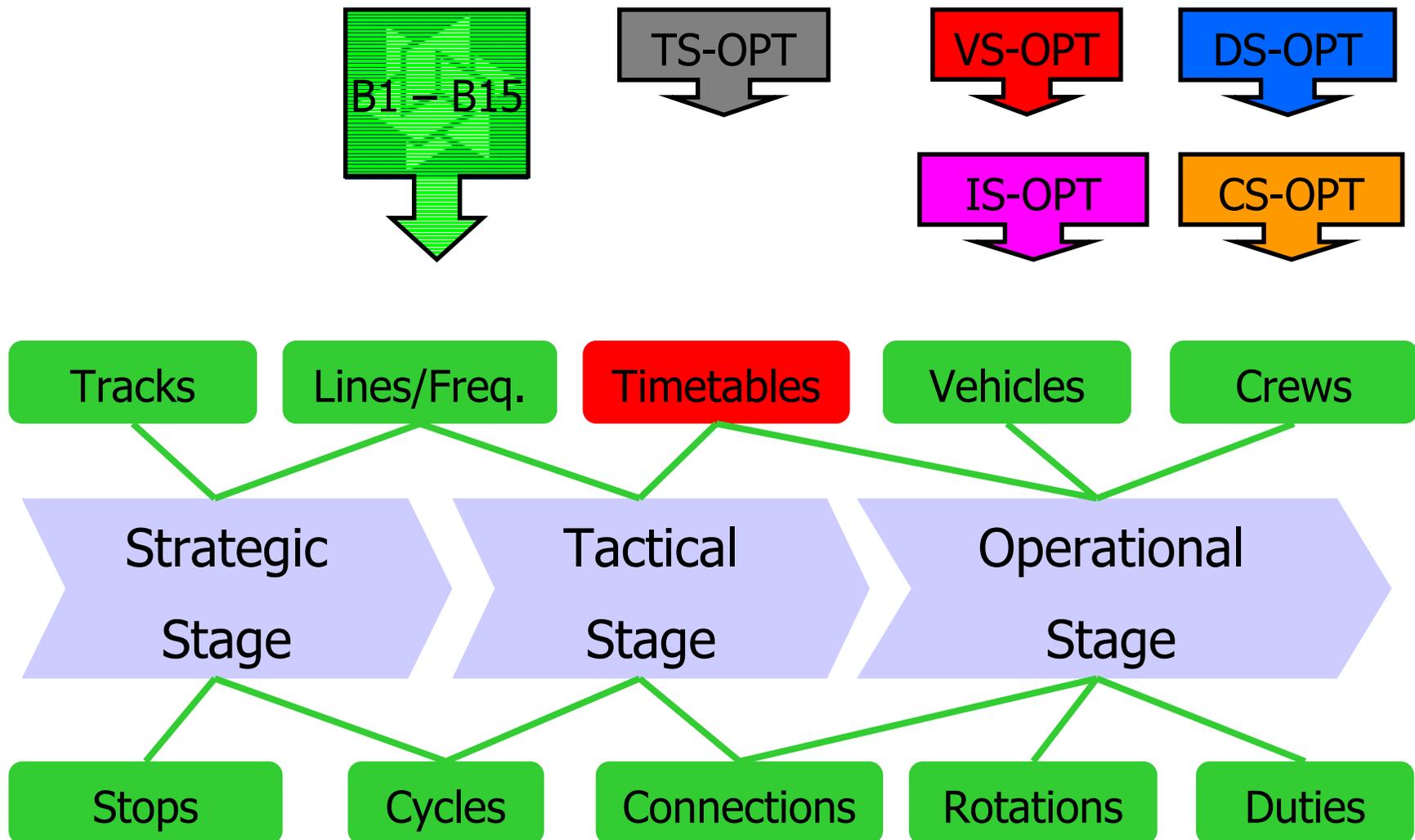
Planning in Public Transport



Traffic Projects @ ZIB



Planning in Public Transport



Overview

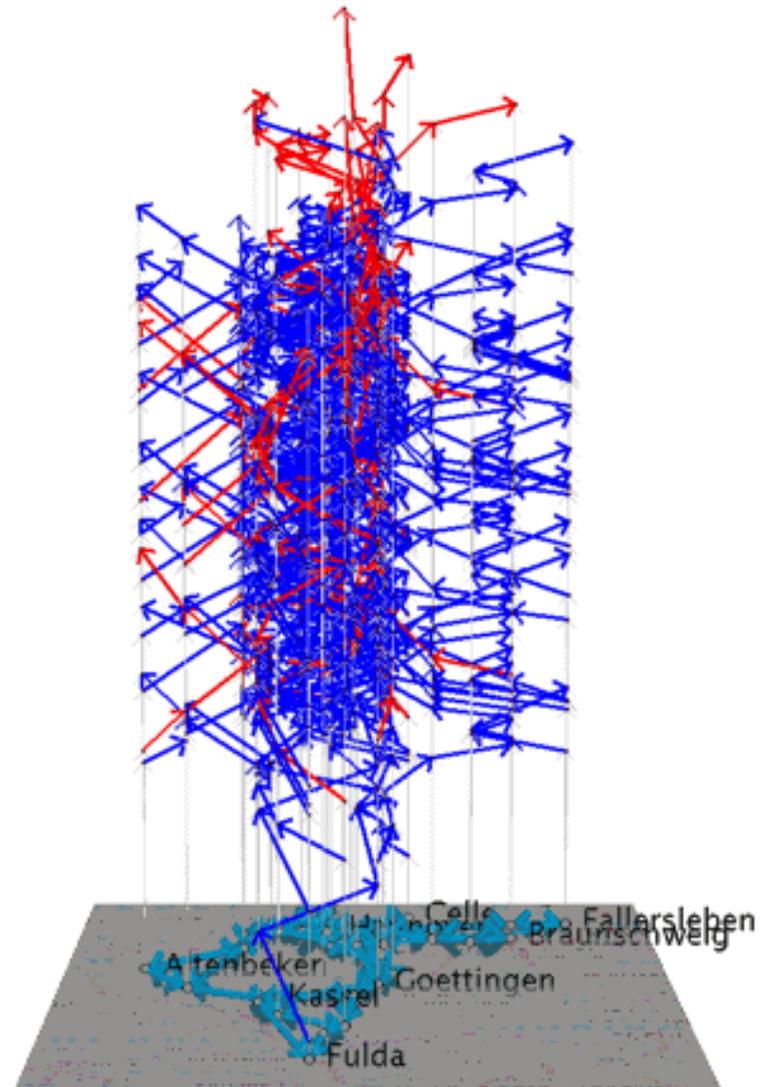
1. Problem Introduction
2. Model Discussion
3. Column Generation Approach
4. Computational Results



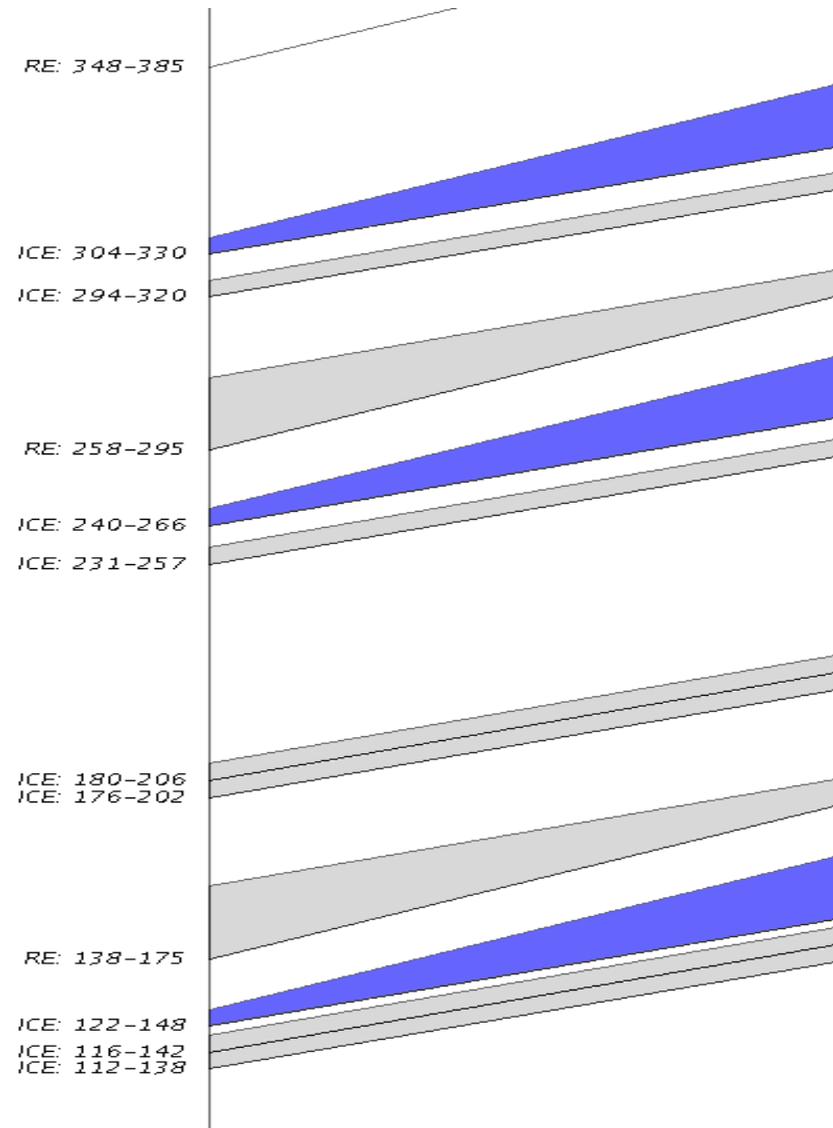
The Problem (TraVis by M.Kinder)



Schedule in 3d



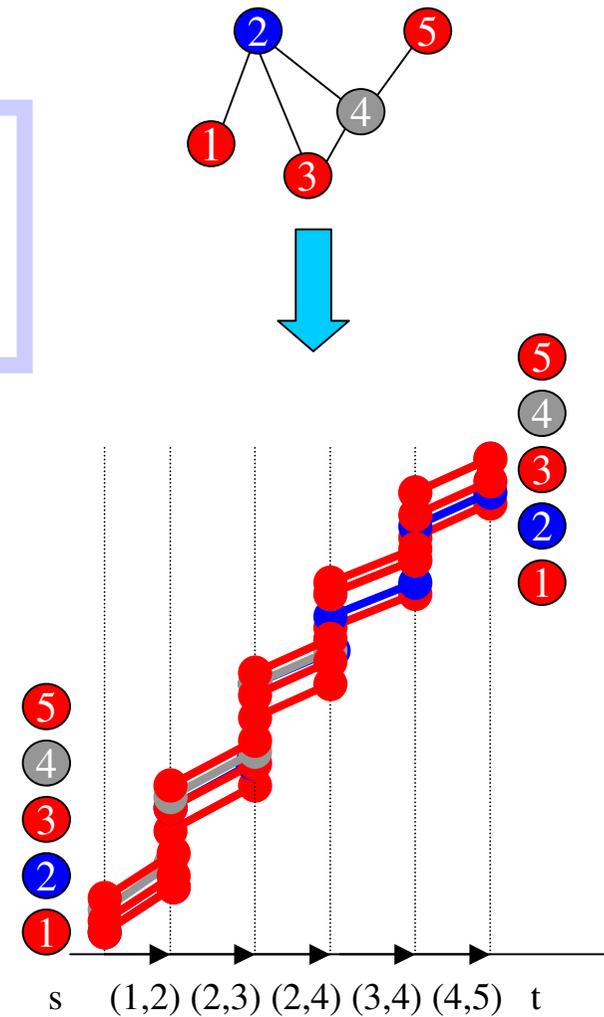
Conflict-Free-Allocation



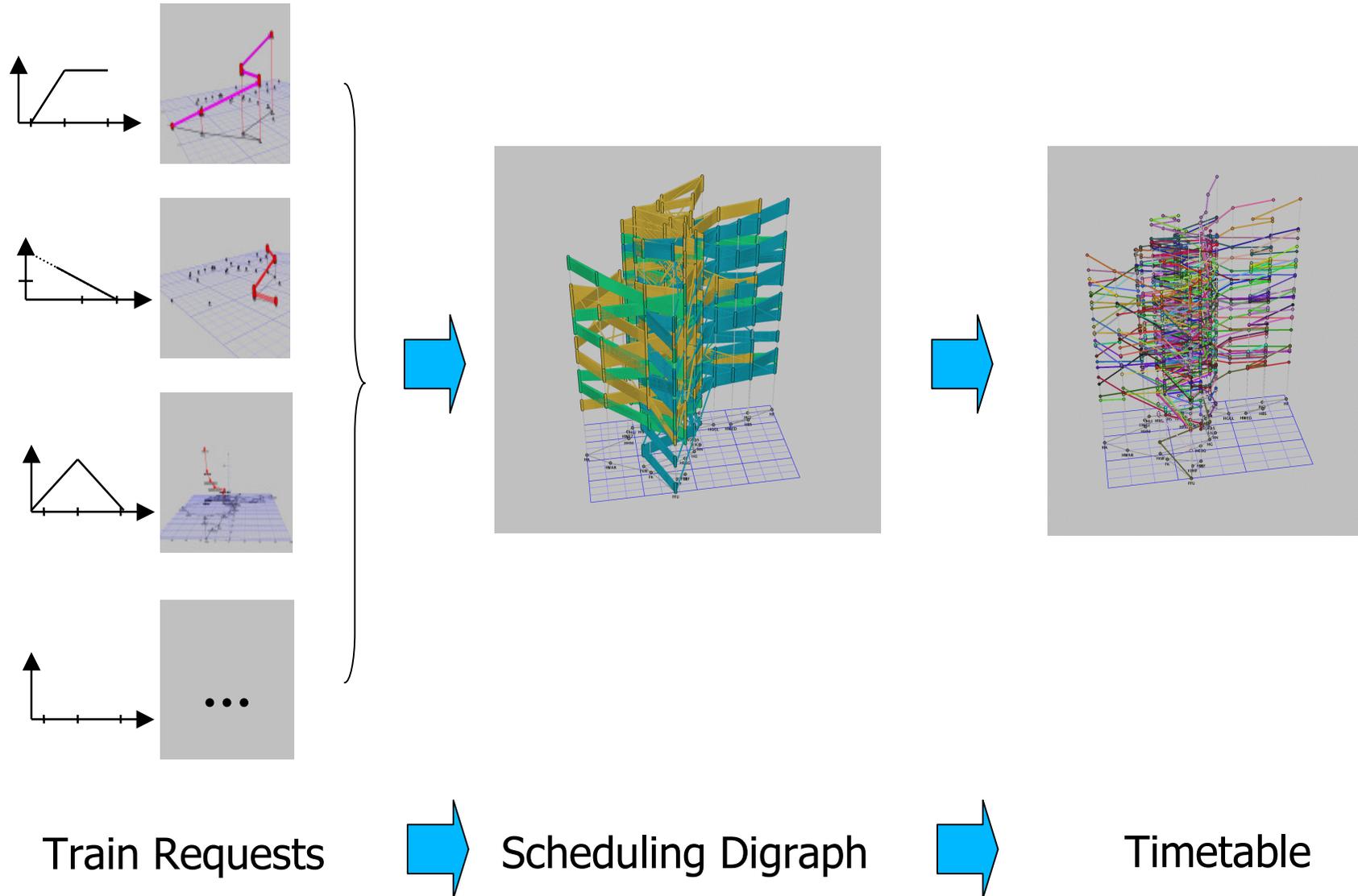
Complexity

Proposition [Caprara, Fischetti, Toth (02)]:
OPTRA/TTP is *NP*-hard.

Proof:
Reduction from Independent-Set.



Track Allocation Problem

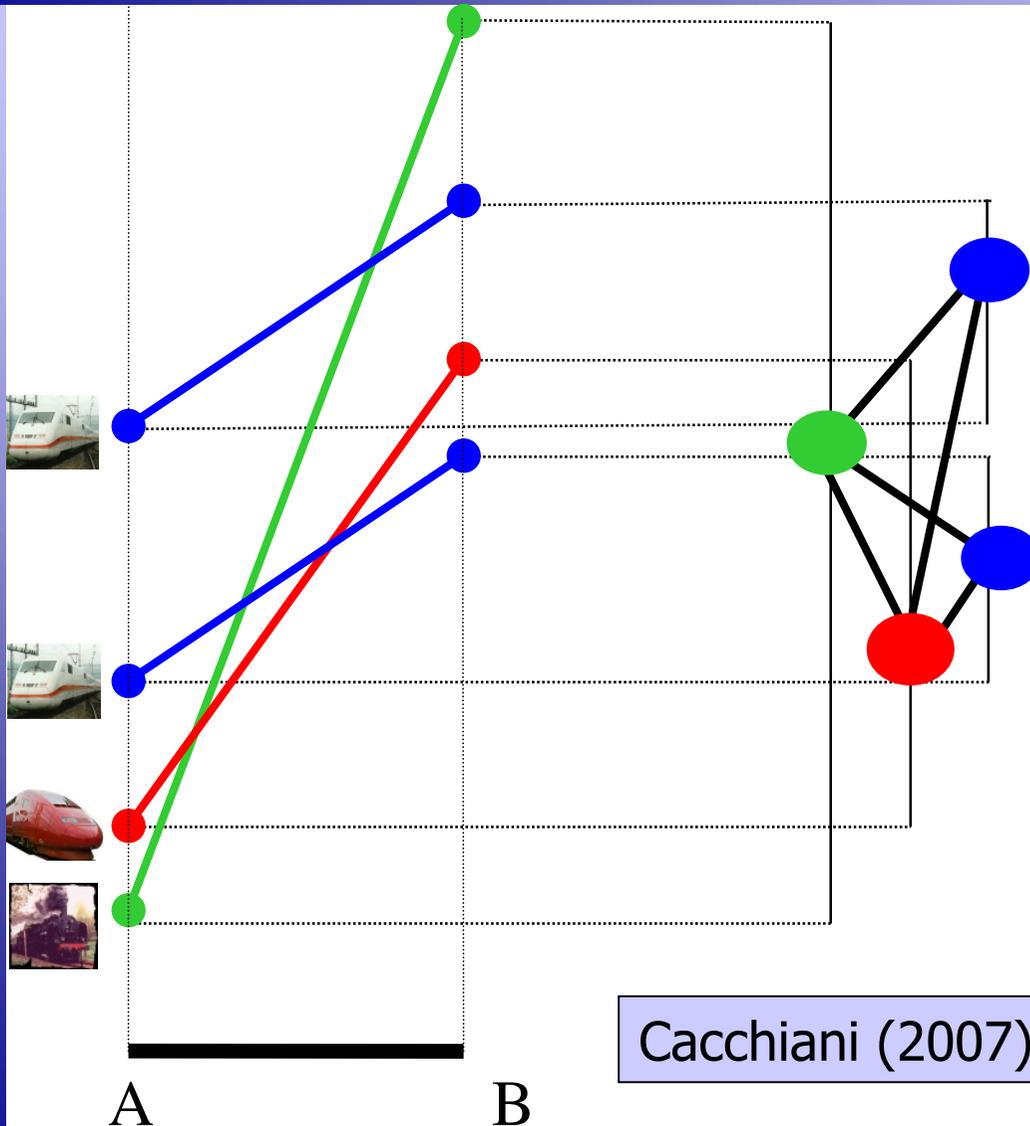


Overview

1. Problem Introduction
2. Model Discussion
3. Column Generation Approach
4. Computational Results



Packing Models



- **Conflict graph**
- Cliques
- Perfect

Cacchiani (2007) – Path Compatibility Graphs

Arc Packing Problem

(APP)

max

$$\sum_{i \in \mathcal{I}} \sum_{a \in A} p_a^i x_a^i$$

s.t.

$$\sum_{a \in \delta_i^{out}(v)} x_a^i - \sum_{a \in \delta_i^{in}(v)} x_a^i \leq \delta_i(v) \quad \forall v \in V, \forall i \in \mathcal{I} \quad (i)$$

$$\sum_{i \in \mathcal{I}} \sum_{a \in A} x_a^i \leq 1 \quad \forall c \in C \quad (ii)$$

$$x_a^i \in \{0, 1\} \quad \forall a \in A, \forall i \in \mathcal{I} \quad (iii)$$

Variables

- Arc occupancy (request i uses arc a)

Constraints

- Flow conservation and
- Arc conflicts (pairwise)

Objective

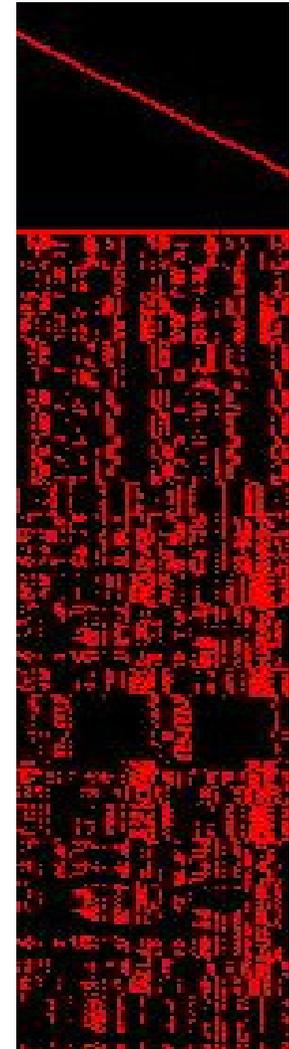
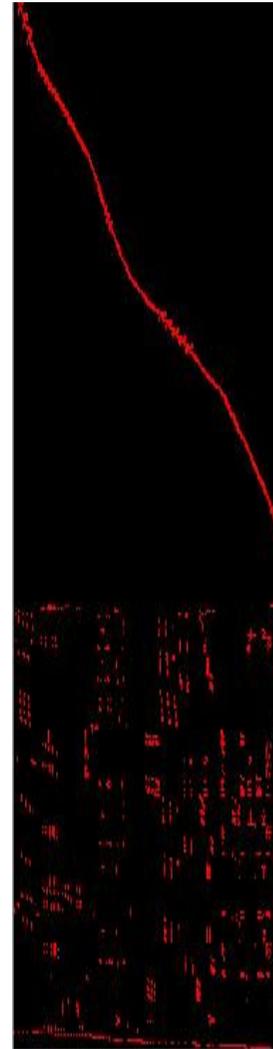
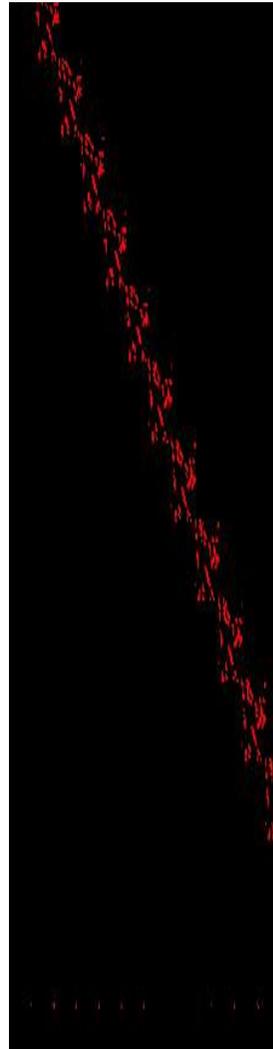
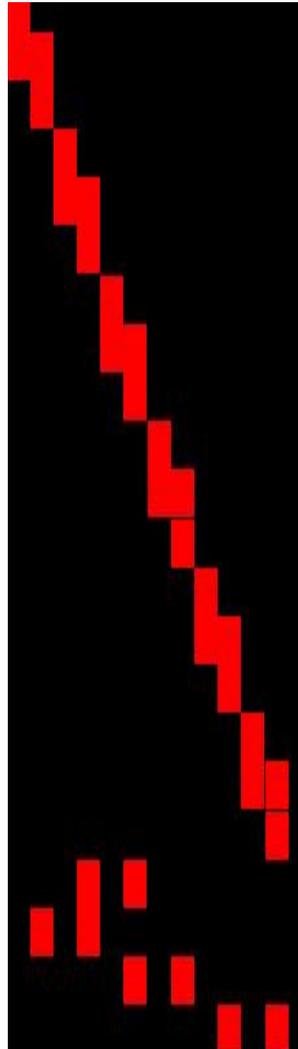
- Maximize proceedings

(PPP) transformation from arc to path variables (see Cachhiani (2007))

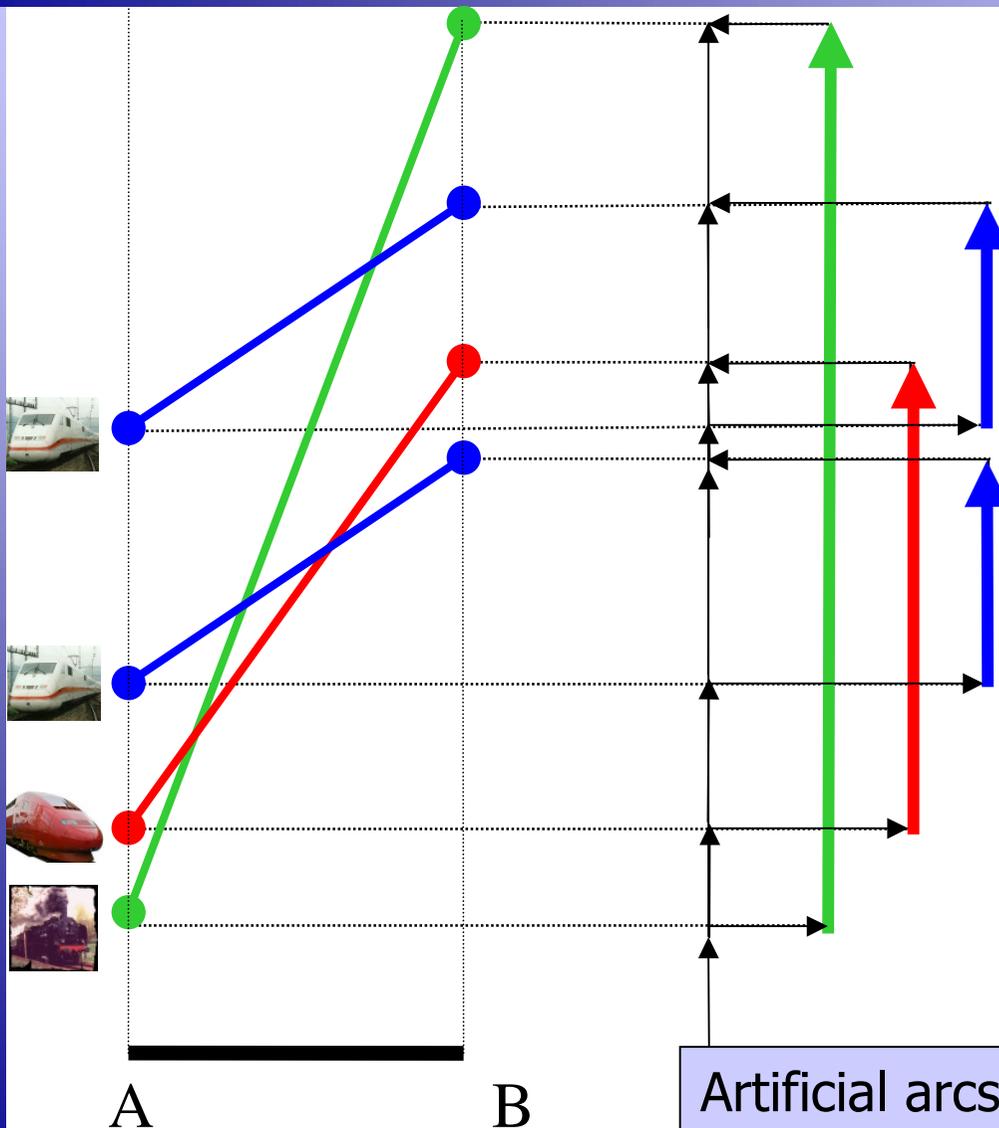


Packing Models

- **Proposition:**
The LP-relaxation of APP can be solved in polynomial time.
- ... and in practice.



Novel Model



- **Track Digraph**
- Timeline(s)
- Config paths

Artificial arcs represent valid successors !

Path Coupling Problem

(PCP)

max

$$\sum_{p \in \mathcal{P}} \sum_{a \in p} p_a^i x_p$$

s.t.

$$\sum_{p \in \mathcal{P}_i} x_p \leq 1 \quad \forall i \in I \quad (\text{i})$$

$$\sum_{q \in \mathcal{Q}_j} y_q \leq 1 \quad \forall j \in J \quad (\text{ii})$$

$$\sum_{a \in p \in \mathcal{P}} x_p - \sum_{a \in q \in \mathcal{Q}} y_q \leq 0 \quad \forall a \in A_I \cap A_J \quad (\text{iii})$$

$$y_q \in \{0, 1\} \quad \forall q \in \mathcal{Q}_j, \forall j \in J \quad (\text{iv})$$

$$x_p \in \{0, 1\} \quad \forall p \in \mathcal{P}_i, \forall i \in I \quad (\text{v})$$

Variables

- Path und config usage (request i uses path p , track j uses config q)

Constraints

- Path and config choice
- Path-config-coupling (track capacity)

Objective Function

- Maximize proceedings

(ACP) transformation from path to arc variables (see Borndörfer, S. (2007))



Linear Relaxation of PCP

(MLP)

max

$$\sum_{p \in \mathcal{P}} \sum_{a \in \mathcal{P}} p_a^i x_p$$

s.t.

$$\sum_{p \in \mathcal{P}_i} x_p \leq 1 \quad \forall i \in I \quad (\text{i})$$

$$\sum_{q \in \mathcal{Q}_j} y_q \leq 1 \quad \forall j \in J \quad (\text{ii})$$

$$\sum_{a \in \mathcal{P} \in \mathcal{P}} x_p - \sum_{a \in \mathcal{Q} \in \mathcal{Q}} y_q \leq 0 \quad \forall a \in A_I \cup A_J \quad (\text{iii})$$

$$0 \leq y_q \leq 1 \quad \forall q \in \mathcal{Q} \quad (\text{iii})$$

$$0 \leq x_p \leq 1 \quad \forall p \in \mathcal{P} \quad (\text{iv})$$

γ_i
 π_j
 λ_a

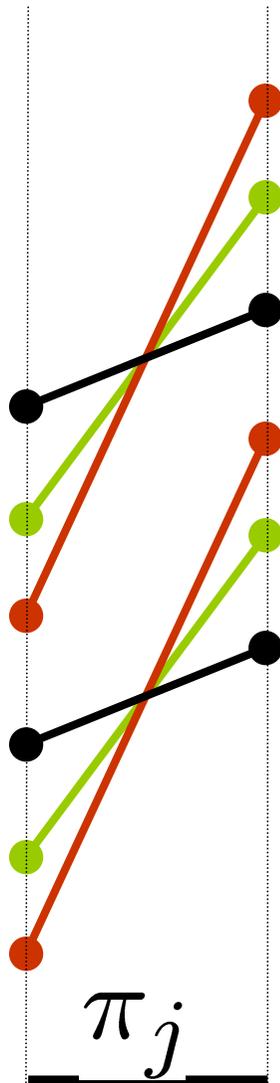
dual variable	information about	useful to
γ_i	bundle price	analyse request
π_j	track price	analyse network
λ_a	arc price	-

Dualization

$$\begin{array}{ll}
 (DLP) & \\
 \min & \sum_{j \in J} \pi_j + \sum_{i \in I} \gamma_i \\
 \text{s.t.} & \gamma_i + \sum_{a \in p} \lambda_a \geq \sum_{a \in p} p_a^i \quad \forall p \in \mathcal{P}_i, \forall i \in I \quad (\text{i}) \\
 & \pi_j - \sum_{a \in q} \lambda_a \geq 0 \quad \forall q \in \mathcal{Q}_j, \forall j \in J \quad (\text{ii}) \\
 & \gamma_i \geq 0 \quad \forall i \in I \quad (\text{iii}) \\
 & \lambda_a \geq 0 \quad \forall a \in A_I \cup A_J \quad (\text{iv}) \\
 & \pi_j \geq 0 \quad \forall j \in J \quad (\text{v})
 \end{array}$$



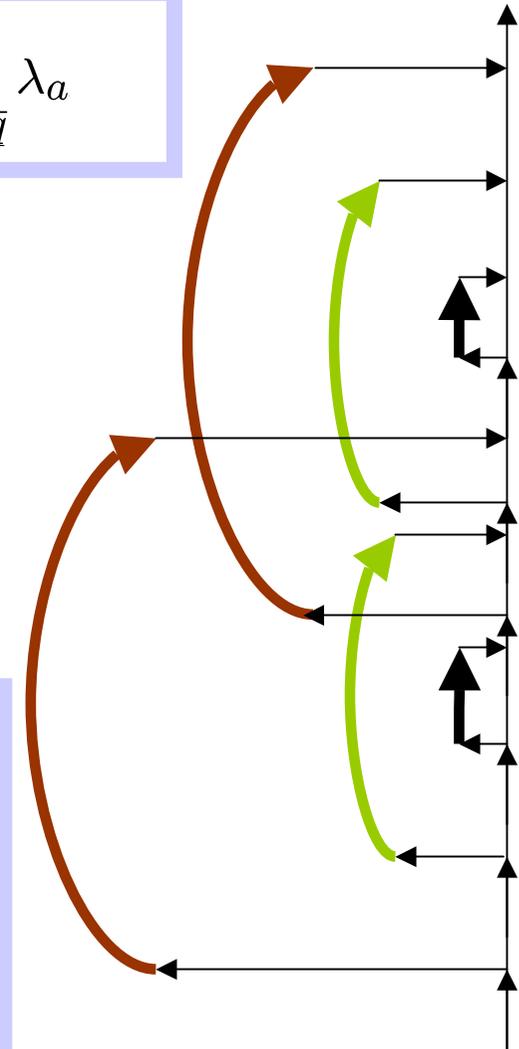
Pricing of y -variables



$$(\text{PRICE } (y)) \quad \exists \bar{q} \in Q_j : \pi_j < \sum_{a \in \bar{q}} \lambda_a$$

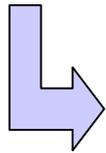
$$c_a = -\lambda_a$$

Pricing Problem(y) :
Acyclic shortest path problem
for each track j with modified
cost function c !

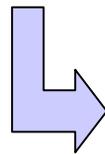


Observation

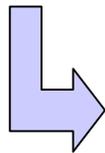
$$(\text{PRICE } (x)) \quad \exists \bar{p} \in \mathcal{P}_i : \gamma_i < \sum_{a \in \bar{p}} (p_a - \lambda_a)$$



$$\eta_i := \max_{p \in \mathcal{P}_i} \sum_{a \in p} (p_a - \lambda_a) - \gamma_i, \quad \forall i \in I$$



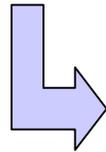
$$\eta_i + \gamma_i \geq \sum_{a \in p} (p_a - \lambda_a) \quad \forall i \in I, p \in \mathcal{P}_i$$



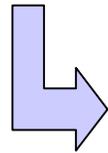
$$\eta_i + \gamma_i \text{ satisfies } (DLP)(i)$$

And analogously ...

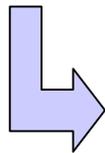
$$\text{(PRICE (y)) } \exists \bar{q} \in Q_j : \pi_j < \sum_{a \in \bar{q}} \lambda_a$$



$$\theta_j := \max_{\bar{q} \in Q_j} \sum_{a \in \bar{q}} \lambda_a - \pi_j, \forall j \in J$$



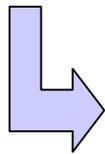
$$\theta_j + \pi_j \geq \sum_{a \in q} \lambda_a \quad \forall j \in J, q \in Q_j$$



$\theta_j + \pi_j$ satisfies $(DLP)(ii)$

Pricing Upper Bound

$(\max\{\eta+\gamma, 0\}, \max\{\theta+\pi, 0\}, \lambda)$ is feasible for (DLP)



$$\beta(\gamma, \pi, \lambda) := \sum_{i \in I} \max\{\gamma_i + \eta_i, 0\} + \sum_{j \in J} \max\{\pi_j + \theta_j, 0\}$$

- **Lemma** [ZR-07-02]: Given (infeasible) dual variables of PCP and let $v_{LP}(PCP)$ be the optimum objective value of the LP-Relaxation of PCP, then:

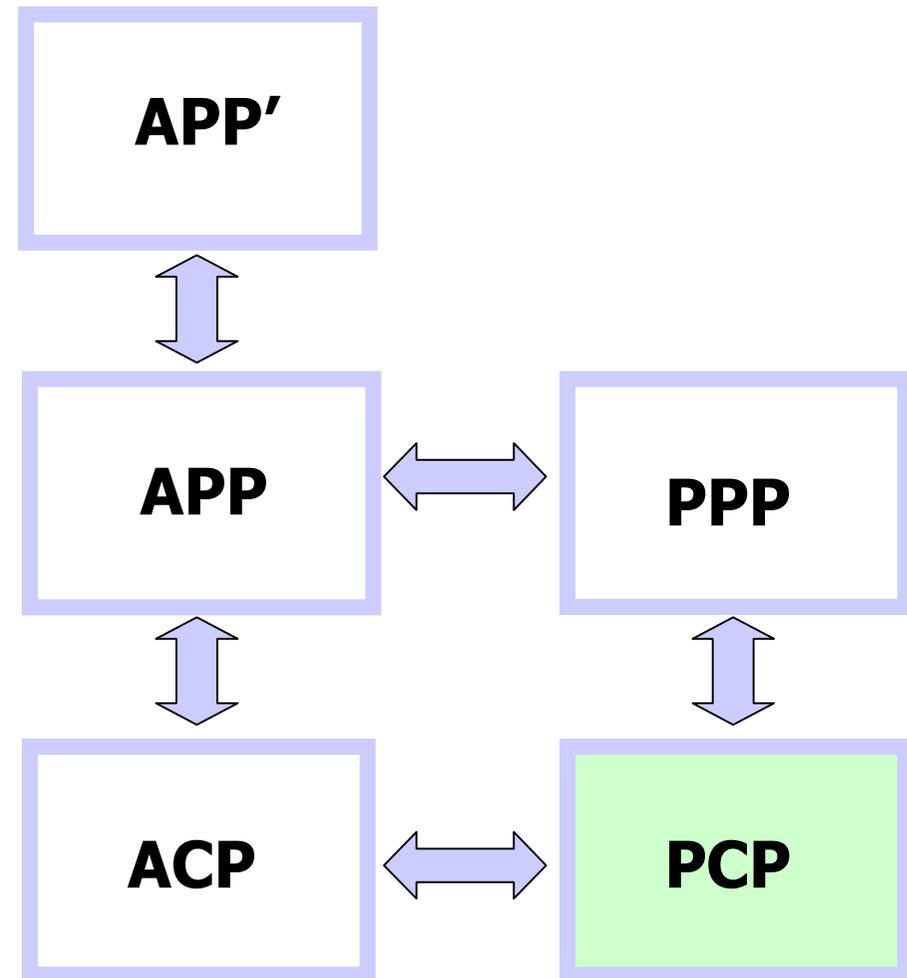
$$v_{LP}(PCP) \leq \beta(\gamma, \pi, \lambda)$$

Model Comparison

- **Theorem** [ZR-07-02]:
The LP-relaxations of ACP and PCP can be solved in polynomial time.

- **Lemma** [ZR-07-02]:

$$\begin{aligned} v_{\text{LP}}(\text{PCP}) &= v_{\text{LP}}(\text{ACP}) \\ &= v_{\text{LP}}(\text{APP}) = v_{\text{LP}}(\text{PPP}) \\ &\leq v_{\text{LP}}(\text{APP}') \end{aligned}$$

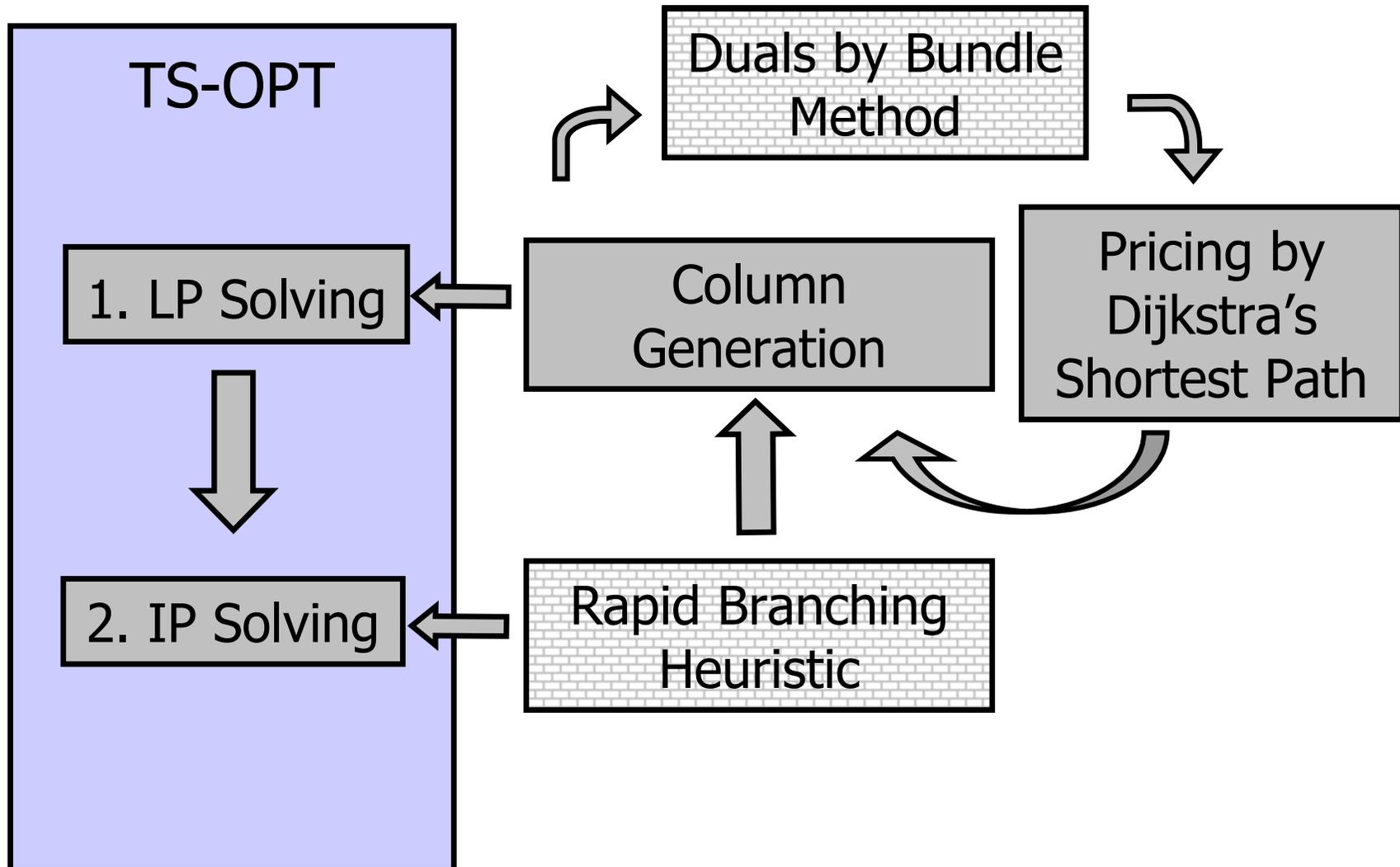


Overview

1. Problem Introduction
2. Model Discussion
3. Column Generation Approach
4. Computational Results



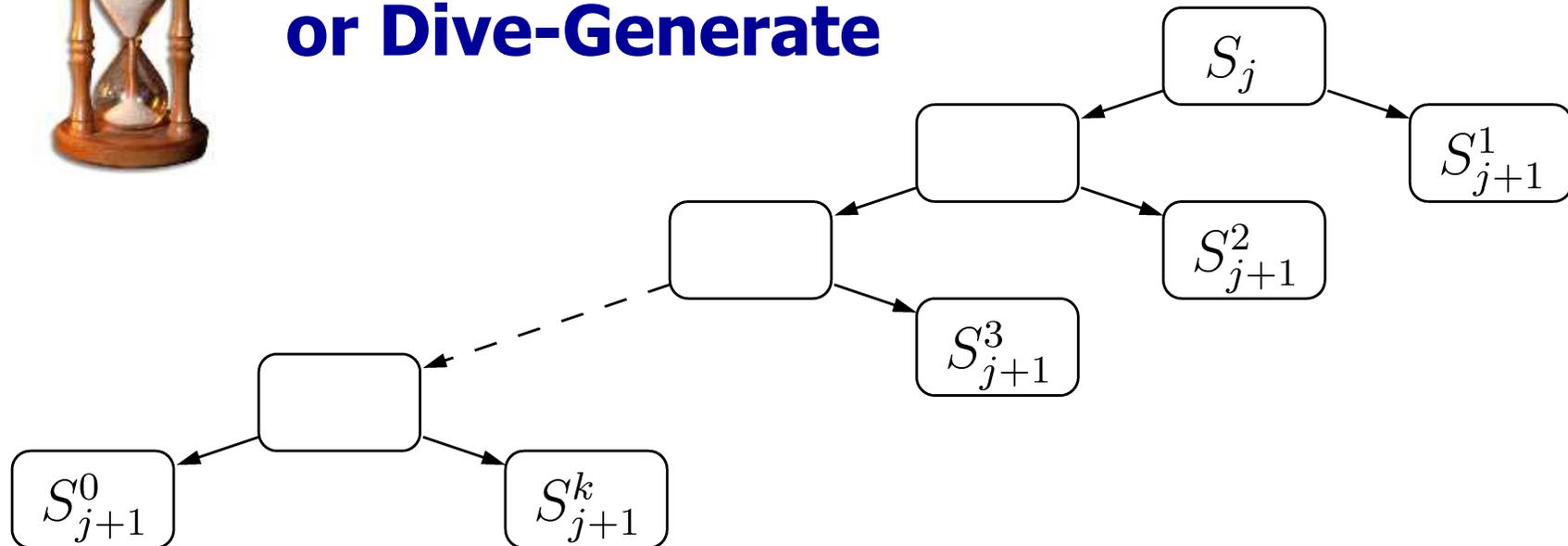
Two Step Approach



Branch-Bound-Price



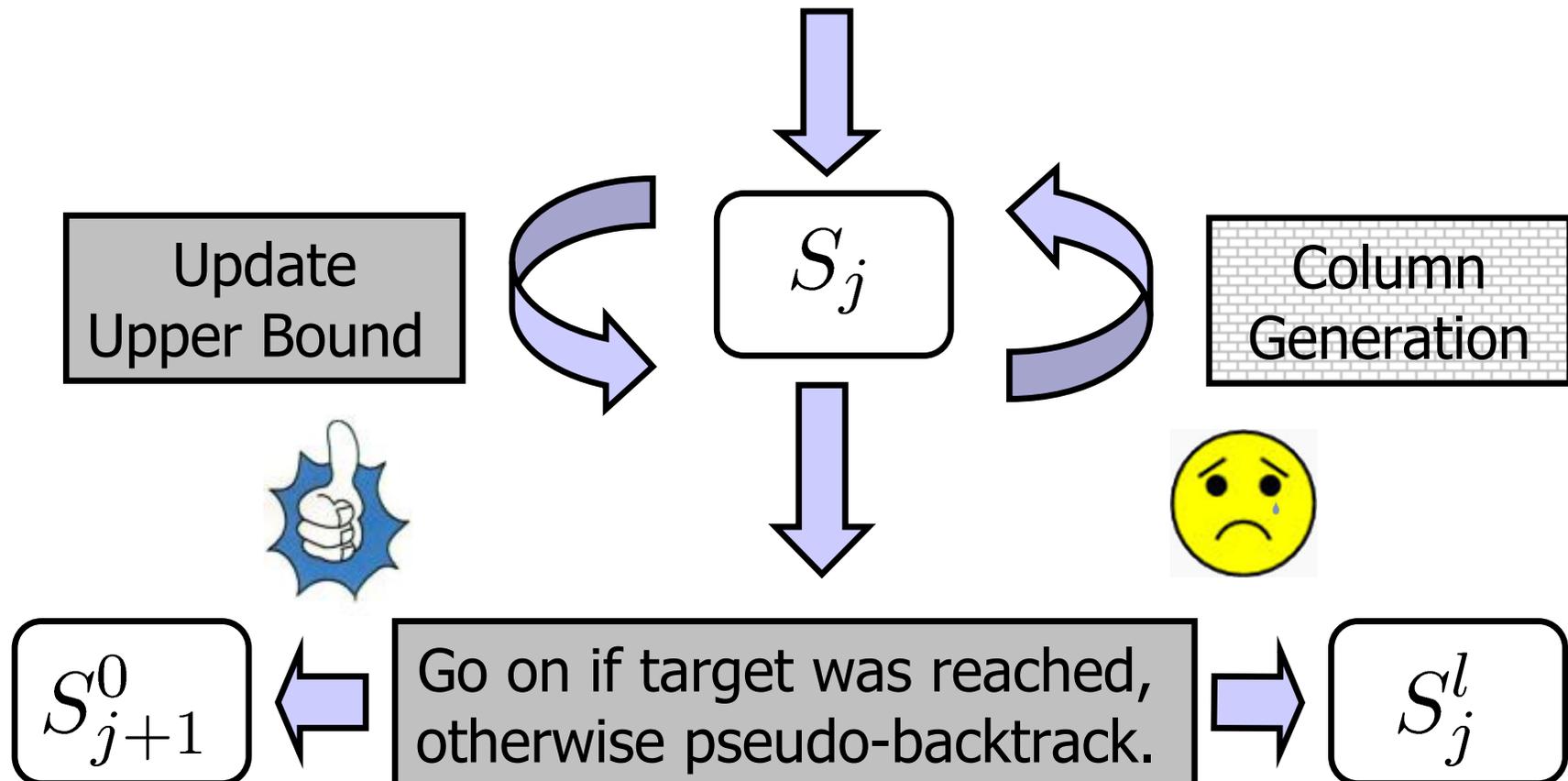
or Dive-Generate



Evaluation of only few highly different sub-problems at iteration j to reach IP-Solutions fast.

Rapid Branching

Node selection of set of fixed to 1 variables by using perturbed cost function (bonus close to 1.0).



Overview

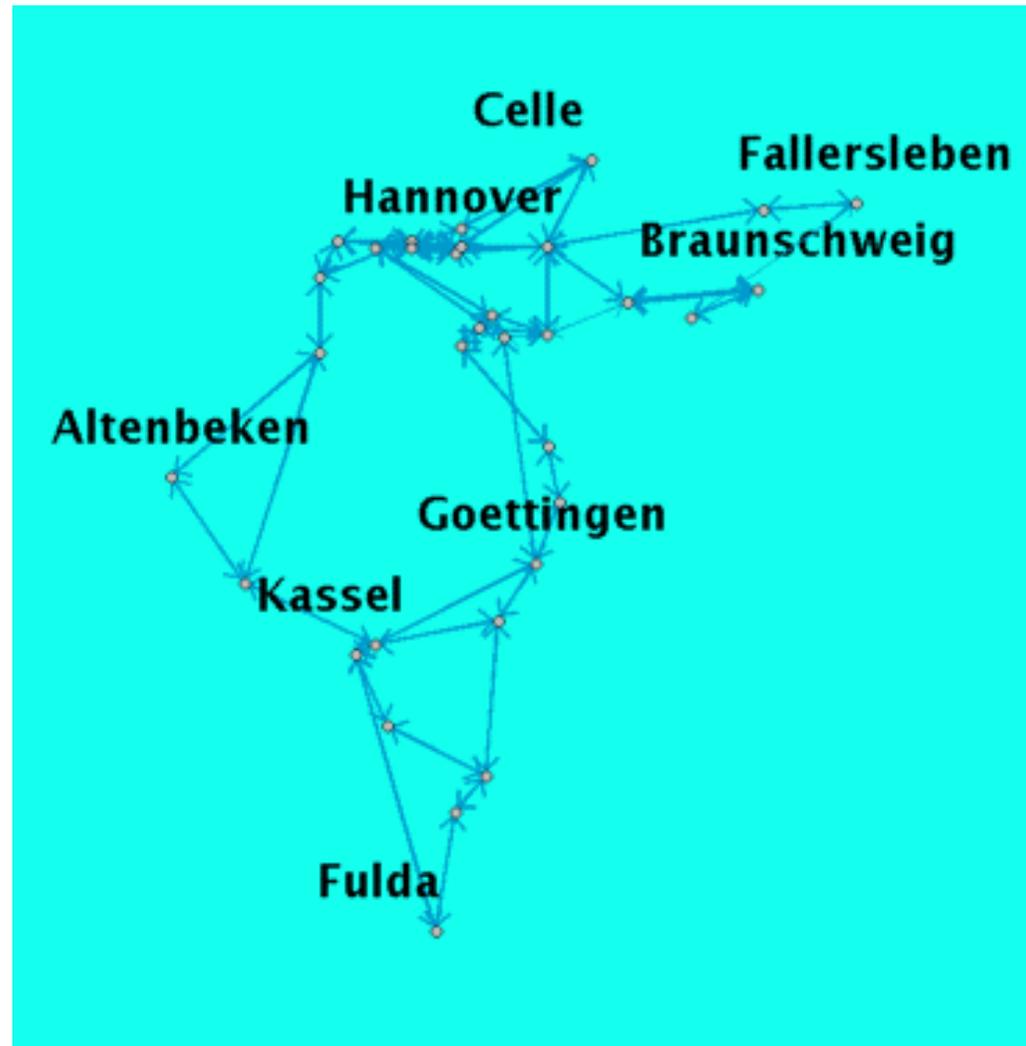
1. Problem Introduction
2. Model Discussion
3. Column Generation Approach
4. Computational Results



Results

• Test Network

- 45 Tracks
- 37 Stations
- 6 Traintypes
- 10 Trainsets
- 146 Nodes
- 1480 Arcs
- 96 Station Capacities
- 4320 Headway Times



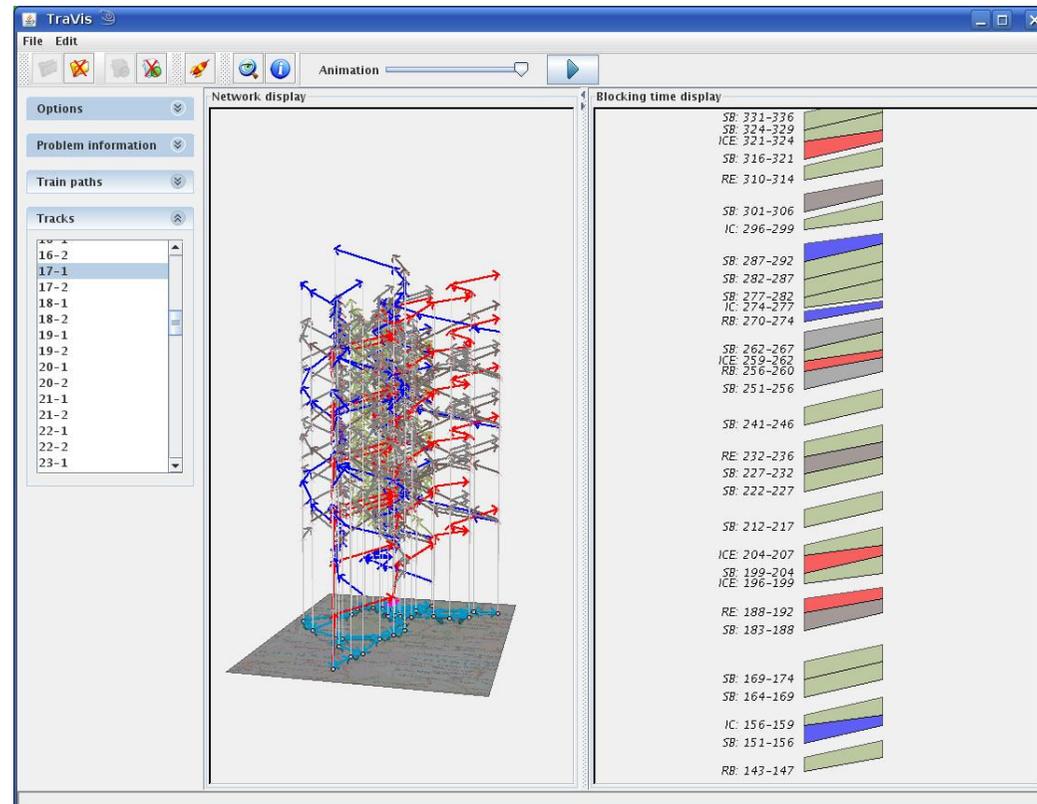
Model Comparison

- **Test Scenarios**

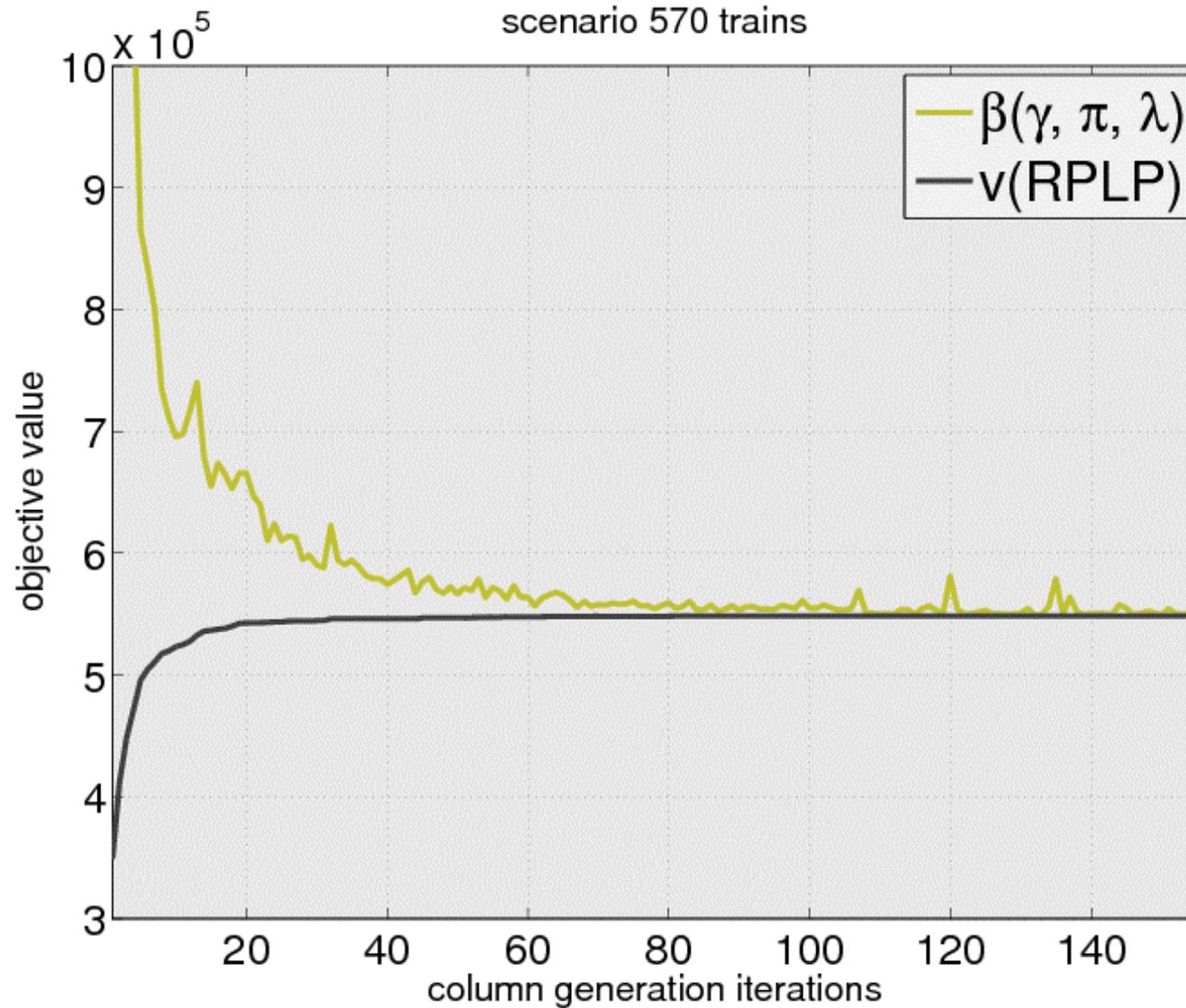
- 146 Train Requests
- 285 Train Requests
- 570 Train Requests

- **Flexibility**

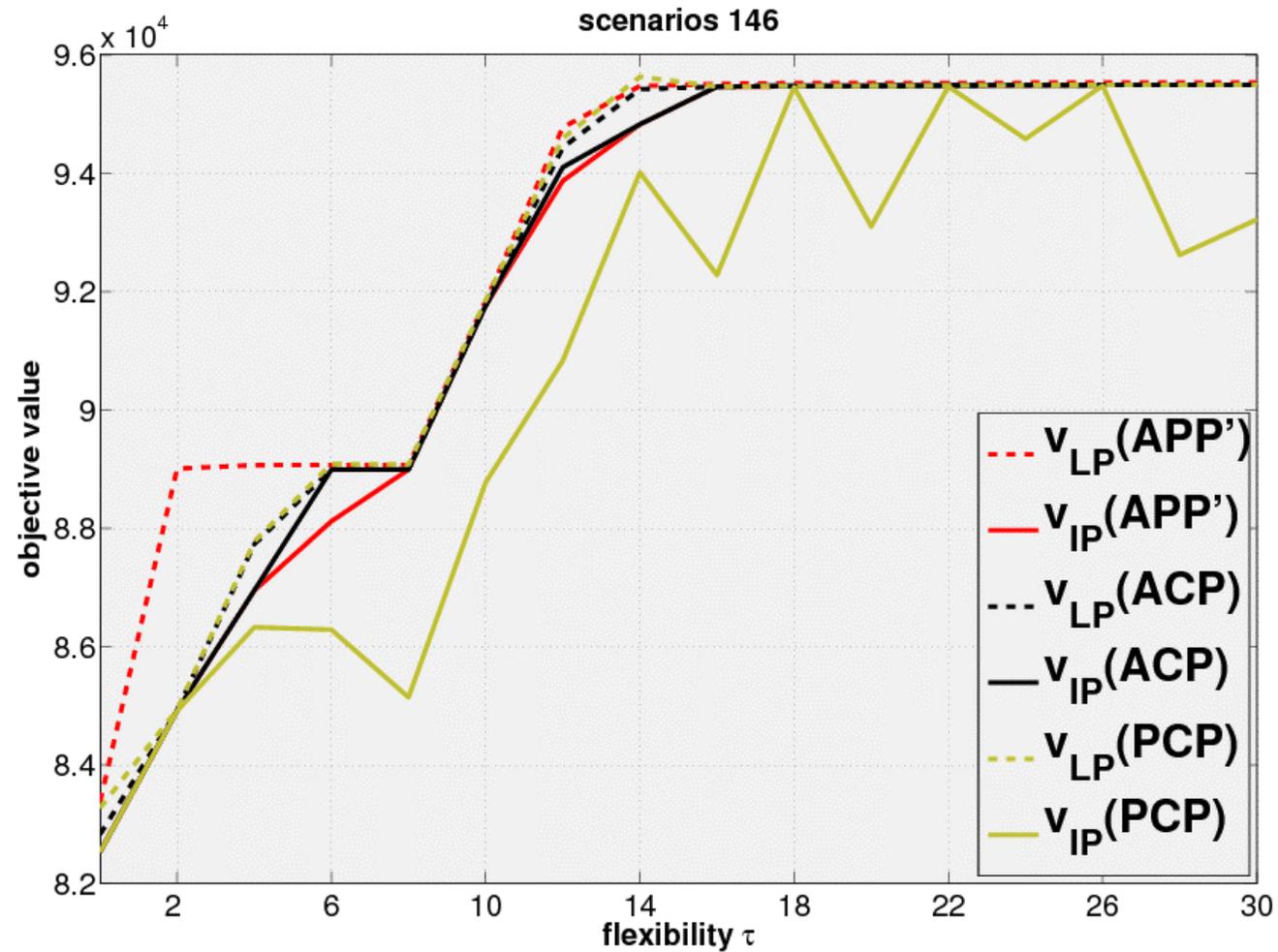
- 0-30 Minutes
- earlier departure penalties
- late arrival penalties
- train type depending profits



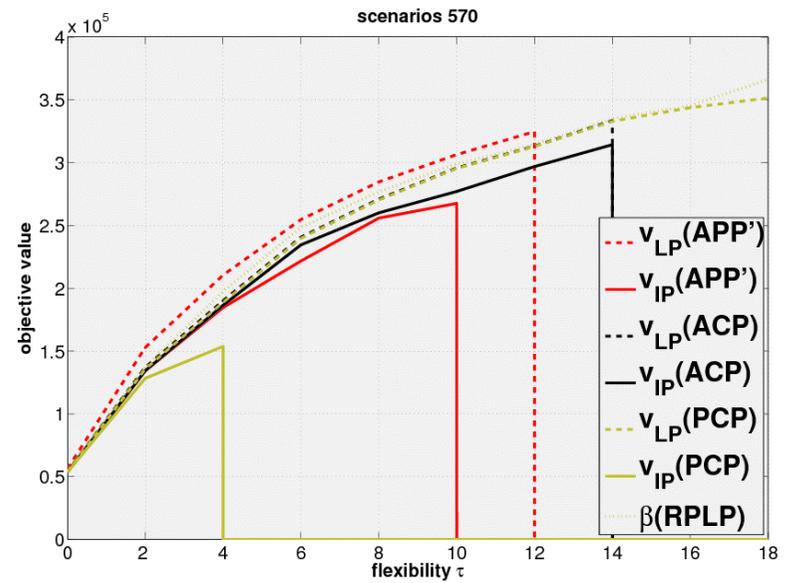
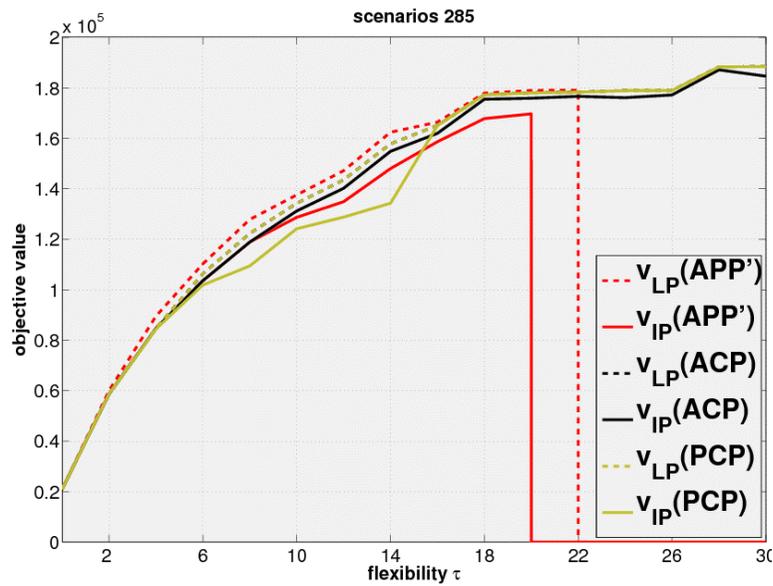
Run of TS-OPT /LP Stage



Model Comparison



Model Comparison



For details see [ZR-07-02, ZR-07-20].



Outlook

Algorithmic Developments

- Bundle method
- Model refinement (connections)
- Adaptive IP Heuristics
- Dynamic Discretization

Simulation of results by





**Thank you
for your attention !**

**Thomas Schlechte
Zuse-Institut Berlin (ZIB)
Takustr. 7, 14195 Berlin
Deutschland**

**Fon (+49 30) 84185-317
Fax (+49 30) 84185-269
schlechte@zib.de
www.zib.de/schlechte**