

# Capacity Estimation Methods

Prof. Dr.-Ing. Ingo A. Hansen

24-6-2013

# Content

- Definition of capacity
- Determinants of capacity
- Capacity balance
- Classification of capacity estimation models
- Macroscopic capacity estimation
- Microscopic infrastructure and rolling stock model
- Analytical capacity estimation
- Waiting times and queuing
- Combinatorial optimisation
- Conclusions

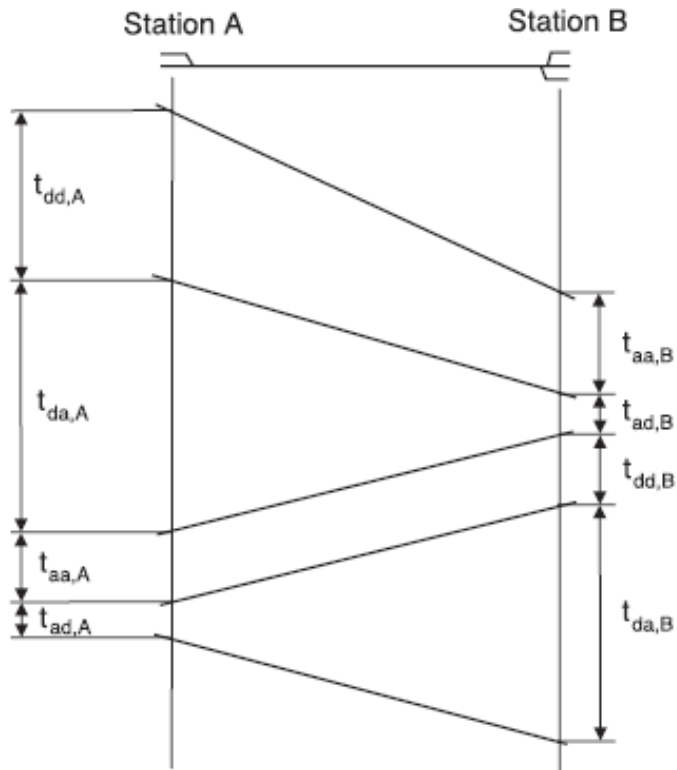
# Definition of Capacity

Maximum number of trains  $N$  that may be operated using a defined part of the infrastructure at the same time during a defined time period  $[1/T]$  ;  $T$ : Time period [24 h = 1440 min; 1 h = 60 min]

- Theoretical capacity  $C = T / \sum(t_{h\ min} + \Delta t)$   
Maximum number of trains  $N_T$  at scheduled order and speed without timetable margins;  $t_{h\ min}$ : minimum headway time  
 $\Delta t$ : running time difference between successive trains
- Practical capacity  $C_p = T / \sum(t_{h\ min} + \Delta t + t_b)$   
Maximum number of trains  $N_p$  at scheduled order and speed including running time supplements  $t_r$ , buffer times  $t_b$  and track possession time for infrastructure inspection and maintenance

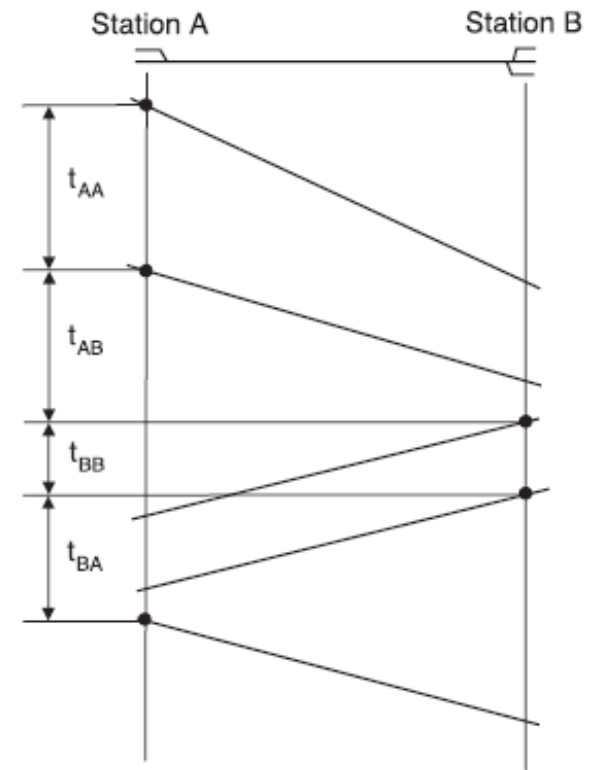
# Assignment of headway times

## A. To stations

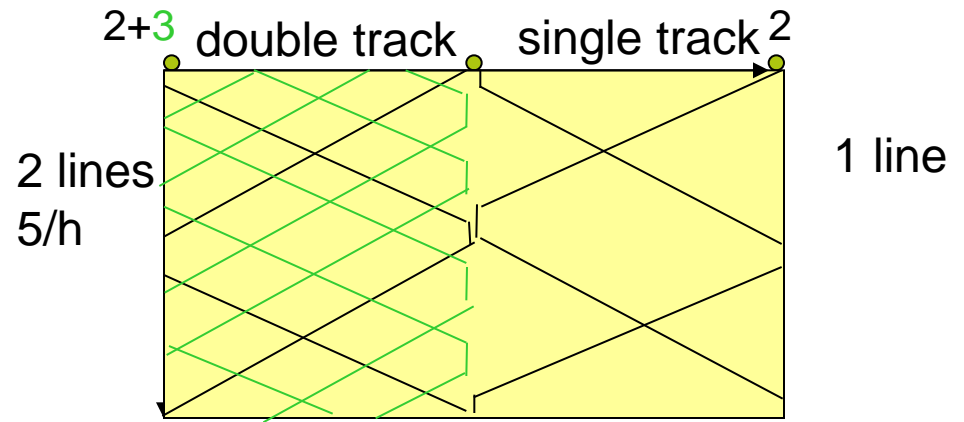
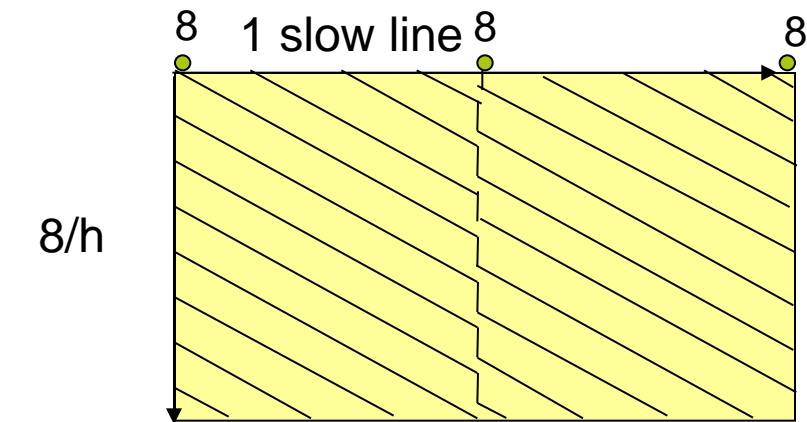
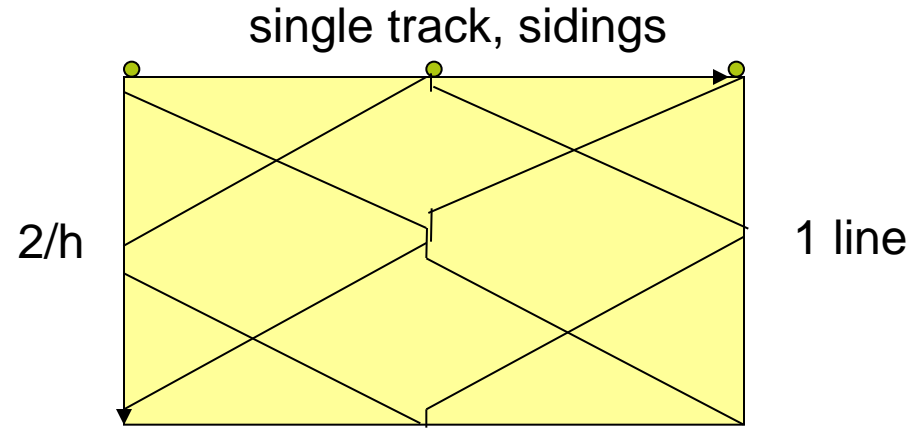
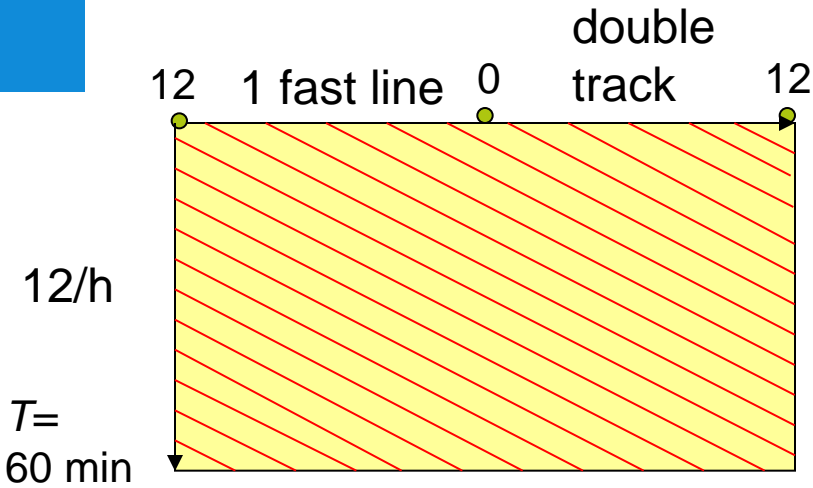


$t_{dd,X}$  "depart-depart" headway at station X  
 $t_{da,X}$  "depart-arrive" headway at station X  
 $t_{aa,X}$  "arrive-arrive" headway at station X  
 $t_{ad,X}$  "arrive-depart" headway at station X

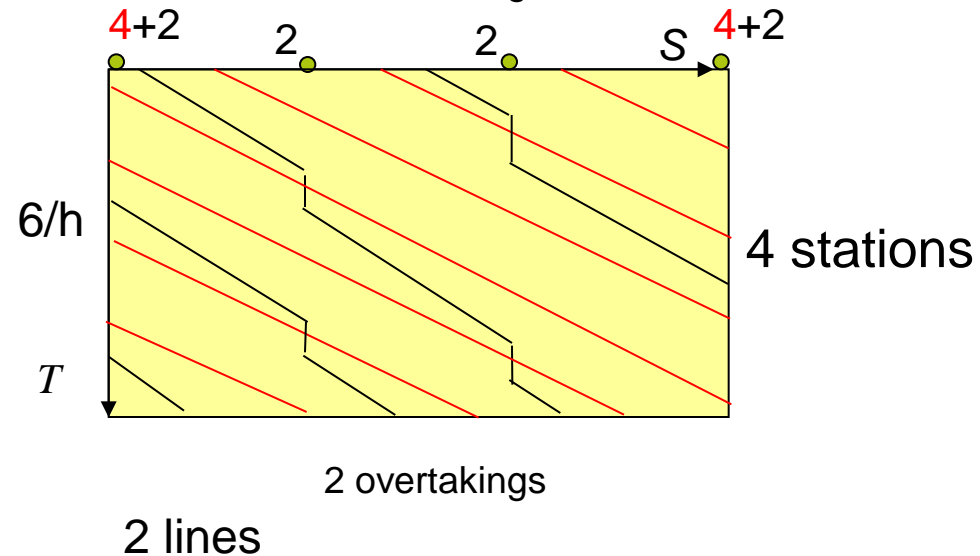
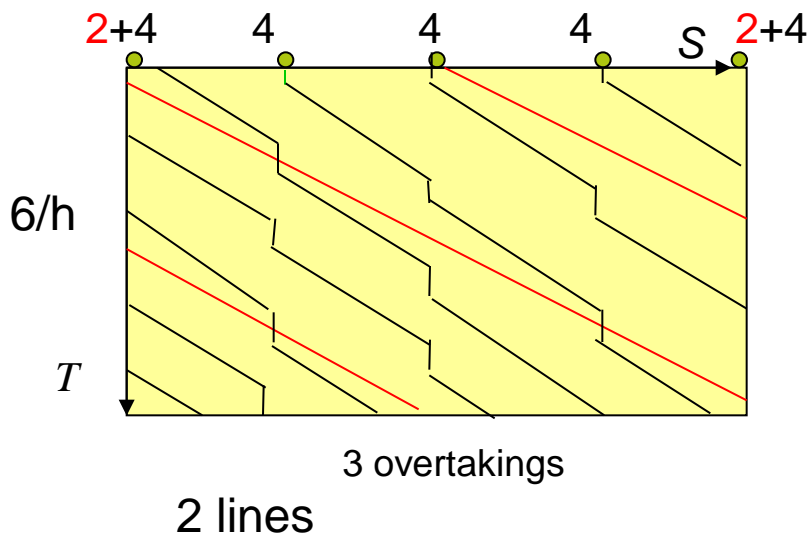
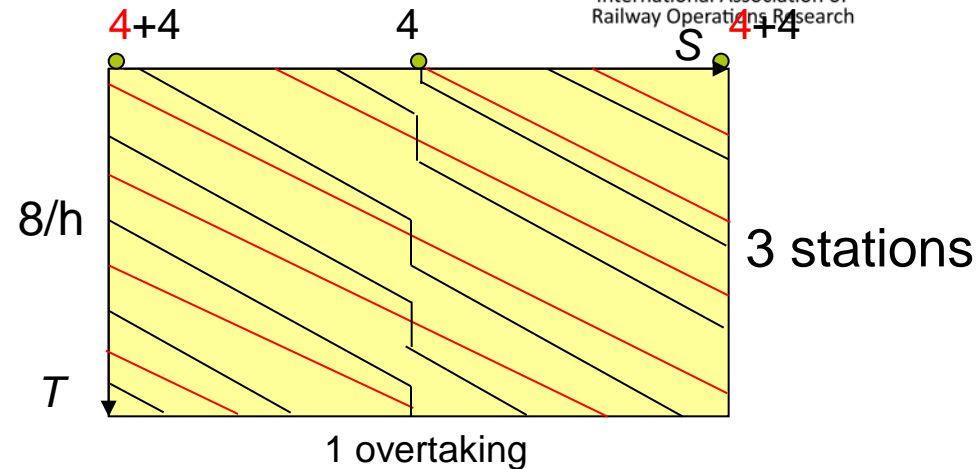
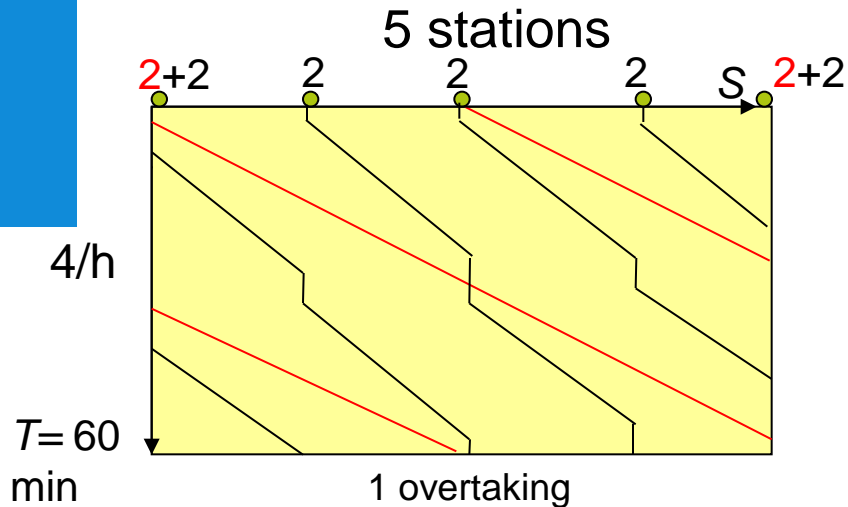
## B. To track sections



# Impact of speed and number of tracks on capacity



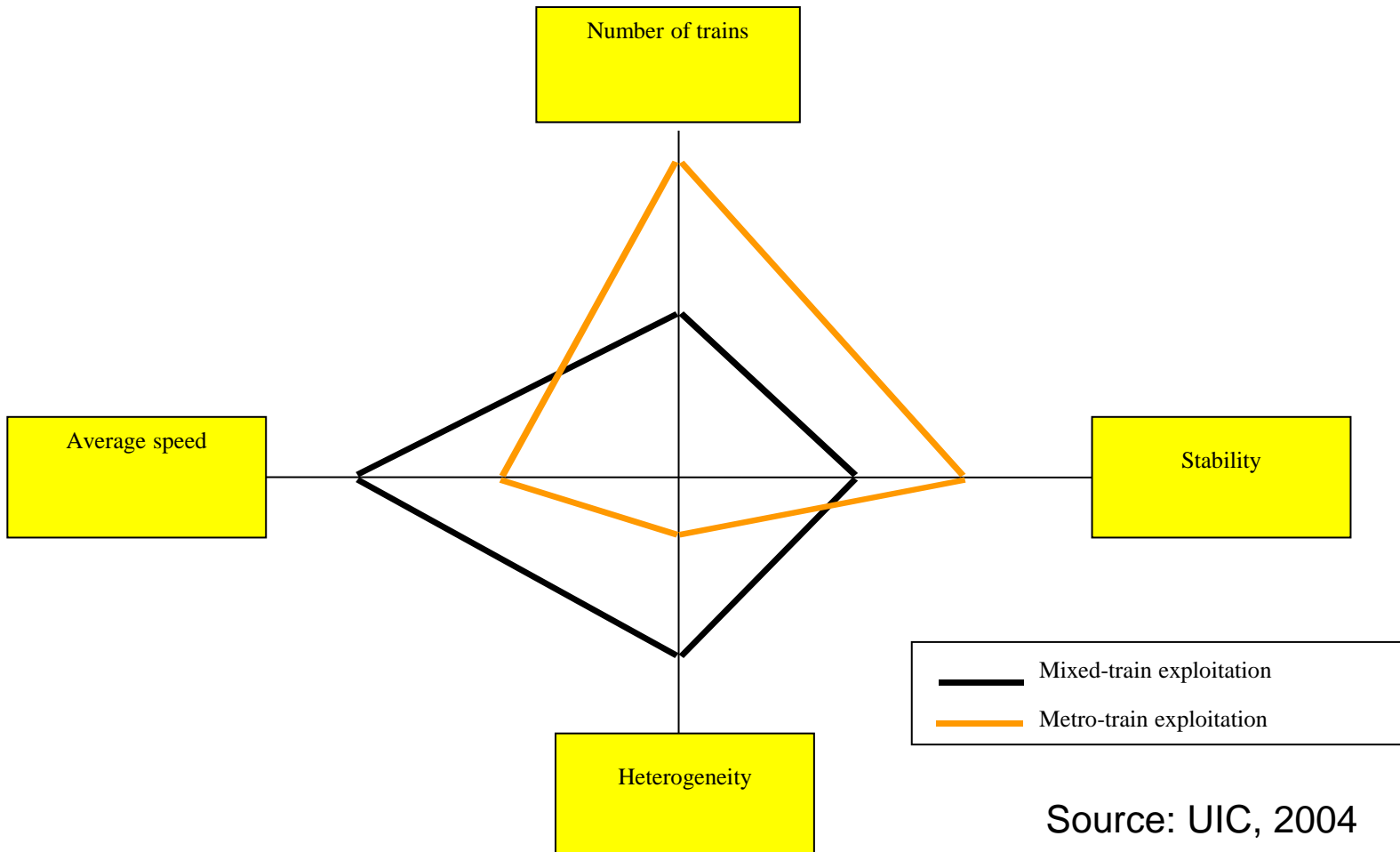
# Impact of station density, service pattern and overtakings on capacity



# Capacity depends on

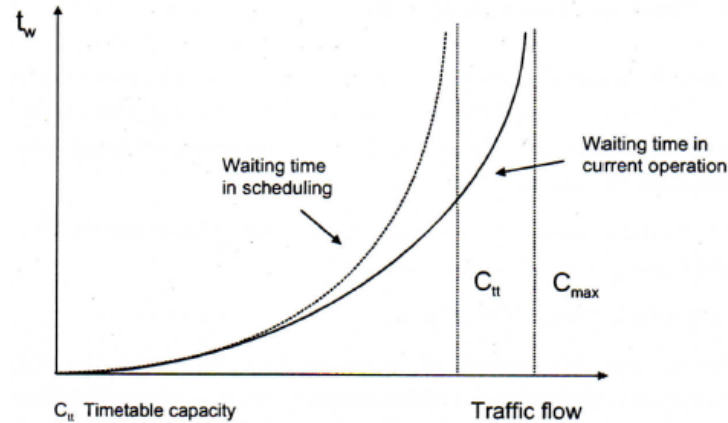
- **Timetable**
    - Train speed and homogeneity
    - Train order
  - **Infrastructure**
    - Alignment
    - Number and length of tracks
    - Number of stations
    - Number of lines
    - Signalling & safety system
  - **Rolling stock**
  - **Weather**
  - **Human behavior**
- Travel time differences
  - Minimum headways
  - Timetable margins
  - At-grade crossings, flyovers, speed reductions, steep gradients
  - Single (bidirectional), passing loop
  - Double, merging/diverging/crossing, terminal, stabling
  - Fixed block {one-section/multiple track sections}
  - Automatic Train Protection (ATP)
  - Automatic Train Control (ATC)
  - Automatic Train Operation (ATO)
  - Moving block

# Capacity balance

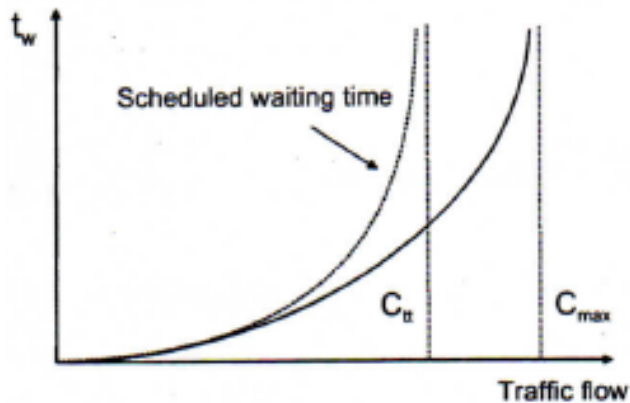


Source: UIC, 2004

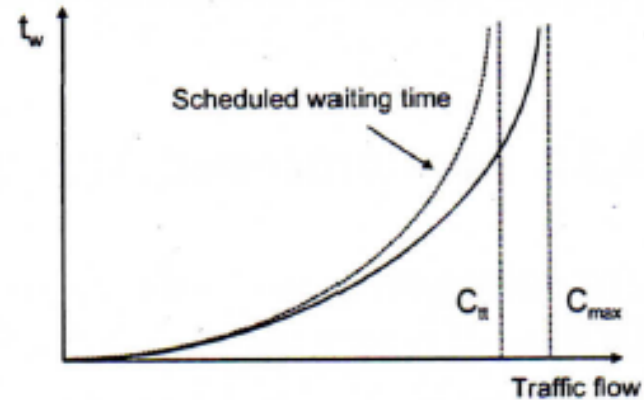
# Trade-off between waiting time and capacity



a) High variation of minimum line headways



b) Low variation of minimum line headways



# Classification of capacity estimation models

## A. Graphical (rule based)

- ❖ Train diagramming

## B. Analytical

- ❖ Track occupation
- ❖ Waiting time, stability margin
- ❖ Queueing

## C. Simulation

- ❖ Macroscopic
- ❖ Microscopic

## D. Combinatorial Optimisation

- ❖ (Mixed) Integer Linear Programming
- ❖ Heuristics (local search, genetic, tabu search)
- ❖ Stochastic programming, Light robustness, Recoverable robustness, ...

	<u>Open track</u>	<u>Station</u>	<u>Network</u>
	✓	✓	✓
	✓	✓	-
	✓	(✓)	(✓)
	✓	(✓)	-
	✓	✓	(✓)
	✓	✓	((✓))
	✓	(✓)	(✓)

✓

✓

-

✓

(✓)

(✓)

✓

(✓)

-

✓

✓

(✓)

✓

✓

((✓))

✓

(✓)

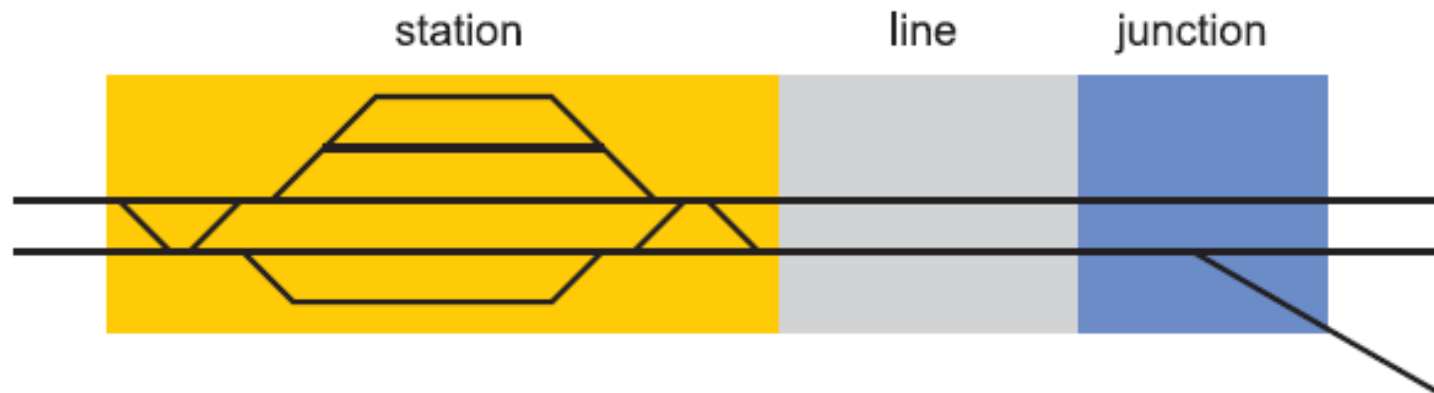
(✓)

# Infrastructure model

## A. Macroscopic model



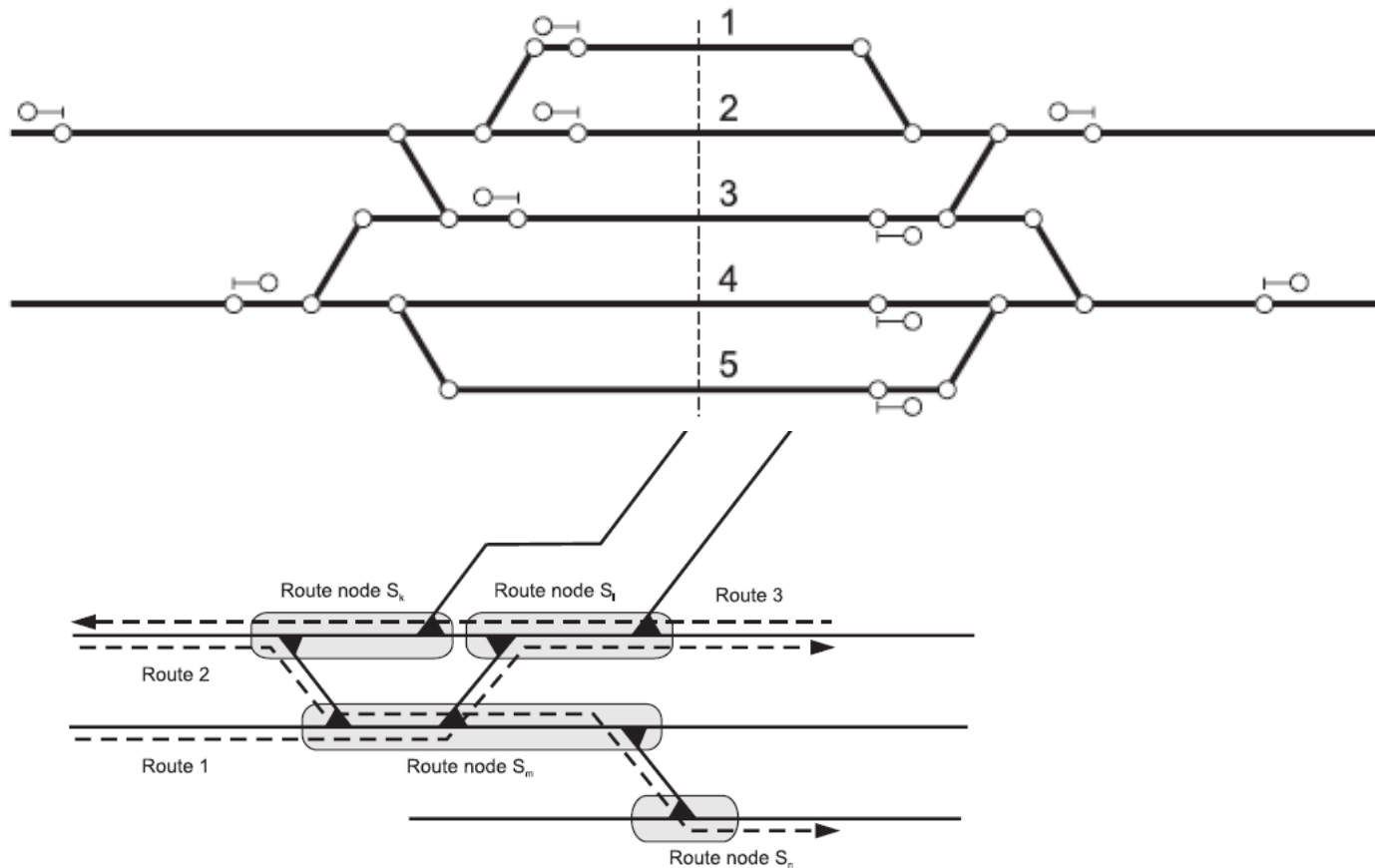
## B. Microscopic model



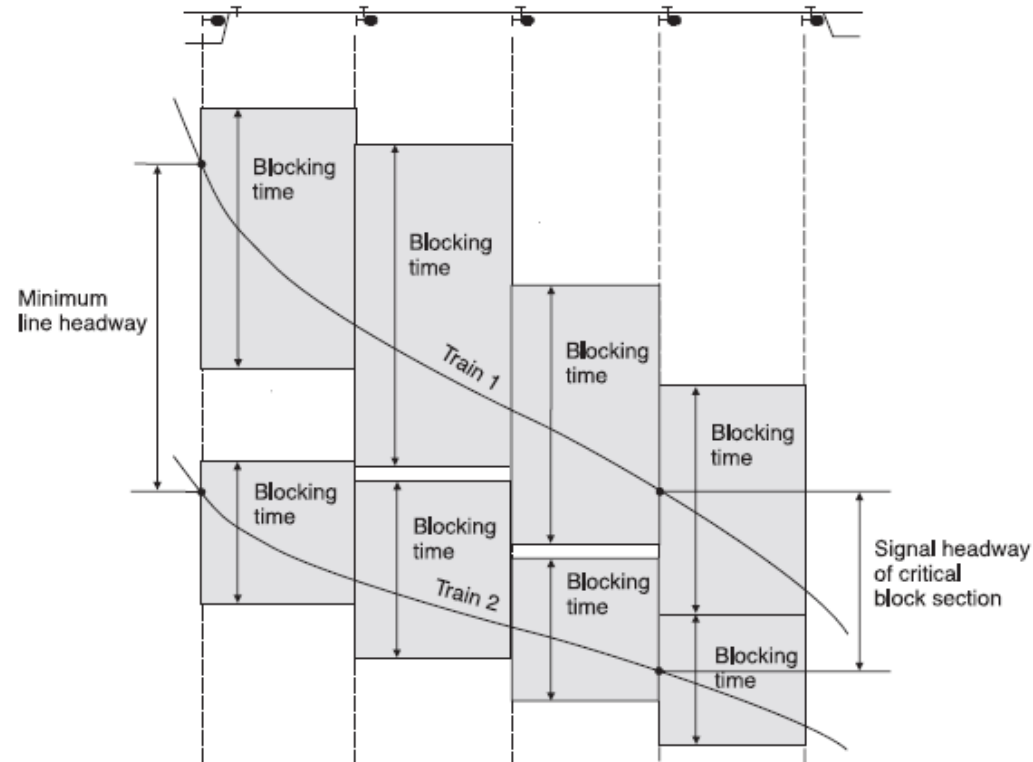
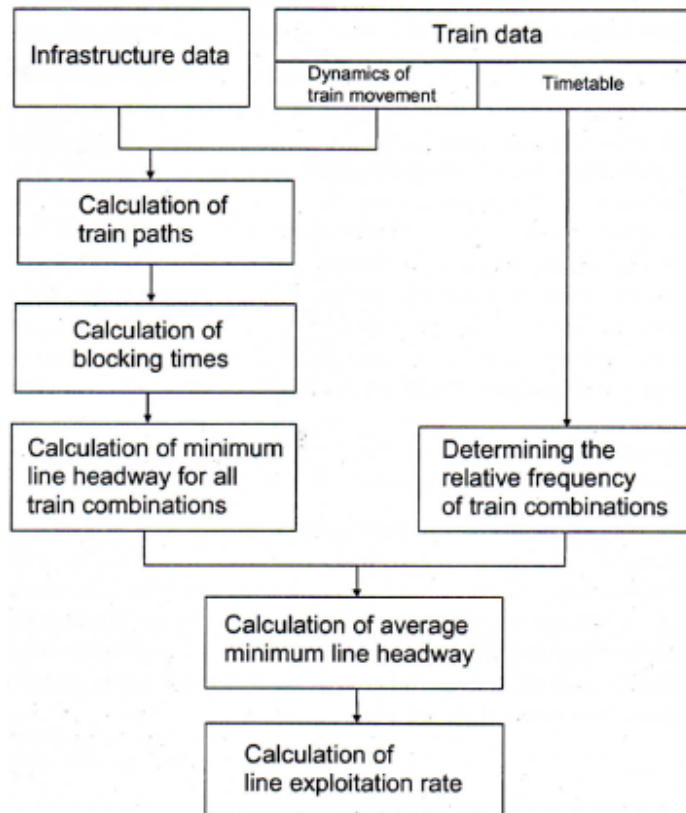
# Drawbacks of macroscopic capacity estimation model

- Inaccuracy of scheduled travel (running and dwell) times
    - Linear train graphs: time loss due to acceleration, coasting and deceleration unknown/not disaggregated
    - Scale: rounded-up to full minutes
    - Discrete point modelling of trains: variation of train length neglected
  - Validation of scheduled minimum time headways missing
    - Use of given standard minimum headway values (safety constraints)
    - Variation of train speed and minimum headway times neglected
    - Impact of ATP, ATC neglected
  - Timetable margins unknown
    - Standard running time supplements not verified nor differentiated
    - Amount of buffer times unknown
- ⇒ Insufficient precision and reliability of capacity estimation!

# Microscopic model of station track yard and route nodes



# Analytical capacity estimation



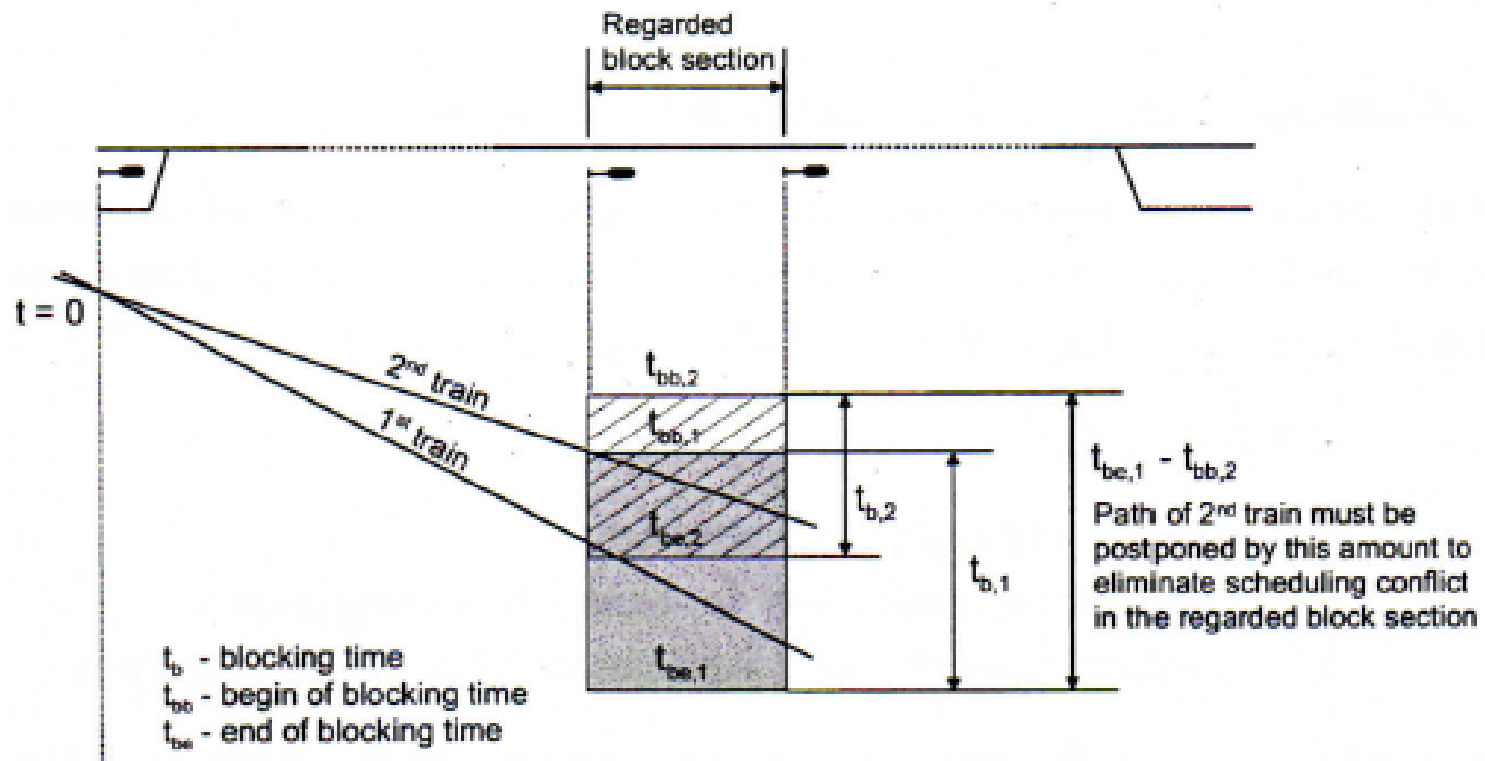
# Microscopic infrastructure and rolling stock model



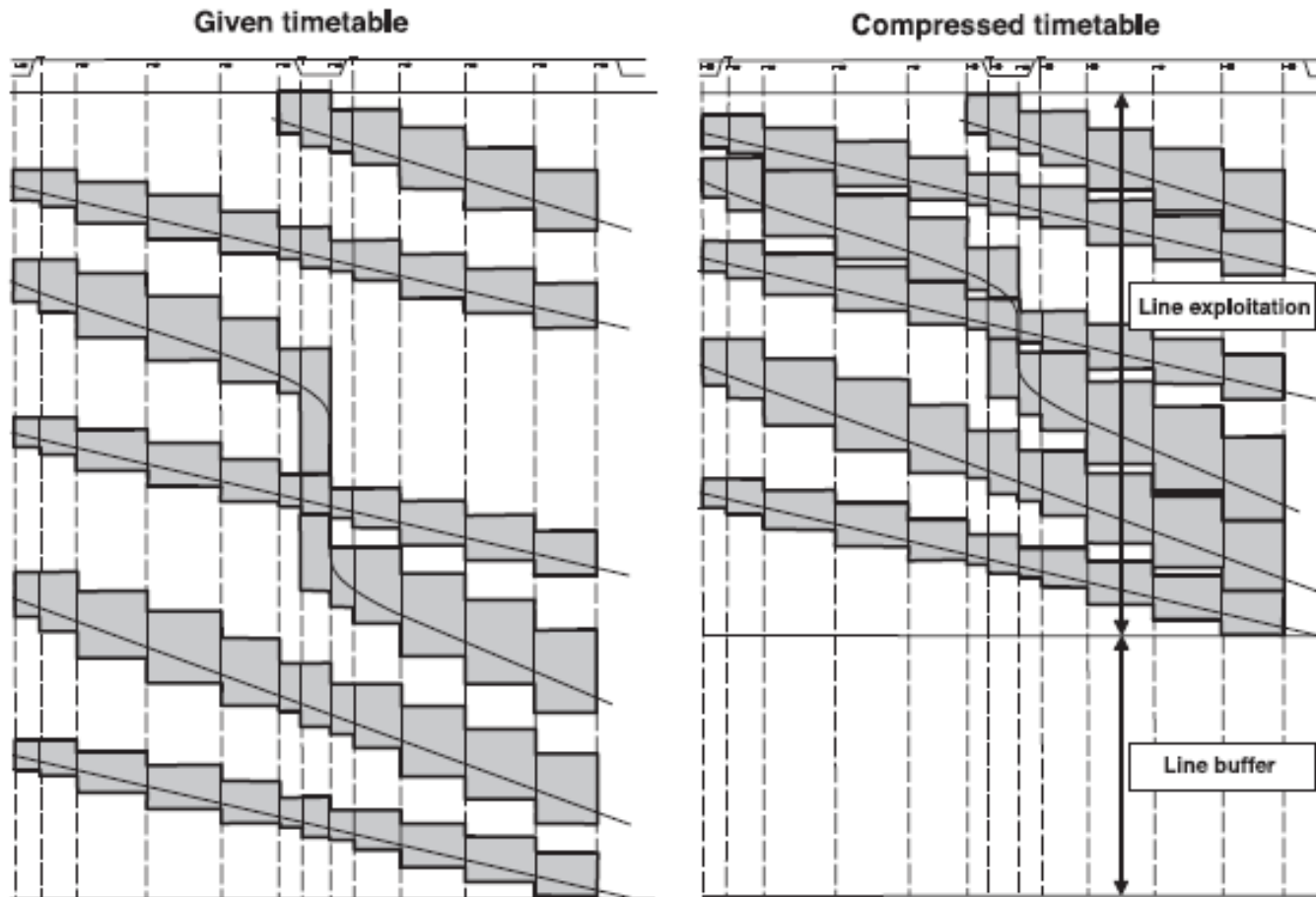
## Input

- Graph modelling of track infrastructure
  - Track section and platform lengths, radii, gradients and max. speed
  - Location and distance between signals, switches, crossings, insulation joints, overhead contact line separators,
- Specification of signalling and safety systems
  - Blocking and clearance, signal aspects, overlaps
  - Train detection, location of track circuits/axle counters/fouling points
  - Interlocking, set-up and (partial) release of routes
  - Train protection, train control
  - Train regulation
- Train length, weight, resistance, tractive effort-speed diagram

# Calculation of (scheduled) blocking time overlap



# Capacity consumption: 'compression' of blocking time diagrams

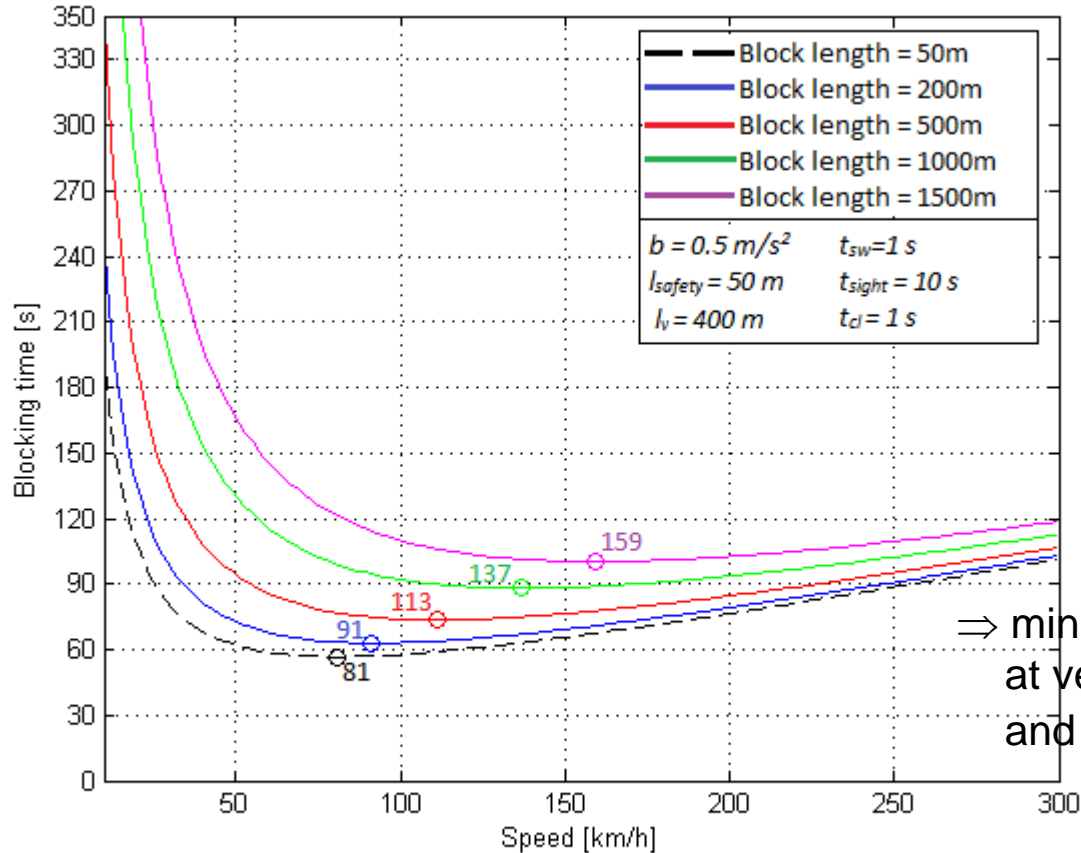


⇒ Drawbacks:  
timetable  
dependency,  
transferability

# Microscopic capacity estimation

1. Estimation of blocking times  $t_{b/i}$  of trains per line
2. Determination of minimal headway time  $t_{hij}$  between trains
  - at departure (according to different train sequences)
  - at arrival (stations)
  - at conflict points (merging/crossing of lines, long block, speed limit)
3. Determination of prevailing minimal headway times;  
mean minimal headway  $t_{hm} = \sum (t_{hij} \cdot p_{ij})$ ;  $p_{ij} = n_i \cdot n_j / n^2$
4. Estimation of (scheduled/feasible) number of train path  $n/n_{max}$
5. Estimation of total track occupation time of compressed  
(scheduled) train graph  $T_{toc} = n \cdot t_{hm}$
6. Estimation of scheduled track occupation  $\rho_s = T_{toc} / T$  [%]
7. Estimation of maximal track occupancy  $\rho_{max} = T_{toc\ max} / T$  [%]

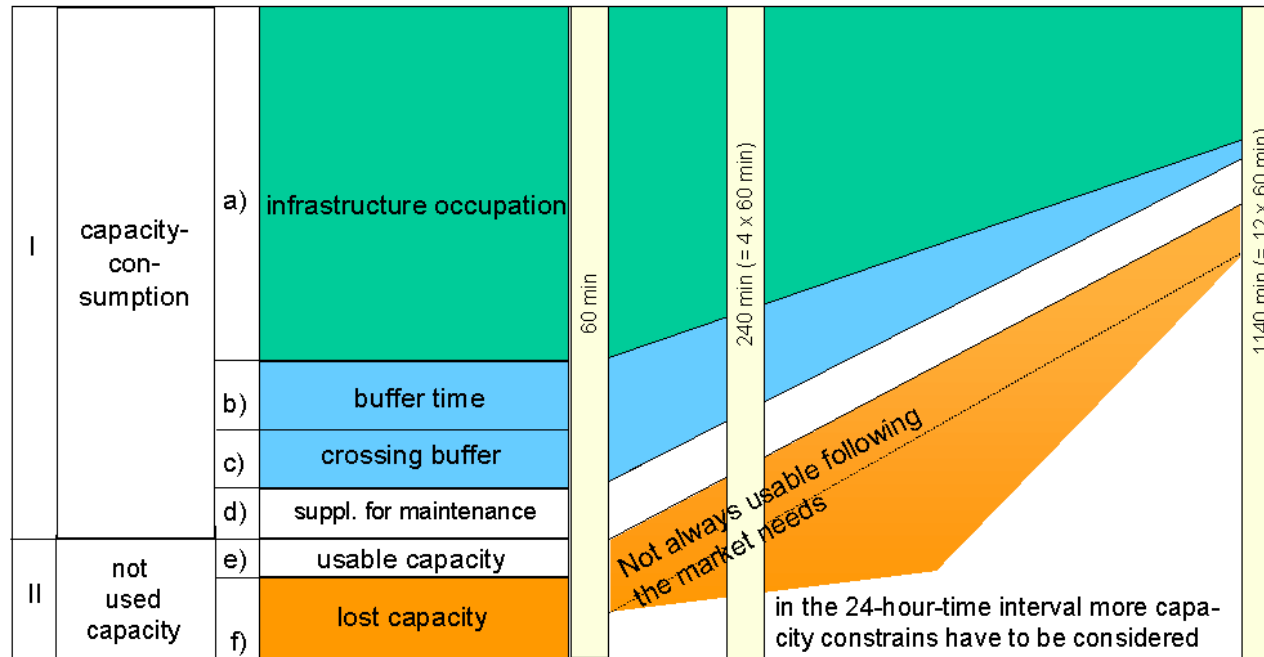
# Impact of block length and speed on blocking time



⇒ minimum blocking time at very short block length and low speed  $\approx 80$  km/h!

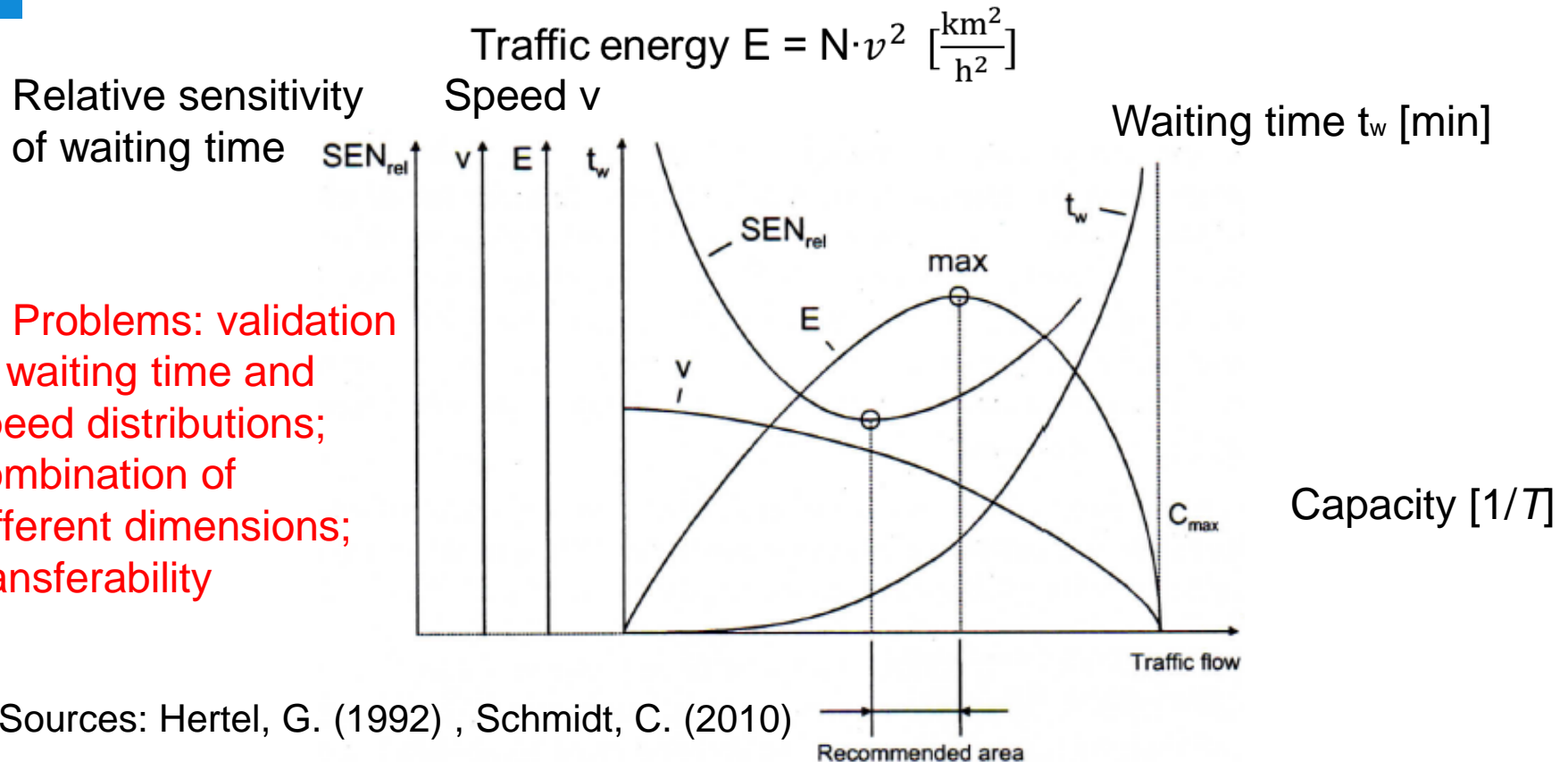
$$T_{block} = t_{sw} + t_{sight} + (l_{br} + l_{block} + l_{safety} + l_v) / v + t_{cl}$$

# Capacity consumption levels recommended by UIC leaflet 406

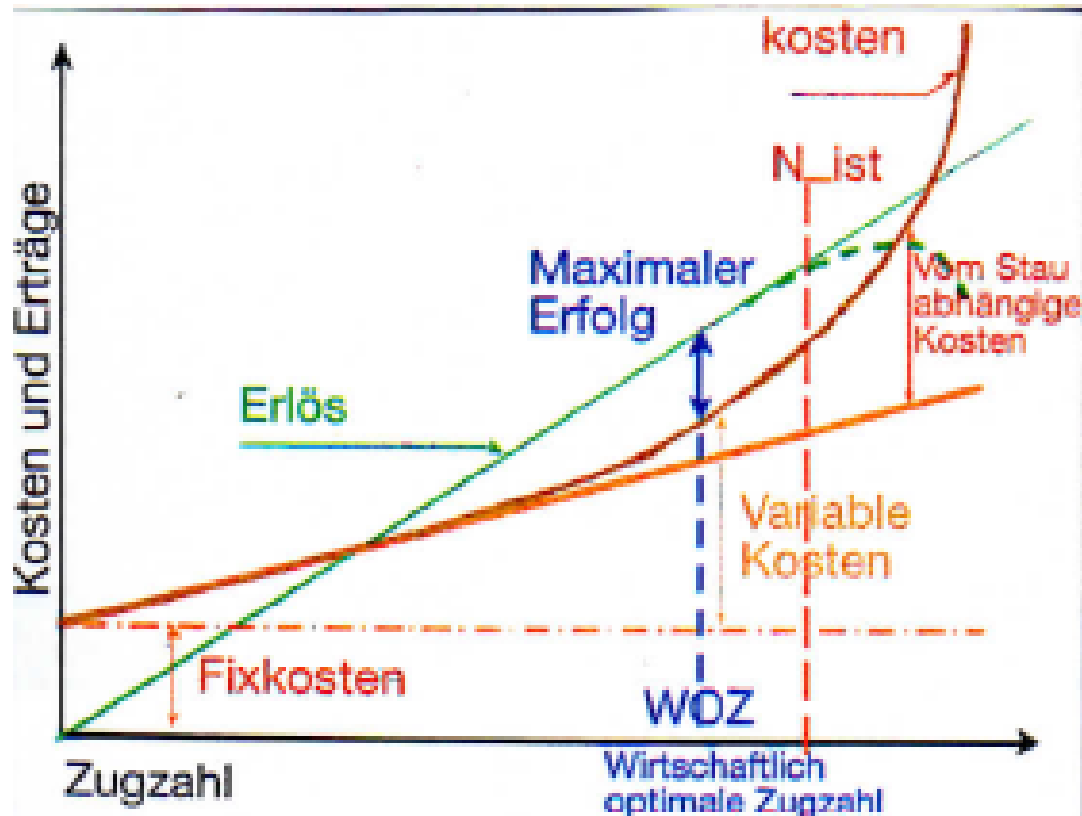


Type of line	Peak hour	Daily period
Dedicated suburban passenger traffic	85 %	70 %
Dedicated high speed lines	75 %	60 %
Mixed traffic lines	75 %	60 %

# Recommended area of traffic flow



# 'Optimal' analytical route capacity and cost-benefit model

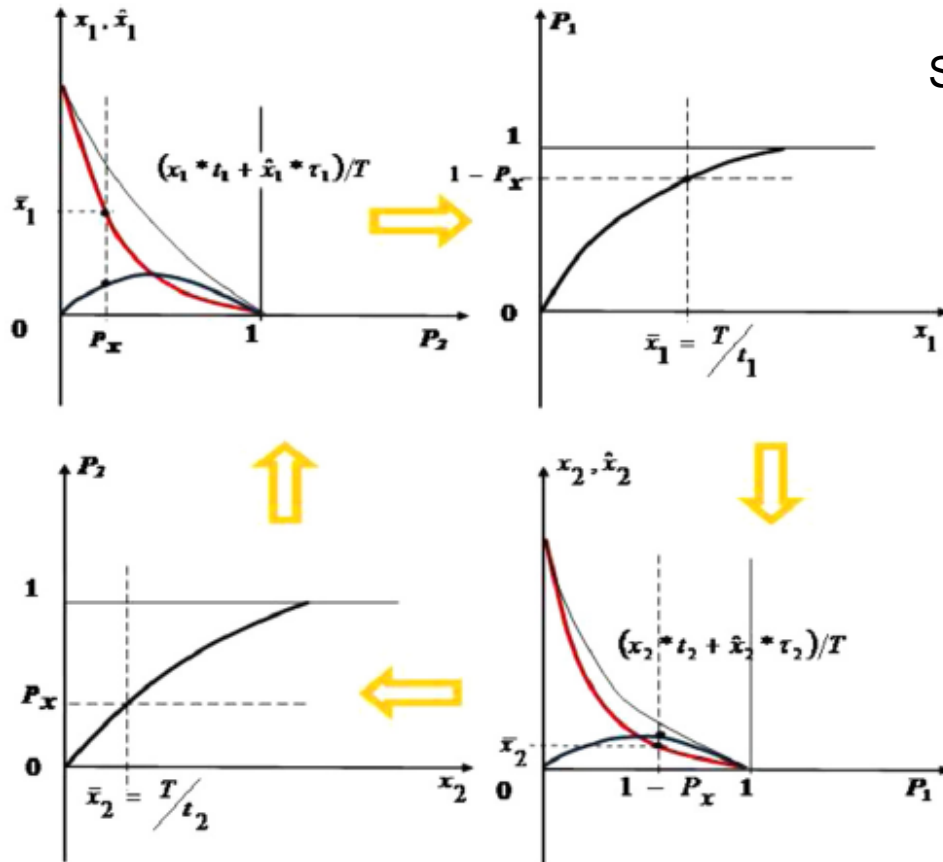


Source: Schwanhäuser, 2009

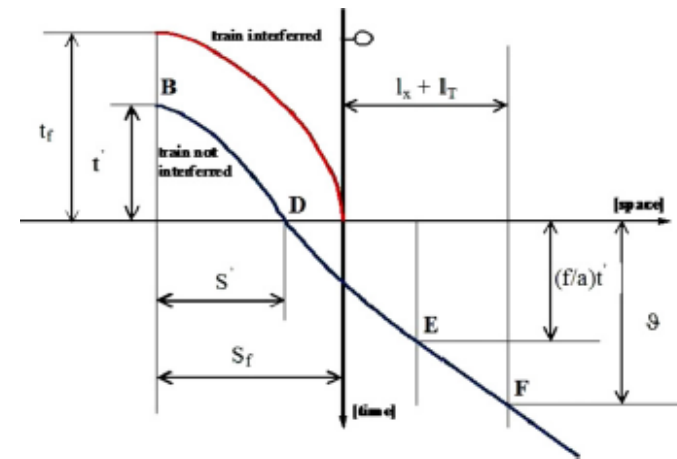
- ⇒ Shortcomings:
- complicated parameter calibration;
  - validation?
  - optimality gap unknown

# Analytical junction occupation model

Source: Mussone & Wolfer Calvo, 2013

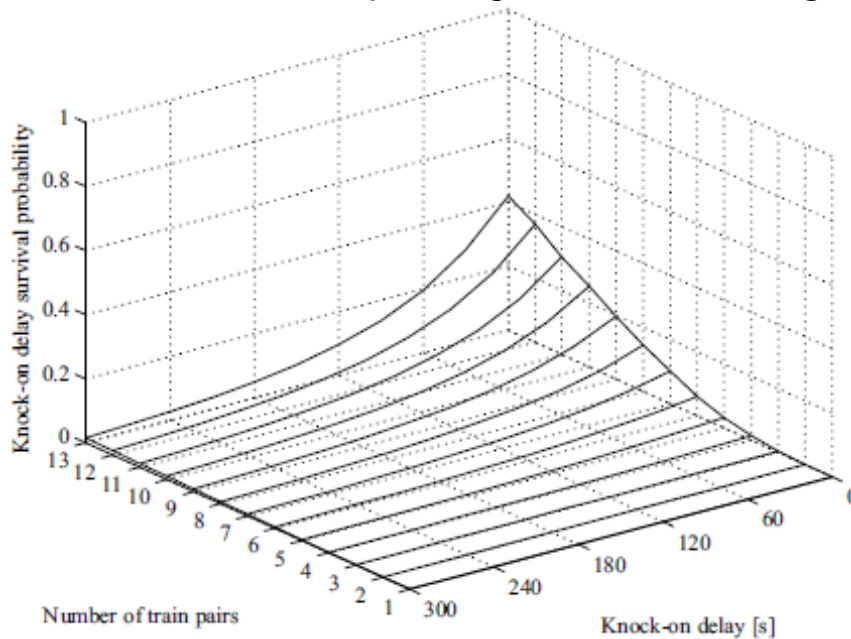


⇒ Shortcoming: Switch, sight and approach time neglected

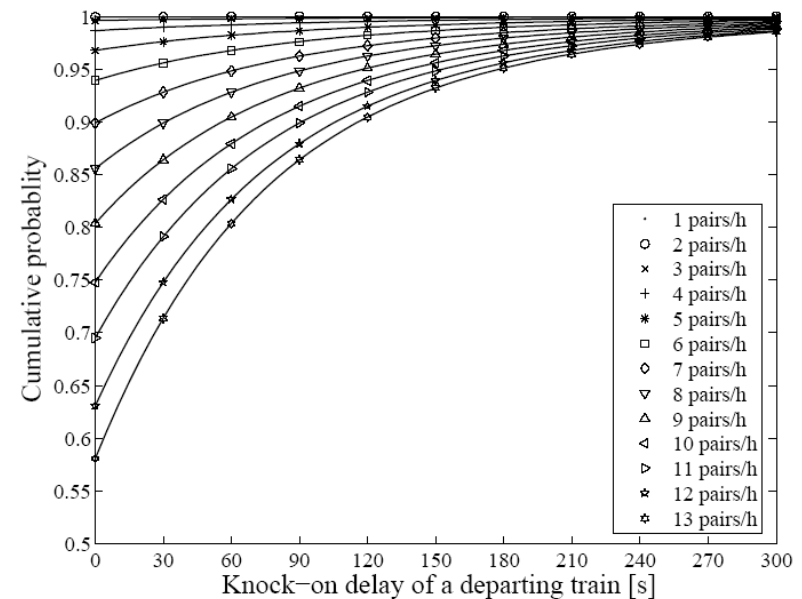


# Conditional probability junction capacity estimation model

Consecutive delay survival probability for northbound departing trains as function of number of trains passing at level crossing



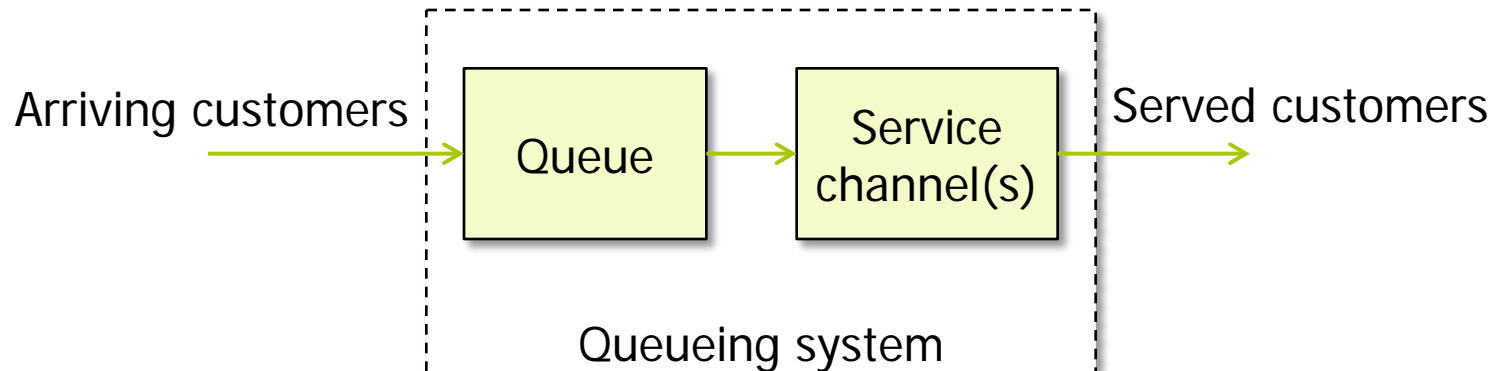
Consecutive delay survival probability for a northbound departing train as function of train frequency passing at level crossing



Source: Yuan, 2006

⇒ Shortcomings:...

# Queuing models to estimate train delays and capacity



- A: Arrival process with interarrival time distribution  $A$ 
  - Arrival time  $t_k$ , interarrival time  $a_k = t_{k+1} - t_k$  of customer  $k$
- B: Service process with service distribution  $B$ 
  - Service time  $b_k$
- $n$ : Number of service channels
- Queue discipline: First Come First Served (FCFS), Priority (PR)

Notation:

- $A/B/n$  are systems with queue capacity  $m = \infty$  (general  $A/B/n/m$ )

# Classification of queuing models

Variation coefficients

$$k = 1/V_A^2 = (\sigma^2 A / E^2 A)^{-1}, \quad V_A: \text{variance of arrival headway times}$$

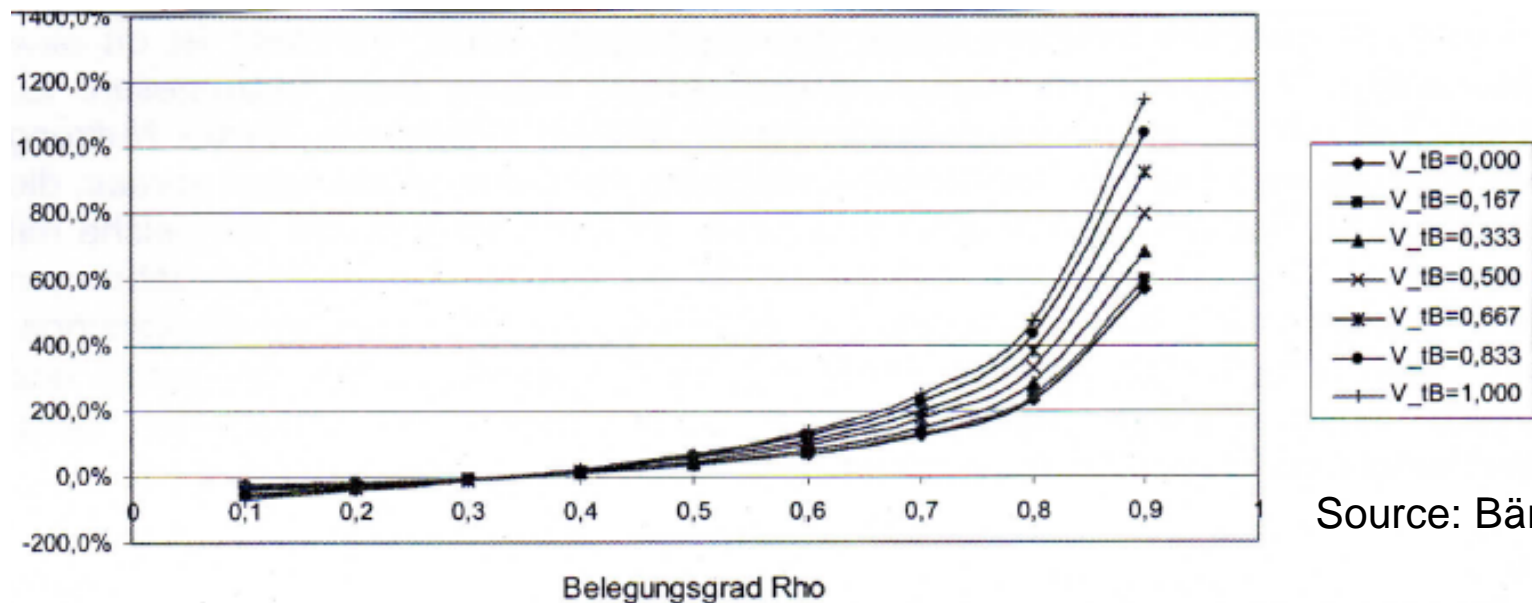
$$l = 1/V_B^2 = (\sigma^2 B / E^2 B)^{-1}, \quad V_B: \text{variance of service times (minimal headway)}$$

Arrival distribution	Service distribution			
	big variance $0 < l < 1$	Markov $l = 1$	small variance $1 < l < \infty$	deterministic $l = \infty$
big variance $0 < k < 1$				
Markov $k = 1$		M/M/1	M/GI/1	M/D/1
small variance $1 < k < \infty$		GI/M/1	GI/GI/1	
deterministic $k = \infty$		D/M/1		D/D/1

M: Markovian process (exponential)  
 GI: General independent process (iid)  
 D: Dirac process (deterministic, variance = 0)

⇒ **Drawbacks:** iid assumption invalid for periodic timetable and heavy traffic; service distributions unknown

# Comparison with estimated waiting time of clockface timetable



Source: Bär, 2009

Relative deviation between  $GI/GI/1$  queuing system and 2-train clockface timetable

⇒ Modelling of clockface process by estimated waiting time and variation coefficient of interarrival time must be **rejected!**

# Combinatorial optimisation models

1. Integer Linear Programming (ILP)  
(Review see Cacciani & Toth, 2012)
2. Genetic Algorithms (GA) }
3. Simulated Annealing (SA) } Disregarded here
4. Constraint Programming (CP) }
5. Job-Shop Scheduling  
(Mascis & Pacciarelli, 2002; D'Ariano et al. 2007, 2008, 2009...  
Burdett & Kozan, 2004, 2006, 2009, 2010; Liu & Kozan, 2011)

# Integer Linear Programming (ILP)

- Objective functions
  - Maximize customer satisfaction (minimize travel times, waiting times, number of transfers)
  - Maximize infrastructure use (train throughput)
  - Minimize changes w.r.t. requested arrival & departure times
  - Minimize costs of train operation
  - Increase robustness (avoid/minimize delay propagation)
- Constraints
  - Periodicity (regularity, running, dwell and transfer times)
  - Safety (minimal headway times)
  - Infrastructure discontinuities (number of tracks, terminals, switches, crossings, platform length)
  - Rolling stock composition, performance and assignment
  - etc.

# Integer Linear Programming (ILP)

## Cyclic train timetabling

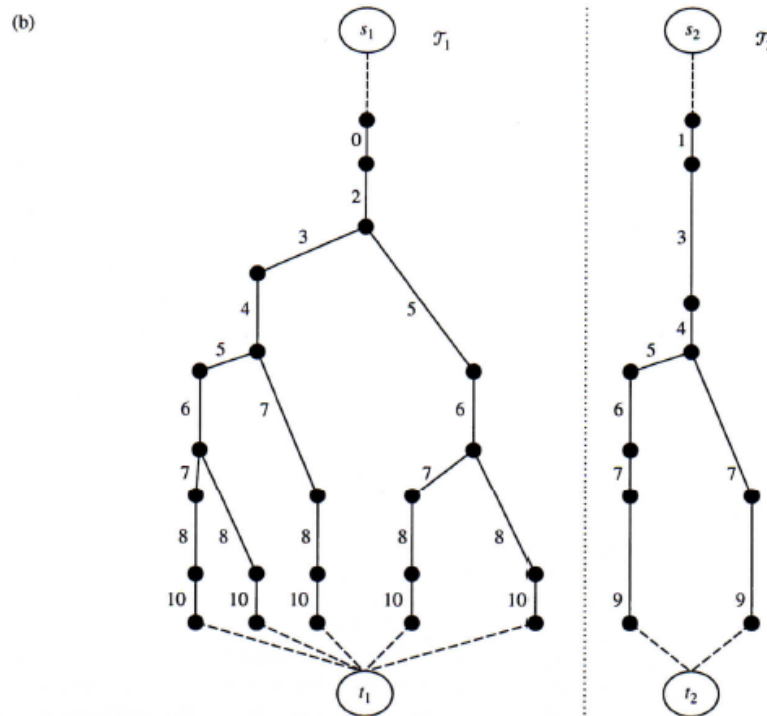
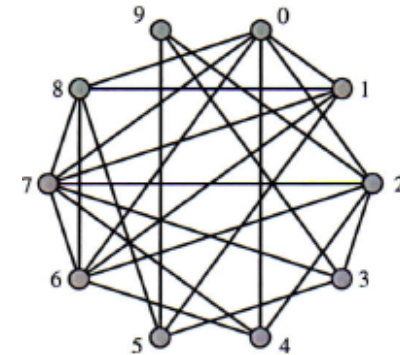
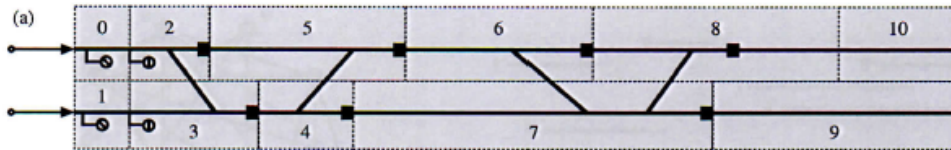
- Periodic Event Scheduling Problem (PESP)
  - Constraint Graph (directed graph with node set  $N$  and arc set  $A$ )
  - Constraint propagation algorithm
- Given timetable of train operator assumed to be *feasible*, i.e. route conflicts within stations (interlocking areas) neglected
- Timetable consists of fixed running times between stations that may be altered within certain bounds (time windows)
- Solving PESP of larger networks and complex nodes is *NP hard*
- Relaxation of (some) constraints in order to find a near-optimal solution and/or speed-up computation time
- Mixed Integer Linear Programming (MILP) tolerates mix of non- and integer constraints

# Integer Linear Programming (ILP)

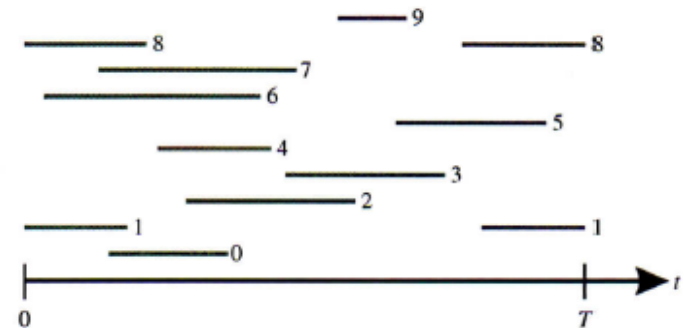
## Non-cyclic train timetabling

- More appropriate in competitive market with generally conflicting train path requests
- Arrival and departure times represented by continuous variables (MILP)
- Binary variables expressing the train order at departures
- Discretising the time (each node corresponding a time instant, arcs representing train travel or stop at station)
- Costs associated e.g. with the deviation from preferred departure time, travel times and dwell times
- Set of arcs representing assumed *feasible* travel times between and dwell times per train at each station (time-space graph equals train path)
- Trains may be cancelled
- Easier to solve than PESP of periodic timetable (more flexible)

# Resource-Tree Conflict Graph



Source: Caimi et al, 2011



# Drawbacks of (M)ILP

Exeption: ILP relaxation of Resource-Tree Conflict Graph by Caimi et al, 2011

- Possible headway conflicts on open and station tracks due to speed limitations and/or longer approach and block signal distances neglected
  - Route conflicts in interlocking areas neglected (*feasibility* assumption)
  - Shunting movements generally neglected
  - Availability of sufficient parking/stabling tracks not guaranteed
- ⇒ Validity of given timetable and used safety constraints not proven!
- ⇒ Lack of accuracy compared to analytical, queuing and microscopic simulation models that may endanger timetable feasibility!
- ⇒ Optimality claim may lead to misunderstanding
- ⇒ Stochastic Programming and Robust MILP train timetabling models cannot provide optimal allocation of timetable margins due to inherent restrictions of macroscopic models!

# Conclusion

- Track capacity is influenced by the timetable, infrastructure, signalling and safety systems, rolling stock, weather and human behaviour
- Macroscopic capacity estimation models simplify infrastructure, route and signalling constraints but can support strategic network and timetable planning
- Microscopic capacity models can accurately estimate minimum headways, capacity consumption and timetable margins for different signalling and safety systems
- (M)ILP cannot substitute more accurate analytical capacity analysis and microscopic timetable simulation
- Micro-macroscopic (meso-scopic) models can bridge the gap between accurate analytical and efficient combinatorial optimisation

# Literature



- Bär, M. (2009), Analytische Modelle für Leistungsuntersuchungen im Eisenbahnnetz unter den speziellen Bedingungen von Taktfahrplänen, 23. Verkehrswiss. Tagung TU Dresden
- Burdett, R.L, Kozan, E. (2010), A disjunctive graph model and framework for constructing new train schedules, *European Journal of Operational Research* 200, 85-98
- Burdett, R., Kozan, E. (2006), Tecxhniques for absolute capacity determination in railways, *Transportation Research Part B*, 40(8), 616-632
- Cacchiani, V., Toth, P. (2012), Nominal and robust train timetabling problems, *EJOR* 219, 727-737
- Caimi, G., Chudak, F., Fuchsberger, M., Laumanns, M., Zenklusen, R. (2011), A New Resource-Constrained Multicommodity Flow Model for Conflict-Free Train Routing and Scheduling, *Transportation Science*, 45(2), 212-227
- Gudehus, T. (1976), Grenzleistung bei absoluter Vorfahrt, *Zeitschrift für Operations Research*, 20, B127-B160
- Hansen, I.A., Pachi, J. (2008), *Railway Timetable and Traffic. Analysis Modelling Simulation*, Hamburg: Eurailpress
- Hansen, I.A. (2006), State-of-the-art of Railway Operations Research, in: J. Allan, C.A. Brebbia, R.J. Hill, G. Sciotto & S. Sone (eds.), *Computers in Railways X*: WIT Press, 565-579
- Hansen. I.A. (2004), Increase of capacity through optimised timetabling, in: J. Allan, C.A. Brebbia, R.J. Hill, G. Sciotto & S. Sone (eds.), *Computers in Railways IX*: WIT Press, 529-538

# Literature



- Hertel, G. (1992), Die maximale Verkehrsleistung und die minimale Fahrplanempfindlichkeit auf Eisenbahnstrecken, *ETR* 41(10), 665-671
- Landex, A. (2009), Evaluation of Railway Networks with Single Track Operation Using the UIC 406 Capacity Method, *Networks and Spatial Economics* 9(1), 7-23
- Lindner, T. (2011), Applicability of the analytical UIC Code 406 compression method for evaluating line and station capacity, *Journal of Rail Transport Planning & Management*, 1(1), 49-57
- Liu, S.Q., Kozan, E. (2011), Scheduling Trains with Priorities: A No-Wait Blocking Parallel-Machine Job-Shop Scheduling Model, *Transportation Science*, 45(2), 175-198
- Mussone, L, Wolfler Calvo, R. (2013), An analytical approach to calculate the capacity of a railway system, *European Journal of Operational Research* 218, 11-23
- Nießen, N., Wendler, E. (2007), Grenzleistung von Gesamtfahrstraßenknoten, 21. Verkehrswiss. Tage TU Dresden
- Pahl, J. (2002), Railway Operation and Control, VTD Rail Publishing
- Schmidt, C. (2010), Experimentelle Bestimmung der Wartezeitfunktion für Leistungsuntersuchungen, *Eisenbahn Technische Rundschau ETR*, 1+2, 33-39

# Literature



- Schwanhäüßer, W. (1994), The Status of German Railway Operations Management in Research and Practice, *Transportation Research Part A*, 28(6), 495-500
- Schwanhäüßer, W. (2009), Wirtschaftlich und betrieblich optimale Zugzahlen auf Eisenbahnstrecken, *ETR*, 9, 488-495
- Union Internationale des Chemins de Fer (UIC) (2004), UIC Code 406 - Capacity, Paris, 1-21
- Vakhtel, S. (2002), Rechnerunterstützte analytische Ermittlung der Kapazität von Eisenbahnnetzen, Dissertation, Veröffentlichung des Verkehrswissenschaftlichen Institutes der RWTH Aachen; H. 59
- Wakob, H. (1985), Ableitung eines generellen Wartemodells zur Ermittlung der planmäßigen Wartezeiten im Eisenbahnbetrieb unter besonderer Berücksichtigung der Aspekte Leistungsfähigkeit und Anlagenbelastung, Dissertation, Veröffentlichung des Verkehrswissenschaftlichen Institutes der RWTH Aachen; H. 36
- Wendler, E. (2007), The scheduled waiting time on railway lines, *Transportation Research Part B*, 41 (2), 148-158
- Wendler, E. (2008), Queueing, in: Hansen & Pacht (eds.) *Railway Timetable & Traffic. Analysis Modelling Simulation*, Hamburg: Eurailpress, 106-117
- Yuan, J., Hansen, I.A. (2007) Optimizing capacity utilization of stations by estimating knock-on train delays, *Transportation Research B* 41(2), 202-217