

# **School on Column Generation:** *Stabilized Column Generation*

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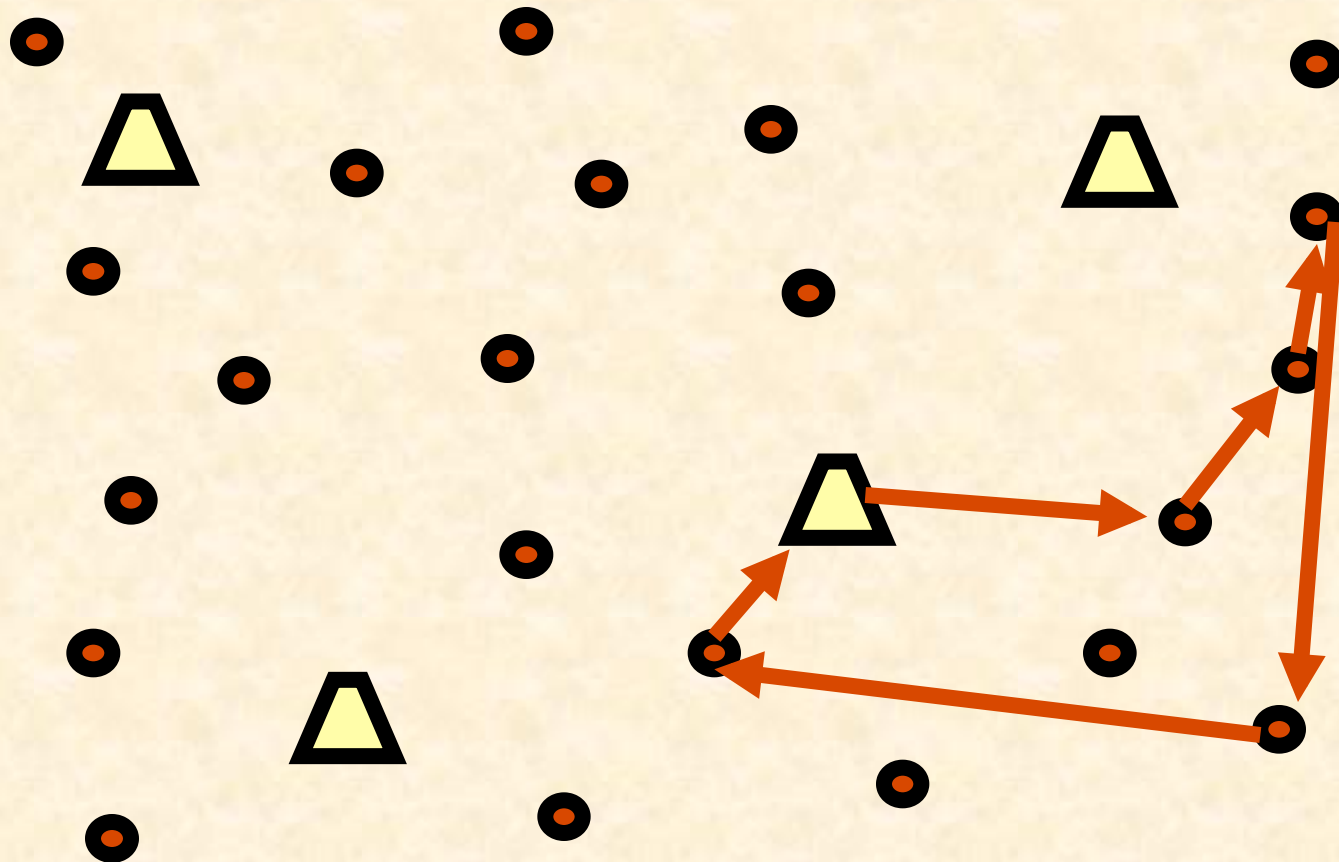
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# MULTIPLE DEPOT VEHICLE SCHEDULING PROBLEM (MDVSP)



## MDVSP: NOTATION

$K$  : set of depots

$N$  : set of scheduled tasks (no time windows)

$t_i$  : time at which service must start for task  $i \in N$

$s_i$  : service duration for task  $i \in N$

$t_{ij}$  : time to go from node  $i$  to node  $j$ , for  $i, j \in N \cup K$

$A$  : set of arcs;  $(i, j) \in A$  if  $t_i + s_i + t_{ij} \leq t_j$ , for  $i, j \in N \cup K$

$\Omega^k$  : set of paths from and to depot  $k \in K$

$c_p^k$  : cost of path  $p \in \Omega^k$

$a_{ip}^k = 1$  if path  $p \in \Omega^k$  covers task  $i \in N$ ; 0 otherwise

$v^k$  : available number of vehicles at depot  $k \in K$

## MDVSP: SET PARTITIONING FORMULATION

$$Z_M = \min \sum_{k \in K} \sum_{p \in \Omega^k} c_p^k \lambda_p^k$$

$$\sum_{k \in K} \sum_{p \in \Omega^k} a_{ip}^k \lambda_p^k = 1, \quad i \in N$$

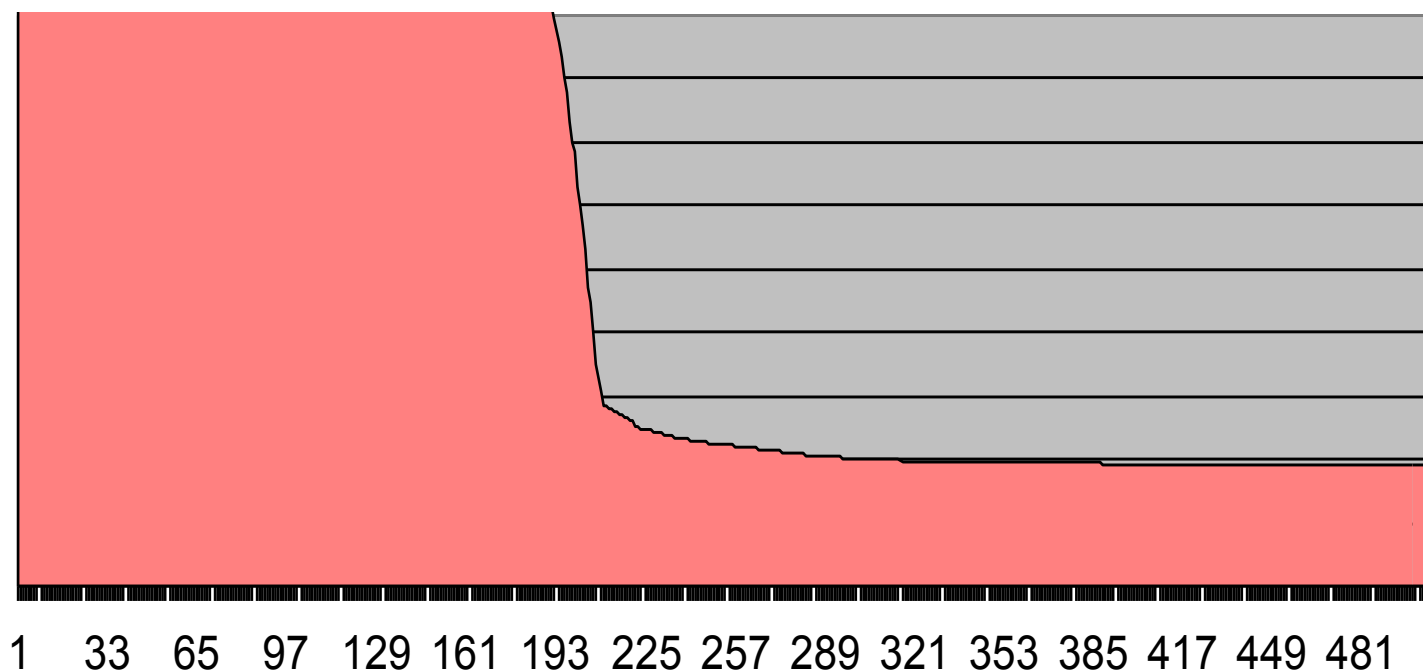
$$\sum_{p \in \Omega^k} \lambda_p^k \leq v^k, \quad k \in K$$

$$\lambda_p^k \text{ binary}, \quad k \in K, p \in \Omega^k$$

## PRIMAL DEGENERACY

<b><i>MDVSP</i></b> <b><i>R 800 (4)</i></b>	<b>cpu total</b>	<b>cpu mp</b>	<b>cpu sp</b>	<b># CG iter</b>	<b># SP cols</b>	<b># MP itr</b>
<b><i>standard CG</i></b>	<b>4178.4</b>	<b>3149.2</b>	1029.2	<b>509</b>	37579	926161

Tailing off effect of the objective function



## PRIMAL DEGENERACY

<b><i>MDVSP</i></b> <b><i>R 800 (4)</i></b>	<b>cpu total</b>	<b>cpu mp</b>	<b>cpu sp</b>	<b># CG iter</b>	<b># SP cols</b>	<b># MP itr</b>
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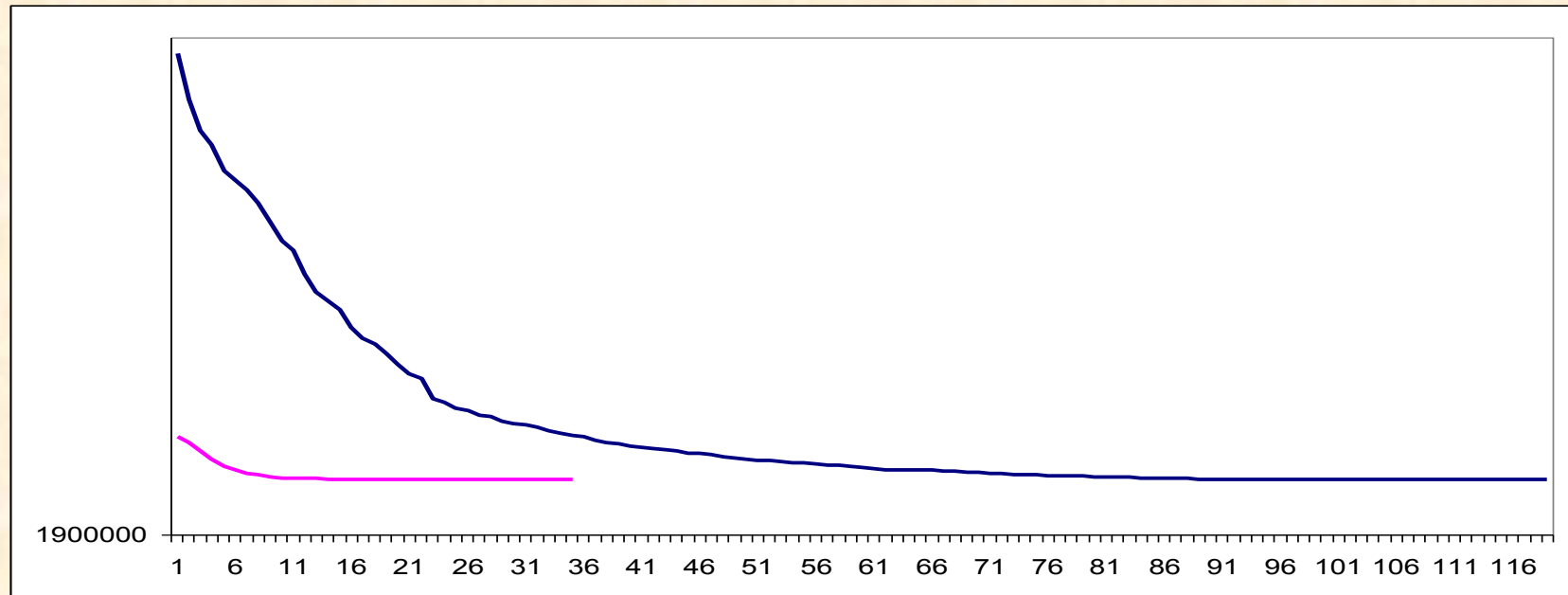
- Master problem requires more than 75% of total cpu time

# IMPACT OF PERFECT DUAL INFORMATION ON CG

<b>Problem</b>			<b>cpu</b>	<b># CG</b>	<b># SP</b>	<b># MP</b>
<b>R800 (4)</b>	<b>Opt sol</b>	<b>Init sol</b>	<b>total</b>	<b>iter</b>	<b>cols</b>	<b>itr</b>
<b>standard</b>	1915589.5	800000000	<b>4178.4</b>	<b>509</b>	37579	926161
<b>dual boxes</b>						
<b>100</b>		2035590.5	<b>835.5</b>	<b>119</b>	9368	279155
<b>10</b>		1927590.5	<b>117.9</b>	<b>35</b>	2789	40599
<b>1</b>		1916790.5	<b>52.0</b>	<b>20</b>	1430	8744
<b>0.1</b>		1915710.5	<b>47.5</b>	<b>19</b>	1333	8630
<b>0.01</b>	1915589.1	1915602.5	<b>37.3</b>	<b>17</b>	1145	6288

# IMPACT OF PERFECT DUAL INFORMATION ON CG

<i>Problem</i>			<i>cpu</i>	<i># CG</i>	<i># SP</i>	<i># MP</i>
<i>R800 (4)</i>	Opt sol	Init sol	<i>total</i>	<i>iter</i>	<i>cols</i>	<i>itr</i>
<i>standard</i>	1915589.5	800000000	<i>4178.4</i>	<i>509</i>	37579	926161
dual boxes						
100		2035590.5	<i>835.5</i>	<i>119</i>	9368	279155
10		1927590.5	<i>117.9</i>	<i>35</i>	2789	40599





## DUAL-OPTIMAL INEQUALITIES

Valério de Carvalho, **Using extra dual cuts to accelerate column generation for the Cutting Stock problem**, *Inform Journal on Computing* (2004).

- Small items ( $i=1, \dots, m$ ) are ranked:

$$l_1 > l_2 > l_3 > \dots \Rightarrow \pi_1 \geq \pi_2 \geq \pi_3 \geq \dots$$

- Additionally:

$$l_i \geq l_j + l_k \Rightarrow \pi_i \geq \pi_j + \pi_k$$

*At most  $2m$  dual constraints (or primal columns) inserted a priori.*

# DUAL-OPTIMAL INEQUALITIES / PRIMAL COLUMNS

Generated cutting patterns			<i>a priori columns</i>						
			1				1		
			-1	1				1	
				-1	1				
					-1				
						...			
							-1		
							-1	-1	
								-1	
									...

*Total cpu time reduced by 40%.*

## TRIPLETS (501 ITEMS)

*EACH ROLL IS CUT INTO EXACTLY TWICE WITHOUT WASTE*

Standard CG	124.2 iterations
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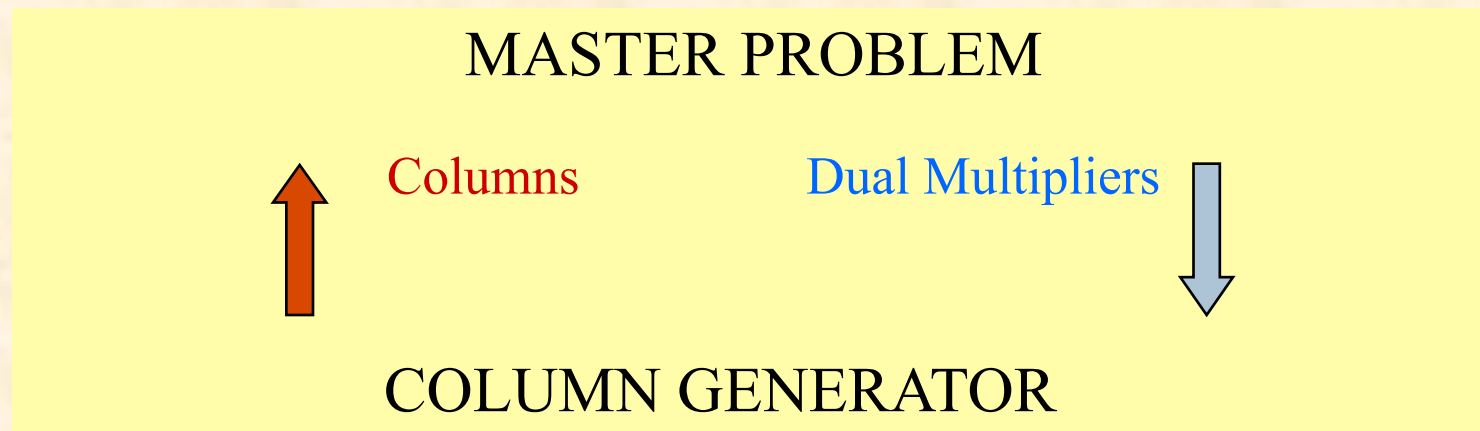
$\pi_1 \geq \pi_2 \geq \pi_3 \geq \dots$	113.3 iterations
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**Perfect dual  
information**

$\pi_i = \ell_i / L, \quad i = 1, \dots, m$	12.2 iterations
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( Average over 10 problems )

# LP COLUMN GENERATION



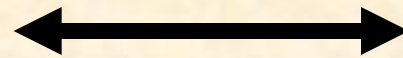
*Stopping rule?*

**Optimality of Dual Multipliers**

## DUAL BOXES

**Primal**

$$\begin{array}{ll}\min & cx \\ & Ax = b \\ & x \geq 0\end{array}$$



$$\begin{array}{ll}\max & b\pi \\ & \pi A \leq c\end{array}$$

**Dual**

$$\begin{array}{ll}\min & cx - d_1 y_1 + d_2 y_2 \\ & Ax - y_1 + y_2 = b \\ & x \geq 0 \\ & y_1 \geq 0, y_2 \geq 0\end{array}$$

**Relaxed Primal**



$$\begin{array}{ll}\max & b\pi \\ & \pi A \leq c \\ & d_1 \leq \pi \leq d_2\end{array}$$

**Restricted Dual**

Surplus & Slack Variables

# DEGENERACY & PERTURBATION

**Primal**

$$\begin{array}{ll}\min & cx \\ & Ax = b \\ & x \geq 0\end{array}$$

*Alternative*

*Perturbed Primal*

$$\begin{array}{ll}\min & cx \\ & Ax - y_1 + y_2 = b \\ & x \geq 0 \\ & 0 \leq y_1 \leq \varepsilon_1, 0 \leq y_2 \leq \varepsilon_2\end{array}$$

*Relaxed Primal*

Surplus & Slack Variables

## PERTURBATION & DUAL BOXES

**Primal**

$$\begin{array}{ll}\min & cx \\ & Ax = b \\ & x \geq 0\end{array}$$

$$\begin{array}{ll}\min & cx \\ & Ax - y_1 + y_2 = b \\ & x \geq 0 \\ & 0 \leq y_1 \leq \varepsilon_1, 0 \leq y_2 \leq \varepsilon_2\end{array}$$

**Relaxed Primal**

$$\begin{array}{ll}\min & cx - d_1 y_1 + d_2 y_2 \\ & Ax - y_1 + y_2 = b \\ & x \geq 0 \\ & y_1 \geq 0, y_2 \geq 0\end{array}$$

**Relaxed Primal**

## STABILIZED PRIMAL PROBLEM

$$\begin{aligned} \min \quad & cx \\ \text{subject to} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

**Primal**  
**P**

$$\begin{aligned} \min \quad & cx - d_1 y_1 + d_2 y_2 \\ \text{subject to} \quad & Ax - y_1 + y_2 = b \\ & x \geq 0 \\ & 0 \leq y_1 \leq \varepsilon_1, \quad 0 \leq y_2 \leq \varepsilon_2 \end{aligned}$$

**Stabilized Primal**

**SP**



## STABILIZED PRIMAL & DUAL PROBLEMS

$$\min \quad cx - d_1 y_1 + d_2 y_2$$

$$Ax - y_1 + y_2 = b \quad \pi$$

$$y_1 \leq \varepsilon_1 \quad -\omega_1 \leq 0$$

$$y_2 \leq \varepsilon_2 \quad -\omega_2 \leq 0$$

$$x \geq 0, y_1 \geq 0, y_2 \geq 0$$

**Stabilized Primal SP**

$$\max \quad b\pi$$

$$\pi A \leq c$$

**Dual D**

$$\max \quad b\pi - \omega_1 \varepsilon_1 - \omega_2 \varepsilon_2$$

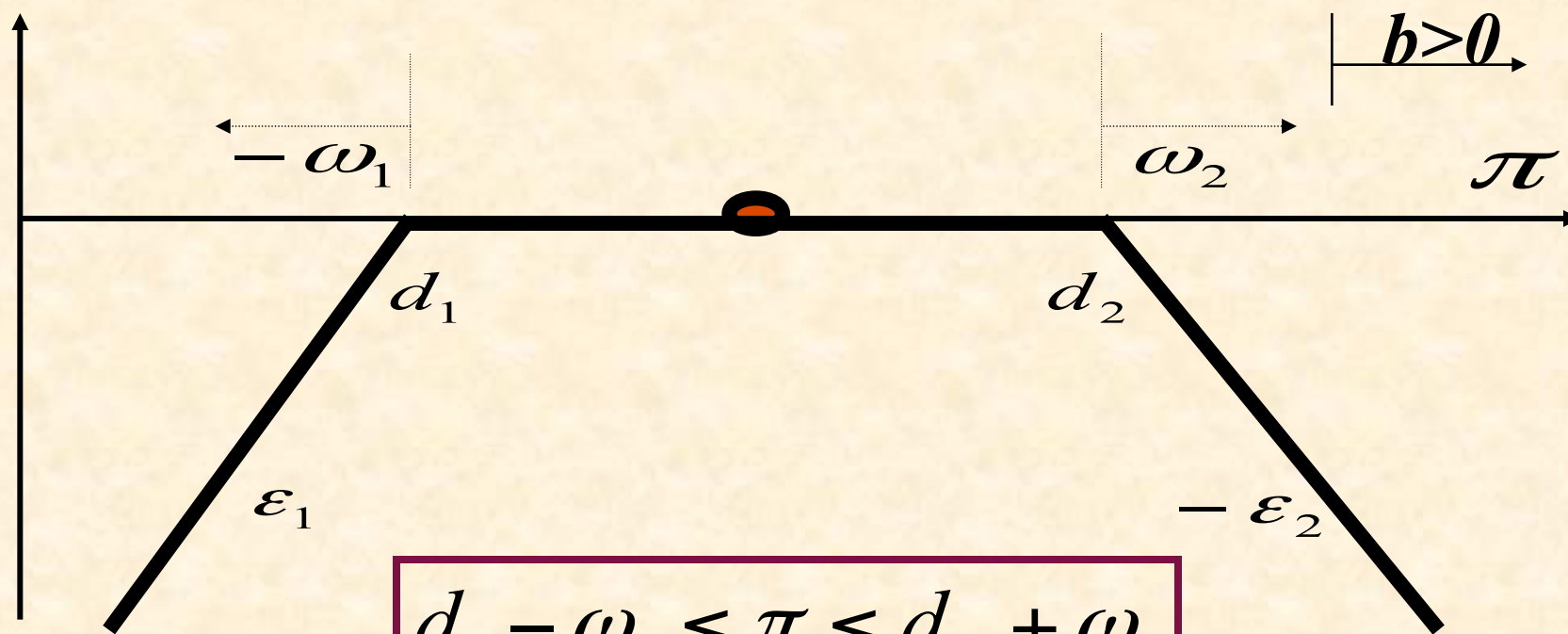
$$\pi A \leq c$$

$$d_1 - \omega_1 \leq \pi \leq d_2 + \omega_2$$

$$\omega_1 \geq 0, \omega_2 \geq 0$$

**Stabilized Dual SD**

## INTERPRETATION IN DUAL SPACE



$$d_1 - \omega_1 \leq \pi \leq d_2 + \omega_2$$

$$\omega_1 \geq 0, \omega_2 \geq 0$$

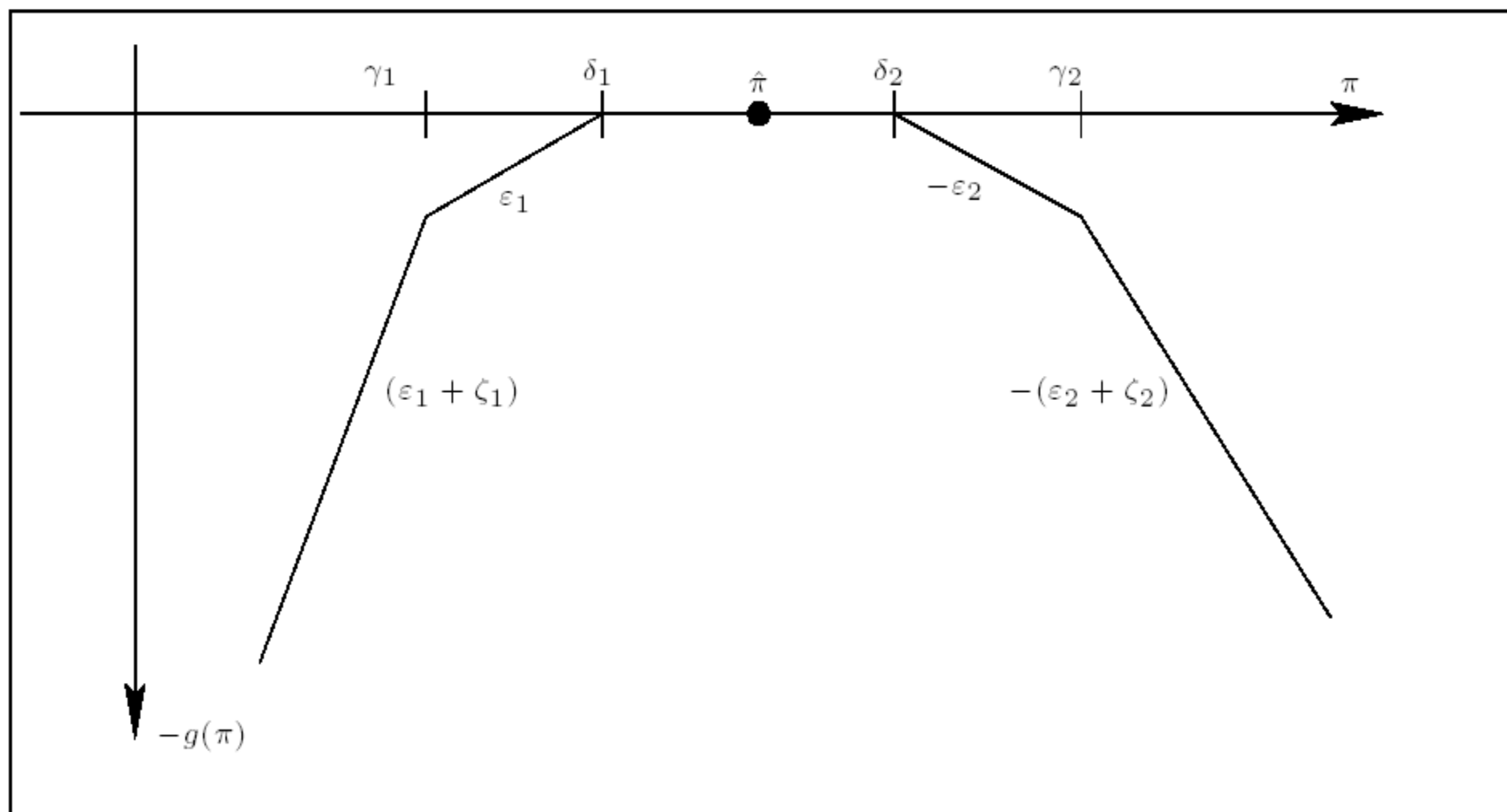


Figure 1: 5-piecewise linear dual penalty function

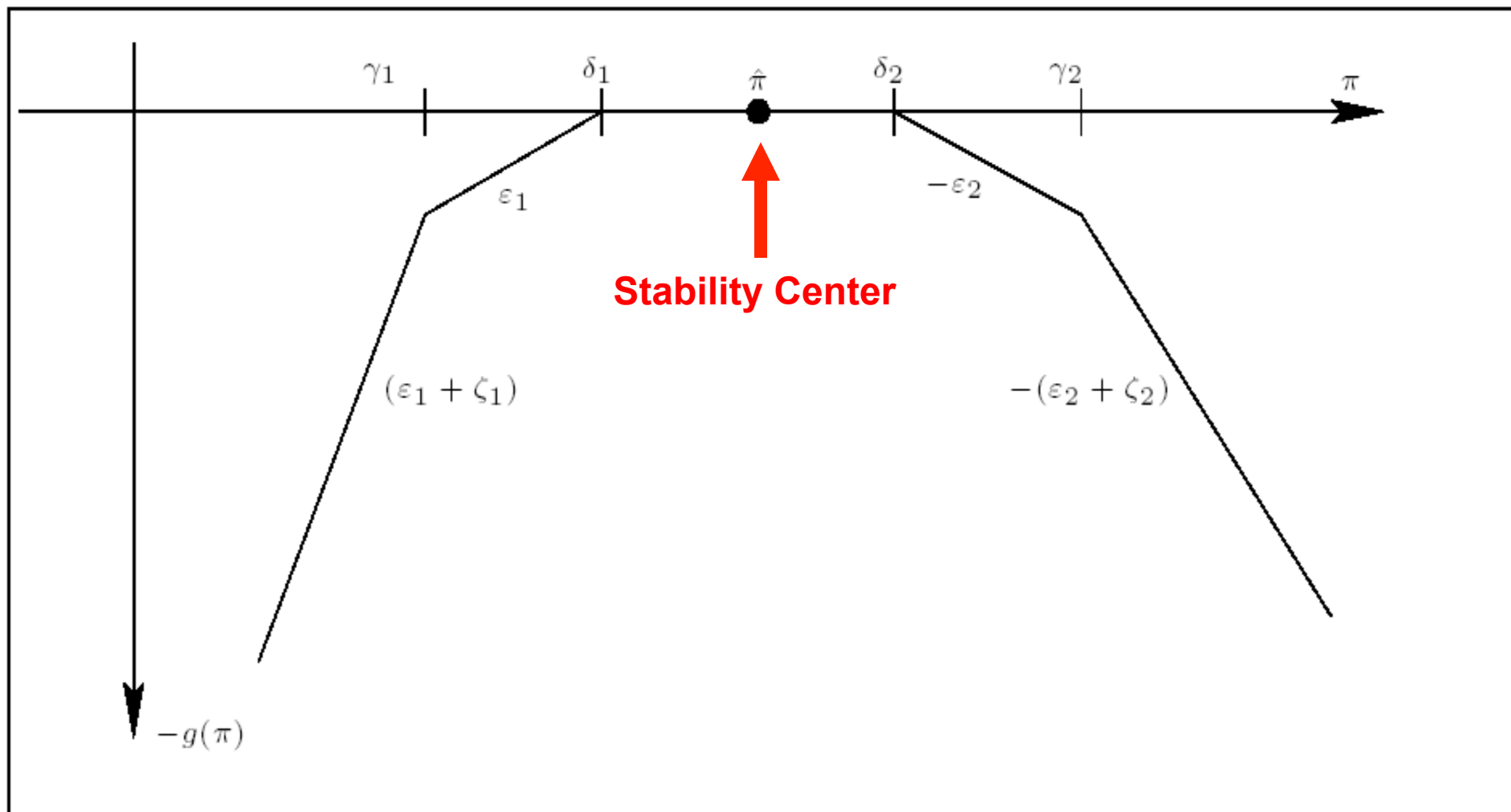


Figure 1: 5-piecewise linear dual penalty function

# MDVSP: SOLUTION STRATEGY

## 1. Preprocessing

- **Assignment problem**: Single depot approximation
  - Lower bound  $Z_L$
  - Optimal number of vehicles  $\nu^*$
- **Transportation problem** to derive a primal integer solution
  - Upper bound  $Z_U$
- **Arc elimination** based reduced cost greater than  $Z_U - Z_L$
- **Estimation of dual multipliers** for all tasks

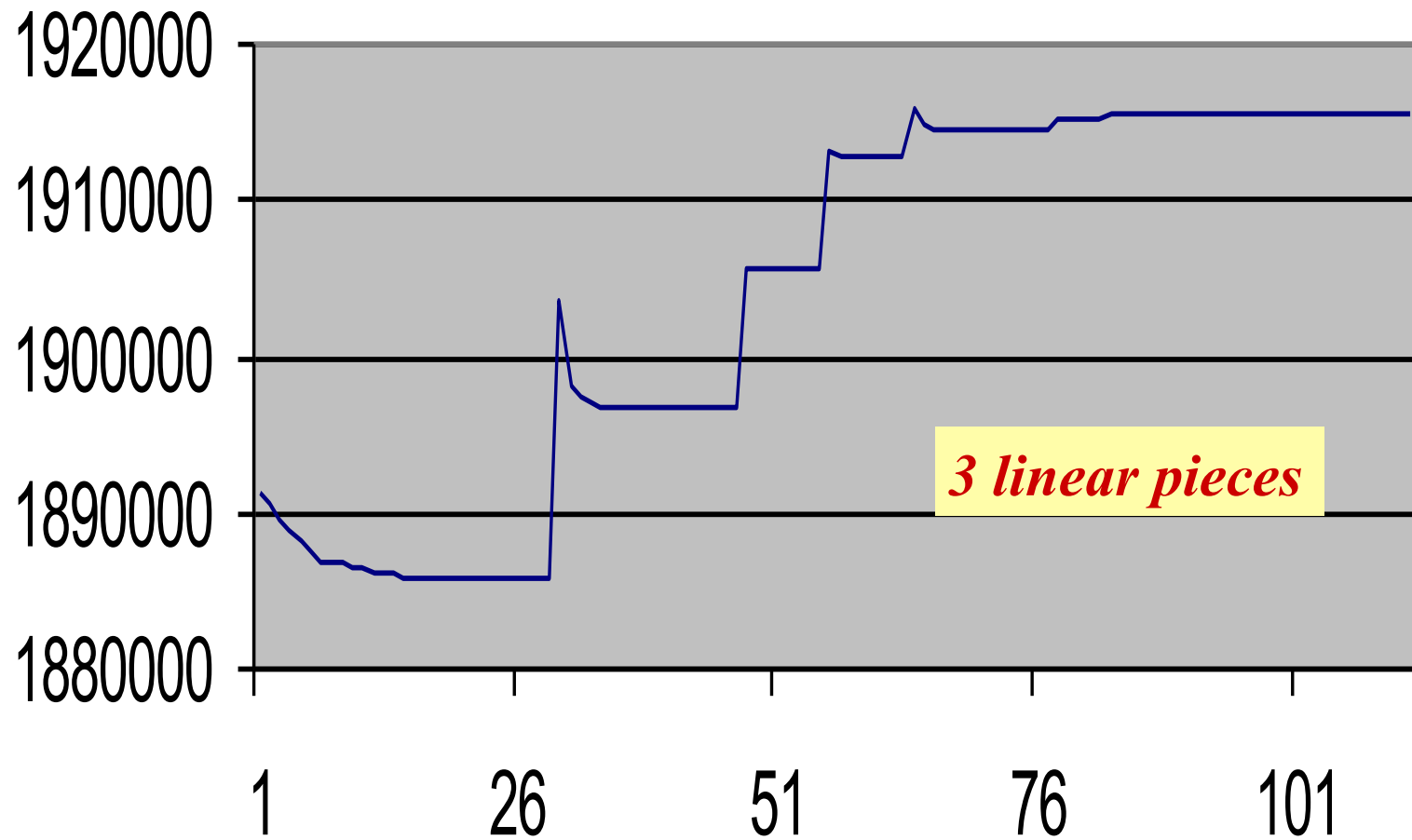
## 2. Stabilization procedure

## STABILIZATION PROCEDURE FOR PROBLEM P

- Initialize approximation problems SP & SD
  - Stability center
  - Trust region without penalties
  - Penalties outside the trust region (3 to 5 pieces)
- Solve stabilized problems SP & SD until P is feasible
  - Otherwise update problems SP & SD
    - Stability center, trust region and penalties

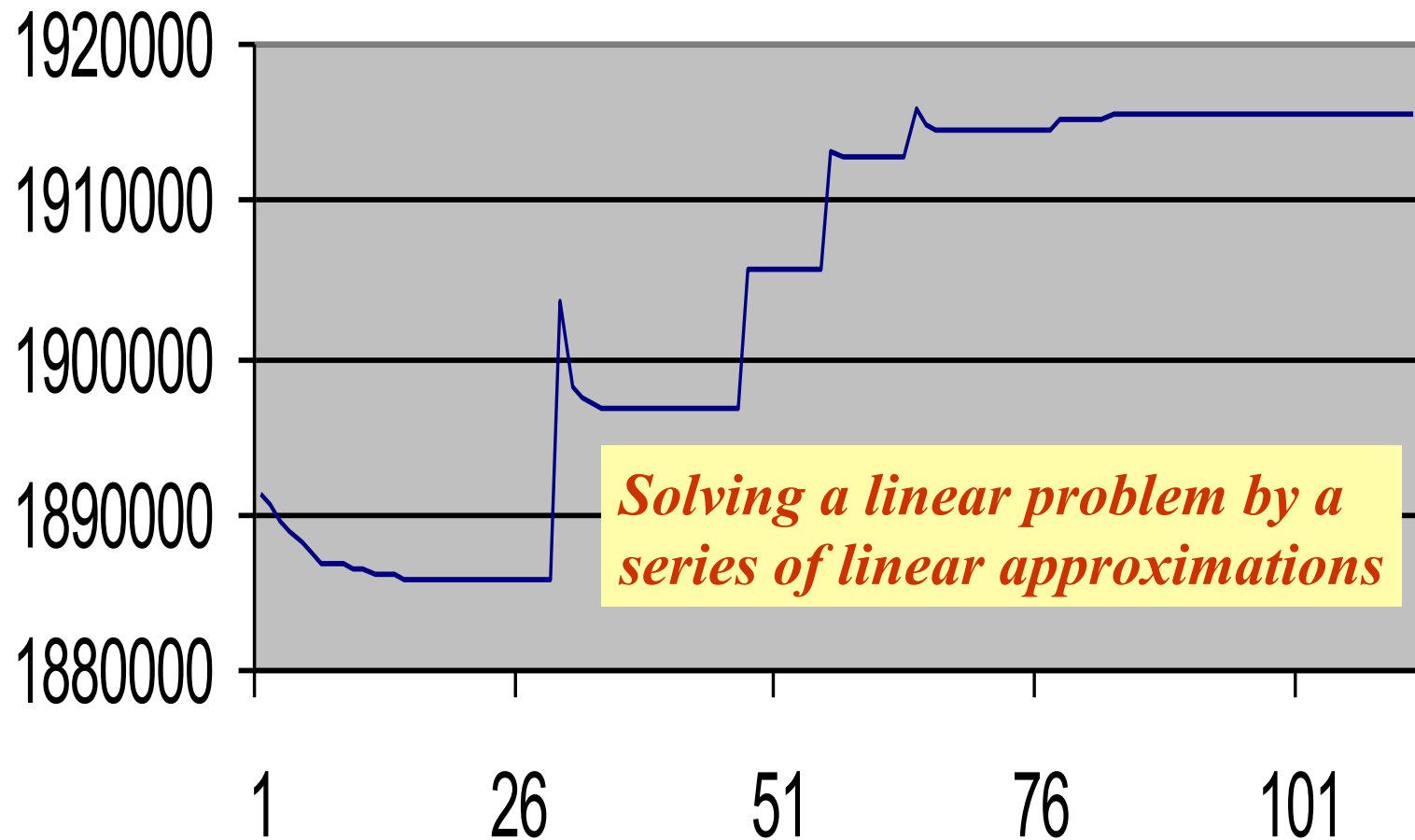
**Problem R800 (4)**

# Objective Function



**Problem R800 (4)**

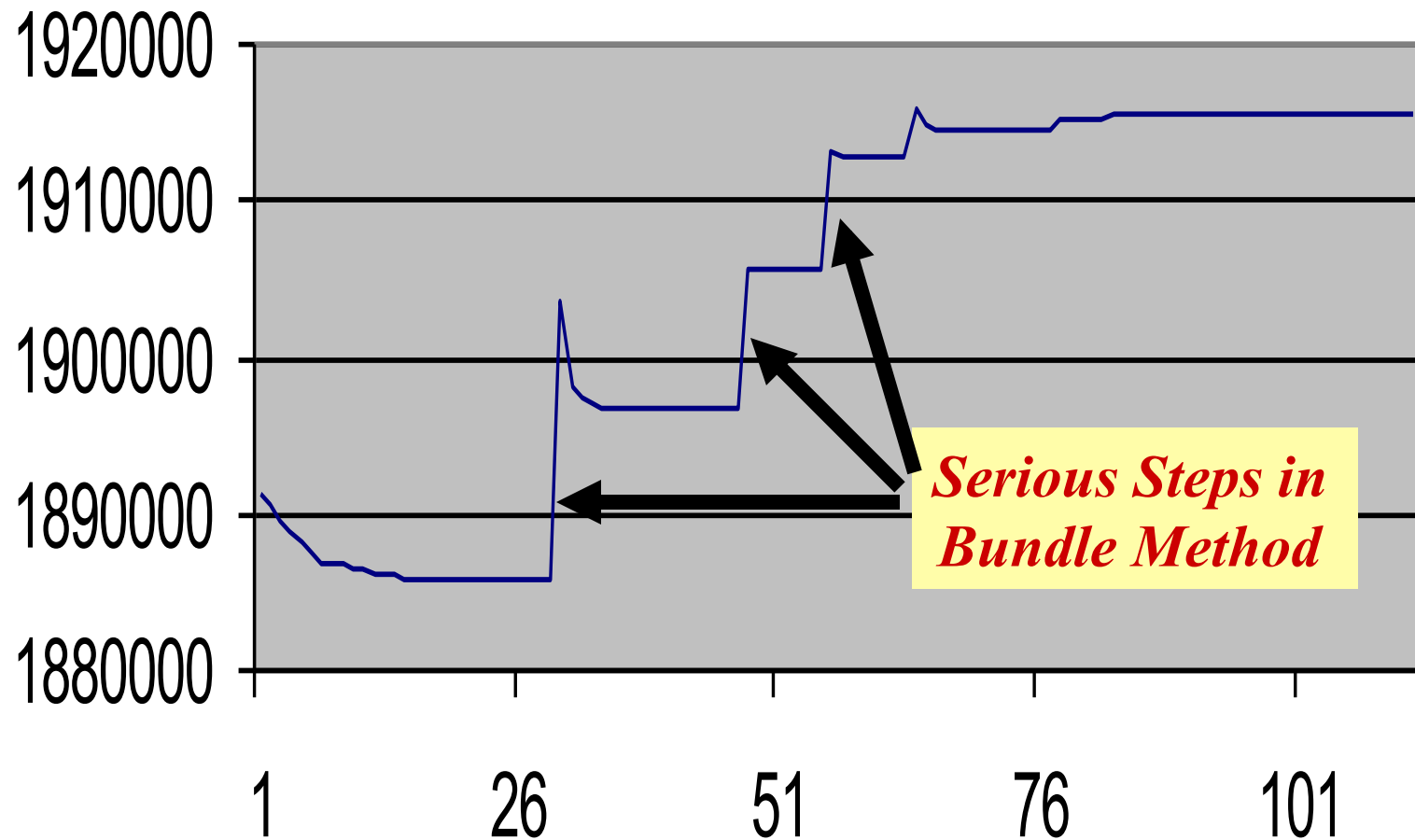
## Objective Function



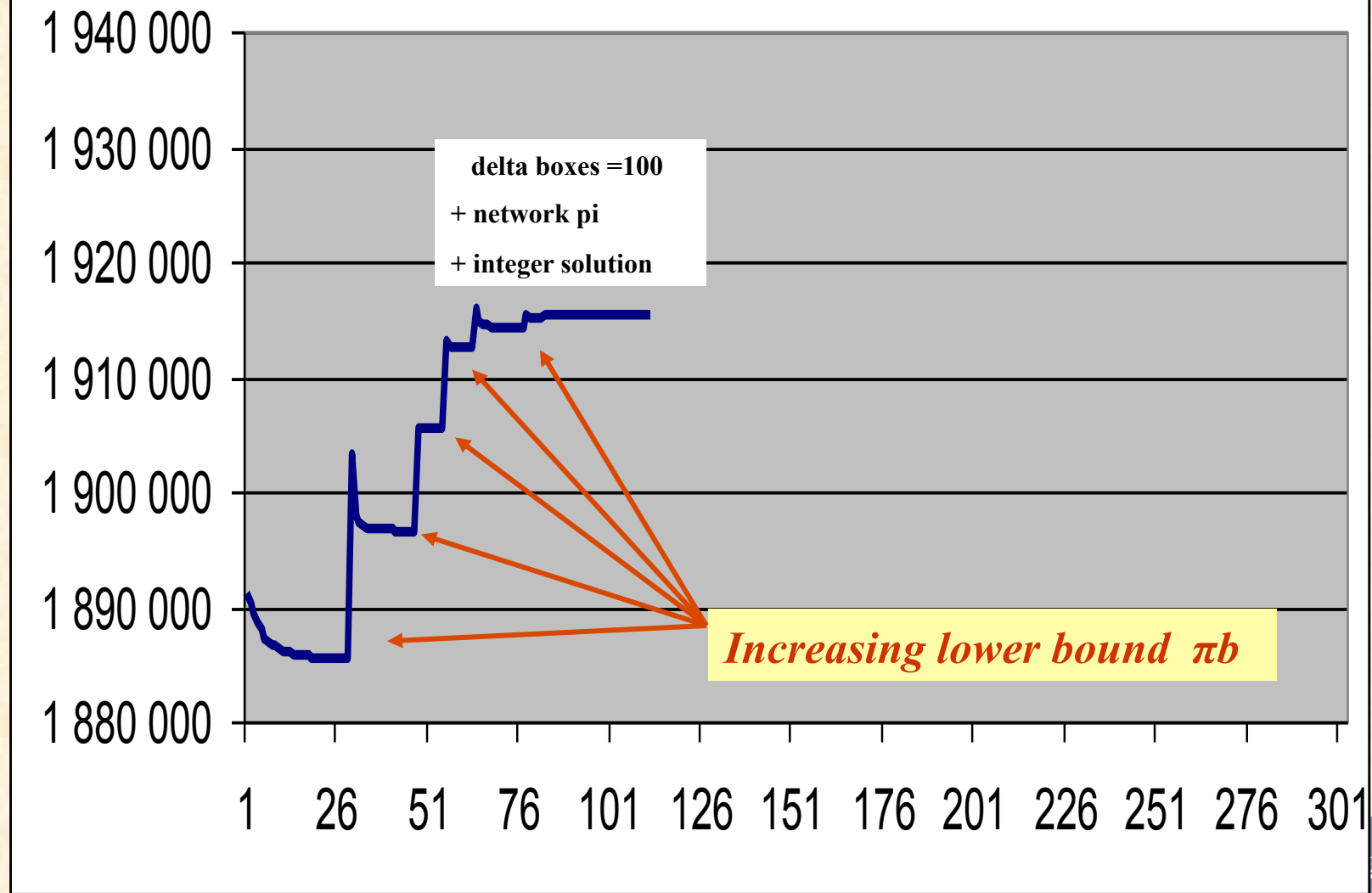


**Problem R800 (4)**

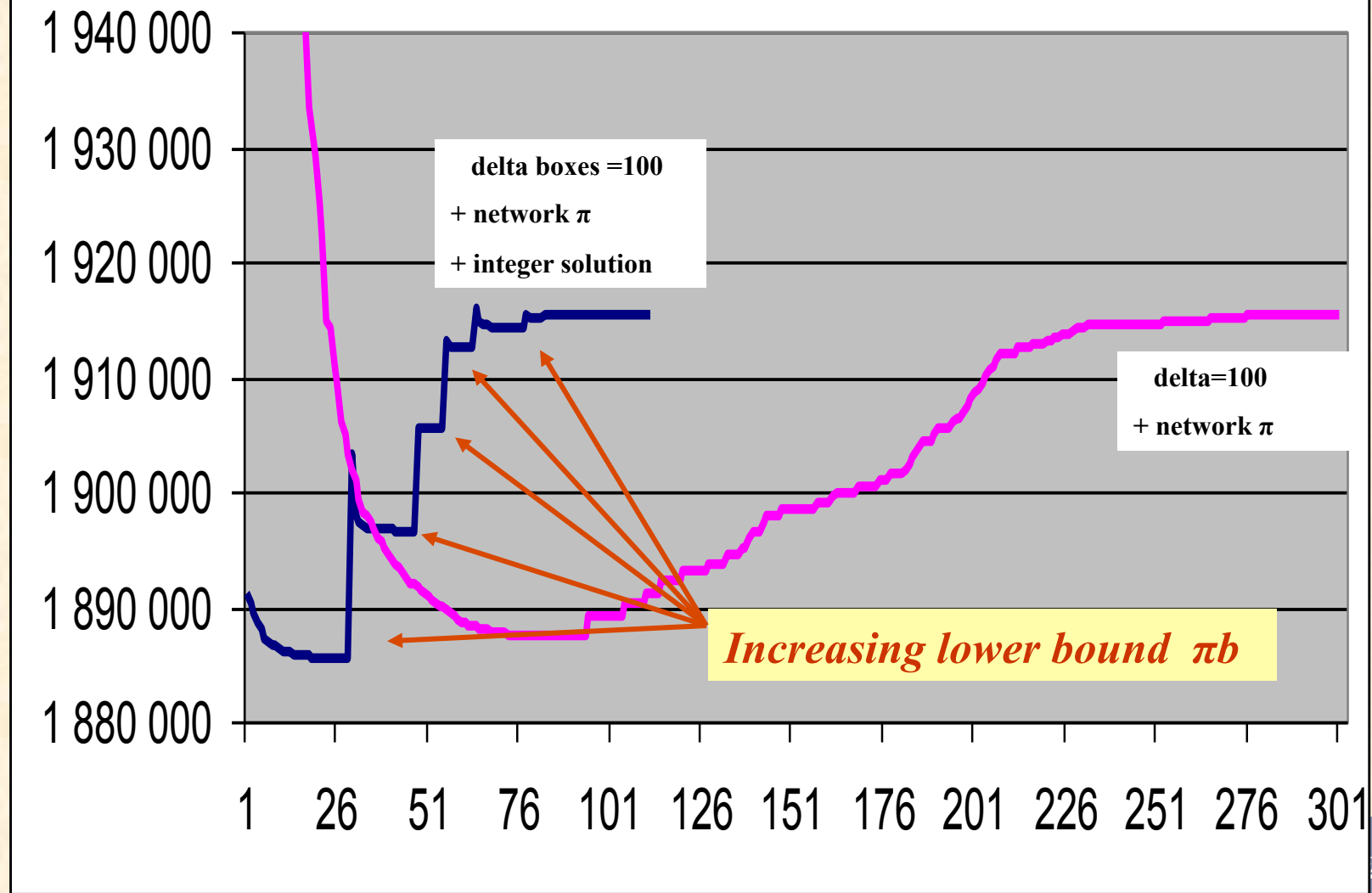
## Objective Function



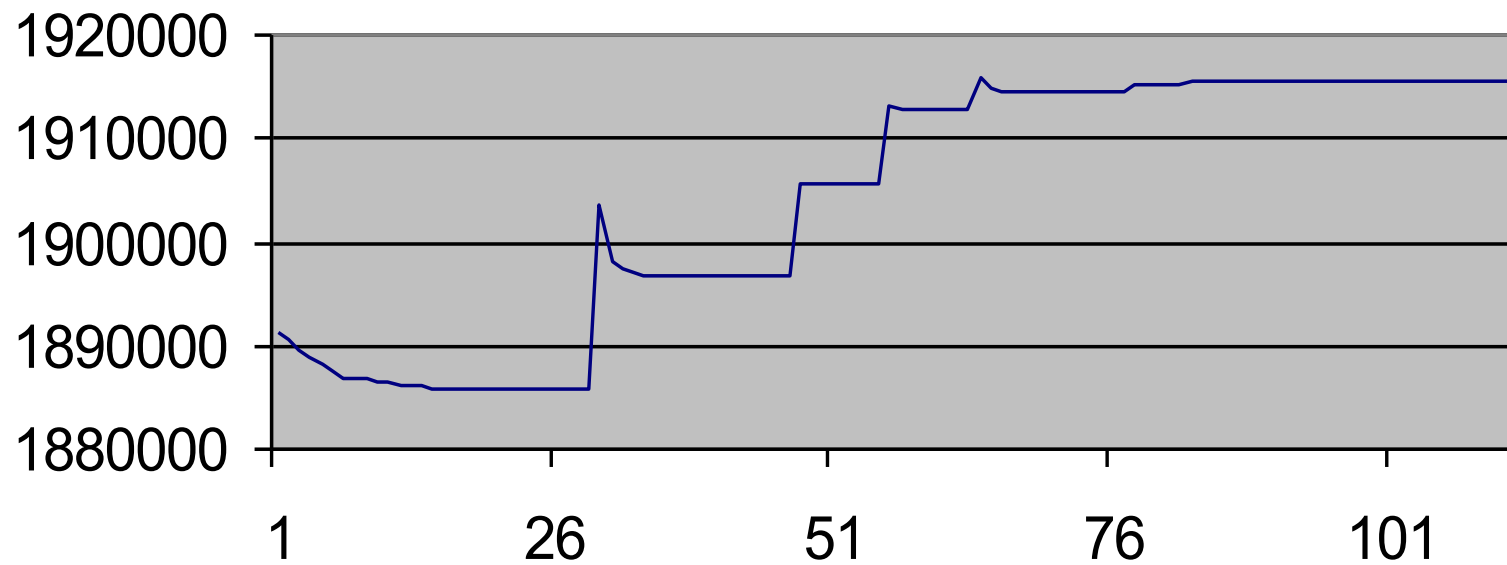
# Objective Function



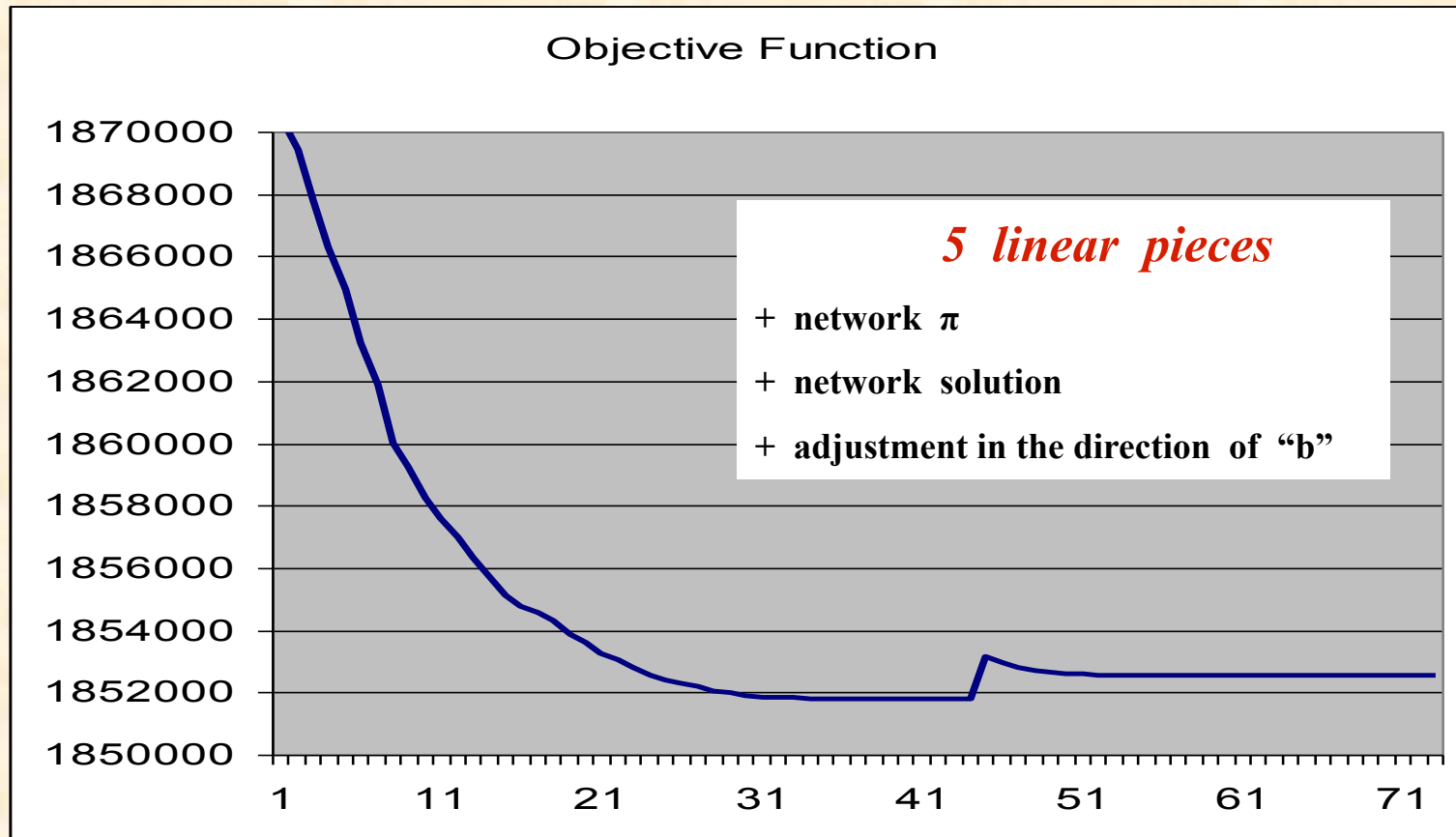
# Objective Function



# Objective Function



<b>Problem</b>				cpu	cpu	# CG	# SP	# MP
<b>R800 (4)</b>	Opt sol	Init sol	<i>cpu tot</i>	mp	sp	iter	cols	itr
<b>standard</b>	1915589.5	8000000000	<b>4178.4</b>	3149.2	1029.2	<b>509</b>	37579	926161
delta = 100		2035590.5	<b>835.5</b>	609.1	226.4	<b>119</b>	9368	279155
network pi		2014429.8	<b>1097.1</b>	518.5	578.6	<b>301</b>	10105	324959
network pi + network sol		1891386.0	<b>439.2</b>	216.2	223.0	<b>112</b>	4749	153420
		% reduction	<b>89.5</b>	<b>93.1</b>	<b>78.3</b>	<b>78.0</b>	<b>87.4</b>	<b>83.4</b>
			<b>9.5 times faster</b>					



Opt sol	Init sol	<i>cpu tot</i>	cpu mp	cpu sp	# CG iter	# SP cols	# MP itr
1852571.5	800000000	<b>3562</b>	886	2676	<b>422</b>	33280	672726
	1870487.8	<b>241</b>	110	131	<b>63</b>		
	% reduction	<b>93.2</b>	<b>87.6</b>	<b>95.1</b>	<b>85.1</b>		
	<b>14.8 times faster</b>						

## PERFECT DUAL INFORMATION: APPLICATIONS

- Useful in the context of **Lagrangian Relaxation** to recover primal feasibility (fractional)

- **Useful to perform *Crossover* from an interior point solution to an extreme point solution**

- H. Ben Amor, J. Desrosiers, and F. Soumis

Recovering an optimal LP basis from an optimal dual solution,  
*Operations Research Letters* (2006)

## CROSSOVER: LARGE CREW ROSTERING INSTANCES

	Constraints	Variables
pb1	12 351	126 326
pb2	12 310	129 046
pb3	13 190	146 013
pb4	13 433	151 654
pb5	13 550	162 914
pb6	13 451	156 839
pb7	13 254	148 025
pb8	13 424	154 205
pb9	13 598	163 707
pb10	13 310	155 313

- CPLEX7.5 **primal simplex algorithm fails** to solve any of these 10 problems in less than **18000 seconds** on a Enterprise 10000 solaris2.7 400MHZ machine (64 CPU, RAM=64G).
- The **dual simplex** needed less than 5000 seconds for two problems and failed to solve one within 18,000 seconds. The seven others require between 8,000 and 13,000 seconds.
- The problems are rather better solved by combining the CPLEX **Barrier algorithm with a primal or dual crossover** method based on problem simplifications followed by a primal or dual simplex algorithm, as proposed in Bixby and Saltzman (2002).
- For the interior point algorithm, we used values  **$10^{-8}$  and  $10^{-10}$**  for the optimality parameter; in both cases, all problems were **solved in less than 1900 seconds**.



## Crossover: CPLEX vs Box Methods (sec.)

Barrier  $10^{-8}$

Crossover

pb1	188	72
pb2	20	1183
pb3	327	435
pb4	8166	2568
pb5	59	2645
pb6	***	1797
pb7	270	1092
pb8	1036	1876
pb9	78	2811
pb10	37	3011

<b>Avg</b>	1834.7	1749
<b>StdD</b>	3350.1	1027.6
<b>Min</b>	20	72
<b>Max</b>	***	3011

## Crossover: CPLEX vs Box Methods (sec.)

Barrier  $10^{-8}$

Crossover	CrPrimal	CrDual	BoxPrimal	BoxDual	
pb1	188	72	80	132	
pb2	20	1183	98	208	
pb3	327	435	189	171	
pb4	8166	2568	249	321	
pb5	59	2645	313	322	
pb6	***	1797	242	229	
pb7	270	1092	162	148	
pb8	1036	1876	205	235	
pb9	78	2811	205	265	
pb10	37	3011	290	426	
<b>Avg</b>	1834.7	1749	203.3	245.7	
<b>StdD</b>	3350.1	1027.6	75.5	90.8	
<b>Min</b>	20	72	80	132	
<b>Max</b>	***	3011	313	426	


## Crossover: CPLEX vs Box Methods (sec.)

*Barrier  $10^{-8}$*

<b>Crossover</b>	<b>CrPrimal</b>	<b>CrDual</b>	<b>BoxPrimal</b>	<b>BoxDual</b>	<b>BoxCrPrimal</b>
pb1	188	72	80	132	<b>9</b>
pb2	20	1183	98	208	<b>13</b>
pb3	327	435	189	171	<b>26</b>
pb4	8166	2568	249	321	<b>35</b>
pb5	<b>59</b>	2645	313	322	<b>45</b>
pb6	***	1797	242	229	<b>86</b>
pb7	270	1092	162	148	<b>22</b>
pb8	1036	1876	205	235	<b>20</b>
pb9	78	2811	205	265	<b>43</b>
pb10	<b>37</b>	3011	290	426	<b>30</b>
<b>Avg</b>	1834.7	1749	203.3	245.7	<b>32.9</b>
<b>StdD</b>	3350.1	1027.6	75.5	90.8	<b>22.1</b>
<b>Min</b>	20	72	80	132	<b>9</b>
<b>Max</b>	***	3011	313	426	<b>86</b>

## Crossover: CPLEX vs Box Methods (sec.)

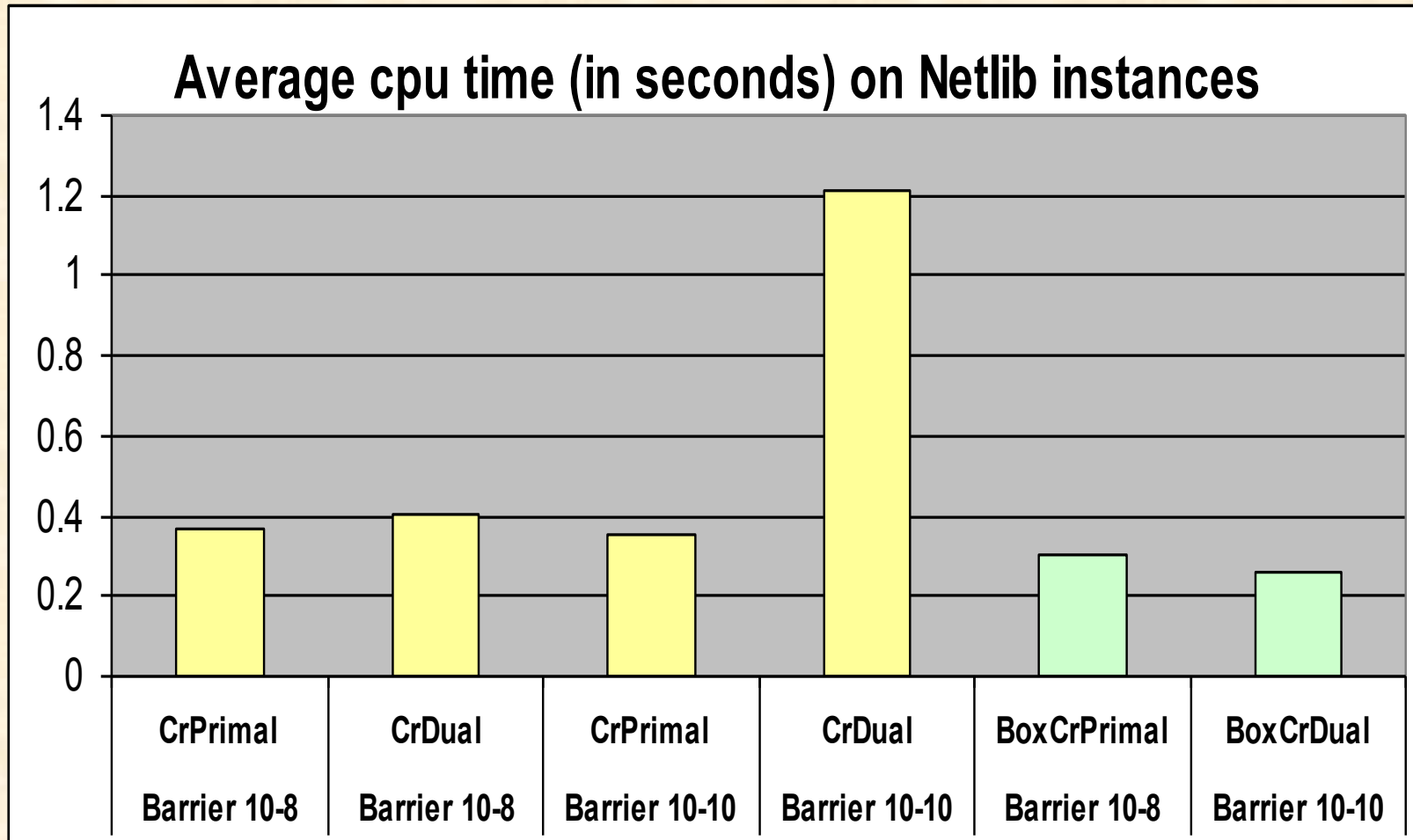
*Barrier  $10^{-10}$*



Crossover	CrPrimal	CrDual	BoxPrimal	BoxDFual	BoxCrPrimal
pb1	101	15	79	129	7
pb2	***	27	85	218	7
pb3	605	69	146	124	43
pb4	232	2121	192	162	16
pb5	89	2819	231	234	25
pb6	72	2328	201	175	22
pb7	192	1025	147	152	13
pb8	1405	1407	168	198	15
pb9	7190	1957	238	308	23
pb10	4159	2832	218	261	15
<b>Avg</b>	2123.5	1460	170.5	196.1	18.6
<b>StdD</b>	2944.5	1127.2	56.5	59.5	10.5
<b>Min</b>	72	15	79	124	7
<b>Max</b>	***	2832	238	308	43

*Efficient... and so simple!*

## Crossover: CPLEX vs Box Methods (sec.)



## Crossover: CPLEX vs Box Methods (sec.)

Crossover	Barrier 10 <sup>-8</sup>	Barrier 10 <sup>-8</sup>	Barrier 10 <sup>-10</sup>	Barrier 10 <sup>-10</sup>	Barrier 10 <sup>-8</sup>	Barrier 10 <sup>-10</sup>
	CrPrimal	CrDual	CrPrimal	CrDual	BoxCrPrimal	BoxCrDual
Netlib 500 ( 59)	0.05	0.14	0.05	0.14	0.04	0.05
Netlib 1000 (14)	0.12	0.13	0.12	0.14	0.14	0.13
Netlib 2000 ( 9)	0.24	0.36	0.24	0.26	0.31	0.27
Netlib 3000 ( 1)	0.10	0.10	0.10	0.10	0.10	0.10
Netlib 4000 ( 1)	6.40	2.00	5.10	84.90	2.60	0.80
Kennington 5000 ( 9)	0.41	0.43	0.50	0.40	0.36	0.34
Kennington 10000 ( 4)	1.70	1.75	1.63	1.73	1.20	1.20
Kennington 30000 ( 5)	2.58	2.78	2.54	2.62	2.52	1.94
Avg	0.367	0.402	0.357	1.212	0.305	0.257
StdD	0.865	0.656	0.77	8.433	0.612	0.455