



New Formulations and Solution Approaches for Train *Eco-driving* Problems [https://ssrn.com/abstract=4928445] (Join work with Zhuang Xiao & Edward Chung) Hongbo Ye

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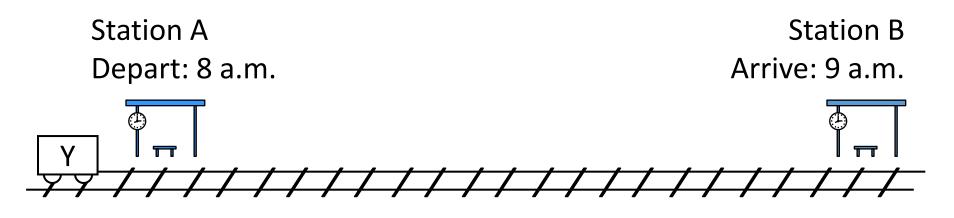
_ Opening Minds ∙ Shaping the Future 啟迪思維 ∙ 成就未來

Outline



- What is eco-driving
- Formulation of eco-driving problem
- Existing solution methods
- Our solution method
 - Reformulation, relaxation, and valid inequalities
- Possible directions for future research



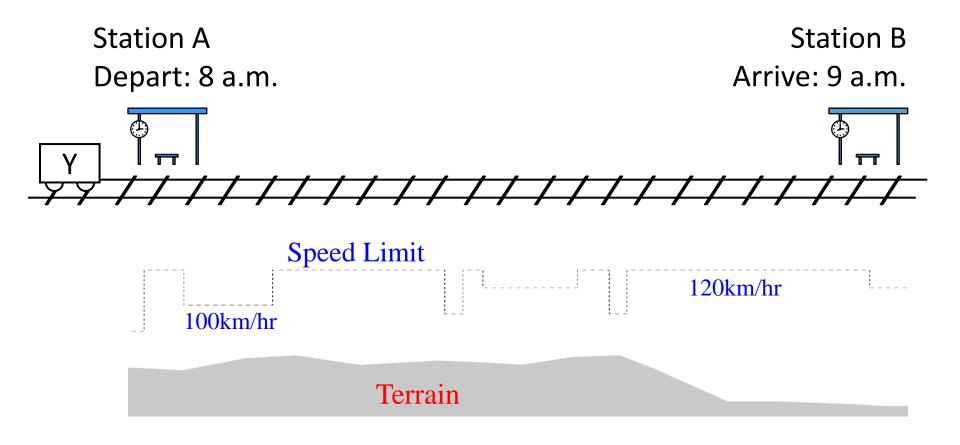


How to drive the train (i.e., control the power and brake) so as to

- Arrive at the station on time
- Consume minimum amount of energy

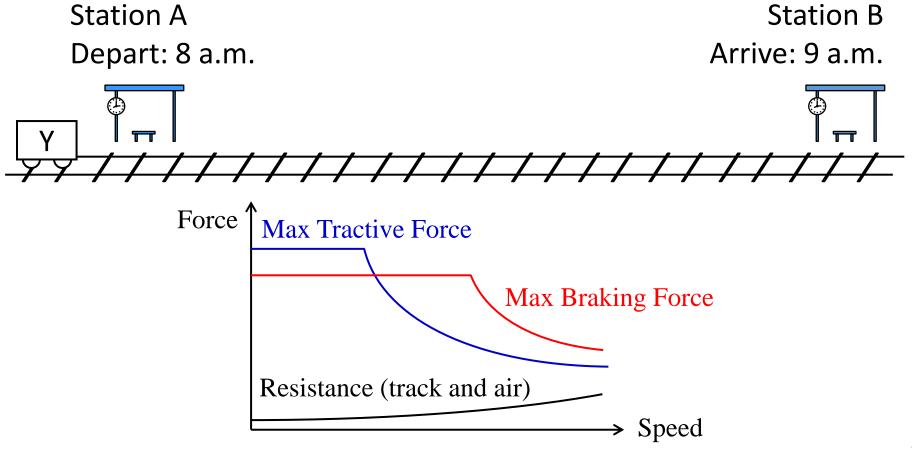
Factors Considered in Eco-driving: Track Condition





Factors Considered in Eco-driving: Train Characteristics

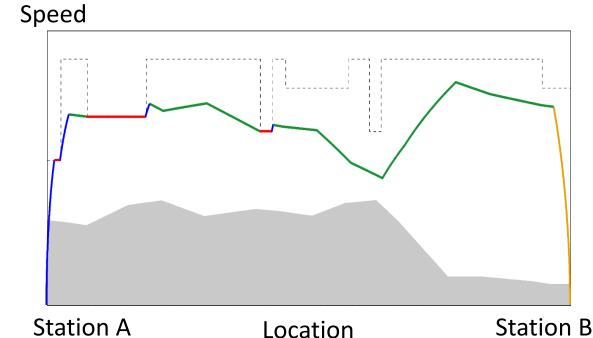






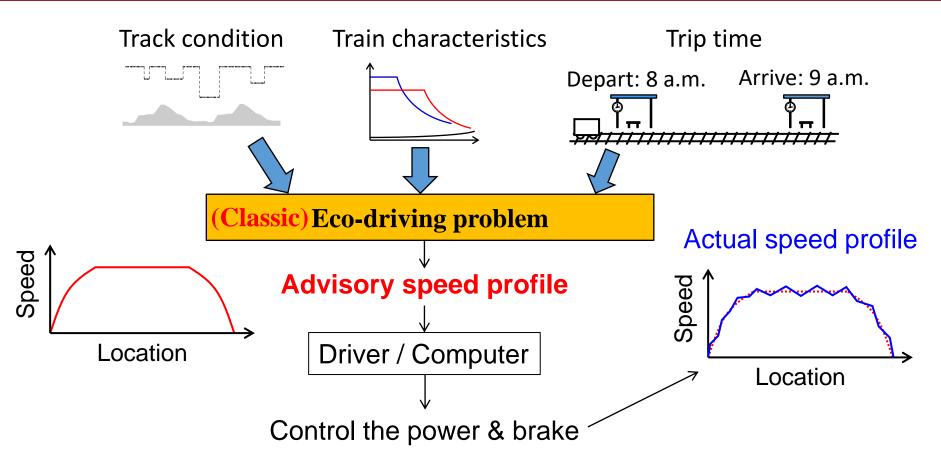
Solving the eco-driving problem, we can get a speed profile that

- Guarantees the train reaching next station on time,
- Minimizes the energy consumption



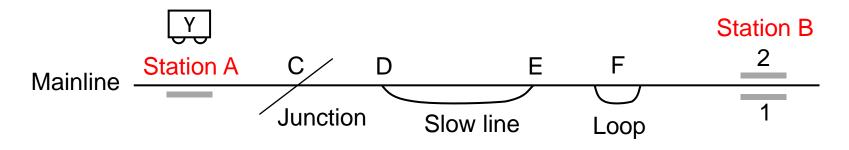
Speed Tracking





Other Factors Considered in Eco-driving: Schedule

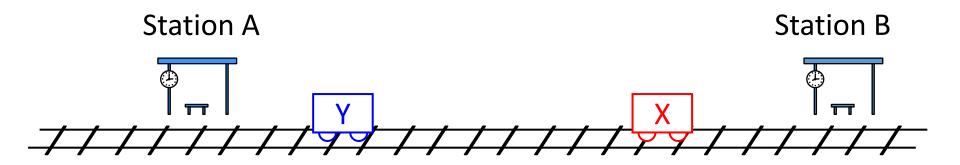




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- What passengers see
 - Train YArrDepStation A-8:00Station B
(platform 2)9:00-
- What train operators/drivers see

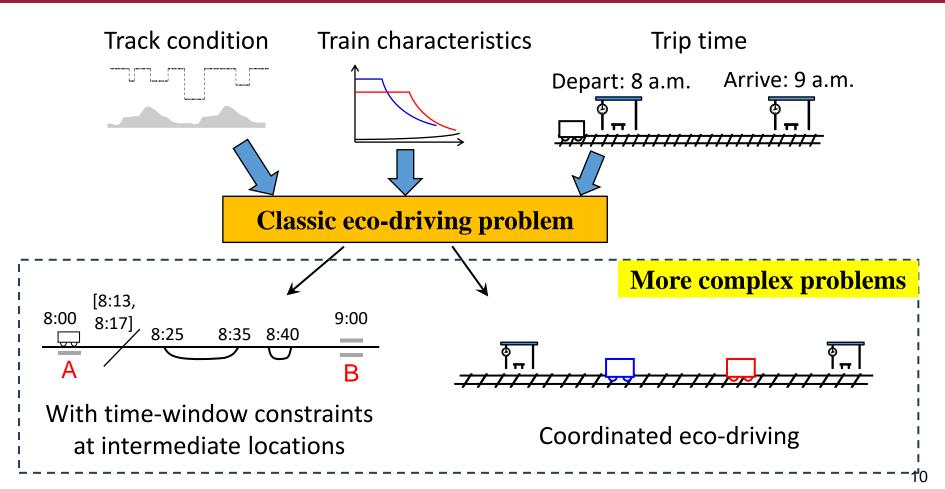




- Train Y must not **collide** with train X
- The eco-driving of trains X and Y can be coordinated

Type of Eco-driving Problems







$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds$$
s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg \sin \alpha(s)}{m \cdot v(s)}$$

$$\frac{dt(s)}{ds} = \frac{1}{v(s)}$$

$$\varepsilon \le v(s) \le \overline{V}(s)$$

$$F_{\text{min}} \le F(s) \le F_{\text{max}}$$

$$P_{\text{min}} \le F(s)v(s) \le P_{\text{max}}$$

$$t(S_f) - t(S_0) \le T$$

$$v(S_0) = V_0, v(S_f) = V_f$$



 $\min_{F(s):s\in[S_0,S_f]}\int_{S_0}^{S_f} \max\{F(s),\eta_{\text{reg}}F(s)\}\,\mathrm{d}s$ Total energy consumption of the trip s.t. $\frac{\mathrm{d}v(s)}{\mathrm{d}s} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg\sin\alpha(s)}{m \cdot v(s)}$ dt(s) = 1s: Location $\frac{1}{\mathrm{d}s} = \frac{1}{v(s)}$ F(s): Applied force at location s $\varepsilon \leq v(s) \leq \overline{V}(s)$ $\eta_{reg} \in [0,1)$: Proportion of braking energy reused $F_{\min} \leq F(s) \leq F_{\max}$ $P_{\min} \leq F(s)v(s) \leq P_{\max}S_0 \& S_f$: Trip start & end position $t(S_f) - t(S_0) \le T$ $v(S_0) = V_0, v(S_f) = V_f$



S

$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds$$
s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg \sin \alpha(s)}{m \cdot v(s)}$$

$$\frac{dt(s)}{ds} = \frac{1}{v(s)}$$

$$\varepsilon \le v(s) \le \overline{V}(s)$$

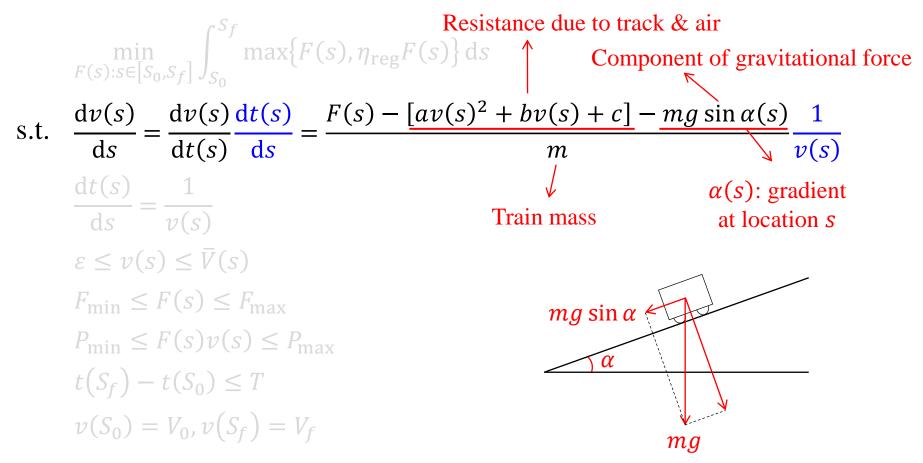
$$F_{\min} \le F(s) \le \overline{V}(s)$$

$$F_{\min} \le F(s) v(s) \le P_{\max}$$

$$t(S_f) - t(S_0) \le T$$

$$v(S_0) = V_0, v(S_f) = V_f$$







$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds$$

s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg \sin \alpha(s)}{m \cdot v(s)}$$

$$\frac{dt(s)}{ds} = \frac{1}{v(s)}$$

$$v(s): \text{ speed at location s}$$

$$\varepsilon \le v(s) \le \overline{V}(s) \quad \text{Speed limit constraint}$$

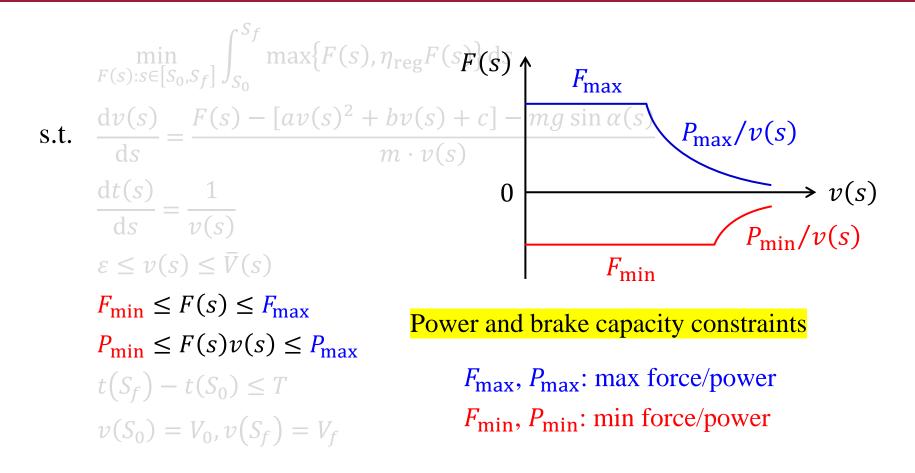
$$F_{\min} \le F(s) \le F_{\max}$$

$$P_{\min} \le F(s)v(s) \le P_{\max}$$

$$t(S_f) - t(S_0) \le T$$

$$v(S_0) = V_0, v(S_f) = V_f$$







 $\min_{F(s):s\in[S_0,S_f]}\int_{S_0}^{s_f}\max\{F(s),\eta_{\mathrm{reg}}F(s)\}\,\mathrm{d}s$ s.t. $\frac{\mathrm{d}v(s)}{\mathrm{d}s} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg\sin\alpha(s)}{m \cdot v(s)}$ dt(s) 1 ds = v(s) $\varepsilon \leq v(s) \leq \overline{V}(s)$ $t(S_f), t(S_0)$: Clock time at trip end & start $F_{\min} \leq F(s) \leq F_{\max}$ T: Scheduled travel time of the trip $P_{\min} \leq F(s)v(s) \leq P_{\max}$ $t(S_f) - t(S_0) \le T$ Total trip time constraint $\nu(S_0) = V_0, \nu(S_f) = V_f$



$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds$$

s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg\sin\alpha(s)}{m \cdot v(s)}$$
$$\frac{dt(s)}{ds} = \frac{1}{v(s)}$$
$$\varepsilon \le v(s) \le \overline{V}(s)$$
$$F_{\min} \le F(s) \le \overline{V}(s)$$
$$F_{\min} \le F(s) \le F_{\max}$$
$$P_{\min} \le F(s)v(s) \le P_{\max}$$
$$t(S_f) - t(S_0) \le T$$
$$v(S_0) = V_0, v(S_f) = V_f \text{ Required speed at trip start and}$$

Solution Methods of Eco-driving Problems



$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds$$
s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg \sin \alpha(s)}{m \cdot v(s)}$$

$$\frac{dt(s)}{ds} = \frac{1}{v(s)}$$
Two main types of solution approaches
$$\varepsilon \le v(s) \le \overline{V}(s)$$
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$$t(S_f) - t(S_0) \le T$$

$$v(S_0) = V_0, v(S_f) = V_f$$
(2) Direct method: via discretization



$$\min_{F(s):s\in[S_0,S_f]} \int_{S_0}^{S_f} \max\{F(s), \eta_{\text{reg}}F(s)\} ds \implies \min_{F_k} \sum_{k=0}^{N-1} \max\{F_k, \eta_{\text{reg}}F_k\} \Delta_k$$

s.t.
$$\frac{dv(s)}{ds} = \frac{F(s) - [av(s)^2 + bv(s) + c] - mg \sin \alpha(s)}{m \cdot v(s)}$$

$$\frac{dt(s)}{ds} = \frac{1}{v(s)} \implies \frac{t_k - t_{k-1}}{\Delta_k} = \frac{1}{v_k}$$

$$\varepsilon \le v(s) \le \overline{V}(s) \implies \varepsilon \le v_k \le \overline{V}_k$$

$$F_{\min} \le F(s) \le F_{\max}$$

$$v(S_0) = V_0, v(S_f) = V_f$$

Discretize the whole trip length into N
segments. Length of segment k is Δ_k
$$t_{k-1}$$

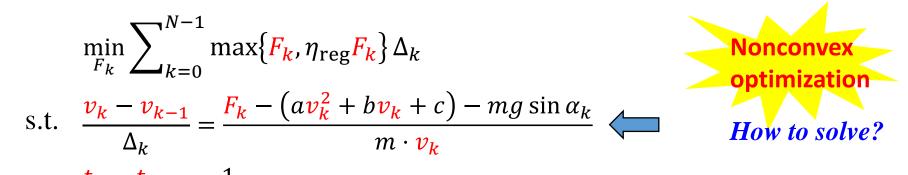
$$F_{k-1}$$

$$v_0 \cdots v_{k-1} v_k \cdots v_N$$

Ist segment $\cdots k$ -th N-th
$$S_0$$

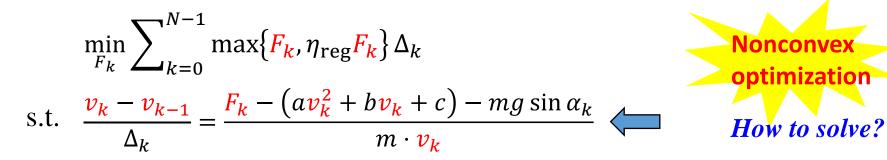
Location S_f



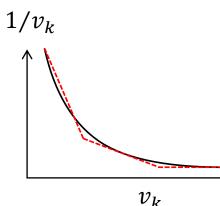


- i. Use off-the-shelf solvers
 - Cannot guarantee to find exact solutions
- $\stackrel{t_k t_{k-1}}{\longrightarrow} = \frac{1}{\nu_k}$ $\varepsilon \leq \nu_k \leq \overline{V}_k$ $F_{\min} \leq F_k \leq F_{\max}$ $\stackrel{r_{\min}}{\longrightarrow} P_{\min} \leq F_k \nu_k \leq P_{\max}$ $t_N t_0 \leq T$ $\nu_0 = V_0, \nu_N = V_f$

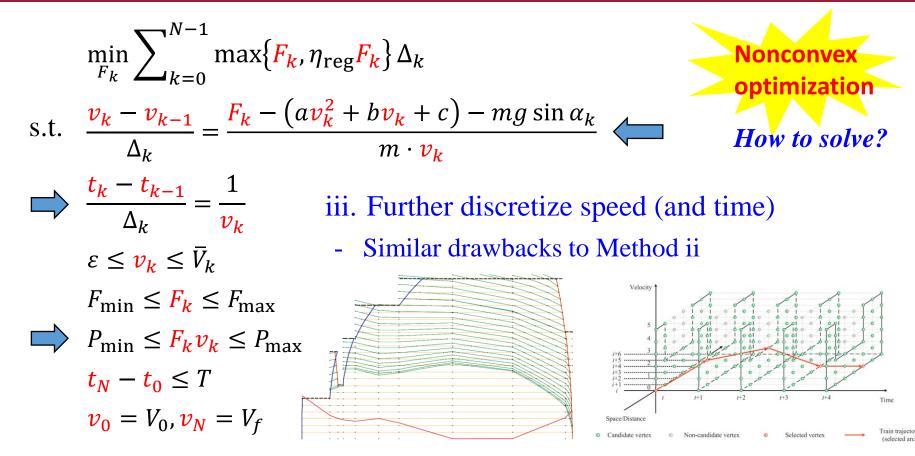




- $\stackrel{}{\longrightarrow} \frac{t_k t_{k-1}}{\Delta_k} = \frac{1}{\nu_k}$ $\varepsilon \le \nu_k \le \overline{V}_k$ $F_{\min} \le F_k \le F_{\max}$ $\stackrel{}{\longrightarrow} P_{\min} \le F_k \nu_k \le P_{\max}$ $t_N t_0 \le T$ $\nu_0 = V_0, \nu_N = V_f$
- ii. Approximate the nonlinear terms (\Rightarrow MILP)
- The solution is not the same as that of the nonconvex problem
- More pieces → longer computing time







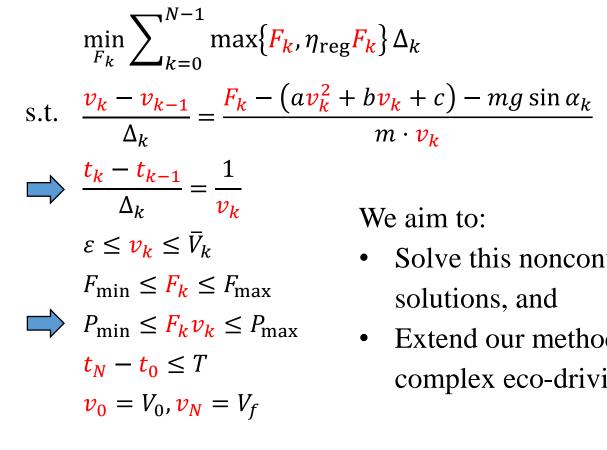
Source: Haahr et al. (2017)

Source: Zhou et al. (2017)



imization

How to solve?



We aim to:

- Solve this nonconvex problem to exact solutions, and
- Extend our method to solve the more • complex eco-driving problems

Our Solution Method: Step 1 – Reformulation

$$\min_{F_k} \sum_{k=0}^{N-1} \max\{F_k, \eta_{\text{reg}}F_k\} \Delta_k$$

s.t.
$$\frac{v_k - v_{k-1}}{\Delta_k} = \frac{F_k - (av_k^2 + bv_k + c) - mg \sin \alpha_k}{m \cdot v_k}$$

$$\implies \frac{t_k - t_{k-1}}{\Delta_k} = \frac{1}{v_k} \qquad \text{Let } z_k = 1/v_k$$

$$\varepsilon \le v_k \le \overline{V}_k \qquad \frac{t_k - t_{k-1}}{\Delta_k} = z_k$$

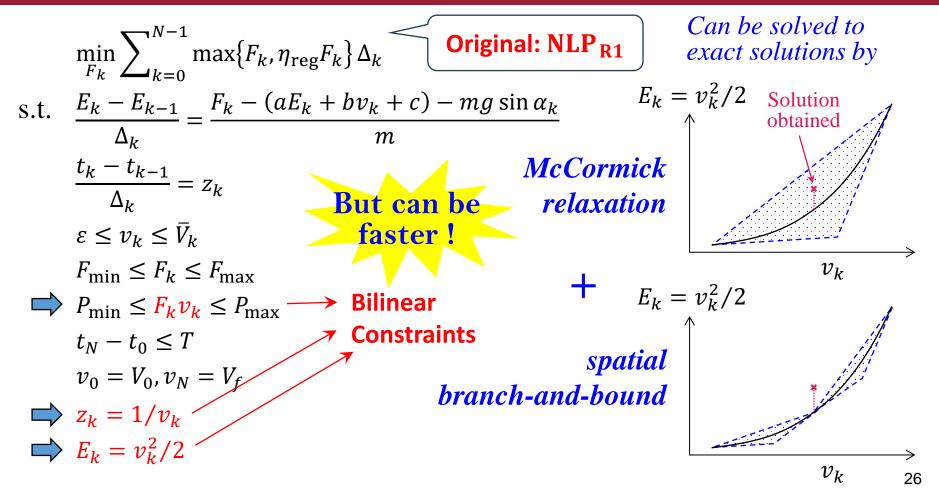
$$\implies P_{\min} \le F_k v_k \le P_{\max} \qquad \frac{t_k - t_{k-1}}{\Delta_k} = z_k$$

$$\implies P_{\min} \le F_k v_k \le P_{\max} \qquad \frac{E_k - E_{k-1}}{\Delta_k} = \frac{F_k - (aE_k + bv_k + c) - mg \sin \alpha_k}{m}$$



Our Solution Method: Step 1 – Reformulation





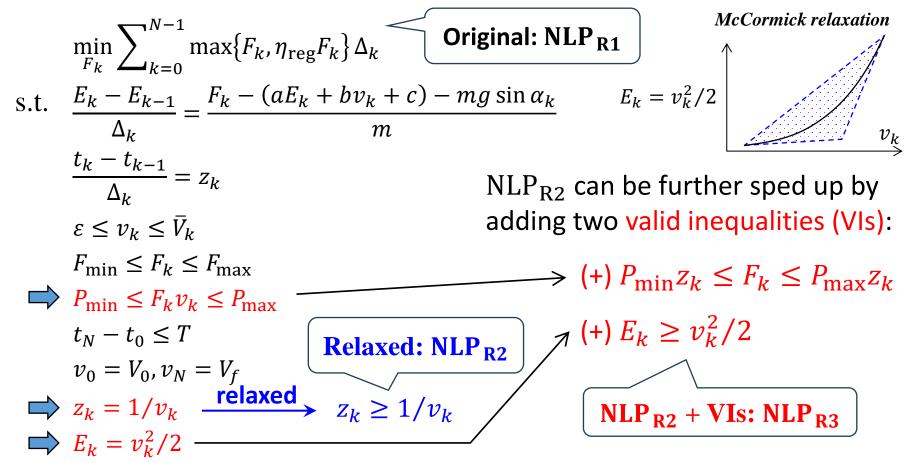
Our Solution Method: Step 2 – Relaxation



$$\begin{array}{l} \min_{F_k} \sum_{k=0}^{N-1} \max\{F_k, \eta_{\operatorname{reg}} F_k\} \Delta_k \qquad \text{Original: NLP}_{R1} \\ \text{S.t.} \quad \frac{E_k - E_{k-1}}{\Delta_k} = \frac{F_k - (aE_k + bv_k + c) - mg \sin \alpha_k}{m} \\ \frac{t_k - t_{k-1}}{\Delta_k} = z_k \\ \varepsilon \le v_k \le \overline{V}_k \\ F_{\min} \le F_k \le F_{\max} \\ F_{\min} \le F_k \le F_{\max} \\ t_N - t_0 \le T \\ v_0 = V_0, v_N = V_f \\ \Rightarrow z_k = 1/v_k \qquad \begin{array}{c} \text{Named NLP}_{R2} \\ \text{relaxed} \\ z_k \ge 1/v_k \\ \Rightarrow E_k = v_k^2/2 \end{array}$$

Our Solution Method: Step 3 – Valid Inequalities





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Scheduled trip time: 540 s Solver: Gurobi 11.0 Speed limit: 140 km/hr Gradient: 0 -0.04 $\mathbf{0}$ \rightarrow Distance (km) 9.01 14.20 15.02



Instance –	MILP with 32 pieces			NLP _{R1} (original)		NLP _{R2} (relaxed)		NLP _{R3} (R2+VIs)	
	Ctime (sec)	Gap (%)	Diff (%)	Ctime (sec)	Gap (%)	Ctime (sec)	Gap (%)	Ctime (sec)	Gap (%)
N = 170	146.43	0	1.21	3.01	0	2.71	0	0.89	0
N = 320	3600.69	1.7	-	8.54	0	5.65	0	2.68	0
N = 395	3600.57	-	-	135.25	0	10.53	0	3.32	0
N = 520				107.27	0	13.48	0	4.43	0
N = 770				3600.71	0.1	35.12	0	11.58	0
N = 1020				3600.55	0.1	80.69	0	19.29	0

Ctime: computing time

Gap: optimality gap when the solver terminated

Diff: difference of energy consumption from the NLP result

Conclusion, Extensions of Our Method, and Future Directions



- **Conclusion**: via reformulation, relaxation and valid inequalities, we developed a (more efficient) method of obtaining the exact solution to the nonlinear optimization problem for the classic single-train eco-driving problem
- We have **extended** our method to solve the eco-driving problems with:
 - Intermediate time-window constraints on a single train;
 - Joint eco-driving of a fleet of trains under green-wave policy for fixedblock signaling
- Possible future directions
 - More general signaling constraints: general fixed-block signaling with signal-dependent speed limit; moving-block; virtual coupling
 - Eco-driving under stochastic factors