

1 **TOWARDS OPTIMIZED DEPLOYMENT OF ELECTRIC BUS SYSTEMS USING**
2 **COOPERATIVE ITS**

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1 ABSTRACT

2 In this paper we analyze the impact of using cooperative intelligent transportation systems (C-ITS)
3 to manage electrical bus systems. A simulation-based study is presented where three control
4 strategies are used to regulate the operations of a line, namely bus holding, Green Light Optimal
5 Dwell Time Adaptation (GLODTA) and Transit Signal Priority (TSP). The results show, using a
6 realistic scenario of a major line in Luxembourg City, that buses are efficiently operated without
7 necessarily providing additional priority to public transport, hence without negatively affecting the
8 capacity of the private vehicles system. Benefits in terms of headway regulations, energy
9 consumption and travel time variance reductions are quantified.

10

11

12 *Keywords:* Public transport · Cooperative ITS · e-buses · Driver Assistance Systems

13

14

1 INTRODUCTION

2
3 Sustainable urban development motivates investments in environmental-friendly and user-centred
4 Public Transport (PT) services. Recent trends towards next generation PT systems show the
5 development of greener vehicles such as battery electric/hybrid buses (e-buses), which are
6 introduced to reduce the emission of pollutants, especially in urban environments (thanks to the
7 implementation of, for instance, zero-emission zones), and the noise caused by traditional
8 combustion engines. Apart from engine technology advancements, increased penetration of e-
9 buses is also favoured by the introduction of new solutions at the level of charging infrastructures
10 (opportunity or on-route charging, flash charging), which allow extending the operational range of
11 electric vehicles.

12 A second trend is observed on bus operators seeking higher service quality beyond conventional
13 performance measures such as service regularity and/or punctuality, for instance through increased
14 ride comfort and reduced emissions and energy consumption via mitigation of stop-and-go driving
15 patterns. These additional features are possible thanks especially to the introduction of sensors
16 allowing real data information retrieval (automatic vehicle location, automatic passenger counts,
17 etc.) and communication between all actors involved in the bus eco-system (vehicle-to-vehicle and
18 vehicle-to-infrastructure communication). These technologies, which define the so-called
19 Cooperative Intelligent Transportation System (C-ITS), offer great potential to improve the overall
20 system performance, and to increase the level of driving control and automation. In particular,
21 more conventional bus control systems (holding, stop skipping, etc.) and Transit Signal Priority
22 (TSP) can be extended and improved thanks to information on e.g. signal times and phases of
23 traffic lights, hence reducing the number of unneeded stops or signal timing changes. C-ITS has
24 also great potential in reducing bus operating costs related to energy consumption and equipment
25 wear and tear during bus operations (*I*).

26 These trends, however, bring new challenges. The first one is due to different operational
27 characteristics and constraints characterising e-buses, e.g. they need to periodically recharge
28 batteries at charging stations placed in terminals and (optionally) in bus stops. Despite fast
29 technological advancements are showing that range extension of e-buses is growing significantly,
30 it is still hard to imagine fully electric buses to operate the whole day without being recharged at
31 some point of the day. This brings additional complexity into PT operations and its costs. The
32 second challenge is to find measures able to provide comfort- and cost-related benefits without
33 negatively impacting the general traffic performance. Relying solely on strategies such as TSP,
34 which prioritises buses at signalised intersections penalising the other traffic streams, might cause
35 congestion effects that could backfire on the PT system itself via blocking back phenomena.

36 In this paper, we demonstrate how these challenges can be effectively addressed by the emerging
37 Cooperative ITS solutions. A novel framework addressing the problem via combination of
38 cooperation and multi-objective optimisation is introduced. In particular, we extend a C-ITS-based
39 driver assistance system, namely the Green Light Optimal Dwell Time Adaptation (GLODTA),
40 which adapt driving speeds and holding times, respectively, to avoid buses to arrive at nearby
41 traffic lights during the red phase, hence reducing the number of stops at intersections. We
42 investigate how the use of such C-ITS based solutions contribute to minimise total energy
43 consumption. We also show that, by adopting these solutions, we reduce the need to resort to transit
44 signal priority, hence in turn we reduce the negative impacts on private traffic.

45 The paper is structured as follows. The next section provides a review of the relevant literature.
46 Then, Section 3 describes the whole methodology. Section 4 presents a realistic case study on an
47 electrified urban line and then compares different strategies both in terms of operational efficiency

1 and in terms of energy consumption. Finally, Section 5 provides the main conclusions and
2 recommendations for future research.

3 4 **LITERATURE REVIEW**

5
6 PT systems consisting of e-buses reduce emissions, energy use, and noise as well as offer smoother
7 rides (2). Currently, e-bus systems are moving from pilot projects to small-scale deployments with
8 single line/operator and with very few charging stations. The potentials and technical needs of
9 large-scale e-bus systems have been recently under investigation by, for instance, the EU's Zero
10 Emission Urban Bus System project (3). Peak demand charges (based on the maximum amount of
11 electrical power drawn from a grid within certain period) are a major barrier to e-bus deployment
12 (4). As opportunity charging provides the technical feasibility for the deployment of fully
13 electrified lines, it comes with high costs for the line operators, and, in future large deployment
14 scenarios, may create issues for the electrical grid, it is important to find measures to reduce the
15 energy consumption during operations to reduce the need for daily charging operations.

16 Therefore, the charging requirements create a strong link between infrastructure planning and bus
17 operations (5). Currently, approaches for optimal recharging of e-bus systems are based on design
18 and economical principles and do not consider in detail the actual energy consumption at the
19 operational level (6). The existing research efforts currently focus on developing a proper system
20 design such as deploying strategic locations of charging stations (7, 8). At operational levels,
21 energy efficiency is currently addressed via energy management strategies for the engine (9), and
22 regenerative braking technologies (10).

23 Additional gains in terms of both operational efficiency and reduced energy consumption can be
24 obtained through prioritising PT systems at intersection through Transit Signal Priority (TSP) (11).
25 Currently, in the literature, TSP, together with holding control strategies (12) have been designed
26 to only support the simple punctuality objectives or aim at regulating headways in frequency-based
27 bus systems. TSP strategies can be seen as cost-efficient, since they overall reduce the number of
28 stops at signals, hence avoiding additional stop-and-go operations. On the other hand, such control
29 measures may have some negative impact on the general performance of the whole urban transport
30 system: excessive use of TSP may reduce the capacity of competing traffic streams.

31 Holding strategy has proven to be effective in restoring regularity and maintain a smooth operation
32 (13–17). However, holding strategies have also some limitation as, by delaying buses at stops, they
33 may increase the total trip times, increase passenger on-board and waiting times, and force line
34 operators to increase the fleet size in order to guarantee a certain service level.

35 Connected vehicle technology can also contribute to reduce the energy consumption, and in the
36 same time improve operational efficiency of bus systems. In particular, the communication of
37 Signal Phase and Timing (SPaT) information obtained from traffic signal controllers allows to
38 switch from signal-centric strategies (for instance, resorting to TSP requests) towards vehicle-
39 centric (18). The two SPaT-based controls that are researched in literature are the Green Light
40 Optimal Speed Advisory (GLOSA) (19, 20) and the Green Light Optimal Dwell Time Advisory
41 (GLODTA) (21). Both solutions have been conceived to mitigate stop-and-go driving. GLOSA
42 does so by providing vehicles with speed guidance, while GLODTA reaches the goal by optimizing
43 dwell time of PT vehicles (i.e. by occasionally holding the buses longer at bus stops). Consequently,
44 performance of traffic flow of buses is improved without the need of changing traffic signal timings.
45 As up to 20% more fuel is used to accelerate from a full stop to a speed of 8 kilometres per hour
46 (in case of a passenger car), there are significant benefits of moving to stop-and-go or slow-and-
47 go patterns.

1 GLOSA has been studied in several projects and field operational tests for both cars and buses, e.g.
2 PREDRIVE C2X (22), DRIVEC2X (23), simTD (24), MobiTraff (18), Compass4D (18)
3 Extensions of GLOSA have been found in the literature in combination with adaptive signal
4 control strategies (25), with vehicle platooning (20), and to generate fuel-efficiency speed profiles
5 (26). Very limited works combine GLOSA with e-vehicles (27). GLODTA advises a prolonged
6 dwell time at bus stops in order to avoid arriving at the next signalized intersection during a red
7 phase (18).

8 Both GLOSA and GLODTA strategies rely on Signal Phase and Timing (SPaT) data continuously
9 communicated from controllers placed along the route. Furthermore, real-time positions of buses
10 in the network are accessed through Automated Vehicle Location (AVL) systems to estimate the
11 speed and the additional dwell times. However, the two aforementioned systems do not yet take
12 into account battery charging requirements of electric buses with on route charging. Recently, the
13 work of Giorgione et al. (28) addressed this issue. The proposed eGLOSA instructs the driver to
14 maintain a specific speed so that the bus traverses the next signalized intersection without stopping
15 and affecting signal timings, and it further considers the energy consumption. On the other hand,
16 eGLODTA determines whether additional dwell time should be advised, considering both schedule
17 adherence criteria and on-route battery charging needs.

18 In this paper we investigate whether C-ITS-based solutions, and in particular GLODTA, can be
19 adopted to reduce the need to resort to TSP and in combination with holding strategies, contribute
20 to increase the regularity of the bus service.

21 This study can therefore be seen as an extension of the work of Giorgione et al. (28), which has
22 focused on introducing energy consumption in the operational objectives of e-bus systems on
23 schedule-based services. We will instead include bus regularity objectives in frequency-based
24 services. This work can also be seen as complementary to recent work of Seredynski and Viti (1),
25 where GLOSA and GLODTA have been studied to reduce the need for TSP, but for conventional
26 bus systems.

27 28 **METHODOLOGY**

29 30 **Modelling assumptions**

31 Our modelling approach consists of two layers: the first represents the physical eco-system,
32 consisting of:

- 33 • a traffic light system managed by a traffic management centre;
- 34 • the bus system; and
- 35 • the charging infrastructure managed by the same bus operator.

36 The second layer corresponds to the cooperative communication environment composed of
37 interconnected vehicles communicating positions and priority levels. The goal of our approach is
38 to manage efficiently three interacting components: 1) the in-vehicle control managed by a Driver
39 Assistance System (DAS) dashboard supporting the bus drivers, 2) a signal control system
40 regulating traffic and eventually providing priority to the buses, and 3) a centralised back-office
41 system, which takes care of the bus operation and of e-bus charging.

42 Traffic lights are assumed as pre-phased, with a-priori stages, cycle times and green/red timings.
43 We assume that all signals are equipped with Dynamic Short-Range Communication (DSRC, also
44 known under the name of ETSI ITS-G5 in Europe), which allows infrastructure-to-vehicle
45 communication within a range of around 200m) and provide SPaT information to all vehicles
46 within this range. Each e-bus is equipped with AVL and APC systems, and it collects real-time
47 information about locations and battery status of each e-bus. We assume that the back-office, via

1 AVL information, seeks for cost-efficient use of e-charging infrastructure with minimisation of
2 impact on operations, and optimised e-charging schedules.

3 Schedules are used as constraints to the operational times of buses so it is important to guarantee
4 punctuality (for scheduled services) and/or regularity (for frequency-based services) in order to
5 guarantee optimal charging operations (29). On the other hand, change in charging plans may
6 always occur during the day as the bus system remains unavoidably stochastic due to boarding and
7 alighting operations, traffic congestion, incidents, etc. However, we leave the problem of re-
8 computing optimal charging schedules due to operational delays to future research. Since full
9 charging takes approximately 6 minutes (4), it is reasonable to assume that a bus leaving the
10 terminal will be fully charged. On the other hand, buses are required to terminate their trip with at
11 least 10% remaining battery to consider a safety margin to prevent the bus to stop while still on-
12 route.

14 **Overview of the model**

15 We present a novel control strategy in which, when a vehicle gets sufficiently close to a traffic
16 light to obtain SPaT information, the decision is not limited to traverse the next intersection during
17 the green phase, but it involves the actual time headway between consecutive vehicles, in order to
18 arrive evenly spaced as best as possible at stops, hence reducing the level of bunching and overall
19 improve line regularity.

20 In particular, at stops where holding is applied (Time Control Points, or TCPs), holding time for
21 regularity is determined by a simple rule subject to the forward and backward headways (Cats et
22 al., 2011). In order to ensure that vehicles, by the time of their departure, will traverse the
23 intersection without stopping, the additional time needed is estimated via GLODTA. In case of a
24 late arrival, only GLODTA time is checked and triggered only if it results in time saving for the
25 line.

26 In the control strategy developed, calls for green time extension and green recall are also
27 considered, expecting to be in line with the findings of previous studies for need of weak TSP
28 instead of strong TSP (21).

30 **Problem Formulation**

31 We assume a bus line i , the route of which consists of J stops and there are K trips conducted.
32 Between stops, there are $J-1$ links that connect the stops with different lengths. Links may contain
33 a signalized intersection, the distance of which from the upstream bus stop is known. A fixed cycle
34 is assumed for all traffic lights. All trips are conducted with the same vehicle type with similar
35 characteristics and same average speed.

36 In the following subsections, we present the different controls that determine the trajectory of the
37 bus along the route.

39 *Holding strategy*

40 As primary objective, bus regularity should be sought in order to guarantee high quality of service
41 for the passengers and for smooth and efficient charging operations at terminals. A well-established
42 control strategy to regularise bus headways is holding (14), in which buses are instructed to hold
43 on a bus stop an additional time such that headway with the preceding and the succeeding vehicles
44 is modified.

45 The holding criterion chosen for this study is the criterion used by Cats et al (16). The criterion
46 has been compared with other holding strategies using simulation for frequency-based services
47 and has proven to be superior (16, 30). The criterion is based on the actual headway of the vehicle

1 subject to its preceding and succeeding buses. Additionally, the maximum allowed holding time is
 2 limited to a specific share of the planned headway. The formula to calculate departure (exit) time
 3 ET_{ijk} is given as equation (1):

$$4 \quad ET_{ijk} = \max(\min(\text{term1}, \text{term2}), AT_{ijk} + DT_{ijk}) \quad (1)$$

6
 7 Where *term 1* and *term 2* are respectively formulated as:

$$8 \quad \text{term1} = AT_{ijk-1} + \frac{AT_{imk+1} + SRT + AT_{ijk-1}}{2} \quad (2)$$

$$9 \quad \text{term2} = AT_{ijk-1} + \alpha PH \quad (3)$$

10
 11 In the equations, AT_{ijk} and DT_{ijk} are the arrival time and the dwell time of bus k at stop placed on
 12 link j and for line i , respectively. *term1* is estimated as the average between the arrival time of the
 13 previous vehicle AT_{ijk-1} and the arrival time of the following vehicle $k+1$. We estimate the arrival
 14 time of the succeeding vehicle $k+1$ from the current stop by summing the scheduled riding time
 15 SRT between the last visited stop m and the current stop j and the arrival time at the last visited
 16 stop m AT_{imk+1} . PH is the planned headway, while α is the threshold ratio parameter that limits
 17 the maximum allowed holding. Threshold ratio parameter ranges between 0.6-0.8 (13, 16). For
 18 simplicity's sake, we will consider a single line control from now on.

20 21 *GLODTA*

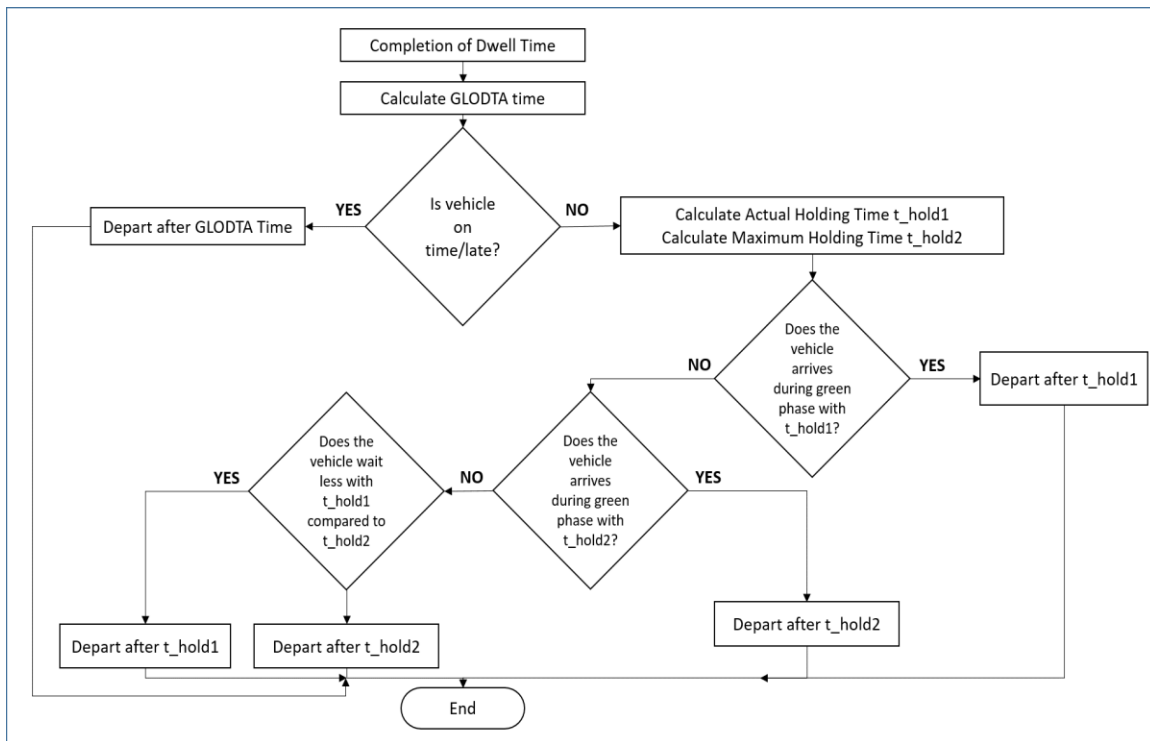
22 Holding is an effective way of controlling buses for regularity objectives. As aforementioned, this
 23 comes with a cost, as buses are delayed at stops, yielding to longer trip times and in turn more
 24 vehicles needed to operate at a certain service level. Similar to holding strategies, GLODTA
 25 introduces additional dwell times at bus stops. Specifically, given that a bus stop is within the
 26 DSRC range distance from the next signalized intersection downstream, when the vehicle is ready
 27 for departure, the driver can be instructed to hold an additional amount of time if the average speed
 28 to reach the signal will allow him or her to traverse the next green phase. This means that eventual
 29 delays incurred at the signal are instead transferred to the upstream stop. The additional advantage
 30 is that the total number of stops for a trip is reduced, and hence less energy is consumed and
 31 emissions are expected to be reduced. Details on how GLODTA can be implemented in practice is
 32 given in Seredynski and Khadraoui (21).

33 34 *R-GLODTA*

35 The aforementioned solutions have been developed to seek different objectives (regularity for the
 36 first and minimization of stops at traffic lights for the other two). However, holding and GLODTA
 37 are based on the same principle of delaying a vehicle by remaining for additional time at a stop.
 38 We explore a potential synergy of both by computing analytically, when a vehicle completes its
 39 dwell time, the holding time depending on its headway from the preceding and the succeeding
 40 vehicles together with the time needed to traverse the next green phase.

41
 42 If the vehicle is either on time or late, when no holding is needed, in order to assist the operation
 43 by saving time at the link, only GLODTA is applied, again if needed. In case of an early arrival,
 44 the vehicle should remain at the stop in order to restore regularity. From equation 1 both terms for
 45 actual holding time needed and the maximum holding are estimated. Then, we check if with either
 46 of the two the vehicle can hit a green phase. If one of the two times meets both regularity and green

1 light criteria, it is selected and counts as a combined controller. If with both holding times we hit
 2 red then the holding time with the less estimated remaining time at the traffic light is selected and
 3 the controller counts simply as a regularity controller. This joint strategy, which we name R-
 4 GLODTA, is shown with the following Figure 1.



6
7
8
9 **FIGURE 1 Operating rules of R-GLODTA**

10 *Transit Signal Priority*

11 Transit Signal Priority can be adopted at traffic lights to avoid buses to stop and in the same time
 12 provide some additional priority to the general traffic stream. This is obtained by detecting bus
 13 arrivals (via detectors or via C-ITS communication) and modifying green and red times. Green
 14 times can be anticipated (or extended) such that vehicles pass without modifying their planned
 15 trajectory. In our approach, before requesting a priority at traffic signals a bus attempts to use
 16 GLOSA and GLODTA advisory systems. That is, first a bus attempts to avoid stopping at signals
 17 by using uniquely the combination of the two systems. If this fails, priority request can be sent to
 18 the traffic controller.

19 **Computational Algorithm**

20 We present here the whole procedure to apply our integrated control method below:

21
22 Given

- 23 • Network layout (number of line segments, bus stop positions);
- 24 • Traffic light position and signal timings;
- 25 • Operational speeds of buses and their scheduled riding time; and
- 26 • Passenger demand profile in terms of arrival rates per OD pair at each bus stop.

27
28 CONTROLS


```

1  1.  GLODTA
2      IF
3          distance between stop and next traffic light is less or equal than 200m
4      THEN
5          Calculate extra holding time Tglodta to pass on green (calculated at the end of the dwell time
6          at the stop)
7      IF
8          Tglodta>0
9      THEN
10         apply GLODTA
11  2.  R-GLODTA
12      IF
13         distance between stop and next traffic light is less or equal than 200m
14      AND
15         vehicle early or on time
16      THEN
17         Calculate extra holding time Tglodta to pass on green (calculated at the end of the dwell time
18         at the stop)
19      ELSE
20         Calculate holding time using regularization strategy (1-3)
21      IF
22         holding time within Tglodta interval to pass on green
23      THEN
24         use holding time
25      ELSE
26         depart for the holding time with the least expected waiting time at traffic light
27  3.  TSP
28      IF
29         GLODTA cannot be applied (out of mix-max values)
30      THEN
31         Check if above boundaries can be met with TSP (extend/anticipate green of max 5s)
32

```

33

34

35 **SIMULATION STUDY**

36

37 The proposed algorithmic scheme is tested and evaluated by simulating a high frequency line.
38 Control is applied at specific stops of a bus line where high passenger demand and delays in terms
39 of travel time are observed. We compare the new control criteria with independent application of
40 holding and the DAS at the selected TCPs and a no-control case is used as a benchmark.

41

42 **Simulation Environment**

43 We developed a simulation environment in MATLAB. The basic elements of the code are the
44 physical network and the passenger demand.

45 During simulation, there are two dimensions that are logged; Time and Passengers. Passengers
46 enter and exit to the network via the stops along the line. The time passengers spend in the network
47 is the time spending waiting for the next vehicle to board at stops and the in-vehicle time (the time
48 between the origin stop and the destination stop including all intermediate stops).

49 In terms of time, the following information is recorded:

- 1 • Dispatching time from terminal;
- 2 • Actual riding time between stops;
- 3 • Arrival time at each stop;
- 4 • Dwell time at each stop; and
- 5 • Departure time at each stop;

6 For passengers:

- 7 • The number of boarding passengers;
- 8 • The number of alighting passengers; and
- 9 • The number of passengers on board.

10 The main sources of stochasticity are the dispatching time, the riding time between stops and the
11 arrival of passengers at stops. Dispatching times are drawn by Gamma distribution given the
12 planned headway between departures and the user can increase the variability by changing the
13 shape and the size of the distribution. For the current setup, we assume that vehicles are dispatched
14 perfectly regular.

15 The calculation of actual riding times on the links between stops depends on the existence of a
16 signalized intersection. If there is no signalized intersection, the riding time is sampled by
17 lognormal distribution with given inputs the scheduled riding time and the desired standard
18 deviation. Since empirical data is not available, the standard deviation is set to 20% of the
19 scheduled riding time. In case of a signalized intersection, actual speed is sampled from a normal
20 distribution, getting values between a minimum and a maximum speed given by the user. If the
21 vehicle pass on green with its actual speed, continues until the next stop and the actual riding time
22 is registered. In case of a stop on a red, the waiting time at red is registered and vehicle continues
23 after resampling a new speed. The actual riding time is the sum of the running times prior and after
24 the intersection and the waiting time at the traffic light.

25 We assume that passengers arrive at stops randomly following Poisson distribution. The simulator
26 gets as input the arrival rate per OD pair and during simulation the actual headway in order to
27 generate the number of passengers at each stop. Alighting passengers at each stop are the share of
28 passengers boarded at the previous stop with destination the specific stop. Occupancy of the bus
29 is defined at each stop by the numbers of boarding and alighting passengers and the last recorded
30 occupancy.

31 During simulation each trip is generated and arrives at the first stop at its dispatching time. At each
32 stop, after the arrival time of the vehicle is registered and the actual headway is calculated and
33 boarding passengers are estimated. Dwell time is calculated as the sum of the product of the
34 boarding and alighting passengers with the corresponding rate with a constant that represents a
35 potential delay at the stop. After the completion of dwell time, if no control is applied the vehicle
36 departs and the departure time is registered. If control is applied the corresponding controller is
37 triggered and returns the additional time a vehicle is help at the stop. If the controller includes a
38 TSP call, the time the green phase of the next traffic light is also returned to be added to the cycle.
39 Overtaking is not allowed. In the case of bunched vehicles, the successor cannot depart prior to its
40 preceding vehicle. As a recovery strategy, the following vehicle departs after an additional time to
41 increase the headway between vehicles.

42 Each replication simulates a 3-hour operation of the bus lines. For each replication, the first three
43 and the last three trips are excluded from the analysis as they run deterministically and constitute
44 the warm up and the cool down period of the simulation.

45

46

1 Performance Indicators

2 The main performance indicators used in this study are the adherence of headway of the line as
3 well as total travel time and its variability. Moreover, we will also analyse the delay at the different
4 intersections and the times the vehicles managed to pass through a green phase, in order to compare
5 the results at both network level and at a local scale. The performance indicators selected for the
6 study are the following:

7 Regularity indicators: Coefficient of variation of headways; bunching;

8 Passengers' cost indicators: in-vehicle time; waiting time at stops;

9 Link performance indicators: stop frequency and delay at traffic light; average speed and travel
10 time variance;

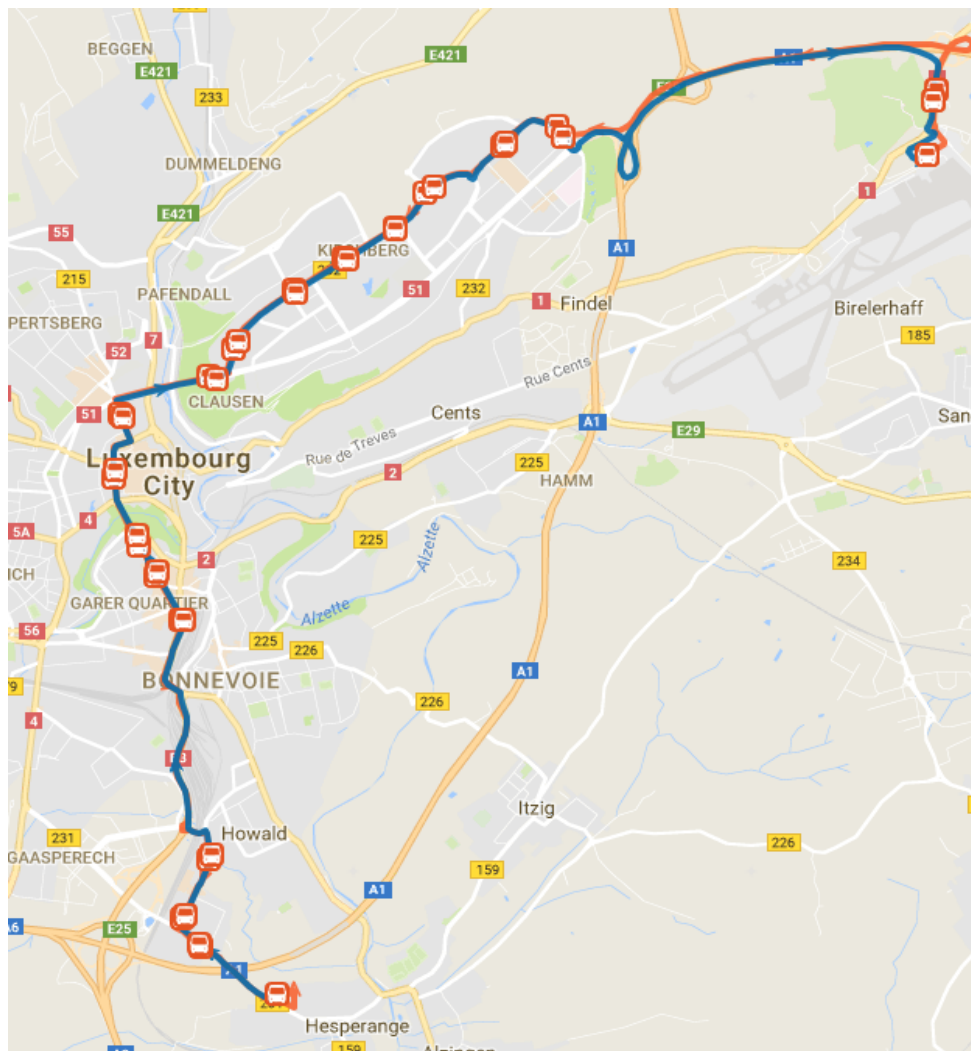
11 Energy consumption: average battery status at the end of the trip.

12

13 Case Study

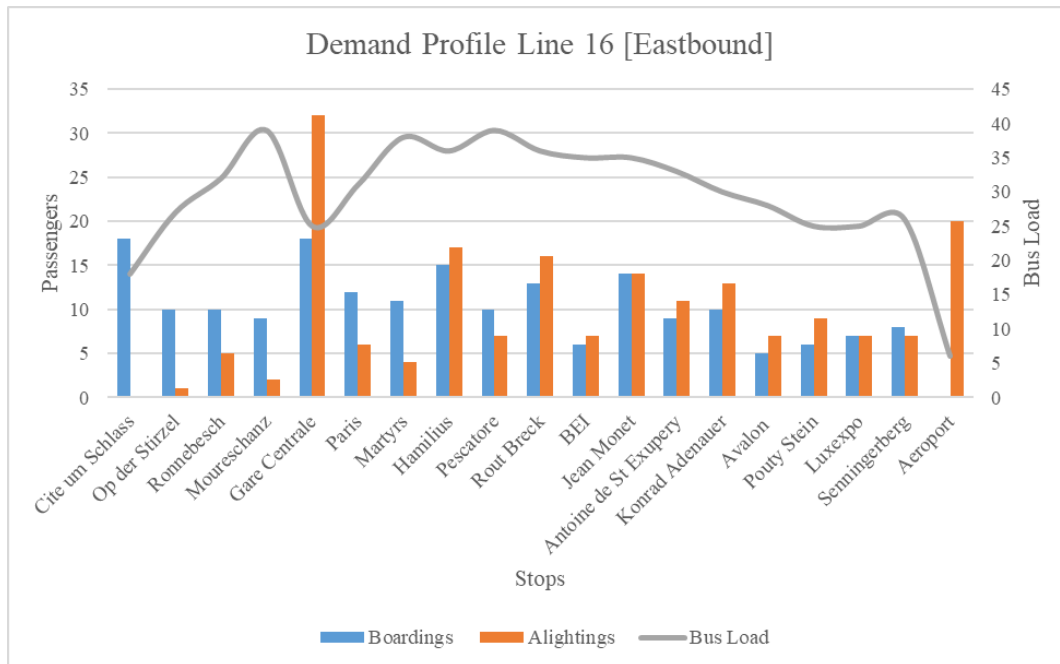
14 We tested the simulated controls for one of the busiest lines of the city of Luxembourg, Line 16.
15 The route of the line is depicted in Figure 2.

16



17
18
19
FIGURE 2 Line 16, Luxembourg City

1 The line consists of 19 stops in the eastbound direction and connects the new activity zone at the
 2 south part of the city with the city centre, the central business district of Kirchberg and finally the
 3 airport of Findel. This line provides also connection with the major transport hubs of the city
 4 (central railway station, Kirchberg multimodal transport hub and the airport). The frequency of the city
 5 line is 10min and articulated buses are used for the trip. The demand profile of the line is displayed
 6 in Figure 3.
 7



8 **FIGURE 3 Demand Profile of Line 16**

9

10

11 All intersections have been assumed equipped with TSP technology, and bus stops have been set
 12 as time control points (TCP). We assume that all traffic lights have the same signal program with
 13 cycle of 120sec (80 green and 40 red) with the red indication first at the simulation environment.
 14 No coordination has been considered between signals.

15 We simulated and compared 5 different scenarios (25 replications were conducted for each
 16 scenario):

17

18

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- No Control: the e-buses are running without any C-ITS support and they do not receive any priority at signals;
- Holding: the e-buses seek for headway regularisation via holding at each bus stop;
- GLODTA: when applicable, additional dwell time is added to the boarding-alighting time at each stop to avoid arriving during red at the next traffic light;
- R-GLODTA: a combination of the two strategies;
- R-GLODTA +TSP: additional to holding at the stop, priority is given at traffic lights.

25

26

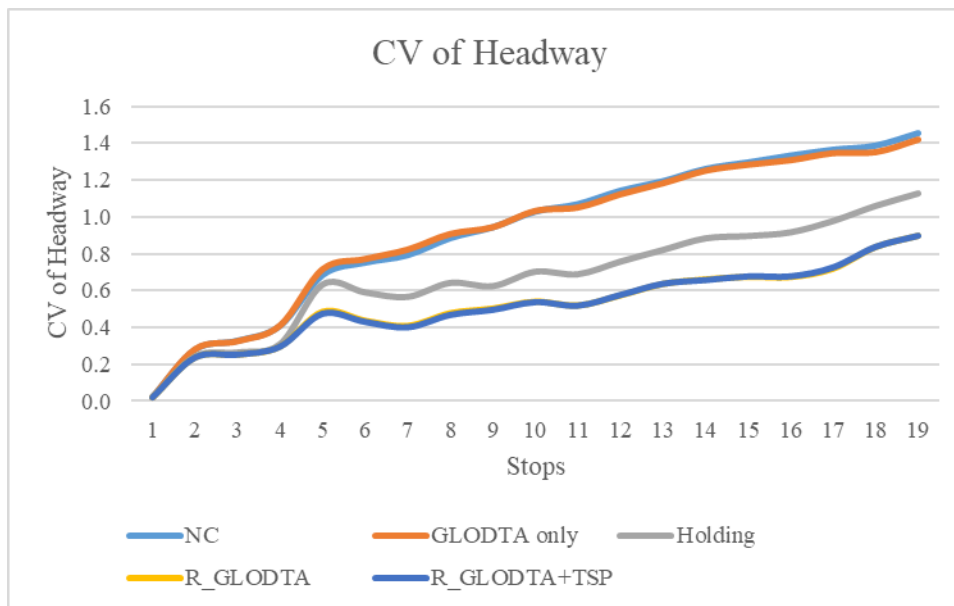
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28

1 SIMULATION RESULTS

3 Line Performance Indicators

4 An index of the variability of headway at each stop but also for the line is the coefficient of
 5 variation of headway (31). Coefficient of Variation of headway is the ratio of the standard deviation
 6 over the average headway. Figure 4 compares the behaviour of the coefficient of variation (CV) of
 7 headways for all simulated scenarios. It is clear from the results that headway variability improves
 8 when the controller accounts for regularity.
 9



10 **FIGURE 4 Coefficient of Variation (CV) at stops for the different scenarios**

11
 12 As one can see, the R-GLODTA outperforms both holding and simple GLODTA strategies, which
 13 straightforwardly do not show any significant improvement with respect to the no-control scenario.
 14 This is because it is more generous than the simple holding strategy when adding time, with the
 15 possibility to allow maximum holding time to ensure also the GLODTA stop-avoidance criterion.
 16 It should also be noted that in terms of CV improvement, R-GLODTA shows similar achievements
 17 then with TSP, hence on the basis of bus regularity, TSP does not provide any additional gain.
 18
 19

20 Regularising headways comes with a price: it penalizes the passengers on board with extra in-
 21 vehicle delay compared to simple holding, as shown in Table 1. On the other hand, vehicle
 22 bunching as well as passengers' waiting times are significantly reduced.
 23
 24

TABLE 1 Performance Indicators of line 16

Scenario	CV Line	Bunching	Waiting time [sec]	In vehicle time [sec]
NC	0.92	0.546	150.17	152.18
GLODTA	0.92	0.53	150.6	154.22
HOLDING	0.66	0.37	150.4	161.8
R_GLODTA	0.52	0.27	145.7	162.5
R_GLODTA+TSP	0.52	0.26	145.6	162.55

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Link Performance

Table 2 shows how R-GLODTA is also very effective in reducing the number of stops, waiting time and overall increase the average speed at the link if compared to no-control and single control strategies. Clearly, in this case TSP provides some additional gain but this is relatively marginal.

TABLE 2 Link performance Indicators

Scenarios	Stop Frequency at traffic light	Average Waiting Time at Traffic Light [sec]	Average Speed
NC	0.329	6.76	21.8
GLODTA	0.261	3.88	22.1
HOLDING	0.314	6.62	21.7
R_GLODTA	0.170	2.81	23.2
R_GLODTA+TSP	0.163	2.67	23.3

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Travel Time

In general, a travel time distribution with less variability allows to the operator to create a more robust schedule and schedule the number of vehicles needed. Figure 5 shows the trip time distribution for each scenario. As shown in the figure with control strategies that account for line regularity, the objective of travel time with reduced variability is achieved. Again, R-GLODTA shows very similar performances with or without resorting to TSP, which one more time indicates that such control strategy may be effectively applied without taking away capacity from opposing traffic streams. On the contrary, by reducing the number of stops at traffic light alone (GLODTA scenario) cannot guarantee to a stable travel time distribution.

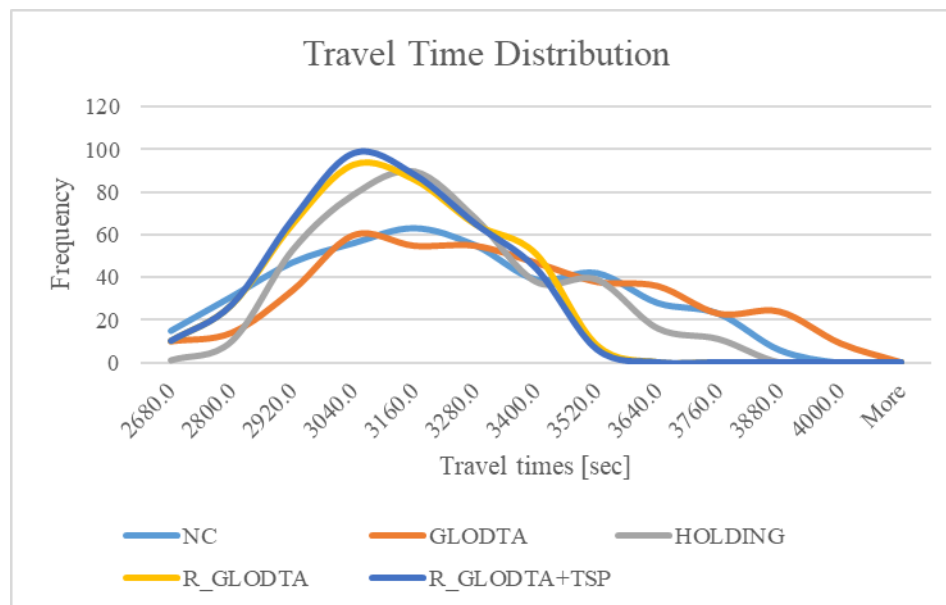


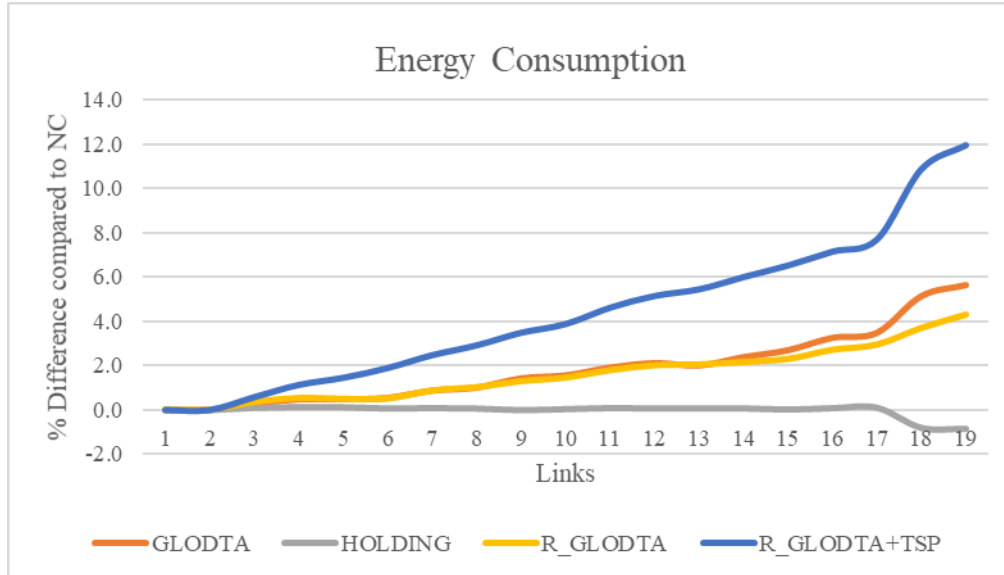
FIGURE 5 Travel time distribution for each scenario

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Energy Consumption

Finally, we investigate the impact of adopting different control strategies affect the electrical energy consumption. We adopted the energy consumption model used in Giorgione et al. (28),

1 slightly adapted to consider additional traction force needed to perform bus accelerations. Figure
 2 6 provides the trend in terms of average energy saved with respect to the no-controlled scenario.
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4 **FIGURE 6 Energy consumption for each scenario**

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 7 As one can observe, holding provides no gain with respect to the NC scenario, and actually it
 8 slightly increases the total energy consumed due to the additional holding times. All other
 9 strategies are positively impacting energy consumption, with R-GLODTA with TSP strongly
 10 outperforming the other methods and reaching more than 11% less energy consumption with
 11 respect to the no-control scenario.

12 **Controller Performance**

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 15 A final analysis is performed to check how many times the different strategies are adopted in the
 16 simulated scenarios. Figure 8 shows the share of each control decision, i.e. when each control was
 17 needed. As one can observe, fixing regularity was needed in around 38% of the controlled cases,
 18 when vehicles were early, while in all other cases only GLODTA was used to ensure that vehicle
 19 would pass during green. In the cases that holding was needed R-GLODTA was used more
 20 frequently used than simple holding ignoring the indication at the intersection downstream. It
 21 should be mentioned that in the simulated environment, control was needed approximately 48%
 22 of times.

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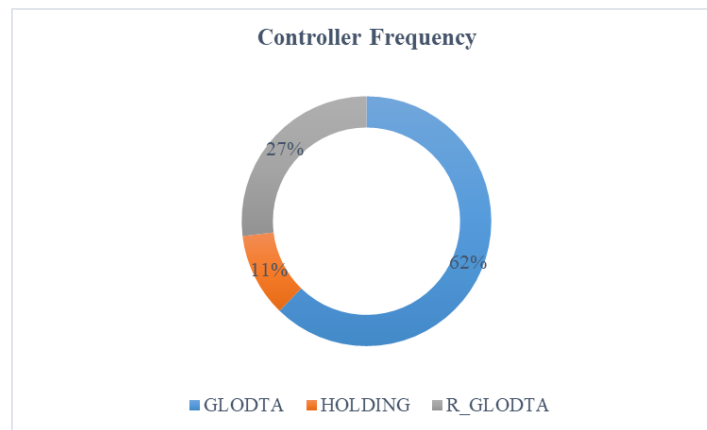


FIGURE 8 Share of times each control strategy was used

CONCLUSIONS

This paper presented a novel control strategy that combines a Cooperative ITS-based driving assistance system, namely Green Light Optimal Speed Adaptation (GLODTA) with bus holding control at bus stops. In particular, benefits of this integrated strategy for the deployment of electrical buses is presented through a realistic simulation scenario.

The logic behind this integrated method is that bus holding can effectively improve different objectives, namely 1) bus regularity performance, 2) passengers costs, 3) trip performances, and 4) energy consumption.

Using a case study inspired by a real line in Luxembourg city, we showed that significant gains are achieved on all four performance indicators. Additionally, we show that the novel R-GLODTA strategy provides effective improvements even when Transit Signal Priorities are not provided to buses.

Future research will focus on testing if similar conclusions are confirmed using the other C-ITS-based solution, i.e. GLOSA, and will extend the current simulation to multiple interacting lines.

AUTHOR CONTRIBUTION STATEMENT

“The authors confirm contribution to the paper as follows: study conception and design: G. Laskaris, M.Seredynski, F. Viti; model(s) formulation and implementation: G. Laskaris, F. Viti; analysis and interpretation of results G. Laskaris, F. Viti; draft manuscript preparation: G. Laskaris, M.Seredynski, F. Viti. All authors reviewed the results and approved the final version of the manuscript.”

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