

# CYCLING AT INTERSECTIONS: A MULTI-OBJECTIVE ASSESSMENT FOR TRAFFIC, EMISSIONS AND SAFETY

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**Abstract.** This paper evaluated in an integrated manner the traffic performance, pollutant emissions and road conflicts between bicycles and motor vehicles at a signalized intersection. Two alternative scenarios were examined: (1) bicycles increment and motor vehicles replacement within the cycle-fixed traffic signal; (2) replacing the existing traffic control by a conventional two-lane roundabout and evaluating the impacts of bicycles increment. For each scenario, bicycle demand was varied from 9 to 270 bicycles per hour (bph). Traffic flow and vehicle dynamic data were collected from a three-leg signalized intersection in Aveiro (Portugal). The microscopic traffic model (VISSIM) paired with an emission (Vehicle Specific Power – VSP) methodology and safety (Surrogate Safety Assessment Methodology – SSAM) model were used to assess intersection-specific operations. The fast Non-Dominated Sorting Genetic Algorithm (NSGA-II) was used to find the optimal bicycle demand scenario (number of stops and travel time reduced in 78 and 14%, respectively;  $CO_2$ ,  $NO_x$ , and HC decreased 9, 7, and 12%, respectively). It was also found that the number of conflicts was significantly reduced (–49%) with this latter layout even in maximum bicycle demand scenario (270 bph). However, roundabout layout lead to more severe conflicts and potential crashes. The analysis showed that bicycle demands of 75, 95 and 110 bph delivered good environmental and safety outcomes for the intersection.

Keywords: bicycle, microscale modelling, multi-objective optimization, traffic, emissions, safety.

#### Introduction and objectives

Cycling demand is increasing every day, notably in highdensity areas (Pucher, Buehler 2008). Cycling offers some important financial, health and social benefits to the users and environment. Bicycle is one of the most important alternatives to motor vehicle, and for short distances might be the best alternative to go easy and fast when the vehicles must stop because of traffic congestion. They are often quicker than motor vehicles over short distances of up to 5 km (DTMR 2018). These facts can explain why the rate of cycling in European small cities is higher than that observed in big cities (Pucher *et al.* 2011).

Signalized intersections are essential traffic control treatments that provide safe and efficient control of traffic congestion (Nguyen *et al.* 2016). Due to the complexity of intersections, high traffic volumes and densities, the interactions of vehicle-to-vehicle and bicycle-to-vehicle are steadily increased. This means that there is a higher risk for motor vehicles and bicycles crashes at intersec-

tions, compared to other urban network areas (Götschi et al. 2016).

Based on the current released data by National Highway Traffic Safety Administration (NHTSA 2017) 45000 cyclists were injured in the United States (57% of them at intersections) in 2015. Among 818 killed cyclists, which represents a 12% increase from 2014, 27% of fatalities occurred at intersections as well.

According to a research that was conducted by Haworth and Debnath (2013) in the State of Queensland of Australia from 2005 to 2009, almost half of the bicycle and motorcycle crashes occurred at intersections. Moreover, most Australian on-road bicycle fatalities (80%) involving motor vehicles in dry road conditions. Haworth and Debnath (2013) showed that drivers performed illegal manoeuvres in crashes at intersections with no traffic control (74.4% of these crashes involved bicycles) and operating traffic lights (18.6% of these involved bicycles).

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This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited. About European countries, a total of 2112 cyclists were killed in 2014, which represents 8% of total number of road accident fatalities. Moreover, 55% of cyclist fatalities occurred in urban areas and almost half of them (27%) were at junctions. Although, there was a significant variation among European countries, but in total bicycles had the third highest fatality rate at junctions.

Frequently reviewing traffic operations at intersections to find a compatible signal operation is one of the ways of improving traffic performance, safety, energy consumption or pollutant emissions levels. A shorter cycle length may result in poor progress while a longer cycle length may result in excessive delays and queue blockage problems (Ramadurai 2015). Kaplan *et al.* (2016) suggested a self-selection effect in which the intersection density is negatively correlated with the probability of cycling but is positively related to the distance cycled.

With respect to safety, number of conflicts between vehicles has significant relationship with the number of crashes in urban network. It is one of the main items that helps traffic engineers and managers to assess safety and to predict accidents before its occurrence (Van Hout 2008). Regarding the probable safety concerns associated with the number of conflicts between motor vehicles and cyclists, roundabouts and intersections with traffic lights and stop-controlled junctions are the critical traffic points (Kaplan et al. 2016). The frequency of conflicts between vehicles and cyclists at signalized intersections influence areas may increase under moderated or high cycling flows. For example, drivers typically wait to turn left to proceed even after the signal is red, cutting the path of vehicles processing on the green signal. In turn, cyclists filter through to the front of stopped queue and therefore they created conflicts with these drivers.

The study of conflicts between bicycle and motor vehicle might be more important than conflicts between motor vehicles since cyclists are more vulnerable and potential exposed to damage of a collision than motor vehicles' drivers (Götschi *et al.* 2016).

The analysis of traffic signals impact on traffic performance and emissions is extensive. The first studies, which date back to the 1970s, focused on the traffic performance and emissions impacts at intersections and can be found in the Environmental Protection Agency (EPA) reports (EPA 2002). Since then a lot of related studies have been performed both in a macroscale and microscale levels. For instance, on a microscale level, there have been several studies about the impact of signal optimization (Zhang *et al.* 2009; Khaki, Haghighat Pour 2014), level of service (LOS) improvement (Hurley, Kalus 2007; Barth, Boriboonsomsin 2008; Papson *et al.* 2012; Mok *et al.* 2013), and vehicular emissions at intersections (Coelho *et al.* 2005, 2006; Li *et al.* 2011; Zhou, Cai 2014).

There were several studies focused on impacts of cycling regarding safety concerns, traffic performance and emissions at intersections – about traffic performancesafety: (AlRaji 2015; Huang *et al.* 2013) and about traffic performance-emissions: (Zhang *et al.* 2009; Khaki, Haghighat Pour 2014; Zhou, Cai 2014). However, there is a lack of research focused on the impacts of cycling at intersections in an integrated way. Stevanovic *et al.* (2015) used multi-objective analysis to evaluate the traffic signal operation integrated impacts on three dimensions; traffic performance, emissions and safety at signalized intersections. However, the research did not include bicycle effects on intersection-specific operations.

The presence of bicycles may dictate a trade-off in the network. On one side, as the number of cyclists increases, the emissions generated by vehicles (assuming a same traffic demand) and the number of traffic conflicts increases. On the other side, the severity of conflicts tends to be high with low demand of bicycles because vehicles can attain high speeds. Since the existing trade-off between the results are associated with different performance measures, the multi-objective analysis can be useful and informative (Bai *et al.* 2012).

Roundabouts have the potential to reduce both the frequency and severity of crashes compared to traditional intersections (signalized or stop-controlled). Several studies carried out show the safety benefits of replacing signalized intersections by roundabouts (Sakshaug *et al.* 2010; Gross *et al.* 2013; Jensen 2013; Brilon 2016).

For instance, Groos *et al.* (2013) used Empirical Bayes method in an observational before–after study to estimate total and injury crashes after replacing 28 signalized intersections by roundabouts in the United States. There was a significant reduction in both total and injury crashes. It was also found that safety benefits were more pronounced in suburban than for urban conversions and for intersections with four approaches compared to those with three.

Jensen (2013) carried out a before–after study to analyse the safety results of converting 332 intersections to roundabouts in Denmark from 1995 to 2009. The total number of crashes and injuries decreased by 27 and 60%, respectively, while the total number of bicycles' crashes and injuries have increased by 65 and 40% respectively after conversions. However, the authors have mentioned that the safety results should not be generalized, since there were some significant differences among converted sites.

In contrast, the positive safety effects of converting intersections to roundabouts are not clear for bicycles. Some studies have been shown significant safety improvement (Schoon, Van Minnen 1994) while in other studies the effect of converting intersections to roundabouts on cyclists' safety was unclear or even negative (Sakshaug *et al.* 2010; Jensen 2013).

Since most of bicycle fatalities involve motor vehicles, studying the interaction between vehicles and specially bicycles-vehicles can help to a better accident prediction in urban areas. In addition, a deeper understanding of the bicycle impacts on road networks is essential to achieve a good balance among traffic performance, emissions and safety parameters. The novelty of this paper is the analysis of cyclists' demand effects at intersection influence area on traffic performance, emissions and conflicts between motor vehicles and cyclists. In addition, this research also identifies some trade-offs among these outputs. Next, several methods to improve intersection-specific operations are examined, namely: (1) replacement of the current intersection layout and (2) proper cyclist demand values according to the intersection-specific characteristics. This paper tests these methods in a real-world three-leg intersection.

Thus, the main objectives of this paper are:

- to evaluate the impact of increasing bicycle demand on traffic performance, global and local pollutant emissions and safety;
- to improve site-specific operation by proposing a different traffic control treatment (two-lane roundabout instead of traffic lights) for the intersection;
- to conduct a multi-objective analysis to find optimum bicycle demands to improve site specific emissions and safety.

### 1. Methodology

The present work uses a microscopic simulation platform of traffic (VISSIM) (PTV 2011) and emissions (Vehicle Specific Power - VSP) (EPA 2002) to analyse traffic operations, and to estimate carbon dioxide CO<sub>2</sub>, nitrogen oxides NO<sub>x</sub>, carbon monoxide CO and hydrocarbons HC emissions, respectively. The Surrogate Safety Assessment Methodology (SSAM) model was applied to estimate conflicts from vehicle-vehicle and vehicle-cyclist interactions and to compute surrogate safety measures. The Time-To-Collision (TTC) and the minimum Post-Encroachment Time (PET) were used to assess conflict severity, while the initial Deceleration Rate (DR), maximum speed (MaxS) and maximum relative speed difference (DeltaS) during the conflicts were used to represent the severity of the potential crashes. The intersection lanes operation was videotaped and necessary data were extracted from these tapes. In addition, a test-equipped vehicle with a Global Position System (GPS) collected second-by-second speed and acceleration/deceleration rates. Subsequently, data were input in the VISSIM model, which was calibrated and validated. Lastly, alternative scenarios were defined and then multi-objective optimization of bicycle demand was carried out. The flowchart of methodology is illustrated in Figure 1.

#### 1.1. Site selection and field data collection

The selected case study is a three-leg intersection controlled by traffic light located in the city of Aveiro (Portugal). It has potential traffic conflicts caused by left-turning vehicles from North to East directions. The segments that were considered to monitor traffic volumes included 300 m upstream the traffic light, as shown in Figure 2. Traffic signal is working based on fixed time operation with two-phases and the cycle time is 80 s. In Phase I, there are two protected turns, without opposing traffic, to right



Figure 1. Methodological framework



Figure 2. Layout of the case study with the identification of videotaping and traffic lights (source: https://www.bing.com/maps)

and left for minor lanes (diverging conflicts). Concerning Phase II, there are three types of vehicular conflicts (diverging, merging and crossing conflicts) for the major lane of the intersection. In addition, sequential conflicts may occur between two vehicles travelling in sequences (one following the other) in both major and minor roads. Figure 3a identifies above-mentioned conflicts. The average traffic volume was 1649 vehicles per hour (vph) and 9 bicycles per hour (bph). Furthermore, the operation of signal time represents 40 s green interval, 3 s yellow interval and 37 s red interval for Phase I that is the cycle time related to the major lanes (North and South directions).



Figure 3. Conflict types observed at three-leg signalized intersection (a) and relationship between observed and simulated conflicts (b)

Data were collected on three typical weekdays (Tuesday to Thursday) in April 2016, using one video camera (Figure 2). Field measurements were conducted in the morning peak hour, from 8:00 a.m. to 10:00 a.m. A camera was placed near the intersection, approximately 5 m above the ground, on the pedestrian bridge over the main road. The video tapes from the study site were used to acquire traffic and bicycle volumes and directional splits by approach, as well as to estimate the observed conflicts, as explained in the following sections. An equipped Light-Duty Vehicle (LDV) performed several trips at the major roads (mainly North-South and South-North movements). The-vehicle was equipped with a GPS data logger to record vehicle speed, distance travelled, and deceleration-acceleration rates in 1-second interval. Approximately 100 GPS travel runs were performed for this paper. GPS data were collected between 8:00 a.m. and 10:00 a.m. while traffic data were collected between 8:00 a.m. and 9:30 a.m.

#### 1.2. Traffic modelling

VISSIM software package was selected to simulate traffic operations (PTV 2011). The capability of VISSIM model in reproducing accurately traffic and bicycle operations at microscale for intersections is one of the main advantages of this traffic model (Mok *et al.* 2013; AlRajie 2015). Simulation experiments was made for the analysis period between 8:20 p.m. and 9:30 p.m. with a 10-min "warmup" period prior to 8:30 p.m. to load the road network adequately with corresponding traffic and bicycle flows.

VISSIM traffic model was initially calibrated to reproduce intersection specific operations from the data collected. Thus, driver behaviour parameters of the VISSIM traffic model were adjusted with the main purpose of assessing their impact on traffic volumes for each coded link. The calibrated driver behaviour measures included the car following - average standstill distance, additive and multiple part safety distance; gap acceptance - visibility, front and rear gaps and safety distance; and lane change - waiting time before diffusion, min-headway, safety distance reduction factor and maximum distance for cooperative breaking. The modified chi-squared statistics Geoffrey E. Havers (GEH) that incorporates both absolute and relative differences in comparison of estimated and observed volumes, was used as the calibration criteria (Buisson et al. 2014). More than 85% of the links must meet the GEH value lower than 4. In addition, the Mean Absolute Percent Error (MAPE) was used to measure the size of the error (deviation) for the observed and estimated data.

#### 1.3. Emissions estimations

The selected methodology to estimate the emissions is based on the concept of VSP. The scope of analysis is focused on vehicular emissions of  $CO_2$ ,  $NO_x$ , CO and HC. VSP is estimated from a second-by-second speed profile based on emission factors from typical LDVs (EPA 2002). Furthermore, VSP is associated with any speed trajectory and has capability to estimate the footprint of emissions at intersections with traffic lights and roundabouts (Coelho *et al.* 2006; Anya *et al.* 2013; Salamati *et al.* 2013, 2015). Equation provides the generic VSP equation from typical LDVs (EPA 2002):

$$VSP = v \cdot (1.1 \cdot a + 9.81 \cdot (a \cdot \tan(\sin(grade)))) + 0.123) + 0.000302 \cdot v^3,$$

where: VSP is vehicle specific power [kW/metric ton]; v is the instantaneous speed [m/s]; a is the acceleration/deceleration rate [m/s<sup>2</sup>]; *grade* is road grade (decimal fraction).

Each VSP bin refers to one of 14 modes. Each VSP mode is defined by a range of VSP values, which are associated to an emission rate. Each calculation of VSP results in a unique classification to a VSP mode (Anya *et al.* 2013; Coelho *et al.* 2009).

The following fleet composition based on the Portuguese car fleet distribution (ACAP 2014) was considered: 44.7% of light duty gasoline vehicles, 34.3% of light duty diesel vehicles and 21.0% of light commercial diesel vehicles. Transit buses, motorcycles and heavy-duty trucks represented only 1% of traffic composition and therefore were excluded from the emissions calculations. The effects of the grade were not included since terrain is flat.

#### 1.4. Safety model

For the safety assessment approach the software developed by the Federal Highway Administration – FHWA (Surrogate Safety Assessment Model – SSAM) was used (Gettman *et al.* 2008).

Traditionally, traffic safety assessment heavily relies on crash data analysis, in which the number or consequences

of crashes were used as measures of effectiveness to evaluate the safety performance of traffic facilities (Huang *et al.* 2013). SSAM automates traffic conflict analysis by processing vehicles trajectories from a microscopic traffic model as VISSIM. For each simulation, SSAM stores the trajectories of vehicles and bicycles from the traffic model and determines whether an interaction between vehicle-to-vehicle or vehicle-to-bicycle satisfies the condition to be deemed a traffic conflict (Fernandes *et al.* 2015). The authors used the TTC as a safety indicator to assess whether a vehicle-vehicle and vehicle-bicycle interaction can result in conflict. TTC is a measure of conflict severity (low values of TTC indicate high severe conflicts).

The threshold value of TTC was set 1.5 s, as suggested for urban areas by Gettman *et al.* (2008). Moreover, TTC  $\leq$  1.5 s was defined as a being a serious conflict point for motor vehicles and bicycles at intersections by Widdel and Kaster (1981). However, other authors (Getman *et al.* 2008) divided TTC values in different severity ranges (e.g. TTC  $\leq$  0.5 s, TTC  $\leq$  1.0 s, and TTC  $\leq$  1.5 s). It must be noted that conflicts with TTC value of 0.5 s (more severe) have higher probability in resulting in a crash than conflicts with a value of 1.0 s. Based on these facts, the authors also analysed the impact of conflicts with TTC lower than 0.5 s in the analysis of the alternative scenarios.

Also, the research team used the minimum PET to assess conflict severity, and the initial DR), MaxS and DeltaS during the conflict to represent the severity of the potential crashes (Gettman *et al.* 2008).

#### 1.5. Alternative scenarios

In order to improve traffic operation and safety levels, two alternative scenarios were defined based on the potential role of bicycles:

- Scenario I: Increasing number of bicycles from 9 to 270 with 30 bph increments for baseline scenario and replacing number of motor vehicles based on occupancy ratio of motor vehicles (1.2 person per 1 motor vehicle) for each volume of bicycle demand (Schultz et al. 2015). It should be noted that, increasing the number of bicycles and replacing the motor vehicles was done based on their distribution rate at each lane of the network. This range of values was justified by the new project implementation (PO SEUR 2016) that aims to encourage as many people as possible to use the bicycle as a regular transport mode, and as a result 240 new bicycles will be provided at campus area of the University of Aveiro. Furthermore, assuming 30 more active bicycles for current situation the impact of new situation was analysed up to 270 bph;
- Scenario II: The existing traffic control was replaced by a two-lane roundabout and evaluating the impacts of new bicycles increment as same as the first proposed scenario. The number of bicycles increased for proposed roundabout and the number of motor vehicles replaced based on occupancy ratio between bicycle and motor vehicle, as it was defined before,

#### 1.6. Multi-objective optimization

The fast Non-Dominated Sorting Genetic Algorithm (NSGA-II) was applied to conduct the multi-objective analysis (Moussouni *et al.* 2007). NSGA-II was reported as one of the effective algorithm in finding a good approximation of an optimal Pareto front (Konak *et al.* 2006).

NSGA-II can be outlined in the following steps (Deb et al. 2002):

- the population (optimal bicycle demand) was initialized considering both objective variables range values and model constraints (in this case, bicycle demand);
- a non-domination criteria was applied to sort the population. Thus, a binary tournament selection based on the rank and crowding distance process was performed to select the parents (bicycle demand in the first generation) from the population that in this case is defined as bicycle optimal demand. An individual (bicycle demand) is selected in the rank if is smaller than the other individual or if crowding distance is greater than the other (this guarantees diversity in the optimal solutions);
- the selected population generated offspring by applying crossover and mutation rates, and further parents and offspring were merged to choose the best individuals in the population (elitism).

After reaching stopping criteria (number of generations) and optimal solutions found in the Pareto Approximate Front, the procedure was finished.

Among the safety indicators, the TTC and MaxS were selected to represent the severity of conflicts and collision respectively. Moreover, the global and local emissions were selected to analyse against these safety indicators to find the existent differences between them.

The following tests were performed: (1)  $CO_2$  versus MaxS; (2)  $CO_2$  versus 1/TTC; (3) CO versus MaxS; (4) CO versus 1/TTC; (5)  $NO_x$  versus MaxS; (6)  $NO_x$  versus 1/TTC; (7) HC versus MaxS; (8) HC versus 1/TTC. Three main reasons supported these tests: (1) to identify hypothetic differences in optimal bicycle demand differences between global ( $CO_2$ ) and local (CO,  $NO_x$  and HC) pollutant emissions; (2) to examine the differences between severe conflicts (TTC) and potential crashes (MaxS); and (3) a two-dimensional (2D) front allows visualizing optimal bicycle demand data sets on an easy way, instead of considering all variables simultaneously.

Once these tests perform, the optimal bicycle data set values were obtained for each case. A set of 15 optimal solutions was considered for this analysis. In addition, the authors gave the same relevance (weight) to all objective variables during the optimization procedure.

## 2. Results and discussion

# 2.1. Traffic and safety model calibration and validation

Calibration of VISSIM parameters was made based on estimated and observed traffic volumes with 15 random seed runs (Hale 1997). A good fit between observed and estimated volumes was obtained ( $R^2 = 0.99$ ) using a linear regression analysis. Also, all the 36 links recorded a GEH value lower than 4, which satisfied the calibration criteria proposed by Dowling *et al.* (2004), while MAPE values were lower than 4%.

Regarding model validation, a comparison of observed and simulated travel time at the two main lanes of intersection, for North-South and South-North movements, was conducted using 100 floating car runs (Dowling *et al.* 2004). Observed travel time were obtained by vehicle dynamic data collected from through movements (North-South and South-North) at major lanes while simulated travel time were computed from vehicle recording tool of VISSIM (PTV 2011). The difference between observed and estimated average travel time was not statistically significant at a 5% significance level, which demonstrated the accuracy of the traffic model in representing traffic and bicycle operations along the study domain.

Lastly, a comparison of SSAM conflicts and observed was conducted. Thus, the videotapes recordings were reviewed for several times to obtain the traffic conflicts (Huang *et al.* 2013), as shown in Figure 3. The authors ran the VISSIM simulation model and the results classified with 15 min time intervals. To be consistent with the conflict types computed by SSAM, the observed conflicts were classified into three types: (1) rear-end conflicts, (2) lane-change conflicts, and (3) crossing conflicts.

Linear regression analysis was used to identify if the simulated traffic conflicts provided reasonable estimates for the observed traffic conflicts. It was found that the relationships between the simulated and the observed conflicts were statistically significant and acceptable (Figure 3b).

#### 2.2. Baseline and alternative Scenario I

This section presents the main results for baseline and alternative Scenario I. Performance measures such as average vehicle and bicycle travel time and number of stops were given from the vehicle recording tool of the VISSIM (PTV 2011) while TTC, PET, MaxS, DeltaS and DR were computed in SSAM (Gettman *et al.* 2008).

Table 1 lists traffic performance, emissions and safety outputs for above scenarios by bicycle demand scenario. As the number of bicycles increased from 9 to 270 bicycles, the emissions generated by vehicles decreased (on average 9, 6, 6 and 8% for  $CO_2$ , CO,  $NO_x$  and HC) and concomitantly the travel time increased about 5% for the motor vehicles. This happens because motor vehicles in the network are more impacted by cyclists and so they spend more time, individually in the road.

The bicycles did not follow the same trend. Due to bicycles increment at the study domain, the number of stops increased from 4 (9 bph) to 148 (270 bph), and the travel time increased from 94.1 to 105.5 s respectively as well. The increase in the number of bicycles resulted in more conflicts (27% more in the 270 bph scenario compared to the existing conditions). The analysis results showed that there was no stable trend in TTC and PET values as bicycle demand increased (these measures were practically

Vahiclas	Dollutants	Baseline	Bicycles [bph]								
venicies	ronutants	9 bicycles [bph]	30	60	90	120	150	180	210	240	270
	CO <sub>2</sub> [kg]	173	172	170	170	168	162	164	162	160	158
Motor vehicles	CO [g]	227	228	226	227	225	217	222	219	216	214
	NO <sub>x</sub> [g]	511	512	509	510	506	487	497	491	485	481
	HC [g]	8.5	8.4	8.3	8.3	7.9	8.1	8.0	7.9	7.8	7.8
	CO <sub>2</sub> [g/km]	260	263	264	268	270	269	273	274	277	279
	Stops [No]	928	853	822	814	817	775	795	795	782	782
	Travel time [s/veh]	48.1	48.1	48.1	48.7	49.7	49.0	49.6	50.0	50.2	50.6
	Speed (average) [km/h]	44.5	44.1	44.0	43.8	43.4	43.1	42.7	42.5	42.4	42.0
Bicycles	Stops [No]	4	15	29	48	67	84	102	112	133	148
	Travel time [s/veh]	94.1	102.8	100.7	103.2	104.1	103.1	103.8	104.4	105.0	105.5
	Speed (average) [km/h]	20.0	19.8	20.0	19.8	19.6	19.6	19.5	19.5	19.3	19.3
	Conflicts [No]	81	82	86	86	92	92	96	98	98	103
Safata	TTC [s]	1.14	1.17	1.17	1.14	1.15	1.13	1.14	1.14	1.13	1.14
	PET [s]	1.61	1.61	1.63	1.55	1.51	1.45	1.45	1.50	1.46	1.51
Salety	MaxS [m/s]	6.39	6.54	6.64	7.00	6.92	7.20	7.05	7.19	7.38	7.14
	DeltaS [m/s]	5.28	5.20	5.07	5.10	4.99	5.18	4.98	4.98	4.96	4.86
	DR [m/s <sup>2</sup> ]	-2.31	-2.37	-2.35	-2.46	-2.38	-2.45	-2.45	-2.44	-2.44	-2.33

Table 1. Summary results of baseline and alternative Scenario I

constant). However, the rate of severe serious conflicts (TTC  $\leq 0.5$  s) slightly increased from 2% (9 bph) to 5% (270 bph). The severity of potential collisions increased as MaxS point of view (more than 10%), but DR (in absolute terms) and DeltaS did not vary among bicycle demand scenarios.

#### 2.3. Alternative roundabout Scenario II

The findings confirmed some improvements on traffic performance and emissions using two-lane roundabout, as presented in Table 2. The results showed that in the first demand (30 bph) the emissions in CO<sub>2</sub>, NO<sub>x</sub> and HC, reduced 1.2, 0.6 and 3.5% respectively, compared to traffic light solution (baseline). However, CO emissions were higher in roundabout scenario (2.6%), which may be due to the acceleration episodes that vehicles experienced both in the downstream and circulating areas (before exiting) of the roundabout. Continuing to increase the number of bicycles, and replacing more motor vehicles as well, the reduction in emissions was more pronounced. For instance, in last demand scenario (270 bph) CO<sub>2</sub>, NO<sub>x</sub>, and HC vehicular emissions decreased 9.0, 7.0 and 12% respectively replacing traffic light by two-lane roundabout. Roundabout solution was also very effective in terms of traffic performance measures (its implementation allowed the number of stops and travel time to be reduced in more than 78 and 14%, respectively in the last demand). Concerning the bicycles, the traffic performance measures dictated notable reduction in the number of stops while travel time did not vary after implementing roundabout.

The findings confirmed significant improvements on safety regarding the number of conflicts. The results showed that in the first demand (30 bph) and last demand (270 bph), the number of conflicts reduced 74 and 49%, respectively, compared to traffic light solution (baseline). However, there was not a consensus about safety variables. Both TTC and PET decreased, which means more severe conflicts. For instance, the total number of severe conflicts (TTC  $\leq$  0.5 s) increased from 7% (9 bph) to 16% (270 bph). Notably, roundabout scenario recorded high MaxS values, especially under high-bicycle demand scenarios (30% higher than those obtained in the signalized solution). This is caused by weaving manoeuvres of vehicles at the circulating area of the roundabout.

In summary, the benefits of roundabout layout in traffic performance and emissions measures were pronounced while the safety benefits were unclear. Roundabout caused more severe conflicts as the number of bicycle users increased, as well as more severe potential crashes (as MaxS criteria) compared to baseline scenario. However, the difference in surrogate measures between scenarios was not statistically significant at a 5% significance level (p-value < 0.05). Several explanations support these results. First, most of conflicts in the baseline occurred near the traffic light and during the left-turning movements (vehicles are waiting for a crossable gap of South approach vehicles), as depicted in Figure 4. In the roundabout, the severity of conflicts can increase due to circulating area of roundabout that drivers experience moderate speed and lane change (as a presence of other vehicles or bicycles), which does not occur in the baseline scenario.

The Figure 4 also shows that the extent of hotspot conflicts locations is higher in Scenario I than in Scenario II. This is explained by the longer queues that vehicles experience due to red signal both in main roads and minor roads.

Vahiclas	Dollutants	Baseline	Bicycles [bph]									
venicies	Pollutants	9 bicycles [bph]	9	30	60	90	120	150	180	210	240	270
Motor vehicles	CO <sub>2</sub> [kg]	173	172	171	168	168	167	163	163	161	159	157
	CO [g]	227	233	233	229	229	229	225	224	222	219	217
	NO <sub>x</sub> [g]	511	508	508	500	501	499	490	489	486	479	475
	HC [g]	8.5	8.2	8.2	8.0	8.0	8.0	7.8	7.8	7.7	7.6	7.5
	CO <sub>2</sub> [g/km]	260	193	195	196	198	200	201	201	203	203	205
	Stops [No]	928	107	111	106	195	197	199	200	200	202	204
	Travel time [s/veh]	48.1	39.1	39.4	39.6	39.9	40.3	40.4	40.5	47.4	41.3	41.3
	Speed (average) [km/h]	44.5	51.1	50.8	50.6	50.3	49.9	49.7	49.6	49.2	48.9	48.8
	Stops [No]	4	1	0	1	3	3	4	4	6	8	11
Bicycles	Travel time [s/veh]	94.1	94.4	94.6	94.6	95.7	96.1	95.1	96.4	96.8	94.3	97.5
	Speed (average) [km/h]	20.0	22.2	21.8	22.3	22.4	21.6	21.6	21.6	21.4	21.4	21.3
	Conflicts [No]	81	19	21	24	22	32	31	35	36	37	41
C a fata	TTC [s]	1.14	1.15	1.11	1.11	1.08	1.08	0.98	1.06	1.03	1.03	1.03
	PET [s]	1.61	1.40	1.23	1.30	1.13	1.18	1.03	1.08	1.04	1.06	1.05
Salety	MaxS [m/s]	6.39	6.72	7.84	8.30	8.90	8.78	9.52	9.61	9.53	9.63	9.64
	DeltaS [m/s]	5.28	4.94	4.85	4.83	5.01	4.93	4.88	4.96	5.11	4.78	5.13
	$DR[m/s^2]$	-2.31	-2.17	-2.57	-2.52	-2.48	-2.44	-2.54	-2.61	-2.73	-2.60	-2.46

Table 2. Summary results of alternative roundabout Scenario II



Figure 4. Hot spot conflicts location for the baseline (a) and roundabout (b) scenarios: crossing conflicts – red colour; lane change – blue colour; rear end – yellow colour

The implementation of the roundabout also was effective in eliminating the traffic conflicts at the exits of the intersection.

Overall, three main conclusions can be drawn from above-mentioned results: (1) two-roundabout improved traffic performance and some pollutant emissions at the selected intersection regardless of the number of bicycle users; (2) safety benefits of this layout were less pronounced under high-bicycle demands; and (3) increasing the number of bicycle users resulted in a degradation of traffic and cyclist's operations from a certain level of demand. With these concerns in mind, the research team decided to develop a multi-objective analysis to search the optimal number of bicycles to improve emissions and safety variables for the best intersection solution (two-lane roundabout).

#### 2.4. Multi-objective optimization

The literature review has shown that, at least for bicycles, it is unclear if roundabouts represent better safety performance than intersections (Sakshaug *et al.* 2010; Jensen 2013). Specifically, this paper represented a multi-objective perspective to find the balance among traffic performance, emissions and safety outputs.

Thus, thus section presents the main multi-objective results for the alternative scenario. Preliminary analysis of genetic algorithm showed that 100 iterations were sufficient to reach convergence of the optimal bicycle demand solutions. In addition, the crossover and mutation were set at 90 and 15% respectively.

The multi-objective resulted solutions from this scenario are illustrated in Figure 5 that shows the Pareto fronts estimated from final population (after 100 generations).

For this Scenario II, two dimensions were CO<sub>2</sub>, CO, NO<sub>x</sub>, and HC emissions (x-axis), and MaxS and the inverse of the TTC (y-axis) were drawn based on bicycles numbers. It is to be noted that the unit of TTC is defined in inverse second (s<sup>-1</sup>) because longer TTC represents less severe conflicts. In Figure 5 for each level of bicycle demand, a two-dimensional Pareto frontier with two objective functions – emissions (x-axis), and safety measure (y-axis) – is defined. Data labels indicate the set of points that represent optimal solutions as bicycle demand value. Clearly, the trade-off between outputs was observed. The optimal solutions, which minimized vehicular emissions were allocated to the upper-left, while the optimal bicycle demand values, which led to the minimal conflict severity and potential crashes were assigned lower-right. For instance, adopting a solution number 8 (bicycle demand of 74 bph), from Figure 5a, which is closest to the abscissa of the graph will conduct to 2.3% of CO2 decreases and 31.5% increase in MaxS, compared to traffic light solution (baseline with 9 bph). Moreover, adopting a solution number 8 (bicycle demand of 68 bph), from Figure 5f, will be conducted to 1.8% of  $NO_x$  decreases and 4.9% increase in 1/TTC, compared to traffic light solution (baseline with 9 bph). Other optimal solutions also suggested 90, 120 or 125 bph to achieve a good balance between emissions and safety.

The results also confirmed that the differences in the optimal bicycle demands were very small between global  $(CO_2)$  and local  $(CO, NO_x \text{ and } HC)$  pollutants, or between safety parameters. Some solutions were found at 75, 95 and 110 bph that seems to represent the better compromise to assure good levels of comfort for bicycles and reduce pollutant emissions.

Summary results of multi-objective solutions for alternative roundabout Scenario II presented in Table 3.



Figure 5. The approximate final Pareto front for alternative Scenario II: a – CO<sub>2</sub>, MaxS; b – CO<sub>2</sub>, 1/TTC; c – CO, MaxS; d – CO, 1/TTC; e – NO<sub>x</sub>, MaxS; f – NO<sub>x</sub>, 1/TTC; g – HC, MaxS; h – HC, 1/TTC

# Conclusions

This paper addressed the impact of bicycle demand at a three-leg signalized intersection on traffic performance, vehicular emissions and safety. In addition, the paper proposed a solution for the intersection by replacing existing traffic light for a two-lane roundabout and compare the results between layouts subjected to increments in the number of bicycle users (and decrease in the number of motor vehicles based on occupancy ratio). Finally, a multi-objective genetic algorithm was used to improve site-specific emissions and safety outcomes.

No	Bicycle demand [bph]	CO <sub>2</sub> [g/km]	MaxS [m/s]	Bicycle demand [bph]	CO [mg/km]	MaxS [m/s]	Bicycle demand [bph]	NO <sub>x</sub> [mg/km]	MaxS [m/s]	Bicycle demand [bph]	HC [mg/km]	MaxS [m/s]
1	270	160	9.5	270	221.1	9.5	270	483	9.5	270	7.6	9.5
2	164	163	9.5	165	224.3	9.5	162	490	9.5	270	7.6	9.5
3	157	164	9.5	149	225.2	9.4	165	490	9.5	165	7.8	9.5
4	143	165	9.4	128	226.4	9.2	144	493	9.4	147	7.9	9.4
5	122	166	9.1	115	227.3	9.0	120	496	9.1	133	7.9	9.3
6	118	166	9.1	91	228.7	8.7	94	499	8.8	123	7.9	9.1
7	94	167	8.8	78	229.4	8.5	78	501	8.5	101	8.0	8.9
8	74	169	8.4	73	229.7	8.4	69	502	8.3	82	8.0	8.6
9	63	169	8.2	66	230.1	8.3	61	503	8.2	79	8.0	8.5
10	44	170	7.9	49	231.1	8.0	53	504	8.0	69	8.0	8.4
11	48	170	7.9	34	232.0	7.7	31	507	7.6	51	8.1	8.0
12	28	171	7.5	25	232.6	7.4	19	509	7.3	18	8.2	7.3
13	12	172	7.1	23	232.7	7.4	18	509	7.3	20	8.2	7.3
14	9	172	7.0	9	233.7	7.0	9	510	7.0	9	8.2	7.0
15	9	172	7.0	9	233.7	7.0	9	510	7.0	9	8.2	7.0
No	Bicycle demand [bph]	CO <sub>2</sub> [g/km]	1/TTC [s <sup>-1</sup> ]	Bicycle demand [bph]	CO [mg/km]	1/TTC [s <sup>-1</sup> ]	Bicycle demand [bph]	NO <sub>x</sub> [mg/km]	1/TTC [s <sup>-1</sup> ]	Bicycle demand [bph]	HC [mg/km]	1/TTC [s <sup>-1</sup> ]
No 1	Bicycle demand [bph] 270	CO <sub>2</sub> [g/km] 157	1/TTC [s <sup>-1</sup> ] 0.967	Bicycle demand [bph] 270	CO [mg/km] 218.1	1/TTC [s <sup>-1</sup> ] 0.967	Bicycle demand [bph] 270	NO <sub>x</sub> [mg/km] 476	1/TTC [s <sup>-1</sup> ] 0.967	Bicycle demand [bph] 270	HC [mg/km] 7.5	1/TTC [s <sup>-1</sup> ] 0.967
No 1 2	Bicycle demand [bph] 270 165	CO <sub>2</sub> [g/km] 157 163	1/TTC [s <sup>-1</sup> ] 0.967 0.967	Bicycle demand [bph] 270 165	CO [mg/km] 218.1 224.3	1/TTC [s <sup>-1</sup> ] 0.967 0.967	Bicycle demand [bph] 270 270	NO <sub>x</sub> [mg/km] 476 476	1/TTC [s <sup>-1</sup> ] 0.967 0.967	Bicycle demand [bph] 270 165	HC [mg/km] 7.5 7.8	1/TTC [s <sup>-1</sup> ] 0.967 0.967
No 1 2 3	Bicycle demand [bph] 270 165 151	CO <sub>2</sub> [g/km] 157 163 164	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.963	Bicycle demand [bph] 270 165 161	CO [mg/km] 218.1 224.3 224.5	1/TTC [s <sup>-1</sup> ] 0.967 0.966	Bicycle demand [bph] 270 270 165	NO <sub>x</sub> [mg/km] 476 476 490	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967	Bicycle demand [bph] 270 165 149	HC [mg/km] 7.5 7.8 7.9	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962
No 1 2 3 4	Bicycle demand [bph] 270 165 151 141	CO <sub>2</sub> [g/km] 157 163 164 165	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956	Bicycle demand [bph] 270 165 161 130	CO [mg/km] 218.1 224.3 224.5 226.3	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955	Bicycle demand [bph] 270 270 165 140	NO <sub>x</sub> [mg/km] 476 476 490 493	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967 0.959	Bicycle demand [bph] 270 165 149 120	HC [mg/km] 7.5 7.8 7.9 7.9	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950
No 1 2 3 4 5	Bicycle demand [bph] 270 165 151 141 131	CO <sub>2</sub> [g/km] 157 163 164 165 165	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.963 0.956 0.955	Bicycle demand [bph] 270 165 161 130 104	CO [mg/km] 218.1 224.3 224.5 226.3 227.9	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.966 0.955 0.942	Bicycle demand [bph] 270 270 165 140 127	NO <sub>x</sub> [mg/km] 476 476 490 493 495	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967 0.959 0.954	Bicycle demand [bph] 270 165 149 120 103	HC [mg/km] 7.5 7.8 7.9 7.9 7.9 8.0	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942
No 1 2 3 4 5 6	Bicycle demand [bph] 270 165 151 141 131 114	CO <sub>2</sub> [g/km] 157 163 164 165 165 166	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948	Bicycle demand [bph] 270 165 161 130 104 97	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939	Bicycle demand [bph] 270 270 165 140 127 112	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967 0.959 0.954 0.947	Bicycle demand [bph] 270 165 149 120 103 83	HC [mg/km] 7.5 7.8 7.9 7.9 7.9 8.0 8.0	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930
No 1 2 3 4 5 6 7	Bicycle demand [bph] 270 165 151 141 131 114 97	CO <sub>2</sub> [g/km] 157 163 164 165 165 166 167	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938	Bicycle demand [bph] 270 165 161 130 104 97 89	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934	Bicycle demand [bph] 270 270 165 140 127 112 95	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967 0.959 0.954 0.947 0.938	Bicycle demand [bph] 270 165 149 120 103 83 72	HC [mg/km] 7.5 7.8 7.9 7.9 7.9 8.0 8.0 8.0 8.0	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930 0.923
No 1 2 3 4 5 6 7 8	Bicycle demand [bph] 270 165 151 141 131 114 97 90	CO <sub>2</sub> [g/km] 157 163 164 165 165 166 166 167 168	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935	Bicycle demand [bph] 270 165 161 130 104 97 89 83	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 228.8 229.6	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.942	Bicycle demand [bph] 270 270 165 140 127 112 95 68	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920	Bicycle demand [bph] 270 165 149 120 103 83 72 58	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930 0.923 0.913
No 1 2 3 4 5 6 7 8 9	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73	CO <sub>2</sub> [g/km] 157 163 164 165 165 165 166 167 168 169	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935 0.923	Bicycle demand [bph] 270 165 161 130 104 97 89 83 83 75	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.966 0.955 0.942 0.939 0.934 0.942 0.930	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1 8.1	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900
No 1 2 3 4 5 6 7 8 9 10	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73 69	CO <sub>2</sub> [g/km] 157 163 164 165 165 165 166 167 168 169 169	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935 0.923 0.920	Bicycle demand [bph] 270 165 161 130 104 97 89 83 75 53	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1 230.9	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.942 0.930 0.908	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54 31	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504 507	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909 0.890	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42 31	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1 8.1 8.1	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900 0.890
No 1 2 3 4 5 6 7 8 9 10 11	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73 69 56	CO <sub>2</sub> [g/km] 157 163 164 165 165 166 167 168 169 169 169 170	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935 0.923 0.920 0.911	Bicycle demand [bph] 270 165 161 130 104 97 89 83 75 53 43	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1 230.9 231.5	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.934 0.942 0.930 0.908 0.901	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54 31 25	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504 507 508	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909 0.890 0.884	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42 31 29	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1 8.1 8.1 8.1	1/TTC [s <sup>-1</sup> ] 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900 0.890 0.888
No 1 2 3 4 5 6 7 8 9 10 11 12	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73 69 56 47	CO <sub>2</sub> [g/km] 157 163 164 165 165 165 166 167 168 169 169 169 170 170	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.963 0.955 0.955 0.948 0.938 0.935 0.923 0.923 0.920 0.911 0.903	Bicycle demand [bph] 270 165 161 130 104 97 89 83 75 53 43 22	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1 230.9 231.5 232.7	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.942 0.930 0.908 0.901 0.881	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54 31 25 20	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504 507 508 508	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909 0.890 0.884 0.880	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42 31 29 18	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1 8.1 8.1 8.1 8.1 8.2	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900 0.890 0.888 0.878
No 1 2 3 4 5 6 7 8 9 10 11 12 13	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73 69 56 47 26	CO <sub>2</sub> [g/km] 157 163 164 165 165 166 167 168 169 169 170 170 171	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935 0.923 0.920 0.911 0.903 0.886	Bicycle demand [bph] 270 165 161 130 104 97 89 83 75 53 43 22 15	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1 230.9 231.5 232.7 233.1	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.942 0.930 0.908 0.901 0.881 0.875	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54 31 25 20 10	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504 507 508 508 508 510	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909 0.890 0.884 0.880 0.870	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42 31 29 18 18	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.1 8.1 8.1 8.1 8.1 8.2 8.2	1/TTC [s <sup>-1</sup> ] 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900 0.890 0.888 0.878 0.878
No 1 2 3 4 5 6 7 8 9 10 11 12 13 14	Bicycle demand [bph] 270 165 151 141 131 114 97 90 73 69 56 47 26 9	CO <sub>2</sub> [g/km] 157 163 164 165 165 166 167 168 169 169 169 170 170 170 171	1/TTC [s <sup>-1</sup> ] 0.967 0.963 0.956 0.955 0.948 0.938 0.935 0.923 0.920 0.911 0.903 0.886 0.869	Bicycle demand [bph] 270 165 161 130 104 97 89 83 75 53 43 22 15 9	CO [mg/km] 218.1 224.3 224.5 226.3 227.9 228.3 228.8 229.6 229.1 230.9 231.5 232.7 233.1 233.5	1/TTC [s <sup>-1</sup> ] 0.967 0.966 0.955 0.942 0.939 0.934 0.934 0.942 0.930 0.908 0.901 0.881 0.875 0.869	Bicycle demand [bph] 270 270 165 140 127 112 95 68 54 31 25 20 10 99	NO <sub>x</sub> [mg/km] 476 476 490 493 495 497 499 502 504 507 508 508 508 510 510	1/TTC [s <sup>-1</sup> ] 0.967 0.967 0.959 0.954 0.947 0.938 0.920 0.909 0.890 0.884 0.880 0.870 0.869	Bicycle demand [bph] 270 165 149 120 103 83 72 58 42 31 29 18 18 9	HC [mg/km] 7.5 7.8 7.9 7.9 8.0 8.0 8.0 8.0 8.0 8.1 8.1 8.1 8.1 8.1 8.1 8.2 8.2 8.2	1/TTC [s <sup>-1</sup> ] 0.967 0.962 0.950 0.942 0.930 0.923 0.913 0.900 0.890 0.888 0.878 0.878 0.869

Table 3. Summary results of multi-objective solutions for alternative roundabout Scenario II

The findings showed a reduction in emissions (from 6 to 9%, depending on the pollutant) by increasing the number of bicycle users from 9 to 270 for signalized intersection, while bicycle travel time increased from 94.1 to 105.5 s. The conversion of traffic lights by a two-lane roundabout lead to a significant reduction in  $CO_2$ ,  $NO_x$  and HC criteria regardless of the bicycle demand scenario. The proposed layout also reduces the number of vehicle stops and traffic conflicts, but this was not hold for surrogate measures (more severe conflicts and potential crashes in roundabout solution). The multi-objective analysis suggested that optimal bicycle demand solutions of 75, 95 and 110 bph dictated a good compromise between global and local pollutants, and the inverse of TTC and MaxS safety variables to

assure comfort levels for all road users in the roundabout.

The main contributions of this paper were the following:

- to understand suitable bicycle demand values at intersections to assure a good balance among traffic performance, emissions and safety outputs;
- to include conflict and potential crash severities criteria to account site specific safety concerns;
- to identify the trade-off among environmental, delay and bicycle safety outcomes;
- to provide some design principles for transportation planners and local governments (e.g., intersection layout, bicycle demand, traffic demand or speed) to mitigate the mutual effect of drivers and cyclists on intersection operations.

Although this study provides a multi-objective tool to assess the safety of cyclists, traffic performance and emissions with varied levels of bicycles demand at intersections, there are some limitations that must be highlighted. Different traffic volumes, speed limits, type of intersection or roundabout design features may result in different safety outcomes. The paper did not include illegal manoeuvres of drivers or cyclists, and pedestrians' presence that can decrease the performance of roundabout. Lastly, the simulation has some limitations in representing cycling and shared space situations. For example, VISSIM allows modelling cyclists using a similar behaviour models as for motor vehicle traffic, simply adapting the parameters of the vehicle behaviour models to reflect the lower speeds of bicycles; this may be not work for all conditions, since cyclists have unique characteristics that are difficult or impossible to represent with model car following models. This should be improved in future work.

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