

1 **Analysis of Road User Behavior and Safety during New York City's**
2 **Summer Streets Program**

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1 **Abstract**

2 Automated computer vision video analysis techniques were used to analyze video
3 data during the operation of New York City’s Summer Streets Program at a major
4 signalized intersection. The main objectives of this study were to: 1) diagnose
5 pedestrian and cyclist safety issues during the “shared space” operation and 2)
6 demonstrate the feasibility of the automatic extraction of road user (e.g.
7 pedestrian, runner, rollerblader, or cyclist) data required for microscopic
8 behavior analysis. Road users’ speeds and pedestrian gait parameters (step
9 frequency and step length) were automatically extracted and analyzed. The
10 results show that pedestrian walking speed was highest during the Summer Street
11 operation (1.49 ± 0.54 m/s) as they had more street space to use and slowest
12 during normal operations (1.30 ± 0.22 m/s). Bike speeds were low during the
13 Summer Streets event (3.62 ± 0.97 m/s), likely because of interaction with
14 pedestrians, but increased during normal traffic operations. Pedestrians and
15 cyclists moving in groups tended to be slower, confirming results found in
16 previous studies. The safety analysis was conducted using traffic conflict
17 techniques (TCT). It was observed that the lowest rate of conflicts between
18 pedestrians and cyclists and between cyclists was found to be during Summer
19 Streets operations. In addition, an analysis of spatial violations show that some
20 road users were not observing traffic rules in the transition period after Summer
21 Streets ceased to operate.

22 **Introduction**

23 Encouraging sustainable modes of transportation and exercise—such as walking,
24 rollerblading, jogging and cycling—is important to build healthy and livable
25 communities. Policies promoting non-motorized modes of travel and events
26 encouraging active lifestyles is helpful in motivating users to shift from motorized
27 transportation modes to other sustainable modes. Active transportation is a
28 simple way to increase the physical activity level of the population, contributing
29 to improvements to overall public health. However, transportation safety
30 problems are a serious concern for vulnerable road users. Vulnerable road users,
31 such as pedestrians and cyclists, are subject to higher safety risks and usually
32 represent the highest share of fatal road collisions [1]. They have higher per-mile
33 casualty rate than car travel, yet pose minimal risk to other road users. Overall,
34 traffic casualty rates are found to decline as walking and cycling increase in a
35 community because drivers are more cautious [2]. Safety experts study pedestrian
36 and cyclist behaviors and develop appropriate techniques to diagnose pedestrian
37 safety issues and recommend pedestrian safety improvements. To enhance the
38 safety of vulnerable road users, it is necessary to develop an understanding of this
39 behavior through developing better tools to study vulnerable road user behavior.

40 In August 2008, New York City initiated the Summer Streets program. This event
41 opened Park Avenue from Brooklyn Bridge to Central Park (approximately 7
42 miles/11 km) to pedestrians, cyclists, rollerbladers, and joggers, and created
43 vehicle-free streets on three consecutive Saturdays from 7AM to 1PM to promote
44 sustainable forms of transportation. All activities at Summer Streets were

1 provided free of charge and designed for people of all ages and ability to share the
2 streets respectfully [3], creating a “shared space.”

3 In this study, video data was collected at a busy signalized intersection in New
4 York City (Park Avenue South and East 29th Street) on August 2nd, 2014 during
5 three operating periods: the Summer Streets event, the transition period from
6 Summer Streets to normal operation and the normal operation. The purpose of
7 this data collection is to demonstrate an automated methodology to diagnose
8 safety issues of non-motorized road users during a busy car-free “shared space”
9 environment. The potential for automatic extraction of non-motorized road users’
10 data required to understand their behavior was also examined. In the past few
11 years, video data collection has been gaining wider acceptance as an effective and
12 practical data collection method that avoids problems associated with manual
13 data collection. Due to the significant advances in the computer vision field, it is
14 possible to automatically and accurately detect and track road users. This
15 automatically collected and analyzed data is more accurate than manually
16 collected and analyzed data [4].

17 The main objectives of this study are:

- 18 1. Diagnose non-motorized road users’ safety issues during Summer Streets’
19 transition period and normal operation. Assess the road user safety issues
20 at the location and identify factors contributing to them. Traffic conflict
21 technique (TCT) [[5], [6]] was applied to assess vulnerable road user’s
22 safety during the three phases of operation at the study location. This
23 approach overcome many shortcomings associated with traditional road
24 safety analysis techniques that depend on collision data [5]. The analysis
25 was conducted using an automated road safety analysis system [[7], [8],
26 and [9]]. This system can detect, track and classify road users as well as
27 measure the severity of conflicts for complex interaction.
- 28 2. Extract data of non-motorized road users, including pedestrians, joggers,
29 cyclists and other users (such as rollerbladers). Traditionally, pedestrian
30 and cyclist walking speed measurement and volume counts have been
31 considered the most important data for behavior analysis. With the
32 advancement of computer vision technology, studying the behavior of
33 vulnerable road users becomes more efficient. The automated video-based
34 system was used successfully in previous studies to automatically collect
35 pedestrian crossing speeds and volume data [[10], [11]] and cyclist speeds
36 and volume data [[12], [13]]. This study is unique in its analysis as it
37 demonstrates the applicability of the system to collect pedestrian,
38 rollerblader, joggers and cyclist data during a complex shared space urban
39 environment. Furthermore, pedestrian gait parameters, such as step
40 frequency and step length, were automatically extracted, which can be
41 used to provide better understanding of pedestrian behavior.

1 Previous Work

2 Pedestrian and Cyclist Tracking

3 The detection and tracking of different road users is a significant application in
4 computer vision technology. Compared to tracking vehicular traffic, tracking
5 smaller road users (such as pedestrians) is more difficult because of various user
6 sizes and their complex movements. Pedestrians, for example, are difficult to track
7 due to their frequent change in direction and other road users can easily occlude
8 them. Common problems with tracking cyclists and pedestrians also include
9 global illumination variations, shadow handling, and multiple object tracking [14].
10 Different technologies are combined to increase the accuracy of pedestrian and
11 cyclist tracking [[15], [16]]. Enzweiler et al. [17] and Bertozzi et al. [18] provide
12 detailed reviews of different methods used to track and detect pedestrians. A
13 detailed methodology for detecting, tracking and classifying bicycles from video
14 scenes can also be found in [19] and [20]. A typical road user detection and
15 tracking sequence starts with object detection, hypothesis generation,
16 classification, and tracking.

17 Conflict Analysis

18 Recently, the TCT has been advocated as a promising approach of evaluating the
19 safety of road users from a broader perspective than relying on collision data,
20 which requires observing collisions for long periods of time. A traffic conflict is
21 defined as “an observable situation in which two or more road users approach
22 each other in space and time to such an extent that there is a risk of collision if
23 their movements remain unchanged” [21]. The objective of the TCT is to observe
24 and analyze the frequent traffic interactions and near misses between road users.
25 The validity of using traffic conflicts in safety studies has been established through
26 the investigating the relationship between conflicts and collisions in recent
27 studies [[22], [23], and [24]]. There are a number of conflict indicators proposed
28 to measure the severity of a traffic conflict. In this study, the Time-To-Collision
29 (TTC) indicator was used, defined as “the time that remains until a collision
30 between two road users would have occurred if the collision course and speed
31 difference are maintained” [25].

32 Speed and Gait Parameters

33 Pedestrian and cyclist speed is considered an important parameter in
34 understanding road user behavior. For pedestrians, previous studies focused on
35 estimating pedestrian speeds to investigate the behavior of pedestrians in
36 different walking environments [e.g., [26]]. In addition, the effect of different
37 pedestrian attributes on walking speed (such as age, gender, and group size,
38 among others) has been investigated in several studies. For example, it was found
39 that single pedestrians have faster crossing speeds compared to pedestrians
40 walking in groups [[26], [27]]. More recently, automated methods using computer
41 vision techniques calculate speed to study pedestrian behavior [4]. There are gaps
42 in the literature analyzing jogger and rollerblader speed, mainly due to the lack of
43 data of these road users. For cyclists, speed and behavior analysis has been an
44 active research topic recently, with cyclist speed being studied within different
45 route configurations [12]. To overcome shortcomings in manual methods,

1 automated video analysis has become increasingly common. One study using
2 computer vision to analyze cyclist speed behavior found that group size, travel
3 path, lane position and helmet usage affect the cycling mean speed [12].

4 Gait analysis has recently emerged as an important approach to understanding
5 pedestrian behavior, analyzing pedestrian step length and frequency. Hediye et
6 al. [4] conducted a study using computer vision techniques to collect pedestrian
7 gait parameters and speed data. Results of the study showed that the step
8 frequency and step length are influenced by factors such as crosswalk grade,
9 pedestrian gender, age, and group size. Hussein et al. [11] used computer vision
10 techniques to conduct a gait analysis showing that walking speed for a single
11 pedestrian is faster than those who walk in groups, males walk faster than females
12 with a longer step length and violators have higher walking speed, which was
13 dependent on step length, not step frequency.

14 Shared Space

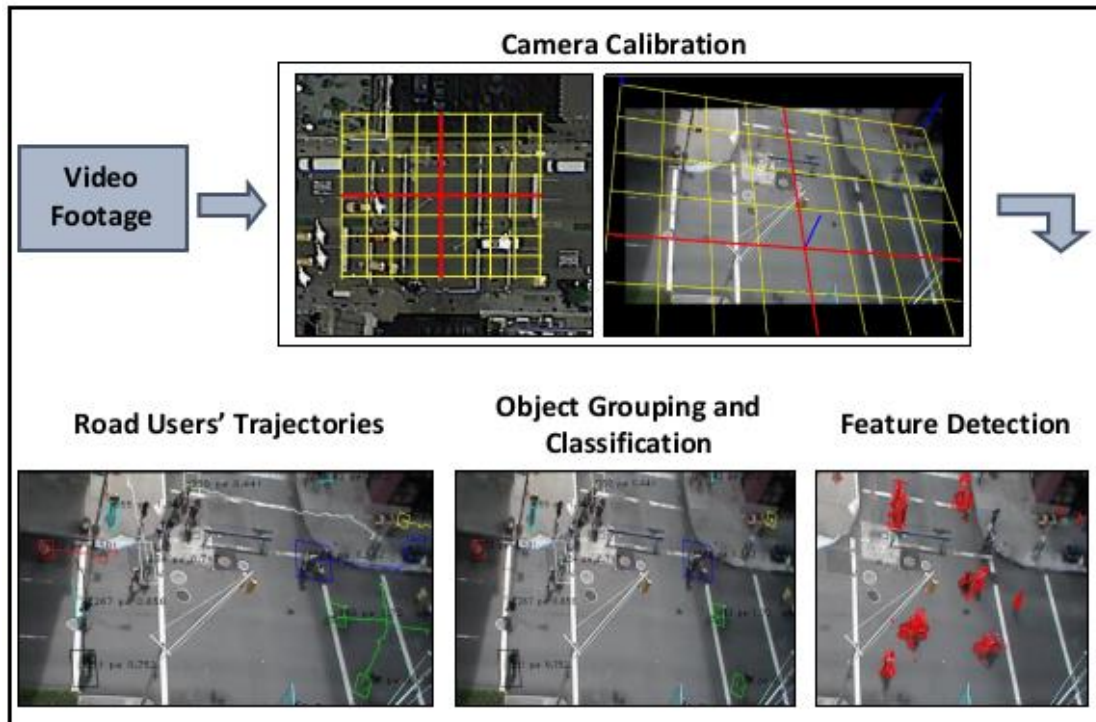
15 Encouraging the mixing of slower speed pedestrians and cyclists with higher
16 speed vehicles in a “shared space” is a novel concept following previous objectives
17 of separating vulnerable road users from vehicles. A “shared space” typically
18 achieves this safely and efficiently by designing the road to reduce the dominance
19 of vehicular traffic by promoting pedestrian and cycling activity while viewing the
20 road as a “place,” in addition to its transportation mobility and accessibility
21 purpose. The theory of “shared space” requires multi-disciplinary professions to
22 collaborate in the development of transportation corridors, which has been
23 argued by many authors [e.g. [28], [29]]. There are gaps in the literature with
24 regards to safety studies of “shared spaces.” Kaparias et al. [30] used a behavior
25 analysis technique instead of a conflict analysis technique to analyze the conduct
26 of pedestrians and vehicles in “shared space,” and found an increased confidence
27 of pedestrians but an unchanged behavior in drivers. Cyclovia or Open Streets
28 events have also become popular festivals, taking place in “shared space” streets
29 around the world. More than 80 North American cities share their streets by
30 offering regularly scheduled events drawing pedestrians and cyclists to enjoy car-
31 free streets [31]. Studies have documented the positive effects of these events on
32 physical activity, environment, social cohesion and businesses in the area [e.g.,
33 [32], [33]].

34 Methodology

35 Road Users’ Trajectory Extraction

36 The detection of road users from a video and the extraction of their trajectories is
37 achieved through an automated video processing system developed at the
38 University of British Columbia [[7], [34]]. As shown in Figure 1, the procedure
39 starts with the feature tracking step, in which discrete points (features) are
40 tracked on moving objects in the video scene. The system relies on the Kanade-
41 Lucas-Tomasi Feature Tracker algorithm to track moving features. The details of
42 the feature tracking process could be found in [34]. Next, features are grouped
43 based on their speeds, proximity and movement patterns to create one object. The
44 position of each object is tracked frame by frame to produce the road user

1 trajectory. These objects are classified into different road users, depending on the
2 classification methodology as described in [35]. However, a vital component of the
3 system is to relate the tracked trajectories in the video image coordinates to their
4 actual positions in the real world. This is done by creating a mapping between the
5 real world coordinated and the image space (camera calibration), which was
6 described in details in [36].



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Figure 1. Trajectory Extraction Process

9 **Conflict Analysis**

10 The conflicts between road users in this study were extracted by predicting the
11 future positions of each pair of road user trajectories and examining the
12 probability of these predicted future positions to collide, in space and time, into
13 one another [7]. Predicting the future positions of road users involve preparing a
14 set of prototypical trajectories from previously learnt motion patterns of road
15 users. Road user trajectories are then matched to these prototypes using the
16 longest common sub-sequence algorithm (LCSS) to provide a set of predicted
17 future positions with relative probabilities [8]. Typically, conflict indicators are
18 used to measure the severity of the detected conflicts. This study used the TTC
19 indicator as a measure of conflict severity. Only traffic conflicts with TTC less than
20 1.5 seconds was considered. TTC is calculated for each frame between conflicting
21 road users until they are no longer on a collision course, and each conflict is
22 associated with a set of TTC values. The minimum TTC is then used to represent
23 the overall severity of the conflict. TTC is used for in this study for demonstration
24 only. Other conflict indicators, such as Post Encroachment Time (PET), could also
25 be used. This procedure of conflict analysis has previously been validated in [37].

1 Road Users' Speed and Pedestrian Gait Parameters Analysis

2 After a road user trajectory is extracted from the video scene, the road user speed
3 profile can be easily produced. The average speed of any road user is estimated
4 directly from the speed profile as the average speed through the lifetime of the
5 trajectory. For the automatic extraction of pedestrian gait parameters, step
6 frequency can be computed automatically by analyzing the speed profile of each
7 pedestrian [38]. It was observed that a pedestrian speed profile typically shows
8 cyclic fluctuations repeated continuously over time. Hediye, et al. [38]
9 determined that each fluctuation in a given speed profile corresponds to a new
10 step taken by the pedestrian. Based on these observations, the pedestrian speed
11 profile can be seen as a time-series. Identifying the step frequency involves
12 evaluating the power spectral density (PSD) [[1], [38]] of the speed signal to
13 detect the dominant periodicity in the speed profile. Once the step frequency and
14 walking speed of a pedestrian is determined, the average step length can be
15 calculated from the fundamental linear relationship:

$$16 \text{ Walking Speed} = \text{Step Frequency} \times \text{Step Length} \quad (1)$$

17 Data Collection and Study Location

18 Video data was collected from the intersection of Park Avenue South and East 29th
19 Street in New York City. The analyzed data could be classified into three phases:
20 First, 30 minutes of data covering the Summer Streets operations where both
21 intersecting streets were closed for motorized vehicles. The second phase covers
22 the transition period starting from the moment the streets are reopen for traffic
23 until road users adapt to the regular operation. The transition period was
24 estimated to be 15 minutes based on the observed road users' behavior in the
25 video scene. Finally, the third phase covers 30 minutes of data of the regular
26 operations of the intersection. During regular operations, Park Avenue South is a
27 two-way north-south roadway with two moving lanes, a raised median, and a
28 parking lane in each direction (total roadway width = 69 feet or 21.0 meters). East
29 29th Street carries one-way westbound traffic (total roadway width approaching
30 Park Avenue South North Bound = 30 feet or 9.1 meters). The intersection is
31 controlled by a pre-timed signal (cycle length = 90 seconds). Data used in this
32 study was collected on August 2nd, 2014, from 11:45 AM to 2:30 PM.

33 Summary of Findings

34 Road Users Count

35 Table 1 summarizes the count of different road users during each of the three
36 phases considered for the analysis. Counts are reported per 15 minutes for
37 consistency as the transition period was only 15 minutes.

1 **Table 1: Road users count for each period**

Road user	Count (Per 15 minutes)		
	Summer Streets	Transition period	Normal operation
Pedestrians	560	398	384
Bikes	230	83	29
Vehicles	0	161	236
Child, bike	10	0	0
Baby carriage	6	0	0
Rollerblade	6	1	0
Jogger	49	4	2

2 **Road User Trajectories**

3 Road user trajectories were extracted according to the methodology described
 4 earlier. Figure 2 shows the spatial distribution of the pedestrian and bicycle
 5 trajectories for each of the three time periods considered. As shown in the figure,
 6 pedestrians use the full width of Park Avenue during the summer street operation
 7 (Figure 2-a). After the summer street operation was terminated, pedestrians start
 8 to shift towards the crosswalks during the transition period until the spatial
 9 distribution of their trajectories get back to normal operation configuration
 10 (Figure 2-e). However, during the transition phase (Figure 2-c), many trajectories
 11 were observed in the middle of Park Avenue, which can be considered as a
 12 violation event. This indicates that the transition period represents a hazard to
 13 pedestrians as they are not fully aware that the summer street operation has
 14 ended, which may lead to serious conflicts with other motorized road users using
 15 the roadway. Similar situations were noticed for bikes despite the absence of bike
 16 infrastructure at the intersection. Cyclists ride their bikes along Park Avenue
 17 during Summer Street operation (Figure 2-b), then move to the right side of the
 18 road when the event is over (Figure 2-f). The problem again appears during the
 19 transition period, when many bikes are still using the full width of Park Avenue
 20 despite the presence of vehicles in the street. These observations show that the
 21 transition period is a time where vulnerable road users have a higher risk of being
 22 involved in serious conflicts with motorized road users due to their non-
 23 compliance to traffic regulations.

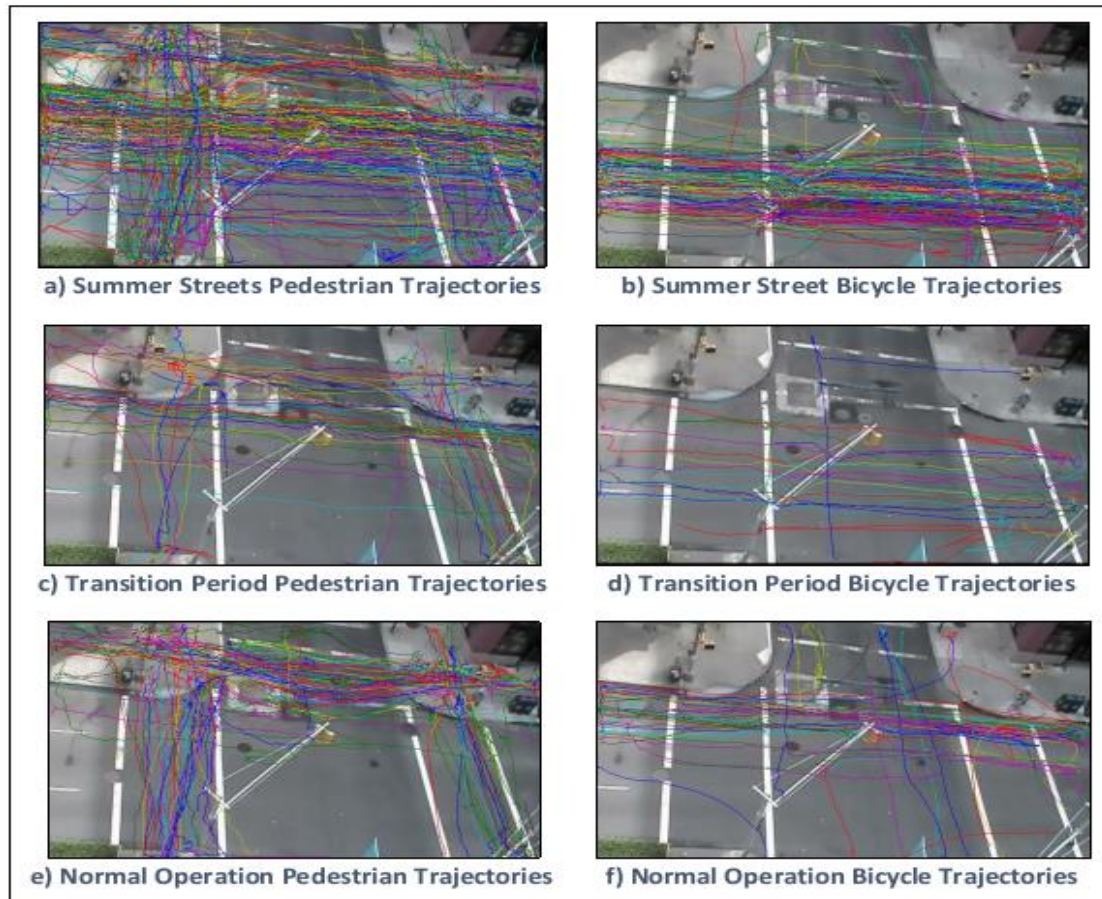
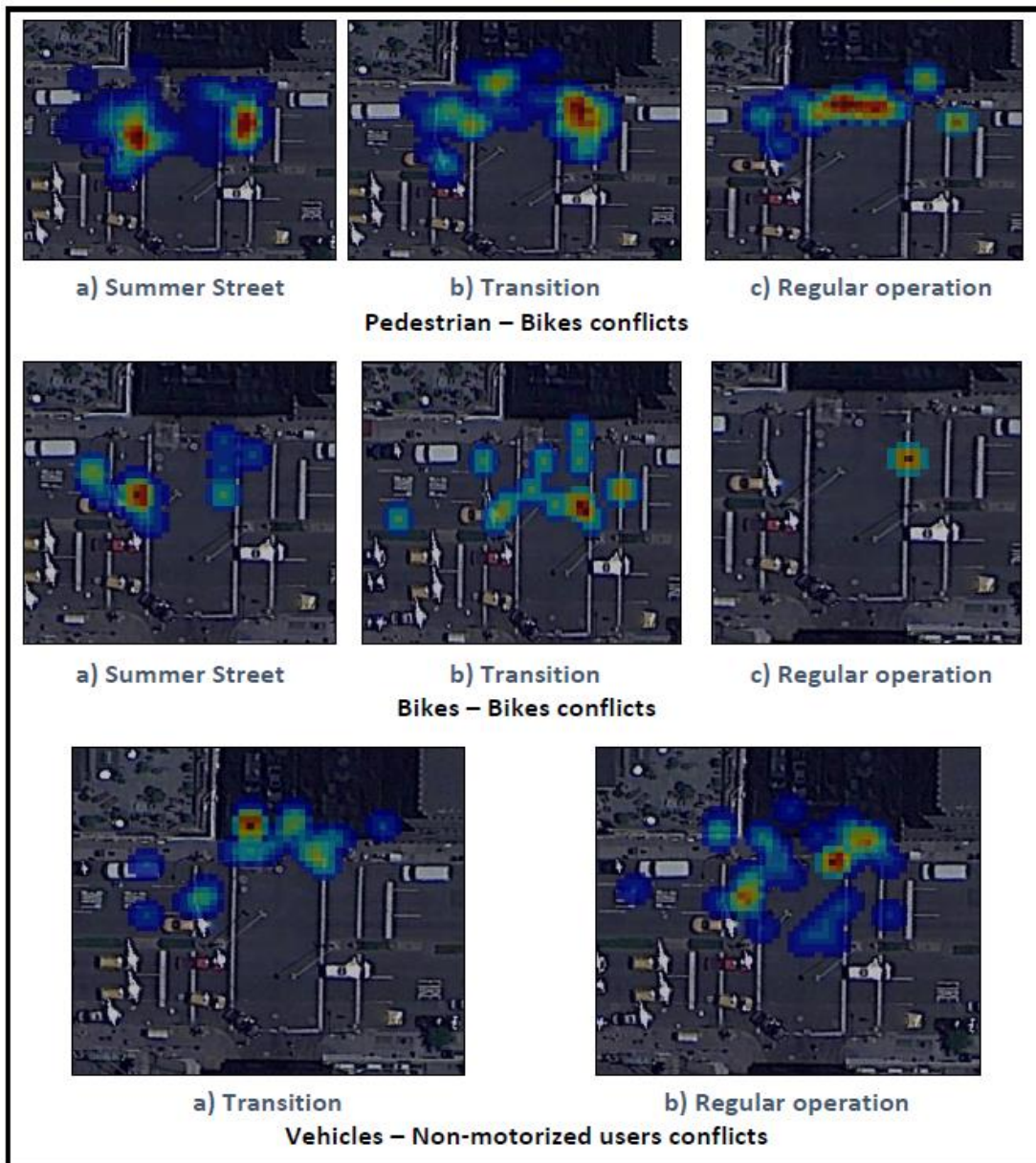


Figure 2. Spatial Distribution of Road Users' Trajectories

Traffic Conflicts

TCT was applied to investigate the frequency of serious interactions between different road users. Three conflict classes were extracted: conflict between bikes and pedestrians, conflicts between bikes and other bikes, and conflicts between motorized vehicles and all non-motorized road users. Extracted conflicts were classified among the three operational periods: Summer Street, transition period, and normal operation. Figure 3 shows the heat maps of automatically identified conflicts of the three classes considered while Table 2 summarizes the number of conflicts for each period and collision type.



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Figure 3. Heat Map of Road Users' Conflicts

3 **Table 2: Number of TTC events for each period and collision type**

Class	TTC events (Per 15 minutes)		
	Summer Street	Transition period	Normal operation
Pedestrian - Bike	127	59	26
Bike - Bike	16	14	1
Motorized Vehicle - Non motorized	4	24	43

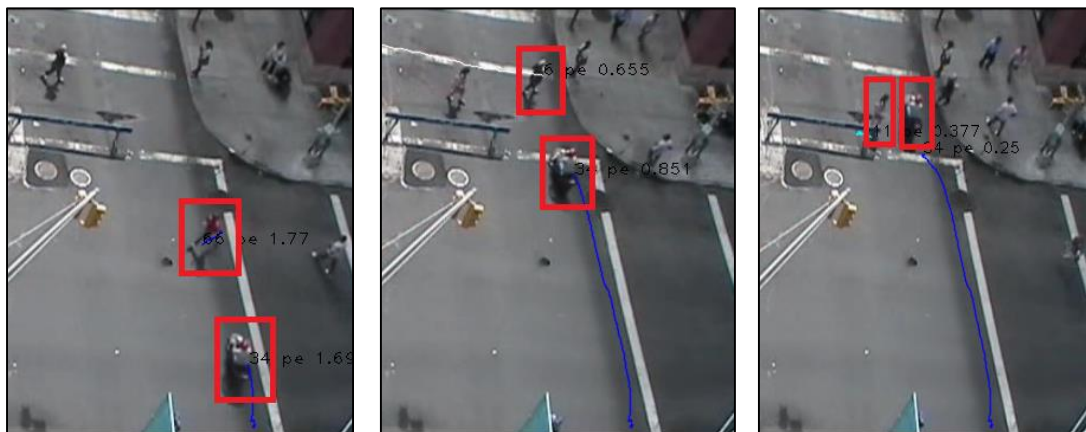
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The results show a large number of conflicts between pedestrians and bikes during the Summer Street period (127 conflicts/15 minutes) and these conflicts are spread almost over the whole study area as shown in Figure 3 (top left image) a. This large number of conflicts is difficult to compare since there are no published studies examining pedestrian-bicycle conflicts in a shared space environment. However, this large number of conflicts was expected because the

1 Summer Streets event drew a large number of pedestrians that used the full width
2 of Park Avenue to walk. In addition, bikes have to navigate through the crowds
3 creating many conflicts with pedestrians. The number of pedestrian–bicycle
4 conflicts was reduced significantly in the transition period and the normal
5 operation period afterwards. As well, conflicts began to concentrate around the
6 crosswalks instead of the whole space, as it appears in Figure 3 (top right image).
7 However, the relatively large number of conflicts in the transition period (59
8 conflicts in 15 minutes) and their spatial distribution (conflicts spread throughout
9 the study space) suggests that the transition between Summer Streets and normal
10 operation represent a hazardous period to road users as they are not yet
11 complying to the traffic rules. This was the reason behind the presence of traffic
12 police officers and traffic control agents to help direct traffic during the transition.

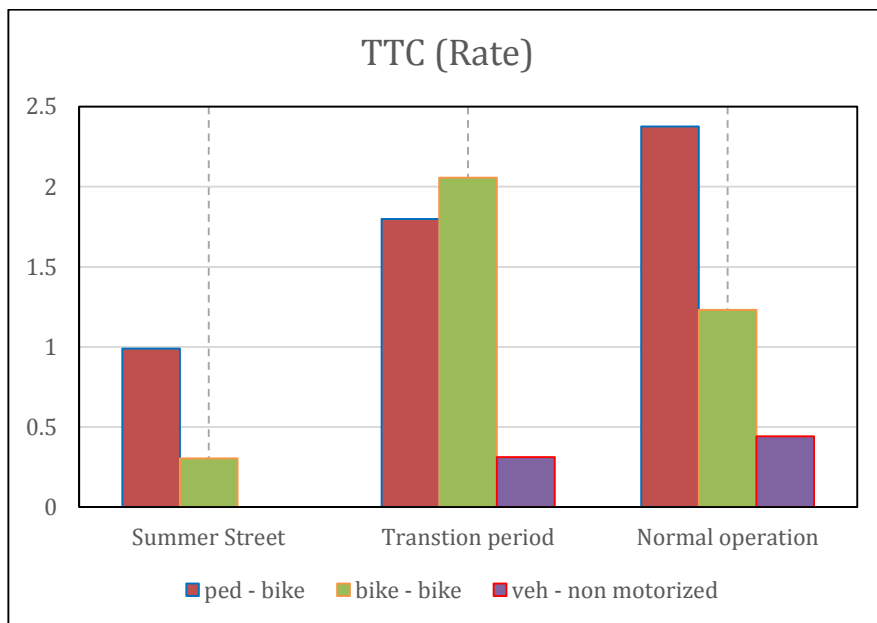
13 Conflicts between bikes were almost constant during the Summer Streets and
14 transition periods despite the fact that the number of bikes during the transition
15 period was significantly lower compared to during Summer Streets, as shown in
16 Table 1. This suggests that the transition period reveals many road users’ non-
17 compliance to traffic rules, which lead to higher rates of conflict. It also shows that
18 when organizing “shared space” events, the transition period should be carefully
19 planned. The conflicts between bikes were not notable during normal operation
20 due to the smaller number of bikes using the intersection during that period.

21 Conflicts between vehicles and non-motorized users were not expected during the
22 Summer Streets period. However, a violation was observed during this period
23 where a motorcycle passed through East 29th Street causing four conflicts with
24 non-motorized road users (Figure 4). During the transition period, conflicts
25 between motorized and non-motorized users mainly occurred at the east
26 crosswalk (crossing East 29th Street), as shown in Figure 3 (the bottom left image).
27 This is mainly because vehicles were allowed to move along East 29th Street for
28 some time before Park Avenue was open for traffic. As such, many transitional
29 period conflicts occur between the westbound traffic on East 29th Street and
30 pedestrians and bikes on Park Avenue. The number of conflicts rises to 43 conflicts
31 per 15 minutes during the normal operation phase, which is high and agrees with
32 reported results [11] that intersections in New York City have high conflict rates
33 between vehicles and non-motorized users.



34 **Figure 4. Motorbike Violation Event during Summer Streets and Associated**
35 **Conflicts**

1 Furthermore, as the number of road users varied significantly among the three
 2 periods analyzed, it is important to investigate the conflict rates in addition to the
 3 conflict frequencies discussed above. Figure 5 presents the different observed
 4 conflicts normalized by the number of conflicting road users of each category.
 5 Although the number of “pedestrian–bicycle” and “bicycle–bicycle” conflicts
 6 during the Summer Street period were significantly higher than the other two
 7 periods, the conflict rates were significantly lower. This indicates that the Summer
 8 Street period is the safest period for road users in terms of serious conflicts with
 9 other road users. Despite many interactions observed during Summer Streets,
 10 road users make use of the larger space available during Summer Streets where
 11 they can navigate freely and avoid serious conflicts, especially as their speeds are
 12 slower during this period as will be discussed later. The “bicycle–bicycle” conflict rate
 13 recorded its peak during the transition period, which serves as confirmation of
 14 safety issues during the transition period. As well, although the conflict rates for
 15 “vehicle–non-motorized users” and “pedestrian–bicycle” conflicts were higher
 16 during the normal operation phase, the rates observed during the transition phase
 17 were still high and very close to those observed during the normal operation. The
 18 difference in observed rates between the two periods was about 24% and 28% for
 19 “pedestrian–bicycle” and “vehicle–non-motorized” conflicts, respectively, despite
 20 the presence of a police officer to regulate traffic during the transition period.



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 22 **Figure 5. Conflict rates**

23 **Road Users’ Speed and Pedestrian Gait Analysis**

24 **Pedestrians and Joggers’ Speed**

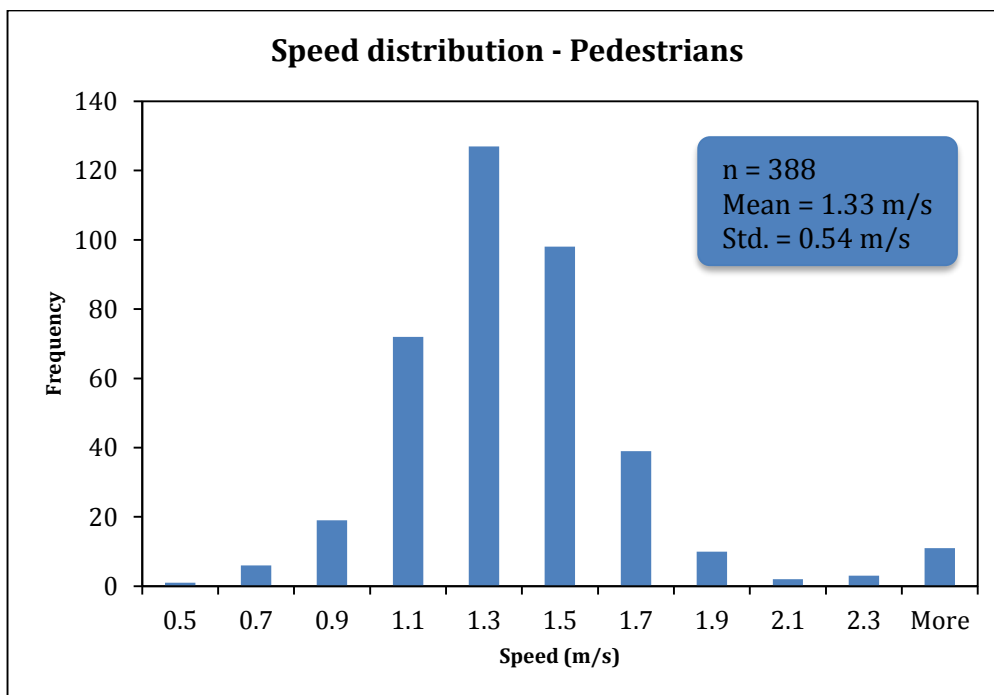
25 Speeds of non-motorized road users (pedestrians, cyclists, joggers, and
 26 rollerbladers) were automatically extracted following the methodology described
 27 earlier. The distribution of both pedestrians’ and joggers’ speed is presented in
 28 figure 6. The mean and standard deviation for pedestrian and jogger speed was
 29 found to be $(1.33 \pm 0.54 \text{ m/s})$ and $(2.81 \pm 0.54 \text{ m/s})$, respectively. The variation of
 30 pedestrian speed during the three time periods of the festival was investigated. As

1 shown in Table 3, pedestrians walked faster during the Summer Streets event and
 2 their speed reduced during the following periods. One hypothesis that can explain
 3 this is that they may be taking advantage of the “shared space” to move freely all
 4 over the width of the intersecting streets, while during the normal operation
 5 phase pedestrians have limited space (the sidewalks and crosswalks), which limit
 6 their speed, especially with the high density of pedestrians in the study location.
 7 Table 3 also shows the variation of pedestrian speed with group size. As shown in
 8 the table, the walking speed decreases as the group size increases which agrees
 9 with results reported in several previous studies [e.g., [11], [38], and [39]].
 10 Unfortunately, the investigation of jogger’s speed variation during different
 11 phases and with group size was not possible, as joggers were only present during
 12 the Summer Street event and only ran individually.

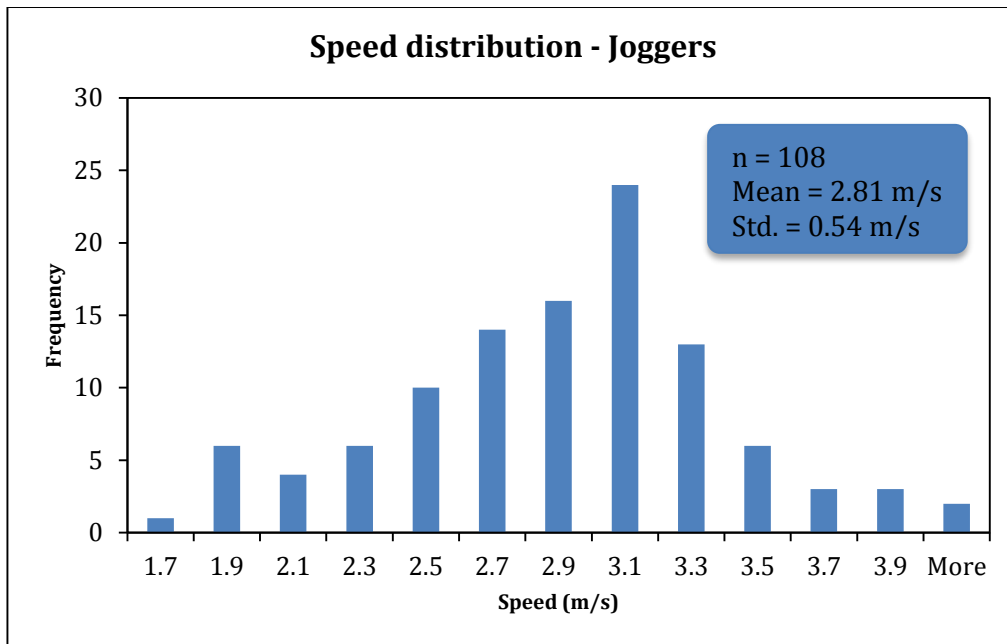
13 **Table 3: Summary of Pedestrian Speed**

	Group Size			Analysis time			Total
	Single	Size = 2	Size = 3 +	Summer Streets	transition period	normal operations	
Count	148	174	66	220	58	92	388
Average speed (m/s)	1.38	1.33 (0.21)	1.24 (0.10)*	1.49	1.39 (0.17)	1.30 (0.21)	1.33
Standard deviation (m/s)	0.58	0.54	0.42	0.74	0.73	0.47	0.54

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2 **Figure 6. Speed distribution of pedestrian and joggers**

3 **Cyclists' Speed**

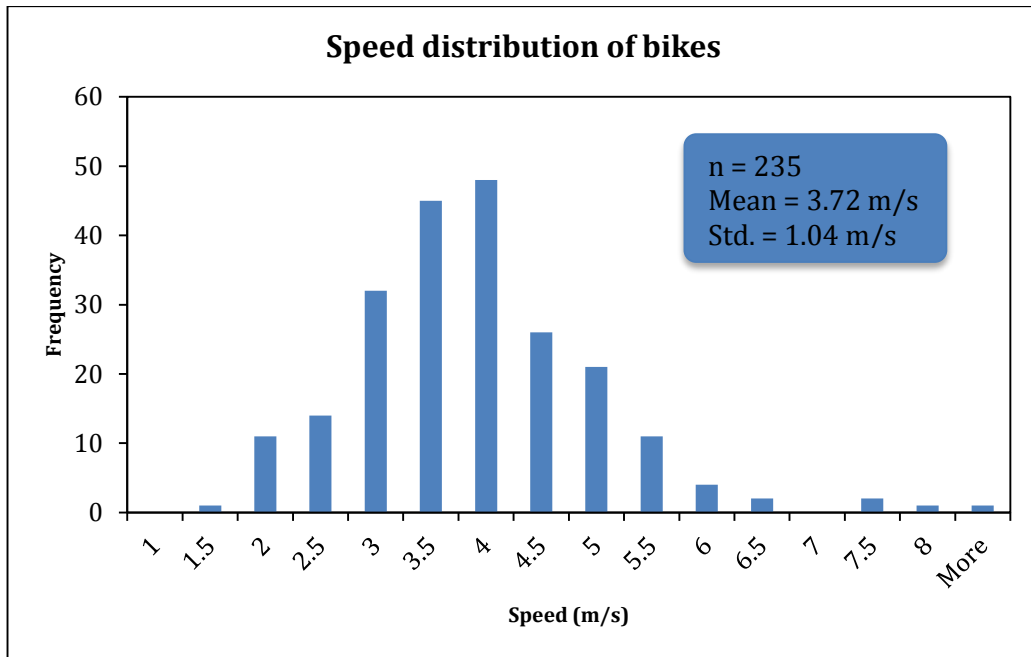
4 The distribution of cyclists' speed is presented in figure 7. The mean and standard
 5 deviation for cyclists' speed was found to be $(3.72 \pm 1.04 \text{ m/s})$ with a minimum
 6 speed of 1.09 m/s and maximum speed of 8.30 m/s. The variation of cyclists' speed
 7 during the three time periods of the festival were investigated. As shown in Table
 8 4, bike speeds were very low during the Summer Streets event and increased
 9 during normal traffic operations. These findings may be attributed to the "shared
 10 space" of Summer Streets attracting bikers of different ages to ride their bikes in
 11 a relaxed manner for leisure and consequently, the average biking speed is
 12 expected to be low. On the other hand, it is expected that during the normal
 13 operations, more cyclists use their bike to commute with higher speed. In addition,
 14 the presence of pedestrians in the street during the summer street event forces
 15 bikes to slow down to avoid collision with pedestrians and to navigate through the
 16 crowds. The effect of group size on biking speeds was also investigated. As was
 17 expected, when the cyclists' bike in groups, their speed on average becomes
 18 slower. Groups of cyclists were observed only during the Summer Streets
 19 operations, when the "shared space" allowed them to take the road and cycle at
 20 slower speeds in larger groups. Group cycling is common in bicycle-friendly cities
 21 where larger proportions of cyclists and separated bicycle infrastructure allow
 22 cyclists to feel safe and relaxed cycling in groups [40].

23 **Table 4: Summary of Cyclist Speeds**

	Group Size		Analysis time			Total
	Single Bike	Size = 2+	Summer Streets	Transition Period	Normal Operations	
Count	101	134	150	34	28	235
Average speed (m/s)	3.93	3.56 ** (.01)	3.62	3.86 (0.13)	4.17 (0.2)	3.72

Standard deviation (m/s)	1.43	0.79	0.97	1.17	1.57	1.04
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Figure 7. Speed distribution of bikes

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Rollerbladers' Speed

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The concept of Summer Streets encourages alternative road users, such as rollerbladers, to utilize and share the road. Although only 13 rollerbladers were observed during the analyzed data, we think it is worth reporting the values obtained as a reference. Further analysis on a larger data set is highly recommended. The average speed of rollerbladers was 3.13 m/s with a standard deviation of 0.69 m/s. The maximum rollerblader speed was 4.39 m/s and the minimum was 2.48 m/s.

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Road users' speed compared to FHWA study

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The speed of different road users during summer street operation were compared to the Federal Highway Administration's *Shared Use Path Level of Service Calculator* [41], presented in Table 5. It was observed that the average speeds for road users during Summer Streets operations are slower than the typical speeds for road users on shared paths. This is most likely because the festival encourages a larger group of people to enjoy "shared spaces" and road users may be slowing down to enjoy the Summer Streets.

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Table 5: Comparison of road users' speed during Summer Street and values extracted from shared use path LOS calculator

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Road User	Summer Streets average Speed (m/s)	Shared Use Path LOS Calculator average Speed (m/s)
Pedestrians	1.38	2.91
Bikes	3.93	5.72

Rollerbladers	3.13	4.52
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2 **Pedestrian Gait parameters**

3 Pedestrian gait parameters (mainly, step frequency and step length) were
4 automatically extracted and compared to the findings reported from the
5 pedestrian gait analysis studied at Park Avenue South and East 28th Street during
6 typical New York City street operations in [11]. The frequency and length
7 parameters were found to follow the normal distribution with 95% confidence, as
8 confirmed by the χ^2 test. Step frequency and step length were found to be $(1.8 \pm$
9 $0.17 \text{ Hz})$ and $(0.7 \pm 0.116 \text{ m})$, respectively. Both the step frequency and lengths
10 were smaller compared to values extracted from typical New York City street
11 operations reported in [11] (reported step frequency in this study was 1.96 ± 0.17
12 Hz and step length was $0.75 \pm 0.14 \text{ m}$). These differences are due to the fact that
13 Summer Street data is collected on pedestrians in a “shared space” environment,
14 while the New York City data is collected during peak hour normal operations. The
15 “shared space” creating a sense of place on the street and the recreational nature
16 of the festival is hypothesized as the reason for smaller pedestrian gait parameters
17 during Summer Streets.

18 **Conclusions and Future Research**

19 Seventy-five minutes of video data were collected at a signalized intersection of
20 Park Avenue South and East 29th Street in New York City during three different
21 periods of operations: Summer Streets “shared space” operations, the transition
22 period, and normal operations. Analysis was conducted on video data by means of
23 computer vision techniques using an automated system developed at the
24 University of British Columbia. The main purposes of the analysis are: 1) to assess
25 pedestrian and cyclist safety issues at the intersection during the “shared space”,
26 transition, and normal operations and to identify factors that contribute to the
27 safety issues; 2) to demonstrate the utility of the automatic extraction of
28 pedestrian, cyclist, rollerblader, and runner data to better understand road user
29 behavior, speed, and pedestrian gait parameters (step length and step frequency)
30 during the “shared space,” transition, and normal operations periods.

31 TCT were adopted to diagnose road user safety at the intersection during the three
32 periods. The study identified pedestrian–bicycle, bicycle–bicycle, and vehicle–
33 non-motorized conflicts during the Summer Streets operation, transition period,
34 and normal operation time periods. Trajectories of pedestrians and cyclists show
35 that the road users were not observing traffic rules in the transition period and
36 normal operations, as they were spatially distributed through the street. The
37 lowest rate of conflicts between pedestrians and cyclists and between cyclists was
38 found to be during Summer Streets “share space” operations. It is hypothesized
39 that this is due to cyclists slowing down because they do not need to compete with
40 vehicle traffic, as well as the street being used recreationally during the festival. It
41 is recommended that North American cities take the opportunity to implement
42 “shared spaces” to create safer streets for vulnerable road users. Many European
43 cities have examples of place-making and improving safety through implementing

1 “shared spaces.” In the meantime, open street festivals that close roads to
2 vehicular traffic encourage sharing space, active lifestyles, and sustainable living.

3 Relatively large number of pedestrian–bicycle conflicts were observed in the
4 transition period (59 conflicts in 15 minutes) and bicycle–bicycle conflicts (14
5 conflicts in 15 minutes) and their spatial distribution (conflicts spread throughout
6 the study space) suggests that the transition between Summer Streets and normal
7 operation represent a hazardous period to road users as they are not yet
8 complying to the traffic rules. This shows that when organizing “shared space”
9 events, the transition period should be given more thought.

10 In addition to conflict analysis, road user (pedestrian, runner, cyclist, and
11 rollerblader) speeds and pedestrian step frequency and step length were
12 automatically extracted. The parameters were found to follow the normal
13 distribution with 95% confidence, as confirmed by the χ^2 test. Comparing Summer
14 Streets operations to typical New York City street operations, both the pedestrian
15 step frequency and lengths were smaller. The “shared space” creates a sense of
16 place on the street and recreational nature of the festival is hypothesized as the
17 reason for smaller pedestrian gait parameters during Summer Streets. Results
18 show that pedestrians were fastest during Summer Street operations (1.49 ± 0.54
19 m/s) and slowest during normal operations (1.30 ± 0.22 m/s). These results show
20 that pedestrians may be taking advantage of the “shared space” festival to get
21 exercise. The results for different pedestrian group sizes of singles, two
22 pedestrians and three pedestrians were found to be (1.38 ± 0.34 m/s), (1.32 ± 0.30
23 m/s) and (1.24 ± 0.18 m/s), respectively, show that as group sizes become larger,
24 pedestrians walk more slowly. This study was the first to automatically extract
25 rollerblader speed and found it to be (3.12 ± 0.69 m/s), however due to small
26 sample size ($n=13$), further analysis is recommended.

27 For cyclists, the results show that the “shared space” allowed them to cycle slower
28 at speeds of (3.62 ± 0.95 m/s) than normal operations where cyclists traveled
29 adjacent to vehicle traffic at speeds of (4.17 ± 2.46 m/s). The results also showed
30 that as cyclist group sizes became larger, the cyclists on average were slower.
31 Group cycling is common in bicycle-friendly cities with separate bicycle
32 infrastructure, and it is a good indicator of a positive safety environment. This was
33 the first study to automatically extract groups of cyclists, and further study on this
34 topic is recommended.

35 Future work includes analyzing additional video data of “shared space” road user
36 behavior and conflict analysis, calculating other conflict indicators (Post
37 Encroachment Time, etc.), recommending specific safety countermeasures to
38 design the optimal “shared space” and potentially conducting a before and after
39 safety evaluation of the implemented “shared space.” More data is required to
40 come to any conclusions about the speed and behavior of rollerbladers, groups of
41 pedestrians, and groups of cyclists. To better understand the safety and road user
42 behavior of “shared spaces” it is important to analyze other similar
43 environments—such as pedestrian-only streets, multi-use pathways, and Open
44 Street events.

1 References

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- [1] Saunier, N., El Husseini, A., Ismail, K., Morency, C., Auberlet, J. M., & Sayed, T., "Estimation of frequency and length of pedestrian stride in urban environments with video sensors," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2264, no. 1, pp. 138-147, 2011.
- [2] Victoria Transportation Institute, "Evaluating Active Transportation Benefits and Costs," 2014.
- [3] New York City Department of Transportation, "Summer Streets," 2015. [Online]. Available: <http://www.nyc.gov/html/dot/summerstreets/html/home/home.shtml>.
- [4] H. Hediye, T. Sayed and M. H. & I. K. Zaki, "Automated analysis of pedestrian crossing speed behavior at scramble-phase signalized intersections using computer vision techniques," *International journal of sustainable transportation*, vol. 8, no. 5, pp. 382-397, 2014.
- [5] T. Sayed and S. Zein, "Traffic Conflict Standards for Intersections," *Transportation Planning and Technology*, vol. 22, pp. 309-323, 1999.
- [6] Chin, H. C., & Quek, S. T., "Measurement of traffic conflicts," *Safety Science*, vol. 26, no. 3, pp. 169-185, 1997.
- [7] Saunier, N.; Sayed, T.; Ismail, K., "Large Scale Automated Analysis of Vehicle Interactions and Collisions," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2147, pp. 42-50, 2010.
- [8] Saunier, Nicolas; Sayed, Tarek, "Automated Analysis of Road Safety with Video Data," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2019, pp. 57-64, 2007.
- [9] J. Autey, T. Sayed and M. Zaki, "Safety Evaluation Of Right-turn Smart Channels Using Automated Traffic Conflict Analysis," *Accident Analysis and Prevention*, vol. 45, pp. 120-130, 2012.
- [10] Zaki, M. H., Sayed, T., Tageldin, A., & Hussein, M., "Application of Computer Vision to Diagnosis of Pedestrian Safety Issues," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2393, no. 1, pp. 75-84, 2013.
- [11] Hussein, M., Sayed, T., Reyad, P., & Kim L., "Automated pedestrian safety diagnosis and behavioral study at signalized intersections in New York City," *Transportation Research Record: Journal of the Transportation Research Board*, vol. In print, 2015.

- [12] Zaki, M., Sayed, T., & Cheung, A., "Computer vision techniques for the automated collection of cyclist data," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2387, pp. 10-19, 2013.
- [13] Sayed, T., Zaki, M. H., & Autey, J., "Automated safety diagnosis of vehicle-bicycle interactions using computer vision analysis," *Safety science*, vol. 59, pp. 163-172, 2013.
- [14] B. Morris and M. Trivedi, "A Survey of Vision-Based Trajectory Learning and Analysis for Surveillance," *IEEE Transactions on Circuits and Systems for Video Technology*, vol. 18, no. 8, pp. 1114-1127, 2008.
- [15] Curio, C.; Edelbrunner, J.; Kalinke, T.; Tzomakas, C.; Seelen, W. von, "Walking pedestrian recognition," in *IEEE/IEEJ/JSAI International Conference on Intelligent Transportation Systems*, 1999.
- [16] Khanloo, B. Y. S., Stefanus, F., Ranjbar, M., Li, Z. N., Saunier, N., Sayed, T., & Mori, G., "A large margin framework for single camera offline tracking with hybrid cues," *Computer Vision and Image Understanding*, vol. 116, no. 6, pp. 676-689, 2012.
- [17] M. Enzweiler and D. Gavrila, "Monocular Pedestrian Detection: Survey and Experiments," *IEEE Transactions on Pattern Analysis and Machine Intelligence*, vol. 31(12), pp. 2179-2195, 2009.
- [18] Bertozzi, M.; Broggi, A.; Cellario, M.; Fascioli, A.; Lombardi, P.; Porta, M., "Artificial vision in road vehicles," *Proceedings of the IEEE*, vol. 90, no. 7, pp. 1258-1271, 2002.
- [19] Buch, N., Orwell, J., & Velastin, S. A., "Urban road user detection and classification using 3D wire frame models," *IET Computer Vision*, vol. 4, no. 2, pp. 105-116, 2010.
- [20] Cho, H., Rybski, P. E., & Zhang, W., "Vision-based bicyclist detection and tracking for intelligent vehicles," in *Intelligent Vehicles Symposium (IV), 2010 IEEE*, 2010.
- [21] F. Amundsen and C. Hydén, "Proceedings of First Workshop on traffic Conflicts, Institute of Economics," in *Lund Institute of Technology*, Oslo, 1977.
- [22] K. El-Basyouny and T. Sayed, "Safety performance functions using traffic conflicts," *Safety Science*, vol. 51, no. 1, pp. 160-164, 2013.
- [23] Sacchi, E., Sayed, T., & deLeur, P., "A comparison of collision-based and conflict-based safety evaluations: The case of right-turn smart channels," *Accident Analysis & Prevention*, vol. 59, pp. 260-266, 2013.

- [24] Zheng, L., Ismail, K., & Meng, X., "Shifted Gamma-Generalized Pareto Distribution model to map the safety continuum and estimate crashes," *Safety Science*, vol. 64, pp. 155-162, 2014.
- [25] J. Hayward, "Near-miss determination through use of a scale of danger," *Highway Research Record*, vol. 384, pp. 24-34, 1968.
- [26] Knoblauch, R. L., Pietrucha, M. T., & Nitzburg, M., "Field studies of pedestrian walking speed and start-up time," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 1538, no. 1, pp. 27-38, 1996.
- [27] Gates, T. J., Noyce, D. A., Bill, A. R., & Van Ee, N., "Recommended walking speeds for timing of pedestrian clearance intervals based on characteristics of the pedestrian population," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 1982, no. 1, pp. 38-47, 2006.
- [28] J. Jacobs, *The death and life of great American cities*, Vintage, 1961.
- [29] Appleyard, D., Gerson, M. S., & Lintell, M., "Livable streets, protected neighborhoods," University of California Press, 1981.
- [30] Kaparias, I., Bell, M. G. H., Biagioli, T., Bellezza, L., & Mount, B., "Behavioural analysis of interactions between pedestrians and vehicles in street designs with elements of shared space," *Transportation Research Part F: Traffic Psychology and Behaviour*, vol. 30, pp. 115-127, 2015.
- [31] E. Beard, "Care-Free, Carefree Streets," Parks & Recreation, 2013.
- [32] Engelberg, J. K., Carlson, J. A., Black, M. L., Ryan, S., & Sallis, J. F., "Ciclovía participation and impacts in San Diego, CA: The first CicloSDias," *Preventive medicine*, vol. 69, pp. S66-S73, 2014.
- [33] Hipp, J. A., Eyler, A. A., Zieff, S. G., & Samuelson, M. A., "Taking physical activity to the streets: the popularity of Ciclovía and Open Streets initiatives in the United States," *American Journal of Health Promotion*, vol. 28, no. sp3, pp. S114-S115, 2014.
- [34] N. Saunier and T. Sayed, "A Feature-Based Tracking Algorithm for Vehicles in Intersections," in *Third Canadian Conference on Computer and Robot Vision, IEEE*, Québec, 2006.
- [35] Zaki, M. H. & Sayed, T., "A framework for automated road-users classification using movement trajectories," *Transportation Research Part C: Emerging Technologies*, vol. 33, pp. 50-73, 2013.

- [36] Ismail, K., Sayed, T., & Saunier, N., "A methodology for precise camera calibration for data collection applications in urban traffic scenes," *Canadian Journal of Civil Engineering*, vol. 40, no. 1, pp. 57-67, 2013.
- [37] Ismail, K.; Sayed, S. & Saunier, N., "Automated Analysis Of Pedestrian-vehicle Conflicts: A Context For Before-and-after Studies.," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2198, pp. 52-64, 2010.
- [38] Hediye, H., Sayed, T., Zaki, M. H., & Mori, G., "Pedestrian gait analysis using automated computer vision techniques," *Transportmetrica A: Transport Science*, vol. 10, no. 3, pp. 214-232, 2014.
- [39] Li, S., Sayed, T., Zaki, M. H., Mori, G., Stefanus, F., Khanloo, B., & Saunier, N., "Automated Collection of Pedestrian Data Through Computer Vision Techniques," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 2299, no. 1, pp. 121-127, 2012.
- [40] Aldred, R., and Jungnickel, K., "Constructing mobile places between 'leisure' and 'transport': a case study of two group cycle rides," *Sociology*, vol. 46, no. 3, pp. 523-539, 2012.
- [41] Patten, R. S., Schneider, R. J., Toole, J. L., Hummer, J. E., & Roupail, N. M. , "Shared-Use Path Level of Service Calculator—A User's Guide," (No. FHWA-HRT-05-138), 2006.
- [42] Bain, L., Gray, B., & Rodgers, D., *Living streets: Strategies for crafting public space*, John Wiley & Sons, 2012.

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