

1 **MODELING DOUBLE PARKING IMPACTS ON URBAN STREET**

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Abstract

1 Double parking (DP) is one of the key contributors to traffic congestion on urban streets.
2 Double parking violations of commercial vehicles while they load and unload at delivery locations
3 with insufficient curbside space can have significant negative impact on traffic. Motivated by the
4 need to study such impact in urban cities, this paper utilizes parking violation records for New
5 York City along with field data collected using video recording, and adopts a comprehensive
6 modeling approach that combines available data with two types of models. The first is an
7 M/M/∞ queueing model used to estimate double parking effect on the average travel time. The
8 second is a micro-simulation model developed and calibrated to study individual and combined
9 effects of various explanatory variables. Both models account for different effects of general
10 vehicles and commercial trucks. Via case studies in Midtown Manhattan and Downtown Brooklyn
11 (New York, US), double parking activities and driver behaviors are investigated and used for
12 comparative analysis. The M/M/∞ queueing model has been empirically validated using field data
13 collected as part of this study. Comparison results show a good fit for uncongested traffic
14 conditions. Micro-simulation results indicate different impact levels for 21 scenarios in four
15 categories namely, travel demand, double parking locations, frequency, and durations. This study
16 can provide traffic agencies a potential approach to quantify the impact of double parking in a
17 large-scale network and insights into the management and alleviation of on-street parking
18 problems including incentives for encouraging off-hour deliveries and more effective enforcement
19 during peak hours.
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1 INTRODUCTION

2 In urban areas, obstructions of traffic such as double parking, commercial vehicle deliveries,
3 pedestrian jaywalking, taxi pick-ups and drop-offs, are potential impediments to road capacity and
4 vehicular speed, and causes traffic delay and safety risks. New York City Department of Finance
5 (NYCDOF) defines “double parking” as “standing or parking a vehicle on the roadway side of a
6 vehicle already stopped, standing or parked at the curb” (1). It is mainly due to the lack of available
7 on-street parking spaces and sometimes makes the street impassable, especially in one-way single-
8 lane situations. According to New York City Department of Transportation (NYCDOT) parking
9 regulations, double parking of passenger vehicles is illegal at all times in New York City (NYC),
10 regardless of location, purpose or duration (2). In most cases, a double parked vehicle will block
11 part of the street and a bus or a bicycle lane, if one is present. Double parking can become a serious
12 problem especially in congested urban areas since it is both obstructive and irritating.

13 Most cities rely on fines when dealing with double parking (3). In 2014, double parking
14 violation had 502,082 ticketed cases and was ranked seventh among all the parking violations in
15 New York City according to NYCDOF records. In reality, the total number of double parking
16 violations is highly underestimated as most of such activities may not even be recorded or ticketed.
17 Notably, it becomes more complex in mixed-used or commercial zones while double-parked
18 commercial vehicle can lead to further reduction in capacity and increase in travel time (TT) and
19 delay.

20 There are several other strategies to reduce double parking of commercial vehicles,
21 especially during the peak hours. A recent study that focused on Off-Hour Delivery (OHD)
22 strategies emphasizes the importance of promoting these strategies to reduce congestion caused by
23 double parking delivery trucks. On the one hand, driving on an OHD route saves about
24 \$9,000/year/OHD-tour in parking fines since it is easier for truck drivers to find legal parking
25 spaces near their delivery destination during the off-hours (4). On the other hand, increased
26 enforcement of parking fines for double parking during the regular hours could encourage more
27 carriers to participate in OHD (5).

28 One of the challenges in quantifying the effects of double parking is the lack of reliable
29 analytical approaches dealing with this important problem. There’s a need to identify a robust
30 analytical methodology so that a large number of sites can be studied efficiently to quantify the
31 impact of double parking as well as various long and short term parking enforcement and
32 management strategies in a large urban network such as New York City. The use of a very detailed
33 microscopic simulation tool for this purpose might not be feasible due to time and budget
34 constraints.

35 As an alternative approach, a stochastic queuing model proposed in Gursoy-Baykal *et al.*
36 (6) to estimate incident induced delays is adopted to estimate average delayed travel time, given
37 the amount and duration of double parking activities. This kind of queuing model applicable to
38 transportation systems, can be applied at the macroscopic level, consists of multiple road segments
39 modeled as ‘individual servers’ that provide service to vehicles that join the queue. It has a set of
40 simplifying assumptions about vehicle arrival distribution, service distribution, and number of
41 servers to derive equations to predict the queue lengths and waiting times. It can be magnitude of
42 order inexpensive and faster to develop and implement such models compared with microscopic
43 simulation models. However, accuracy of such macroscopic models for various spatial-temporal
44 characteristics of a highly complex urban transportation network has to be validated carefully. This
45 paper describes such two models and discusses pros and cons of each approach along with
46 suggestions for improving both models.

1 LITERATURE REVIEW

2 Various studies have been taken into consideration for on-street parking and traffic congestion. In
3 2009, Portilla *et al.* (7) extended a model developed by Baykal-Gürsoy *et al.* (6) to quantify the
4 influence of parking maneuvers and badly parked cars on average link journey times. Baykal-
5 Gürsoy *et al.*'s model (6) was a M/M/ ∞ queueing model subjects to random interruptions of
6 exponentially distributed durations that was originally developed to simulate the effects incidents
7 have on delays that she had proposed in a previous article (8). It describes a queueing system where
8 arrival and service processes of the vehicles are all Poisson processes with infinite number of
9 servers. In Portilla *et al.*'s study, badly parked cars are introduced as time events while parking
10 maneuvers are introduced as high-frequency short duration event. The results showed a 15%-199%
11 increase in average journey time and a 6%-55% reduction in capacity. Guo *et al.* (9) evaluated
12 influence factors of on-street parking by using a proportional hazard-based duration model.
13 Narrow lanes, frequent turnover rates, and parking occupancy show a negative effect on travel
14 time.

15 However, only a few studies focus on the occurrence of double parking. Recent efforts
16 have primarily concentrated on the price strategies or applying microsimulation models to explore
17 on-street parking impact on traffic. Millard-Ball *et al.* (10) evaluated the first two-year
18 performance of SFpark (11) – a demand-responsive rate adjustment parking system embarked by
19 San Francisco Municipal Transportation Agency. Rate changes have helped achieve the city's
20 occupancy goal (60%~80%) and reduced cruising by 50%. Another pilot called "PARK Smart"
21 (12) was implemented in New York City by NYCDOT. Schaller *et al.* (13) summarized that
22 pricing can be effective in achieving the goals of commercial loading availability, increasing
23 turnover, and parking availability. Morillo and Campos (14) showed that the economic cost differs
24 depending on the DP location and the type of typology.

25 Besides pricing strategies and economic impact, Kladeftiras and Antoniou (15) conducted
26 a research project using microsimulation to estimate the impact of average speed, delay and
27 stopped time, as well as environmental impact caused by double parking. An auxiliary lane is
28 created in the model for one-lane roadways for the vehicles to overtake the double-parked vehicle.
29 Results showed that by eliminating double-parking activities, it can increase 10% - 15% in average
30 speed and reduce 15% and 20% in delay and stopped time. Researchers in Italy also studied illegal
31 parking impact in a high-density urban area in Palermo, Italy (16). Results showed that parking
32 durations over 5 minutes would have significant negative impacts.

33 To sum up, most studies in the literature only provide insights into policy and economic
34 aspects of double parking. Only a few micro-simulation models are used to simulate double
35 parking impacts, but these are mainly outside the US and usually shown to be time-consuming and
36 labor intensive for network-wide implementation. Therefore, there is a need to identify a more
37 efficient macroscopic approach such as the queueing based model described in Gürsoy-Baykal *et al.*
38 (6) to study the impact of double parking activities. Moreover, such macroscopic methods have
39 to be validated using real-world data and/or microscopic simulations. The feasibility of using
40 microscopic simulation for modeling double parking under various street configurations has to be
41 established since none of the major simulation tools has a default functionality for modeling double
42 parking. This paper attempts to address both these needs by collecting and analyzing field data and
43 validating a queueing based macroscopic model, as well as demonstrating a Paramics based
44 microscopic simulation model extended through a special application program interface (API)
45 developed by the research team to realistically modeling of double parking on urban streets.

1 DATA DESCRIPTION

2 NYCDOF's parking violation records in 2014 are summarized to investigate the frequency of
 3 double parking activities in NYC. As shown in the following table (TABLE 1), double parking
 4 ranked as the seventh highest parking violation in NYC, which represents 5.0% of the total
 5 NYCDOF's parking violation records in 2014. Since most of the double parking events are
 6 difficult to record, this number can be assumed to be significantly less than the real number of such
 7 violations.

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 9 **TABLE 1 Top 10 NYC Parking Violations in 2014**

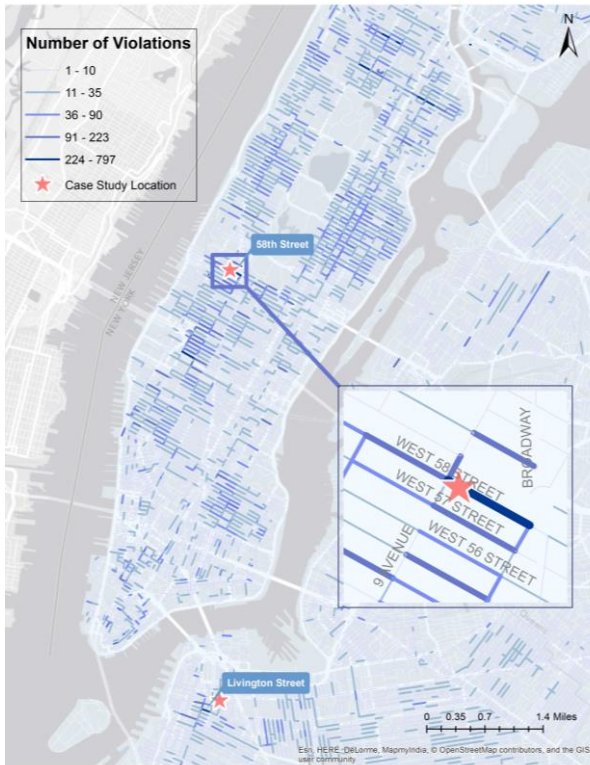
Rank	Violation Code	Tickets	Percentage	Violation Description
1	21	1,337,993	13.4%	Stand or park during street cleaning
2	38	1,302,810	13.1%	Muni meter: failing to show a receipt
3	14	884,828	8.9%	General no standing
4	37	745,096	7.5%	Muni Meter: excess of allowed time
5	20	584,334	5.9%	General no parking
6	71	556,677	5.6%	No Inspection Sticker
7	46	502,082	5.0%	Double parking
8	7	482,051	4.8%	Not stop at red light
9	40	479,191	4.8%	Stand or park beside a fire hydrant
10	36	457,141	4.6%	Exceeding speed limit in school zone

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 11 An automatic geocoding program is developed by the research team based on Google
 12 Geocoding API (17) to convert the original address information into geographic coordinates that
 13 can be employed to geolocate each event on a map.

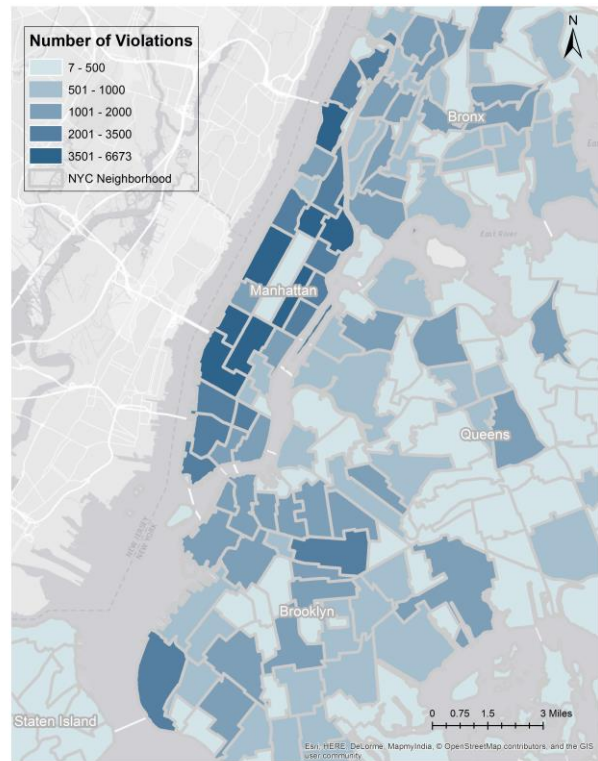
14 Violation records are also visualized onto streets and neighborhoods to identify DP
 15 "hotspots". The goal of the geocoding and visualization is to understand the spatial-temporal
 16 nature of double parking violations in New York City.

17 FIGURE 1 shows the results of four-month data from July to October in 2014. On the
 18 neighborhood level, the neighborhood of Upper East Side, Upper West Side, and Midtown
 19 Manhattan have the highest double parking violation records for all type of vehicles. The
 20 neighborhood of Upper East Side, Upper West Side, and Hudson Yards-Chelsea-Flatiron-Union
 21 Square ranked top three of double parking violation records for commercial vehicles. Double
 22 parking has more violation records in commercial districts or mixed commercial/residential
 23 districts, while commercial DP has less records compare to total DP in boroughs other than
 24 Manhattan that are mainly residential. It must also be noted that Police enforcement is a crucial
 25 factor to the number of summons for passenger cars. However, this may not necessarily reduce the
 26 occurrence of commercial vehicle double parking in large and highly congested urban areas such
 27 as NYC, since their loading and unloading activities are often along a short distance at multiple
 28 destinations with not many opportunities of legal parking during a given time frame.

29 On the street level, 58th Street and Livingston Street ranked the first in Manhattan and
 30 Brooklyn. The highest number is up to 797 total violations and 420 commercial violations on a
 31 single street block in July to October in 2014.



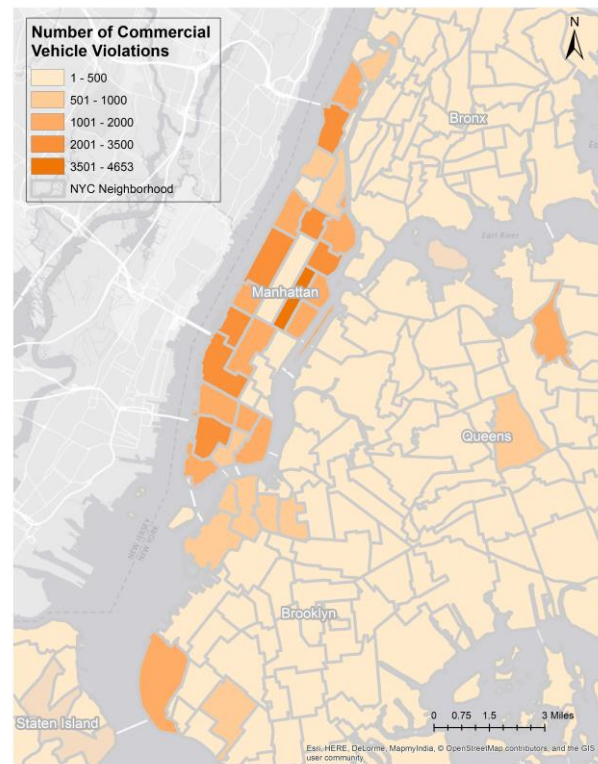
(a) Total DP by street



(b) Total DP by neighborhood



(c) Commercial DP by street



(d) Commercial DP by neighborhood

FIGURE 1 NYC double parking violation tickets, 2014 July to October.

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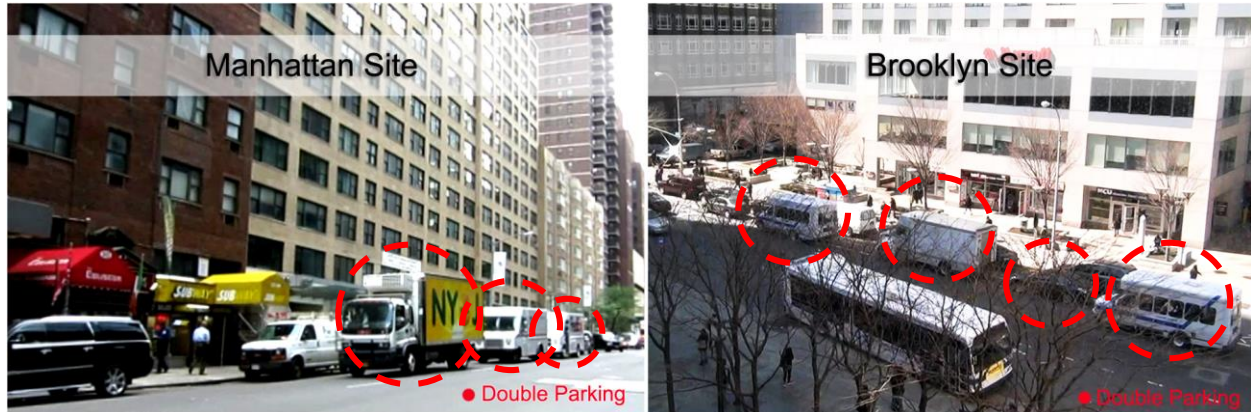
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1 Case Study Sites

2 After identifying the hotspots, two case study sites, where the occurrence of double parking is
 3 found to be significant from the above analysis, were chosen specifically (FIGURE 2).

- 4 1. Manhattan: 58th Street between 8th Avenue and 9th Avenue (highest number of records)
- 5 2. Brooklyn: Jay Street between Myrtle Avenue and Metrotech Center (observation of high
 6 DP occurrence)

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FIGURE 2 Double parking activities in study sites.

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11 Traffic movements were recorded in April, July and November in 2015 during AM (8-
 12 9AM), Midday (12-1PM), and PM (5-6PM) time periods for both sites to obtain traffic flow, speed,
 13 queue length, DP information, and driver behaviors. The following observations are found from
 14 the recorded videos:

- 15 • Double parked trucks have more impact than general cars (e.g. further reduction on the
 16 effective lane width).
- 17 • Manhattan site has a shorter duration of DP events compared to the Brooklyn site.
- 18 • The passing speed is generally higher than the default Paramics incident passing speed
 19 (10MPH) which indicates that double parked events should not be treated same as
 20 incidents.
- 21 • Speed reduction is subject to multiple factors such as the location of DP.
- 22 • For both sites, AM and Midday have noticeable number of DP events (20-30/hour), while
 23 PM has very few DP events (1-2/hour). The regular curbside parking is not fully
 24 occupied and the turnover rate is high during PM period.

25

26 Double parking information includes location, arrival time, departure time, vehicle type
 27 and duration. A sample count sheet is shown in TABLE 2. The location of the DP events is
 28 according to the n th ($n = 1, 2, 3 \dots$) regular curb-parked vehicle measured from the upstream end
 29 of the link. So the “location” shown in TABLE 2 stands for the double parking location parallel to
 30 the regular curb parking. AM period is used for both study sites to better distinguish the delay
 31 caused by double parking from the delay caused by midblock jaywalking pedestrians, which
 32 occurs a lot during Midday.

33

1 **TABLE 2 Sample DP Events**

Site	Location	Arrival Time	Departure Time	Duration	Vehicle Type
Brooklyn	5	8:41:18	8:41:52	0:00:34	Car
Brooklyn	5	8:46:45	8:49:15	0:02:30	Car
Brooklyn	6	8:50:11	8:52:08	0:01:57	Truck
Brooklyn	2	8:51:15	8:51:55	0:00:40	Car
Brooklyn	5	8:52:52	8:56:53	0:04:01	Car
Brooklyn	2	8:57:09	9:04:17	0:07:08	Car
Manhattan	4	8:02:30	8:02:53	0:00:23	Taxi
Manhattan	1	8:04:40	8:06:16	0:01:36	Car
Manhattan	1	8:34:20	8:35:10	0:00:50	Truck
Manhattan	5	8:45:10	8:46:25	0:01:15	Car

2

3 Traffic information is collected and summarized for every 15 minutes (TABLE 3). One
 4 important observation for the Brooklyn site is that the traffic was experiencing severe downstream
 5 blocking during the last 15 minutes of the observed AM hour.

6

7 **TABLE 3 Traffic Information for Case Studies**

Location	Time	Volume	HV%	# DP	Status	Avg.TT(s)	Standard deviation(s)	
Brooklyn	8:00-8:15 AM	66	12.1%	6	DP	5.52	1.19	
					w/o DP	5.26	0.93	
	8:15-8:30 AM	58	19.0%	3	DP	5.47	0.98	
					w/o DP	5.07	0.91	
	8:30-8:45 AM	59	18.6%	4	DP	6.01	1.83	
					w/o DP	6.00	0.82	
	8:45-9:00 AM	65	12.3%	8	DP	14.03	15.49	
					w/o DP	4.61	0.02	
	Manhattan	8:00-8:15 AM	114	17.5%	8	DP	7.86	2.54
						w/o DP	6.74	1.43
8:15-8:30 AM		108	13.0%	6	DP	8.09	1.27	
					w/o DP	7.77	0.12	
8:30-8:45 AM		107	15.9%	6	DP	8.15	1.40	
					w/o DP	6.93	0.98	
8:45-9:00 AM		127	29.9%	7	DP	8.57	0.63	
					w/o DP	6.74	1.14	

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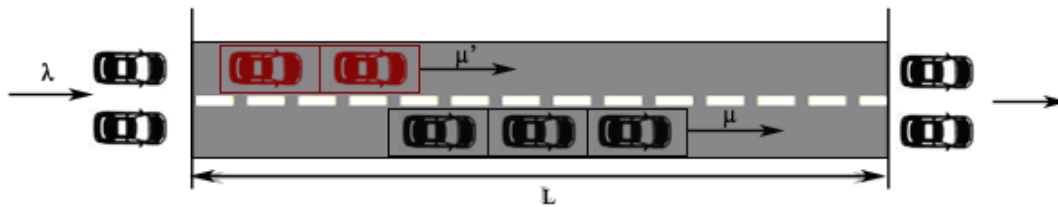
9 **METHODOLOGY**10 **M/M/∞ queueing model**

11 A queueing model assumes that the space occupied by an vehicle on the roadway link represents
 12 one queueing “server”, which starts its service once a vehicle joins the link and carries the
 13 “service” until the end of the link is reached (18). The notation of queueing models is classified by
 14 arrival distribution, service distribution and the number of servers. A queueing model that has a
 15 Markovian arrival rate, a Markov modulated service rates, and an infinite number of servers in the
 16 system is called M/M/∞ queueing model. A Markov process defines discrete states and the

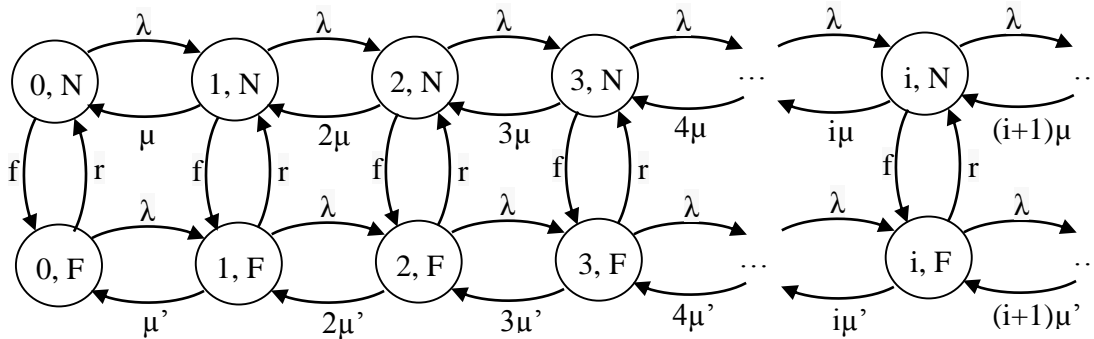
1 transition probability from one state to another. One of the key features is that it solely depends on
 2 the current state. The M/M/∞ model is a stochastic process that has the following features:

- 3 • Arrivals process is a Poisson process at rate λ .
- 4 • Exponential service times with parameter μ . There are always sufficient servers in the
 5 system that every vehicle is served immediately after arriving.
- 6 • When a vehicle arrives or departs, the system moves to an adjacent state.

7 When double parking occurs, it indicates the high occupancy of the curbside parking lane.
 8 This is also observed in the video with a turnover rate of only one or two vehicles per hour for
 9 regular curbside parking. Thus, the regular curbside parking impact is negligible in this case study.
 10 Adopted from (6, 7, 18), the system is designed with two server states “Failure (F)” or “Normal
 11 (N)” to represent the conditions with or without double parking events. Let r denote the rate of DP
 12 clearance time and f the frequency of DP events. The system will be in state (i, F) if there are i
 13 vehicles in the system that are interrupted by double-parked vehicles, while the system will be in
 14 state (i, N) if there are i vehicles in the system that are travelling without interruption. The
 15 depiction of the modified M/M/∞ queueing model on a two-lane link is shown in FIGURE 3. The
 16 servers in the systems will work at a low service rate μ' when a double parking event happens
 17 (FIGURE 4).



18 **FIGURE 3 Modified M/M/∞ queueing model on a two-lane link (based on (6)).**



20 **FIGURE 4 State transitions for M/M/∞ model with two server states (based on(18)).**

21
 22 One important result from Baykal-Gursoy and Xiao (8) showed that the expected number of
 23 vehicles on the link can be represented as below when the M/M/∞ queueing system is experiencing
 24 service interruptions:

$$25 \quad E(X) = \frac{\lambda}{\mu} + \frac{\lambda f(\mu - \mu')}{\mu^2(r + f)} \left(1 + \frac{(f + \mu)(\mu - \mu')}{(r\mu + f\mu' + \mu\mu')} \right) \quad (1)$$

26
 27 Average travel time on the link is:

$$W = \frac{E(X)}{\lambda} = \frac{1}{\mu} + \frac{f(\mu - \mu')}{\mu^2(r + f)} \left(1 + \frac{(f + \mu)(\mu - \mu')}{(r\mu + f\mu' + \mu\mu')}\right) \quad (2)$$

To reflect road segment characteristics, Equation (2) is reformatted by Portilla *et al.* (7) to compute average travel time t :

$$t = \frac{L}{v} \left[1 + \frac{f(1 - \frac{v'}{v})}{\frac{1}{d} + f} \left(1 + \frac{(f + \frac{v}{L})(v - v')}{(\frac{v}{d} + fv' + \frac{v - v'}{L})}\right)\right] \quad (3)$$

Where L =length of the link (mile); $D=\lambda$ =traffic demand (veh/h); f =frequency of double parking (events/hour), $d=1/r$ =average duration time of double parking (h); $v=\mu L$ =average speed without double parking (mph); and $v'=\mu' L$ =average speed with double parking (mph).

With the intention of accommodating commercial vehicle violations, a double parked commercial vehicle is treated as two double parking events in this study. Moreover, double parking impact can vary among different street segments due to traffic volume, number of lanes, gross leasable area (GLA) of commercial properties and so on. To capture these effects, equation (3) is modified by adjusting a correction factor C_f as follows:

$$t = C_f \frac{L}{v} \left[1 + \frac{f(1 - \frac{v'}{v})}{\frac{1}{d} + f} \left(1 + \frac{(f + \frac{v}{L})(v - v')}{(\frac{v}{d} + fv' + \frac{v - v'}{L})}\right)\right] \quad (4)$$

And,

$$C_f = \frac{1}{N} \sum_{i=1}^N \frac{\text{Real TT}}{\text{Model TT}} \quad (5)$$

Where *Real TT* = field travel time (s), *Model TT* = queueing model travel time (s), N = number of observations.

We propose a correction factor of $C_f = 1.07$ for Manhattan site and $C_f = 1.02$ for Brooklyn site based on the observed data. However, since the observed data is limited, we are not capable of generalizing C_f for all possible influencing factors. This will be the focus of future study where automated data collection methods can be deployed.

Microsimulation Model

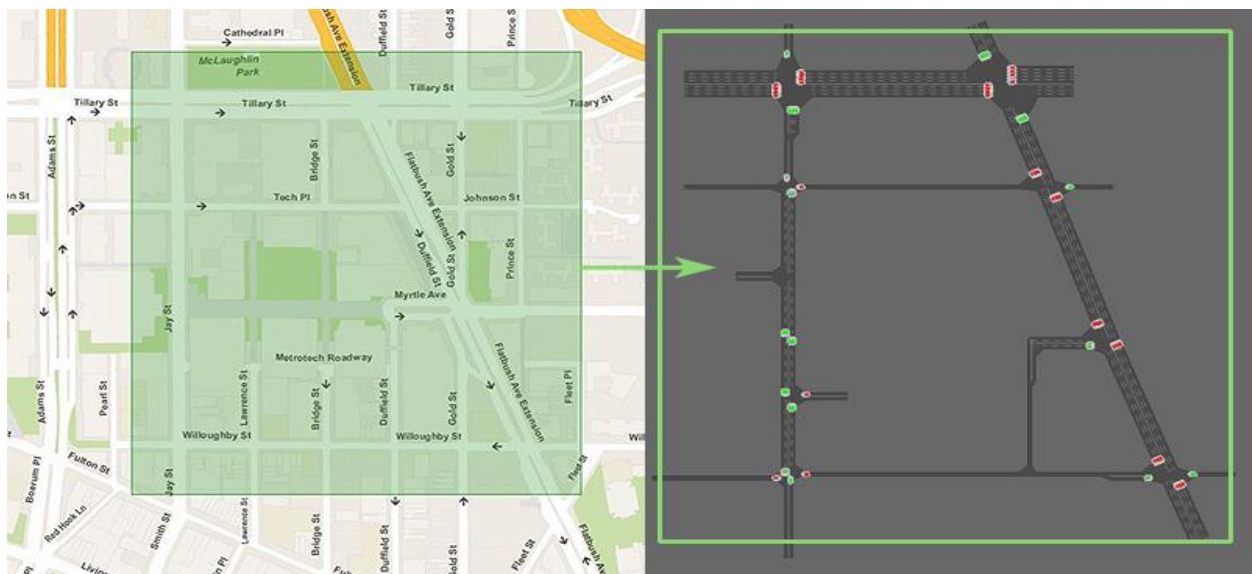
A Paramics (19) micro-simulation (FIGURE 5) is developed for the Brooklyn site to simulate complex conditions especially when downstream blocking happens. The model is bounded by Jay Street to the West, Flatbush Avenue Extension to the East, Tillary Street to the North, and Willoughby Street to the South. The particular interest of this study focuses on the road segment between Myrtle Avenue and Metrotech Center on Jay Street where DP data is available. It's a two-way segment has one 10-ft travel lane, one 4-ft bike lane and one 10-ft curbside parking lane in each direction and is referred as the reference link in this study.

Paramics is a high-performance software that model the movement and behavior of individual vehicles (19). To simulate double parking activities, an "incident" file coded in C programming language is used to specify where, when and how long the "double parking event" should occur. Double parking is programmed as a special incident type that has the event duration,

1 lane specification and overspills. The occurrence can be defined as either random or fixed on each
 2 link. In order to test the location impact of DP event, this case study uses fixed incident with
 3 defined distance and time.

4 Since the incident feature does not work well for single-lane situations while the vehicle
 5 behind the double-parked event will not move until the event is cleared, a narrow auxiliary lane is
 6 used along with a customized API plug-in to allow vehicles to overtake the DP vehicle. The API
 7 plug-in is developed to slow down vehicle on a certain distance of the road link to simulate driver
 8 behaviors when passing a DP vehicle. The location, length, and speed of the slow-down area can
 9 be customized. The micro-simulation model is calibrated based on field traffic counts, average link
 10 travel time and observed behaviors collected in April, 2015. It is also validated with November's
 11 data in the same year to demonstrate reasonable prediction capability. Validation results shows a
 12 GEH (20) of traffic volume less than 0.20 and a root mean square error (RMSE) of travel time less
 13 than 1.20 for all time intervals.

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15 **FIGURE 5 Case study area in Downtown Brooklyn and Paramics network.**

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The Traffic Engineering Handbook (21), along with other publications (7, 22), have addressed the effect of curb parking, mainly caused by the lack of a curb lane plus the parking activity next to a moving traffic stream. To estimate the impact on the average travel time due to different factors under DP condition, 21 different scenarios were designed into four categories:

Category 1: Location of Double Parking Events Scenarios

Scenario 1-3: Double parking events happen at the beginning of the block, midblock, and the end of the block

Category 2: Demand Scenarios

Scenario 4-7: 110%, 120%, 130%, 140% of morning peak demand

Category 3: Frequency of Double Parking Events Scenarios

Scenario 8-11: Increase one, two, three, and four double parking event every 15 minutes

Category 4: Duration of Double Parking Events Scenarios

Scenario 12-21: Average duration of the events is increased by one minute in each scenario (e.g. Scenario 12 has one-minute average duration, Scenario 21 has 10-minute average duration).

RESULTS AND DISCUSSION

First, M/M/∞ queueing model is applied to estimate average travel time with a 15 minute time interval. FIGURE 6 and TABLE 4 provide a summary of the results. The result implies that the M/M/∞ queueing model has a percentage difference of less than 8% for both sites in terms of average travel times compared with the actual field travel times when no downstream queueing exists. While applying correction factors to the queueing model, the percentage difference is further reduced to less than 4%. In this study, the correction factor is tied to site-specific conditions. However, by introducing it, we intend to introduce the idea of a general factor that will help modelers to capture the impact of factors on traffic as a result of double parking. Many parameters such as traffic volume, number of lanes, GLA of commercial properties may have an effect on the correction factor. Further study and data collections are needed to define it as a general function in terms of all these different characteristics of city streets as part of ongoing efforts.

Average field travel time significantly increased due to the downstream blocking of the Brooklyn site in the last 15 minutes of the observed hour (FIGURE 6(a)). Under this circumstance, M/M/∞ queueing model is found to be incapable of capturing certain spillover conditions which resulted in the difference between the field data and the queueing model output. The reason for this, is that M/M/∞ queueing model assumes independence of different queues, as well as the independence of the arrival of double parking events, and does not account for spatial queue or queue capacity.

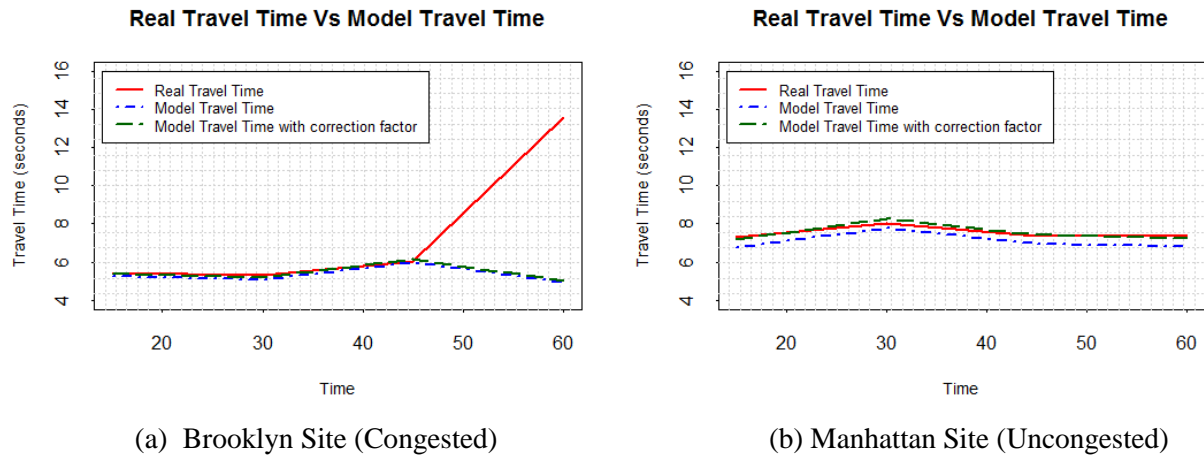


FIGURE 6 M/M/∞ queueing model results.

TABLE 4 Summary of Field, Queueing Model, and Microsimulation Result

Site	Time	Field TT	Model TT	Model TT with C_f	Microsimulation TT
Brooklyn	8:00-8:15 AM	5.41	5.27	5.39	5.54
	8:15-8:30 AM	5.30	5.08	5.20	5.83
	8:30-8:45 AM	6.01	6.00	6.14	6.21
	8:45-9:00 AM	13.60	4.95	5.07	12.91
Manhattan	8:00-8:15 AM	7.32	6.77	7.21	-
	8:15-8:30 AM	8.03	7.78	8.29	-
	8:30-8:45 AM	7.38	6.98	7.43	-
	8:45-9:00 AM	7.40	6.82	7.26	-

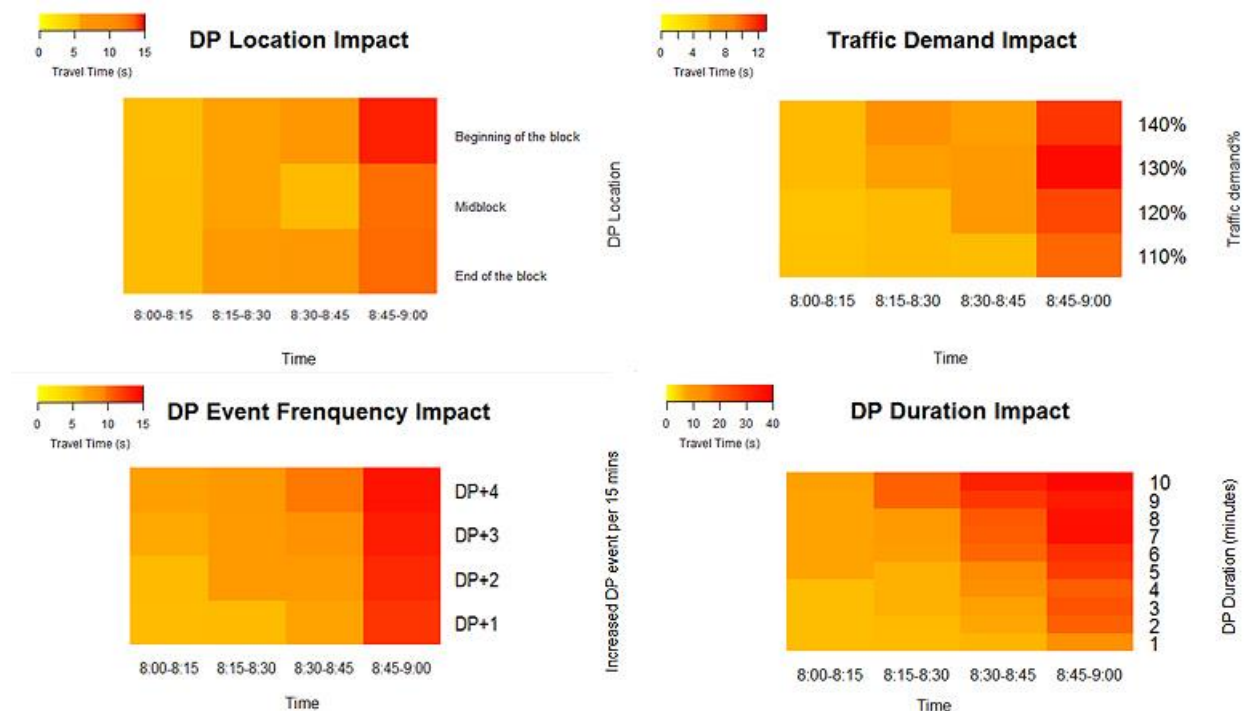
1 The micro-simulation model is then used to quantify complex characteristics of double
 2 parking events such as DP location. Travel time results are shown in the following heat maps
 3 (FIGURE 7) for the four scenarios described in the previous section.

4 The first category tested the impact due to different DP locations. DP events at the end of
 5 the block and midblock have similar mild effects on the average TT. Events occurring at the
 6 beginning of the block have a 5.15% higher impact in terms of the average TT increase over the
 7 AM peak hour. This matches the field observation that once a vehicle is double-parked near the
 8 beginning of the block (close to the upstream end of the road in the direction of traffic flow),
 9 drivers are more likely to slow down. DP vehicle under this condition usually reduces the effective
 10 lane width and causes potential conflicts with other activities at the intersection such as pedestrian
 11 crossing.

12 The second category of simulation scenarios examined the impact on the travel time of
 13 applying 110%, 120%, 130% and 140% of morning peak demand. Generally, higher demand
 14 resulted in higher travel time, as expected. The increase of the average travel time is 7.3%, 18.5%
 15 and 18.6% for 20%, 30% and 40% increase in the traffic demand, respectively, while 10% increase
 16 in demand does not cause a significant change in average travel time.

17 The third category of simulation scenarios accounted for the travel time effect if the
 18 frequency of DP event increased. The hourly average travel time increased by 3.1%, 13.6%,
 19 20.5%, and 27.6% when there is an increase of DP event frequency by 1, 2, 3, and 4 vehicles per
 20 15 minutes, respectively.

21 The fourth category of scenarios specifies various duration time for DP events from 1
 22 minute to 10 minutes. As shown in FIGURE 7, the longer the event goes on, the greater the increase
 23 in the average travel time. Average travel time can be more than two times higher in the worst-
 24 case scenario where the average duration of DP event is 10 minutes.
 25



26
 27 **FIGURE 7** Microsimulation results.

1 CONCLUSIONS AND FUTURE WORK

2 This study examined a macroscopic M/M/∞ queueing model and micro-simulation for
3 estimating average travel time in the presence of double parking activities. Under uncongested
4 traffic conditions without downstream blocking, applying the M/M/∞ queueing model produced
5 a good fit with the field data. The observations obtained from video recording were used to develop
6 and calibrate a microsimulation model for one of the two case study areas namely, Downtown
7 Brooklyn. Simulation output shows that the location, frequency, and duration of the double parking
8 event and overall traffic demand have a certain impact on the average travel time.

9 Overall, M/M/∞ queueing model is an effective approach to compute average travel time
10 for unsaturated traffic flows under double parking conditions. For a large urban network, it has the
11 potential to be implemented instead of micro-simulation models mainly due to its computational
12 efficiency, if it can be combined with an appropriate correction factor. Effect of commercial
13 vehicles is also estimated in the model where one double-parked truck is assumed to account for
14 two DP events. However, when downstream blocking happens, the traffic condition does not
15 completely depend on what is happening on the current road segment. Micro-simulation is a more
16 powerful tool than the M/M/∞ queueing model for evaluating such congested scenarios and can
17 be used to examine individual and combined effects of various explanatory variables.

18 Future research will focus on collecting more field data using automated image processing
19 techniques, investigating specific impacts of double-parked commercial vehicles, exploration of
20 the accurate estimation of the correction factor function, correction factor for various spatio-
21 temporal characteristics of the transportation network, and developing new APIs to improve the
22 micro-simulation model for more challenging geometric and traffic conditions.

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