

Research Article

Formation and Propagation of Local Traffic Jam

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Large scale traffic congestion often stems from local traffic jam in single road or intersection. In this paper, macroscopic method was used to explore the formation and propagation of local traffic jam. It is found that (1) the propagation of traffic jam can be seen as the propagation of traffic signal parameters, that is, virtual split and virtual green time; (2) for a road with endogenous flow, entrance location influences the jam propagation. With the same demand (upstream links flow and entrance flow), the upstream got more influence; (3) when a one-lane road is thoroughly congested, virtual signal parameters everywhere are the same as that at stop line; for a basic road, the virtual signals work in a cooperative manner; (4) phase sequence is one important parameter that influences traffic performances during peak hour where spill back of channelization takes place. The same phase plan for left-turn flow and through flow would be preferred; (5) signal coordination plays an important role in traffic jam propagation and hence effective network signal parameters should be designed to prevent jam from propagation to the whole network. These findings would serve as a basis for future network traffic congestion control.

1. Introduction

Urban road network has been considered as the “life line” of urban daily life. It serves as the basis of urban economy. The assurance of its normal operation is very important. However, as the development of urbanization, the number of vehicles relatively exceeds the capacity of road network which often results in large scale traffic jam during peak hours. This leads to many problems such as pollution, and noise. Large scale traffic jam often stems from local jam which takes place in a road or single node. According to Wright and Roberg [1], there are three sources of traffic jam: a temporary obstruction, stochastic fluctuation in demand, and permanent capacity bottleneck. These types of congestion cannot be prevented efficiently because they are related tightly with road geometry, land usage, or other factors which cannot be changed easily. However, if we can control the propagation of traffic jam, then, network traffic jam will be prevented. The propagation process looks like “domino phenomena” which is characterized by connected set of events. Many literatures are focusing on this subject.

Simulation methods including microscopic and macroscopic types are among the most used methods. Microscopic tools include cellular automata [2] and car following models [3]. They are based on dynamic motion of single vehicle and hence can grasp the traffic performance of the whole network [1, 4, 5]; macroscopic tools often adopted a discrete form [6, 7] of LWR model which stemmed from Lighthill and Whitham [8] and Richards [9]. This method divided links into a series of consecutive cells and time into equal time steps. By checking and tracking basic traffic parameters (flow, density, and velocity) of every cell during each step, the dynamic nature of traffic jam can be derived. This kind of method greatly relies on underlying models and the simulation networks are often regular [10, 11] which lacks generality. Furthermore, no analytical conclusion can be obtained through this approach.

Some researchers tried to establish analytical results of traffic jam. Wright and Roberg [1] proposed a simple analytical model for incident-based jam growth and discussed the effect of the length of the channelized part of roads and stop line width assignment on jam formation; Michalopoulos

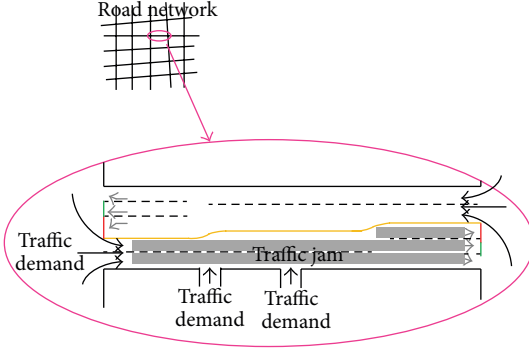


FIGURE 1: Simple representation of road geometry.

et al. [12] formulated a rigorous but complex mathematical modeling for signalized links; Skabardonis and Geroliminis [13, 14] developed an analytical model based on kinetic wave theory to construct monitor methods for arterials. Based on the derivation of waves produced by traffic signal, vehicle trajectory was identified and hence travel time function was formulated; Daganzo [15] divided the network into connected neighborhoods and constructed the relationship between defined traffic variables and formulated the so-called “gridlock” condition. Through these researches we can get a deep understanding of traffic jam. If we examine the general road network, we can see that it is comprised of basic road (some may be simpler) as shown in Figure 1. The road is multilane structure controlled by traffic signal. The traffic demand may come from upstream links or entrance within roads. Compared with this situation, current researches are not enough to describe traffic flow operation on basic road segment, not to mention the network traffic jam propagation.

Some scholars used data collected from control system (mainly loop detectors) to monitor long queue and spillover [16, 17]. However, under congestion hour traffic jam involves various traffic flows that include multilane flow and endogenous flow from within the entrance, and these are difficult to distinguish from loop detector data. These factors may undermine the applicability of these methods.

The dynamic nature of network traffic jam (critical condition for its formation, propagation speed, propagation route, and so on) is complex in that they are influenced by many factors such as road physical characteristics, network topology, and control parameters. Understanding of these relationships is important and premise for further control and management. However, they were not fully accounted for in traditional methods. In this paper, with the consideration that network jam always stems from traffic jam formed within single road, attention was put on the traffic jam formation on signal controlled network comprised of basic road as in Figure 1 that is generally existed. After description of critical condition of jam formation for one-way link (Section 2.2), decomposition method is used to divide the road into two multilane segment (Section 2.3). Factors include insufficient split or excess demand (Section 3.1), endogenous flow (Section 3.2) and channelization spillback (Section 3.3) are discussed. The influences of road spillback on upstream links

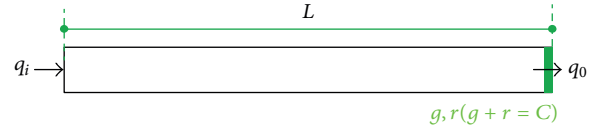


FIGURE 2: A Single one-way link.

are then analyzed (Section 4). We end with some highlights for further researches.

2. Preliminary Results

2.1. Notations

q_m : optimal flow rate,

k_m : optimal density,

k_j : jam density,

g, r, C, λ : green time, red time, cycle and split respectively,

r_v : virtual red time,

u_0, u_1, u_2 : stopping wave, starting wave, and wave that emerges after u_0 and u_2 meet,

l_{max} : maximal queue length,

q_{ol}, q_{os}, q_{or} : output flow rate of left-turn flow, through flow and right-turn flow. “o” means output,

g_{ol}, g_{os}, g_{or} : green time for left-turn flow, through flow and right-turn flow.

2.2. Wave Dynamics under Simple Condition. Consider a one-lane road controlled by signal without endogenous flow as shown in Figure 2 which is the simplest condition in the road network (such as a road regulated by one-way traffic organization), assuming that fundamental diagram of traffic flow is parabolic function as shown in Figure 3. Upstream flow is q (point A in Figure 2) and from parabolic function k can be obtained (density of upstream flow is always smaller than optimum density under maximum flow; that is, point A should be on left side of the curve, otherwise stopping wave speed would be bigger than starting wave which denotes that queue would not disperse forever, which can be seen from Figure 3):

$$k = \frac{k_j - \sqrt{k_j^2 - 4(k_m^2/q_m)q}}{2}. \quad (1)$$

Formation and dispersion of a queue behind stop line at signal controlled intersection are shown in Figure 4(a). At first (the beginning of red time), stopping wave (red lines in Figure 4 which represent queue back) propagates upstream with velocity u_0 and a queue forms. When green time begins a starting wave (blue line in Figure 4) emerges and also propagates upstream with bigger speed (denoted as u_1). After t' starting wave catches up with stopping wave and queue has dispersed and a new wave (denoted by its speed u_2) forms. It takes t'' for wave u_2 to pass through stop line. If green time, g ,

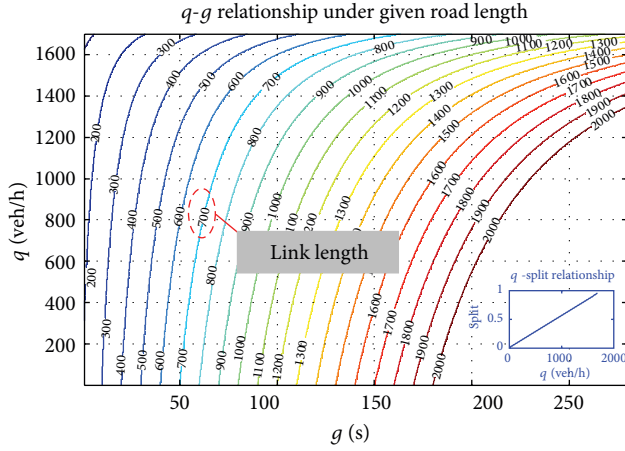


FIGURE 5: Critical q - g relationship under different link length.

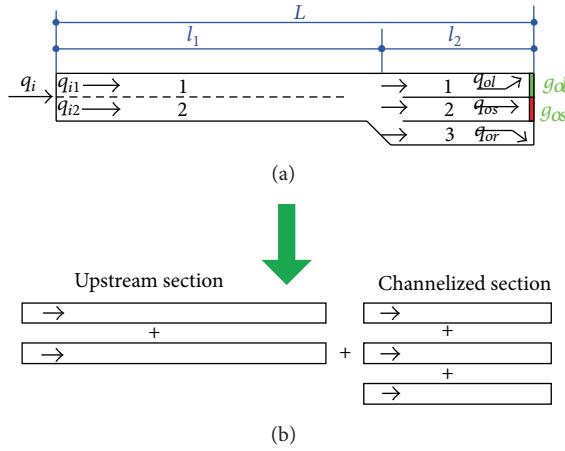


FIGURE 6: Sketch map of basic road and its decomposition.

same symbol expression method is for signal parameters, that is, g_{ol} , g_{os} , r_{ol} , and r_{os} , so we can get

$$\begin{aligned}
 q_i &= q_{ol} + q_{os} + q_{or}, \\
 q_{ol} &= q_i p_{ol}, \\
 q_{os} &= q_i p_{os}, \\
 q_{or} &= q_i p_{or}.
 \end{aligned} \tag{5}$$

Decompose the basic road as shown in Figure 6(b). The basic road has been divided into two sections: upstream section and channelized section. Both are comprised of some single one-way links. Channelized sections are controlled by traffic signal (except right-turn section). The analysis hence is simplified that the method in Section 2 can be used in further analysis.

3. Complex Circumstances

During daily operation, circumstances often hold more complex nature. This section deals with this problem by analyzing

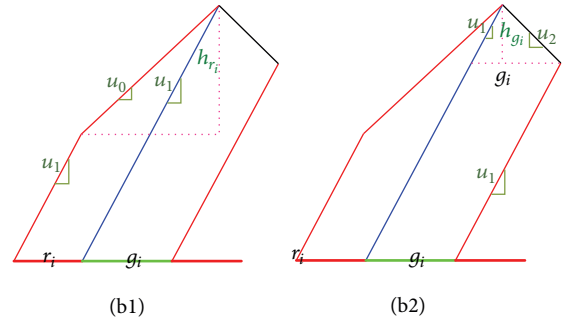
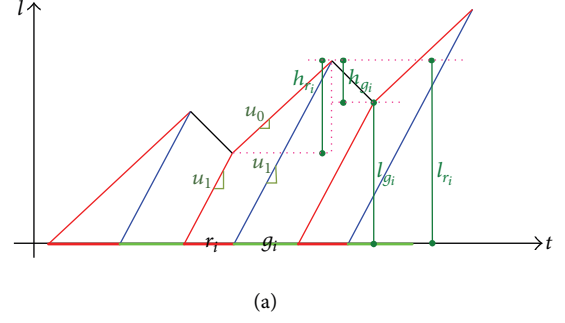


FIGURE 7: Stopping wave and starting wave with insufficient split.

dynamic and insufficient split, endogenous flow of basic road, and the spillover of channelized section.

3.1. Dynamic and Insufficient Split. Usually oversaturation is developed gradually and queue back extends cycle by cycle, that is, domino phenomenon is the result of oversaturation extending in several cycles as shown in Figure 7(a). Each green signal tends to pull back queue back while each red signal prolongs queue back. Furthermore, the length pulled back or prolonged is proportional to time duration of traffic signal because in the figure slope of wave trajectory; that is, wave speed including starting wave and stopping wave is fixed.

Given green time g_i of cycle i (or red time r_i) queue length h_{g_i} pulled back (h_{r_i} prolonged) can be computed from Figures 7(b1) and 7(b2):

$$\begin{aligned}
 h_{r_i} &= \frac{r_i u_0 u_1}{u_1 - u_0}, \\
 h_{g_i} &= \frac{g_i u_2 u_1}{u_1 + u_2}.
 \end{aligned} \tag{6}$$

After i th red time r_i (i th green time g_i) location of queue back is

$$\begin{aligned}
 l_{r_i} &= l_0 + \sum_{i-1} (h_{r_i} - h_{g_i}) + h_{r_i} = l_{g_{i-1}} + h_{r_i}, \\
 l_{g_i} &= l_0 + \sum_i (h_{r_i} - h_{g_i}),
 \end{aligned} \tag{7}$$

where l_0 denotes the length of initial oversaturated queue. It can also be calculated easily by assuming that the traffic is

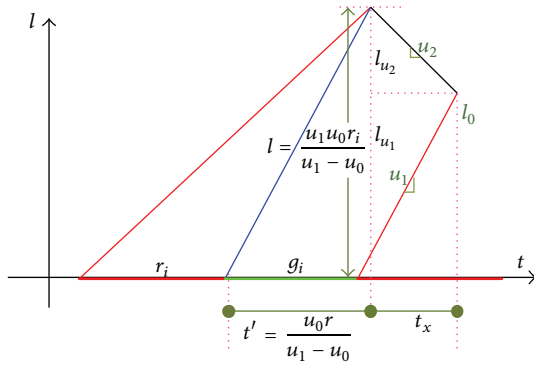


FIGURE 8: Initial oversaturated queue.

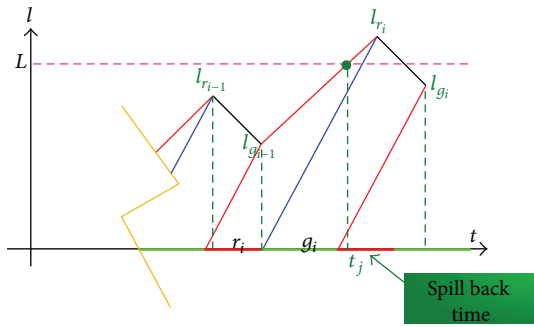


FIGURE 9: Spillover of road queue.

undersaturated at first, as shown in Figure 8. Symbols and derivations are as follows:

$$\begin{aligned}
 l_{u_2} + l_{u_1} &= l, \\
 u_2 t_x + (t_x + t' - g_i) u_1 &= l, \\
 t_x &= \frac{l - (t' - g_i) u_1}{u_1 + u_2}, \\
 l_0 &= (t_x + t' - g_i) u_1, \\
 l_0 &= \left(\frac{l - (t' - g_i) u_1}{u_1 + u_2} + t' - g_i \right) u_1.
 \end{aligned} \tag{8}$$

Replace l_0 in (7) queue back can be calculated at any time. Given changeable r_i and g_i and upstream traffic demand q , h_{r_i} and h_{g_i} hence each l_{r_i} also can be determined. When queue back exceeds road length, that is, $l_{r_i} > L$ then spillover emerges (t_j in Figure 9 is the time when queue back exceeds road length).

However, spillover itself does not result domino phenomena because there may be no upstream traffic demand, this problem will be discussed in Section 4.

In order to describe jam propagation upstream, the method of SVS (speed of virtual signal) is used. The principle is that there is a virtual signal located everywhere. When the jam has not reached a specific location, the virtual signal

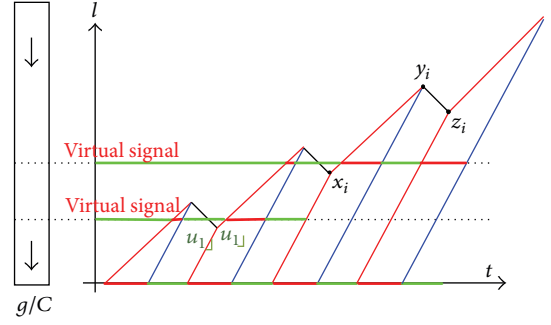


FIGURE 10: Method of virtual signal.

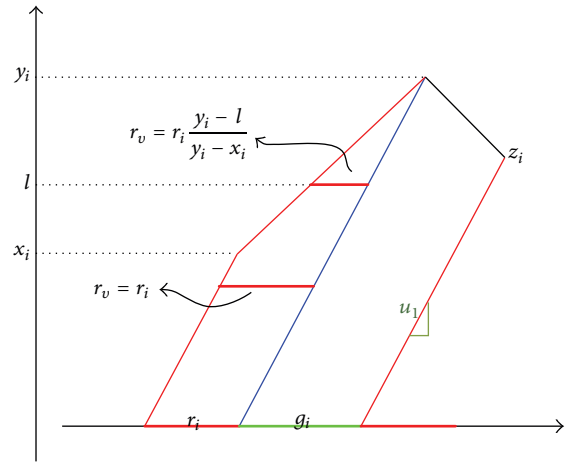


FIGURE 11: Virtual split calculation.

is always green. When the location is occupied by a queue (i.e., jam propagates to this position), the virtual signal shows red. When the queue is cleared, the virtual signal turns to green. From this description, we can see that the propagation of jam can be transformed to the propagation of virtual signal parameter. We use split to explore the propagation of jam. If we obtain locations of x_i , y_i , z_i shown in Figure 10, then we can derive the split dynamics along time and space (see Figure 11). If demand exceeds supply, at the end of each cycle, the residual vehicle number is $Cq - gq_m$ (the queue that remains after past cycles is not included in this formula), so three locations for each cycle can be derived:

$$\begin{aligned}
 x_i &= x_{i-1} + \frac{Cq - gq_m}{k_j}, \\
 y_i &= x_i + h_{r_i}, \\
 z_i &= y_i - h_{g_{i-1}}.
 \end{aligned} \tag{9}$$

After obtaining x_i , y_i , z_i for every cycle, together with Figure 11, split everywhere for any cycle can be determined.

Figures 12 and 13 present a simple numerical example for propagation of split. From Figure 12, for a specified location, the gradual change of virtual signal cycle by cycle can be easily seen. Figure 13 gives the split on space-time (expressed as

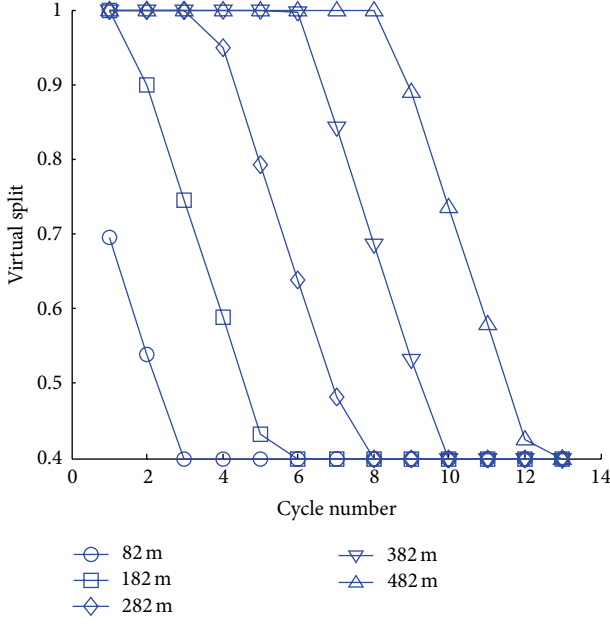


FIGURE 12: Virtual split propagation, $C = 100$ s, $g = 40$ s, $r = 50$ s, $q = 1000$ veh/h, $k_j = 180$ veh/km, $q_m = 1800$ veh/h.

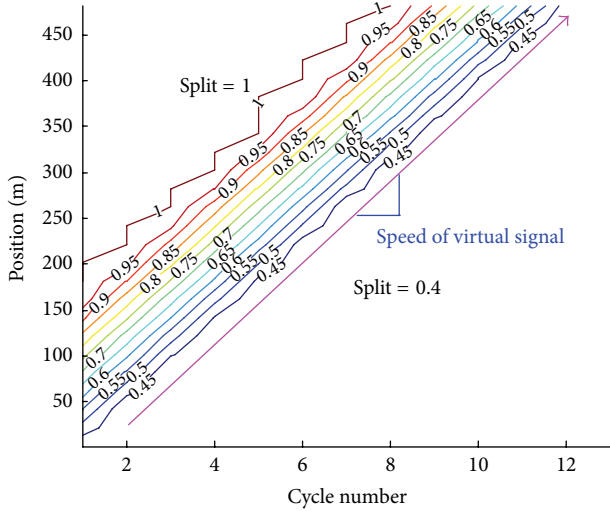


FIGURE 13: Contour of split on space-time diagram and parameters are the same as in Figure 12.

cycle number) diagram. The slope can be seen as the speed of jam propagation.

In order to develop a measure of this speed, an index which incorporates stopping wave and starting wave is proposed later on. For a cycle of jam propagation, the increment of queue back can be got:

$$\Delta l_i = h_{r_i} - h_{g_{i-1}}. \quad (10)$$

If the control parameters are constant, then we define speed of virtual signal (SVS) to express the severity of jam,

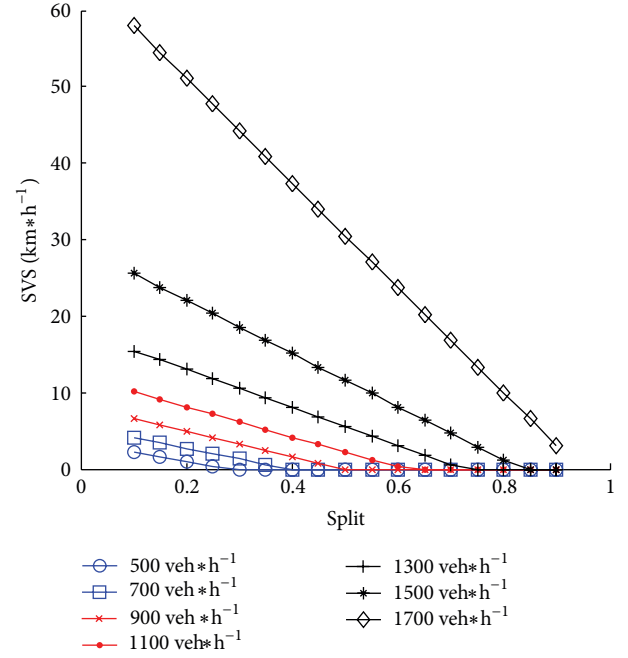


FIGURE 14: SVS versus split; $k_j = 180$ veh/km, $q_m = 1800$ veh/h.

that is, the propagation of congestion. During a cycle, queue back propagates Δl_i , and the average speed is then

$$\begin{aligned} \text{SVS} &= \frac{\Delta l_i}{C} = \frac{r_i u_0 u_1 / (u_1 - u_0) - g_i u_2 u_1 / (u_1 + u_2)}{C} \\ &= \frac{(1 - \lambda) u_0 u_1}{u_1 - u_0} - \frac{\lambda u_2 u_1}{u_1 + u_2}, \end{aligned} \quad (11)$$

$$\text{SVS} = \frac{u_0 u_1}{u_1 - u_0} - \lambda \left[\frac{u_2 u_1}{u_1 + u_2} + \frac{u_0 u_1}{u_1 - u_0} \right] = f(q, \lambda). \quad (12)$$

From the expression we can see that SVS is in linear relationship with split.

Figures 14 and 15 present the relationship between SVS and split, arriving flow rate, respectively. From the figure we can see that SVS is very sensitive to traffic state under saturation condition. The results above are based on parabolic fundamental, which is symmetrical about optimal flow. Figures 16 and 17 plot the SVS versus split and arriving flow under triangular fundamental diagram like Figure 3(b). The parameters are given beneath the figure.

Similar trend is also found for triangular fundamental diagram. However, it is evident from both figures that change of SVS is somewhat moderate compared with parabolic diagram. This phenomena stem from the different nature embedded in these two fundamental diagrams: wave speed for triangular diagram is constant when the density is below k_m , but for parabolic fundamental diagram, wave speed (tangent line of specific point) is changing simultaneously with traffic demand.

Equation (12) assumes that the arriving flow is constant; however, it is not the case. But we can generalize it to dynamic

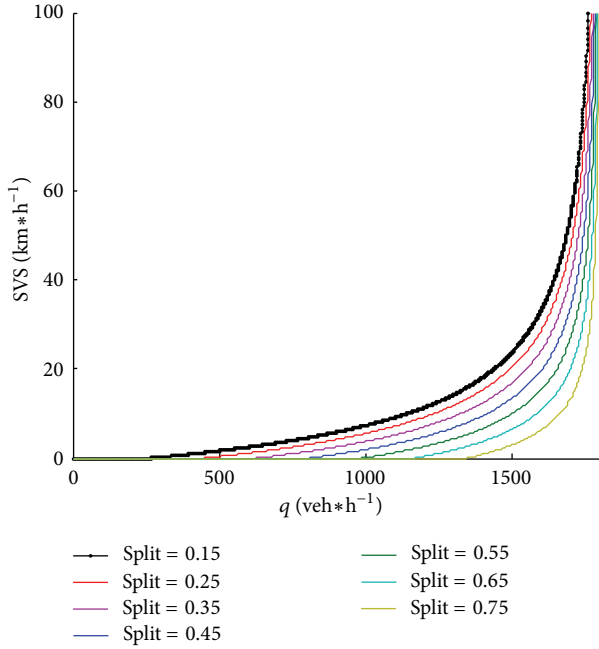


FIGURE 15: SVS versus arriving flow rate; $k_j = 180$ veh/km, $q_m = 1800$ veh/h.

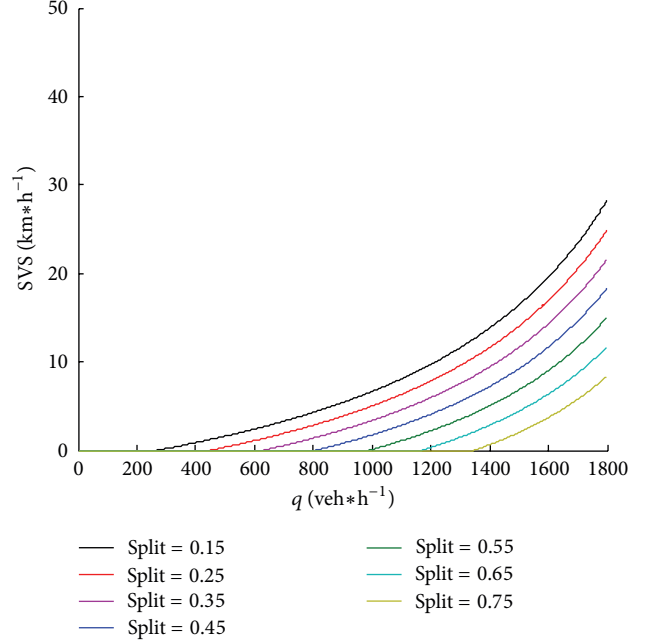


FIGURE 17: SVS versus arriving flow rate with triangular fundamental diagram; $k_j = 180$ veh/km, $q_m = 1800$ veh/h, $v_f = 50$ km/h, and $w_a = 12.5$ km/h.

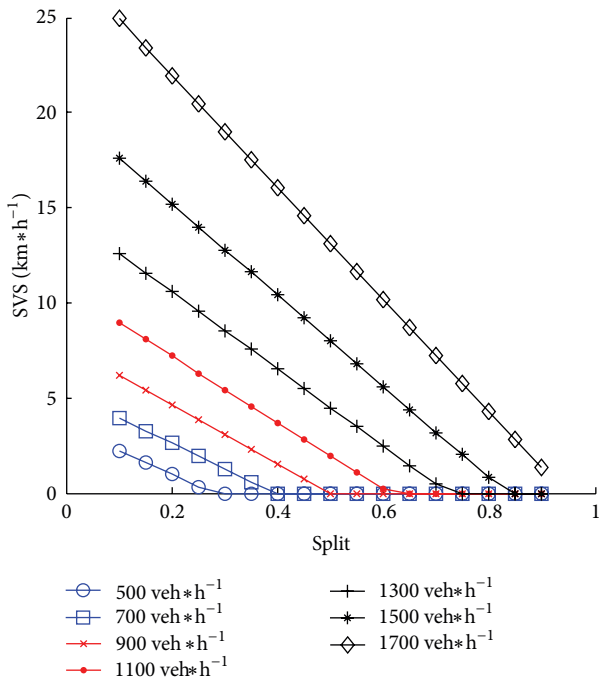


FIGURE 16: SVS versus split with triangular fundamental diagram; $k_j = 180$ veh/km, $q_m = 1800$ veh/h, $v_f = 50$ km/h, and $w_a = 12.5$ km/h.

condition where within one cycle the arriving is stable but different from cycle to cycle, that is,

$$SVS = \frac{u_0(i) u_1}{u_1 - u_0(i)} - \lambda \left[\frac{u_2(i) u_1}{u_1 + u_2(i)} + \frac{u_0(i) u_1}{u_1 - u_0(i)} \right], \quad (13)$$

where $u_0(i)$ is stop wave speed at cycle i .

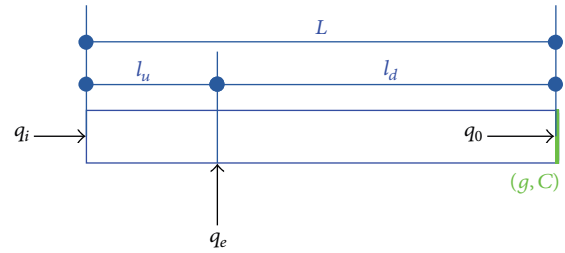


FIGURE 18: Single road with endogenous traffic flow.

3.2. Influence of Endogenous Flow. Traditionally it is assumed that flow is generated at nodes (i.e., intersections within the network or originations of a road network) as we did in network traffic analysis [18]. However, intersections inside road network actually do not generate flow (of course, from the viewpoint of downstream intersection, flow is “produced” at upstream intersection). In fact, most traffic stems from entrance within road (e.g., shopping malls, schools, and official departments about the road). It is necessary to take endogenous traffic flow into account. Due to complexity in analyzing multilane road, we only deal with one-way road with entrance inside which the flow is denoted as q_e (shown in Figure 18). The basic assumption is major road priority; that is, vehicles in major road have priority to cross.

When a formed queue back has not reached entrance location, the upstream flow is $q_i + q_e$; when queue length exceeds l_d (distance between stop line and endogenous flow entrance location) upstream flow becomes q_i , so the stopping wave and starting wave propagate differently from that without endogenous flow (Figure 4(a)). Given q_i and

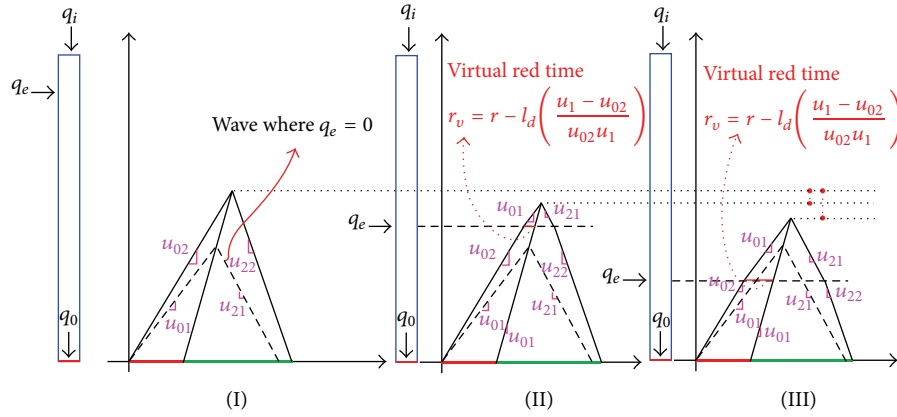


FIGURE 20: Stopping wave and starting wave with endogenous flow.

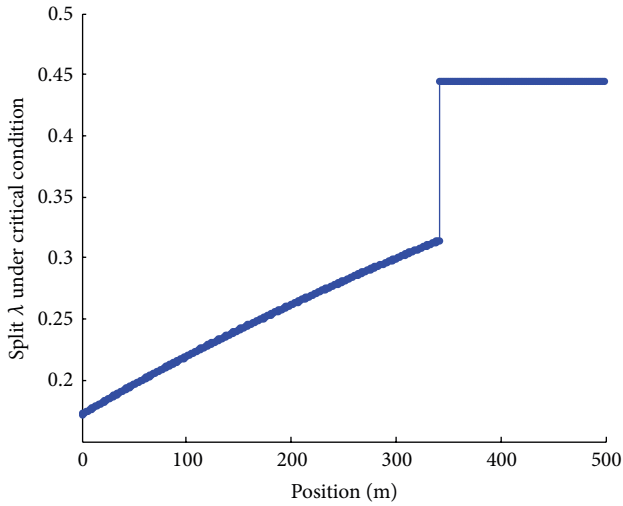


FIGURE 21: Critical split for jam formation versus different entrance position (l_d), $q_i = 500$ veh/h, $q_e = 300$ veh/h.

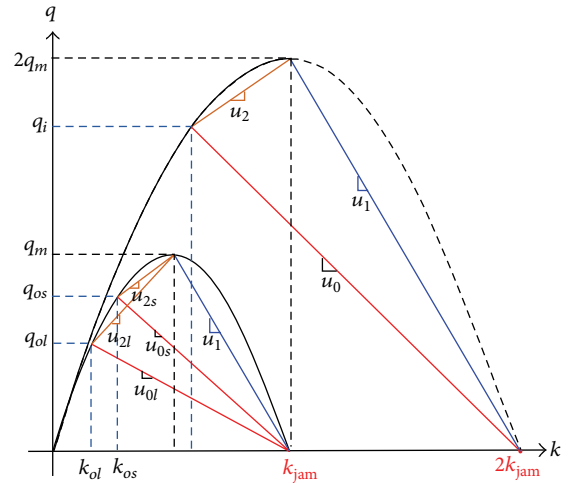


FIGURE 22: q - k relationship of two sections.

When $p_{ol} < p_{os}$ (i.e., $q_{ol} < q_{os}$), upstream section holds less flow averagely than channelized section, so the wave propagation profile should be the pattern as C in Figure 24. Its stopping wave speed u_{0c} is smaller than u_0 and u_{2c} is greater than u_2 . When green time g_{ol} extends over X then the queue will be cleared completely. Critical conditions for domino formation are deduced:

$$\left(\frac{l_2}{u_0} + \frac{l_1}{u_{0c}}\right) + \left(\frac{l_1}{u_{2c}} + \frac{l_2}{u_2}\right) = r_{ol} + g_{ol}. \quad (18)$$

From geometry and Figure 24, we can deduce virtual red time for interface when critical condition (i.e., (17) and (18)) is not reached:

$$r_v = r_{ol} + \frac{l_2}{u_1} - \frac{l_2}{u_0} = r_{ol} - l_2 \left(\frac{u_1 - u_0}{u_0 u_1}\right), \quad (19)$$

where r_v is virtual red for interface. From it split can also be determined: $\lambda_v = (C - r_v)/C$. Since $(u_1 - u_0)/u_0 u_1 > 0$, from

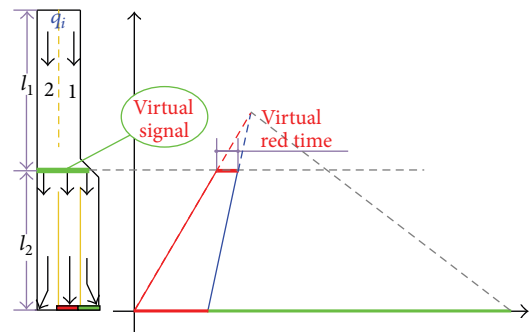


FIGURE 23: Virtual signal and virtual red time.

(19) we can see that virtual red time increases against channelization section length, and opposite relationship holds for split and channelization section length.

Now critical condition of (4a), (4b), and (4c) can be used to determine supply-demand of upstream section. Note that there is no channelization spill back for through traffic flow. Since virtual split of interface is bigger than $(C - r_{ol})/C$ (from

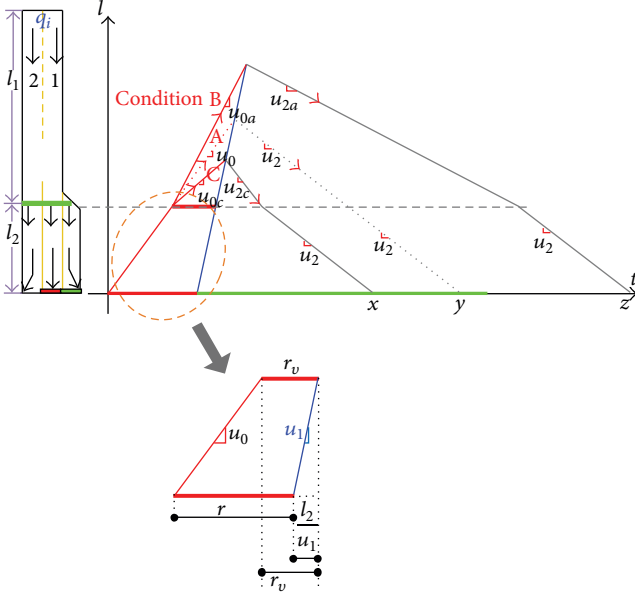


FIGURE 24: Wave profile when channelized section spillover exists.

(19) relationship that $r_v < r_{ol}$ is evident and also $\lambda_v < \lambda_{ol}$, for condition A and C in Figure 24, if critical condition of (18) has not reached, demand of upstream section is always satisfied because

$$\frac{q_{ol} + q_{os}}{\lambda_v q_m + \lambda_v q_m} < \frac{2q_{ol}}{2\lambda_v q_m} = \frac{q_{ol}}{\lambda_v q_m} < \frac{q_{ol}}{\lambda_{ol} q_m}. \quad (20)$$

For condition B, if the critical relationship of (4a), (4b), and (4c) is not hold, traffic jam will form and propagate upstream.

If channelization spillover of only one direction happens, capacity of the other two directions may be wasted, which imply the importance of well-designed phase sequences. From (4a), (4b), and (4c), it implies that the equation below should hold to prevent wasted capacity:

$$l_{\max} = u_1 t' = \frac{u_1 u_0 r_{ol}}{u_1 - u_0}, \quad (21a)$$

$$g_{ol} > \frac{l_{\max}}{u_1} + \frac{l_{\max}}{u_2}, \quad (21b)$$

$$l_{\max} < l_2. \quad (21c)$$

Equations (21a) and (21b) assure that the demand can be satisfied and (21c) guarantees queue back would not reach interface such that block cannot happen.

3.3.2. Two Section Spillovers. Sometimes during peak hour left-turn demand and through demand may both exceed respective capacity which results in spillover of both sections. In such case the problem may be complicated but can be explained conveniently. Spillover of two sections is just as we set two signals in the interface at the same time. When both virtual signals show green then overall virtual signal is

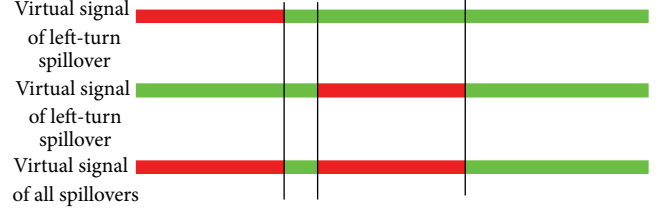


FIGURE 25: Virtual signal of both spillover.

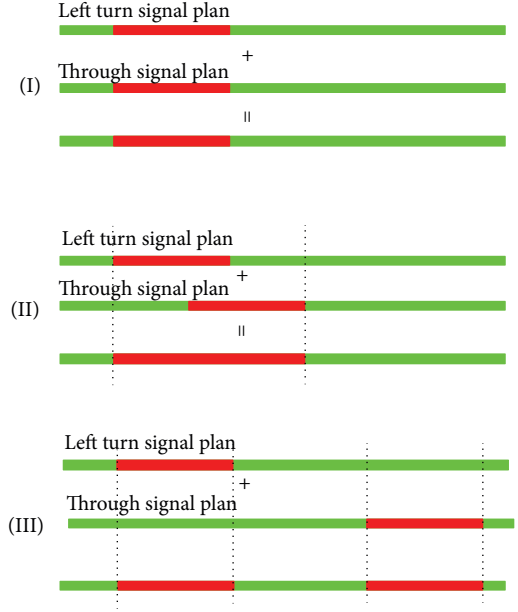


FIGURE 26: Virtual signal for three cases.

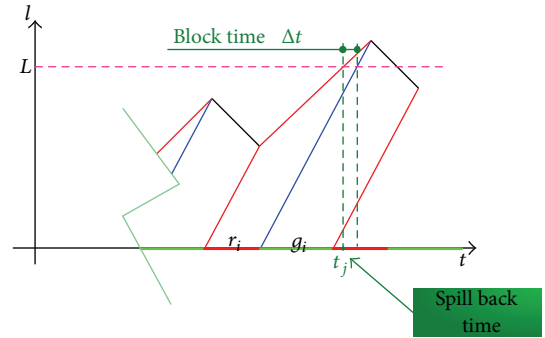


FIGURE 27: Spillover time and block time.

green otherwise the overall signal is red. One case is shown in Figure 25.

According to the signal assignment of left-turn flow and through flow, there will be three cases as shown in Figure 26. Apparently for the same split, different “offset” will produce different spillover virtual signal which implies the importance of phase sequence under spillover condition. We can see that during congestion hours (where channelized section spill back take place); it would be better to put through flow and left-turn flow in the same phase.

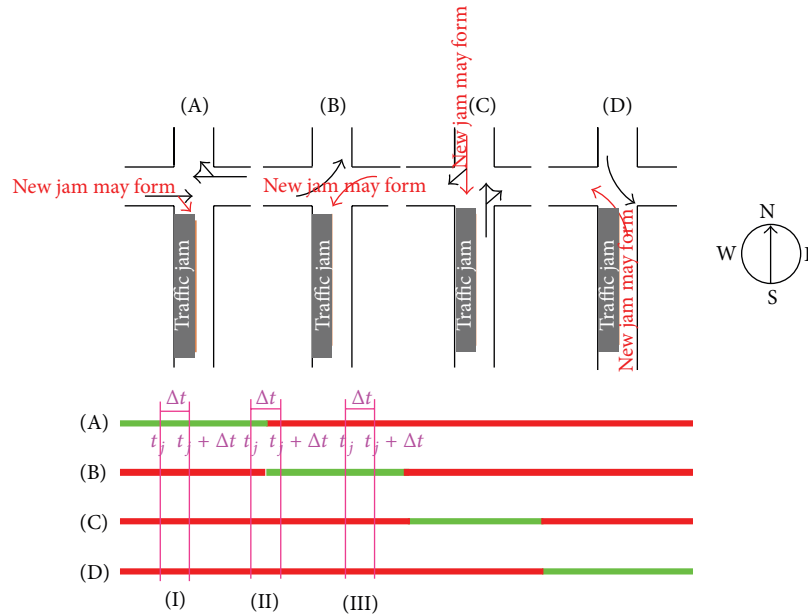


FIGURE 28: Block of upstream intersection.

4. Impact of Spillover on Upstream Intersection

Spillover is the precondition of congestion propagation. However, spillover itself is not enough for domino phenomena formation because if there is not enough upstream demand congestion would not spread. Thus the sufficient and necessary condition for domino phenomena is first, spillover of downstream road and second, there is continuous demand upstream.

When a queue forms and spill back and blocks upstream intersection, there are basically two events to be analyzed: first, how long it will block upstream intersection and second, its influence on upstream flows. When queue back reaches upstream intersection at time, say, t_j (as shown in Figure 27) then it blocks the intersection till the starting wave catches up with it.

We denote that the time duration of the intersection is blocked (i.e., blocked time) as Δt , then from the viewpoint of upstream signal phases there are three types (using a typical four-phase intersection shown in Figure 28 as an example) of block. Take phase A to describe as following.

- (I) Block fully happens at green time. Then the phase will lose Δt effective green time which will decrease capacity respectively.
- (II) Block fully happens at red time. Since no matter whether the intersection is blocked, this movement always has not got RoW (Right of Way), so this type of block holds no impact.
- (III) Block happens partially at green time and partially at red time (let us neglect amber time because it will not

harm the conclusion). This block will encroach some green time which will reduce some capacity.

Due to reduced capacity caused by downstream block, specific phase upstream will be under oversaturation temporarily and long queue forms which may be block upstream intersection of this crossroad.

The block influences upstream link flow differently. In Figure 28 the most possible flow blocked by downstream jam is highlighted in red line for each phase. Green time for these flows then may be wasted hence new jam may form at upstream links. This can be seen as domino phenomena of traffic jam in the whole network.

From the description above we can see that badly designed signal may contribute to network jam propagation, which suggests the importance of network signal coordination during peak hours.

5. Conclusion

Due to the complexity of urban traffic jam, many researches were taken under simplified assumption or used simulated methods. In this paper, urban traffic jam formation is studied. From the research we can see that many factors (such as road geometry, flow structure, and signal settings) contribute to the traffic jam formation and propagation which should be took into account in the control of jam during peak hours.

However, limited by the macroscopic method, we cannot look into deeply mutual influence of different flow, (left-turn flow, right-turn flow, and through flow) which may contribute to the formation and propagation of traffic jam. This may be solved by using microscopic analysis methods. Furthermore, formation of congestion is only prelude of

large scale urban traffic jam. Propagation of urban traffic jam (denominated as “domino phenomena” here) itself is influenced by many factors. Some further researches could be highlighted.

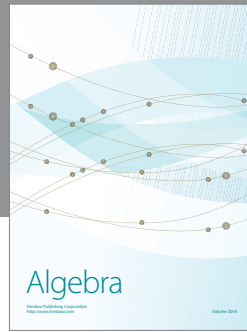
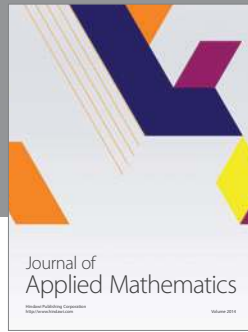
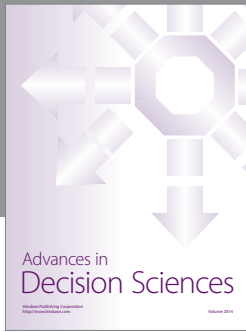
- (1) The calibration of fundamental diagram: the relationship between flow, density, and speed is a basic problem of traffic engineering field and exerts great influence on jam propagation (severity and propagation speed). A thorough examination of the influence is needed.
- (2) The assumption of uniformly mixed flow at upstream section should be revised. In reality, the distribution may be not averagely. Whether this feature prompts the propagation of traffic jam can be included in our model with a coefficient that needs to be explored closely.
- (3) Isolated intersection signal design at peak hours. From this research we can see that traditional control principle of minimizing delay may be ineffective in dealing with peak hour flow due to the spillover of long queue. Till now the control strategy that takes this condition into account is absent.

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