Article

A System Dynamics Approach to Integrating Sustainable Land Use with Transportation System Based on Access Management

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Abstract: The coordination planning between land use and transportation system is an important premise of solving urban transportation problems and realizing land use integration. This study investigates the interactive and feedback relationship between land use and transportation system from the perspective of access management. By integrating the land use and traffic data from Las Vegas Metropolitan area with the system dynamics model, the causal relationship and causal loop diagrams (CLDs) are introduced to analyze the cause-and-effect relationship and quantitative relationship between the factors of the combined system of land use and transportation, and then sub-models partition and system simulation are performed. The systems dynamics model is established by analyzing the relationship between a series of access management techniques, traffic characteristics, and land use features. The results show that system dynamics model can be used as an effective alternative to model the symbiosis relationship of land use and transportation system for urban planning and construction.

Keywords: System Dynamics; Land Use; Transportation Systems; Access Management

1. Introduction

It is well known that urban roadway resources are limited, and it takes enormous capital and time for construction and maintenance. Moreover, the roadway growth speed is far less than that of the travel demand. Therefore, as for the limited roadway resources, by managing the accesses the roadway lifecycle can be extended, and the safety can be increased, while the traffic congestion can be reduced, and the appearance and quality of the environment can be improved. More importantly, access management not only ameliorates the basic functions of roadways, but also preserves the long-term property value and economic viability of adjacent land [1].

Access management (AM) is to control the location, spacing design and operation of driveways, median treatments, median openings, and related auxiliary lanes systematically. The goal of access management is to perform the roadway functional hierarchy in the transportation planning, which provides the accessibility to the adjacent land as well as the safety and efficiency of the transportation system, particularly along arterials and other primary roads [1]. Consequently, a comprehensive access management program is required to support the safe and efficient operation of the main arterials, while providing convenient accesses to the surrounding commercial sites.

On the contrary, without access management, the function and operation of major corridors may deteriorate in a few years, especially for the newly-built roadways. In recent years, this phenomenon has frequently occurred in China. Inadequate access management can not only cause high costs for government and related agencies, but also have a negative impact on the corridor business. With the rapid growth of travel demand, the access management problem is now being considered seriously, as the closely-spaced and poorly-designed major corridors make it difficult for
customers to enter and exit the commercial areas safely and efficiently (the commercial sites at the intersection corner may even be blocked by the queuing traffic). As a result, the newly opened businesses may seek more attractive locations where access is more convenient, safer, and less congested. More customers shall stream into these commercial sites with better accesses. And as the situation keeps going on, the previously developed areas may start to deteriorate due to the inadequate access management, and more future businesses will turn to the better-spaced and better-designed corridors.

These symptoms are caused by inadequate coordination of transportation and land use decisions, and the results may lead to a series of disastrous events that serve as a cycle of functional obsolescence of the major corridors. In the beginning, the accessibility of adjacent land is ameliorated with significant improvements in the corridor, while the improved accessibility increases the value of land and stimulates the real estate development. With the time moving on, the traffic around the land increases and traffic conflicts also increase, leading to deterioration of traffic conditions. As a result, the corridors may need to be improved again to restore the safety and capacity of the system, and repeatedly this loop will continue. As a systematic control method, system dynamics can reflect interaction and feedback features of this loop, and address the accessibility and sustainability of along the major corridor [2]. Furthermore, this loop can be managed effectively by considering both the adjacent land and transport system based on access management program.

Although land use and transportation interaction has been studied over the past 60 years, few has investigated using system dynamics from the perspective of access management. Therefore, the purpose of this study is to propose a system dynamics model that integrates both land use and road transport from the perspective of access management because the interaction and feedback between land use and transportation is consistent with the characteristics of system dynamics, and more importantly system dynamics can deal with qualitative and quantitative features systematically. The model is established with the causality functions and causal loop diagrams (CLDs) by adopting a series of AM techniques, traffic characteristics, land use features and related variables that are considered as model components. The results are beneficial to transportation planning, design and maintenance of roadway system, as well as to local government and agencies to develop policies, regulations, and improvement programs.

2. Literature Review

In transportation areas, so far there have been a variety of models, methods and software focused on the interaction between land use and transportation system. Since the study is based on system dynamics, two major types of literature are summarized, namely, general land use/transportation interaction models and system dynamics approach.

General Land Use/Transportation Models

Over the last 60 years, land use and transportation interaction has been extensively studied [3-6]. Generally, there are four types of land use and transportation models, including econometrics models, heuristic models, simulation models and scenario techniques [7].

As for the econometric models, Vu et al. (2006) [8] explored the effects of access management on accessibility and patronage. Safety and congestion impacts were concentrated along six major commercial corridors in Western Washington State with simultaneous logit model. The findings show that business type and operational variables, corridor and street environment variables, as well as willingness-to-pay measures are significant factors affecting commercial corridors. Xu et al. (2013) [9] identified access management factors associated with safety of urban arterials mid-blocks with panel data simultaneous equation models. The interaction between land use and transport was addressed from the perspective of access management, which is the base of this study. Generally speaking, most of the studies presented econometric models oriented on accessibility and travel efficiency, but due to the assumptions, complicated modeling and harsh terms, it is difficult to put the models into practice and realize the real value.

From the perspective of heuristic algorithms, Srinivasan (2005) [10] linked land use and transportation in a spatial context in Delhi, India, cellular automation and Markov models were
used to understand land use change at the regional scale and discrete choice models to predict change at the local level. The results showed that land use in the Delhi metropolitan area is intensifying and losing diversity, while linking land use and transportation models is not from real world, but from data-based planning of policy framework. As for the equity based on land use and transportation problem, Lee et al. (2006) [11] proposed bi-level programming model, in which the upper level maximized traffic production incorporating equity constraints, while the lower level combined trip distribution with assignment user equilibrium, and penalty function method with a Genetic Algorithm (GA) based approach was employed to solve the proposed model. The results offered a tool in modeling equity-based transportation and land use development. The study by Li et al. (2014) [12] developed an integrated model of a sustainable land use and transportation system with uncertainty in future population. A two-stage robust optimization model was proposed, in which the first stage was to optimize the land use and transportation system and the second stage was a scenario-based stochastic location and route choice equilibrium problem. The findings showed that the integrated model of sustainable land use and transportation system depends on the level of uncertainty in future population, capital budget for urban development, and confidence levels of the chance constraints significantly. However, Dutta et al. (2014) [13] conducted the calibration of an integrated land use and transportation model using maximum-likelihood estimation (MLE). The performance of the MLE optimization problem was displayed better compared to the Grenoble land-use model. Generally, heuristic algorithms have been widely utilized in this area, but the process is relatively complicated, which may influence the application of the algorithms.

From the aspects of simulation packages, MEPLAN [4, 14], UrbanSim [15], PECAS [16], ILUTE [17], and Uplan and Emme [6, 18] have widely used transportation and land use models based on econometric modeling approach [19]. Clay and Johnston (2006) [14] attempted to analyze the impact of uncertain model inputs and parameters on the final-year model output with MEPLAN, one of the most theoretically consistent and practical integrated land use and transportation model, in Sacramento, California region. The results presented that the MEPLAN model shows a positive characteristics on model outputs, compared with UrbanSim and DRAM/EMPAL packages. Differently, Waddell et al. (2007) [15] incorporated land use in the transportation planning of the Greater Wasatch Front area, Utah by adopting UrbanSim simulation model system. The combined land use and travel model system were described and the results of the sensitivity analysis suggested that the key transportation evaluation measures, such as vehicle miles traveled, vehicle hours traveled, and hours of congestion delay, play an important role in accounting for the land use effects of a regional transportation plan. Later, Waddell (2011) [20] explored the challenges to integrated land use and transportation modeling and planning. UrbanSim was used to draw out lessons from more than a decade of experiences in attempting to bridge the gaps between academic research and the messy complications of applying models in the real world. Moreover, for small to medium-sized cities, Clay (2010) [16] developed an integrated land-use and transportation model in Montgomery, Alabama. The PECAS was chosen to simulate a number of policy alternatives and model economic changes. The conclusions were made that PECAS can be successfully adapted to the needs of a small region. The study by Su et al. (2014) [18] developed a three-stage-two-feedback method for both land use allocation and the transportation policy options in China. Three land use allocation strategies and three transportation policy options were evaluated with UPlan and Emme. The results showed that a proper allocation of land use can have significant impacts in the transportation system with a reduction of VMT (vehicles-miles-travelled) and a better coverage of transit systems. Besides those packages, Farooq and Miller (2012) [17] presented a dynamic disequilibrium based micro-simulation framework towards integrated land use and transportation. The framework was employed to operationalize the housing market by using Integrated Land Use Transportation and Environment (ILUTE) modeling system. The heterogeneity was captured within simulation results because of types of dwellings and different market conditions. All the software packages above may have certain assumptions for land use and transport, and are only suitable for certain conditions.
Scenario techniques have been widely used in land use and transportation planning, which provides various constructive dialogues about the future of the communities, and serve as the basis for assertive action to direct the future planning [21]. From the developers’ perceptions of the market for pedestrian-and transit-oriented development, Levine and Inam (2004) [22] discussed the market for transportation-land use integration through a national survey. The survey indicated that respondents prefer land-policy reform based on the expansion of choice in transportation and land use. Bartholomew (2007) [21] summarized 80 scenario planning projects from more than 50 U.S. metropolitan areas, and public participation, methodology and institutional structures were discussed. The study concluded that the institutional structures behind regional land use-transportation scenario planning needs to be bolstered. Meanwhile, Nijkamp et al. (2007) [23] offered a methodological contribution to sustainable mobility policy in the Netherlands. Scenario analysis was used to design combined land-use/transportation plans, and long-range choice options were evaluated using a hierarchical Regime method. The results are critical to monitor and control in a quantitative way the effects of the policy scenarios proposed for the sustainability objective. Furthermore, in order to analyze the land-use-transportation future, Lemp et al. (2008) [24] compared visioning and modeling methods in Austin metropolitan area as a case study. The results verified that visioning offers the greatest potential for public involvement in identifying regional development goals while the land-use models provide a strong theoretical foundation and allow for interactions with transportation model, and the combination of the two approaches would be beneficial for future planning. The current studies tend to take the qualitative direction, but in practice both quantitative and qualitative studies are required.

As far as concerned, there have been some studies on the accessibility consequences of land use and transportation interaction. Bertolini et al. (2005) [25] employed the concept of accessibility as a framework for the interactive design of integrated transport and land use plans in Netherlands. Similar study by Geurs et al. (2010) [26] examined the accessibility benefits associated with land use/transport policy strategies for Netherlands. The land use and transport interaction model TIGRIS XL was used to measure accessibility, which accounted for both changes in generalized transport costs and destination utility. Then Coppola and Papa (2013) [27] proposed accessibility planning tool for land use/transport interaction based on a zone-specific indicator, marginal activity access cost, and the application in Rome validated the proposed tool for zones clusters identified by accessibility to public transport.

As a summary, some scholars summarized the transportation and land use models from different perspectives. Badoe and Miller (2000) [28] summarized the literature on empirical studies of transportation–land-use interaction in North America. From macro and micro aspects the studies are reviewed and found that the data limitations and methodological weaknesses are the reason why various studies have different impacts on land-use policies. The conclusions recommend providing an integrated transportation-land-use model and taking into account all the major factors and the interactions. From the operation aspect of land use and transportation, Hunt et al. (2005) [4] provided a detailed review of six frameworks (ITLUP, MEPLAN, TRANUS, MUSSA, NTYMTCLUM and UrbanSim). Iacono et al. (2008) [19] reviewed some of the four important theoretical frameworks for modeling transportation and land use. Aggregate models of spatial interaction, econometric models, simulation models, activity-based travel model, multi-agent models and cell-based models were discussed and criticized. The first two “top-down” “frameworks specified the interaction between transportation networks and location based on the behavior of representative individuals, while the last two “bottom-up” frameworks offered an alternative mechanism for representing the dynamics of land use change. Recently, Acheampong and Silva (2015) [6] provided comprehensive and up-to-date material for both researchers and practitioners interested in land use and transportation interaction modeling. All of the reviews give the direction of our study.

System Dynamics Approach

The System Dynamics (SD) is an approach to understanding the nonlinear behavior of complex system over time by reflecting both qualitative and quantitative features, which is an alternative
and systematic approach for land use and transportation problem, and has been widely used all over the world in various fields [2, 7, 29, 30, 31, 32] since presented by Prof. Forrester in 1958.

From the aspect of engineering, Ahmad et al. (2016) [29] summarized the application of SD in electricity sector modeling, including policy assessment, generation capacity expansion, financial instruments, demand side management, mixing methods and finally micro-worlds models. From the aspect of management, Cosenz and Noto (2016) [30] explored and discussed the theoretical and empirical development of SD in the strategic management so as to highlight the main research themes and developmental patterns of the key contributions; From the perspective of environment, public policy and resources, Brent et al., (2017) [31] examined the extent of utilization of SD in South Africa between 2003 and 2014, which is useful to facilitate further discussions on practical applications of SD. The latest study by Currie et al. (2018) [32] applied SD modeling to environmental health decision-making and policy, and revealed that SD modeling was increasingly used but still limited.

In transportation field, Abbs and Bell (1994) [29] made the first attempt to apply system dynamics approach for land use and transportation study and demonstrated that it can be a good alternative. From methodology and application, Haghani et al. (2003) [7, 30] presented a system dynamics approach to land use/transportation system performance modeling simultaneously. Later study by Shen et al. (2009) [31] applied a system dynamics model for the sustainable land use and urban development in Hong Kong. Transport sector and urban land sector of SD models were established as two sub-systems, and simulated separately. Although the results support compact and high-density development in Hong Kong, the interaction between land use and transportation is the weakness. Pointed to this interaction, Pfaffenbichler et al. (2010) [32] developed a system dynamics model for land use/transport problem with Metropolitan Activity Relocation Simulator (MARS) in VENSIM (one simulation software developed by Ventana Systems, Inc. used for SD), which is capable of analyzing policy combinations at the city/regional level and evaluating their impacts over a 30-year planning period. Shepherd (2014) [2] reviewed six aspects of SD models applied in transportation, including the uptake of alternate fuel vehicles, supply chain management, highway maintenance/construction, strategic policy at urban, regional and national levels, airlines and airports, and the emerging areas. The results verify that SD is better suited to providing a holistic system model dealing with feedbacks and delays in the system.

Summarized from the studies above, most concentrate on either land use/transport, or system dynamics separately or individually, and there have been very limited studies integrating land use and transportation with access management together. Moreover, the current methods are concentrated either on quantitative or qualitative aspect. Thus, this study will fill the gap by integrating land use/transport with access management, while addressing the interaction and feedback with system dynamics from both quantitative and qualitative aspects.

3. System Dynamics Methodology

3.1. Fundamental rules

Compared with other alternative techniques, SD can reflect the interaction and feedback between land use and transportation based on access management, as well as addressing qualitative and quantitative features systematically. The qualitative feature is from the causal loop diagrams (CLDs), which can be defined by the feedback loops, influencing factors and cause-and-effect relationship, while with the CLDs the quantitative attribution could be converted from the abstract social and economical phenomena to concrete technical one [2], and then the dynamic behavior model of the main system can be established.

Usually a CLD consists of the components and connected arrows, in which the former represents the unspecified quantity affecting other components and themselves affected by other components, while the unspecified quantity should be able to reflect the components’ increasing or decreasing trend; the latter denotes the causal interactions among variables, symbolized by causal links with arrows. Each causal link indicates the specific same or opposite relation, reflecting the
direction that the dependent component varies with the independent one. If the dependent component varies with the independent one at the same direction, the causal link is denoted as “+”, otherwise denoted as “−” at the opposite direction. How to illustrate the method of CLD can be exemplified in Figure 1.

Figure 1 Feedback Loop Structure

In Figure 1, assume two variables, travel demand and traffic conflict, and the feedback loop formed. If the travel demand is increased, the traffic conflicts are increased, or if the traffic conflicts are increased, the travel demand is increased, thus both links are expressed as “+”. Similarly, assume two variables, travel demand and congestion, and the feedback loop formed. An increase in travel demand may increase the congestion chances, and a decrease in travel demand implies the congestion is reduced, so the link is denoted as “+”; contrarily, an increase in congestion may be due to the increase of travel demand, while a decrease in congestion may attribute to the increase of travel demand, thus the link is expressed as “−”.

Moreover, Figure 1 indicates that the two components, traffic conflict and congestion not only affect the traffic travel demand, but are affected by the travel demand. Hence, the travel demand includes two feedback loops: the reinforcing loop (R) and the balancing loop (B). The R loop reflects the continuing trend of growth or decline, which makes the system apart from an equilibrium point, while the B loop features causal relationships, which tends to decrease the discrepancy between the current state and an expected state. More details about fundamental rules of SD can be referred to [2, 7, 31,32].

3.2. Modeling procedure

With the CLDs above, the qualitative features of SD can be revealed from the structure of a system and dynamic behavior, meanwhile the quantitative principles can be reflected with differential variables and equations varying over time, and can be addressed via a stock-flow diagram, which makes the model visualized and understood easily.

Based on the CLDs, the stock-flow diagram can be transferred to provide accumulation and/or evacuation of stock and flow for quantities in the system with the following steps:

Step 1: Key variable identification: main driven force should be identified according to the problem description;
Step 2: Stock identification: the system resources (i.e. stocks) related to the main driven force should be determined;
Step 3: Flow module development: the rates of change should be provided, and the processes dominating each stock should be presented;
Step 4: Qualitative analysis: additional driven force overlooked and other controllable variables should be identified, and causal relationships requiring further analysis with specific approaches should be determined.

Through the steps, the system can be represented in terms of stocks and flows, and the quantification of the process can be represented and computed with the equations or algorithms.
Figure 2 Stock and flow example

Shown from Figure 2, the stock-flow diagram of SD can be represented by 3 major types of variables, stock level variables, rate variables and auxiliary variables. Stock level variables can represent any variable that accumulates or integrates over consecutive time periods, while rate variables denote activities or variables making the stock change. The auxiliary variables are described as identifying rate variables or other auxiliary ones. Each variable is expressed by an equation that is concerned with its level, rate and auxiliary equations. The stock level equation is denoted as a differential equation, while the rest are general algebraic equations. As exemplified by Figure 2, the inflow to a stock represents the traffic flow into certain land, while an outflow from a stock stands for the traffic flow out of the land, and both the inflow rate and outflow rate are controlled by the land types. Thus, the stock level can be described by the following equation:

\[ \int_{t_1}^{t_2} [\text{Inflow}(s) - \text{Outflow}(s)] ds + \text{Stock}(t_1) \]  

where the \( t_1 \) and \( t_2 \) represent starting time and ending time moments respectively, inflow (s) and outflow (s) are the arrival flow and departure flow of the land between the initial time \( t_1 \) and the current time \( t_2 \), correspondingly, and stock \( \text{Stock}(t_1) \) stands for the initial traffic volume at time \( t_1 \). The net rate of the stock can be expressed by the differential equation:

\[ \frac{d(\text{Stock})}{dt} = \text{Inflow}(t) - \text{Outflow}(t) \]  

(2)

The auxiliary variable, such as congestion load factor, is a function of stock, which can be used to formulate and calibrate the model. Stock level variables, rate variables and auxiliary variables, connected by arrows, are formed into the stock-flow diagram, which can be utilized to establish the cause-and-effect relationships between the model variables by transferring the CLDs information into the differential equations. With these equations, the simulation can be performed with software program VENSIM and represent the dynamic behavior of the system variation.

4. Modeling for Land Use and Transportation

The main structural interrelationships are shown in Figure 3 with four sub-models, and the positive and negative feedback loops can be identified.

**Accessibility-Land use degree-Travel demand-Congestion Feedback Loop (positive)**

Improved accessibility can make the land use degree (residential or commercial lands by the total lands) reach the higher level, and then attract more travel demand, and finally cause more congestion along major corridors. The increase of accessibility encourages more land use development: on the one hand, the implementation of access management techniques improves the safety and mobility around the adjacent lands; on the other hand, improvement of accesses enhances the commercial and residential lands along the corridors’ safety and efficiency [8]. In this way, the land with more convenient and safer accesses would generate more trips, thus the travel demand would be increased. Although the improvement of accessibility and travel demand has increased the roadway capacity significantly, the growth can’t be infinite, and when the traffic volume reaches the capacity boundary, the roadway mobility would be deteriorated till the congestion occurs. Consequently, the accessibility needs to be improved again to restore the safety and mobility of the system.

Four Sub-models:

**Accessibility-AM techniques Feedback Loop (positive)**

The implementation of AM techniques can improve the roadway accessibility clearly, e.g. appropriate signal spacing, driveway spacing, median alternatives, median openings, access circulation etc., which not only improve the traffic flow organization smoothly, but make safe and efficient access points. On the contrary, the increase of accessibility brings the application of AM techniques [1].
Figure 3 General Model Structure

*Land Use Degree-Land Value Feedback Loop (positive)*

The larger the land use degree is, the higher the land value is. The more developed land encourages more patronages to invest more, whether the commercial or residential lands are, thus the land value is increased significantly. In the same way, the higher land value promotes the land use to be developed more efficiently.

*Travel Demand-Traffic Conflict Feedback Loop (negative)*

The increased traffic generation brings an increase of traffic conflicts. However, as traffic conflicts progress, the trips generated are reduced, thus the travel demand is decreased.

*Congestion-Roadway Supply Feedback Loop (negative)*

The severe traffic congestion can make the roadway capacity reach its peak, which may bring about the development of new roadways (positive effect). On the other hand, increase of roadway capacity by broadening or expansion of lane-miles alleviates the traffic congestion (negative effect) at least as a whole.

According to the feedback loops, the CLDs of general model and four sub-models can be integrated and formed into the following diagram, as shown in Figure 4.

In Figure 4, on the one hand, accessibility, land use degree and travel demand are considered as the influencing factors of traffic congestion in the general model, i.e., accessibility, land use degree and travel demand are regarded as input in the system model while traffic congestion is treated as output. On the other hand, in the sub-models, accessibility, land use degree and travel demand are considered as dependent variables influenced by sub-factors. For instance, in the travel demand sub-model, travel demand is influenced by traffic conflicts, and considered as the dependent variable, while traffic conflicts are influenced by AADT, number of accidents, road length and so on, which can be considered as independent variables, thus forming into travel demand sub-model loop.
Variables and Parameters

The major variables, parameters, and auxiliary variables in this model system are described as follows.

Level Variables

According to the level equation, the level variables are defined as the quantity accumulated during certain time periods, and in this study two major level variables are considered, land use degree and congestion, since the main purpose is to find out the cause-and-effect relation between land use and transportation.

Rate Variables and Auxiliary Variables

In the level variables equation, the rate variables represent the varying parameter of inflow and outflow, and an auxiliary variable denotes the quantity during a certain period, which can help explain the various factors associated with the model components.

Model Equations

By integrating major variables, parameters and auxiliary variables with SD principles, software VENSIM not only provides the qualitative relations among all the variables as Figure 4, but offers the quantitative equations as follows. The equations are obtained by several trials with the dataset till the optimal ones are achieved, with which the quantitative features of SD can be reflected appropriately.

Accessibility Sub-model

In this sub-model there are several auxiliary variables developed by the following equations:

\[ AY = \alpha_0 + \alpha_1 \times MD + \alpha_2 \times 1.5SDY + \alpha_3 \times DD \]  

where \( AY \) = Accessibility; \( MD \) = Median-opening density, number of median openings per mile; \( SDY \) = Signal density, number of signals per mile; \( DD \) = Driveway density, number of driveways per mile.

Equation 3 states that accessibility (AY) can be defined as a function of median density (MD), signal density (SDY), and driveway density (DD). \( \alpha_0 \) denotes the intercept, and \( \alpha_1, \alpha_2, \alpha_3 \) stand for the coefficients.

Land Use Sub-model

Based on the causal loop diagram in Figure 4, level variables \( LUD \) is defined by the following equation:

\[ LUD_K = LUD_L + \int_{L}^{K} LVdt \]  

Equation 4 states that the number of present land use degree (\( LUD_K \)) is the land use degree in previous time period (\( LUD_L \)) plus the net change during one time period caused by land value. \( K \) and
L. represent the certain time. Rate variable LV (land value) is deduced from the following algebraic equations:

\[
LV = (\beta_0 - \beta_1 \times \cos((RL) + 1) + \beta_2 \times CL - \beta_3 \times \ln(TL) + 0.5) \times \ln(Time + 2.7183)
\]  
\[\text{(5)}\]

where: RL=Residential Land Use Degree, proportion of residential land to total land; CL=Commercial Land Use Degree, proportion of commercial land to total land; TL=Total Land Use Degree (Residential Land plus Commercial Land).

\[\beta_0 \text{ is for the intercept, } \beta_1, \beta_2, \beta_3 \text{ represent the coefficients.}\]

**Travel Demand Sub-model**

In this part there are two major variables: Traffic conflict (TC) and Traffic demand (TD). Each of them can be developed from VENSIM as follows:

\[
TC = (NA \times 10^3/(AADT \times 365 \times LS)) \times \ln(Time + 2.7183)
\]  
\[\text{(6)}\]

\[
TD = \delta - \alpha \times TC
\]  
\[\text{(7)}\]

where: TC=traffic conflict; NA=number of accidents; AADT=annual average daily traffic; LS=length of street; TD=traffic demand; \(\delta\) is the intercept, and \(\alpha\) is the coefficient.

**Congestion Sub-model**

From the causal loop diagram in Figure 4, the level variable Congestion (CN) can be defined as the following equation:

\[
TCN_k = TCN_{L} + \int_{t_{0}}^{t_{k}} CRdt
\]  
\[\text{(8)}\]

It states that the number of present traffic congestion \((TCN_k)\) is the traffic congestion in previous time period \((TCN_L)\) plus the net change during one time period caused by congestion rate. Rate variable \(CR\) (congestion rate) is deduced from the following algebraic equations:

\[
CR = \gamma_0 + \gamma_1 \times TD + \gamma_2 \times \ln(AY/(LUD + 1)^{0.8}) + TCN
\]  
\[\text{(9)}\]

where: TCN=traffic conflict; CR=congestion rate; TD=traffic demand; LUD=land use degree; \(\gamma_0\) is the intercept, \(\gamma_1, \gamma_2\) are the coefficients.

**5. Parameter Estimation**

According to the SD principle, the coefficients of models for the accessibility equation, land use equation, traffic demand equation, and congestions equation are assessed through Ordinary Least Square (OLS) estimation with the corresponding data. The data were collected from 2003 to 2005 in Las Vegas Metropolitan area maintained by the Nevada Department of Transportation, and 356 roadway segments were included from 19 major and minor arterials. More details about the dataset can be referred to [9, 33].

**Estimation of the Coefficients for the Accessibility Equations:**

Correlation test was conducted before the final model was obtained. The variables with high correlation, such as, two-way and total median opening density, only one is kept and the other is neglected. Table 1 gives the estimated coefficient values for all the significant variables at 95% confidence interval. R\(^2\) value and F value indicate that the estimated equation is meaningful for use. It can be seen that median-opening density, signal density and driveway density are significant for accessibility, which is consistent with the results in Xu et al. (2013) [9]. The R\(^2\) value reflects that there is high relation between dependent variable and independent variables.

<table>
<thead>
<tr>
<th>Coefficient</th>
<th>Independent Variable</th>
<th>Accessibility Sub-model</th>
</tr>
</thead>
<tbody>
<tr>
<td>(a_1)</td>
<td>MD</td>
<td>0.926</td>
</tr>
<tr>
<td>(a_2)</td>
<td>SDY</td>
<td>0.034</td>
</tr>
<tr>
<td>(a_3)</td>
<td>DD</td>
<td>1.018</td>
</tr>
<tr>
<td>(a_0)</td>
<td>Constant</td>
<td>2.318</td>
</tr>
<tr>
<td>(R^2)</td>
<td>value</td>
<td>0.996</td>
</tr>
</tbody>
</table>
Estimation of the Coefficients for the Land Use Equations:

Similarly, land use sub-model, travel demand sub-model, and congestion sub-model have been tested various times before the final equations were determined, and coefficients values are listed from Table 2 to Table 4. The results show that all coefficient values are statistically significant at 95% confidence interval. $R^2$ values and F values also indicate that there lies in high relation between dependent variable and independent variables. With all the equations provided, the interaction and feedback relations among the variables in Figure 3 and 4 can be verified, which addresses the quantitative features of SD.

### Table 2. Estimation of Coefficients for the Land Use Equations

<table>
<thead>
<tr>
<th>Coefficient</th>
<th>Independent Variable</th>
<th>Land Use Sub-model</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\beta_1$</td>
<td>RL</td>
<td>-0.044</td>
</tr>
<tr>
<td>$\beta_2$</td>
<td>CL</td>
<td>0.023</td>
</tr>
<tr>
<td>$\beta_3$</td>
<td>TL</td>
<td>-1.172</td>
</tr>
<tr>
<td>$\beta_4$</td>
<td>Constant</td>
<td>0.819</td>
</tr>
<tr>
<td>$R^2$ value</td>
<td></td>
<td>0.858</td>
</tr>
<tr>
<td>F value</td>
<td></td>
<td>786.798</td>
</tr>
</tbody>
</table>

### Table 3. Estimation of Coefficients for the Traffic demand Equations

<table>
<thead>
<tr>
<th>Coefficient</th>
<th>Independent Variable</th>
<th>Travel Demand Sub-model</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha$</td>
<td>TC</td>
<td>0.244</td>
</tr>
<tr>
<td>$\delta$</td>
<td>Constant</td>
<td>48.078</td>
</tr>
<tr>
<td>$R^2$ value</td>
<td></td>
<td>0.844</td>
</tr>
<tr>
<td>F value</td>
<td></td>
<td>19.098</td>
</tr>
</tbody>
</table>

### Table 4. Estimation of Coefficients for the Congestion Equations

<table>
<thead>
<tr>
<th>Coefficient</th>
<th>Independent Variable</th>
<th>Congestion Sub-model</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\gamma_1$</td>
<td>TD</td>
<td>0.065</td>
</tr>
<tr>
<td>$\gamma_2$</td>
<td>AY/LUD</td>
<td>0.015</td>
</tr>
<tr>
<td>$\gamma_0$</td>
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<tr>
<td>F value</td>
<td></td>
<td>1429.678</td>
</tr>
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</table>

6. Simulation Results and Discussion

According to the principle of SD, the model performance can be reflected from the equations and figures, which is accomplished by simulation with software VENSIM. The following figures display the variation trend of variables according to equations listed above. Figure 5 shows the variation trend of accessibility within five years. It can be seen that as time goes on, the accessibility of the region increases year by year, and the increasing rate of the first three years is larger, but gradually tends to be gentle, which is in line with the actual situation. As a matter of fact, in the beginning of the construction, the accessibility of the region is significantly improved due to the introduction of various facilities, such as raised median, driveways, median openings, etc.; however, with the completion of a variety of facilities, accessibility tends to be stable gradually after certain time period.
In this accessibility sub-model, four factors play important roles in determining the accessibility: signal density, number of driveway, road length and median treatments. The sensitivity test for each individual factor is performed while holding other variables constant, as the followings:

1. When the value of signal density is increased by 100%, the value of accessibility is raised by 0.682%;
2. When the value of driveway is increased by 100%, the value of accessibility is increased by 86.091%;
3. When the value of median density is increased by 100%, the value of accessibility is increased by 7.5785%;
4. When the value of road length is increased by 100%, the value of accessibility is decreased by 46.835%;

It can be concluded that the number of driveways has the greatest impact on accessibility, followed by the length of the road (46.835%), which is much greater than the other two variables. The density of traffic signals has little effect on accessibility. In addition, the length of road has a negative impact on accessibility, while the other three are positive. Therefore, it can be suggested that although more driveways can improve accessibility, the number of driveways should be controlled since more driveways may produce more conflicts; longer roadway segments may improve the vehicle mobility instead of accessibility, thus the roadway length should be designed to some appropriate extent, which is consistent with the results by Xu et al. (2013, 2015) [9,33]. The density of traffic signals depends on the roadway length and structure to a certain extent, showing less effect on accessibility, which can be understood.

Similarly, the variation trend of land use over time is performed as shown in Figure 6. Although the land use degree varies year by year, the trend is increasing but much gentler compared to that of accessibility trend. As for this land use sub-model, there are three critical variables influencing land use degree: commercial land, residential land and total land. The sensitivity tests are conducted as follows:

1. When the value of the commercial land is increased by 100%, the value of land is increased by 17.68%;
2. When the value of the residential land is increased by 100%, the value of land is reduced by 13.24%;
3. When the value of the total land increased by 100%, the value of land decreased by 92.5%;

There are two noteworthy aspects of the sensitivity test: one is that total land has the greatest impact on the land use, followed by commercial land; the other is that the variation direction of commercial land is consistent with that of land value, while the other two are negative, indicating that the development of commercial land has significantly contributed to the growth of land value. However, the total land use should be controlled appropriately to keep the commercial-land and residential-land balanced to meet the requirements of urban development.
Figure 6 Land Use Variation Trend

Figure 7 shows the varying trend of travel demand over time. In this sub-model, travel demand is affected by traffic conflicts, while the latter depends on various factors, such as the number of accidents, AADT, length of street, etc. Suppose that AADT is kept unchanged, the travel demand is decreased year by year due to the increase of traffic conflicts in the model. The increase of traffic conflicts may cause the deterioration of the roadway conditions, thus raising people’s travel costs, and finally leading to the travel demand down.

Figure 7 Travel Demand Trend

Consequently, the travel demand is affected by the three factors: the number of accidents, AADT, and length of street. The sensitivity test for each factor is performed as follows:

1. When the number of accidents is increased by 100%, the value of travel demand is decreased by 10.79%;
2. When the value of AADT is increased by 100%, the value of travel demand is raised by 5.4%;
3. When the value of the road length is increased by 100%, the value of travel demand is increased by 5.4%;

As shown in the results of the sensitivity test, the number of accidents has the greatest impact on travel demand, followed by AADT, while being equal to road length; Moreover, the variation direction of road length and AADT is uniform with that of travel demand, while the number of accidents is opposite. The reason that the number of accidents reduces the travel demand is because of more accidents worsening the roadway conditions and deterring the vehicles along the corridors, thus leading to the purpose of reducing the travel demand [9, 33]. Therefore, more measures should be taken to reduce the number of accidents so as to improve the travel demand. On the other side, AADT can be considered as an exposure of travel demand, meaning that more traffic volume would generate more travel demand, whereas AADT along the corridors should be adjusted in order to avoid traffic congestion. Similarly, longer roadway may produce more travel demand, same principle as that the freeway/expressway attracts more traffic, whereas for the urban corridor the segment length should be suited, matching with the signal design, driveway and median treatments.

From Equations (3) to (9), the trend of traffic congestion is displayed in Figure 8. It can be seen that the variation of traffic congestion in the first three years is relatively gentle, but then becomes more and more intense. For this dataset, one unit is considered as the critical value, and when the
value of traffic congestion is greater than one, the road is under a congested state. The greater the value, the more congested. Shown from Figure 8, after 3.5 years, the roadway begins to be in a state of congestion, and as time goes by, the roadway becomes more and more congested.

![Figure 8 Traffic Congestion Trend](image)

Figure 8 Traffic Congestion Trend

In this congestion sub-model, congestion rate is influenced by three factors, namely, travel demand, accessibility and land use degree. The sensitivity test is performed individually as follows:

1. When the value of travel demand is increased by 100%, the value of congestion rate is increased by 110%;
2. When the value of accessibility is increased by 100%, the value of congestion rate is raised by 0.03%;
3. When the value of land use is increased by 100%, the value of congestion rate is decreased by 0.04%;

The sensitivity test shows that travel demand has the greatest impact on traffic congestion, followed by land use, and accessibility. Moreover, the variation direction of travel demand and accessibility is consistent with that of congestion rate, while the land use is opposite. Among the three factors, traffic congestion is contributed to travel demand mostly, implying that travel demand is of utmost and direct source of the congestion, and various counter measures should be taken according to travel demand management. Land use and accessibility, as for the matured corridors, are already standardized, and have no extra alternation, which should be improved as much as possible [1].

7 Conclusions

In this study, the system dynamics model is used to integrate the transportation system and land use in a novel way from the perspective of access management. By combining the real-world data of Las Vegas metropolitan area with the system dynamics model, the causal relationship and CLDs are introduced to analyze the cause-and-effect relationship and quantitative relationship between the factors of the system formed by the land use and transportation system, and then sub-models partition and system simulation are performed. Through the use of a series of AM techniques, traffic characteristics, and land use features, the system dynamics model between land use and transportation system is established.

The findings are summarized as follows:

1. In this study, accessibility is considered as the driven force of system dynamics, and the driveway, median treatments and the density of traffic signals are selected to reflect accessibility comprehensively. According to the accessibility sub-model, the accessibility is the most sensitive factor to the variation of driveway, followed by median treatments and the signal density, which is uniform to the results of Xu et al. (2013) [9]. Therefore, in order to improve the accessibility of the major corridors, an appropriate number of driveways should be built firstly, and then suitable median types and the density of traffic signals should be taken into account on the basis of economic benefits and efficiency.

2. In this study, the capital appreciation of land (especially the commercial land) causes the land use degree to change, resulting in traffic congestion problem more severely, which reflects the impact of land use on urban transportation. Hence, in the process of urban planning, land use
planning and transportation planning should be taken into account synergistically, and the interactive relationship between land use and transportation needs to be emphasized. Subsequently, reasonable guidance should be provided to the traffic flow to avoid the over-concentration, and even traffic congestion.

(3) In this study, it is found that travel demand rely on three main factors, the number of accidents, AADT and street length. The simulation results show that, if AADT is kept relatively constant, the travel demand decreases year by year with the increase of traffic conflicts.

(4) To sum up, accessibility, land use and travel demand are integrated into factors that affect traffic congestion. From the above analysis, travel demand has the greatest impact on traffic congestion, followed by accessibility and land use. Therefore, in order to reduce traffic congestion, on the one hand, proper management and control can be conducted and some effective measures can be adopted to reduce unnecessary travel demand; on the other hand, some access management measures should be implemented to improve the accessibility of the region and regulate the traffic flow effectively. Last but not least, land use layout should be managed during the planning period to avoid the unreasonable distribution of traffic flow.

In the end, since the interaction between land use and transportation system is dynamic and complex, this study only investigates part of the interaction from the perspective of access management, whereas other factors, e.g., socio-economic parameters (GDP, income, population etc.) and the network level of service are neglected due to the limitation of the dataset. Moreover, other limitations, such as lack of spatial representation, lack of representation of travel modes, lack of endogeneity analysis among variables, etc. may need further consideration. In the future work, the calibration/validation of the model could be improved with more parameters and data. Therefore, it is necessary to conduct more in-depth investigation to improve the model such that the performance of the model is approximate to the actual situation.

Author Contributions: Conceptualization, X. Xu and X. Luo; Methodology, X. Xu; Estimation and Simulation, X. Luo; Formal Analysis, X. Xu and X. Luo; Discussion, X. Luo; Data Curation, X. Xu; Writing-Original Draft Preparation, X. Xu; Writing-Review & Editing, X. Xu; Project Administration, X. Xu; Funding-Original Draft Preparation, X. Xu; Writing-Review & Editing, X. Xu; Project Administration, X. Xu; Funding-Original Draft Preparation, X. Xu.

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