

A System Dynamics Model for Simulating the Logistics Demand Dynamics of Metropolitans: A Case Study of Beijing, China

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Abstract:

Purpose: We attempted to propose an approach to simulate the dynamics of Beijing's logistics demand, which can do some help to find out the dynamics path of the needed storage and shipment, put forward with logistics policies and enhance logistics service.

Design/methodology/approach: We present a paper with system dynamics (SD) methodology, which was run by the software of Vensim[®].

Findings: With SD model, causal loop diagram and stock and flow diagram are constructed, as well as some experiments and policy analysis. The research findings revealed that the increase of average shipping capacity for a vehicle will bring a decrease in congestion and CO₂ emission directly and the decrease of the average fuel use for a vehicle can help with the reduction of CO₂ emission directly. Both the two parameters are the indirect causes of logistics demand dynamics in Beijing.

Originality/value: Researches of this paper are aiming at handling logistics demand dynamics of Beijing, problems belonging to the area of complex systems, with SD model, where, to the best of our knowledge, no significant research has been done.

Keywords: logistics demand, system dynamics, Beijing

1. Introduction

The emergency of city is definitely a great progress and worthless treasure of human civilization's history, which is sharply developed in the past 200 years around the world with the rise of vast emerging powers like China, Russia and India. The urbanization level of China reached 53.7% in 2013, which is forecasted to be 56%-58% around 2020 and 70% in 20-30 years. According to the policies put forward in 18th National Congress of CPC and 12th NPC, majority of the areas in China will experience a higher level of urbanization in the future, leading to more big cities and congregate population. Along with the speeding-up development of the urbanization, metropolitans like Beijing are faced with a dilemma. Demands, generated from production and civil life are ascending with city-expansions, which can't be satisfied well with the existing supplies. Pollutions, resource scarcity, booming of population and congestion, etc. are breaking down the living environment of civil people. The dilemmas of metropolitans ask for an efficient and environmental solution, with which materials and products serving for production and civil life can be safely stored and quickly shipped—a better urban logistics system. As people known in research of economic or social systems, the fundamental thing is the study of demand, which is the original intention of this paper to choose such a topic.

We adopted the system dynamics (SD) methodology in this work for its merits in solving complicated systematic problems (Cai, 2008) as a modeling and analysis tool tackling the complicated issues of the simulation of logistics demand amount in Beijing. The next section presents a literature review. Section 3 presents some description of the studying area and SD methodology, while section 4 provides the model description, as well as the causal-loop diagram, feedback diagram and equations. Section 5 exposes experiments based on SD model. Finally, in section 6, conclusions will be presented.

2. Literature Review

System dynamics is a powerful methodology for obtaining insights into problems of dynamic complexity and policy resistance (Georgiadis & Besiou, 2008). J.W. Forrester developed SD methodology in 1961 to model and simulate dynamic management problems of operation and stock in companies (Forrester, 1961). And then, Forrester gave out the structure and principles of SD model in 1968 (Forrester, 1968). The next year in 1969, Forrester introduced SD model to the wider area of social science and summarized the evolution of American cities. From then, SD methodology began to be used in some huge and comprehensive areas to solve the dynamic problems. In the 1970s, Forrester together with the Club of Rome published "World Dynamics" (Forrester JW, 1971)" and "The limits to growth" (Meadows, Meadows, Randers & Behrens, 1972), in which they analyzed the interactions and feedbacks of the five fundamental factors (population, agriculture, natural resource, industrial production and pollution) of global development. Forrester was the first to study the interactions between natural resources,

technology and economic sectors (Georgiadis & Besiou, 2008; Meadows et al., 1972). Researches of SD was booming since 1970s, which is being applied to areas of natural science, social science and engineering, etc. nowadays.

SD methodology is widely used in the researches of sustainability. Jin, Xu and Yang (2009) attempts to incorporate system dynamics (SD) into EF to develop a dynamic EF forecasting framework, and provide a platform to support policy making for urban sustainability improvement. Bockermann, Meyer, Omann and Spangenberg (2005) adopted econometric method and SD method modelling sustainability, both of which were used to test sustainability strategies for their environmental, social and economic impacts. Collins, Neufville, Claro, Oliveira and Pacheco (2013) explored how interactions between physical and political systems in forest fire management impact the effectiveness of different allocations by SD model and conducted a case study of Portugal. Ansari and Seifi (2013) presented a system dynamics model to analyze energy consumption and CO₂ emission in Iranian cement industry under various production and export scenarios.

Forrester (1969) firstly applied SD methodology to simulate the evolution of American cities in his book "Urban Dynamics", which initiated the use of SD to research on some society-related issues. Mavrommati, Bithas and Panayiotidis (2013) propose a dynamics approach for Ecologically Sustainable Development (ESD) in urban coastal systems, which can facilitate decision makers to define paths of development that comply with the principles of ESD. Musango, Brent, Amigun, Pretorius and Muller (2012) assessed the technology sustainability based on the SD approach and analyzed the case of biodiesel developments in South Africa. Egilmez and Tatari (2012) studied US highway system sustainability and tested three potential strategies for policy making by SD methodology. Researches for the society-related issues with SD model have been combined with 3S (RS, GPS, GIS) technologies these days, with which people could develop a dynamic, visible, updating and computer-transforming system (Cai, 2008). Xu and Coors (2012) developed an integrated system (GISSD System) to assess the sustainability of urban residential development with the integration of SD methodology, GIS and 3D visualization.

SD methodology has been also widely used in business management, policy design, strategy making, etc., which has been well known and proven in strategic decision-making (Georgiadis, Vlachos and Iakovou, 2005). Quadrat-Ullah (2013) presented a dynamic simulation model drawing on SD model to analyze the complex electricity system in Canada. Ghazvini and Shukur (2013) discussed the use of SD as an effective approach to understand the impact of aspects and components of e-health system on policy making. Faezipour and Ferreir (2013) applied SD methodology to analyze the social aspect in healthcare systems and explored important factors and relationships related to patient satisfaction. Georgiadis (2013) developed an integrated SD model for strategic capacity planning in closed-loop recycling networks, which combined with the simulation discipline and the feedback control theory. Zaim, Bayyurt, Tarim,

Zaim and Guc (2013) examined how the activities and variables of knowledge management process interact with each other and how they affect organization with SD model. Bouloiz, Garbolino, Tkiouat and Guarnieri (2013) developed a SD model to analyze causal interdependencies between safety factors of the complex chemical storage system. Tong and Dou (2014) developed a simulation system with SD model to help with decision-making and classifying safety investments by structurally analyzing the causality between safety investments and their influence factors.

In the area of logistics or supply chain management, a great amount of papers can be found developed by SD methodology. Teimoury, Nedaei, Ansari and Sabbaghi (2013) proposed the SD approach to study the behaviors and relationships with in the supply chain of perishable fruits and vegetables, taking into account the influence of import quota policy. Tako and Robinson (2012) explored the application of discrete event simulation and SD methodology as decision support systems for logistics and supply chain management by looking at the nature and level of issues modelled. Poles (2013) modelled a production and inventory system for remanufacturing using a SD approach to explore the dynamics of the remanufacturing process and to evaluate system improvement strategies. Shouping, Qiang and Lifang (2005) analyzed the area logistics system and established a SD model, illustrating by the case of Guangzhou. Rasjidin, Kumar, Alam and Abosuliman (2012) used SD methodology to investigate the influence of weather and forward contract conditions on the fluctuation of energy supply and demand in order to minimize energy retailer's cost.

In this work, we develop a SD methodology to simulate the evolution of Beijing's logistics demand, taking economic, social and environmental factors into account, which is, to some extent, creative for the use of SD methodology, combing with some achievements in the area of society, city and area economic theories, logistics management theories, as well as some ideas of sustainability.

3. Study Area and Methodology

3.1. Area of Beijing

Beijing city, as the capital of China, is one of the fastest economic growing and most densely populated cities in the world, which is playing a pivotal role in the area of politics and culture for the whole country. Beijing is composed with 16 districts and counties like Dongcheng District and Miyun County, which covers 16410.54 kilometers. Every district or county plays different roles in Beijing, with which Beijing government grouped all the districts and counties for four different functional zone. Up to 2013, more than 21.14 million people lived in Beijing, who brought Beijing's GDP to 1950 billion with a growth of 7.7%. Modern service industry in Beijing plays increasingly important role, which account for 76.9% in the whole GDP. In the

context of the strategy for integration of Beijing, Hebei and Tianjin, Beijing will be trapped with a totally new situation, as well as logistics in Beijing. As illustrated in Figure 1, Beijing city, the studying area is located in the east of China between the range of 115.7°E-117.4°E and 39.4°N-41.6°N, next to Hebei Province and Tianjin City.



Figure 1. Studying area

3.2. System Dynamics Methodology

System Dynamics is an integrated methodology that combined system scientific theory with computer simulation, believing that the internal structure is the determination of the behavior model and features of a system (Zhong, Jia & Qian, 2013).

The structure of a system in SD methodology is exhibited by causal-loop diagram (CLD) (Georgiadis et al., 2005); the CLD reflects the major feedback mechanisms. There are many variables in a CLD connected with arrows and influence lines, forming several causal chains and loops. The direction of influence lines means the direction of the effect of a causal chain; in details, when there's a sign "+" at the upper end of an influence line, the variables on the two sides of the influence line change in the same direction, otherwise they change in the

opposite one. All the variables and influence lines sketch the system of negative (balancing) and positive (reinforcing) feedback loops. In a negative feedback loop, the system seeks to a balancing situation. To the opposite, an unstable equilibrium can be exhibited in a positive feedback loop.

Stock and flow diagram (SFD) presents another important scheme in SD methodology with the stock and flow variables, which helps people do the quantitative analysis. Stock variables are the reflections of the accumulation, which picture the states of the system. While flow variables are the only reason diversifying stocks, representing the flows in the system.

Nowadays, with the help of computer, great simulation programs like DYNAMO, iThink, Vensim® and Powersim® are supporting the researches on different areas with SD methodology.

4. Model Description

4.1. Boundary and Structure

SD model must be established with a closed-system boundary, within which the system interactions take place that give the system its characteristic behavior (Forrester, 1969). Therefore, it is fundamental to determine where the boundary is when build the simulating model for Beijing's logistics demand and illustrate it appropriately. Although the boundary of a complex social system is difficult to illustrate, the components or the subsystems of this system can be structured, as illustrated in Figure 2. The whole model can be divided into five subsystems, which separately are production subsystem, civil life subsystem, urban logistics subsystem, natural environment subsystem and social subsystem. There will be some key indexes that connect different subsystems, as shown in Table 1, which will help with the construction of causal loop diagram.

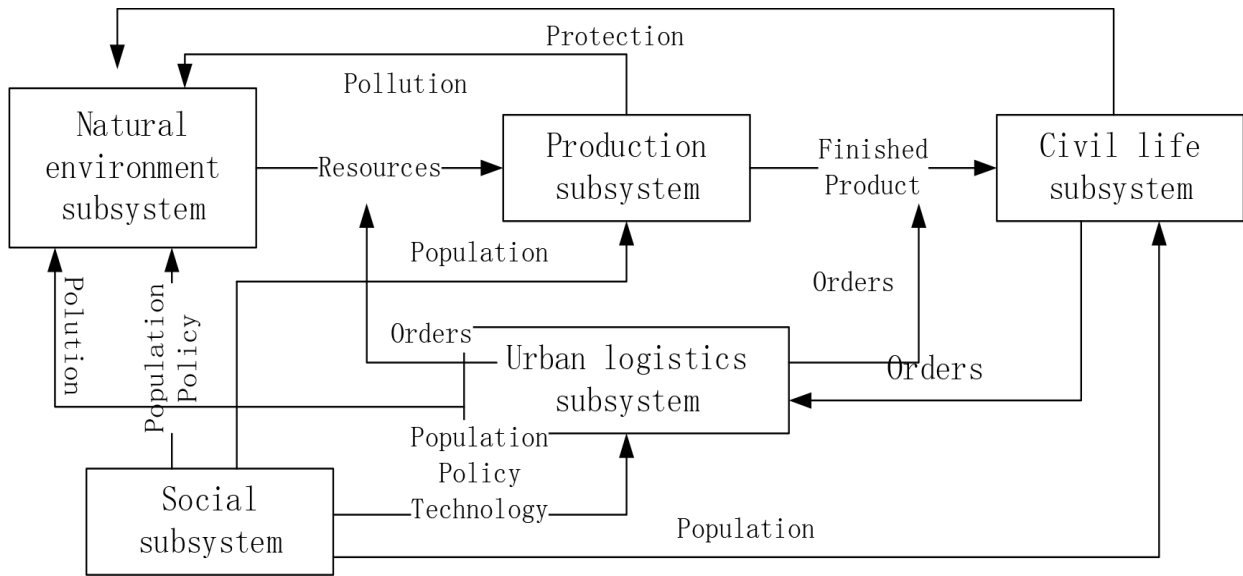


Figure 2. Boundary and structure of SD model

Subsystems	Production	Civil life	Urban logistics	Natural environment	Social
Production	--	Finished product	Orders	Pollution	Population
Civil life	Finished product	--	Orders	Protection	Population
Urban logistics	Orders	Orders	--	Pollution	Population Policy Technology
Natural environment	Resources Pollution	Protection	Orders	--	Population Policy
Social	Population	Population	Population Policy Technology	Population Policy	--

Table 1. Key indexes between subsystems

4.2. Identification of Parameters

Based on all the subsystems, key performance indicators can be put forward as Table 2 illustrated. Parameters are composed of stocks, flows and auxiliaries, which are clearly shown in Table 2 with their notations, descriptions, types and units. Parameter of Freight Volume is the representative of Beijing’s logistics demand, while CO₂ Emission will explain the effect of production and logistics to environment. At the same time, parameter of Average Shipping Capacity for A Vehicle will tell the influence of logistics technology and Bearing Capacity illustrates the competition between congestion and the increase of vehicles. To tackle with some dilemmas, the government has proposed some important policies to decrease

congestions and environment pollution in Beijing. Based on all the parameters, we can do some analysis and tests on the policies by changing the value of some important parameters, which will be presented in Section 5.

Notation	Parameters	Description	Type	Unit
L	Labor	Total amount of labor of production in Beijing	Stock	Million persons
Po	Population	Total population in Beijing	Stock	Million persons
GDP	GDP	The amount of annual GDP of Beijing	Stock	10 million Yuan
OA	Orders accomplished	Total weight of accomplished orders in Beijing	Stock	T
FGI	Finished Goods Inventory	Total weight of goods inventory in Beijing	Stock	T
CE	CO ₂ emission	Total amount of CO ₂ emitted	Stock	Cubic meter
VA	Vehicle amount	The amount of freight wagons in Beijing	Stock	Unit
FV	Freight volume	Total weight of annual freight in Beijing	Stock	T
LI	Labor Increase	Total amount of labor of production increased in Beijing during one year	Flow	Million persons/year
PI	Population Increase	Total population increased in Beijing during one year	Flow	Million persons/year
GDPI	GDP Increase	The amount of annual GDP increase in Beijing	Flow	10 million Yuan/year
Pr1	Production	Total weight produced in Beijing in one year	Flow	T/year
S	Sales	Total weight sold in Beijing in one year	Flow	T/year
OP	Order Production	Total weight ordered in Beijing in one year	Flow	T/year
VAI	Vehicle Amount Increase	The amount of freight wagon increased in Beijing during one year	Flow	Unit/year
VAD	Vehicle Amount Decrease	The amount of freight wagon decreased in Beijing during one year	Flow	Unit/year
FVI	Freight Volume Increase	Freight increased in Beijing during one year	Flow	T/year
FVD	Freight Volume Decrease	Freight decreased in Beijing during one year	Flow	T/year
LIR	Labor Increase Rate	Labor in production increased rate every year	Auxiliary	--
Pr2	Productivity	Productivity of per labor	Auxiliary	T
PIF	Population Increase Fraction	Population increased rate every year	Auxiliary	--
GDPIR	GDP Increase Rate	GDP increased rate every year	Auxiliary	--

Notation	Parameters	Description	Type	Unit
GDPPC	GDP Per Capital	GDP in a year/population of the year	Auxiliary	Yuan
RM	Raw Material	Weight of material used in production in a year	Auxiliary	T
PC	Production Capacity	Production capacity for all labors in a year	Auxiliary	T
FU	Fuel Use	Fuel used for freight in a year	Auxiliary	L
AFUAV	Average Fuel Use for A Vehicle	Average fuel use for the freight wagon	Auxiliary	L/Vehicle
BC	Bearing Capacity	Road capacity in Beijing	Auxiliary	Unit
C	Congestion	Index of congestion in Beijing	Auxiliary	--
OAPP	Order Amount Per Person	Average weight of order per person	Auxiliary	T
OPR	Order Production Rate	Rate of population who will make an order	Auxiliary	--
BO	Backlog Order	Order cannot be accomplished	Auxiliary	T
ASCFV	Average Shipping Capacity for a vehicle	Average carries of the freight wagon	Auxiliary	T

Table 2. Parameters list of SD model

4.3. Causal-Loop Diagram

Causal-loop diagrams are the skeleton of SD model, which should be constructed after the identification of parameters. Figure 3 shows the causal-loop diagram of this paper, within which the cause and effect relationships among parameters can be found.

There are six causal loops in total and two of them are reinforcing and the other four are balancing. Following is the explanations of every causal loop. Production will be the power to promote the operations of the system, which absorbs the help of raw material and labor. Increase of production will result in the increase of finished goods and then the shipping to customers, which finally forced the freight volume. Being different from production, population will be another force in the system, whose increase brings the growth of labor and orders, rise of freight volume will be the result. Technology is definitely the pioneering promotion to the circulatory rise of the system, which will bring the increase of orders and decrease the pollution and congestion.

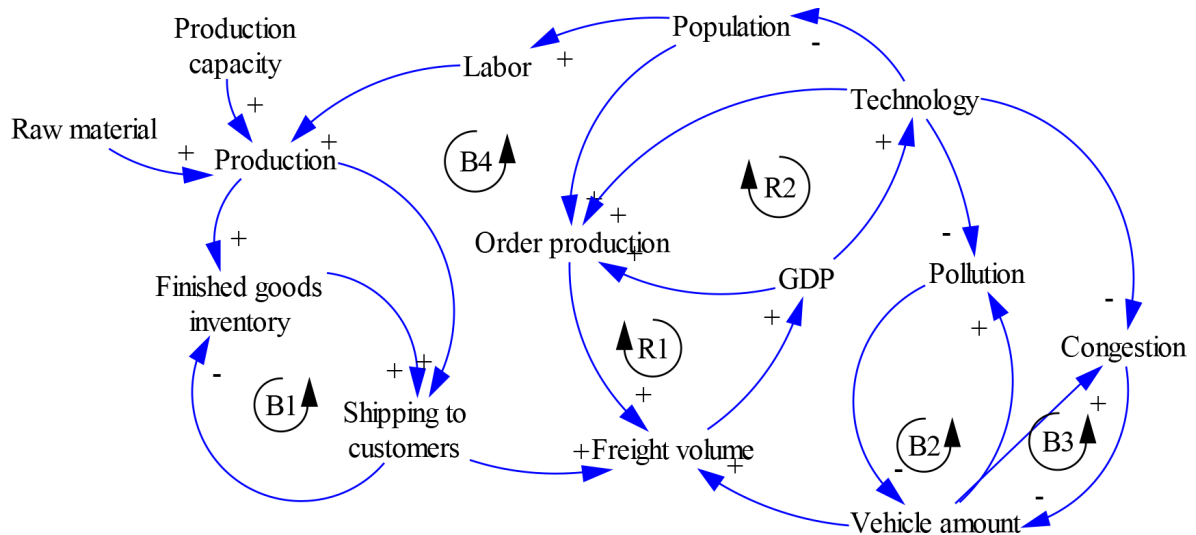


Figure 3. Causal loop diagram of SD model

- (1) Finished inventory → shipping to customer → Finished inventory (Balancing-1)
- (2) Order production → Freight volume → GDP → Order production (Reinforcing-1)
- (3) Order production → Freight volume → GDP → Technology → order production (Reinforcing-2)
- (4) Vehicles amount → Pollution → Vehicles amount (Balancing-2)
- (5) Vehicles amount → Congestion → Vehicles amount (Balancing-3)
- (6) Production → Shipping to customer → Freight volume → GDP → Technology → Population → Labor → Production (Balancing-4)

4.4. Stock and Flow Diagrams

Stock and flow diagram is the algebraic representation of model based on the causal loops identified (Egilmez and Tatari, 2012). In this section, stock and flow diagram is shown as Figure 4. In SD model, stocks can be calculated with the integration of their flows, whose structure is presented in (1) as an integration equation. Stocks will be first defined and then are the flows and auxiliaries.

$$Stock(t) = \int_{t_0}^t [Inflows(s) - Outflows(s)]ds + Stock(t_0) \tag{1}$$

Following is the separate explanations for some key variables.

- Labor (L): Labor is a stock parameter, which will be calculated as follows:

$$L(t) = \int_0^t LI dt + L(0) \quad (2)$$

$$LI(t) = L(t - 1) * LIR \quad t \geq 1 \quad (3)$$

Where LIR will be decided by the change of Population Increase Fraction (PIF), which will be calculated by the multiplication of the PIF and a proper factor.

- Population(P):

$$P(t) = \int_0^t PI dt + P(0) \quad (4)$$

$$PI(t) = P(t - 1) * PIF \quad t \geq 1 \quad (5)$$

Where PIF is calculate by a lookup function with data from Beijing's annual statistical book.

- GDP:

$$GDP(t) = \int_0^t GDPI dt + GDP(0) \quad (6)$$

Where GDPI is the GDP increase every year calculated as (7), the GDP Increase Rate (GDPIR) is calculated by a lookup function with data from Beijing's annual statistical book.

$$GDPI(t) = GDP(t - 1) * GDPIR \quad (7)$$

- Finished Goods Inventory (FGI):

$$FGI(t) = \int_0^t (Pr1 - S) dt + FGI(0) \quad (8)$$

Parameter Pr1 will be affected by two other indexes of RM and PC, the minimum value between the two parameters will decide the value of Pr1, whose calculation will be

$$Pr1 = Min(PC, \alpha RM) \quad (9)$$

α is a changeable factor that means a transfer rate from raw material to finished products. RM is assumed as a constant variable and PC is calculated as:

$$PC = L * Pr2 \quad (10)$$

- Orders Accomplished (OA):

$$OA(t) = MIN\{FGI, \int_0^t (OP - BO)dt + OA(0)\} \quad (11)$$

As the model structured, OP is assume the same as S, which is affected by parameters of OAPP, GDPPC, P and OPR, will be calculated as

$$OP = P * OPR * (1 + e^{1/GDPPC}) * OAPP \quad (12)$$

- CO₂ Emission

CE will arise by procedure of production and the fuel use for freight, whose calculation is

$$CE = \beta P + \gamma FU \quad (13)$$

β and γ is the rate of CO₂ from production and fuel use for freight.

$$FU = AFU * VA \quad (14)$$

AFU is treated as a constant variable.

CE is related to congestions in cities. People have to use less vehicles if CO₂ emission is very high, what helps decrease congestion coincidentally. So, the calculation for Congestion (C) is:

$$C = VA * (1 - e^{\frac{1}{CO_2} Emission}) / BC \quad (15)$$

Bearing Capacity (BC) represents the total amount of vehicles could be drive on the road, which is assumed as a constant number since there are not so many spaces for Beijing to the road construction.

- Vehicle Amount (VA):

$$VA(t) = \int_0^t (VAI - VAD)dt + VA(0) \quad (16)$$

$$VAI = \frac{S}{ASCAV} - VA \quad (17)$$

Where the S is the same as P and the ASCAV is a constant variable.

$$VAD = OA / ASCAV * e^{(1/CE)} * C \quad (18)$$

- Freight Volume (FV):

FV is treated as the representative of logistics demand in this paper, whose calculation is:

$$FV(t) = \int_0^t (FVI - FVD)dt + FV(0) \quad (19)$$

$$FVI = VAI * ASCAV \quad (20)$$

$$FVD = VAD * ASCAV \quad (21)$$

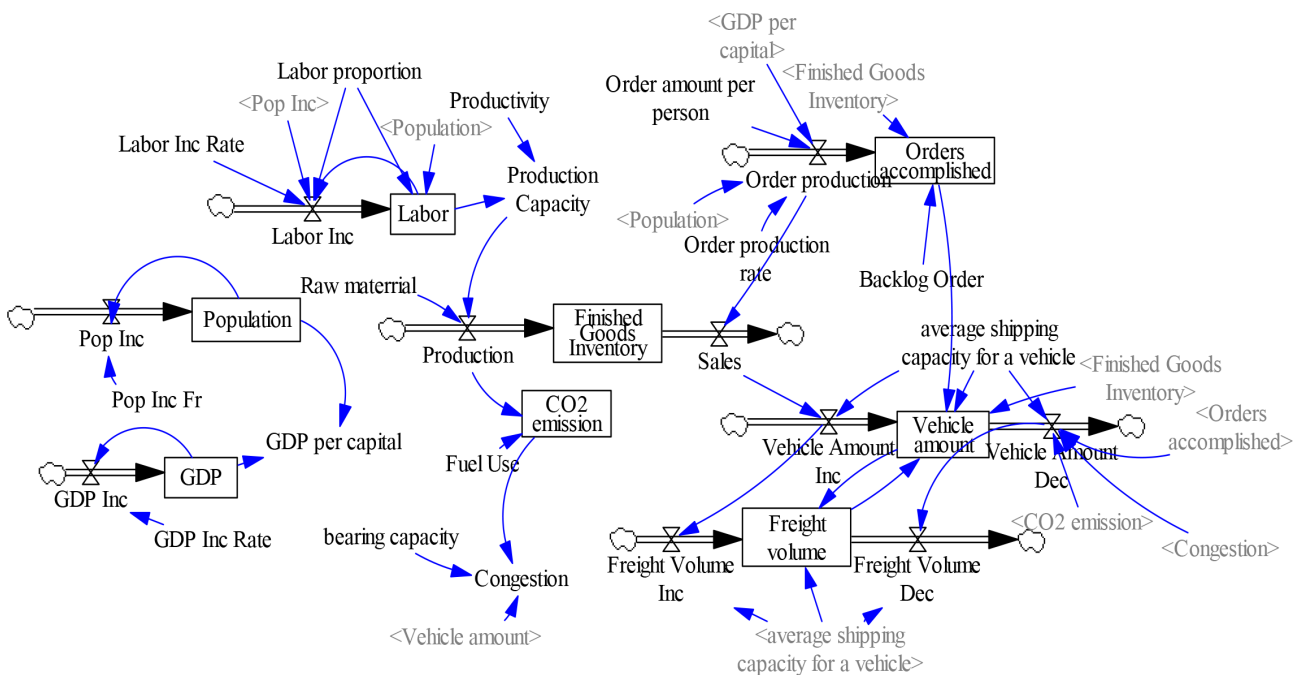


Figure 4. Stock and flow diagram of SD model

5. Policy Analysis

As the political center of China, Beijing is the city where most of the key policies are put forward. With the speeding up expansion of the capital, Beijing is faced with more and more problems, which forced Beijing's government to issue some powerful measures and polices. Some of the policies were closely related to Beijing's logistics system.

Promotion of logistics technology, especially some technologies to increase logistics efficiency and decrease the pollution to environment.

Ways to enhancing the bearing capacity of Beijing city, no matter the by the way of construction or by the way of management.

Based on the structured SD model, we will try to verify the policies with the simulations.

5.1. Promotion of Logistics Technology

In this paper, parameter of ASCAV is treated as the representative of logistics technology, which means that if the average shipping capacity of a vehicle increased, then we thought that logistics technology developed.

5.1.1. Enhance the Average Shipping Capacity of a Vehicle

To verify this kind of policy, this paper selected the parameter of ASCAV. The initial value of ASCAV is 0, which was set as 40t, 80t, 120t and represented with a green line, a red line and a blue line separately. And we got pictures as shown in Figure 5, 6 and 7. In Figure 5, FV doesn't present any difference when the ASCAV changes, which means this kind of technology is not the force of Beijing's logistics demand. But we do find some changes in Figure 6 and Figure 7, in which the amount of vehicle and CO₂ emission show a downward trend with the increase of ASCAV. The experiment results give three enlightenments to us:

- Technology on improving freight facilities won't stimulate demand, which means if the government wants to obtain an increasing logistics demand, some other kinds of technology like e-commerce or rush delivery may be proposed.
- Bigger wagons can bring less vehicle amount.
- Bigger wagons can bring less CO₂ emission.
- Recently, vehicles with yellow license plate are forbidden to appear inside the Fifth Ring Road during the day time to ensure the safe of the city and unimpeded traffic, which means all the freight wagons couldn't come into the city center unless the goods in freight wagons are changed into little vans with blue license plate. The result of this phenomenon is the increase of vehicles and against with the experiment results based on SD model, which will finally lead to higher CO₂ emission. Beijing government has no choice to put forward with the forbidden policy, though the consequence will do harm to efficiency and environment.

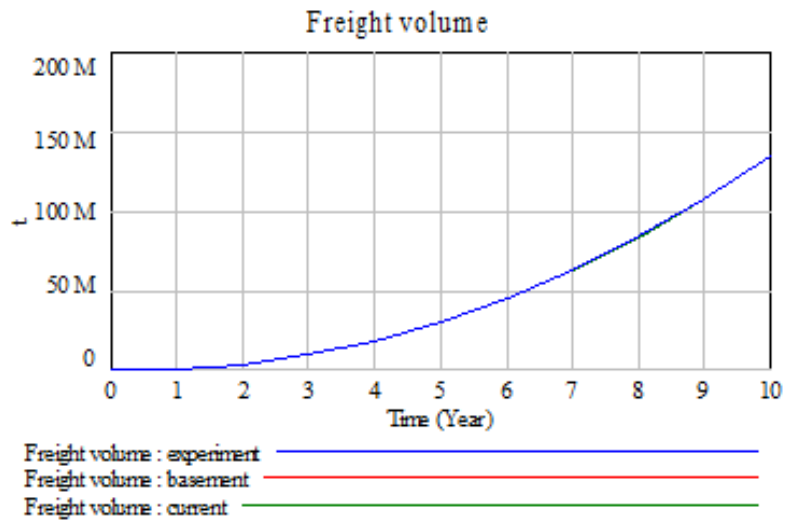


Figure 5. Experiment for freight volume with changes of ASCAV

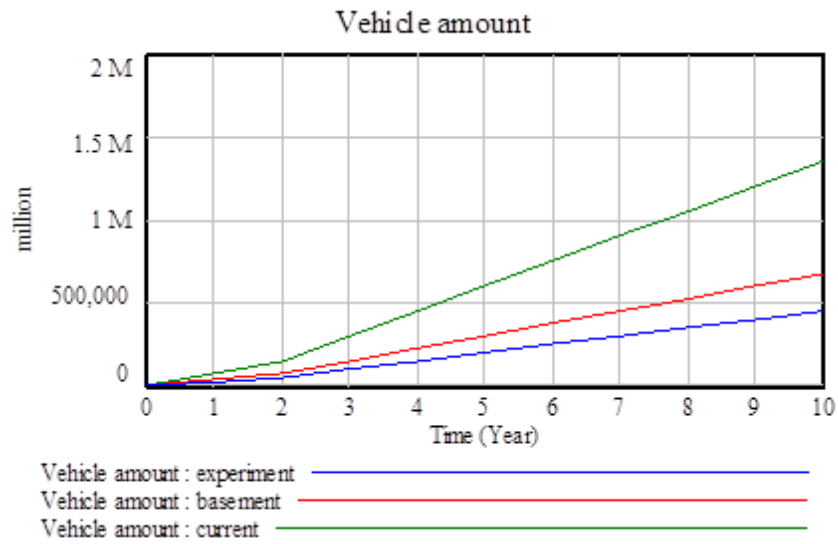


Figure 6. Experiment for vehicle amount with changes of ASCAV

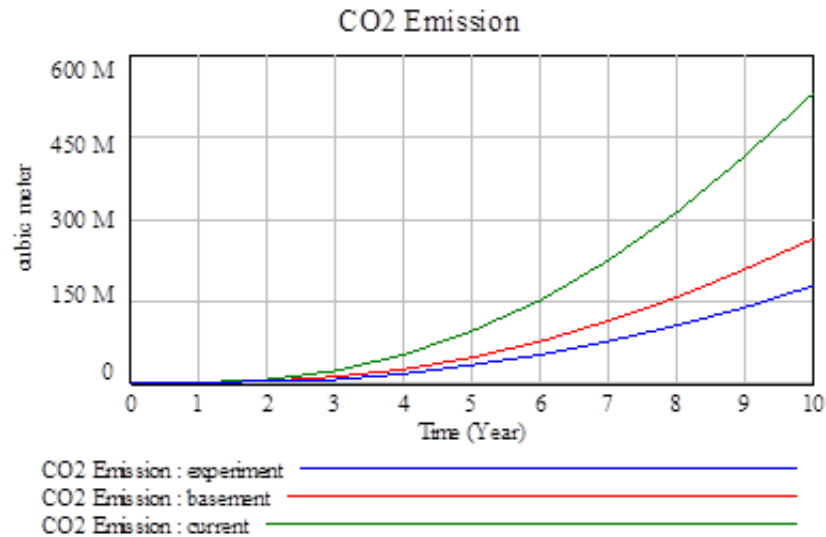


Figure 7. Experiment for CO₂ emission with changes of ASCAV

5.1.2. Reduce Average Fuel Use for a Vehicle

Beijing government has issued some policies to vehicle users for the decrease of average fuel use for a vehicle, which won't bring a directly change of logistics demand either, but will be beneficial to environment protection. AFUAV is an auxiliary in SD model, which is regarded as the representative of technology, especially for protect environment from being polluted from freight. So AFUAV is designed as 5000, 10000, 15000 and illustrated with a green line, red line and blue line. The result of experiment tells that with the decreasing fuel use CO₂ emission and the rate of the increase of the emission will go down.

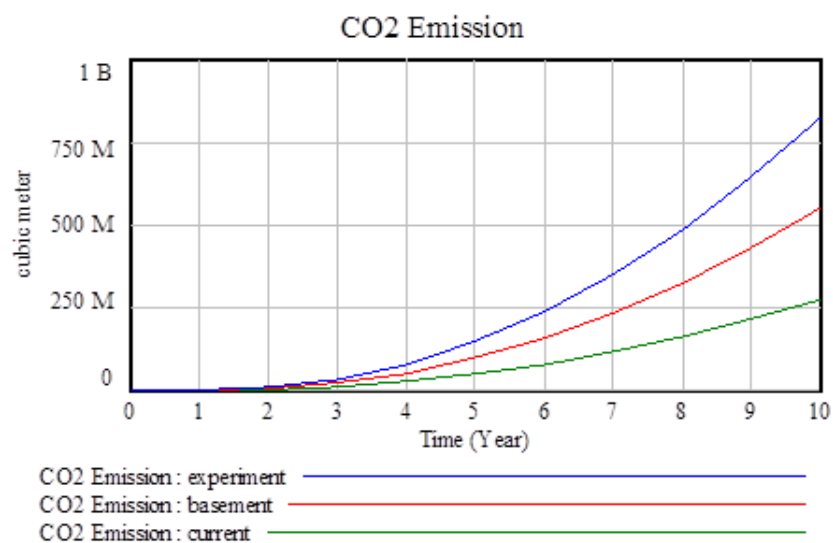


Figure 8. Experiment for CO₂ emission with changes of AFUAV

Based on the above experiments and analysis, we make sure that ASCAV and AFUAV won't bring a change to freight volume directly, though the benefit to social and environment is obvious. But on the other side, decrease of vehicle amount can help with the reduction of congestion, which could bring a better traffic condition for urban logistics. If the CO₂ emission keeps going on, it is definitely that government will force people to reduce the holding volume of vehicles, which will make many orders delay and reduce urban logistics demand mediately.

5.2. Ways to Enhance the Bearing Capacity

Parameter of BC represents the bearing capacity of the road in Beijing. The limit of bearing capacity in Beijing has been broken through since the booming of population and vehicles. Heavy congestions are everywhere in Beijing, which lead to delay of freight and decrease of logistics demand mediately. We assume the value of BC as 200, 500, 800 and illustrated the changes of congestions with green line, red line and blue line, shown as Figure 9. The experiment result shows that with the increase of a cities bearing capacity, congestion will be greatly decreased.

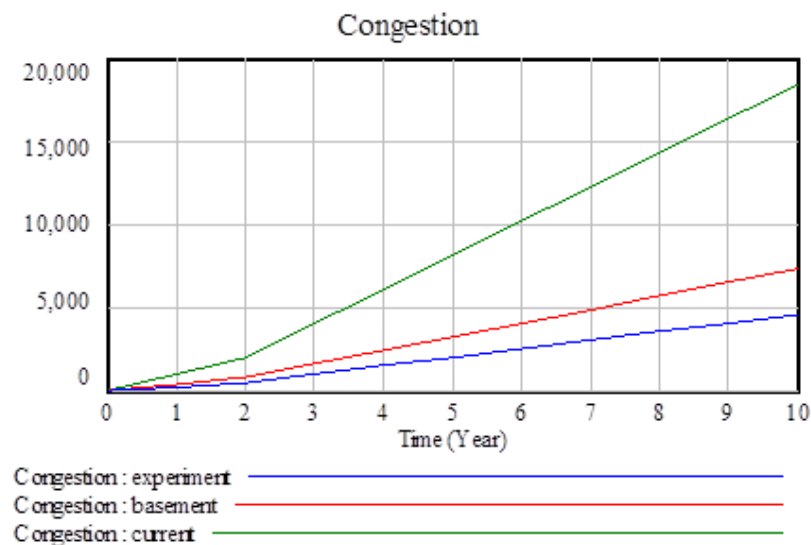


Figure 9. Experiment for CO₂ emission with changes of BC

Actually, congestions usually appear in the center and main road of Beijing, where no space can be used for construction. These years the government put forward with some other policies to enhance the bearing capacity:

- Time limited-entry policy.
- Wagon limited-entry policy
- Limited –entry policy to vehicles authorized in other provinces
- Better traffic management
- Constructions for subways

6. Conclusion

Due to complicated characteristics of urban logistics and sophisticated relations with other systems, the simulation for urban logistics demand is definitely a systematic work, which contains not only the conditions of urban logistics demand itself, but also requires a large amount of work studying on some other related factors like society, population, economy, environment and etc. Based on that, SD methodology is an excellent tool with a systematic thought, which can help us figure out the relationships of numerous variables and find out the variables' changing direction.

This paper gave out a clear list of parameters representing the key performances of Beijing's logistics system. Based on those parameters, causal loop diagram and stock and flow diagram were constructed. With the tested SD model, we realized some experiments by changing some of the auxiliaries in SD model. We found that: (1) technologies like enhancing the average shipping capacity for a vehicle and decreasing the average fuel use for a vehicle will not bring a direct change of logistics demand in Beijing. Some forbidden policies issued by Beijing government will do aggravate congestion and environment pollution; (2) ways to enhance the bearing capacity will decrease the CO₂ emission directly and promote logistics demand mediately. Some limited policies authorized by Beijing government is helpful to decrease congestions and CO₂ emission, and will help with the increase of Beijing's logistics demand.

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