

Road Traffic Speed Prediction using Multi-Source Data

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Abstract :Traffic flow forecasting is of great importance for traffic safety, energy conservation and protection of the environment. Nonetheless, making reliable assumptions about traffic volumes is a difficult problem in the face of many external uncertainties. Many previous types of research only explore the usefulness of a single factor in their prediction and do multi-factor research rarely. As for the prediction of traffic flow, many past types of research focus primarily on the temporal distribution of traffic flow at a single point on the road, ignoring the correlation between spaces. As regards global forecasting, the mechanical view of traffic as images was logically far-fetched. Given the effects of several exogenous variables and the interaction between monitor locations, we are proposing a hybrid model to simultaneously predict traffic flow in multiple positions by combining the layer-wise structure with the Markov transition matrix (MTM). Specifically, we use the layer wise structure to capture traffic flow's periodicity, pattern, and nonlinearity characteristics and then generate the MTM that captures the dynamics expressed in the data and generates the corresponding distributions. We apply the methodology on the traffic data, and the experimental results show that we can achieve satisfactory predictions using our model, which shows the value of the transition matrix in traffic forecasts. We also introduce the point of interest and analyze its impact on the results of forecasts.

Keywords-Markov transition matrix (MTM), traffic, Xiamen

INTELLIGENT TRANSPORTATION SYSTEM (ITS), TOPIC-ENHANCED GAUSSIAN PROCESS AGGREGATION MODEL (TEGPAM)

I. INTRODUCTION TO DOMAIN

The exponential rise within the variety of motorized vehicles creates tie up on urban roads, with the expansion of cities and also the improvement of living standards for folks. This not solely affects the travel expertise of people, however additionally reduces the transport potency, resulting in a decline in social productivity with all. if potential traffic patterns may well be forecast properly, positive steps may well be taken prior to to avoid tie up and also the associated negative impacts. several analysisers have undertaken relevant research to tackle this drawback by mistreatment historical traffic knowledge, that plays a major role within the intelligent transport system. Anacleto et al. [8] extended the linear multi-regressive dynamic model to resolve the measure error caused by errors in knowledge assortment, and additionally incontestible however shut the approximate prediction limit is to truth prediction limit. The growing graded self-organizing map model was imply by Chiou et al. [9] to assist to assist patterns into associate degree acceptable variety of clusters, so develop a genetic programming model for every cluster so as to predict traffic flow characteristics. Besides, Polson and Sokolov [10] developed a Bayesian particle filter to trace

non-linear and discontinuous flows in flow dynamics. Work et al. [11] used partial equation (PDE) to form the model appropriate for any route network. a site is that the target subject of a bug. it's a term utilized in computer code engineering. Formally it represents the target subject of a particular programming project, whether or not narrowly or loosely outlined. A Networking Domain mistreatment during this project. A network domain is associate degree body grouping of multiple non-public pc networks or hosts among an equivalent infrastructure. Domains is known employing a domain name; domains which require to be accessible from the general public web is allotted a globally distinctive name among the name System (DNS).

How does a domain network work?

The name System (DNS) is that the telephone book of the web. ... net browsers move through net Protocol (IP) addresses. DNS interprets domain names to science addresses thus browsers will load net resources. every device connected to the web contains a distinctive science address that different machines use to seek out the device.

ISSUES AND CHALLENGES

- Forecasting the traffic flow is greatly significant for traffic safety, energy conservation, and environmental protection.
- However, in the face of many external uncertainties, making accurate predictions about traffic volumes is a challenging issue.
- Road traffic speed prediction is a challenging problem in intelligent transportation system (ITS) and has gained increasing attentions.

II. RELATED WORK

[1] J. Zhang, F.-Y. Wang, K. Wang, W.-H. Lin, X. Xu, and C. Chen, “Data driven intelligent transportation systems: A survey,” *IEEE Trans. Intell. Transp. Syst.*, vol. 12, no. 4, pp. 1624–1639, Dec. 2011.

For the last twenty years, intelligent transportation systems (ITS) have emerged as associate economical method of up the performance of transportation systems, enhancing travel security, and providing additional decisions to travelers. a major amendment in ITS in recent years is that way more knowledge square measure collected from a range of sources and might be processed into varied forms for various stakeholders. the provision of an outsized quantity of knowledge will probably cause a revolution in ITS development, dynamical associate ITS from a standard technology-driven system into a additional powerful multifunctional data-driven intelligent installation (D two ITS) : a system that's vision, multisource, and learning algorithmic program driven to optimize its performance. what is more, D two ITS is trending to become a privacy-aware people-centric additional intelligent system. during this paper, we offer a survey on the event of D two ITS, discussing the practicality of its key parts and a few preparation problems related to D two ITS Future analysis directions for the event of D two ITS is additionally given.

[2] C. L. P. Chen and C.-Y. Zhang, “Data-intensive applications, challenges, techniques and technologies: A survey on big data,” *Inf. Sci.*, vol. 275, pp. 314–347, Aug. 2014.

This paper is aimed to demonstrate a close-up read regarding huge information, together with huge information applications, huge information opportunities and challenges, we tend to also because the progressive techniques and technologies we presently adopt to subsume the large information issues. we have a tendency to additionally discuss many underlying methodologies to handle the information deluge, as an example, granular computing, cloud computing, bio-inspired computing, and quantum computing.

[3] X. Sun, H. Zhang, W. Meng, R. Zhang, K. Li, and T. Peng, “Primary resonance analysis and vibrations suppression of the harmonically excited nonlinear suspension system using a pair of symmetric viscoelastic buffers,” *Nonlinear Dyn.*, vol. 94, no. 2, pp. 1243–1265, 2018.

The suspension module system model has been established supported MIMO (multiple input and multiple output) state feedback linearization. we've completed decoupling between double suspension points, and therefore the new decoupling technique has been applied to CMS04 magnetic suspension vehicle in national mid-low-speed rail technology experiment field of urban center town in China. Double mechanical system model is incredibly correct for investigation stability property of rail technology system. Once magnetic flux signal is taken back to the suspension system, the suspension module's anti jam capability for resisting suspension load selection has been tried. Also, the external force interference has been increased. As a result, the lustiness and stability properties of double-electromagnet suspension system are increased.

[4] A. Rahim et al., “Vehicular social networks: A survey,” *Pervasive Mobile Compute.*, vol. 43, pp. 96–113, Jan. 2018.

A conveyance Social Network (VSN) is AN rising field of communication wherever relevant ideas are being borrowed from 2 completely different disciplines, i.e., conveyance ad-hoc networks (VANETs) and mobile social networks (MSNs). This rising paradigm presents new analysis fields for content sharing, information dissemination, and delivery services. supported social network analysis (SNA) applications and methodologies, interdependencies of network entities are often exploited in VSNs for prospective applications. VSNs involve social interactions of commuters having similar objectives, interests, or quality patterns within the virtual community of vehicles, passengers, and drivers on the roads. during this paper, considering social networking in a very conveyance atmosphere, we tend to investigate the potential applications of VSNs and communication design. VSNs enjoy the social behaviors and quality of nodes to develop novel recommendation systems and route designing. we tend to gift a progressive literature review on socially-aware applications of VSNs, information dissemination, and quality modeling. Further, we tend to provide an summary of various recommendation systems and path designing protocols supported crowd sourcing and cloud-computing with future analysis directions.

PROPOSED SYSTEM

Given the effects of several exogenous variables and the interaction between monitor locations, we are proposing a hybrid model to simultaneously predict traffic flow in multiple positions by combining the layer-wise structure with the Markov transition matrix (MTM). Specifically, we use the layerwise structure

to capture traffic flow's periodicity, pattern, and nonlinearity characteristics and then generate the MTM that captures the dynamics expressed in the data and generates the corresponding distributions. We apply the methodology on the traffic data, and the experimental results show that we can achieve satisfactory predictions using our model, which shows the value of the transition matrix in traffic forecasts. We also introduce the point of interest and analyze its impact on the results of forecasts

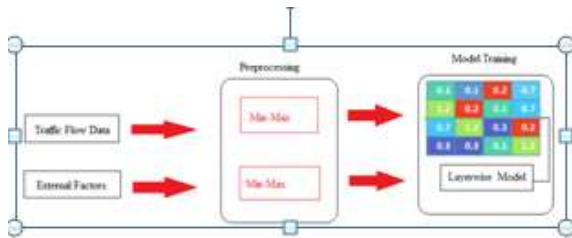


Fig. 1 Proposed System

III. METHODOLOGY

A Markov transition matrix is a square matrix that describes the probabilities of moving within a dynamic system from one state to another. In each row are the probabilities of moving from that row to the other states, from the state represented. The rows of a Markov transition matrix therefore each add to one.

In the first-order Markov model, the center of which is the Markov transition matrix, there is a simple assumption that the transformation of each state depends only on its prior state. This matrix is commonly used in various kinds of fields for data processing. Each matrix entity is a non-negative real number representing the probability of a single-time step from status i to state j .

Modules

1.Load Dataset – This module is used to load the dataset which consists of the details of the traffic flow. The traffic data for this experiment was collected from the detectors distributed at intersections and different lanes in the city. Technologies such as advanced photoelectric image processing and pattern recognition were utilized to preprocess the real-time data of passing vehicles.

2.View Traffic – Used to view the traffic. **Traffic on roads** consists of **road** users including pedestrians, ridden or herded animals, vehicles, streetcars, buses and other conveyances, either singly or together, while using the public way for purposes of travel. ... Organization typically produces a better combination of travel safety and efficiency.

3.Traffic flow forecast – traffic flow forecast displays with time, lane and speed of the current lane. HMM exhibits good results in the analysis of non-stationary and obviously fluctuating feature data systems, and the actual road traffic system is also a complex and non-stationary system. Essentially, traffic flow as an observation sequence is continuous numerical data. Considering that is the fourdimensional vector traffic flow at four entrances of an intersection, while is the traffic state at the intersection in a certain period, the state of is one of all hypothetical states.

□ **Comparison** – Gives the comparison of the model with the previous model.

3.2 algorithm

Hidden Markov models (HMM) represent one of methods that are suitable for congestion prediction .in this paper a new model and contrast is proposed to define the traffic states during peak hours in two dimensional space(2D). The proposed model uses mean speed and contrast to capture the variability in traffic patterns.

Hidden markov models:

- Set of status: process moves from one state to another generating sequence of
- previous state
- States are not visible ,but each state randomly generates one of M observations (visible states)
- To define status
- Markov chain property :probability of each subsequent state depends only on what was the hidden markov model the following probabilities have to be specified :
- Matrix of transition probabilities $A=(a_{ij})$,
 $a_{ij} = P(S_i|S_j)$,matrix of observation probabilities $B=(b_i(v_m))$,
 $b_i(v_m) = p(v_m | s_i)$ and a vector of initial probabilities $\pi = (\pi)$,
 $\pi = p(s_i)$ Model is represented by $M=(A,B \pi)$

IMPLEMENTATION



Fig 4:Menu

This is a menu which consists of Load dataset, view traffic, external parametrs, traffic flow forecasting, previous model comparison

Time	VehicleID	Speed	Dist_Front	Section ID	Avg_Sect_Speed	Sect_Density
0	0	0	0	000001	000001	0000
0	0	0	0	000002	000002	0000
01	00	00	0	000003	000003	0000
01	01	00	00	000004	000004	0000
01	02	00	00	000005	000005	0000
01	03	00	00	000006	000006	0000
01	04	00	00	000007	000007	0000
01	05	00	00	000008	000008	0000
01	06	00	00	000009	000009	0000
01	07	00	00	000010	000010	0000
01	08	00	00	000011	000011	0000
01	09	00	00	000012	000012	0000
01	10	00	00	000013	000013	0000
01	11	00	00	000014	000014	0000
01	12	00	00	000015	000015	0000
01	13	00	00	000016	000016	0000
01	14	00	00	000017	000017	0000
01	15	00	00	000018	000018	0000
01	16	00	00	000019	000019	0000
01	17	00	00	000020	000020	0000
01	18	00	00	000021	000021	0000
01	19	00	00	000022	000022	0000
01	20	00	00	000023	000023	0000
01	21	00	00	000024	000024	0000
01	22	00	00	000025	000025	0000
01	23	00	00	000026	000026	0000
01	24	00	00	000027	000027	0000
01	25	00	00	000028	000028	0000
01	26	00	00	000029	000029	0000
01	27	00	00	000030	000030	0000
01	28	00	00	000031	000031	0000
01	29	00	00	000032	000032	0000
01	30	00	00	000033	000033	0000
01	31	00	00	000034	000034	0000
01	32	00	00	000035	000035	0000
01	33	00	00	000036	000036	0000
01	34	00	00	000037	000037	0000
01	35	00	00	000038	000038	0000
01	36	00	00	000039	000039	0000
01	37	00	00	000040	000040	0000
01	38	00	00	000041	000041	0000
01	39	00	00	000042	000042	0000
01	40	00	00	000043	000043	0000
01	41	00	00	000044	000044	0000
01	42	00	00	000045	000045	0000
01	43	00	00	000046	000046	0000
01	44	00	00	000047	000047	0000
01	45	00	00	000048	000048	0000
01	46	00	00	000049	000049	0000
01	47	00	00	000050	000050	0000
01	48	00	00	000051	000051	0000
01	49	00	00	000052	000052	0000
01	50	00	00	000053	000053	0000
01	51	00	00	000054	000054	0000
01	52	00	00	000055	000055	0000
01	53	00	00	000056	000056	0000
01	54	00	00	000057	000057	0000
01	55	00	00	000058	000058	0000
01	56	00	00	000059	000059	0000
01	57	00	00	000060	000060	0000
01	58	00	00	000061	000061	0000
01	59	00	00	000062	000062	0000
01	60	00	00	000063	000063	0000
01	61	00	00	000064	000064	0000
01	62	00	00	000065	000065	0000
01	63	00	00	000066	000066	0000
01	64	00	00	000067	000067	0000
01	65	00	00	000068	000068	0000
01	66	00	00	000069	000069	0000
01	67	00	00	000070	000070	0000
01	68	00	00	000071	000071	0000
01	69	00	00	000072	000072	0000
01	70	00	00	000073	000073	0000
01	71	00	00	000074	000074	0000
01	72	00	00	000075	000075	0000
01	73	00	00	000076	000076	0000
01	74	00	00	000077	000077	0000
01	75	00	00	000078	000078	0000
01	76	00	00	000079	000079	0000
01	77	00	00	000080	000080	0000
01	78	00	00	000081	000081	0000
01	79	00	00	000082	000082	0000
01	80	00	00	000083	000083	0000
01	81	00	00	000084	000084	0000
01	82	00	00	000085	000085	0000
01	83	00	00	000086	000086	0000
01	84	00	00	000087	000087	0000
01	85	00	00	000088	000088	0000
01	86	00	00	000089	000089	0000
01	87	00	00	000090	000090	0000
01	88	00	00	000091	000091	0000
01	89	00	00	000092	000092	0000
01	90	00	00	000093	000093	0000
01	91	00	00	000094	000094	0000
01	92	00	00	000095	000095	0000
01	93	00	00	000096	000096	0000
01	94	00	00	000097	000097	0000
01	95	00	00	000098	000098	0000
01	96	00	00	000099	000099	0000
01	97	00	00	000100	000100	0000
01	98	00	00	000101	000101	0000
01	99	00	00	000102	000102	0000
01	100	00	00	000103	000103	0000

Fig 5:Load Dataset

This module loads the dataset and displays dataset which consists of Time, vehicle id, speed, dist_front,section ID, avg_sect_speed, sect density

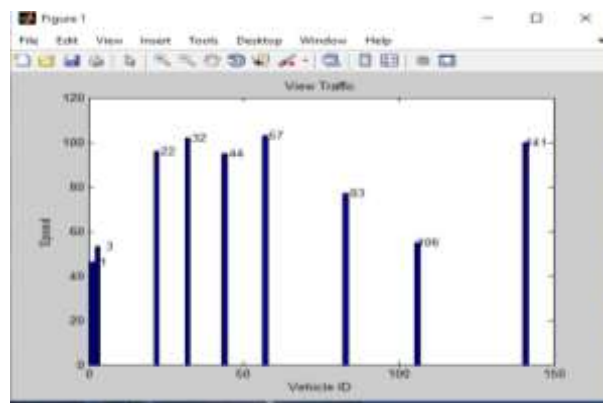


Fig 6: View Traffic

This module displays the traffic flow. It gives the vehicle ID vs Speed. At what speed each vehicle is moving on the road.



Fig 7: Traffic flow forecasting with vehicle ID, speed and vehicle counting

This module displays the traffic flow forecasting with vehicle ID, speed and vehicle counting

Conclusion

IN THIS PAPER WE PROPOSE A NOVEL HYBRID MODEL TO SIMULTANEOUSLY FORECAST THE TRAFFIC VOLUME FROM THE HISTORICAL TRAFFIC PATTERNS OF MULTIPLE LOCATIONS. TAKING INTO ACCOUNT THE TEMPORAL AND SPATIAL CHARACTERISTICS OF TRAFFIC DATA , THE PROPOSED MODEL, WHICH INCORPORATES LAYERWISE STRUCTURE AND MTM, INTEGRATES EXTERNAL FACTORS SUCH AS WEATHER , TEMPERATURE, HOLIDAYS, AND ACTUAL ROAD NETWORK SIZE. IT HAS ALSO BEEN ESTABLISHED THAT THE PROPER UTILIZATION OF POIS AND MTM WILL BENEFIT MODEL PERFORMANCE IMPROVEMENTS. THROUGH COMPARING VARIOUS MODELS IN TERMS OF PREDICTION ERROR, THE LSAE-MTM APPROACH WAS FOUND TO BE MORE SUCCESSFUL AND RELIABLE FOR PREDICTION OF TRAFFIC FLOWS.

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