

The Approaches for Availability of Parking Spots on On- Street and Off- Streets

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Abstract

Abstract-Parking guidance and information (PGI) systems are becoming important parts of intelligent transportation systems due to the fact that cars and infrastructure are becoming more and more connected. One major challenge in developing efficient PGI systems is the uncertain nature of parking availability in parking facilities (both on-street and off-street). A reliable PGI system should have the capability of predicting the availability of parking at the arrival time with reliable accuracy. In this paper, we study the model and approaches used for providing availability parking spots to vehicles. These models are used to predict parking availability with high accuracy. The prediction errors are used to recommend the parking location with the highest probability of having at least one parking spot available at the estimated arrival time.

Keywords

Parking guidance and information (PGI), Intelligent Parking Systems (IPS), WAVE, IPAS, SNN

I. Introduction

For over thirty years, traffic information has been provided to avail motorists make en-route decisions. In more recent years, the development of Intelligent Transportation Systems (ITS) and Advanced Traffic Management Systems (ATMS) have commenced to ameliorate transportation through the utilization of technology. Along the same lines, Perspicacious Transportation Highway Systems that “acquire, analyze, communicate, and present information to avail surface transportation peregrinators in peregrinating from a commencement location (inception) to their desired destination,” can now be utilized for en-route assistance as well as traffic data amassment. And determinately, technology is commencing to agonize the consequentiality of post-trip information dissemination by providing information on the location and availability of parking. Authentic-time information can be accurately provided to motorists through Intelligent Parking Systems (IPS) to reduce congestion in or near parking areas, inadequate utilization of the available parking space stock, road congestion caused by space-probing traffic, access quandaries and safety hazards caused by illicit parking, and environmental strains. Vehicle users must store their transportations at trip end for varying lengths of time. Studies have shown that the average passenger car is at kinetics for only 500 hours and at rest for more than 8,000 hours per year.⁴ In many locations, especially high-density traffic areas such as central business districts (CBDs), shopping centers, university campuses, and airports, parking supply cannot meet parking demand; or at least the perception of the public is that parking is inadequate at these venues.

With the advent of advanced peregrinator information systems and in particular route guidance systems (RGS), the presage of short-term link peregrinate times has become increasingly paramount. Intuitively, the RGS’s route cull algorithms should utilize link peregrinate times that are predicated on the time at which the driver is expected to arrive at a given link rather than use link peregrinate times that are predicated on current conditions. This would be concretely consequential for trips where the expected advent time at a link is relatively far into the future, and it is unlikely that the current link peregrinate time will remain stable.

Because drivers implicitly predicate their routes on the anticipated link peregrinate time when they arrive at a particular link, the RGS should have the same capabilities. Artificial neural

networks (ANN) and in particular multilayer perception neural networks that utilize a back-propagation algorithm have been applied prosperously for forecasting link peregrinates time and other traffic parameters. Hereafter, the term conventional ANN will refer to multilayer perception neural networks that utilize a back-propagation algorithm unless mentioned otherwise. The prosperous applications of conventional ANN may be attributed to the nonlinear and multidimensional nature of many transportation quandaries. However, there are numerous practical shortcomings associated with conventional ANN and a number of methodologies including modular neural networks (MNN) have been proposed to alleviate these shortcomings

Parking guidance and information (PGI) systems aim to provide information to drivers concerning the location, direction and availability of parking spaces. The location or direction information is conventionally static information intended for guiding unfamiliar drivers to car parks. The availability information shows the degree of congestion in a car park such as PLENARY, CONGESTED, SPACES or CLOSED. The operation of the PGI systems are expected to redistribute parking demand among car parks and to reduce queues at the most popular car parks. However, it is not conspicuous how much the PGI systems contribute to reduce queues and congestions. The availability information shown to drivers is conventionally FULWSPACES information. However, it is not sure if it is still efficacious under congested situation. In order to develop an Keenly intellectual Parking System that will accommodate the goals and inductively authorizes of ITS in the twenty-first century, it is consequential to understand the state-of-the art. An in-depth literature search was conducted to locate IPS systems in the design, implementation, and evaluation stages. The studies located can be relegated into the following topics: (1) venues utilizing IPS, (2) rudimentary IPS system requisites, and (3) IPS evaluations. The venues utilizing IPS range from universities to airports. The categorical projects opted to review for each particular venue reflect the state-of-the-art in the Cumulated States and Europe giving a good background of what current perspicacious parking systems entail.

A. Venues Utilizing IPS

The projects reviewed in this section fixate on the following venues or areas: (1) central business district (CBD), (2) University/transit

and (3) airport. The central business district systems reviewed are generally utilized during special events only or when two or more special events occur simultaneously downtown. For example, in Phoenix, Arizona, the presence of the Arizona Diamondbacks, Phoenix Suns, Herberger Theater and other cultural and sporting centres require an IPS system that can accommodate 40,000 to 80,000 visitors. The CBD system designs can have kindred goals for special events, but may vary extensively depending on number of people. University and transit systems are lumped together because a case study at Oregon State University is reviewed that is endeavouring to route a transit shuttle through a parking lot predicated on the location of the entering motorist's parking space. The system will route the motorist to a space then route the shuttle to them.

B. Basic IPS Requirements

Most Intelligent parking systems researched in this literature review were utilized or are currently being designed or deployed in an immensely colossal downtown central business district area. The other concrete cases involving university transit or airport applications contain kindred attributes to those CBD applications including the utilization of variable message signage however, the CBD applications go above and beyond the general utilization of VMS to manufacture a whole system that relates into the local ITS architecture. The CBD applications withal demonstrate the competency to coordinate an authentic-time system over more than one parking facility. The rudimentary perspicacious parking system requisites discussed in this portion of the literature review are consequently predicated on the state-of-the-art CBD applications discussed in the antecedent section. These requisites can be facilely scaled down to accommodate a more minuscule venue such as a university transit or airport parking area. The rudimentary system that coordinates a multiple facility intelligent parking system contains the three main elements of (1) parking facility equipment, (2) central computer and connections, and (3) signage. The parking facility equipment includes the transportation counters, space monitors and processing units found on site to monitor ingress and egress traffic, which is sent conventionally by modem to a central computer. The central computer then controls the variable message signage engendering the desired LED exhibits to direct traffic to open garages or other parking areas. The central computer can withal be programmed to send simultaneous messages by radio frequency, dedicated phone line, or Ethernet connection to the local radio, television station, or Internet.

C. IPS Evaluations

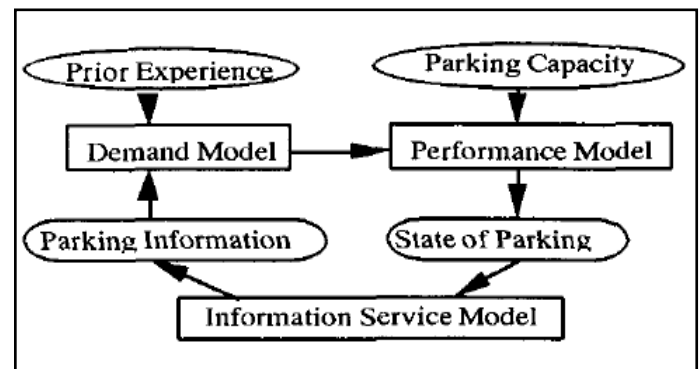
Keenly intellectual Parking Systems can be grouped into categories of "parking management tactics." Parking management tactics are "actions taken to alter the supply, operation, and/or parking injective authorization of a jurisdiction's parking system to further the procurement of local transportation, economic, environmental, and other objectives." More concretely, IPS alters the operation of the parking systems themselves. Because goals of parking management can range from "way-finding" to air emissions reduction to incrementing transit utilization and carpooling, concrete goals should be set afore an IPS is deployed so that the system may be congruously evaluated later. This evaluation section addresses a pre-design strategy utilized in Phoenix to identify IPS need, the outcomes of evaluations in Minnesota and Frankfort, Germany of deployed systems, and the human factors

involved in culling an authentic parking location.

II. A Simulation Model Approach

A. Model Framework

In evaluating the effects of parking information accommodation utilizing a simulation model, it is required that the model describe the dynamic interaction between demand and system performance. In other words, the model must describe temporal fluctuation of driver's parking cull decisions and resulting congestion in car parks. The model is additionally expected to distinguish the difference of a driver's parking cull demeanor between with and without availability information, and to compare the effects of the variants of availability information.



The model consists of three sub-models; demand, performance and information accommodation models. The ordinate dictation model describes a driver's parking cull comportment. The car park cull probability of a driver is calculated through a disaggregate cull model. Transportation is then allocated to a car park due to the cull probability. The inputs of the cull model are a driver's prior cognizance such as parking charge, ambulating distance and availability information when it is shown to drivers. The performance model describes the state change of each car park. Between sequential time periods, the state of a car park may change even if transportation does not arrive at the car park, as a parked transportation may finish its accommodation and leaves the car park. In the performance model, the magnification of the queue in front of a car park and parking accommodation time are numerically calculated with deference to time. Capacity of car parks and average accommodation rate (reciprocal of average parking accommodation time) are postulated given. The decision making process of the parking accommodation ascendancy is described in the information accommodation sub-model. The ascendancy observes the state of car parks and culls the opportune available information exhibited to drivers. Variants of information are additionally engendered in the model, for example, the number of vacant spaces and the waiting time in the queue as well as FULWSPACES information

B. Demand Model

Drivers are categorized into two groups: apprised and non-informed drivers. All drivers are postulated to have prior erudition on the attributes of car parks such as ambulating distance to their final destinations and parking charge. Unless they are apprised through PGI systems, they do not ken the availability such as the congestion level of car parks. While a non-apprised driver has to opiate a car park without availability information, an apprised driver can utilize availability information in advisement to the prior cognizance. A multi-nominal logit (MNL) model is applied

for formulating the car park cull behavior since a non-apprised driver has some information on location and situation of car parks except for availability, his/her utility function can be expounded by those variables. When a driver obtains availability information, the driver will utilize the information as well as prior erudition and the variable betokening parking information is involved in the utility function. The deterministic part of the utility function for choosing the *i*-th car park is denoted by V_i . Note that the subscript of an individual is omitted.

Then functional form of V_i is written as,

$$V_i = \sum_{k \in A_1} \beta_k X_{ik} + \delta \sum_{h \in A_2} \gamma_h I_{ih}$$

C. Performance Model

Two variables are habituated to describe the states of car parks. NQ_i and NS_i , denote the number of queuing transportations and the number of parked transportations in car park *i*, respectively.

D. Information Accommodation Model

When we monitor the states of car parks, it is possible to produce several types of the availability information; FULL/SPACES information, the number of vacant spaces and the expected waiting time. For simplicity, the strategy of the parking ascendancy is surmised to show the current states of car parks to drivers without any modification and presage.

III. Spectral Basis Neural Networks (SNN)

It amalgamates a pre-transformation of the input features and conventional ANN for forecasting multiple-periods link peregrinates times into the future. The pre-transformation is predicated on sinusoidal functions and is performed on the input features to obtain linearly separable input features. The goal is to convert the intricate, non-monotonic function that relates future peregrinate times to the input features into a monotonic function. The data utilized as input was link peregrinate times from Houston that were accumulated as a component of the Automatic Transportation Identification (AVI) system of the Houston Transat system.

1. Input Feature Transformation (SNN)

This paper proposes an SNN that incorporates an input transformation or pre-mapping. The SNN is withal kened as an expanded input classifier, functional expansion, and higher order or augmented neural network. The motivation for this architecture is that relegation errors can be reduced and the ANN performance ameliorated, if the hard vectors are transformed into another set of vectors (that is, into a different dimension) so that they are more facile to relegate. The hard vectors are converted utilizing a nonlinear transformation such that the input features become simpler to relegate without authentically introducing any incipient information (Han 1997). In this paper the nonlinear transformation is predicated on a sinusoidal transformation. Pao (1989), Ersoy and Hong (1990), and Eck and Shih (1994) introduced kindred approaches utilizing variant transformation functions. The proposed SNN is explicated with a simple example. The figure represents an example of a non-monotonic function with 1D input *x*. The output *y* has a bipolar value (i.e., 1 or -1). Surmise that A1 and A2 belong to class "A" that has a target value of 1 for *y* and that B1 and B2 belong to class "B" with a target value of -1 for *y*. The requisite is to construct a pattern classifier that engenders

output 1 in replication to input patterns A1 and A2 and output 21 in replication to input patterns B1 and B2. However, note that the *y* value cannot be calculated from the input value *x* utilizing any linear function. It is relatively facile to show that this quandary is not linearly separable with a one-layer structure and would require a three layer structure to relegate the A and B regions (Hagan et al. 1995).

In contrast to the above linear perception approach, the problem can be classified using only one layer if an appropriate nonlinear transformation is employed. Eq. Shows a nonlinear transformation based on sinusoidal functions

$$x_1 = \sin(2\pi x); \quad x_2 = \sin(4\pi x)$$

The test bed for this study was US 290, which is a radial six-lane urban freeway in Houston as shown in below Fig... It has a barrier-separated HOV lane that runs along the centerline of the freeway for approximately 19 km, and the data utilized were from the non-HOV section of the freeway.

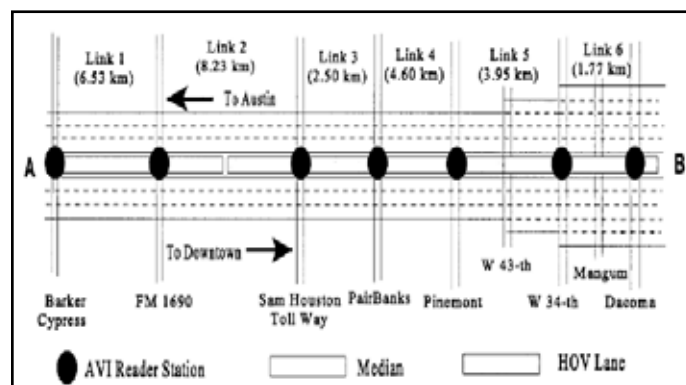


Fig: Study Urban Freeway Corridor US 290 in Houston

2. Study Design of SNN

The proposed SNN was habituated to forecast peregrinate times on link 4 for one through five time periods into the future. The sigmoid function was utilized as a neuron transfer function for the obnubilated layer(s) and a linear function were utilized for the output layer (Haykin 1994). The number of neurons in the obnubilated layer depends on the pattern and intricacy of the approximated function and the transfer function of the layers. The congruous number of neurons was culled through a preliminary analysis. The best SNN structure was identified after a sensitivity analysis involving the number of obnubilated layers and the number of neurons. To minimize the squared error during training, a steepest descent algorithm was incorporated. The SNN was entirely coded in FORTRAN and run on a 133- MHz Pentium desktop computer.

Intuitively, the average peregrinates times of the target link in the preceding time periods are paramount parameters for identifying future peregrinate time patterns. The peregrinate time information from the anterior five time periods was utilized, as this range had given the best results in an antecedent study. Another paramount parameter is the link peregrinates times experienced on the upstream and downstream links during the preceding time periods. A shockwave that composed upstream or downstream from the target link has the potential to affect the target link sometime in the future. In this study, the link peregrinates times of the links immediately upstream (link 3) and downstream (link 5) were culled for inclusion predicated on anterior results. Given the nature of the AVI system, other consequential parameters such as volume

and density were not available and ergo could not be utilized in the analysis. The output space consists of the link peregrinate forecasts for the next five consecutive time periods. The best model was chosen based on the average prediction error for all time periods

$$MAPE_{\delta} = \frac{\sum_{k=1}^n \frac{|\pi(k + \delta)_p - \pi(k + \delta)_o|}{\pi(k + \delta)_o}}{n} \times 100, \quad \delta = 1, 2, \dots, 5$$

IV. Intelligent Parking Assistant System (IPAS)

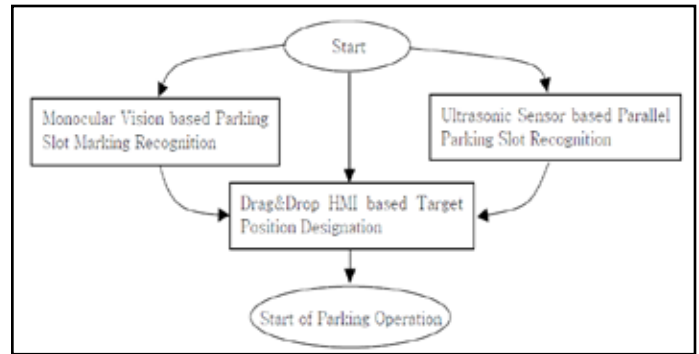
IPAS sanctions driver to designate target position by three complimentary methods: monocular vision predicated parking slot marking apperception, ultrasonic sensor predicated parallel parking slot apperception, and drag & drop GUI (Graphical Utilizer Interface). IPAS engenders an optimal path to reach the designated target position. During parking operations, IPAS estimates ego-transportation pose utilizing ESP (Electric Stability Program) sensors such as wheel speed sensors, braking pedal switch and steering angle sensor. IPAS automatically controls braking and steering to achieve orchestrated trajectory by sending required actuations to ESP and EPS (Electric Power Steering) via CAN. Furthermore, IPAS apprises driver of the perpetual parking operation by exhibiting estimated trajectory on the rear view image. Proposed system is validated by transportation experiments.

A. System Configuration

IPAS consists of six components: target position localization, ego-transportation pose estimation, path engenderer & tracker, active braking system, active steering system, and HMI (Human Machine Interface). IPAS can implement not only semi-automatic parking assistant system, in which steering operation is automated, but additionally automatic parking avail system, in which steering and braking operation is automated. Target position localization designates the target position of automatic/semi-automatic parking operation. We developed three complimentary methods: monocular vision predicated parking slot marking apperception, ultrasonic sensor predicated parallel parking slot apperception, and drag & drop GUI (Graphical Utilizer Interface). Ego-transportation pose estimation implements Ackerman model predicated transportation pose estimation by utilizing sundry sensors including wheel speed sensors, steering angle sensor, braking pedal switch, and wheel angle sensor. These sensors are annexed to ESP (or, ESC) and EPS and report their quantifications via CAN. Path engenderer makes an optimal path to reach the designated target position. Path tracker controls steering and braking actuation to make the estimated ego-transportation pose follow the orchestrated path. During parking operation, IPAS shows rear-view image to driver through the HMI and overlaps estimated trajectory on it.

B. 3D Localization of Target Position

Our 3D localization of target position supports 2 automatic methods and 1 manual method as shown in below Figure



Monocular vision based parking slot marking recognition designates target position in case of perpendicular parking situation with parking slot markings. Ultrasonic sensor based parallel parking slot recognition designates target position in case of parallel parking situation with both forward and backward vehicle. Once driver gets target position using automatic method, driver can adjust or refine the target position by manual method. Naturally, if automatic target position localization methods are not available, user can designate target position by manual method, i.e. drag & drop concept based GUI. Although there have been researches to detect target position by recognizing 3D information of adjacent vehicles [1][2][3], such kind of methods are supposed to be expensive for the first generation. Monocular vision based parking slot marking recognition uses a monocular CMOS camera installed on the backend of vehicle. The vision system uses a wide-angle lens to cover wide FOV (Field Of View). It rectifies input image to compensate the radial distortion of wide-angle lens, and then transforms the undistorted image into bird's eye view image. Bird's eye view image is a kind of virtual image taken by a flying bird. Vision system starts recognition procedure from the seed point, which is designated by driver with touch screen.

C. Drag & Drop HMI Predicated Target Position Designation

Captured scene of parking site is exhibited on minute touch screen monitor and current target position is exhibited. The inside region of rectangular target position is utilized as a moving cursor. The outside region of rectangular target position is utilized as a rotating cursor. Utilizer can facilely designate the target position by the moving and rotating cursor. Manual target designation method has two usages. First, it can refine the target position acquired by automatic target localization methods. Second, it can be utilized at any time driver want to manually designate target position without automatic methods. Because manual method is supposed to be inevitably ineluctable in spite of the development of automatic methods, research focus should be on the amendment of efficiency and accommodation. To verify the efficiency of the proposed manual method, we quantify the operation time and clicking number, and then compare drag & drop predicated method with multiple-arrow predicated method.

D. Path Orchestrating and Tracking

After the localization of target position, parking controller engenders parking path for semi-automatic guidance. Orchestrated path of garage parking situation regimentally consists of two lines: line elongating current vehicle's rear direction, line elongating the central line of target position. A circular segment whose radius is the most minute rotating radius of installed steering system connects the transition part of two lines. Current semi-automatic parking avail system does not consider the multi-turn parking tribulation

situation yet. System checks whether it can track the orchestrated path afore the commencement of parking operation.

V. Wireless Access For Vehicular Environment (WAVE)

Wireless local area networks (WLAN, or Wi-Fi), researchers and developers have found their wide application perspectives in a vehicular environment, which spawns an incipient type of vehicular network: wireless access for vehicular environment (WAVE) system. The initiative of WAVE system is to establish physical platform for both perspicacious transportation systems (ITS) and vehicular infrastructure integration (VII) through providing drivers and passengers on board with authentic-time messages cognate to safety, traffic and data accommodations. To amend driving safety, WAVE system is expected to provide drivers with warning messages when possible accident subsists. To avail make WAVE an authenticity, a functional WAVE prototype is developed. Some researchers and designers have several developed dedicated short range communications (DSRC) prototypes by simply adopting subsisting indoor wireless technologies, such as Wi-Fi, to a vehicular environment. However, the expeditious fading and Doppler shift due to the mobility of transportations make WAVE signals experience doubly selective fading when propagating over transportation-to-transportation (V2V) and transportation-to-infrastructure (V2I) channels. Meanwhile, the well-studied mobile ad hoc network (MANET) medium access control (MAC) and routing protocols are proven to be inadequate or not optimized for WAVE system due to their distinctive features. For example, transportations are supposed to drive along roads in lieu of arbitrarily kinetically circulate. Both theoretical analysis and experiment have verified that Wi-Fi systems are unstable and non-optimized when applied into a vehicular environment. Industry is calling for an incipient and dedicated standard for DSRC.

A. Prototype Architecture

The design of a WAVE prototype involves in the setup of a WAVE radio and the development of a graphic utilizer interface (GUI). The WAVE radio is setup predicated on a software defined radio (SDR) development kit and a RF front end. The expeditious varying and rigorous vehicular environment establish several fresh research topics on the study of WAVE systems. The challenging topics include mobile channel modelling, the study of Doppler shift, expeditious synchronizations, expeditious channel estimate and prognostication, capacity evaluation when adopting multiple-input multiple-output (MIMO), keenly intellectual antenna and beam forming, adaptive modulation, novel network configuration (delay tolerance networks, DTN), efficacious media access control (MAC) protocols, and robust routing and congestion control schemes.

The prototype is to be with a cordial GUI with touch screen buttons and voice controlled command inputs, exhibiting authentic-time traffic distribution, image/video predicated traffic monitor and sign vigilance, weather and road hazard admonition, road and traffic accommodation, high-speed data exchange and Internet access, and even future intra-transportation broadcasting and communications. The GUI accommodates in two-fold purposes: to illustrate the designed functions and to provide probes for concrete performance or function testing. The GUI is developed in the Matlab environment, which benefits from the features of a short develop cycle, feasibility, scalability and good maintenance. Which is capable of providing following functions: warning information for

safety, authentic-time traffic, live image/video of road situations, voice and data communications, high-speed Internet access and multimedia stream for regalement. An authentic-time WAVE prototype is setup to demonstrate the defined functions of WAVE systems, which has consummated the concept proven stage for WAVE system and can be utilized the paradigm for the next step application categorical integrated circuit (ASIC) development. Besides, the prototype can be utilized as a platform for applying advance wireless technologies to WAVE systems

VI. Multivariate Spatiotemporal Model

INTELLIGENT TRANSPORTATION SYSTEM (ITS) has been transformed from a luxury option to a desideratum in most cars during recent years. Today more mainstream cars are connected directly to the cyber world than in the past, via 3G or 4G networks. This caliber of connectivity has enabled engineers to develop custom-made applications/software for each car and has unhardened numerous ITS application scenarios. As sensors become less extravagant, and both cars and infrastructure get more connected, the amount of available information increases dramatically. The information from infrastructure and cars along with the computational power which subsists in modern cars will have a great impact on the expeditious implementation of different ITS technologies.

One of the paramount and growing fields in ITS is parking avail and Parking Guidance and Information (PGI) Systems. Finding parking in sizably voluminous cities is characterized by frustration and waste of time and mazuma due to the lack of precise information about where parking spots are available at the time they are needed. Shoup states that the average probing time for parking spots in Incipient York city between 11 A.M. and 2 P.M. during a weekday is 10.6 minutes. The adventure of probing for parking in a congested environment leads to circling around blocks causing supplemental air pollution and fuel consumption and contributing to adscititious congestion. In integration, the scarcity of parking locations and the transmuting of parking rules and restrictions make parking more inelegant and arduous, especially to drivers that are not acclimated with the area. Most of the issues associated with finding parking can be solved or reduced by utilizing incipient technologies. A consequential impediment in implementing an efficacious PGI system is the entelechy of transportation to infrastructure (V2I) and infrastructure to transportation (I2V) communications. Several methods have been proposed, ranging from local radio stations to text messages and cellular networks. The utilization of these technologies alone however will not achieve the best possible result with veneration to parking availability when needed. One quandary which arises in the on-street parking guiding systems is that the authentic-time parking availability data is subsidiary only when the driver is very proximate to the parking location. The presage of parking availability at the time of advent is vigorously connected with both the traffic situation and the estimated advent times predicated on presaged traffic flows and peregrinate times. An extensive range of presage models have been studied in the literature and used to presage traffic flows in transportation networks.

The system which analyses the available information from the infrastructure ranging from authentic-time parking occupancy data to parking inhibitions/regulations and provides the driver with guidance regarding the likelihood of finding parking at the Estimated Time of Advent (ETA). This uses historical and authentic time data to make online prognostications regarding the

availability of parking in the area where the driver is orchestrating to park. First analyse the parking availability data characteristics and then, develop a model for parking availability presage. We utilize trending and detrending techniques employed in time series analysis for financial applications in order to disunite the deterministic part of data from the desultory components.

A. Modelling of Parking Utilization

The parking spot as the space in which a single vehicle can park and a parking location an area that consists of one or more parking spots. A parking location can be an off-street parking garage/lot/valet or a street block with several on-street parking spots. Parking availability data is usually aggregated and available for a parking location rather than single parking spots. We focus on parking locations instead of individual parking spots. A parking tile is defined as a group of parking locations in an area. In order to study the effects of the high/low parking demand in a certain area on the neighbouring areas, it is useful to study a group of parking locations rather than single locations. This approach also reduces the computational power needed for predictions.

B. Training of the Model

We use the historical parking information to estimate the parameters of the model. We define the residual error as the difference between the observed parking flow and the predicted flow i.e., $V_t - \hat{V}_t$. The normalized mean square residual error R for the last T time steps leading to time k is defined as

$$R(A) = \sum_{t=k-T}^k (\tilde{V}_t - \hat{V}_t)' W (\tilde{V}_t - \hat{V}_t)$$

$$= \sum_{t=k-T}^k \left(\tilde{V}_t - \sum_{m=1}^M \hat{A}_m \tilde{V}_{t-m} \right)' W \left(\tilde{V}_t - \sum_{m=1}^M \hat{A}_m \tilde{V}_{t-m} \right)$$

Where $W > 0$ is a symmetric normalizing weight matrix. In order to find the parameters of the model, we minimize the residuals with respect to $A = \{A_m ; m= 1, 2, \dots, M\}$. In the following sections, we find the parameters by implementing batch processing and recursive processing schemes. The recursive scheme is more computationally efficient than the batch (non-recursive) scheme. We first use the non-recursive scheme to generate an initial estimate of the parameters which are then used as the initial condition for the parameter estimates generated by the recursive scheme.

C. Prediction of Parking Availability

We use the developed model to predict the parking availability in a fashion which is usable in PGI systems. The parameters of the model $\Theta = [A_1 A_2 \dots A_m]$ using the real-time and historical parking availability information. Now we will use this to predict the actual availability for different parking locations. In order to make it easier for computational and scaling purposes In order to use the predictions of parking availabilities, we need to study the prediction error, and how this error affects the probability of finding parking. The probability of error is given by

$$P_i(q) \triangleq \Pr\{q - 0.5 \leq \text{err}_i < q + 0.5\}$$

$$= \frac{1}{\sqrt{2\pi\sigma_i^2}} \int_{q-0.5}^{q+0.5} \exp(-x^2/2\sigma_i^2) dx$$

The PGI system recommends the parking location with highest

probability of having at least one available spot to the user. Therefore, it is important to study the recommendation error which is defined as; I) the system predicts there would be at least one parking spot at a certain location at the estimated arrival time, however the driver finds no available spots at the arrival, or II) the system predicts no availability, and the driver finds at least one spot. In this scenario we consider the sum of these two types of error. These two cases affect the reliability of the PGI systems

VII. Conclusion

Utilizing authentic-time parking availability data in intelligent transportation systems and parking guidance systems proves to be challenging due to the fact that parking availability is astochastic process. Available parking spots might be taken by the time the driver arrives at the parking location. Ergo, the models that prognosticate parking availability at the estimated advent time will enhance the performance and amend the acceptance of intelligent systems. In this paper we surveyed the varies already proposed models for availability of car parking spots in more astronomically immense cities. These models can be habituated to sooth say the parking availability for both on-street and off-street parking locations at the estimated advent time of the driver. This is utilized to recommend the parking location with the highest probability. The results denote that the proposed model system recommends parking locations to drivers with high reliability..

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