

Transactions Letters

An Exact Markov Process for Multihop Connectivity via Intervehicle Communication on Parallel Roads

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Abstract—This paper identifies a Markov process for the multihop connectivity along two parallel roads through intervehicle communication. Vehicles are assumed to follow two Poisson processes on both roads. Exact mean, variance and probability distribution of the instantaneous propagation distance are derived. A closed form approximation for the expected distance is also proposed.

Index Terms—Network inter-vehicle communication; vehicular ad hoc networks; Markov process.

I. INTRODUCTION AND THE RESEARCH PROBLEM

Ad hoc vehicular networks (VANETs) with short range wireless communication have been under intensive study in recent years because of their promise to improve roadway mobility and safety. As locations of equipped vehicles traveling on highways exhibit high randomness, connectivity between vehicles is an important notion critical to the system design in terms of efficiency and reliability [1]–[6].

We study instantaneous information propagation through traffic along two parallel roads by developing exact models, different from those in the literature that are only on single road traffic [7]–[9] (or treat multiple roadways as one [10], [11]). Earlier, Wang *et al.* [12] study the same problem by identifying an *approximate* Bernoulli process for traffic on two roads. This study contrasts with the literature of node connectivity on mobile ad hoc networks assuming a continuous two-dimensional space ([13], [14]).

In the study problem, two sufficiently long straight roadways R_1 and R_2 of traffic are a distance d apart. Traffic densities on the two roads are λ_1 and λ_2 , respectively. Vehicles follow independent Poisson processes on both roads. Starting from a vehicle N_1 on one road, information is propagated in a direction. Vehicles on both roads within a transmission range

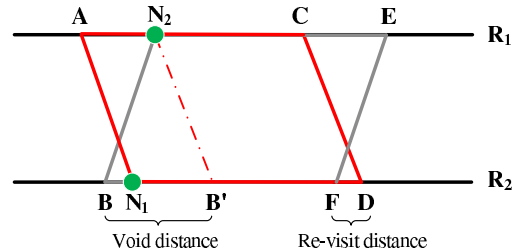


Fig. 1. Transition from type 2 to 1 region as N_2 receives from N_1 .

L are able to receive and instantly further transmit the information forward. When there is no new vehicle present within range, the propagation process terminates. The propagation distance measures from the initiating vehicle to the furthest receiving one along the direction of the two roads. The objective is to characterize the distance of information propagation as a function of the vehicle densities, road separation distance and transmission range.

The notions of node and vehicle are interchangeable here. The connectivity between vehicles is topologically measured by a transmission range. In addition, we only consider the case $0 \leq d \leq \frac{\sqrt{3}L}{2}$ because of technical tractability.

II. MODELING WITH A MARKOV PROCESS

A. Transmission Regions

We have identified a Markov process that depends on a concept of transmission region. Take an example in which a transmitter node N_1 is on road R_2 as in Figure 1. Suppose C and D are the two right most nodes that N_1 can reach on the two roads, assuming propagation goes *rightwards*. Consider a parallelogram N_1ACD where $|N_1D| = |N_1C| = |AC| = L$. We refer to N_1ACD as the *transmission region* of type 2 associated with N_1 . Similarly, a node N_2 present on road R_1 also has its transmission region N_2BFE , referred to as transmission region of type 1, as in Figure 1. The type of transmission region depends on whether the associated vehicle is present on R_1 or R_2 . An inherent characteristic is a segment AB' as in Figure 2, whose length is denoted by r , where $\frac{r}{2} = L - \sqrt{L^2 - d^2}$.

The case $d \leq \frac{\sqrt{3}L}{2}$ guarantees that an entire transmission region be within the transmission range of the according vehicle. If no vehicle is present in the transmission region,

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satisfies the following relationship.

$$\begin{aligned}
D_i(v, u) = & \int_0^L \lambda_i e^{-t\lambda_i - [t-v]_+ \lambda_j} \left(t + D'_i([v-t]_+, [u-t]_+) \right) dt \\
& + \int_v^L \lambda_j e^{-(t-v)\lambda_j - t\lambda_i} \left(t - \frac{r}{2} + D'_j(r, [r-t]_+) \right) dt \\
& + \int_L^{L+u} \lambda_j e^{-(t-v)\lambda_j - L\lambda_i} \left(t - \frac{r}{2} + D'_j(r - (t-L), 0) \right) dt \\
& + e^{-L\lambda_i - (L+u-v)\lambda_j} \left[u - \frac{r}{2} \right]_+, \quad (1)
\end{aligned}$$

where $D_i(v, u)$ and $D'_i(v, u)$ are i.i.d. for $i = 1, 2$; $[x]_+ = \max\{0, x\}$. Replacing each $D_i(\cdot, \cdot)$ and $D'_i(\cdot, \cdot)$ with the according $d_i(\cdot, \cdot)$ gives the expected distances, $i = 1, 2$.

$$\begin{aligned}
V_i(v, u) = & \int_0^L \lambda_i e^{-t\lambda_i - [t-v]_+ \lambda_j} [(t + d_i([v-t]_+, [u-t]_+) - d_i(v, u))^2 \\
& + V_i([v-t]_+, [u-t]_+)] dt \\
& + \int_v^L \lambda_j e^{-(t-v)\lambda_j - t\lambda_i} [(t - \frac{r}{2} + d_j(r, [r-t]_+) - d_i(v, u))^2 \\
& + V_j(r, [r-t]_+)] dt \\
& + \int_L^{L+u} \lambda_j e^{-(t-v)\lambda_j - L\lambda_i} [(t - \frac{r}{2} + d_j(r - (t-L), 0) \\
& - d_i(v, u))^2 + V_j(r - (t-L), 0)] dt \\
& + e^{-L\lambda_i - (L+u-v)\lambda_j} \left(\left[u - \frac{r}{2} \right]_+ - d_i(v, u) \right)^2. \quad (2)
\end{aligned}$$

where $(i, j) = (1, 2)$ or $(2, 1)$ (End of Proposition 2).

The variance $V_i(v, u)$ is easily derived by using the Law of Total Variance and by conditioning on the *first* vehicle location. In addition, Proposition 1 applies in Proposition 2 above.

We denote by $P_i(x, v, u)$ the probability of propagation beyond a horizontal distance x starting with a state $S_i(v, u)$. Regarding probability distribution of propagation distance, we first define a function $P_i(x, v, u) = 1$, if $x \leq \max\{0, u - \frac{r}{2}\}$, $i = 1, 2$. Detailed rationale is provided in [15].

Proposition 3: The propagation probability satisfies the following with $(i, j) = (1, 2)$ or $(2, 1)$.

$$\begin{aligned}
P_i(x, v, u) = & \int_0^L \lambda_i e^{-t\lambda_i - [t-v]_+ \lambda_j} P_i(x - t, [v-t]_+, [u-t]_+) dt \\
& + \int_v^L \lambda_j e^{-(t-v)\lambda_j - t\lambda_i} P_j(x - t + \frac{r}{2}, r, [r-t]_+) dt \quad (3) \\
& + \int_L^{L+u} \lambda_j e^{-(t-v)\lambda_j - L\lambda_i} P_j(x - t + \frac{r}{2}, r - t + L, 0) dt.
\end{aligned}$$

Our additional numerical tests in [15] show a clear Gamma type of curve for $P_i(x, v, u)$ when the traffic densities are high.

III. CLOSED FORM APPROXIMATION WITH A SINGLE ROAD

A meaningful question is whether there is a way to consolidate the two road densities to use as on a single road for study of propagation distance. Based on a rough observation that road R_1 of a length $L - \frac{r}{2}$ is horizontally

right of the transmitting vehicle on road R_2 in a Type 2 region (refer to Figure 2), we propose the following formula when consolidating the two densities, which is equivalent to evening out the traffic in $L - \frac{r}{2}$ on road R_1 onto a longer distance L on R_2 , $\lambda = \lambda_1 + \lambda_2 - \alpha \cdot \frac{r}{2L} \cdot \lambda_1$, where $\lambda_2 \geq \lambda_1$. The value λ is then used in the formulas for a single road as in [9]: Expectation = $\lambda^{-1}(e^{\lambda L} - L\lambda - 1)$ and Variance = $\lambda^{-2}(e^{2\lambda L} - 2L\lambda e^{\lambda L} - 1)$.

Typically $\alpha \in [1.0, 2.0]$. Our test shows that $\alpha = 1.0$ leads to overestimates while $\alpha = 2.0$ leads to underestimates of the expectation. Therefore, we propose $\alpha = 1.5$. As special cases, when the road separation distance d is zero, the consolidated density becomes the sum of the two.

IV. NUMERICAL TESTS

We solve the integral equations by simply discretizing the integrations and solving the resultant arrays of equations. Because r is not a large value, the number of states (v, u) resulting from discretization is not too large. Analytical results are validated by simulation. Each simulation generates a snapshot of traffic and measures the propagation distance. The vertical bars in Figure 3 and 4 are one standard deviation about the corresponding mean expectation and mean variance based on 20 runs, each conducting 2000 simulations and giving an ‘expectation’ and a ‘variance’.

In the numerical tests, we scale the transmission range to be a standard unit 1.0. All other measures including the node density are scaled accordingly. In light of [16], an average density for freeway at level of service A (a stable density) is about 5.5 vehicles per mile per lane (vpml). When a percentage of equipped vehicles (also referred to as market penetration rate) is applied, a significantly lower equipped vehicle density could result. Additionally, we set R_2 to have the higher traffic density.

Effect of Road Separation on Propagation Distance As in Figure 3, road separation decreases the propagation distance at an increasing rate, especially at large traffic densities. When the road separation distance increases from zero to about 0.4 at low traffic density, the decrease of both expectation and variance is insignificant.

Effect of Traffic Density on Propagation Distance Figure 4(a) and 4(b) indicate that the expected propagation distance increases at a slightly increasingly rate with traffic density and that the variance of propagation distance increases with traffic density at a larger rate than the expected distance.

Performance of Approximation In terms of both the expected distance and the variance, the approximation agrees exceptionally well with the exact solution when the vehicle density is relatively low and when the road separation is small as in Figure 3.

V. CONCLUSION

Interactions of nodes on the two roads clearly undermine the applicability of the connectivity models developed for a single road. We explicitly consider road separation in this paper and identify an inherent Markov process that equivalently determines the propagation distance. Exact models are proposed for the information propagation distance. This research measures

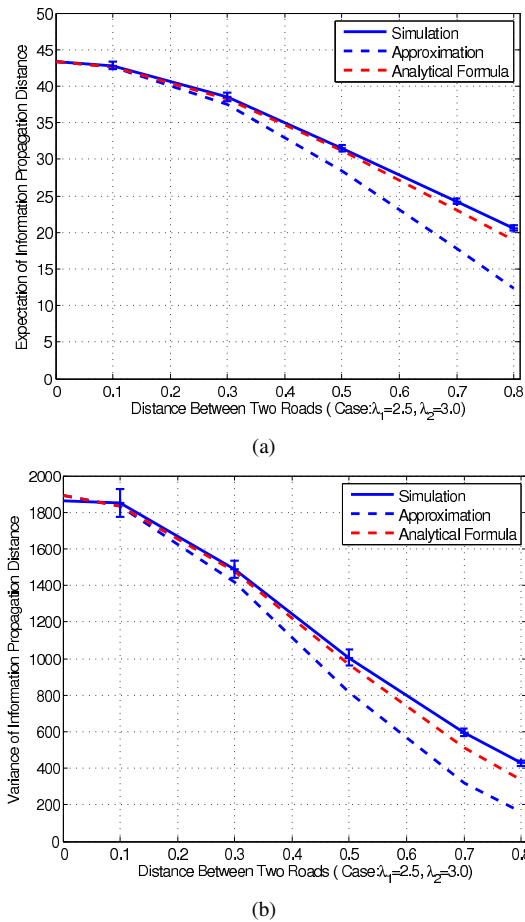


Fig. 3. Propagation distance at $\lambda_1 = 2.5$, $\lambda_2 = 3.0$.

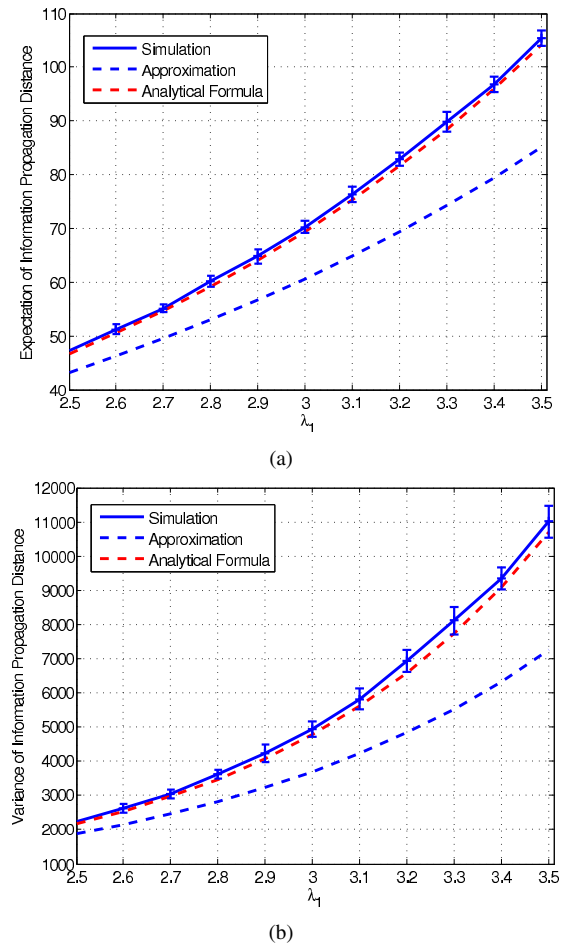


Fig. 4. Effect of traffic density at $d = 0.5$ and $\lambda_2 = 3.5$.

the topological connectivity of vehicles through a transmission range, which provides an upper bound to cases that consider factors of channel conflicts, signal fading, etc. These models may serve as a framework for future efforts to consider those other factors.

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