Stability conditions for a discrete-time decentralised medium access algorithm

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Abstract

We consider a stochastic queueing system modelling the behaviour of a wireless network with nodes employing a discrete-time version of the standard decentralised medium access algorithm. The system is unsaturated – each node receives an exogenous flow of packets at the rate $\lambda$ packets per time slot. Each packet takes one slot to transmit, but neighboring nodes cannot transmit simultaneously. The algorithm we study is standard in that: a node with empty queue does not compete for medium access; the access procedure by a node does not depend on its queue length, as long as it is non-zero. Two system topologies are considered, with nodes arranged in a circle and in a line. We prove that, for either topology, the system is stochastically stable under condition $\lambda < \frac{2}{5}$. This result is intuitive for the circle topology as the throughput each node receives in a saturated system (with infinite queues) is equal to the so called parking constant, which is larger than $\frac{2}{5}$. (The latter fact, however, does not help to prove our result.) The result is not intuitive at all for the line topology as in a saturated system some nodes receive a throughput lower than $\frac{2}{5}$.

1 Introduction

In this paper we consider a stochastic queueing model, primarily motivated by the MAC (Medium Access Control) algorithms in wireless networks.

Wireless devices are an indispensable part of everyday life for most of us. Wireless networks are used to connect devices at home and at businesses, to connect cell phones, sensors and even satellites. The sizes of wireless networks are growing at astonishing speeds, and with the development of Internet of Things will only continue to grow.

Wireless transmissions inevitably create noise and if many devices transmit simultaneously, the packets may be lost. There is therefore a need for a MAC algorithm that will control which devices are going to transmit at a given time. In wireless networks there is a natural notion of neighbouring devices meaning that the transmissions generated by these devices interfere with each other. The model in this paper corresponds to those MAC protocols that prevent neighbours from transmitting simultaneously, thus making any collision and loss of packets impossible.

A wireless network is called single-hop, if each generated packet only needs to be transmitted once, by its source node (where the packet was generated). A multi-hop network is more general – each generated packet has a source and a destination node and needs to find its way from the source to the destination, potentially passing a number of intermediate nodes on its way.

In this paper we are interested in stability of a network, i.e. the ability of the nodes in the network to transmit all generated packets, without the packet queues building up to infinity. If an access algorithm guarantees stability for all packet arrival processes for which stability is in principle possible, such an algorithm is called maximally stable (or throughput-optimal).

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There are known maximally stable algorithms, which utilise queue lengths (or packet delays) information while making access decisions. The celebrated MaxWeight (or, BackPressure) algorithms have been introduced in [24] and have been shown to be maximally stable for both single-hop and multi-hop networks. They have later been extended and generalised in a number of directions. Maximum stability is also known to be achieved in the single-hop case by a class of α-fair algorithms (see [14, 15, 17] for introduction of the fair-allocation concepts and [3, 10] for stability proofs). All these algorithms are centralised, in that they need a central controller to know the states of the queues of all nodes and then to solve an optimisation problem to make every access control decision. Having such a central entity (and its ability to solve an optimisation problem for a vast number of devices) is not feasible in large networks. There is therefore a need for the so-called decentralised algorithms where each node regulates its own access to the medium.

A well-known and widely used (most notably in IEEE 802.11) example of a decentralised MAC algorithm preventing collisions is the so-called CSMA (Carrier Sense Multiple Access) protocol, where each node has a random back-off time before attempting to access the channel and if, on expiration of the back-off time, it does not sense any other node transmitting, the node will start its own transmission. Let us call this type of multiple access a standard CSMA, where ‘standard’ refers to the following properties: (a) each node does not know (and does not try to explicitly learn) its neighbours, (b) the access procedure is same regardless of the node queue length, as long as it is non-zero, and (c) the node does not access the channel when its queue length is zero (i.e., no packets to transmit).

Most results on networks governed by the standard CSMA protocol assume that the system is saturated, i.e. each node is assumed to always have a packet to transmit (or, has infinite queue), and the question of interest is the throughput of the system, or an individual node. The typical assumption is that the service times, as well as back-off times, are Exponentially distributed, and the state of the system is fully described by the activity process tracking which nodes are currently transmitting. This Markov process turns out to be reversible and it thus has a product-form stationary distribution, providing values of throughputs of individual nodes (see [4, 11, 26]).

In practice the nodes generate packets at random time instances and thus do not always compete for medium access. This calls for the analysis of unsaturated systems, modelled by each node having its own exogenous arrivals of generated packets. Such unsaturated systems may not be approximated by saturated ones, or even bounded by them. For example, if one exchanges an unsaturated node for a saturated one, this will be detrimental to the performance of the node’s neighbours but beneficial for the neighbours’ neighbours (with whom the original node does not interfere). In other words, the random process describing such a system does not have the monotonicity property, which, informally speaking, means that two versions of the process, with initial state of one dominating that of the other, can be coupled so that this dominance relation persists at all times. The absence of monotonicity substantially complicates the process analysis, including establishing stability conditions. (The only case where the monotonicity for the CSMA model holds is when the interference structure is given by a complete graph, so that at most one transmission is possible at a time.) It is also impractical in applications to keep a device accessing the network if it does not have packets to transmit. (Hence property (c) of a standard CSMA is important.) The analysis of general unsaturated networks is extremely difficult, because, primarily due to lack of monotonicity, the queue dynamics and transmission schedule of any node depends on the states of all other nodes, across the entire network, in a very complicated manner.

Very few results are known for unsaturated systems ([7, 25]). In [25] the authors consider a continuous-time model and study the question of stability. They show that, unsurprisingly due to the monotonicity we mentioned in the previous paragraph, stability of an unsaturated system with a complete interference graph is achieved when the arrival intensity for each node
is smaller than the throughput of the node in the saturated system. They also demonstrate with an example that for a more general interference graph this condition is not sufficient. This means, in particular, that standard CSMA does not achieve maximum stability.

The absence of maximal stability of the standard CSMA protocol led to the development, starting with [13], of the queue-based algorithms where the distribution of the back-off duration of a node depends on the state of the queue of the node. In particular, an algorithm of this type was proposed in [19] for a continuous-time model, which guarantees maximal stability for single-hop systems on any graph. However, despite providing maximal stability, the queue-based algorithms are difficult to implement and are known to lead to high delays. (Hence, property (b) of a standard CSMA is important).

The analysis of multi-hop systems governed by CSMA protocols is even more difficult than that of single-hop systems. In multi-hop networks a packet needs to reach its destination node and may need to be transmitted by several nodes in the network, in which case, upon successful transmission by one node it may enter the queue of another node, thus complicating the queue dynamics further. There are no known results on stability of such networks even assuming a single source-destination pair and assuming that all packets have to take the same path through a sequence of nodes, making the graph a line segment. Studying the end-to-end throughput of such a network is also extremely difficult. Experiments show that the throughput may decrease when the arrival rate increases beyond the stability threshold (see [1]), a phenomenon that has not been explained rigorously. To the best of our knowledge, the only rigorous results on multi-hop networks are presented in [20] where the authors calculate the end-to-end throughput of a 3-node network in full saturation, i.e. when the source node is assumed to always have packets to transmit.

In this paper, our goal is to characterise the performance of a standard CSMA algorithm. We would like to stress again that standard CSMA algorithms are important, because they are decentralised (and therefore easy to implement), because implementing a queue-based scheme leads to long delays, and because it is impractical in the unsaturated situation to keep all nodes active at all times. We will focus on two simple single-hop systems, consisting of \( N \) nodes, with the nodes arranged in a circle or in a line. The systems operate in discrete time. To model standard CSMA in discrete, slotted time, we will assume that at the beginning of each time slot the nodes are given access priorities, forming a permutation of numbers 1, \ldots, \( N \), picked independently (across time slots), uniformly at random from all possible permutations. The node with the highest priority transmits in this slot if its queue is non-empty. Node with the second-highest priority transmits in the slot if its queue is non-empty and if none of its neighbour nodes is transmitting. And so on until all nodes are checked in their priority order. All transmission times are equal to 1, so at the beginning of the next time slot no transmission is ongoing, and the medium access process is repeated independently. Note that this algorithm is easy to implement in a decentralised fashion, with an arbitrarily small loss in throughput (we provide a precise algorithm in the model definition).

Our main result concerns the homogeneous system, where the arrival rates into each node are equal. Non-homogeneous systems are notoriously difficult and even the simpler random access algorithm ALOHA does not allow for the exact description of the stability region (see, e.g., [2, 3, 4, 23]).

The maximum stability for a standard CSMA system on a circle or on a line is given by \( \lambda < \lfloor N/2 \rfloor /N \), where \( \lambda \) is the arrival rate into each node, i.e. the average number of new generated packets per unit of time. (If \( N \) is even, the condition is simply \( \lambda < 1/2 \).) This is not achievable by the CSMA protocol on a circle. Indeed, if all nodes have non-empty queues, then the process of transmission scheduling by the nodes in a time slot is equivalent to the discrete parking problem (see, e.g., [12, 16]), and there are known expressions for the expected total number of transmissions per time slot (see [21] for expressions adopted to CSMA setting). On a circle, if all nodes have non-empty queues, then by symmetry the expected number of
transmissions per slot is the same for all nodes and is equal to the so-called parking constant (sometimes referred to as jamming density), which we denote by $c_p = c_p(N)$. Therefore, if $\lambda$ exceeds $c_p$, the system is unstable. The parking constant $c_p$ for a circle of $N$ nodes is equal to $1/2$ if $N = 4$, $2/5$ if $N = 5$ and it decreases over even values of $N$ and increases over odd values of $N$ to the same limit $1/2(1 - e^{-2}) \approx 0.4323$.

On the other hand, each non-empty node always transmits if its priority is higher than the priorities of both its neighbours, which happens with probability $1/3$. Therefore $\lambda < 1/3$ trivially leads to stability. One of the goals of our work is to study how close the stability region of the standard CSMA algorithm on a circle is to $\lambda < c_p$ (the best achievable for this algorithm).

The main result of the paper for the circle topology is that, under the standard CSMA, $\lambda < 2/5$ leads to stability for a system with $N \geq 4$. We conjecture that, in fact, $\lambda < c_p$ is sufficient for stability, but this is not proved.

It appears to be a lot more difficult to conjecture the stability condition for a system on a line. If all nodes have non-empty queues, then the process of transmission scheduling in a slot is again equivalent to the discrete parking problem, but in this situation the expected number of transmissions differs from node to node. For instance, the nodes on the edges will have transmission probabilities larger than $1/2$ and tending to $1 - e^{-1} \approx 0.6321$ as $N \to \infty$, but the nodes right next to them will have much lower transmission probabilities tending to $e^{-1} \approx 0.3679$ (see [21]).

Our main result for the line topology is that, just like for the circle system, $\lambda < 2/5$ leads to stability for a system with $N \geq 4$. This result is not intuitive at all, given that in the saturated system the transmission probability of the second node from the edge is strictly less than $2/5$ (see above). The result also stresses once again that the saturated system does not provide a bound for the unsaturated one, that it is unreasonable to make all nodes (including empty) to compete for the transmission at all times, and that it is important to study unsaturated systems.

Our main results are proved using the fluid limit technique. The main difficulty in the proofs consists in dealing with situations when a number of queues in the fluid limit are at zero. One has to study rather complicated structure of the occupancy and activation processes of the neighbourhood of non-zero queues. We believe that the technique we developed may be used to study the behaviour of CSMA and similar decentralised algorithms on more complicated topologies.

We also provide a relatively short direct proof of stability for both topologies under condition $\lambda < 3/8$. This is weaker than our main results but we find that the proof provides a good insight into the behaviour of the model. Note that the lack of efficiency of the random-access algorithm compared to the ideal (centralised) algorithm in the system on a circle or a line comes from the fact that at each time slot a number of pairs of neighbouring nodes will not transmit. Therefore we consider the Lyapunov function $\sum_i (Q_i + Q_{i+1})^2$, where $Q_i$ is the queue length at node $i$. The analysis of the drift of this function in one step leads to stability provided $\lambda < 3/8$. Another advantage of this approach is that it provides some sufficient conditions for stability of inhomogeneous systems (with different arrival rates at the nodes). This is not captured by our fluid-limit proof of the main results.

The paper is structured as follows. We introduce the model and notation in Section 2. In Section 3 we formulate and prove auxiliary results and formulate our main results. Section 4 is devoted to the proof of the main results. We provide some conjectures and discuss our further research plans in Section 5.

2 Model and notation

Assume that we have $N$ nodes (transmitters) on a circle or on a line, and for any node we are going to say that the nodes to its immediate left and right are its neighbours. More formally, if
we consider the circle topology, we will denote the neighbourhood of node $i$ as

$$\mathcal{N}_c(i) = \begin{cases} \{i-1, i+1\} & \text{for } i = 2, \ldots, N-1, \\ \{N, 2\} & \text{for } i = 1, \\ \{N-1, 1\} & \text{for } i = N, \end{cases}$$

and if we consider the line topology, we will denote the neighbourhood of node $i$ as

$$\mathcal{N}_l(i) = \begin{cases} \{i-1, i+1\} & \text{for } i = 2, \ldots, N-1, \\ \{2\} & \text{for } i = 1, \\ \{N-1\} & \text{for } i = N. \end{cases}$$

Assume that each node has unlimited buffer space to queue packets waiting for transmission. Time is slotted. At the beginning of each time slot, first new transmissions are initiated, and then the arrivals of new packets (messages) occur. Every message transmission time is equal to 1, so any transmission is completed before the beginning of the next time slot.

Denote by $Q_i(n)$ the queue length at node $i$ at time $n$. Denote by $\xi_i(n)$ the number of new arrivals into node $i$ at time $n$. We assume that $\xi_i(n)$ are all i.i.d. with $\mathbb{E}\xi_i(n) = \lambda$.

We will assume that the neighbours cannot transmit in the same time slot. The competition for medium access is as follows. At the beginning of each time slot $n$, an allocation of priorities \{U_1(n), \ldots, U_N(n)\} for nodes is chosen uniformly at random from all possible permutations of the set \{1, \ldots, N\}. The node with the highest priority 1 will transmit if its queue is not empty. The further procedure is defined by induction: the node with the next-highest priority will transmit if its queue is not empty and if no node in its neighbourhood is already transmitting. At each time slot this procedure is applied until the set of all nodes is exhausted, and this procedure is repeated independently at each time slot. Upon completion of its transmission, the packet leaves the system.

Denote by $D_i(n)$ the number of packets transmitted by node $i$ in time slot $n$. Note that $D_i(n)$ can only take values 0 and 1. Then the evolution of queue $i$ length is as follows:

$$Q_i(n+1) = Q_i(n) - D_i(n) + \xi_i(n).$$

Remark 1. Note that the discrete time medium access procedure described above may be implemented in a decentralised way with a negligible loss in efficiency. Indeed, assume that at the beginning of each time slot $n$, every node with a non-empty queue (every node $i$ such that $Q_i(n) > 0$) generates a random variable $W_i(n)$ uniformly in $(0, \varepsilon)$. We assume that the random variables \{W_i(n)\} are independent in $i$ and $n$. These random variables serve as back-off times for transmitters: node $i$ will wait exactly $W_i(n)$ and at that time, if no node in its neighbourhood is transmitting, will start its transmission. If at least one of the nodes in the neighbourhood of node $i$ is transmitting at time $n + W_i(n)$, then node $i$ stays silent for the duration of the time slot. As we can see, only an $\varepsilon$ proportion of each time slot needs to be devoted to this competition, and $\varepsilon$ may be made arbitrarily small.

Some basic notation we use throughout the paper: the norm of a vector $q = (q_1, \ldots, q_N)$ is $\|q\| = \sum_i |q_i|$; $[x]$ is the largest integer not greater than $x$.

3 Results

In this section we present the results of our paper. We start (in Section 3.1) by formulating and proving some properties of expected numbers of transmissions on intervals which are used in our proofs and which are also interesting in their own right. We then proceed (in Section 3.2) to prove maximum stability achievable by our algorithm (standard CSMA), for a system
where nodes change their locations at random at the beginning of every time slot. In Section 3.3 we turn our attention back to the original system and prove stability for both circle and line topologies in the case $\lambda < 3/8$ using a one-step Lyapunov function. Finally, in Section 3.4 we present our main results stating that the original system for both circle and line topologies is stable provided $\lambda < 2/5$. The proof of the main results is based on fluid limits and is given in Section 4.

### 3.1 Some properties of parking constants for a segment

Denote by $L_k$ the expected number of departures from $k$ non-empty nodes on a line segment. Note again that the transmission initiation process on a set of non-empty nodes is equivalent to the discrete-time parking problem, therefore $L_k$ may be thought of as the expected number of cars parked on the segment, and $L_k/k$ as the parking constant, or jamming density. We note here that the original papers (see, e.g. [12][10]) on the parking problem are devoted to the expected amount of cites occupied by the parked cars (and each car needs two cites), which is closely related to but slightly different from the number of expected transmissions in our case. Formulas adapted to the CSMA situation are provided in [21] and we are going to refer to this paper hereinafter.

**Lemma 2.** $L_n/n$ is non-increasing for $n \geq 3$.

**Proof.** We need to show that $L_n/n \leq L_{n-1}/(n-1)$ for all $n \geq 3$, which is equivalent to $n(L_n - L_{n-1}) - L_n \leq 0$ for $n \geq 4$. Note that, using the expression for $L_n$ from [21] Corollary 2,

$$L_n - L_{n-1} = \sum_{k=1}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!} (n - k + 1) - \sum_{k=1}^{n-1} (-1)^{k+1} \frac{2^{k-1}}{k!} (n - k)$$

$$= \sum_{k=1}^{n-1} (-1)^{k+1} \frac{2^{k-1}}{k!} + (-1)^{n+1} \frac{2^{n-1}}{n!} = \sum_{k=1}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!},$$

and hence

$$n(L_n - L_{n-1}) - L_n = n \sum_{k=1}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!} - \sum_{k=1}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!} (n - k + 1)$$

$$= \sum_{k=1}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!} (k - 1) = \sum_{k=2}^{n} (-1)^{k+1} \frac{2^{k-1}}{k!} (k - 1) = \sum_{k=2}^{n} (-1)^{k+1} c_k,$$

where we denoted $c_k = \frac{2^{k-1}}{k!} (k - 1)$. It may be checked directly that the sum on the RHS of the last expression is negative for $n = 4$ and $n = 5$. Also, for any $n = 2m$, the sum is smaller than the sum for $n = 2m - 1$, so it is sufficient to consider odd values of $n$. And the sums for odd values of $n \geq 5$ are negative as $c_{2m+1} \leq c_{2m}$ for all $m \geq 3$. 

**Lemma 3.** Denote by $L_{k,m}$ the expected number of departures from $k$ first nodes in a line of $m$ nodes. Then

$$L_{k,m} \leq L_k$$

for any $k \leq m$.

The statement of the lemma may be described in words as "by adding interference, we reduce the expected number of departures".

**Proof.** We use induction. The statement is trivial for $m = 1$. Assume that the statement is correct for all $m \leq M - 1$ and write

$$L_{k,M} = \sum_{i=1}^{k} \frac{1}{M} (1 + L_{i-2} + L_{k-i-1;M-i-1}) + \frac{1}{M} L_{k-1} + \sum_{i=k+2}^{M} \frac{1}{M} L_{k;i-2},$$

6
with appropriate conventions.

We can also write

\[ L_k = \sum_{i=1}^{k} \frac{1}{M}(1 + L_{i-2} + L_{k-i-1}) + \sum_{i=k+1}^{M} \frac{1}{M}L_k, \]

by adding \( M - k \) dummy nodes to the right of node \( k \). The statement is then easily seen by comparing the two formulas.

Formula (1) immediately implies

\[ L_{k+m} \leq L_k + L_m \quad (2) \]

for any \( k \) and \( m \). This can be described as "by stacking two non-empty pieces together, we make things worse in terms of average number of departures". Or, again, "added interference reduces expected number of departures".

### 3.2 Stability of a system with reshuffling

Using Lemma 2, we easily obtain that condition \( \lambda < \min\{L_N/N, 1/2\} \) implies \( \lambda < L_k/k \) for all \( k \leq N \).

Denote by \( C_k \) the expected number of departures from \( k \) non-empty nodes on a circle. For all \( k \geq 1 \), \( C_k \leq L_k \). Indeed, this is trivial for \( k \leq 2 \). For \( k \geq 3 \), obviously, \( C_k = L_{k-3} + 1 \), and using (2) it is easy to check that \( L_{k-3} + 1 \leq L_k \). Using \( C_N \leq L_N \), we can easily check that \( \lambda < C_N/N \) implies \( \lambda < \min\{L_N/N, 1/2\} \), and then implies \( \lambda < L_k/k \) for all \( k < N \).

This means that, under condition \( \lambda < \min\{L_N/N, 1/2\} \) for the circle topology, or \( \lambda < C_N/N \) for the line topology, on any non-empty segment of any length (including all nodes being non-empty), the expected number of arrivals in a time slot is smaller than the expected number of departures.

For either topology, define the following system, which we will refer to as the system with reshuffling. In fact, we define three versions of it; in all of them arrivals happen according to the same rule as before. Version 1: at the beginning of each slot, all system queues are reshuffled uniformly at random; i.e. queue locations are determined by a permutation of their indices, chosen uniformly at random. Version 2: at the beginning of each slot, all empty queues stay where they are, but all non-empty queues are reshuffled uniformly at random. Version 3: at the beginning of each slot, all empty queues stay where they are, but non-empty queues are reshuffled uniformly at random within each non-empty segment.

Given the property of non-empty segments derived above, we have the following result.

**Theorem 4.** For the circle topology, the system with reshuffling (any of the three versions) is stable if \( \lambda < C_N/N \). For the line topology, the system with reshuffling (any of the three versions) is stable if \( \lambda < \min\{L_N/N, 1/2\} \).

### 3.3 Stability in the case \( \lambda < 3/8 \)

In this section we prove that stability of the original system in either topology is guaranteed if \( \lambda < 3/8 \). This is weaker than our main results (also, the following theorem requires the second moment of the arrival process to be finite, which is not needed in our main result).

However, we think that the proof sheds some light on the behaviour of the system as it is based on the Lyapunov function that penalises the cases when pairs of neighbouring nodes do not transmit - exactly what makes a random access inefficient on a circle or a line segment. The following theorem also has a simple generalisation leading to sufficient stability conditions for inhomogeneous arrival processes - something our main result does not allow.
**Theorem 5.** Let $\xi$ be a random variable equal in distribution to $\xi_i(n)$ for all $i$ and $n$. If $\lambda < 3/8$ and $\mathbb{E}(\xi^2) < \infty$, then the system in either topology is stable.

**Proof.** We start with a technical lemma.

**Lemma 6.** Consider nodes $k$ and $k + 1$ such that $Q_k > 0$ and $Q_{k+1} > 0$. Let $l = \max\{i \geq 0 : Q_{k-i} > 0\}$ and $r = \max\{i \geq 0 : Q_{k+1+r} > 0\}$. For all $l$ and $r$

$$\mathbb{E}(D_k + D_{k+1}) \geq 3/4.$$  

**Proof of Lemma 6.** First note that $\mathbb{E}(D_k + D_{k+1}) = 1$ if either $l = 0$ or $r = 0$ (as out of two nodes at the edge of a line of non-empty nodes, exactly one will always transmit). It can also be calculated directly that $\mathbb{E}(D_k + D_{k+1}) = 3/4$ when $l = 1$ and $r = 1$.

The rest of the proof is based on induction. Denote by $D(l, r) = \mathbb{E}(D_k + D_{k+1})$ for given values of $l$ and $r$. Assume that the statement holds for all $l$ and $r$ such that $l \geq 1$, $r \geq 1$ and $l + r \leq K$ and let us look at $l$ and $r$ such that $l \geq 1$, $r \geq 1$ and $l + r \leq K + 1$. Conditioning on the node that has the highest priority, write

$$D(l, r) = \sum_{i=k-l}^{k-2} \frac{1}{l + r + 2} D(2 - i - k, r) + \frac{1}{l + r + 2} p_1(r + 1) + \frac{2}{l + r + 2}$$

$$+ \frac{1}{l + r + 2} p_1(l + 1) + \sum_{i=k+1}^{k+r+1} \frac{1}{l + r + 2} D(l, i - k - 3),$$

where by $p_1(n)$ we denote the probability that the first node in a line of $n$ non-empty nodes transmits. It is straightforward to note (see also Lemma 19) that $p_1(n) \geq 1/2$ for any $n$. Using this and the induction assumption, we can conclude that

$$D(l, r) \geq 3 \frac{l - 1}{4(l + r + 2)} + 3 \frac{r - 1}{4(l + r + 2)} = 3/4.$$  

\[ \square \]

Let us now prove theorem 5. The proof uses the standard Foster-Lyapunov technique. We start with the circle topology. Consider a function $L : \mathbb{R}_+^N \to \mathbb{R}_+$ given by

$$L(x) = \sum_{i=1}^{N} (x_i + x_{i+1})^2,$$

with the convention $x_{N+1} = x_1$.

It is sufficient to show that if $\lambda < 3/8$, then there exist $K < \infty$ and $\varepsilon > 0$ such that

$$\mathbb{E} (L(Q(1)) - L(Q(0)) | Q(0) = Q) < -\varepsilon$$  

for all initial conditions $Q(0) = (Q_1(0), \ldots, Q_N(0)) = (Q_1, \ldots, Q_N) = Q$ with $\sum_{i=1}^{N} Q_i \geq K$.

To show this, write

$$\mathbb{E} (L(Q(1)) - L(Q(0)) | Q(0) = Q) = \sum_{i=1}^{N} \mathbb{E} ((Q_i + Q_{i+1} + \xi_i + \xi_{i+1} - D_i - D_{i+1})^2 - (Q_i + Q_{i+1})^2)$$

$$\leq 2 \sum_{i=1}^{N} (Q_i + Q_{i+1})(2\lambda - \mathbb{E}(D_i) - \mathbb{E}(D_{i+1})) + \sum_{i=1}^{N} \mathbb{E}(\xi_i + \xi_{i+1})^2 + \sum_{i=1}^{N} \mathbb{E}(D_i + D_{i+1})^2,$$  

(4)
where for ease of notation we wrote $\xi_k$ for $\xi_k(0)$ and $D_k$ for $D_k(0)$. Note that in the last expression
\[
\sum_{i=1}^{N} \mathbb{E}(\xi_i + \xi_{i+1})^2 + \sum_{i=1}^{N} \mathbb{E}(D_i + D_{i+1})^2 \leq 2 \sum_{i=1}^{N} \mathbb{E}(\xi_i^2) + 2N\lambda^2 + 4N.
\]
Denote the constant on the RHS of the last expression by $C < \infty$. Let us now bound the first sum on the RHS of \((4)\). In order to do this, rewrite it as
\[
\sum_{i=1}^{N} (Q_i + Q_{i+1})(2\lambda - \mathbb{E}(D_i) - \mathbb{E}(D_{i+1})) = \sum_{i=1}^{N} Q_i(4\lambda - \mathbb{E}(D_{i-1}) - 2\mathbb{E}(D_i) - \mathbb{E}(D_{i+1})).
\]
Now we consider all possible options for the states of the nearest neighbours of a non-empty node $i$.

If $Q_{i-1} = Q_{i+1} = 0$, then
\[
\mathbb{E}(D_{i-1}) + 2\mathbb{E}(D_i) + \mathbb{E}(D_{i+1}) = 2.
\]
If $Q_{i-1} = 1$ and $Q_i = 0$ (or the other way around), then
\[
\mathbb{E}(D_{i-1}) + 2\mathbb{E}(D_i) + \mathbb{E}(D_{i+1}) = \mathbb{E}(D_{i-1}) + 2\mathbb{E}(D_i) = (\mathbb{E}(D_{i-1}) + \mathbb{E}(D_i)) + \mathbb{E}(D_i) = 1 + \mathbb{E}(D_i) \geq 3/2,
\]
as in this case exactly one of nodes $i-1$ and $i$ is transmitting, and with a probability at least $1/2$ it is node $i$.

Finally, if $Q_{i-1} = Q_{i+1} = 1$, then
\[
\mathbb{E}(D_{i-1}) + 2\mathbb{E}(D_i) + \mathbb{E}(D_{i+1}) = (\mathbb{E}(D_{i-1}) + \mathbb{E}(D_i)) + (\mathbb{E}(D_i) + \mathbb{E}(D_{i+1})) \geq 3/2,
\]
due to Lemma \[4\]
Combining these estimates, we can write
\[
\sum_{i=1}^{N} Q_i(4\lambda - \mathbb{E}(D_{i-1}) - 2\mathbb{E}(D_i) - \mathbb{E}(D_{i+1})) \leq (4\lambda - 3/2) \sum_{i=1}^{N} Q_i,
\]
and if $\lambda < 3/8$, then the RHS of \((4)\) is does not exceed $-\varepsilon$, provided $\sum_{i=1}^{N} Q_i \geq K = (C + \varepsilon)/(2(3/2 - 4\lambda))$. This completes our proof for the circle topology.

The proof of Theorem \[5\] for the line topology follows the same lines and is based on the Lyapunov function
\[
L(\vec{x}) = \sum_{i=1}^{N-1} (x_i + x_{i+1})^2.
\]
Its expected drift in one step may be bounded from above by
\[
Q_1(2\lambda - \mathbb{E}(D_1) - \mathbb{E}(D_2)) + \sum_{i=2}^{N-1} Q_i(4\lambda - \mathbb{E}(D_{i-1}) - 2\mathbb{E}(D_i) - \mathbb{E}(D_{i+1})) + Q_N(2\lambda - \mathbb{E}(D_{N-1}) - \mathbb{E}(D_N)) + C.
\]
Now if $Q_1 > 0$, then $\mathbb{E}(D_1) + \mathbb{E}(D_2) = 1$ and the same holds for $Q_N$. The exact same arguments as used in the proof for the circle topology are applied to the rest of the drift, and we conclude that it is negative provided $\lambda < 3/8$.

Remark 7. From the proof of Theorem \[5\] we can also see that for system with inhomogeneous arrivals, if we denote by $\lambda_i = \mathbb{E}(\xi_i(n))$, then
\[
\lambda_{i-1} + 2\lambda_i + \lambda_{i+1} < 3/2
\]
is sufficient for stability. The above is guaranteed, for instance, by
\[
\lambda_i + \lambda_{i+1} < 3/4
\]
for every $i$. 

9
3.4 Main results

The following lemma shows that for a circle of length \( N \geq 4 \), the parking constant is at least \( \frac{2}{5} \).

**Lemma 8.** For any \( N \geq 4 \), \( C_N/N \geq \frac{2}{5} \).

The proof of this lemma is presented in the Appendix.

Our main result for the system on a circle is the following **Theorem 9.** (Its proof is given in Section 4.)

**Theorem 9.** The system with the circle topology and \( N \geq 4 \) is stable if \( \lambda < \frac{2}{5} \).

**Remark 10.** Note that the system we consider is trivially stable under the condition \( \lambda < C_N/N \), if all nodes compete for transmission at all times, even when they are empty. However, we want to emphasise that this fact does not imply Theorem 9 and does not help to prove it. Recall that our main interest is the model of a standard CSMA, under which each node is “completely unaware” of its surrounding and competes for the channel when and only when it has packets to transmit. The difference between the models when all nodes are competing all the time and when empty nodes remain silent, is even more evident from our next result for the line topology.

The main result for the line system has the same form as for the circle. (Its proof is also in Section 4.)

**Theorem 11.** The system with the line topology and \( N \geq 4 \) is stable if \( \lambda < \frac{2}{5} \).

**Remark 12.** This result is more surprising than that for a circle as on a line, if all nodes have non-empty queues, their transmission probabilities are different. In fact, the transmission probability of the second node (or node \( N - 1 \)) is equal to \( 3/8 \) if \( N = 4 \), \( 11/30 \) if \( N = 5 \) and can be shown to decrease over even values of \( N \) and increase over odd values of \( N \) to the same limit \( 1 - e^{-1} \approx 0.3679 \) (see [21]), thus never exceeding \( 2/5 \). This demonstrates that in order to obtain correct stability conditions, it is not sufficient to consider the throughputs of nodes in saturation.

4 Proof of main results

4.1 Fluid limits

We will prove Theorem 9 and Theorem 11 using the fluid limit technique [18, 9, 22]. (For the application of the fluid limit technique to discrete time processes, see e.g. [8].) This section gives definitions and preliminaries that will be needed in both proofs.

Denote by \( N = \{1, \ldots, N\} \) the set of all nodes, and by \( Q(n) = (Q_i(n), \ i \in N) \) the queue length vector. The Markov process that we consider, \( Q(n), n = 0,1,\ldots, \) is a discrete time countable irreducible Markov chain. To apply the fluid limit technique, consider a sequence of processes \( Q^r(\cdot) \) with increasing initial state, \( \|Q^r(0)\| = r \uparrow \infty \), is fixed. (Here and below, the upper index \( r \), which is the norm of initial state, is used as a process label.) To establish the process stability (positive recurrence) is suffices to prove that for some fixed integer \( T > 0 \) any such sequence of processes is such that

\[ \mathbb{E} \frac{1}{r}\|Q^r(rT)\| \to 0, \ \ r \to \infty. \]  (5)

First, we will view process \( Q^r(\cdot) \) (and all other system processes) in continuous time \( t \geq 0 \), by adopting the convention \( Q^r(t) = Q^r([t]) \), \( t \geq 0 \). Define the associated processes as follows: \( F^r_i(t) \) is the total number of exogenous arrivals in node \( i \) by time \( t \); \( H^r_i(t) \) is the total number
of departures from (transmissions at) node $i$ by time $t$. We will also need an additional family of processes, $G^r_B(t)$, whose definition requires some details, given next.

Denote by $s$ a system occupancy state (at a given time), $s = (s_i, i \in N)$, where $s_i$ is 1 [resp. 0] if node $i$ is occupied [resp. empty]. There is, of course, only a finite number $2^N$ of possible $s$. At time $n$ the occupancy state is given by $s_t = \{Q_i(n) > 0\}$. Denote by $u = (u_i, i \in N)$ a ranking realisation (at a given time), where $u_i \in N$ is the ranking that node $i$ receives. In other words, $u \in U$, where $U_N$ is the set of permutations of $(1, \ldots, N)$. Denote by $d$ a transmission realisation (at a given time), $d = (d_i, i \in N)$, $d_i$ is 1 [resp. 0] if node $i$ transmits [resp. does not transmit]. Clearly, at any time, $d = \phi(s, u)$ is a deterministic function of $s$ and $u$. Denote by $\Psi$ the set of possible values of a triple $(s, u, d)$. Then, for $B \subseteq \Psi$, denote $G^r_B(t)$ is the total number of time slots, up to and including time $t$, in which $(s, u, d) \in B$. (For example, $B$ can be a subset – or, “event” – like $\{Q_i > 0$ and node $i + 1$ transmits}.) To simplify notation, we often write $G^r_s(t)$ to mean $G^r_{s=\sigma}(t)$.

To summarize, the processes that we will consider are $Q^r_t(\cdot), F^r_t(\cdot), H^r_t(\cdot)$ for $i \in N$, and $G^r_B(\cdot)$ for $B \in \Psi$.

Recall that the processes with different index $r$ have different fixed initial states. We will construct all these processes (for all $r$) on a common probability space as follows. The arrival process $F^r_t(\cdot)$ for a given $i$ (and for all $r$) is driven by an independent i.i.d. sequence of random variables $\xi_i(1), \xi_i(2), \ldots$, with finite mean $\mathbb{E}\xi_i(1) = \lambda$; namely,

$$F^r_t(t) = \sum_{1 \leq n \leq t} \xi_i(n).$$

The processes that drive transmissions (for all $r$) are as follows. For each $s$ we define an independent i.i.d. driving sequence of random elements $u_s(\ell), \ell = 1, 2, \ldots$, distributed randomly uniformly in $U_N$. When the system is in occupancy state $s$ for the $\ell$th time, the ranking realisation $u_s(\ell)$ is taken, which uniquely determines the transmission realisation $d = \phi(s, u_s(\ell))$. Clearly, this collection of driving processes defines the entire process uniquely for each initial state.

Denote by $\Theta$ the set of all possible pairs $(u, d)$. For a given $s$, denote by $\{P_s(u, d), (u, d) \in \Theta\}$, the probability distribution of a random outcome of $(u, d)$ when the system occupancy state is $s$. In other words, $P_s(u, d)$ the probability distribution on $\Theta$, induced by $u$ distributed uniformly in $U_N$, and (deterministic) mapping $d = \phi(s, u)$; clearly,

$$P_s(u, d) = \frac{1}{N!} \mathbb{I}\{d = \phi(s, u)\}.$$

We use notation

$$P_s(A) = \sum_{(u, d) \in A} P_s(u, d), \quad A \subseteq \Theta,$$

for the probabilities of events, and $P_s(A|A')$ for conditional probabilities.

Note that the driving processes are such that the following FSLLN properties hold. With probability 1,

$$\frac{1}{r} F^r_t(rt) \rightarrow \lambda t, \quad \text{u.o.c., } \forall i \in N, \quad (6)$$

$$\frac{1}{r} \sum_{\ell \leq rt} \mathbb{I}\{u_s(\ell) \in A\} \rightarrow P_s(A) t, \quad \text{u.o.c., } \forall A \subseteq \Theta, \forall s. \quad (7)$$

For each $r$ we define fluid-scaled processes:

$$q^r_i(t) = \frac{1}{r} Q^r_i(rt), \quad f^r_i(t) = \frac{1}{r} F^r_i(rt), \quad h^r_i(t) = \frac{1}{r} H^r_i(rt),$$

$$q^r_B(t) = \frac{1}{r} G^r_B(rt).$$
Definition 13. A collection of deterministic functions \( \chi = [(q_i(\cdot), f_i(\cdot), h_i(\cdot), i \in \mathcal{N}), (g_B(\cdot), B \subseteq \Psi)] \) is called a fluid limit if there exists a subsequence of \( r \) and the corresponding sequence of the (scaled) process realisations \( \chi^r = [(q_i^r(\cdot), f_i^r(\cdot), h_i^r(\cdot), i \in \mathcal{N}), (g_B^r(\cdot), B \subseteq \Psi)] \) such that, as \( r \to \infty \) along this subsequence, \([\Box] \) and \([\Box] \) hold, and

\[ \chi^r \to \chi, \text{ u.o.c. for each component function.} \]

Lemma 14. Any fluid limit \( \chi \) has the following basic properties.

(i) All component functions are Lipschitz continuous. Therefore, a.e. w.r.t. Lebesgue measure, all component functions have derivatives; time points, where this holds are called regular.

(ii) \( \|q(0)\| = 1 \).

(iii) \( q_i(t) = q_i(0) + f_i(t) - h_i(t), \forall i. \)

(iv) \( f_i(t) = \lambda t, t \geq 0, \forall i. \)

(v) \( h_i(t) = g_{(d_i=1)}(t), t \geq 0, \forall i. \)

(vi) For any occupancy state \( \sigma \) and any \( A \subseteq \Theta \),

\[ g_{(s=\sigma,(u,d)\in A)}(t) = P_\sigma(A)g_\sigma(t). \]

For a regular time point \( t \), it will be convenient to use notation

\[ \pi_t(B) = (d/dt)g_B(t); \]

and we sometimes will write \( \pi_t(\sigma) \) to mean \( \pi_t(s = \sigma) \). Clearly, \( \pi_t(\cdot) \) is a probability distribution on \( \Psi; \pi_t(B|B') \) denotes conditional probabilities.

Lemma 15. The sequence of fluid-scaled processes \( \chi^r = [(q_i^r(\cdot), f_i^r(\cdot), h_i^r(\cdot), i \in \mathcal{N}), (g_B^r(\cdot), B \subseteq \Psi)] \) is such that the following holds w.p.1. For any subsequence of \( r \) there is a further subsequence, along which

\[ \chi^r \to \chi, \text{ u.o.c. for each component function,} \]

where \( \chi \) is a fluid limit.

To prove Theorem \([\Box]\) it suffices to show that for some fixed \( T > 0 \) and any fluid limit \( \chi \),

\[ \|q(t)\| = 0, \forall t \geq T. \]  \hspace{1cm} (8)

(The family of random variables \( \|q'(T)\| \equiv (1/r)\|Q'(rT)\| \) is easily seen to be uniformly integrable. This, along with \([\Box]\) and Lemma \([\Box]\) implies \([\Box]\).)

4.2 Auxiliary definitions and facts.

4.2.1 Ranking and conditional ranking realisation constructions.

Consider a set of \( m \) nodes. They can be arranged in a circle or in a line. (For the definition that follows it is only required that a neighbour relation between the nodes is defined.) For a given occupancy state \( s \) the random process which determines the nodes’ ranking realisation – and the transmission realisation as well – is as follows. The occupied nodes are initialised as active and idle nodes as inactive. The ranking, from 1 to \( m \), is assigned to nodes in the order of being chosen (or, picked). The first node is chosen uniformly at random among all \( m \) nodes; if this node is active and occupied, it transmits and then deactivates itself and its neighbours; if this node is inactive, no further action is taken. The second node is chosen among the \( m - 1 \) nodes not chosen before, and the action is taken according to the same rule as for the first node. And so on, until all nodes are chosen – this produces the ranking and transmission realisations of all nodes.

Let two nodes \( i \) and \( j \) be fixed. Later on we will need to generate the ranking (and transmission) realisations, conditioned on the event \( u_i < u_j \). It is easy to check that this can be done
via the following construction. The process, producing the ranking/transmission realisation, is the same as described above, with the following modification. As long as both nodes $i$ and $j$ are not chosen yet, when the next node is being chosen, nodes $i$ and $j$ are considered as a pair, which is picked with the probability twice greater than that for a single node. If/when the pair $(i, j)$ is chosen, the node $i$ is picked. After that, the process runs as usual without modifications.

4.2.2 Properties of occupied segments of nodes

**Definition 16.** For given occupancy state $s$, a set of $k$ consecutive occupied nodes, bordering on idle nodes (or no nodes) on both sides, is called an occupied segment of length $k$. Without loss of generality, for each $k$, we can and will consider the occupied segment consisting of nodes $1, \ldots, k$, with nodes $N$ (in the case of a circle) and $k + 1$ being idle, and will denote it $[1, k]$.

We are interested in the transmission probabilities of nodes within one occupied segment (and one time slot). Clearly, these probabilities depend only on the random mutual ranking of the nodes within this segment. The mutual ranking $u' = (u'_1, \ldots, u'_k) \in U_k$ is determined by total ranking $u$: $u'_i < u'_j$ if and only if $u_i < u_j$. The distribution of the mutual ranking $u'$ is uniform on the permutations set $U_k$. Formally speaking, we are interested in the probabilities $P_s(A)$ in the case when $s$ is such that $[1, k]$ is an occupied segment, and $A$ depends only on the mutual ranking and transmissions within this segment. For this reason, the corresponding probabilities (with a slight abuse of notation) are denoted by $P_k(A)$.

**Definition 17.** Nodes $i$ and $j$ within same occupied segment are called friends [resp. foes] if $|j - i|$ is even [resp. odd].

The following lemma means that if we add node $k + 1$ or a pair of nodes $(k + 1, k + 2)$ to a occupied segment $[1, k]$, this addition can only increase the transmission probabilities of the nodes in $[1, k]$ that are friends of node $k + 1$, and can only decrease the transmission probabilities of the nodes in $[1, k]$ that are foes of $k + 1$.

**Lemma 18.** Consider an occupied segment $[1, k]$, and a subset $L$ of its nodes which are all mutual friends. Consider also the larger occupied segments $[1, k + 1]$ and $[1, k + 2]$. Then, if node $k + 1$ is a friend [resp. foe] of the nodes in $L$, we have

\[ P_k(d_i = 1, i \in L) \geq [\text{resp. } \leq] P_{k+1}(d_i = 1, i \in L), \tag{9} \]

\[ P_k(d_i = 1, i \in L) \geq [\text{resp. } \leq] P_{k+2}(d_i = 1, i \in L). \tag{10} \]

**Proof.** Let us prove (9). (The proof of (10) is completely analogous.) We will prove the following (stronger) statement. The ranking and transmission realisations of the segments $[1, k]$ and $[1, k + 1]$ can be coupled (constructed on a common probability space), so that for any $i = 1, \ldots, k$: $(d_i = 1$ in $[1, k]$-segment) implies $(d_i = 1$ in $[1, k + 1]$-segment) when $i$ is a friend of $k + 1$; and $(d_i = 0$ in $[1, k]$-segment) implies $(d_i = 0$ in $[1, k + 1]$-segment) when $i$ is a foe of $k + 1$. This property is proved by induction on $k$. The statement is trivially true for $k = 0$, if we adopt the convention that this case corresponds to the “empty” segment – it has no nodes and, therefore, no friends or foes of any other node. This is the base of induction. Suppose, the statement is true for $k = 0, 1, \ldots, m - 1$. Let us prove it for $k = m$. We are constructing the ranking and transmission realisations, jointly for the segments $[1, m]$ and $[1, m + 1]$. Segment $[1, m + 1]$ initialised to have all nodes occupied and active; segment $[1, m]$ initialised to have all nodes occupied and active, but it is augmented by a “dummy” node $m + 1$, which is initialised to be inactive. The process for ranking/transmission realisation runs for $m + 1$ nodes, jointly in both systems. We stop this process when either (a) all nodes are inactive or (b) node $m + 1$ is chosen. In the case (a), the transmission realisation for nodes $1, \ldots, m$ in both systems is same. In the case (b), there are two sub-cases: (b.1) node $m$ was already inactive and (b.2)
node \( m \) was active. In the case (b.1) the transmission of node \( m + 1 \) in the \([1, m + 1]\)-system and its non-transmission in the \([1, m]\)-system will not affect nodes 1, \ldots, \( m \) in either system, and therefore the rest of the ranking/transmission realisations in both systems can obviously be coupled to be the same. The remaining case to consider is (b.2). In the \([1, m + 1]\)-system, the node \( m \) — a foe of \( m + 1 \) — is deactivated, and will not transmit; in the \([1, m]\)-system the node \( m \) remains active, and therefore may eventually transmit; so, the property we are proving holds for node \( m \). The remaining active nodes in both systems are same, except the right-most occupied segment (if we consider only active occupied nodes) in the \([1, m]\)-system has the additional node \( m \) on the right. (As a degenerate special case, this segment may consist of the single node \( m \) in the \([1, m]\)-system and be “empty” in the \([1, m + 1]\)-system.) This right-most segment has length at most \( m \) in the \([1, m]\)-system. It remains to apply the induction hypothesis to complete the construction of coupling. □

**Lemma 19.** For any occupied segment of length \( k \geq 1 \)

\[ P_k(d_1 = 1) \geq 1/2. \]

**Proof.** \( P_k(d_1 = 1) \geq P_k(u_1 < u_2) = 1/2. \)

**Lemma 20.** For any occupied segment of length \( k \geq 2 \)

\[ P_k(d_1 = 1) \geq P_k(d_2 = 1). \]

That is, an end node of a segment has the transmission probability not smaller than that of its neighbour.

**Proof.** Clearly, \( P_k(d_1 = 1 \text{ or } d_2 = 1) = 1 \), and events \( \{d_1 = 1\} \) and \( \{d_2 = 1\} \) are mutually exclusive. So, \( P_k(d_1 = 1) + P_k(d_2 = 1) = 1 \). And \( P_k(d_1 = 1) \geq 1/2 \) (Lemma 19). □

### 4.2.3 Additional properties of fluid limits.

First, note some standard properties that hold at a regular point of a fluid limit: if \( q_i(t) = 0 \), then necessarily \( q'_i(t) = 0 \), and then \( h'_i(t) = g'_i(d_i = 1) = \pi_t(d_i = 1) = \lambda \); if \( q_i(t) > 0 \), then necessarily \( \pi_t(s_i = 1) = 1 \).

**Lemma 21.** Consider a regular point \( t \) such that \( \pi_t(d_{i-1} = 1) > 0, q_i(t) > 0, q_{i+1}(t) > 0 \). Then,

\[ \pi_t(d_{i+1} = 1 \mid d_{i-1} = 1) \geq 1/2. \]

**Proof.** From \( q_i(t) > 0, q_{i+1}(t) > 0 \) we know that \( \pi_t(s_i = 1, s_{i+1} = 1) = 1 \). Observe that

\[
\pi_t(d_{i+1} = 1, d_{i-1} = 1) \geq \pi_t(d_{i+1} = 1, d_{i-1} = 1, u_{i+1} < u_{i+2}) = \pi_t(s_{i+1} = 1, d_{i+1} = 1, d_{i-1} = 1, u_{i+1} < u_{i+2}) = \pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}),
\]

where we used \( \pi_t(s_{i+1} = 1) = 1 \) and the fact that \( (s_{i+1} = 1, d_{i-1} = 1, u_{i+1} < u_{i+2}) \) implies \( d_{i+1} = 1 \). Therefore, to prove the lemma it suffices to prove

\[ \pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}) \geq (1/2)\pi_t(d_{i-1} = 1), \]

which is what we proceed to do.

We have

\[
\pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}) = \sum_{\sigma} \pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}, s = \sigma) = \sum_{\sigma} \pi_t(\sigma)P_\sigma(d_{i-1} = 1, u_{i+1} < u_{i+2})
\]

\[
= \sum_{\sigma} \pi_t(\sigma)P_\sigma(u_{i+1} < u_{i+2})P_\sigma(d_{i-1} = 1 \mid u_{i+1} < u_{i+2}) = (1/2) \sum_{\sigma} \pi_t(\sigma)P_\sigma(d_{i-1} = 1 \mid u_{i+1} < u_{i+2})
\]

14
\[
\pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}) = (1/2) \sum_{\sigma: \sigma_{i-1} = \sigma_i = \sigma_{i+1} = 1} \pi_t(\sigma) P_\sigma(d_{i-1} = 1 \mid u_{i+1} < u_{i+2})
\]

Analogously,
\[
\pi_t(d_{i-1} = 1, u_{i+1} > u_{i+2}) = (1/2) \sum_{\sigma: \sigma_{i-1} = \sigma_i = \sigma_{i+1} = 1} \pi_t(\sigma) P_\sigma(d_{i-1} = 1 \mid u_{i+1} > u_{i+2}).
\]

We claim that for any \( s \) such that \( s_{i-1} = s_i = s_{i+1} = 1 \), we have
\[
P_s(d_{i-1} = 1 \mid u_{i+1} < u_{i+2}) \geq P_s(d_{i-1} = 1 \mid u_{i+1} > u_{i+2}). \tag{12}
\]

The proof of (12) is by coupling the ranking and transmission realisations under the conditions \( \{u_{i+1} < u_{i+2}\} \) and \( \{u_{i+1} > u_{i+2}\} \), which we label as (a) and (b), respectively. The construction is as follows. We use the process for producing a conditional ranking of this type, described in Section 12.4.1 for conditions \( \{u_{i+1} < u_{i+2}\} \) and \( \{u_{i+1} > u_{i+2}\} \), but we couple these processes. Specifically, as long as neither \( i + 1 \) nor \( i + 2 \) is chosen yet, when the next node is being chosen these two nodes are considered as a pair, which is picked with the probability twice greater than that a single node. If/when the pair \( (i + 1, i + 2) \) is chosen, the node \( i + 1 \) or \( i + 2 \) is picked deterministically, depending on whether the realisation is under condition (a) or (b), respectively. The process runs like this, and is stopped when (c1) node \( i - 1 \) transmits or is deactivated, or (c2) the pair \( (i + 1, i + 2) \) is chosen. In the former case, obviously, \( u_{i-1} \) is same for conditions (a) and (b), so it remains to consider the case (c2) \( \setminus \) (c1). Note that before the pair \( (i + 1, i + 2) \) is chosen, nodes \( i - 1, i, i + 1 \) are active. If node \( i + 2 \) was inactive, then under condition (a) node \( i - 1 \) becomes an end node of an occupied segment (if we only consider occupied active nodes), and under condition (b) node \( i - 1 \) is within the same occupied segment, but with nodes \( i \) and \( i + 1 \) appended “on the right”; therefore, starting from this state of the access process, by Lemma 18 the probability that node \( i - 1 \) eventually transmits is higher under condition (a). Suppose node \( i + 2 \) was active (and then necessarily occupied). Consider two subcases: (c2.1) not all nodes were occupied and active and (c2.2) all nodes were occupied and active. (For the line topology, we only need to consider the case (c2.1).) In case (c2.1), under condition (a) node \( i - 1 \) becomes an end node of an occupied segment (if we only consider occupied active nodes), and under condition (b) node \( i - 1 \) is within the same occupied segment, but with node \( i \) appended “on the right”; therefore, starting from this state of the access process, by Lemma 18 the probability that node \( i - 1 \) eventually transmits is higher under condition (a). In case (c2.2), under condition (a) node \( i - 1 \) becomes an end node of the occupied segment of length \( N - 3 \) (if we only consider occupied active nodes), and under condition (b) node \( i - 1 \) is within the occupied segment of same length \( N - 3 \), but with node \( i - 1 \) being second from the end; therefore, starting from this state of the access process, by Lemma 20 the probability that node \( i - 1 \) eventually transmits is higher under condition (a).

This completes the proof of claim (12).

Using (12) we obtain
\[
\pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}) \geq \pi_t(d_{i-1} = 1, u_{i+1} > u_{i+2}).
\]

Given that
\[
\pi_t(d_{i-1} = 1, u_{i+1} < u_{i+2}) + \pi_t(d_{i-1} = 1, u_{i+1} > u_{i+2}) = \pi_t(d_{i-1} = 1),
\]
we obtain (11). \( \square \)

### 4.3 Proof of Theorem 9

We are going to show that there exists \( \epsilon > 0 \) such that for any fluid limit, at any regular point \( t \) such that \( \sum_{t=1}^{N} q_i(t) > 0 \),
\[
\sum_{i=1}^{N} q_i(t) \leq -\epsilon, \tag{13}
\]
This will prove stability, because, clearly, (13) implies (5).

Consider first the case when \( q_i(0) > 0 \) for all \( i \). In this case \( \pi_t(s_i = 1) = 1 \) for all \( i \) and then, by symmetry, \( \pi_t(d_i = 1) = C_N/N \) for all \( i \). Then, \( q'_i(t) = \lambda - C_N/N < 0 \) for all \( i \) by Lemma 8 and (13) follows (for any \( \epsilon < C_N/N - \lambda \)).

We now turn to the case when there is at least one \( i \) with \( q_i(t) = 0 \). A group of nodes \((k+1, k+2, \ldots, k+l)\) such that \( q_k(t) = q_{k+1}(t) = 0 \) and \( q_{k+i}(t) > 0 \) for all \( i = 1, \ldots, l \), we will call a positive group of size \( l \). We will show that

\[
\sum_{i=1}^{l} q'_{k+i}(t) < -\epsilon_1 < 0,
\]

for any positive group \((k+1, k+2, \ldots, k+l)\) and, moreover, the constant \( \epsilon_1 \) in our estimates will depend only on the size \( l \) of the group. Since there is only a finite number of possible values of \( l \), (13) will follow.

We are going to present our proof of (13) for different values of \( l \).

1. First consider the case \( l = 1 \). Recall \( \pi_t(s_{k+1} = 1) = 1 \). Let \( a = \pi_t(s_k = 1, s_{k+2} = 1) \) and let \( b = \pi_t(s_{k-1} = 1, s_{k+3} = 1 \mid s_k = 1, s_{k+2} = 1) \).

From Lemma 19 we immediately obtain

\[
\pi_t(d_{k+1} = 1 \mid s_k = 0 \cup s_{k+2} = 0) \geq 1/2.
\]

And from Lemma 18 we get

\[
\pi_t(d_{k+1} = 1 \mid s_k = 1, s_{k+2} = 1, s_{k-1} = 0 \cup s_{k+3} = 0) \geq 1/3
\]

and

\[
\pi_t(d_{k+1} = 1 \mid s_k = 1, s_{k+2} = 1, s_{k-1} = 1, s_{k+3} = 1) \geq 179/420.
\]

Indeed, due to Lemma 18 in the first case the worst possible \( s \) (appearing in the condition) in terms of the smallest \( P_s(d_{k+1} = 1) \) is such that \( k+1 \) is the middle node in an occupied segment of length 3, and in the second case - such that it is the middle node in an occupied segment of length 7. The corresponding probabilities may either be calculated directly, or using [21] Theorem 1.

We obtain:

\[
q'_{k+1}(t) \leq \lambda - 1/2(1 - a) - a((1 - b)(1/3) + b(179/420)) = \lambda - 1/2 + a/2 - a(1/3 + b(13/140)) = \lambda - 1/2 + (1/2)a(1/3 - b(13/70)).
\]

Therefore, if

\[
a(1/3 - b(13/70)) < \lambda/2,
\]

then \( q'_{k+1}(t) < \lambda - 1/2 + \lambda/4 < 0 \), provided \( \lambda < 2/5 \), and (13) follows.

We can also write a different estimate of \( q'_{k+1}(t) \):

\[
q'_{k+1}(t) = \lambda - \pi_t(d_k = 0, d_{k+2} = 0) = \lambda - 1 + \pi_t(d_k = 1 \cup d_{k+2} = 1) = \lambda - 1 + \pi_t(d_k = 1) + \pi_t(d_{k+2} = 1) - \pi_t(d_k = 1, d_{k+2} = 1)
\]

\[
= \lambda - 1 + 2\lambda - \pi_t(d_k = 1, d_{k+2} = 1) = 3\lambda - 1 - a\pi_t(d_k = 1, d_{k+2} = 1 \mid s_k = 1, s_{k+2} = 1),
\]

where \( a \) is the probability that \( s_k = s_{k+2} = 1 \) and \( d_k = 1 \). This follows from the fact that \( \pi_t(s_k = 1) = 1 \).

We obtain:

\[
q'_{k+1}(t) < \lambda - 1/2 + \lambda/4 < 0,
\]

which completes the proof for \( l = 1 \). The general case can be handled similarly.
and to prove (14) it is sufficient to show that
\[ a\pi_t(d_k = 1, d_{k+2} = 1|s_k = 1, s_{k+2} = 1) > \lambda/2. \] (16)

To summarise, in order to prove (14) in this case, it is sufficient to prove either (15) or (16).

Note that, thanks to Lemma 18,
\[ \pi_t(d_k = 1, d_{k+2} = 1|s_k = 1, s_{k+2} = 1, s_{k-1} = 1, s_{k+3} = 1) \geq 1/5 \]
and
\[ \pi_t(d_k = 1, d_{k+2} = 1|s_k = 1, s_{k+2} = 1, s_{k-1} = 0 \cup s_{k+3} = 0) \geq 3/8. \]

Indeed, due to Lemma 18 in the first case the worst possible \( s \) (appearing in the condition) in terms of the smallest \( P_s(d_k = 1, d_{k+2} = 1) \) is such that \( k \) and \( k + 2 \) are nodes 2 and 4 in an occupied segment of length 5, and in the second case - such that they are nodes 2 and 4 in an occupied segment of length 4. The corresponding probabilities are easy to verify directly.

We then have
\[ \pi_t(d_k = 1, d_{k+2} = 1|s_k = 1, s_{k+2} = 1) \geq (1/5)b + (3/8)(1 - b) = 3/8 - b(7/40), \]
and, in order to show (16), it is sufficient to show that
\[ a(3/8 - b(7/40)) > \lambda/2. \]

Simple algebra shows that either the above or (15) holds for any values of \( a > 0 \) and \( b > 0. \)

2. Assume now that \( l = 2. \) Note that \( \pi_t(d_{k+1} + d_{k+2} = 1 \mid d_k = 0) = 1. \) (When node \( k \) is not transmitting, exactly one of nodes \( k + 1 \) and \( k + 2 \) will transmit.) Therefore,
\[ q'_{k+1}(t) + q'_{k+2}(t) = 2\lambda - \pi_t(d_k = 1)\pi_t(d_{k+2} = 1|d_k = 1) - \pi_t(d_k = 0) = 3\lambda - 1 - \lambda\pi_t(d_{k+2} = 1|d_k = 1) \leq (5/2)\lambda - 1 < 0, \]
because (from Lemma 24)
\[ \pi_t(d_{k+2} = 1|d_k = 1) \geq 1/2. \]

3. Assume now that \( l = 3 \) and note that in this case
\[ q'_{k+1}(t) + q'_{k+2}(t) + q'_{k+3}(t) = 3\lambda - (\pi_t(d_{k+1} = 1) + \pi_t(d_{k+2} = 1) + \pi_t(d_{k+3} = 1)) = 3\lambda - (\pi_t(d_{k+1} = 1) + \pi_t(d_{k+3} = 1)) - (1 - \pi_t(d_{k+1} = 1 \cup d_{k+3} = 1)) = 3\lambda - (1 - \pi_t(d_{k+1} = 1 + \pi_t(d_{k+3} = 1)) - 1 + \pi_t(d_{k+1} = 1) + \pi_t(d_{k+3} = 1)) - \pi_t(d_{k+1} = 1, d_{k+3} = 1) = 3\lambda - 1 - \pi_t(d_{k+1} = 1, d_{k+3} = 1) \leq 3\lambda - 1 - 1/5 < 0 \]
if \( \lambda < 2/5, \) where the second-last inequality above follows from Lemma 18 again by considering \( s \) for which \( \pi_t(d_{k+1} = 1, d_{k+3} = 1) \) is the smallest, equal 1/5.

4. Finally, consider the case \( l \geq 4. \) Let us first focus on nodes \( k + 1 \) and \( k + 2. \) Note that in our proof for the case \( l = 2 \) we only used the fact that for the node to the left, node \( k, \)
\[ \pi_t(d_k = 1) = \lambda. \] Therefore the proof may be repeated to show that
\[ q'_{k+1}(t) + q'_{k+2}(t) < 0. \]
The same argument of course implies that

\[ q_{k+l-1}(t) + q_{k+l}(t) < 0. \]

Note also that, due to Lemma 18, for any \( 3 \leq i \leq l - 2 \), \( \pi_i(d_{k+i} = 1) \geq 179/420 \), as the “worst” occupancy state \( s \) (among those with \( s_j = 1 \) for \( i - 2 \leq j \leq i + 2 \)), in the sense of the smallest \( P_s(d_i = 1) \), is the one where \( i \) is the middle node within an occupied segment of length 7 (and the corresponding probability may be calculated directly of from [21, Theorem 1]). As \( 179/420 > 2/5 \), our proof is complete.

\[ \square \]

### 4.4 Proof of Theorem 11

The proof of Theorem 11 follows the same lines as that of Theorem 9.

Note that for all \( N \geq 4 \), \( C_N = 1 + L_{N-3} \leq L_N \) and therefore, due to Lemma 8, \( L_N/N \geq 2/5 \) and

\[ \sum_{i=1}^{N} q_i(t) = N\lambda - L_N < -\varepsilon \]

with a positive \( \varepsilon \), if \( q_i(t) > 0 \) for all \( i \).

We can therefore consider positive groups of nodes bordering at least on one side an existing node \( i \) with \( q_i(t) = 0 \). Note that having no node on one side is beneficial for a group of \( l \geq 2 \) compared to having an existing node with \( q_i(t) = 0 \). Therefore the proof of the negativity of the total drift of a group of \( l \geq 2 \) nodes carries over from Theorem 9 without changes.

The proof in the case \( l = 1 \) is unchanged again if the node \( k + 1 \) under consideration is \( 3, \ldots, N-2 \) as it relies only on considering nodes at a distance at most 2 from the target one. If node 1 (or node \( N \)) is the node under consideration, then, due to Lemma 19, its drift is at most \( \lambda - 1/2 < 0 \). Therefore it remains to look at the case when the node under consideration is 2 (the case of node \( N - 1 \) is exactly the same by symmetry). We are going to show that the drift of node 2 is negative in this case. The proof follows the same lines as the one given for the case \( l = 1 \) in Theorem 9, with small changes in the worst cases for transmissions of certain (groups of) nodes. We provide the argument here for completeness.

Recall \( \pi_t(s_2 = 1) = 1 \). Let

\[ a = \pi_t(s_1 = 1, s_3 = 1) \]

and let

\[ b = \pi_t(s_4 = 1 \mid s_1 = 1, s_3 = 1) \]

From Lemma 19 we immediately obtain

\[ \pi_t(d_2 = 1 \mid s_4 = 0 \cup s_3 = 0) \geq 1/2. \]

Trivially,

\[ \pi_t(d_2 = 1 \mid s_1 = 1, s_3 = 1, s_4 = 0) = 1/3 \]

and from Lemma 18 we get

\[ \pi_t(d_2 = 1 \mid s_1 = 1, s_3 = 1, s_4 = 1) \geq 11/30. \]

Indeed, due to Lemma 18 in the last case the worst possible \( s \) (appearing in the condition) in terms of the smallest \( P_s(d_2 = 1) \) is such that it is node 2 in a segment of 5 non-empty nodes. The corresponding probability may either be calculated directly, or using formulas from [21].
We obtain:

\[ q_2'(t) \leq \lambda - 1/2(1 - a) - a((1 - b)(1/3) + b(11/30)) \]
\[ = \lambda - 1/2 + a/2 - a(1/3 + b(1/30)) = \lambda - 1/2 + (1/2)a(1/3 - b(1/15)). \]

Therefore, if
\[ a(1/3 - b(1/15)) < \lambda/2, \]
then \( q_2'(t) < \lambda - 1/2 + \lambda/4 < 0 \), provided \( \lambda < 2/5 \), and \( q_2'(t) < 0 \).

We can also write a different estimate of \( q_2'(t) \):

\[ q_2'(t) = \lambda - \pi_t(d_1 = 0, d_3 = 0) = \lambda - 1 + \pi_t(d_1 = 1 \bigcup d_3 = 1) \]
\[ = \lambda - 1 + \pi_t(d_1 = 1) + \pi_t(d_3 = 1) - \pi_t(d_1 = 1, d_3 = 1) \]
\[ = \lambda - 1 + 2\lambda - \pi_t(d_1 = 1, d_3 = 1) = 3\lambda - 1 - a\pi_t(d_1 = 1, d_3 = 1|s_1 = 1, s_3 = 1), \]
and to prove that \( q_2'(t) < 0 \) it is sufficient to show that
\[ a\pi_t(d_1 = 1, d_3 = 1|s_1 = 1, s_3 = 1) > \lambda/2. \]

To summarise, in order to prove that \( q_2'(t) < 0 \) in this case, it is sufficient to prove either (17) or (18).

Note that, thanks to Lemma 18, in the first case the worst possible \( s \) (appearing in the condition) in terms of the smallest \( P_s(d_1 = 1, d_3 = 1) \) is such that 1 and 3 are nodes 1 and 3 in an occupied segment of length 4.

We then have
\[ \pi_t(d_1 = 1, d_3 = 1|s_1 = 1, s_3 = 1, s_4 = 1) \geq 3/8 \]
and
\[ \pi_t(d_1 = 1, d_3 = 1|s_1 = 1, s_3 = 1, s_4 = 0) = 2/3. \]

Indeed, due to Lemma 18, in the first case the worst possible \( s \) (appearing in the condition) in terms of the smallest \( P_s(d_1 = 1, d_3 = 1) \) is such that 1 and 3 are nodes 1 and 3 in an occupied segment of length 4.

We then have
\[ \pi_t(d_1 = 1, d_3 = 1|s_1 = 1, s_3 = 1) \geq (3/8)b + (2/3)(1 - b) = 2/3 - b(7/24), \]
and, in order to show (16), it is sufficient to show that
\[ a(2/3 - b(7/24)) > \lambda/2. \]

Simple algebra shows that either the above or (17) holds for any values of \( a > 0 \) and \( 0 < b < 1 \).

\[ \square \]

5 Conclusion, conjectures and further work

We considered a networks of \( N \) nodes arranged in a circle or a line where neighbouring nodes cannot transmit at the same time and each node’s medium access is governed by a discrete-time CSMA protocol. Our main result shows that the system in either topology is stable for any \( N \geq 4 \) if the per-node arrival rate \( \lambda < 2/5 \). This provides an upper bound on the loss in throughput the standard (decentralised) CSMA algorithm suffers compared to the maximum possible throughput of \( \lfloor N/2 \rfloor /N \) that is achieved by an ideal centralised protocol.

The result is intuitive for the circle topology. Indeed, instability for a system of \( N \) nodes on a circle may be easily proved if \( \lambda > C_N/N \), where \( C_N/N \) is the parking constant which is higher than 2/5 and converges, as \( N \) goes to infinity, to \( 1/2(1 - e^{-2}) \approx 0.4323 \). Instability
holds because, if all nodes are active, then each of them receives a throughput of $C_N/N$. We conjecture that stability in fact holds if $\lambda < C_N/N$. This is an open question.

Our result is not intuitive in the case of the line topology because, if all nodes are active, some receive a throughput of less than $2/5$. This illustrates how intricate the system behaviour is and how the analysis of unsaturated systems is difficult even in such simple topologies. The exact stability condition, even for a homogeneous system, is an open problem for the line topology; this condition is even not easy to conjecture.

We are currently also working on more difficult multi-hop networks. For a line we consider traffic arriving at node 1 and requiring to reach node $N$. Each message thus needs to be transmitted by every node in turn. We are interested in both stability and end-to-end throughput of the system.

For a circle one can define a multi-hop network in, for instance, the following way. Fix a constant $m \geq 1$ and assume that each node gets on average $\lambda/m$ new packets per time slot. Assume that medium-access competition is done in the same way as for the single-hop system but, upon a successful transmission from node $i$, a message leaves the system with probability $1/m$ and goes to the queue of node $i+1$ with probability $1 - 1/m$. It is easy to show that the average nominal traffic load for any node is $\lambda$ and we conjecture that $\lambda < C_N/N$ is sufficient for stability for any $m \geq 1$. As $m \to \infty$ and $N \to \infty$ the model on a circle may serve as an approximation for the behaviour of a large line segment far away from source.

We can in fact prove the above multi-hop stability conjecture for $N = 4$ using a simplified version of the argument of the proof of Theorem 9. The proof however heavily relies on the specific structure of the system of 4 nodes in that at every time slot there are, essentially, at most two possible outcomes of the medium-access competition. This proof does not generalise to larger values of $N$. Stability for all $N$ and all $m \geq 1$ is a challenging and exciting question for further study.

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Consider now

\[
\frac{C_{n+2}}{n+2} - \frac{C_n}{n} = (-1)^{n-1} \frac{2^{n-3}}{(n-2)!} + (-1)^{n} \frac{2^{n-2}}{(n-1)!} - \frac{1}{n+2}(-1)^{n} \frac{2^{n-1}}{(n-1)!} + \frac{1}{n}(-1)^{n} \frac{2^{n-3}}{(n-3)!}
\]

\[
= (-1)^{n} \frac{2^{n-3}}{(n-3)!} \left( -\frac{1}{n-2} + \frac{2}{(n-2)(n-1)} - \frac{4}{(n+2)(n-1)(n-2)} + \frac{1}{n} \right)
\]

\[
= (-1)^{n} \frac{2^{n-3}}{(n-3)!} \left( -\frac{1}{n-2} + \frac{2n}{(n+2)(n-2)(n-1)} + \frac{1}{n} \right)
\]

\[
= (-1)^{n} \frac{2^{n-3}}{(n-3)!} \left( -\frac{n+1}{(n+2)(n-1)} + \frac{1}{n} \right) = (-1)^{n} \frac{2^{n-3}}{(n-3)!} \left( -\frac{2}{n(n+2)(n-1)} \right),
\]

and this completes the proof. \qed