Dynamic Time-domain Duplexing for Self-backhauled Millimeter Wave Cellular Networks

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Abstract—Millimeter wave (mmW) bands between 30 and 300 GHz have attracted considerable attention for next-generation cellular networks due to vast quantities of available spectrum and the possibility of very high-dimensional antenna arrays. However, a key issue in these systems is range: mmW signals are extremely vulnerable to shadowing and poor high-frequency propagation. Multi-hop relaying is therefore a natural technology for such systems to improve cell range and cell edge rates without the addition of wired access points. This paper studies the problem of scheduling for a simple infrastructure cellular relay system where communication between wired base stations and User Equipment follow a hierarchical tree structure through fixed relay nodes. Such a systems builds naturally on existing cellular mmW backhaul by adding mmW in the access links. A key feature of the proposed system is that TDD duplexing selections can be made on a link-by-link basis due to directional isolation from other links. We devise an efficient, greedy algorithm for centralized scheduling that maximizes network utility by jointly optimizing the duplexing schedule and resources allocation for dense, relay-enhanced OFDMA/TDD mmW networks. The proposed algorithm can dynamically adapt to loading, channel conditions and traffic demands. Significant throughput gains and improved resource utilization offered by our algorithm over the static, globally-synchronized TDD patterns are demonstrated through simulations based on empirically-derived channel models at 28 GHz.

I. INTRODUCTION

Millimeter wave (mmW) networks are an attractive candidate for beyond 4G and 5G cellular system evolution. Such systems can potentially offer tremendous increases in bandwidth along with further gains from highly directional antenna arrays [1], [2]. However, a key issue in these systems is cell coverage and range due to the extreme susceptibility of these high-frequency signals to shadowing and high isotropic propagation loss. Given these range limitations, multi-hop relaying is a natural technology in the mmW space [3], [4]. Furthermore, multi-hop relaying is particularly attractive for mmW systems since many cellular systems already use mmW backhaul and thus the addition of access links will naturally give rise to multi-hop systems.

Most mmW systems designs have assumed a time-division duplex (TDD) structure to fully exploit beamforming and eliminate the need for paired bands. In current TDD cellular standards, such as TD-LTE, all subframes are globally synchronized with base stations transmitting in one common set of DL time slots and User Equipment (UEs) in transmitting in the complementary UL set [5]. The DL/UL transmission mode patterns are essentially static and cannot be adjusted for load balancing or changing channel conditions experienced by mobile nodes. Additionally, in the current LTE-A specifications [6], in-band backhaul communication between the “donor” eNodeB and Relay Node (RN) is restricted to designated subframes, which, as we will show in Section V, could result in the wireless backhaul being severely bottlenecked. In addition, backhaul and access communication cannot occur simultaneously, meaning the system misses out on a potential multiple-access gain.

Such static, synchronized duplexing may not be necessary and, in fact, may be particularly disadvantageous for mmW systems and wireless systems that use high-gain, directional antennas. In these systems, interference from transmitters can be isolated even if there are significant power disparities (as found in [7]). In this work, we thus consider a Dynamic Time-Domain Duplexing (DTDD) scheme where individual links can select their own transmit-receive duplexing pattern. Specifically, we consider a system where subframes or slots are synchronized network-wide in time, but the transmit/receive selections in each slot can be made on a link-by-link basis – a feature uniquely available in the mmW range due to directional isolation. This flexibility enables the duplexing pattern to be dynamically optimized to current traffic loading and channel conditions. In addition, the duplexing can be adapted to local topological constraints. This adaptation is particularly valuable since the number of hops and their capacity are likely to vary significantly due to different cell sizes, propagation obstacles and availability and quality of wired backhaul.

Related work: There is now a large body of work in optimization, scheduling, power control and relay selection in OFDMA/TDD cellular networks [8], [9], [10], [11], [12], [13], including several that propose dynamic TDD algorithms designed to take advantage of the new LTE-B enhanced Interference Mitigation and Traffic Adaptation (eIMTA) capabilities [14], [15], [16]. These works focus mainly on scheduling and ICIC in the context of 4G microwave networks under the assumption of an interference-limited regime, whereas we assume constant interference due to directional isolation. We also take the approach of centralized scheduling, which is in contrast to works on distributed MAC schemes for mesh networks, as in [17]. Also, while our analysis is based on simulation, we point out that stochastic geometry analysis of self-backhaul mmW networks has recently appeared in [18] as well as a scaling law analysis in [19].
II. System Model

A. Network Topology

We consider an OFDMA-TDD cellular network with directive smart antennas and multi-hop, in-band relaying as shown in Fig. 1. We assume a tree-structured network, where an eNodeB base station (denoted BS) with wireline backhaul provides a root to which mobile N_{UE} user equipment (UE1,..., UE_{N_{UE}}) can connect via N_{RN} relay nodes RN1...RN_{N_{RN}}, which are self-backhauled, decode-and-forward stations. RNs are essentially indistinguishable in operation from the BS but utilize wireless backhaul to the primary BS, which, unlike the BS’s wired backhaul connection, is subject to the impairments of the mmW channel. For exposition, we assume a two-hop network, although all the methods apply to multihop networks as well. Each user is associated with either the BS (via direct-link) or a relay RN. The rate on the link between RN and its parent BS is denoted R_{i,j} and R_{j,i} for the downlink (DL) and uplink (UL), respectively. The rates on the DL and UL links between UEi and its parent BS or RN are likewise written R_{i,j} and R_{j,i}.

We define $\mathcal{F}$ to be the set of flows in the network. Users have one DL flow and one UL flow, which gives $N_\mathcal{F} = 2N_{UE}$ total flows in $\mathcal{F}$, but it can be easily extended to allow for multiple differentiated types of traffic. Flows have a throughput $R_f$ and utility $U_f(R_f)$. We assume a standard proportional-fair metric [20]

$$U_f(R_f) = \log(R_f),$$

although other concave utilities may also be used.

Users are scheduled in a series of epochs or frames of period $T_f$, which are further subdivided into $N_{sf}$ subframes of period $T_{sf} = T_f/N_{sf}$. From the perspective of each node, each subframe can be designated for DL or UL transmission indicating whether the transmissions are away from or toward the BS. A subframe can also be muted (i.e. unutilized). Within each subframe, we assume Orthogonal Frequency Division Multiple Access (OFDMA) is employed, allowing multiple users to be allocated orthogonal frequency resources within same time interval. Specific OFDMA parameters, discussed in Section V, are derived from the LTE-like mmW system design proposed by Samsung in [21].

In contrast to the semi-static and globally synchronized TDD configurations supported by relay-enhanced TD-LTE networks, we allow each individual BS and relay node to dynamically select the transmission mode of each subframe. The assignment is centrally-coordinated through control messaging from the BS, however the operation of the specific MAC protocol is beyond the scope of this work.

Additionally, we make the following assumptions about resource assignment and the network, in general:

i. **Half-duplex** – Transmission and reception cannot occur simultaneously (i.e. during the same subframe) between pairs of adjacent nodes.

ii. **Constant-interference** – Wireless links behave nearly like point-to-point links due to the spatial isolation of directional beams. This is the key insight that allows us to analyze each routing tree, rooted at the BS, separately, and schedule resources without need for interference coordination. Based on the findings in [7], we assume that high-gain, directional beamforming, enabled by multi-element antenna arrays, provides for minimal interference. However, receiving nodes may experience some small but non-negligible interference power. To be conservative, the interference at each node is taken to be the power assuming all nodes in the network were to be simultaneously transmitting at full power, with averaged TX/RX beamforming directions.

iii. **Single-stream** – A node can transmit to only one receiver in the same time slot and frequency resources. In later work, we shall consider BS and RN nodes capable of multi-user Space Division Multiple Access (SDMA), which can more efficiently utilize the degrees of freedom of the channel than standard OFDMA [22].

iv. **Constant-channel** – As in [7], we assume that shadowing and other large-scale channel parameters vary slower than the duration being simulated. Since resource allocation is only modeled over a small sequence of frames, these stochastic parameters remain constant for the entire sequence. Small-scale fading is accounted for in our formulation of achievable rate.

v. **Single-path and static routing** – Each RN has a single link back to the BS and UEs are associated only with the BS or RN that offers the least path loss. The topology of the network is therefore fixed over the entire sequence of frames. Furthermore, it is assumed that relays are deployed by the operator as to guarantee a minimum capacity for the mmW backhaul link between the RN and wired BS, meaning that UEs will not associate an RN that cannot provide a quality backhaul connection.
B. Channel Model

For our simulations, we adopt the distance-based path loss model for a dense, urban environment at 28 GHz developed in [7], which itself is empirically derived from extensive measurements taken in New York City [23], [24]. Each link from an access point (BS or RN) to a UE is randomly in one of three states: non-line-of-sight (NLOS), line-of-sight (LOS) or outage (i.e. the signal is blocked completely by some obstruction in the environment). For nodes with a NLOS path to their selected transmitter, i.e. the serving BS or RN, the path loss is represented by (2),

\[ PL_{dB}(d) = \alpha + \beta 10 \log 10(d) + \xi \quad \xi \sim \mathcal{N}(0, \sigma^2), \]

where \( d \) is in meters and the parameters \( \alpha, \beta \) and \( \sigma \) are provided in Table I – see [7] for details.

As earlier noted, UE nodes initially select their serving BS or RN node with the least path loss. The long-term beamforming gain is then applied, which is a function of the expected gain given the optimal TX and RX beamforming vectors, which, in turn, depends on the number of antenna elements \( (n_{TX} \text{ and } n_{RX}) \) in the \( \lambda/2 \) planar antenna array. We again defer to [7] for further details on how the long-term beamforming gain is calculated, as this subject is outside the scope of this paper.

C. SINR and Rate Calculation

As discussed in the Introduction, for the purpose of the optimization, we assume a conservative interference model where the Signal-to-Interference-Noise Ratio (SINR) from any transmitting node \( i \) to a receiving node \( j \) is computed as

\[ \text{SINR}_{ij} = \frac{P_{RX,ij}}{P_{int-max,j} + P_{N,j}}, \]

where \( P_{RX,ij} \) is the received power from node \( i \) to node \( j \) assuming the path loss and optimal long-term beamforming gain; \( P_{int-max,j} \) is the maximum average interference and \( P_{N,j} \) is the thermal noise. The maximum average interference is computed assuming that (a) all other nodes are transmitting and (b) each of the other nodes select a beamforming direction to a random node that it could possibly transmit to. When a node such as a BS or RN transmits to multiple other nodes, the transmit power is scaled by the bandwidth allocated to each of the receivers. Given the SINR and bandwidth allocation, we assume that maximum data rate for (DL or UL) link \( l_{ij} \) is calculated based on the formulation in [25], which abstracts the achievable capacity for LTE/LTE-A systems as

\[ C_{ij} = \eta W_{ij} \log_2(1 + 10^{-0.1(\text{SINR}_{ij} - \Delta_{loss})}), \]

where \( W_{ij} \) is the bandwidth allocated to the link, \( \eta \) is the bandwidth overhead and \( \Delta_{loss} \) accounts from loss from Shannon capacity due to small-scale fading, channel coding and inaccuracy of Channel State Information (CSI). Following [7], we take \( \eta = 0.8 \) and \( \Delta_{loss} = 3 \text{ dB} \).

III. Optimization Formulation

We consider the problem of maximizing the sum utility,

\[ U(\mathcal{R}) := \sum_f U_f(R_f), \]

where \( \mathcal{R} \) is the vector of the flow rates \( R_f, f \in \mathcal{F} \). We divide the frame into \( N_{sf} \) subframes, and in each subframe, \( t = 1, \ldots, N_{sf} \), and each link \( \ell \), we let \( x_f^{\ell} \in \{0, 1\} \) be the binary variable indicating whether there is a transmission on that link in that subframe. Determining the binary variables \( x_f^{\ell} \), in effect, determines the TDD scheduling pattern. To enforce the half-duplex constraint, we require that in each subframe \( t \),

\[ x_f^{\ell} + x_f^{\ell'} \leq 1, \forall \ell = (p, n), \ell' = (n, c), \]

\[ x_f^{\ell} + x_f^{\ell'} \leq 1, \forall \ell = (n, p), \ell' = (c, n), \]

for all links \( \ell = (p, n) \) and \( (n, p) \) between a node \( n \) and its parent \( p \) and \( (n, c) \) and \( (c, n) \), which are the links from \( n \) to child \( c \). In addition to the transmission schedule variables, the optimization will also allocate bandwidths \( W_f^\ell \) for each link \( \ell \) in subframe \( t \) with a maximum bandwidth constraint

\[ \sum_{\ell \in \mathcal{L}(n)} W_f^\ell \leq W_{\text{max}}, \]

where \( W_{\text{max}} \) is the total available bandwidth and \( \mathcal{L}(n) \) are the set of links adjacent to node \( n \). We additionally require that \( W_f^\ell = 0 \) when \( x_f^{\ell} = 0 \). The bandwidth allocations and spectral efficiency determine the link capacity \( C_t^{\ell} \), as in (4). We let \( R_{f,\ell,t} \) be the rate allocated to flow \( f \) on link \( \ell \) in subframe \( t \) which must satisfy the flow and capacity constraints

\[ \sum_{f \in \mathcal{F}(t)} R_{f,\ell,t} \leq C_t^{\ell}, \quad R_f \leq \sum_{t=1}^{N_{sf}} R_{f,\ell,t} \quad \forall \ell \in \mathcal{L}(f), \]

where \( \mathcal{F}(t) \) is the set of the flows over link \( t \) and \( \mathcal{L}(f) \) are the links traversed by flow \( f \).

Observe that, though typical OFDMA systems such as LTE only permit scheduling of frequency-domain resources with the granularity of discrete subcarriers or groups of subcarriers (i.e. Resource Blocks), we have assumed that an arbitrary bandwidth fraction can be allocated. After scheduling, these fractions can be rounded to the nearest integer number of subcarriers that is compatible with the numerology of the practical OFDMA system.

IV. Proposed Algorithm

The problem as formulated is a binary mixed-integer optimization, which is non-convex. A number of algorithms and heuristics have been proposed for OFDMA-TDD scheduling, many of which employ mixed-integer methods to optimize over the integral search space formed by the subcarrier and time slot indices [10], [12], [13]. These techniques are designed to deal with the computationally hard problem posed by an interference-limited network. As shall be demonstrated in the following section, two key aspects reduce the complexity of our problem. Firstly, subframe allocation can be performed individually for each BS and RN without regard to interference.
at other nodes. Secondly, the hierarchical structure of the network is exploited to identify bottlenecks and strategically reassign resources to links to incrementally improve utility.

The suboptimal, greedy algorithm in 1 operates by recursively updating slot configurations on each subtree. The procedure \texttt{TestSubtree} takes as arguments, a root node \( n \), an initial TX/RX mode indicator matrix \( X = (x^t_i) \) for all nodes \( i \) in the subtree rooted at \( n \) and an optimal utility for that allocation. Note that \( x^t_i \) represents whether a node can be transmitting \((x^t_i = +1)\) or receiving \((x^t_i = -1)\) in \( t \), which differs from the notation introduced earlier for the link allocation \( x_i \). In this scheme, the link allocation \( x_i \) is determined by whether one adjacent node is transmitting while the other is receiving.

\[
x^t_i = 1, \forall \ell = (p, n) \leftrightarrow x^t_p = +1 \land x^t_n = -1, \quad (11)
\]
\[
x^t_i = 1, \forall \ell = (n, p) \leftrightarrow x^t_p = -1 \land x^t_n = +1, \quad (12)
\]

The procedure returns a tuple \((X, U)\) of the improved duplexing schedule and utility for that subtree. The procedure operates recursively, where at each level of the tree, it attempts to allocate additional subframes in either the DL and UL (by switching the TX/RX mode of at that level) and calling the inner optimizer, here denoted as \textsf{Opt}(), to test if utility has improved. The inner optimizer utilizes the fact that, for any candidate duplexing schedule, the optimization problem is the maximization of a concave function with linear constraints.

We initially call \texttt{TestSubtree()} with the root node \( n = BS \) to perform the following depth-first procedure on the entire tree. In lines 3-10, we make a recursive call to reschedule each RN subtree with the initial \( X \). This tests the case where the current configuration is optimal for links adjacent to the root but possibly not for subtrees rooted at a lower level. We then reschedule a UL subframe in the DL (i.e. \( x_{new} = +1 \)) at the BS by calling \texttt{Realloc} in line 13. We naturally set each UE in the set of children \( C(n) \) of the root node (the BS, in this case) to RX mode in the same \( t \); otherwise, these links would be unutilized in this subframe. When \texttt{Realloc} is called on a RN, it selects the first \( t \) for rescheduling such that its transmission mode is the same as its parent, as this link was previously unutilized. If no such \( t \) exists, it simply selects the next subframe in succession. The inner optimizer is then called, which returns the intermediate utility \( U_{int} \). While \( U_{int} \) may be less than the initial utility \( U \) at this point, there is still the opportunity for it to improve by rescheduling subframes in each relay subtree (lines 18-29). Therefore, the procedure inside the \textit{while} loop continues as long as the intermediate utility is increasing \textit{within each subtree} (i.e. \( U_{int} > U \) at each level), but not necessarily globally. In the base case, where the base of the tree has been reached or we have returned from testing all subtrees, we then attempt to improve utility by rescheduling slots in the UL using the same recursive procedure.

V. SIMULATION METHODOLOGY & RESULTS

Apart from the parameters taken from [7], we base our simulations on the guidelines laid by the 3GPP and ITU-R for modeling LTE/LTE-A networks [26], [27]. We design our simulation scenarios to compare the performance of our dynamic TDD algorithm with the static TDD and relay configurations supported by LTE. For the static case, we test every valid combination of the TDD patterns in table 4.2-2 of [28] and the possible reserved eNB-RN transmission subframes in table 5.2-2 of [6]. For each static configuration, we run the inner OFDMA optimization to provide a fair comparison with the dynamic scheduler.

We test a random network topology with 2 and 4 relay nodes. For each scenario and topology, a random drop is performed where nodes are placed uniformly randomly in the area, with the constraint that RNs must be over 50m from the wired BS. Note also that for RN-to-BS links, we assume that the deployment of RNs is planned by the operator to ensure a LOS backhaul connection.

The marked gains offered by optimized DTDD scheduling over static LTE TDD reusing are shown in Figures 3a and 3b. Also from table III we see how DTDD provides a gain of about 1.5x in average downlink rate. We find an even more distinct improvement in average uplink flow rate: 2.2x and 2.9x for the 2- and 4-RN scenarios. Improvement in median rate is substantial, particularly for the uplink case, which further underscores the poor performance of statically-configured LTE relays for uplink traffic. Edge rates, defined as the lowest 5%
of rates, also improve notably in all cases. Under proportional fairness, the much lower rates of median and edge users in the static relay case result in a large increase in system utility with DTDD allocation.

We also find that, despite adding 4.76 dB of SINR on average (see Figure 2), the gains of the 4-relay system over the 2-relay system are only slight for static TDD, whereas they are much more pronounced under DTDD. This result highlights how adding relays provides little benefit when the eNB-RN links are bottlenecked, which, as we have shown in our simulations, is a characteristic weakness of LTE-style TDD scheduling for mmW relay networks under full-buffer traffic.

It is important to note that the results shown for the static LTE-relay case represent the average performance over all TDD schemes with 8 usable subframes. This does not illustrate the point that the network must initially select a single configuration, which may result in the ratio of uplink to downlink subframes and eNB-to-RN and RN-to-UE subframes being very poorly matched to the given the network loading. User data traffic in cellular networks is highly variable, so by this fact alone, the gains of DTDD should be even more significant than shown by these simulations. Also since different relays may have traffic loads that vary significantly, a single global frame configuration will likely not be appropriate for each relay and its users. In future work, we intend to demonstrate the full benefits of our scheme by developing a network model that supports mobility, with time-varying channels and traffic.

Table I: Network model parameters

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>TX power</td>
<td>$P_{TX} = 30dBm$ (BS and RN)</td>
</tr>
<tr>
<td></td>
<td>$P_{TX} = 20dBm$ (UE)</td>
</tr>
<tr>
<td>Antenna</td>
<td>$n_{ant} = 64$, $8x8$ λ/2 planar array (BS and RN)</td>
</tr>
<tr>
<td></td>
<td>$n_{ant} = 16$, $4x4$ λ/2 planar array (UE)</td>
</tr>
<tr>
<td>Noise figure</td>
<td>$N_{F_{TX}} = 5$ (BS and RN)</td>
</tr>
<tr>
<td></td>
<td>$N_{F_{TX}} = 7$ (UE)</td>
</tr>
<tr>
<td>Bandwidth</td>
<td>$W = 1000$ MHz</td>
</tr>
<tr>
<td>Carrier freq.</td>
<td>$f_c = 28$ GHz</td>
</tr>
<tr>
<td>Path loss</td>
<td>$\alpha = 72.0$, $\beta = 2.92$, $\sigma = 8.7$dB (NLOS)</td>
</tr>
<tr>
<td></td>
<td>$\alpha = 61.4$, $\beta = 2$, $\sigma = 5.8$dB (LOS) [7]</td>
</tr>
<tr>
<td>LOS-NLOS-outage prob.</td>
<td>$a_{out} = 0.0334m^{-1}$, $b_{out} = 5.2$, $\sigma_{out} = 0.0149m^{-1}$ [7]</td>
</tr>
<tr>
<td>Subframe allocation</td>
<td>Case 1: Static with reserved BS ↔ RN SFs</td>
</tr>
<tr>
<td></td>
<td>Case 2: Dynamic</td>
</tr>
<tr>
<td>Inter-site Distance (ISD)</td>
<td>RRs placed at &gt; 500m from BS</td>
</tr>
<tr>
<td>Number of nodes</td>
<td>Case 1: $n_{BS} = 1$, $n_{RN} = 2n_{UE} = 10$</td>
</tr>
<tr>
<td></td>
<td>Case 2: $n_{BS} = 1$, $n_{RN} = 4n_{UE} = 10$</td>
</tr>
<tr>
<td>Traffic</td>
<td>Full buffer</td>
</tr>
<tr>
<td>Number of drops $N_{drop}$</td>
<td>$N_{drop} = 25$</td>
</tr>
</tbody>
</table>

Table II: Mean fractions of LOS, NLOS, and outage UEs

<table>
<thead>
<tr>
<th></th>
<th>LOS</th>
<th>NLOS</th>
<th>Outage</th>
</tr>
</thead>
<tbody>
<tr>
<td>2RN 10UE</td>
<td>0.388</td>
<td>0.600</td>
<td>0.012</td>
</tr>
<tr>
<td>4RN 10UE</td>
<td>0.536</td>
<td>0.460</td>
<td>0.004</td>
</tr>
</tbody>
</table>

Table III: UE flow mean, median and cell edge (worst 5%) rate for dynamic and static TDD (units in Mbps)

<table>
<thead>
<tr>
<th></th>
<th>Mean</th>
<th>Median</th>
<th>Cell edge</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>DL</td>
<td>UL</td>
<td>DL</td>
</tr>
<tr>
<td>2RN</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DTDD</td>
<td>55.47</td>
<td>40.84</td>
<td>24.44</td>
</tr>
<tr>
<td>static</td>
<td>22.32</td>
<td>13.86</td>
<td>8.61</td>
</tr>
<tr>
<td>gain</td>
<td>1.50X</td>
<td>2.25X</td>
<td>2.84X</td>
</tr>
<tr>
<td>4RN</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DTDD</td>
<td>50.19</td>
<td>52.29</td>
<td>41.63</td>
</tr>
<tr>
<td>static</td>
<td>34.05</td>
<td>17.93</td>
<td>12.01</td>
</tr>
<tr>
<td>gain</td>
<td>1.47X</td>
<td>2.92X</td>
<td>2.47X</td>
</tr>
</tbody>
</table>

Figure 2: CDF of DL and UL SINR for non-outage UEs after beamforming with omnidirectional interference

Figure 3: CDF of flow rates
Runtime Performance and Optimality: From comparing the empirical runtime performance of our suboptimal DTDD algorithm to the optimal allocation, which is computed via brute-force search, we find that for the 2-RN/10-UE case, DTDD achieves 13% and 7% off from the optimal DL and UL mean flow rate and is within 5% of the optimal utility on average. Our heuristic took 12 iterations (of calling the inner optimizer), on-average, and a maximum of 19 iterations to complete, as opposed to 329 for brute-force (found by numerically generating and counting unique permutations). We separately perform these simulations over only 4 subframes since the set of unique configurations we must search over increases exponentially with the number of subframes and would be intractable for 8 subframes. Clearly, the number of permutations of $N_{sf}$ subframes for a single BS and $N_{RN}$ relays is upper-bounded by $\mathcal{O}(N_{sf}(N_{RN}+1))$, and while the number of unique permutations is much fewer, it still demonstrates that a brute-force approach is computationally unmanageable.

VI. CONCLUSIONS

Multi-hop relaying provides an attractive technology for extending coverage in mmW systems which are inherently limited by range and blocking. In conventional cellular systems, the full benefits of relays are difficult to realize due to the need to have global, common DL-UL subframe or slot assignments with reserved subframes for relays. However, given the advantages of directional isolation with high-gain antennas, we have argued that duplexing schedules in mmW systems can be selected individually on each link. To exploit this flexibility, we have proposed a dynamic TDD system, where duplex schedules can adapt to local channel and traffic conditions. We have formulated the selection of the duplex schedules as a joint optimization, which allocates power and bandwidth to maximize a global utility. The resulting optimization is non-convex, but we have developed a computationally efficient heuristic based on recursive searching within subtrees. Simulations of realistic deployments demonstrate considerable gains over static allocations, even under fairly uniform traffic.

REFERENCES


