Multi-User Frequency Assignment for Ultra-Reliable mmWave Two-Ray Channels

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Abstract—We consider a multi-user two-ray ground reflection scenario with unknown distances between transmitter and receivers. By using two frequencies per user in parallel, we can mitigate possible destructive interference and ensure ultra-reliability with only very limited knowledge at the transmitter. In this work, we consider the problem of assigning two frequencies to each receiver in a multi-user communication system such that the average minimum receive power is maximized. In order to solve this problem, we introduce a generalization of the quadratic multiple knapsack problem to include heterogeneous profits and develop an algorithm to solve it. Compared to random frequency assignment, we report a gain of around 6 dB in numerical simulations.

Index Terms—Ultra-reliable communications, Two-ray ground reflection, Knapsack problems, mmWave, Worst-case design.

I. INTRODUCTION

Reliability is a major requirement for many modern applications of wireless communication systems [1], [2]. In particular, this includes autonomous vehicles, e.g., self-driving cars and unmanned aerial vehicles (UAVs). It is therefore of great interest to develop techniques, which enable ultra-reliable communications. This is especially important for scenarios where only limited information, e.g., channel state information (CSI), is available at the communication parties, e.g., due to high mobility or in frequency-division duplex (FDD) systems.

Among others, it has been observed that negative dependency between channel gains can significantly improve reliability [3]–[5]. The basic idea is to establish diversity and ensure that always one communication link is available, if the others fail. In the following, we will use this idea to develop a simple frequency diversity scheme that enables ultra-reliable communications in two-ray ground reflection scenarios. In this two-ray model, it is assumed that only one significant multipath component exists in addition to a line-of-sight (LoS) connection. The second component is typically caused by a single reflection on a ground surface. This could occur in flat outdoor terrain [6], on large concrete areas, e.g., airports [7], and for a UAV flying above water [8], [9]. It has also been observed that the two-ray model can be appropriate for vehicle-to-vehicle (V2V) communication scenarios [10], [11]. This includes high frequency bands like millimeter wave (mmWave) [12]–[14].

In general, the curvature of the Earth’s surface needs to be considered for long-distance outdoor settings and accurate models like the curved-Earth model [8], [15], [16] exist. However, when considering relatively short distances, the flat-Earth model is a valid approximation [8], [16], which we adopt throughout this work.

When varying the distance between transmitter and receiver, the relative phase of the two received signal components varies and they may interfere constructively or destructively. A destructive interference causes a drop of receive power, which in turn could cause an outage of the communication link. In order to mitigate drops of the signal power on one frequency, a second frequency can be used in parallel. The use of multiple frequencies in parallel to create diversity and improve the reliability in ground reflection scenarios has already been proposed in [3] and [17]. In [18], it is analyzed and experimentally verified that frequency diversity improves the performance of distance measurements in outdoor ground reflection scenarios. Instead of using multiple frequencies in parallel, it was shown experimentally in [7] that using multiple antennas and carefully choosing the spacing between them can also improve the received power.

In practical communication systems, we are often given a fixed set of available frequencies. If there exist multiple users with unknown distances, the following questions immediately arise: how should the frequencies be assigned to the users in order to maximize the worst-case receive power? Such an optimization problem of assigning discrete items to users belongs to the broad class of knapsack problems [19].

In this work, we consider the problem of assigning each user in a multi-user communication system two frequencies such that the average worst-case receive power is maximized. Our main contributions are summarized as follows.

- We analyze the worst-case received power for a two-ray ground reflection model with unknown distance between transmitter and receiver when employing only a single frequency (Section III-A) and when employing two frequencies in parallel (Section III-B).
- We formulate the problem of assigning frequencies to users with unknown distances as a knapsack problem. (Section IV)
In order to solve this problem, we introduce a novel generalization of the standard quadratic multiple knapsack problem (QMKP) to support heterogeneous profits. We then present an algorithm to find a solution to it. (Section V)

Finally, we demonstrate the effectiveness of our proposed solution by numerical examples and report a gain of up to 6 dB compared to a random frequency assignment. (Section VI)

In [20], we analyze a similar two-ray model with two frequencies in parallel. However, we only consider a single user and focus on optimizing the frequency spacing instead of assuming a fixed set of available frequencies.

**NOTATION**

An overview of the most commonly used variable notation can be found in Table I.

| Variable | Description |
|----------|-------------|
| $U = \{1, 2, \ldots, K\}$ | Set of users |
| $F = \{f_1, f_2, \ldots, f_N\}$ | Set of available frequencies |
| $P = \{P_1, P_2, \ldots, P_K\}$ | Profit $P_u$ for user $u$ |
| $A = \{A_1, \ldots, A_K\}$ | Set $A_u$ of assigned frequencies to user $u$ |
| $d_u$ | Distance between transmitter and receiver $u$ (on the ground) [m] |
| $h_{tx}$ | Height of the transmitter [m] |
| $h_{rx,u}$ | Height of receiver $u$ [m] |
| $\ell_u$ | Length of the LoS path to user $u$ [m] |
| $\ell_u$ | Total length of reflection path to user $u$ [m] |
| $c$ | Speed of light $= 299 792 458$ m/s |
| $\omega = 2\pi f$ | (Angular) frequency $([\text{rad/s}])$ [Hz] |
| $P_i$ | Transmit power [W] |
| $P_r$ | Receive power (single frequency) [W] |
| $P_s$ | Received sum power (two frequencies) [W] |
| $P_k$ | Lower bound of the receive sum power [W] |
| $d_k$ | Distance at which the $k$-th local minimum of the receive power occurs [m] |
| $\Delta \omega = 2\pi \Delta f = \omega_2 - \omega_1$ | (Angular) frequency spacing $([\text{rad/s}])$ [Hz] |

In order to simplify the notation, we will omit variables on which functions depend when their value is clear from the context, e.g., we will write $f(x)$ instead of $f(x, y)$ when the value of $y$ is fixed.

Since the angular frequency $\omega = 2\pi f$ is a simple scaling of the frequency $f$, we will treat them somewhat interchangeably. Especially for calculations, it is more convenient to use $\omega$, while $f$ is relevant for actual system design. We will therefore use the frequency $f$ for the numerical examples while expressing all formulas in terms of the angular frequency $\omega$.

The uniform distribution on the interval $[a, b]$ is denoted as $\text{unif}(a, b)$.

**II. SYSTEM MODEL AND PROBLEM FORMULATION**

We consider a multi-user communication system with $K$ single-antenna users at unknown distances $d_u$ from the single-antenna transmitter. For each user $u$, only a range of possible distances is known, i.e., it is known that $d_u \in [d_{\min,u}, d_{\max,u}]$. An illustration can be found in Fig. 1. It should be noted that the intervals of the individual users may overlap or even be the same. The transmitter is fixed at height $h_{tx}$ above the ground while the users are located at heights $h_{rx,u}$ above the ground.

The scenario is considered to be in a flat terrain, e.g., UAVs flying above a flat water surface. Based on this, the propagation environment is approximated as a plane reflecting ground surface and therefore modeled by the classical two-ray ground reflection model [21, Chap. 4.6]. This geometrical model for a single receiver is depicted in Fig. 2.

Based on the setup, it can be seen that the transmitted signal is propagated via two separate paths to the receiver. On the one hand, there exists a LoS propagation with path length $\ell_u$. On the other hand, the signal is also reflected by the ground, which leads to the second component. The total length of the second ray is $\tilde{\ell}_u$. Finally, these two components superimpose at the receiver.

From basic trigonometric considerations, the path lengths can be calculated as

\[
\ell_u^2 = (h_{tx} - h_{rx,u})^2 + d_u^2
\]

(1)

\[
\tilde{\ell}_u^2 = (h_{tx} + h_{rx,u})^2 + d_u^2
\]

(2)
When transmitting on a single frequency $\omega = 2\pi f$, the received power $P_r$ of user $u$ at distance $d_u$ is given as [3, Eq. (2)], [21, Chap. 4.6]

$$P_r(d_u, \omega; h_{\text{TX}}, h_{\text{RX},u}, P_t) = P_t \left( \frac{c}{2\omega} \right)^2 \left( \frac{1}{\ell_u^2} + \frac{1}{\ell_t^2} - \frac{2}{\ell_u \ell_t} \cos \left( \frac{\omega}{c} \left( \ell_u - \ell_t \right) \right) \right) \Delta \phi$$

with the additional notation from Table 1. For simplicity, we assume a perfectly reflecting ground throughout this work. However, all of the calculations can be adapted to include an additional attenuation due to the absorption on the ground.

It is well-known that the two components can interfere constructively or destructively at the receiver, depending on the distance $d$. This leads to local minima in the receive power at certain distances, which in turn can lead to outages in the transmission. In order to mitigate these drops in receive power, we propose to use a second frequency in parallel. This leads to local minima in the receive power when the direct and reflected signals interfere destructively [22]. This occurs whenever the phase difference $\Delta \phi$ is a multiple of $2\pi$, i.e.,

$$\Delta \phi = \frac{\omega}{c} \left( \ell - \ell' \right) = 2\pi k, \quad k \in \mathbb{N}_0.$$  

It can easily be verified that $\Delta \phi$ is a decreasing function in $d$ and it therefore follows that

$$\Delta \phi_{\text{max}} = \lim_{d \to 0} \Delta \phi = \frac{2\omega}{c} \min \{h_{\text{TX}}, h_{\text{RX}}\}$$

and

$$\lim_{d \to \infty} \Delta \phi = 0.$$

This shows that $\Delta \phi$ decreases from a finite value $\Delta \phi_{\text{max}}$ to 0. Hence, there always exists a finite number of multiples of $2\pi$ with $k_{\text{max}} = \left\lfloor \frac{\Delta \phi_{\text{max}}}{2\pi} \right\rfloor$, i.e., there exist $k_{\text{max}}$ local minima of the receive power.

The distance $d_k$ at which the $k$-th minimum occurs, is given by solving (5) as

$$d_k^2(\omega) = \frac{\left( (c\pi k)^2 - (\omega h_{\text{RX}})^2 \right) \left( (c\pi k)^2 - (\omega h_{\text{TX}})^2 \right)}{(c\pi k)^2}$$

with $k = 1, \ldots, k_{\text{max}}$.

Example 1. An illustration of $\Delta \phi$ can be found in Fig. 3. The parameters are set to $\omega/c = 10$ (f = 477 MHz), $h_{\text{TX}} = 10$ m, and $h_{\text{RX}} = 1.5$ m. Additionally, we indicate the distances $d_k$. Since $k_{\text{max}} = 4$, there exist four $d_k$ at which a local minimum occurs. For the selected parameters, they are evaluated to $d_1 = 46.7$ m, $d_2 = 21.6$ m, $d_3 = 12.3$ m, and $d_4 = 6.5$ m. The corresponding received power from (3) is shown in Fig. 4. It can be seen that a minimum occurs at $d = d_k$ with the lowest peak being at the smallest $k$, i.e., $k = 1$, which corresponds to the highest distance of all $d_k$. For comparison, we additionally show the received power for $f_2 = 2.4$ GHz in Fig. 4.
2) Worst-Case Receive Power: When \( d \) is anywhere in the interval \([d_{\text{min}}, d_{\text{max}}]\), the lowest peak is at the smallest \( k \) such that \( d_k \) is still in \([d_{\text{min}}, d_{\text{max}}]\). However, in order to determine the global minimum of the receive power in \([d_{\text{min}}, d_{\text{max}}]\), the boundary points \( d_{\text{min}} \) and \( d_{\text{max}} \) need to be taken into account. This leads to the following result of the minimal receive power when only a single frequency is used.

**Theorem 1** (Minimal Receive Power (Single Frequency)). Consider the described two-ray ground reflection model with a single frequency \( \omega = 2\pi f \). The distance \( d \) between transmitter and receiver is in the interval \([d_{\text{min}}, d_{\text{max}}]\). The minimal receive power is then given as

\[
\min_{d \in [d_{\text{min}}, d_{\text{max}}]} P_r(d) = \min \left\{ P_r(d_{\text{min}}), P_r(d_{\text{max}}), P_r\left( \max_{d_k \in [d_{\text{min}}, d_{\text{max}}]} d_k \right) \right\} . \tag{7}
\]

**Example 2** (Single Frequency Worst-Case Receive Power). For a numerical example, we take the parameters that are used in Example 1 and additionally compare it to a higher frequency scenario. In particular, we fix \( h_{\text{Tx}} = 10 \text{ m}, h_{\text{Rx}} = 1.5 \text{ m}, \) and \( P_t = 1 \). The receiver is assumed to be randomly located at a distance between \( d_{\text{min}} = 30 \text{ m} \) and \( d_{\text{max}} = 100 \text{ m} \) from the transmitter.

For the lower frequency \( \omega_1 = 10\pi \), i.e., \( f_1 = 477 \text{ MHz} \), we get \( P_r(d_{\text{min}}, \omega_1) = -50 \text{ dB}, P_r(d_{\text{max}}, \omega_1) = -60 \text{ dB}, \) and \( P_r(d_1(\omega_1), \omega_1) = -97 \text{ dB} \) with \( d_1(\omega_1) = 46.7 \text{ m} \). Based on (7) from Theorem 1, we determine that the worst-case receive power is equal to \(-97 \text{ dB}\).

In contrast, for a higher frequency \( f_2 = 2.4 \text{ GHz} \), there are multiple local minima at locations \( d_k \), which lie in the interval \([d_{\text{min}}, d_{\text{max}}]\). According to (7), we need to determine the maximum of all \( d_k \in [d_{\text{min}}, d_{\text{max}}] \). For the considered parameters, this is calculated to \( d_3(\omega_2) = 79.4 \text{ m} \) with \( P_r(d_3(\omega_2), \omega_2) = -125 \text{ dB} \). The received powers at the boundary points are evaluated to \( P_r(d_{\text{min}}, \omega_2) = -64 \text{ dB} \) and \( P_r(d_{\text{max}}, \omega_2) = -75 \text{ dB} \). Hence, the worst-case receive power when using only frequency \( f_2 \) is \(-125 \text{ dB}\).

**B. Two Frequencies**

Since the drops in receive power due to destructive interference of the two rays cannot be avoided when a single frequency is used, we will now employ a second frequency to mitigate these minima. As described in Problem Statement 1, we aim to optimize the frequency assignment such that the minimum receive power is maximized.

The received power at distance \( d \) is given as the sum power \( P_r = P_r(d, \omega_1) + P_r(d, \omega_2) \). For a fair comparison with the single frequency case, we assume that the total transmit power \( P_t \) remains the same. Thus, only half of the transmit power is used for the individual frequencies. This leads to the expression for the total received power in (8) at the top of the next page.

Since we are particularly interested in improving the worst-case performance, we will consider a lower bound on \( P_r \) in the following.

**Lemma 1** (Lower Bound on the Sum Power for Two Frequencies). For the described two-ray ground reflection system using two frequencies \( \omega_1, \omega_2 \) in parallel, the received sum power \( P_r \), is lower bounded by \( P_r^* \) from (9) at the top of the next page, where \( \Delta \omega = \omega_2 - \omega_1 \) denotes the frequency spacing.

**Proof.** The proof can be found in Appendix A. \( \square \)

Based on this lower bound and Theorem 1, we can immediately state the worst-case receive power in \([d_{\text{min}}, d_{\text{max}}]\) when the two frequencies \( \omega_1 \) and \( \omega_2 \) are used in parallel.

**Theorem 2** (Minimal Receive Power (Two Frequencies)). Consider the described two-ray ground reflection model with two frequency \( \omega_1 = 2\pi f_1 \) and \( \omega_2 = 2\pi f_2 \). The distance \( d \)
\[ P_s = \frac{P_t}{2} \left( \frac{c}{2} \right)^2 \left[ \left( \frac{1}{\omega_1^2} + \frac{1}{\omega_2^2} \right) \left( \frac{1}{\ell^2} + \frac{1}{\ell^2} \right) - \frac{2}{\ell \ell} \left( \cos \left( \frac{\omega_1}{c} (\ell - \ell) \right) + \cos \left( \frac{\omega_2}{c} (\ell - \ell) \right) \right) \right] \] (8)

\[ P_s(d, \Delta \omega; \omega_1, h_{Tx}, h_{Rx}, P_t) = \frac{P_t}{2} \left( \frac{c}{2} \right)^2 \left[ \left( \frac{1}{\omega_1^2} + \frac{1}{\omega_2^2} \right) \left( \frac{1}{\ell^2} + \frac{1}{\ell^2} \right) - \frac{2}{\ell \ell} \left( \frac{1}{\omega_1^2} + \frac{1}{\omega_2^2} \right) \right] \] (9)

IV. FREQUENCY ASSIGNMENT AS KNAPSACK PROBLEM

Since we are now able to calculate the worst-case receive power for user \( u \) with given distance interval \([d_{\min, u}, d_{\max, u}]\) and frequencies \( f_1 \) and \( f_2 \), we now want to reformulate the original problem from Problem Statement 1 in order to solve it.

The general problem of assigning items to multiple bins (or knapsacks) is referred to as a multiple knapsack problem (MKP) [19]. Each item \( i \) has a weight \( w_i \) and each knapsack \( u \) has a weight capacity \( c_u \). When item \( i \) is placed in any knapsack, it yields the profit \( p_i \). In the QMKP, there additionally exist joint profits \( p_{ij} \) for placing items \( i \) and \( j \) in the same knapsack.

This basic setup closely resembles our considered problem of assigning frequencies to users. In particular, we have the following correspondences. The items to be assigned are the \( N \) available frequencies \( f_i, i = 1, \ldots, N \). The knapsacks, which receive the items, correspond to the individual users \( U = \{1, 2, \ldots, K\} \). Since we limit the number of frequencies that can be assigned to each user to two, we set the weight capacity \( c_u \) of each user to \( c_u = 2 \) and the weight of each frequency to \( w_i = 1 \). The profit \( p_i \) of assigning frequency \( f_i \) to a user corresponds to the minimum receive power from (7). When adding a second frequency \( f_j \), the overall profit \( p_i + p_j + p_{ij} \) should equal the minimum receive power from (10). However, it needs to be emphasized that these profits are different for each user \( u \), since the parameters \([d_{\min, u}, d_{\max, u}]\) and \( h_{Rx, u} \) may vary between users. We therefore have the heterogeneous profits

\[
p_{u,i} = \min_{d \in [d_{\min, u}, d_{\max, u}]} P_s(d, f_i) \quad (11)
\]

\[
p_{u,ij} = \min_{d \in [d_{\min, u}, d_{\max, u}]} P_s(d, f_i, f_j) - p_{u,ij} \quad (12)
\]

Since this is different compared to the standard definition of the QMKP, we generalize it to include heterogeneous profits in the following.

V. QUADRATIC MULTIPLE KNAPSACK PROBLEM WITH HETEROGENEOUS PROFITS

For the standard QMKP with homogeneous profits, there are multiple (heuristic) algorithms available to find a solution [23–27]. However, since the standard formulation does not include heterogeneous profits, these algorithms do not support it. While generalizations to the QMKP exist, e.g., considering groups of items [28, 29], they are not directly applicable to our considered problem.

Figure 5. Received power for two parallel frequencies with \( f_1 = 2.4 \) GHz, \( \Delta f = 250 \) MHz, \( h_{Tx} = 10 \) m, and \( h_{Rx} = 1.5 \) m. Both the actual value \( P_s \) from (8) and the lower bound \( P_s \) from (9) are shown. (Example 3)
Therefore, we introduce a novel generalization as quadratic multiple knapsack problem with heterogeneous profits (QMKP-HP) in the following.

**Problem Statement 2** (Quadratic Multiple Knapsack Problem with Heterogeneous Profits (QMKP-HP)). Let $\mathcal{F} = \{1, 2, \ldots, N\}$ be a set of items (or objects) and $\mathcal{U} = \{1, 2, \ldots, K\}$ be a set of knapsacks. Each item $i \in \mathcal{F}$ has weight $w_i$ and can be assigned to at most one knapsack $u \in \mathcal{U}$. Each knapsack has a weight capacity $c_u$. If item $i$ is assigned to knapsack $u$, it yields the profit $p_{u,i}$. Additionally, each pair of items $i, j \in \mathcal{F}$, $i \neq j$, has a joint profit $p_{u,ij}$, if both items are assigned to the same knapsack $u$. The set $A_u \subset \mathcal{F}$ describes the allocation of items to knapsack $u$.

The objective of the QMKP-HP is to find allocations $A = \{A_1, \ldots, A_K\}$ such that the overall profit is maximized,

$$\max \sum_{u \in \mathcal{U}} \left( \sum_{i \in A_u} p_{u,i} + \sum_{j \in A_u \setminus i} p_{u,ij} \right)$$  \hspace{1cm} (13a)

s.t. \hspace{1cm} \sum_{i \in A_u} w_i \leq c_u \quad \forall u \in \mathcal{U}  \hspace{1cm} (13b)

$$K \sum_{u=1}^{K} a_{iu} \leq 1 \quad \forall 1 \leq i \leq N$$  \hspace{1cm} (13c)

where $a_{iu} = 1$ if item $i$ is assigned to user $u$, i.e., if $i \in A_u$, and $a_{iu} = 0$ otherwise.

**Remark 1.** When the profits $p_u$ of the individual knapsacks are all the same, i.e., $P_1 = P_2 = \cdots = P_K$, the QMKP-HP reduces to the standard QMKP.

In the following, we develop a greedy algorithm to solve Problem (13). The algorithm is inspired by the constructive procedure from [24] for the classical QMKP, cf. [24, Alg. 1].

As an essential guideline in the algorithm, we use the common notion of value density [24], [27]. This quantity describes the value of adding a particular item $i$ to a knapsack $u$ if it already has items $S$ assigned to it, i.e.,

$$v_{d,u}(i, S) = \frac{1}{w_i} \left( p_{u,i} + \sum_{j \in S \setminus i} p_{u,ij} \right).$$  \hspace{1cm} (14)

For a more convenient notation, we define its matrix form as

$$vd(S) = \begin{pmatrix}
vd(1,S) & vd(2,S) & \cdots & vd(K,S) \\
vd(1,S) & vd(2,S) & \cdots & vd(K,S) \\
vd(1,S) & vd(2,S) & \cdots & vd(K,S) \\
vd(1,S) & vd(2,S) & \cdots & vd(K,S)
\end{pmatrix}. $$  \hspace{1cm} (15)

After an initialization, the main idea of the proposed Algorithm 1 at the bottom of the next page is the following. We start with the combination of item $i$ and knapsack $u$ which yields the largest value density. If there is still room for item $i$ in knapsack $u$, i.e., if $w_i \leq c_u$, we add the item to the knapsack. If there is no room for it, we proceed with the next best combination of item and knapsack, until we can assign an item to a knapsack. After an item has been assigned, it is removed from the list of unassigned items and the value densities for the remaining ones are updated. These steps are repeated until there are no more items to assign or all remaining items’ weights exceed the remaining capacities of all knapsacks.

**Remark 2.** Algorithm 1 can also be used to complete a partial solution $A^{(0)}$. We simply need to take $A^{(0)}$ into account while initializing the set of unassigned items. If no initial assignments exist, we set $A^{(0)} = \emptyset$.

VI. Numerical Examples

In this section, we demonstrate the effectiveness of our proposed solution to Problem Statement 1 by numerical examples. The source code to reproduce all of the following results can be found at [30]. All simulations are run in Python on an off-the-shelf laptop computer with Intel i7-8565 CPU.

We compare different problem sizes with varying numbers of users $K$ and frequencies $N$. For each setup, we uniformly select $N$ frequencies in the interval $2.4 \text{ GHz}$ to $2.5 \text{ GHz}$. The height of the transmitter is fixed to $h_{TX} = 10 \text{ m}$. We then randomly generate $K$ user parameters with $h_{Rx,u} \sim \text{unif}[1, 3] \text{ m}$, $d_{\text{min},u} \sim \text{unif}[20, 40] \text{ m}$, and $(d_{\text{max},u} - d_{\text{min},u}) \sim \text{unif}[10, 100] \text{ m}$. While we assume $N \geq 2K$ in our examples, this is not necessary. However, in the case of $N < 2K$, not every user gets two frequencies assigned.

Table II shows the performance of the proposed QMKP-HP approach for different problem sizes. The reported run times for Algorithm 1 are averages over 100 trials with minimum and maximum in parentheses. The last two columns indicate the achieved average worst-case receive powers, i.e., the objective of (4) divided by the number of users $K$. For comparison, we additionally evaluated the performance for a random frequency assignment and three round-robin (RR) schemes. The first one (RR-simple) starts by assigning user 1 $f_1$, user 2 $f_2$ and continues for each user. In a second round, each user gets assigned a second frequency, i.e., at the end, user $u$ is assigned $f_u$ and $f_{u+N}$. The second RR scheme (RR-block) is similar to the first one but assigns two frequencies to each user at once, i.e., user $u$ gets frequencies $f_{2u-1}$ and $f_{2u}$. As a final comparison, we add a RR scheme that takes the user profits $P_u$ into account (RR-profits). With this, the users take turns selecting the frequency that yields the highest profit for them.

From the results in Table II, it can clearly be seen that the proposed algorithm consistently improves the average worst-case receive power by $5 \text{ dB}$ to $6 \text{ dB}$ over a random frequency assignment. Additionally, it can be run in feasible times on standard hardware, even for larger problem sizes of 100 frequency blocks and 45 users.

In comparison to the RR schemes, the proposed QMKP-HP approach also shows gains in performance. The worst performance is found for RR-block, since each user is assigned neighboring frequencies, which does not effectively leverage the benefit of using two frequencies in parallel. The RR-simple scheme improves this in general. However, depending
on the number of available frequencies and users, the assigned
frequencies can still be very close together. In contrast, the
profit-aware RR algorithm achieves results close to the QMKP-
HP approach. However, there is still a gap of up to 2.8 dB, which
is maximal in the case where $N$ is close to $2K$. In these
scenarios it can happen in the RR scheme that users take
frequencies before others, which would have yielded higher
benefits for the overall system.

**VII. CONCLUSION**

In this work, we have considered the problem of assigning
frequencies to users in a two-ray ground reflection scenario
with unknown distances between transmitter and receivers.

We have investigated the receive power in such systems and
shown that using two frequencies in parallel can significantly
improve the worst-case receive power over using only a
single frequency. In order to solve the optimization problem
of assigning up to two frequencies to users for worst-case
design, we introduced a generalization of the QMKP with
heterogeneous profits together with an algorithm to solve it.
Finally, we have illustrated the effectiveness of our approach
by numerical examples.

While only two parallel frequencies are considered in this
work, it could be extended to multiple frequencies in future
work. This is a promising research direction to further improve
the reliability of the described communication systems with
limited knowledge at the transmitter.

**APPENDIX A**

**PROOF OF LEMMA 1**

The lower bound on $P_s$ from (8) is calculated as the (lower)
envelope of the function, which is given as the absolute value
of its analytic function [31]. However, we are only interested in
bounding the oscillating part of $P_s$ given by the cosine terms
as

$$
\frac{\cos \left(\frac{\omega_2}{\ell} (\ell - \ell)\right)}{\omega_1^2} + \frac{\cos \left(\frac{\omega_2}{\ell} (\ell - \ell)\right)}{\omega_2^2}.
$$

| $K$ | $N$ | Time Alg. 1 [s] | Alg. 1 [dB] | Random [dB] | RR-Simple [dB] | RR-Block [dB] | RR-Profits [dB] |
|-----|-----|----------------|-------------|-------------|--------------|--------------|----------------|
| 3   | 10  | 0.0093 (0.0089–0.0117) | −82.14      | −86.87      | −90.00       | −99.40       | −83.17         |
| 3   | 100 | 0.8694 (0.7691–0.9412) | −81.36      | −87.23      | −108.73      | −113.03      | −81.11         |
| 10  | 100 | 2.9911 (2.5696–3.3221) | −81.17      | −87.21      | −99.44       | −112.60      | −81.61         |
| 20  | 50  | 1.4986 (1.2925–1.6559) | −81.37      | −87.25      | −87.38       | −110.41      | −83.84         |
| 20  | 100 | 6.1032 (5.1878–6.4554) | −81.13      | −87.58      | −93.47       | −112.71      | −82.21         |
| 45  | 100 | 14.4117 (11.9171–14.7878) | −81.69      | −87.56      | −86.62       | −112.70      | −84.51         |

**Algorithm 1** Constructive procedure for solving the QMKP-HP from Problem Statement 2

**TABLE II**

**MINIMUM RECEIVE POWER COMPARISON FOR THE FREQUENCY ASSIGNMENT PROBLEM (AVERAGED OVER 100 TRIALS)**

1: **Input:** Items $\mathcal{F}$, Users $\mathcal{U}$, Profits $\mathcal{P}$, Initial Assignments $\mathcal{A}^{(0)}$
2: **Output:** Final Assignments $\mathcal{A}$
3: **procedure** CONSTRUCTIVEPROCEDURE($\mathcal{F}$, $\mathcal{U}$, $\mathcal{P}$, $\mathcal{A}^{(0)}$)
4: Initialize unassigned items $\mathcal{S} = \mathcal{F} \setminus \bigcup_{u} \mathcal{A}^{(0)}$
5: Initialize assignments $\mathcal{A}_u = \mathcal{A}^{(0)}_u, \forall u \in \mathcal{U}$
6: Calculate initial value densities $V = \text{vd}(\mathcal{S})$
7: **repeat**
8: \begin{align*}
   V = \text{SORTDESCENDING}(V) & \quad \text{Sort } V \text{ in descending order and store it in } \tilde{V}, \text{ i.e., } \max V_{iu} = \tilde{V}_1 \geq \tilde{V}_2 \geq \ldots \\
   \text{Initialize counter } j = 1 &
\end{align*}
9: **repeat**
10: \begin{align*}
   \tilde{V}_j &= V_{iu} & \quad \text{if } w_i \leq c_u \\
   \mathcal{A}_u &= \mathcal{A}_u \cup \{i\} & \quad \text{Assign element } i \text{ to user } u \\
   \mathcal{S} &= \mathcal{S} \setminus \{i\} & \quad \text{Remove element } i \text{ from set of unassigned items } \mathcal{S} \\
   c_u &= c_u - w_i & \quad \text{Adjust the remaining capacity } c_u \text{ of knapsack } u \\
   j &= j + 1 & \quad \text{Increase counter and proceed with value next in size}
\end{align*}
11: **until** (An item is assigned to a user or $j = |\tilde{V}|$)
12: \begin{align*}
   V_{iu} &= \text{vd}_{iu}(i, \mathcal{A}_u), \forall u \in \mathcal{U}, i \in \mathcal{S} & \quad \text{Update value densities for the remaining items} \\
   \text{until } (S = \emptyset \text{ or } c_u < w_i, \forall u \in \mathcal{U}, i \in \mathcal{S}) & \quad \text{All items are assigned or no more space in knapsacks}
\end{align*}
13: **return** $\mathcal{A} = \{\mathcal{A}_1, \ldots, \mathcal{A}_K\}$
14: **end procedure**
The analytic function of $s$ is given as $s + j\hat{s}$, where $\hat{s} = \mathcal{H}\{s\}$ is the Hilbert transform of $s$. The envelope of $s$ is then given as $|s + j\hat{s}|$. With the correspondence $\mathcal{H}\{\cos(\omega t)\} = \sin(\omega t)$ [32], we obtain the analytic signal

$$s + j\hat{s} = \frac{\cos(\omega_1 t)}{\omega_1^2} + \frac{\cos(\omega_2 t)}{\omega_2^2} + j \left( \frac{\sin(\omega_1 t)}{\omega_1^2} + \frac{\sin(\omega_2 t)}{\omega_2^2} \right)$$

where we use the shorthand $t = \frac{\ell - \ell}{c}$. The absolute value can then be calculated as

$$|s + j\hat{s}|^2 = \frac{1}{\omega_1^2} + \frac{1}{\omega_2^2} + 2 \frac{\cos (\omega_2 - \omega_1) t}{\omega_1^2 \omega_2^2}.$$

Applying the definitions of $t$ and $\Delta \omega = \omega_2 - \omega_1$ and substituting the envelope of $s$ into (9), we obtain (9).

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