

Compromise in CDMA Network Planning

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Abstract

CDMA network planning, for example in 3G UMTS networks, is an important task whether for upgrading existing networks or planning new networks. It is a time consuming, computationally hard, task and generally requires the consideration of both downlink and uplink requirements. Simulation experiments presented here suggest that if time is a major consideration in the planning process then as a compromise only uplink needs to be considered.

Keywords: CDMA, Network Planning, Optimization, Simulation

1. Introduction

The past decade has seen the emergence of many computational approaches for cellular network site selection, configuration and dimensioning. Many of these contributions have paid attention to planning wide-area FTDMA systems such as second generation GSM where planning is generally carried out using a downlink transmission model and independent criteria for coverage and capacity, e.g [1].

A number of researchers have considered the rather more complex problem of network planning for UMTS networks. Amaldi *et al* ([2-5]) propose a mathematical programming model which accounts for both the uplink and downlink directions as well as for base station configuration issues including location, height, tilt and azimuth. To allow solutions to be sought in reasonable time, approximate solutions are sought via application of the tabu search meta-heuristic. In [6] and [7], Zhang, Yang, *et al.* propose a mathematical framework for UMTS network planning that considers fast power control, soft handover and pilot signal power in the uplink and downlink directions. Again, solutions are sought via the application of meta-heuristics (SA and evolutionary SA). Ben Jamaa *et al.* ([8,9]) propose an approach which employs a multi-objective GA (MOGA) to simultaneously optimize capacity and coverage by adjusting antenna parameters and common channel transmitted powers (antenna locations fixed). A multi-objective fitness function is employed which can consider objectives such as coverage, capacity and cost. The result of the MOGA is a Pareto set of non-dominated solutions.

For third generation systems such as UMTS, the planning

problem is significantly more complex than for FTDMA systems due to the dependency between capacity and coverage. The underlying CDMA protocol requires that on each link, a target signal to noise ratio (SNIR) is maintained, and consequently per-link power allocation is required before user service coverage and cell load can be accurately assessed. However, determining this is non-trivial as one user transmission is seen as interference by all other users, making coverage/capacity evaluation sensitive to other users. Transmission power minimization is important and the real-time UMTS system achieves this by fast power control. However for modeling purposes, this is costly to repeatedly simulate because all links are required to frequently re-evaluate their SNIR and adjust power accordingly.

Whitaker R. M. *et al* describe two efficient heuristic algorithms that enable the evaluation of service coverage and cell loading in both the uplink and downlink directions. In this paper, we investigate the application of these heuristics to the problem of cell planning for UMTS networks. The cell planning problem (CPP) is concerned with the selection of *antennae* from a set of *candidate antennae*, and the configuration of these antennae, such that an optimal configuration is achieved. As for the frequency assignment problem (FAP), the CPP has *NP*-complete computational complexity. This dictates that exact solutions to the CPP cannot be attained in practice. Hence we consider a meta-heuristic optimization approach.

The remainder of this paper is organized as follows. Section 2 describes the model and Section 3 provides a brief overview of the uplink and downlink service coverage/load evaluation heuristics. Section 5 outlines the

optimization problem and the meta-heuristic employed. A number of test problems are defined in Section 5 and the results and analysis of applying our optimization approaches to these problems can be seen in Section 6.

2. Model

The uplink (UL) and downlink (DL) dedicated channels and the pilot signal is included in our model. Parameters are described in **Table 1**, **Table 2** and **Table 3** and are

defined relative to the link direction under consideration.

The terms *cell*, *antenna* and *transmitter* are used interchangeably when describing aspects of coverage. A number of candidate antenna locations are defined for a given region. A planning/optimization process is employed to select and configure antennae based on defined objectives. Discrete *test points* from the region are used to sample service coverage. Each test point is a physical position (expressed in two dimensional Cartesian co-ordinates). Two types of test point are defined in our

Table 1. Global parameters.

Symbol	Description
W	CDMA chip rate.
R_i	Data rate for service i .
S_A	Set of all antennas in the working region.
S_{stp}	The set of covered pilot test points.
S_{stp}	The set of service test points.
O_{stp}	An ordering of the stp .

Table 2. Uplink parameters.

Symbol	Description
p_{xy}^{UL}	Received power from stp x at a cell y .
I_{own}^{UL}	Total received power from stp active in cell y .
I_{oth}^{UL}	Total received power from stp active in cells other than y .
I_y^{UL}	Total received power from all active stp .
N	Noise power seen at the antennas receiver in an empty cell.
$(E_b/N_o)_{UL}^*$	Target threshold for E_b/N_o ratio at an stp for the dedicated UL channel (service dependent).
$\eta_{UL,y}$	Uplink load at cell y .

Table 3. Downlink parameters.

Symbol	Description
I_{own}^{DL}	Total power received from serving cell (all links and pilot).
I_{oth}^{DL}	Total power received from all cells other than the serving cell.
α	Orthogonality Factor.
P_n^{DL}	Noise power (thermal and equipment) seen at a test point.
p_{xy}^{DL}	Power allocated by cell y for stp x as received at stp x .
p_{xy}^{pilot}	Pilot power from cell y as received at stp x .
$(E_c/I_o)_{pilot}^{DL}$	Target threshold for pilot E_c/I_o ratio.
$(E_b/N_o)_{UL}^*$	Target threshold for E_b/N_o ratio at an stp for a dedicated DL channel (service dependent).
$\eta_{DL,y}$	Downlink load at cell y .
$PtxTotal_y$	Total of allocated transmit powers in cell y .
$Ptxmax_y$	Maximum transmit capability of cell y .
$\eta_{pilot,y}$	Proportion of $Ptxmax_y$ allocated for pilot signal at cell y .

model: service test points (stp) and pilot test points (ptp). The ptp are used to assess pilot signal quality. At an stp, quality of both UL and DL dedicated channels are assessed for a particular service, which is defined prior to evaluation.

2.1. Test Point Coverage and Cell Load

The pilot signal is transmitted at a proportion $\eta_{pilot,y}$ of the maximum cell power. A ptp x is served by antenna y when the received energy per chip relative to the total spectral density E_c/I_o at least meets the target $E_c/I_o|_{pilot}$. Letting $I_y = I_{own} + I_{oth}$, then x is served if and only if:

$$\frac{P_{xy}^{pilot}}{N+I_y} \geq (E_c/I_o)_{pilot} \quad (1)$$

An stp is *covered* in a particular link direction if energy per bit relative to spectral noise density (Eb/N_0) at least meets the required target threshold. For an stp x connected to antenna y , x is *UL covered* if and only if:

$$\frac{W}{R_i} \cdot \frac{P_{xy}^{UL}}{I_y - P_{xy} + N} \geq (Eb/N_0)_{UL}^* \quad (2)$$

In the downlink, for an stp x and serving antenna y , x is *DL covered* if and only if:

$$\frac{W}{R_i} \cdot \frac{P_{xy}^{DL}}{I_{own}(1-\alpha) + I_{oth} + P_n} \geq (Eb/N_0)_{DL}^* \quad (3)$$

There are various ways in which cell loading can be assessed. Wideband power-based measurement is used in this model because it directly identifies the resources being allocated. The downlink load at cell y is estimated by:

$$\eta_{DL,y} = \frac{P_{txTotal,y}}{P_{txmax,y}} \quad (4)$$

while the uplink load at cell y is estimated by:

$$\eta_{UL,y} = \frac{I_y}{I_y + N} \quad (5)$$

Note that a ptp's ability to be served depends on downlink cell load. Consequently a ptp is covered if and only if it is served when all cells y are operating at maximum permitted downlink load $\eta_{DL,y}^{\max}$. Covered ptp can see the pilot signal independent of traffic and are collectively denoted S_{ptp} .

To ensure that an stp can see the pilot signal, it is required that $S_{stp} \subseteq S_{ptp}$. A list O_{stp} of the set S_{stp} is also required to specify the order in which stp are prioritized for admission. The ordering is defined based on the received signal strength from the best serving antenna with those with the strongest signal given priority.

3. Evaluation Heuristics

Calculating off-line transmission power for target Eb/N_0

attainment on a link requires knowledge of interference levels or equivalently cell loads. However, interference/cell loads depend on per-link transmission powers. This dependency has led to the analytical characterization of the problem [11]. We employ an algorithmic approach which initially over-estimates interference/cell loading and then uses a feedback mechanism to iteratively update and reduce the conservative error. When this feedback mechanism is applied, the heuristic can converge to a state where inaccuracy in power allocation and cell loading is negligible. From this, stp coverage and cell loads can be directly obtained.

Detailed discussion of the uplink and downlink evaluation heuristics used here can be found in [10].

4. Optimization Problem

It is assumed that for optimization, the objective is to select/activate and configure (where appropriate) a subset of antennae from the set of candidate antennae such that coverage is maximized for a specified number of active transmitters. After experimentation with a number of meta-heuristics it was determined that tabu search (TS) was the most effective approach for this optimization problem. The TS algorithm employed is summarized in **Figure 1**. A detailed description of the TS meta-heuristic can be found in [12].

The operation of our tabu search approach can be characterized by the following components: starting configuration; moves; evaluation type and cost function.

4.1. Starting Configuration

The starting configuration can impact on the final configuration achieved by the TS. Having investigated a number of starting configurations (*i.e.*, all transmitters inactive, all transmitters active, random transmitters active and Halton configuration - approximately random uniformly distributed) it has been shown that whilst the

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Generate starting configuration.
Evaluate cost of starting configuration.
Set best_cost_so_far = 0.
Initialise memory structures.
FOR i = 0 to max_iterations DO
    Evaluate all possible moves.
    Sort moves on cost (prioritization).
    Accept first move where move is non-tabu or is tabu but meets
    aspiration criteria.
    Update memory structures.
    IF cost of configuration after move improves upon
    best_cost_so_far
        THEN
            Set best_configuration = current configuration.
            Set best_cost_so_far = cost of current configuration.
        END IF
    END FOR

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Figure 1. Generic tabu search algorithm

effectiveness of each starting configuration is dependent on the problem scenario and other parameter settings, starting with all transmitters inactive leads to the best solutions in general.

4.2. Moves

A range of different moves are employed by the TS. At each iteration of the TS, the impact of each of the available moves is evaluated by applying the move to each candidate antenna in order to determine the best possible move at that instance. The quality of each move is determined by the cost function. A range of moves have been implemented from which a subset of moves can be selected for evaluation:

- Activate an inactive transmitter *i.e.* make operational.
- Deactivate an active transmitter *i.e.* shut down.
- Swap transmitters:
- Deactivate an active transmitter and activate a randomly selected inactive transmitter.
- Activate an inactive transmitter and deactivate a randomly selected active transmitter.
- Determine the best azimuth for a transmitter - the azimuth for a given transmitter is varied (controlled by *azimuth_increment*) such that all available azimuth configurations in the sector are evaluated. The configuration with the best cost can then be determined.
- Determine the best tilt for a transmitter - the tilt for a given transmitter is varied (controlled by *tilt_increment*) such that all available tilt configurations in the range *tilt_max* to *tilt_min* are evaluated. The configuration with the best cost can then be determined.

4.3. Evaluation Type

When evaluating moves, cost functions are employed in conjunction with an evaluation type. The evaluation type defines the evaluation heuristic used to determine service (*e.g.* uplink or downlink evaluation heuristic) and any constraints on the feedback mechanism employed to determine total load. In its least constrained form the iterative feedback mechanism repeats until the variation in the load is less than a predefined threshold. Iterating to convergence may require much iteration. This is time consuming and additional iterations achieve decreasing returns. As a result, in order to achieve an acceptable runtime for the TS (which need to perform large numbers of evaluations) the number of feedback iterations is constrained when employed for evaluating moves, only running to convergence for the final TS solution.

4.4. Cost Function

TS requires that a cost is associated with problem configurations such that an optimal or near optimal configuration can be sought. A weighted cost function is

employed to enable the following objectives to be considered:

- 1) Meet the constraint on the number of transmitters.
- 2) Maximise coverage.
- 3) Favour configurations with lower total loads.

The tabu search seeks to minimize the total cost of a configuration, defined as:

$$\text{total_cost} = (\text{covg_cost} * w_c) + (\text{actv_cost} * w_a) + (\text{ld_cost} * w_l) \quad (6)$$

where w_c , w_a and w_l are weightings for *coverage cost*, *active cost* and *load cost* respectively.

Coverage cost is defined as

$$\text{covg_cost} = 100 - \text{coverage} \quad (7)$$

where *coverage* is the percentage of *stp* that are covered in the downlink or uplink direction.

Active cost is defined as

$$\text{actv_cost} = \text{trans_thrs} - \text{active_trans} \quad (8)$$

where *trans_thrs* is the desired number of active transmitters and *active_trans* is the number of active transmitters.

Load cost is defined as

$$\text{ld_cost} = \sum \eta_{DL} \quad (9)$$

for all active transmitters in the downlink direction and

$$\text{ld_cost} = \sum \eta_{UL} \quad (10)$$

for all active transmitters in the uplink direction.

It should be noted that as some objectives are competing (*e.g.* coverage and total load) it may not be possible to determine the configuration which exhibits the optimal trade-off between objectives. This would require a more complex (and time consuming) multi-objective approach.

5. Experimentation

The purpose of experimentation is to compare the performance of network configurations generated using uplink and downlink evaluation heuristics. Performing optimization for both link directions is time consuming. Consequently, it would be useful if we could identify an optimization configuration that provides a good trade-off between optimizing for uplink and for downlink, *i.e.*, a single approach that produces network configurations that perform well in both link directions. Whilst this trade-off may not be acceptable when producing final configurations, it could be beneficial in preliminary stages of network planning where some accuracy can be traded for the decreased evaluation time associated with optimizing for a single link direction.

5.1. Test Problems

All experiments consider a 3km x 3km transmission region containing 36 directional candidate antennae lo-

cated at 12 uniformly distributed sites. Eight different problem scenarios have been considered as summarized in **Table 4**. Each scenario consists of a number of uniformly distributed *stp* with defined service requirements. Different scenarios have been generated by varying the number of *stp* considered and the distribution of services over these *stp*. Scenarios 5a, 5b and 5c have the same number of *stp* and number of *stp* with each service requirement, but a different distribution of these services over the *stp*. Signal attenuation is defined by the Hata path loss model.

5.2. TS Configuration

The tabu search was constrained to run for a maximum of 400 iterations and terminate after 50 iterations in which there is no improving move.

For each scenario and evaluation heuristic, two sets of moves are employed:

- 1) Activate/deactivate transmitter and swap transmitter activity (AS).
- 2) Activate/deactivate transmitter, swap transmitter activity, best tilt and best azimuth (ASBTBA)¹.

These sets of moves were selected in order to investigate the impact of tuning the configuration of the antennae.

6. Results

In this section we present the results of optimization for each problem scenario. For each problem scenario, a number of *problem instances* are considered, each with different constraints on the maximum number of antennae allowed. Four *optimization approaches* are considered:

- 1) DL optimisation heuristics with AS
- 2) DL optimisation heuristics with ASBTBA
- 3) UL optimisation heuristics with AS
- 4) UL optimisation heuristic with ASBTBA

On completion of the TS the resulting configuration is evaluated for the opposite link direction, i.e.:

- DL optimize and evaluate configuration for DL and UL percentage coverage/service, and load.

- UL optimize and evaluate configuration for DL and UL percentage coverage/service, and load.

This enables us to determine how well configurations optimized in one link direction perform with respect to the opposite link direction. Further analysis is undertaken to determine:

1) *Coverage Difference* ('Covg Diff' column in the tables) - the difference between uplink and downlink coverage for each instance of a problem scenario. This gives an indication of how optimizing for one link direction impacts on the other.

2) *Maximum DL Coverage Difference* ('Max DL Covg Diff' column) - for each instance of a problem scenario the *maximum percentage DL coverage* ('Max % DL Covg' column) for all optimization approaches is identified i.e. the maximum percentage DL coverage obtained from the DL or UL optimized AS or ASBTBA method. From this, the *Maximum DL Coverage Difference* is determined (by subtracting the coverage obtained for a problem instance from the maximum percentage DL coverage value), i.e. this indicates how well an optimization approach performs with respect to the best result. Consequently, this value gives an indication of which optimization approach performs best for each problem instance, and over all problem instances (based on the mean value).

3) *Maximum UL Coverage Difference* ('Max UL Covg Diff' column) - as Maximum DL Coverage Difference, but in the uplink direction.

6.1. Sample Results

Due to the volume of results generated from experimentation, only a subset of results is presented here². For Problem 1, the results of all optimization approaches i.e. AS and ASBTBA are included (see **Tables A1 to A4** in Appendix A). For other problems, the results for optimization approach ASBTBA are presented only (see **Tables**

Table 4. Problem scenarios.

Scenario	No. stp	No. stp assigned per service type						Total Capacity Req (kbps)
		Pilot	12.2 kbps	64kbps	144 kbps	384 kbps		
1	441	100	220	44	44	23		20,668
2	441	147	0	294	0	0		18,816
3	441	392	0	0	49	0		18,816
4	961	630	220	44	44	23		20,668
5a, 5b, 5c	961	299	440	88	88	46		41,336
6	3721	1116	1675	372	372	186		169,235

¹For best tilt and best azimuth moves, an increment of 1 degree is applied.

²A complete set of results can be found in Appendix B at: www.cs.cf.ac.uk/bounds/documentation.htm

Table 5. Analysis summary.

Link Direction/ Optimisation Method	Highest Mean DL Covg	Highest Mean UL Covg	Lowest Mean Covg Difference	Lowest Mean Max DL Covg Differnce	Lowest Mean Max UL Covg Difference	Lowest Combined Mean Covg Difference
DL AS	1		3	1		
DL ASBTBA	2, 3, 4, 5a, 5b, 5c, 6			2, 3, 4, 5a, 5b, 5c, 6		
UL AS		3		3		2
UL ASBTBA		1,2,3,4,5a, 5b, 5c, 6	1, 2, 4, 5a, 5b, 5c, 6		1, 2, 3, 4, 5a, 5b, 5c, 6	1, 3, 4, 5a, 5b, 5c, 6

A5 to A18 in Appendix A) as these performs best in general (see Section 6.2).

6.2. Analysis of Results and Conclusions

In order to compare the effectiveness of the different optimization approaches, we examine for each scenario which approach gives the best performance with respect to a number of metrics:

- 1) highest mean DL % coverage ('Highest Mean DL Covg' column);
- 2) highest mean UL % coverage ('Highest Mean UL Covg' column);
- 3) lowest mean coverage difference ('Lowest Mean Covg Difference' column), i.e. for each scenario which method gives the lowest *Coverage Difference* value;
- 4) lowest mean maximum DL coverage difference ('Lowest Mean Max DL Covg Difference' column) i.e. for each scenario which method gives the lowest mean *Maximum DL Coverage Difference* value;
- 5) lowest mean maximum UL coverage difference ('Lowest Mean Max UL Covg Difference' column), i.e. as above but defined for UL, and
- 6) lowest combined (i.e., DL and UL) mean coverage difference ('Lowest Combined Mean Covg Difference' column) i.e. for each scenario which method gives the lowest value when adding the maximum DL and UL coverage difference means.

The results of this summary analysis can be seen in **Table 5** which indicates the optimization method/ scenario combination that gives the best result for each of the above six metrics. As expected, the results show that including antenna configuration moves during optimization leads to increased coverage³. The results show that the DL ASBTBA optimization approach generally leads to the best downlink coverage and therefore the best i.e. lowest mean maximum DL coverage difference. Similarly the UL ASBTBA approach is consistently the best in terms of uplink.

Furthermore UL ASBTBA leads to the best (lowest) combined mean coverage difference i.e. for scenarios 1,3,4,5a,5b,5c and 6. This indicates that although optimization using the UL ASBTBA approach does not lead

to the best levels of downlink coverage (though it is competitive in many places as illustrated by low mean maximum DL coverage difference values) it does result in the best overall combined mean coverage difference indicating that it performs better in terms of DL coverage than DL optimization does in terms of UL coverage. As a result, where time for planning a network is limited, the experimental results presented here suggest that a compromise in many cases is to optimize for uplink only rather than optimizing in both the uplink and downlink directions.

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³This is also confirmed by Amaldi *et al* in [2] who have also shown that it is preferable to simultaneously optimize antenna location and configuration than to do so separately.

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