University of Würzburg Institute of Computer Science Research Report Series

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Report No. 268

February 2001

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#### Abstract

Third generation (3G) wireless systems will be operating with Wideband Code Division Multiple Access (WCDMA) over the air interface. This access technology requires a tight power control by the base station on the uplink in order to reduce the interference in the cell and thus increase the capacity. Furthermore, with the introduction of 3G systems, data services will be offered that permit data rates of up to 2 Mbps and varying quality of service requirements in terms of received bitenergy-to-noise ratio. In this paper we will present a performance evaluation of a WCDMA system that is based on a simulation of the power control loops on the uplink in order to investigate the influence of the traffic given by the user distribution and traffic mix on the interference in a cell. Furthermore, we will show that it is possible to use approximations of the power control loops in order to evaluate more complex scenarios.

## **1 INTRODUCTION**

Wideband Code Division Multiple Access (WCDMA) technology will be the most prominent access technology over the air interface for the upcoming third generation wireless systems (3G) [1]. Contrary to second generation networks which are rather limited to voice services, 3G systems like the Universal Mobile Telecommunication System (UMTS) in Europe are dedicated to offer data services with high bit rates and flexible capabilities. Another difference is that the cell capacity is no longer a fixed term, but also depends on the spatial distribution of the users and their individual services [2]. Thus, the capacity of a UMTS system is limited by the interference caused by each user, his maximum transmit power, and the type of service given by the data rate and the activity factor.

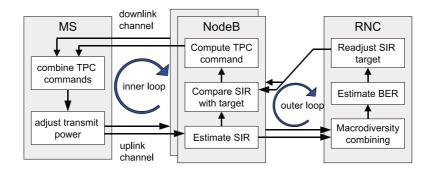
This interference limitation leads to the necessity of reducing the transmission power of each user to a minimum. Mobile stations (MS) will be located at varying distances from the base station (BS, or NodeB). However, if a mobile near the NodeB transmits with a power level that is too high, it causes too much interference for other MS farther away ("near-far" problem). Additionally, due to shadowing and multi-path fading there will be variations in the signal strength received at the NodeB. All of this is overcome by a tight power control performed on the *uplink* (mobile–to–base station path). With this power control algorithm the NodeB tries to perform a balancing of the received *bit-energy-to-noise ratios* ( $E_b/N_0$ ) of all users in the cell. For these reasons, an evaluation of the capacity depends heavily on the performance of the power control loops.

Closed loop power control based on local SIR estimates has been studied by many authors. While some of the studies contain analytical approaches, e.g. [3, 4, 5, 6], many researchers have used simulations to investigate the effects of the system parameters on the efficiency of power control, see [7]. Most of these previously mentioned studies focus on IS-95 systems operating with a single class of users, however, the complexity of the model increases rapidly when considering several different classes of users. This often leads to the case where an analytical performance evaluation reaches its limits and evaluation by simulation is required [8, 9].

In this paper we will describe a simulation model of the uplink closed loop power control in UMTS. This model will permit an evaluation of the influence of the user distribution and the traffic mix on the interference in a cell. The user distribution will be modeled with a homogeneous spatial Poisson process and we will derive empirical distributions of the interference caused by the users in the same cell and users from neighboring cells. Based on the total interference, we will show how the traffic density impacts the cell loading factor.

This paper is organized as follows. In Section 2 we will give a description of the detailed closed loop power control model and describe an approximation model that reduces the computation time. We will show the relationship between the user density, traffic mix, and the interference in Section 3 and demonstrate that the cell loading factor is a useful indicator for cell capacity. Numerical results will also be given that indicate

the applicability of the approximation method which permits an evaluation of complex real world scenarios in Section 4.



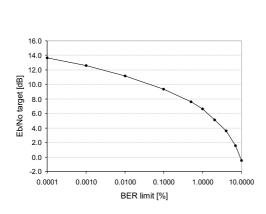
# 2 MODELING OF UMTS POWER CONTROL

Figure 1: WCDMA uplink closed loop power control

The closed loop power control on the uplink consists of two interoperating loops [10, 11], see Fig. 1. Within the *inner loop*, the BTS continuously monitors the quality of the uplink in terms of the received  $E_b/N_0$  level and compares it with the outer loop target. If the received value is too high, the MS causes too much interference to its environment and it is told to decrease its power by a fixed step size. On the other hand, if the received  $E_b/N_0$  is too low, the link quality is not sufficient and a "power-up" command is sent to the mobile. Such an inner loop correction is performed for each power control group, 15 times per frame, resulting in a rate of 1500 updates per second. It should also be noted that an MS in soft handover will be communicating with the *radio network controller* (RNC) via all base stations in its active set. In this case it will also receive several power control commands and will only increase its power if all base stations in the active set will demand an increase in transmit power. In our case, we will consider an inner loop correction step size of 0.5 dB.

After the transmission of a frame is complete, power control enters the *outer loop* [12]. A CRC check is performed for each frame at the RNC and the outer loop target

value is updated based on the current frame error rate (FER). While the actual outer loop uses the received FER for updating its target value, we will consider a simplified case in which we evaluate the bit error rate instead. The reason for this approximation is that the relationship between  $E_b/N_0$  and FER is not straightforward, as it requires an inclusion of the error rates at modulation, interleaving, and channel coding. Therefore, we will map the relationship between  $E_b/N_0$  and BER based on the modulation scheme. Fig. 2 presents the mean  $E_b/N_0$  target results from simulations based on different BER limits.



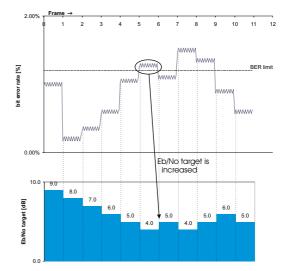


Figure 2: Measurement of  $E_b/N_0$  target in proportion to BER limit

Figure 3: Correlation between the BER and  $E_b/N_0$  target

For a better understanding of the effects of the bit error rate on the resulting  $E_b/N_0$ target an example is given in Fig. 3. The development of both BER and  $E_b/N_0$  target is presented over a period of 11 frames. The bit error rate exhibits a typical course with variations from frame to frame. At the end of each frame the outer loop control adjusts the  $E_b/N_0$  target depending on the current bit error rate. This means that if the BER exceeds the defined BER limit, the  $E_b/N_0$  target has to be increased. Otherwise, this variable will be reduced. So the course of the  $E_b/N_0$  target has to be seen with regard to the bit error rate of each frame.

#### 2.1 Approximation Methods for Uplink Power Control

The major computational part of the simulation algorithm (cf. Fig. 1) consists of the inner loop computation. Furthermore, the purpose of the simulation of power control groups lies in the adjustment of the mobile stations' transmit powers. Since the calculation time required for the simulation of a scenario often turns out as a limiting factor during detailed examinations, an inner loop algorithm is presented in this section that includes both, an approximation of the actual transmit power and a significant simplification of the extensive inner loop mechanism.

The basic idea behind this approximation algorithm is an estimation of the mean transmit power applied by the power control groups of an inner loop cycle. Since the inner loop mechanism adjusts the transmit power so that the current  $E_b/N_0$  adapts to the  $E_b/N_0$  target, a reverse calculation scheme is used as basis of the approximation (cf. Fig. 4).

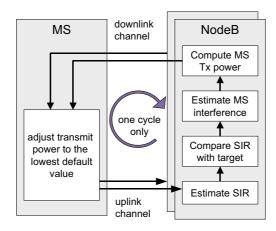


Figure 4: Approximated uplink power control loop

The mobile station's required received power at the NodeB in order to achieve his  $E_b/N_0$  target depends on the total interference affecting the cell. This is the sum of all received powers including the considered mobile station and can be approximated by

	12.2 kbps	144 kbps	384 kbps
$E_b/N_0$ target	6 dB	3 dB	3 dB
mix1	100%	—	—
mix2	—	100%	—
mix3	—	—	100%
mix4	75%	20%	5%

Table 1: Traffic mixes and  $E_b/N_0$  targets

taking the value from the previous iteration step. Following this, the received power can be computed and with consideration of the path loss this leads to the transmit power. Because the complete inner loop cycle containing the simulation of the power control groups is now reduced to a single calculation step the required simulation time is reduced by the factor of 15 (corresponding to the 15 power control groups per frame in UMTS). This opens the way to simulating more detailed and complex scenarios as we will show in Section 4.

# **3 DERIVATION OF CELL LOADING AND NUMERICAL RESULTS**

In this section we will give a description of the underlying simulation scenarios and the user and traffic distributions. For different reference scenarios, we will show the impacts of the user density on the distribution of the interference received at a certain cell site. The interference for different traffic distributions leads to a derivation of the cell loading, which reduces the parameterization of traffic to a single value.

#### 3.1 Scenario Description

Let us consider in the following a regular base station layout as depicted in Figure 5. Cell sizes are illustrated here with their pilot signal ranges, a parameter which we kept constant throughout the simulation. Within the considered area, users were randomly generated following a spatial homogenous Poisson process. In order to consider individual data rates, we selected the traffic mixes for all cells and the desired  $E_b/N_0$  target level from Table 1, cf. [13]. A fixed distance of 2 km between each base station was

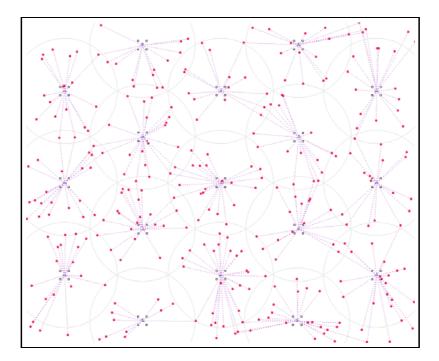


Figure 5: Simulated cell layout

chosen. All simulation runs were performed over a duration of 15000 power control groups corresponding to a simulation time of 10 seconds after a warm-up period of 5 seconds.

# 3.2 Evaluation of Inter-Cell and Intra-Cell Interference

The first value that we will investigate is the cumulated interference, distinguished in interference caused by users in the same cell due to the quasi-orthogonality of the codes (*own-cell interference*) and the influence from users of the surrounding tiers of cells on our observed center cell (*other-cell interference*). Both terms together constitute the term which we will denote as *total interference*  $I_{tot}$ .

In Fig. 6 the cumulative probability distribution function (CDF) of the own-cell interference is depicted for a fixed scenario with a mean traffic density of 4 users per km<sup>2</sup>. Due to the higher spreading gain for 12.2 kbps users in mix1, the least interference is created here. The higher the data rates are, the higher is also the interference.

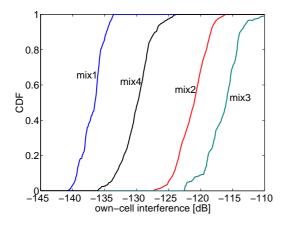


Figure 6: CDF of own-cell interference for traffic mix 1-4

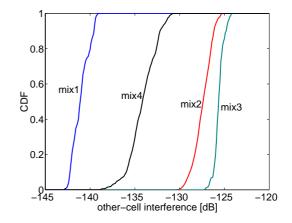


Figure 7: CDF of other-cell interference for traffic mix 1-4

Furthermore, it can be seen that mix4, which is a combination of different data rates has a higher variance indicated by the more gradual steepness of the curve. In Fig. 7 a similar observation can be made for the other-cell interference. However, it can be also seen that the other-cell interference has a smaller mean and variance than the curves in Fig. 6. Again, the mix of several different data rates results in a higher variance.

The influence of the user density on the mean total interference is illustrated in Fig. 8. Depending on the traffic mix, the mean interference increases with the user density, for higher data rates the increase is steeper and saturates when the user density gets higher and the power control loops fail to achieve the target values. The outage rate will then increase as well. The error bars in Fig. 8 give the 95% confidence interval obtained from 30 runs.

Fig. 9 shows the CDF of the total interference power for traffic mix4, where the full inner and outer loop simulation is compared with the inner loop approximation described in Section 2.1. It can be seen that for low interference the approximation matches the full loop quite well, however, for higher values the lack of the fine tuning in the inner loop leads to a slight overestimation of the transmit power, which results in a higher overall interference.

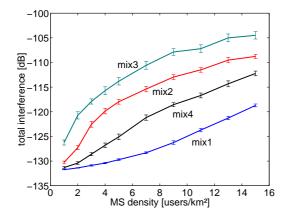


Figure 8: Mean total interference as function of MS density

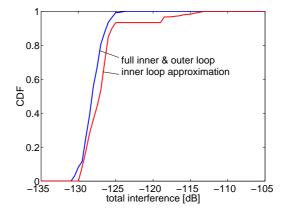


Figure 9: Comparison of inner loop approximation

## 3.3 Definition of Cell Loading

In CDMA systems with only a single class of users, capacity can be easily described by the number of users served in the cell and the theoretical maximum limits can be computed as *pole capacity* [14]. However, third generation networks will provide many different classes of services, each with an individual data rate, activity behavior, and QoS requirement given by the maximum BER that is tolerable for this application. In order to evaluate the capacity of a UMTS cell we will examine the term *cell loading*  $\rho$ which we will define as

$$\rho = \frac{I_{\text{tot}}}{I_{\text{tot}} + N_0}.$$
(1)

Here,  $I_{tot}$  denotes the total interference at the observed cell and  $N_0$  is the thermal noise density. For an empty system without any user, the total interference  $I_{tot}$  will be zero and thus yields a load of zero. For increasing  $I_{tot}$ , the influence of the  $N_0$  term in the denominator will reduce and  $\rho$  approaches one. Note that often a different definition of the loading factor is used, which considers the ratio of own-cell interference to total interference [15]. This leads to the view of the load created in the considered cell, while we are more interested in a normalized interference term. In Fig. 10 the cell loading factor is shown as a function of the MS density. It can be seen that depending on the traffic mix, the value  $\rho$  describes with one value the load caused by a certain traffic situation, regardless what traffic mix or user density exists. Thus, a load of 0.8 could represent a load created when an average user density of 10.5 users/km<sup>2</sup> with 12.2 kbps is considered or a density of approximately 5.5 when 75% operate with 12.2 kbps, 20% with 144 kbps, and 5% with 384 kbps.

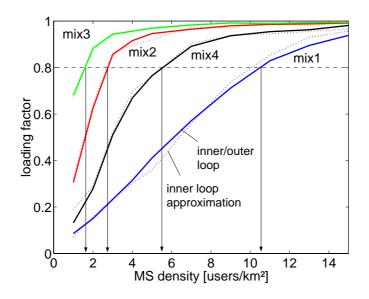
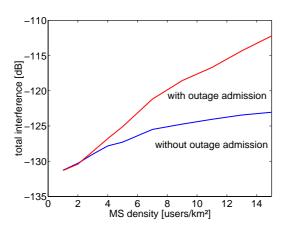


Figure 10: Cell loading as function of MS density

# 3.4 Consideration of Outage Admission

So far, we considered that all MS would perform their power control loops regardless of the actually received  $E_b/N_0$  of their connection. In the following we will consider the case when users are removed from the system, whenever their outage requirements are not fulfilled. We consider all users to be in *outage*, whenever their current  $E_b/N_0$  falls below 0.5 dB under the  $E_b/N_0$  target for a period of 1 second. Since we so far modeled all users to be active from the beginning of the simulation, we chose the following method to remove users in outage. At first all users which exceeded the  $E_b/N_0$  target most were dropped and if after a recalculation of the interference situation other users were still in outage these would be dropped step by step as well.



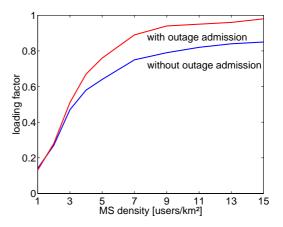


Figure 11: Mean total interference with outage admissions

Figure 12: Influence of outage admissions on cell loading

In Fig. 11 and Fig. 12 the results from Section 3.2 and 3.3 for traffic mix4 are compared to the case when we remove users that are over a period of 1s in outage. It can be seen that as long as we also consider users in outage for our interference calculation that would in reality drop their connection, we overestimate the total interference and cell loading. In fact the removal of outage users reduces the total interference up to 10 dB and the maximum load to about 85% instead of approaching 100%.

# 4 EVALUATION OF COMPLEX SCENARIOS

The inner-loop approximation now enables simulations of larger scenarios with complex characteristics. This section therefore demonstrates the measurement of cell parameters for a more realistic scenario. In view of network planning these results can be used to improve the base station arrangement in order to increase the coverage and enhance the quality of service of each user, see [2].

The scenario in this section uses data from the area around the city of Würzburg. A



Figure 13: Realistic simulation scenario of the city of Würzburg

map of this area with an extension of  $6.7 \times 4.2$  km is illustrated in Fig. 13. We assume that demands in teletraffic correspond to the density distribution of conventional traffic, i.e. more subscribers can be found on the main streets and in the city center than on country roads. Only mobile subscribers are used in this study. The fixed number of subscribers is thereby distributed over the streets at the beginning of each simulation. These test users move during the simulation randomly on the road system – main streets are chosen preferably. Subscribers leaving the simulation area reenter the map again randomly at another incoming street. The number of test users therefore remains constant for the entire simulation time.

Contrary to the simplified studies, communication is now simulated on the basis of calls. This means, each test user begins and ends calls randomly. Therefore, we will use a slightly different method than in Section 3.4 to cope with outages. Since we have a session activity in this case, not all users will begin their connection at the same time instant. The criterion to drop outage users is given whenever the target BER can not be reached within 2 seconds.

The cell of interest is cell 5, where we will focus on the highlighted sector covering most parts of the city center with lots of streets with high density. In order to compare the results to a homogeneous case, we chose also a Poisson process with the same average number of connections in a sector. However, unlike the Poisson case, in our Würzburg scenario there are no stationary users.

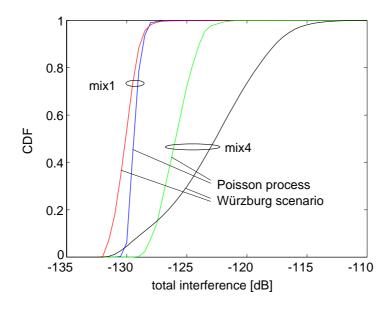


Figure 14: CDF of total interference for mix4

In Fig. 14 the CDF of the total interference is given for traffic mix1 and mix4. It can be recognized that for mix4 there is a difference in mean of about 3 dB which shows that the homogeneous assumption leads to an underestimation of the total interference. For the more homogeneous traffic of mix1 (only voice users with 12.2 kbps) a good match between the Poisson model and the Würzburg model can be seen. Additionally, the inclusion of mobility in the Würzburg scenario causes that the standard deviation of the total interference increases by about 3 dB for mix4 and only 0.5 dB for mix1. Fig. 14 leads to the conclusion that it is possible to capture the effects of the traffic parameters on the interference in a homogeneous scenario, however, in order to perform an accurate prediction of the interference and load a detailed modeling of the location and movement of the users plays an essential role.

# 5 CONCLUSION

In this paper we presented a simulation model for evaluating the performance of a 3G wireless network operating with WCDMA. The capacity of such a system is greatly influenced by the performance of the closed loop power control. We therefore modeled the inner and outer loop uplink power control in great detail and derived empirical distributions of the own-cell and other-cell interference. It could be seen that changing the mix of traffic in the cells has an impact on the mean and variance of the received interference. Furthermore, we examined the loading factor as a descriptor of the traffic load which integrates the user density in terms of mean number of users per unit area size and the mix of data rates into a single parameter value.

We also showed that it is important to include the user reaction to outages since this results in calls to be dropped from not being able to maintain the  $E_b/N_0$  target level and thus reduces the load in the cell. This also indicates the need for an accurate model of the connection admission control mechanism implemented in the system.

The simulation time could be reduced by a factor of the inner loop updates when we only considered the outer loop with an approximated inner loop. It could be shown that the results were accurate enough for evaluation of the cell loading and total interference. With this approximation of the power control loops, it is possible to investigate more complex scenarios with mobility like the real world case which we considered in Section 4.

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