



Joint User Pairing and Resource Allocation for Multiuser SC-FDMA Transmission

Wolfgang H. Gerstacker and Michael A. Ruder

*University of Erlangen-Nürnberg
Chair of Mobile Communications*

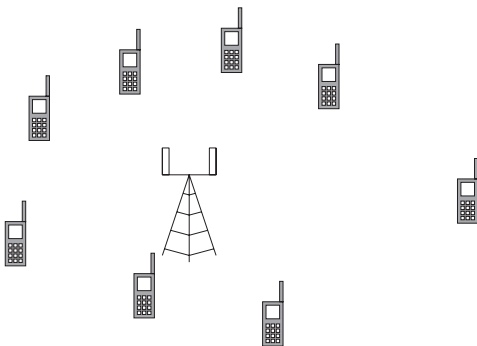
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- 3 Equalizers for SC-FDMA
- 4 Criteria for User Grouping and Resource Allocation
- 5 Algorithms for Time / Frequency Domain Resource Allocation
- 6 Numerical Results
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Basic Idea

System with

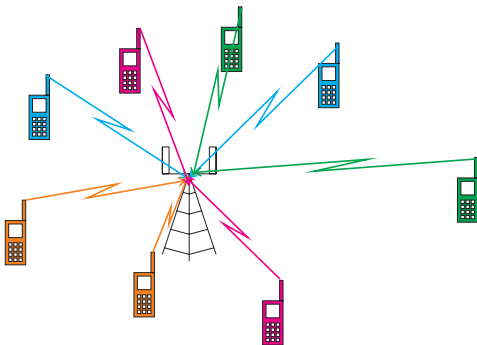
- N_{user} users
- One transmit antenna per user
- One base station with N_{u} antennas



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System with

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Considered Scenarios

- User Pairing/Grouping:
 - N_u users transmit on the same time and frequency resource
 - $N_u = 2 \Rightarrow$ user pairing
 - $N_u > 2 \Rightarrow$ user grouping
- Pairing/Grouping Criteria:
 - Capacity
 - BER
- Resource Allocation:
 - TDMA scheme: transmit in subsequent time slots, using entire bandwidth
 - FDMA scheme: transmit in same time slot, on different frequency chunks

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SC-FDMA Transmitter

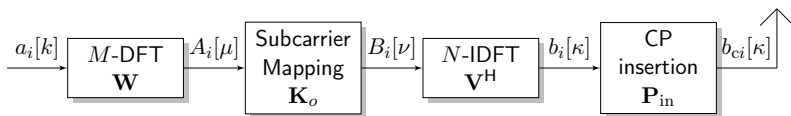


Figure: SC-FDMA transmitter structure for user i in pair o .

\mathbf{W} : M -point DFT matrix

\mathbf{K}_o : subcarrier mapping matrix for group $o \in \{1, \dots, O\}$

\mathbf{V} : N -point DFT matrix

\mathbf{P}_{in} : cyclic prefix insertion matrix

SC-FDMA Receiver

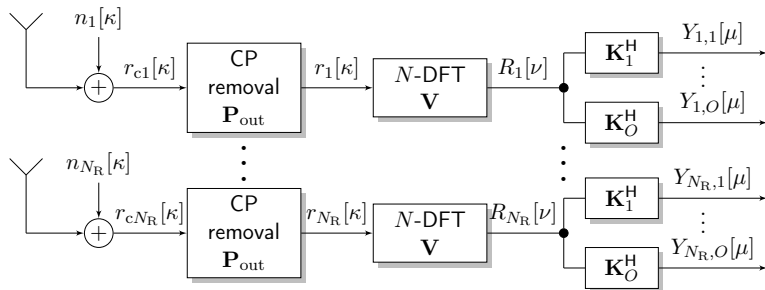


Figure: SC-FDMA BS receiver structure.

$n_l[\kappa]$: spatially and temporally white Gaussian noise at receive antenna l

\mathbf{P}_{out} : cyclic prefix removal matrix

\mathbf{V} : N -point DFT matrix

\mathbf{K}_o : subcarrier mapping matrix for group $o \in \{1, \dots, O\}$

V-MIMO Channel

Signal at the l th receive antenna:

$$r_{cl}[\kappa] = \sum_{i=1}^{N_{\text{tx}}} \sum_{\lambda=0}^{q_h} h_{l,i}[\lambda] b_{ci}[\kappa - \lambda] + n_l[\kappa]$$

$h_{l,i}[\lambda]$: discrete-time subchannel impulse response characterizes the transmission from user i to the l th receive antenna including transmit and receiver input filtering

$\lambda \in \{0, 1, \dots, q_h\}$

q_h : channel order

N_{tx} : number of users being transmitted within one subframe (N_u for TDMA scheme, N_{users} for FDMA scheme)

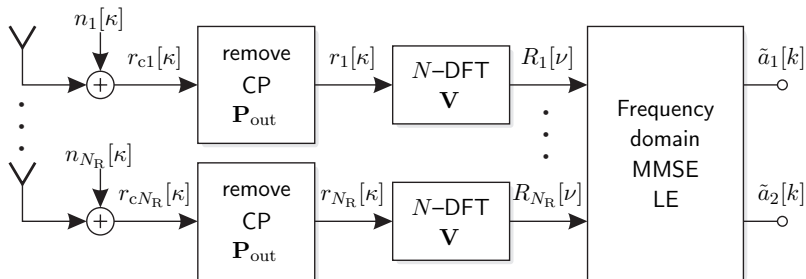
V-MIMO Channel

Signal at the l th receive antenna in matrix-vector notation:

$$\mathbf{r}_{cl} = \sum_{i=1}^{N_{tx}} \mathbf{H}_{l,i} \mathbf{b}_{ci} + \mathbf{n}_l,$$

$\mathbf{H}_{l,i}$: channel convolution matrix for transmission from user i
to receive antenna l

Linear MMSE Equalizer for SC-FDMA



- Filter matrix \mathbf{F}_o for each pair o in frequency domain
- Error variance of the p th user within pair o :

$$\sigma_{e_{u(o,p)}}^2 = \frac{1}{M} \sum_{\mu=0}^{M-1} \Sigma_{pp}^2[\nu_{\text{offset}} + \nu_o + \mu]$$

- $\Sigma_{pp}^2[\nu]$: error variance on ν th subcarrier for user p

SIC Equalizer for SC-FDMA

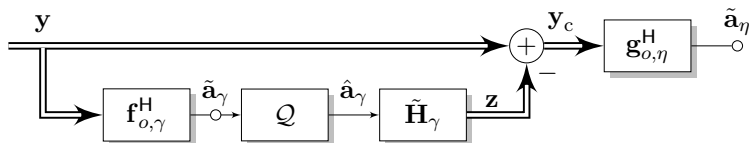


Figure: SC-FDMA SIC equalizer structure for two users.

- Filter vectors $\mathbf{f}_{o,\gamma}$ and $\mathbf{g}_{o,\eta}$ for each user γ and η of pair o
- Equalization order depends on SINR after equalization

$$\text{SINR}_{\text{biased},p}\{o\} = \sigma_{a_{u(o,p)}}^2 / \sigma_{e_{u(o,p)}}^2$$

with

$$\sigma_{e_{u(o,p)}}^2 = \sigma_{a_{u(o,p)}}^2 - \mathbf{f}_{o,p}^H \boldsymbol{\varphi}_{\mathbf{y}_o a_{u(o,p)}}[0]$$

Random Grouping and Resource Allocation

- Random grouping
- Random allocation of users to resources

⇒ Huge potential is wasted

Capacity Grouping

Capacity of group o :

$$C_o = \frac{1}{M} \sum_{\mu=0}^{M-1} \text{ld det} \left(\mathbf{I}_{N_R} + \frac{\text{SNR}}{N_u} \mathbf{H}_o[\nu_{\text{offset}} + \nu_o + \mu] \right. \\ \left. \times \mathbf{H}_o^H[\nu_{\text{offset}} + \nu_o + \mu] \right)$$

Aim: find a mapping function $u(o, p)$ that maximizes the sum of the capacities of all individual groups:

$$C_{\text{sum}} = \sum_{o=1}^O C_o.$$



Bit Error Rate Grouping I

- Approximation for BER of an uncoded transmission:

$$\text{BER} \approx \frac{N_{\min}}{\text{ld}(\mathcal{M})} \text{Q} \left(\sqrt{d_{\min}^2 \frac{E_b}{\mathcal{N}_0}} \right),$$

N_{\min} : average number of nearest neighbors of a signal point of the modulation alphabet

\mathcal{M} : size of the modulation alphabet

$\text{Q}(\cdot)$: complementary Gaussian error integral

d_{\min}^2 : normalized minimum squared Euclidean distance
 $= 3 \text{ld}(\mathcal{M}) / (\mathcal{M} - 1)$ for QAM constellations

Bit Error Rate Grouping I

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- BER for each user p within the group o can be approximated by

$$\text{BER}_{u(o,p)} \approx \frac{N_{\min}}{\text{ld}(\mathcal{M})} \text{Q} \left(\sqrt{\zeta \text{SINR}_{\text{ub},u(o,p)} G_c} \right),$$

$$\zeta = \frac{3}{R(\mathcal{M}-1)}$$

G_c : gain of the channel code

Bit Error Rate Grouping II

- Average BER for group o

$$\text{BER}\{o\} = 1/N_u \sum_{p=1}^{N_u} \text{BER}_{u(o,p)},$$

- Average BER for all N_{user} users is

$$\text{BER}_{\text{total}} = \frac{N_u}{N_{\text{user}}} \sum_{o=1}^O \text{BER}\{o\},$$

Algorithm for Time Domain Resource Allocation

- Two-dimensional optimization problem
- Hungarian Algorithm (HA) finds solution fast
- HA needs polynomial time vs. double factorial for full search
- Symmetric cost matrix with entries for each pair



FD Resource Allocation – Problem Formulation

- Combined optimization of resource allocation and user pairing
⇒ three-dimensional optimization problem
- A three-dimensional assignment problem is in general \mathcal{NP} -hard
- Task: find that mapping function $u(o, p)$, that either maximizes the sum capacity or minimizes the average BER for all users
- Each value of $o \in \{1, 2, \dots, O\}$ stands for the number of the considered frequency chunk



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FD Resource Allocation – Full Search

- Compute sum capacity (C_{sum}) or BER for all possible groups ($\text{BER}_{\text{total}}$) and group/chunk combinations $u(o, p)$
- Choose combination with the maximum sum capacity/minimum average BER
- If the frequency bandwidth is divided into $O = N_{\text{user}}/N_u$ disjoint chunks the total number of combinations is

$$\begin{aligned} \mathcal{O}_{\text{FS}} &= \binom{N_{\text{user}}}{N_u} \cdot \binom{N_{\text{user}} - N_u}{N_u} \cdots \binom{2 \cdot N_u}{N_u} \cdot \binom{N_u}{N_u} \\ &= \frac{N_{\text{user}}!}{(N_u!)^{N_{\text{user}}/N_u}} \end{aligned}$$



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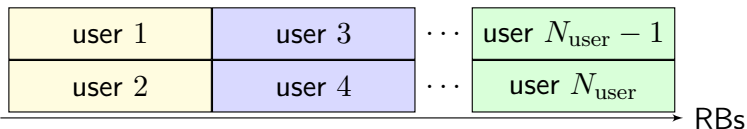
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FD Resource Allocation – Hungarian Algorithm

Before HA



After HA

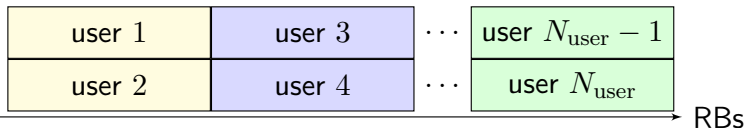


Figure: Example for result of HA for the optimization of the resource allocation of pairs.

HA finds the best assignment of O given user groups to O chunks.

FD Resource Allocation – Hungarian Algorithm

Before HA



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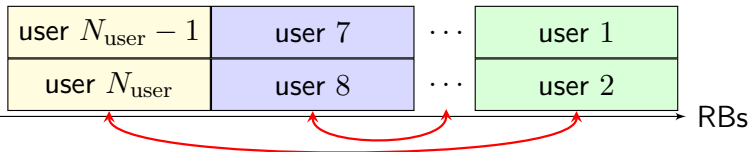


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FD Resource Allocation – Binary Switching Algorithm I

- An exchange of users of different groups can be realized by the binary switching algorithm (BSA)
- Based on
 - Simply try to exchange users between different groups
 - Compute the resulting cost/weight function $\text{BER}_{\text{total}}/C_{\text{sum}}$
 - Perform the exchange which yields the maximum decrease/increase of the cost/weight function of all considered trials
- To limit complexity of the BSA \Rightarrow confine the number of users that can be switched simultaneously to N_{sw}
- Number of total switches for $N_{\text{sw}} = 1$

$$\begin{aligned} O_{\text{BSA}} &= N_{\text{u}} \cdot (N_{\text{user}} - N_{\text{u}}) + N_{\text{u}} \cdot (N_{\text{user}} - 2N_{\text{u}}) + \dots + N_{\text{u}} \cdot N_{\text{u}} \\ &= \frac{1}{2} (N_{\text{user}}^2 - N_{\text{u}} \cdot N_{\text{user}}). \end{aligned}$$

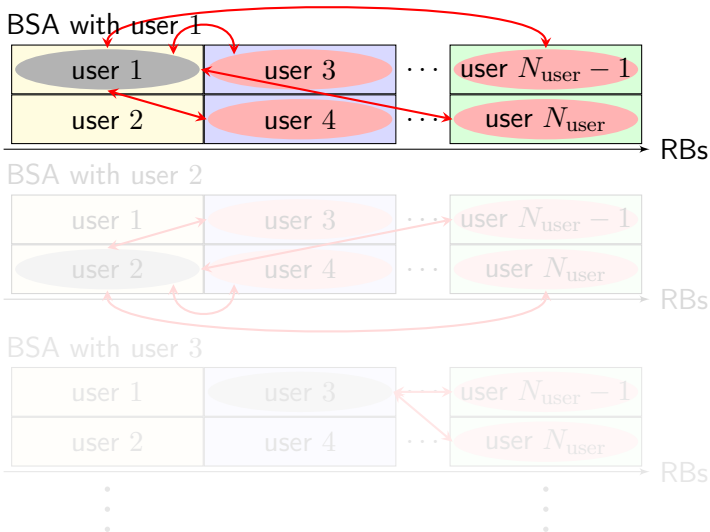


FD Resource Allocation – Binary Switching Algorithm I

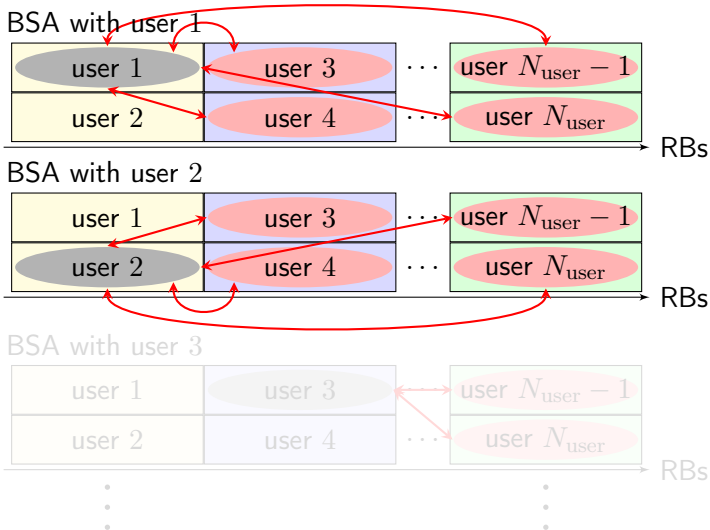
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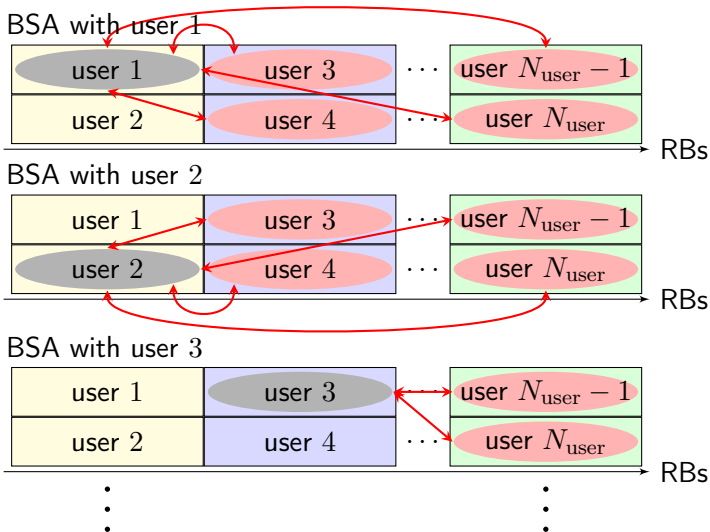
FD Resource Allocation – Binary Switching Algorithm II



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FD Resource Allocation – Binary Switching Algorithm II



FD Resource Allocation – Combined HA+BSA I

- **Initialization:** random groups assigned to random chunks
- **First step:** use HA to find optimum allocation of given groups to chunks
BUT HA does not exchange users between chunks
- **Second step:** BSA exchanges $N_{sw} = 1$ users between chunks
- **Iterations:** repeat first and second step for N_{it} iterations
- **Convergence** of the algorithm is guaranteed in principle
BUT resulting solution might not be the optimum one

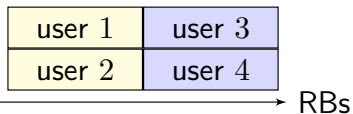


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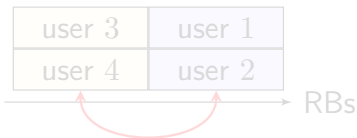
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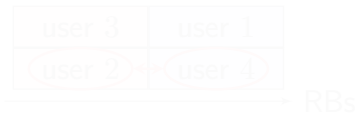
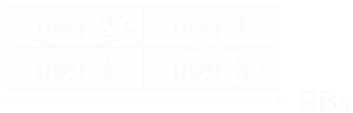
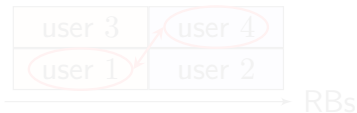
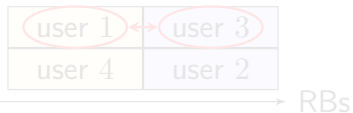
Starting point



HA test



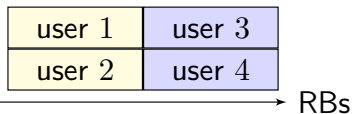
BSA tests



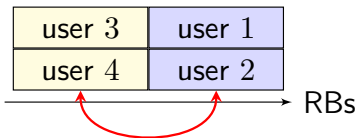
Example for HA+BSA for the pairing and resource allocation of $N_{\text{user}} = 4$ users and $N_u = 2$ users per group.

FD Resource Allocation – Combined HA+BSA II

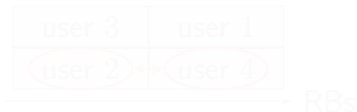
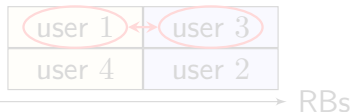
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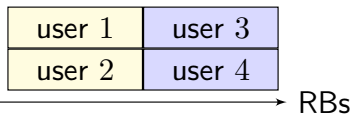
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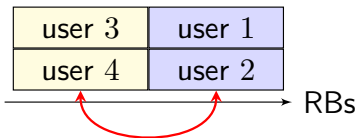
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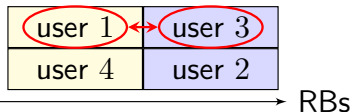
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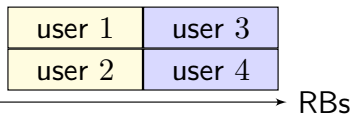
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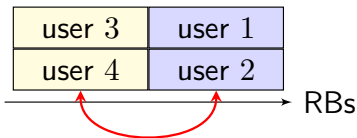
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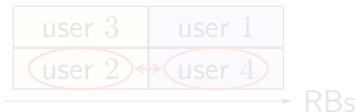
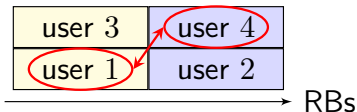
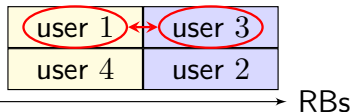
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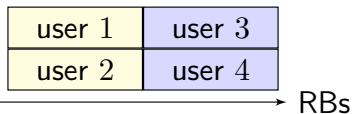
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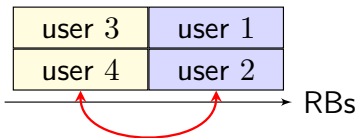
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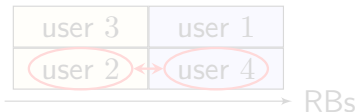
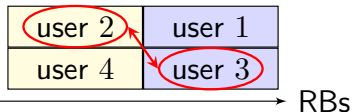
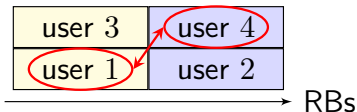
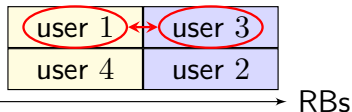
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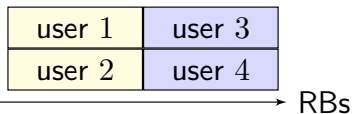
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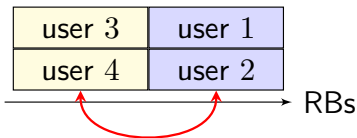
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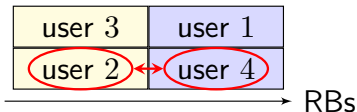
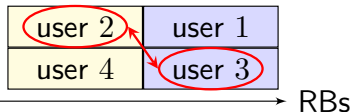
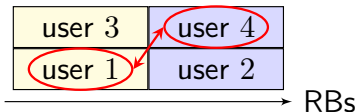
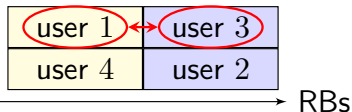
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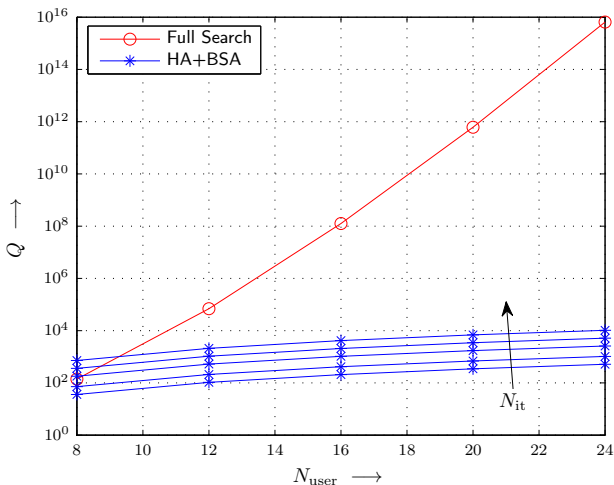
FD Resource Allocation – Complexity Analysis

Number of computations of the capacity/BER for the HA+BSA with N_{it} iterations

$$\begin{aligned} Q_{\text{HA+BSA}} &= N_{it} \cdot \left(\frac{1}{N_u^2} N_{\text{user}}^2 + (N_{\text{user}}^2 - N_u \cdot N_{\text{user}}) \right) \\ &= N_{it} \cdot \left(\left(\frac{1}{N_u^2} + 1 \right) N_{\text{user}}^2 - N_u \cdot N_{\text{user}} \right). \end{aligned}$$

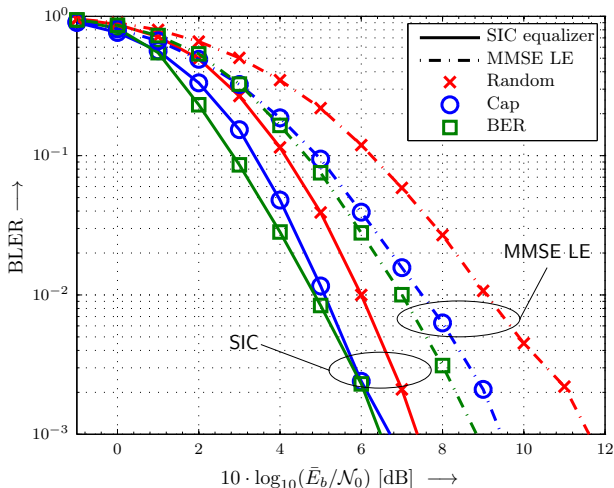


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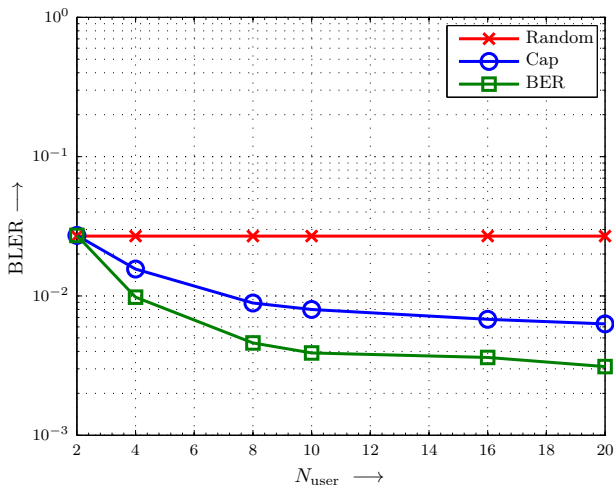
Complexity Q of the full search and of the HA+BSA for $N_u = 4$ and $N_{\text{it}} \in \{1, 2, 5, 10, 20\}$, respectively.

Numerical Results I



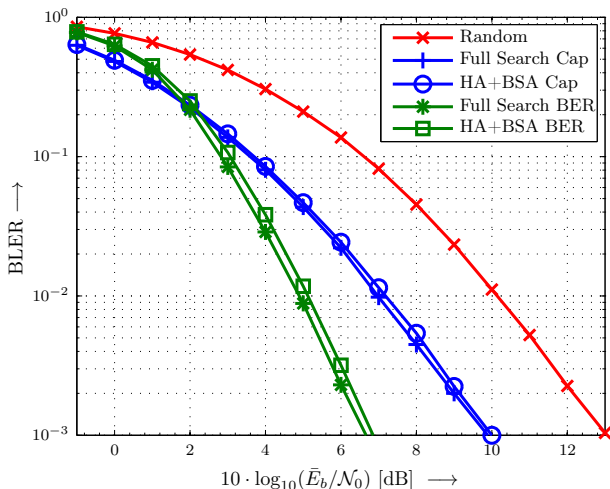
BLER versus \bar{E}_b/N_0 for TD resource allocation, $N_{\text{user}} = 20$, $N_u = 2$, different pairing criteria, 10 MHz bandwidth.

Numerical Results II



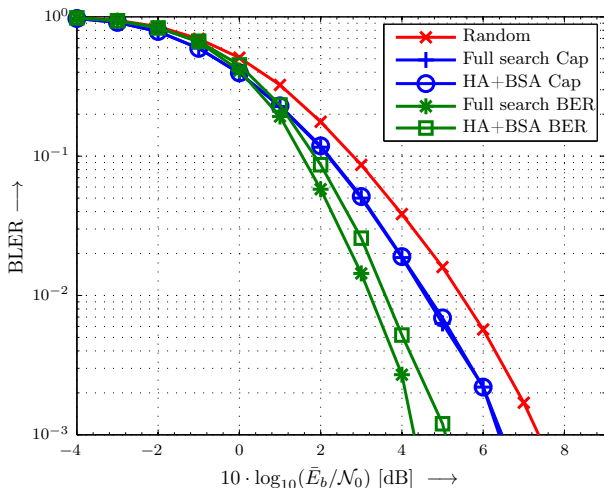
BLER versus N_{user} for TD resource allocation, $\bar{E}_b/\mathcal{N}_0 = 8$ dB, $N_u = 2$, different pairing criteria, MMSE LE, 10 MHz bandwidth.

Numerical Results III



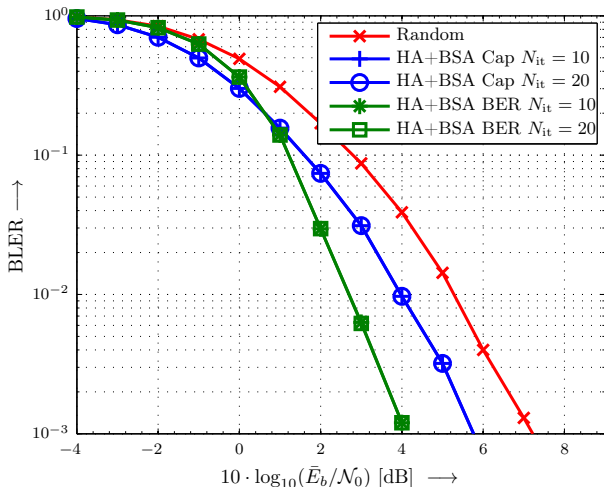
BLER versus \bar{E}_b/N_0 for FD res. alloc., $N_{\text{user}} = 10$, $N_u = 2$, $N_{\text{it}} = 20$, different pairing criteria, MMSE LE, 10 MHz bandwidth.

Numerical Results IV



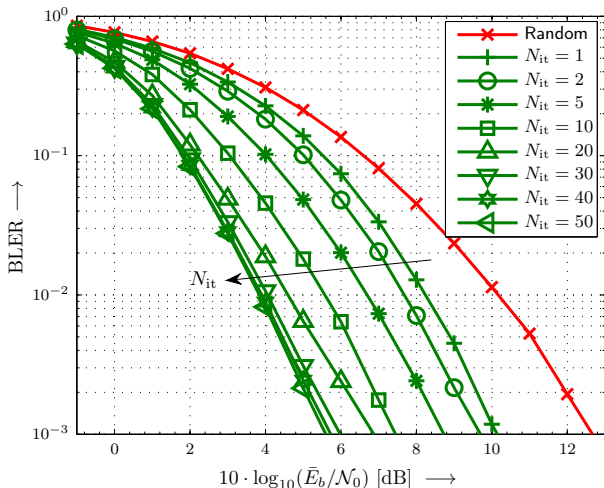
BLER versus \bar{E}_b/N_0 for FD resource allocation, $N_{\text{user}} = 8$, $N_u = 4$, $N_{\text{it}} = 10$, MMSE LE, 20 MHz bandwidth.

Numerical Results V



BLER versus \bar{E}_b/N_0 for FD resource allocation, $N_{\text{user}} = 16$, $N_u = 4$, MMSE LE, 20 MHz bandwidth.

Numerical Results VI



BLER versus \bar{E}_b/\mathcal{N}_0 for FD resource allocation, $N_{\text{user}} = 20$, $N_u = 2$, BER grouping, MMSE LE, 20 MHz bandwidth.

Summary

- Random user grouping and resource allocations wastes huge potential
- Hungarian Algorithm finds optimum solution for two dimensional time domain resource allocation and user pairing problem
- Full search finds optimum solution for three dimensional frequency domain resource allocation and user grouping problem with extremely high complexity
- Proposed iterative HA+BSA finds close to optimum solution for frequency domain resource allocation and user grouping problem at much lower complexity



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Future Work

- Lower bound for capacity pairing (orthogonality in random Gaussian channels, famous balls-and-bins problem)
- Power allocation not yet implemented
- Rate adaptation not considered
- 4×4 V-MIMO with two transmit antennas per mobile
- \Rightarrow Use of network simulator necessary



Joint User Pairing and Resource Allocation for Multiuser SC-FDMA Transmission

Wolfgang H. Gerstacker and Michael A. Ruder

*University of Erlangen-Nürnberg
Chair of Mobile Communications*

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