

Optimisation Methods for Dynamic Frequency Planning in OFDMA Networks

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Outline

- Introduction & Motivation
- OFDMA
 - OFDMA/TDD Frame
 - Frequency Reuse Schemes
- Dynamic Frequency Planning
 - Model
 - Inputs: Demand Vector & Restriction Matrix
 - Optimization Problem: Mixed Integer Program
 - Solving DFP: Simulated Annealing, Tabu Search & Greedy Algorithms
- Results
- Conclusions

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Introduction

- **OFDMA** networks such as WiMAX and LTE are pointed by the industry as the most suitable technologies for the future deployment of cellular networks.
- In order to avoid **inter-cell interference** OFDMA networks are flexible in terms of radio resource management and *FRSs* (*Frequency Reuse Scheme*).
- However, most of these techniques are not suitable for **dynamic scenarios** where the traffic demands and channel conditions rapidly change.

Introduction

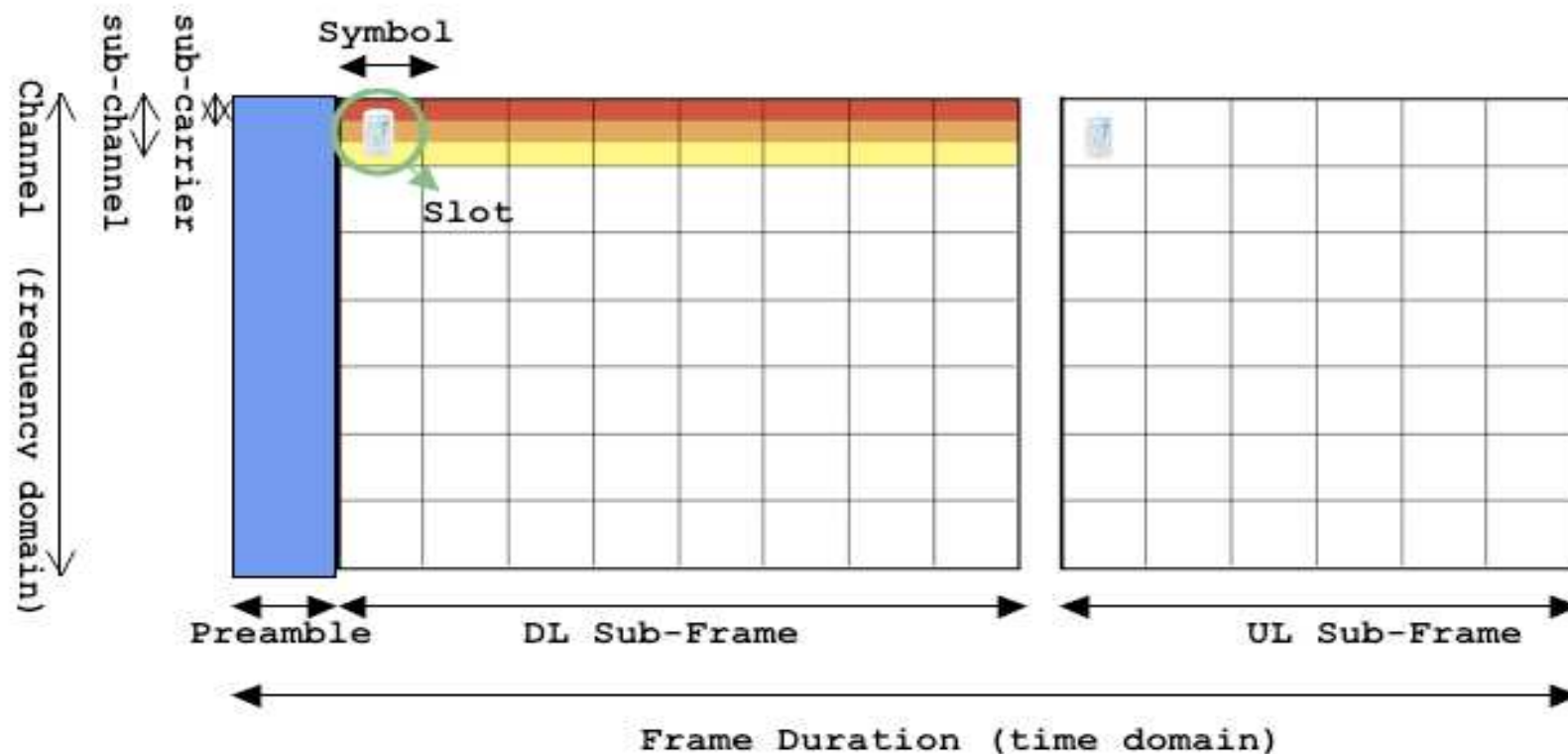
- In this work, *DFP* (*Dynamic Frequency Planning*) is presented as a new approach to the frequency assignment problem tailored to OFDMA networks.
- DFP may **decrease the inter-cell interference** and increase the capacity of the network by dynamically adapting the radio frequency parameters to the specific scenario.
- DFP uses novel algorithms to **sense the radio channel**, and adapted versions of *Simulated Annealing*, *Tabu Search*, or *Greedy Algorithms* to **solve the optimization problem**.

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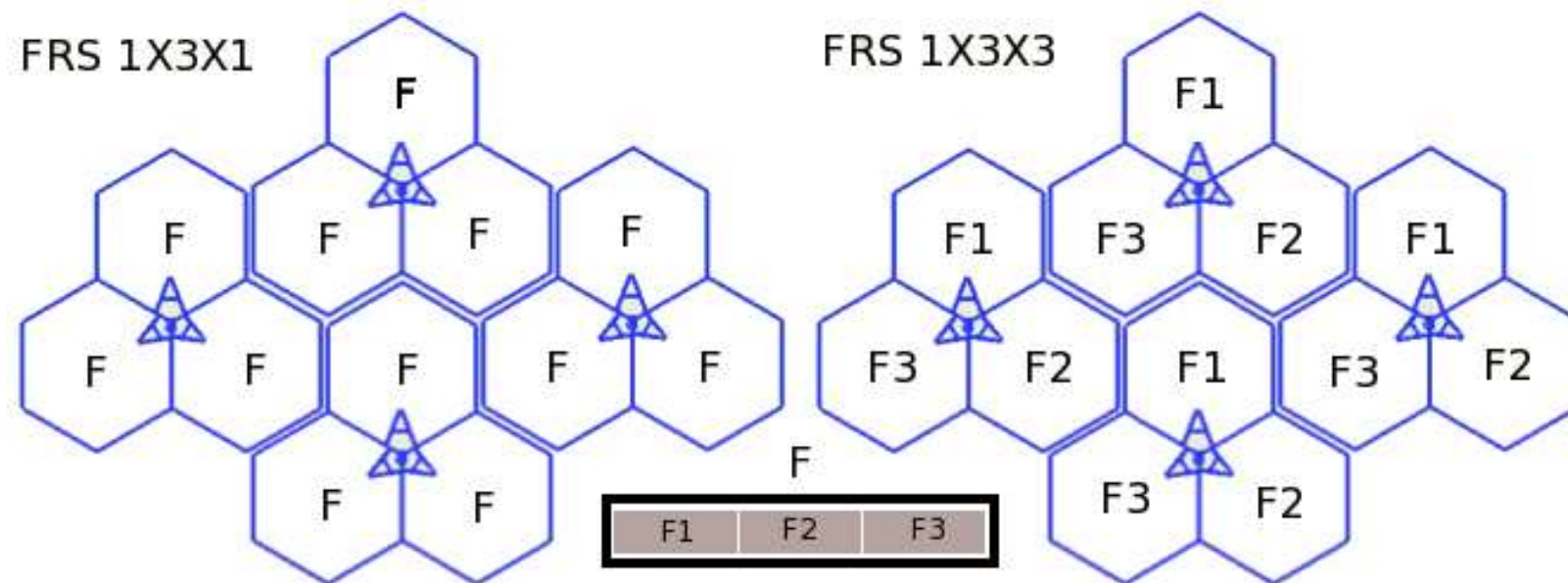
OFDMA/TDD Frame

- When using **OFDMA**, different users are allocated to different subsets of *sub-carriers* called *sub-channels*, allowing the use of *multi-user* or *frequency* diversity.



OFDMA Frequency Reuse Schemes

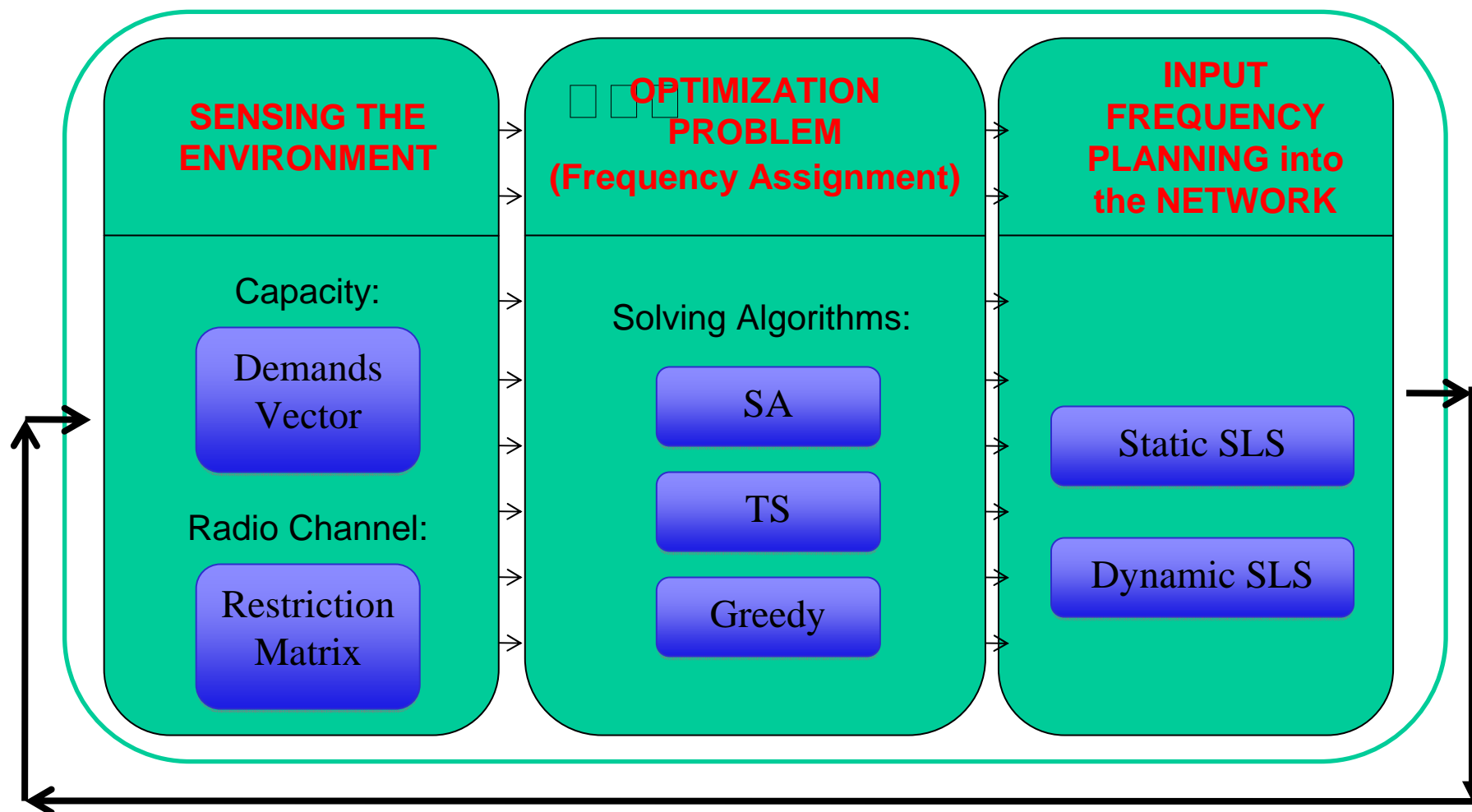
- Many **FRSs** have been proposed to reduce inter-cell interference. However, they are based on **static forecasts**. Therefore, they are not able to cope with the changing behavior of the users and the environment.



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DFP - Model



DFP – Demand Vector

- The demand vector indicates the number of sub-channels needed per sector to satisfy the users demands.
- It is calculated as follows:
 - Number of users able to decode the preamble in each sector
 - Requested slots for the i -th user (1)(2)
 - Requested sub-channel for the S_i -th sector (3)

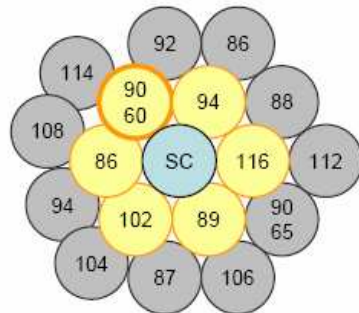
$$(1) \text{ Slot throughput (Kbps)} = \frac{\text{Subcarriers}}{\text{slot}} \cdot \frac{\text{bits}}{\text{subcarrier}} \cdot \frac{1}{5 \text{ ms}}$$

$$(2) \text{ Slot }_i = \frac{\text{Requested Throuhgput }_i \text{ (Kbps)}}{\text{Slot throughput }_i \text{ (Kbps)}}$$

$$(3) C_{S_j} = \sum_{i \in S_j} \text{Slots }_i \cdot \frac{\text{slots}}{\text{sub - channels}}$$

DFP – Restriction Matrix (I)

- **Measurement Report** contains information about Serving Cell and the strongest Neighbouring Cell



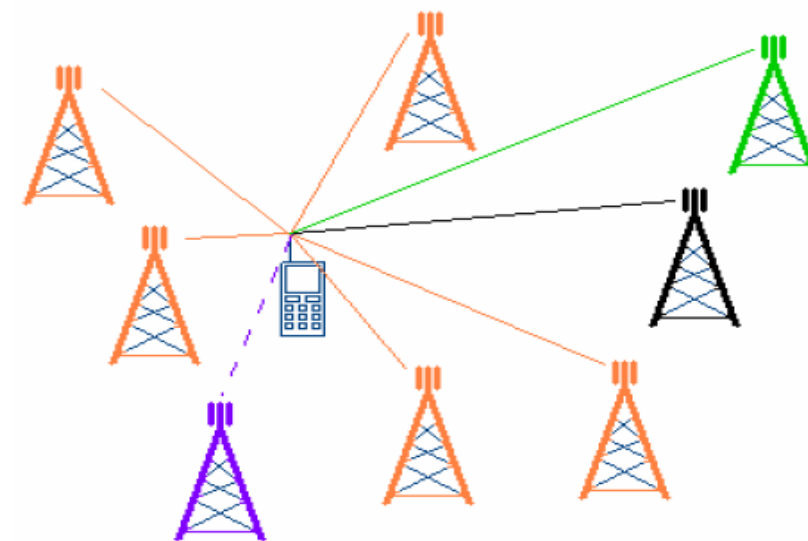
BA List (only def. HO NB):
{94, 116, 89, 102, 86, 90}

Serving Cell Measurements

	RXLEV	RXQUAL
FULL	94	6
SUB	82	1

Neighbouring Cell Measurements

NB	RXLEV	BCCH	BSIC
1	38	94	47
2	42	116	34
3	31	89	62
4	12	102	07
5	10	86	11
6	18	90	60

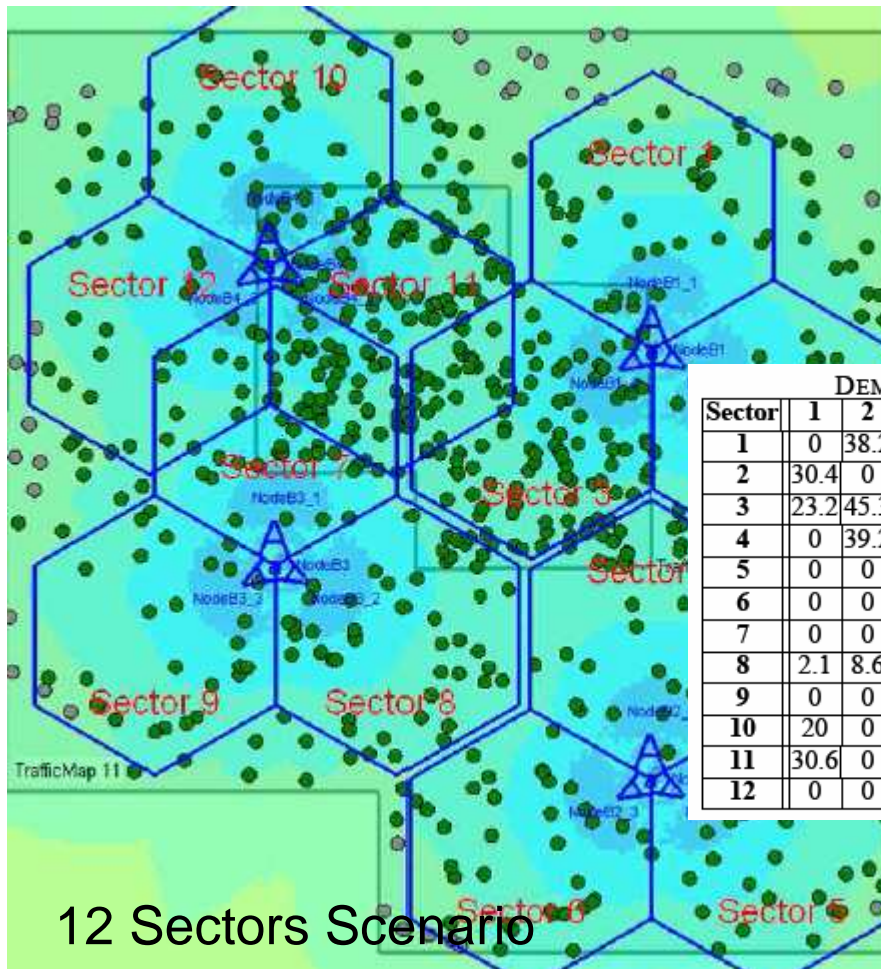


— Serving Cell
— Neighbour Cell
— Other Cells

DFP – Restriction Matrix (II)

- When a MR is received by the BS, the received signal power of the Serving Cell and the Neighbour Cell are compared.
- An **Interference Event** happens if the signal from the Serving Cell is stronger / weaker with respect to Neighbour Cell by specified threshold value defined by the operator.
- **Dividing # I** (Interference Event) by **# MR** (Measurement Report) for a given threshold (e.g.12dB) indicates the amount of overlapping from the Serving Cell and the Neighbour Cell in terms of % of time.
- Finally, the **Restriction Matrix** can be build. The size of the restriction matrix is [Sector x Sector].

DFP – Restriction Matrix (III)



Example of:
Capacity Estimation
Restriction Matrix

DEMAND VECTOR AND RESTRICTION MATRIX

Sector	1	2	3	4	5	6	7	8	9	10	11	12	D_i
1	0	38.2	41.1	0	0	0	2.9	0	0	11.7	26.4	0	4
2	30.4	0	47.8	34.7	4.3	0	0	0	0	0	0	0	3
3	23.2	45.3	0	24.4	0	3.4	31.3	37.2	0	4.6	45.3	0	8
4	0	39.2	46.4	0	28.5	25	0	32.1	0	0	0	0	4
5	0	0	0	52.9	0	29.4	0	0	0	0	0	0	4
6	0	0	0	39.2	42.8	0	0	21.4	10.7	0	0	0	4
7	0	0	23.4	4.2	0	0	0	44.6	23.40	0	76.5	65.9	6
8	2.1	8.6	36.9	56.5	0	43.4	39.13	0	34.7	0	23.9	2.1	6
9	0	0	0	0	0	0	39.2	42.8	0	0	3.57	3.14	4
10	20	0	12.5	0	0	0	17.1	0	0	0	45	22.5	5
11	30.6	0	45.5	0	0	0	50.49	19.8	0	37.6	0	29.7	8
12	0	0	0	0	0	0	65.2	0	4.3	39.1	30.4	0	3

DFP – Frequency Assignment Problem

The problem is to :

$$\min \sum_{i \in N} \sum_{j \in N} \sum_{k \in NF} \frac{W_{ij}}{D_i \cdot D_j} y_{i,j,k}$$

subject to :

$$\sum_{k=0}^{NF} x_{i,k} = D_i \quad \forall i,k$$

$$y_{i,j,k} \geq x_{i,k} + x_{j,k} - 1 \quad \text{and} \quad y_{i,j,k} \geq 0 \quad \forall i,j,k$$

$$x_{i,k} \in \{0,1\} \quad \forall i,k$$

N number of sectors, $\{1, \dots, n, \dots, N\}$

NF number of subchannels, $\{1, \dots, f_k, \dots, NF\}$

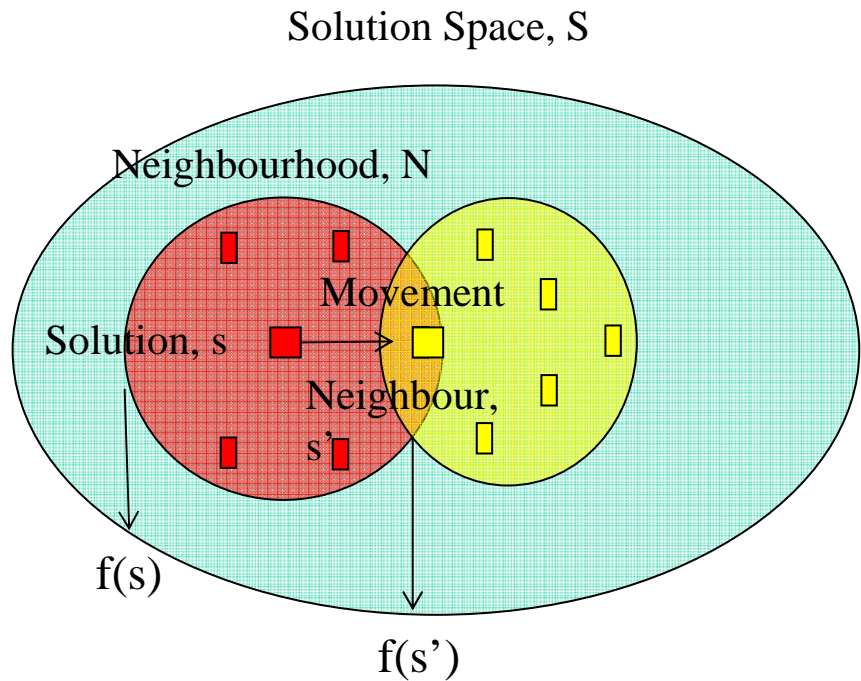
D_i requested subchannels by the i - sector

W_{ij} interference restriction between the i - sector and the j - sector

$y_{i,j,k} \begin{cases} 1, & \text{the } i\text{-th and } j\text{-th sector are using the } k\text{-frequency} \\ 0, & \text{otherwise} \end{cases}$

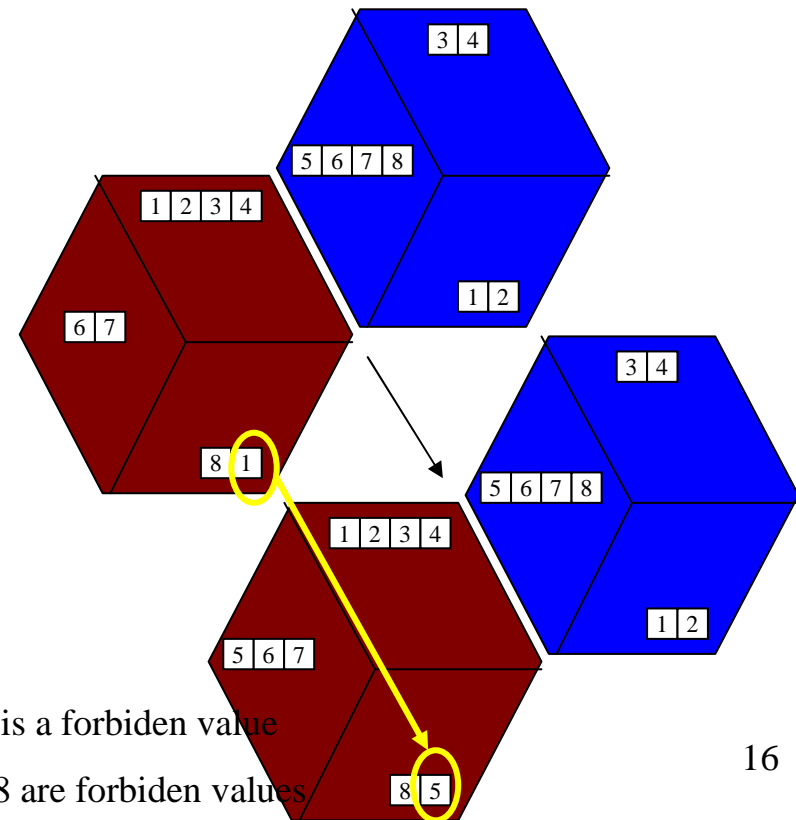
DFP – Solving Algorithms

- Meta-heuristics



best neighbour = $\min f(s')$

NEIGHBOUR DEFINITION



DFP – Solving Algorithms (I)

- Simulated Annealing

```
s := s0; e := f(s)           // Random Solution  
sb := s; eb := e           // Best Solution  
T := high value           // Temperature  
k := 0                     // Iterations  
while T > Tmin and e > emax // While T > Tthreshold or not good enough:  
    sn := neighbour(s)      // Random neighbour  
    en := f(sn)             // Compute its energy  
    if en < eb then         // Is this a new best?  
        sb := sn; eb := en   // Yes, save it.  
    elseif P(e, en, temp) > rnd() // Should we move to it?  
        s := sn; e := en     // Yes, change state.  
    k := k + 1              // Increase iterations  
    T *= annealing factor    // Update temperature  
return sb                  // Return the best solution found.
```

DFP – Solving Algorithms (II)

■ Tabu Search

```
s := s0; e := f(s)           // Random Solution
sb := s; eb := e           // Best Solution
T := 0 ; k := 0           // Tabu List & Iterations
while k < kmax and e > emax // While time remains and not good enough:
    k := k + 1             // Increase iterations
    for (neighbourhood(s))
        sn := neighbour(s) ; en := f(sn) // Neighbour & Compute its energy
        [sbn, ebn] := select_best_neighbourh
        if ebn < eb then    // Is this a new best?
            sb := sn; eb := en // Yes, save it.
        elseif ebn < e then // Is the best neighbour better than the
            current?
            s := sbn; e := ebn // Yes, we move to it
        else                // Otherwise
            s :=random(s); e := f(s) // No, we select a random neighbour
    Update tabu list       // Update tabu list
return sb                 // Return the best solution found
```

DFP – Solving Algorithms (III)

■ Greedy Algorithms

Operation = Insertion: We start with empty assignments. At each iteration, we first compute the frequency which increases the cost function the least for all sectors.

Then:

- Random Insertion (RI): Select a random sector
- Minimum Insertion (MinI): Select the sector increasing the cost function the least.
- Maximum Insertion (MaxI): Select the sector increasing the cost function the most.

DFP – Solving Algorithms (IV)

■ Reverse Greedy Algorithms

Operation = Remove: We start with full assignments. At each iteration, we first compute the frequency which decreases the cost function the most for all sectors.

Then:

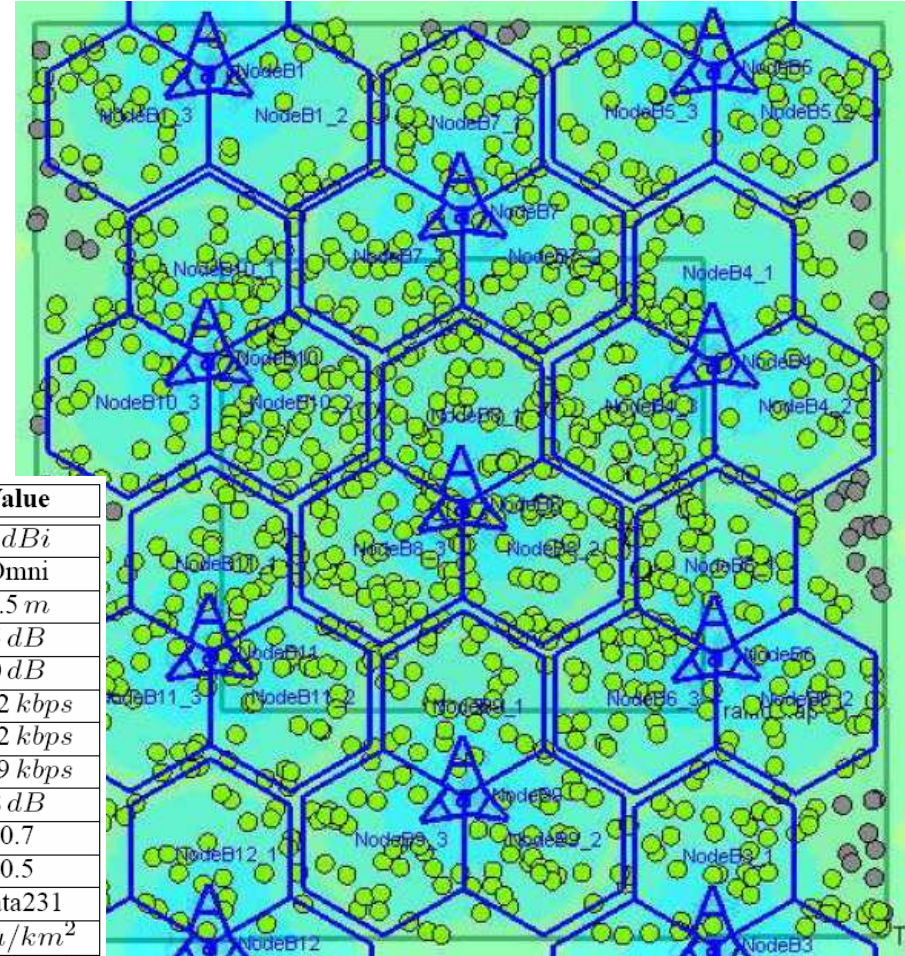
- Random Insertion (RR): Select a random sector
- Minimum Insertion (MaxR): Select the sector decreasing the cost function the most.
- Maximum Insertion (MinR): Select the sector decreasing the cost function the least.

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Results - Scenario

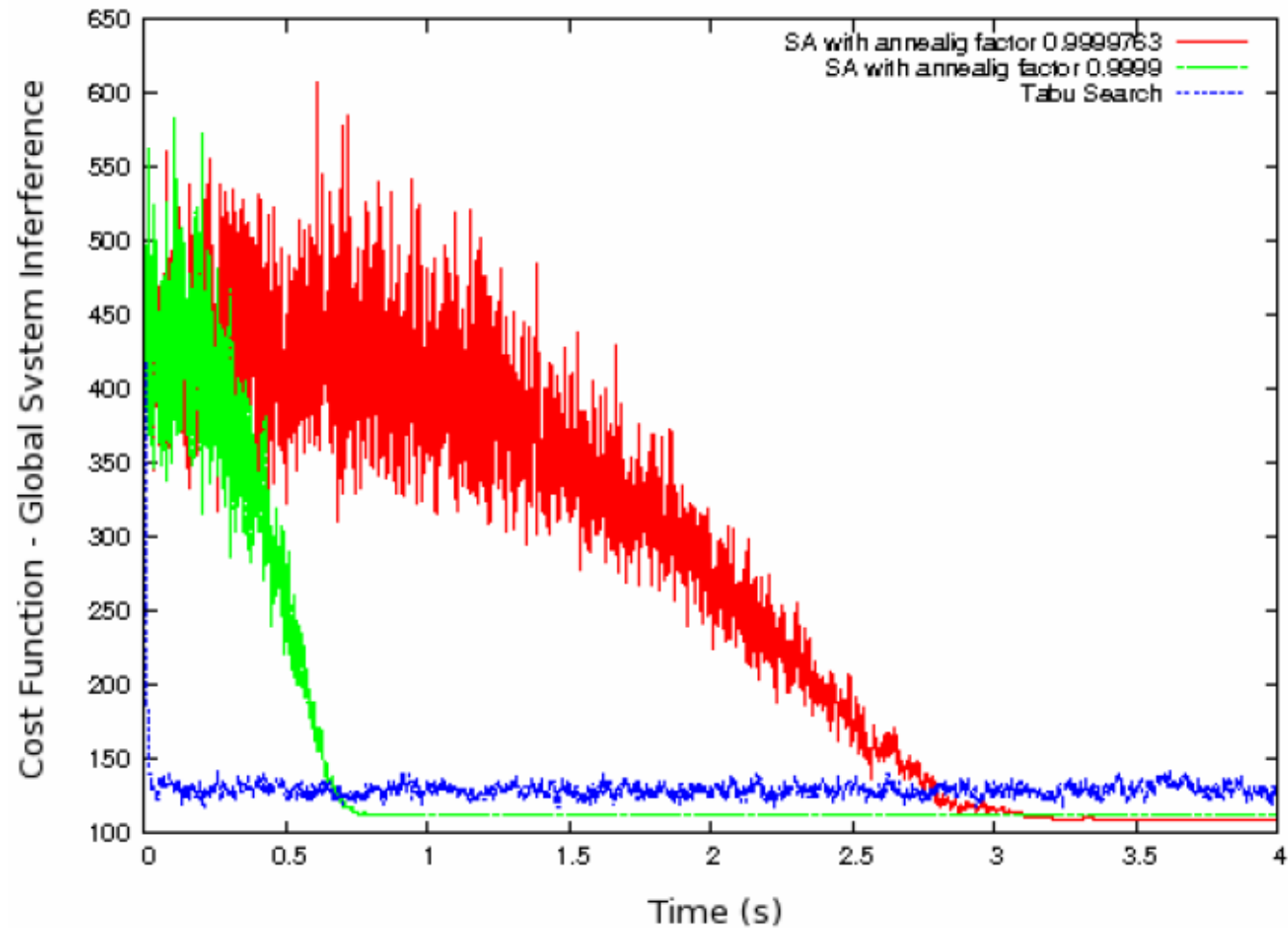
Overlapping Sectors
Heavy Interference
(DL case)



Parameter	Value	Parameter	Value
Sites	11	CPE Ant. Gain	0 dBi
Sector/site	3	CPE Ant. Pattern	Omni
Carrier Frequency	3.5 GHz	CPE Ant. Height	1.5 m
Channel Bandwidth	10 MHz	CPE Noise Figure	5 dB
DL:UL Ratio	1:1	CPE Cable Loss	0 dB
Permutation Scheme	AMC	Min Service TP	12.2 kbps
Frame Duration	5 ms	Max Service TP	42.2 kbps
Sub-channels	16	Average Symbol Eff.	19.9 kbps
DL symbols	19	σ (Shadow Fading)	8 dB
BS TX Power	43 dBm	Intra BS correlation	0.7
BS Ant. Gain	18 dBi	Inter BS correlation	0.5
BS Ant. Beam Width	120	Path Loss Model	Hata231
BS Ant. Height	30 m	Traffic Map Density	40 u/km ²
BS Ant. Tilt	3	Hot spot Density	80 u/km ²
BS Noise Figure	4 dB	Admission	BP ¹
BS Cable Loss	3 dB	Resource Allocation	CA ²
CPE Tx Power	23 dBm	Snapshots	20

¹Best Path loss, ²Contiguous Allocation

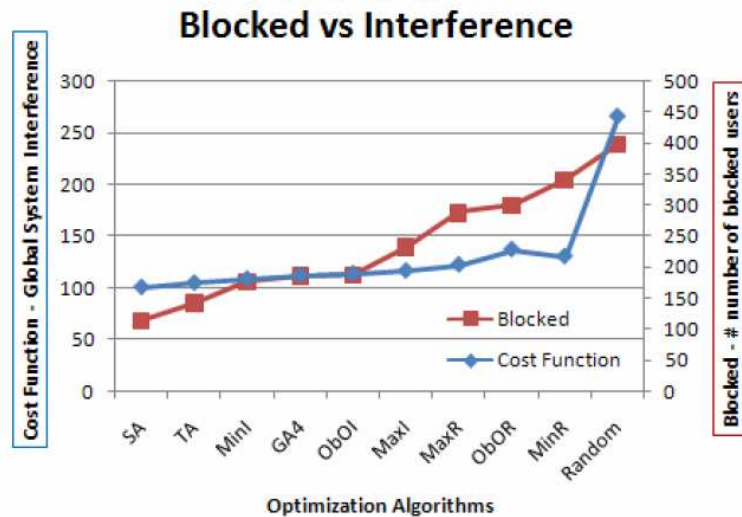
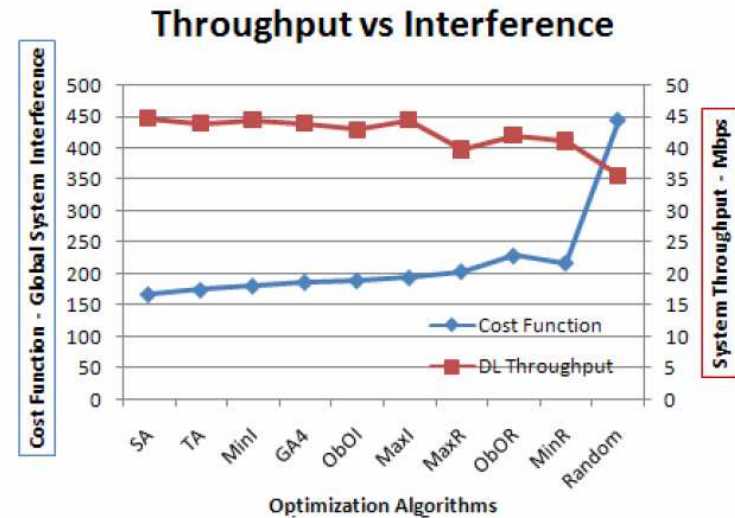
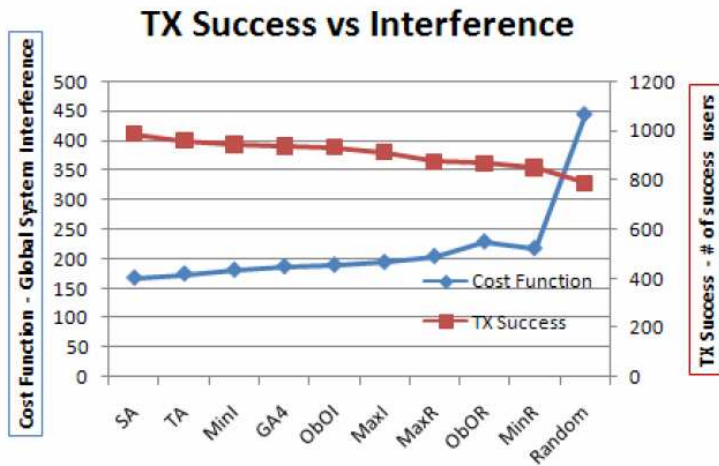
Results – SA Vs TS



Results – Heuristics Vs Greedy Algorithm

Scenario	1 st			2 nd			3 rd		
	users/km ²	20/40	30/60	40/80	20/40	30/60	40/80	20/40	30/60
random	397.6	374.3	353.6	1086.2	941.6	991.2	435.1	435.1	444.0
SA	56.2	180.7	222.3	151.7	292.2	500.4	1.0	52.5	108.1
TS	55.2	181.8	224.3	182.4	317.3	516.8	1.0	53.3	115.7
SI	79.5	197.7	246.2	283.4	331.9	547.3	40.9	111.5	137.5
RI	86.1	192.1	236.9	274.3	377.8	569.7	33.2	88.4	134.1
MinI	98.8	188.4	230.2	222.9	392.9	587.9	27.6	84.5	137.7
MaxI	60.8	199.1	249.1	231.7	343.5	579.9	25.8	79.2	151.2
RR	116.5	203.7	243.7	346.9	439.2	599.8	71.0	97.9	174.4
MaxR	80.0	215.9	230.1	337.1	365.1	570.4	121.2	274.6	224.4
MinR	113.4	224.4	247.3	2928.4	395.8	571.4	4192.0	105.4	127.6
Best MH	55.2	180.7	222.3	151.7	292.2	500.4	1.0	52.5	108.1
Improv(%)	86.1	51.7	93.7	86.0	69.0	49.5	99.8	87.9	75.7
Best GA	60.8	188.4	230.1	222.9	331.9	547.3	25.8	79.2	127.6
Improv(%)	84.7	49.7	34.9	79.5	64.8	44.8	94.1	81.8	71.3
GAvsBest(%)	-1.4	-2.0	-2.2	-6.6	-4.2	-4.7	-5.7	-6.3	-4.4

Results – Simulation



of **success users** and **throughput** increases when interference decreases.

of **blocked users** on the other hand decreases

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Conclusions

- This work has introduced a **novel approach** to the frequency assignment problem tailored to OFDMA networks.
- DFP proposed a **new model** for capacity and inter-cell interference estimation in OFDMA networks.
- **Adapted SA, TS and greedy algorithms** have been used to solve the optimization problem.
- Greedy algorithm have been demonstrated very useful since they **run very fast** and they **produce slightly worse** solutions than meta-heuristics (by around 6%).
- DFP significantly **reduces inter-cell interference** (by around 70%) and **increases system capacity** (by around 20%)

Thank You

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