

# **University of Sarajevo Faculty of Electrical Engineering Department of Computing and Informatics**



Advanced Databases

and Data Mining

**MODELING OF COMPUTER AIDED SIMULATOR IN CONTROL OF NSCLC TREATMENT BASED ON EGFR GENE MUTATIONS' ARTIFICIAL NEURAL NETWORK CLASSIFIER** AND MICROARRAY EXPRESSION ANALYSIS

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EGFR EXONS

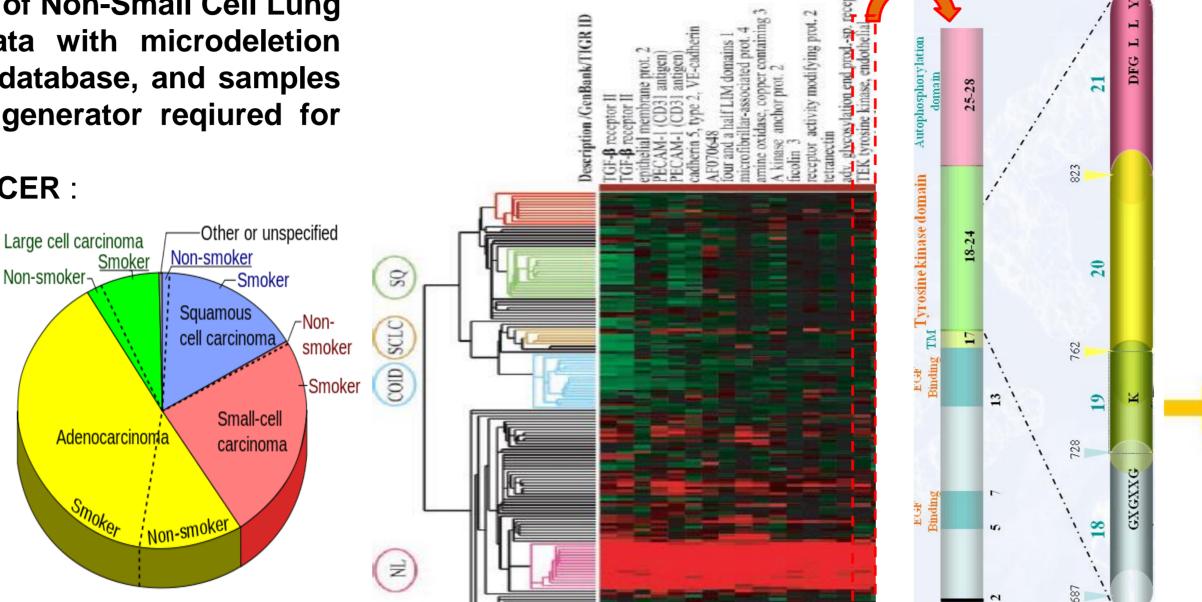
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# LUNG CANCER

This research has as its goal to improve diagnosis of Non-Small Cell Lung Cancer (NSCLC) based on sample patients' data with microdeletion mutations extracted from online EGFR mutation database, and samples with microdeletion mutations generated in own generator required for simulation.

## THE MAIN TYPES OF LUNG CANCER :

1. Non-small-cell lung carcinoma (NSCLC) 1.1 Adenocarcinomas are often found in an outer



# **GENE MICROARRAY**

# EGFR GENE / NCBI / NG\_007726

Different combinations of mutations (micro-deletions) exist within the EGFR kinase, and the most frequently observed mutations are on the exon 18, 9, 20 and 21.

## EXON 18: 159890-160012

MUTATIONS: 96-97 : Patient Class: 17

ACTGAATTCAAAAAGATCAAAGTGCTGGG--CCGGTGCGTTCGGCACGGTGTATAAG

## EXON 19: 160691-160789

MUTATIONS: 50-64. Patient class 1

GGACTCTGGATCCCAGAAGGTGAGAAAGTTAAAATTCCCGTCGCTATCA CAACATCTCCGAAAGCCAACAAGGAAATCCTCGAT

### MUTATIONS: 51-65, Patient Class 2

- area of the lung.
- 1.2 Squamous cell carcinomas are usually found in the center of the lung next to an air tube (bronchus).
- 1.3 Large cell carcinomas can occur in any part of the lung. They tend to grow and spread faster than the other two types.
- Small-cell lung carcinoma (SCLC), 2.1 Small cell carcinoma (oat cell cancer). 2.2 Combined small cell carcinoma.

GGACTCTGGATCCCAGAAGGTGAGAAAGTTAAAATTCCCGTCGCTATCAAG-ACATCTCCGAAAGCCAACAAGGAAATCCTCGAT

### EXON 20: 167262-167447

### MUTATIONS:25-26 : Patient Class: 15

GAAGCCTACGTGATGGCCAGCGT—ACAACCCCCACGTGTGCCGCCTGCTGGGCATCTGCCTCA CCTCCACCGTGCAGCTCATCACGCAGC TCATGCCCTTCGGCTGCCTCCTGGACTATGTCCG GGAACACAAAGACAATATTGGCTCCCAGTACCTGCTCAACTGGTGTGTGCAGATCGCAAAG

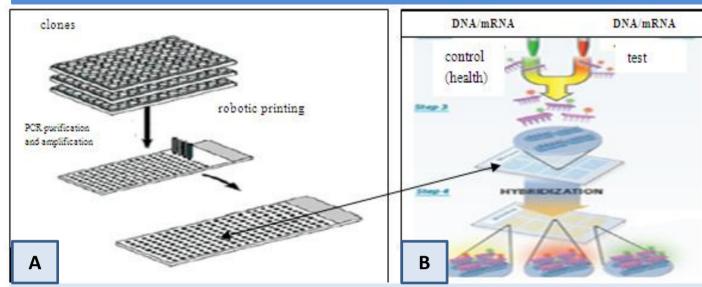
### EXON 21: 177688-177843

For example, exon 21 with convereted nucleotide G into A, and corresponding patient class 25

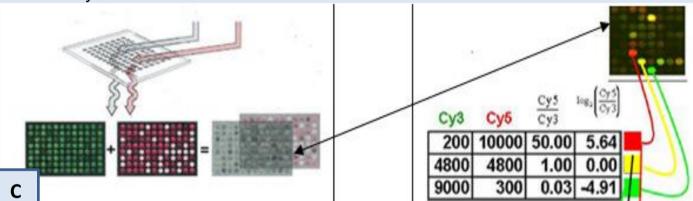
GGCATGAACTACTTGGAGGACCGTCGCTTGGTGCACCGCGACCTGGCAGCCAGGAACGTACTGGT AAGAATACCATGCAGAAGGAGGCAAA

# **MICROARRAY GENE COMPARATOR**

## **MICROARRAY EXPERIMENT**



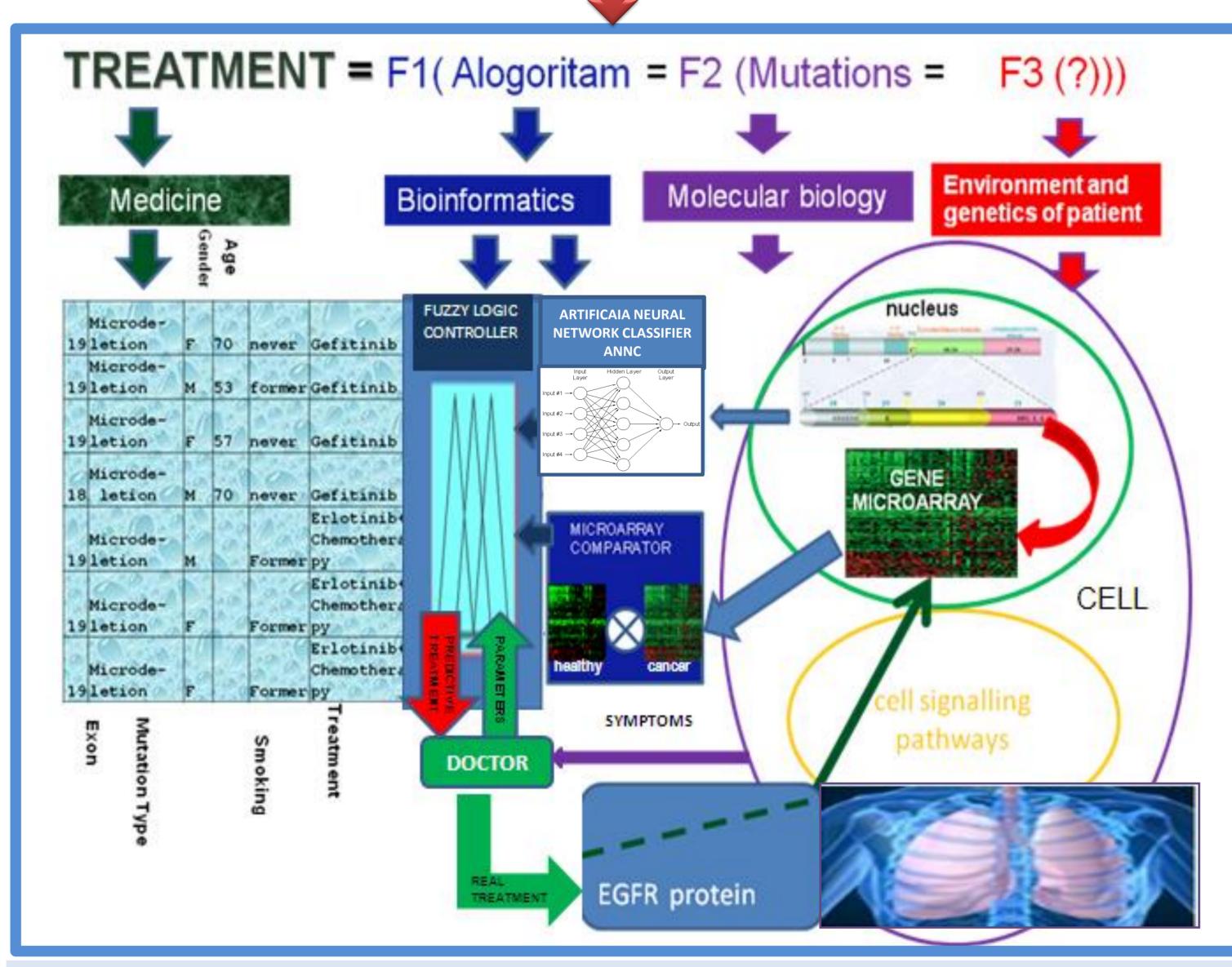
A)Genes of interest are made available as DNA clones. A polymerase chain reaction is used to amplify each gene to a sufficient amount to allow 'printing' onto array.B) RNA is isolated from reference (control) cells and test (experimental) cells. cDNA is synthesized from each RNA population by a reverse transcription process. The resulting two cDNA samples (control and experimentation) are labeled with two different fluorescent dyes, mixed, and hybridized to the targets on the microarray



# **INTELLIGENT COMPUTER AIDED** SIMULATOR FOR NSCLC TREATMENT

# **ARTIFICIAL NEURAL NETWORK CLASSIFIER**

## **STATISTICS FOR MUTATION GENERATOR**

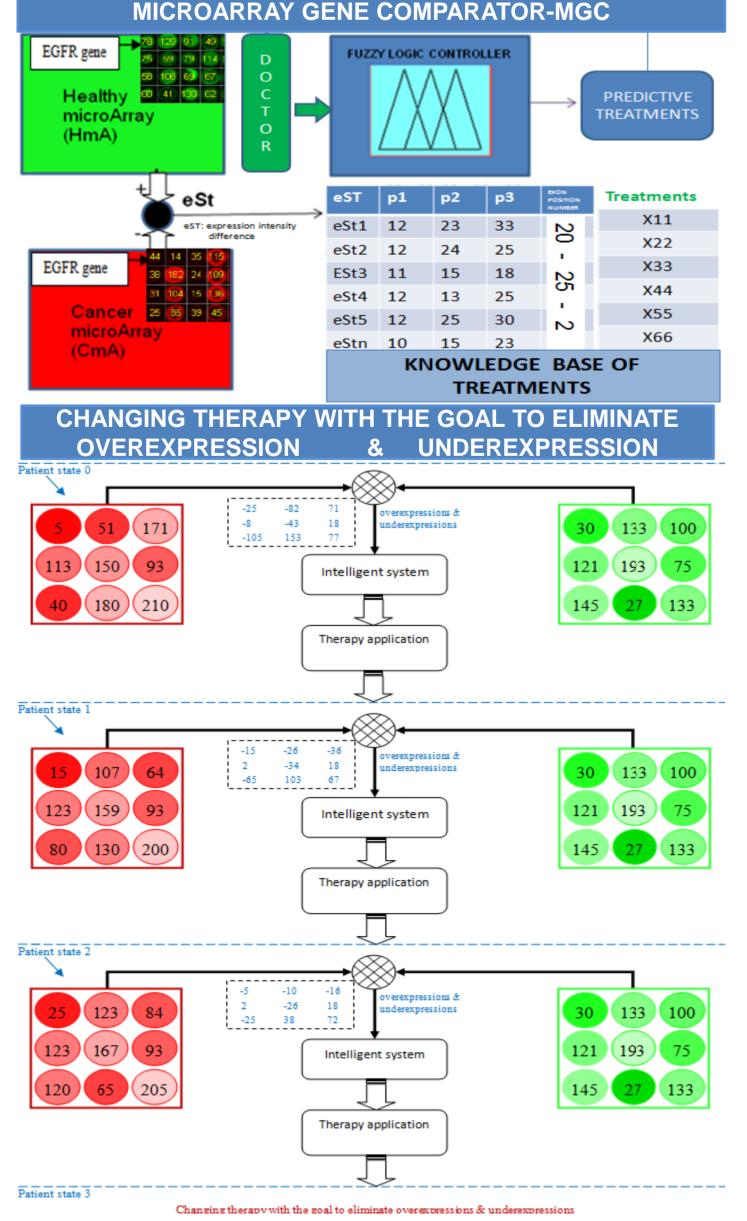


1	2	3	4	5	6	7	8
1	160691	160789	50	64	166	19	'-'
2 2	160691	160789	51	65	60	19	'-'
3 3	160691	160789	69	92	1	19	1-1 1
4 4	160691	160789	55	72	33	19	'-'
5 5	160691	160789	55	69	6	19	'-'
6 6	160691	160789	54	71	7	19	'-'
7 7	160691	160789	52	66	7	19	'-'
8 8	160691	160789	53	67	2	19	1-1 1
9 9	160691	160789	53	70	2	19	'-'
10 10	160691	160789	52	69	3	19	1-1 1
11 11	160691	160789	54	62	2	19	1-1 1
12 12	160691	160789	54	68	1	19	1-1 1
13 13	160691	160789	60	68	1	19	1-1
14 14	160691	160789	68	91	2	19	1-1 1
15 15	167262	167447	25	26	2	20	1-1 1
16 16	160691	160789	51	56	1	19	1-1
17 17	159890	160012	96	97	1	18	1-1
18 18	160691	160789	53	62	1	19	1-1
19 19	160691	160789	69	70	1	19	1-1
20 20	160691	160789	50	51	2	19	1-1 1
21 21	160691	160789	55	66	3	19	1-1 1
22 22	160691	160789	44	51	1	19	1-1 1
23 23	177688	177843	103	103	305	21	'G>T'
24 24	177688	177843	112	112	7	21	'A>T'
25 25	177688	177843	7	7	2	21	'G>A'
26 26	177688	177843	103	104	2	21	'TG>GT'

## **MUTATION PREDICTION GENERATOR**

EXON 18	7627	MUTATED NUCLEOTIDES
EXON 19	4951	MUTATED NUCLEOTIDES
EXON 20	17392	MUTATED NUCLEOTIDES

C) Hybridization is monitored through measurement of the amount of fluorescence associated with individual target molecules on the array. Since the two dyes fluoresce at different wavelengths, it is possible to distinguish hybridization due to the control cDNA population from that due to the test cDNA population. D) Mixture of control cDNApopulation with test cDNA population. R/G ratio represents relative abundance of transcripts.



The goal of this integration is to obtain a simulator that will process the appropriate inputs and generate predictive treatment. The inputs into Fuzzy Logic Controller-FLC are: (1) value of signal eSt, expression intensity difference between healthy microarray, and cancer microarray, (2) parameters from doctors (symptoms of patient), and (3) output from Artificial Neural Network Classifier-ANNC that means mutated exon, position in exon and number of mutated nucleotides. The outputs from FLC are: (1) type of therapy (for example, Gefitinib), (2)time duration of drug, (3) quantity of drug, (4) rules of behavior, (5) nutrition rules.

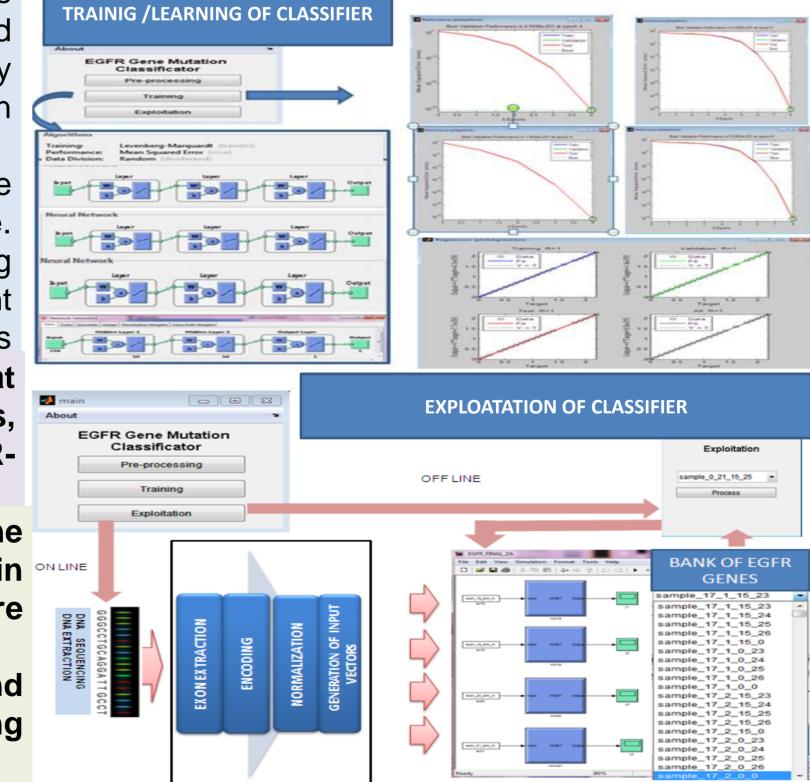
Knowing the principles and outputs of each module, we can now follow the functioning of our complete system. ANNC for input DNA sample reveals whether the observed gene is a healthy or a mutated one. Or, to be more precise, it identifies the exon of more of them on which mutations are present. According to mutations, the algorithm detects a certain treatment, or their combination, from the treatment knowledge base. A physician is free to accept that treatment as it is, or to adjust it to his previous New cancer treatments target and turn off EGFR signals. These therapies use antibodies that recognize only EGFR, stick to it, and block it from sending messages. By interrupting the signals, cancer cells are no longer told to overgrow, and eventually die. These treatments are called EGFRtargeted therapies.

•Gefitinib (Iressa) inhibits EGFR tyrosine kinase domain by binding to the adenosine triphosphate(ATP)-binding site of the enzyme. Thus the function of the EGFR tyrosine kinase in activating the anti-apoptotic RAS signal transduction cascade is inhibited, and malignant cells are inhibited.

It is not enough to perform the classification of patients into one already wellknown statistical group, but if there is a patient with a new mutated exons, our algorithm need to determine the position and number of mutated nucleotides. Because of that we have developed more powerfull mutation prediction generator for microdeletion mutations that take place over the nucleotides in consecutive order (the sum of all mutations with one deletion, two deletions and so on to the number of microdeletion corresponding to the length exon.

## **PREPROCESSING-GRAPHICAL USER INTERFACE**

	NO 007708	- 4. Encode Sequer			
NCBI / Accession Number	NG_007726	Encode	exon_18	3	- Show Data
LOCAL / .mat File	egfr_gene_dna.mat		-	2	
Get Data Show Data		5. Normalize Sequ	iences—		
2. Extract Exons		Normalize	exon_18	3	Show Data
tatistics File lung_cance	er_statistics.mat				
Extract Show Exons	Show Stats	6. Building training	g data set	s	
		exon_20	-	5	Build
Generate Mutated Sequences					Show Input >
Statistical / # sequences	100	ANN	-		Show Target 2
Predictive					
	_				



•Erloinib (Tarceva) specifically targets the EGFR tyrosine kinase, which is highly expressed and occasionally mutated. It binds in a reversible fashion to the adenosine triphosphate (ATP) binding site of the receptor.

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Patient state N

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# PERSPECTIVES

#### **TREATMENT = F1(** Alogoritam = F2 (Mutations = F3 (?))

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## DEVELOPMENT

Microarray miRNA will provide Comparator which information on changes of miRNAs that target EGFR gene during the process of treatment. But there are open questions as how many targeted genes would be included and how many miRNAs we need to include?

Cell Signaling Pathways Module, which will enable us to monitor signals (downstream from EGFR protein to EGFR gene) through the regulatory miRNAs that target EGFR gene in continuous mode, not only in switching mode. But there are open question as how many reactions (interactions) we need to include?

# **FUTURE PLAN**

Due to combinatorial nature of all possible mutations calculated number of mutated combinations would be:

EXON 18	<b>2</b> <sup>123</sup>	1.063 x 10 <sup>37</sup>
EXON 19	<b>2</b> <sup>99</sup>	6.33 x 10 <sup>29</sup>
EXON 20	<b>2</b> <sup>186</sup>	9.80 x 10 <sup>55</sup>
EXON 21	2 <sup>74</sup> · 3 <sup>35</sup>	9.45 x 10 <sup>38</sup>

To develop prediction mutation generator that cover all combinations of distributed nucleotides mutations related to positions in exons and number of mutated nucleotides would be very long process, and is not in line with human life. Because of that we look for a solution in new technologies (parallel multiprocessing cores), and the application of new models based on fusion of artificial intelligence methods.