

RESPIRATORY SOCIETY CONGRESS – SARAJEVO 2013

Faculty of Electrical Engineering
Department of Computer Science

Faculty of Medicine
Center for Genetics

COMPUTER AIDED LUNG CANCER CLASSIFICATION OF MUTATED EGFR EXONS USING ARTIFICIAL INTELLIGENCE METHODS

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Problem Domain Description

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Research fields
Artificial Intelligence, Bioinformatics, Molecular Genetics, Biomedical Engineering

Molecular diagnostics is a rapidly advancing field in which insights into:
microarray gene expression analysis,
microarray miRNA expression analysis,
cell signaling pathways,
 may provide invaluable information on:
disease pathology,
disease progression,
resistance to treatment,
response to cellular microenvironments,
 and ultimately may need fusion of:
artificial intelligent methods,
bioinformatics,
mathematics & statistics

TO IMPROVE EARLY DIAGNOSIS AND CREATE INNOVATIVE THERAPEUTIC APPROACHES FOR 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 data with microdeletion mutations generated in own generator required for simulation. We have developed an integrated software suit based on module for preprocessing data (extraction, encoding, and normalization), module for exon microdeletions generation (statistical data base), module for training/learning of artificial neural networks (ANN), and module for postprocessing (classification, and evaluation). We have made experiments on eleven different training/learning algorithms in combination with different number of cells, layers, and activation functions. The best results have been achieved with cascade-forward backpropagation algorithm based on Levenberg-Marquardt learning mechanism, including best performance (error 9.3083e-031) with the minimum epochs (training iterations 8), and the regression fit curves (training, validation and testing R=1). Dividing training sets into smaller subsets using ensemble-voting methods on each exons would improve system reliability in situations of possible single neural network fail.

Lung Cancer

The main types of lung cancer are:
 1. **non-small-cell lung carcinoma (NSCLC)**, and
 2. **small-cell lung carcinoma (SCLC)**, also called oat cell cancer.

There are three common forms of NSCLC:
 1.1 Adenocarcinomas are often found in an outer 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.

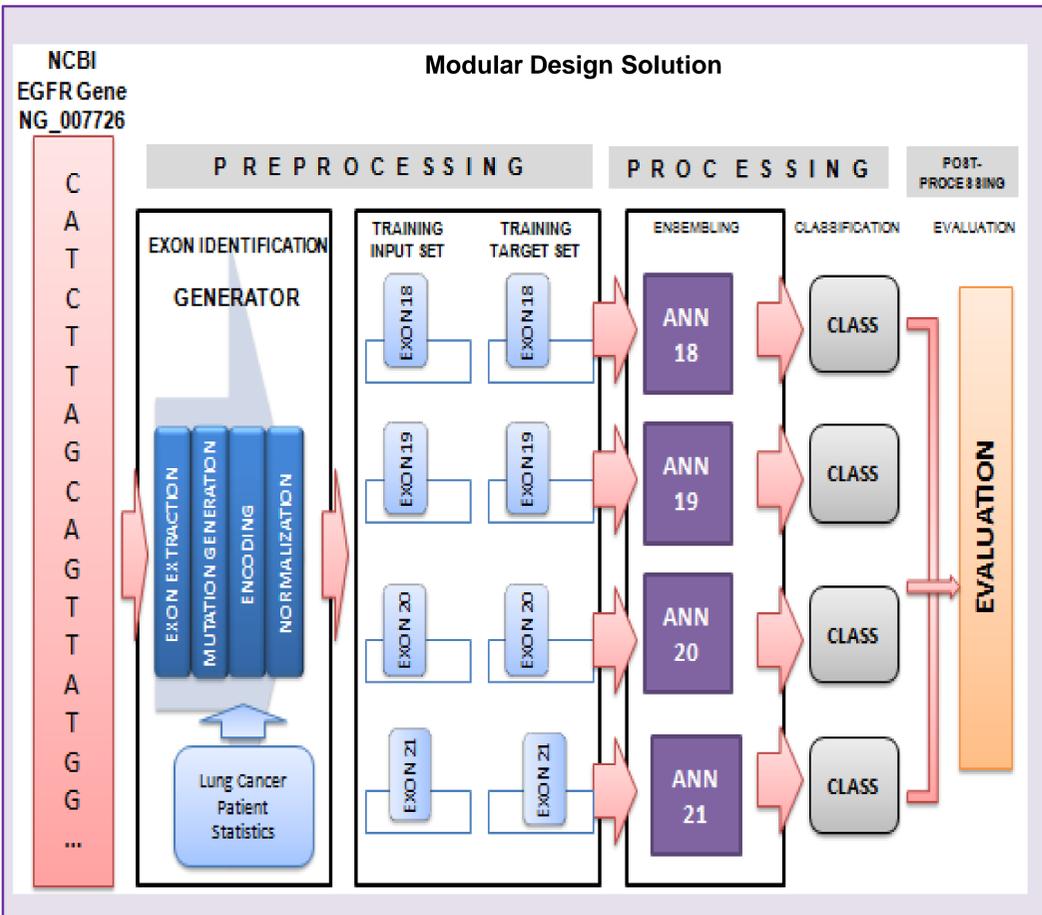
There are two different types of SCLC:
 2.1 Small cell carcinoma (oat cell cancer).
 2.2 Combined small cell carcinoma.

Statistics on Lung Cancer

- The most common cause of lung cancer is long-term exposure to tobacco smoke which causes 80%-90% of lung cancers.
- Nonsmokers account for 10-15% of lung cancer cases and these cases are often attributed to a combination of genetic factors, radon gas, asbestos and air pollution including second hand smoke.
- Different combinations of mutations (micro-deletions) exist within the EGFR kinase, and the most frequently observed mutations are on the exon 18, 19, 20 and 21.

Statistics for Mutation Generator

lung_cancer_mutation_statistics	POSITIONS OF MUTATIONS				PATIENTS	EXONS	CONVERSIONS	
	1	2	3	4	5	6	7	8
1	160691	160789	50	64	166	19	-	-
2	160691	160789	51	65	60	19	-	-
3	160691	160789	69	92	1	19	-	-
4	160691	160789	55	72	33	19	-	-
5	160691	160789	55	69	6	19	-	-
6	160691	160789	54	71	7	19	-	-
7	160691	160789	52	66	7	19	-	-
8	160691	160789	53	67	2	19	-	-
9	160691	160789	53	70	2	19	-	-
10	160691	160789	52	69	3	19	-	-
11	160691	160789	54	62	2	19	-	-
12	160691	160789	54	68	1	19	-	-
13	160691	160789	60	68	1	19	-	-
14	160691	160789	68	91	2	19	-	-
15	167262	167447	25	26	2	20	-	-
16	160691	160789	51	56	1	19	-	-
17	159890	160012	96	97	1	18	-	-
18	160691	160789	53	62	1	19	-	-
19	160691	160789	69	70	1	19	-	-
20	160691	160789	50	51	2	19	-	-
21	160691	160789	55	66	3	19	-	-
22	160691	160789	44	51	1	19	-	-
23	177688	177843	103	103	305	21	GT>T	-
24	177688	177843	112	112	7	21	AT>T	-
25	177688	177843	7	7	2	21	GA>A	-
26	177688	177843	103	104	2	21	TG>GT	-



Graphical User Interface

Neural Network: Layer 1, Layer 2, Layer 3, Output

Algorithms: Levenberg-Marquardt (trainlm), Mean Squared Error (mse), Random (dividerand)

Progress: Epoch 0, 8 iterations, 1000. Performance: 0.468, 9.36e-31, 0.00. Gradient: 1.00, 5.10e-15, 1.00e-10. Mu: 0.00100, 1.00e-07, 1.00e-10. Validation Checks: 0, 0, 6.

Training State (plot:trainstate): Gradient = 5.0969e-016, at epoch 8. Mu = 1e-007, at epoch 8. Validation Checks = 0, at epoch 8.

Performance (plot:perform): Best Validation Performance is 9.0383e-031 at epoch 8.

Epidermal Growth Factor Receptor (EGFR) Targeted Therapy

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 EGFR-targeted therapies.

STATISTICS → DATA PREPROCESSING → TRAINING → ARTIFICIAL NEURAL NETWORK SYSTEM → TESTING → THERAPIES → CHOOSE THERAPY → USE? → YES → RTKI → DNA EXTRACTION → DNA SEQUENCING → STATISTICS

Patient ID	Pat No.	Mutation Type	Nucleotide Change	Researcher Author	Gender	Age	Ethnicity/ Putative	Smoking status	Treatment
2051	19	Microdeletion	c.2235_2249del	Sasaki H, Endo K, et al.	F		Putative Japanese		
2148	19	Microdeletion	c.2235_2249del	Takano T, Ohe Y, et al.	F	70	Putative Japanese	never	Gefitinib
2149	19	Microdeletion	c.2235_2249del	Takano T, Ohe Y, et al.	M	53	Putative Japanese	former	Gefitinib
2150	19	Microdeletion	c.2235_2249del	Takano T, Ohe Y, et al.	F	57	Putative Japanese	never	Gefitinib
2151	19	Microdeletion	c.2235_2249del	Takano T, Ohe Y, et al.	M	70	Putative Japanese	never	Gefitinib
2132	19	Microdeletion	c.2235_2249del	Eberhard DA, Johnson BE, et al.	M	N/A	Former		Erlotinib+Chemotherapy
2127	19	Microdeletion	c.2235_2249del	Eberhard DA, Johnson BE, et al.	F	N/A	Former		Erlotinib+Chemotherapy
2128	19	Microdeletion	c.2235_2249del	Eberhard DA, Johnson BE, et al.	F	N/A	Former		Erlotinib+Chemotherapy
2129	19	Microdeletion	c.2235_2249del	Eberhard DA, Johnson BE, et al.	M	N/A	Former		Erlotinib+Chemotherapy

Future Plan

In order to complete this work our next task is to develop a more powerful mutation prediction generator for exon 18, 19, and 20 covering 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). Mutations on exon 21 will be based on translations of one kind nucleotide to another (G -> T, G -> A, A -> T), and translation of a pair of nucleotides (TG -> GT). This work is only first step within our:

LONG TERM RESEARCH

GLOBAL APPROACH based on **microARRAY, microRNA AND CELL SIGNALLING PATHWAYS** using **FUSION OF ARTIFICIAL INTELLIGENCE METHODS** to improve **EARLY DIAGNOSIS AND INNOVATIVE METHODS FOR CANCER**

References

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