UNIVERSITY OF OSLO

Faculty of Mathematics and Natural Sciences

Exam in INF3490/4490 — Biologically Inspired Computing

Day of exam: December 9th, 2015

Exam hours: 09:00 – 13:00

This examination paper consists of 7 pages.

Appendices: 1
Permitted materials: None

Make sure that your copy of this examination paper is complete before answering.

The exam text consists of problems 1-35 (multiple choice questions) to be answered on the form that is enclosed in the appendix and problems 36-39 which are answered on the usual sheets (in English or Norwegian). Problems 1-35 have a total weight of 70%, while problems 36-39 have a weight of 30%.

About problem 1-35:

Each problem consists of a topic in the left column and a number of statements each indicated by a capital letter. Problems are answered by marking true statements with a clear cross (X) in the corresponding row and column in the attached form, and leaving false statements unmarked. Each problem has a variable number of true statements, but there is always *at least one* true and false statement for each problem. 0.5 points are given for each marked true statement and for each false statement left unmarked, resulting in a score ranging from 0 to 60.

You can use the right column of the text as a draft. The form in the appendix is the one to be handed in (remember to include your candidate number).

Problem 1

Biologically	A	Optimization	
inspired	В	Modelling	
computation is	C	Safety critical systems	
appropriate for	D	Simulation	

Exhaustive search	Α	Not guaranteed to find the optimal solution	
	В	Test all possible solutions, pick the best	
	C	Relevant for continuous problems by using approximation	
	D	Most relevant for large search problems	

Which of the	A	Travelling salesman problem	
following are	В	Robot control	
discrete	С	Chess playing program	
optimization	D	Prediction of stock prices	
problems?		•	

Problem 4

Gradient ascent	A	The direction of the move is towards a larger value	
	В	Relevant for discrete optimization	
	C	Is not guaranteed to find the optimal solution	
	D	The ascent continues until the gradient is very small	

Problem 5

Exploration in	A	Concerned with improving the current best solution by	
search is		local search	
	В	Combined with exploitation in evolutionary algorithms	
	С	Often resulting in getting stuck in local optima	
	D	Concerned with global search	

Problem 6

What controls the	A	Time	
search in simulated	В	Temperature	
annealing?	С	Initial solution	
	D	Final solution	

Problem 7

Evolutionary	A	Individuals are normally generated randomly	
algorithm:	В	Is concerned with generating candidate solutions	
Initialization	C	Mutation of candidates is normally also taking place during	
		the initialization	
	D	Heuristics for generating candidates can be applied	

Problem 8

Evolutionary	A	Is a selection operator	
algorithm:	В	Act on population level	
Variation operators	С	Act on individual level	
	D	Are crossover and mutation	

Evolutionary	Α	Also known as crossover	
algorithm:	В	Combines elements of two or more genotypes	
Recombination	C	Also known as mutation	
	D	Also known as representation	

Evolutionary	A	Is often stochastic	
algorithm:	В	Also known as replacement	
Survivor selection	С	Can be fitness based	
	D	Can be age based	

Problem 11

Evolutionary	Α	Several termination criteria can be combined	
algorithm:	В	Determines when to compute the fitness for a population	
Termination	С	Is checked in every generation	
condition	D	Should be avoided to get faster evolution	

Problem 12

Permutation	A	Is used for problems where each variable can only appear	
representation		once	
	В	Bit-flip mutation is applicable	
	С	A mutation operator that swaps at least two values is applicable	
	D	Is used for problems where each variable can appear multiple times	

Problem 13

Tree representation	Α	Is used in Genetic Programming	
	В	Mutation results in replacing a randomly	
		chosen subtree by a randomly generated tree	
	С	Not suited for representing computer programs	
	D	Is used in Genetic Algorithms	

Problem 14

Selection pressure	Α	Should be high to avoid premature convergence	
	В	The higher pressure, the harder for the fittest solutions to	
		survive	
	C	Fitness-proportionate selection avoids selection pressure	
	D	Rank-based selection can adjust and control the pressure	

Problem 15

Rank based	Α	Use relative rather than absolute fitness	
selection	В	Use absolute rather than relative fitness	
	C	Results in less control of the selection pressure than	
		fitness-proportionate selection	
	D	Ranking can be either linear or non-linear	

Multimodality	A	In crowding, offspring competes with their nearest parent	
	В	In fitness sharing, the fitness decreases if there are many	
		candidates in a niche	
	C	The problem has only one locally optimal solution	
	D	Periodic migration is not relevant in the island model	

Simple Genetic	Α	Children compete with parents in survival selection	
Algorithm (GA)	В	Both crossover and mutation are applied in each generation	
	С	The whole population is replaced with the resulting	
		offspring	
	D	Uses real-valued representation	

Problem 18

Evolutionary	A	(μ,λ) : Select survivors among parents and offspring	
Strategies (ES)	В	(μ+λ): Select survivors among parents and offspring	
	C	(μ-λ): Select survivors among offspring only	
	D	(μ:λ): Select survivors among offspring only	

Problem 19

What is most	Α	Do multiple runs until a good solution is found	
important to be	В	Execute one run until the solution is good enough	
concerned with in	С	Get a reasonably good solution every time	
the evolution of	D	Get a very good result just once	
repetitive		, c	
problems?			

Problem 20

What are normally	A	Number of evaluations	
the two best	В	Elapsed time	
measurement units	С	CPU time	
for an evolutionary	D	Number of generations	
algorithm?			

Problem 21

Multiobjective	A	The travelling salesman problem is an example of a MOP	
optimisation	В	Concurrent optimisation of n possibly conflicting	
problems (MOPs)		objectives	
	C	The Pareto front represents the best solutions found	
	D	The Pareto front consists of dominated solutions	

Problem 22

Learning in neural	Α	Learning takes place in the neurons	
networks	В	An error is computed on axon outputs in the human brain	
	С	Learning takes place in the connections between neurons	
	D	Weights in a perceptron represent the strengths of synapses	

Supervised	Α	Desired outputs are not included	
learning	В	Desired outputs are included	
	С	Error between desired outputs and actual outputs are	
		computed during training	
	D	The multi-layer perceptron is trained by supervised	
		learning	

Artificial neural	A	Are trained by adjusting the network size	
networks	В	Are trained by adjusting weights	
	С	The weights are either all positive or all negative	
	D	The learning rate controls the amount of weight change	

Problem 25

Why use Multi	Α	Faster learning			
Layer Perceptron	В	asier programming			
instead of a single	С	Can solve more complex problems			
layer perceptron?	D	Can learn multiple decision boundaries			

Problem 26

When can the	A	In the forward pass	
weights be	В	In the backward pass	
adjusted in a	C	In both forward and backward pass	
multilayer	D	After computing output values of each training vector	
perceptron?			

Problem 27

The activation	Α	Does thresholding to 0 or 1	
function in a	В	Is used to compute the output value of a node	
multilayer	С	Is used for initialization of the network	
perceptron	D	Makes it possible to train non-linear decision boundaries	

Problem 28

Cartesian Genetic	A	s more restricted than the general Genetic Programming				
Programming	В	In evolving circuits, the genes determines function and				
		input to each node				
C The l		The level back parameter decides the number of columns in				
	the node-array					
	D	The problem of bloat is larger than for the general Genetic				
		Programming				

Swarm intelligence	A	Global behaviour appears as a result of centralized control	
	В	In Particle Swarm Optimization, velocity and position of	
		particles are updated	
	C Communication through the environment is called		
		stigmergy	
	D The probability of choosing a new edge in ant colony		
		optimization is proportional with the pheromone level of	
		the edge	

Support vector	Α	Only data vectors defining the margins are needed to					
machines		resent the support vectors					
	В	Can only classify linearly separable data					
	C	Map inputs into a higher-dimensional space					
	D	Margins can be increased by using soft margins					

Problem 31

Ensemble learning	Α	Multiple classifiers are trained to be slightly different	
	В	Only the best classifier is applied after training	
	С	Training vectors can be assigned weights during training	
	D All training vectors available should be used for training		
		each classifier	

Problem 32

Principal	A	Performs mapping to higher dimensions	
component	В	Can be applied for feature extraction	
analysis	С	Components represent the directions along with the most	
		variation in the data	
	D	Is a non-linear transformation	

Problem 33

Unsupervised	A	Can be used for training with data sets containing only				
learning		inputs				
	В	No specific error function is used for training				
	C	Self-organizing maps are increasing dimensions in the data				
	D	A multi-layer perceptron can be trained with unsupervised				
		earning				

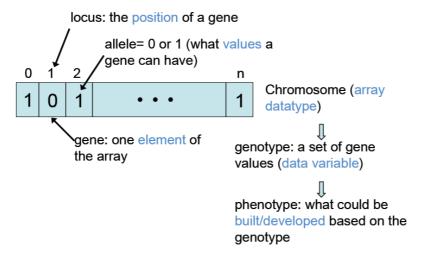
Problem 34

K-means	A	Need to know the number of clusters in advance		
clustering	В	Need to know which cluster a data point belongs to		
	C Each cluster center is moved most in the beginning			
	D The method always results in the global optimal solution			

Reinforcement	Α	The algorithm is told when the answer is wrong, and how					
learning		correct it					
	В	Is training using rewards					
	С	A policy defines how actions are chosen					
	D	A discount factor is used to discount future rewards					

Problem 36 (8%)

- a) Briefly explain the evolutionary algorithm terms chromosome, gene, locus and allele by including a figure of a chromosome.
- b) Explain briefly what a genotype and phenotype are and give an example of each of them.



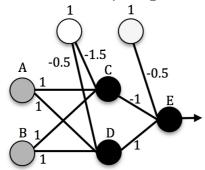
Problem 37 (5%)

In a population of three individuals, they have fitness 2, 3 and 5, respectively. What is the probability for selecting each of them when using a roulette wheel?

Total fitness= 2+3+5=10, thus, probability for selection is 1/10 for each of the fitness values: 0.2, 0.3 and 0.5.

Problem 38 (9%)

a) Show how the following multi-layer perceptron realizes a XOR-function by computing the output of each node and putting the results into a table:



Each perceptron accepts inputs being 0 or 1 and contains a threshold activation function.

В	A	D (BT)	C (BT)	D (AT)	C (AT)	E (BT)	E (AT)
0	0	-0.5	-1.5	0	0	-0.5	0
0	1	1-0.5	1-1.5	1	0	1-0.5	1
1	0	1-0.5	1-1.5	1	0	1-0.5	1
1	1	2-0.5	2-1.5	1	1	1-1-0.5	0

BT: Before threshold, AT: After threshold.

b) What values should the weights in the output layer have to make an inverted XOR function (XNOR)? All output layer weights must be negated (including to the bias). Weight_{CE}= 1, Weight_{DE}= -1 and Weight_{bias}= 0.5.

Problem 39 (8%)

List and explain, with one sentence each, up to four of the ethical recommendations for commercial robots the Euronet Roboethics Atelier came up with.

- *Safety:* There must be mechanisms (or opportunities for an operator) to control and limit a robot's autonomy.
- **Security:** There must be a password or other keys to avoid inappropriate and illegal use of a robot.
- *Traceability:* Similarly as aircraft, robots should have a "black box" to record and document their own behavior.
- *Identifiability:* Robots should have serial numbers and registration number similar cars.
- *Privacy policy:* Software and hardware should be used to encrypt and password protect sensitive data that the robot needs to save.

INF3490/INF4490 Answers problems 1 – 35 for candidate no:

Problem	A	В	C	D
1				
2				
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INF3490/INF4490 Answers problems 1 – 35 for candidate no:

Problem	A	В	C	D
1	0	0		0
2		0	0	
3	0		0	
5 6	0 0		0 0	0
5		0		0
6		0 0 0		
7	0	0		0
8			0	0
9	0	0		
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11 12	0		0	
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29		O	O	0
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31	0		O	
32		0	O	
33	0	0		
34	0		0	
35		О	0	0