ınit	question	option1	option2	option3	option4	answer
		Fuzzy Computing,	Francis Nationalisa and	A	Namel Cairner and Canadia	
1	Core of soft Computing is	Neural Computing, Genetic Algorithms	Fuzzy Networks and Artificial Intelligence	Artificial Intelligence and Neural Science	Neural Science and Genetic Science	option1
_	Who initiated the idea of Soft Computing	Charles Darwin	Lofti A Zadeh	Rechenberg	Mc_Culloch	option
				J	_	
				deals with		
				information which is		
				vague, imprecise,		
		mimics human	doesnt deal with 2	uncertain, ambiguous, inexact,		
1	Fuzzy Computing	behaviour	valued logic	or probabilistic	All of the above	option4
	, , ,					'
			information	Both (a) and (b)	None of the above	
1	Neural Computing	mimics human brain	processing paradigm			option3
				are adaptive		
				heuristic search		
				algorithm based on		
			inspired by Darwin's	the evolutionary		
			theory about	ideas of natural		
		Evolutionary		selection and	All of the above	
_1	Genetic Algorithm are a part of	Computing	the fittest"	genetics		option ²
		Improvised and	supervised and	Layered and		
1	What are the 2 types of learning	unimprovised	unsupervised	unlayered	None of the above	option2
		learning with the	learning without	learning with the	learning with computers	
1	Supervised Learning is	help of examples	teacher	help of teacher	as supervisor	option3
4	llanina italihannina ita	learning without	problem based	learning from		
1	Unsupervised learning is	computers	learning	environment	learning from teachers	option3
		Conventional	Conventional Artificial			
		Artificial Intelligence	Intelligence methods			
		deal with prdicate	are limited by symbols			
		logic where as soft	where as soft			
4	Conventional Artificial Intelligence is different from soft	computing deal with	computing is based on	Dath (a) and (b)	None of the above	antion?
	computing in the sense	fuzzy logic	empirical data	Both (a) and (b)	Notice of the above	option3
		classes are not		classes are not		
	In supervised learning	predefined	classes are predefined	required	classification is not done	option2
	ANN is composed of large number of highly interconnected					
	processing elements(neurons) working in unison to solve					ا ا
1	problems	TRUE	FALSE			option1
1	Artificial neural network used for	Pattern Recognition	Classification	Clustering	All of these	option4
		_		IF-The-Else Analysis		
1	A Neural Network can answer	For Loop questions	what-if questions	Questions	None of these	option2
	In artificial Nouval Notwork interconnects -					
1	In artificial Neural Network interconnected processing elements are called	nodes or neurons	weights	axons	Soma	option1
1	Neuron can send signal at a time.	Multiple	One	Two	Three	option2
		<u>'</u>				
			control actions are		algorithm which can easily	
			unambiguous and	Control action is	adapt with the change of	
1	Any soft-computing methodology is characterized with	precise solutions	accurate	formally defined	dynamic environment	option4
	An equivalence between Fuzzy vs. Probability to that of			Probability=Forecasti		
1	Prediction vs. Forecasting is	Fuzzy=Prediction	Fuzzy= Forecasting	ng	None of these	option2
	-					
			Artificial neural	In each, no precise	F	
	Both fuzzy logic and artificial nouval natives of	Poth gives presies	network gives	mathematical model	Fuzzy gives exact result but artificial neural	
1	Both fuzzy logic and artificial neural network are soft computing techniques because	Both gives precise and accurate results.	accurate result, but fuzzy logic does not.	of the problem is required	network does not.	option3
				Date of birth of a		570000
1	Which of the following cannot be stated using fuzzy logic?	Color of an apple	Height of a person	student	Speed of a car	option3
		Supervised training	Supervised training	Unsupervised		
	For the same size of training data as input, the fastest learning technique is	with gradient descent	with stochastic	training without	Unsupervised training	00+:
		error correction	method	error calculation	with Hebbian method.	option1

		Input layer	Hidden layer	Output layer		
1 In case of layer calculation, the r	naximum time involved in	computation.	computation.	computation.	Equal effort in each layer.	optio
				Both input and		
1 In supervised learning, training s	et of data includes	Input	Output	output	None	optio
			A	 		
			Artificial neural	In each, no precise		
			network gives	mathematical model	Fuzzy gives exact result	
Both fuzzy logic and artificial neu	iral network are soft	Both gives precise	accurate result but	of the problem is	but artificial neural	
computing techniques because,		and accurate results.	fuzzy logic does not.	required.	network does not.	optio
		Fuzzy logic in parallel				
		with the Genetic	Fuzzy logic controlled	Genetic algorithm		
Fuzzy – Genetic Hybrid system is		algorithm	Genetic algorithm	controlled Fuzzy logic	None of the above	optio
In which of the following, one ted	chnology calls the other					
technology as subroutine to proc	ess or manipulate	Embedded hybrid	Sequential hybrid	Auxiliary hybrid		
information needed		system	system	system	Parallel hybrid system	optio
		Embedded hybrid	Sequential hybrid	Auxiliary hybrid		
Which of the following is not a h	ybrid system?	system	system	system	Parallel hybrid system	optio
Command to start matlab fuzzy	toolbox is	fis	fuzzy	fuzzybox	fuzzytool	optio
		Supervised learning				
Training Perceptron is based on		technique.	Unsupervised learning	Reinforced learning	Stochastic learning	optio
				_		
A batch mode of training is gener	rally implemented through	Minimization of	Maximization of	Maximization of	Minimization of mean	
	in error calculation	median square error	median square error	mean square error	square error	optio
			,			1
is/are the way/	s to represent uncertainty.	Fuzzy Logic	Probabilty	Entropy	All of the mentioned	optio
is, are the way,	op. coone anoch tunity.	1 221 20810			or are mentioned	John
An artificial neurons receives n ir	inuts x1 x2 vn					
with weights w1,w2,,wn attac						
-	·					
weighted sum is compu	· ·					
non-linear filter ø calledactivation	n function to release the					l
output.		Σwi	Σχί	Σ xi+Σ wi	Σ χί.Σωί	optio
Who invented the Single-Layer P		Frank Rosenblatt	Marvin Minsky	Seymour Papert	None of these	optio
Japanese were the first to utilize	fuzzy logic practically on					
high-speed trains in Sendai.		True	False			optio
			Geographical			
Which AI system provides a diagr	nosis to a specific problem?	Intelligent agent	information system	Data mining system	Expertsystem	optio
				Because they are		
		Because they are the	Because they are the	the only	Because they are the only	
		only class of problems	only mathematical	mathematical	class of problems a	
Why are linearly separable proble	ems of interest to neural	that a network can	functions that are	functions you can	perceptron can solve	
network researchers?		solve sucessfully	continuous	draw	successfully	optio
		When there is an	When there is an OR	In De-Morgan's		
Where is the minimum criterion	used?	AND operation	operation	theorem	None of these	optio
		Error correction	Reinforcement			† <u>'</u>
Perceptron learning, Delta learn	ing and I MS learning are	learning - learning	learning - learning		Competitive learning -	
learning methods which falls und		with a teacher	with a critic	Hebbian learning	learning without a teacher	optio
Generally, AI systems analyze im		a teacher	a critic			Jopin
		Blurred data	Fuzzy logic	Inclusive information	Dirty data	on+i-
information. This information is o		Blurred data	Fuzzy logic	Inclusive information	Dirty data	optio
Which AI system will work for yo	u to linu information on	 	Name No.	Companie Alexandre	[<u> </u>
the internet?		Intelligent agent	Neural Network	Genetic Algorithms	Expertsystem	optio
Which AI system will continue to	analyze a problem until it	1.1.18			<u></u>	
finds the best solution?		Intelligent agent	Neural Network	Genetic Algorithms	Expertsystem	optio
Which Intelligent Agent will mon		[
back to you when there is a prob		Shopping bot	Buyer agent	Information agent	Predictive agent	optio
Which Intelligent Agent can play	an Internet game on your					
behalf?		Information agent	User agent	Predictive agen	Game agent	optio
In terms of computing we have		antecedent	consequent	mapping function	All of the mentioned	optio
			Hand written			
Example of hard computing		Robot movement	character recognition	money allocation	searching problem	optio
1						
Hard computing produce		precise solutions	fuzzy solution	approximate solution	None of these	optio
Hard computing is strictly		Parallel	sequential	Both	None of these	optio
1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2		T	granitation and the second			1 2000
Principle compunent of soft com	nuting is	Neural Network	Fuzzy Loguc	Genetic Algorithms	All of the mentioned	optio
			. ally logue	Concue Aigoritimis	or the mentioned	Jopin
Which is supervised learning		clustoring	rograssion	association	dimonsionality reduction	Cont.
Which is supervised learning		clustering	regression	association	dimensionality reduction	optio
Example of hybrid system is		Auxillary	Embedde	Neuro Fuzzy	Neuro Gastro	optio
		1	1	allows parallel	1	
		L. II	1.1		la de la compansión de la	
Soft computing is NOT Fuzzy logic is a form of		allows stochasticity Two valued logic	deterministic Many valued logic	computation Crisp set logic	tolerant of imprecision Binary set logic	optio

	T	T	1	I	1
2 Traditional set theory is also known as Crisp set theory.	True	False			option1
2 Traditional set theory is also known as crisp set theory.	Titue	raise	$\lim x \to -\infty \mu(x) = 0$		Орионт
	$\lim x \to -\infty \mu(x) = 1$	$\lim x \to -\infty \mu(x) = \lim$	and $\lim x \to +\infty$ (x) =	$\lim x \to -\infty \mu (x) = \lim x \to$	
2 A fuzzy set A is closed if:	1	$ x \rightarrow +\infty (x) = 0$	1	$+\infty$ (x) = 1	option2
How is Fuzzy Logic different from conventional control	and min x / 1 = (x) =	X / 1 = (x) = 0	1		Optionz
2 methods?	IF and THEN Approach	FOR Approach	WHILE Approach	DO Approach	option1
The height h(A) of a fuzzy set A is defined as h(A) = sup A(x)	ii diid iii Eiv Approdeii	1 OK Approach	WITTEL Approach	DO Approach	Ориона
where x belongs to A. Then the fuzzy set A is called normal					
2 when	h(A)=0	h(A)<0	h(A)=1	h(A)>1	option3
For k>1, which of the following concept can be used to	Concentration and	II(A)<0	(A)-1	(A)>1	Ориона
2 generate other linguistic hedge	Dilation	Dilation	Concentration	None of the above	ontion?
Z generate other linguistic nedge	Dilation	Dilation	Concentration	Notic of the above	option3
2) The manufaction of matical and another annually manual and in	Tabulas Fassa	Coonbinal Famo	NA-+1	Lasiaal Farm	
2 The membership functions are generally represented in	Tabular Form	Graphical Form	Mathematical Form	Logical Form	option2
		Fuzzy Algorithm,			
Three main basic features involved in characterizing	Intution, Inference,	Neural network,	Core, Support ,	Weighted Average, center	l
2 membership function are	Rank Ordering	Genetic Algorithm	Boundary	of Sums, Median	option3
Membership function defines the fuzziness in a fuzzy set					
irrespective of the elements in the set, which are discrete or					
2 continuous.	True	False			option1
Membership function can be thought of as a technique to					1
2 solve empirical problems on the basis of	knowledge	examples	learning	experience	option4
The region of universe that is characterized by complete					
2 membership in the set is called	Core	Support	Boundary	Fuzzy	option1
A fuzzy set whose membership function has at least one			,	·	<u> </u>
element x in the universe whose membership value is unity					
2 is called	sub normal fuzzy sets	normal fuzzy set	convex fuzzy set	concave fuzzy set	option2
2 In a Fuzzy set a prototypical element has a value	1	0	Infinite	Not Defined	option1
A fuzzy set whose membership function has no members	-		THIN ICC	Not Bernied	Ортюпі
2 whose membership value is 1	sub normal fuzzy sets	normal fuzzy set	convex fuzzy set	concave fuzzy set	option1
2 Whose membership value is 1	Sub Horrital Tuzzy Sets	Hormai ruzzy set	CONVEX TUZZY SEL	concave ruzzy set	Орион
values are strictly monotonically increasing or strictly monotonically decreasing or strictly monotonically increasing than strictly monotonically decreasing with					
2 increasing values for elements in the universe	sub normal fuzzy sets	normal fuzzy set	convex fuzzy set	concave fuzzy set	option3
The membership values of the membership function are nor strictly monotonically increasing or decreasing or strictly monoronically increasing than decreasing. The crossover points of a membership function are defined	convex fuzzy set	concave fuzzy set	non convex fuzzy set	non concave fuzzy set	option3
as the elements in the universe for which a particular fuzzy					
2 set has values equal to	infinite	1	lo	0.5	option4
If if x ia A then y is B else y is C . The output of the given	IIIIIIIICE	1		0.5	Орионч
2 fuzzy rule is	a fuzzy set	a crisp set	a fuzzy relation	a membership function	option3
2 The cardinality of the given set A={1,2,3,4,5}	4	5	2	3	+
The cardinality of the given set A-(1,2,5,4,5) The cardinality of the fuzzy sets defined with continuous	4	3	2	3	option2
1	infinit.		1		antion1
2 membership function on any universe is Two fuzzy sets A and B with membership functions µA(x)and	infinity	0	1	-1	option1
μ B(x), respectively defined as below. A = Hot Climate with					
μ A (x) as the MF B = Cold Climate with μ B(x)as the MF		(, , , , , , , , , , , , , , , , , , ,		,	l
2 Pleasant climate is given by	1 – μB(x)	max(μA(x),μB(x))	min(μA(x),μB(x))	1 – μA(x)	option
What is the Bandwidth of fuzzy set A which is given as					
follow A ={ (10, 0.1), (15, 0.2), (20, 0.5), (25, 0.4), (30, 0.4),					
2 (35, 0.5), (40, 0.2), (45, 0.1) }	10	15	20	25	option2
Given that "x is Sweet " with T(x) = 0.8 and" y is Sweet "					
with $T(y) = 0.6$. The fuzzy truth value of "If x is Sweet then y is					
2 Sweet " is:	0.2	0.4	0.6	0.8	option4
2 $(\sim (P \land Q) \Rightarrow R) \land P \land Q$ is equivalent to	(P ∧ Q)	(P ∧ Q)VR	Р	(~ P ∨ Q)	option1
	(A × B) + (B × C)	(A × B) ∪ (A ⁻ × C)	$(A \times B) \to (B \times C)$	(A × C) U (B × C)	option2
2 If if x ia A then y is B else y is C . Then relation is eqivalent to				· AUTOMATIC CONTROL	
2 If if x ia A then y is B else y is C . Then relation is eqivalent to	Wireless services,	Restrict power usage,	Cincolline 1 1 1 1	Automatic control,	
	Wireless services, heat control and	telephone lines and	Simulink, boiler and	decision analysis and data	
 2 If if x ia A then y is B else y is C . Then relation is eqivalent to 2 What are the applications of Fuzzy Inference Systems? 	Wireless services,		Simulink, boiler and CD recording	· ·	option4
2 What are the applications of Fuzzy Inference Systems?	Wireless services, heat control and printers	telephone lines and sort data	CD recording	decision analysis and data classification	option4
2 What are the applications of Fuzzy Inference Systems? Let X = {a, b, c, d} and Y = {1, 2, 3, 4} and A = {(a, 0.0), (b,	Wireless services, heat control and printers	telephone lines and sort data	CD recording 0 0.4 1 0.8	decision analysis and data classification 0 0.4 1 0.8	option4
2 What are the applications of Fuzzy Inference Systems? Let X = {a, b, c, d} and Y = {1, 2, 3, 4} and A = {(a, 0.0), (b, 0.8), (c, 0.6), (d, 1.0)}, B = {(1, 0.2), (2, 1.0), (3, 0.8), (4, 0.0)},	Wireless services, heat control and printers 0 0 0 0 0.2 0.8 0.8 0	telephone lines and sort data 1 1 1 1 0.2 0.8 0.8 0.2	CD recording 0 0.4 1 0.8 0.2 0.8 0.8 0.2	decision analysis and data classification 0 0.4 1 0.8 0 0.2 0.2 0.2	option4
2 What are the applications of Fuzzy Inference Systems? Let X = {a, b, c, d} and Y = {1, 2, 3, 4} and A = {(a, 0.0), (b,	Wireless services, heat control and printers	telephone lines and sort data	CD recording 0 0.4 1 0.8	decision analysis and data classification 0 0.4 1 0.8	option4

2	what are the following sequence of steps taken in designing a fuzzy logic machine?	Fuzzification -> Rule Evaluation> Defuzzification	Rule Evaluation >Fuzzification - >Defuzzification	Defuzzification >Rule Evaluation >Fuzzification	Fuzzy Sets >Defuzzification>Rule Evaluation	option1
		LE TUEN ELCE : L	LE TUEN - L	Both IF-THEN-ELSE	No. of the control of	
2	Fuzzy logic is usually represented as	IF-THEN-ELSE rules	IF-THEN rules	rules & IF-THEN rules	None of the mentioned	option2
2	Which of the following is not true regarding the principles of fuzzy logic?	Fuzzy logic follows the principle of Aristotle and Buddha	Japan is currently the most active users of fuzzy logic	Fuzzy logic is a concept of 'certain degree'	Boolean logic is a subset of fuzzy logic	option1
2	The room temperature is hot. Here the hot (use of linguistic variable is used) can be represented by	Fuzzy Set	Crisp Set	Fuzzy & Crisp Set	None of these	option1
2	Considering a graphical representation of the 'tallness' of people using its appropriate member function, which of the following combinations are true? (I)TALL is usually the fuzzy subset. (II) HEIGHT is usually the fuzzy set. (III) PEOPLE is usually the universe of discourse	1,11,111	I,II	1,111	11,111	option1
	, , , , , , , , , , , , , , , , , , , ,	, ,	,	,	,	-,
ı		A fuzzy sytem can	The conversion of	A continoussytem		
2	What is the Fuzzy Approximation Theorem(FAT) ?	model any continoussytem	fuzzy logic to probability.	can model a fuzzy system	Fuzzy patches covering a series of fuzzy rules	option2
	What is the main difference between probability and fuzzy logic	Fuzzy logic is probability in disguise	Fuzzy logic is the likelihood of an event occuring and probability is the extent of that event	Probability is ADDITIVE, meaning all its values must add up to one	Probability dissipates with decreasing information	option1
	Fuzzy Set theory defines fuzzy operators. Choose the fuzzy	probability in disguise	extent of that event	add up to one	decreasing information	Optioni
2	operators from the following.	AND	OR	NOT	All of the mentioned	option4
2	There are also other operators, more linguistic in nature, called that can be applied to fuzzy set theory.	Hedges	Lingual Variable	Fuzzy Variable	None of the mentioned	option1
	Consider a fuzzy set A defined on the interval X = [0, 10] of	Treages	Elligadi variable	Tuzzy variable	None of the mentioned	Optioni
2	integers by the membership Junction $\mu A(x)=x\ /\ (x+2)$ Then the α cut corresponding to α = 0.5 will be	{0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10}	{1, 2, 3, 4, 5, 6, 7, 8, 9, 10}	{2, 3, 4, 5, 6, 7, 8, 9, 10}	None of these	option3
	If A and B are two fuzzy sets with membership functions: $\mu a(\chi) = \{0.2,0.5.,0.6,0.1,0.9\}$ $\mu b \; (\chi) = \{0.1,0.5,0.2,0.7,0.8\}$					
2	then the value of μa ∩ μb will be	{0.2,0.5,0.6,0.7,0.9}	{0.2, 0.5,0.2, 0.1,0.8}	{0.1, 0.5, 0.6, 0.1,0.8}	{0.1, 0.5, 0.2, 0.1,0.8}	option4
2	A point of a fuzzy set A is a point $x \in X$ at which $\mu A(x) = 0.5$	Core	Support	Cross-over	None of these	option3
	Given U = {1,2,3,4,5,6,7} A = {(3, 0.7), (5, 1), (6, 0.8)}		{(4, 0.3.): (5, 0), (6.	{(I, 1), (2, 1), (3, 0.3),		
2	then A $^{\sim}$ will be: (where $^{\sim} \rightarrow$ complement) In Lamda-cut method the value of λ can be	{(4, 0.7), (2,1), (1,0.8)} Greater than 10		(4, 1), (6,0.2), (7, 1)}	{(3, 0.3), (6.0.2)}	option3
	Suppose, a fuzzy set Young is defined as follows Young =	Oreater tilali 10	Between 1 and 10	Between 0 and 1	Any value	option3
2	(10, 0.5), (20, 0.8), (30, 0.8), (40, 0.5), (50, 0.3) Then the crisp value of Young using MoM method is	20	25	30	35	option2
	If the fuzzy set has two sub regions, then the centre of	with the meditf	with the most of all			
2	gravity of the sub region can be used to calculate the defuzzified value.	with the median of all the area	with the mean of all the area	with the largest area	with the smallest area	option3
	Sales are delegatived value.	Centre of gravity	Centre of sum	Centre of area	are smallest area	350000
	Which of the following is not a centroid method?	method (CoG)	method (CoS)	method (CoA)	Centre of Mass (CoM)	option4
2	Let A be a fuzzy set. Then 1-cut of A is usually called	support	height	core	alpha-cut	option3
2	Each fuzzy complement has at mostequilibrium.	1	2	3	None of these	option1
	Equilibrium of a fuzzy complement c is a solution of the					.,
2	equation	c(a)-a=1	c(a)-a=2	c(a)=2a	c(a)-a=0	option4
า	Defuzzification is done to obtain	Crisp output	Fuzzy Output	The best rule to	None of those	ontion1
	Defuzzification is done to obtain If Z is a set of elements with a generic element z, i.e. Z = {z}, then this set is called	Crisp output Universe set	Fuzzy Output Universe of discourse	follow Derived set	None of these None of these	option1 option2
	A fuzzy convexity is set			increasing and then		27.0112
2	, , , ,	increasing	decreasing	decreasing	All of the mentioned	option4

3 What 3 When 3 What 3 What 4 What 3 Operat 3 What	at Is Fuzzy Inference Systems? at Are The Two Types Of Fuzzy Inference Systems? ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems?	Used to respond to questions in a humanlike way The process of formulating the mapping from a given input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables The input is a single	A new programming language used to program animation The process of formulating the mapping from a given input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations. Defuzzification	c) The result of fuzzy thinking Having a larger output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the antecedent	Having a smaller output than the input Mihni-Type and Sujgani-Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option1 option3 option4 option4 option3
3 What 3 When 3 What 3 What 4 What 3 Operat 3 What What 4 What 4 What 4 What 5 Operat 6 What 6 What 7 What 7 What 7 What 8 What 9 What 9 What 1 What 1 What 1 What 1 What	at Is Fuzzy Inference Systems? at Are The Two Types Of Fuzzy Inference Systems? are Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? and Is Another Name For Fuzzy Inference Systems?	The process of formulating the mapping from a given input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	program animation The process of formulating the mapping from a given input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	thinking Having a larger output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	one Having a smaller output than the input Mihni-Type and Sujgani-Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option1 option3 option4 option4
3 What 3 When 3 What 3 What 4 What 3 Operat 3 What What 4 What 4 What 4 What 5 Operat 6 What 6 What 7 What 7 What 7 What 8 What 9 What 9 What 1 What 1 What 1 What 1 What	at Is Fuzzy Inference Systems? at Are The Two Types Of Fuzzy Inference Systems? are Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? and Is Another Name For Fuzzy Inference Systems?	The process of formulating the mapping from a given input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	The process of formulating the mapping from a given input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Having a larger output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Having a smaller output than the input Mihni-Type and Sujgani-Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option1 option3 option4 option4
3 What 3 What 3 What 4 What 3 What 3 What 4 What 4 What 5 Operat 6 What 7 What 8 What 9 What 1 What 1 What 1 What 1 What 1 What	at Is Fuzzy Inference Systems? at Are The Two Types Of Fuzzy Inference Systems? are Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? andani's Fuzzy Inference Method Was Designed To mpt What?	formulating the mapping from a given input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	formulating the mapping from a given input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	than the input Mihni-Type and Sujgani- Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 What 3 What 3 What 4 What 3 What 3 What 4 What 4 What 5 Operat 6 What 7 What 8 What 9 What 1 What 1 What 1 What 1 What 1 What	at Is Fuzzy Inference Systems? at Are The Two Types Of Fuzzy Inference Systems? are Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? andani's Fuzzy Inference Method Was Designed To mpt What?	formulating the mapping from a given input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	formulating the mapping from a given input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	than the input Mihni-Type and Sujgani- Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 What 3 What 3 What 4 What 3 What 3 What 4 What 4 What 5 Operat 6 What 7 What 8 What 9 What 1 What 1 What 1 What 1 What 1 What	at Are The Two Types Of Fuzzy Inference Systems? ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	input to an output using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	input to an output using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	than the input Mihni-Type and Sujgani- Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 What 3 What 3 What 4 What 3 What 3 What 4 What 4 What 5 Operat 6 What 7 What 8 What 9 What 1 What 1 What 1 What 1 What 1 What	at Are The Two Types Of Fuzzy Inference Systems? ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	using fuzzy logic Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	using fuzzy logic Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	output than the input Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	than the input Mihni-Type and Sujgani- Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 What 3 What 3 What 4 What 3 What 3 What 4 What 4 What 5 Operat 6 What 7 What 8 What 9 What 1 What 1 What 1 What 1 What 1 What	at Are The Two Types Of Fuzzy Inference Systems? ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	Model-Type and System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	Momfred-Type and Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Mamdani-Type and Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Mihni-Type and Sujgani-Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 Where 3 What 4 Mamo 3 Attem 3 What I 4 What 3 Operat 3 What What What	ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	System-Type Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	Semigi-Type Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Sugeno-Type Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Type Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
3 Where 3 What 4 Mamo 3 Attem 3 What I 4 What 3 Operat 3 What What What	ere Has Fuzzy Inference Systems Been Implemented? at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	Wireless services, heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	Wireless services, heat control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Simulink, boiler and CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Automatic control, decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
Mamo 3 Attem 3 What I What 3 Operat 3 What What	at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	heat control and printers Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	decision analysis and data classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
Mamo 3 Attem 3 What I What 3 Operat 3 What What	at Is Another Name For Fuzzy Inference Systems? mdani's Fuzzy Inference Method Was Designed To mpt What?	Fuzzy Expert System Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	control and printers Fuzzy Modelling Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	CD recording Fuzzy Logic Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	classification All of the mentioned Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option4
Mamo 3 Attem 3 What I What 3 Operat 3 What	ndani's Fuzzy Inference Method Was Designed To mpt What?	Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Controller Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	
Mamo 3 Attem 3 What I What 3 Operat 3 What	ndani's Fuzzy Inference Method Was Designed To mpt What?	Control any two combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	Control a television and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	Control a steam engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	Control a air craft and feul level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	
3 What I What 3 Opera: 3 What What What	ndani's Fuzzy Inference Method Was Designed To mpt What?	combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	ndani's Fuzzy Inference Method Was Designed To mpt What?	combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	ndani's Fuzzy Inference Method Was Designed To mpt What?	combinations of any two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	and remote combination by synthesising a set of linguistic control rules obtained from experienced human operations.	engine and a boiler combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	ndani's Fuzzy Inference Method Was Designed To mpt What?	two products by synthesising a set of linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	combination by synthesising a set of linguistic control rules obtained from experienced human operations.	combination by synthesising a set of linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	level combination by synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	ndani's Fuzzy Inference Method Was Designed To mpt What?	linguistic control rules obtained from experienced human operations. Fuzzification of the input variables	synthesising a set of linguistic control rules obtained from experienced human operations.	linguistic control rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	synthesising a set of linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	mpt What?	obtained from experienced human operations. Fuzzification of the input variables	obtained from experienced human operations.	rules obtained from experienced human operations Application of the fuzzy operator (AND or OR) in the	linguistic control rules obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	mpt What?	experienced human operations. Fuzzification of the input variables	experienced human operations.	experienced human operations Application of the fuzzy operator (AND or OR) in the	obtained from experienced human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera: 3 What What What	mpt What?	Fuzzification of the input variables	operations.	Application of the fuzzy operator (AND or OR) in the	human operations Aggregation of the consequents across the	option3
3 What I What 3 Opera 3 What		Fuzzification of the input variables	·	Application of the fuzzy operator (AND or OR) in the	Aggregation of the consequents across the	Options
What 3 Opera 3 What	t Is The First Step Of Fuzzy Logic Toolbox?	input variables	Defuzzification	fuzzy operator (AND or OR) in the	consequents across the	
What 3 Opera 3 What	t Is The First Step Of Fuzzy Logic Toolbox?	input variables	Defuzzification	l '	· .	
What 3 Opera 3 What	t Is The First Step Of Fuzzy Logic Toolbox?		Defuzzification	antecedent	rules	
3 Operation 3 What		The input is a single				option1
3 Operation 3 What		The input is a single	The factor of the constitute			
3 Operation 3 What		truth value and the	The input is a value greater than one and	The input and	The input has two or more	
3 Operation 3 What	at Is The Input And Output Of Step 2 - Apply Fuzzy	output has two or	the output is a value	l '	values and the output has	
What		more values	less than the input	same values	· '	option4
What		Probor (a,b) = a-b +		Probor (a,b) = a+b -		
1	at Is The Equation For Probabilistic?	ab	Probor (a,b) = ab + ab	ab	Probor (a,b) = a/b x ab	option3
I		Input is a fuzzy set	Input is a whole value			
	at Is The Input And Output Of Step 3 - Apply Implication	but the output is a	but the output can be	Input and output	Input is a smaller value	
		whole value	a fuzzy set	have the same value	'	option2
		To gather all the		To gather all the		
		different fuzzy set outputs and combine	To gather all the possible inputs and	different fuzzy set	To subtract all the output	
		them into a single	use the average to	outputs and average them out to get a	To subtract all the output fuzzy set values from the	
3 What	at Is The Purpose Of Aggregation?	fuzzy set outputs	gain an output	single value	'	option1
	1 55 5	, ,			'	
I .	eralizations of ordinary fuzzy sets which involve fuzzy					
I .	defined within a universal set whose elements are					
3 ordina	nary fuzzy sets constitute afuzzy set	Level 1	Level 2	Level 3	Level 4	option2
3 Exam	mple of an idempotent t-norm is	Algebraic Product	Bounded Difference	Drastic intersection	Standard intersection	option4
- J LAGITI	o. an idempotent t norm is	bestale i Toudet	Takagi-Sugeno	274300 IIICI3ECUOII	Standard Intersection	Sparin
			approach does not			
		Mamdani approach	require any			
		needs defuzzification	fuzzification module	Takagi-Sugeno		
	difference habitan Manuala 1 and 1 and 1 and 1	module whereas	whereas Mamdani	approach is more		
1	difference between Mamdani approach and Takagi-	Takagi-Sugeno approach does not	approach needs	interpretable but less	All of the mentioned	ontion1
		approach does not	necus	accurate Takagi-Sugeno	All of the mentioned	option1
	eno approach to FLC design is that	Mamdani approach	Searching a rule in	approach consider a		
		i waniaanii appi oacii	_	large number of	Computation of each rule	
- 1		considers a less	Mamdani approach is		1	
3 more e	eno approach to FLC design is that ugi-Sugeno approach to FLC design is computationally		simple and hence less	rules in fuzzy rule	in Takagi-Sugeno approach	
3 "The t	eno approach to FLC design is that	considers a less	1	_	''	option4
	eno approach to FLC design is that ugi-Sugeno approach to FLC design is computationally	considers a less number of rules in fuzzy rule base	simple and hence less	rules in fuzzy rule base	is more time consuming	option4 option1

			T			I
	An expert system differs from a database program in that	contains declarative	contains procedural	features the retrieval	expects users to draw	
- 1	only an expert system:	knowledge	knowledge		their own conclusions	option2
1	is the process of formulating the mapping from a	owicage	- Innownedge	or stored information	then own conductions	Ориона
3 8	given input to an output using Fuzzy logic.	FIZ	FIS	FOZ	None of these	option2
	MILL CH. C.H	+ * · · · · · · · ·	+	6	All City I	
_	Which of the following is a type of Membership function? Which of the following is not a type of Membership	Triangular	Trapezoidal	Sigmoid	All of the above	option ²
- 1	function?	S-shape	Bell shape	Truncated Gaussian	None of these	option4
_	What is the independent variable of fuzzy output?	Maturity	Membership	Generic Element	None of these	option1
_	Gausian MF is specified by parameters	1	2	3	4	option2
3 (Cauchy MF is	Gausian MF	Triangular MF	Generalised Bell MF	Trapezoidal MF	option3
		{(1, 0.2), (2, 0.1),		{(1, 0.5),(2,		
3 ([3, 0.5)}, find AUB	(3, 0.4)}		0.9),(3, 0.6)}	{(x1,0.1),(x2,0.03),(x3,0.20)}	option2
٦,		{(1, 0.2), (2, 0.1),	1	{(1, 0.5),(2,	((4 0.4) (2 0.02) (2 0.20))	
3 ($[3, 0.5]$, find $A \cap B$ = {(1, 0.5), (2, 0.1), (3, 0.4)} and = {(1, 0.2), (2, 0.3),	(3, 0.4)} {(1, 0.2), (2, 0.1),	(3, 0.5)} {(1, 0.5), (2, 0.3),	0.9),(3, 0.6)} {(1, 0.5),(2,	{(x1,0.1),(x2,0.03),(x3,0.20)}	option
3 (3, 0.5)}, find A complement	(3, 0.4)}	(3, 0.5)}	0.9),(3, 0.6)}	[(x1,0.1),(x2,0.03),(x3,0.20)]	ontion ²
1	$= \{(1, 0.5), (2, 0.1), (3, 0.4)\}$ and $= \{(1, 0.2), (2, 0.3), (2, 0.3), (3, 0.4)\}$	{(1, 0.2), (2, 0.1),	{(1, 0.5), (2, 0.3),	{(1, 0.5),(2,	((//2/012/)(//2/0100/)(//0/0120/)	Ортон
3 ((3, 0.5)}, find A.B	(3, 0.4)}		0.9),(3, 0.6)}	{(x1,0.1),(x2,0.03),(x3,0.20)}	option4
	Methods of fuzzy approximate reasoning are	Syllogistic	Categorical	Dispositional	All of the mentioned	option4
3 ١	Which fuzzy approximate reasoning uses "usually"	Syllogistic	Categorical	Dispositional	Qualitative	option3
- 1	Like relational databases there does exists fuzzy relational					
3 0	databases	True	False			option1
	n this mode of approximate reasoning, the antecedents,					
- 1	containing no fuzzy quantifiers and fuzzy probabilities	Syllogistic	Categorical	Dispositional	Qualitative	option2
- 1	In this mode of approximate reasoning, the antecedents and					
3 0	consequents have fuzzy linguistic variables	Syllogistic	Categorical	Dispositional	Qualitative	option ²
	In this mode of approximation reasoning, antecedents with					
- 1	fuzzy quantifiers are related to inference rules	Syllogistic	Categorical	Dispositional	Qualitative	option1
╣	uzzy quantiners are related to interence rules	Conjuctive system of	DISjuctive system of	Dispositional	Quantative	Ориона
3 1	Method of aggregation of fuzzy rule	rules	rules	Both	None of these	option3
-	Fuzzy Expert system does NOT consist of	Knowledge base	User Interface	Inference Engine	None of these	option4
ľ	f P:Mary is efficient, Q: Ram is efficient, T(P)=0.8, T(Q)=0.6					
_	find value of Mary is not efficient	0.2	0.4	0.6	0.8	option1
- 1	f P:Mary is efficient, Q: Ram is efficient, T(P)=0.8, T(Q)=0.6					
_	find value of Mary is efficient and so is Ram. f P:Mary is efficient, Q: Ram is efficient, T(P)=0.8, T(Q)=0.6	0.2	0.4	0.6	0.8	option3
- 1	, , , , , , , , , , , , , , , , , , , ,	0.2	0.4	0.6	0.8	option4
Ť	,					
l'	f P:Mary is efficient, Q: Ram is efficient, T(P)=0.8, T(Q)=0.6					
3 f	find value of If Mary is efficient then so is Ram.					
		0.2	0.4	0.6	0.8	option3
- 1						
_	Fuzzy Implication is also known as	Fuzzy logic	Fuzzy IF-THEN rule	Fuzzy expert system	None of these	option2
_	Fuzzy Implication is also known as Ways to compute fuzzy rule A-> B is					option2
_		Fuzzy logic A coupled with B	Fuzzy IF-THEN rule	Fuzzy expert system Both	None of these	option2
_		Fuzzy logic A coupled with B Zadeh's Max Product	Fuzzy IF-THEN rule A entails B	Fuzzy expert system Both Zadeh's Max Product	None of these None of these	option2
3 \		Fuzzy logic A coupled with B	Fuzzy IF-THEN rule	Fuzzy expert system Both	None of these	option2
3 \	Ways to compute fuzzy rule A-> B is	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then	None of these None of these Zadeh's Max Min rule for <i>If</i>	option2
3 N	Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C	None of these None of these Zadeh's Max Min rule for If x is A then y is B	option2
3 \ 3 F 3 S	Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO Which method is good for embedding linear controller	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C	None of these None of these Zadeh's Max Min rule for If x is A then y is B	option2 option3 option4
3 N 3 F 3 S	Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO Which method is good for embedding linear controller Suppose, a fuzzy set Young is defined as follows Young =	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C Mamdani	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B Sugeno	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C Takagi	None of these None of these Zadeh's Max Min rule for If x is A then y is B None of these	option2 option3 option4
3 \ 3 \ 3 \ 3 \ \ ((Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO Which method is good for embedding linear controller Suppose, a fuzzy set Young is defined as follows Young = (10, 0.5), (20, 0.8), (30, 0.8), (40, 0.5), (50, 0.3) Then the crisp	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C Mamdani Mamdani	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B Sugeno Sugeno	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C Takagi Takagi	None of these None of these Zadeh's Max Min rule for If x is A then y is B None of these None of these	option2 option2 option2
3 \ 3 \ 3 \ 3 \ 3 \ ((3 \)	Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO Which method is good for embedding linear controller Suppose, a fuzzy set Young is defined as follows Young = (10, 0.5), (20, 0.8), (30, 0.8), (40, 0.5), (50, 0.3) Then the crisp value of Young using LoM method is	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C Mamdani	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B Sugeno	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C Takagi	None of these None of these Zadeh's Max Min rule for If x is A then y is B None of these	option2
3 F 3 S 3 V	Ways to compute fuzzy rule A-> B is R=(AXB)U(ÃXY) is System is used foe both MISO and MIMO Which method is good for embedding linear controller Suppose, a fuzzy set Young is defined as follows Young = (10, 0.5), (20, 0.8), (30, 0.8), (40, 0.5), (50, 0.3) Then the crisp value of Young using LoM method is Suppose, a fuzzy set Young is defined as follows Young =	Fuzzy logic A coupled with B Zadeh's Max Product rule for If x is A then y is B else y is C Mamdani Mamdani	Fuzzy IF-THEN rule A entails B Zadeh's Max Min rule for If x is A then y is B Sugeno Sugeno	Fuzzy expert system Both Zadeh's Max Product rule for If x is A then y is B else y is C Takagi Takagi	None of these None of these Zadeh's Max Min rule for If x is A then y is B None of these None of these	option2 option2 option2
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			continuous	combinatorial		1
4	Problem domain of Evolutionary Algorithm is	discrete optimization	optimization	optimization	None of these	option2
	Evolutionary Algorithm solves problem	NP-complete	NP-hard	P	None of these	option2
	, 0	initialization-> selection->mutation- >crossover-	initialization-> selection->crossover-	initialization-> selection->crossover- >mutation-		
4	Sequence of steps in EA	>termination	>termination	>termination	None of these	option3
4	Which of the following MOEA algorithm is based on the concept of "elitism"?	MOGA	NPGA	NSGA	NSGA-II	option4
	Which of the following solution is non-Pareto based a					
4	posteriori technique?	SOEA	MOGA	VEGA	Lexicographic ordering	option3
4	A three-objective optimizations are solved using a MOEA algorithm and a few objective vectors are listed below. (Assume all objectives are to be minimized) [3,5,10], [5,3,10], [3,10,5], [10,5,3], [10,3,5]. Which of the following is true?	All are non- dominating solutions	Solution [3,5,10] dominates all other solution	Solution [10,5,3] dominated by the solution [3,5,10]	Solution [3,10,5] dominates solution [5,3,10]	option1
4	MOOP follows "ranking" followed by "fitness averaging"	MOGA	NPGA	NSGA	NSGA-II	option1
	Which of the following MOEA techniques follows					T .
4	Tournament selection strategy?	MOGA	NPGA	NSGA	NSGA-II	option2
4	Niche counts of two solutions x1 and x2 are 10, 20. This means that	x1 is surrounded by more neighbors than that of X2	x1 is surrounded by less neighbors than that of X2	x1 would be less desirable to provide population diversity .	x1 would be more desirable to provide population diversity .	option2
4	Which of the following statement is not correct?	A set of solutions is called trade-off solutions, which lie on the Pareto optimal front.	A solution is called a trade-off solution, if it is not dominated by any other solution in the solution space	A front is called Pareto-optimal front on which all optimal solutions lie.	A front containing a non- dominated set of solutions obtained over an exhaustive search space is called pareto optimal front	option3
4	In the following, only one statement is correct. Select the correct statement.	Stochastic selection with remainder supports low selection pressure.	Crowding tournament selection scheme supports low population diversity.	There is no selection scheme in NPGA.	In MOGA, Rank based selection can be applied to select parent chromosome for mating pool creation	option4
4	Select the wrong statement	MOGA uses the concept of ranking whereas NSGA uses the concept of niching to assign fitness values to parent chromosomes.	MOGA assigns different fitness values to all solution with the same rank whereas NSGA assigns the same fitness value to all solutions belonging to the same front.	MOGA assigns the same fitness values to all solution with the same rank whereas NSGA assigns different fitness value to all solutions belonging to the same front.	NSGA yields more accurate Pareto front than MOGA.	option3
		Stochastic remainder	Crowding	Roulette wheel		
4	To create mating pool, NSGA follows	selection	Tournament selection	selection	Canonical Selection	option1
	A similarity between NPGA and NSGA is that	Both are based on the concept of ranking.	Both assign fitness values to the parent chromosomes prior to their selection for mating pool.	Both calculates niched count to maintain population diversity.	Both are computationally very expensive.	option3
4	Ifc1 and c2 are two offspring chromosomes, then according to NPGA, c1 will be preferable to be selected for mating pool if	if c1's niche count is higher than that of c2.	if c1's niche count is equal to that of c2.	if c1's niche count is less than that of c2.	None of these	option3
4	A priori high level information that is required in "Lexicographic ordering" is	the scalar weights of each objective function.	the descending ordering of the rank of the importance of objective functions.	independent of objective function.	None of these	option2

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How many genes will be in the alphabet of the algorithm? n*(n-1)/2	opti
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x = abcdefgh with a fixed length of eight genes. Each gene can be any digit between 0 and 9. Let the fitness of individual x be calculated as: f(x) = (a + b) - (c + d) + (e + f) - (g + h), and let the initial population consist of four individuals with the following chromosomes: x1 = 65 4 1 3 5 3 2 x2 = 8 7 1 2 6 6 0 1 x3 = 2 3 9 2 1 2 8 5	opti
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chromosomes: x1 = 65 4 1 3 5 3 2 x2 = 8 7 1 2 6 6 0 1 x3 = 2 3 9 2 1 2 8 5	
x1 = 65 4 1 3 5 3 2 x2 = 8 7 1 2 6 6 0 1 x3 = 2 3 9 2 1 2 8 5	
x2 = 87126601 x3 = 23921285	
x3 = 2 3 9 2 1 2 8 5	
arrange chromosomes in order with the fittest first and the	
4 least fit last. x1,x2,x3,x4 x2,x1,x3,x4 x1,x3,x4,x2 x2,x3,x1,x4	
one–point crossover at the middle point on	onti
x2 = 87126601	opti
4 x1 = 65 4 1 3 5 3 2 will generate	opti
A budget ariline company operates 3 plains and employs 5	optio
cabin crews. Only one crew can operate on any plain on a	
single day, and each crew cannot work for more than two	
days in a row. The company uses all planes every day. A	
Genetic Algorithm is used to work out the best combination	
4 of crews on any particular day. What is alphabet size in it? 3 5 8 15	

				T	
In individuals are represented as a Finite State 4 Machine	EP	ES	GA	PSO	option1
4 In individuals are represented as a Real valued vector	EP	ES	GA	PSO	option2
4 In individuals are represented as a binary string	EP	ES	GA	PSO	option3
4 Parent selection in is stochastic using tournament	EP	ES	GA	PSO	option1
4 Parent selection in is deterministic	EP	ES	GA	PSO	option2
4 Parent selection in is either stochastic or deterministic	EP	ES	GA	PSO	option3
4does not use recombination to produce offspring.	EP	ES	GA	PSO	option1
4 uses crossover to produce offspring.	EP	ES	GA	PSO	option2
4 uses various recombination operators.	EP	ES	GA	PSO	option3
Is it advisable to apply genetic algorithm for all kinds of					
4 optimization problem Ineach individual survives for exactly one	Yes	No			option2
4 generation	Generation Model	Steady state Model	Population Model	Tournament Model	option1
4 Category of EP based on scaling function	Non adaptive	Non Dynamic	Non self-adaptive	Adaptive	option1
Evolutionary algorithm differes from genetic algorithm as it	Non adaptive	TVOIT Dynamic	ivon sen adaptive	Adaptive	Ориона
4 does not have	Crossover	mutation	Selection	Reproduction	option1
	Evolutionary				
4 Evolutionary algorithms classified as	Programming	Evolutionary Strategies	Genetic Algorithms	All of the mentioned	option4
In the choice phase of problem solving, normative models	_				
5 involve selecting an optimal or best outcome Analytical techniques for problem solving are best for	True	False			option1
5 unstructured rather than structured problems.	True	False			option2
Heuristic approaches are typically used to solve more					
5 complex problems.	True	False			option1
Genetic algorithms are heuristic methods that do not	T	Falsa			
5 guarantee an optimal solution to a problem A "what-if" model is most typically used for the most	True	False			option1
5 structured problems	True	False			option2
The use of simulation models is desirable because they can					
usually be solved in one pass, without incurring the time and					
5 cost of iterations	True	False			option2
An advantage of simulation is that it allows model builders					
to solve problems with minimal interaction with users or	_				
5 managers Time compression in a simulation allows managers to test	True	False			option2
5 certain strategies with less risk.	True	False			option1
Simulation solutions cannot easily be transferred from one	Truc	Tuise			Орионт
5 problem domain to another	True	False			option1
Determining the duration of the simulation occurs before	_				
5 the model is validated and teste	True	False			option2
Which of the following(s) is/are the pre-requisite(s) when					
Genetic Algorithms are applied to solve					
problems?					
(i) Encoding of solutions.					
(ii) Well-understood search space.					
(iii) Method of evaluating the suitability of the solutions. 5 (iv) Contain only one optimal solution.	i and ii only	i and iii only	ii and iii only	All of the mentioned	option2
Spirit contain only one optimal solution.	, and it offity	r and in only	Reproduction to	, iii or the mentioned	OPTIONZ
	Initial population	Selection of sub-	produce next		
5 Which GA operation is computationally most expensive?	creation	l	generation	Convergence testing	option3
		Ţ.	If an optimization	_	
			problem has more		
		It is guaranteed to	than one solution,	It is an iterative process	
T Militar afails fallening of the Control of the Co	It is a probabilistic	give global optimum	then it will return all	suitable for parallel	
5 Which of the following is not true for Genetic algorithms? Which one of the following is not necessarily be considered.	search algorithm	solutions the obtainable	the solutions	programming	option2
Which one of the following is not necessarily be considered 5 as GA parameters?	the population size.	accuracy	the mutation probability	the average fitness score	option4
	paracion size.	,			
Which of the following optimization problem(s) can be		Travelling salesman	Job shop scheduling	Optimal binary search tree	
5 better solved with Order GA?	0-1 Knapsack problem		problem	construction problem	option2
5 Optimal binary search tree construction problem	13572468	ABDECFHG	10011001	14.6 -23.4 177.23	option3
	Fitness values are	Fitness values are non-uniformly	Needs low selection	Needs high population	
5 Roulette wheel selection scheme is preferable when	uniformly distributed	distributed	pressure	diversity	option1
Spromette wheel selection scheme is preferable when	Januarining distributed	מוטנו וטענכע	hi cooni c	urversity	Ιοριισιιτ

	What GA encoding scheme suffers from Hamming cliff					
5	problem?	Binary coded GA	Real coded GA	Order GA	Tree coded GA	option1
	Which selection strategy is susceptible to a high selection	Roulette-wheel		Tournament		
5	pressure and low population diversity?	selection.	Rank based selection.	selection.	All of the mentioned	option1
	Which of the following is not a mutation operation in real			Polynomial	All are mutation operation	
5	coded GA?	Flipping	Random mutation.	mutation.	in real coded GA	option1
	Two parent chromosomes in Order GA encoding scheme is					
	given as follows:					
	*					
	12345678910					
	10987654321					
	A k-th point is selected at 4th location according to single					
	point crossover technique. Which					
5	of the following off-spring is not possible?	12341098765	78910654321	10987123456	5 6 7 8 9 10 1 2 3 4	option4
		To check whether all				
		individual satisfies the constraints given in	To decide the	To select the	To identify the individual	
5	The purpose of the fitness evaluation operation is	the problem	termination point.	bestindividuals	To identify the individual with worst costfunction	option3
	The purpose of the fitness evaluation operation is	the problem	termination point.	Destillarviduais	With Worst Costiunction	ориона
	If crossover between chromosomes in search space does					
	not produce significantly different offspring, what does it					
	imply? (if offspring consist of one half of each parent)					
	(i) The crossover operation is not successful.					
	(ii) Solution is about to be reached.					
	(iii) Diversity is so poor that the parents involved in the					
	crossover operation are similar.					
5	(iv) The search space of the problem is not ideal for GAs to operate	ii, iii & iv only	ii, iii only	i, iii & iv only	All of the mentioned	option2
	operate	ii, iii Q iV oiiiy	ii, iii oiiiy	i, iii & iv oiiiy	7 th of the mentioned	Орстопа
		The % area to be	Two or more			
		occupied by an	individuals with the	Individuals are	The proportionate based	
		individual , is given by	same fitness values	arranged in a	selection scheme is	
	In Rank-based selection scheme, which of the following is	average of sumation	should have the same	descending order of	followed based on the	
5	not correct	of elements	rank.	their fitness values.	assigned rank.	option3
		Low population	High population	NA		
		diversity and moderate selection	diversity and Moderate selection	Moderate population diversity and high	High population diversity	
5	Tournament Selection has	pressure	pressure	selection pressure	and low selection pressure	option3
	Tournament delegation has	The search focuses	p. 6354. 6	pressure	and town selection pressure	ориона
		only on good			Leads to pre-mature	
		individuals (in terms			convergence of the	
		of fitness) at the	It loses the	Lower rate of	solution to a sub-optimal	
5	If selection pressure is HIGH, which one is FALSE	moment	population diversity.	convergence.	solution.	option3
		In the event of		The qualities of		
		restricted accessto information, GAs win		The qualities of solutions offered by		
		out in that they		GAs for any	GAs could be applied to	
		require much	Under any	problems are always	any problem, whereas	
		fewer information to	circumstances, GAs	better than	certain algorithms are	
		operate than other	always outperform	those provided by	applicable to	
_	Which of the following comparison is true?	search.	other algorithms.	other search	limited domains.	option1
5	Which of the following is a fitness scaling approach?	Linear scaling	Sigma scaling	Power law scaling	All of the mentioned	option4
		Definition of chicative	Implementation of	Implementation of		
5	Important aspect of GA	Definition of objective function	genetic representation	genetic operators	All of the mentioned	option4
,	portaine appear or ort	. and on	periode representation	Dericale operators	Number of coded design	570012
5	Term in GA for chromosome	coded design vector	coded design variable	every bit	variable	option1
					Number of coded design	
5	Term in GA for substring	coded design vector	coded design variable	every bit	variable	option2
					Number of coded design	
5	Term in GA for gene	coded design vector	coded design variable	every bit	variable	option:
_	Torm in CA for nonulation	and add design	and add design to the	ayanı bi+	Number of coded design	0.54
5	Term in GA for population	coded design vector	coded design variable	every bit population of design	variable	option
Ť		1		vector after 1	Number of coded design	
		1				i
	Term in GA for generation	coded design vector	coded design variable		_	option?
	Term in GA for generation	coded design vector evolutionary	coded design variable	iteration	variable	option
5	Term in GA for generation Which of these are adaptive heuristic search algorithms		coded design variable		_	option3

selection	termination	recombination	initialization	optio
deterministic	probabilistic	both	None of these	option
Yes	No			option
103	110			Optio
Provide optimization	Provide optimization	They break on slight		
•	· '	, ,	None of these	optio
over small space area	over large space area	change in input	TVOIC OF THESE	Орио
10111001	01010110	10111000	01010101	ontio
10111001	01010110	10111000	01010101	optio
			N Cul	
Crossover point	Crossover rate	Cross-over population	None of these	optio
123456789		 		optio
Generational	Centralized	Distributed	None of these	optio
Generational	Steady state	Distributed	None of these	optio
Generation limit	No change in fitness	Elapsed time	All of the mentioned	optio
	Divide n conquer			
Pittsburg Approach	approach	Michi Approach	None of these	optio
Parallel Genetic	Proportional Genetic	Perceptron based		
Algorithm	Algorithm	Genetic Algorithm	None of these	optio
		Termination		optio
1	2	3	4	optic
		-		- Spene
	Darticle Course	Ant Colonia		
C			All -f +h+:	
Genetic Algorithms	Optimization	Optimization	All of the mentioned	optic
ı				
of decentralized, self-				
organized systems,				
natural or artificial	fuzzy logic system	crisp logic concept	None of these	optic
Gerardo Beni and	Gerardo Beni and	George M and Jing		
Jing Wang	James Gosling	Wang	None of these	optic
	Ŭ	ŭ		
swarm robotics	swarm behaviour	robotics	None of these	optic
		1000000	Trone or these	Optio
	l '	Doth	None of these	antia
<u>'</u>	'	1 1		optio
				l
	· ·	<u> </u>	Optimization	optic
Analytical Colony	Ant Colony	Ant Coding		
Optimization	Optimization	optimization	None of these	optic
pbest	gbest	Both pbest, gbest	None of these	optic
personal best	personal based	proper based	proper best	optic
global best	general best	global based	general based	optic
progress	loops	iteration	None of these	optio
	<u> </u>	 		optic
neness value	velocity	poest	7 in or the mentioned	Optio
1	2		Many	ontic
<u>T</u>	2	3	Ivially	optio
velocity	gbest	Both velocity ,gbest	None of these	optio
		L	l	
Vi(t+1)=w*Vi -	Vi(t+1)=w*Vi	Vi(t+1)=w*Vi	Vi(t+1)=w*Vi	
a1*rand*/nhast vi);	+c1*rand*(pbest-xi)-	+c1*rand*(gbest-xi)+	+c1*rand*(pbest-xi)+	
c1*rand*(pbest-xi)+			c2*rand*(gbest-xi)	lantic
c2*rand*(gbest-xi)	c2*rand*(gbest-xi)	c2*rand*(pbest-xi)		Optic
	c2*rand*(gbest-xi)	c2*rand*(pbest-xi)		Орис
	c2*rand*(gbest-xi)	c2*rand*(pbest-xi) Vi(t+1)=w*Vi	Vi(t+1)=w*Vi	Орис
	c2*rand*(gbest-xi)	Vi(t+1)=w*Vi	` '	Орис
c2*rand*(gbest-xi)		Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+	+c1*rand*(pbest-xi)-	
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1)	c2*rand*(gbest-xi) Xi(t+1)=Xit-Vi(t+1)	Vi(t+1)=w*Vi	` '	
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration	Xi(t+1)=Xit-Vi(t+1)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1)		Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+	+c1*rand*(pbest-xi)-	optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient	Xi(t+1)=Xit-Vi(t+1)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optio
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optio
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient	Xi(t+1)=Xit-Vi(t+1)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optio
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optio
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optio
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)	Xi(t+1)=Xit-Vi(t+1) Cognitive Component	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi)	optic optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component c1*rand*(pbest-xi)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi) Inertia term None of these	optic optic optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)+ c2*rand*(gbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component c1*rand*(pbest-xi)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi) Inertia term None of these	optic optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)+ c2*rand*(gbest-xi)+ c2*rand*(gbest-xi)+ c1*rand*(pbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component c1*rand*(pbest-xi) c1*rand*(pbest-xi)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component c2*rand*(gbest-xi)	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi) Inertia term None of these	optic optic
c2*rand*(gbest-xi) Xi(t+1)=Xit+Vi(t+1) Acceleration Coeficient c1*rand*(pbest-xi)+ c2*rand*(gbest-xi)+ c2*rand*(gbest-xi)+	Xi(t+1)=Xit-Vi(t+1) Cognitive Component c1*rand*(pbest-xi)	Vi(t+1)=w*Vi +c1*rand*(pbest-xi)+ c2*rand*(gbest-xi) Social Component	+c1*rand*(pbest-xi)- c2*rand*(gbest-xi) Inertia term None of these	optic optic
	deterministic Yes Provide optimization over small space area 1 0 1 1 1 0 0 1 Crossover point 1 2 3 4 5 6 7 8 9 Generational Generational Generation limit Pittsburg Approach Parallel Genetic Algorithm Analysis 1 Genetic Algorithms collective behavior of decentralized, selforganized systems, natural or artificial Gerardo Beni and Jing Wang swarm robotics Particle swarm optimization Particle Swarm Optimization Particle Swarm Optimization Analytical Colony Optimization phest personal best global best progress fitness value 1 velocity Vi(t+1)=w*Vi -	deterministic Yes No Provide optimization over small space area 1 0 1 1 1 0 0 1 Crossover point Crossover rate 1 2 3 4 5 6 7 8 9 Generational Generational Generation limit No change in fitness Divide n conquer approach Parallel Genetic Algorithm Analysis Computation 1 2 Particle Swarm Optimization Gerardo Beni and Jing Wang Swarm robotics Particle Swarm Optimization Optimization Analytical Colony Optimization Optimization Optimization Dest personal best personal best global best progress loops fitness value Vi(t+1)=w*Vi Vi(t+1)=w*Vi	deterministic probabilistic both Yes No Provide optimization over large space area over large space area over large space area over large space area over large in input 10111001 0101110 10111000 Crossover point Crossover rate Cross-over population 123456789 153426789 154326789 Generational Centralized Distributed Distributed Divide neonquer approach Michi Approach Parallel Genetic Algorithm Analysis Computation Termination 1 2 2 3 4 5 6 7 8 9 15 4 3 2 6 7 8 9 15 4 3 2 6 7 8 9 10 10 10 10 10 10 10 10 10 10 10 10 10	deterministic Probabilistic both None of these No No Provide optimization over small space area or large s

	<u> </u>	1	T	1	ı
	Once PSO traps in	Once PSO traps in	Once PSO traps in		
	global optimum, it is	local optimum, it is	local optimum, it is	Once PSO traps in global	
	dificult to jump out of	dificult to jump out of	dificult to jump out	optimum, it is dificult to	
6 Premature convergence of PSO is	global optimum	local optimum	of global optimum	jump out of local optimum	option
6 PSO topology	Star	Ring	Adaptive random	All of the mentioned	option
6 In binary PSO, solution in is binary string	Population	chromosome	swarm	velocity	option
6 Chemical substance release by ants is	pheromone	perchloric	Both	None of these	optio
6 ACO is typically used to solve based problem	searching	graph	database	None of these	optio
6 ACO inspired by	ant colony foraging	bird flocking	both	None of these	optio
6 PSO inspired by	ant colony foraging	bird flocking	both	None of these	optio
6 ACO is best at finding solution	discreate	continuous	both	None of these	optio
6 PSO is best at finding solution	discreate	continuous	both	None of these	optio
o F3O is best at illiding solution	Cost matrix,	Cost Matrix, Velocity	Velocity Matrix,	Velocity Matrix,	οριιο
Clar ACO there evict two metric	Pheromone Matrix	Matrix	Pheromone Matrix	Pheromone Matrix	antia
6 In ACO there exist two matrix	Prieromone Matrix	IVIALITIX	Prieromone Matrix	Prieromone Matrix	optio
6 Ants are that move along between nodes in a graph.	Participants	Population	Agents	None of these	option
6 ACO is optimization	Heuristic	Metaheuristic	Probabilistic	None of these	optio
6 Ants uses communication via pheromone trails	Direct	Stigmergic	eulogic	None of these	optio
			Condition occur at		
		Negative feedback	termination of	Allows ants to explore less	
6 Autocatalysis is	Positive feedback loop	"	algorithm	promising areas	optio
	Autocatalytic		Autocatalytic		<u> </u>
	negative feedback	Autocatalytic positive	forward feedback	Autocatalytic backward	
6 ACO algorithms also called as	algorithm	feedback algorithm	algorithm	feedback algorithm	optio
6 Ant System first introduced by	James Kennedy	Russell E.	Marco Dorigo	L. A. Zadeh	optio
6 Application of Ant systems	Scheduling Problem	Assignment Problem	Set Problem	All of the mentioned	optio
			Clobal phoromono		
6			Global pheromone	Too a siti a a mula	
avoids premature convergence in ACO	distributed computing	pneromone	update	Transition rule	optio
ACO performed poorly for TSP problems larger than	50	65	70		
6 cities	60	65	70	75	optio
In ACO search is extreamly important to obtain good					l
6 result	local	global	distributed	None of these	optio
6 Optimization in PSO is	Metaheuristic	Stochastic	Discreate	Continuous	optio
		reach target with	find local finest		
6 PSO	find shortest path	minimal duration	solution	find best among others	optio
6 Quadratic Assignment problem can be solved by	ES	GA	PSO	ACO	optio
6 Vehicle routing can be solved by	ES	GA	PSO	ACO	optio
6 Automatic programming uses	ES	GA	PSO	ACO	optio
6 Economic models uses	ES	GA	PSO	ACO	optio
6 Heating system planning problem uses	ES	GA	PSO	ACO	optic
	1/length_of_tour, if	1/length_of_tour, if	length_of_tour, if	length_of_tour, if path is	
6 Amount of pheromone deposited	path is not used	path is used	path is not used	used	optio
When all ants have completed a solution, the trails are			Txy=(1-p)*		
6 updated with	Txy=(1-p)* Txy-∑ΔTxy	Txy=(1-p)+ΣΔTxy	Τχγ+ΣΔΤχγ	Txy=(1-p)* Txy-Txy	optio