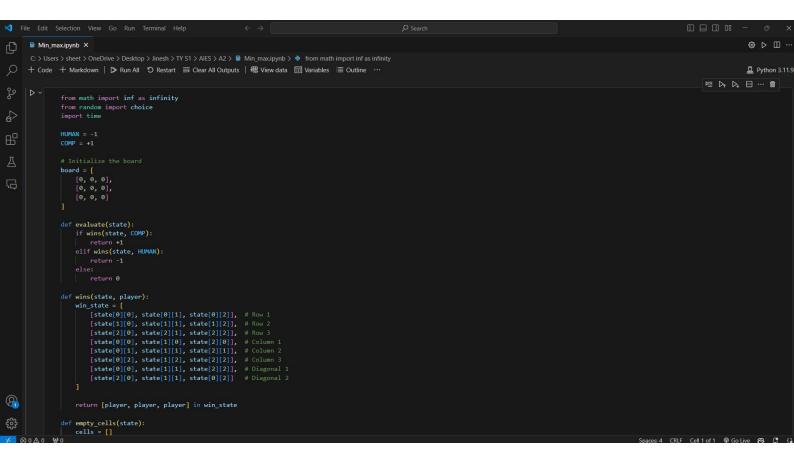
Jine	esh. N	Page No. :
oll N	o : ¿Topic:	Date.:
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arc	Assignment - 02 AIE	57
	71379	
*	Tittle:	. Land and
	Implementation of Minmax Alg	= for tir-tar
	toe game.	
		1
*	Aim :	•
	Solving tic tocke game using	minmax ago.
		P W SCAR STATE
*	Objective ! To study and implement minmax	a landard
	To study and implement mining	La contract de la contraction
		July Andrews
	. ,	- 10
(1)	Adversial search: It is a search strategy used in problems where multiple agents	decision making
	It is a search strategy	conpete
	It is commonly used in the t	field of game
	It is commonly	I has head from
	theory and AI.	and the state of
,	Tic Tac Toe solving steps : 6	ame represm
(2)	initial state, move generation	evaluation of
	game states, minmax algo.	alpha bela
	frunning.	in a second second second second
		For Educational Use Only

	Page No.:	
	Topic:	
	Topic .	-
3.	Data Structures +	
	Key features	
	Game tree: nodes, edge	
	Board represe : A 2D array	
	More list	
	Evaluate func*	
	and the second of the second o	
	Key Consol.	
	Key Concepts:	
->		
→	depth and pruning	
->	optimal strategy	
	utility Valuer.	
×15 1	The second section of the second seco	
*	Input:	
	Initial state	_
	as to blance and the property of the	
*	Output:	18
	Sola goal state with optimal path	-
	describe the described the second of	
+	Algorithm:	
	Minmax	
	The state of the s	
*	Plat form:	
	Platform: / linux	
		4
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	Page No.:
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	+ FAg's:
	1. Compare informed search and adversial
9.	
	seach.
	> Informed search:
	- Utilises domain specialists on finding the - Goal oriented : Focuses on finding the chartest costly path to a
	and tate
	- A CLARE LOCK.
	- txplorer path in a serimate cost to
	goal.
T ale	- single agent Problems.
A.E. My	
	Adversial search in the came
	- Represents all possible moves in
	trom current game.
	- Alpha beta pruning
	- Multi agent problems
	- Explore game trees.
9.2	Explain alpha-beta Pruning.
<i>→</i>	Alpha bet is an optimism technique for the
	mirmax algo used in adversial search
	scenarios.
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		No.:
	key Concepts:	41 +
	(1) Alpha	
l.	(1) seta man house	
	(1) Pruning	
		- 1
	Howit works:	
	Initial call	
	recarrive search	
	pruning search	
	Min Min	
	Max Mode	
	Møde	
	Lace Lace Lace Lace	
-/-	689	
	(2)	
-11 -10 3	and the same of the same that	
	The same Experience made	+
	was the state of t	
	Land to the second of the seco	
	A AIRLINE 2 TO LONG LOS A LONG LOS ALLES	
3/4	A LO CONTRACTOR AND A LOCAL AN	
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	The state of the s	
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| No fire to Section | New Go Ran Torman | New Co Ran Torman | Ne
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XI File Edit Selection View Go Run Terminal Help
                                                                                                                                                                                                                                                                                        ⊕ ⊳ □ ..
        ■ Min_max.ipynb ×
        + Code + Markdown | ▶ Run All り Restart 云 Clear All Outputs | 甥 View data 园 Variables 這 Outline …
                                                                                                                                                                                                                                                                                                                 Python 3.11.
                            str_line = '---+----
print('\n' + str_line)
                             for row in state:
    row_str = '|'.join(chars[cell] for cell in row)
    print(f' {row_str} ')
                       def ai_turn(c_choice, h_choice):
    depth = len(empty_cells(board))
                             if depth == 0 or game_over(board):
                             print(f'Computer turn [{c_choice}]')
render(board, c_choice, h_choice)
                             if depth == 9:
    x = choice([0, 1, 2])
    y = choice([0, 1, 2])
                             else:

move = minimax(board, depth, COMP)

x, y = move[0], move[1]
                             set_move(x, y, COMP)
time.sleep(1)
                       def human_turn(c_choice, h_choice):
    depth = len(empty_cells(board))
                              if depth == 0 or game_over(board):
                              moves = {
    1: [0, 0], 2: [0, 1], 3: [0, 2],
    4: [1, 0], 5: [1, 1], 6: [1, 2],
    7: [2, 0], 8: [2, 1], 9: [2, 2]
8
 £55
```

