



IT426: ARTIFICIAL INTELLIGENCE SYSTEMS
INFORMATION TECHNOLOGY DEPARTMENT

PROGRAMMING ASSIGNMENT

ASSIGNED: OCT 4TH 2020
DUE: NOV 8TH 2020 @ 12:00 PM

Best Meal Synthesis using GA Optimization

RULES:

1. You are allowed to work on this assignment in a group approved in the shared excel sheet.
2. You are NOT allowed to copy online published solutions.
3. Some questions are supposed to encourage you to read more on the topics taught in class.

You are going out on the weekend to have a meal in international cuisine restaurant. You went to your favorite international restaurant and found the menu in the table below. Your goal is to optimize your meal based on type of food, number of calories and the price. Each meal consists of items from each of the following category (starters, main-course, dessert, hot beverage, cold beverage) given your dietary preference (see note below), budget and calories intake.

You decide to implement a GA algorithm to find the best combination of items for each course. The code should work as follows; the user enters an item type (options include regular, vegetarian and vegan), calories = [CAL 0 – CAL ∞], and budget= [SAR 0.0 – SAR ∞]. Your GA will then run to find the best items from each category that maximize your fitness function and output the items.

Your fitness function includes three parameters; item type, calories, and budget. The item type and budget are equally important, and the calories is half important. Your fitness function can be designed as *weighted sum* and importance of factors can be encoded as weights w_i in the fitness function, where $\sum w_i = 1$. Please note that your fitness function value must be in the range [0,1].

Dietary preference notes:

1. Vegan individuals can choose only vegan items.
2. Vegetarian individuals can choose vegan or vegetarian items.
3. People with no dietary preferences can choose vegan, vegetarian and/or regular items.

Restaurant Menu:

CATEGORY	ITEM	CALORIES	TYPE	PRICE
STARTER	Dynamite shrimp	770	Vegetarian	74.00 SR
	Halloumi fries	530	Vegetarian	48.00 SR
	Baked eggplant	270	Vegan	36.00 SR
	Fish cake	480	Vegetarian	60.00 SR
	Lentil soup	290	Vegan	26.00 SR
	Chicken soup	280	Regular	32.00 SR
	Lamb samosa	650	Regular	52.00 SR
	Kale salad	280	Vegan	38.00 SR
	Quinoa salad	320	Vegan	36.00 SR
	Burrata salad	360	Vegetarian	65.00 SR
	Shredded beef salad	520	Regular	52.00 SR
	Hummus and bread	620	Vegan	28.00 SR
MAIN COURSE	Margarita Pizza	740	Vegetarian	35.00 SR
	Veggie Pizzas	850	Vegan	45.00 SR
	Pepperoni Pizza	1100	Regular	40.00 SR
	Fish & Chips	2000	Regular	100.00 SR
	Fried Shrimp	2050	Regular	103.00 SR
	Grilled Salmon	1130	Regular	124.00 SR
	Classic Burger	1360	Regular	84.00 SR
	Mushroom Burger	1490	Regular	76.00 SR
	Veggie Burger	1140	Vegan,	35.00 SR
	Grilled Chicken Sandwich	1200	Regular	84.00 SR
	Chicken Parmesan Sandwich	2050	Regular	87.00 SR
	Tomato Basil Pasta	1460	Vegan	59.00 SR
	Fettuccini Alfredo	1990	Vegetarian	65.00 SR
	Spaghetti with Meatballs	1650	Regular	85.00 SR
	Four Cheese Pasta	1240	Vegetarian	65.00 SR
	Grilled Rib Eye Steak	1300	Regular	130.00 SR
	Vegetable Biryani	1240	Vegetarian	47.00 SR
	Biryani Rice	1350	Vegan	25.00 SR
DESSERT	Chocolate Fondant	587	Vegetarian	45.00 SR
	Date Pudding	322	Vegetarian	38.00 SR
	Coconut Crunch	896	Vegetarian	46.00 SR
	Mango & Passion Fruit Crèmeux	498	Vegetarian	34.00 SR
	Carrot Cake	290	Vegetarian	26.00 SR
	Salted Caramel & Pecan Pie	880	Vegetarian	32.00 SR
	Almond Delight	509	Vegan	42.00 SR
	Coconut and almond vanilla tart	320	Vegan	45.00 SR
	Cashew Apple pie	920	Vegan	40.00 SR
	Vegan coconut ice cream	520	Vegan	25.00 SR
	Assorted fruits platter	238	Vegan	22.00 SR
	Vegan chocolate cake	829	Vegan	28.00 SR
	Chocolate Brownie	580	Vegetarian	57.00 SR
	Millefeuille	296	Vegetarian	52.00 SR
	Classic Cheesecake	798	Vegetarian	64.00 SR
	Almond pistachio croissant	533	Vegan	52.00 SR
	Signature Apricot almond tart	334	Vegan	68.00 SR
	Almond Pistachio Pie	940	Vegan	56.00 SR
HOT DRINK	Americano	15	Vegan	18.00 SR
	Espresso	9	Vegan	13.0 SR
	Cappuccino	120	Vegetarian	19.0 SR
	Cafe Latte	190	Vegetarian	19.0 SR
	Macchiato	250	Vegetarian	20.0 SR
	Black tea	0	Vegan	15.00 SR

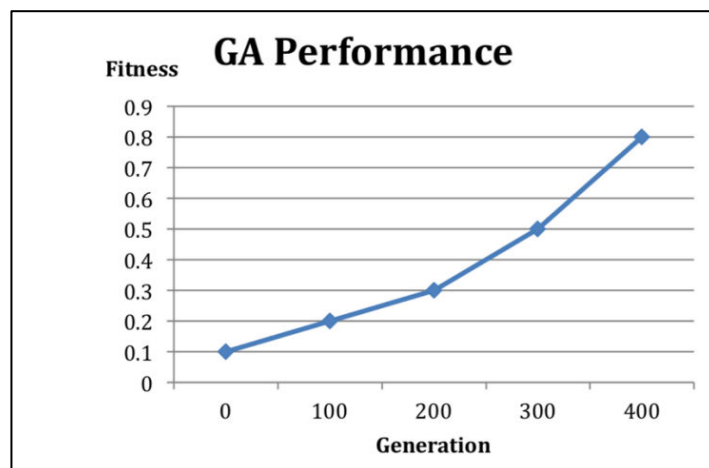
COLD DRINK	Green tea	0	Vegan	15.00 SR
	Hot Chocolate	350	Vegetarian	21.00 SR
	Pepsi	120	Vegan	10.0 SR
	Diet Pepsi	0	Vegan	10.0 SR
	7up	120	Vegan	10.0 SR
	Mirinda	90	Vegan	10.0 SR
	Orange Juice	130	Vegan	16.0 SR
	Strawberry milkshake	210	Vegetarian	20.0 SR
	Tropical Juice	340	Vegan	25.0 SR
	Ice Latte	160	Vegetarian	18.0 SR
	Large Water	0	Vegan	10.0 SR
	Small Water	0	Vegan	10.0 SR
	Raspberry Mojito	129.4	Vegan	28.0 SR
	Blueberry Mojito	136	Vegan	30.0 SR

Deliverables:

1. Python Code.
2. Report
3. Termination condition.
4. Result:

Graph of GA performance containing the generation vs fitness function as shown in the figure below for each combination of parameter setting in the table (Total of 3^3 combinations).
containing explanation of your choice of:

1. Solution representation.
2. Fitness function.
3. Genetic operators:
 - i. Crossover.
 - ii. Mutation.
 - iii. Selection by roulette wheel selection (See below).
 - iv. Replacement.
4. Analysis of your results



Experimental Settings:

1. Runs: Run your GA 20 times and report the average fitness.
2. Initialize your first generation randomly in the search range.

GA Parameter	Value	Number of combinations to try
Population size	variable (multiples of 10)	3
Selection method	Roulette wheel selection	-
Crossover type	Random one point	-
Crossover rate, c	Variable $0 < c < 1$, the higher the better	3
Mutation rate, m	Variable $0 < m < 1$ The lower the better	3
Termination Condition	1. Try different conditions and choose the best. For example, terminate when the error value is less than 10^{-8} , or after 20000 generations. (The error value is the difference between the objective function value and the optimal solution)	-

Helpful links:

GA Explanation:

<http://www.cs.ucc.ie/~dgb/courses/tai/notes/handout12.pdf>
https://www.doc.ic.ac.uk/~nd/surprise_96/journal/vol1/hmw/article1.html
<https://www.slideshare.net/pbpimpale/genetic-algorithms-200688>

Roulette wheel Selection:

https://en.wikipedia.org/wiki/Fitness_proportionate_selection
<http://www.edc.ncl.ac.uk/highlight/rhjanuary2007g02.php>
<https://www.youtube.com/watch?v=9JzFcGdpT8E>