FAQ

Module-1		
Question 1.	What exactly is Optimization in the context of Data Science?	
Answer	Optimization in the realm of data science refers to the process of tuning or adjusting a model or algorithm to make it as effective as possible, usually by minimising or maximising some measure of interest, known as the objective function. For instance, in a machine learning model, optimization might involve adjusting parameters to minimise prediction errors.	
Question 2.	Why are Objective Functions so crucial in the optimization process?	
Answer	Objective functions play a pivotal role because they define the goal of the optimization. In simpler terms, they provide a measure of how "good" or "bad" a particular solution is. By defining what we want to maximise (like accuracy) or minimise (like error), objective functions guide the optimization algorithm towards the best possible outcome. In the context of data science, various objective functions, like mean squared error or cross-entropy, quantify the difference between predicted and actual outcomes, thereby directing models to improve their	

	predictions.
Question 3.	Can you explain the significance of Decision Variables
	in optimization problems?
Answer	Certainly! Decision variables are the variables that we
	decide upon or control to achieve the best outcome in an
	optimization problem. They represent the potential
	solutions to the problem. For instance, if you're trying to
	find the best mix of investments for a portfolio, the
	percentage allocation to each investment option would be
	your decision variables. In data science, decision variables
	might represent model parameters, and the optimization
	process will adjust these variables to achieve the best
	model performance. Their significance lies in the fact that
	by adjusting these variables, we can directly influence the
	result or outcome of our optimization problem.
Question 4.	Why might one choose Python over Excel for
	optimization problems?
Answer	Both Excel and Python have their merits for tackling
	optimization problems. However, Python offers a few
	notable advantages:
	1. Scalability: Python can handle larger datasets and
	more complex problems, especially when using
	specialised libraries.
	2. Flexibility: With Python, you can integrate various

	
	data sources, use diverse algorithms, and tap into
	advanced optimization libraries like SciPy and
	CVXPY.
	3. Automation: Python allows for more automation
	and can be integrated into larger systems or
	workflows.
	4. Community Support: The vast community of
	Python developers often provides updates, new
	libraries, and solutions to emerging challenges in
	optimization. While Excel's Solver tool is
	user-friendly and suitable for simpler problems,
	Python is often the go-to for more intricate and
	large-scale optimization tasks in data science.
Question 5.	Can every optimization problem be solved to find a
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used to enhance the chances of finding a global optimum. However, in some scenarios, finding an exact global optimum might be computationally infeasible, and an approximate solution or local optimum might be satisfactory.