

**CO3 PROJECT**

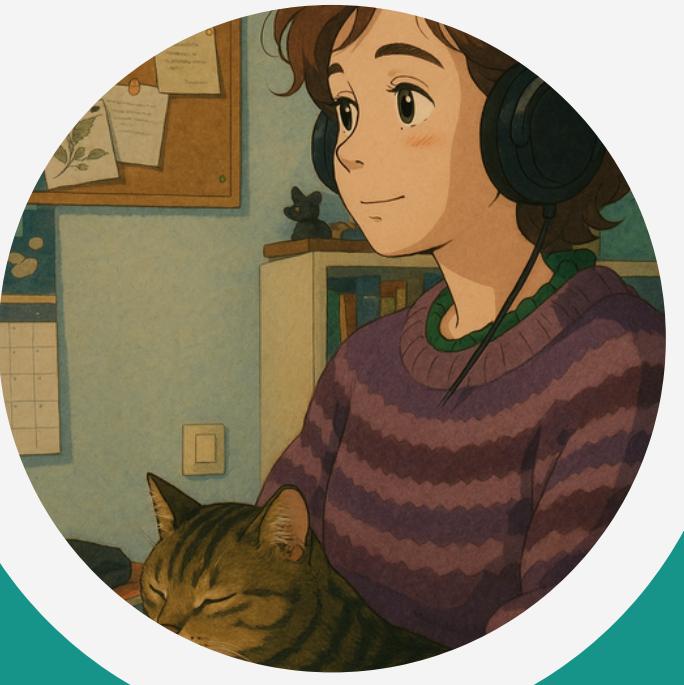
# **BIO INSPIRED ALGORITHMS**

# OUR TEAM

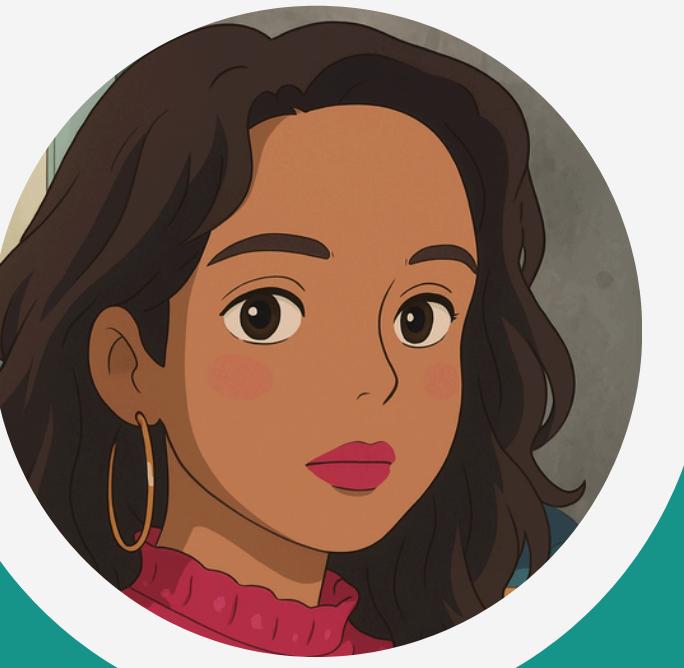
Group 4



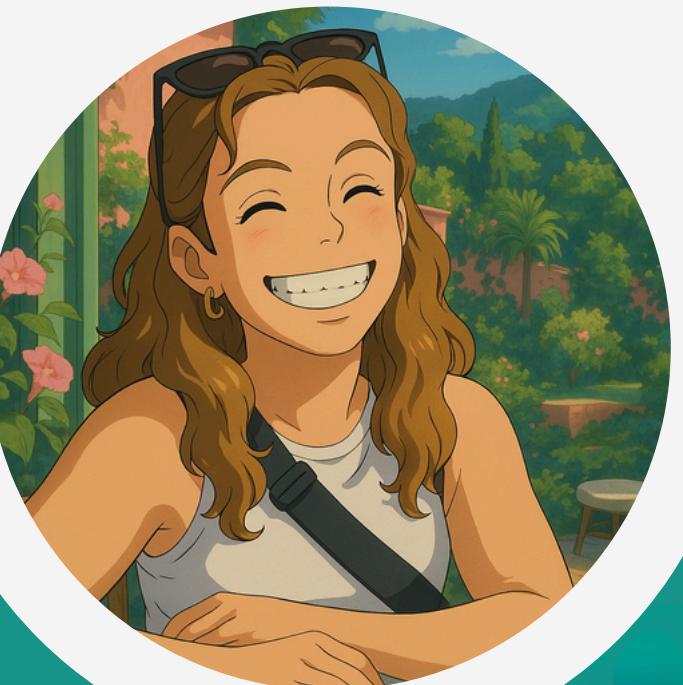
Mohan Adluru



Sara Kasraian



Shweta Dayal



Vanessa Geiss

# EMILY CARTER

**Age : 45 Yrs**

**Body Weight : 80 Kg**

**Volume Density : 200 mL/Kg**

**Insulin Sensitivity Factor : 20 mg/dL per unit**

**Carbs intake : 100 - 150 g**

**Insulin Dose : 1 - 10**

**Safe Glucose Range : 80 - 130 mg/dL**

**Metabolic Equivalent (met) : 4**

**r : 0.5 mg/dL per kcal burned**

**Exercise Duration : Rarely**



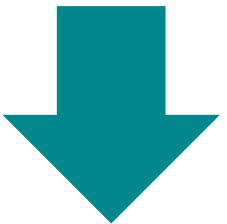
# Problem Statement

What insulin-meal-exercise daily plan keeps her glucose as close as possible to 100 mg/dL and in the safe range of 80 -130mg/dl?

# Mathematical Modelling

## Discrete Model

$$G_{\text{next}} = G_{\text{current}} + G_{\text{rise}} - G_{\text{drop}} - G_{\text{clear}} - G_{\text{ex}}$$



## Mini Bergmann

$$\frac{dG}{dt} = G_{\text{meal}}(t) - X \cdot G - G_{\text{ex}} - G_{\text{clear}}(t)$$

$$\frac{dX}{dt} = -p2 \cdot X + p3 \cdot (I - I_b)$$

$$\frac{dI}{dt} = -n \cdot (I - I_b) + u(t)$$



### Glucose Rise Due to Carbohydrates

$$G_{\text{rise}}(t) = \frac{C(t) \cdot k}{BW \cdot V_d}$$

### Glucose Reduction Due to Insulin:

$$G_{\text{red}}(t) = D(t) \times S$$

### Glucose Clearance Rate:

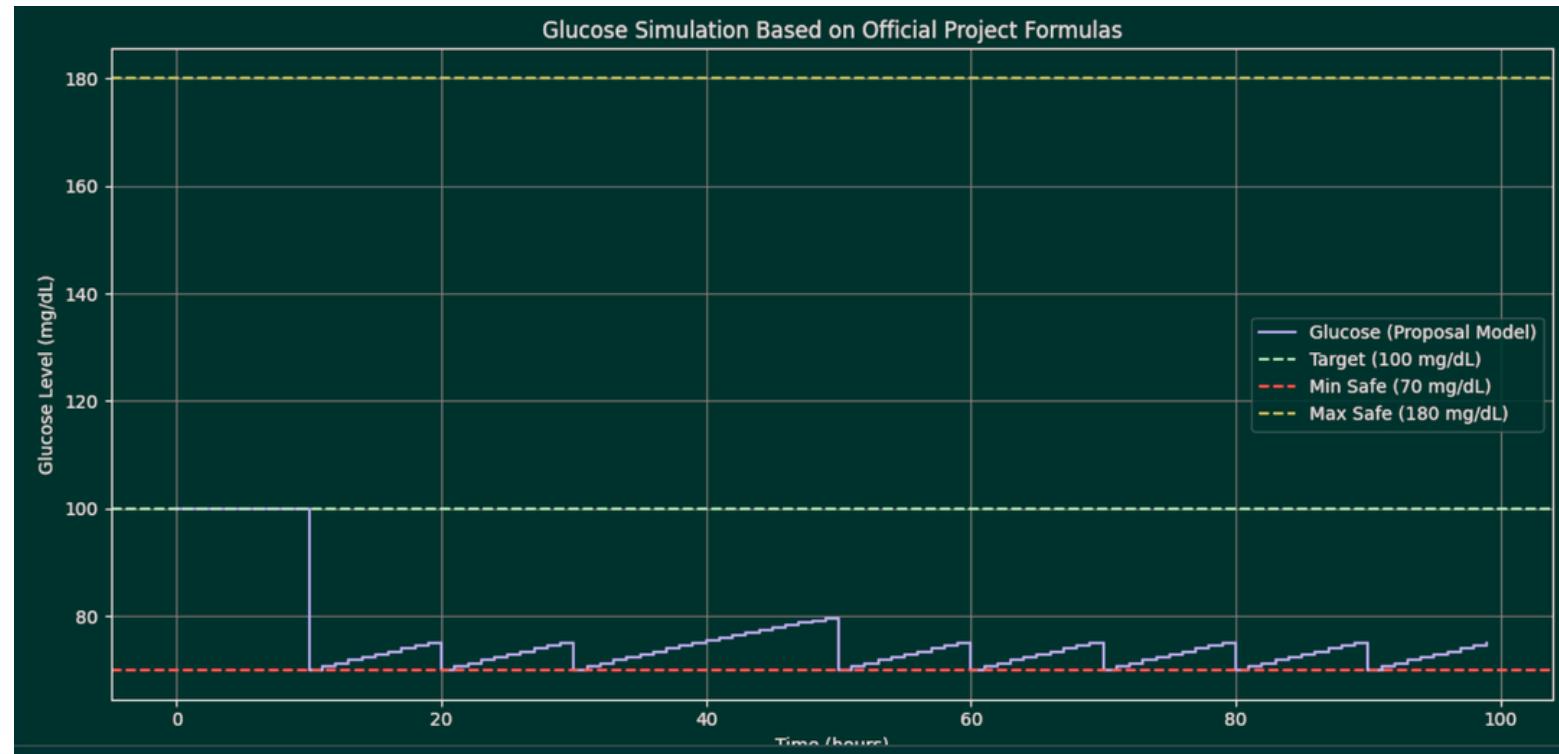
$$G_{\text{clear}}(t) = \frac{(G(t) - G_{\text{target}})}{4} \times T$$

### Calories Burned

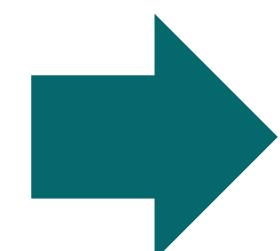
$$G_{\text{exercise}}(t) = C_{\text{burn}}(t) \times R$$

$$C_{\text{burn}}(t) = BW \times MET \times \text{Duration}(t)$$

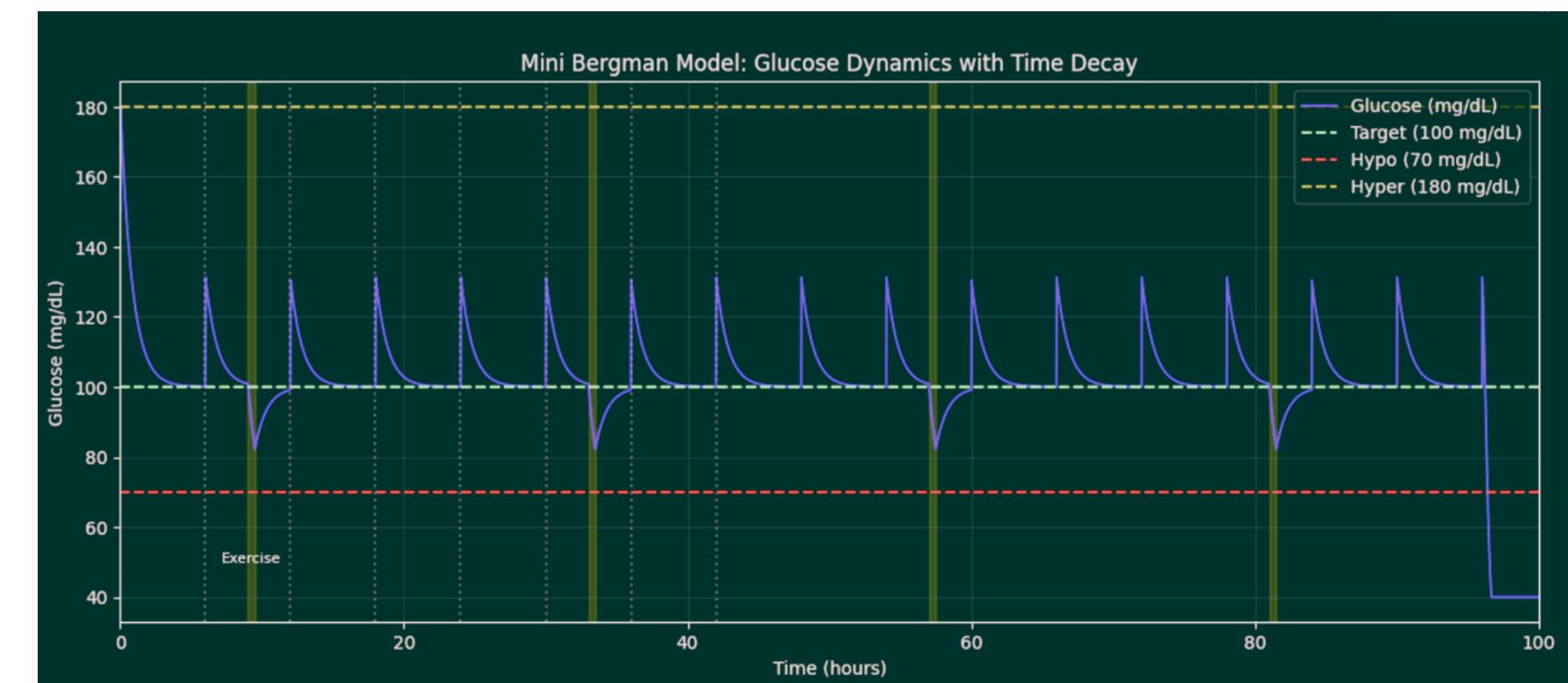
# Mathematical Modelling



Mini Bergmann



Discrete Model



# Constraints

## 01 Glucose level constraints

$$G_{\min} \leq G + G_{rise} - G_{red} - G_{clear} - G_{burn} \leq G_{\max}$$

## 02 Insulin dose constraints

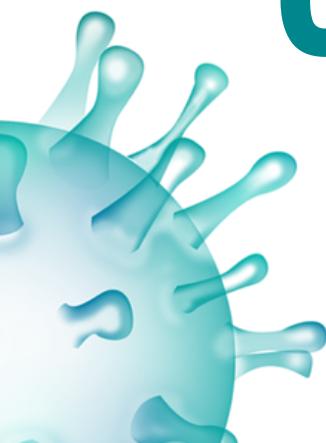
$$D_{\min} \leq D \leq D_{\max}$$

## 03 Carbohydrate intake constraints

$$C_{\min} \leq C \leq C_{\max}$$

## 04 Exercise constraints

$$0 \leq T \leq T_{\max}$$



# Objective Function

$$\frac{1}{T} \sum_{t=1}^T |G(t) - G_{\text{target}}| + \text{penalty}$$

# Grid Search

(1.0, 101.525)

## Constant Constraints

Parameter	Possible Values
Insulin Dosage	[2, 4, 6, 8] (4 values)
Carb Intake	[30, 50, 70] (3 values)
Exercise Duration	[20, 40] (1 values)
Exercise Time	[6 AM, 9 AM, 12 PM] (1 values)

Systematic Parameter Search  
Defined Search Space  
Computationally Expensive

## Dynamic Constraints

Parameter	Possible Values
Insulin Dosage	20 values (0.5 to 10 in 0.5 steps)
Carb Intake	30 to 80 (random integer)
Exercise Duration	20 to 60 minutes (random)
Exercise Time	Any hour in a 24-hour cycle (random)

# Why Genetic Algorithm?

Feature	Grid Search	Genetic Algorithm
Dynamic Insuline Adjustments	✗ Hard to implement efficiently	✓ Adjusts insulin adaptively
Meal Timing Constrains	✗ Static, doesn't adapt	✓ Adapts meal-insulin timing dynamically
Exercise Optimization	✗ Fixed values only	✓ Learns best times/intensity dynamically
Computational Efficiency	✗ Explodes with more constraints	✓ Learns efficiently over generations

# Genetic Algorithm

## 01 Initialization

Initialize Population P

02

## Mating

Recombine Parent Population P using Evolutionary Operators and create Offspring O

Merge the Population P and Offspring M = P' O

04

## Survival

Return the fittest individuals of M as the new Population P  
[Stopping Criterion reached]

05

## Termination

Stopping Criterion based on number of Function Evaluations, Generations, etc.

## Selection

Select Parents for Recombination

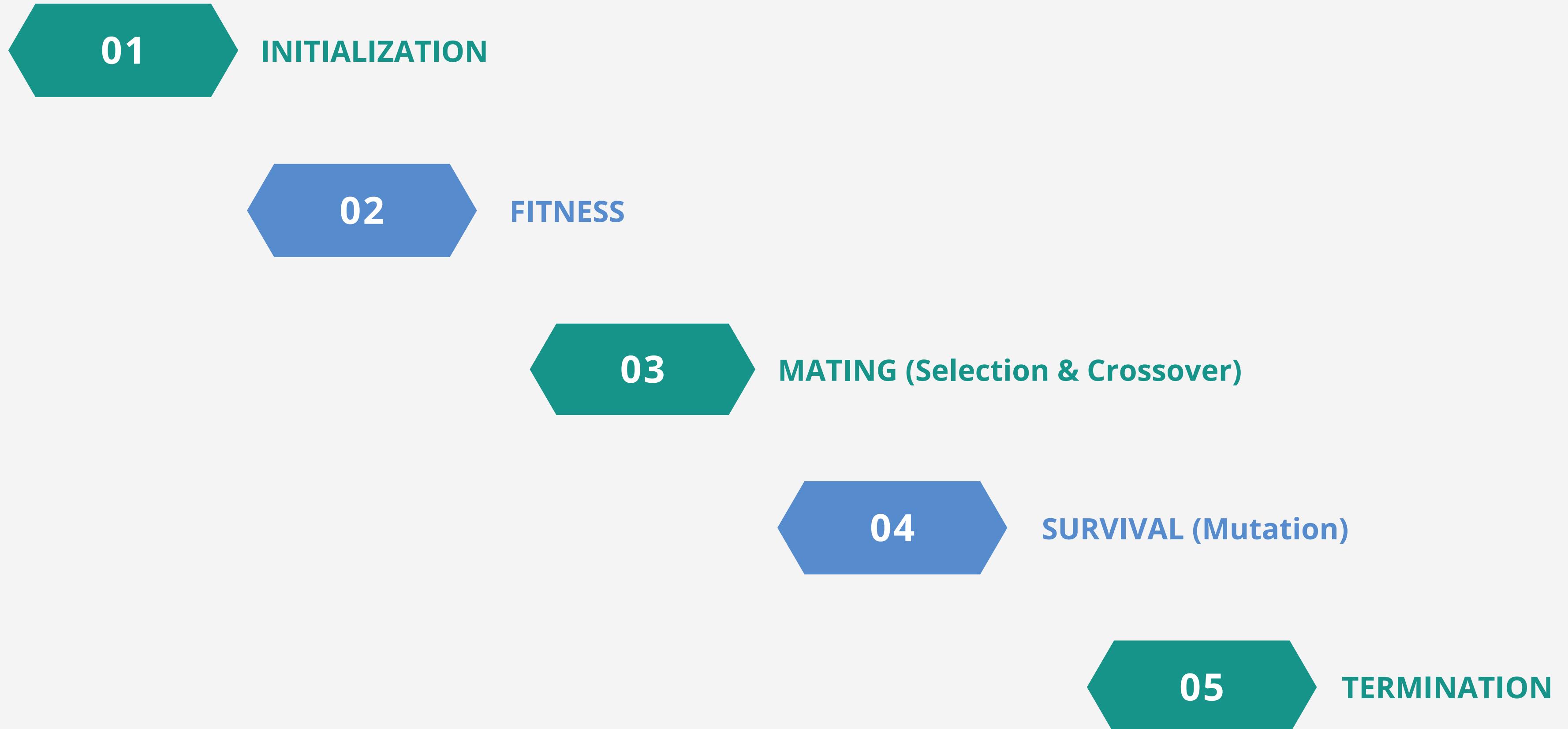
## Crossover

Create offspring from Parents

## Mutation

Mutate the Offspring

# Genetic Algorithm



# Grid Search vs Genetic Algorithm

Difference	Grid Search	Genetic Algorithm
Method	<b>Brute-force method: tries every combination from a pre-defined grid</b>	A bio-inspired, evolutionary method that uses selection, crossover, mutation to improve solutions over time
Start	<b>Define a grid of values</b>	Start with random population of solutions
Evaluation	<b>Try all combinations</b>	Evaluate fitness of each individual
Improvement	<b>No learning - just exhaustive</b>	Select best → crossover → mutate → repeat
Stopping	<b>When all combos are tested</b>	After fixed number of generations

# Results

**with GA more possible Day plans with same Fitness score**

```
==== Optimal Solution ===
```

```
Carb Intake Schedule: [50, 60, 80, 30]
```

```
Total Glucose Deviation: 16.57
```

```
==== Glucose Levels at Each Time Step ===
```

```
Time 8: 101.54 mg/dL
```

```
Time 12: 97.17 mg/dL
```

```
Time 18: 94.31 mg/dL
```

```
Time 21: 92.92 mg/dL
```

```
Best Deviation Score: 428.20
```

```
Meals: [(233, 34), (574, 61), (1186, 41)]
```

```
Insulin: [(75, 8.2), (468, 8.5), (1209, 9.3)]
```

```
Exercise: [(1184, 12, 7.5)]
```



# SAMUEL THOMPSON

Age : 60 Yrs

Body Weight : 85 Kg

Volume Density : 150 mL/Kg

Insulin Sensitivity Factor : 10 mg/dL per unit

Carbs intake : 50 - 100 g

Insulin Dose : 1 - 10 Units

Safe Glucose Range : 80 - 130 mg/dL

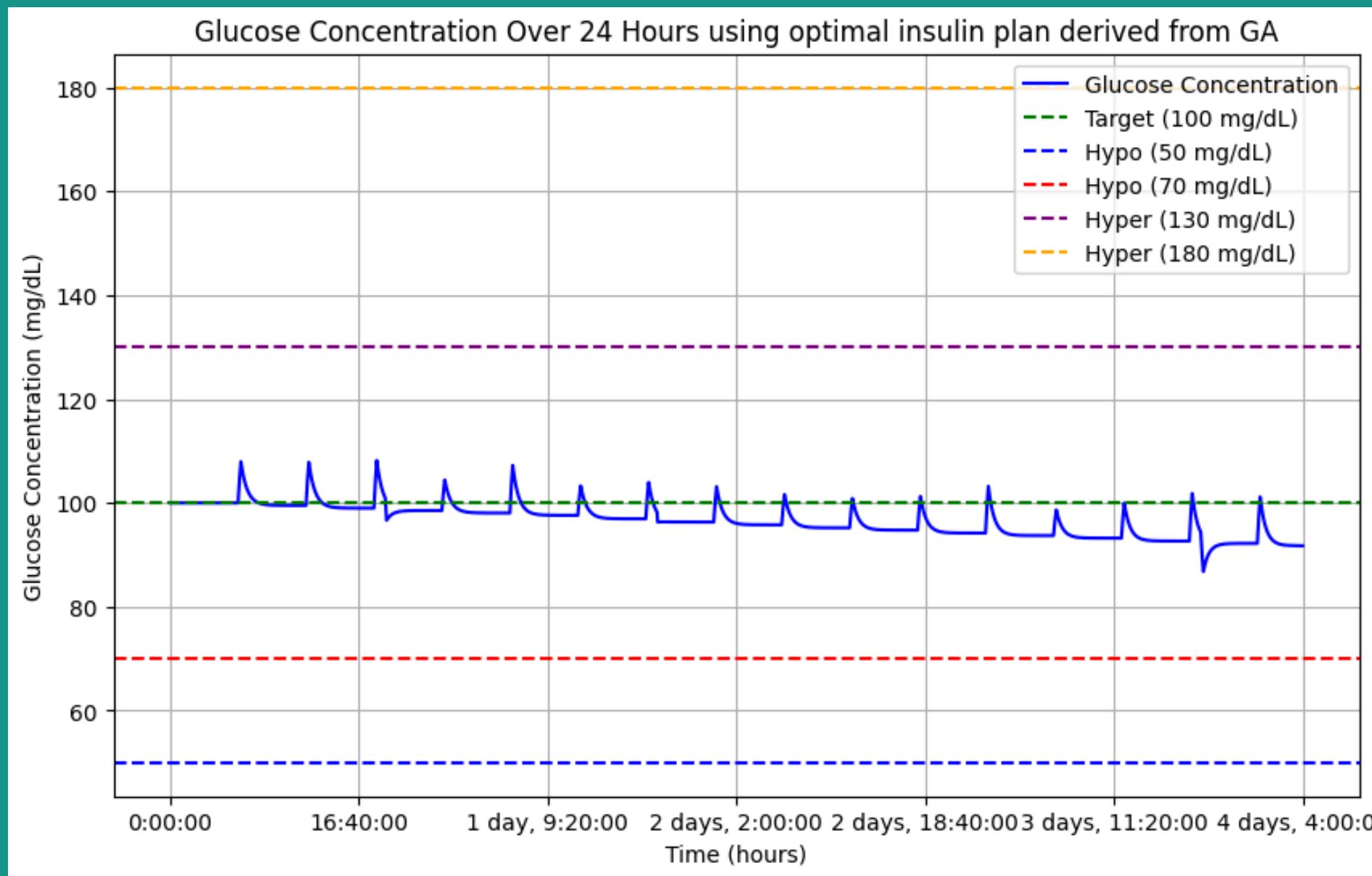
Metabolic Equivalent (met) : 4

Exercise Duration : 15 - 60 min

r : 0.1 mg/dL per kcal burned



# Results



Optimizing Insulin Dosage

## Optimized Time

✓ Best Plan Found:

Meal Times: [134, 433, 933] min

Insulin Times: [194, 754, 908] min

Exercise Time: 930 min

Exercise Duration: 30 min

Exercise Intensity (MET): 2.68

Glucose Deviation Score: 403.33

GA completed in 58.86 seconds

Total combinations explored: 2500

Best Deviation Score: 416.31

Plan 1

Score: 416.31 | REM: 2.15

Meals: [(170, 35), (550, 32), (1130, 31)]

Insulin: [(277, 7.8), (628, 3.5), (1126, 3.9)]

Exercise: [(962, 15, 2.9)]

Search Space like quadrillions of possible plans

# LISA

**Age : 30 Yrs**

**Body Weight : 60 Kg**

**Carbs intake : 50 - 80 g**

**Meal Plan: Plant-based (*Vegan*)**

**Volume Density : 200 mL/Kg**

**Insulin Sensitivity Factor : 50 mg/dL per unit**

**Insulin Dose : 0 Units**

**Exercise Duration : 60 - 90 min**

**r : 0.5 mg/dL per k**



# LISA

## RESULTS !?

Fun Exercise To Try 😊



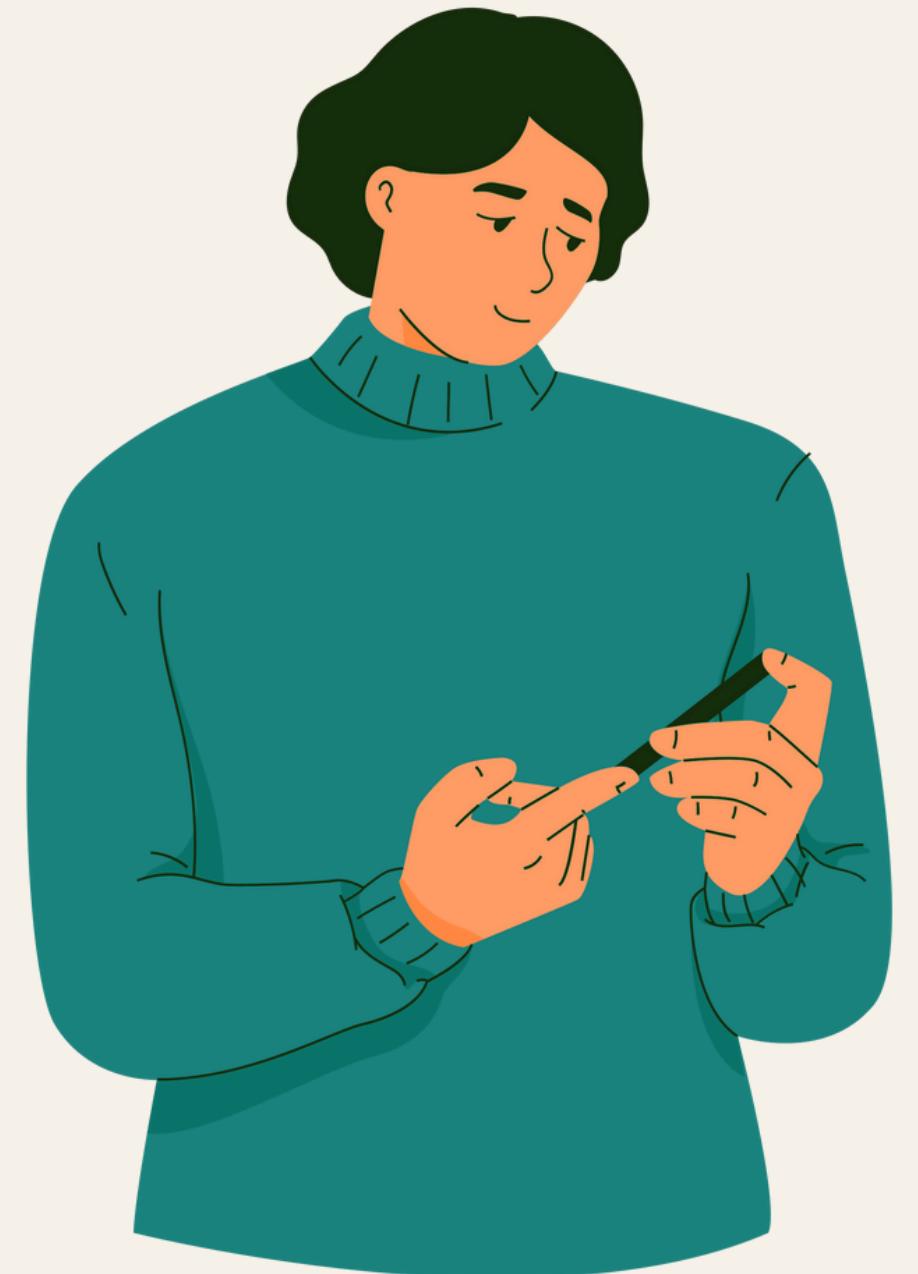
# Conclusion

**Further refinement  
is possible**

**Small steps make  
big problems  
solvable**

**Future work  
could  
include testing  
additional  
algorithmic  
strategies**

# THANK YOU!



## QUESTIONS?