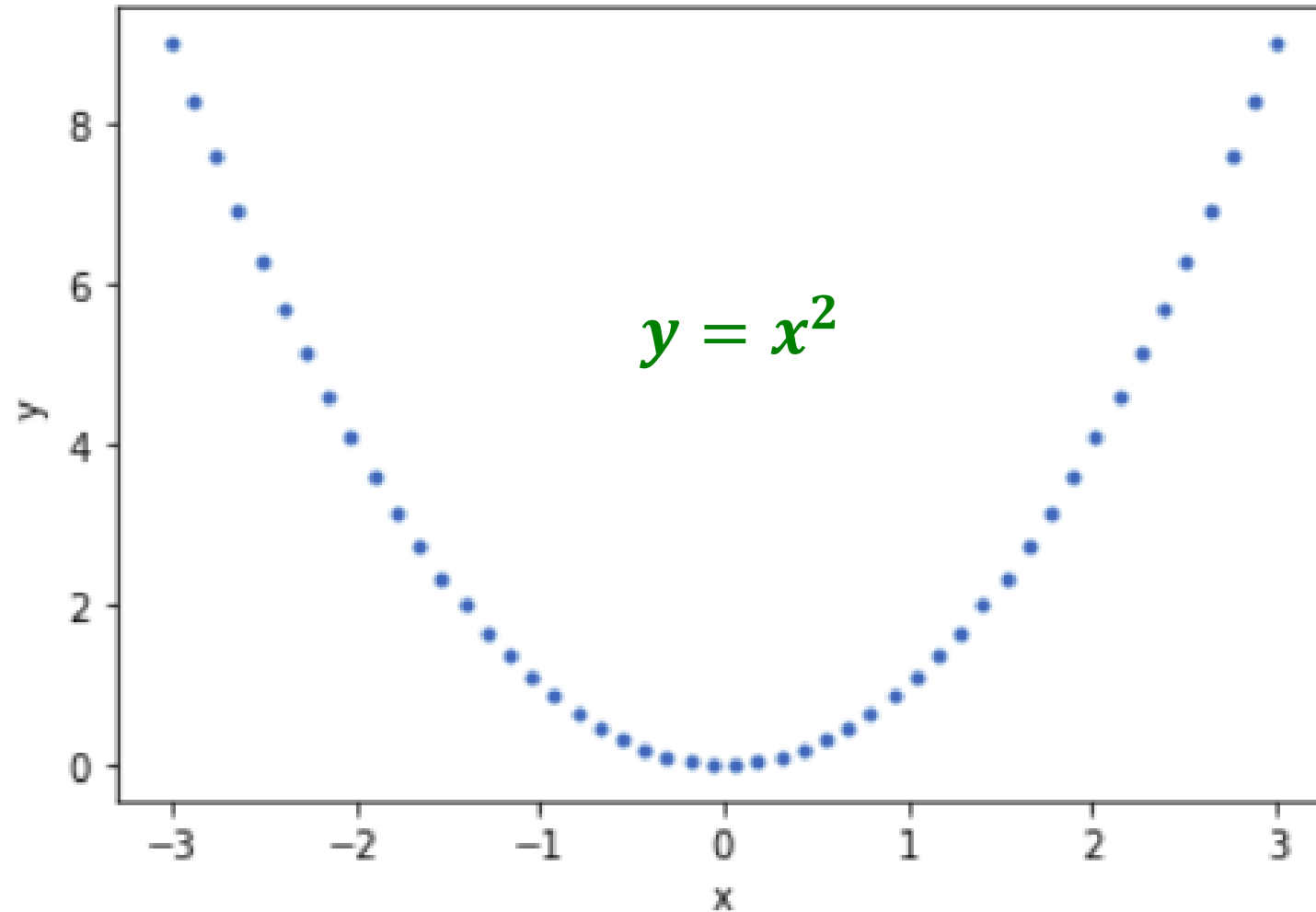




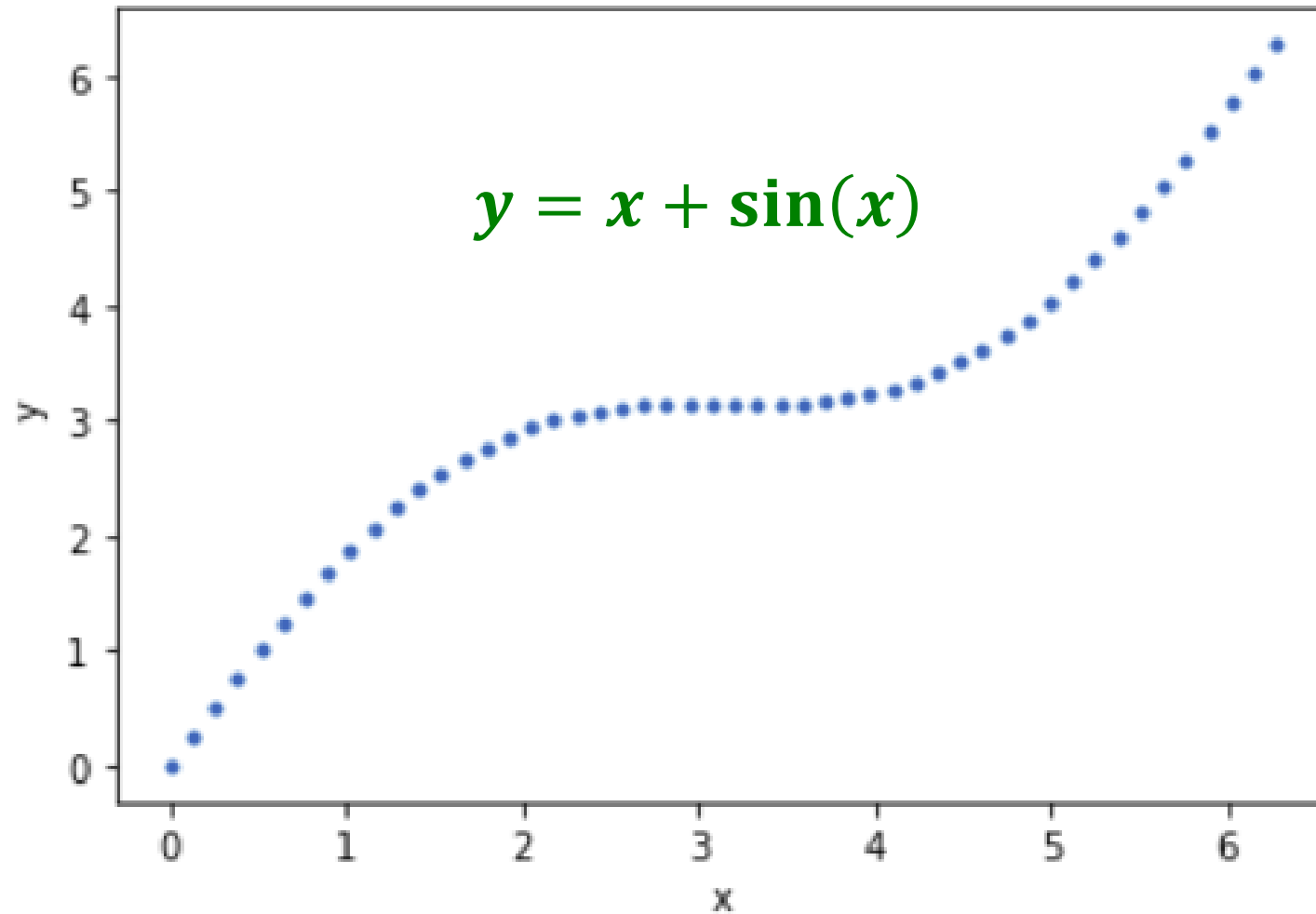
Symbolic Regression with Genetic Programming

Geoffrey Bomarito

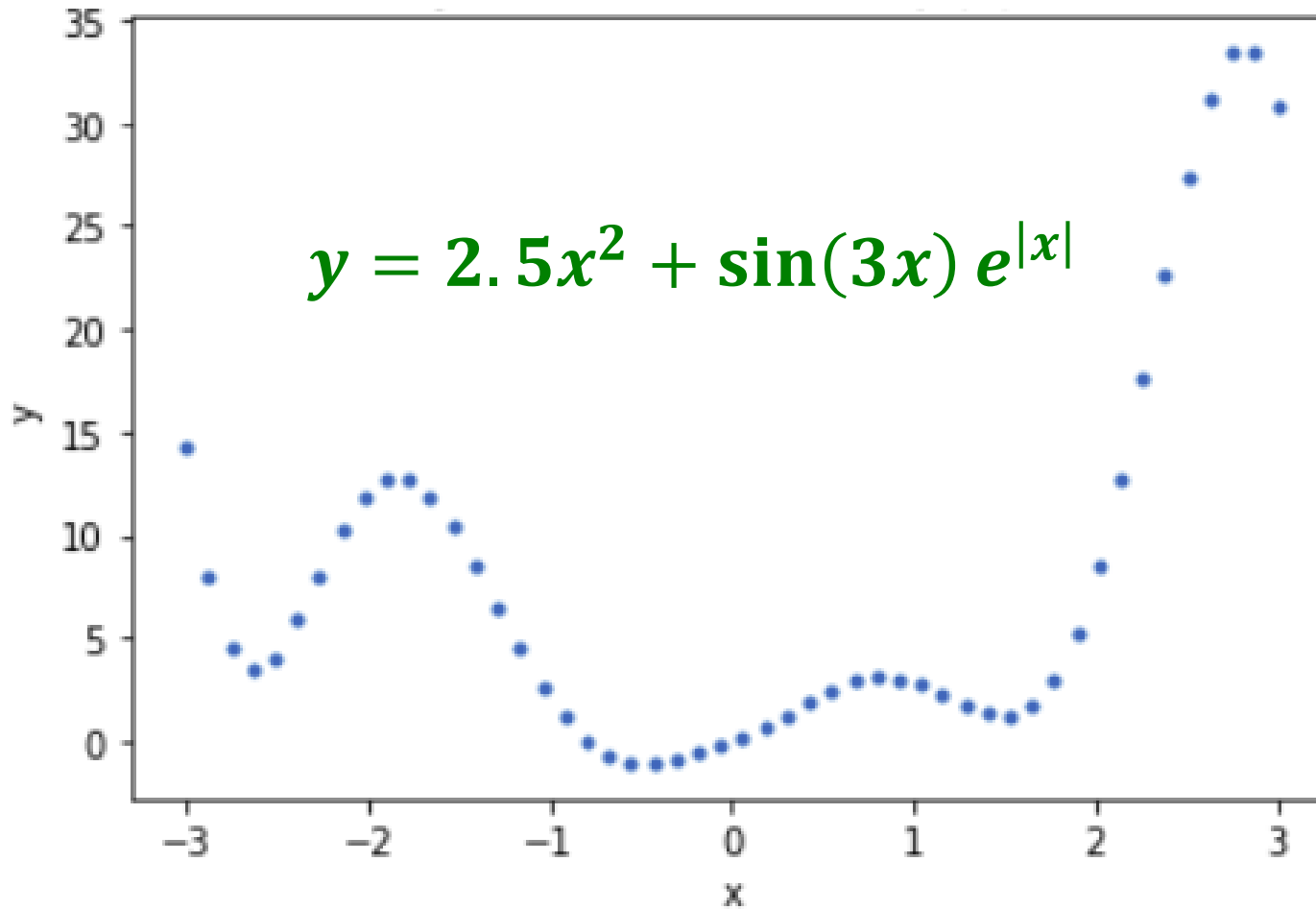
Can you name the equation?



Can you name the equation?



Can you name the equation?



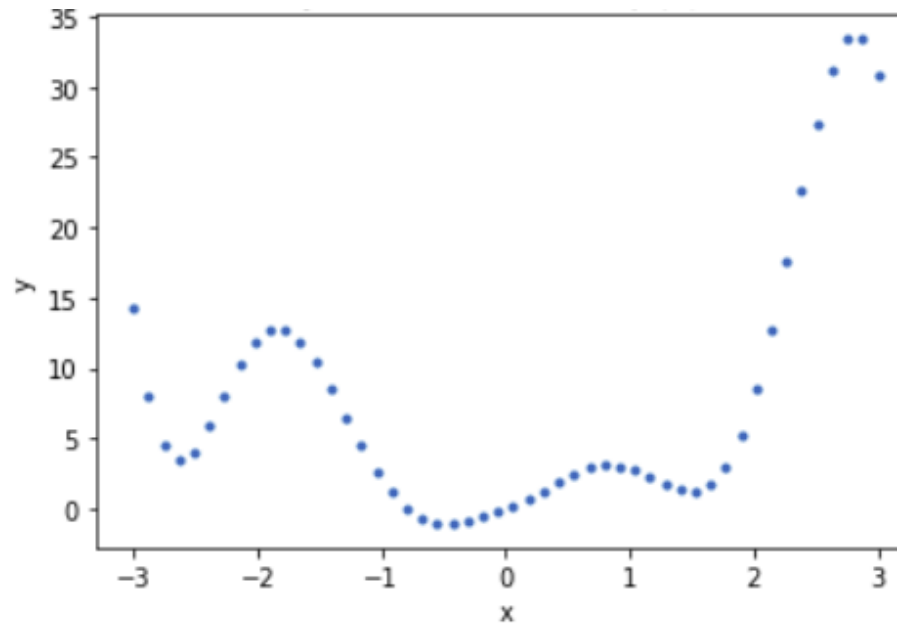
Symbolic regression

$$y = a_1x + a_2x^2 + a_3x^3 + \dots$$

$$y = a_1 \sin(a_2x) + a_3x + \dots$$

Standard regression

$$y = 2.5x^2 + \sin(3x) e^{|x|} \quad \text{Symbolic regression}$$



Genetic Programming (GP)



The evolutionary approach to symbolic regression

- Why GP?
 - Discrete optimization
 - Not the most compute efficient
 - But has seen huge success
- Alternatives:
 - Deep symbolic regression
 - Prioritized grammar enumeration

$$y = a_1 \sin(a_2 x) + a_3 x$$



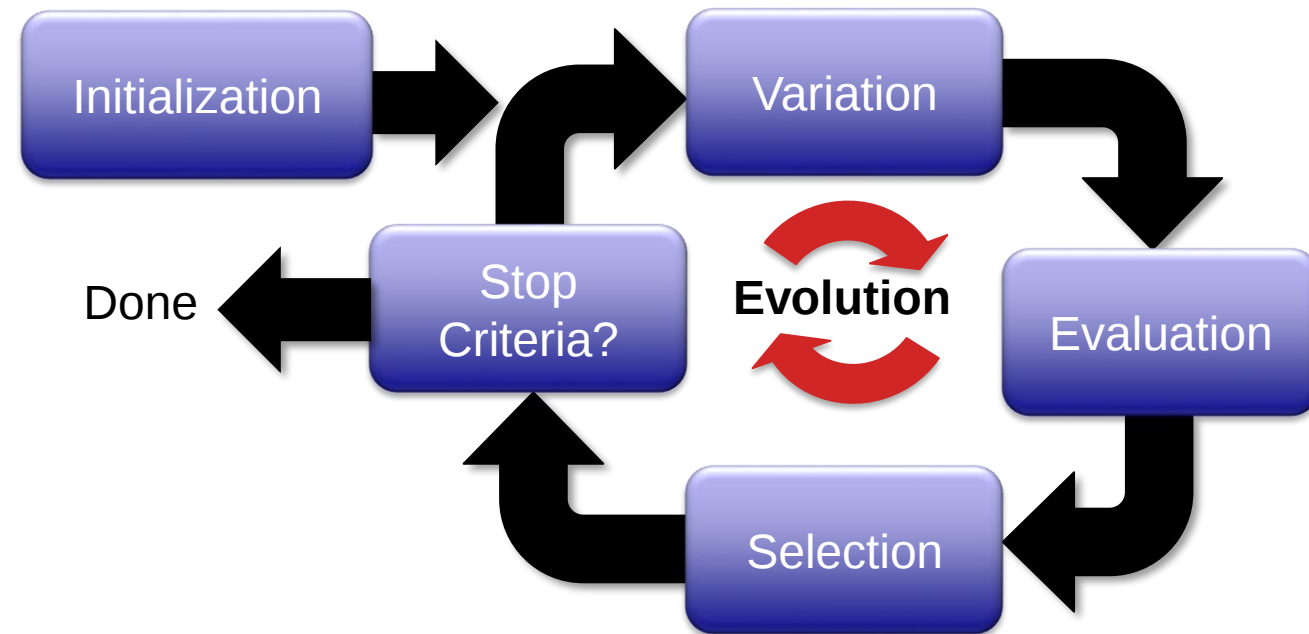
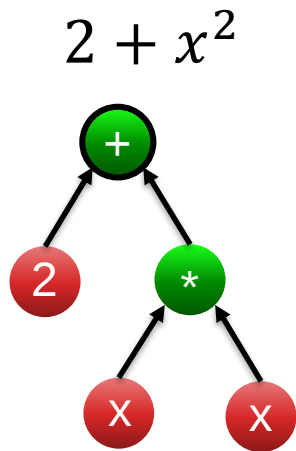
$$y = a_1 + a_3 x^2 + \frac{a}{x}$$

Genetic Programming (GP)



The evolutionary approach to symbolic regression

- Evolving equations to match data
- Equations as graphs



Initialization

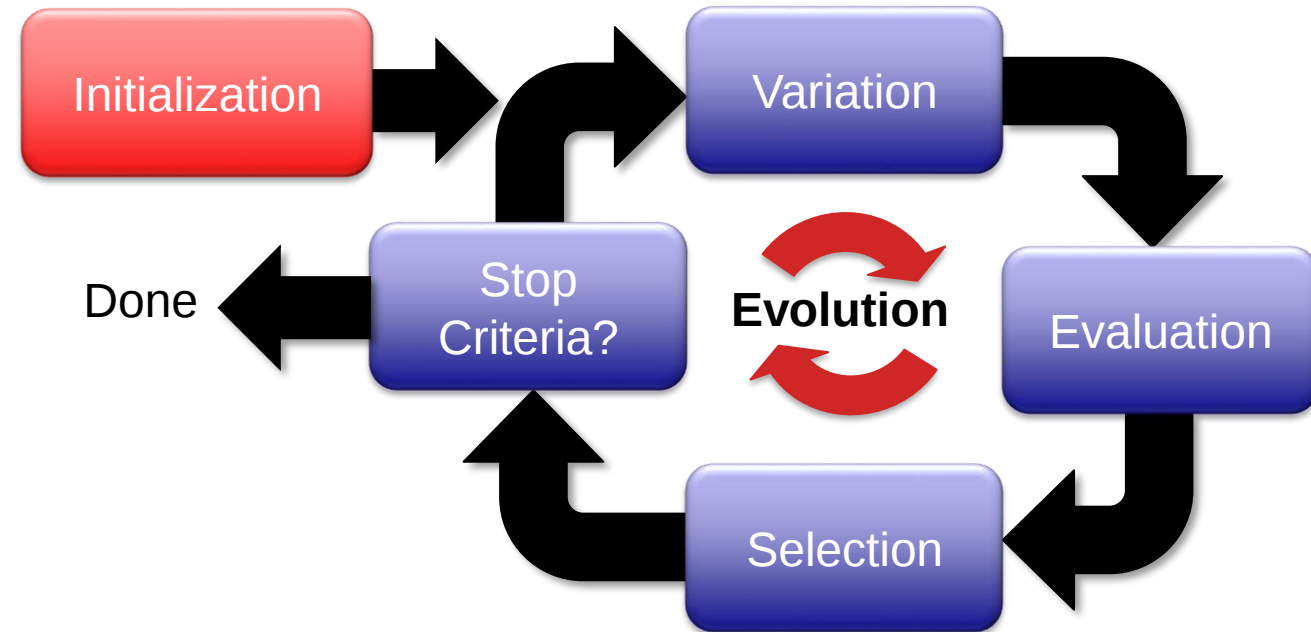
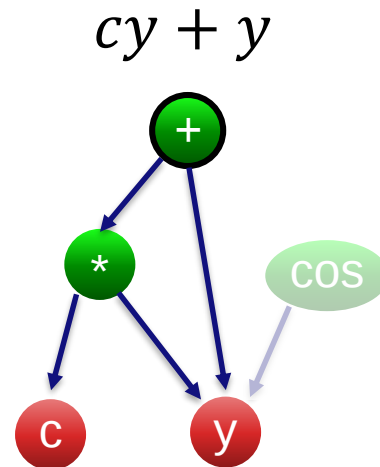
Randomly add nodes (and connections if necessary)

- Set of operators

+ - * COS

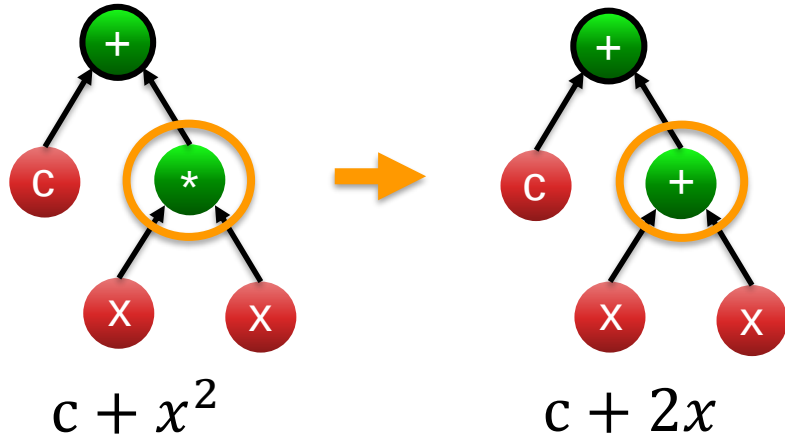
- Set of terminals

c x y

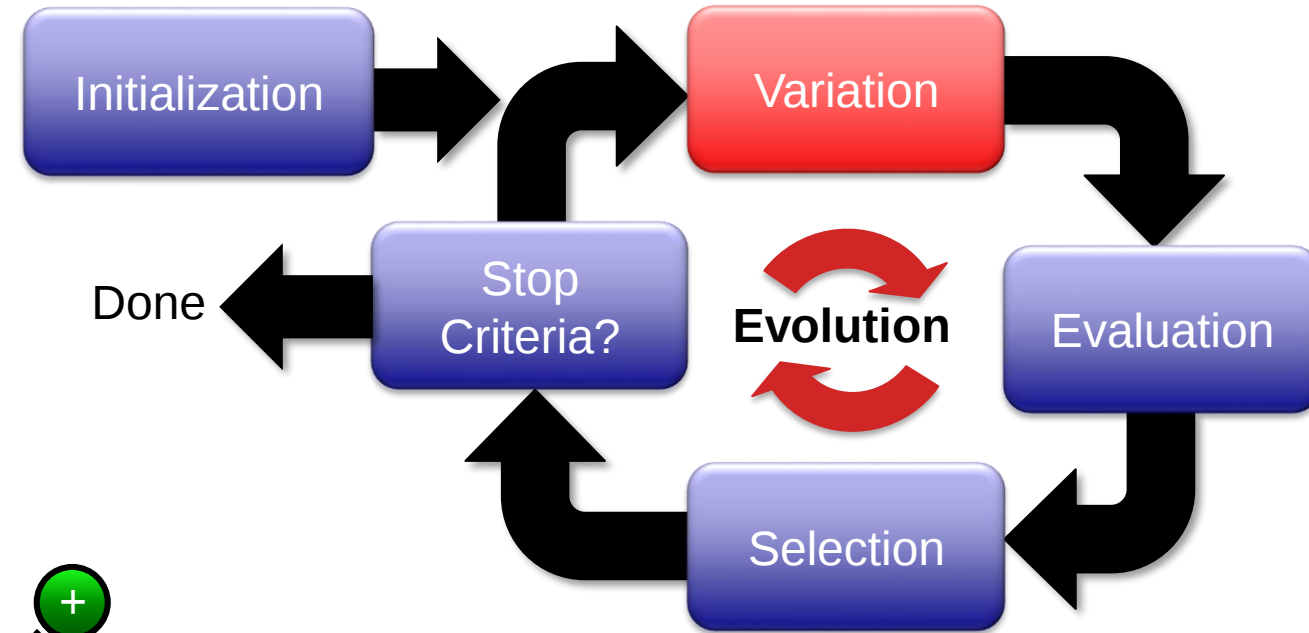
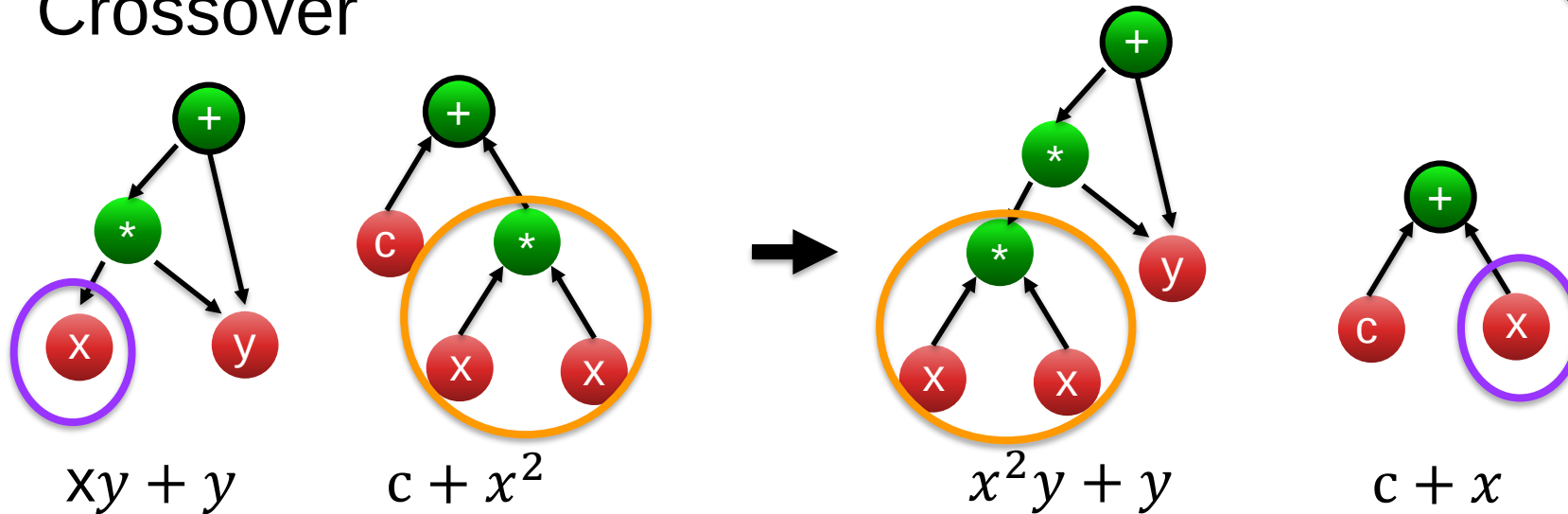


Variation

Mutation



Crossover

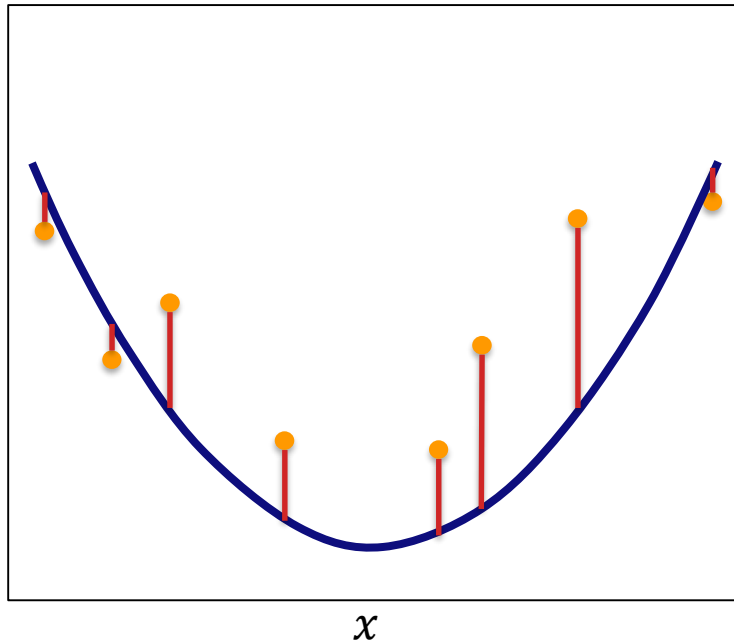


Evaluation

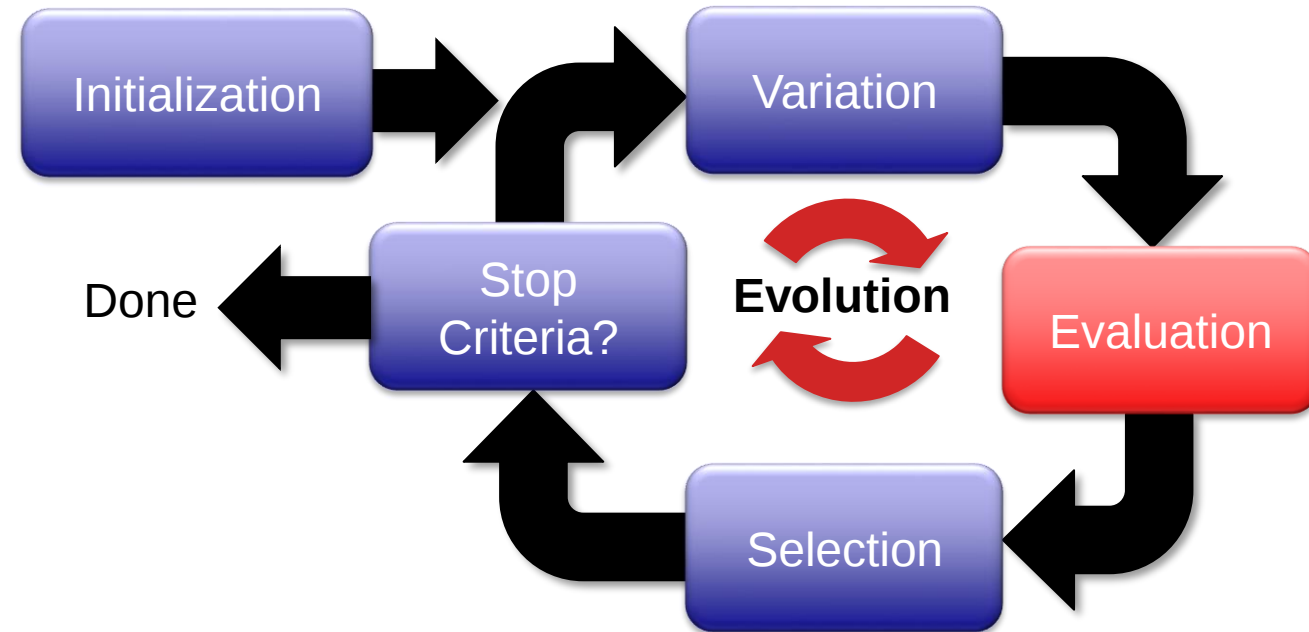
Fitness: How well the equation matches the training data

Explicit Regression

$$c + x^2$$



Mean absolute error
Mean squared error
Etc.

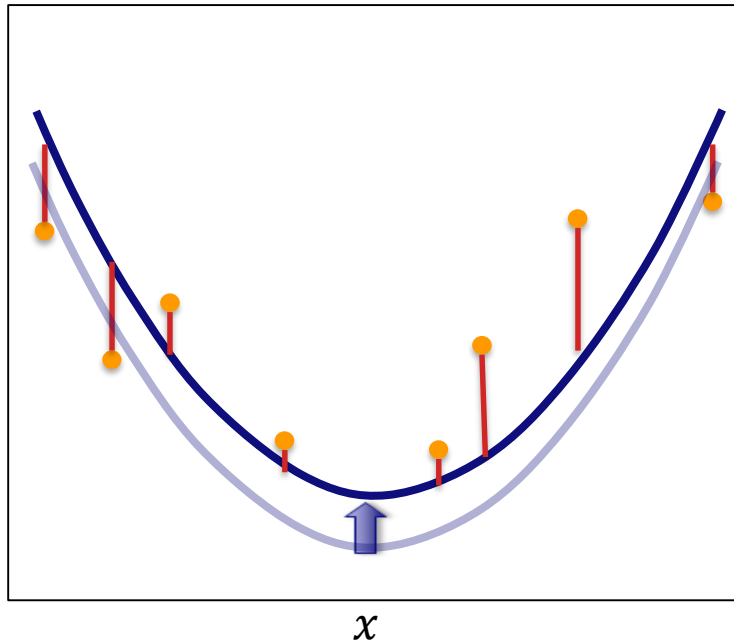


Evaluation

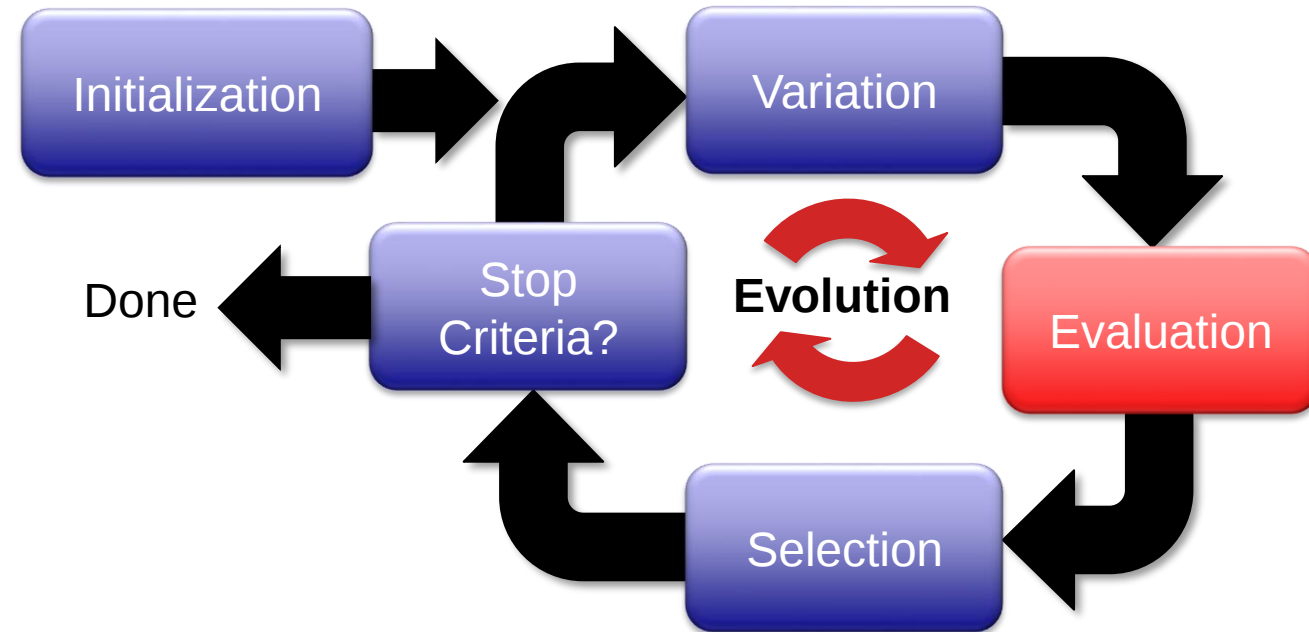
Fitness: How well the equation matches the training data

Explicit Regression

$$c + x^2 \rightarrow 2.7 + x^2$$



Local optimization performed for constants

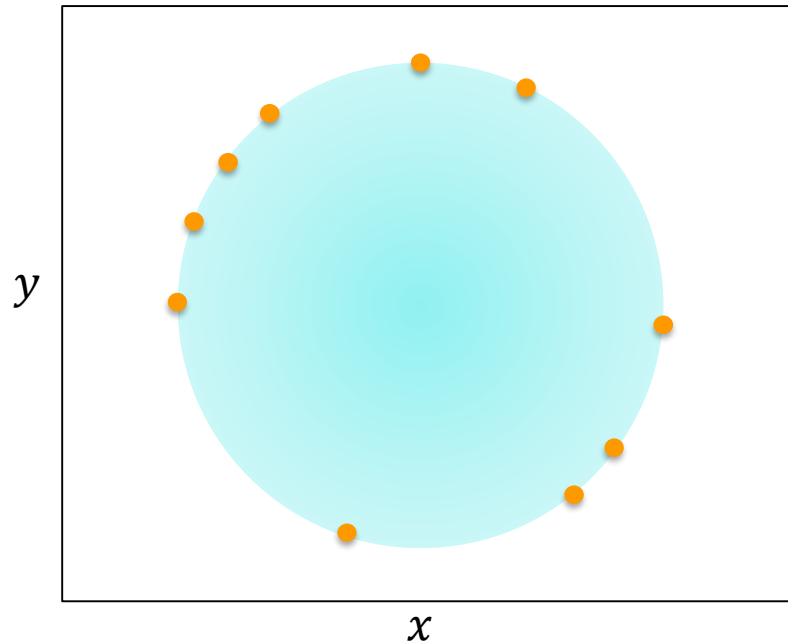


Evaluation

Fitness: How well the equation matches the training data

Implicit Regression

Searching for invariants in the data



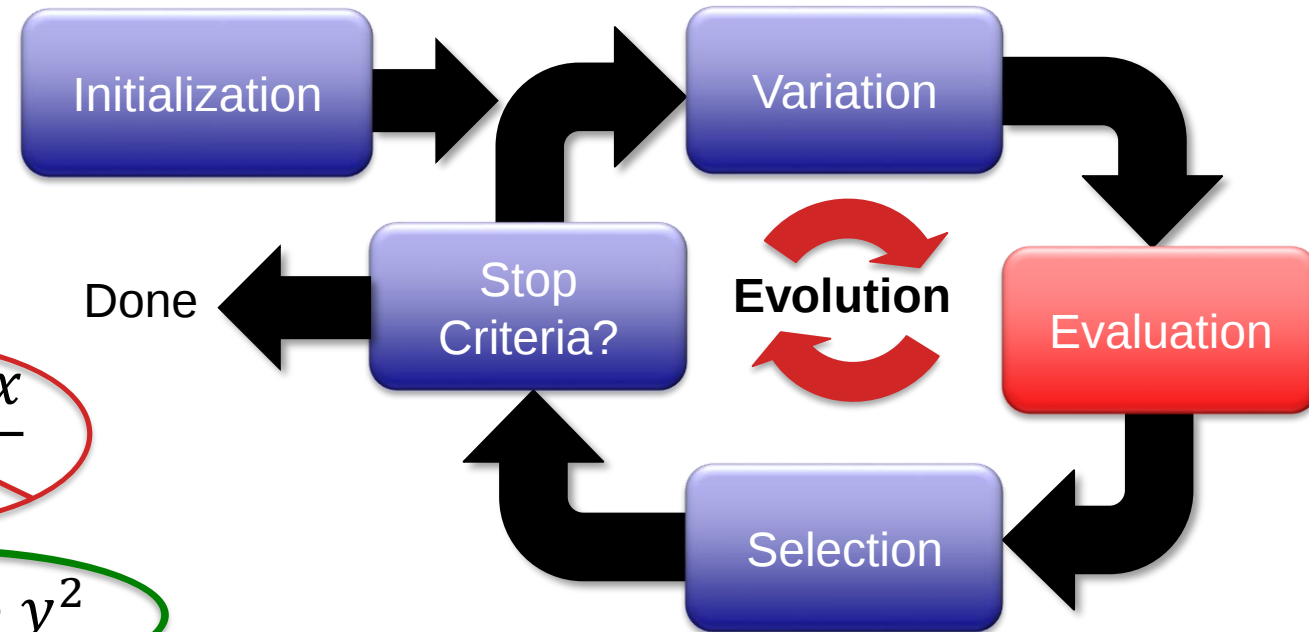
$$f(x, y) = \text{constant}$$

Three candidate equations are shown, each in a red oval with a diagonal line through it, indicating they are rejected:

- $f = 0$
- $f = \frac{x + x}{x}$

The correct equation is shown in a green oval:

$$f = x^2 + y^2$$



Survival of the fittest

■ Tournament

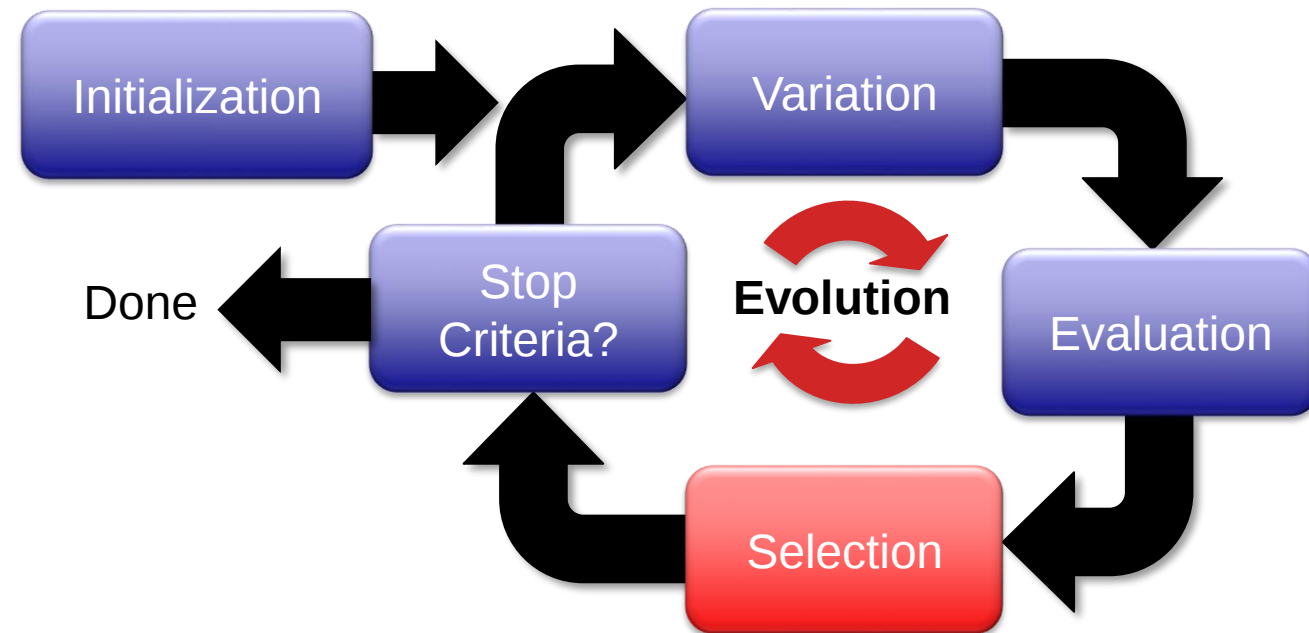
- Randomly selected groups compete
- **Most common**

■ Deterministic crowding

- Children compete against parents
- **Promotes niching**

■ Age-fitness

- Equations compete against other similarly aged equations
- **Mitigates premature convergence**



Stopping criteria

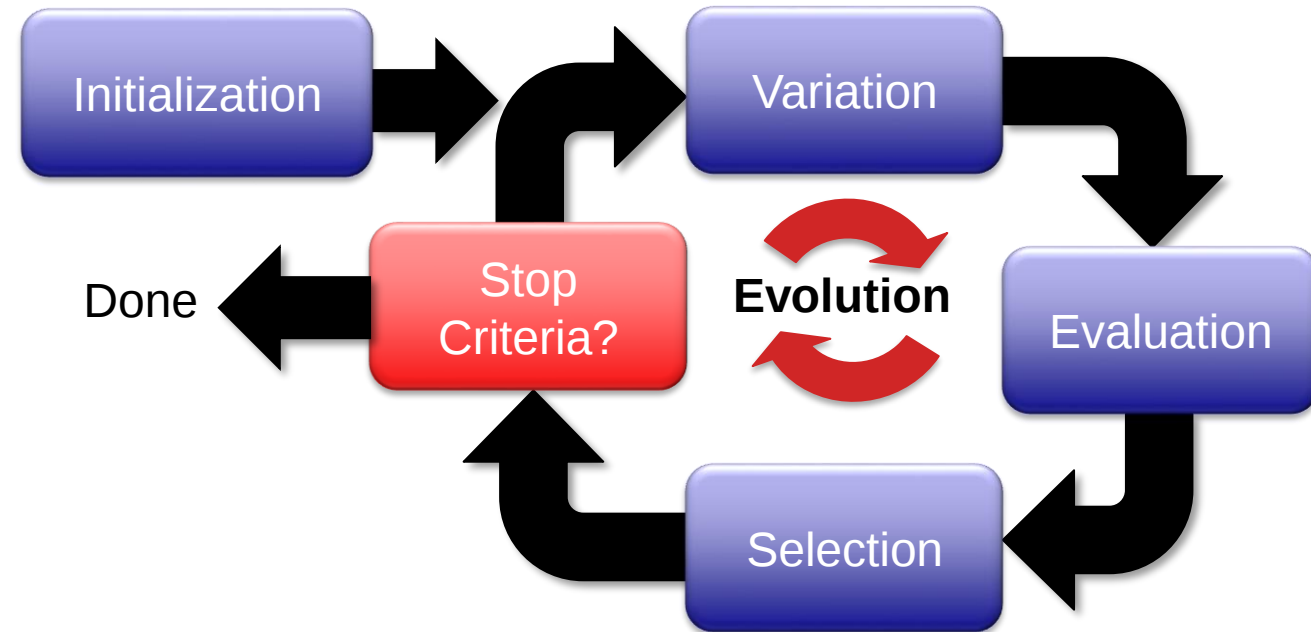


Is evolution done?

- Convergence
- Stagnation
- Time limit

NO:
repeat

YES:
results in equation(s) that
best match data

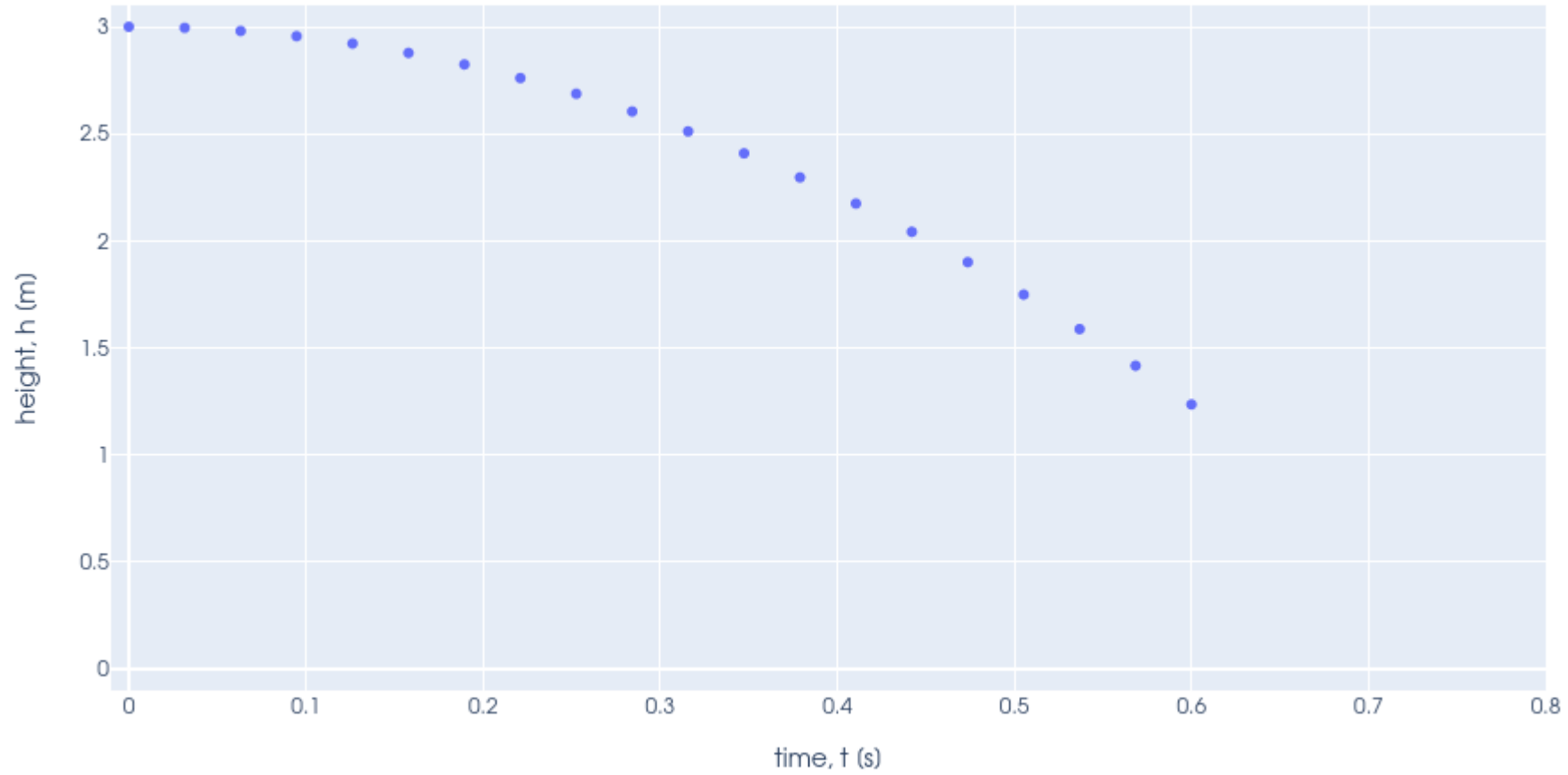




Live Demo

Apple dropping example

Explicit Regression



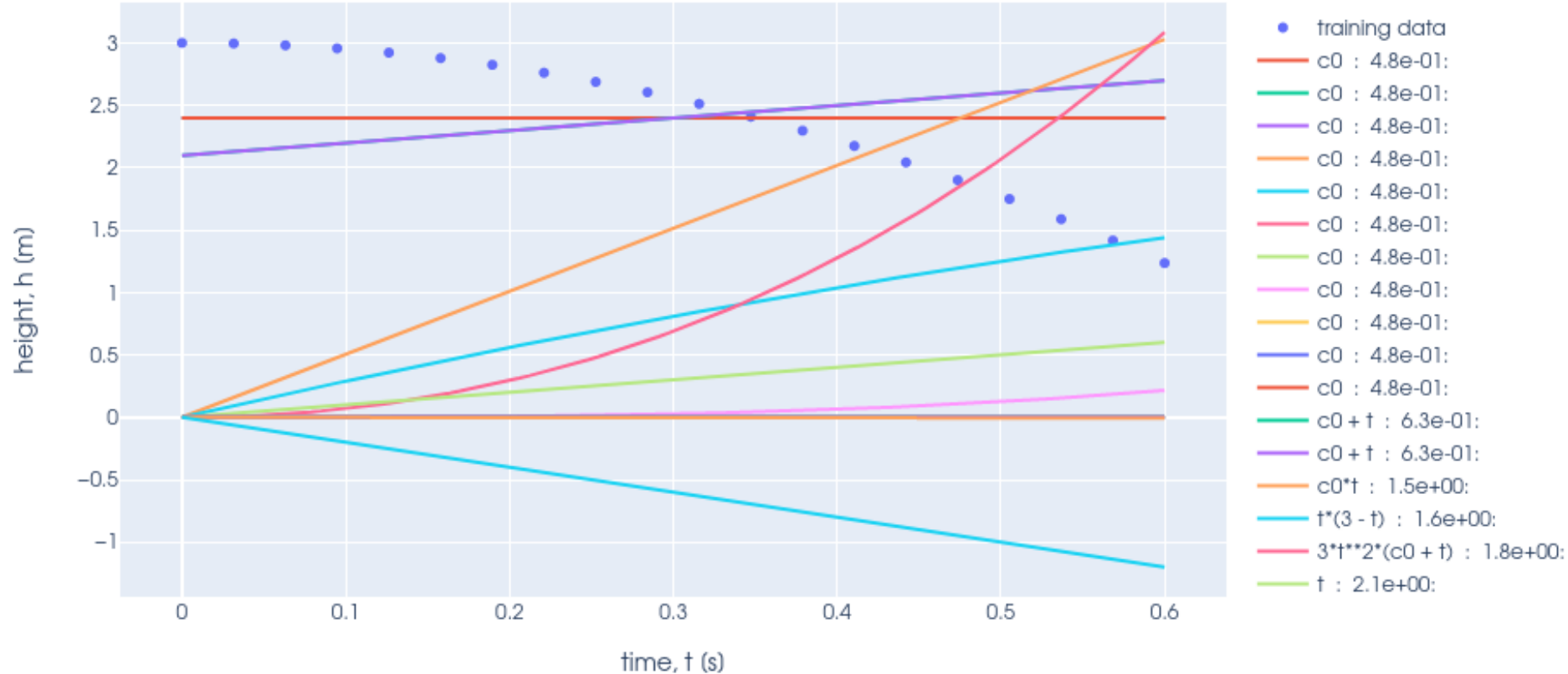
Initial Population

Inputs: Variable: t
Operators: $+$, $-$, $*$ Training data from previous slide

```

0: t**5*(t - 1)*(2*t - 1)
1: c0
2: c0
3: c0
4: c0 + t
5: 0
6: c0
7: c0 + t
8: 0
9: 0
10: t**3
11: c0*t
12: c0
13: t
14: c0
15: 0
16: 0
17: c0
18: c0
19: t*(3 - t)
20: c0
21: 3*t**2*(c0 + t)
22: c0
23: -2*t
24: c0
    
```

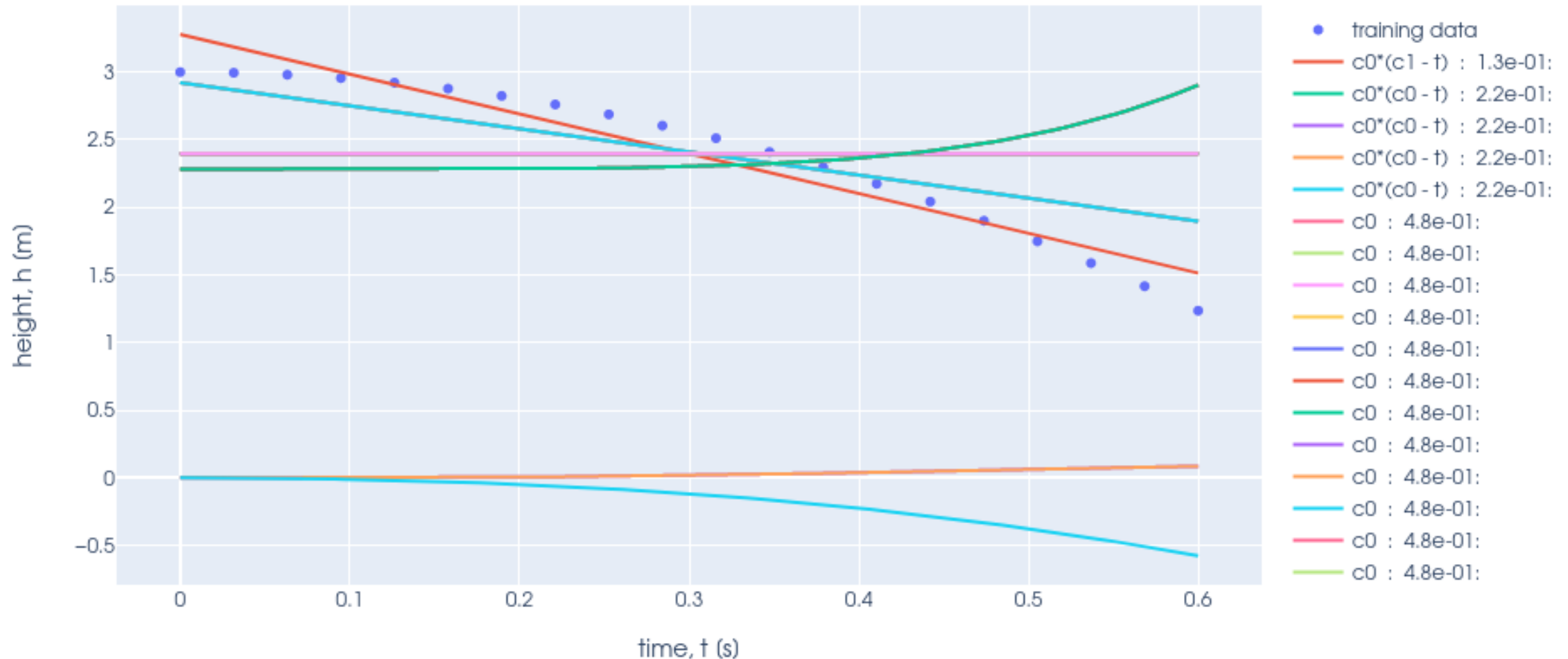
Population_age: 0 Best: $h = 2.396526315789522$



Population Evolution



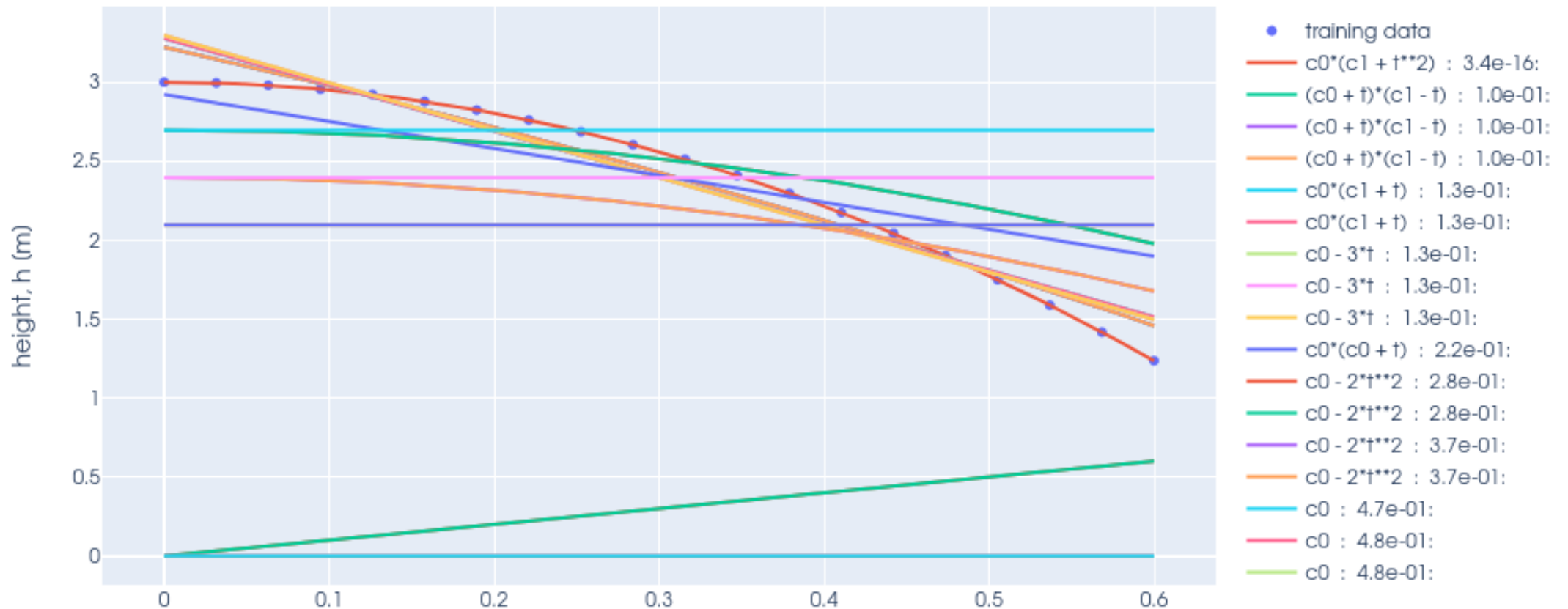
Population_age: 10 Best: $h = 3.2785263147135386 - 2.9399999959915752 * t$



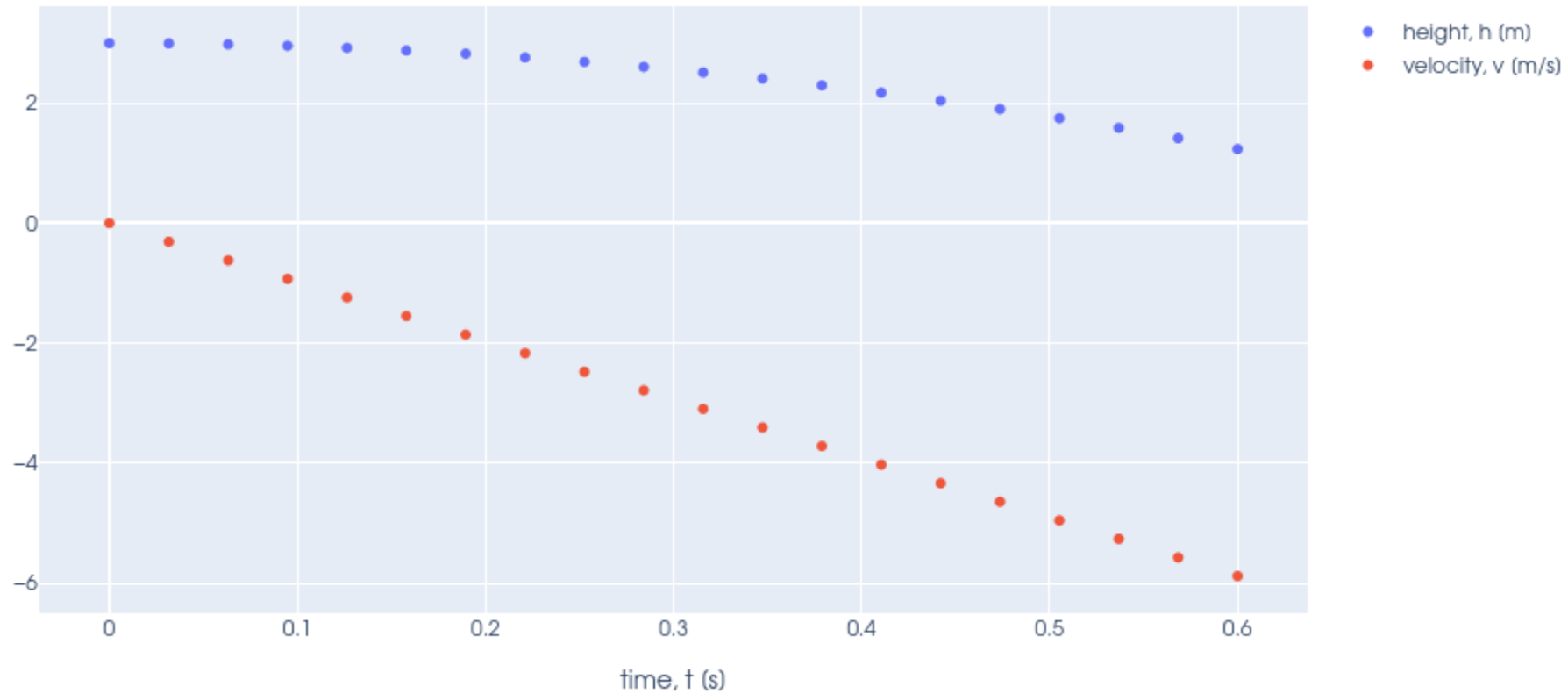
Population Evolution

h_0 $\frac{1}{2}g$
 Population_age: 187 Best: $h = 3.0 - 4.9000000000000001 \cdot t^{**2}$

$$h = h_0 + \frac{1}{2}gt^2$$



Implicit Regression



$$f(h, v) = \text{constant}$$

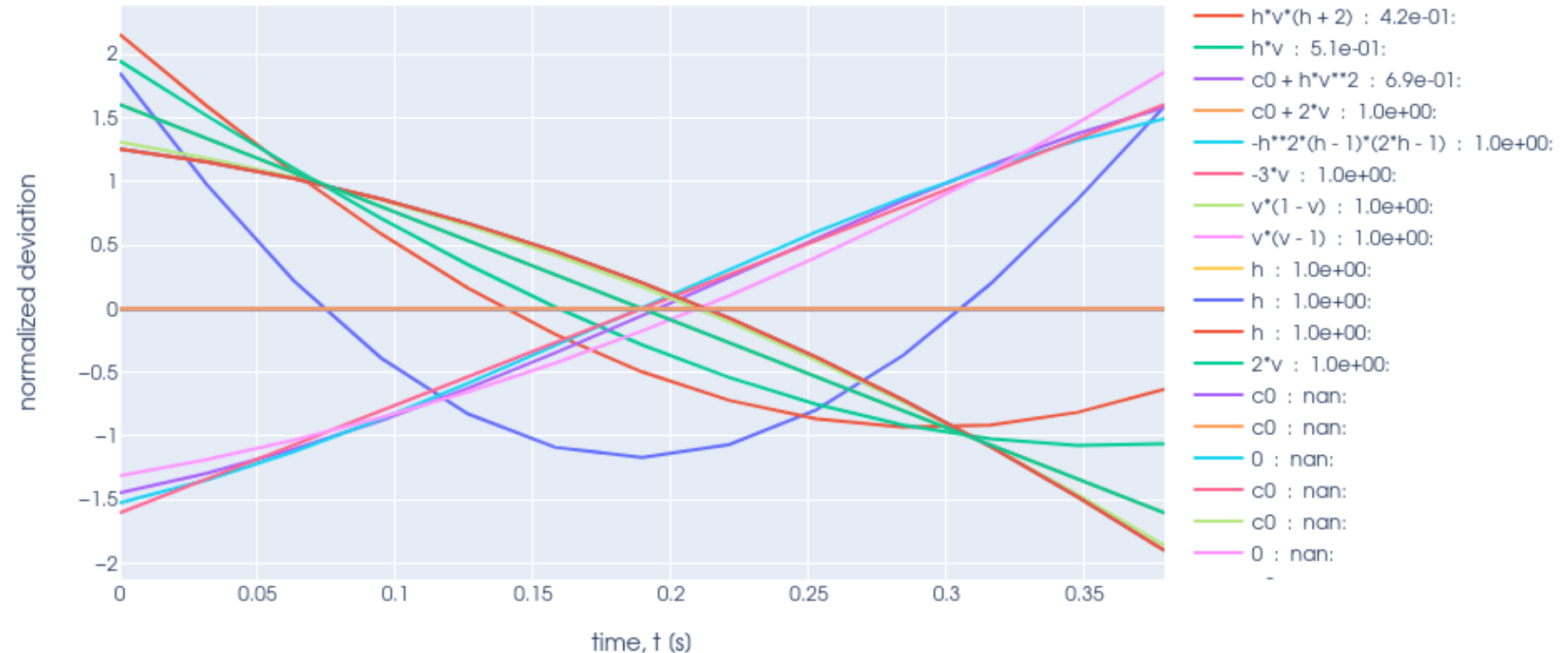
Initial Population

Inputs: Variable: h, v
Operators: $+, -, *$ Training data from previous slide

```

0: 0
1: c0
2: c0
3: 0
4: c0
5: h
6: h*v*(h**2 - 1)
7: h
8: 2*v
9: h*v
10: v*(v - 1)
11: c0
12: c0 + 2*v
13: 0
14: v*(1 - v)
15: c0
16: c0 + h*v**2
17: -3*v
18: c0
19: h*v*(h + 2)
20: 0
21: 0
22: -h**2*(h - 1)*(2*h - 1)
23: c0
24: h
    
```

Population_age: 0 Best: $h*v*(h**2 - 1) = -33.93795160580491$

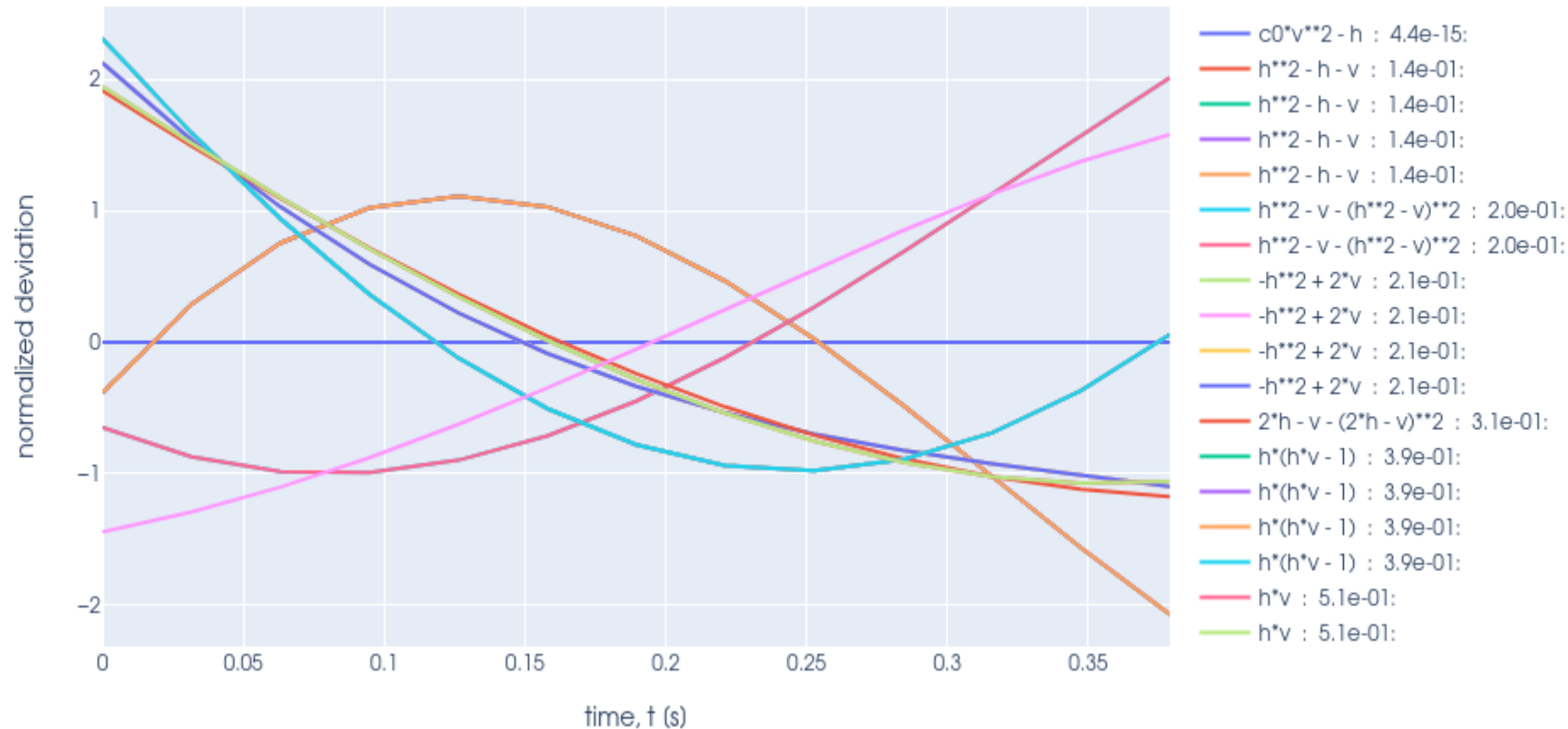


Population Evolution



Best: -h - 0.05102040816326488*v**2 = -2.999999999999997

Population_age: 152 Best: -h - 0.05102040816326488*v**2 = -2.999999999999997



$$h + 0.05102v^2 = 3$$

$$9.8h + \frac{1}{2}v^2 = 3(9.8)$$

$$mgh + \frac{1}{2}mv^2 = \text{constant} = 3mg$$