

Genetic Programming with DEAP

Lang Lovdog Inu Oókami

Genetic programming is a strategy for creating programs that solve a specific problem using operations as the chromosomes and the full set of operations as the individuals.

1 The demo using DEAP

Here are the requirements for our program, which mainly are DEAP, matplotlib and numpy.

```
!pip install -r requirements.txt &> /dev/null
```

Here the modules are imported

```
import operator
import random
from deap import algorithms, base, creator, tools, gp
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import rcParams
import seaborn as sns

# Setting matplotlib pyplot style to dark
plt.style.use('deepseek_lovdog_plasma')
# Set the dark purplish Plasma theme
sns.set_theme(
    style="darkgrid", # Dark background with grid
    context="notebook",
    font="sans-serif",
    font_scale=1.1,
    rc={
        # Core colors
        'axes.facecolor': '#1a0b2e', # Deep purple background
        'figure.facecolor': '#0d0518', # Even darker figure bg
        'grid.color': '#2a1a4a', # Grid lines
        'axes.edgecolor': '#6d3bb5', # Axis spines
        'xtick.color': '#b58ae6', # Tick colors
        'ytick.color': '#b58ae6',

        # Text
        'text.color': '#e0d6ff', # Off-white text
        'axes.labelcolor': '#c9b8ff',
        'axes.titlecolor': '#ffffff',

        # Lines
        'lines.color': '#9d65ff', # Vibrant purple
        'lines.linewidth': 2.5,

        # Color cycle (custom Plasma-like palette)
        'axes.prop_cycle': plt.cycler('color', [
```

```

        '#9d65ff', # Purple
        '#ff6eae', # Pink
        '#6eff87', # Mint
        '#ffb86c', # Orange
        '#6cebff'  # Cyan
    ])
}
)

# Optional: Improve font rendering
rcParams['font.family'] = 'sans-serif'
rcParams['font.sans-serif'] = ['Arial', 'DejaVuSans', 'Helvetica']

```

2 Target Function and error evaluation function

$$f(x) = x^2 + x \tag{1}$$

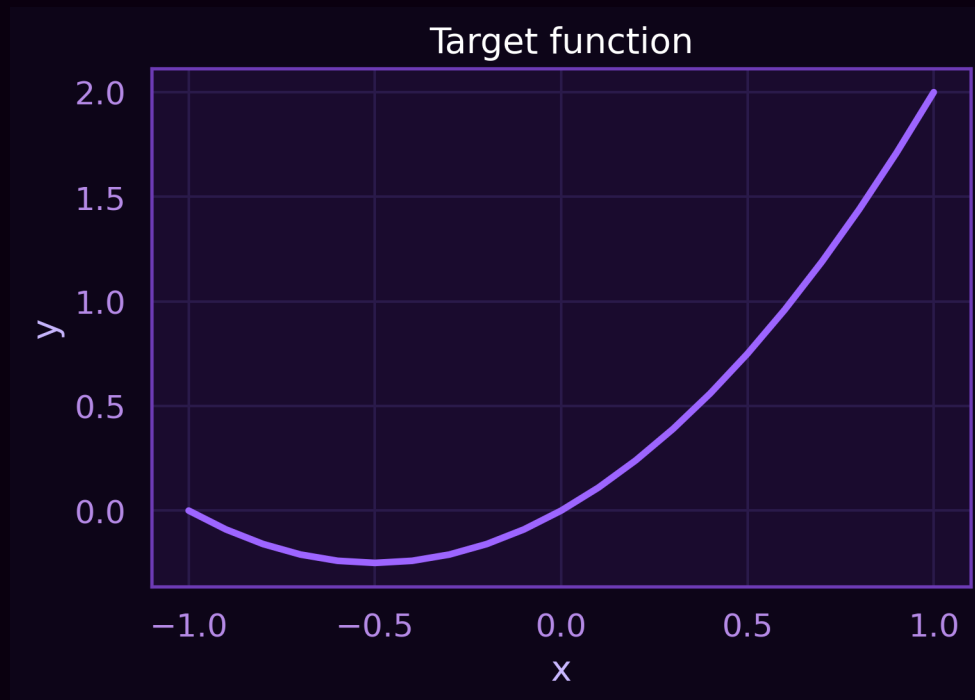
```

# 1. Define the target function we want to approximate (e.g., x^2 + x)
# And error function
def target_func(x):
    return x**2 + x

# 2. Define evaluation (mean squared error)
def evaluate(individual):
    func = gp.compile(individual, pset)
    points = [x/10.0 for x in range(-10, 11)] # Test points from -1 to 1
    error = sum((func(x) - target_func(x))**2 for x in points)
    return error,

x = [i/10.0 for i in range(-10, 11)]
plt.plot(x, [target_func(xi) for xi in x])
# legend and stuff
plt.xlabel("x")
plt.ylabel("y")
plt.title("Target_function")
plt.show()

```



3 Hyperparameters

In this part the primitives are created with add and multiplication operations with a unique variable.

```
# 3. Create primitive set (building blocks)
pset = gp.PrimitiveSet("MAIN", arity=1) # 1 input variable (x)
pset.addPrimitive(operator.add, 2, name="add") # Addition
pset.addPrimitive(operator.mul, 2, name="mul") # Multiplication
pset.addTerminal(1) # Constant 1
pset.renameArguments(ARG0="x") # Rename input to 'x'
```

Population options, minimum fitness and individual type

```
# 5. Set up fitness and individual types
creator.create("FitnessMin", base.Fitness, weights=(-1.0,)) # Minimize error
creator.create("Individual", gp.PrimitiveTree, fitness=creator.FitnessMin)
```

Procedure definitions

```
# 6. Initialize toolbox
toolbox = base.Toolbox()
toolbox.register("expr", gp.genHalfAndHalf, pset=pset, min_=1, max_=3)
toolbox.register("individual", tools.initIterate, creator.Individual, toolbox.expr)
toolbox.register("population", tools.initRepeat, list, toolbox.individual)
toolbox.register("evaluate", evaluate)
toolbox.register("select", tools.selTournament, tournsize=3)
toolbox.register("mate", gp.cxOnePoint)
toolbox.register("mutate", gp.mutUniform, expr=toolbox.expr, pset=pset)
```

4 Running the evolution simulation

```
# 7. Run evolution
population = toolbox.population(n=50)
hof = tools.HallOfFame(1)
stats = tools.Statistics(lambda ind: ind.fitness.values)
stats.register("avg", np.mean)
stats.register("min", np.min)

result, log = algorithms.eaSimple(
    population, toolbox, cxpb=0.7, mutpb=0.2, ngen=10,
    stats=stats, halloffame=hof, verbose=True
)
```

gen	nevals	avg	min
0	50	68.4375	1.52361
1	31	18.5112	1.73334e-32
2	39	11.189	0
3	37	7.32568	0
4	46	18.1242	0
5	43	8.26689	0
6	41	20.0019	0
7	41	17.6553	0
8	45	9.96535	0
9	34	74.0928	0
10	31	22.0356	0

5 Results of the program

```
# 8. Results
best_expr = str(hof[0]) # String representation of the best tree
print(f"\nBest expression: {best_expr}")

# Convert to readable math
readable_expr = best_expr.replace("add", "+").replace("mul", "*")
print(f"Human-readable: {readable_expr}")

# Plot results
y_true = [target_func(xi) for xi in x]
y_pred = [gp.compile(hof[0], pset)(xi) for xi in x]

## Interactive 3D plot separating target and evolved Function in Z axis
fig = plt.figure()
fig2 = plt.figure()
ax = fig.add_subplot(111, projection='3d')
# Plot functions
ax.plot(x, y_true, zs=0, zdir='z',
        color='#9d65ff', linewidth=3, label=f"Target: {x^2+{x}}")
ax.plot(x, y_pred, '--', zs=1, zdir='z',
        color='#ff6eae', linewidth=3, label=f"Evolved: {readable_expr}")

# Axis labels (smaller font, better padding)
ax.set_xlabel("x", labelpad=12, fontsize=10, color='#c9b8ff')
ax.set_ylabel("y", labelpad=12, fontsize=10, color='#c9b8ff')
ax.set_zlabel("function", labelpad=12, fontsize=10, color='#c9b8ff')

# Z-axis ticks (customized)
ax.set_zticks([0, 1])
ax.set_zticklabels(["Target", "Evolved"], fontsize=9, color='#e0d6ff')
```

```

# Legend (moved outside plot)
ax.legend(
    loc='upper_left',
    bbox_to_anchor=(1.05, 1), # Outside right
    facecolor='#2a1a4a',
    edgecolor='#6d3bb5',
    fontsize=10,
    labelcolor='linecolor' # Matches line colors
)

# Viewing angle and grid
ax.view_init(elev=25, azim=-45) # Better perspective
ax.grid(color='#2a1a4a', alpha=0.3, linewidth=0.5)

# Customize panes (transparent with colored edges)
ax.xaxis.pane.fill = False
ax.yaxis.pane.fill = False
ax.zaxis.pane.fill = False
ax.xaxis.pane.set_edgecolor('#2a1a4a')
ax.yaxis.pane.set_edgecolor('#2a1a4a')
ax.zaxis.pane.set_edgecolor('#2a1a4a')

#ax = fig.add_subplot(111, projection='3d')
#
#ax.plot(x, y_true, zs=0, zdir='z', label="Target:  $x^2 + x$ ")
#ax.plot(x, y_pred, '--', zs=1, zdir='z', label=f"Evolved: {readable_expr}")
#ax.set_xlabel("x")
#ax.set_ylabel("y")
#ax.set_zlabel("function")
### Set ticks in z [ Target, Evolved ]
#ax.set_zticks([0, 1])
#ax.set_zticklabels(["Target", "Evolved"])
#ax.legend()

## 2d plot

ax2 = fig2.add_subplot(111)
ax2.plot(x, y_true, label="Target:  $x^2 + x$ ")
ax2.plot(x, y_pred, "--", label=f"Evolved: {readable_expr}")
ax2.legend()
plt.show()

Best expression: add(mul(mul(x, 1), x), x)
Human-readable: +(*((x, 1), x), x)

```

