



**ICML**  
International Conference  
On Machine Learning

# DeltaEvolve: Accelerating Scientific Discovery through Momentum-Driven Evolution

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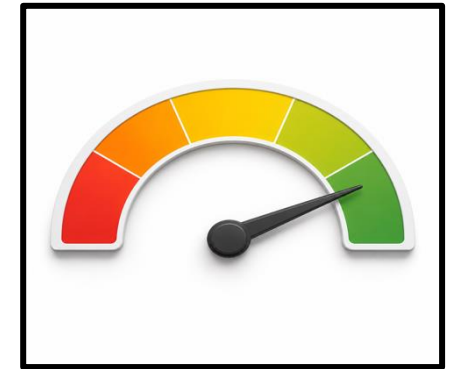
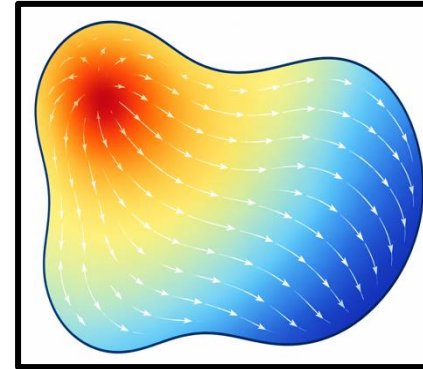
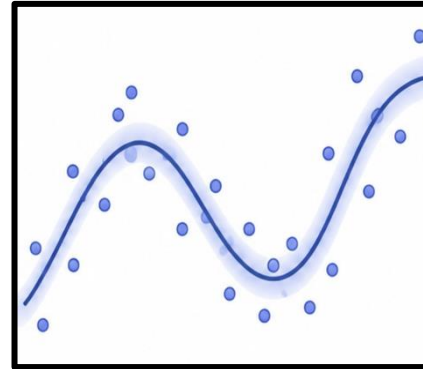
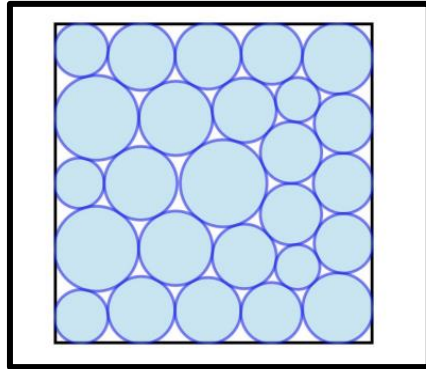
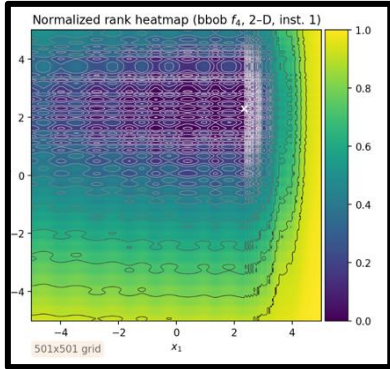
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# Open-Ended Discovery Tasks



## BlackBox Optimization

- Bbob\_f1
- Bbob\_f2
- ...
- Bbob\_f24

## Combinational Optimization

- Aircraft landing
- Bin Packing
- ...
- Travelling Salesman

## Symbolic Regression

- Population Growth
- Chemistry Reaction
- ...
- Physics

## PDE

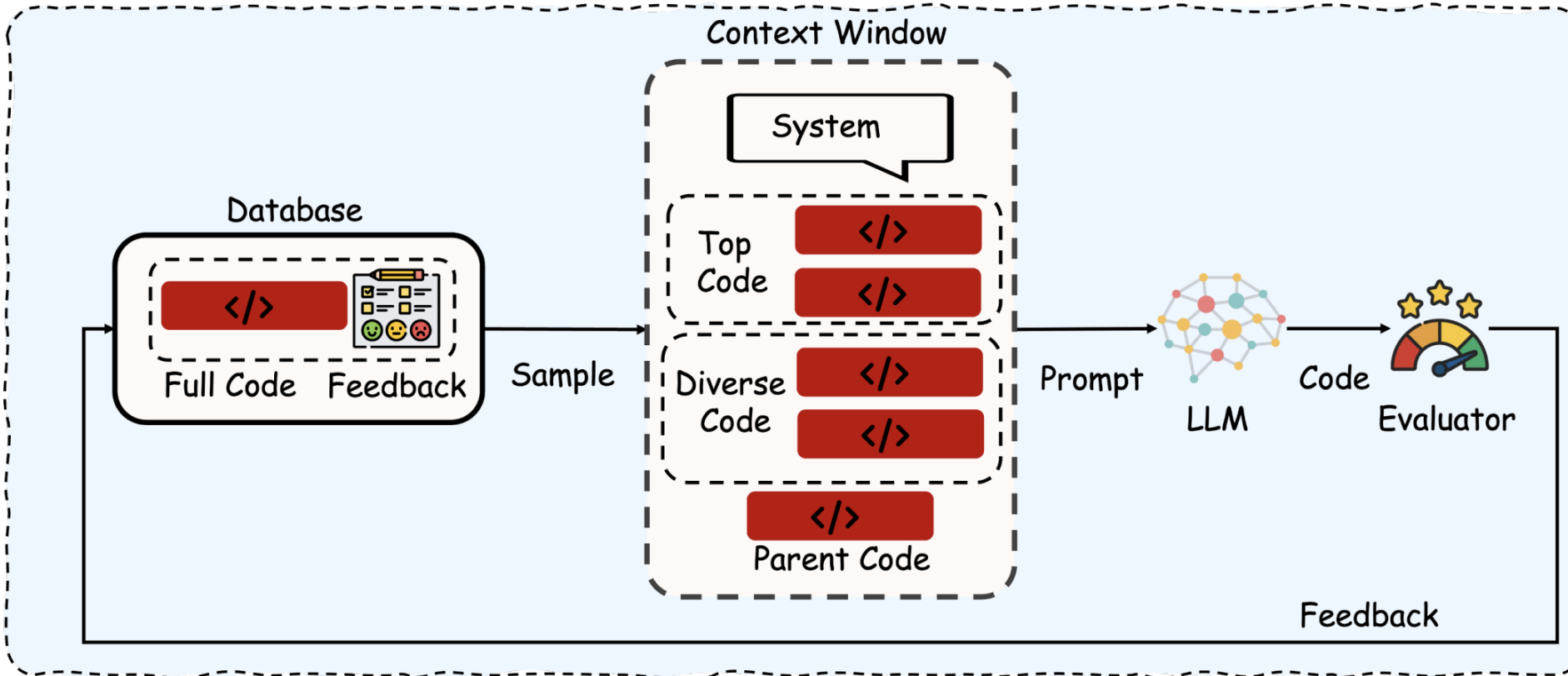
- Advection
- ...
- Burgers

## Algotune

- Convolve 2d
- Eigenvector
- ...
- Affine Transform

# AlphaEvolve

## AlphaEvolve



AlphaEvolve[1]

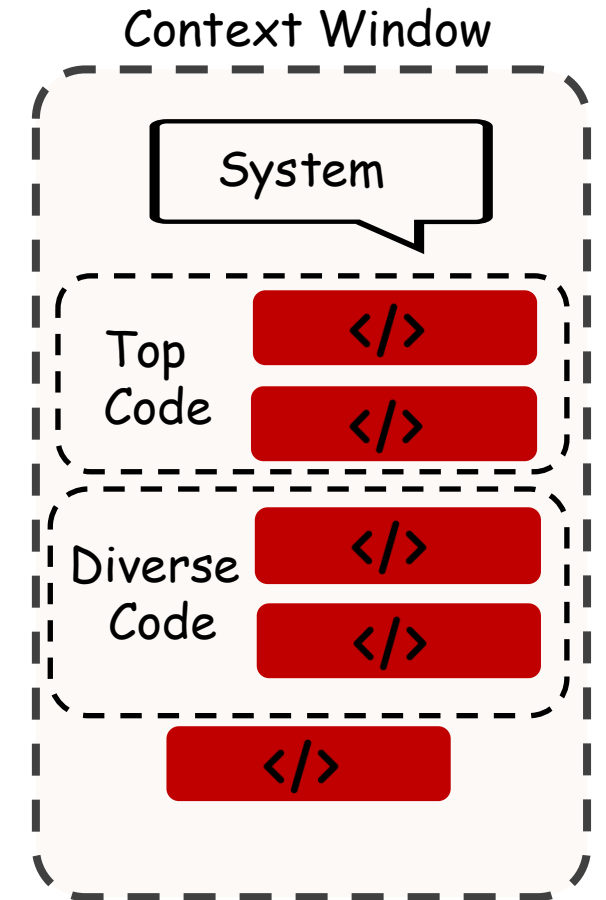
# Issue of Full Code

## 1. Limited context window

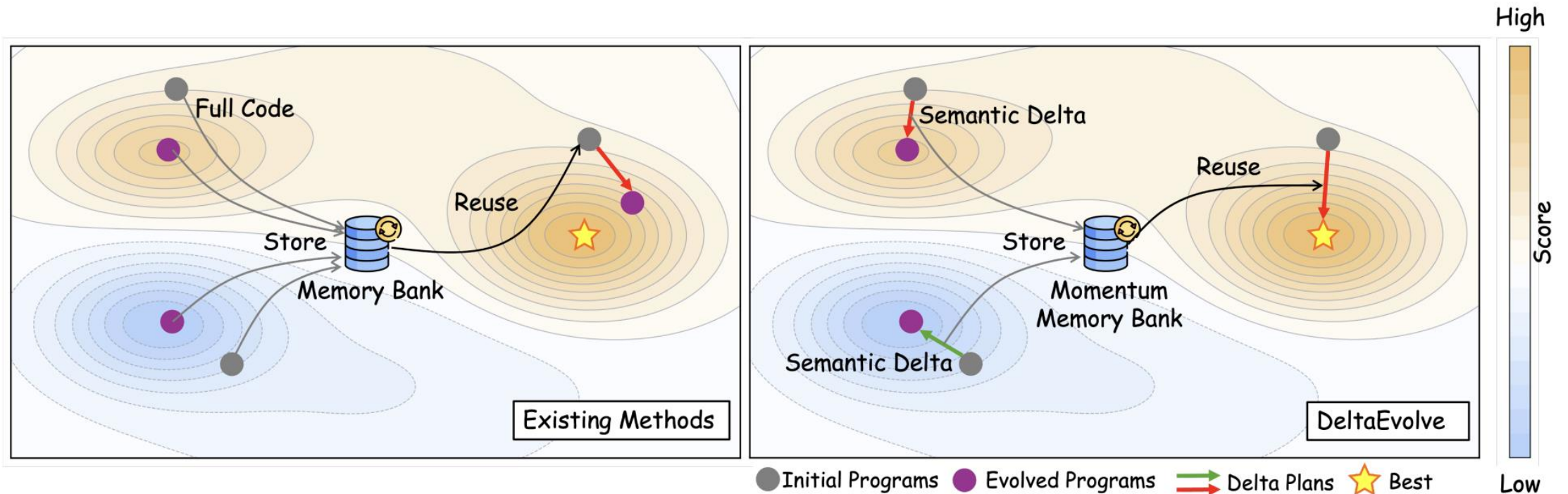
Programs are long, context is finite — only a handful survive, most history is discarded.

## 2. Insufficient evolutionary guidance

A full program buries the idea under detail. The model can't tell which change raised the score, or why.



# DeltaEvolve



## Semantic delta $\Delta$

A short, plain-language note of **what changed** from parent to child, and whether it **improved or degraded** the score.

## Momentum memory

Like momentum stacks differences between iterates, stacked deltas point in the **direction of improvement** — and transfer across programs.

# Progressive Disclosure Sampler

## Multi-Level Database

$$N_t = (\delta_t^{(1)}, \delta_t^{(2)}, p_t)$$

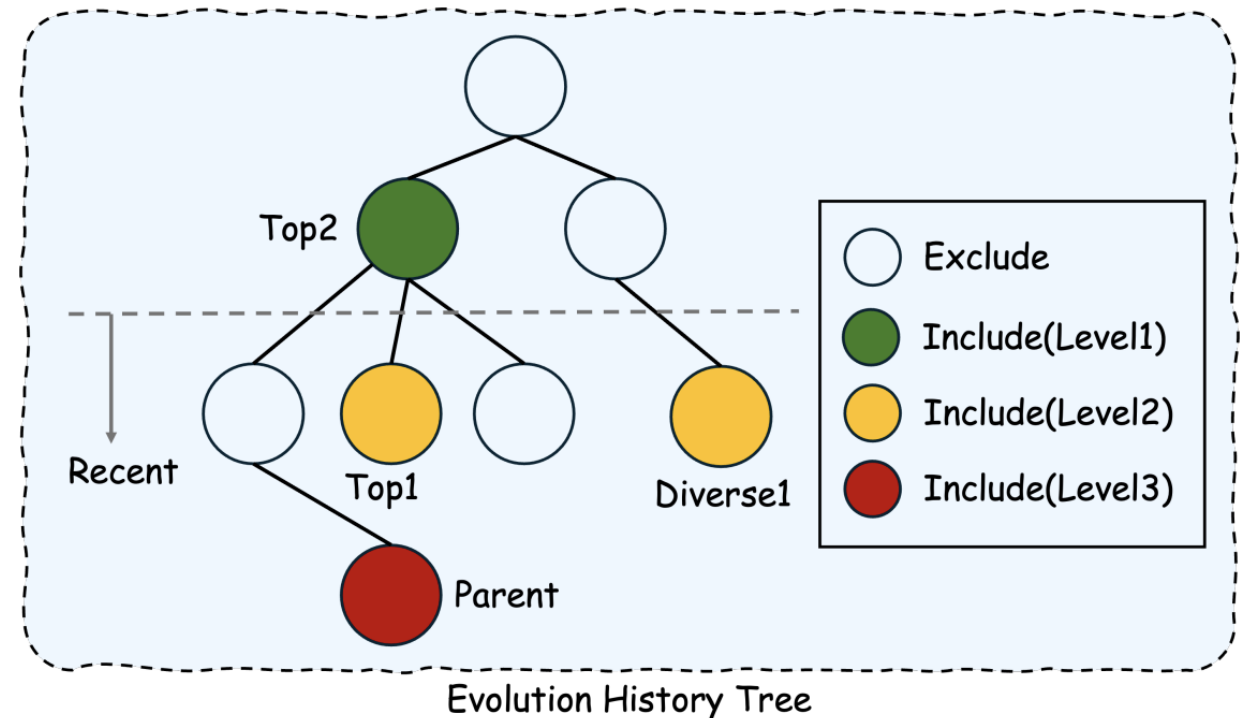
- **L1: Delta Summary**

$$\delta_t^{(1)} := \langle \text{FROM: } s_{t-1} \quad \text{TO: } s_t \rangle,$$

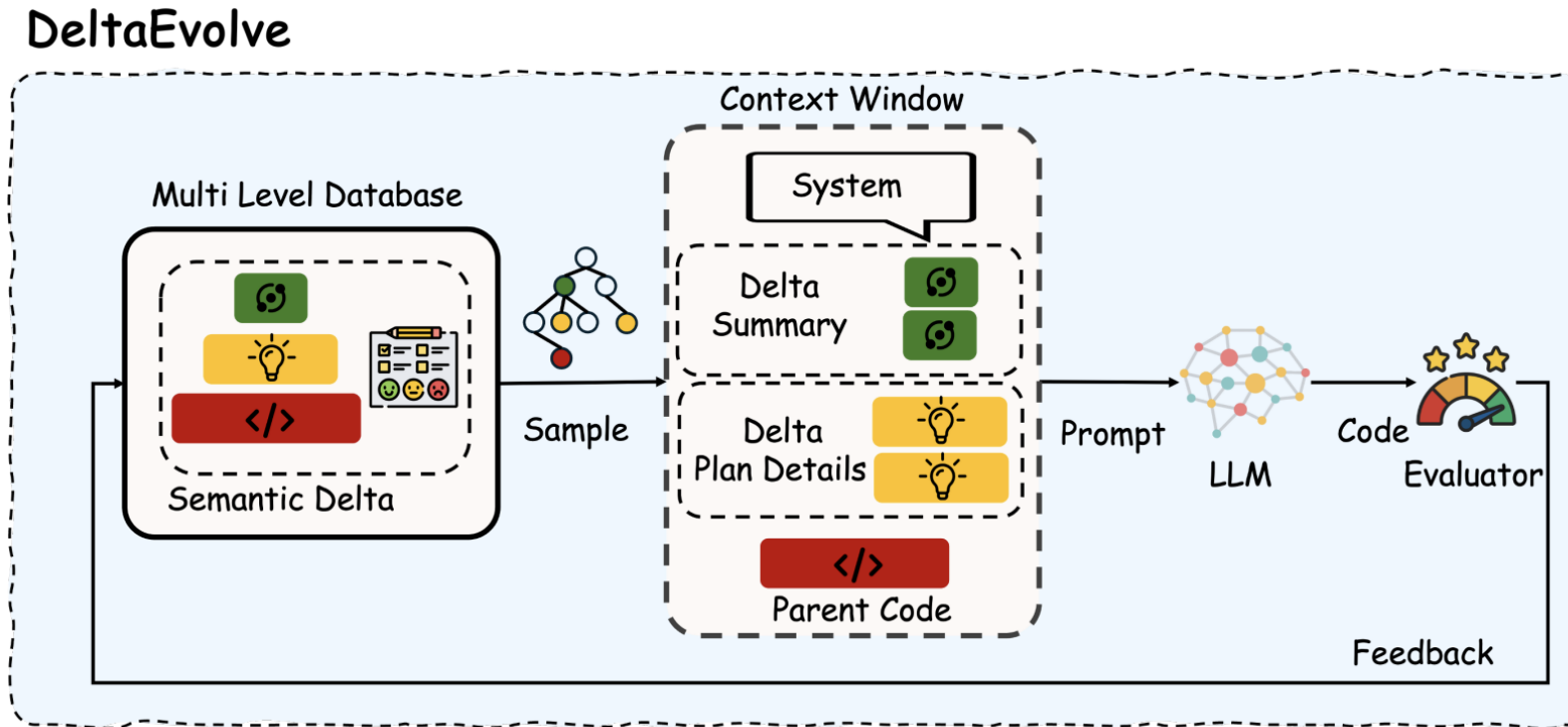
- **L2: Delta Plan Details**

$$\delta_t^{(2)} := \{(\ell_{t-1}^{(j)}, \ell_t^{(j)}, \text{hyp}_t^{(j)})\}_{j=1}^J,$$

- **L3: Full Code**



# DeltaEvolve



## Longer history, same budget

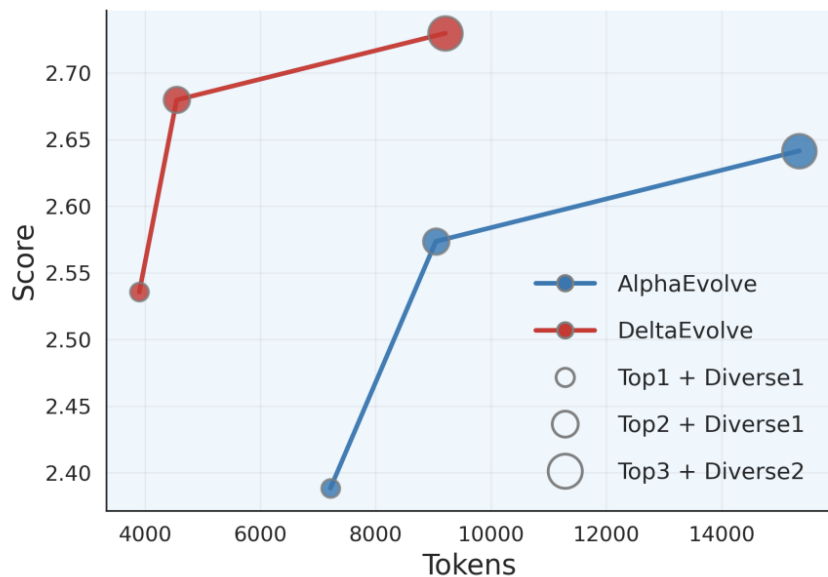
A delta is far shorter than a program, so much more history fits in context.

## Sharper signal

The model sees the direction of improvement directly, so the search is more focused.

# Experiments

- **RQ1 (Solution Quality):** *Can DeltaEvolve discover solutions with superior objective scores compared to state-of-the-art evolutionary baselines across diverse scientific domains?*
- **RQ2 (Token Consumption):** *Does DeltaEvolve reduce the computational cost (input tokens) required to discovery high-quality solutions?*



Task Domain	Method	GPT-5-mini + o3-mini		Gemini-2.5-flash-lite + Gemini-2.5-flash	
		Best Score(↑)	Token Consump.(↓)	Best Score(↑)	Token Consump.(↓)
1. BlackBox Optimization	Parallel Sampling	0.4161	390872	0.4161	390872
	Greedy Refine	2.2618	555430	2.3403	1054566
	AlphaEvolve (Full Code)	<u>2.6415</u>	1852841	<u>2.5221</u>	1894890
	<b>DeltaEvolve (Ours)</b>	<b>2.7297</b>	1390709	<b>3.9372</b>	1227388
2. Hexagon Packing	Parallel Sampling	0.4913	565016	0.4913	565016
	Greedy Refine	0.8508	717697	0.4913	814092
	AlphaEvolve (Full Code)	<u>0.9721</u>	905249	<u>0.7859</u>	1334339
	<b>DeltaEvolve (Ours)</b>	<b>0.9821</b>	827884	<b>0.8804</b>	893749
3. Symbolic Regression	Parallel Sampling	1.8527	503098	1.8535	503098
	Greedy Refine	2.9553	811404	2.9550	756762
	AlphaEvolve (Full Code)	<u>3.2657</u>	1660699	<u>3.2174</u>	1545091
	<b>DeltaEvolve (Ours)</b>	<b>3.4174</b>	810354	<b>3.2198</b>	832164
4. PDE Solver	Parallel Sampling	0.7506	154016	0.7506	154016
	Greedy Refine	0.7506	375186	0.7506	255332
	AlphaEvolve (Full Code)	<u>0.8850</u>	711298	<u>0.9901</u>	595094
	<b>DeltaEvolve (Ours)</b>	<b>0.8915</b>	562848	<b>0.9931</b>	253719
5. Efficient Convolution	Parallel Sampling	0.8035	139958	0.8036	139958
	Greedy Refine	0.8035	240473	0.8037	305334
	AlphaEvolve (Full Code)	<u>0.8974</u>	683439	<u>0.8219</u>	885592
	<b>DeltaEvolve (Ours)</b>	<b>0.9067</b>	348539	<b>0.9032</b>	517281

# Acknowledge & Thanks

We acknowledge support from NSF grants IIS-2312840 and IIS-2402952. We gratefully acknowledge Ismail Alkhouri, Yuxin Dong, Xia Ning, and Huan Sun for valuable discussions.

**Thanks for Listening !**

