## FunSearch & AlphaEvolve

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### **Motivation**

#### Preconditions:

- A problem of finding optimal heuristic/program.
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- An automated evaluator returning a scalar score.

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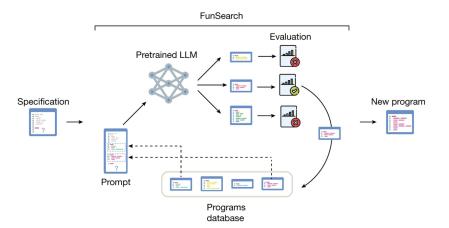
But instead of sampling independently, can we **incorporate the evaluator feedback** into subsequent generations?

### **FunSearch**

### Key ingredients:

- best-shot prompting,
- a growing database of programs,
- an evolutionary strategy acting on it.

### **FunSearch**

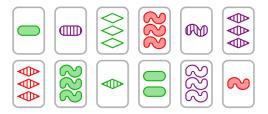


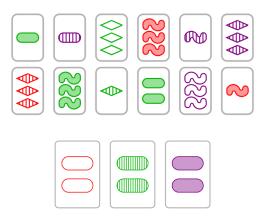
## **Database of programs**

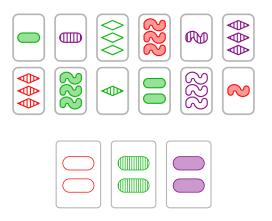
- Several island/subpopulations growing independently.
- Higher-scoring programs, but also shorter ones, are prioritized.

## **Best-shot prompting**

- *k* good programs per prompt sampled, for each island.
- Information which one is better incorporated into the prompt.







n	3	4	5	6	7	8
Best known	9	20	45	112	236	496
FunSearch	9	20	45	112	236	512

## Cap set solution template

```
"""Finds large cap sets."""
import numpy as np
import utils capset
def main(n):
  """Runs `solve` on `n`-dimensional cap set and

→ evaluates the output."""

 solution = solve(n)
 return evaluate (solution, n)
def evaluate(candidate set, n):
  """Returns size of candidate set if it is a cap

→ set, None otherwise, """

 if utils capset is capset (candidate set, n):
   return len(candidate set)
 else:
   return None
def solve(n) .
  """Builds a cap set of dimension `n` using

→ `priority` function."""
 # Precompute all priority scores.
 elements = utils capset.get all elements(n)
 scores = [priority(el, n) for el in elements]
 # Sort elements according to the scores.
 elements = elements[np.argsort(scores,

        kind='stable')[::-1]]
 # Build `capset` greedily, using scores for
 → prioritization.
 capset = []
 for element in elements:
   if utils capset.can be added(element, capset):
     capset.append(element)
 return capset
# Function to be evolved by FunSearch.
def priority(element, n):
  """Returns the priority with which we want to add
 → `element` to the cap set."""
 return 0.0
```

# Cap set solution (priority function)

```
def priority(el: tuple[int,...],

    n: int) -> float:
 score = n
  in el = 0
  el count = el.count(0)
  if el count == 0:
   score += n**2
   if el[1] == el[-1]:
    score *= 1.5
   if el[2] == el[-2]:
    score *= 1.5
   if el[3] == el[-3]:
     score *= 1.5
  else:
   if el[1] == el[-1]:
    score *= 0.5
   if el[2] == el[-2]:
     score *= 0.5
  for e in el:
   if e == 0:
     if in el == 0:
       score *= n * 0.5
     elif in el == el count - 1:
       score *= 0.5
     else:
       score *= n * 0.5 ** in el
```

## **AlphaEvolve**

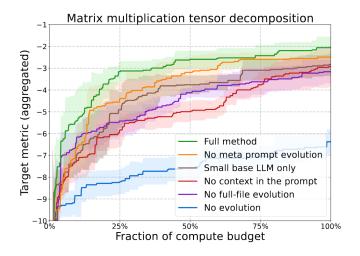
### tldr: FunSearch scaled-up in multiple dimensions.

FunSearch [83]	AlphaEvolve
evolves single function	evolves entire code file
evolves up to 10-20 lines of code	evolves up to hundreds of lines of code
evolves code in Python	evolves any language
needs fast evaluation (≤ 20min on 1 CPU)	can evaluate for hours, in parallel, on accelerators
millions of LLM samples used	thousands of LLM samples suffice
small LLMs used; no benefit from larger	benefits from SOTA LLMs
minimal context (only previous solutions)	rich context and feedback in prompts
optimizes single metric	can simultaneously optimize multiple metrics

## **Optimizong matrix multiplications**

$\langle m,n,p\rangle$	best known [reference]	AlphaEvolve
$\langle 2, 4, 5 \rangle$	33 [42]	32
$\langle 2, 4, 7 \rangle$	46 [93]	45
$\langle 2, 4, 8 \rangle$	52 [93]	51
$\langle 2, 5, 6 \rangle$	48 [93]	47
$\langle 3, 3, 3 \rangle$	23 [52]	23
$\langle 3, 4, 6 \rangle$	56 [48]	54
$\langle 3, 4, 7 \rangle$	66 [91]	63
$\langle 3, 4, 8 \rangle$	75 [91]	74
$\langle 3, 5, 6 \rangle$	70 [48]	68
$\langle 3, 5, 7 \rangle$	82 [91]	80
$\langle 4, 4, 4 \rangle$	49 [95]	48
$\langle 4, 4, 5 \rangle$	62 [47]	61
$\langle 4, 4, 7 \rangle$	87 [93]	85
$\langle 4, 4, 8 \rangle$	98 [95]	96
(4, 5, 6)	93 [48]	90
$\langle 5, 5, 5 \rangle$	93 [72]	93

### **Ablations**



# **Closing remarks**

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