Multi-paradigm programming in Python

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Previse (https://previ.se/)

Quick Intro

- Elias Mistler
- Previse
 - Invoice financing
 - based on ML
 - corporate data
 - improve SME cashflow
- Machine Learning Engineer
 - ML integration into invoice processing platform
 - Buyer data intake and mapping
 - Operational tooling

Contents

- Introduction
- Code Structure
- Data Structures
- State Handling
- Multiple implementations
- Summary

Introduction

- Python = multi-paradigm (unlike OO Java / FP Clojure)
- OOP and FP are **concepts**, not tied to syntax (class or def)

Object-oriented principles

- mutable data structures
- (relies on rich type system)
- class hierarchies
 - inheritance
 - abstraction
 - encapsulation
 - polymorphism

Functional programming Principles

- immutable data structures
- (relies on simple data types)
- pure functions
 - no side-effects
 - idempotent
- (often lazy evaluation)

Sudoku

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
8 4 7			8		3			1
7				2				6
	6					2	8	
			4	1	9			5 9
				8			7	9

- 9 x 9 grid
- numbers from 1 9
- each row/column/block should contain each digit

Code structure: high- vs. low-context

Example: parse raw Sudoku string (from OpenSudoku (https://opensudoku.moire.org/) to array

```
In [2]: raw_example = '7001500000030020978004701265003902000300100500080270019750310041207
00900000065002'
```

Factory function (00)

```
In [3]:
        @dataclass
         class Sudoku:
             grid: np.array
             @classmethod
             def from string(cls, raw):
                 values = []
                 for digit in raw:
                     values.append(int(digit))
                 grid = np.array(values, dtype='int64').reshape((9, 9))
                 return cls(grid)
In [4]:
        Sudoku.from string(raw example)
         Sudoku(grid=array([[7, 0, 0, 1, 5, 0, 0, 0],
Out[4]:
                [0, 0, 3, 0, 0, 2, 0, 9, 7],
                [8, 0, 0, 4, 7, 0, 1, 2, 6],
                [5, 0, 0, 3, 9, 0, 2, 0, 0],
                [0, 3, 0, 0, 1, 0, 0, 5, 0],
                [0, 0, 8, 0, 2, 7, 0, 0, 1],
                [9, 7, 5, 0, 3, 1, 0, 0, 4],
                [1, 2, 0, 7, 0, 0, 9, 0, 0],
                [0, 0, 0, 0, 6, 5, 0, 0, 2]]))
```

- explicit, high-context
- easy to find and use

Isolated function (FP)

- free of assumptions about the use case
- easy to reuse or generalise

Multi-paradigm solution

Generalised, low-context pure function, use in high-context class

```
In [7]: def parse_raw(raw):
    size = int(math.sqrt(len(raw)))
    return np.array(list(map(int, raw)), dtype='int64').reshape((size, size))

class Sudoku:
    @classmethod
    def from_string(cls, raw):
        values = parse_raw(raw)
        return cls(values)
```

- low-context pure functions and high-context class
- tidy, reusable code
- generalises well
- works in any context
- easy to use and explore

That tedious for-loop

- easy to write
- tedious to read and reconstruct
- comparatively far from high-level intention
- error prone

"I WOULD HAVE
WRITTEN A
SHORTER
LETTER, BUT I
DID NOT HAVE
THE TIME."

Blaise Pascal

The alternative:

```
In [9]: values = tuple(map(int, raw_example))
    values[:5]
Out[9]: (7, 0, 0, 1, 5)
In [10]: values = thread_last(raw_example, (map, int), tuple)
```

- concise
- reflects the intention
- easy to read
- can take longer to write

Also useful: list comprehension

```
In [11]: values = [int(digit) for digit in raw_example]
```

- Pythonic middle ground
- easy to both read and write
- do not use *lambda functions* in list comprehensions!
- combine with pure functions for best results

Further example - display/format

Object-oriented

Implement __repr__

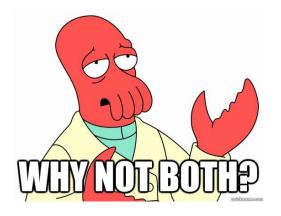
```
In [12]:
         from sudoku.oo.base import Sudoku
          Sudoku.from_string(raw_example)
Out[12]:
```

Functional

explicit functions

```
from sudoku.fp.load import format_sudoku
In [13]:
         thread_last(raw_example, parse_raw, format_sudoku, print)
```

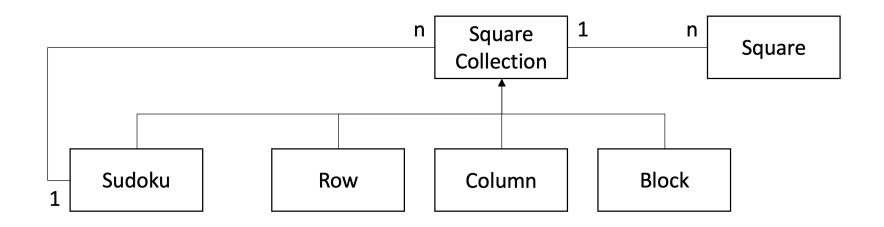
Multi-paradigm



Data Structures: explicit vs. minimalist

Example: The Sudoku grid

Class Hierarchy (00)



- assumes certain usage patterns
- intuitive to explore
- fairly rigid
- requires lots of boilerplate

```
In [17]: # even with `dataclass` and without many getters, setters etc:
!wc ./sudoku/oo/base.py
```

121 335 3148 ./sudoku/oo/base.py

Simplicity (FP)

```
In [18]:
         import schema
          sudoku schema = schema.And(np.ndarray,
                                     lambda a: a.shape == (9, 9),
                                     lambda a: a.dtype == 'int64')
In [19]:
         thread last(raw example, parse raw, sudoku schema.validate)
          array([[7, 0, 0, 1, 5, 0, 0, 0, 0],
Out[191:
                 [0, 0, 3, 0, 0, 2, 0, 9, 7],
                 [8, 0, 0, 4, 7, 0, 1, 2, 6],
                 [5, 0, 0, 3, 9, 0, 2, 0, 0],
                 [0, 3, 0, 0, 1, 0, 0, 5, 0],
                 [0, 0, 8, 0, 2, 7, 0, 0, 1],
                 [9, 7, 5, 0, 3, 1, 0, 0, 4],
                 [1, 2, 0, 7, 0, 0, 9, 0, 0],
                 [0, 0, 0, 0, 6, 5, 0, 0, 2]])
```

- minimalist approach with basic data types
- zero boilerplate
- no context on the data structure itself
- harder to explore
- easier to reuse

Multi-paradigm solution

```
In [20]: @dataclass
class Sudoku:
    grid: np.ndarray

    @property
    def remaining_blanks(self):
        return (self.grid == 0).sum()

    def __repr__(self):
        ...
```

- "shallow" class
- saves a lot of boilerplate code
- adds context for user

State handling - mutable vs. immutable

Example: Fill digits into Sudoku

Using a multi-paradigm implementation, inspired by pandas:

Mutable (00)

```
In [22]:
          sudoku.set_digit(0, 0, 7, inplace=True)
          sudoku
Out[22]:
```

- changed in-place
- seems "natural"
- no way back / history

Immutable (FP)

```
In [23]:
         sudoku.set_digit(2, 2, 4, inplace=False)
Out[23]:
```

In [24]: sudoku Out[24]:

- easy to reuse or parallelise (efficienct, avoids concurrecny errors)
- natural versioning
- lends itself well to pipelines or method chaining

Method Chaining

```
In [25]:
          (sudoku
           .set_digit(2, 8, 9)
           .set_digit(1, 0, 9)
           .set_digit(0, 3, 9))
Out[25]:
```

Recommendation

- make use of immutable data structures like @dataclass(frozen=True),
 NamedTuple, frozendict and pyrsistent.pmap)
- use mutable data structures in immutable ways (try the toolz library!)
- keep functions pure and idempotent use classes where configuration and state is required

Example in Pandas

0 0.902117 0.616821 0.341237 1.860174

 2
 0.062307
 0.200615
 0.328240
 0.591162
 0.105397

 4
 0.423682
 0.935371
 0.089573
 1.448627
 0.292472

```
In [26]:
           import pandas as pd
           df = pd.DataFrame(np.random.random((5,3)), columns=list('abc'))
           df
Out[26]:
            0 0.902117 0.616821 0.341237
            1 0.474177 0.159596 0.566260
            2 0.062307 0.200615 0.328240
            3 0.521419 0.147103 0.210795
            4 0.423682 0.935371 0.089573
In [27]:
           (df
            .assign(sum=lambda df: df.sum(axis=1))
            .assign(a percent=lambda df: df['a'] / df['sum'])
            .drop(index=[1,3]))
Out[27]:
                                        sum a_percent
```

0.484964

```
In [28]: d
```

Out[28]:

	а	b	С
0	0.902117	0.616821	0.341237
1	0.474177	0.159596	0.566260
2	0.062307	0.200615	0.328240
3	0.521419	0.147103	0.210795
4	0.423682	0.935371	0.089573

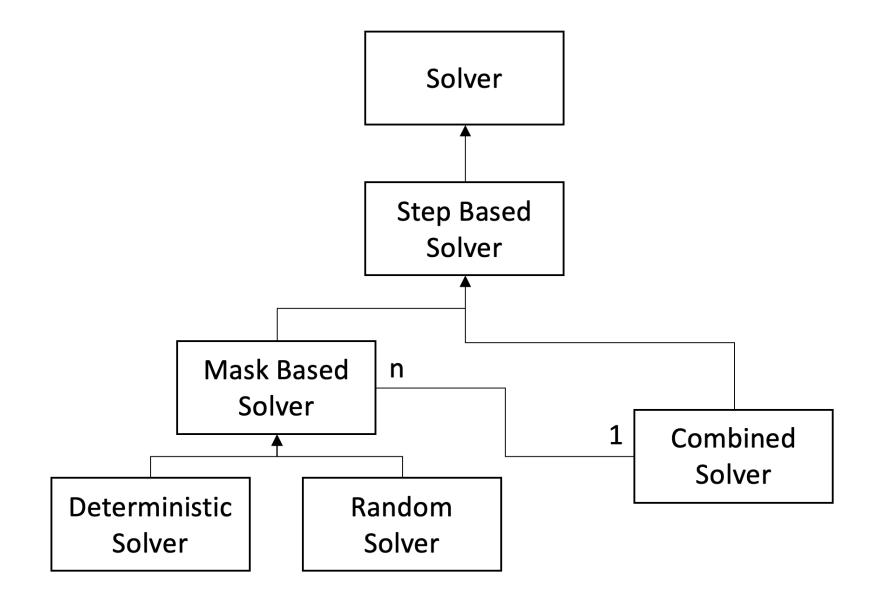
- cleaner Jupyter notebooks (execution order...)
- better reusability
- close to production-ready

Multiple implementations: polymorphism vs. function composition

Example: Different Sudoku solvers

- Deterministic (mask, fill unambiguous, repeat) insufficient
- Random (mask, fill random, repeat) prohibitively slow
- Combined (deterministic as much as possible, random step, repeat)

00 - Solver class hierarchy



```
In [29]: from sudoku.oo.solver import *
    sudoku = Sudoku.from_string(raw_example)
    solver = DeterministicSolver(sudoku)
    solver.solve()
    sudoku
```

Out[29]:

```
        | 7 | 6 | 2 | 1 | 5 | 9 | 4 | 8 | 3 |

        | 4 | 1 | 3 | 6 | 8 | 2 | 5 | 9 | 7 |

        | 4 | 1 | 3 | 6 | 8 | 2 | 5 | 9 | 7 |

        | 8 | 5 | 9 | 4 | 7 | 3 | 1 | 2 | 6 |

        | 5 | 4 | 1 | 3 | 9 | 6 | 2 | 7 | 8 |

        | 2 | 3 | 7 | 8 | 1 | 4 | 6 | 5 | 9 |

        | 6 | 9 | 8 | 5 | 2 | 7 | 3 | 4 | 1 |

        | 7 | 5 | 2 | 3 | 1 | 8 | 6 | 4 |

        | 1 | 2 | 6 | 7 | 4 | 8 | 9 | 3 | 5 |

        | 3 | 8 | 4 | 9 | 6 | 5 | 7 | 1 | 2 |

        | 3 | 8 | 4 | 9 | 6 | 5 | 7 | 1 | 2 |

        | 4 | 1 | 2 | 6 | 7 | 4 | 8 | 9 | 3 | 5 |

        | 3 | 8 | 4 | 9 | 6 | 5 | 7 | 1 | 2 |

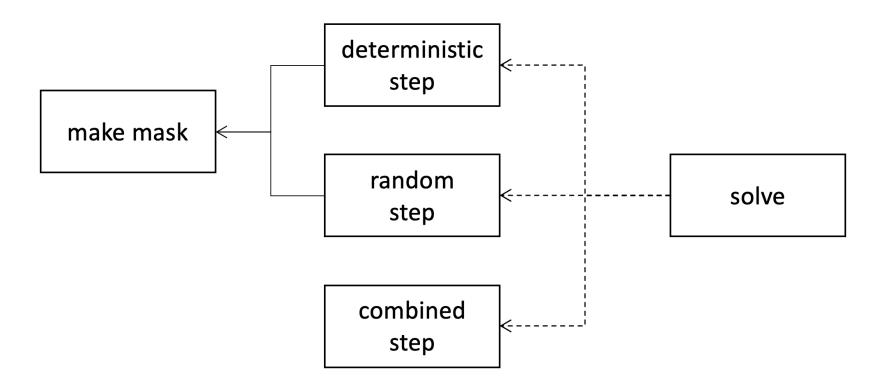
        | 3 | 8 | 4 | 9 | 6 | 5 | 7 | 1 | 2 |
```

- Mutable data access (as before)
- Single-method classes excessive (boilerplate!)
- Complicated design for simple functionality

In [30]: !wc ./sudoku/oo/solver.py

123 297 3852 ./sudoku/oo/solver.py

FP - solving function composition



```
In [31]: from sudoku.fp.solve import *
    from sudoku.fp.load import *
    solve_combined = partial(solve, step_function=combined_step)
    thread_last(raw_example, parse_raw, solve_combined, format_sudoku, print)
```

- very clear responsibilities per function
- simple, pragmatic design
- easy to introspect
- much more concise (and no base module!)

Multi-paradigm solution

- simplicity and clarity of FP
- takes and returns high-context Sudoku objects

Or, with more context:

```
In [33]: from sudoku.mp.solve import solve

@dataclass(frozen=True)
class Solver:
    step_function: Callable
    max_tries: int = 1

def __call__(self, sudoku: Sudoku):
    return solve(sudoku, self.step_function, self.max_tries)
```

Out[34]:

Key Takeaways

Object-orientation

- "top-down" design
- larger, topical structures
- explicit, high-context
- functionality and data intertwined

leads to:

- intuitive use cases
- high explorability

Functional programming

- "bottom-up" design
- simplistic thinking
- small chunks of reusable logic, separate from data
- high isolation, low context

leads to

- high reusability
- tidy, concise code
- flexible use cases

Multi-paradigm programming

pick & mix of both worlds:

- pure functions in mutable context
 - brings the simplicity and elegance of FP into OO
 - make your code explorable and easy to understand
 - remember: no side effects, no problem!
- mutable data in immutable context
 - use your favourite OO libaries in concise FP code
 - remember: copy-and-modify mutable data structures!

leads to (ideally) - best of both worlds:

- intuitive and flexible use cases
- high explorability **and** reusability

My preferred Approach

- iterate with a REPL
- use immutable data types and pure functions where possible
- create classes where either:
 - required due to syntax or library
 - high-context use cases are required

Thank you for your attention!

References & Further Reading

Full notebook and code available at https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku (https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku (https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku (https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku (https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku (https://github.com/eliasmistler/europython2020-multi-paradigm-sudoku)