

Search Algorithms: Object-Oriented Implementation (Part C)

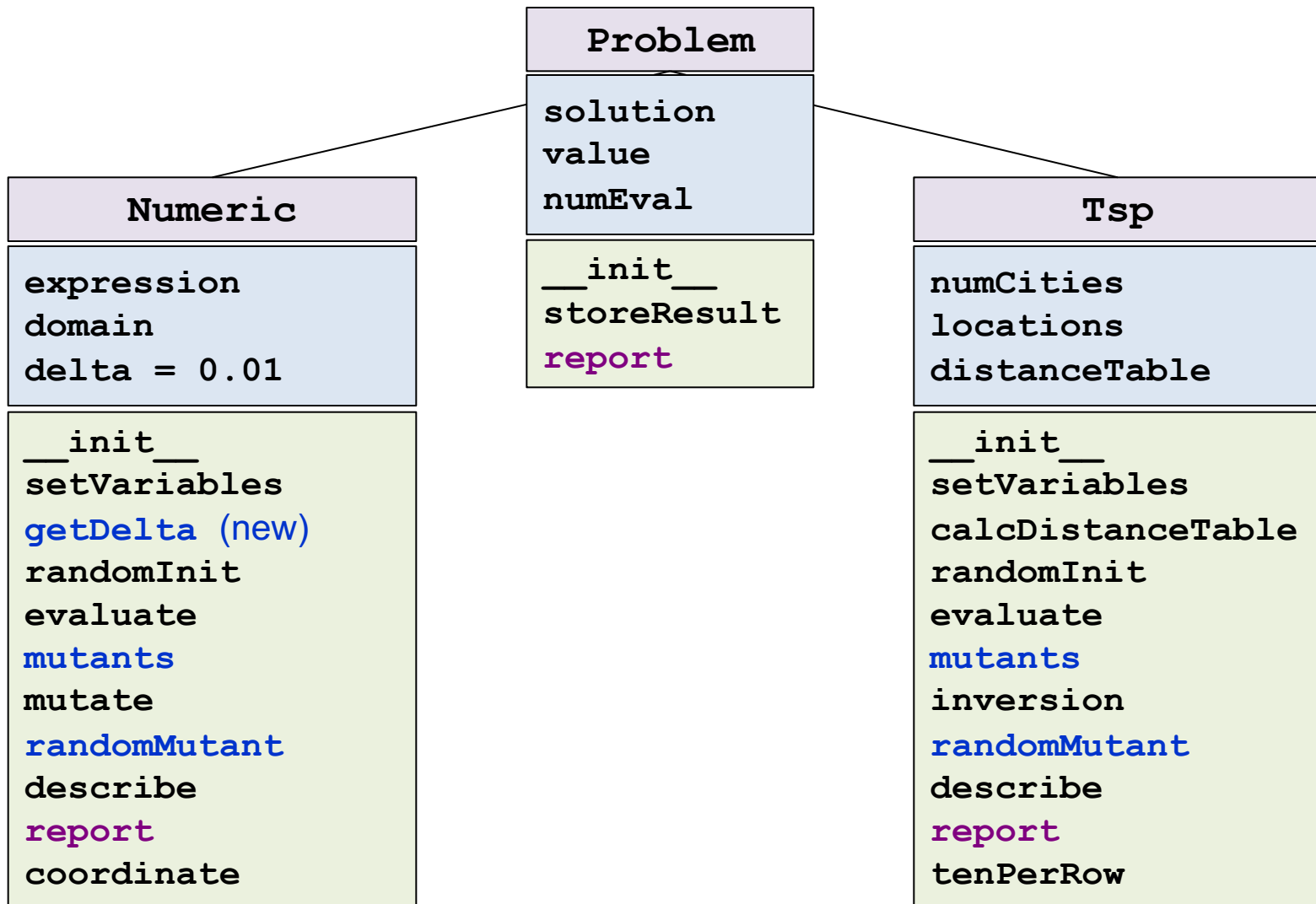
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Defining 'Problem' Class

- We define a hierarchy of classes whose base class is named as **Problem**
 - Information about the problem to be solved can be stored in an organized fashion
 - Codes of the same purposes but with different problem-specific implementations can be organized nicely under this class hierarchy
 - The result of search can also be handled nicely depending on the type of problem solved
- **Problem** has two subclasses **Numeric** and **Tsp**
 - Functions in the 'numeric' module become methods of **Numeric**
 - Similarly, those in the 'tsp' module become methods of **Tsp**

Defining 'Problem' Class



Defining 'Problem' Class

- Some functions are renamed after migration:
 - `createProblem` → `setVariables`
 - `describeProblem` → `describe`
 - `displayResult` → `report`
- The subclass `Numeric` has three variables for storing specifics about the problem:
 - `expression`: function expression of a numerical optimization problem
 - `domain`: lower and upper bounds of each variable of the function
 - `delta`: step size of axis-parallel mutation

Defining 'Problem' Class

- `delta` takes the role of `DELTA` that was previously the named constant of the 'numeric' module
 - `delta` is referred to by `mutants` and `randomMutant`
 - Its value is preset to 0.01 for the time being for convenience
 - The method `getDelta` is newly made for being used by the `displaySetting` function of the main program when displaying the value of `delta`
- The subclass `Tsp` also has three variables:
 - `numCities`: number of cities
 - `locations`: coordinates of city locations in a 100×100 square
 - `distanceTable`: matrix of distances of every pair of cities

Defining 'Problem' Class

- Both `Numeric` and `Tsp` have the `setVariables` method that reads in the specifics of the problem to be solved and stores them in the relevant class variables
 - It is renamed from its previous version `createProblem`
 - Unlike `createProblem`, `setVariables` does not return anything
- Notice that the functions `mutants` and `randomMutant` can also be converted to methods of both `Numeric` and `Tsp` classes just like `mutate` or `inversion`
 - `mutants` appears in both SAHC-N and SAHC-T
 - `randomMutant` appears in both FCHC-N and FCHC-T
- This makes the main programs simpler and thus easier for later unification

Defining 'Problem' Class

- The base class `Problem` has three variables for storing the result of search:
 - `solution`: final solution found by the search algorithm
 - `value`: its objective value
 - `numEval`: total number of evaluations taken for the search

However, the way to report the result is different depending on the type of problem solved

- The values of the two variables `solution` and `value` are set by the `storeResult` method that is called by the search algorithms
 - Now the search algorithms do not have the `return` statement

Defining 'Problem' Class

- The value of `numEval` is initially 0 and incremented whenever the `evaluate` method is called
 - Previously, reporting the value of the global variable `NumEval` was done by `displayResult` of both 'numeric' and 'tsp' modules
 - Recall that `NumEval` appeared in duplicate in both modules
 - Now, printing `numEval` is done by the `report` method

Defining 'Problem' Class

- **report** (previously **displayResult**) is defined in both the base class and the subclasses
 - **report** in the base class prints **numEval** that is a common information to both the subclasses
 - The one in the base case handles general information and is inherited to the subclasses
 - **report** in each subclass prints further information specific to the problem type
 - To inherit a method of a superclass, it should be stated explicitly (e.g., **Problem.report(self)**)
- We can see that **object-oriented programming** provides us with the opportunity to organize codes in a way easier to maintain

Defining 'Problem' Class

- **Numeric** and **Tsp** have many methods of the same names but with different implementations
 - Polymorphism allows us to write codes that look the same regardless of the type of problem to be solved
 - E.g., `evaluate` of **Numeric** and `evaluate` of **Tsp** are of the same name but are implemented differently depending on the type of problem to be solved (numerical optimization or TSP)
- By introducing the classes and taking advantage of polymorphism, we will eventually be able to unite the main programs of different search algorithms into a single program
 - Duplications among different main programs can be avoided

Defining 'Problem' Class

- We store **Problem** in a separate file named 'problem.py'
 - 'random.py' and 'math.py' that were imported to the previous modules should now be imported to the 'problem.py' file
 - Each main program needs to import either **Numeric** or **Tsp** from the 'problem.py' file

Changes to the Main Program

- It is the `main` function that creates an instance of `Problem` object (`p = Numeric()` or `p = Tsp()`)
 - Then, the `setVariables` method (`p.setVariables`) is executed to store the corresponding values in the relevant class variables
 - Previously this was done by the function `createProblem`
- After creating problem `p`, the `main` function calls the search algorithm with `p` as its argument
 - The search algorithm calls the methods such as `randomInit`, `evaluate`, and `mutants` to conduct the search and these methods refer to the relevant class variables when executed

Changes to the Main Program

- The `main` function, after running the search algorithm, makes calls to relevant functions or methods to show the specifics of the problem solved (`p.describe`), to display the settings of the search algorithm (`displaySetting`), and then to display the result (`p.report`)
- There were two versions of `mutants` for the steepest-ascent hill climbing, one for numeric optimization and the other for TSPs
 - They are now migrated to the `Numeric` and `Tsp` classes
- Similarly, `randomMutant` of the first-choice hill climbing are migrated to the `Numeric` and `Tsp` classes, too
- The functions `displaySetting` and `bestof` still remain in the main program because they are tied with the search algorithms used rather than the types of the problems solved

Adding Gradient Descent

- Gradient descent is the same as the steepest-ascent except the way a next point is created from the current point
 - Steepest ascent generates m neighbors from which to select a successor to move to
 - m evaluations are needed
 - Gradient descent computes gradient at the current point and apply the gradient update rule to move to a next point
 - n evaluations are needed to calculate partial derivatives in all the dimensions, where n is the dimension of the objective function
 - One additional evaluation is needed to evaluate the next point obtained by applying the update rule using the gradient
- Gradient descent is applicable only to numerical optimization

Adding Gradient Descent

- Two variables are newly added to the `Numeric` subclass:
 - `alpha`: update rate for gradient descent
 - Set to a default value of 0.01 for the time being
 - Referenced by the method `takeStep`
 - `dx`: size of the increment used when calculating derivative
 - Set to a default value of 10^{-4} for the time being
 - Referenced by the method `gradient`

$$x \leftarrow x - \alpha \nabla f(x)$$

$$\frac{df(x)}{dx} = \lim_{dx \rightarrow 0} \frac{f(x + dx) - f(x)}{dx}$$

Adding Gradient Descent

- Also, five methods are newly added to the `Numeric` subclass:
 - `takeStep(self, x, v):`
 - Computes the gradient (`gradient`) of the current point `x` whose objective value is `v`

$$\nabla f(\mathbf{x}) = \left(\frac{\partial f(\mathbf{x})}{\partial x_1}, \frac{\partial f(\mathbf{x})}{\partial x_2}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_d} \right)^T$$

- Makes a copy of `x` and changes it to a new one by applying the gradient update rule as long as the new one is within the domain (`isLegal`)

$$(\mathbf{x} - \alpha \nabla f(\mathbf{x}))_i = x_i - \alpha \frac{\partial f(\mathbf{x})}{\partial x_i}$$

Adding Gradient Descent

- `gradient(self, x, v)`

- Calculates partial derivatives at \mathbf{x}

$$\frac{\partial f(\mathbf{x})}{\partial x_i} = \frac{f(\mathbf{x}') - f(\mathbf{x})}{\delta}$$

$$\mathbf{x}' = (x_1, \dots, x_{i-1}, x_i + \delta, x_{i+1}, \dots, x_d)^T$$

- Returns the gradient $\nabla f(\mathbf{x})$

- `isLegal(self, x)`

- Checks if \mathbf{x} is within the domain

- `getAlpha(self)`

- `getDx(self)`

- `getAlpha` and `getDx` are called from `displaySetting` of the main program when reporting the update rate and the increment size for calculating derivative