

## Table:

parallel cellular algorithms & programs

### Algorithm

1. Initialization
  - i) Define the grid dimensions
  - ii) Initialize each cell based on the problem
  - iii) Define the neighbourhood for each cell
  - iv) Define transition rule
2. Set initial condition
  - . populate the grid with all initial states for all cells
3. parallel update
  - for each time step:
    - for each  $(i, j, \dots)$  in grid (in parallel):
      - 1. -> get the current state of  $(i, j, \dots)$
      - 2. -> set the states of the neighbouring cells
      - 3. -> apply the transition rule to complete if  $c(i, j, \dots)$
    - then update all cells simultaneously
  - 4. Check Stopping condition
    - Terminate if a predefined condition met
    - A fixed no. of iterations completed
    - System reaches a steady state

```
import numpy as np
```

```
def fitness - function (position):
    return np.sum(position ** 2)
```

```
def initialize_grid(grid_size, dim):
    return np.random.uniform(-10, 10, (grid_size, grid_size, dim))
```

```
def evaluate_fitness():
    return np.apply_along_axis(fitness,
                               0, 2, grid)
```

```
def update_cell(cell, neighbors, learning_rate=0.5):
```

```
    return cell + learning_rate * np.mean(neighbors) - cell, 0.1 - 0.1
```

~~```
def get_neighbors(grid, cell)
```~~~~```
for i in [-1, 0, 1]
    for j in [-1, 0, 1]
```~~~~```
        if i == 0 & j == 0
            continue
```~~~~```
        xi, yi = (x + i) / Grid.shape[0],
                  (y + j) / Grid.shape[1]
        neighbors.append(Grid[xi, yi])
```~~~~```
return np.array(neighbors)
```~~

~~def AlgoGridSize = 10, clm = 2, maxIteration = 100,~~

~~grid = initialize\_grid(GridSize, clm),~~

~~best\_so\_far = None~~

~~best\_fitness = float('inf')~~

for x in range(maxIteration):  
 fitness\_grid = evaluate\_fitness(grid)

for x in range(GridSize - size):

for y in range(GridSize):

neighbors = get\_neighbours(grid, x, y)  
 grid[x, y] = update\_cell(grid, y, neighbors)

max\_fitness = np.max(fitness\_grid)

if max\_fitness > best\_fitness:

best\_fitness = max\_fitness

best\_solution = grid[np.unravel\_index

(np.argmax(fitness\_grid), shape)]

return best\_solution, best\_fitness

best\_solution[0], best\_fitness = parallel\_csmula

print("Best solution: ", best\_solution)  
print("Best fitness: ", best\_fitness)

about:

Best solution= [-0.5305 0.4845]  
Best fitness= 0.2002485