

Optimization Methods

Lab 9 Session



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Task1

Problem 1: Minimize $x_1 + x_2$
subject to $x_1 + x_2 \geq 1$
 $\mathbf{0} \leq \mathbf{x} \leq \mathbf{1}$

Since $x_1 + x_2 \geq 1$, to minimize $x_1 + x_2$, it is simple to find know that the minimum is 1.

$$x_1 = t, x_2 = 1 - t \text{ where } 0 \leq t \leq 1$$



Task1

Problem 2: Minimize $x_1 + x_2 + x_3 + x_4$
subject to $x_1 + x_2 = 1$
 $x_2 + x_3 = 1$
 $x_3 + x_4 = 1$
 $x_1 + x_4 = 1$
 $\mathbf{0} \leq \mathbf{x} \leq \mathbf{1}$

If we add the first and third constraint, we can directly get that $x_1 + x_2 + x_3 + x_4 = 1$, which means that the minimum value is 1



Task2

In this task , I use these 3 LP solvers: **GLPK (GNU Linear Programming Kit)**, **Gurobi**, **PuLP**, and get the result for two problems shown in the following tables.

Problem1

| | x1 | x2 | Object function |
|--------|-----|-----|-----------------|
| GLPK | 1.0 | 0.0 | 1.0 |
| Gurobi | 1.0 | 0.0 | 1.0 |
| PuLP | 1.0 | 0.0 | 1.0 |

Problem2

| | x1 | x2 | x3 | x4 | Object function |
|--------|-----|-----|-----|-----|-----------------|
| GLPK | 0.0 | 1.0 | 0.0 | 1.0 | 2.0 |
| Gurobi | 1.0 | 0.0 | 1.0 | 0.0 | 2.0 |
| PuLP | 0.0 | 1.0 | 0.0 | 1.0 | 2.0 |



Task3

Task 3: Compare the following four methods on your 100-item knapsack problem (used in the lab session on Simulated Annealing) with respect to the solution quality (i.e., the objective function value), and the total computation time.

- (i) A heuristic method used to generate an initial solution for SA.
- (ii) Simulated algorithm with your parameter setting
- (iii) Use of an LP solver (i.e., use of the LP relaxation problem), and create a feasible solution from the LP solutions.
- (iv) Use of an integer LP (ILP) solver

Approach to change LP to integer problem:

Set boundary = 1;

Set element = (element < boundary)

While constraint is satisfied:

Set element = (element < boundary)

boundary -= 0.01

| | (i) | (ii) | (iii) | (iv) |
|--------------|--------|-------|--------|-------|
| Running time | 0.06ms | 200ms | 1.14ms | 431ms |
| Object value | 1667 | 2953 | 2987 | 3093 |



Task4

200-item

| | (i) | (ii) | (iii) | (iv) |
|--------------|--------|-------|--------|-------|
| Running time | 0.43ms | 413ms | 2.14ms | 631ms |
| Object value | 4190 | 6248 | 6260 | 6341 |

400-item

| | (i) | (ii) | (iii) | (iv) |
|--------------|--------|-------|--------|---------|
| Running time | 0.89ms | 723ms | 3.41ms | 90067ms |
| Object value | 9841 | 12918 | 12929 | 12936 |

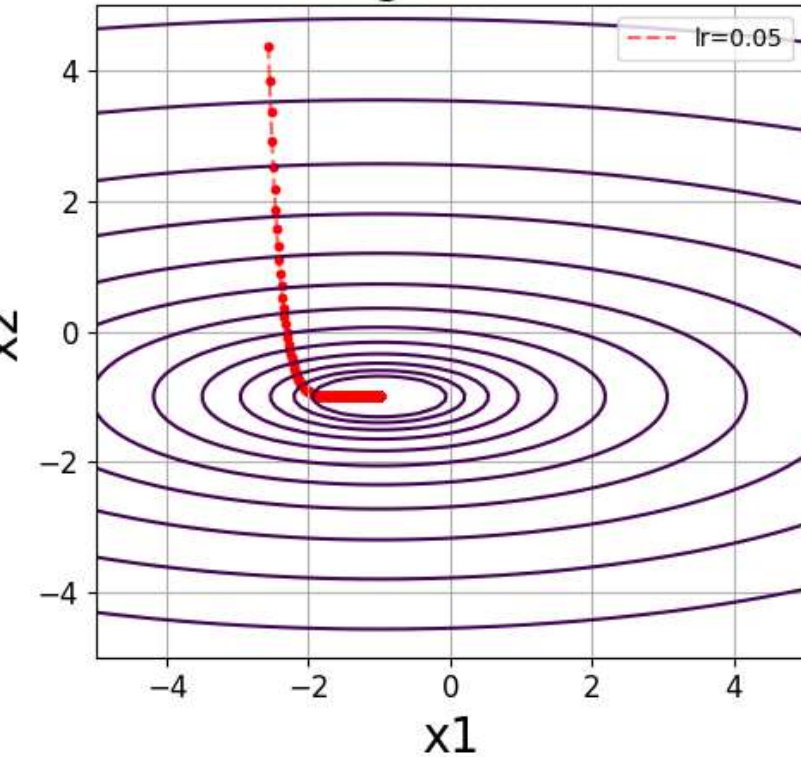
For using an integer LP (ILP) solver, the running time increase very fast when the scale of the problem grows, while using an LP solver increase nearly linearly.



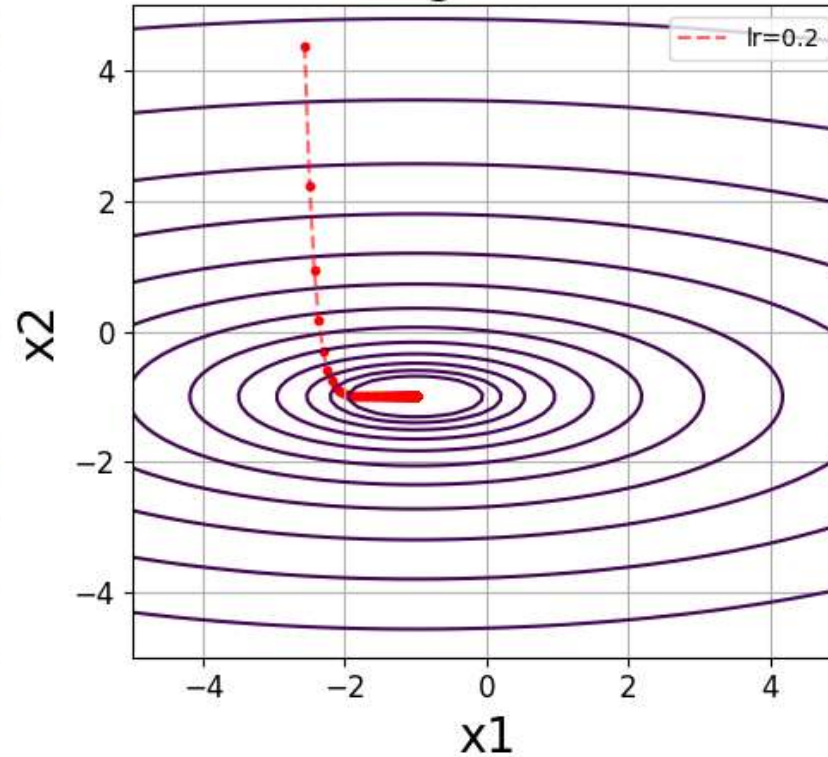
Task5

I set 3 learning rate: 0.05, 0.2 and 0.9

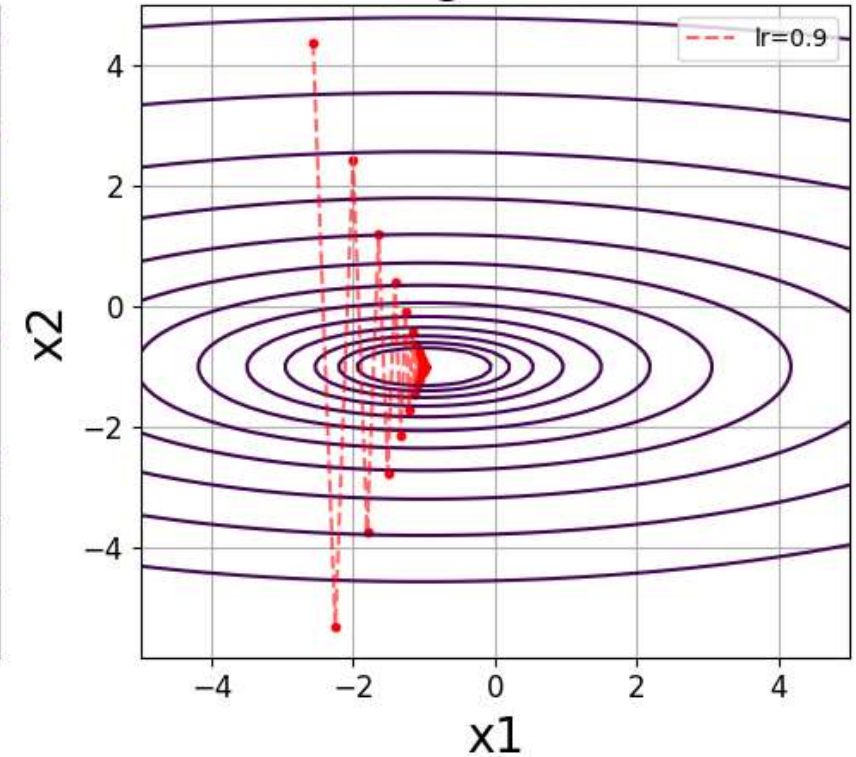
Learning Rate = 0.05



Learning Rate = 0.2



Learning Rate = 0.9



When learning rate is too small, it will take a long time to search at beginning. And if learning rate is too large, it will cause shaking. Therefore, select learning rate with a median value like 0.2 is reasonable .

