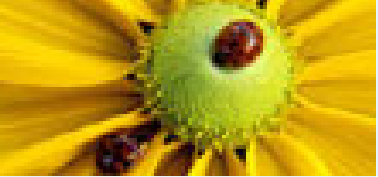


## **Week 9**

# ***Integer Programming: Algorithms - 3***

OPR 992

Applied Mathematical Programming



#### Column Generation Algorithms

- Dantzig-Wolfe Reformulation
- Example
- Strength of the Linear Programming Master

Heuristic Algorithms

# Column Generation Algorithms

# Dantzig-Wolfe Reformulation

## Column Generation Algorithms

### ● Dantzig-Wolfe Reformulation

#### ● Example

#### ● Strength of the Linear Programming Master

#### Heuristic Algorithms

Consider an IP of the form

$$\begin{array}{ll} \max & \sum_{k=1}^K (c^k)^T x^k \\ \text{s.t.} & \sum_{k=1}^K (A_k)^T x^k \leq b^k \\ & x^k \in X^k, \quad k = 1, \dots, K \end{array}$$

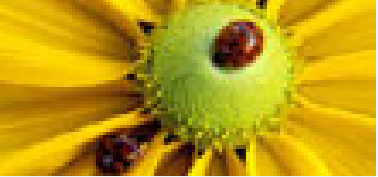
Assuming that

$$X^k = \{x^{k,t}\}_{t=1}^{T_k}, \quad k = 1, \dots, K,$$

we can introduce  $\lambda_{k,t} \in \{0, 1\}$  such that

$$X^k = \sum_{t=1}^{T_k} \lambda_{k,t} x^{k,t}, \quad \text{and} \quad \sum_{t=1}^{T_k} \lambda_{k,t} = 1.$$

# Example



## Column Generation Algorithms

- Dantzig-Wolfe Reformulation

- Example

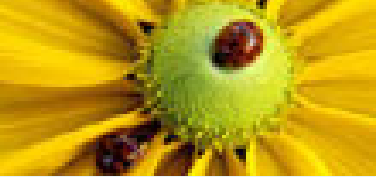
- Strength of the Linear Programming Master

## Heuristic Algorithms

Solve an instance of STSP with the following distance matrix:

$$\begin{bmatrix} - & 7 & 2 & 1 & 5 \\ - & - & 3 & 6 & 8 \\ - & - & - & 4 & 2 \\ - & - & - & - & 9 \\ - & - & - & - & - \end{bmatrix}$$

# Strength of the Linear Programming Master



## Column Generation Algorithms

- Dantzig-Wolfe Reformulation
- Example

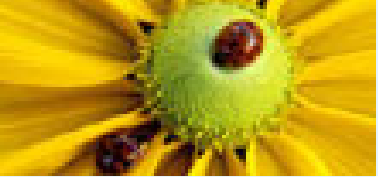
## ● Strength of the Linear Programming Master

## Heuristic Algorithms

$$z^{LPM} = \max \left\{ \sum_{k=1}^K c^k x^k : \sum_{k=1}^K A^k x^k = b, x^k \in \text{conv}(X^k), k = 1, \dots, K \right\}.$$

$$z^{LPM} = w_{LD} = z^{CUT}.$$

Choose the right algorithm based on speed.

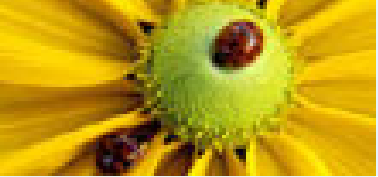


## Column Generation Algorithms

### Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

# Heuristic Algorithms



# Introduction

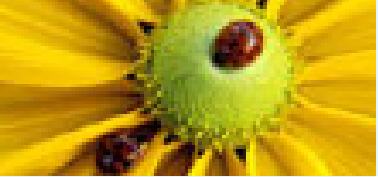
Column Generation Algorithms

Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

When should we use a heuristic?

- A solution is required rapidly.
- The instance is too large to formulate as whole problem of reasonable size.
- Once formulated, known algorithms cannot solve it in real time.
- It is much easier to find solutions by inspection than by solving using a general-purpose algorithm.



# Tabu Search

Column Generation Algorithms

Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

Local search for a constrained problem of the form

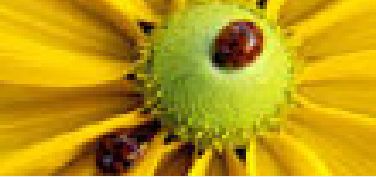
$$\min\{c(x) : g(x) = 0\}$$

can involve a goal function

$$c(x) + \alpha g(x).$$

For different values of  $\alpha$ , the local search can cycle among solutions.

Solution: Make certain solution forbidden to avoid cycling.



# Simulated Annealing

Column Generation Algorithms

Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

Choose a neighbor randomly.

The probability of choosing a neighbor depends on the effect on the goal function.

Parameters of a simulated annealing algorithm:

1. The initial temperature  $T$
2. The cooling ratio  $r$
3. The loop length  $L$
4. The definition of “frozen,” or the stopping criterion.

# Simulated Annealing - The Algorithm

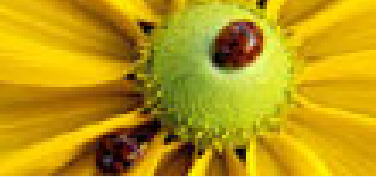


Column Generation Algorithms

Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

1. Get an initial solution  $S$ .
2. Get an initial temperature  $T$  and a cooling ratio  $r$  with  $0 < r < 1$ .
3. While not yet frozen, do the following:
  - (a) Perform the following loop  $L$  times:
    - i. Pick a random neighbor  $S'$  of  $S$ .
    - ii. Let  $\Delta = f(S') - f(S)$ .
    - iii. If  $\Delta \leq 0$ , set  $S = S'$ .
    - iv. If  $\Delta > 0$ , set  $S = S'$  with probability  $e^{-\Delta/T}$ .
  - (b) Reduce the temperature by setting  $T \leftarrow rT$ .
4. Return the best solution found.



# Genetic Algorithms

Column Generation Algorithms

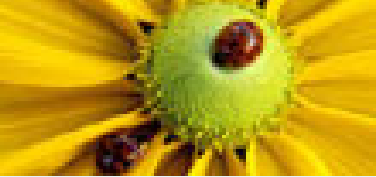
Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

Start with a finite population of solutions and evolve it from one generation to the next.

Each iteration performs the following steps:

1. Evaluation: The fitness of the individuals is evaluated.
2. Parent Selection: Certain pairs of solutions (parents) are selected based on their fitness.
3. Crossover: Each pair of parents combines to produce one or two new solutions (offspring).
4. Mutation: Some of the offspring are randomly modified.
5. Population Selection: Based on their fitness, a new population is selected replacing some or all of the original population by an identical number of offspring.



# Local Search Wrap-Up

Column Generation Algorithms

Heuristic Algorithms

- Introduction
- Tabu Search
- Simulated Annealing
- Genetic Algorithms

In any local search algorithm, it is important to balance:

1. Communication
2. Diversification
3. Intensification