

Optimization Problems in Network Design

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Outline

1. Background: Optimization Problems in Network Design
2. Progress in Content Connectivity Research

Resource Allocation in Optical Networks

- *Fixed-grid*: Routing and Wavelength Assignment (RWA) to select path and wavelength for lightpaths
- *Flex-grid*: Routing and Spectrum Allocation (RSA) including parameters of tunable transponders

Traffic Models for Resource Allocation

- Planning phase:
 - ✓ Network is empty.
 - ✓ Connection requests known in advance
 - ✓ Offline and static
- Operational phase:
 - ✓ Connection requests arrive dynamically.
 - ✓ Taking into account network current utilization
 - ✓ Online and dynamic

Optical Network Optimization Problems

- RSA and RWA are NP-complete (no polynomial time solutions)
- In next slides: A review of algorithmic techniques

Network Optimization Problems

- Problem definition:

Minimize $f_0(\mathbf{x})$ subject to $f_i(\mathbf{x}) \leq b_i, i = 1, \dots, m$

where:

- ✓ $\mathbf{x} = (x_1, x_2, \dots, x_n)^T \in R^{n \times 1}$
 - ✓ $f_0: R^n \rightarrow R$: optimization function
 - ✓ $f_i: R^n \rightarrow R$: constraint functions.
- Linear Programming (LP), Integer Linear Programming (ILP), Connection of ILP and LP

Linear Programming

- Problem definition:

$$\text{Minimize } \mathbf{c}^T \cdot \mathbf{x} \text{ subject to } \mathbf{A} \cdot \mathbf{x} \leq \mathbf{b}$$

where:

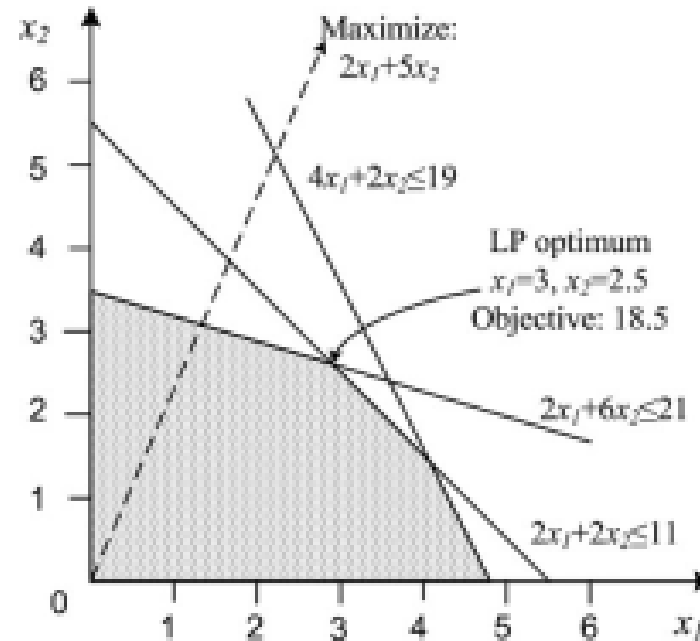
✓ $\mathbf{x} = (x_1, x_2, \dots, x_n)^T \in R^{n \times 1}$

✓ $\mathbf{c}: n \times 1$

✓ $\mathbf{A}: m \times n$

✓ $\mathbf{b}: m \times 1$

- Both optimization function and constraints are linear.
- An optimal solution always exists.



Maximize : $2x_1 + 5x_2$
subject to
 $2x_1 + 6x_2 \leq 21$
 $4x_1 + 2x_2 \leq 19$
 $2x_1 + 2x_2 \leq 11$
 $x_1 \geq 0, x_2 \geq 0$

Integer Linear Programming

- Problem definition:

Minimize $\mathbf{c}^T \cdot \mathbf{x}$ subject to $\mathbf{A} \cdot \mathbf{x} \leq \mathbf{b}$

where:

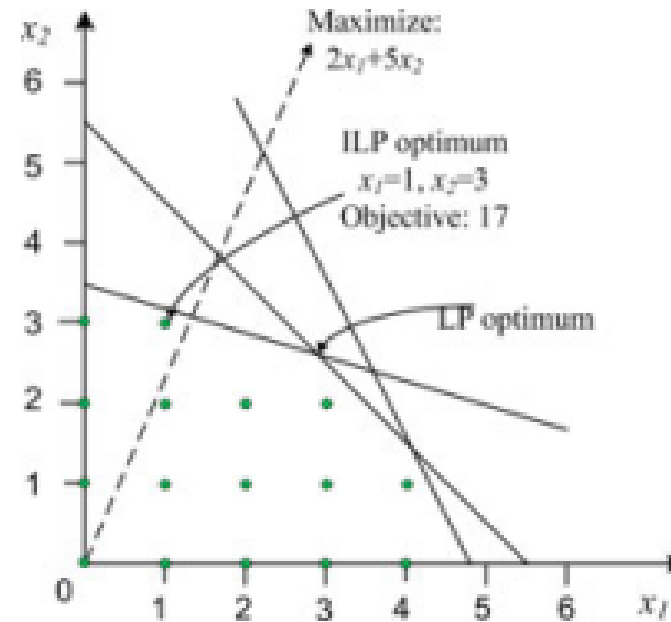
✓ $\mathbf{x} = (x_1, x_2, \dots, x_n)^T \in \mathbb{Z}^{n \times 1}$

✓ $\mathbf{c}: n \times 1$

✓ $\mathbf{A}: m \times n$

✓ $\mathbf{b}: m \times 1$

- Both optimization function and constraints are linear.
- An optimal solution is not guaranteed.



Maximize: $2x_1 + 5x_2$
subject to
 $2x_1 + 6x_2 \leq 21$
 $4x_1 + 2x_2 \leq 19$
 $2x_1 + 2x_2 \leq 11$
 $x_1 \geq 0, x_2 \geq 0$
 x_1, x_2 integers

Connection of ILP and LP

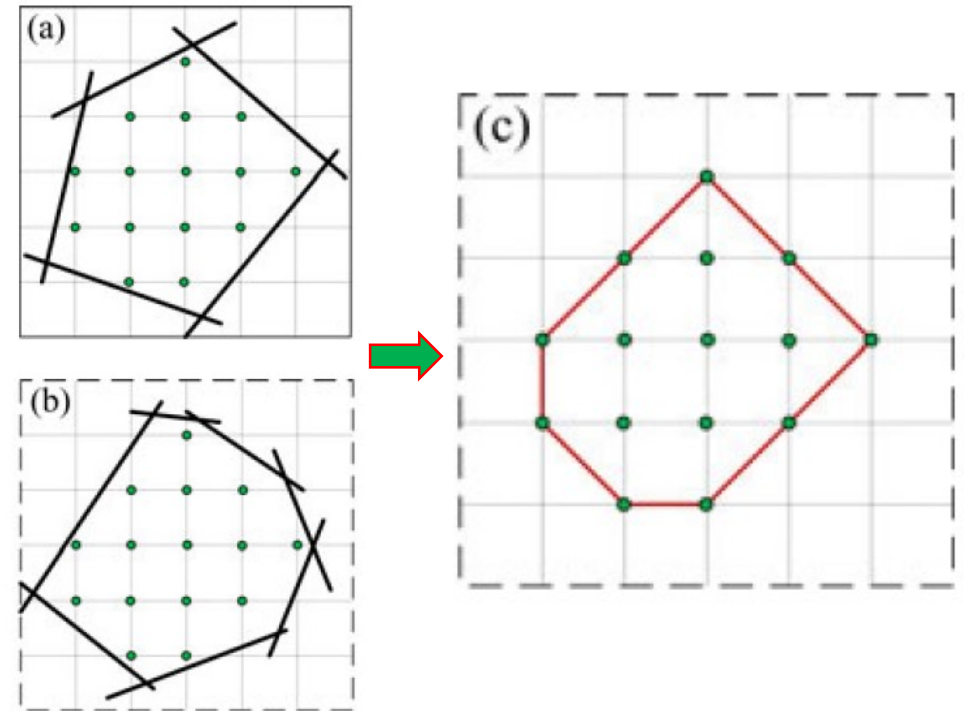
- **LP-Relaxation:**

- ✓ Solve the same ILP problem without variables to be integers.
- ✓ If solutions are integers, then they are optimal solutions.
- ✓ Providing a lower and upper bounds of optimization function
- ✓ Sophisticated techniques to increase probability the solutions are integers

Connection of ILP and LP

- **Convex Hull:**

- ✓ Feasible solutions of an ILP can be represented by n -dimension polyhedrons.
- ✓ Convex hull: minimum convex set including all integer solutions
- ✓ ILP + convex hull constraints = LP



Connection of ILP and LP

- **Good ILP Formulations:**

- ✓ Constraints close to be a Convex Hull
- ✓ Boolean variables
- ✓ Avoid Integer and Big-M variables

Heuristic Algorithms

- For non-convex problems and NP-hard
- Local search
- Greedy algorithms
- Metaheuristic algorithms

Local Search

Algorithm 20 Local search (*best-fit*)

```
1: Initialization: Set  $x$  initial solution
2: do
3:    $next = x$ .
4:   for each  $y \in \mathcal{V}(x)$ 
5:     if  $f(y) < f(x)$    comment:  $f$  is the cost function to minimize.
6:        $next = y$    comment: Improving solution
7:     comment: If first-fit then execute break here to leave the for each loop
8:   while  $next \neq x$    comment: End when no improving solution
9: return  $x$ 
```

- Parameters: Vicinity Size, Connectivity of Solutions, and Best-fit or First-fit



Simulated Annealing

Algorithm 21 Simulated annealing

```
1: Initialization:
2:   Set  $T = T(0)$ , initial temperature,  $x$  initial solution,  $x_{best} = x_0$  incumbent solution
3: do   comment: Outer loop
4:   do   comment: Inner loop
5:      $v =$  random chosen neighbor in  $\mathcal{V}(x)$ 
6:     if  $c(v) < c(x)$  or with probability  $e^{-\frac{c(v)-c(x)}{T}}$ 
7:        $x = v$ , update  $x_{best}$  if needed. comment: Jump to  $v$ 
8:     while tempDecreaseCriterion
9:     Decrease  $T$ .
10: while stopCriterion
11: return  $x_{best}$ 
```

- Better solution, accept; If not, still accept with the probability; High T , non-improving solutions are always accepted; Low T , non-improving solutions are never accepted.



Tabu Search

Algorithm 22 Tabu search (basic scheme)

```
1: Initialization:
2:   Set  $x$  initial solution,  $x_{best} = x$ 
3:   Set  $\mathcal{TL} = \emptyset$    comment: Tabu list initially empty
4: do
5:    $x_{bestNeighbor} = \emptyset$ .
6:   for each  $y \in \mathcal{V}(x)$ ,  $y$  not tabu
7:     if  $f(y) < f(x_{bestNeighbor})$ 
8:        $x_{bestNeighbor} = y$    comment: Improving solution
9:    $x = x_{bestNeighbor}$    comment: Jump to the best neighbor
10:   $\mathcal{TL} = \mathcal{TL} \cup a(x)$    comment: Update the tabu list
11:  if  $|\mathcal{TL}| > T$  then remove oldest element in  $\mathcal{TL}$ 
12:  if  $f(x) < f(x_{best})$  then  $x_{best} = x$    comment: Update incumbent solution
13: while stop criterion not met
14: return  $x$ 
```

- Variation of Local Search permitting jumping to non-improving locations

Heuristic: Considerations

- Intensification and Diversification (clever and balanced search)
- Stop conditions: optimal bounds, max. runtime or iterations, no improvement
- Vicinity size

Network Connectivity and Content Connectivity

ILP Problem Statement

Given:

1. Physical topo.
2. Logical topo.
3. Datacenter locations
4. Wavelengths/link
5. Cut-set space
(already in Logical topo.)

Cost function:

Minimize total wavelength channels

Output Results:

1. f_{NC} (vs. connectivity degree, SRG size – number of link and node failures)
2. f_{CC} (vs. connectivity degree, SRG size – number of link and node failures)
3. Result comparison
4. Running time (scalable, heuristic)

ILP Formulation + Heuristic

Presolve eliminates 2693 constraints and 12890 variables.

Adjusted problem:

9176 variables, all binary

12620 constraints, all linear; 40080 nonzeros

2080 equality constraints

10540 inequality constraints

1 linear objective; 1810 nonzeros.

Sorry, a demo licenses for AMPL is limited to 500 variables and 500 constraints and objectives (after presolve) for linear problems. You have 19608 variables, 12620 constraints, and 1 objective.
ampl: include CC1w-deg6.run;

In progress

- 1) Replicate the ILP results in AMPL + CPLEX (Online Server)
- 2) Develop and simulate Heuristic Algorithms for the ILP

