

A Network-based Column Generation Approach for Volumetric Modulated Arc Therapy

Amirhossein Vaeztehrani, Houra Mahmoudzadeh

amir.vaeztehrani@uwaterloo.ca, houra.mahmoudzadeh@uwaterloo.ca

Abstract

Background

- Radiation therapy (RT) is used in more than 50% of cancer treatment cases
- **Goal:** irradiating the tumor while protecting adjacent organs

Volumetric Modulated Arc Therapy

- **VMAT:** a fast RT modality with continuously rotating beam around patient
- Apertures change by a MLC to avoid irradiating critical organs

Purpose

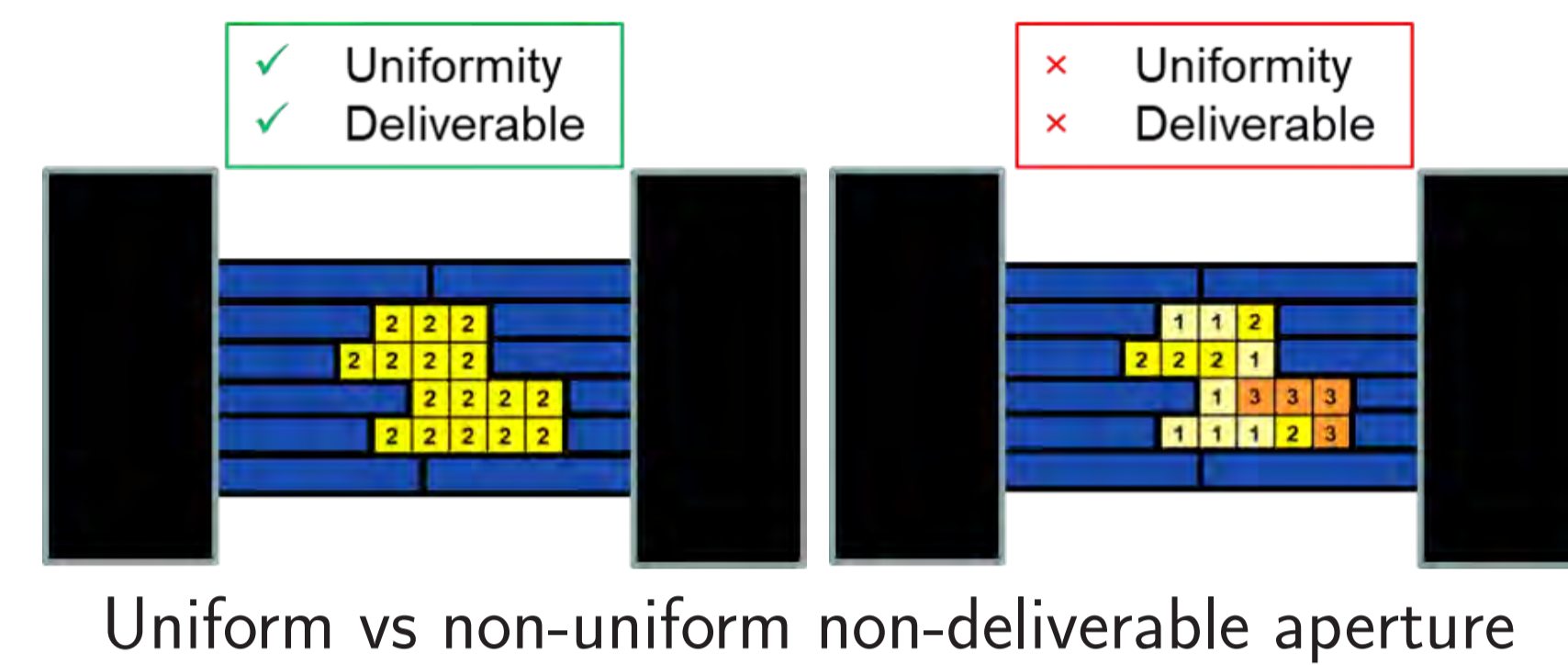
- We model the clinical and deliverability constraints by a large-scale MIP and solve it

Method

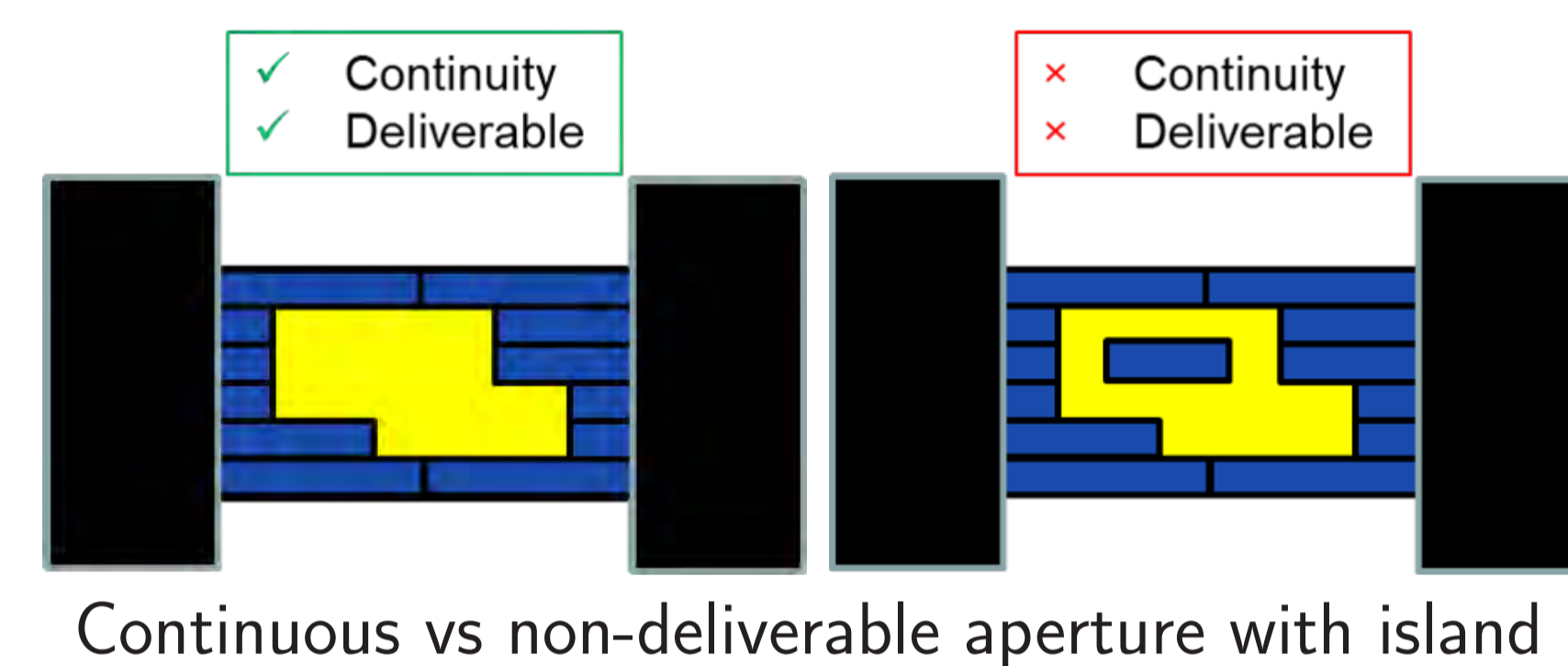
- We develop a network-based column-generation algorithm which optimizes the intensity and shape of apertures.

Deliverability constraints

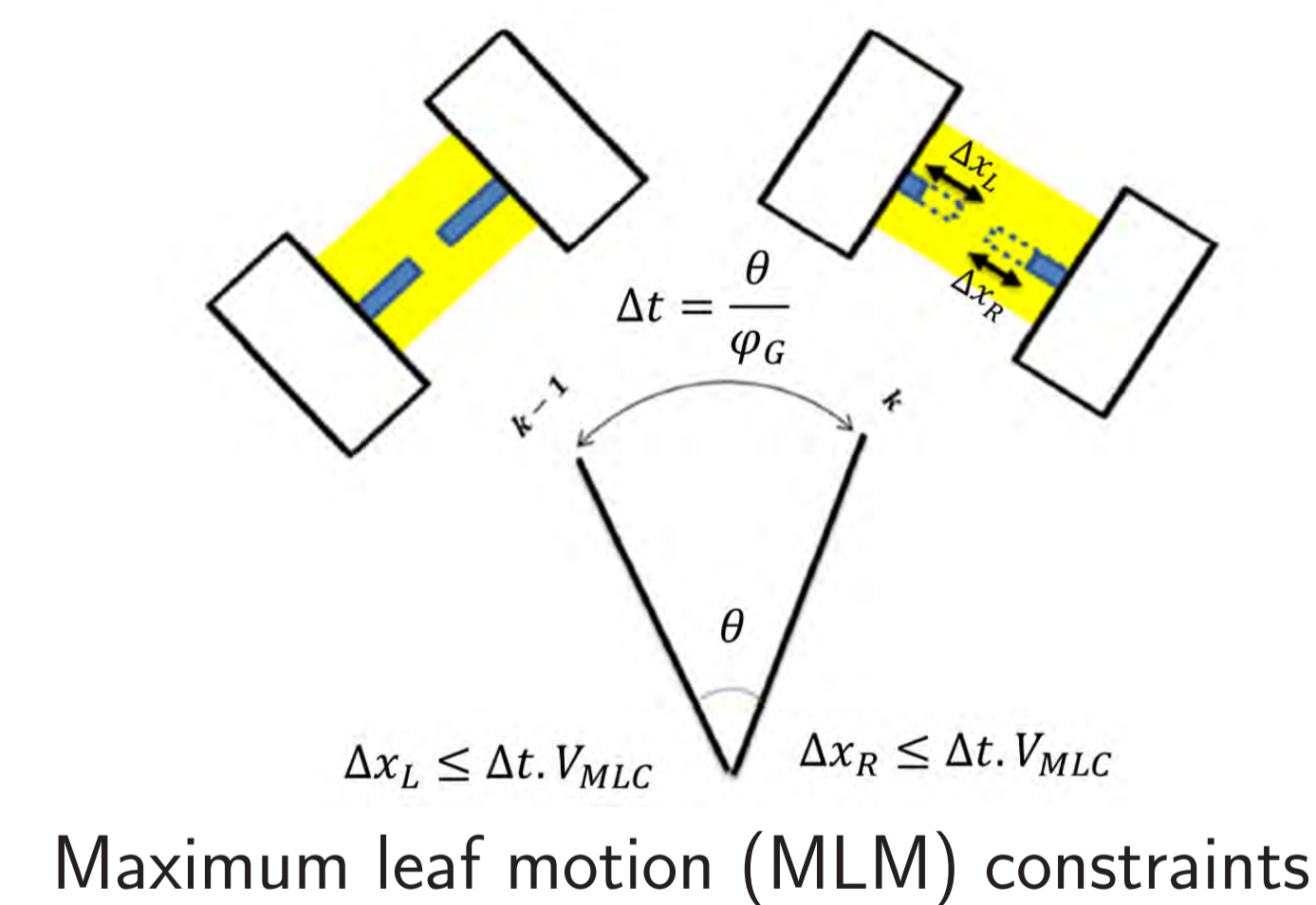
1. Uniformity constraints: Apertures deliver uniform dose



2. Continuity constraints: Apertures are continuous



3. Leaf motion constraints: Within an arc θ , MLC leaves can move no more than Δx



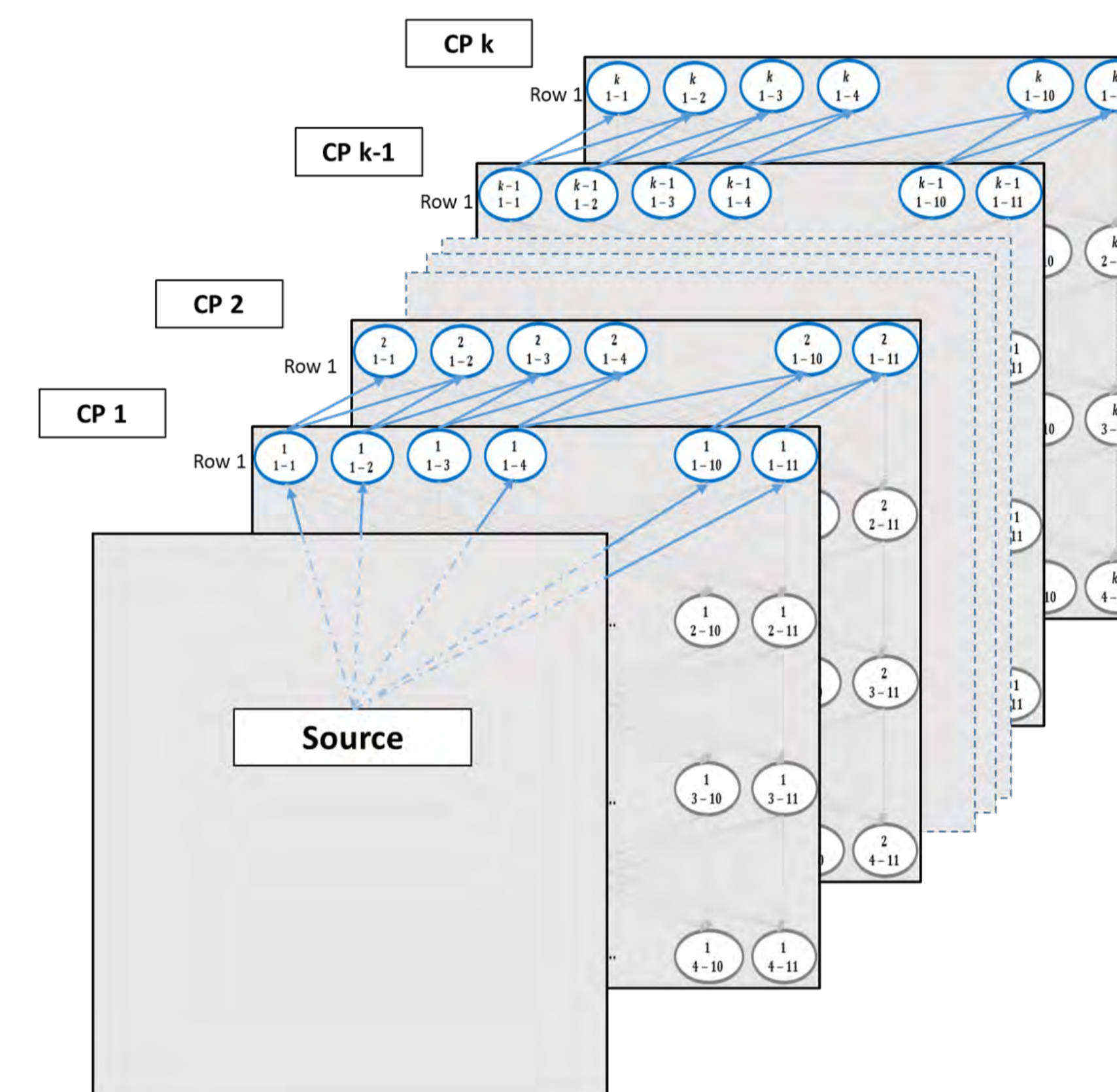
Column generation

Master problem (MP)

- Optimizes the intensity over the available set of apertures
- A linear FMO problem

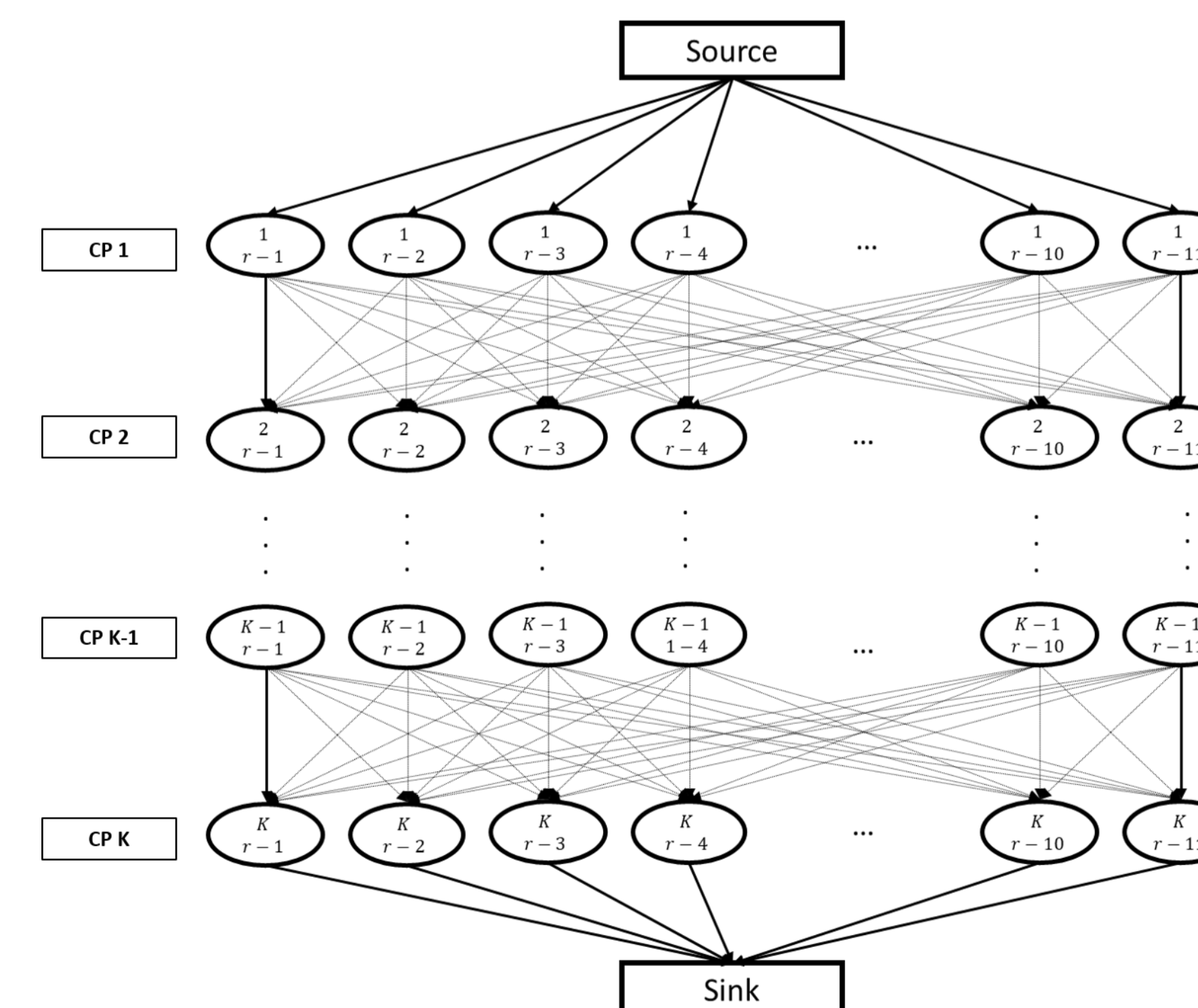
Pricing problem (PP)

- Optimizes the shape of apertures
- MIP PP is modeled by a network



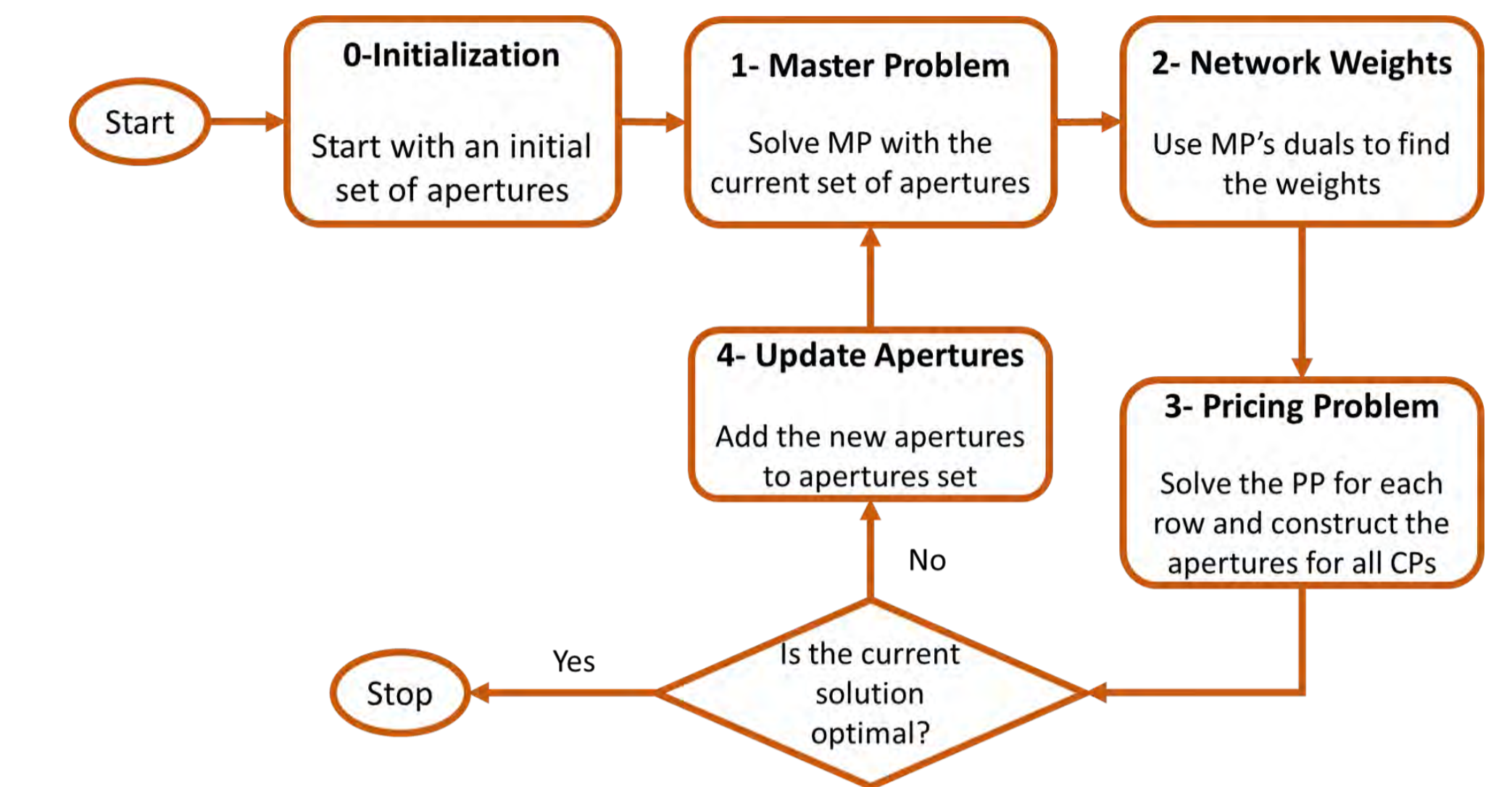
Modeling the MLC at each CP by a network

- Leaves in MLC can move independently.
- Position of each leaf pairs is dependent to that of the adjacent CP.
- A shortest path problem is solved for each MLC row.



The network for the first row of MLC

Algorithm



Results

A numerical example

- An example with 5 CPs, constant speed and dose rate
- For dose of 200 (Gy) and $\Delta x = 1$
- Optimal solution after 7 iterations

Iteration	Arc	Optimal Intensity (Gy)	Objective
0		118	1633
1		173	1106
...
7		215	745

CG iteration for the example

Problem statement

Standard assumption

- We approximate the continuous path by discrete control points (CPs) and solve a DAO problem at each CP

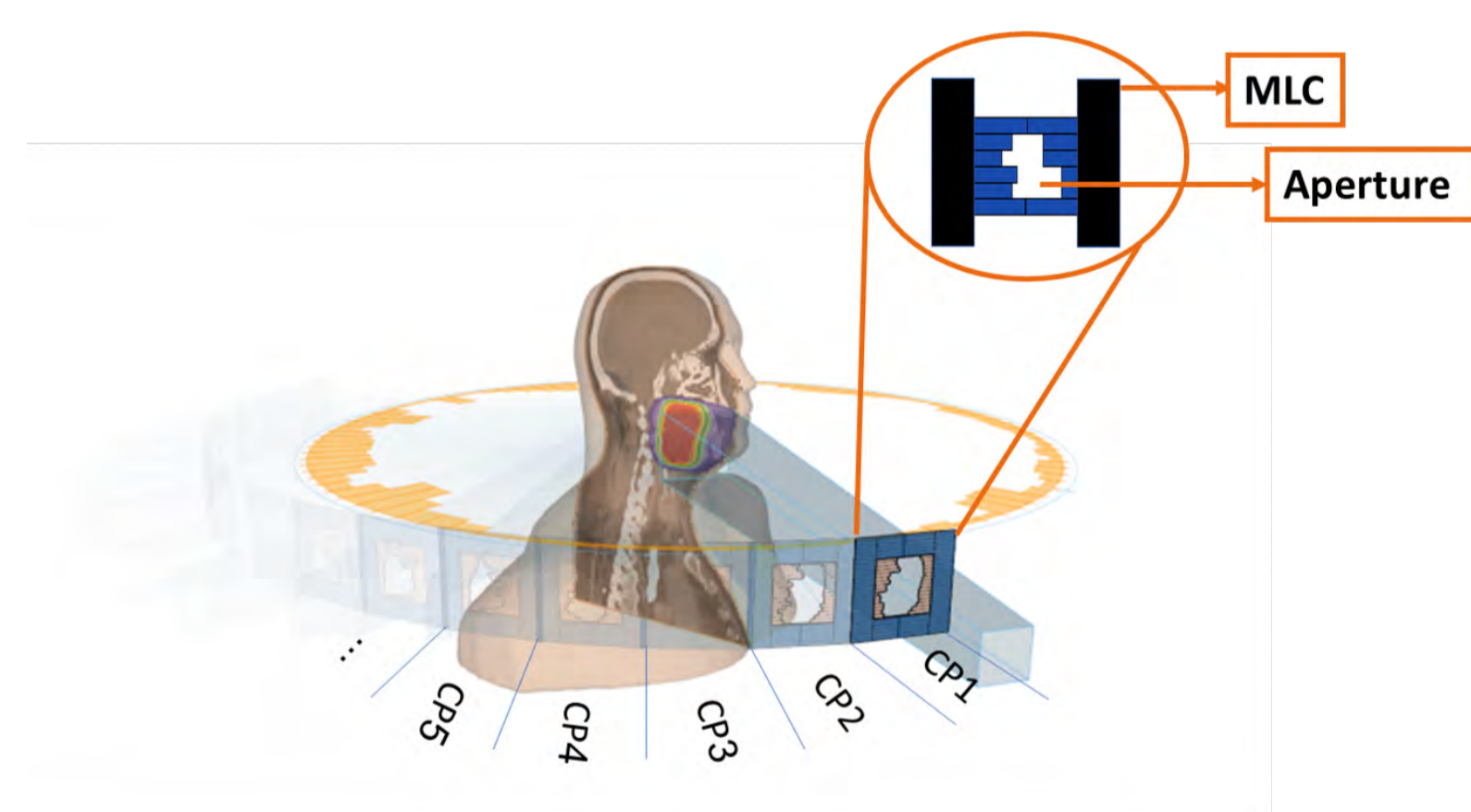
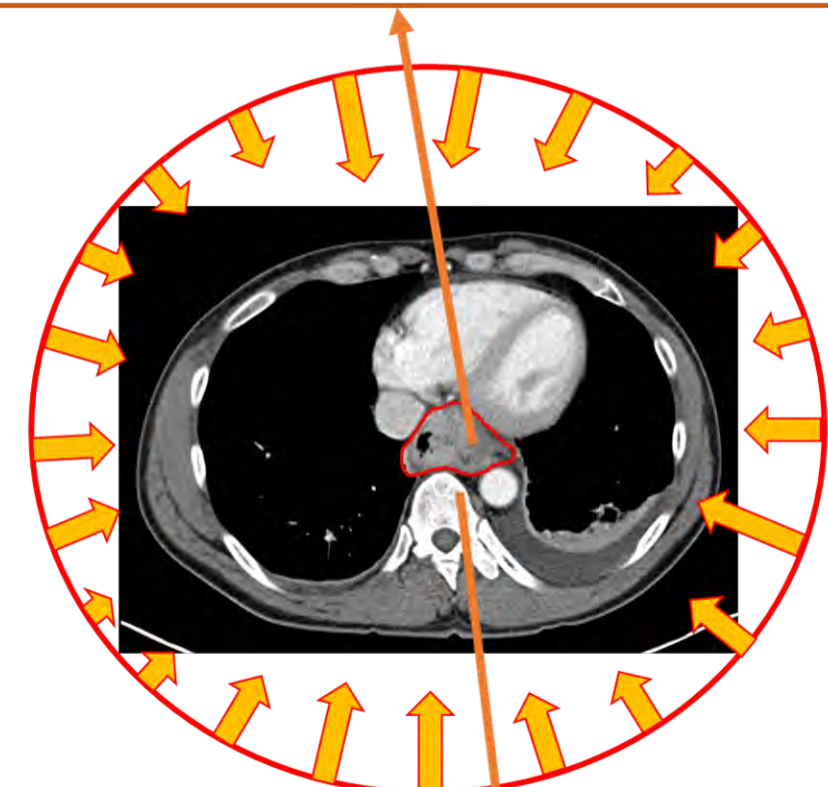


Image from www.rayresearchlab.com

Model

The tumor must receive the prescribed dose (**Clinical constraints**)



Dose to adjacent organs is minimized (**Objective function**)

Challenges

Advantages

- ✓ Fast delivery
- ✓ Less patient discomfort
- ✓ Less positioning error
- ✓ Less waiting time
- ✓ More machine availability
- ✓ Potential for higher quality

Challenges

- Practical**
 - ❖ Feasible solution (Not optimal)
 - ❖ Limited usage
- Mathematical**
 - ❖ Excessively large problem
 - ❖ MIP constraints

VMAT advantages and challenges

Conclusion and future works

Conclusion

- Our CG converges to optimal solution very fast
- At each iteration the hard Deliverability constraints in PP is modeled and solved by a network efficiently

Future works

- We incorporate uncertainty in our VMAT model using robust optimization
- We will incorporate more complicated row-dependent deliverability constraints and modify our network to solve PP