# Optimization Models for Container Inspection

#### **Endre Boros**

RUTCOR, Rutgers University

Joint work with L. Fedzhora and P.B. Kantor (Rutgers),

and K. Saeger and P. Stroud (LANL)

ション ふゆ マ キャット マックシン

#### Problem

- Finding ways to intercept illicit nuclear materials and weapons destined for the U.S. via the maritime transportation system is an exceedingly difficult task. Today, only a small percentage of containers arriving to U.S. ports are inspected.
- Inspection involves checking paperwork, using various imaging sensors, and manual inspection.
- Objectives involve maximizing **detection rate**, minimizing **unit cost of inspection**, **rate of false positives**, **time delays**, etc.

#### Problem

- Finding ways to intercept illicit nuclear materials and weapons destined for the U.S. via the maritime transportation system is an exceedingly difficult task. Today, only a small percentage of containers arriving to U.S. ports are inspected.
- Inspection involves checking paperwork, using various imaging sensors, and manual inspection.
- Objectives involve maximizing **detection rate**, minimizing **unit cost of inspection**, **rate of false positives**, **time delays**, etc.

#### Problem

- Finding ways to intercept illicit nuclear materials and weapons destined for the U.S. via the maritime transportation system is an exceedingly difficult task. Today, only a small percentage of containers arriving to U.S. ports are inspected.
- Inspection involves checking paperwork, using various imaging sensors, and manual inspection.
- Objectives involve maximizing detection rate, minimizing unit cost of inspection, rate of false positives, time delays, etc.

### A small example involving two sensors



▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

#### A small example involving two sensors





◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへぐ

#### A small example involving two sensors







 $0.4 C_{CHK} + C_a$ 

Detection rate

 $\mathbf{60}\%$ 

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへ⊙

### A small example involving two sensors





▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

Detection rate

60%

### A small example involving two sensors





### A small example involving two sensors







$0.4 C_{CHK}$	$0.1 \rm C_{\rm CHK}$
$+C_{a}$	$+C_{a}$
	$+0.5 C_{\mathrm{b}}$

Detection rate 60%

64%

▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

### A small example involving two sensors







rate



▲ロト ▲圖ト ▲画ト ▲画ト 三直 - のへで

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- $\bullet$  over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

• Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.

#### Maximize detection rate $\Delta(\mathbf{D}, \mathbf{t})$

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

#### A possible solution (Stroud and Saeger, 2003)

• Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

Maximize detection rate  $\Delta(\mathbf{D}, \mathbf{t})$ 

- over all decision trees D and threshold selections t
- subject to **budget**, **capacity**, and **delay** constraints

A possible solution (Stroud and Saeger, 2003)

- Enumerate all possible (binary) decision trees and compute best possible threshold selections for each.
  - Number of decision trees is doubly exponential!

- Enumeration is possible only for  $s \le 4!$
- Too expensive to analyze tradeoffs!
- Why only 1-1 thresholds?
- Why a single decision tree?

# Large Scale LP Formulation

• Developed a polyhedral description of all possible decision trees.

- Formulated a large scale LP model for optimal inspection policy; maximization of **detection rate**, while limiting **unit cost of inspection**, **rate of false positives**, and **time delays**, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

イロト 不得下 イヨト イヨト

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Large Scale LP Formulation

- Developed a polyhedral description of all possible decision trees.
- Formulated a large scale LP model for optimal inspection policy; maximization of detection rate, while limiting unit cost of inspection, rate of false positives, and time delays, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

・ロト ・ 西ト ・ ヨト ・ 日 ・

ъ

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Large Scale LP Formulation

- Developed a polyhedral description of all possible decision trees.
- Formulated a large scale LP model for optimal inspection policy; maximization of detection rate, while limiting unit cost of inspection, rate of false positives, and time delays, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Large Scale LP Formulation

- Developed a polyhedral description of all possible decision trees.
- Formulated a large scale LP model for optimal inspection policy; maximization of detection rate, while limiting unit cost of inspection, rate of false positives, and time delays, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Large Scale LP Formulation

- Developed a polyhedral description of all possible decision trees.
- Formulated a large scale LP model for optimal inspection policy; maximization of detection rate, while limiting unit cost of inspection, rate of false positives, and time delays, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Large Scale LP Formulation

- Developed a polyhedral description of all possible decision trees.
- Formulated a large scale LP model for optimal inspection policy; maximization of detection rate, while limiting unit cost of inspection, rate of false positives, and time delays, etc.
- Off the shelf LP packages can find optimal inspection strategies up to 6-8 sensors.

- Detection rate unit inspection cost ROC curve can be tabulated.
- Effects of capacity and time delay limitations can be analyzed.
- Benefits of new sensor technologies can be evaluated.

# Experiments with 4 sensors (Stroud and Saeger, 2003)



- Detection rate  $\geq 81.5\%$
- Threshold-optimized pure strategy found by Stroud and Saeger (2003)

うして ふゆう ふほう ふほう ふしつ

• Non-optimized threshold grid; savings of  $\approx 10\%$ 

### Experiments with 4 sensors (Stroud and Saeger, 2003)



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへぐ