# Applying Cost Curves to Marine Cargo Container Inspection

DIMACS / DyDAn / LPS Workshop on Port Security, Safety, Inspection, Risk Analysis and Modeling

> Rutgers University November 18<sup>th</sup>, 2008

> > **Richard Hoshino**

Laboratory and Scientific Services Directorate Canada Border Services Agency





- Each year in Canada, approximately 20,000 marine containers are *referred* for a full examination.
- Some of these containers have been fumigated with chemical compounds to kill invasive alien species.
- If these marine containers are not ventilated properly, fumigants may pose a risk to the health and safety of border service officers.

#### **Flowchart of Current Process**



# A Simple Yet Powerful Insight

- We can create a mathematical model that predicts whether a container has been fumigated.
- For containers predicted to have been fumigated, we ventilate *prior to* testing.
- Deploying a reliable binary classifier would reduce the overall costs of inspection, creating a more efficient and effective port.

### **Flowchart of Proposed Process**





- The misclassification cost of the Status Quo is
  M<sub>1</sub> = #P × \$C<sup>-</sup>
- The misclassification cost of the Binary Classifier is
  M<sub>2</sub> = #FN × \$C<sup>-</sup> + #FP × \$C<sup>+</sup>
  = (FNR × #P) × \$C<sup>-</sup> + (FPR × #N) × \$C<sup>+</sup>
  = (1 TPR) × #P × \$C<sup>-</sup> + FPR × #N × \$C<sup>+</sup>
- Given a predictive model, its optimal binary classifier is the classifier that minimizes the misclassification cost M<sub>2</sub>.

# The Improvement Curve

- We introduce the **improvement curve**, inspired by the theory of cost curves (Drummond & Holte, 2000).
- Improvement curves measure a model's performance
  - Over <u>all</u> possible class distributions (**#P** vs. **#N**)
  - Over <u>all</u> possible misclassification costs (\$C<sup>-</sup> and \$C<sup>+</sup>)
- Define the *improvement* to be  $I = (M_1 M_2) \div M_1$ .



### Definition of x-axis and y-axis

• The *x*-axis of the improvement curve is the following expression, denoted as *probability times cost*:

 $x = PC(+) = (\#P \times C^{-}) \div (\#P \times C^{-} + \#N \times C^{+}).$ 

 The y-axis is the *improvement*, the percentage reduction in misclassification cost by replacing the status quo with the model's optimal classifier:

y = 
$$I(x) = (M_1 - M_2) \div M_1$$
  
=  $TPR - [FPR \times (\#N \times $C^+) \div (\#P \times $C^-)].$ 

It is straightforward to show that  $0 \le x \le 1$  and  $0 \le y \le 1$ .

# Illustrating the Theory

- We built a simple predictive model based on a 4,200 container data set. The model's four features are:
  - Origin Country
  - Canadian Port of Arrival
  - HS section (e.g. Section 5 = Mineral Products)
  - HS chapter (e.g. Chapter 26 = Ores, Slag, Ash)
- The model consists of 2<sup>4</sup>=16 disjoint classes.
- The data was split 70/30 for Training/Testing.

#### **ROC Curve of Model**



## Improvement Curve of Model



#### **Improvement Curve Interpretation**

- Suppose #N ÷ #P = 6 and \$C<sup>-</sup> ÷ \$C<sup>+</sup> = 4. Then,
  x = PC(+) = (#P × \$C<sup>-</sup>) ÷ (#P × \$C<sup>-</sup> + #N × \$C<sup>+</sup>) = 0.4.
- From the improvement curve, we have y = 28% (Reading from the <u>Testing</u> Set).
- This simple 4-feature model would have reduced our misclassification cost by 28%.



## Conclusion

- The Improvement Curve does the following:
  - Addresses the limitations of ROC curves and the ROC AUC.
  - Measures performance over <u>all</u> possible values of PC(+).
  - Determines a simple condition for when the status quo should be retained.
  - Compares the optimal classifiers of two predictive models.
- The Improvement Curve is an evaluation metric that
  - Is extremely accessible to a non-specialist.
  - Has numerous applications to operations research beyond marine container inspection.