



Using AI to Identify and Expedite Environmental Goods at Customs

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Global Drivers and Operational Gaps

Global Frameworks for Green Trade



UN SDGs

Guide sustainable production, climate action, and international partnerships



Doha Declaration

Calls for lowering barriers and labeling to environmental goods



WCO Green Customs Action Plan

Greening Customs, enabling green trade, and fostering long-term innovation



ESCAP Framework

Four pillars: Digital procedures, environmental goods, green transport, NTMs

Operational Challenges at Customs



Classification Problem

Environmental and non-environmental goods have same HS codes.



Operational Inefficiency

Time-consuming and inconsistent manual identification processes



Policy Gap

60% of Customs do not have a trade facilitation policy for environmental goods
(WCO annual survey, 2022)



Technology Adoption

Only 10% of Customs use AI in operations
(WCO smart Customs survey, 2024)

Objective and Scope of the Study




Objective


To develop and evaluate an AI-based model for the identification and risk assessment of environmental goods at Customs.



Scope

1 

Uses Customs declaration data from Nepal.

2 

Applied AI models for classification and fraud detection of EGs

3 

Based on reference lists of EGs: APEC list of environmental goods and others

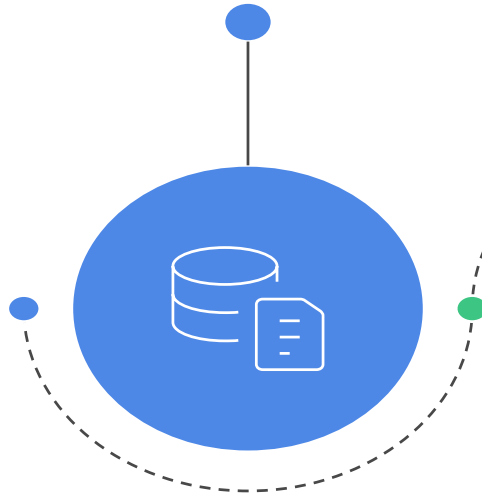
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Addresses both policy relevance and operational feasibility in Customs

Methodology

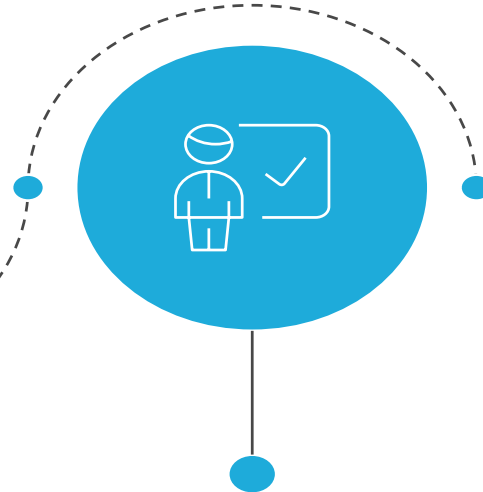
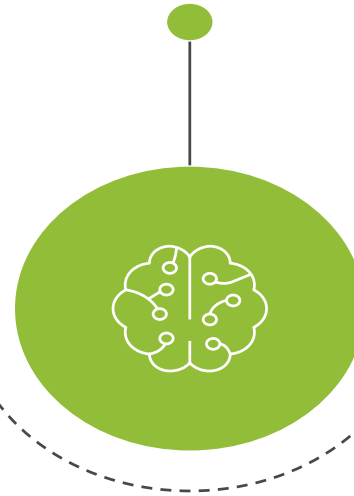
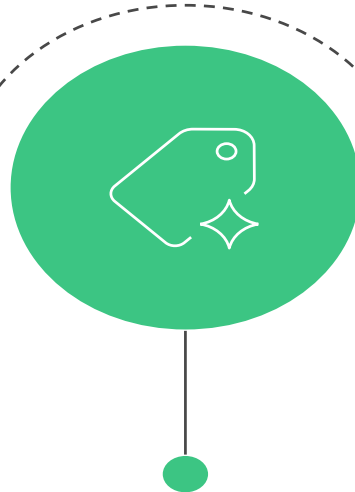
Data Collection

Data from Nepal Customs



AI Models Development

Model for environmental goods classification
Model for fraud detection



Labeling and Pre-processing

- Environmental goods labeled using three references
- Fraud labels derived from existing seizure database
- Feature engineering

Model Evaluation

Accuracy, Precision, Recall and custom FDR@k metrics

FDR@K: Fraud detection rate at certain inspection level thresholds

Key Findings

Environmental Goods Classification

- ❑ 6.4% of the dataset were environmental goods
- ❑ FastText outperformed all tested models
- ❑ Achieved 99.5% overall accuracy on test data
- ❑ Predicted 1,229 goods as green; 1,201 were correct → 97.7% precision
- ❑ Out of 1,281 actual green goods, 1,201 were correctly identified → 93.8% recall

Model Performance Metrics

99.5%

Accuracy

97.7%

Precision

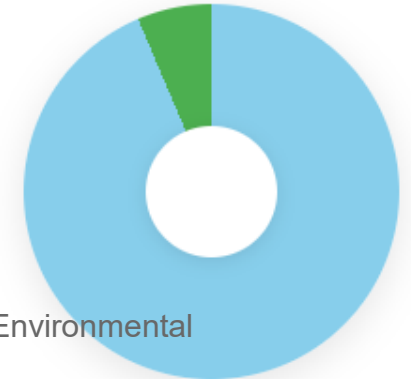
93.8%

Recall

95.7%

F1-Score

6.4% Environmental Goods



93.6% Non Environmental

Confusion Matrix

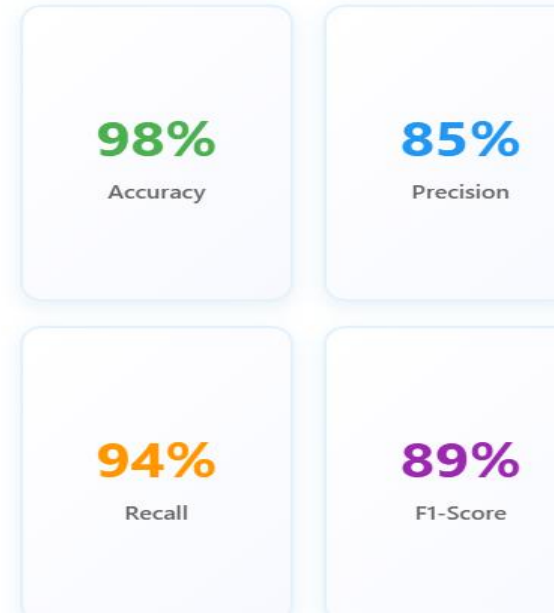
	Non Green	Green
Non Green	18691 (99.9%)	28 (0.1%)
Green	80 (6.2%)	1201 (93.8%)

Key Findings

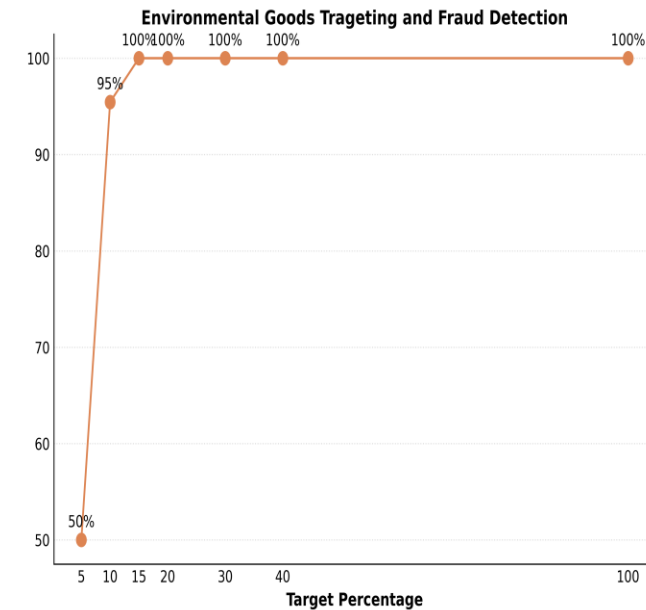
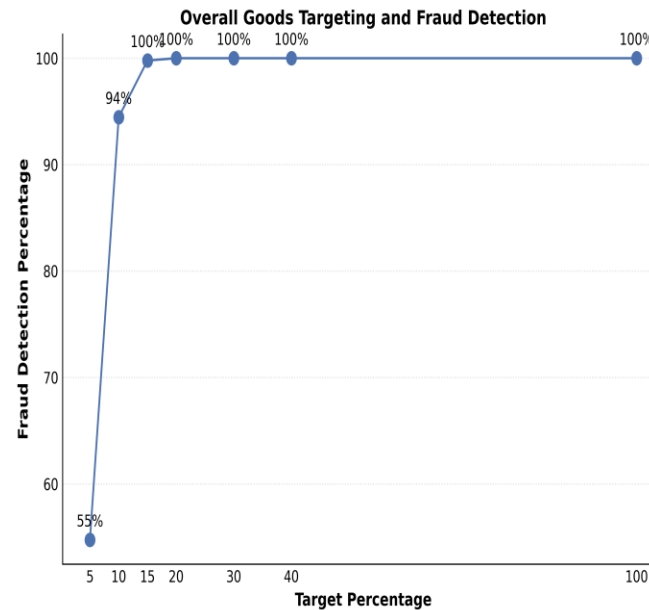
Fraud detection

- ❑ 8.9% of the transaction has some kind of fraud.
- ❑ CatBoost model outperformed all other tested
- ❑ Achieved 98% accuracy and 94% recall on test data
- ❑ Custom FDR@K metrics shows:
 - Targeting top 10% of risky consignments detects 94% of fraud
 - Targeting 15% detects nearly all fraud currently caught by Customs

Model Performance Metrics



91.1% Non Fraud 8.9% Fraud





Thank you

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