Exploratory Data Analysis & Hyperparameter Tuning

1. Exploratory Data Analysis (EDA)

During the EDA phase, we performed the following steps and observations:

- Class Distribution: Plotted counts of each "Action Taken" category (Logged, Blocked, Ignored) to confirm balance and identify any skew.
- Feature Overview: Displayed dataset info to verify non-null counts and data types across all 25 original columns.
- Statistical Summary:
- Source/Destination Port: range 1 024–65 535, mean ≈33 000, high variability.
- Packet Length: mean ≈780 bytes, σ ≈416, min 64, max 1 500.
- Anomaly Scores: uniform distribution on [0,100], mean \approx 50, σ \approx 29.
- Missing Values: Identified five columns (\sim 50% missing) and replaced nulls with explicit categories ("None", "No Data", "No Detection") to preserve all samples.
- Categorical Distributions: Visualized protocol usage (TCP/UDP/ICMP), traffic types (HTTP, DNS, FTP, etc.), and attack types (DDoS, Intrusion, Malware).

```
→ n iterations: 10
      n_required_iterations: 10
n_possible_iterations: 10
      min resources: 8
max resources: 4235
aggressive_elimination: False
factor: 2
      n_candidates: 529
      n_{\rm resources}: 8 Fitting 2 folds for each of 529 candidates, totalling 1058 fits
      iter: 1
n_candidates: 265
      n_resources: 16
Fitting 2 folds for each of 265 candidates, totalling 530 fits
      iter: 2
n_candidates: 133
n_resources: 32
Fitting 2 folds for each of 133 candidates, totalling 266 fits
     iter: 3
n_candidates: 67
n_resources: 64
Fitting 2 folds for each of 67 candidates, totalling 134 fits
   iter: 4
n_candidates: 34
n_resources: 128
Fitting 2 folds for each of 34 candidates, totalling 68 fits
   n\_resources\colon 256 Fitting 2 folds for each of 17 candidates, totalling 34 fits
   iter: 6
n_candidates: 9
n_resources: 512
Fitting 2 folds for each of 9 candidates, totalling 18 fits
   n\_resources\colon 1024 Fitting 2 folds for each of 5 candidates, totalling 10 fits
   n_-^{\rm resources}\colon 2048 Fitting 2 folds for each of 3 candidates, totalling 6 fits
```

2. Hyperparameter Tuning (Random Forest)

To optimize the Random Forest classifier efficiently, we applied successive halving search:

- Subsampling: Extracted 10% of training data to accelerate tuning.
- Halving Strategy: Used HalvingRandomSearchCV with factor=2 and 2-fold CV, progressively narrowing from 529 to 2 candidates over 10 iterations.
- Parameter Space: n_estimators \in [50,200], max_depth \in [5,20].
- Results: Best params = {n_estimators:178, max_depth:12}, CV accuracy = 62.91%.

This method reduced computational cost compared to a full search while effectively finding strong hyperparameters.

