

WorstCase MIA attack result report

Introduction

This report provides a summary of a series of simulated attack experiments performed on the model outputs provided. An attack model is trained to attempt to distinguish between outputs from training (in-sample) and testing (out-of-sample) data. The metrics below describe the success of this classifier. A successful classifier indicates that the original model is unsafe and should not be allowed to be released from the TRE.

In particular, the simulation splits the data provided into test and train sets (each will in- and out-of-sample examples). The classifier is trained on the train set and evaluated on the test set. This is repeated with different train/test splits a user-specified number of times.

To help place the results in context, the code may also have run a series of baseline experiments. In these, random model outputs for hypothetical in- and out-of-sample data are generated with identical statistical properties. In these baseline cases, there is no signal that an attacker could leverage and therefore these values provide a baseline against which the actual values can be compared.

For some metrics (FDIF and AUC), we are able to compute p-values. In each case, shown below (in the Global metrics sections) is the number of repetitions that exceeded the p-value threshold both without, and with correction for multiple testing (Benjamini-Hochberg procedure).

ROC curves for all real (red) and dummy (blue) repetitions are provided. These are shown in log space (as recommended here [ADD URL]) to emphasise the region in which risk is highest -- the bottom left (are high true positive rates possible with low false positive rates).

A description of the metrics and how to interpret them within the context of an attack is given below.

Experiment summary

```
n_reps: 10
reproduce_split: [5, 25, 36, 49, 64, 81, 100, 121, 144, 169]
p_thresh: 0.05
n_dummy_reps: 10
train_beta: 1
test_beta: 1
test_prop: 0.2
n_rows_in: 1094
n_rows_out: 186
training_preds_filename: None
test_preds_filename: None
output_dir: ./1
report_name: attack_output
include_model_correct_feature: False
sort_probs: True
mia_attack_model: <class
'sklearn.ensemble._forest.RandomForestClassifier'>
mia_attack_model_hyp: {'min_samples_split': 20, 'min_samples_leaf':
10, 'max_depth': 5}
attack_metric_success_name: P_HIGHER_AUC
attack_metric_success_thresh: 0.05
attack_metric_success_comp_type: lte
attack_metric_success_count_thresh: 5
attack_fail_fast: False
attack_config_json_file_name: None
target_path: None
```

Global metrics

```
null_auc_3sd_range: 0.3458 -> 0.6542
n_sig_auc_p_vals: 1
n_sig_auc_p_vals_corrected: 0
```

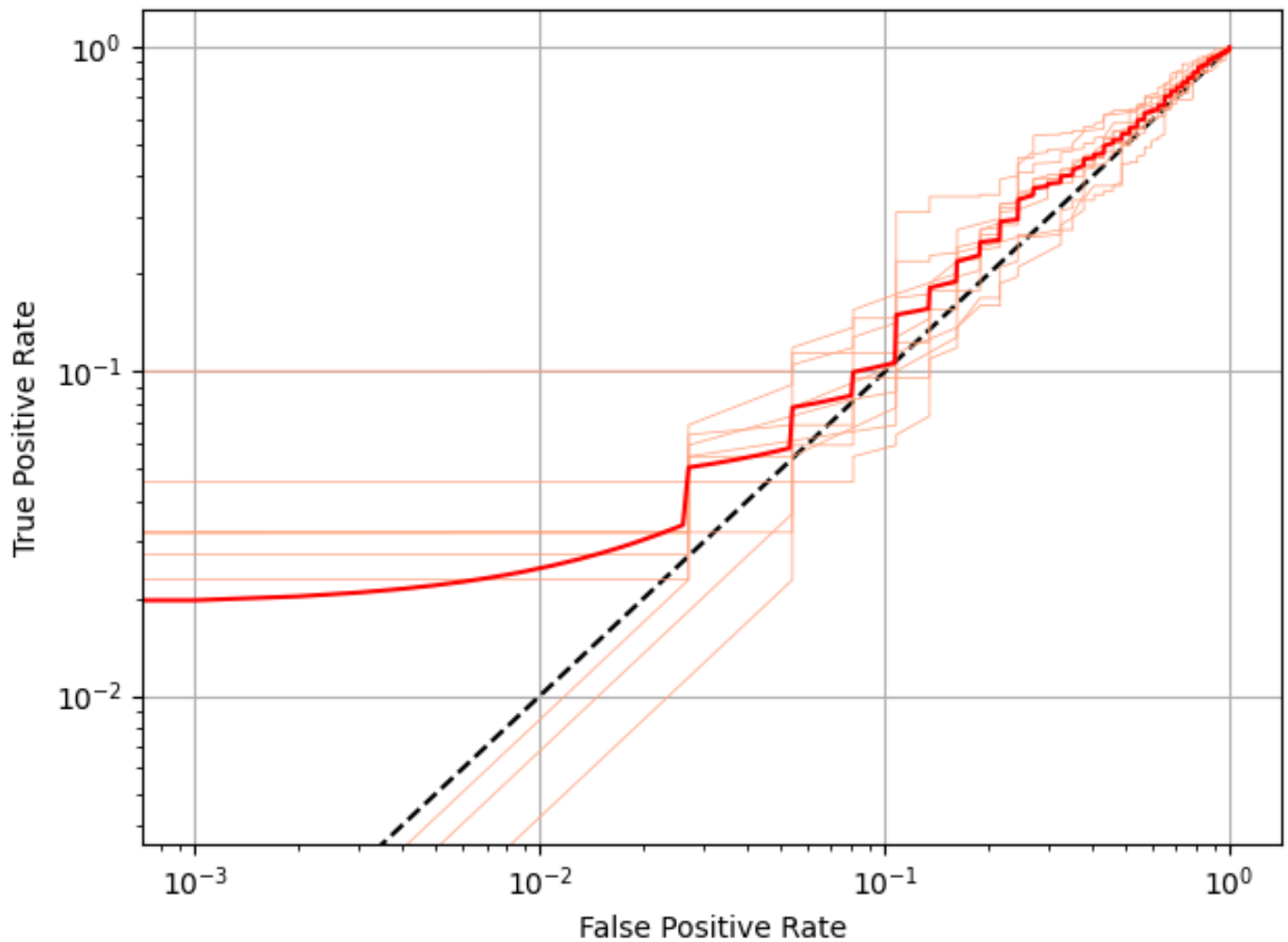
```
n_sig_pdif_vals: 1
n_sig_pdif_vals_corrected: 0
```

Metrics

The following show summaries of the attack metrics over the repetitions

```
AUC mean = 0.54, var = 0.0016, min = 0.48, max = 0.62
ACC mean = 0.86, var = 0.0000, min = 0.86, max = 0.86
Advantage mean = 0.00, var = 0.0000, min = 0.00, max = 0.00
FDIF01 mean = 0.06, var = 0.0045, min = -0.08, max = 0.19
PDIF01 mean = 0.30, var = 0.0417, min = 0.02, max = 0.78
TPR@0.1 mean = 0.10, var = 0.0012, min = 0.06, max = 0.17
TPR@0.01 mean = 0.02, var = 0.0005, min = 0.00, max = 0.08
TPR@0.001 mean = 0.02, var = 0.0004, min = 0.00, max = 0.07
TPR@1e-05 mean = 0.02, var = 0.0004, min = 0.00, max = 0.07
```

Log ROC



This plot shows the False Positive Rate (x) versus the True Positive Rate (y). The axes are in log space enabling us to focus on areas where the False Positive Rate is low (left hand area). Curves above the $y = x$ line (black dashes) in this region represent a disclosure risk as an attacker can obtain many more true than false positives. The solid coloured lines show the curves for the attack simulations with the true model outputs. The lighter grey lines show the curves for randomly generated outputs with no structure (i.e. in- and out-of- sample predictions are generated from the same distributions). Solid curves consistently higher than the grey curves in the left hand part of the plot are a sign of concern.

Glossary

AUC

Area

True Positive Rate (TPR)

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ACC

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Likelihood Ratio Attack Report

Introduction

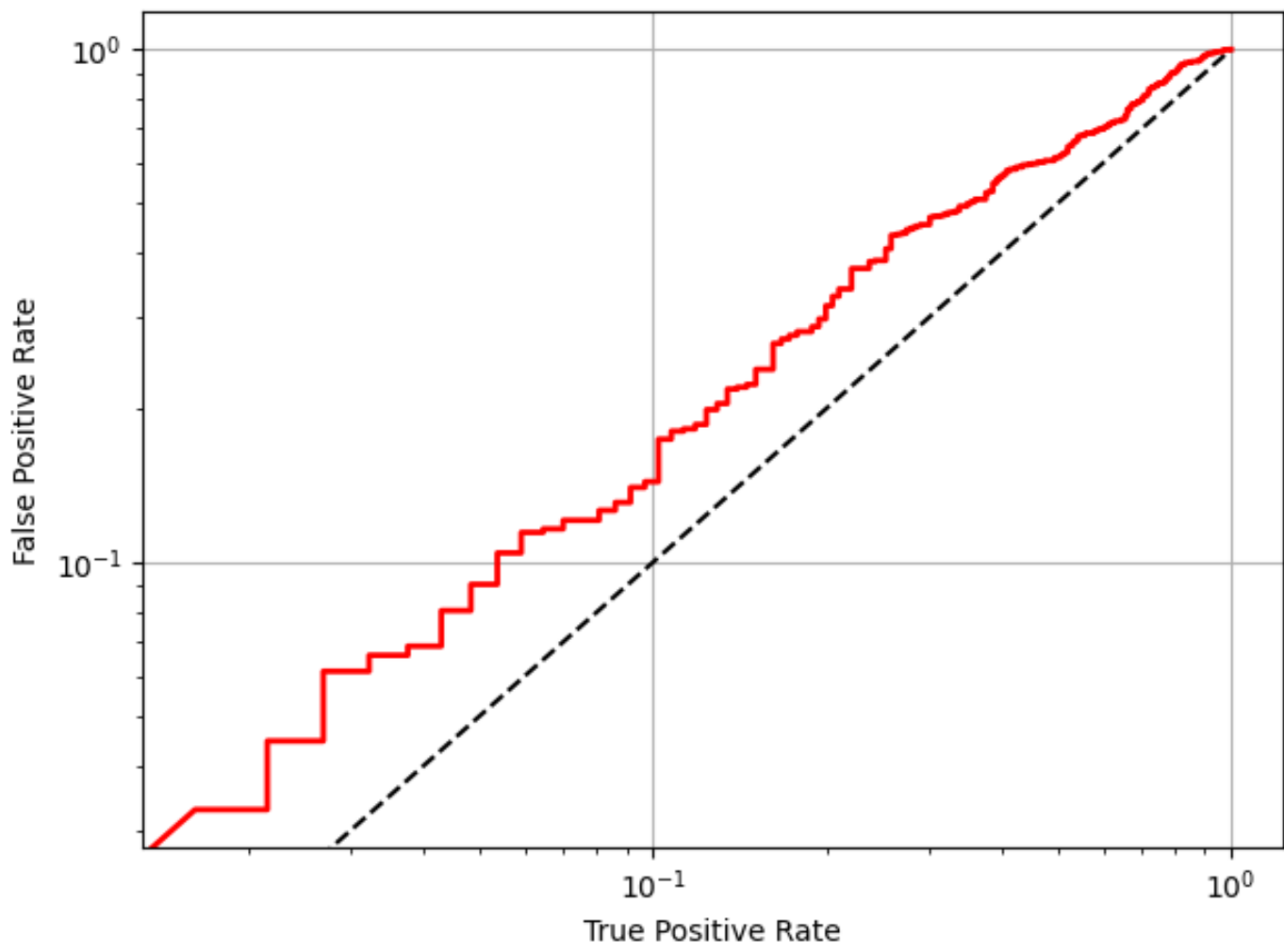
Metadata

```
n_shadow_models: 100
  p_thresh: 0.05
  output_dir: ./1
  report_name: attack_output
training_data_filename: train_data.csv
test_data_filename: test_data.csv
training_preds_filename: train_preds.csv
test_preds_filename: test_preds.csv
target_model: ['sklearn.svm']
target_model_hyp: {'C': 10, 'gamma': 'scale'}
attack_config_json_file_name: ./1\lira_config.json
n_shadow_rows_confidences_min: 10
shadow_models_fail_fast: False
target_path: None
PDIF_sig: Significant at p=0.05
AUC_sig: Significant at p=0.05
null_auc_3sd_range: 0.4312868175323611 -> 0.5687131824676389
```

Metrics

```
TPR: 0.3007
FPR: 0.1989
FAR: 0.1011
TNR: 0.8011
PPV: 0.8989
NPV: 0.1630
FNR: 0.6993
ACC: 0.3734
Flscore: 0.4507
Advantage: 0.1018
AUC: 0.6010
P_HIGHER_AUC: 0.0000
FMAX01: 0.9141
FMIN01: 0.7266
FDIF01: 0.1875
PDIF01: 0.0000
FMAX02: 0.9023
FMIN02: 0.7930
FDIF02: 0.1094
PDIF02: 8.4085
FMAX001: 1.0000
FMIN001: 0.5385
FDIF001: 0.4615
PDIF001: 7.7741
pred_prob_var: 0.1560
```

ROC Curve



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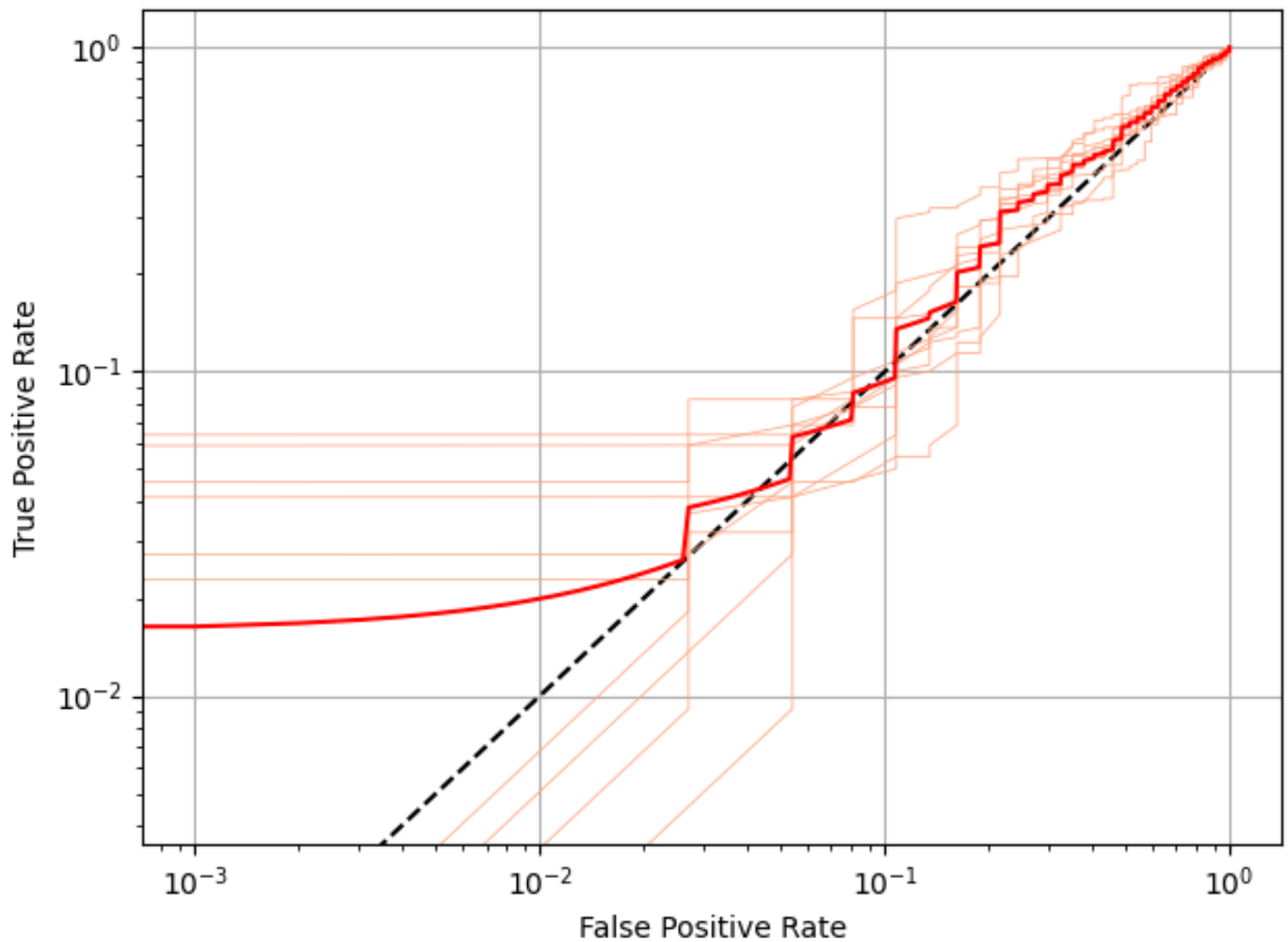
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FDIF01 mean = 0.02, var = 0.0027, min = -0.08, max = 0.12
PDIF01 mean = 0.43, var = 0.0355, min = 0.12, max = 0.78
TPR@0.1 mean = 0.09, var = 0.0014, min = 0.05, max = 0.17
TPR@0.01 mean = 0.02, var = 0.0003, min = 0.00, max = 0.05
TPR@0.001 mean = 0.02, var = 0.0003, min = 0.00, max = 0.05
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