



# TRUSTY SECURITY SOLUTIONS

*Securing IoT Environments*



# Trusty Security Solutions

## Background and Purpose

The Internet of Things (IoT) paradigm has evolved into a technology for building smart environments. Widespread use of the Internet of Things (IoT) in various domains like smart city, healthcare, supply chain and transportation make IoT a target of potentially malicious attacks which affect smart environment applications. Thus, there is a crucial need for Intrusion Detection Systems (IDSs) designed for IoT environments to mitigate IoT-related security attacks that exploit some of these security vulnerabilities.

## Mission Statement

Our mission is to thwart malicious attacks in IoT systems and mitigate these threats.

# About our Product

## IDS for IoT

An Intrusion Detection System(IDS) detects malicious attacks.

## ML as it's soul

It based on a ML model which upgrades and learns as it geos along to keep up with new threats.

**Faster**

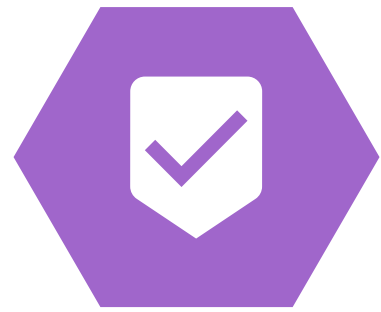
**Robust & Reliable**

# **Why TSS?**

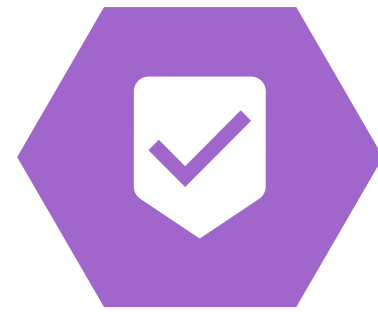
**High Detection Rate**

**Low False Positive  
Rate**

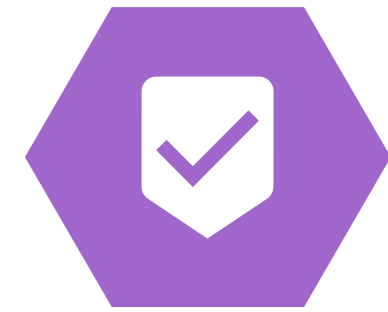
# About Our IDS



**Anomaly Based**



**Multiple Classifiers**



**Accurate & Precise**

# Methods of Intrusion Detection

## Anomaly Based

- **Less Data Intensive and hence Faster**
- **Can Detect new Threats**

Our Anomaly based IDS (A-IDS) is trained using NSL-KDD dataset and it also learns the behaviour of the network to predict whether a connection is normal or a threat. Thus, our A-IDS can check for threats and can even detect new threats, for which the model was not trained.

## Signature Based

- **More Data Intensive**
- **Cannot Detect new Threats**

Signature based IDS (S-IDS) compares the signature of a connection with signatures in a database to detect threats. Hence it is more data intensive and consumes more time. Also, it cannot predict a new threat unless the data set is updated and hence is not reliable for vulnerable IoT environment.

# How it Works



## **KNN**

Every connection is classified as normal or threat using KNN classifier and a probability is assigned to it.

## **Random Forest**

Every connection is classified as normal or threat using Random Forrest classifier and a probability is assigned to it.

## **Combine Results**

The results of KNN and RF are combined using a voting classifier to get a combined probability.

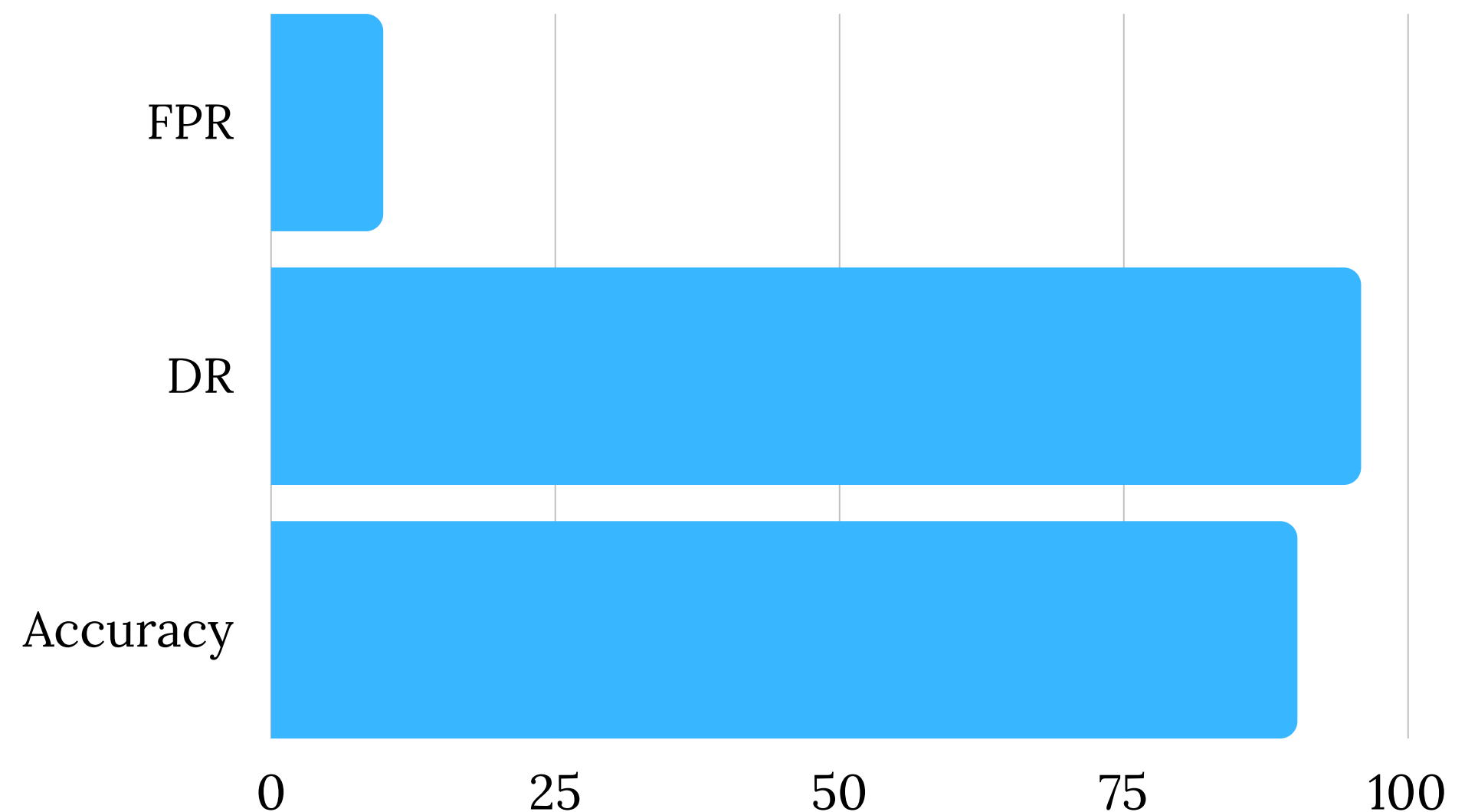
## **Result**

Based on a certain threshold probability, the connection is classified as either normal or threat.

# How our Model Performs

## Using Threshold for optimizing

For different thresholds, False Positive Rate (FPR), Detection Rate (DR) and Accuracy. Our aim is to maximise DR for a significantly less FPR. We have achieved 95.8% DR which is 90.2% accurate for FPR less than 10%.







# Thank you!

Feel free to approach us if  
you have any questions.

# Contact us

Reach out if you have any questions or clarifications

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