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Machine learning for a 5G future

# Message Collision Identification Approach Using Machine Learning

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## Introduction

#### **ITU CONTEXT**

- Extend terrestrial ADS-B coverage with LEO satellites.
- Inclusion to the WRC-19 Agenda.

#### ADS-B MESSAGES

 Message collisions in highly congested areas.

#### PURPOSE OF THIS STUDY Use of machine learning to estimate collisions



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### **Detection Problem**

 $\begin{cases} \omega_0: x = b & \longrightarrow hypothesis: noise \\ \omega_1: x = b + s & \longrightarrow hypothesis: identifiable \\ signal \end{cases}$ 

$$P_e(d) = p(d(X) \neq Y)$$

 $\begin{cases} H_0: X \in \omega_0, & X \sim p(X|\omega_0) \\ H_1: X \in \omega_1, & X \sim p(X|\omega_1) \end{cases}$ 

The general detection problem

- Minimize the error probability
- A dataset obtained by sampling is compared against a synthetic dataset from an idealized model.
- The distribution of the random proces
  X has to be known to use the former method.



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## **Machine Learning**

 A learning system that use past data to infer new solutions

Supervised ML

- Classification Problem
- SVM and kNN

#### **Features Extraction**

- Derived information from the input data relevant to the problem
- Make algorithms more efficient





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### **ML Methods**



**K-Nearest Neighbors** 

#### **Support Vector Machine**







# **Training Simulator**

Models the input data

Labels the input data



Generator

Supervisor

#### **ADS-B** Messages

- Duration of 120 µs
- Random periodicity with mean of 161 ms
- Manchester-coded
  - Two sates per bit







### **Signal Generator & Supervisor**

#### Scenario

- LEO satellite orbiting at 400km.
- 1000 planes uniformly distributed in it's footprint.
- Specific gain for each aircraft-satellite channel.

#### Classification

- Two classes defined
  - It is possible to decodify
  - It is not possible to decodify
- Depends on
  - Message collisions
  - Received power above receiver's sensitivity





## **ADS-B Signal Features Extraction**

#### **Feature Vector**

- Samples of the same length.
- The means of consecutive pair of samples.
- The variance of each set was calculated.
- The greatest and the smallest value were the two chosen features.
- The vector X was generated with the value of the two features for each sample.





### **ML Classification**



**kNN Performance** 

#### • $P_e = 0.059$

**SVM Performance** 

•  $P_e = 0.049$ 



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#### Results







### Conclusions

#### **Future Work**

- Improvements in signal modeling.
- Improvements in feature extraction.
- Experimental tests.
- Evaluation of reduction of energy due to improvement in efficiency.
- Work on development of International Recommendations.





### Conclusions

#### **Potential Impact**

- Using ML can increase overall efficiency.
  - The lifespan of satellites could be improved.
  - Launch costs could be significantly lower.
- Including details of useful ML techniques in International Recommendations could benefit all stackeholders.
- ML is an enabler of a new real-time global flight tracking system.
  - Impact in the safety, ecology and costs of flights.



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Thank you