



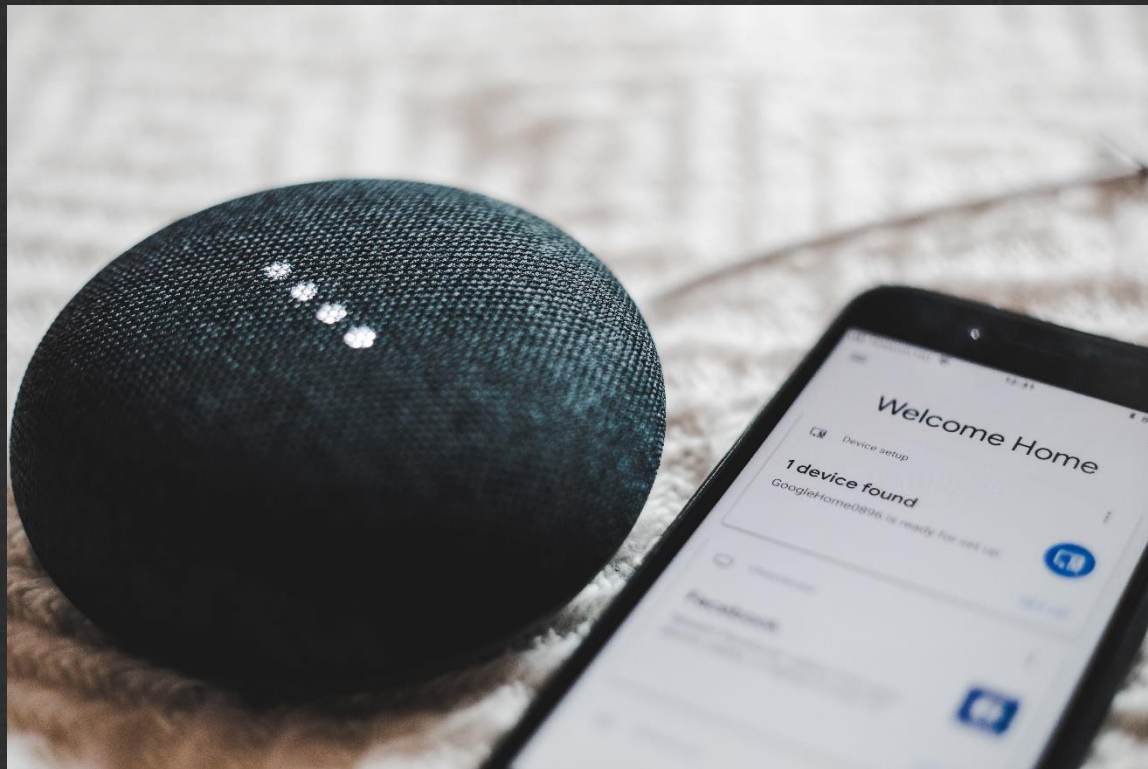
# Studio, progettazione e implementazione di un algoritmo genetico per l'individuazione di problemi di privacy in sistemi IoT

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**Candidato: Davide La Gamba**  
**Mat: 0512106292**



# Introduzione & Background





# Introduzione & Background

**10 miliardi di dispositivi IoT** nel 2021

**25,4 miliardi di dispositivi** nel 2030

**1.1 trilioni di dollari** spesi nel 2023 per l'IoT

**73.1 ZB** (1 Zettabyte = circa 1.000.000.000 Terabyte) di dati generati nel 2025

# Introduzione & Background

Lavoro esistente di Al-fuhaidi et al.  
con algoritmi genetici impiegati per  
problemi di sicurezza nell'IoT

13<sup>th</sup> International Conference on  
AEROSPACE SCIENCES & AVIATION TECHNOLOGY,  
ASAT- 13, May 26 – 28, 2009, E-Mail: [asat@mtc.edu.eg](mailto:asat@mtc.edu.eg)  
Military Technical College, Kobry Elkhobba, Cairo, Egypt  
Tel.: +(202) 24025292 – 24036138, Fax: +(202) 22621908



## Performance Evaluation of a Genetic Algorithm Based Approach to Network Intrusion Detection System

B. Abdullah\*, I. Abd-alghafar\*\*, Gouda I. Salama\*\* and A. Abd-alhafez\*\*

**Abstract:** The purpose of the work described in this paper is to provide an intrusion detection system (IDS), by applying genetic algorithm (GA) to network intrusion detection system. Parameters and evolution process for GA are discussed in detail and implemented. This approach uses information theory to filter the traffic data and thus reduce the complexity. We use a linear structure rule to classify the network behaviors into normal and abnormal behaviors. This approach applied to the KDD99 benchmark dataset and obtained high detection rate up to 99.87% as well as low false positive rate 0.003%. Finally the results of this approach compared with available machine learning techniques.

**Keywords:** Intrusion Detection System, Genetic Algorithm, Open Source Weka software.

### 1. Introduction

Internet and local area networks are expanding at an amazing rate in recent years, not just in the terms of size, but also in the terms of changing the services offered and the mobility of users that make them more vulnerable to various kinds of complex attacks. While we are benefiting from the convenience that new technology has brought us, computer systems are exposed to increasing number and complexity of security threats.

Of particular importance, thus, is the ability of applying rapidly new network security policies in order to detect and react as quickly as possible to the occurring attacks. Different techniques have been developed and deployed to protect computer systems against network attacks (anti-virus software, firewall, message encryption, secured network protocols, password protection). Despite all the efforts, it is impossible to have a completely secured system. Therefore, intrusion detection is becoming an increasingly important technique that monitors network traffic and identifies network intrusions such as anomalous network behaviors, unauthorized network access, or malicious attacks to computer systems.

Intrusion detection systems are typically classified with respect to placement as: host based or network based [1]. A host based IDS will monitor resources such as system logs, file systems and disk resources; whereas a network based intrusion detection system monitors the data passing through the network.

There are two general categories of intrusion detection systems (IDSs) [2] as: misuse detection and anomaly based. Misuse detection systems are most widely used and they detect

\* Yemeni Armed Forces , [belalarh@gmail.com](mailto:belalarh@gmail.com)

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problemi di sicurezza nell'IoT

Obiettivo: estendere il lavoro in  
esame, aumentando la  
capacità di individuazione di  
problemi di privacy

Risultati: regole di  
classificazione con Detection  
Rate migliorato e maggior  
varietà di attacchi identificati



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<https://github.com/davide-lagamba>

Algoritmo genetico

per problemi di privacy in sistemi IoT

Davide La Gamba

# Descrizione dell'approccio utilizzato

Individuazione dataset (KDDCUP99)  
e lavoro di riferimento (Al-fuhaidi et al.)

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Scelta delle codifiche



# Descrizione dell'approccio utilizzato

Individuazione dataset (KDDCUP99)  
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Progettazione algoritmo genetico e obiettivi

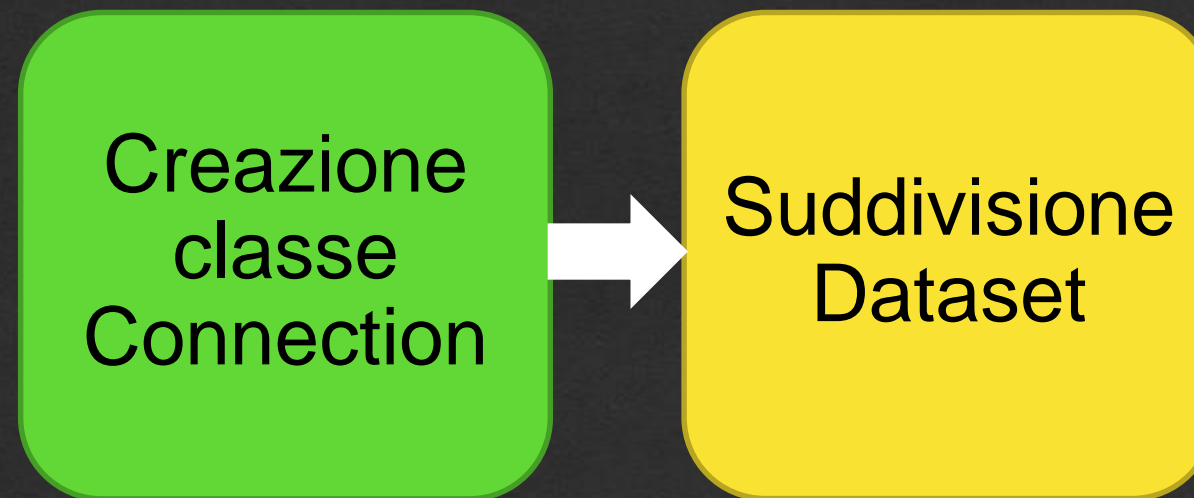
Scelta delle codifiche

Scelta degli altri parametri e degli operatori

# Descrizione dell'approccio utilizzato

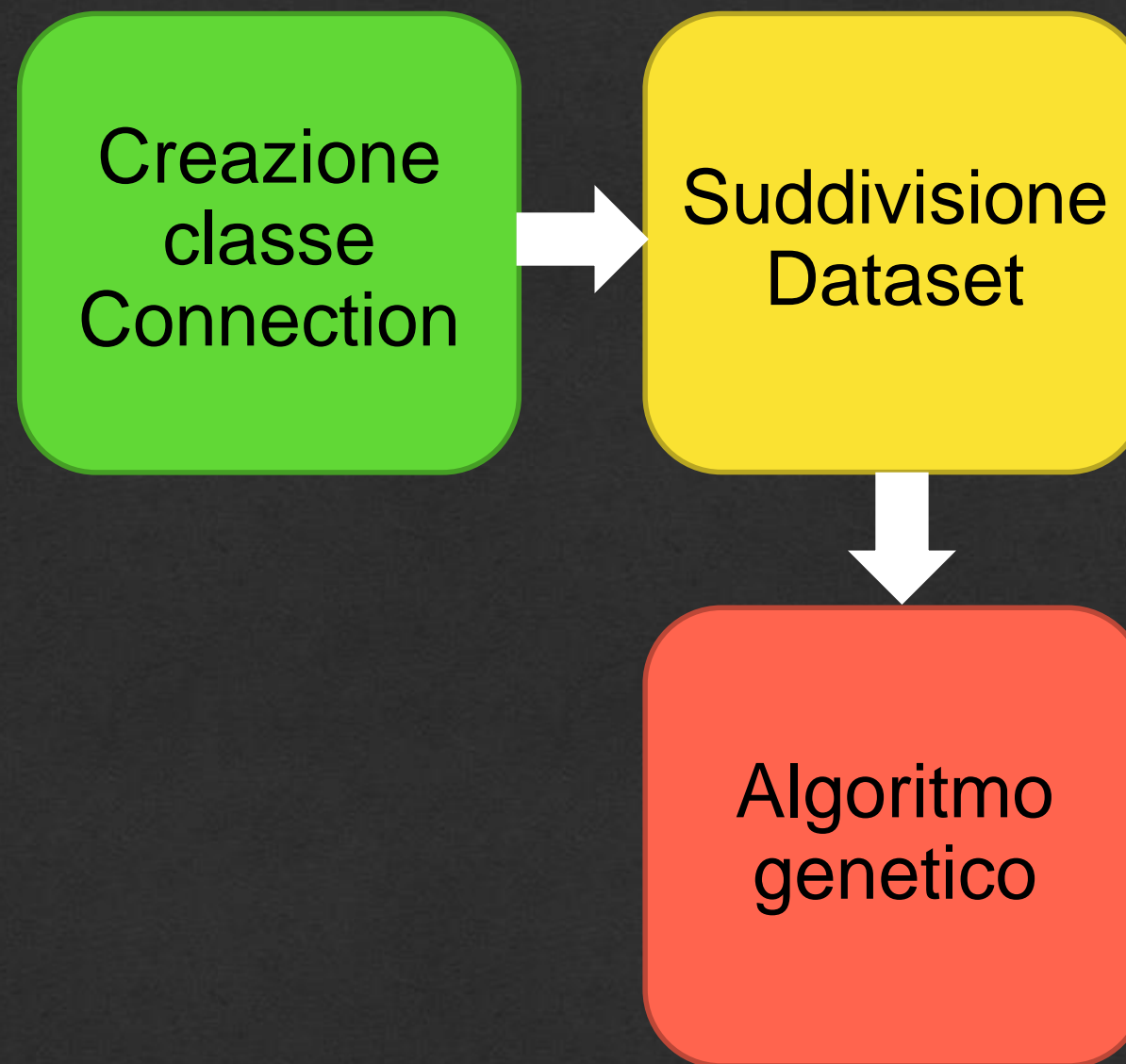
Creazione  
classe  
Connection

# Descrizione dell'approccio utilizzato

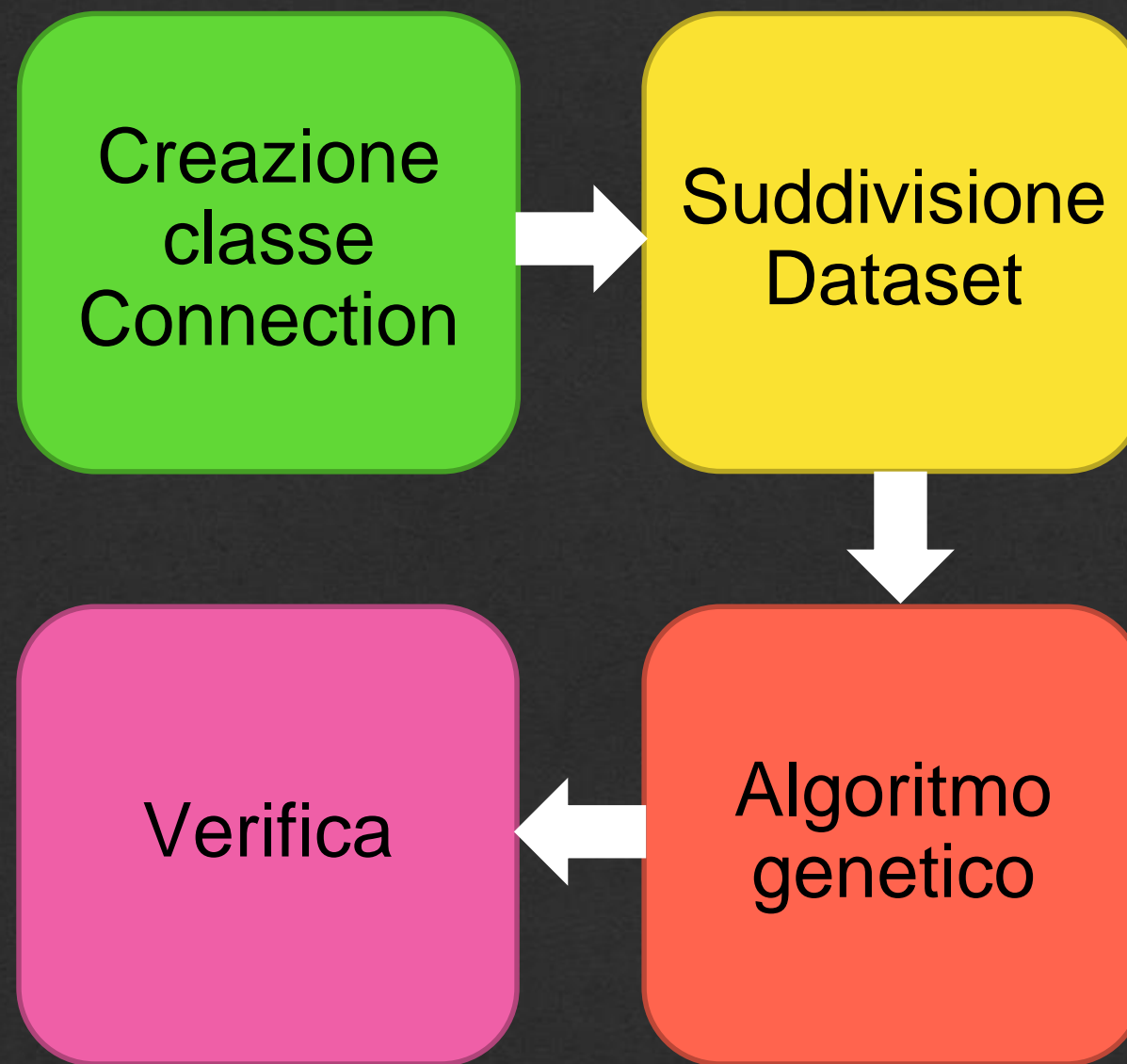




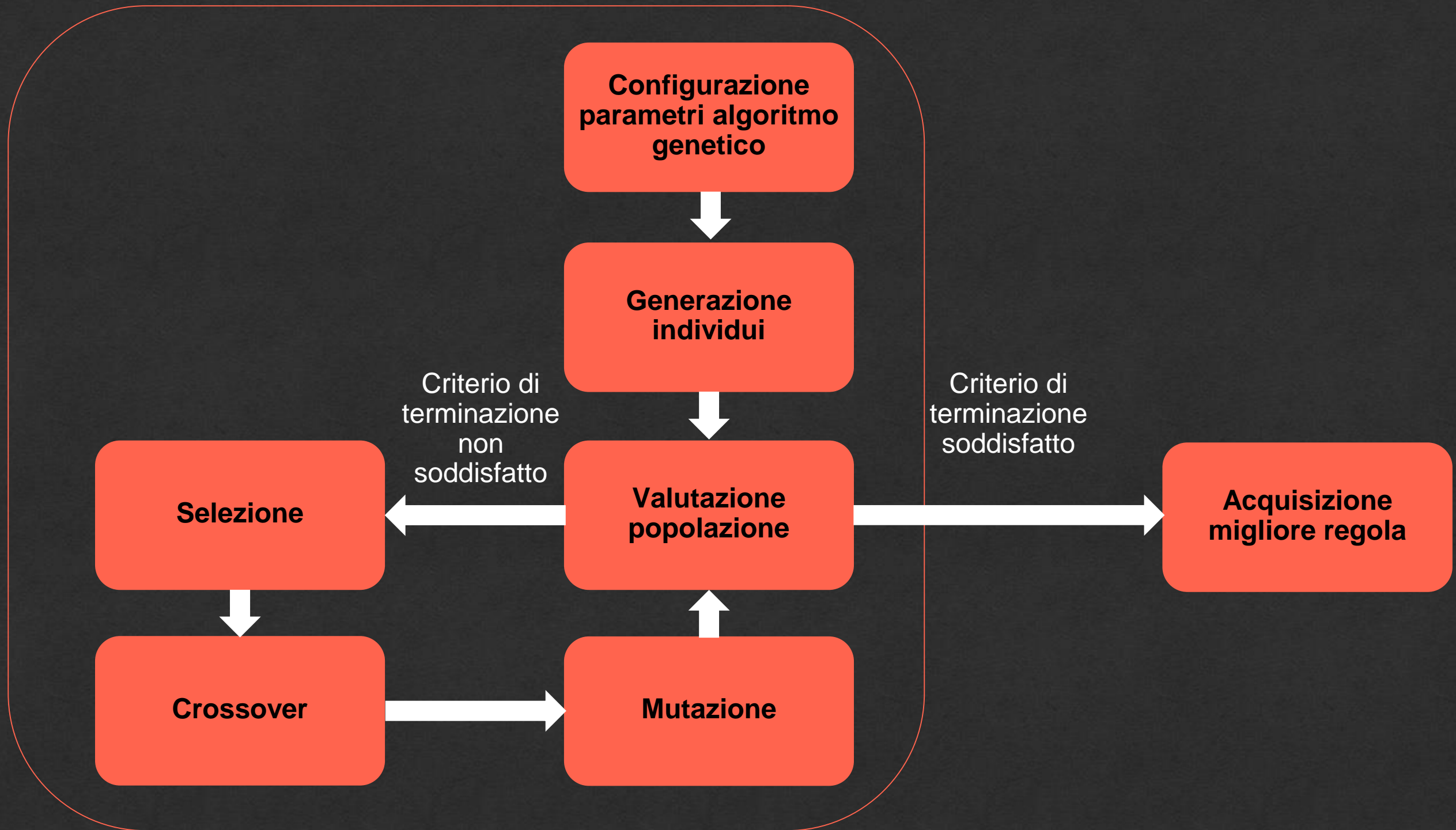
# Descrizione dell'approccio utilizzato



# Descrizione dell'approccio utilizzato



# Descrizione dell'approccio utilizzato





# Descrizione dell'approccio utilizzato

Individuo = serie di geni (valori numerici) raccolti in cromosomi con valori di min e max

Rappresentano sia i valori dei parametri da analizzare per ogni connessione sia i versi delle disuguaglianze

*if (duration <= 5855 AND protocolType == icmp AND service == ecr\_i  
AND flag == SF AND ... AND sameSrvRate <= 1.00 AND diffSrvRate  
<= 0.90 AND srvDiffHostRate <= 1.00)  
{attacco trovato}*

# Descrizione dell'approccio utilizzato

$$\text{Funzione di fitness} = \frac{ab}{A} - \frac{a}{B}$$

*ab = attacchi correttamente individuati*  
*a = connessioni erroneamente individuate*  
*A = totale degli attacchi*  
*B = totale delle connessioni normali*

Algoritmo di selezione: Elite Selector

Algoritmo di crossover: Single Point Crossover (probabilità 90%)

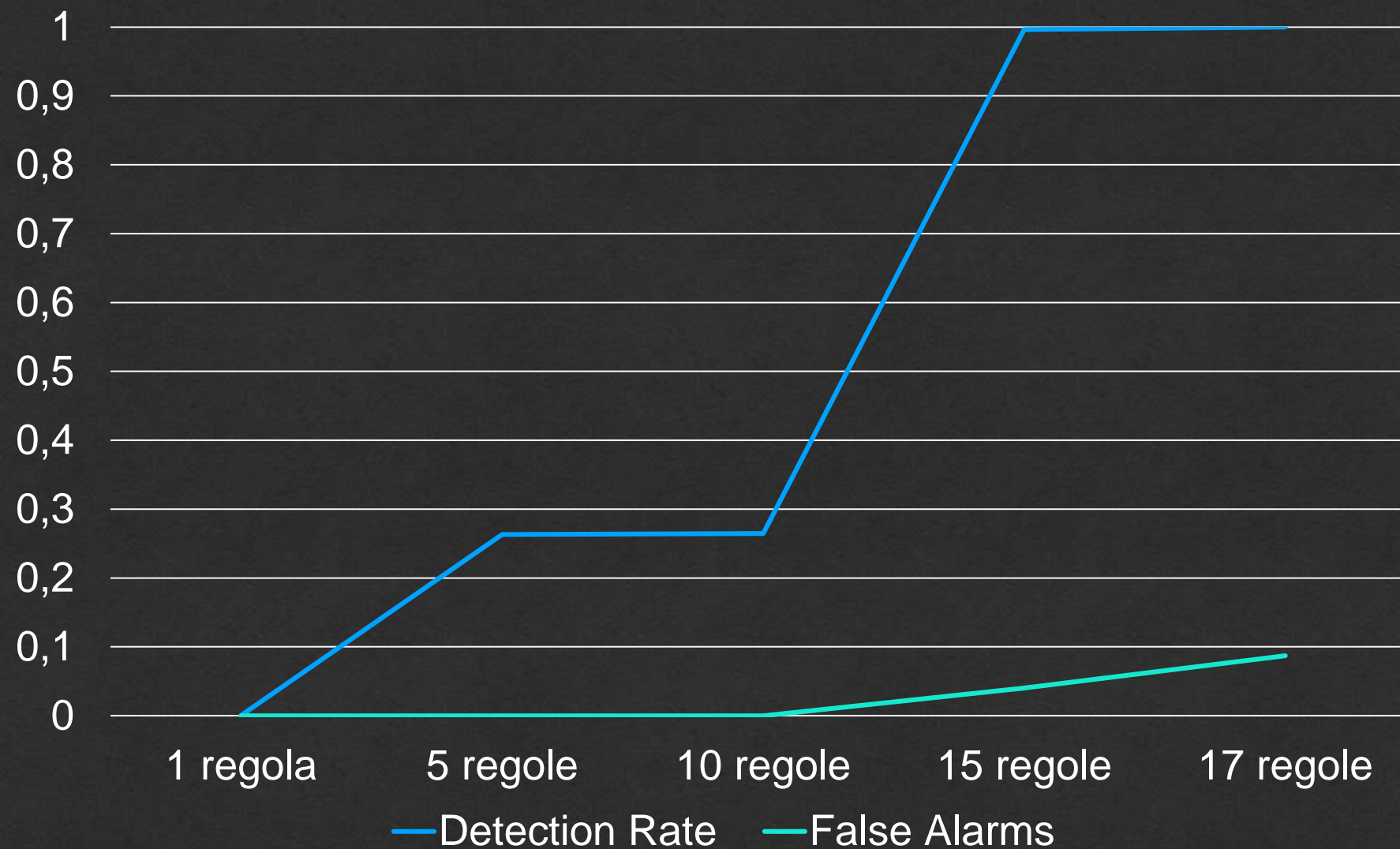
Probabilità di mutazione: 10%

1000 (o 5000) individui

1000 generazioni

# Sperimentazione e Risultati

Andamento metriche



Numero di regole	17
Detection Rate	0.99989
F-Measure	0.98925
Precision	0.97884
Accuracy	0.98259
Specificity	0.91288
MCC	0.94500
False Alarms	0.08711



# Sperimentazione e Risultati

Metrica	Algoritmo Genetico Soluzione 100%	Naive Bayes	Random Forest
Detection Rate	0.999	0,992	1,000

# Sperimentazione e Risultati

Metrica	Algoritmo Genetico Soluzione 100%	Algoritmo Genetico Soluzione 10%	Algoritmo Genetico di Al-fuhaidi et al.
Detection Rate	99.989%	99.938%	99.87%
False Alarms	08.711%	02.572%	0.003%

# Conclusioni e Sviluppi futuri

## Pro:

- Flessibilità nella codifica degli individui
- Alta comprensibilità



# Conclusioni e Sviluppi futuri

## Pro:

- Flessibilità nella codifica degli individui
- Alta comprensibilità

## Contro:

- Delicata fase di scelta dei parametri
- Prestazioni leggermente inferiori

# Conclusioni e Sviluppi futuri



**Implementazione  
generica da utilizzare  
in un contesto reale**

# Conclusioni e Sviluppi futuri



**Implementazione  
generica da utilizzare  
in un contesto reale**



**Classificazione  
specifica degli  
attacchi**



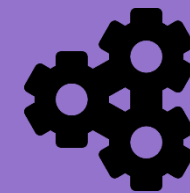
# Conclusioni e Sviluppi futuri



**Implementazione  
generica da utilizzare  
in un contesto reale**



**Classificazione  
specifica degli  
attacchi**



**Utilizzare  
funzione multi-  
obiettivo**

# Conclusioni e Sviluppi futuri

Vi ringrazio per l'attenzione

QR Code Repository GitHub



QR Code PDF Tesi



# Conclusioni e Sviluppi futuri



<https://www.flaticon.com/free-icons/code> Code icons created by Freepik - Flaticon



<https://www.flaticon.com/free-icons/malware> Malware icons created by Smashicons - Flaticon



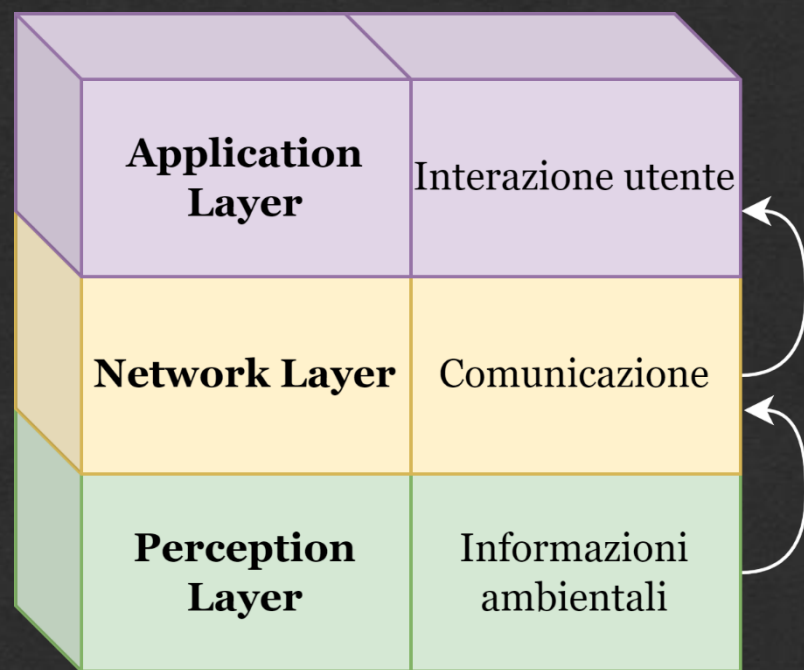
<https://www.flaticon.com/free-icons/world> World icons created by Hilmy Abiyyu A. - Flaticon



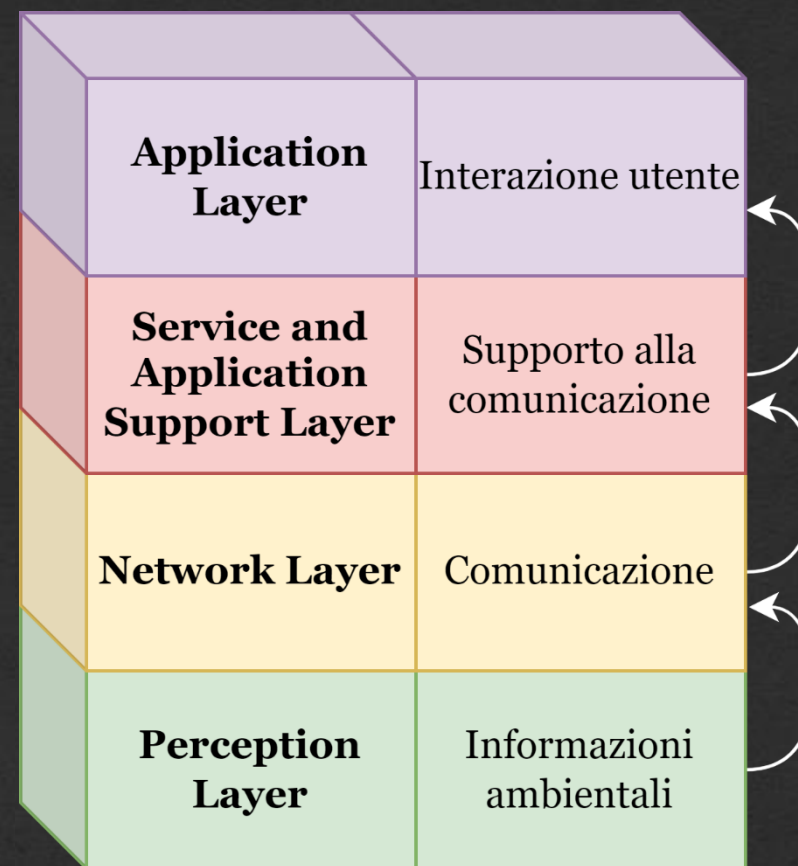
<https://www.flaticon.com/free-icons/gear> Gear icons created by Freepik - Flaticon



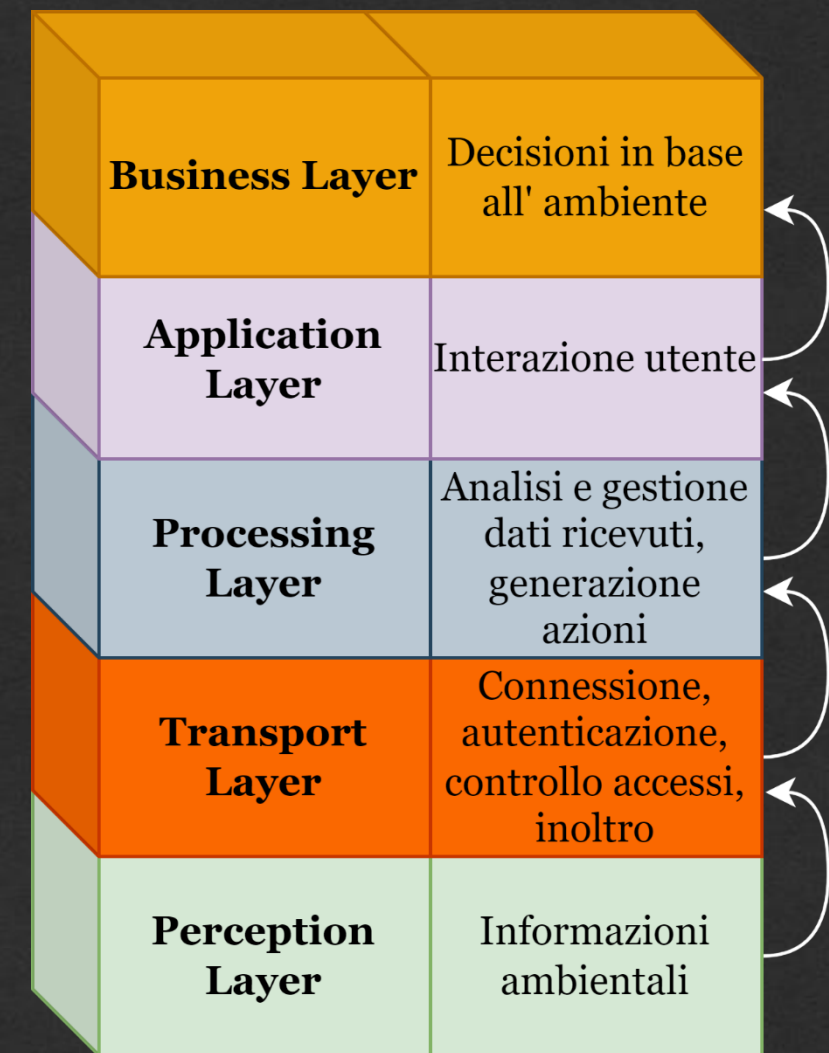
# Introduzione & Background



Tre livelli



Quattro livelli



Cinque livelli

# Introduzione & Background

- Individuazione di vulnerabilità
- Evoluzione di malware
- Feature Selection
- Applicazioni in sistemi blockchain
- Individuazione di intrusioni

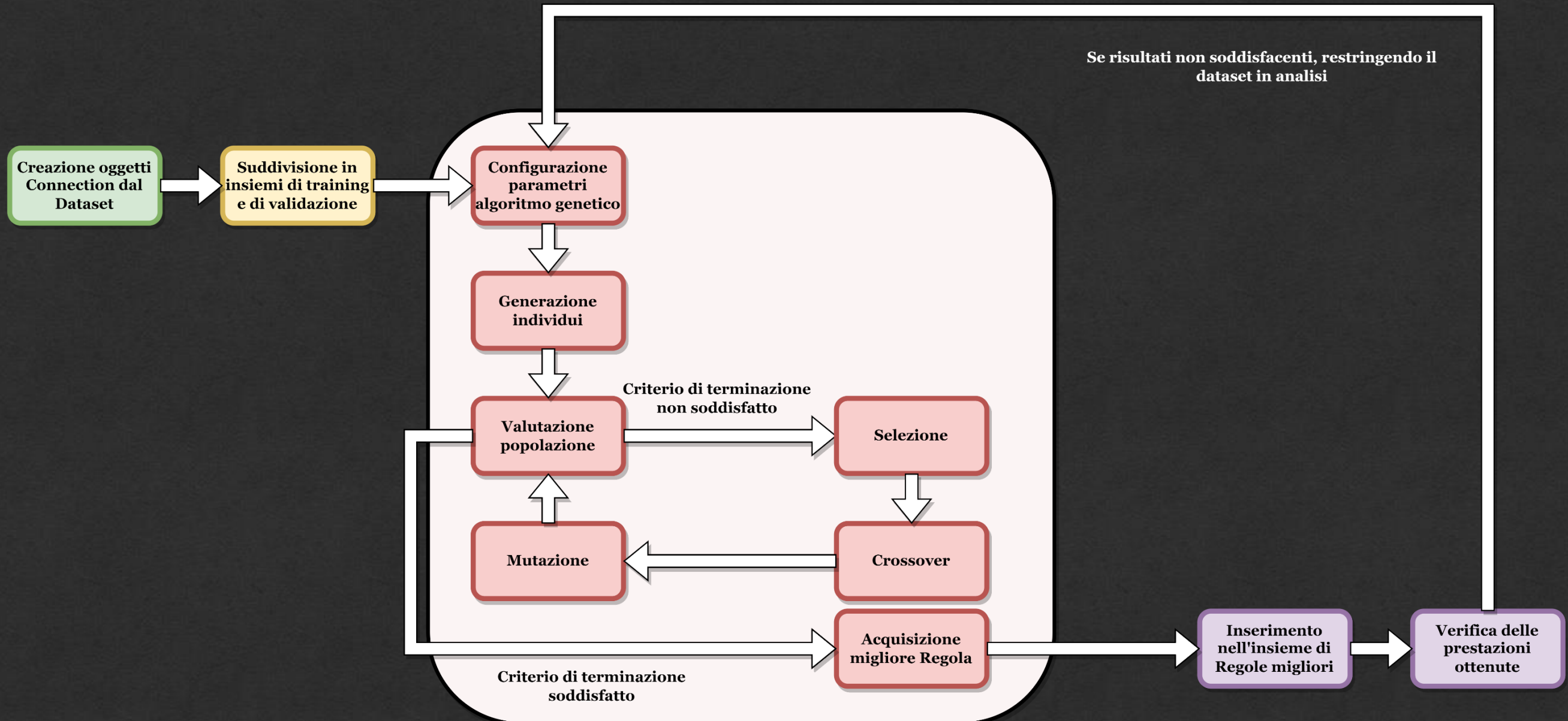
# Descrizione dell'approccio utilizzato

Classe  
Connection

Suddivisione  
Dataset

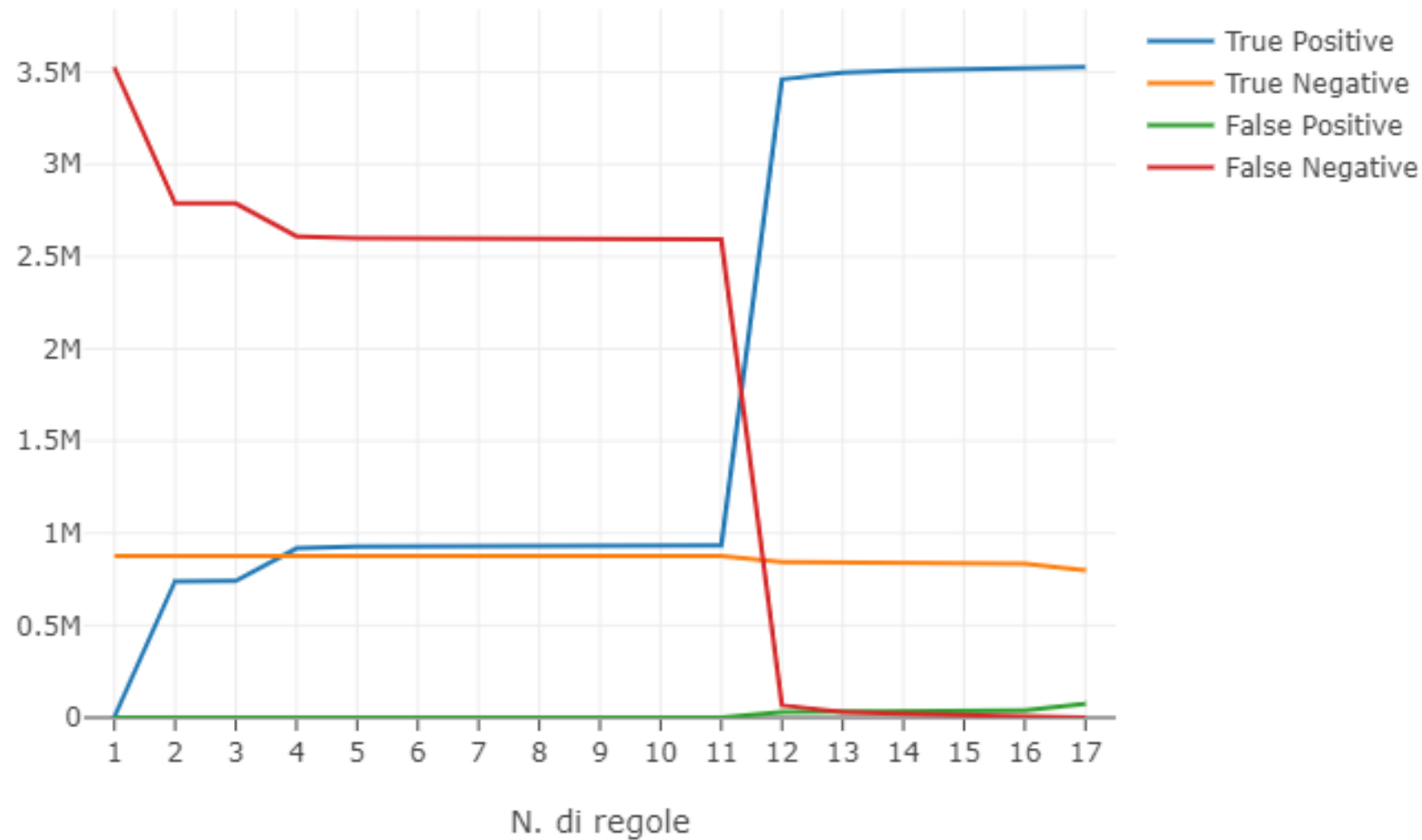
Algoritmo  
Genetico

Verifica





# Sperimentazione e Risultati



Numero di regole	17
True Positive	3528525
True Negative	799237
False Positive	76267
False Negative	382