

CONFERENCIA DE POSGRADO

Clasificación de ciberataques mediante un modelo de ensemble learning



Andrés Caro Lindo
Director
Cátedra VIEWNEXT-UEX

Presentación



Introducción

MACHINE LEARNING

¿Extracción de **conocimiento** a partir de

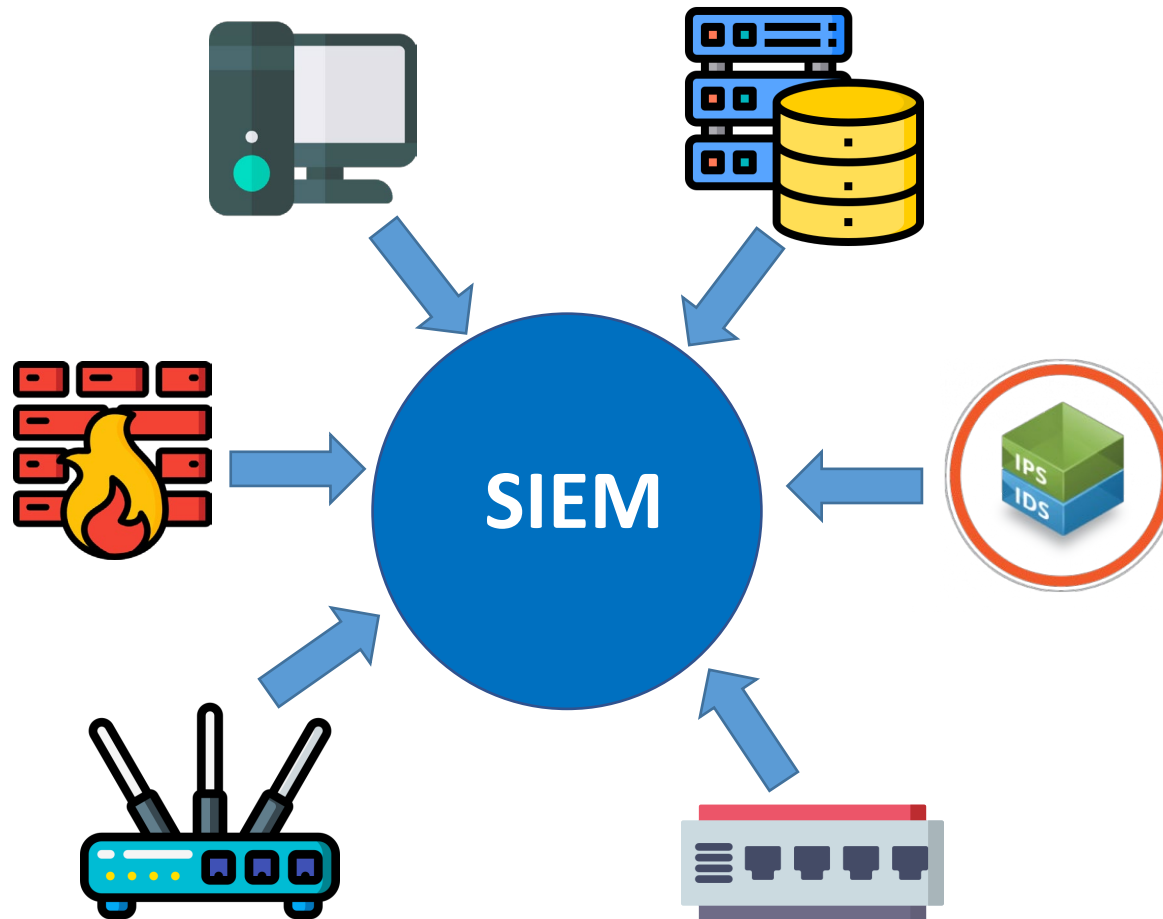
DATASETS

fuentes de datos procedentes de la

SIEM

monitorización de eventos de seguridad?

Introducción



Dataset



UNSW-NB15

Cyber Range Lab at UNSW Canberra

	Train	Train (%)	Test	Test (%)
Analysis	2000	1.14	677	0.82
Backdoor	1746	1.00	583	0.71
DoS	12264	6.99	4089	4.97
Exploits	33393	19.04	11132	13.52
Fuzzers	18184	10.37	6062	7.36
Generic	40000	22.81	18871	22.92
Normal	56000	31.94	37000	44.94
Reconnaissance	10491	5.98	3496	4.25
Shellcode	1133	0.65	378	0.46
Worms	130	0.07	44	0.05
TOTAL	175341	100%	82332	100%

Dataset



Técnicas de balanceo

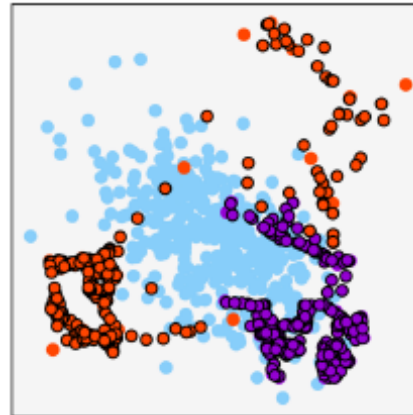
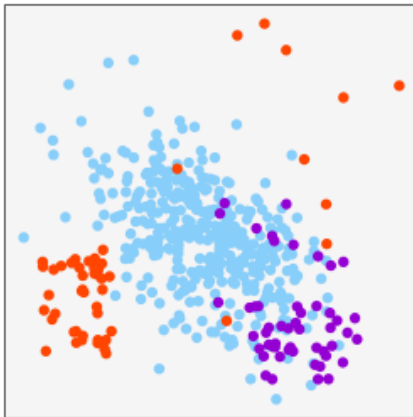
- **OVERSAMPLING** - GENERAR MÁS MUESTRAS, EN LAS CLASES MINORITARIAS
 - **SMOTE** (SYNTHETIC MINORITY OVERSAMPLING TECHNIQUE)
 - SMOTE-BORDERLINE1 (**BLINE-1**)
 - ADAPTATIVE SYNTHETIC (**ADASYN**)
- **UNDERSAMPLING** - REDUCIR EL NÚMERO DE MUESTRAS
 - RANDOM UNDERSAMPLING (**RU**)
 - CLUSTER CENTROIDS (**CC**)
 - NEAR MISS (NM)
 - REPEATED EDITED NEAREST NEIGHBOUR (**RENN**)
 - TOMKEK LINKS (**TKL**)

Dataset



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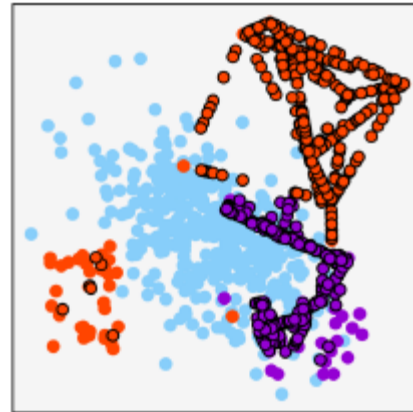
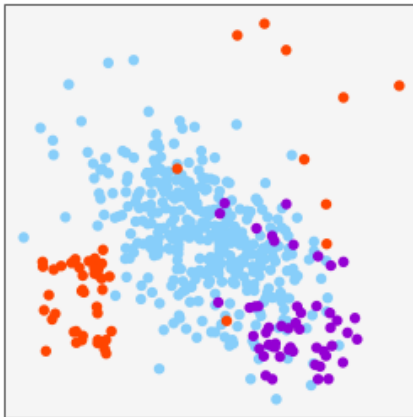


Dataset



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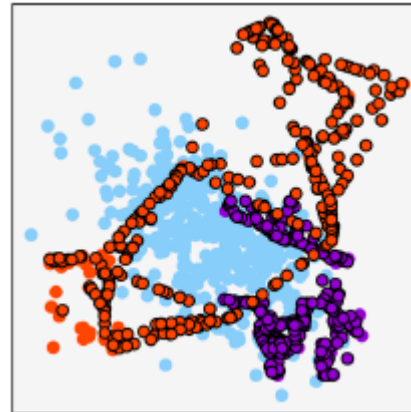
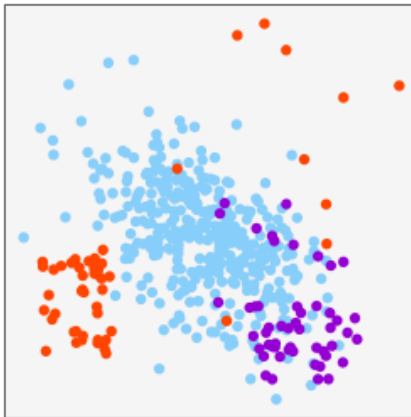


Dataset



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Dataset



Técnicas de balanceo

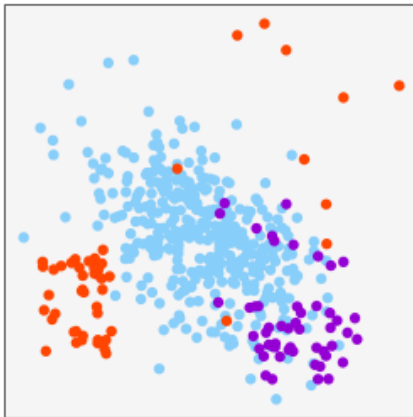
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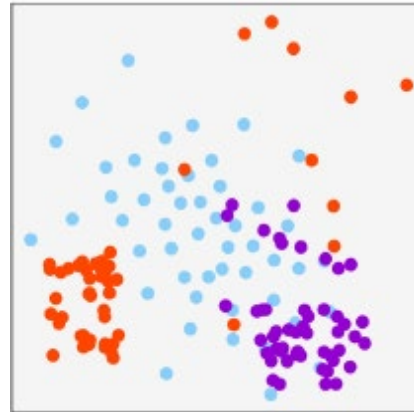
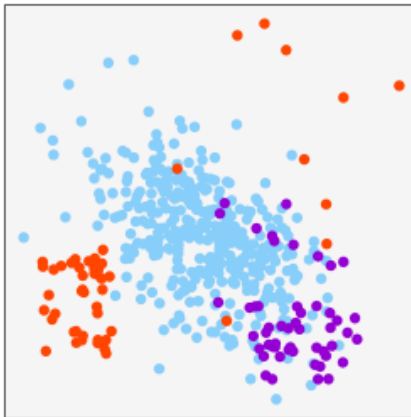


Dataset



Técnicas de balanceo

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 - CLUSTER CENTROIDS (CC)

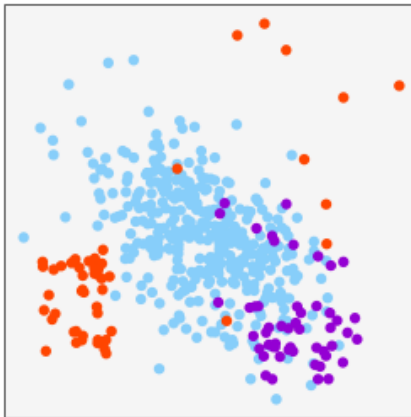


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Técnicas de balanceo

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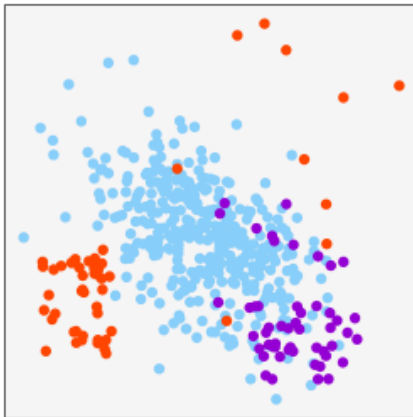


Dataset



Técnicas de balanceo

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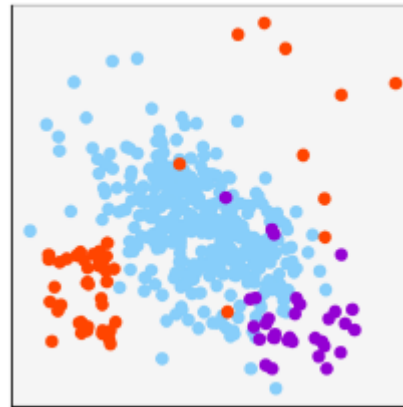
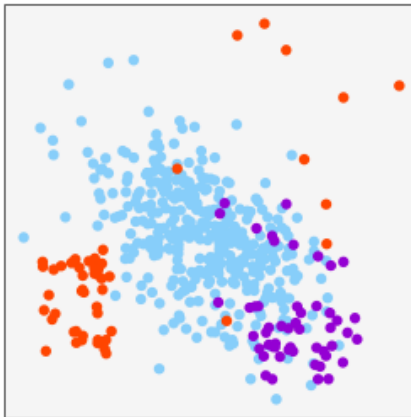


Dataset



Técnicas de balanceo

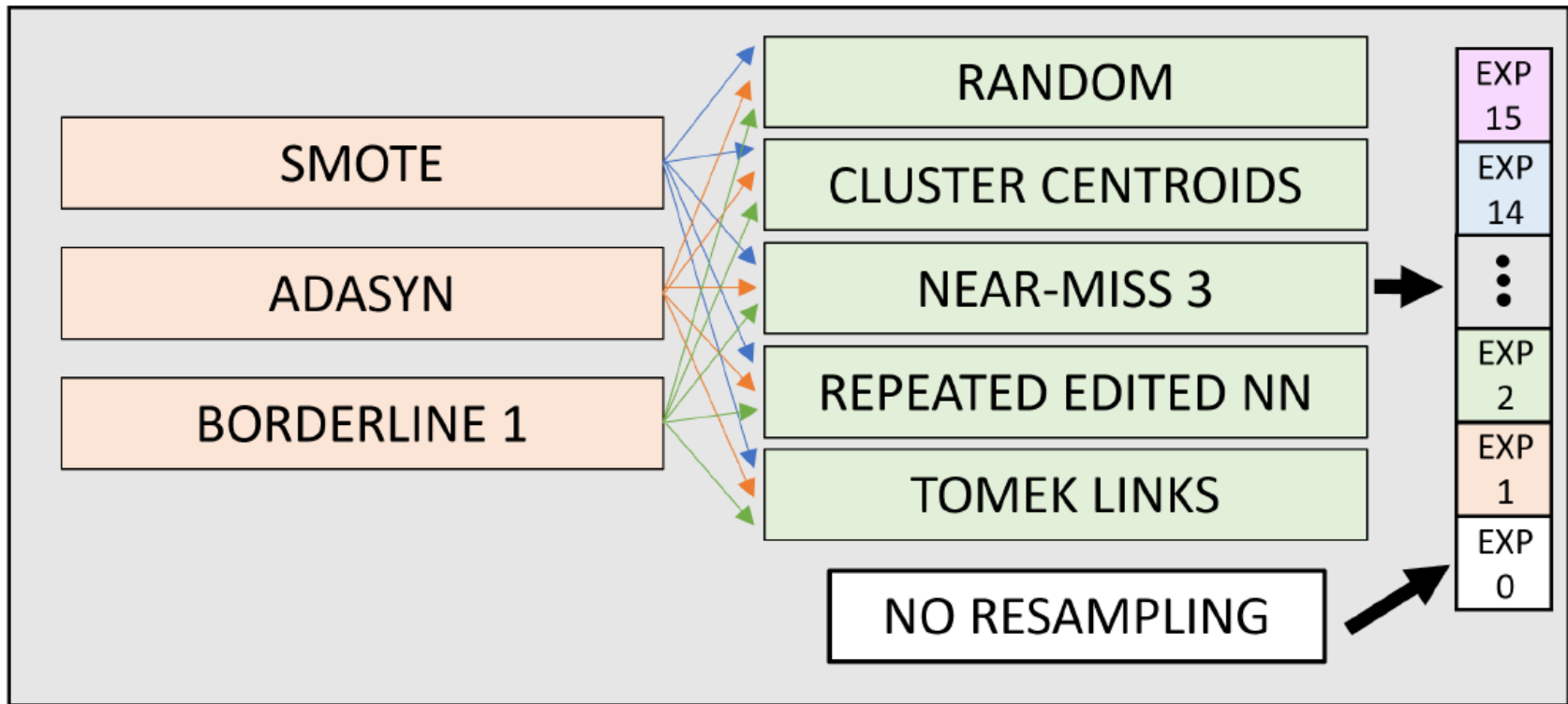
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Experimentos

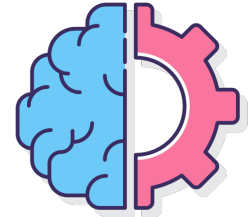


Técnicas de balanceo



Experimentos

Machine Learning



- SUPPORT VECTOR MACHINE (SVM)
- DECISION TREES (DT)
- K NEAREST NEIGHBORS (KNN)
- NAÏVE BAYES (NB)



UNSW-
NB15

SUPPORT VECTOR MACHINE

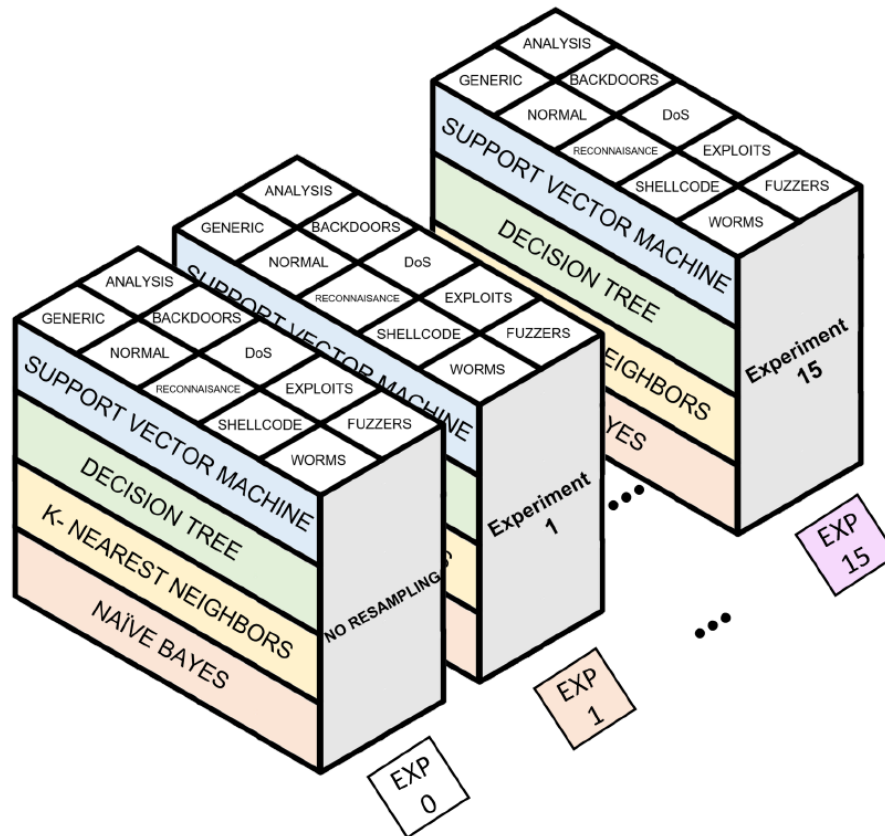
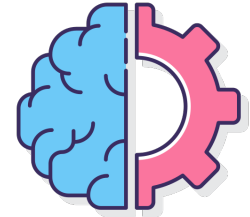
DECISION TREE

K-NEARES NEIGHBORS

NAÏVE BAYES

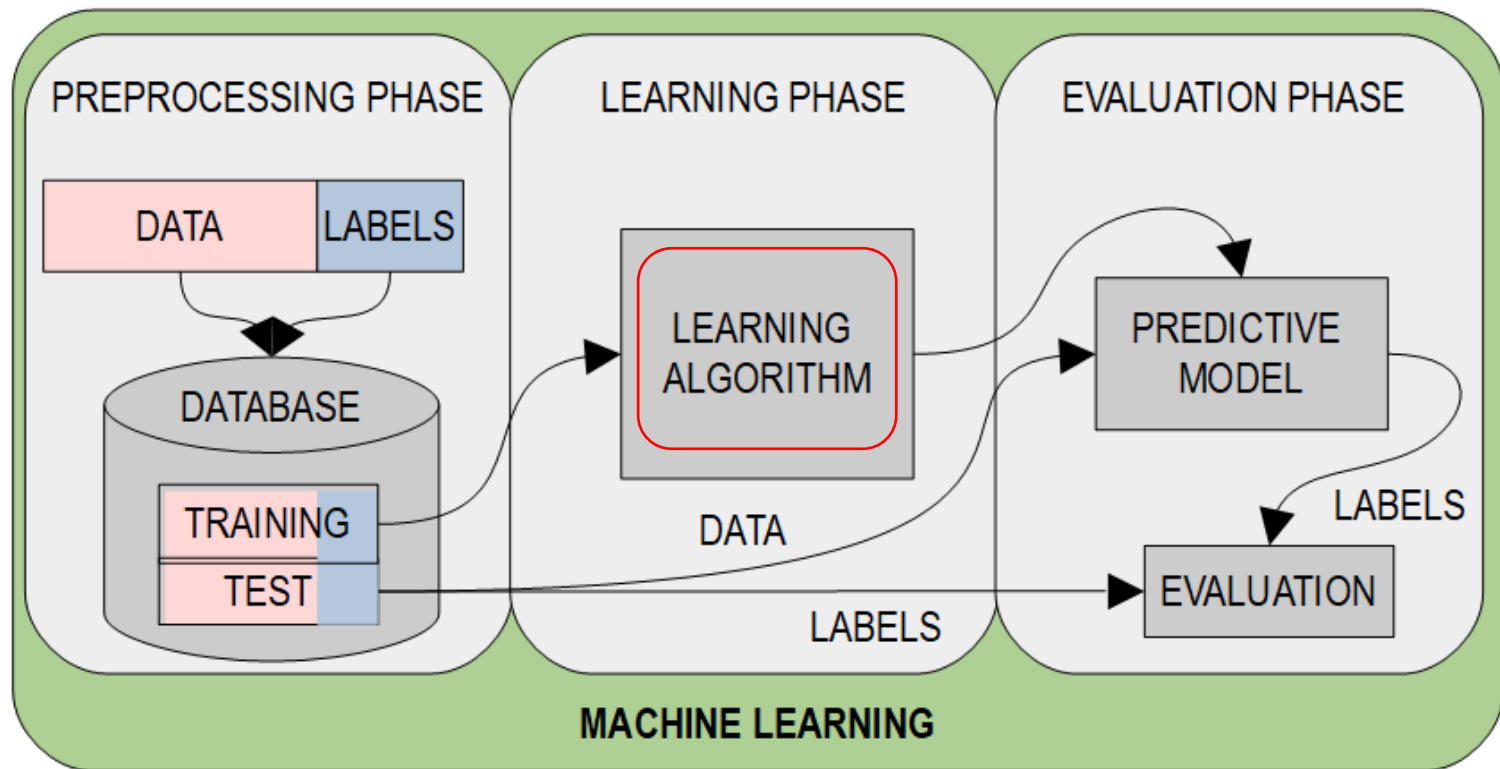
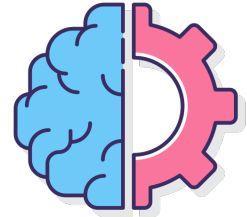
Experimentos

ML + BALANCEO



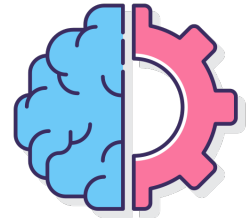
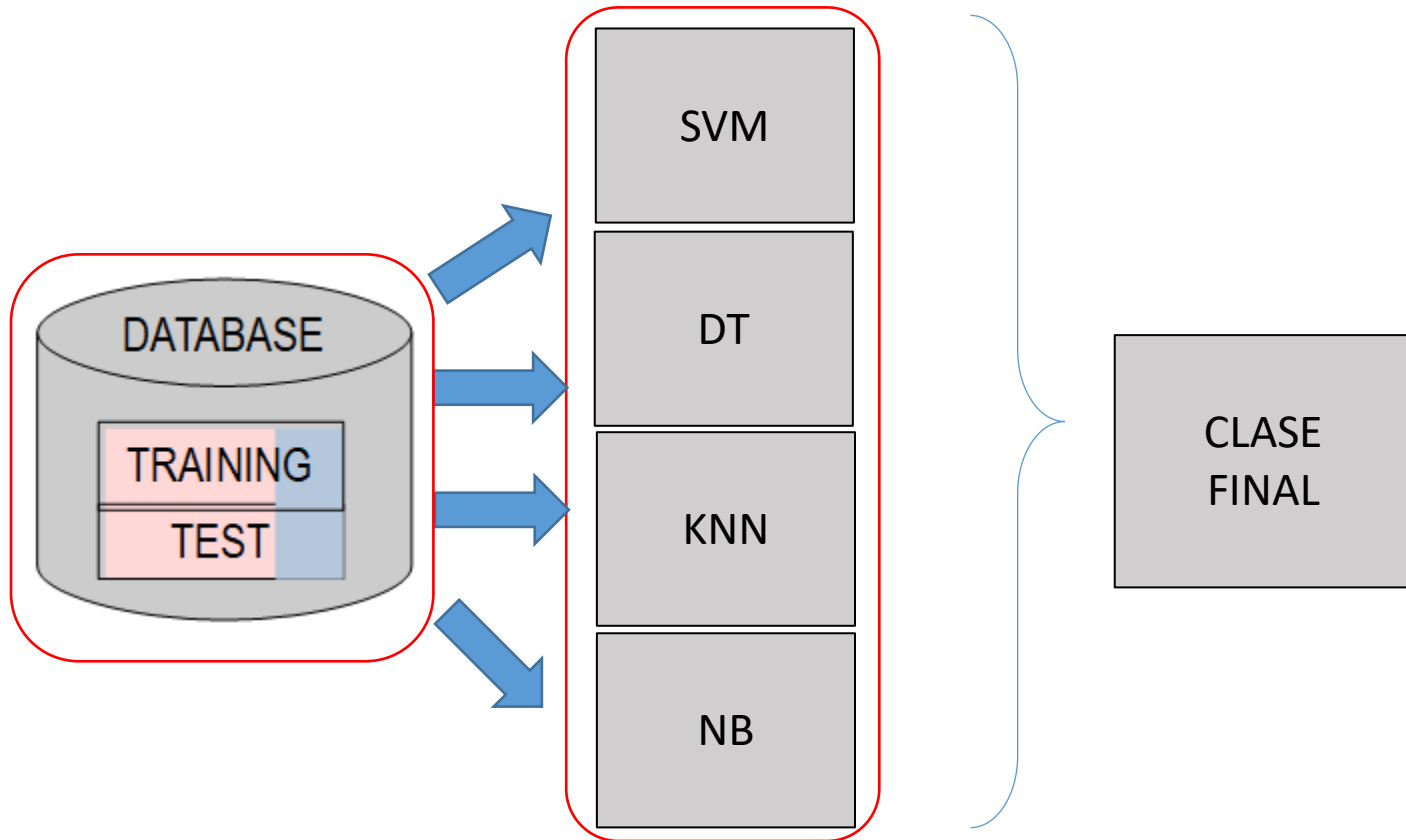
Experimentos

Machine Learning



Experimentos

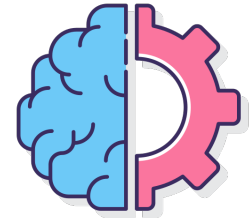
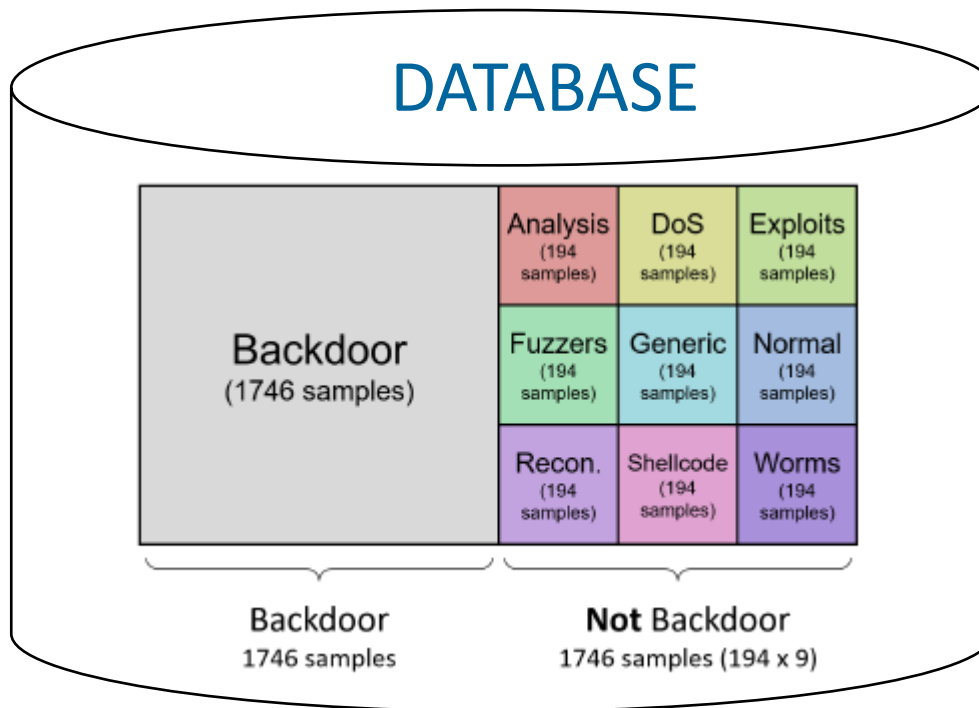
Ensemble Learning



-  Analysis
-  Backdoor
-  DDoS
-  Exploits
-  Fuzzers
-  Generic
-  Normal
-  Reconnaissance
-  Shellcode
-  Worms

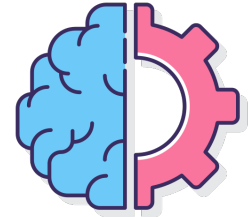
Experimentos

Ensemble Learning



- Analysis
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Experimentos



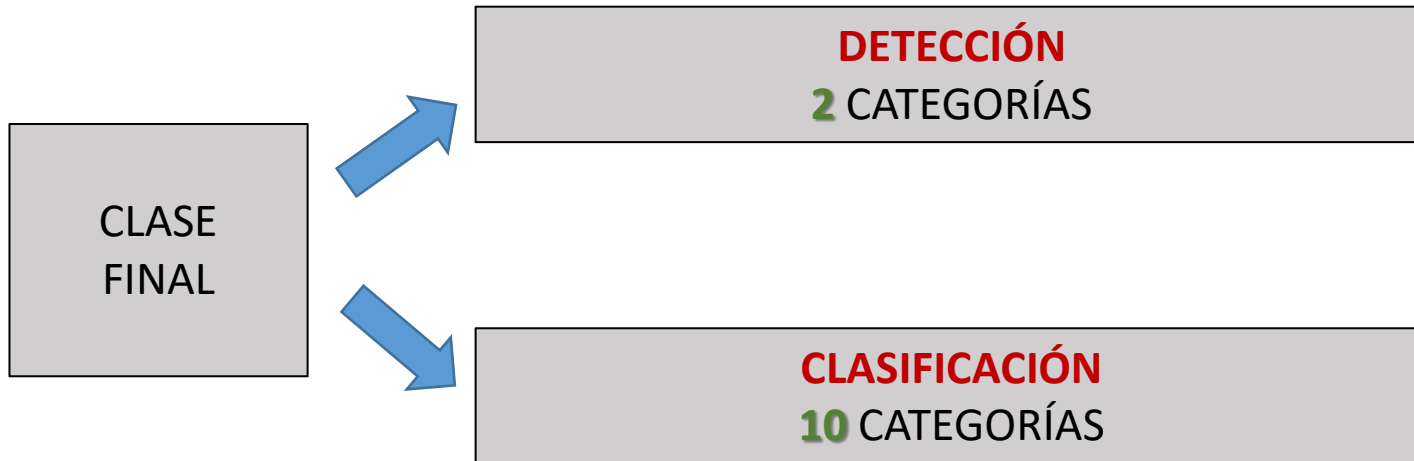
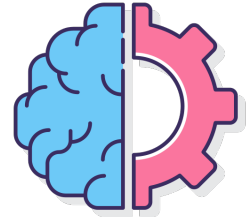
Ensemble Learning

Medicina general

- Oftalmología
- Otorrinolaringología
- Alergología
- Digestivo
- Cardiología
- Dermatología

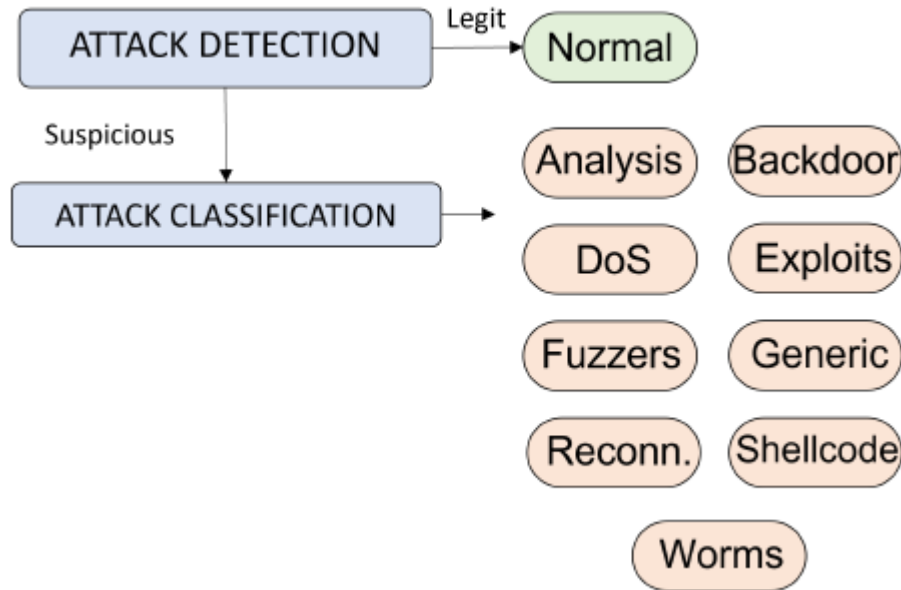
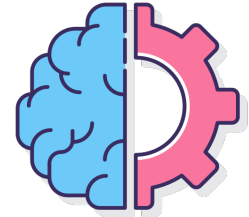
Experimentos

Ensemble Learning



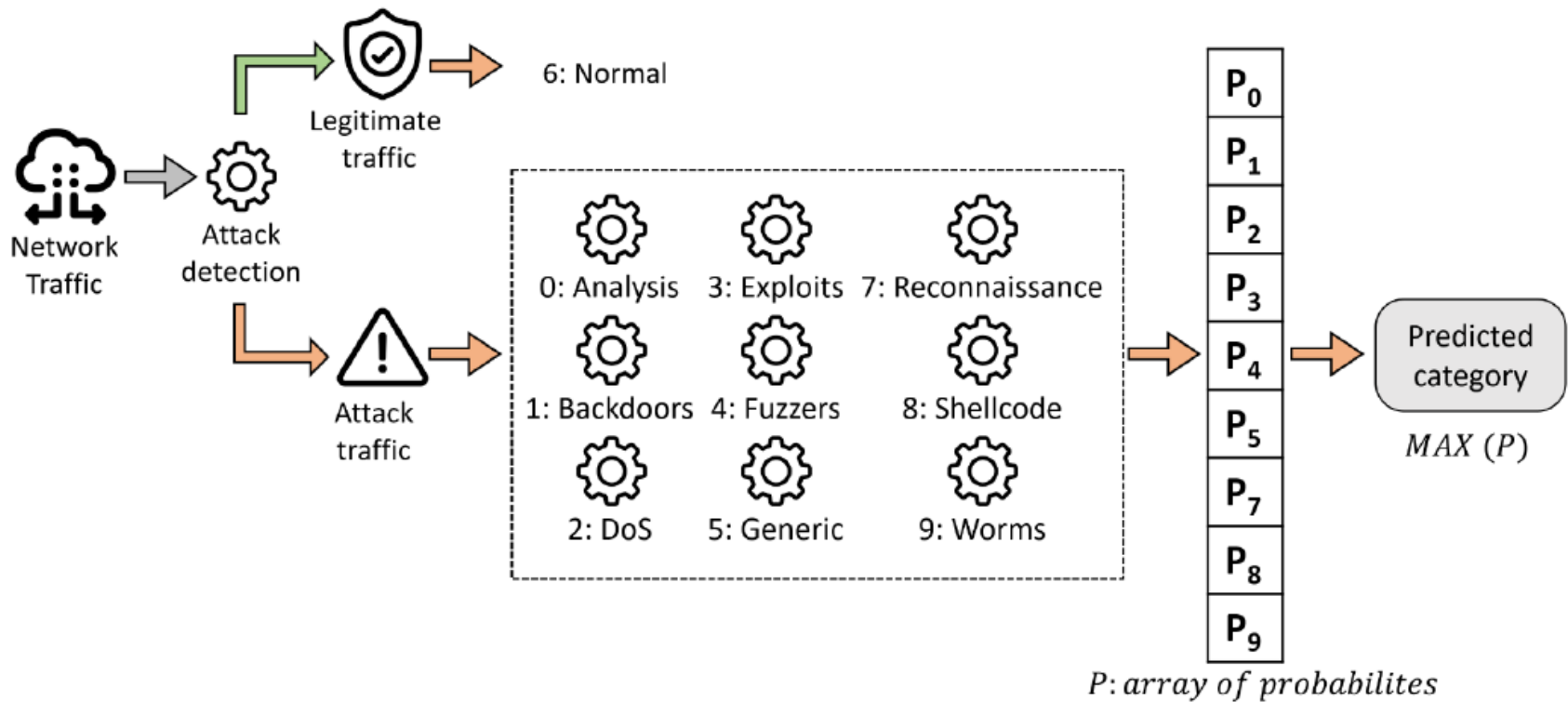
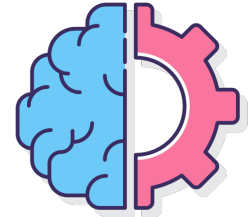
Experimentos

Ensemble Learning



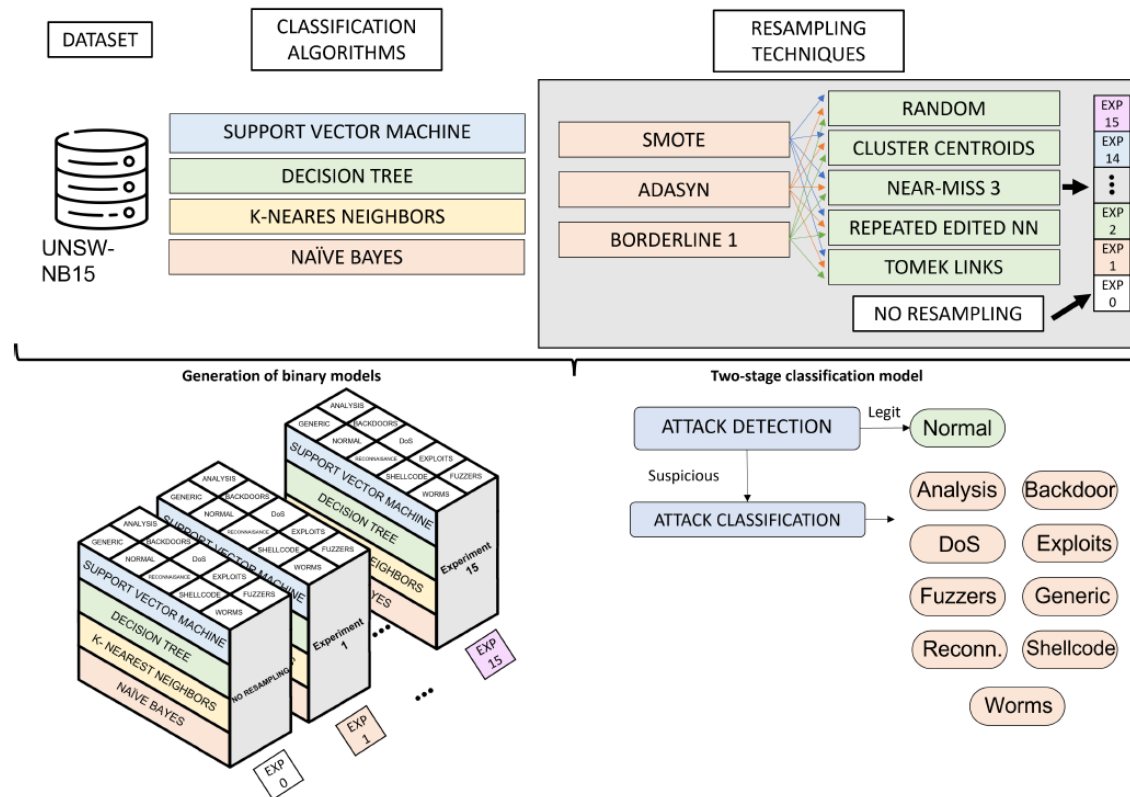
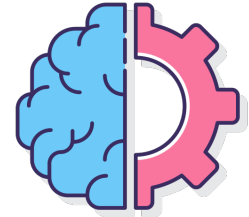
Experimentos

Ensemble Learning



Experimentos

Ensemble Learning



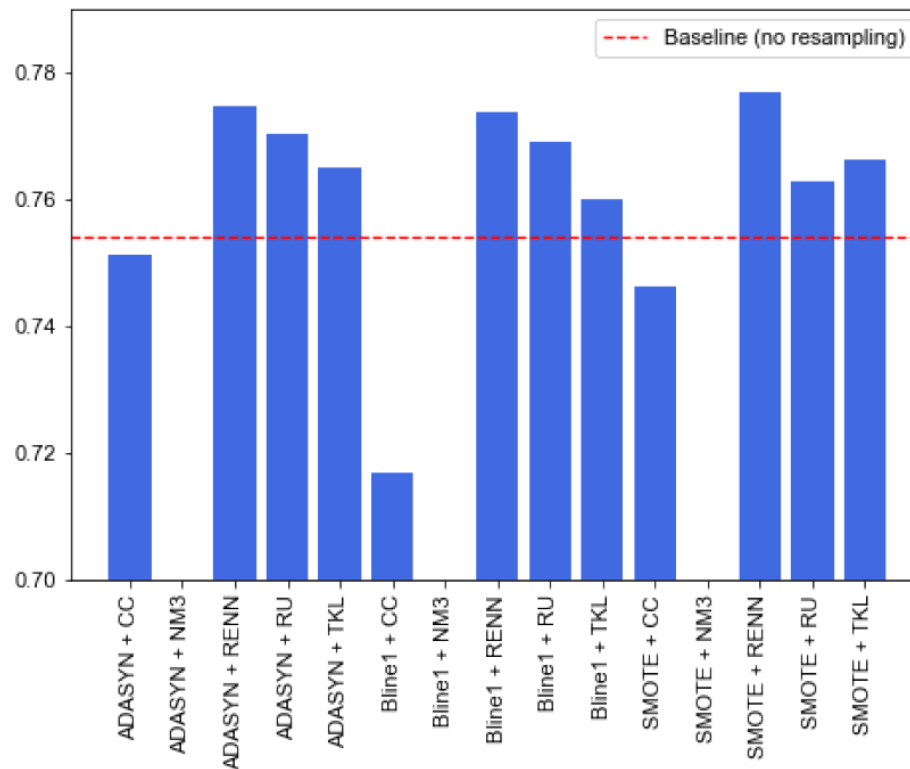
Resultados

Undersampling - Oversampling

Models	Composition									
	An.	Ba.	DoS	Ex.	Fu.	Ge.	No.	Re.	Sh.	Wo.
Analysis	-	US	US	US	US	US	US	US	US	OS
Backdoor	US	-	US	US	US	US	US	US	US	OS
DoS	US	US	-	US	US	US	US	US	OS	OS
Exploits	OS	OS	US	-	US	US	US	US	OS	OS
Fuzzers	OS	OS	US	US	-	US	US	US	OS	OS
Generic	OS	OS	US	US	US	-	US	US	OS	OS
Normal	OS	OS	US	US	US	US	-	US	OS	OS
Reconnaissance	US	US	US	US	US	US	US	-	OS	OS
Shellcode	US	US	US	US	US	US	US	US	-	US
Worms	US	US	US	US	US	US	US	US	US	-

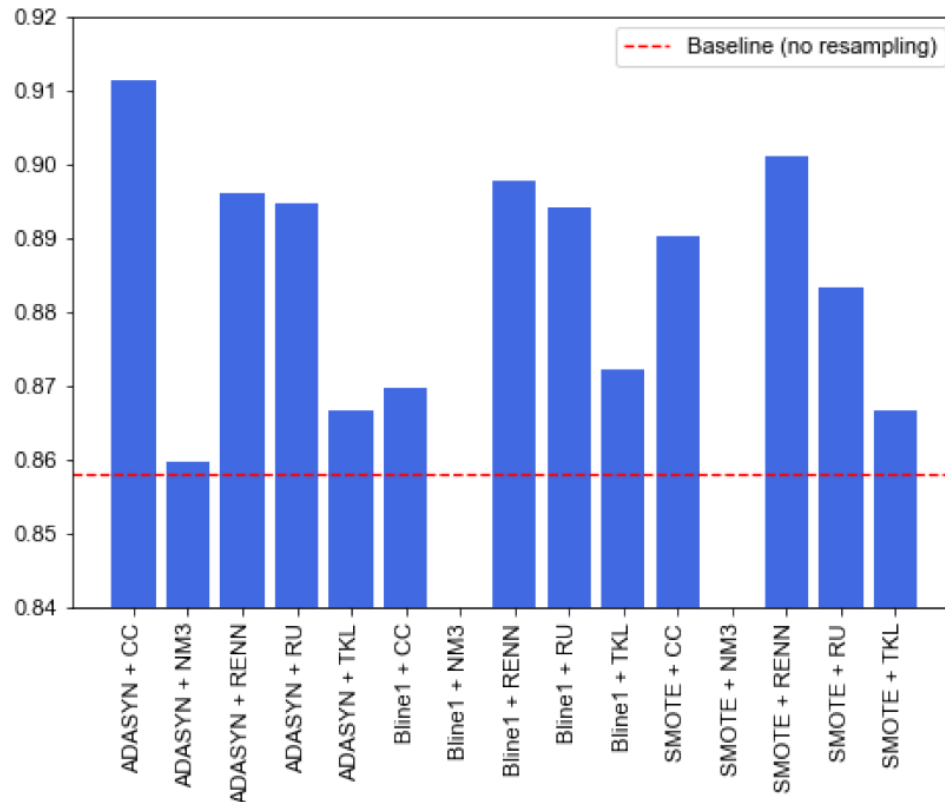
Resultados

Multiclass classification



Resultados

Clasificación en dos pasos

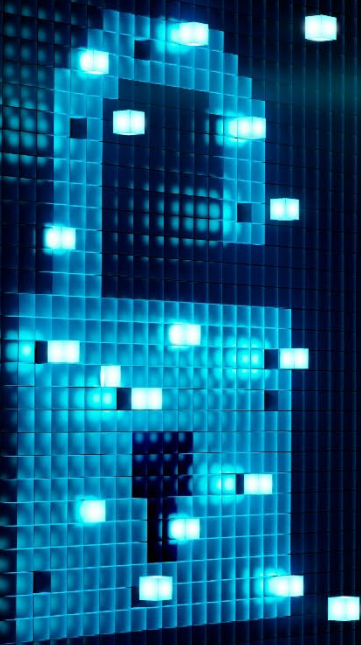


Resultados

Clasificación en dos pasos

Category	Acc.	Prec.	Recall	F1	FPR	Train Time
Analysis	0.8312	0.8388	0.8312	0.8298	0.1723	105 ms
Backdoor	0.8553	0.8747	0.8553	0.8530	0.1502	98 ms
DoS	0.7907	0.7936	0.7907	0.7909	0.2076	10.6 s
Exploits	0.7983	0.7975	0.7983	0.7963	0.2157	692 ms
Fuzzers	0.8127	0.8247	0.8127	0.8128	0.1823	47.4 s
Generic	0.9835	0.9839	0.9835	0.9835	0.0144	2 s
Normal	0.9112	0.9112	0.9112	0.9112	0.0952	2.4 s
Reconn.	0.9183	0.9255	0.9183	0.9183	0.0785	170 ms
Shellcode	0.9643	0.9644	0.9643	0.9643	0.0357	88 ms
Worms	0.8876	0.8884	0.8876	0.8876	0.1126	58 m

MUCHAS GRACIAS



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