	Malware detection with machine learning and Deep learning data science project My LinkedIn: Ismail Ameskour Build ML aand DL models to detect malware (we will make three algorithms: random forest, logistic regression, neural network (DL)) Make comparison of all model and select the best model
	 41,323(original file) binaries (exe,dll) - legitimate 96,724 malware files
In [166	<pre>import pandas as pd import sklearn from sklearn.feature_selection import SelectFromModel from sklearn.ensemble import ExtraTreesClassifier from sklearn.metrics import confusion_matrix from sklearn.model_selection import train_test_split from sklearn.model_selection import cross_validate import matplotlib.pyplot as plt</pre>
In [167 In [168	<pre>MalwareDataset = pd.read_csv('MalwareData.csv', sep=' ', low_memory =True)</pre> MalwareDataset.head()
Out[168	Name md5 Machine SizeOfOptionalHeader Characteristics MajorLinkerVersion MinorLinkerVersion SizeOfCode SizeOfInitializedData SizeOfUninitializedData 0 memtest.exe 631ea355665f28d4707448e442fbf5b8 332 224 258 9 0 361984 115712 1 ose.exe 9d10f99a6712e28f8acd5641e3a7ea6b 332 224 3330 9 0 130560 19968 2 setup.exe 4d92f518527353c0db88a70fddcfd390 332 224 3330 9 0 517120 621568 3 DW20.EXE a41e524f8d45f0074fd07805ff0c9b12 332 224 258 9 0 585728 369152 4 dwtrig20.exe c87e561258f2f8650cef999bf643a731 332 224 258 9 0 294912 247296
In [169	5 rows × 57 columns MalwareDataset.shape
Out[169 In [170	(138047, 57) MalwareDataset.describe() Machine SizeOfOrtionelllooder Characteristics MajortinkerVersion MinortinkerVersion SizeOfOrtionellist Research Address OfForty Point Research
Out[170	Machine SizeOfOptionalHeader Characteristics MajorLinkerVersion MinorLinkerVersion SizeOfCode SizeOfInitializedData SizeOfUninitializedData AddressOfEntryPoint Based count 138047.000000 138047.000000 138047.000000 138047.000000 1.380470e+05 1.380470e+05
	75% 332.000000 224.000000 8226.000000 10.000000 0.000000 1.203200e+05 3.850240e+05 0.000000e+00 6.157800e+04 4.09600 max 34404.000000 352.000000 49551.000000 255.000000 1.818587e+09 4.294966e+09 4.294941e+09 1.074484e+09 2.0287
In [171	<pre>Legit = MalwareDataset[0:41323].drop(['legitimate'], axis=1) Malware = MalwareDataset[41323::].drop(['legitimate'], axis=1) print("The shape of the legit dataset is : %s samples, %s features"%(Legit.shape[0], Legit.shape[1])) print("The shape of the mal dataset is : %s samples, %s features"%(Malware.shape[0], Malware.shape[1]))</pre> The shape of the legit dataset is : 41323 samples, 56 features
In [172	The shape of the mal dataset is: 96724 samples, 56 features fig = plt.figure() ax = fig.add_axes([0,0,1,1]) ax.hist(MalwareDataset['legitimate'],20) plt.show() # there are 2 number of legitimate in datasets is 1 and 0 , for 1 is original file while 0 is malware # so we can see output that 0 which were that malware files these are 96 000 something , and then 1 which were original #files somthing 40000 100000 400000 -
	20000 -
In [173	0 00 02 04 06 08 10 Data cleaning
	y = MalwareDataset['legitimate'] MalwareDataset=MalwareDataset.drop(['legitimate'],axis=1) Data preprocessing
In [174	MalwareDataset=MalwareDataset.drop(['Name'],axis=1) MalwareDataset=MalwareDataset.drop(['md5'],axis=1) print("the Name and md5 variables are removed successfully") the Name and md5 variables are removed successfully
In [175	Spliting the dataset into test and train #from sklearn.model_selection import train_test_split #X_train , X_test ,y_train , y_test = train_test_split(MalwareDataset, test_size=0.2 , random_state=42)
In [176	<pre>X_train, X_test, y_train, y_test = train_test_split(MalwareDataset, y,test_size=0.2, random_state=42) X_train.shape</pre>
Out[176 In [177	Model Building 1- Random Forest from sklearn.ensemble import RandomForestClassifier from sklearn.datasets import make_classification
	clf = RandomForestClassifier(max_depth=2 , random_state=0) randomModel = clf.fit(X_train , y_train) Random forest Evaluation on test data
In [178 In [179	<pre>from sklearn.metrics import f1_score , accuracy_score, plot_confusion_matrix , auc , confusion_matrix #accuracy on the train dataset train_pred = randomModel.predict(X_train) accuracy_score(y_train,train_pred)</pre>
Out[179 In [180	0.9828318407780001 #accuracy on the test dataset
Out[180 In [181	<pre>prediction=randomModel.predict(X_test) accuracy_score(y_test, prediction) 0.9838102136906918</pre>
Out[181	f1_score(y_test,prediction) 0.9730933606212002 Confusion matrix
In [182	<pre>titles_options = [("Confion matrix, without normalization", None),</pre>
	<pre>display_labels=['malware', 'legitimate'],</pre>
	<pre>print(title) print(disp.confusion_matrix) plt.show() Confion matrix, without normalization</pre>
	[[19080 170] [277 8083]] Confion matrix, without normalization malware - 19080 170 - 17500
	malware - 19080 170 - 12500 - 10000 - 7500 - 5000 - 2500
	malware legitimate Predicted label Normalized confusion matrix [[0.99116883 0.00883117] [0.03313397 0.96686603]]
	Normalized confusion matrix malware - 0.99
	legitimate - 0.033 0.97 - 0.2
In [183	2- Logistic regression from sklearn.linear_model import LogisticRegression
	<pre>clf = LogisticRegression(random_state=0) logModel=clf.fit(X_train, y_train) c:\users\ismail\appdata\local\programs\python\python39\lib\site-packages\sklearn\linear_model_logistic.py:763: ConvergenceWarning: lbfgs failed to c</pre>
	<pre>verge (status=2): ABNORMAL_TERMINATION_IN_LNSRCH. Increase the number of iterations (max_iter) or scale the data as shown in: https://scikit-learn.org/stable/modules/preprocessing.html Please also refer to the documentation for alternative solver options: https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression</pre>
In [184…	n_iter_i = _check_optimize_result(Mode Evaluaiton # Accuracy on the dataset
Out[184	train_log=logModel.predict(X_train) accuracy_score(y_train, train_log) 0.7015221347917817
In [185 Out[185	<pre>#Accuracy on the dataset pred=logModel.predict(X_test) accuracy_score(y_test,pred) 0.6972111553784861</pre>
In [186 Out[186	<pre>f1_score(y_test, pred) 0.0</pre>
In [187	<pre>Confusion matrix titles_options = [("Confion matrix, without normalization", None),</pre>
	normalize=normalize) disp.axset_title(title) print(title) print(disp.confusion_matrix)
	plt.show() Confion matrix, without normalization [[19250 0] [8360 0]] Confion matrix, without normalization
	malware - 19250 0 - 15000 - 12500
	10000
	malware legitimate Predicted label Normalized confusion matrix [[1. 0.] [1. 0.]]
	Normalized confusion matrix malware - 1 0 -0.8
	0.6
	malware legitimate Predicted label
In [188	3- Neural Netework import tensorflow as tf from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense
In [189	<pre>#Define model model=Sequential() model.add(Dense(16,input_dim=54,activation="relu")) model.add(Dense(8,activation="relu"))</pre>
	<pre>model.add(Dense(4,activation="relu")) model.add(Dense(1,activation="sigmoid")) model.summary() # Print model summary Model: "sequential_7"</pre> Model: "sequential_7"
	Layer (type)
	dense_16 (Dense) (None, 1) 5 ====================================
In [190…	<pre>Mon-trainable params: 0 #Compile model model.compile(loss="binary_crossentropy", optimizer="rmsprop", metrics=["accuracy"])</pre>
In [191	#fit Modeel model.fit(X_train,y_train,epochs=5,batch_size=33) Epoch 1/5 3347/3347 [====================================
	3347/3347 [====================================
L	3347/3347 [====================================
In [192	<pre>#Accuracy on the test dataset y_prediction=model.predict(X_test) y_prediction=[1 if y>=0.5 else 0 for y in y_prediction] accuracy_score(y_test,y_prediction) 863/863 [====================================</pre>
Out[192 In [193	0.9590365809489315 confusion_matrix(y_test,y_prediction)
Out[193 In [194 Out[194	array([[19002, 248],
L = 04	