Esempio Modulo Machine Learning

May 4, 2022

1 Esempi di Utilizzo del MODULO MACHINE LEARNING della libreria IntelligenzaArtificiale

1.1 Installare la libreria

```
[]: #Per installare la libreria sul tuo computer puoi usare il comando :
pip3 install intelligenzaartificiale

#se invece utilizzi google colab puoi usare:
!pip install intelligenzaartificiale
```

1.2 Importare la libreria

```
[1]: from intelligenzaartificiale import dataset as dt from intelligenzaartificiale import machinelearning as ml

il_mio_dataset = dt.leggi_csv("exams.csv")
```

Tempo impiegato per leggere il file: 0.0034322738647460938

1.3 Vedere se il dataset è Pulito

```
[2]: from intelligenzaartificiale import statistica as st

print("\n ------- Numero Valori nulli per colonna")
print(st.valori_nan(il_mio_dataset))
print("\n\n ----- Dataset \n")
print(il_mio_dataset)
```

```
----- Numero Valori nulli per colonna gender 0 sace/ethnicity 8 parental level of education 7 lunch 1 test preparation course 5 math score 3 reading score 8
```

writing score dtype: int64

----- Dataset

gender	race/ethnicity p	parental leve	l of education	lunch	\
male	group E	bac	helor's degree	standard	
female	group D		some college	free/reduced	
male	group E		high school	free/reduced	
male	group C	m	aster's degree	free/reduced	
male	NaN	m	aster's degree	free/reduced	
•••	•••		•••	•••	
female	group B		high school	standard	
female	group C		some college	standard	
male	group C	so	me high school	free/reduced	
female	group D	m	aster's degree	standard	
male	group A		high school	standard	
test pre	eparation course	math score	reading score	writing score	
	none	78.0	59.0	64.0	
	none	47.0	52.0	50.0	
	none	62.0	47.0	46.0	
	completed	55.0	65.0	68.0	
	none	61.0	54.0	55.0	
	•••	•••	•••	•••	
	none	33.0	36.0	33.0	
	none	52.0	59.0	64.0	
	none	66.0	64.0	62.0	
	completed	99.0	100.0	100.0	
	none	46.0	33.0	30.0	
	male female male male female female female male	male group E female group D male group E male group C male NaN female group B female group C male group C female group C state group D male group D male group A test preparation course none none completed none none completed completed completed completed	male group E back female group D male group E male group C m male NaN m female group B female group C male group C male group C male group C male group A test preparation course math score none 78.0 none 47.0 none 62.0 completed 55.0 none 61.0 none 33.0 none 52.0 none 66.0 completed 99.0	female group D some college male group E high school male group C master's degree male NaN master's degree male group B high school female group C some college male group C some high school female group D master's degree male group A high school test preparation course math score reading score none 78.0 59.0 none 47.0 52.0 none 62.0 47.0 completed 55.0 65.0 none 61.0 54.0 none 52.0 59.0 none 52.0 59.0 none 66.0 64.0 completed 99.0 100.0	male group E bachelor's degree standard female group D some college free/reduced male group E high school free/reduced male group C master's degree free/reduced male group B high school standard female group C some college standard male group C some high school free/reduced female group D master's degree standard male group A high school free/reduced female group C some high school free/reduced female group C some high school free/reduced female group C some high school free/reduced female group D master's degree standard female group A high school free/reduced female group A high school free/reduced female group A high school free/reduced

2

[1000 rows x 8 columns]

1.4 PreProcessiamo il nostro Set di Dati

```
[3]: from intelligenzaartificiale import preprocessing as pp

#leviamo i valori nulli
il_mio_dataset = pp.rimuovi_nan(il_mio_dataset)

#effettuare il labelencoding su più colonne
il_mio_dataset = pp.label_encoding_multiplo(il_mio_dataset,["test preparation_\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u00fc\u
```

	gender	race/eth	${ t nicity}$	parental	level	of e	${\tt ducation}$	lunch	\
0	:	L	4				1	1	
1	()	3				4	0	
2		L	4				2	0	
3	:	L	2				3	0	
5	()	3				1	1	
			•••						
995	()	1				2	1	
996	()	2				4	1	
997		L	2				5	0	
998	()	3				3	1	
999	_	L	0				2		
	test p	reparation	course	math sc	ore r	eadin	g score	writing	score
0	•	•	1		3.0		59.0	J	64.0
1			1	4	7.0		52.0		50.0
2			1		2.0		47.0		46.0
3			0		5.0		65.0		68.0
5			1		0.0		58.0		61.0
			•••	•••			•••	•••	
995			1	33	3.0		36.0		33.0
996			1		2.0		59.0		64.0
997			1		5.0		64.0		62.0
998			0		9.0		100.0		100.0
			•		-				

[966 rows x 8 columns]

999

1.5 Scoprire l'algoritmo più performante

1

```
[4]: colonne_x = ["test preparation course", "gender", "race/ethnicity", "parental
location", "lunch", "reading score", "writing score"]

target = "math score"

# scoprire algoritmo di regressione più performante
modello = ml.performance_modelli_regressione(il_mio_dataset, colonne_x, target)

# scoprire algoritmo di classificazione più performante
#modello = ml.performance_modelli_classificazione(il_mio_dataset, colonne_x, locatione)

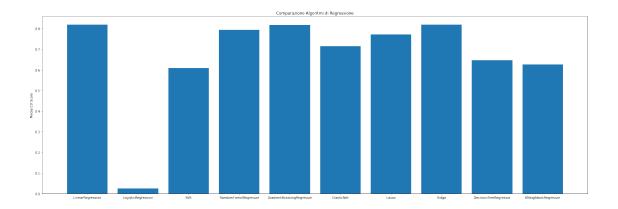
target)

print("il modello migliore è : " + str(modello))
```

46.0

33.0

30.0



il modello migliore è : LinearRegression()

1.6 Creare il modelli di regressione [SEMPLICE]

```
[5]: # dopo aver scoperto l'algoritmo più performante lo potrai implementare in unau sola riga
il_mio_modello = ml.regressione_gradient_boosting(il_mio_dataset, colonne_x,ustarget)

print("Il modello è stato allenato " + str(il_mio_modello))
```

Il modello è stato allenato GradientBoostingRegressor()

1.6.1 Lista di modelli Disponibili:

Regressori:

- modello1 = ml.regressione lineare(il mio dataset, colonne x, target)
- modello2 = ml.regressione logistica(il mio dataset, colonne x, target)
- modello3 = ml.regressione SVR(il mio dataset, colonne x, target)
- modello4 = ml.regressione_SVC(il_mio_dataset, colonne_x, target)
- modello5 = ml.regressione_random_forest(il_mio_dataset, colonne_x, target)
- modello6 = ml.regressione_gradient_boosting(il_mio_dataset, colonne_x, target)
- modello7 = ml.regressione decision tree(il mio dataset, colonne x, target)
- modello8 = ml.regressione_knn(il_mio_dataset, colonne_x, target)
- modello9 = ml.modello_elastic_net(il_mio_dataset, colonne_x, target)
- modello 10 = ml.modello lasso(il mio dataset, colonne x, target)
- modello11 = ml.modello ridge(il mio dataset, colonne x, target)

Classificatori:

- modello1 = ml.classificatore random forest(il mio dataset, colonne x, target)
- modello2 = ml.classificatore_gradient_boosting(il_mio_dataset, colonne_x, target)
- modello3 = ml.classificatore decision tree(il mio dataset, colonne x, target)
- modello4 = ml.classificatore_knn(il_mio_dataset, colonne_x, target)

- modello5 = ml.classificatore_logistico(il_mio_dataset, colonne_x, target)
- modello6 = ml.classificatore_naivebayes(il_mio_dataset, colonne_x, target)
- modello7 = ml.classificatore_svm(il_mio_dataset, colonne_x, target)

1.7 Valutare, Prevedere, Salvare e Caricare un modello

```
[6]: # Valutare un modello
     print(ml.valutazione modello(il mio modello,il mio dataset, colonne x, target))
     # Spiegare un modello
     ml.spiega modello(il_mio_modello,il_mio_dataset, colonne_x, target)
     # Previsione con un modello
     nuovo_dataset = dt.leggi_csv("dataset_test.csv")
     #leviamo i valori nulli
     nuovo_dataset = pp.rimuovi_nan(nuovo_dataset)
     #effettuare il labelencoding su più colonne
     nuovo_dataset = pp.label_encoding_multiplo(nuovo_dataset,["test preparation_
      ⇔course", "gender", "race/ethnicity", "parental level of education", ⊔

¬"lunch"])
     previsioni = ml.predizione_y(il_mio_modello,nuovo_dataset)
     print(previsioni)
     # Salvare e Caricare un modello
     ml.salva_modello(il_mio_modello,"nome_modello")
     vecchio_modello = ml.carica_modello("nome_modello")
```

0.8720584862249142

Feature 0 : test preparation course

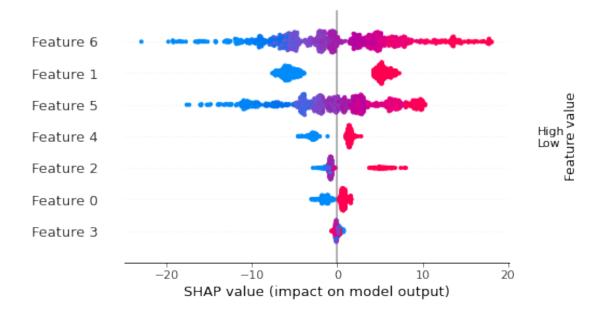
Feature 1 : gender

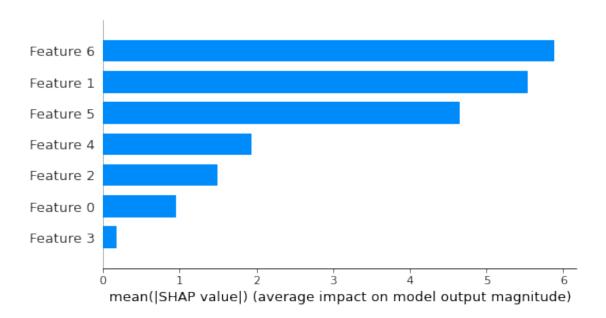
Feature 2 : race/ethnicity

Feature 3 : parental level of education

Feature 4 : lunch

Feature 5 : reading score Feature 6 : writing score





Tempo impiegato per leggere il file: 0.0020928382873535156 [66.84271213 61.83662268 52.31382548 66.11906333 59.42655177 62.77534272 68.74875798 86.42373963 71.39188096 73.9828311 88.72600086 83.36908134 54.01600518 68.60257738 49.95320093 93.82308641 61.41654404 74.8149285 61.80346554 69.1610261 77.57524143 83.19602997 72.16056839 54.30939007 50.07943696 75.93225747 81.64408563 86.12767091 49.12534691 86.31633209 85.07435461 59.7625009 72.59924757 59.51509962 91.08612808 67.57389125 95.45736007 67.73130997 50.33766591 64.91773883 79.09882821 52.75166538

```
90.99868585 86.15748795 59.91171349 53.04888707 69.52448189 81.11994678 90.52348981 69.69634439 67.81845839 70.13043753 99.11857026 81.37451939 51.15747059 69.12878077 64.9219627 68.42990118 93.81985533 90.15230664 86.28322134 80.67966904 79.03667897 84.26331199 81.64408563 80.24193654 56.81337801 66.47213791 58.06671441 67.70444244 90.18778551 74.79730823 58.7771907 78.6735824 75.80291878 81.37510074 51.25464091 73.24810364 72.4639118 50.93059133]
```

1.8 Altre risorse

- Documentazione Ufficiale
- Blog Ufficiale
- Corsi Gratis
- Ebook Gratis
- Progetti Python Open Source
- Dataset Pubblici
- Editor Python Online per il M.L.
- 2 Per favore citaci se usi la Libreria.