

#Diseases Dataset

#This is a machine learning project on different type of diseases.  
#Are there any potential relationships with different epidermis issues.  
#Does these diseases have anything in common with each other  
#Are there any other factors that can cause or make these diseases worse

!date

Sat 20 Jul 2024 02:16:40 PM EDT

from IPython import display  
display.Image('/home/harlohutch77/python/python\_master/rash.jpeg',height=400, width=800)



# importing libraries

import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
%matplotlib inline

#This is a csv that is being made into a dataframe  
df =pd.read\_csv('/home/harlohutch77/python/python\_master/d\_training.csv')  
df.head()

	itching	skin_rash	nodal_skin_eruptions	continuous_sneezing	shivering	chills	joint_pain	stomach_pain	acidity	ulcers_on_tongue	...
0	1	1		1	0	0	0	0	0	0	...
1	0	1		1	0	0	0	0	0	0	...
2	1	0		1	0	0	0	0	0	0	...
3	1	1		0	0	0	0	0	0	0	...
4	1	1		1	0	0	0	0	0	0	...

5 rows × 134 columns

# other examples of removing columns  
# Drop column 'B'  
#df = df.drop('B', axis=1)  
  
# Drop columns multiple column  
#df = df.drop(['B', 'C'], axis=1)

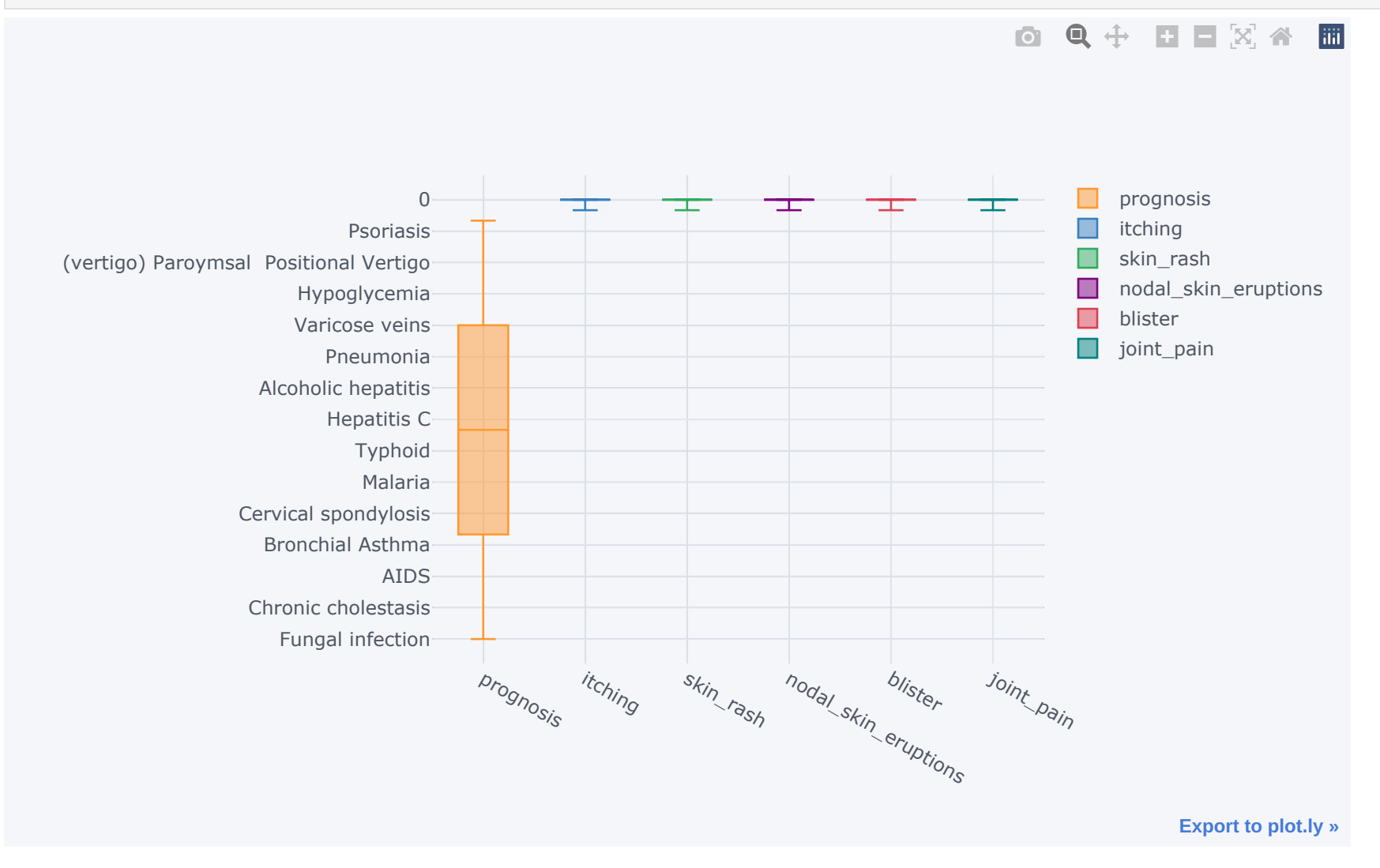
#del df['column\_name'] deletes column from dataframe  
del df['Unnamed: 133']

import seaborn as sns  
import cufflinks as cf

from plotly import \_\_version\_\_  
from plotly.offline import download\_plotlyjs, init\_notebook\_mode, plot, iplot  
  
print(\_\_version\_\_) # requires version >= 1.9.0  
  
# For Notebooks - jupyter notebook  
init\_notebook\_mode(connected=True)  
  
cf.go\_offline()

5.7.0

#This visualization show the prognosis as negative or postive; 0 = don't have and  
#1 = does have  
df[['prognosis', 'itching', 'skin\_rash', 'nodal\_skin\_eruptions', 'blister', 'joint\_pain']].iplot(kind='box')



[Export to plot.ly »](#)

Machine Learning

from sklearn.model\_selection import train\_test\_split

X = df.drop('prognosis', axis=1)  
y = df['prognosis']

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X,y, test\_size=0.30, random\_state=42)

Random Forest Machine Learning

from sklearn.ensemble import RandomForestClassifier  
df = RandomForestClassifier(n\_estimators=100)  
df.fit(X\_train, y\_train)

RandomForestClassifier()

df\_pred = df.predict(X\_test)

from sklearn.metrics import classification\_report, confusion\_matrix

#Confusion Matrix  
print(confusion\_matrix(y\_test,df\_pred))

[[32 0 0 ... 0 0 0]  
[ 0 39 0 ... 0 0 0]  
[ 0 0 41 ... 0 0 0]  
...  
[ 0 0 0 ... 36 0 0]  
[ 0 0 0 ... 0 37 0]  
[ 0 0 0 ... 0 0 39]]

print(classification\_report(y\_test,df\_pred))

		precision	recall	f1-score	support
(vertigo) Paroysmal	Positional Vertigo	1.00	1.00	1.00	32
	AIDS	1.00	1.00	1.00	39
	Acne	1.00	1.00	1.00	41
	Alcoholic hepatitis	1.00	1.00	1.00	36
	Allergy	1.00	1.00	1.00	35
	Arthritis	1.00	1.00	1.00	36
	Bronchial Asthma	1.00	1.00	1.00	44
	Cervical spondylosis	1.00	1.00	1.00	32
	Chicken pox	1.00	1.00	1.00	35
	Chronic cholestasis	1.00	1.00	1.00	30
	Common Cold	1.00	1.00	1.00	31
	Dengue	1.00	1.00	1.00	40
	Diabetes	1.00	1.00	1.00	33
	Dimorphic hemmorhoids(piles)	1.00	1.00	1.00	45
	Drug Reaction	1.00	1.00	1.00	35
	Fungal infection	1.00	1.00	1.00	28
	GERD	1.00	1.00	1.00	41
	Gastroenteritis	1.00	1.00	1.00	40
	Heart attack	1.00	1.00	1.00	36
	Hepatitis B	1.00	1.00	1.00	46
	Hepatitis C	1.00	1.00	1.00	40
	Hepatitis D	1.00	1.00	1.00	35
	Hepatitis E	1.00	1.00	1.00	47
	Hypertension	1.00	1.00	1.00	31
	Hyperthyroidism	1.00	1.00	1.00	37
	Hypoglycemia	1.00	1.00	1.00	42
	Hypothyroidism	1.00	1.00	1.00	33
	Impetigo	1.00	1.00	1.00	35
	Jaundice	1.00	1.00	1.00	30
	Malaria	1.00	1.00	1.00	30
	Migraine	1.00	1.00	1.00	36
	Osteoarthritis	1.00	1.00	1.00	28
	Paralysis (brain hemorrhage)	1.00	1.00	1.00	34
	Peptic ulcer disease	1.00	1.00	1.00	27
	Pneumonia	1.00	1.00	1.00	39
	Psoriasis	1.00	1.00	1.00	32
	Tuberculosis	1.00	1.00	1.00	35
	Typhoid	1.00	1.00	1.00	34
	Urinary tract infection	1.00	1.00	1.00	36
	Varicose veins	1.00	1.00	1.00	37
	hepatitis A	1.00	1.00	1.00	39
	accuracy			1.00	1476
	macro avg	1.00	1.00	1.00	1476
	weighted avg	1.00	1.00	1.00	1476

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5 rows × 134 columns

#I am focusing on the epidermis issues to see if there is some kind of commonality

#the amount of patients with this issue  
df['nodal\_skin\_eruptions'].sum()

108

#this is how many patients that had this issue  
df['small\_dents\_in\_nails'].sum()

114

#this is for the full count of patients accounted for  
df['small\_dents\_in\_nails'].count()

4920

#this is the amount of patients that have a skin rash  
df['skin\_rash'].sum()

786

#  
df['skin\_rash'].describe()

count 4920.000000  
mean 0.159756  
std 0.366417  
min 0.000000  
25% 0.000000  
50% 0.000000  
75% 0.000000  
max 1.000000  
Name: skin\_rash, dtype: float64

#description and explanation of prognosis  
df['prognosis'].describe()

count 4920  
unique 41  
top Fungal infection  
freq 120  
Name: prognosis, dtype: object

#114 patients have Psoriasis and skin\_rash  
df[df['prognosis']=='Psoriasis']['skin\_rash'].sum()

114

df[df['prognosis']=='impetigo']['skin\_rash'].sum()

0

#this is to cross reference skin issues with fungal infection  
df[df['prognosis']=='Fungal infection'][['skin\_rash', 'itching', 'nodal\_skin\_eruptions']].sum()

skin\_rash 108  
itching 108  
nodal\_skin\_eruptions 108  
dtype: int64

# 4920 prognosis's in this dataset  
# Is there a correlation between certain epidermis issues  
# What other factors are at play

from IPython import display  
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